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2 **Extension of a multisensor satellite radiance-based**

3 **evaluation for cloud system resolving models**

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30

Abstract

31

32

33 As an alternative approach to the previous multisensor satellite evaluation method of
34 cloud system resolving models, a method is presented using combined infrared and
35 microwave channels for precipitation clouds in cloud system resolving models over the
36 ocean. This method determines characteristics of cloud-top temperatures and ice
37 scatterings for clouds using infrared 11- μm and microwave high frequencies (89.0 GHz)
38 brightness temperatures (TBs). The threshold of the TB at low frequencies (18.7 GHz) is
39 also used to identify precipitation regions. This method extends the previous approach
40 via the wider swath of the passive microwave sensor and sensitivities to ice clouds
41 compared to the previous Tropical Rainfall Measuring Mission (TRMM)-based analysis
42 method using the narrower coverage of the Precipitation Radar.

43 The numerical results of the non-hydrostatic icosahedral atmospheric model (NICAM)
44 with two cloud microphysics schemes are evaluated over the tropical open ocean using
45 this method. The intensities of the scatterings in the two simulations at 89.0 GHz are
46 different due to the parameterizations of the snow and graupel size distributions. A
47 bimodal size distribution of the snow improved the underestimation of the TBs at 89.0
48 GHz. These results have a similar structure to the joint histograms of cloud-top
49 temperatures and precipitation-top heights in the previous method: the overestimated
50 intensity of scattering and the frequencies of high precipitation-top heights above 12 km

51 in the control experiment. We find that the change in the snow size distribution in the
52 cloud microphysics scheme can lead to better agreements of simulated TBs at 89.0 GHz
53 with observations. We further investigate impacts of non-spherical assumptions for snow
54 using a satellite simulator. The effect of a non-spherical shape of snow in the radiative
55 transfer model causes a smaller change of TBs at 89.0 GHz compared to the difference
56 between the TBs of the two simulations without non-spherical assumptions.

57

58 **Keywords** Cloud system resolving models; a satellite simulator; evaluations of cloud
59 microphysics; passive microwave satellites

60

61 **Introduction**

62 Recently, various methods have been proposed to evaluate and improve the cloud
63 microphysics schemes of cloud system resolving models (CSRMs) using satellite data. One
64 method is a radiance-based evaluation to avoid making assumptions about the
65 microphysical properties between retrieval algorithms and CSRMs using a satellite
66 simulator (Masunaga 2010; Hashino 2013; Matsui 2014). Using the Tropical Rainfall
67 Measuring Mission (TRMM) and a satellite simulator (Matsui et al. 2009, 2016), Roh and
68 Satoh (2014) (hereafter RS14) improved cloud properties over the tropical Pacific Ocean
69 simulated by the non-hydrostatic icosahedral atmospheric model (NICAM; Tomita and
70 Satoh 2004; Satoh et al. 2008; Satoh et al. 2014), such as the precipitation cloud statistics
71 in terms of the cloud-top temperatures (CTT) and the precipitation-top heights (PTHs), and
72 the contoured frequency altitude diagrams of the radar reflectivities. RS14 improved their
73 single-moment bulk microphysics scheme using the preferred size distributions from
74 multiple sensitivity tests to reproduce realistic cloud statistics, accumulated precipitation,
75 and outgoing longwave radiation (OLR). Roh et al. (2017) expanded RS14 to evaluate
76 global simulations with 3.5-km mesh horizontal resolutions and found improvements in
77 various types of clouds over the global domain in comparison with the TRMM and CloudSat
78 observations.

79 RS14 and Roh et al. (2017) used the method proposed by Matsui et al. (2009) known as
80 the TRMM Triple-Sensor Three-Step Evaluation Framework (T3EF). This method is

81 innovative because multisensor observations are used to evaluate the CSRMs using
82 satellite simulators. However, because TRMM precipitation radar (PR) has a relatively
83 narrow 250-km swath, the applicability of T3EF is limited to sampling volumes for relatively
84 short-term and small domain simulations.

85 Passive microwave observations in satellites have been used to estimate quantitative
86 precipitation. The low frequencies (<20 GHz) of microwave sensors have a direct physical
87 relationship between the path integrated water content of the rain and radiances over the
88 ocean. For high frequencies (>80 GHz), the radiances are depressed by the scattering of
89 large ice particles such as snow, graupel, and hail. The microwave observations enable us
90 to obtain precipitation information, including ice clouds, particularly in deep convective
91 systems.

92 Previous studies have used microwave observations to evaluate the precipitation systems
93 of CSRMs (Eito and Aonashi (2009); Matsui et al. 2009; Han et al. 2010). For example, Eito
94 and Aonashi (2009) found fairly good agreement between the simulated and observed
95 brightness temperatures (TBs) at 18.7 GHz but a stronger intensity of scattering at high
96 frequencies (36.5 GHz and 89.0 GHz) than in the observations, using the Japan
97 Meteorological Agency non-hydrostatic model (JMA-NHM) and observations of the
98 Advanced Microwave Scanning Radiometer for Earth Observing System (AMSR-E).

99 In this study, we introduce a method of radiance-based evaluation using microwave and
100 infrared channels and satellite simulators over the ocean. This method is an extension of

101 T3EF and has two advantages: (1) a wider range of coverage compared to TRMM PR and
102 (2) ice cloud information. We evaluate the numerical results from NICAM using two cloud
103 microphysics schemes over the tropical open ocean and investigate the impact of the snow
104 and graupel size distributions.

105

106 **Numerical experimental design and observational data**

107 NICAM is a global non-hydrostatic model that can be used as a regional model by
108 transforming the horizontal grid system to focus on a region of interest (the stretched
109 NICAM; Tomita 2008b). Simulations using the stretched version of NICAM are evaluated
110 following the case of RS14. The analysis domain is over the tropical central Pacific region
111 between latitudes 10° S and 10° N and longitudes 170° E and 170° W. Mesoscale convective
112 systems (MCSs) are dominant in this case and the convective band occurred at
113 approximately 5° N latitude over the analysis domain. The minimum of the horizontal grid
114 spacing of the domain is 2.4 km at the central point of the domain and most of the grid sizes
115 are less than 5 km. We focus on the period from 06 UTC January 1 to 06 UTC January 6
116 2007. Two microphysics schemes are used and evaluated: the original NICAM
117 Single-Moment Water (NSW6) scheme (hereafter, CON; Tomita, 2008a) and the modified
118 NSW6 scheme (hereafter, MODI) following Roh et al. (2017).

119 The microwave data are from AMSR-E on the Aqua satellite. AMSR-E has a 1450-km
120 swath at the surface and measures microwave emissions at six separate frequencies

121 (6.925 GHz, 10.65 GHz, 18.7 GHz, 23.8 GHz, 36.5 GHz, and 89.0 GHz). The
122 dual-polarized microwave radiometer operates at the six separate frequencies for both the
123 vertical and horizontal polarizations. The polarization differences between the vertically
124 polarized and horizontally polarized brightness temperatures at 18.9 GHz (D19; Liu and
125 Curry 1998; Eito and Aonashi 2009) and the polarization corrected 89-GHz brightness
126 temperatures (PCT89, Spencer et al. 1989) are derived from AMSR-E L2A (Ashcroft and
127 Wentz 2013), which has a spatial resolution of 39 km at 18.7 GHz and 12 km at 89.0 GHz.
128 D19 is representative of the atmospheric emission from liquid water itself and is close to 0
129 in the strong emission areas (Liu and Curry 1998). PCT89 is calculated in order to reduce
130 an impact of the inhomogeneity of surface emissivity (Kidd 1998)

131 The CTTs are taken from the 11- μm infrared channels with 0.04° resolution and 30 minutes
132 interval on the Multifunction Transport Satellite (MTSAT) geostationary satellite. The CTTs
133 are matched with AMSR-E data having the closest time. We sampled observation data from
134 the same period of simulations.

135 To compare our results with RS14, joint histograms of the CTT and PTH are constructed
136 (Masunaga and Kummerow 2006; Matsui et al. 2009) following RS14. The PTH is identified
137 as the highest altitude of the layer above a PR reflectivity of 20 dBZ. The infrared 11- μm
138 TBs for the CTTs from the TRMM 1B01 product and the 13.8 GHz reflectivity and orbital
139 precipitation from TRMM 2A25 are used for the PTH.

140 We used two satellite simulators to compare the radiances of the observational data with

141 those of the NICAM data: the first, satellite simulator developed at Florida University
142 (hereafter the Liu simulator; Liu 1998; Liu 2008) for the 18.7 GHz and 89.0 GHz microwave
143 TBs of AMSR-E, and the second, Satellite Data Simulator Unit (SDSU; Masunaga et al.
144 2010) version 2.1.4 for the 11- μm TBs and radar reflectivities at 13.8 GHz. The four-stream
145 radiative transfer model is used and the non-spherical database for the ice clouds is
146 implemented in the Liu simulator. The Gaussian beam convolution is applied to the two
147 microwave channels in the Liu simulator following Masunaga and Kummerow (2005) to
148 reduce the uncertainty from the antenna pattern of passive microwave sensors. The
149 extinction and scattering properties of the hydrometeors are calculated based on the Mie
150 calculation in the two simulators. The effective dielectric constant of snow and graupel is
151 calculated based on the Maxwell-Garnett approach (Maxwell-Garnett 1904), which was
152 generalized by Bohren and Battan (1982). We assumed the snow density (ρ_s) like $\rho_s = 100$
153 kg m^{-3} in CON, and $\rho_s = 0.15 D^{-1} \text{ kg m}^{-3}$ in MODI, where D is the diameter of snow in m. In
154 Section 4, we compare the results of the spherical assumption with those of the
155 non-spherical databases in the Liu simulator (Liu 2008).

156

157 **Methodology**

158 Liu et al. (1995) presented a cloud classification method using CTT of geostationary
159 meteorological satellite (GMS)-4 and a microwave index from scattering and emission
160 channels of the special sensor microwave/imager (SSM/I) over ocean. Matsui et al. (2014)

161 introduced an evaluation method based on joint diagrams using the CTT (MODIS 11- μm
162 TBs) and PCT89 of AMSR-E to classify cloud types over land following the Liu et al. (1995).
163 The two previous studies use microwave index and PCT89 to distinguish between
164 non-precipitation and precipitation clouds. In this study, we introduce an evaluation method
165 for only precipitation clouds using CTT and PCT89 as a scattering intensity from large ice
166 particles. We discriminate the precipitation clouds by additionally introducing D19 to
167 improve the detection of shallow precipitation clouds than Matsui et al. (2014). This method
168 enables us to investigate high and low precipitation systems using CTT and PCT89 over
169 the ocean. The precipitation areas are identified using the D19 threshold.

170 Figure 1a shows the joint histogram in terms of the CTT and PCT89 over the domain.
171 There are high frequencies of mixtures of clear sky and cloudy regions with higher CTT and
172 PCT89 than 280 K (Fig. 1a). Figure 1a shows all clouds and does not clearly distinguish
173 between non-precipitation and shallow precipitation. Therefore, we use D19 to focus on the
174 precipitation areas by using a D19 threshold of 50 K; this corresponds to approximately 1
175 mm hr^{-1} for a typical tropical profile according to Liu and Curry 1996 (in their Fig. 1a).
176 Figure 1b shows the joint histogram of the CTT and PCT89 for the areas whose D19 is
177 lower than 50 K. Using this histogram, we can classify the precipitation clouds into three
178 categories, “warm,” “cold,” and “deep” clouds, using the thresholds of CTT and PCT89. The
179 systems are clearly divided into low and high precipitation clouds by a CTT of 245 K. We
180 refer to clouds with a CTT higher than 245 K as “warm clouds.” For clouds with a CTT less

181 than 245 K, we further divide them into two categories: “cold clouds” for clouds with a
182 PCT89 larger than 250 K and “deep clouds” for clouds with a PCT89 less than 250 K.

183 Note that the terminology introduced in this study does not precisely follow conventional
184 definitions. “Warm clouds” are mixtures of shallow and cumulus congestus precipitations.
185 According to Machado et al. (1998), the infrared threshold of clouds associated with deep
186 convection is estimated to be 245 K. “Cold clouds” consist of stratiform precipitation in
187 MCSs and shallow/congestus with overlapped cirrus ($CTT < 245 \text{ K}$ and $PCT89 > 250 \text{ K}$).
188 “Deep clouds” ($CTT < 245 \text{ K}$ and $PCT89 < 250 \text{ K}$) are convective and stratiform
189 precipitation of MCSs related to the depression of PCT89 due to snow and graupel.

190

191 **Evaluation of NICAM**

192 Using the method introduced in the previous section, we evaluated two simulation results
193 using NICAM. Figure 2 shows the relationship between the averaged D19 and rain rate
194 from the simulations with the two microphysics schemes over the tropical open ocean using
195 the Liu simulator. In these simulations, as in the observations (Liu and Curry 1998), the D19
196 decreases as the precipitation increases. This is because the emission of precipitating
197 liquid hydrometeors at 18.7 GHz weakens D19 from the polarized ocean surface. D19 has
198 limitations when observing heavy precipitation of over 12 mm hr^{-1} due to the saturation of
199 the microwave radiation. The precipitation rate for a D19 of 50 K is approximately 1 mm hr^{-1} .
200 These characteristics linking D19 and the rain rate are consistent with the tropical case with

201 no ice (Liu and Curry 1998). We found that D19 is not sensitive to the microphysics
202 schemes in CON and MODI, even though the D19 of CON reproduces slightly larger
203 averaged precipitation rates than MODI in the D19 regime between 25 K and 60 K.
204 According to Sasaki et al. (2007), the effects from various rain size distributions are small
205 on the simulated brightness temperature at 19 GHz. This lack of sensitivity to cloud
206 microphysics schemes indicates that D19 is a good index of the surface precipitation.

207 Figure 3 shows the joint histograms of the CTT and PCT89 from simulation results of CON
208 and MODI using the satellite simulators. MODI (Fig. 3b) is closer to the observations (Fig.
209 1b) than is CON (Fig. 3a); MODI shows structures similar to the observations such as a
210 high frequency of cold clouds and a minimum value of PCT89 near 210 K. Conversely, in
211 the case of CON, the frequency of deep clouds is overestimated and the frequency of cold
212 clouds is underestimated compared to the observations and MODI. Deep clouds in CON
213 have PCT89 much lower than 200 K, which is not seen in the observations or MODI. MODI
214 reproduces a smaller fraction of the deep clouds (5.7%) than is seen in the observations
215 (10.6%). Both of the simulated warm clouds are overestimated compared to the
216 observations.

217 To understand how MODI improves PCT89, we examined its sensitivity to the size
218 distribution of the ice hydrometeors (Fig. 4). First, only the snow size distribution in MODI
219 was changed from MODI to CON (Fig. 4a). The use of the snow size distribution of CON
220 leads to underestimations of PCT89 compared to the case with the snow size distribution of

221 MODI. That is, the assumptions in MODI, in which the bimodal size distribution is
222 parameterized by temperatures and the ice water contents are based on Field et al. (2005),
223 lead to an increase in PCT89 compared to that in CON, in which a constant intercept
224 parameter with a negative exponential distribution is assumed. Second, only the graupel
225 size distribution in MODI was exchanged from MODI to CON. The graupel size distribution
226 had a small impact compared to that of the snow size distribution. The joint histogram is
227 more widespread than that of the original MODI. Because the interceptor parameter of
228 graupel in MODI was increased by 100 times compared to that of CON, MODI produces
229 smaller sizes of graupel than does CON. From the above sensitivity, we conclude that the
230 snow size distribution in MODI is the primary factor influencing the improvement in PCT89.
231 In Figure 5, we compare the above joint histogram of the CTT and PCT89 with that of CTH
232 and PTH as used in RS14. To capture precipitation regions, we focus on the regions with
233 larger surface radar reflectivity, above 20 dBZ; note that a TRMM PR of 21.6 dBZ
234 corresponds to approximately 1 mm hr^{-1} for convective systems (Fig. 9 of Schumacher and
235 Houze 2010). We divided the cold and deep clouds using a PTH of 6 km instead of a
236 PCT89 of 250 K. The two joint histograms capture a similar contrast between CON and
237 MODI. For example, CON shows higher frequencies below 220 K compared to those above
238 a PTH of 12 km (Figs. 3b and 5b) and CON shows a systematic bias, such as at high
239 frequencies with a PTH of 12 km, which corresponds to high frequencies below 220 K that
240 are related to deep clouds. Deep clouds with higher PTH indicate larger sizes of graupel

241 and snow located at higher altitudes. According to RS14, the snow size distribution affects
242 the distribution of PTH. In this study, we found that the size distribution of snow primarily
243 affects the distribution of PCT89.

244

245 **Snow shape dependency**

246 The advantage of using PCT89 is that it is sensitive to ice clouds. Snow has various
247 shapes and affects the scattering properties of high frequency microwave channels like
248 PCT89. Snow shapes have not been fully taken into account in CSRMs and therefore
249 create uncertainties (Geer and Baordo 2014). The previous results are based on a soft
250 sphere assumption (SP) according to Mie calculation with effective dielectric constant from
251 the Maxwell-Garnett method. The impacts of the non-spherical assumptions of snow are
252 investigated on the normalized probability distribution of PCT89 over the analysis domain
253 using the Liu simulator in off-line (Fig. 6). We used the pre-computed optical properties of
254 five snow shapes, short column (SN1), thin plate (SN4), six bullet rosettes (SN8), dendrite
255 snow (SN10), and aggregate (SN11) according to the discrete dipole approximation (DDA,
256 Purcell and Pennypacker 1973) in a database of microwave single-scattering properties for
257 non-spherical ice particles (Liu 2008; Nowell et al. 2013). The normalized probability
258 distribution of PCT89 shows that MODI reproduces a more realistic distribution compared
259 to the observations than does CON. PCT89 in CON are underestimated compared to MODI
260 and the observations due to the overestimation of scattering from ice particles. In CON,

261 simple structures of snow such as the short column lead to lower PCT89 compared to
262 complicated shapes such six bullet rosettes, dendrite snow, and aggregate. Dendrite snow
263 has a similar distribution to that of the soft sphere assumption in CON and MODI. The
264 impact of the non-spherical assumption is different for CON and MODI. MODI experiences
265 a smaller impact from the snow shapes than does CON because the size of the snow is
266 smaller in MODI than in CON. The non-spherical assumption has a strong impact on the
267 large size snow. The PCT89 distribution pattern more strongly depends on the size
268 distribution of the snow than does the non-spherical assumptions.

269

270 **Summary**

271 We present an evaluation method for precipitation clouds in CSRMs using 11- μm TB from
272 a geostationary satellite and PCT89 and D19 from a passive microwave satellite over the
273 ocean using satellite simulators based on Liu et al. (1995) and Matsui et al. (2014). This
274 method can investigate precipitation clouds detected by D19 using two indices related to
275 the CTT and PCT89, which represent the cloud-top height and scattering intensities from
276 large ice particles, respectively. This method enables us to quantitatively evaluate shallow
277 and deep precipitation clouds in simulations by comparing them to their observational
278 counterparts.

279 Two NICAM simulations using different cloud microphysics schemes were evaluated over
280 the tropical open ocean. We found that the simulation results with the modified cloud

281 microphysics scheme (MODI) developed by RS14 closer to the observations than the
282 original scheme (CON). We found that the improvement of PCT89 in RS14 is primarily
283 derived from the change in the size distribution of the snow rather than that of the graupel in
284 the cloud microphysics scheme.

285 The new method presented in this study shows similar results to those for the
286 TRMM-based analysis in RS14. The passive microwave satellite has limitations when
287 distinguishing shallow and congestus precipitation compared to PTH from TRMM PR. The
288 distribution of the joint histograms of PTH and CTT clearly shows three modes: shallow
289 clouds ($PTH < 4$ km and $CTT < 273$ K), congestus ($4 < PTH < 6$ km and $CTT > 245$ K), and
290 deep clouds ($PTH > 6$ km and $CTT < 245$ K). The advantage of the joint histogram with CTT
291 and PTH is that it can distinguish shallow and congestus clouds clearly using a PTH of 4 km
292 and a CTT of 245 K (Fig. 5a). TRMM PR has a finer horizontal resolution (5 km) for
293 detecting precipitation regions compared to D19 (approximately 18 km). However, passive
294 microwave satellites are abundant, e.g., AQUA, TRMM, and Global Precipitation
295 Measurement (GPM), and the swath that is covered is approximately 1450 km compared to
296 the 247 km covered by TRMM PR. This means that the present method is more appropriate
297 when collecting sampling data with wider coverage than those of TRMM PR.

298 In addition to the wider swath coverage, passive microwave satellite sensors are sensitive
299 to ice cloud properties. We tested non-spherical assumptions for snow compared to the Mie
300 theory using a database of optical properties from DDA with a satellite simulator. The size

301 distribution of snow had a strong impact on the depression of PCT89 compared to the
302 non-spherical assumptions. The impact of the non-spherical assumption on the distribution
303 of PCT89 depends on the parameterization of the snow size distribution in the cloud
304 microphysics schemes.

305 Many more satellites carry passive microwave sensors than carry precipitation radar
306 sensors. That means that more sampling data are available from passive microwave
307 sensors. Therefore, it is possible to evaluate the simulations of CSRMs with a smaller
308 domain and shorter integration time than is possible with a method using active microwave
309 sensors. In addition, it should be useful to investigate quantitatively warm and deep
310 precipitation clouds in the different environmental conditions such as sea surface
311 temperatures and, more importantly, ice cloud properties, which are one of the most
312 uncertain parts of CSRMs, should be evaluated and improved.

313

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Fig. 1

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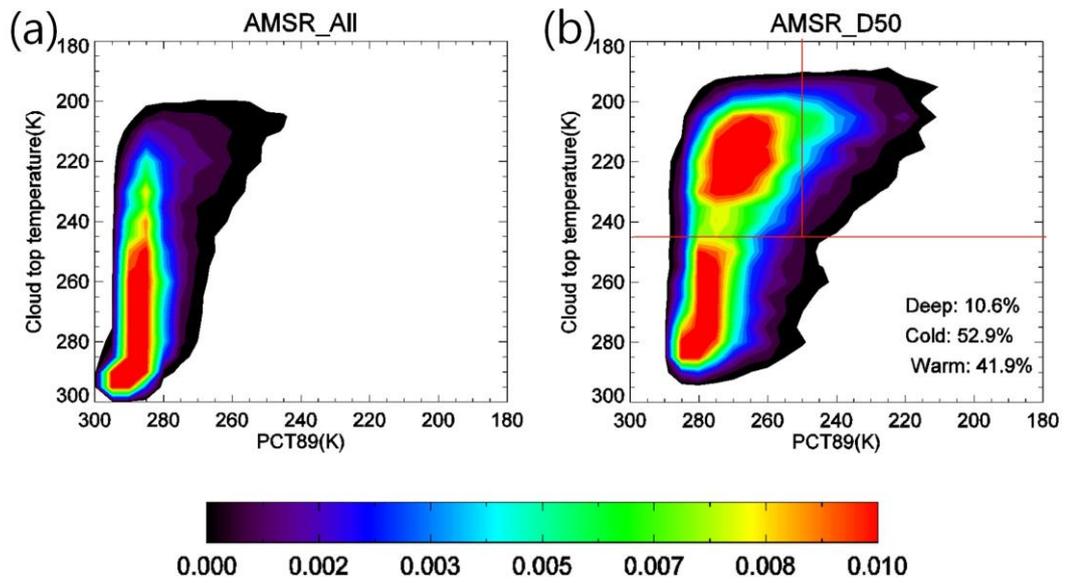
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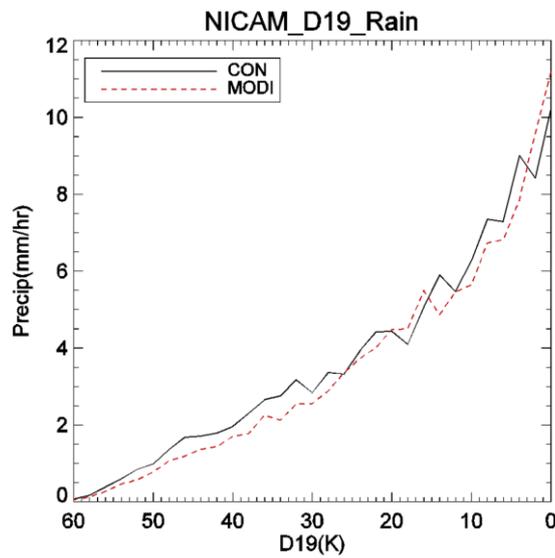


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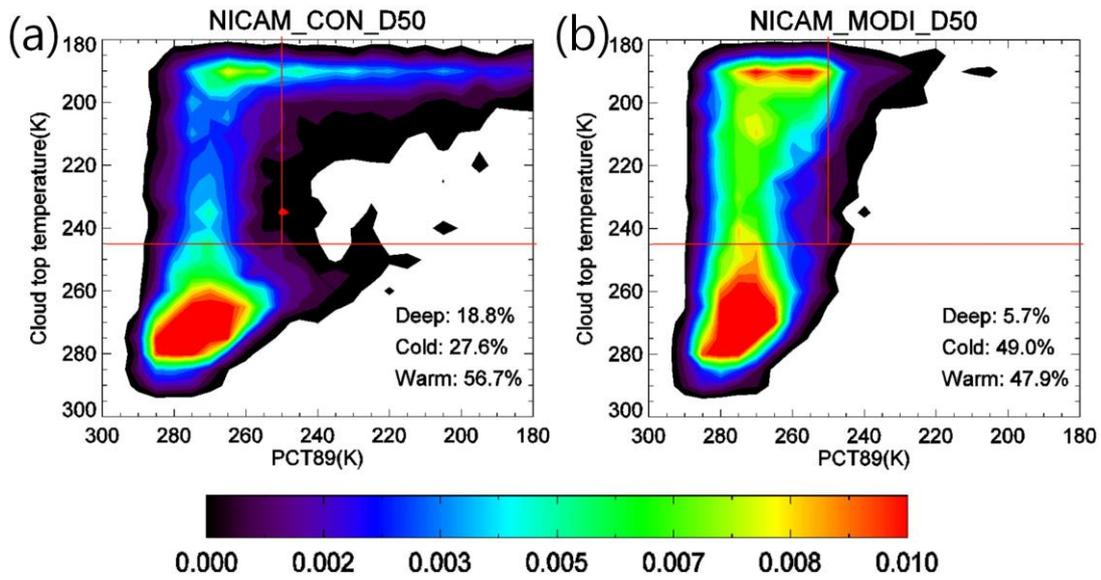


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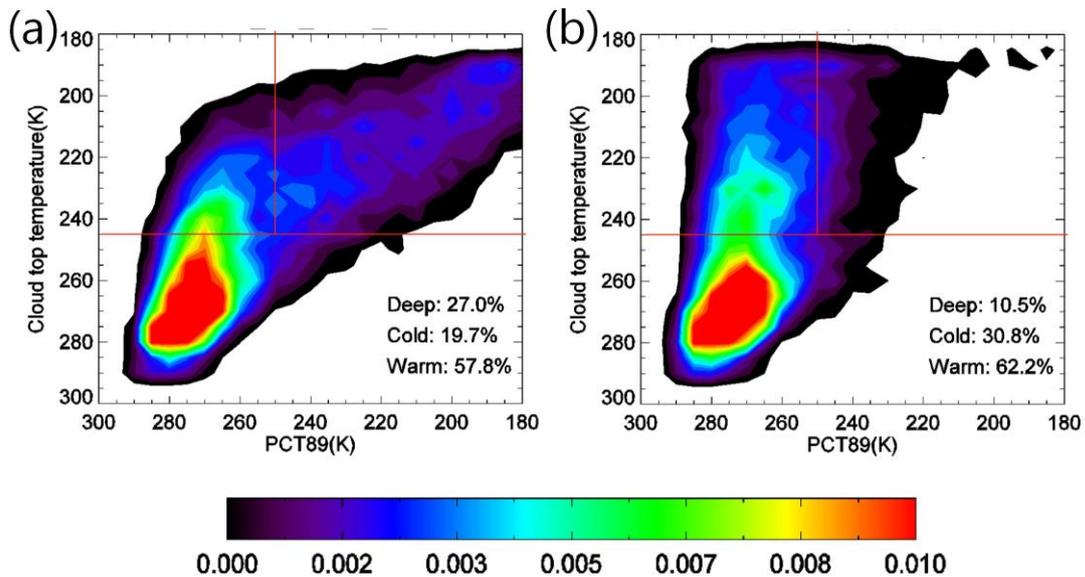
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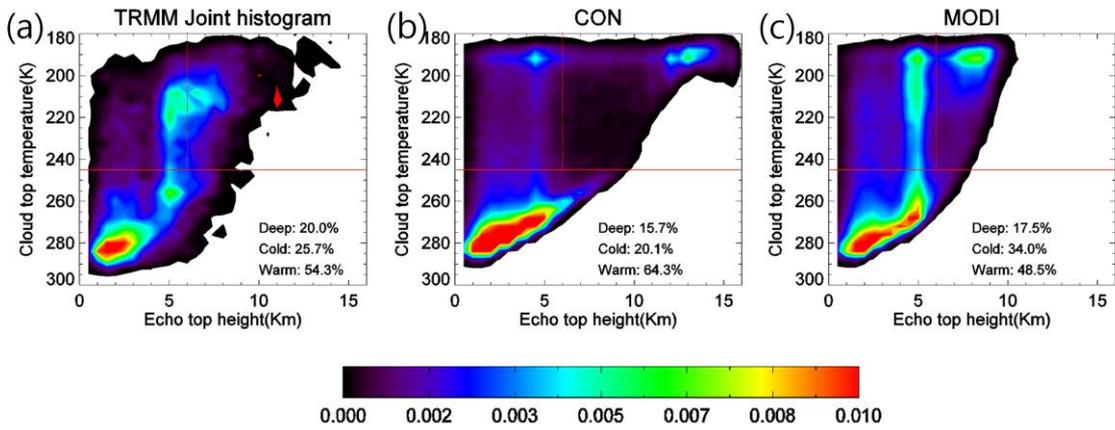


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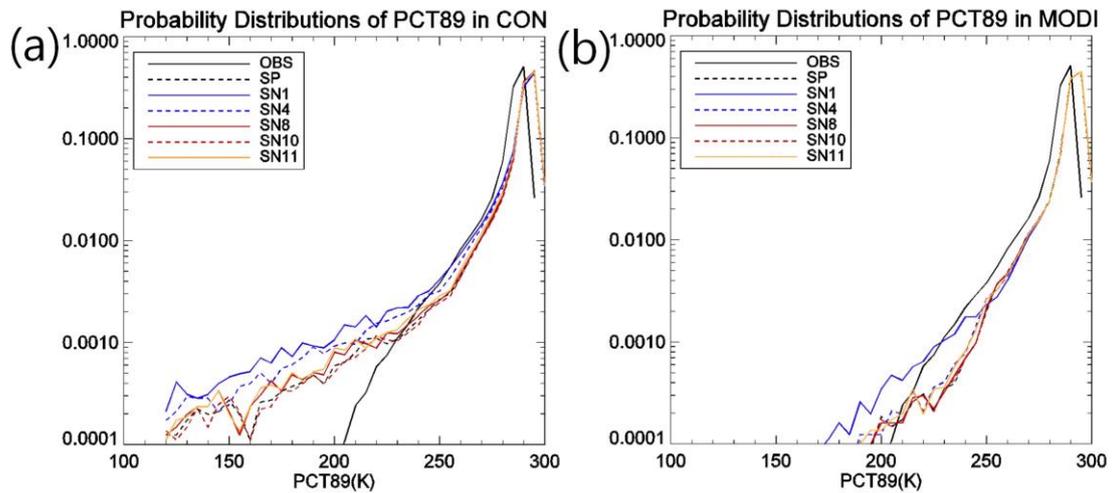


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