

1 **Pacific Walker Circulation Variability in**
2 **Coupled and Uncoupled Climate Models**

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33 **Abstract:** There is still considerable uncertainty concerning twentieth century trends in
34 the Pacific Walker Circulation (PWC). In this paper, observational datasets, coupled
35 (CMIP5) and uncoupled (AGCM) model simulations, and additional numerical sensitivity
36 experiments are analyzed to investigate twentieth century changes in the PWC and their
37 physical mechanisms. The PWC weakens over the century in the CMIP5 simulations, but
38 strengthens in the AGCM simulations and also in the observational twentieth century
39 reanalysis (20CR) dataset. It is argued that the weakening in the CMIP5 simulations is
40 not a consequence of a reduced global convective mass flux expected from simple
41 considerations of the global hydrological response to global warming, but is rather due to
42 a weakening of the zonal equatorial Pacific sea surface temperature (SST) gradient.
43 Further clarification is provided by additional uncoupled atmospheric general circulation
44 model simulations in which the ENSO-unrelated and ENSO-related portions of the
45 observed SST changes are prescribed as lower boundary conditions. Both sets of SST
46 forcing fields have a global warming trend, and both sets of simulations produce a
47 weakening of the global convective mass flux. However, consistent with the strong role
48 of the zonal SST gradient, the PWC strengthens in the simulations with the ENSO-
49 unrelated SST forcing, which has a strengthening zonal SST gradient, despite the
50 weakening of the global convective mass flux. Overall, our results suggest that the PWC
51 strengthened during 20th century global warming, but also that this strengthening was
52 partly masked by a weakening trend associated with ENSO-related PWC variability.

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54 *Keywords: Pacific Walker Circulation; Hydrological Cycle; ENSO*

55 **1. Introduction**

56 The Pacific Walker Circulation (PWC), a planetary-scale east-west
57 overturning atmospheric circulation in the equatorial belt with ascent over the
58 western and descent over the eastern Pacific Ocean, is an important component of
59 the global climate system. There is growing interest in how the PWC has been
60 affected by global warming over the past century (Vecchi and Soden 2007;
61 Vecchi et al. 2006; Meng et al. 2012; Tokinaga et al. 2011; Tokinaga et al. 2012).
62 As yet, however, there is no clear consensus on whether it has weakened,
63 strengthened, or remained unchanged. In an early modeling study, Knutson and
64 Manabe (1995) found that the time-mean upward motion weakened in a 4xCO₂
65 perturbation experiment, despite a 15% increase of precipitation, in the rising

66 branch of the PWC over the Indo-Pacific Warm Pool, and was accompanied by a
67 reduction of the zonal sea surface temperature (SST) gradient along the equatorial
68 Pacific. Held and Soden (2006) argued that a general slowdown of overturning
69 atmospheric circulations was physically necessary to reconcile the much weaker
70 response of global precipitation to global warming compared to the $\sim 7\% / \text{K}$
71 increase of saturation water vapor pressure obtained in many climate models.
72 Based on this theoretical expectation, Vecchi and Soden (2007) linked the
73 weakening of the PWC in their climate model simulations (and also apparently in
74 observations) to a global reduction of the convective mass flux. They further
75 noted that the conditions over the tropical Pacific resembled an El Niño-like
76 situation in a warming climate. The fact that they could also simulate a PWC
77 weakening in an atmospheric GCM coupled to a slab ocean instead of a dynamic
78 ocean suggested that it was not strongly influenced by ocean dynamics.

79 More recent studies have yielded conflicting results regarding twentieth
80 century PWC trends. Some have supported the weakening theory (Tokinaga et al.
81 2011; 2012), whereas others have found either no trend (Compo et al. 2011) or a
82 strengthening trend (Luo et al. 2012; Meng et al. 2012; Sohn and Park 2010;
83 Wang et al. 2012; L'Heureux et al. 2013). However, because these studies covered
84 different time periods, direct intercomparisons are inappropriate. Given the
85 existence of substantial natural variability on multi-decadal scales in the climate
86 system, it is clearly inappropriate to compare trends obtained over relatively short
87 (30-yr) intervals with those over longer (~ 100 -yr) intervals that are likely more
88 associated with radiatively forced climate change (L'Heureux et al. 2013).

89 It is important to note in this context that there is also uncertainty in how
90 global warming has affected and will affect tropical Pacific SSTs, with some
91 climate models generating an El Niño-like and others a La Niña-like response
92 (Collins 2005; Latif and Keenlyside 2009; Collins et al. 2010). Multi-model
93 ensemble mean responses often resemble an El Niño-like pattern (Knutson and
94 Manabe 1995; Solomon et al. 2007; Xie et al. 2010). A robust evaluation of the
95 models is complicated by uncertainties in observational SST reconstructions,
96 including disagreements as to whether the eastern equatorial Pacific warmed or
97 cooled over the 20th century (Deser et al. 2010). These disagreements have been
98 shown to be largely associated with differences in the representation of the El
99 Niño Southern Oscillation (ENSO) in the SST reconstructions. When the ENSO-

100 related variations are removed, the residual ENSO-unrelated 20th century SST
101 trend patterns indicate a strengthening of the zonal Pacific SST gradient (Compo
102 and Sardeshmukh 2010; Solomon and Newman 2012), consistent with previous
103 studies that found a strengthening of the full and the ENSO-unrelated zonal SST
104 gradient in some datasets (Cane et al. 1997; Guan and Nigam 2008). Interestingly,
105 the associated ENSO-unrelated trend pattern of Sea Level Pressure (SLP) does not
106 indicate a clear trend in an SLP-based index of the PWC (Solomon and Newman
107 2012). This suggests that (mis-) representations of ENSO-related variability in
108 observational datasets may have contributed substantially to the disagreements
109 regarding 20th century PWC variability and trends. An investigation of 20th
110 century PWC variability in coupled climate model simulations (Power and
111 Kociuba 2011) found that 30 to 70% of the simulated PWC weakening was
112 accounted for by external forcing and the rest by internal model variability, and
113 that models with relatively weak negative PWC trends were also generally
114 deficient in simulating ENSO.

115 The disagreements among observational and modeling studies concerning
116 the long-term PWC trend have generated a vigorous debate as to whether the
117 observations or models are in error. Tokinaga et al. (2012) argued that uncoupled
118 AGCMs fail to simulate a weakening of the PWC because of errors in the
119 observational SSTs that are used to force those models. To support their
120 argument, they forced their particular AGCMs with trends in observed night-time
121 marine air temperatures blended with SSTs, and obtained a weakening of the
122 PWC. L'Heureux et al. (2013), on the other hand, questioned the long-term
123 weakening of the PWC itself claimed in observational studies, stressing that
124 biases in the SLP observations and reconstructions compromise those claims,
125 given especially the low signal-to-noise ratios associated with sparse *in situ*
126 measurements before the 1950s. Focusing on the better-observed period since
127 1950, they found that the PWC has strengthened, not weakened. Observational
128 SLP data quality issues were also raised by DiNezio et al. (2013), who found the
129 PWC weakening trend in the CMIP5 coupled model simulations to be an order of
130 magnitude smaller than in observations and reconstructions. Given that only 26 of
131 the 101 CMIP5 simulations analyzed by them were consistent with the observed
132 PWC trend, they suggested that the weakening trend in the observations could be

133 a result of biases over data sparse oceanic regions such as the region around
134 Tahiti, which is data-poor before the 1940s.

135 Meng et al. (2012) examined 20th century PWC variability in observations
136 and in a suite of uncoupled atmospheric GCM simulations. They found that the
137 Pacific SST gradient, rather than external radiative forcing *per se*, drove the PWC
138 variability and that the PWC strengthened, not weakened, over the century.
139 However, Tokinaga et al. (2011; 2012), using in situ ocean and land observations
140 concluded that the PWC weakened over the second half of the century. Tokinaga
141 et al. (2011) also found a reduction of land precipitation over the maritime
142 continent, which is inconsistent with the model results of Knutson and Manabe
143 (1995), who obtained an increase in this general area in their 4xCO₂ experiment.
144 Knutson and Manabe reconciled their precipitation increase (and the associated
145 increase of latent heat release) with the weakening of upward motion (and
146 associated slowdown of the PWC) by noting an increase also of the static stability
147 and radiative cooling in the region. Their results were thus not inconsistent with
148 the dominant heat balance in areas of organized deep convection in the tropics
149 between diabatic heating and the adiabatic cooling of ascent.

150 The above very brief summary of the current inadequate understanding of
151 long-term PWC variability and trends is in sharp contrast to the understanding of
152 shorter-term interannual PWC variability. On interannual time scales the PWC,
153 regardless of whether it is defined in terms of vertical velocities in its vertical
154 branches, or zonal winds or pressure gradients in its horizontal branches, weakens
155 (strengthens) during El Niño (La Niña) events, and this is associated with a
156 reduction (intensification) of the zonal SST gradient in the equatorial Pacific. On
157 longer time scales, however, other physics and dynamics come into play as noted
158 above, so it is not obvious to what extent a simple relationship between the PWC
159 and the zonal SST gradient still holds (Karnauskas et al. 2009).

160 We have three aims in this paper. One is to reassess to what extent the
161 PWC weakened or strengthened over the *entire* 20th century in the longest
162 available observational and reanalysis datasets, and in newly available CMIP5
163 model simulations (Taylor et al. 2011) performed using the latest generation of
164 coupled models. Our second aim is to determine to what extent the 20th century
165 PWC trends were “ENSO-unrelated” or “ENSO-related”, by performing
166 uncoupled atmospheric GCM simulations in which the ENSO-unrelated and

167 ENSO-related portions of observed global SST anomalies, isolated using the
168 method of Compo and Sardeshmukh (2010) on a month-by-month basis
169 throughout the century, are prescribed as lower boundary conditions. Our third
170 aim is to investigate the tacit assumption made in numerous studies that a
171 weakening of the PWC is necessarily associated with the reduction of the global
172 convective mass flux expected from a weak global mean precipitation response to
173 global warming. In our view, this assumption has been the root cause of
174 skepticism toward observational studies that suggest a long-term strengthening of
175 the PWC, but is in fact unjustified. Our uncoupled atmospheric GCM simulations
176 are especially relevant in this regard. Both our ENSO-unrelated and ENSO-related
177 SST anomaly fields have a global warming trend over the century, and both
178 reduce the global convective mass flux, but the ENSO-unrelated SST forcing
179 strengthens the PWC, whereas the ENSO-related forcing weakens it. This is
180 mainly because over the equatorial Pacific, the ENSO-unrelated trend fields have
181 a steepening zonal SST gradient, whereas the ENSO-related trend fields have a
182 weakening SST gradient.

183 The data and methods used are summarized in Section 2. This is followed
184 by results and a discussion in Section 3, and concluding remarks in Section 4.

185 **2. Data and methods**

186 We used monthly mean sea level pressure (SLP), convective mass flux
187 (M_c), pressure vertical velocity at 500 hPa (ω_{500}), Surface Temperature (TS), and
188 Sea Surface Temperature (SST) from coupled GCM (CGCM) and uncoupled
189 atmospheric GCM (AGCM) simulations, the Twentieth Century Reanalysis
190 (20CR; (Compo et al. 2011)) dataset, and observational/reconstruction SLP (Allan
191 and Ansell 2006) and SST (Rayner et al. 2003) datasets from the Hadley Center
192 (HadSLP2 and HadISST1.1). We used one ensemble member from each ensemble
193 of simulations generated using 12 different CGCMs (see Table 1) for the CMIP5
194 project, all spanning the 20th century and using specified observed time-varying
195 concentrations of greenhouse gases (GHGs), ozone, and aerosols as radiative
196 perturbations. We also used four ensemble members of 20th century simulations
197 by an uncoupled AGCM (GISS-E2-R) (Schmidt et al. 2006) participating in the
198 Atmospheric Model Inter-comparison Project (AMIP). The SSTs used in these
199 GISS-E2-R simulations were a merged version of the HadISST1.1 and NOAA OI

200 v2 with HadISST1.1 anomalies plus the NOAA OIv2 (Reynolds et al. 2002)
201 climatology (1971-2000) for data spanning 1871 to November 1981, and the
202 NOAA OIv2 data from December 1981 to the present (Hurrell et al. 2008).

203 We additionally used a three-member ensemble of AGCM simulations
204 generated in-house with the National Center for Atmospheric Research (NCAR)
205 Community Atmospheric Model version 4 (CAM4), which is the atmospheric
206 component of the Community Earth System Model (CESM1; Gent et al. (2011)).
207 For these simulations, the CAM4 was configured at a horizontal resolution of 0.9°
208 $\times 1.25^\circ$ (lat \times lon) and 26 hybrid sigma-pressure levels in the vertical. The first
209 ensemble member was generated using observed/reconstructed Hadley Center
210 SSTs and sea ice concentrations (HadISST1.1) as lower boundary conditions, in
211 addition to other 20th century atmospheric forcings used by the coupled models,
212 for the 1900 – 2005 period. The other two ensemble members were generated
213 using Centennial *in situ* Observation-Based Estimates of SST (COBE SST) (Ishii
214 et al. 2005) and NOAA Extended Reconstructed SST version 3 (ERSST v3b)
215 (Smith and Reynolds 2004) SST fields respectively, and with all other forcings
216 identical to those for the first ensemble member. The first year of the model runs
217 was regarded as spin-up and not considered in the subsequent analysis. These
218 CAM4 runs will be referred as control (CTRL) simulations.

219 In order to estimate the ENSO-related contribution to PWC variability, we
220 performed two sensitivity experiments with the CAM4 model, prescribing time-
221 evolving ENSO-related and ENSO-unrelated monthly SST anomaly fields plus
222 the long-term mean seasonally varying SST fields as lower boundary conditions,
223 and with all other forcings identical to those in the CTRL runs. Each of these
224 experiments consisted of three ensemble members, as in the case of the CTRL
225 experiment. The ENSO-related and ENSO-unrelated SST anomaly fields from
226 HadISST1.1, COBE SST, and ERSST v3b were derived using the dynamical
227 ENSO filter described in Compo and Sardeshmukh (2010).

228 The PWC index was defined as the difference in area-averaged monthly
229 SLP between the eastern and western Pacific as in Vecchi et al. (2006). The
230 eastern Pacific ($160^\circ\text{W} - 80^\circ\text{W}$ and $5^\circ\text{S} - 5^\circ\text{N}$) and western Pacific ($80^\circ\text{E} -$
231 160°E and $5^\circ\text{S} - 5^\circ\text{N}$) areas used for this purpose are shown as rectangular boxes
232 in Fig. 6(a). An index of the zonal equatorial Pacific sea surface temperature
233 gradient (ΔT_S) was similarly defined, but using the SST difference between the

234 western and eastern Pacific. The land areas in these regional boxes were masked
235 out in the model and 20CR data when computing ΔTS . The M_c output from the
236 different models was integrated vertically from $p_{\text{bottom}} = 1000$ hPa to $p_{\text{top}} = 100$
237 hPa and normalized by $P_{\text{bottom}} - P_{\text{top}}$. The trends in all quantities presented here
238 were calculated using least-squares linear regressions and their level of
239 significance was estimated using a t -test. The number of degrees of freedom (dof)
240 were adjusted for the lag 1 autocorrelation of the residuals of the linear regression
241 (r_1), when r_1 was significant at the 5% level, as $n' = n(1 - r_1)/(1 + r_1)$, where n' is
242 the adjusted dof and n is the original dof (Wilks 2011). Unless noted otherwise,
243 all trends and changes stated below as “significant” were estimated to be
244 significant at the 5% level. The multi-model ensemble means of M_c , ω_{500} , and
245 precipitation were computed after regridding to a common $2.5^\circ \times 2.5^\circ$ latitude-
246 longitude grid (using the ‘linint2’ function in the NCAR Command Language,
247 NCL (NCAR 2012)).

248 We performed a Clausius-Clapeyron (C-C) scaling of the column
249 integrated water vapor w and precipitation as in Held and Soden (2006). The
250 values of M_c were similarly scaled. The mean responses of w , precipitation, and
251 M_c were determined as differences between 1996–2005 averages and 1901–1910
252 averages, similar to Held and Soden (2006) and Vecchi and Soden (2007).

253 For ease of discussion below we quantified the similarities of the SLP, M_c ,
254 ω_{500} , and TS trend fields in the uncoupled CAM4 and coupled CMIP5 simulations
255 in terms of the centered spatial correlations (Wilks 2011) of these fields over the
256 oceanic regions of the tropics ($30^\circ\text{S} - 30^\circ\text{N}$).

257 **3. Results and discussion**

258 The close relationship between the zonal SST gradient and easterly Trade
259 Winds over the equatorial Pacific, which are an important component of the PWC,
260 is well known (Bjerknes 1966; Lindzen and Nigam 1987). Fig. 1 compares the
261 20th century variability of ΔTS and the SLP-based PWC index in the statistical
262 reconstructions, 20CR, and model simulations. The curves show anomalies from
263 the long-term mean annual cycle, and the straight lines are least-squares linear
264 trends. Fig. 1a shows a significant ($p < 0.05$) weakening PWC trend ($-4 \text{ Pa decade}^{-1}$)
265 in the HadSLP2 dataset and an apparently inconsistent strengthening ΔTS trend
266 ($0.04 \text{ K decade}^{-1}$, $p < 0.05$) in the HadSST1.1 dataset. As already discussed in the

267 introduction, errors in the SST, SLP, or both reconstructions could be behind this
268 apparent inconsistency. A possibility of opposite SST and SLP trends involving
269 ocean dynamical mechanisms has also been suggested previously (Karnauskas et
270 al. 2009), but we cannot confirm that the strengthening of Δ TS presented here
271 resulted from an ocean dynamical process. Note also that although the trends in
272 the PWC and Δ TS reconstructions are opposite, the two time series themselves are
273 positively correlated with a correlation coefficient R of 0.66. Strengthening Δ TS
274 trends are seen in other SST reconstructions, such as the ERSST3 ($0.01 \text{ K decade}^{-1}$)
275 and COBE SST ($0.03 \text{ K decade}^{-1}$) datasets as well, although only the latter is
276 statistically significant at the 5% level. The PWC and Δ TS time series derived
277 from the 20CR, shown in Fig. 1b, both show significant strengthening trends, and
278 the time series are well correlated ($R = 0.84$). It should be noted that the 20CR
279 uses HadISST1.1 as the SST boundary condition in its data assimilation system
280 (Compo et al. 2011). The ensemble means of the three CAM4 CTRL simulations
281 shown in Fig. 1c also have significant strengthening trends in the PWC and Δ TS,
282 and the two time series are highly correlated ($R = 0.93$). Note that the
283 strengthening Δ TS trend in Fig. 1c reflects an average over three different
284 observational SST datasets: HadISST1.1, COBE, and ERSSTv3b. The
285 strengthening of the PWC in the 20CR and CTRL simulations is consistent with
286 the results of Meng et al. (2012). On the other hand, the CMIP5 model
287 simulations (Fig. 1d) show slight but statistically significant weakening trends in
288 both the PWC and Δ TS. Similar to Figs 1b and 1c, the ensemble-mean CMIP5
289 PWC and Δ TS time series in Fig. 1d are also highly correlated ($R=0.89$). Indeed
290 this is also true for the individual models in the CMIP5 ensemble, with R ranging
291 between 0.7 (GFDL-ESM2G) and 0.93 (CCSM4). The weakening ensemble-mean
292 PWC trend in this CMIP5 ensemble is consistent with previously reported
293 weakening trends in coupled model simulations with anthropogenic forcing
294 (Vecchi et al. 2006); however, its magnitude is smaller ($\sim -1.7 \text{ Pa decade}^{-1}$ vs. $\sim -$
295 $2.5 \text{ Pa decade}^{-1}$, see Fig. 2 in Vecchi et al. (2006)).

296 We next estimate the ENSO-related PWC variability by considering the
297 CAM4 model forced with ENSO-related and ENSO-unrelated SSTs in two
298 separate sensitivity experiments. The ENSO-related SST anomalies were filtered
299 on a month-by-month basis from the HadISST1.1, COBE SST, and ERSST v3b
300 data using a novel method, which unlike conventional regression or band-pass

301 filtering methods does not define ENSO in terms of a single index or temporal
302 frequency, but rather treats it as an evolving dynamical process (Penland and
303 Matrosova (2006); Penland and Sardeshmukh (1995); see Compo and
304 Sardeshmukh (2010) for details). Fig. 2a shows that both the PWC and ΔTS have
305 significant weakening trends in the ensemble mean of the three CAM4
306 simulations forced with the ENSO-related SSTs. On the other hand, as shown in
307 Fig. 2b, they both have strengthening trends in the simulations forced with the
308 ENSO-unrelated SSTs that are consistent with the strengthening trends in the
309 20CR dataset and CTRL simulations in Fig. 1. The PWC has an ensemble mean
310 trend of about 2.3 Pa/decade in the CTRL ensemble, and about 10.1 and -7.4
311 Pa/decade in the ENSO-unrelated and ENSO-related SST forced ensembles,
312 respectively. Note that the strengthening ENSO-unrelated ΔTS trend in Fig. 2b is
313 the dominant contributor to the strengthening full ΔTS trend in Fig. 1a-c.

314 These intercomparisons of long-term trends in Figs. 1 and 2 suggest a
315 dominant role of ENSO-unrelated SST changes in the dynamics of the PWC
316 trend. They also suggest that the large ENSO-related PWC variability
317 compromises the estimation of the PWC trend, depending on how that variability
318 is represented in observational datasets of limited length and quality. This is
319 evident from the similar large magnitudes of interannual PWC variability in the
320 reconstruction and 20CR datasets and CTRL simulations in Fig. 1 and the ENSO-
321 related PWC variability in Fig. 2a.

322 As already mentioned, a weakening of the PWC has generally been
323 assumed to be associated with a global slowdown of overturning atmospheric
324 circulations, i.e., a decrease of global mean M_c , expected from the relatively
325 muted response of global mean precipitation (compared to that of global mean
326 column integrated water vapor) to global warming. Here we investigate how
327 closely the PWC changes follow those of the globally, as well as tropically,
328 averaged M_c in the model simulations. In the CTRL simulations (Fig. 3a), both the
329 globally and tropically averaged M_c show significant decreasing trends, that are
330 relatively large over the second half of the century. The simulated PWC time
331 series is repeated in Fig. 3a for convenience, and all time series are presented as
332 36 month running means for clarity. The correlation between the PWC and
333 globally (tropically) averaged M_c time series is -0.03 (0.24), which indicates a
334 very weak association between the two variables. The same analysis repeated for a

335 four-member ensemble mean of uncoupled GISS-E2-R AMIP (Schmidt et al.
336 2006) simulations (Fig. 3b) yields similar results, with a statistically significant
337 weakening of tropically and globally averaged M_c and strengthening of the PWC.
338 As in the case of the CTRL simulations using CAM4, the PWC in the GISS-E2-R
339 ensemble is also poorly correlated with the global M_c ($R = 0.07$) as well as the
340 tropical M_c ($R = 0.26$). The CMIP5 ensemble (Fig. 3c) shows a weakening of both
341 the globally and tropically averaged M_c , in agreement with previous model studies
342 (Held and Soden 2006; Vecchi and Soden 2007). However, the relationship
343 between the PWC and M_c is weak, considering the much smaller magnitude of the
344 PWC trend here than in the CTRL runs and also the weak correlation of the two
345 time series, with $R = 0.15$ (0.28) for the globally (tropically) averaged M_c . The
346 PWC and M_c in the CMIP5 simulations are somewhat better correlated at longer
347 time scales (10-yr running means) with $R = 0.44$ (0.44) for global (tropical)
348 averages. However, the corresponding correlations of $R = -0.16$ (-0.04) remain
349 weak in the CTRL simulations.

350 The M_c and PWC changes in the ENSO-related and ENSO-unrelated SST
351 forcing experiments are shown in Fig. 4. The globally and tropically averaged M_c
352 weaken significantly over the century in both experiments. However, the PWC
353 weakens in the ENSO-related forcing experiment (Fig. 4a) whereas it strengthens
354 in the ENSO-unrelated forcing experiment (Fig. 4b). The correlations of the PWC
355 and globally (tropically) averaged M_c time series are 0.13 (0.42) in the ENSO-
356 related forcing experiment and -0.25 (-0.1) in the ENSO-unrelated forcing
357 experiment.

358 Taken together, the results in Figures 3 and 4 suggest a rather tenuous
359 relationship between the PWC and the globally or tropically averaged convective
360 mass flux M_c . They highlight the danger of linking global energy balance
361 constraints to regional phenomena such as the Pacific Walker Circulation.
362 Tokinaga et al. (2012) also suggested that the PWC trend is more closely
363 associated with the trend in the Indo-Pacific SST gradient, rather than the
364 reduction in global mean M_c .

365 To assess the consistency of the simulated changes in M_c with the scaling
366 arguments of Held and Soden (2006), we related the global mean responses of
367 precipitation, column integrated water vapor, and M_c to the change in global mean
368 TS in each of our datasets (Table 2). The difference (1996–2005 average minus

369 1901–1910 average) in global mean TS (δTS) yields a value of 0.90 K for the
370 coupled CMIP5 ensemble, which is closely matched by the δTS of 0.91K for the
371 uncoupled CTRL ensemble. The ENSO-related and ENSO-unrelated ensembles
372 show a weaker warming with δTS values of 0.52 K and 0.64 K respectively. The
373 scaled responses of the global mean precipitation to δTS are $1.2 \% K^{-1}$, $1.6 \% K^{-1}$,
374 $0.4 \% K^{-1}$, and $0.2 \% K^{-1}$ respectively in the CMIP5, CTRL, ENSO-related, and
375 ENSO-unrelated simulations. The responses in the CMIP5 and CTRL simulations
376 are consistent with the precipitation responses in the $1 - 2 \% K^{-1}$ range reported
377 by Held and Soden (2006) for the A1B scenario simulations of the CMIP3
378 coupled models. The weaker responses in the ENSO-related and ENSO-unrelated
379 simulations are not surprising, since they each specify only a part of the δTS
380 changes, and only over the oceans. Nonetheless, the substantial response to the
381 ENSO-related SST changes suggests that ENSO plays an important role in the
382 variability of even global mean precipitation.

383 The scaled responses of column-integrated water vapor range between 7%
384 K^{-1} and $7.4 \% K^{-1}$ in the CMIP5, CTRL, and ENSO-related simulations, consistent
385 with previous studies (e.g. (Held and Soden 2006; Vecchi and Soden 2007),
386 although the $5.5 \% K^{-1}$ response obtained in the ENSO-unrelated simulations is
387 somewhat weaker. The scaled M_c responses of $-3.1 \% K^{-1}$ (CMIP5), $-2.4 \% K^{-1}$
388 (CTRL), $-4.7 \% K^{-1}$ (ENSO-related), and $-2.4 \% K^{-1}$ (ENSO-unrelated) show a
389 reduction of M_c in not just the coupled but in all the uncoupled simulations as
390 well. Consistent with the arguments for a reduction in global M_c , the (ENSO-
391 related forcing) experiment with the largest difference between the scaled
392 responses of global mean column integrated water vapor and precipitation also
393 has the largest scaled reduction in M_c , and experiments with relatively weaker
394 differences between $\delta w/\delta TS$ and $\delta P/\delta TS$ have relatively weaker scaled reductions
395 in M_c .

396 The time series of δTS and the fractional changes of global precipitation
397 ($\delta P/P$), column integrated water vapor ($\delta w/w$), and convective mass flux ($\delta M_c/M_c$)
398 are shown in Fig. 5 for the four sets of simulations. The increases of $\delta w/w$ (Fig.
399 5c) with increases of δTS (Fig. 5a) in all simulations are generally consistent with
400 C-C scaling. The decreases of $\delta M_c/M_c$ (Fig. 5d) in all simulations are also
401 consistent with the weak increases of $\delta P/P$ (Fig. 5b) according to the global
402 energy balance arguments of Held and Soden (2006).

403 It is interesting to examine the spatial patterns of the SLP, TS, and M_c
404 trends, and also of the ω_{500} and precipitation trends, to better understand the above
405 results in a larger geographical context. The 1901–2005 linear trends of these
406 quantities in the CTRL simulations are shown in Fig. 6. Focusing on the 10°S-
407 10°N equatorial zone, it is clear that the strong positive SLP trend over the eastern
408 Pacific and the weak positive and negative trends over the maritime continent are
409 both responsible for the strengthening of the PWC in the CTRL simulations. The
410 trend patterns of vertical velocity (Fig. 6b) and convective mass flux (Fig. 6c) are
411 consistent with that of SLP (Table 3). Further, the surface warming trend over the
412 western Pacific, and the slight cooling trend over the eastern Pacific (Fig. 6d),
413 both contribute to steepening the zonal SST gradient over the equatorial Pacific.
414 Together, these results indicate a robust strengthening PWC trend associated with
415 both enhanced ascent in its ascending branch and enhanced descent in its
416 descending branch, and with enhanced zonal SLP and SST gradients in its surface
417 branch. Considering the wider tropical belt (30°S-30°N), it is evident that the
418 convective mass flux M_c has a larger area of negative trends than positive trends,
419 which contributes to a general weakening of M_c as highlighted in Fig. 5d. Figures
420 5d and 6 show that a general weakening of area averaged M_c can be compatible
421 with a strengthening of the PWC. This suggests that air-sea interactions over the
422 tropical Pacific dominated over radiative forcing in determining the PWC
423 variability and trend over the 20th century, in line with Xie et al. (2010).

424 Figure 7 shows the spatial patterns of the trends in the coupled CMIP5
425 simulations, in an identical format to that of Fig 6. Focusing again on the 10°S-
426 10°N equatorial zone, it is evident that the SLP, vertical velocity, convective mass
427 flux, and TS trends have spatial variations that are very different, and generally
428 *opposite*, to those in the CTRL simulations over the Pacific sector. It is
429 particularly interesting that the stronger surface warming trend over the cold
430 tongue region of the eastern equatorial Pacific (Fig. 6d), together with a slightly
431 weaker warming over the western Pacific, resembles an El Niño-like pattern,
432 which was also seen in many of the coupled CMIP3 simulations (Collins 2005;
433 Solomon et al. 2007) as well as in the 4xCO₂ experiments of Knutson and
434 Manabe (1995). Together, they indicate a robust *weakening* PWC trend associated
435 with both reduced ascent in its ascending branch and reduced descent in its
436 descending branch, and reduced zonal SLP/SST gradients in its surface branch.

437 Considering the wider tropical belt (30°S-30°N), the convective mass flux M_c has
438 a larger area of negative trends than positive trends, similar to the CTRL runs but
439 over an even larger area, which makes a major contribution to the general
440 weakening trend of globally and tropically averaged M_c in these CMIP5
441 simulations.

442 Table 3 shows the spatial correlations of various pairs of trend patterns in
443 Figs. 6 and 7, computed over the oceanic regions of the tropics (30°S – 30°N). The
444 SLP trend is moderately correlated with the ω_{500} and M_c trends in the CTRL
445 ensemble ($R = 0.41$ and -0.38) as well as in the CMIP5 ensemble ($R = 0.46$ and $-$
446 0.34). The SLP trend is well correlated with the TS trend in the CTRL ensemble
447 ($R=-0.73$), but moderately correlated in the CMIP5 ensemble ($R=-0.4$). The ω_{500}
448 and M_c trends are highly correlated in the CTRL ensemble ($R=-0.91$) and also in
449 the CMIP5 ensemble ($R=-0.85$). The ω_{500} and TS trends are moderately correlated
450 in the CTRL ensemble ($R=-0.46$) and also in the CMIP5 ensemble ($R=-0.34$).
451 Similarly, the M_c and TS trends are moderately correlated in the CTRL ensemble
452 ($R=0.49$) and in the CMIP5 ensemble ($R=0.52$). Finally, the trend patterns of SLP
453 in the CTRL and CMIP5 ensembles are less strongly correlated with one another
454 ($R=0.36$), and those of ω_{500} , M_c , and TS are poorly correlated ($R=-0.01$, 0.07 , and
455 0.03 respectively).

456 The spatial correlations presented in Table 3 support the interpretation of
457 the trend patterns in Figs. 6 and 7 that changes of the PWC (defined by an SLP
458 index) cannot be explained by changes of either the globally or tropically
459 averaged convective mass flux. In general, the SLP variability is strongly tied to
460 the TS variability in the uncoupled AGCM simulations, but much less strongly in
461 the coupled CMIP5 simulations. The M_c and ω_{500} variations, on the other hand,
462 are strongly correlated in both AGCM and CMIP5 simulations.

463 We also computed the linear trends of SLP and ω_{500} in the 20CR dataset.
464 Their patterns are shown in Fig. 8. The SLP trends (Fig. 8a) are broadly consistent
465 with those in the CTRL simulations (Fig. 6a), although they have somewhat larger
466 magnitudes, and also have different signs over Asia. Importantly for our purposes,
467 however, both the 20CR and CTRL SLP trends are positive over the eastern
468 tropical Pacific and negative over the equatorial western Pacific and Indian
469 Oceans. In other words, both SLP trend fields indicate a strengthening PWC
470 trend. The ω_{500} trend pattern in the 20CR dataset (Fig. 8b) is also consistent with

471 that of SLP in terms of strengthening ascent over the maritime continent, in the
472 area of maximum equatorial SLP decrease, and descent over the equatorial eastern
473 Pacific, in the area of maximum equatorial SLP increase. Both of these features
474 are consistent with a strengthening of the PWC.

475 As a final check that we really do obtain consistently different PWC trend
476 responses in the CTRL simulations with prescribed observed SSTs (with due
477 acknowledgement of their uncertainties) and in the CMIP5 simulations with
478 imposed radiative perturbations in which the SSTs are obtained as part of the
479 response (and which may also have errors), we examine the precipitation trends in
480 the simulations. As noted earlier, Knutson and Manabe (1995) obtained a 15%
481 increase of precipitation over the maritime continent in response to a 4xCO₂
482 forcing. Some observational studies, however, suggest a decreasing recent
483 precipitation trend over the maritime continent (Tokinaga et al. 2011). Our CTRL
484 simulations do not show a significant precipitation trend over the maritime
485 continent (Fig. 9a), but they do show a decreasing precipitation trend over the
486 eastern equatorial Pacific consistent with a strengthening of the PWC. The result
487 for the CMIP5 simulations (Fig. 9b) is a decreasing precipitation trend over the
488 maritime continent and an increasing trend over the central and eastern equatorial
489 Pacific, consistent with a weakening of the PWC. This pattern bears some
490 resemblance to that associated with El Niño events (Dai and Wigley 2000). It is
491 also consistent with the El Niño-like SST warming trend pattern in the CMIP5
492 simulations (Fig. 6d). The precipitation trend pattern in the uncoupled simulations
493 with ENSO-related SST forcing (Fig. 9c) further supports the notion that the
494 precipitation trend in the CMIP5 simulations is “El Niño-like”, although of much
495 weaker magnitude. The precipitation trend pattern in the uncoupled simulations
496 with ENSO-unrelated SST forcing (Fig. 9d) shows an increasing trend over the
497 western Pacific and a decreasing trend over the central and eastern tropical
498 Pacific, consistent with the strengthening of the PWC.

499 Similar to our interpretation of the PWC trend, the precipitation trend
500 patterns in Fig. 9 also suggest that the underlying tropical SST trend *pattern*, as
501 opposed to the overall tropical SST warming trend associated with global
502 warming, plays an important role in determining atmospheric circulation trends
503 over the tropical Indo-Pacific. The main reason the PWC trends are opposite in
504 the uncoupled CTRL and coupled CMIP5 simulations is because the trends in the

505 zonal SST gradients are opposite in these simulations, and not because the trends
506 in the global convective mass flux are opposite (they are, in fact, the same).

507 **4. Concluding remarks**

508 In this paper we investigated changes in the Pacific Walker Circulation over the
509 20th century using observational and reanalysis datasets and coupled
510 (atmosphere-ocean) as well as uncoupled (atmosphere only, prescribed SST)
511 model simulations. Consistent with Meng et al. (2012), our results suggest that
512 changes in the zonal SST gradient over the tropical Pacific, associated with
513 coupled air-sea processes rather than radiative forcing *per se*, are crucial in
514 determining PWC variability and trends. We challenge the notion that a
515 weakening or strengthening PWC trend is strongly tied to a weakening or
516 strengthening of the global convective mass flux M_c . This challenge is justified by
517 the poor correlation of the PWC and M_c variations in the simulations examined
518 here and also the fact that one can obtain a PWC trend of either sign in the
519 presence of a weakening M_c trend. With regard to the question of whether the
520 PWC strengthened or weakened over the 20th century in association with global
521 warming, our results suggest that it strengthened, but that this strengthening was
522 partly masked by a weakening trend associated with ENSO-related PWC
523 variability. We noted that this component of the PWC variability also
524 compromises attempts to pin down the sign of the PWC trend using observational
525 datasets of limited length and quality. The reanalysis data could also be affected
526 by the quality of the observations that are assimilated. We suspect that the
527 weakening PWC trend obtained in the CMIP5 models is ultimately due to their
528 misrepresentation and systematic underestimation of the spatial variation of
529 tropical SST trends (clearly demonstrated by Shin and Sardeshmukh (2011) for
530 the CMIP3 models), but more work is needed to confirm or reject this suspicion.

531

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752 **Table 1:** Details of CMIP5 models used in this study

Model	Horizontal grids (latxlon)	Reference
BCC-CSM1-1	64x128	Wu (2012)
CanESM2	64x128	Chylek et al. (2011)
CCSM4	192x288	Gent et al. (2011)
CESM1-BGC	192x288	Riley et al. (2011)
CESM1-WACCM	96x144	Richter et al. (2008)
GFDL-CM3	90x144	Griffies et al. (2011)
GFDL-ESM2G	90x144	Dunne et al. (2012)
GFDL-ESM2M	90x144	Dunne et al. (2012)
MIROC5	128x256	Watanabe et al. (2010)
MRI-CGCM3	160x320	Yukimoto et al. (2012)
NorESM1-M	96x144	Bentsen et al. (2012)
NorESM1-ME	96x144	Tjiputra et al. (2012)

753

754 **Table 2:** Decadal change of global averaged surface temperature (δTS) for 1996–
755 2005 average minus 1901–1910 average and associated scaled changes in column
756 integrated water vapor (δw), Precipitation (δP), and Convective mass flux (ΔM_c).

757 Clausius-Claperyon scaling for $\delta w/\delta TS$ is $\sim 7\% K^{-1}$ assuming constant relative
 758 humidity.

Experiment/Variab le	δTS (K)	$\delta w/\delta TS$ (% K^{-1})	$\delta P/\delta TS$ (% K^{-1})	$\Delta M_c/\delta TS$ (% K^{-1})
CMIP	0.90	7.4	1.2	-3.1
CTRL	0.91	7.0	1.6	-2.4
ENSO-rel	0.52	7.3	0.4	-4.7
ENSO-unrel	0.64	5.5	0.2	-2.4

759

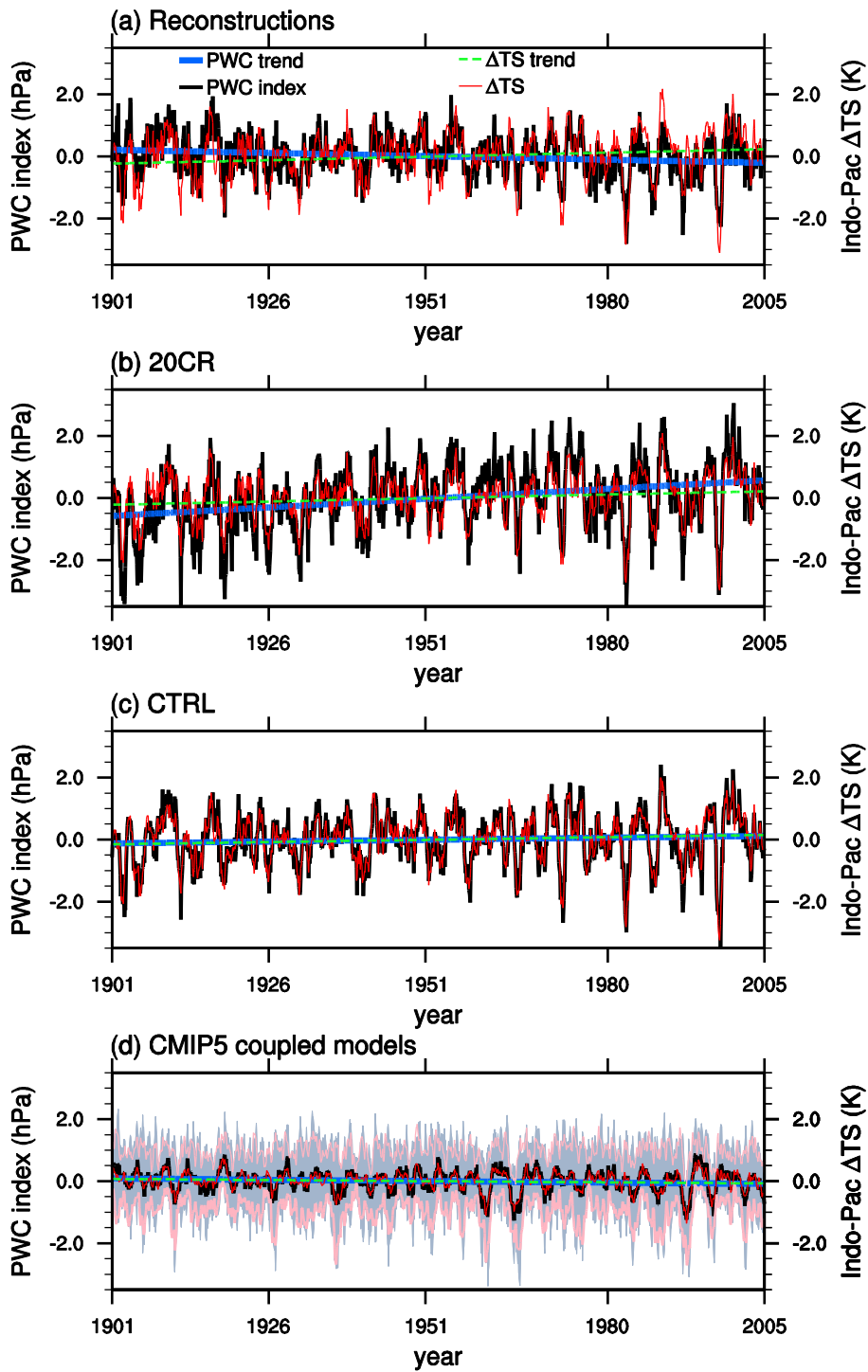
760 **Table 3:** Spatial correlations over tropical Oceanic region between trend maps of
 761 SLP, ω_{500} , M_c , and TS simulated by CAM4 CTRL (Green boxes), CMIP5 (blue
 762 boxes), and those between trend maps of respective variables simulated by CAM4
 763 CTRL and CMIP5 (yellow boxes).

Variable	SLP	ω_{500}	M_c	TS
SLP	0.36	0.41	-0.38	-0.73
ω_{500}	0.46	-0.01	-0.91	-0.46
M_c	-0.34	-0.85	0.07	0.49
TS	-0.40	-0.34	0.52	0.03

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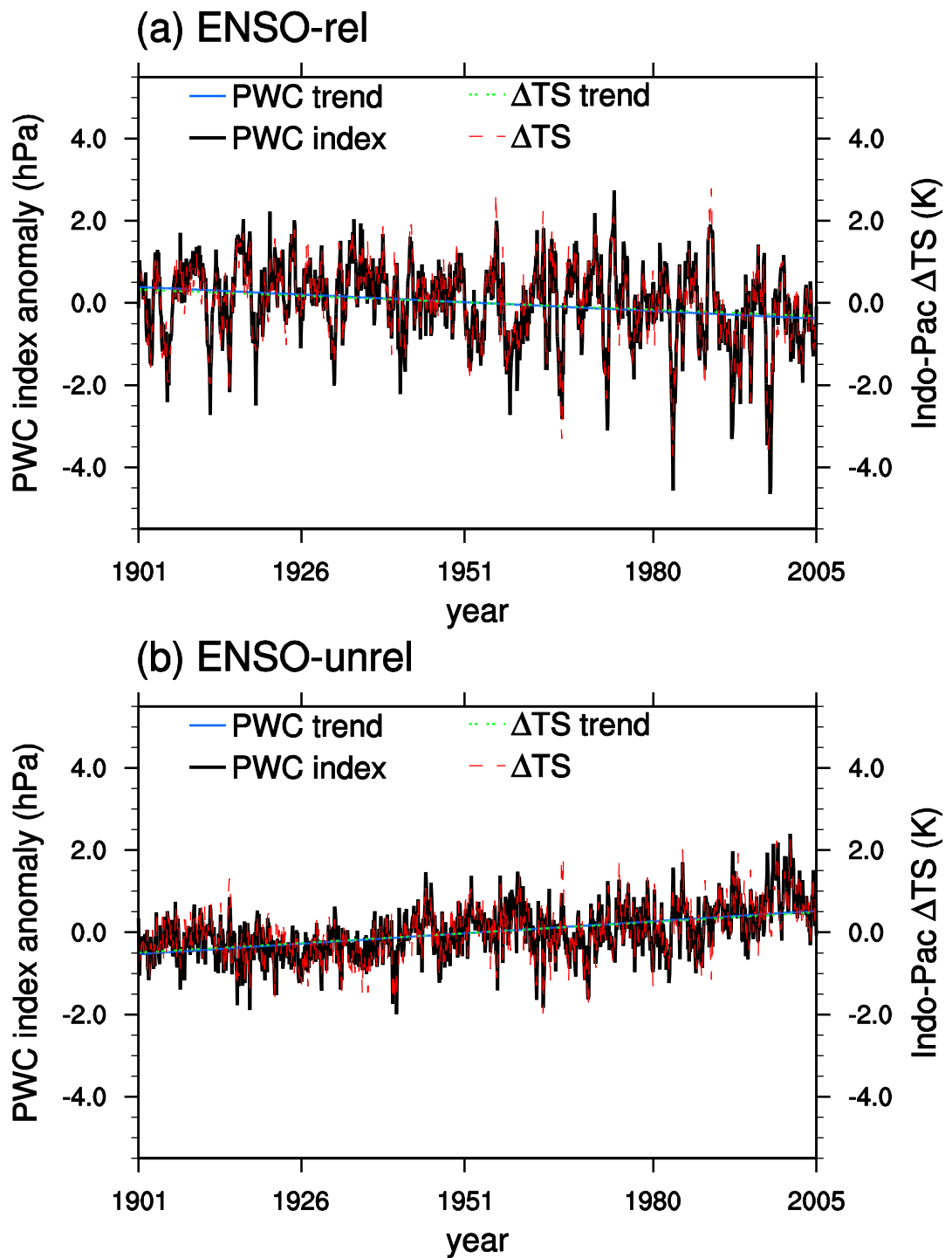
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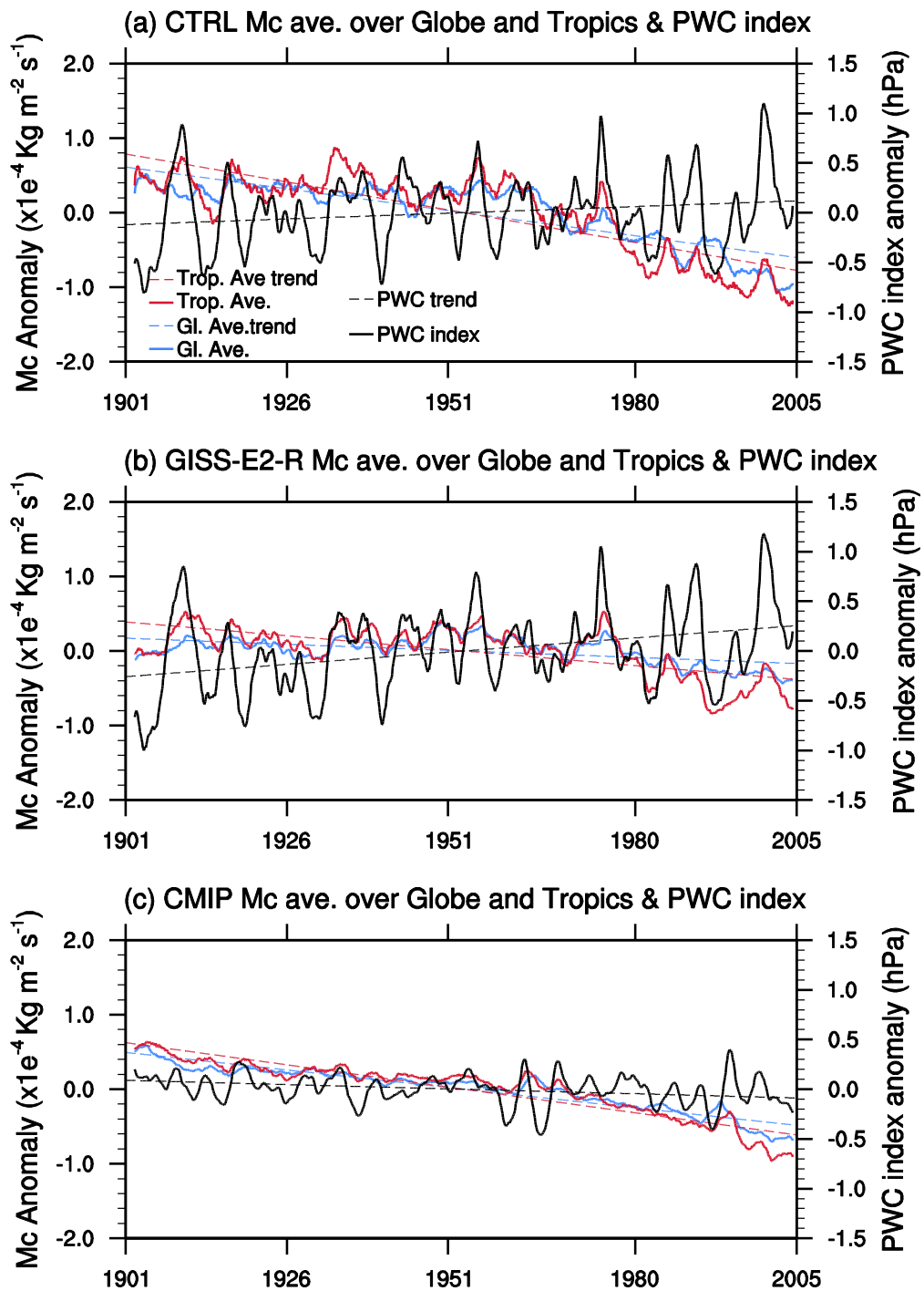
767

768 **Figure 1:** Pacific Walker Circulation (PWC, black curve) index and
 769 Pacific East West Temperature difference (Δ TS, red curve) of regions
 770 in Fig. 6a from (a) Reconstructions, (b) 20CR, (c) CAM4 CTRL
 771 ensemble, and (d) CMIP5 models (Table 1). Blue (green dashed) line
 772 shows the least-squares linear trend fit to the PWC (Δ TS).



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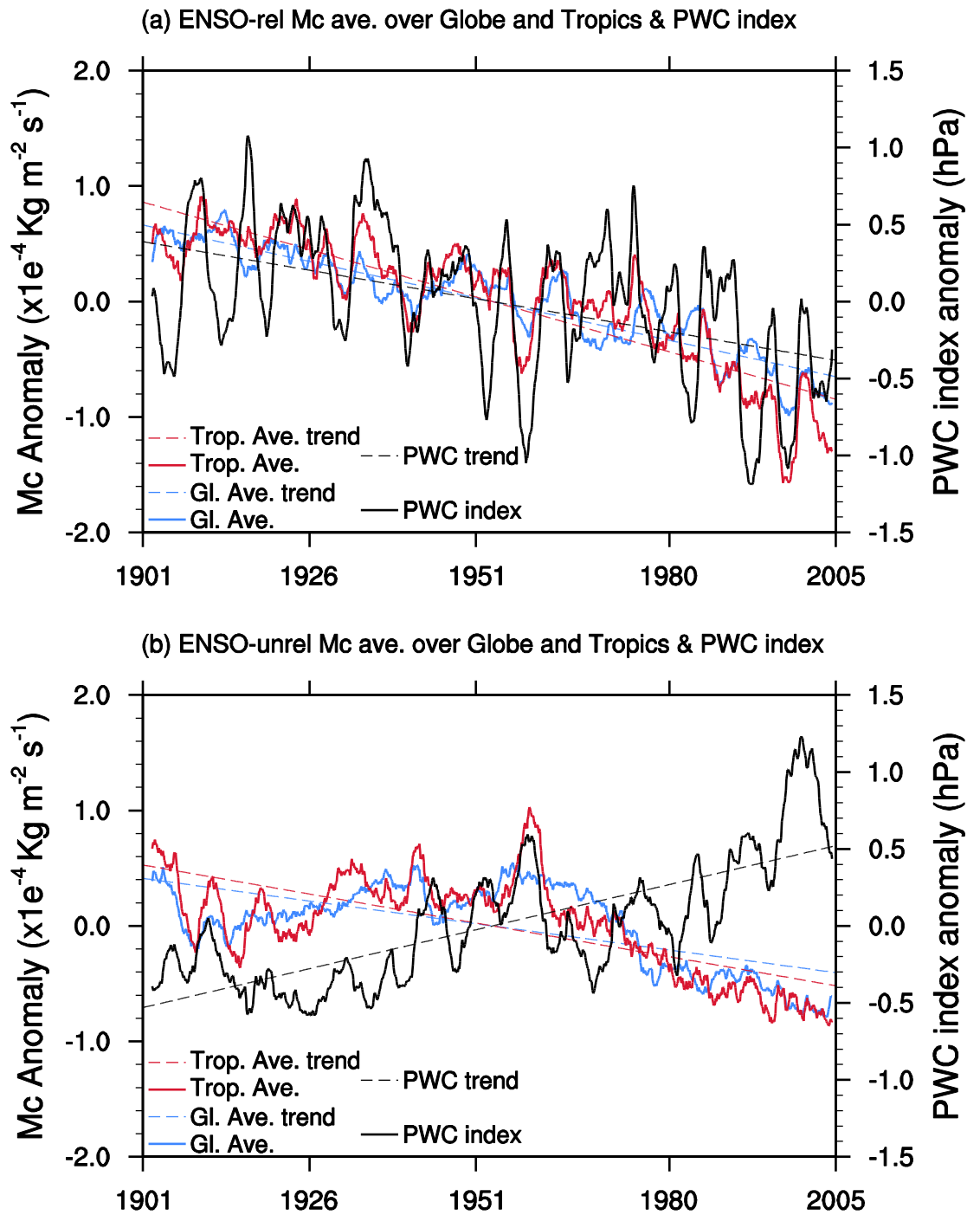
774 **Figure 2:** Ensemble mean PWC index (black curve) and ΔTS (red
 775 dashed curve) from CAM4 simulations of (a) ENSO-related SST and
 776 (b) ENSO-unrelated SST. Blue (green dashed) line shows the least-
 777 squares linear trend fit to the PWC (ΔTS).



778

779 **Figure 3:** PWC index (black curve) and Convective mass flux
 780 anomaly M_c averaged over the globe (blue curve) and the tropics (red
 781 curve) from (a) CTRL and (b) GISS-E2-R, and (c) CMIP5
 782 simulations. Black dashed line shows the least-squares linear trend fit
 783 to the PWC. Red (blue) dashed line shows the trend fit to the tropical
 784 (global) average M_c .

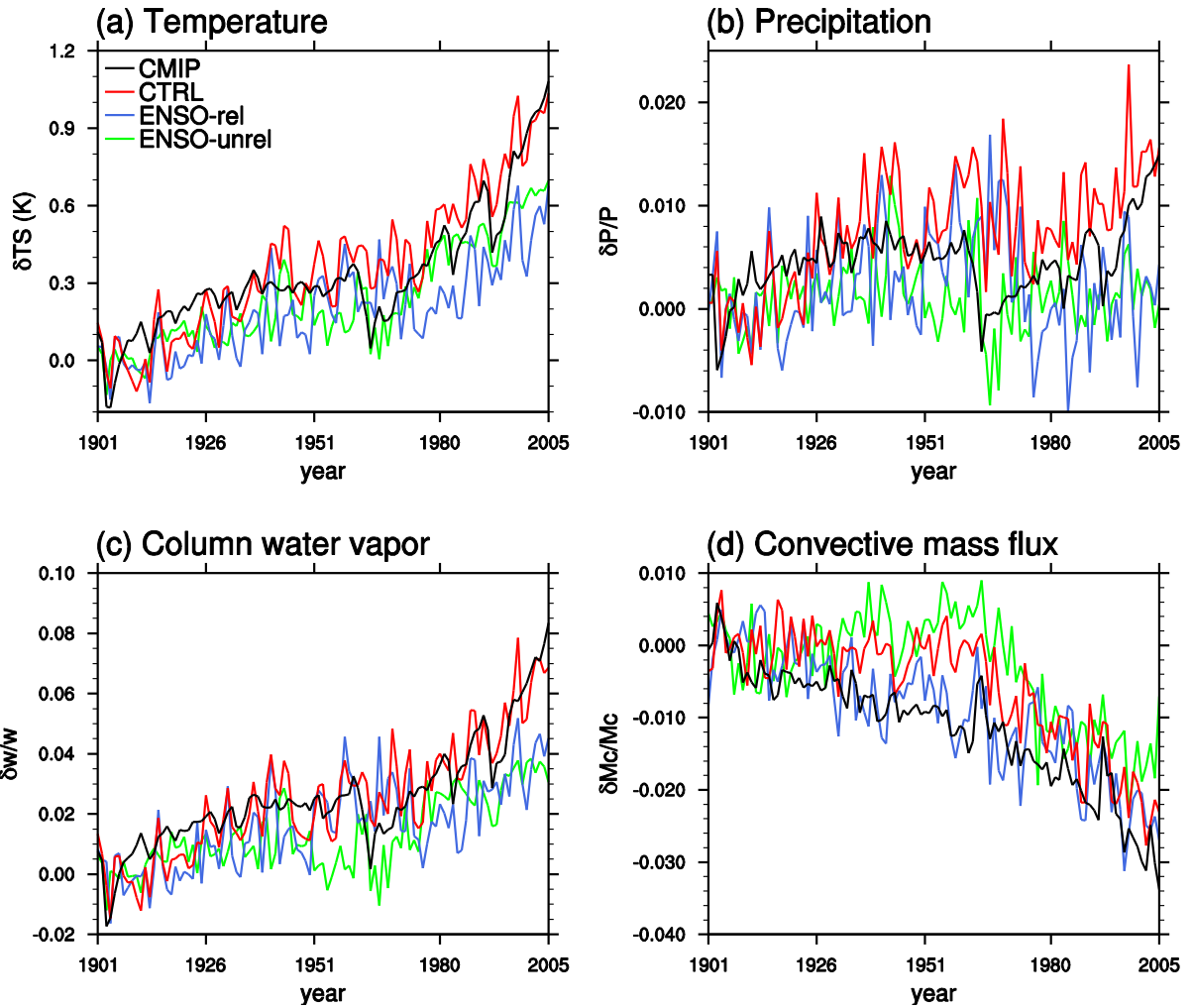
785



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787 **Figure 4:** Same as Fig. 3, except for CAM4 simulations with ENSO-

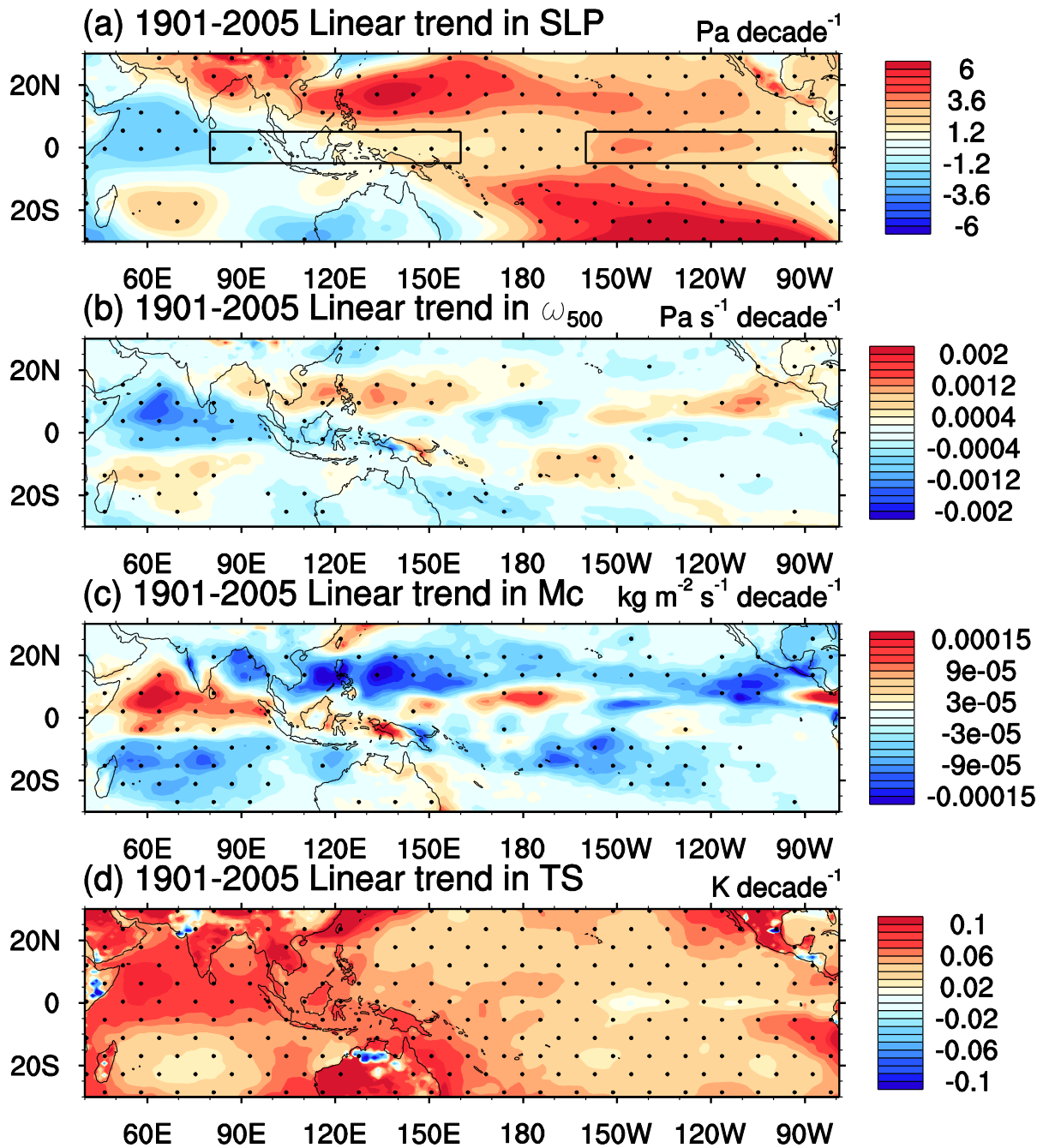
788 related and ENSO-unrelated SSTs



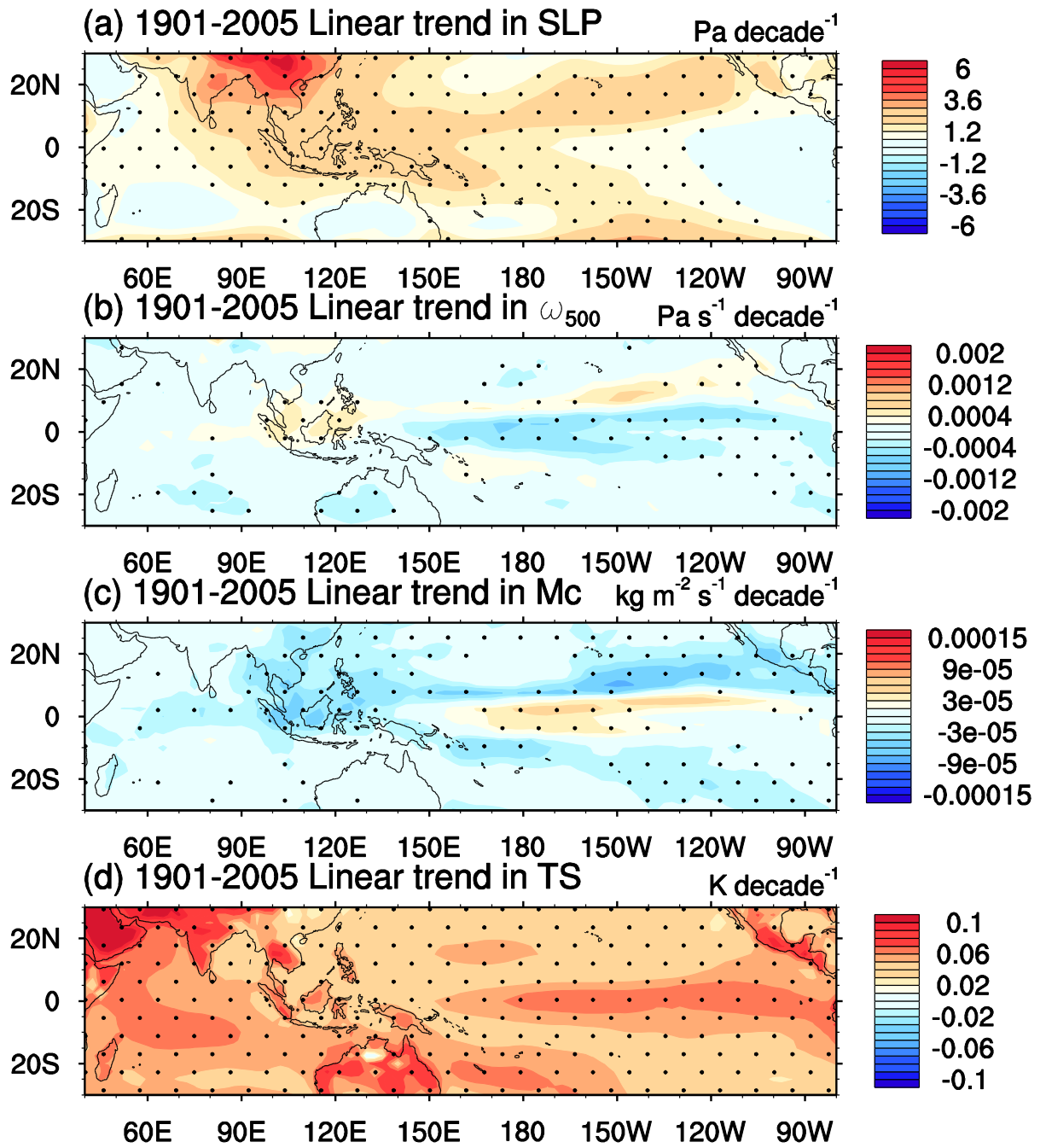
789

790 **Figure 5:** Time series of global mean changes in (a) temperature (K),
 791 (b) precipitation, (c) column integrated water vapor, and (d)
 792 convective mass flux. All quantities except temperature are expressed
 793 as fractional change relative to 1901 – 1910 mean. All quantities are
 794 annual means and are computed by first global averaging, and then
 795 differencing with the average computed over 1901 – 1910. Curves
 796 show results from the ensemble means of (black) CMIP5; (red) CAM4
 797 CTRL; (blue) CAM4 forced with ENSO-related SSTs; (green) CAM4
 798 forced with ENSO-unrelated SSTs.

799

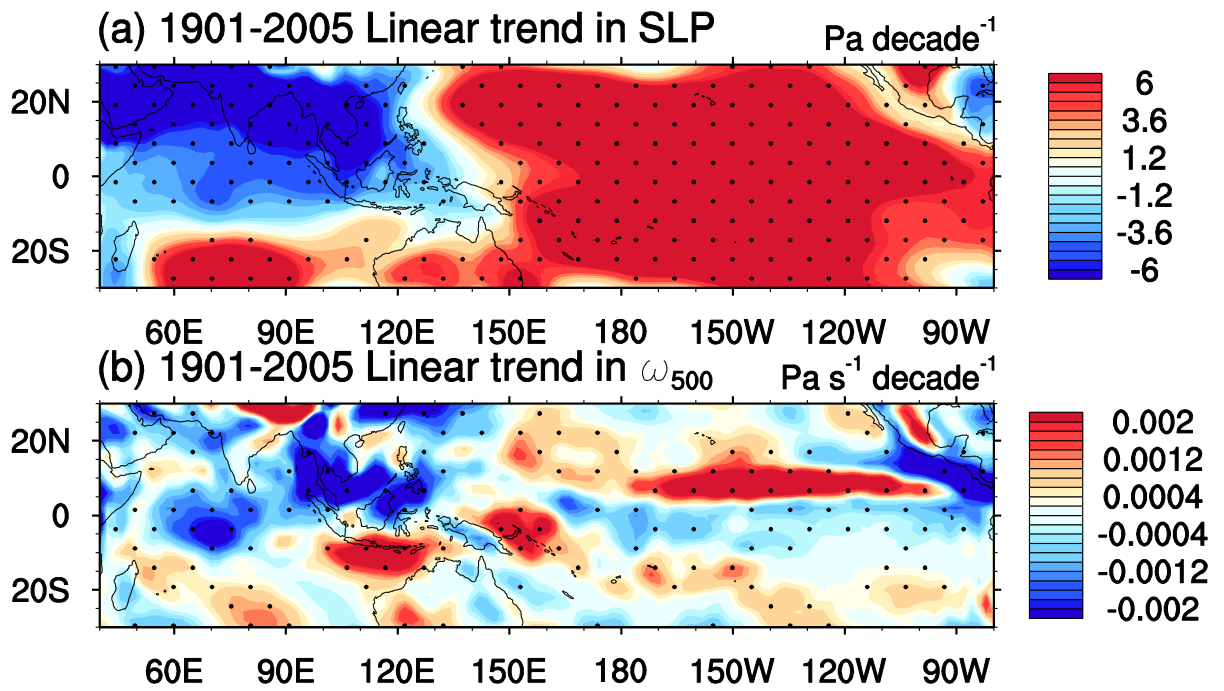


800
 801 **Figure 6:** Spatial patterns of 1901-2005 trend in (a) SLP, (b) vertical
 802 velocity at 500 hPa, (c) convective mass flux, and (d) surface
 803 temperature from CTRL simulations. Stippling show trends that are
 804 statistically significant at 5% level, as revealed by a t-test. The boxes
 805 in (a) show the regions, over which SLP and TS are averaged for
 806 PWC index and Δ TS calculations. Negative values in 5b indicates
 807 upward motion.

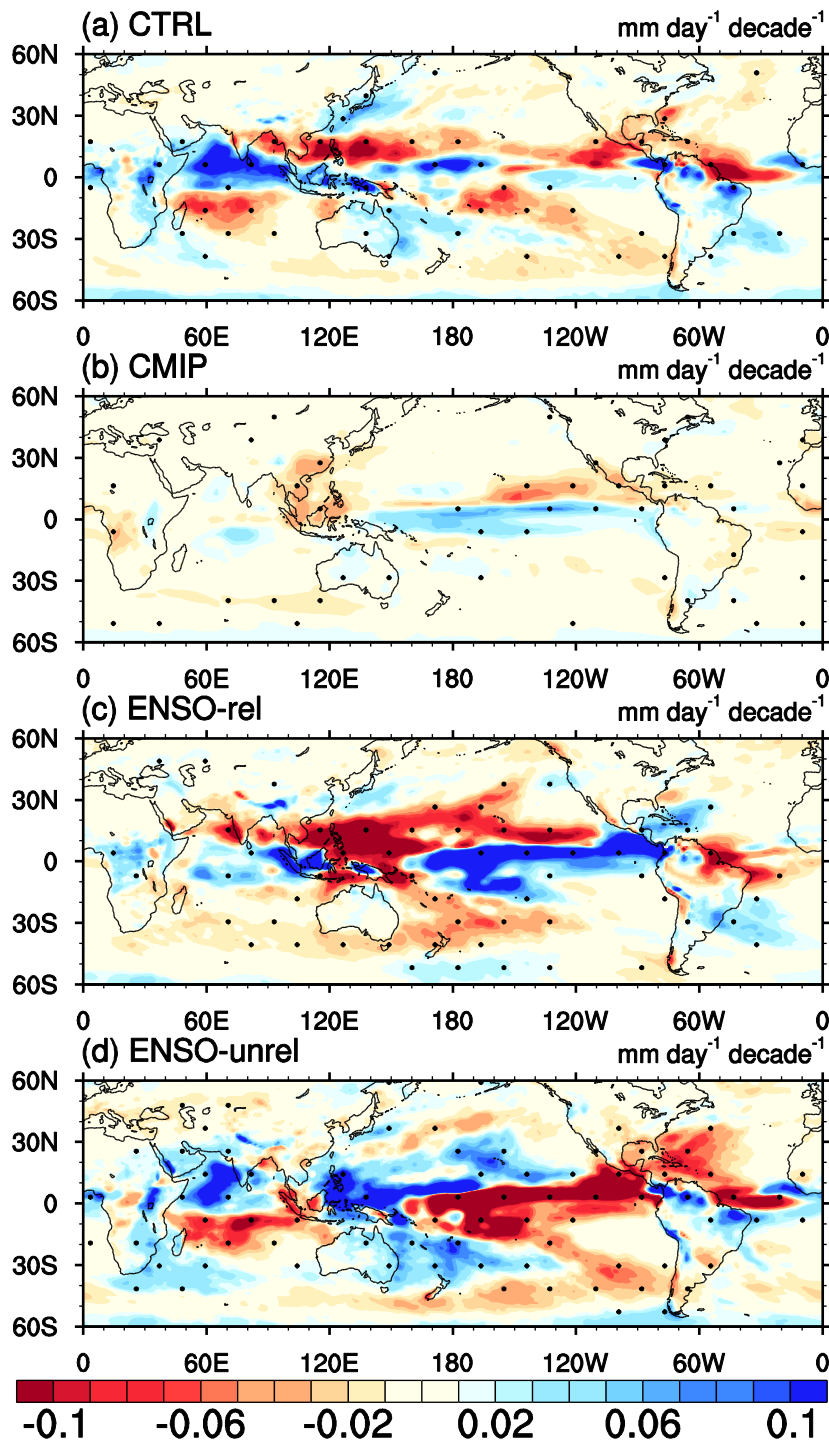


808

809 **Figure 7:** Same as Fig. 6, except for CMIP5 simulations



811 **Figure 8:** Spatial patterns of 1901-2005 trend in (a) SLP, (b) vertical
 812 velocity at 500 hPa calculated from 20CR. Stippling show trends that
 813 are statistically significant at 5% level, as revealed by a t-test.



814

815 **Figure 9:** Spatial Patterns of trends in 1901-2005 precipitation from
 816 (a) CTRL, (b) CMIP5, (c) ENSO-related, and (d) ENSO-unrelated
 817 simulations.