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2 Asymmetric Responses of the Western Tropical Pacific Sea Level to  
3 El Niño and La Niña

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16    **Key Points.**

- 17        1 )        The responses of the western tropical Pacific sea level to ENSO are  
18                   obviously asymmetric, and the response to El Niño is evidently stronger.
- 19        2 )        The different surface wind anomaly structure between El Niño and La Niña  
20                   is critical for the asymmetric response of sea level.
- 21        3 )        The asymmetric ocean responses to ENSO may contribute to the ENSO  
22                   asymmetry.

## 23    **Abstract**

24            The western tropical Pacific (WTP) exhibits large interannual sea level  
25    anomalies (SLAs), and the sea level falling in El Niño is evidently stronger than the  
26    rising in La Niña. The asymmetry is most prominent near 160°E with the response to  
27    El Niño larger by three times and becomes less obvious near the western boundary.  
28    Sensitivity experiments of a simplified ocean model suggest that the asymmetry in  
29    surface wind forcing structure between El Niño and La Niña is critical. The El Niño's  
30    westerly wind anomaly patch locates more east than the La Niña's easterly wind patch  
31    during the mature stage, and its upwelling effects are accumulated over a wider  
32    longitude range and cause stronger negative SLAs in the WTP. Near the western  
33    boundary, however, upwelling effects are attenuated by easterly wind anomalies  
34    during El Niño conditions. The asymmetric ocean responses to ENSO winds may  
35    participate in the asymmetry of ENSO cycle.

## 36    **Plain Language Summary**

37            ENSO is the most influential climate variability mode of the Pacific and causes  
38    strong interannual sea level anomalies (SLAs) in the western tropical Pacific (WTP).  
39    We notice that the WTP's sea level falling in El Niño condition is stronger than its sea  
40    level rising in La Niña. This difference is most prominent near 160°E, where the  
41    falling in El Niño is stronger by three times. This phenomenon becomes much less  
42    evident near the western boundary of the Pacific basin. We find that the difference in  
43    surface wind anomaly structures between El Niño and La Niña is the primary cause.  
44    El Niño's westerly wind anomaly center locates more east than La Niña's easterly

45 wind anomaly center in their mature phase, and there are easterly wind anomalies  
46 emerging near the western boundary during El Niño. Therefore, effect of El Niño's  
47 westerly wind anomaly is accumulated over a wider longitude range and causes  
48 stronger sea level falling in the WTP. But this effect becomes weaker near the western  
49 boundary, due to easterly wind anomalies there. By contrast, effect of La Niña's  
50 easterly wind anomaly strengthens monotonically from east to west, but the produced  
51 sea level rising signatures are mainly confined to the WTP.

52 **Keywords**

53 Western Tropical Pacific, Sea Level, ENSO, ENSO asymmetry, Interannual

54 Variability

## 55    **1. Introduction**

56        Regional sea level change, as an essential aspect of climate change, is attracting  
57    increasing attention of scientific communities and the general public, because of its  
58    threats on eco-systems and people of coastal residence (e.g., Nicholls & Cazenave,  
59    2010; Cazenave & Cozannet, 2014). There is a paramount demand for improved  
60    understanding of regional sea level changes on various timescales (e.g., Milne et al.,  
61    2009; Church et al., 2013; Stammer et al., 2013). Sea level changes over the tropical  
62    Pacific Ocean are particularly pronounced on interannual timescale, as largely  
63    modulated by El Niño-Southern Oscillation (ENSO) (e.g., Cazenave et al., 2008;  
64    Antonov et al., 2005; Cheng et al., 2008). It is evident in Figure 1a that the western  
65    tropical Pacific (WTP) shows stronger interannual sea level anomalies (SLAs) than  
66    other regions, as quantified the standard deviation of low-passed SLAs. The WTP  
67    shows strong sea level rising (falling) during the La Niña (El Niño) condition, as a  
68    result of prevailing easterly (westerly) wind anomalies over the tropical Pacific basin  
69    (Wyrtki, 1975; Zebiak, 1984; Alory & Delcroix, 2002; Gu & Li, 2009; Nerem et al.,  
70    2010; Merrifield, 2011; Zhang & Church, 2012; Chang et al., 2013; Becker et al.,  
71    2016; Hamlington et al., 2016; Wang, 2018). The sea level rising during La Niña  
72    conditions is expected to aggravate coastal erosion, extreme marine flooding, or  
73    saltwater intrusion in coastal aquifers (Nicholls & Tol, 2006; Nicholls et al., 2007;  
74    Nicholls & Cazenave, 2010), thus becomes a major threat for the densely populated  
75    coastal regions. However, the sea level falling during El Niño causes the coral reef  
76    exposure and could damage the region's ecosystem. There are many

77 heavily-populated coasts and islands in the WTP, so the sea level changes during  
78 ENSO are well worth studying.

79 El Niño and La Niña are, however, not mirrors of each other. They exhibit  
80 obvious asymmetry in magnitude, duration time, and occurrence frequency (e.g., Yu  
81 et al., 1997; Burgers & Stephenson, 1999; Ohba & Ueda, 2007; Gergis & Fowler,  
82 2009; An & Choi, 2009; Okumura & Deser, 2010). Their signatures on sea level are  
83 also asymmetric, owing to not only ENSO's asymmetry but also the nonlinearity in  
84 ocean response (Niedzielski & Kosek, 2010; Swierczynska et al., 2013; Im et al.,  
85 2015; An & Kim, 2017). Previous studies have demonstrated that ENSO asymmetry is  
86 contributed by nonlinear dynamical heating of ocean current advection (e.g., Kang &  
87 Kug, 2002; Jin et al., 2003; An & Jin, 2004; Su et al., 2010). As dynamically  
88 associated with the upper-ocean currents through pressure gradient, the asymmetric  
89 SLAs induced by ENSO may affect ocean current advection of heat and thereby  
90 participate in the ENSO asymmetry.

91 In this study, we aim to investigate the asymmetric response of the WTP sea  
92 level to ENSO and the underlying dynamical processes. This effort is of paramount  
93 need for the prediction and projection of regional sea level change and adaptations of  
94 the low-lying coasts and islands, as well as for understanding the dynamics of ENSO  
95 asymmetry. The rest of the paper is structured as follows. Section 2 describes the data  
96 and models. Section 3 describes the asymmetric SLAs of the WTP in response to  
97 ENSO. Section 4 explores the underlying dynamics through a simplified ocean model.  
98 Section 5 presents concluding remarks.

## 99    **2. Data and Models**

### 100    **2.1. Data**

101        In this study we use  $0.25^\circ \times 0.25^\circ$ , monthly satellite SLA data of 1993-2016  
102    from Archiving Validation, and Interpretation of Satellite Oceanography (AVISO; Le  
103    Traon et al., 1998) and sea level records of three tidal gauge stations in the WTP:  
104    Malakal at  $7.33^\circ\text{N}$ ,  $134.45^\circ\text{E}$  for 1979-2016, Kapingamrangi at  $1.1^\circ\text{N}$ ,  $154.78^\circ\text{E}$  for  
105    1929-2016, and Lombrum at  $2.04^\circ\text{S}$ ,  $147.47^\circ\text{E}$  for 1995-2016 (Figure S1 in the  
106    Supporting Information). The Hadley Centre Sea Ice and Sea Surface Temperature  
107    (HadISST) data set (Kennedy et al., 2011) of the Met Office during 1979-2016 with  
108    horizontal resolution of  $1^\circ \times 1^\circ$  is used to compute Niño-3.4 index and identify El  
109    Niño and La Niña conditions. Surface winds and other surface atmospheric fields of  
110    1979-2016 are taken from the  $0.75^\circ$  monthly dataset of the European Centre for  
111    Medium-Range Weather Forecasts (ECMWF) ERA-interim (Dee et al., 2011).

### 112    **2.2. HYCOM**

113        The HYbrid Coordinate Ocean Model (HYCOM) version 2.2.18 (Bleck, 2002)  
114    are used to simulate interannual SLAs in the tropical Pacific Ocean. HYCOM is  
115    configured to the Pacific Ocean basin between  $48^\circ\text{S}$ - $48^\circ\text{N}$ ,  $110^\circ\text{E}$ - $70^\circ\text{W}$ , with a  
116    horizontal resolution of  $1/3^\circ \times 1/3^\circ$  and 26 hybrid vertical layers (Li et al., 2015; Ren et  
117    al., 2020). Surface atmospheric forcing fields (winds, heat fluxes, precipitation, etc.)  
118    are taken from ERA-interim. Three sponge layers are applied to the western, southern  
119    and northern open-ocean boundaries, where model temperature and salinity are  
120    related to the World Ocean Atlas 2009 (WOA09) climatology (Antonov et al., 2010).

121 More details of model configuration are described in Ren et al. (2020). Subsequent to  
122 the spin-up run of 30 years under monthly climatologic forcing, two parallel  
123 experiments are performed under daily ERA-Interim fields for the period of  
124 1979-2016. The control run (HYCOM-CTL) is forced with the original daily  
125 atmospheric fields and assumed to contain the complete processes. This experiment  
126 can well reproduce the amplitude and spatial distribution of interannual SLA and  
127 upper-ocean circulation variation in observation (Figure 1b; Ren et al., 2020). Another  
128 experiment, HYCOM-TAU, uses daily wind stress forcing as HYCOM-CTL, but all  
129 the other forcing fields (heat and freshwater fluxes) are fixed to monthly climatology.  
130 HYCOM-TAU is used to evaluate the effects of wind forcing on sea level.

### 131 **2.3. Reduced-Gravity Ocean Model**

132 To achieve more in-depth understanding, a series of experiments are performed  
133 with a 1.5-layer nonlinear reduced-gravity ocean (RGO) model. This model mainly  
134 represents the 1<sup>st</sup>-mode baroclinic response of the ocean to surface wind forcing,  
135 which is the dominant source of large-scale interannual variability in sea level and  
136 upper-ocean circulation of the WTP (e.g., Qiu & Chen, 2010, 2012). The model is  
137 configured to the Pacific Ocean basin between 40°S-40°N, 100°E-70°W with  
138 horizontal resolutions of 0.25°×0.25° and forced by monthly ERA-Interim surface  
139 winds. Readers are referred to Duan et al. (2019) for more details of the model  
140 configuration. After a spin-up of 10 years under climatological wind forcing, the  
141 control run of RGO model (RGO-CTL) is forced by realistic monthly winds. RGO  
142 model experiments (Table S1) are forced by idealized wind forcing to examine the



143 role of wind forcing asymmetry and are described in Section 4.

### 144 **3. Asymmetric Responses to El Niño and La Niña**

145 To highlight the interannual variations associated with ENSO, we analyze the  
146 13-month low-pass filtered anomaly fields with the monthly climatology removed.  
147 Skewness  $S$  is a measure of the distribution asymmetry with  $S = 0$  indicating a normal  
148 distribution (White, 1980). Figure 1c shows the skewness of the observed SLA over  
149 the tropical Pacific for 1993-2016. The eastern tropical Pacific is positively skewed  
150 with the maximum  $S$  of  $\sim 2.0$ , while the central-western tropical Pacific is negatively  
151 skewed with the minimum  $S$  of  $-2.0$  and a horseshoe structure extending from the  
152 equator to extratropical regions in both hemispheres. This distribution of SLA  
153 skewness resembles that of SST anomaly (An & Jin, 2004; Niedzielski & Kosek,  
154 2010) and is likely associated with the positive skewness of ENSO (Nerem et al.,  
155 2010). The El Niño condition is characterized by positive SLAs in the eastern Pacific  
156 and negative SLAs in the WTP, and these anomalies are stronger in amplitude than  
157 the opposite SLAs occurring in La Niña condition (Niedzielski & Kosek, 2010;  
158 Figure S1).

159 In addition to the asymmetry residing in ENSO (e.g., as quantified by the  
160 skewness of Niño-3.4 index), the asymmetric responses of sea level to El Niño and La  
161 Niña also contribute to the SLA asymmetry shown in Figure 1c. We regress SLAs  
162 onto the normalized Niño-3.4 index separately for the El Niño condition (Niño-3.4 >  
163 0) and the La Niña condition (Niño-3.4 < 0). The corresponding regression  
164 coefficients, namely  $k_{\text{Niño}}$  and  $k_{\text{Niña}}$ , are used to quantify the responses of SLA to El

165 Niño and La Niña, respectively (Figures 1d and 1e). The response time of sea level to  
 166 ENSO shows spatial variation, as indicated by the lead-lag correlation (Figure S2a).  
 167 For each grid point, the lead-lag time of the maximal correlation is used to compute  
 168 the regression coefficient. The results are not dramatically different from those of  
 169 simultaneous regression (Figure S2b). Figures 1d and 1e show distributions of  $k_{\text{Nino}}$   
 170 and  $k_{\text{Nina}}$ . In the WTP, the maximal  $k_{\text{Nino}}$  of  $\sim 0.2$  m is located  $20^\circ$ - $30^\circ$  away from the  
 171 western boundary, and  $k_{\text{Nino}}$  decreases in magnitude as approaching the western  
 172 boundary. By contrast, the peak  $k_{\text{Nina}}$  values of  $\sim 0.15$  m are close to the western  
 173 boundary. We further use the ratio of  $k_{\text{Nino}}$  to  $k_{\text{Nina}}$  to quantify the response asymmetry,  
 174 
$$R_k = \frac{k_{\text{Nino}}}{k_{\text{Nina}}}. \quad (1)$$
  
 175  $R_k = 1$  denotes symmetric response of SLAs to El Niño and La Niña, while  $R_k > 1$  and  
 176  $R_k < 1$  indicates stronger and weaker response to El Niño than to La Niña, respectively.  
 177 As shown in Figure 1f,  $R_k$  reaches the largest value near  $160^\circ\text{E}$  with values exceeding  
 178 3.0, indicating that the response of SLA to El Niño is stronger by at least 3 times than  
 179 the response to La Niña.  $R_k$  is weakened to  $\sim 1.0$  near the western boundary, implying  
 180 SLA responses there are nearly symmetric. This interesting distribution of  $R_k$  in the  
 181 WTP and underlying dynamics are worthy of systematic investigation. One may  
 182 notice that  $R_k$  is  $< 1$  in the central Pacific and is  $> 1$  in the eastern Pacific, indicative  
 183 of prevailing asymmetry over the tropical Pacific basin. In the following, we focus on  
 184 explaining the  $R_k$  distribution in the WTP.

185 Simulations of HYCOM and RGO model have faithfully reproduced the  
 186 observed interannual variations of sea level at three tidal gauge stations (Figure S3).

187 The upper-layer thickness (ULT) anomaly of the 1.5-layer RGO model is a good  
188 proxy of SLA in the tropical Pacific (e.g., Qiu & Chen, 2010, 2012; Chang et al., 2013;  
189 Duan et al., 2019). The correlations among AVISO, HYCOM-CTL, HYCOM-TAU  
190 and RGO-CTL at the three tidal gauge stations are all above 0.85. The asymmetric  
191 responses of SLAs to El Niño and La Niña in the WTP during 1993-2016 can be  
192 realistically reproduced by HYCOM-CTL, HYCOM-TAU, and RGO-CTL (Figures  
193 2a-2c), although RGO-CTL fails to capture the features in the eastern Pacific. The  
194 good performance of HYCOM-TAU and RGO-CTL indicates that the interannual  
195 SLAs in the WTP and their asymmetric features are primarily the results of ENSO  
196 wind forcing, and the underlying dynamics can be explored by sensitive model  
197 experiments with prescribed wind forcing fields.

#### 198 **4. Dynamics**

199 To include more ENSO events, we use the period of 1979-2016 to perform  
200 model experiments (Table S1), although  $R_k$  of 1979-2016 shows detailed differences  
201 from that of 1993-2016 in the northwest Pacific (Figures 2d-2f). According to existing  
202 studies of ENSO asymmetry, the different spatial and temporal characteristics of wind  
203 anomalies are essential to cause the SST asymmetry between El Niño and La Niña  
204 (Kang & Kug, 2002; An & Kim, 2017). To examine which aspect of ENSO's wind  
205 forcing is critical in regulating the asymmetric responses of SLAs, we adopt a  
206 statistical model based on the singular value decomposition (SVD) of wind stress and  
207 SST (Kang & Kug, 2000), which is expressed as

$$208 \quad \mathbf{\tau}'(x, y, t) = \sum_n^N c(n) [\sum_{x,y} V_{\text{SST}}(x, y, n) T(x, y, t)] V_\tau(x, y, n), \quad (2)$$

209 where  $x$ ,  $y$ , and  $t$  represent longitude, latitude, and time, respectively,  $V_{\text{SST}}$  and  $V_{\tau}$  are  
 210 the SVD singular vectors for SST and wind stress,  $T$  is the SST anomaly field, and  $n$   
 211  $= 1, 2, \dots, N$  indicates the  $n$ th mode of SVD,  $\sum_{x,y}$  indicates spatial integration over  
 212 the region of 100°-290°E, 40°S-40°N.  $c(n)$  represents the correlation between SST  
 213 and wind stress anomalies,

$$214 \quad c(n) = \frac{\sum_t t_{\text{SST}}(t,n)t_{\tau}(t,n)}{\sum_t t_{\text{SST}}(t,n)^2}, \quad (3)$$

215 where  $t_{\text{SST}}$  and  $t_{\tau}$  are the corresponding time series, the numerator is the covariance  
 216 of  $t_{\text{SST}}$  and  $t_{\tau}$  and the denominator is the variance of  $t_{\text{SST}}$ , and  $\sum_t$  indicates temporal  
 217 integration from January 1979 to December 2016. Zonal and meridional components  
 218 of  $\tau'$  are separately computed. All the RGO sensitive experiments are forced by  
 219 monthly wind stress anomalies constructed by Eq. 2 (representing ENSO wind  
 220 forcing) plus monthly climatological winds.

221 We first perform two experiments, namely EXP1 and EXP2. EXP1 uses different  
 222 wind stress anomaly fields for El Niño and La Niña conditions. Specifically,  $\tau'_{\text{Niño}}$  and  
 223  $\tau'_{\text{Niña}}$  are reconstructed separated for Niño-3.4  $\geq 0$  and Niño-3.4  $< 0$  conditions  
 224 (Figures S4a and S4b) using Eq. 2, respectively, so that the synthesized  $\tau'$  still retains  
 225 the difference in spatial structure between El Niño and La Niña conditions. By  
 226 contrast, EXP2 does not distinguish El Niño and La Niña conditions and uses  
 227 reconstructed  $\tau'$  for the entire model period (Figure S4c). As such, the difference  
 228 between EXP1 and EXP2 represents the effect of different wind anomaly structures  
 229 between El Niño and La Niña on SLAs. In EXP1 and EXP2, we use only the leading  
 230 mode ( $n = 1$ ) of SVD to reconstruct  $\tau'$  (Figure S4), which explains  $> 85\%$  of the total

231 covariance and mainly represents ENSO's mature phase (Figure S5). The higher SVD  
232 modes largely represent the transition stages between El Niño and La Niña polarities  
233 (Figure S5) and have limited impacts on the asymmetry of SLAs (Figure S6).

234       Figures 3a and 3b show  $R_k$  distributions of ULT produced by EXP1 and EXP2,  
235 respectively. EXP1 is able to reproduce large  $R_k$  values in the WTP as RGO-CTL,  
236 whereas  $R_k$  is generally close to 1.0 in EXP2. These results suggest that the  
237 asymmetry in surface wind structures between El Niño and La Niña is largely  
238 responsible for asymmetric responses of the WTP sea level. Note that EXP2 still  
239 retains some asymmetric characteristics of ENSO winds, such as the asymmetries in  
240 intensity, frequency, and temporal evolution. Figure 3b however indicates that these  
241 factors have little contributions to SLA asymmetry. This is confirmed by an additional  
242 experiment EXP3, which adopts an idealized sine time series for  $\tau'$  (Table S1) and  
243 achieves similar results to EXP2 (Figure S7).

244       In the equatorial zone, zonal component of wind stress  $\tau^x$  is much more  
245 influential for the ocean than the meridional component. We repeat EXP1 and EXP2  
246 using only  $\tau^x$  anomaly, and  $\tau^y$  is fixed to monthly climatology (EXP1-TAUX and  
247 EXP2-TAUX). The results of two experiments achieve are broadly consistent with  
248 EXP1 and EXP2 (Figures 3c and 3d). Therefore, the spatial structure of  $\tau^x$  is critical  
249 for the asymmetric responses. The typical structures of ULT and  $\tau^x$  in EXP1-TAUX  
250 for warm and cold phases are shown in Figures 3e and 3f, respectively, as represented  
251 by the leading SVD mode of  $\tau^x$  and ULT. Notice that westerly wind anomaly patch of  
252 El Niño locates more east than the easterly wind anomaly patch of La Niña in their

253 mature phase, and correspondingly the zero value of ULT anomaly of El Niño also  
 254 locates more east (Jin, 1997; Kang & Kug, 2002). During El Niño there are easterly  
 255 wind anomalies near the western boundary. The strong negative ULT anomalies in El  
 256 Niño are seen over a wide longitude range and weaken as approaching the western  
 257 boundary, while the positive ULT anomalies in La Niña are confined to the far WTP  
 258 and generally strengthen westward.

259 To better elucidate how the wind forcing structure cause asymmetric SLAs, we  
 260 show in Figure 4 the zonal distributions of  $\tau^x$  and ULT in the equatorial band  
 261 ( $5^\circ\text{S}$ - $5^\circ\text{N}$ ). It is clearly discernible in Figure 4a that the El Niño's westerly wind  
 262 anomaly patch locates more east than the La Niña's easterly wind anomaly patch. The  
 263 forcing effect of equatorial zonal wind stress on sea level and ULT slopes can be  
 264 roughly expressed in a linear relationship (Sverdrup, 1947; McCreary, 1977; Alory &  
 265 Delcroix, 2002; Palanisamy et al., 2014),

$$266 \quad \frac{dh}{dx} = \tau^x, \quad (4)$$

267 where  $h$  is ULT anomaly or SLA. Therefore,  $h(x)$  at a given longitude  $x$  can be  
 268 determined by the integration of Eq. (4) from the eastern boundary  $x_E = 70^\circ\text{W}$  along  
 269 equator,

$$270 \quad h(x) = h(x_E) + \int_{x_E}^x \tau^x dx. \quad (5)$$

271 Since  $h(x_E)$  diffuses quickly away from the eastern boundary as free Rossby waves  
 272 and has little impact on the interior Pacific (Qiu et al., 2013),  $h(x)$  is primarily  
 273 determined by the zonal integral of  $\tau^x$  from  $x_E$  to  $x$  (second term). As such, the  
 274 negative ULT anomaly in El Niño in the WTP is much stronger than the positive ULT

275 anomaly in La Niña owing to the much wider westerly wind anomaly patch to its east.  
 276 West of 160°E, easterly wind anomaly near the western boundary causes downwelling  
 277 of the ocean, resulting in the attenuation of negative ULT anomalies there. By contrast,  
 278 the La Niña's positive ULT anomalies are continuously strengthened by easterly wind  
 279 anomalies. As a result, the ULT anomalies of El Niño and La Niña are comparable in  
 280 amplitude in the far WTP, and the asymmetry of response is no longer evident there  
 281 (right panel of Figure 4a). In EXP2-TAUX (Figure 4b), without the difference in wind  
 282 forcing structure (left panel), the asymmetric responses cannot be reproduced (middle  
 283 and right panels).

284 Figure 4c shows the zonal distributions of ULT anomaly and regression  
 285 coefficients predicted by the linear theory, which compare favorably with Figure 4a. It  
 286 indicates that the linear theory can to a large extent capture the processes causing  
 287 asymmetric responses to ENSO winds and confirms the critical role played by the  
 288 structure of zonal wind anomaly. Under this theoretical framework, the response  
 289 asymmetry  $R_k$  can be theoretically expressed as,

$$290 \quad R_k(x) \approx \frac{\int_{x_E}^x \tau_{Nino}^x dx}{\int_{x_E}^x \tau_{Nina}^x dx}, \quad (6)$$

291 where  $\tau_{Nino}^x$  and  $\tau_{Nina}^x$  are the zonal wind stress anomaly in El Niño and La Niña  
 292 conditions, respectively. Eq. (6) clearly suggests the sensitivity of  $R_k(x)$  to the  
 293 distribution of  $\tau^x$  from  $x$  to the eastern boundary.

## 294 **5. Concluding Remarks**

295 The WTP exhibits large interannual variations of sea level, and the sea level

296 falling in El Niño is stronger than the rising in La Niña. Here we show that this  
297 asymmetry is most prominent near 160°E with the response to El Niño larger by ~3  
298 times and becomes much less obvious near the western boundary. RGO model  
299 experiments suggest that the asymmetric surface wind anomaly structure between El  
300 Niño and La Niña conditions is critical. El Niño's westerly wind anomaly patch  
301 locates more east than La Niña's easterly wind anomaly patch in their mature stages.  
302 As such, the upwelling effects of westerly wind anomalies are accumulated over a  
303 wider longitude range and cause stronger negative SLAs in the WTP. As approaching  
304 further toward the western boundary, positive SLAs in La Niña continue to amplify,  
305 while negative SLAs of El Niño are attenuated by easterly wind anomalies in the far  
306 WTP.

307       Here we reveal the sensitivity of the asymmetric SLAs in the WTP to the ENSO  
308 wind structures. It is interesting to investigate whether the asymmetric SLAs in turn  
309 contributes to the ENSO asymmetry in amplitude and temporal evolutions. This can  
310 be investigated through careful heat budget analysis that evaluate the effects of  
311 asymmetric current advection on SST variability. In addition, the zonal surface wind  
312 patch dominates the SLA asymmetry along the equator, and its off-equatorial structure  
313 may affect the asymmetry beyond the equator by modifying the wind stress curl (An  
314 & Bong, 2016). In addition to wind forcing, the effects of local nonlinear processes,  
315 such as mesoscale eddies, on sea level in the WTP are not resolved by the RGO  
316 model experiments (Chen et al., 2015; Qiu et al., 2015), which can be also examined  
317 in the future study.



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322    at <http://marine.copernicus.eu/services-portfolio/access-to-products/>. Tidal gauges data  
323    are available at <https://www.psmsl.org/>. ERA-Interim wind data are available at  
324    <https://apps.ecmwf.int/datasets/>. HadISST data are downloaded from Met-Office  
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493

494 **Figure captions**

495 **Figure 1.** Standard deviation (STD) of sea level anomalies (SLAs, m) of 1993-2016  
496 from (a) AVISO sea level product and (b) HYCOM-CTL. (c) Skewness of SLAs. (d)  
497 Regression coefficient  $k_{\text{Nino}}$  (m) of SLAs onto the normalized Niño-3.4 index for El  
498 Niño condition (Niño-3.4 > 0). (e) Same as (b) but for La Niña condition (Niño-3.4 <  
499 0). (f) Ratio of  $k_{\text{Nino}}$  to  $k_{\text{Nina}}$ , i.e.,  $R_k = k_{\text{Nino}}/k_{\text{Nina}}$ . SLA data in (c)-(f) are derived from  
500 AVISO sea level product of 1993-2016.

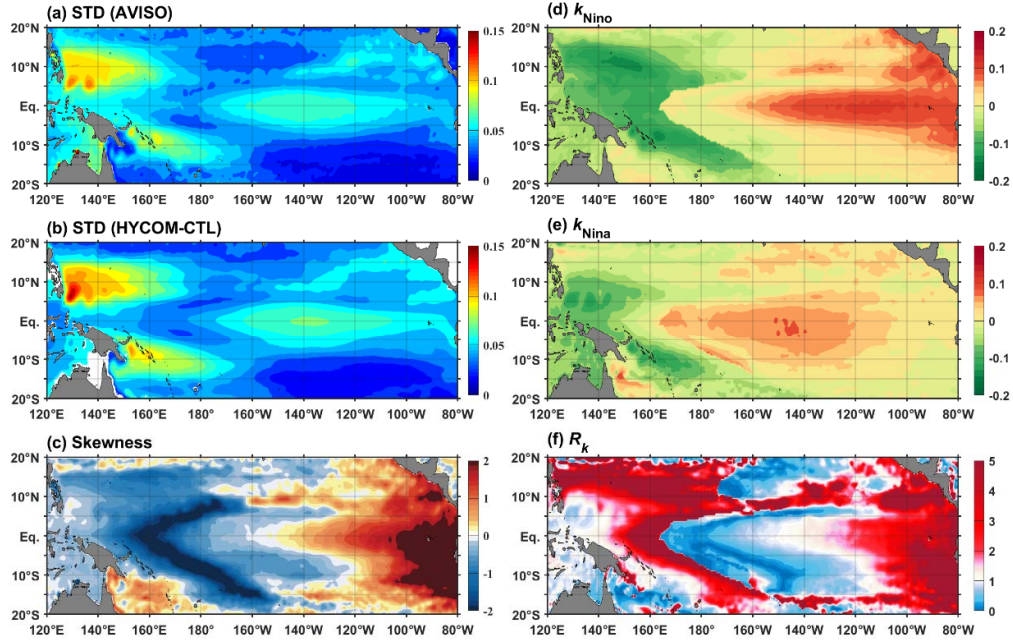
501 **Figure 2.**  $R_k$  distributions during 1993-2016 simulated by (a) HYCOM-CTL, (b)  
502 HYCOM-TAU, and (c) RGO-CTL. (d)-(f) are the same as (a)-(c), but for the  
503 1979-2016 period.

504 **Figure 3.**  $R_k$  distributions from (a) EXP1, (b) EXP2, (c) EXP1-TAUX and (d)  
505 EXP2-TAUX. (e) and (f) show the 1<sup>st</sup> singular value decomposition (SVD) modes for  
506 zonal wind stress  $\tau^x$  ( $\text{N m}^{-2}$ ; solid and dashed contours for positive and negative  
507 values) and upper-layer thickness ULT (m; color shading) of EXP1-TAUX1 for El  
508 Niño and La Niña conditions, respectively.

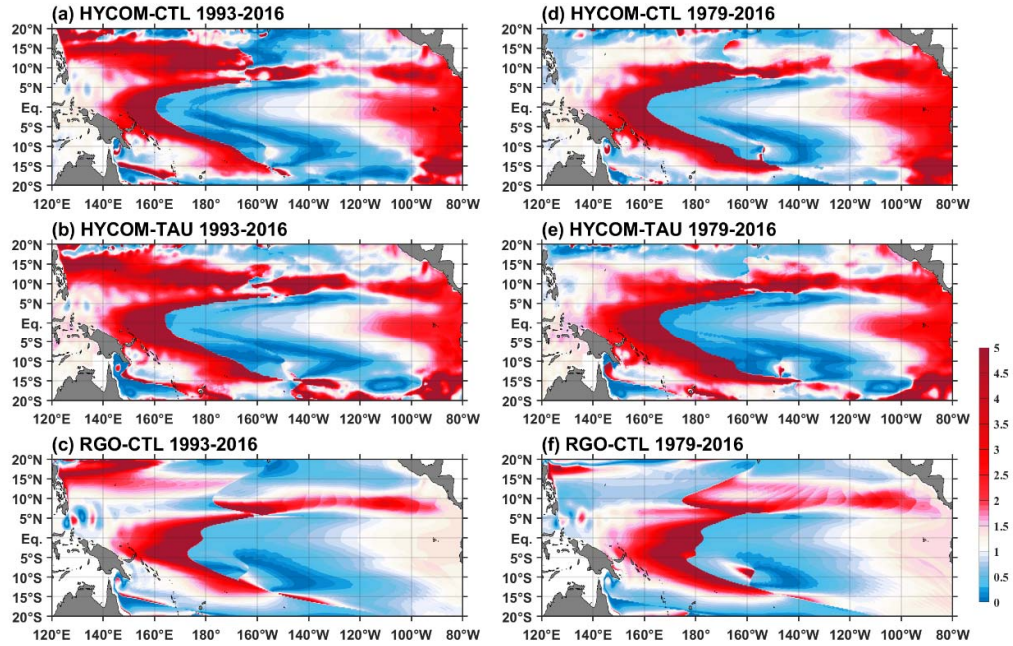
509 **Figure 4.** (a) Zonal structure of the 5°S-5°N average  $\tau^x$  (left), and ULT (middle) of  
510 the 1<sup>st</sup> SVD mode, and regression coefficients ( $k_{\text{Nino}}$  and  $k_{\text{Nina}}$ ; right) in EXP1-TAUX,  
511 computed separated for the for El Niño and La Niña conditions. (b) Same as (a) but  
512 for EXP2-TAUX. (c) is the same as (a), but for  $\tau^x$  of the 1<sup>st</sup> SVD mode in  
513 RGO-CTL, and theoretically-predicted ULT and regression coefficients (see the text  
514 for details), and they are normalized by the standard deviation.

515

516 **Figures**



517  
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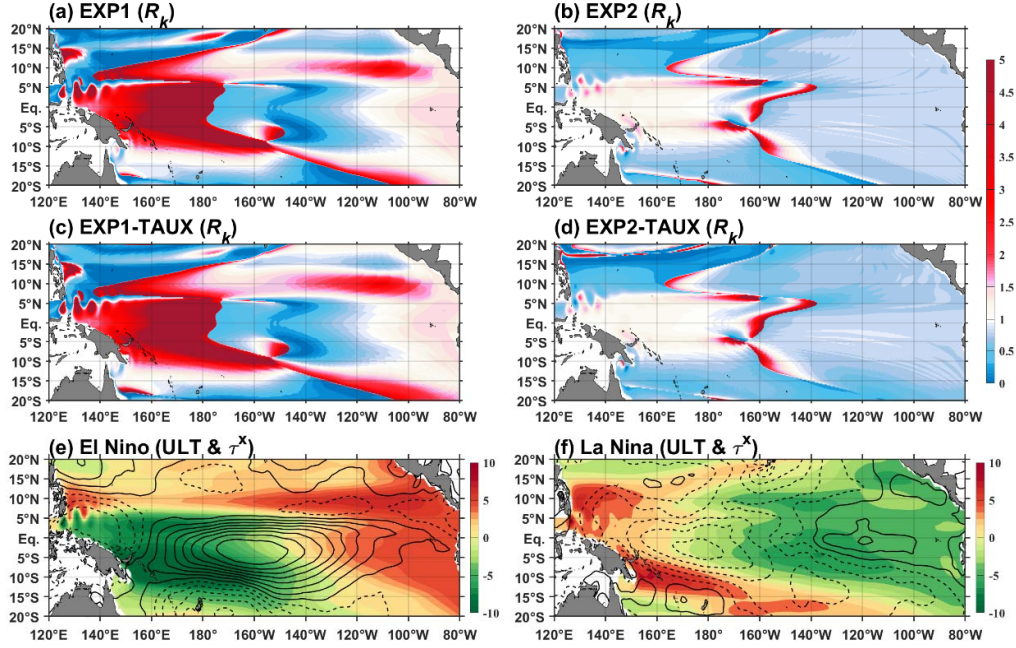


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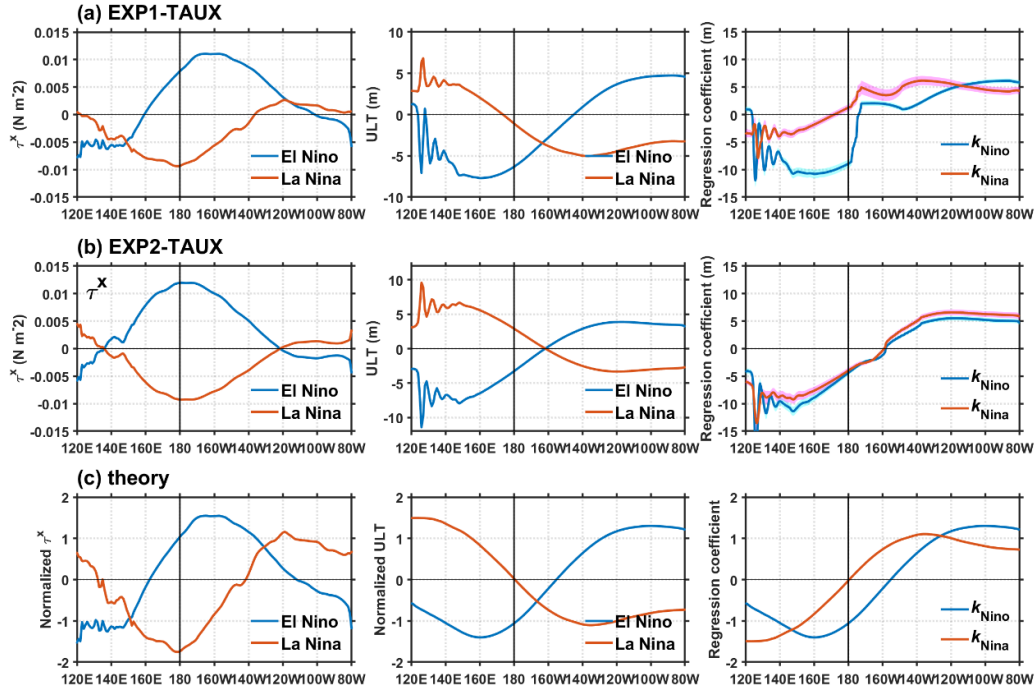
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