Annex IV: Modes of Variability

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AIV.1 Introduction

This annex describes the fundamental features of the main modes of large-scale climate variability assessed across chapters in the AR6 WGI Report. Modes are defined as recurrent space-time structures of variability of the climate system with intrinsic spatial patterns, seasonality and time scales. They can arise through the dynamical characteristics of atmospheric circulation, but also through coupling between the ocean and the atmosphere, with some interactions with land surfaces and sea ice. The variability of the climate system at ocean- or continental-basin scales, and in particular, on seasonal-to-multi-decadal time scales, can be described to a large extent by the occurrence and often combination of several modes of climate variability which lead to local impacts and remote responses through teleconnection processes on top of externally forced trends. More precisely, the concept of ‘teleconnection’ refers to the ability of modes of variability to relate climate in remote regions through associated atmospheric or oceanic pathways. Regional climate variations are thus the complex outcome of local physical processes, such as thermodynamical and land–atmosphere feedback processes, and non-local large-scale phenomena.

Historically, modes have been identified based on evidence of correlations between regional climate variations at widely separated, mostly geographically fixed spatial locations. One of the oldest known space-time structured patterns of variability is the North Atlantic Oscillation (NAO; Section AIV.2.1), with some early description from seafaring activities available since several centuries ago (Stephenson et al., 2003). Similarly, empirical, albeit remarkably precise, documentation of trans-Pacific coherent variability in atmospheric and oceanic conditions can be traced back to the 19th century for the El Niño–Southern Oscillation (ENSO; Section AIV.2.3). The Southern Oscillation was named in 1924 and the first use of the name ‘El Niño’ to describe a climatic phenomenon appeared in South America in 1893 (see for instance Philander, 1983 for ENSO history). Since then, the increased spatial coverage of observations and the measurement of new climatic parameters have improved significantly. These advancements, along with the development of new theoretical background and statistical methods for combining different sources of information, have allowed substantial improvements in our understanding of the characteristic spatial scale of the coherent structures of variability as well as of the characteristic temporal expression of the modes.

There are several ways to document and describe the modes of climate variability based on more or less sophisticated statistical techniques and the dynamical nature, interpretation and understanding of the variability (i.e., Feldstein and Franzke, 2017; Hannachi et al., 2017; Franzke et al., 2020; Ghil and Lucarini, 2020). The simplest method to evaluate the temporal evolution of a spatially coherent phenomenon is to take one or more spatial averages of a climate variable such as sea surface temperature (SST) or sea level pressure (SLP), as examples, within a latitude-longitude box. Another classical technique relies on principal component analysis (PCA), also known as empirical orthogonal function (EOF) analysis, to extract the patterns in the chosen climate variable with maximal spatial covariance. The method also extracts an associated climate index time series, that is, principal component (PC), describing the temporal evolution tied to the spatial pattern. PCA/EOF is widely used with gridded datasets and model outputs. Correlation or regression maps between the climate index time series and other space-time variables, such as precipitation, temperature or wind, are then calculated to quantify the teleconnection patterns. Altogether, beyond their physical relevance, modes of climate variability and their teleconnection patterns provide a simplified representation of climate variation affecting weather, from regional (i.e., oceanic or continental scale) to worldwide scales (i.e., for some specific phenomena like ENSO), having important impacts on human activities and ecosystems. Modes are often associated with climatic impact-drivers (CIDs, as defined in Chapter 12, Section 12.1) including droughts, heavy rains, storms, heatwaves and cold spells, affecting agriculture, water resources, availability and consumption of energy, and fire risks.

Many modes of variability are driven by internal climate processes and provide a substantial potential source of climate predictability on sub-seasonal to decadal time scales. It is essential to understand the physical processes behind the past evolution of the modes of climate variability in order to assess, with confidence, their future changes. External forcing may affect their temporal (occurrence, variance, seasonality and persistence, etc.) or spatial properties and associated teleconnections.

Previous IPCC reports have described the observed and modelled behaviour of a large number of modes of variability over the instrumental period and in a paleoclimate context. The projected evolution of those modes has been also documented based on future climate scenarios. In AR5, large-scale modes of variability and teleconnections were assessed in depth in Chapter 14, setting the physical basis of future regional climate changes contingent on global mean temperature increase. Modes were defined through so-called ‘indices of climate variability’ in AR5 Chapter 2, dedicated to observed atmospheric and surface changes. Two boxes (AR5 Box 2.5 and AR5 Box 14.1) introduced the main concepts. AR4 had a similar structure, with a specific section on modes in the chapter dedicated to the observations (Chapter 3). In AR6, a different perspective has been adopted because the modes are discussed in various chapters and assessed with respect to the specific objectives of each of them. Space-time past changes of the modes using proxy records and instrumental data are assessed in Chapter 2. Human influence on modes is assessed in Chapter 3 over the instrumental period, while their future changes are addressed in Chapter 4. Their influence at regional scale through large-scale teleconnection is assessed in Chapters 10, 11 and 12, and in the Atlas, while the contribution of the modes in past and future hydrological changes and related uncertainties are assessed in Chapter 8. Chapter 9 focuses on oceanic expression of the modes. The aim of this annex is to set a common background for the assessment of the modes of climate variability and related regional climate anomalies for all the AR6 WGI chapters, and to provide this information for the readers of the report. In doing so, the annex provides a more accessible summary of the main modes of variability allowing a more coherent assessment of this important aspect of the climate system.
The annex summarizes the fundamental features of spatial structure, seasonality and teleconnections for a selection of modes of climate variability corresponding to those addressed in the report, namely: the NAO and Northern Annular Mode, the Southern Annular Mode, ENSO, the Indian Ocean Basin and Dipole modes, the Atlantic Meridional and Zonal modes, the Pacific Decadal Variability, the Atlantic Multi-decadal Variability, and the Madden–Julian Oscillation. For each mode, the description follows a common structure, starting with a short definition of the mode (based on simple box-averaged indices and EOF) and related key physical mechanisms, then a description of the associated teleconnections (based on regression techniques), and when relevant, finishes with a short summary of paleoclimate reconstruction of the mode based on proxy-based records. An illustrative figure including patterns and time series is also provided for each mode.

AIV.2 The Main Modes of Climate Variability Assessed in AR6

AIV.2.1 North Atlantic Oscillation and Northern Annular Mode

AIV.2.1.1 Definition of the Mode

The North Atlantic Oscillation (NAO) is the leading mode of large-scale atmospheric variability in the North Atlantic basin. It is characterized by alternating (seesaw) variations in SLP between the Azores High in the sub-tropics and the Icelandic Low in the mid- to high latitudes, with some northward extension deep into the Arctic (Hurrell et al., 2003). The positive phase of the NAO corresponds to an anomalously strong Azores High and concomitant deepening of the Icelandic Low with some penetration in the sub-Arctic basin (Greenland/Norwegian/ Barents Sea), accompanied by a reinforced and northward-shifted jet and storm track over a broad North Atlantic-European sector (Figure AIV.1a). The opposite, notwithstanding marginal spatial asymmetries, occurs for the negative phase.

The NAO is the main mode of climate variability over a broad North Atlantic-European region in all seasons (Hurrell and Deser, 2009) for intra-seasonal to multi-decadal time scales, albeit with seasonally dependent physical characteristics (fraction of explained variance, latitudinal shift and intensity of the related anomalous atmospheric circulation and SST fingerprint, etc.; Folland et al., 2009; Woollings et al., 2015). The phases of the NAO control a significant fraction of the variance of temperature and precipitation over the North Atlantic and surrounding continents, and of the prevailing westerly winds and the related storm tracks (Woollings et al., 2014). It is also strongly related to the occurrence and intensity of blocking conditions, especially over Greenland (Woollings et al., 2010; Davini et al., 2012), and controls a large part of the incidence of extremes (cold waves, very strong wind episodes related to explosive storminess, heavy precipitation events, etc.) over Europe (Matthews et al., 2014; You et al., 2017) and eastern North America (Durkee et al., 2008; Whan and Zwiers, 2017).

The NAO can be viewed as the regional expression of a hemispheric-scale mode of atmospheric variability that was identified in the late 1990s known as the Arctic Oscillation (AO; Thompson and Wallace, 1998) or Northern Annular Mode (NAM; Thompson and Wallace, 2000; Ambaum et al., 2001) to reflect the longitudinal scale that transcends the entire hemisphere. An additional centre of action is present for NAM over the North Pacific, albeit much weaker than its Atlantic counterpart, with a slackened Aleutian Low in the positive phase (Figure AIV.1b). The AO/NAM can be defined as the leading mode of variability across a range of atmospheric levels from the surface to the stratosphere (Baldwin and Dunkerton, 1999; Gerber et al., 2010). In the stratosphere, the AO/NAM corresponds to fluctuations in the strength of the boreal winter polar vortex over the Arctic. In the troposphere, the AO/NAM primarily corresponds to fluctuations in the position of the mid-latitude westerly jet stream. Interactions between the stratospheric and tropospheric annular modes have been extensively studied and are important for the variability and predictability of the mode at intra-seasonal time scales (i.e., see Domeisen, 2019).

Despite the fact that NAM and NAO indices are strongly correlated (Figure AIV.1c), the debate on the relevance of the NAM as a mode of variability per se (Deser, 2000; Honda and Nakamura, 2001; Itoh, 2008) is not closed yet, despite progress in understanding the dynamical mechanisms linking the Pacific and Atlantic SLP cores on the basis of wave breaking theory that prevails at intra-seasonal time scales (e.g., Rivière and Drouard, 2015). Consequently, NAO and NAM are treated as indistinguishable in some studies (Feldstein and Franzke, 2006) while others still consider the NAO as a stand-alone mode even if associated circumpolar patterns exist (García-Serrano and Haarsma, 2017). In the latter case, the NAM is interpreted as a statistical artefact of two regional independent modes, namely the NAO and the Pacific-North American pattern (Ambaum et al., 2001; Douville et al., 2017). Throughout the report, NAO and NAM are often considered as the same entity because the NAO explains most of the variance on the zonally-averaged hemispheric circulation and because the associated time series are highly correlated.

The NAO/NAM arises as large-scale signals primarily driven by the internal atmospheric variability of mid-latitude dynamics (e.g., Lorenz and Hartmann, 2003). By nature, these signals correspond to barotropic phenomena being partly controlled by eddy-mean flow interaction (Feldstein and Franzke, 2017). Their fingerprint can be found along the entire atmospheric column with maximum amplitude in the winter season when storminess is the most pronounced. In addition, the tropospheric NAO/NAM variability has been shown in boreal winter to be linked to the strength of the stratospheric polar vortex (Baldwin and Dunkerton, 1999), including sudden stratospheric warmings (Domeisen, 2019). On intra-seasonal time scales, downward propagation of associated geopotential height anomalies occurs from the lower stratosphere to the troposphere through wave-mean flow interaction resulting in some predictability at the surface for CIDs, through temperature, precipitation and wind anomalies at large spatial scales (e.g., Karpechko et al., 2017 over Europe).
Figure AIV.1 | The boreal winter North Atlantic Oscillation (NAO) and Northern Annular Mode (NAM) extracted as the leading empirical orthogonal function (EOF) of December–January–February (DJF) sea level pressure (SLP) anomalies over (a) the North Atlantic–Europe sector (20°N–80°N, 90°W–30°E, indicated by the red box) and (b) the Northern Hemisphere (north of 20°N, black domain), respectively, based on JRA-55 reanalysis over 1959–2019. (a, b) SLP anomalies regressed onto the leading principal component (PC) time series shown in (c). The NAO-PC is represented by red and blue bars, the NAM-PC by the black curve in (c); the percentages of variance explained by the mode are shown in the legends. The NAO index based on weather stations (cyan dots in (a)) is given in cyan and the zonal-NAM index (latitude circles in (b)) is given in purple. Correlation between each index and the NAO-PC time series is given in the legend. (d, e) Regression maps of blended sea surface temperature and surface air temperature over land from the ERSS and Berkeley Earth datasets, respectively, for 1959–2019 (d) and of precipitation anomalies (e); shading based on GPCC for 1959–2016 and contours based on CMAP for 1979–2019 for every 0.1 mm day⁻¹. On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicate no significance. Significance for CMAP precipitation is materialized by red contours in (e). All fields have been linearly detrended prior to computation.
The NAO/NAM includes a forced component arising from SST anomalies: key oceanic regions are identified but are dependent on the time scale of variability that is investigated (Baker et al., 2019). There is also growing evidence for the importance of cyrosphere-atmosphere coupling to explain part of the observed variability of the NAO in response to recent Arctic sea ice decline (Cohen et al., 2014), even if discrepancies remain in model estimates (Screen et al., 2018) and may not be stationary in time (Blackport and Screen, 2020). Finally, part of the NAO/NAM variability has also been documented to be influenced by external forcing such as volcanic eruptions (Swingedouw et al., 2017) and solar forcing (Ineson et al., 2011; Gray et al., 2016; Le Mouël et al., 2019) as well as anthropogenic factors (Gillett et al., 2003). However, internal variability is predominant and, because of signal-to-noise issues, very long observational records and large ensembles of climate simulations are needed to clearly evaluate the fraction of the forced component in the observed NAO/NAM, as assessed in detail in Sections 2.4.1.1 and 3.7.1. Sea ice-NAO/NAM coupling is comprehensively assessed in Cross-Chapter Box 10.1.

Given their equivalent barotropic structure, the NAO/NAM is usually defined from SLP fields or geopotential height anomalies at various levels (Fu et al., 2016). The NAO/NAM can be defined using pattern-based statistical techniques such as EOF or clustering techniques applied to seasonal gridded atmospheric circulation fields (SLP or geopotential height anomalies) from observations or reanalysis data. The NAM pattern is traditionally defined as the leading EOF of pressure fields computed over the entire hemisphere, northward of 20°N (Figure AIV.1b) and its related PC defines the NAM index (Figure AIV.1c). Similarly the NAO can be defined as the PC of the leading EOF of the pressure field over the North Atlantic sector (Figure AIV.1a) or the temporal occurrence of the seasonal weather regimes extracted from clustering (Corti et al., 1999; Cassou et al., 2011; Hannachi et al., 2017).

An alternative and simpler estimate of the NAM/NAO uses differences in SLP between fixed locations. The NAM index can be calculated from the difference in normalized zonally averaged SLP anomalies between 35°N and 65°N (Li and Wang, 2003), which reflects the latitudinal fluctuation of the position and strength of the mid-latitude westerlies at the hemispheric scale. Similarly, for the NAO, a weather station-based index is traditionally calculated as the difference of normalized SLP anomalies located in the Azores (Ponta Delgada) or continental Iberia (Lisbon/Gibraltar) minus those along the northern centre of action around Iceland (Stykkisholmur/Reykjavik) (Jones et al., 1997). Cornes et al. (2013) alternatively uses the Paris-London difference because of data availability back to the 17th century. The station-based NAO index is simple to compute, extends the longest over the instrumental period, and can be used year round, but caveats stand in its limitation to capture the seasonal latitudinal migration of the NAO variability pattern as well as potential shifts in the North Atlantic pressure centres of action at interannual to decadal time scales (Ulbrich and Christoph, 1999). Because it uses fixed locations, the index captures part of the variability that is not directly related to the latitudinal seesaw pressure balance and associated large-scale fluctuations of the zonal flow that characterize the NAO dynamics, and it also fails representing the spatial asymmetries between the positive and negative phases of the mode (Cassou et al., 2004).

The temporal correlation between the EOF and fixed latitude-based zonal NAM estimations is equal to 0.91 for the December–January–February (DJF) average and both are strongly correlated with the NAO indices (Figure AIV.1c). Correlations between all the NAO estimates are spatially and temporally very high in all the observational products over the modern reanalyses period (Gerber and Martineau, 2018); for instance, the correlation between the station-based and EOF-based NAO indices over 1959–2018 in boreal winter using JRA-55 is equal to 0.89 (Figure AIV.1c). EOF-based NAO and NAM indices are used in Section 3.7.1 for model evaluation. The fixed-latitude NAM index is used in Sections 2.4.1.1, 4.3.3.1, 4.4.3.1 and 4.5.3.1, and in Section 3.7.1 for attribution of past changes.

**AIV.2.1.2 Teleconnections and Regional Influence**

Large-scale changes in the position and strength of the North Atlantic jet stream and storm tracks associated with the NAO/NAM, as well as the related incidence in blocking episodes traditionally linked to extreme events, affect surface climate conditions over the entire North Atlantic and surrounding land masses from daily to multi-decadal time scales (Hurrell, 1995). Warmer and wetter weather over Northern Europe extending eastward in western Siberia with simultaneous cooler and drier conditions over a broad Mediterranean basin are distinct features of the positive boreal winter NAO/NAM (Figure AIV.1d,e). Over North America, regions around the Labrador Sea are significantly colder, with excess sea ice formation, while milder conditions dominate the Great Plains, forming altogether a typical quadripolar anomaly pattern for temperature linked to the winter NAO/NAM. The latter structure is tightly related to the latitudinal shift of the storm tracks that is particularly marked in precipitation over the ocean (Figure AIV.1e). More remotely, changes in North Atlantic storm track position and transient eddy activity associated with the positive (negative) phase of the NAO contribute downstream to negative (positive) SLP anomalies in north-eastern East Asia during boreal summer (Linderholm et al., 2011).

Related changes in both westerly winds at mid-latitudes and trade winds in the tropics over the North Atlantic result in anomalous fluxes of sensible and latent heat at sea surface, which create ocean temperature anomalies that extend down to the base of the deep mixed layer during boreal winter (Deser and Timlin, 1997; Seager et al., 2000). The SST imprint of the NAO has a tripolar shape with positive anomalies at mid-latitudes stranded by colder surface ocean in the subtropics and along the entire subpolar gyre (Figure AIV.1d). Weak feedbacks from SST on boreal winter NAO are found through the forced Rossby wave mechanism (Baker et al., 2019) and local phenomena such as oceanic re-emergence in the North Atlantic Nordic seas (Cassou et al., 2007). Negative NAO phases are characterized by a wavier jet stream and a southward shift of storm tracks, which favours cold outbreaks over Northern Europe (Cattiaux et al., 2010) and reinforced storminess along the Mediterranean basin leading to heavy precipitation events (Sanchez-Gomez et al., 2008) and storm surges in southern Europe (Cid et al., 2016). Those are also often related to a collapse/split/displacement of the low-level stratospheric polar vortex and the occurrence of stratospheric sudden warming (Sections 2.3.1.4.5 and 3.3.3.4). In boreal winter, NAM and NAO teleconnections are very similar over the Atlantic-surrounding continents.
There is a strong seasonal asymmetry in observed NAO teleconnection patterns and NAO temporal evolution. Impacts over Europe are opposite between summer and winter months with dominant warmer and drier weather conditions over northwestern Europe in summer concurrently with wetter conditions over southern Europe during positive NAO phases (Folland et al., 2009; Bladé et al., 2012). Extreme events (heatwave occurrence, droughts, etc.) are also linked to NAO phases (Cassou et al., 2005; Drouard et al., 2019) due to the related latitudinal shift of the North Atlantic storm tracks (Dong et al., 2013). Summer NAO trends also affect the Arctic climate as assessed in Sections 3.4.1 and 9.4.1. The NAO is associated with storms or periods of persistent mild winds and dry/sunny spells affecting availability and power consumption (Jerez et al., 2013; Zubiate et al., 2017).

NAO fluctuations generate ocean responses due to altered buoyancy fluxes and wind stress anomalies (Barrier et al., 2014). Signatures of the NAO can be found in the gyre circulations (Marshall et al. 2001) and in the Atlantic Meridional Overturning Circulation (AMOC; Yeager and Danabasoglu, 2014), with a sea level imprint in observations (Han et al., 2017). The oceanic response is dependent on the time scale of the NAO variability (Delworth and Zeng, 2016), especially for AMOC in connection with the so-called Atlantic Multi-decadal Variability (AMV; Section AIV.2.7). This is assessed in Section 9.2.3.1.

Implications of NAO/NAM variability in terms of observed and projected precipitation at a regional scale (Europe, Mediterranean basin, etc.) is comprehensively assessed in Sections 8.3.2.9, 8.4.2.9 and 8.5.2. NAO implications on extremes are assessed in Sections 11.3.1 and 11.4.1, and regional manifestations of NAM/NAO are further assessed in Atlas.5, Atlas.7, Atlas.8 and Atlas.9, and Cross-Chapter Box Atlas.2.

**AIV.2.3 Available Proxy-based Reconstruction**

Sea level pressure records from Stykkisholmur, Akureyri, Reykjavik and from Gibraltar, Lisbon and Ponta Delgada weather stations, all located close to the NAO centres of action (i.e., the Icelandic Low and the Azores High), allowed the reconstruction of the NAO index back to the mid-19th Century (Cropper et al., 2015; Hernández et al., 2020). The Paris–London sea level pressure difference used by Cornes et al. (2013) extends back to 1692, but it captures only part of the large-scale NAO phenomena because of the spatial proximity of the weather stations used. More reliable NAO reconstructions were recently developed based on the reconstruction of the sea level pressure field all over the North Atlantic and Europe (Delaygue et al., 2019; Mellado-Cano et al., 2019). Other long-term reconstructions involve proxy-based information from documents (Barriopedro et al., 2014), tree rings (Cook et al., 2019), ice cores (Sjolte et al., 2018), stalagmites (Baker et al., 2015), sediments (Faust et al., 2016) and a combination of multi-proxy information (Ortega et al., 2015), with large uncertainties due to the diverse methodological approaches used for reconstruction (Michel et al., 2020) and disparate correlations between proxies and associated temporal resolution (Hernández et al., 2020). An additional difficulty for reconstruction lies on the fact that the spatial structure of North Atlantic variability might be time-scale dependent (Woollings et al., 2015). It is shown that on the shorter time scale (interannual), the NAO is dominated by variations in the latitude of the North Atlantic jet and storm track, whereas on the longer time scale (multi-decadal and beyond) it represents changes in their strength. The two time scales also exhibit different regional impacts on temperature and precipitation, and different relationships to SSTs, which makes proxy reconstructions of the NAO problematic.

Most of the reconstructed NAO/NAM indices describe the boreal winter patterns of the last 500 years because the variability is the most pronounced and is on a larger scale for that season (Luterbacher, 2001; Dezileau et al., 2011). Tree rings can however allow NAO boreal summer reconstruction back to the 16th century (Folland et al., 2009; Linderholm et al., 2009). Recent efforts to develop reconstructions of the NAO phases using pollen data (Mauri et al., 2014), speleothems (Ait Brahim et al., 2019) and lake sediments (Olsen et al., 2012) provide new evidence for multi-centennial NAO-type variability along the Holocene (Hernández et al., 2020). A detailed description of the latest proxy-based reconstructions of the NAO and related conclusions placing the observations since the pre-industrial period in a broader context of variability are given in Section 2.4.1.1.
Figure AIV.2 | The austral summer Southern Annular Mode (SAM) extracted as the leading empirical orthogonal function (EOF) of December–January–February (DJF) sea level pressure (SLP) anomalies over the Southern Hemisphere (south of 20°S, a) based on JRA-55 reanalysis over 1979–2019. (a) SLP anomalies regressed onto the leading principal component (PC) time series shown in (b); the variance explained is given in the legend in (b). The SAM-PC is represented by red and blue bars and the zonal-SAM index (latitude circles in a) is given in purple. Correlation between the zonal SAM index and the SAM PC time series is given in the legend. (c) same as (a) but for sea surface temperature and surface air temperature anomalies over land from the ERSST and Berkeley Earth datasets, respectively, over 1979–2019. (d) same as (a) but for precipitation anomalies (shading based on GPCP for 1979–2016 and contours based on CMAP for 1979–2019 for every 0.1 mm day$^{-1}$). On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicate no significance. Significance for CMAP precipitation is materialized by red contours in (d). All fields have been linearly detrended prior to computation.
positive when pressure over Antarctica is anomalously low, and vice versa. While the SAM pattern is more zonally symmetric than its Northern Hemisphere counterpart (NAM; Section AIV.2.1), it does exhibit some asymmetries particularly in the mid-latitude node over the Pacific Ocean and Indian Ocean sectors (Fogt et al., 2012). Alternative interpretations of the SAM as a regional variability pattern in the South Pacific are also considered (Spensberger et al., 2020). The zonal asymmetries are present for all seasons, and illustrated in Figure AIV.2a for austral summer, to support the assessment in Section 3.7.2. A number of different SAM indices have been produced, and several are compared by Ho et al. (2012) and Barrucand et al. (2018), showing good agreement after the second half of the 20th century. Over the 1979–2019 period, the correlation between the EOF-based and zonal SLP-based indices for the summer SAM reaches 0.94 (Figure AIV.2b).

During the positive phase of the SAM, the Antarctic air mass is isolated by an enhanced westerly jet that is located further south than during the negative phase. This leads to colder conditions over the entire continent and negative SST anomalies along the surrounding ocean (Figure AIV.2). During the opposite, negative phase, an equatorward displacement of the jet and related storm track brings low-pressure systems toward the north with a strong effect on the southern tip of South America (Mariani and Fletcher, 2016). The SAM is often used as a proxy for the location of the mid-latitude westerly wind belt at seasonal to interannual time scales. However, Swart et al. (2015) noted that trends in the SAM and the Southern Hemisphere surface westerly jet are not directly interchangeable because multiple factors other than SAM can affect trends in the position, width and strength of the jet and surface wind.

The SAM is mainly driven by extratropical atmospheric dynamics, but its phase is affected by remote SST anomalies such as ENSO (Section AIV.2.3) and anthropogenic forcing as assessed in Section 3.7.2.

The EOF-based SAM index is used for model evaluation in Section 3.7.2. The fixed-latitude SAM index is used for the attribution assessment of the observed positive trend of the SAM that is maximum over the second half of the 20th century but reduced since about 2000 (Section 2.4.1.2) due to slackened ozone influence (Section 3.7.2). The fixed-latitude SAM index is also used in Sections 4.3.3.1, 4.4.3.1 and 4.5.3.1 to assess the future changes in Southern Hemisphere zonal circulation as a function of emissions scenarios and lead times.

AIV.2.2.2 Teleconnections and Regional Influence

Recent observational studies have enriched our understanding of the widespread influence of SAM variability on mean climate and related environmental parameters of the entire Southern Hemisphere extratropics (e.g., Manatsa et al., 2013; Waugh et al., 2013; Raut et al., 2014; Jones et al., 2016; Holz et al., 2017; Kostov et al., 2017; Marshall et al., 2017). The influence of the SAM extends over the mid-latitude continental land masses with precipitation anomalies over southern Australia (Cai and Watterson, 2002) and South-Eastern South America (Silvestri and Vera, 2003) in connection with the latitudinal migration and strength of the extratropical storm and mean westerly circulation. Consistently with NAM/NAO (Section AIV.2.1), there is a strong seasonality in the teleconnection associated with SAM. In austral summer, during a positive phase of the SAM, drier conditions affect most of the southern Patagonia region and the southernmost tip of South America, while wetter conditions are found in South Africa and Madagascar, as well as in south-eastern Australia where colder temperatures also prevail (Figure AIV.2c, d). Positive temperature anomalies are found in New Zealand, Tasmania and the southern tip of South America. These anomalies appear as the continental fingerprint of a warmer belt located around 40°S that is characterized by significant hemispheric-scale SST changes consistent with the southern shift of the summer storm track.

Moreover, several studies have described equatorward teleconnections linking the SAM to Asian precipitation across the equator (e.g., Prabhu et al., 2017; Liu et al., 2018). Coupling of seasonal stratospheric anomalies to the SAM can influence spring heat extremes in eastern Australia (Lim et al., 2019). Marshall et al. (2013) described a reversal in the relationship between the SAM and East Antarctic temperatures during austral summer/autumn in the first decade of the 21st century, due to regional variability of the pressure pattern associated with the zonal wavenumber 3. Other studies have analysed the temporal variability in the interactions between the SAM and ENSO at different time scales (e.g., Yu et al., 2015; Clem et al., 2016), being positive and significant during 1986–2014 (Vera and Osman, 2018).

Implications of SAM on observed and projected changes in the water cycle and specific regional influences are assessed in Sections 8.3.2.9 and 8.4.2.9, and in Sections 12.4, Atlas.7 and Atlas.11, respectively.

AIV.2.2.3 Available Proxy-based Reconstruction

The regional influence of the SAM is recorded in tree rings (Holz et al., 2017; Dätwyler et al., 2018), lake sediments (Fletcher et al., 2018; Moreno et al., 2018), ice cores (Goodwin et al., 2004) and corals (Goodwin and Harvey, 2008) across the Southern Hemisphere extratropics. Multi-proxy-based reconstructions of the SAM have been developed using temperature-sensitive proxy records that cover the last millennium (Villalba et al., 2012; Abram et al., 2014; Dätwyler et al., 2018). Villalba et al. (2012) used tree ring records to reconstruct the summer SAM since 1409, while Abram et al. (2014) combined ice-core records with tree ring data to derive a weighted composite-plus-scale reconstruction of the annual SAM index for the past millennium. Dätwyler et al. (2018) further included proxies from corals, speleothems and lake sediments using different stationarity criteria for the proxy record selection to build an austral summer SAM reconstruction. Hessl et al. (2017) compared three palaeoclimatic reconstructions of the SAM, highlighting the need for additional proxy records over the Indian Ocean sector, to fully account for the annular structural properties of this mode of variability. Uncertainties also arise from the lack of proxy data related to cold season conditions, given that reconstructions relying on tree rings might be biased towards summer thus accounting only for SAM austral summer variability (Hernández et al., 2020). Several studies attempted to reconstruct the SAM behaviour during the Holocene, inferred through changes in the Southern Hemisphere westerly wind belt, but uncertainties remain large without consensus in the SAM phases for most of the periods at multi-decadal time scales (Hernández et al., 2020). These SAM reconstructions are assessed in Section 2.4.1.2 and compared with simulations in Section 3.7.2.
AIV.2.3 El Niño–Southern Oscillation

AIV.2.3.1 Definition of the Mode

The El Niño–Southern Oscillation (ENSO) refers to the large-scale alternation between anomalous warming and cooling of central/eastern equatorial Pacific SSTs that coincide with changes in the overlying winds and precipitation (Philander, 1990; Neelin et al., 1998; Wang, 2018). The main features of ENSO are displayed in Figure AIV.3. ENSO is the primary mode of tropical variability on interannual time scales and is considered as an internally occurring coupled ocean–atmosphere phenomenon. Whilst the ENSO phenomenon itself largely occurs in the tropical Pacific, it triggers climate teleconnections over the entire basin and in many other parts of the world (Figure AIV.3c,d). ENSO is consistently the main modulator of the global surface temperature at interannual time scales (Pan and Oort, 1983; Trenberth et al., 2002). ENSO is the main source of climate predictability on seasonal to interannual time scales (Philander, 1990; Smith et al., 2012) and its status is consequently closely monitored by meteorological institutes around the globe. El Niño events are typically identified by warmer than normal central/eastern equatorial Pacific SSTs, negative SLP anomaly difference between the eastern and western portions of the tropical Pacific Ocean, anomalous westerly surface winds, and increases in cloudiness and precipitation over the central and eastern equatorial Pacific and adjacent land areas. By contrast, La Niña events are typically identified by cooler than normal central/eastern equatorial Pacific SSTs, positive surface pressure anomaly difference between the eastern and western portions of the tropical Pacific Ocean and anomalous easterly surface winds reflecting altogether an intensification of the climatological east–west thermal gradient across the equatorial Pacific Ocean.

ENSO has been monitored through a range of indices, which are based on SST or SLP anomalies during the instrumental period, and various proxies on longer time scales. Differences between indices and thresholds used to define the occurrence of an El Niño or La Niña event imply that there is no commonly accepted list of events and associated magnitude. The three most prominent (Pepler, 2016; L’Heureux et al., 2017) indices used to monitor ENSO are the averaged SST anomalies over the so-called NINO3 (5°S–5°N, 150°W–90°W) or NINO3.4 (5°S–5°N, 170°W–120°W) regions and the Southern Oscillation Index (SOI) based on the normalized SLP difference between Darwin and Tahiti (Troup, 1965). Decomposition into EOFs of the tropical SST is also used in the literature. Those indices are highly correlated, as shown in Figure AIV.3b. The correlation between NINO3.4 SST and the EOF-SST index is about 0.97, and between NINO3.4 and SOI it is about 0.89 over the 1959–2019 period.

SST anomalies from the NINO3 and NINO3.4 regions provide a direct measure of the events via its surface ocean fingerprint, while the SOI provides a measure in terms of atmospheric perturbation. Data are available back to 1876 for the SOI computation, whilst the uncertainty in SST-based indices, particularly those associated with the central equatorial Pacific, is large prior to the 1950s when the number of SST observations was much lower (Kennedy et al., 2019). The complexity of ENSO is at least partly demonstrated by the observed events, which are all different in the magnitude, spatial structure and/or seasonal timing. As a direct consequence, it is difficult to accurately represent ENSO with one single index. Over time, further indices have been introduced to monitor ENSO event evolution (i.e., Trans Niño Index – TNI, Trenberth and Stepaniak, 2001) and spatial structure (i.e., El Niño Modoki Index – EMI, G. Li et al., 2010). The limited length and quality of the observational records may further affect the choice of the index, whose relevance can depend on the specific application. The variety in ENSO indices also explains the non-universality in the qualification of ENSO status delivered by meteorological agencies responsible for operational monitoring, operational forecast and climate services for users.

A particular aspect of ENSO diversity is whether the event is centred in the eastern Pacific (EP events) or central Pacific (CP or ‘Modoki’ events). This distinction is important as the type of event leads to different large-scale teleconnections and worldwide climatic impacts (e.g., Ashok et al., 2007; Ratnam et al., 2014; Capotondi et al., 2015; Timmermann et al., 2018; Taschetto et al., 2020). Differences can be significant at regional scale because of the displaced source of diabatic heating associated with the position of the main SST anomalies. For instance, stronger precipitation teleconnections occur over Australia during CP than during EP events (Wang and Hendon, 2007; Taschetto et al., 2009). It is noteworthy that central and eastern Pacific SST indices are more highly correlated with each other during La Niña than during El Niño events. This indicates that La Niña events have less geographic diversity than El Niño events (Kug and Ham, 2011) and that the CP/EP distinction is therefore more relevant for the latter. In the historical instrumental records, Banholzer and Donner (2014) found that CP events had a weaker influence on global surface temperature than EP events did.

However, the distinction between CP and EP events is not necessarily clearly defined, as different methods yield an inconsistent classification of the events, which implies that no robust consensus on the classification of CP or EP events is possible (Wiedermann et al., 2016). Furthermore, Giese and Ray (2011) found that the central longitude of peak SST anomalies associated with individual El Niño events had a distribution indistinguishable from a Gaussian distribution peaked at 140°W, as opposed to a bimodal distribution implied by the EP/CP paradigm. There are also occasional ‘coastal Niño’ events, such as those of 1925 (Takahashi and Martínez, 2019) or 2017 (Rodríguez-Morata et al., 2019), where abnormally warm waters are confined to the South American coast. Even though these events are not conventionally classified as El Niños, they have some El Niño-like teleconnections in South America. Instrumental data have historically been of limited value in assessing the classification of events before 1955, because of the almost total lack of SST measurements in the western half of the NINO3.4 region (L’Heureux et al., 2013). It is not yet known whether newer SST dataset versions, which benefit from improved data archives (Section 2.3.1.1), have reduced this uncertainty.

The description of ENSO in this Annex supports the assessments of Sections 2.4.2, 3.7.3, 4.3.3.2, 4.4.3.2 and 4.5.3.2. ENSO’s influence on ocean circulation, heat transport and sea levels are assessed in Sections 9.2 and 9.6.
Figure AIV.3 | The boreal winter El Niño–Southern Oscillation (ENSO) mode defined by December–January–February (DJF) sea surface temperature (SST) anomalies averaged over the NINO3.4 region (5°S–5°N, 170°W–120°W, cyan box) or extracted as the leading EOF mode over the tropical Pacific Ocean (the region denoted by the black box in (a)) for 1958–2019 using ERSSTv5. (a) SST anomalies regressed onto the NINO3.4 time series, which is shown in (b) as red and blue bars. Black curve in (b) represents the standardized leading principal component and the purple curve stands for the SOI index. The percentage of explained variance of the leading EOF is given in the legend, as well as the correlation between all indices and the NINO3.4 time series. (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth) and 10 m level wind (arrows; m s⁻¹ based on JRA-55) anomalies over 1958–2018. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1958–2016 and contours based on CMAP for 1979–2019 for every 0.5 mm day⁻¹). Magenta dots in (c, d) stands for the Darwin and Tahiti weather station used for the SOI computation. On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance. Significance for CMAP precipitation is materialized by red contours in (d). All fields have been linearly detrended prior to computation.
AIV.2.3.2 Teleconnections and Regional Influence

ENSO has a broad range of teleconnections, both in and outside the Pacific region (Figure AIV.3c,d). Amongst the best-known teleconnections are the association of El Niño events with increased precipitation in the south-western and south-eastern USA, eastern Africa, coastal Ecuador and Peru, and northern Argentina, and with decreased precipitation in eastern Australia and the Maritime Continent, southern Africa, the Amazonian basin and much of the Indian subcontinent. A general reverse association is found during La Niña events although there are some asymmetries (Cai et al., 2010; Taschetto et al., 2020). ENSO also has a substantial impact on tropical cyclone occurrence (Wang and Chan, 2002; Chu, 2004; Kuleshov et al., 2008), with El Niño typically being associated with reduced tropical cyclone activity in the North Atlantic, and an eastward displacement of western North Pacific and South Pacific tropical cyclone genesis away from the Philippine Sea and the Australian continent and into the central Pacific. ENSO has been shown to control part of the interannual variability of fire, including wildfire and human-triggered fires, in many regions of the world including all the continents surrounding the Pacific (see for instance the devastating extreme fire emissions in Indonesia in 2015; Chisholm et al., 2016) but also remotely, for example, over the Arctic (Monks et al., 2012).

Tropical climate impacts during ENSO events are largely driven by a reorganization of the zonal Walker circulation that results from anomalous diabatic heating associated with deep atmospheric convection changes (e.g., Gill, 1980; Klein et al., 1999; Chiang and Sobel, 2002; Choi et al., 2015). The latter are also responsible for atmospheric teleconnections to higher latitudes (Yeh et al., 2018), acting as a source of forced large-scale Rossby waves (Hoskins and Karoly, 1981). Specific phases of the Pacific-North American (PNA) pattern (Horel and Wallace, 1981) and Pacific-South American (PSA) pattern (Karoly, 1989) are then favoured in the Northern and Southern hemispheres, respectively, and are associated with shift and modulation of the mean storm tracks. These teleconnections influence temperature and rainfall around the globe, including extreme events (King et al., 2014). For instance, mild winter conditions dominate Alaska/Canada, connected to the south-eastward displacement and reinforcement of the Aleutian Low during El Niño, which generates anomalous southerlies and advects warmer air northward. By contrast, the anomalous cyclonic circulation favours cold outbreaks over the far-east Siberian region and Kamtchatka (Figure AIV.3c,d).

Being the dominant actor of variability at interannual time scales, ENSO is linked to many other modes of variability both in the tropics and mid-latitudes, either in phase or with some seasonal lag. For instance, connection has been documented between EP ENSO events and SAM during austral summer (Ciasto et al., 2015). A relationship exists between ENSO and the NAO/NAM in boreal winter through various mechanisms such as stratosphere–troposphere coupling (Domeisen et al., 2015) and tropospheric pathways, including the eastward extension of the PNA during El Niño events and also through the alteration of the propagation of tropical Rossby wave trains from the Pacific to the Atlantic (Drouard et al., 2015). The tropical-extratropical teleconnection is not symmetrical with the phase of ENSO and non-stationary with considerable dependence on epochs being modulated by the inherent strong internal variability (Deser et al., 2017; Drouard and Cassou, 2019). In the Atlantic, a relationship between ENSO and Atlantic Niño (Section AIV.2.5) is found in boreal spring (Tokinaga et al., 2019) while the South Atlantic Subtropical Dipole mode is mostly connected to ENSO in austral summer (Rodrigues et al., 2015).

Implications of ENSO in water cycle changes are further assessed in Sections 8.3.2.9 and 8.4.2.9. The association of ENSO with extremes and specific regional climate anomalies is assessed in Sections 11.4, 11.6, 11.7 and 12.4, in Box 11.4, and in Atlas.4, Atlas.5, Atlas.6, Atlas.7 and Atlas.9.

AIV.2.3.3 Available Proxy-based Reconstruction

Model evidence suggests that ENSO characteristics vary on decadal to centennial time scales (e.g., Lewis and LeGrande, 2015; Deser et al., 2017), implying that a minimum of several hundred years of observations, which is much longer than the instrumental record, are required to determine the full natural range of ENSO complexity (Wittenberg, 2009). Thus, one way to understand if anthropogenic forcing has modulated ENSO is to understand ENSO behaviour prior to the observational record. Paleoclimate archives can be used to fill this gap as they provide quantitative estimates of ENSO behaviour during the pre-industrial era, while also providing information of ENSO characteristics under a variety of past natural climate background states controlled by external forcings. Extensive proxy-based studies, using a range of ENSO-sensitive records, have been carried out to assess the ENSO changes on various paleoclimate time scales, specifically the last few hundred years and back to the mid-Holocene (Section 2.4.2).

Reconstructions of past ENSO behaviour traditionally merges information from multiple proxies that are considered sensitive to ENSO (Lu et al., 2018). This merging process may have multiple purposes, but it is generally used to extend the reconstruction length and also to increase its signal-to-noise ratio. The resulting reconstructions can be broadly placed into two main categories: (i) those that utilize proxies in the central and eastern equatorial Pacific itself; and (ii) those that utilize proxies from other regions impacted through teleconnections, along with those that mix both categories. ENSO reconstructions from category (i) are considered to be a more accurate measure of ENSO behaviour. However, proxies from the central and eastern equatorial Pacific are generally rarer than those outside the region, due to the small number of land/reefs, and they also tend to be of shorter duration. Thus, the ideal case is a combination of both categories to get a robust quantitative estimate of ENSO behaviour in the past.

AIV.2.4 Indian Ocean Basin and Dipole Modes

AIV.2.4.1 Definition of the Mode

The Indian Ocean Basin (IOB) and Indian Ocean Dipole (IOD) modes are the two dominant modes of interannual climate variability in the Indian Ocean. The IOB mode features the temporal alternation of basin-wide warming and cooling of sea surface in the positive and negative phases, respectively (Figure AIV.4), while the IOD mode is...
Figure AIV.4 | The boreal spring Indian Ocean Basin (IOB) mode defined by March–April–May (MAM) SST anomalies averaged over the Indian Ocean (the region denoted by the cyan box in a) or extracted as the leading EOF mode over the same domain (black lines) for 1958–2019 using ERSSTv5. (a) SST anomalies regressed onto the IOB index time series shown in (b) in red and blue bars, while the black curve in (b) represents the leading principal component time series. Explained variance and correlation between indices are given in the legend in (b). (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth) and 10 m level wind (arrows; m s⁻¹ based on JRA-55) anomalies for 1958–2018. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1958–2016 and contours based on CMAP for 1979–2019 for every 0.3 mm day⁻¹). On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance. Significance for CMAP precipitation is materialized by red contours in (d). All fields have been linearly detrended prior to computation.
The Indian Ocean Dipole Mode (IOD)

a. SST spatial pattern for IOD in Sept-Nov (SON)

b. SON IOD temporal evolution from several metrics

Annex IV

Figure AIV.5 | The boreal autumn Indian Ocean Dipole (IOD) mode defined by September–October–November (SON) standardized SST difference between the western (10°S–10°N, 50°E–70°E) and eastern (10°S–0°, 90°E–110°E) equatorial Indian Ocean denoted by cyan boxes in (a) or extracted as the leading empirical orthogonal function (EOF) mode over the Indian Ocean (the region denoted by black lines in (a)) over 1958–2019 using ERSSTv5. (a) SST anomalies regressed onto the IOD index time series shown in (b) in red and blue bars, while the black curve represents the leading principal component time series. Explained variance and correlation between indices are given in the legend in (b). (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth) and 10 m level wind (arrows); m s⁻¹ based on JRA-55 anomalies for 1958–2018. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1958–2016 and contours based on CMAP for 1979–2019 for every 0.3 mm day⁻¹). On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance. Significance for CMAP precipitation is materialized by red contours in (d). All fields have been linearly detrended prior to computation.
characterized by a zonal dipole of SST anomalies in the tropics, with one lobe centred off Sumatra and Java and the other covering a large portion of the western basin (Figure AIV.5). The two modes mostly vary on interannual time scales and are both correlated positively with ENSO. In particular, the IOD tends to emerge during ENSO development from boreal summer to autumn (Stuecker et al., 2017), followed by the IOB which grows in ENSO peak season and persists for a few subsequent seasons (Tokinaga and Tamimto, 2004).

The IOB mode is considered primarily as a response to ENSO associated with large-scale alteration of the Walker circulation, and thus matures in boreal spring, a season after the typical ENSO peak (Klein et al., 1999; Lau and Nath, 2000, 2003). By contrast, the Bjerknes feedback is key to the formation of the IOD. This feedback requires background surface easterlies and thermocline shallowing in the eastern part of the basin along the equator, and this condition sets in boreal spring and persists until autumn, then dissipating by early winter in association with climatological monsoon wind swing. The IOB mode therefore develops in boreal summer and reaches its maximum amplitude in autumn, then rapidly terminates before winter (Schott et al., 2009). Differently from the IOB, strong internal feedbacks enable the IOD mode to develop also without influence from ENSO, as noted for some specific events (Saji and Yamagata, 2003b; Meyers et al., 2007).

The IOB and IOD are identified as the first and second EOF modes, respectively, of monthly SST variability in the tropical Indian Ocean (Weare, 1979; Saji et al., 1999). The IOB can be extracted also as the first EOF of SST variability in boreal autumn, although the pattern depends on the analysis period (Piilai et al., 2019). The IOD mode has a sub-surface signature and can be also calculated as the leading EOF mode of the 20°C isotherm depth of the Indian Ocean seawater temperature (Saji et al., 2006). Apart from the EOF definitions, the tropical Indian Ocean averaged SST (20°S–20°N, 40°E–100°E or 40°E–120°E) is often used as the IOB mode index (Figure AIV.4a). For March–April–May (MAM), its correlation with the leading PC of Indian Ocean SST (Figure AIV.4b) reaches 0.98. For the IOD, the difference between normalized SST anomalies between the western (10°S–10°N, 50°E–100°E) and eastern (10°S–0°, 90°E–110°E) equatorial Indian Ocean is a widely used index (Saji et al., 1999; Figure AIV.5). It is highly correlated (at 0.76) with the leading principal component of the Indian Ocean September–October–November SST when assessed over 1958–2019 (Figure AIV.5b).

The IOB-like (Han et al., 2014) and IOD-like (Ashok et al., 2004a; Tozuka et al., 2007) variability can also be identified at decadal time scales. The decadal IOB mode is defined as the leading EOF mode of linearly detrended monthly SST after decadal time filtering (Han et al., 2014; Dong and McPhaden, 2017). Ten-year low-pass filtered Indian Ocean basin-mean SST is an alternative index for the decadal IOB mode (Dong et al., 2016), with the global-mean SST subtracted in some cases as an estimation of the response component to radiative forcing (Huang et al., 2019). The decadal IOD mode can be extracted as the second EOF mode of 10-year low-pass filtered Indian Ocean SST in boreal autumn (Yang et al., 2017), while other studies use the 10-year low-pass filtered SST anomaly difference between the western and eastern equatorial Indian Ocean (Ashok et al., 2004a; Tozuka et al., 2007).

Paleo fluctuations of IOB and IOD as well as the changes over the instrumental record period are assessed in Section 2.4.3, while their model representations are assessed in Section 3.7.4. Future variations under global warming are assessed in Sections 4.4.3.3 (near term) and 4.5.3.3 (mid- to long terms).

AIV.2.4.2 Teleconnections and Regional Influence

Since the IOB mode is tightly associated with ENSO, related climate anomalies can be viewed as part of ENSO’s influence from boreal winter to subsequent spring. Still, the IOB mode plays an important role in modulating the ENSO influence on several regions. The positive phase of the IOB plays a major role in suppressing precipitation in the Maritime Continent from boreal winter to spring (Watanabe and Jin, 2002; Annamalai et al., 2005) and in Northern Australia in austral summer (Taschetto et al., 2011) in conjunction with El Niño events (Figure AIV.3d). An anomalous surface anticyclone then forms over the Philippine Sea linked to the precipitation reduction in the tropical western Pacific and brings more precipitation to East Asia during El Niño winter and spring (Figure AIV.3d; Wang et al., 2000).

ENSO’s equatorial Pacific SST anomalies typically dissipate by boreal summer, when the lingering IOB then plays the leading role in exerting climate anomalies in Asia and the western North Pacific (the Indian Ocean capacitor effect; Xie et al., 2009). The boreal summer IOB induces meridionally dipolar anomalies of precipitation and surface air temperature in South East and East Asia, with wetter and cooler summers in mid-latitude East Asia and drier and warmer summers in South East Asia during positive IOB events (Kosaka et al., 2013; Wang et al., 2013). These conditions affect the occurrence of extremes such as heavy rainfall (often caused through atmospheric rivers; Kamae et al., 2017) and heatwaves (Hu et al., 2012; Deng et al., 2019). The positive IOB suppresses summer tropical cyclone formation in the western North Pacific (Du et al., 2011; Ueda et al., 2018), leading to a delayed onset of the typhoon season (Zhao et al., 2019). These regional influences are already present in the spring season (Figure AIV.4c,d) but tend to persist and even get reinforced due to the persistence of the IOB, while following the seasonal climatological displacement of tropical dynamics (monsoon, etc.). In South Asia, tripolar precipitation anomalies form in boreal summer, with precipitation increase along the Western Ghats and in Bangladesh, and decrease around the Ganges Delta during the positive IOB (Chowdary et al., 2013, 2019).

The IOD is associated with climate anomalies in broad regions (Figure AIV.5c,d). However, the related anomalies are affected by influences from concomitant ENSO. The IOD affects rainfall in eastern Africa, South Asia, Indonesia and Australia (Figure AIV.5c,d). The IOD in the positive phase increases rainfall in eastern tropical Africa in boreal autumn to early winter (Figure AIV.5d), while the negative phase induces the opposite anomalies. These anomalies are also found in association with ENSO (Figure AIV.3d), but the explained variance by IOD is larger (Saji and Yamagata, 2003a). Both positive IOD and El Niño events reduce precipitation and temperature over the entire Maritime Continent (Figures AIV.5c,d and AIV.3c,d) because of strong low-level divergence coupled with upper-level convergence related to diminished convection, while the opposite anomalies are induced by negative IOD and La Niña episodes (Saji and Yamagata, 2003a).
The positive phase of the IOD also decreases rainfall in western and south-eastern Australia (Ashok et al., 2003; Risbey et al., 2009), acting as preconditioning for fires (Cai et al., 2009). The rainfall anomalies cover the whole of Australia when combined with simultaneous ENSO with the same polarity (Risbey et al., 2009). The IOD influences South Asian summer monsoon rainfall either directly through moisture transport over the western Indian Ocean or modification of the local Hadley cell (e.g., Ashok et al., 2001; Gadgil et al., 2004; Ashok and Saji, 2007; Behera and Ratnam, 2018), or indirectly in the framework of the tropospheric biennial oscillation (e.g., Meihl et al., 2003; Li et al., 2006; Webster and Hoyos, 2010). IOD and ENSO often interfere in their influence on the South Asian summer monsoon (e.g., Ashok et al., 2001, 2004b; Pepler et al., 2014; Crépit et al., 2017). Opposing influences of IOD and ENSO have been found on surface temperature in summer in northern East Asia (Saji and Yamagata, 2003a).

Implications of these modes on water cycle changes and specific regional influences are further assessed in Sections 8.3.2.9, 8.4.2.9 and 12.4, and Atlas.5, Atlas.6, Atlas.7 and Cross-Chapter Box Atlas.2, respectively.

### AIV.2.4.3 Available Proxy-based Reconstruction

Corals record interannual and longer variability in the Indian Ocean (Zinke et al., 2005, 2009). Several studies have developed IOD reconstructions based on oxygen isotope ratios from corals. These reconstructions include those based on samples from Kenya since 1887 (Kayanne et al., 2006), from the Mentawai Islands in western Indonesia for the last 6500 years (Abram et al., 2007), and in combination with samples from Bali in southern Indonesia and the Seychelles since 1846 (Abram et al., 2008). Abram et al. (2020) developed a coral-based reconstruction of the IOD for the last millennium, albeit with some discontinuity. Niedermeyer et al. (2014) use isotope compositions in terrestrial plant wax to reconstruct rainfall changes over north-western Sumatra, which are highly correlated with the IOD. Compared to the IOD, proxy-based IOD reconstruction is limited. Yet Du et al. (2014) find that coral oxygen isotope records from the Seychelles are well correlated with the IOD, suggesting that they can be used as an IOD proxy. Evidence from other proxies that are potentially associated with the Indian Ocean variability is assessed in Section 2.4.3.

### AIV.2.5 Atlantic Meridional and Zonal Modes

#### AIV.2.5.1 Definition of the Mode

The Atlantic Zonal Mode (AZM), also commonly referred to as the Atlantic Niño, and the Atlantic Meridional Mode (AMM) are the two dominant modes of tropical Atlantic climate variability on interannual to decadal time scales. The AZM peaks during the boreal summer upwelling season with maximum SST anomalies in the eastern equatorial Atlantic and is commonly regarded as the Atlantic counterpart of the Pacific El Niño (Zebiak, 1993; Keenlyside and Latif, 2007; Lübbecke et al., 2018; Foltz et al., 2019). The AMM, which is more active on longer time scales, peaks during boreal spring and exhibits a cross-equatorial gradient of SST with opposite signs in the northern and southern portion of the Atlantic. Both modes are associated with changes in the Inter-tropical Convergence Zone (ITCZ) and related winds, and exert a strong influence on the climate in adjacent and remote regions.

The AMM is thought to be primarily driven by air-sea heat fluxes and therefore thermodynamic in nature (Chang et al., 2000; Foltz and McPhaden, 2010), while various mechanisms for generating AMM events have been discussed. They mainly concern fluctuations in the wind field over the equatorial Atlantic. These wind-stress anomalies can be excited by local SST changes, as part of Bjerknes feedback (i.e., weakening of trade winds and thermocline slope adjustments via oceanic equatorial Kelvin wave propagation; Keenlyside and Latif, 2007; Deppenmeier et al., 2016). They can also be excited by meridional advection of temperature anomalies (Richter et al., 2013), generated remotely as a response to ENSO (Latif and Grötzner, 2000; Tokinaga et al., 2019) or to variations in the South Atlantic Subtropical High (Lübbecke et al., 2014), or occur as stochastic wind variability (Richter et al., 2014). In contrast to previous mechanisms involving coupled atmosphere–ocean dynamics, Bellomo et al. (2015) and Nnamchi et al. (2015) suggested that a large part of the SST variability in the eastern equatorial Atlantic can be explained by thermodynamic feedbacks excited by stochastic atmospheric perturbations. A recent study based on observational and reanalysis products by Nnamchi et al. (2021) indicates that the seasonality of AZM is indeed largely controlled by the variability of the atmospheric diabatic heating. However, the balance between thermodynamic and dynamic processes determining SST variability in the equatorial Atlantic is subtle, and other studies (e.g., Jouanno et al., 2017; Dippe et al., 2018) conclude that, even if the thermodynamic component is not negligible, the dynamics is the dominant driver of the AZM.

The AZM and AMM represent the first and second EOF patterns of the tropical Atlantic variability identified using SST anomalies in this region (Figures AIV.6a and AIV.7a). These modes may also be defined using box averages, with the AZM defined as SST anomalies averaged over the equatorial Atlantic region called ATL3 (3°S–3°N, 0°–20°W), and the AMM defined as the difference between the normalized SST anomalies averaged over the tropical North Atlantic and tropical South Atlantic (i.e., 5°N–30°N, 20°W–60°W minus 5°N–20°S, 5°E–25°W). In boreal summer, the ATL3 and the EOF-AMM PC are correlated at 0.87 during 1958–2019 (Figure AIV.6b). The correlation between the AMM-box index and the EOF-AMM PC is lower at 0.39 (Figure AIV.7b) but increases to 0.67 over 1900–2019.

For the AMM, the latitudinal polarity in SST anomalies is associated with wind gradient and flow from the cooler to the warmer hemisphere (Figure AIV.7c; Ruiz-Barradas et al., 2000; Chiang and Vimont, 2004; Lübbecke et al., 2018; Foltz et al., 2019). For the AZM, similarly to ENSO in the Pacific Ocean, anomalous westerlies lead to low-level convergence over warmer SST along the climatological cold tongue (Figure AIV.6c).

The temporal evolution of the Atlantic modes of variability and their model representation are assessed in Sections 2.4.4 and 3.7.5, respectively. Their near-term and long-term evolution is assessed in Sections 4.4.3.4 and 4.5.3.4, respectively.
Figure AIV.6 | The boreal summer Atlantic Zonal Mode (AZM) defined by June–July–August (JJA) box-averaged sea surface temperature (SST) anomalies over the ATL3 region (3°S–3°N, 0°–20°W, cyan box in a) or estimated as the leading empirical orthogonal function (EOF) over the tropical Atlantic Ocean (the region denoted by the black box in a) for 1958–2019 using ERSSTv5. (a) SST anomalies regressed onto the ATL3 time series, which is shown in (b) as red and blue bars, while the black curve represents the leading principal component time series. Explained variance and correlation between indices are given in the legend in (b). (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth) and 10 m level wind (arrows; m s⁻¹ based on JRA-55 for 1958–2019) anomalies. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1958–2016 and contours based on CMAP for 1979–2019 for every 0.2 mm day⁻¹). On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance. Significance for CMAP precipitation is materialized by red contours in (d). All fields have been linearly detrended prior to computation.
The Atlantic Meridional Mode (AMM)

**a. SST spatial pattern for AMM in June-Aug (JJA)**

Not significant at 10% level

°C (ERSSTv5)

**b. JJA AMM temporal evolution from several metrics**

Normalized index


AMM box index

EOF1 (25% Corr=0.39)

**c. JJA AMM teleconnection for 2m-temperature/10m-wind**

°C (Berkely)

-0.36 -0.24 -0.12 0 0.12 0.24 0.36

**d. JJA AMM teleconnection for precipitation**

CMAP dataset

Significant at 10% level

-0.5 -0.4 -0.3 -0.2 -0.1 0 0.1 0.2 0.3 0.4 0.5 mm/day (GPCC)

Figure AIV.7 | The boreal summer Atlantic Meridional Mode (AMM) defined by June–July–August (JJA) standardized sea surface temperature (SST) difference between the north (5°N–30°N, 20°W–60°W) and south (5°S–20°S, 5°E–25°W) tropical Atlantic Ocean shown by the cyan boxes in (a) or estimated as the leading empirical orthogonal function (EOF) over the tropical Atlantic Ocean (the region denoted by the black box in a) for 1958–2019 using ERSSTv5. (a) SST anomalies regressed onto the AMM time series, which is shown in (b) as red and blue bars, while the black curve represents the leading principal component time series. Explained variance and correlation between indices are given in the legend in (b). (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth) and 10 m level wind (arrows; m s⁻¹ based on JRA-55 for 1958–2019) anomalies. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1958–2016 and contours based on CMAP for 1979–2019 for every 0.2 mm day⁻¹). On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance. Significance for CMAP precipitation is materialized by red contours in (d). All fields have been linearly detrended prior to computation.
AIV.2.5.2 Teleconnections and Regional Influence

The tropical Atlantic variability has robust teleconnections with climate variability across the global tropics and beyond. There is evidence for two-way teleconnections between the tropical Atlantic and Pacific on interannual to decadal time scales, such that the tropical Atlantic variability responds and feeds back to the Pacific ENSO (Section AIV.2.3) and Pacific Decadal Variability (PDV; Section AIV.2.6). A number of studies suggest the existence of a possible connection between the tropical North Atlantic and tropical Pacific variability (Wu et al., 2007; Ham et al., 2013a, b; L. Wang et al., 2017) at interannual time scales. This teleconnection appears to be modulated at a decadal time scale by the AMV phases (L. Wang et al., 2017) and it is also affected by global warming (Dong and Zhou, 2014).

The AZM-related boreal summer variability is linked with ENSO in the following boreal winter (Polo et al., 2008; Rodríguez-Fonseca et al., 2009; Ding et al., 2012; Keenlyside et al., 2013; Martín-Rey et al., 2015; Cai et al., 2019). This relationship is strongest during negative AMV phases due to enhanced equatorial Atlantic variability (Martín-Rey et al., 2014, 2018; Polo et al., 2015). On the other hand, the persistence of SST anomalies during multi-year ENSO events affects the equatorial Atlantic variability through modifications of the Walker circulation (Tokinaga et al., 2019). An exhaustive analysis of the AZM characteristics and related teleconnections is given in Lübbecke et al. (2018) and by Cabos et al. (2019).

The AZM affects the Indian summer monsoon (ISM), altering the ENSO-ISM connection (Kucharski et al., 2007, 2008; Wang et al., 2009; Barimalala et al., 2012, 2013; Kucharski and Joshi, 2017). Specifically, warm SST in the equatorial Atlantic leads to decreased monsoon rainfall over central India, and the opposite occurs for negative Atlantic SST anomalies via strengthening of the Somali Jet and low-level convergence (Kucharski et al., 2007, 2008; Pottapinjara et al., 2016). This latter relationship opens up the possibility of using AZM phases to predict ISM rainfall (Sabeerali et al., 2019). In this framework, the AZM modulates SST variability in the Indian Ocean (Kajtar et al., 2017). The number of monsoon depressions over the Bay of Bengal increases during the cold phase of the AZM, which is remotely responsible for changes in low-level cyclonic vorticity and mid-tropospheric humidity (Pottapinjara et al., 2014). A relationship between AZM and the West African Monsoon is documented as well, with a reduction of rainfall over the Sahel and an increase over Guinea (Losada et al., 2010). But the relationship with the Sahel shows significant changes across the 20th century (Joly and Voldoire, 2010; Losada et al., 2012), which is strongly indicative for non-stationarity of the teleconnection.

On the western side of the basin, a warm equatorial Atlantic delays the northward migration of the ITCZ and can thus influence rainfall over North-Eastern South America (Nobre and Srukla, 1996), even before the peak of the AZM (Losada et al., 2010; Mohino et al., 2011; Martín-Rey et al., 2018). When the Atlantic and Pacific basins act together, the response over North-Eastern South America is found to be strongly enhanced (Torrabala et al., 2015). In the extratropics, the decaying phase of the AZM in boreal autumn can modify the atmospheric circulation over Europe via a circum-global teleconnection pattern (Haarsma and Hazeleger, 2007; García-Serrano et al., 2011), and can remotely affect the NAO (Dréevillon et al., 2003; Peng et al., 2005). The AZM can also increase precipitation over southern Europe and the Mediterranean Sea during boreal summer through extratropical Rossby waves (Losada et al., 2012).

The fluctuation of the meridional SST gradient in the tropical Atlantic modulates the seasonal march of the ITCZ, thus influencing regional rainfall over north-eastern Brazil and the Sahel (Hastenrath and Heller, 1977; Folland et al., 1986; Foltz et al., 2012). During a negative phase of the AMM, northerly wind anomalies are associated with a southward shift of the ITCZ (Cabos et al., 2019). During boreal spring, warmer SST over the tropical North Atlantic modulates winds over the western Indian Ocean and may strengthen the monsoon circulation over India, affecting the continental rainfall (Vittal et al., 2020). The AMM is also well known to affect the Atlantic hurricane activity (Vimont and Kossin, 2007; Patricola et al., 2014) and tropical cyclones over the western North Pacific (Zhang et al., 2017).

AIV.2.5.3 Available Proxy-based Reconstruction

Only a few paleo-reconstructions of AMM and AZM are available. Inter-hemispheric cross-equatorial SST gradients linked to changes in ITCZ locations characteristic of the AMM have been found during the Last Glacial Maximum, Heinrich Stadial 1 and the mid-Holocene (McGee et al., 2014). Similarly, the dipole-like SST pattern in the South Atlantic sub tropics, which is related to the AZM (Lübbecke et al., 2018; Foltz et al., 2019), has been reconstructed with SST proxies from marine sediment cores during the past 12 kyr (Wainer et al., 2014).

Evidence from proxies associated with the AZM and AMM is assessed in Section 2.4.4.

AIV.2.6 Pacific Decadal Variability

AIV.2.6.1 Definition of the Mode

The Pacific Decadal Variability (PDV) describes the large-scale fluctuations typically observed beyond ENSO time scales in a variety of instrumental records and proxy reconstructions over the entire Pacific Ocean and surrounding continents.

Historically, climate variability of the Pacific Ocean on time scales longer than ~8–10 years has been described in terms of statistical modes, usually assessed through the decomposition into EOFs of the SST over oceanic sub-basins. The leading EOF mode of SST decadal variability in the extratropical North Pacific is called the Pacific Decadal Oscillation (PDO; Mantua et al., 1997; Mantua and Hare, 2002). The PDO is recognized to correspond to a collection of multiple processes (Newman et al., 2016). On interannual-to-decadal time scales, the PDO is thought to be driven by atmospheric forcing linked to stochastic changes in the Aleutian Low integrated by ocean mixed layer dynamics and related re-emergence processes, as well as to ENSO-forced teleconnections (Schneider and Cornuelle, 2005; Nidheesh et al., 2017). On decadal-to-inter-decadal time scales, subpolar–subtropical gyre dynamics and related zonal advection
of temperature anomalies in the Kuroshio-Oyashio extension and westward-propagating oceanic Rossby waves are thought to be key mechanisms (Qiu et al., 2007; Taguchi et al., 2007; Wills et al., 2019).

The PDO’s South Pacific counterpart, called the South Pacific Decadal Oscillation (SPDO; Chen and Wallace, 2015) is defined as the leading EOF mode of SST in the extratropical South Pacific. As for the PDO, the SPDO is thought to be driven by internal climate variability associated with extratropical atmospheric modes (e.g., SAM; Section AIV.2.2), ENSO teleconnections (Section AIV.2.3) and ocean dynamics through gyre circulation (Shakun and Shaman, 2009; Zhang et al., 2018).

In the tropical Pacific, while ENSO has a clear interannual spectral peak, decadal variability is also present with a similar spatial structure (Zhang et al., 1997), called tropical Pacific Decadal Variability (TPDV) or decadal ENSO. Relative occurrence of El Niño and La Niña events in a given decade correlates with the phase of the TPDV (Power and Colman, 2006).

Despite the PDO, SPDO and TPDV being considered to be physically distinct modes (Newman et al., 2016), they are highly correlated in observations (Chen and Wallace 2015) and they collectively constitute an equatorially symmetric pattern of basin-wide variability, often called the Inter-decadal Pacific Oscillation (IPO; Power et al., 1999; Folland et al., 2002; Henley et al., 2015). There is evidence that ENSO and the TPDV drive the PDO and SPDO in the North and South Pacific through tropical–extratropical teleconnections and the ‘reddening’ (i.e., selective retention of decadal and longer variability) in the ocean, ensuring the synchronization of decadal variability in both hemispheres (Newman et al., 2016). Besides, positive feedbacks are suggested between the TPDV and the extratropical North Pacific through the Pacific Meridional Mode (Stuecker, 2018), which is the Pacific analogue of the AMM (Section AIV.2.5; Chiang and Vimont, 2004), and modulations of the shallow subtropical cell of the Pacific ocean circulation (Farneti et al., 2014).

Whether the IPO is a distinct mode of climate variability (Henley et al., 2017) or not (Tung et al., 2019) remains to be further assessed and clarified. However, since the PDO, SPDO, TPDV and IPO modes are highly correlated both spatially and temporally, they can be referred to collectively as the Pacific Decadal Variability mode (PDV; Henley, 2017; Liu and Di Lorenzo 2018), similarly to the Atlantic (Section AIV.2.7). Besides, it is noteworthy that none of these modes are actually oscillations with clear spectral peaks in the instrumental period (Mann et al., 2020), which advocates for the use of the PDV acronym across the report.

De facto, there is no unique way to assess the PDV. It can be extracted from the IPO, defined as the leading EOF mode of decadal low-pass filtered SST variability over the entire Pacific to remove the overwhelming influence of ENSO at interannual time scales. Henley et al. (2015) introduced the tripole index (TPI) as the difference of SST anomalies between an equatorial Pacific domain (10°S–10°N, 170°E–90°W) and an average of mid-latitude North (25°N–45°N, 140°E–145°W) and South (50°–15°S, 150°E–160°W) Pacific domains (Figure AIV.8a). Section 2.4.5 describes the changes in the PDV phases over the instrumental record using a variety of SST products.

A positive phase of the PDV is characterized by warm conditions in the central-eastern tropical Pacific from the dateline to the coasts of the Americas, encircled to the west in the mid-latitudes by negative SST anomalies forming a so-called horseshoe pattern. The pattern has a certain similarity with ENSO and related anomalous SST fingerprints through atmospheric teleconnections. However, the PDV features meridionally broader SST anomalies in the tropical Pacific extending to the subtropics compared to ENSO, and relatively stronger extratropical SST anomalies. Alternatively, the low-pass filtered SST regression on the PDO index, defined as the leading principal components of SST (Mantua et al., 1997) or sea surface height (Di Lorenzo et al., 2008) over the extratropical North Pacific north of 20°N, are used in literature for PDV. Their corresponding spatial patterns are, by definition, weighted to the respective hemisphere of the Pacific. Figure AIV.8b provides evidence that they are all highly correlated to each other. The correlation value between the TPI and the PDO index is equal to 0.76, and when low-pass filtered, it is greater than 0.9 for all the combination of indices.

In the following paragraphs and in Section 3.7.6 evaluating the model performance in reproducing PDV and assessing the human influence on the mode evolution, the TPI index is preferred. The latter definition is found to be less sensitive than classical EOF analysis to the possible interplay between internal processes that intrinsically drive the PDV (Wills et al., 2018) and the externally-forced SST response simulated in historical simulations. There is no unique way to remove the impact of the external forcing in the observations at decadal to multi-decadal time scales, and any chosen method may have significant implications for the interpretation of the PDV expression during the instrumental era.

Since AR5, there has been considerable progress in the understanding of PDV itself, but also in its critical role in modulating the evolution of observed global surface temperature (Cross-Chapter Box 3.1) as well as regional climate through teleconnections. On interannual time scales, ENSO is the leading internal driver of global surface temperature variability (Section AIV.2.3). Since PDV encompasses decadal modulations of El Niño versus La Niña occurrences, PDV affects global surface temperature on decadal time scales, with its positive and negative phases tied to higher and lower global surface temperature, respectively. Indeed, studies identify PDV as the leading mode of variability associated with unforced decadal global surface temperature fluctuations in observations and a majority of CMIP5 models (Brown et al., 2015; Dai et al., 2015), with additional influence from AMV (Section AIV.2.7; Tung and Zhou, 2013). PDV thus plays an important role in decadal acceleration and slowdown of global surface temperature trends (Cross-Chapter Box 3.1). So far, little predictability has been found for decadal PDV, despite its global relevance, and decadal prediction in the Pacific remains challenging with the current decadal forecast system (Section 4.4.3.5). Finally, there is an increasing body of evidence about the existence of a network of teleconnections related to decadal climate-variability phenomena (Cassou et al., 2018; Cai et al., 2019). For instance, AMV has been hypothesized to influence in part the phase of the PDV (McGregor et al., 2014; Ruprich-Robert et al., 2017), and consistently there is a significant negative correlation between PDV and AMV. Other studies suggest that the PDV may be the main driver of the
The Pacific Decadal Variability (PDV)

a. SST spatial pattern for PDV (annual mean)

b. Annual PDV temporal evolution from several metrics

- PDO (Corr=0.66)
- IPO (Corr=0.96)
- DCPP-CMIP6 (Corr=0.88)
- TPI index
- 10-yr MP TPI

Figure AIV.8 | The Pacific Decadal Variability (PDV) based on the tripole index (TPI) for 1900–2014 after 10-year low-pass filtering. (a) Sea surface temperature (SST) anomalies regressed onto TPI based on ERSSTv5. The cyan boxes indicate regions for the TPI definition. (b) Temporal evolution of the raw TPI (bars), the 10-year low-pass filtered TPI (cyan curve), the empirical orthogonal function (EOF)-based Pacific Decadal Oscillation (PDO) index (dashed black) and two estimations of the Inter-decadal Pacific Oscillation (IPO) EOF-based indices. In solid black, linear detrending is applied to annual data prior to the computation of the EOF of the low-pass filtered SST to remove the forced component in the SST field. In orange, the forced signal is estimated from CMIP5 historical simulations following Ting et al. (2009). The latter metric is used in CMIP6 DCPP-C sensitivity experiments (Boer et al. 2016). The correlation between the low-pass filtered TPI and all PDV indices is given in the legend. (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth for 1900–2014) and 10 m level wind (arrows; m s⁻¹ based on JRA-55 for 1958–2014) anomalies. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1900–2014). All are based on annual mean. On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance.
decadal IOB and IOD modes (Section AIV.2.4), while influence from the Indian Ocean SST on PDV has been also suggested (Luo et al., 2012; Mochizuki et al., 2016).

The temporal evolution of PDV in the instrumental records and its model representation are assessed in Sections 2.4.5 and 3.7.6, respectively. Skill in predicting near-term temporal evolution of the PDV and mid- to long-term PDV changes are assessed in Sections 4.4.3.5 and 4.5.3.5, respectively.

AIV.2.6.2 Teleconnections and Regional Influence

Due to the similarity of the SST-anomaly pattern with ENSO in the tropical Pacific, atmospheric teleconnections associated with the PDV are rather similar to those of ENSO but with a longer temporal expression. This includes the long-term alteration of the Walker circulation, the modification of the extratropical background driven by tropically forced Rossby waves then favouring particular phases of the PNA and PSA patterns of atmospheric internal variability that are responsible for persistent influences over the continents (e.g., persistent drought over California and Australia, etc.). Those atmospheric patterns imprint PDO-like and SPDO-like SST patterns with a reddening by the extratropical oceans, forming the pan-Pacific SST anomaly pattern of the PDV as described above (Newman et al., 2016; Henley, 2017; Liu and Di Lorenzo, 2018).

A positive phase of the PDV brings anomalously warm conditions in the north-western corner of North America, with maximum loading over Alaska linked to reinforced and eastward-displaced Aleutian Low and related anomalous southerlies (Figure AIV.8c). In contrast, concurrent cold conditions dominate far-eastern Siberia due to strengthened advection of Arctic air. The tail end of the storm track is strengthened leading to more precipitation along the Pacific coast of North America, especially in boreal winter. A positive PDV also induces a cooler and wetter climate in the southern part of the North American continent, with the wet anomalies being the most pronounced in boreal summer (Figure AIV.8d). The negative phase of the PDV is an important cause of droughts in the western and central USA (Dai, 2013). Besides the direct influence of atmospheric teleconnections from the tropical branch of the mode, extratropical SST anomalies have been shown to modulate the amplitude of the precipitation anomalies in North America (Burgman and Jang, 2015).

In the tropics, a positive PDV is associated with warm and dry anomalies in the Maritime Continent and Amazon (Meehl and Teng, 2014; Figure AIV.8c, d). The Indian and African monsoons also tend to weaken (Meehl and Hu, 2006; Figure AIV.8d). Warm and dry anomalies also cover most of Australia while wet anomalies dominate in South America south of 15°S. In particular, a positive PDV increases drought risk in Australia (Power et al., 1999; Vance et al., 2015). In the oceans, a positive PDV increases the occurrence of marine heatwaves in the Bering Sea and subtropical eastern Pacific, while a negative PDV increases ocean heatwaves in the Kuroshio–Oyashio extension, Melanesia and over the tropical Indian Ocean (Holbrook et al., 2019). PDV also modulates tropical cyclone activity: a positive PDV increases tropical cyclone genesis over the Philippine Sea and the tropical eastern North Pacific and decreases it over the tropical North Atlantic and western South Pacific (Grant and Walsh, 2001; Liu and Chan, 2008; Li et al., 2015; Zhao et al., 2018, 2020). PDV, together with ENSO, have been shown to control part of the interannual to multi-decadal wildfire variability, through PDV-related persisting droughts and severe heatwaves acting as preconditioning in many regions (see for instance, Kitzberger et al., 2007; Ward et al., 2016).

Implications of PDV on water cycle changes are further assessed in Sections 8.3.2.4 and 8.4.2.4. Section 10.4 highlights the importance of PDV in regional climate change attribution, with an example in Section 10.6. The influence of PDV climate anomalies in specific regions, including extremes, are assessed in Sections 11.3, 11.4, 11.6 and 11.7, and in Atlas.4, Atlas.5, Atlas.7 and Cross-Chapter Box Atlas.2. Its implications for ocean circulation and sea levels are assessed in Sections 9.2 and 9.6.

AIV.2.7 Atlantic Multi-decadal Variability

AIV.2.7.1 Definition of the Mode

The Atlantic Multi-decadal Variability (AMV) describes the large-scale slow fluctuations observed from one decade to the next in a variety of instrumental records and proxy reconstructions over the entire North Atlantic Ocean and surrounding continents. Fingerprints of the AMV can be found at the surface ocean, which is characterized by swings in basin-scale SST anomalies reflecting the interaction with the atmosphere, with a nominal period in the order of 70 years (Deser et al., 2010). AMV markers can be also tracked in the subsurface ocean in terms of heat content and density anomalies. The latter have been traditionally linked to the low-frequency fluctuations in the Atlantic Meridional Overturning Circulation (AMOC) and related oceanic meridional heat/salinity transport and water masses formation through deep convection in the sub-Arctic Seas (Zhang, 2017).

The positive phase of the AMV is characterized by anomalous warming over the entire North Atlantic, with the strongest amplitude in the subpolar gyre and along sea ice margin zones in the Labrador Sea and Greenland/Barents Sea (approximately +0.5°C) and in the
Figure AIV.9 | The Atlantic Multi-decadal Variability (AMV) based on the AMV index defined from Trenberth and Shea (2006) for 1900–2014 after 10-year low-pass filtering. The index corresponds to the annual sea surface temperature (SST) averaged over the North Atlantic domain (cyan box) minus the near-global average (60°N–60°S). (a) SST anomalies regressed onto the AMV index based on ERSSTv5. (b) The cyan curve indicates the 10-year low-pass filtered AMV index shown in bars. The orange curve stands for the AMV index used for the CMIP6 DCPP-C sensitivity experiments (Boer et al. 2016). (c) Same as (a) but for land surface air temperature (shading; based on Berkeley Earth for 1900–2014) and 10 m level wind (arrows; m s⁻¹ based on JRA-55 for 1958–2014) anomalies. (d) Same as (a) but for precipitation anomalies (shading based on GPCC for 1900–2014). All is based on annual mean. On maps, no overlay indicates regions where the regressions are significant based on t-test accounting for false detection rates at 10% and crosses indicates no significance.
subtropical North Atlantic basin to a lower extent (Figure AIV.9a). Concurrent, albeit weaker, anomalous cooling is found over the South Atlantic but also along the Pacific cold tongue sandwiched by warmer SST, a pattern reminiscent of the PDV (Section AIV.2.6). What stands out is the apparent inter-basin connectivity in AMV-related SST anomalies, which is strongly indicative of the existence of a network of teleconnections linking neighbouring ocean basins, the tropics and the extratropics, including Northern Hemisphere polar regions, and the oceans and land regions (Cassou et al., 2018). The AMV, together with the PDV, has been shown to modulate the observed time-evolving global surface temperature on multi-decadal time scales since pre-industrial time (Cross-Chapter Box 3.1), whilst the level of their respective influence remains difficult to quantify because of their inter-connectivity (Tung and Zhou, 2013; Wu et al., 2019; Li et al., 2020).

The term ‘Atlantic Multi-decadal Oscillation’ (AMO) has been introduced in the literature (Kerr, 2000) and used in previous IPCC reports to describe this phenomenon and in particular to make the distinction with the so-called North Atlantic tripole pattern connected to the NAO (Section AIV.2.1) at interannual time scales (Enfield et al., 2001). However, since the observed slow variations of the North Atlantic SSTs consist of a broad band of low-frequency signals (Zhang, 2017; Sutton et al., 2018) rather than a single preferred time scale, as also confirmed from last-millennium reconstructions (e.g., Singh et al., 2018), the denomination ‘Atlantic Multi-decadal Variability’ wherein a distinct oscillatory time scale is less clearly implied, appears more appropriate to describe decadale climate variability phenomena as also suggested for the Pacific (Section AIV.2.6). The term AMV will be accordingly used subsequently and throughout the entire report.

The observed AMV is interpreted as the regional residual imprint in SST anomalies after removal of the externally forced signal. There is no unique and preferred way to remove the impact of external forcing in observations. Linear detrending is used in the traditional definition of the AMV index, which corresponds to the average over the entire North Atlantic basin of the linearily detrended SST yearly anomalies on which a 10-year running mean is applied, as proposed by Enfield et al. (2001). To remove the residual nonlinear global-scale signal from the North Atlantic mean SST, different approaches have been proposed either based on observations applying specific statistics (e.g., Trenberth and Shea, 2006; Frajka-Williams et al., 2017; Frankignoul et al., 2017; Sutton et al., 2018; Yan et al., 2019) or based on model estimates of the externally forced signal including both natural (solar and volcanoes) and anthropogenic (greenhouse gases and aerosols) influence assessed from historical simulations (Ting et al., 2009; Terray, 2012; Steinman et al., 2015; Tandon and Kushner, 2015). More specifically, in Trenberth and Shea (2006), the annual observed global-mean SST anomaly time series is subtracted from the observed annual North Atlantic spatially averaged time series to obtain the raw unfiltered AMV index. In Ting et al. (2009), a signal-to-noise maximizing EOF analysis is applied to global annual mean SSTs derived from the CMIP5 multi-model ensemble to extract an estimation of the forced component, which is a priori removed before computation of the AMV index, then defined as a low-frequency residual. The latter technique has been retained in the so-called CMIP6 DCPP-C sensitivity experiments (Boer et al., 2016) aimed at furthering a better understanding of the teleconnections associated with the AMV. The two latter indices are shown in Figure AIV.9b and compared to the traditional AMV index.

All AMV time series consistently show a few distinct phases over the approximately 120-year instrumental record, namely warm periods over 1930–1965 and since 1995, and cool periods between 1900 and 1925 and over 1965–1995 (Figure AIV.9b). The uncertainty introduced by the different definitions of the index is mainly related to the strength of the phase of the AMV and in particular the most recent ones, which are the most sensitive to the estimation of the externally forced signal.

The temporal evolution of AMV in the instrumental records is assessed in Section 2.4.6. The transition between phases has been shown to involve various atmospheric and oceanic processes, leading to some predictability, but also influences from external forcing (both anthropogenic and natural), whose respective weight and interplay remain outstanding issues, as assessed in detail in Sections 3.7.7 and 9.2.3.

**AIV.2.7.2 Teleconnections and Regional Influence**

The important role of the AMV in observed low-frequency climate signals at a regional scale has been recognized for decades. The AMV is a key driver of temperature and precipitation anomalies along the continents surrounding the North Atlantic but also remotely through global atmospheric teleconnections (e.g., monsoons). Considering AMV-induced regional influence is crucial because the AMV acts as a pronounced modulator of the impacts driven by external natural and anthropogenic forcings. For instance, the AMV can have a pronounced fingerprint in time-integrating processes, such as river flows, and could explain the largest fraction of observed variance in some specific local areas since the 1900s (e.g., Bonnet et al., 2020). It is thus a key phenomenon for regional-scale or process-based attribution of past observed climate variations as introduced in Cross-Chapter Box 1.4 and extensively developed in Section 10.4.1.

Positive AMV is associated with much warmer conditions over the north-western corner of the Atlantic, with marked positive anomalies on land around the Labrador Sea (Figure AIV.9c). There is a strong seasonality in the AMV teleconnection and this pattern is mostly explained by boreal winter atmospheric dynamics and a local reduction of sea ice formation. There is a large-scale decrease of the winter mid-latitude westerlies and, concurrently, a weakening of the trade winds in the North Atlantic tropical basin, which is reminiscent of a negative phase of the NAO (Peings and Magnusdottir, 2014; Gastineau and Frankignoul, 2015). Over Europe, AMV-related signals are weak on yearly average (Figure AIV.9c) but mask a pronounced seasonality of the teleconnection as well as some interplay between dynamical (i.e., atmospheric circulation) versus thermodynamical changes controlled by the AMV (O’Reilly et al., 2017; Ruprich-Robert et al., 2017, Qasmi et al. 2020). Summer tends to be warmer during positive AMV in Europe (Sutton and Dong, 2012; Nicoli et al., 2020) with constructive contribution of both dynamical and thermodynamical responses while destructive interference prevails in boreal winter leading to colder conditions...
with a reinforced occurrence of blocking (Rimbu et al., 2014; Davini et al., 2015) and a southward shift of the jet (Simpson et al., 2018). Anomalies are positive all year round over a broad Mediterranean region, including North Africa and the Middle East. Reduction of Arctic sea ice is also found during positive AMV with local warming extension over the adjacent polar land (Day et al., 2012; Yu et al., 2017; Castruccio et al., 2018).

In terms of precipitation (Figure AIV.9d), positive AMV is associated with a northward shift of the Atlantic ITCZ in response to changes in the meridional SST gradient with respect to the equator (Martin and Thornicroft, 2014; Friedman et al., 2017; Green et al., 2017). The AMV influence is the strongest in the boreal summer affecting the West African Monsoon, leading to reinforced Sahel rainfall, and wetter conditions in the Caribbean basin through the modulation of the occurrence of hurricanes (Dunstone et al., 2011; Hermanson et al., 2014). By contrast, dry conditions prevail in north-easter Brazil and in the south-eastern corner of South America as well as in the Great Plains in North America. The signal of the positive AMV lead to intensified precipitation and river runoff over northern Eurasia (Nicoli et al., 2020) and controls the multi-decadal variations in river flows over much of Europe (Boé and Habets, 2014; Hodgkins et al., 2017). Because of global teleconnections, especially in the tropics, the AMV also affects the other monsoon systems with signals found over India, South East Asia and the Maritime Continent and, more broadly, over the entire tropical Pacific (Chylek et al., 2014; Kucharski et al., 2016; Ruprich-Robert et al., 2017).

The weight of the AMV in observed low-frequency variations is assessed in detail in Section 10.4.2 for specific regional case studies, including the sectors listed above. Because of its long time scale due to persistent heat content anomalies, the AMV can be considered as a potential source of near-term climate predictability. The skill of the current decadal forecast systems in predicting the AMV temporal evolution and associated teleconnection at various lead times is assessed in Section 4.4.3.6, while long-term changes are assessed in Section 4.5.3.6.

AIV.2.7.3 Available Proxy-based Reconstruction

Paleoclimate reconstructions of AMV-related signals have been developed using terrestrial, marine and ice-core proxies. The AMV reconstruction in the last 1200 years by J. Wang et al. (2017), using terrestrial proxy records, shows a clear broad band of enhanced spectral power at multi-decadal time scales but no evidence of a distinct multi-decadal or centennial spectral peak (Singh et al., 2018). This is largely consistent with the record of the instrumental period and with previous proxy-based studies (Gray et al., 2004; Mann et al., 2009) in the overlapping periods. Studies focusing on marine proxies also show persistent multi-decadal variability over the last three centuries (Kilbourne et al., 2014; Svendsen et al., 2014; Moore et al., 2017). Multi-decadal Arctic temperature variability related to the AMV is also evident in ice-core records over the last millennium (Chylek et al., 2011; Zhou et al., 2016) and beyond (Knudsen et al., 2011). High-resolution sedimentary records from the Arctic were recently analysed to reconstruct the AMV over the last 2900 years (Lapointe et al., 2020). However, this new reconstruction refers to the raw North Atlantic SST variability rather than to the residual of the Atlantic signal obtained after removal of the estimation of the external forcing influence. A multi-perspective assessment of the North Atlantic variability is thus necessary for understanding the origin of the AMV, its physics and its climatic implications over the paleoclimatic period.

The temporal evolution of the AMV is described and assessed in detail in Section 2.4.6.

AIV.2.8 Madden–Julian Oscillation

AIV.2.8.1 Definition of the Mode

The Madden–Julian Oscillation (MJO) is the leading mode of tropical intra-seasonal variability with 20–90 days’ time scale (Madden and Julian, 1994), and it is an important source of regional climate variability and predictability across the globe at sub-seasonal time scales from a week to months (Madden and Julian, 1994; Zhang, 2005; Vitart, 2017; Wheeler et al., 2017; Kim et al., 2018). It is characterized by planetary-scale disturbances (zonal wavenumber 1–3) of pressure, wind, clouds and rainfall moving predominantly eastward along the equator at average speed of 5 m s\(^{-1}\) (Roxy et al., 2019). More precisely, the MJO is characterized by alternating regions of suppressed and enhanced convection coupled to an anomalous zonal overturning circulation of the atmosphere (Zhang, 2005). The MJO is driven by internal ocean–atmosphere processes that occur all year round but it exhibits prominent seasonality with more pronounced signals in boreal winter. During boreal summer, centres of convective activity associated with the MJO are shifted away from the equator to 10°N–20°N and propagate northward in the Asian monsoon region (Yasunari, 1980; Madden, 1986) in addition to eastward propagation, often called as the boreal summer intra-seasonal oscillation (BSISO; Madden, 1986; Kikuchi et al., 2012). Whereas the MJO is considered present in all seasons, albeit with weaker amplitude in boreal summer, the BSISO tends to prevail during boreal summer, with influences on monsoons, tropical cyclones and other water cycle aspects (Kikuchi et al., 2012; Lee et al., 2013).

The real-time multivariate MJO (RMM) index has been widely used to detect the MJO, based on the first two combined EOF modes of daily outgoing longwave radiation (OLR) and zonal winds at 850 and 200 hPa averaged over 15°N–15°S (Wheeler and Hendon, 2004). The OLR represents convective activity and zonal winds at the lower and upper troposphere depict the overturning zonal circulation along the equator. The composite maps of convective activity and atmospheric circulation (Figure AIV.10a) and the phase diagram (Figure AIV.10b) based on eight phases constructed by the two PCs time series provide effective ways to understand and monitor spatial and temporal evolution of the MJO event. In general, the MJO disturbances are initiated in the equatorial western Indian Ocean and travel eastward mainly over the Indo-Pacific warm pool region. However dynamical fields in composite maps indicate that the MJO also propagates over climatological non-convective regions (e.g., the eastern Pacific, part of the tropical Atlantic; Figure AIV.10a), then manifesting the planetary nature of the tropical intra-seasonal oscillation (Cassou, 2008).
The Madden–Julian Oscillation (MJO)

a. The 8 phases of the MJO in DJFM

b. Observed daily MJO phases for 2007-2008 DJFM

Figure AIV.10 | (a) The spatio-temporal properties of the Madden–Julian Oscillation (MJO). Daily composites of outgoing longwave radiation (OLR, W m⁻², shaded) anomalies and stream-function anomalies at 300 hPa (STF, contours every 0.5 \times 10⁶ m² s⁻¹) for the eight MJO phases in boreal winter (1 December to 31 March) based on NOAA interpolated satellite products and NCEP–NCAR reanalysis over 1979–2019, respectively. Greenish (brownish) colours indicate enhanced (reduced) convective activity associated with wet (dry) conditions. Solid (dashed) contours show for positive (negative) STF anomalies which stand for anomalous anticyclonic circulation in the Northern (Southern) Hemisphere. (b) Phase diagram of the MJO constructed from the two leading principal components drawn from empirical orthogonal decomposition of combined wind at 300 hPa + OLR fields following Wheeler and Hendon (2004) for winter 2007–2008 chosen as an example of active MJO years. (c) Schematic representation of MJO-related tropical–Northern Hemisphere interactions and teleconnections inferred from observational and numerical studies (adapted from Stan et al., 2017). Because the interactions and teleconnections of the Southern Hemisphere are not as well understood except for the Southern Annular Mode, they are not included in the current schematic.
Each phase has a nominal persistence of about seven to eight days and the anti-clockwise rotation in this two-dimensional diagram materializes the propagation of the MJO packets from west to east. Figure AIV.10b provides an example of that RMM index computed for the year 1997, with months highlighted in colours. When the value of the index lies in the centre of the diagram, the MJO is considered to be inactive; outside the circle, its distance from the centre quantifies the amplitude and spatial coherence of the MJO episode.

The evaluation of the MJO in climate models is done through phenomenological (Waliser et al., 2009) and process-oriented (Xavier, 2012) diagnostics. MJO simulation remains a challenge in both CMIP3 and CMIP5 models, even though overall improvements have been reported in the most recent model versions (Waliser et al., 2003; Lin et al., 2006; Sperber and Annamalai, 2008; Kim et al., 2009; Xavier et al., 2010; Jiang et al., 2016). Poor MJO simulation in coupled climate models has been attributed to many factors, including an insufficient buildup of atmospheric moisture for large-scale organized convection to occur (Kim et al., 2012; Mizuta et al., 2012), biases in models’ mean state (Inness et al., 2003; Hannah et al., 2015; Kim, 2017), coupling frequency with the ocean and representation of air–sea interactions (Bernie et al., 2008; DeMott et al., 2015). In CMIP5 simulations, the eastward propagation of boreal winter MJO convection from the Indian Ocean into the western Pacific (Hung et al., 2013) and the northward propagation during boreal summer (Sperber et al., 2013) are slightly better represented, even if major biases persist and still affect the monsoon systems. These improvements have been attributed to advances in parametrizing moist turbulence and convection (Hohenegger and Bretherton, 2011; Mapes and Neale, 2011; Del Genio et al., 2012; Kim et al., 2012; Hourdin et al., 2013), leading to greater MJO predictive skill of up to four weeks in operational sub-seasonal prediction centres (Vitart, 2014). Super-parametrized models (Benedict and Randall, 2009), global cloud system-resolving models (Miyakawa et al., 2014) and high-resolution global climate models with an improved seasonal cycle (Mizuta et al., 2012) have been shown to produce a more realistic simulation of the MJO. Progresses in the representation of the MJO across model generations are assessed in Sections 1.5.4.6 and 8.3.2.9.1.

The assessment of changes in the MJO and related processes, mostly in terms of its implication for water cycle changes in the recent past and in future projections is provided in Sections 8.3.2.9 and 8.4.2.9, respectively. The MJO influence over South East Asia and Central and South America is further assessed in Atlas.4, Atlas.5 and Atlas.7, and Cross-Chapter Box Atlas.2.

AIV.2.8.2 Teleconnections and Regional Influence

In AR5, the MJO has been described as a tropical phenomenon, yet with prominent impacts across the globe through teleconnections. The MJO exerts a strong influence on occurrences of extreme weather events in the tropics, including tropical cyclones, droughts, flooding, heatwaves and fires, among others (Zhang, 2013), as well as those in the extratropics via teleconnections (Stan et al., 2017). This has opened new opportunities and scientific endeavours in the emerging field of sub-seasonal to seasonal forecasts of extremes (Vitart and Robertson, 2018) and for the development of new observational networks needed to initialize the forecast systems (Subramanian et al., 2019). The MJO can also impact the atmospheric composition, such as ozone, CO, CO$_2$ and aerosols (Tian et al., 2007, 2011; K.-F. Li et al., 2010; Lau et al., 2012).

The schematic in Figure AIV.10c shows the main regions of influence of the MJO. During boreal winter, the MJO fingerprint is mostly confined to the deep tropics with a zonal alternation of wet and dry zones depending on the MJO phase. For instance, in phase 4, enhanced rainfall is found over the Maritime Continent while suppression of convection is present over the central Pacific, northeast Brazil and equatorial Africa (Figure AIV.10a). The opposite is found during phase 8. Anomalous upper-level divergence in the deep tropics associated with altered convection and diabatic heating leads to anomalous vorticity in the subtropics (shown by stream-function dipoles straddling the equator in Figure AIV.10a) acting as an efficient and primary source of teleconnections outside the tropics (L’Heureux and Higgins, 2008; Lin et al., 2009; Lukens et al., 2017). The extratropical influence of the MJO may occur via the excitation of forced tropospheric Rossby waves (Ferranti et al., 1990; Matthews et al., 2004; Cassou, 2008; Stan et al., 2017) that affect the climate variability over the North Pacific (Ferranti et al., 1990; Higgins and Mo, 1997), the North Atlantic (Ferranti et al., 1990; Cassou, 2008) and the Arctic (Zhou and Miller, 2005). For example, tropical convective heating associated with MJO phase 5 contributes to Arctic warming, while the opposite occurs for MJO phase 1 (Yoo et al., 2011, 2012). Phase 2–3 of the MJO is clearly associated with the prevalence of the positive phase of the NAO with a 10- to 15-day lag after the MJO forcing. Stratospheric pathways are also reported to support tropical–extratropical teleconnections though the modified occurrence of sudden stratospheric warming (Liu et al., 2014; Garfinkel and Schwartz, 2017) with 15- to 20-day delay leading to sub-seasonal predictability.

During boreal summer, the MJO/BSISO convective activities and related impacts propagate northward/north-eastward over the Indian Ocean–South Asia region (Yasunari, 1980; Annamalai and Sperber, 2005), and northward/north-westward over the western North Pacific–East Asia region (Kembal-Cook and Wang, 2001; Lee et al., 2013). The MJO/BSISO significantly modulate occurrences of extreme rainfall and heatwaves over Asia (Hsu et al., 2016, 2017; Chen and Zhai, 2017; Diao et al., 2018) and North America (Moon et al., 2013; Lee and Grotjahn, 2019). They also regulate tropical cyclone activities in the Indian (Frank and Roundy, 2006), Pacific (Maloney and Hartmann, 2000; Diamond and Renwick, 2015; Zhao et al., 2015) and Atlantic (Klotzbach and Oliver, 2015) tropical oceans. They contribute to the intra-seasonal fluctuations of monsoons, including timing of onset and termination during summer (Maloney and Shaman, 2008; Wheeler et al., 2009; Lee et al., 2013; Grimm, 2019), and active/break phases (Pohl and Camberlin, 2006; Maloney and Shaman, 2008; Joseph et al., 2009; Lee et al., 2017).


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