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1 **Impact of resolution and parameterized**
2 **convection on the diurnal cycle of**
3 **precipitation in a global nonhydrostatic**
4 **model**

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Abstract

20 A series of 40-day non-hydrostatic global simulations was run with the
21 NASA Goddard Earth Observing System (GEOS) model with horizontal
22 grid spacing ranging from 50 km to 3.5 km. Here we evaluate the diurnal
23 cycle of precipitation and organized convection as a function of resolution.
24 For validation we use the TRMM 3B42 and IMERG precipitation prod-
25 ucts and 4 km Merged Infrared brightness temperature, focusing on three
26 regions: the contiguous United States (CONUS), the Maritime Continent,
27 and Amazonia. We find that higher resolution has mixed impacts on di-
28 urnal phase. Regions dominated by non-local propagating convection show
29 the greatest improvement, with better representation of organized convec-
30 tive systems. Precipitation in regions dominated by local thermodynamic
31 forcing tends to peak too early at high resolution. Diurnal amplitudes in all
32 regions develop unrealistic small-scale variability at high resolution, while
33 amplitudes tend to be underestimated at low resolution. The GEOS model
34 uses the Grell-Freitas scale-aware convection scheme, which smoothly re-
35 duces parameterized deep convection with increasing resolution. We find
36 that some parameterized convection is beneficial for the diurnal amplitude
37 and phase even with a 3.5 km model grid, but only when throttled with
38 the scale-aware approach. An additional 3.5 km experiment employing the
39 GFDL microphysics scheme and higher vertical resolution shows further im-

40 provement in propagating convection, but an earlier rainfall peak in locally
41 forced regions.

42 **Keywords** diurnal; convection; parameterization; precipitation; nonhydro-
43 static

44 **1. Introduction**

45 The diurnal variation of clouds and precipitation is an important facet
46 of the energy and water cycles, which general circulation models (GCMs)
47 have historically struggled to represent. The most widely documented de-
48 ficiency has been a daily maximum over land that is too tightly coupled
49 to surface heating, peaking around local noon instead of late afternoon
50 or evening (Yang and Slingo 2001; Betts and Jakob 2002; Dai and Tren-
51 berth 2003; Clark et al. 2007; Brockhaus et al. 2008; Stratton and Stirling
52 2012). In addition to direct impacts on short-range forecasts, diurnal biases
53 can rectify onto longer timescales by creating energy and water imbalances
54 (Bergman and Salby 1997). Cloud fraction that incorrectly peaks near local
55 noon can amplify cloud shortwave forcing and alter the surface energy bal-
56 ance. Similarly, mid-day rainfall is more prone to evaporation, potentially
57 inducing a surface dry bias (Del Genio 2012). These issues limit the ability
58 of models to represent climate sensitivity, drought and flood conditions, and
59 other aspects of the Earth system.

60 Most weather and climate models use horizontal grid spacing of tens of
61 kilometers, and smaller scale processes such as moist convection and bound-

62 ary layer turbulence must be represented with subgrid parameterizations.
63 Because the diurnal variation is driven by large and well defined external
64 forcing, model representation of the diurnal cycle offers an ideal test of the
65 parameterized physics (Yang and Slingo 2001).

66 Many efforts to improve diurnal cycle simulation have focused on pa-
67 rameterized convection. Experiments with cloud resolving models (CRMs)
68 have suggested that entrainment rates vary diurnally as convection tran-
69 sitions from shallow to deep (Grabowski et al. 2006; Del Genio and Wu
70 2010), and that this transition is often mediated by boundary layer cold
71 pools, which organize updrafts on larger scales (Khairoutdinov and Randall
72 2006; Kuang and Bretherton 2006).

73 Studies with parameterized convection have found that increasing en-
74 trainment rates can improve the diurnal cycle by inhibiting deep convection
75 (Bechtold et al. 2004; Wang et al. 2007). Stratton and Stirling (2012) tied
76 convective entrainment rates to the lifting condensation height to improve
77 the diurnal cycle in the Met Office climate model. Rio et al. (2009) included
78 a representation of sub-grid boundary layer processes, such as gust fronts,
79 to improve diurnal rainfall in the LMDZ model. Efforts have also consid-
80 ered other aspects of convection parameterization, including boundary layer
81 coupling and trigger functions (Lin et al. 2000; Lee et al. 2007; Suhas and
82 Zhang 2014).

83 Although deficiencies in parameterized convection have rightly received
84 much attention, GCMs likely struggle with the diurnal cycle for different
85 reasons in different regions. Over the central United States, warm season
86 precipitation has a nocturnal peak associated with propagating mesoscale
87 convective systems (MCSs), which account for up to half of summer rainfall
88 (Jiang et al. 2006). Developing leeward of the Rocky Mountains thousands
89 of kilometers to the west, orogenic MCSs can persist for many hours or even
90 days. Although there have been recent efforts to parameterize (Moncrieff
91 et al. 2017) or explicitly simulate (Pritchard et al. 2011) such systems, in
92 general they remain poorly represented in coarse-grid GCMs.

93 Similarly, the diurnal cycle in coastal regions is often associated with
94 land-sea breezes, driven by the differing heat capacities of water and dry
95 land, and the diurnally varying thermal contrast that results. Low level
96 convergence associated with sea breeze fronts can trigger convective storms,
97 which in turn can grow upscale into MCSs (Carbone et al. 2000). Propaga-
98 tion of the MCSs is often guided offshore by coupled gravity wave dynamics,
99 which destabilize and moisten the lower troposphere ahead of the MCS. Such
100 diurnal waves can be forced by stratiform heating associated with the MCS,
101 or potentially influenced by orogenic systems excited by nearby topography
102 (Ruppert et al. 2020; Mapes et al. 2003). These dynamics are particu-
103 larly important for the diurnal cycle over the Maritime Continent, which

104 consists of an extensive network of islands, varying greatly in size and with
105 often mountainous topography (Yang and Slingo 2001). As with continen-
106 tal mesoscale systems, the relevant horizontal scales are measured in tens
107 of kilometers, and the dynamics are poorly resolved in most global models.
108 Neale and Slingo (2003) demonstrated the difficulty of correctly simulating
109 the Maritime Continent diurnal cycle in a model with inadequate resolution.

110 Some of the above issues can be remedied with finer model grid spac-
111 ing, as topographic features, land-sea contrasts and mesoscale organization
112 begin to be resolved. Regional modeling studies have generally shown im-
113 proved diurnal variability with higher resolution. Gao et al. (2017) found
114 improved convection propagation and diurnal timing using the Weather Re-
115 search and Forecasting (WRF) model over North America when grid spacing
116 was reduced from 36 km to 4 km. Pearson et al. (2010) and Kendon et al.
117 (2012) both showed similar improvement over West Africa in regional ex-
118 periments with the Met Office Unified Model, and Love et al. (2011) found
119 realistic diurnal propagation offshore of the Maritime Continent with 4 km
120 grid spacing. Notably, Pearson et al. (2014) argued that the improvement
121 seen over West Africa was a consequence of the convection representation,
122 rather than the increased resolution itself. Their experiments with 12 km
123 and 4km spacing both showed similar skill, as long as the parameterized
124 convection was similarly restricted.

125 Global models, too, have been used to explore this resolution depen-
126 dence, although in more limited number given the computational expense.
127 Dirmeyer et al. (2012) considered the diurnal cycle in three GCMs over
128 a wide range of grid spacing (125 km to 10 km), and found that mod-
129 els with higher resolution generally outperformed the coarser cases. A
130 super-parameterized model, in which the convection parameterization was
131 replaced with embedded two-dimensional cloud resolving models, outper-
132 formed aspects of the traditional GCMs but still trailed the high resolution
133 global model. Sato et al. (2009) showed that the Nonhydrostatic ICosa-
134 hedral Atmospheric Model (NICAM) also exhibits a resolution dependence
135 for grid spacing between 14 km and 3.5 km, particularly pronounced over
136 land, where the diurnal peaks at lower resolutions increasingly lagged ob-
137 servations.

138 Models used for global numerical weather prediction (NWP) now em-
139 ploy resolutions fine enough to permit mesoscale organization, though still
140 insufficient to resolve individual updrafts. The present study examines the
141 diurnal cycle of precipitation in one such model, the NASA Goddard Earth
142 Observing System (GEOS). The same GEOS executable is used in appli-
143 cations including NWP (12 km), seasonal forecasting and reanalysis pro-
144 duction (50 km; Borovikov et al. 2019), and global mesoscale modeling (6
145 km; Putman and Suarez 2011), with current typical grid spacing indicated.

146 Science-driven applications on specialized grids, such as the global stretched
147 grid or doubly-periodic domain (Arnold and Putman 2018), further expand
148 the possible model configurations. Scale-aware parameterizations become
149 necessary to ensure realistic simulation across resolutions.

150 Here we conduct a set of short simulations with globally quasi-uniform
151 grid spacing ranging from 3.5 km to 50 km. These are supplemented by ex-
152 periments in which the strength of parameterized deep convection is varied,
153 along with its closure assumptions, and a 3.5 km case using an alternative
154 microphysics scheme. We aim to evaluate the diurnal cycle as a function of
155 resolution across regions with a range of diurnal mechanisms.

156 In section 2, we describe the GEOS model, the experiment configura-
157 tion, and the datasets used for evaluation of the diurnal cycle. In section 3
158 we describe the simulated mean state to provide context for the analysis
159 to follow. Section 4 presents the diurnal cycle over the contiguous United
160 States (CONUS), and Sections 5 and 6 present analogous results over the
161 Maritime Continent and Amazonia. Section 7 describes experiments mod-
162 ulating the strength of parameterized convection, Section 8 examines the
163 role of microphysics, and in Section 9 we evaluate the distribution of cloud
164 sizes over CONUS and their diurnal variation. Conclusions are made in
165 Section 10.

166 2. Model and data description

167 2.1 Model

168 The Goddard Earth Observing System (GEOS) is a modular Earth sys-
169 tem model used for numerical weather prediction, seasonal forecasting, re-
170 analysis production and global mesoscale modeling. Deep convection is pa-
171 rameterized with the Grell-Freitas scheme (Grell and Freitas 2014; Freitas
172 et al. 2018). It is aerosol and scale-aware, with cloud condensation nuclei
173 (CCN)-dependent autoconversion and re-evaporation, and a dependence on
174 horizontal grid spacing based on Arakawa and Wu (2013). For reference,
175 the $1\text{-}\sigma$ scaling factor used here is roughly 0.2 at 12 km. Two plumes,
176 representing congestus and deep convection, are active here. The scheme
177 employs the non-equilibrium closure of Bechtold, et al. (2014), which re-
178 duces the available CAPE associated with rapid changes in boundary layer
179 forcing. Shallow convection is based on Park and Bretherton (2009), with
180 boundary layer turbulent mixing following Lock et al. (2000) and Louis
181 (1979). Longwave and shortwave radiation are calculated with the Rapid
182 Radiative Transfer Model for GCMs (Iacono et al. 2008, RRTMG), and
183 the land surface uses the catchment-based model of Koster et al. (2000).
184 The single-moment microphysics of Bacmeister et al. (2006), is used except
185 as noted below. We note that this model configuration is nearly identical

186 to that of the GEOS forward processing (FP) NWP system as of January
187 2020.

188 The experiments presented here are based on the DYnamics of the At-
189 mospheric general circulation Modeled On Non-hydrostatic Domains (DYA-
190 MOND) protocol (Stevens et al. 2019). DYAMOND is an intercomparison
191 project aimed at global convection-permitting non-hydrostatic models. The
192 present study is an extension of the baseline DYAMOND experiments, with
193 a wider range of horizontal grid spacing and use of parameterized convec-
194 tion. Most experiments presented here use 72 vertical levels. The single
195 exception is the official GEOS submission to the DYAMOND intercompari-
196 son, which uses 132 levels, and also employs the GFDL microphysics scheme
197 (based on Zhao and Carr, 1997). Results from this experiment (labeled “3
198 km GFDL”) are included in order to illustrate the impact of microphysics
199 and vertical resolution. All experiments are initialized on July 30, 2016,
200 and run for 40 days. Daily, time-varying sea surface temperature is taken
201 from 1/8 degree Operational Sea Surface Temperature and Sea Ice Analysis
202 (OSTIA). All simulations are run with the FV3 non-hydrostatic dynamical
203 core on a cubed-sphere grid (Putman and Lin 2007).

204 2.2 Data

205 To evaluate the simulated diurnal cycle we use several satellite datasets.
206 Precipitation was taken from both Version 7 of the Tropical Rainfall Mea-
207 suring Mission (TRMM) Multi-Satellite 3B42 0.25 degree dataset (Huffman
208 et al. 2007), and from version 6B of the 0.1 degree Integrated Multi-satellitE
209 Retrievals for GPM (IMERG; Tan et al. 2019). We find that, for the di-
210 urnal amplitude and phase studied here, the two datasets are almost iden-
211 tical. Most plots are based on the TRMM dataset, with IMERG reserved
212 for time-series, where its higher temporal resolution is beneficial. Outgo-
213 ing longwave radiation was taken from Edition 4 of the Clouds and Earth’s
214 Radiant Energy System (CERES) - Energy Balanced And Filled (EBAF)
215 top-of-atmosphere data (Loeb et al. 2018). Finally, the size distribution of
216 cloud clusters was evaluated against the NCEP/CPC global merged bright-
217 ness temperature dataset (Janowiak et al. 2001). Based on the 11-micron
218 channel from GMS-5, GOES-8, GOES-10, Meteosat-7 and Meteosat-5, it is
219 available 60°S-60°N every half hour on a roughly 4 km latitude/longitude
220 grid.

221 3. Mean precipitation

222 The mean precipitation for August 2016 is shown in Fig. 1 for the TRMM
223 dataset and GEOS with grid spacing from 3.5 km to 50 km (note 3.5 km

224 case is labeled 3 km in figures). Persistent model departures from observa-
225 tions include a slight underestimation of storm track precipitation in mid-
226 latitudes, and an overestimation of weak precipitation in the subtropical
227 subsidence regions, though some of this may be a result of missing driz-
228 zle in the TRMM product. There is also some resolution dependence to
229 regional precipitation biases over land, with grid spacing 12 km and finer
230 associated with an overestimation of precipitation over Africa, and an un-
231 derestimate over the North American Great Plains. Internal variability may
232 also contribute to regional disagreements. For example, August of 2016 was
233 an anomalously wet month over the Great Plains, with some areas receiv-
234 ing double the 10-year mean precipitation. Although the model was run
235 with historical forcing, the free-running land and atmosphere are unlikely
236 to reproduce observed weather events over the entire month.

237 The 50°S-50°N mean total precipitation is relatively constant with model
238 resolution (Table 1). Table 1 also lists the mean convective precipitation and
239 outgoing longwave radiation (OLR). The convective precipitation, meaning
240 that produced by the subgrid parameterizations, is seen to smoothly de-
241 crease with resolution, from roughly two thirds to one third of the total.
242 The mean OLR is, like the total precipitation, roughly constant across res-
243 olutions, and remains within 3 W m^{-2} of the observed value in all cases.

Table 1

244 4. Diurnal cycle over CONUS

245 Before evaluating the diurnal cycle, we first interpolate the precipitation
246 at each model resolution onto the TRMM 0.25 degree grid. The diurnal
247 harmonic is then calculated through a Fourier transform, and the diurnal
248 amplitude is defined from the real and imaginary Fourier components, a
249 and b , as $\sqrt{a^2 + b^2}$. The phase is defined as the hour of the first maximum
250 in the diurnal harmonic, and then shifted to local solar time (LST) such
251 that hour 12 corresponds to maximum top-of-atmosphere insolation.

252 Figure 2 shows the diurnal amplitude over the contiguous United States
253 (CONUS). Amplitudes smaller than 0.25 mm day^{-1} are masked. The
254 TRMM values using an August climatology from 2007-2016 are shown in
255 top left, while August 2016 alone is in the top right. This gives a sense of
256 the interannual variability in the August diurnal amplitude. The observed
257 August 2016 rainfall was marked by a historic flood event in Louisiana, and
258 above average rainfall across the Midwest. Note that the GEOS simulations
259 should only roughly reproduce historical weather events in the first few days
260 of the simulation, after which the growth of initial errors would cause the
261 model to diverge from observed history.

262 The TRMM climatology shows maximum diurnal amplitudes over the
263 southeast United States, over the northern Gulf of Mexico, and along the
264 Gulf of California. These features also appear in 2016, along with enhanced

265 precipitation over the central US. The lowest resolution GEOS experiments
266 closely resemble the climatology, with the exception of the Gulf of Mex-
267 ico, where the model consistently underestimates precipitation. The GEOS
268 model has a known subsidence bias over the Gulf of Mexico, which may
269 be linked to excessive mean precipitation and large-scale ascent in the east
270 Pacific ITCZ along the Central American Coast, and along the eastern
271 United States and Gulf Stream. Amplitudes increase at higher resolution,
272 with greater small-scale spatial variability, suggestive of excessive grid-scale
273 precipitation.

274 The phase is shown in Fig. 3. On this metric there is less difference
275 between the TRMM decadal and 2016 values. Both indicate late afternoon
276 peaks in precipitation over the southeast and mountain west, where convec-
277 tion is dominated by local thermodynamic instability. The ocean regions
278 show peaks in late morning and early afternoon, with a gradient consis-
279 tent with offshore propagation, while the central plains exhibit a nocturnal
280 peak associated with organized and long-lived convective systems (Wallace
281 1975; Carbone et al. 2002).

282 The model largely reproduces the late afternoon peak in the southeast,
283 although it is somewhat delayed in the 50 km case. As we will show below,
284 this is largely due to the non-equilibrium closure in the Grell-Freitas scheme
285 (Freitas et al. 2018). At low resolutions, the model delays precipitation too

286 much in the mountain west, with a peak in the evening rather than late
287 afternoon. This improves with resolution, and the 6 km and 3.5 km cases
288 are close to the TRMM phasing. Example time-series averaged over the
289 Mountain West, Great Plains, and Southeast are shown in Fig. 4, which
290 clearly illustrate these differences. The time-series also highlight the incon-
291 sistent amplitude in the 3.5 km case, which is reasonable in the southeast,
292 but too strong over the Mountain West and underestimated over the Great
293 Plains.

294 In the 50 km case, peak rainfall over the Great Plains occurs around
295 1800 LST, with a sharp drop into the late evening and early morning. This
296 contrasts with observations, which show persistent strong rainfall through
297 the night, and is consistent with the lack of propagating systems evident
298 in Fig. 3. The propagation appears to strengthen at 12, 6 and 3.5 km, al-
299 though it is still underestimated relative to observations. The improvement
300 in propagation is better illustrated in the Hovmoller diagrams shown in
301 Fig. 5. These show the composite diurnal hourly precipitation, normalized
302 by the August mean for each case, and meridionally averaged between 38°N
303 and 45°N. The IMERG product shows precipitation originating in the west
304 around 0000 UTC and then propagating eastward from 105°W to 95°W
305 over roughly eight hours. A white line indicating a 24 m s^{-1} propagation
306 speed is superimposed on the precipitation. In the 25 km and 50 km cases,

307 this eastward propagation is largely absent, and the precipitation over the
308 eastern US peaks coincident with the west. A more realistic slope is visi-
309 ble in the 12, 6 and 3.5 km cases, though not as robust as in the IMERG
310 dataset.

311 As noted above, the ability of the Grell-Freitas scheme to represent di-
312 urnal timing is in large part due to its non-equilibrium closure. To illustrate
313 the closure’s impact, we show in Fig. 6 the amplitude and phase of the diur-
314 nal precipitation over CONUS, for the original 50 km case, and an otherwise
315 identical case with the non-equilibrium closure disabled (DC0). In the DC0
316 case, the diurnal amplitude is slightly larger over the southeast, but other-
317 wise quite similar to the control case. However, the phase is significantly
318 altered, with precipitation peaking roughly six to eight hours earlier, around
319 local noon.

320 **5. Diurnal cycle over Maritime Continent**

321 Another region in which the diurnal cycle might be expected to show
322 sensitivity to model resolution is the Maritime Continent, where the diurnal
323 cycle is dominated by land-sea circulations (Mori et al. 2004). The diurnal
324 amplitude over the Maritime Continent is shown in Fig. 7. The observed
325 amplitudes are generally largest over and adjacent to the largest islands,
326 although in 2016 TRMM shows comparable amplitudes in many ocean re-

327 gions as well. The amplitudes in the model are somewhat underestimated
328 over ocean in the 50 km case, but increase monotonically with resolution,
329 and are larger than observed when grid spacing is below 12 km. This trend
330 is more exaggerated over land, with the 3.5 km case showing a significant
331 overestimation of diurnal amplitude.

332 The diurnal phase is shown in Fig. 8. Over land, the observed precipi-
333 tation is characterized by a peak in late afternoon or early evening. Over
334 oceans nearest the islands, the peak generally occurs in the early morn-
335 ing around 0800, and precipitation then propagates out to several hundred
336 kilometers away from the coast. Two notable exceptions to this are the
337 southwest coastline of Sumatra and the eastern coast of Malaysia, where
338 the coastal peaks begin 2-4 hours earlier.

339 The model captures the overall geographic dependence of diurnal phase
340 quite well. The near-coastal oceans generally match the observed peak
341 around 0800, although at lower resolutions that timing extends too far sea-
342 ward, often including regions which are observed to peak around 1000. Over
343 land, particularly near the coasts, the peak is too early.

344 Time-series of precipitation averaged 11°S-9°N over land and ocean are
345 shown in Fig. 9. These make clear the differences in phase, with the model
346 precipitation over land similar to observations in early morning, but with
347 a too-rapid increase during the day and a peak two hours early. As in the

348 amplitude plots (Fig. 7), the 3.5 km case has a more significant overestimate
349 of the diurnal amplitude than the 50 km case. Over ocean, the simulated
350 phase agrees well with observations, and the diurnal amplitude is generally
351 small relative to that over land, though it still visibly increases at higher
352 resolution. Precipitation associated with parameterized convection (dotted
353 lines in Fig. 9) comprises most of the total at 50 km, but less than half at
354 3.5 km.

355 We make a closer examination of the Sumatran land-sea circulation by
356 constructing hovmoller diagrams of precipitation and 10m wind. Figure 10
357 shows precipitation averaged as a function of distance from the southwest-
358 ern Sumatran coastline, with vectors indicating the strength and direction
359 of the onshore wind component. Negative distances indicate points over wa-
360 ter. The model shows little dependence on resolution, except for an increase
361 in the diurnal peak precipitation over land, and a weaker increase in the off-
362 shore diurnal amplitude, as grid spacing is reduced. Each case captures the
363 offshore propagation with similar timing, lagging the IMERG precipitation
364 by roughly two hours, despite peaking too early over land. Offshore, the
365 model generally underestimates the mean precipitation relative to IMERG.

366 6. Diurnal cycle over Amazonia

367 Lastly, we consider the diurnal cycle in a continental tropical regime,
368 over South America. The diurnal amplitudes are shown in Fig. 11. Observed
369 amplitudes are largest along the northern coastline, where mean precipita-
370 tion is contiguous with the inter-tropical convergence zone in northern sum-
371 mer. The model at low resolution tends to underestimate amplitude along
372 the northern coast, and produces excessive precipitation over the southern
373 Amazon basin. Both of these issues are improved at high resolution, but
374 the model then develops a band of excess precipitation along the Bolivian
375 Andes, and suffers from the general appearance of strong grid-scale precip-
376 itation, as noted previously over CONUS.

377 The diurnal phase is shown in Fig. 12. Observed precipitation over
378 the Amazon basin is mostly characterized by a peak in late afternoon or
379 early evening, while the northern mountainous regions, from Colombia to
380 Venezuela and French Guiana, show a nocturnal peak. These regions also
381 show evidence of propagation and greater mean rainfall (Fig. 1), consistent
382 with mechanisms that favor organized convection. An early morning peak
383 is seen in Peru and Bolivia on the eastern side of the Andes.

384 The 25 km case has the simulated phase most similar to observations.
385 The phase over northern mountainous regions is best represented at coarser
386 resolutions, but precipitation is delayed over significant areas of the Amazon

387 basin. At finer resolutions, the diurnal peak over much of Amazonia is close
388 to local noon, roughly four hours too early. This behavior is similar to that
389 seen over southeastern CONUS in Fig. 3. Here it again suggests that the
390 parameterized convection acts to delay the diurnal peak.

391 **7. The effect of parameterized convection at 3.5 km**

392 To gain further insight into the role of the Grell-Freitas deep convection
393 at high resolution, we conduct two additional experiments with 3.5 km grid
394 spacing. In the first, denoted GF0, the Grell-Freitas parameterization is
395 simply disabled, and all deep convection is handled by resolved motions. In
396 the second experiment, denoted GF1, we disable the scale-aware function
397 in Grell-Freitas such that the full tendency of parameterized convection is
398 applied, even with 3.5 km grid spacing. These experiments can be viewed
399 as limiting cases, with the original 3.5 km case in between. The Park and
400 Bretherton (2009) shallow convection remains active in both cases.

401 Figure 13 shows the diurnal amplitude and phase for the three cases over
402 CONUS, with the strength of parameterized convection increasing from top
403 to bottom. With no parameterized deep convection (GF0), the diurnal
404 amplitude is larger over the southeast and central US, but reduced over the
405 Gulf of Mexico. As parameterized convection increases (3.5 km and GF1),
406 the amplitude field becomes smoother and generally more similar to the

407 TRMM climatology.

408 The GF0 phase plot indicates that convection develops too early in the
409 Southeast and West, while the nocturnal peak over the Great Plains occurs
410 too late. Both of these tendencies are reduced as the parameterized con-
411 vection increases. The implication from both phase and amplitude results
412 is that the exclusive use of explicit convection can be improved upon with
413 some degree of parameterized convection.

414 The results over the Maritime Continent are shown in Fig. 14. Disabling
415 Grell-Freitas completely (GF0) has little effect, with phase and amplitude
416 largely indistinguishable from the 3.5 km case. However, the GF1 case is
417 dramatically different. The precipitation pattern becomes more land-locked,
418 with reduced diurnal amplitudes over most ocean regions, and again, less
419 evidence of grid-scale storms. The phase shows somewhat earlier peaks
420 over land, and broader near-coastal regions with peak rainfall after solar
421 midnight (pink shading), rather than morning (blues).

422 Finally, Fig. 15 shows the phase and amplitude over Amazonia. Peak
423 amplitudes around Colombia and along the Andes are somewhat reduced
424 with increasing parameterized convection (GF1), and the continental in-
425 terior and ocean both show less grid-scale variability. The parameterized
426 convection has an effect along the northern coastline similar to that seen
427 around the Maritime Continent, with a broader band of peak rainfall after

428 midnight (pink shading). In the interior, the hints of propagating systems
429 evident with GF0 are mostly absent in GF1.

430 **8. The effect of microphysics at 3.5 km**

431 The choice of microphysics also plays a role in diurnal variability. While
432 the simulations discussed above all utilized the single-moment microphysics
433 scheme of (Bacmeister et al. 2006), a 3.5 km case using the GFDL micro-
434 physics scheme (based on Zhao and Carr, 1997, with significant modifica-
435 tions) was also examined. This simulation was the official GMAO submis-
436 sion to the DYAMOND intercomparison. The phase and amplitude over
437 CONUS are shown in the top left of Figs. 2 and 3. There is significant
438 further improvement in phase over the Great Plains, presumably benefiting
439 from more realistic convective organization. At the same time, precipita-
440 tion peaks even earlier over the Southeast, and, although reduced, hints of
441 excessive grid-scale precipitation still appear in the amplitude plot.

442 Over the Maritime Continent, Fig. 7 shows a significant reduction in
443 diurnal amplitude over both land and ocean regions when the GFDL mi-
444 crophysics is used, bringing the amplitudes generally closer to observations,
445 though now somewhat underestimated and even smaller than the 50 km
446 case. The phase, shown in Fig. 8, is also impacted by the microphysics.
447 Here, as over the southeastern CONUS, there is a widespread shift toward

448 earlier rainfall peaks. The shift occurs over both land and ocean, and gen-
449 erally pulls the model away from the observed timing.

450 Finally, over Amazonia, the diurnal amplitudes in Fig. 11 are somewhat
451 reduced. There is a notable reduction in the unrealistic amplitude along
452 the Andes, but also in the northern regions where amplitudes become un-
453 derestimated. In contrast to the microphysics' impact over CONUS, the
454 phase plot in Fig. 12 suggests a reduction in propagating systems, with
455 a relatively uniform late afternoon peak across most of the interior. The
456 nocturnal peaks in the North, while already limited in the 3.5 km case,
457 are further reduced with GFDL microphysics, as the late afternoon peaks
458 extend fully to the coastline in most regions.

459 9. Cloud clusters over CONUS

460 The spatial organization of convection and cloud cover also varies diur-
461 nally, and is expected to depend strongly on model resolution. It therefore
462 offers a complementary metric to the amplitude and phase of precipitation
463 analyzed above. In this section we examine the size distribution of convec-
464 tive cloud clusters and their diurnal variability over CONUS. Cloud clusters
465 are defined as contiguous regions of brightness temperature (T_b) less than
466 230 K, and area larger than 100 km^2 , contained within the CONUS domain.
467 Similar criteria have been used in previous work to consider generic statis-

468 tical properties of convection (e.g., Mapes and Houze 1993), although more
469 stringent criteria are typically applied in studies of mesoscale convective
470 systems (MCSs). The identified clusters are then binned by size to produce
471 histograms shown in Fig. 16.

472 With grid spacing of 25 km or 50 km, the model significantly under-
473 estimates the number of all clusters smaller than 10^4 km². Note that the
474 minimum area representable on a 50 km grid is roughly 2500 km², which
475 falls into the third bin, spanning 1000 km² to 3000 km². The number of
476 small clusters increases monotonically with resolution and ultimately ex-
477 ceeds the observations in the 6 km and 3.5 km cases. The number of larger
478 clusters (above 10^4 km²) varies less with model resolution, and generally
479 agrees with observations within a factor of two.

480 We also consider the intensity of precipitation within convective clusters.
481 A given cluster's intensity is defined as the instantaneous mean precipitation
482 rate within the 230 K T_b contour. An observational estimate is created
483 by first re-gridding the 0.1 degree IMERG dataset to the 4 km Merged
484 IR grid, and calculating intensities following the procedure above. The
485 intensities are averaged across all clusters within each size bin and shown in
486 the bottom panel of Fig. 16. The low resolution GEOS cases underestimate
487 precipitation intensity for all size bins, but intensity increases monotonically
488 with resolution, such that the 3.5 km GEOS case is comparable to the 4 km

489 observations.

490 Figure 17 shows the diurnal cycle of cluster number for each area bin,
491 normalized by the daily mean for each case. The observed size distributions
492 exhibit a pronounced diurnal cycle, with peak numbers in the late afternoon
493 and early evening. The peak for the largest clusters is delayed by 1-2 hours
494 relative to the smaller clusters, suggestive of a lifecycle effect of upscale
495 convective growth, as isolated deep convection transitions into organized
496 mesoscale systems. In the 25 km and 12 km GEOS cases, the diurnal cycle
497 is relatively muted, particularly for the smaller clusters. The amplitude of
498 diurnal variation is more realistic in the 6 km and 3.5 km cases, generally
499 comparable to observations, although the smallest clusters are too numerous
500 during the early day, and their late evening peak is underestimated.

501 Also included in Fig. 16 are the GF0 and GF1 3.5 km experiments.
502 When Grell-Freitas convection is disabled (GF0), there is little impact on
503 the cluster size distribution. However, the precipitation intensity curve
504 significantly overshoots the observations for clusters smaller than 10^4 km².
505 On the other hand, when Grell-Freitas is allowed to run at full strength
506 (GF1), there is a further increase in the number of small clusters over the
507 observed counts, exacerbating the 3.5 km bias. The precipitation intensities
508 with GF1 are dramatically reduced, similar to those of the 50 km case.
509 Overall, the scale-aware function in the Grell-Freitas scheme seems to allow

510 a more optimal balance between parameterized convection and the resolved
511 dynamics.

512 The distribution of precipitation intensity over CONUS is compared
513 with IMERG in Fig. 18, using hourly mean model and IMERG data inter-
514 polated to a common 0.5 degree grid for consistency. The model generally
515 overestimates light precipitation, under 5 mm day^{-1} . Simulation of heavier
516 precipitation depends strongly on resolution, with the 25 km and 50 km
517 cases producing more at moderate rates (10 to 100 mm day^{-1}), and higher
518 resolutions producing more above 100 mm day^{-1} . An inflection point is
519 seen around 80 mm day^{-1} in the 3-12 km curves, likely associated with
520 the reduced parameterized convection in those cases, which allows more
521 convective precipitation from resolved updrafts. The GF0 and GF1 exper-
522 iments (dashed curves in Fig. 18), show that a 3.5 km run with increased
523 parameterized convection looks similar to the 25 km and 50 km cases, while
524 a 3.5 km run with no parameterized convection has significantly stronger
525 precipitation rates, though it is not necessarily a better match to IMERG.

526 **10. Summary and conclusions**

527 We have evaluated the diurnal cycle of precipitation in a set of non-
528 hydrostatic AGCM simulations with nominal grid spacing ranging from 3.5
529 km to 50 km. Finer resolution is often expected to improve representation of

530 diurnal variability by reducing reliance on sub-grid parameterizations that
531 introduce uncertainty into model formulation. While we do find that some
532 aspects of the diurnal cycle improve with resolution, these improvements are
533 partially offset by degradations in other areas. The results emphasize the
534 complicated and regional nature of the diurnal cycle and the many physical
535 mechanisms that govern it.

536 In general, we find that amplitudes of the diurnal harmonic appear more
537 similar to the observed multi-year August climatology in the low resolution
538 cases, while the 3.5 km and 6 km cases appear to suffer from excessive
539 small-scale variability. This overproduction of strong small-scale storms
540 has been reported in other studies with explicit convection (Kendon et al.
541 2012; Hanley et al. 2019), and can be made worse when parameterized
542 convection is removed entirely (Pearson et al. 2014). We find that resolution
543 has no consistent impact on the regional-scale amplitudes, with some regions
544 showing larger amplitude at high resolution (e.g., the western United States
545 and Maritime Continent), and other regions at low resolution (e.g., the
546 southern Amazon).

547 Over regions where the diurnal cycle is dominated by local thermody-
548 namic forcing, such as over the southeastern United States, precipitation in
549 the higher resolution cases generally peaks several hours earlier than with
550 low resolution, and typically earlier than observations. The delay at low

551 resolution is primarily due to the Grell-Freitas parameterized convection,
552 which employs the non-equilibrium closure of Bechtold et al. 2014. When
553 this closure is disabled, the low resolution precipitation peaks even earlier
554 than the high resolution cases.

555 Higher resolution generally offers improvement in regions where diur-
556 nal variability is dependent on organized propagating convection. Over
557 CONUS, more realistic mesoscale organization enables eastward propagat-
558 ing systems that produce a more realistic nocturnal precipitation peak over
559 the Great Plains, which is largely missing at low resolution. This is con-
560 sistent with previous studies using regional models over CONUS (e.g., Gao
561 et al. 2017). However, the improvement is not global, or monotonic with
562 resolution. For example, the intensity of propagating rainfall offshore of
563 Sumatra is arguably best in the 6 km simulation (Fig. 10).

564 We also examined the statistics of convective cloud clusters, identified
565 using a brightness temperature threshold, and their dependence on resolu-
566 tion. We find that at high resolution the intensity of precipitation varies
567 more realistically with convective cluster size, and the diurnal cycle of cloud
568 cluster number better matches observations. On the other hand, while cloud
569 cluster size histograms indicate that coarse resolutions are unable to repre-
570 sent the smallest clusters, the relative number of small cloud clusters be-
571 comes overestimated when model grid spacing drops below 12 km. Unlike

572 the excessive small-scale precipitation noted above, this bias actually grows
573 worse with stronger parameterized convection, and instead may be related
574 to issues of upscale convective growth, discussed below.

575 The role of the Grell-Freitas (GF) parameterization at 3.5 km grid spac-
576 ing was explicitly examined in “mechanism-denial” experiments, in which
577 the parameterization was either switched off entirely (GF0), or fully en-
578 abled by removing its scale-aware throttling function (GF1). The analysis
579 shows that even at 3.5 km, GF produces a diurnal cycle amplitude and
580 phase quite similar to that of the 50 km case. This implies, as argued by
581 Pearson et al. (2014), that differences between the low and high resolution
582 cases are largely driven by scaling of the parameterized convection, rather
583 than changes in resolution itself. The precipitation intensity as a function of
584 cluster size in GF1 similarly resembles the 50 km case, while simultaneously
585 worsening the size distribution bias toward small cloud clusters.

586 These results highlight the continued importance of model formulation
587 even at convection-permitting resolutions. Some deficiencies in the 3.5 km
588 case, such as the too early diurnal peaks in locally forced regimes, and over-
589 estimated small-scale variability, may be associated with insufficient inhi-
590 bition of resolved updrafts. Many convection-permitting models include an
591 explicit parameterization of horizontal subgrid mixing (e.g., a Smagorinsky-
592 Lilly scheme) that contributes to the dilution of buoyant air in resolved

593 updrafts (Kendon et al. 2012; Hanley et al. 2019), and indeed, diurnal
594 variability can be sensitive to its formulation (Pearson et al. 2014). How-
595 ever, the GEOS model does not currently include any such mixing, and as a
596 consequence explicit convection may be unrealistically vigorous and insen-
597 sitive to environmental conditions. Incorporating a subgrid mixing scheme
598 should be analogous to increasing the entrainment rate with parameter-
599 ized convection (Bechtold et al. 2004), with potentially similar impacts on
600 diurnal timing.

601 Other issues, such as the over-estimated number of small cloud clusters,
602 and weak propagation relative to observations, may be associated with in-
603 sufficient upscale convective growth. A number of errors may contribute to
604 insufficient growth, including the simulated convective environment (such
605 as inadequate CAPE, shear or moisture), misrepresentation of cold pools,
606 or microphysical issues (Coniglio et al. 2010; Thielen and Jr. 2019). Cold
607 pools, generated by hydrometeor loading and evaporative cooling, are an
608 integral part of mesoscale convective systems, and may aid more generally
609 in diurnal transitions from shallow to deep convection (Khairoutdinov and
610 Randall 2006; Schlemmer and Hohenegger 2014). An analysis of cold pool
611 statistics in these simulations would be a valuable future study, both as a
612 factor in upscale convective development and as an indicator of problems
613 with microphysics.

614 While the 3.5 km case with GFDL microphysics does show more realistic
615 eastward propagation over the Great Plains (Fig. 3), the improvement does
616 not extend to the Maritime Continent or Amazonia. That case produces
617 unrealistically small diurnal amplitudes in all three regions, as well as early
618 timing in locally forced regimes, consistent with the hypothesis that that
619 problem is due to insufficient subgrid mixing or another non-microphysical
620 issue. Future research and model development with GEOS will explore these
621 and other issues to achieve a realistic balance of variability and intensity in
622 convective regimes.

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629 Public access to the model source code is available at [github.com/GEOS-](https://github.com/GEOS-ESM/GEOSgcm)
630 [ESM/GEOSgcm](https://github.com/GEOS-ESM/GEOSgcm) on tag Jason-3_0.

631 References

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Fig. 18

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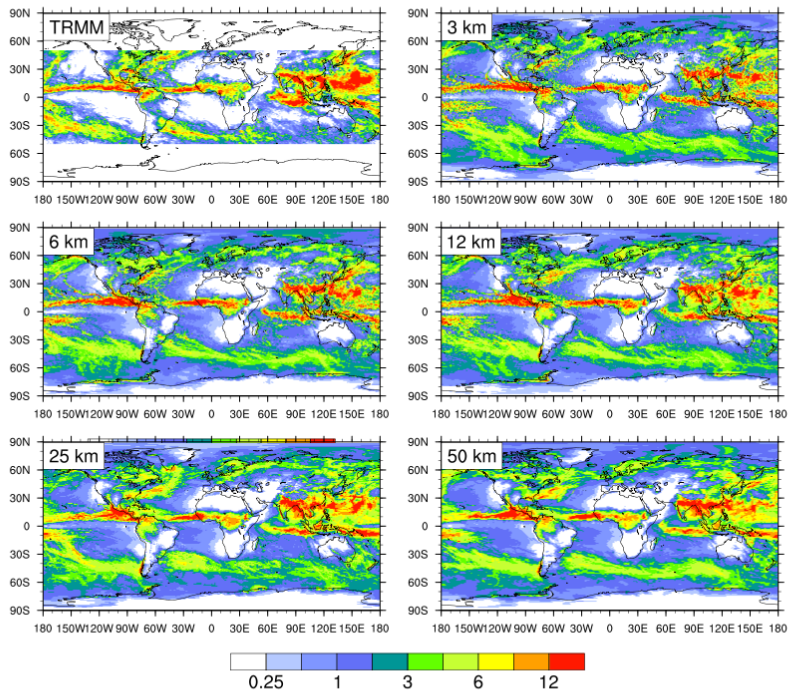


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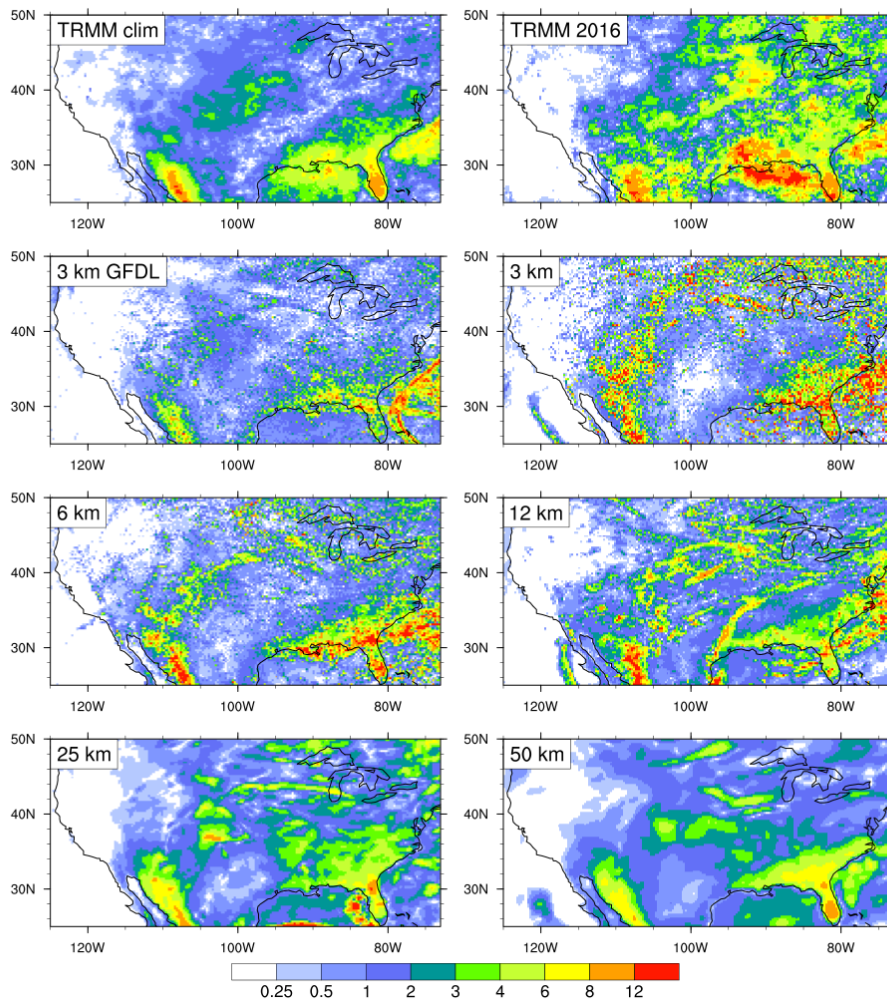


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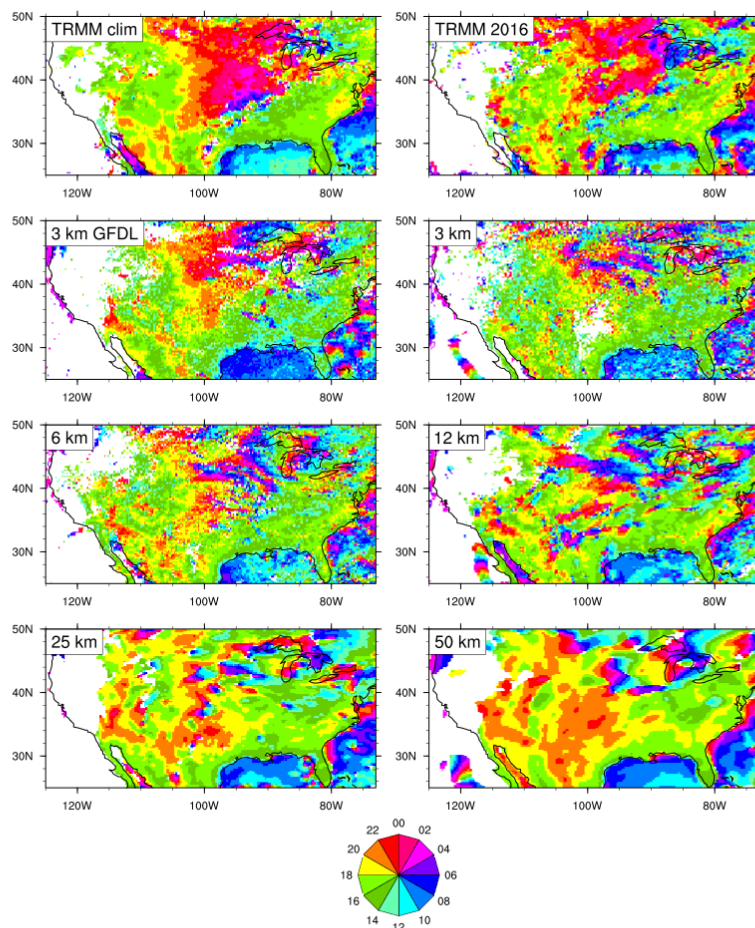


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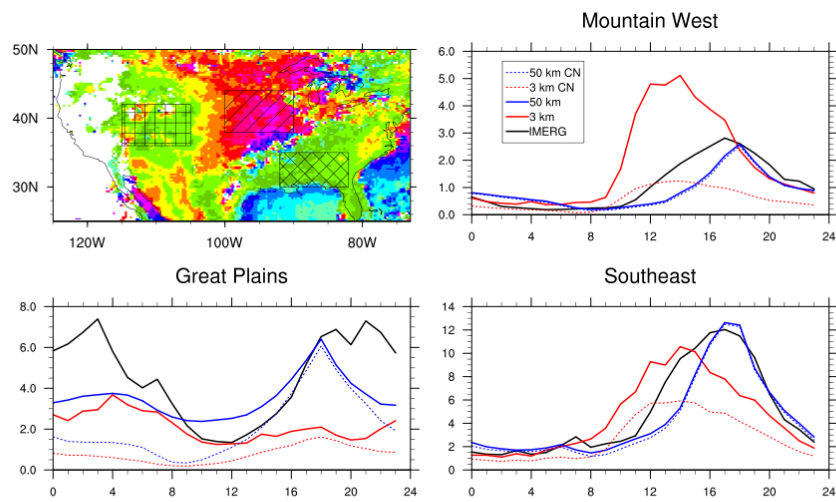


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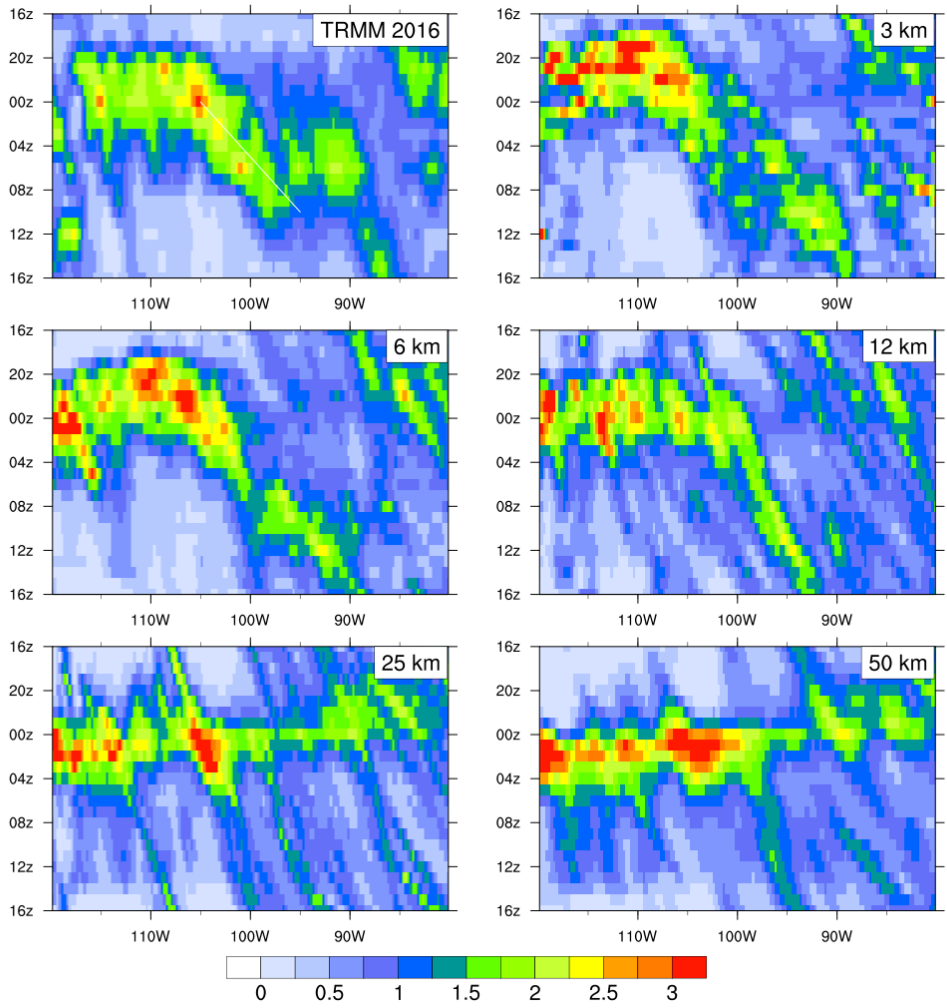


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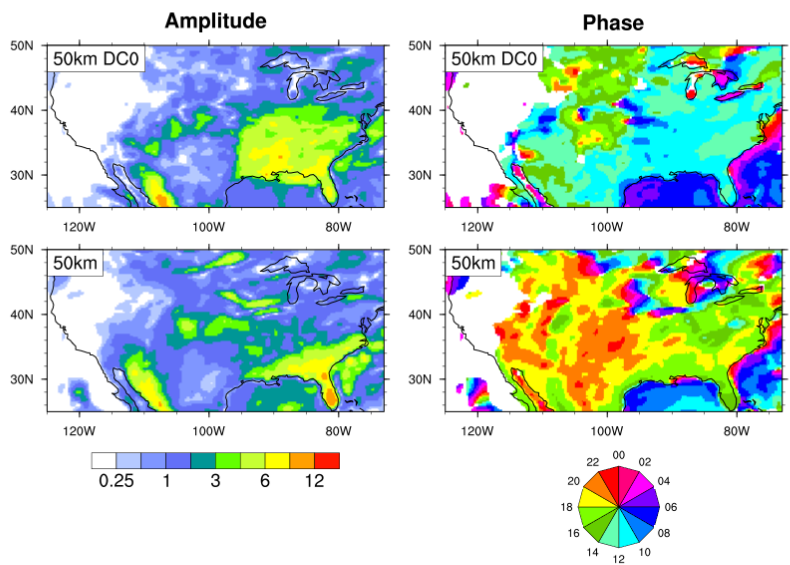


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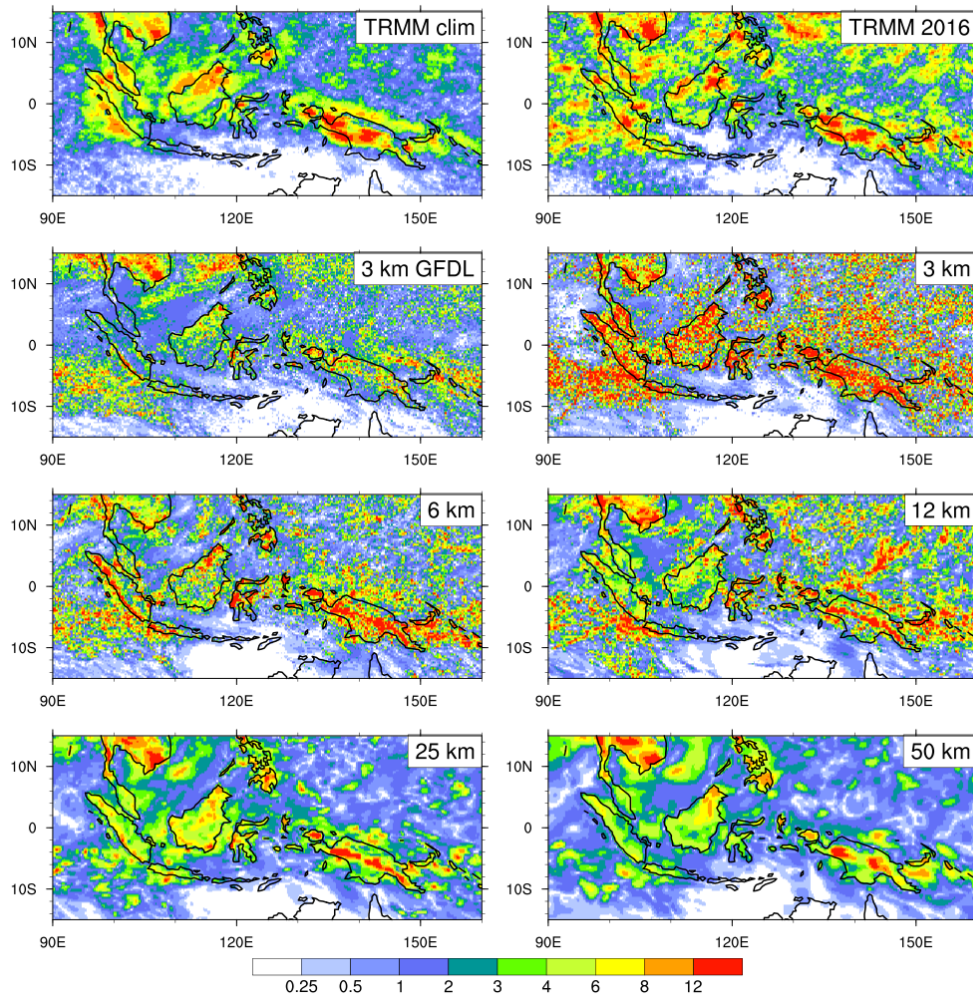


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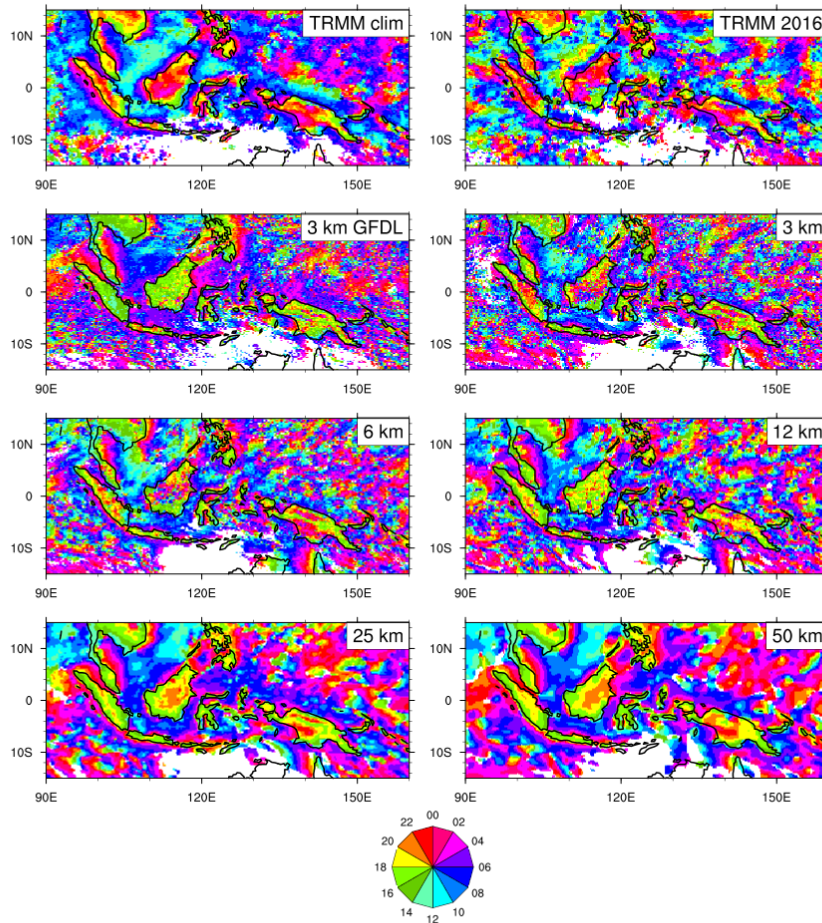


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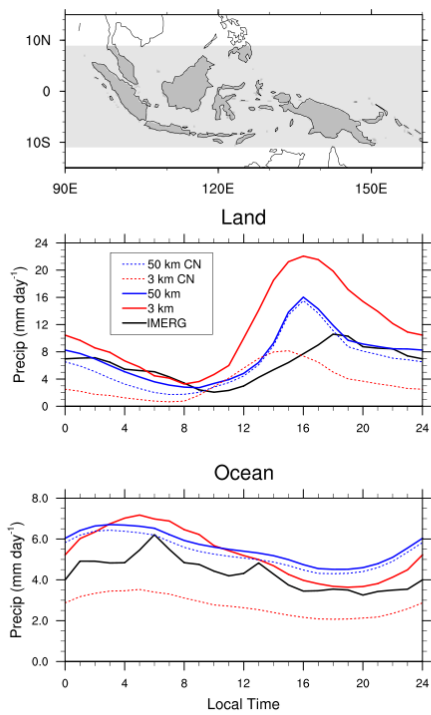


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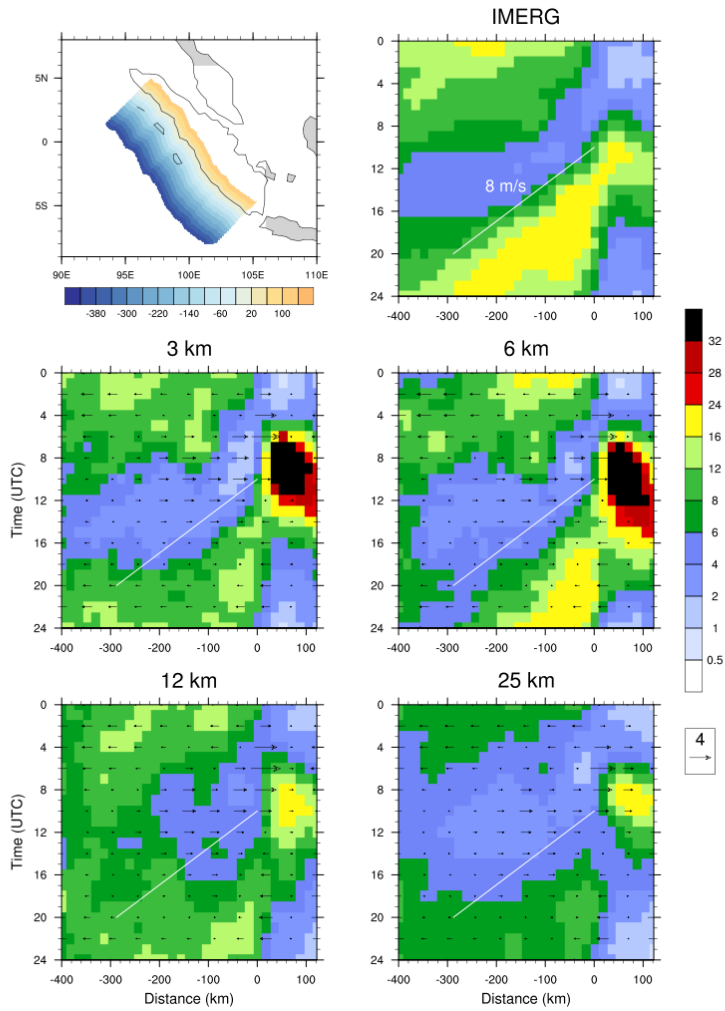


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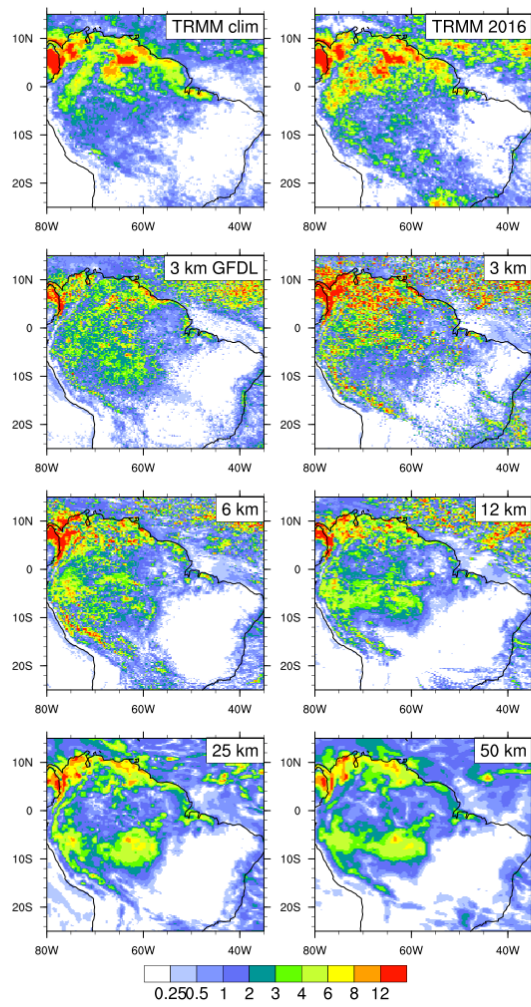


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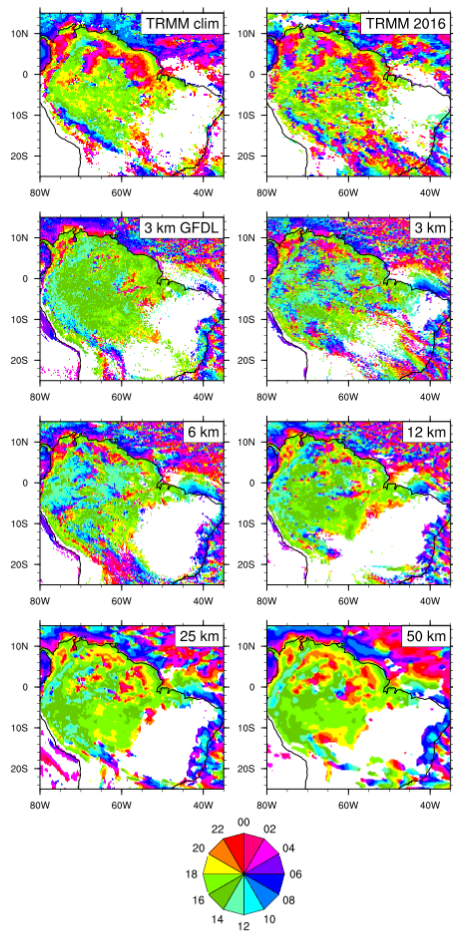


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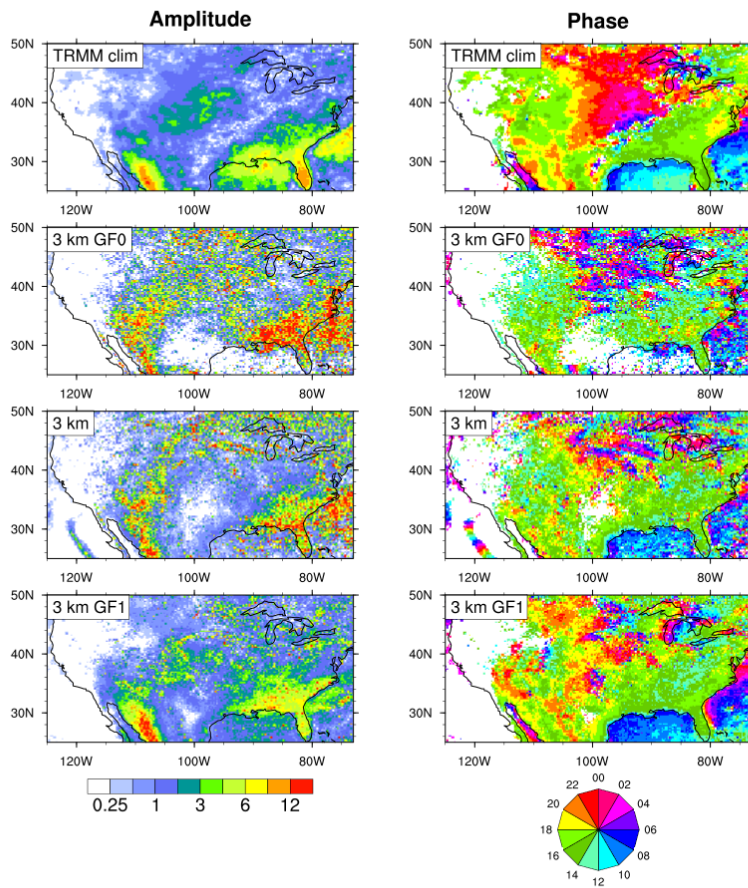


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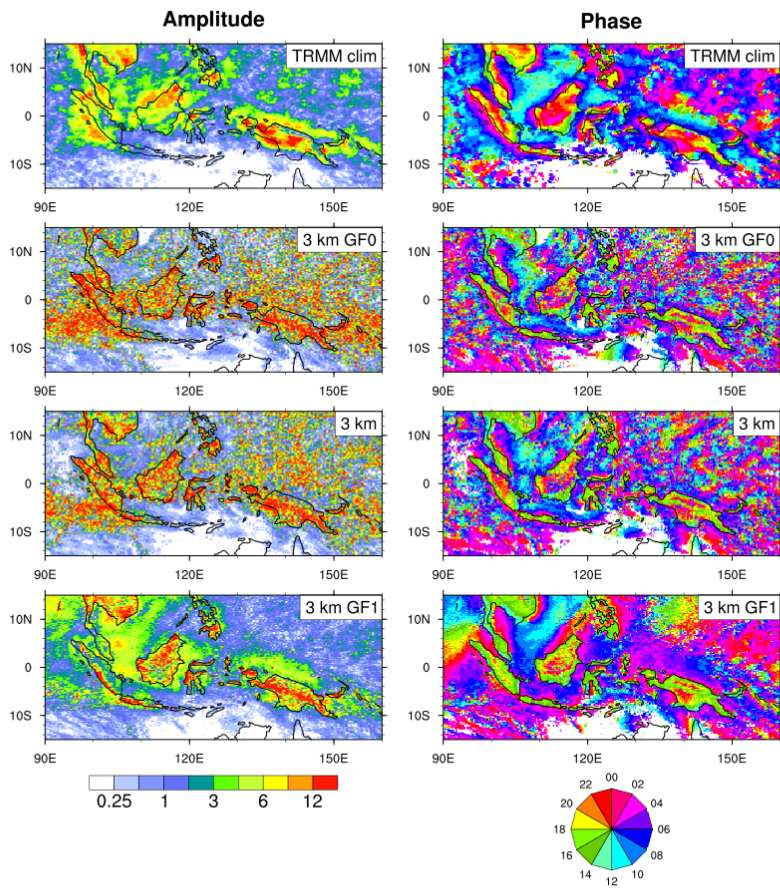


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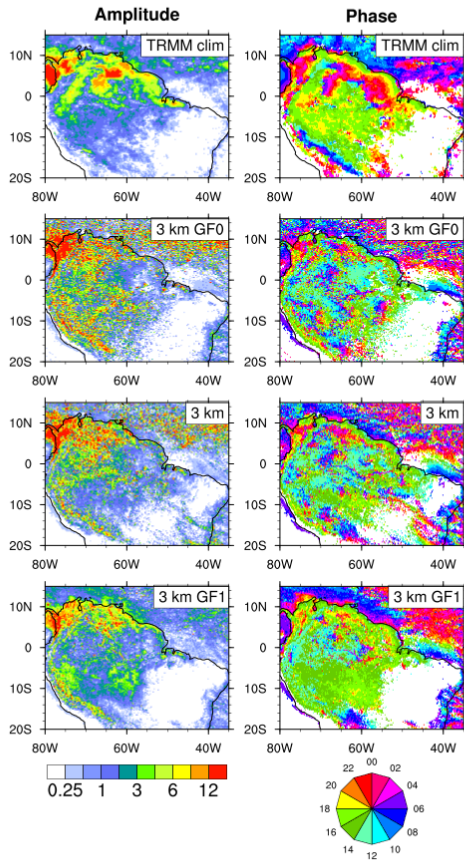


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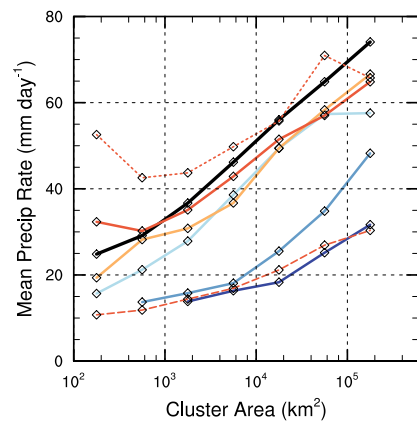
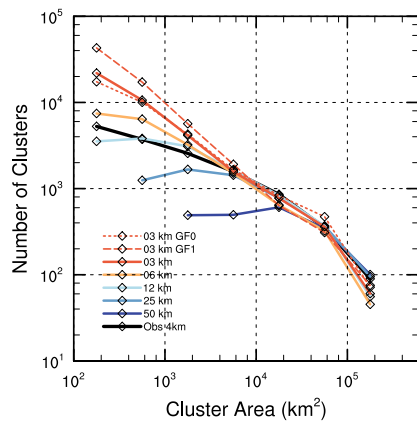


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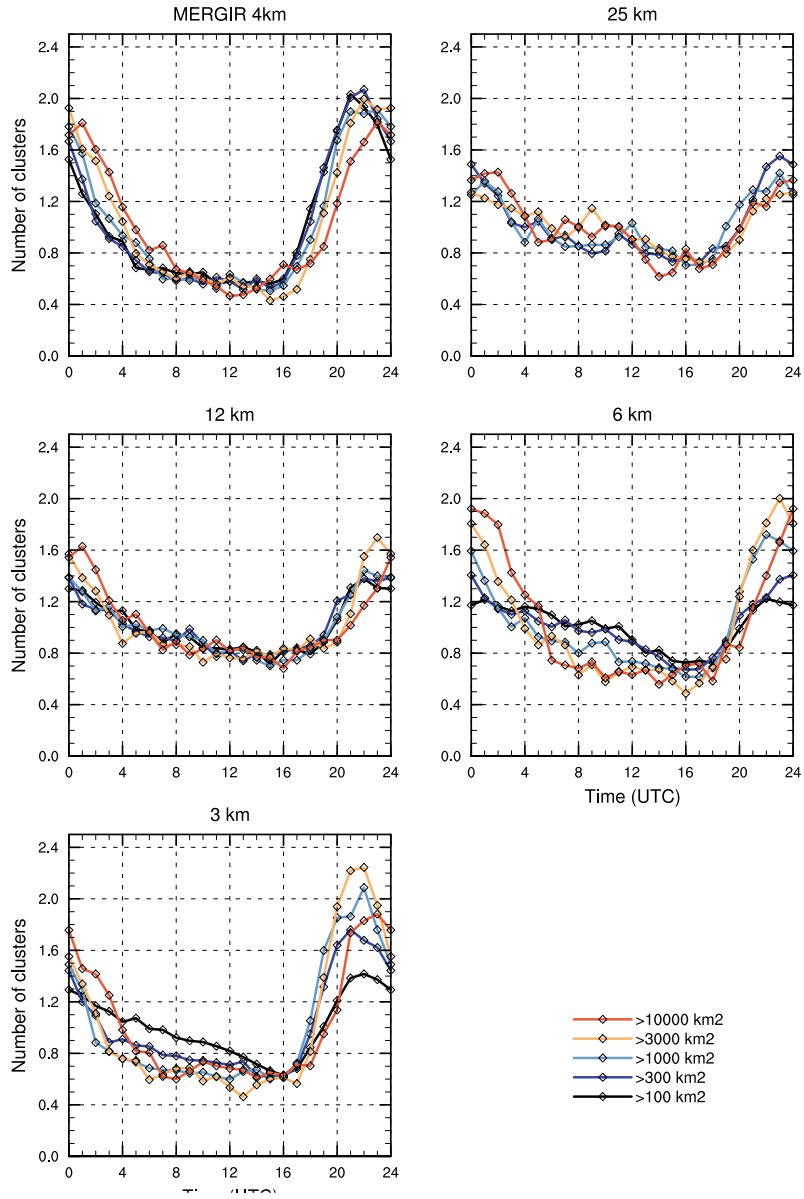


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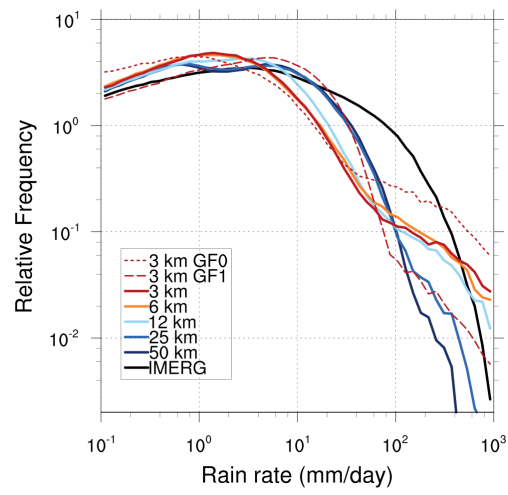


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Table 1. Mean precipitation (50°S-50°N) and outgoing longwave radiation (OLR) across resolutions.

	Total Precip (mm d ⁻¹)	Convective Precip (mm d ⁻¹)	OLR (W m ⁻²)
TRMM/CERES	3.05	-	244.4
GEOS 50 km	3.28	2.27	244.0
GEOS 25 km	3.30	2.18	243.6
GEOS 12 km	3.31	1.60	245.5
GEOS 6 km	3.30	1.28	247.0
GEOS 3 km	3.27	1.17	245.8