<table>
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<tr>
<th>SPM Page:Line</th>
<th>Chapter/Supp. Material</th>
<th>Chapter Page:Line</th>
<th>Summary of edit to be made</th>
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<tr>
<td>4:25-4:29</td>
<td>2</td>
<td>6:16</td>
<td>Immediately after bolded text add: GMST in the first two decades of the 21st century (2001-2020) was 0.99 [0.84-1.10] °C higher than 1850-1900. Each of the last 4 decades has successively been warmer than all preceding decades since 1850.</td>
</tr>
<tr>
<td>6:45-48</td>
<td>2</td>
<td>7:34</td>
<td>Add after ‘2-3 decades.’ as follows: A long-term increase in surface open ocean pH occurred over the past 50 million years (high confidence), and surface ocean pH as low as recent times is uncommon in the last 2 million years (medium confidence)</td>
</tr>
<tr>
<td>4:25-4:29</td>
<td>2</td>
<td>39:26</td>
<td>To Cross-Chapter Box 2.3 Table 1 add the following values into a new row (to be positioned immediately before the existing row for 1850-1900 to 1995-2014): 1850-1900 to 2001-2020; 0.99 [0.84-1.10]; 0.99[0.81-1.14]; [empty]; Warming to first two decades of 21st century</td>
</tr>
<tr>
<td>12:23</td>
<td>2</td>
<td>40:35</td>
<td>Rephrase comparison to SR1.5 so as to connect more effectively to the SPM. Replace “about 10 years earlier than the midpoint” with “in the early part”.</td>
</tr>
<tr>
<td>12:23</td>
<td>2</td>
<td>40:37</td>
<td>Rephrase comparison to SR1.5 so as to connect more effectively to the SPM. Replace “the ten-year” with “this”.</td>
</tr>
<tr>
<td>12:27</td>
<td>2</td>
<td>40:40</td>
<td>Add sentence referring to another location in the SR1.5. After “(medium confidence)”, add: “When considering scenarios similar to SSP1-1.9 instead of linear extrapolation, the SR1.5 estimate of when 1.5°C global warming is crossed is close to the central estimate reported here (SR1.5, Table 2.SM.12)”</td>
</tr>
<tr>
<td>4:25-4:29</td>
<td>2</td>
<td>46:1</td>
<td>Add column to Table as detailed in attached document (appendix 1). Change is to add a column of change to 2001-2020 to support its addition at SPM level.</td>
</tr>
<tr>
<td>4:25-4:29</td>
<td>2</td>
<td>47:1</td>
<td>before ‘and to the most recent decade’, add ‘to the first two decades of the 21st century (2001-2020) by 0.99 [0.84-1.10] °C’.</td>
</tr>
</tbody>
</table>
## Appendix 1

<table>
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<th>Dataset</th>
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## Appendix 1

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The corrigenda listed below will be implemented in the Chapter during copy-editing.

### CHAPTER 2

<table>
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<tr>
<th>Document (Chapter, Annex, Supp. Mat…)</th>
<th>Section</th>
<th>Page :Line (based on the final pdf FGD version)</th>
<th>Detailed info on correction to make</th>
</tr>
</thead>
<tbody>
<tr>
<td>Chapter 2</td>
<td>Author 1:31</td>
<td>Replace “Hong Kong” with “Hong Kong, China”</td>
<td></td>
</tr>
<tr>
<td>Chapter 2</td>
<td>ES 6:16-17</td>
<td>Replace 'medium confidence' with &quot;High confidence&quot; to read: Since around 1950, GMST has increased at an observed rate unprecedented for any 50-year period in at least the last 2000 years (high confidence).” Required for consistency with changes made to SPM</td>
<td></td>
</tr>
<tr>
<td>Chapter 2</td>
<td>ES 6:10-12</td>
<td>Replace entire of un-bolded text with: Temperatures as high as during the most recent decade (2011-2020) exceed the warmest centennial-scale range reconstructed for the present interglacial, around 6,500 years ago [0.2-1°C relative to 1850-1900] (medium confidence). The next older warm period is the last interglacial when the multi-centennial temperature range about 125,000 years ago [0.5-1.5°C relative to 1850-1900] encompasses the recent decade values (medium confidence).” Required for consistency with changes made to SPM Paragraph to read: Observed changes in the atmosphere, oceans, cryosphere and biosphere provide unequivocal evidence of a world that has warmed. Over the past several decades, key indicators of the climate system are increasingly at levels unseen in centuries to millennia, and are changing at rates unprecedented in at least the last 2000 years (high confidence). Temperatures as high as during the most recent decade (2011-2020) exceed the warmest centennial-scale range reconstructed for the present interglacial, around 6,500 years ago [0.2-1°C relative to 1850-1900] (medium confidence). The next older warm period is the last interglacial when the multi-centennial temperature range about 125,000 years ago [0.5-1.5°C relative to 1850-1900] encompasses the recent decade values (medium confidence).</td>
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<tr>
<td>Chapter 2</td>
<td>ES 7:13</td>
<td>An abbreviation for AIS is introduced but not GrIS (abbreviation for GrIS is introduced at 14:15). For consistency: Introduce Abbreviation for both or none.</td>
<td></td>
</tr>
<tr>
<td>Chapter 2</td>
<td>ES 7:21</td>
<td>Please add new sentence after bolded GMSL text: The average rate of sea level rise was 1.3 [0.6 to 2.1] mm yr-1 between 1901 and 1971, increasing to 1.9 [0.8 to 2.9] mm yr-1 between 1971 and 2006, and further increasing to 3.7 [3.2 to 4.2] mm yr-1 between 2006 and 2018 (high confidence). Required for consistency with changes made to SPM</td>
<td></td>
</tr>
<tr>
<td>Chapter 2</td>
<td>CCB2.1, Table 11</td>
<td>Replace “Steinthorsdottir et al., 2020” by “Steinthorsdottir et al., 2021”</td>
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<tr>
<td>Chapter 2</td>
<td>CCB2.1, Table 11</td>
<td>Replace “Kageyama et al., 2017; 2021” by “Kageyama et al., 2017, 2021a”</td>
<td></td>
</tr>
<tr>
<td>Chapter 2</td>
<td>2.2.1 13 : 15; 13 : 20</td>
<td>Replace “Wu et al., 2018” by “C.-J. Wu et al., 2018”</td>
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</tbody>
</table>
Chapter 2 2.2.1   13 : 16 Replace “Egorova, T. et al., 2018” by “Egorova et al., 2018”
Chapter 2 2.2.1   13 : 33 Replace “Yeo et al., 2015, 2017a” by “Yeo et al., 2015, 2017”
Chapter 2 2.2.2   14 : 45 Replace “Luo et al, 2018” by “Luo, 2018”
Chapter 2 2.2.3.1   15 : 23-24 Replace “Zhang et al., 2019” by “Y. Zhang et al., 2019”
Chapter 2 2.2.3.1   15 : 29 Replace “Egorova, T. et al., 2018” by “Egorova et al., 2018”
Chapter 2 2.2.3.2.1   17 : 31-32 Replace “Marcott et al., 2014” by “Marcott et al., 2014a”
Chapter 2 2.2.3.2.2   18 : 7-8; 18 : 10-11 Replace “WAIS Divide Project Members, 2015” by “Buizert et al., 2015”
Chapter 2 2.2.3.2.2   18 : 15-16; 18 : 19 Replace “Mitchell et al., 2013b” by “L. Mitchell et al., 2014”
Chapter 2 2.2.4   21 : 12 Replace “Engel & Rigby, 2018; Montzka & Velders, 2018” by “Engel et al., 2018; Montzka et al., 2018b”
Chapter 2 2.2.4.1   21 : 29 Replace “Montzka et al., 2018, 2021” by “Montzka et al., 2018a, 2021”
Chapter 2 2.2.6   26 : 45 Replace “Yu et al., 2020” by “H. Yu et al., 2020”
Chapter 2 2.2.6   26 : 49 Replace “Zhao et al., 2017” by “S.-P. Zhao et al., 2017”
Chapter 2 2.2.6   27 : 28; 27 : 31; Replace “Santer et al., (2008a)” by “Santer et al. (2008)”
Chapter 2 2.2.7   28 : 1 Replace “Li et al., 2020” by “F. Li et al., 2020”
Chapter 2 2.2.7   28 : 22 Replace “Smith et al., (2020)” by “C.J. Smith et al., (2020)”
Chapter 2 2.3.3.3   28 : 28 Replace “Kopp et al., 2016b; Kemp et al., 2018a” by “Kopp et al., 2016; Kemp et al., 2018”
Chapter 2 2.3.1.1   32 : 48; 33 : 3; 33 : 37; 33 : 51 Replace “Hansen et al. (2013c)” by “J. Hansen et al. (2013)”
Chapter 2 2.3.1.1.2   35 : 38 Replace “Santer et al., (2008a)” by “Santer et al. (2008)”
Chapter 2 2.3.1.1.2   35 : 25-26 Replace ‘medium confidence’ with “high confidence” to read:
“Since around 1950, GMST has increased at an observed rate unprecedented for any 50-year period in at least the last 2000 years (high confidence).”
Required for consistency with changes made to SPM
Chapter 2 CCB 2.3   36 : 19; 36 : 29 Replace “Kadow et al. (2020a)” by “Kadow et al. (2020)”
Chapter 2 CCB 2.3   37 : 50-51 Replace “Junod and Christy (2020a)” by “Junod and Christy (2020)”
Chapter 2 CCB 2.3   37 : 50-51 Replace “Zhou et al., 2020” by “Zhou et al., 2020a”
Chapter 2 CCB 2.3   38 : 36 Replace “Liu W. et al., 2015a” by “W. Liu et al., 2015”
Chapter 2 CCB 2.3, Table 1 caption   39:21-24 Update caption to the following:
Summary of key observationally-based global warming estimates (in °C) to various reference periods in the present report and selected prior reports (AR5 WGI and SR1.5) and their principal applications (see Section 1.4.1 for further information on reference periods). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Agreed change with TSU and Bureau
Chapter 2   39:25 All brackets in this table should be [] and not ().
Chapter 2 CCB 2.3, Table 1   39:whole table Please change all brackets to [] as all fall under the very likely range
Agreed change with TSU and Bureau
Chapter 2 CCB 2.3, Table 1   39:Table Row 1, Column 1: Add ‘Reference’ before Period to read: “Reference Period”.
Row 1, Column 5: Replace ‘Principle use of this period in reports’ with “Principal use of this period in this report and previous reports”
Row 3, Column 5 (R3,C5): Replace “Attributable warming in AR6 WGI” with “Attributable warming assessment period in WGI AR6.”

R4,C5: Replace 'Energy budget constraints in AR6 on ECS and TCR' with “WGI AR6 warming estimate as a line of evidence for energy budget constraints to estimate ECS and TCR.”

R5,C5: Remove “GSAT value used for remaining carbon budgets (RCBs) in SR1.5” to just show “Warming to date in SR1.5” (the first sentence)

R6, C5: Swap AR5 and WG1 to read: “Warming to date in WGI AR5.”

R8, C5: Add following: “This difference is used to report in this box the implications of the AR6 historical global surface temperature assessment in a way that is directly comparable to the AR5 estimate.” To read in full: “Warming to recent past in AR5 WGI. This difference is used to report in this box the implications of the AR6 historical global surface temperature assessment in a way that is directly comparable to the AR5 estimate.”

R9,C5: Include AR6 and a full stop to read: Warming to reference period recommended by WMO for national-level data sets used for climate change assessment (included in the AR6 WGI Atlas).

R10, C5: Replace full sentence with “Warming trend to date in AR5 WGI SPM and AR5 SYR.”

Agreed changes with TSU and Bureau (contact: Anna)

Chapter 2  CCB 2.3  40:18-19 Edit sentence to reflect the following (new in bold):

“… metric being compared (GMST/GSAT, method used to calculate a trend or change between two periods, the exact reference period used) with the best estimates (with the exception of the SR1.5 GSAT estimate) falling between 0.07 and 0.12 °C. The choice of 1850-1900 to 1986-2005 as the basis is due to the widespread use of this period across AR5 and SR1.5 in several contexts. The AR6-assessed GMST warming……”

Agreed changes with TSU and Bureau

Chapter 2  CCB 2.3 Figure 1  42:11 Please replace ‘headline’ with ‘assessment’ to read “… various steps from AR5 assessment warming-to-date…”

Agreed change with TSU and Bureau

Chapter 2  2.3.1.1.3  43 : 24 Replace “Junod & Christy, 2019” by “Junod and Christy, 2020”

Chapter 2  2.3.1.1.3  43 : 52 Replace “Sun et al., 2018” by “X. Sun et al., 2018”

Chapter 2  2.3.1.1.3  44 : 2 Replace “Shi et al., 2019” by “Z. Shi et al., 2019”

Chapter 2  2.3.1.1.3  44 : 10 Replace “Simmons et al., 2010a” by “Simmons et al., 2010”

Chapter 2  2.3.1.1.3  44 : 13; Table 2.3 Replace “Kadow et al. (2020a)” by “Kadow et al. (2020)”

Chapter 2  2.3.1.1.3  45 : 7 Replace “Huang et al., 2019” by “B. Huang et al., 2019”

Chapter 2  2.3.1.1.3  45 : 15 Replace “Kobayashi et al., 2015” by “Harada et al., 2016”

Chapter 2  2.3.1.1.3  45 : 40 Replace “Santer et al., (2008a)” by “Santer et al. (2008)”

Chapter 2  2.3.1.2.1  47 : 18-19 Replace “Zhou et al., 2020” by “Zhou et al., 2020b”

Chapter 2  2.3.1.2.1  47 : 47 Replace “Steiner et al., 2013, 2020a” by “Steiner et al., 2013, 2020”

Chapter 2  2.3.1.2.1  47 : 49 Replace “Steiner et al., 2020a” by “Steiner et al., 2020”

Chapter 2  2.3.1.2.1  48 : 31 Replace “Xian and Homeyer, 2018” by “Xian and Homeyer, 2019”

Chapter 2  2.3.1.2.1  48 : 38 Replace “Scherllin-Pirscher et al., 2020” by “Scherllin-Pirscher et al., 2021”
Chapter 2 2.3.2.1.1 65:8  References: (Hansen E. et al., 2013; King et al., 2018; Renner et al., 2014; Rösel et al., 2018; Spreen et al., 2020), are not all in chronological order.

Chapter 2 2.3.2.1.1 65 : 6-7  Replace “King et al., 2018a” by “King et al., 2018”

Chapter 2 2.3.2.4.1 69:35  Insert comma after ‘et al.’

Chapter 2 2.3.2.4.1 Figure 2.24 Caption 70: 16-17 Issue: Error in data range – was not updated to reflect that time series is for 1992-2020 for both ice sheets.


Note from Author (Sharon Smith): The link to this updated version of the data set is https://doi.org/10.5285/77B64C55-7166-4A06-9DEF-2E400398E452 Antarctic and Greenland Ice Sheet mass balance 1992-2020 for IPCC AR6

This is not the IMBIE Consortium link where the ice sheet data resides (and link used for FAIR data table) and that site has not been updated yet. I’m assuming it will eventually be updated with this new dataset but for now it seems to be a data set specifically updated for AR6 with its own doi number. I’m not sure whether this link should go in the figure caption itself or FAIR data table.

Chapter 2 2.3.2.4.2 70:50 and reference list “Wilson et al. 2018” reference incorrect. Wrong Wilson et al. 2018 paper, please replace with:


And remove currently existing Wilson et al. 2018 from reference list (Glacial lakes of central and Patagonian Andes) as it is no longer cited.

Chapter 2 2.3.2.4.2 70 : 50 Change reference details for “Wilson et al., 2018” to https://www.nature.com/articles/s41586-018-0501-8

Chapter 2 2.3.3.1 73:51 Insert comma after ‘et al.’

Chapter 2 2.3.3.2 75 : 11 Replace “Unesco, 1981” by “UNESCO/ICES/SCOR/IAPSO, 1981”

Chapter 2 2.3.3.3 78 : 21 Replace “Kemp et al., 2018b” by “Kemp et al., 2018”

Chapter 2 2.3.3.3 78 : 34 Replace “Kopp et al., 2016a; Kemp et al., 2018a” by “Kopp et al., 2016; Kemp et al., 2018”

Chapter 2 2.3.3.3 79:8-10 Please replace entire opening sentence with:

Based on the ensemble approach of Palmer et al (2021) and an updated WCRP (2018) assessment (Figure 2.28) GMSL rose at a rate of 1.32 [0.58 to 2.06] mm yr⁻¹ for the period 1901–1971, increasing to 1.87 [0.82 to 2.92] mm yr⁻¹ between 1971 and 2006, and further increasing to 3.69 [3.21 to 4.17] mm yr⁻¹ for 2006–2018 (high confidence).

Original sentence for search accuracy: “Based on the ensemble approach of Palmer et al (2021) and an updated WCRP (2018) assessment (Figure 2.28) GMSL rose at a
| Chapter 2 | 2.3.3.3 | 79:13 | Insert comma after ‘et al.’ |
| Chapter 2 | 2.3.3.4 | 80 : 24 | Replace “Worthington et al., 2020” by “Worthington et al., 2021” |
| Chapter 2 | 2.3.3.4.2 | 81 : 32 | Replace “Yang H et al., 2020” by “Yang et al., 2020” |
| Chapter 2 | 2.3.3.4.2 | 81 : 35 | Replace “Y.-L. Wang et al., 2016” by “Y.-L. Wang et al., 2016” |
| Chapter 2 | 2.3.3.4.2 | 81 : 52 | Replace “Liu et al. (2015)” by “Q.-Y. Liu et al. (2015)” |
| Chapter 2 | 2.3.3.4.2 | 81 : 54 | Replace “Li et al., 2018” by “M. Li et al., 2018” |
| Chapter 2 | 2.3.3.5 | 82 : 28 | Replace “Turner et al., 2018” by “Kirtland Turner et al., 2018” |
| Chapter 2 | 83:5 | Insert comma after ‘et al.’ |
| Chapter 2 | 83:15 | Insert comma after ‘et al.’ |
| Chapter 2 | 84:22 | Insert comma after ‘et al.’ |
| Chapter 2 | 2.3.3.6 | 84 : 31; 84 : 36-37 | Replace “Naqvi et al., 2018” by “Al-Said et al., 2018” |
| Chapter 2 | 2.3.4 | 85 : 11-12 | Replace “Graven et al., 2013b” by “Graven et al., 2013” |
| Chapter 2 | 2.3.4 | 85 : 14 | Replace “Graven et al., 2013a” by “Graven et al., 2013” |
| Chapter 2 | 2.3.4.2.1 | 86 : 22-23 | Replace “Racault et al., 2017b” by “Racault et al., 2017a” |
| Chapter 2 | 2.3.4.2 | 86 : 32 | Replace “Santer et al (2008c)” by “Santer et al. (2008)” |
| Chapter 2 | 87:8 | Insert comma after ‘et al.’ |
| Chapter 2 | 2.3.4.2.3 | 89 : 34 | Replace “Donat et al., 2013” by “Donat et al., 2013b” |
| Chapter 2 | 2.3.4.3.1 | 90 : 6 | Replace “Sha, 2019” by “Sha et al., 2019” |
| Chapter 2 | 2.3.4.3.2 | 90 : 11 | Replace “Marquier et al., 2017c” by “Marquier et al., 2017” |
| Chapter 2 | 2.3.4.3.1 | 90 : 39 | Replace “Harsch et al., 2009b” by “Harsch et al., 2009” |
| Chapter 2 | 2.3.4.3.3 | 91 : 14 | Replace “(Liu et al., 2015c). Pan et al. (2018a)” by “(Liu et al., 2015c). Pan et al. (2018a)” |
| Chapter 2 | 2.3.4.3.3 | 91 : 18-19 | Replace “Chen et al., 2019” by “C. Chen et al., 2019” |
| Chapter 2 | 2.3.4.3.3 | 91 : 27 | Replace “Chen et al., 2019” by “X. Chen et al., 2019” |
| Chapter 2 | 2.3.4.3.3 | 91 : 29 | Replace “Wang et al., 2018” by “Y. Wang et al., 2018” |
| Chapter 2 | 2.3.4.3.3 | 91 : 33 | Replace “Pan et al., 2018a” by “T.-Y. Pan et al., 2018” |
| Chapter 2 | 2.3.4.3.3 | 91 : 40 | Replace “Santer et al (2008c)” by “Santer et al. (2008)” |
| Chapter 2 | 2.3.4.3.3 | 91 : 51-52 | Replace “Pan et al. (2018a)” by “N. Pan et al. (2018)” |
| Chapter 2 | 2.3.4.3.3 | 92 : 3 | Replace “Wang et al., 2016” by “S. Wang et al., 2016” |
| Chapter 2 | CCB2.4 | 94 : 24 | Replace “Huang X. et al., 2019” by “X. Huang et al., 2019a”. Replace “Zhang et al., 2019” by “R. Zhang et al., 2019”. |
| Chapter 2 | 2.4.1.2 | 96 : 44-46 | Replace “Lee J. et al., 2019a” by “J. Lee et al., 2019” |
| Chapter 2 | 2.4.1.2 | 96 : 47 | Replace “Jones et al., 2016b” by “J.M. Jones et al., 2016” |
| Chapter 2 | 2.4.1.2 | 96 : 49 | Replace “IPCC, 2019b” by “IPCC, 2019” |
| Chapter 2 | 99:27 | Insert comma after ‘et al.’ |
| Chapter 2 | 2.4.3 | 100 : 9 | Replace “Abram et al., 2020a” by “Abram et al., 2020b” |
| Chapter 2 | 2.4.3 | 100 : 13 | Replace “Abram et al., 2020b” by “Abram et al., 2020a” |
| Chapter 2 | 2.4.4 | 101 : 11 | Replace “Svendsen et al., 2014” by “Svendsen et al., 2014a” |
| Chapter 2 | 2.4.5 | 102 : 4 | Replace “Han et al., 2013” by “Han et al., 2014” |
| Chapter 2 | 2.4.5 | 102 : 45-46 | Replace “Wang et al., 2017b” by “J. Wang et al., 2017b” |
| Chapter 2 | 2.4.5 | 102 : 45-46 | Replace “Svendsen et al., 2014a” by “Svendsen et al., 2014b” |
| Chapter 2 | 2.4.5 | 102 : 49 | Replace “Wang J. et al., 2017b” by “J. Wang et al., 2017b” |
| Chapter 2 | 2.4.5 | 103 : 6 | Replace “Svendsen et al., 2014a” by “Svendsen et al., 2014b” |
| Chapter 2 | 112 :39 | Reference “Blunden and Arndt, 2019” still included in list, but not cited anymore. Propose to remove from reference list. |
| Chapter 2 | 131 :44 | TSU (Robin) previously notified of following reference update: Hugonnet et al. 2021 is now published, but here still listed as in press at the end of the reference. see link: https://www.nature.com/articles/s41586-021-03436-z |
| Chapter 2 | Figure 2.1 | 170 | replace with updated visual roadmap, as all visual roadmaps have been harmonised (to have a set with a consistent visual identity. This does not alter the content of the chapter.) |
| Chapter 2 | Figure 2.10 | 180:1 | **Changes made: New version uploaded to figure manager.**  
Minor changes made to Figure 2.10  
The y-axis for the inset plot ("rate of change anthropogenic ERF") was stretched accidentally. Not all data fit into the resulting plot. This has now been rectified. |
| Chapter 2 | Figures CCB 2.3 Figure 1 | 183:1 | **Changes made: New version uploaded to figure manager.**  
Several minor amendments made to Cross-Chapter Box 2.3 Figure 1:  
Replaced second “changes” with “estimates” in main title.  
Replaced ‘Anomaly from 1850-1900 mean’ with ‘Anomaly relative to 1961-1990 mean’ on vertical main axis.  
Flipped the colours of AR5 and AR6 in the legend to reflect the data shown. AR5 should be Orange and AR6 should be Blue.  
Changed “AR5-AR6” to “AR6-AR5” in plot and right hand vertical axis.  
Changed ‘headline’ to ‘assessment’ in (a). |
| Chapter 2 | Figures CCB 2.3 Figure 1 | 183:4 | Change 'headline' to 'assessment' to read:  
“… various steps from AR5 assessment warming-to-date…” |
| Chapter 2 | Figure 2.12 | 184:7 | Add ‘(c)’ so that the sentence reads:  
“… no the absolute trends. (b) (c) Trends in temperature at various…” |
| Chapter 2 | Figures, Figure 2.26 | 198:3 | Remove capital letter from ‘Ocean’ OR add capital letters to ‘Heat Content’ to read either ‘Ocean Heat Content (OHC)’ or ‘ocean heat content (OHC)’. |
| Chapter 2 | Figures, Figure 2.26 | 198:6/7 | Remove 'all of which are broadly similar’ to read:  
“(a) but for 0-2000 m depths only and reflecting the broad range of available estimates over this period.” |
| Chapter 2 | Figures, Figure 2.30 | 202: 1 | **Changes made: New version uploaded to figure manager.**  
Seasonal Amplitude (ppm) Barrow data updated due to an error in the original processing. |
Chapter 2: Changing state of the climate system

Coordinating Lead Authors:
Sergey K. Gulev (Russian Federation), Peter W. Thorne (Ireland/United Kingdom)

Lead Authors:
Jinho Ahn (Republic of Korea), Frank J. Dentener (EU/The Netherlands), Catia M. Domingues (Australia, United Kingdom/Brazil), Sebastian Gerland (Norway/Germany), Daoyi Gong (China), Darrell S. Kaufman (United States of America), Hyacinth C. Nnamchi (Nigeria, Germany/Nigeria), Johannes Quaas (Germany), Juan A. Rivera (Argentina), Subba Sathyendranath (United Kingdom/Canada), Overseas Citizen of India, United Kingdom), Sharon L. Smith (Canada), Blair Trewin (Australia), Karina von Schuckmann (France/Germany), Russell S. Vose (United States of America),

Contributing Authors:
Guðfinna Aðalgeirsdóttir (Iceland), Samuel Albani (Italy), Richard P. Allan (United Kingdom), Richard A. Betts (United Kingdom), Lea Beusch (Switzerland), Kenfie Beyene (Ethiopia), Jason E. Box (Denmark/United States of America), Denise Breitburg (United States of America), Kevin D. Burke (United States of America), Michael P. Byrne (United Kingdom/Ireland), John A. Church (Australia), Sloane Coats (United States of America), Naftali Cohen (United States of America), William Collins (United Kingdom), Owen R. Cooper (United States of America), Pedro Di Nezio (United States of America), Fabio Boeira Dias (Finland/Brazil), Ed J. Dlugokencky (United States of America), Timothy Dunkerton (United States of America), Paul J. Durack (United States of America), Tamsin L. Edwards (United Kingdom), Veronika Eyring (Germany), Chris Fairall (United States of America), Vitali Fioletov (Canada), Piers Forster (United Kingdom), Gavin L. Foster (United Kingdom), Baylor Fox-Kemper (United States of America), Qiang Fu (United States of America), Jan S. Fuglestad (Norway), John C. Fyfe (Canada), Marie-José Gaillard (Sweden/Switzerland, Sweden), Joelle Gergis (Australia), Nathan P. Gillett (Canada), Hans Gleisner (Denmark/Sweden), Nadine Gobron (EU/France), Nicholas R. Golledge (New Zealand/United Kingdom), Bradley Hall (United States of America), Ed Hawkins (United Kingdom), Alan M. Haywood (United Kingdom), Armand Hernandez (Spain), Forrest M. Hoffman (United States of America), Yongyun Hu (China), Dale F. Hurst (United States of America), Masao Ishii (Japan), Samuel Jaccard (Switzerland), Dabang Jia (China), Christopher Jones (United Kingdom), Bror Jönsson (United Kingdom/Sweden), Andreas Kääb (Norway/Germany), Ralph Keeling (United States of America), Noel S. Keenlyside (Norway/Australia, United Kingdom), Joan Kennedy (United Kingdom), Elizabeth Kent (United Kingdom), Nichol S. Khan (Hong Kong, China/United States of America), Wolfgang Kienzle (Germany), Stefan Kinne (Germany), Robert E. Kopf (United States of America), Svitlana Krakovska (Ukraine), Elmar Kriegl (Germany), Gerhard Kriinner (France/Germany, France), Natalie Krivova (Germany), Paul B. Krummel (Australia), Werner L. Kutsch (EU/Germany), Ron Kwok (United States of America), Florian Ladstädter (Austria), Peter Landschützer (Germany/Austria), June-Yi Lee (Republic of Korea), Andrew Lenton (Australia), Lisa A. Levin (United States of America), Daniel J. Lunt (United Kingdom), Jochem Marotzke (Germany), Gareth J. Marshall (United Kingdom), Robert A. Massom (Australia), Katja Matthes (Germany), Damon H. Matthews (Canada), Thorsten Mauritsen (Sweden/Denmark), Gerard D. McCarthy (Ireland), Erin L. McClymont (United Kingdom), Shayne McGregor (Australia), Jerry F. McManus (United States of America), Walter N. Meier (United States of America), Alan Mix (United States of America), Olaf Morgenstern (New Zealand/Germany), Lawrence R. Mudryk (Canada), Jens Mühle (United States of America/Germany), Dirk Notz (Germany), Lisa C. Orme (Ireland/United Kingdom), Scott M. Osprey (United Kingdom), Matthew D. Palmer (United Kingdom), Anna Pirani (Italy/United Kingdom, Italy), Chris Polashenski (United States of America), Elvira Poloczanska (Australia/United Kingdom), Anthony Richardson (Australia), Belén Rodríguez-Fonseca (Spain), Joeri Rogelj (United Kingdom/Belgium), Steven K. Rose (United States of America), Yair Rosenthal (United States of America/Israel, United States of America, United Kingdom)
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2.5 Final remarks

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Acknowledgements

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Executive Summary

Chapter 2 assesses observed large-scale changes in climate system drivers, key climate indicators and principal modes of variability. Chapter 3 considers model performance and detection/attribution, and Chapter 4 covers projections for a subset of these same indicators and modes of variability. Collectively, these chapters provide the basis for later chapters, which focus upon processes and regional changes. Within Chapter 2, changes are assessed from in situ and remotely sensed data and products and from indirect evidence of longer-term changes based upon a diverse range of climate proxies. The time-evolving availability of observations and proxy information dictate the periods that can be assessed. Wherever possible, recent changes are assessed for their significance in a longer-term context, including target proxy periods, both in terms of mean state and rates of change.

Changes in Climate System Drivers

Climate system drivers lead to climate change by altering the Earth’s energy balance. The influence of a climate driver is described in terms of its effective radiative forcing (ERF), measured in W m\(^{-2}\). Positive ERF values exert a warming influence and negative ERF values exert a cooling influence (Chapter 7).

Present-day global concentrations of atmospheric carbon dioxide (CO\(_2\)) are at higher levels than at any time in at least the past two million years (high confidence). Changes in ERF since the late 19th century are dominated by increases in concentrations of greenhouse gases and trends in aerosols; the net ERF is positive and changing at an increasing rate since the 1970s (medium confidence). [2.2, 7.3, 7.3]

Change in ERF from natural factors since 1750 is negligible in comparison to anthropogenic drivers (very high confidence). Solar activity since 1900 was high but not exceptional compared to the past 9000 years (high confidence). The average magnitude and variability of volcanic aerosol forcing since 1900 have not been unusual compared to the past 2500 years (medium confidence). [2.2.1, 2.2.2]

In 2019, concentrations of CO\(_2\), methane (CH\(_4\)) and nitrous oxide (N\(_2\)O) reached levels of 409.9 (±0.4) ppm, 1866.3 (±3.3) ppb and 332.1 (±0.4) ppb, respectively. Since 1850, these well-mixed greenhouse gases (GHGs) have increased at rates that have no precedent on centennial time scales in at least the past 800,000 years. Concentrations of CO\(_2\), CH\(_4\) and N\(_2\)O increased from 1750 to 2019 by 131.6 ± 2.9 ppm (47.3%), 1137 ± 10 ppb (156%), and 62 ± 6 ppb (23.0%) respectively. These changes are larger than those between glacial and interglacial periods over the last 800,000 years for CO\(_2\) and CH\(_4\) and of comparable magnitude for N\(_2\)O (very high confidence). The best estimate of the total ERF from CO\(_2\), CH\(_4\) and N\(_2\)O in 2019 relative to 1750 is 2.9 W m\(^{-2}\), an increase of 12.5 % from 2011. ERF from halogenated components in 2019 was 0.4 W m\(^{-2}\), an increase of 3.5% since 2011. [2.2.3, 2.2.4, 7.3.2]

Tropospheric aerosol concentrations across the Northern Hemisphere mid-latitudes increased from 1700 to the last quarter of the 20th century, but have subsequently declined (high confidence). Aerosol optical depth (AOD) has decreased since 2000 over Northern Hemisphere mid-latitudes and Southern Hemisphere mid-latitude continents, but increased over South Asia and East Africa (high confidence). These trends are even more pronounced in AOD from sub-micrometre aerosols for which the anthropogenic contribution is particularly large. The best-estimate of aerosol ERF in 2019 relative to 1750 is –1.1 W m\(^{-2}\). [2.2.6, 7.3.3]

Changes in other short-lived gases are associated with an overall positive ERF (medium confidence). Stratospheric ozone has declined between 60°S and 60°N by 2.2% from the 1980s to 2014–2017 (high confidence). Since the mid-20th century, tropospheric ozone has increased by 30–70% across the Northern Hemisphere (medium confidence). Since the mid-1990s, free tropospheric ozone increases were 2–7% per decade in the northern mid-latitudes (high confidence), 2–12% per decade in the tropics (high confidence) and <5% per decade in southern mid-latitudes (medium confidence). The best estimate of ozone column ERF (0.5 W m\(^{-2}\) relative to 1750) is dominated by changes in tropospheric ozone. Due to discrepancies in satellite and in situ records, there is low confidence in estimates of stratospheric water vapour change. [2.2.5, 7.3.2]
Changes in Key Indicators of Global Climate Change

Observed changes in the atmosphere, oceans, cryosphere and biosphere provide unequivocal evidence of a world that has warmed. Over the past several decades, key indicators of the climate system have increased at levels unseen in centuries to millennia, and are changing at rates unprecedented in at least the last 2000 years (high confidence). The increase in GMST since the mid-19th century was preceded by a slow decrease that began in the mid-Holocene (around 6500 years ago) (medium confidence). {Cross-Chapter Box 2.1}

GMST increased by 0.85 [0.69 to 0.95] °C between 1850–1900 and 1995–2014 and by 1.09 [0.95 to 1.20] °C between 1850–1900 and 2011–2020. From 1850–1900 to 2011–2020, the temperature increase over land (1.59 [1.34 to 1.83] °C) has been faster than over the oceans (0.88 [0.68 to 1.01] °C). Over the last 50 years, observed GMST has increased at a rate unprecedented in at least the last 2000 years (medium confidence). The increase in GMST since the mid-19th century was preceded by a slow decrease that began in the mid-Holocene (around 6500 years ago) (medium confidence). {2.3.1.1, Cross-Chapter Box 2.1}

Changes in GMST and global surface air temperature (GSAT) over time differ by at most 10% in either direction (high confidence), and the long-term changes in GMST and GSAT are presently assessed to be identical. There is expanded uncertainty in GSAT estimates, with the assessed change from 1850–1900 to 1995–2014 being 0.85 [0.67 to 0.98] °C. {Cross-Chapter box 2.3}

The troposphere has warmed since at least the 1950s, and it is virtually certain that the stratosphere has cooled. In the Tropics, the upper troposphere has warmed faster than the near-surface since at least 2001, the period over which new observation techniques permit more robust quantification (medium confidence). It is virtually certain that the tropopause height has risen globally over 1980–2018, but there is low confidence in the magnitude. {2.3.1.2}

Changes in several components of the global hydrological cycle provide evidence for overall strengthening since at least 1980 (high confidence). However, there is low confidence in comparing recent changes with past variations due to limitations in paleoclimate records at continental and global scales. Global land precipitation has likely increased since 1950, with a faster increase since the 1980s (medium confidence). Near-surface specific humidity has increased over both land (very likely) and the oceans (likely) since at least the 1970s. Relative humidity has very likely decreased over land areas since 2000. Global total column water vapour content has very likely increased during the satellite era.

Observational uncertainty leads to low confidence in global trends in precipitation minus evaporation and river runoff. {2.3.1.3}

Several aspects of the large-scale atmospheric circulation have likely changed since the mid-20th century, but limited proxy evidence yields low confidence in how these changes compare to longer-term climate. The Hadley circulation has very likely widened since at least the 1980s, and extratropical storm tracks have likely shifted poleward in both hemispheres. Global monsoon precipitation has likely increased since the 1980s, mainly in the Northern Hemisphere (medium confidence). Since the 1970s, near-surface winds have likely weakened over land. Over the oceans, near-surface winds likely strengthened over 1980–2000, but divergent estimates lead to low confidence in the sign of change thereafter. It is likely that the northern stratospheric polar vortex has weakened since the 1980s and experienced more frequent excursions toward Eurasia. {2.3.1.4}

Current Arctic sea ice coverage levels are the lowest since at least 1850 for both annual mean and late-summer values (high confidence) and for the past 1000 years for late-summer values (medium confidence). Between 1979 and 2019, Arctic sea ice area has decreased in both summer and winter, with sea ice becoming younger, thinner and more dynamic (very high confidence). Decadal means for Arctic sea ice area decreased from 6.23 million km² in 1979–1988 to 3.76 million km² in 2010–2019 for September and
from 14.52 to 13.42 million km$^2$ for March. Antarctic sea ice area has experienced little net change since 1979 (high confidence), with only minor differences between sea ice area decadal means for 1979–1988 (2.04 million km$^2$ for February, 15.39 million km$^2$ for September) and 2010–2019 (2.17 million km$^2$ for February, 15.75 million km$^2$ for September). {2.3.2.1}

Changes across the terrestrial cryosphere are widespread, with several indicators now in states unprecedented in centuries (high confidence). Reductions in spring snow cover extent have occurred across the Northern Hemisphere since at least 1978 (very high confidence). With few exceptions, glaciers have retreated since the second half of the 19th century and continued to retreat with increased rates since the 1990s (very high confidence); this behaviour is unprecedented in at least the last 2000 years (medium confidence). Greenland Ice Sheet mass loss has increased substantially since 2000 (high confidence). The Greenland Ice Sheet was smaller than at present during the Last Interglacial period (high confidence) and the mid-Holocene (high confidence). The Antarctic Ice Sheet (AIS) lost mass between 1992 and 2020 (very high confidence), with an increasing rate of mass loss over this period (medium confidence). Although permafrost persists in areas of the Northern Hemisphere where it was absent prior to 3000 years ago, increases in temperatures in the upper 30 m over the past three to four decades have been widespread (high confidence). {2.3.2}

Global mean sea level (GMSL) is rising, and the rate of GMSL rise since the 20th century is faster than over any preceding century in at least the last three millennia (high confidence). Since 1901, GMSL has risen by 0.20 [0.15–0.25] m, and the rate of rise is accelerating. Further back in time, there is medium confidence that GMSL was within –3.5 to 0.5 m (very likely) of present during the mid-Holocene (6000 years ago), 5 to 10 m (likely) higher during the Last Interglacial (125,000 years ago), and 5 to 25 m (very likely) higher during the mid-Pliocene Warm Period (MPWP) (3.3 million years ago). {2.3.3.3}

Recent ocean changes are widespread, and key ocean indicators are in states unprecedented for centuries to millennia (high confidence). Since 1971, it is virtually certain that global ocean heat content has increased for the upper (0–700 m) layer, very likely for the intermediate (700–2000 m) layer and likely below 2000 m, and is currently increasing faster than at any point since at least the last deglacial transition (18–11 thousand years ago) (medium confidence). It is virtually certain that large-scale near-surface salinity contrasts have intensified since at least 1950. The Atlantic Meridional Overturning Circulation (AMOC) was relatively stable during the past 8000 years (medium confidence) but declined during the 20th century (low confidence). Ocean pH has declined globally at the surface over the past four decades (virtually certain) and in all ocean basins in the ocean interior (high confidence) over the past 2–3 decades. Deoxygenation has occurred in most open ocean regions during the mid 20th to early 21st centuries (high confidence), with decadal variability (medium confidence). Oxygen minimum zones are expanding at many locations (high confidence). {2.3.3}

Changes in the marine biosphere are consistent with large-scale warming and changes in ocean geochemistry (high confidence). The ranges of many marine organisms are shifting towards the poles and towards greater depths (high confidence), but a minority of organisms are shifting in the opposite directions. This mismatch in responses across species means that the species composition of ecosystems is changing (medium confidence). At multiple locations, various phenological metrics for marine organisms have changed in the last 50 years, with the nature of the changes varying with location and with species (high confidence). In the last two decades, the concentration of phytoplankton at the base of the marine food web, as indexed by chlorophyll concentration, has shown weak and variable trends in low and mid-latitudes and an increase in high latitudes (medium confidence). Global marine primary production decreased slightly from 1998–2018, with increasing production in the Arctic (medium confidence). {2.3.4.2}

Changes in key global aspects of the terrestrial biosphere are consistent with large-scale warming (high confidence). Over the last century, there have been poleward and upslope shifts in the distributions of many land species (very high confidence) as well as increases in species turnover within many ecosystems (high confidence). Over the past half century, climate zones have shifted poleward, accompanied by an increase in the length of the growing season in the Northern Hemisphere extratropics and an increase in the amplitude of the seasonal cycle of atmospheric CO$_2$ above 45°N (high confidence). Since the early 1980s, there has been a global-scale increase in the greenness of the terrestrial surface (high confidence). {2.3.4.1,
During the Mid-Pliocene warm period (MPWP, 3.3–3.0 million years ago) slowly changing large-scale indicators reflect a world that was warmer than present, with CO₂ similar to current levels. CO₂ levels during the MPWP were similar to present for a sustained period, within a range of 360–420 ppm (medium confidence). Relative to the present, GMST, GMSL and precipitation rate were all higher, the Northern Hemisphere latitudinal temperature gradient was lower, and major terrestrial biomes were shifted northward (very high confidence). There is high confidence that cryospheric indicators were diminished and medium confidence that the Pacific longitudinal temperature gradient weakened and monsoon systems strengthened. [2.3, Cross-Chapter Box 2.4, 9.6.2]

Inferences from past climate states based on proxy records can be compared with climate projections over coming centuries to place the range of possible futures into a longer-term context. There is medium confidence in the following mappings between selected paleo periods and future projections: During the Last Interglacial, GMST is estimated to have been 0.5°C–1.5°C warmer than the 1850–1900 reference for a sustained period, which overlaps the low end of the range of warming projected under SSP1-2.6, including its negative-emissions extension to the end of the 23rd century [1.0–2.2] °C. During the mid-Pliocene Warm Period, the GMST estimate [2.5–4.0] °C is similar to the range projected under SSP2-4.5 for the end of the 23rd century [2.3–4.6] °C. GMST estimates for the Miocene Climatic Optimum [5–10] °C and Early Eocene Climatic Optimum [10–18] °C, about 15 and 50 million years ago, respectively, overlap with the range projected for the end of the 23rd century under SSP5-8.5 [6.6–14.1] °C. [Cross-Chapter Box 2.1, 2.3.1, 4.3.1.1, 4.7.1.1]

Changes in Modes of Variability

Since the late 19th century, major modes of climate variability show no sustained trends but do exhibit fluctuations in frequency and magnitude at inter-decadal time scales, with the notable exception of the Southern Annular Mode, which has become systematically more positive (high confidence). There is high confidence that these modes of variability have existed for millennia or longer, but low confidence in detailed reconstructions of most modes prior to direct instrumental records. Both polar annular modes have exhibited strong positive trends toward increased zonality of midlatitude circulation over multi-decadal periods, but these trends have not been sustained for the Northern Annular Mode since the early 1990s (high confidence). For tropical ocean modes, a sustained shift beyond multi-centennial variability has not been observed for El Niño–Southern Oscillation (medium confidence), but there is low evidence and low agreement about the long-term behaviour of other tropical ocean modes. Modes of decadal and multi-decadal variability over the Pacific and Atlantic oceans exhibit no significant trends over the period of observational records (high confidence). [2.4]
2.1 Introduction

This chapter assesses the evidence-basis for large-scale past changes in selected components of the climate system. As such, it combines much of the assessment performed in Chapters 2 through 5 of the Fifth Assessment Report (AR5) WGI contribution (IPCC, 2013) that, taken together, supported a finding of unequivocal recent warming of the climate system. The Sixth Assessment Report (AR6) WGI Report structure differs substantially from that in AR5 (Section 1.1.2). The current chapter focuses upon observed changes in climate system drivers and changes in key selected large-scale indicators of climate change and in important modes of variability (Cross-Chapter Box 2.2), which allow for an assessment of changes in the global climate system in an integrated manner. This chapter is complemented by Chapters 3 and 4, which respectively consider model assessment / detection and attribution, and future climate projections for subsets of these same indicators and modes. It does not consider changes in observed extremes, which are assessed in Chapter 11. The chapter structure is outlined in the visual abstract (Figure 2.1).

[START FIGURE 2.1 HERE]

Figure 2.1: Chapter 2 visual abstract of contents.

[END FIGURE 2.1 HERE]

Use is made of paleoclimate, in situ, ground- and satellite-based remote sensing, and reanalysis data products where applicable (Section 1.5). All observational products used in the chapter are detailed in Annex I, and information on data sources and processing for each figure and table can be found in the associated chapter table 2.SM.1 available as an electronic supplement to the chapter. Use of common periods ranging from 56 million years ago through to the recent past is applied to the extent permitted by available data (Section 1.4.1 and Cross-Chapter Box 2.1). In all cases, the narrative proceeds from as far in the past as the data permit through to the present. Each sub-section starts by highlighting the key findings from AR5 and any relevant AR6-cycle special reports (SROCC, SR1.5, SRCCL), and then outlines the new evidence-basis arising from a combination of: i) new findings reported in the literature, including new datasets and new versions of existing datasets; and ii) recently observed changes, before closing with a new summary assessment.

Trends, when calculated as part of this assessment, have wherever possible been calculated using a common approach following that adopted in Box 2.2 of Chapter 2 of AR5 (Hartmann et al., 2013). In addition to trends, consideration is also made of changes between various time slices / periods in performing the assessment (Section 1.4.1 and Cross-Chapter Box 2.1). Statistical significance of trends and changes are assessed at the two-tailed 90% confidence (very likely) level unless otherwise stated. Limited use is also made of published analyses that have employed a range of methodological choices. In each such case the method / metric is stated.

There exist a variety of inevitable and, in some cases, irreducible uncertainties in performing an assessment of the observational evidence for climate change. In some instances, a combination of sources of uncertainty is important. For example, the assessment of global surface temperature over the instrumental record in Section 2.3.1.1.3 considers a combination of observational-dataset and trend-estimate uncertainties. Furthermore, estimates of parametric uncertainty are often not comprehensive in their consideration of all possible factors and, when such estimates are constructed in distinct manners, there are often significant limitations to their direct comparability (Hartmann et al., 2013; their Box 2.1).
Cross-Chapter Box 2.1:  Paleoclimate Reference Periods in the Assessment Report

**Contributing Authors:** Darrell Kaufman (USA), Kevin Burke (USA), Samuel Jaccard (Switzerland), Chris Jones (UK), Wolfgang Kiessling (Germany), Daniel Lunt (UK), Olaf Morgenstern (New Zealand), John W. Williams (USA)

Over the long evolution of the Earth’s climate system, several periods have been extensively studied as examples of distinct climate states. This Cross-Chapter Box places multiple paleoclimate reference periods into the unifying context of Earth’s long-term climate history, and points to sections in the report with additional information about each period. Other reference periods, including those of the industrialized era, are described in Section 1.4.1.

The reference periods represent times that were both colder and warmer than present, and periods of rapid climate change, many with informative parallels to projected climate (Cross-Chapter Box 2.1, Table 1). They are used to address a wide variety of questions related to natural climate variations in the past (FAQ 1.3).

Most of them are used as targets to evaluate the performance of climate models under different climate forcings (Section 3.8.2), while also providing insight into the ocean-atmospheric circulation changes associated with various radiative forcings and geographical changes.

Global Mean Surface Temperature (GMST) is a key indicator of the changing state of the climate system. Earth’s mean temperature history during the current geological era (Cenozoic, beginning 66 Ma) can be broadly characterized as follows (Cross-Chapter Box 2.1, Figure 1): (1) transient warming during the first 15 million years of the Cenozoic, punctuated by the Paleocene-Eocene Thermal Maximum; (2) a long-term cooling over tens of millions of years beginning around 50 Ma, driven by (among other factors) the slow drift of tectonic plates, which drove mountain building, erosion and volcanism, and reconfigured ocean passages, all of which ultimately moved carbon from the atmosphere to other reservoirs and led to the development of the Antarctic Ice Sheet (AIS) about 35–30 Ma; (2) the intensification of cooling by climate feedbacks involving interactions among tectonics, ice albedo, ocean circulation, land cover and greenhouse gases, causing ice sheets to develop in the Northern Hemisphere (NH) by about 3 Ma; (3) glacial-interglacial fluctuations paced by slow changes in Earth’s astronomical configuration (orbital forcing) and modulated by changes in the global carbon cycle and ice sheets on time scales of tens to hundreds of thousands of years, with particular prominence during the last 1 Myr; (4) a transition with both gradual and abrupt shifts from the Last Glacial Maximum to the present interglacial epoch (Holocene), with sporadic ice-sheet breakup disrupting ocean circulation; (5) continued warming followed by minor cooling following the mid-Holocene, with superposed centennial- to decadal-scale fluctuations caused by volcanic activity, among other factors; and (6) recent warming related to the build-up of anthropogenic greenhouse gases (Sections 2.2.3, 3.3.1).

GMST estimated for each of the reference periods based on proxy evidence (Section 2.3.1.1) can be compared with climate projections over coming centuries to place the range of possible futures into a longer-term context (Cross-Chapter Box 2.1, Figure 1). Here, the very likely range of GMST for the warmer world reference periods are compared with the very likely range of GSAT projected for the end the 21st century (2080–2100; Table 4.5 and the likely range for the end of the 23rd century (2300; Table 4.9) under multiple Shared Socioeconomic Pathway (SSP) scenarios. From this comparison, there is medium confidence in the following: GMST estimated for the warmest long-term period of the Last Interglacial about 125 ka (0.5–1.5°C relative to 1850–1900) overlaps with the low end of the range of temperatures projected under SSP1-2.6 including its negative emissions extension to the end of the 23rd century [1.0 to 2.2°C]. GMST estimated for a period of prolonged warmth during the mid-Pliocene Warm Period about 3 Ma [2.5 to 4.0°C] is similar to temperatures projected under SSP2-4.5 for the end of the 23rd century [2.3 to 4.6°C]. GMST estimated for the Miocene Climatic Optimum [5°C to 10°C] and Early Eocene Climatic Optimum [10 to 18°C], about 15 and 50 million years ago, respectively, overlap with the range projected for the end of the 23rd century under SSP5-8.5 [6.6 to 14.1°C].
### Cross-Chapter Box 2.1, Table 1: Paleo reference periods, listed from oldest to youngest. See ‘AR6 sections’ for literature citations related to each ‘Sketch of the climate state.’ See Working Group II (Chapters 1, 2, 3) for citations related to paleontological changes. See Interactive Atlas for simulated climate variables for MPWP, LIG, LGM and MH.

<table>
<thead>
<tr>
<th>Period</th>
<th>Age/year*</th>
<th>Sketch of the climate state (relative to 1850–1900), and model experiment protocols. Values for large-scale climate indicators including global temperature, sea level and atmospheric CO₂ are shown in Figure 2.34.</th>
<th>AR6 sections partial list</th>
</tr>
</thead>
<tbody>
<tr>
<td>Paleocene-Eocene thermal maximum (PETM)</td>
<td>55.9 – 55.7 Ma</td>
<td>A geologically rapid, large-magnitude warming event at the start of the Eocene when a large pulse of carbon was released to the ocean-atmosphere system, decreasing ocean pH and oxygen content. Terrestrial plant and animal communities changed composition, and species distributions shifted poleward. Many deep-sea species went extinct and tropical coral reefs diminished. DeepMIP (Lunt et al., 2017)</td>
<td>2.2.3.1, 2.3.1.1.1, 5.1.2.1, 5.3.1.1</td>
</tr>
<tr>
<td>Early Eocene climatic optimum** (EECO)</td>
<td>53–49 Ma</td>
<td>Prolonged “hothouse” period with atmospheric CO₂ concentration &gt;1000 ppm similar to SSP5-8.5 end-of-century values. Continental positions were somewhat different to present due to tectonic plate movements; polar ice was absent and there was more warming at high latitudes than in the equatorial regions. Neatropical forests grew at 70°S, despite seasonal polar darkness. DeepMIP, About 50 Ma (Lunt et al., 2017; 2021)</td>
<td>2.2.3.1, 2.3.1.1.1</td>
</tr>
<tr>
<td>Miocene climatic optimum** (MCO)</td>
<td>16.9 – 14.7 Ma</td>
<td>Prolonged warm period with atmospheric CO₂ concentrations 400–600 ppm, similar to SSP2-4.5 end-of-century values. Continental geography was broadly similar to modern. At times, Arctic sea ice may have been absent, and the AIS was much smaller or perhaps absent. Peak in Cenozoic reef development. MidMIP, Early and Middle Miocene (Steinthorsdottir et al., 2020)</td>
<td>2.2.3.1, 2.3.1.1.1</td>
</tr>
<tr>
<td>Mid-Pliocene warm period (MPWP)</td>
<td>3.3–3.0 Ma</td>
<td>Warm period when atmospheric CO₂ concentration was similar to present (Cross-Chapter Box 2.4). The Arctic was much warmer, but tropical temperatures were only slightly warmer. Sea level was high. Voyager extended to the northern coastline of the NH continents. Also called, “Pacemanzan warm period.” PMIP4 midPliocene-eoi400, 3.2 Ma (Haywood et al., 2016; 2020)</td>
<td>CCB2.4, 7.4.4.1.2, 7.5.3.3, 8.2.2.2, 9.6.2</td>
</tr>
<tr>
<td>Last Interglacial (LIG)</td>
<td>129–116 ka</td>
<td>Most recent interglacial period, similar to mid-Holocene, but with more pronounced seasonal insolation cycle. Northern high latitudes were warmer, with reduced sea ice. Greenland and West Antarctic Ice Sheets were smaller and sea level was higher. Monsoon was enhanced. Boreal forests extended into Greenland and subtropical animals such as <em>Hippopotamus</em> occupied Britain. Coral reefs expanded latitudinally and contracted equatorially. PMIP4 ligg127k, 127 ka (Otto-Bliesner, et al., 2017; 2021)</td>
<td>2.2.3.2, 2.3.1.1.1, 2.3.3.3, 9.2.2.1, 9.6.2</td>
</tr>
<tr>
<td>Last Glacial Maximum (LGM)</td>
<td>23–19 ka</td>
<td>Most recent glaciation when global temperatures were lower, with greater cooling toward the poles. Ice sheets covered much of North America and northwest Eurasia, and sea level was commensurately lower. Atmospheric CO₂ was lower; more carbon was sequestered in the ocean interior. Precipitation was generally lower over most regions; the atmosphere was drier, and ranges of many plant species contracted into glacial refugia; forest extent and coral reef distribution was reduced worldwide. PMIP4-lgm, 21 ka (Kageyama et al., 2017; 2021)</td>
<td>2.2.3.2, 2.3.1.1.1, 3.3.1.1, 3.8.2.1, 5.1.2.2, 7.4.4.1.2, 7.5.3.1, 8.3.2.4, 9.6.2</td>
</tr>
<tr>
<td>Last deglacial transition (LDT)</td>
<td>18–11 ka</td>
<td>Warming that followed the Last Glacial Maximum, with decreases in the extent of the cryosphere in both polar regions. Sea level, ocean meridional overturning circulation, and atmospheric CO₂ increased during two main steps. Temperate and boreal species ranges expanded northwards. Community turnover was large. Megafauna populations declined or went extinct.</td>
<td>2.2.3.2, 5.1.2.2, 5.3.1.2, 8.6.1, 9.6.2</td>
</tr>
<tr>
<td>Mid-Holocene (MH)</td>
<td>6.5–5.5 ka</td>
<td>Middle of the present interglacial when the CO₂ concentration was similar to the onset of the industrial era, but the orbital configuration led to warming and shifts in the hydrological cycle, especially NH monsoons. Approximate time during the current interglacial and before the onset of major industrial activities when GMST was highest. Biome-scale loss of African grasslands caused by weakened monsoons and collapses of temperate tree populations linked to hydroclimate variability. PMIP4 mid-Holocene, 6 ka (Brierley et al., 2020; Otto-Bliesner et al., 2017)</td>
<td>2.3.1.1.2, 2.3.2.4, 2.3.3.3, 3.3.1.1, 3.8.2.1, 8.3.2.4, 8.6.2.2, 9.6.2</td>
</tr>
<tr>
<td>Last millennium***</td>
<td>850–1850 CE</td>
<td>Climate variability during this period is better documented on annual to centennial scales than during previous reference periods. Climate changes were driven by solar, volcanic, land cover, and anthropogenic forcings, including strong increases in</td>
<td>2.3.1.1.2, 2.3.2.3, 8.3.1.6</td>
</tr>
</tbody>
</table>
cross-sections at decadal scale are stated explicitly when possible.

** The word “optimum” is traditionally used in geosciences to refer to the warmest interval of a geologic period.

*** The terms “Little Ice Age” and “Medieval Warm Period” (or “Medieval Climate Anomaly”) are not used extensively in this report because the timing of these episodes is not well defined and varies regionally. Since AR5, new proxy records have improved climate reconstructions at decadal scale across the last millennium. Therefore, the dates of events within these two roughly defined periods are stated explicitly when possible.

* CE: Common Era; ka: thousands of years ago; Ma: millions of years ago.

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**END CROSS-CHAPTER BOX 2.1, TABLE 1 HERE**

**START CROSS-CHAPTER BOX 2.1, FIGURE 1 HERE**

Cross-Chapter Box 2.1, Figure 1: Global mean surface temperature (GMST) over the past 60 million years relative to 1850–1900 shown on three time scales. Information about each of the nine paleo reference periods (blue font) and sections in AR6 that discuss these periods are listed in Cross-Chapter Box 2.1 Table 1. Grey horizontal bars at the top mark important events. Characteristic uncertainties are based on expert judgement and are representative of the approximate midpoint of their respective time scales; uncertainties decrease forward in time. GMST estimates for most paleo reference periods (Figure 2.34) overlap with this reconstruction, but take into account multiple lines of evidence. Future projections span the range of global surface air temperature best estimates for SSP1–2.6 and SSP5–8.5 scenarios described in Section 1.6. Range shown for 2100 is based on CMIP6 multi-model mean for 2081–2100 from Table 4.5; range for 2300 is based upon an emulator and taken from Table 4.9. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

**END CROSS-CHAPTER BOX 2.1, FIGURE 1 HERE**

**END CROSS-CHAPTER BOX 2.1 HERE**

### 2.2 Changes in Climate Drivers

This section assesses the magnitude and rates of changes in both natural and anthropogenically mediated climate drivers over a range of time scales. First, changes in insolation (orbital and solar; 2.2.1), and volcanic stratospheric aerosol (2.2.2) are assessed. Next, well-mixed greenhouse gases (GHGs) (CO₂, N₂O and CH₄) are covered in 2.2.3, with climate feedbacks and other processes involved in the carbon cycle assessed in Chapter 5. The section continues with the assessment of changes in halogenated GHGs (2.2.4), stratospheric water vapour, stratospheric and tropospheric ozone (2.2.5), and tropospheric aerosols (2.2.6). Short-lived Climate Forcers (SLCFs), their precursor emissions and key processes are assessed in more detail in Chapter 6. Section 2.2.7 assesses the effect of historical land cover change on climate, including biophysical and biogeochemical processes. Section 2.2.8 summarizes the changes in the Earth’s energy balance since 1750 using the comprehensive assessment of Effective Radiative Forcing (ERF) performed in Section 7.3. For some SLCFs with insufficient spatial or temporal observational coverage, ERFs are based on model estimates, but also reported here for completeness and context. Tabulated global mixing ratios of all well-mixed GHGs and ERFs from 1750–2019 are provided in Annex III.

#### 2.2.1 Solar and Orbital Forcing

AR5 assessed solar variability over multiple time scales, concluding that total solar irradiance (TSI) multi-millennial fluctuations over the past 9 kyr were <1 W m⁻², but with no assessment of confidence provided. For multi-decadal to centennial variability over the last millennium, AR5 emphasized reconstructions of TSI that show little change (<0.1%) since the Maunder Minimum (1645–1715) when solar activity was particularly low, again without providing a confidence level. AR5 further concluded that the best estimate of radiative forcing due to TSI changes for the period 1750–2011 was 0.05–0.10 W m⁻² (medium confidence),
and that TSI very likely changed by \(-0.04\) [-0.08–0.00 W m\(^{-2}\)] between 1986 and 2008. Potential solar influences on climate due to feedbacks arising from interactions with Galactic Cosmic Rays are assessed in Section 7.3.4.5.

Slow periodic changes in the Earth’s orbit around the Sun mainly cause variations in seasonal and latitudinal receipt of incoming solar radiation. Precise calculations of orbital variations are available for tens of millions of years (Berger and Loutre, 1991; Laskar et al., 2011). The range of insolation averaged over boreal summer at 65°N was about 83 W m\(^{-2}\) during the past million years, and 3.2 W m\(^{-2}\) during the past millennium, but there was no substantial effect upon global average radiative forcing (0.02 W m\(^{-2}\) during the past millennium).

A new reconstruction of solar irradiance extends back 9 kyr based upon updated cosmogenic isotope datasets and improved models for production and deposition of cosmogenic nuclides (Poluianov et al., 2016), and shows that solar activity during the second half of the 20th century was in the upper decile of the range. TSI features millennial-scale changes with typical magnitudes of 1.5 [1.4 to 2.1] W m\(^{-2}\) (Wu et al., 2018). Although stronger variations in the deeper past cannot be ruled out completely (Egorova et al., 2018; Reinhold et al., 2019), there is no indication of such changes having happened over the last 9 kyr.

Recent estimates of TSI and spectral solar irradiance (SSI) for the past millennium are based upon updated irradiance models (e.g., Egorova et al., 2018; Wu et al., 2018) and employ updated and revised direct sunspot observations over the last three centuries (Clette et al., 2014; Chatzistergos et al., 2017) as well as records of sunspot numbers reconstructed from cosmogenic isotope data prior to this (Usoskin et al., 2016). These reconstructed TSI timeseries (Figure 2.2a) feature little variation in TSI averaged over the past millennium. The TSI between the Maunder Minimum (1645–1715) and second half of the 20th century increased by 0.7–2.7 W m\(^{-2}\) (Jungclaus et al., 2017; Egorova et al., 2018; Lean, 2018; Wu et al., 2018; Yeo et al., 2020; Lockwood & Ball, 2020). This TSI increase implies a change in ERF of 0.09–0.35 W m\(^{-2}\) (Section 7.3.4.4).

Estimation of TSI changes since 1900 (Figure 2.2b) has further strengthened, and confirms a small (less than about 0.1 W m\(^{-2}\)) contribution to global climate forcing (Section 7.3.4.4). New reconstructions of TSI over the 20th century (Lean, 2018; Wu et al., 2018) support previous results that the TSI averaged over the solar cycle very likely increased during the first seven decades of the 20th century and decreased thereafter (Figure 2.2b). TSI did not change significantly between 1986 and 2019. Improved insights (Coddington et al., 2016; Yeo et al., 2015, 2017a; Krivova et al., 2006) show that variability in the 200–400 nm UV range was greater than previously assumed. Building on these results, the forcing proposed by Matthes et al. (2017) has a 16% stronger contribution to TSI variability in this wavelength range compared to the forcing used in the 5th Phase of the Coupled Model Intercomparison Project (CMIP5).

To conclude, solar activity since the late 19th century was relatively high but not exceptional in the context of the past 9 kyr (high confidence). The associated global-mean ERF is in the range of \(-0.06\) to 0.08 W m\(^{-2}\) (Section 7.3.4.4).

[START FIGURE 2.2 HERE]

Figure 2.2: Time series of solar and volcanic forcing for the past 2.5 kyr (panels a, c) and since 1850 (panels b, d). (a) Total solar irradiance (TSI) reconstruction (10-year running averages) recommended for CMIP6 / PMIP4 millennial experiments based on the radiocarbon dataset before 1850 (blue) scaled to the CMIP6 historical forcing after 1850 (purple). (b) TSI time series (6-month running averages) from CMIP6 historical forcing as inferred from sunspot numbers (blue), compared to CMIP5 forcing based on (red) and an update to CMIP6 by a TSI composite (orange). (c) Volcanic forcing represented as reconstructed Stratospheric aerosol optical depth (SAOD; as presented in Section 7.3.4.6) at 550 nm. Estimates covering 500 BCE to 1900 CE (green) and 1850–2015 (blue). (d) SAOD reconstruction from CMIP6 (v 4) (blue), compared to CMIP5 forcing (red). Note the change in y-axis range between panels c and d. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.2 HERE]
2.2.2 Volcanic Aerosol Forcing

AR5 concluded that, on inter-annual time scales, the radiative effects of volcanic aerosols are a dominant natural driver of climate variability, with the greatest effects occurring within the first 2–5 years following a strong eruption. Reconstructions of radiative forcing by volcanic aerosols used in the Paleoclimate Modelling Intercomparison Project Phase III (PMIP3) simulations and in AR5 featured short-lived perturbations of a range of magnitudes, with events of greater magnitude than $-1 \text{ W m}^{-2}$ (annual mean) occurring on average every 35–40 years, although no associated assessment of confidence was given. This section focuses on advances in reconstructions of stratospheric aerosol optical depth (SAOD), whereas Chapter 7 focuses on the effective radiative forcing of volcanic aerosols, and Chapter 5 assesses volcanic emissions of CO$_2$ and CH$_4$; tropospheric aerosols are discussed in Section 2.2.6. Cross-Chapter Box 4.1 undertakes an integrative assessment of volcanic effects including potential for 21st century effects.

Advances in analysis of sulphate records from the Greenland Ice Sheet (GrIS) and AIS have resulted in improved dating and completeness of SAOD reconstructions over the past 2.5 kyr (Sigl et al., 2015), a more uncertain extension back to 10 ka (Kobashi et al., 2017; Toohey and Sigl, 2017), and a better differentiation of sulphates that reach high latitudes via stratospheric (strong eruptions) versus tropospheric pathways (Burke et al., 2019; Gautier et al., 2019). The PMIP4 volcanic reconstruction extends the period analysed in AR5 by 1 kyr (Jungclaus et al., 2017; Figure 2.2c) and features multiple strong events that were previously misdated, underestimated or not detected, particularly before about 1500 CE. The period between successive large volcanic eruptions (Negative ERF greater than $-1 \text{ W m}^{-2}$), ranges from 3–130 years, with an average of 43 ± 7.5 years between such eruptions over the past 2.5 kyr (data from Toohey & Sigl, 2017). The most recent such eruption was that of Mt Pinatubo in 1991. Century-long periods that lack such large eruptions occurred once every 400 years on average. Systematic uncertainties related to the scaling of sulphate abundance in glacier ice to radiative forcing have been estimated to be about 60% (Hegerl et al., 2006). Uncertainty in the timing of eruptions in the proxy record is ± 2 years (95% confidence interval) back to 1.5 ka and ± 4 years before (Toohey and Sigl, 2017).

SAOD averaged over the period 950–1250 CE (0.012) was lower than for the period 1450–1850 CE (0.017) and similar to the period 1850–1900 (0.011). Uncertainties associated with these inter-period differences are not well quantified but have little effect because the uncertainties are mainly systematic throughout the record. Over the past 100 years, SAOD averaged 14% lower than the mean of the previous 24 centuries (back to 2.5 ka), and well within the range of centennial-scale variability (Toohey and Sigl, 2017).

Direct observations of volcanic gas-phase sulphur emissions (mostly SO$_2$), sulphate aerosols, and their radiative effects are available from a variety of sources (Kremser et al., 2016). New estimates of SO$_2$ emissions from explosive eruptions have been derived from satellite (beginning in 1979) and in situ measurements (Höppner et al., 2015; Carr et al., 2016; Neely and Schmidt, 2016; Brühl, 2018). Satellite observations of aerosol extinction after recent eruptions have uncertainties of about 15–25% (Vernier et al., 2011; Bourassa et al., 2012). Additional uncertainties occur when gaps in the satellite records are filled by complementary observations or using statistical methods (Thomason et al., 2018). Merged datasets (Thomason et al., 2018) and sparse ground-based measurements (Stothers, 1997) allow for volcanic forcing estimates back to 1850. In contrast to the CMIP5 historical volcanic forcing datasets (Ammann et al., 2003), updated time series (Luo et al., 2018; Figure 2.2d) feature a more comprehensive set of optical properties including latitude-, height- and wavelength-dependent aerosol extinction, single scattering albedo and asymmetry parameters. A series of small-to-moderate eruptions since 2000 resulted in perturbations in SAOD of 0.004–0.006 (Andersson et al., 2015; Schmidt et al., 2018).

To conclude, strong individual volcanic eruptions cause multi-annual variations in radiative forcing. However, the average magnitude and variability of SAOD and its associated volcanic aerosol forcing since 1900 are not unusual in the context of at least the past 2.5 kyr (medium confidence).
2.2.3 Well-mixed Greenhouse Gases (WMGHGs)

Well-mixed greenhouse gases generally have lifetimes of more than several years. AR5 assigned medium confidence to the values of atmospheric CO₂ concentrations (mixing ratios) during the warm geological periods of the early Eocene and Pliocene. It concluded with very high confidence that, by 2011, the mixing ratios of CO₂, CH₄, and N₂O in the atmosphere exceeded the range derived from ice cores for the previous 800 kyr, and that the observed rates of increase of the greenhouse gases were unprecedented on centennial timescales over at least the past 22 kyr. It reported that over 2005–2011 atmospheric burdens of CO₂, CH₄, and N₂O increased, with 2011 levels of 390.5 ppm, 1803.2 ppb and 324.2 ppb, respectively. Increases of CO₂ and N₂O over 2005–2011 were comparable to those over 1996–2005, while CH₄ resumed increasing in 2007, after remaining nearly constant over 1999–2006. A comprehensive process-based assessment of changes in CO₂, CH₄, and N₂O is undertaken in Chapter 5.

2.2.3.1 CO₂ during 450 Ma to 800 ka

Isotopes from continental and marine sediments using improved analytical techniques and sampling resolution have reinforced the understanding of long-term changes in atmospheric CO₂ during the past 450 Myr (Table 2.1, Figure 2.3). In particular, for the last 60 Myr, sampling resolution and accuracy of the boron isotope proxy in ocean sediments has improved (Anagnostou et al., 2016; 2020; Babila et al., 2018; Chalk et al., 2017; de la Vega et al., 2020; Dyez et al., 2018; Gutjahr et al., 2017; Harper et al., 2020; Henehan et al., 2019; 2020; Penman et al., 2014; Raitzsch et al., 2018; Sosdian et al., 2018), the understanding of the alkenone CO₂ proxy has increased (e.g., Badger et al., 2019; Rae et al., 2021; Stoll et al., 2019; Zhang et al., 2019; 2020) and new phytoplankton proxies have been developed and applied (e.g., Witkowski et al., 2018). Understanding of the boron isotope CO₂ proxy has improved since AR5 with studies showing very good agreement between boron-CO₂ estimates and co-existing ice core CO₂ (Chalk et al., 2017; Foster, 2008; Henehan et al., 2013; Hönisch & Hemming, 2005; Raitzsch et al., 2018; Figure 2.3c). Such independent validation has proven difficult to achieve with the other available CO₂ proxies (e.g., Badger et al., 2019; Da et al., 2019; Stoll et al., 2019; Zhang et al., 2019). Remaining uncertainties in these ocean sediment based proxies (Hollis et al., 2019) partly limit the applicability of the alkenone δ¹³C and boron δ¹⁰B proxies beyond the Cenozoic, although new records are emerging, e.g. Jurikova et al., (2020). CO₂ estimates from the terrestrial CO₂ proxies, such as stomatal density in fossil plants and δ¹³C of palaeosol carbonates, are available for much of the last 420 Myr. Given the low sampling density, relatively large CO₂ uncertainty, and high age uncertainty (relative to marine sediments) of the terrestrial proxies, preference here is given to the marine based proxies (and boron in particular) where possible.

Levels were close to 1750 values during at least one prolonged interval during the Carboniferous and Permian (350–252 Ma). During the Triassic (251.9–201.3 Ma), atmospheric CO₂ mixing ratios reached a maximum between 2000–5000 ppm (200–220 Ma). During the PETM (56 Ma) CO₂ rapidly rose from about 900 ppm to about 2000 ppm (Anagnostou et al., 2020; Gutjahr et al., 2017; Schubert & Järens, 2013; Table 2.1) in 3–20 kyr (Gutjahr et al., 2017; Turner, 2018; Zeebe et al., 2016). Estimated multi-millennial rates of CO₂ accumulation during this event range from 0.3–1.5 PgC yr⁻¹ (Gingerich, 2019); were at least 4-5 times lower than current centennial rates (Section 5.3.1.1). Based on boron and carbon isotope data, supported by other proxies (Hollis et al., 2019), atmospheric CO₂ during the EECO (50 Ma) was between 1150 and 2500 ppm (medium confidence), and then gradually declined over the last 50 Myr at a long-term rate of about 16 ppm Myr⁻¹ (Figure 2.3). The last time the CO₂ mixing ratio was as high as 1000 ppm (the level reached by some high emission scenarios by 2100; Annex III) was prior to the Eocene-Oligocene transition (33.5 Ma; Figure 2.3) that was associated with the first major advance of the AIS (Anagnostou et al., 2016; Hollis et al., 2019; Pagani et al., 2011; Pearson et al., 2009; Witkowski et al., 2018). The compilation of Foster et al., (2017) constrained CO₂ concentration to between 290 and 450 ppm during the MPWP, based primarily on the boron-isotope data reported by Martínez-Boti et al. (2015b), consistent with the AR5 range of 300–450 ppm. A more recent high-resolution boron isotope-based study revealed that CO₂ cycled during the MPWP from about 330 to about 390 ppm on orbital timescales, with a mean of about 370 ppm (de la Vega et al., 2020). Although data from other proxy types (e.g., stomatal density or δ¹³C of alkenones) have too low resolution to resolve the orbital variability of CO₂ during this interval (e.g. Kürschner et al., 1996; Stoll et al.,...
2019) there is general agreement among the different proxy types with the boron-derived mean (e.g. Stoll et al., 2019). High-resolution sampling (about 1 sample per 3 kyr) with the boron-isotope proxy indicates mean CO₂ mixing ratios for the Marine Isotope Stage KM5c interglacial were 360–420 ppm (medium confidence) (de la Vega et al., 2020).

Following the MPWP, the atmospheric CO₂ mixing ratio generally decreased at a rate of about 30 ppm Myr⁻¹. It is very likely that CO₂ levels as high as the present were not experienced in the last 2 Myr (Bartoli et al., 2011; Chalk et al., 2017; Da et al., 2019; Dyez et al., 2018; Hönsch et al., 2009; Martínez-Botí et al., 2015a; Stoll et al., 2019). Related to the shift of glacial-interglacial cycle frequency from 40 to 100 kyr at 0.8–1.2 Ma, there was a decrease of glacial-period CO₂ (Chalk et al., 2017; Dyez et al., 2018). These boron isotope-based CO₂ results agree with available records based on ancient ice exposed near the surface of the AIS (Yan et al., 2019), however, direct comparison is limited due to a lack of ancient ice cores with sufficiently continuous stratigraphy (Higgins et al., 2015; Brook and Buizert, 2018).

To conclude, there is high confidence that average EECO and MPWP (KM5c) CO₂ concentration was higher than those preceding industrialisation at 1150-2500 ppm and 360-420 ppm, respectively. Although there is some uncertainty due to the non-continuous nature of marine sediment records, the last time atmospheric CO₂ mixing ratio was as high as present was very likely more than 2 million years ago.

[START TABLE 2.1 HERE]

<table>
<thead>
<tr>
<th>Reference period</th>
<th>CO₂ concentration (ppm) and dataset details</th>
<th>Rate of change (ppm per century)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Modern (1995-2014)</td>
<td>359.6 to 360.4 → 396.7 to 397.5 (AR6)</td>
<td>192.3 to 198.3* (AR6)</td>
</tr>
<tr>
<td>Last 100 years (1919-2019)</td>
<td>302.8 to 306.0 → 409.5 to 410.3 (AR6)</td>
<td>103.9 to 107.1 (AR6)</td>
</tr>
<tr>
<td>Approximate pre-industrial baseline (1850-1900) (See CCB1.2)</td>
<td>283.4 to 287.6 → 294.8 to 298.0 (AR6); 284.3 → 295.7† (CMIP6)</td>
<td>16.5 to 27.1† (AR6) 22.8* (CMIP6)</td>
</tr>
<tr>
<td>Last millennium (1000-1750)</td>
<td>278.0 to 285.0 (AR6; average of WAIS Divide, Law Dome and EDML core data)</td>
<td>-6.9 ~ 4.7† (Law Dome); -1.9 ~ 3.2† (EDML); -5.2 ~ 4.2† (WAIS Divide)</td>
</tr>
<tr>
<td>MH</td>
<td>260.1 to 268.1 (Dome C; CMIP6)</td>
<td></td>
</tr>
<tr>
<td>LDT</td>
<td>193.2† → 271.2‡ (AR6); 195.2† → 265.3‡ (Dome C); 191.2† → 277.0‡ (WAIS Divide)</td>
<td>9.6† (WAIS Divide); 7.1† (Dome C)</td>
</tr>
</tbody>
</table>
Table 2.1

<table>
<thead>
<tr>
<th>Event</th>
<th>Time Period (Kyr)</th>
<th>Source (AR6)</th>
</tr>
</thead>
<tbody>
<tr>
<td>LGM</td>
<td>188.4 to 194.2</td>
<td>(AR6); 190.5 to 200.1 (WAIS Divide); 186.8 to 202.0 (Byrd); 184.9 to 193.1 (Dome C); 180.5 to 192.7 (Siple Dome); 190° (PMIP6); 174.2 to 205.8 (δ13B proxy)</td>
</tr>
<tr>
<td>LIG</td>
<td>265.9 to 281.5</td>
<td>(AR6); 259.4 to 283.8 (Vostok); 266.2 to 285.4 (Dome C); 275° (PMIP4) 282.2 to 305.8 (δ13B proxy)</td>
</tr>
<tr>
<td>MPWP (KM5c)</td>
<td>360 to 420</td>
<td>(AR6)</td>
</tr>
<tr>
<td>EECO</td>
<td>1150 to 2500</td>
<td>(AR6)</td>
</tr>
<tr>
<td>PETM</td>
<td>800 to 1000</td>
<td>→ 1400 to 3150 (AR6)</td>
</tr>
</tbody>
</table>

*centennial rate of change estimated by extrapolation of data from a shorter time period. The values (x to y) represent very likely ranges (90% CIs).

Data uncertainty is not estimated.

[END TABLE 2.1 HERE]

Figure 2.3: The evolution of atmospheric CO₂ through the last 450 million years. The periods covered are 0–450 Ma (a), 0–58 Ma (b), and 0–3500 ka (c), reconstructed from continental rock, marine sediment and ice core records. Note different timescales and axes ranges in panels (a), (b) and (c). Dark and light green bands in (a) are uncertainty envelopes at 68% and 95% uncertainty, respectively. 100 ppm in each panel is shown by the marker in the lower right hand corner to aid comparison between panels. In panel (b) and (c) the major paleoclimate reference periods (CCB2.1) have been labelled, and in addition: MPT (Mid Pleistocene Transition), MCO (Miocene Climatic Optimum). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.3 HERE]

2.2.3.2 Glacial-interglacial WMGHG Fluctuations from 800 ka

Since AR5, the number of ice cores for the last 800 kyr has increased and their temporal resolution has improved (Figure 2.4), especially for the last 60 kyr and when combined with analyses of firn air, leading to improved quantification of greenhouse gas concentrations prior to the mid-20th century.

2.2.3.2.1 CO₂

Records of CO₂ from the AIS formed during the last glacial period and the LDT show century-scale fluctuations of up to 9.6 ppm (Ahn et al., 2012; Ahn & Brook, 2014; Bauska et al., 2015; Marcott et al., 2014; Rubino et al., 2019). Although these rates are an order of magnitude lower than those directly observed over 1919–2019 CE (Section 2.2.3.3.1), they provide information on non-linear responses of climate-biogeochemical feedbacks (Section 5.1.2). Multiple records for 0–1850 CE show CO₂ mixing ratios of 274–285 ppm. Offsets among ice core records are about 1%, but the long-term trends agree well and show coherent multi-centennial variations of about 10 ppm (Ahn et al., 2012; Bauska et al., 2015; Rubino et al., 2019). Multiple records show CO₂ concentrations of 278.3 ± 2.9 ppm in 1750 and 285.5 ± 2.1 ppm in 1850 (Ahn et al., 2012, 2019; Bauska et al., 2015; MacFarling Meure et al., 2006; Siegenthaler et al., 2005). CO₂ concentration increased by 5.0 ± 0.8 ppm during 970–1130 CE, followed by a decrease of 4.6 ± 1.7 ppm during 1580–1700 CE. The greatest rate of change over the CE prior to 1750 is observed at about 1600 CE, and ranges from –6.9 to 4.7 ppm per century in multiple high-resolution ice core records (Ahn et al., 2012; Bauska et al., 2015; MacFarling Meure et al., 2006; Rubino et al., 2019; Siegenthaler et al., 2005). Although ice core records present low-pass filtered time series due to gas diffusion and gradual bubble close-off in the snow layer over the ice sheet (Fourteau et al., 2020), the rate of increase since 1850 CE (about 125 ppm
increase over about 170 years) is far greater than implied for any 170-year period by ice core records that cover the last 800 ka (very high confidence).

2.2.3.2 CH₄
CH₄ concentrations over the past 110 kyr are higher in the Northern Hemisphere (NH) than in the Southern Hemisphere (SH), but closely correlated on centennial and millennial timescales (WAIS Divide Project Members, 2015). On glacial to interglacial cycles, approximately 450 ppb oscillations in CH₄ concentrations have occurred (Lougue et al., 2008). On millennial timescales, most rapid climate changes observed in Greenland and other regions are coincident with rapid CH₄ changes (Rhodes et al., 2015, 2017; WAIS Divide Project Members, 2015). The variability of CH₄ on centennial timescales during the early Holocene does not significantly differ from that of the late Holocene prior to about 1850 (Rhodes et al., 2013; Yang et al., 2017). The LGM concentration was 390.5 ± 6.0 ppb (Kageyama et al., 2017). The global mean concentrations during 0–1850 CE varied between 625 and 807 ppb. High-resolution ice core records from Antarctica and Greenland exhibit the same trends with an inter-polar difference of 36–47 ppb (Mitchell et al., 2013b; Sapart et al., 2012). There is a long-term positive trend of about 0.5 ppb per decade during the Common Era (CE) until 1750 CE. The most rapid CH₄ changes prior to industrialisation were as large as 30–50 ppb on multi-decadal timescales. Global mean CH₄ concentrations estimated from Antarctic and Greenland ice cores are 729.2 ± 9.4 ppb in 1750 and 807.6 ± 13.8 ppb in 1850 (Mitchell et al., 2013b).

2.2.3.3 N₂O
New records show that N₂O concentration changes are associated with glacial-interglacial transitions (Schilt et al., 2014). The most rapid change during the last glacial termination is a 30 ppb increase in a 200-year period, which is an order of magnitude smaller than the modern rate (Section 2.2.3.3). During the LGM, N₂O was 208.5 ± 7.7 ppb (Kageyama et al., 2017). Over the Holocene the lowest value was 257 ± 6.6 ppb during 6–8 ka, but millennial variation is not clearly detectable due to analytical uncertainty and insufficient ice core quality (Flückiger et al., 2002; Schilt et al., 2010). Recently acquired high-resolution records from Greenland and Antarctica for the last 2 kyr consistently show multi-centennial variations of about 5–10 ppb (Figure 2.4), although the magnitudes vary over time (Ryu et al., 2020). Three high temporal resolution records exhibit a short-term minimum at about 600 CE of 261 ± 4 ppb (MacFarling Meure et al., 2006; Ryu et al., 2020). It is very likely that industrial N₂O increase started before 1900 CE (Machida et al., 1995; Sowers, 2001; MacFarling Meure et al., 2006; Ryu et al., 2020). Multiple ice cores show N₂O concentrations of 270.1 ± 6.0 ppb in 1750 and 272.1 ± 5.7 ppb in 1850 (Flückiger et al., 1999; Machida et al., 1995; Rubino et al., 2019; Ryu et al., 2020; Sowers, 2001).

[START FIGURE 2.4 HERE]

Figure 2.4: Atmospheric WMGHG concentrations from ice cores. (a) Records during the last 800 kyr with the LGM to Holocene transition as inset. (b) Multiple high-resolution records over the CE. The horizontal black bars in the panel a inset indicate Last Glacial Maximum (LGM) and Last Deglacial Termination (LDT) respectively. The red and blue lines in (b) are 100-year running averages for CO₂ and N₂O concentrations, respectively. The numbers with vertical arrows in (b) are instrumentally measured concentrations in 2019. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.4 HERE]

2.2.3.3 Modern measurements of WMGHGs

In this section and for calculation of ERF, surface global averages are determined from measurements representative of the well-mixed lower troposphere. Global averages that include sites subject to significant anthropogenic activities or influenced by strong regional biospheric emissions are typically larger than those
from remote sites, and require weighting accordingly (Table 2.2). This section focusses on global mean mixing ratios estimated from networks with global spatial coverage, and updated from the CMIP6 historical dataset (Meinshausen et al., 2017) for periods prior to the existence of global networks.

### [START TABLE 2.2 HERE]

#### Table 2.2:

Atmospheric global annual mean mixing ratios (dry-air mole fraction) observed in 2011 and 2019, and relative changes since 2011, for selected well-mixed, radiatively important gases (ERF > 0.001 W m⁻²), estimated from various measurement networks or compilations. Units are parts-per-million (ppm) for CO₂, parts-per-billion (ppb) for CH₄ and N₂O, parts-per-trillion (ppt) for all other gases. Time series since 1750, data for additional gases, references, and network information can be found in Annex III and the corresponding electronic supplement. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

<table>
<thead>
<tr>
<th>Species</th>
<th>Lifetime, AR6, ERF</th>
<th>2011</th>
<th>2019</th>
<th>Chg.</th>
<th>Network</th>
</tr>
</thead>
<tbody>
<tr>
<td>CO₂</td>
<td>409.9 (0.4)</td>
<td>390.5</td>
<td>409.9 (0.17)</td>
<td>5.0%</td>
<td>SIO</td>
</tr>
<tr>
<td></td>
<td>2.156</td>
<td>390.2</td>
<td>409.6 (0.31)</td>
<td>5.0%</td>
<td>CSIRO</td>
</tr>
<tr>
<td></td>
<td></td>
<td>390.9</td>
<td>410.5 (0.30)</td>
<td>5.0%</td>
<td>WMO</td>
</tr>
<tr>
<td></td>
<td></td>
<td>390.9</td>
<td></td>
<td></td>
<td>CMIP6</td>
</tr>
<tr>
<td>CH₄</td>
<td>1803.1 (1.0)</td>
<td>1803.1</td>
<td>1803.1 (1.0)</td>
<td>3.5%</td>
<td>NOAA*</td>
</tr>
<tr>
<td></td>
<td>1803.6 (2.0)</td>
<td>1803.6</td>
<td>1803.6 (2.0)</td>
<td>3.5%</td>
<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td>0.544</td>
<td>1791.8</td>
<td>1860.8 (3.5)</td>
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<td>UCI</td>
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<tr>
<td></td>
<td></td>
<td>1802.3</td>
<td>1862.5 (2.3)</td>
<td>3.3%</td>
<td>CSIRO</td>
</tr>
<tr>
<td></td>
<td></td>
<td>1813</td>
<td>1877 (3)</td>
<td>3.5%</td>
<td>WMO</td>
</tr>
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<td></td>
<td></td>
<td>1813.1</td>
<td></td>
<td></td>
<td>CMIP6</td>
</tr>
<tr>
<td>N₂O</td>
<td>324.2</td>
<td>324.2</td>
<td>331.9 (0.2)</td>
<td>2.4%</td>
<td>NOAA*</td>
</tr>
<tr>
<td></td>
<td>324.3</td>
<td>324.3</td>
<td>332.0 (0.2)</td>
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<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td>0.180</td>
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<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>503.1 (3.2)</td>
<td>502.6</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
</tr>
<tr>
<td></td>
<td></td>
<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
</tr>
<tr>
<td></td>
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<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
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</tr>
<tr>
<td>CFC-12</td>
<td>502.5</td>
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<td>504.6 (0.2)</td>
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<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td>502.5</td>
<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
</tr>
<tr>
<td></td>
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<td>502.5</td>
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</tr>
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<td>2.4%</td>
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<td>504.6 (0.2)</td>
<td>2.4%</td>
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<td>504.6 (0.2)</td>
<td>2.4%</td>
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<td>504.6 (0.2)</td>
<td>2.4%</td>
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<td>2.4%</td>
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</tr>
<tr>
<td></td>
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<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
</tr>
<tr>
<td></td>
<td></td>
<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
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<td>502.5</td>
<td>504.6 (0.2)</td>
<td>2.4%</td>
<td>AGAGE*</td>
</tr>
</tbody>
</table>

### [END TABLE 2.2 HERE]
2.2.3.3.1 Carbon Dioxide (CO$_2$)

There has been a positive trend in globally averaged surface CO$_2$ mixing ratios since 1958 (Figure 2.5a), that reflects the imbalance of sources and sinks (Section 5.2). The growth rate has increased overall since the 1960s (Figure 2.5a inset), while annual growth rates have varied substantially, e.g. reaching a peak during the strong El Niño events of 1997–1998 and 2015–2016 (Bastos et al., 2013; Betts et al., 2016). The average annual CO$_2$ increase from 2000 through 2011 was 2.0 ppm yr$^{-1}$ (standard deviation 0.3 ppm yr$^{-1}$), similar to what was reported in AR5. From 2011 through 2019 it was 2.4 ppm yr$^{-1}$ (standard deviation 0.5 ppm yr$^{-1}$), which is higher than that of any comparable time period since global measurements began. Global networks consistently show that the globally averaged annual mean CO$_2$ has increased by 5.0% since 2011, reaching 409.9 ± 0.4 ppm in 2019 (NOAA measurements). Further assessment of changing seasonality is undertaken in Section 2.3.4.1.

[START FIGURE 2.5 HERE]

Figure 2.5: Globally averaged dry-air mole fractions of well-mixed greenhouse gases. (a) CO$_2$ from SIO, CSIRO, and NOAA/GML. (b) CH$_4$ from NOAA, AGAGE, CSIRO, and UCI; and (c) N$_2$O from NOAA, AGAGE, and CSIRO (see Table 2.2). Growth rates, calculated as the time derivative of the global means after removing seasonal cycle are shown as inset figures. Note that the CO$_2$ series is 1958–2019 whereas CH$_4$, and N$_2$O are 1979–2019. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.5 HERE]

2.2.3.3.2 Methane (CH$_4$)

The globally averaged surface mixing ratio of CH$_4$ in 2019 was 1866.3 ± 3.3 ppb, which is 3.5% higher than 2011, while observed increases from various networks range from 3.3–3.9% (Table 2.2; Figure 2.5b). There are marked growth rate changes over the period of direct observations, with a decreasing rate from the late-1970s through the late-1990s, very little change in concentrations from 1999–2006, and resumed increases since 2006. Atmospheric CH$_4$ fluctuations result from complex variations of sources and sinks. A detailed discussion of recent methane trends and our understanding of their causes is presented in Cross-Chapter Box 2.5.

2.2.3.3.3 Nitrous Oxide (N$_2$O)

AR5 reported 324.2 ± 0.1 ppb for global surface annual mean N$_2$O in 2011; since then, it has increased by 2.4% to 332.1 ± 0.4 ppb in 2019. Independent measurement networks agree well for both the global mean mixing ratio and relative change since 2011 (Table 2.2). Over 1995–2011, N$_2$O increased at an average rate of 0.79 ± 0.05 ppb yr$^{-1}$. The growth rate has been higher in recent years, amounting to 0.96 ± 0.05 ppb yr$^{-1}$ from 2012 to 2019 (Figure 2.5c) (Section 5.2.3.5).
2.2.4 Halogenated Greenhouse Gases (CFCs, HCFCs, HFCs, PFCs, SF6 and others)

This category includes ozone depleting substances (ODS), their replacements, and gases used industrially or produced as by-products. Some have natural sources (Section 6.2.2.4). AR5 reported that atmospheric abundances of chlorofluorocarbons (CFCs) were decreasing in response to controls on production and consumption mandated by the Montreal Protocol on Substances that Deplete the Ozone Layer and its amendments. In contrast, abundances of both hydrochlorofluorocarbons (HCFCs, replacements for CFCs) and hydrofluorocarbons (HFCs, replacements for HCFCs) were increasing. Atmospheric abundances of perfluorocarbons (PFCs), SF6, and NF3 were also increasing.

Further details on ODS and other minor greenhouse gases can be found in the Scientific Assessment of Ozone Depletion: 2018 (Engel & Rigby, 2018; Montzka & Velders, 2018). Updated mixing ratios of the most radiatively important gases (ERF > 0.001 W m-2) are reported in Table 2.2, and additional gases (ERF <0.001 W m-2) are shown in Annex III.

2.2.4.1 CFCs

Atmospheric abundances of most CFCs have continued to decline since 2011 (AR5). The globally averaged abundance of CFC-12 decreased by 25 ppt (4.8%) from 2011 to 2019, while CFC-11 decreased by about 11 ppt (4.7%) over the same period (Table 2.2, Figure 2.6). Atmospheric abundances of some minor CFCs (CFC-13, CFC-115, CFC-113a) have increased since 2011 (Annex III), possibly related to use of HFCs (Laube et al., 2014). Overall, as of 2019 the ERF from CFCs has declined by 9 ± 0.5% from its maximum in 2000, and 4.7 ± 0.6% since 2011 (Table 7.5).

While global reporting indicated that CFC-11 production had essentially ceased by 2010, and the atmospheric abundance of CFC-11 is still decreasing, emissions inferred from atmospheric observations began increasing in 2013–2014 and remained elevated for 5–6 years, suggesting renewed and unreported production (Montzka et al., 2018; 2021; Park et al., 2021; Rigby et al., 2019). The global lifetimes of several ozone-depleting substances have been updated (SPARC, 2013), in particular for CFC-11 from 45 to 52 years.

[START FIGURE 2.6 HERE]

Figure 2.6: Global mean atmospheric mixing ratios of select ozone-depleting substances and other greenhouse gases. Data shown are based on the CMIP6 historical dataset and data from NOAA and AGAGE global networks. PFCs include CF3, C2F6, and C3F8, and c-C3F8; Halons include halon-1211, halon-1301, and halon-2402; other HFCs include HFC-23, HFC-32, HFC-125, HFC-143a, HFC-152a, HFC-227ea, HFC-245fa, and HFC-365mfc, and HFC-43-10mee. Note that the y-axis range is different for a, b and c and a 25 ppt yardstick is given next to each panel to aid interpretation. Further data are in Annex III, and details on data sources and processing are available in the chapter data table (Table 2SM.1).

[END FIGURE 2.6 HERE]

2.2.4.2 HCFCs

The atmospheric abundances of the major HCFCs (HCFC-22, HCFC-141b, HCFC-142b), primarily used in refrigeration and foam blowing, are increasing, but rates of increase have slowed in recent years (Figure 2.6). Global mean mixing ratios (Table 2.2) showed good concordance at the time of AR5 for the period 2005–2011. For the period 2011–2019, the UCI network detected larger increases in HCFC-22, HCFC-141b, and HCFC-142b compared to the NOAA and AGAGE networks. Reasons for the discrepancy are presently unverified, but could be related to differences in sampling locations in the networks (Simpson et al., 2012). Emissions of HCFC-22, derived from atmospheric data, have remained relatively stable since 2012, while those of HCFC-141b and HCFC-142b have declined (Engel et al., 2018). Minor HCFCs, HCFC-133a and
HCFC-31, have been detected in the atmosphere (currently less than 1 ppt) and may be unintentional by-products of HFC production (Engel et al., 2018).

2.2.4.3 HFCs, PFCs, SF6 and other radiatively important halogenated gases

Hydrofluorocarbons (HFCs) are replacements for CFCs and HCFCs. The atmospheric abundances of many HFCs increased between 2011 and 2019. HFC-134a (mobile air conditioning, foam blowing, and domestic refrigerators) increased by 71% from 63 ppt in 2011 to 107.6 ppt in 2019 (Table 2.2). The UCI network detected a slightly smaller relative increase (53%). HFC-23, which is emitted as a by-product of HCFC-22 production, increased by 8.4 ppt (35%) over 2011–2019, HFC-32 used as a substitute for HCFC-22, increased at least by 300%, and HFC-143a and HFC-125 showed increases of 100% and 187%, respectively. While the ERF of HFC-245fa is currently < 0.001 W m$^{-2}$, its atmospheric abundance doubled since 2011 to 3.1 ppt in 2019 (Annex III). In contrast, HFC-152a is showing signs of stable (steady-state) abundance.

Other radiatively important gases with predominantly anthropogenic sources also continue to increase in abundance. SF$_6$, used in electrical distribution systems, magnesium production, and semiconductor manufacturing, increased from 7.3 ppt in 2011 to 10.0 ppt in 2019 (+36%). Alternatives to SF$_6$, or SF$_6$-free equipment for electrical systems have become available in recent years, but SF$_6$ is still widely in use in electrical switch gear (Simmonds et al., 2020). The global lifetime of SF$_6$ has been revised from 5200 years to about 1000 years (Kovács et al., 2017; Ray et al., 2017) with implications for climate emission metrics (Section 7.6.2). NF$_3$, which is used in the semiconductor industry, increased 147% over the same period to 2.05 ppt in 2019. Its contribution to ERF remains small, however, at 0.0004 W m$^{-2}$. The atmospheric abundance of SO$_2$F$_2$, which is used as a fumigant in place of ozone-depleting methyl bromide, reached 2.5 ppt in 2019, a 46% increase from 2011. Its ERF also remains small at 0.0005 W m$^{-2}$.

The global abundance of CCl$_4$ continues to decline, down about 9.6% since 2011. Following a revision of the global lifetime from 26 to 32 years, and discovery of previously unknown sources (e.g. biproduct industrial emissions), knowledge of the CCl$_4$ budget has improved. There is now better agreement between top-down emissions estimates (based on atmospheric measurements) and industry-based estimates (Engel et al., 2018). Halon-1211, mainly used for fire suppression, is also declining, and its ERF dropped below 0.001 W m$^{-2}$ in 2019. While CH$_3$Cl has a short atmospheric lifetime (6 months), and is not well-mixed, its abundance is increasing and its ERF is approaching 0.001 W m$^{-2}$.

Perfluorocarbons CF$_4$ and C$_2$F$_6$, which have exceedingly long global lifetimes, showed modest increases from 2011 to 2019. CF$_4$, which has both natural and anthropogenic sources, increased 8.2% to 85.5 ppt, and C$_2$F$_6$ increased 16.3% to 4.85 ppt. C$_3$F$_8$, which is used in the electronics industry and may also be generated during the production of polytetrafluoroethylene (PTFE, “Teflon”) and other fluoropolymers (Mühlle et al., 2019) has increased 34% since 2011 to 1.75 ppt, although its ERF remains below 0.001 W m$^{-2}$. Other PFCs, present at mixing ratios < 1 ppt, have also been quantified (Droste et al., 2020; Annex III).

2.2.4.4 Summary of changes in halogenated gases

In summary, by 2019 the ERF of halogenated GHGs has increased 3.5% since 2011, reflecting predominantly a decrease in the atmospheric mixing ratios of CFCs and an increase in their replacements. However, average annual ERF growth rates associated with halogenated gases since 2011 are a factor of seven lower than in the 1970s and 1980s. Direct radiative forcings from CFCs, HCFCs, HFCs, and other halogenated greenhouse gases were 0.28, 0.06, 0.04, and 0.03 W m$^{-2}$ respectively, totalling 0.41 ± 0.07 W m$^{-2}$ in 2019 (see Table 7.5).
2.2.5 Other short-lived gases

2.2.5.1 Stratospheric water vapour

AR5 assessed low confidence in stratospheric water vapour (SWV) trends based on substantial seasonal and interannual variability in satellite data from 1992 to 2011. The 1980–2010 record of balloon-borne frost point hygrometer measurements over Boulder, Colorado (40°N), showed an average net increase of 1.0 ± 0.2 ppm (27 ± 6%) in the 16–26 km layer.

Since AR5, bias-adjusted spatially comprehensive SWV measurements by different satellite sensors were merged to form continuous records (Hegglin et al., 2014; Froidevaux et al., 2015; Davis et al., 2016). These indicate no net global increase of SWV in the lower stratosphere since the late 1980s. Hegglin et al. (2014) reported a latitudinal dependence of SWV trends and suggested that the upward trend over Boulder should not be considered representative of the global stratosphere, while Lossow et al. (2018) showed insignificant differences between SWV trends at Boulder and those for the 35–45°N zonal mean from 1980 to 2010 using model simulations and satellite observations.

Recent studies of dynamical influences on SWV (Eguchi et al., 2015; Evan et al., 2015; Tao et al., 2015; Konopka et al., 2016; Diallo et al., 2018; Garfinkel et al., 2018) have demonstrated that the quasi-biennial oscillation (QBO), El Niño-Southern Oscillation (ENSO), Sudden Stratospheric Warming (SSW) events and possibly also Pacific Decadal Variability (PDV) (Wang W. et al., 2016) can significantly influence SWV abundance and the tropical cold point tropopause temperatures that largely control water vapour entering the stratosphere. It has also been shown that the convective lofting of ice can moisten the lower stratosphere over large regions (Dessler et al., 2016; Anderson et al., 2017; Avery et al., 2017). Near-global observations of SWV have revealed unusually strong and abrupt interannual changes, especially in the tropical lower stratosphere. Between December 2015 and November 2016, the tropical mean SWV anomaly at 82 hPa dropped from 0.9 ± 0.1 ppm to −1.0 ± 0.1 ppm, accompanied by highly anomalous QBO-related dynamics in the tropical stratosphere (Newman P. et al., 2016; Tweedy et al., 2017) and the transition of ENSO from strong El Niño to La Niña conditions (Davis et al., 2017). The tropical mean SWV anomaly then rose sharply to 0.7 ± 0.1 ppm in June 2017 as warm westerlies returned to the tropical lower stratosphere and ENSO neutral conditions prevailed (Davis et al., 2017).

In summary, in situ measurements at a single mid-latitude location indicate about a 25% net increase in stratospheric water vapour since 1980, while merged satellite data records since the late 1980s suggest little net change. Recent studies of dynamical influences on SWV have highlighted their substantial roles in driving large interannual variability that complicates trend detection. There thus continues to be low confidence in trends of SWV over the instrumental period. Disregarding dynamic influences on SWV, an ERF of 0.05 ± 0.05 W m$^{-2}$ is estimated for SWV produced by CH$_{4}$ oxidation (Section 7.3.2.6), unchanged from AR5.

2.2.5.2 Stratospheric ozone

AR5 assessed that it was certain that global stratospheric ozone from the mid-1990s to 2011 was nearly constant and about 3.5% lower than in the reference period 1964–1980. Most of the declines occurred prior to the mid-1990s.

Global annual mean total ozone (Figure 2.7) significantly declined by about 3.5% during the 1980s and the early 1990s and by 2.5% over 60°S–60°N (near-global). Then, during 2000–2017, both global and near-global concentrations increased slightly, but not significantly, all in line with production and consumption limits of ODS regulated under the Montreal Protocol and its amendments. Near-global 2014–2017 mean total ozone is about 2.2% below the pre-ozone depletion 1964–1980 average (Braesicke et al., 2018). At southern and northern mid-latitudes, declines are 5.5% and 3.0% compared to the 1964–1980 average respectively. Total ozone remained practically unchanged in the tropics (Braesicke et al., 2018). Emission of ODS started before 1980 and some estimates suggest that as much as 40% of the long-term ozone loss...
occurred between 1960 and 1980 (Shepherd et al., 2014), lowering the 1964–1980 baseline values by about 1% (outside the polar regions), a value close to observational uncertainties. The world’s longest record of total ozone measurements from Arosa, Switzerland, initiated in 1926, does not show any substantial long-term changes before about 1980 (Staehelin et al., 2018).

[START FIGURE 2.7 HERE]

Figure 2.7:  Time series of annual mean total column ozone from 1964-2019. Values are in Dobson Units (DU), a good proxy for vertically integrated stratospheric ozone. Time series are shown for (a) near-global domain, (b-d) three zonal bands and (e) polar (60°-90°) total ozone in March (NH) and October (SH); the months when polar ozone losses usually are largest. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.7 HERE]

ERF depends strongly on the altitude of ozone changes. Two stratospheric regions are mainly responsible for long-term changes outside the polar regions. In the upper stratosphere (35–45 km), there was a strong decline (about 10%) from the start of observations in 1979 up to the mid-1990s and a subsequent increase by about 4% to present (SPARC/IO3C/GAW, 2019). In the lower stratosphere (20–25 km), there also was a statistically significant decline (7–8%) up to the mid-1990s, followed by stabilization or a small further decline (Ball et al., 2018; 2019), although the natural variability is too strong to make a conclusive statement (Chipperfield et al., 2018).

The strongest ozone loss in the stratosphere continues to occur in austral spring over Antarctica (ozone hole) with emergent signs of recovery after 2000 (Langematz et al., 2018). Interannual variability in polar stratospheric ozone is driven by large scale winds and temperatures, and, to a lesser extent, by the stratospheric aerosol loading and the solar cycle. This variability is particularly large in the Arctic, where the largest depletion events, comparable to a typical event in the Antarctic, occurred in 2011 (Manney et al., 2011; Langematz et al., 2018) and again in 2020 (Grooß and Müller, 2020; Manney et al., 2020). Further details on trends and ERF can be found in sections 6.3.2 and 7.3.2.5.

In summary, compared to the 1964–1980 average, stratospheric ozone columns outside polar regions (60°S–60°N) declined by about 2.5% over 1980–1995, and stabilized after 2000, with 2.2% lower values in 2014–2017. Large ozone depletions continue to appear in spring in the Antarctic and, in particularly cold years, also in the Arctic. Model based estimates disagree on the sign of the ERF due to stratospheric ozone changes, but agree that it is much smaller in magnitude than that due to tropospheric ozone changes (Section 7.3.2.5).

2.2.5.3  Tropospheric ozone

AR5 assessed medium confidence in large-scale increases of tropospheric ozone at rural surface sites across the NH (1970–2010), and in a doubling of European surface ozone during the 20th century, with the increases of surface ozone in the SH being of low confidence. Surface ozone likely increased in East Asia, but levelled off or decreased in the eastern USA and Western Europe. Free tropospheric trends (1971–2010) from ozonesondes and aircraft showed positive trends in most, but not all, assessed regions, and for most seasons and altitudes. This section focuses on large scale ozone changes; chemical and physical processes and regional changes in tropospheric ozone are assessed in Section 6.3.2.1 and Section 7.3.2.5 assesses radiative forcing.

Prior to 1850 ozone observations do not exist, but a recent analysis using clumped-isotope composition of molecular oxygen (18O/16O in O2) trapped in polar firm and ice, combined with atmospheric chemistry model simulations, constrains the global tropospheric ozone increase to less than 40% between 1850 and 2005, with most of this increase occurring between 1950 and 1980 (Yeung et al., 2019). Recently, the Tropospheric Ozone Assessment Report identified and evaluated 60 records of surface ozone observations collected at
rural locations worldwide between 1896 and 1975, which were based on a range of measurement techniques with potentially large uncertainties (Tarasick et al., 2019). They found that from the mid-20th century (1930s to the early 1970s) to 1990–2014, rural surface ozone increased by 30–70% across the northern extra-tropics. This is smaller than the 100% 20th century increase reported in AR5, which relied on far fewer measurement sites, all in Europe. In the northern tropics limited low-elevation historical data (1954–1975) provide no clear indication of surface ozone increases (Tarasick et al., 2019). However, similar to the northern mid-latitude increases, lower-free tropospheric ozone at Mauna Loa, Hawaii increased by approximately 50% from the late 1950s to present (Cooper et al., 2020). Historical observations are too limited to draw conclusions on surface ozone trends in the SH tropics and mid-latitudes since the mid-20th century, with tropospheric ozone exhibiting little change across Antarctica (Tarasick et al., 2019; Cooper et al., 2020). Based on reliable UV absorption measurements at remote locations (surface and lower troposphere), ozone trends since the mid-1990s varied spatially at northern mid-latitudes, but increased in the northern tropics (2-17%; 1-6 ppbv per decade) (Cooper et al., 2020; Gaudel et al., 2020). Across the SH these more recent observations are too limited to determine zonal trends (e.g. tropics, mid-latitudes, high latitudes).

The earliest observations of free tropospheric ozone (1934–1955) are available from northern mid-latitudes where limited data indicate a tropospheric column ozone increase of 48 ± 30% up to 1990–2012 (Tarasick et al., 2019). Starting in the 1960s, records from ozonesondes show no significant changes in the free troposphere over the Arctic and mid-latitude regions of Canada, but trends are mainly positive elsewhere in the northern mid-latitudes (Cooper et al., 2020; Oltmans et al., 2013). Tropospheric column and free tropospheric trends since the mid-1990s based on commercial aircraft, ozonesonde observations and satellite retrievals (Figure 2.8b,c), are overwhelmingly positive across the northern mid-latitudes (2–7%; 1–4 ppbv per decade) and tropics (2–14%; 1–5 ppbv per decade), with the largest increases (8–14%; 3–6 ppbv per decade) in the northern tropics in the vicinity of southern Asia and Indonesia. Observations in the SH are limited, but indicate average tropospheric column ozone increases of 2–12% (1–5 ppbv per decade) in the tropics (Figure 2.8c), and weak tropospheric column ozone increases (< 5%, < 1 ppbv per decade) at mid-latitudes (Cooper et al., 2020). Above Antarctica, mid-tropospheric ozone has increased since the late 20th century (Oltmans et al., 2013). The total ozone ERF from 1750 to 2019 best estimate is assessed as 0.47 W m² (7.3.2.5) and this is dominated by increases in the troposphere. The underlying modelled global tropospheric ozone column increase (Skeie et al., 2020) from 1850 to 2010 of 40–60%, is somewhat higher than the isotope based upper-limit of Yeung et al., (2019). At mid-latitudes (30°–60° N) model increases of 30–40% since the mid-20th century are broadly consistent with observations.

In summary, limited available isotopic evidence constrains the global tropospheric ozone increase to less than 40% between 1850 and 2005 (low confidence). Based on sparse historical surface/low altitude data tropospheric ozone has increased since the mid-20th century by 30–70% across the NH (medium confidence). Surface / low altitude ozone trends since the mid-1990s are variable at northern mid-latitudes, but positive in the tropics (2 to 17% per decade) (high confidence). Since the mid-1990s, free tropospheric ozone has increased by 2–7% per decade in most regions of the northern mid-latitudes, and 2–12% per decade in the sampled regions of the northern and southern tropics (high confidence). Limited coverage by surface observations precludes identification of zonal trends in the SH, while observations of tropospheric column ozone indicate increases of less than 5% per decade at southern mid-latitudes (medium confidence).

[START FIGURE 2.8 HERE]

Figure 2.8: Surface and tropospheric ozone trends. (a) Decadal ozone trends by latitude at 28 remote surface sites and in the lower free troposphere (650 hPa, about 3.5 km) as measured by IAGOS aircraft above 11 regions. All trends are estimated for the time series up to the most recently available year, but begin in 1995 or 1994. Colours indicate significance (p-value) as denoted in the in-line key. See Figure 6.5 for a depiction of these trends globally. (b) Trends of ozone since 1994 as measured by IAGOS aircraft in 11 regions in the mid-troposphere (700–300 hPa; about 3–9 km) and upper troposphere (about 10-12 km), as measured by IAGOS aircraft and ozonesondes. (c) Trends of average tropospheric column ozone mixing ratios from the TOST composite ozonesonde product and three composite satellite products based on TOMS, OMI/MLS (Sat1), GOME, SCIAMACHY, OMI, GOME-2A, GOME-2B (Sat2), and GOME, SCIAMACHY, GOME-II (Sat3). Vertical bars indicate the latitude range of each product, while horizontal lines indicate the very likely uncertainty range. Further details on data sources and processing...
are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.8 HERE]

2.2.6 Aerosols

AR5 assessed large-scale Aerosol Optical Depth (AOD) trends over 2000–2009, concluding that there was low confidence in a global trend, but that AOD very likely decreased from 1990 onwards over Europe and the eastern USA, and increased since 2000 over eastern and southern Asia. The ERF associated with aerosol-radiation interactions for 2011 (relative to 1750) was estimated to be -0.45 ± 0.5 W m⁻² and of aerosol-cloud interaction estimated as -0.45 [-1.2–0.0] W m⁻². Aerosol ERF uncertainty was assessed as the largest contributor to the overall ERF uncertainty since 1750.

This section assesses the observed large-scale temporal evolution of tropospheric aerosols. Aerosol-related processes, chemical and physical properties, and links to air quality, are assessed in Chapter 6. An in-depth assessment of aerosol interactions with radiation and clouds is provided in Section 7.3.3.

Aerosol proxy records of improved temporal resolution and quality are now available (Dornelas et al., 2018; Jacobel et al., 2017; Kylander et al., 2016; Middleton et al., 2018; Stevens et al., 2016; 2018), which further advance synthesis of new global compilations of aerosol loadings (Lambert et al., 2015; Albani et al., 2016). Estimates of the glacial / interglacial ratio in global dust deposition are within the range of 2–4 (Albani et al., 2015; Lambert et al., 2015). New reconstructions indicate a ratio of 3–5 for the glacial / interglacial loadings for mid- and high-latitude ocean of both hemispheres (Lamy et al., 2014; Martinez García et al., 2014; Serno et al., 2015). Improved quantification of changes in dust deposition from North Africa and North Atlantic sediment records confirms dust deposition rates lower by a factor 2–5 during the African Humid Period (10–5 ka) compared to the late Holocene (McGee et al., 2013; Albani et al., 2015; Middleton et al., 2018; Palchan & Torfstein, 2019). During the Holocene, biogenic emissions and volcanic activity drove significant variability (up to one order of magnitude) in sulphate concentrations (Schüpbach et al., 2018).

Ice cores allow for estimation of multi-centennial trends in mid- and high-latitude aerosol deposition, including those for sulphate and black carbon (Figure 2.9 a, b). Sulphate in ice cores increased by a factor of 8 from the end of the 19th century to the 1970s in continental Europe, by a factor of 4 from the 1940s to the 1970s in Russia, and by a factor of 3 from the end of the 19th century to 1950 in the Arctic (Svalbard). In all regions studied, concentrations have declined by about a factor of 2 following their peak (around 1970 in Europe and Russia, and 1950 in the Arctic). Strong increases of black carbon (BC) were observed in the 20th century over Europe, Russia, Greenland (primarily originating from emissions from North America), and in the Arctic (Svalbard). South America exhibits a small positive trend (Figure 2.9). BC concentrations in various Antarctic ice cores were below 1 ng g⁻¹ without a clear trend.

Spatially resolved trends of AOD derived from Aqua/Terra MISR and MODIS instruments over 2000–2019 range between -2% and 2% per year (Figure 2.9c). Ground-based solar attenuation networks help to constrain and improve the satellite-derived retrievals of AOD, and trends derived from the AERONET network (Figure 2.9c, d) corroborate satellite results (Bauer et al., 2020; Georgoulas et al., 2016; Wei et al., 2019; Yu et al., 2020) in particular for declines over Europe (Stjern et al., 2011; Cherian et al., 2014; Li et al., 2014) and the USA (Jongeward et al., 2016; Li et al., 2014). The tendency in AOD over East Asia reversed from positive (2000–2010) to negative (since 2010) (Filonchyk et al., 2019; Ma et al., 2019; Samset et al., 2019; Sogacheva et al., 2018). Over southern Asia, however, AOD from satellite (MODIS / MISR) and AERONET retrievals show continuing increases (Li et al., 2014; Zhao et al., 2017), with similar trends from UV-based aerosol retrievals from the Ozone Monitoring Instrument (OMI) on the Aura satellite (Dahutia et al., 2018; Hammer et al., 2018). A comparison of MODIS and MISR radiometric observations with the broadband CERES satellite instrument (Corbett and Loeb, 2015) showed that drifts in calibration are unlikely to affect the satellite derived trends. CERES shows patterns for clear-sky broadband radiation consistent with the aerosol spatio-temporal changes (Loeb et al., 2018; Paulot et al., 2018).

Satellite-derived trends are further supported by in situ regional surface concentration measurements,
Anthropogenic aerosol is predominantly found in the fraction of particles with radii <1 μm that comprise the fine-mode AOD (AODf) (Kinne, 2019) (Figure 2.9d). A significant decline in AODf of more than 1.5% per year from 2000 to 2019 has occurred over Europe and North America, while there have been positive trends of up to 1.5% per year over Southern Asia and East Africa. The global-scale trend in AODf of ~0.03% per year (Figure 2.9) is significant. The results are consistent with trend estimates from an aerosol reanalysis (Bellouin et al., 2020), and the trends in satellite-derived cloud droplet number concentrations are consistent with the aerosol trends (Cherian and Quaas, 2020). Cloudiness and cloud radiative properties trends are, however, less conclusive possibly due to their large variability (Norris et al., 2016; Cherian and Quaas, 2020). Further details on aerosol-cloud interactions are assessed in Section 7.3.3.2.

[START FIGURE 2.9 HERE]

Figure 2.9: Aerosol evolution from ice-core measurements. Changes are shown as 10-yr averaged time series (a, b) and trends in remote-sensing aerosol optical depth (AOD) and AODf (c, d). (a) Concentrations of non-sea salt (nss) sulphate (ng g⁻¹), (b) Black carbon (BC) in glacier ice from the Arctic (Lomonosovfonna), Russia (Belukha), Europe (Colle Gnifetti), South America (Illimani), Antarctica (stacked sulphate record, and BC from the B40 core), and BC from Greenland (stacked rBC record from Greenland and Eastern Europe (Elbrus)). (c) Linear trend in annual mean AOD retrieved from satellite data for the 2000–2019 period (% yr⁻¹). The average trend from MOderate Resolution Imaging Spectroradiometer (MODIS) and Multi-Angle Imaging Spectroradiometer (MISR) is shown. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al., (2008a). Superimposed are the trends in annual-mean AOD from the AERONET surface sunphotometer network for 2000–2019. (d) Linear trend in 2000–2019 as in (c), but for fine-mode AOD, AODf, and using only MISR over land. Significance is assessed following AR(1) adjustment after Santer et al., (2008a), ‘x’ marks denote non-significant trends. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.9 HERE]

To conclude, atmospheric aerosols sampled by ice cores, influenced by northern mid-latitude emissions, show positive trends from 1700 until the last quarter of the 20th century and decreases thereafter (high confidence), but there is low confidence in observations of systematic changes in other parts of the world in these periods. Satellite data and ground-based records indicate that AOD exhibits predominantly negative trends since 2000 over NH mid-latitudes and SH continents, but increased over South Asia and East Africa (high confidence). A globally decreasing aerosol abundance is thus assessed with medium confidence. This implies increasing net positive ERF, since the overall negative aerosol ERF has become smaller.

2.2.7 Land use and land cover

AR5 assessed that land use change very likely increased the Earth’s albedo with a radiative forcing of -0.15 (± 0.10) W m⁻². AR5 also assessed that a net cooling of the surface, accounting for processes that are not limited to the albedo, was about as likely as not. SRCCL concluded with medium confidence that the biophysical effects of land cover change (mainly increased albedo) had a cooling effect on surface temperatures. SRCCL also concluded with very high confidence that the biogeochemical effects of land cover change (i.e., GHG emissions) resulted in a mean annual surface warming.

Much of the global land surface has been modified or managed to some extent by human activities during the Holocene. Reconstructions based on pollen data indicate that natural vegetation probably covered most of
the Earth’s ice-free terrestrial surface until roughly the mid-Holocene (Harrison et al., 2020; Li et al., 2020; Marquer et al., 2017). Reconstructions based on pollen, archaeological, and historical data indicate deforestation at the regional scale since at least 6 ka (Harrison et al., 2020; Li et al., 2020; Marquer et al., 2017; Stephens et al., 2019). From a global perspective, land-use forcing datasets (Lawrence et al., 2016) estimate that changes in land use (and related deforestation) were small on the global scale until the mid-19th century and accelerated markedly thereafter, with larger uncertainties prior to industrialization (Kaplan et al., 2017). Since the early 1980s, about 60% of all land cover changes have been associated with direct human activities, with spatial patterns emphasizing the regional character of land use and land management, including tropical deforestation, temperate afforestation, cropland intensification, and increased urbanization (Song et al., 2018, Zeng et al., 2018). At present, nearly three-quarters of the ice-free terrestrial surface is under some form of human use (Venter et al., 2016; Erb et al., 2017), particularly in agriculture and forest management.

The impact of historical land-cover change on global climate is assessed with model simulations that consider multiple climate and biophysical processes (e.g., changes in albedo, evapotranspiration, and roughness) and/or biogeochemical processes (e.g., changes in atmospheric composition such as carbon release from deforestation). The dominant biophysical response to land cover changes is albedo, which is estimated (using a MODIS albedo product and a historical land-use harmonization product) to have increased gradually prior to the mid-19th century and then strongly through the mid-20th century, with a slightly slower rise thereafter (Ghimire et al., 2014). Recent radiative forcing estimates arising from biophysical processes generally fall at the lower end of the AR5 assessed range. For instance, based on historical simulations from 13 CMIP6 models, Smith et al., (2020) estimated that the ERF from surface albedo changes (including snow cover and leaf area) was $-0.08 (-0.22$ to $0.06) \text{ W m}^{-2}$ since 1850. Similarly, based on simulations from 13 CMIP5 models, Lejeune et al., (2020) estimated the radiative forcing from transitions between trees, crops, and grasslands was $-0.11 (-0.16$ to $0.04) \text{ W m}^{-2}$ since 1860. Andrews et al., (2017) identified an ERF of $-0.40 \text{ W m}^{-2}$ since 1860, ascribing much of the effect to increases in albedo (including the unmasking of underlying snow cover); notably, however, the analysis was based on a single model with a known tendency to overestimate the ERF (Collins et al., 2011). Ward et al. (2014) examined the combined effects of biophysical and biogeochemical processes, obtaining an RF of $0.9 \pm 0.5 \text{ W m}^{-2}$ since 1850 that was driven primarily by increases in land-use related GHG emissions from deforestation and agriculture (Ward and Mahowald, 2015). According to a large suite of historical simulations, the biophysical effects of changes in land cover (i.e. increased surface albedo and decreased turbulent heat fluxes) led to a net global cooling of $0.10 \pm 0.14 ^\circ \text{C}$ at the surface (SRCCL). Available model simulations suggest that biophysical and biogeochemical effects jointly may have contributed to a small global warming of $0.078 \pm 0.093 ^\circ \text{C}$ at the surface over about the past two centuries (SRCCL), with a potentially even larger warming contribution over the Holocene as a whole (He et al., 2014).

In summary, biophysical effects from historical changes in land use have an overall negative ERF (medium confidence). The best-estimate ERF from the increase in global albedo is $-0.15 \text{ W m}^{-2}$ since 1700 and $-0.12 \text{ W m}^{-2}$ since 1850 (medium confidence; Section 7.3.4.1). Biophysical effects of land-use change likely resulted in a net global cooling of about $0.1 ^\circ \text{C}$ since 1750 (medium confidence; Section 7.3.5.3).

### 2.2.8 Effective radiative forcing (ERF) exerted by the assessed climate drivers

AR5 concluded that changes in climate drivers over the industrial period corresponded to a positive ERF which increased more rapidly after 1970 than before. There was very high confidence in the positive ERF due to WMGHG, with CO$_2$ the single largest contributor. AR5 concluded that there was high confidence that aerosols have offset a substantial portion of the WMGHG forcing.

This section reports the evolution in ERF with respect to 1750 as assessed in Section 7.3 and relies on the observed changes in climate drivers as assessed in Section 2.2 wherever possible, and models otherwise. The ERF is assessed using the methods and details described in Section 7.3.1 and includes, in addition to the radiative forcing, the rapid adjustments, especially implied by clouds. The time series are shown in Figure 2.10.
Increasing TSI (Section 2.2.1) implies a small ERF of less than 0.1 W m\(^{-2}\) between 1900 and 1980. TSI varies over the 11-year solar cycle with ERF of order ± 0.1 W m\(^{-2}\) in the assessed period. Strong volcanic eruptions (Section 2.2.2) with periods of strong negative ERF lasting 2–5 years in duration occurred in the late 19th and early 20th centuries. There followed a relatively quiescent period between about 1920 and 1960, and then three strong eruptions in 1963, 1982 and 1991, and only small-to-moderate eruptions thereafter (Schmidt et al., 2018).

The atmospheric concentrations of WMGHGs (Section 2.2.3) have continuously increased since the early 19th century, with CO\(_2\) contributing the largest share of the positive ERF. Compared to the last two decades of the 20th century, the growth rate of CO\(_2\) in the atmosphere increased in the 21st century, showed strong fluctuations for CH\(_4\), and was about constant for N\(_2\)O. Mixing ratios of the most abundant CFCs declined (Section 2.2.4). Mixing ratios of HCFCs increased, but growth rates are starting to decelerate. Mixing ratios of HFCs and some other human-made components are increasing (Section 2.2.4). The ERF for CO\(_2\) alone is stronger than for all the other anthropogenic WMGHGs taken together throughout the industrial period, and its relative importance has increased in recent years (Figures 2.10, 7.6).

Among the gaseous short-lived climate forcers (Section 2.2.5, Chapter 6, Section 7.3; excluding CH\(_4\) here), ozone (O\(_3\)) is the component with the largest (positive) ERF. Concentrations from direct observations have increased since the mid-20th century and, mostly based on models, this extends to since 1750. Other gaseous short-lived climate forcers have small contributions to total ERF.

The net effect of aerosols (Sections 2.2.6, 6.4) on the radiation budget, including their effect on clouds, and cloud adjustments, as well as the deposition of black carbon on snow (Section 7.3.4.3), was negative throughout the industrial period (high confidence). The net effect strengthened (becoming more negative) over most of the 20th century, but more likely than not weakened (becoming less negative) since the late 20th century. These trends are reflected in measurements of surface solar radiation (Section 7.2.2.3) and the Earth’s energy imbalance (Section 7.2.2.1). The relative importance of aerosol forcing compared to other forcing agents has decreased globally in the most recent 30 years (medium confidence) and the reduction of the negative forcing in the 21st century enhances the overall positive ERF.

Land use and land cover changes (Section 2.2.7) over the industrial period introduce a negative radiative forcing by increasing the surface albedo. This effect increased since 1750, reaching current values of about –0.20 W m\(^{-2}\) (medium confidence). This ERF value is taken from Section 7.3.4.1 and is different from the assessment in Section 2.2.7 in that it also includes the effect of irrigation. It also includes uncertain rapid adjustments and thus there is low confidence in its magnitude. Biogeochemical feedbacks can be substantial (Section 5.4) and are not included in ERF.

In conclusion, the net ERF due to all observed changes in climate drivers is positive, except for short periods (up to a few years in duration) following moderate to large volcanic eruptions, and has grown in magnitude since the late 19th century. The rate of change likely has increased in the last 30 years, since CO\(_2\) concentrations increased at an increasing rate due to growing CO\(_2\) emissions (very likely), and since the aerosol forcing became less negative (more likely than not).
2.3 Changes in large-scale climate

[START CROSS-CHAPTER BOX 2.2 HERE]

Cross-Chapter Box 2.2: Large-scale indicators of climate change

Contributing Authors: Veronika Eyring (Germany), Nathan Gillett (Canada), Sergey Gulev (Russian Federation), Jochen Marotzke (Germany), June-Yi Lee (Republic of Korea), Peter Thorne (Ireland/UK)

Chapters 2, 3 and 4 assess the current evidence basis for climatic changes, their causes, and their potential future under different possible emissions pathways using a combination of observations and state-of-the-art Earth System Models (ESMs). The assessment in these chapters focuses on selected large-scale indicators and modes as defined in this Box. These indicators and modes of variability taken together characterise overall changes to the climate system as a whole.

Defining large-scale
Understanding of large-scale climate variability and change requires knowledge of both the response to forcings and the role of internal variability. Many forcings have substantial hemispheric or continental scale variations. Modes of climate variability are generally driven by ocean basin scale processes. The climate system involves process interactions from the micro- to the global-scale and as such, any threshold for defining "large-scale" is arbitrary, but, within these chapters on the basis of these considerations large-scale is defined to include ocean basin and continental scales as well as hemispheric and global scales.

Defining a key set of climate indicators
Key climate indicators should constitute a finite set of distinct variables and / or metrics that may collectively point to important overall changes in the climate system that provide a synthesis of climate system evolution and are of broad societal relevance. Key indicators have been selected across the atmospheric, oceanic, cryospheric and biospheric domains, with land as a cross-cutting component. These indicators, and their use across chapters 2 to 4, as well as the broader report, are summarized in Cross-Chapter Box 2.2 Table 1. All selected indicators are Essential Climate Variables as defined by the Global Climate Observing System (Bojinski et al., 2014).

[START CROSS-CHAPTER BOX 2.2, TABLE 1 HERE]

Cross-Chapter Box 2.2, Table 1: Summary of the large-scale indicators used across chapters 2 through 4 and their principal applications in remaining chapters. Indicators are sub-divided by Earth-system domain and their inclusion in individual chapters is indicated by an orange solid fill. The list of additional chapters is limited to those where the variable is a principal consideration.

<table>
<thead>
<tr>
<th>Selected large-scale indicator of climate change</th>
<th>Used in Chapters</th>
<th>Additional regional or process-based assessments undertaken in Chapters</th>
</tr>
</thead>
<tbody>
<tr>
<td>Atmosphere and surface</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Surface and upper air temperatures</td>
<td>2, 3, 4</td>
<td>7, 10, 11, 12, Atlas</td>
</tr>
<tr>
<td>Hydrological cycle components (surface humidity, precipitable water vapour, precipitation, runoff, Precipitation-Evaporation)</td>
<td></td>
<td>8, 10, 11, 12, Atlas</td>
</tr>
<tr>
<td>Atmospheric circulation (sea level pressure and winds, Hadley / Walker circulation, global monsoons, blocking, storm tracks and jets, sudden stratospheric warmings)</td>
<td></td>
<td>8, 10, 11, 12</td>
</tr>
<tr>
<td>Cryosphere</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Sea ice extent / area, seasonality and thickness</td>
<td>2, 3, 4</td>
<td>9, 12, Atlas</td>
</tr>
<tr>
<td>Terrestrial snow cover</td>
<td></td>
<td>8, 9, 12, Atlas</td>
</tr>
<tr>
<td>Glacier mass and extent</td>
<td></td>
<td>8, 9, 12</td>
</tr>
<tr>
<td>Ice sheet mass and extent</td>
<td></td>
<td>9, 12, Atlas</td>
</tr>
<tr>
<td>Terrestrial permafrost temperature and active layer</td>
<td></td>
<td>5, 9, 12</td>
</tr>
</tbody>
</table>
Atmospheric indicators
Monitoring surface temperatures is integral to the negotiations of UNFCCC and the global mean temperature goals of the Paris Agreement. Upper-air temperatures are a key indicator of different causal mechanisms underlying climate change and underpinned the very first conclusion in the Second Assessment Report of a ‘discernible human influence’ (Santer et al., 1996). To ascertain large-scale changes of and human influence on the global hydrological cycle (which includes terrestrial and oceanic components), a small subset of indicators across this cycle are chosen: ocean and land precipitation-evaporation (P-E), global precipitation, total column water vapour, surface humidity (specific and relative), and global river runoff. Chapter 8 performs a substantive and holistic assessment of a much broader range of components. Finally, a warming world may be accompanied by a change in large-scale circulation patterns linked through energy / mass / momentum constraints such as the extent and strength of the Hadley circulation (HC), monsoon systems, and / or the position and strength of the sub-tropical and polar jets.

Cryospheric indicators
Changes in ice sheets are indicators of the longest-term impacts of climate change and associated with changes in global and regional sea level. Seasonal snow cover has many implications for mid- to high-latitude regions (albedo, hydrological cycle, etc.) with impacts on biospheric components of the system.
Changes in sea ice extent, seasonality and thickness have potential impacts for hemispheric-scale circulation (Cross-Chapter Box 10.1). Changes in glacier mass balance contribute to changes in sea level but also have substantial implications for water supply for a substantial proportion of the global population. Finally, changes in permafrost and the seasonally thawed active layer have substantial implications in mid- to high-latitudes and have been hypothesised to be important in potential feedbacks through degassing of WMGHGs as the permafrost thaws.

Oceanic indicators
Most of the energy imbalance (Box 7.2) in the climate system is taken up by the ocean, resulting in changes in ocean temperature and heat content. Salinity changes indicate broad-scale hydrological cycle and circulation changes. Global mean sea-level change is a key indicator of the impacts of both global warming and changes in global ice volume. Furthermore, it is integral to assessing the global energy budget (Cross-Chapter Box 9.1). The oceanic overturning circulation redistributes heat, carbon, oxygen and salinity within the ocean. Declines in ocean pH result from air-sea exchange of carbon dioxide and loss of ocean oxygen results from ocean warming; both lead to changes in marine ecosystems.

Biospheric indicators
The seasonal cycle of CO₂ is an integrated measure of the biogeochemical activity across the global biosphere. Changes in marine and terrestrial ecosystems can also be observed directly at large scales. For small, free-floating organisms such as phytoplankton, the dynamics can be rapid in nature, whereas on land slower changes in plant assemblages may occur, with commensurate changes in altitude and latitude of the tree-line. Lengthening of the growing season and the associated changes in phenology, distribution and abundance of species would be expected in most of the extratropics. Biospheric indicators and their impacts

<table>
<thead>
<tr>
<th>thickness</th>
<th>Ocean</th>
<th>5, 7, 9, 12, Atlas</th>
</tr>
</thead>
<tbody>
<tr>
<td>Temperature / ocean heat content</td>
<td>Salinity</td>
<td>8, 9, 12</td>
</tr>
<tr>
<td>Sea-level</td>
<td>Circulation</td>
<td>9, 12</td>
</tr>
<tr>
<td>pH and deoxygenation</td>
<td>Biosphere</td>
<td>5, 9</td>
</tr>
<tr>
<td>Seasonal cycle of CO₂</td>
<td>Marine biosphere (distribution of marine biota, primary production, phenology)</td>
<td>5, 12</td>
</tr>
<tr>
<td>Terrestrial biosphere (distribution of terrestrial biota, global greening / browning, growing season)</td>
<td>5, 12, Atlas</td>
<td></td>
</tr>
</tbody>
</table>
Defining a selection of modes of variability

Many modes of climate variability affect global, hemispheric or regional climate across a range of timescales. Conversely, their behaviour may be influenced by global climate change. Modes were selected for inclusion that: i) have effects at large spatial scales; and ii) have substantial potential to modify interannual to multidecadal climate. The selected modes are considered in multiple chapters (Cross-Chapter Box 2.2 Table 2) and are defined in Annex IV.

### Cross-Chapter Box 2.2, Table 2

<table>
<thead>
<tr>
<th>Selected mode of variability</th>
<th>Used in Chapters</th>
<th>Additional regional or process-based assessments undertaken in chapters</th>
</tr>
</thead>
<tbody>
<tr>
<td>El Niño Southern Oscillation</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>Indian Ocean Basin and dipole modes</td>
<td>8, 10</td>
<td></td>
</tr>
<tr>
<td>Atlantic Multidecadal Variability</td>
<td>8, 10</td>
<td></td>
</tr>
<tr>
<td>Pacific Decadal Variability</td>
<td>8, 10</td>
<td></td>
</tr>
<tr>
<td>Annular modes (NAO / NAM, SAM)</td>
<td>8, 10</td>
<td></td>
</tr>
<tr>
<td>Atlantic Meridional and Zonal Modes</td>
<td>8</td>
<td></td>
</tr>
</tbody>
</table>

2.3.1 Atmosphere and Earth’s surface

2.3.1.1 Surface temperatures

2.3.1.1.1 Temperatures of the deep past (65 Ma to 8 ka)

This assessment of the paleo reference periods (Cross-Chapter Box 2.1) draws from studies based mostly or entirely on indirect observational evidence from geological archives (i.e. proxy records) rather than reconstructions that rely more heavily on modelled parameters and those based on deep-ocean temperatures (e.g. Köhler et al., 2015; Friedrich et al., 2016). In contrast to AR5, temperature estimates from climate models are not included in the assessed values for paleo reference periods in this chapter. AR5 concluded that the reconstructed GMST during the PETM was 4°C–7°C warmer than pre-PETM mean climate (low confidence), and that the EECO and the MPWP were 9°C–14°C and 1.9°C–3.6°C warmer than pre-industrial, respectively (medium confidence). The GMST during the LIG was assessed at 1°C–2°C warmer than pre-industrial (medium confidence), whereas SROCC narrowed the range to 0.5°C–1.0°C warmer, but did not state a confidence level. AR5 further concluded that it was very likely that the LGM was 3°C–8°C colder than pre-industrial, and likely that the maximum rate of global warming during the subsequent deglacial period was 1°C–1.5°C kyr⁻¹.

For the PETM, new reconstructions agree with those assessed by AR5. A major new compilation of proxy temperature data (Hollis et al., 2019) analysed using multiple statistical approaches (Inglis et al., 2020) indicates that GMST was 10°C–25°C (90% range) warmer than 1850–1900, or about 5°C warmer relative to the pre-PETM state. A related synthesis study also estimates that PETM warmed by 5°C (no uncertainty assigned; Zhu et al., 2019). A recent benthic isotope compilation (Westerhold et al., 2020) transformed to GMST based on the formulation by Hansen et al. (2013c) (Cross-Chapter Box 2.1, Figure 1), and adjusted to 1850–1900 by adding 0.36°C, shows an increase of GMST by about 10°C during the PETM. This reflects the...
expected higher variability at single sites that were used to splice together the composite time series, compared to the globally averaged composite time series of Zachos et al. (2008). The latter was originally used by Hansen et al. (2013c) to reconstruct GMST, and is the preferred representation of the global average bottom water conditions, despite its less well-refined chronology.

For the EECO, new GMST reconstructions fall at the high end of the range assessed by AR5. These include estimates of 7°C–18°C (90% range; Inglis et al., 2020) and 12°C–18°C (95% range; Zhe et al., 2019) warmer than 1850–1900, and 10°C–16°C warmer than 1995-2014 “recent past” conditions (2 standard error range; Caballero & Huber, 2013). Together, they indicate that GMST was 10°C–18°C warmer during the EECO compared with 1850–1900 (medium confidence).

AR5 did not assess the GMST for the MCO. Reconstructions based on data from multiple study sites include estimates of about 4°C (uncertainty range not specified; You et al., 2009) and 5°C–10°C (2 standard error range; Goldner et al., 2014) warmer than 1850–1900. Together, these studies indicate that GMST was 4°C–10°C warmer during the MCO (medium confidence).

For the MPWP, new proxy-based estimates of global sea surface temperatures (SST) are about 2.0°C–3.5°C warmer than 1850–1900, depending on which proxy types are included in the analysis (Foley and Dowsett, 2019; McClymont et al., 2020). On the basis of model-derived relationships between land versus sea surface temperatures under different climate states (Section 3.3.1.1, Figure 3.2b), the increase in GMST is estimated to have been roughly 15% greater than the increase in global SST. Therefore, GMST during the MPWP is estimated to have been 2.5°C–4.0°C warmer than 1850–1900 (medium confidence).

For the LIG (Cross-Chapter Box 2.1, Figure 1; Figure 2.1), a major new compilation of marine proxy data (Turney et al., 2020) from 203 sites indicates that the average SST from 129–125 ka was 1.0°C ± 0.2°C (2SD) warmer than 1850–1900 (reported relative to 1981–2010 and adjusted here by 0.8°C). These temperatures represent the time of peak warmth, which may not have been synchronous among these sites. This compares with two other SST estimates for 125 ka of 0.5°C ± 0.3°C (± 2 SD) warmer at 125 ka relative to 1870–1889 (Hoffmann et al., 2017), and about 1.4°C (no uncertainty stated) warmer at 125 ka relative to 1850–1900 (Friedrich and Timmermann, 2020; reported relative to 10–5 ka and adjusted here by 0.4°C; (Kaufman et al., 2020a)). The average of these post-AR5 global SST anomalies is 1°C. Commensurately (Figure 3.2b), GMST is estimated to have been roughly 1.1°C above 1850-1900 values, although this value could be too high if peak warmth was not globally synchronous (Capron et al., 2017). A further estimate of peak GMST anomalies of 1.0°C–3.5°C (90% range; adjusted here to 1850–1900 by adding 0.2°C) based on 59 marine sediment cores (Snyder, 2016) is considerably warmer than remaining estimates and are therefore given less weight in the final assessment. The warmest millennium of the LIG GMST reconstruction in the Hansen et al. (2013c) is 1.5°C above 1850-1900. In summary, GMST during the warmest millennia of the LIG (within the interval of around 129–125 ka) is estimated to have reached 0.5°C–1.5°C higher values than the 1850–1900 reference period (medium confidence).

New GMST reconstructions for the LGM fall near the middle of AR5’s very likely range, which was based on a combination of proxy reconstructions and model simulations. Two of these new reconstructions use marine proxies to reconstruct global SST that were scaled to GMST based on different assumptions. One indicates that GMST was 6.2°C [4.5 to 8.1°C; 95% range] cooler than the late Holocene average (Snyder, 2016), and the other, 5.7°C ± 0.8°C (2 SD) cooler than the average of the first part of the Holocene (10–5 ka) (Friedrich and Timmermann, 2020). A third new estimate (Tierney et al., 2020) uses a much larger compilation of marine proxies along with a data-assimilation procedure, rather than scaling, to reconstruct a GMST of 6.1°C ± 0.4°C (2 SD) cooler than the late Holocene. Assuming that the 1850–1900 reference period was 0.2°C and 0.4°C cooler than the late and first part of the Holocene, respectively (Kaufman et al., 2020a), the midpoints of these three new GMST reconstructions average –5.8°C relative to 1850–1900. The coldest multi-century period of the LGM in the Hansen et al. (2013c) reconstruction is 4.3°C colder than 1850–1900. This compares to land- and SST-only estimates of about –6.1°C ± 2°C and –2.2°C ± 1°C, respectively (2 SD), which are based on AR5-generation studies that imply a warmer GMST than more recent reconstructions (Fig. 1c in Harrison et al., 2015; Fig. 7 in Harrison et al., 2016). A major new pollen-based data-assimilation reconstruction averages 6.9°C cooler over northern extratropical land (Cleator et al., 2020). LGM temperature variability on centennial scales was about four times higher globally than during
the Holocene, and even greater at high latitudes (Rehfeld et al., 2018). In summary, GMST is estimated to have been 5°C–7°C lower during the LGM (around 23–19 ka) compared with 1850–1900 (medium confidence).

For the LDT (Cross-Chapter Box 2.1, Figure 1), no new large-scale studies have been published since AR5 (Shakun et al., 2012) to further assess the rate of GMST change during this period of rapid global warming (estimated at 1°C–1.5°C per kyr). The reconstruction of Shakun et al. (2012) was based primarily on SST records and therefore underrepresents the change in GMST during the LDT. Temperature over Greenland increased by about ten times that rate during the centuries of most rapid warming (Jansen et al., 2020).

2.3.1.1.2 Temperatures of the post-glacial period (past 7000 years)

AR5 did not include an assessment of large-scale temperature estimates for the MH, although it assigned high confidence to the long-term cooling trend over mid- to high-latitudes of the NH during the 5 kyr that preceded recent warming. For average annual NH temperatures, the period 1983–2012 was assessed as very likely the warmest 30-year period of the past 800 years (high confidence) and likely the warmest 30-year period of the past 1.4 kyr (medium confidence); the warm multi-decadal periods prior to the 20th century were unsynchronized across regions, in contrast to the warming since the mid-20th century (high confidence), although only sparse information was available from the SH.

This section concerns the Holocene period prior to industrialization when GMST was overall highest. Whereas SR1.5 focussed upon the ‘Holocene thermal maximum’ when regional temperatures were up to 1°C higher than 1850–1900, though peak warming occurred regionally at different times between around 10 and 5 ka greatly complicating interpretation. A multi-method reconstruction (Kaufman et al., 2020a) based on a quality-controlled, multi-proxy synthesis of paleo-temperature records from 470 terrestrial and 209 marine sites globally (Kaufman et al., 2020b) indicates that the median GMST of the warmest two-century-long interval was 0.7 [0.3 to 1.8] °C warmer than 1800–1900 (which averaged 0.03°C colder than 1850–1900; PAGES 2k Consortium, 2019), and was centred around 6.5 ka. This is similar to Marcott et al. (2013), which is based on a smaller dataset (73 sites) and different procedures to estimate a maximum warmth of 0.8°C ± 0.3°C (2 SD) at around 7.0 ka, adjusted here by adding 0.3°C to account for differences in reference periods. These may be underestimated because averaging inherently smoothed proxy records with uncertain chronologies reduces the variability in the temperature reconstruction (e.g., Dolman & Laepple (2018) for sedimentary archives). However, the general coincidence between peak warmth and astronomically driven boreal summer insolation might reflect a bias toward summer conditions (Bova et al., 2021; Hou et al., 2019; Liu et al., 2014), suggesting that the estimate is too high. This possibility is supported by AR5-generation proxy data focusing on 6 ka (Harrison et al., 2014), the long-standing MH modelling target (Cross-Chapter Box 2.1), that indicate surface temperatures for land and ocean were indistinguishable from ‘pre-industrial’ climate (Fig. 1e in Harrison et al., 2015; Fig. 7 in Harrison et al., 2016). In contrast, the GMST estimate from the multi-method global reconstruction (Kaufman et al., 2020a) for the millennium centred on 6 ka is only about 0.1°C colder than the warmest millennium.

Taking all lines of evidence into account, the GMST averaged over the warmest centuries of the current glacial period (sometime between around 6 and 7 ka) is estimated to have been 0.2°C–1.0°C higher than 1850–1900 (medium confidence). It is therefore more likely than not that no multi-centennial interval during the post-glacial period was warmer globally than the most recent decade (which was 1.1°C warmer than 1850–1900; Section 2.3.1.1.3); the LIG (129–116 ka) is the next most recent candidate for a period of higher global temperature. Zonally averaged mean annual temperature reconstructions (Routson et al., 2019) indicate that MH warmth was most pronounced north of 30°N latitude, and that GMST subsequently decreased in general, albeit with multi-century variability, with greater cooling in the NH than in the SH (Kaufman et al., 2020a).

The temperature history of the last millennium and the methods used to reconstruct it have been studied extensively, both prior to and following AR5, as summarized recently by Smerdon and Pollack (2016) and Christiansen and Ljungqvist (2017). New regional (e.g., Shi et al., 2015; Stenni et al., 2017; Werner et al., 2018), global ocean (McGregor et al., 2015), quasi-hemispheric (Neukom et al., 2014; Schneider et al., 2015;
Anchukaitis et al., 2017, and global (Tardif et al., 2019) temperature reconstructions, and new regional proxy data syntheses (Lüning et al., 2019a,b) have been published, extending back 1-2 kyr. In addition, a major new global compilation of multiproxy, annually resolved paleo-temperature records for the CE (PAGES 2k Consortium, 2017) has been analysed using a variety of statistical methods for reconstructing temperature (PAGES 2k Consortium, 2019). The median of the multi-method GMST reconstruction from this synthesis (Figure 2.11a) generally agrees with the AR5 assessment, while affording more robust estimates of the following major features of GMST during the CE: (1) an overall millennial-scale cooling trend of –0.18 [–0.28 to 0.00] °C kyr⁻¹ prior to 1850; (2) a multi-centennial period of relatively low temperature beginning around the 15th century, with GMST averaging –0.03 [–0.30 to 0.06] °C between 1450 and 1850 relative to 1850–1900; (3) the warmest multi-decadal period occurring most recently; and (4) the rate of warming during the second half of the 20th century (from instrumental data) exceeding the 99th percentile of all 51-year trends over the past 2 kyr. Moreover, the new proxy data compilation shows that the warming of the 20th century was more spatially uniform than any other century-scale temperature change of the CE (medium confidence) (Neukom et al., 2019). A new independent temperature reconstruction extending back to 1580 is based on an expanded database of subsurface borehole temperature profiles, along with refined methods for inverse modelling (Cuesta-Valero et al., 2021). The borehole data, converted to GMST based on the modelled relation between changes in land versus sea surface temperature outlined previously, indicate that average GMST for 1600–1650 was 0.12°C colder than 1850–1900, which is similar to the PAGES 2k reconstruction (0.09°C colder), although both estimates are associated with relatively large uncertainties (0.8°C (95% range) and 0.5°C (90% range), respectively).

To conclude, following approximately 6 ka, GMST generally decreased, culminating in the coldest multi-century interval of the post-glacial period (since 8 ka), which occurred between around 1450 and 1850 (high confidence). This multi-millennial cooling trend was reversed in the mid-19th century. Since around 1950, GMST has increased at an observed rate unprecedented for any 50-year period in at least the last 2000 years (medium confidence).

[START FIGURE 2.11 HERE]

Figure 2.11: Earth’s surface temperature history with key findings annotated within each panel. (a) GMST over the Holocene divided into three time scales. (i) 12 kyr–1 kyr in 100-year time steps, (ii) 1000–1900 CE, 10-year smooth, and (iii) 1900–2020 CE (from panel c). Median of the multi-method reconstruction (bold lines), with 5th and 95th percentiles of the ensemble members (thin lines). Vertical bars are the assessed medium confidence ranges of GMST for the Last Interglacial and mid-Holocene (Section 2.3.1.1). The last decade value and very likely range arises from 2.3.1.1.3. (b) Spatially resolved trends (°C per decade) for HadCRUT5 over (upper map) 1900–1980, and (lower map) 1981–2020. Significance is assessed following AR(1) adjustment after Santer et al., (2008a), ‘x’ marks denote non-significant trends. (c) Temperature from instrumental data for 1850–2020, including (upper panel) multi-product mean annual timeseries assessed in Section 2.3.1.1.3 for temperature over the ocean (blue line) and temperature over the land (red line) and indicating the warming to the most recent 10 years; and annually (middle panel) and decadally (bottom panel) resolved averages for the GMST datasets assessed in Section 2.3.1.1.3. The grey shading in each panel shows the uncertainty associated with the HadCRUT5 estimate (Morie et al., 2021). All temperatures relative to the 1850–1900 reference period. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.11 HERE]

[START CROSS-CHAPTER BOX 2.3 HERE]

Cross-Chapter Box 2.3: New estimates of global warming to date and key implications

Contributing Authors: Blair Trewin (Australia), Peter Thorne (Ireland / UK), Richard Allan (UK), Richard Betts (UK), Lea Beusch (Switzerland), Chris Fairall (USA), Piers Forster (UK), Baylor Fox-Kemper (USA), Jan Sigurd Fuglestvedt (Norway), John Fyfe (Canada), Nathan Gillett (Canada), Ed Hawkins (UK), Chris
This Cross-Chapter Box presents the AR6 WGI assessment of observed global warming and describes improvements and updates since AR5 and subsequent Special Reports. The revised estimates result from: the availability of new and revised observational datasets; the occurrence of recent record warm years; and the evaluation of the two primary metrics used to estimate global warming in past IPCC reports: “Global mean surface temperature” (GMST) and “Global surface air temperature” (GSAT). Implications for threshold crossing times, remaining carbon budgets and impacts assessments across AR6 WGs are discussed.

Dataset innovations

Since AR5, all major datasets used for assessing observed temperature change based upon GMST have been updated and improved (Section 2.3.1.1.3). A number of new products have also become available, including new datasets (e.g. Berkeley Earth, Rohde & Hausfather, 2020) and new interpolations based on existing datasets (e.g. Cowtan and Way (2014) and Kadow et al. (2020a)). These various estimates are not fully independent.

Improvements in global temperature datasets since AR5 have addressed two major systematic issues. First, new SST datasets (Huang et al., 2017; Kennedy et al., 2019) address deficiencies previously identified in AR5 relating to the shift from predominantly ship-based to buoy-based measurements; these improvements result in larger warming trends, particularly in recent decades. Second, all datasets now employ interpolation to improve spatial coverage. This is particularly important in the Arctic, which has warmed faster than the rest of the globe in recent decades (Atlas 5.9.2.2); under-sampling of the Arctic leads to a cool bias in recent decades (Simmons et al., 2017; Benestad et al., 2019). Some datasets are now spatially complete (Cowtan and Way, 2014; Kadow et al., 2020a) while others have expanded spatial coverage (Lenssen et al., 2019; Morice et al., 2021; Rohde & Hausfather, 2020; Vose et al., 2021). Several interpolation methods have been benchmarked against test cases (e.g. Lenssen et al., 2019), and comparisons with reanalyses further confirm the value of such interpolation (Simmons et al., 2017). It is extremely likely that interpolation produces an improved estimate of the changes in GMST compared to ignoring data void regions.

Overall, dataset innovations and the availability of new datasets have led to an assessment of increased GMST change relative to the directly equivalent estimates reported in AR5 (Cross-Chapter Box 2.3, Table 1 and Figure 1).

Effects of warming since AR5 and choice of metrics of global mean temperature change

Each of the six years from 2015 to 2020 has likely been warmer than any prior year in the instrumental record. GMST for the decade 2011–2020 has been 0.19 [0.16 to 0.22] °C warmer than 2003–2012, the most recent decade used in AR5 (Cross-Chapter Box 2.3, Figure 1). A linear trend has become a poorer representation of observed change over time since most of the sustained warming has occurred after the 1970s (Cross-Chapter Box 2.3 Figure 1) and all values since 2012 are at least 0.2°C above a linear trendline for 1850–2020. For this reason, the primary method used to assess observed warming in this report is the change in temperature from 1850–1900 to the most recent decade (2011–2020) or the recent past (1995–2014), replacing the trend-based methods used in AR5 and earlier assessments. The effect of this change from trend-based to change-based metrics is currently relatively minor at -0.03°C (<5%) for the most recent decade, but this may not remain the case in future (high confidence).

Observed changes in global mean temperature since the pre-industrial era

AR5 used 1850–1900 as an approximate pre-industrial baseline for global temperature change, whilst using an earlier pre-industrial baseline of 1750 for radiative forcings. Cross-Chapter Box 1.2 assesses that there
was an observed GMST change from the period around 1750 to 1850 of around 0.1°C (likely range -0.1 to +0.3°C, medium confidence). This additional global temperature change before 1850–1900 is not included when making AR6 assessments on global warming to date, global temperature threshold crossing times, or remaining carbon budgets to ensure consistency with previous ARs.

Addressing the non-equivalence of GMST and GSAT

GMST is a combination of land surface air temperatures (LSAT) and SSTs, whereas GSAT is a combination of LSAT and marine air temperatures (MATs). Although GMST and GSAT are closely related, the two measures are physically distinct. The implications have become more apparent since AR5 (Cowtan et al., 2015; IPCC, 2018 (SR1.5); Merchant et al., 2013; Richardson et al., 2018; Simmons et al., 2017), and it has been shown (Rubino et al., 2020) that MAT and SST can show distinct multidecadal-scale trends and patterns of interannual variability. Although the SR1.5 used GMST for observational-based and GSAT for model-based headline warming statements, they noted the importance of the difference for their assessment (SR1.5 Section 1.2.1.1). SR1.5 used information from CMIP5 models to estimate a GSAT equivalent from observation-based GMST for certain applications such as remaining carbon budgets. The following subsections assess available lines of evidence related to the equivalence between GMST and GSAT.

Physical understanding

A well-understood physical constraint on the vertical gradient between the air and sea surface temperature is that it is approximately proportional to the turbulent sensible heat flux in the atmospheric surface layer (Chor et al., 2020). Similarly, the latent heat flux scales with the vertical humidity gradient and, in the global mean and in most oceanic regions, the latent heat flux is substantially larger than the sensible heat flux (Sections 7.2.1, 9.2.1.3). If GSAT were to warm faster than GMST, the sensible surface heat flux would respond so as to reduce this difference. However, it is the sum of the sensible, latent, and radiative heat fluxes that controls GMST, so the sensible heat flux effect cannot be considered in isolation. Attempts to further constrain the combination of fluxes (e.g., Lorenz et al., 2010; Siler et al., 2019) rely on parameterisations or output from Earth system models (ESMs) or reanalyses and so are not considered independent. Apart from the above global considerations, regional and seasonal effects such as changes to the frequency and intensity of storms, sea state, cloudiness, sea ice cover, vegetation and land use may all affect the GSAT to GMST difference, either directly or by altering the relationships between gradients and energy fluxes. These changing energy flux relationships are monitored through observing the stratification of the upper ocean (Section 9.2.1.3) and the response of upper ocean processes (Cross-Chapter Box 5.3) in ESMs and reanalyses, but such monitoring tasks rival the observational challenge of directly observing SSTs and 2 m air temperature under a wide range of conditions. In summary, because of the lack of physical constraints and the complexity of processes driving changes in the GSAT to GMST temperature differences, there is no simple explanation based on physical grounds alone for how this difference responds to climate change.

Direct observational evidence

There is currently no regularly updated, entirely observation-based dataset for GSAT. The best available observations of near-surface air temperature over ocean are datasets of night-time marine air temperature (NMAT) (e.g. Cornes et al., 2020; Junod & Christy, 2020), though spatial coverage is less extensive than for SST. Night-time measurements are used to avoid potential biases from daytime heating of ship superstructures. Kennedy et al. (2019) show little difference between HadNMAT2 and HadSST4 between 1920 and 1990, but a warming of SST relative to NMAT manifesting as a step change of 0.05°C–0.10°C in the early 1990s, which may reflect an actual change, the impact of increasingly divergent spatial coverage between SST and MAT measurements, or unresolved structural uncertainties in one or both datasets. This leads to NMAT warming around 10% more slowly than SST over the last century. In contrast, Junod and Christy (2020a) find NMAT trends which are 8–17% larger than those for SST in the ERSSTv4 and HadISST datasets for the period 1900 to 2010, but 11–15% smaller than the SST trends for the same datasets from 1979 to 2010. However, ERSSTv4 uses NMAT data as a basis for homogeneity adjustment so is not fully independent. Kent and Kennedy (2021) note sensitivity to methodological choices in comparisons but find that NMAT is warming more slowly than SST products over most periods considered. Rubino et al.
(2020) exploit tropical Pacific moored buoy arrays, available since the early 1980s, and find differences in NMAT and SST anomalies, which are sensitive to the choice of period and show spatio-temporal ENSO-related (Annex IV) signals in the differences.

Overall, with medium evidence and low agreement, available observational products suggest that NMAT is warming less than SST by up to 15%. Given that these ocean observations cover roughly two thirds of the globe, this implies that GMST is warming up to at most 10% faster than GSAT. Substantial uncertainty remains and the effect is highly sensitive to the choice of both time period and choice of NMAT and SST observational products to compare. Observed NMAT warming faster than observed SST cannot be precluded.

CMIP model-based evidence

CMIP historical simulations and projections agree that GSAT increases faster than GMST, the reverse of what is indicated by many marine observations. Several studies approximate the approach used to derive GMST from observations by blending SST over open ocean and SAT over land and sea ice from model output (Cowtan et al., 2015; Richardson et al., 2018; Beusch et al., 2020; Gillett et al., 2021). Cowtan et al. (2015) found that trends in GSAT are of the order of 9% larger than for GMST in CMIP5, based on data from 1850–2100 (historical + RCP8.5), if anomalies are blended and sea ice is allowed to vary over time. Broadly consistent numbers are found for both CMIP5 and CMIP6, across a range of SSP and RCP scenarios and time periods (Richardson et al., 2018; Beusch et al., 2020; Gillett et al., 2021). Blending monthly anomalies and allowing sea ice to vary, the change in GSAT for 2010–2019 relative to 1850–1900 is 2–8% larger than spatially-complete GMST in CMIP6 historical and SSP2.4.5 simulations (Gillett et al., 2021), and 6–12% larger in CMIP5 historical and RCP2.6 and 8.5 simulations for 2007–2016 relative to 1861–1880 (Richardson et al., 2018). However, a true like-for-like comparison to observational products is challenging because methodological choices have a large impact on the relationship between modelled GMST and GSAT and none of these studies fully reproduces the methods used to derive estimates of GMST in recent observational datasets, which use various ways to infill areas lacking in situ observations (Jones, 2020).

Marine boundary layer behaviour and parameterisations in all CMIP models are based upon Monin-Obukhov similarity theory (e.g. Businger et al., 1971), which informs assumptions around gradients in the near-surface boundary layer dependent upon temperature, wind speed and humidity. This leaves open the possibility of a common model bias, while Druzhinin et al. (2019) also point to departures of temperature profiles from theoretical predictions under certain conditions. There remain inadequacies in understanding and modelling of key processes (Edwards et al., 2020), and biases in the representation of the absolute SST-MAT difference have been identified in climate models and reanalyses (Găiniușă-Bogdan et al., 2015; Zhou et al., 2020).

Reanalysis-based evidence

Simmons et al. (2017) found that in JRA-55 and ERA-Interim (following an adjustment to account for an apparent discontinuity), GSAT increased 2–4% faster than GMST over the period 1979–2016. In atmospheric reanalyses, SST is given as a lower boundary condition from an observed globally interpolated product (such as HadISST; Rayner et al., 2003) whereas the air temperature is reliant upon model parameterisations and assimilated observations that do not include MAT observations (Simmons et al., 2017), thereby limiting their capability to constrain differences in GMST and GSAT trends. Furthermore, it is unclear what the lack of dynamic coupling at the ocean-atmosphere interface might imply for the representativeness of reanalysis-based estimates.

Representation of surface temperatures in sea ice regions

There is a significant issue in areas where sea-ice melts or grows, where the quantity used in observational-based GMST estimates switches between air temperature and sea surface temperature. This primarily affects analyses combining SAT anomalies over land and ice with SST anomalies over ocean. In areas where sea ice has recently melted, the climatological value changes from an air-temperature based estimate to an SST estimate based upon the freezing point of seawater (−1.8°C). This switch in climatology to, in general, a warmer climatology, leads to a bias towards reduced warming in anomalies compared with analyses based
on absolute temperatures. Richardson et al. (2018) found this underestimation to amount to approximately 3% of observed warming in historical model simulations. Given the projected future sea-ice losses, the effect will grow in future (low confidence), with potential effects of the order of 0.1°C in the second half of the 21st century under high warming scenarios, although with some uncertainty arising from the large spread of sea-ice loss in model projections (Tokarska et al., 2019).

Summary of lines of evidence

GMST and GSAT are physically distinct. There is high confidence that long-term changes in GMST and GSAT differ by at most 10% in either direction. However, conflicting lines of evidence from models and direct observations combined with limitations in theoretical understanding lead to low confidence in the sign of any difference in long-term trends. The very likely range of estimated historical GMST warming is combined with the assessed ±10% uncertainty in the relationship between GMST and GSAT changes to infer a GSAT equivalent, accounting for any possible real-world physical difference. Improvements in understanding may yield a robust basis to apply a scaling-factor to account for the difference in future assessments.

[START CROSS-CHAPTER BOX 2.3, TABLE 1 HERE]

Cross Chapter Box 2.3, Table 1: Summary of key observationally-based global warming estimates (in °C) to various periods in the present report and selected prior reports and their principal applications. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

<table>
<thead>
<tr>
<th>Period</th>
<th>AR6 GMST (°C)</th>
<th>AR6 GSAT* (°C)</th>
<th>AR5 and / or SR1.5 (only where reported) (°C)</th>
<th>Principal use of this period in reports</th>
</tr>
</thead>
<tbody>
<tr>
<td>1850–1900 to 2011–2020</td>
<td>1.09 (0.95–1.20)</td>
<td>1.09 (0.91–1.23)</td>
<td></td>
<td>Warming to present in AR6 WGI</td>
</tr>
<tr>
<td>1850–1900 to 2010–2019</td>
<td>1.06 (0.92–1.17)</td>
<td>1.06 (0.88–1.24)</td>
<td></td>
<td>Attributable warming in AR6 WGI</td>
</tr>
<tr>
<td>1850–1900 to 2006–2019</td>
<td>1.03 (0.89–1.14)</td>
<td>1.03 (0.86–1.18)</td>
<td></td>
<td>Energy budget constraints in AR6 on ECS and TCR</td>
</tr>
<tr>
<td>1850–1900 to 2006–2015</td>
<td>0.94 (0.79–1.04)</td>
<td>0.94 (0.76–1.08)</td>
<td>0.87 (0.75–0.99) – GMST 0.97 (0.85–1.09) – GSAT*</td>
<td>Warming to date in SR1.5. GSAT value used for remaining carbon budgets (RCBs) in SR1.5.</td>
</tr>
<tr>
<td>1850–1900 to 2003–2012</td>
<td>0.90 (0.74–1.00)</td>
<td>0.90 (0.72–1.03)</td>
<td>0.78 (0.72–0.85)</td>
<td>Warming to date in AR5 WGI.</td>
</tr>
<tr>
<td>1850–1900 to 1995–2014</td>
<td>0.85 (0.69–0.95)</td>
<td>0.85 (0.67–0.98)</td>
<td></td>
<td>Warming to recent past in AR6 WGI.</td>
</tr>
<tr>
<td>Year Range</td>
<td>OLS Trend</td>
<td>Global Warming Level (GWL) Crossing Time</td>
<td>Notes</td>
<td></td>
</tr>
<tr>
<td>------------</td>
<td>-----------</td>
<td>------------------------------------------</td>
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<td></td>
</tr>
<tr>
<td>1850–1900 to 1986–2005</td>
<td>0.69 (0.54–0.79)</td>
<td>0.69 (0.52–0.82)</td>
<td>0.61 (0.55–0.67)^</td>
<td>Warming to recent past in AR5 WGI.</td>
</tr>
<tr>
<td>1850–1900 to 1961–1990</td>
<td>0.36 (0.23–0.44)</td>
<td>0.36 (0.22–0.45)</td>
<td>0.85 (0.65–1.06)</td>
<td>Warming to reference period recommended by WMO for national-level data sets used for climate change assessment (included in the WGI Atlas).</td>
</tr>
<tr>
<td>1880–2012 OLS trend</td>
<td>0.92 (0.68–1.17)</td>
<td>0.85 (0.65–1.06)</td>
<td>0.85 (0.65–1.06)</td>
<td>Warming to date in AR5 WGI</td>
</tr>
</tbody>
</table>

* As the uncertainty in the relationship between GMST and GSAT changes is independent of the uncertainty in the assessed change in GMST, these uncertainties are combined in quadrature.

# SR1.5 derived a GSAT estimate by taking the CMIP5 ensemble mean GSAT change of 0.99°C, sub-sampling to HadCRUT4.6, noting the offset in trends (0.84°C HadCRUT4 observed GMST vs. 0.86°C modelled GMST) and adjusting by this to arrive at an estimate of 0.97°C change in GSAT. The likely uncertainty range of ±0.12°C was not further adjusted.

^ Note that the AR5 approach for the change from 1850–1900 to both 1986–2005 and 2003–2012 was based upon one dataset (HadCRUT4) and its parametric uncertainty estimates are known to underestimate the true uncertainty.

[END CROSS-CHAPTER BOX 2.3, TABLE 1 HERE]

Mapping between AR5 and AR6 assessments

The AR5 assessed estimate for historical warming between 1850–1900 and 1986–2005 is 0.61 [0.55 to 0.67] °C. The equivalent in AR6 is 0.69 [0.54 to 0.79] °C, and the 0.08 [-0.01 to 0.12] °C difference is an estimate of the contribution of changes in observational understanding alone (Cross-Chapter Box 2.3, Table 1). The exact value of this contribution depends upon the metric being compared (GMST/GSAT, trend or change, exact period). The AR6-assessed GMST warming between 1850–1900 and 2011–2020 is 1.09 [0.95 to 1.20] °C. An AR5-equivalent assessment using this estimated difference in observational understanding is thus 1.01 [0.94 to 1.08] °C. These updates and improvements in observational datasets affect other quantities that derive from the assessment of GSAT warming, including estimates of remaining carbon budgets and estimates of crossing times of 1.5°C and 2°C of global warming (See Cross Chapter Box 2.3 Table 1).

Updates to estimated Global Warming Level (GWL) crossing times

The updated estimate of historical warming is one contribution to the revised time of projected crossing of the threshold of 1.5°C global warming in comparison with SR1.5, but is not the only reason for this update. The AR6 assessment of future change in GSAT (Section 4.3.4, Table 4.5) results in the following threshold-crossing times, based on 20-year moving averages. The threshold-crossing time is defined as the midpoint of the first 20-year period during which the average GSAT exceeds the threshold. During the near term (2021–2040), a 1.5°C GSAT increase relative to the average over the period 1850–1900 is very likely to occur in scenario SSP5-8.5, likely to occur in scenarios SSP2-4.5 and SSP3-7.0, and more likely than not to occur in scenarios SSP1-1.9 and SSP1-2.6. In all scenarios assessed here except SSP5-8.5, the central estimate of crossing the 1.5°C global warming level lies in the early 2030s. This is about ten years earlier than the midpoint of the likely range (2030–2052) assessed in the SR1.5, which assumed continuation of the then-reported warming rate; this estimated rate has been confirmed in AR6 (Section 3.3.1). Roughly half of the ten-year difference arises from the higher diagnosed historical warming in AR6. The other half arises because, for central estimates of climate sensitivity, most scenarios show stronger warming over the near term than was assessed as ‘current’ in SR1.5 (medium confidence).

Implications for assessment of emission scenarios and remaining carbon budgets

To estimate the global warming implications of emissions scenarios, AR5 and SR1.5 combined estimates of observed GMST changes from 1850–1900 to 1986–2005 (Cross-Chapter Box 2.3, Table 1) with GSAT projections of subsequent warming. AR6 undertakes three changes to this approach. First, the AR6...
assessment of improved observational records is used. Second, the recent past baseline period is updated from 1986–2005 to 1995–2014, and, third, historical estimates are expressed in GSAT instead of GMST for consistency of historical estimates with future projections. The updated estimates of warming to date in AR6 lead to higher estimates of future warming, all else being equal. The temperature classification of emissions scenarios in the WGIII report adopts the definition of temperature classes as introduced in SR1.5, and assigns emission scenarios to these classes based on their AR6 assessed GSAT outcomes (Cross-Chapter Box 7.1, WGIII Annex C.II.2.4).

Both in AR5 and SR1.5, remaining carbon budgets were expressed as a function of GSAT warming, while also highlighting the implications of using historical warming estimates expressed in GMST. AR5 reported total carbon budgets for GSAT warming relative to 1861–1880. AR5 SYR also includes remaining carbon budget estimates based on AR5 WGIII scenario projections that use the method for AR5 scenario projections described above. SR1.5 integrated several methodological advancements to estimate remaining carbon budgets and reported budgets for additional GSAT warming since the 2006–2015 period, estimating, following the application of an adjustment (Richardson et al., 2016, Table 1.1. SR1.5) to GMST, that 0.97°C (± 0.12°C) of GSAT warming occurred historically between 1850–1900 and 2006–2015. The AR6 assessment, above, leads to an estimate of 0.94°C of warming between 1850–1900 and 2006–2015. All other factors considered equal, the AR6 estimate thus implies that 0.03°C more warming is considered for remaining carbon budgets compared to SR1.5. Combining this 0.03°C value with the SR1.5 transient climate response to cumulative emissions of CO₂ (TCRE) translates into remaining carbon budgets about 70 GtCO₂, [40 to 140 GtCO₂] larger compared to SR1.5 on a like-for-like basis. Meanwhile, on the same like-for-like basis, updates to historical observational products would reduce remaining carbon budgets reported in AR5 SYR based on WGIII scenario projections by about 180 GtCO₂, [120 to 370 GtCO₂]. Box 5.2 provides a further overview of updates to estimates of the remaining carbon budget since AR5.

Implications for assessment of impacts and adaptation

The assessment of global warming to date now being larger than previously assessed has no consequence on the assessment of past climate impacts, nor does it generally imply that projected climate impacts are now expected to occur earlier. The implications are mainly that the level of warming associated with a particular impact has been revised. This has very limited practical implications for the assessment of the benefits of limiting global warming to specific levels, as well as for the urgency of adaptation action. For example, impacts that occurred in the period 1986–2005 were previously associated with a GMST increase of 0.61°C relative to 1850–1900, relative to AR5 estimates. These impacts are now instead associated with a GMST increase of 0.69°C, relative to the assessment in this Report. The impacts themselves have not changed. Similarly, the impacts previously associated with a GMST or GSAT increase of 1.5°C will now generally be associated with a slightly different global warming level. This is because projections of future warming and its impacts relative to 1850–1900 are normally made by adding projected warming from a recent past baseline to an estimate of the observed warming from 1850–1900, as in AR5 and SR1.5.

Most of the previously projected impacts and risks associated with global warming of 1.5°C have therefore not changed and are still associated with the same level of future warming (0.89°C) relative to 1986–2005. With this warming now estimated as 0.08°C larger than in AR5, the future impacts previously associated with 1.5°C warming are now associated with 1.58°C warming. Similarly, the impacts now associated with 1.5°C warming would have previously been associated with 1.42°C warming. There are exceptions where impacts studies have used a baseline earlier than 1986–2005 (e.g. King et al., 2017), for which the new estimate of the historical warming would mean an earlier occurrence of the projected impacts. However, even in these cases, the ostensible difference in impacts associated with a 0.08°C difference in global mean temperature will be small in comparison with the uncertainties. There are also substantial uncertainties in regional climate changes and the magnitude of climate impact-drivers projected to occur with global warming of 1.5°C (Betts et al., 2018; Seneviratne et al., 2018). Furthermore, the time of reaching global warming of 1.5°C is subject to uncertainties of approximately ± 10 years associated with uncertainties in climate sensitivity, and ± 3 to 4 years associated with the different SSP forcing scenarios (Section 4.3.4, Table 4.5 see prior discussion). There is therefore high confidence that assessment of the magnitude and timing of impacts-related climate quantities at 1.5°C is not substantially affected by the revised estimate of...
historical global warming.

The assessment of the implications of limiting global warming to 1.5°C compared to 2°C will also remain broadly unchanged by the updated estimate of historical warming, as this depends on the relative impacts rather than the absolute impacts at any specific definition of global temperature anomaly (high confidence).

[START CROSS-CHAPTER BOX 2.3, FIGURE 1 HERE]

Cross-chapter Box 2.3, Figure 1: Changes in assessed historical surface temperature changes since AR5. (a) Summary of the impact of various steps from AR5 headline warming-to-date number for 1880–2012 using a linear trend fit to the AR6 assessment based upon the difference between 1850–1900 and 2011–2020. Whiskers provide 90% (very likely) ranges. AR6 assessment in addition denotes additional warming since the period around 1750 (Cross-Chapter Box 1.2). (b) Time series of the average of assessed AR5 series (orange, faint prior to 1880 when only HadCRUT4 was available) and AR6 assessed series (blue) and their differences (offset) including an illustration of the two trend fitting metrics used in AR5 and AR6. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END CROSS-CHAPTER BOX 2.3, FIGURE 1 HERE]

[END CROSS-CHAPTER BOX 2.3 HERE]

2.3.1.1.3 Temperatures during the instrumental period – surface

AR5 concluded that it was certain that GMST had increased since the late 19th century. Total warming in GMST was assessed as 0.85 [0.65 to 1.06] °C over 1880–2012, whilst the change from 1850–1900 to 2003–2012 was assessed at 0.78 [0.72 to 0.85] °C, and from 1850–1900 to 1986–2005 at 0.61 [0.55 to 0.67] °C. SR1.5 reported warming of GMST from 1850–1900 to 2006–2015 of 0.87°C, with an 1880–2012 trend of 0.86°C and an 1880–2015 trend of 0.92°C. SRCCL concluded that since the pre-industrial period, surface air temperature over land areas has risen nearly twice as much as the global mean surface temperature (high confidence).

Since AR5, there have been substantial improvements in the availability of instrumental archive data both over the ocean and on land. A new version of the International Comprehensive Ocean-Atmosphere Dataset (ICOADS Release 3.0, Freeman et al., 2017) comprises over 450 million in situ marine reports and incorporates newly digitised data, increasing coverage in data sparse regions and times (e.g. polar oceans and World War I). The International Surface Temperature Initiative released a much improved collection of fundamental land surface air temperature records (Rennie et al., 2014) comprising more than 35,000 station records. These advances, both of which have substantially improved spatial coverage, have reduced uncertainties in assessments of both land and marine data.

Marine domain

For SST analyses, three products – HadSST4 (1850–present, Kennedy et al., 2019), ERSSTv5 (1850–present, Huang et al., 2017) and COBE SST2 (1880–present, (Hirahara et al., 2014) – now have bias adjustments applied throughout the record. The new SST datasets account for two major issues previously identified in AR5: that globally averaged buoy SSTs are about 0.12°C cooler than ship-based SSTs (Huang et al., 2015; Kennedy et al., 2011), and that SSTs from ship engine room intakes may have biases for individual ships depending upon the sensor set-up (Kent & Kaplan, 2006) but have an overall warm bias when globally aggregated (Kennedy et al., 2019). The first issue primarily affects data since 1990, when buoys began to increasingly contribute to the observation network (Woodruff et al., 2011), and the second issue has its largest effect from the 1940s to the 1970s. From the standpoint of uncertainty, ERSSTv4 (Liu W. et al., 2015a; Huang et al., 2016) and subsequent versions (Huang et al., 2017), and HadSST4 have
estimates presented as ensembles that sample parametric uncertainty. Comparisons between these independently-derived analyses and the assessed uncertainties (Kennedy, 2014; Kent et al., 2017) show unambiguously that global mean SST increased since the start of the 20th century, a conclusion that is insensitive to the method used to treat gaps in data coverage (Kennedy, 2014).

A number of recent studies also corroborate important components of the SST record (Cowtan et al., 2018; Hausfather et al., 2017; Kennedy et al., 2019; Kent et al., 2017). In particular, ATSR SST satellite retrievals (Merchant et al., 2012; Berry et al., 2018), the near-surface records from hydrographical profiles (Gouretski et al., 2012; Huang et al., 2018), and coastal observations (Cowtan et al., 2018) have all been shown to be broadly consistent with the homogenized SST analyses. Hausfather et al. (2017) also confirmed the new estimate of the rate of warming seen in ERSSTv4 since the late 1990s through comparison with independent SST data sources such as Argo floats and satellite retrievals. Nevertheless, dataset differences remain in the mid-20th century when there were major, poorly-documented, changes in instrumentation and observational practices (Kent et al., 2017), particularly during World War II, when ship observations were limited and disproportionately originated from US naval sources (Thompson et al., 2008). Kennedy et al. (2019) also identify differences between the new HadSST4 dataset and other SST datasets in the 1980s and 1990s, indicating that some level of structural uncertainty remains during this period, whilst Chan et al. (2019) and Davis et al. (2019) document residual uncertainties in the early and later 20th century records respectively.

Historically, SST has been used as a basis for global temperature assessment on the premise that the less variable SST data provides a better estimate of marine temperature changes than marine air temperature (MAT) (Kent & Kennedy, 2021). However, MAT products are used to adjust SST biases in the NOAA SST product because they are assessed to be more homogeneous (Huang et al., 2017). Observational datasets exist for night-marine air temperature (NMAT) (e.g. (Cornes et al., 2020; Junod & Christy, 2019; Rayner et al., 2020) and there are methods to adjust daytime MATs (Berry et al., 2004), but there is to date no regularly updated dataset which combines MAT with temperatures over land. MAT datasets are more sparse in recent decades than SST datasets as marine datasets have become increasingly dependent on drifting buoys (Centurioni et al., 2019) which generally measure SST but not MAT, and there are almost no recent winter MAT data south of 40°S (Swart et al., 2019). However, the situation reverses in the 19th Century with a greater prevalence of MAT than SST measurements available in the ICOADS data repository (Freeman et al., 2017, 2019; Kent & Kennedy, 2021).

Land domain

The GHCN-M4v dataset (Menne et al., 2018) includes many more land stations than GHCN-Mv3, arising from the databank efforts of Rennie et al. (2014), and calculates a 100-member parametric uncertainty ensemble drawing upon the benchmarking analysis of Williams et al. (2012), as well as accounting for sampling effects. A new version of the CRUTEM dataset (CRUTEM5, Osborn et al., 2021) has increased data completeness and additional quality control measures. A new global land dataset, the China Land Surface Air Temperature (CLSAT) dataset (Xu et al., 2018) has higher network density in some regions (particularly Asia) than previously existing datasets. Global trends derived from CLSAT are generally consistent with those derived from other land datasets through 2014 (Xu et al., 2018).

AR5 identified diurnal temperature range (DTR) as a substantial knowledge gap. The most recent analysis of Thorne et al. (2016a,b) compared a broad range of gridded estimates of change in DTR, including a new estimate derived from the ISTI databank release using the pairwise homogenization algorithm used to create GHCMV4, and estimates derived from the Vose et al. (2005), HadEX2 (Donat et al., 2013a), HadGHCND (Donat et al., 2013a), GHCNDEX (Donat et al., 2013b), Berkeley Earth (Rohde et al., 2013), and CRU TS (Harris et al., 2014). The analysis highlighted substantial ambiguity in pre-1950 estimates arising from sparse data availability. After 1950 estimates agreed that DTR had decreased globally with most of that decrease occurring over the period 1960–1980. A subsequent DTR analysis using CLSAT further confirmed this behaviour (Sun et al., 2018).

No recent literature has emerged to alter the AR5 finding that it is unlikely that any uncorrected effects from urbanization (Box 10.3), or from changes in land use or land cover (Section 2.2.7), have raised global Land Surface Air Temperature (LSAT) trends by more than 10%, although larger signals have been identified in the 20th century when there were major, poorly-documented, changes in instrumentation and observational practices (Kent et al., 2017), particularly during World War II, when ship observations were limited and disproportionately originated from US naval sources (Thompson et al., 2008). Kennedy et al. (2019) also identify differences between the new HadSST4 dataset and other SST datasets in the 1980s and 1990s, indicating that some level of structural uncertainty remains during this period, whilst Chan et al. (2019) and Davis et al. (2019) document residual uncertainties in the early and later 20th century records respectively.
some specific regions, especially rapidly urbanizing areas such as eastern China (Li Y. et al., 2013; Liao et al., 2017; Shi et al., 2019). There is also no clear indication that site-specific data homogeneity issues have had any significant impact on global trends since the early 20th century; there is more uncertainty in the 19th century, mainly arising from a lack of standardization of instrument shelters, which has been largely accounted for in data from central Europe (Jones et al., 2012), but less so elsewhere.

Combined data products

At the time of AR5 a limitation of conventional datasets was the lack of coverage, especially in high latitudes, which although recognised as an issue (Simmons et al., 2010a) had not been addressed in most products. Interpolation involves the statistical imputation of values across regions with limited data and can add both systematic and random uncertainties (Lenssen et al., 2019). Cowtan and Way (2014) applied a kriging-based method to extend existing datasets to polar regions, while Kadow et al. (2020a) used an artificial intelligence-based method, and Vaccaro et al. (2021) used gaussian random Markov fields, for the same purpose, although only Kadow et al. (2020) uses the most recent generation of datasets as its base. The Berkeley Earth merged product (Rohde & Hausfather, 2020), HadCRUT5 (Morice et al., 2021) and NOAA GlobalTemp - Interim (Vose et al., 2021) all include interpolation over reasonable distances across data sparse regions which results in quasi-global estimates from the late 1950s when continuous Antarctic observations commenced. Interpolated datasets with substantial coverage of high latitudes show generally stronger warming of GMST than those with limited data in polar regions (Vose et al., 2021), and their strong warming at high northern latitudes is consistent with independent estimates from reanalyses (Simmons et al., 2017; Lenssen et al., 2019) and satellites (Cowtan and Way, 2014). Given the spatial scales of surface temperature variations and the verification of the methods, it is extremely likely that interpolation results in a less-biased estimate of the actual global temperature change than ignoring regions with limited or no data.

In total there are five conventional datasets which meet spatial coverage requirements and draw from the most recent generation of SST analyses, four of which have sufficient data in the 1850–1900 period to allow an assessment of changes from that baseline (Table 2.3). A fifth dataset is added to the assessment for changes over land areas. Datasets share SST and LSAT data products and in several cases differ solely in the post-processing interpolation applied meaning that there are far fewer methodological degrees of freedom than implied by a straight count of the number of available estimates.

[START TABLE 2.3 HERE]

Table 2.3: Principal characteristics of GMST in-situ data products considered in AR6 WGI highlighting interdependencies in underlying land and SST products and whether inclusion criteria are met.

<table>
<thead>
<tr>
<th>Dataset</th>
<th>Period of record</th>
<th>Land component</th>
<th>SST component</th>
<th>Ensemble uncertainties?</th>
<th>Meets all inclusion criteria?</th>
<th>Principal reference</th>
</tr>
</thead>
<tbody>
<tr>
<td>HadCRUT5</td>
<td>1850-2020</td>
<td>CRUTEM5</td>
<td>HadSST4</td>
<td>Yes</td>
<td>Yes</td>
<td>Morice et al. (2021)</td>
</tr>
<tr>
<td>NOAA GlobalTemp - Interim</td>
<td>1850-2020</td>
<td>GHCNv4</td>
<td>ERSSTv5</td>
<td>Yes, on earlier version</td>
<td>Yes</td>
<td>Vose et al. (2021)</td>
</tr>
<tr>
<td>Berkeley Earth</td>
<td>1850-2020</td>
<td>Berkeley</td>
<td>HadSST4</td>
<td>No</td>
<td>Yes</td>
<td>Rohde and Hausfather (2020)</td>
</tr>
<tr>
<td>Kadow et al</td>
<td>1850-2020</td>
<td>CRUTEM5</td>
<td>HadSST4</td>
<td>No</td>
<td>Yes</td>
<td>Kadow et al (2020a)</td>
</tr>
<tr>
<td>China - MST</td>
<td>1856-2020</td>
<td>CLSAT</td>
<td>ERSSTv5</td>
<td>No</td>
<td>Land only</td>
<td>Sun et al. (2021)</td>
</tr>
</tbody>
</table>
Estimates of GMST have also benefitted from improved estimation of parametric uncertainties. New versions of three long-standing products from NASA GISTEMP v4 (Lenssen et al., 2019), NOAA GlobalTempv5 (Huang et al., 2019) and HadCRUT5 (Morce et al., 2021) are all now available as ensemble estimates. These ensembles each account for a variety of systematic and random uncertainty effects in slightly different ways, giving broadly similar results, which are incorporated into the present assessment, with the total uncertainty generally declining up until the mid-20th century as data coverage improves.

Another significant development has been the incorporation of reanalysis products (Section 1.5.2) into operational monitoring of GSAT. It was reported in AR5 that various reanalyses were broadly consistent with conventional surface datasets in the representation of trends since the mid-20th century. Since that time, Simmons et al. (2017) found that the ERA-Interim (Dee et al., 2011) and JRA-55 (Kobayashi et al., 2015) reanalyses continued to be consistent, over the last 20 years, with those surface datasets which fully represented the polar regions. GSAT trends from ERA5 reanalysis (Hersbach et al., 2020) are also broadly consistent with GMST trends from conventional surface datasets. However, the MERRA-2 reanalysis (Gelaro et al., 2017) GSAT spuriously cooled sharply relative to ERA-Interim and JRA-55 in about 2007 (Funk et al., 2019). Since the early 2000s, analyses of surface temperature, from which near-surface temperature may be derived, have also been available from various satellites (Famiglietti et al., 2018; Prakash et al., 2018; Susskind et al., 2019), which have the potential to improve assessments of temperature changes over data-sparse regions.

Most land areas in the extratropical NH have warmed faster than the GMST average over both the 1900–2020 and 1980–2020 periods (Figure 2.11b), although at more regional scales, particularly in data sparse regions, considerable uncertainty is introduced by sometimes large differences in trends between different LSAT datasets (Rao et al., 2018). Temperatures averaged over land areas globally have warmed by 1.59 [1.34 to 1.83] °C from 1850–1900 to 2011–2020, substantially higher than the SST warming of 0.88 [0.68 to 1.01] °C. The four conventional surface temperature products which meet all criteria to be included in the final assessment (Table 2.4) agree that each of the last four decades has consecutively been the warmest globally since the beginning of their respective records (Figure 2.11c, Table 2.4). Each of the six years 2015 to 2020 has very likely been at least 0.9 °C warmer than the 1850–1900 average.

### Table 2.4: Observed increase (°C) in GMST and underlying LSAT and SST estimates in various datasets. Numbers in square brackets indicate 5–95% confidence ranges. Trend values are calculated with ordinary least squares following Santer et al. (2008a) and expressed as a total change over the stated period. Datasets considered in this table are those with data for at least 90% of global gridpoints in each year from 1960 onwards. GMST and SST are shown only for data sets which use air temperature (as opposed to climatological SST values) over sea ice. Changes from an 1850–1900 baseline are calculated only for those datasets which have data in at least 80% of years over 1850–1900. GMST values for each year are calculated as the mean of hemispheric means for the NH and SH, while LSAT and SST values are calculated from hemispheric means weighted according to the proportion of land (ocean) in the two hemispheres. This may vary from the methods used by individual data set providers in their own reporting. Products which meet all criteria to be included in the final assessment and contribute to the average are shown in italics. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

<table>
<thead>
<tr>
<th>Method</th>
<th>Year Period</th>
<th>CRUTEM4</th>
<th>HadSST3</th>
<th>Include</th>
<th>Output</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cowtan and Way</td>
<td>1850-2020</td>
<td>Yes</td>
<td>No</td>
<td>Cowtan and Way (2014)</td>
<td></td>
</tr>
<tr>
<td>Vaccaro et al</td>
<td>1850-2020</td>
<td>No</td>
<td>No</td>
<td>Vaccaro et al (2021)</td>
<td></td>
</tr>
<tr>
<td>Diagnostic/ Dataset</td>
<td>1850–1900 to 1995–2014 (°C)</td>
<td>1850–1900 to 2011–2020 (°C)</td>
<td>Trend 1880–2020 (°C)</td>
<td>Trend 1960–2020 (°C)</td>
<td>Trend 1980–2020 (°C)</td>
</tr>
<tr>
<td>----------------------------</td>
<td>-----------------------------</td>
<td>-----------------------------</td>
<td>----------------------</td>
<td>----------------------</td>
<td>----------------------</td>
</tr>
<tr>
<td>HadCRUT5</td>
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<td></td>
<td>1.10</td>
<td>1.04</td>
<td>0.76</td>
</tr>
<tr>
<td>GMST</td>
<td>0.87</td>
<td>[0.81–0.94]</td>
<td>[1.06–1.18]</td>
<td>[0.89–1.32]</td>
<td>[0.93–1.14]</td>
</tr>
<tr>
<td>LSAT</td>
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<td>[1.06–1.38]</td>
<td>1.43</td>
<td>1.50</td>
<td>1.20</td>
</tr>
<tr>
<td>SST</td>
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<td>[0.69–0.78]</td>
<td>1.03</td>
<td>0.90</td>
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<td></td>
<td>1.01</td>
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</tr>
<tr>
<td>GMST</td>
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<td>1.02</td>
<td>[0.80–1.32]</td>
<td>[0.90–1.11]</td>
<td>[0.63–0.87]</td>
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<tr>
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<td>1.69</td>
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<td>1.54</td>
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<tr>
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<td>0.75</td>
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<tr>
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<td>[0.80–1.34]</td>
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<td>0.72</td>
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</tr>
<tr>
<td>LSAT</td>
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<td>[1.19–1.78]</td>
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<td>SST</td>
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<td>[0.65–1.17]</td>
<td>1.40–1.72</td>
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</tr>
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<tr>
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<td></td>
</tr>
<tr>
<td>SST</td>
<td>0.73</td>
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<td>China-MST</td>
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<td>1.48</td>
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<td></td>
<td></td>
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<td>1.66</td>
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<tr>
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</tr>
<tr>
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<tr>
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<td>0.94</td>
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<td>GMST</td>
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<td>0.99</td>
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<td>SST</td>
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<td>GSAT</td>
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<td>[0.64–0.92]</td>
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<tr>
<td>LSAT</td>
<td>1.21</td>
<td>[1.02–1.40]</td>
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</tr>
<tr>
<td>Average – GMST</td>
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<td>1.09</td>
<td>1.11</td>
<td>1.04</td>
<td>0.76</td>
</tr>
<tr>
<td>Average – LSAT</td>
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<td>1.59</td>
<td>1.50</td>
<td>1.51</td>
<td>1.18</td>
</tr>
<tr>
<td>Average – SST</td>
<td>0.67</td>
<td>0.88</td>
<td>0.96</td>
<td>0.86</td>
<td>0.60</td>
</tr>
</tbody>
</table>

[END TABLE 2.4 HERE]
To conclude, from 1850–1900 to 1995–2014, GMST increased by 0.85 [0.69 to 0.95] °C, and to the most recent decade (2011–2020) by 1.09 [0.95 to 1.20] °C. Each of the last four decades has in turn been warmer than any decade that preceded it since 1850. Temperatures have increased faster over land than over the oceans since 1850–1900, with warming to 2011–2020 of 1.59 [1.34 to 1.83] °C versus 0.88 [0.68 to 1.01] °C, respectively.

2.3.1.2 Temperatures during the instrumental period – free atmosphere

AR5 reported that it was virtually certain that tropospheric temperatures have risen, and stratospheric temperatures fallen, since the mid-20th century, but that assessments of the rate of change and its vertical structure had only medium confidence in the NH extratropics and low confidence elsewhere. In particular there was low confidence in the vertical structure of temperature trends in the upper tropical troposphere.

2.3.1.2.1 Dataset developments

There have been updated radiosonde estimates from the University of Vienna (RAOB2CORE and RICH; Haimberger et al., 2012) and a new dataset from the State University of New York (SUNY, Zhou et al., 2020). There are new versions of AMSU products from the University of Alabama in Huntsville (UAHv6.0; Spencer et al., 2017) and Remote Sensing Systems (RSSv4.0; Mears and Wentz, 2017). These updates have led to convergence in the lower stratosphere layer (Maycock et al., 2018), in particular, the move to UAHv6.0 has addressed homogeneity issues identified by Seidel et al. (2016), although residual differences remain (Christy et al., 2018). Reanalyses products had identified limitations near the 300 hPa level where the contribution of aircraft observations has increased rapidly in recent years (Dee et al., 2011; Gelaro et al., 2017), leading to identified biases (Dee and Uppala, 2009), that have been addressed in ERA5 (Hersbach et al., 2020). Modern reanalyses are generally well aligned with radiosonde and satellite observations in the middle and lower stratosphere and lower stratosphere. A new operational mid- and upper-stratospheric dataset (STAR) has been developed by Zou & Qian (2016), merging the previous 1979–2006 SSU dataset (Zou et al., 2014) with a dataset from 1998 onwards drawn from relevant AMSU channels (Wang & Zou, 2014). Further stratospheric satellite-based datasets from various combinations of satellites have been developed by McLandress et al. (2015) and Randel et al. (2016).

New assessments of free-atmosphere temperature are available through radio occultation (RO) and Atmospheric Infrared Sounder (AIRS) products which begin in the early 2000s (Section 1.5.1.1). Global Navigation Satellite System (GNSS)-RO datasets have been compared against AMSU data records, finding almost identical trends (Khaykin et al., 2017). Comparison of RO with collocated radiosondes, Vaisala RS90/92 and GCOS Reference Upper Air Network data (RS92-HAP; Dirksen et al., 2014), show very good correspondence with global annual mean differences of less than 0.2°C in the upper troposphere and lower stratosphere. Radiosonde daytime radiation biases were identified at higher altitudes (Ho et al., 2017; Ladstädter et al., 2015). The stability of RO makes this data a useful comparator for AMSU (Chen & Zou, 2014) and radiosondes (Ho et al., 2017; Tradowsky et al., 2017), as well as anchoring post-2006 reanalyses datasets and improving their consistency in the lower and middle stratosphere (Ho et al., 2020; Long et al., 2017). The effective vertical resolution of RO measurements in the upper troposphere and lower stratosphere was found to be up to 100 m at the tropical tropopause (Zeng et al., 2019a), which is favourable for resolving atmospheric variability (Scherllin-Pirscher et al., 2012; Stocker et al., 2019; Wilhelmsen et al., 2018).

Temperature trends in RO products are most consistent with each other and with other observations between 8 km and 25 km (Ho et al., 2012; Steiner et al., 2013, 2020a). The uncertainty increases above 25 km for the early RO period, for which data are based on the single-satellite CHAMP mission, but data at higher altitudes become more reliable for later missions based on advanced receivers (Steiner et al., 2020a), along with the application of corrections for ionospheric effects (Danzer et al., 2020). The uncertainty due to the changing number of observations is reduced by correcting for the sampling uncertainty in RO climatological fields (e.g., Scherllin-Pirscher et al., 2011). For AIRS, thus far, stability of the instrument has been constrained to less than 0.03 °C per decade for selected window channels in a comparison to SSTs measured by ocean buoys (Aumann et al., 2019). Trends were inter-compared with trends in RO data and reanalysis data to assess systematic uncertainties (Leroy et al., 2018).
2.3.1.2.2 Assessment of trends

Warming has continued in the lower troposphere according to all radiosonde, reanalyses and satellite datasets, with a rate over 1980–2019 similar to surface warming rates (Table 2.5; c.f. Table 2.4). Radiosonde-based products generally show greater warming rates for 1980–2019 than satellite-based products and reanalyses. They also extend further back to the 1950s and trends since quasi-global coverage around 1960 also show warming (Table 2.5). Trends in RO and AIRS data, supported by radiosonde datasets, exhibit a warming trend in most of the mid- to upper-troposphere at all non-polar latitudes over 2002–2019. These also exhibit faster warming rates in the tropics in the upper troposphere than those observed at or near the surface (Figure 2.12); with the lowermost stratosphere also warming while above it is cooling. There is some spread between different data types in the tropics near the 15 km level, although these differences are reduced to near zero if a subset of radiosonde data, using only high-quality instruments, is used (Steiner et al., 2020b). AMSU tropical middle troposphere data also show that warming rates are near or above those in the lower troposphere, but they are measuring much broader layers which greatly complicates interpretation (Steiner et al., 2020b).

Temperatures averaged through the full lower stratosphere (roughly 10–25 km) have decreased over 1980–2019 in all data products, with the bulk of the decrease prior to 2000. The decrease holds even if the influence of the El Chichon (1982) and Pinatubo (1991) volcanic eruptions on the trend, found by Steiner et al. (2020b) to have increased the 1979–2018 cooling trend by 0.06°C per decade, is removed. Most datasets show no significant or only marginally significant trends over 2000–2019, and the results of Philipona et al. (2018) show weak increases over 2000–2015 in the very lowermost stratosphere sampled by radiosondes.

The STAR dataset shows cooling in the middle and upper stratosphere with a trend of −0.56°C ± 0.16°C per decade for the mid-stratosphere and −0.62°C ± 0.29°C per decade for the upper stratosphere over 1980–2019, although both cooling rates have slowed substantially since the mid-1990s. The overall post-1980 trend is reduced in magnitude by about 0.10°C per decade at both levels if the influences of the El Chichon and Pinatubo eruptions, and the solar cycle, are removed (Zou and Qian, 2016). The results obtained by McLandress et al. (2015) for 1980–2012, Randel et al. (2016) for 1979–2015, and Maycock et al. (2018) for 1979–2016 are broadly consistent with this.

A rise in the tropopause height of 40 to 120 m per decade between 1981 and 2015 was determined from both radiosonde and reanalysis datasets (Xian and Homeyer, 2018). Local studies (e.g. Tang et al., 2017; Chen X. et al., 2019) found stronger trends in some regions near the subtropical jet linked to tropical expansion (Section 2.3.1.4.1). Whilst Seidel and Randel (2006) found that the tropopause height was more closely coupled with temperatures in the stratosphere than those in the troposphere, it is not yet clear whether the rate of increase in tropopause height has experienced a similar recent slowdown to that of the cooling of the lower stratosphere, as short-period trends are typically inconclusive due to significant natural variability (Scherllin-Pirscher et al., 2020). RO data (Gao P. et al., 2015) indicate little change in tropopause height over the short period from 2006 to 2014, but a warming below the tropopause is observed over 2002 to 2019 (Figure 2.12).

In summary, the troposphere has warmed since the mid-20th century. There is medium confidence that temperatures in the tropical upper troposphere have warmed faster than those at the surface since 2001, but low confidence in changes prior to 2001. It is virtually certain that the lower stratosphere has cooled since the mid-20th century. However, most datasets show that lower stratospheric temperatures have stabilized since the mid-1990s with no significant change over the last 20 years. It is likely that middle and upper stratospheric temperatures have decreased since 1980, but there is low confidence in the magnitude. It is virtually certain that the tropopause height has risen over 1980–2019 but there is low confidence in the magnitude of this rise, or whether the rate of change has reduced commensurate with stabilized lower stratospheric temperatures.
Table 2.5: Observed change (°C) in free atmospheric temperatures in various datasets, for the lower tropospheric and lower stratospheric layers. Numbers in square brackets indicate 5–95% confidence ranges. Trend values are calculated with ordinary least squares following (Santer et al., 2008b) and are expressed as a total change over the stated period. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

<table>
<thead>
<tr>
<th></th>
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</thead>
<tbody>
<tr>
<td><strong>Lower troposphere</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>RAOBCORE</td>
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<td>0.74 [0.57–0.91]</td>
<td>0.52 [0.26–0.78]</td>
</tr>
<tr>
<td>RICH</td>
<td>1.20 [1.06–1.34]</td>
<td>0.79 [0.63–0.96]</td>
<td>0.53 [0.28–0.77]</td>
</tr>
<tr>
<td>SUNY</td>
<td>0.97 [0.80–1.13]</td>
<td>0.91 [0.76–1.05]</td>
<td>0.53 [0.35–0.72]</td>
</tr>
<tr>
<td>UAH</td>
<td></td>
<td>0.51 [0.37–0.65]</td>
<td>0.29 [0.07–0.50]</td>
</tr>
<tr>
<td>RSS</td>
<td></td>
<td>0.79 [0.66–0.92]</td>
<td>0.41 [0.24–0.58]</td>
</tr>
<tr>
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<td></td>
<td>0.68 [0.53–0.84]</td>
<td>0.54 [0.32–0.75]</td>
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<tr>
<td><strong>Average</strong></td>
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<td>0.70</td>
<td>0.46</td>
</tr>
<tr>
<td><strong>Lower stratosphere</strong></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>RAOBCORE</td>
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<td>–1.00 [–1.56–0.45]</td>
<td>–0.05 [–0.20–0.09]</td>
</tr>
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<td>RICH</td>
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<td>–1.19 [–1.95–0.42]</td>
<td>0.02 [–0.20–0.23]</td>
</tr>
<tr>
<td>SUNY</td>
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<td>–0.79 [–1.16–0.43]</td>
<td>–0.11 [–0.25–0.03]</td>
</tr>
<tr>
<td>UAH</td>
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<td>–0.24 [–0.37–0.12]</td>
</tr>
<tr>
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<tr>
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<td>–0.17 [–0.29–0.04]</td>
</tr>
<tr>
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<td>0.33 [0.07–0.60]</td>
</tr>
<tr>
<td><strong>Average</strong></td>
<td>–1.41</td>
<td>–1.07</td>
<td>–0.04</td>
</tr>
</tbody>
</table>

Figure 2.12: Temperature trends in the upper air. (a) Zonal cross-section of temperature anomaly trends (2007–2016 baseline) for 2002–2019 in the upper troposphere and lower stratosphere region. The climatological tropopause altitude is marked as a grey line. Significance is not indicated due to the short period over which trends are shown, and because the assessment findings associated to this figure relate to difference
2.3.1.3 Global hydrological cycle

This section focuses on large-scale changes in a subset of components of the hydrological cycle (Cross-
Chapter Box 2.2). Chapter 8 undertakes a holistic assessment of changes in the hydrological cycle
integrating observations, modelling and theoretical understanding, while Chapter 11 assesses hydrological
cycle extremes such as droughts and floods.

2.3.1.3.1 Paleo perspective of the global hydrological cycle

AR5 assessed large-scale indicators of terrestrial paleo hydroclimate, including as part of its assessment of
paleo floods and droughts, but did not assess proxy evidence for paleo hydroclimate indicators over
continental and larger scales. The paleoclimate evidence assessed in AR5 was broadly consistent with global
hydroclimate scaling with temperature: warmer periods were wetter (e.g., the Pliocene; increased
precipitation) with colder periods being drier (e.g., the LGM; decreased precipitation).

Substantial limitations exist in reconstructing the global hydrological cycle prior to the Quaternary,
particularly during the Eocene, due to the lack of high-resolution proxy records and their sparsity. Spatial
heterogeneity complicates identification of wetting and drying signals during the PETM and the EECCO, with
paleo data and model simulations suggesting an intensified global hydrological cycle (Carmichael et al.,
2016, 2017; Hyland et al., 2017; West et al., 2020), in particular an increased specific humidity (Winnick et
al., 2015; van Dijk et al., 2020). Conditions wetter than present were inferred for the MPWP (Cross Chapter
Box 2.4), with intensified Asian monsoons (An et al., 2015) and with nevertheless drier conditions over
tropical and subtropical SH locations (Pontes et al., 2020). A new global reconstruction of hydroclimate
proxies for the LIG points to stronger boreal precipitation compared to 1850–1900 over high latitudes and
especially over monsoon areas, with a more heterogeneous signal for the SH (Scussolini et al., 2019). This
heterogeneity is also present in the tropics, characterized by large zonal differences in precipitation change
due to the variations in the intensity of Walker circulation (section 2.3.1.4.1). Available records indicate
reduced global vegetation cover and abundant atmospheric dust deposition during the LGM (increased
aridity), particularly over the tropics and high latitudes (Lamy et al., 2014; Újvári et al., 2017). This agrees
with models and moisture-sensitive proxies, suggesting an overall decrease in global precipitation during the
LGM relative to recent decades, albeit with regional-scale heterogeneity (Cao et al., 2019). Despite lower
global precipitation amounts, research since AR5 has identified a wetting of mid-latitudes during the LGM
(Putnam and Broecker, 2017; Lowry and Morrill, 2018; Morrill et al., 2018), thereby complicating the
characterization of the LGM as a relatively “dry” period. Low evaporation rates and increased top-soil
moisture during the LGM may have contributed to elevated levels of large closed-basin lakes located in the
30°–45° latitudinal belts (Putnam and Broecker, 2017; Scheff et al., 2017), such as the southwest US (e.g.
Ibarra et al., 2018), southern Australia (Fitzsimmons et al., 2015; Petherick et al., 2013; Sniderman et al.,
2019) and Patagonia (e.g. Quade & Kaplan, 2017).

New analyses suggest that during the Holocene, the NH mid-latitudes became increasingly wet, in phase
with the strength of the latitudinal temperature and insolation gradients (Shuman and Marsicek, 2016;
Routson et al., 2019). Nevertheless, there was also considerable spatial heterogeneity and variability on
centennial to millennial timescales (Liefert & Shuman, 2020; Newby et al., 2014; Shuman & Marsicek,
2016; Zhang H. et al., 2018). The NH tropics and many regions of the SH deep tropics experienced wetting
up until the early to mid-Holocene but drying thereafter (Muñoz et al., 2017; Nash et al., 2016; Quade et al.,
2018; Shanahan et al., 2015). Evidence for the SH is limited, with a wetting trend during the Holocene in low
latitudes of South America (Kanner et al., 2013; Mollier-Vogel et al., 2013) and parts of the African tropics
(Scheufler et al., 2011; Chevalier and Chase, 2015) but a drying tendency over southern Australia and New
Zealand (Barr et al., 2019; van den Bos et al., 2018) and South America (Moreno et al., 2018; Quade & Kaplan, 2017).

For the CE, new proxy records have led to the creation of continental drought atlases (Cook et al., 2015; Morales et al., 2020; Palmer et al., 2015; Stahle et al., 2016) and millennium reanalyses (Steiger et al., 2018; Tardif et al., 2019). These reconstructions highlighted the occurrence of multi-decadal regional mega-droughts in the NH before 1600 CE, particularly during 800–1200 CE, with a predominance of wet periods after 1700 CE (Cook et al., 2015; Rodysill et al., 2018; Shuman et al., 2018). In the SH, much of South America and the African tropics experienced a reduction of precipitation during 900–1200 CE and a wetting peak during 1500–1800 CE (Campos et al., 2019; Fletcher et al., 2018; Lüning et al., 2018; Nash et al., 2016; Tierney et al., 2015), with an opposite pattern in southern subtropical Africa (Lüning et al., 2018; Woodborne et al., 2015). Large multidecadal variability was documented over Australia and New Zealand during the 800–1300 CE period, followed by a well-defined wet period during 1500–1800 CE (Barr et al., 2014; Evans et al., 2019).

To summarise, since AR5 there has been considerable progress in detecting the variations of the global hydrological cycle prior to the instrumental period. There are indications from multiple sources of a wetting trend during the Holocene, particularly for the NH and parts of the SH tropics (medium confidence). Hydroclimate during the CE is dominated by regional variability, generally precluding definitive statements on changes at continental and larger scales, with a general reduction of mega-drought occurrences over the last about 500 years (medium confidence). Availability of proxy data for assessing Holocene hydroclimate variability is biased towards the NH, with medium evidence but low agreement for the assessment of SH changes.

### 2.3.1.3.2 Surface humidity

AR5 reported very likely widespread increases in near-surface air specific humidity since the 1970s, abating from around 2000 to 2012 (medium confidence). This abatement resulted in a recent decline in relative humidity over the land.

Near surface humidity has been monitored using in situ data (e.g. NOCSv2.0; Berry and Kent, 2011), satellite-derived estimations (e.g. HOAPS3, Liman et al., 2018; J-OFURO3, Tomita et al., 2019), global gridded products such as HadISDH (Willett et al., 2014; 2020), and reanalyses (e.g. ERA5, JRA-55 and 20CRv3). In situ based humidity products suffer from uncertainties over poorly sampled regions particularly in the SH (Berry & Kent, 2011; Kent et al., 2014; Willett et al., 2014). There is general consensus in the inter-annual variability and sign of trends implying high confidence in increasing specific humidity since the 1970s and decreasing relative humidity since 2000, particularly over land (Simmons et al., 2010; Willett et al., 2014; 2020). Since 2012, specific humidity over land and ocean has remained well above the 1973–2019 average and reached record or near-record values (Figure 2.13b), with the strong 2015–2016 El Niño event boosting surface moisture levels (Byrne and O’Gorman, 2018). The abatement from around 2000 to 2012 reported in AR5 has not persisted. This is consistent with increases in total column water vapour (Section 2.3.1.3.3) and a resumption of rapid warming in surface temperatures (Section 2.3.1.1.3). The global averaged relative humidity however has remained depressed since 2000 (Figure 2.13d) (Dunn et al., 2017; Simmons et al., 2010; Vicente-Serrano et al., 2018; Willett et al., 2014; 2020).

Since 1973, increases in specific humidity have been widespread and significant across the majority of the land and ocean regions where observations are available (Figure 2.13a). In contrast, trends in relative humidity show distinct spatial patterns with generally increasing trends over the higher latitudes and the tropics and generally decreasing trends over the sub-tropics and mid-latitudes, particularly over land areas (Figure 2.13c). Near-surface specific humidity over the oceans has increased since the 1970s according to several in situ, satellite and reanalysis data records (Kent et al., 2014; Robertson et al., 2020; Willett et al., 2020). According to the HadISDH product, increases in specific humidity and decreases in relative humidity are significant particularly over the NH mid-latitudes (Figure 2.13a,c). Poor data coverage over the SH south of 20°S does not allow for the robust assessment of trends. Sources of uncertainty include the initial measurement accuracy, homogenisation over land, observational height at ships and instrument bias.
adjustment over ocean, and sparse spatio-temporal sampling (Prytherch et al., 2015; Roberts et al., 2019; Willett et al., 2020).

In summary, observations since the 1970s show a very likely increase in near surface specific humidity over both land and oceans. A very likely decrease in relative humidity has occurred over much of the global land area since 2000, particularly over mid-latitude regions of the NH, with increases at northern high latitudes.

[START FIGURE 2.13 HERE]

Figure 2.13: Changes in surface humidity. (a) Trends in surface specific humidity over 1973–2019. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al (2008a) (‘x’ marks denote non-significant trends). (b) Global average surface specific humidity annual anomalies (1981–2010 base period). (c) as (a) but for the relative humidity. (d) as (b) but for the global average surface relative humidity annual anomalies. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.13 HERE]

2.3.1.3.3 Total column water vapour

AR5 concluded that total column water vapour (TCWV) very likely increased since the 1970s, at a rate that was overall consistent with the Clausius-Clapeyron relationship (about 7% per °C) given the observed increase in atmospheric temperature.

Records prior to the instigation of quasi-global coverage by radiosondes require the use of statistical relationships to infer TCWV from historical SST observations or the evaluation of centennial-scale reanalysis products (Smith and Arkin, 2015). These approaches reveal two periods of positive trends, one during 1910–1940 and the other from 1975 onwards (Zhang et al., 2013; Mieruch et al., 2014; Shi et al., 2018), concurrent with periods of positive SST trends (Figure 2.11). Potential sources of errors in the SST-based estimation of TCWV include both uncertainties in historical SST and uncertainties in the parameters that define the relationship between the variables (Smith & Arkin, 2015). Trends based on 20CRv2c, ERA-20C and ERA-20CM indicate an increase in TCWV over much of the global ocean since the beginning of the 20th century, particularly over the tropics (Bordi et al., 2015; Poli et al., 2016; Smith & Arkin, 2015). TCWV trends estimated since the middle of the 20th century from radiosonde observations show significant increases over North America and large portions of Eurasia, while decreases are restricted to Australia, Eastern Asia and the Mediterranean region (Zhang Y. et al., 2018). Overall, there is a significant increase in TCWV over global land areas since 1979 (Chen & Liu, 2016).

Since the late 1970s a range of satellite missions permit a quasi-global assessment of TCWV. Several satellite products provide water vapour retrievals based upon distinct spectral domains, in addition to products from radiosondes, reanalyses and GNSS radio occultation. The GEWEX Water Vapour Assessment (G-VAP) provided an intercomparison of several TCWV data records, with global coverage but limited timespan (Schröder et al., 2018). The various global products generally exhibit a positive trend since 1979 (Figure 2.14) (Allan et al., 2014; Mieruch et al., 2014; Schröder et al., 2014; Wang X.L. et al., 2016a), most evident over the tropics (Chen & Liu, 2016; Gu & Adler, 2013; Mears et al., 2018; Salamalikis et al., 2021; Wang R. & Liu, 2020). The existence of apparent breakpoints in several products, which are generally coincident with changes in the observing system, lead to trend estimates that are not in line with theoretical expectations imposed by the Clausius-Clapeyron relationship (Schröder et al., 2019a), although other factors such as regional moisture divergence/convergence could account for the observed TCWV-temperature scaling. Substantial potential inhomogeneities affect trend estimates based on satellite, reanalysis and merged products in particular over Central Africa, the Sahara and central South America (Schröder et al., 2016, 2019; Wang et al., X.L., 2016). Moreover, data gaps in observations from ground-based GNSS receivers and radiosondes lead to low confidence in TCWV estimation in these regions.
In summary, positive trends in global total column water vapour are very likely since 1979 when globally representative direct observations began, although uncertainties associated with changes in the observing system imply medium confidence in estimation of the trend magnitudes. Low confidence in longer-term trends arises from uncertainties in the SST-TCWV relationship and current centennial scale reanalyses, particularly during the first half of the 20th century.

2.3.1.3.4 Global precipitation

AR5 concluded that there was low confidence in precipitation change averaged over global land areas prior to 1950, and medium confidence thereafter with no significant global trends. There was a likely overall increase in precipitation in the well-sampled NH mid-latitudes, with high confidence after 1951.

In situ precipitation records over land extend back for centuries in a few locations, and to the early to mid-20th century quasi-globally. Datasets differ in their input data, completeness of records, period covered, and the gridding procedures which, given spatial clustering and the small spatial scales of precipitation, results in differences in global and regional estimates of precipitation changes (Nogueira, 2020; Sun Q. et al., 2018). The spatial variability of observed long-term trends (1901–2019) based on GPCC V2020 and CRU TS 4.04 (Figure 2.15a,b) indicates significant increases in precipitation mainly over eastern North America, northern Eurasia, southern South America and northwestern Australia. Decreases are strongest across tropical western and equatorial Africa and southern Asia. The temporal evolution of global annual land precipitation anomalies exhibits little consistency between GPCC V2020, CRU TS 4.04 and GHCNv4 datasets, especially prior to 1950, that is associated with limitations in data coverage (Figure 2.15c, (Gu & Adler, 2015; Shen et al., 2014; Wu et al., 2013). These disagreements between datasets prior to the 1950s result in differences in trend estimates over global land (Table 2.6). A qualitative consistency in decadal and interdecadal variations between the products is only observed since the 1950s, with primarily positive land precipitation anomalies during the 1950s, 1970s and after the 2000s (Figure 2.15c).

Several satellite-based precipitation datasets improve the representation of the spatio-temporal changes since the late 20th century. Some of these are based exclusively on satellite data (e.g. CMORPH, Joyce et al., 2004; GSMaP, Okamoto et al., 2005), with others being combinations of in situ observations, reanalyses and satellite retrievals (e.g. CMAP, Xie and Arkin, 1997; TRMM 3B43 V7, Huffman et al., 2007; PERSIANN-CDR, Ashouri et al., 2015; CHIRPS; Funk et al., 2015; GPCP V2.3, Adler et al., 2018). These can be affected by systematic and random uncertainties due to inhomogeneities in the satellite-derived precipitation and station data and the uncertainties of blending algorithms (Hegerl et al., 2015; Sun et al., 2018a). The spatial coverage of these products is near-global, with available estimations formally covering 60°S–60°N with decreasing quality from low to high latitudes, depending on the sensors and algorithms used (Hu Q. et al., 2019). A detailed description of the most relevant satellite products is provided in section 10.2.1.1.

Recent trends (1980–2019) for GPCC V2020, CRU TS 4.04 and GPCP V2.3 show significant increases in land precipitation over tropical Africa, the eastern portions of Europe and North America, central Asia and the Maritime Continent (Figure 2.14, d, e and f). Significant decreases are observed over central South America, western North America, northern Africa and the Middle East. A detailed assessment of the recent regional precipitation trends using the same datasets can be found in the Atlas. Global trends for 1980–2019 show a general increase in annual precipitation over land, which is particularly marked for CRU TS 4.04 and GHCNv4 (Table 2.6). These changes have been accompanied by a strengthening of precipitation seasonality over tropical land areas, although with broad spread between different satellite-based (GPCP, MSWEP_V1.2, PERSIANN-CDR) and in situ gridded datasets (GPCC, CRU TS) (Chou et al., 2013; Li X.
et al., 2016; Tan et al., 2020). Increasing trends since 1980, in contrast to longer-term declining trends since 1901, are particularly evident over much of Africa, while more widespread negative trends were observed over much of southern South America in the more recent period (Knutson and Zeng, 2018) (Atlas 7.2). A faster recent increase in precipitation over global land is inferred comparing the precipitation trends over 1960–2019 with 1980–2019 (Table 2.6). Over the global ocean, the comparison between precipitation datasets is compromised by the different measurement periods, as well as the spatial coverage of the available products (Adler et al., 2017; Jaber & Abu-Allaban, 2020; Nguyen et al., 2018; Nogueira, 2020), limiting the ability to assess the sign and magnitude of precipitation trends. The GPCPv2.3 database (Adler et al., 2017, 2018) exhibits an increase of 2.94 mm yr\(^{-1}\) per decade over 1980–2019, principally due to the trends over the Indian ocean and in the tropical western Pacific (Figure 2.15f). The regional patterns of recent trends are consistent with the documented increase in precipitation over tropical wet regions and the decrease over dry areas, estimated through GPCP v2.2 data (Kao et al., 2017; Liu & Allan, 2013; Polson & Hegerl, 2017; Trammell et al., 2015).

![START FIGURE 2.15 HERE]

**Figure 2.15:** Changes in observed precipitation. (a, b) Spatial variability of observed precipitation trends over land for 1901–2019 for two global in-situ products. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al (2008a) ('x' marks denote nonsignificant trends). (c) Annual time series and decadal means from 1891 to date relative to a 1981–2010 climatology (note that different products commence at distinct times). (d, e) as (a, b), but for the periods starting in 1980. (f) is for the same period for the globally complete merged GPCP v2.3 product. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

![END FIGURE 2.15 HERE]

![START TABLE 2.6 HERE]

**Table 2.6:** Globally averaged trend estimates over land and 90% confidence intervals for annual precipitation for each time series in Figure 2.15(c) over three periods all ending in 2019. Trends are calculated using OLS regression with significance assessed after Santer et al (2008a). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

<table>
<thead>
<tr>
<th>Dataset</th>
<th>Trends in annual precipitation (mm yr(^{-1}) per decade)</th>
</tr>
</thead>
<tbody>
<tr>
<td>GPCPv2020</td>
<td>1.01* ± 0.99</td>
</tr>
<tr>
<td>CRUTS 4.04</td>
<td>0.57 ± 2.08</td>
</tr>
<tr>
<td>GHCNv4</td>
<td>3.19* ± 1.48</td>
</tr>
<tr>
<td>GPCPv2.3</td>
<td>4.19* ± 2.08</td>
</tr>
</tbody>
</table>

* Trend values significant at the 10% level.

![END TABLE 2.6 HERE]

In summary, globally averaged land precipitation has **likely** increased since the middle of the 20th century (**medium confidence**), with **low confidence** in trends prior to 1950. A faster increase in global land precipitation was observed since the 1980s (**medium confidence**), with large interannual variability and regional heterogeneity. Over the global ocean there is **low confidence** in the estimates of precipitation trends, linked to uncertainties in satellite retrievals, merging procedures and limited in situ observations.

### 2.3.1.3.5 Precipitation minus Evaporation

AR5 concluded that the pattern of precipitation minus evaporation (P-E) over the ocean had been enhanced since the 1950s (**medium confidence**). Saline surface waters had become saltier, while the relatively fresh
surface waters had become fresher. The inferred changes in P-E were consistent with the observed increased TCWV, although uncertainties in the available products prevented identifying robust trends.

Estimating global-scale trends in P-E using direct observations alone is challenging due to limited evaporation measurements and inhomogeneities in satellite-derived precipitation and evaporation datasets (Hegerl et al., 2015; López et al., 2017). Hence, the assessment of global P-E trends is generally performed using reanalyses, although changes in the observing system imply considerable uncertainty (Skliris et al., 2014a). Since the second half of the 20th century, several reanalyses and observational datasets have shown increases in P-E over global land, although 75% of land areas exhibit no significant changes and both internal variability and observational uncertainty are substantial (Greve et al., 2014; Robertson et al., 2016).

The recently released ERA5 (Hersbach et al., 2020) showed improvements in the representation of tropical precipitation, although it overestimates global precipitation trends in comparison to ERA-Interim and GPCP (Nogueira, 2020), and suffers from temporal changes in the annual balance between precipitation and evaporation (Hersbach et al., 2020). The spatial pattern of P-E trends over 1980–2019 (Figure 2.16a) are largely consistent with the trends in the GPCP v2.3 precipitation dataset (Figure 2.15f; Section 2.3.1.3.4) and agrees in sign with the trends from other reanalyses such as JRA-55 and MERRA-2 (Yu L. et al., 2020).

A variety of reanalysis products exhibit diverse temporal evolutions of P-E (Figure 2.16b–d). Globally, MERRA-2, ERA20C and ERA20CM exhibit little change whereas JRA-55, ERA5 and 20CRv3 all imply long-term changes (Figure 2.16d). A potential limitation in estimating P-E from some reanalysis products is readily apparent when considering the temporal evolution of global P-E from CFSR and MERRA (Figure 2.16d) which both exhibit strong discontinuities over the global ocean in the late 1990s. Over global land as a whole, precipitation exceeds evaporation (P-E > 0) for all the reanalysis products (Figure 2.16c), with decreasing trends in P-E for ERA5 and JRA-55 and increasing trends for MERRA-2 and CFSR. The P-E over the global ocean is negative (evaporation exceeding precipitation) for most reanalyses (Figure 2.16d), with declining trends in ERA5 and MERRA-2 dominated by trends in evaporation (Hersbach et al., 2020; Craig et al., 2017) and in satellite data (Andersson et al., 2011; Robertson et al., 2014), although with considerable differences between available estimates (Chandanpurkar et al., 2017; Yu L. et al., 2020). An alternative indirect approach to estimate P-E changes is based on near-surface ocean salinity (Section 2.3.3.2), which is partially driven by the freshwater flux at the ocean surface. The near-surface salinity trends are more spatially coherent compared to those revealed by P-E estimates from reanalyses, with an intensification of the water cycle over oceans, especially in subtropical regions (Yu L. et al., 2020, Durack et al., 2012; Skliris et al., 2014). However, the precise rate of water cycle intensification implied by salinity trends is sensitive to methodological choices (e.g., Skliris et al., 2016; Zika et al., 2018).

[START FIGURE 2.16 HERE]

Figure 2.16: Changes in precipitation minus evaporation. (a) Trends in precipitation minus evaporation (P-E) between 1980 and 2019. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after (Santer et al., 2008; ‘x’ marks denote non-significant trends). Time series of (b) global, (c) land-only and (d) ocean-only average annual P-E (mm/day). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.16 HERE]

In conclusion, observational uncertainty yields low confidence in globally averaged trends in P-E over the 20th century, with a spatial pattern dominated by precipitation changes over land and by evaporation increases over the ocean. Different reanalyses disagree on the sign of long-term changes in the global mean P-E.

2.3.1.3.6 Streamflow AR5 concluded that there was low confidence in a positive trend in global river discharge during the 20th
century. It noted that many of the largest rivers with long term streamflow records have been impacted by non-climatic human influences such as dam construction or land-use change.

River discharge is monitored widely, although gaps remain at a subcontinental scale over central Asia and Africa (Wei et al., 2020). Substantial recent efforts have been made to generate new global streamflow datasets, consolidating observations from many stream gauges to create streamflow indices (Do et al., 2018; Gudmundsson et al., 2018) and gridded products using neural networks (Barbarossa et al., 2018) or combinations between observations and reanalyses (Suzuki et al., 2018; Ghiggi et al., 2019).

Human intervention on river discharge linked to increases in evapotranspiration and some reduction of intra-annual streamflow variability (Jaramillo and Destouni, 2015, Chai et al., 2020) might affect the detection of trends in extreme daily streamflow events (Do et al., 2017; Gudmundsson et al., 2019). However, these activities have a minor impact on annual streamflow compared to climate variations (Dai et al., 2009; Alkama et al., 2013). Available global studies post-1950 generally concur that there have been more rivers experiencing decreases than increases in runoff (Do et al., 2017; Gudmundsson et al., 2019; X. Shi et al., 2019; Su et al., 2018). Most of the rivers have not experienced statistically significant changes in streamflow, and when globally aggregated there is no significant change (Dai and Zhao, 2017). Global streamflow variability is strongly modulated by ENSO and PDV, with below-normal global streamflow, as a response to El Niño events and vice-versa during La Niña episodes (Dai, 2016; Kim, 2019; Liang et al., 2016). The response of streamflow to changes in precipitation associated with ENSO and PDV has heterogeneous regional patterns at subcontinental scales (Section 8.3.2.9.1). No significant trends are found for reanalysis-based discharge estimates over 1993 to 2015 (Chandanpurkar et al., 2017). Uncertainties in global streamflow trends arise predominantly from changes in instrumentation, gauge restoration, recalibration of rating curves, flow regulation or channel engineering (Alkama et al., 2011; Gudmundsson et al., 2018; Ghiggi et al., 2019).

In summary, the sign of global streamflow trends remains uncertain, with slightly more globally gauged rivers experiencing significantly decreasing flows than significantly increasing flows since the 1950s (low confidence).

### 2.3.1.4 Atmospheric circulation

This section focuses on large-scale changes in a subset of components of the atmospheric circulation (Chapter Box 2.2). Chapter 8 assesses large scale as well as regional aspects of circulation components and their impact on the hydrological cycle, while Chapter 11 assesses the association of circulation changes and variability with extreme events.

#### 2.3.1.4.1 The Hadley and Walker circulations

AR5 reported low confidence in trends in the strength of the Hadley circulation (HC) and the Walker circulation (WC) due to uncertainties in available reanalysis datasets and the large interannual-to-decadal variability of associated circulation patterns. However, AR5 indicated a likely widening of the tropical belt since the 1970s, albeit with large uncertainty in the magnitude of this change. There was high confidence that the post-1990s strengthening trends of the Pacific WC reversed its weakening observed from the mid-19th century to the 1990s.

Paleo reconstructions of rainfall and trade winds extending over the last 100 kyr show an intensification of the NH HC concurrently with a weakening of the SH HC and a southward shift of the inter tropical convergence zone (ITCZ) during Heinrich stadials (Deplazes et al., 2013; McGee et al., 2018; Strikis et al., 2018; Wendt et al., 2019). An intensification of the HC associated with conditions similar to La Niña (northward migrations of both the ITCZ and the SH westerlies) was found in reconstructions for the MH (McGee et al., 2014; Mollier-Vogel et al., 2019). Changes in insolation from the mid to late Holocene favoured a southward migration in the position of the ITCZ and the descending branch of the HC in the NH, approaching its current width and position (Wirth et al., 2013a; Thatcher et al., 2020). Tree ring chronologies...
from the NH mid-latitudes over the last 800 years show that the northern edge of the HC tended to migrate southward during positive phases of ENSO and PDV, with northward shifts during negative phases (Alfaro-Sánchez et al., 2018). Between 1400 and 1850 CE the HC over both hemispheres and the ITCZ were displaced southward, consistent with occurrence of drought conditions in several NH regions (Alfaro-Sánchez et al., 2018; Burn and Palmer, 2014; Flores-Aqueveque et al., 2020; Lechleitner et al., 2017; Wirth et al., 2013). Moreover, several proxy records showed not only inter-hemispheric shifts in the ITCZ but a contraction of the tropical belt during 1400–1850 CE, which followed an expansion during 950–1250 CE (Denniston et al., 2016; Griffiths et al., 2016).

From centennial-scale reanalyses, Liu et al. (2012) and D’Agostino and Lionello (2017) found divergent results on HC extent over the last 150 years, although with unanimity upon an intensification of the SH HC. A substantial discrepancy between HC characteristics in centennial-scale reanalyses and in ERA-Interim (D’Agostino and Lionello, 2017) since 1979 yields significant questions regarding their ability to capture changes in HC behaviour. Taken together with the existence of apparent non-climatic artefacts in the dataset (Nguyen et al., 2015), this implies low confidence in changes in the extent and intensity of HC derived from centennial-scale reanalyses. However, using multiple observational datasets and centennial-scale reanalyses, Bronnmann et al. (2015) identified a southward shift in the NH HC edge from 1945 to 1980 of about 0.25° latitude per decade, consistent with observed changes in global land monsoon precipitation (Section 2.3.1.4.2).

Since AR5 several studies based upon a range of metrics and different reanalyses products have suggested that the annual mean HC extent has shifted poleward at an approximate rate of 0.1°–0.5° latitude per decade over the last about 40 years (Allen & Kovilakam, 2017; Davis & Birner, 2017; Grise et al., 2018; Staten et al., 2020; Staten et al., 2018; Studholme & Gulev, 2018; Grise & Davis, 2020). The observed widening of the annual mean HC, revealed by a variety of metrics, is primarily due to poleward shift of the NH HC. There have been stronger upward trends in the NH extent of HC after 1992 (Figure 2.17a). The magnitude of the recent changes based on modern-era reanalyses is not as large as that in AR5, due to apparent biases in older-generation reanalyses (Grise et al., 2019). Moreover, large interannual variability leads to uncertainties in estimates of long-term changes (Nguyen et al., 2013, Garfinkel et al., 2015b; Seviour et al., 2018; Staten et al., 2018), particularly for the NH given its zonal asymmetries (Staten et al., 2020; Wang et al., 2020). These large-scale features of the HC based on reanalyses agree with estimates revealed from the Integrated Global Radiosonde Archive (IGRA) during 1979–2012 (Lucas & Nguyen, 2015; Mathew et al., 2016). Recent trends based on reanalyses indicate a larger seasonal widening in the HC for summer and autumn in each hemisphere, although the magnitude of changes in HC extent is strongly dependent on dataset and metrics used (Grise et al., 2018; Hu Y. et al., 2018; Staten et al., 2018). The shifts in the HC position were accompanied by a narrowing ITCZ over the Atlantic and Pacific basins, with no significant change in its location and increases in the precipitation intensity (Byrne et al., 2018).

Trends in the HC intensity since 1979 differ between reanalyses, although there is a tendency toward HC intensification (Figure 2.17b) (Nguyen et al., 2013; Chen et al., 2014; D’Agostino and Lionello, 2017; Huang et al., 2019), which is more marked in the NH than the SH (Studholme and Gulev, 2018). However, the ability of reanalyses to represent the HC strength has been questioned due to inaccurate representation of latent heating distribution, which is directly related to tropical convection and influences the HC dynamics (Chenck and Polvani, 2019; Mathew and Kumar, 2019).

[START FIGURE 2.17 HERE]

Figure 2.17: Timeseries of the annual mean Northern Hemisphere (NH, top curves) and Southern Hemisphere (SH, bottom curves) Hadley Cell extent (a) and Hadley Cell intensity (b) since 1979. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.17 HERE]

Paleo evidence during the LGM indicates a weaker WC over the Indian Ocean (Windler et al., 2019).
DiNezio et al., 2018), with a stronger Pacific WC (DiNezio and Tierney, 2013). During the Holocene, a transition from a strong WC located more westward during the Early-to-Mid Holocene towards a weak and eastward shifted WC during the late Holocene was inferred from proxy records from the Pacific Warm Pool and South-eastern Asia (Barr et al., 2019; Dang et al., 2020; Griffiths et al., 2020), in concurrence with changes in ENSO activity (section 2.4.2). Reconstructions for the CE showed weakened WC during 1000–1250 and since 1850, with an intensified circulation during 1500–1850 CE (Deng et al., 2017; Xu et al., 2016).

Considering instrumental records, there is considerable interdecadal variability in the strength of the WC, resulting in time-period dependent magnitude and even sign of trends (Bordbar et al., 2017; Carilli et al., 2015; Hou et al., 2018), with some studies reporting weakening over the 20th century (e.g. Liu et al., 2019; Power and Kociuba, 2011), while others reported strengthening (Li Z. et al., 2020), particularly over the last 30–40 years (e.g. Hu et al., 2013; L’Heureux et al., 2013, Yim et al., 2017). Based on estimation of changes in mid-tropospheric velocity from changes in observed cloud cover, Bellomo and Clement (2015) suggest a weakening and eastward shift of the WC over 1920–2010, however the robustness of this signal is questionable due to high uncertainty in the ship reported cloud data records used before 1954. Using centennial-scale 20CR reanalysis Tseng et al. (2019) showed that the vertical westerly wind shear over the western Pacific does not indicate any long-term change during 1900–1980, but shows a marked increase since the 1980s that is not present in ERA-Interim and JRA-55, again calling into question the ability of centennial-scale reanalyses to capture tropical circulation changes. Recent strengthening together with a westward shift of the WC (Bayr et al., 2014; Ma & Zhou, 2016) was identified across several reanalysis products and observational datasets, and using different metrics for quantifying WC. Nevertheless, satellite observations of precipitation and analyses of upper tropospheric humidity suggest substantially weaker strengthening of the WC than implied by reanalyses (Chung et al., 2019). This recent strengthening in the WC is associated with enhanced precipitation in the tropical western Pacific, anomalous westerlies in the upper troposphere, strengthened downwelling in the central and eastern tropical Pacific, and anomalous surface easterlies in the western and central tropical Pacific (Dong and Lu, 2013; Choi et al., 2016, McGregor et al., 2014). Positive trends in sea level pressure over the eastern Pacific and concurrent negative trends over the Indonesian region result in a pattern implying a shift towards a La Niña-like WC regime, with strengthening of the Pacific Trade Winds mainly over 1979–2012 (England et al., 2014; L’Heureux et al., 2013; Sohn et al., 2016; Zhao & Allen, 2019). Seasonal assessment of the WC showed significant changes in the vertical westerly wind shear over the Pacific during the austral summer and autumn implying a strengthening (Clem et al., 2017).

In summary, there has been a likely widening of the Hadley circulation since the 1980s, mostly due to its extension in the NH, although there is only medium confidence in the extent of the changes. This has been accompanied by a strengthening of the Hadley circulation, particularly in the NH (medium confidence). There is low confidence in the estimation of long-term trends in the strength of the Walker circulation, which are time period dependent and subject to dataset uncertainties. Trends since 1980 are better characterised and consistent with a very likely strengthening that resembles a La Niña-like Walker circulation and a westward shift of the Walker circulation, although with medium confidence in the magnitude of the changes, arising from the differences between satellite observations and reanalysis products.

2.3.1.4.2 Global monsoon changes

AR5 reported a weakening of the global monsoon (GM) circulation as well as a decrease of global land monsoon rainfall over the second half of the 20th century. Nevertheless, there was low confidence in the observed circulation trends due to uncertainties in reanalysis products and in the definition of the monsoon area. From a paleo perspective, AR5 only assessed regional monsoon changes.

New research based on high-resolution proxies reinforces previous findings on the influence of orbital cycles on GM variability on millennial time scales. The intensity of the monsoon systems is generally out of phase between hemispheres, being associated with the precession cycle (~21–23 kyr) (An et al., 2015; Seth et al., 2019; Wang P.X. et al., 2017), with intensified NH monsoon systems during precession minima (Toucanne et al., 2015; Wagner et al., 2019). The eccentricity forcing (about 100 kyr cycle) shows stronger GM during
interglacial periods (An et al., 2015; Mohtadi et al., 2016; Wang P. X. et al., 2014; 2017). Changes in obliquity (about 41 kyr cycle) modify the strength of monsoon systems, with increased summer monsoon rainfall when obliquity is maximal (Liu Yi et al., 2015; Mohtadi et al., 2016). Millennial scale variability in GM during the LDT was also linked to the occurrences of Heinrich stadials, resulting in weakened NH monsoons and intensified SH monsoons (An et al., 2015; Margari et al., 2020; Wang P.X. et al., 2017).

An intensification of the NH monsoons in the early to mid-Holocene with increased precipitation and regional expansions of rainfall areas identified through a variety of proxy records is shown by Biasutti et al., (2018) and Wang P.X. et al. (2017). The response for the SH monsoons during this period indicates a weakening in both summer and winter precipitation (Sachs et al., 2018; Wang P.X. et al., 2014; 2017). A decline in GM precipitation and a retraction of the northern fringes of monsoon areas was inferred from the mid-Holocene onwards, with some regions experiencing wetter conditions during the mid to late Holocene compared with present and a strengthening of the SH monsoons (Sachs et al., 2018; Wang P.X. et al., 2014; 2017). For the CE, GM reconstructions exhibit inter-hemispheric contrast during the period 950–1250 CE, with intensified NH monsoons and weakened SH monsoons, and the opposite pattern during 1400–1850 CE (An et al., 2015; Wang P.X. et al., 2014).

Direct observations highlight that the GM land precipitation, particularly over the NH, experienced a slight increase from 1900 through the early 1950s, followed by an overall decrease from the 1950s to the 1980s, and then an increase to present (Huang X. et al., 2019; Kitoh et al., 2013; Wang B. et al., 2018; 2020). This highlights the existence of multidecadal variations in the NH monsoon circulation patterns and precipitation intensity (Monerie et al., 2019; Wang B. et al., 2013; Wang P.X. et al., 2014; 2017). An overall increase in monsoon precipitation during extended boreal summer (JJAS) over the NH since 1979 is revealed by GPCP (Deng et al., 2018; Han et al., 2019) and CMAP for 1980–2010 (Jiang et al., 2016). SH summer monsoon behaviour is dominated by strong interannual variability and large regional differences (Deng et al., 2018; Han et al., 2019; Jiang et al., 2016; Kamae et al., 2017; Kitoh et al., 2013; Lin et al., 2014), with no significant trends reported by GPCP and CMAP (Deng et al., 2018). Uncertainty predominantly arises from the observed increase in tropical precipitation seasonality (Feng et al., 2013) and the estimation of GM precipitation over the ocean areas, leading to a large apparent spread across datasets (Kitoh et al., 2013b; Kamae et al., 2017).

In summary, observed trends during the last century indicate that the GM precipitation decline reported in AR5 has reversed since the 1980s, with a likely increase mainly due to a significant positive trend in the NH summer monsoon precipitation (medium confidence). However, GM precipitation has exhibited large multidecadal variability over the last century, creating low confidence in the existence of centennial-length trends in the instrumental record. Proxy reconstructions show a likely NH monsoons weakening since the mid-Holocene, with opposite behaviour for the SH monsoons.

2.3.1.4.3 Extratropical jets, storm tracks, and blocking

AR5 reported a likely poleward shift of storm tracks and jet streams since the 1970s from different datasets, variables and approaches. These trends were consistent with the HC widening and the poleward shifting of the circulation features since the 1970s. There was low confidence in any large-scale change in blocking.

Proxy records consistent with modeller results imply a southward shift of the storm tracks over the North Atlantic during the LGM (Raible et al., 2021). A variety of proxies are available for the changes in the position of the extratropical jets / westerlies during the Holocene. Recent syntheses of moisture-sensitive proxy records indicate drier-than-present conditions over mid-latitudes of western North America (Hermann et al., 2018; Liefert and Shuman, 2020) during the MH, which together with a weakened Aleutian Low (Bailey et al., 2018) implies that the winter North Pacific jetstream was shifted northward. A synthesis of lines of evidence from the SH indicates that the westerly winds were stronger over 14–5 ka, followed by regional asymmetry after 5 ka (Fletcher & Moreno, 2012). There is no consensus on the shifts of the SH westerlies with some studies implying poleward migrations (Lamy et al., 2010; Voigt et al., 2015; Turney et al., 2017; Anderson et al., 2018) and others suggesting an equatorward shift (Kaplan et al., 2016) in the MH.
During 950–1400 CE, hydroclimate indicators suggest a northward shift of Pacific storm tracks over North America (McCabe-Glynn et al., 2013; Steinman et al., 2014) which was comparable in magnitude to that over 1979–2015 (Wang et al., 2017). Storm tracks over the North Atlantic-European sector shifted northward as indicated by multi-proxy indicators over the North Atlantic (Wirth et al., 2013; Orme et al., 2017) and Mediterranean (Roberts et al., 2012). Reconstructed westerly winds in the SH suggest a poleward shift (Goodwin et al., 2014; Koffman et al., 2014; Lamy et al., 2010; Moreno et al., 2018; Schimpf et al., 2011), with latitudinal change comparable to that during recent decades (Swart and Fyfe, 2012; Manney and Hegglin, 2018).

Multiple reanalyses show that since 1979 the subtropical jet wind speeds have generally increased in winter and decreased in summer in both hemispheres, but the trends are regionally dependent (Pena-Ortiz et al., 2013; Manney and Hegglin, 2018; Lee et al., 2019). Over NH mid-latitudes, the summer zonal wind speeds have weakened in the mid-troposphere (Francis and Vavrus, 2012; Coumou et al., 2014, 2015; Haimberger and Mayer, 2017). Meanwhile there are indications of enhanced jetstream meandering in boreal autumn at the hemispheric scale (Francis and Vavrus, 2015; Di Capua and Coumou, 2016), whereas the regional arrangement of meandering depends on the background atmospheric state (Cohen et al., 2020). These meandering trends, however, are sensitive to the metrics used (Cattiaux et al., 2016; Hassanzadeh et al., 2014; Screen & Simmonds, 2013; Vavrus, 2018). Hypothesised links to Arctic warming are assessed in Cross-Chapter Box 10.1.

Multiple reanalyses and radiosonde observations show an increasing number of extratropical cyclones over the NH since the 1950s (Chang & Yau, 2016; Wang X. L. et al., 2016). The positive trends are generally consistent among reanalyses since 1979, though with considerable spread (Tilinina et al., 2013; Wang X. L. et al., 2016). In recent decades the number of deep extratropical cyclones has increased over the SH (Reboita et al., 2015; Wang X. L. et al., 2016) (Section 8.3.2.8.1, Figure 8.12), while the number of deep cyclones has decreased in the NH in both winter and summer (Neu et al., 2013; Wang et al., 2017, Chang et al., 2016; Coumou et al., 2015; Gertler & O’Gorman, 2019). The regional changes for different intensity extratropical cyclones are assessed in Section 8.3.2.8.1. The assessment of trends is complicated by strong interannual-decadal variability, sensitivity to dataset choice and resolution (Tilinina et al., 2013; Lucas et al., 2014; Wang et al., 2016; Pepler et al., 2018; Rohrer et al., 2018) and cyclone identification / tracking methods (Neu et al., 2013; Grieger et al., 2018). Thus there is overall low confidence for recent changes in global extratropical storm tracks.

A consistent poleward shift of the tropospheric extratropical jets since 1979 is reported by multiple reanalyses (Davis and Rosenlof, 2012; Davis and Birner, 2013; Pena-Ortiz et al., 2013; Manney and Hegglin, 2018) (Figure 2.18) and radiosonde winds (Allen et al., 2012). This is generally consistent with the earlier reported shifts retrieved from satellite temperature observations (Fu and Lin, 2011; Davis and Rosenlof, 2012). After the 1990s the magnitude of meridional shifts in extratropical jets over both the North Atlantic and North Pacific in August is enhanced compared to multi-century variability (Trouet et al., 2018). Despite some regional differences (Ma & Zhang, 2018; Melamed-Turkis et al., 2018; Norris et al., 2016; Wang J. et al., 2017a; Woolings et al., 2014; Xue & Zhang, 2018), overall poleward deflection of storm tracks in boreal winter over both the North Atlantic and the North Pacific was identified during 1979–2010 (Tilinina et al., 2013). Over the SH extra-tropics there is a similarly robust poleward shift in the polar jet since 1979 (Pena-Ortiz et al., 2013; Manney and Hegglin, 2018; WMO, 2018), although after 2000 the December-January-February (DJF) tendency to poleward shift of the SH jet stream position ceased (Banerjee et al., 2020). The general poleward movement in midlatitude jet streams (Lucas et al., 2014) is consistent with the expansion of the tropical circulation (Section 2.3.1.4.1). The changes of extratropical jets and westerlies are also related to the annual modes of variability (Section 2.4, Annex IV).

Robust trends in blocking have only been found in certain regions and specific seasons during recent decades. Increases in blocking frequency have occurred over low-latitude regions in the North Atlantic in boreal winter (Davini et al., 2012), the South Atlantic in austral summer (Dennison et al., 2016) and the southern Indian Ocean in austral spring (Schemm, 2018). Over the subpolar North Atlantic sustained periods of positive Greenland blocking were identified during 1870–1900 and from the late 1990s to 2015 (Hanna et al., 2015). Further analysis of association of Greenland blocking with the NAM is provided in Section 2.4.1.1. Meanwhile, a reduced blocking frequency has been found over winter in Siberia (Davini et al., 2012).
and the southwestern Pacific in austral spring (Schemm, 2018). Over Eastern European Russia and Western Siberia (40–100E) a tendency towards longer blocking events was reported by Luo et al. (2016) for 2000–2013 and by Tyrilis et al. (2020) for 1979–2017. Inter-annual variance in the number of blocking events over the SH (Oliveira et al., 2014) and North Atlantic (Kim & Ha, 2015) has enhanced. Blocking events and their trends are sensitive to choice of datasets, calculation periods and methods (Barnes et al., 2014; Cheung et al., 2013; Kononova & Lupo, 2020; Pepler et al., 2018; Rohrer et al., 2018; Woollings et al., 2018b). As a result, hemispheric and global trends in blocking frequency have overall low confidence.

In summary, the total number of extratropical cyclones has likely increased since the 1980s in the NH (low confidence), but with fewer deep cyclones particularly in summer. The number of strong extratropical cyclones has likely increased in the SH (medium confidence). The extratropical jets and cyclone tracks have likely been shifting poleward in both hemispheres since the 1980s with marked seasonality in trends (medium confidence). There is low confidence in shifting of extratropical jets in the NH during the mid-Holocene and over 950–1400 CE to latitudes that likely were similar to those since 1979. There is low confidence in observed global-scale changes in the occurrence of blocking events.

**[START FIGURE 2.18 HERE]**

Figure 2.18: Trends in ERA5 zonal-mean zonal wind speed. Shown are (a) DJF (December-January-February), (b) MAM (March-April-May), (c) JJA (June-July-August) and (d) SON (September-October-November). Climatological zonal winds during the data period are shown in solid contour lines for westerly winds and in dashed lines for easterly. Trends are calculated using OLS regression after Santer et al. (2008a) (‘x’ marks denote non-significant trends). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

**[END FIGURE 2.18 HERE]**

2.3.1.4.4  *Surface wind and sea level pressure*

AR5 concluded that surface winds over land had generally weakened. The confidence for both land and ocean surface wind trends was low owing to uncertainties in datasets and measures used. Sea level pressure (SLP) was assessed to have likely decreased from 1979–2012 over the tropical Atlantic and increased over large regions of the Pacific and South Atlantic, but trends were sensitive to the period analysed.

Terrestrial in situ wind datasets have been updated and the quality-control procedures have been improved, with particular attention to homogeneity and to better retaining true extreme values (Dunn et al., 2012; 2014; 2016). Global mean land wind speed (excluding Australia) from HadISD for 1979–2018 shows a reduction (stilling) of 0.063 m s\(^{-1}\) per decade (Azorin-Molina et al., 2019). Trends are broadly insensitive to the subsets of stations used. Although the meteorological stations are unevenly distributed worldwide and scarce in South America and Africa, the majority exhibit stilling particularly in the NH (Figure 2.19). Regionally, strong decreasing trends are reported in central Asia and North America (−0.106 and −0.084 m s\(^{-1}\) per decade respectively) during 1979–2018 (McVicar et al., 2012; Vautard et al., 2012; Wu et al., 2018; Azorin-Molina et al., 2019). This stilling tendency has reversed after 2010 and the global mean surface winds have strengthened (Azorin-Molina et al., 2020; Zeng et al., 2019), although the robustness of this reversal is unclear given the short period and interannual variability (Kousari et al., 2013; Kim and Paik, 2015; Azorin-Molina et al., 2019).

**[START FIGURE 2.19 HERE]**

Figure 2.19: Trends in surface wind speed. (a) Station observed winds from the integrated surface database (HadISD v2.0.2.2017f); (b) Cross-Calibrated Multi-Platform wind product; (c) ERA5; and (d) wind speed from the Objectively Analyzed Air-Sea Heat Fluxes dataset, release 3. White areas indicate incomplete or missing data. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al. (2008c) (‘x’ marks denote non-significant trends). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

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Over the ocean, datasets demonstrate considerable disagreement in surface wind speed trends and spatial features (Kent et al., 2013). Global ocean surface winds from NOCV2.0 demonstrate upward trends of about 0.11 m s⁻¹ per decade (1979–2015) with somewhat smaller trends from WASwind for 1979–2011 (Azorin-Molina et al., 2017, 2019). The datasets are consistent until 1998, but diverge thereafter. Both ERA5 and JRA-55 reanalyses show consistently increasing global marine wind speeds over 1979–2015, though flattening since 2000, whereas MERRA-2 agrees until 1998, but then exhibits increased variability and an overall decrease in the last two decades (Azorin-Molina et al., 2019). This agrees with estimates by Sharma et al. (2021) showing upward ocean wind trends from 1979 to 2000 which are consistent in ERA-Interim, ERA5 and MERRA-2, but disagree with CFSR trends for the same period. Over 2000–2019 all reanalyses show diverging tendencies. An updated multiphase satellite database (comprising data from altimeters, radiometers, and scatterometers) from 1985–2018 shows small increases in mean wind speed over the global ocean, with the largest increase observed in the Southern Ocean (Young & Ribal, 2019), consistent with signals in ERA-Interim, ERA5 and MERRA-2 (Sharmar et al., 2021). Overall, most products suggest positive trends over the Southern Ocean, western North Atlantic and the tropical eastern Pacific since the early 1980s.

The modern era reanalyses exhibit SLP increases over the SH sub tropics with stronger increases in austral winter over 1979–2018. Over the NH, SLP increased over the mid-latitude Pacific in boreal winter and decreased over the eastern subtropical and mid-latitude North Atlantic in boreal summer. Discrepancies in the low-frequency variations during the first half of the 20th century exist in the centennial-scale reanalysis products (Befort et al., 2016). Overall, modern reanalysis datasets support the AR5 conclusion that there is no clear signal for trends in the strength and position of the permanent and quasi-permanent pressure centres of action since the 1950s. Instead, they highlight multi-decadal variations. Large-scale SLP is strongly associated with the changes in modes of variability (Section 2.4, Annex IV).

In summary, since the 1970s a worldwide weakening of surface wind has likely occurred over land, particularly marked in the NH, with low confidence in a recent partial recovery since around 2010. Differences between available wind speed estimates lead to low confidence in trends over the global ocean as a whole but with most estimates showing strengthening globally over 1980–2000 and over the last four decades in the Southern Ocean, western North Atlantic and the tropical eastern Pacific.

### 2.3.1.4.5 Stratospheric polar vortex and sudden warming events

AR5 assessed changes in the polar vortices and reported a likely decrease in the lower-stratospheric geopotential heights over Antarctica in spring and summer at least since 1979.

Multiple definitions for the polar vortex strength and sudden stratospheric warming (SSW) events have been proposed and compared (Butler et al., 2015; Palmeiro et al., 2015; Waugh et al., 2017; Butler and Gerber, 2018), and new techniques identifying daily vortex patterns and SSWs have been developed (Kretschmer et al., 2018; Mitchell et al., 2013a). Errors in reanalysis stratospheric winds were assessed and discrepancies in stratospheric atmospheric circulation and temperatures between reanalyses, satellites and radiosondes have been reported (Mitchell et al., 2013a; Duruissseau et al., 2017).

The northern stratospheric polar vortex has varied seasonally and with altitude during recent decades. Multiple reanalysis and radiosonde datasets show that the midwinter lower stratospheric geopotential height (150 hPa) over the polar region north of 60°N has increased significantly since the early 1980s (Bohlinger et al., 2014; Garfinkel et al., 2017). This signal extends to the middle and upper stratosphere. In January-February zonal winds north of 60°N at 10 hPa have been weakening (Kim et al., 2014; Kretschmer et al., 2018). Daily atmospheric circulation patterns over the northern polar stratosphere exhibit a decreasing frequency of strong vortex events and commensurate increase in more-persistent weak events, which largely explains the observed significant weakening of the vortex during 1979–2015 (Kretschmer et al., 2018). The northern polar vortex has weakened in early winter but strengthened during late winter (Bohlinger et al., 2014; Garfinkel et al., 2017).
The SSW, a phenomenon of rapid stratospheric air temperature increases (sometimes by more than 50°C in 1–2 days), is tightly associated with the reversal of upper stratospheric zonal winds, and a resulting collapse or substantial weakening of the stratospheric polar vortex (Butler et al., 2015; Butler and Gerber, 2018) and on average occurs approximately 6 times per decade in the NH winter (Charlton et al., 2007; Butler et al., 2015). The SSW record from all modern reanalyses is very consistent. There is a higher occurrence of major midwinter SSWs in the 1980s and 2000s with no SSW events during 1990–1997 (Reichler et al., 2012, Butler et al., 2015). An assessment of multidecadal variability and change in SSW events is sensitive to both chosen metric and methods (Palmeiro et al., 2015). Due to the lack of assimilation of upper air data, the centennial-scale reanalyses do not capture SSW events, even for the most recent decades (Butler et al., 2015, 2017) and hence cannot inform on earlier behaviour. There has been considerably less study of trends in the southern hemispheric stratosphere vortex strength despite the interest in the ozone hole and the likely impact of the southern hemispheric stratosphere vortex strength on it. The occurrence of SSW events in the SH is not as frequent as in the NH, with only 3 documented events in the last 40 years (Shen et al., 2020).

In summary, it is likely that the northern lower stratospheric vortex has weakened since the 1980s in midwinter, and its location has shifted more frequently toward the Eurasian continent. The short record and substantial decadal variability yields low confidence in any trends in the occurrence of SSW events in the NH winter and such events in the SH are rare.

2.3.2 Cryosphere

This section focuses on large-scale changes in a subset of components of the cryosphere (Cross-Chapter Box 2.2). Chapter 9 undertakes a holistic assessment of past and possible future changes at the process level in the cryosphere, including those at regional scales, integrating observations, modelling and theoretical understanding, while, here in chapter 2, the focus is on past large-scale, observation-based cryospheric changes.

2.3.2.1 Sea ice coverage and thickness

2.3.2.1.1 Arctic Sea Ice

AR5 reported that the annual mean Arctic sea ice extent (SIE) very likely decreased by 3.5–4.1% per decade between 1979 and 2012 with the summer sea ice minimum (perennial sea ice) very likely decreasing by 9.4–13.6% per decade. This was confirmed by SROCC reporting the strongest reductions in September (12.8 ± 2.3% per decade; 1979–2018) and stating that these changes were likely unprecedented in at least 1 kyr (medium confidence). The spatial extent had decreased in all seasons, with the largest decrease for September (high confidence). AR5 reported also that the average winter sea ice thickness within the Arctic Basin had likely decreased by between 1.3 m and 2.3 m from 1980 to 2008 (high confidence), consistent with the decline in multi-year and perennial ice extent. SROCC stated further that it was virtually certain that Arctic sea ice had thinned, concurrent with a shift to younger ice. Lower sea ice volume in 2010–2012 compared to 2003–2008 was documented in AR5 (medium confidence). There was high confidence that, where the sea ice thickness had decreased, the sea ice drift speed had increased.

Proxy records are used in combination with modelling to assess Arctic paleo-sea-ice conditions to the extent possible. For the Pliocene, limited proxy evidence of a reduced sea-ice cover compared to ‘modern’ winter conditions (Knies et al., 2014; Clotten et al., 2018) and model simulations of a largely ice-free Arctic Ocean during summer (Feng et al., 2019; Howell et al., 2016; Li F. et al., 2020) imply medium confidence that the
Arctic Ocean was seasonally ice covered. Over the LIG, sparse proxy reconstructions (Stein et al., 2017; Kremer et al., 2018) and proxy evidence from marine sediment (Kageyama et al., 2021) provide medium confidence of perennial sea-ice cover.

Over the past 13 kyr proxy records suggest extensive sea ice coverage during the Younger Dryas (at the end of the LDT), followed by a decrease in sea ice coverage during the Early Holocene, and increasing sea ice coverage from the MH to the mid-1400s (De Vernal et al., 2013; Belt et al., 2015; Cabello-Sanz et al., 2016; Armand et al., 2017; Belt, 2018). There is limited evidence that the Canadian Arctic had less multiyear sea ice during the Early Holocene than today (Spolaor et al., 2016). For more regional details on paleo arctic sea ice see Section 9.3.1.1.

Current pan-Arctic SIE conditions (annual means and late summer) are unprecedented since at least 1850 (Brennan et al., 2020; Walsh et al., 2017; 2019, Figure 2.20a), while, as reported in SROCC, there remains medium confidence that the current September (late summer) Arctic sea ice loss is unprecedented during the past 1 kyr. Sea ice charts since 1850 (Walsh et al., 2017; 2019) suggest that there was no significant trend before the 1990s, but the uncertainty of these estimates is large and could mask a trend, a possibility illustrated by Brennan et al. (2020), who found a loss of Arctic sea ice between 1910 and 1940 in an estimate based on a data assimilation approach.

There has been a continuing decline in SIE and Arctic sea ice area (SIA) in recent years (Figure 2.20a). To reduce grid-geometry associated biases and uncertainties (Notz, 2014; Ivanova et al., 2016; Meier and Stewart, 2019) SIA is used in addition to, or instead of SIE herein (see also section 9.3.1). A record-low Arctic SIA since the start of the satellite era (1979) occurred in September 2012 (Figure 2.20a). Decadal SIA means based on the average of three different satellite products decreased from 6.23–5.76 million km$^2$ for September and 14.52–13.42 million km$^2$ for March SIA (Figure 2.20a). Initial SIA data for 2020 (OSISAF) are within the range of these recent decadal means or slightly below (Figure 2.20a). SIA has declined since 1979 across the seasonal cycle (Figure 9.13). Most of this decline in SIA has occurred after 2000, and is superimposed by substantial interannual variability. The sharp decline in Arctic summer SIA coincides with earlier surface melt onset (Mortin et al., 2016; Bliss et al., 2017), later freeze-up, and thus a longer ice retreat and open water period (Stammerjohn et al., 2012; Parkinson, 2014; Peng et al., 2018).

**[START FIGURE 2.20 HERE]**

**Figure 2.20: Changes in Arctic and Antarctic sea ice area.** (a) Three time series of Arctic sea ice area (SIA) for March and September from 1979 to 2020 (passive microwave satellite era). In addition, the range of SIA from 1850–1978 is indicated by the vertical bar to the left. Decadal means for the three series for the first and most recent decades of observations are shown by horizontal lines in grey (1979–1988) and black (2010–2019). (b) Three time series of Antarctic sea ice area for September and February (1979–2020). Sea ice area values have been calculated from sea ice concentration fields. Available data for 2020 (OSISAF) is shown in both (a) and (b). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

**[END FIGURE 2.20 HERE]**

Over the past two decades, first-year sea ice has become more dominant and the oldest multiyear ice (older than 4 years) which in March 1985 made up 33% of the Arctic sea ice cover, has nearly disappeared, making up 1.2% in March 2019 (Perovich et al., 2020). The loss of older ice is indicative of a thinning overall of ice cover (Tschudi et al., 2016), but also the remaining older ice has become thinner (Hansen E. et al., 2013). Since in situ ice thickness measurements are sparse, information about ice thickness is mainly based on airborne and satellite surveys. Records from a combination of different platforms show for the central and western Arctic Ocean (Arctic Ocean north of Canada and Alaska) negative trends since the mid-1970s (Kwok, 2018; Lindsay & Schweiger, 2015), with a particularly rapid decline during the 2000s, which coincided with a large loss of multiyear sea ice. Direct observations from 2004 and 2017 indicate a decrease
of modal ice thickness in the Arctic Ocean north of Greenland by 0.75 m, but with little thinning between 2014 and 2017 (Haas et al., 2017). This agrees with data based on satellite altimetry and airborne observations, showing no discernible thickness trend since 2010 (Kwok & Cunningham, 2015; Kwok, 2018; Kwok and Kacimi, 2018) (Figure 2.21). However, sea ice thickness derived from airborne and spaceborne data is still subject to uncertainties imposed by snow loading. For radar altimeters, insufficient penetration of radar signal into the snowpack results in overestimation of ice thickness (e.g., Ricker et al., 2015; King et al., 2018a; Nandan et al., 2020). Negative trends in ice thickness since the 1990s are also reported from the Fram Strait in the Greenland Sea, and north of Svalbard (Hansen E. et al., 2013; King et al., 2018; Renner et al., 2014; Rössel et al., 2018; Spreen et al., 2020). Thickness data collected in the Fram Strait originate from ice exported from the interior of the Arctic Basin and are representative of a larger geographical area upstream in the transpolar drift. A reduction of survival rates of sea ice exported from the Siberian shelves by 15% per decade has interrupted the transpolar drift and affected the long-range transport of sea ice (Krumpen et al., 2019). The thinner and on average younger ice has less resistance to dynamic forcing, resulting in a more dynamic ice cover (Hakkinen et al., 2008; Spreen et al., 2011; Vihma et al., 2012; Kwok et al., 2013).

[SSTART FIGURE 2.21 HERE]

Figure 2.21: Arctic sea ice thickness changes (means) for autumn (red/dotted red) and winter (blue/dotted blue). Shadings (blue and red) show 1 S.E. ranges from the regression analysis of submarine ice thickness and expected uncertainties in satellite ice thickness estimates. Data release area of submarine data ice thickness data is shown in inset. Satellite ice thickness estimates are for the Arctic south of 88°N. Thickness estimates from more localized airborne/ground electromagnetic surveys near the North Pole (diamonds) and from Operation IceBridge (circles) are shown within the context of the larger scale changes in the submarine and satellite records. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.21 HERE]

SROCC noted the lack of continuous records of snow on sea ice. Nevertheless in recent decades, more snow on sea ice has been observed in the Atlantic sector in the Arctic than in the Western Arctic Ocean (Webster et al., 2018). Previously, Warren et al. (1999) showed that over 1954–1991 there were weak trends towards declining snow depth on sea ice in the Pacific sector. Recent observations indicate a substantial thinning of the spring snowpack in the western Arctic (Cavalieri et al., 2012; Brucker and Markus, 2013; Kurtz et al., 2013; Laxon et al., 2013; Webster et al., 2018). In contrast, thick snow over Arctic sea ice in the Atlantic sector north of Svalbard (snow thickness around 0.4 m or more) has been observed in the 1970s and since the 1990s (Rössel et al., 2018), but data are too sparse to detect trends.

In summary, over 1979–2019 Arctic SIA has decreased for all months, with the strongest decrease in summer (very high confidence). Decadal means for SIA decreased from the first to the last decade in that period from 6.23 to 3.76 million km² for September, and from 14.52 to 13.42 million km² for March. Arctic sea ice has become younger, thinner and faster moving (very high confidence). Snow thickness on sea ice has decreased in the western Arctic Ocean (medium confidence). Since the Younger Dryas at the end of the LDT, proxy indicators show that Arctic sea ice has fluctuated on multiple time scales with a decrease in sea ice coverage during the Early Holocene and an increase from the MH to the mid-1400s. Current pan-Arctic sea ice coverage levels (annual mean and late summer) are unprecedentedly low since 1850 (high confidence), and with medium confidence for late summer for at least the past 1 kyr.

2.3.2.1.2 Antarctic Sea Ice

AR5 reported a small but significant increase in the total annual mean Antarctic SIE that was very likely in the range of 1.2–1.8% per decade between 1979 and 2012 (0.13–0.20 million km² per decade) (very high confidence), while SROCC reported that total Antarctic sea ice coverage exhibited no significant trend over the period of satellite observations (1979–2018) (high confidence). SROCC noted that a significant positive
trend in mean annual ice cover between 1979 and 2015 had not persisted, due to three consecutive years of below-average ice cover (2016–2018). SROCC stated also that historical Antarctic sea ice data from different sources indicated a decrease in overall Antarctic sea ice cover since the early 1960s, but was too small to be separated from natural variability (high confidence).

There is only limited evidence from predominantly regional paleo proxies for the evolution of Southern Ocean sea ice before the instrumental record and estimates are not available for all proxy target periods (Section 9.3.2). Proxies from marine sediments for intervals preceding and following the MPWP indicate open water conditions with less sea-ice than modern conditions (Taylor-Silva and Rieselman, 2018; Ishino and Suto, 2020). During the LGM, proxies indicate that summer sea ice coverage reached the polar ocean front (e.g., Nair et al., 2019). More recently, sea ice coverage appears to have fluctuated substantially throughout the Holocene (e.g., for the western Amundsen Sea, Lamping et al., 2020). At the beginning of the CE, regional summer sea ice coverage in the north-western Ross Sea was lower than today (Tesi et al., 2020). Crosta et al. (2021) suggest, based on different proxies, four different phases with 7–10 months periods of sea ice occurrence per year in the Antarctic region off Adelie Land during the CE, where each phase was several hundred years long.

More recent sea ice reconstructions are based on diverse sources including whaling records (De La Mare, 1997, 2009; Cotté and Guinet, 2007), old ship logbooks (Ackley et al., 2003; Edinburgh and Day, 2016), and ice core records (Curran et al., 2003; Abram et al., 2010; Sinclair et al., 2014), amongst other methods (e.g. Murphy et al., 2014). These reconstructions, in combination with recent satellite-based observations indicate: i) a decrease in summer SIE across all Antarctic sectors since the early- to mid-20th century; ii) a decrease in winter SIE in the East Antarctic, and Amundsen-Bellingshausen Seas sectors starting in the 1960s; and iii) small fluctuations in winter SIE in the Weddell Sea over the 20th century (Hobbs et al., 2016a,b). There are also ice-core indications that the pronounced Ross Sea increase dates back to the mid-1960s (Sinclair et al., 2014; Thomas & Abram, 2016). While there is reasonable broad-scale concurrence across these estimates, the uncertainties are large, there is considerable interannual variability, and reconstructions require further validation (Hobbs et al., 2016a,b). New reconstructions (Thomas et al., 2019) from Antarctic land ice cores show that SIE in the Ross Sea had increased between 1900 and 1990, while the Bellingshausen Sea had experienced a decline in SIE; this dipole pattern is consistent with satellite-based observations from 1979 to 2019 (Parkinson, 2019), but the recent rate of change then has been larger. Records of Antarctic SIE for the late 19th and early 20th centuries (Edinburgh and Day, 2016), show SIE comparable with the satellite era, although with marked spatial heterogeneity (e.g., Thomas et al., 2019).

Early Nimbus satellite visible and infrared imagery from the 1960s (Meier et al., 2013; Gallacher et al., 2014) indicate higher overall SIE compared to 1979–2013 (Hobbs et al., 2016a,b), but with large uncertainties and poorly quantified biases (NA SEM, 2017). The continuous satellite passive-microwave record shows that there was a modest increase in overall Antarctic SIA of 2.5% ± 0.2% per decade (1 standard error over 1979–2015 (Comiso et al., 2017). For overall ice coverage and for this period, positive long-term trends were most pronounced during austral autumn advance (Maksym, 2019), being moderate in summer and winter, and lowest in spring (Comiso et al., 2017; Hobbs et al., 2016a,b; Holland, 2014; Turner et al., 2015). Since 2014, overall Antarctic SIE (and SIA) has exhibited major fluctuations from record-high to record-low satellite era extents (Comiso et al., 2017; Massonnet et al., 2015; Parkinson, 2019; Reid, and Massom, 2015; Reid et al., 2015). After setting record-high extents each September from 2012 through 2014, Antarctic SIE (and SIA) dipped rapidly in mid-2016 and remained predominantly below average through 2019 (Reid et al., 2020). For the most recent decade of observations (2010–2019), the decadal means of three SIA products (Figure 2.20b) were 2.17 million km$^2$ for February and 15.75 million km$^2$ for September, respectively. The corresponding levels for the means for the first decade of recordings (1979–1988) were 2.04 million km$^2$ for February and 15.39 million km$^2$ for September indicating little overall change. Initial SIA data for 2020 (OSISAF) show SIA for September above, and for February slightly below the recent decadal means (Fig. 2.20b). The 2020 September level (OSISAF) remains below the levels observed over 2012–2014.

In summary, Antarctic sea ice has experienced both increases and decreases in SIA over 1979–2019, and substantively lower levels since 2016, with only minor differences between decadal means of SIA for the first (for February 2.04 million km$^2$, for September 15.39 million km$^2$) and last decades (for February 2.17 million km$^2$, for September 15.75 million km$^2$) of satellite observations (high confidence). There remains low
2.3.2.2 Terrestrial snow cover

AR5 concluded that snow cover extent (SCE) had decreased in the NH, especially in spring (very high confidence). For 1967–2012, the largest change was in June and March-April SCE very likely declined. No trends were identified for the SH due to limited records and large variability. SROCC concluded with high confidence that Arctic June SCE declined between 1967 and 2018 and in nearly all mountain regions, snow cover declined in recent decades.

Analysis of the combined in situ observations (Brown, 2002) and the multi-observation product (Mudryk et al. 2020) indicates that since 1922, April SCE in the NH has declined by 0.29 million km² per decade, with significant interannual variability (Figure 2.22) and regional differences (Section 9.5.3.1). The limited pre-satellite era data does not allow for a similar assessment for the entire spring-summer period. Assessment of SCE trends in the NH since 1978 indicates that for the October to February period there is substantial uncertainty in trends with the sign dependent on the observational product. Analysis using the NOAA Climate Data Record shows an increase in October to February SCE (Hernández-Henríquez et al., 2015; Kunkel et al., 2016) while analyses based on satellite borne optical sensors (Hori et al., 2017) or multi-observation products (Mudryk et al., 2020) show a negative trend for all seasons (Section 9.5.3.1, Figure 9.23). The greatest declines in SCE have occurred during boreal spring and summer, although the estimated magnitude is dataset dependent (Rupp et al., 2013; Estilow et al., 2015; Bokhorst et al., 2016; Thackeray et al., 2016; Connolly et al., 2019).

Figure 2.22: April snow cover extent (SCE) for the Northern Hemisphere (1922–2018). Shading shows very likely range. The trend over the entire 1922–2018 period (black line) is −0.29 (± 0.07) million km² per decade. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

There has been a commensurate decrease in the snow-cover duration and persistence, particularly in higher latitudes due to earlier spring melt and, in some cases, later autumn onset of snow cover (Chen Xiaona et al., 2015; Derksen et al., 2015; Hammond et al., 2018; Hori et al., 2017). Arctic snow-cover duration has decreased by 2–4 days per decade since the 1970s (Brown et al., 2017). Significant decreases in snow-cover duration have been documented over western Eurasia since 1978 (Hori et al., 2017).

For the NH, maximum snow depth has generally decreased since the 1960s, with more robust trends for North America and greater uncertainty for Eurasia (Kunkel et al., 2016). Several satellite-based passive microwave and other products indicate general declines in pre-melt snow water equivalent since 1981 although there is regional and inter-dataset variability (Brown et al., 2017; Jeong et al., 2017; Marty et al., 2017; Mottin et al., 2020; Mudryk et al., 2020; Pulliainen et al., 2020, Section 9.5.3).

In summary, substantial reductions in spring snow cover extent have occurred in the NH since 1978 (very high confidence) with limited evidence that this decline extends back to the early 20th century. Since 1981 there has been a general decline in NH spring snow water equivalent (high confidence).

2.3.2.3 Glacier mass

AR5 concluded with high confidence that, during the Holocene, glaciers were at times smaller than at the
end of the 20th century. AR5 stated further with very high confidence that most glaciers had been shrinking since the mid-1800s, and the mass loss from all glaciers worldwide very likely increased from 1970 to 2009. SROCC reported a globally coherent picture of continued glacier recession in recent decades (very high confidence) based on in situ and satellite observations of changes in glacier area, length and mass, although there were considerable inter-annual and regional variations. Between 2006 and 2015 the global glacier mass change assessed by SROCC was $-278 \pm 113 \text{ Gt yr}^{-1}$.

Two recent global reviews on glaciers over the Holocene (Solomina et al., 2015) and the past 2 kyr (Solomina et al., 2016) summarize the chronologies of respectively 189 and 275 glaciers. The former shows that glaciers retreated during the LDT and retracted to their minimum extent between 8 ka and 6 ka. Except for some glaciers in the SH and tropics, glaciers expanded thereafter, reaching their maximum extent beyond their present-day margins during the mid-1400s to late-1800s CE. With few exceptions, glacier margins worldwide have retreated since the 19th century, with the rate of retreat and its global character since the late 20th century being unusual in the context of the Holocene (Solomina et al., 2016, Figure 2.23a). However, the areal extents of modern glaciers in most places in the NH are still larger than those of the early and/or middle Holocene (Solomina et al., 2015). When considering Holocene and present glaciers extents, it is important to account for the relatively long adjustment time of glaciers (often referred to as response time; Section 9.5.1.3); the majority of modern glaciers are currently out of equilibrium with current climate, even without further global warming (Christian et al., 2018; Marzeion et al., 2018; Mermild et al., 2013; Zekollari et al., 2020). The size of glaciers during other periods warmer than the Early to Mid-Holocene, such as the MPWP and LIG, is largely unknown because the deposits marking previous extents were in almost all cases over-ridden by later glaciations. For Arctic glaciers, different regional studies consistently indicate that in many places glaciers are now smaller than they have been in millennia (Harning et al., 2016, 2018; Lowell et al., 2013; Miller et al., 2013, 2017; Pendleton et al., 2019; Schweinsberg et al., 2017, 2018).

New glacier outline (RGI Consortium, 2017) and glacier mass compilations (Ciraci et al., 2020; Hugonnet et al., 2021; Zemp et al., 2019; 2020) improve, refine and update the quantification of glacier areal and mass changes based on observations from in situ and remote sensing data. Observations between the 1960s and 2019 indicate that mass loss has increased over recent decades (Figure 2.23b). The overall global glacier mass loss rate has increased from 240 ± 9 Gt yr$^{-1}$ over 2000–2009 to 290 ± 10 Gt yr$^{-1}$ over 2010–2019 (Hugonnet et al., 2021), confirming that the last decade exhibits the most negative glacier mass balance since the beginning of the observational record. Observations are in general consistent with trends revealed by global glacier mass change modelling for almost the entire 20th century (1901–1990) implying an estimated mass loss (without uncharted glaciers (Parkes and Marzeion, 2018) and excluding peripheral glaciers of Greenland and Antarctica) of very likely 210 ± 90 Gt yr$^{-1}$ and very likely 170 ± 80 Gt yr$^{-1}$ for the period 1971–2019 (Marzeion et al., 2015; Section 9.5.1 and Table 9.5).

[START FIGURE 2.23 HERE]

Figure 2.23: Glacier advance and annual mass change. (a) Number of a finite selection of surveyed glaciers that advanced during the past 2000 years. (b) Annual and decadal global glacier mass change (Gt yr$^{-1}$) from 1961 until 2018. In addition, mass change mean estimates are shown. Ranges show the 90% confidence interval. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.23 HERE]

In summary, there is very high confidence that, with few exceptions, glaciers worldwide have retreated since the second half of the 19th century, and continue to retreat. The current global character of glacier mass loss is highly unusual (almost all glaciers simultaneously receding) in the context of at least the last 2 kyr (medium confidence). Glacier mass loss rates have increased since the 1970s (high confidence). Although many surveyed glaciers are currently more extensive than during the MH (high confidence), they generally are in disequilibrium with respect to current climate conditions and hence are committed to further ice loss.
2.3.2.4 Ice sheet mass and extent

During glacial periods, ice sheets were more extensive and the state of knowledge on their paleo-reconstruction can be found in recent publications (e.g. Batchelor et al., 2019; Stokes et al., 2015). This section focuses only on the large-scale aspects of those ice sheets, Greenland and Antarctic, that still exist today.

2.3.2.4.1 Greenland Ice Sheet

AR5 concluded the volume of the Greenland Ice Sheet (GrIS) was reduced compared to present during periods of the past few million years that were globally warmer than present (high confidence). It reported that the GrIS had lost ice during the prior two decades (very high confidence), that the ice loss had occurred in several sectors, and that high rates of mass loss had both expanded to higher elevations (high confidence) and very likely accelerated since 1992. SROCC concluded that it was extremely likely that ice loss increased through the early 21st century. SROCC also found that summer melting rate had increased since the 1990s to a rate unprecedented over the last 350 years (very high confidence), being two to five times greater than the pre-industrial rates (medium confidence).

Details of the history of GrIS fluctuations during warm interglacials continue to be elucidated. Oscillations over the past 7.5 Myr, including the Pliocene and through the glacial-interglacial cycles of the Pliocene are not well-constrained, but most studies indicate that Greenland was at least partially glaciated over this time with extended periods when it was predominantly deglaciated (Schaefer et al., 2016; Bierman et al., 2016). Geological evidence and modelling studies suggest periods of glacial intensification during the Pliocene at 4.9 Myr, 4.0 Myr, 3.6 Myr and 3.3 Myr (De Schepper et al., 2014; Bierman et al., 2016; Bachem et al., 2017). Retreat of the GrIS occurred during the MPWP and GrIS extent was reduced compared to today with some studies suggesting that the ice sheet was limited to the highest elevations (Blake-Mizen et al., 2019; De Schepper et al., 2014; Haywood et al., 2016; Koenig et al., 2015). There is apparent glacial intensification following the MPWP, 2.75–2.72 Myr (Nielsen and Kuijpers, 2013; De Schepper et al., 2014; Blake-Mizen et al., 2019; Knutz et al., 2019). Several studies agree that during the LIG the total GrIS extent was likely less than present day (Section 9.4.1, Figure 9.17) with the total mass loss ranging from 0.3–6.2 m sea level equivalent (SLE), although timing and magnitude of this mass loss are not well constrained (Clark et al., 2020; Goelzer et al., 2016; Helsen et al., 2013; Sinclair et al., 2016; Stone et al., 2013; Vasskog et al., 2015; Yau et al., 2016). During the LGM, the GrIS reached a peak ice volume greater than present (2–5 m SLE), as revealed by limited number of available geological records (Lecavalier et al., 2014; Simpson et al., 2009; Batchelor et al. 2019).

Recent studies of marine and lake sediments, glacier ice, and geomorphic features show that the GrIS retreated rapidly during the early Holocene but halted periodically, with a complex ice-margin chronology (Briner et al., 2016; Carlson et al., 2014; Larsen et al., 2014, 2015; Young et al., 2020; Young & Briner, 2015). It is probable that its total volume during 8–3 ka was smaller than today (Briner et al., 2016; Larsen et al., 2015; Young & Briner, 2015), but uncertainties exist regarding precisely when the minimum MH extent and volume was reached, due to uncertainties in reconstructions. The GrIS then re-advanced reaching its maximum extent in most places during 1450–1850 CE, although the timing and extent of this maximum differed by sector (Briner et al., 2016; Larsen et al., 2015).

Greenland-wide estimates of mass change based on direct observations were limited prior to 1992 at the time of AR5 (Kjeldsen et al., 2015). Combined records based on airborne observations, model-based estimates and geodetic approaches indicate an average mass loss of 75 ± 29.4 Gt yr⁻¹ for 1900–1983 (Kjeldsen et al., 2015). Integration of proxies and modelling indicates that the last time the rate of mass loss of GrIS was plausibly similar to 20th century rates was during the early Holocene (Briner et al., 2020; Buizert et al., 2018).

Since AR5, a combination of remote sensing, in situ observations and modelling has provided new insights regarding surface processes and their contribution to recent GrIS mass changes (AMAP, 2017; Bamber et al., 2018; IMBIE Consortium, 2020; Khan et al., 2020; Mouginot et al., 2019; van den Broeke et al., 2017).
Estimates of total ice loss during the post-1850 period (Kjeldsen et al., 2015) and recent observations show that the rate of loss has increased since the beginning of the 21st century (IMBIE Consortium, 2020; Sasgen et al., 2020; Velicogna et al., 2020) (Figure 2.24, Section 9.4.1.1 Figure 9.17).

The GrIS lost 4890 [4140–5640] Gt (SLE 13.5 [11.4–15.6] mm) of ice between 1992 and 2020 (IMBIE Consortium, 2020) (Figure 2.24, Section 9.4.1). The ice sheet was close to mass balance in the 1990s, but increases in mass loss have occurred since (Bamber et al., 2018; WCRP Global Sea Level Budget Group, 2018; Mouginot et al., 2019; IMBIE Consortium, 2020). The rate of ice sheet (including peripheral glaciers) mass loss rose from 120 [70 to 170] Gt yr\(^{-1}\) (SLE 0.33 [0.18 to 0.47] mm yr\(^{-1}\)) in 1901–1990 to 330 [290 to 370] Gt yr\(^{-1}\) (SLE 0.91 [0.79 to 1.02] mm yr\(^{-1}\)) for 2006–2018 (Section 9.4.1, Table 9.5).

[START FIGURE 2.24 HERE]

Figure 2.24: Cumulative Antarctic Ice Sheet (AIS) and Greenland Ice Sheet (GrIS) mass changes. Values shown are in gigatons and come from satellite-based measurements (IMBIE Consortium, 2018, 2020) for the period 1992–2018 for GrIS and 1992–2017 for AIS. The estimated uncertainties are very likely range, for the respective cumulative changes are shaded. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.24 HERE]

In summary, the GrIS was smaller than present during the MPWP (medium confidence), LIG (high confidence) and the MH (high confidence). GrIS mass loss began following a peak volume attained during the 1450–1850 period and the rate of loss has increased substantially since the turn of the 21st century (high confidence).

2.3.2.4.2 Antarctic Ice Sheet

AR5 reported that there was high confidence that the AIS was losing mass. The average ice mass loss from Antarctica was 97 [58 to 135] Gt yr\(^{-1}\) (GMSL equivalent of 0.27 [0.16 to 0.37] mm yr\(^{-1}\)) over 1993–2010, and 147 [74 to 221] Gt yr\(^{-1}\) (0.41 [0.20 to 0.61] mm yr\(^{-1}\)) over 2005–2010. These assessments included the Antarctic peripheral glaciers. AR5 reported with high confidence that the volume of the West Antarctic Ice Sheet (WAIS) was reduced during warm periods of the past few million years. SROCC concluded that over 2006–2015, the AIS lost mass at an average rate of 155 ± 19 Gt yr\(^{-1}\) (very high confidence). SROCC stated also that it is virtually certain that the Antarctic Peninsula and WAIS combined have cumulatively lost mass since widespread measurements began in 1992, and that the rate of loss has increased since around 2006.

Process understanding and, to some extent, paleoclimate records show that changes in parts of the AIS can occur over multi-century time scales (<2 kyr) (e.g. Dowdeswell et al., 2020; Sections 9.4.2.3 and 9.6.2). Based on physical understanding, paleo evidence and numerical simulations, it is very likely that the AIS has been smaller than today during at least some past warm climates (such as MCO and LIG), in particular the WAIS (de Boer et al., 2015; Deconto & Pollard, 2016; Golledge et al., 2014; Levy et al., 2016) (Figure 9.18). Results from sediment studies suggest a smaller AIS during the MPWP compared with current levels, with main differences in the WAIS (SROCC, Bertram et al., 2018; Shakun et al., 2018, Section 9.6.2). Marine sediments indicate that during the Pleistocene repeated ungrounding and loss of large marine-based parts of the AIS occurred during interglacial periods, with at least seven transitions between floating and grounded ice in the Ross Sea during the last 780 kyr (McKay et al., 2012) and at least three reductions in ice volume in the Wilkes Basin during the last 500 kyr (Wilson et al., 2018). Proxies, modelling and process understanding (Rohling et al., 2019; Clark et al., 2020) indicate that the AIS was smaller during the LIG than present.

Geological evidence has been used to reconstruct Holocene glacial fluctuations of the ice sheet margin and lowerings of its surface, which occurred at different times in different places, as recently reviewed by Noble et al. (2020). In West Antarctica, marine sediments below the ice sheet (Kingslake et al., 2018) corroborate a
Improved estimates of surface mass balance (SMB) in Antarctica from 67 ice core records do not show any substantial changes in accumulation rates over most of Antarctica since 1200 CE (Frezzotti et al., 2013). The SMB growth rate in Antarctica is estimated to be 7.0 ± 0.1 Gt per decade between 1800 and 2010 and 14.0 ± 1.8 Gt per decade since 1900 (Thomas et al., 2017). For the period 1979–2000, an insignificant Antarctic-wide negative SMB trend has been estimated (Medley and Thomas, 2019). The Antarctic Ice Sheet lost 2670 [1800 to 3540] Gt (SLE 7.4 [5.0 to 9.8] mm) of ice between 1992 and 2020. The rate of ice sheet (including peripheral glaciers) mass loss rose from 0 [–36 to 40] Gt yr⁻¹ (SLE 0.0 [–0.10 to 0.11] mm yr⁻¹) in 1991–1990 to 192 [145 to 239] Gt yr⁻¹ (SLE 0.54 [0.47 to 0.61] mm yr⁻¹) for 2006–2018 (Figure 2.24, Section 9.4.2, Table 9.5). Within quantified uncertainties, this estimate agrees with other recent estimates (Rignot et al., 2019; Smith et al., 2020; Velicogna et al., 2020). There is therefore very high confidence that the AIS has been losing mass over 1992–2020 (Figure 2.24, Section 9.4.2.1). Major contributions to recent AIS changes arise from West Antarctica and Wilkes Land in East Antarctica (Rignot et al., 2019). For the East Antarctic most studies suggest that the mass balance is not significantly different from zero (Bamber et al., 2018; IMBIE Consortium, 2018; Mohajerani et al., 2018; Rignot et al., 2019).

In summary, the AIS has lost mass between 1992 and 2020 (very high confidence), and there is medium confidence that this mass loss has increased. During the MPWP and LIG, the ice sheet was smaller than present (medium confidence). There is low confidence as to whether the total mass of the ice sheet was larger or smaller around 6 ka compared to now.

### 2.3.2.5 Terrestrial permafrost

AR5 concluded that in most regions and at most monitoring sites permafrost temperatures since the 1980s had increased (high confidence). Negligible change was observed at a few sites, mainly where permafrost temperatures were close to 0°C, with slight cooling at a limited number of sites. AR5 also noted positive trends in active layer thickness (ALT; the seasonally thawed layer above the permafrost) since the 1990s for many high latitude sites (medium confidence). SROCC concluded permafrost temperatures have increased to record high levels since the 1980s (very high confidence) with a recent increase by 0.29°C ± 0.12°C from 2007 to 2016 averaged across polar and high mountain regions globally.

Permafrost occurrence during the Pliocene has been inferred from pollen in lake sediments in NE Arctic Russia and permafrost-vegetation relationships which indicate that permafrost was absent during the MPWP in this region (Birgham, Grette et al., 2013; Herzschuh et al., 2016). Analysis of speleothem records in Siberian caves, indicates that permafrost was absent in the current continuous permafrost zone at 60°N at the start of the 1.5 Ma record, with degradation occurring around 0.4 Ma (Vaks et al., 2020). There are indications of extensive permafrost thaw during subsequent interglacials especially further south in the current permafrost zone (Vaks et al., 2013). Reconstruction of permafrost distribution during the LGM indicates that permafrost was more extensive in exposed areas (Vandenberghe et al., 2014). In non-glaciated areas of the North American Arctic there is permafrost that survived the LIG (French and Millar, 2014).

Trends and timing of permafrost degradation and thaw over the last 6 kyr in peatlands of the NH were recently summarized (Hiemstra, 2018; Treat and Jones, 2018a). Three multi-century periods (ending 1000 Before the Common Era (BCE), 500 CE and 1850 CE) of permafrost degradation, associated with neoglacial periods are inferred resulting in more extensive permafrost in peatlands of the present-day discontinuous permafrost zone, which reached a peak approximately 250 years ago, with thawing occurring...
concurrently with post 1850 warming (Treat and Jones, 2018b). Although permafrost persists in peatlands at
the southern extent of the permafrost zone where it was absent prior to 3 ka, there has been thawing since the
1960s (Holloway & Lewkowicz, 2020; James et al., 2013; Jones et al., 2016).

Records of permafrost temperature measured in several boreholes located throughout the northern polar
regions indicate general warming of permafrost over the last 3–4 decades (Figure 2.25), with marked
regional variations (Romanovsky et al., 2017a,b; 2020; Biskaborn et al., 2019). Recent (2018–2019)
permafrost temperatures in the upper 20–30 m layer (at depths where seasonal variation is minimal) were the
highest ever directly observed at most sites (Romanovsky et al., 2020), with temperatures in colder
permafrost of northern North America being more than 1°C higher than they were in 1978. Increases in
temperature of colder Arctic permafrost are larger (average 0.4°C–0.6°C per decade) than for warmer
(temperature >–2°C) permafrost (average 0.17°C per decade) of sub-Arctic regions (Figures 2.25, 9.22).

[START FIGURE 2.25 HERE]

Figure 2.25: Changes in permafrost temperature. Average departures of permafrost temperature (measured in the
upper 20–30 m) from a baseline established during International Polar Year (2007–2009) for Arctic
regions. Further details on data sources and processing are available in the chapter data table (Table
2.SM.1).

[END FIGURE 2.25 HERE]

Increases in permafrost temperature over the last 10–30 years of up to 0.3°C per decade have been
documented at depths of about 20 m in high elevation regions in the NH (European Alps, the Tibetan Plateau
and some other high elevation areas in Asia; Biskaborn et al., 2019; Cao et al., 2018; Liu G. et al., 2017;
Noetzli et al., 2020; Zhao et al., 2020). In Antarctica, where records are limited and short (most < 10 years)
trends are less evident (Noetzli et al. 2019).

Assessment of trends in ALT is complicated by considerable ALT interannual variability. For example, in
north-western North America during the extreme warm year of 1998, ALT was greater than in prior years.
Although ALT decreased over the following few years, it has generally increased again since the late 2000s
(Duchesne, et al. 2015; Romanovsky et al., 2017b; 2020). However, at some sites there has been little change
in ALT due to ground subsidence that accompanies thaw of ice-rich permafrost (Streletskiy et al., 2017;
O’Neill et al., 2019). In the European and Russian Arctic there has been a broad-scale increase in ALT
during the 21st century (Streletskiy et al., 2015; Romanovsky et al., 2020). In high elevation areas in Europe
and Asia, increases in ALT have occurred since the mid-1990s (Cao et al., 2018; Liu, Y. et al., 2017;
Noetzli et al., 2019, 2020; Zhao et al., 2020). Limited and shorter records for Antarctica show marked interannual
variability and no apparent trend with ALT being relatively stable or decreasing at some sites since 2006
(Hrbáček et al., 2018).

Observations of ground subsidence and other landscape change (e.g. thermokarst, slope instability) since the
middle of the 20th century in the Arctic associated with ground ice melting have been documented in several
studies and provide additional indications of thawing permafrost (Borge et al., 2017; Derksen et al., 2019;
Farquharson et al., 2019; Kokelj et al., 2017; Lewkowicz & Way, 2019; Liljedahl et al., 2016; Nitze et al.,
2017; O’Neill et al., 2019; Séjourné et al., 2015; Streletskiy et al., 2017; Section 9.5.2.1). In mountain areas,
destabilization and acceleration of rock glacier complexes that may be associated with warming permafrost
have also been observed (Eriksen et al., 2018; Marcer et al., 2019).

In summary, increases in permafrost temperatures in the upper 30 m have been observed since the start of
observational programs over the past three to four decades throughout the permafrost regions (high
confidence). Limited evidence suggests that permafrost was less extensive during the MPWP (low
confidence). Permafrost that formed after 3ka still persists in areas of the NH, but there are indications of
thaw after the mid-1800s (medium confidence).
2.3.3 Ocean

This section focuses on large-scale changes in a subset of physical components of the ocean (Cross-Chapter Box 2.2). Chapter 7 assesses the role of the ocean in Earth system heating and evaluates the Earth’s energy budget. Chapter 9 undertakes a holistic assessment of changes in the ocean integrating observations, modelling and theoretical understanding. Chapter 11 assesses extremes such as marine heat waves and storm surges. SSTs are assessed in Section 2.3.1.1 as they constitute a critical component of GMST estimation.

2.3.3.1 Ocean temperature, heat content and thermal expansion

AR5 assessed that since 1971, global ocean warming was virtually certain for the upper 700 m and likely for the 700–2000 m layer. SROCC reported linear warming trends for the 0–700 m and 700–2000 m layers of the ocean of 4.35 ± 0.8 and 2.25 ± 0.64 ZJ yr⁻¹ over 1970–2017; 6.28 ± 0.48 and 3.86 ± 2.09 ZJ yr⁻¹ over 1993–2017; and 5.31 ± 0.48 and 4.02 ± 0.97 ZJ yr⁻¹ over 2005–2017. Both AR5 and SROCC assessed that the ocean below 2000 m had likely warmed since 1992. SROCC reported global mean thermosteric sea level (ThSL) rise, associated with thermal expansion of the ocean, with a trend of 0.89 ± 0.05 mm yr⁻¹ for 1970–2015; 1.36 ± 0.40 mm yr⁻¹ for 1993–2015; and 1.40 ± 0.40 mm yr⁻¹ for 2006–2015, and also reported that the rate of ocean warming over 1993–2017 had likely more than doubled since 1969–1992.

New ocean heat content (OHC) reconstructions derived from paleo proxies (Baggenstos et al., 2019; Bereiter et al., 2018; Gebbie, 2021; Shackleton et al., 2019) indicate that the global ocean warmed by 2.57°C ± 0.24°C, at an average rate of ~0.3°C ka⁻¹ (equivalent to an OHC change rate of 1.3 ZJ yr⁻¹) from the LGM (~20 ka) to the early Holocene (~10 ka) (Figure 9.9, Section 9.2.2.1). Over the LDT, ocean warming occurred in two stages, offset by some heat loss during the Antarctic Cold Reversal (14.58–12.75 ka). Only during a short period of rapid warming at the end of the Younger Dryas (12.75–11.55 ka) were rates comparable to those observed since the 1970s (Bereiter et al., 2018; Shackleton et al., 2019). Ice cores imply a small decrease in the global mean ocean temperature during the early Holocene (<0.4°C) (Baggenstos et al., 2019; Bereiter et al., 2018). Sediment cores from the equatorial Pacific and Atlantic Ocean (0–1000 m) indicate a stronger regional cooling (compared to mean ocean temperature) of 1.0°C ± 0.7°C to 1.8°C ± 0.4°C from the early/mid-Holocene to ca. 1750 CE (Kalanovsky et al., 2015; Morley et al., 2014; Rosenthal et al., 2017, 2013). Sediment cores from the western equatorial Pacific suggest 0.8°C ± 0.1°C higher temperatures in the upper 700 m of the ocean during 950–1100 CE compared to 1400–1750 CE. These changes are consistent with a global estimate derived from combined surface and subsurface ocean temperature proxy records (McGregor et al., 2015; PAGES 2k Consortium, 2013). A combined study of model and observational data further confirmed these results, treating temperature as a passive tracer (Gebbie & Huybers, 2019) and addressing the role of circulation dynamics (Scheen and Stocker, 2020). Collectively, the proxy records indicate a global OHC decrease of ~400 ± 70 ZJ (~170 ± 100 ZJ in the Pacific) in the upper 700 m between 950–1100 CE and 1400–1750 CE, and also suggest that the deep Pacific is still adjusting to this cooling (Rosenthal et al., 2013), partially offsetting the global increase since 1750 CE (Gebbie & Huybers, 2019; Gebbie, 2021).

For the instrumental era, since AR5 and SROCC, new and updated OHC (and ThSL) observation-based analyses (Johnson et al., 2020; von Schuckmann et al., 2020) enhance an existing large ensemble of direct and indirect OHC estimates (Figure 2.26), although some rely to varying degrees upon information from ocean-climate models. Direct estimates benefit from improved: bias adjustments (e.g. Bagnell & DeVries, 2020; Cheng et al., 2018; Gouretski & Cheng, 2020; Leahy et al., 2018; Palmer et al., 2018; Ribeiro et al., 2018; Wang et al., 2018); interpolation methods (Kuusela & Stein, 2018; Su et al., 2020); and characterization of sources of uncertainty (e.g. Allison et al., 2019; Garry et al., 2019; Good, 2017; Meyssignac et al., 2019; Palmer et al., 2021; Wunsch, 2018), including those originating from forced and intrinsic ocean variability (Penduff et al., 2018). After 2006 direct OHC estimates for the upper 2000 m layer benefit from the near-global ARGO array with its superior coverage over 60°S–60°N (Roemmich et al., 2019). Indirect estimates include OHC and ThSL series inferred from satellite altimetry and gravimetry since 2003 (Meyssignac et al., 2019), the passive uptake of OHC (ThSL) at centennial timescales inferred from observed SST anomalies, and time-invariant circulation processes from an ocean state estimation (e.g. Zanna...
et al., 2019). Resplandy et al. (2019) estimate the rate of global OHC uptake over 1991–2016 from changes in atmospheric composition and physical relationships based on CMIP5 model simulations. The uncertainties are broader than from direct estimates but the estimate is qualitatively consistent.

Collectively, the new and updated analyses strengthen AR5 and SROCC findings of a sustained increase in global OHC (Figure 2.26, Table 2.7) and associated ThSL rise. Larger warming rates are observed in the upper 700 m compared to deeper layers, with more areas exhibiting significant warming than significant cooling (Johnson & Lyman, 2020). There is an improved consistency among available estimates of OHC rates in the upper 2000 m since 2006. Cheng et al. (2020), von Schuckmann et al. (2020) and Johnson et al. (2020) have further confirmed that the central estimates of rates of OHC change in the upper 2000 m depths have increased after 1993 and particularly since 2010 (Section 3.5.1.3, Figures 2.26, 3.26), although uncertainties are large (Table 2.7). Ocean reanalyses support findings of continued upper ocean warming (Balmaseda et al., 2013; Meyssignac et al., 2019; von Schuckmann et al., 2018), albeit with higher spread than solely observational estimates, particularly in the poorly sampled deep ocean below 2000 m (Palmer et al., 2018; Storto et al., 2017).

In summary, current multidecadal to centennial rates of OHC gain are greater than at any point since the last deglaciation (medium confidence). At multi-centennial timescales, changes in OHC based upon proxy indicators demonstrate a tight link with surface temperature changes during the last deglaciation (high confidence), as well as during the Holocene and CE (low confidence). It is likely the global ocean has warmed since 1871, consistent with the observed increase in sea surface temperature. It is virtually certain that OHC increased between 1971 and 2018 in the upper 700 m and very likely in the 700–2000 m layer, with high confidence since 2006. It is likely the OHC below 2000 m has increased since 1992. Confidence in the assessment of multidecadal OHC increase is further strengthened by consistent closure of both global sea level and energy budgets (Section 7.2.2.2, Box 7.2, Cross-Chapter Box 9.1).

[START FIGURE 2.26 HERE]

Figure 2.26: Changes in Ocean heat content (OHC). Changes are shown over (a) full depth of the ocean from 1871 to 2019 from a selection of indirect and direct measurement methods. The series from Table 2.7 is shown in solid black in both (a) and (b) (see Table 2.7 caption for details). (b) as (a) but for 0–2000 m depths only and reflecting the broad range of available estimates over this period all of which are broadly similar. For further details see chapter data table (Table 2.SM.1).

[END FIGURE 2.26 HERE]

[START TABLE 2.7 HERE]

Table 2.7: Rates of global ocean heat content (OHC) and global mean thermosteric sea level ThSL change for four depth integrations over different periods. For the period up to 1971, the assessment for all depth layers is based on Zanna et al (2019). From 1971 onwards, consistent with AR5, Domingues et al. (2008, updated) is the central estimate for 0–700 m along with uncertainty from a 5-member ensemble (Cheng et al., 2017; Domingues et al., 2008, updated; Good et al., 2013; Ishii et al., 2017; Levitus et al., 2012), following the approach of Palmer et al (2021). Similarly, Ishii et al. (2017) is the central estimate for 700–2000 m with uncertainty based on a 3-member ensemble (Cheng et al., 2017; Ishii et al., 2017; Levitus et al., 2012). For depths below 2000 m, both central estimate and uncertainty are from Purkey and Johnson (2010, updated). In cases when OHC estimates do not have a ThSL counterpart (e.g. Cheng et al., 2017; Good et al., 2013), OHC was converted into ThSL using the average linear regression coefficients for 0–700 m and 700–2000 m from all available ensemble members. For consistency with the energy and sea-level budgets presented in Chapters 7 and 9, reported rates are based on the difference between the first and last annual mean value in each period (Palmer et al., 2021, Box 7.2, Cross-Chapter Box 9.1). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Prior to the instrumental record, reconstructions of near-surface salinity change are accomplished by combining isotopic and elemental proxy data from microfossil plankton shells and skeletons preserved in deep-sea sediments. These data highlight changes in the salinity contrast between the Pacific and Atlantic oceans during past ice ages (Broecker, 1989; Keigwin and Cook, 2007; Costa et al., 2018) and for repeated

<table>
<thead>
<tr>
<th>Depth</th>
<th>Period</th>
<th>OHC rate (ZJ yr⁻¹)</th>
<th>ThSL rate (mm yr⁻¹)</th>
<th>Relative full ocean depth contribution</th>
</tr>
</thead>
<tbody>
<tr>
<td>0 to 700 m</td>
<td>1901-1990</td>
<td>2.50 [1.16–3.85]</td>
<td>0.31 [0.16–0.45]</td>
<td>81% 86%</td>
</tr>
<tr>
<td></td>
<td>1901-2018</td>
<td>3.11 [2.18–4.04]</td>
<td>0.40 [0.30–0.50]</td>
<td>66% 74%</td>
</tr>
<tr>
<td></td>
<td>1971-2018</td>
<td>5.14 [3.46–6.82]</td>
<td>0.71 [0.51–0.90]</td>
<td>61% 70%</td>
</tr>
<tr>
<td></td>
<td>1993-2018</td>
<td>6.06 [4.56–7.55]</td>
<td>0.89 [0.69–1.10]</td>
<td>58% 68%</td>
</tr>
<tr>
<td></td>
<td>2006-2018</td>
<td>6.28 [4.06–8.50]</td>
<td>0.91 [0.51–1.31]</td>
<td>54% 65%</td>
</tr>
<tr>
<td>700 to 2000 m</td>
<td>1901-1990</td>
<td>0.50 [−0.59–1.60]</td>
<td>0.04 [−0.07–0.16]</td>
<td>16% 11%</td>
</tr>
<tr>
<td></td>
<td>1901-2018</td>
<td>1.26 [0.43–2.09]</td>
<td>0.11 [0.02–0.19]</td>
<td>27% 20%</td>
</tr>
<tr>
<td></td>
<td>1971-2018</td>
<td>2.62 [2.04–3.20]</td>
<td>0.23 [0.16–0.31]</td>
<td>31% 23%</td>
</tr>
<tr>
<td></td>
<td>1993-2018</td>
<td>3.31 [2.40–4.22]</td>
<td>0.30 [0.19–0.41]</td>
<td>32% 23%</td>
</tr>
<tr>
<td></td>
<td>2006-2018</td>
<td>4.14 [2.41–5.86]</td>
<td>0.36 [0.15–0.58]</td>
<td>36% 26%</td>
</tr>
<tr>
<td>&gt;2000 m</td>
<td>1901-1990</td>
<td>0.07 [0.02–0.12]</td>
<td>0.01 [0.00–0.01]</td>
<td>2% 3%</td>
</tr>
<tr>
<td></td>
<td>1901-2018</td>
<td>0.32 [0.18–0.46]</td>
<td>0.03 [0.02–0.05]</td>
<td>7% 6%</td>
</tr>
<tr>
<td></td>
<td>1971-2018</td>
<td>0.66 [0.33–0.99]</td>
<td>0.07 [0.03–0.10]</td>
<td>8% 7%</td>
</tr>
<tr>
<td></td>
<td>1993-2018</td>
<td>1.15 [0.58–1.72]</td>
<td>0.12 [0.06–0.18]</td>
<td>11% 9%</td>
</tr>
<tr>
<td></td>
<td>2006-2018</td>
<td>1.15 [0.58–1.72]</td>
<td>0.12 [0.06–0.18]</td>
<td>10% 9%</td>
</tr>
<tr>
<td>Full-depth</td>
<td>1901-1990</td>
<td>3.08 [1.36–4.79]</td>
<td>0.36 [0.17–0.54]</td>
<td></td>
</tr>
<tr>
<td></td>
<td>1901-2018</td>
<td>4.68 [3.45–5.92]</td>
<td>0.54 [0.40–0.68]</td>
<td></td>
</tr>
<tr>
<td></td>
<td>1971-2018</td>
<td>8.42 [6.08–10.77]</td>
<td>1.01 [0.73–1.29]</td>
<td></td>
</tr>
<tr>
<td></td>
<td>1993-2018</td>
<td>10.52 [7.26–13.28]</td>
<td>1.31 [0.95–1.66]</td>
<td></td>
</tr>
<tr>
<td></td>
<td>2006-2018</td>
<td>11.57 [7.20–15.94]</td>
<td>1.39 [0.74–2.05]</td>
<td></td>
</tr>
</tbody>
</table>

2.3.3.2 Ocean salinity

AR5 concluded that subtropical regions of high salinity (where evaporation dominates over precipitation) had become more saline, while regions of low salinity (mostly in the tropics and high latitudes) had very likely become fresher since the 1950s, both at the near-surface, and in the ocean interior along ventilation pathways. From 1950 to 2008, the mean surface contrast between high- and low-salinity regions increased by 0.13 [0.08 to 0.17] (PSS-78, Unesco, 1981). Across basins, the Atlantic Ocean had become saltier and the Pacific and Southern Oceans had freshened (very likely).
episodes of increased subtropical salinity (Schmidt et al., 2004, 2006) and subpolar freshening (Cortijo et al., 1997; Thornalley et al., 2011) in the North Atlantic ocean. These episodes were associated with disruptions to the large-scale deep ocean circulation (Buizert et al., 2015; Henry et al., 2016; Lynch-Stieglitz, 2017). Further quantification of paleo salinity changes is complicated by incomplete understanding on how proxy-salinity relationships and the relative influence of atmospheric and ocean processes vary across regions and paleo periods (Conroy et al., 2017; Holloway et al., 2016; LeGrande & Schmidt, 2011; Rohling, 2007).

Since AR5, new and extended multidecadal analyses have strengthened the observational support for increased contrast between high and low near-surface salinity regions and inter-basin contrast since the mid-20th century (Section 9.2.2.2; Aretxabaleta et al., 2017; Cheng et al., 2020; Durack & Wijffels, 2010; Good et al., 2013; Skliris et al., 2014). These analyses employ different statistical algorithms for interpolation, and only Cheng et al., (2020) use CMIP5 model simulations to constrain observation-based signals in data-sparse regions.

The 1950–2019 trends reveal near-surface freshening of the northern and western Warm (and fresh) Pool of the Pacific and increased salinity maxima in the subtropical Atlantic, strengthening the inter-basin contrast (Figure 2.27a). There are indications that the subpolar freshening and subtropical salinification of the Atlantic ocean may extend back to at least 1896 (Friedman et al., 2017). Over recent decades, new observations from Argo floats and ocean reanalyses provide general support that changes in the global patterns of near-surface salinity contrast are broadly associated with an intensification of the hydrological cycle (Sections 2.3.1.3.5; 8.3.1.1). However, this assessment is complicated by changing observational techniques (Section 1.5.1), temporally and spatially inhomogeneous sampling and uncertainties in interpolation algorithms and the substantial influence of modes of natural variability and ocean circulation processes over interannual timescales (Aretxabaleta et al., 2017; Durack, 2015; Grist et al., 2016; Liu et al., 2020; Skliris et al., 2014; Vinogradova & Ponte, 2017). Following AR5, based on the updated analysis from Durack and Wijffels (2010) which infills in situ gaps to recover large-scale patterns the mean salinity contrast between high- and low- near-surface salinity regions increased by 0.14 [0.07 to 0.20] from 1950 to 2019.

Changes in the global patterns of near-surface salinity contrast are transferred to the ocean interior via ventilation pathways (Figure 2.27b). Large scale similarities in subsurface salinity changes across observational estimates point to decreasing (increasing) salinity in regions where salinity is lower (higher) than the global average, with freshening in subpolar regions and salinification in the subtropical gyres (Aretxabaleta et al., 2017; Cheng et al., 2020; Durack, 2015; Durack & Wijffels, 2010; Good et al., 2013; Skliris et al., 2014). Regional changes in salinity are assessed in Section 9.2.2.2.

In summary, it is virtually certain that since 1950 near-surface high salinity regions have become more saline, while low salinity regions have become fresher, and it is very likely that this extends to the ocean interior along ventilation pathways. Across basins, it is very likely that the Atlantic has become saltier and the Pacific and Southern oceans have freshened. The differences between high-salinity and low-salinity regions are linked to an intensification of the hydrological cycle (medium confidence).

[START FIGURE 2.27 HERE]

Figure 2.27: Changes in ocean salinity. Estimates of salinity trends using a total least absolute differences fitting method for (a) global near-surface salinity (SSS) changes and (b) global zonal mean subsurface salinity changes. Black contours show the associated climatological mean salinity (either near-surface (a) or subsurface (b)) for the analysis period (1950–2019). Both panels represent changes of Practical Salinity Scale 1978 [PSS-78], per decade. In both panels green denotes freshening regions and orange/brown denotes regions with enhanced salinities. ‘x’ marks denote non-significant changes. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.27 HERE]

2.3.3.3 Sea level

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AR5 concluded based on proxy and instrumental data that the rate of global mean sea level (GMSL) rise since the mid-19th century was larger than the mean rate during the previous two millennia (high confidence). SROCC reported with high confidence that GMSL increases were 1.5 [1.1 to 1.9] mm yr\(^{-1}\) for 1902–2010 (with an acceleration rate between -0.002–0.019 mm yr\(^{-2}\)). 2.1 [1.8 to 2.3] mm yr\(^{-1}\) for 1970–2015, 3.2 [2.8 to 3.5] mm yr\(^{-1}\) for 1993–2015 and 3.6 [3.1 to 4.1] mm yr\(^{-1}\) for 2006–2015. AR5 reported that GMSL during the LIG, over several thousand years, between 5 and 10 m higher than 1985–2004 (medium confidence) whereas SROCC concluded it was virtually certain that GMSL exceeded current levels (high confidence), and reached a peak that was likely 6–9 m higher than today, but did not exceed 10 m (medium confidence). AR5 concluded with high confidence that there were two intra-LIG GMSL peaks and that the millennial-scale rate during these periods exceeded 2 mm yr\(^{-1}\). AR5 had high confidence that GMSL during the MPWP did not exceed 20 m above present. Based on new understanding, SROCC placed the upper bound at 25 m but with low confidence.

The Earth was largely ice free during the EECO (Cramer et al., 2011; Miller et al., 2020, Section 9.6.2), and complete loss of current land ice reservoirs would raise GMSL by 65.6 ± 1.8 m (Farinotti et al., 2019; Morlighem et al., 2017; 2020). Given that GMSL change must be due to some combination of transient land ice growth and changes in terrestrial water storage, additional global mean thermosteric sea level increase of 7 ± 2 m (Fischer et al., 2018) implies a peak EECO GMSL of 70–76 m (low confidence). Changes in ocean basin size driven by plate tectonics contributed a comparable amount to global mean geocentric sea level in the Eocene, but are definitionally excluded from GMSL assessment (Wright et al., 2020).

For the MPWP, several studies of coastal features have provided additional quantitative sea-level estimates of: 5.6–19.2 m from Spain (Dumitruc et al., 2019), approximately 14 m from South Africa (Hearty et al., 2020), 15 m from the United States (Moucha and Ruetenik, 2017), and 25 m from New Zealand (Grant et al., 2019). Thus, consistent with SROCC, GMSL during the MPWP was higher than present by 5–25 m (medium confidence).

Reconstructions of GMSL from marine oxygen isotopes in foraminifera shells show variations of more than 100 m over intervals of 10–100 kyr during glacial-interglacial cycles of the Quaternary (McManus et al., 1999; Miller et al., 2020; Shackleton, 1987; Waelbroeck et al., 2002). Correction for past temperatures and a calibration for ice-volume changes implies uncertainty estimates of ±10–13 m (1 SD) (Grant et al., 2014; Shakun et al., 2015; Spratt & Lisiecki, 2016). A recent marine oxygen-isotope-based GMSL reconstruction (Spratt and Lisiecki, 2016) agrees with previous reconstructions, while focusing on the past 800 kyr (Figure 2.28). It shows that GMSL during the Holocene was among the highest over this entire interval, and was surpassed only during the LIG (Marine Isotope Stage (MIS 5e)) and MIS 11 (medium confidence); however, relatively brief (about 2 kyr) highstands during other interglacial periods might be obscured by dating limitations.

Few sites globally have well-preserved MIS 11 sea-level indicators (Dutton et al., 2015). As reported in AR5, Raymo and Mitrovica (2012) used glacial isostatic adjustment models to correct the elevation of MIS 11 sea-level proxies from Bermuda and Bahamas to estimate a peak MIS 11 GMSL between 6 and 13 m above present-day. This agrees with the elevation of 13 ± 2 m for the MIS 11 subtidal–intertidal transition in South Africa (Roberts et al., 2012). A revised glacial isostatic adjustment at this location resulted in a peak GMSL estimate of 8–11.5 m (Chen et al., 2014). In light of these data, and the review by Dutton et al. (2015), the AR5 estimate of 6–13 m for MIS 11 remains the best available (medium confidence).

Recent studies have highlighted uncertainties in estimates of GMSL during the LIG, including the extent of GMSL variability (Capron et al., 2019). Vertical land motions (Austermann et al., 2017) are starting to be considered quantitatively (e.g., Stephenson et al., 2019), but are still bounded by large uncertainties. The distribution and thickness of pre-LIG ice sheets (Dendy et al., 2017; Rohling et al., 2017) and isostasy driven by sediment loading since the LIG (Pico, 2020) add further uncertainty. In light of these recent studies and previous assessments, there is medium confidence that peak GMSL during the LIG was likely between 5 and 10 m higher than modern. Relative sea-level estimates from some sites (e.g., Bahamas and Seychelles) report ephemeral, metre-scale fluctuations (Vyverberg et al., 2018). Different generations of LIG reef growth at other sites (e.g., Yucatan Peninsula, Western Australia) suggest the occurrence of sudden accelerations in
GMSL change (Blanchon et al., 2009; O’Leary et al., 2013). However, other sites (e.g., South Australia, Mediterranean), indicate that LIG sea level was substantially stable (Pan, Y-T. et al., 2018; Polyak et al., 2018). In addition, there are uncertainties in the interpretation of local relative sea level from some GMSL reconstructions (Barlow et al., 2018). Therefore, low confidence is assigned to any GMSL rate of change estimated within the LIG.

New geological proxies and glacial isostatic adjustment (GIA) modelling studies confirm that, at the LGM, GMSL was 125–134 m below present (Lambeck et al., 2014; Yokoyama et al., 2018). During the LDT, GMSL rose from approximately -120 m to -50 m, implying an average rate of about 10 mm yr\(^{-1}\) (Lambeck et al., 2014). The fastest rise occurred during Meltwater Pulse 1A, at about 14.6–14.3 ka (Deschamps et al., 2012; Sunborn et al., 2017), when GMSL rose by between 8 m and 15 m (medium confidence) (Liu J. et al., 2016) at an average rate of 24–44 mm yr\(^{-1}\).

Recent GIA modelling studies tuned to both near- and far-field relative sea level (RSL) data yield MH GMSL estimates of -3.8 to -1.0 m (Lambeck et al., 2014; Peltier, 2017). Estimates from relatively stable locations where the effects of GIA are small and relatively insensitive to parameters defining Earth rheology, and where RSL is expected to approximate GMSL to within about 1 m (e.g., Milne & Mitrovica, 2008), suggest that RSL was between about -6 to 1.5 m at around 6 ka at multiple locations (Braithwaite et al., 2000; Camoin et al., 1997; Frank et al., 2006; Hibbert et al., 2018; Khan et al., 2017; Montaggioni & Faure, 2008; Vacchi et al., 2016). The assessment of GMSL change at 6 ky is challenging considering the proportionately large GIA effect (Kopp et al., 2016a). Insufficient resolution of marine geochemical proxies (\(\delta^{18}O\), Mg/Ca) and uncertainties in the contribution of the Antarctic Ice Sheet during the MH (Section 2.3.2.4). The possibility that GMSL was at least somewhat higher than present cannot be excluded.

For the last 3 kyr, GMSL has been estimated from global databases of sea-level proxies, including numerous densely-sampled high-resolution salt-marsh records with decimetre-scale vertical resolution and sub-centennial temporal resolution (Kopp et al., 2016b; Kemp et al., 2018a). Over the last about 1.5 kyr, the most prominent century-scale GMSL trends include average maximum rates of lowering and rising of -0.7 ± 0.5 mm yr\(^{-1}\) (2 SD) over 1020–1120 CE, and 0.3 ± 0.5 (2 SD) over 1460–1560, respectively. Between 1000 and 1750 CE, GMSL is estimated to have been within the range of about -0.11–0.09 m relative to 1900 (Kemp et al., 2018b). This was followed by a sustained increase of GMSL that began between 1820 and 1860 and has continued to the present day. New analyses demonstrate that it is very likely that GMSL rise over the 20th century was faster than over any preceding century in at least 3 kyr (Kopp et al., 2016a; Kemp et al., 2018a) (Figure 2.28).

[START FIGURE 2.28 HERE]

Figure 2.28: Changes in global mean sea level. (a) Reconstruction of sea-level from ice core oxygen isotope analysis for the last 800 kyr. For target paleo periods (CCB2.1) and MIS11 the estimates based upon a broader range of sources are given as box whiskers. Note the much broader axis range (200 m) than for later panels (tenths of metres). (b) Reconstructions for the last 2500 years based upon a range of proxy sources with direct instrumental records superposed since the late 19th century. (c) Tide-gauge and, more latterly, altimeter-based estimates since 1850. The consensus estimate used in various calculations in Chapters 7 and 9 is shown in black. (d) The most recent period of record from tide-gauge and altimeter-based records. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.28 HERE]

Since SROCC, two new tide gauge reconstructions of 20th century GMSL change have been published, although both rely upon CMIP models to varying degrees (Figure 2.28). Frederikse et al (2020) used a “virtual station” method and a probabilistic framework to estimate GMSL change and its uncertainties since 1900. Dangendorf et al (2019) combined a Kalman Smoother (Hay et al., 2015) with Reduced Space
Optimal Interpolation (Church & White, 2011; Ray & Douglas, 2011) in an effort to better represent both the long-term GMSL change while preserving information on sea-level variability. In addition, new ensemble-based methods for quantifying GMSL change have been presented that account for both structural and parametric uncertainty (Palmer et al., 2021). Altimeter timeseries of GMSL change (Figure 2.28) have been extended to 2019/2020 but bias adjustments (Ablain et al., 2019; Beckley et al., 2017; Dieng et al., 2017; Legeais et al., 2020; Watson et al., 2015) did not change since SROCC.

Based on the ensemble approach of Palmer et al. (2021) and an updated WCRP (2018) assessment (Figure 2.28) GMSL rose at a rate of 1.35 [0.78 to 1.92] mm yr\(^{-1}\) for the period 1901–1990 and 3.25 [2.88 to 3.61] mm yr\(^{-1}\) for 1993–2018 (high confidence). The average rate for 1901–2018 was 1.73 [1.28 to 2.17] mm yr\(^{-1}\) with a total rise of 0.20 [0.15 to 0.25] m (Table 9.5). The acceleration rate very likely is 0.094 [0.082 to 0.115] mm yr\(^{-2}\) for 1993–2018 (WCRP, 2018, updated), consistent with other estimates (Ablain et al., 2019; Chen X. et al., 2017; Legeais et al. 2020; Nerem et al., 2018; Watson et al., 2015; WCRP, 2018). For the period 1902–2010 the updated tide gauge reconstructions published since the SROCC also show a robust acceleration over the 20th century and the ensemble estimate of Palmer et al (2021) gives a value of 0.0053 [0.0042 to 0.0073] mm yr\(^{-2}\), based on an unweighted quadratic fit.

In summary, GMSL is rising, and the rate of GMSL rise since the 20th century is faster than over any preceding century in at least the last three millennia (high confidence). Since 1901, GMSL has risen by 0.20 [0.15 to 0.25] m at an accelerating rate. Further back in time, there is medium confidence that GMSL was within -3.5 to -0.5 m (very likely range) of present during the MH, 5–10 m higher (likely range) during the LIG, and 5–25 m higher (very likely range) during the MPWP.

### 2.3.3.4 Ocean circulation

#### 2.3.3.4.1 Atlantic Meridional Overturning circulation (AMOC)

AR5 concluded that there was no evidence of a trend in the AMOC during the period of instrumental observations. However, AR5 also stressed insufficient evidence to support a finding of change in the heat transport of the AMOC. SROCC assessed that there was emerging evidence in sustained observations, both in situ (2004–2017) and revealed from SST-based reconstructions, that the AMOC had weakened during the instrumental era relative to 1850–1900 (medium confidence), although there were insufficient data to quantify the magnitude of the weakening. SROCC also concluded with low confidence an increase of the Southern Ocean upper cell overturning circulation. SROCC also reported with medium confidence that the production of Antarctic Bottom Water had decreased since the 1950s consistent with a decreased lower cell overturning circulation, and potentially modulating the strength of the AMOC.

On multi-millennial timescales, proxy evidence indicates that the AMOC varied repeatedly in strength and vertical structure. During the last glacial period, particularly around the LGM, AMOC was estimated to be shallower than present, although there is continued debate about the magnitude of the shoaling (Gebbie, 2014; Lynch-Stieglitz et al., 2007), and whether this change was associated with a weaker overturning (Ritz et al., 2013; Menviel et al., 2017; Muglia et al., 2018). There are indications that substantial variations in AMOC were associated with abrupt climate changes during the glacial intervals, including Dansgaard-Oeschger and Heinrich events (14–70 ka) (Böhm et al., 2015; Henry et al., 2016; Lynch-Stieglitz, 2017; McManus et al., 2004). During these millennial-scale oscillations, weakened AMOC was associated with dramatic cooling in the northern hemisphere and warming in the southern hemisphere (Buizert et al., 2015; Henry et al., 2016), while hemispheric changes of opposite sign accompanied strengthened AMOC. After the final demise of the Laurentide ice sheet about 8 ka, the mean overall strength of AMOC has been relatively stable throughout the Holocene compared to the preceding 100 kyr (Hoffmann et al., 2018; Lippold et al., 2019). There are however indications of episodic variations in AMOC during the Holocene (Bianchi and McCave, 1999; Oppo et al., 2003; Thormanley et al., 2013; Ayache et al., 2018), and past interglacial intervals (Galaaen et al., 2014, 2020; Hayes et al., 2014; Huang et al., 2020; Mokeddem et al., 2014). Over the last 3 kyr, there are indications that AMOC variability was potentially linked to decreasing production of Labrador Sea Water (LSW), one of the water masses contributing to AMOC (Alonso-Garcia et al., 2017; McClymont et al., 2020; Moffa-Sánchez et al., 2019; Moffa-Sánchez & Hall, 2017).
Numerous proxy records collectively imply that AMOC is currently at its weakest point in the past 1.6 ka (Caesar et al., 2018, 2021; Rahmstorf et al., 2015; Thibodeau et al., 2018; Thornalley et al., 2018). Caesar et al. (2021) analyse a compilation of various available indirect AMOC proxies from marine sediments, in situ-based reconstructions and terrestrial proxies, which show a decline beginning in the late 19th century and over the 20th century superimposed by large decadal variability in the second half of the 20th century. Indirect reconstructions of AMOC components based on coastal sea level records in the western North Atlantic (Ezer, 2013; McCarthy et al., 2015; Piecuch, 2020) show an AMOC decline since the late 1950s, with only a short period of recovery during the 1990s.

However, other studies highlight that proxy records do not show such clear signals (Moffa-Sánchez et al., 2019), and the use of SST- and coastal sea level-based proxies of AMOC places uncertainties on these results (Jackson & Wood, 2020; Little et al., 2019; Menary et al., 2020). For instance, SSTs are additionally influenced by atmospheric and non-AMOC related ocean variability (Josey et al., 2018; Keil et al., 2020; Menary et al., 2020), while sea level responds to a variety of factors (e.g. atmospheric pressure and local winds) independent of the AMOC (Woodworth et al., 2014; Piecuch and Ponte, 2015; Piecuch et al., 2016).

Finally, large decadal variability is present in many reconstructions and obscures estimation of the long-term trend over the 20th century (Caesar et al., 2021; Ezer, 2013; McCarthy et al., 2015; Thornalley et al., 2018; Yashayaev & Loder, 2016). It is also noted that the proxy reported AMOC decline, beginning in the late 19th century, is not supported by model-based evidence (Sections 3.5.4.1. and 9.2.3.1).

Since the 1980s, multiple lines of observational evidence for AMOC change exist. Ship-based hydrographic estimates of AMOC as far back as the 1980s show no overall decline in AMOC strength (Fu et al., 2020; Worthington et al., 2020). Direct indications from in-situ observations report a $-2.5 \pm 1.4$ Sv change between 1993 and 2010 across the OVIDE section, superimposed on large interannual to decadal variability (Mercier et al., 2015). At 41°N and 26°N, a decline of $-3.1 \pm 3.2$ Sv per decade and $-2.5 \pm 2.1$ Sv per decade respectively has been reported over 2004–2016 (Baringer et al., 2018; Smeed et al., 2018). However, Moat et al. (2020) report an increase in AMOC strength at 26°N over 2009–2018. Recent time series of moored observations at 11°S (Humells et al., 2015), 34°S (Meinen et al., 2018; Kersalé et al., 2020), and between 57 and 60°N (Lozier et al., 2019a) are currently too short to permit robust conclusions about changes. The directly observed AMOC weakening since 2004, while significant, is over too short a period to assess whether it is part of a longer term trend or dominated by decadal-scale internal variability (Smeed et al., 2014; Collins et al., 2019; Moat et al., 2020). Notably an increase and subsequent decline in the 1990s is present in estimates of AMOC and associated heat transport constructed from reanalyses or auxiliary data (Frajka-Williams, 2015; Jackson & Wood, 2020; Jackson et al., 2016; Trenberth & Fasullo, 2017) (Section 9.2.3.1).

Repeated full depth in situ measurements report that deep convection - a major driver for AMOC - has recently returned to the Labrador Sea (particularly in 2015; Rhein et al., 2017; Yashayaev & Loder, 2016), and Irminger Sea (de Jong et al., 2018; de Jong & de Steur, 2016; Gladyshev et al., 2016) following an extended period with weak convection since 2000. An associated strengthening of the outflow from the Labrador Sea has not been observed (Zantopp et al., 2017; Lozier et al., 2019b), while strengthening of the AMOC is tentative (Desbruyeres et al., 2019; Moat et al., 2020). A long-term increase of the upper overturning cell in the Southern Ocean since the 1990s can be assessed with low confidence, and there is medium confidence of a decrease in Antarctic bottom water (AABW) volume and circulation, which has potential implications for the strength of the AMOC (Section 9.2.3.2).

In summary, proxy-based reconstructions suggested that the AMOC was relatively stable during the past 8 kyr (medium confidence), with a weakening beginning since the late 19th century (medium confidence), but due to a lack of direct observations, confidence in an overall decline of AMOC during the 20th century is low. From mid-2000s to mid-2010s, the directly observed weakening in AMOC (high confidence) cannot be distinguished between decadal-scale variability or a long-term trend (high confidence).
2.3.3.4.2 Western boundary currents and inter-basin exchanges

Both AR5 and SROCC reported that western boundary currents (WBCs) have undergone an intensification, warming and poleward expansion, except for the Gulf Stream and the Kuroshio, but did not provide confidence statements. AR5 reported with medium to high confidence intensification of the North Pacific subpolar gyre, the South Pacific subtropical gyre, and the subtropical cells, along with an expansion of the North Pacific subtropical gyre since the 1990s. It was pointed out that these changes are likely predominantly due to interannual-to-decadal variability, and in the case of the subtropical cells represent a reversal of earlier multi-decadal change. SROCC concluded that it was unlikely that there has been a statistically significant net southward movement of the mean Antarctic Circumpolar Current (ACC) position over the past 20 years, in contrast to AR5, where this change had been assessed with medium confidence.

The intensity of the Kuroshio current system in the northwest Pacific varied in conjunction with the glaciation cycles over the last 1 Myr, with some limited glacial-interglacial variability in position (Jian et al., 2000; Gallagher et al., 2015). The Agulhas current has strengthened substantially during the warming associated with deglaciations of the past 1 Myr (Peeters et al., 2004; Bard and Rickaby, 2009; Martínez-Méndez et al., 2010; Marino et al., 2013; Ballalai et al., 2019). According to sediment core analyses, the Agulhas leakage varied by about 10 Sv during major climatic transitions over the past 640 ky (Caley et al., 2014). Available data suggests that there was relatively little change in the net flow of the ACC in the LGM, with no consensus on the sign of changes (Lamy et al., 2015; Lynch-Stieglitz et al., 2016; McCall et al., 2013), except at one location at the northern edge of the Drake Passage where a 40% decrease of transport had been reported (Lamy et al., 2015). Longer time series from the northern entrance to Drake Passage suggest a consistent transport variability of 6–16% through glacial climate cycles, with higher current speeds during interglacial times and reduced current speeds during glacial intervals (Toyos et al., 2020). Inferred variability in the size and strength of the North Atlantic subpolar gyre was substantial, and included rapid changes on millennial time scales during both interglacial and glacial intervals over the last 150 ky (Born and Levermann, 2010; Mokeddem et al., 2014b; Irvali et al., 2016; Mokeddem and McManus, 2016). North Atlantic – Arctic exchange has also varied in the past, with indications of an increasing inflow of Atlantic waters into the Arctic during the late Holocene (Ślubowska et al., 2005) with an acceleration to the recent inflow that is now the largest of the past 2 ky (Spielhagen et al., 2011).

A latitudinal shift of subtropical/subpolar gyres on the order of 0.1 ± 0.04° per decade is derived by an indirect method using remote sensing data during 1993–2018 (Yang H. et al., 2020). Direct observations show a systematic poleward migration of WBCs (Bisagni et al., 2017; Wu L. et al., 2012; Yang H. et al., 2016, 2020). However, they do not support an intensification of WBCs, with a weakening, broadening, or little change reported for the Kuroshio (Collins et al., 2019; Wang & Wu, 2018; Wang et al., 2016), Gulf Stream (Andres et al., 2020; Collins et al., 2019; Dong et al., 2019; McCarthy et al., 2018), Agulhas (Beal and Elipot, 2016; Elipot and Beal, 2018) and East Australian (Sloyan and O’Kane, 2015) currents. The Gulf Stream has recently reversed a long-term poleward migration (Bisagni et al., 2017). Multidecadal variability of the strength and position of WBCs (Bisagni et al., 2017; Hsin, 2015; McCarthy et al., 2018) and short records from direct observations obscure the detection of any long-term trends (Yang H. et al., 2020).

The Pacific to Arctic exchange at the Bering Strait plays a minor role in the total Arctic exchange with the global ocean, which has increased from 0.8 Sv to 1.0 Sv over 1990–2015 (Woodgate, 2018). For Atlantic–Arctic exchange, major branches of Atlantic Water inflow from the North Atlantic into the Arctic across the Greenland–Scotland Ridge have remained stable since the mid-1990s (Berx et al., 2013; Hansen et al., 2015; Jochumsen et al., 2017; Østerhus et al., 2019), with only the smaller pathway of Atlantic Water north of Iceland showing a strengthening trend during 1993–2018 (Casanova-Masjoan et al., 2020), but with associated heat transport strengthening through the 1990s (Rosby et al., 2020; Tsuouchi et al., 2021). The Arctic outflow remained broadly stable from the mid-1990s to the mid 2010 (Østerhus et al., 2019). The heat and mass transport of the Indonesian throughflow (ITF) shows substantial variability at seasonal to decadal time scales (Feng M. et al., 2017, 2018; Li M. et al., 2018; Liu Q.-Y. et al., 2015; Sprintall et al., 2019; Susanto & Song, 2015; Xie T. et al., 2019; Zhuang et al., 2013). Liu et al. (2015) reported an increasing trend in the ITF geostrophic transport of 1 Sv per decade over 1984–2013, consistent with direct estimates (Sprintall et al., 2014), and results from reanalyses (Li et al., 2018), and this appears to be linked to multi-decadal scale variability rather than a long-term trend (England et al., 2014; Kosaka & Xie, 2013; Lee S.-K.)
et al., 2015). Southern Ocean circulation changes are assessed in SROCC (Meredith et al., 2019), and are confirmed and synthesized in Section 9.2.3.2 which shows that there is no indication of ACC transport change, and that it is unlikely that the mean meridional position of the ACC has moved southward in recent decades.

In summary, over the past 3–4 decades, the WBC strength is highly variable (high confidence), and WBCs and subtropical gyres have shifted poleward since 1993 (medium confidence). Net Arctic Ocean volume exchanges with the other ocean basins remained stable over the mid-1990s to the mid-2010s (high confidence). There is high confidence that the ITF shows strong multi-decadal scale variability since the 1980s.

2.3.3.5 Ocean pH

AR5 assessed with high confidence that the pH of the ocean surface has decreased since preindustrial times, primarily as a result of ocean uptake of CO₂. SROCC concluded that the global ocean absorbed 20–30% of total CO₂ emissions since the 1980s, with virtually certain ocean surface pH decline. The SROCC assessed a rate of surface pH decline of 0.017–0.027 pH units per decade across a range of time series of pH observations longer than 15 years. The decline in surface open ocean pH was assessed by SROCC as having very likely already emerged from background natural variability for more than 95% of the global surface open ocean.

Understanding of changes in surface pH at paleo time-scales has increased since AR5 (Anagnostou et al., 2020; Clarkson et al., 2015; Foster & Rae, 2016; Gutjahr et al., 2017; Harper et al., 2020; Henehan et al., 2019; Müller et al., 2020; Sosdian et al., 2018; Zeebe et al., 2016). Over the last 65 million years there have been several intervals when the pH of surface waters varied concurrently with climate change such as during the PETM, EECO, and MCO (Figure 2.29a) (Section 5.3.1.1). However, only during the PETM is the change sufficiently well-constrained to allow for a direct comparison with recent and current trends (Turner, 2018). This event was associated with profound perturbations of the global carbon cycle, ocean warming, deoxygenation and a surface ocean pH decrease likely ranging from 0.15 to 0.30 units (Penman et al., 2014; Gutjahr et al., 2017; Babila et al., 2018) - a rate that was likely at least an order of magnitude slower than today (Cui et al., 2011; Bowen et al., 2015; Frieling et al., 2016; Zeebe et al., 2016; Gutjahr et al., 2017; Kirtland Turner, 2018; Gingerich, 2019).

[start figure 2.29 here]

Figure 2.29: Low-latitude surface ocean pH over the last 65 million years. (a) Low-latitude (30°N–30°S) surface ocean pH over the last 65 million years, reconstructed using boron isotopes in foraminifera. (b) as (a) but for the last 3.5 million years. Double headed arrow shows the approximate magnitude of glacial-interglacial pH changes. (c) Multisite composite of surface pH. In a)–c), uncertainty is shown at 95% confidence as a shaded band. Relevant paleoclimate reference periods (CCB2.1) have been labelled. Period windows for succeeding panels are shown as horizontal black lines in a) and b). (d) Estimated low-latitude surface pH from direct observations (BATS, HOT) and global mean pH (65°S–65°N) from two indirect estimates (CMEMS, OCEAN-SODA). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[end figure 2.29 here]

Paleo evidence suggests that surface ocean pH has gradually increased over the last 50 Myr (Anagnostou et al., 2016; 2020; Sosdian et al., 2018) (Figure 2.29a). Global mean surface pH values as low as observed during recent decades are uncommon in the last 2 Myr (Figure 2.29b) (Chalk et al., 2017; Dyez et al., 2018; Martínez-Botí et al., 2015a; Sosdian et al., 2018), and have not been experienced in at least the last 25 kyr (Figure 2.29c) (Ezat et al., 2017; Foster, 2008; Gray et al., 2018; Henehan et al., 2013; Körschke et al., 2013; Martínez-Botí et al., 2015a; Naik et al., 2015; Palmer et al., 2010; Palmer & Pearson, 2003; Shao et al.,...
The magnitude of pH change during the Pleistocene glacial-interglacial cycles was 0.1–0.15 pH units - similar to recent changes in the modern era (Figure 2.29c) (Hönisch et al., 2009; Chalk et al., 2017; Shao et al., 2019) (Section 5.3.1.2). Maximum rates of pH change during the LDT, inferred from changes in atmospheric CO₂ recorded in ice cores (Marrat et al., 2014b) and the established relationships between pH and CO₂ changes and the boron isotope proxy (Hain et al., 2018), reached -0.02 pH units per century at about 11.7 ka, about 14.8 ka and about 16.3 ka, as previously sequestered CO₂ was transferred from the ocean interior to the subsurface ocean (Jaccard et al., 2016; Martínez-Botí et al., 2015a; Rae et al., 2018).

Since the 1980s, the global ocean has experienced a decline in surface pH of 0.016 ± 0.006 pH units per decade based on indirect pH products (Figure 2.29d) (Gehlen et al., 2020; Hurd et al., 2018; IPCC, 2019; Lauvset et al., 2015) that agrees with the decline of 0.017–0.025 pH units per decade assessed in SROCC from direct time-series measurements of pH. Section 5.3.2.2 assesses a decline that ranges from 0.01 to 0.026 pH units per decade for the tropical and subtropical open ocean areas, and 0.003–0.026 pH units per decade for the polar and subpolar open ocean regions by using time series and ship-based datasets from the surface ocean CO₂ measurement network (Bakker et al., 2016; Gehlen et al., 2020; Gregor & Gruber, 2021). There is general consensus that global surface ocean pH trends over the past two decades have exceeded the natural background variability (Bindoff et al., 2019; Gehlen et al., 2020; Lauvset et al., 2015). However, for some areas sparse data coverage, and large year-to-year variations hinders the detection of long-term surface ocean pH trends; for example in the Southern Ocean (Bindoff et al., 2019; Lauvset et al., 2015) and in the Arctic Ocean (Bindoff et al., 2019; Lauvset et al., 2015; Meredith et al., 2019).

For subsurface pH changes, estimates arise from direct ship measurements from repeated hydrography programs (Carter et al., 2019), indirect estimates of pH through calcite and aragonite saturation horizons (Osborne et al., 2020; Ross et al., 2020), and the very recent biogeochemical Argo floats equipped with pH sensors (Claustre et al., 2020). Global subsurface pH has decreased over the past 20–30 years, with signals observed to at least 1000 m depths (Lauvset et al., 2020). Global findings are supplemented by regional findings from the Pacific Ocean (Carter et al., 2019; Ross et al., 2020), the South Atlantic (Salt et al., 2015) and Southern Ocean (Jones et al., 2017); the North Atlantic Ocean and along the AMOC (Woosley et al., 2016; Perez et al., 2018), the Arctic Ocean (Qi et al., 2017) and marginal seas (Chen C.-T. A. et al., 2017). Further details are given in Section 5.3.3.1.

To conclude, it is virtually certain that surface open ocean pH has declined globally over the last 40 years by 0.003–0.026 pH per decade, and a decline in the ocean interior has been observed in all ocean basins over the past 2–3 decades (high confidence). A long-term increase in surface open ocean pH occurred over the past 50 Myr (high confidence), and surface open ocean pH as low as recent times is uncommon in the last 2 Myr (medium confidence). There is very high confidence that open ocean surface pH is now the lowest it has been for at least 26 kyr and current rates of pH change are unprecedented since at least that time.

2.3.3.6 Ocean deoxygenation

SROCC concluded that there has very likely been a net loss of oxygen over all ocean depths since the 1960s linked to global ocean deoxygenation at a range of 0.3–2.0%, and that the oxygen levels in the global upper 1000 m of the ocean had decreased by 0.5–3.3% during 1970–2010 (medium confidence), alongside an expansion of oxygen minimum zones (OMZ) by 3–8%. For the surface ocean (0–100 m) and the thermocline at 100–600 m, the very likely range of oxygen decline was assessed to be 0.2–2.1% and 0.7–3.5%, respectively. Multidecadal rates of deoxygenation showed variability throughout the water column and across ocean basins (high confidence).

Since AR5 evidence for changes in oxygen content based on new proxy reconstructions has increased (Anderson et al., 2019; Gottschalk et al., 2016, 2020; Hoogakker et al., 2015, 2018). Paleo records point to past periods of reduced oceanic oxygen levels during the late Permian (about 250 Ma), the Jurassic and Cretaceous (180–100 Ma), alongside associated global scale disturbances of the global carbon cycle.

Emerging studies for the Cenozoic Era suggest more stable ocean oxygenation conditions throughout the interval on million-year time scales (Wang X. et al., 2016). Sedimentary proxy data indicate, however, that the seemingly stable Cenozoic was punctuated by transient, widespread deoxygenation during the PETM.
(Dickson et al., 2012; Winguth et al., 2012; Remmelzwaal et al., 2019), with parts of the ocean reaching anoxic levels (Yao et al., 2018) (Section 5.3.1.1). Since the LGM, there was an overall emergence and expansion of low-oxygen waters into the ocean intermediate depths as a result of rapid warming and a reorganization of the global overturning circulation (Galbraith and Jaccard, 2015). The maximum expansion of oxygen-depleted waters during the LDT occurred coincidently with rapid warming in the NH at 14.7–12.9 ka (Hoogakker et al., 2018; Jaccard & Galbraith, 2012; Moffitt et al., 2015). Deep (> 1500m) ocean oxygen levels increased by 100–150 μmol kg⁻¹ since the LGM, reaching modern oxygen levels at about 10 ka (Anderson et al., 2019; Gottschalk et al., 2016; Hoogakker et al., 2015) (Section 5.3.1.2).

New findings for ocean oxygen content since SROCC are limited to regional scale assessments. The magnitude of change is difficult to compare across regions to arrive at a global assessment due to differences in depth range, time period, baseline climatology, methodology, and particularly the use of different units. To facilitate comparisons, data are presented as change per decade, and conversions of the SROCC global mean percentage of oxygen decline estimates are provided as a loss of 3.2 μmol kg⁻¹ in the upper 1000 m of the global ocean (1.93%), and 2.0 μmol kg⁻¹ (0.8 μmol kg⁻¹ per decade) in the upper 1000 m of the global ocean (1.93%), and 2.0 μmol kg⁻¹ (0.5 μmol kg⁻¹ per decade) in the entire water column (1.15%) between 1970 and 2010 (Bindoff et al., 2019). Oxygen change also shows decadal variability (Ito et al., 2016; Stramma et al., 2020) that can influence estimates of trends.

Subsurface (100–400 m) oxygen in the California Current system is estimated to have declined by 24 ± 2 μmol kg⁻¹ (1.0 μmol kg⁻¹ per decade) between 1993 and 2018, a rate similar to the global upper 1000 m average (Bindoff et al. 2019). In some locations, however, the magnitude of oxygen loss substantially exceeds global averages (Queste et al., 2018; Bronselaer et al., 2020; Cummins and Ross, 2020; Stramma et al., 2020). For example, a decline in oxygen content of 11.7 ± 3.5% in the upper 4000 m, including a decline of 20.4 ± 7.2% in the upper 1550 m, is reported in the North Pacific over the period 1958–2018 (Cummins and Ross, 2020) (equivalent to 2.3 μmol kg⁻¹ per decade in the upper 1550 m and 2.0 μmol kg⁻¹ decade⁻¹ throughout the 4000 m water column). In some regions of the Southern Ocean south of 65°S oxygen in the upper 2000 m has declined by 60 μmol kg⁻¹ (~52 μmol kg⁻¹ per decade) based on comparisons of 2014–2019 and 1985–2005 observations (Bronselaer et al., 2020). Within some OMZs regions of the Indian ocean, oxygen has declined from 6–12 to < 2 μmol kg⁻¹ between the 1960s and 2015–2016 (Bristow et al., 2017; Naqvi et al., 2018; Queste et al., 2018).

Findings since SROCC provide further support that the volume of severely oxygen-depleted waters has expanded in some locations of the global ocean (Section 5.3.3.2). For example, vertical expansion of low oxygen zones is reported in the North Pacific at a rate of 3.1 ± 0.5 m yr⁻¹ (Ross et al., 2020), and suboxic waters have increased by 20% at a rate of about 19 m per decade from 1982–2010 in the Arabian Sea (Naqvi et al., 2018; Lachkar et al., 2019), and expanded off the coast of Mexico (Sánchez-Velasco et al., 2019).

In summary, episodes of widespread and long-lasting (100 ka scales) open-ocean deoxygenation were related to warm climate intervals of the Permian-Cretaceous, with conditions becoming generally better oxygenated as the climate cooled over the course of the Cenozoic (high confidence). The largest expansions of oxygen depleted waters over the past 25 ka were strongly linked to rapid warming rates (medium confidence). Open-ocean deoxygenation has occurred in most regions of the open ocean during the mid-20th to early 21st centuries (high confidence), and shows decadal variability (medium confidence). Evidence further confirms SROCC that OMZs are expanding at many locations (high confidence).

### 2.3.4 Biosphere

This section is limited to a few biological indicators to demonstrate the close links between the biosphere and physical forcing. In selecting the indicators, we focussed on those (i) that are observable at large (global) spatial scales and over long (two decades or more) temporal scales, using standardised and consistent procedures; and (ii) those that are illustrative of the influence of the physical system on the biological realm. Chapters 2 and 3 of WGII AR6 undertake a more holistic assessment of biospheric impacts.
2.3.4.1 Seasonal Cycle of CO₂

AR5 noted that because CO₂ uptake by photosynthesis occurs only during the growing season, the greater land mass in the NH imparts a characteristic ‘sawtooth’ seasonal cycle in atmospheric CO₂. SRCCL similarly stated that due to strong seasonal patterns of growth, NH terrestrial ecosystems are largely responsible for the seasonal variations in global atmospheric CO₂ concentrations. Neither AR5 nor SRCCL made a confidence statement about observed changes in the amplitude of the seasonal cycle of CO₂.

In situ observations of CO₂ generally depict a rising amplitude of the seasonal cycle over the past half century, especially north of about 45°N (Figure 2.30). For example, an amplitude increase of 6 ± 2.6% per decade has been observed at the Barrow surface observatory in Alaska over 1961–2011 (Graven et al., 2013), with slightly slower increases thereafter. Aircraft data north of 45°N exhibit an amplitude increase of 57 ± 7% at 500 mb versus an increase of 26 ± 18% for 35°N–45°N between field campaigns in 1958–1961 and 2009–2011 (Graven et al., 2013b). Increases in amplitude for the period 1980–2012 are apparent at eight surface observatories north of 50°N (Piao et al., 2018), related primarily to a larger drawdown in June and July. Trends in seasonal cycle amplitude at lower latitudes are smaller (if present at all); for instance, the increase at the Mauna Loa observatory in Hawaii since the early 1960s is only about half as large as at Barrow (Graven et al., 2013a), and only one other low-latitude observatory has a significant increase from 1980–2012 (Piao et al., 2018). There is a weak signal of an increase in amplitude at the Sinhagad observatory in western India in recent years (Chakraborty et al., 2020). Generally speaking, larger increases in the Arctic and boreal regions are indicative of changes in vegetation and carbon cycle dynamics in northern ecosystems (Forkel et al., 2016), though increased carbon uptake can also result from other factors such as warmer- and wetter-than-normal conditions.

Recent satellite-based, global-scale estimates of seasonal variations in atmospheric CO₂ for the period 2003–2018 show that the seasonal variations in the SH are out of phase with those in the NH (Reuter et al., 2020), which is consistent with the phenological shifts in primary productivity between hemispheres. The net effect of the phase shift between the two hemispheres is to dampen the amplitude of the global average seasonal cycle. These integrated results also show that the amplitude of the oscillations has been increasing in the SH, from about 2009, but comparison with data from Baring head suggests that periods of high seasonal oscillation had occurred at that location in the SH prior to 1995.

Figure 2.30: Changes in the amplitude of the seasonal cycle of CO₂. (a) Observed peak-to-trough seasonal amplitude given by the day of year of downward zero crossing, of CO₂ concentration at Barrow (71°N, blue) and Mauna Loa (20°N, black). Seasonal CO₂ cycles observed at (b) Barrow and (c) Mauna Loa for the 1961–1963 or 1958–1963 and 2017–2019 time periods. The first six months of the year are repeated. Reprinted with permission from AAAS. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

In summary, there is very high confidence that the amplitude of the seasonal cycle of atmospheric CO₂ has increased at mid-to-high NH latitudes since the early 1960s. The observed increase is generally consistent with greater greening during the growing season and an increase in the length of the growing season over the high northern high latitudes. Similarly, globally-integrated results from the SH also show an increase in seasonal amplitude of atmospheric CO₂ signal, from around 2009 to 2018 (low confidence).

2.3.4.2 Marine biosphere

2.3.4.2.1 Large-scale distribution of marine biota

SROCC pointed out that long-term global observations of many key ocean variables, including...
phytoplankton, have not reached the density and accuracy necessary for detecting change. But SROCC noted the good comparability between short time-scale single-sensor ocean-colour products and a longer time-scale, climate-quality time series of multi-sensor, inter-sensor-bias-corrected, and error characterised, global data on chlorophyll-a concentration in the surface layers of the ocean. With respect to oligotrophic gyres, AR5 WGII concluded that the oligotrophic subtropical gyres of the Atlantic and Pacific Oceans are expanding and that they indicate declining phytoplankton stocks in these waters (limited evidence, low agreement). With respect to distributions of marine organisms, AR5 WGII reported range shifts of benthic, pelagic, and demersal species and communities (high confidence), though the shifts were not uniform.

Phytoplankton are responsible for marine primary production through photosynthesis; they are a major player in the ocean carbon cycle. They have a high metabolic rate, and respond fast to changes in environmental conditions (light, temperature, nutrients, mixing), and as such, serve as a sentinel for change in marine ecosystems. Concentration of chlorophyll-a, the major photosynthetic pigment in all phytoplankton, is often used as a measure of phytoplankton biomass. As primary producers, they are also food for organisms at higher trophic levels. The multi-sensor time series of chlorophyll-a concentration has now been updated (Sathyendranath et al., 2019) to cover 1998–2018. Figure 2.31 shows that global trends in chlorophyll-a for the last two decades are insignificant over large areas of the global ocean (von Schuckmann et al., 2019), but some regions exhibit significant trends, with positive trends in parts of the Arctic and the Antarctic waters (>3% yr⁻¹), and both negative and positive trends (within ±3% yr⁻¹) in parts of the tropics, subtropics and temperate waters. The interannual variability in chlorophyll-a data in many regions is strongly tied to indices of climate variability (Section 2.4, Annex IV) and changes in total concentration are typically associated with changes in phytoplankton community structure (e.g., Brewin et al., 2012; Racault et al., 2017a). Variability in community structure related to El Niño has, in turn, been linked to variability in fisheries, for example in the catch of anchovy (Engraulis ringens) in the Humboldt current ecosystem (Jackson et al., 2011).

[START FIGURE 2.31 HERE]

Figure 2.31: Phytoplankton dynamics in the ocean. (a) Climatology of chlorophyll-a concentration derived from ocean-colour data (1998–2018); (b) Linear trends in chlorophyll concentration. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al (2008c) (‘x’ marks denote non-significant changes). (c) Histogram of linear trends in chlorophyll concentration, after area weighting and with per-pixel uncertainty estimates based on comparison with in situ data. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.31 HERE]

Since AR5 WGII, analysis of a longer time series of ocean-colour data (1998–2012) has shown (Aiken et al., 2017) that the expansion of the low nutrient part of the North Atlantic oligotrophic gyre was significant, at 0.27 x 10⁶ km² per decade, but that the rate was much lower than that reported earlier by Polovina et al. (2008). Furthermore, Aiken et al. (2017) reported no significant trend in the oligotrophic area of the South Atlantic Gyre. With the time series extended to 2016, von Schuckmann et al. (2018) reported that since 2007, there was a general decreasing trend in the areas of the North and South Pacific oligotrophic Gyres, while the North and South Atlantic oligotrophic Gyres remained stable, with little change in area, consistent with Aiken et al. (2017). The changing sign of trends in the areal extent of the oligotrophic gyres with increase in the length of the time series raises the possibility that these changes arise from interannual to multi-decadal variability. The time series of ocean-colour data is too short to discern any trend that might be superimposed on such variability.

Similarly, there is limited consistent and long-term information on large-scale distributions of marine organisms at higher trophic levels. But there are increased indications since AR5 and SROCC that the distributions of various higher trophic-level organisms are shifting both polewards and to deeper levels (Atkinson et al., 2019; Edwards et al., 2016; Haug et al., 2017; Lenoir et al., 2020; Pinsky et al., 2020), mostly consistent with changes in temperature. However observations also show a smaller set of counter-
intuitive migrations towards warmer and shallower waters, which could be related to changes in phenology and in larval transport by currents (Fuchs et al., 2020). There are also strengthening indications of greater representation by species with warm-water affinity in marine communities, consistent with expectations under observed warming (Burrows et al., 2019). There are indications that pre-1850 CE plankton communities are different from their modern counterparts globally (Jonkers et al., 2019). Indicators of geographical distributions of species (mostly from coastal waters) suggest that the rates at which some species are leaving or arriving at an ecosystem are variable, leading to changes in community composition (Blowes et al., 2019), with likely greater representation of warm-water species in some locations (Burrows et al., 2019).

In summary, there is high confidence that the latitudinal and depth limits of the distribution of various organisms in the marine biome are changing. There is medium confidence that there are differences in the responses of individual species relative to each other, such that the species compositions of ecosystems are changing. There is medium confidence that chlorophyll concentration in the surface shows weak negative and positive trends in parts of low and mid latitudes, and weak positive trends in some high-latitude areas. There is medium confidence that the large-scale distribution of the oligotrophic gyre provinces is subject to significant inter-annual variations, but low confidence in the long-term trends in the areal extent of these provinces because of insufficient length of direct observations.

2.3.4.2.2 Marine Primary production

SROCC expressed low confidence in satellite-based estimates of trends in marine primary production, citing insufficient length of the time series and lack of corroborating in situ measurements and independent validation time series. The report also cites significant mismatches in absolute values and decadal trends in primary production when different satellite-based products are compared.

Recent model-based results with assimilation of satellite data (Gregg and Rousseaux, 2019), show global annual mean marine primary production of around 38 (±1.13) PgC yr⁻¹ over 1998–2015. This new result lies towards the low end of values reported in earlier, satellite-based studies (range 36.5–67 PgC yr⁻¹, reported in Sathyendranath et al., 2020). Reconciling the results of Gregg & Rousseaux (2019) with earlier satellite-based studies leads to a mean of 47 (±7.8) PgC yr⁻¹. There is a strong correlation between interannual regional variability in marine primary production and climate variability (Gregg & Rousseaux, 2019; Racault et al., 2017b). The increase in primary production in the Arctic has been associated with retreating sea ice and with increases in nutrient supply and chlorophyll concentration (Lewis et al., 2020). Gregg and Rousseaux (2019) reported a decreasing trend in marine primary production, of -0.8 PgC (-2.1%) per decade globally. There is low confidence in this trend because of the small number of studies and the short length of the time series (<20 years).

In conclusion, there is low confidence because of the small number of recent studies and the insufficient length of the time series analysed that marine primary production is 47 (±7.8) PgC yr⁻¹. A small decrease in productivity is evident globally for the period 1998–2015, but regional changes are larger and of opposing signs (low confidence).

2.3.4.2.3 Marine phenology

Phenology is the study of the timing of important events in the annual life cycle of organisms (plants or animals) (see also Annex VII: Glossary). AR5 WGII noted that the timing of various seasonal biological events in the ocean had advanced by more than four days per decade over the previous 50-year period and concluded that there was high confidence in observed changes in the phenological metrics of marine organisms. AR5 WGII further reported that, of those observations that showed a response, 81% of changes in phenology, distribution and abundance were consistent with anticipated responses to climate warming according to theoretical expectations, corroborated by updates in SROCC. The consequent current and future impacts on interactions between species, including competition and predator-prey dynamics, were noted with high confidence.
There are additional indications that phenological metrics related to different species are changing, but not always in a similar manner. For example, many seabirds are breeding earlier, while others are breeding later (Sydeman et al., 2017). Planktonic organisms in the North Atlantic are also responding differently to each other when subjected to the same environmental changes (Edwards & Richardson, 2004). Furthermore, different factors could be responsible for triggering phenological responses in different stages in the life cycle of a single organism (Koeller et al., 2009). The shift in the distribution of many benthic invertebrates on the North-West Atlantic shelf, including some commercially-important shellfish, could be explained by phenology and larval transport and the shift and contraction of species range have been associated with higher mortality (Fuchs et al., 2020). Changes in phytoplankton phenological indicators globally (Racault et al., 2012; Sapiano et al., 2012) have been linked to indicators of climate variability, such as the multivariate ENSO Index (Racault et al., 2017a) with responses varying across ecological provinces of the ocean (Longhurst, 2007).

Phenological links between multiple components of an ecosystem have to be maintained intact, to retain system integrity. Since all higher pelagic organisms depend on phytoplankton for their food, either directly or indirectly, a match favours survival, and a mismatch is antagonistic to survival. Match represents synchronicity in the phenological events of both prey and predator. There are indications from ocean-colour data used in conjunction with fisheries data that the survival rate of various larger marine organisms depends on phenological metrics related to the seasonality of phytoplankton growth. Such links have been demonstrated, for example, for haddock (Melanogrammus aeglefinus) in the North-West Atlantic (Platt et al., 2003); northern shrimp in the North Atlantic (Koeller et al., 2009; Ouellet et al., 2011); sardine (Sardinella aurita) off the Ivory coast (Kassi et al., 2018); cod (Gadus morhua) and haddock (Melanogrammus aeglefinus) larvae in the North-West Atlantic (Trzcinski et al., 2013); and oil sardine (Sardinella longiceps) off the south-west coast of India. Borstad et al. (2011) showed that fledgling production rate of rhinoceros auklets (Cerorhinca monocerata) on a remote island in coastal north-eastern Pacific was related to seasonal values of chlorophyll-a biomass in the vicinity of the island.

In summary, new in situ data as well as satellite observations strengthen AR5 and SROCC findings that various phenological metrics for many species of marine organisms have changed in the last half century (high confidence), though many regions and many species of marine organisms remain under-sampled or even unsampled. The changes vary with location and with species (high confidence). There is a strong dependence of survival in higher trophic-level organisms (fish, exploited invertebrates, birds) on the availability of food at various stages in their life cycle, which in turn depends on phenologies of both (high confidence). There is a gap in our understanding of how the varied responses of marine organisms to climate change, from a phenological perspective, might threaten the stability and integrity of entire ecosystems.

2.3.4.3 Terrestrial biosphere

2.3.4.3.1 Growing season and phenology changes

AR5 WGII briefly discussed large-scale changes in the length of the growing season but made no confidence statement about observed trends. However, AR5 did conclude with high confidence that warming contributed to an overall spring advancement in the NH.

Recent in situ analyses document increases in the length of the thermal growing season (i.e., the period of the year when temperatures are warm enough to support growth) over much of the extratropical land surface since at least the mid-20th century. Over the NH as a whole, an increase of about 2.0 days per decade is evident for 1951–2018 (Dunn et al., 2020), with slightly larger increases north of 45°N (Barichivich et al., 2013). Over North America, a rise of about 1.3 days per decade is apparent in the United States for 1900–2014 (Kukal and Irmak, 2018), with larger increases after 1980 (McCabe et al., 2015); likewise, all ecozones in Canada experienced increases from 1950–2010 (Pedlar et al., 2015). Growing season length in China increased by at least 1.0 days per decade since 1960 (Xia et al., 2018) and by several days per decade in South Korea since 1970 (Jung et al., 2015). In general, changes in phenological indicators are consistent with the increase in growing season length documented by instrumental data (Parmesan and Hanley, 2015).

Several long-term, site-specific records illustrate the unusualness of recent phenological changes relative to...
interannual variability; for example, peak bloom dates for cherry blossoms in Kyoto, Japan have occurred progressively earlier in the growing season in recent decades, as have grape harvest dates in Beaune, France (Figure 2.32).

[START FIGURE 2.32 HERE]

Figure 2.32: Phenological indicators of changes in growing season. (a) cherry blossom peak bloom in Kyoto, Japan; (b) grape harvest in Beaune, France; (c) spring phenology index in eastern China; (d) full flower of Piedmont species in Philadelphia, USA; (e) Grape harvest in Central Victoria, Australia; (f) start of growing season in Tibetan Plateau, China. Red lines depict the 25-year moving average (top row) or the 9-year moving average (middle and bottom rows) with the minimum roughness boundary constraint of Mann (2004). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.32 HERE]

Changes in the length of the photosynthetically active growing season (derived from the Normalised Difference Vegetation Index (NDVII)) are also evident over many land areas since the early 1980s. Increases of about 2.0 days per decade are apparent north of 45°N since the early 1980s (centred over mid-latitude Eurasia and north-eastern North America), with indications of a reversal to a decline in season length starting in the early 2000s (Barichivich et al., 2013; Garonna et al., 2016; Liu Q. et al., 2016; Park et al., 2016; Zhao J. et al., 2015). Satellite-based records suggest that most NH regions have experienced both an earlier start and a later end to the growing season, a finding supported by ground-based data (Piao et al., 2020). A number of studies also capture increases in growing season length over the Canadian Arctic (Chen, 2016), Fennoscandia (Hogda et al., 2013), most of Europe (Garonna et al., 2014), and parts of sub-Saharan Africa (Vrielings et al., 2013).

The general consistency between in situ and satellite estimates over the NH is noteworthy given that many factors independently contribute to uncertainty in observed changes. For example, there is no universally accepted definition of growing season length across in situ analyses; some define the growing season as the period based on a temperature threshold (e.g., 5°C) whereas others use the frost-free period. Spatial and temporal coverage can also affect conclusions based upon in situ studies (Donat et al., 2013). For satellite analyses, uncertainties can be related to the satellite datasets themselves (e.g., satellite drift, sensor differences, calibration uncertainties, atmospheric effects); and to the methods for determining phenological metrics (e.g., start, end, and length of season) (Wang S. et al., 2016).

In summary, based on multiple independent analyses of in situ, satellite, and phenological data, there is high confidence that the length of the growing season has increased over much of the extratropical NH since at least the mid-20th century.

2.3.4.3.2 Terrestrial ecosystems

AR5 WGII concluded that many terrestrial species have shifted their geographic ranges in recent decades (high confidence). Similarly, SRCCL assessed that many land species have experienced range size and location changes as well as altered abundances over recent decades (high confidence). SROCC noted that species composition and abundance have markedly changed in high mountain ecosystems in recent decades (very high confidence).

Paleoclimate reconstructions document large-scale biome shifts from the deep past through the Holocene (e.g., Hoogakker et al., 2016). The northernmost location of the treeline is a representative indicator in this regard (Figure 2.34). During the MPWP, boreal forest extended to the Arctic coast, with the northernmost treeline being about 4° to 10° latitude further north than at present; temperate forests and grasslands were also shifted poleward (with reduced tundra extent), while savannahs and woodlands were more expansive in Africa and Australia at the expense of deserts (Salzmann et al., 2008, 2013; Sniderman et al, 2016; Andrae et
Long-term ecological records capture extensive range shifts during the 20th and early 21st centuries (Lenoir & Svenning, 2015; Pecl et al., 2017). Research has been most extensive for North America and western Eurasia, with fewer studies for central Africa, eastern Asia, South America, Greenland, and Antarctica (Lenoir & Svenning, 2015). Most documented changes are toward cooler conditions, i.e., poleward and upslope (Lenoir et al., 2008; Harsh et al., 2009; Elmendorf et al., 2015; Parmesan and Hagley, 2015; Evans and Brown, 2017). Notably, a large, quasi-global analysis (Chen et al., 2011) estimated that many insect, bird, and plant species had shifted by 17 (± 3) km per decade toward higher latitudes and 14 (± 3) m per decade toward higher elevations since the mid-20th century, with changes in both the leading and trailing edges of species ranges (Rumpf et al., 2018). Over the past century, long-term ecological surveys also show that species turnover (i.e., the total number of gains and losses of species within an area) has significantly increased across a broad array of ecosystems (Dornelas et al., 2014, 2019), including undisturbed montane areas worldwide (Gibson-Reinemer et al., 2015). Despite global losses to biodiversity, however, most local assemblages have experienced a change in biodiversity rather than a systematic loss (Pimm et al., 2014).

With increased species turnover, the novelty of contemporary communities relative to historical baselines has risen (Hobbs et al., 2009; Radeloff et al., 2015) due to greater spatial homogenization, mixtures of exotic and native species, altered disturbance regimes, and legacies of current or historic land use (Olden and Rooney, 2006; Schulte et al., 2007; Thompson et al., 2013; Goring et al., 2016). In general, terrestrial species have had lower rates of turnover than marine species (Dornelas et al., 2018; Blowes et al., 2019).

There are exceptions to the general pattern of poleward / upslope migration. For some species, various biotic and abiotic factors (such as precipitation and land use) supersede the physiological effects of temperature (Vanderwal et al., 2013; Gibson-Reinemer and Rahel, 2015; Ordonez et al., 2016; Scheffers et al., 2016; Lenoir et al., 2020). For other species, poleward migration is slower than expectations from the observed temperature increases. Trees are one such example because of their long lifespan and gradual maturity (Renwick and Rocca, 2015); in fact, poleward advance is only evident at about half of the sites in a large global dataset of treeline dynamics for 1900–present (Harsch et al., 2009b). Furthermore, the northernmost extent of treeline at present (roughly 73° N) is actually somewhat south of its location in the MH (MacDonald et al., 2008) despite an expanding growing season in the extratropical NH since the mid-20th century (Section 2.3.4.3.1).

Consistent with species range shifts, SRCCL noted that there have been changes in the geographical distribution of climate zones. Poleward shifts in temperate and continental climates are evident across the globe over 1950–2010, with decreases in the area (and increases in the average elevation) of polar climates (Chan and Wu, 2015). Zonal changes towards higher latitudes in winter plant hardiness regions are apparent since the 1970s over the central and eastern United States, with elevational changes also being important in the western United States (Daly et al., 2012). A clear northward shift in winter plant hardiness zones is detectable across western Canada since 1930, with somewhat lesser changes in the south eastern part of the country (McKenney et al., 2014). A northward migration of agro-climate zones is also evident over Europe since the mid-1970s (Ceglar et al., 2019). In addition, a shift toward more arid climate zones is apparent in some areas, such as the Asian monsoon region (Son and Bae, 2015) as well as parts of South America and Africa (Spinoni et al., 2015).
In summary, there is very high confidence that many terrestrial species have shifted their geographic ranges poleward and/or upslope over the past century, with increased rates of species turnover. There is high confidence that the geographical distribution of climate zones has shifted in many parts of the world.

2.3.4.3.3 Global greening and browning

AR5 WGII briefly discussed changes in global vegetation greenness derived from satellite proxies for photosynthetic activity. Observed trends varied in their strength and consistency, and AR5 thus made no confidence statement on observed changes. SRCCGL subsequently concluded that greening had increased globally over the past 2–3 decades (high confidence).

Vegetation index data derived from AVHRR and MODIS depicts increases in aspects of vegetation greenness (i.e., green leaf area and/or mass) over the past four decades (Piao et al., 2020). NDVI increased globally from the early 1980s through the early 2010s (Liu et al., 2015c). Pan et al. (2018a) found NDVI increases over about 70% of the Earth’s vegetated surface through 2013, and Osborne et al. (2018) noted strong upward changes in NDVI in the circumpolar Arctic through 2016. Globally integrated Leaf Area Index (LAI) also rose from the early 1980s through at least the early 2010s (Zhu et al., 2016; Forzieri et al., 2017; Jiang et al., 2017; Xiao et al., 2017) and probably through near-present; for example, Chen et al. (2019) documented an LAI increase over one-third of the global vegetated area from 2000–2017. Although less frequently analysed for temporal trends, Fraction of Absorbed Photosynthetically Active Radiation (FAPAR) likewise increased over many global land areas (particularly China, India, and Eastern Europe) in the past two decades (Figure 2.33) (Forkel et al., 2014; Gobron, 2018; Keenan & Riley, 2018). There are also documented changes in specific vegetation types, such as a 7% rise in global tree cover for 1982–2016 (Song et al., 2018) and an expansion of shrub extent in the Arctic tundra over 1982–2017 (Myers-Smith et al., 2020). The increased greening is largely consistent with CO₂ fertilization at the global scale, with other changes being noteworthy at the regional level (Piao et al., 2020): examples include agricultural intensification in China and India (Chen et al., 2019; Gao et al., 2019) and temperature increases in the northern high latitudes (Kong et al., 2017; Keenan and Riley, 2018) and in other areas such as the Loess Plateau in central China (Wang et al., 2018). Notably, some areas (such as parts of Amazonia, central Asia, and the Congo basin) have experienced browning (i.e., decreases in green leaf area and/or mass) (Anderson et al., 2019; Gottschalk et al., 2016; Hoogakker et al., 2015). Because rates of browning have exceeded rates of greening in some regions since the late 1990s, the increase in global greening has been somewhat slower in the last two decades (Pan et al., 2018a).

[START FIGURE 2.33 HERE]

Figure 2.33: Satellite-based trends in Fraction of Absorbed Photosynthetically Active Radiation (per decade) for 1998–2019. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al. (2008c) (‘x’ marks denote non-significant trend). Unvegetated areas such as barren deserts (grey) and ice sheets (white) have no trend in FAPAR. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.33 HERE]

Global-scale linear trends differ substantially across products for the same periods and trend metrics used (Jiang et al., 2017). Several factors contribute to this large span in estimated changes. Remotely sensed vegetation products vary in their spatial and temporal completeness as well as resolution and are sensitive to contamination from atmospheric composition, clouds, snow cover, and anisotropy, as well as orbital changes and sensor degradations (de Jong et al., 2012; Zhu et al., 2016; Jiang et al., 2017; Xiao et al., 2017; Pan et al., 2018a). Ground-based measurements suitable for calibration and validation are scarce before 2000 (Xiao et al., 2017), and the recalibration of satellite records (e.g., as in from MODIS Collection 5 to 6) can affect trends (Piao et al., 2020). It is possible that the increase in greenerness over 2000–2015 is larger than the increase in gross primary production (based on flux tower measurements and MODIS Collection 6 data) (Zhang L. et al., 2018). Land use changes and altered disturbance regimes (e.g. floods, fires, diseases) may...
mask large-scale signals (Franklin et al., 2016). In addition, there is a plethora of models for the
identification of phenological metrics from satellite data as well as a variety of statistical techniques for
analysing historical changes (Wang et al., 2016).

In summary, there is high confidence that vegetation greenness (i.e., green leaf area and/or mass) has
increased globally since the early 1980s. However, there is low confidence in the magnitude of this increase
owing to the large range in available estimates.

2.3.5 Synthesis of evidence for past changes

Section 2.3 has assessed the observational evidence for changes in key indicators across the atmosphere,
cryosphere, ocean and biosphere starting, where applicable, from paleoclimate proxy records and coming up
to the present day. This synthesis serves as an assessment of the evidence for change across the climate
system as represented by the instrumental record and its unusualness in the longer-term context. Building
upon previous sections assessing the observational evidence for each key indicator individually, this section
integrates the evidence across multiple indicators to arrive at a holistic and robust final assessment.

Climate has varied across a broad range of timescales (Figure 2.34). During the Cenozoic Era temperatures
generally decreased over tens of millions of years, leading to the development of ice sheets. During the last
two million years, climate has fluctuated between glacial and interglacial periods. Within the current Holocene
interglacial and, with increasing detail in the CE, it is possible to reconstruct a history both of more
indicators of the climate system and, with increasing fidelity, the rates of change. Solely for the last 150
years or so are instrumental observations of globally distributed climate indicators available. However, only
since the late 20th century have observational systems attained essentially global monitoring capabilities.
The direct observations point unequivocally to rapid change across many indicators of the climate system
since the mid-19th century. These are all consistent in indicating a world that has warmed rapidly.

Assessing the long-term context of recent changes is key to understanding their potential importance and
implications. The climate system consists of many observable aspects that vary over a very broad range of
timescales. Some biogeochemical indicators of change such as atmospheric CO₂ concentrations and ocean
pH have shifted rapidly and CO₂ concentrations are currently at levels unseen in at least 800 kyr (the period
of continuous polar ice-core records) and very likely for millions of years. The GMST in the past decade is
likely warmer than it has been on a centennially-averaged basis in the CE and more likely than not since the
peak of the LIG. Many more integrative components of the climate system (e.g., glaciers, GMSL) are
experiencing conditions unseen in millennia, whereas the most slowly responding components (e.g. ice-sheet
extent, permafrost, tree line) are at levels unseen in centuries (high confidence). The rate at which several
assessed climate indicators (e.g., GMSL, OHC, GSAT) have changed over recent decades is highly unusual
in the context of preceding slower changes during the current post-glacial period (high confidence).

In summary, directly observed changes in the atmosphere, ocean, cryosphere and biosphere are unequivocal
evidence of a warming world. Key climate indicators are now at levels not experienced for centuries to
millennia. Since the late 19th century many indicators of the global climate system have changed at a rate
unprecedented over at least the last two thousand years.

[START FIGURE 2.34 HERE]

Figure 2.34: Selected large-scale climate indicators during paleoclimate and recent reference periods of the
Cenozoic Era. Values are based upon assessments carried out in this chapter, with confidence levels
ranging from low to very high. Refer to Cross-Chapter Box 2.1 for description of paleoclimate reference
periods and Section 1.4.1 for recent reference periods. Values are reported as either the very likely range
(x to y), or best estimates from beginning to end of the reference period with no stated uncertainty (x →
y), or lowest and highest values with no stated uncertainty (x ~ y). Temperature is global mean surface
temperature. Glacier extent is relative and colour scale is inverted so that more extensive glacier extent is
intuitively blue.
Cross-Chapter Box 2.4: The climate of the Pliocene (around 3 million years ago), when CO₂ concentrations were last similar to those of present day

**Contributing Authors:** Alan Haywood (UK), Darrell Kaufman (USA), Nicholas Golledge (New Zealand), Dabang Jiang (China), Daniel Lunt (UK), Erin McClymont (UK), Ulrich Salzmann (UK), Jessica Tierney (USA)

Throughout this Report, information about past climate states is presented in the context of specific climate variables, processes or regions. This Cross-Chapter Box focuses on a single paleoclimate reference period as an example of how proxy data, models and process understanding come together to form a more complete representation of a warm climate state that occurred during the relatively recent geologic past.

**Introduction.** The Pliocene Epoch is one of the best-documented examples of a warmer world during which the slow responding components of the climate system were approximately in balance with concentrations of atmospheric CO₂, similar to present (e.g., Haywood et al., 2016). It provides a means to constrain Earth’s equilibrium climate sensitivity (Section 7.5.3) and to assess climate model simulations (Section 7.4.4.1.2). During the Pliocene, continental configurations were similar to present (Cross-Chapter Box 2.4 Figure 1a), and many plant and animal species living then also exist today. These similarities increase reliability of paleo-environmental reconstructions compared with those for older geological periods. Within the well-studied mid-Pliocene Warm Period (MPWP, also called the mid-Piacenzian Warm Period, 3.3–3.0 Ma), the interglacial period KM5c (3.212–3.187 Ma) has become a focus of research because its orbital configuration, and therefore insolation forcing, was similar to present (global mean insolation = −0.022 W m⁻² relative to modern; Haywood et al., 2013), allowing for the climatic state associated with relatively high atmospheric CO₂ to be assessed with fewer confounding variables.

**Major global climate indicators.** During the KM5c interglacial, atmospheric CO₂ concentration was typically between 360 and 420 ppm (Section 2.2.3.1). New climate simulations of this interval from the Pliocene Model Intercomparison Project Phase 2 (PlioMIP2) show a multi-model mean global surface air temperature of 3.2°C [2.1–4.8] °C warmer than control simulations (Haywood et al., 2020; Cross-Chapter Box 2.4 Figure 1a). This is consistent with proxy evidence for the broader MPWP, which indicates that global mean surface temperature was 2.5°C–4.0°C higher than 1850–1900 (Section 2.3.1.1.1). Global mean sea level was between 5 and 25 m higher than present (Section 2.3.3.3). Geological evidence (Section 2.3.2.4) and ice-sheet modelling (Section 9.6.2) indicate that both the Antarctic and Greenland Ice Sheets were substantially smaller than present (Cross-Chapter Box 2.4 Figure 1c). Attribution of sea level highstands to particular ice sheet sources (Section 9.6.2) is challenging (DeConto & Pollard, 2016; Golledge, 2020), but improving (Berends et al., 2019; Grant et al., 2019).

**Northern high latitudes.** The latitudinal temperature gradient during the MPWP was reduced relative to present-day and the consistency between proxy and modelled temperatures has improved since AR5 (Section 7.4.4.1.2). Northern high latitude (> 60°N) SSTs were up to 7°C higher than 1850–1900 (Bachem et al., 2016; McClymont et al., 2020; Sánchez-Montes et al., 2020), and terrestrial biomes were displaced poleward (e.g., Dowsett et al., 2019) (Cross-Chapter Box 2.4, Figure 1b). Arctic tundra regions currently underlain by permafrost were warm enough to support boreal forests, which shifted northward by approximately 250 km in Siberia, and up to 2000 km in the Canadian Arctic Archipelago (Fletcher et al., 2017; Salzmann et al., 2013). The shift caused high-latitude surface albedo changes, which further amplified the Pliocene global warming (Zhang & Jiang, 2014). Vegetation changes in northeast Siberia indicate that MPWP summer temperatures were up to 6°C higher than present day (Brigham-Grette et al., 2013). Farther south, modern boreal forest regions in Russia and eastern North America were covered with temperate forests and grasslands, whereas highly diverse, warm-temperate forests with subtropical taxa were widespread in central...
and eastern Europe (Cross-Chapter Box 2.4, Figure 1). While seasonal sea ice was present in the North
Atlantic and Arctic oceans, its winter extent was reduced relative to present (Knies et al., 2014; Clotten et al.,
2018), and some models suggest that the Arctic was sea ice free during the summer (Feng et al., 2020;
Howell et al., 2016).

**Tropical Pacific.** The average longitudinal temperature gradient in the tropical Pacific was weaker during
the Pliocene than during 1850–1900 (Section 7.4.4.2.2). Changes in Pacific SSTs and SST gradients had far-
reaching impacts on regional climates through atmospheric teleconnections, affecting rainfall patterns in
western North America (Burls & Fedorov, 2017; Ibarra et al., 2018). The reduced zonal SST gradient has led
some to propose that the Pliocene Pacific experienced a “permanent El Niño” state (Molnar and Cane, 2002;
Fedorov, 2006). However, there is no direct geological evidence, nor support from climate models, that
ENSO variability collapsed during the Pliocene. Although not located in the centre-of-action region for
ENSO, Pliocene corals show temperature variability over 3–7 year timescales (Watanabe et al., 2011). In
addition, a multi-model intercomparison indicates that ENSO existed, albeit with reduced variability
(Brierley, 2015). Thus, there is high confidence that ENSO variability existed during the Pliocene.

**Hydrological cycle.** Vegetation reconstructions for the late Pliocene indicate regionally wetter conditions
resulting in an expansion of tropical savannas and woodlands in Africa and Australia at the expense of
deserts (Cross-Chapter Box 2.4, Figure 1b). PlioMIP2 climate models generally simulate higher rates of
mean annual precipitation in the tropics and high latitudes, and a decrease in the subtropics, with a multi-
model mean global increase of 0.19 mm day$^{-1}$ [0.13–0.32 mm day$^{-1}$] relative to control simulations
(Haywood et al., 2020) (Cross-Chapter Box 2.4, Figure 1a). Both simulations and limited proxy evidence
indicate stronger monsoons in northern Africa, Asia, and northern Australia relative to present, but trends are
uncertain in other monsoon regions (Huang X. et al., 2019; Li X. et al., 2018; Yang et al., 2018; Zhang et al.,
2019). There is thus medium confidence that monsoon systems were stronger during the Pliocene.

Simulations of MPWP climate show that global tropical cyclone intensity and duration increased during the
MPWP (Yan et al., 2016); however, there is low confidence in this result because inter-model variability is
high.

**Summary.** During the MPWP (3.3–3.0 Ma) the atmospheric CO$_2$ concentration was similar to present, and
the slow-response, large-scale indicators reflect a world that was warmer than present. With very high
confidence, relative to present, global surface temperature, sea level, and precipitation rate were higher, NH
latitudinal temperature gradient was lower, and major terrestrial biomes expanded poleward. With medium
confidence from proxy-based evidence alone (Section 2.3.2), combined with numerical modelling, analysis
of the sea-level budget, and process understanding (Section 9.6.2), there is high confidence that cryospheric
indicators were diminished. There is medium confidence that the Pacific longitudinal temperature gradient
was weaker and monsoon systems were stronger.

[START CROSS-CHAPTER BOX 2.4, FIGURE 1 HERE]

Cross-Chapter Box 2.4, Figure 1: Climate indicators of the mid-Pliocene Warm Period (3.3–3.0 Ma) from models
and proxy data. (a) Simulated surface air temperature (left) and precipitation rate anomaly (right) anomaly (relative to 1850–1900) from the Pliocene Model
Intercomparison Project Phase 2 multi-model mean, including CMIP6 (n = 12) and
non-CMIP6 (n = 12) models. Symbols represent site-level proxy-based estimates of
sea-surface temperature for KM5c (n = 32), and terrestrial temperature (n = 8) and
precipitation rate for the MPWP (n = 8). (b) Distribution of terrestrial biomes was
considerably different during the Piacenzian Stage (3.6–2.6 Ma) (upper) compared
with present-day (lower). Biome distributions simulated with a model (BIOME4) in
which Pliocene biome classifications are based on 208 locations, with model-
predicted biomes filling spatial gaps, and the present day, with the model adjusted
for CO$_2$ concentration of 324 ppm. (c) Ice-sheet extent predicted using modelled
climate forcing and showing where multiple models consistently predict the former
presence or absence of ice on Greenland (n = 8 total) and Antarctica (n = 10 total).
Further details on data sources and processing are available in the chapter data table
(Table 2.SM.1).
2.4 Changes in modes of variability

Modes of climate variability that are important for large-scale climate on interannual and longer timescales (Cross-Chapter Box 2.2) are assessed herein and defined and summarised in Annex IV. Though the modes of variability discussed here are assessed and classified on the basis of physical variables, it is important to recognise that the distribution and function of various components of the terrestrial and marine biospheres are modified in response to them.

2.4.1 Annular modes

2.4.1.1 Northern Annular Mode (NAM) / North Atlantic Oscillation (NAO)

AR5 reported that the shift towards a positive NAO, a mode of variability in the North Atlantic that is closely related to the hemispheric-scale NAM, from the 1950s to the 1990s was largely reversed by more recent changes (high confidence). Moreover, periods of persistent negative or positive NAO states observed during the latter part of the 20th century were not unusual, based on NAO reconstructions spanning the last half millennium (high confidence).

New multi-millennial-timescale NAO reconstructions are derived from marine and lake sediments, speleothems, tree rings and ice cores (Ortega et al., 2015; Faust et al., 2016; Sjolte et al., 2018). NAO variability over the past 8 kyr suggests the presence of a significant 1.5 kyr periodicity (Darby et al., 2012). Positive NAO conditions dominated during the MH, while the prevailing NAO sign during the early and late Holocene was negative according to most reconstructions (Brahim et al., 2019; Hernández et al., 2020; Olsen et al., 2012; Røthe et al., 2019). For the CE, reconstructions developed since AR5 indicate no dominant NAO phase during 1000–1300 CE (Baker et al., 2015; Hernández et al., 2020; Jones et al., 2014; Lasher & Axford, 2019; Ortega et al., 2015), with either negative (Baker et al., 2015; Faust et al., 2016; Mellado-Cano et al., 2019) or a more variable phase of the NAO (Jones et al., 2014; Ortega et al., 2015; Sjolte et al., 2018; Cook et al., 2019) between 1400 and 1850 CE. Several instrument-based NAO reconstructions extending back to the 17th and 18th centuries highlight the presence of multidecadal variations in the NAO phases (Corones et al., 2013; Cropper et al., 2015), although these studies have limitations considering the seasonality of the centres of action and the locations of the stations used. Recent reconstructions of the large scale sea level pressure field yield more robust NAO analysis, showing a persistently negative NAO phase from the 1820s to the 1870s, with positive values dominating during the beginning of the 20th century followed by a declining trend over 1920–1970, with a recovery thereafter to a period of consistently high values between 1970 and the early 1990s (Délaygue et al., 2019; Mellado-Cano et al., 2019). Based on the evaluation of several NAO reconstructions for recent centuries, Hernández et al. (2020) highlighted that the strong positive NAO phases of the 1990s and early 21st century were not unusual.

The predominantly positive phase during the 1990s was followed by partial reversal and a tendency towards stronger variability in boreal winter NAM and NAO since the late 1990s (Hanna et al., 2018; Pinto & Raible, 2012). This is particularly evident in December NAO (Hanna et al., 2015) and NAM (Overland and Wang, 2015) indices, and is not unusual on multidecadal time scales (Woolings et al., 2018a). Since the 1990s, a statistically significant summer NAO decline was reported, which is, to a lesser extent, also evident in the winter NAO, linked to an enhanced blocking activity over Greenland (Hanna et al., 2015; 2016; Wachowicz et al., 2021). However, this was moderated by a persistent positive NAO values since 2015 (Annex IV 2.1). Based on observations and reanalysis datasets, multidecadal fluctuations were found for the NAM patterns: the Atlantic centre remained unchanged throughout 1920–2010, whereas the Pacific centre was stronger during 1920–1959 and 1986–2010 and weaker during 1960–1985 (Gong et al., 2018). Multidecadal changes were also observed in the position of the centres of action of the NAO (Moore et al., 2013; Zuo et al., 2016).
In summary, positive trends for the NAM / NAO winter indices were observed between the 1960s and the early 1990s, but these indices have become less positive or even negative thereafter (high confidence). The NAO variability in the instrumental record was very likely not unusual in the millennial and multi-centennial context.

2.4.1.2 Southern Annular Mode (SAM)

AR5 concluded that it was likely that the SAM had become more positive since the 1950s and that this increase was unusual in the context of the prior 400 years (medium confidence). Both AR5 and SROCC reported statistically significant trends in the SAM during the instrumental period for the austral summer and autumn.

Several studies have attempted to reconstruct the evolution of the SAM during the Holocene using proxies of the position and strength of the zonal winds, although with no clear consensus regarding the timing and phase of the SAM (Hernández et al., 2020). The early Holocene was dominated by SAM positive phases (Moreno et al., 2018; Reynhout et al., 2019), consistent with increasing westerly wind strength (Lamy et al., 2010), with some reconstructions showing significant centennial and millennial variability but no consistent trend after 5 ka (Hernández et al., 2020). For the CE, enhanced westerly winds occurred over 0–1000 CE, as reflected in increased burning activity in Patagonia (Turney et al., 2016a) and tree ring records from southern New Zealand (Turney et al., 2016b), implying a predominantly positive SAM phase. Pollen records and lake sediments from Tasmania, southern mainland Australia, New Zealand and southern South America, inferred the period of 1000 to 1400 CE to be characterized by anomalously dry conditions south of 40°S, implying a positive SAM (Evans et al., 2019; Fletcher et al., 2018; Matley et al., 2020; Moreno et al., 2014).

Nevertheless, proxy reconstructions of the SAM based on temperature-sensitive records from tree rings, ice cores, lake sediments and corals spanning the mid-to-polar latitudes show altering positive and negative phases (Figure 2.35).

Prolonged periods of negative SAM values were identified during the period 1400–1700 CE in several reconstructions (Figure 2.35, Abram et al., 2014; Dätwyler et al., 2018; Villalba et al., 2012), with a minimum identified during the 15th century (Hernández et al., 2020), although some disagreements exist between proxy records before 1800 CE (Hessl et al., 2017). Abram et al. (2014) concluded that the mean SAM index during recent decades is at its highest levels for at least the last 1 kyr. Similarly, the summer SAM reconstruction by Dätwyler et al. (2018) indicates a strengthening over the last 60 years that is outside the very likely range of the last millennium natural variability. The largest 30- and 50-year trends in the annual SAM index occurred at the end of the 20th century (after 1969 and 1950 respectively), indicating that the recent increase in the SAM is unprecedented in the context of at least the past three centuries (Yang & Xiao, 2018).

Before the mid-1950s, SAM indices derived from station-based datasets, and centennial reanalyses show pronounced interannual and decadal variability but no significant trends, with low correlation between SAM indices due to the diversity across different datasets and sensitivity to the definition used for the index calculation (Barrucand et al., 2018; Lee J. et al., 2019a; Schneider & Fogt, 2018). Various SAM indices exhibit significant positive trends since the 1950s, particularly during austral summer and autumn (Barrucand et al., 2018; Lee J. et al., 2019a; Schneider & Fogt, 2018), unprecedented for austral summer over the last 150 years (Fogt & Marshall, 2020; Jones et al., 2016b). This indicates a strengthening of the surface westerly winds around Antarctica, related to both the position and intensity of the subpolar jet in the SH (IPCC, 2019b; Ivy et al., 2017) (Section 2.3.1.4.3). The SAM trends have slightly weakened after about 2000 (Fogt and Marshall, 2020).

[START FIGURE 2.35 HERE]

Figure 2.35: Southern Annular Mode (SAM) reconstruction over the last millennium. (a) SAM reconstructions as 7-year moving averages (thin lines) and 70-year LOESS filter (thick lines). (b) observed SAM index

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AR5 reported with medium confidence that ENSO-like variability existed, at least sporadically, during epochs of millions of years (including the MPWP; Chapter Box 2.4), with proxy records indicating that this was the case even when cross-Pacific SST gradients were much weaker than present. There is substantial disagreement between proxy records for ENSO activity during the early Holocene (White et al., 2018; Zhang et al., 2014), and for ENSO activity and mean state at the LGM (Ford et al., 2015; 2018; Koutavas & Joanides, 2012; Leduc et al., 2009; Sadekov et al., 2013; Tierney et al., 2020; Zhu et al., 2017). A number of studies (Carré et al., 2014; Cobb et al., 2013; Emile-Geay et al., 2016; Grothe et al., 2019; McGregor H. et al., 2013; Thompson et al., 2017; Tian et al., 2017; White et al., 2018) have found that ENSO was substantially weaker than at present at various times in the mid-Holocene within the period from 6 to 3 ka, with stronger decreases in variability revealed by remote proxies than by those close to the core region of ENSO activity. However, Karamperidou et al. (2015) find that weakening in ENSO-related variability in eastern Pacific proxies does not necessarily correspond to weakening in central Pacific proxies. Barrett et al. (2018) concluded that multi-proxy reconstructions are more efficient at identifying eastern Pacific than central Pacific events. This suggests that a weakening of proxy-based signals may indicate an along-equatorial shift in ENSO activity rather than a weakening of ENSO during some periods. Following the period of weak ENSO variability in the mid-Holocene, a number of studies find an increase in ENSO activity which, depending upon the study, commences between 4.4 and 3 ka (Chen et al., 2016; Cobb et al., 2013; Du et al., 2021; Emile-Geay et al., 2016; Koutavas & Joanides, 2012; Thompson et al., 2017; Zhang et al., 2014). Numerous studies (Hope et al., 2017; Li J. et al., 2013; Liu Y. et al., 2017; McGregor S. et al., 2013; Rustic et al., 2015) find substantial variability in ENSO activity on multi-decadal to centennial timescales over the last 500 to 1 kyr (Figure 2.36). Different proxies show a wide spread in the specific timing and magnitude of events in the pre-instrumental period (e.g. Dätwyler et al., 2019). Most investigators find that ENSO activity in recent decades was higher than the most recent centuries prior to the instrumental period. Grothe et al. (2019) also found that ENSO variance of the last 50 years was 25% higher than the average of the last millennium, and was substantially higher than the average of the mid- to late Holocene. McGregor S. et al. (2010, 2013) looked for common variance changes in pre-existing ENSO proxies, finding stronger ENSO variance for the 30-year period 1979–2009 compared to any 30-year period within the timespan 1590–1880 CE. This finding also holds when adding more recently developed ENSO proxies (Figure 2.36). Koutavas & Joanides (2012), Ledru et al. (2013) and Thompson et al. (2017) identify various periods within the range 1000 BCE to 1300 CE when ENSO activity was greater than in the following centuries, and more
closely comparable to the mid-20th century onwards behaviour.

[START FIGURE 2.36 HERE]

Figure 2.36: Reconstructed and historical variance ratio of El Niño–Southern Oscillation (ENSO). (a) 30-year running variance of the reconstructed annual mean Niño 3.4 or related indicators from various published reconstructions. (b) variance of June-November Southern Oscillation Index (SOI) and April-March mean Niño 3.4 (1981–2010 base period) along with the mean reconstruction from (a). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

[END FIGURE 2.36 HERE]

Since AR5, updates to datasets used widely in prior ENSO assessments resulted in substantial and important revisions to observed tropical Pacific SST data (Section 2.3.1.1). In particular, ERSSTv4, and then ERSSTv5, addressed known SST biases in ERSSTv3 in the equatorial Pacific which affected the derived mean state and amplitude of indices based on that dataset (Huang et al., 2015). During the instrumental period, there is no robust indication of any significant century-scale trend in the east-west SST gradient across the equatorial Pacific Ocean, with periods when gradients have been stronger and weaker than the long-term average on decadal timescales, associated with a predominance of La Niña or El Niño events respectively. The frequency of El Niño and La Niña events is also subject to considerable decadal variability (e.g. Hu et al., 2013) but with no indication of a long-term signal in the frequency of events. The ENSO amplitude since 1950 has increased relative to the 1910–1950 period, as confirmed by independent proxy records (e.g., Gergis & Fowler, 2009), the Southern Oscillation Index (SOI) (Braganza et al., 2009) and SSTs (e.g., Obba, 2013; Yu & Kim, 2013), although there is a spread between different proxy and instrumental sources as to the magnitude of that increase (Figure 2.36). The El Niño events of 1982–1983, 1997–1998 and 2015–2016 had the strongest anomalies in the Niño 3.4 SST index since 1950. Their predominance was less evident from indices based on de-trended data such as the Oceanic Niño Index (ONI) (which still ranked them as the three strongest events since 1950, but only by a small margin), and in the SOI. Huang B. et al. (2019d) also note that analyses based upon buoy and Argo data, which are only available since the 1990s, are more capable of resolving strong events than analyses which do not include such data.

Prior to the 1950s, SST observations in the tropical Pacific were much sparser and hence uncertainties in Niño indices are much larger (Huang B. et al., 2020). SOI data and some newer SST-based studies show high ENSO amplitude, comparable to the post-1950 period, in the period from the mid-late 19th century to about 1910, but proxy indicators generally indicate that the late 19th and early 20th century were less active than the late 20th century (Figure 2.36). Yu & Kim’s (2013) implementation of the ONI found a number of events with the ONI above 1.5°C between 1888 and 1905, then no such events until 1972, whilst the SOI indicates comparable or stronger events to the three strongest post-1950 events in 1896 and 1905. Giese & Ray (2011) also found a number of such events between 1890 and 1920 in the SODA ocean reanalysis, corroborated further by Huang B. et al. (2020a) and Vaccaro et al. (2021), who found that the strength of the 1877–1878 event is comparable with that of the 1982–1983, 1997–1998 and 2015–2016 events. There have also been a number of strong La Niña events (e.g. 1973–1974, 1975–1976 and 2010–2011), with few clear analogues in the 1920–1970 period; the proxy-based analysis of McGregor et al. (2010) indicates that the mid-1970s La Niña period was also extreme in a multi-centennial context. There is no indication that the frequency of high-amplitude events since the 1970s reflects a long-term trend which can be separated from multi-decadal variability, given apparent presence of several high-amplitude events in the late 19th and early 20th centuries, and the relatively large uncertainty in pre-1950 SST data in the tropical Pacific region.

There is a distinction (Annex IV 2.3.1) between El Niño events centred in the eastern Pacific (‘Eastern Pacific’ (EP) or ‘classical’ events) and those centred in the Central Pacific (‘Central Pacific’ (CP) or ‘Modoki’ events), which have different typical teleconnections (e.g. Ashok et al., 2007; Capotondi et al., 2015; Ratnam et al., 2014; Timmermann et al., 2018). A number of studies, using a range of indicators, have found an increase in recent decades of the fraction of CP El Niño events, particularly after 2000 (Yu and
There is no robust indication of any changes in ENSO teleconnections over multi-centennial timescales (Hernández et al., 2020) despite multi-decadal variability. Shi and Wang (2018) found that teleconnections with the broader Asian summer monsoon, including the Indian and the East Asian monsoon, were generally stable since the 17th century during the developing phase of the monsoon, and showed substantial decadal variability, but no clear trend, during the decaying phase. They also found that the weakening of teleconnections between the Indian summer monsoon and ENSO in recent decades had numerous precedents over the last few centuries. Räsänen et al. (2016) also found substantial decadal variability, but little trend, in the strength of the relationship between ENSO and monsoon precipitation in South-East Asia between 1650 and 2000. Dütwyler et al., (2019) identified a number of multi-decadal periods with apparently changed teleconnections at times over the last 400 years.

In the instrumental period, teleconnections associated with ENSO are well known to vary on decadal to multi-decadal timescales (e.g. Ashcroft et al., 2016; He et al., 2013; Jin et al., 2016; Lee & Ha, 2015; Wang Q. et al., 2019). Yun and Timmermann (2018) found that decadal variations in teleconnections between ENSO and the Indian monsoon did not extend beyond what would be expected from a stochastic process. Many observed decadal changes in teleconnections in the instrumental period are consistent with a shift to more central Pacific El Niño events (Evtushevsky et al., 2018; Yeh et al., 2018; Yu & Sun, 2018; Zhao & Wang, 2019). Effects of the PDV (Kwon et al., 2013; Wang S. et al., 2014; Dong et al., 2018) and the AMV (Kayano et al., 2019) can also modulate ENSO teleconnections, and affect the frequency of CP versus EP events (Ashok et al., 2007). Chiodi and Harrison (2015) found that teleconnections over the most recent decades are broadly consistent with those over the last 100 years. Variability in teleconnections can also occur on timescales longer than characteristic PDV timescales (e.g. Gallant et al., 2013).

In summary, there is medium confidence that both ENSO amplitude and the frequency of high-magnitude events since 1850 are higher than over the period from 1850 as far back as 1400, but low confidence that they are outside the range of variability over periods prior to 1400, or higher than the average of the Holocene as a whole. Overall, there is no indication of a recent sustained shift in ENSO or associated features such as the Walker Circulation, or in teleconnections associated with these, beyond the range of variability on decadal to millennial timescales. A high proportion of El Niño events in the last 20–30 years has been based in the central, rather than eastern Pacific, but there is low confidence that this represents a long-term change.

### 2.4.3 Indian Ocean Basin and dipole modes

AR5 did not provide an assessment of the Indian Ocean Dipole (IOD) records based on paleo reconstructions. For the instrumental era, AR5 reported that there were no trends in the IOD behaviour. However, the strength of the Indian Ocean Basin-wide (IOB) mode, quantified by the basin-scale averaged SST index, increased in all assessed periods except 1979–2012, but the AR5 neither quantified trends nor provided a confidence statement.

For the LGM, enhanced equatorial Indian Ocean productivity in marine sediment records was associated with strengthened westerly jets, in line with a shallower central-western Indian Ocean thermocline and stronger negative IOD events (Punyu et al., 2014). The LGM Indian Ocean basin was substantially modified by the exposure of the tropical shelves (DiNezio et al., 2018) and this has been associated with an Indian Ocean "El Niño" (Thirumalai et al., 2019). Wurtzel et al. (2018) contend that during the LDT, including the
Younger Dryas event, Indian Ocean precipitation did not mirror the zonal asymmetry observed in Indian Ocean SSTs in the Holocene based upon a speleothem record from Sumatra, but instead reflected shifts in moisture transport pathways and sources. Using Seychelles corals, representing the western pole of the IOD, spanning from the MH to present, Zinke et al. (2014) identified changes in seasonality, with the lowest seasonal SST range in the MH and then again around 2 ka, while the largest seasonal range occurred around 4.6 ka and then again during the near-present (1990–2003). Reconstructions from fossil corals for the eastern Indian Ocean point to stronger negative IOD SST anomalies due to the enhanced upwelling and cooling driven by a stronger monsoon with enhanced anomalous easterly winds in the eastern Indian Ocean during the MH (Abram et al., 2020a). Niedermeyer et al. (2014) from the analysis of stable isotopes in terrestrial plant waxes, suggest that the period 6.5 ka to 4.5 ka was characterized by an anomalously positive IOD mean state. During various parts of the Holocene, periods of a mean positive IOD-like state were associated with increased IOD variability, including events that exceed the magnitude of the strongest events during the instrumental period (Abram et al., 2020b).

From the coral δ¹⁸O record from the Seychelles over 1854–1994, Du et al. (2014) showed a 3–7 year dominant period associated with the IOB in response to ENSO forcing. They identified multi-decadal variability in the IOB, with more active IOB phases during 1870–1890, 1930–1955, and 1975–1992, while decadal variability in the IOB dominated during 1940–1975 (Du Y. et al., 2014). Evidence for changes in IOB characteristics during earlier periods (e.g., MH, LGM) is limited.

The role of decadal to multi-decadal variability has recently emerged as an important aspect of the IOD with many indications of the effects of Pacific Ocean processes on IOD variability through atmospheric and oceanic mechanisms (Dong et al., 2016; Jin et al., 2018; Krishnamurthy & Krishnamurthy, 2016; Zhou et al., 2017). Positive events in the 1960s and 1990s were linked to a relatively shallow eastern Indian Ocean thermocline, and a primarily negative IOD state in the 1970s and 1980s was related to a deeper thermocline (Ummenhofer et al., 2016). Positive IOD events may have increased in frequency during the second half of the 20th century (Abram et al., 2020a,b). Earlier observations of apparent changes in the frequency and / or magnitude of the IOD events are considered unreliable, particularly prior to the 1960s (Hernández et al., 2020). Although the seasonal evolution and the type of ENSO (Section 2.4.2) may influence the character of the IOD (Guo et al., 2015; Zhang et al., 2015; Fan et al., 2016), the occurrence of some IOD events may be independent of ENSO (Sun et al., 2015).

To summarize, there is low confidence in any multi-decadal IOD variability trend in the instrumental period due to data uncertainty especially before the 1960s. In addition to data uncertainty, understanding of the IOD variability during the instrumental period is also limited by large-scale warming of the Indian Ocean. Neither the IOD nor the IOB have exhibited behaviour outside the range implied by proxy records (low confidence).

**[START FIGURE 2.37 HERE]**

Figure 2.37: Indices of interannual climate variability from 1950–2019 based upon several sea surface temperature based data products. Shown are the following indices from top to bottom: IOB mode, IOD, Niño4, AMM and AZM. All indices are based on area-averaged annual data (see Technical Annex IV). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

**[END FIGURE 2.37 HERE]**

### 2.4.4 Atlantic Meridional and Zonal Modes

AR5 reported no changes in the Atlantic Meridional Mode (AMM) during the 20th century or shorter periods thereof. For the Atlantic Zonal Mode (AZM), also referred as the Atlantic Niño, the AR5 reported increases during the 1950–2012 period but neither assessed trends nor provided a confidence statement. The AR5 did not assess paleo evidence for the AZM and AMM.

Paleo-reconstructions of these two modes remain rather limited. Nonetheless, the interhemispheric cross-
equatorial SST gradients linked to changes in ITCZ locations characteristic of the AMM has been found during the LGM, Heinrich Stadial 1 and the MH, with the largest shift occurring during HS1 (McGee et al., 2014). Similarly, the dipole-like SST pattern in the South Atlantic subtropics, which is related to the AZM (Foltz et al., 2019; Lübbecke et al., 2018; Moriya et al., 2011; Nnamchi et al., 2016, 2017; Rouault et al., 2018; Venegas et al., 1996), has been reconstructed using SST proxies from marine sediment cores during the past 12 kyr (Wainer et al., 2014). The reconstructed index captures two significant cold events that occurred during the 12.9–11.6 ka and 8.6–8.0 ka periods in the South Atlantic (Wainer et al., 2014).

During the observational period the AZM and AMM (Figure 2.37) are related to the AMV largely controlling the interhemispheric gradient of the SST at decadal to multi-decadal timescales (Li et al., 2015; Lübbecke et al., 2018; Polo et al., 2013; Svendsen et al., 2014; Tokinaga & Xie, 2011). The AZM interannual variability is enhanced (Foltz et al., 2019; Lübbecke et al., 2018), and is more strongly related to ENSO during the negative phase of the AMV (Martín-Rey et al., 2014, 2018; Polo et al., 2015; Nnamchi et al., 2020). The AZM displayed a persistent weakened variability over the 1960–2009 period associated with declined cold tongue upwelling (Tokinaga and Xie, 2011), which became pronounced since 2000 (Prigent et al., 2020a,b). Despite these multi-decadal fluctuations, there is limited evidence for any sustained change in the AMM (Chang et al., 2011; Martín-Rey et al., 2018) and AZM (Martín-Rey et al., 2018; Nnamchi et al., 2020) during the instrumental period. The AZM and AMM interact on interannual timescales (Foltz & McPhaden, 2010; Pottapinjara et al., 2019; Servain et al., 1999) leading in 2009 to extremes of both modes in which the negative phase of the AMM (Burmeister et al., 2016) (Foltz et al., 2012) preceded an equatorial cold tongue cold event that was unprecedented in the prior 30 years (Burmeister et al., 2016; Foltz & McPhaden, 2010).

In summary, confidence is low in any sustained changes to the AZM and AMM variability in instrumental observations. There is very low confidence in changes of the paleo AZM and AMM due to extremely limited availability of paleo reconstructions.

2.4.5 Pacific Decadal Variability

Pacific Decadal Variability (PDV) refers to the ocean-atmosphere climate variability over the Pacific Ocean at decadal-to-interdecadal time scales and is usually described by the Pacific Decadal Oscillation (PDO) or the Interdecadal Pacific Oscillation (IPO) indices. AR5 and SROCC reported a large shift of the PDO in the late 1970s, with a predominantly positive phase until the end of the 1990s, being mainly negative afterwards. There was no significant change assessed in the PDO during the instrumental period as a whole, and no confidence level was assigned. Changes in the pre-instrumental era PDO were not assessed in AR5.

The existence of the PDV in the centuries prior to the instrumental period is evidenced by a variety of proxy records based on tree rings (Biondi et al., 2001; D’Arrigo and Ummerhofer, 2015), corals (Deng et al., 2013; Felis et al., 2010; Linsley et al., 2015) and sediments (Lapointe et al., 2017; O’Marra et al., 2019). There is little coherence between the various paleo-proxy indices prior to the instrumental record, and neither these nor the instrumental records provide indications of a clearly defined spectral peak (Buckley et al., 2019; Chen & Wallace, 2015; Henley, 2017; Newman M. et al., 2016; Zhang L. et al., 2018). For instance, spectral analysis from millennia length PDV reconstructions shows spectral peaks at multi-decadal, centennial and bi-centennial time scales (Beaufort and Grelaud, 2017), while only multi-decadal oscillations can be detected in the shorter (less than 400 years into the past) paleoclimate reconstructions. A variety of proxies suggest a shift in the PDV from the early-mid Holocene, which was characterized by a persistently negative phase of the PDO (i.e. weak Aleutian Low), to the late Holocene, and a more variable and more positive PDO (i.e. strong Aleutian Low) conditions. This shift at around 4.5 ka is also evident in the PDO periodicities, changing from bidecadal and pentadecadal variability in the early Holocene to only pentadecadal periodicities in the late Holocene (Hernández et al., 2020). Several proxy records indicate that the strengthening in the Aleutian Low inferred since the late 17th century is unprecedented over the last millennium (Liu Z. et al., 2017; Osterberg et al., 2017; Winski et al., 2017), in line with an increase in PDV low-frequency variability (Hernández et al., 2020; Williams et al., 2017).
The PDO and IPO indices are significantly correlated during the instrumental period, showing regime shifts in the 1920s, 1940s, 1970s and around 1999. Positive PDV phases were observed from the 1920s to the mid-1940s and from the late 1970s to the late 1990s, while negative phases occurred from mid-1940s until the late 1970s, and since 1999 (Figure 2.38; Han et al., 2013; Chen and Wallace, 2015; Newman M. et al., 2016). The associated spatial patterns are quite similar, but the PDO pattern exhibits stronger SST anomalies in the extra-tropical North Pacific than the IPO (Chen & Wallace, 2015). The strength and structure of the SST patterns also differ among the periods (Newman M. et al., 2016).

Figure 2.38: Indices of multi-decadal climate variability from 1854–2019 based upon several sea surface temperature based data products. Shown are the indices of the AMV and PDV based on area averages for the regions indicated in Technical Annex IV. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).

In summary, the PDV in the instrumental record is dominated by (multi-) decadal-scale shifts between positive and negative phases over the last 150 years with no overall trend (high confidence). There is low confidence in the classification of the PDO / IPO phase among several indices, even during recent years with the availability of high-quality data. Nevertheless, the teleconnection patterns are robust regardless of the index used to characterize the PDO (McAfee, 2017). Analysis of time series of PDO and IPO highlights the (multi-) decadal nature of this mode of variability with no significant trends, but highlights a recent switch from a positive to a negative phase since 1999 / 2000 across all indicators (England et al., 2014; Henley, 2017).

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2.4.6 Atlantic Multidecadal Variability

AR5 reported no robust changes in Atlantic Multidecadal Variability (AMV) reconstructions based on paleo records due to low consistency between different AMV reconstructions prior to 1900. AR5 concluded that there have been no significant trends in the AMV index during the instrumental period and there was difficulty in interpreting the AMV signal because of the long-term underlying SST warming trend.

The AR5 conclusions about large uncertainties in AMV paleo reconstructions (Hernández et al., 2020) have been reinforced by recent studies of tree rings (Wang et al., 2017b), Greenland ice (Chylek et al., 2012), and corals (Kilbourne et al., 2014; Svendsen et al., 2014a; Wang et al., 2017b). The AMV exhibited a generally positive state over the first millennium of the CE (Mann et al., 2009; Singh et al., 2018). Paleow reconstructions over the last millennium consistently show a negative AMV phase during 1400–1850 CE and a positive phase during 900–1200 CE (Mann et al., 2009; Singh et al., 2018; Wang J. et al., 2017), consistent with warmer surface temperatures from tropical Atlantic records (Kilbourne et al., 2014).

Instrumental observations show that AMV is characterized by basin-wide warm and cool periods with an average variation in SST of about 0.4°C, but with larger variations in the North Atlantic subpolar gyre. Despite small differences in indices used to define the AMV (Annex IV), they all show warm periods occurring approximately between 1880–1900, 1940–1960, and from the mid-1990s to present, with cool periods in between (Figure 2.38) but no overall sustained change during the instrumental period (Bellomo et
al., 2018; Booth et al., 2012; Gulev et al., 2013; Sun et al., 2017).

The oceanic changes are seen in salinity and temperature variations over the upper 3000 m of the North Atlantic (Polyakov et al., 2005; Keenlyside et al., 2015), and in sea level variations in the Western North Atlantic along the Gulf Stream passage (McCarthy et al., 2015). The pattern and strength of the AMV differs among the periods (e.g., Svendsen et al., 2014a; Reynolds et al., 2018) and there are indications that there may have been a shift since 2005 toward a negative phase of the AMV (Robson et al., 2016).

In summary, no sustained change in AMV indices has been observed over the instrumental period (high confidence). However, instrumental records may not be long enough to distinguish any oscillatory behaviour from trends in the AMV. There is low confidence in the paleo AMV reconstructions due to a paucity of high-resolution records.

2.5 Final remarks

The assessment in this chapter is based upon an ever-expanding volume of available proxy and observational records and a growing body of literature. Since AR5, improvements have been made in all aspects of data collection, data curation, data provision and data analysis permitting improved scientific insights. The chapter has also benefitted from the availability of new products from a range of emerging, mainly space-based, observing capabilities and new generations of reanalysis products. However, a number of key challenges still remain which, if addressed, would serve to strengthen future reports.

- Development of new techniques and exploitation of existing and new proxy sources may help address challenges around the low temporal resolution of most palaeoclimate proxy records, particularly prior to the Common Era, and ambiguities around converting palaeoclimate proxy data into estimates of climate-relevant variables. Conversions rely upon important proxy-specific assumptions and biases can be large due to limited accounting of seasonality, non-climatic effects, or the influence of multiple climate variables. These challenges currently limit the ability to ascertain the historical unusualness of recent directly observed climate changes for many indicators. {2.2, 2.3}

- Improved sharing of historical instrumental and proxy records (including metadata) along with significant efforts at data rescue of presently undigitised records would serve to significantly strengthen many aspects of the present assessment. Longer observational time series help to better understand variability, and any underlying periodicity in climate indicators and climate forcers, especially for components of the climate system where the dominant response is on multi-decadal and longer time scales. Targeted rescue of early marine data records, especially from waters of the tropical oceans and the Southern Ocean could help constrain modes of variability and important teleconnection changes in, for example, the hydrological cycle and patterns of global temperature change. {2.2, 2.3, 2.4}

- Contemporary observing systems have limitations in parts of the Earth system with limited accessibility or coverage, for example coastal and shallow ocean, polar regions, marine biosphere, and the deep sea, leading to seasonal-dependency and regional variability in data coverage and uncertainty which serve to limit confidence in the present assessment of changes. {2.3}

- Contradictory lines of evidence exist between observations and models on the relationship between the rates of warming in GMST and GSAT, compounded by limitations in theoretical understanding. Improvements in air temperature datasets over the ocean and an improved understanding of the representation of the lowermost atmosphere over the ocean in models would reduce uncertainty in assessed changes in GSAT. {Cross-Chapter Box 2.3}
Frequently asked questions

FAQ 2.1: The Earth’s temperature has varied before. How is the current warming any different?

_Earth’s climate has always changed naturally, but both the global extent and rate of recent warming are unusual. The recent warming has reversed a slow, long-term cooling trend, and research indicates that global surface temperature is higher now than it has been for millennia._

While climate can be characterised by many variables, temperature is a key indicator of the overall climate state, and global surface temperature is fundamental to characterising and understanding global climate change, including Earth’s energy budget. A rich variety of geological evidence shows that temperature has changed throughout Earth’s history. A variety of natural archives from around the planet, such as ocean and lake sediments, glacier ice and tree rings, shows that there were times in the past when the planet was cooler, and times when it was warmer. While our confidence in quantifying large-scale temperature changes generally decreases the farther back in time we look, scientists can still identify at least four major differences between the recent warming and those of the past.

_It’s warming almost everywhere._ During decades and centuries of the past 2000 years, some regions warmed more than the global average while, at the same time, other regions cooled. For example, between the 10th and 13th centuries, the North Atlantic region warmed more than many other regions. In contrast, the pattern of recent surface warming is globally more uniform than for other decadal to centennial climate fluctuations over at least the past two millennia.

_It’s warming rapidly._ Over the past 2 million years, Earth’s climate has fluctuated between relatively warm interglacial periods and cooler glacial periods, when ice sheets grew over vast areas of the northern continents. Intervals of rapid warming coincided with the collapse of major ice sheets, heralding interglacial periods such as the present Holocene Epoch, which began about 12,000 years ago. During the shift from the last glacial period to the current interglacial, the total temperature increase was about 5°C. That change took about 5000 years, with a maximum warming rate of about 1.5°C per thousand years, although the transition was not smooth. In contrast, Earth’s surface has warmed approximately 1.1°C since 1850–1900. However, even the best reconstruction of global surface temperature during the last deglacial period is too coarsely resolved for direct comparison with a period as short as the past 150 years. But for the past 2000 years, we have higher-resolution records that show that the rate of global warming during the last 50 years has exceeded the rate of any other 50-year period.

_Recent warming reversed a long-term global cooling trend._ Following the last major glacial period, global surface temperature peaked by around 6500 years ago, then slowly cooled. The long-term cooling trend was punctuated by warmer decades and centuries. These fluctuations were minor compared with the persistent and prominent warming that began in the mid-19th century when the millennial-scale cooling trend was reversed.

_It’s been a long time since it’s been this warm._ Averaged over the globe, surface temperatures of the past decade were probably warmer than when the long cooling trend began around 6500 years ago. If so, we need to look back to at least the previous interglacial period, around 125,000 years ago, to find evidence for multi-centennial global surface temperatures that were warmer than now.

Previous temperature fluctuations were caused by large-scale natural processes, while the current warming is largely due to human causes (see, for example, FAQ 1.3, FAQ 3.1). But understanding how and why temperatures have changed in the past is critical for understanding the current warming and how human and natural influences will interact to determine what happens in the future. Studying past climate changes also makes it clear that, unlike previous climate changes, the effects of recent warming are occurring on top of stresses that make humans and nature vulnerable to changes in ways that they have never before experienced (for example, see FAQ 11.2, FAQ12.3).
[START FAQ 2.1, FIGURE 1 HERE]

FAQ 2.1, Figure 1: Evidence for the unusualness of recent warming.

[END FAQ 2.1, FIGURE 1 HERE]

FAQ 2.2: What is the evidence for climate change?

The evidence for climate change rests on more than just increasing surface temperatures. A broad range of indicators collectively leads to the inescapable conclusion that we are witnessing rapid changes to many aspects of our global climate. We are seeing changes in the atmosphere, ocean, cryosphere, and biosphere. Our scientific understanding depicts a coherent picture of a warming world.

We have long observed our changing climate. From the earliest scientists taking meteorological observations in the 16th and 17th centuries to the present, we have seen a revolution in our ability to observe and diagnose our changing climate. Today we can observe diverse aspects of our climate system from space, from aircraft and weather balloons, using a range of ground-based observing technologies, and using instruments that can measure to great depths in the ocean.

Observed changes in key indicators point to warming over land areas. Global surface temperature over land has increased since the late 19th century, and changes are apparent in a variety of societally relevant temperature extremes. Since the mid-1950s the troposphere (i.e., the lowest 6–10 km of the atmosphere) has warmed, and precipitation over land has increased. Near-surface specific humidity (i.e., water vapour) over land has increased since at least the 1970s. Aspects of atmospheric circulation have also evolved since the mid-20th century, including a poleward shift of mid-latitude storm tracks.

Changes in the global ocean point to warming as well. Global average sea surface temperature has increased since the late 19th century. The heat content of the global ocean has increased since the 19th century, with more than 90% of the excess energy accumulated in the climate system being stored in the ocean. This ocean warming has caused ocean waters to expand, which has contributed to the increase in global sea level in the past century. The relative acidity of the ocean has also increased since the early 20th century, caused by the uptake of carbon dioxide from the atmosphere, and oxygen loss is evident in the upper ocean since the 1970s.

Significant changes are also evident over the cryosphere – the portion of the Earth where water is seasonally or continuously frozen as snow or ice. There have been decreases in Arctic sea ice area and thickness and changes in Antarctic sea ice extent since the mid-1970s. Spring snow cover in the Northern Hemisphere has decreased since the late-1970s, along with an observed warming and thawing of permafrost (perennially frozen ground). The Greenland and Antarctic ice sheets are shrinking, as are the vast majority of glaciers worldwide, contributing strongly to the observed sea level rise.

Many aspects of the biosphere are also changing. Over the last century, long-term ecological surveys show that many land species have generally moved poleward and to higher elevations. There have been increases in green leaf area and/or mass (i.e., global greenness) since the early 1980s, and the length of the growing season has increased over much of the extratropical Northern Hemisphere since at least the mid-20th century. There is also strong evidence that various phenological metrics (such as the timing of fish migrations) for many marine species have changed in the last half century.

Change is apparent across many components of the climate system. It has been observed using a very broad range of techniques and analysed independently by numerous groups around the world. The changes are consistent in pointing to a climate system that has undergone rapid warming since the industrial revolution.

[START FAQ 2.2, FIGURE 1 HERE]
FAQ 2.2, Figure 1: Synthesis of significant changes observed in the climate system over the past several decades. Upwards, downwards and circling arrows indicate increases, decreases and changes, respectively. Independent analyses of many components of the climate system that would be expected to change in a warming world exhibit trends consistent with warming. Note that this list is not comprehensive.

[END FAQ 2.2, FIGURE 1 HERE]
Acknowledgements

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2.4.2 El Niño Southern Oscillation

Figure 2.1: Chapter 2 visual abstract of contents.
Cross-Chapter Box 2.1, Figure 1: Global mean surface temperature (GMST) over the past 60 million years relative to 1850–1900 shown on three time scales. Information about each of the nine paleo reference periods (blue font) and sections in AR6 that discuss these periods are listed in Cross-Chapter Box 2.1 Table 1. Grey horizontal bars at the top mark important events. Characteristic uncertainties are based on expert judgement and are representative of the approximate midpoint of their respective time scales; uncertainties decrease forward in time. GMST estimates for most paleo reference periods (Figure 2.34) overlap with this reconstruction, but take into account multiple lines of evidence. Future projections span the range of global surface air temperature best estimates for SSP1–2.6 and SSP5–8.5 scenarios described in Section 1.6. Range shown for 2100 is based on CMIP6 multi-model mean for 2081–2100 from Table 4.5; range for 2300 is based upon an emulator and taken from Table 4.9. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
**Figure 2.2:** Time series of solar and volcanic forcing for the past 2.5 kyr (panels a, c) and since 1850 (panels b, d). (a) Total solar irradiance (TSI) reconstruction (10-year running averages) recommended for CMIP6 / PMIP4 millennial experiments based on the radiocarbon dataset before 1850 (blue) scaled to the CMIP6 historical forcing after 1850 (purple). (b) TSI time series (6-month running averages) from CMIP6 historical forcing as inferred from sunspot numbers (blue), compared to CMIP5 forcing based on (red) and an update to CMIP6 by a TSI composite (orange). (c) Volcanic forcing represented as reconstructed Stratospheric aerosol optical depth (SAOD; as presented in Section 7.3.4.6) at 550 nm. Estimates covering 500 BCE to 1900 CE (green) and 1850–2015 (blue). (d) SAOD reconstruction from CMIP6 (v 4) (blue), compared to CMIP5 forcing (red). Note the change in y-axis range between panels c and d.

Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.3: The evolution of atmospheric CO$_2$ through the last 450 million years. The periods covered are 0–450 Ma (a), 0–58 Ma (b), and 0–3500 ka (c), reconstructed from continental rock, marine sediment and ice core records. Note different timescales and axes ranges in panels (a), (b) and (c). Dark and light green bands in (a) are uncertainty envelopes at 68% and 95% uncertainty, respectively. 100 ppm in each panel is shown by the marker in the lower right hand corner to aid comparison between panels. In panel (b) and (c) the major paleoclimate reference periods (CCB2.1) have been labelled, and in addition: MPT (Mid Pleistocene Transition), MCO (Miocene Climatic Optimum). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.4: Atmospheric WMGHG concentrations from ice cores. (a) Records during the last 800 kyr with the LGM to Holocene transition as inset. (b) Multiple high-resolution records over the CE. The horizontal black bars in the panel a inset indicate Last Glacial Maximum (LGM) and Last Deglacial Termination (LDT) respectively. The red and blue lines in (b) are 100-year running averages for CO₂ and N₂O concentrations, respectively. The numbers with vertical arrows in (b) are instrumentally measured concentrations in 2019. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.5: Globally averaged dry-air mole fractions of greenhouse gases. (a) CO$_2$ from SIO, CSIRO, and NOAA/GML (b) CH$_4$ from NOAA, AGAGE, CSIRO, and UCI; and (c) N$_2$O from NOAA, AGAGE, and CSIRO (see Table 2.2). Growth rates, calculated as the time derivative of the global means after removing seasonal cycle are shown as inset figures. Note that the CO$_2$ series is 1958–2019 whereas CH$_4$ and N$_2$O are 1979–2019. Further details on data are in Annex III, and on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.6: Global mean atmospheric mixing ratios of select ozone-depleting substances and other greenhouse gases. Data shown are based on the CMIP6 historical dataset and data from NOAA and AGAGE global networks. PFCs include CF$_4$, C$_2$F$_6$, and C$_3$F$_8$, and c-C$_4$F$_8$; Halons include halon-1211, halon-1301, and halon-2402; other HFCs include HFC-23, HFC-32, HFC-125, HFC-143a, HFC-152a, HFC-227ea, HFC-236fa, HFC-245fa, and HFC-365mfc, and HFC-43-10mee. Note that the y-axis range is different for a, b and c and a 25 ppt yardstick is given next to each panel to aid interpretation. Further data are in Annex III, and details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Mean total column ozone in six regions

Figure 2.7: Time series of annual mean total column ozone from 1964-2019. Values are in Dobson Units (DU), a good proxy for vertically integrated stratospheric ozone. Time series are shown for (a) near-global domain, (b-d) three zonal bands and (e) polar (60°–90°) total ozone in March (NH) and October (SH); the months when polar ozone losses usually are largest. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.8: Surface and tropospheric ozone trends. (a) Decadal ozone trends by latitude at 28 remote surface sites and in the lower free troposphere (650 hPa, about 3.5 km) as measured by IAGOS aircraft above 11 regions. All trends are estimated for the time series up to the most recently available year, but begin in 1995 or 1994. Colours indicate significance (p-value) as denoted in the in-line key. See Figure 6.5 for a depiction of these trends globally. (b) Trends of ozone since 1994 as measured by IAGOS aircraft in 11 regions in the mid-troposphere (700–300 hPa; about 3–9 km) and upper troposphere (about 10–12 km), as measured by IAGOS aircraft and ozonesondes. (c) Trends of average tropospheric column ozone mixing ratios from the TOST composite ozonesonde product and three composite satellite products based on TOMS, OMI/MLS (Sat1), GOME, SCIAMACHY, OMI, GOME-2A, GOME-2B (Sat2), and GOME, SCIAMACHY, GOME-II (Sat3). Vertical bars indicate the latitude range of each product, while horizontal lines indicate the very likely uncertainty range.. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.9: **Aerosol evolution from ice-core measurements.** Changes are shown as 10-yr averaged time series (a, b) and trends in remote-sensing aerosol optical depth (AOD) and AODf (c, d). (a) Concentrations of non-sea salt (nss) sulphate (ng g\(^{-1}\)). (b) Black carbon (BC) in glacier ice from the Arctic (Lomonosovfonna), Russia (Belukha), Europe (Colle Gnifetti), South America (Illimani), Antarctica (stacked sulphate record, and BC from the B40 core), and BC from Greenland (stacked rBC record from Greenland and Eastern Europe (Elbrus)). (c) Linear trend in annual mean AOD retrieved from satellite data for the 2000–2019 period (% yr\(^{-1}\)). The average trend from MODerate Resolution Imaging Spectroradiometer (MODIS) and Multi-Angle Imaging Spectroradiometer (MISR) is shown. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al., (2008a). Superimposed are the trends in annual mean AOD from the AERONET surface sunphotometer network for 2000–2019. (d) Linear trend in 2000–2019 as in (c), but for fine-mode AOD, AODf, and using only MISR over land. ‘x’ marks denote non-significant trends. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.10: Temporal evolution of effective radiative forcing (ERF) related to the drivers assessed in Section 2.2. ERFs are based upon the calculations described in Chapter 7, of which the global annual mean, central assessment values are shown as lines and the 5 to 95% uncertainty range as shading (Section 7.3, see Figures 7.6 to 7.8 for more detail on uncertainties). The inset plot shows the rate of change (linear trend) in total anthropogenic ERF (total without TSI and volcanic ERF) for 30-year periods centred at each dot. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.11: Earth’s surface temperature history with key findings annotated within each panel. (a) GMST over the Holocene divided into three time scales: (i) 12 kyr–1 kyr in 100-year time steps, (ii) 1000–1900 CE, 10-year smooth, and (iii) 1900–2020 CE (from panel c). Median of the multi-method reconstruction (bold lines), with 5th and 95th percentiles of the ensemble members (thin lines). Vertical bars are the assessed medium confidence ranges of GMST for the Last Interglacial and mid-Holocene (Section 2.3.1.1). The last decade value and very likely range arises from 2.3.1.3. (b) Spatially resolved trends (°C per decade) for HadCRUTv5 over (upper map) 1900–1980, and (lower map) 1981–2020. ‘x’ marks denote non-significant trends. (c) Temperature from instrumental data for 1850–2020, including (upper panel) multi-
product mean annual timeseries assessed in Section 2.3.1.1.3 for temperature over the oceans (blue line) and temperature over the land (red line) and indicating the warming to the most recent 10 years; and annually (middle panel) and decadal (bottom panel) resolved averages for the GMST datasets assessed in Section 2.3.1.1.3. The grey shading in each panel shows the uncertainty associated with the HadCRUT5 estimate (Morice et al., 2021). All temperatures relative to the 1850–1900 reference period. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Cross-Chapter Box 2.3, Figure 1: Changes in assessed historical surface temperature changes since AR5. (a) Summary of the impact of various steps from AR5 headline warming-to-date number for 1880–2012 using a linear trend fit to the AR6 assessment based upon the difference between 1850–1900 and 2011–2020. Whiskers provide 90% (very likely) ranges. AR6 assessment in addition denotes additional warming since the period around 1750 AR6 assessment in addition denotes additional warming since the period around 1750 (Cross-Chapter Box 1.2). (b) Time series of the average of assessed AR5 series (orange, faint prior to 1880 when only HadCRUT4 was available) and AR6 assessed series (blue) and their differences (offset) including an illustration of the two trend fitting metrics used in AR5 and AR6. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.12: Temperature trends in the upper air. (a) Zonal cross-section of temperature anomaly trends (2007–2016 baseline) for 2002–2019 in the upper troposphere and lower stratosphere region. The climatological tropopause altitude is marked as a grey line. Significance is not indicated due to the short period over which trends are shown, and because the assessment findings associated to this figure relate to difference between trends at different heights, not the absolute trends. (b) Trends in temperature at various atmospheric heights for 1980–2019 and 2002–2019 for the near-global (70°N–70°S) domain. (d) (e) as for (b) (c) but for the tropical (20°N–20°S) region. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.13: Changes in surface humidity. (a) Trends in surface specific humidity over 1973–2019. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al (2008a) ('x' marks denote non-significant trends). (b) Global average surface specific humidity annual anomalies (1981–2010 base period). (c) as (a) but for the relative humidity. (d) as (b) but for the global average surface relative humidity annual anomalies. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.14: Timeseries of global mean total column water vapour annual anomalies (mm) relative to a 1988–2008 base period. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.15: Changes in observed precipitation. (a, b) Spatial variability of observed precipitation trends over land for 1901–2019 for two global in-situ products. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al. (2008a) (‘x’ marks denote non-significant trends). (c) Annual time series and decadal means from 1891 to date relative to a 1981–2010 climatology (note that different products commence at distinct times). (d, e) as (a, b), but for the periods starting in 1980. (f) is for the same period for the globally complete merged GPCP v2.3 product. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Changes in precipitation minus evaporation

(a) Spatial pattern of ERA5 P-E trends of 1980-2019

(b) Global

(c) Land

(d) Ocean

Figure 2.16: Changes in precipitation minus evaporation. (a) Trends in precipitation minus evaporation (P-E) between 1980 and 2019. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after (Santer et al., 2008; ‘x’ marks denote non-significant trends). Time series of (b) global, (c) land-only and (d) ocean-only average annual P-E (mm/day). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.17: Timeseries of the annual mean Northern Hemisphere (NH, top curves) and Southern Hemisphere (SH, bottom curves) Hadley Cell extent (a) and Hadley Cell intensity (b) since 1979. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.18: Trends in ERA5 zonal-mean zonal wind speed. Shown are (a) DJF (December-January-February), (b) MAM (March-April-May), (c) JJA (June-July-August) and (d) SON (September-October-November). Climatological zonal winds during the data period are shown in solid contour lines for westerly winds and in dashed lines for easterly. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al (2008a) (‘x’ marks denote non-significant trends). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.19: Trends in surface wind speed. (a) Station observed winds from the integrated surface database (HadISD v2.0.2.2017); (b) Cross-Calibrated Multi-Platform wind product; (c) ERA5; and (d) wind speed from the Objectively Analyzed Air-Sea Heat Fluxes dataset, release 3. White areas indicate incomplete or missing data. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al. (2008b) (‘x’ marks denote non-significant trends). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.20: Changes in Arctic and Antarctic sea ice area. (a) Three time series of Arctic sea ice area (SIA) for March and September from 1979 to 2020 (passive microwave satellite era). In addition, the range of SIA from 1850–1978 is indicated by the vertical bar to the left. Decadal means for the three series for the first and most recent decades of observations are shown by horizontal lines in grey (1979–1988) and black (2010–2019). (b) Three time series of Antarctic sea ice area for September and February (1979–2020). Sea ice area values have been calculated from sea ice concentration fields. Available data for 2020 (OSISAF) is shown in both (a) and (b). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.21: Arctic sea ice thickness changes (means) for autumn (red/dotted red) and winter (blue/dotted blue). Shadings (blue and red) show 1 S.E. ranges from the regression analysis of submarine ice thickness and expected uncertainties in satellite ice thickness estimates. Data release area of submarine data ice thickness data is shown in inset. Satellite ice thickness estimates are for the Arctic south of 88°N. Thickness estimates from more localized airborne/ground electromagnetic surveys near the North Pole (diamonds) and from Operation IceBridge (circles) are shown within the context of the larger scale changes in the submarine and satellite records. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.22: April snow cover extent (SCE) for the Northern Hemisphere (1922–2018). Shading shows very likely range. The trend over the entire 1922–2018 period (black line) is −0.29 (± 0.07) million km² per decade. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.23: Glacier advance and annual mass change. (a) Number of a finite selection of surveyed glaciers that advanced during the past 2000 years. (b) Annual and decadal global glacier mass change (Gt yr\(^{-1}\)) from 1961 until 2018. In addition, mass change mean estimates are shown. Ranges show the 90% confidence interval. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.24: Cumulative Antarctic Ice Sheet (AIS) and Greenland Ice Sheet (GrIS) mass changes. Values shown are in gigatons and come from satellite-based measurements (IMBIE Consortium, 2018, 2020) for the period 1992–2018 for GrIS and 1992–2017 for AIS. The estimated uncertainties, very likely range, for the respective cumulative changes are shaded. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.25: Changes in permafrost temperature. Average departures of permafrost temperature (measured in the upper 20–30m) from a baseline established during International Polar Year (2007–2009) for Arctic regions. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.26: Changes in ocean heat content (OHC). Changes are shown over (a) full depth of the ocean from 1871 to 2019 from a selection of indirect and direct measurement methods. The series from Table 2.7 is shown in solid black in both (a) and (b) (see Table 2.7 caption for details). (b) as (a) but for 0–2000 m depths only and reflecting the broad range of available estimates over this period all of which are broadly similar. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.27: Changes in ocean salinity. Estimates of salinity trends using a total least absolute differences fitting method for (a) global near-surface salinity (SSS) changes and (b) global zonal mean subsurface salinity changes. Black contours show the associated climatological mean salinity (either near-surface (a) or subsurface (b)) for the analysis period (1950–2019). Both panels represent changes of Practical Salinity Scale 1978 [PSS-78], per decade. In both panels green denotes freshening regions and orange/brown denotes regions with enhanced salinities. ‘x’ marks denote non-significant changes. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.28: Changes in global mean sea level. (a) Reconstruction of sea-level from ice core oxygen isotope analysis for the last 800 kyr. For target paleo periods (CCB2.1) and MIS11 the estimates based upon a broader range of sources are given as box whiskers. Note the much broader axis range (200 m) than for later panels (tenths of metres). (b) Reconstructions for the last 2500 years based upon a range of proxy sources with direct instrumental records superposed since the late 19th century. (c) Tide-gauge and, more latterly, altimeter based estimates since 1850. The consensus estimate used in various calculations in Chapters 7 and 9 is shown in black. (d) The most recent period of record from tide-gauge and altimeter based records. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.29: Low latitude surface ocean pH over the last 65 million years. (a) Low-latitude (30°N–30°S) surface ocean pH over the last 65 million years, reconstructed using boron isotopes in foraminifera. (b) as (a) but for the last 3.5 million years. Double headed arrow shows the approximate magnitude of glacial–interglacial pH changes. (c) Multisite composite of surface pH. In a)–c), uncertainty is shown at 95% confidence as a shaded band. Relevant paleoclimate reference periods (CCB2.1) have been labelled. Period windows for succeeding panels are shown as horizontal black lines in a) and b). (d) Estimated low-latitude surface pH from direct observations (BATS, HOT) and global mean pH (65°S–65°N) from two indirect estimates (CMEMS, OCEAN-SODA). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.30: Changes in the amplitude of the seasonal cycle of CO$_2$. (a) Observed peak-to-trough seasonal amplitude given by the day of year of downward zero crossing, of CO$_2$ concentration at Barrow (71°N, blue) and Mauna Loa (20°N, black). Seasonal CO$_2$ cycles observed at (b) Barrow and (c) Mauna Loa for the 1961–1963 or 1958–1963 and 2017–2019 time periods. The first six months of the year are repeated. Reprinted with permission from AAAS. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.31: Phytoplankton dynamics in the ocean. (a) Climatology of chlorophyll-a concentration derived from ocean-colour data (1998–2018); (b) Linear trends in chlorophyll concentration. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al. (2008b) (‘x’ marks denote non-significant changes). (c) Histogram of linear trends in chlorophyll concentration, after area weighting and with per-pixel uncertainty estimates based on comparison with in situ data. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.32: Phenological indicators of changes in growing season. (a) cherry blossom peak bloom in Kyoto, Japan; (b) grape harvest in Beaune, France; (c) spring phenology index in eastern China; (d) full flower of Piedmont species in Philadelphia, USA; (e) Grape harvest in Central Victoria, Australia; (f) start of growing season in Tibetan Plateau, China. Red lines depict the 25-year moving average (top row) or the 9-year moving average (middle and bottom rows) with the minimum roughness boundary constraint of Mann (2004). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.33: Satellite-based trends in Fraction of Absorbed Photosynthetically Active Radiation (per decade) for 1998–2019. Trends are calculated using OLS regression with significance assessed following AR(1) adjustment after Santer et al (2008b) (‘x’ marks denote non-significant trend). Unvegetated areas such as barren deserts (grey) and ice sheets (white) have no trend in FAPAR. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.34: Selected large-scale climate indicators during paleoclimate and recent reference periods of the Cenozoic Era. Values are based upon assessments carried out in this chapter, with confidence levels ranging from low to very high. Refer to Cross-Chapter Box 2.1 for description of paleoclimate reference periods and Section 1.4.1 for recent reference periods. Values are reported as either the very likely range (x to y), or best estimates from beginning to end of the reference period with no stated uncertainty (x → y), or lowest and highest values with no stated uncertainty (x ~ y). Temperature is global mean surface temperature. Glacier extent is relative and colour scale is inverted so that more extensive glacier extent is intuitively blue.

### Figure 2.34: Selected large-scale climate indicators during paleoclimate and recent reference periods of the Cenozoic Era

#### Atmosphere
- **CO₂ (ppm)**
  - Approximate pre-industrial (1850–1900 CE): 286 – 296
  - Mid-Holocene (6.5–6.5 ka): 260 – 268
  - Last Deglacial Transition (18–11 ka): 193 – 271
  - Last Glacial Maximum (23–19 ka): 188 to 194
  - Last Interstadial (129–130 ka): 266 to 282
  - Mid-Pleistocene warm period (3.5–3.0 Ma): 360 to 420
  - Early Eocene climate optimum (55–50 Ma): 1150 to 2500
  - Paleocene–Eocene thermal maximum (56.5–50.7 Ma): 900 to 2000

#### Cryosphere
- **Temperature relative to 1850–1900 (°C)**
  - Recent past (1995–2014 CE): 0.66 to 1.00
  - Approximate pre-industrial (1850–1900 CE): 0.15 to 0.25
  - Last Millennium (850–1850 CE): 0.0 to 1.0
  - Mid-Holocene (6.5–6.5 ka): 0.14 to 0.24
  - Last Deglacial Transition (18–11 ka): 0
  - Last Glacial Maximum (23–19 ka): 0
  - Last Interstadial (129–130 ka): 0.5 to 1.5
  - Mid-Pleistocene warm period (3.5–3.0 Ma): 2.5 to 4.0
  - Early Eocene climate optimum (55–50 Ma): 10 to 15
  - Paleocene–Eocene thermal maximum (56.5–50.7 Ma): 4 to 42

#### Biosphere
- **Glacier extent relative to 1850–1900 (unitless)**
  - Recent past (1995–2014 CE): 0.15 to 0.25
  - Approximate pre-industrial (1850–1900 CE): 0.03 to 0.00
  - Last Millennium (850–1850 CE): 0.05 to 0.03
  - Mid-Holocene (6.5–6.5 ka): 3.5 to 0.5
  - Last Deglacial Transition (18–11 ka): 3.5 to 0.5
  - Last Glacial Maximum (23–19 ka): 4 to 42
  - Last Interstadial (129–130 ka): 8 to 2
  - Mid-Pleistocene warm period (3.5–3.0 Ma): 4 to 10
  - Early Eocene climate optimum (55–50 Ma): 10 to 15
  - Paleocene–Eocene thermal maximum (56.5–50.7 Ma): 10 to 25

#### Ocean
- **Sea level relative to 1900 (m)**
  - Recent past (1995–2014 CE): 2.9 to 3.6
  - Approximate pre-industrial (1850–1900 CE): 0.4 to 0.6
  - Last Millennium (850–1850 CE): 1.1 to 0.7
  - Mid-Holocene (6.5–6.5 ka): 3.5 to 0.5
  - Last Deglacial Transition (18–11 ka): 24 to 44
  - Last Glacial Maximum (23–19 ka): 10 to 25
  - Last Interstadial (129–130 ka): 5 to 10
  - Mid-Pleistocene warm period (3.5–3.0 Ma): 5 to 25
  - Early Eocene climate optimum (55–50 Ma): 70 to 76
  - Paleocene–Eocene thermal maximum (56.5–50.7 Ma): Insufficient data or not assessed

#### Sea level rate of change (mm/yr or m/1000 yr)
- Recent past (1995–2014 CE): 2.9 to 3.6
- Approximate pre-industrial (1850–1900 CE): 0.4 to 0.6
- Last Millennium (850–1850 CE): -1.1 to 0.7
- Mid-Holocene (6.5–6.5 ka): 3.5 to 0.5
- Last Deglacial Transition (18–11 ka): 24 to 44
- Last Glacial Maximum (23–19 ka): 10 to 25
- Last Interstadial (129–130 ka): 5 to 10
- Mid-Pleistocene warm period (3.5–3.0 Ma): 5 to 25
- Early Eocene climate optimum (55–50 Ma): 70 to 76
- Paleocene–Eocene thermal maximum (56.5–50.7 Ma): Insufficient data or not assessed
Cross-Chapter Box 2.4, Figure 1: Climate indicators of the mid-Pliocene Warm Period (3.3–3.0 Ma) from models and proxy data. (a) Simulated surface air temperature (left) and precipitation rate anomaly (right) anomaly (relative to 1850–1900) from the Pliocene Model Intercomparison Project Phase 2 multi-model mean, including CMIP6 (n = 4) and non-CMIP6 (n = 12) models. Symbols represent site-level proxy-based estimates of sea-surface temperature for KM5c (n = 32), and terrestrial temperature (n = 8) and precipitation rate for the MPWP (n = 8). (b) Distribution of terrestrial biomes was considerably different during the Piacenzian Stage (3.6–2.6 Ma) (upper) compared with present-day (lower). Biome distributions simulated with a model (BIOME4) in which Pliocene biome classifications are based on 208 locations, with model-predicted biomes filling spatial gaps, and the present day, with the model adjusted for CO₂ concentration of 324 ppm. (c) Ice-sheet extent predicted using modelled climate forcing and showing where multiple models consistently predict the former presence or absence of ice on Greenland (n = 8 total) and Antarctica (n = 10 total). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.35: Southern Annular Mode (SAM) reconstruction over the last millennium. (a) SAM reconstructions as 7-year moving averages (thin lines) and 70-year LOESS filter (thick lines). (b) observed SAM index during 1900–2019. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.36: Reconstructed and historical variance ratio of El Niño–Southern Oscillation (ENSO). (a) 30-year running variance of the reconstructed annual mean Niño 3.4 or related indicators from various published reconstructions. (b) Variance of June-November Southern Oscillation Index (SOI) and April-March mean Niño 3.4 (1981–2010 base period) along with the mean reconstruction from (a). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.37: Indices of interannual climate variability from 1950–2019 based upon several sea surface temperature based data products. Shown are the following indices from top to bottom: IOB mode, IOD, Niño4, AMM and AZM. All indices are based on area-averaged annual data (see Technical Annex IV). Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
Figure 2.38: Indices of multi-decadal climate variability from 1854–2019 based upon several sea surface temperature based data products. Shown are the indices of the AMV and PDV based on area averages for the regions indicated in Technical Annex IV. Further details on data sources and processing are available in the chapter data table (Table 2.SM.1).
FAQ 2.1: How is this global warming different to before?
Climate has always changed, but warming like that of recent decades has not been seen for millennia or longer.

- It is warming almost everywhere.
- It is warming rapidly.
- The warming reversed a long-term cooling.
- It has been a long time since it has been this warm.

FAQ 2.1, Figure 1: Evidence for the unusualness of recent warming.
FAQ 2.2, Figure 1: Synthesis of significant changes observed in the climate system over the past several decades. Upwards, downwards and circling arrows indicate increases, decreases and changes, respectively. Independent analyses of many components of the climate system that would be expected to change in a warming world exhibit trends consistent with warming. Note that this list is not comprehensive.