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3 **Observed Recent Trends in Tropical Cyclone Rainfall**
4 **Over Major Ocean Basins**
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Abstract

In this study, we use Tropical Rainfall Measuring Mission (TRMM) and Global Precipitation Climatology Project (GPCP) rainfall data together with historical storm track records to examine the trend of tropical cyclone (TC) rainfall in major ocean basins during recent decades (1980-2007). We find that accumulated total rainfall along storm tracks for all tropical cyclones shows a weak positive trend over the whole tropics. However, total rainfall associated with weak storms, and intense storms (Category 4-5) both show significant positive trends, while total rainfall associated with intermediate storms (Category 1-3) show a significant negative trend. Storm intensity defined as total rain produced per unit storm also shows increasing trend for all storm types. Basin-wide, from the first half (1980-1993) to the second half (1994-2007) of the data period, the North Atlantic shows the pronounced increase in TC number and TC rainfall while the Northeast Pacific shows a significant decrease in all storm types. Except for the Northeast Pacific, all other major basins (North Atlantic, Northwest Pacific, Southern Oceans, and Northern Indian Ocean) show a significant increase in total number and rainfall amount in Category 4-5 storms. Overall, trends in TC rainfall in different ocean basins are consistent with long-term changes in the ambient large-scale environment, including SST, vertical wind shear, sea level pressure, mid-tropospheric humidity, and Maximum Potential Intensity (MPI). Notably the pronounced positive (negative) trend of TC rainfall in the North Atlantic (Northeast Pacific) appears to be related to the most (least) rapid increase in SST and MPI, and the largest decrease (increase) in vertical wind shear in the region, relative to other ocean basins.

72 Introduction

73 Tropical cyclones (TC) are among nature's most destructive forces. Whether the number and
74 the intensity of TCs have changed or will change in a warming climate has been the subject
75 of many studies. However the issue is far from settled, due to large interannual and
76 interdecadal variability in the frequency and intensity of TCs and limitation in the availability
77 and quality of global long-term historical records of TCs [Pielke et al., 2005; Shepherd and
78 Knutson, 2006; Landsea et al., 2006; 2008; Knutson et al., 2010]. Despite large uncertainties
79 in observed long-term TC frequency and intensity change, numerical models have predicted
80 that TC related rainfall rates are likely to increase with greenhouse warming [Knutson and
81 Tuleya, 2004; Bengtsson, 2007]. Such models have also projected substantial increase in
82 storm-centered rainfall in the late 21st century in the range of +3-37%, resulting from
83 increased evaporation from the warmer ocean and increased moisture in the atmosphere. For
84 TC's, where moisture convergence plays a key role, increased available moisture in the
85 atmosphere is expected to lead to greater rain rates.

86 Previous studies found that the frequency of extreme rainfall has increased in the last
87 several decades over land from land-based precipitation data set [Karl and Knight, 1998;
88 Groisman et al., 2004], as well as over the tropical oceans based on long-term satellite
89 precipitation data [Lau and Wu, 2007]. In the coastal regions of South China and
90 southeastern United States, TCs account for a significant portion of total rainfall, and even
91 more so for extreme rainfall, and an increase in major TCs may be related to extreme rain
92 events [Wu and Zhai, 2007; Shepherd et al, 2007]. Lau et al. [2008] recently studied the
93 relationship between TC related rainfall and extreme rain events in North Atlantic and
94 Northwest Pacific using nearly 30-year of GPCP pentad data. They found that TCs contribute

95 to increasing amount of extreme rain events in the two ocean basins, with more pronounced
96 signal in the North Atlantic Ocean than the Northwest Pacific, in part due to the larger
97 increasing in warm pool size in the former. Despite well founded theoretical expectation and
98 modeling projections, a long-term ($>$ decades) trend in observed TC related rainfall has not
99 been established [Knutson et al. 2010].

100 In this study, possible long-term trends in TC related rainfall is examined for all the
101 ocean basins from 1980 to 2007 using GPCP data set and historical storm track data. Section
102 2 describes the data and methodology. Section 3 examines the relationship between TC rain
103 and TC intensity from more recent high resolution TRMM data and GPCP pentad data.
104 Section 4 shows results of long-term trend in TC rain. Section 5 examines trends in large-
105 scale environment that may be responsible for the observed changes in TC rainfall. Section 6
106 summarizes the results.

107 **1. Data and methodology**

108 The precipitation data used in this study include the pentad rainfall from the Global
109 Precipitation Climatology Project (GPCP) for 1979–2007 [Xie et al. 2003], the 3-hourly
110 Tropical Rainfall Measurement Mission (TRMM) Multi-satellite Precipitation Analysis
111 (TMPA) data for 1998–2007 [Huffman et al., 2007]. The 3-hourly TMPA data is used to
112 generate daily gridded precipitation data. The 6-hourly best storm track data for the North
113 Atlantic, Northeast Pacific are from the National Hurricane Center [Jarvinen et al. 1984] and
114 those for the Western North Pacific, Northern Indian Ocean and Southern Ocean are from
115 Joint Typhoon Warning Center [Chu et al., 2000], with a correction of wind speed [Emanuel,
116 2005]. Data for sea surface temperature (SST) are from the Hadley Center [Rayner et al.
117 2003]. The NCEP-NCAR Reanalysis-1 [Kalnay et al., 1996] surface air temperature and
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120 vertical temperature, humidity, and wind profiles are used to compute changes in TC-related
121 large-scale environmental factors such as vertical wind shear, relative humidity, atmosphere
122 instability and the Maximum Potential Index (MPI). The spatial resolution is 2.5° latitude x
123 2.5° longitude for GPCP, NCAR-NCEP Reanalysis, 0.25° x 0.25° for TRMM, and 1° x 1° for
124 SST.

125 For this study, we estimate the TC-related rainfall for each storm over its life-time
126 (hereafter denoted as TC-rain) using the 6-hourly storm track data. Following Larson *et al.*
127 [2005] and Rodgers *et al.* [2000, 2001], we define TC-rain as the accumulated amount that
128 falls within an area of 500 km radius, from the center of the TC within the same day (for
129 TRMM data) and the same pentad (for the GPCP pentad data) along the entire track. TC rain
130 estimated from TRMM 3B42 is found to be higher than those estimated from TRMM Radar
131 (2A25) and Microwave imager (2A12) estimates, probably due to overestimate of rainfall by
132 IR-based rainfall retrievals for high and cold clouds [Jiang and Zipser, 2010]. However,
133 because of GPCP's coarse spatial and temporal resolutions, counting the entire pentad as TC-
134 rain can result in some over-estimation of actual TC-rain because non-TC related rain might
135 occur within the same pentad for the grid box. As a result, TC-rain amount calculated from
136 GPCP pentad data is more than those estimated from TRMM TMPA for the same pentad.
137 Since latent heat is released in rainfall, TC-rain can be identified as the total energy generated
138 by a given TC. To facilitate comparison with other forms of energy, hereafter, we convert
139 the TC-rain amount to the unit of Energy Year ($EY = 5.10 \times 10^{20}$ Joule), which is the
140 estimated world primary energy consumption in 2007 (<http://eia.gov/aer/txt/ptb1103.html>).

141 Figure 1 shows scatter plots of TC-rain estimated from TRMM and GPCP data for
142 different TC categories according to the Simpson-Saffir classification. The GPCP estimates

143 are systematically higher than that of TRMM by 30-40 percent for all hurricane categories,
144 possibly due to the coarse spatial and temporal resolution of the GPCP data. However, within
145 GPCP there are no systematic differences among TC categories, and for different oceans
146 basins (figure not shown). Therefore it is reasonable to assume that despite its coarse
147 estimate, the GPCP estimate of TC-rain would be self-consistent over time and can be
148 compared over different ocean basins.

149 Another concern related to the use of GPCP pentad data has to do with temporal
150 inhomogeneity inherent in many long-term satellite data sets [Lau et al., 2008; Zhou and Lau,
151 2010]. The GPCP rainfall is a merged product that combines available surface rain gauge and
152 operational satellite rainfall estimates to provide a state-of-the-art, multi-decadal global
153 dataset for climate studies [Adler et al., 2003; Xie et al., 2003]. Examination of the pentad
154 data shows significant shifts in tropical rainfall Probability Distribution Function (PDF) with
155 notably more extreme rainfall after inclusion of SSM/I data in 1987 [Zhou and Lau, 2010].
156 However, the TC-rain data subset we construct for this study does not show any obvious bias
157 between pre-SSM/I and post-SSMI periods (see Figure 3), presumably the bias are embedded
158 more in other heavy rainfall systems such as ITCZ and monsoons, which are not included in
159 our TC-rain construction. While uncertainties undoubtedly exist with respect to some
160 inhomogeneity in the GPCP data, we are reasonably sure that the qualitative results regarding
161 TC-rain are not affected.

162

163 **2. TC-rain versus TC-intensity**

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165 Numerical simulations suggest that TC rainfall will increase as storm intensity
166 increases [Knutson and Tuleya, 2004]. TC rain amount over the life-time of a storm, defined
167 as TC-rain, depends on multiple factors governing storm intensity and duration. The radial

168 distribution of azimuthally averaged instantaneous rain rates from storm center shows high
169 correlation with storm intensity [Lonfat et al., 2004], but they are strongly modulated by
170 vertical wind shear, total precipitable water, horizontal moisture convergence, ocean surface
171 flux and topography [Lonfat et al., 2007, Jiang et al., 2008]. So far no study has examined
172 the relationship between TC-rain and storm intensity. Figure 2 shows the mean and standard
173 deviation of TC-energy (TC- rain in units of energy) per storm versus the Simpson-Saffir
174 storm category based on GPCP and TRMM TMPA data. TC-energy increases almost linearly
175 with storm intensity, with TS and Category 1, 2 storms on average generating less than 2×10^{20}
176 Joule of energy (0.5~1.0 EY) per storm, while for Category 3, 4 and 5, producing 5×10^{20}
177 Joule (1.0~1.5 EY) per storm (Figure. 2a). The more concurrent relationship between
178 daily mean storm wind (averaged from 6 measurements per day) versus daily storm rain also
179 shows positive correlation (Figure not shown), as reported by [Lonfat, 2004 and Jiang et al.,
180 2008]. Large standard deviations are expected due to many other factors affecting TC-rain
181 [Jiang et al., 2008]. The mean TC-rain from GPCP is larger than that from TRMM (Figure.
182 2b) for the same reasons discussed before. The above results demonstrate that GPCP and
183 TRMM TC-rain data have similar properties with respect to TC intensity for the overlapping
184 period and for all rainfall categories, providing some reassurance that GPCP TC-rain can be
185 used for studies for the longer data period (1979-2007).

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187 **3. Recent trends in TC-Rain**

188 In order to examine the trend in TC-rain for each basin and for different TC
189 categories, we compute TC-rain from each category and for each ocean basin. The total TC-
190 rain for each basin is calculated by adding the TC-rain from all the storms in the year from

191 January to December for the northern hemisphere. For the southern hemisphere, the tropical
192 cyclone year starts from the previous July to current June. Figure 3 shows the global total
193 number of TCs and TC-energy from each storm category from 1980 to 2007. The linear
194 trend of total TC number and TC-energy are computed by linear regression. Globally the
195 total number of tropical storms (TS) increases slightly by 3.8% per decade, with 90%
196 confidence level (c.l.) from the beginning to the end of the data period. The TC numbers in
197 Category 1, 2 and 3 decrease by 6.7%, 13.3% and 17.6% per decade, respectively at 90% c.l..
198 For Category 4 and 5, the TC numbers increases by 21% and 42% per decade, respectively,
199 with 99% c.l.. These trends are consistent with Webster et al. [2005], but substantially higher
200 than Klotzbach [2006] and Kossin et al. [2007]. The TC energy, converted from TC-rain,
201 generally follows the trend of TC number. However, for TS and Category 4 and 5 storms,
202 the percentage increase of TC energy is 6.2%, 24.8%, and 43.2% per decade, respectively,
203 slightly higher than the corresponding percentage increase in TC number. For Category 1, 2
204 and 3 storms, the decrease in total TC-energy is less than the rate of decrease in TC number
205 and statistically insignificant, indicating that TC-rain-per-storm might have increased in all
206 the categories. Indeed, calculation shows that the mean energy per storm has increased by 2.6%
207 for TS and by 9.5% for Category 3 TCs (figures omitted).

208 To examine the trend of TC-rain in each basin, we divide the TCs into 3 groups: TS,
209 Cat 1-3, Cat 4-5, identified as weak, intermediate and intense storms, respectively. This
210 grouping is consistent with the previously discussed linear trends within each group, and has
211 the advantage of enabling relatively homogeneous number of samples in each group. To
212 focus on the long-term (>decadal) changes, we compute the TC-rain and related statistics
213 separately for two equal halves of 14 years each, i.e., 1980-1993, and 1994-2007, and

214 compare the statistics between the two halves. Figure 4 shows the TC numbers in each
215 group in the first and second half of the period for each basin and global total (All-TC). For
216 the North Atlantic, TC numbers in all categories has significantly increased from the first to
217 the second period. The percentage increase for TS, Cat 1-3, Cat 4-5, and All-TC are 59%,
218 27%, 149% and 53%, respectively, all significant at 90% level. For the Northwest Pacific,
219 the numbers of TS and Cat 4-5 have increased by 31% and 25%, respectively, while the
220 storm of medium intensity (Cat 1-3) has decreased by 23%. Overall the total TC number still
221 shows a significant increase of 8%. For the Northeast Pacific, TC counts in all groups have
222 decreased, with the number of medium TCs showing a pronounced and significant decrease
223 of 36%, culminating in a total basin-wide reduction of 23%. The inverse relationship
224 between TC activities in Northeast Pacific and North Atlantic is consistent with previous
225 study which reported out-of-phase variations of TC activities in the two ocean basins on both
226 interannual and multi-decadal time scales due to response to opposite changes of atmosphere
227 instability and vertical wind shear associated with ENSO and Atlantic Multi-decadal
228 Oscillations [Wang and Lee, 2009; 2010]. For the Southern Oceans (including both the
229 Southern Indian and the Southern Pacific Ocean), there is a significant increase in the
230 number of Cat 4-5 storms, but not in TS and Cat 1-3 storms. Compared to other ocean basins,
231 the Northern Indian Ocean has the least number of TCs with an average of 4 to 5 per year.
232 Here we note a significant increase of Cat 1-3 storms and Cat 4-5 storms in Northern Indian
233 Ocean. However because of the small number of TCs, the Northern Indian Ocean storms
234 have little effects on the global statistics. Globally, both TS and Cat 4-5 have significant
235 increase in numbers while Cat 1-3 has a 13% decrease. The global total number of TC has a
236 small but significant 5% increase.

237 The TC-energy by category shows similar patterns as the total number, indicating that
238 total numbers of TC are dominating the TC energy (Figure 5). It is interesting to note that 4
239 out of 5 basins (Northeast Pacific being the exception), the TC-energy in Cat 4-5 storms has
240 increased. Globally, the increase in TC-energy from Cat 4-5 has significantly increased by
241 54%, considerably faster than the rate of increase of the TC numbers (Figure 4). The global
242 total TC –rain from all TCs has also increased by a significant 14% (90% c.l.). This suggests
243 that overall there is stronger signal, or higher sensitivity of TC-rain compared to TC number
244 to the changing tropical environment (see further discussion in Section 5). Figure 6 shows
245 the change in energy per-storm (EPS) for each basin and the global total, by storm categories.
246 Over all ocean basins, there is no significant change in EPS for TS. For Cat 1-3, only two
247 ocean basins, *i.e.*, the North Atlantic and the Southern Oceans, show significant 39% and 15 %
248 increase in EPS, respectively. Interestingly, there is a significant reduction in Cat 1-3 EPS
249 over the Northeast Pacific. For Cat 4-5 storm, only the Northwest Pacific shows significant
250 increase. Globally, there is a significant overall 8% EPS increase for Cat 4-5, and 8%
251 increase for all categories. Further computations have shown (figures omitted) that there is no
252 significant increase in TC duration, implying that the EPS increase is largely due to heavier
253 TC-rain, with more latent energy release over the approximately same storm life-span for the
254 two periods.

255 The systematic change in EPS during the two periods is further illustrated by
256 examining the change in the mean probability distribution functions (PDFs) of rain-per-storm
257 from the first to second period (Fig. 7). Notably Northwest Pacific has the most energetic
258 (wettest) storm, with the top 10% TC-rain EPS threshold exceeding 2.7 EY, followed in
259 order by the Southern Ocean (2.0 EY), the North Atlantic (1.9 EY), the Northeast Pacific (1.5

260 EY) and the least energetic Indian Ocean (1.4 EY). Globally, 50% of the storm produced
261 more than an equivalent of 1.0 EY of latent energy, while 10% of the storms produced close
262 to 2.2 EY. The difference of PDF shows clearly that there is an internal redistribution of the
263 TC-energy among storms within each basin during the two periods. Over the North Atlantic
264 and Southern Oceans, the PDF-difference suggests a shift of the rainfall spectrum to the right,
265 *i.e.*, more frequent occurrence of energetic storms and less of weak (below 50th percentile)
266 storms. On the other hand, the Northwest Pacific and Northeast Pacific show a flattening of
267 the PDF, *i.e.*, increased frequency of occurrence of the most and the least energetic storms
268 respectively, coupled with a reduction in moderately energetic storms. Interestingly, the PDF
269 difference in the Northern Indian Ocean suggests a shift towards a sharpening of the PDF
270 toward the intermediate and highly energetic storm. Here the small sample size may have
271 contributed to large uncertainties towards the tail end of the spectrum ($EY > 2$). The internal
272 redistribution of the TC-rain PDF may also be affected by inter-basin teleconnection through
273 adjustment in the large scale tropical circulations to long-term climate forcing, affecting the
274 environment for TC development (see further discussion in Section 5). The net result of
275 these redistribution and adjustment processes is a shift in the global EPS PDF, with an
276 increase in more energetic storm ($EY > 1.5$ EY), and a reduction in the lesser and moderately
277 energetic storms ($EY < 1.5$ EY), over all major tropical oceans.

278

2794. Long-term changes in the large-scale environment

280

281 It is well known that the global or basin-scale tropical cyclone activities are affected
282 by large-scale environment such as sea surface temperature (SST), vertical wind shear, and
283 atmospheric stability [Gray 1968, 1984; Landsea et al., 1998; Shapiro and Goldenberg, 1996;
284 Cheung, 2004; Vecchi and Soden, 2007; Chan 2009]. For individual TC, the moisture

285 content in the atmosphere and moisture convergence are important factors affecting storm
286 rainfall [Jiang, 2008; Knutson et al., 2010]. To examine various large-scale environmental
287 factors that might affect the basin total TC rainfall, we compute the linear trends of quantities
288 that are known to affect tropical cyclone formation, genesis and intensity. These include SST,
289 vertical wind shear (magnitude of vector difference of horizontal wind between 850 mb and
290 200 mb), sea level pressure, mid-tropospheric relative humidity, and the Maximum Potential
291 Intensity (MPI) for velocity [Emanuel, 1986; 1995]. The MPI is a measure of the
292 thermodynamic limit on the intensity of a storm based on sea surface temperature and
293 atmospheric temperature and moisture profiles. The trends are computed for the period
294 1980-2009 for July-November (JASON) and December-April (DJFMA) for the TC seasons
295 in the northern and southern hemispheres respectively.

296 Figure 8a shows that in JASON, there is a widespread warming trend over the
297 tropical and subtropical Pacific, the Indian Ocean and the North Atlantic, and cooling over
298 the eastern and southeast Pacific, South Atlantic and the southwestern Indian Ocean. The
299 magnitude of the SST increase in the North Atlantic TC domain (dotted rectangle in Figure.
300 8a), including the Maximum Development Region (MDR, 10°N-20°N, 20°W-85°W) is
301 notably the largest over the other major basins. The lesser warming over the western Atlantic
302 and the Gulf of Mexico region may reflect the increased occurrence of more energetic TCs,
303 whose stronger surface winds could cool the surface ocean by enhanced Ekman upwelling.
304 The Northwest Pacific TC domain generally lies under warmer water in recent decades, but
305 the large increase in SST occurs in the midlatitudes around 35°N-40°N outside of the TC-
306 domain. Similarly a large body of warmer water is found in the South Oceans domain
307 (Figure. 8d) in the region of the South Pacific Convergence Zone. The warmer water

308 enhances evaporation, and may have contributed to the more energetic storms found in the
309 North Atlantic, the northwestern Pacific and the Southern Oceans. In contrast, a large area of
310 cooling SST is found in the Eastern Pacific TC domain possibly a contributing factor to the
311 decreasing trend in the TC number and TC rainfall in the Northeast Pacific. Over the Indian
312 Ocean TC domain, the SST signal seems to be quite weak.

313 In JASON, the atmospheric vertical wind shear trend pattern shows near- zonal bands
314 of negative and positive features (Figure 8b). The negative trends appear along the equator
315 while the positive trends appear in the margin of ITCZ, probably associated with the shift of
316 ITCZ and Hadley circulation [Zhou et al., 2011]. Large reduction of vertical wind shear,
317 favorable for enhanced tropical cyclone activities is found in the tropical Atlantic,
318 particularly over the MDR. Most of the TC domain in the northern Indian Ocean observes
319 decreased wind shear, while the TC domain in the northwest Pacific has a band of increasing
320 wind shear near 20°N surrounded by reduced wind shear to the south and the north. On the
321 contrary, the TC domain in the Northeast Pacific is dominated by increasing vertical wind
322 shear, which suppresses tropical cyclone activities. For the DJFMA season, the wind shear
323 trend over the TC domain of the Southern Oceans (Fig. 8e) is generally negative especially in
324 the equatorward side of the domain, favoring TC genesis there. All basins in the northern
325 hemispheres, except the Northeast Pacific, show positive MPI trend, indicating a tendency
326 for increased TC development. In the Northeast Pacific, an area of negative trend in the
327 MPI is found, consistent with suppressed TC activities in this basin as noted in discussions in
328 previous sections.

329 Table 1 shows a summary of the percentage change of TC rainfall and the 5 major
330 large-scale environmental factors *i.e.*, SST, wind shear, sea level pressure, mid-tropospheric

331 relative humidity, and MPI averaged over different ocean domains (as marked by dotted
332 rectangles in Fig. 8). The percentage changes per decade are computed from the slope of
333 linear regression for the period 1980-2007. The North Atlantic shows the strongest
334 alignment of factors favoring increased TC activities. Three out of the 5 factors show strong
335 (>99% c.l.) in the linear trends, i.e., increased SST, reduced sea level pressure, and increased
336 MPI. All these three factors, together with a significant reduction in vertical wind shear (>90%
337 c.l.) may account for the significant increase (>95.8%) in TC rain in the North Atlantic. For
338 other ocean basins, the trend signals are generally weaker and alignments are not as clear.
339 To provide a quantitative assessment of the degree of alignment of factors among ocean
340 basins, we use an *ad hoc* ranking scheme, by numerically assigning a number from 1 to 5 for
341 each ocean basin and for each factor. In this scheme, #1 representing the largest percentage
342 change per decade favoring TC formation, and #5 the lowest, with the proper sign of the
343 trends taking into account. For example, the North Pacific ranks #1 in both sea surface
344 temperature and sea level pressure, having the largest percentage changes in SST (+1.9%)
345 and SLP (-0.09%) respectively, and also #1 in wind shear with the strongest reduction in
346 wind shear (-4.2%) among all oceans. The cumulative score for each ocean basin is then the
347 sum of the ranks for all five factors. Accordingly, the range of cumulative score is 5 to 25,
348 with the lower score showing the more alignment of factors favoring enhanced TC activities.
349 Based on the cumulative score (Table 1), the North Atlantic and the Northeast Pacific ranks
350 as the most favorable and least favorable for enhanced TC activities respectively. The two
351 most important factors in the Northeast Pacific that are in sharp contrast to other ocean basins
352 are the large positive wind shear (+4% per decade), and the least value of MPI. These two
353 factors could be key to the significant reduction in TC rainfall (-36.7%) over the Northeast

354 Pacific. This is consistent with Wang and Lee [2009, 2010], which showed increased TC
355 activities in the North Atlantic coupled to reduced TC activities in the Northeast Pacific,
356 possibly through changes in the Walker Circulation [Zhou et al., 2011]. In the Northwest
357 Pacific, contributing factors favoring increased TC-rain include increased MPI and SST, and
358 reduced SLP and wind shear. However, a reduced mid-tropospheric relative humidity (-0.3%)
359 may oppose TC development there. In the Southern Oceans, a number of factors, *i.e.*, SST,
360 wind shear and MPI, would favor TC activities and TC-rain, but none of these factors ranked
361 the strongest among ocean basins. In the Northern Indian Ocean, moderate increase in SST
362 and MPI, and moderate reduction of wind shear and SLP tend to favor TC activities, but the
363 small reduction of RH may oppose the tendency. Overall, for the Northwest Pacific,
364 Northern Indian Ocean and Southern Oceans, the mixed signals in ambient environmental
365 trends and possible compensations among competing factors (for or against TC development)
366 are consistent with the moderate increase, *i.e.*, +21.3%, +53.8%, + 22.2%, respectively
367 found in these ocean basins. In addition, the weak signal in TC-rain in Northwest Pacific has
368 low confidence level due to the previously documented large interannual and inter-decadal
369 variability [Chan 2006, 2009; Lau et al. 2008] (Table 1).

370

371 **6. Summary and discussion**

372

373 Using pentad GPCP data, we have estimated rainfall associated with tropical cyclones
374 for all storms from 1980 to 2007 from major ocean basins and examined the long-term trends
375 of tropical cyclone rainfall by storm categories in each basin and in the global mean. From
376 linear regression, we find that globally, tropical cyclone associated rainfall has increased
377 significantly for TS and for Category 4 and 5 by 6.2%, 24.8% and 43.2% per decade, while

378 decreased by 0.8 %, 9.2% and 5.5% per decade for Category 1, 2 and 3 storms, respectively.
379 The trend in TC rainfall in each category is largely determined by the trend in TC numbers,
380 but the percentage increase in rainfall is larger than the percentage increase in TC numbers.
381 Accumulated total TC rain for all tropical cyclones has a positive trend (8.4% per decade)
382 over the entire tropics.

383 Large differences in the TC-rain statistics exist among individual basins. The North
384 Atlantic observes increases in both TC number and TC rainfall in weak (TS), intermediate
385 (Cat 1-3), and intense (Cat 4-5) storms, while the Northeast Pacific shows decrease in storm
386 number and storm rain amount for all TC categories. This result is consistent with previous
387 studies [Wang and Lee 2009, 2010] which found an apparent inverse relationship between
388 TC activities between the North Atlantic and Northeast Pacific. The Northwest Pacific
389 observes increases in weak and intense storms while decreases in intermediate storms. Both
390 the Indian Ocean and Southern Oceans show significant increase in the number of intense TC
391 and the associated TC-rain. All basins except the Northeast Pacific show significantly
392 increased TC-rain in Cat 4-5 storms. There is a significant (90% c.l.) increasing trend for
393 energy-per-storm in Cat 1-3 storms in the North Atlantic, and Southern Oceans, and in Cat 4-
394 5 storms in the Northwest Pacific. In contrast, a significant decreasing trend is found in Cat
395 1-3 storms in the Northeast Pacific. Analyses of TC-energy PDFs find a shift towards greater
396 number of energetic storms at the expense of less energetic TCs is found in the North
397 Atlantic and the Southern Ocean. Over the Northwest and Northeast Pacific, we find a
398 flattening of the PDF, signifying a higher probability energetically for more strong, and weak
399 storms, but less moderate storms. In contrast, Indian Ocean shows a sharpening of the TC
400 rainfall PDF towards moderately energetic storms.

401 Large-scale environmental variables such as SST, vertical wind shear, sea level
402 pressure, mid-tropospheric humidity, and MPI are analyzed to shed light on the trend
403 differences, and the shift in rainfall PDF among ocean basins. Notably, the North Atlantic
404 observes largest increase in SST and reduction in wind shear, which may explain the largest
405 increase in TC-rain. On the other hand, the Northeast Pacific is unique as the only major
406 ocean basin showing increased vertical wind shear in addition to having the smallest increase
407 in SST, and MPI, which are less favorable for TC activities compared to other ocean basins,
408 consistent with the observed significant negative trend in TC-rain there. For other ocean
409 basins, the trend signals in environment signals are mixed, but overall favoring an increase in
410 TC-rain. Our results suggest that long-term TC-rain changes may be associated with a re-
411 distribution in the latent energy content among the population of storms within and among
412 major ocean basins, associated with the alignment of changes in key large-scale
413 environmental factors [Wu and Wang; 2008]. Further studies are needed to examine the
414 mechanisms of the alignment processes.

415 The present results are consistent with recent numerical modeling experiments
416 indicating increase in number of simulated high-intensity TCs and increasing TC rainfall in a
417 warming environment [Knutson and Tuleya 2004; Bengtsson et al., 2007; Knutson et al,
418 2010]. However, large interannual variability in TC activities and short duration of satellite
419 data make it difficult to infer long-term relationship between tropical cyclone rainfall and
420 global warming. Therefore we cannot extrapolate our results to the pre-satellite period, e.g.,
421 active TC activity in North Atlantic during late 1920s and late 1960s [Goldenberg et al. 2001;
422 Zhang and Delworth 2009]. For Northwest Pacific, large decadal variation seems to
423 dominate the variation in the last three decades. Therefore, even using the state-of-the-art

424 multi-decadal rainfall data set, the record length is still too short for a definitive assessment
425 of the long-term trend in TC-rain [Lau et al., 2008]. Furthermore, the results in this work are
426 subject to some uncertainties in the satellite rainfall data as discussed in Section 2. These
427 considerations underscore the need for extending and improving long-term rainfall data sets
428 to obtain reliable estimates of tropical cyclone rainfall such as planned under the Global
429 Precipitation Measurement (GPM) program [Hou et al., 2008].

430

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433 is obtained from the NOAA/OAR/ESRL PSD, Boulder, Colorado, USA, from their Web site
434 at <http://www.esrl.noaa.gov/psd/>

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617 Table 1. Summary of the percentage change of TC-rain and 5 major large-scale
618 environmental factors averaged over different ocean domains (as marked by dotted
619 rectangles in Fig. 8). The percentage changes are computed from the slope (% per
620 decade) of linear regression multiplied by the number of decades (=2.8) for the period
621 1980-2007. The TC-rains are computed as total for the TC seasons for JASON, and
622 ONDJF for the northern and southern hemisphere respectively. Trends with confidence
623 levels greater than 90%, 95% and 99% are indicated with superscripts ⁺, * and bold fonts,
624 respectively.
625

Change%	North Atlantic	Northeast Pacific	Northwest Pacific	N. Indian Ocean	Southern Oceans
SST	1.9	0.8 ⁺	1.4	1.0*	1.0
WS	-4.2 ⁺	4.0 ⁺	-3.4	-4.0*	-2.1
SLP	-0.09	-0.01	-0.03 ⁺	-0.04 ⁺	-0.01
RH	2.2	8.3*	-0.3	-0.04	2.5 ⁺
MPI	11.3	4.6*	12.1	11.0	9.6
Score	8	20	14	15	17
TC-rain	95.8	-36.7 ⁺	21.3	53.8 ⁺	22.2 ⁺

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628 **Figure captions**

629

630 Figure 1. TC-rain in energy unit from TRMM versus from GPCP for different TC

631 categories during the overlapping period 1998-2007.

632 Figure 2. Mean (green bar) and standard deviation (black line) of TC-rain per storm in

633 energy unit for different TC categories from (a) TRMM and (b) GPCP.

634 Figure 3. Time series of global total TC-rain in energy units (solid line) and TC number

635 (dotted line) for different TC categories.

636 Figure 4. Annual mean number of TCs from three TC groups and all TCs for the first 14

637 years (1980-1993, left bars) and latter 14 years (1994-2007, right bars) in each

638 ocean basin and global total. Orange and blue bars indicate significant (above 90%

639 confidence level) positive and negative changes. Numerical values are shown for

640 significant percentage changes.

641 Figure 5. Same as Figure 4, but for basin total TC-rain in energy unit (EY) from different

642 TC categories.

643 Figure 6. Same as Figure 4, but for the mean TC-energy per storm for different TC

644 categories.

645 Figure 7. Frequency distributions of TC as a function of accumulated TC-rain in energy

646 unit (EY) over its life time. The blue lines show the climatological mean

647 distributions over the entire period (1980-2007). The color filled lines show the

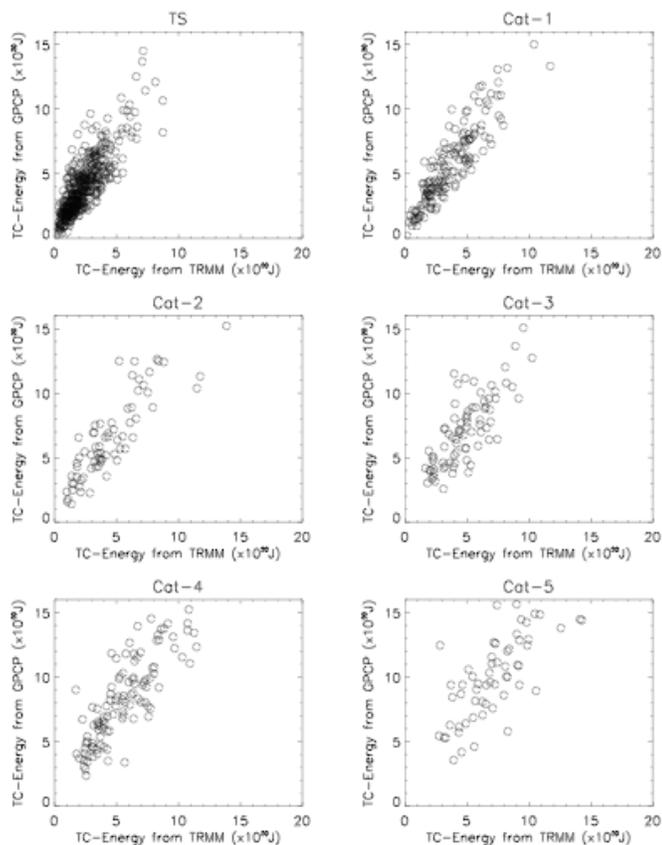
648 difference between the latter 14 years (1994-2007) and earlier 14 years (1980-1993),

649 positive in red and negative in green. .

650 Figure 8. **Left:** Linear trends of (a) SST, (b) vertical wind shear, and (c) MPI for wind in

651 July-November that affect TC activities in the northern oceans from 1980 to 2007.

652 **Right:** the same as the left panels except for trends in December-April that affects
653 TC activities in the southern oceans. The dotted rectangles indicate domains where
654 the mean values of each environmental variables in Table 1 are computed.
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658 Figure 1. TC-rain in energy unit (10^{20} Joules) from TRMM versus from GPCP for
 659 different TC categories during the overlapping period 1998-2007.

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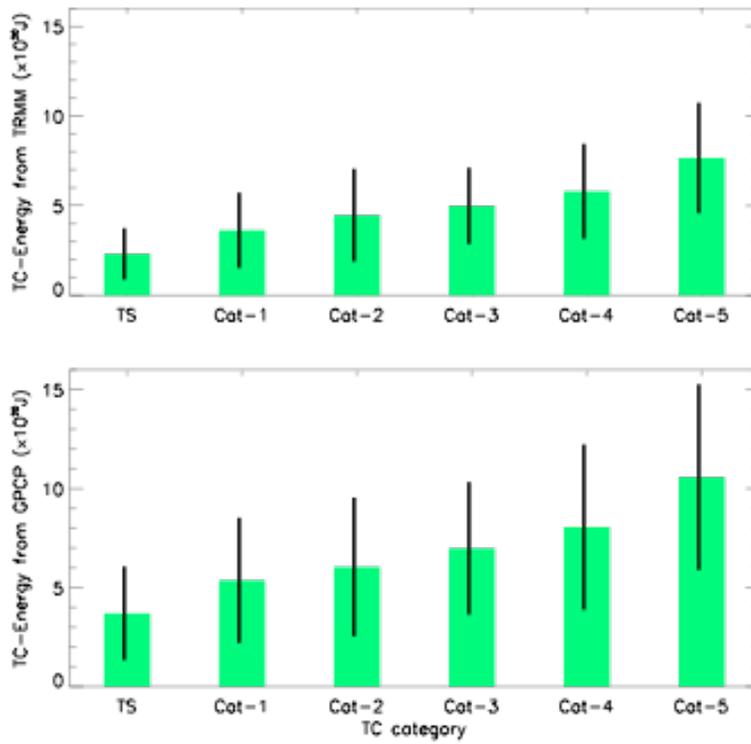
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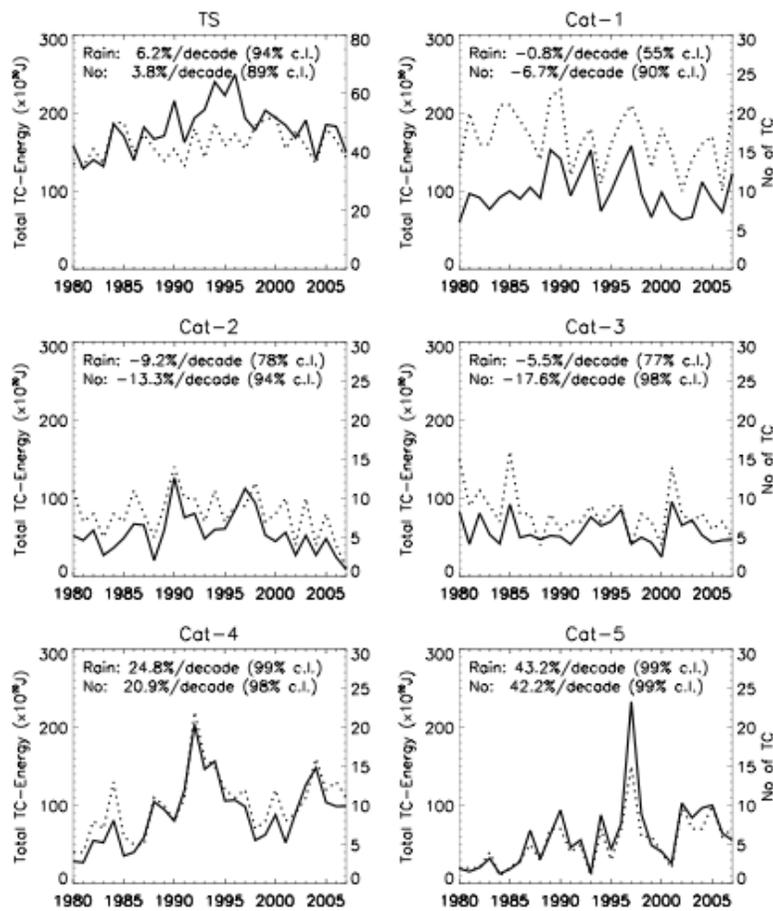
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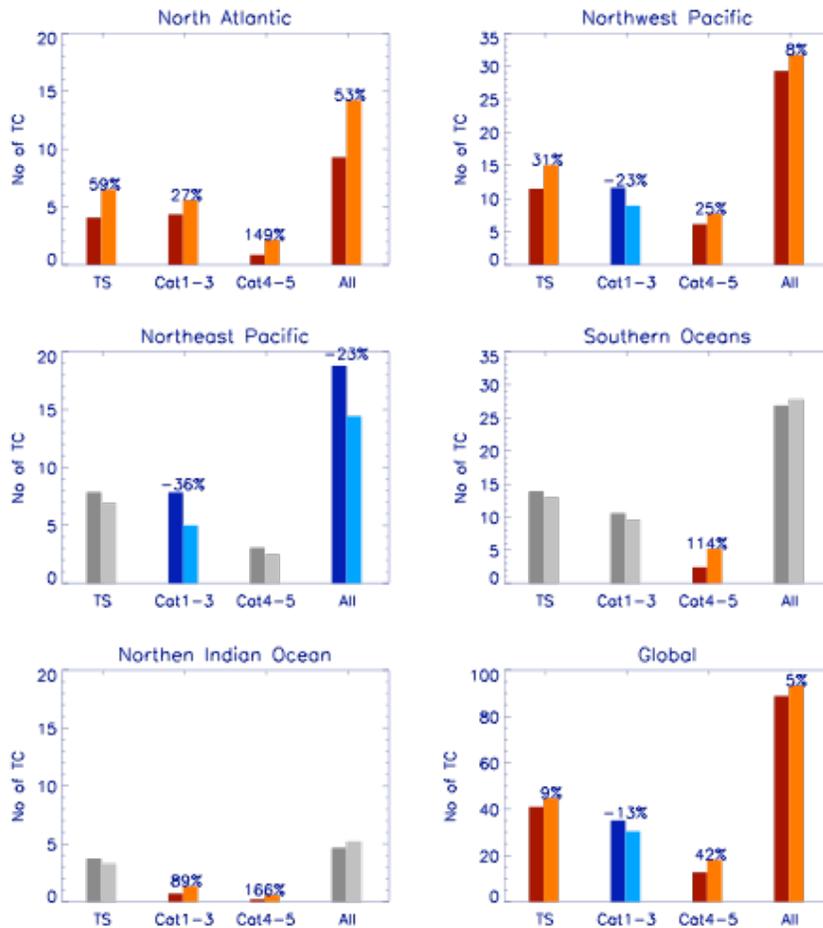


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Figure 2. Mean (green bar) and standard deviation (black line) of TC-rain per storm in energy unit (10^{+20} Joules) for different TC categories from (a) TRMM and (b) GPCP.

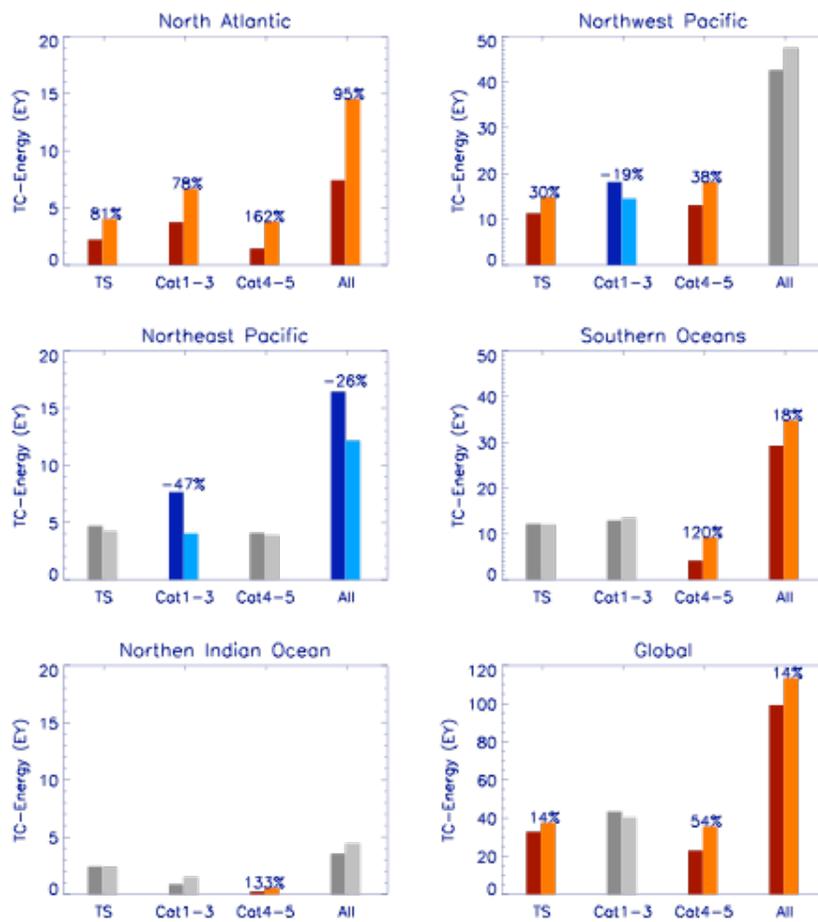


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 688 Figure 3. Time series of global total TC-rain in energy unit (solid line) and TC number
 689 (dotted line) for different TC categories.
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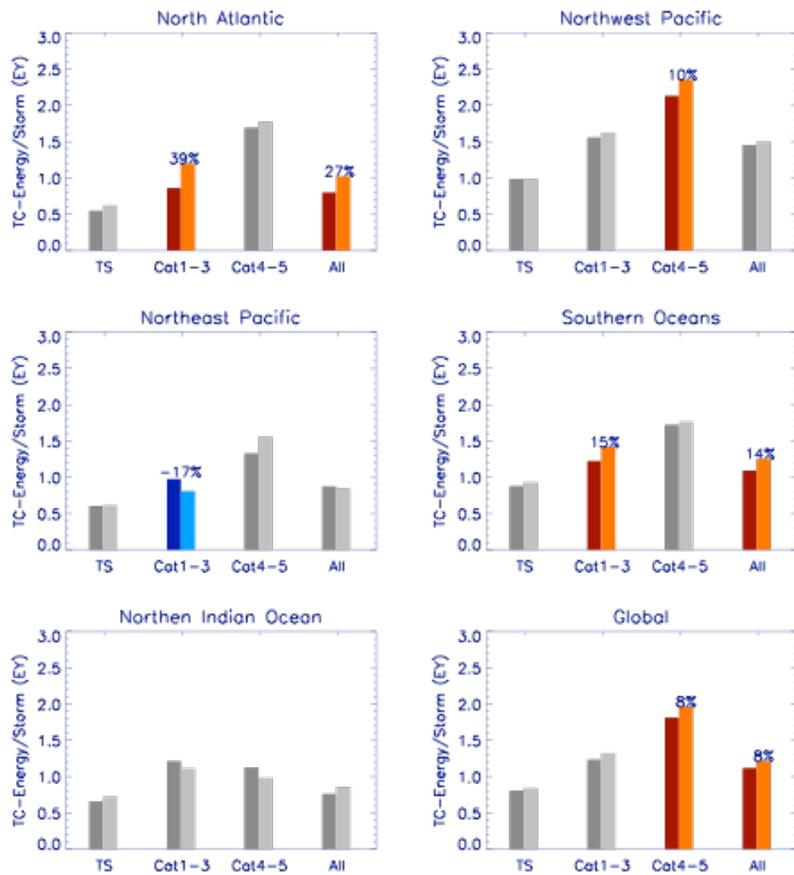
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Figure 4. Annual mean TC numbers from three TC groups and all TCs for the first 14 years (1980-1993, left bars) and latter 14 years (1994-2007, right bars) in each ocean basin and global total. Orange and blue bars indicate significant (above 90% confidence level) positive and negative changes. Numerical values are shown for significant percentage changes.



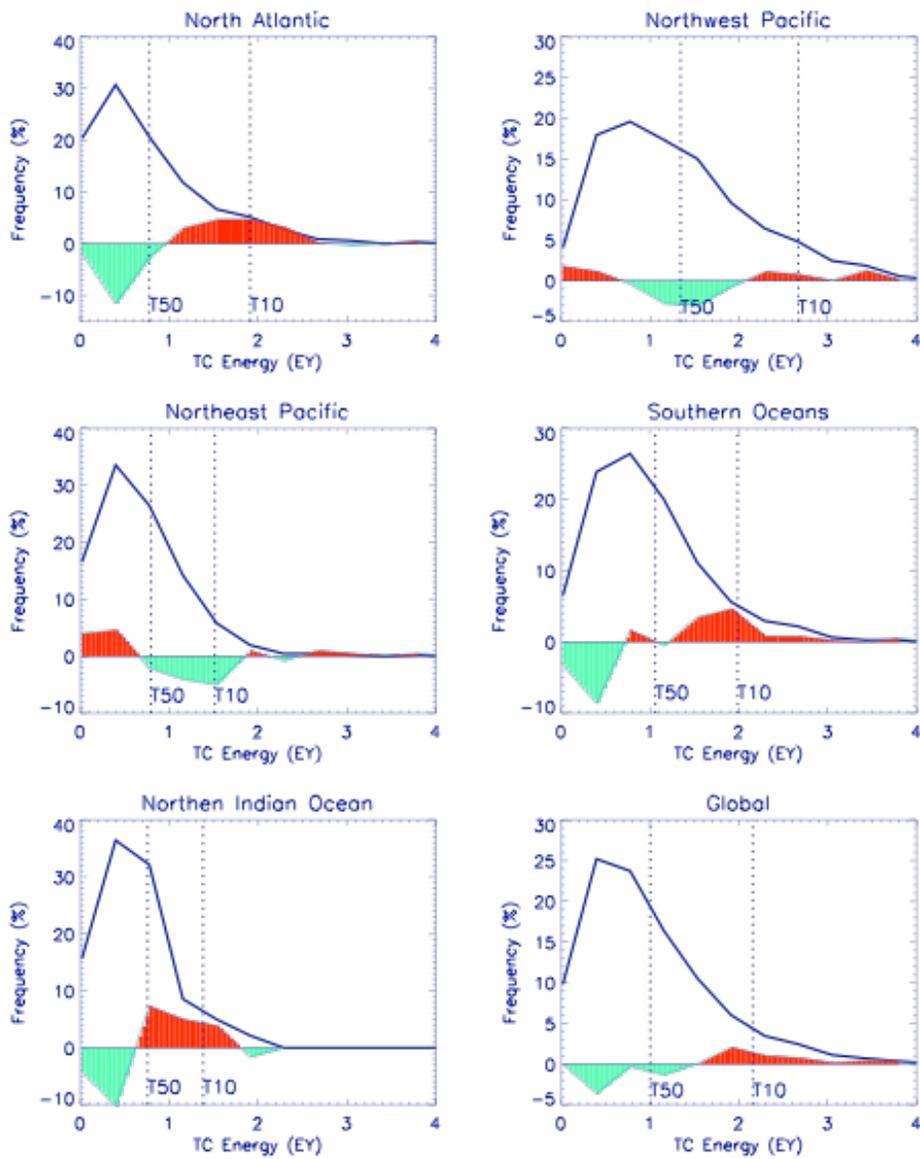
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Figure 5. Same as Figure 4, but for basin total TC-rain in energy unit (EY) from different TC categories.



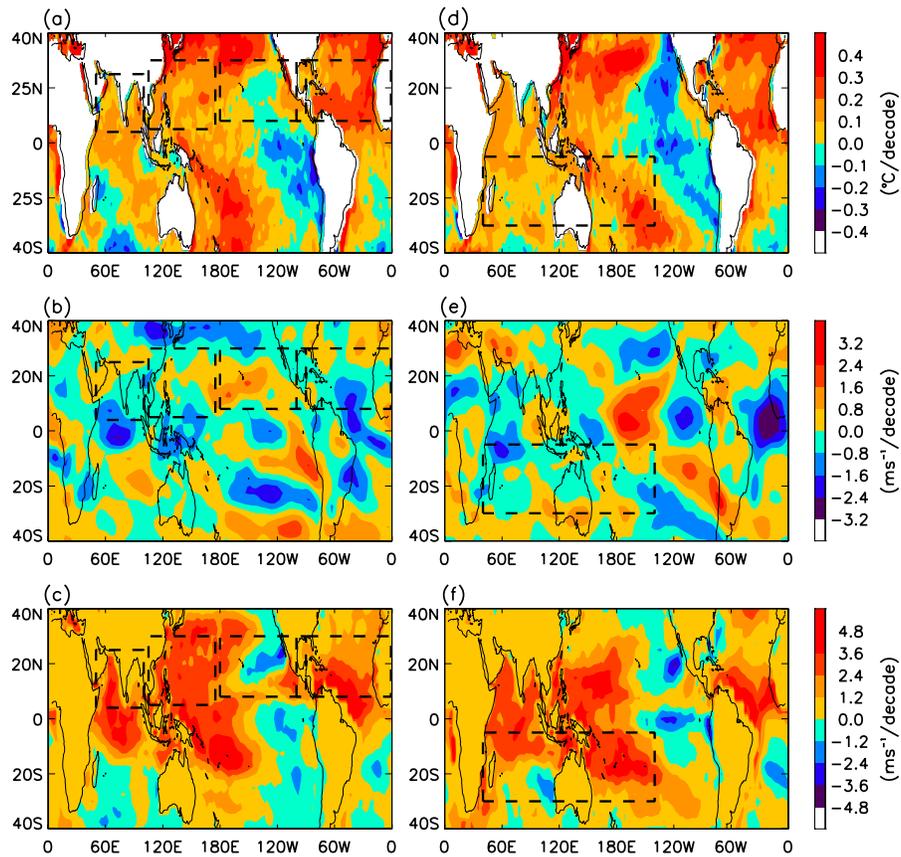
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Figure 6. Same as Figure 4, but for the mean TC-energy per storm for different TC categories.



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Figure 7. Frequency distributions of TC as a function of accumulated TC-rain in energy unit (EY) over its life time. The blue lines show the climatological mean distributions over the entire period (1980-2007). The color filled lines show the difference between the latter 14 years (1994-2007) and earlier 14 years (1980-1993), positive in red and negative in green. .



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Figure 8. **Left:** Linear trends of (a) SST, (b) vertical wind shear, and (c) MPI for wind in July-November that affect TC activities in the northern oceans from 1980 to 2007. **Right:** the same as the left panels except for trends in December-April that affects TC activities in the southern oceans. The dotted rectangles indicate domains where the mean values of each environmental variables in Table 1 are computed.