

1 Assessing possible dynamical effects of condensate in
2 high resolution climate simulations

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3 In areas of heavy precipitation, condensed water species can add signif-
4 icant mass to an atmospheric column. This mass can create positive pres-
5 sure anomalies of up to several hPa at the surface. This pressure is expected
6 to force a divergent component in the low-level flow that may have an im-
7 pact on the evolution of the precipitating system. In this study we examine
8 results from a cloud resolving model simulation of tropical convection to es-
9 timate the pressure induced by condensates. A simple parameterization of
10 this condensate loading as a function of surface rain rate is derived and im-
11 plemented in the National Center for Atmospheric Research's Community
12 Atmosphere Model version 5 (CAM5). Our results suggest that at horizon-
13 tal resolutions of 25 km condensate loading is an important factor in con-
14 trolling the frequency of intense rain rates in the model.

1. Introduction

15 The contribution of condensed water species to atmospheric mass has long been known
16 to be a significant factor in the dynamics of moist convection [e.g. Emanuel, 1994; Xu and
17 Randall, 2001]. The weight of condensates in convecting parcels can have a major impact
18 on their buoyancy and may be a dominant control on the global statistics of convection
19 [Emanuel, 1994].

20 The major contribution to condensate mass comes from precipitating species such as
21 rain, hail, snow, and graupel. Microphysics schemes for models at climate resolutions
22 typically use diagnostic rather than prognostic treatments for precipitating condensate.
23 Climate models correctly account for the removal of mass by precipitation in the models'
24 mass budget. However, the diagnostic treatments in climate models view this removal as
25 occurring instantaneously. In reality precipitating condensate may exist in a deep column
26 that persists for a significant time, i.e., comparable to a typical climate model time step of
27 15 to 30 minutes. The weight of this column contributes to the pressure field (condensate
28 loading, henceforth abbreviated CL) and can have direct dynamical effects on the flow.
29 The dynamical effects of CL are not currently present in most climate models that use
30 diagnostic treatments of precipitation.

31 Neglect of CL may be justified at the horizontal resolutions of 100's of km. However,
32 as climate model resolution increases we believe this neglect is no longer justified. This
33 study will assess the potential impact of CL at 25 km resolutions by quantifying the
34 condensate contribution to the pressure field in much finer cloud resolving simulations
35 using the National Center for Atmospheric Research's (NCAR's) Weather Research and

36 Forecasting model (WRF) [Skamarock and Klemp, 2008]. A simple parameterization of
37 this pressure is developed and implemented in NCAR’s Community Atmosphere Model
38 version 5 (CAM5) [Neale et al., 2010]. Our results suggest that horizontal resolutions
39 of 25 km and finer require some representation of CL. This resolution range is already
40 accessible to global climate simulations, and will likely become the default for leading
41 edge simulations in the next ten years. In passing it will be shown that at 25 km and
42 even at 5 km resolutions, CL effects are significantly more important than nonhydrostatic
43 effects.

44 The paper is structured as follows: In Section 2, we describe the models used in this
45 study. In Section 3 we analyze cloud-resolving model (CRM) results and describe a simple
46 parameterization of CL effects based on surface precipitation rates. In Section 4 we show
47 results from CAM5 including this parameterized CL. In Section 5 we summarize and
48 discuss our results.

2. Models and Experimental Setup

2.1. CAM5

49 The Community Atmosphere Model version 5 is a state of the art global climate model.
50 Major differences from earlier versions of CAM include a new 2-moment, 2-phase prognos-
51 tic cloud condensate scheme ,advanced boundary layer and shallow convection schemes
52 and deep convection with enhanced plume entrainment and momentum transport . Com-
53 plete documentation of CAM5 is provided in [Neale et al., 2010]. In this study we use
54 the finite-volume (FV) dynamical core with a horizontal resolution of $0.23^{\circ}\text{lat}\times 0.31^{\circ}\text{lon}$
55 and 30 layers in the vertical. A physics time-step of 15 minutes is used. We will examine

56 results from experiments forced by observed sea-surface temperatures (SST) initialized on
57 Jan 1 2005. The experiments ran for 18 months, but in this study we will examine results
58 from the first 13 months only. Currently CAM5 does not incorporate any condensed water
59 species in its atmospheric mass field.

2.2. WRF

60 The Weather Research and Forecasting model is a well established nonhydrostatic dy-
61 namical model with flexible nesting capability [Skamarock and Klemp, 2008]. WRF solves
62 the full Euler equations on a dry mass vertical coordinate. Prognostic equations for pres-
63 sure, 3D velocity, heat, and water species are included. In this study we will examine
64 results from the innermost domain of triply-nested simulation initialized by ERA interim
65 reanalysis. The simulation period covers February 22 through 27 2005. The innermost
66 domain has a resolution of 500 m in both zonal and meridional directions with a size of
67 1000×800 gridpoints. It is chosen to overlap with the TOGA-COARE domain; 112°E to
68 117°E and 5°S to 1°S . Fifty vertical levels are used with a top close to 25 km, and Δz
69 ranging from 50 m close the surface to 500 m in the midtroposphere. A 2 second time
70 step is used and data are saved every 15 minutes.

71 WRF offers a large number of options for parameterizing physical processes, includ-
72 ing cloud microphysics. The experiment examined here used the Hong and Lim [2006]
73 microphysics option. This is a 6-category bulk scheme that incorporates graupel rather
74 than hail, as appropriate for tropical, oceanic convection. The innermost domain in the
75 simulation discussed here did not employ a deep convection parameterization.

3. Development of condensate loading (CL) parameterization

3.1. Preliminary analysis of WRF results

WRF uses a complete prognostic pressure equation, which includes nonhydrostatic effects as well as a complete atmospheric mass field, including the contribution of all condensed water species. Our assessment of the potential CL effect at high climate resolutions (25 km) involves two steps. First, the CRM fields are coarse grained to 25 km resolution by averaging over 50×50 gridpoint subdomains. Second, coarse-grained profiles of potential temperature $\bar{\theta}$ (K), water vapor mixing ratio with respect to dry air \bar{q}_w (kg kg^{-1}), and condensed water mass mixing ratios with respect to dry air $\bar{q}_{[l,i,r,s,g]}$ (kg kg^{-1}), where the subscripts l , i , r , s , and g refer to cloud liquid, cloud ice, rain, snow, and graupel, are used to calculate diagnostic hydrostatic pressure fields with and without CL

$$\bar{\pi}_{hyd,[v,c]}(x, y, z, t) = \bar{\pi}_{top} + \int_z^{z_{top}} \frac{g}{c_p \bar{\theta}_{[v,c]}(x, y, z', t)} dz', \quad (1)$$

where g is gravitational constant (9.81 m s^{-2}), c_p is specific heat capacity of dry air (1003 J kg^{-1}). The Exner pressure π_{hyd} and the pressure are related by

$$\bar{p}_{hyd} = p_{00} (\bar{\pi}_{hyd})^{c_p/R}, \quad (2)$$

where R is gas constant for dry air (286 J kg^{-1}), and p_{00} is a reference pressure of 1000 hPa. We use two versions of thermodynamic variable $\bar{\theta}_{[v,c]}$ in the hydrostatic integral (1)

$$\bar{\theta}_v = \bar{\theta} \frac{1 + \bar{q}_w/\epsilon}{1 + \bar{q}_w} \quad (3)$$

$$\bar{\theta}_c = \bar{\theta} \frac{1 + \bar{q}_w/\epsilon}{1 + \bar{q}_w + \bar{q}_l + \bar{q}_i + \bar{q}_r + \bar{q}_s + \bar{q}_g} \quad (4)$$

[e.g. Emanuel, 1994, (4.3.6)]. With $\bar{\theta}_v$ the diagnostic hydrostatic pressure $\bar{p}_{hyd,v}$ obtained from (1) will include virtual effects but not CL. This quantity is the pressure variable used

95 in most state-of-the-art climate models. Using $\bar{\theta}_c$ in (1) yields an approximate hydrostatic
96 pressure including both virtual effects and CL.

97 Figure 1 shows joint frequency distributions (JFDs) of $\bar{p}_{hyd,c}$ and $\bar{p}_{hyd,v}$ at the surface
98 versus the coarse grained prognostic pressure from WRF \bar{p} . Fig. 1b shows that ignoring
99 CL even at 25 km resolutions leads to frequent, large surface pressure departures from the
100 WRF value. Underestimates of several hPa are common. A clear implication of this result
101 is that high-resolution climate model surface pressures in regions of strong precipitation
102 may be systematically underestimated by several hPa. In the tropics, pressure anomalies
103 of this size may be dynamically-significant. The CL effect on pressure should act against
104 low-level convergence, and should therefore weaken CISK(Conditional Instability of the
105 Second Kind)-interactions between moist heating and flow in the boundary layer.

106 The close agreement between $\bar{p}_{hyd,c}$ and \bar{p} in Fig. 1a implies that nonhydrostatic dynam-
107 ics are unimportant at the 25 km scale. So, while small, intense, nonhydrostatic updrafts
108 may be critical in determining vertical fluxes, their detailed structure has negligible im-
109 pact on the pressure field at scales of 25 km. Based on this analysis there is no reason
110 to suspect that explicitly-resolved convection in a model with 25 km resolution would
111 be inherently “pathological”. We repeated this analysis using a coarse-graining scale of
112 5 km (10×10 WRF points). Results are summarized in Table 1. As expected, the root
113 mean square (RMS) difference between $\bar{p}_{hyd,c}$ and \bar{p} is larger than for the 25 km scale.
114 Nevertheless, even at 5 km the difference between $\bar{p}_{hyd,v}$ and \bar{p} is still much larger than
115 that between $\bar{p}_{hyd,c}$ and \bar{p} , suggesting that CL remains more critical at 5 km resolution
116 than nonhydrostatic dynamics.

3.2. Implementation of CL parameterization in CAM5

117 In order to quickly assess potential CL impacts, we designed a simple parameterization
 118 for CAM5 based on surface precipitation rates. Figure 2 shows a JFD of CL-induced
 119 surface pressure, i.e., $p'_{CL} \equiv \bar{p}_{hyd,c} - \bar{p}_{hyd,v}$ at $z=0$, and 15-minute average, surface rain rate
 120 \mathcal{R}_{sfc} from our WRF simulation. The plot shows that a reasonably-compact, relationship
 121 exists between these variables. The additional hydrostatic pressure induced by CL at any
 122 height z is given by

$$123 \quad p'_{CL}(x, y, z, t) = \int_z^{\infty} g \rho_c(x, y, z', t) \, dz' \quad (5)$$

124 where ρ_c is the density of all condensate in the atmosphere. Examination of condensate
 125 density profiles from our WRF simulations binned by \mathcal{R}_{sfc} (supplemental Figure 1) sug-
 126 gests ρ_c is reasonably constant from the surface to around 5000 m and then begins to drop
 127 off rapidly somewhere between 5000 and 10000 m. This general shape seems to hold for
 128 moderate to intense \mathcal{R}_{sfc} (100 to 1000 mm d⁻¹).

129 As a crude first approximation we set ρ_c to a constant value ρ_{c0} between the surface
 130 and a height H_{CL} and set $\rho_c=0$ above. The density ρ_{c0} is then specified as a function of
 131 \mathcal{R}_{sfc} and a terminal velocity w_f ,

$$132 \quad \rho_{c0} = \rho_{L0} \frac{\mathcal{R}_{sfc}}{w_f}. \quad (6)$$

133 where ρ_{L0} is the density of liquid water (1000 kg m⁻³) and \mathcal{R}_{sfc} is expressed in units of
 134 m s⁻¹. Combining (5) and (6) and incorporating our assumptions about the shape of
 135 the condensate profile we obtain an expression for the time-varying, fully-3D, hydrostatic

136 pressure perturbation induced by CL

$$137 \quad p'_{CL}(x, y, z, t) = \begin{cases} g\rho_{L0} \frac{\mathcal{R}_{sfc}}{w_f} \times (H_{CL} - z) & z \leq H_{CL}, \\ 0 & z > H_{CL}. \end{cases} \quad (7)$$

138 For \mathcal{R}_{sfc} we use the instantaneous CAM5 total surface precipitation (convective+large-
139 scale) at each time step. We simply use the hydrostatically-determined heights of the
140 CAM5 half-levels or layer edges to define the condensate column. When the upper-edge
141 of a layer falls below H_{CL} it is included in the column, otherwise it is left out. This can
142 lead to some variation in the actual thickness of the condensate layers.

143 The condensate pressure p'_{CL} is added to the dynamical pressure in the FV dynamical
144 core immediately before horizontal pressure gradient forces are calculated. In the present
145 implementation p'_{CL} has no other effects in the simulation, so that its horizontal gradient
146 can simply be regarded as another parameterized body force similar to gravity wave drag.

147 We tried 2 different forms for p'_{CL} (Table 2) whose surface signatures are shown by the
148 white lines in Fig. 2. These two experiments are intended to explore the sensitivity of
149 the model to the depth of CL while maintaining the CL pressure signature at the surface
150 approximately constant. Clearly, CL1 with $H_{CL} \approx 8500\text{m}$ is closer to the WRF condensate
151 profiles (supplemental Fig. 1) than is CL2 with $H_{CL} \approx 2000\text{m}$. However, it should be kept
152 in mind that these profiles are from a single 5-day period dominated by deep convection.
153 Furthermore, as will be seen below CL2 reveals interesting sensitivities to H_{CL} .

4. CAM5 results

154 Figure 3a shows probability density functions (PDFs) of instantaneous precipitation
155 intensity (30°S-30°N) in our CAM5 experiments, accumulated during August 2005 from
156 data written every 3 hours. The PDF from the CAM5 control (CTR) is shown in black.

157 The observational PDF for precipitation estimated from the Tropical Rainfall Measuring
158 Mission (TRMM) 3B42 product [Huffman et al., 2007] is also shown (dashed red). CTR
159 clearly overestimates the likelihood of precipitation rates greater than 200 mm d⁻¹ with
160 respect to TRMM-3B42. There is some uncertainty about whether the TRMM-3B42
161 precipitation rates represent instantaneous values or longer three hour averages. In any
162 case, there is minor impact on the model PDFs in Figure 3a when three hour average
163 precipitation rates are used (see supplemental Figure 2). Boyle and Klein [2010] note that
164 excessive extreme precipitation becomes more pronounced in CAM as resolution increases.
165 We also note that observations of intense precipitation frequency are likely to depend on
166 the area sampled, with smaller sample area yielding more frequent intense events.

167 With parameterized pressure gradient forces from CL, the frequency of intense precip-
168 itation rates ($\mathcal{R}_{sfc} > 200$ mm d⁻¹) is dramatically reduced. In CL1 only a small excess
169 with respect to TRMM at these rates remains (Fig. 3a, green curve). The result for
170 CL2 (magenta curve) is similar to CL1 for $\mathcal{R}_{sfc} < 1000$ mm d⁻¹, but at higher rates CL2
171 is less effective at reducing occurrence probabilities. This suggests that a deep pressure
172 perturbation is better at suppressing these wildly extreme events. In all cases, the effect
173 of CL is remarkably well targeted at reducing the frequency of intense precipitation. At
174 rates below 100 mm d⁻¹ little effect from CL is noticeable.

175 Figure 3b shows PDFs of vertical motion near 850 hPa (ω_{850}) over tropical ocean.
176 There is a clear connection between large precipitation rates and integrated low-level
177 convergence, indicated by $\omega_{850} < 0$. The control simulation exhibits a pronounced skew
178 towards strong convergence events. This skew is significantly reduced in CL1. In CL2

179 moderate convergence (-4000 to -2000 hPa d^{-1}) is noticeably more suppressed than in
180 CL1, but strong convergence (<-6000 hPa d^{-1}) is almost as frequent as in CTR again
181 suggesting a connection between large H_{CL} and suppression of extremes.

182 Figure 4 shows 12-month mean precipitation from CAM5 compared with observational
183 estimates from the Global Precipitation Climatology Project [GPCP, Adler et al., 2003].
184 All CAM5 experiments exhibit positive precipitation biases in the Pacific intertropical
185 convergence zone (ITCZ) with respect to GPCP. Modest improvements over CTR (Fig.
186 4a) are evident in CL1 (Fig. 4b) particularly south of the Equator where the model's
187 double ITCZ bias has been reduced. Interestingly, in CL2 (Fig 4c) clearer improvements
188 in mean precipitation are seen, with peak values dropping by around 3 mm d^{-1} over
189 much of the northern ITCZ. This suggests that the suppression of moderate low-level
190 convergence seen in Fig. 3b may be more significant in determining some aspects of mean
191 model climate than the suppression of extremes.

5. Summary and Discussion

192 We have shown that a simple but plausible parameterization of condensate loading (CL)
193 has appreciable impacts on simulations with the Community Atmosphere Model version
194 5 (CAM5) at a horizontal resolution of $0.23^{\circ}\text{lat}\times 0.31^{\circ}\text{lon}$. Our parameterization assumes
195 a one-to-one relationship between instantaneous surface precipitation rates and the total
196 mass of condensates in the atmospheric column above. The condensates are assumed
197 to have constant density in a layer of specified thickness H_{CL} and to fall with terminal
198 velocity w_f . Three CAM5 experiments with CL were performed using values of w_f and
199 H_{CL} given in Table 2. The surface signatures of the resulting condensate pressure are

200 compared with those from a 5-day cloud resolving simulation of tropical convection using
201 the Weather Research and Forecasting model (WRF) in Figure 2.

202 The best overall fit to the WRF results (CL1) yields significant reductions in the fre-
203 quency of intense precipitation ($\mathcal{R}_{sfc} > 200$ mm d⁻¹) and intense low-level convergence
204 ($\omega_{850} < -6000$ hPa d⁻¹) (Fig. 3), as well as modest improvements in annual mean precipi-
205 tation patterns (Fig. 4). Reducing the assumed thickness of the condensate layer (CL2),
206 while maintaining the surface pressure signature close to that in CL1, reduces the impact
207 of CL on the frequency of intense precipitation and convergence. On the other hand,
208 moderate convergence events (Fig. 3b) and annual mean precipitation in the ITCZ (Fig.
209 4c) are more strongly suppressed in CL2. This suggests that letting H_{CL} increase with
210 \mathcal{R}_{sfc} could yield improvements in both climate means and extreme event statistics.

211 We note that our parameter choices in designing the CL parameterization are based
212 on a single 5-day WRF experiment. Both the meteorological background state and the
213 choice of microphysics scheme for the WRF experiment could affect the estimate of CL
214 as a function of surface precipitation rate as well the vertical profile shape chosen to
215 represent the condensates. However, there is no reason to believe that the WRF results
216 used here grossly misrepresent these quantities, and for an exploratory study such as this,
217 we believe this is sufficient.

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References

- 221 Adler, R., et al., 2003: The Version 2 Global Precipitation Climatology Project (GPCP)
222 Monthly Precipitation Analysis (1979- Present). *Journal of Hydrometeorology*, **4** (1),
223 1147–1167.
- 224 Boyle, J. and S. Klein, 2010: Impact of horizontal resolution on climate model forecasts
225 of tropical precipitation and diabatic heating for the TWP-ICE period. *Journal of Geo-*
226 *physical Research*, **115** (D23), D23 113.
- 227 Emanuel, K., 1994: *Atmospheric Convection*. Oxford University Press, USA.
- 228 Hong, S. and J. Lim, 2006: The WRF single-moment 6-class microphysics scheme
229 (WSM6). *J. Korean Meteor. Soc.*, **42** (2), 129–151.
- 230 Huffman, G., et al., 2007: The TRMM Multisatellite Precipitation Analysis (TMPA):
231 Quasi-global, multiyear, combined-sensor precipitation estimates at fine scales. *Journal*
232 *of Hydrometeorology*, **8** (1), 38–55.
- 233 Neale, R., C.-C. Chen, A. Gettelman, P. Lauritzen, S. Park, D. Williamson, and et al.,
234 2010: Description of the NCAR Community Atmosphere Model (CAM 5.0). *NCAR*
235 *Technical Note*, **NCAR TN (486)**.
- 236 Skamarock, W. and J. Klemp, 2008: A time-split nonhydrostatic atmospheric model
237 for weather research and forecasting applications. *Journal of Computational Physics*,
238 **227** (7), 3465–3485.
- 239 Xu, K. and D. Randall, 2001: Updraft and downdraft statistics of simulated tropical
240 and midlatitude cumulus convection. *Journal of the Atmospheric Sciences*, **58** (13),
241 1630–1649.

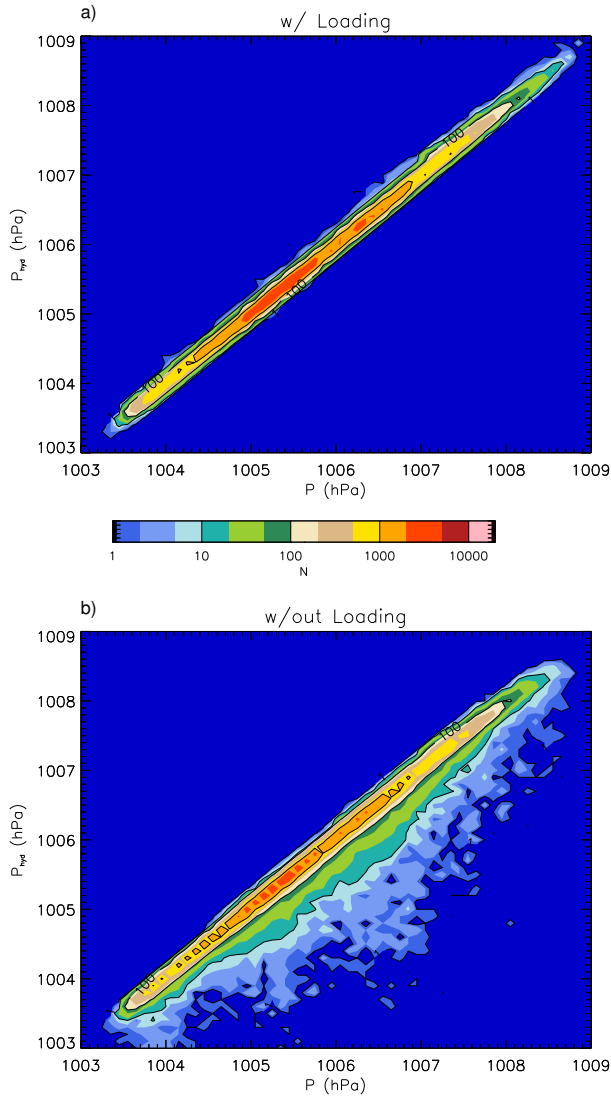


Figure 1. JFDs of prognostic WRF pressure (horizontal) vs. diagnostic hydrostatic pressure calculations (vertical). Hydrostatic pressures are calculated using fields coarse-grained to $25\text{km} \times 25\text{km}$ subdomains. Panel a shows result with a hydrostatic calculation including mass of all condensed species. Panel b shows result for hydrostatic calculation ignoring condensate masses (see text). N is the number of occurrences in each 0.1×0.1 (hPa²) bin

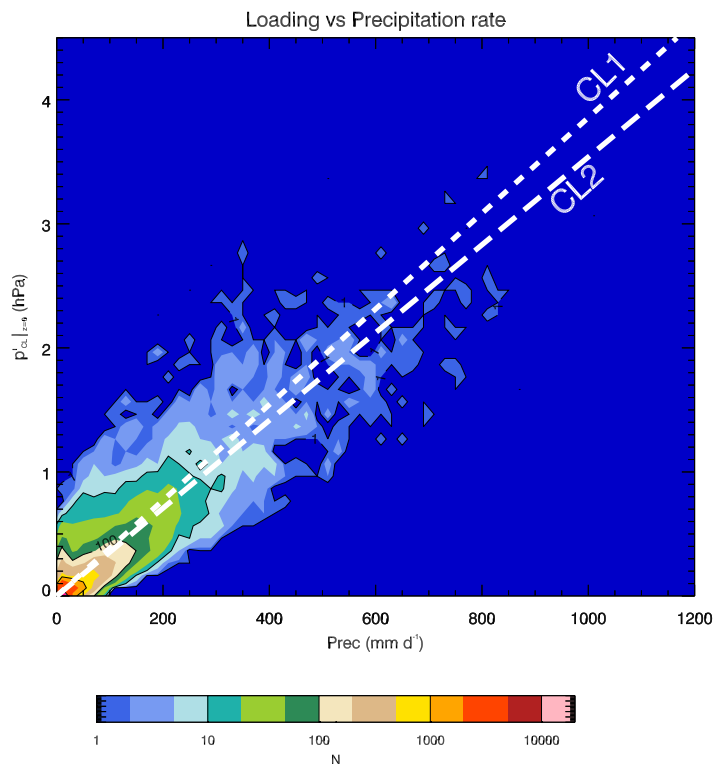


Figure 2. JFD of pressure loading at the surface from condensates (hPa, vertical axis) and surface precipitation rates \mathcal{R}_{sfc} (mm d^{-1} , horizontal axis) in $25\text{km} \times 25\text{km}$ subdomains. Dashed White lines show $p'_{CL}|_{z=0}$ for CL1 and CL2 defined in Table 2. N is the number of occurrences in each 20×0.1 ($\text{mm d}^{-1} \times \text{hPa}$) bin.

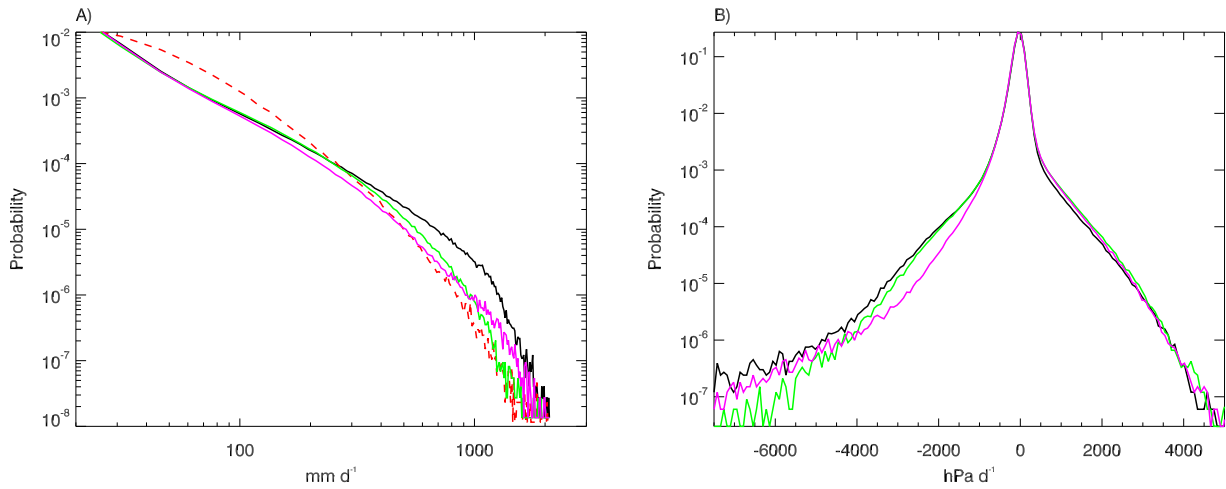


Figure 3. a) PDFs of precipitation rates for August 2005 between 30°S and 30°N for experiments defined in Table 2: CTR (black curve); CL1 (green curve); and CL2 (magenta curve). The corresponding TRMM 3B42 observational estimate is shown by the dashed red curve. Note results are displayed in log-log form. b) Same as except for vertical motion around 850 hPa (ω_{850}) over ocean, between 12°S and 25°S . Note only vertical axis is logarithmic in panel b. Probabilities are with respect to bins of 15 mm d^{-1} (a) and 80 hPa d^{-1} (b).

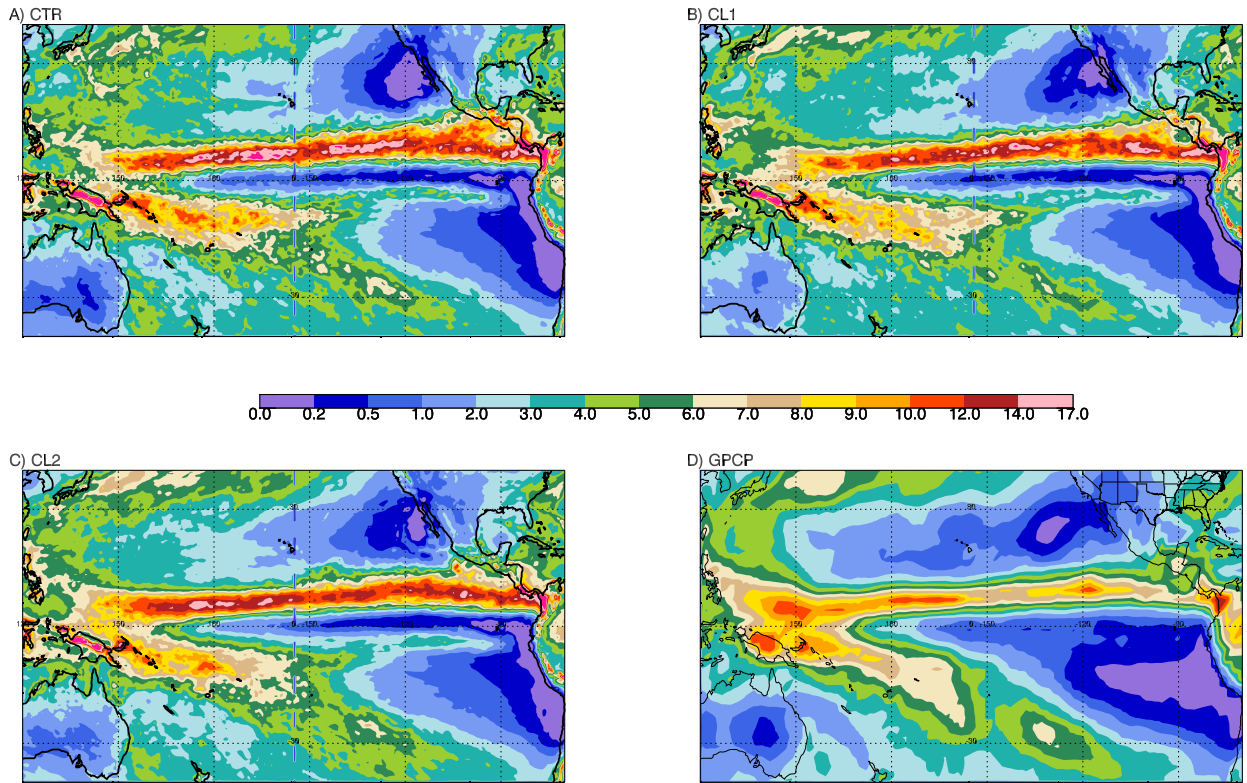


Figure 4. 12-month mean surface precipitation rate for 2/2005-1/2006 as a function of longitude and latitude for: a) CTR; b) CL1; c) CL2; and d) from the GPCP observational estimate.

Table 1. RMS differences between p_{wrf} and p_{hyd} for different coarse-graining scales

Coarse-graining scale	w/ loading (hPa)	w/out loading (hPa)
25 km	0.062	0.17
5 km	0.098	0.25

Table 2. CAM5 experiments and parameters for p'_{CL}

Experiment	w_f	H_{CL}
CTR	control, no loading	
CL1	2.5 ms^{-1}	8500 m
CL2	0.625 ms^{-1}	2000 m