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1 **Convective self-aggregation and tropical cyclogenesis**

2 **under the hypohydrostatic rescaling**

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ABSTRACT

8 The behavior of rotating and non-rotating aggregated convection is exam-
9 ined at various horizontal resolutions using the hypohydrostatic, or Reduced
10 Acceleration in the VERTICAL (RAVE), rescaling. This modification of the
11 equations of motion reduces the scale separation between convective and
12 larger-scale motions, enabling the simultaneous and explicit representation
13 of both types of flow in a single model without convective parameterization.
14 Without the RAVE rescaling, a dry bias develops when simulations of non-
15 rotating radiative-convective equilibrium are integrated at coarse resolution
16 in domains large enough to permit convective self-aggregation. The rescal-
17 ing reduces this dry bias, and here it is suggested that the rescaling moistens
18 the troposphere by weakening the amplitude and slowing the group veloc-
19 ity of gravity waves, thus reducing the subsidence drying around aggregated
20 convection. Separate simulations of rotating radiative-convective equilibrium
21 exhibit tropical cyclogenesis; as horizontal resolution is coarsened without
22 the rescaling, the resulting storms intensify more slowly and achieve lower
23 peak intensities. At a given horizontal resolution, using RAVE increases peak
24 storm intensity and reduces the time needed for tropical cyclogenesis, effects
25 here suggested to be caused at least in part by the environmental moistening
26 produced by RAVE. Consequently, the RAVE rescaling has the potential to
27 improve simulations of tropical cyclones and other aggregated convection in
28 models with horizontal resolutions of $O(10-100 \text{ km})$.

29 **1. Introduction**

30 The representation of moist convection in numerical models of atmospheric flow is a problem
31 that has stymied the scientific community for decades. Computing power is typically insufficient to
32 provide the spatial resolutions needed to successfully simulate moist convective motions in model
33 domains large enough to represent planetary scale flow. At the same time, poor understanding of
34 the net effects of convective motions has prevented the development of unbiased approximations
35 of the subgrid-scale effects of moist convection; some argue that this sort of parameterization may
36 not even be possible (for a review see Arakawa 2004).

37 These issues are particularly vexing when attempting to represent organized convection having
38 horizontal scales on the order of 1-100 km, such as occurs in mesoscale convective systems and
39 tropical cyclones. Such circulations lie in the gap that is sometimes assumed to exist between
40 convective motions with horizontal scales of 0.1 - 1 km and the “large-scale” motions that can
41 be explicitly represented in global models with horizontal grid spacings on the order of 100 km.
42 While individual occurrences of organized convection can be simulated at extremely fine resolu-
43 tions because only short times and relatively small domains need to be represented, simulation
44 of the global distribution of organized convection is hindered by limited model resolution and
45 inadequate convective parameterization.

46 Study of the effect of climate change on the global distribution of tropical cyclones (TCs) has
47 been especially limited by these issues. Explicit representation of the $O(10\text{ km})$ -diameter TC eye-
48 wall is impossible at typical global climate model resolutions, so even the latest generation of those
49 models can only simulate “TC-like storms” (e.g. Camargo 2013; Merlis et al. 2013). Although the
50 space-time distribution of these TC-like storms is similar to the distribution of observed TCs, the
51 model storms are larger and weaker than observed TCs (e.g. Manabe et al. 1970; McBride 1984;

52 Vitart et al. 1997). Even when regional models of the western North Pacific and tropical Atlantic
53 were integrated at the relatively fine horizontal resolution of 18 km, the most intense simulated
54 storms would be classified in Saffir-Simpson category three (Knutson et al. 2007; Wu et al. 2014).
55 The question of how the characteristics of the most intense TCs (i.e. categories four and five) vary
56 with the global climate state thus cannot be answered directly by most global and even regional
57 numerical models. Downscaling methodologies have been developed in attempts to bypass this
58 problem, using grid-scale fields from coarse-resolution global models as inputs to statistical or
59 dynamical simulations of individual TCs (e.g. Emanuel et al. 2008; Bender et al. 2010; Zhao and
60 Held 2010; Fedorov et al. 2010; Villarini and Vecchi 2012; Knutson et al. 2013). However, it
61 seems fair to say that explicit representation of the most intense category of TCs in a global model
62 remains a much sought after goal of the atmospheric science community.

63 Previous studies as well as this work show that faithful representation of TC structure and in-
64 tensity requires model horizontal resolutions on the order of 1 km. For example, Gentry and
65 Lackmann (2010) found that storm intensity increased as horizontal grid spacing was reduced
66 from 8 km to 1 km, and they suggested that horizontal resolutions of 2-3 km are needed to resolve
67 the eyewall processes that are important for operational prediction. Other studies find a more am-
68 biguous dependence of storm intensity on horizontal resolution for grid spacings in the range of
69 1-5 km, and suggest that subgrid-scale parameterizations are at least as important as resolution for
70 such grid spacings (Fierro et al. 2009; Sun et al. 2013). But it seems clear that coarsening horizon-
71 tal resolution beyond 5-10 km greatly reduces the peak intensity achievable in simulated tropical
72 cyclones: Murakami and Sugi (2010) found that 20-km grid spacing produced a large underesti-
73 mate in the number of storms with intensities higher than Saffir-Simpson category 2. Peak storm
74 intensity generally decreases as horizontal resolution is further coarsened past 10-20 km, so that
75 typical global climate models, even at “high” resolutions of 25 or 50 km, do not simulate tropical

76 cyclones with intensities greater than category 2 or 3 (Walsh et al. 2013; Strachan et al. 2013).
 77 This conclusion is confirmed by various simulations with global and regional atmospheric models
 78 (Zhao et al. 2009; Knutson et al. 2008, 2013) and coupled global climate models (e.g. Gualdi et al.
 79 2008; Scoccimarro et al. 2011; Bell et al. 2013). A few global atmospheric models do produce
 80 TCs of higher intensity at 25-km resolution (e.g. Zarzycki and Jablonowski 2014), but this seems
 81 to depend on specifics of the convective parameterization used. The frequency and intensity of TCs
 82 simulated by global models with O(25-50 km) horizontal resolution is highly sensitive, sometimes
 83 in nonmonotonic and counterintuitive ways, to parameterizations of subgrid-scale physics and to
 84 the numerical damping used to suppress grid-scale noise (Zhao et al. 2012).

85 A new approach to the representation of moist convection in numerical models, proposed by
 86 Kuang et al. (2005, hereafter KBB), modifies the equations of motion to reduce the scale separation
 87 between convective and large-scale motions and thus allows explicit representation of both in the
 88 same model. This approach can be implemented and interpreted in multiple ways, but perhaps the
 89 simplest involves reducing the vertical acceleration of fluid parcels by introducing a factor $\gamma > 1$
 90 in the vertical momentum equation,

$$\gamma^2 \frac{Dw}{Dt} = -\frac{1}{\rho} \frac{\partial p}{\partial z} - g + F_z. \quad (1)$$

91 Here D/Dt is the material derivative, F_z is the vertical acceleration due to diffusion or other pro-
 92 cesses (typically acting on the subgrid scale), and other symbols have their usual meteorological
 93 meanings. This implementation, which KBB called Reduced Acceleration in the VERTICAL (RAVE),
 94 reduces the vertical velocities and increases the horizontal length scales of smaller convective mo-
 95 tions, making them closer in size to those of the unaltered large-scale, hydrostatic flow. RAVE has
 96 also been referred to as the hypohydrostatic rescaling because it artificially increases the inertia
 97 of vertical motions (Pauluis et al. 2006; Garner et al. 2007). Although the RAVE/hypohydrostatic

98 approach has received new attention in the past decade for its effects on moist convection, the
99 same modification of the vertical momentum equation was used years earlier in so-called quasi-
100 nonhydrostatic (QNH) models used for numerical weather prediction (MacDonald et al. 2000a;
101 Lee and MacDonald 2000; Browning and Kreiss 1986; Skamarock and Klemp 1994). These QNH
102 models were shown to be more numerically stable and resistant to small-scale error growth when
103 subjected to impulsive forcings such as moist convective heating or initialization with an out-of-
104 balance state. MacDonald et al. (2000b) showed that using (1) for the vertical momentum equation
105 suppressed gravity wave generation below a certain length scale, thus slowing the adjustment to
106 geostrophic or gradient wind balance while leaving Rossby waves and the large-scale response to
107 diabatic heating unchanged.

108 As discussed by KBB, the RAVE rescaling is equivalent to the Diabatic Acceleration and Rescal-
109 ing (DARE) approach, in which the planetary rotation rate is increased by a factor of γ , the plane-
110 tary radius is decreased by γ , and diabatic processes such as radiative and surface enthalpy fluxes
111 are increased by γ . The DARE approach shrinks the time and space scales of the large-scale
112 dynamics (e.g. the Rossby deformation radius), bringing them closer to the scales of convective
113 motions. Yet another mathematically equivalent approach is known as the Deep Earth rescaling,
114 in which the gravitational acceleration is decreased and the vertical coordinate (z) is increased in
115 scale by the factor γ (Pauluis et al. 2006). Although all of these treatments are mathematically
116 identical, here we use the RAVE approach because it has perhaps the simplest physical interpreta-
117 tion and is easily implemented in numerical models. For more background and history on RAVE
118 and equivalent rescalings, see KBB, Pauluis et al. (2006), and references therein.

119 RAVE and equivalent rescalings have been used to study a number of phenomena involving
120 moist convection, but to our knowledge have not been used for studying TCs or tropical cycloge-
121 nesis. KBB presented preliminary results from an equatorial beta-plane simulation of the tropo-

122 spheric general circulation forced by an equatorial sea surface temperature (SST) maximum, with
123 some emphasis on the spectrum of convectively coupled equatorial waves. Garner et al. (2007)
124 conducted global aquaplanet simulations with large RAVE factors (i.e. $\gamma \geq 100$) and found that
125 the extratropical circulation was largely unaltered by use of even these extreme rescalings; they
126 noted that use of RAVE with $\gamma \sim 3$ and horizontal resolutions on the order of 10 km may provide
127 a promising alternative to convective parameterization. Boos and Kuang (2010) used RAVE in an
128 equatorial beta-plane model to examine the mechanisms involved in tropical intraseasonal vari-
129 ability during boreal summer, with horizontal resolutions of about 30 km and $\gamma = 15$. Ma et al.
130 (2014) examined the influence of topography on the South Asian monsoon using RAVE in a global
131 model (on a sphere) without convective parameterization, at a horizontal resolution of 40 km with
132 $\gamma = 10$.

133 The use of RAVE was criticized by Pauluis et al. (2006), who argued that for the same compu-
134 tational cost, coarse-resolution integrations without convective parameterizations more accurately
135 reproduced the statistics of deep moist convection than integrations with RAVE. They based this
136 argument on simulations of radiative-convective equilibrium in doubly-periodic domains in which
137 the horizontal grid spacing was varied while the number of model grid points was held constant
138 (thus larger domains were used at coarser resolutions). They simulated a 16 day period and ana-
139 lyzed the last 8 days, and found that tropospheric specific humidity decreased as resolution was
140 coarsened and that RAVE enhanced the amplitude of this dry bias. In contrast, here we conduct
141 much longer integrations holding domain size constant, and find that instead of enhancing a dry
142 bias, RAVE actually reduces the dry bias caused by use of coarse resolution in simulations of
143 radiative-convective equilibrium. We attribute this contrasting result to our use of a fixed domain
144 size and longer simulation, and provide a more detailed comparison with the results of Pauluis
145 et al. (2006) in an appendix.

146 The main goal of this paper is to examine the effects of RAVE on convective self-aggregation,
147 both with and without rotation, in doubly periodic domains large enough to contain one TC. We
148 present possible mechanisms by which RAVE influences the humidity field in radiative-convective
149 equilibrium and in tropical cyclogenesis. This work is in some ways a methodological study
150 showing that RAVE can compensate for some of the deleterious effects of coarse resolution on
151 simulated TC intensity. But it also provides a closer look at the ways in which RAVE alters the
152 interaction of organized moist convection with its environment. In particular, we apply some of
153 the ideas of MacDonald et al. (2000b) for the effect of RAVE on the internal wave field to the topic
154 of moist convection and the spatial distribution of moisture.

155 The next section of this paper presents details of the numerical model used in this work. Subse-
156 quent sections show results from simulations in doubly periodic domains, both with and without
157 rotation. The paper ends with a summary and discussion of the method's possible value for future
158 studies of TCs and aggregated convection.

159 **2. Model details**

160 This study uses the System for Atmospheric Modeling (SAM) version 6.3 by Khairoutdinov and
161 Randall (2003), which integrates the anelastic equations of motion in Cartesian coordinates. The
162 model has prognostic equations for liquid/ice water moist static energy, total precipitating water,
163 and total nonprecipitating water. A five-class bulk microphysics scheme diagnoses rain, snow,
164 graupel, cloud water, and ice. Numerous studies of radiative-convective equilibrium states have
165 used this model, including some examinations of convective self-aggregation and spontaneous TC
166 genesis (e.g. Bretherton et al. 2005; Muller and Held 2012).

167 We conducted integrations with and without rotation, all in domains of square, doubly periodic
168 horizontal dimensionality. Integrations without rotation (i.e. $f = 0$) had domain widths of either

169 96 km, 384 km, or 768 km, referred to hereafter as the 1°-, 4°-, and 8°-wide domains, respectively.
170 Integrations with rotation used domain widths of 1280 km and a Coriolis parameter equal to that
171 found at 20°N on Earth. These domain widths may inhibit the size of simulated TCs and artificially
172 confine the gravity waves that propagate away from organized convection. Indeed, Chavas and
173 Emanuel (2014) found that the size of a simulated axisymmetric TC increased with domain size
174 until the diameter of the domain was approximately 6000 km. Unfortunately, computational limits
175 prevented use of larger domains in this work. Nevertheless, our chosen domain width of 1280 km
176 is larger than that used in most previous studies of convective aggregation and spontaneous TC
177 genesis (e.g. Bretherton et al. 2005; Muller and Held 2012; Wing and Emanuel 2013). Also, since
178 individual occurrences of aggregated convection do not exist in isolation in the real world, use of
179 larger idealized domains may not better represent reality.

180 All integrations used 64 vertical levels with vertical grid spacing ranging from 80 m near the
181 surface to 400 m in the bulk of the troposphere and 1.2 km near the rigid lid at 27 km. To reduce
182 gravity wave reflection and resonance, Newtonian damping was applied to wind, temperature, and
183 water vapor in the top 30 percent of the domain with a time scale that decayed linearly from 2
184 minutes at the top to 2 hours at the bottom of this sponge layer. Time steps ranged from 4 to 50
185 seconds, depending on model resolution, with SAM automatically halving the time step when high
186 Courant numbers were achieved.

187 All integrations used an oceanic lower boundary condition with an SST of 301 K and wind-
188 dependent surface sensible and latent heat fluxes computed using a bulk surface flux formula with
189 a prescribed minimum surface wind speed of 1 m s^{-1} . Fully interactive radiation was represented
190 using parameterizations from the National Center for Atmospheric Research (NCAR) Community
191 Climate Model (CCM) version 3.5 (Kiehl et al. 1998) with radiative fluxes calculated about once
192 every 15 minutes. Insolation was set to its perpetual, diurnal mean equinox value at the equator

193 for the nonrotating runs and at 20°N for the rotating runs. Ocean surface albedo depends on
194 solar zenith angle and is about 0.08 for these integrations. Note that although use of interactive
195 radiation does not alter the SST, moisture-dependent radiation has been shown to be essential
196 for achieving self-aggregation of convection in SAM (Bretherton et al. 2005; Wing and Emanuel
197 2013; Emanuel et al. 2013). All integrations were initialized using a horizontally homogeneous
198 tropical mean sounding, and symmetry was broken by adding white noise to the dry static energy
199 field in the lowest five model levels with amplitude decreasing from 0.1 K at the lowest level to
200 0.02 K at the fifth level. Simulations with rotation were initialized using a moister mean sounding
201 for reasons discussed below.

202 The RAVE rescaling was used in many integrations, with values of γ ranging from 1 to 16 (γ
203 = 1 represents no rescaling). A Smagorinsky-type closure was used to represent subgrid-scale
204 turbulence, with the parameterized stresses scaled by γ to account for the fact that RAVE alters the
205 aspect ratio of the resolved eddies (see discussion in appendix of Pauluis et al. 2006). We did not
206 directly modify microphysical processes when RAVE was used in our simulations, consistent with
207 the methodology used in Pauluis et al. (2006). In particular, we did not scale the terminal velocity
208 of falling condensate by γ .

209 **3. Results for non-rotating domains**

210 We first present results illustrating the effects of horizontal resolution and RAVE rescaling on
211 non-rotating radiative-convective equilibrium. For domains larger than a few hundred kilometers
212 in width, moist convection evolves over tens of days to an aggregated state consisting of a moist
213 precipitating cluster surrounded by a dry nonconvecting region, as in our simulations conducted
214 in 8°-wide domains (Figs. 1b, c). This self-aggregation of moist convection has been explored in
215 detail in previous studies, and has been shown to require a minimum domain size of about 200 km

216 and to be caused by feedbacks between tropospheric moisture and radiation (e.g. Muller and Held
217 2012; Wing and Emanuel 2013). Consistent with those studies, we find that the smallest domain
218 (1° wide) has convective activity and a moisture field that is horizontally homogeneous in the time
219 mean (Fig. 1a).

220 Models integrated at coarser resolutions undergo convective aggregation more quickly and have
221 lower domain-mean moisture content than integrations conducted at finer resolutions. This can be
222 seen in the instantaneous distribution of precipitable water (PW) 50 days after model initializa-
223 tion, which has a drier and more horizontally extensive non-convecting region in the integration
224 conducted at 16 km horizontal resolution than in the standard 2 km-resolution run (Fig. 1c). The
225 16 km-resolution integration also took about half the time to aggregate as the 2 km-resolution in-
226 tegration (Fig. 2). While the dependence of aggregation time on horizontal resolution is generally
227 monotonic in the integrations shown here and in others not shown, the dependence of domain-
228 mean PW on horizontal resolution is less regular. For instance, an integration conducted at 8 km
229 resolution exhibits nearly the same equilibrium PW as the 2 km run, although oscillations in the
230 PW field in both runs make comparison difficult. Similar oscillations in the PW field were seen
231 in some of the simulations of Bretherton et al. (2005, their Fig. 5a), and are associated with vari-
232 ations in the size of the moist region. But in general, coarser resolutions typically produce a drier
233 domain. The 16 km-resolution integration equilibrates with a domain mean PW several mm lower
234 than that of the 2 km-resolution integration. This general pattern can be seen in vertical profiles
235 of specific humidity in this 8° -wide domain and in a 4° -wide domain (Fig. 3). Coarse resolution
236 runs also typically have a colder troposphere and a warmer stratosphere than fine resolution runs.
237 The dry bias may cause at least some of the temperature bias, because one expects a tropospheric
238 cooling and stratospheric warming as the longwave emissivity of the column decreases. However,

239 the warmer stratosphere is also consistent with a simple reduction in the height of the tropopause,
240 as occurs during a tropospheric cooling, combined with an unchanged stratospheric lapse rate.

241 Use of the RAVE rescaling generally produces a moister equilibrium state. The dry bias seen
242 in the equilibrium state of runs using the 4° -wide domain at 12 km resolution is nearly eliminated
243 when a rescaling factor of $\gamma = 3$ is used at the same horizontal resolution (Fig. 4a); there is a moist
244 bias near the surface and a weak dry bias above that nearly cancel in a vertical integral. Increasing γ
245 to 6 produces a moist bias that is roughly the same amplitude as the original dry bias (as a reminder,
246 all biases are assessed relative to integrations using the same domain size at 2 km resolution with
247 $\gamma = 1$). A further increase of γ to 12 produces a humidity profile that looks more like that obtained
248 for $\gamma = 3$. This nonmonotonic dependence of the moisture field on γ is associated with the system
249 transitioning to a non-aggregated state after roughly 120 days of model time for $\gamma = 6$ but not for
250 the larger value of $\gamma = 12$. Given that the combination of a domain size of 4° and an SST of 301
251 K lies near a threshold in the parameter space for self-aggregation (e.g. Wing and Emanuel 2013;
252 Muller and Held 2012), it is perhaps not surprising that some combinations of parameters produce
253 runs that occasionally slip into a moister, non-aggregated state. These transitions back to a non-
254 aggregated state do not occur in any simulations we conducted using the larger domain width of
255 8° . For example, for integrations with 16 km resolution and an 8° -wide domain, increasing γ from
256 1 to 8 and then to 16 produced a monotonic moistening of the troposphere (Fig. 4c). The RAVE
257 rescaling also produces a temperature bias that is of similar magnitude but opposite sign to that
258 seen in coarse resolution runs without the rescaling: RAVE creates a warm bias in the lower to
259 middle troposphere and a cold bias in the upper troposphere and lower stratosphere.

260 In summary, coarse resolution integrations without RAVE (i.e. $\gamma = 1$) produce a troposphere that
261 is too dry and too cold compared to fine resolution simulations. Convective self-aggregation also
262 occurs too quickly in coarse resolution simulations. When coarse resolution models are integrated

263 with moderate values of γ , the dry bias is reduced to give PW values near those seen in the fine
264 resolution control runs, although RAVE does typically produce an overly strong moistening of
265 the near-surface air together with a more moderate reduction of the dry bias at higher altitudes.
266 Previous studies (e.g. Pauluis et al. 2006) arguing that RAVE does not reduce and may even am-
267 plify the dry bias seen in coarse resolution runs may have failed to account for the aggregation of
268 convection that occurs as domain size is increased. In an appendix we present additional model
269 integrations that allow for better comparison with that previous work.

270 **4. A hypothesis for the effect of RAVE on aggregated convection**

271 We now present a mechanism that may explain how RAVE moistens the troposphere in simu-
272 lations of radiative-convective equilibrium. Previous arguments for the influence of RAVE on the
273 moisture field have taken a local view in which RAVE slows convective overturnings, with these
274 overturnings directly influencing the moisture field through vertical advection (e.g. Pauluis et al.
275 2006). However, such arguments are relevant only to regions in which convection is active, and
276 when convection has aggregated the domain-mean humidity is dominated by values in the non-
277 convecting region. The key issue, then, is how RAVE alters humidity in the non-convecting region
278 outside a convective cluster.

279 Locations outside of active convection (but within a deformation radius) are known to have
280 temperatures set by spreading gravity wave-like disturbances that produce net subsidence and
281 adiabatic warming as they pass (e.g. Bretherton and Smolarkiewicz 1989; Nicholls et al. 1991;
282 Mapes 1993; Cohen and Craig 2004). The subsidence produced by these buoyancy bores also dries
283 areas around the convecting region by advecting dry air downward from the upper troposphere.
284 At the same time, shallower and slower-moving buoyancy bores have been argued to lift and
285 destabilize the environment near the original convection. Thus, while properties of the convecting

286 region might be set primarily by the convective motions themselves, properties of the environment
 287 around an actively convecting region are controlled indirectly via the gravity waves that emanate
 288 from the convection. Here we discuss how the modification of the internal wave field by RAVE,
 289 which was studied by MacDonald et al. (2000b) in the context of numerical weather prediction,
 290 might alter humidity and subsequent convection around a convecting cluster.

291 Skamarock and Klemp (1994) showed that RAVE modifies the dispersion relation for linear,
 292 Boussinesq, nonhydrostatic gravity waves to be

$$v^2 = \frac{f^2 m^2 + N^2 k^2}{\gamma^2 k^2 + m^2}, \quad (2)$$

293 where m is the vertical wavenumber, k is the total horizontal wavenumber, and v is frequency.
 294 Anticipating results that will be presented in the next section, we have included rotation using an
 295 f -plane approximation. The horizontal phase speed and group velocity are, respectively,

$$c_x = \frac{N}{\sqrt{\gamma^2 k^2 + m^2}} \sqrt{1 + \frac{f^2 m^2}{N^2 k^2}} \quad (3)$$

$$c_{gx} = \frac{m^2 \left(N - \frac{f^2 \gamma^2}{N} \right)}{(\gamma^2 k^2 + m^2)^{\frac{3}{2}} \sqrt{1 + \frac{f^2 m^2}{N^2 k^2}}}. \quad (4)$$

296 The use of RAVE thus decreases both the horizontal phase speed and group velocity of gravity
 297 waves. If $\gamma \lesssim 10$, then $f\gamma \ll N$ for typical atmospheric values of f and N , and the group velocity
 298 is well approximated by

$$c_{gx} \simeq \frac{Nm^2}{(\gamma^2 k^2 + m^2)^{\frac{3}{2}} \sqrt{1 + \frac{f^2 m^2}{N^2 k^2}}}. \quad (5)$$

299 The coefficient by which RAVE reduces the group velocity is then

$$\frac{c_{gx,RAVE}}{c_{gx}} = \left(\frac{k^2 + m^2}{\gamma^2 k^2 + m^2} \right)^{3/2} \quad (6)$$

300 which is valid without the assumption of hydrostatic balance. This ratio is plotted in Fig. 5a for
 301 a vertical wavelength of 28 km, $N = 0.01 \text{ s}^{-1}$, and f at 30° latitude. These results are consis-
 302 tent with the frequency response found by MacDonald et al. (2000b), but here we focus on wave

303 speeds to better connect with previous work showing that these wave speeds set the rate at which
304 subsidence warming spreads away from a pulse of convection (e.g. Bretherton and Smolarkiewicz
305 1989; Mapes 1993; Cohen and Craig 2004). MacDonald et al. (2000b) also showed that the am-
306 plitude of the vertical velocity response to an initial temperature perturbation or a heating impulse
307 is proportional to the frequency ν and so is also reduced by RAVE. For this reason we also plot
308 the ratio by which RAVE reduces the frequency, ν_{RAVE}/ν , in Fig. 6.

309 As expected, the use of RAVE has the largest effect on nonhydrostatic waves (i.e. those with
310 horizontal wavelengths shorter than 100 km), for which the group velocity scales like γ^{-3} and the
311 amplitude scales like γ^{-1} . But RAVE also substantially reduces the horizontal group velocity and
312 amplitude of longer waves for which the hydrostatic dispersion relation would, without RAVE, be
313 a good approximation. For example, the group velocity and frequency (which is proportional to
314 amplitude) are reduced to about 60% and 80% of their standard values, respectively, for a horizon-
315 tal wavelength of 200 km and the moderate value of $\gamma = 5$. RAVE alters c_{gx} and ν less for higher
316 vertical wavenumbers, which is illustrated by evaluating (6) and (2) for a vertical wavelength of
317 14 km (Figs. 5b and 6b).

318 Our core argument is that RAVE slows the geostrophic adjustment process that returns the envi-
319 ronment around episodic convective heating to a balanced state, and this slower adjustment process
320 includes weaker subsidence that produces less vertical advective drying of the environment. The
321 slowing of the geostrophic adjustment process is accomplished through a reduction in the horizon-
322 tal speed and amplitude of inertia-gravity waves, and this reduction acts more strongly on waves
323 with longer vertical wavelengths and shorter horizontal wavelengths, as detailed by previous au-
324 thors (Skamarock and Klemp 1994; MacDonald et al. 2000b) and discussed above. In particular,
325 the deeper waves that suppress remote convection are more strongly slowed and weakened by
326 RAVE, which will reduce static stability and subsidence drying far from the original convection

327 and enhance static stability and drying near the convection. The shallow internal waves hypothe-
328 sized to be responsible for initiating convection via low-level lifting adjacent to an initial convec-
329 tive disturbance (e.g. Mapes 1993) are less affected by RAVE, so they can continue to propagate
330 into and initiate convection in the far field. All of these effects act to encourage future convection
331 in the far-field environment and to suppress it near the original convection, reducing the horizontal
332 variance of humidity and convection in the domain.

333 This is analogous to the “rotational trapping” of gravity waves, which Liu and Moncrieff (2004)
334 argued makes convective clustering less likely at higher magnitudes of the Coriolis parameter (see
335 also Bretherton et al. 2005). One notable difference is that rotation provides an inherent length
336 scale in the Rossby deformation radius, while RAVE alters the rate at which the geostrophic ad-
337 justment occurs. Given a sufficiently long time after an episode of convective heating, RAVE
338 would thus make little to no difference in the final adjusted state. However, we are working with
339 a radiative-convective equilibrium state that is constantly destabilized by radiative cooling and
340 surface heat fluxes, and so hypothesize that reducing the rate at which the environment around a
341 convecting cluster warms will foster convection in that environment by reducing its static stability.
342 We also hypothesize that reducing the rate of subsidence drying will result in a moister environ-
343 ment because that environment is constantly moistened by eddy fluxes of moisture, perhaps mostly
344 due to vertical transports by shallow convective motions.

345 One caveat is that the convective heating that initiates any geostrophic adjustment is itself af-
346 fected by RAVE in ways that may modify the above arguments. RAVE increases the horizontal
347 scale of convective updrafts, and so the modified convection may excite longer wavelengths of
348 gravity waves. This would compensate for the reduction in wave speed that occurs at a given wave-
349 length because, in the absence of rotation, the transformation $k \rightarrow k/\gamma$ results in $c_{gx,RAVE} \rightarrow c_{gx}$
350 (with rotation, the group velocity still decreases as γ increases, but the functional form is different

351 and this would not be relevant to our results in non-rotating domains). However, the frequency,
352 and thus the amplitude, of the wave response continues to scale like γ^{-1} under this transformation:

$$v^2 \rightarrow \frac{N^2 k^2}{\gamma^2 (k^2 + m^2)} \quad (7)$$

353 where we have again neglected rotation. So even if the broadening of convective updrafts by RAVE
354 resulted in no net change in the group velocity of the convectively excited gravity waves, we would
355 still expect a reduction in wave amplitude. Nevertheless, it is important to emphasize that many
356 of the above arguments neglect potentially important feedbacks. One of these is the radiative
357 feedback of the modified humidity field, which one might expect to counter some of the effect of
358 RAVE on the environmental temperature field (e.g. less subsidence warming and drying produces
359 a moister atmosphere that cools less efficiently via radiation, albeit with some dependence on the
360 vertical moisture profile). This is one reason why we have not attempted to argue that the effects of
361 RAVE should be apparent in the time-mean temperature field. Furthermore, any simple horizontal
362 average of temperature (such as that presented in Fig. 4) would by construction include the near-
363 field increase and far-field decrease in static stability hypothesized to be induced by RAVE.

364 Another possibility is that RAVE may influence shallow convective motions that moisten the
365 lower troposphere. RAVE is expected to increase the horizontal scale of convective motions in
366 general, which might allow shallow convection to be better resolved at a given horizontal resolu-
367 tion. Pauluis and Garner (2006) attributed the dry bias that occurs in coarse-resolution models of
368 radiative-convective equilibrium to the inability of those simulations to represent the vertical mix-
369 ing caused by shallow clouds. While shallow turbulence occurs on sub-kilometer length scales
370 that cannot be resolved at any of the grid spacings used here, substantial vertical mixing is also
371 generated by cold pools with length scales of a few to tens of kilometers (Moeng et al. 2009). We
372 find that the variance of eddy vertical velocity at 1 km altitude decreases greatly as resolution is

373 coarsened in our 8° -wide domain (Fig. 7). The enhanced variance that occurs for grid spacings
374 finer than 16 km occurs entirely on length scales smaller than 16 km, as indicated by the fact that
375 there is no change in variance as a function of resolution when w is block-averaged to 16×16
376 km before calculating its eddy variance. But the more salient point is that RAVE does not increase
377 the variance of eddy vertical velocity as γ is increased from 1 to 16 at 16-km resolution. So these
378 results are consistent with the idea that vertical mixing by shallow convection is inhibited at coarse
379 resolutions, but they do not support the idea that RAVE moistens the troposphere by amplifying
380 shallow overturnings.

381 **5. Results for rotating domains**

382 *a. Effect of resolution*

383 We now examine the spontaneous cyclogenesis that occurs in simulations of radiative-convective
384 equilibrium in rotating domains. Emanuel and Nolan (2004) and Bretherton et al. (2005) showed
385 that when simulations of radiative-convective equilibrium are performed on an f -plane, convec-
386 tion self-aggregates and evolves into a TC. This behavior is reproduced in our simulations using
387 domains 1280 km wide with the Coriolis parameter equal to that at 20°N . When these simulations
388 were conducted at 1 km horizontal resolution, cyclogenesis took roughly 30 days to occur and
389 the simulated TC occupied about half the domain with a drier, non-convecting region occupying
390 the other half (Figs. 8a, 9a). Integrations conducted at the coarser horizontal resolutions of 5 and
391 8 km took considerably longer — typically 40 to 50 days — to undergo cyclogenesis. There is
392 a fair amount of variability in the time needed for cyclogenesis at a particular resolution; these
393 integrations were initialized with small-amplitude random noise in the low-level dry static energy
394 field, and small ensembles of integrations (4 ensemble members at each resolution) reveal that the

395 time needed to produce a category 1 TC can vary by tens of days at a given resolution. But there
396 is a clear delay of cyclogenesis as resolution is coarsened from 1 to 5 and then to 8 km (Fig. 10a).
397 As resolution is further coarsened to 16 km, the time needed to achieve category 1 intensity actu-
398 ally decreases, but the cyclone only barely achieves that intensity and the intensification process
399 lacks the abrupt character seen in the fine-resolution runs. Zhao et al. (2012) found that the fre-
400 quency of TC genesis changed nonmonotonically as a horizontal cumulus mixing rate in a GCM
401 was increased, although it is unclear whether this has any relation to the resolution-dependence
402 illustrated here.

403 The delay of cyclogenesis at coarser resolutions contrasts sharply with the effect of resolution on
404 self-aggregation time in non-rotating domains, where coarser resolutions produced faster aggre-
405 gation (e.g. Fig. 2). Indeed, convective self-aggregation and tropical cyclogenesis seem to be two
406 separate processes that take place on different time scales. To better illustrate this, we initialized
407 all rotating simulations (including those discussed above) with a moister sounding having precip-
408 itable water of over 50 mm (integrations without rotation were initialized with a sounding having
409 PW of about 37 mm). Use of this moister sounding does not qualitatively change any of the results
410 presented here, as was confirmed by repeating all runs with the drier initial sounding, but allows
411 for better illustration of the initial convective self-aggregation. Fig. 8b shows that the initial de-
412 crease of PW occurs more rapidly at coarser resolutions, as it did in the non-rotating simulations.
413 This domain-mean drying accompanies the convective self-aggregation process, as illustrated in
414 the previous section and discussed by Bretherton et al. (2005), and indicates that the time needed
415 for self-aggregation decreases monotonically as resolution is coarsened. The faster aggregation
416 at coarser resolutions is consistent with the larger peak surface wind speeds achieved at coarser
417 resolutions in the first 15 days of model time (Fig. 8a). At 1 km resolution, the self-aggregation
418 time scale is similar to the time needed for cyclogenesis, so that it seems like the two might occur

419 simultaneously as part of a single process. But at resolutions of 5 and 8 km, only about 20 days is
420 needed to form a single moist cluster surrounded by a dry region, while about 30 additional days
421 are needed for this cluster to undergo intensification to a category 1 TC. Once peak surface wind
422 speeds of almost 20 m s^{-1} are achieved (i.e. tropical storm intensity), the increase to peak intensity
423 occurs quite rapidly — within about 5 days — and has greater rapidity at finer resolutions. The PW
424 and surface wind speeds thus suggest the existence of multiple time scales: a 20-day time scale
425 associated with the initial convective aggregation and domain-mean drying, a subsequent 10-40
426 day time scale associated with the formation of a tropical storm, and a 5-day time scale of rapid
427 intensification of the storm to hurricane strength (these particular numbers might vary for different
428 basin sizes, initial soundings, etc.). This is consistent with the idea that a “preconditioning” period
429 exists prior to TC genesis, during which there is a moistening and cooling of the lower troposphere
430 (Bister and Emanuel 1997) and/or diabatic production of low-level potential vorticity anomalies
431 (Hendricks et al. 2004).

432 The peak intensity achieved by TCs decreased as resolution was coarsened. As the horizontal
433 grid spacing of the rotating simulations was increased from 1 to 16 km, the peak intensity fell
434 monotonically from Saffir-Simpson category 5 to category 1 (Figs. 8a and 10). The peak intensity
435 achieved for each horizontal resolution exhibited less variability among ensemble members than
436 did the time needed to reach category 1 intensity. These results are consistent with the propensity
437 for coarse resolution regional models to simulate TCs with an intensity distribution that peaks at
438 category 2-3 (e.g. Wu et al. 2014), and for global models with $O(100 \text{ km})$ horizontal grid spacing
439 to simulate only “TC-like” vortices. More generally, the peak intensity of simulated TCs has
440 been shown to decrease as model horizontal resolution is coarsened, although most studies of this
441 effect have used realistic initial and boundary conditions to simulate an observed TC. Gentry and
442 Lackmann (2010) and Sun et al. (2013) found that peak intensity was reduced as grid spacing

443 increased from 1 to 8 km in simulations of Hurricane Ivan (2004) and Typhoon Shanshan (2006),
444 respectively. Fierro et al. (2009) found substantial changes in the simulated structure of Hurricane
445 Rita (2005) as resolution was coarsened from 1 to 5 km, but little change in typical measures of
446 storm intensity; further coarsening of resolution beyond 5 km did reduce peak intensity. All of
447 these studies prescribed an initial vortex and did not note any strong effect of resolution on the
448 time needed for TC intensification.

449 Here we suggest that TC intensity is limited and cyclogenesis delayed at coarse resolutions at
450 least in part because coarse-resolution simulations produce a drier environment around the storm.
451 Multiple studies have argued that a near-saturated troposphere is required for TC genesis (e.g.
452 Emanuel 1989, 1995; Bister and Emanuel 1997; Frisius 2006; Raymond et al. 2007), so it seems
453 plausible that a dry bias might inhibit the intensification of a TC in a model. We do not seek to
454 determine whether the dry bias directly inhibits precipitating ascent in the TC eyewall or whether
455 it acts indirectly by enhancing convective downdrafts outside the eyewall that cool and dry the
456 subcloud layer; instead we simply invoke the general idea that TC genesis and intensity are in-
457 hibited in dry environments. The moisture field in our integrations with rotation was affected by
458 resolution in a way qualitatively similar to that seen in the non-rotating integrations. During the
459 first 20 days of integration, the time-mean, domain-mean PW decreased monotonically as resolu-
460 tion was coarsened from 1 to 16 km (Fig. 8b). On the day the TC achieved category 1 intensity, the
461 environment around the storm was drier in integrations conducted at 8 km resolution than in those
462 conducted at 1 km resolution (e.g. Fig. 9). The radial moisture gradient is enhanced even more at
463 coarser resolutions when the TCs achieve category 2 intensity and the eyewall, as indicated by the
464 surface wind speed distribution, is larger and somewhat more ragged at coarser resolutions (e.g.
465 Fig. 11a, b).

466 To illustrate the resolution-dependence of the humidity field more quantitatively, we present
467 horizontal distributions of 6-hour averages of PW during the first two days on which the storm
468 had an intensity of category 1, with all model output coarsened via block averaging to the same 16
469 km grid. The mode of the horizontal distribution of PW clearly shifts to lower values at coarser
470 resolutions (Fig. 12b). At the same time, the moist region (i.e. the eye and eyewall of the TC)
471 became moister as resolution was coarsened, as evidenced by the upward shift in the upper tail of
472 the PW distribution and by the example shown in Fig. 11.

473 These changes in humidity are accompanied by changes in the distribution of vertical velocities.
474 Explicit simulations of convection are expected to produce slower ascending motions as horizon-
475 tal resolution is coarsened beyond about 1 km because updrafts at these resolutions typically have
476 a width of a single grid cell and thus a more shallow aspect ratio. Buoyant parcels with a shal-
477 low aspect ratio rise more slowly because they are closer to the hydrostatic limit in which the
478 buoyancy-induced vertical pressure gradient force balances the buoyancy force itself. In contrast,
479 for narrow parcels, the pressure gradient force facilitates ascent by horizontally diverging air above
480 the parcel out of the parcel's upward path (e.g. Houze 1993). Slower ascent of individual parcels
481 was documented in the simulations of Pauluis and Garner (2006), who derived a theoretical scal-
482 ing for the dependence of updraft speed on model horizontal resolution. Slower ascent is seen at
483 coarser resolutions in our rotating simulations (e.g. the left tail of the 500 hPa vertical velocity
484 distributions in Fig. 12a), although this may represent organized ascent in the TC eyewall that is
485 less directly controlled by the local vertical buoyancy force.

486 While previous studies have examined how updraft speed depends on horizontal resolution, less
487 attention has been given to how resolution affects the subsiding motions that adiabatically warm
488 and dry the environment around a precipitating cluster. In our simulations, downward velocities

489 increase as resolution is coarsened, as evidenced by the distribution of vertical velocity at 500 hPa
490 on the two days after the TCs achieved category 1 intensity (Fig. 12a).

491 Lane and Knievel (2005) examined the spectrum of gravity waves excited by a buoyancy
492 anomaly that was 10 km wide in models with variable horizontal resolution, and found that coarser
493 resolutions produced more power at longer wavelengths. Although they did not discuss the far-
494 field subsidence, their figures show that the spreading gravity wave front travelled a greater dis-
495 tance after one hour of simulation at a resolution of 1.5 km than it did at the very fine resolution of
496 63 m. The group velocity of nonhydrostatic gravity waves increases with horizontal wavelength,
497 as can be seen from (5), which for the nonrotating case without RAVE reduces to

$$c_{gx} \simeq \frac{Nm^2}{(k^2 + m^2)^{\frac{3}{2}}}. \quad (8)$$

498 Although some discussions of the remote response to a convective cluster assumed a hydrostatic
499 gravity wave field, for which $c_{gx} \simeq N/m$ (e.g. Mapes 1993), non-hydrostatic effects slow the group
500 velocity to about 90% of its hydrostatic value for a horizontal wavelength of 100 km and to about
501 50% of its hydrostatic value for a wavelength of 40 km. Given that Lane and Knievel (2005) found
502 a peak response at 10 km wavelength in the simulated spectrum of stratospheric gravity waves ex-
503 cited by idealized convection, these nonhydrostatic effects would seem to matter for determining
504 the speed at which gravity wave pulses propagate away from a convecting region. Coarser hori-
505 zontal resolutions would be expected to partition more energy into the longer-wavelength part of
506 the gravity wave spectrum, and it is that part of the spectrum that has faster group velocities. It is
507 unclear whether this faster spreading of gravity waves away from the precipitating cluster would
508 result in stronger environmental subsidence. If radiative cooling balances subsidence warming in
509 the time-mean, then an increase in the rate at which subsiding motions enter a region might re-

510 sult in stronger subsidence. But unlike the effects of RAVE, we have no theory to show that the
511 amplitude of vertical velocities is influenced directly by horizontal resolution.

512 In summary, we suggest that coarser resolutions produce slower and wider updrafts together
513 with a geostrophic adjustment process that is accomplished by gravity waves with longer horizon-
514 tal wavelengths and thus faster group velocities. Our simulations produce stronger subsidence at
515 coarser resolutions, as part of a more general reduction in skewness seen in the histograms of 500
516 hPa vertical motion. Whether this can be explained by the faster group velocity of gravity waves
517 is unclear, and so the cause of the stronger subsidence at coarser resolutions is unclear. Vertical
518 velocity distributions in the nonrotating simulations exhibit a similar sensitivity to horizontal res-
519 olution (not shown), which suggests that the resolution-dependence of the vertical velocity skew-
520 ness arises from a mechanism that is general to aggregated convection and not one that involves,
521 for example, TC eyewall dynamics.

522 *b. Effect of RAVE*

523 Since we showed that RAVE reduces the group velocity and amplitude of gravity waves and
524 compensates for the dry bias seen in coarse-resolution simulations of convectively aggregated
525 states, it seems natural to ask whether RAVE might reduce some of the bias seen in simulations
526 of cyclogenesis at coarse resolutions. In our rotating simulations, the use of RAVE does produce
527 a moister domain and decrease the time needed for TC genesis. Increasing γ from 1 to 8 in
528 simulations with 8 km resolution reduces the time needed to achieve category 1 intensity from
529 about 50 days to 15 days (Fig. 13). It also increases the peak surface wind speeds by 10 m s^{-1} or
530 more, which corresponds to an increase of one to two Saffir-Simpson intensity categories.

531 At the time at which the storm first achieves category 1 intensity, RAVE produces a large increase
532 in the domain-mean precipitable water, compensating for the dry bias seen in the coarse-resolution

533 simulations without RAVE (Fig. 14, compare with Fig. 9). On the day the storm achieves category
534 2 intensity, the use of $\gamma = 4$ appears to have eliminated, and perhaps even slightly overcompen-
535 sated for the biases in radial moisture gradient and eyewall size that arose from using the coarser
536 resolution of 8 km (Fig. 11). The histogram of PW values narrows as γ is increased from 1 to 4 so
537 that it becomes more like the PW histogram for the control simulation at 1 km resolution without
538 RAVE; the bias in the histogram of vertical velocity is also reduced (Fig. 12c, d). Use of $\gamma = 8$ at
539 8 km resolution produces a domain that is too moist and results in genesis that occurs too quickly
540 relative to the fine-resolution control run. Eyewall diameter also becomes smaller as γ is increased.
541 Gentry and Lackmann (2010) found that the radius of maximum wind increased when horizontal
542 grid spacing was increased in simulations of an observed hurricane, so this decrease in storm size
543 might also be seen as correcting a bias caused by low resolution (though an over-correction clearly
544 occurs for $\gamma = 8$).

545 At 16 km resolution, use of $\gamma = 16$ produces little change in the time needed to achieve category
546 1 intensity, but does allow the peak storm intensity to increase by a full Saffir-Simpson category
547 (Fig. 13). These changes may be desirable from the perspective of bias reduction since the 16 km-
548 resolution simulations with $\gamma = 1$ did not exhibit a delay in the time needed to achieve category 1
549 intensity but did exhibit overly weak storm intensities.

550 **6. Summary and discussion**

551 Using simulations of radiative-convective equilibrium in a cloud system-resolving model, we
552 demonstrated that the environment around a precipitating cluster becomes drier as model reso-
553 lution is coarsened from 1 km to $O(10)$ km). This dry bias occurs at coarse resolutions in both
554 rotating and non-rotating domains, and is accompanied by overly intense subsidence outside the
555 moist cluster. In rotating domains, the convective cluster spontaneously evolves into a tropical

556 cyclone, but this genesis is delayed and the peak TC intensity is reduced as model resolution is
557 coarsened.

558 Many of these biases that occur at coarse resolutions can be compensated for by the RAVE
559 rescaling: the dry bias is reduced, TC genesis occurs at earlier times, and the peak TC intensity
560 is increased for $\gamma > 1$. We suggest that these effects of RAVE are caused at least in part by the
561 influence of the rescaling on the domain-mean moisture field via a reduction in the amplitude
562 and horizontal group velocity of gravity waves. Although this hypothesis may seem somewhat
563 speculative at this stage because we have not provided any explicit evidence that the moisture
564 field was altered by changes in gravity wave characteristics, it is a corollary of several existing
565 ideas. Multiple previous studies have shown that the dry region outside a convective cluster is
566 produced by subsidence created by the buoyancy bores that spread outward from the convective
567 source (e.g. Bretherton and Smolarkiewicz 1989; Nicholls et al. 1991; Mapes 1993). Reducing the
568 group velocity of these gravity wave packets by increasing the Coriolis parameter has been argued
569 to trap subsidence closer to the convective source and thus to reduce the contrast in subsidence and
570 humidity between the convective cluster and its far-field environment (Liu and Moncrieff 2004;
571 Bretherton et al. 2005). Here we argue that a reduction in the amplitude and group velocity of
572 gravity waves produced by RAVE should have an analogous effect on the distributions of verti-
573 cal motion, humidity, and convection. Previous studies showed that RAVE slows the process of
574 adjustment to a balanced state by altering the gravity wave dispersion relation (Skamarock and
575 Klemp 1994; MacDonald et al. 2000b), but did not consider the implications for the distribution
576 of humidity or subsequent moist convection. The idea that RAVE reduces the horizontal contrast
577 in humidity by trapping buoyancy anomalies near an initial pulse of moist convection constitutes
578 a new view of the manner in which RAVE influences organized convection. Prior work focused
579 on how RAVE slows the vertical convective motions themselves (e.g. Kuang et al. 2005; Pauluis

580 et al. 2006) rather than the slower subsiding motions that operate on the much larger length scale
581 of a Rossby deformation radius.

582 The effects of RAVE on the group velocity of gravity waves may be desirable if the group veloc-
583 ity is biased high in coarse-resolution simulations of convection. While we have not demonstrated
584 this definitively here, Lane and Knievel (2005) showed that the spectrum of gravity waves excited
585 by convection is centered at longer wavelengths in simulations conducted at coarser resolutions,
586 and longer waves do have a faster group velocity. Furthermore, RAVE reduces the bias in the
587 horizontal distribution of mid-tropospheric vertical velocity that occurs in coarse-resolution sim-
588 ulations of TCs. It is thus possible that RAVE directly addresses the cause of the dry bias seen in
589 coarse-resolution simulations, rather than introducing a moist bias by some mechanism indepen-
590 dent of that which creates the original dry bias. Yet there is still much to explore here, and the
591 effects of resolution and RAVE on the vertical moisture fluxes produced by shallow convection
592 merit further examination.

593 The use of large RAVE factors can overcompensate for the biases seen in coarse-resolution
594 simulations. While using $\gamma = 4$ at 8 km resolution produced PW distributions similar to those
595 seen in the fine-resolution control run, using $\gamma = 8$ produced a domain that was too moist and
596 TC genesis that occurred too soon. Similarly, the dry bias seen in 16 km-resolution simulations
597 with $\gamma = 1$ was largely eliminated when γ was increased to 8, but became a moist bias at $\gamma = 16$.
598 One would probably want to avoid using values of γ that are large enough to create a moist bias
599 stronger than the original dry bias. Since the nature and magnitude of the biases created by use
600 of coarse resolution likely depend on model numerics and grid, it may not be possible to provide
601 a universal recommendation for the optimal value of γ . However, our results indicate that values
602 of γ that would provide a resolution “equivalent” to that of the control run (e.g. $\gamma = 8$ at 8 km
603 resolution for a control run at 1 km resolution) may be too high. This may be because it is more

604 important to tune the gravity wave group velocity rather than some sort of effective resolution, and
605 that group velocity scales like γ^{-3} .

606 While we have focused on the effects of resolution and RAVE on the ambient moisture field,
607 the influence on organized convection might manifest via other mechanisms. For instance, resolu-
608 tion might limit peak TC intensity through its influence on the explicit representation of eyewall
609 structure (e.g. Fierro et al. 2009; Gentry and Lackmann 2010). Or resolution might produce en-
610 vironmental drying through microphysical effects not explored here, such as less precipitation
611 falling outside of saturated updrafts when those updrafts are wider at coarser resolutions, thereby
612 increasing the precipitation efficiency. RAVE might enhance the intensity of TCs through the
613 cooling effect it seemed to have in the upper troposphere and lower stratosphere (Emanuel 1988),
614 but simple estimates suggest that this cooling should produce an increase in the maximum po-
615 tential intensity of 1-2% for the RAVE numbers considered here. Given that coarse resolution
616 integrations without RAVE can also have large temperature biases in the upper troposphere and
617 lower stratosphere, it is not obvious that this effect of RAVE is any worse than the effect of inte-
618 grating at coarse resolution or using parameterized convection. Another possibility is that RAVE
619 might make a TC less sensitive to mid-tropospheric dry air by enhancing the horizontal scale of
620 the precipitating upward mass flux in the eyewall and thereby reducing any drying by eddies that
621 act diffusively, which is distinct from making ambient air less dry. Finally, it should be remem-
622 bered that the limited domain size used in our simulations may influence TC size, TC intensity,
623 and the rapidity of cyclogenesis in ways that depend on resolution or RAVE factor (e.g. Chavas
624 and Emanuel 2014). Despite all these caveats, the environmental moisture field is thought to play
625 an important role in TC genesis and intensification (e.g. Bister and Emanuel 1997; Frisius 2006;
626 Raymond et al. 2007), and our results show that this field is influenced by resolution and RAVE.

627 Further assessment will be required before RAVE can be used routinely in operational or re-
628 search models. However, this study shows that even without RAVE, numerical models with hor-
629 izontal resolutions coarser than a few kilometers have large biases in the simulation of organized
630 convection and its non-convecting environment. In deciding whether to use RAVE in numerical
631 models of organized convection, one should remember that traditional convective parameteriza-
632 tion introduces its own set of biases and that use of coarse resolution distorts vertical motions in
633 and around organized convection, with consequences for the moisture field.

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640 APPENDIX

641 642 **Comparison with results of Pauluis et al. (2006)**

643 As stated in the Introduction, Pauluis et al. (2006, hereafter P06) concluded that RAVE enhanced
644 the amplitude of the dry bias that occurs at coarse horizontal resolutions in simulations of radiative-
645 convective equilibrium. We demonstrated the opposite result in this paper and suggest that the
646 difference may arise from the fact that P06 used a much shorter period for their time averages
647 while also increasing their model domain size as resolution was coarsened (holding the number of

648 model grid points constant). Here we present a few relevant model analyses and discuss possible
649 reasons for our contrasting results.

650 Other studies have shown that larger domains favor the spontaneous aggregation of moist con-
651 vection and an associated domain-mean drying (e.g. Bretherton et al. 2005; Khairoutdinov and
652 Emanuel 2010), so it seems possible that the domain-mean drying P06 found at coarse resolutions
653 is due to convective aggregation in those correspondingly larger domains. Our 4°-wide domain
654 at 8-km resolution has the same number of grid points as our 1°-wide domain at 2-km resolution,
655 and the former does have a dry bias relative to the latter when 150-day time- and horizontal-means
656 are compared (solid blue line in Fig. 15). This dry bias is about twice as large as those found
657 when resolution is coarsened while holding domain size constant (e.g. Fig. 3). Applying RAVE
658 with $\gamma = 4$ to the 8-km resolution run reduces the dry bias by roughly 30% (relative to the same
659 2-km resolution run without RAVE). From this one would conclude that RAVE moistens the en-
660 vironment even when the number of grid points is held constant as resolution is coarsened, which
661 is opposite to the finding of P06. However, when we average the specific humidity over an 8-
662 day period, starting after the first 8 days of our simulations, RAVE actually enhances the dry bias
663 (compare solid and dashed black lines in Fig. 15). One might speculate that RAVE speeds up the
664 convective self-aggregation that is associated with the domain mean drying while moistening the
665 final aggregated state, but further analysis would be needed to draw a firm conclusion.

666 Other differences between the model configurations used here and by P06 might also contribute
667 to our contrasting results. P06 imposed vertical wind shear in their models by relaxing horizontal
668 winds to either a weak or strong shear profile (our simulations did not use imposed shear). They
669 noted that strong shear produces organized convection, consistent with previous studies showing
670 that shear can cause convection to arrange into bands or arcs (e.g. Robe and Emanuel 2001).
671 Vertical wind shear can also inhibit the self-aggregation of convection into a single cluster (e.g.

672 Tompkins 2001), although Bretherton et al. (2005) applied the same magnitude of shear used in
673 the “weak shear” simulations of P06 and found that it slowed but did not halt the self-aggregation
674 process. The implications of this slowing for integrations that only last 16 days are unclear. This
675 study and P06 also used models with different parameterizations of microphysics and radiation,
676 which might influence the character of any organization or whether aggregation occurs. Thus, it
677 is not possible to definitively determine the reasons for our different results, but this appendix has
678 presented some evidence in support of the hypothesis that the difference is primarily caused by the
679 use in P06 of a short averaging period and a methodology that enlarged domain size as resolution
680 was coarsened.

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LIST OF FIGURES

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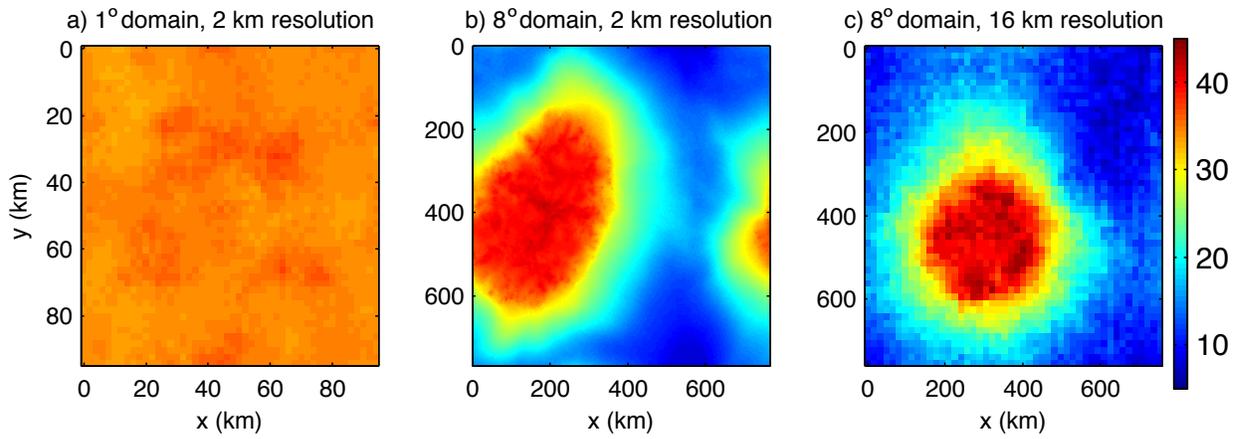
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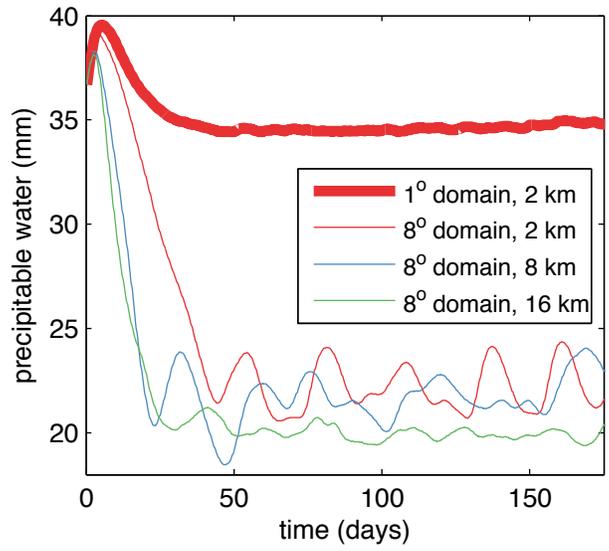
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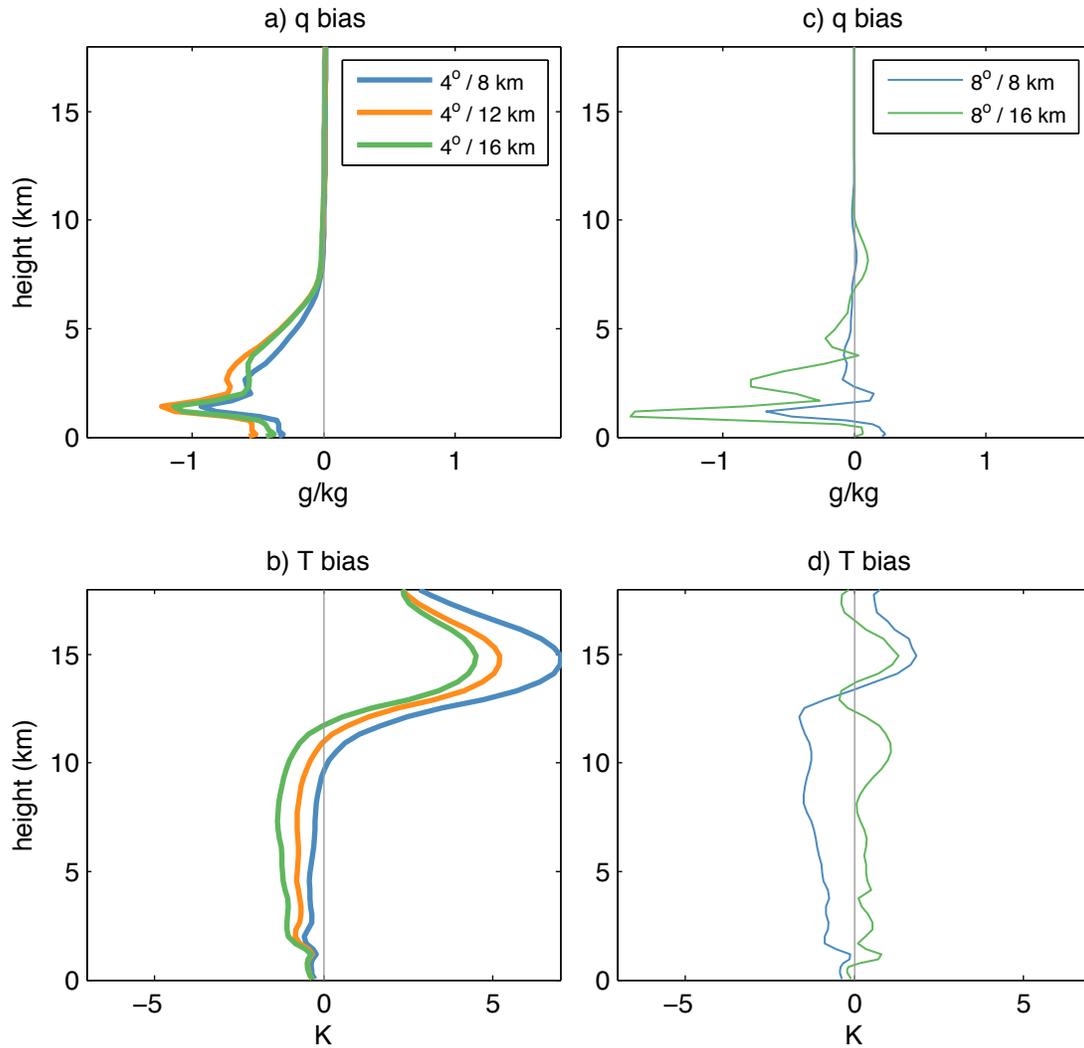
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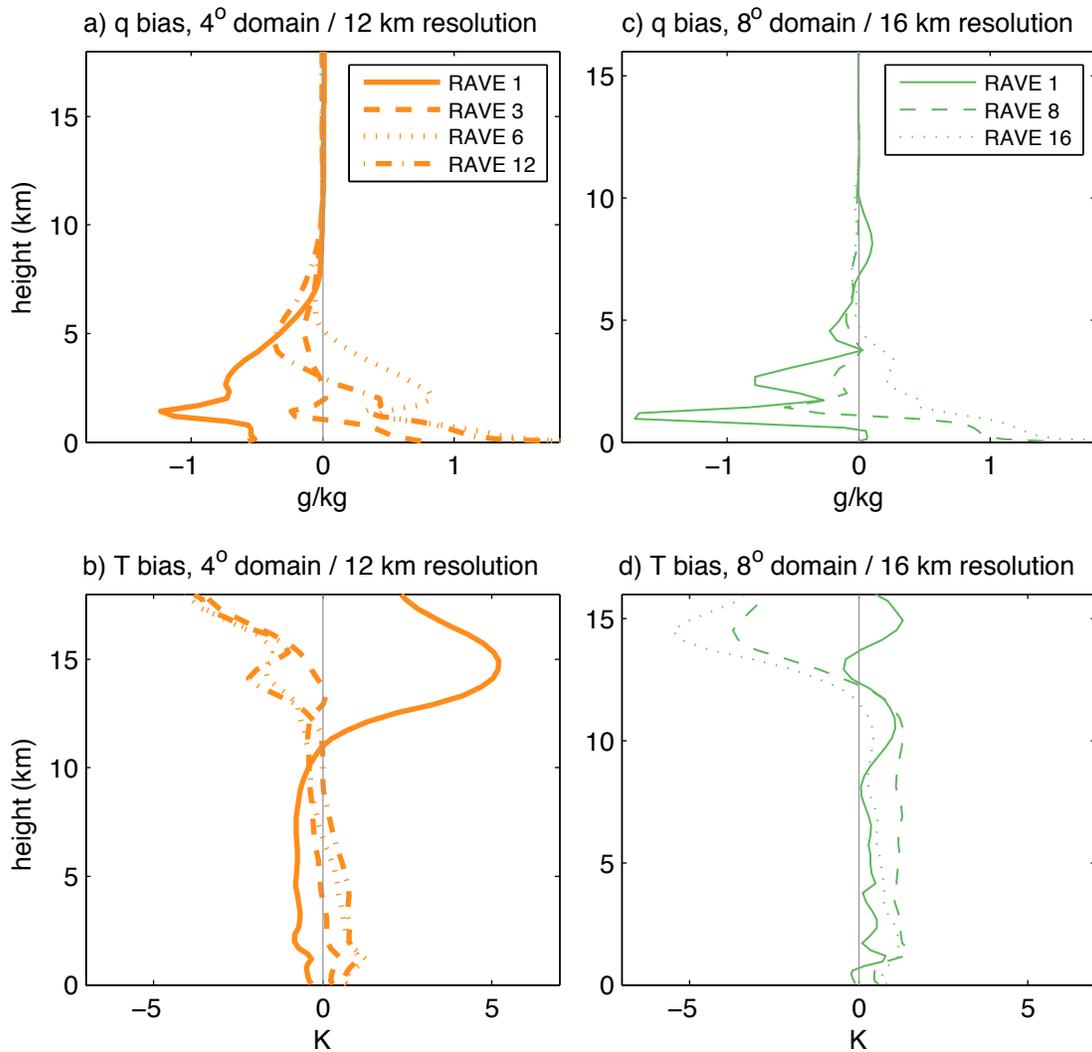
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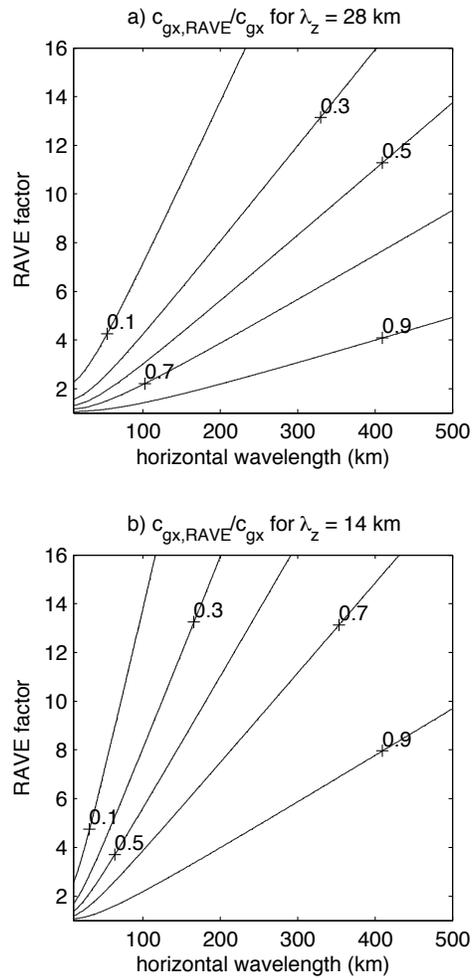
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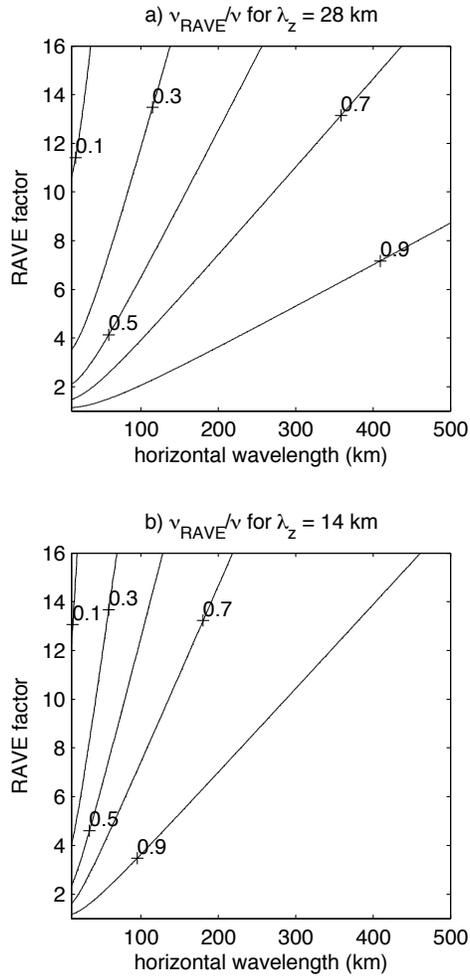
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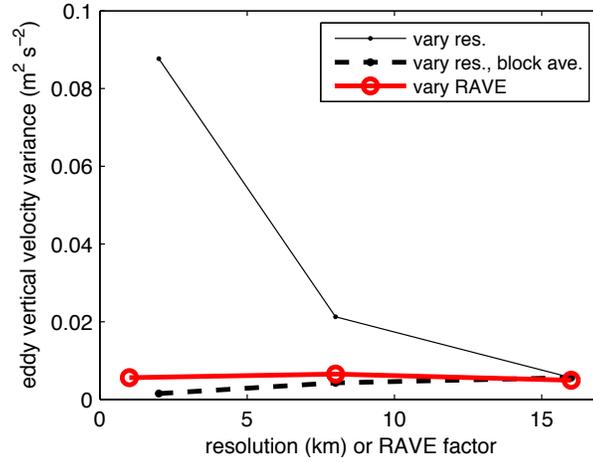
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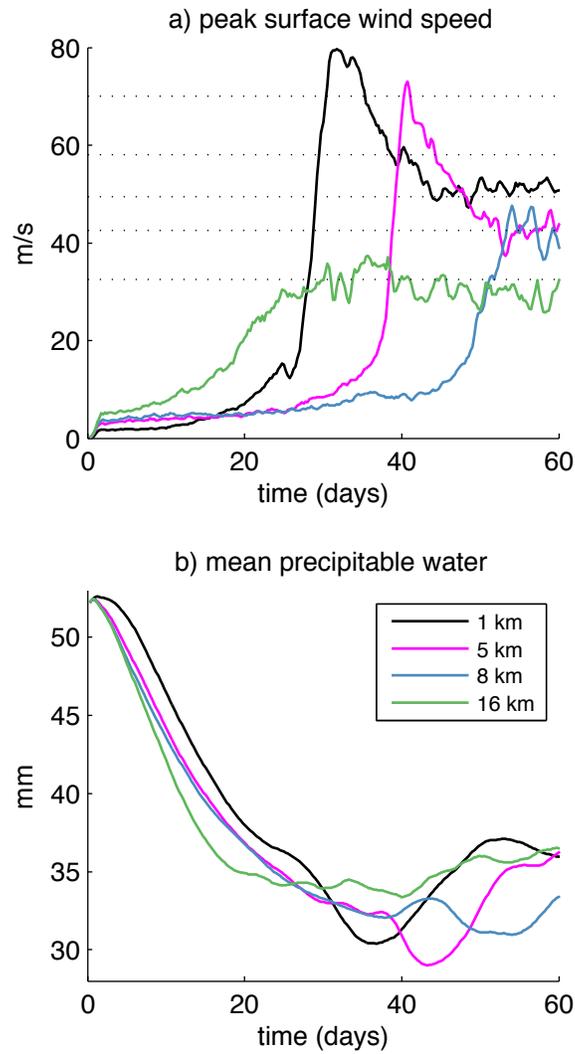
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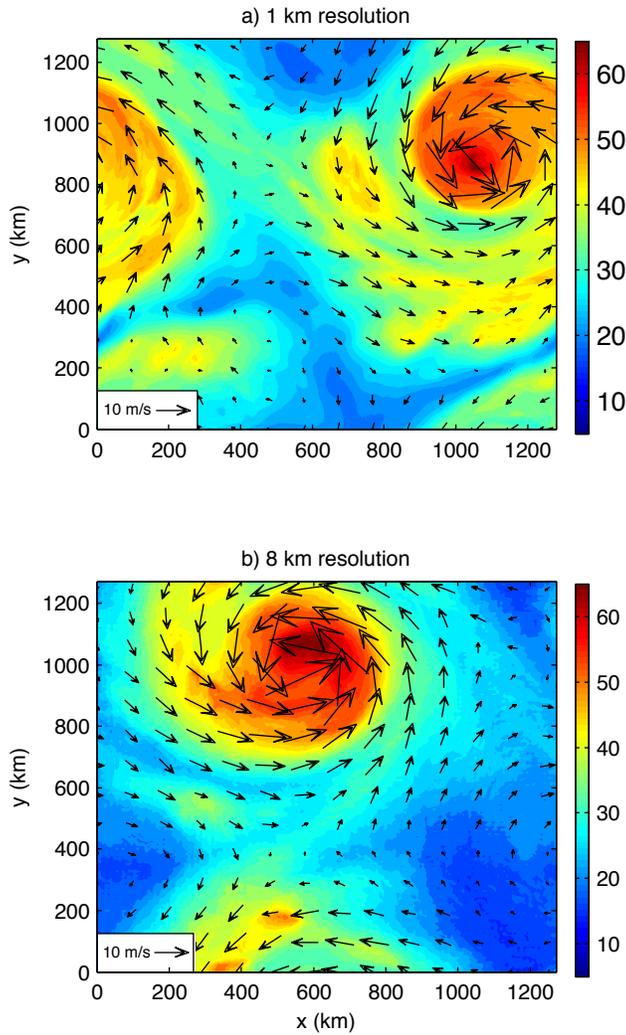
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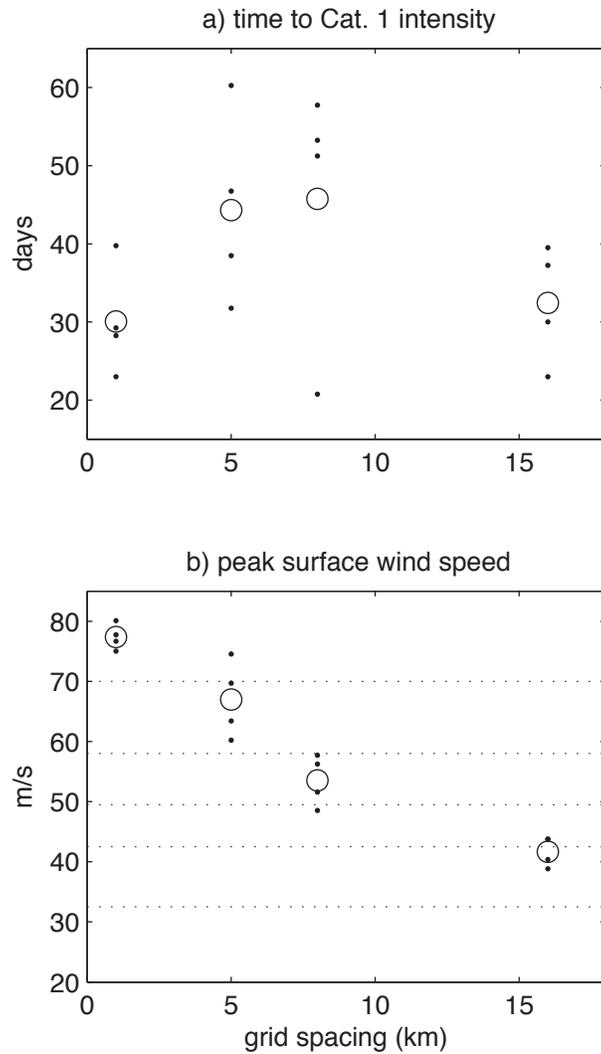
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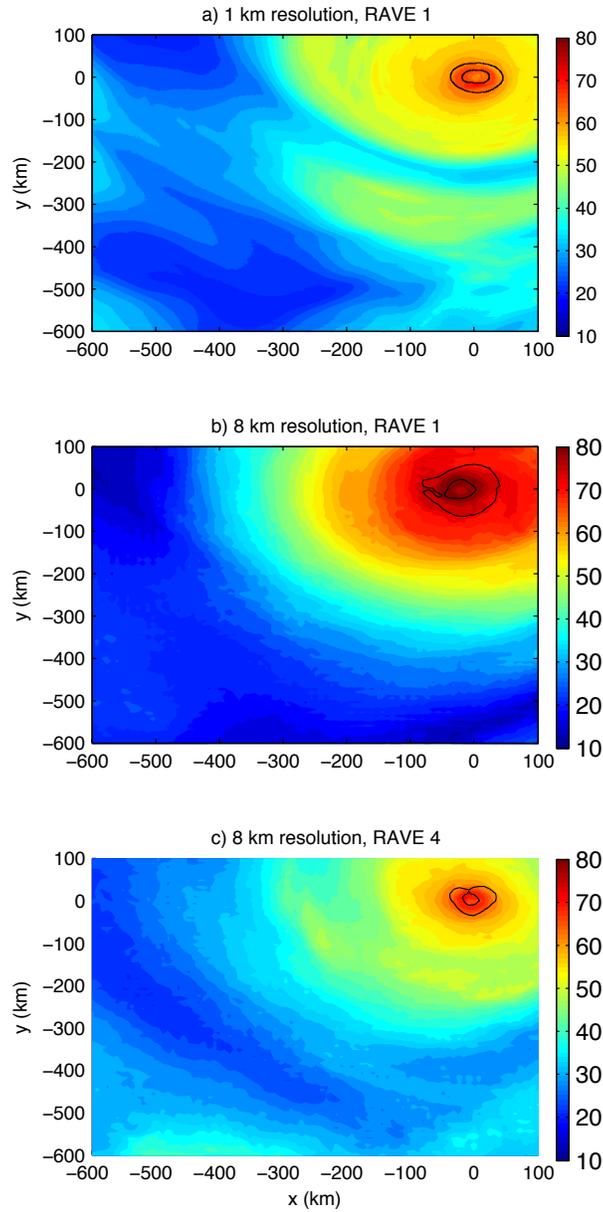
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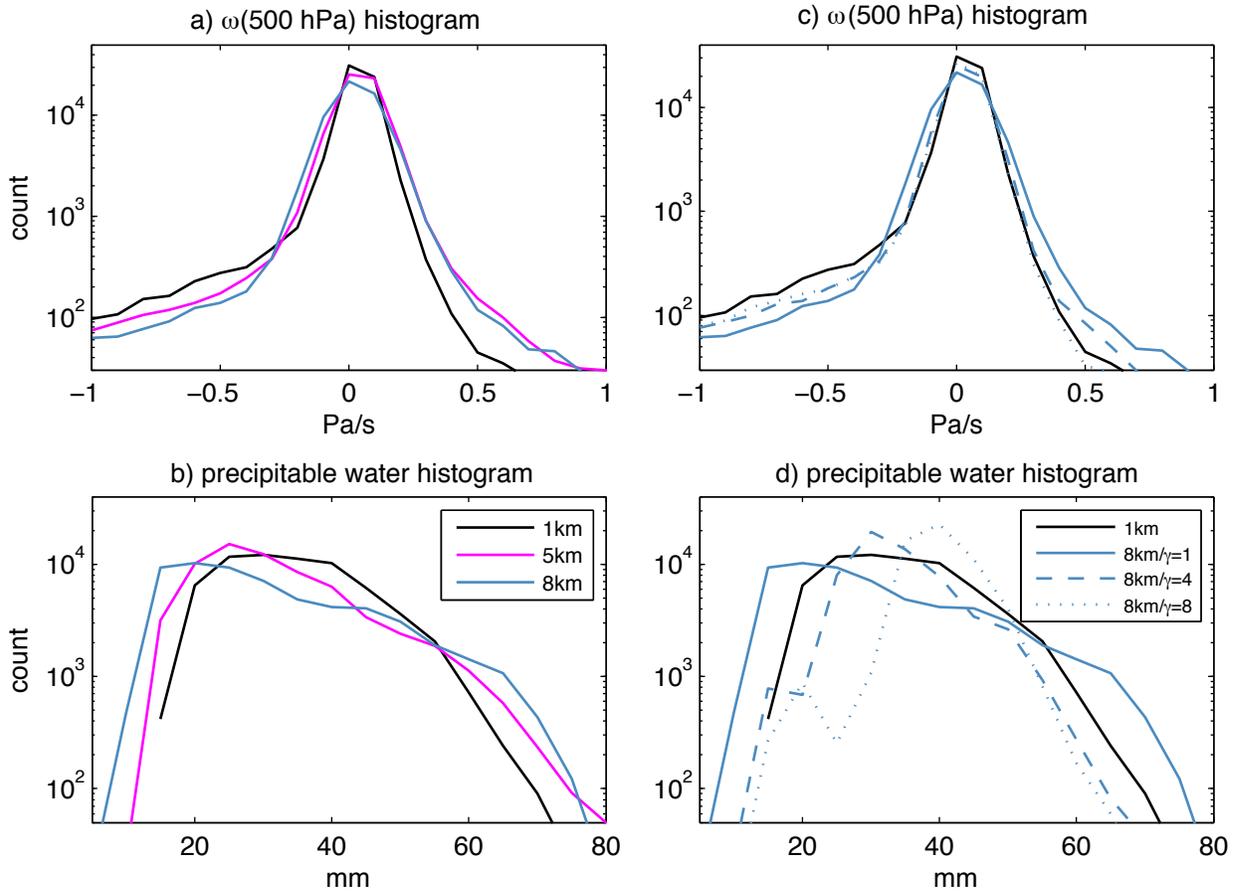
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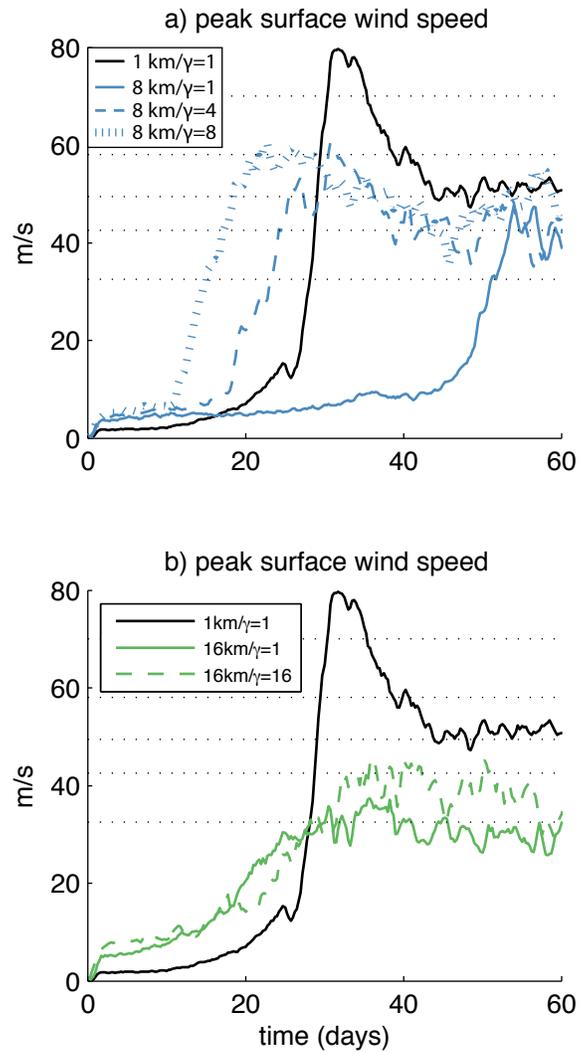
933 FIG. 10. Metrics of tropical cyclogenesis for ensembles of integrations conducted in rotating domains at
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 935 model initialization and the day on which the TC achieves Saffir-Simpson category 1 intensity. b) Maximum
 936 6-hourly averaged surface wind speed during the entire simulation. Small black dots show values for individual
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949 FIG. 13. As in Fig. 8a but for varying RAVE factor at a) 8 km resolution and b) 16 km resolutions. The result
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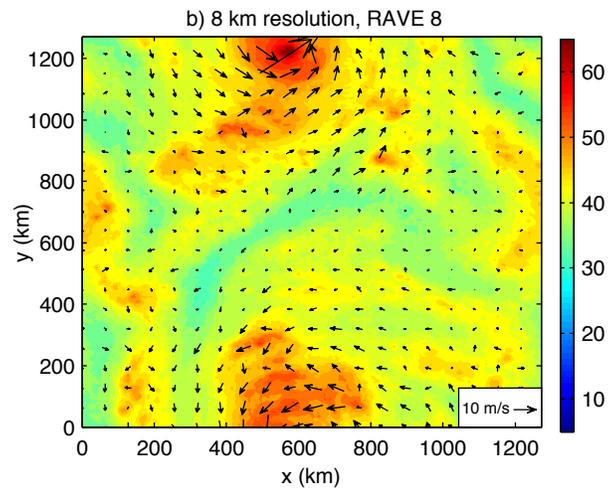
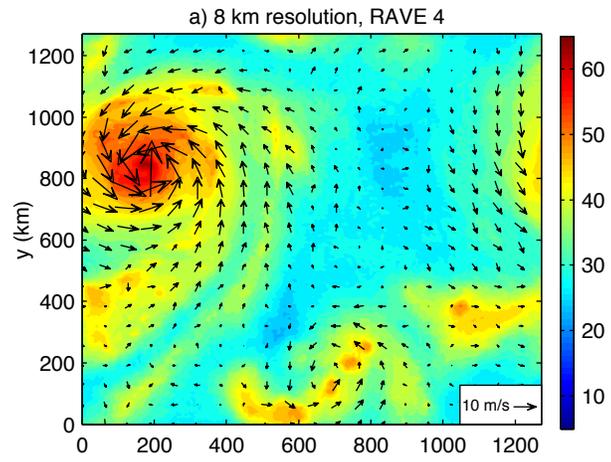
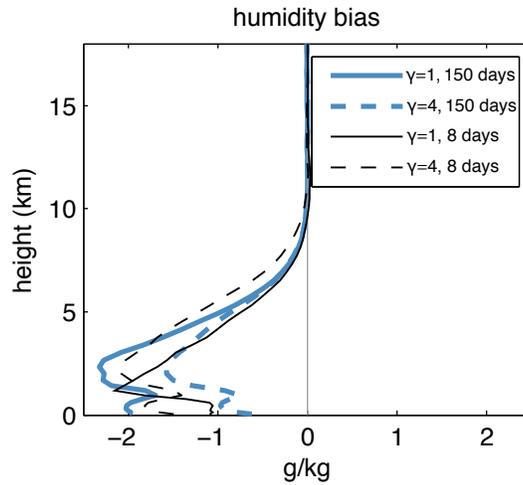


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951 FIG. 15. Time- and horizontal-mean difference in specific humidity between integrations conducted at 8 km
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