

2

3 **Chapter 13. ENSO Response to Greenhouse Forcing**

4 Wenju Cai^{1,2}, Agus Santoso^{2,3}, Guojian Wang^{1,2}, Lixin Wu¹, Mat Collins⁴, Matthieu
5 Lengaigne⁵, Scott Power⁶, Axel Timmermann^{7,8,9}

6

7 ¹Key Laboratory of Physical Oceanography/Institute for Advanced Ocean Studies, Ocean University
8 of China and Qingdao National Laboratory for Marine Science and Technology, Qingdao, China.

9 ²Centre for Southern Hemisphere Oceans Research (CSHOR), CSIRO Oceans and Atmosphere,
10 Hobart, Tasmania, Australia

11 ³Australian Research Council (ARC) Centre of Excellence for Climate Extremes and Climate Change
12 Research Centre, The University of New South Wales, Sydney, NSW, Australia.

13 ⁴College of Engineering, Mathematics, and Physical Sciences, University of Exeter, Exeter, UK

14 ⁵LOCEAN/IPSL, Sorbonne Universités/UPMC-CNRS-IRD-MNHN, Paris, France

15 ⁶Australian Bureau of Meteorology, Melbourne, Victoria, Australia

16 ⁷Center for Climate Physics, Institute for Basic Science (IBS), Busan, South Korea

17 ⁸Pusan National University, Busan, South Korea

18 ⁹International Pacific Research Center, University of Hawaii at Manoa, Honolulu, HI, USA

19

20 Corresponding email: wenju.cai@csiro.au

27 How ENSO responds to an increasing concentration of greenhouse gases in the atmosphere has
28 remained an elusive issue for decades. Climate models produce widely diverging results based
29 on the traditional sea surface temperature (SST) metrics of ENSO. Some models show stronger
30 ENSO, some weaker, some show no clear change. Steering away from these static measures,
31 but more carefully examining the underlying processes and the associated key physical
32 characteristics of ENSO, a clearer picture begins to emerge. Due to the nonlinear response of
33 the atmosphere to SSTs, an enhancement in ENSO-driven precipitation is favored. Such a
34 response tends to be robust across models featuring relatively strong inter-model agreement in
35 the projected changes of the Pacific mean climate, marked by equatorially enhanced warming
36 and weakened Walker Circulation. These mean-state changes facilitate increased frequency of
37 extreme El Niño events in models that are able to simulate nonlinear properties of ENSO closer
38 to observations. In this ensemble of selected models, the frequency of extreme La Niña events
39 is also projected to increase, as facilitated by faster warming of the Maritime Continent than
40 the surrounding ocean waters. A projected increase in upper-ocean stratification further favors
41 increased variability and occurrences of Eastern Pacific El Niño. Uncertainties however
42 remain due to persistent model biases, highlighting the need to further improving climate
43 models, as well as sustaining reliable observations to constrain model projections. Nonetheless,
44 these projections underscore a possible heightened impact of ENSO-driven changes in a
45 warming climate.

46

47 **13.1. Introduction**

48 Given the significant global-scale impact of ENSO on society, economy, and the environment,
49 understanding how ENSO responds to greenhouse forcing is an urgent critical issue as global
50 greenhouse-gas emissions continue unabated. While ENSO as a naturally occurring climate
51 phenomenon will continue to operate in the warmer future, crucial questions remain which are
52 of relevance to climate and disaster risk management. Will it change in character? Will it
53 become more or less active, stronger or weaker? And if so, what are the mechanisms? Are the
54 climate models used to make the projections reliable? This chapter discusses the current state
55 of understanding that is relevant to addressing these questions, indicative to the possibility that
56 ENSO will respond to future anthropogenic greenhouse forcing in some significant ways.

57 Considering the multitude delicate processes governing ENSO characteristics and evolution as
58 described in earlier chapters of this book, it is not surprising that projecting future ENSO
59 behavior is a challenging undertaking. Every element of the dynamical processes can respond
60 directly to greenhouse forcing, or indirectly via changes in other components of the climate
61 system. This complexity has also posed a great challenge in simulating a realistic ENSO in
62 climate models that are used to make the projections (Chapter 9). ENSO is tightly linked to
63 the tropical Pacific mean climate upon which it evolves (Chapter 8), so it is necessary to
64 understand how the mean climate might change under greenhouse forcing. This will be first
65 discussed in section 13.2. Section 13.3 explains the fundamental reasons for why at a first
66 impression there was no agreement among climate models in the projected ENSO changes. It
67 is not until recently that more robust projections started to emerge, owing to more careful
68 considerations of the nonlinear nature of ENSO. This is to be covered in section 13.4 outlining
69 the projected changes in atmosphere and oceanic aspects of ENSO dynamics along with the
70 associated mechanisms. Section 13.5 discusses the uncertainties underlying the future
71 projections tied to model biases. The chapter concludes in section 13.6 with a summary and
72 concluding remarks.

73

74 **13.2. Forced changes in background climate**

75 Under ENSO neutral conditions, the tropical Pacific climate is characterized by the westward
76 blowing Trade Winds that pile up warm waters in the Western Pacific. The Western Pacific
77 warm pool exhibits annual mean SSTs above 28°C, a necessary condition for maintaining deep
78 atmospheric convection above the ocean (Graham & Barnett, 1987). The warm pool hence
79 hosts an almost permanent deep atmospheric convection, in which mid-tropospheric heating
80 drives the ascending branch of the Walker Circulation and surface easterlies over the eastern
81 and central equatorial Pacific. The rising air flows eastward in the upper troposphere and
82 subsides over the relatively cold and dry eastern equatorial Pacific (see Fig. 1 in Chapter 1).
83 The equatorial Trade Winds generate strong upwelling in the eastern equatorial Pacific through
84 Ekman divergence. The easterly momentum transfer from the equatorial trades to the ocean is
85 balanced by a zonal oceanic pressure gradient, which in turn causes the thermocline to shoal
86 (deepen) in the eastern (western) tropical Pacific. The off-equatorial and equatorial trade winds
87 also generate subtropical meridional oceanic circulation cells (McCreary & Lu, 1994), that

88 transport the upwelled equatorial water poleward, while bringing back subducted subtropical
89 waters towards the equator within the mean thermocline layer at about 100-200 m depths (e.g.,
90 Schott et al., 2004).

91 In response to increasing greenhouse gas emissions many climate models simulate a reduction
92 of the eastern equatorial Pacific cold tongue and stronger warming along the equator than off-
93 equator (Timmermann et al., 1999). This “enhanced equatorial warming” pattern (Liu, et al.
94 2005; Xie et al., 2010; Cai et al., 2015) is due to the fact that the evaporative damping of CO₂-
95 induced warming is weaker in the equatorial strip compared to the off-equatorial regions,
96 because the latent heat flux scales with the mean wind-speed, and the mean wind speed
97 decreases towards the equator (Seager & Murtugudde, 1997; Xie et al., 2010). In addition to
98 the characteristic meridional warming structure, most CMIP3 and CMIP5 models also simulate
99 a slightly stronger warming in the eastern Pacific, compared to the west (e.g., Xie et al. 2010;
100 Power et al., 2013). Climate models with a thermodynamic mixed-layer slab ocean can capture
101 this response (e.g., Vecchi & Soden, 2007), sometimes even stronger than Coupled General
102 Circulation Models (CGCMs).

103 Factors that may contribute to this zonally asymmetric response include a shallower mixed
104 layer in the eastern Pacific, and asymmetric cloud feedbacks (stratus cloud feedback in the east,
105 cumulus cloud/cirrus feedback in the west) (Meehl & Washington, 1996), but also a deepening
106 of the eastern equatorial thermocline in response to weaker equatorial trade winds (see
107 discussion below). However, the fact that some climate models (e.g., Kohyama et al., 2017)
108 instead simulate a negative zonal SST gradient change along the equator in response to transient
109 CO₂ radiative forcing, suggests that the upwelling of relatively cold water in the eastern Pacific
110 and the associated thermodynamic damping (“dynamical thermostat”; Clement et al., 1996)
111 can partly offset the CO₂-induced surface warming. This argument relies on the assumption
112 that the older waters that upwell in the eastern equatorial Pacific had been exposed to lower
113 atmospheric CO₂-induced warming the last time they were in contact with the surface. This
114 non-equilibrium effect is likely to play an important role in transient climate change
115 simulations, but not under equilibrated conditions in which the amount of warming at
116 subsurface has caught up with that at the surface.

117 The enhanced eastern equatorial warming plays a key role in intensifying rainfall in the
118 equatorial region over the 21st Century (e.g., Vecchi & Soden, 2007; Cai et al. 2014; Power and

119 Delage 2018), due to an increased low-level moisture convergence. Such pattern of rainfall
120 changes is consistent with a ‘warmer gets wetter’ hypothesis (Xie et al., 2010) in which
121 precipitation changes are closely tied to shifts in atmospheric circulation and moisture
122 convergence (Chadwick et al., 2012; Widlansky et al., 2013).

123 The majority of climate models forced with increasing greenhouse gas concentrations show a
124 slow-down of the Walker Circulation (Vecchi et al., 2007; Power & Kociuba, 2011a; Kociuba
125 & Power 2015). Following the Clausius-Clapeyron relation (Clausius, 1850; Clapeyron, 1834),
126 the saturated water vapor in the lower troposphere increases at a global mean rate of about 7%
127 K^{-1} of global warming. Climate models support this thermo-dynamical relationship – at least
128 on a global scale (Collins et al., 2010). However, the rate of precipitation increase is much
129 lower ($\sim 2\% \text{K}^{-1}$). Because the precipitation increase does not keep up with the increase in
130 humidity, there must be a reduction in the mass flux from the moist boundary layer into the
131 dryer air aloft (Held & Soden, 2006). As a consequence, atmospheric vertical motion over
132 tropical convective regions, such as the Western Pacific Warm Pool, is expected to decrease
133 ($\sim 5\text{-}10\% \text{K}^{-1}$), leading to a slow-down in the atmospheric overturning circulation. Even though
134 this argument links the thermodynamics of global warming with atmospheric dynamics, it does
135 not entirely explain how the atmospheric circulation will slow down and whether the slow-
136 down will affect the Walker and Hadley circulation cells. Other factors, such as the enhanced
137 equatorial warming (e.g., Liu et al., 2005) and off-equatorial rainfall increase driven by the
138 background sea surface warming (Kug et al., 2011), can further contribute to the Walker
139 Circulation slow down. The wind change also leads to a reduction of Ekman divergence and
140 equatorial upwelling (Vecchi & Soden, 2007; DiNezio et al., 2009; see Collins et al., 2010 and
141 Cai et al., 2015 for a review). As a result, the east-west tilted thermocline flattens, which can
142 further amplify the initial equatorial warming, in particular in the eastern equatorial Pacific.

143 Regarding the Hadley Circulation, the situation is more complex, because on top of an overall
144 slow-down of the meridional mass stream function (e.g., Vecchi & Soden, 2007; Lu et al., 2007;
145 Seo et al., 2014), there is also evidence for a “deep tropical squeeze” (Lau & Kim, 2015), which
146 tends to enhance the near-equatorial circulation characteristics. Furthermore, given the Hadley
147 cells are influenced by atmospheric stratification, meridional surface temperature gradients,
148 and extratropical eddy dynamics (Schneider, 1977; Seo et al., 2014; Walker & Schneider, 2006),
149 their future strength and properties can be influenced by changes in these various elements.

150 It should be noted here that even though the tropical Pacific warming pattern and an El Niño
151 bear some similarities in terms of their surface temperature characteristics, the underlying
152 processes responsible for the pattern formation are very different. The greenhouse gas-induced
153 westerly wind anomalies cover the entire equatorial Pacific, whereas westerly wind anomalies
154 during an El Niño are confined to the western to central part of the Pacific basin. The weaker
155 Walker Circulation under global warming reduces poleward heat transport, which is near
156 symmetrical about the equator (e.g., Liu et al., 2017). In contrast, there tends to be a net
157 northward heat transport across the equator during an El Niño event (e.g., McGregor et al.
158 2014). The seasonality of the anomalous equatorial Pacific warming may also be different:
159 For instance, in the GFDL CM2.1 model, during an El Niño it tends to peak toward boreal
160 winter, whereas the greenhouse warming response peaks around mid-calendar year (Xie et al.,
161 2010). Further, sea level pressure over tropical South Pacific decreases during an El Niño, but
162 an increase may be detected under global warming (Xie et al., 2010) as would be reflected in
163 the Southern Oscillation Index (SOI) which is negative during El Niño years, whereas under
164 global warming the SOI increases (Power and Kociuba, 2011b). This contrast arises because
165 the SOI depends on MSLP spatial differences, and MSLP changes over the Pacific under global
166 warming and during El Niño years tend to have a different spatial structure (Power & Kociuba,
167 2011b).

168 While the projected mean-state changes outlined above tend to be consistent across models,
169 confidence in these projections is reduced by the existence of observational uncertainty and
170 model biases, and uncertainty in estimates of long-term climate trends derived from different
171 observational datasets.

172 Disentangling long-term observed SST trends in the tropical Pacific in terms of externally
173 forced signals and corresponding feedbacks and natural variability has been hampered by the
174 fact that the different SST observational products show opposing patterns, in particular in the
175 eastern equatorial Pacific (Vecchi et al., 2007; Deser et al., 2010; An et al., 2012). A
176 multivariate statistical decomposition which includes a removal of ENSO, suggests a more
177 robust strengthening of the prevailing SST gradients (Solomon & Newman, 2012) over the
178 period from 1900-2010 – the extent of which is not captured by CMIP5 models (Coats &
179 Karnauskas, 2017). This result would support the role of the “ocean dynamical thermostat”
180 (Clement et al., 1996, Cane et al., 1997), particularly during the boreal fall when the
181 climatological SST gradient is the strongest (Karnauskas et al., 2009). This strengthening of

182 the east-west SST gradient occurs despite a weakening of sea level pressure gradient which
183 indicates a weakened Walker Circulation and Trade Winds (Karnauskas et al., 2009). This near
184 centennial trend could suggest a greenhouse forcing effect that may manifest in stronger zonal
185 SST gradients, although uncertainties still exist in the observational data prior to 1950 (see
186 Chapter 3) and due to the fact that the greenhouse warming signal, natural interdecadal
187 variability and ENSO share some pattern characteristics. In addition, these patterns may also
188 vary over a shorter period. Over 1950-2009, for instance, the zonal SST gradient was found to
189 weaken (Tokinaga et al., 2012a), consistent with the weaker winds (Tokinaga et al., 2012b).
190 However, these are inconsistent with analysis of sea level pressure based on several
191 observational products which instead suggested a stronger Walker Circulation (L'Heureux et
192 al., 2013). These inconsistent changes between zonal SST gradient and Walker Circulation
193 again differ from the case of El Niño in which weaker east-west SST gradient associated with
194 warmer eastern equatorial Pacific is accompanied by weaker Trade Winds through the Bjerknes
195 feedback – a positive feedback loop that sustains ENSO event development (Chapters 1, 2).
196 Under greenhouse warming though, stronger zonal SST gradient may not necessarily be
197 accompanied by stronger Walker Circulation (An, 2011).

198 Diagnosing greenhouse effect in the relatively short observational record is complicated by
199 naturally occurring decadal variability, such as the Interdecadal Pacific Oscillation (IPO; Power
200 et al. 1999). Shifts in the Pacific climate associated with the IPO were observed in the mid
201 70's from a negative IPO to a positive IPO, followed by a shift to a negative in the late 90's.
202 The positive IPO phase was characterized by an SST trend, with stronger warming east of the
203 Dateline and weaker Trade Winds (e.g., Meehl & Washington, 1996). The latter was marked
204 by unprecedented acceleration of the Walker Circulation (Kociuba and Power, 2015), along
205 with a cooler tropical Pacific that contributed to the global warming hiatus (Kosaka and Xie,
206 2013, England et al. 2014). Correspondingly, there were marked changes in ENSO properties.
207 The positive IPO state saw stronger ENSO variability in the eastern equatorial Pacific marked
208 by the 1982/83 and 1997/98 extreme El Niño events, and the negative IPO state had stronger
209 variability in the central Pacific (e.g., Wang and An, 2001; Lee & McPhaden, 2012; Santoso et
210 al. 2017). The link between IPO and ENSO variability is still a topic of intense research, which
211 is further complicated by the fact that mean state changes can influence ENSO properties (e.g.,
212 Fedorov & Philander, 2000; Wang and An, 2002; Power et al. 2013), while changes in ENSO
213 variability can in turn influence multi-decadal variability (Timmermann, 2003; Rodgers et al.,
214 2004; Power & Colman, 2006; Sun et al., 2014; Newman et al., 2016).

215 It is necessary to stress that climate models still suffer from persistent biases (oftentimes larger
216 than their projected global warming responses), thus leaving uncertainties in the projected
217 mean-state changes despite the reasonably strong inter-model consensus on the projections. A
218 well-known model deficiency is the Pacific “cold tongue” bias in which the ribbon of cool
219 eastern equatorial Pacific water extends too far west into the Western Pacific Warm Pool, with
220 the stronger than observed Trade Winds, and a double Intertropical Convergence Zone bias
221 (Chapter 9). It is not entirely clear what ramifications such biases have on future projections,
222 but recent research (Li et al., 2016; Ying et al., 2018) indicates that models with less severe
223 cold-tongue bias tend to project a warming pattern with stronger SST warming in the east
224 Pacific (see Section 13.5 for a discussion).

225

226 **13.3. Elusive projections of ENSO**

227 Despite relative inter-model agreement on the projected change in the 21st Century mean
228 climate, there is a lack of consensus on the change in ENSO as typically diagnosed in terms of
229 SST variability at fixed locations in the equatorial Pacific, such as in the Niño3, Niño3.4, and
230 Niño4 regions. SST is a core variable for ENSO as it is the main way through which ocean-
231 atmosphere feedbacks are mediated, thus the Niño indices have been widely used across
232 research and prediction platforms to characterize and monitor ENSO events.

233 Investigations of how ENSO SST variability could change in response to global warming
234 started in the ‘90s, using climate models that were considered advanced at the time. These
235 studies found little or no changes in future ENSO behavior (e.g., Meehl et al., 1993; Tett, 1995;
236 Knutson et al., 1997). However, the reliability of the results was questionable given the models
237 deficiencies in simulating the complex interacting processes involved in ENSO. Using a model
238 with a more realistic representation of ENSO in part due to higher resolution that can better
239 resolve equatorial wave dynamics, Timmermann et al. (1999) found more frequent El Niños
240 and stronger cold events in the eastern equatorial Pacific under a future emission scenario. The
241 authors argued that a long-term increase of vertical stratification in the eastern tropical Pacific
242 enhanced the sensitivity of SST to ENSO-related wind stress forcing. This higher sensitivity
243 would strengthen the thermocline feedback and may thus contribute to ENSO amplitude
244 changes. A follow-up study (Timmermann, 2001) then presented evidence for a major change
245 of ENSO stability during this greenhouse warming simulation, which translated into rapid

246 amplitude shifts. Stronger and more frequent ENSO were also found in another model, the
 247 Hadley Centre coupled model version 2 (HadCM2), under a four-time pre-industrial CO₂ levels
 248 (Collins, 2000a). In stark contrast, in spite of being forced with the same greenhouse gas
 249 concentration, no appreciable response was found in the third version of the Hadley Centre
 250 model (HadCM3) which had enhanced horizontal ocean resolution, exclusion of flux
 251 adjustments, and subtle changes in the sub-grid scale parameterization schemes that can affect
 252 cloud formation (Collins, 2000b). This discrepancy is an early example of model-based
 253 uncertainties in ENSO projections.

254 Since the '90s, the performance of climate models in simulating ENSO has notably improved,
 255 despite stubborn common biases (see Chapter 9). Facilitated through CMIP, an increasing body
 256 of studies have now analyzed ensembles of different climate models, run with the same forcing
 257 under equivalent emission scenarios. Results from the third phase of CMIP (CMIP3) still
 258 showed no inter-model consensus in terms of amplitude and frequency of ENSO SST
 259 variability (van Oldenborgh et al., 2005; Merryfield, 2006; Guilyardi, 2006; Yeh & Kirtman,
 260 2007; Latif & Keenlyside, 2009; Collins et al., 2010), nor did models that participated in the
 261 later inter-comparison project (CMIP5) (Stevenson, 2012; Santoso et al., 2013; Bellenger et al.,
 262 2014; Taschetto et al., 2014; Kim et al., 2014; Cai et al., 2015; Chen et al., 2017). Some models
 263 showed an increase in the amplitude of ENSO SST variability, some a decrease, and some
 264 showed no appreciable change.

265 One reason for the variety of responses can be explained by the fact that ENSO SST variability
 266 essentially arises from an imbalance between positive and negative feedback processes. For
 267 instance, during the growth phase of an El Niño, positive feedback processes dominate the
 268 negative ones (Stein et al., 2010), resulting in a positive temperature tendency or heating rate
 269 (T_t). If the negative feedback processes dominate, then T_t is negative, resulting in a cooling of
 270 the mixed layer. This can be expressed mathematically in terms of a mixed-layer heat budget,
 271 which decomposes the rate of change of potential temperature T into the different contributing
 272 terms:

$$273 \int_{-H_m}^0 T_t dz = \int_{-H_m}^0 \{Q' - [(u'T'_x + \bar{u} T'_x + u'\bar{T}_x) + (v'T'_y + \bar{v} T'_y + v'\bar{T}_y) + (w'T'_z + \bar{w}T'_z +$$

$$274 w'\bar{T}_z)]\} dz + Res \quad , \quad (1)$$

275 where Q is the net balance between shortwave and long-wave radiations, and latent-heat and
 276 sensible heat fluxes at the air-sea interface [divided by the product of a reference density of

277 $\sim 1026 \text{ kg m}^{-3}$, specific heat capacity of seawater ($3986 \text{ J kg}^{-1} \text{ K}^{-1}$), and mixed layer depth (~ 50
278 m]. The ocean current variables, u , v , and w , denote currents in the zonal, meridional, and
279 vertical direction. The subscripts x , y , z are spatial derivatives in the zonal, meridional, and
280 vertical direction, respectively, and prime indicates deviation from the climatological state
281 denoted by the overbar. These collective terms are integrated across the surface mixed layer
282 of depth H_m (50 m is a reasonable estimate for the average mixed layer depth in the equatorial
283 Pacific), and the residual term, Res , contains unresolved processes such as mixing and diffusion.

284 The main positive feedback processes for ENSO are contained in the square bracketed terms
285 in Eq. (1) of which the current and temperature anomalies are linked to anomalous winds and
286 thermocline. The zonal advective feedback is contained in the $-u'\bar{T}_x$ term, where anomalous
287 zonal current acts on the zonal gradient of the climatological mixed-layer temperature. During
288 a developing El Niño, this term is positive (anomalous eastward advection of warmer western
289 tropical Pacific waters), as the climatological zonal temperature gradient is negative (cooler
290 temperature in the eastern Pacific than in the west) and u' is positive due to westerly wind
291 anomaly and deeper thermocline along equator than off-equator, following the Ekman and
292 geostrophic relations (see Chapter 6). The Ekman pumping feedback is associated with the
293 $-w'\bar{T}_z$ term, which is also positive during an El Niño, since weaker zonal winds lead to weaker
294 upwelling (i.e., $w' < 0$) and \bar{T}_z is positive as climatologically the surface water is warmer than
295 subsurface. The thermocline feedback is associated with the $-\bar{w}T'_z$, which is again positive
296 during an El Niño development, as warm anomalies are stronger in the subsurface than at the
297 sea surface (i.e., $T'_z < 0$) and the climatological vertical velocity anomaly is upward positive
298 for upwelling. Q' is typically a negative feedback or damping term for ENSO SST anomalies:
299 warmer sea surface produces more cloud that reduces incoming solar radiation and enhances
300 latent heat, together acting to cool the surface waters. The opposite generally applies for La
301 Niña, although significant asymmetries exist (Chapter 4), originating from the non-linear terms
302 in Eq. (1).

303 Thus, the heating rate of the east Pacific mixed layer and sea surface temperatures is a net result
304 of a delicate balance across several various positive and negative feedback processes. Because
305 both the positive and negative feedback processes are, in general, projected to increase under
306 global warming (e.g., Philip & van Oldenborgh, 2006; Kim et al., 2011), it may not be very
307 surprising to find diverse changes in ENSO variability across models. Also notice that these
308 feedback processes are a function of the mean basic states, as illustrated in Eq. (1) (see also

309 Chapters 6-8). Thus, changes in the mean state described in section 13.2 affect ENSO
310 projections. Most climate models simulate stronger sensitivity between the wind forcing and
311 ocean dynamical response (i.e., currents and thermocline) over the 21st Century (e.g., Kim et
312 al., 2011; Chen et al., 2017). This is likely due to increased upper-ocean stratification
313 associated with surface intensified warming and thinner mixed layer due to weaker winds, thus
314 bringing the thermocline closer to the surface. Along with the consistent mean-state changes
315 across models, the increased sensitivities tend to enhance the positive feedbacks across models.
316 At the same time, the negative feedback of air-sea thermodynamic damping (i.e., sensitivity
317 between sea surface temperature and net air-sea heat flux, Q) is also projected to enhance across
318 models, due to climatologically increased evaporation and cloudiness (Knutson and Manabe
319 1995).

320 The results discussed above are generally based on ensembles of any available models without
321 scrutinizing their performance in simulating the relevant processes. More recently, performing
322 model selection based on the fidelity in simulating certain aspects of the observed ENSO (e.g.,
323 van Oldenborgh et al., 2005) has become common practice. In many cases, such approach has
324 led to statistically significant detectable changes. For example, when examining only models
325 that most realistically simulate the aforementioned linear feedback processes (Jin et al., 2006),
326 Kim et al. (2014) found an inter-model consensus in the ENSO SST amplitude response that is
327 time varying. In these “more realistic” models, the ENSO amplitude strengthens before 2040
328 when the eastern Pacific warms faster than the maritime region, before decreasing thereafter
329 when the maritime region warming catches up, although the reason for this behavior is not
330 entirely clear. On the other hand, the excluded models exhibit a persistent ENSO amplitude
331 increase in the ensemble mean, but with large inconsistency in the trend across the models. It
332 turned out that the excluded models tend to exhibit stronger cold tongue bias, thus potentially
333 indicating that model bias may affect uncertainty in ENSO projections. In addition, the time-
334 varying nature of the response itself also contributes to weaker inter-model consensus when
335 examining changes averaged across the entire 21st Century.

336 Despite the elusive projections of ENSO SST variability at fixed locations discussed above,
337 there is relatively strong inter-model agreement on how global warming will impact ENSO-
338 driven precipitation in the tropical Pacific (Power et al., 2013). This will be reviewed in the
339 next section, along with how consideration of atmospheric and non-linear processes, as well as
340 application of model selection can lead to further insights about how ENSO behavior will

341 change in the future.

342

343 **13.4. Process-based ENSO projections**

344 Even though the conventional SST metrics for ENSO do not provide a clear consensus picture
345 of how ENSO amplitude will respond to greenhouse warming (Section 13.3), it does not
346 necessarily imply there are no robust changes in the behavior and characteristics of ENSO.
347 There are a number of atmosphere and oceanic processes involved in setting the spatial and
348 temporal evolution of ENSO. These specific underlying processes, which are to a large extent
349 nonlinear and linked to the mean state, could respond to and in turn contribute to climate change,
350 potentially affecting some aspects of ENSO in a robust and significant way. Thus, to better
351 understand how greenhouse warming would affect ENSO, it is important to look closely into
352 each of these processes. This section illustrates that robust changes in ENSO could manifest in
353 terms of the frequency of ENSO extremes arising from the interaction between the mean state
354 and nonlinear processes. The non-linear processes are not well captured by every model which
355 is in part why the multi-model picture does not tell the whole story.

356 *13.4.1 ENSO-driven precipitation response*

357 We first begin the discussion by stressing the results that the nonlinear atmospheric response
358 to ENSO-driven SST variability in a warming world gives rise to greater agreement among
359 models on changes in ENSO-driven precipitation variability under various warming scenarios,
360 than they do for changes in ENSO-driven SST variability (Cai et al. 2012; Power et al., 2013;
361 Watanabe et al., 2014; Cai et al., 2014; Chung and Power, 2014; Chung et al., 2014; Huang &
362 Xie, 2015; Huang, 2016). This is primarily due to two factors. Firstly, the atmospheric response
363 to SST anomalies in the equatorial Pacific is nonlinear (e.g. Hoerling et al., 1997; Kang & Kug,
364 2002; Philip & van Oldenborgh, 2009; Dommenges, et al. 2013; Power et al. 2013; Chung et
365 al., 2014; Takahashi & Dewitte, 2016). Secondly, there is a degree of inter-model agreement in
366 the mean-state change (Section 13.2).

367 The tropical Pacific is characterized by surface waters warmer than 27.5°C west of the dateline
368 toward the maritime continent, and along the major rainfall bands, the ITCZ and SPCZ.
369 Elsewhere, the surface temperature gradually drops to about 20°C toward the coastal regions

370 of the eastern equatorial Pacific. In this “cold tongue” region, atmospheric deep convection is
371 inhibited, and strong positive SST anomalies such as those that occur during an extreme El
372 Niño are required to elevate SSTs and weaken lateral SST gradients to trigger deep convection
373 and thus intense precipitation in the eastern equatorial Pacific. This anomalous convection
374 tends to occur on the western flank of the core SST anomalies, given that SSTs are warmer
375 toward the maritime continent. On the other hand, the impact of negative SST anomalies
376 during a La Niña on eastern equatorial Pacific rainfall is considerably weaker since the cold-
377 tongue region is already cold, and so atmospheric convection cannot decrease any further.
378 However, with increasingly strong cold anomalies, the atmospheric response to La Niña SST
379 anomalies can be large in regions where SSTs are climatologically high, leading to poleward
380 displacement of the ITCZ and SPCZ, and westward shift of maximum equatorial rainfall in the
381 western Pacific warm pool (Chung et al., 2014).

382 The nonlinear interaction between background warming and ENSO-related SST anomalies
383 could hence increase the rainfall response to ENSO SST (Power et al. 2013; Chung et al. 2014;
384 Huang & Xie 2015; Huang, 2016). As the eastern equatorial Pacific is projected to warm in the
385 future (according to multi-model ensemble mean projections), convection can indeed be more
386 easily triggered, even for moderate El Niño events, thereby increasing rainfall considerably in
387 response to a given El Niño SST anomaly (Cai et al., 2012; 2014). But this is achieved through
388 severe weakening of zonal and meridional background SST gradients (Cai et al. 2012; 2014),
389 rather than by simply exceeding the convective SST threshold since the threshold itself also
390 increases in a warmer climate (Johnson & Xie, 2010) due to changing atmospheric background
391 stratification. This spatial pattern of changes in background SST also shift the response pattern
392 of rainfall eastward to the equatorial eastern Pacific and equatorward (Kug et al., 2011; Power
393 et al. 2013; Cai et al., 2012; 2014; Zhou et al. 2014; Huang 2016), leading to more frequent
394 swings of the SPCZ and ITCZ toward the equator (Cai et al., 2012; 2014), and the ENSO-
395 related rainfall asymmetry is enhanced in some locations (Huang & Chen, 2017; Bonfils et al.,
396 2015; Power & Delage, 2017). In essence, the reasons behind the robust projection of ENSO-
397 related precipitation are atmospheric nonlinearity and inter-model agreement in the mean-state
398 change, despite the uncertainty in future ENSO amplitude change (Power et al., 2013;
399 Watanabe et al., 2014).

400 The nonlinear atmospheric response means that heavy rainfall can occur in the usually dry and
401 cold eastern equatorial Pacific when SSTs in the region increase to significantly weaken or

402 even reverse the otherwise positive meridional SST gradient. This occurs during an extreme
403 El Niño when averaged boreal winter rainfall over the Niño3 region (5°S-5°N, 150°W-90°W)
404 reaches more than 5 mm day⁻¹, as observed during the 1982/83, 1997/98, and 2015/16 extreme
405 El Niño events (Fig. 13.3a). However, such a nonlinear feature that characterizes an extreme
406 El Niño is not captured by all climate models (Cai et al., 2014), because of severity of the cold-
407 tongue bias and the ability of models in simulating convection and cloud processes in the
408 eastern equatorial Pacific (see Section 13.5 for discussions). Considering only models in the
409 CMIP3 and CMIP5 archives that are able to simulate this atmospheric nonlinearity, Cai et al.
410 (2014) found a doubling in the frequency of future El Niño events, as characterized by their
411 atmospheric rainfall response, under a “business-as-usual” emission scenario (Fig. 13.3c, d).
412 The frequency increase is not simply due to increased climatological rainfall in the cold-tongue
413 region, but to increased likelihood of convective activity, as also indicated by increases in
414 atmospheric vertical velocity (Cai et al., 2017), linked to changes in the mean state discussed
415 above. From the perspective of rainfall impact, this indicates an increased frequency of
416 extreme El Niño events under greenhouse warming.

417 *13.4.2 Nonlinearity in oceanic variables*

418 Robust future ENSO changes are not limited to the atmosphere, but are also found in the ocean.
419 The weakened Walker Circulation associated with the eastern Pacific warming, leads to weaker
420 equatorial Pacific currents (Vecchi & Soden, 2017), which can influence the evolution of ENSO
421 SST anomalies through heat advection. An asymmetric propagation feature in ENSO SST
422 anomalies along the equatorial Pacific puzzled researchers for decades (Wallace et al., 1998;
423 Neelin et al., 1998; An and Jin, 2004). Specifically, why do SST anomalies appear to propagate
424 westward during La Niña and moderate El Niño events, but eastward during an extreme El
425 Niño of 1982/83 and 1997/98 (McPhaden & Zhang, 2009)? Linear theories showed that
426 eastward propagation arises if the thermocline feedback dominates over other positive feedback
427 processes, otherwise a westward propagation would result (Jin & Neelin, 1993). This would
428 apply to both El Niño and La Niña, in stark contrast to the observed asymmetry. The effect of
429 the positive feedback processes on zonal propagation was confirmed through a mixed layer
430 heat budget analysis (Santoso et al., 2013). However this occurs amidst the interplay between
431 nonlinear zonal advection ($-u'T'_x$) and the advection by the mean current ($-\bar{u}T'_x$) that was
432 found to be a key factor in giving rise to the El Niño-La Niña propagation asymmetry in
433 observations and models that are able to simulate this propagation asymmetry (Santoso et al.,

434 2013). In essence, the strong eastward current anomaly during strong El Niño events is
435 sufficiently large to overcome the westward mean current, such that the total current is eastward
436 ($u' + \bar{u} > 0$), thereby advecting heat anomaly to the east. In all other cases (La Niña and
437 moderate El Niño) the total current remains westward. Thus, as the mean current is projected
438 to weaken under global warming, it would be easier for the current to reverse eastward during
439 an El Niño. This leads to increased frequency of future eastward propagating El Niño events
440 (Santoso et al., 2013). It should be noted that not all simulated extreme El Niño events
441 identified based on the Niño3 rainfall threshold exhibit eastward propagation feature, but the
442 converse is true (Cai et al., 2015b). In a similar token, the 2015/16 extreme El Niño exhibited
443 large Niño3 rainfall, but did not exhibit a clear eastward propagation (Fig. 13.3a) due to the
444 weak eastward current reversal (Santoso et al., 2017).

445 Aside from the projected faster warming of the eastern equatorial Pacific than the surrounding
446 waters, the maritime continent is also projected to warm at a considerably faster rate than the
447 surrounding waters (Cai et al., 2015). This leaves the equatorial central Pacific (Niño4 region;
448 5°S - 5°N , 160°E - 150°W) to be relatively cooler, thus increasing the likelihood for atmospheric
449 subsidence and anomalously strong easterly winds to occur, which are conducive for strong La
450 Niña events. In contrast to strong El Niño events which tend to have SST anomalies peaking
451 in the eastern equatorial Pacific, strong La Niña events tend to peak in the central Pacific (e.g.,
452 Takahashi et al., 2011; Dommenget et al., 2013). Strongly negative Niño4 index distinguishes
453 extreme La Niña events from moderate events (Fig. 13.3.b; Cai et al. 2015). As a salient feature
454 of ENSO asymmetry (see Chapter 4), significant El Niño events tend to be followed by a La
455 Niña, rather than the other way around – associated with El Niño-induced strong warm water
456 volume discharge. The 1998/99 and 1988/1989 extreme La Niña events followed significant
457 El Niño events in the previous year which left the equatorial Pacific subsurface cooler than
458 normal (see Chapter 6). As a result, the thermocline shoals across the equatorial Pacific,
459 promoting surface cooling that can initiate the Bjerknes feedback through Ekman pumping
460 ($-w'\bar{T}_z$) involving easterly wind anomaly. The easterly wind anomaly promotes warming in
461 the western Pacific, and this enhances zonal gradient of the anomalous surface temperature
462 between the central Pacific and maritime continent, thereby further enhancing the Bjerknes
463 feedback through nonlinear zonal advection ($-u'T'_x$). Under climate change, the faster
464 warming of the Maritime Continent facilitates stronger nonlinear zonal advection, and the
465 surface intensified warming favors the Ekman pumping feedback. In an ensemble of models
466 that are able to simulate extreme El Niño (Cai et al., 2014), these favorable conditions for

467 anomalous central Pacific cooling were found to support a near doubling in future occurrences
468 of extreme La Niña events (Cai et al., 2015). The majority of the increase was found to be due
469 to more frequent extreme El Niño events, thus depicting a future scenario of more extreme
470 swings from an extreme El Niño to an extreme La Niña the following year (Fig. 13.3.c, d; Cai
471 et al., 2015).

472 *13.4.3 SST-based metrics revisited*

473 The finding of increased frequency of ENSO extremes, which are inherently nonlinear, stresses
474 the importance of considering nonlinear processes in understanding the effect of greenhouse
475 warming on ENSO. While linear theories (e.g. the recharge oscillator) can explain much of
476 ENSO evolution (Chapter 6), there is a substantial degree of nonlinearity in its basic spatial
477 and temporal properties (Timmermann et al., 2018). The tendency for extreme El Niño
478 anomalies to peak in the eastern equatorial Pacific, and extreme La Niña to peak in the western-
479 central Pacific, results in a positively skewed SST anomaly in the Niño3 region and negative
480 skewness in Niño4 region (Timmermann, 1999). These tendencies can be illustrated by
481 decomposing the tropical SST anomaly into the first and second principal components (PC1
482 and PC2) via an empirical orthogonal function technique. The spatial pattern associated with
483 PC1 represents the canonical El Niño with warm SST anomaly covering the eastern equatorial
484 Pacific; whilst the second mode captures weaker positive SST anomaly straddling the central
485 equatorial Pacific with a negative anomaly off the coast of Peru (see Fig. 3 of Takahashi et al.,
486 11). Plotting PC1 against PC2, which respectively correspond to the Niño3 and the Trans-Niño
487 indices (e.g., Santoso et al., 2017), results in an inverted ‘V-shape’ (Fig. 13.4a), underscoring
488 the nonlinearity in ENSO patterns and magnitude. The combination between the two principal
489 components depicts an ENSO diversity which encapsulates the tendency for strong El Niño
490 events to peak in the eastern Pacific, and for moderate El Niño and strong La Niña events to
491 peak in the central Pacific (Takahashi et al., 2011; Dommenges et al., 2013; Capotondi et al.,
492 2015; see also Fig. 3a,b in Chapter 4). While the quasi-continuous distribution of SST
493 anomalies in the PC1-PC2 space means that an ENSO event can peak anywhere across the
494 equatorial Pacific, it also reveals two regimes of Eastern Pacific (EP) El Niño and Central
495 Pacific (CP) El Niño, the former of which tends to be a stronger event (see Chapter 4).

496 Yeh et al. (2009) found that the frequency of CP El Niño would be greater than that of EP El
497 Niño under greenhouse warming based on the premise that the weakened Trade winds would

498 shoal the thermocline in the central Pacific which would in turn promote SST variability there.
499 On the other hand, the reduced wind-driven upwelling should weaken the thermocline feedback,
500 which governs EP El Niño. The result was based on a small sample of CMIP3 models that
501 simulate the relative frequency of EP and CP El Niño closest to observations. There is no
502 robust change in a larger sample of CMIP5 models (Taschetto et al., 2014), and the observed
503 relative frequency itself can be affected by internal variability (Yeh et al., 2011). However,
504 Power et al. (2013) showed that while CMIP5 models do a poor job in simulating the spatial
505 pattern of CP El Niños, the four models best able to simulate them tend to exhibit an increase
506 in the frequency of CP El Niños in response to global warming under RCP8.5. Nevertheless,
507 the fact the SST biases are still larger than the SST amplitude of a typical CP ENSO event,
508 suggests only low confidence in these results.

509 The degree of the nonlinear relationship between PC1 and PC2 can be quantified by fitting a
510 quadratic function in the PC1-PC2 space: $PC2(t) = \alpha[PC1(t)]^2 + \beta PC1(t) + \gamma$. Not all climate
511 models can simulate the observed nonlinearity α , or equivalently the degree of ENSO diversity
512 as seen in observations (Ham & Kug, 2012; Karamperidou et al., 2017). Models that capture
513 the degree of observed α parameter tend to simulate a more realistic amplitude of the Bjerknes
514 feedback components and shortwave damping in the eastern Pacific (Karamperidou et al.,
515 2017), and these correspond to projection of eastern Pacific enhanced warming. The more
516 realistic models should also be able to simulate shifts in atmospheric convection following the
517 peak of El Niño warm surface anomaly, from the western to eastern Pacific as compared to the
518 less realistic models which have convection more statically confined in the west due to the
519 more severe cold-tongue bias (Ham & Kug, 2012). Examining CMIP5 models that have more
520 realistic α parameter, Cai et al. (2018) showed that models with greater amplitude of α simulate
521 greater amplitude of positive skewness in the Eastern Pacific and negative skewness in the
522 Central Pacific, i.e., better simulating the nonlinear processes responsible for the skewness.
523 Further, models producing greater amplitude of α simulate more distinct centres of CP and EP
524 ENSO.

525 The core of SST anomalies associated with EP ENSO varies in location across the models,
526 especially when the excluded models (based on the fidelity of the α parameter; Fig. 13.4d) are
527 included in the ensemble. This reduces the robustness of projected ENSO amplitude when
528 considering SST variability averaged over a region that is fixed across models (e.g., Niño3
529 index; see section 13.3). Using models that produce the two distinct anomaly centres, Cai et

530 al. (2018) found a statistically 15% increase in SST variability associated with EP ENSO under
531 greenhouse warming scenario (Fig. 13.4b, c, e). The enhanced variability translates to about
532 45% increase in the occurrences of strong EP El Niño. The cause for the increase is attributed
533 to greenhouse gas-induced increased vertical stratification that enhances the coupling between
534 wind and ocean which supports the Bjerknes feedback, as originally proposed by Timmermann
535 et al. (1999).

536

537 **13.5 Uncertainties and model biases**

538 Errors and biases in models can influence the characteristics of the modelled ENSO and its
539 projected response under climate change. Most CMIP models simulate an excessive westward
540 extension of cold tongue and insufficient equatorial western Pacific precipitation (e.g. Mechoso
541 et al., 1995; Li et al., 2015), which translates into an unrealistic extension of ENSO-related
542 SST anomalies to the tropical western Pacific and westward shift of ENSO-related zonal wind
543 stress and rainfall variability (Ham & Kug, 2014; 2015). In addition, most models still struggle
544 to accurately represent feedback processes that control ENSO evolution. This includes a
545 weaker than observed thermo-dynamical damping (e.g. Bellenger et al., 2014) as well as an
546 underestimated positive ENSO feedbacks, which can also somewhat be connected to these
547 aforementioned Pacific mean-state biases (e.g. Kim et al., 2014b). These compensating errors
548 may still lead to realistic ENSO characteristics but with incorrect underlying ENSO dynamics
549 (Bayr et al., 2018) and unrealistic sensitivities with respect to radiative perturbations. This fact
550 may partly explain the diverse changes in ENSO SST variability across models (Kim et al.,
551 2014b).

552 Several studies have attempted to identify sources of inter-model uncertainty in ENSO
553 amplitude change, which appear to be linked with uncertainty in the climatological mean state.
554 This can be done by examining inter-model relationship between changes in ENSO and various
555 ocean-atmosphere parameters (e.g., Ham & Kug, 2016; Rashid et al., 2016; Chen et al., 2017).
556 Some of the findings include a link between ENSO amplitude change with the climatological
557 location of the convergence zones in present-day simulation, which controls the air-sea
558 coupling strength change and the amplitude of ENSO variability (Ham & Kug, 2016). Chen
559 et al. (2017) found that the CMIP5 inter-model divergence in the ENSO amplitude change is
560 closely tied to the spread in the thermocline feedback changes which are in turn linked to

561 changes in the mean equatorial upwelling and thermocline. Further, there is a statistically
562 significant inter-model correlation between the change in ENSO amplitude and the relative
563 SST warming pattern: models that project stronger ENSO amplitude tend to project stronger
564 warming in the eastern Pacific (Zheng et al., 2016). Based on this inter-model relationship and
565 to the extent that correcting present-day climatological biases, at least in a statistical sense,
566 favours an eastern Pacific enhanced warming (Li et al., 2016; Ying et al., 2018), an increase in
567 ENSO-related SST variance was likely under global warming (Zheng et al., 2016). This
568 appears to support the overarching conclusion of more extreme ENSO events presented in
569 section 13.4 based on an ensemble of selected models. There are however several other factors
570 that remain to be investigated, which may affect such statistic-based corrections. For instance,
571 Kohyama et al. (2018) suggested, based on a comparison of two climate models, that the more
572 diffused thermocline in models compared to observations may instead render a weaker ENSO
573 amplitude under greenhouse warming.

574 Several strategies have been proposed to alleviate the influence of model errors and biases on
575 tropical Pacific future projections. A simple way is to perform a selection of models based on
576 their ability to best simulate ENSO dynamics and characteristics (section 13.4). Another way
577 is to dynamically correct the present-day model mean state. For instance, the impact of present-
578 day CMIP biases on rainfall projections can be reduced by forcing an atmosphere-only model
579 with bias-corrected SSTs (e.g. Knutson et al., 2008), i.e. to add the global warming CMIP SST
580 pattern to the present-day observed mean state. Several authors have used this strategy to assess
581 the sensitivity of the tropical Pacific rainfall response to CMIP mean state biases (Power et al.,
582 2013; Widlansky et al., 2013; Chung et al., 2014; Dutheil et al., 2018). These studies
583 demonstrate that the increase in equatorial Pacific rainfall variability in response to climate
584 change is similar to that found in the CMIP ensemble projection but that the reduction in the
585 cold-tongue bias further enhances the eastward shift in the location of main ENSO-related
586 anomalies in convection.

587 It must however be mentioned that, while this bias correction approach is likely to improve the
588 reliability of future projection, atmosphere-only present-day simulations still exhibit significant
589 biases in their representation of the atmospheric response to ENSO (e.g., Zhang and Sun 2014;
590 Ferret et al., 2017; Tang and Yu, 2018), which may still impair the reliability of the projected
591 changes. Another dynamical approach is to use a flux correction strategy directly in coupled
592 models to reduce present-day mean-state bias. By applying such a strategy, Cai et al. (2014)

593 suggested that the projected frequency increase in extreme El Niño events derived from CMIP
594 model analyses should be even larger when correcting for model biases. However, minimizing
595 mean-state biases through flux adjustment may not necessarily remove important biases in the
596 dominant ENSO feedbacks and uncertainties in ENSO projections (Neelin & Dijkstra, 1995;
597 Ferret & Collins; 2016).

598 Given ENSO is influenced by variability outside the tropical Pacific Ocean (Chapter 11), model
599 biases in remote oceans would also be an important factor for ENSO simulation and projection.
600 For instance, biases in Indian Ocean mean state and variability can affect the amplitude and
601 frequency of simulated ENSO (Yu et al., 2009; Santos et al., 2012). Further, recent studies
602 have pointed out the underestimation of Pacific Trade Winds acceleration during the early 21st
603 Century global warming hiatus period by climate models (e.g., Kociuba & Power, 2015)
604 appears to be linked to biases in the Atlantic (Kajtar et al., 2018; McGregor et al., 2018), and
605 underestimation of inter-basin relative warming (Luo et al., 2017). How this may impact on
606 ENSO simulation remains to be investigated, and so does the ramifications for future
607 projections of ENSO, in particular considering the potential that pan-tropical inter-basin
608 interactions could be underestimated in climate models (Cai et al., 2019).

609

610 **13.6. Summary and concluding remarks**

611 From the discussions presented in this chapter based on various existing studies, there appears
612 to be three factors that are key toward an understanding of how ENSO would change under
613 global warming: 1) the tropical Pacific mean-state change; 2) changes in non-linear processes
614 that interact with the mean climate; and 3) climate model fidelity. Most climate models project
615 equatorially enhanced surface warming in the Pacific Ocean, with a weakening of the Walker
616 Circulation (section 13.2; Fig. 13.1). This was found to lead to robust enhancement of ENSO-
617 driven rainfall variability along the equatorial Pacific (section 13.4.1; Fig. 13.2), even though
618 models do not tend to agree on changes in ENSO as measured using traditional SST metrics
619 (section 13.3; Fig. 13.2).

620 On the other hand, considering only models that are able to simulate key nonlinear processes,
621 such as rainfall response to eastern equatorial Pacific SSTs, eastward propagation of ENSO
622 SST anomalies, large anomalous central Pacific cooling during a La Niña, it has led to a

623 projection of increased frequency of extreme El Niño and extreme La Niña events (section
624 13.4). These models also project that extreme swing of extreme El Niño to extreme La Niña
625 in the following year – a rare catastrophic sequence that occurred in 1997-1999 – would be
626 more than double in frequency under increasing greenhouse-gas concentration in the
627 atmosphere (Fig. 13.3). Finally, accounting for model ability to simulate ENSO flavors has led
628 to a projection of enhanced Eastern Pacific ENSO variability (Fig. 13.4), stemming from more
629 frequent occurrences of strong El Niño events (section 13.4.3). All these taken together point
630 to a possibility of increased activity of extreme ENSO events in a warming climate.
631 Nonetheless, there are still uncertainties surrounding these projections given persistent model
632 biases (section 13.5). Even the best models are still not free of errors and biases.

633 Another influential factor to consider is internal variability. ENSO potentially exhibits highly
634 diverse behavior in the absence of external forcing, as demonstrated by millennial-long
635 simulations (e.g., Wittenberg, 2009; Borlace et al., 2013). Multi-model averaging may not
636 entirely remove internal variability (e.g., Frankcombe et al., 2015), but requiring large
637 ensembles with individual models (e.g., Maher et al., 2018). Studies based on a climate model
638 with a large ensemble has shown that the range of internal variability in ENSO amplitude is
639 substantial and comparable to the projection uncertainty based on the CMIP5 multi models
640 (Zheng et al., 2018). The way forward to study future ENSO changes is to utilize large
641 ensemble members in a multi-model environment to account for both differences in model
642 physics and natural variability. In addition, the impact of the background warming on ENSO
643 changes requires further investigation in this context (e.g., An & Choi, 2015; Zheng et al.,
644 2018); particularly since the interaction between ENSO and the mean climate, as well as the
645 nature of decadal climate variability, are still subjects of active investigation (Chapter 8). How
646 the annual cycle could change under global warming is also still an open question, which is
647 highly relevant for understanding future ENSO changes (see Chapter 21 for discussions).

648 Nonetheless, despite the projection uncertainties, recent studies based on models, observations,
649 and paleo-reconstructions have suggested that the Pacific climate and ENSO variability might
650 already have been altered by anthropogenic forcing. For example, Wang et al. (2015)
651 concluded that the warming since 1960 observed in the western tropical Pacific and in high
652 quality surface temperature records from small island states in the west Pacific can be
653 reproduced by climate models only if they include anthropogenic forcing. More recently,
654 Power et al. (2017) showed that the overwhelming majority of CMIP5 models exhibit an

655 increase in the frequency of disruption to tropical Pacific precipitation that ENSO causes (Fig.
656 13.5). The model results imply that the ENSO-driven rainfall disruption experienced in the
657 real world in the late 20th Century could be due to increases in anthropogenic greenhouse
658 forcing. This is consistent with paleoclimatic studies that have shown that ENSO-driven
659 variability seen in paleo records had already increased in the late 20th Century (McGregor et
660 al., 2013; Cobb et al., 2013; see Chapters 5, 21). The results of Power et al. (2017) further
661 suggested that the risks would be elevated in the future, should the rate of global warming
662 continues. In addition, the risk would be locked in for at least the rest of the 21st Century even
663 if global warming is limited below 2°C under the Paris Agreement. This last result is consistent
664 with Wang et al. (2017) who found that the increased frequency of extreme El Niño would
665 continue for about a century upon stabilization of global warming at 1.5°C, although the risk
666 associated with extreme La Niña would be averted. Future changes in ENSO behavior are
667 expected to impact on global rainfall, tropical cyclones, marine ecosystems, fisheries and the
668 global carbon cycle (Chapters 16-20) via atmospheric and oceanic teleconnections (Chapters
669 14, 15).

670 To sum up, latest research has suggested that the response of ENSO to greenhouse forcing may
671 manifest in more frequent stronger ENSO events. Given the large impact of ENSO on society,
672 economy, and the environment, this result has important ramifications for risk management in
673 a warming world. There are uncertainties associated with the projections due to model biases.
674 However, preliminary approaches in bias correction and/or consideration as discussed above
675 have indicated a possibility that the risk could have been underestimated. While we wait for
676 improved models for more reliable projections, it is also important to sustain and enhance
677 observations and paleo-reconstruction capabilities.

678

679 **Acknowledgments**

680 We thank two anonymous reviewers for their constructive feedback. W.C. is supported by
681 National Key R&D Program of China (2018YFA0605700). L.W. is supported by National
682 Natural Science Foundation of China (NSFC). W.C., G.W., and A.S. are supported by
683 CSHOR and the Earth System and Climate Change Hub of the Australian Government's
684 National Environment Science Program (NESP). CSHOR is a joint research Centre for

685 Southern Hemisphere Oceans Research between QNLM and CSIRO. S. P. is supported by
686 NESP.

687

688

689 **References**

690 An, S. I. (2011). Atmospheric responses of Gill-type and Lindzen–Nigam models to global warming.
691 *Journal of Climate*, 24(23), 6165–6173.

692 An, S.-I., & Choi, J. (2015). Why the twenty-first century tropical Pacific trend pattern cannot
693 significantly influence ENSO amplitude? *Climate Dynamics*, 44, 133–146.

694 An, S.-I. & Jin, F.-F. (2004). Nonlinearity and asymmetry of ENSO. *Journal of Climate*, 17, 2399–2412.

695 An, S.-I., Kim, J.-W., Im, S.-H., Kim, B.-M., & Park, J.-H. (2012). Recent and future sea surface
696 temperature trends in tropical Pacific warm pool and cold tongue regions. *Climate Dynamics*, 39, 1373–
697 1383.

698 Bayr, T., Latif, M., Dommenges, D., Wengel, C., Harlaß, J., & Park, W. (2018). Mean-state dependence
699 of ENSO atmospheric feedbacks in climate models. *Climate Dynamics*, 50(9–10), 3171–3194.

700 Bellenger, H., Guilyardi, E., Leloup, J. Lengaigne, M. and Vialard, J. (2014). ENSO representation in
701 climate models: from CMIP3 to CMIP5. *Climate Dynamics*, 42, 1999–2018.

702 Bonfils, C. J., Santer, B. D., Phillips, T. J., Marvel, K., Leung, L. R., Doutriaux, C., & Capotondi, A.
703 (2015). Relative contributions of mean-state shifts and ENSO-driven variability to precipitation changes
704 in a warming climate. *Journal of Climate*, 28(24), 9997–10013.

705 Borlace, S., Cai, W., & Santoso, A. (2013). Multidecadal ENSO amplitude variability in a 1000-yr
706 simulation of a coupled global climate model: Implications for observed ENSO variability. *Journal of*
707 *Climate*, 26(23), 9399–9407. <https://doi.org/10.1175/JCU-D-13-00281.1>

708 Cai, W., Borlace, S., Lengaigne, M., van Rensch, P., Collins, M., Vecchi, G., et al. (2014). Increasing
709 frequency of extreme El Niño events due to greenhouse warming. *Nature Climate Change*, 4(2), 111–
710 116. <https://doi.org/10.1038/nclimate2100>

711 Cai, W., Wang, G., Santoso, A., McPhaden, M. J., Wu, L., Jin, F.-F. et al. (2015a). Increased frequency
712 of extreme La Niña events under greenhouse warming. *Nature Climate Change*, 5(2), 132–137.
713 <https://doi.org/10.1038/nclimate2492>

714 Cai, W., Santoso, A., Wang, G., Yeh, S.-W., An, S.-I., Cobb, K. M., et al. (2015b). ENSO and greenhouse
715 warming. *Nature Climate Change*, 5(9), 849–859.

716 Cai, W., Wang, G., Santoso, A., Lin, X., & Wu, L. (2017). Definition of extreme El Niño and its impact
717 on projected increase in extreme El Niño frequency. *Geophysical Research Letters*, 44,
718 doi:10.1002/2017GL075635.

719 Cai, W., et al. (2018). Increased variability of eastern Pacific El Niño under greenhouse
720 warming. *Nature*, 564, 201–206.

721 Cai, W., Wu, L., et al. (2019). Pantropical climate interactions. *Science*, 363, doi:

722 10.1126/science.aav4236.

723 Cane, M. A., Clement, A. C., Kaplan, A., Kushnir, Y., Pozdnyakov, D., Seager, R., et al. (1997).
724 Twentieth-century sea surface temperature trends. *Science*, 275, 957-960.

725 Capotondi, A., Wittenberg, A. T., Newman, M., Di Lorenzo, E., Yu, J.-Y., Braconnot, P., Dewitte, B.
726 (2015). Understanding ENSO diversity. *Bulletin of the American Meteorological Society*, 96(6), 921–
727 938. <https://doi.org/10.1175/BAMS-D-13-00117.1>

728 Chadwick, R., Boutle, I., Martin, G. (2012). Spatial patterns of precipitation change in CMIP5: Why
729 the rich don't get richer in the Tropics. *Journal of Climate*, doi:10.1175/JCLI-D-12-00543.1.

730 Chen, L., Li, T., Yu, Y., et al. (2017). A possible explanation for the divergent projection of ENSO
731 amplitude change under global warming. *Climate Dynamics*, 49, 3799-3811.

732 Chung, C. T. Y., Power, S. B., Arblaster, J. M., Rashid, H. A. & Roff, G. L. (2014). Nonlinear
733 precipitation response to El Niño and global warming in the Indo-Pacific. *Climate Dynamics*, 42, 1837–
734 1856 (2014).

735 Chung, C. T. Y. & Power, S. B. (2014). Precipitation response to La Niña and global warming in the
736 Indo-Pacific. *Climate Dynamics*, 43, 3293–3307.

737 Clapeyron, M. C. (1834). "Mémoire sur la puissance motrice de la chaleur". *Journal de l'École*
738 *Polytechnique* (in French). 23, 153–190.

739 Clausius, R. (1850). "Ueber die bewegende Kraft der Wärme und die Gesetze, welche sich daraus für
740 die Wärmelehre selbst ableiten lassen" *Annalen der Physik*, 155(4): 500–524.

741 Clement, A. C., Seager, R., Cane, M. A., & Zebiak, S. E., (1996). An ocean dynamical thermostat.
742 *Journal of Climate*, 9, 2190-2196.

743 Coats, S., & Karnauskas, K. B. (2017). Are simulated and observed twentieth century tropical pacific
744 sea surface temperature trends significant relative to internal variability? *Geophysical Research Letters*,
745 44(19), 9928-9937.

746 Cobb, K. M. et al. (2013). Highly Variable El Niño-Southern Oscillation Throughout the Holocene.
747 *Science*, 339, 67-70, doi:10.1126/science.1228246.

748 Collins, M. (2000a). The El-Niño Southern Oscillation in the second Hadley Centre coupled model and
749 its response to greenhouse warming. *Journal of Climate*, 13, 1299-1312.

750 Collins, M. (2000b). Understanding uncertainties in the response of ENSO to greenhouse warming.
751 *Geophysical Research Letters*, 27(21), 3509-3512.

752 Collins, M., et al. (2010). The impact of global warming on the tropical Pacific Ocean and El Niño.
753 *Nature Geoscience*, 3, 391-397.

754 Collins, M., et al. (2011). Climate model errors, feedbacks and forcings: a comparison of perturbed
755 physics and multi-model ensembles. *Climate Dynamics*, 36, 1737-1766.

756 Deser, C., Phillips, A.S., & Alexander, M. A. (2010). Twentieth century tropical sea surface temperature
757 trends revisited. *Geophysical Research Letters*, 37, L10701.

758 DiNezio, P. N. et al. Climate response of the equatorial Pacific to global warming. *Journal of Climate*,
759 22, 4873–4892 (2009).

760 Dommenges, D., Bayr, T., & Frauen, C. (2013). Analysis of the non-linearity in the pattern and time
761 evolution of El Niño Southern Oscillation. *Climate Dynamics*, 40(11-12), 2825–2847.
762 <https://doi.org/10.1007/s00382-012-1475-0>

763 Dutheil, C., et al. (2018). Impact of surface temperature biases on climate change projections of the
764 South Pacific Convergence Zone. *Climate Dynamics*, in press.

765 England, M., et al. (2014). Recent intensification of wind-driven circulation in the Pacific and the
766 ongoing warming hiatus. *Nature Climate Change*, 4, 222–227, doi:10.1038/nclimate2106.

767 Fedorov, A. V., & Philander, S. G. (2000). Is El Niño changing? *Science*, 288(5473), 1997–2002.
768 <https://doi.org/10.1126/science.288.5473.1997>

769 Ferrett, S., Collins, M., & Ren, H. L. (2017). Understanding bias in the evaporative damping of El
770 Niño–Southern Oscillation events in CMIP5 models. *Journal of Climate*, 30(16), 6351–6370.

771 Ferrett, S., & Collins, M. (2016). ENSO feedbacks and their relationships with the mean state in a flux
772 adjusted ensemble. *Climate Dynamics*, 1–20.

773 Frankcombe, L. M., England, M. H., Mann, M. E., & Steinman, B. A. (2015). Separating internal
774 variability from the externally forced climate response. *Journal of Climate*, 28(20), 8184–
775 8202. doi:10.1175/JCLI-D-15-0069.1

776 Graham, N. E., & Barnett, T. P. (1987). Sea surface temperature, surface wind divergence, and
777 convection over tropical oceans. *Science*, 238, 657–659.

778 Guilyardi, E. (2006). El Niño mean state seasonal cycle interactions in a multi-model ensemble. *Climate
779 Dynamics*, 26, 329–348.

780 Ham, Y.-G., & Kug, J.-S. (2012). How well do current climate models simulate two types of El Niño?
781 *Climate Dynamics*, 39, 383–398.

782 Ham, Y.-G., & Kug, J.-S. (2014). ENSO phase-locking to the boreal winter in CMIP3 and CMIP5
783 models. *Climate Dyn.*, 43, 305–318, doi:<https://doi.org/10.1007/s00382-014-2064-1>.

784 Ham, Y.-G., & Kug, J.-S. (2015). Improvement of ENSO simulation based on intermodel diversity,
785 *Journal of Climate*, 28, 998–1015.

786 Ham Y.-G., & Kug, J.-S. (2016). ENSO amplitude changes due to greenhouse warming in CMIP5: Role
787 of mean tropical precipitation in the twentieth century. *Geophysical Research Letters*, 43,
788 doi: [10.1002/2015gl066864](https://doi.org/10.1002/2015gl066864)

789 Held, I. M., & Soden, B. J. (2006). Robust response of the hydrological cycle to global warming.
790 *Journal of Climate*, 19, 5686–5699.

791 Hoerling, M. P., Kumar, A., & Zhong, M. (1997). El Niño, La Niña, and the nonlinearity of their
792 teleconnections. *Journal of Climate*, 10, 1769–1786, [https://doi.org/10.1175/1520-0442\(1997\)010<1769:ENOLNA>2.0.CO;2](https://doi.org/10.1175/1520-0442(1997)010<1769:ENOLNA>2.0.CO;2)

794 Huang, P., & Xie, S. P. (2015). Mechanisms of change in ENSO-induced tropical Pacific rainfall

795 variability in a warming climate. *Nature Geoscience*, 8(12), 922.

796 Huang, P. (2016). Time-varying response of ENSO-induced tropical Pacific rainfall to global warming
797 in CMIP5 models. Part I: Multimodel ensemble results. *Journal of Climate*, 29(16), 5763-5778.

798 Huang, P., & Chen, D. (2017). Enlarged asymmetry of tropical Pacific rainfall anomalies induced by El
799 Niño and La Niña under global warming. *Journal of Climate*, 30(4), 1327-1343.

800 Jin, F.-F., Kim, S.-T., Bejarano, L. (2006). A coupled-stability index for ENSO. *Geophysical Research*
801 *Letters*, doi.org/10.1029/2006GL027221.

802 Jin, F.-F., & Neelin, J. D. (1993). Modes of interannual tropical ocean-atmosphere interaction-a unified
803 view. Part I: Numerical results. *Journal of the Atmospheric Sciences*, 50(21), 3477–3503.
804 [https://doi.org/10.1175/1520-0469\(1993\)050%3C3477:MOITOI%3E2.0.CO;2](https://doi.org/10.1175/1520-0469(1993)050%3C3477:MOITOI%3E2.0.CO;2)

805 Johnson, N., & Xie, S.-P. (2010). Changes in the sea surface temperature threshold for tropical
806 convection. *Nature Geoscience*, 3, 842-845.

807 Kajtar, J. B., Santoso, A., McGregor, S., & England, M. H. (2018). Model under-representation of
808 decadal Pacific trade wind trends and its link to tropical Atlantic bias. *Climate Dynamics*, 50, 1471-
809 1484.

810 Kang, I. - S., & Kug, J. - S. (2002). El Niño and La Niña sea surface temperature anomalies:
811 Asymmetry characteristics associated with their wind stress anomalies, *J. Geophys. Res.*, 107(D19),
812 4372, doi:[10.1029/2001JD000393](https://doi.org/10.1029/2001JD000393).

813 Karamperidou, C., Jin, F.-F. & Conroy, J. L. (2017). The importance of ENSO nonlinearities in tropical
814 Pacific response to external forcing. *Climate Dynamics*, 49, 2695–2704.

815 Karnauskas, K. B., Seager, R., Kaplan, A., Kushnir, Y. & Cane, M. A. (2009). Observed strengthening
816 of the zonal sea surface temperature gradient across the equatorial Pacific Ocean. *Journal of Climate*
817 22, 4316–4321.

818 Kim, S.-T., & Jin, F.-F. (2011). An ENSO stability analysis. Part II: results from the twentieth and
819 twenty-first century simulations of the CMIP3 models. *Climate Dynamics*, 36, 1609-1627.

820 Kim, S.-T., Cai, W., Jin, F.-F., Santoso, A., Wu, L., Guilyardi, E., & An, S.-I. (2014a). Response of El
821 Niño sea surface temperature variability to greenhouse warming. *Nature Climate Change*, 4, 786-790.

822 Kim, S.-T., W. Cai, F.-F. Jin, and J.-Y. Yu (2014b). ENSO stability in coupled climate models and its
823 association with mean state. *Climate Dynamics*, 42, 3313–3321, doi:<https://doi.org/10.1007>

824 Knutson, T. R., & Manabe, S. (1995). Time-mean response over the tropical Pacific to increased CO2
825 in a coupled ocean-atmosphere model. *Journal of Climate*, 8, 2181-2199.

826 Knutson, T. R., Manabe, S. & Gu, D. (1997). Simulated ENSO in a global coupled ocean-atmosphere
827 model: multidecadal amplitude modulation and CO2 sensitivity. *Journal of Climate*, 10, 138-161.

828 Knutson, T.R. Sirutis, J.J. Garner, S.T. Vecchi, G.A. and Held, I.M. (2008). Simulated reduction in
829 Atlantic hurricane frequency under twenty-first-century warming conditions. *Nature Geoscience*, 1,
830 359–364.

831 Kociuba, G. & Power, S. B. (2015). Inability of CMIP5 models to simulate recent strengthening of the
832 Walker Circulation: implications for projections. *Journal of Climate*, 28, 20-35, doi: 10.1175/JCLI-D-
833 13-00752.1.

834 Kohyama, T., Hartmann, D.L. & Battisti, D. S. (2017). La Niña-like Mean-State Response to Global
835 Warming and Potential Oceanic Roles. *Journal of Climate*, 30, 4207–
836 4225, <https://doi.org/10.1175/JCLI-D-16-0441.1>

837 Kohyama, T., Hartmann, D. L., & Battisti, D. S. (2018). Weakening of nonlinear ENSO under global
838 warming. *Geophysical Research Letters*, 45(16), 8557-8567.

839 Kosaka, Y. & Xie, S. P. (2013). Recent global-warming hiatus tied to equatorial Pacific surface cooling.
840 *Nature*, 50158 (7467), 403-407, doi:10.1038/nature12534.

841 Kug, J. S., Sooraj, K. P., Jin, F. F., Ham, Y. G., & Kim, D. (2011). A possible mechanism for El Niño-
842 like warming in response to the future greenhouse warming. *International Journal of Climatology*,
843 31(10), 1567-1572.

844 Latif, M., & Keenlyside, N. (2009). El Niño/Southern Oscillation response to global warming.
845 *Proceedings of the National Academy of Sciences*, 106, 20578-20583.

846 Lau, W. K. M., & Kim, K.-M. (2015). Robust Hadley Circulation changes and increasing global dryness
847 due to CO2 warming from CMIP5 model projections. *PNAS*, 112(12), 3630-3635.

848 Lee, T., & McPhaden, M. J. (2010). Increasing intensity of El Niño in the central-equatorial Pacific.
849 *Geophysical Research Letters*, 37, L14603. <https://doi.org/10.1029/2010GL044007>

850 L’Heureux, M. L., S. Lee, and B. Lyon, 2013: Recent multidecadal strengthening of the Walker
851 circulation across the tropical Pacific. *Nature Climate Change*, 3, 571–576,
852 doi:<https://doi.org/10.1038/nclimate1840>.

853 Li, G. Du, Y. Xu, H. and Ren, B. (2015). An Intermodel Approach to Identify the Source of Excessive
854 Equatorial Pacific Cold Tongue in CMIP5 Models and Uncertainty in Observational Datasets. *Journal*
855 *of Climate*, 28, 7630–7640.

856 Li, G., Xie, S.-P., Du., Y., & Luo Y. (2016). Effects of excessive equatorial cold tongue bias on the
857 projections of tropical Pacific climate change. Part I: the warming pattern in CMIP5 multi-model
858 ensemble. *Climate Dynamics*, 47, 3817-3831.

859 Liu, Z. (1998). On the role of ocean in the response of tropical climatology to global warming: The
860 west-east SST contrast. *Journal of Climate*, 11, 864-875.

861 Liu, Z., Vavrus, S., He, F., Wen, N., & Zhong, Y., (2005). Rethinking tropical ocean response to global
862 warming: The enhanced equatorial warming. *Journal of Climate*, 18, 4684-4700.

863 Liu, F., Luo, Y., Lu, J., & Wan, X. (2017). Response of the tropical Pacific Ocean to El Niño versus
864 global warming. *Climate Dynamics*, 48, 935-956.

865 Lu, J., Vecchi, G. A., & Reichler, T. (2007). Expansion of the Hadley cell under global
866 warming, *Geophysical Research Letters*, 34, L06805, doi:10.1029/2006GL028443.

867 McCreary, J. P., & Lu, P. (1994). Interaction between the subtropical and equatorial ocean circulations:

868 The subtropical cell. *Journal of Physical Oceanography*, 24, 466–497.

869 Luo, J.J., Wang, G., & Dommenges, D. (2018). May common model biases reduce CMIP5's ability to
870 simulate the recent Pacific La Niña-like cooling? *Climate Dynamics*, 50, 1335.
871 <https://doi.org/10.1007/s00382-017-3688-8>

872 Maher, N., Matei, D., Milinski, S., & Marotzke, J. (2018). ENSO change in climate projections: Forced
873 response or internal variability? *Geophysical Research
874 Letters*, 45, 11,390– 11,398. <https://doi.org/10.1029/2018GL079764>

875 McGregor, S., Timmermann, A., England, M. H., Timm, O. E., & Wittenberg, A. T. (2013). Inferred
876 changes in El Niño-Southern Oscillation variance over the past six centuries. *Climate of the Past*, 9,
877 2269-2284, doi:10.5194/cp-9-2269-2013.

878 McGregor, S., Spence, P., Schwarzkopf, F. U., England, M. H., Santoso, A., Kessler, W., Timmermann,
879 A., Boening, C. (2014). ENSO driven interhemispheric Pacific mass transports. *Journal of Geophysical
880 Research (Oceans)*, 119, 6221-6237.

881 McGregor, S., et al. (2018). Model tropical Atlantic biases underpin diminished Pacific decadal
882 variability. *Nature Climate Change*, 8, doi:10.1038/s41558-018-0163-4.

883 McPhaden, M. J., & Zhang, X. (2009). Asymmetry in zonal phase propagation of ENSO sea surface
884 temperature anomalies. *Geophysical Research Letters*, 36, L13703,
885 <http://dx.doi.org/10.1029/2009GL038774>.

886 Mechoso, C., et al. (1995). The Seasonal Cycle over the Tropical Pacific in Coupled Ocean–Atmosphere
887 General Circulation Models. *Monthly Weather Review*, 123, 2825–2838.

888 Meehl, G. A., & Washington, W. M. (1996). El Niño-like climate change in a model with increased
889 atmospheric CO₂ concentrations. *Nature*, 382,56-60.

890 Meehl, G. A., Branstator, G.W. & Washington, W. M. (1993). Tropical Pacific interannual variability
891 and CO₂ climate change. *Journal of Climate*, 6, 42-63.

892 Merryfield, W. J. (2006). Changes to ENSO under CO₂ doubling in a multimodel ensemble. *Journal of
893 Climate*, 19, 4009–4027.

894 Neelin, D. J., et al. (1998). ENSO theory. *Journal of Geophysical Research*, 103, 14261–14290.

895 Neelin, J. D. & Dijkstra, H.A. (1995). Ocean-Atmosphere Interaction and the Tropical Climatology.
896 Part I: The Dangers of Flux Correction. *Journal of Climate*, 8, 1325–1342,
897 [https://doi.org/10.1175/1520-0442\(1995\)008<1325:OAIATT>2.0.CO;2](https://doi.org/10.1175/1520-0442(1995)008<1325:OAIATT>2.0.CO;2)

898 Newman, M., Alexander, M. A., Ault, T. R., Cobb, K. M., Deser, C., DiLorenzo, E., et al. (2016). The
899 Pacific Decadal Oscillation, revisited. *Journal of Climate*, 29(12), 4399–4427.
900 <https://doi.org/10.1175/JCLI-D-15-0508.1>

901 Philip, S. Y., & van Oldenborgh, G. J. (2006). Shifts in ENSO coupling processes under global warming.
902 *Geophysical Research Letters*, 33, L11704. doi: 10.1029/2006GL026196.

903 Philip, S. Y., & van Oldenborgh, G. J. (2009). Significant atmospheric nonlinearities in the ENSO cycle.
904 *Journal of Climate*, 22, 4014-4028.

905 Power, S. B., & Delage, F. P. (2018). El Niño–Southern Oscillation and Associated Climatic Conditions
906 around the World during the Latter Half of the Twenty-First Century. *Journal of Climate*, 31, 6189–
907 6207, <https://doi.org/10.1175/JCLI-D-18-0138.1>

908 Power, S. B., & Kociuba, G. (2011a). What caused the observed twentieth century weakening of the
909 Walker Circulation? *Journal of Climate*, 24, 6501-6514.

910 Power, S. B., & Kociuba, G. (2011b). Impact of global warming on the SOI. *Climate Dynamics*, 37,
911 1745-1754, doi: 10.1007/s00382-010-0951-7.

912 Power, S. B., & Colman, R. (2006). Multi-year predictability in a coupled general circulation model.
913 *Climate Dynamics*, 26, 247-272.

914 Power, S. B., Delage, F., Chung, C., Kociuba, G., & Keay, K. (2013). Robust twenty-first century
915 projections of El Niño and related precipitation variability, *Nature*, 502, 541-545,
916 doi:10.1038/nature12580.

917 Power, S., Folland, C., Colman, A., & Mehta, V. (1999). Inter-decadal modulation of the impact of
918 ENSO on Australia. *Climate Dynamics*, 15, 319-324.

919 Power, S., Haylock, M., Colman, R., & Wang, X. (2006). The predictability of interdecadal changes in
920 ENSO and ENSO teleconnections. *Journal of Climate*, 8, 2161-2180.

921 Power, S. B., et al. (2017). Humans have already increased the risk of major disruptions to Pacific
922 rainfall. *Nature Communications*, 8, 14368.

923 Rashid, H. A., Hirst, A. C., & Marsland, S. J. (2016). An atmospheric mechanism for ENSO amplitude
924 changes under an abrupt quadrupling of CO2 concentration in CMIP5 models. *Geophysical Research*
925 *Letters*, 43, doi: 10.1002/2015gl066768.

926 Rodgers, K.B., Friederichs, P., & Latif, M. (2004). Tropical Pacific decadal variability and its relation
927 to decadal modulations of ENSO. *J. Climate*, 17, 3761–3774, [https://doi.org/10.1175/1520-0442\(2004\)017<3761:TPDVAI>2.0.CO;2](https://doi.org/10.1175/1520-0442(2004)017<3761:TPDVAI>2.0.CO;2)

929 Santoso A., England, M. H., & Cai, W. (2012). Impact of Indo-Pacific feedback interactions on ENSO
930 dynamics diagnosed using ensemble climate simulations. *Journal of Climate*, 25, 7743-7763.

931 Santoso, A., McGregor, S., Jin, F. F., Cai, W., England, M. H., An, S. I., McPhaden, M., & Guilyardi,
932 E. (2013). Late-twentieth-century emergence of the El Niño propagation asymmetry and future
933 projections. *Nature*, 504(7478), 126.

934 Santoso, A., McPhaden, M. J., & Cai, W. (2017). The defining characteristics of ENSO extremes and
935 the strong 2015/16 El Niño. *Reviews of Geophysics*, doi:10.1002/2017RG000560.

936 Schneider, E. K. (1977). Axially symmetric steady-state models of the basic state for instability and
937 climate studies, Part II. Nonlinear calculations, *Journal of Atmospheric Science*, 34, 280–296.

938 Schott, F. A., J. P. McCreary, & G. C. Johnson (2004). Shallow overturning circulations of the tropical–
939 subtropical oceans. *Earth's Climate: The Ocean–Atmosphere Interaction, Geophysical Monograph*, Vol.
940 147, Amer. Geophys. Union, 261–304.

941 Seager, R., & Murtugudde, R. (1997). Ocean dynamics, thermocline adjustment and regulation of

942 tropical SST. *Journal of Climate*, 10, 521–534.

943 Seo, K.-H., D. M. W. Frierson, and J.-H. Son (2014). A mechanism for future changes in Hadley
944 circulation strength in CMIP5 climate change simulations. *Geophysical Research Letters*, 40, 5251–
945 5258, doi:10.1002/2014GL060868.

946 Solomon, A., & Newman, M. (2012). Reconciling disparate twentieth-century Indo-Pacific ocean
947 temperature trends in the instrumental record. *Nature Climate Change*, 2, 691-699.

948 Stein, K., Schneider, N., Timmermann, A., & Jin, F.-F. (2010). Seasonal synchronization of ENSO
949 events in a linear stochastic model. *Journal of Climate*, 23, 5629-5643.

950 Stevenson, S. L. (2012). Significant changes to ENSO strength and impacts in the twenty-first century:
951 Results from CMIP5. *Geophysical Research Letters*, 39, L17703. [doi:10.1029/2012GL052759](https://doi.org/10.1029/2012GL052759)

952 Sun, D.-Z., Zhang, T., Sun, Y., & Yu, Y. (2014). Rectification of El Niño–Southern Oscillation into
953 climate anomalies of decadal and longer time scales: Results from forced ocean GCM experiments.
954 *Journal of Climate*, 27(7), 2545–2561. <https://doi.org/10.1175/JCLI13-00390.1>

955 Takahashi, K., Montecinos, A., Goubanova, K., & Dewitte, B. (2011). ENSO regimes: Reinterpreting
956 the canonical and Modoki El Niño. *Geophysical Research Letters*, 38, L10704.
957 <https://doi.org/10.1029/2011GL047364>

958 Takahashi, K., & Dewitte, B. (2016). Strong and moderate nonlinear El Niño regimes. *Climate*
959 *Dynamics*, 46, 1627-1645.

960 Tang, S. L., & Yu, Y. Q. (2018). Evaluation of the zonal wind stress response to SST in the CMIP5
961 AMIP simulations. *Atmospheric and Oceanic Science Letters*, 11(2), 157-164.

962 Taschetto, A. S., Sen Gupta, A., Jourdain, N., Santoso, A., Ummenhofer, C. C., & England, M. H.
963 (2014). Cold tongue and warm pool ENSO events in CMIP5: mean state and future
964 projections. *Journal of Climate*, 27, 2861-2885.

965 Tett, S. (1995). Simulation of El Niño–Southern Oscillation-like variability in a global AOGCM and its
966 response to CO2 increase. *Journal of Climate*, 8, 1473-1502.

967 Timmermann, A. (2003). Decadal ENSO amplitude modulations: A nonlinear paradigm. *Global and*
968 *Planetary Change*, 37, 135-156.

969 Timmermann, A., Oberhuber, J., Bacher, A., Esch, M., Latif, M., & Roeckner, E. (1999). Increased E1-
970 Niño frequency in a climate model forced by future greenhouse warming, *Nature*, 398, 694- 696.

971 Timmermann, A. (2001). Changes of ENSO stability due to greenhouse warming. *Geophysical*
972 *Research Letters*, 28, 2016-2064.

973 Timmermann, A., et al. (2018). El Niño–Southern Oscillation complexity. *Nature*, 559, 535-545.

974 Tokinaga, H., Xie, S.-P., Deser, C., Kosaka, Y., & Okumura Y. M. (2012a). Slowdown of the Walker
975 circulation driven by tropical Indo-Pacific warming. *Nature*, 491, 439-443.

976 Tokinaga, H., Xie, S.-P., Timmermann, A., McGregor, S., Ogata T., et al. (2012b). Regional patterns of
977 tropical Indo-Pacific climate change: Evidence of the Walker circulation weakening. *Journal of Climate*,

978 25, 1689-1710.

979 van Oldenborgh, G. J., Philip, S. & Collins, M. (2005). El Niño in a changing climate: a multi-model
980 study. *Ocean Sci.*, 2, 267–298.

981 Vecchi, G. A. et al. (2006). Weakening of tropical Pacific atmospheric circulation due to anthropogenic
982 forcing. *Nature*, 441, 73–76.

983 Vecchi, G. A. & Soden, B. J. (2007). Global warming and the weakening of the tropical circulation. *J.*
984 *Climate* 20, 4316–4340.

985 Walker, C. C., & Schneider, T. (2006). Eddy-influences on Hadley circulations: Simulations with an
986 idealized GCM, *Journal of Atmospheric Science*, 63, 3333–3350.

987 Wallace, J.M., et al. (1998). On the structure and evolution of ENSO-related climate variability in the
988 tropical Pacific: lessons from TOGA. *Journal of Geophysical Research*, 103, 14241–14259.

989 Wang, B., & An, S.-I. (2001). Why the properties of El Niño changed during the late 1970s.
990 *Geophysical Research Letters*, 28, 3709-3712.

991 Wang, B., & An, S. (2002). A mechanism for decadal changes of ENSO behavior: Roles of background
992 wind changes. *Climate Dynamics*, 18(6), 475–486. <https://doi.org/10.1007/s00382-001-0189-5>

993 Wang, G., Power, S. B., & McGree, S. (2015). Unambiguous warming in the western tropical Pacific
994 primarily caused by anthropogenic forcing. *International Journal of Climatology*, 36, 933-944, doi:
995 10.1002/joc.4395.

996 Wang, G., Cai, W., Gan, B., Wu, L., Santoso, A., Lin, X., Chen, Z., & McPhaden, M. (2017). Continued
997 increase of extreme El Niño frequency long after 1.5C warming stabilization. *Nature Climate Change*,
998 7, 568-572.

999 Watanabe, M., et al. (2012). Uncertainty in the ENSO amplitude change from the past to the future.
1000 *Geophysical Research Letters*, 39, L20703.

1001 Watanabe, M., Kamae, Y., & Kimoto, M. (2014). Robust increase of the equatorial Pacific rainfall and
1002 its variability in a warmed climate. *Geophysical Research Letters*, 41(9), 3227-3232.

1003 Widlansky, M. J., Timmermann, A., Stein, K., McGregor, S., Schneider, N., England, M.H., Lengaigne,
1004 M., & Cai, W. (2013). Changes in South Pacific rainfall bands in a warming climate. *Nature Climate*
1005 *Change*, 3, 417–423.

1006 Wittenberg, A. T. (2009). Are historical records sufficient to constrain ENSO simulations? *Geophysical*
1007 *Research Letters*, 36, L12702. <https://doi.org/10.1029/2009GL038710>

1008 Xie, S.-P., et al. (2010). Global warming pattern formation: Sea surface temperature and rainfall.
1009 *Journal of Climate*, 23, 966-986.

1010 Yeh, S.-W., & Kirtman, B. (2007). ENSO amplitude changes due to climate change projections in
1011 different coupled models. *Journal of Climate*, 20, 203-217.

1012 Yeh, S.-W., et al. (2009). El Niño in a changing climate, *Nature*, 461, 511–514,
1013 doi:10.1038/nature08316

1014 Yeh, S.-W., Kirtman, B. P., Kug, J.-S., Park, W., & Latif, M. (2011). Natural variability of the central
1015 Pacific El Niño event on multi-centennial timescales. *Geophysical Research Letters*, 38, L02704,
1016 doi:10.1029/2010GL045886.

1017 Ying, J., Huang, P., Lian, T., & Tan, H. (2018). Understanding the effect of an excessive cold tongue
1018 bias on projecting the tropical Pacific SST warming pattern in CMIP5 models. *Climate Dynamics*,
1019 <https://doi.org/10.1007/s00382-018-4219-y>

1020 Yu, J.-Y., Sun, F., & Kao, H.-Y. (2009). Contributions of Indian Ocean and monsoon biases to the
1021 excessive biennial ENSO in CCSM3. *Journal of Climate*, 22, 1850–1858.

1022 Zhang, M. & Song, H. Evidence of deceleration of atmospheric vertical overturning circulation over
1023 the tropical Pacific. *Geophysical Research Letters*, 33, L12701 (2006).

1024 Zhang, T., & Sun, D. Z. (2014). ENSO asymmetry in CMIP5 models. *Journal of Climate*, 27(11), 4070-
1025 4093.

1026 Zheng, X. T., Xie, S. P., Lv, L. H., & Zhou, Z. Q. (2016). Intermodel uncertainty in ENSO amplitude
1027 change tied to Pacific Ocean warming pattern. *Journal of Climate*, 29(20), 7265-7279.

1028 Zheng, X. - T., Hui, C., & Yeh, S. - W. (2018). Response of ENSO amplitude to global warming in
1029 CESM large ensemble: Uncertainty due to internal variability. *Climate*
1030 *Dynamics*, 50, 4019. doi:10.1007/s00382-017-3859-7

1031 Zhou, Z.-Q., S.-P. Xie, X.-T. Zheng, Q. Liu, and H. Wang, 2014: Global warming–induced changes in
1032 El Niño teleconnections over the North Pacific and North America. *Journal of Climate*, 27, 9050–9064,
1033 <https://doi.org/10.1175/JCLI-D-14-00254.1>.

1034

1035

1036

1037

1038

1039

1040

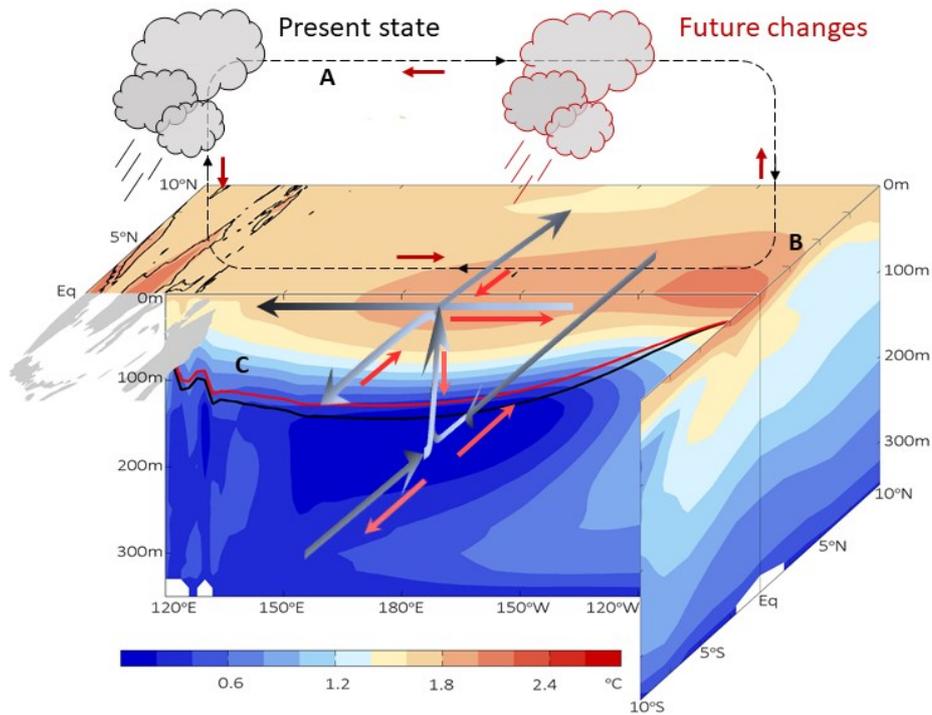
1041

1042

1043

1044

1045



1046

1047 **Figure 13.1:** Schematic of tropical Pacific mean-state changes due to greenhouse forcing. Red
 1048 arrows and red outlined clouds indicate greenhouse-induced changes. Present-day mean states are
 1049 featured in black or gray (e.g., gray arrows indicate mean ocean circulations). Color shading
 1050 denotes temperature change from present to future. (A) The Walker Circulation (dashed arrow)
 1051 slows down, resulting in weaker westward flowing Trade Winds and ocean currents. (B) The
 1052 equatorial region warms faster than off equator, with the eastern region and Maritime Continent
 1053 warming faster than in the central Pacific. Atmospheric convection shifts towards eastern
 1054 equatorial Pacific due to the reduced meridional and zonal temperature gradients. (C) The vertical
 1055 ocean temperature gradient increases in response to increased radiative forcing at the surface,
 1056 leading to shoaled thermocline. Present-day thermocline is indicated by the black curve in the ocean
 1057 interior which shoals eastward, shallowing in the future as indicated by the red curve. Adapted
 1058 from Cai et al. (2015b).

1059

1060

1061

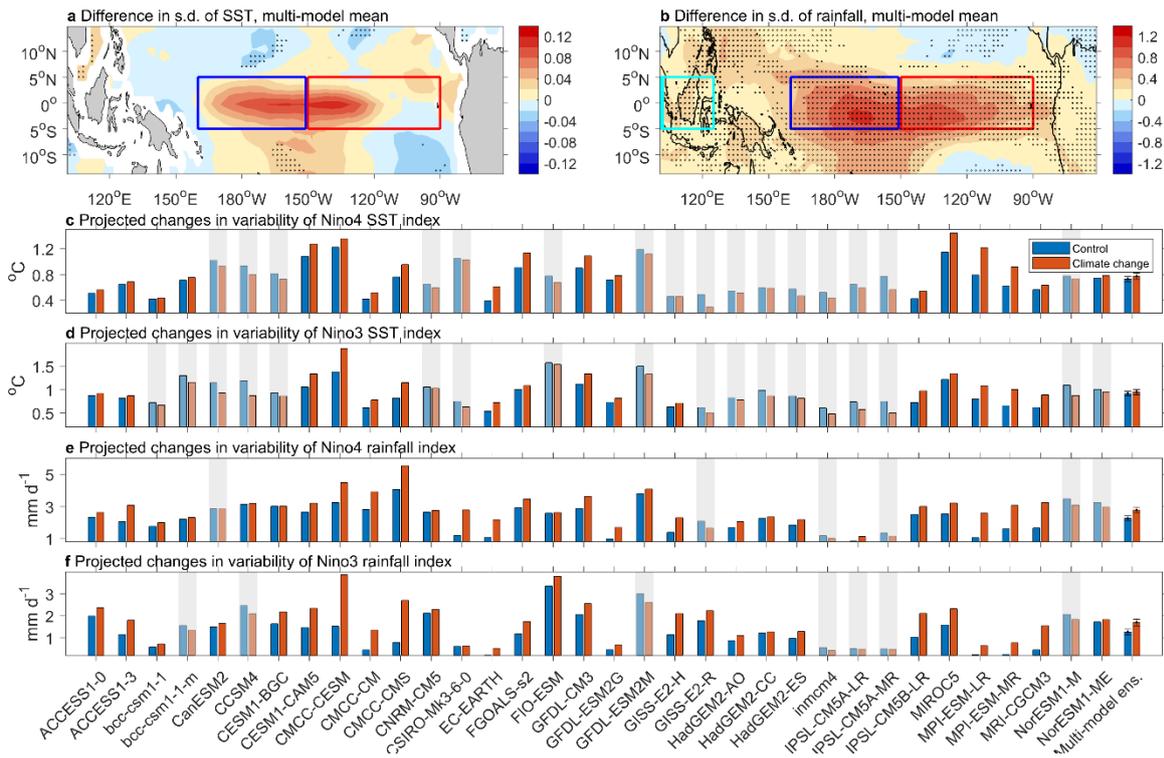
1062

1063

1064

1065

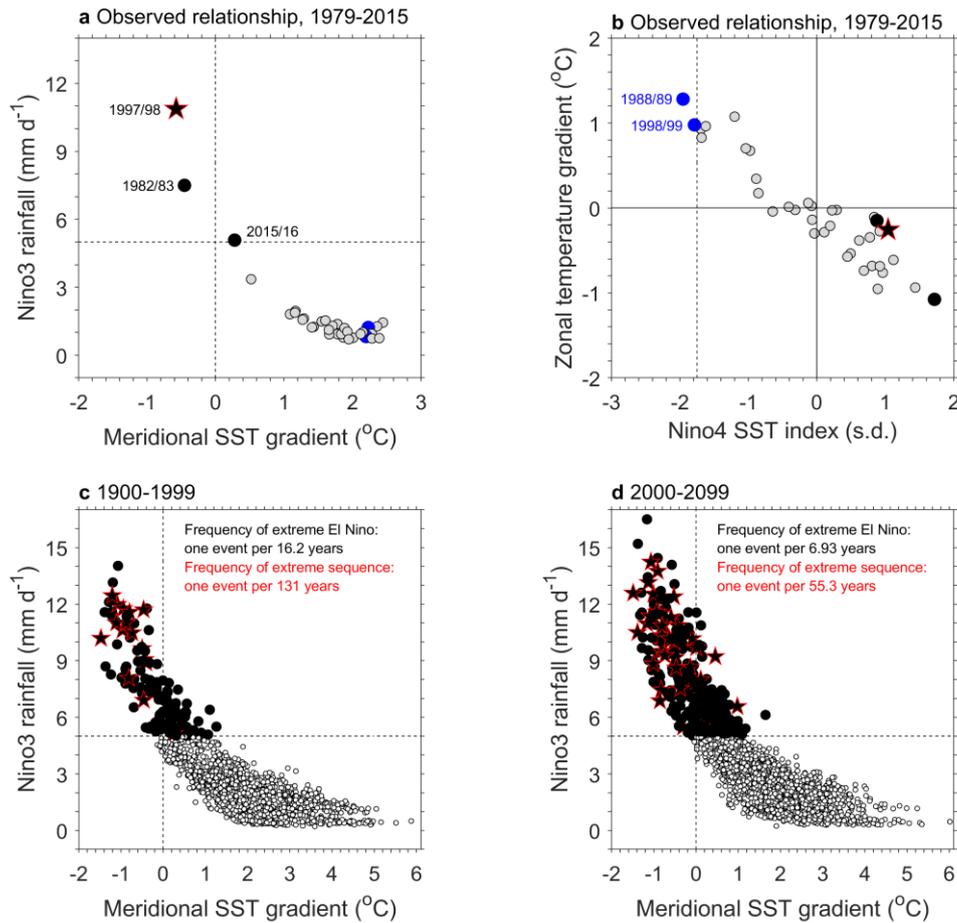
1066



1067

1068 **Figure 13.2:** Greenhouse-warming induced change in sea surface temperature and precipitation
 1069 variability over tropical Pacific. Results shown are based on CMIP5 outputs under historical and
 1070 RCP8.5 emission scenario using 34 models, focusing on boreal winter (December-February
 1071 average). **a**, Multi-model-mean change in SST standard deviation between future (2000-2099) and
 1072 present-day (1900-1999). Niño4 (160°E-150°W, 5°S-5°N) and Niño3 (150°W-90°W, 5°S-5°N)
 1073 regions are indicated by blue box and red box, respectively. Dotted areas mean that more than 70%
 1074 of models (>~24 models) generate a change in phase with multi-model-mean value. **b**, As in **a**,
 1075 but for rainfall (mm per day). Maritime Continent region (100°E -125°E, 5°S-5°N) used to construct
 1076 zonal temperature gradient in Fig. 13.3b is indicated by cyan box. **c** and **d**, Projected changes in
 1077 SST variability over Niño4 and Niño3 regions, respectively. Models that simulate a reduction are
 1078 grayed out. Multi-model ensemble is also shown in both panels with error bars corresponding to
 1079 the 95% confidence interval based on a bootstrap method. **e** and **f**, As in **c** and **d**, but for rainfall.

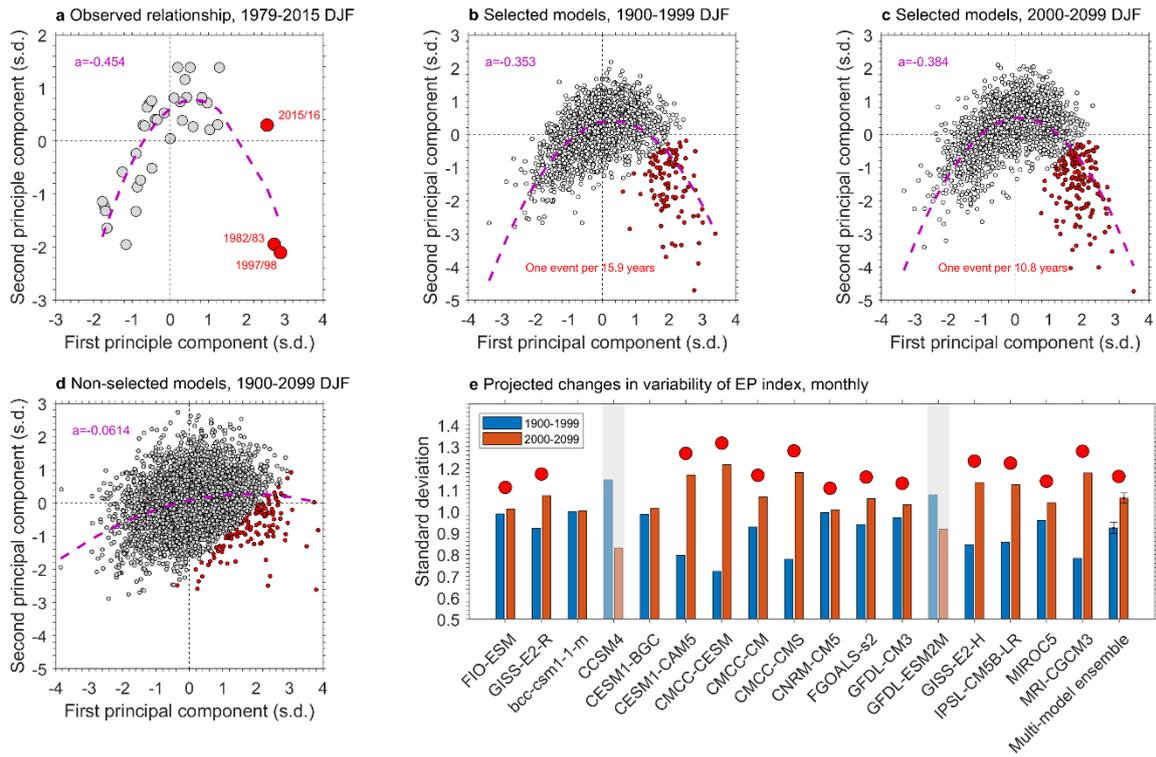
1080



1081

1082 **Figure 13.3:** Observed climate extremes and greenhouse-warming induced changes. **a**, Observed
 1083 relationship between Niño3 rainfall and meridional SST gradient (Cai et al 2014; 150°W - 90°W ,
 1084 5°N - 10°N minus 150°W - 90°W , 2.5°S - 2.5°N). Black dots indicate extreme El Niño events, defined
 1085 as when Niño3 rainfall is greater than 5 mm per day (marked by the horizontal line). Red star
 1086 marks the 1997/98 event which was an extreme El Niño with concurrent eastward-propagating SST
 1087 anomalies (Santoso et al., 2013) that was also followed by an extreme La Niña the following year
 1088 (Cai et al., 2015b). **b**, Observed relationship between Niño4 SST and surface air temperature
 1089 gradient between the Maritime Continent region (Fig. 13.2b) and Niño4 (Cai et al., 2015a). Blue
 1090 dots indicate extreme La Niña events, defined as when Niño4 SST is negative and greater than 1.75
 1091 s.d. in amplitude. **c** and **d**, As in **a**, but using 21 selected CMIP5 models which can produce extreme
 1092 El Niño events as in Cai et al. (2015b), under present-day (1900-1999) and future (2000-2099)
 1093 respectively. Gray dots indicate events that are not extreme El Niño. As indicated within each panel,
 1094 extreme El Niño occurs more frequently in future climate, from approximately one event per 16
 1095 years to one event per 7 years. A rare sequence of extreme events like in 1997-1998 is also
 1096 projected to occur more often, from approximately one event per 131 years to one in 55 years.

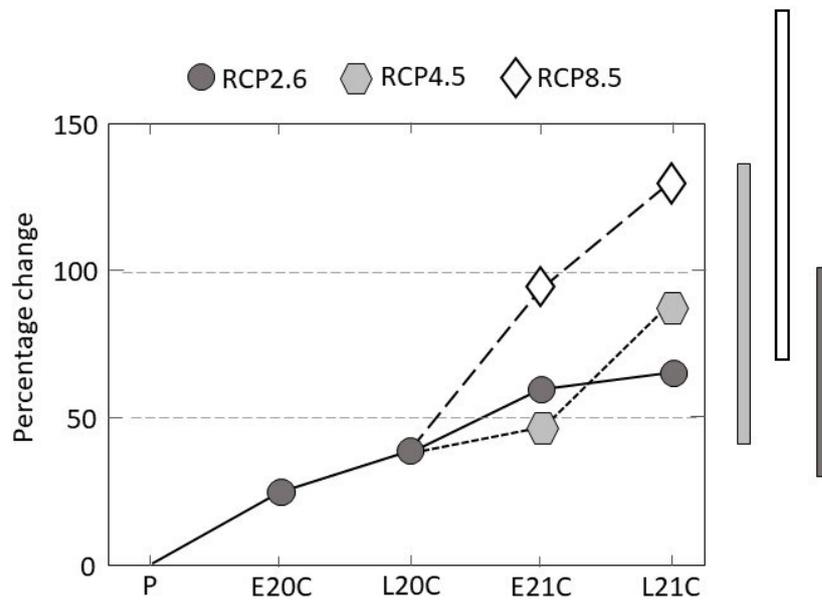
1097



1098

1099

1100 **Figure 13.4:** Eastern Pacific El Niño events defined by SST anomalies, and their projected changes
 1101 under future climate. **a**, Observed relationship between the first two principal components using
 1102 SST anomalies averaged over December-February ENSO peak phase. The EOF analysis for
 1103 observations is applied to monthly SST anomalies referenced to their long-term monthly mean,
 1104 over an equatorial domain (15°S – 15°N , 140°E – 80°W) to de-convolve spatio-temporal variability
 1105 into orthogonal modes, each described by a principal spatial pattern and an associated principal
 1106 component (PC) time series. The PC time series is scaled to have a standard deviation of one. The
 1107 nonlinear relationship is determined by a quadratic fitting between PC1 and PC2: $PC2(t) = \alpha$
 1108 $PC1(t)^2 + \beta PC1(t) + \gamma$ (Cai et al., 2018; Dommenget et al., 2013), as indicated by the purple curve.
 1109 Red dots indicate strong EP El Niño events, defined by an E-index = $(PC1(t) - PC2(t))/\sqrt{2}$
 1110 (Takahashi et al., 2011) and when it is greater than 1.5 s.d. **b** and **c**, As in **a**, but using 17 models
 1111 which are able to simulate at least half of the observed degree of nonlinearity measured by the
 1112 parameter α (a in the panels), under present-day (1900-1999) and future (2000-2099), respectively
 1113 (Cai et al., 2018). The EOF analysis for models is applied to monthly SST anomalies (from 1900
 1114 to 2099 forced with historical anthropogenic and natural forcings, and future greenhouse gases
 1115 under the RCP8.5 scenario) which were firstly referenced to the monthly climatology of the first
 1116 100 years and then quadratically de-trended over the whole 200 years. Strong Eastern Pacific El
 1117 Niño events are projected to occur more frequently under future climate, from approx. one in 16
 1118 years to one in 11 years. **d**, As in **a**, but using the remaining 17 models with weaker nonlinearity
 1119 which is not sufficient to separate EP and CP El Niño. **e**, Projected changes in variability of the E-
 1120 index using the 17 selected models. Blue and red bars indicate standard deviation of the E-index
 1121 under present-day and future climate, respectively. Models simulating a decreased variability in E-
 1122 index is greyed out. 15 out of 17 models produce an increased variability, in contrast to the poor
 1123 inter-model agreement on the changes in Niño3 SST variability (Collins et al., 2010). Models with
 1124 a red dot are able to project more occurrence of strong EP El Niño events. Increased E-index
 1125 variability is associated with more occurrence of strong EP El Niño events in most models.



1126
 1127 **Figure 13.5:** Percentage change in the frequency of major disruptions to precipitation over the tr
 1128 opical Pacific (140°E-240°E, 25°S-15°N) in the twentieth and twenty-first centuries, relative to th
 1129 e corresponding pre-industrial value of approximately (1.0/9.0 yr⁻¹). Results for the early 20th cent
 1130 ury (E20C), late twentieth century (L20C), early 21st century (E21C) and late 21st century (L21C)
 1131 are shown. The results are based on changes obtained from twenty CMIP5 models forced using pr
 1132 e-industrial control, historical forcing and three warming scenarios for the 21st century: RCP2.6 (d
 1133 ark grey circle), RCP4.5 (light grey hexagon) and RCP8.5 (diamond). The lines and markers indic
 1134 ate percentage changes in the frequency of major disruption, relative to the pre-industrial value. T
 1135 he result is significant at the 90% level, under the assumption that models are independent. Adapt
 1136 ed from Power et al. (2017).

1137
 1138
 1139
 1140
 1141
 1142
 1143