

1 Future changes to El Niño-Southern Oscillation temperature and precipitation teleconnections

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8

9 Key Points

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11 • The majority of CMIP5 models project robust increases in the spatial extent of ENSO
12 temperature and precipitation teleconnections over land.

13

14 • The increase in area is related to the amplified ENSO-driven precipitation across the
15 equatorial Pacific in the future.

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17 • Despite the robust increase in area over land, we do not find a consistent
18 strengthening of these teleconnections in the individual models.

19

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21

22 Abstract

23

24 Potential changes to the El Niño-Southern Oscillation (ENSO) resulting from climate change
25 may have far reaching impacts through atmospheric teleconnections. Here, ENSO
26 temperature and precipitation teleconnections between the historical and high-emission future
27 simulations are compared in 40 models from phase 5 of the Coupled Model Intercomparison
28 Project (CMIP5). Focusing on the global land area only, we show that there are robust
29 increases in the spatial extent of ENSO teleconnections during austral summer in 2040-2089
30 of ~19% for temperature and ~12% for precipitation in the multi-model mean (MMM),
31 relative to the 1950-1999 period. The MMM further shows the expansion of ENSO
32 teleconnection extent is at least partly related to a strengthening of ENSO teleconnections
33 over continental regions, however, a consistent strengthening is not found across the
34 individual models. This suggests that while more land may be affected by ENSO, the existing
35 teleconnections may not be simply strengthened.

36

37 Introduction

38

39 The El Niño-Southern Oscillation (ENSO) is the largest source of interannual climate
40 variability, contributing to substantial changes in rainfall, temperature, and extreme weather
41 around the globe. Although there are robust projections of climate change impacts for
42 changes in the long-term average temperature and precipitation for the next century (*IPCC*,
43 2014), there remains uncertainty around how the characteristics of ENSO and its
44 teleconnections may change in the future (*Collins et al.*, 2010), despite models showing an

45 increase in the frequency of extreme ENSO events (*Cai et al.*, 2015). The uncertainty arises
46 from the different representations of the coupled ocean and atmosphere feedback processes
47 that control ENSO, with the CMIP5 models showing no clear consensus on whether ENSO
48 sea surface temperature (SST) variability will increase or decrease (*Watanabe et al.*, 2012;
49 *Guilyardi et al.*, 2009; *Stevenson et al.*, 2012).

50

51 Despite the lack of agreement in projected changes to ENSO SST variability, recent studies
52 have shown that there is a consistent projected strengthening of the atmospheric response to
53 ENSO across the equatorial Pacific. *Power et al.* (2013) show that for ENSO SST anomalies
54 (SSTAs) of the same structure and magnitude, anomalous convection is greater in the 21st
55 century compared to the 20th century due to background warming in the equatorial Pacific.
56 They further show that the nonlinear contribution to intensified precipitation from
57 background warming can be enhanced or damped by projected changes in ENSO variance
58 and structural changes to ENSO SSTAs, which have less intermodel agreement. *Cai et al.*
59 (2014) also show that the faster rates of warming projected for the equatorial and eastern
60 Pacific reduces the zonal and meridional gradients of SST, increasing the frequency of deep
61 convection anomalies including the southward shift of the intertropical convergence zone
62 which signifies extreme El Niño events.

63

64 The enhanced thermodynamic response in the central and eastern Pacific is the result of the
65 interaction between the warmer and moister atmosphere (*Huang and Xie*, 2015; *Held and*
66 *Soden*, 2006) and the total SST arising from ENSO SSTAs superimposed upon the mean
67 warming of the equatorial Pacific (*Johnson and Xie*, 2010), forming conditions that are more
68 conducive to deep convection and contributing to an eastward shift of the convective heating
69 anomaly (*Power et al.*, 2013; *Cai et al.*, 2014). As the latent heat release from anomalous

70 deep convection is the driving mechanism of atmospheric teleconnections that cause the
71 remote impacts of ENSO (*Hoskins and Karoly, 1981; Chiang and Sobel, 2002; Trenberth et*
72 *al., 1999*), it follows that increases in ENSO-driven convection in the tropical Pacific may
73 cause changes in ENSO teleconnections to remote regions (*Watanabe et al., 2014*).

74

75 Recent model studies have begun to provide insight into potential changes to ENSO
76 teleconnections. Using atmosphere-only models, *Zhou et al. (2014)* show that the eastward
77 shift of deep convection during El Niño combined with mean warming drives a strengthened
78 Pacific North America (PNA) teleconnection pattern, intensifying the ENSO-driven
79 precipitation variability across North America. This is also consistent with other studies
80 which suggest geographical changes in ENSO teleconnections, the magnitude of which
81 depend on both external forcing and ENSO amplitude changes (e.g. *Stevenson, 2012; Kug et*
82 *al. 2010; Meehl and Teng, 2007*). For example, in a global study, *Bonfils et al. (2015)* project
83 the observed pattern of ENSO variability onto CMIP5 model output from the historical and
84 RCP8.5 simulations to obtain a “typical” ENSO SST pattern and associated precipitation
85 teleconnections for the current and future climates. They find intensified ENSO-driven
86 precipitation in the future simulations, and show that this increase in precipitation response is
87 modulated by the projected change in the amplitude of ENSO variability in each model
88 (*Bonfils et al., 2015*). To date, there has been limited research investigating future changes in
89 temperature teleconnections, or examining in detail changes over land where ENSO events
90 have the greatest societal impacts. Here we expand upon the analysis of *Power et al. (2013)*
91 by examining a larger ensemble of CMIP5 models, and broadening the focus from the
92 equatorial Pacific Ocean to identify the changes in ENSO-driven variability that occur over
93 remote land areas.

94

95 Methods

96

97 *2.1 Data*

98

99 We analyse monthly precipitation and surface temperature output from 40 CMIP5 models
100 (see *Taylor et al.*, 2012; Supplementary Table 1). We compare the ENSO teleconnections
101 simulated in different climate conditions over two fifty-year periods: between 1950-1999 in
102 the historical simulations, and between 2040-2089 in the high-emissions RCP8.5 simulations,
103 where anthropogenic radiative forcing reaches 8.5 W m^2 by 2100. For models that have
104 archived multiple historical and RCP8.5 simulations, we use output from the first ensemble
105 member only. The data processing and identification of the ENSO signal follows *Power et al.*
106 (2013). Surface temperature and precipitation data are interpolated to a 1.5° latitude by 1.5°
107 longitude grid and a high-pass spectral filter is applied to remove variability with a period
108 greater than 13 years including the climatological mean, global warming signal and any
109 multidecadal variability. Subsequent analysis focuses on the austral summer season,
110 December-February (DJF), when ENSO variability peaks. For comparison with observed
111 ENSO teleconnections, we analyse observations of temperature from the NCEP/NCAR
112 reanalysis (*Kalnay et al.*, 1996) for a comparable fifty-year period between 1953-2002, and
113 precipitation from the CMAP dataset (*Xie and Arkin*, 1997) between 1984-2010, applying the
114 same temporal filtering outlined above. The first and last five years of the datasets are
115 discarded to minimise edge effects introduced by the spectral filter.

116

117 *2.2 Identifying ENSO and ENSO teleconnections*

118

119 Following *Power et al.* (2013), ENSO patterns for models and observations observations
120 were calculated using an Empirical Orthogonal Function (EOF) analysis of the equatorial
121 ocean domain (30°S-30°N, 0-360°E). The first EOF of both the equatorial ocean SST and
122 precipitation represents the ENSO mode, with its associated principle component (PC)
123 illustrating the temporal evolution of El Niño and La Niña events. Separate EOFs were
124 calculated for each model for each 50-year time period, and for both SST and precipitation, to
125 identify if structural changes in ENSO or ENSO's teleconnections are apparent. The first PCs
126 for temperature and precipitation are highly correlated for each model with an ensemble
127 median correlation of 0.94 (interquartile range (IQR) of 0.91-0.97) for the historical period.
128 The ENSO signal is insensitive to the zonal domain: the correlation between the first PC for
129 the equatorial ocean and the equatorial Pacific Ocean exceeds 0.98 for the majority of
130 models.

131

132 To identify global ENSO teleconnection patterns, the normalised SST and precipitation PCs
133 were linearly regressed against the respective global surface temperature or precipitation
134 filtered time series at all locations. The resulting regression maps (Figure 1) illustrates the
135 mean change in temperature or precipitation (in degrees Celsius or millimetres per day) at
136 each grid point that is associated with a one standard deviation change in the ENSO PC. The
137 strongest relationships are expected over the equatorial Pacific Ocean, the source of ENSO
138 variability, with significant ($p < 0.05$) regressions elsewhere indicative of remote ENSO
139 teleconnections. The proportion of global land area having a significant regression is
140 hereafter referred to as the spatial extent of the ENSO teleconnections. We further test where
141 there is agreement across the model ensemble in the location of significant ENSO
142 teleconnections. When at least 25 out of the 40 models show a statistically significant
143 regression with ENSO at a given location the level of intermodel agreement (62.5%) is

144 considered significant at $p < 0.1$ based on a binomial distribution (*Power et al.*, 2012),
145 assuming that each model is an independent sample (*Annan and Hargreaves*, 2017). Regions
146 that meet this requirement for intermodel agreement are hereafter referred to as having a
147 significant MMM teleconnection.

148

149 Results

150

151 *3.1 Comparison to observed ENSO teleconnections*

152

153 Previous studies have shown that the CMIP5 models are capable of simulating the dynamics
154 of ENSO with varying degrees of fidelity (e.g. *Bellenger et al.*, 2014; *Taschetto et al.*, 2014;
155 *Weare*, 2013). Comparison of the multi-model mean (MMM) teleconnection maps from the
156 historical period with the observed teleconnection maps (Figure 1), indicates that the MMM
157 captures the broad-scale observed ENSO teleconnections. The spatial correlation between the
158 MMM temperature teleconnection for the historical period and the observed teleconnection
159 for all land areas is 0.61. Significant ($p < 0.05$) observed temperature teleconnections cover
160 36.2% of the global land area (stippling in Figure 1a). This is within the range of the
161 individual CMIP5 models and is similar to the median spatial extent of 37.0% (IQR of 30.2-
162 47.0%) (Supplementary Table 1). However, only 17.4% of the global land area has a
163 significant MMM teleconnection (where at least 25 out of 40 models agree, corresponding to
164 $p < 0.1$). This area is shown by the stippling in Figure 1c, and includes parts of northern
165 South America, Central Africa, the Maritime Continent and northern Australia. Despite the
166 broad similarity of the observed and MMM teleconnection, the spatial correlation between
167 the observed and simulated temperature teleconnections over land in the individual models is
168 relatively low, with an ensemble median of 0.34 (IQR of 0.22-0.50).

169

170 The spatial correlation between the MMM precipitation teleconnection map for the historical
171 period and the observed teleconnection map for the global land area is 0.56. This is again
172 considerably higher than the spatial correlation between the observed and simulated
173 precipitation teleconnections over land for most individual models, which have a median
174 correlation of 0.38 (IQR of 0.30-0.44). The spatial extent of the observed precipitation
175 teleconnection is smaller than the temperature teleconnection, covering 23.5% of the global
176 land area (stippling in Figure 1b). This observed area of significant teleconnections over land
177 falls in the lower range of the individual model spread, where the median percentage of land
178 with a significant precipitation teleconnection is 26.4% (IQR of 18.9-32.7%) (Supplementary
179 Table 1). As with temperature, the models show limited agreement in the locations where
180 ENSO teleconnections are significant, with a significant MMM teleconnection covering only
181 3.12% of the global land area for precipitation (stippling in Figure 1d).

182

183 For both temperature and precipitation, the limited intermodel agreement in the MMM
184 teleconnections, and the low spatial correlation when compared to the observed
185 teleconnections, are influenced by: i) the differences in the individual models representation
186 of ENSO SSTAs and subsequently its teleconnections (*Bellenger et al.*, 2013; *Taschetto et*
187 *al.*, 2014; *Weare*, 2013); and ii) stochastic noise arising from the use of relatively short fifty-
188 year time periods to derive the ENSO teleconnections (*Batehup et al.*, 2015; *Wittenberg*,
189 2009). The effect of the short record is further discussed in Section 3.2.

190

191 *3.2 Global changes to ENSO teleconnections*

192

193 We now focus on how ENSO teleconnections over land are projected to change subject to
194 continued high emissions as per the RCP8.5 simulations. The spatial similarities between the
195 1950-1999 and 2040-2089 periods are clear from the MMM teleconnection maps (Figure 1c-
196 f), as reflected by the spatial correlation over land of 0.95 for both temperature and
197 precipitation. Despite the similarity of the MMM patterns, the median spatial correlation for
198 the global land area across the models between the two periods is 0.44 (IQR of 0.32-0.67) for
199 temperature, and 0.69 (IQR of 0.54-0.80) for precipitation. To examine if the projected
200 changes in the teleconnection patterns are a result of external forcing or internal variability,
201 we compare the spatial correlations of DJF ENSO teleconnection maps for an ensemble of
202 simulations from two models, HadGem2-ES and CCSM4 (Supplementary Table 2 and 3).
203 We seek to determine if the range of spatial correlations over land between the historical and
204 future simulations in these models lies outside the range of stochastic variability sampled
205 between model ensemble members during the historical period. To this end, a two-sample t-
206 test was calculated between the two distributions (historical-historical versus historical-
207 future) for each model and for temperature and precipitation. Three of the four comparisons
208 revealed correlations between the historical and future teleconnections that were not
209 statistically different ($p < 0.05$) from comparison within the historical period itself. This
210 suggests that the stochasticity of the climate system is likely to have a strong influence on the
211 low spatial correlation in individual ensemble members, and as such it will be hard to
212 separate from any forced signal.

213

214 To further examine whether changes are occurring in ENSO teleconnections over land, we
215 calculate the change in area of significant teleconnections between the two time periods.
216 Examining the models individually, we find that the land area with significant ENSO
217 temperature teleconnections increases in 28 out of 40 models (Figure 2a; Supplementary

218 Table 1). This level of model agreement is significant above the 99% level ($p < 0.01$) based
219 on a binomial distribution. For the individual models, the mean change in spatial extent is
220 10.15% (IQR of -5.68 to 26.6%) (Supplementary Table 1), relative to the 1950-1999 period,
221 and is significantly different from 0 ($p < 0.05$) using a t-test. The land area with a significant
222 MMM teleconnection (where at least 25 out of 40 models agree, corresponding to $p < 0.1$)
223 increases by approximately 19% in the 2040-2089 period, relative to the 1950-1999 period.
224 This is most prominent over equatorial Africa, South America and Australia (Figure 1c and
225 e).

226

227 For precipitation, the land area with a significant precipitation teleconnection is also
228 increasing in 27 out of 40 models (Figure 2b; Supplementary Table 1), with an equivalent
229 significance level of $p < 0.02$ based on a binomial distribution. The mean change over land in
230 the individual models, relative to the 1950-1999 period, is 5.93% (IQR of -3.70 to 14.5%)
231 (Supplementary Table 1), however this change is not significant ($p < 0.05$) using a t-test. The
232 area with a significant precipitation teleconnection in the MMM similarly shows an increase
233 in the future period, although the change is smaller than that found for temperature. The
234 increase in the land area with a significant MMM teleconnection is approximately 12%
235 relative to the 1950-1999 period, and primarily occurs over equatorial Africa and South
236 America, including an area over northern Chile that is significant in the future period only
237 (Figure 1d and f).

238

239 To determine if this change in ENSO teleconnections is seasonally dependent, we repeat our
240 analysis for the other seasons (MAM, JJA, SON), and for an annual average calculated
241 between July and June of the following year. The spatial extent of significant teleconnections
242 in the MMM and individual models differs with the seasons for both temperature and

243 precipitation (Supplementary Figure 1). However, the spatial extent of significant
244 temperature teleconnections over land for the individual models is increasing for all seasons
245 and the annual average. The level of intermodel agreement for the increase in area is
246 significant above the 90% level ($p < 0.1$) in all cases (Supplementary Figure 1). In contrast,
247 the precipitation teleconnection is found to have significant intermodel agreement ($p < 0.02$)
248 on the increase in the area of significant teleconnections over land for DJF only
249 (Supplementary Figure 1). For the annual average and all other seasons there is no significant
250 agreement across the ensemble in the change in teleconnections over land for precipitation.

251

252 *3.3 Relationship to changes in ENSO variance*

253

254 We next examine whether the projected change in teleconnection spatial extent is dependent
255 on the change in ENSO variability to determine if those models with enhanced (suppressed)
256 ENSO variability in the future simulations also have increased (decreased) ENSO
257 teleconnections. We compare the change in area of significant ENSO teleconnections to the
258 change in standard deviation of the SSTA in the Niño 3.4 region (190-240°E, 5°S-5°N),
259 calculated using high-pass filtered DJF means for each model. The change in standard
260 deviation between the two time periods was normalised by the historical standard deviation
261 for each model to illustrate the proportional change in variability relative to the models initial
262 ENSO variability. Figure 3 (a and b) shows that there is a significant ($p < 0.05$) intermodel
263 relationship between the change in ENSO variability and the change in the area of significant
264 ENSO teleconnections during DJF for both temperature and precipitation, suggesting that
265 part of the change in teleconnection area can be associated with changes in ENSO variability
266 in the future. For example, each model with enhanced ENSO variability also shows an
267 increase in the temperature teleconnection spatial extent, but not all the models with reduced

268 ENSO variability show decreases in spatial extent. This contributes to the positive offset of
269 the regression model, which suggests that models displaying no change in ENSO variability
270 will on average display an increase in the spatial extent of significant temperature
271 teleconnections over land. A similar relationship is also evident for the precipitation
272 teleconnection, however there are a few models with increased ENSO SSTA variability that
273 also show decreasing teleconnection spatial extent (Figure 3b). Thus, the positive offsets of
274 the regression models (Figure 3a and b) suggest an increase in the teleconnection area that is
275 unrelated to the changes in ENSO SSTA variability.

276

277 As previous studies have shown that the precipitation response in the tropics to unchanged
278 ENSO SSTAs is enhanced in future climate conditions (*Chung et al., 2014; Power et al.,*
279 *2013*), we further compare the change in teleconnection area to the change in the total ENSO-
280 driven precipitation over the equatorial Pacific Ocean. We define the change in ENSO-driven
281 precipitation as the difference between the cumulative sum of precipitation regression
282 coefficients over the equatorial Pacific Ocean between 120-290°E, 10°S-10°N (c.f. Figure 1d
283 and f). A significant ($p < 0.05$) intermodel relationship is found for the change in ENSO-
284 driven precipitation and teleconnection area for both temperature and precipitation (Figure 3c
285 and d). For temperature, the majority of models show increased (decreased) teleconnections
286 associated with increased (decreased) ENSO-driven precipitation, with the change in ENSO-
287 driven precipitation explaining a greater portion of the change in teleconnection area than the
288 change in ENSO SSTA variance. Further, the positive offset is reduced, suggesting that the
289 increased ENSO-driven precipitation response is at least partly responsible for the increased
290 ENSO temperature teleconnections over land. However, the result is not as clear for
291 precipitation, as although the positive offset is again slightly reduced, ENSO SSTA variance

292 explains more of the change in precipitation teleconnections than the change in precipitation
293 over the equatorial Pacific Ocean.

294

295 *3.4 Changes to the strength of the ENSO teleconnections over land*

296

297 Here we examine whether the intensity of the ENSO teleconnections over land during DJF is
298 also strengthened in the future period. Comparison of the MMM difference between the
299 historical and future teleconnection maps (Supplementary Figure 2) indicates that there are
300 regions where the strength of ENSO's teleconnections, calculated as the magnitude of the
301 regression coefficients, are projected to change. A two-sided t-test was calculated to
302 determine if the mean of the regression coefficients for the individual models at each grid
303 point is significantly different ($p < 0.05$) between the historical and future simulations
304 (stippling in Supplementary Figure 2). Based on regions which have a significant
305 teleconnection in the MMM in either period, increases in the temperature teleconnection
306 strength are evident over areas of South America, eastern and western Africa. Changes in the
307 strength of the precipitation teleconnection are evident for central and eastern Africa
308 (Supplementary Figure 2).

309

310 We quantify the magnitude of the change in teleconnection strength over land by comparing
311 the difference between the regression coefficients (2040-2089 – 1950-1999) at the locations
312 where there is a significant teleconnection in the MMM in either period (stippling in Figure
313 1c-f) against the historical period (Figure 4). The regression slope (0.15 ± 0.014) indicates
314 that the MMM temperature teleconnection at these locations is on average stronger in the
315 future period (Figure 4a). However, the relatively low R^2 of 0.25 suggests that this
316 strengthening is not a uniform intensification of the historical MMM temperature

317 teleconnection over land (cf. Supplementary Figure 3a). This is also the case for the MMM
318 precipitation teleconnection (Figure 4b), which displays a slight strengthening in the future
319 period (0.04 +/- 0.023), while the low R^2 of 0.03 indicates little similarity between the
320 teleconnection change and the historical teleconnection (cf. Supplementary Figure 3b).
321 Comparing the change in teleconnection strength for the same land area as the MMM in the
322 individual models also reveals there is no significant intermodel agreement (based on the
323 binomial distribution) on a strengthening of the teleconnections over land (Supplementary
324 Figure 4). This is at least partly influenced by noise in the climate system, which is reduced
325 in the MMM, as discussed in Section 3.2. However, we do find a clear relationship between
326 the change in each models ENSO teleconnection strength and the change in the
327 teleconnection spatial extent over land, with larger changes in teleconnection area associated
328 with strengthened teleconnections (Supplementary Figure 4).

329

330 4 Summary and Conclusion

331

332 Recent studies have shown a consistent projected strengthening of the atmospheric response
333 to ENSO SSTAs across the equatorial Pacific (*Power et al.*, 2013; *Cai et al.*, 2014), which
334 may enhance remote teleconnections due to the increase in latent heating (*Watanabe et al.*,
335 2014). Our analysis shows that with future warming under the RCP8.5 simulations, the
336 spatial extent of ENSO temperature teleconnections in the MMM (where at least 25 out of 40
337 models agree on the location of teleconnections, corresponding to $p < 0.1$) increases by
338 approximately 19% in the 2040-2089 period relative to 1950-1999, and this increase is seen
339 regardless of the season analysed. The MMM precipitation teleconnection increases in area
340 by approximately 12%, however a significant change only occurs during DJF. Although there
341 is spread across the individual ensemble members regarding the change in teleconnection

342 area, we find a significant level of agreement for an increase in area amongst the individual
343 models that is also robust when separating the models according to the teleconnection skill
344 (Supplementary Table 1). This may have implications for reconstructions of past ENSO
345 variability from remote proxies that assume ENSO teleconnections are stationary (*McGregor*
346 *et al.*, 2013, 2010; *Li et al.*, 2011, 2013).

347

348 In agreement with previous studies (*Bonfils et al.*, 2015), we show that changes in the spatial
349 extent of the ENSO teleconnections are at least partly related to the change in amplitude of
350 SSTAs in the Niño 3.4 region. However, the amplitude changes cannot entirely explain the
351 mean increase in the spatial extent of ENSO teleconnections over land. This is supported by
352 the fact that approximately half of all models displaying reduced future ENSO amplitudes
353 display an increase in the teleconnection spatial extent. We further show there is a
354 relationship between this mean change in area and ENSO-driven precipitation across the
355 equatorial Pacific, consistent with the suggestions of *Power et al.* (2013), *Cai et al.* (2014)
356 and *Watanabe et al.* (2014). It is important to note that neither predictor explains more than
357 half of the intermodel differences in the change in spatial extent, suggesting there may be
358 additional mechanisms that contribute to this change in teleconnection area that have not yet
359 been considered. Further exploration of the underlying mechanisms remains an avenue for
360 future work.

361

362 For land areas where there are significant MMM teleconnections, we show that although the
363 MMM teleconnection is strengthened in the future period this strengthening varies by
364 location for both temperature or precipitation. Further, analysis of the individual models does
365 not show a consistent strengthening of the teleconnection over land. As the strengthening of
366 the future teleconnection over land areas lacks consistency across the individual models, we

367 suggest that while the land area that is being impacted by ENSO variability is increasing, it is
368 unlikely that the future teleconnections will simply reflect an enhancement of the historical
369 teleconnections.

370

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475

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481 **Figure 1:** Regression of normalised ENSO index (defined in Section 2.2) on global surface
482 temperature and precipitation. (a) NCEP/NCAR Reanalysis temperature observations (1953-
483 2002), (b) CMAP precipitation observations (1984-2010), (c, d) MMM historical simulations
484 (1950-1999), (e, f) MMM RCP8.5 simulations (2040-2089). Stippling in a and b shows
485 significant regressions at $p < 0.05$. Stippling in c-f shows significant teleconnections in the
486 MMM (where 25 out of 40 models agree on the location of significant teleconnections,
487 corresponding to a significance level of $p < 0.1$).

488

489 **Figure 2:** Percentage of the global land area that has a significant regression with ENSO ($p <$
490 0.05) for the historical and future periods in each model for (a) temperature and (b)
491 precipitation. The relationship between the two time periods for the ensemble is given by the
492 regression equation y (solid red line) including the upper and lower 95% confidence intervals
493 for the slope (dashed red lines). Reference numbers correspond to models given in
494 Supplementary Table 1. The blue line indicates the percentage of land with a significant
495 ENSO teleconnection in the observations. The grey dashed line illustrates the 1:1 line.

496

497 **Figure 3:** The relationship between the change in teleconnection spatial extent over land
498 versus the change in ENSO variability. In (a, b) ENSO variability is the change in standard
499 deviation of the Niño3.4 region SSTAs, (c, d) the change in the total sum of the precipitation
500 regression coefficients over the tropical Pacific Ocean (120-290°E, 10°S-10°N). The
501 regression relationship for the ensemble is shown by equation y (solid red lines) including the
502 upper and lower 95% confidence intervals for the slope (dashed red lines). Reference
503 numbers correspond to models given in Supplementary Table 1.

504

505 **Figure 4:** The relationship between the change in teleconnections (future-historical MMM
506 regression coefficients) and the historical teleconnections for land areas where there is a
507 significant teleconnection in the MMM (where 25 out of 40 models agree on the location of
508 significant teleconnections, corresponding to a significance level of $p < 0.1$), for (a)
509 temperature and (b) precipitation. Each point represents a land grid point that is significant in
510 the MMM in one or both time periods. The regression relationship is given by the equation y
511 (solid red lines) including the upper and lower 95% confidence intervals on the slope (dashed
512 red lines). The grey dashed line illustrates the 1:1 line.

Figure 1.

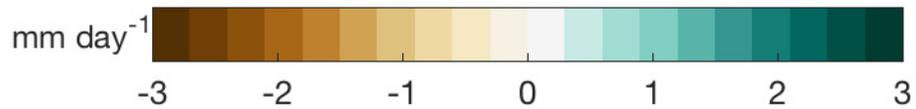
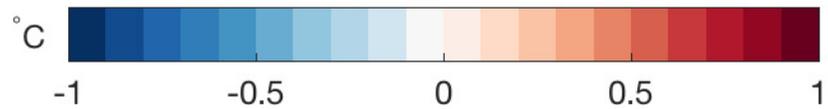
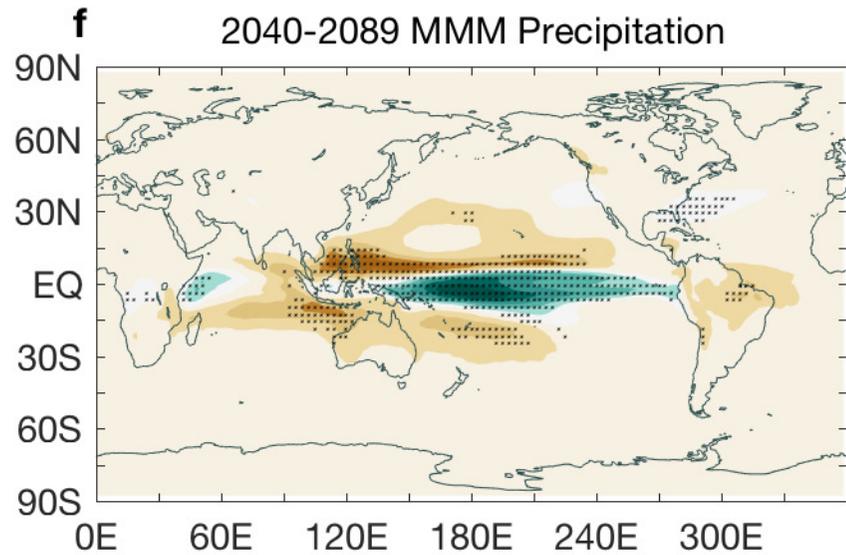
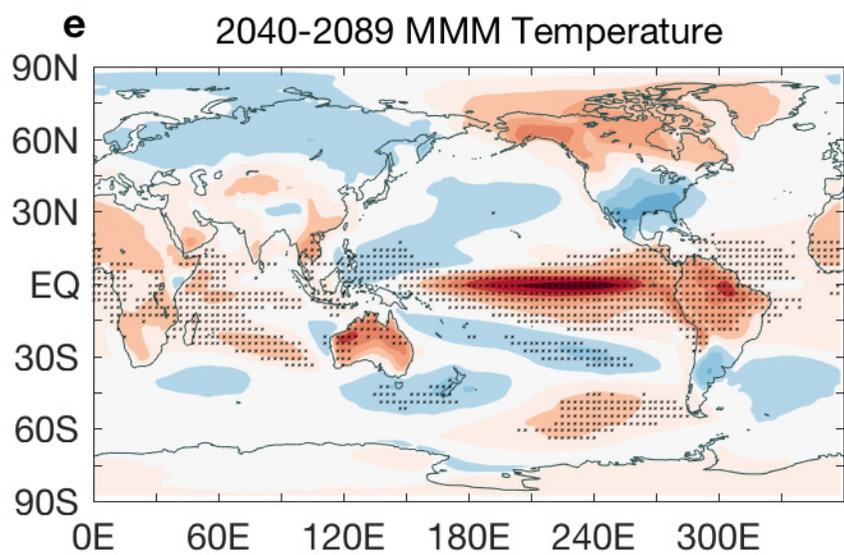
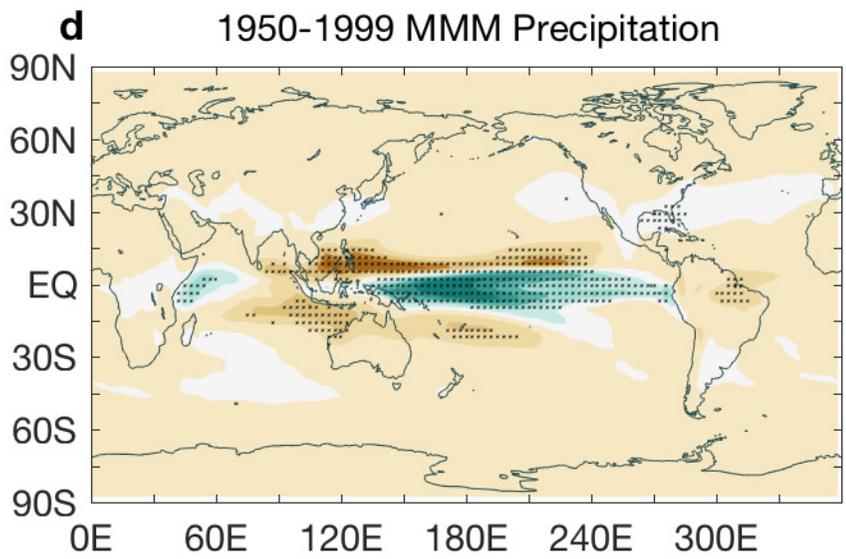
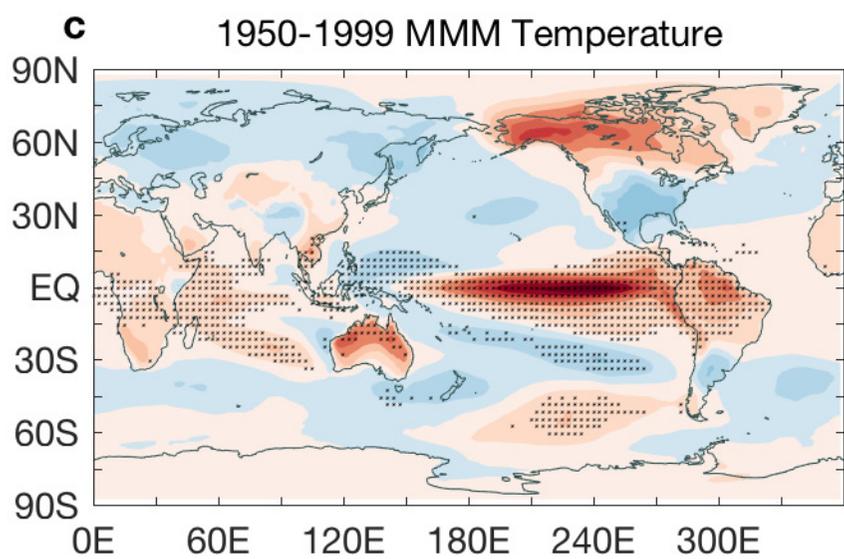
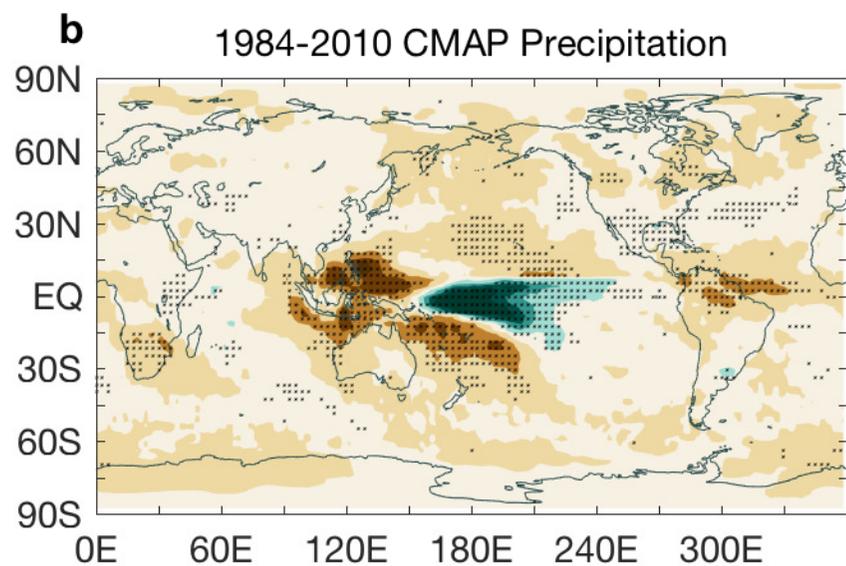
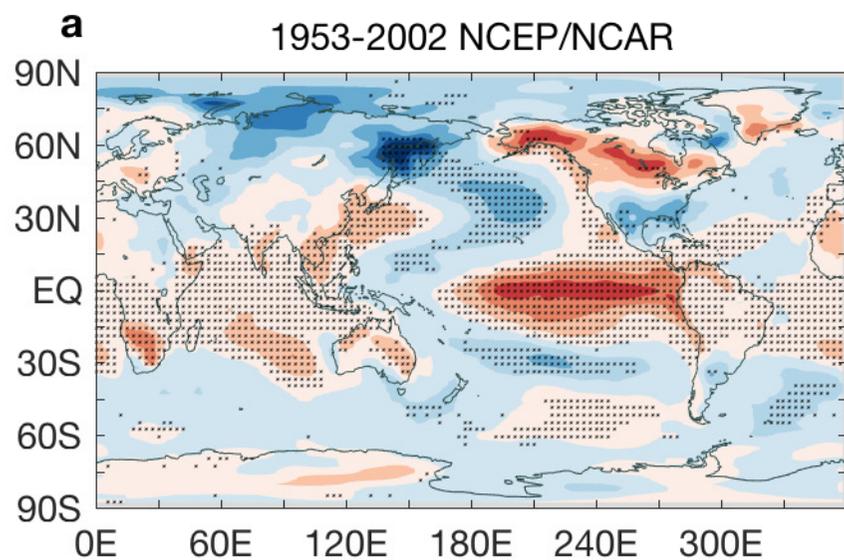


Figure 2.

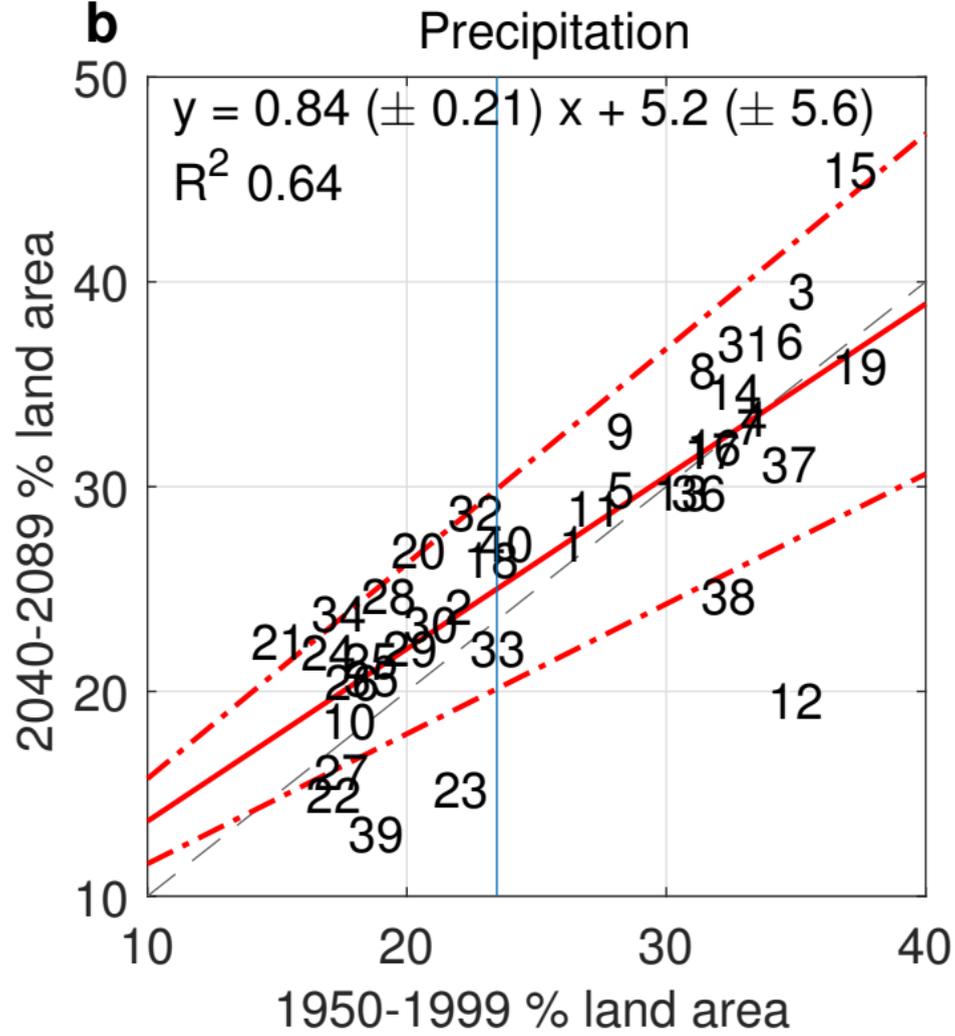
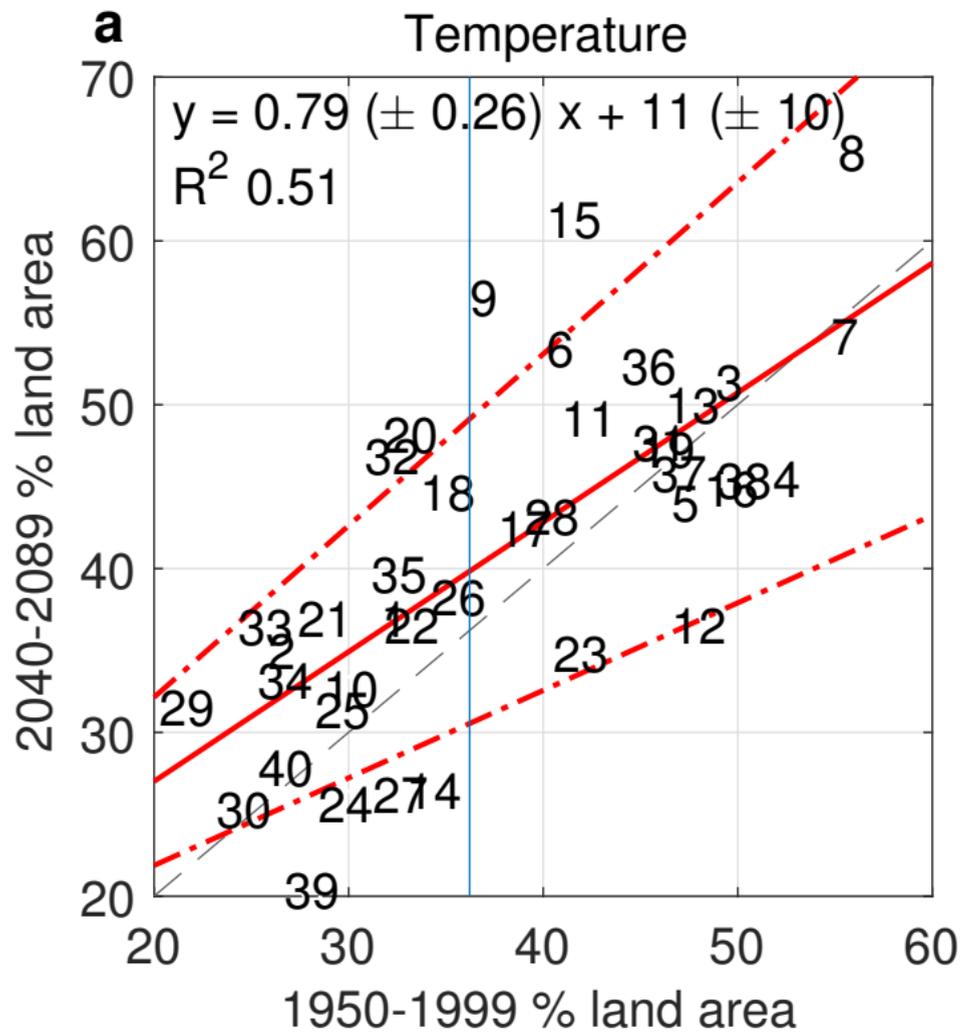


Figure 3.

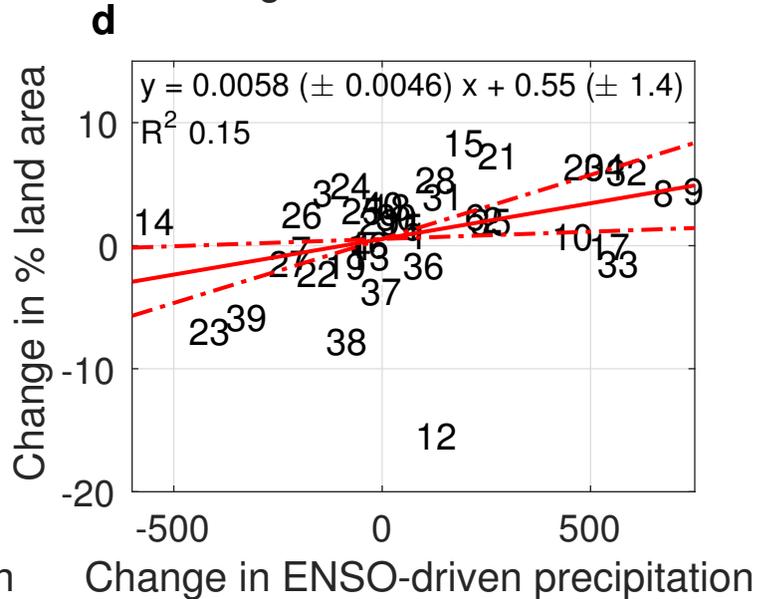
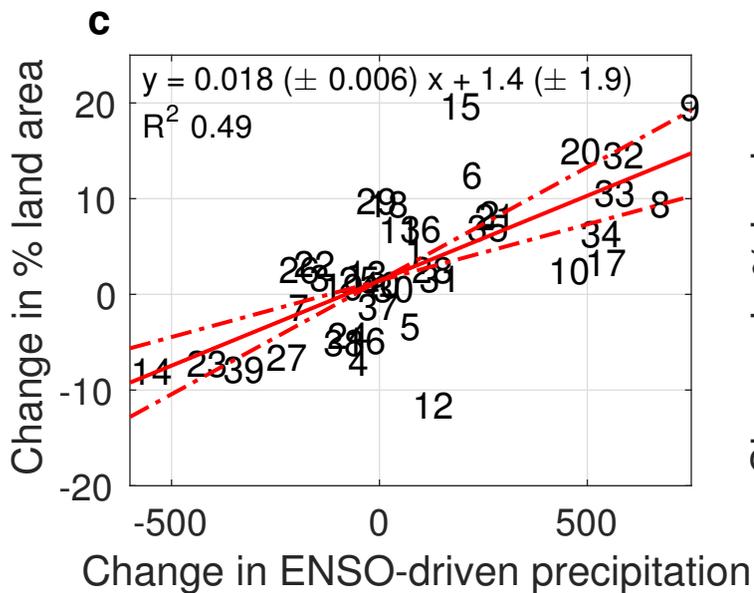
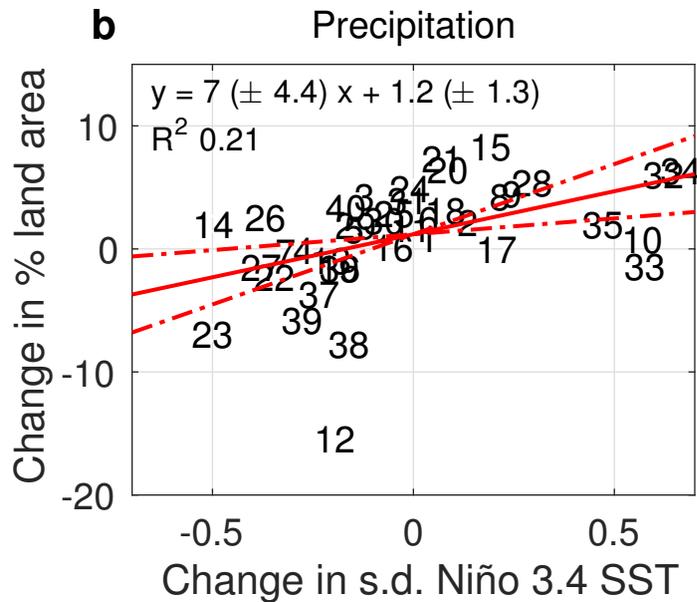
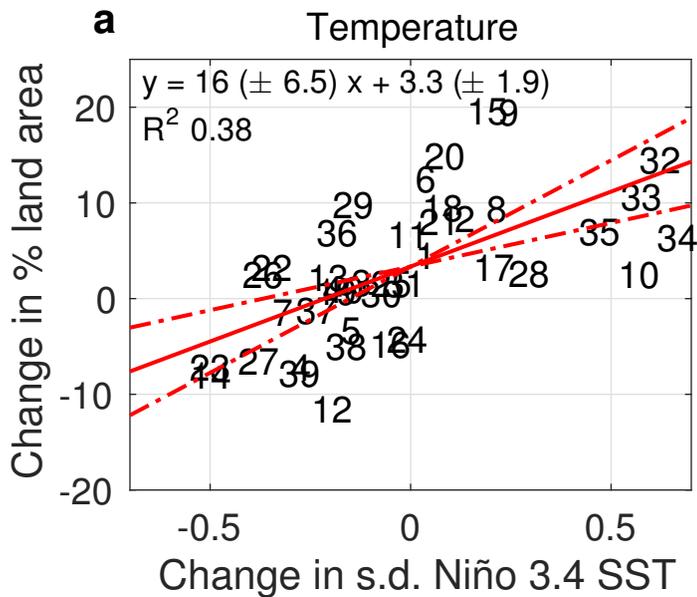


Figure 4.

