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Effect of Tropical Cyclone Intensity on the Relationship between Hydrometeor Distribution and Rapid Intensification by GPM GMI

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Key Points:

- Hydrometeor contents increase in the whole area of a weak TC but mainly in the inner-core of a strong TC during rapid intensification (RI).
- Hydrometeor contents are higher in RI than in slow intensification, and the maxima locate related to TC intensity and intensification rate.
- The cloud water content in the inner-core area have the largest correlation with the TC intensification rate, especially in a weak TC.

20 **Abstract**

21 This study analyzes hydrometeor evolution during rapid intensification (RI) and tropical cyclone
22 (TC) intensity dependence using satellite data. Previous studies have suggested that ice cloud
23 water or non-convective precipitation can serve as predictors of RI from different perspectives.
24 However, few studies have focused on the impact of TC intensity or comprehensive comparisons
25 to identify better indicators. During RI, the hydrometeor contents in weak TCs (WTCs) increase
26 over the entire region, whereas they increase mainly in the inner-core region and decrease in
27 advance in the outer-core region for strong TCs (STCs). Hydrometeor contents are higher in RI
28 than in slow intensification, and their maximum locations are related to TC intensity and
29 intensification rate. Cloud water content (CWC) in the inner-core region has the largest
30 correlation with the intensification rate, especially in WTCs. Therefore, the CWC can serve as a
31 predictor of RI and can be applied to all TC intensities.

32 **Plain Language Summary**

33 Stronger tropical cyclones (TCs), which often undergo rapid intensification (RI), cause
34 significant damage. However, accurately predicting its intensity is difficult. A close relationship
35 exists between TC intensity changes and cloud content. Previous studies have shown a
36 connection between the ice water content, precipitation, and RI of TCs. However, few studies
37 have comprehensively compared the hydrometeors and precipitation associated with RI and the
38 impact of TC intensity. During the RI, various cloud contents increase in the entire area of a
39 weak TC, while they increase mainly in the inner-core area and decrease in advance in the outer-
40 core area of a strong TC. The cloud contents are higher in RI than in slow intensification, and
41 their maximum locations are related to the TC intensity and intensification rate. Among the
42 cloud content and precipitation types, the water cloud content in the TC inner-core area is the

43 best indicator of the intensification rate, applicable to all TC intensities, especially for weak TCs.
44 Understanding the relationship between clouds and TC intensification facilitates prediction of
45 TC intensity and reduces the impact of disasters.

46 **1 Introduction**

47 Tropical cyclones (TCs) are one of the most destructive weather systems worldwide. Significant
48 progress has been achieved in predicting TC intensity; however, predicting the rapid
49 intensification (RI) of TC remains challenging (Kaplan et al., 2010; Courtney et al., 2019;
50 DeMaria et al., 2021). This is primarily due to the limited understanding of the physical
51 processes responsible for TC intensity changes, especially for RI (Elsberry et al., 2007; Kaplan et
52 al., 2010; Emanuel, 2018). Over the past few decades, the frequency and magnitude of RI have
53 gradually increased, and the location of RI has notably become considerably closer to the coast
54 (Zhao et al., 2018; Song et al., 2020; Klotzbach et al., 2022; Liu & Chan, 2022; Li et al., 2023).
55 Therefore, understanding the RI process from various perspectives is imperative to improve the
56 accuracy of RI forecasting and reduce disaster losses.

57 Previous theoretical studies have demonstrated that the evolution of TC intensity is closely
58 related to latent heat release and that the maximum intensifying efficiency occurs when latent
59 heating is within the radius of the maximum wind (RMW) (Shapiro & Willoughby, 1982;
60 Schubert & Hack, 1982; Vigh & Schubert, 2009; Rogers et al., 2013). Simulation analysis has
61 also shown a strong correlation between the total heating and total condensate water within the
62 TC (Nolan et al., 2019). Recent studies have suggested that convective bursts (CBs) can serve as
63 precursors to the RI, especially the frequency of CBs in the core region, which increases before
64 the RI (Rogers et al., 2013; Wang & Wang, 2014; Miller et al., 2015). Tang et al. (2019) found
65 that deep convection and the associated latent heat increased substantially, shifting inward before

66 the onset of RI. Further studies have revealed that ice-phase microphysical processes in the upper
67 troposphere play an important role in RI (Zhao et al., 2020). Though most of simulation studies
68 have focused on the core area where the cloud microphysical processes are sensitive to TCs
69 structures and intensities (Lord et al., 1984; Sawada & Iwasaki, 2007; Zhu & Zhang, 2006;
70 Pattnaik & Krishnamurti, 2007), diabatic heating resulting from cloud microphysical processes
71 in outer spiral rainbands can also affect TC intensity (Wang, 2009).

72 Although it is challenging to directly measure the amount of latent heat that can be converted
73 into potential and kinetic energies within a TC using observational data, hydrometeors are a
74 feasible proxy (Adler & Rodgers, 1977; Rodgers et al., 1994; Rodgers & Pierce, 1995).
75 Microwave radiation observations have been used for decades to identify TC intensification
76 signals. Ice-scattering signals in microwave bands are highly correlated with future TC intensity
77 (Cecil & Zipser, 1999; Kieper & Jiang, 2012; Harnos & Nesbitt, 2016). Harnos and Nesbitt
78 (2016) found that the 85-GHz channels bright temperatures of intensifying TCs are lower than
79 that of weakening TCs, implying a stronger signal of ice-phase particle scattering. More
80 precipitation and colder clouds have been observed with TC intensification (Ruan and Wu,
81 2018). Therefore, the ice-phase condensate content can serve as an indicator of TC
82 intensification (Wu & Soden, 2017; Su et al., 2020; Zhang et al., 2019). Furthermore, Wu et al.
83 (2020) revealed that the ice water content in the upper troposphere is highly correlated with the
84 TC intensification rate, and Liu et al. (2022) suggested that the radial distribution of the deep
85 convective cloud percentage and temperature could be used to identify the impending RI.

86 However, some studies have considered that deep convective activities are more likely to be a
87 response rather than a precursor to RI, as they increase significantly after the onset of RI (Jiang
88 & Ramirez, 2013; Zagrodnik & Jiang, 2014; Tao & Jiang, 2015; Tao et al., 2017; Fisher et al.,

89 2018). Statistical analysis of the precipitation characteristics of intensifying TCs by Jiang and
90 Ramirez (2013) indicated that the maximum convective precipitation intensity in the inner-core
91 region is not necessarily more intense for an RI episode than for a non-RI episode. Tao and Jiang
92 (2015) showed that increased and widespread shallow precipitation has occurred in the TC core
93 area before RI and which contributes to higher total volumetric rainfall and latent heat than non-
94 RI, suggesting that RI is probably triggered by the increase of shallow-moderate precipitation.
95 Therefore, stratiform precipitation may be a dominant factor in the RI (Zagrodnik & Jiang, 2014;
96 Fisher et al., 2018; Tao et al., 2017). Due to the different study perspectives it is hard to know if
97 IWC or non-convective precipitation is the better predictor, though they varies coinstantaneous
98 during TC intensification.

99 Observational and modeling studies have indicated a relationship between TC intensification rate
100 and intensity (Shapiro and Willoughby, 1982; Xu and Wang, 2015). Stronger TCs demonstrate
101 higher inertial stability within the RMW region, resulting in increasing intensification efficiency
102 and tangential wind tendencies in response to the latent heat in the eyewall (Schubert and Hack.,
103 1982). Nolan et al. (2007) suggested that the intensification efficiency is positively related to the
104 intensity of weak storms. Wu et al. (2020) found that TC experiencing an RI tends to have a
105 higher initial intensity than more slowly intensifying TC. The relationship between TC intensity
106 and the intensification rate is complex and nonlinear (Xu & Wang, 2015; Wang et al., 2021).
107 Therefore, questions arise regarding the main differences in cloud microphysical characteristics
108 between RI and slow-intensification (SI) processes in the context of different TC intensities.
109 Furthermore, which is the key indicator of the RI process, hydrometeor or precipitation, and are
110 they influenced by TC intensity? The objective of this study is to answer these questions using
111 Global Precipitation Measurement (GPM) satellite observations, which have been widely used in

112 cloud and precipitation studies. The remainder of this paper is organized as follows: The data and
113 methods are described in Section 2, the main results are presented in Section 3, and the
114 conclusions are discussed in Section 4.

115 **2 Data and Methods**

116 In this study, we utilize GPM observations of the rain water content (RWC), ice water content
117 (IWC, solid), cloud water content (CWC, liquid), and the column mass-integrated contents
118 including rainwater path (RWP), ice water path (IWP), and cloud water path (CWP) (Hou et al.,
119 2014, Randel et al., 2020). Surface and convective precipitation rates are also used. These data
120 are retrieved from the passive microwave data product (2AGRPOFGMI) on the GPM with a
121 resolution of 13×13 km and an interval of approximately 90 min. Pixel-level surface
122 precipitation rate and vertical hydrometeor content distribution files are generated using the
123 Goddard Profiling Algorithm (GPROF, Randel et al., 2020).

124 The International Best Track Archive for Climate Stewardship (IBTrACS, Knapp et al., 2010)
125 provides the TC location and maximum wind speed at 6-h intervals from March 2014 to
126 December 2022 over the western North Pacific (WNP). Following the definition given by
127 Kaplan and Demaria (2003), an RI case is defined as an increase of TC surface maximum wind
128 speed of at least 30 kts within 24 h ($\Delta V_{RI} \geq 30\text{kt}$, $\Delta V = V_{24h} - V_{0h}$). For comparison, we also select
129 slow intensification (SI) cases with an intensification rate of less than 30 kts per 24-h (30 kts
130 $> \Delta V_{SI} > 0$ kt). The radius of maximum wind (RMW) data is provided by IBTrACS.

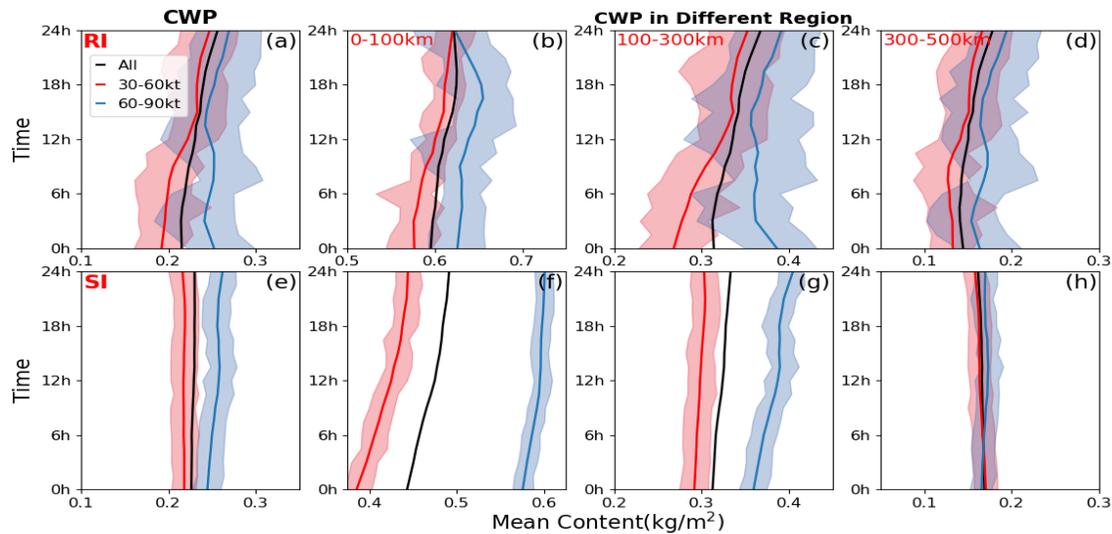
131 Owing to the time inconsistency between GPM observations and TC data, the former has a
132 higher temporal resolution at a 90-min interval while the latter is at a 6-h interval. The TC data
133 are first linearly interpolated to match the observation time of GPM satellite. Then, swath data
134 within 500 km of the TC center are directly extracted. Owing to the satellite orbit and scanning

135 method, the number of TCs that can be scanned by the GPM is far less than the total number of
136 TCs. However, their frequency distributions based on TC intensity for RI and SI are almost the
137 same, indicating that the episodes scanned by GPM are representative of the initial intensities of
138 the intensification episodes (Fig. S1; the numbers of cases are shown in Table S1).
139 The initial intensity of the intensification episodes is predominantly distributed in the 30–90 kts
140 (Table S1). Therefore, in this study, the RI and SI cases are categorized into two groups based on
141 the initial intensity: $TC_{30-60kt}$ (the initial intensity in the range of 30–60 kts) and $TC_{60-90kt}$ (the
142 initial intensity in the range of 60–90 kts) representing weak and strong TCs, respectively. All
143 cases represented RI or SI with an initial intensity of 30–90 kts. In addition, we define the inner-
144 core, outer-core, and outer regions of the TC with radii of 100, 100–300, and 300–500 km from
145 the TC center, respectively. The radii of the inner-core and outer-core regions are similar to those
146 defined by Weatherford and Gray (1988a). Weatherford and Gray (1988a and 1988b) found that
147 the outer-core region wind strength is weakly related to the inner-core intensity unless the eye
148 structures are considered, which are closely linked to the TC intensity.

149 **3 Results**

150 The time evolutions of the hydrometeor content within a 500 km radius of TC are analyzed for
151 both the RI and SI cases. Because the evolution features of RWP, IWP, and CWP are similar,
152 only CWP is presented in Fig. 1 as a representative example; those of RWP and IWP are shown
153 in Fig. S2 for details. For all intensifying cases (black curves in Fig. 1a and 1e, Fig. S2), RWP,
154 IWP, and CWP increase during the RI process but are essentially unchanged in SI; therefore,
155 after 24-h the hydrometeor contents of RI are larger than those of SI, although they are opposite
156 at the onset. This implies that the change in hydrometeor content is closely related to the change
157 in TC intensity. The hydrometeor contents of the $TC_{60-90kt}$ cases are higher than those of the

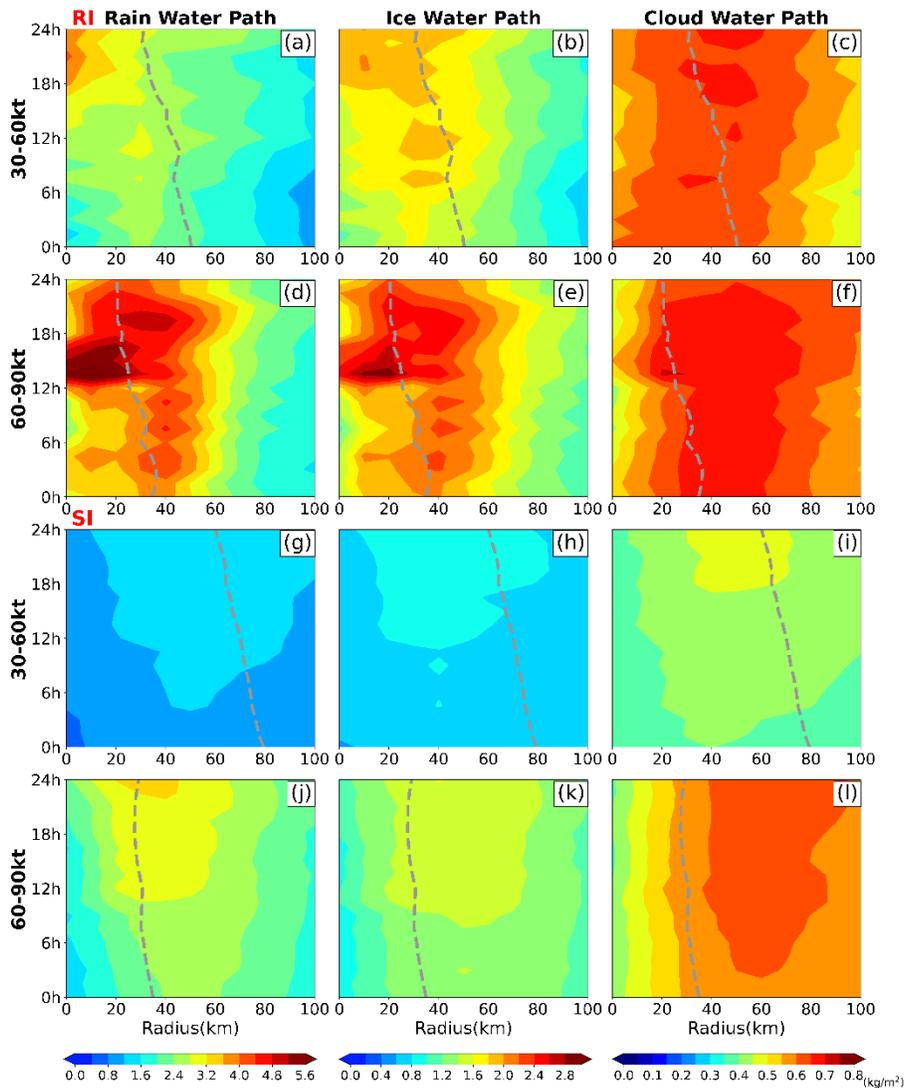
158 TC_{30-60kt} cases both in the RI and SI processes, although their increasing rates are relatively
 159 different (Fig. 1a and 1e). During RI, the TC_{30-60kt} cases have higher changes in hydrometeor
 160 contents than those of the TC_{60-90kt} cases, while the opposite is true in SI. There are two notable
 161 phenomena: the hydrometeor content changes in the 24-h period of the TC_{60-90kt} cases in RI and
 162 SI are nearly the same, and the hydrometeor content changes of TC_{30-60kt} during SI are nearly
 163 zero. This implies that the hydrometeor content changes vary significantly with TC intensity and
 164 intensification rate.



165
 166 Fig. 1 Evolution of averaged CWP (kg/m^2) within the radius of 500 km in RI (a) and SI (e) for
 167 all intensifying cases. Black, red, and blue lines denote the average contents for all, TC_{30-60kt}, and
 168 TC_{60-90kt} cases, respectively. (b-d) and (f-h) similar to (a, e) but over an area within radii of 100,
 169 100–300, and 100–500 km, respectively. The shading represents 95% confidence intervals for
 170 average CWP.

171
 172 Fig. 1b-1d and Fig. 1f-1h further show the CWP evolutions in the inner-core, outer-core and
 173 outer regions of the TC. A significant difference is observed in hydrometeor content among these

174 regions (see Fig. S3 for the consecutive radial distributions). The CWP is predominantly largest
175 in the core area of the TC, regardless of the RI or SI cases (Fig. 1b and 1f), where strong updrafts
176 appeared with dominant water vapor convergence in the lower troposphere. CWP in the TC_{60-90kt}
177 inner-core region increases slowly at first and starts to rapidly increase in 9 h after the onset of
178 RI, reaching a maximum at approximately 16 h and then weakening (Fig. 1b). The evolution of
179 CWP in the outer-core region for strong TCs is almost opposite, decreasing rapidly in the first 3
180 h, slowly decreasing, and then reaching a minimum approximately 1 h early than the time of
181 maximum in the inner-core region (Fig. 1c). This shows the contraction feature of the CWP in
182 the outer-core region of strong TCs first, followed by an increase in the CWP in the inner-core
183 region. In TC_{30-60kt}, however, the CWP evolution for RI is significantly different, increasing in
184 almost all regions for RI (Fig. 1b-1d). The different features of weak and strong TCs can be
185 demonstrated through the lead-lag correlation analysis between these two regions (Table S2).
186 TC_{30-60kt} cases exhibit a significant positive correlation between CWPs in the inner-core and
187 outer-core regions, while TC_{60-90kt} show a significant negative correlation, especially when the
188 evolution of the hydrometeor in the outer-core region led to 3 h. This might be the different
189 models of hydrometeor content increase in the inner-core region for the weak and strong TCs
190 associated with those in the outer-core region. In a recent study on the super-Typhoon Nanmadol
191 (2022), it is found that there is a small amount of large particle region approximately 300 km
192 from the eye in the post-period of RI (Wu et al., 2023). This implies that the outer-core region of
193 strong TCs is a particular area of hydrometeor evolution during the RI. The CWP in the outer
194 region is much smaller than that in the core regions, and is its change over a short time (Fig. 1d
195 and 1h).



196

197 Fig. 2 Hovmöller diagrams of azimuthally averaged RWP (a,d,g,j), IWP(b,e,h,k), and CWP
 198 (c,f,i,l) (shaded, kg/m^2) in RI (a-f) and SI (g-l) for $\text{TC}_{30-60\text{kt}}$ and $\text{TC}_{60-90\text{kt}}$. The y-axis represents
 199 the time related to RI onset, and the x-axis represents the radius from the TC center. The gray
 200 dashed lines denote the RMW.

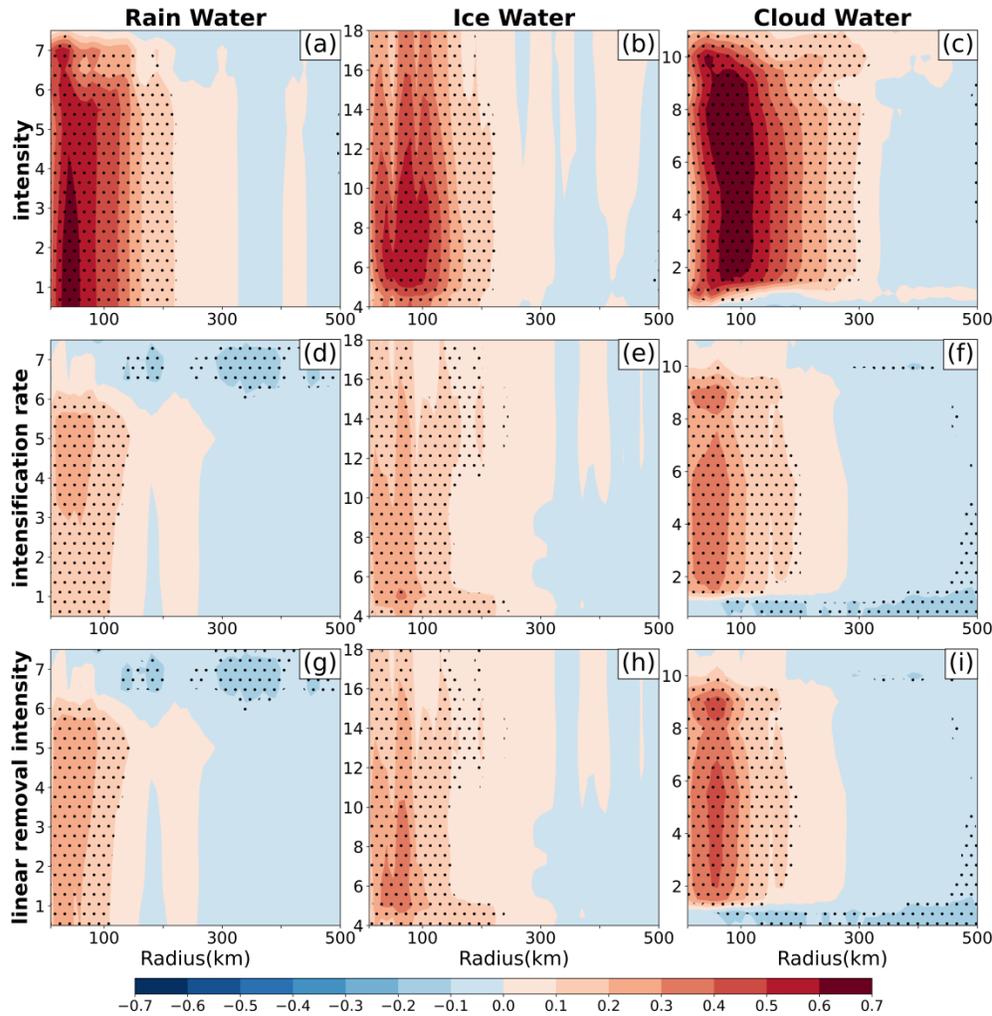
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202 The maximum increasing rates of CWP in the inner-core regions of $\text{TC}_{60-90\text{kt}}$ and $\text{TC}_{30-60\text{kt}}$ during
 203 the RI are approximately the same. However, notably, the maximum rate of increase of CWP in
 204 the inner-core region of $\text{TC}_{30-60\text{kt}}$ during SI is comparable to that of $\text{TC}_{30-60\text{kt}}$ during RI, which is

205 greater than that of $TC_{60-90kt}$ in SI (Fig. 1b-1c and Fig. 1f-1g). Owing to both the magnitude and
206 location of the latent heat impact on TC intensification, the relationship between the RMW and
207 hydrometeor distribution is examined. Fig. 2 shows that both the peak hydrometeor contents in
208 $TC_{30-60kt}$ and $TC_{60-90kt}$ of RI are distributed near or within the RMW, while those of SI are
209 distributed near or within the RMW of $TC_{30-60kt}$ and outside the RMW of $TC_{60-90kt}$. Although the
210 peak hydrometeor contents of $TC_{30-60kt}$ in RI and SI are both within the RMW, the magnitude in
211 RI is larger than that in SI (Fig. 2a-2c and Fig. 2g-2i), which is consistent with their
212 intensification rates. A theoretical study by Schubert and Hack (1982) stated that the latent heat
213 release from stronger TCs within the RMW results in higher intensification efficiencies. Fig. 2 is
214 the observation appearance of the magnitude and location of latent heat release, RMW, TC
215 intensity, and intensification rate. Owing to the offset of surface frictional dissipation, only when
216 an appropriate latent heat increase at the right place is conducive to the maintenance and
217 intensification of TC.

218 TC intensification is often related to the contraction of RMW (Shapiro & Willoughby, 1982;
219 Schubert & Hack, 1982; Willoughby & Shoreibah, 1982; Quan et al., 2023). Fig. 2a-2f show that
220 the RMW contraction is more evident in $TC_{30-60kt}$ than in $TC_{60-90kt}$ during RI. It is mainly due to
221 the stronger curvature of RMW in strong TC than in weak TC. The contraction of RMW can be
222 prohibited by increasing curvature of the RMW (Li et al., 2019). No significant difference in the
223 RMW change is identified between the SI and RI (Fig. 2g-2l). This suggests that the contraction
224 of RMW is a fundamental characteristic of TC intensification and the radius contraction rate is
225 unrelated directly to the TC intensification rate.

226



227

228 Fig. 3 Correlation coefficient radial profiles of azimuthal mean RWC, IWC, and CWC with TC
 229 intensity (a-c), intensification rate (d-f), and intensification rate after linearly removing the TC
 230 intensity effect (g-i). Dots denote the 99% confidence level of the correlation coefficient. The y-
 231 axis is the height (km), and the x-axis is the radius (km) from the TC center. Vertical areas with
 232 trifling hydrometeor content are omitted.

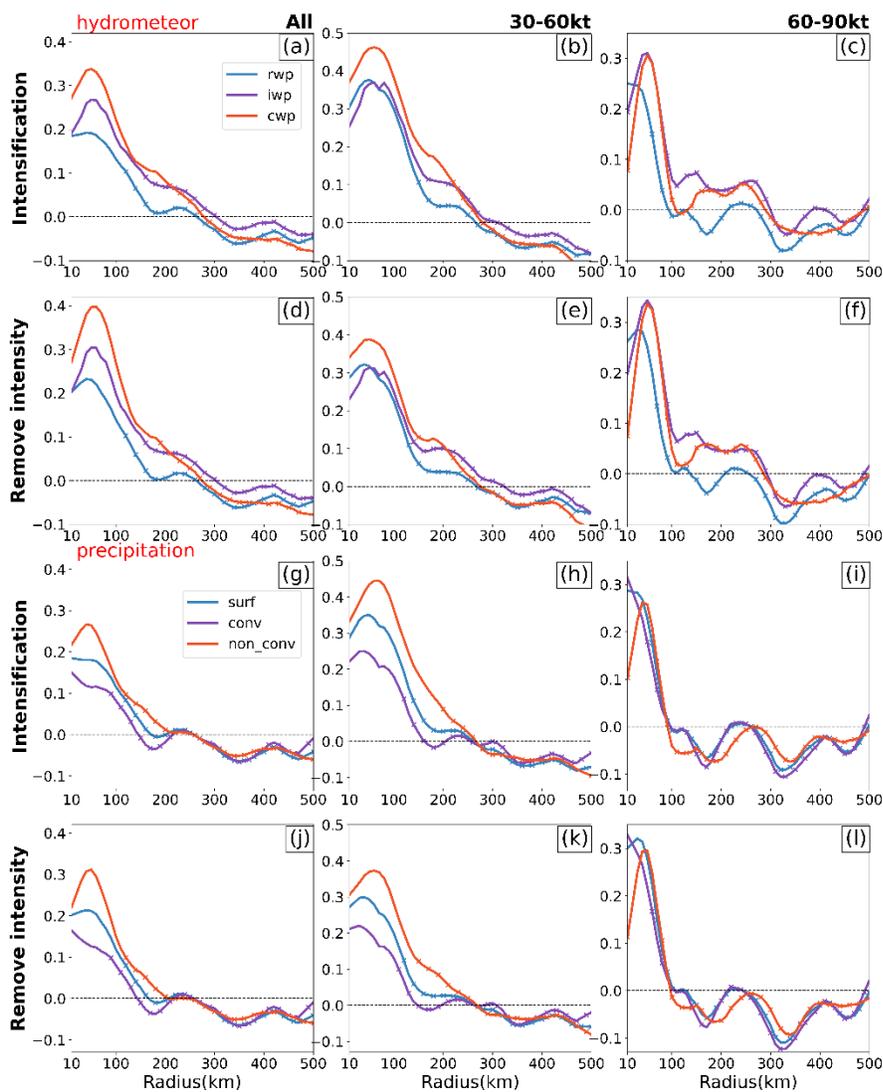
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234 Wu et al. (2020) observed that tropical cyclones with an RI usually exhibit a larger IWC at the
 235 onset of intensification, emphasizing a strong correlation between the IWC near the storm center
 236 and the intensification rate. They have suggested that the IWC can serve as a predictor of the RI

237 process. In this study, we calculated the correlation coefficients of all hydrometeor contents,
238 including RWC, IWC, and CWC, with TC intensity and intensification rate for the intensifying
239 cases (Fig. 3). The results show robust positive correlations between hydrometeors and TC
240 intensity, primarily in the inner-core areas. CWC and RWC reveal the largest linear correlation
241 coefficients in the lower and upper layers, respectively, and the former extended significantly to
242 a radius of 300 km (Fig. 3a-3c). The IWC has the smallest linear correlation, although the radius
243 of the significant correlation is approximately the same (200 km) as that of the RWC. The linear
244 correlations between the hydrometeors and intensification rate (Figs. 3d-3f) are also significant,
245 although with smaller coefficients. The radii of the maximum correlation coefficients between
246 hydrometeors and intensity (Fig. 3a-3c) are shorter than those between hydrometeors and
247 intensification rate (Figs. 3d-3f). This is the appearance of hydrometeor distribution on TC
248 intensification, which is consistent with theoretical and simulation studies from various
249 perspectives (Shapiro & Willoughby, 1982; Quan et al., 2023).

250 Theoretical and observational studies have demonstrated a close relationship between TC
251 intensification rate and initial intensity (DeMaria, 2009; Xu & Wang, 2015). TC intensity exerts
252 two effects on the intensification rate. The positive effect boosts the inertial stability in the
253 eyewall region to enhance the heating efficiency, and the negative effect counteracts the effect of
254 latent heating through frictional dissipation represented in the cube of the TC wind speed
255 (Schubert and Hack., 1982; Wang et al., 2021). To examine the impact of TC intensity on the
256 relationship between hydrometeors and the intensification rate, we remove the linear effect of TC
257 intensity and recalculated the correlation. The correlation coefficients between the hydrometeors
258 and intensification rates increase (Fig. 3g-3i). This indicates that TC intensity acts as a
259 confounding factor for the relationship between the intensification rate and hydrometeor content,

260 which should be considered in the prediction. As shown in Fig. 3i, the correlation coefficient
 261 associated with CWC exceeded 0.4, which is more significant than those for IWC and RWC.
 262 Fig. 3i has also shown that there are two maximum correlation coefficients in the middle and
 263 upper layers. This is similar to the modeling study of Li et al. (2019), in which two net heating
 264 peaks at 5 and 9 km are observed during the CB process in the RI, which are attributed to the
 265 associated peak value of the vertical velocity (Li et al., 2019; Qin et al., 2023).



267 Fig. 4 Radial distribution of the correlation coefficients between the azimuthal-averaged column
268 contents of hydrometeors and intensification rates (a-c) for all cases (a), $TC_{30-60kt}$ (b), and TC_{60-}
269 $90kt$ (c); (d-f) are similar to (a-c), but after the removal of the TC intensity effect; (g-l) is similar to
270 (a-f), but from the perspective of the precipitation rate. Lines marked with an "x" indicate not
271 passing the 95% significance test.

272

273 How is surface precipitation related to the intensification rate compared with hydrometeors? Fig.
274 4 shows the correlation coefficients along the radius of precipitation and column hydrometeor
275 content with the TC intensification rate based on the TC intensity. For all TCs, the correlation
276 associated with non-convective precipitation has a larger coefficient than that with convective
277 precipitation in the core region but smaller than that with CWP (Fig. 4a and 4g). In the $TC_{30-60kt}$
278 cases, the coefficient associated with CWP is the largest in the core region among RWP, IWP,
279 and precipitation, although the difference between those associated with CWP and non-
280 convective precipitation decreased compared to that of all cases (Fig. 4b and 4h). After linearly
281 removing the TC intensity effect, the coefficients decrease (Fig. 4e and 4k), which is opposite to
282 that for all TCs (Fig. 4d and 4j). However, the situation is different for the $TC_{60-90kt}$ cases, in
283 which the CWP, IWP, and convective precipitation have similar correlation coefficients with the
284 intensification rate of TC (Fig. 4c and 4i); after removal of the TC intensity effect, the
285 coefficients increase, and those associated with CWP and IWP are slightly larger than others
286 (Fig. 4f and 4l). The difference in performances between $TC_{30-60kt}$ and $TC_{60-90kt}$ is due to the
287 relationship between the initial intensity and intensification rate of the TCs (Fig. S4), which has
288 shown similar features in a study by Xu and Wang (2015). That is, a stronger initial intensity of
289 the TC promotes a higher intensification rate in weak TCs, whereas for strong TC cases, a

290 stronger initial intensity is associated with a lower intensification rate in the future. Keep in mind
291 that the threshold for distinguishing between weak and strong TCs is defined based on the
292 intensity at the peak of the curve representing the evolution of TC's intensification over the
293 Western North Pacific (WNP). The threshold value varies with ocean region, with approximately
294 60 kts over WNP in this study and 80 kts in the North Atlantic [refer to Fig. 1a of Xu & Wang
295 (2015)]. Fig. 4 reveals that the differences among previous studies may be partly affected by the
296 TC intensity or the proportion of different TC intensities for a group of cases. Among all
297 hydrometeors and precipitates, CWC might be a better precursor for rapid TC intensification,
298 regardless of the TC intensity.

299 **4 Conclusions**

300 This study uses GPM satellite observations to examine the impact of TC intensity on
301 hydrometeors and precipitation evolution characteristics during TC RI in the WNP. Generally,
302 higher hydrometeor contents can be expected in RI cases than in SI cases, which increase with
303 TC intensity. The hydrometeor content changes during the RI are larger than those during the SI.
304 In the RI process, the hydrometeor contents of weak TC increase throughout the entire region. In
305 contrast, those of strong TC significantly increase in the inner-core region in a short time while
306 decreasing in the outer-core region, showing a feature of hydrometeor concentration towards the
307 center. This implies that there are different increasing models for weak and strong TCs.
308 Hydrometeors are mainly distributed within the RMW for weak TCs during intensification,
309 whereas they are within the RMW during RI and outer RMW during SI for strong TCs. This
310 shows that the relationship between the intensification rate of TC and hydrometeor content and
311 location is intimately associated with TC intensity. Hydrometeors in the inner-core region are
312 significantly correlated with the TC intensification rate; the peaks of the correlation coefficients

313 are closer to the center than those correlations with TC intensity, indicating the characteristics of
314 TC intensification. Among the hydrometeors, the correlation between the CWC in the inner-core
315 region and the TC intensification rate is the most significant for all TCs. After linearly removing
316 the intensity effect, the correlation coefficients increase for all TC groups and strong TCs,
317 whereas it decreases for weak TCs, indicating that TC intensity interferes with the relationship
318 between hydrometeors and intensification rate. From the precipitation perspective, non-
319 convective precipitation is a better predictor than convective precipitation, although it is inferior
320 to CWC. A comprehensive comparison suggests that the CWC in the core region of the TC is the
321 best predictor of RI; however, the influence of TC intensity should also be considered. This
322 study provides some observational features of hydrometeors in the RI with GPM. This supports
323 previous studies on hydrometeors as predictors of TC intensification and further suggests that the
324 CWC may serve as a better predictor of RI among all hydrometeors and precipitation, as well as
325 the impact characteristics of TC intensity. Future research should explore the physical
326 mechanisms of different hydrometeor evolutions in RI processes between weak and strong TC
327 using high-resolution numerical simulations.

328

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332

333 **Data Availability Statement**

334 GPM satellite data are downloaded from the NASA Goddard Earth Sciences Data and
335 Information Services Center

336 (https://disc.gsfc.nasa.gov/datasets/GPM_2AGPROFGPMGMI_07/summary). The IBTrACS
337 data are obtained from the National Centers for Environmental Information
338 (<https://www.ncei.noaa.gov/products/international-best-track-archive>).

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