

1 **Response to Reviewers (Manuscript ID: amt-2019-223)**

2

3 First of all, we would like to thank the reviewers for their valuable comments. In the revised manuscript,
4 we have accommodated all the suggested changes into consideration and revised the manuscript
5 accordingly. The reviewers' comments are copied here as texts in BLACK. The authors' responses are
6 followed in BLUE, and our changes in the manuscript are in *italics*.

7

8

9 **Reviewer # 1**

10 This paper by Yao et al., evaluates qualities of cloud properties in three reanalysis datasets, namely,
11 China Meteorological Administration Reanalysis data (CRA), ECMWF's Fifth-generation Reanalysis
12 (ERA5), and Modern-Era Retrospective Analysis for Applications version 2 (MERRA-2). A
13 radiance-based evaluation approach is utilized with reflectance and brightness temperature observations
14 from the Advanced Himawari Imager (AHI) onboard the Himawari-8 satellite. A radiative transfer
15 model (CRTM) is used to link cloud related variables from reanalysis to satellite observations.

16 Overall, I believe this work is very valuable, which enhances our understanding of cloud representation
17 in those reanalysis products. However, I have some concerns about the structure and some details of this
18 paper.

19

20 **Several major concerns I have about this paper include:**

21. This paper uses observations from AHI/Himawari-8 to evaluate reanalysis. It is very important to
22 mention that which satellite products (in particular cloud related datasets) are used as input in the three
23 reanalysis products.

24 **Response:** Thanks for the suggestion. Yes, it is necessary to introduce satellite observations assimilated
25 for the reanalysis, because the differences on satellite datasets assimilated may be a potential reason for
26 different performances of the reanalysis. Thus, we added the related contents in Section 2. Both ERA5
27 and CRA consider Himawari-8 observations, whereas MERRA-2 does not. This may be one of the
28 reasons that MERRA-2 has relatively poor performance in the Asian region. To address the reviewer's
29 concern, we included the following discussion in the revision (Lines 400-402):

30

31 *"It should be noticed that both ERA5 and CRA reanalysis consider Himawari-8 observations for*
32 *assimilation (see Section 2), whereas MERRA-2 dose not. This may be one of the reasons that*
33 *MERRA-2 has relatively poor performance on cloud representation in the Asian region."*

34

352. The advantages of a radiance-based evaluation approach are discussed in the abstract and introduction. I
36 don't understand why the authors still use a lot of space describing AHI cloud products in Section 4?

37 **Response:** In the original submission, we try to demonstrate more clearly that direct retrieval-based
38 evaluation may be problematic, so Figures 1 and 2 as well as the corresponding discussions give
39 comparisons based on the cloud products retrieved based on different bands (i.e., the solar channels and
40 thermal infrared channels). We agree with the reviewer that the purpose of the study is to evaluate
41 different reanalysis datasets based on the radiance-based approach. Considering that the Introduction
42 Section is clear enough to demonstrate the disadvantage and uncertainties related to the retrieval-based
43 evaluation (as noticed by the reviewer), we have removed the details related to the retrieval-based
44 evaluation (i.e., Figs. 1 and 2 as well as the corresponding discussions), and the part related to AHI
45 cloud products has also been removed.

46

473. This paper uses almost 4-pages to describe a case (a snapshot on a particular day) assessment, which I
48 think is not necessary. In my point of view, the authors should pay more attention on long-term cloud
49 representation (e.g., cloud monthly mean, seasonal/annual variability).

50 **Response:** Actually, the "case study" mentioned in this study is not a snapshot for a particular day, and
51 we consider results over eight days with over 30 realizations. To avoid such misunderstanding, we have
52 the added the following sentence in the revision (Lines 211-212):

53 *"Noted that even for this case study, we consider a period over eight days covering 32 time steps."*

54

55 We think the case assessment is meaningful as well for the following reasons:

56 (1). The results in Figures 11 and 13 indicate that the evaluations are generally stable over time. The
57 results of the case study are universalistic and representative, and the corresponding conclusions are
58 actually consistent with those from the long-term evaluation. However, because the forward radiative
59 transfer simulation is computationally expensive, this study considers results from a typical case with
60 eight days and a generally evaluation with 144 realizations over one year.

61 (2). In fact, we use the case study results to present more details of the three reanalysis, whereas use the
62 long-term results for the general evaluation. As a result, we think both parts are necessary.

63 (3). Both the case study and the 144 realizations spanning over one year indicate that our methodology,
64 i.e., the radiance-based evaluation, is feasible, and the results are reliable.

65 Meanwhile, we agree with the reviewer that more attentions should also be paid to cloud monthly mean,
66 seasonal/annual variability, and we have extended these discussions. Furthermore, we would like to
67 investigate the long-term cloud representation in details in our future studies.

68

69 **Some minor suggestions include:**

70 1. Page 2, large advantages of spatial distributions → large advantages of spatial coverages.

71 **Response:** Thanks for your suggestion, and the phrase is corrected.

72

73 2. Page 6, CTT from two satellite retrieved cloud datasets (i.e., from solar and thermal infrared) How
74 to use AHI solar bands to get CTT, can you give more details on this?

75 **Response:** Sorry for the confusion because of my incorrect description. The cloud top in the product
76 from Letu et al. (2018) is retrieved based on the observations in the infrared window channel (11.2 μm),
77 and the cloud product of Iwabuchi et al. (2018) is based on observations in the 10.4 μm channel.
78 However, the atmospheric profiles used in the cloud retrieval are different, and Letu et al. (2018) and
79 Iwabuchi et al. (2018) cloud products use profiles from the GPV (the Grid Point Values of atmospheric)
80 and MERRA reanalysis, respectively. As mentioned above, we think this study should focus on the
81 radiance-based evaluation, so we have removed the section on cloud retrieval products.

82

83 3. Figures 3, 5, and 7. The plots in Figures 5 and 7 use all pixels (i.e., clear + cloudy) in Figure 3? If
84 yes, I suggest remove clear pixels or only focus on the regions of interest. I noticed that a large number
85 of pixels in Australia are clear and reflectances from models are much higher (brighter) than AHI
86 observations. This can significantly bias your plots in Figs. 5 and 7, and statistics.

87 **Response:** Yes, both clear and cloudy pixels are considered in Figs 5 and 7. Because we consider
88 different clouds by using different BTs or BTDs, even with all pixels considered, the problems related
89 to the reanalysis over cloudy regions can be illustrated by the figures. We think the reviewer gives an
90 excellent comment to consider only cloudy pixels, so we added a new Figure 5 in the revision with clear
91 and cloudy pixels considered separately. We found that the cloud property representation contributes
92 more to the differences than the atmospheric profiles.

93 Meanwhile, as there is no “truth” for the classification of clear/cloudy pixels (again, we do not want to
94 use the retrieval results due to their own uncertainties), we can only use reanalysis data for the
95 classification. This is also a reason that we mostly consider all pixels in the discussions.

96

97 4. Figures 11 and 12 and corresponding text: The authors use BT 11 μ m as a proxy to differentiate
98 clouds on low, mid, and high levels. This is problematic since high and thin cirrus may be attributed to
99 low clouds.

100 **Response:** Thanks for the suggestion. In the revision, the widely-used thresholds based on BTDs
101 between the 6.2- and 11.2- μ m channels are used to differentiate clouds over different layers (Mecikalski
102 and Bedka, 2006; Yao et al., 2018). Because of strong water vapor absorption in the 6.2- μ m channel
103 and the temperature lapse rate within the troposphere, the BTDs between 6.2- and 11.2- μ m are usually
104 negative. The BTDs increase as the cloud top height increases and larger negative BTDs often
105 corresponds to clear-sky pixels. We use the thresholds of -45 to -30 K to infer pixels with low cloud
106 tops, and those with low- to mid-layer cloud are represented by BTDs between -30 and -10 K following
107 Mecikalski and Bedka (2006). The BTDs less than -45 K normally correspond to clear pixels and those
108 larger than -10 K are from high cloud pixels. With the improved classification, most results and
109 conclusion are similar, and slight differences are noticed for mid-layer clouds (The mid-layer cloud in
110 CRA is closest to the observation.) Thanks for your suggestions, and we have updated the
111 corresponding classification, figures, and the corresponding discussion in the revision.

112
113
114 **References:**

- 115 Letu, H., Nagao, T. M., Nakajima, T. Y., Riedi, J., Ishimoto, H., et al.: Ice cloud properties from
116 Himawari-8/AHI next-generation geostationary satellite: capability of the AHI to monitor the DC
117 cloud generation process, *IEEE Trans. Geosci. Remote Sens.* 12, 1-11, 2018.
- 118 Iwabuchi, H., Putri, N. S., Saito, M., Toloro, Y., Sekiguchi, M., et al.: Cloud property retrieval from
119 multiband infrared measurements by Himawari-8, *J. Meteor. Soc. Jpn*, 96, 27-42, 2018.
- 120 Mecikalski, J. R. and Bedka, K. M.: Forecasting convective initiation by monitoring the evolution of
121 moving cumulus in daytime GOES imagery, *Mon. Wea. Rev.*, 134, 49-78, 2006.
- 122 Yao, B., Liu, C., Yin, Y., Zhang, P., Min, M., and Han, W.: Radiance-based evaluation of WRF cloud
123 properties over East Asia: Direct comparison with FY-2E observations, *J. Geophys. Res.*, 123,
124 4613-4629, 2018.

130 **Reviewer # 2**

131 General Comments: This paper is about assessment of cloud properties from the re- analysis with
132 satellite data over East Asia. Three sets of reanalysis data are used, including the newly developed
133 China Meteorological Administration Reanalysis data (CRA), the ECMWF's Fifth-generation
134 Reanalysis (ERA5), and the Modern-Era Retrospective Analysis for Applications, Version 2
135 (MERRA-2). And, to avoid the unrealistic assumptions and uncertainties on satellite retrieval
136 algorithms and products, a radiative transfer model (CRTM) is used to transform reanalysis data into
137 radiance/brightness temperature that can be directly compared with the Himawari-8 satellite data.
138 Although cloud properties from CRA, ERA5, and MERRA-2 have their own advantages, the results
139 show that ERA5 reanalysis data is best representative of cloudy atmosphere over East Asia, while the
140 results in CRA are close to those in ERA5. This study may contribute to the improvement of cloudy
141 property representation in models and satellite observations. This paper is within the scope of
142 Atmospheric Measurement Techniques but some improvement should be conducted before the paper
143 could be accepted for publication.

144

145 **Major concerns:**

146 1. The authors claim that the radiance-based evaluation approach could avoid unrealistic assumptions
147 and uncertainties on satellite retrieval algorithms and products, and thus it is a better way to carry out
148 the assessment of cloud properties from various reanalysis. However, I would say I only partially agree
149 with the authors on the perspective that the conventional way to compare cloud variables could be still
150 indispensable. Without knowing the quantitative and qualitative differences in cloud properties, it is still
151 hard to explain the radiance/brightness temperature differences resulting from the radiative transfer
152 modeling. Thus, more discussion about the cloud optical properties should be added.

153 **Response:** We agree with the reviewer that the comparisons with retrieved cloud products are still
154 necessary for assessment of model simulations. As we have discussed in the Introduction Section (as
155 well as Figs. 1 and 2 in the original submission), such direct comparison may be also problematic due to
156 the uncertainties related to retrieval product. Of course, the radiance-based evaluation has its own
157 disadvantages as well. Thus, we decided to focus only on the radiance-based evaluation, and more
158 detailed quantitative and qualitative evaluation based on direct comparison is suggested be performed in
159 further independent studies. Besides removing the retrieval-based evaluation parts, we also included the
160 following discussion in the revision (Lines 71-74):

161

162 *“The retrieval-based evaluation is still an indispensable approach in the evaluation of atmospheric*
163 *properties from various simulations, and quantitative and qualitative analysis of the cloud optical*
164 *properties, e.g., the cloud effective radius and optical depth, can be evaluated directly. However, to*
165 *avoid uncertainties associated with satellite retrieval algorithms and platforms, another alternative*
166 *radiance-based comparison is chosen for the cloud properties assessment in our study.”*

167

168 2. Previous studies (i.e., Yi et al., JGR, 2017a, b) indicate that a consistent cloud optical property
169 parameterization scheme should be used in satellite retrievals and modeling studies to well simulate the
170 radiance/flux at the top of the atmosphere under cloudy sky. Any mismatch in cloud optics
171 parameterization could induce large bias in the retrieval and simulations. Taking that into account, it
172 seems the study here using CRTM with a new set of cloud optical property look up tables (it is also not
173 clear what kind of ice cloud particle model is used) that is inconsistent with the Himawari-8 cloud
174 retrieval algorithm, could be potentially problematic in the satellite radiance/brightness temperature
175 simulation. The authors may need to consider using the Voronoi ice scattering model by Letu et al.
176 (2016; 2018).

177 **Response:** We agree with the reviewer that inconsistent cloud optical property models could be a
178 potential problem for the differences in different satellite retrievals. This is the reason that we think the
179 retrieval-based evaluation can be problematic. We have omitted the figures showing the direct
180 comparison. In our radiance-based evaluation, no satellite cloud product is used, so such differences for
181 different cloud product will not influence our results.

182 Meanwhile, we clarified that the optical properties of aggregate columns with eight elements and severe
183 surface roughness are used for CRTM. We think it is interesting to check the influence of cloud optical
184 property parameterization on our evaluation, and this is suggested as a future study as following (Lines
185 169-172):

186

187 *“It should be noted that schemes for both cloud optical properties (e.g., ice cloud model) in the RTM*
188 *and coupling between atmospheric reanalysis and RTM (e.g., approximation of cloud effective radius)*
189 *may influence simulated BTs/reflectances, although the influences are relatively minor compared to*
190 *presences of clouds (cloud amount). The potential numerical uncertainties due to different schemes will*
191 *be performed with more details in further studies.”*

192

193 3. Apart from the potential problem in cloud optical property, another important issue is about the
194 differences in the atmospheric profiles. The simulated radiance/brightness temperature is closely related

195 with the atmospheric profiles. Whereas, differences in the atmospheric profiles among the reanalysis
196 datasets are prevalent. And these differences may contribute to the simulated results under cloudy sky.
197 Thus, I think it would be best that the authors provide some analysis of the clear-sky evaluations
198 (maybe in appendix). This would be helpful for the reader to distinguish the impacts of atmospheric
199 profiles and the cloud properties.

200 **Response:** Thanks for the suggestion. It is interesting and meaningful to consider the cloudy and
201 clear-sky pixels separately and to evaluation the contributions from cloud or atmospheric profiles.

202 (1) First, for the solar channel results (Figs. 1 and 3 in the new version), the differences are almost all
203 contributed by cloud representation, because atmospheric profiles have little effect on the
204 reflectance in the 0.64- and 1.6- μm channel. We added brief discussion and analysis in the revised
205 paper.

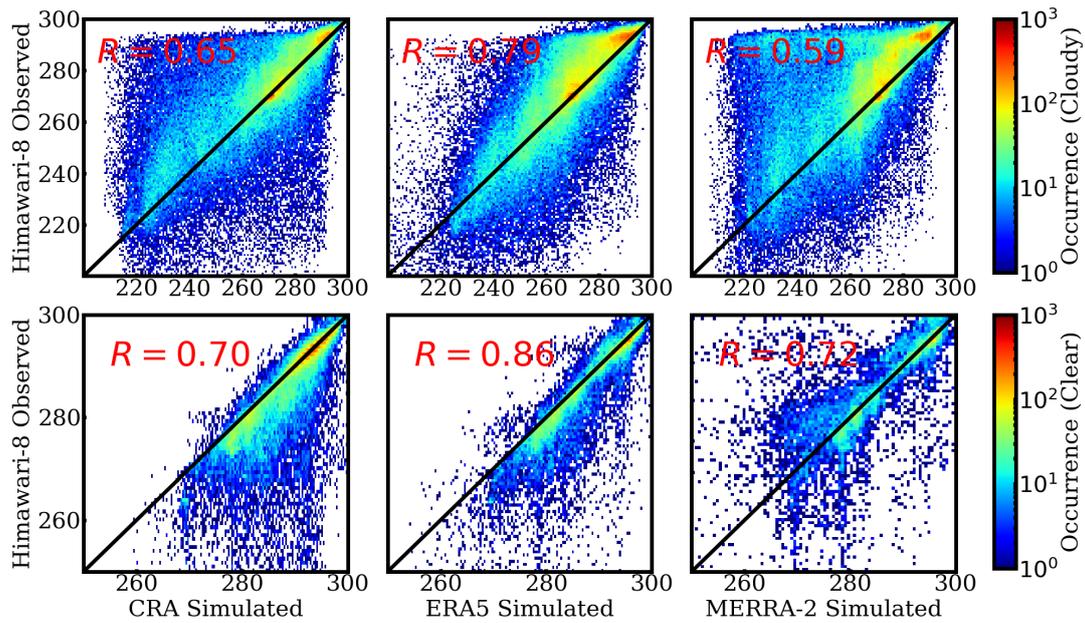
206 (2) Comparison between simulated and observed BTs in the IR channels does show the overall
207 performances of the reanalysis data due to both cloudy and atmospheric profiles. However, the
208 discussion and classification based on BTDs can significantly remove the influence of atmospheric
209 profiles, because the BTDs between the selected channels are mostly influenced by the cloud
210 properties (e.g., cloud height and cloud amount).

211 (3) Furthermore, we include the following discussions in the revision. If pixels are separated as cloudy
212 or clear ones based on a criterion of 0.1 for the integrated column cloud optical depth in each pixel,
213 the figure below shows the pixel-to-pixel comparisons between observed and simulated BTs in the
214 11.2- μm channel. The top row is for cloudy pixels, and the bottom one is for clear-sky pixels.
215 Larger correlation values for the clear pixels indicate that the cloud properties do significantly
216 contribute to the differences.

217 (4) Last but not the least, the reviewer raised an interesting and important point, which should and will
218 be done in the future, we have added the following discussion (Lines 409-412):

219
220 *“The radiance-based approach is a reliable choice for the evaluation to avoid uncertainties due to*
221 *retrieval products, and its drawbacks may be investigated in further studies. For examples, differences*
222 *between simulated and observed radiances can be contributed by both cloudy and atmospheric*
223 *variables, and these may be distinguished by considering the same atmospheric profiles in the RTM*
224 *simulations.”*

225



226

227 **Figure 1.** Pixel-to-pixel comparisons between the observed and simulated BTs in the 11.2- μm channel.
 228 Top panels indicate the comparison for cloudy pixels, and the bottom panels show the comparison for
 229 clear pixels. The results are taken at 00:00 (UTC) on 12 September 2016.

230

231 4. In part 3: methodology, to derive the necessary cloud property inputs for RTM, the authors also make
 232 quite a few assumptions. Especially in deriving the effective radius (Line 145), the used definition is
 233 somewhat different from those normally used in parameterization. As the effective radius is a very
 234 important quantity that decides the cloud optical properties in the parameterization, the authors need to
 235 analyze how the differences in the definition of effective radius will influence the results.

236 **Response:** The reviewer noticed an important point of our study. In fact, the couple between reanalysis
 237 cloud variables and RT simulations is one of the most essential parts of this study. We have tried our
 238 best to avoid empirical relationships for cloud property estimation.

239 (1) For water cloud, the effective radius scheme is based on Thompson et al. (2004) a popular scheme
 240 in mesoscale meteorological forecast models (e.g., the WRF model). The cloud number
 241 concentration over continent and ocean regions are assumed as typical and widely used values
 242 (Miles et al. 2000; Thompson et al., 2004; Wendisch and Yang, 2012).

243 (2) For ice clouds, the effective radius is physically estimated by mass extinction coefficient, which is
 244 given by an empirical relationship related to ice water content (Heymsfield and McFarquhar, 1996;
 245 Platt, 1997; Heymsfield et al. 2003), and the ice water content is from reanalysis directly.

246 (3) As also noticed by the reviewer, the coupling is far from being a done work. There could be
 247 multiple ways to estimate the effective radius. For example, in our previous study (Yao et al. 2018),

the effective radius of ice particle is calculated based on ice crystal mass and mass-radius relation (Hong et al. 2004). The following table compares observations with simulated BTs calculated based on the schemes used in this study (Scheme A) and the previous study (Scheme B, Yao et al. 2018). The correlations between observations and simulations from two different radius parameterized schemes are close to each other, and slight differences are noticed for the mean BT differences (MBTD) and BTD standard deviation (SBTD). This indicates that the schemes for effective radius estimation matter, whereas the influences are limited. Considering the length and focus of this study, we will not include such discussion in the manuscript, but we do think such sensitive study is interesting for a further study.

Table 1. The mean BT difference (MBTD), BTD standard deviation (SBTD), and correlation coefficient (R) between the observation and simulations (simulations based on two different particle effective radius estimations).

Variables	6.2- μm		11.2- μm	
	Scheme A	Scheme B	Scheme A	Scheme B
R	0.87	0.85	0.70	0.68
MBTD (K)	-0.52	-1.71	-1.71	-6.43
SBTD (K)	4.98	4.98	16.13	18.50

5. There are quite a few places in the text that are not clearly stated and are difficult to understand. For example:

Line 301: It is not clear how the probability and cumulative probability are calculated here. And how do you “obviously” figure out from Figure 7 that “total cloud is overestimated in ERA5 and MERRA-2”?

Response: Here the probability and cumulative probability indicate the occurrence of pixels with certain BTs.

The probability (P_{BT_o}) is numerically calculated as:

$$P_{BT_o} = \frac{\text{Number of pixels with BT between } BT_o - \Delta BT \text{ and } BT_o + \Delta BT}{\text{Total pixel number}}$$

, and the cumulative probability (C_{BT_o}) is given by:

$$C_{BT_o} = \frac{\text{Number of pixels with BT less than } BT_o}{\text{Total pixel number}}$$

271 The cumulative probability distribution is a good metric to give the occurrence of cloud. If we simply
272 use a BT threshold of $\sim 275\text{K}$ in the $11.2\text{-}\mu\text{m}$ channel to distinguish the cloud ($\text{BT} < \text{the threshold}$) and
273 clear-sky ($\text{BT} > \text{the threshold}$) pixels, the cumulative probability with BTs less than $\sim 275\text{ K}$ is
274 approximate 0.8 and 0.7 for MERRA-2 and ERA5, respectively, whereas the cumulative probability
275 with BTs less than $\sim 275\text{ K}$ for CRA and Himawari-8 observation is only 0.6. This suggests that over
276 the observational domain, $\sim 80\%$ of the MERRA-2 and $\sim 70\%$ of the ERA are covered by clouds, which
277 is larger than that from the observation.

278 We have rephrased the discussion and analysis in the corresponding paragraph.

279

280 Line 348: How do you define “ratio of the simulation-to-observation frequency of pixels with particular
281 BTs”?

282 **Response:** The “ratio of the simulation-to-observation frequency of pixels with particular BTs” is
283 defined by the ratio of number of pixels with particular BT interval in simulation and observation. The
284 value (RA) is numerically given by:

$$RA = \frac{\text{Number of pixels with simulated BT between } BT_a \text{ and } BT_b}{\text{Number of pixels with observed BT between } BT_a \text{ and } BT_b}$$

285 To better distinguish different clouds, the threshold of BTs of $6.2 - 11.2\text{-}\mu\text{m}$ is used in the revision,
286 and the corresponding explanation and discussion in the paragraph are rephrased.

287

288 Line 353: What does TCC mean?

289 **Response:** TCC here is the abbreviation of Total Cloud Cover, we have add the full name of it.

290

291 Line 376-377: How do you define mean error (MBTD) and standard error (SBTD) ?

292 **Response:** For each snapshot, the MBTD is the mean BTs over the entire comparing region, and the
293 SBTD is the corresponding standard deviation. The MBTD and SBTD are calculated over the whole
294 Himawari-8 observation domain between simulated and observed BTs. We have clarified this in the
295 revision.

296

297 6. Figure captions in this paper are not clear enough to show what the figures are about. For example:
298 Figure 7 “Probability and cumulative probability density for the observed and simulated results . . .” –
299 what kind of “results” do you have here? The authors failed to state the name of the variable.

300 **Response:** Sorry for the confusion. The “results” means the observed and simulated BTs or reflectances.
301 We have rephrased the captions.

302

303 Figure 8 “The results are from Figure 4 marked by blue dashed lines” – couldn’t see the “blue dash line”
304 in Figure 4, and actually, there are too many elements in Figure 4.

305 **Response:** Sorry for the mistake. The caption has been changed into “*The profiles are for the track*
306 *marked by blue solid lines Figure 2.*” The regions or tracks particular discussed in the text are marked
307 by boxes or lines in the new Figure 2, and we have improved the figure. Furthermore, to present Figure
308 8 (Figure 7 in revision) more clearly, we have removed the cloud mixing ratio panels.

309

310 **Minor problems:**

311 Line 33: “The ERA5 reanalysis is found the most capability . . .” should be “The ERA5 reanalysis is
312 found to have the most capability . . .”

313 **Response:** Thanks, and we have updated the sentence.

314

315 Line 97: Do you have some references for the CRA-interim?

316 **Response:** Because the CRA reanalysis dataset is producing and it will be released in 2020, and only a
317 few papers have been published. Two papers by Liao et al. (2018) and Wang et al. (2018), which
318 discuss the datasets assimilated in the CRA, have been referred in the revision.

319

320 Line 142: “Ignore the uncertainties . . .” should be “Ignoring the uncertainties . . .”; In addition, is it
321 reasonable to assume mixed phase cloud can be ignored?

322 **Response:** Thanks. We have changed the “Ignore the uncertainties . . .” to “Ignoring the uncertainties”.
323 In our study, we distinguish cloud with different phases based on the temperature profiles, so the mixed
324 clouds are treated ice cloud and they are not ignored. We have tested that this would lead little bias, and
325 clarified this in the revision.

326

327 Line 187: “The correlation between the two is small.” – This sentence is vague, as it is not clear about
328 what are “the two”.

329 **Response:** It should be “the correlation between the CTT from CRA and the CTT from satellite
330 retrieval based on the solar measurement”. The section has been removed in the revision.

331

332 Line 191: “We notice that . . .” should be “It is noted that . . .”

333 **Response:** Thanks and we have removed the paragraph.

334

335 Line 215-217: The authors mentioned the cloud scattering properties in the CRTM are recalculated.
336 Then some necessary validation and description are needed to prove the validity of the new
337 implementation.

338 **Response:** The validation of the CRTM was done in our previous study (Yao et al., 2018). As discussed
339 in Figure 1 of Yao et al. (2018), the BTDs between the CRTM and rigorous (DISROT+LBLRTM)
340 simulations for ice and water clouds in different channels are generally less than 1 K, and they coverage
341 to 0 K as cloud optical thickness increases to 10 or larger. We have added some discussion on the
342 validation of the cloud optical properties in the CRTM model in the revision.

343

344 Line 230: “From” should be “from”

345 **Response:** Corrected.

346

347 Line 272: “with a mean BTs of . . .” should be “with a mean BT of ...”

348 **Response:** Thanks, and it has been corrected.

349

350 Line 324-325: “an abnormal excessive cloud mixing ratio” should be “an abnormally excessive cloud
351 mixing ratio”

352 **Response:** Corrected.

353

354 Line 373: “as marked in region A in Figure A” – where is Figure A?

355 **Response:** It should be Figure 2 in the revision, and we have changed it.

356

357 Line 390: “the in-site observation”?

358 **Response:** We have changed it into “the in-situ observation”

359

360 Line 413: “demonstrate that . . .” should be “demonstrating that . . .”

361 **Response:** Thanks, and it has been corrected.

362

363 **References:**

364 Miles, N. Y., Verlinde, J., and Clothiaux, E. E.: Cloud Droplet Size Distributions in Low-Level
365 Stratiform Clouds, *J. Atmos. Sci.*, *57*, 295-311, 2000.

366 Thompson, G., Rasmussen, R. M., and Manning, K.: Explicit forecasts of winter precipitation using an
367 improved bulk microphysics scheme, Part I: Description and sensitivity analysis, *Mon. Wea. Rev.*,
368 132, 519-542, 2004.

369 Wendish, M., and Yang, P.: *Theory of Atmospheric Radiative Transfer. A Comprehensive Introduction*,
370 1st ed., WILEY-VCH Verlag Gmbh, Weinheim, Germany, 2012.

371 Heymsfield, A. J., and McFarquhar, G. M.: On the high albedos of anvil cirrus in the tropical Pacific
372 warm pool: Microphysical interpretations from CEPEX and from Kwajalein, Marshall Islands, *J.*
373 *Atmos. Sci.*, 53, 2424-2451, 1996.

374 Platt, C. W. R.: A Parameterization of the Visible Extinction Coefficient of Ice Clouds in Terms of the
375 Ice/Water Content, *J. Atmos. Sci.*, 54, 2083-2098, 1997.

376 Heymsfield, A. J., Matrosov, J. S., and Baum, B.: Ice water path optical depth relationships for cirrus
377 and deep stratiform ice cloud layers, *J. Appl. Meteor.*, 42, 1369-1390, 2003.

378 Yao, B., Liu, C., Yin, Y., Zhang, P., Min, M., and Han, W.: Radiance-based evaluation o WRF cloud
379 properties over East Asia: Direct comparison with FY-2E observations, *J. Geophys. Res.*, 123,
380 4613-4629, 2018.

381 Liao, J., Hu, K., Jiang, H., Cao, J., Jiang, L., Li, Q., Zhou, Z., Liu, Z., Zhang, T., and Wang, H.:
382 Pre-Process and Data Selection for Assimilation of Conventional Observations in the CMA Global
383 Atmospheric Reanalysis, *Advances in Met S&T.*, 8, 133-142, 2018.

384 Wang, M., Yao, S., Jiang, L., Liu, Z., Shi, C., Hu, K., Zhang, T., Zhang, Z., and Liu, J.: Collection and
385 Pre-Processing of Satellite Remote-Sensing Data in CRA-40 (CMA's Global Atmospheric
386 ReAnalysis), *Advances in Met S&T.*, 8, 158-163, 2018.

387