

1 Higher Surface Bowen Ratios Ineffective at
2 increasing Updraft Intensity

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3 The sensitivity of various metrics of convective intensity to changes in bound-
4 ary layer depth via changes in the surface Bowen ratio is explored with radiative-
5 convective equilibrium (RCE) and initial condition simulations in the Sys-
6 tem for Atmospheric Modeling, a cloud resolving model (CRM). In the RCE
7 simulations, high percentile updrafts showed little change in response to changes
8 in the surface Bowen ratio. Initial condition simulations showed low surface
9 Bowen ratios having stronger updrafts than high surface Bowen ratios. A par-
10 cel model was used to explore whether RCE results could be explained with
11 an entrainment parameter independent of boundary layer depth. It was found
12 that for every set of simulations in RCE, entrainment rates independent of
13 boundary layer depth could explain the lack of change in updraft velocities
14 with boundary layer depth. Given the indifference of high percentile updraft
15 velocities in our simulations to changes in the surface Bowen ratio, we con-
16 clude that convective intensity as measured by this quantity in the cloud re-
17 solving model is not sensitive to this forcing.

1. Introduction

18 Tropical deep convection has large variations in the occurrence of high intensity con-
19 vective storms over land versus over the ocean [Zipser, 2003]. This is seen in lightning
20 flash rate measurements [Cecil *et al.*, 2014], as well as in the few direct measurements of
21 updraft velocity from convective cumulus [Lucas *et al.*, 1994]. In general, lightning flash
22 rate can be considered a good proxy for storm updraft velocity because it is thought that
23 lightning flash rate increases monotonically with the updraft velocity of the storm [Boc-
24 cippio, 2002]. Measurements of convective cumulus overshooting the tropopause have also
25 been shown to be much more frequent over land than over ocean [Liu and Zipser, 2005].
26 Minimum brightness temperatures of the 37 and 85 GHz channels as observed from the
27 Tropical Rainfall Measuring Mission Satellite, as well as the heights of 40 dBZ echo tops
28 indicate that storms over land are stronger than those over the ocean [Zipser *et al.*, 2006].
29 In general, it seems that by many measures of intensity, from updraft velocity, to lightning
30 flash rate, to overshooting top frequency, convective events over tropical landmasses tend
31 to be much stronger than those over tropical oceans. It is our goal to test and explore one
32 of the more popular hypotheses regarding what controls tropical convective intensity: the
33 surface Bowen ratio (SBR) and by proxy, boundary layer depth using a cloud-resolving
34 model. The $SBR = SHF/LHF$, where SHF is the sensible heat flux, and LHF is the
35 latent heat flux.

36 Land surfaces typically experience a more notable diurnal cycle in temperature than
37 oceans do, mainly due to the lower heat capacity of land. Therefore, it seems plausible
38 that convective available potential energy (CAPE) over land could have larger values

39 than over the oceans in the tropics. However, available data does not show a marked
40 difference of CAPE between land and ocean: Observations are sparse, from one of the few
41 publications available, CAPE does not vary too much between land and ocean [*Williams*
42 *and Renno*, 1993]. Reanalysis data indicates some differences in mean CAPE between
43 tropical continents and oceans, but they are within an order of magnitude, and over the
44 maritime continent, CAPE has values very similar to nearby oceans, despite a very large
45 difference in lightning activity in those regions [*Riemann-Campe et al.*, 2009; *Cecil et al.*,
46 2014]. *Lucas et al.* [1994] stated that there “is no basis at all for attributing updraft
47 velocity differences [between various field campaigns] to CAPE over land and water.”
48 With that in mind, we explore a different hypothesis that could explain the observed
49 land-ocean differences in lightning flash rate.

50 Variations of the fractional entrainment rate of environmental air into a rising moist
51 parcel is a popular hypothesis for why intensity metrics like updraft velocity and lightning
52 flash rate may be higher over landmasses than over oceans in the tropics [*Zipser*, 2003;
53 *Williams and Stanfill*, 2002]. If there is less dilution of a parcel, it follows that intensity
54 metrics related to moist buoyancy are larger than in cases when there is more dilution
55 by drier environmental air [*Williams and Stanfill*, 2002]. Some researchers [*Lucas et al.*,
56 1994, 1996; *Zipser*, 2003; *Williams and Stanfill*, 2002; *Williams et al.*, 2005] have argued
57 that entrainment rates over land and ocean could vary due to potential differences in
58 updraft width. *Williams and Stanfill* [2002] described the concept that as boundary layer
59 depth increases, the width of updrafts reaching cloud base increases, and with a wider

60 updraft it is harder for environmental air to impinge into the updraft core, thus keeping
61 the convective plume less diluted and more buoyant.

62 Boundary layer depth can be controlled by the surface Bowen ratio. If we consider the
63 tropics to have a constant total surface flux, increasing the SBR will result in a deeper
64 boundary layer. Typical values for ocean SBRs are around 0.05, while those over land
65 tend to go from 0.25 and up [*Williams and Stanfill, 2002*]. Thus, boundary layer depth
66 over land is greater than that over the ocean.

67 It has been found that one measure of convective intensity, lightning flash rate, is pro-
68 portional to boundary layer depth [*Williams et al., 2005*]. Ice water path, another variable
69 that could influence lightning frequency has also been found to be strongly correlated with
70 boundary layer depth [*Leung, 2011*]. *Williams and Stanfill [2002]* suggested that the influ-
71 ence of the SBR on boundary layer depths and hence updraft widths is a plausible control
72 for land-ocean differences in convective intensity in the tropics. Comparison of updraft
73 widths between GATE [*LeMone and Zipser, 2005*] and those of the Thunderstorm Project
74 [*Byers and Braham, 1949*], showed that the updrafts over land were notably wider than
75 those over the ocean. This combined support for the idea that regional variations in the
76 surface Bowen ratio could lead to convective intensity differences via their influence on
77 updraft width is what led us to test this hypothesis in a CRM framework.

78 Given the physical evidence indicating that deeper boundary layers resulting from higher
79 SBRs might enhance convective intensity by limiting entrainment, this study focuses on
80 examining this mechanism in a cloud-resolving model. Our simulations were used to test
81 whether variations in SBR could control high percentile updraft velocities, a statistic that

82 can be used as a corollary to convective intensity [as in e.g. *Lucas et al.*, 1994; *Zipser*,
83 2003].

84 Another intensity metric that we examine is frictional dissipation by falling precipitation
85 and precipitation dissipation scale height [*Pauluis and Dias*, 2013], which is the former,
86 normalized by precipitation, outputting a distance variable. Greater scale heights imply
87 more recirculation of precipitation in convective cumulus, indicating that there may be
88 more ice available to generate lightning. Previous work has shown that the precipitation
89 scale height seems to effectively highlight a land-ocean convective intensity difference
90 [*Pauluis and Dias*, 2013]. We also examine what we refer to as graupel dissipation and
91 scale height, which is likely more directly applicable to the development of lightning.

92 We examine radiative-convective equilibrium [RCE, as in e.g. *Parodi and Emanuel*,
93 2009; *Singh and O’Gorman*, 2014], and initial condition simulations as in [*Robinson et al.*,
94 2011]. In examining all of our simulations, we asked the following questions:

- 95 • Does increasing the SBR in CRM simulations produce stronger high intensity updraft
96 statistics?
- 97 • How do other convective intensity metrics vary?
- 98 • Does entrainment have to depend on the surface Bowen ratio to explain our results?

99 There are potential limitations to using RCE and cloud resolving models as a framework
100 for measuring convective intensity. *Varble et al.* [2014] found that many cloud resolving
101 models (including the model we use) attempting to replicate field campaign data over-
102 estimated vertical velocities in deep convective storms. Our goal isn’t to reproduce real
103 world convection, but to test the sensitivity of convection to major differences in surface

104 forcing. While it's certainly possible that these models will not respond to forcings as
105 deep convection responds in real life, they do include processes that should influence the
106 response to variations in surface Bowen ratio. Hence it is of interest to see what they tell
107 us about how surface Bowen ratios influence the intensity of deep convection, even if this
108 will not definitively tell us about what real world convection would do. The methodology
109 that we develop could be applied to other models when they are able to more accurately
110 reproduce real-world convective intensities.

111 The methodology and setup for our cloud resolving model is described in section 2.
112 Section 3 first presents intensity metrics directly from the CRM, and then uses a sim-
113 ple parcel model to explore the CRM environments. Section 4 discusses the potential
114 implications of our results and concludes.

2. Methodology

115 Our simulations were conducted using the System For Atmospheric Modeling [SAM,
116 version 6.10.3 *Khairoutdinov and Randall, 2003*]. Our tabulated simulations were run in
117 both 2D and 3D, with horizontal resolutions of 200m in 3D, and 400m and 200m in 2D,
118 always with 64 vertical levels. The model was run into radiative-convective equilibrium
119 for the majority of cases, taking approximate 40 days. In all 2D simulations, the total
120 run time was 50 days, while the 3D simulations had run times of 45 days in the low SBR
121 case and 65 days in the high SBR case (as it took longer to reach RCE). The non-RCE
122 cases were initial condition simulations, which were run for 1 model day, with statistics
123 gathered over that day. All statistics were initially gathered at a 30 minute sampling
124 interval.

125 All simulations used doubly or singly periodic boundary conditions with no Corio-
 126 lis effect. All simulations utilized a 1.5 order subgrid-scale turbulence closure. In the
 127 simulations we show/tabulate results for, we used Morrison two-moment microphysics
 128 [*H. Morrison and Khvorostyanov, 2005*]. We also ran 2D and 3D simulations using Lin
 129 single-moment microphysics [*Yuh-Lang Lin and Orville, 1983*], including a 256km x 256km
 130 1km resolution simulation where statistical convergence with domain size was reached (e.g.
 131 updraft statistics didn't change with doubled domain size). Because the qualitative results
 132 were generally the same, we show/tabulate a 3D, 200m Morrison microphysics runs with
 133 a domain size of 51.2km x 51.2km instead. In 2D, we were unable to achieve statistical
 134 convergence with domain size, and so ran domains of 410km and 820km for each set of
 135 resolutions. The qualitative results of the different domain sizes were the same, and so
 136 only the results from the 820km domain simulations in 2D will be shown here.

137 This study explores the impact of changing surface Bowen ratio, and by extension,
 138 boundary layer depth, on tropical convective intensity. For each combination of domain
 139 size and resolution, a control simulation meant to represent a tropical ocean is used with
 140 a fixed sea surface temperature (SST) of 300K. For each setup, we also run a simulation
 141 with a more land-like SBR and deeper boundary layer. Exact surface Bowen ratios and
 142 boundary layer depths of simulations are shown in table 1.

143 The SBR is altered by using an evaporative conductance parameter α inserted in the
 144 bulk equation for latent heat flux:

$$LHF = \alpha C_e |v| (q_s - q) \quad (1)$$

145 where LHF is the latent heat flux, C_e is the bulk transfer coefficient, $|v|$ is the magnitude
146 of the wind speed at some height above the surface, q_s is the saturation specific humidity
147 near the surface, and q is the near surface specific humidity. When α is 1, the SBR is at
148 the standard oceanic value. As α decreases, latent heat flux is reduced.

149 Free tropospheric temperature is a known control on convective intensity [*Singh and*
150 *O’Gorman, 2014*], and in the tropics horizontal temperature gradients in the free tropo-
151 sphere tend to be weak [*Charney, 1963*]. When the SBR and boundary layer depth of our
152 simulations is increased by increasing α and maintaining the same SST, free tropospheric
153 temperature falls. This is because a) the boundary layer follows the dry adiabat to higher
154 heights in the high SBR case, and b) increased differences between SST and first model
155 level temperature are needed in order for surface fluxes to balance radiative cooling. To
156 account for this, we increase our SST in the high SBR cases in order to get all simulations’
157 free tropospheric temperature to be nearly the same. Thus we fixed SSTs at 304K in for
158 our high (land-like) SBR simulations. An iterative process was used to yield α values that
159 gave free tropospheric temperatures that were nearly the same in each pair of varying α
160 runs. For our high SBR simulations, we found an α value of 0.342 in the 2D cases, and an
161 α value of 0.250 in the 3D case. The higher SBR runs had slightly higher free tropospheric
162 temperatures (0.05K-0.50K) to exclude the possibility of low SBR cases having stronger
163 high intensity updrafts due to a temperature difference.

164 Our setup for the initial condition simulations was much simpler: 3D, 200m resolution,
165 102.4 km x 102.4km domain size, interactive radiative cooling with Lin microphysics,
166 and a TOGA COARE initial sounding, with no large scale tendencies. Instead of using

167 an evaporative conductance parameter, we simply specified our sensible and latent heat
168 fluxes for a total surface flux of 100 Wm^{-2} with SBRs of 0, 0.25, 0.5, and 1.

3. Results

3.1. Results from the CRM

169 The most intense updrafts of the high SBR cases were never stronger than those from
170 the low SBR case. The top row of figure 1 shows a cumulative distribution function of
171 500 hPa instantaneous vertical velocity highlighting the upper tail of the distribution. At
172 the 99.99th percentile, both cases have nearly the same vertical velocities, at odds with
173 the initial hypothesis that thunderstorms with high SBRs have higher vertical velocities
174 than those with low SBRs.

175 The bottom row of figure 1 shows maximum updraft velocity over a range of sampling
176 intervals. There were no significant differences between simulation pairs, using a Mann-
177 Whitney U test [*Mann and Whitney, 1947*] at the 95% significance level. The discrepancy
178 in maximum vertical velocity between the 2D and 3D simulations may be explainable by
179 the number of samples per time step being lower in the 2D simulations. Vertical profiles
180 of buoyancy flux appear to be very similar over the entire troposphere, except in the
181 boundary layer, which is to be expected given the different SBRs.

182 Inferred entrainment can be visualized from plots of the mass flux per moist static energy
183 bin [*Pauluis and Mrowiec, 2013*], shown in figure 2. As we are interested in metrics of
184 high intensity convection, we took the log of the mass flux, which highlights where rare
185 parcel paths are occurring. Dilution of updrafts by environmental air can be described as
186 follows:

$$\frac{dh_u}{dz} = \epsilon(h_{env} - h_u) \quad (2)$$

187 This gives the rate of change of the updraft moist static energy (MSE), h_u with height,
 188 z as a function of ϵ , entrainment rate, and environmental MSE, h_{env} . h_u is always larger
 189 than h_{env} for updrafts, meaning that entrainment decreases the buoyancy of convective
 190 plumes. Moist static energy is:

$$h = C_p T + gz + L_v q \quad (3)$$

191 Where h is the MSE, C_p is the specific heat capacity of dry air at constant pressure,
 192 T is the temperature, g is gravity, z is height above the surface, L_v is the latent heat of
 193 vaporization, and q is the specific humidity of water vapor.

194 By looking at the portions of the mass flux diagram with the highest moist static
 195 energies we can infer the which simulations are entraining more or less environmental air.
 196 If we assume entrainment rate is a known function of height, we can infer thermodynamic
 197 paths of the high MSE-air parcels. Because parcel MSE is controlled by the amount of
 198 entrainment experienced, higher MSE indicates lower entrainment rates and vice versa.
 199 For each pair of simulations, the maximum moist static energy of anomalously high MSE
 200 updrafts appears to be approximately the same, though the 3D & 2D 200m resolution
 201 simulations have slightly higher peak MSEs for the high SBR case, potentially indicating
 202 that some updrafts experiences slightly less entrainment than the low SBR case. These
 203 difference may also be due to the high SBR cases have slightly higher free tropospheric
 204 temperatures.

3.2. Precipitation Dissipation and related metrics

205 Precipitation dissipation in our simulations had no systematic response to changes in
206 the SBR, as seen in table 1. However, when normalized by the surface precipitation, we
207 see scale height differences that would suggest a land-ocean convective intensity difference,
208 as one sees in lightning per unit precipitation observations. A caveat however: precipi-
209 tation at cloud base is balanced by radiative cooling of the free troposphere [*O’Gorman*
210 *et al.*, 2012], and the free troposphere shrinks when increasing the size of the boundary
211 layer and maintaining free tropospheric temperature. Combined with the expectation of
212 increased precipitation evaporation in a deeper boundary layer, it makes sense that when
213 we normalize our precipitation dissipation by surface precipitation, we get larger values
214 in the high SBR simulations.

215 Electric charges resulting in lightning are likely produced by collisions between graupel
216 and small ice crystals in the presence of water droplets [pp 93, *Rakov and Uman*, 2003],
217 so we examine a dissipation metric that only uses graupel, shown in Table 1. Frictional
218 dissipation by graupel doesn’t depend on SBR in a systematic way. We also calculated
219 the graupel scale height by normalizing the graupel dissipation by the total falling pre-
220 cipitation at the lowest level of graupel existence. This graupel scale height shows a
221 notable trend: high SBR cases have higher graupel scale heights than the low SBR cases.
222 Thus there is more graupel dissipation per unit precipitation at freezing level, potentially
223 providing a source for increased lightning that is independent of updraft velocity. Lin
224 microphysics runs (not shown) did not show this tendency. It would be interesting to
225 further explore the relationship of this metric to lightning in observations.

3.3. Initial Condition Simulations

226 In the initial condition simulations, we examined the mean maximum vertical velocity
 227 over the first day (30 minute sampling interval). For SBR 0, this metric was 39 ms^{-1} ,
 228 for SBR 0.25: 37 ms^{-1} , for SBR 0.5: 34 ms^{-1} , and for SBR 1: 32 ms^{-1} . Hence, low
 229 SBR cases were notably stronger than high SBR cases. Other studies have suggested
 230 that idealized initial condition simulations can be used to produce results in line with real
 231 world observations of the intensity of convection as a function of island size [*Robinson*
 232 *et al.*, 2011], though the proposed mechanism is different than what we tested.

3.4. Analysis with Parcel Model

233 Our simulation pairs are producing high percentile updraft velocities that are very
 234 similar to one another. Using a 1-D parcel model, we check if the same distribution of
 235 entrainment rates gives similar vertical velocities in our simulation pairs to see if we can
 236 explain these results with an entrainment that is independent of boundary layer depth.

237 Undoubtedly, simple functions that approximate entrainment are not 100% correct.
 238 However they have received extensive use: in cumulus parameterizations [*Tokioka et al.*,
 239 1988], for analysis from observations [*Holloway and Neelin*, 2009], as well as for the de-
 240 velopment of frameworks used to explain variations in convection [*Lucas et al.*, 1994;
 241 *Williams and Stanfill*, 2002; *Williams et al.*, 2005; *Singh and O’Gorman*, 2013].

242 We modified the 1-D parcel model from *Singh and O’Gorman* [2013] to use the envi-
 243 ronment from our RCE simulations to calculate entraining parcel paths. In the parcel
 244 model calculation, the parcel is lifted dry adiabatically to the lifting condensation level,
 245 then moist adiabatically to the level of neutral buoyancy. During moist ascent, entrain-

246 ment occurs and modifies parcel moist static energy as in equation 2. Condensate loading
247 is parameterized as a fraction of the total condensed water produced (30% for the fig-
248 ures). Total integrated buoyancy is calculated using the difference in density temperature
249 between the environment and the parcel, and is integrated between the level of free con-
250 vection (LFC) and the level of neutral buoyancy.

251 The following results use the 3D 200m resolution simulation pair. Other cases are
252 qualitatively the same. Figure 3a shows the mean thermodynamic environment of our
253 simulations along with two possible parcel paths with different fixed entrainment rates,
254 ϵ (dashed lines). Figure 3b shows the difference in temperature between the high SBR
255 simulation and the low SBR simulation, as well as their respective relative humidities as
256 a function of height. Figure 3c shows potential maximum updraft velocity (square root of
257 two times integrated buoyancy) as a function of fixed entrainment rate. Figure 3d shows
258 maximum vertical velocity using an explicit vertical momentum equation [*Bretherton et al.*
259 , 2004].

260 We first look at how well our parcel model produces the vertical velocity distributions
261 seen in our CRM by using a range of specified ϵ . This is shown in figure 3c. Clearly, a
262 fixed entrainment rate can give the same lack of variation in updraft velocity as was seen
263 in figure 1.

264 The hypothesis that deeper boundary layers and higher SBRs have greater intensity
265 due to larger updraft proportion [*Lucas et al.*, 1994, 1996; *Zipser*, 2003; *Williams and*
266 *Stanfill*, 2002; *Williams et al.*, 2005] would be modeled with a function of entrainment
267 that is inversely proportional to boundary layer depth:

$$\epsilon_{bl} = \frac{e}{Z_{bl}} \quad (4)$$

268 Where e is a specified entrainment parameter, and Z_{bl} is boundary layer depth. If this
 269 conceptual model were representative of what we see in our simulations, then we would
 270 expect differing ϵ_{bl} (by the factor $\frac{Z_{bl_{highSBR}}}{Z_{bl_{lowSBR}}}$), to produce the same vertical velocities in our
 271 simulation pairs. Clearly, that is not the case, as shown in figure 3c. Instead, the results
 272 of our simulations can be explained with a fixed entrainment rate that is independent of
 273 boundary layer depth.

274 Lower environmental relative humidities can lead to greater effective buoyancy in undi-
 275 lute plumes, potentially favoring a drier (more land-like) environment in the case of the
 276 most intense updrafts [*Singh and O’Gorman, 2013*]. The high and low SBR cases have
 277 very similar free tropospheric relative humidity. However, one might expect more land-like
 278 environments in nature to sometimes have lower free tropospheric humidities than occur
 279 over the ocean. It would be interesting to further explore the effects of environmental
 280 humidity on updraft velocities in the type of simulation setup we used here.

281 We looked at the sensitivity of the results shown in Figure 3c to condensate loading
 282 (from 0% - 100%), as well as our definition for the LFC, but found qualitatively the same
 283 general results, namely that for each pair of simulations, a fixed entrainment rate yields
 284 nearly the same vertical velocities. Variations on our LFC calculation that we used were
 285 a) defining the LFC as where the saturation moist static energy above the surface reached
 286 the value of the moist static energy of the surface, and b) looking for the level of maximum
 287 cloud water, as used in *Kuang and Bretherton [2006]*.

288 Depending on the condensate loading used in the parcel model calculation, it appears
 289 possible that some updrafts may be undilute. Condensate loading values above approxi-
 290 mately 70% give undilute maximum vertical velocity values around $55ms^{-1}$, which would
 291 coincide with an approximately 1 in 1,000,000 occurrence in our 3D simulations. These
 292 values are not observed in our 2D simulations, possibly because there are simply too few
 293 samples.

We also consider the effects of entrainment drag using the following explicit equation for vertical motion:

$$\frac{1}{2} \frac{\partial w^2}{\partial z} = aB - b\epsilon w^2 \quad (5)$$

294 where a and b are constants, and B is the buoyancy from the parcel model. Figure 3d
 295 shows maximum vertical velocities using parameters from *Bretherton et al.* [2004], setting
 296 $a = 1$ and $b = 2$. The magnitude of the maximum updraft velocity is much lower, but
 297 we are able to replicate maximum updraft velocities from figure 3c when altering the
 298 parameters. The distribution of updraft velocities as a function of entrainment rate tends
 299 to be similar between simulations using this model.

3.5. Cloud Widths

300 We checked the mean width of our clouds in these simulations by looking at the number
 301 of connected cloud water points at a given height, and calculating the area. For the 3D
 302 simulations we converted the area into an effective radius, while for 2D simulations, we
 303 used the width. As shown in table 1, the cloud widths at the lowest level of cloudiness were
 304 typically higher for the high SBR simulation, while cloud widths at the level of maximum
 305 cloudiness were generally very similar between simulation pairs. This might explain why

306 our model wasn't producing differences in high intensity updraft statistics: differences in
307 cloud width were the argument made by [*Williams and Stanfill, 2002; Williams et al.,*
308 2005] as to why storms over land entrain less than those over the ocean.

4. Discussion

309 Our goal was to test the hypothesis that land-like surface Bowen ratios contribute to
310 observed convective intensity variations between land and ocean, including differences
311 in lightning flash rate (updraft velocity by proxy). Our CRM simulations showed that
312 changes in the SBR did not seem to produce notable differences in high percentile and
313 maximum updraft velocities. Initial condition simulations produced weaker updrafts when
314 SBRs were higher. Thus, in SAM, variations in the surface Bowen ratio do not produce
315 the convective intensity differences that had previously been predicted, when intensity is
316 described by updraft velocities.

317 The one metric that showed some response to variations in the SBR was the graupel scale
318 height. These results imply that there may be some convective intensity metrics related to
319 lightning flash rate that respond to SBR and are independent of vertical velocity. There
320 were also some variations in the mass flux that indicated high SBR simulations may be
321 entraining less, possibly allowing the high SBR case to have the same vertical velocities
322 despite the higher undilute CAPE in the lower SBR simulations. However, variations in
323 boundary layer depth are not strongly driving entrainment variations as these differences
324 are small.

325 There are questions about the ability of CRMs to produce intense convection repre-
326 sentative of real world processes [e.g. *Varble et al., 2014*]. It is certainly possible that

327 something nonphysical is preventing SAM from producing the expected updraft velocity
328 difference in response to varying Bowen ratios. However, the model includes physics that
329 leads to different boundary layer depths and cloud width differences at cloud base, so
330 should be able to simulate the entrainment processes which have been hypothesized to
331 lead to variations in updraft velocities. In the RCE simulations, cloud widths at the level
332 of maximum cloudiness seem to be consistent with little dependence on SBR.

333 Analysis with a parcel model suggests that the updraft velocity results we are seeing
334 are also consistent with entrainment not depending on SBR. A constant entrainment
335 rate independent of boundary layer depth does a good job of approximating the lack of
336 variation in vertical velocity seen in figure 1.

337 It remains to be seen what actually causes the convective intensity differences between
338 land and ocean in the tropics. A more in-depth analysis of tropical CAPE distributions
339 would be helpful in confirming that CAPE is not a controlling factor for regional variations
340 in convective intensity. Diurnal cycles and the development and decay of convective
341 inhibition could potentially predispose land surfaces towards more intense convection.
342 Larger diurnal cycles occur over regions with higher SBRs, potentially explaining the
343 observed relationship between boundary layer depth and convective intensity. Surface
344 heterogeneities are another potential avenue of pursuit [*Rieck et al.*, 2014], which could
345 potentially follow well with exploration of the surface Bowen ratio.

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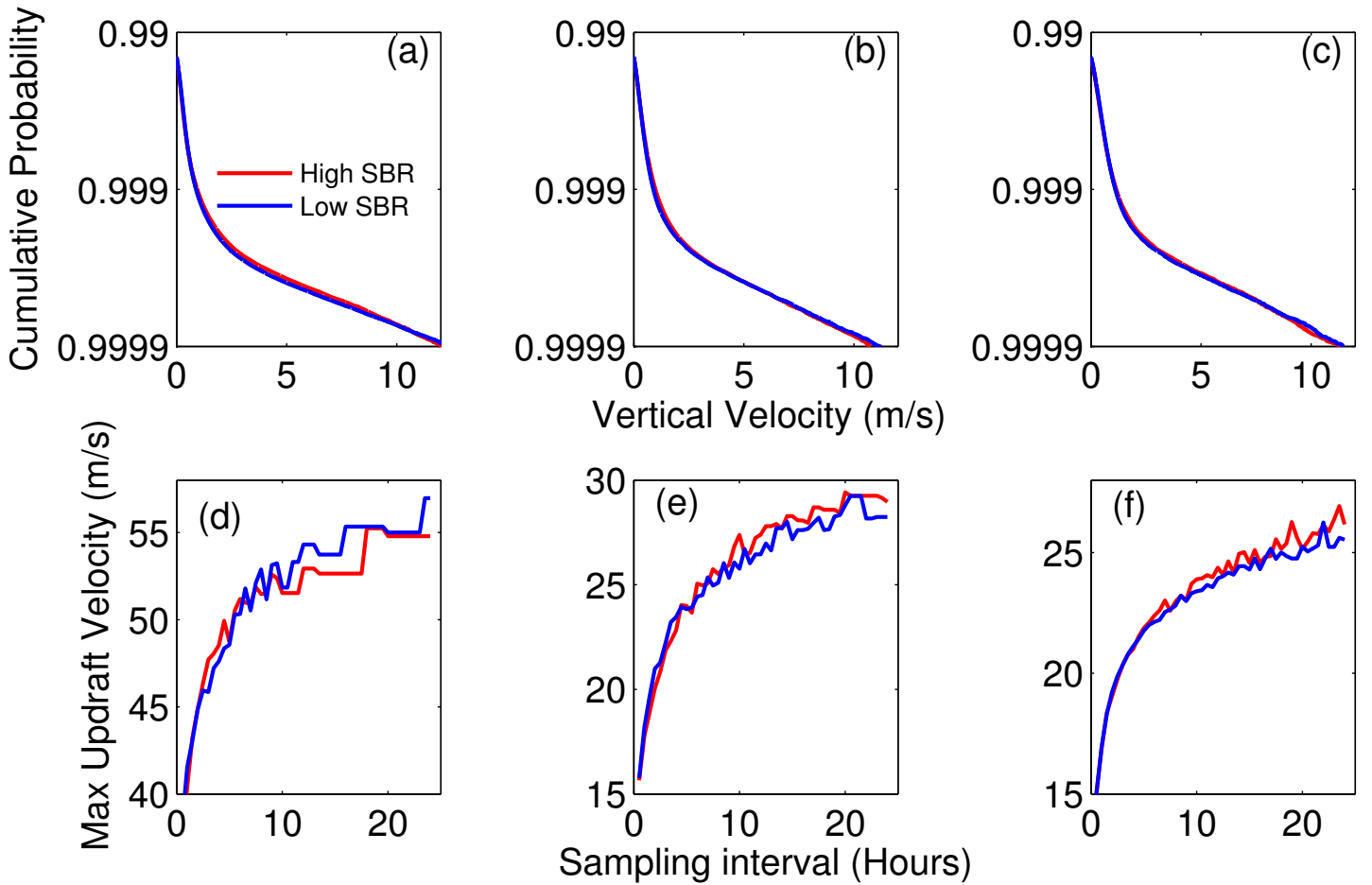


Figure 1. Cumulative distribution function of 500hPa vertical velocity (top row) and maximum updraft velocity as a function of sampling interval (bottom row). Red lines represent high or land-like SBR cases, while blue lines represent low or ocean-like SBR cases. (a) and (d) are 3D 200m resolution Morrison microphysics case, (b) and (e) are the 2D 400m resolution Morrison microphysics case, (c) and (f) are the 2D 200m resolution Morrison microphysics case.

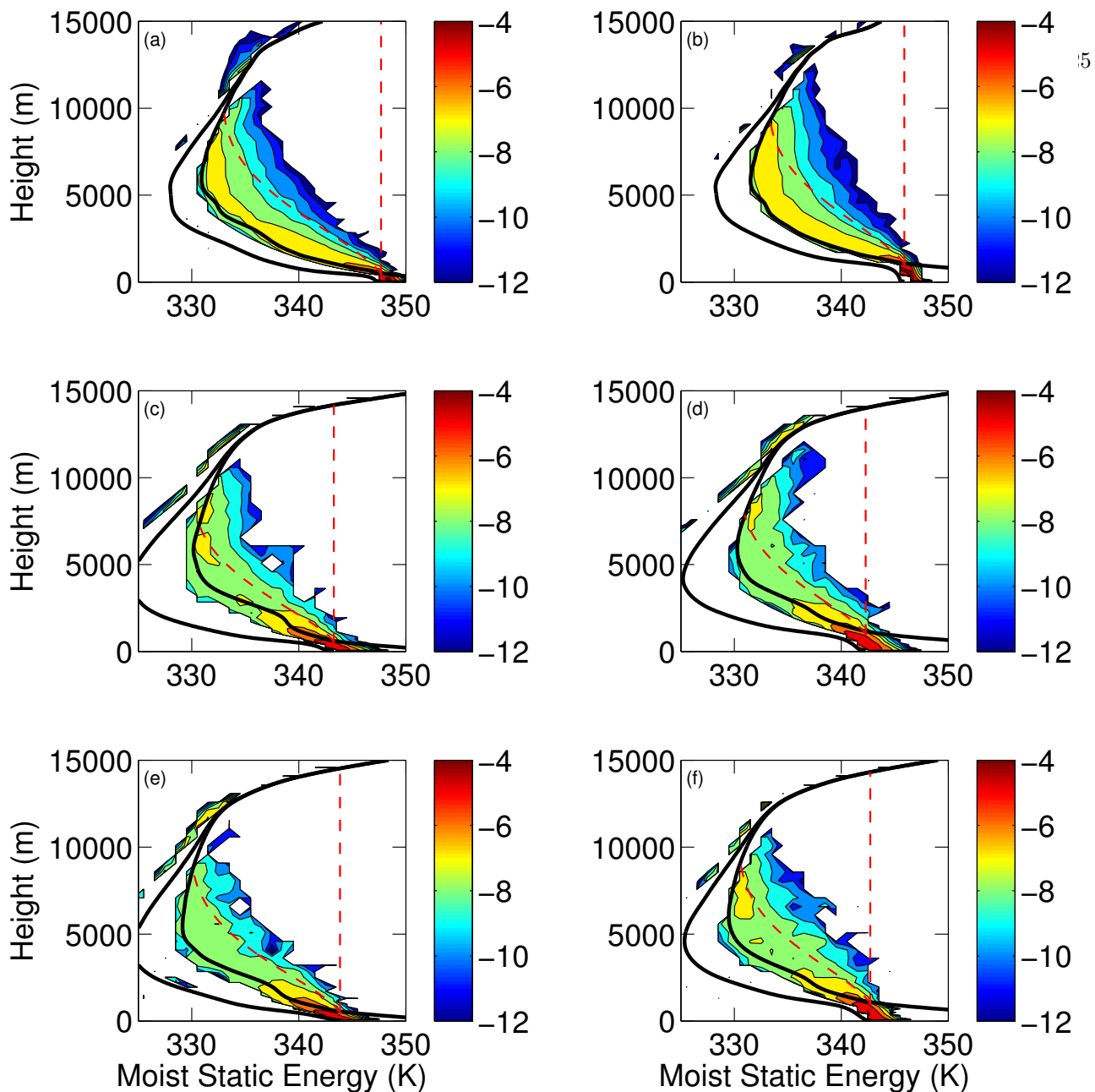


Figure 2. Log of positive mass flux per 1K MSE bin vs height for low SBR cases (left column) and high SBR cases (right column). Black lines are MSE and saturation MSE, while red dashed lines show potential entraining parcel paths with fixed entrainment rates of 0 and 0.3 km^{-1} . (a) and (b) are 3D 200m resolution Morrison microphysics, (c) and (d) are 2D 400m resolution Morrison microphysics simulations, (e) and (f) are 2D 200m resolution Morrison microphysics simulations.

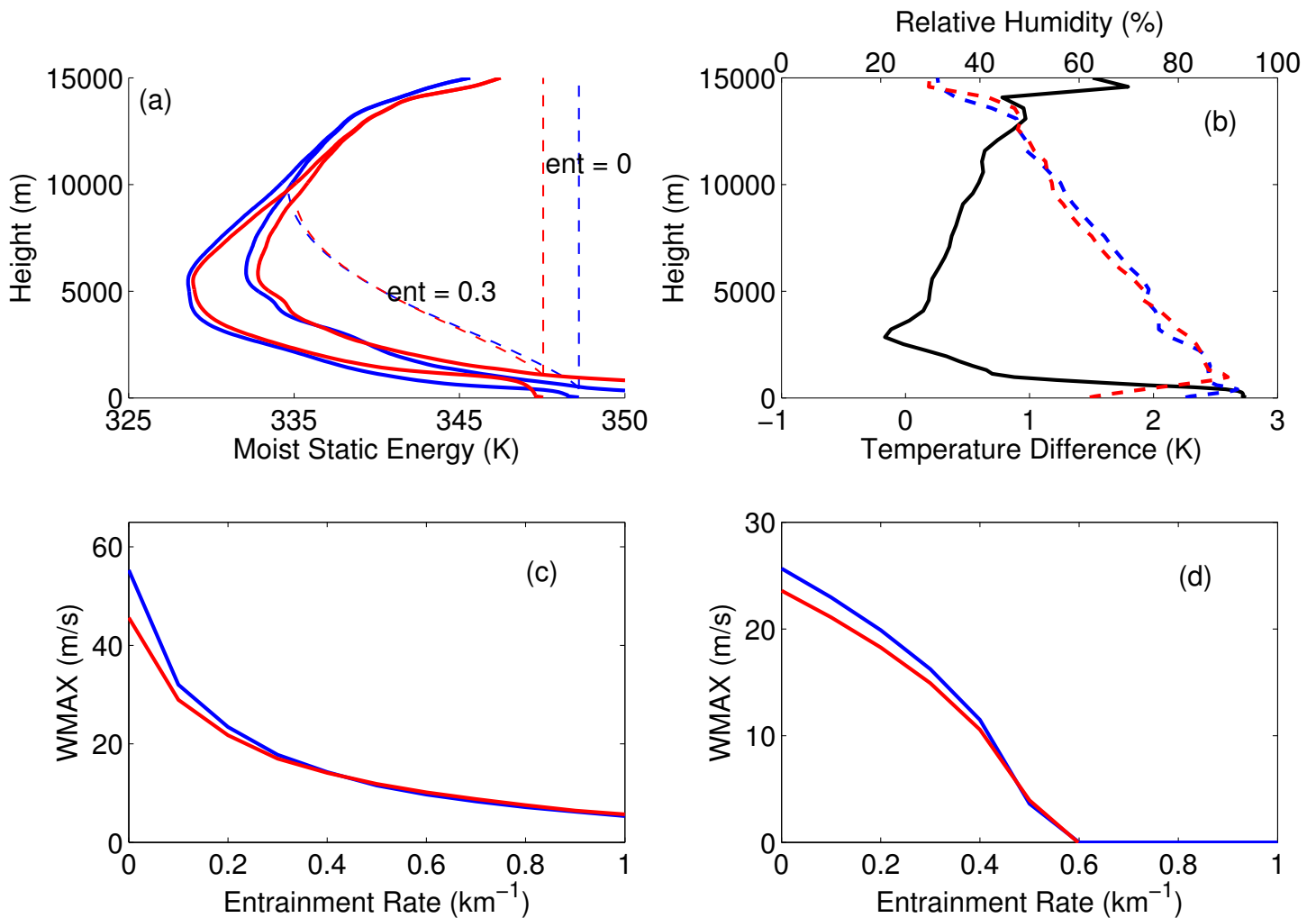


Figure 3. Thermodynamic characteristics, and results from the entraining parcel model for the 3D 200m resolution simulations. (a) Shows the thermodynamic diagram as in figure 2, but for both high and low SBR, where ent is a fixed entrainment rate (km^{-1}). (b) The temperature difference between the high SBR case and the low SBR case (black line) and their respective relative humidities (dashed lines). (c) and (d) give potential maximum updraft velocity as a function of entrainment rate. (c) integrates buoyancy to produce maximum vertical velocity, and (d) shows maximum calculated vertical velocity from the explicit vertical velocity equation. Red lines represent the high or land-like SBR case, while blue lines represent low or ocean-like SBR case.

Simulation	SBR	BLD	D_P	D_G	SH_P	SH_G	X_{Cl}	X_{Cmax}
3D 200m Morrison Low SBR	0.08	475	1.09	0.51	3.93	8.76	1.30	1.70
3D 200m Morrison High SBR	0.39	1025	1.22	0.60	5.45	13.3	1.49	1.64
2D 400m Morrison Low SBR	0.08	575	2.37	0.39	6.33	5.87	0.80	1.4
2D 400m Morrison High SBR	0.23	1025	2.00	0.53	7.83	8.32	1.1	1.4
2D 200m Morrison Low SBR	0.08	525	1.09	0.39	3.70	5.17	1.0	1.7
2D 200m Morrison High SBR	0.23	975	1.00	0.37	4.60	6.43	1.3	1.6

Table 1. The surface Bowen ratio (SBR), boundary layer depth (BLD (m)) defined by the height of the lifting condensation level, precipitation and graupel dissipation (D_P and D_G respectively) as well as their scale heights (SH_P and SH_G), and cloud widths (2D simulation) or effective radius (3D simulation) at the lowest level of cloudiness (X_{Cl}) and the level of maximum cloudiness (X_{Cmax}) for each of the three simulation pairs. Boundary layer depth is meters, dissipation is in units of watts per meter squared and scale height is in units of kilometers. Cloud width is in units of kilometers.