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2 **Main Manuscript for**

3 **Summertime Stationary Waves Integrate Tropical and Extratropical Impacts on**
4 **Tropical Cyclone Activity**

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19 **This PDF file includes:**

20 Main Text, Figures 1 to 4, Table 1

21

Abstract

22 Tropical cyclones (TC) are one of the most severe storm systems on Earth and cause significant
23 loss of life and property upon landfall in coastal areas. A better understanding of their variability
24 mechanisms will help improve the TC seasonal prediction skill and mitigate the destructive
25 impacts of the storms. Early studies focused primarily on tropical processes in regulating the
26 variability of TC activity, while recent studies suggested also some long-range impacts of
27 extratropical processes, such as lateral transport of dry air and potential vorticity by large-scale
28 waves. Here we show that stationary waves in the Northern Hemisphere integrate tropical and
29 extratropical impacts on TC activity in July through October. In particular, the tropical upper-
30 tropospheric troughs (TUTTs), as part of the summertime stationary waves, are associated with
31 the variability of large-scale environmental conditions in the tropical North Atlantic and North
32 Pacific and significantly correlated to the variability of TC activity in these basins. TUTTs are
33 subject to the modulation of diabatic heating in various regions and are the preferred locations for
34 extratropical Rossby wave breaking (RWB). A strong TUTT in a basin is associated with enhanced
35 RWB and tropical-extratropical stirring in that basin, and the resultant changes in the tropical
36 atmospheric conditions modulate TC activity. In addition, the anticorrelation of TUTTs between
37 the North Atlantic and North Pacific makes the TC activity indices over the two basins compensate
38 each other, rendering the global TC activity less variable than otherwise would be the case if
39 TUTTs were independent.

40

Significance Statement

41 Skillful seasonal prediction of tropical cyclone (TC) activity helps hurricane preparedness
42 and mitigation, especially when the storm impacts are expected to worsen with increasing sea level
43 and water vapor capacity in a warming climate. Slowly varying tropical oceanic conditions have
44 been regarded as a primary source of predictability for TC activity, but recent studies suggested
45 that TC activity is also subject to some long-range impacts of extratropical processes. We show
46 that summertime stationary waves in the Northern Hemisphere, including tropical upper-
47 tropospheric troughs, integrate tropical and extratropical impacts into a unified framework and
48 provide a hemispheric perspective that helps understand the variability and predictability of TC
49 activity over the North Atlantic and North Pacific.

50 **Main Text**

51 **1. Introduction**

52 The impacts of tropical cyclones (TC) are expected to worsen with increasing sea level and
53 water vapor capacity in a warming climate. Prior to a hurricane season, the public is concerned
54 whether the upcoming season will be more, or less, active than average. Skillful seasonal
55 prediction of TC activity provides valuable information for storm preparedness and has received
56 increased attention in recent years. While TC prediction has benefitted substantially from advances
57 in numerical models with higher resolution and improved physics parameterizations, a better
58 understanding of the variability mechanisms will also contribute to improved TC prediction.

59 Tropical cyclone formation, intensification and motion are strongly modulated by tropical
60 atmospheric conditions, including vertical wind shear and tropospheric humidity (1, 2). Although
61 the evolution of individual storms cannot be predicted deterministically beyond the synoptic time
62 scale, TC statistics spanning a longer time scale may be predictable because the large-scale
63 atmospheric circulation in the tropics is closely coupled to slowly varying tropical oceans (3, 4),
64 which are regarded as the primary source of predictability for tropical atmosphere and TC activity
65 (5-9). In addition to coupling to tropical oceans, tropical atmospheric circulation also interacts
66 actively with extratropical atmosphere. On the one hand, the tropics, as the primary terrestrial
67 source of heat, moisture, and angular momentum for the global climate system, modulate
68 extratropical atmospheric circulation via teleconnections and the meridional overturning
69 circulation (10). On the other hand, tropical atmospheric circulation is subject to the impacts of
70 extratropical processes. In a seminal study (11), Charney proposed the view that the large-scale
71 tropical atmospheric circulation is driven by the lateral coupling with precipitating regions and

72 with the extratropics given the weak coupling between vertical motion and horizontal circulation
73 in the tropics. In other words, the extratropics not only respond to but also feed back to the tropics.

74 Rossby waves represent an important process for extratropical atmospheric circulation to
75 feed back to tropical atmospheric circulation. Extratropical Rossby wavetrains and the attendant
76 Rossby wave breaking (RWB), aided by westerly ducts (12), may penetrate into the tropics and
77 affect tropical convection and change the atmospheric tracer distribution (13-15). In contrast to
78 teleconnections associated with low-frequency climate modes, RWB is a transient, nonlinear
79 process (16). Although RWB occurs on the synoptic time scale, repeated occurrence of RWB
80 enhances the mixing between the tropics and extratropics and can lead to significant anomalies in
81 wind, temperature and humidity fields on the subseasonal and longer time scales and thereby
82 modulate TC activity (17-20). Rossby wave breaking is affected by both tropical and extratropical
83 processes (21-23). Semi-idealized regional numerical experiments demonstrate that the
84 extratropical processes contribute to the interannual variability of Atlantic TC activity and that the
85 impacts can exceed the direct impacts of local SST in some years (24, 25). Overall, these studies
86 suggest that we need to look beyond the tropics to understand the variability of TC activity.

87 The objective of this study is to provide a unified framework integrating tropical and
88 extratropical impacts on TC activity, highlighting the tropical-extratropical connection. We will
89 invoke the concepts of tropical upper-tropospheric troughs (TUTTs) and summertime stationary
90 waves. TUTTs, also known as mid-ocean troughs (26, 27), are the preferred locations for RWB
91 (28, 29). Characterized by reduced meridional potential vorticity (PV) gradient, TUTTs “break”
92 the barrier of strong PV gradient along the subtropical jet that hinders geostrophic mixing (Fig. S1)
93 and can be regarded as “windows” for active tropical-extratropical interaction (12, 30). Along with

94 the upper-level monsoon anticyclones, TUTTs constitute the summertime subtropical stationary
95 waves (31, 32). The variability of TUTTs can be understood in the context of stationary waves,
96 which are planetary waves modulated by global diabatic heating, topographic effect, and transient
97 eddy feedback (31, 32). We will show that summertime stationary waves integrate tropical and
98 extratropical impacts and provide a hemispheric perspective on the variability of TC activity over
99 the North Atlantic and North Pacific basins.

100 **2. Results**

101 *2.1 TUTTs and the large-scale circulation anomalies modulating TC activity*

102 TUTTs are characterized by a cold-core, shallow structure in the upper troposphere (27).
103 They are present over the North Pacific and North Atlantic in boreal summer and extend
104 equatorward and westward from the subtropics to the tropics (Fig. S1). To quantify the relationship
105 between TUTTs and TC activity, a TUTT index is defined based on the equatorward extension of
106 the upper-level westerly flow over a subtropical ocean (see the definition of a TUTT index in “Data
107 and Methods”). We focus on the bulk of TC season over the North Atlantic and North Pacific,
108 July-October (JASO), during 1979-2018. Two TUTT indices are derived, one representing the
109 North Pacific TUTT (TUTT_Pac) and the other the North Atlantic TUTT (TUTT_Atl) (Fig. S2a).
110 TUTT_Pac extends from the subtropical East Pacific to the tropical Central Pacific, while
111 TUTT_Atl has a smaller latitude range but spans the entire Atlantic in the east-west direction,
112 including the Caribbean Sea in most years. It is worth noting that the tropical Central Pacific,
113 where TUTT_Pac resides, is relatively devoid of climatological TC activity, with most TCs to its
114 west or east, whereas the Atlantic has considerable activity that is highly variable both in regard
115 to the basin-wide storm frequency and regional distribution.

116 We will first explore the link between TUTTs and the large-scale atmospheric circulation
117 anomalies that modulate TC activity. Composite differences are constructed between strong and
118 weak TUTT years in the respective basins (see composite years in Table S1). As shown in Figs.
119 1a and 1b, strong TUTT_Atl years are characterized by enhanced vertical wind shear (VWS,
120 defined as the magnitude of the vector wind difference between 200 and 850 hPa) and reduced
121 column water vapor (CWV) in the tropical/subtropical Atlantic, including a major portion of the
122 main development region (MDR; defined here as 10°–25°N, 20°–80°W). Since TUTT_Atl is
123 defined based on the zonal wind, a strong link between the TUTT index and VWS is expected,
124 consistent with enhanced upper-level westerlies for a deepened TUTT. Additionally, strong CWV
125 composite anomalies suggest that the TUTT index also relates to the thermodynamics of
126 atmospheric circulation. Furthermore, a strong TUTT is associated with a stronger subtropical high
127 in the lower troposphere and higher sea level pressure over the tropical/subtropical Atlantic (33).

128 It is worth noting that the composites of VWS and CWV based on TUTT_Atl resemble
129 closely the corresponding composites based on an RWB index that represents the seasonal
130 frequency of RWB over the North Atlantic (Fig. 8 in ref. 18). The similarities can be explained by
131 the strong correlation between the RWB and TUTT_Atl indices ($r=0.80$). RWB frequency
132 increases significantly in strong TUTT years (Fig. 1c), especially over the western Atlantic, which
133 transports cold and dry extratropical air equatorward and increases the thickness gradient and
134 tropospheric dryness to its south (17-19). The extensive upper-level westerlies associated with a
135 deepened TUTT facilitate equatorward Rossby wave propagation (12) and breaking. Meanwhile,
136 transient eddy feedback likely helps maintain or amplify a TUTT (28, 30). The seasonal variability
137 of TUTTs thus reflects the cumulative impacts of RWB beyond the synoptic time scale.

138 Large-scale circulation anomalies are also found associated with TUTT_Pac (Figs. 1d-f).
139 During strong TUTT_Pac years, CWV is reduced over the central and eastern tropical Pacific and
140 enhanced in the western Pacific; VWS increases in the central and eastern tropical Pacific and
141 decreases in the western tropical Pacific. Similar to TUTT_Atl, TUTT_Pac is related to the
142 variability of RWB, and RWB occurs more frequently over the Central Pacific during strong
143 TUTT_Pac years. However, we caution that the large-scale circulation anomalies associated with
144 TUTTs should not be entirely attributed to RWB. As shown in section 2.3, other factors also
145 modulate TUTTs in addition to the transient eddy feedback related to RWB.

146 *2.2 TUTTs and Tropical Cyclone Activity*

147 Previous studies have shown that upper-level troughs or TUTT cells may facilitate the
148 development of a TC and affect the TC track at the synoptic time scale (34, 35). Here we will focus
149 on the seasonal time scale. Given the circulation anomalies (Fig. 1), it is not a surprise to see a
150 strong link between the seasonal TC activity and TUTTs. As shown in Fig. 2a, strong Atlantic
151 TUTT years are characterized by a basin-wide reduction of track density function (TDF) in the
152 North Atlantic, consistent with reduced CWV and enhanced VWS over the Atlantic MDR (33).
153 Significant negative correlations are found between TUTT_Atl and the basin-wide TC frequency
154 (TCF), hurricane frequency (HURR) and accumulative cyclone energy (ACE) (Table 1). In fact,
155 the seasonal correlations of the Atlantic TC indices with TUTT_Atl are much stronger than the
156 corresponding correlations with the Nino3.4 or the Atlantic MDR SST index (Table 1).

157 Over the eastern Pacific, reduced CWV and enhanced VWS in strong TUTT_Pac years (Figs.
158 1d-f) lead to a decrease in TC activity except in a small region south of the Gulf of California (Fig.
159 2b), where enhanced TC activity may be related to Central American gap winds (36). Over the

160 western Pacific, the changes of CWV and VWS have a more complicated spatial pattern (Figs. 1d-
161 f). East of 150°E, negative CWV anomalies and positive VWS anomalies are dominant between
162 10°–25°N, the main latitude band for TC genesis, while positive CWV anomalies and negative
163 VWS anomalies largely occur poleward or equatorward of this latitude band. West of 150°E,
164 enhanced CWV occurs off the coast along with weak VWS anomalies. As a consequence, tropical
165 cyclogenesis frequency decreases east of 150°E and does not change much west of 150°E (Fig. S3),
166 corresponding to a westward shift (37) in addition to a basin-wide reduction. The reduced genesis
167 frequency east of 150°E contributes to decreased TDF both locally and downstream. Negative TDF
168 anomalies thus prevail over the western Pacific except in the East China Sea region, where the
169 environmental conditions are favorable. The pattern bears a resemblance to the TC activity
170 anomalies related to the El Niño-Southern Oscillation (ENSO) reported in previous studies (38,
171 39), but the significant negative correlations between TC indices and TUTT_Pac over the western
172 and eastern Pacific are stronger than the corresponding correlations with an ENSO index, some of
173 which are insignificant or even close to zero (Table 1; also see 40).

174 Another interesting feature in the TDF composites is the out-of-phase TC variability between
175 the Atlantic and Pacific basins, which is consistent with the seesaw tendency of the Pacific and
176 Atlantic TUTTs (Fig. S2b). The anticorrelation between TUTT_Pac and TUTT_Atl ($r=-0.59$) can
177 be explained by the PV impermeability theorem (41). If the atmosphere is in a quasi-steady state
178 and the PV fluxes associated with diabatic heating and friction are negligible, the PV
179 impermeability theorem requires that the poleward advective PV flux closely balance the
180 equatorward advective PV flux (42, 43). The poleward advective PV flux is mainly determined by
181 the zonal mean meridional overturning circulation, while the equatorward advective PV flux
182 occurs primarily in the TUTT regions during summer (41). The variability of equatorward PV

183 fluxes over the two basins tends to compensate each other unless there is a substantial change in
184 the poleward advective PV flux. An important implication is the out-of-phase relationship of TC
185 activities between the Atlantic and Pacific basins [$r=-0.47$, -0.44 and -0.42 for TCF, HURR and
186 ACE between the North Pacific (the sum of the western and eastern Pacific) and North Atlantic,
187 respectively]. Since the North Pacific and North Atlantic together account for more than 60% of
188 the global TC frequency, the anticorrelation between the two basins makes the global TC activity
189 less variable (44, 45). This TC anticorrelation between the North Pacific and North Atlantic has
190 been examined in some previous studies and attributed to the modulation by the Walker circulation
191 or changes in tropospheric static stability related to relative SST (46-49). TUTTs offer an
192 additional mechanism for this inter-basin relationship. Further analysis shows that the
193 anticorrelation between TUTT_Atl and TUTT_Pac is much weaker on the monthly time scale,
194 possibly because the quasi-steady state assumption breaks down on the shorter time scales.

195 *2.3 Summertime Stationary waves*

196 Given the strong link between TUTTs and TCs, one may ponder what drives the variability
197 of TUTTs. Recalling that TUTTs are part of the summertime stationary waves, we carried out an
198 empirical orthogonal function (EOF) analysis to extract the dominant mode of variability of 200-
199 hPa streamfunction field between 15°S - 75°N . To focus on stationary waves, the zonal mean was
200 removed and the data were detrended prior to the EOF analysis. The leading EOF mode (EOF1)
201 explains 46% of the total variance and is well separated from the remaining modes (Fig. S4). It
202 consists of a wavenumber-one pattern in the tropics/subtropics and higher wavenumber patterns
203 in the extratropics (Fig. 3a). The former is associated with the variability of the Walker circulation,
204 reminiscent of the impacts of the ENSO (Fig. S5). In the extratropics, of particular interest to this

205 study is a wavetrain pattern emanating from the Central Pacific. It spans across the North Pacific
206 and North America and extends to the subtropical North Atlantic, following a great-circle route.
207 The wavetrain contributes to an out-of-phase relation between the subtropical Central Pacific and
208 the subtropical Atlantic, and it resembles the wavetrain that modulates RWB occurrence over the
209 West Atlantic (Fig. 1d in ref. 23). The time series of EOF1 is significantly correlated with
210 TUTT_Pac ($r=0.72$) and TUTT_Atl ($r=-0.82$) (Fig. 3b) and also strongly correlated to the HURR
211 and ACE indices in the three basins (Table 1).

212 EOF1 is characterized by a baroclinic vertical structure in the tropics/subtropics and a
213 barotropic structure in the extratropics (Fig. S6). The different vertical structures are consistent
214 with previous studies and indicate the role of diabatic heating in maintaining the stationary waves
215 in the tropics/subtropics and the importance of Rossby wave energy dispersion, topographic effect,
216 and transient eddy feedback in the extratropics (32, 50). To investigate the forcing mechanisms of
217 EOF1, correlations are calculated between the time series of EOF1 and SST/precipitation at each
218 grid point.

219 A horseshoe pattern of SST signals (Fig. 4a) is present over the Pacific, with significant
220 positive correlations over the Central and East Pacific and negative correlations extending from
221 the equatorial West Pacific poleward in both hemispheres. The pattern has stronger signals in the
222 extratropical Pacific than the ENSO pattern (Fig. S5), and resembles the Pacific Decadal
223 Oscillation (PDO). Meanwhile, negative correlations prevail over the tropical and extratropical
224 North Atlantic and are reminiscent of the Atlantic tripolar SST pattern. EOF1 is found significantly
225 correlated to the ENSO and PDO indices (Table S2). Additionally, the positive phase of EOF1 is
226 associated with reduced precipitation over the tropical Central and East Pacific and enhanced

227 precipitation over the Indian summer monsoon region, the Maritime Content, Australia, Central
228 America, the tropical/subtropical Atlantic, and the African monsoon region (Fig. 4b).

229 Although the SST and precipitation correlation maps strongly resemble the impacts of the
230 ENSO (Fig. S5), we stress that the variability of the summertime stationary waves or TUTTs
231 cannot be completely attributed to the ENSO, which is much weaker in summer than in winter. To
232 better illustrate this point, the linear impacts of the ENSO are removed from the time series of
233 EOF1 and the field variables of interest using the linear least squares regression on the Nino3.4
234 index, and correlation maps are constructed with the residual signals (right panels in Fig. 4). After
235 the removal of the ENSO, the SST signals are weakened in the tropical Pacific but remain largely
236 unchanged in the extratropical Pacific and are slightly enhanced in the North Atlantic; significant
237 correlations with precipitation remain over the subtropical Pacific, the Maritime Continent, the
238 Indian summer monsoon region, Central America and the tropical Atlantic; and the partial
239 correlation map between EOF1 and H200 has weaker signals over the tropical Pacific, but the
240 signals in the northern extratropics, including the wavetrain spanning from the Central Pacific to
241 the subtropical North Atlantic, remain significant or are even slightly enhanced (Figs. 4c and 4f).
242 These results are consistent with previous findings that the diabatic heating over various monsoon
243 regions helps maintain stationary waves (32, 50). In particular, it was suggested that a monsoon
244 anticyclone and the TUTT to its east are associated with an east-west overturning circulation, and
245 the dryness in the descending branch over the ocean helps maintain the TUTT via radiative cooling
246 (51, 33). Since the subtropical stationary waves have a first-baroclinic mode structure (31, 32), the
247 enhanced shear is a byproduct of a stronger upper-level trough sitting over a stronger low-level
248 subtropical high. Over the Pacific, a strong TUTT is also accompanied by a weakened monsoon
249 trough over the western Pacific (Fig. S7). These conditions all contribute to suppressed TC activity

250 in strong TUTT years. The large spatial scale of stationary waves highlights the global nature of
251 TC variability.

252 The correlations between TUTT indices and various climate indices are examined as well.
253 Each TUTT index is significantly correlated to several climate indices, including the Nino3.4,
254 PDO, the Atlantic meridional mode (AMM, 7), the Atlantic Multidecadal Oscillation (AMO), and
255 the Atlantic MDR SST (Table S2) (52, 53, 33). The PDO and TUTT_Pac, and the AMO and
256 TUTT_Atl, may be linked via RWB in the respective basins (54, 55). Although the ENSO
257 contributes to the anticorrelation between TUTT_Pac and TUTT_Atl, the partial correlation
258 between TUTT_Pac and TUTT_Atl remains significant ($r=-0.39$) after the influence of the ENSO
259 is removed using linear regression. Additionally, the negative correlation between the TUTT and
260 ACE indices of the same basin remains significant after the removal of the linear impacts of the
261 ENSO (Fig. S8).

262 We stress that TUTTs (or stationary waves) not only reflect the contribution from the slowly
263 varying tropical SST but also extratropical impacts. The latter is supported by the strong statistical
264 link between TUTT indices and RWB (Fig. 1). To further demonstrate this point, the Atlantic ACE
265 index is reconstructed using a linear regression model with various indices. The MDR SST alone
266 explains 30.3% of the ACE variance during 1979-2018; the MDR SST and the tropical mean SST
267 (averaged between 30°S-30°N) together explain 39.7% of the variance; and the explained variance
268 increases to 62.4% by adding TUTT_Atl (Fig. S9). This suggests that skillful prediction of
269 stationary waves and TUTTs will help improve TC seasonal prediction.

270 *2.4 A statistical assessment of TUTT predictability*

Given the abovementioned statistical linkages, we attempt to predict the EOF1 and TUTT indices in JASO using SST indices during April-June. We tested different pairs of possible predictors among the AMO, PDO, Nino3.4 and the Atlantic MDR SST, and constructed multiple linear regression (MLR) models. The AMO and Nino3.4 are the best pair to predict TUTT_Atl and EOF1, and the PDO and Atlantic MDR SST are the best pair to predict TUTT_Pac (Fig. S10), suggesting that a TUTT is modulated by SST in both the local basin and a remote basin. The anomaly correlation coefficients between the predicted and observed time series using the leave-five-out method (see data and methods) are 0.72, 0.53 and 0.57 for TUTT_Atl, TUTT_Pac and EOF1, respectively. The prediction skill using this simple statistical model represents the lower bound of the seasonal predictability of summertime stationary waves and TUTTs with the caveat that the sample size is not very large and cross validation may overestimate the prediction skill (56). Although the predictors have information on tropical SST, we caution that the AMO and PDO both include extratropical information, and that the physical processes linking tropical SST and TUTTs (or stationary waves) may not be purely tropical as implied by the recent semi-idealized numerical model studies (24, 25). A better understanding of the TUTT variability that is unrelated to tropical SST will help better understand the predictability of TC activity and merits further study.

3. Summary and Discussion

We demonstrated the strong link between TUTTs and TC activity over the North Pacific and North Atlantic. As part of the summertime stationary waves, TUTTs are related to tropical and extratropical SST and precipitation anomalies in various regions, including the ENSO and monsoons. Additionally, TUTTs are the preferred regions for RWB, and reflect the cumulative

293 effects of RWB beyond the synoptic scale. Active RWB in a strong TUTT year contributes to
294 enhanced vertical wind shear and reduced tropospheric humidity, and may help amplify the TUTT
295 via the transient eddy feedback. Radiative forcing associated with the changes in the humidity field
296 may play a role in maintaining TUTTs as well (33). Our analysis suggests that stationary waves
297 integrate tropical and extratropical impacts on TC activity and provide a hemispheric perspective
298 on the variability of TC activity over the North Pacific and North Atlantic. In addition, the Atlantic
299 TUTT and the Pacific TUTT tend to vary out of phase due to the PV impermeability nature, and
300 the seesaw relationship of the Pacific and Atlantic TUTTs contributes to an anticorrelation of TC
301 activity between the two basins, which makes the global TC activity less variable. Finally, because
302 TUTTs are connected to monsoons and extratropical Rossby waves, they introduce a factor other
303 than tropical SST for the variability of TC activity. Overall, this study advocates a hemispheric
304 perspective that helps understand the variability and predictability of TC activity over the North
305 Atlantic and North Pacific. This view may also help understand the projection of TC activity in
306 future climate.

307 **4. Materials and Methods**

308 The JASO seasonal mean data on isobaric surfaces from the ERA-Interim reanalysis are
309 used to examine the atmospheric circulation anomalies, and the 6-hourly PV field on the 350-K
310 isentropic surface is used to detect anticyclonic RWB (57, 17). RWB frequency is in units of
311 percent and is defined as the frequency of high-PV tongue centroids associated with RWB in a $5^\circ \times$
312 5° resolution grid mesh, smoothed with four-point averaging. Additionally, we use precipitation
313 data from the Global Precipitation Climatology Project, SST from the Extended Reconstructed Sea
314 Surface Temperature Version 5, and TC track and intensity data from the IBTrACS. The Nino3.4

315 SST index is used to represent the ENSO, and along with the other indices, is downloaded from
316 the NOAA Physical Science Laboratory.

317 To quantitatively evaluate the variability of the TUTTs and their links to TCs and RWB, a
318 TUTT index is defined using the 200-hPa geopotential height (H200) field. First, the geostrophic
319 zonal wind is derived from H200 with a fixed Coriolis parameter (f) at 15°N (denoted as u_g). The
320 zonal mean latitude of the circumglobal contour of the long-term seasonal mean $u_g = 1.0 \text{ m s}^{-1}$,
321 which turns out to be just south of 20°N, is chosen as a reference latitude, and the area where the
322 circumglobal contour of the seasonal mean $u_g = 1.0 \text{ m s}^{-1}$ extends equatorward of the reference
323 latitude is defined as the TUTT index of an oceanic basin. We chose to use the constant- f
324 geostrophic zonal wind rather than the total zonal wind because the weak westerly flow in the
325 latter field occasionally extends across the equator and connects to westerlies in the Southern
326 hemisphere. The choice of a small positive value 1.0 m s^{-1} , instead of zero, helps increase the
327 robustness of the results, and varying this value from 0.5 to 1.5 m s^{-1} does not qualitatively
328 change our results. Varying the Coriolis parameter in the calculation of u_g is equivalent to
329 choosing a different contour threshold for u_g . A TUTT index defined this way focuses on the
330 westerly flow in the eastern half of a TUTT and mainly describes the extent of a TUTT, although
331 one can define the intensity and the longitudinal and latitudinal locations of a TUTT as well.

332 The leave-five-out method is used to assess the skill of an MLR model. For a time series of
333 n observations, we leave five consecutive observations out as a test dataset (e.g., 1, 2, ..., 5) and
334 develop an MLR based on the remaining $n-5$ observations (e.g., 5, 6, ..., n). The model is then
335 used to predict the five test data points. This procedure is repeated for different test datasets (1-5,
336 6-10, etc.) to predict all observations, yielding a time series of the predicted variable. The

337 correlation between the predicted and observed time series is calculated to assess the prediction
338 skill. The leave-five-out cross validation is chosen over leave-one-out to take care of the biennial
339 tendency of the large-scale atmospheric variability.

340 ***Data Availability***

341 The IBTrACS data are available at <https://www.ncdc.noaa.gov/ibtracs/>. The climate indices are
342 available from the NOAA Physical Science Laboratory (<https://www.psl.noaa.gov/data/>). The
343 normalized TUTT indices are available in Dataset S1. The ERA-Interim reanalysis data are
344 downloaded from the National Center for Atmospheric Research (NCAR) Research Data
345 Archive (<https://rda.ucar.edu/datasets/ds627.0/>). The GPCP precipitation data are available at
346 https://psl.noaa.gov/data/gridded/data_gpcp.html. The ERSST data are available at
347 <https://www.ncdc.noaa.gov/data-access/marineocean-data/extended-reconstructed-sea-surface-temperature-ersst-v5>.

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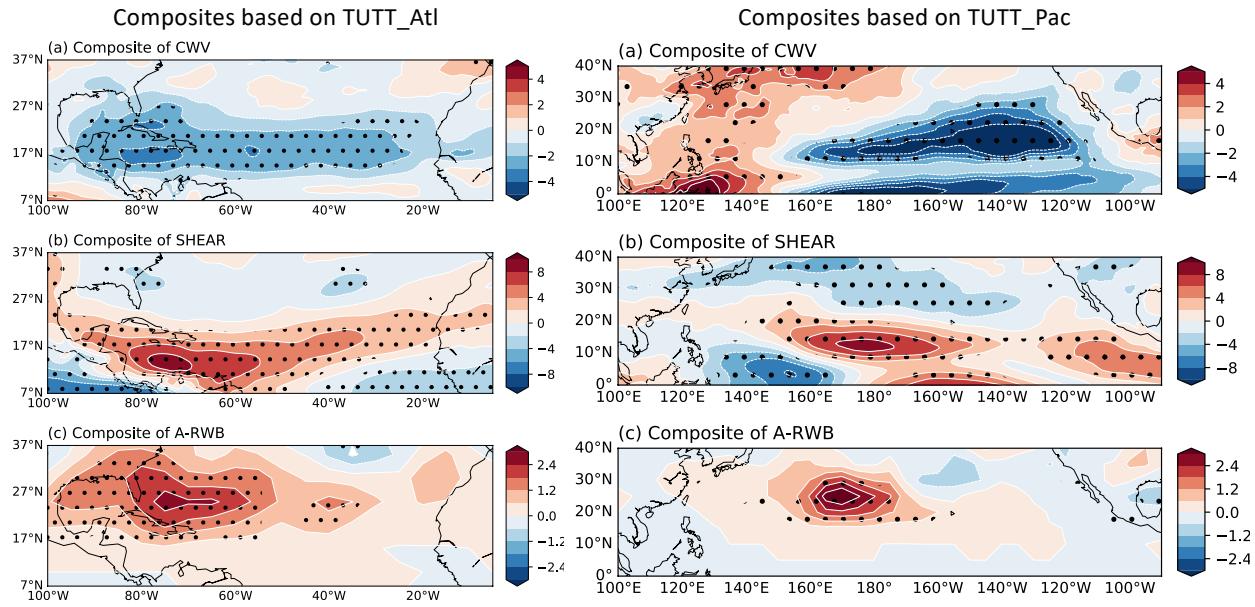
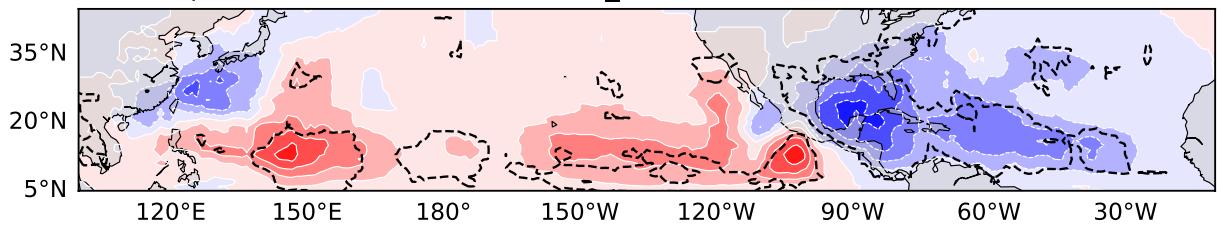
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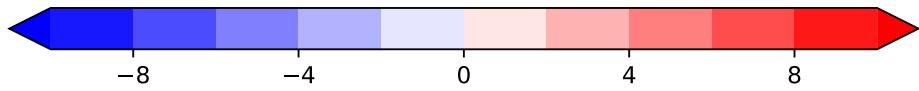
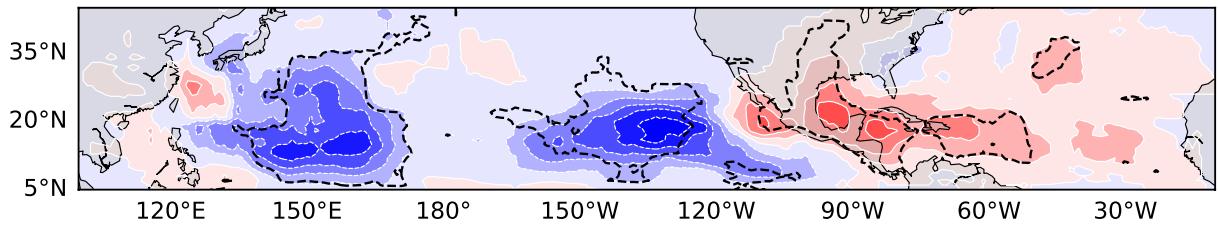
Figure 1 Composite anomalies of (a, d) column water vapor (mm), (b, e) vertical wind shear ($m s^{-1}$), and (c, f) anticyclonic RWB frequency (%) for the Atlantic basin based on the TUTT_Atl index (left) and for the Pacific basin based on the TUTT_Pac index (right). Black dots highlight the anomalies exceeding the 95% confidence level. Note that the latitude-longitude ranges of the plots are different for the two basins.

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(a) Composites of TDF based on TUTT_Atl



(b) Composites of TDF based on TUTT_Pac



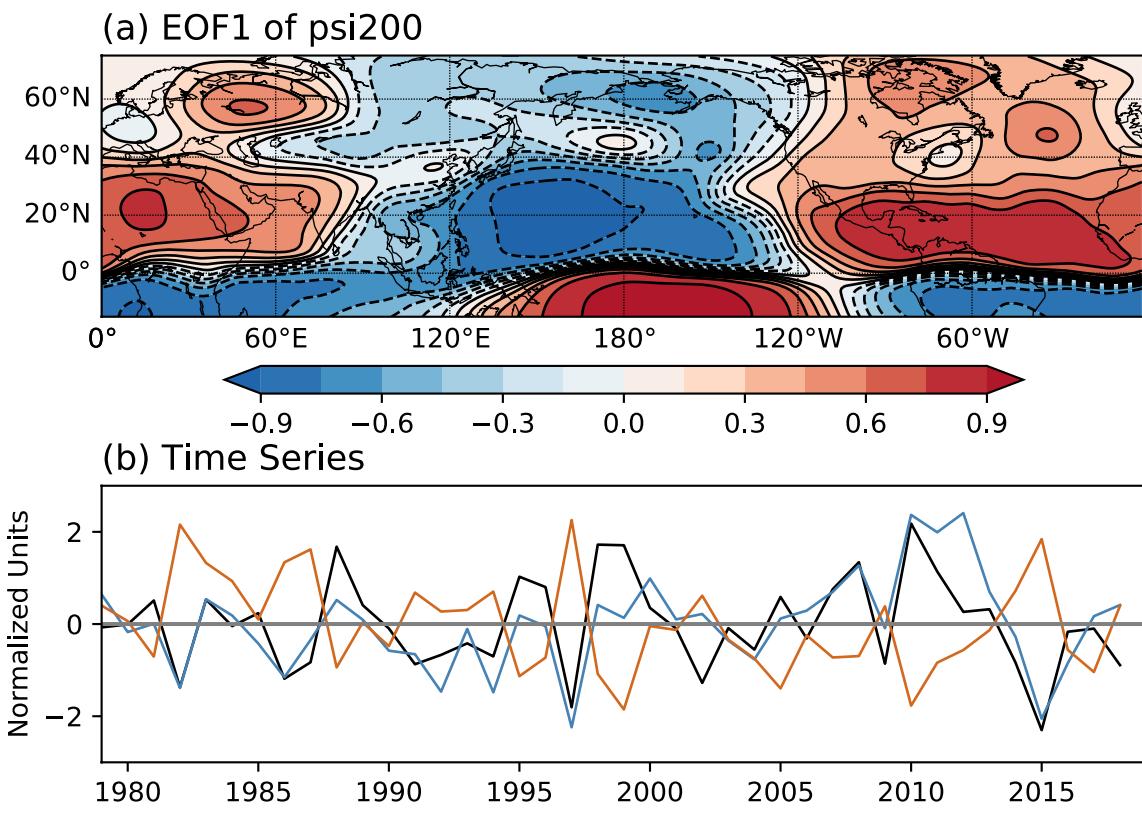
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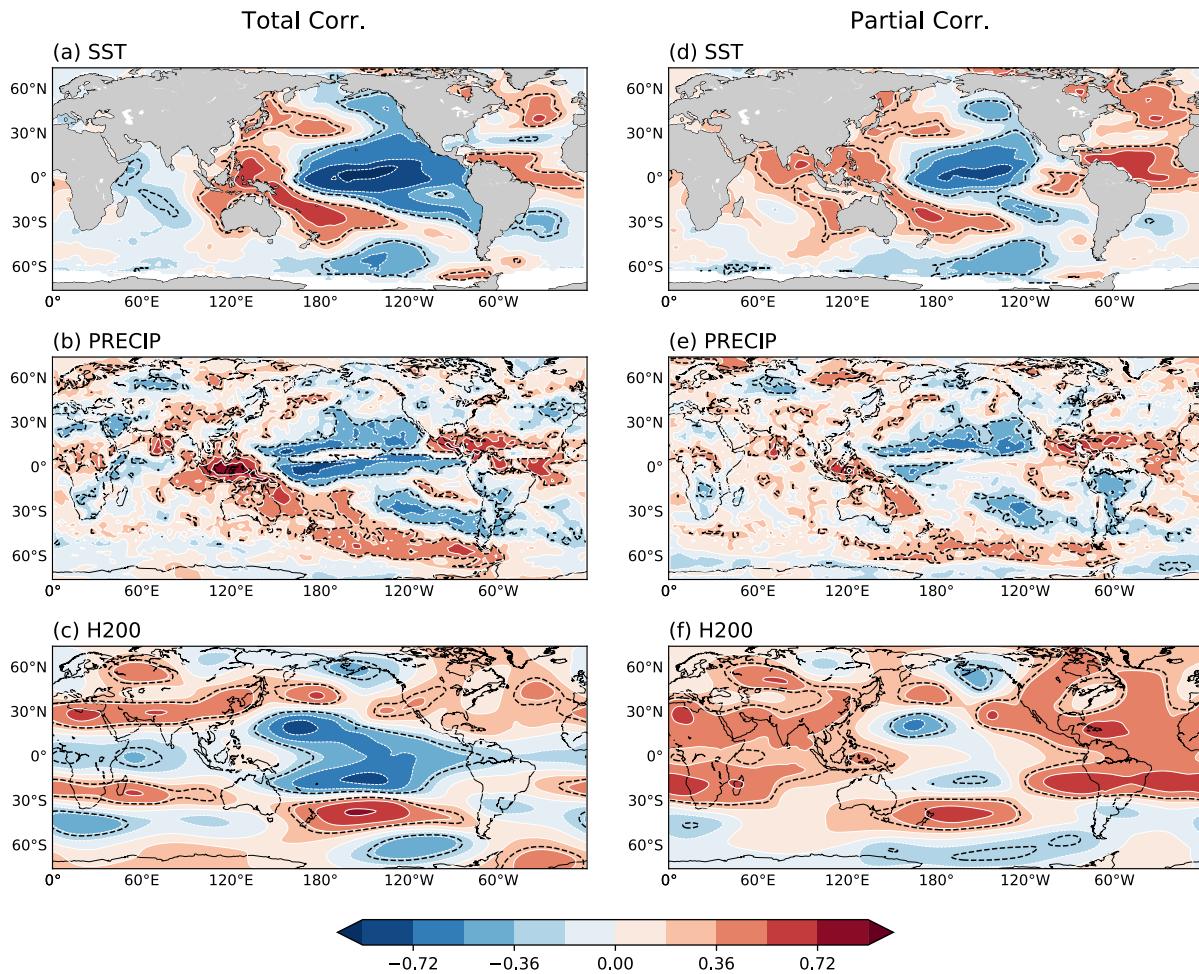
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Figure 2 Composites of tropical cyclone track density function (units: number of TCs per month within a $10^\circ \times 10^\circ$ grid box) based on (a) TUTT_Atl and (b) TUTT_Pac. Dashed contours depict the anomalies exceeding the 95% confidence level.



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Figure 3 (a) The first EOF mode of 200-hPa streamfunction (scaled to unit variance); (b) the normalized time series of EOF1 (black), TUTT_Pac (blue), and TUTT_Atl (brown).



501
 502 Figure 4 Correlations of the EOF1 time series with (top) SST, (middle) precipitation and
 503 (bottom) 200-hPa geopotential height. The total correlations are shown in the left column (a-c),
 504 and the right column (d-f) shows the partial correlations with the Niño3.4 index controlled.
 505 Dashed contours depict the correlation coefficients exceeding the 95% confidence level.

506 Table 1 Correlation coefficients between different climate indices and the TC activity indices over
 507 three basins during JASO 1979-2018. RWB is the RWB frequency between 10-85°W, south of the
 508 jet axis and north of 20°N (see Zhang et al. 2017 for more details). Most correlations exceed the
 509 95% confidence level, and those below the 95% confidence level are highlighted by an asterisk.
 510

Corr.	TCF	HURR	ACE
Atlantic TC Indices			
TUTT_Atl	-0.73	-0.76	-0.75
MDR	0.59	0.56	0.55
Nino3.4	-0.34	-0.38	-0.32
RWB	-0.46	-0.56	-0.68
EOF1	0.54	0.58	0.50
East Pacific TC Indices			
TUTT_Pac	-0.60	-0.58	-0.63
Nino3.4	0.39	0.28*	0.46
EOF1	-0.60	-0.50	-0.58
West Pacific TC Indices			
TUTT_Pac	-0.45	-0.53	-0.61
Nino3.4	0.01*	0.18*	0.56
EOF1	-0.30*	-0.46	-0.69

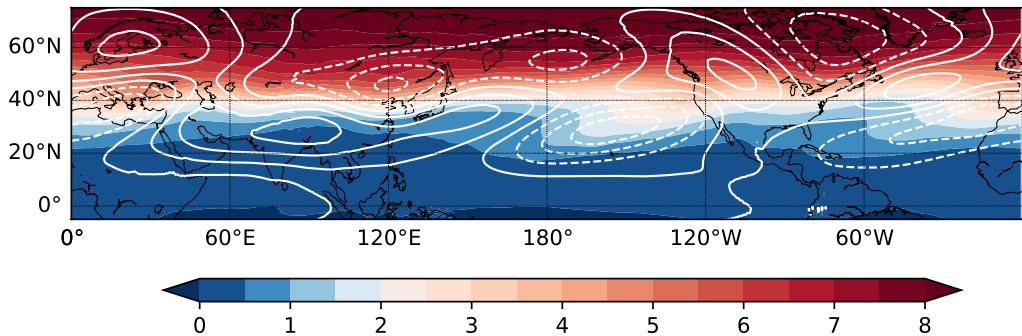
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519 Supplementary Information for
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521 **Summertime Stationary Waves Integrate Tropical and Extratropical Impacts**
522 **on Tropical Cyclone Activity**
523 Zhuo Wang, Gan Zhang, Timothy J. Dunkerton and Fei-Fei Jin
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525 **corresponding author:** Zhuo Wang
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529 **This PDF file includes:**
530 Figures S1 to S10
531 Tables S1 to S2

Long-Term Mean PV200 and Eddy H200

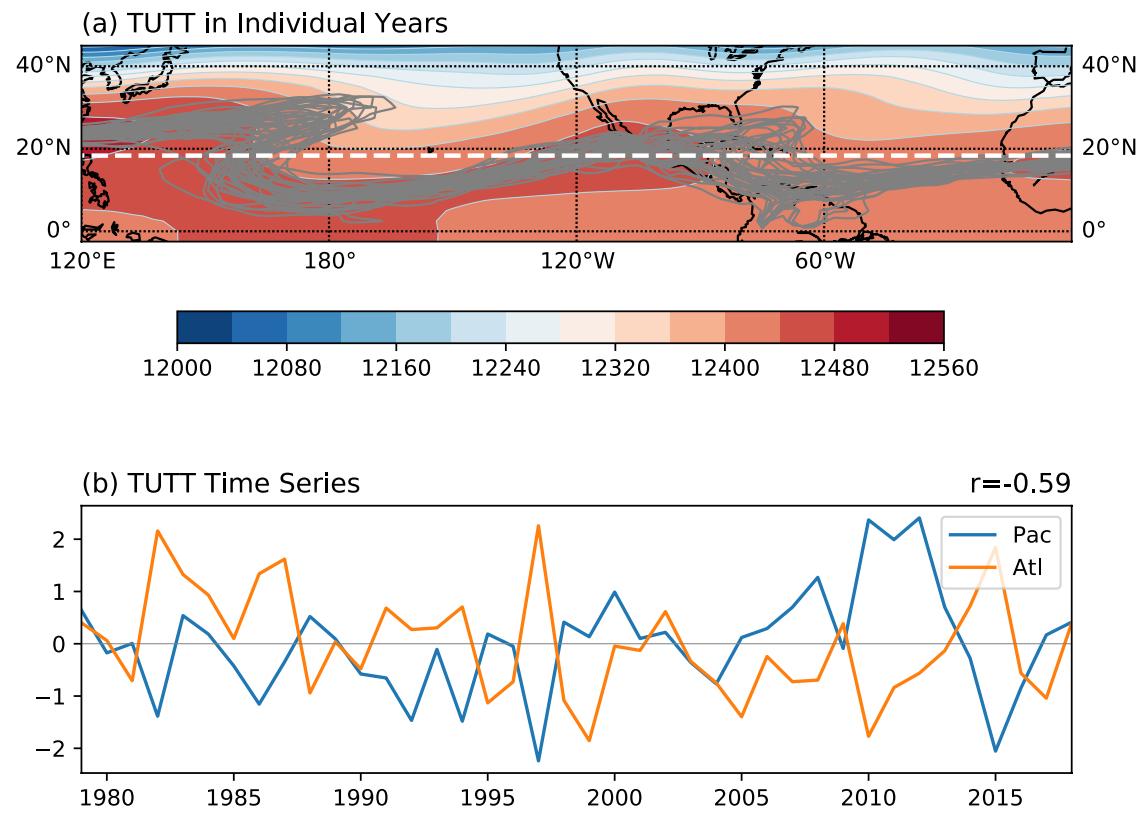


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Fig. S1 Long-term mean (1979-2018) 200-hPa potential vorticity (shading: units: PVU) and 200-hPa geopotential height (contours with the interval of 250 m). The zonal mean component is removed from the geopotential height field.

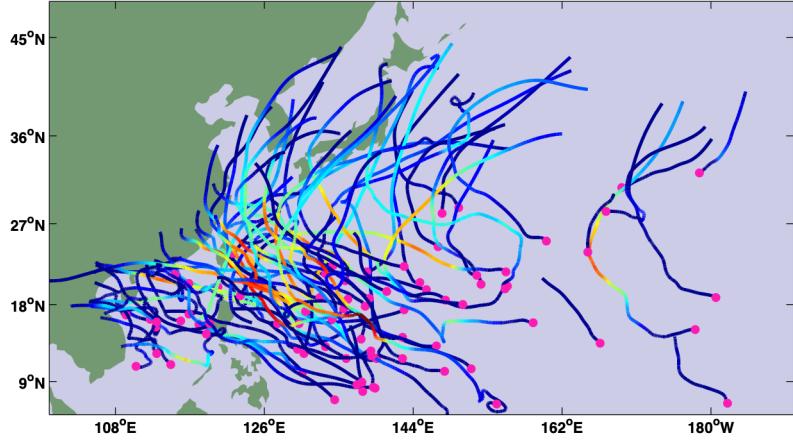
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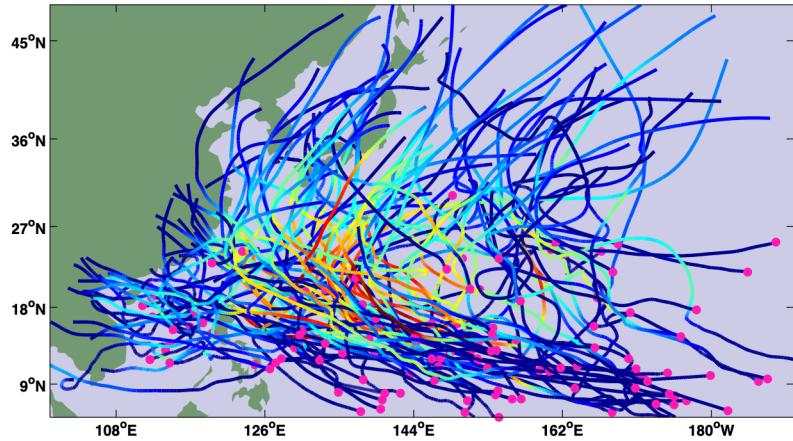


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540 Fig. S2 (a) Long-term mean 200-hPa geopotential height (shading; units: m), the reference
541 latitude for TUTTs (white dashed line; see text for details), and the contours of $u_g =$
542 1.0 m s^{-1} for individual years from 1979-2018; (b) the time series of the normalized TUTT_Pac
543 and TUTT_Atl indices. The correlation between TUTT_Atl and TUTT_Pac is shown at the
544 upper right corner of panel (b).

(a) TUTT_Pac: positive (87 TCs)



(b) TUTT_Pac: negative (127 TCs)



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546 Fig. S3 Composites of TCs over the western North Pacific during (a) eight strong TUTT_Pac years
547 and (b) eight weak TUTT_Pac years. Pink dots represent genesis locations, and colors along TC
548 tracks indicate TC intensity in terms of the maximum surface wind speed. The numbers inside
549 parentheses show the total number of TCs for each composite plot.

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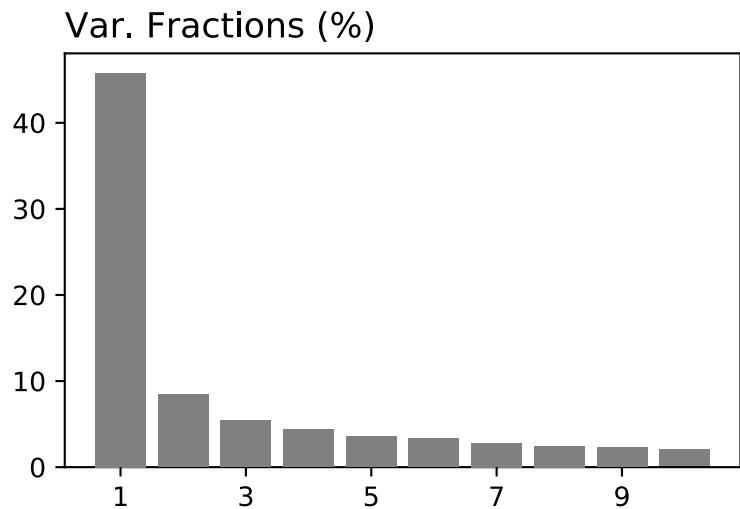
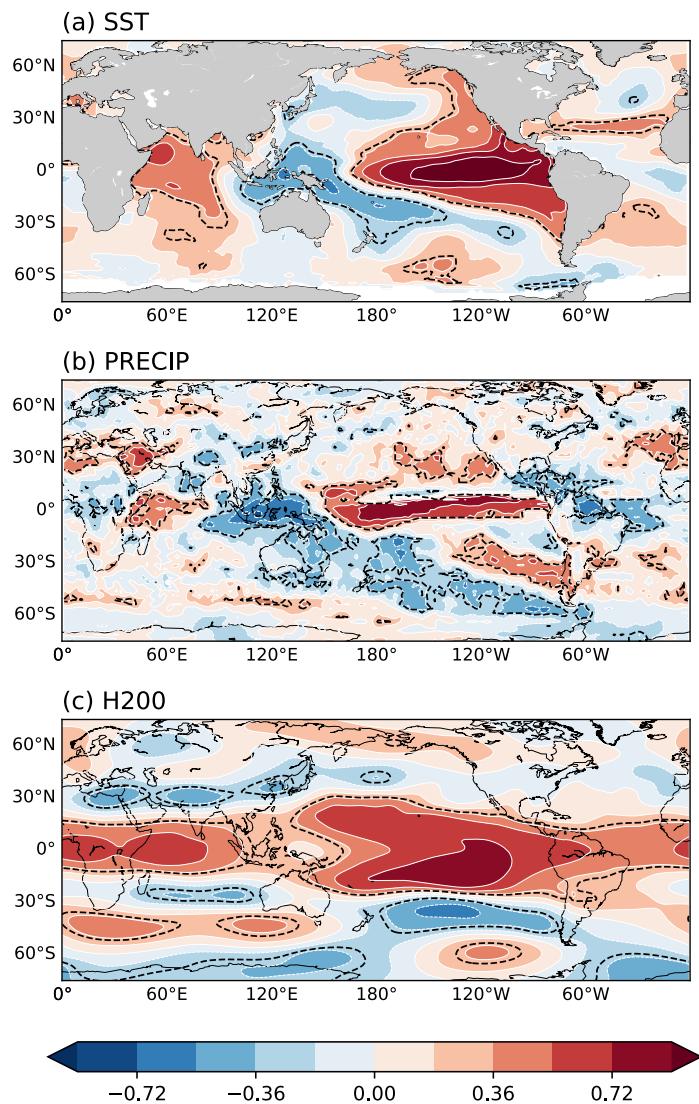


Fig. S4 The fractional contribution of the first ten EOF modes to the total variance.

Corr.with the ENSO



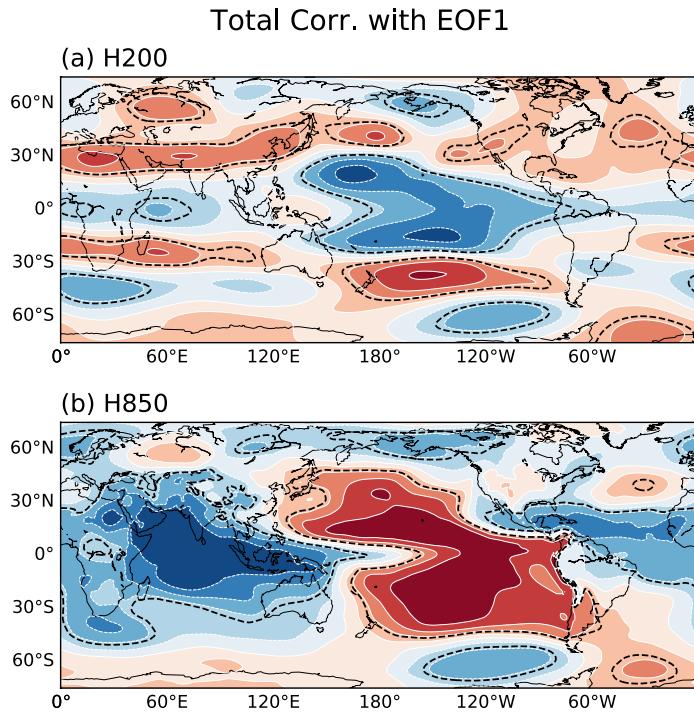
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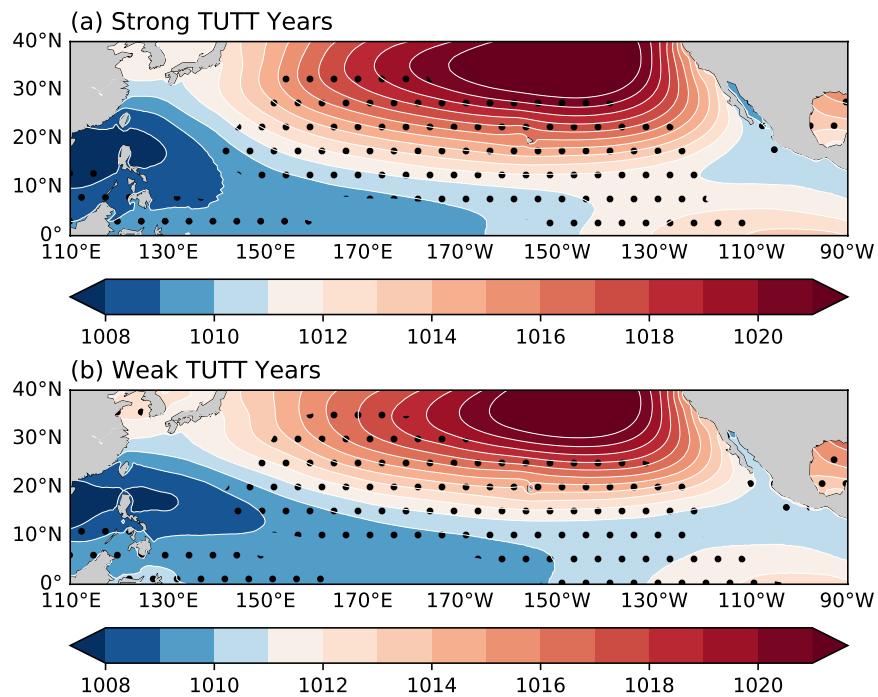
Fig. S5 Correlations of the Nino3.4 index with (a) SST, (b) precipitation, (c) H850 and (d) H200 during 1979-2018 JASO. Dashed contours depict the correlation coefficients exceeding the 95% confidence level.



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557 Fig. S6 Correlations of the EOF1 time series with (a) H200 and (b) H850. Dashed contours
558 depict the correlation coefficients exceeding the 95% confidence level.

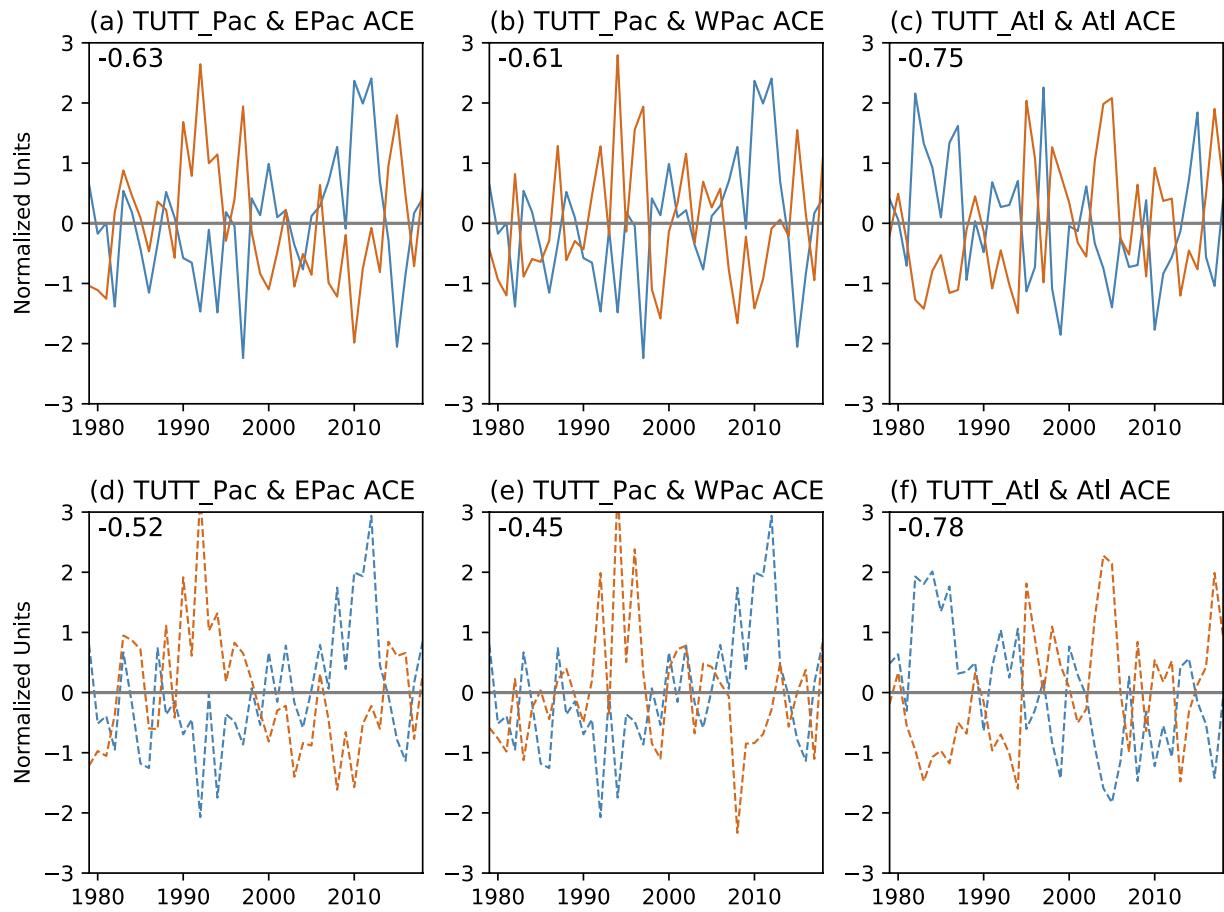
Composites of SLP



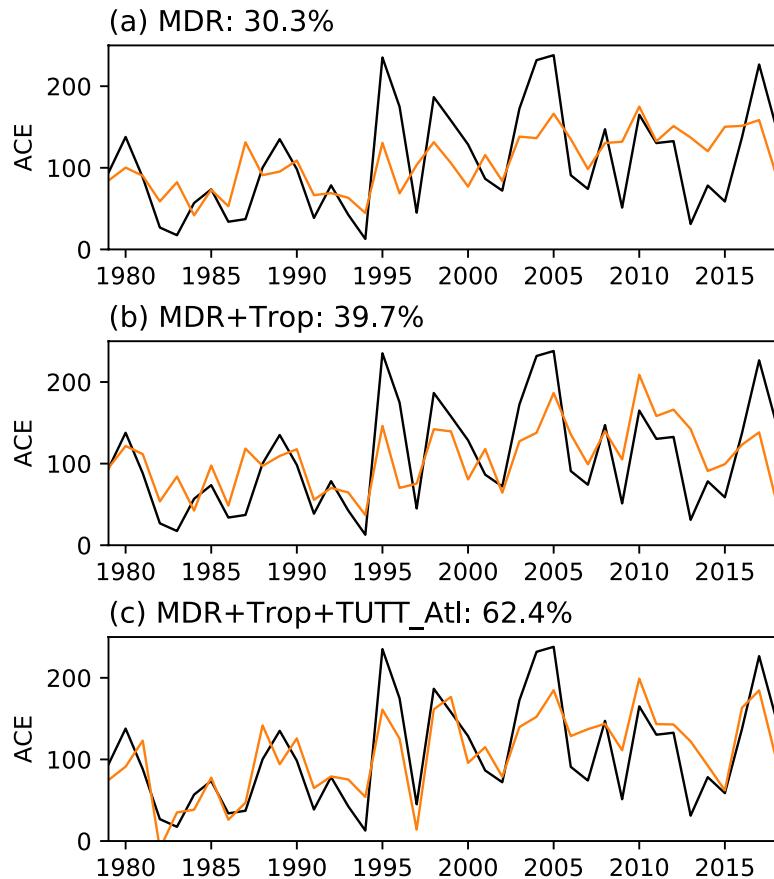
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560 Fig. S7 Composite mean sea level pressure (SLP; hPa) for (a) strong and (b) weak TUTT_Pac
561 years. Dots highlight where SLP differs significantly between the two phases.

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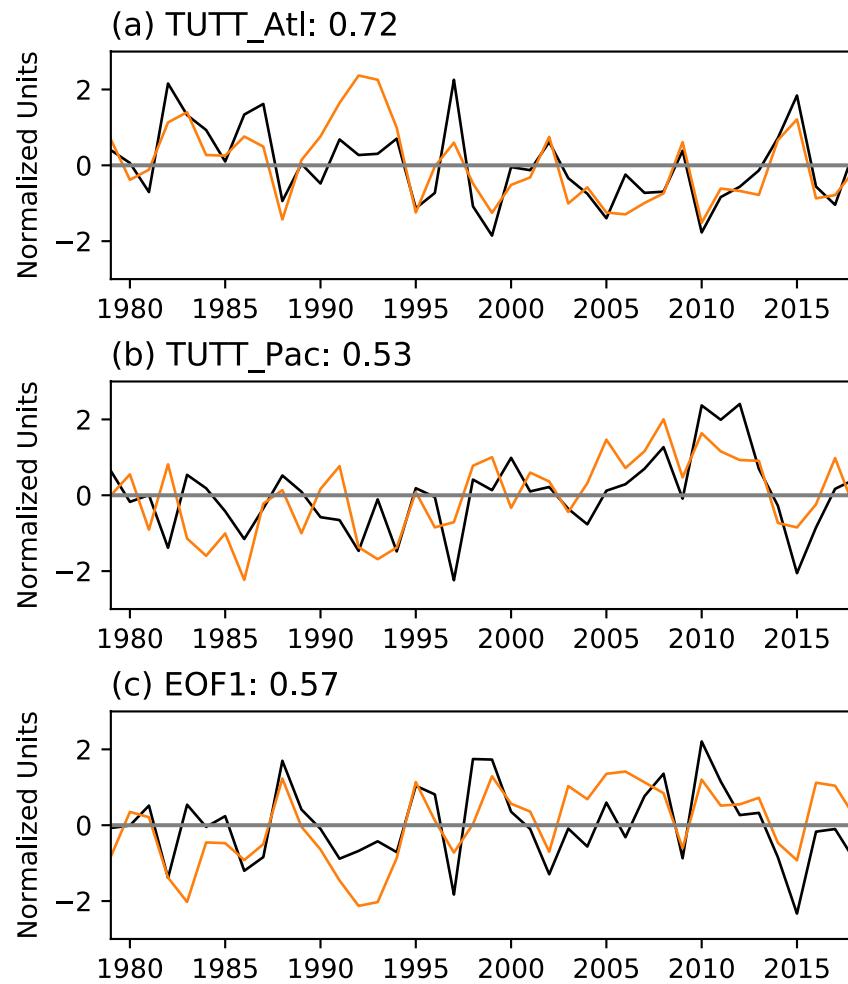


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 565 Fig. S8 (left) Time series of TUTT_Pac (blue) and the East Pacific ACE (brown); (middle) time
 566 series of TUTT_Pac (blue) and the West Pacific ACE (brown); and (right) time series of
 567 TUTT_atl (blue) and the North Atlantic ACE (brown). The top panels show the original data,
 568 and the bottom panels show the time series after the linear impacts of the ENSO are removed.
 569 The number at the upper left corner of a panel is the correlation between the illustrated time
 570 series. All the time series are normalized and shown for 1979-2018.



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Fig. S9 The Atlantic ACE time series derived from IBTrACS (black) and reconstructed (orange) based on the linear regression of (a) MDR SST, (b) MDR SST and the tropical mean SST (Trop), and (c) MDR SST, tropical mean SST and TUTT_Atl. The numbers indicate the observed variance explained by the reconstructed ACE.



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Fig. S10 Time series of the observed (black) and predicted (orange) time series of (a) TUTT_Atl, (b) TUTT_Pac, and (c) EOF1 (see the main text for more information). The numbers indicate the correlations between the predicted and observed time series.

580 Table S1 Composite years for the positive and negative phases of TUTT_Atl.

	TUTT_Atl	TUTT_Pac
Positive	1997, 1982, 2015, 1987, 1986, 1983, 1984, 2014	2012, 2010, 2011, 2008, 2000, 2013, 2007, 1979
Negative	1999, 2010, 2005, 1995, 1998, 2017, 1988, 2011	1997, 2015, 1994, 1992, 1982, 1986, 2016, 2004

581

582 Table S2 Correlations with various climate indices during JASO 1979-2018. Correlations below
583 the 95% confidence level are highlighted by an asterisk.

	Nino3.4	PDO	PMM	MDR	AMM	AMO
EOF1	-0.79	-0.59	-0.26*	0.24*	0.50	0.25*
TUTT_Pac	-0.50	-0.65	-0.45	0.34	0.39	0.35
TUTT_Atl	0.71	0.62	-0.03*	-0.46	-0.65	-0.48

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