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The DOI for this manuscript is

DOI:10.2151/jmsj.2021-072

J-STAGE Advance published date: September 6th, 2021

The final manuscript after publication will replace the preliminary version at the above DOI once it is available.

1 **The elements of the thermodynamic**
2 **structure of the tropical atmosphere**

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8 August 26, 2021

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Abstract

Understanding of the tropical atmosphere is elaborated around two elementary ideas, one being that density is homogenized on isobars, which is referred to as the weak temperature gradient (WTG), the other being that the vertical structure follows a moist-adiabatic lapse rate. This study uses simulations from global storm-resolving models to investigate the accuracy of these ideas. Our results show that horizontally the density temperature appears to be homogeneous, but only in the mid- and lower troposphere (between 400 hPa and 800 hPa). To achieve a homogeneous density temperature, the horizontal absolute temperature structure adjusts to balance the horizontal moisture difference. Thus, water vapor plays an important role in the horizontal temperature distribution. Density temperature patterns in the mid- and lower troposphere vary by about 0.3 K on the scale of individual ocean basins, but differ by 1 K among basins. We use equivalent potential temperature to explore the vertical structure of the tropical atmosphere and we compare the results assuming pseudo-adiabat and the reversible-adiabat (isentropic) with the effect of condensate loading. Our results suggest that the tropical atmosphere in saturated convective regions tends to adopt a thermal structure that is isentropic below the zero-degree isotherm and pseudo-adiabatic above. However, the tropical mean temperature is substantially colder, and is set by the bulk of convection which is

30 affected by entrainment in the lower troposphere.

31 **Keywords** tropical temperature; weak temperature gradient; lapse rate;
32 water vapor; cloud-resolving models

33 **1. Introduction**

34 The two principles underpinning the tropical atmosphere are that gravity
35 waves are effective in homogenizing the horizontal temperature in the free
36 troposphere (Charney, 1963; Bretherton and Smolarkiewicz, 1989) and that
37 in convecting regions the thermodynamic stratification follows the moist-
38 adiabatic lapse rate of the near surface air (Betts, 1982; Xu and Emanuel,
39 1989).

40 These principles are thought to work together to set the thermal struc-
41 ture of the free troposphere even in regions well removed from deep pre-
42 cipitating convection, i.e., across the broader tropics. Were it not for the
43 gravity waves, then convection would arise everywhere to set the vertical
44 stratification to that associated with the saturated ascent of the near sur-
45 face air, and thereby counter the destabilization of the tropical atmosphere
46 by radiative processes. In this case, however, the free troposphere would
47 adopt a thermal structure that mirrored the inhomogeneities of the under-
48 lying surface. The smallness of the Coriolis parameter combined with the
49 fast speed of deep gravity waves instead adjusts, or homogenizes, the tem-
50 perature in the atmosphere, an effect encoded in the Weak Temperature

51 Gradient approximation (WTG, Sobel and Bretherton, 2000). The grav-
52 ity waves adjust the temperature in the non-convective regions to that in
53 the convecting regions, effectively inhibiting convection at a lower convective
54 temperature. This process titrates the convection, concentrating it to
55 the area where the convective temperature is higher than some threshold
56 whose value allows just enough convection to balance the radiative cooling
57 globally. Hence these two elements (vertical homogenization by convection
58 along the local moist adiabat, and non-local horizontal homogenization by
59 gravity waves) provide the theoretical underpinnings of our broader understanding
60 of the tropical atmosphere. However, in reality – or as nearly as
61 we can approximate it – how accurate are these two principles?

62 Above, and more generally, the term ‘moist adiabat’ is applied loosely,
63 so that it is often unclear which thermodynamic process it actually is meant
64 to encapsulate. As discussed by Betts (1982) and others, when picturing
65 a moist-adiabatic process as saturated ascent of moist air, it makes a difference
66 whether condensate precipitates or remains suspended, for instance
67 by updrafts. The pseudo-adiabatic process assumes that the condensate is
68 removed from the atmosphere immediately upon formation, which makes it
69 an irreversible process. A reversible, and hence isentropic, process requires
70 not only the absence of external heating, but also for the condensed water
71 to remain in the updraft. In this case the condensate loading must be

72 accounted for when calculating the density. Ice processes compound these
73 differences, as the available fusion enthalpy is contingent on the amount of
74 condensate that can be frozen. More generally, the thermodynamic pro-
75 cesses affecting the ascent of air in deep convection may also deviate from
76 being strictly adiabatic, as mixing and radiative processes may also play a
77 role.

78 Both Betts (1982) and Xu and Emanuel (1989) analyzed soundings in
79 areas of deep convection and concluded that the tropical atmosphere is neu-
80 trally stratified with respect to the reversible-adiabatic ascent from the sub-
81 cloud layer, suggesting that the tropical temperature follows a reversible-
82 adiabatic (or isentropic) lapse rate. They justified their finding by two
83 arguments. The first was that condensate loading is maintained either by
84 virtue of being suspended in convective updrafts, or by continuity of the
85 precipitate field – precipitation lost to the air below is balanced by what
86 is gained from above. The second argument was that although mixing is
87 a characteristic of clouds, in a convecting atmosphere, some favored air-
88 parcels are shielded from their environment, thus rising without dilution,
89 and it is the thermodynamic properties of this air which determines the
90 overall stratification.

91 The idea that the undilute ascent of air within convecting regions deter-
92 mines the stratification of the tropical atmosphere is disputed by a number

93 of studies. In simulations of radiative convective equilibrium (RCE), Romps
94 and Kuang (2010) found that undilute ascent is very rare, with only 1 %
95 of the convective mass flux at 4 km could qualify as undilute. Romps and
96 Kuang (2010), and later Singh and O’Gorman (2013) using similarly config-
97 ured simulations, also showed that the convective available potential energy
98 (CAPE) associated with undilute ascent is substantial, but that the actual
99 cloud buoyancy fails to realize this CAPE due to dilution through entrain-
100 ment. Interpreting these studies is difficult. For one reason, because the
101 simulations don’t represent organized convection, but also because of sub-
102 tleties in how parcel buoyancy is calculated. If gravity waves efficiently
103 homogenize the density temperature, then the actual temperature – equiva-
104 lently the saturation moist static energy – must increase with its saturation
105 deficit. Hence, estimating the buoyancy of undilute ascent based on the
106 saturated moist-static energy of unsaturated air – as is done in these mod-
107 eling studies – will over-state the buoyancy in proportion to the saturation
108 deficit.

109 Studies analyzing sounding data taken from non-convective regions, also
110 make the case for entrainment playing a fundamental role in setting the
111 stratification of the tropical troposphere. Mapes (2001) computed the tem-
112 perature lapse rates for both Western and Eastern Pacific mean soundings,
113 and found a region of anomalously weak stability between 2 km and 5 km.

114 The stability of this layer is described as anomalously weak in that it is
115 in accord with what one would expect if the temperature were decreas-
116 ing more rapidly with height than it would following either a reversible or
117 pseudo-adiabat. This could arise through the entrainment of dry air. Later
118 Folkins and Martin (2005) also found similar deviations. However, both
119 studies estimated the stability based on the temperature, rather than the
120 density temperature, which again must decrease more rapidly with altitude
121 in progressively more sub-saturated conditions, if the density temperature
122 is constrained to be homogeneous.

123 These considerations highlight how efforts to understand the processes
124 that determine the vertical stratification of the tropical atmosphere are in-
125 tertwined with assumptions as to the what determines the horizontal distri-
126 bution of this stratification, i.e., the efficacy of gravity waves in annihilating
127 horizontal gradients of buoyancy.

128 So far our understanding of the tropical atmosphere has been built upon
129 radiosonde measurements, global modeling with parameterized convection,
130 or simulations that explicitly represent convection, albeit for very idealized
131 situations over small domains. The main limitation of radiosondes is their
132 sparse spatial coverage, particularly in the tropics. Additionally, radioson-
133 des can be influenced by biases, from poor moisture sensors, or the effects
134 of solar heating or sensor wetting during the period of severe weather. In

135 terms of modeling, simulations over a large domain in the tropics are mainly
136 achieved by global climate or numerical weather prediction models. How-
137 ever, convection usually happens at kilometer scales, which are much finer
138 than the grid size of most global models (typically at 50 km to 200 km).
139 Therefore, moist convection has to be parameterized as a sub-grid scale
140 process. These parameterizations are problematic, and often incorporate
141 assumptions about precisely those things we wish to test. In addition, moist
142 convection is an important source of gravity waves in the tropics, whether it
143 is resolved or parameterized has been shown to be able to impact the char-
144 acteristics of these waves (Müller and Hohenegger, 2020; Stephan et al.,
145 2019), something that might influence the temperature adjustment through
146 the troposphere.

147 To overcome the parameterization dilemma, one can increase the model
148 resolution to avoid the use of convective parameterization. Models that
149 adopt this approach are often referred to as convection-permitting mod-
150 els (CPMs) or storm-resolving models (SRMs). Kilometer-scale resolution
151 (typical grid spacings for such models are 3 km to 5 km) is computation-
152 ally expensive, something which in the past has limited the simulations to
153 relatively small domains, either through regional ‘downscaling’ approaches
154 or by adopting an idealized configuration (Prein et al., 2017; Wing et al.,
155 2018a; Bao and Sherwood, 2019). In recent years, with the increasing

156 computational capacity, it has become possible to carry out global storm-
157 resolving model (GSRM) simulations (Stevens et al., 2019; Satoh et al.,
158 2019). The first intercomparison project of GSRMs, DYAMOND, which
159 stands for The DYnamics of the Atmospheric general circulation Modeled
160 On Non-hydrostatic Domains, has been initiated in 2017 (Stevens et al.,
161 2019). These simulations discard convective parameterization while at the
162 same time ensuring a global domain, offering an opportunity to rejoin the
163 questions outlined above in ways that were not previously possible

164 In this paper, we investigate the thermodynamic structure of the trop-
165 ical atmosphere using output from the DYAMOND project. Our aim is
166 to answer two – and as we saw above, intertwined – questions: how ho-
167 mogeneous is tropical temperature horizontally and which process sets the
168 vertical thermodynamic structure of the troposphere. In addition, we ask
169 to what extent the answer depends on processes that are still un-, or poorly,
170 resolved in models with a grid spacing of a few kilometers. We describe the
171 data in §2 and theory and methodology in §3. In §4 we analyze to what
172 extent and on what scales gravity waves can horizontally homogenize the
173 vertical stratification in the tropics. In §5 we develop techniques that al-
174 low us to infer the stratification in the convecting regions from its value in
175 the non-convecting region, under the assumption of the weak-temperature
176 gradient. §6 presents a discussion of our findings and our main conclusions.

177 2. Data

178 We analyze the model output simulated by ICOSahedral Non-hydrostatic
179 model (ICON; Zängl et al., 2015) with a quasi-uniform horizontal mesh of
180 2.5 km. The simulations follow the experimental protocol for DYAMOND
181 (Stevens et al., 2019) in which models are required to run at storm-resolving
182 scales (5 km or less) for 40 days from August 1 in 2016. The model is
183 initialized with the global meteorological analysis at a grid spacing of 9.5 km
184 from the European Center for Medium Range Weather Forecasts (ECMWF)
185 and daily observed sea surface temperatures are forced as lower boundary
186 conditions.

187 The parameterizations used in this version of ICON are typical for
188 GSRMs. Convective parameterization is switched off for both shallow and
189 deep convection. Physical parameterizations include a microphysics scheme
190 with five hydrometeors (cloud water, cloud ice, rain, snow and graupel;
191 Baldauf et al., 2011), a turbulent mixing scheme, RRTM (Rapid Radiative
192 Transfer Model) radiation scheme (Mlawer et al., 1997) and an interactive
193 surface flux scheme. Further details about the ICON model as configured
194 for DYAMOND are provided by Hohenegger et al. (2020).

195 The data from the last 10 days of the simulations are used in the analysis.
196 As the thermodynamic structure of the tropical atmosphere is a relatively
197 stable characteristic (particularly when averaging spatially), having a short

198 time span of data is not a severe limitation. Most of the analysis is focused
199 on the tropical oceanic grids (10°N-10°S). The key conclusions from ICON
200 are compared with other models in the DYAMOND project listed in Ta-
201 ble 1. Radiosonde observational data at two tropical sounding stations –
202 Ponape (6.96°N, 158.21°E) and Pago Pago (14.33°S, 170.71°W) – retrieved
203 from the University of Wyoming archive¹ are also utilized to assist in the
204 interpretation of the simulations.

Table 1

205 **3. Theory and methodology**

206 *3.1 Notation and Definitions*

207 Thermodynamic quantities are defined following Stevens and Siebesma
208 (2020). Thereby the atmosphere is represented as two component fluid,
209 consisting of dry air and water. Subscripts indicate component properties,
210 e.g., subscript ‘d’ refers to dry air, whereas subscript v, l, i, t denote gaseous
211 (vapor), liquid, solid (ice), and total water (sum over all phases). Subscript
212 s denotes a saturation value. As examples, q_d denotes the specific mass of
213 dry air, and p_s the saturation vapor pressure. The ‘equivalent’ reference
214 state is denoted by ‘e’ and corresponds to the hypothetical situation in
215 which $q_t = q_l$.

¹<http://weather.uwyo.edu/upperair/sounding.html>

216 In this system the density temperature T_ρ , is an effective temperature
 217 that measures the ratio between the air pressure, p , and density, ρ , such
 218 that

$$219 \quad T_\rho \equiv \frac{p}{\rho R_d} = T(1 + \varepsilon_2 q_v - q_l - q_i), \quad (1)$$

220 where T is the temperature, $\varepsilon_2 = 1/\varepsilon_1 - 1$ and $\varepsilon_1 = R_d/R_v$, where R_x is the
 221 specific gas constant of component x . In most regions, q_l and q_i are negligible
 222 and do not contribute substantially to the spatial variance in T_ρ .

223 The equivalent potential temperature (θ_e) of the air is conserved for
 224 isentropic transformations of the closed system. Subject to a few common
 225 and simple assumptions, and with c_p denoting (composition dependent) the
 226 isobaric specific heat, θ_e can be expressed as

$$227 \quad \theta_e = T \left(\frac{p_0}{p} \right)^{\frac{R_e}{c_{pe}}} \left(\frac{R}{R_e} \right)^{\frac{R_e}{c_{pe}}} \left(\frac{p_v}{p_s} \right)^{\frac{-q_v R_v}{c_{pe}}} \exp \left(\frac{q_v \ell_v}{c_{pe} T} \right), \quad (2)$$

whereby

$$q_v = \begin{cases} q_s(T, p) & \text{for } q_t \geq q_s(T, p) \\ q_t, & \text{otherwise} \end{cases}$$

228 Here, following the definition of the ‘equivalent’ state, its specific heat ca-
 229 pacity and gas constant are given as

$$230 \quad c_{pe} = c_{pd} + (c_l - c_{pd})q_t, \quad \text{and} \quad R_e = R_d(1 - q_t). \quad (3)$$

231 These definitions allow one to write the (composition dependent) gas con-
 232 stant as

$$R = R_e + q_v R_v. \quad (4)$$

233 Physically θ_e can be thought of as a condensation (potential) tempera-
 234 ture (cf Betts, 1982). It measures the temperature the air would have after
 235 a two step process: (i) an adiabatic expansion that results (asymptotically)
 236 in all vapor condensing to liquid; (ii) an adiabatic compression to stan-
 237 dard pressure (p_0) of the air-condensate system, with the two components
 238 in thermal, but not mechanical, equilibrium. The second step retains the
 239 water in its condensate phase, and thus loses none of the enthalpy gained
 240 through condensation (first step) to re-vaporization. We subsequently refer
 241 to the process that conserves θ_e as isentropic. On a thermodynamic diagram
 242 θ_e -isopleths are called isentropes, which assumes the system is closed.

243 The pseudo-equivalent potential temperature is defined following Bolton
 244 (1980) as

$$\begin{aligned} \tilde{\theta}_e = T \left(\frac{p_0}{p - p_v} \right)^{\frac{R_d}{c_{pd}}} \left(\frac{T}{T_L} \right)^{0.28r_v} \\ \exp \left[\left(\frac{3036}{T_L} - 1.78 \right) (r_v + 0.448r_v^2) \right] \end{aligned} \quad (5)$$

248 where

$$T_L = \frac{2840}{3.5 \ln(T) - \ln(0.01p_v) - 4.805} + 55, \quad (6)$$

250 is an approximate equation for the temperature at the lifting condensation
 251 level and $r_v = q_v/(1 - q_v)$, describes the humidity content in the form

252 of a mixing ratio. The $\tilde{\theta}_e$ has the advantage that for the special case of
253 saturated air it reduces to a simple function of T and p , which we denote $\tilde{\theta}_s$.
254 Because θ_e varies with q_t , rather than q_v , the seemingly analogous quantity,
255 θ_s , does not have a ready physical interpretation. As we are interested in the
256 temperature profile set by convection, i.e., in a saturated atmosphere, we
257 work with $\tilde{\theta}_s$ rather than $\tilde{\theta}_e$. Processes that conserve $\tilde{\theta}_s$ are called pseudo-
258 adiabatic. On a thermodynamic diagram $\tilde{\theta}_s$ -isopleths are called pseudo-
259 adiabats.

260 3.2 θ_e versus $\tilde{\theta}_s$ coordinates

261 If moist air undergoes an isentropic expansion without any exchange of
262 mass, then T would change in a way that keeps θ_e constant as p decreases
263 for the given q_t . Choosing θ_e as a coordinate (with q_t specified) results in
264 this quantity remaining unchanged. Likewise psuedo-adiabats are vertical
265 lines in a coordinate system whose abscissa measures $\tilde{\theta}_s$.

266 The advantage of describing the state of the atmosphere using either θ_e
267 or $\tilde{\theta}_s$ as a coordinate is that these quantities are not expected to change
268 under certain types of transformations. Hence, measuring how much θ_e or
269 $\tilde{\theta}_s$ does change can be indicative of the thermodynamic processes associated
270 with a particular process, for instance deep moist convection. The trivial
271 example, and the one many researchers employ as a mental model, is that of

272 moist convection being pseudo-adiabatic and gravity waves efficiently act-
273 ing within the free troposphere to adjust the temperature along isobars to
274 its value in the convective region. In this example, $\tilde{\theta}_s$ would adopt a single
275 value throughout the free troposphere. This expectation motivates analyses
276 of the thermodynamic structure of the troposphere with $\tilde{\theta}_s$ as a thermody-
277 namic coordinate. However, convection may not be pseudo-adiabatic. Con-
278 sider the case that, as argued by Betts (1982) and Xu and Emanuel (1989),
279 convection follows an isentrope. In that case, if one adopted θ_e as a thermo-
280 dynamic coordinate, then θ_e profile should exhibit a constant vertical line,
281 varying only with p . However this will only be the case if θ_e is computed
282 with the value of q_t in the saturated convective region where the isentropic
283 process occurs, which we denote $q_{t,c}$.

284 To avoid local variations in q_t masking an isentropic temperature profile,
285 one can fix q_t in the calculation of θ_e to the value, $q_{t,c}$, it has in the satu-
286 rated convective region. To indicate when we calculate θ_e in this fashion we
287 write $\theta_e(T, p; q_{t,c})$. The semi-colon notation indicates that when evaluating
288 Eq. (2), q_t is fixed as a parameter with value $q_{t,c}$, which is either known
289 or must be estimated. Fortunately, the bias from over or under-estimating
290 $q_{t,c}$ by a small amount is also small, and estimates of $q_{t,c}$ are strongly con-
291 strained by the constancy of cloud base in the convective region. This
292 relative insensitivity of $\theta_e(T, p; q_{t,c})$ to the estimate of $q_{t,c}$ can be inferred

293 by inspection of Eq. (2). The main effect of the q_t on θ_e is through the q_v
294 term. As long as $q_t > q_s$, $q_v = q_s$. Hence is given by T and p . The small
295 influence on the specific heat results in a small (0.25 K) decrease in θ_e in the
296 lower troposphere for a 1 g kg^{-1} overestimation of $q_{t,c}$. This bias increases
297 with height, to a value about twice as large in the upper troposphere, but
298 two times a small number is still small.

299 To help interpret the state of the atmosphere using θ_e and p as thermody-
300 namic coordinates, Fig. 1 illustrates the fundamental lines associated with
301 different processes when plotted in these coordinates. The profile represent-
302 ing an isentropic process shows a constant line in θ_e coordinate, whereas
303 the pseudo-adiabatic profile computed in θ_e coordinate decreases roughly
304 linearly with geometric height (and hence exponentially with pressure), so
305 that values in the upper troposphere will be reduced by as much as 10 K.
306 The fundamental lines that incorporate additional processes, such as ice for-
307 mation, show similar deviations. For instance, an isentrope that allows for
308 freezing implies considerably larger values of θ_e , starting with the release of
309 fusion enthalpy as liquid-condensate freezes at the triple point temperature.

310 The situation is reversed if one adopts $\tilde{\theta}_s$ as a thermodynamic coordinate
311 (Fig. 1b). In that case, should T follow an isentrope it implies a progressive
312 increase in $\tilde{\theta}_s$, mirroring the decrease of θ_e associated with pseudo-adiabatic
313 temperature profiles. Understanding this difference also aids the interpre-

314 tation of the other fundamental lines, for instance for an entraining plume,
315 which for this simple example is modeled as an exponential relaxation to a
316 5 K lower θ_e over a 150 hPa layer.

Fig. 1

317 3.3 *Estimating the effective convective temperature profile T_c*

318 The question this manuscript poses is whether profiles of T and q_t
319 throughout the global tropics can inform us about the effective convec-
320 tive temperature profile, T_c . For our purposes, and unlike what is done in
321 most other studies, T_c is not associated with any preconceived idea of con-
322 vection, rather it is the temperature profile that the tropical troposphere
323 appears to be adjusting too. As such it should be identifiable from pro-
324 files of T and q_t throughout the global tropics. The reason for adopting
325 this method to estimate the effective convective temperature profiles rather
326 than to analyze the profiles in the actual convective regions is that we do
327 not know exactly which convection sets the temperature horizontally. With
328 this method, we can compare across the effective convective temperature
329 profiles inferred from all grid points over the tropical oceans and then de-
330 termine what fraction of convection sets the temperature in the tropical
331 mean state. Irrespective of what process determine T_c , we do not expect
332 this to determine T throughout the global tropics. If anything, the profile of
333 the density temperature, T_ρ , is what will be adjusted by gravity waves. In

334 that case, one expects isopleths of T_ρ to be parallel to isobars. This makes
 335 inferring the profile of T_c from profiles of T more delicate, as doing so must
 336 properly account for differences in $q_{t,c}$ and q_t .

337 Our approach is illustrated with the help of the schematic in Fig. 2.
 338 Rather than guessing which grid columns are representative of the con-
 339 vecting regions, we attempt to infer T_c from local (usually non-convective)
 340 profiles of T_ρ . Assuming T_ρ is constant on isobars, this implies that $T_{\rho,c} \approx T_\rho$.
 341 Depending on the disposition of the condensate in the convecting regions,
 342 two possibilities bound our thinking. The first is that T_c follows an isen-
 343 tropic process. In this case, condensate is present in the convecting region,
 344 and

$$345 \quad T_{\rho,c} = T_c [1 + (\epsilon_2 + 1)q_s - q_{t,c}] \quad (7)$$

346 and $q_{t,c} \geq q_s(T_c, p)$ is constant, but must be additionally specified. The
 347 second possibility is that T_c follows a pseudo-adiabatic process, whereby
 348 $q_{t,c} = q_{v,c} = q_s(T_c, p)$ and is thus known. Given $T_{\rho,c}$ as a function of pres-
 349 sure, one can invert Equation (7) to derive T_c subject to one or the other
 350 assumption regarding $q_{t,c}$. For consistency, the first method is used when
 351 representing estimates of T_c using θ_e as a thermodynamic coordinate, the
 352 second when T_c is represented with $\tilde{\theta}_s$ as the thermodynamic coordinate.

Fig. 2

353 A difficulty that arises when estimating T_c from $T_{\rho,c}$ is that the re-
 354 sulting profile is sensitive to what one assumes about $q_{t,c}$. Two examples

355 illustrate this point. For the first example we take the case of pseudo-
 356 adiabatic atmosphere, but is assumed to be isentropic in the calculation so
 357 that $q_{t,c}$ is held constant. According to Equation (7), incorrectly assuming
 358 an isentropic profile implies $T_{\rho,c}$ is increasingly (with height) burdened by
 359 condensate loading, which must be balanced by an overestimation of T_c for
 360 a given $T_{\rho,c}$. As a result θ_e increases with height. Fig. 3 shows the result,
 361 whereby θ_e increases to a maximum in the middle-upper troposphere (solid
 362 line). The reversal and progressive decrease of θ_e in the upper troposphere
 363 arises from an increasingly important and countervailing bias that arises
 364 by failing to account for the loss of condensate enthalpy associated with a
 365 pseudo-adiabatic temperature profile (e.g., as shown by the grey line in the
 366 left panel of Fig. 1). The second example, shows how the situation reverses
 367 (dotted line in Fig. 3) if T_c follows an isentrope but is estimated from its
 368 remote T_ρ profile by assuming it follows an pseudo-adiabat.

Fig. 3

369 4. Horizontal structure

Fig. 4

370 Before applying the above theory to vertical profiles of model output,
 371 or data, in this section we first explore how well the Weak ‘Temperature’
 372 Gradient is satisfied in the simulations. We begin our analysis by exam-
 373 ining the ICON-simulated spatial distribution of precipitable water (PW),
 374 which is the total vertically integrated atmospheric water vapor (Fig. 4).

375 Figure 4 illustrates that dry and moist regions are well separated. PW is
376 high mainly over regions near the Maritime Continent. Besides, there is a
377 long narrow band of high PW at around 10°N , indicating the location of the
378 Inter Tropical Convergence Zone (ITCZ). PW is low mainly in the South-
379 ern Hemisphere including the Eastern and Central Pacific and the South
380 Atlantic. The PW distribution reflects the location of convection as well as
381 non-convecting environment due to the effect of convective moistening or
382 subsidence drying.

Fig. 5

383 To investigate the horizontal temperature distribution, we choose two
384 levels: 300 hPa and 600 hPa representing the upper and mid-troposphere
385 respectively. Figure 5 shows the spatial distribution of temperature (T)
386 and the density temperature (T_ρ) anomaly (relative to domain-mean value)
387 at 300 hPa and 600 hPa. The difference in T and T_ρ indicates mainly the
388 impact of water vapor. At 300 hPa, T and T_ρ are almost identical due to
389 little water vapor existing there (Fig. 5). Moist regions like the Western
390 Pacific and oceans near the Maritime Continent are generally warmer than
391 dry regions like the Eastern Pacific. The maximum anomaly between the
392 Western and Eastern Pacific is over 3.5 K. However, at 600 hPa, both T
393 and T_ρ are more homogeneous: over the Pacific Ocean, the maximum T_ρ
394 anomaly is less than 1 K, and over the Atlantic Ocean, the T_ρ anomaly is
395 also reduced. Despite being more homogeneous locally at 600 hPa, struc-

396 ture is evident on large (60° of longitude) scales, which appear to align with
397 different ocean basins. The 1 K difference between a colder Atlantic and a
398 warmer Eastern Pacific, is particularly pronounced. This seems to suggest
399 that the different ocean basins are adjusting to convection at different tem-
400 peratures, and that inter basin communication may be hindered either by
401 the distances between the basins or by land masses, where orography and
402 the diurnal cycle influence the atmospheric structure.

403 Because T_ρ is horizontally more homogeneous at 600 hPa than at 300 hPa,
404 there must be a larger lapse rate in places where T_ρ at 300 hPa is smaller.
405 This is confirmed in Fig. 6a, which plots $\delta_z T_\rho = T_\rho|_{600 \text{ hPa}} - T_\rho|_{300 \text{ hPa}}$. Over
406 the Eastern Pacific and the Southern Atlantic $\delta_z T_\rho$ is anomalously large,
407 whereas over the Western Pacific and the Maritime Continent it is anoma-
408 lously small. The pattern of $\delta_z T_\rho$ resembles the pattern of PW. This ap-
409 parent correlation is quantified in Fig. 6c which shows that PW and $\delta_z T_\rho$
410 anomalies are negatively correlated with a correlation coefficient of -0.66.

411 The negative correlation between PW and $\delta_z T_\rho$ is not due to the va-
412 por buoyancy effect, as water vapor is included in the calculation of T_ρ ,
413 and therefore should act to reduce the horizontal heterogeneities in T_ρ . In-
414 stead, it implies that gravity waves are less effective at homogenizing the
415 buoyancy field in the upper (300 hPa) troposphere than they are in the
416 mid-troposphere (600 hPa). If gravity waves were equally effective at ho-

417 mogenizing T_ρ at both levels, there would be no difference in $\delta_z T_\rho$ horizon-
418 tally. Without taking into account the vapor buoyancy effect, $\delta_z T$ anomaly
419 is larger (Fig. 6b) and the negative correlation becomes more robust be-
420 tween PW and $\delta_z T$ anomaly (Fig. 6d). The vapor buoyancy effect can be
421 interpreted by considering a simple idealized case where T_ρ is homogeneous
422 throughout the entire free troposphere. According to Eq.1, T and T_ρ differs
423 when water vapor exists, this means that in the upper troposphere T is
424 almost homogeneous while in the mid- to lower troposphere T varies de-
425 pending on the horizontal differences in water vapor. Given the same T_ρ ,
426 the difference in water vapor enhances T in dry regions and reduces T in
427 moist regions because moist air is less dense than dry air at the same tem-
428 perature. As the only deciding factor is water vapor, the vapor buoyancy
429 effect would lead to a strong negative relationship between PW and $\delta_z T$.
430 However, the results in Fig. 6 indicate that the vapor buoyancy effect is not
431 the dominant factor, but contributing to the negative relationship between
432 PW and $\delta_z T$.

Fig. 6

433 To better understand the temperature structure and the vapor buoy-
434 ancy effect, we calculate the pattern correlation between PW and T , or T_ρ ,
435 at different pressure levels. The variation of the correlation coefficient with
436 pressure is plotted in Fig. 7. Differences in how T and T_ρ correlate with PW
437 is indicative the vapor buoyancy effect. If two profiles overlap, it means that

438 either there is little water vapor (as is the case above 300 hPa) or the vapor
439 buoyancy effect is not dominant (in the boundary layer). For comparison we
440 plot the same figure with data from the Radiative-Convective Equilibrium
441 Model Intercomparison Project (RCEMIP; Wing et al., 2018b). The data
442 that we use are from the ICON-LEM (Dipankar et al., 2015) configured over
443 an elongated channel domain (6000 km \times 400 km) and employing a horizon-
444 tal grid spacing of 3 km. As there is no rotation and the domain is small
445 compared to the global simulations (albeit orders of magnitude larger than
446 the simulation domains used in many previous studies), T_ρ is extremely ho-
447 mogeneous throughout the entire free troposphere. This is illustrated by
448 near-zero correlation coefficients between PW and T_ρ . Given the homoge-
449 neous T_ρ , water vapor becomes the only factor impacting T which leads to
450 strong negative correlations between PW and T . Thus, idealized simula-
451 tions of radiative-convective equilibrium provides a setting where gravity
452 waves function effectively throughout the free troposphere. In reality, and
453 on larger-domains, we expect the vapor buoyancy effect to become more
454 dominant in the relationship between PW and T under the condition that
455 T_ρ becomes more homogeneous.

Fig. 7

456 For the DYAMOND simulations of a more realistic setting, first we focus
457 on the ICON output over the region from 10°N-10°S. Figure 7b indicates
458 that there are two positive correlation maxima: one near the top of the

459 boundary layer and one near 300 hPa. The high correlations in the bound-
460 ary layer are expected as the boundary layer is well mixed and feels strongly
461 the imprint of the temperature at the sea-surface. The other peak at 300 hPa
462 confirms that T is not homogeneous in the upper troposphere, but varies
463 similarly as PW. Between 400 hPa and 800 hPa, PW has weak positive corre-
464 lation with T_ρ , and negative correlation with T . This means that the vapor
465 buoyancy effect becomes more important in the mid-troposphere, therefore,
466 denoting a more homogeneous T_ρ .

467 When the analysis is performed over the broader tropics, to also in-
468 clude the subtropics, both T_ρ and T exhibit positive correlations with PW
469 (Fig. 7). The correlation coefficients are above 0.5 throughout the entire
470 troposphere over 30°N-30°S, implying that T_ρ is not homogeneous even in
471 the mid-troposphere and such a large area cannot be effectively influenced
472 by the tropical convection through gravity waves. The poleward increase of
473 the coriolis parameter increasingly allows the atmosphere to balance density
474 gradients away from the equator. This analysis indicates that to the extent
475 it is a valid approximation, the weak gradients of T_ρ or weak buoyancy gradi-
476 ent (WBG) describes the thermal structure of the atmosphere equatorward
477 of 10° or maybe 20°, and mostly between 400 hPa and 800 hPa.

478 We hypothesize that differences in the degree to which T_ρ is homogenized
479 with height reflects the effectiveness of gravity waves in communicating,

480 and hence homogenizing, density anomalies there. The gravity waves that
481 cause widespread subsidence over non-convective regions are deep, with a
482 half-wavelength which spans the depth of the heating layer (Mapes, 1993).
483 However, the wave transports of buoyancy become less effective near the up-
484 per and lower boundaries, both because it is hard to get strong subsidence
485 motion near these boundaries (Bretherton and Smolarkiewicz, 1989) and
486 because the gravity wave propagation speed is proportional to the vertical
487 wavelength. Hence, proportionally smaller vertical modes are required to
488 homogenize density anomalies confined to shallower layers. Shallow density
489 anomalies arising from imbalances between diabatic (radiative) heating and
490 subsidence warming are thus less effectively homogenized by gravity waves.
491 We speculate that far away from the convection the ability of convectively
492 generated gravity waves to generate sufficient subsidence to balance the ra-
493 diative cooling, thereby equilibrating the temperature to that in the convec-
494 tive region, is thus diminished. Therefore, upper-tropospheric temperature
495 in non-convective regions is colder.

496 The above results highlight the important role that water vapor plays in
497 the temperature lapse rate and reaffirm that horizontally T_ρ is homogeneous
498 mostly in the middle of the troposphere (400 hPa to 800 hPa) in the deep
499 (10°S-10°N) tropics.

500 5. Vertical structure

501 In this section, we focus on the vertical temperature structure. From the
502 previous section we saw that although gravity waves do more efficiently ad-
503 just T_ρ than T , variations in T_ρ of about 1 K emerge in the mid-troposphere
504 across the inner tropics, and that these temperatures become more pro-
505 nounced and positively correlated with PW in the upper and lower tropo-
506 sphere. These differences should help guide our interpretation of the value
507 and vertical structure of T_c as deduced from profiles of T and q_t .

508 5.1 *Estimating T_c from global profiles of T and q_t*

509 Here we use profiles of T and q_t simulated by ICON over the inner
510 (10°N - 10°S) tropics. From these we use the methodology described in §3.3
511 to infer profiles of T_c which we then render in θ_e and $\tilde{\theta}_s$ coordinates to see
512 how they vary, both in the vertical and as a function of PW. The former
513 should be indicative of the thermodynamic processes in the saturated con-
514 vective regions that, to a first approximation, set the thermal structure of
515 the tropics; the latter should be indicative of the extent to which other, non-
516 convective processes, cause deviations from this. Recall that the manner in
517 which T_c is estimated differs depending on which thermodynamic coordi-
518 nate is adopted. For reference we also show uncompensated temperature
519 profiles using the $\tilde{\theta}_s$ coordinate, which amounts to taking $T_c = T$. The three

520 coordinates are summarized with the help of Table 2, which also sets the
521 nomenclature.

Table 2

522 Following the outline of Table 2, $\tilde{\theta}_s(T)$, $\tilde{\theta}_s(T_c)$ and $\theta_e(T_c; q_{t,c})$ are plot-
523 ted in Fig. 8. Profiles are constructed for different values of PW, thereby
524 showing how T_c varies across moisture space in the tropics. Calculation
525 of $\tilde{\theta}_s(T_c)$ and $\theta_e(T_c; q_{t,c})$ assumes that T_ρ is homogenized by gravity waves.
526 As discussed in the previous section this is most approximately true in the
527 free troposphere, near 600 hPa, but not in the unstratified boundary layer,
528 where waves are not supported. This point notwithstanding Fig. 8 shows
529 variations in $\theta_e(T_c; q_{t,c})$ (equivalently $\tilde{\theta}_s(T_c)$) depend only weakly on PW in
530 the free troposphere (above 800 hPa). Additionally, and most importantly,
531 the use of T_c inferred from T_ρ better collapses (groups) the data than does
532 T . This is most evident in the elimination of the apparent local maximum
533 in $\tilde{\theta}_s(T)$ that emerges in the driest columns near 800 hPa (Fig. 8a).

Fig. 8

534 Whereas $\tilde{\theta}_s(T)$ is larger in moist regions and smaller in dry regions in the
535 upper troposphere (200 hPa to 400 hPa), the opposite is true in the lower
536 and middle troposphere (400 hPa to 800 hPa). This implies large differ-
537 ences in temperature lapse rates, consistent with the observational analyses
538 by Mapes (2001) and Folkins and Martin (2005), which also were based on
539 T . To a large extent, the differences in the mid- and lower-tropospheric
540 $\tilde{\theta}_s(T)$ can be traced to the impact of water vapor, as $\tilde{\theta}_s(T_c)$ becomes more

541 uniform by applying T_c assuming constant T_ρ . This suggests that the ap-
542 parently strong deviations from the pseudo-adiabat or the isentrope that
543 these studies identified in the lower troposphere (600 hPa to 800 hPa), may
544 have resulted from neglecting the water vapor effect on buoyancy.

545 A prominent feature in all three panels of Fig. 8 is its increase with mois-
546 ture in the upper troposphere. This implies that neither T , nor T_ρ is homo-
547 geneous (irrespective of how one estimates T_c) and the wave-homogenization
548 mechanism there may not function as well as that in the mid-troposphere.
549 These profiles are consistent with the analysis in §4, and indicates that in
550 the upper troposphere, regions close to deep convection are expected to be
551 warmer than more distant regions. Such differences can be expected to
552 support a large-scale circulation in the upper troposphere analagous to that
553 discussed by Mapes (2001).

Fig. 9

554 The profiles in Fig. 8 suggest that an isentrope (constant $\theta_e(T_c; q_{t,c})$) is a
555 good description of the lower troposphere (below roughly 600 hPa). Above
556 600 hPa, $\theta_e(T_c; q_{t,c})$ more closely approximates the ice-pseudo-adiabat, as
557 inferred by comparison to the theoretical profiles in Fig. (1). In this inter-
558 pretation our eye is drawn to the increase with height above 600 hPa (to
559 a local maximum near 400 hPa) and subsequent fall off with height above
560 that point, reaching a local minimum value between 250 hPa to 200 hPa.

561 Similar profiles are also found in the atmosphere with the most extreme

562 values of PW, which we take to be representative of regions of the deepest
 563 convection. Fig. 9 presents profiles of $\tilde{\theta}_s(T)$, $\tilde{\theta}_s(T_c)$ and $\theta_e(T_c; q_{t,c})$ for the
 564 99.999 percentile of PW (which comprises roughly 100 samples per time-
 565 step). Because the profile of $\tilde{\theta}_s(T_c)$ in Fig. 9 do not differ from $\tilde{\theta}_s(T)$, this
 566 confirms that our selection identifies saturated grid-columns. As none of
 567 the profiles exhibits a constant structure throughout the full depth of the
 568 troposphere, it suggests that even in the most water-laden columns no sin-
 569 gle process (either pseudo-adiabatic or isentropic) can describe the thermal
 570 structure in these saturated regions alone. However, a combination of the
 571 pseudo-adiabatic and the isentropic processes seems like a good description:
 572 below about 600 hPa the profile follows more closely a saturated isentrope,
 573 whereas a pseudo-adiabat appears a good representation of the thermal
 574 structure above.

575 There is a temptation to conclude that because the mean profile of
 576 $\theta_e(T_c; q_{t,c})$ is similar in shape to the profile in the moistest regions, these
 577 latter regions dictate the thermal structure of the tropical troposphere.
 578 A substantially larger mid-troposphere $\theta_e(T_c; q_{t,c})$ in the extremely moist
 579 columns (345 K), as compared to the average (342 K), suggests that this is
 580 not the case. Nor can it be concluded that just because a profile follows one
 581 or the other fundamental line that it is determined by the process associated
 582 with this line. For instance, A constant-like $\theta_e(T_c; q_{t,c})$ for the mean state in

583 the lower troposphere could also arise as a result of several processes com-
584 pensating each other. In the next section, by compositing on progressively
585 moister columns, we explore both of these points in more depth.

Fig. 10

586 5.2 *Processes determining the mean thermal structure of the* 587 *troposphere*

588 It may seem contradictory that in §4 we conclude that T_ρ is horizontally
589 homogeneous especially in the mid-troposphere, yet above identify relatively
590 large (3 K) deviations of mid-troposphere $\theta_e(T_c; q_{t,c})$ in the very moistest
591 columns. Because regions of such extreme PW are so rare, their ability
592 to influence the structure of the troposphere as a whole is likely limited,
593 likewise their ability to exist out of balance with the mean structure of
594 the troposphere will be considerable. So while not contradictory, it does
595 raise the question as to what fraction of the convecting atmosphere, or
596 which percentile of the PW distribution, is responsible for setting the mean
597 properties of the tropical troposphere.

598 To investigate this issue more systematically, we compare the T_ρ anomaly
599 at 600 hPa as a function of percentiles of PW. Figure 10 shows that the
600 T_ρ anomaly changes relatively little (≈ 0.3 K) below the 99th percentile of
601 PW. In contrast, T_ρ anomaly increases sharply (note the log-axis) above the
602 99th percentile. From this we infer that the tropical temperature profile is

603 adjusting to the average temperature profile set by convection in roughly
 604 the moistest (as measured by PW) one percent of the tropics. Profiles of
 605 $\theta_e(T_c; q_{t,c})$ and $\tilde{\theta}_s(T_c)$ for columns within the upper PW decile are plotted in
 606 Figure 11 and support this inference. The columns with yet more extreme
 607 values of PW are considerably warmer than the mean, but that already at
 608 the 99th percentile, the temperature is very close to the tropical mean.

Fig. 11

609 Profiles of $\theta_e(T_c; q_{t,c})$ also hint at what processes might be influencing
 610 the thermal structure of the troposphere in the mean state (Fig. 11). A
 611 feature that captures our attention is the systematic increase (with decreasing
 612 percentile of PW) of the $\theta_e(T_c; q_{t,c})$ lapse rate below 600 hPa. Whereas
 613 the 99.999th percentile has a slightly increasing value of $\theta_e(T_c; q_{t,c})$ with
 614 height, the profile of the 99th and 90th percentile is slightly decreasing
 615 (larger lapse rate). Increasing $\theta_e(T_c; q_{t,c})$ is a signature of pseudo-adiabatic
 616 effects, decreasing $\theta_e(T_c; q_{t,c})$ is a signature of entrainment. However, in
 617 a drier atmosphere (as expected in the lower percentiles), entrainment is
 618 more effective in reducing the updraft buoyancy (temperature), so even if
 619 the most moist convection is entraining the same as convection in drier re-
 620 gions, it will be less evident. This supports the idea that in the tropical
 621 mean state, the isentropic-like profiles of T_c in the lower troposphere arise
 622 from our analysis not because the convection follows a saturated isentrope,
 623 rather due to the competing effects of a pseudo-adiabatic process on the

624 relationship between buoyancy and temperature that we use to diagnose
625 T_c , and the effect of entrainment on T_c directly. Our analysis of the ICON
626 simulations thus supports arguments by Singh and O’Gorman (2013), that
627 the tropical mean condition is not determined by the warmest air parcels
628 that are nearly undiluted, but rather by the bulk of convection subject to
629 the influence of entrainment in the lower troposphere.

630 The shape of the profiles above 600 hPa is more difficult to interpret.
631 The moistest profiles (99.999th percentile) appear more pseudo-adiabatic,
632 in which case ice processes are only a small perturbation. However, the drier
633 profiles are more stably stratified, as they approach the moister profiles with
634 decreasing pressure. We speculate that the convection that can reach the
635 upper troposphere is very rare, and only those with very high boundary-
636 layer θ_e in the saturated environment can survive in the upper troposphere,
637 whereas more convection can get to the mid-troposphere. Thus, the high
638 stability in the drier profiles in the upper-troposphere indicates that with
639 decreasing pressure, the temperature is more controlled by the convection
640 with higher θ_e .

641 In summary, the saturated regions with deep convection in ICON ap-
642 pear to be well described by an isentropic profile below 600 hPa and by a
643 pseudo-adiabat aloft. However this does not appear to be indicative of the
644 mean state actually being described by these processes, but rather through

645 a compensation of competing effects, with different balances in the lower
646 versus upper troposphere.

Fig. 12

647 5.3 *Testing the robustness of inferences from ICON output*

648 A sensible question to ask is whether the above conclusions hold in other
649 DYAMOND models or in data from tropical radiosondes. Fig. 12 shows
650 profiles of $\theta_e(T_c; q_{t,c})$ from six DYAMOND models for the mean and the
651 humid conditions. A first impression of Fig. 12 is that most models convect
652 at a similar temperature (at 600 hPa, $\theta_e(T_c; q_{t,c}) \approx 342$ K), GEOS being
653 somewhat warmer and SAM being somewhat colder than the other models.
654 The models also appear to differ with respect to the exact thermodynamic
655 process which sets the temperature structure in the convective regions. SAM
656 and IFS show a tendency for $\theta_e(T_c; q_t)$ to decrease with height, which can
657 only be explained by a greater role for entrainment. GEOS and FV3 are
658 similar to ICON, NICAM has more pronounced increase in $\theta_e(T_c; q_{t,c})$ with
659 height, indicative of a slightly more pseudo-adiabatic profile below 600 hPa.
660 Notwithstanding these differences, some further inferences from the analysis
661 of ICON hold across these models. First, most models show that the tropical
662 mean $\theta_e(T_c; q_{t,c})$ overlaps with $\theta_e(T_c; q_{t,c})$ over the upper percentile of PW
663 in the mid (600 hPa) troposphere. Second, all models have a tropical mean
664 $\theta_e(T_c; q_{t,c})$ that decreases between 400 hPa and a local minimum near 200 hPa,

665 as one would expect if convection followed a pseudo-adiabat in the upper
666 troposphere. Third, most models show that $\theta_e(T_c; q_{t,c})$ in the most humid
667 regions is significantly larger than the tropical mean value, and the ‘cold-
668 point’ temperature locates at lower pressure.

669 Finally, and as a sanity check, we compare the model results with mea-
670 surements by tropical radiosondes. Although there is no reliable obser-
671 vational product covering the entire tropics, the advantage of the above
672 analysis is that it shows that for many questions one can infer the convec-
673 tive profiles from anywhere in the tropics. Only the apparent dependence
674 of $\theta_e(T_c; q_{t,c})$ on PW in the upper troposphere needs soundings that ade-
675 quately sample moist and dry regions. For many places in the tropics, the
676 seasonal migration of ITCZ allows an individual station to sample the dry
677 and the moist tropics and hence, by adjusting for sampling biases, to ad-
678 dress this question. Here we show results from two stations in the tropical
679 Pacific (Fig. 13). In general, the observations corroborate the main findings
680 from ICON. That the two soundings appear less consistent with respect to
681 which convection sets the tropical mean lapse rate. That the mean profile
682 at Pago Pago (14.33°S, 170.71°W), is less well adjusted to the upper decile,
683 or percentile, may reflect its distance from the equator, which influences the
684 adjustment process as shown in the analysis of § 4, Fig. 7.

Fig. 13

685 6. Conclusions

686 This paper presents our analysis of simulations from a global storm-
687 resolving model (ICON) to investigate the validity of the two important
688 principles of the tropical atmosphere: the horizontal temperature in the
689 free troposphere is homogeneous, which is referred to as the weak tempera-
690 ture gradient (WTG) approximation, and that the vertical structure follows
691 a moist-adiabatic lapse rate – albeit often without a precise definition of the
692 moist adiabat. Our results show that, horizontally, the density temperature
693 (T_ρ) is roughly homogeneous in the mid- and lower troposphere except those
694 regions with deep convection ($\sim 1\%$) being substantially warmer than the
695 rest of the tropical domain. Vertically, the tropical atmosphere in the satu-
696 rated convective regions tends to adopt a thermal structure that is isentropic
697 below the zero-degree isotherm and pseudo-adiabatic above. However, the
698 tropical mean temperature is substantially colder, and is set by the bulk of
699 convection which is affected by entrainment in the lower troposphere.

700 The model results highlight the important role that water vapor plays
701 in the horizontal temperature (T) distribution. The vapor buoyancy effect
702 arises from the unbearable lightness of the water molecule (H_2O) is much
703 smaller than that cocktail of N_2 , O_2 and Ar known as ‘dry air’. At the
704 same pressure and temperature, moist air is less dense than dry air. In the
705 tropics, where the horizontal buoyancy differences are efficiently eliminated

706 by gravity waves, the density temperature (T_ρ), a compensated temperature
707 that includes the density effect of water vapor (and condensate loading when
708 present), is expected to be homogeneous. Hence, for T_ρ to be horizontally
709 homogeneous, T has to vary with the specific humidity. The model re-
710 sults show that T_ρ is relatively homogeneous between 400 hPa and 800 hPa,
711 which defines the mid, and lower mid-troposphere. Because of the effect
712 of vapor on air density, the absolute temperature is colder in moist regions
713 and warmer in dry regions. The latter gives rise to an apparent inversion
714 in the dry regions. Above 400 hPa both the absolute temperature and the
715 density temperature are also less homogeneous, and vary as a function of
716 moisture. This is indicative of a less effective homogenization by gravity
717 waves at these levels and, we speculate, the tendency of the upper tropo-
718 sphere to be more strongly influenced by more θ_e -rich convection, whose
719 rareness makes its effects most pronounced in its local environment.

720 We use equivalent potential temperature to explore the vertical structure
721 of the tropical atmosphere. Two thermodynamic coordinates are adopted.
722 One, $\tilde{\theta}_s$, is constant for a pseudo-adiabat, the other, θ_e , is invariant follow-
723 ing a saturated isentrope. Deviations of the atmospheric thermal structure
724 from an isopleth in these coordinates are used to explore thermodynamic
725 processes that set the thermal structure in the convecting regions – albeit
726 without the need to first identify these regions. To perform this analysis

727 it is necessary to estimate the convective profile, $T_c(p)$ consistent with the
728 local temperature and moisture profile and an assumed buoyancy homg-
729 enization (WTG). In ICON in the most saturated regions of the tropical
730 troposphere, this analysis identifies a thermal structure that is isentropic
731 below the zero-degree isotherm and pseudo-adiabatic above. This structure
732 is also evident in the mean. Nonetheless, by comparing profiles conditioned
733 on PW, we conclude that in the mean state, the apparent isentropic pro-
734 file in the lower troposphere is a result of entrainment masking the effects
735 of pseudo-adiabatic ascent and its implication for the buoyancy, if not the
736 temperature, profile. This contradicts early observational studies that trop-
737 ical atmosphere is neutral to the isentropic ascent from the sub-cloud layer
738 (Betts, 1986; Xu and Emanuel, 1989), but supports recent work using ide-
739 alized simulations in which the fundamental role of entrainment in tropical
740 lapse rate has been recognized (Singh and O’Gorman, 2013; Seeley and
741 Romps, 2015).

742 Using the the effective convective temperature profile, T_c to calculate
743 $\tilde{\theta}_s(T_c)$, also greatly reduces the horizontal spread in the mid- to lower tro-
744 posphere. Furthermore, we show that in the lower troposphere the vapor
745 buoyancy effect strongly conditions T , in ways that easily bias the interpre-
746 tation of $\tilde{\theta}_s(T)$ profiles. Our finding recasts work by Yang and Seidel (2020)
747 who has previously also emphasized how a large vapor buoyancy effect can

748 lead to 1.5 K horizontal temperature differences in the lower troposphere,
749 and explored the implications of this for radiative transfer. As convective
750 instability is often inferred from the profile of T , apparently unstable pro-
751 files may arise due to vertical gradients of water vapor (which condition the
752 gradients of T). Raymond and Flores (2016) defined an instability index
753 using the saturation moist entropy averaged over 1 km to 3 km minus that
754 over 5 km to 7 km. By basing this calculation on T , the dry tropics, i.e., non-
755 convecting areas, will appear more unstable due to an apparent decrease in
756 moist entropy, which arises from a disproportionate effect of water on the
757 temperature at lower levels. Using the effective convective temperature, T_c
758 as we define it, avoids this bias. Another consequence of the atmosphere
759 being generally dry is that estimating upper-tropospheric warming as be-
760 ing proportional to lower-tropospheric temperatures without accounting for
761 differences in the absolute humidity, will overstate the warming, because
762 the lapse rate in a dry atmosphere is often larger than that in a moist at-
763 mosphere due partly to the density effect of water vapor. To what extent
764 this might matter for controversies regarding the expected versus measured
765 upper-tropospheric warming remains to be evaluated.

766 To what extent the WTG holds in the tropical free-troposphere depends
767 on how one defines ‘weak’. The idea of a weak buoyancy, or density, gradient
768 is better founded, but even this is limited in its applicability. Already

769 poleward of 10° , we begin to see large departures from the assumption of
770 a weak density gradient in the mid-troposphere. Even across ocean basins
771 the density temperatures can vary substantially, as it does above and below
772 the lower middle and middle (400 hPa to 800 hPa) troposphere. The larger
773 deviations from the weak buoyancy gradient (WBG) approximation that
774 we note in the upper and lower troposphere are less evident in idealized
775 simulations, even within relatively large domain RCE studies. This suggests
776 that despite support from idealized studies of how the troposphere adjusts
777 to convective heating, an unqualified application of WTG (or WBG) and
778 the moist adiabat, while an attractive simplification, is not something that
779 can be taken for granted. Possible deviations from this balance need to be
780 evaluated for quantitative work.

781 Most of the key results from our analysis of ICON can be generalized
782 to other DYAMOND models and are also apparent in observed tropical
783 soundings. Among the models, however, differences are apparent in terms
784 of the vertical thermal structure. These may be a signature of differences
785 in their treatment of thermodynamic or microphysical processes, a question
786 that we are looking forward to investigating further.

787

Acknowledgements

788 We would like to thank the modeling groups participating in the DYA-
789 MOND and RCEMIP project. The DYAMOND and RCEMIP data as well
790 as further management were provided by the German Climate Computing
791 Center (DKRZ). DYAMOND project was supported through the projects
792 ESiWACE and ESiWACE2. The projects ESiWACE and ESiWACE2 have
793 received funding from the European Union’s Horizon 2020 research and in-
794 novation programme under grant agreements No 675191 and 823988. We
795 also would like to thank the University of Wyoming for providing access to
796 the upper atmospheric sounding data. We thank Steven Sherwood, Kerry
797 Emanuel, Claudia Stephan and Haile Xue for enlightening discussions, and
798 Tobias Becker for the internal review. The primary data and scripts used in
799 the analysis that may be useful in reproducing the results of this paper are
800 archived by the Max Planck Institute for Meteorology and can be obtained
801 by contacting publications@mpimet.mpg.de.

802 **References**

803 Baldauf, M., A. Seifert, J. Förstner, D. Majewski, M. Raschendorfer, and
804 T. Reinhardt, 2011: Operational convective-scale numerical weather
805 prediction with the cosmo model: Description and sensitivities.
806 *Monthly Weather Review*, **139(12)**, 3887–3905.

- 807 Bao, J., and S. C. Sherwood, 2019: The role of convective self-aggregation
808 in extreme instantaneous versus daily precipitation. *Journal of Ad-
809 vances in Modeling Earth Systems*, **11(1)**, 19–33.
- 810 Betts, A. K., 1982: Saturation point analysis of moist convective overturn-
811 ing. *Journal of the Atmospheric Sciences*, **39**, 1484–1505.
- 812 Betts, A. K., 1986: A new convective adjustment scheme. part i: Observa-
813 tional and theoretical basis. *Quarterly Journal of the Royal Meteo-
814 rological Society*, **112(473)**, 677–691.
- 815 Bolton, D., 1980: The computation of equivalent potential temperature.
816 *Monthly Weather Review*, **108(7)**, 1046–1053.
- 817 Bretherton, C. S., and P. K. Smolarkiewicz, 1989: Gravity waves, compen-
818 sating subsidence and detrainment around cumulus clouds. *Journal
819 of the Atmospheric Sciences*, **46(6)**, 740–759.
- 820 Charney, J. G., 1963: A Note on Large-Scale Motions in the Tropics. *Journal
821 of the Atmospheric Sciences*, **20(6)**, 607–609.
- 822 Dipankar, A., B. Stevens, R. Heinze, C. Moseley, G. Zängl, M. Giorgetta,
823 and S. Brdar, 2015: Large eddy simulation using the general circu-
824 lation model icon. *Journal of Advances in Modeling Earth Systems*,
825 **7(3)**, 963–986.

826 Folkins, I., and R. V. Martin, 2005: The vertical structure of tropical con-
827 vection and its impact on the budgets of water vapor and ozone.
828 *Journal of the Atmospheric Sciences*, **62**, 1560–1573.

829 Hohenegger, C., L. Kornblueh, D. Klocke, T. Becker, G. Cioni, J. F. Engels,
830 U. Schulzweida, and B. Stevens, 2020: Climate Statistics in Global
831 Simulations of the Atmosphere, from 80 to 2.5 km Grid Spacing.
832 *Journal of the Meteorological Society of Japan. Ser. II*, **98(1)**, 2020–
833 005–91.

834 Khairoutdinov, M. F., and D. A. Randall, 2003: Cloud resolving modeling of
835 the arm summer 1997 iop: Model formulation, results, uncertainties,
836 and sensitivities. *Journal of the Atmospheric Sciences*, **60(4)**, 607–
837 625.

838 Malardel, S., W. Nils, D. Willem, D. Michail, K. Christian, G. Mozdzynski,
839 M. Hamrud, and S. Piotr, 2016: A new grid for the ifs. *ECMWF*
840 *Newsletter*, (**146**).

841 Mapes, B. E., 1993: Gregarious tropical convection. *Journal of the Atmo-*
842 *spheric Sciences*, **50(13)**, 2026–2037.

843 Mapes, B. E., 2001: Water’s two height scales: The moist adiabat and the
844 radiative troposphere. *Quarterly Journal of the Royal Meteorological*
845 *Society*, **127(577)**, 2353–2366.

846 Mlawer, E. J., S. J. Taubman, P. D. Brown, M. J. Iacono, and S. A. Clough,
847 1997: Radiative transfer for inhomogeneous atmospheres: Rrtm, a
848 validated correlated-k model for the longwave. *Journal of Geophysi-
849 cal Research: Atmospheres*, **102(D14)**, 16663–16682.

850 Müller, S. K., and C. Hohenegger, 2020: Self-aggregation of convection in
851 spatially varying sea surface temperatures. *Journal of Advances in
852 Modeling Earth Systems*, **12**, e2019MS001698.

853 Prein, A. F., R. M. Rasmussen, K. Ikeda, C. Liu, M. P. Clark, and G. J.
854 Holland, 2017: The future intensification of hourly precipitation ex-
855 tremes. *Nature Climate Change*, **7(1)**, 48–52.

856 Putman, W. M., and S.-J. Lin, 2007: Finite-volume transport on various
857 cubed-sphere grids. *Journal of Computational Physics*, **227(1)**, 55–
858 78.

859 Putman, W. M., and M. Suarez, 2011: Cloud-system resolving simulations
860 with the nasa goddard earth observing system global atmospheric
861 model (geos-5). *Geophysical Research Letters*, **38(16)**.

862 Raymond, D. J., and M. M. Flores, 2016: Predicting convective rainfall over
863 tropical oceans from environmental conditions. *Journal of Advances
864 in Modeling Earth Systems*, **8(2)**, 703–718.

- 865 Romps, D. M., and Z. Kuang, 2010: Do undiluted convective plumes ex-
866 ist in the upper tropical troposphere? *Journal of the Atmospheric*
867 *Sciences*, **67(2)**, 468–484.
- 868 Satoh, M., T. Matsuno, H. Tomita, H. Miura, T. Nasuno, and S. Iga, 2008:
869 Nonhydrostatic icosahedral atmospheric model (nicam) for global
870 cloud resolving simulations. *Journal of Computational Physics*,
871 **227(7)**, 3486–3514.
- 872 Satoh, M., B. Stevens, F. Judd, M. Khairoutdinov, S.-J. Lin, W. M. Putman,
873 and P. Düben, 2019: Global Cloud-Resolving Models. *Curr Clim*
874 *Change Rep*, **5(3)**, 172–184.
- 875 Seeley, J. T., and D. M. Romps, 2015: Why does tropical convective avail-
876 able potential energy (cape) increase with warming? *Geophysical*
877 *Research Letters*, **42(23)**, 10,429–10,437.
- 878 Singh, M. S., and P. A. O’Gorman, 2013: Influence of entrainment on the
879 thermal stratification in simulations of radiative-convective equilib-
880 rium. *Geophysical Research Letters*, **40(16)**, 4398–4403.
- 881 Sobel, A. H., and C. S. Bretherton, 2000: Modeling tropical precipitation
882 in a single column. *Journal of Climate*, **13**, 4378–4392.
- 883 Stephan, C., C. Strube, D. Klocke, M. Ern, L. Hoffmann, P. Preusse, and

- 884 H. Schmidt, 2019: Gravity waves in global high-resolution simula-
885 tions with explicit and parameterized convection. *Journal of Geo-*
886 *physical Research: Atmospheres*, **124**, 4446–4459.
- 887 Stevens, B., M. Satoh, L. Auger, J. Biercamp, C. S. Bretherton, X. Chen,
888 P. Düben, F. Judt, M. Khairoutdinov, D. Klocke, C. Kodama,
889 L. Kornbluh, S.-J. Lin, P. Neumann, W. M. Putman, N. Röber,
890 R. Shibuya, B. Vanniere, P. L. Vidale, N. Wedi, and L. Zhou, 2019:
891 Dyamond: the dynamics of the atmospheric general circulation mod-
892 eled on non-hydrostatic domains. *Progress in Earth and Planetary*
893 *Science*, **6(1)**, 61.
- 894 Stevens, B., and A. P. Siebesma, 2020: *Clouds as Fluids* 35–73. Cambridge
895 University Press.
- 896 Wing, A. A., K. Emanuel, C. E. Holloway, and C. Muller, 2018a: *Convective*
897 *Self-Aggregation in Numerical Simulations: A Review* 1–25. Cham:
898 Springer International Publishing.
- 899 Wing, A. A., K. A. Reed, M. Satoh, B. Stevens, S. Bony, and T. Ohno,
900 2018b: Radiative–convective equilibrium model intercomparison
901 project. *Geosci. Model Dev.*, **11(2)**, 793–813.
- 902 Xu, K.-M., and K. A. Emanuel, 1989: Is the tropical atmosphere condition-
903 ally unstable? *Monthly Weather Review*, **117**, 1471–1479.

- 904 Yang, D., and S. D. Seidel, 2020: The incredible lightness of water vapor.
905 *Journal of Climate*, **0(0)**, null.
- 906 Zängl, G., D. Reinert, P. Rípodas, and M. Baldauf, 2015: The icon (icosahe-
907 dral non-hydrostatic) modelling framework of dwd and mpi-m: De-
908 scription of the non-hydrostatic dynamical core. *Quarterly Journal*
909 *of the Royal Meteorological Society*, **141(687)**, 563–579.

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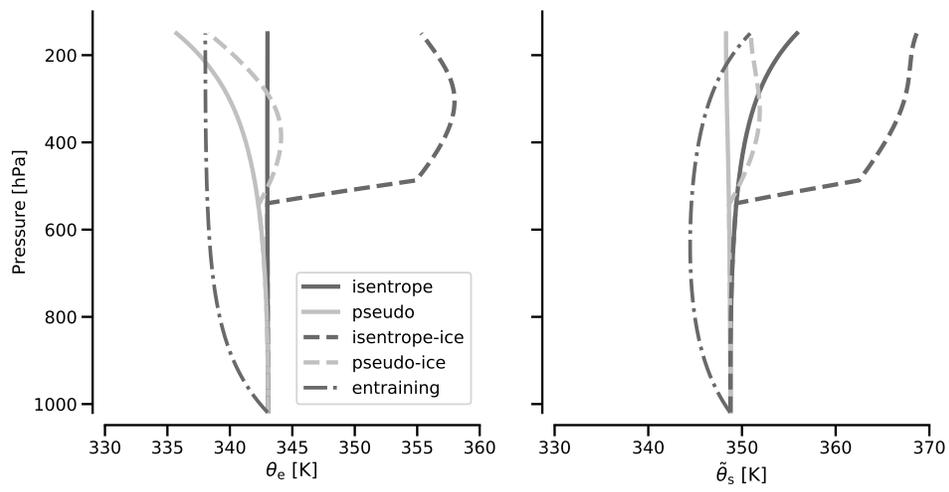


Fig. 1. Profiles of T derived from idealized processes plotted in θ_e coordinate (left) and $\tilde{\theta}_s$ coordinate (right).

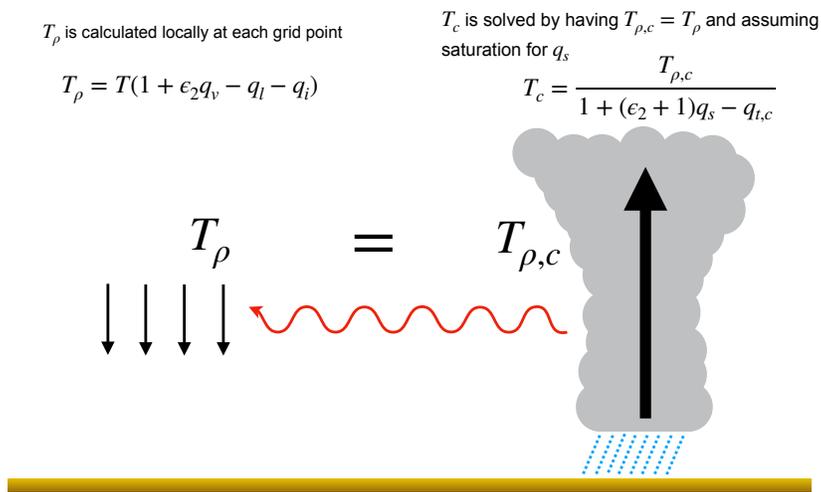


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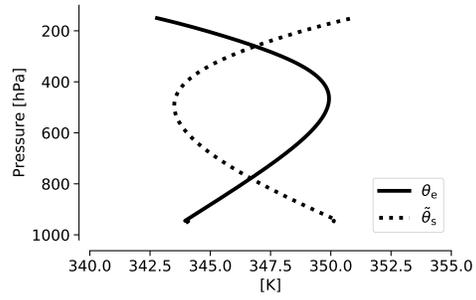


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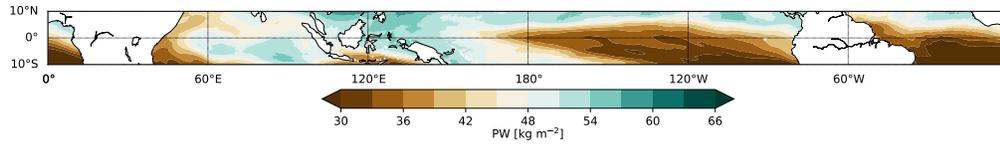


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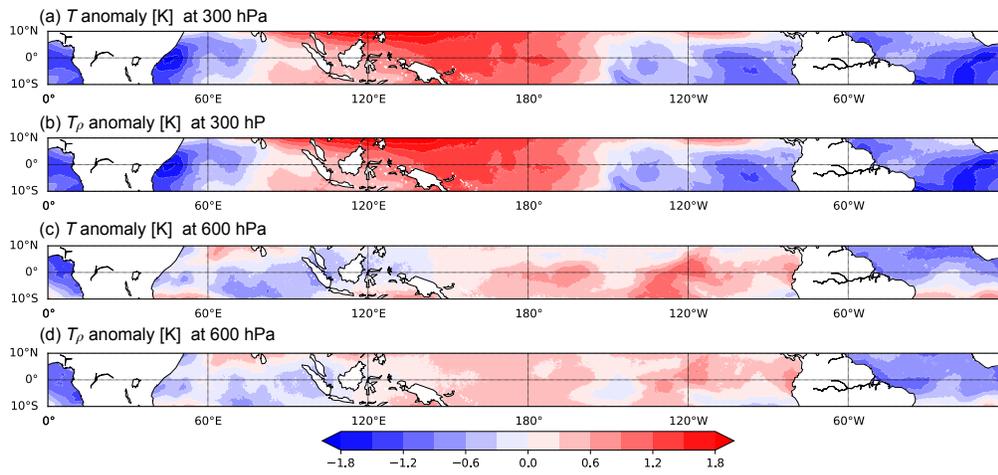


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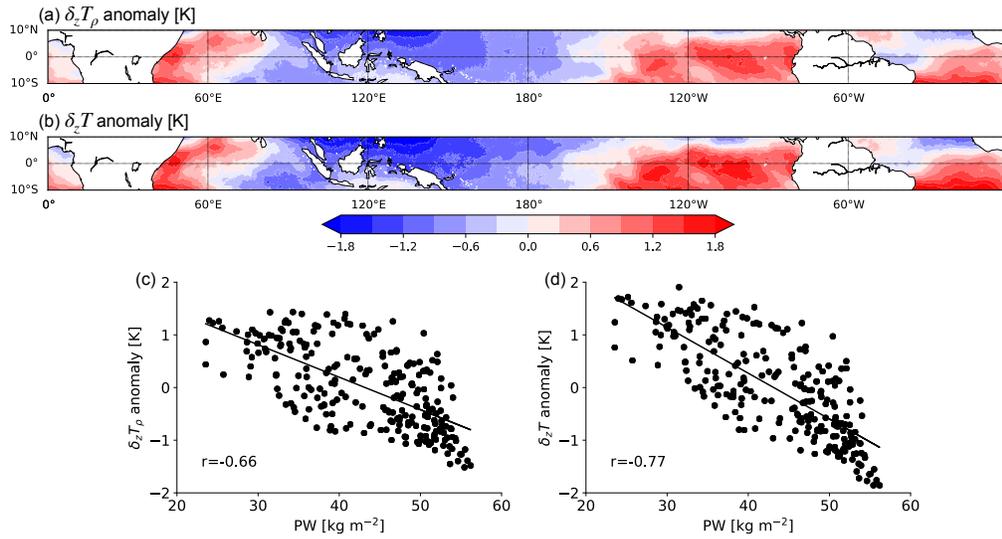


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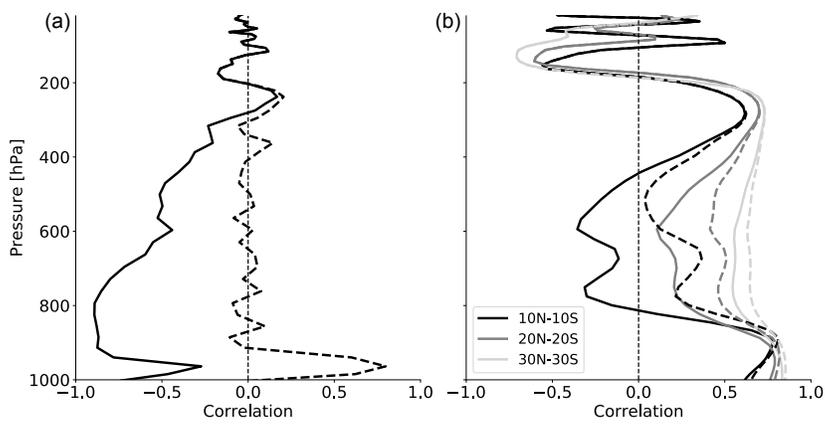


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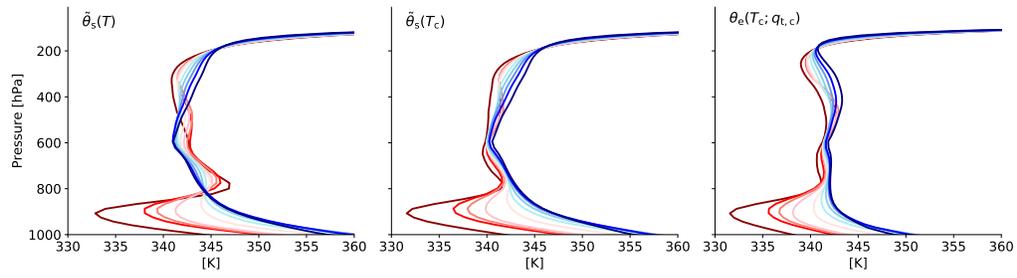


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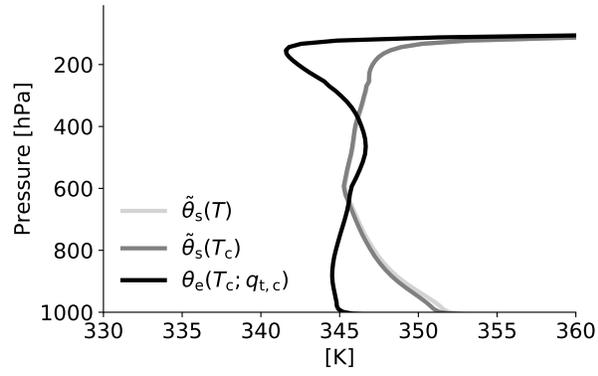


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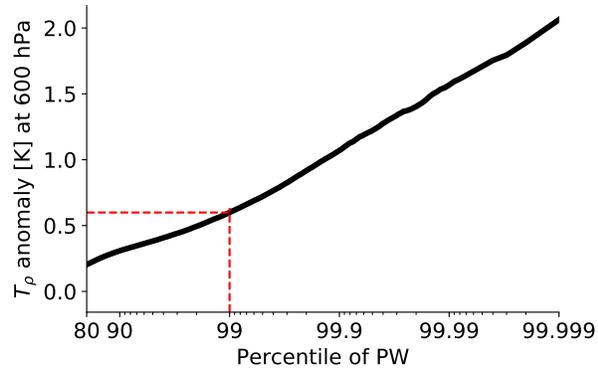


Fig. 10. T_ρ anomaly (relative to the domain-mean value) at 600 hPa as function of percentiles of PW. Red dashed lines are references corresponding to the 99th percentile of PW.

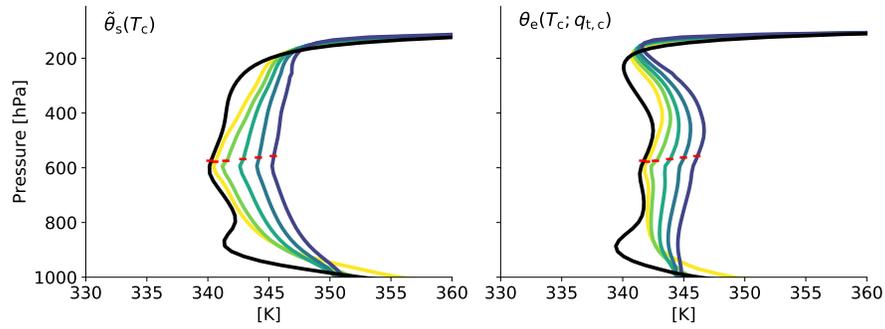


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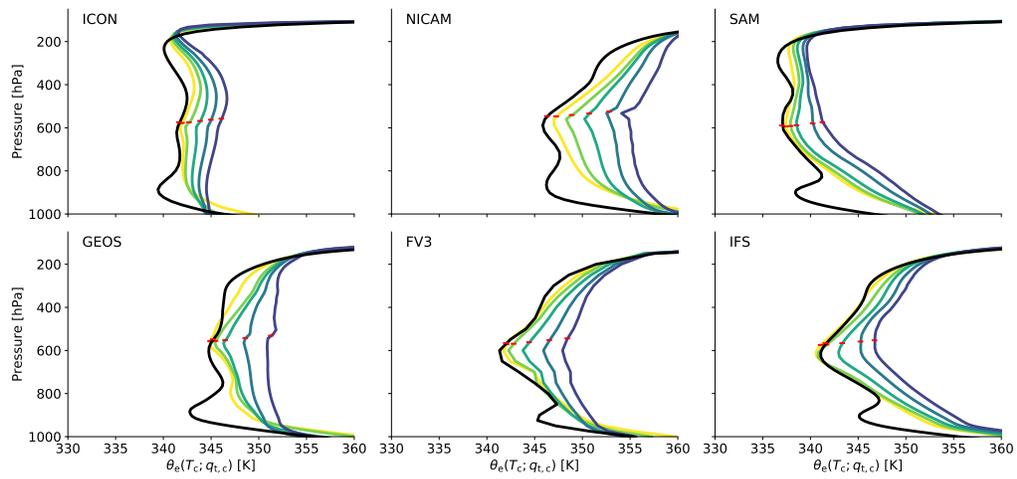


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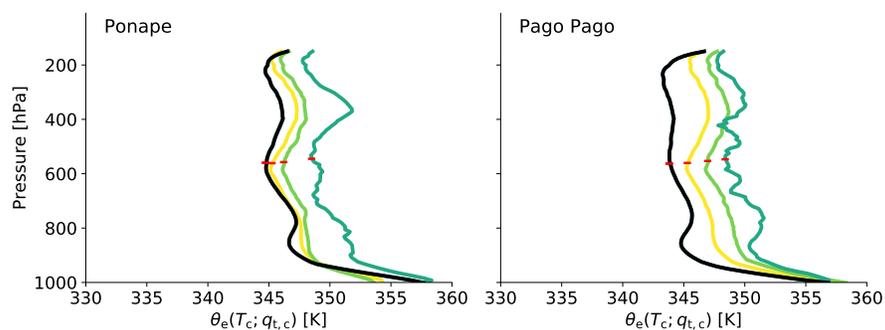


Fig. 13. Mean profiles of $\theta_e(T_c; q_{t,c})$ averaged over all grid points (black) and the extremely humid grid points (colors from yellow to green correspond to the 90th percentile to 99.9th percentile of PW) from two tropical soundings. Freezing levels are marked in red.

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Table 1. List of DYAMOND models whose output is used in this study.

Short name	References
ICON	Zängl et al. (2015)
NICAM	Satoh et al. (2008)
SAM	Khairoutdinov and Randall (2003)
FV3	Putman and Lin (2007)
GEOS	Putman and Suarez (2011)
IFS	Malardel et al. (2016)

Table 2. Thermodynamic coordinates.

Coordinate	Process	T_c from
$\tilde{\theta}_s(T)$	Pseudo-adiabatic	T
$\tilde{\theta}_s(T_c)$	Pseudo-adiabatic	$T_\rho _{q_t=q_s(T_c,p)}$
$\theta_e(T_c; q_{t,c})$	Isentropic	$T_\rho _{q_{t,c}=18 \text{ g kg}^{-1}}$