

Decadal climate variability and predictability: challenges and opportunities

Article

Accepted Version

Cassou, C., Kushnir, Y., Hawkins, E. ORCID: <https://orcid.org/0000-0001-9477-3677>, Pirani, A., Kucharski, F., Kang, I.-S. and Caltabiano, N. (2018) Decadal climate variability and predictability: challenges and opportunities. *Bulletin of the American Meteorological Society*, 99 (3). pp. 479-490. ISSN 1520-0477 doi: 10.1175/BAMS-D-16-0286.1 Available at <https://centaur.reading.ac.uk/73146/>

It is advisable to refer to the publisher's version if you intend to cite from the work. See [Guidance on citing](#).

To link to this article DOI: <http://dx.doi.org/10.1175/BAMS-D-16-0286.1>

Publisher: American Meteorological Society

All outputs in CentAUR are protected by Intellectual Property Rights law, including copyright law. Copyright and IPR is retained by the creators or other copyright holders. Terms and conditions for use of this material are defined in the [End User Agreement](#).

www.reading.ac.uk/centaur

CentAUR

Central Archive at the University of Reading

Reading's research outputs online



AMERICAN METEOROLOGICAL SOCIETY

Bulletin of the American Meteorological Society

EARLY ONLINE RELEASE

This is a preliminary PDF of the author-produced manuscript that has been peer-reviewed and accepted for publication. Since it is being posted so soon after acceptance, it has not yet been copyedited, formatted, or processed by AMS Publications. This preliminary version of the manuscript may be downloaded, distributed, and cited, but please be aware that there will be visual differences and possibly some content differences between this version and the final published version.

The DOI for this manuscript is doi: 10.1175/BAMS-D-16-0286.1

The final published version of this manuscript will replace the preliminary version at the above DOI once it is available.

If you would like to cite this EOR in a separate work, please use the following full citation:

Cassou, C., Y. Kushnir, E. Hawkins, A. Pirani, F. Kucharski, I. Kang, and N. Caltabiano, 2017: Decadal Climate Variability and Predictability: Challenges and opportunities. *Bull. Amer. Meteor. Soc.* doi:10.1175/BAMS-D-16-0286.1, in press.

© 2017 American Meteorological Society



Decadal Climate Variability and Predictability:

Challenges and opportunities

Christophe Cassou (1), Yochanan Kushnir (2), Ed Hawkins (3), Anna Pirani (4,5),
Fred Kucharski (5), In-Sik Kang (6) and Nico Caltabiano (7)

Manuscript submitted to BAMS InBox section

1st revision (July, 5th 2017)

2nd revision (September, 24th 2017)

(1) CECI, CNRS-Cerfacs, Université de Toulouse, Toulouse, France (cassou@cerfacs.fr).

(2) Lamont Doherty Earth Observatory, Palisades NY, USA (kushnir@ldeo.columbia.edu).

(3) National Centre for Atmospheric Science, Department of Meteorology, University of
Reading, Reading, United Kingdom (e.hawkins@reading.ac.uk)

(4) Université Paris-Saclay, Saint Aubin, France (anna.pirani@universite-paris-saclay.fr)

(5) The Abdus Salam International Centre for Theoretical Physics, Trieste, Italy
(kucharsk@ictp.trieste.it)

(6) Center of Excellence for Climate Change Research, Department of Meteorology, King
Abdulaziz University, Jeddah, Saudi Arabia (kang@climate.snu.ac.kr)

(7) International CLIVAR global Project Office, Southampton, United Kingdom
(nico.caltabiano@clivar.org)

Corresponding author: Christophe Cassou, cassou@cerfacs.fr

25 **Decadal phenomena and their characteristics**

26 The slowdown in the rate of global surface warming in the early 2000s, and
27 especially its regional characteristics, highlights the importance of “decadal climate
28 variability” (DCV) as modulator of long-term warming trends due to ever-increasing
29 anthropogenic forcings (Medhaug et al. 2017). This event, which was termed in the
30 scientific and public domain as a “pause” or “hiatus” in global warming
31 (Lewandowsky et al. 2016), was argued by scientists to be associated with long-
32 recognized (see for instance the 2nd IPCC report published in 1996 for an early
33 assessment), multi-year phenomena and in particular the undulation of the tropical
34 Pacific ocean-atmosphere system (Kosaka and Xie 2013; Meehl et al. 2016).
35 According to several studies, changes in Earth energy balance at the top of the
36 atmosphere partly related to volcanic aerosols from a series of moderate eruptions in
37 the early 2000s also contributed to the recent ‘hiatus’ (Huber and Knutti 2014; Santer
38 et al. 2017). Yet, because of uncertainties in observational estimates in both radiative
39 forcing and global temperature measures, it is impossible to stringently attribute the
40 early 2000s ‘hiatus’ to a specific origin (Hedemann et al. 2017); it should rather be
41 interpreted as a combination of several factors (Medhaug et al. 2017).

42 Whether in cases of external forcing due to natural (solar & volcanic) or
43 anthropogenic factors, or during internal climate-system interactions, the oceans play
44 a central role in DCV due to their thermal and dynamical inertia. Decadal variations of
45 both regional and global mean surface temperature can be associated with, and
46 often attributed to, changes in ocean heat uptake and heat redistribution (Yan et al.
47 2016). The study of the ocean's role in climate has traditionally involved the
48 diagnosis of sea surface temperature (SST) variability as the variable reflecting the
49 interaction with the atmosphere. While the major oceanic basins are usually

50 examined separately when identifying the main climate variability phenomena, what
51 stands out is the related nature of the global patterns and some similarities in their
52 temporal evolution. Specifically, large-scale, long-term SST variability in one ocean
53 basin is associated with variability in other basins. The global view of these patterns
54 reveals a "network of teleconnections", linking neighboring ocean basins, the tropics
55 and extratropics, and the oceans and land regions (Fig 1). The two most prominent
56 patterns in this respect are associated with the Atlantic Multidecadal Oscillation
57 (AMO, Fig.1a, b) and the Pacific Decadal Oscillation (PDO, Fig.1c, d) or its nearly
58 interchangeable companion referred to as Interdecadal Pacific Oscillation (IPO, Han
59 et al. 2014, Dong & Dai 2015)¹, although none of these are true 'oscillations'.

60 The apparent inter-basin connectivity seen in DCV SST footprints is likely
61 indicative of the global nature of the associated atmospheric mechanisms that either
62 force, or respond to, the oceanic variability. However, it is challenging to clearly
63 elucidate the origins of DCV phenomena and especially the underlying oceanic
64 mechanisms because of the short observational records and sparse spatial sampling,
65 and because of consistency in terms of instrumental biases. This is even more a
66 limiting factor in the Southern Ocean where low-frequency variability seems to be
67 present (Fan et al. 2014) but cannot be robustly identified because of data issues.
68 Nonetheless it is now well recognized that a superposition of multiple processes
69 underlies both the Atlantic and Pacific DCV (Yeager & Robson 2017; Newman et al.
70 2016, respectively).

71 In the Atlantic, coupled model integrations (e.g. Medhaug & Fluverik 2011,
72 Ruprich-Robert & Cassou 2015 and references therein), including decadal hindcasts
73 (Robson et al. 2012, Msadek et al. 2014) and forced-ocean model simulations

¹ In the rest of the paper, considering the difficulty to distinguish between PDO and IPO, we preferably use the latter term.

74 (Danabasoglu et al. 2016), demonstrate that the AMO is linked to the variability of the
75 Atlantic Meridional Overturning Circulation (AMOC). Through its associated
76 fluctuations of salt and heat transport and through deep convection-related
77 mechanisms in the North Atlantic subarctic seas, the AMOC is considered as a driver
78 of the upper ocean heat content variability and related SST anomalies that project
79 upon the AMO (Buckley and Marshall, 2016). Fingerprints of AMOC variation were
80 found in the subpolar gyre in the form of SST, salinity and, hence, density anomalies
81 (Hakkinen et al. 2011), particularly in deep-water formation regions. Related changes
82 in the state of the ocean are found in the tropical Atlantic, expressed as anti-
83 correlated variations of surface and subsurface ocean temperatures (Zhang et al.
84 2013), a relationship which is supported by paleoclimate proxy evidence (Parker et
85 al. 2015), although the connection between AMOC variations and tropical Atlantic
86 SSTs remains ambiguous and the mechanism producing tropical Atlantic SST
87 decadal variability remains elusive.

88 In addition to such ocean dynamical processes, decadal North Atlantic SST
89 evolution since the early 20th century has also been attributed to external radiative
90 forcing by either anthropogenic (Booth et al. 2012, but see Zhang et al. 2013 for
91 further discussion) or natural aerosols. Results combining paleoclimate
92 reconstructions, simulations and observations, have pointed to decadal climatic
93 impacts of volcanic eruptions and to their possible role in the modulation or phase-
94 locking of the AMOC/AMO pattern (Ottera et al. 2010, Swingedouw et al. 2015).

95 In the Pacific domain, evidence suggests that the PDO should be viewed as
96 the superposition of several phenomena, each governed by different physical
97 processes, each of which is manifested in SST variability in a different part of the
98 basin, rather than being a coherent, single, physical mode. Possible physical drivers

99 include the integration of extratropical atmospheric weather noise by the extratropical
100 ocean basins, teleconnections through Rossby wave propagation from the decadal
101 tropical footprint of ENSO, re-emergence of persistent subsurface heat content
102 anomalies, and variations in wind stress driven gyre circulation (Newman et al.
103 2016). The subtle distinctions between the PDO and IPO may be linked with the
104 multiplicity of governing processes.

105 According to several studies, the IPO underlies part of the slowdown in the
106 rate of global warming during the early 2000s (see Xie & Kosaka 2017 for a review)
107 and could also be considered as the primary driver of past-observed warming
108 slowdowns and accelerations over the historical period (Fyfe et al. 2016, Kosaka &
109 Xie 2016). Model control simulations (with constant external forcing) exhibit IPO-like
110 variability, thus pointing at an internal, i.e. unforced origin of this phenomenon.
111 However, similarly to AMO, it was suggested that aerosol forcing played a role in
112 determining the temporal phase of the IPO in the late 1990's (Smith et al., 2016, but
113 see Kuntz and Schlag 2017 for further discussion). In any case, determining the
114 cause for the particular IPO event that led to the slowdown in warming in the early
115 2000s, still faces lingering uncertainties (Hedemann et al. 2017).

116 The climate literature is rich with studies that have linked the AMO and IPO to
117 past observed worldwide climate variations over land. For example the AMO has
118 been related to the multi-decadal fluctuation in Atlantic tropical cyclone activity (Wang
119 et al. 2012), and the variations of rainfall in the Sahel, which led to the devastating
120 droughts and famines of the 1970s & 1980s. Similarly, the multi-year pulses of
121 droughts in the US and Mexico (e.g., the “dust bowl” in the 1930s and the most
122 recent protracted dry period in California over 2007-2016) have been associated with
123 the combined effect of the AMO and IPO (Schubert et al. 2004, Chylek et al. 2014),

124 while the so-called Millennium drought and warming in Australia has been attributed
125 to a combination of DCV phenomena in the Indo-Pacific and Southern Ocean regions
126 (van Dijk et al. 2013). Controlled experiments in which global climate models are
127 forced to follow AMO and IPO temporal evolution, confirm that many of the past
128 observed changes in the climate could be partly attributable to these oceanic
129 anomalies (Zhang & Delworth 2006; Kosaka & Xie 2013, Ruprich-Robert et al. 2017).
130 This motivates the ongoing development of "initialized decadal prediction", in which
131 coupled models are initialized with the observed state of the ocean while
132 anthropogenic and natural forcings are prescribed. Models are integrated for multiple
133 years to try to predict the combined impact of radiative forcing and natural climate
134 variability since the 1960s (Kirtman et al. 2013, among others).

135

136 **Challenges and outstanding questions**

137 In the context of adaptation to (and mitigation of) the impact of climate change,
138 it is important for a broad range of stakeholders and decision makers to know how
139 the response to anthropogenic forcing and the impact of natural DCV phenomena will
140 mix together to shape the near term evolution of climate, especially at regional
141 scales. A major objective of CLIVAR's study of DCV and its predictability, the
142 research focus on Decadal Climate Variability and Predictability (DCVP²), is to
143 advance the dynamical understanding of the associated phenomena and improving
144 their representation in models. This is an essential component of the World Climate
145 Research Programme (WCRP) plan to facilitate a robust and reliable provision of
146 decadal information to society under its Grand Challenge on Near Term Climate

² <http://www.clivar.org/research-foci/dcvp>

147 Prediction³. Decadal forecasts were part of the 5th phase of the Coupled Model
148 Intercomparison Project (CMIP5, Taylor et al. 2012) and the Decadal Climate
149 Prediction Project (DCPP⁴) ensures the coordination and design of the modeling
150 experiments in the ongoing 6th edition (Boer et al. 2016).

151 In the Atlantic, initialized near-term prediction systems demonstrate substantial
152 skill in hindcasting decadal AMO-like SST anomalies (Oldenborgh et al., 2012; Kim et
153 al., 2012, Marotzke et al. 2016, among others) but noticeably, the added-value of the
154 initialization in temperature and precipitation over adjacent land is lost very rapidly
155 (beyond a couple of years), especially over Europe and the American continents
156 (Doblas-Reyes et al., 2013; Meehl et al., 2014). More robust skill seems to be found
157 when statistical methods to overcome sampling issues are applied (e.g. Robson et al.
158 2013) or when investigations are focused on case studies (such as the 1960s or
159 1990s AMO phase transitions). This is perplexing in light of the findings that coupled
160 models with restored SST to observed anomalies over the Atlantic are able to display
161 part of the ocean's influence on land climate. Part of the loss of predictability over
162 land may be related to systematic errors in the simulation of the spatial structure and
163 temporal properties of the AMO in free-coupled models and/or the weak associated
164 teleconnections that communicate ocean signals to the continents (Eade et al.2014,
165 Qasmi et al. 2017). Skill in a given prediction system might be also dependent on the
166 model representation of processes or chain of events leading to AMO phenomena. In
167 particular, the AMOC-AMO relationship varies considerably from model to model and
168 leads to a wide range of amplitudes, spatial properties and preferred timescales of
169 the simulated AMO and associated teleconnections (Ruiz-Barradas et al. 2013,
170 Zhang and Wang 2013, Ba et al., 2014; Peings et al., 2016). Moreover, recent

³ <https://www.wcrp-climate.org/grand-challenges/gc-near-term-climate-prediction>

⁴ <https://www.wcrp-climate.org/dcpp-overview>

171 studies also showed that a pattern akin to the AMO SST expression could be
172 identified in a model in which the atmosphere is coupled to a motionless, slab-ocean
173 model, calling into question the active role of the ocean dynamics in the phenomenon
174 that could be consequently attributed to random atmospheric forcing (Clement et al.
175 2015, Cane et al. 2017 but also Zhang et al. 2016 and O'Reilly et al. 2016 for further
176 discussion of this controversy). In that context, it is notable that the frequency-
177 dependence of the relationship between the AMV and the North Atlantic Oscillation
178 (NAO), which displayed substantial variability on multidecadal time scale during the
179 20th century, seems to be a key aspect for DCV mechanisms (Delworth et al. 2017).
180 Altogether, the role of atmosphere and ocean thermodynamics and dynamics in the
181 surface expression of the AMO and its remote impacts over land remains an active
182 area of research. Resolving the reasons for the very large diversity in model
183 simulations of the AMO is an outstanding challenge, especially the nature and
184 respective weight of the physical mechanisms that contribute to the development of
185 DCV in the Atlantic.

186 In the Pacific, even if such conclusions are still topics of contention and
187 discussion, initialized coupled models have shown some success in simulating the
188 recent IPO cool phase in the late 1990s (Ding et al. 2013, Meehl et al. 2014); yet,
189 overall performance and results remain inferior to the Atlantic case (Doblas-Reyes et
190 al., 2013). Here too, more research is needed to elucidate the role of the relevant
191 processes and mechanisms that govern the IPO. Fluctuations in tropical Pacific trade
192 winds have been suggested to be a key factor (England et al. 2014) but models fail in
193 reproducing the observed strengthening over the last 30 years (Kociuba and Power
194 2015). In non-initialized simulations, the basin-wide spatial characteristics of the IPO
195 are reasonably well represented (Henley et al. 2017), but the overall ratio of decadal-

196 to-total variance for SST and oceanic fields is underestimated in most of the models
197 and errors are pronounced in the extratropics (Farneti, 2016; Nidheesh et al. 2017).
198 Discrepancies could be explained by the (in)ability of the coupled models to
199 effectively simulate all processes generating the IPO (Newman et al. 2016) and/or by
200 their respective ability to represent the different processes that govern the
201 phenomenon (e.g., the models ability to simulate oceanic re-emergence, Kuroshiro
202 dynamics, stochastic midlatitude atmospheric forcing associated with the variability of
203 the Aleutian Low, etc.).

204 For both Atlantic and Pacific DCV, part of the inter-model divergence can be
205 explained by model biases of both mean state and intrinsic variability. For instance,
206 the formation of Labrador Sea Water, that is critical for AMOC variations leading to
207 DCV, is either controlled by salinity or temperature variations as a function of
208 systematic errors in the model's mean state (Menary et al. 2015); this is expected to
209 influence the spatio-temporal characteristics of the simulated AMO. For the IPO, the
210 failure in reproducing the pan-Pacific imprint of the decadal mode in some models is
211 related to their intrinsic deficiencies in simulating interannual teleconnections linked
212 to ENSO (Nidheesh et al. 2017).

213 Yet, evaluating the authenticity of model simulated DCV is challenging due to
214 sparse and short observational records. Based on long control experiments, Qasmi
215 et al. (2017) provides evidence for non-stationarity, in a statistical sense, of the
216 intrinsic spatio-temporal properties of the AMO (variance, persistence, etc.) and
217 associated teleconnections in all CMIP5 models. The identification of the processes
218 that have led to observed DCV since the end of the 19th century is further
219 complicated by the presence of external forcing, both natural and anthropogenic.
220 Uncertainties and controversies remain concerning the role of anthropogenic

221 aerosols for both the AMO and IPO transitions. This remains an outstanding issue
222 because of the interplay between the response of SST patterns to aerosols and
223 GHGs (Xie et al. 2013) and because external forcing can interfere with internal DCV
224 phenomena (Ottera et al. 2010, Tandon and Kushner 2015). In addition, there is no
225 unique and perfect way to remove the impact of external forcing in the observations
226 and the chosen method may have significant implications for the outcome and hence
227 the interpretations of DCV expression during the historical period (Tandon and
228 Kushner 2015, Lyu and Yu, 2017).

229 Finally, the presence of drifts and initial shocks in decadal hindcasts is a
230 hindrance for prediction and understanding of recent DCV events. Several drift-
231 adjustment procedures, which are mandatory prior to evaluation, exist but are not yet
232 satisfactory in an operational context (Hawkins et al. 2014). Difficulties are mostly
233 related to the temporal non-stationarity of the model drifts due to errors in initial
234 conditions (Fučkar et al. 2014, Sanchez-Gomez et al. 2015, Pohlmann et al. 2016,
235 among others) and interactions with the model representation and responses to
236 external forcings (Kharin et al. 2012).

237 Addressing all these challenges and distinguishing the processes contributing
238 to DCV is all the more important since both the latest observations and initialized
239 decadal forecasts are suggesting that the AMO is undergoing a transition from a
240 warm to a cold phase (Hermanson et al. 2014, Robson et al. 2016) and that the IPO
241 is trending toward its warm phase (Thoma et al. 2015, Meehl et al. 2016). Over the
242 Arctic, Yeager et al. (2015) also reported a prediction for a near-term decadal
243 transition, signaling that the decline rate of sea-ice due to GHG forcing is expected to
244 slow down due to a weakened AMOC in the 2010s associated with the transition of
245 the AMO to its cool phase. These predictions will be put to test in the next few years

246 (e.g. Smith et al. 2013) and there is urgency to assess their reliability based on a
247 process-based evaluation.

248

249 **Framework for progress and recommendations**

250 All the above-listed obstacles and challenges test our ability to *attribute* past
251 variations to the combined role of internal variability and external forcing, as well as
252 *predict* the near-term future climate evolution on a global and regional scale.

253 Research is needed to deepen our understanding of the physical mechanisms
254 contributing to DCV phenomena and to make progress, we propose to address DCV
255 issues in terms of teleconnections based on the intrinsic nature of the observed SST
256 variability (Fig.1). More specifically:

- 257 1. *tropical-extratropical teleconnections*. Studies are needed to better understand
258 the oceanic basin scale properties of the DCV patterns and, in particular, the
259 respective origins and roles of their tropical versus extratropical components.
260 Over the Atlantic, a focus is recommended on the processes that facilitate the
261 link between decadal subpolar gyre and tropical Atlantic SST variations -- the
262 latter being the primary bridge to global climate variability (Chikamoto et al.
263 2016, Ruprich-Robert et al. 2017). The interaction and possible feedbacks
264 between the AMOC and AMO variations and the NAO and other atmospheric
265 circulation anomalies (such as blocking events), is a key question in this
266 respect as these are also very important for driving basin-wide Atlantic Ocean
267 variability (Barrier et al. 2014). In addition, uncertainties remain regarding the
268 influence of the external forcing on the North Atlantic modes of variability. The
269 attribution of the late 20th century observed trend of the NAO to external
270 forcing (Gillett et al. 2003) has been disputed in some recent studies (e.g.

271 Kelly et al. 2012, Cattiaux and Cassou 2013, Deser et al. 2016) and a better
272 evaluation and understanding of the weight of external forcing in the recent
273 decadal changes of the North Atlantic climate is still a research priority. Over
274 the Pacific, it is essential to resolve and understand the exchange of heat
275 between the extra-tropical and tropical Pacific in governing the IPO as well as
276 the role of midlatitude atmospheric stochastic forcings (Drijfhout et al. 2014;
277 Newman et al. 2016).

278 2. *inter-basin teleconnections*. Motivated by the temporal quasi-synchronization
279 of the main modes of DCV (Fig.1), research is needed to better characterize
280 and understand the interconnectivity of the different oceanic basins as well as
281 with the polar regions, and to evaluate the robustness and stationarity of these
282 relationships. The role of the AMV in the phase shift of the IPV in the late
283 1990s (Kucharski et al. 2016) and more broadly in the modulation of ENSO
284 variance (Dong et al. 2006), the role of the IPO/AMO in the cooling of the
285 Southern Ocean and increased sea-ice cover over the last 30 years, the
286 atmospheric bridge between tropical oceanic basins through Walker cell
287 recent reinforcement, among others, are key questions to elevate our
288 understanding of decadal variability and predictability, especially if robust
289 lagged relationships between modes are identified.

290 3. *ocean-land teleconnections*. Motivated by the strong relationship at regional
291 scales between observed oceanic DCV modes and land anomalies, emphasis
292 should be placed on improved understanding of the ocean-land connection at
293 low-frequency, especially as CMIP5 models tend to underestimate the overall
294 relationship (Qasmi et al. 2017 for the AMO) and decadal hindcast skill is
295 critically reduced over the continents. The spatial imprint of the observed early

296 2000s 'hiatus' (strong cooling over North America and Eurasia during
297 wintertime as an important contributor to the slower rise in annual global
298 temperature) as well as recent modeling studies highlight the importance of
299 investigating these issues on a seasonal basis and in terms of changes in the
300 entire probability density function of key parameters such as surface air
301 temperature and precipitation (surmising that the ocean-land connection could
302 take place through changes in the probability of extremes).

303 4. *vertical teleconnections*. Motivated by (i) the importance of surface-subsurface
304 relationship in the ocean as key factor for DCV genesis, (ii) the huge
305 discrepancies between oceanic reanalysis products on the three dimensional
306 thermohaline properties of the water masses (Karspeck et al. 2015), (iii) the
307 recent evidence for the important role of the troposphere-stratosphere
308 coupling in the atmosphere in response to natural decadal forcing (Dunstone
309 et al. 2016), process-oriented approaches and emerging high resolution model
310 configurations (stratosphere-resolving atmosphere and refined vertical levels
311 in the ocean component to better simulate vertical processes), should be
312 promoted to tackle this issue. Resolving and understanding the exchange of
313 heat between the upper and deep ocean in governing the AMO and the IPO is
314 essential for progress.

315 Across all the four teleconnections, it is essential to assess the respective roles of
316 internal variability and external forcing in DCV at *regional* scale, i.e., characterize and
317 understand the observed and simulated spatial responses to external forcing agents
318 and to how these forcing agents affect circulation change in the ocean and the
319 atmosphere and related teleconnections, in presence of internal variability.

320 Within this "*drivers of teleconnectivity*" framework, recommendations are to:

321 (1) *use a hierarchy of models and dedicated modeling protocols to study DCV*
322 *mechanisms and improve their simulations.* Both coupled historical simulations and
323 initialized hindcast experiments, together with targeted model sensitivity experiments
324 (e.g. pacemaker or partial coupling configurations, etc.), are crucial for mechanistic
325 and process understanding of DCV that is necessary to ultimately advance decadal
326 climate prediction skill and capacities, especially at regional scale. Empirical
327 approaches to make decadal predictions and utilize observed teleconnections
328 between ocean and land should be also encouraged (Lean & Rind 2009; Suckling et
329 al. 2017).

330 (2) *reduce model biases and drifts.* Presence of systematic errors in models
331 has been shown to hamper the representation of DCV patterns and variability. In
332 forecast mode, drift and shock when models are initialized from observed conditions,
333 correspond to a sequence of physical processes by which models adjust towards
334 their own biased equilibria. The determination of the physical origin of model drift and
335 systematic errors should be promoted since such a framework provides clues for the
336 mechanisms underlying DCV in models (Toniazzo & Woolnough 2014; Sanchez-
337 Gomez et al. 2015) and is therefore a promising pathway toward improved simulation
338 and related prediction capability/skill of DCV.

339 (3) *improve the knowledge and observational capacity necessary to track the*
340 *energy flows through the climate system.* This is critical for a better understanding of
341 the relationships/interactions between external climate forcing and internal DCV in
342 affecting past and near-term future changes. Accurate measurement and physical
343 interpretation of three-dimensional ocean heat content (heat uptake and heat
344 redistribution) are crucial to make progress in DCV (Trenberth et al, 2016). This
345 should be combined with the precise estimation and analysis of the planetary energy

346 budget at the top of the atmosphere and at the surface (Allan et al. 2014, Smith et al.
347 2015), from global to regional scale.

348 (4) *combine the analysis of instrumental observations and models together*
349 *with the information emerging from high-resolution paleoclimate proxy data.* Evidence
350 of coherent spatial and temporal patterns of DCV emerge from tree ring atlases, coral
351 and mollusc records and other paleoclimatic archives (e.g Linsley et al., 2015,
352 Tierney et al., 2015, Emile-Geay et al., 2017, Reynolds et al., 2017). These can shed
353 light on underlying physical processes and provide validation to patterns emerging
354 from the short observational record. Instrumental data rescue efforts should also be
355 expanded to recover the wealth of un-digitized records, filling gaps and increasing
356 overlap with the paleoclimatic archives (e.g. the ACRE and OldWeather.org
357 initiatives, Allan et al., 2011).

358 (5) *revisit the traditional definition of DCV phenomena.* Care is required when
359 representing and interpreting the climate system with single univariate and basin-
360 wide indices, as is too often done in the case of the AMO and IPO indices (see
361 examples in Yeager and Robson 2017's review). It is recommended that the
362 community adopts (i) the use of multivariate metrics ('fingerprints') to identify and
363 discriminate between processes involved, especially when comparing model
364 simulations to observations and (ii) the notation of decadal variability acronyms (e.g.
365 AMV and IPV) to think in terms of broad spectrum variability instead of temporal
366 oscillations with preferred time-scales.

367 The "drivers of teleconnectivity" theme provides a way to approach the goal of
368 advancing the study of DCV and predictability from different research directions. It
369 goes beyond a preoccupation with changes in the global mean temperature
370 "popularized" by the early 2000s 'hiatus', a shorthand term for a very complex

371 phenomenon, and directly promotes regional analyses in line with societal needs for
372 local climate information. It fosters an approach to address the intrinsic trans-
373 disciplinary issues that challenge progress in DCV and predictability and encourage
374 strengthened links across international community programs such as the CLIVAR
375 DCVP and CONCEPT-Heat⁵ research foci, PAGES2K⁶, WCRP GEWEX⁷, DCP, P,
376 VolMIP⁸ and the WCRP Near-Term Prediction Grand Challenge. Because of the
377 difficulties to observe and model the Earth's climate at timescales of a decade or
378 longer, this area of research is wholly dependent on emerging connections between
379 those who perform, collect and analyze instrumental observations of the present,
380 those who develop and analyze proxies of past climate, and with scientists who
381 develop models and perform dedicated modeling experiments. This “drivers of
382 teleconnectivity” framework has been discussed and adopted at the end of the DCVP
383 Research-Focus workshop organized by CLIVAR and the International Centre for
384 Theoretical Physics in Trieste (Italy) in November 2015⁹.

385 More specifically, the continued PAGES2K effort towards collecting, analyzing
386 and interpreting high-resolution climate records, including the development of spatio-
387 temporal reconstructions of climate (temperature and precipitation) variability both in
388 the tropics and at midlatitudes, especially for the last two millennia, should help better
389 understand tropical-extratropical and inter-basin teleconnections at decadal
390 timescales. Progress on the interpretation of ocean-land connectivity would gain from

⁵ <http://www.clivar.org/research-foci/heat-budget>

⁶ <http://pastglobalchanges.org/ini/wg/2k-network/>

⁷ <http://www.gewex.org/>

⁸ <http://volmip.org/index.html>

⁹ The Workshop was organized towards meeting the objectives of the CLIVAR Research Focus on DCVP to advance the understanding of the mechanisms and predictability of decadal climate variability. The organizers integrated the plans to establish a new WCRP Grand Challenge on “Near Term Climate Prediction” and the development of coordinated modeling strategies to address decadal variability and prediction under CMIP6 led by DCP. The Workshop was attended by around 150 participants and was streamed live to a global audience.

391 studying the relationship between land (e.g. tree rings) and ocean proxies (corals,
392 shells etc.).

393 On the modeling side, coordinated experiments based on “pacemaker” or
394 partial coupling modeling strategies have been designed to directly address
395 teleconnectivity (DCPP Component C; Boer et al., 2016) and thus perfectly fit within
396 the proposed framework. The emergence of high-resolution models now compatible
397 with DCV studies, together with increasing computing resources, is a promising
398 opportunity for improved model fidelity and better simulation of the four above-listed
399 teleconnections. Recent work has highlighted (i) the role of meso-scale oceanic
400 fronts in driving the atmosphere and (ii) the role of the stratosphere and storm-tracks
401 as transmitters of SST variability towards the continents (Kidston et al. 2015, Ma et
402 al. 2016). To make progress, increasing computing resources should be also
403 prioritized to promote the production of large ensembles to correctly characterize
404 internal variability; this is essential to accurately attribute observed DCV events, their
405 nature and physical origins and interaction with external forcing.

406

407 **Acknowledgements:** We would like to acknowledge the following organizations for
408 sponsorship and help in enabling and in organizing the Trieste Workshop at the
409 origin of this paper: The International CLIVAR Project Office (ICPO); the Past Global
410 Changes (PAGES) project; and U.S. CLIVAR, and the National Oceanic and
411 Atmospheric Administration (NOAA). Most sincere thanks go to the The Abdus
412 Salam International Centre for Theoretical Physics (ICTP) for their generous support
413 in hosting the Workshop and providing an excellent environment for the
414 presentations and discussions. The ICTP organized a hands-on training session on
415 decadal climate variability and predictability. Students were sponsored to attend both

416 the workshop and training. We are grateful to the organizing committee and to all the
417 members of the CLIVAR DCVP research focus for very fruitful discussions. Access to
418 the Workshop program and lectures (slides and videos) are available at:
419 <http://indico.ictp.it/event/a14266/other-view?view=ictp timetable>. The authors are very
420 thankful to the four anonymous reviewers who helped considerably improve the
421 original manuscript. CC is supported by CNRS and by the MORDICUS grant under
422 contract ANR-13-SENV-0002-01.
423

424 **References:**

425 Allan, R., P. Brohan, G. P. Compo, R. Stone, J. Luterbacher, and S. Bronnimann,
426 2011: The international atmospheric circulation reconstructions over the earth
427 (ACRE) initiative. *Bull. Amer. Meteor. Soc.*, **92**, 1421, doi:10.1175/2011BAMS3218.1.

428

429 Allan, R. P., C. Liu, N. G. Loeb, M. D. Palmer, M. Roberts, D. Smith, and P.-L. Vidale,
430 2014: Changes in global net radiative imbalance 1985–2012, *Geophys. Res. Lett.*,
431 **41**, 5588–5597, doi:10.1002/2014GL060962.

432

433 Ba, J. et al., 2014: A multi-model comparison of Atlantic multidecadal variability, *Clim.*
434 *Dyn.*, **43**, 2333-2348, doi:10.1007/s00382-014-2056-1.

435

436 Barrier, N., C. Cassou, J. Deshayes, and A.-M. Treguier, 2014: Response of North
437 Atlantic Ocean circulation to atmospheric weather regimes. *J. Phys. Oceanogr.*, **44**,
438 179–201, doi:10.1175/JPO-D-12-0217.1.

439

440 Boer, G. and co-authors, 2016: The Decadal Climate Prediction Project (DCPP)
441 contribution to CMIP6. *Geosci. Model Dev.*, **9**, 3751-3777, doi:10.5194/gmd-2016-78.

442

443 Booth, B. B. B., N. J. Dunstone, P. R. Halloran, T. Andrews, and N. Bellouin, 2012:
444 Aerosols implicated as a prime driver of twentieth-century North Atlantic climate
445 variability. *Nature*, **484**, 228–232.

446

447 Buckley, M. W. and J. Marshall (2016): Observations, inferences, and mechanisms of
448 the Atlantic Meridional Overturning Circulation: A review. *Rev. Geophys.*, **54**, 5-63,
449 doi:10.1002/2015RG000493.

450

451 Cane, M.A., A.C. Clement, L.N. Murphy, and K. Bellomo, 2017: Low-Pass Filtering,
452 Heat Flux, and Atlantic Multidecadal Variability. *J. Climate*, **30**, 7529–7553,
453 doi:10.1175/JCLI-D-16-0810.1

454

455 Cattiaux, J., and C. Cassou, 2013: Opposite CMIP3/CMIP5 trends in the wintertime
456 Northern Annular Mode explained by combined local sea ice and remote tropical
457 influences. *Geophys. Res. Lett.*, **40**, 3682-3687, doi:10.1002/grl.50643.

458

459 Chikamoto, Y., T. Mochizuki, A. Timmermann, M. Kimoto, and M. Watanabe, 2016:
460 Potential tropical Atlantic impacts on Pacific decadal climate trends, *Geophys. Res.*
461 *Lett.*, **43**, 7143–7151, doi:10.1002/2016GL069544.

462

463 Chylek, P., M. K. Dubey, G. Lesins, J. Li, and N. Hengartner, 2014: Imprint of the
464 Atlantic multi-decadal oscillation and Pacific decadal oscillation on southwestern US
465 climate: Past, present, and future. *Clim. Dyn.*, **43**, 119–129, doi:10.1007/s00382-013-
466 1933-3.

467

468 Clement, A., K. Bellomo, L. N. Murphy, M. A. Cane, T. Mauritsen, G. Radel, and B.
469 Stevens, 2015: The Atlantic multidecadal oscillation without a role for ocean
470 circulation. *Science*, **350**, 320–324, doi:10.1126/science.aab3980.

471

472 Danabasoglu, G., S.G. Yeager, W.M. Kim, E. Behrens, M. Bentsen, D. Bi et al.,
473 2016: North Atlantic simulations in Coordinated Ocean-ice Reference Experiments
474 phase II (CORE-II). Part II: inter-annual to decadal variability. *Ocean Model.*, **97**, 65-
475 90, doi:10.1016/j.ocemod.2015.11.007
476

477 Delworth, T.L., F. Zeng, L. Zhang, R. Zhang, G.A. Vecchi, and X. Yang, 2017: The
478 Central Role of Ocean Dynamics in Connecting the North Atlantic Oscillation to the
479 Extratropical Component of the Atlantic Multidecadal Oscillation. *J. Climate*, **30**,
480 3789–3805, doi:10.1175/JCLI-D-16-0358.1
481

482 Deser, C., J. W. Hurrell and A. S. Phillips, 2016: The Role of the North Atlantic
483 Oscillation in European Climate Projections. *Clim. Dyn.*, doi: 10.1007/s00382-016-
484 3502-z.
485

486 Ding, H., R.J. Greatbatch, M. Latif, W. Park, and R. Gerdes, 2013: Hindcast of the
487 1976/77 and 1998/99 Climate Shifts in the Pacific. *J. Climate*, **26**, 7650–7661,
488 doi:10.1175/JCLI-D-12-00626.1
489

490 Doblas-Reyes F.J., I. Andreu-Burillo, Y. Chikamoto, J. García-Serrano, V. Guemas,
491 M. Kimoto, T. Mochizuki, L.R. Rodrigues and G.J. van Oldenborgh, 2013: Initialized
492 near-term regional climate change prediction. *Nat Commun.*, **4**,1715,
493 doi:10.1038/ncomms2704
494

495 Dong, B., R. T. Sutton, and A. A. Scaife, 2006: Multidecadal modulation of El Niño–
496 Southern Oscillation (ENSO) variance by Atlantic Ocean sea surface temperatures.
497 *Geophys. Res. Lett.*, **33**, L08705, doi:10.1029/2006GL025766.

498

499 Dong, B. and A. Dai, 2015: The influence of the Interdecadal Pacific Oscillation on
500 Temperature and Precipitation over the Globe. *Clim. Dyn.*, **45**, 2667-2681,
501 doi:10.1007/s00382-015-2500-x

502

503 Drijfhout, S. S., A. T. Blaker, S. A. Josey, A. J. G. Nurser, B. Sinha, and M. A.
504 Balmaseda, 2014: Surface warming hiatus caused by increased heat uptake across
505 multiple ocean basins, *Geophys. Res. Lett.*, **41**, 7868–7874,
506 doi:10.1002/2014GL061456.

507

508 Dunstone N., D. Smith, A. Scaife, L. Hermanson, R. Eade, N. Robinson, M. Andrews
509 and J. Knight, 2016: Skilful predictions of the winter North Atlantic Oscillation one
510 year ahead, *Nature Geoscience*, **9**, 809–814, doi:10.1038/ngeo2824

511

512 Eade, R., D. Smith, A. Scaife, E. Wallace, N. Dunstone, L. Hermanson, and N.
513 Robinson, 2014: Do seasonal-to-decadal climate predictions under-estimate the
514 predictability of the real world?, *Geophys. Res. Lett.*, **41**, 5620-5628
515 doi:10.1002/2014GL061146

516

517 England, M. H., M. H. England, S. McGregor, P. Spence, G. A. Meehl, A.
518 Timmermann, W. Cai, A. S. Gupta, M. J. McPhaden, A. Purich and A/ Santosoal.,

519 2014: Slowdown of surface greenhouse warming due to recent Pacific trade wind
520 acceleration. *Nature Clim. Change*, **4**, 222-227, doi:10.1038/nclimate2106.

521

522 Emile-Geay, J. et al., 2017: A global multiproxy database for temperature
523 reconstructions of the Common Era. *Scientific Data*, ISSN 2052-4463 (In Press)

524

525 Fan, T., C. Deser, and D. P. Schneider, 2014: Recent Antarctic sea ice trends in the
526 context of Southern Ocean surface climate variations since 1950. *Geophys. Res.
527 Lett.*, **41**, 2419-2426, doi:10.1002/2014GL059239.

528

529 Farneti R., 2016: Modelling Interdecadal Climate Variability and the role of the ocean.
530 *WIREs Climate Change*, **8**, doi:10.1002/wcc.441.

531

532 Fučkar, N. S., D. Volpi, V. Guemas, and F. J. Doblas-Reyes, 2014: A posteriori
533 adjustment of near-term climate predictions: Accounting for the drift dependence on
534 the initial conditions. *Geophys. Res. Lett.*, **41**, 5200–5207,
535 doi:10.1002/2014GL060815.

536

537 Fyfe J. C., G. A. Meehl, M. H. England, M. E. Mann, B. D. Santer, G. M. Flato, E.

538 Hawkins, N. P. Gillett, S-P. Xie, Y. Kosaka and N. C. Swart, 2016: Making sense of
539 the early-2000s global warming slowdown, *Nature Climate Change*, **6**, 224-228,
540 doi:10.1038/nclimate2938

541

542 Gillett N. P., F. W. Zwiers, A. J. Weaver and P. A. Stott, 2003: Detection of human
543 influence on sea level pressure. *Nature*, **422**, 292-294.

544

545 Han, W. et al., 2014: Intensification of decadal and multi-decadal sea level variability
546 in the western tropical Pacific during recent decades. *Clim. Dyn.*, **43**, 1357-1379, doi:
547 10.1007/s00382-013-1951-1

548

549 Häkkinen, S., P. B. Rhines and D. L. Worthen, 2011: Warm and saline events
550 embedded in the meridional circulation of the northern North Atlantic. *J. Geophys.*
551 *Res.*, **116**, C03006, doi:10.1029/2010JC006275.

552

553 Hawkins E., B. Dong, J. Robson, and R. Sutton, 2014: The Interpretation and Use of
554 Biases in Decadal Climate Predictions. *J. Climate*, **27**, 2931-2947, doi: 10.1175/JCLI-
555 D-13-00473.1

556

557 Hedemann, C., T. Mauritsen, J. Jungclaus and J. Marotzke, 2017: The subtle origins
558 of surface-warming hiatuses. *Nat. Clim. Change*, **7**, 336–339.
559 doi:10.1038/nclimate3274.

560

561 Henley, B.J., and co-authors, 2017: Spatial and temporal agreement in climate
562 model simulations of the Interdecadal Pacific Oscillation, *Env. Res. Lett.*, **12**, 044011,
563 doi: 10.1088/1748-9326/aa5cc8

564

565 Hermanson L., R. Eade, N.H. Robinson, N.J. Dunstone, M.B. Andrews, J.R. Knight,
566 A. A. Scaife and D. Smith, 2014: Forecast cooling of the Atlantic subpolar gyre and
567 associated impacts. *Geophys Res Lett.*, **41**, 5167-5174. doi:10.1002/2014GL060420

568

569 Huber, M. and R. Knutti, 2014: Natural variability, radiative forcing and climate
570 response in the recent hiatus reconciled. *Nat. Geosci.*, **7**, 651–656,
571 doi:10.1038/ngeo2228.

572

573 Karspeck, A. R., and Coauthors, 2015: Comparison of the Atlantic meridional
574 overturning circulation between 1960 and 2007 in six ocean reanalysis products.
575 *Clim. Dyn.*, doi:10.1007/s00382-015-2787-7

576

577 Kelley, C., M. Ting, R. Seager and Y. Kushnir, 2012: The relative contributions of
578 radiative forcing and internal climate variability to the late 20th century winter drying
579 of the Mediterranean region. *Clim. Dyn.*, 38, 2001-2015, doi: 10.1007/s00382-011-
580 1221-z

581

582 Kim, H.-M., P. J. Webster, and J. A. Curry, 2012: Evaluation of short-term climate
583 change prediction in multi-model CMIP5 decadal hindcasts, *Geophys. Res. Lett.*, **39**,
584 L10701, doi:10.1029/2012GL051644.

585

586 Kirtman, B., S.B. Power, J.A. Adedoyin, G.J. Boer, R. Bojariu, I. Camilloni, F.J.
587 Doblas-Reyes, A.M. Fiore, M. Kimoto, G.A. Meehl, M. Prather, A. Sarr, C. Schär, R.
588 Sutton, G.J. van Oldenborgh, G. Vecchi and H.J. Wang, 2013: Near-term Climate
589 Change: Projections and Predictability. In: *Climate Change 2013: The Physical*
590 *Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the*
591 *Intergovernmental Panel on Climate Change* [Stocker, T.F., D. Qin, G.-K. Plattner,
592 M. Tignor, S.K. Allen, J. Boschung, A. Nauels, Y. Xia, V. Bex and P.M.

593 Midgley (eds.)]. Cambridge University Press, Cambridge, United Kingdom and New
594 York, NY, USA.

595

596 Kharin, V. V., G. J. Boer, W. J. Merryfield, J. F. Scinocca, and W.-S. Lee, 2012:
597 Statistical adjustment of decadal predictions in a changing climate, *Geophys. Res.*
598 *Lett.*, **39**, L19705, doi:10.1029/2012GL052647.

599

600 Kidston J., A. A. Scaife, S. C. Hardiman, D. M. Mitchell, N. Butchart, M. P. Baldwin
601 and L. J. Gray, 2015: Stratospheric influence on tropospheric jet streams, storm
602 tracks and surface weather. *Nat. Geoscience*, **8**, 433-440, doi:10.1038/ngeo2424.

603

604 Kosaka, Y. and S.-P. Xie, 2013: Recent global-warming hiatus tied to equatorial
605 Pacific surface cooling. *Nature*, 501, 403–407, doi:10.1038/nature12534.

606

607 Kosaka, Y., and S.-P. Xie, 2016: The tropical Pacific as a key pacemaker of the
608 variable rates of global warming. *Nature Geo.*, **9**, 669-673,doi:10.1038/ngeo2770.

609

610 Kociuba, G. and S.B. Power, 2015: Inability of CMIP5 models to simulate recent
611 strengthening of the Walker circulation: implications for projections. *J. Climate*, **28**,
612 20-35.

613

614 Kuntz, L.B. and D.P. Schrag, 2016: Impact of Asian aerosol forcing on tropical Pacific
615 circulation, and the relationship to global temperature trends. *J. Geophys. Res.*
616 *Atmos*, **121**, 403-414, doi:10.1002/2016JD025430.

617

618 Kucharski F., F. Ikram, F. Molteni, R. Farneti, I-S. Kang, H-H. No, M. P. King, G.
619 Giuliani and K. Mogensen, 2016: Atlantic forcing of Pacific decadal variability. *Clim*
620 *Dyn.*, **46**, 2337-2351, doi:10.1007/s00382-015-2705-z

621

622 Lean, J. L., and D. H. Rind, 2009): How will Earth's surface temperature change in
623 future decades? *Geophys. Res. Lett.*, **36**, L15708, doi:10.1029/2009GL038932.

624

625 Lewandowsky, S., J.S. Risbey and N.Oreskes, 2016: The “pause” in global warming:
626 turning a routine fluctuation into a science problem”, *Bull. Amer. Meteor. Soc.*, **97**,
627 723-733, doi:10.1175/BAMS-D-14-00106.1.

628

629 Linsley, B. K., H. C. Wu, E. P. Dassié, and D. P. Schrag, 2015: Decadal changes in
630 South Pacific sea surface temperatures and the relationship to the Pacific decadal
631 oscillation and upper ocean heat content. *Geophys. Res. Lett.*, **42**, 2358–2366. doi:
632 10.1002/2015GL063045.

633

634 Lyu, K. and J.-Y. Yu, 2017: Climate impacts of the Atlantic Multidecadal Oscillation
635 simulated in the CMIP5 models: a re-evaluation based on a revised index, *Geophys.*
636 *Res. Lett.*, 2017GL072681, doi:10.1002/2017GL072681.

637

638 Ma, X., Z. Jing, P. Chang, X. Liu, R. Montuoro, R. J. Small, F.O. Bryan, R. J.
639 Greatbatch, P. Brandt, D. Wu, X. Lin and L. Wu, 2016: Western boundary currents
640 regulated by interaction between ocean eddies and the atmosphere. *Nature*, **535**,
641 533-537, doi:39510.1038/nature1864

642

643 Marotzke, J. et al., 2016: MiKlip - a National Research Project on Decadal Climate
644 Prediction. *Bull. Amer. Met. Soc.*, **97**, 2379-2394, doi: 10.1175/BAMS-D-15-00184.1
645

646 Medhaug I. and T., Furevik, 2011: North Atlantic 20th century multidecadal variability
647 in coupled climate models: sea surface temperature and ocean overturning
648 circulation. *Ocean Sci.*, **7**, 389-404, doi: 10.5194/os-7-389-2011.
649

650 Medhaug, I., M.B. Slope, E.M. Fischer, R. Knutti, 2017: Reconciling controversies
651 about the 'global warming hiatus'. *Nature*, **545**, 41-47, doi: 10.1038/nature22315
652

653 Meehl, G.A., H. Teng and J.M. Arblaster, 2014: Climate model simulations of the
654 observed early-2000s hiatus of global warming. *Nature Clim. Change*, **4**, 898—902,
655 doi: 10.1038/nclimate2357.
656

657 Meehl, G.A., L. Goddard, G. Boer, R. Burgman, G. Branstator, C. Cassou, S. Corti,
658 G. Danabasoglu, F. Doblas-Reyes, E. Hawkins, A. Karspeck, M. Kimoto, A. Kumar,
659 D. Matei, J. Mignot, R. Msadek, A. Navarra, H. Pohlmann, M. Rienecker, T. Rosati, E.
660 Schneider, D. Smith, R. Sutton, H.Y. Teng, G.J. van Oldenborgh, G. Vecchi, S.
661 Yeager, 2014: Decadal Climate Prediction: An Update from the Trenches, *Bull. Amer.*
662 *Meteor. Soc.*, **95**, 243-267, doi: 10.1175/BAMS-D-12-00241.1.
663

664 Meehl, G. A., A. Hu, B. D. Santer and S.-P. Xie, 2016 : Contribution of the
665 Interdecadal Pacific Oscillation to twentieth-century global surface temperature
666 trends. *Nat. Clim. Chang.*, **6**, 1005–1008, doi:10.1038/nclimate3107.
667

668 Meehl, G.A., A. Hu, and H. Teng, 2016: Initialized decadal prediction for transition to
669 positive phase of the Interdecadal Pacific Oscillation. *Nature Comm.*, **7**,
670 doi:10.1038/ncomms11718.

671

672 Menary M.B., D.L.R. Hodson, J.I. Robson, R.T. Sutton, R.A. Wood, J.A. Hunt, 2015:
673 Exploring the impact of CMIP5 model biases on the simulation of North Atlantic
674 decadal variability. *Geophys Res Lett.*, **42**, 5926-5934, doi:10.1002/2015GL064360

675

676 Oldenborgh G.J., F. J. Doblas-Reyes, B. Wouters, W. Hazeleger, 2014: Decadal
677 prediction skill in a multi-model ensemble. *Clim Dyn.*, **38**, 1263–80,
678 doi:10.1007/s00382-012-1313-4

679

680 Msadek, R., T.L. Delworth, A. Rosati, W. Anderson, G. Vecchi, Y.S. Chang et al.
681 (2014): Predicting a decadal shift in North Atlantic climate variability using the GFDL
682 forecast system. *J Clim.*, **27**, 6472–6496, doi:10.1175/JCLI-D-13-00476.1.

683

684 Newman, M., M.A. Alexander, T.R. Ault, K.M. Cobb, C. Deser, E. Di Lorenzo, N.J.
685 Mantua, A.J. Miller, S. Minobe, H. Nakamura, N. Schneider, D.J. Vimont, A.S.
686 Phillips, J.D. Scott, and C.A. Smith, 2016: The Pacific Decadal Oscillation, Revisited.
687 *J. Climate*, **29**, 4399–4427, doi: 10.1175/JCLI-D-15-0508.

688

689 O'Reilly, C. H., M. Huber, T. Woollings, and L. Zanna, 2016: The signature of low-
690 frequency oceanic forcing in the Atlantic Multidecadal Oscillation. *Geophys. Res.*
691 *Lett.*, **43**, 2810–2818, doi:10.1002/2016GL067925.

692

693 Otterå, O.H, M. Bentsen, H. Drange and L. Suo, 2010: External forcing as a
694 metronome for the Atlantic Multidecadal Variability, *Nature. Geo.*, **3**, 688-694, doi:
695 10.1038/ngeo955
696
697 Parker A.O., M.W. Schmidt, P. Chang, 2015: Tropical North Atlantic subsurface
698 warming events as a fingerprint for AMOC variability during Marine Isotope Stage 3.
699 *Paleoceanography*, **30**,1425–1436
700
701 Peings, Y., G. Simpkins, and G. Magnúsdóttir, 2016: Multidecadal fluctuations of the
702 North Atlantic Ocean and feedback on the winter climate in CMIP5 control
703 simulations, *J. Geophys. Res. Atmos.*, **121**, 2015JD024107,
704 doi:10.1002/2015JD024107.
705
706 Pohlmann H., J. Kröger, R. J. Greatbatch, W. Müller, 2016: Initialization shock in
707 decadal hindcasts due to errors in wind stress over the tropical Pacific. *Clim. Dyn.*
708 doi:10.1007/s00382-016-3486-8
709
710 Qasmi, S., C. Cassou and J. Boé, 2017: The teleconnection between Atlantic
711 Multidecadal Variability and European temperature: understanding of the CMIP5
712 models diversity and model evaluation. *Geophys. Res. Lett.*, submitted.
713
714 Reynolds, D., C.A. Richardson, J.D. Scourse, P.G. Butler, P. Hollyman, A. Román-
715 González, I.R. Hall, 2017: Reconstructing North Atlantic marine climate variability
716 using an absolutely-dated sclerochronological network. *Palaeogeography,*
717 *Palaeoclimatology, Palaeoecology*, **645**, 333-346, doi: 10.1016/j.palaeo.2016.08.006

718

719 Robson, J., R. Sutton, K. Lohmann, D. Smith and M.D. Palmer, 2012: Causes of the
720 rapid warming of the North Atlantic Ocean in the mid-1990s. *J Clim.*, **25**, 4116-4134,
721 doi:10.1175/JCLI-D-11-00443.1.

722

723 Robson J., R. Sutton and D. Smith, 2013: Predictable climate impacts of the decadal
724 changes in the ocean in the 1990s. *J Clim.*, **26**, 6329-6339. doi:10.1175/JCLI-D-12-
725 00827.1

726

727 Robson, J. P. Ortega and R. Sutton, 2016: A reversal of climate trends in the North
728 Atlantic since 2005. *Nature Geosci.*, **9**, 513-517.

729

730 Ruiz-Barradas, A., S. Nigam, and A. Kavvada, 2013: The Atlantic Multidecadal
731 Oscillation in twentieth century climate simulations: uneven progress from CMIP3 to
732 CMIP5, *Clim. Dyn.*, **41**, 3301–3315, doi:10.1007/s00382-013-1810-0.

733

734 Ruprich-Robert, Y. and C. Cassou, 2015: Combined influences of seasonal East
735 Atlantic Pattern and North Atlantic Oscillation to excite Atlantic multidecadal
736 variability in a climate model. *Clim. Dyn.*, **44**, 229-253, doi:10.1007/s00382-014-
737 2176-7 .

738

739 Ruprich-Robert, Y., R. Msadek, F. Castruccio, S. Yeager, T. Delworth and G.
740 Danabasoglu, 2016: Assessing the climate impacts of the observed Atlantic
741 multidecadal variability using the GFDL CM2.1 and NCAR CESM1 global climate
742 models. *J. Climate*, **30**, 2585-2810, doi: 10.1175/JCLI-D-16-0127.s1

743

744 Sanchez-Gomez, E., C. Cassou, Y. Ruprich-Robert, E. Fernandez and L. Terray,
745 2015: Drifts dynamics in a coupled model initialized for decadal forecasts. *Clim. Dyn.*,
746 **46**, 1819-1840.

747

748 Santer, B.D., and co-authors, 2017: Causes of differences between model and
749 satellite tropospheric warming rates. *Nature Geoscience*, **10**, 478-485,
750 doi:10.1038/ngeo2973

751

752 Schubert, S.D., M.J. Suarez, P.J. Pegion, R.D. Koster and J.T. Bacmeister, 2004: On
753 the cause of the 1930s dust bowl. *Science*, **303**, 1855-1859, doi:
754 10.1126/science.1095048

755

756 Smith, D.M. et al., 2013: Real-time multi-model decadal climate predictions. *Clim.*
757 *Dyn.*, **41**, 2875-2888, doi:10.1007/s00382-012-1600-0

758

759 Smith, D. M., Allan, R. P., Coward, A. C., Eade, R., Hyder, P., Liu, C., Loeb, N. G.,
760 Palmer, M. D., Roberts, C. D. and A. A. Scaife, 2015: Earth's energy imbalance since
761 1960 in observations and CMIP5 models. *Geophys. Res. Lett.*, **42**, 1205–1213, doi:
762 10.1002/2014GL062669.

763

764 Smith, D.M., B.B.B. Booth, N.J. Dunkstone, R. Eade, L. Hermanson, G.S. Jones,
765 A.A. Scaife, K.L. Sheen and V. Thompson, 2016: Role of the volcanic and
766 anthropogenic aerosols in the recent global surface warming slowdown. *Nature Clim.*
767 *Change*, **6**, 936-940, doi:10.1038/nclimate3058

768

769 Suckling, E, G. J. van Oldenborgh, J. M. Eden and E. Hawkins, 2017: An empirical
770 model for probabilistic decadal prediction: A global attribution and regional hindcasts.

771 *Clim. Dyn.*, 48, 3115-3138, doi: 10.1007/s00382-016-3255-8

772

773 Swingedouw D., P. Ortega, J. Mignot, E. Guilyardi, V. Masson-Delmotte, P.G. Butler,

774 Khodri, M. and R. S  f  rian (2015): Bidecadal North Atlantic ocean circulation

775 variability controlled by timing of volcanic eruptions. *Nat. Commun.*, **6**, 6545, doi:

776 10.1038/ncomms7545.

777

778 Tandon, N.F., and P.J. Kushner, 2015: Does External Forcing Interfere with the

779 AMOC's Influence on North Atlantic Sea Surface Temperature?, *J. Climate*, **28**,

780 6309-6323. DOI:10.1175/JCLI-D-14-00664.1

781

782 Tierney, J. E., N. J. Abram, K. J. Anchukaitis, M. N. Evans, C. Giry, K. H. Kilbourne,

783 C. P. Saenger, H. C. Wu, and J. Zinke, 2015: Tropical sea surface temperatures for

784 the past four centuries reconstructed from coral archives. *Paleoceanography*, **30**,

785 226–252. doi: 10.1002/2014PA002717.

786

787 Thoma, M., R.J. Greatbatch, C. Kadow, and R. Gerdes, 2015: Decadal hindcasts

788 initialized using observed surface wind stress: evaluation and prediction out to 2024.

789 *Geophys. Res. Lett.*, **42**, 6454-6461.

790

791 Toniazzo, T. and S. Woolnough, 2014: Development of warm SST errors in the
792 southern tropical Atlantic in CMIP5 decadal hindcasts, *Clim. Dyn.*, **43**, 2889-2913,
793 doi:10.1007/s00382-013-1691-2

794

795 Trenberth, K.E., J.T. Fasullo, K. von Schuckmann, and L. Cheng, 2016: Insights into
796 Earth's Energy Imbalance from Multiple Sources. *J. Climate*, **29**, 7495–7505, doi:
797 10.1175/JCLI-D-16-0339.1.

798

799 van Dijk, A. I. J. M., H. E. Beck, R. S. Crosbie, R. A. M. de Jeu, Y. Y. Liu, G. M.
800 Podger, B. Timbal, and N. R. Viney, 2013: The Millennium Drought in southeast
801 Australia (2001–2009): Natural and human causes and implications for water
802 resources, ecosystems, economy, and society, *Water Resour. Res.*, **49**,
803 doi:10.1002/wrcr.20123.

804

805 Wang, C., S. Dong, A. T. Evan, G. R. Foltz, and S.-K. Lee, 2012: Multidecadal
806 covariability of North Atlantic sea surface temperature, African dust, Sahel rainfall,
807 and Atlantic hurricanes. *J. Climate*, **25**, 5404–5415, doi:10.1175/JCLI-D-11-00413.1.

808

809 Xie, S.-P., B. Lu and B. Xiang, 2013: Similar spatial patterns of climate responses to
810 aerosol and greenhouse gas changes. *Nature Geo.*, **6**, 828-832,
811 doi:10.1038/ngeo1931.

812

813 Xie, S.-P., Y. Kosaka and Y. Okumura, 2016: Distinct energy budgets for
814 anthropogenic and natural changes during global warming hiatus. *Nature Geo.*, **9**,
815 29-33, doi:10.1038/ngeo2581.

816

817 Xie, S.P. and Y. Kosaka (2017): What caused the surface warming hiatus from 1998-
818 2013? *Curr. Clim. Change Rep.*, **3**, 128, doi:10.1007/s40641-017-0063-0

819

820 Yan, X.-H., Boyer, T., Trenberth, K., Karl, T. R., Xie, S.-P., Nieves, V., Tung, K.-K.
821 and Roemmich, D., 2016: The global warming hiatus: Slowdown or redistribution?.
822 *Earth's Future*, **4**, 472–482, doi:10.1002/2016EF000417

823

824 Yeager, S.G., A.R. Karspeck and G. Danabasoglu, 2015: Predicted slowdown in the
825 rate of Arctic sea ice loss. *Geophys. Res. Lett.*, **10**, 10704-10713, doi:
826 10.002/2015GL065364

827

828 Yeager, S.G. and J.J. Robson, 2017: Recent progress in understanding and
829 predicting decadal climate variability. *Curr. Clim. Change Rep.*, **3**, 112-127, doi:
830 10.1007/s40641-017-0064-z

831

832 Zhang, R. and T.L. Delworth, 2006: Impact of Atlantic multidecadal oscillations on
833 India/Sahel rainfall and Atlantic hurricanes. *Geophys. Res. Lett.*, **33**, L17712,
834 doi:10.1029/2006GL026267.

835

836 Zhang, R., T.L. Delworth, R. Sutton, D.L. Hodson, K.W. Dixon, I.M. Held, Y. Kushnir,
837 J. Marshall, Y. Ming, R. Msadek, J. Robson, A.J. Rosati, M. Ting, and G.A. Vecchi,
838 2013: Have Aerosols Caused the Observed Atlantic Multidecadal Variability?. *J.*
839 *Atmos. Sci.*, **70**, 1135-1144, doi:10.1175/JAS-D-12-0331.1

840

841 Zhang, R. et al., 2016: Comment on “The Atlantic Multidecadal Oscillation without a
842 role for ocean circulation. *Science*, **352**,1527, doi:10.1126/science.aff1660.

843

844 Zhang, L., and C. Wang, 2013: Multidecadal North Atlantic sea surface temperature
845 and Atlantic meridional overturning circulation variability in CMIP5 historical
846 simulations, *J. Geophys. Res. Oceans*, **118**, 5772–5791, doi:10.1002/jgrc.20390.

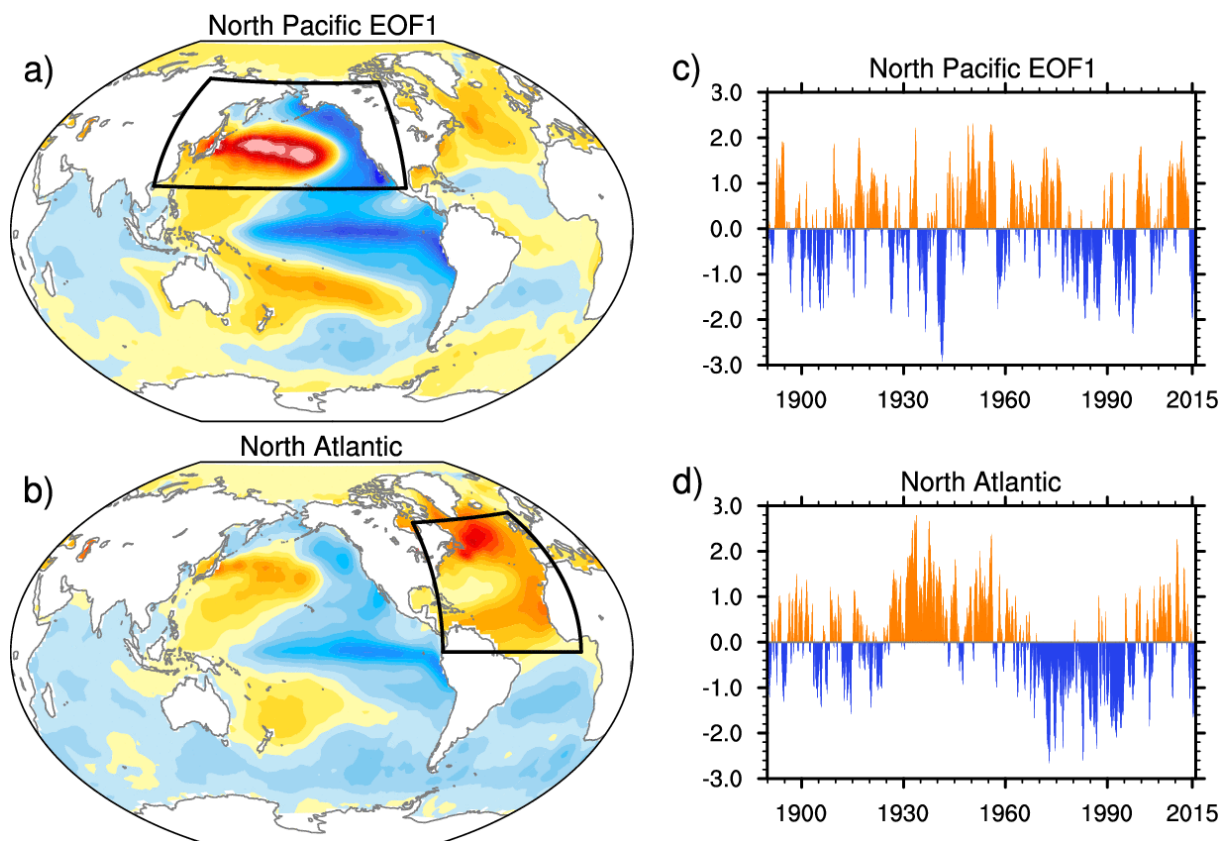
847

848 **Figure captions**

849 **Figure 1:** Spatial and temporal characteristics of sea surface temperature anomaly
850 (SSTA) variability in selected ocean basins. (Left column) Global SSTA regression
851 maps (degrees C) based on the (a) leading principal component of North Pacific
852 SSTA, (b) North Atlantic SSTA. All indices were standardized prior to computing the
853 regression maps. Index regions are outlined by black boxes. (Right column)
854 Standardized 3-month running mean time series of the (a) leading principal
855 component of North Pacific SSTA, (b) North Atlantic SSTA. Global mean SSTA was
856 removed prior to computing the time series and regression maps. Figure adapted
857 from Deser and Phillips (2017), CLIVAR exchanges.

858

859 **Figure 1:**



860
861
862
863
864
865
866
867
868
869
870
871

Figure 1. Spatial and temporal characteristics of sea surface temperature anomaly (SSTA) variability in selected ocean basins. (Left column) Global SSTA regression maps (degrees C) based on the (a) leading principal component of North Pacific SSTA, (b) North Atlantic SSTA. All indices were standardized prior to computing the regression maps. Index regions are outlined by black boxes. (Right column) Standardized 3-month running mean time series of the (a) leading principal component of North Pacific SSTA, (b) North Atlantic SSTA. Global mean SSTA was removed prior to computing the time series and regression maps. Figure adapted from Deser and Phillips (2017), CLIVAR exchanges.