

1
2
3
4 Modulation of Tropical Cyclones over the Eastern Pacific by the Intraseasonal
5 Variability Simulated in an AGCM
6
7

8 XIANAN JIANG^{1,2}, MING ZHAO³, and DUANE E. WALISER²
9

10
11 ¹*Joint Institute for Regional Earth System Science & Engineering, University of California, Los Angeles,*
12 *California*

13 ²*Jet Propulsion Laboratory, California Institute of Technology, Pasadena, California*

14 ³*NOAA/Geophysical Fluid Dynamics Laboratory, Princeton, New Jersey*
15
16
17
18
19
20

21 Submitted 09/15/11, Journal Climate
22 Revised 02/13/12
23
24
25
26
27
28
29
30
31
32

33
34

Corresponding author address: Dr. Xianan Jiang, Jet Propulsion Laboratory, California Institute of
35 Technology, MS 233-300, 4800 Oak Grove Drive, Pasadena, CA 91109. Email: xianan@jifresse.ucla.edu.
36

37 Copyright © 2012. All rights reserved.

38
39
40
41
42
43
44
45
46
47
48
49
50
51
52
53
54
55
56
57
58
59
60
61

Abstract

This study illustrates that observed modulations of tropical cyclone (TC) genesis over the eastern Pacific (EPAC) by large-scale intraseasonal variability (ISV) are represented well in a recently developed high-resolution atmospheric model (HiRAM) at NOAA’s Geophysical Fluid Dynamics Laboratory (GFDL) with a horizontal resolution of about 50km. Considering the intrinsic predictability of the ISV of 2-4 weeks, this analysis thus has significant implications for dynamically based TC predictions on intraseasonal time scales. Analysis indicates that the genesis potential index (GPI) anomalies associated with the ISV can generally well depict ISV modulations of EPAC TC genesis in both observations and HiRAM simulations. Further investigation is conducted to explore the key factors associated with ISV modulation of TC activity based on an analysis of budget terms of the observed GPI during the ISV life cycle. It is found that, while relative roles of GPI factors are dependent on ISV phase and location, lower-level cyclonic vorticity, enhanced mid-level relative humidity, and reduced vertical wind shear can all contribute to the observed active TC genesis over the EPAC during particular ISV phases. In general, the observed anomalous ISV patterns of these large-scale GPI factors are well represented in HiRAM. Model deficiencies are also noted particularly in the anomalous mid-level relative humidity patterns and amplitude of vertical wind shear associated with the EPAC ISV.

1. Introduction

Due to the great threat of tropical cyclones (TCs) to life and property worldwide, improving prediction of TC activity has profound socio-economic impacts. While short- and medium-range forecasts (up to one week) and seasonal outlooks of TC activity have been routinely conducted, predictions of TCs on intraseasonal time scales (one week to several weeks) are still at an experimental stage in most operational prediction centers¹. This is largely due to our limited understanding of predictability sources for TC prediction on this intermediate time scale. For short-

¹ Intraseasonal TC forecasts have been operational at the European Centre for Medium-Range Weather Forecasts (ECMWF) since January 2011.

62 to-medium range TC forecasts, the skill is essentially from the initial condition, and forecasts are
63 usually realized by integrating high-resolution, in some cases limited-area, dynamical models with
64 sophisticated data assimilation systems. Moreover, the predictability of seasonal TC forecasts is
65 largely dependent on slowly varying large-scale elements, such as sea surface temperature (SST;
66 e.g., Gray 1984; Shapiro 1987; Goldenberg and Shapiro 1996). For TC prediction on the
67 intraseasonal time scale, predictability is generally considered rooted in the tropical intraseasonal
68 variability (ISV, e.g., Madden-Julian Oscillation, Madden and Julian 1994), although SST and other
69 tropical waves could also provide partial predictability (e.g., Leroy and Wheeler 2008; Frank and
70 Roundy 2006). The significant impacts of the ISV on TC genesis and movement over various ocean
71 basins have been widely reported (e.g., *Western Pacific*: Nakazawa 1988; Liebmann et al. 1994;
72 Wang and Zhou 2008; *Indian Ocean*: Bessafi and Wheeler 2006; Ho et al. 2006; Kikuchi and Wang
73 2010; *Eastern Pacific*: Molinari et al. 1997; Maloney and Hartmann 2000b; Aiyyer and Molinari
74 2008; *Gulf of Mexico*: Maloney and Hartmann 2000a; Mo 2000; Higgins and Shi 2001; *Atlantic*
75 *Ocean*: Maloney and Shaman 2008; Klotzbach 2010; Camargo et al. 2007b; *Australian Region*: Ho
76 et al. 2006). It is found that the frequency of TC genesis during convectively active ISV phases is
77 up to four times that during suppressed ISV phases (e.g., Maloney and Hartmann 2000a; Barrett and
78 Leslie 2009). Due to this intimate linkage between the ISV and TC activity, and considering the
79 intrinsic predictability of the ISV of 2-4 weeks (Waliser 2006), the reliable prediction of the ISV
80 state several weeks ahead will provide an important foundation for TC prediction on intraseasonal
81 time scales.

82 While significant progress over the past 2-3 decades has been made in understanding the
83 fundamental physics of the ISV (Zhang 2005; Lau and Waliser 2011), achieving credible ISV
84 simulations with general circulation models (GCMs) has been challenging (e.g., Slingo et al. 1996;
85 Waliser et al. 2003; Lin et al. 2006; Kim et al. 2009; Jiang et al. 2011). Until recently, useful
86 predictive skill of the ISV had generally been limited to only 1–2 weeks, with statistical models
87 often outperforming dynamical models (e.g., Waliser 2006; Jiang et al. 2008). This becomes one of

88 the main reasons that current intraseasonal TC predictions, although still limited, are mainly based
89 on statistical approaches (e.g., Frank and Roundy 2006; Leroy and Wheeler 2008). However, in
90 recent years, the improvements in model physics, spatial resolution and data assimilation systems
91 have led to significant increases in the predictive skill for ISV in a few models (Waliser 2011). For
92 example, the ECMWF Integrated Forecast System (IFS; Bechtold et al. 2008; Vitart and Molteni
93 2010), NCEP Coupled Forecast System v2 (Zhang and v. d. Dool 2011) and the Predictive Ocean-
94 Atmosphere Model for Australia (Rashid et al. 2011), exhibit useful ISV predictive skill out to 2-4
95 weeks. This improved skill, along with the resolutions to reasonably resolve TCs, motivated Vitart
96 (2009) to analyze the ISV-TC connections in a set of 46-day hindcasts based on a recent version of
97 the ECMWF IFS with a horizontal resolution of 80km. Vitart (2009) demonstrated that the
98 ECMWF forecast model is generally capable of depicting the observed modulations of the ISV on
99 TC activity including genesis and risk of landfall. This study, for the first time, illustrates that it is
100 possible for a conventional cumulus-parameterized global model to represent both ISV and TCs, as
101 well as their intimate relationships, thus suggesting a plausible new dynamical approach for
102 intraseasonal TC forecasts. Inspired by this encouraging work, there has been increasing enthusiasm
103 in exploring the predictive skill and predictability of dynamical TC forecasts based on ECMWF IFS
104 hindcasts / forecasts (e.g., Belanger et al. 2010; Vitart et al. 2010; Elsberry et al. 2010).

105 In this study, we present results from an analysis of the ISV-TC relationships over the eastern
106 Pacific (EPAC) as simulated in a high resolution GCM (hereafter HiRAM) recently developed at
107 National Oceanic and Atmospheric Administration (NOAA)'s Geophysical Fluid Dynamics
108 Laboratory (GFDL; Zhao et al. 2009). In the EPAC, the ITCZ and warm SST are generally
109 confined north of the equator, particularly in boreal summer, with the warm water region often
110 referred to as the EPAC warm pool. Most TCs over the EPAC are spawned over this warm pool
111 region. The EPAC ITCZ also exhibits vigorous intraseasonal variations in winds and convection
112 (Knutson and Weickmann 1987; Kayano and Kousky 1999; Maloney and Esbensen 2003, 2007; de
113 Szoeké and Bretherton 2005; Jiang and Waliser 2008, 2009; Jiang et al. 2011), which exert broad

114 impacts on regional weather and short-term climate (e.g., Magana et al. 1999; Higgins and Shi
115 2001; Lorenz and Hartmann 2006; Small et al. 2007; Wu et al. 2009; Serra et al. 2010; Martin and
116 Schumacher 2010), including TC activity over the EPAC and the Gulf of Mexico as mentioned
117 earlier. Two dominant ISV periodicities, e.g., a 40-day variability mode and a quasi-biweekly mode,
118 were reported over the EPAC (Jiang and Waliser 2009). A recent multi-model evaluation study by
119 Jiang et al. (2011) illustrates that HiRAM is one of the best in depicting both dominant EPAC ISV
120 modes among nine GCMs analyzed. Moreover, at a horizontal resolution of about 50km, HiRAM
121 also shows great potential in representing global TC activity, including its climate distribution,
122 interannual and multi-decadal variability, as well as responses to global warming (Zhao et al. 2009;
123 Zhao and Held 2010; Held and Zhao 2011).

124 As HiRAM is capable of capturing both the ISV and TC activity over the EPAC, it is of great
125 interest to explore how observed ISV-TC relationships are depicted in this model. This study, on
126 one hand, will have significant implications on the intraseasonal dynamical TC forecasts as
127 discussed earlier. Moreover, it is noted that ECMWF IFS hindcasts analyzed by Vitart (2009) were
128 initialized by observations; thus, skill could be gained from the specified initial conditions. In the
129 present study, integrations from a long-term climatological run based on HiRAM are analyzed.

130 **2. Model, Datasets, and Approaches**

131 Simulations from 1998 to 2008 based on GFDL HiRAM analyzed in this study are similar to
132 those utilized in Zhao et al. (2009) with a horizontal resolution of 50km and prescribed observed
133 monthly mean sea surface temperatures (SSTs) and sea ice. Model storms are detected and tracked
134 every six hours following the algorithm described in Zhao et al. (2009). At each time, a potential
135 tropical storm is identified by a local 850hPa relative vorticity maximum exceeding $1.6 \times 10^{-4} \text{ s}^{-1}$, a
136 minimum in sea surface pressure, a maximum in near-surface wind speed greater than 17 ms^{-1} , and
137 a warm core structure with maximum temperature between 300 and 500hPa. Daily rainfall output
138 from HiRAM is employed to identify EPAC ISV events. Additionally, daily 3D winds, relative

139 humidity, and temperature fields from HiRAM are also analyzed to explore key factors associated
140 with the ISV in modulating TC genesis in model simulations.

141 Six-hourly tropical storm “best tracks” over the EPAC and Atlantic Ocean from 1998 to 2008
142 are obtained from the hurricane database (HURDAT) maintained by the NOAA National Hurricane
143 Center (Jarvinen et al. 1984). Rainfall observations are based on Tropical Rainfall Measuring
144 Mission (TRMM, version 3B42; Huffman et al. 1995). The raw 3-hourly TRMM rainfall with a
145 0.25-degree spatial resolution was interpolated onto $1^\circ \times 1^\circ$ daily data. Daily atmospheric fields
146 including 3D winds, relative humidity, temperature, and SST during the same period from the
147 recent ECMWF ERA-Interim reanalysis (Dee et al. 2011) with a horizontal resolution of $1.5^\circ \times 1.5^\circ$,
148 are also analyzed. In this study, our analysis will focus on TCs exceeding tropical storm strength in
149 both observations and HiRAM simulations

150 Following Jiang et al. (2011), extended empirical orthogonal function (EEOF) analyses
151 (Weare and Nasstrom 1982) of daily 10-90-day band-pass filtered rainfall anomalies over the
152 EPAC (140°W – 90°W ; EQ - 30°N) from 1998-2008 are performed to define the leading ISV modes
153 in both observations and simulations. Temporal lags of 21 days are adopted for EEOF analyses.
154 Similar to EEOF results as presented in Jiang et al. (2011), the first two leading EEOF modes of
155 both observed and simulated daily rainfall anomalies capture the dominant 40-day ISV modes
156 associated with the EPAC ITCZ. Then these two leading principal components (PCs), which are in
157 quadrature with each other, are used to determine daily ISV amplitudes and phases (ranging from 1
158 to 8, then back to phase 1) following a similar method employed in Wheeler and Hendon (2004) for
159 the MJO. Composite analysis for observed and simulated rainfall can then be accomplished by
160 averaging the band-pass filtered rainfall anomalies over each ISV phase based on selected strong
161 ISV events ($\sqrt{PC_1^2 + PC_2^2} \geq 1.0$) during the EPAC hurricane season (June-October) from 1998-2008.

162 It is noteworthy that rainfall associated with TCs may contribute to ISV variances as
163 illustrated by a study over the western Pacific (Hsu et al. 2008). Sensitivity tests are thus conducted
164 by conducting EOF analyses of the same 10-90-day band-pass filtered TRMM rainfall dataset to

165 affirm that the dominant ISV modes over the EPAC derived by the EEOF approach are independent
166 from TC activity. The first two leading EOF modes of observed rainfall anomalies during May-
167 October of 1998-2008 (e.g., total 2024 days) capture similar ISV patterns as depicted by the two
168 leading EEOF modes; the two quadratic patterns of EOF₁ and EOF₂ modes represent anomalous
169 rainfall patterns at phase 8 and phase 2 derived by the two leading EEOF modes, respectively
170 (figure not shown). An additional EOF analysis of TRMM rainfall is then conducted by excluding
171 any day during the 11 summers when a TC was reported anywhere over the EPAC. Results show
172 that the two leading EOF modes of rainfall anomalies during these non-TC days (total 1106 days)
173 are largely identical to those based on rainfall during the full period of the 11 summers. Therefore,
174 the 40-day ISV mode over the EPAC as extracted by EEOF or EOF analysis represents an intrinsic
175 regional low-frequency convective variability mode independent from TC activities. Since the
176 EEOF approach exerts stronger constraints to the derived dominant ISV modes than the EOF
177 analysis by providing temporal evolution of the spatial patterns associated with the leading ISV
178 modes, we employ the PCs corresponding to the two leading EEOFs to define the daily phases of
179 the 40-day ISV mode over the EPAC.

180 **3. Modulations of the ISV on TC activity in GCM simulations**

181 Left panels (a-d) in Figure 1 illustrate observed rainfall anomalies (shaded) along with TC
182 genesis locations from 1998-2008 during a life cycle of the ISV. As previously documented (Jiang
183 and Waliser 2008; Maloney et al. 2008), the EPAC ISV exhibits both eastward and northward
184 propagation. In agreement with previous studies (e.g., Maloney and Hartmann 2000b), strong
185 modulations of TC genesis over the EPAC by the ISV is clearly observed. TC genesis largely
186 occurs over regions with enhanced ISV convection. The northward migration of the TC genesis
187 zone from phases 6+7 (Fig. 1d) to phases 1+8 (Fig. 1a), and then to phases 2+3 (Fig. 1b), coupled
188 with the northward movement of enhanced ISV convection is clearly evident. The largely
189 suppressed ISV convection over the EPAC during phases 4+5 is also consistent with significantly

190 reduced TC genesis at this time. It is also noteworthy that a clear poleward shift of the TC genesis
191 belt relative to maximum ISV convection is discerned for ISV phases 6+7 (Fig. 1d).

192 An identical analysis to that performed on the observations was carried out on the HiRAM
193 output. The bulk features of rainfall anomalies associated with the observed ISV over the EPAC
194 are captured well in the HiRAM simulations (shaded; Figs. 1e-h), which have been
195 comprehensively examined in Jiang et al. (2011). Of particular interest is that strong modulation of
196 TC genesis by the model ISV, as exhibited in the observations, is depicted well in HiRAM,
197 including the northward migration of TC genesis belt from phases 6+7 to phases 2+3, and the
198 considerably smaller TC genesis during phases 4+5. In addition, the poleward shift of TC genesis
199 relative to the convection maximum is also noted in model simulations during phases 6+7 (Fig. 1h),
200 although most model TC genesis favors longitudes east of 120°W, while genesis is still evident west
201 of 120°W in the observations (Fig. 1d). This could be associated with the eastward displacement of
202 the maximum convection center in HiRAM simulations (Fig. 1h) relative to its observed
203 counterpart (Fig. 1d).

204 Figure 2 further illustrates strong impacts of the ISV on TC genesis by displaying TC genesis
205 counts over the EPAC during each ISV phase from 1 to 8, as well as during a weak ISV period as
206 denoted by ISV phase “0” with gray bars in both the observation and HiRAM simulation plots.
207 Again it is clearly illustrated (Fig. 2a) that many more TCs form during ISV phases 1, 7, and 8, with
208 only a few cases occurring during phases 4-6. A relatively small number of TCs originate during
209 weak ISV phases. HiRAM captures the observed variation of TC counts as a function of ISV phase
210 very well (Fig. 2b).

211 To demonstrate how ISV modulates TC movements, in addition to TC genesis locations (i.e.,
212 Fig. 1), six-hourly TC positions during each ISV phase are plotted in Fig. 3 for both observations
213 and simulations. Again, significant impacts of the ISV on TC movements over the EPAC as seen in
214 the observations are well captured in HiRAM. In both observations and model simulations, TCs
215 largely cluster over regions with enhanced ISV convection, while TC activity is greatly reduced

216 over regions with suppressed ISV convection. Consistent with the previous discussion, TC activity
 217 exhibits a slight northward shift relative to active ISV convection near 10°N during phases 6+7, in
 218 both observations (Fig. 3d) and simulations (Fig. 3h).

219 **4. Factors controlling the intraseasonal TC activity over the EPAC**

220 In this section, we explore key factors associated with the ISV in modulating TC genesis over
 221 the EPAC by utilizing an empirical genesis potential index (GPI; Emanuel and Nolan 2004), which
 222 is defined as follows,

$$223 \quad GPI = \frac{\left|10^5 \zeta\right|^{3/2}}{\eta} \left(\frac{\gamma}{50}\right)^3 \left(\frac{PI}{70}\right)^3 \frac{(1 + 0.1 \cdot V_{Shear})^{-2}}{s} \quad (1)$$

224 where ζ is the 850hPa absolute vorticity (s^{-1}), γ is the relative humidity at 600hPa (%), PI is the
 225 potential intensity in terms of wind speed (ms^{-1}), which gives an upper bound on TC intensity
 226 (Bister and Emanuel 2002); and V_{shear} is the magnitude of the vertical wind shear between 850 and
 227 200hPa (ms^{-1}). Terms η , r , ψ , and s denote the GPI components associated with 850hPa absolute
 228 vorticity, 600hPa relative humidity, potential intensity, and vertical wind shear, respectively.

229 It has been illustrated that the GPI is able to replicate the observed climatological annual cycle
 230 as well as interannual variations of TC genesis in several different basins (Camargo et al. 2007a).
 231 On intraseasonal time scales, the GPI also exhibits capability in representing global TC genesis
 232 (Camargo et al. 2009; hereafter CWS09). Based on an analysis of budget terms of anomalous GPI,
 233 CWS09 further examined the roles of the four environmental variables in Eq. (1) associated with
 234 the MJO in modulating TC genesis over the global oceans. It is suggested that mid-level humidity
 235 plays a leading role in the MJO modulation of TC genesis frequency, with a secondary role for the
 236 vorticity and a very weak influence of vertical shear and potential intensity. This differs from
 237 results from previous studies (Maloney and Hartmann 2000b, a; Hall et al. 2001; Bessafi and
 238 Wheeler 2006) that pointed to the dynamical factors, including low-level vorticity and vertical wind
 239 shear, as being most relevant for the MJO modulation of TCs. Also in CWS09, the first-position
 240

241 density field was used to represent TC genesis frequency with nine grid points smoothing, which
242 makes it difficult to characterize detailed features of TC genesis over a small ocean basin such as
243 the EPAC. Here, we analyze the actual TC genesis location during ISV phases. Also the new ERA-
244 Interim reanalysis with improved quality and horizontal resolution is employed for the calculation
245 of daily GPI patterns. Moreover, GPI anomalies associated with the ISV are also calculated based
246 on HiRAM simulations, which may provide further indications of physics regulating TC genesis in
247 the observations.

248 Figures 4a-d show evolution of anomalous GPI (shaded) along with TC genesis during ISV
249 phases based on observations. Observed rainfall anomalies in each phase as displayed in Figs. 1a-d
250 are duplicated with green contours. It is shown that positive (negative) GPI anomalies are generally
251 collocated with enhanced (suppressed) convection. The majority of TC genesis occurs over positive
252 anomalous GPI regions. Particularly noteworthy is the poleward shift of a positive GPI pattern
253 relative to enhanced convection during phases 6+7 (Fig. 4d). TC genesis during this period is well
254 depicted by positive GPI anomalies rather than convection itself as discussed in Fig. 1. A small
255 number of TCs are also observed over negative GPI regions, e.g., during phases 4+5, which could
256 be ascribed to factors that are not included in Eq. (1) or not associated with the 40-day ISV mode
257 discussed here (e.g., quasi-biweekly ISV mode, or smaller scale convectively coupled waves and
258 easterly waves).

259 The evolution of anomalous GPI patterns along with rainfall anomalies and TC genesis during
260 different ISV phases in HiRAM simulations is displayed in Figs. 4e-h. While simulated GPI
261 anomalies exhibit slightly weaker amplitudes than their observed counterparts, model TC genesis
262 also largely occurs over regions with positive GPI anomalies as in the observations. Of particular
263 interest is that the poleward displacement of positive GPI relative to maximum convection during
264 phases 6+7 as previously noted in the observations is also clearly evident in HiRAM simulations
265 (Fig. 4h). Also in agreement with the observations, positive GPI anomalies show better consistency
266 with model TC genesis than convection anomalies.

267 Since the anomalous GPI patterns generally represent TC genesis during ISV evolution very
 268 well in both observations and simulations, we proceed to assess the relative roles of the four
 269 elements in Eq. (1) in contributing to total GPI anomalies. For clarification purposes, any variable
 270 A can be decomposed into a climatological annual cycle component (\bar{A}) and a departure from
 271 climate component (A'), which contains variability on various time scales from synoptic, to
 272 intraseasonal, and to interannual time scales; i.e., $A = \bar{A} + A'$, where A represents each GPI
 273 component in Eq. (1), η , r , ψ , and s . Then the anomalous GPI associated with the ISV (GPI^{isv})
 274 can be written as follows,

$$275 \quad GPI^{isv} = \bar{r}\bar{\psi}\bar{s}\eta^{isv} + \bar{\eta}\bar{\psi}\bar{s}r^{isv} + \bar{\eta}\bar{r}\bar{s}\psi^{isv} + \bar{\eta}\bar{r}\bar{\psi}s^{isv} + \{\bar{\psi}\bar{s}(r'\eta')^{isv} + \bar{r}\bar{\psi}(s'\eta')^{isv} + \dots + \bar{\psi}(s'r'\eta')^{isv} + \dots + (\eta'r'\psi's')^{isv}\} \quad (2)$$

276 Similar to CWS09, the total GPI anomalies associated with the ISV can be decomposed into four
 277 linear terms which are associated with the ISV of each of the four variables while the other three
 278 terms are kept at their summer mean values (i.e., first four terms on the r.h.s. of Eq. 2). Moreover,
 279 nonlinear effects can also be evaluated by terms which include high-order variances of two or more
 280 terms of the four variables (total 11 terms included in brackets in Eq. 2). The results suggest that
 281 contributions by ISV potential intensity to the total GPI anomalies, i.e., $\bar{\eta}\bar{r}\bar{s}\psi^{isv}$ in Eq. (2), are
 282 relatively smaller than those from the ISV of the other three variables in both observations and
 283 HiRAM simulations (to be shown in Figs. 6 and 8). Figure 5 displays observed anomalous 850hPa
 284 vorticity and winds, 600hPa relative humidity, and V_{shear} during different ISV phases along with
 285 total GPI anomalies (shaded, duplicated from Figs. 4a-d). It is shown that positive (negative)
 286 850hPa vorticity patterns associated with the anomalous winds induced by ISV convection are in
 287 good agreement with positive (negative) GPI anomalies during each ISV phase, suggesting the
 288 important role of low-level vorticity in regulating TC genesis. Meanwhile, enhanced mid-level
 289 moisture is also found to be conducive for active TC genesis over the coastal regions off Central
 290 America during ISV phases 1+8 and 2+3 (Figs. 5e-f). An east-west dipole in 600hPa moisture
 291 anomalies is evident during phases 6+7 (Fig. 5h). While positive moisture anomalies to the west of
 292 120°W tends to sustain local positive GPI anomalies between 10-15°N, negative anomalies to the

293 east of 120°W counteract observed positive GPI anomalies. The maximum V_{shear} anomalies over the
294 far EPAC are largely confined south of 15°N. During phases 6+7, reduced vertical shear to the east
295 of 120°W (Fig. 5l), largely due to weakening of low-level mean westerly flow by the northeasterly
296 anomalous winds (Fig. 5d), contributes to locally positive GPI anomalies.

297 Figure 6 displays budget terms of the total anomalous GPI in Eq. (2) over three selected
298 regions where active TC genesis is observed (see green boxes in Fig. 5). As mentioned above, the
299 ISV of potential intensity plays a relatively minor role in contributing to total GPI anomalies over
300 all three regions. Over the northern box region off Central America coast (110-95°W; 13-18°N),
301 positive GPI anomalies during ISV phases 1-2 are dominated by two major terms, e.g., 850hPa
302 vorticity and 600hPa relative humidity. For the region of 140-120°W and 10-15°N, while both
303 850hPa vorticity and 600hPa moisture are again the two most important terms in contributing to
304 maximum positive GPI anomalies during phases 6-7, nonlinear terms also make comparable
305 contributions particularly in phase 6. And for the third region of 115-95°W and 9-13°N, in addition
306 to the contribution by 850hPa vorticity, the major contribution to the maximum positive GPI during
307 phases 6-7 is by the reduced vertical wind shear; mid-level humidity associated with the ISV tends
308 to reduce the total GPI anomalies in phase 6. Meanwhile, contributions from nonlinear terms are
309 also noted over this region.

310 Further analyses illustrate that the nonlinear contributions to observed GPI anomalies
311 associated with the 40-day ISV mode during phases 6-7 over the two box regions near 10°N (Figs.
312 6b and 6c) are largely due to covariance terms of 850hPa vorticity and mid-level humidity
313 anomalies (i.e., $\overline{\psi' s'(r'\eta')^{isv}}$ in Eq. 2) on both synoptic (< 8days) and high-frequency ISV scales (8-
314 20 days). While further studies are needed to explore physics responsible for this upscale
315 contribution to the GPI anomalies associated with the 40-day ISV mode, this result is largely in
316 concert with vigorous activity of easterly waves and the quasi-biweekly ISV mode observed over
317 the fast EPAC near 10°N during boreal summer (e.g., Serra et al. 2010; Jiang and Waliser 2009).

318 The above results suggest that the poleward shift of TC genesis relative to maximum ISV
319 convection observed in ISV phases 6+7 (Fig. 1d) are largely due to the presence of positive low-
320 level vorticity, enhanced mid-level moisture, and reduced vertical wind shear to the north of
321 convection (cf. Figs. 4d, 5d, 5h, and 5l). The northward shift of positive vorticity relative to ISV
322 convection could result from the coupling between ISV convection and summer mean monsoonal
323 flow with an easterly vertical shear (Jiang et al. 2004; Jiang and Waliser 2008). As the maximum
324 amplitude of easterly vertical wind shear over the EPAC is present near 8°N (Jiang and Waliser
325 2008), this may explain why the strongest northward displacement of GPI anomalies relative to the
326 convection is observed during phases 6+7 when enhanced convection is largely confined south of
327 10°N (Fig. 4d). While a northward shift of lower tropospheric specific humidity anomalies to ISV
328 convection has been illustrated on a pressure-latitude plane in Jiang and Waliser (2008), further
329 analysis is warranted for better understanding of 3D anomalous relative humidity pattern associated
330 with the EPAC ISV.

331 To further understand physics in regulating TC genesis in HiRAM, we proceed to examine
332 roles of the four GPI factors for total simulated GPI anomalies. Figure 7 displays similar anomalous
333 850hPa vorticity and winds, 600hPa relative humidity, and V_{shear} during different ISV phases along
334 with total GPI anomalies as in Fig. 5 but based on HiRAM simulations. In general, similar
335 evolutionary features in these anomalous fields associated with the ISV as in the observations are
336 noted in simulations. Positive 850hPa vorticity is in accord with positive GPI anomalies during the
337 ISV evolution (Figs.7a-d), again suggesting the importance of low-level vorticity in modulating TC
338 genesis in HiRAM. Consistency between enhanced mid-level relative humidity and positive GPI
339 anomalies over the coastal region off Central America during phases 1+8 and phases 2+3 (Figs. 7e-
340 f) is also discerned in model simulations. Meanwhile, enhanced (reduced) vertical wind shear
341 during ISV phases 1+8 and 2+3 (phases 4+5 and 6+7) over the far EPAC is also generally
342 consistent with observations, though model deficiencies in anomalous humidity and wind shear
343 patterns are also noticed. A positive moisture center during phases 6+7 is found near 112°W, 8°N in

344 GCM (Fig. 7h); instead, the observed positive relative humidity center is located near 135°W, 17°N
345 during this period (Fig. 5h). Since this positive 600hPa relative humidity plays an active role in
346 contributing the observed positive GPI over the western part of the EPAC along 10°N during phases
347 6+7 (see Figs. 5h and 6b), the model biases in simulating 600hPa moisture anomalies may partially
348 explain the less elongated positive GPI band during phases 6+7 in HiRAM, leaving the model TC
349 genesis largely confined to the eastern part of the EPAC rather than more wide-spread along 10°N
350 in the observations. Moreover, the amplitude of reduced vertical wind shear during phases 6+7 in
351 HiRAM is much weaker than its observational counterpart (c.f., Figs. 5l and 7l), which could lead
352 to the weaker positive GPI anomalies during phases 6+7 in HiRAM than in the observations.

353 Figure 8 further displays budget terms of total GPI anomalies based on model simulations
354 over two of the three rectangle regions previously shown for observations. (Due to rather weak GPI
355 anomalies over the region of 140-120°W, 10-15°N in HiRAM, model results are not displayed for
356 this region). It is shown that the role of potential intensity associated with the ISV for total
357 simulated GPI anomalies is negligible throughout the life cycle of ISV over both of the two box
358 regions. This could be largely ascribed to the lack of the ISV in the SST fields in the GCM as
359 monthly mean SSTs have been specified in these AMIP-type integrations. For the northern box
360 region (110-95°W; 13-18°N), similarly as in the observations, both low-level vorticity and mid-
361 level relative humidity play comparable roles for total GPI anomalies in the GCM. For the southern
362 region along 10°N (115-95°E; 9-13°N), the predominant role of reduced vertical shear for the total
363 GPI anomalies during phase 6 in the observations is not evident in HiRAM (c.f., Figs. 6c and 8b);
364 instead, low-level vorticity takes a leading role in contributing total GPI anomalies in HiRAM,
365 which again could be due to the aforementioned weaker amplitude of reduced vertical wind shear
366 during this period in the GCM. Moreover, simulated total GPI anomalies in phase 7 are dominated
367 by low-level vorticity and mid-level specific humidity. The role of nonlinear effects for total GPI
368 anomalies during this period as noted in the observations is not evident in simulations. As
369 previously discussed, the nonlinear effects for the observed GPI anomalies along 10°N are closely

370 associated with synoptic-scale wave activity, which could be rather challenging to be well resolved
371 by a GCM at 50km horizontal resolution. It remains an intriguing topic for future study to see how
372 the representation of these multi-scale interactive processes can be improved in a GCM with
373 increased horizontal resolutions.

374 **5. Summary and discussion**

375 In the present study, we illustrate that the observed modulations of TC genesis over the EPAC
376 by the ISV can be faithfully represented in the GFDL HiRAM with a horizontal resolution of about
377 50km. This is a remarkable achievement for a climate simulation model (i.e., one that is not
378 initialized from observations) to be capable of depicting the ISV, TC activity, as well as the
379 intimate ISV-TC relationships. Considering the typical predictability of 2-3 weeks for ISV, these
380 results thus suggest great potential for dynamical intraseasonal TC forecasts based on HiRAM.

381 Further analysis indicates that anomalous GPI patterns associated with the EPAC ISV can
382 generally well depict the modulation of TC genesis during different phases of the ISV in both
383 observations and model simulations. Further investigation based on observations is performed to
384 examine the key factors associated with the ISV in modulating TC activity by examining the budget
385 terms of the total GPI anomalies during the ISV life cycle. While Camargo et al. (2009) suggested
386 that mid-level humidity and low-level vorticity are the two most important contributors to the MJO
387 composite GPI anomalies based on a similar analysis for global oceans, the present study based on
388 the EPAC suggests that roles of several factors could be location- as well as ISV phase-dependent.
389 Both the low-level vorticity and mid-level humidity play major roles for observed active TC genesis
390 over northern EPAC warm pool during ISV phases 1+8, and to the west of 120°W between 10-15°N
391 during phases 6+7. Reduced vertical wind shear and the 850hPa vorticity are the two most
392 important factors in modulating TC genesis to the east of 120°W between 10-15°N during ISV
393 phases 6+7. Nonlinear terms are also found to play more active roles over the 10-15°N zone.

394 While in general the observed anomalous patterns of low-level vorticity, mid-level relative
395 humidity and vertical wind shear associated with the EPAC ISV are well represented in HiRAM,

396 model deficiencies are also noted in simulating the anomalous mid-level relative humidity patterns
397 and amplitude of vertical wind shear associated with the EPAC ISV, particularly during ISV phases
398 6+7. These model biases could be responsible for the eastward shift of model TC genesis during
399 this period.

400

401

402

403 *Acknowledgments:* XJ acknowledges support by NOAA MAPP program under Awards
404 NA09OAR4310191 and NA11OAR4310086, and NSF Climate and Large-Scale Dynamics under
405 Award ATM-0934285. DW acknowledges support by U.S. Office of Naval Research under Grant
406 (N00014-10-1-0541) through UCSD, and by NOAA CTB Program through U. HI. We thank the
407 editor, Dr. K. Walsh, and anonymous reviewers for their constructive comments on an earlier
408 version of this manuscript, and B. Kahn and I.-S. Kang for insightful discussions. Thanks also to T.
409 Kubar for his comments and editorial assistance. The Fortran code for GPI calculation was obtained
410 from K. Emanuel's website. Part of this research was carried out at the Jet Propulsion Laboratory,
411 California Institute of Technology, under a contract with the National Aeronautics and Space
412 Administration.

References

413

- 414 Aiyyer, A. and J. Molinari, 2008: MJO and Tropical Cyclogenesis in the Gulf of Mexico and
415 Eastern Pacific: Case Study and Idealized Numerical Modeling. *J. Atmos. Sci.*, **65**, 2691-
416 2704.
- 417 Barrett, B. S. and L. M. Leslie, 2009: Links between Tropical Cyclone Activity and Madden-Julian
418 Oscillation Phase in the North Atlantic and Northeast Pacific Basins. *Mon. Weather Rev.*, **137**,
419 doi:10.1175/2008MWR2602.1, 727-744.
- 420 Bechtold, P., M. Kohler, T. Jung, F. Doblas-Reyes, M. Leutbecher, M. J. Rodwell, F. Vitart, and G.
421 Balsamo, 2008: Advances in simulating atmospheric variability with the ECMWF model:
422 From synoptic to decadal time-scales. *Quart. J. Roy. Meteor. Soc.*, **134**, Doi 10.1002/Qj.289,
423 1337-1351.
- 424 Belanger, J. I., J. A. Curry, and P. J. Webster, 2010: Predictability of North Atlantic Tropical
425 Cyclone Activity on Intraseasonal Time Scales. *Mon. Weather Rev.*, **138**,
426 10.1175/2010mwr3460.1, 4362-4374.
- 427 Bessafi, M. and M. C. Wheeler, 2006: Modulation of south Indian ocean tropical cyclones by the
428 Madden-Julian oscillation and convectively coupled equatorial waves. *Mon. Weather Rev.*,
429 **134**, 638-656.
- 430 Bister, M. and K. A. Emanuel, 2002: Low frequency variability of tropical cyclone potential
431 intensity - 1. Interannual to interdecadal variability. *J. Geophys. Res.*, **107**,
432 10.1029/2001JD000776.
- 433 Camargo, S. J., K. A. Emanuel, and A. H. Sobel, 2007a: Use of a Genesis Potential Index to
434 Diagnose ENSO Effects on Tropical Cyclone Genesis. *J. Clim.*, **20**, doi:10.1175/JCLI4282.1,
435 4819-4834.
- 436 Camargo, S. J., M. C. Wheeler, and A. H. Sobel, 2009: Diagnosis of the MJO Modulation of
437 Tropical Cyclogenesis Using an Empirical Index. *J. Atmos. Sci.*, **66**, Doi
438 10.1175/2009jas3101.1, 3061-3074.
- 439 Camargo, S. J., A. W. Robertson, S. J. Gaffney, P. Smyth, and M. Ghil, 2007b: Cluster analysis of
440 typhoon tracks. Part I: General properties. *J. Clim.*, **20**, 10.1175/JCLI4188.1, 3635-3653.
- 441 de Szoeko, S. P. and C. S. Bretherton, 2005: Variability in the southerly flow into the eastern
442 Pacific ITCZ. *J. Atmos. Sci.*, **62**, 4400-4411.
- 443 Dee, D. P., et al., 2011: The ERA-Interim reanalysis: configuration and performance of the data
444 assimilation system. *Quart. J. Roy. Meteor. Soc.*, **137**, 10.1002/qj.828, 553-597.

445 Elsberry, R. L., M. S. Jordan, and F. Vitart, 2010: Predictability of tropical cyclone events on
446 intraseasonal timescale with the ECMWF monthly forecast model. *Asia-Pacific J. Atmos.*
447 *Sci.*, **46**, 135-153.

448 Emanuel, K. A. and D. S. Nolan, 2004: Tropical cyclone activity and global climate. *Preprints,*
449 *26th Conf. on Hurricanes and Tropical Meteorology*, Amer. Meteor. Soc., Miami, FL, 240-
450 241.

451 Frank, W. M. and P. E. Roundy, 2006: The role of tropical waves in tropical cyclogenesis. *Mon.*
452 *Weather Rev.*, **134**, 2397-2417.

453 Goldenberg, S. B. and L. J. Shapiro, 1996: Physical Mechanisms for the Association of El Niño and
454 West African Rainfall with Atlantic Major Hurricane Activity. *J. Clim.*, **9**, 10.1175/1520-
455 0442, 1169-1187.

456 Gray, W. M., 1984: Atlantic seasonal hurricane frequency I: El-Niño and 30mb Quasi-Biennial
457 Oscillation influences. *Mon. Weather Rev.*, **112**, 1649-1668.

458 Hall, J. D., A. J. Matthews, and D. J. Karoly, 2001: The modulation of tropical cyclone activity in
459 the Australian region by the Madden-Julian oscillation. *Mon. Weather Rev.*, **129**, 2970-2982.

460 Held, I. M. and M. Zhao, 2011: The response of tropical cyclone statistics to an increase in CO2
461 with fixed sea surface temperatures. *J. Clim.*, 10.1175/jcli-d-11-00050.1.

462 Higgins, R. W. and W. Shi, 2001: Intercomparison of the Principal Modes of Interannual and
463 Intraseasonal Variability of the North American Monsoon System. *J. Clim.*, **14**, 403-417.

464 Ho, C. H., J. H. Kim, J. H. Jeong, H. S. Kim, and D. L. Chen, 2006: Variation of tropical cyclone
465 activity in the South Indian Ocean: El Niño-Southern Oscillation and Madden-Julian
466 Oscillation effects. *J. Geophys. Res.*, **111**, 10.1029/2006JD007289.

467 Hsu, H.-H., C.-H. Hung, A.-K. Lo, C.-C. Wu, and C.-W. Hung, 2008: Influence of Tropical
468 Cyclones on the Estimation of Climate Variability in the Tropical Western North Pacific. *J.*
469 *Clim.*, **21**, 10.1175/2007jcli1847.1, 2960-2975.

470 Huffman, G. J., R. F. Adler, B. Rudolf, U. Schneider, and P. R. Keehn, 1995: Global Precipitation
471 Estimates Based on a Technique for Combining Satellite-Based Estimates, Rain-Gauge
472 Analysis, and Nwp Model Precipitation Information. *J. Clim.*, **8**, 1284-1295.

473 Jarvinen, B. R., C. J. Neumann, and M. A. S. Davis, 1984: A tropical cyclone data tape for the
474 North Atlantic Basin, 1886-1983: Contents, limitations, and uses. NOAA Tech. Memo. NWS
475 NHC 22, 21pp.

476 Jiang, X. and D. E. Waliser, 2008: Northward propagation of the subseasonal variability over the
477 eastern Pacific warm pool. *Geophys. Res. Lett.*, **35**, L09814, doi:10.1029/2008GL033723.

478 ———, 2009: Two dominant subseasonal variability modes of the eastern Pacific ITCZ. *Geophys.*
479 *Res. Lett.*, **36**, L04704, doi: 10.1029/2008gl036820.

480 Jiang, X., T. Li, and B. Wang, 2004: Structures and Mechanisms of the Northward Propagating
481 Boreal Summer Intraseasonal Oscillation. *J. Clim.*, **17**, 1022-1039.

482 Jiang, X., D. E. Waliser, M. C. Wheeler, C. Jones, M. N. Lee, and S. D. Schuert, 2008: Assessing
483 the skill of an all-season statistical forecast model for the Madden-Julian oscillation. *Mon.*
484 *Weather Rev.*, **136**, DOI:10.1175/2007MWR2305.1, 1940-1956.

485 Jiang, X., D. Waliser, D. Kim, M. Zhao, K. Sperber, W. Stern, S. Schubert, G. Zhang, W. Wang, M.
486 Khairoutdinov, R. Neale, and M.-I. Lee, 2011: Simulation of the intraseasonal variability over
487 the Eastern Pacific ITCZ in climate models. *Climate Dyn.*, 10.1007/s00382-011-1098-x, 1-20.

488 Kayano, M. T. and V. E. Kousky, 1999: Intraseasonal (30-60 day) variability in the global tropics:
489 principal modes and their evolution. *Tellus Series a-Dynamic Meteorology and*
490 *Oceanography*, **51**, 373-386.

491 Kikuchi, K. and B. Wang, 2010: Formation of Tropical Cyclones in the Northern Indian Ocean
492 Associated with Two Types of Tropical Intraseasonal Oscillation Modes. *J. Meteorol. Soc.*
493 *Japan*, **88**, 475-496.

494 Kim, D., et al., 2009: Application of MJO Simulation Diagnostics to Climate Models. *J. Clim.*, **22**,
495 Doi 10.1175/2009jcli3063.1, 6413-6436.

496 Klotzbach, P. J., 2010: On the Madden-Julian Oscillation-Atlantic Hurricane Relationship. *J. Clim.*,
497 **23**, doi:10.1175/2009JCLI2978.1, 282-293.

498 Knutson, T. R. and K. M. Weickmann, 1987: 30-60 Day Atmospheric Oscillations: Composite Life
499 Cycles of Convection and Circulation Anomalies. *Mon. Weather Rev.*, **115**, 1407-1436.

500 Lau, W. K.-M. and D. E. Waliser, 2011: *Intraseasonal Variability in the Atmosphere-Ocean*
501 *Climate System*. Second ed. Springer, in press, Heidelberg, Germany.

502 Leroy, A. and M. C. Wheeler, 2008: Statistical Prediction of Weekly Tropical Cyclone Activity in
503 the Southern Hemisphere. *Mon. Weather Rev.*, **136**, Doi 10.1175/2008mwr2426.1, 3637-
504 3654.

505 Liebmann, B., H. H. Hendon, and J. D. Glick, 1994: The Relationship between Tropical Cyclones
506 of the Western Pacific and Indian Oceans and the Madden-Julian Oscillation. *J. Meteorol.*
507 *Soc. Japan*, **72**, 401-412.

508 Lin, J.-L., et al., 2006: Tropical Intraseasonal Variability in 14 IPCC AR4 Climate Models. Part I:
509 Convective Signals. *J. Clim.*, **19**, 2665-2690.

510 Lorenz, D. J. and D. L. Hartmann, 2006: The Effect of the MJO on the North American Monsoon.
511 *J. Clim.*, **19**, 333-343.

512 Madden, R. A. and P. R. Julian, 1994: Observations of the 40-50-Day Tropical Oscillation: A
513 Review. *Mon. Weather Rev.*, **122**, 814-837.

514 Magana, V., J. A. Amador, and S. Medina, 1999: The midsummer drought over Mexico and Central
515 America. *J. Clim.*, **12**, 1577-1588.

516 Maloney, E. D. and D. L. Hartmann, 2000a: Modulation of hurricane activity in the Gulf of Mexico
517 by the Madden-Julian oscillation. *Science*, **287**, 2002-2004.

518 ———, 2000b: Modulation of Eastern North Pacific Hurricanes by the Madden-Julian Oscillation. *J.*
519 *Clim.*, **13**, 1451-1460.

520 Maloney, E. D. and S. K. Esbensen, 2003: The Amplification of East Pacific Madden-Julian
521 Oscillation Convection and Wind Anomalies during June-November. *J. Clim.*, **16**, 3482-3497.

522 Maloney, E. D. and S. K. Esbensen, 2007: Satellite and Buoy Observations of Boreal Summer
523 Intraseasonal Variability in the Tropical Northeast Pacific. *Mon. Weather Rev.*, **135**, 3-19.

524 Maloney, E. D. and J. Shaman, 2008: Intraseasonal Variability of the West African Monsoon and
525 Atlantic ITCZ. *J. Clim.*, **21**, 2898-2918.

526 Maloney, E. D., D. B. Chelton, and S. K. Esbensen, 2008: Subseasonal SST variability in the
527 tropical eastern north Pacific during boreal summer. *J. Clim.*, **21**, Doi
528 10.1175/2007jcli1856.1, 4149-4167.

529 Martin, E. R. and C. Schumacher, 2010: Modulation of Caribbean Precipitation by the Madden-
530 Julian Oscillation. *J. Clim.*, doi:10.1175/2010JCLI3773.1, in press.

531 Mo, K. C., 2000: The association between intraseasonal oscillations and tropical storms in the
532 Atlantic basin. *Mon. Weather Rev.*, **128**, 4097-4107.

533 Molinari, J., D. Knight, M. Dickinson, D. Vollaro, and S. Skubis, 1997: Potential vorticity, easterly
534 waves, and eastern Pacific tropical cyclogenesis. *Mon. Weather Rev.*, **125**, 2699-2708.

535 Nakazawa, T., 1988: Tropical Super Clusters within Intraseasonal Variations over the Western
536 Pacific. *J. Meteorol. Soc. Japan*, **66**, 823-839.

537 Rashid, H. A., H. H. Hendon, M. C. Wheeler, and O. Alves, 2011: Prediction of the Madden-Julian
538 oscillation with the POAMA dynamical prediction system. *Climate Dyn.*, **36**,
539 10.1007/s00382-010-0754-x, 649-661.

540 Serra, Y. L., G. N. Kiladis, and K. I. Hodges, 2010: Tracking and Mean Structure of Easterly
541 Waves over the Intra-Americas Sea. *J. Clim.*, **23**, doi:10.1175/2010JCLI3223.1, 4823-4840.

542 Shapiro, L. J., 1987: Month-to-Month Variability of the Atlantic Tropical Circulation and Its
543 Relationship to Tropical Storm Formation. *Mon. Weather Rev.*, **115**, 10.1175/1520-
544 0493(1987)115<2598:mtmvot>2.0.co;2, 2598-2614.

545 Slingo, J. M., et al., 1996: Intraseasonal oscillations in 15 atmospheric general circulation models:
546 Results from an AMIP diagnostic subproject. *Climate Dyn.*, **12**, 325-357.

547 Small, R. J., S. P. De Szoeke, and S.-P. Xie, 2007: The Central American midsummer drought:
548 Regional aspects and large-scale forcing. *J. Clim.*, **20**, Doi 10.1175/Jcli4261.1, 4853-4873.

549 Vitart, F., 2009: Impact of the Madden Julian Oscillation on tropical storms and risk of landfall in
550 the ECMWF forecast system. *Geophys. Res. Lett.*, **36**, L15802, Doi 10.1029/2009gl039089, -.

551 Vitart, F. and F. Molteni, 2010: Simulation of the Madden-Julian Oscillation and its teleconnections
552 in the ECMWF forecast system. *Quart. J. Roy. Meteor. Soc.*, **136**, Doi 10.1002/Qj.623, 842-
553 855.

554 Vitart, F., A. Leroy, and M. C. Wheeler, 2010: A Comparison of Dynamical and Statistical
555 Predictions of Weekly Tropical Cyclone Activity in the Southern Hemisphere. *Mon. Weather
556 Rev.*, **138**, Doi 10.1175/2010mwr3343.1, 3671-3682.

557 Waliser, D. E., 2006: Predictability of tropical intraseasonal variability. *Predictability of weather
558 and climate*, T. Palmer and R. Hagedorn, Eds., Cambridge University Press 718.

559 ———, 2011: Predictability and Forecasting. *Intraseasonal Variability in the Atmosphere-Ocean
560 Climate System*, Second ed. W. K. M. Lau and D. E. Waliser, Eds., Springer, Heidelberg,
561 Germany, In Press.

562 Waliser, D. E., et al., 2003: AGCM simulations of intraseasonal variability associated with the
563 Asian summer monsoon. *Climate Dyn.*, **21**, DOI 10.1007/s00382-003-0337-1, 423-446.

564 Wang, B. and X. Zhou, 2008: Climate variation and prediction of rapid intensification in tropical
565 cyclones in the western North Pacific. *Meteorol. Atmos. Phys.*, **99**, 10.1007/s00703-006-0238-
566 z, 1-16.

567 Weare, B. C. and J. S. Nasstrom, 1982: Examples of Extended Empirical Orthogonal Function
568 Analyses. *Mon. Weather Rev.*, **110**, 481-485.

569 Wheeler, M. C. and H. H. Hendon, 2004: An All-Season Real-Time Multivariate MJO Index:
570 Development of an Index for Monitoring and Prediction. *Mon. Weather Rev.*, **132**, 1917-1932.

571 Wu, M. L. C., S. D. Schubert, M. J. Suarez, and N. E. Huang, 2009: An Analysis of Moisture
572 Fluxes into the Gulf of California. *J. Clim.*, **22**, Doi 10.1175/2008jcli2525.1, 2216-2239.

573 Zhang, C. D., 2005: Madden-Julian oscillation. *Rev. Geophys.*, **43**, RG2003, DOI:
574 10.1029/2004RG000158, 36.

575 Zhang, Q. and H. v. d. Dool, 2011: Relative merit of model improvement versus availability of
576 retrospective forecasts: The case of CFS MJO prediction. *Geophys. Res. Lett.*, submitted.

577 Zhao, M. and I. M. Held, 2010: An Analysis of the Effect of Global Warming on the Intensity of
578 Atlantic Hurricanes Using a GCM with Statistical Refinement. *J. Clim.*, **23**, Doi
579 10.1175/2010jcli3837.1, 6382-6393.

580 Zhao, M., I. M. Held, S. J. Lin, and G. A. Vecchi, 2009: Simulations of Global Hurricane
581 Climatology, Interannual Variability, and Response to Global Warming Using a 50-km
582 Resolution GCM. *J. Clim.*, **22**, Doi 10.1175/2009jcli3049.1, 6653-6678.

583

584 **List of Figures**

585 **Figure 1** TC genesis over the EPAC from 1998-2008 during different ISV phases in the
586 observations (left) and HiRAM simulations (right). Shading in each panel represents observed
587 or simulated anomalous rainfall pattern with the bottom color bar for scale.

588 **Figure 2** Counts of TC genesis over the EPAC as a function of ISV phases (phases 1 to 8 for strong
589 ISVs while phase “0” represents weak ISV period): (a) observations and (b) HiRAM
590 simulations. The weaker ISV is defined by amplitude of $\sqrt{PC_1^2 + PC_2^2} < 1.0$.

591 **Figure 3** Modulation of TC movements by the EPAC ISV: (a) observations; (b) HiRAM
592 simulations. Each mark “+” denotes the six-hourly TC location in each ISV phase in
593 observations or GCM.

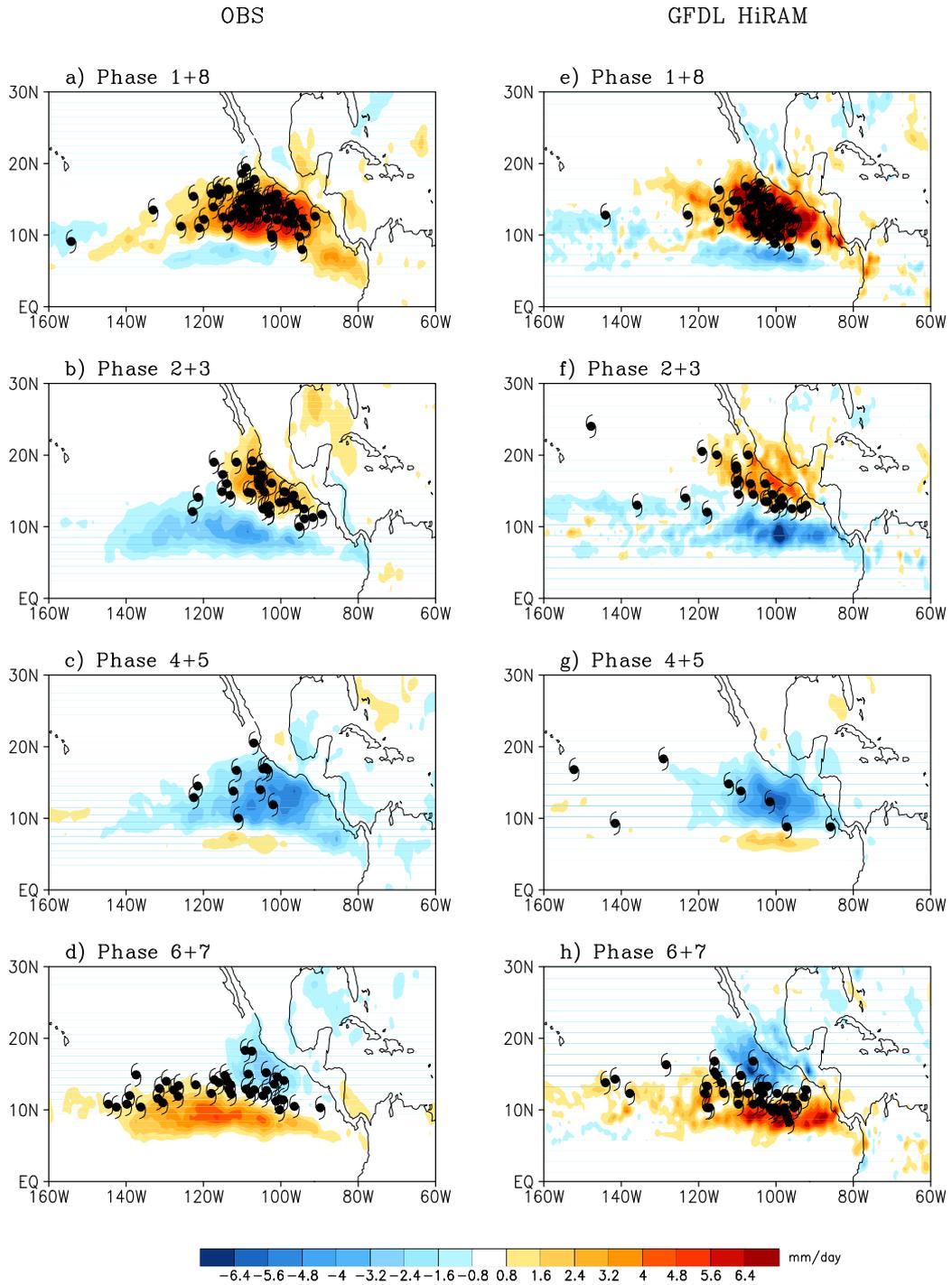
594 **Figure 4** Composite GPI anomalies (shaded; see color scales on the right) at different ISV phases
595 along with corresponding anomalous rainfall distributions (contours; units: mm day^{-1}) and TC
596 genesis events based on observations (left panels) and HiRAM simulations (right panels).
597 Solid (dashed) contours represent positive (negative) rainfall anomalies with an interval of 1
598 mm day^{-1} , and contours “0” are omitted.

599 **Figure 5** (Shaded) Composite total GPI anomalies during each ISV phase (see color bar at the
600 bottom). Left: vorticity (contours with an interval of $1 \times 10^{-6} \text{ s}^{-1}$) and winds (vectors; see scale
601 in Panel a); Middle: 600hPa relative humidity (contours with an interval of 1%); Right: 200-
602 850hPa wind shear (contours with interval of 0.5 ms^{-1}). Solid (dashed) contours denote
603 positive (negative) values, and contours “0” are omitted in all the panels. All these above
604 fields are based on ERA-Interim reanalysis.

605 **Figure 6** Budget terms of the total observed GPI anomalies over three selected regions with active
606 TC genesis (see green boxes in Figure 5) as a function of ISV phase.

607 **Figure 7** Same as in Fig. 5 but for HiRAM simulations.

608 **Figure 8** Same as in Fig. 6 but for HiRAM simulations over two selected regions denoted in Fig. 7e
609 and 7l.



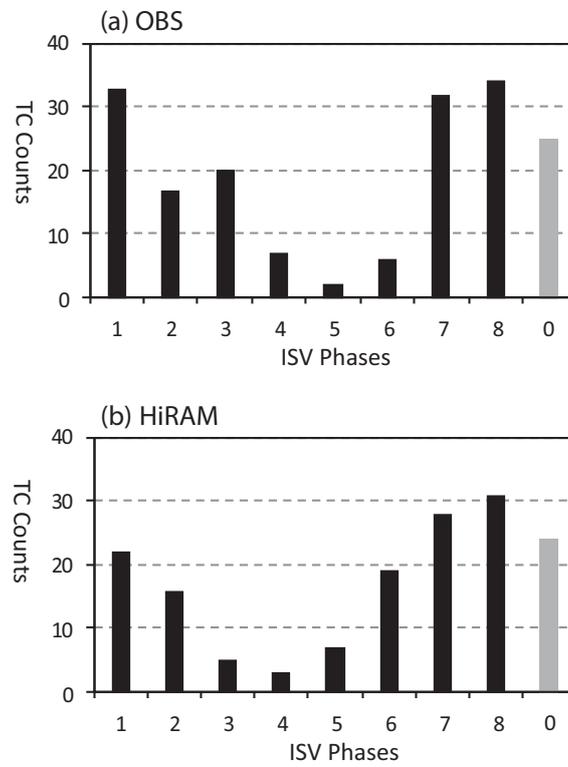
611

612 Figure 1 TC genesis over the EPAC from 1998-2008 during different ISV phases in the
 613 observations (left) and HiRAM simulations (right). Shading in each panel represents observed
 614 or simulated anomalous rainfall pattern with the bottom color bar for scale.

615

616

617



618

619 Figure 2 Counts of TC genesis over the EPAC as a function of ISV phases (phases 1 to 8 for strong

620 ISVs while phase “0” represents weak ISV period): (a) observations and (b) HiRAM

621 simulations. The weaker ISV is defined by amplitude of $\sqrt{PC_1^2 + PC_2^2} < 1.0$.

622

623

624

625

626

627

628

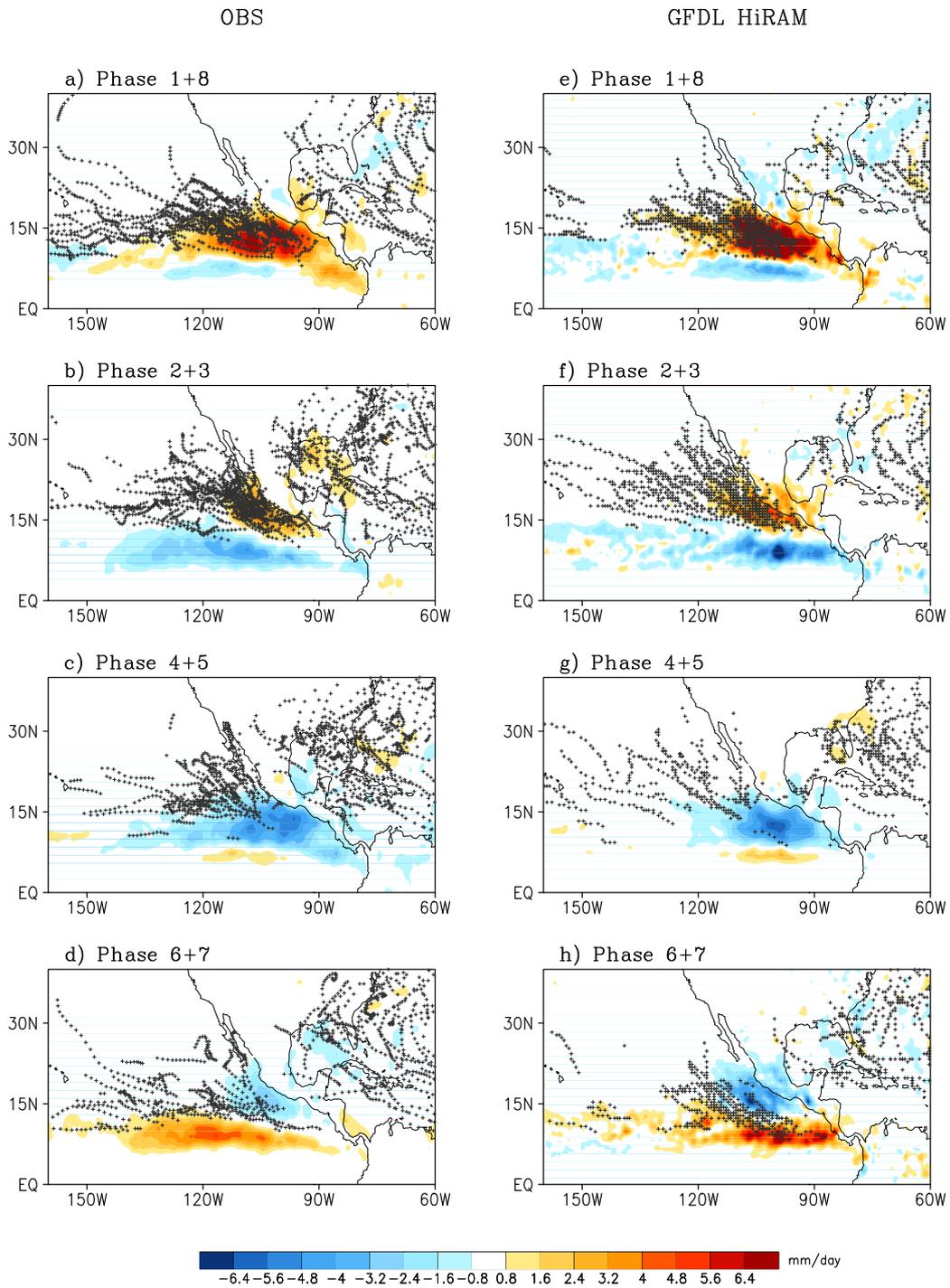
629

630

631

632

633



635

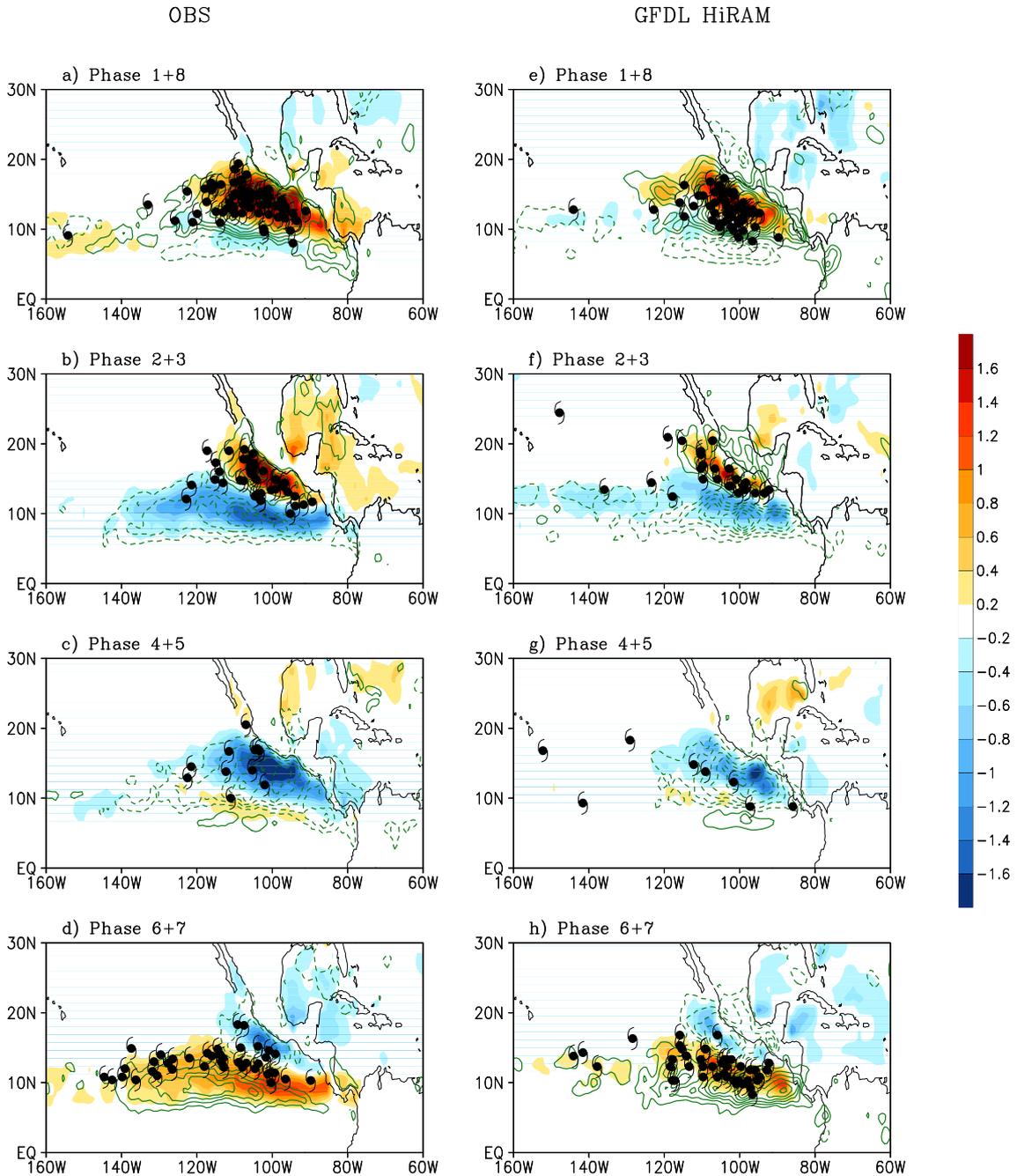
636

637

638

639

Figure 3 Modulation of TC movements by the EPAC ISV: (a) observations; (b) HiRAM simulations. Each mark “+” denotes the six-hourly TC location in each ISV phase in observations or GCM.

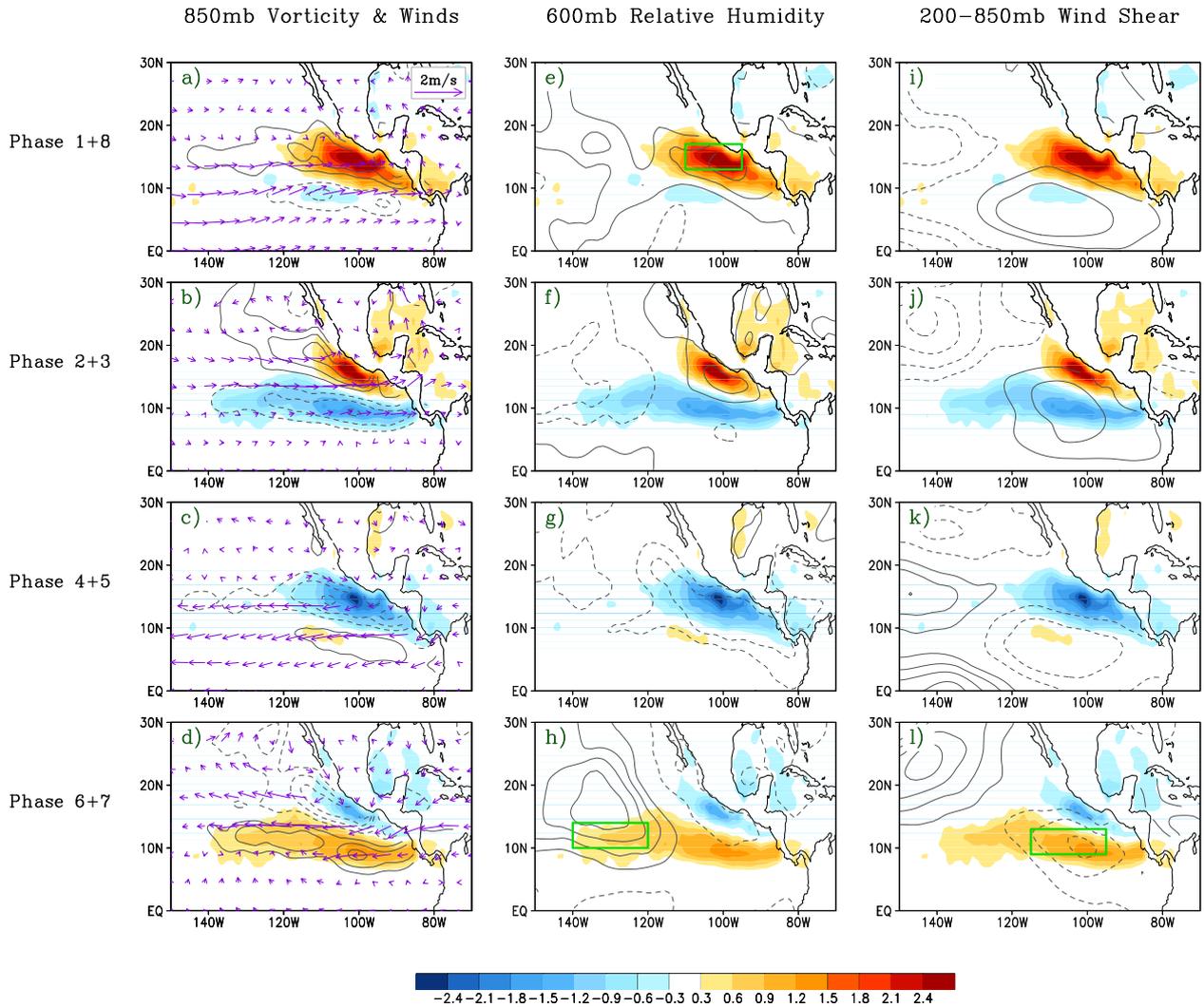


641

642

Figure 4 Composite GPI anomalies (shaded; see color scales on the right) at different ISV phases
 643 along with corresponding anomalous rainfall distributions (contours; units: mm day^{-1}) and TC
 644 genesis events based on observations (left panels) and HiRAM simulations (right panels).
 645 Solid (dashed) contours represent positive (negative) rainfall anomalies with an interval of 1
 646 mm day^{-1} , and contours “0” are omitted.

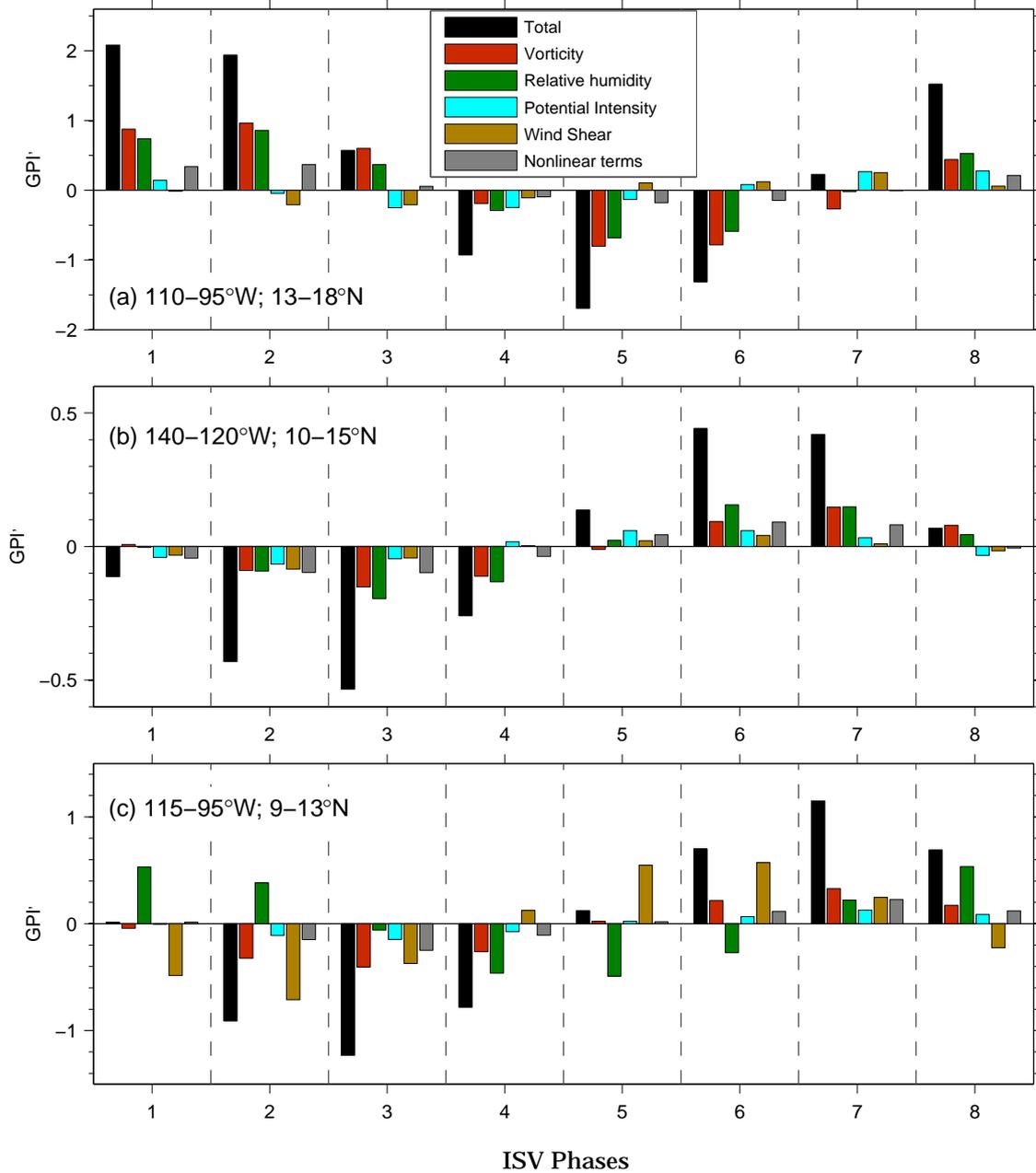
647
648
649
650



651
652
653
654
655
656
657
658
659
660

Figure 5 (Shaded) Composite total GPI anomalies during each ISV phase (see color bar at the bottom). Left: vorticity (contours with an interval of $1 \times 10^{-6} \text{ s}^{-1}$) and winds (vectors; see scale in Panel a); Middle: 600hPa relative humidity (contours with an interval of 1%); Right: 200-850hPa wind shear (contours with interval of 0.5 ms^{-1}). Solid (dashed) contours denote positive (negative) values, and contours “0” are omitted in all the panels. All these above fields are based on observations.

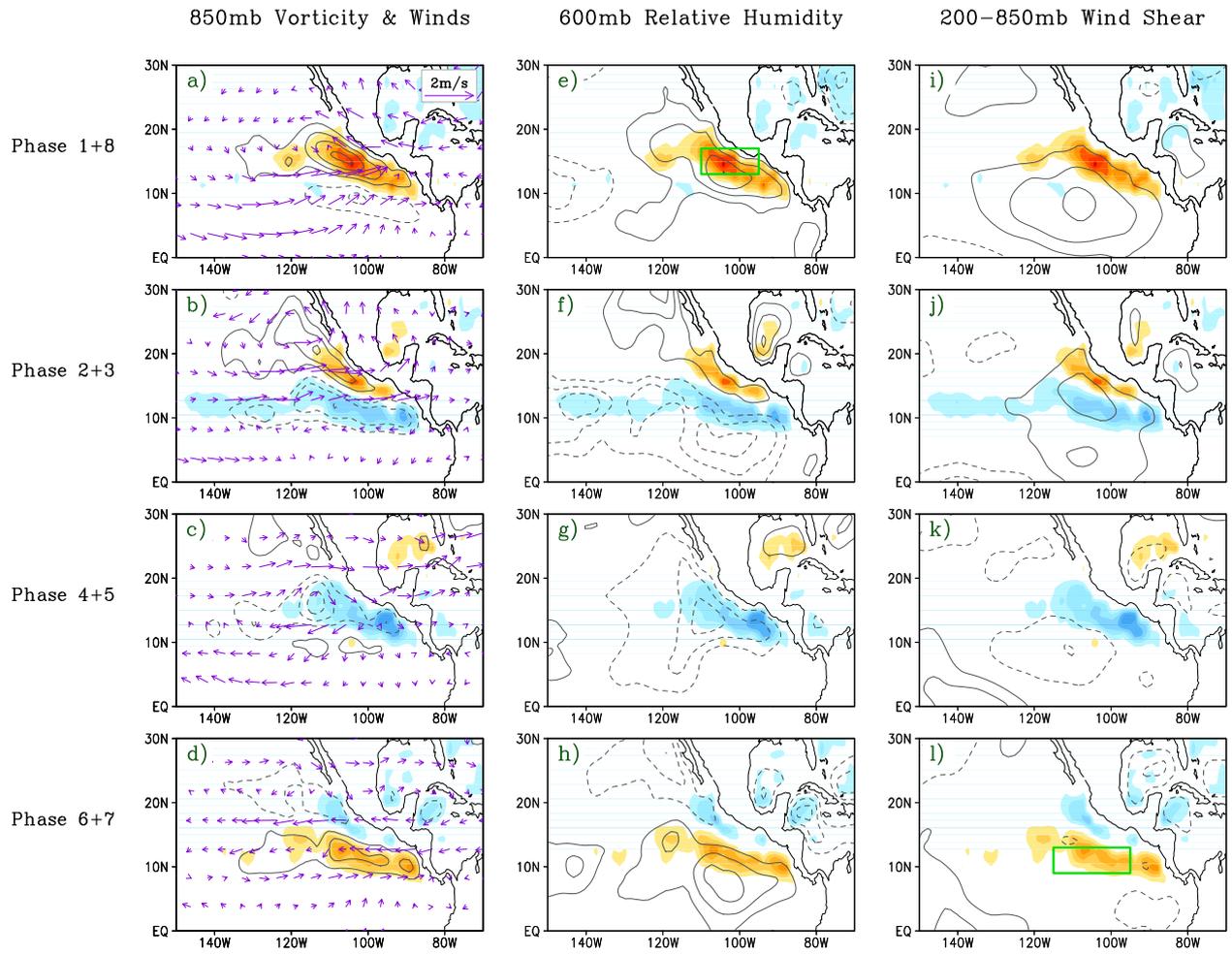
661
662
663
664



665
666
667
668
669

Figure 6 Budget terms of the total observed GPI anomalies over three selected regions with active TC genesis (see green boxes in Figure 5) as a function of ISV phase.

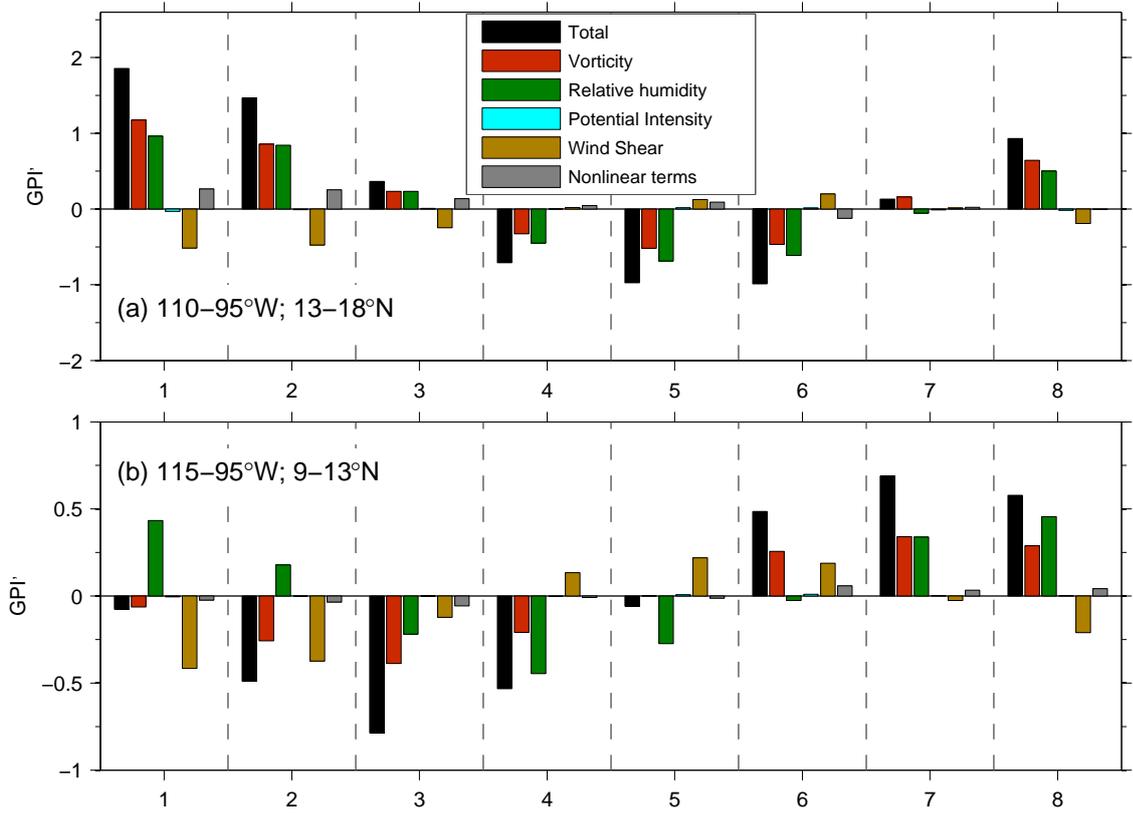
670
671
672
673
674
675



676
677
678
679
680
681
682
683

Figure 7 Same as in Fig. 5 but for HiRAM simulations.

684
685
686
687
688



689
690 Figure 8 Same as in Fig. 6 but for HiRAM simulations over the two selected rectangle regions in
691 Fig. 7e and 7l.