

The Influence of Soil Moisture and Surface Roughness on an Idealized Tropical Cyclone

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Abstract

On occasion, tropical cyclones (TCs) have been shown to strengthen over land, provided that the land is warm and moist. The emergent hypothesis is that the moist surface provides sustaining latent heat flux that is reminiscent of an oceanic environment. To test this hypothesis, numerical simulations of idealized TCs with various profiles of soil moisture availability (SMA) and surface roughness were conducted. SMA gradients are shown to have a large influence on precipitation beyond uniform SMA. The sensitivity of accumulated precipitation to SMA is larger with enhanced friction. The maximum wind speed is more sensitive to differences in SMA under lower surface roughness. Results provide a foundation for refining emerging theories about land–atmosphere interactions with landfalling tropical systems. Additionally, these findings may inform forecasters to consider land-surface conditions when assessing intensity trends for landfalling tropical cyclones, particularly since assimilation of soil moisture and surface characteristics can yield differing impacts.

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The Influence of Soil Moisture and Surface Roughness on an Idealized Tropical Cyclone

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Key Points:

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- Increases in surface roughness can increase precipitation, but at the cost of maximum wind speed.

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- Intensity is more sensitive to soil moisture in areas with a low surface roughness, while the precipitation is more sensitive to areas with a higher surface roughness.

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- Soil moisture gradients were shown to have a larger impact on tropical cyclone precipitation in areas with a higher surface roughness.

14

15 **Abstract**

16 On occasion, tropical cyclones (TCs) have been shown to strengthen over land, provided
17 that the land is warm and moist. The emergent hypothesis is that the moist surface provides
18 sustaining latent heat flux that is reminiscent of an oceanic environment. To test this hypothesis,
19 numerical simulations of idealized TCs with various profiles of soil moisture availability (SMA)
20 and surface roughness were conducted. SMA gradients are shown to have a large influence on
21 precipitation beyond uniform SMA. The sensitivity of accumulated precipitation to SMA is
22 larger with enhanced friction. The maximum wind speed is more sensitive to differences in SMA
23 under lower surface roughness. Results provide a foundation for refining emerging theories
24 about land–atmosphere interactions with landfalling tropical systems. Additionally, these
25 findings may inform forecasters to consider land-surface conditions when assessing intensity
26 trends for landfalling tropical cyclones, particularly since assimilation of soil moisture and
27 surface characteristics can yield differing impacts.

28 **Plain Language Summary**

29 An emerging set of studies have shown that some tropical cyclones either maintain their
30 intensity or strengthen over land. The idea is that in areas with a large amount of soil moisture,
31 the water evaporated from the surface can mimic an ocean environment and impact the cyclone.
32 In order to investigate this emerging concept, a simulation of a hurricane within a controlled box
33 was conducted with a weather model. An array of additional experiments were conducted with
34 various land conditions and soil moisture instead of water. It was found that more precipitation,
35 but weaker winds, were found in the hurricane over cropland rather than bare ground. Spatial
36 differences in soil moisture were also very influential to rainfall, especially for accumulated
37 rainfall.

38 **1 Introduction**

39 It is commonly accepted that a tropical cyclone (TC) typically dissipates after landfall
40 (John Kaplan & DeMaria, 1995; Zhu, 2008) unless it undergoes extratropical transition (Evans et
41 al., 2017; Keller et al., 2018). The reasons for the dissipation include the absence of moisture
42 source (Shen et al., 2002; Tuleya & Kurihara, 1978), and the existence of shear (J. Kaplan &
43 Demaria, 2001). The influence of soil moisture may slow the dissipation to a negligible rate, and
44 even reverse the dissipation rate while maintaining characteristics of a TC (Andersen et al., 2013;
45 Andersen & Shepherd, 2014; Arndt et al., 2009; Nair et al., 2019). This is termed the Brown
46 Ocean Effect (BOE), as the ground surface is presumed to be so moist and warm, that the
47 moisture flux is comparable to that from over an ocean (Andersen et al., 2013). The impact of the
48 BOE is not a binary categorization, but rather a signal with a spectrum of influence (Yoo et al.,
49 2020). The BOE differs from other studies that hypothesize the existence of cyclogenesis over
50 land (Cronin & Chavas, 2019; Mrowiec et al., 2011), in that the tropical cyclone has been formed
51 a priori and is influenced by surface fluxes post-landfall.

52 The BOE hypotheses have primarily assumed a constant soil moisture distribution.
53 However, the assumption of uniform soil moisture is not appropriate for a realistic environment,
54 but is still consistent within the literature that supports the BOE (Andersen & Shepherd, 2014).
55 Previous studies that examined the influence of soil moisture gradients on TCs include Tropical
56 Storm Erin (2007; Arndt et al., 2009; Kellner et al., 2012; Monteverdi & Edwards, 2010) and
57 Hurricane Danny (1997) and Hurricane Fran (1996) (Kehoe et al., 2010). Kellner et al. (2012)

58 hypothesized that the soil moisture gradient helped produce a gradient in vorticity, which helped
59 to reintensify Tropical Storm Erin, which studies have suggested to be a BOE case (Arndt et al.,
60 2009; Monteverti & Edwards, 2010). This finding is consistent with Evans et al. (2011). Kehoe
61 et al. (2010) proposed that the enhancement of Hurricane Fran was due to soil moisture
62 gradients, drawing from an analogy of other mesoscale circulations induced by differences in
63 land use (Hong et al., 1995; Ookouchi et al., 1984). Kehoe et al. (2010) also indicated that
64 Hurricane Danny had local maxima in precipitation in areas where the soil moisture gradient was
65 prominent.

66 Previous studies also suggest that the intensity of TCs is dependent on the surface drag
67 coefficient (Bryan, 2013; Emanuel, 1995; Malkus & Riehl, 1960). The surface drag coefficient is
68 dependent on surface roughness length and the Monin-Obukhov length (Powell et al., 2003;
69 Stull, 2009). The surface roughness is a featural difference between the land surface and oceanic
70 surface which is another aspect in which the BOE is different from the typical intensification of
71 TCs. Changes in the roughness length may reduce tropical cyclone intensity overall, but may
72 also induce convergence, enhancing local winds (Zhu, 2008). Increases in surface drag have also
73 been proposed to be a mechanism for the enhancement of precipitation in tropical cyclones
74 (Zhang et al., 2018).

75 The goal of this research is to demonstrate the validity of the BOE from a theoretical
76 perspective, as well as test the aforementioned deviations from a typical water surface, which is
77 conducive to tropical cyclone intensification, to a land surface with varying characteristics. A
78 simulation of an idealized tropical cyclone was used to conduct a series of experiments replacing
79 the water surface with surface roughness and patterns of soil moisture availability (SMA; Lee &
80 Pielke, 1992) beneath a developed cyclone. Section 2 provides an overview of the data and
81 methodology, and results are presented in section 3. Section 4 summarized key conclusions and
82 points of discussion.

83 **2 Data and Methods**

84 The Weather Research and Forecasting Model (WRF) version 3.8 was used to simulate
85 an idealized TC. Specific changes to the default configuration of WRF include the deactivation
86 of radiation, convective, and land-surface parameterization, as well as a domain of 984 km x 984
87 km with 4 km resolution. A control simulation (CTRL) with a water surface was run for a 10 day
88 period. After a two day period, the restart file of CTRL was altered by replacing the water
89 surface with different land-use types and SMA profiles. Two different land-use categories were
90 used, namely “Bare Ground” (BG; $z_0=0.01$ m) and “Mixed Cropland” (MC; $z_0=0.1$ m) land use
91 types. Since the land-surface parameterization was deactivated, the SMA profile was non-variant
92 with unintended feedback mechanisms suppressed. One limitation with this approach is that
93 while the TC may move, the soil moisture profile does not change. Eleven of the fourteen SMA
94 profiles consisted of uniform SMA, ranging from 0 to 1. Three non-uniform SMA profiles were
95 also used: a parabolically weighted Gaussian distribution (wG), inverse of the weighted Gaussian
96 distribution (iwG), and piecewise (Pw). Those three non-uniform SMA profiles are described by
97 Table 1, where x' and y' are normalized coordinates relative to the minimum central pressure,
98 and R is the radius from the minimum central pressure. Particular simulations will be referred to
99 as land use type, followed by the SMA profile. For example, BG-wG will refer to the bare
100 ground simulation with the weighted Gaussian distribution and MC-U0.3 will refer to the mixed
101 cropland simulation with a uniform SMA of 0.3.

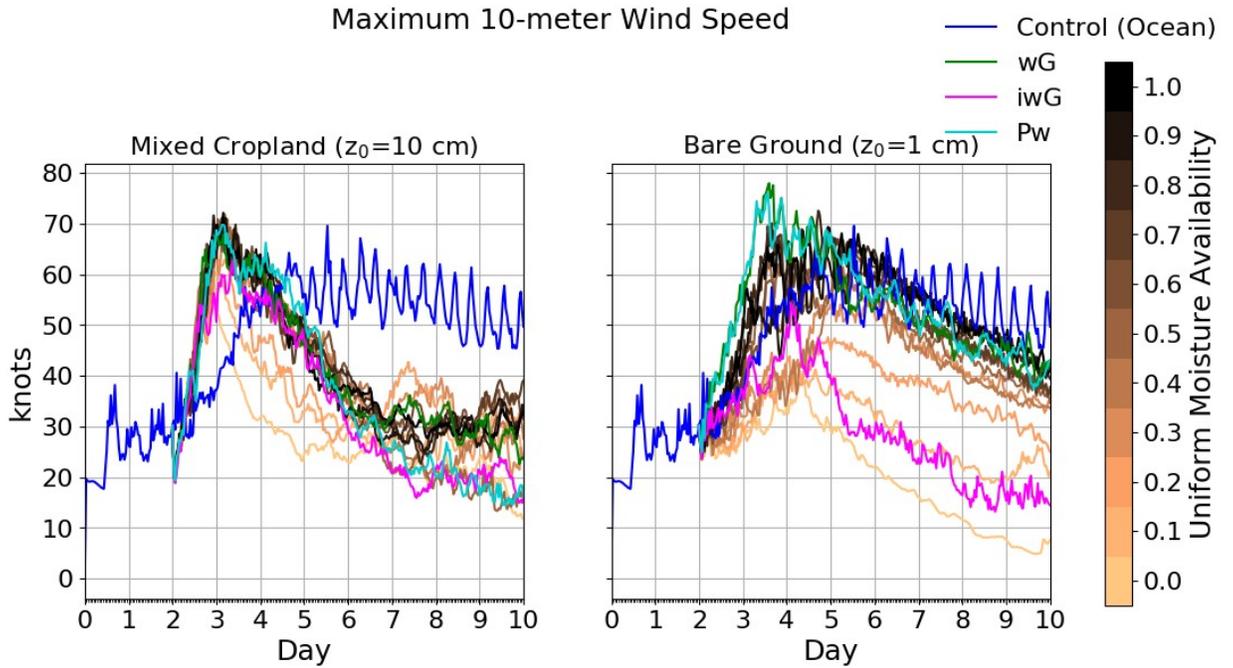
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103 **Table 1:** Details and equations describing the non-uniform soil moisture availability profiles.

Long Name	Abbreviated Name	Expression	Reason
Parabolically weighted Gaussian	wG	$\left(1 - \frac{[x']^2}{2} - \frac{[y']^2}{2}\right) \exp(-\dots)$	Moist near center, dry at edge of domain
Inverse weighted Gaussian	iwG	$1 - \left(1 - \frac{[x']^2}{2} - \frac{[y']^2}{2}\right) \exp(-\dots)$	Dry near center, moist at edge of domain
Piecewise	Pw	$\begin{cases} 0 & \text{if } R > 250 \text{ km} \\ 1 & \text{if } R < 250 \text{ km} \end{cases}$	Moist near center. Strongest SMA gradient.

104 **3 Results**

105 Figure 1 shows the maximum instantaneous wind speed for the BG and MC land use
106 types for all 14 SMA profiles. CTRL achieves an asymptotically stable (Kieu, 2015) quasi-
107 steady state (QSS) shortly after rapid intensification. Although CTRL achieves a QSS, the BOE
108 experiments decay at varying rates, consistent with Kaplan & DeMaria (1995). As expected, the
109 maximum wind speed for all of the BG simulations were generally greater than the MC
110 simulations. One important difference between the CTRL simulation and the BG/MC
111 simulations is the onset of rapid intensification (RI), which is defined as the increase of the
112 maximum wind speed by 30 knots in 24 hours or less. RI onset occurs earlier in the BG/MC
113 simulations, with the exception of the iwG and drier uniform SMA distributions of the BG
114 simulations. The onset of RI happens earlier in the MC simulations than the CTRL. Moreover,
115 the BG-wG and BG-Pw have a more pronounced period of RI. The maximum wind speed of the
116 uniform BG simulations during the QSS are more incremental than the MC simulations, which
117 shows less distinguishable pattern. To this extent, the maximum wind speed in the QSS for the
118 MC-Pw and MC-iwG simulations are almost indistinguishable. The maximum wind speed in the
119 BG simulations during the QSS are more sensitive to changes in the SMA, especially near the
120 center of the TC, than the MC simulations.



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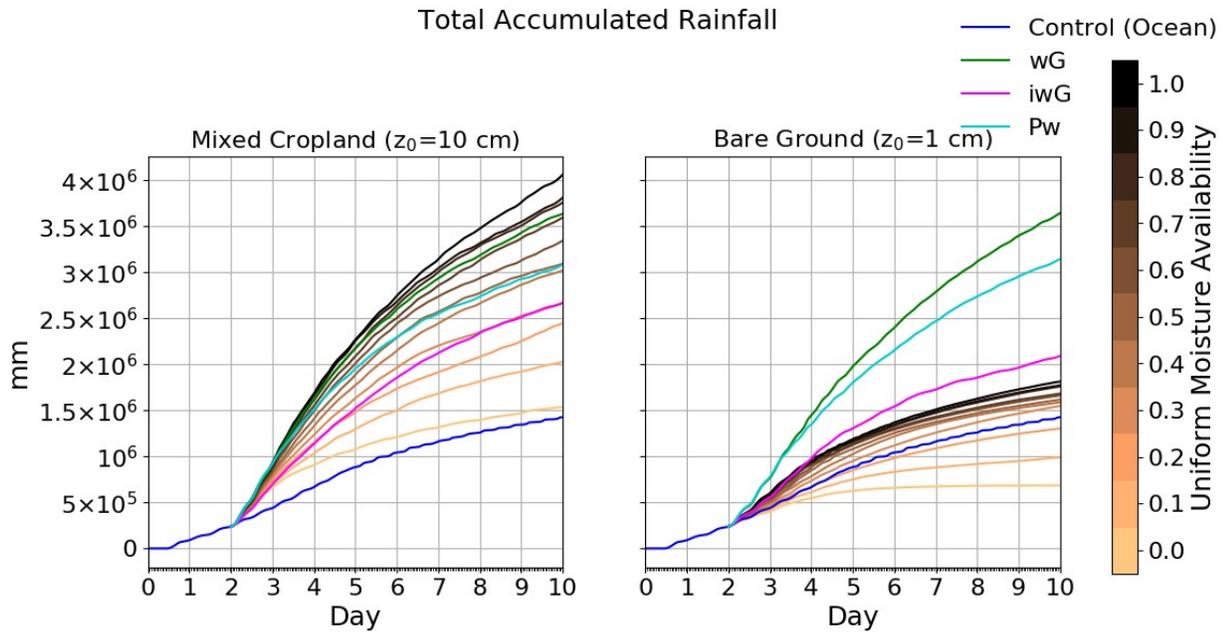
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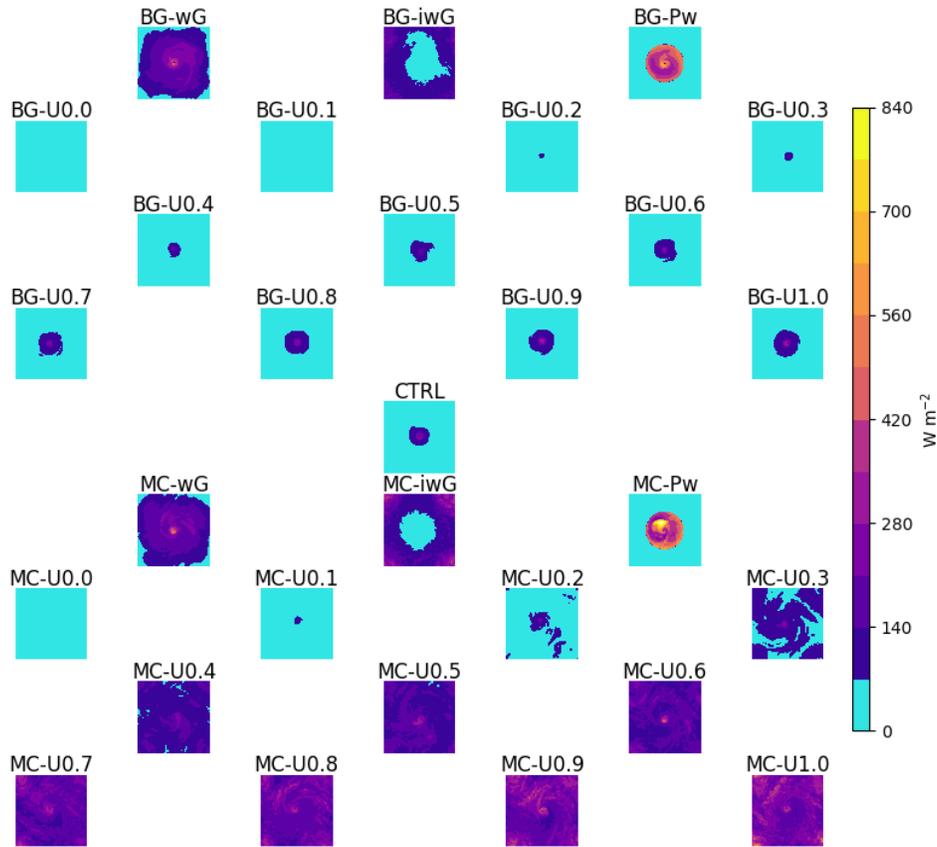
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Figure 1: Maximum instantaneous wind speed (knots) for idealized tropical cyclones over a water surface (CTRL), over Mixed Cropland (left) and Bare Ground (right). The uniform SMA profiles are labeled according to the colorbar, while the CTRL and non-uniform SMA profiles are labeled in the legend.



127 **Figure 2:** Domain-summed accumulated precipitation (mm) for idealized tropical
 128 cyclones over a water surface (CTRL), over Mixed Cropland (left) and Bare Ground (right). The
 129 uniform SMA profiles are labeled according to the colorbar, while the CTRL and non-uniform
 130 SMA profiles are labeled in the legend.

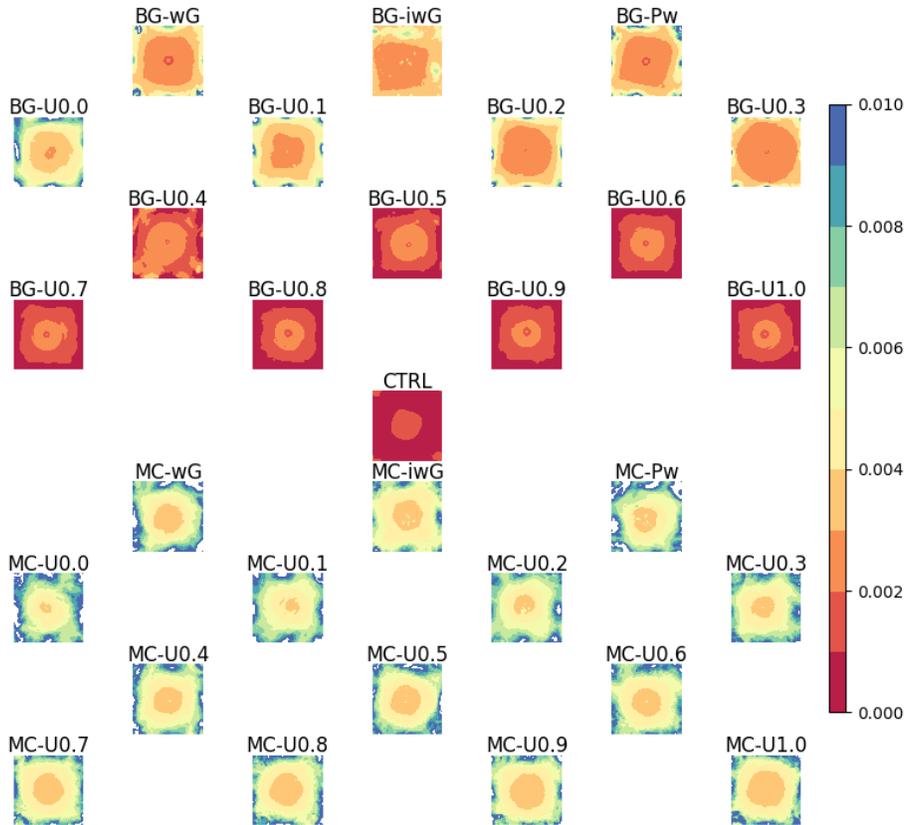
131 Figure 2 shows the domain-summed total accumulated precipitation over MC (left) and
 132 BG (right). All of the MC simulations produced more precipitation than the CTRL simulation.
 133 The CTRL simulation, however, produced more precipitation simulations with a lower SMA
 134 than the BG-U0.3. The presence of a gradient in SMA had a large impact in the total
 135 accumulated precipitation. The influence of the land use type was stronger on the uniform SMA
 136 profiles than the SMA profiles that had a SMA gradient. That is, the BG-iwG produced more
 137 precipitation than the BG-U1.0, while the MC-U1.0 produced more precipitation than the MC-
 138 wG. Land use has a small impact on the total accumulated precipitation generated by SMA
 139 gradients. BG-wG and MC-wG, as well as MC-Pw and BG-Pw, have comparable total
 140 accumulated rainfall amounts, however MC-iwG produced more precipitation than BG-iwG
 141 indicating that there may be a radial dependence to this relationship. The cause of this is likely
 142 due to the differences in sensitivity of the surface latent heat flux.



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Figure 3: Surface latent heat flux (W m^{-2}) for each simulation at hour 180.



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Figure 4: Enthalpy exchange coefficient after 180 hours of the CTRL simulation.

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Figure 3 shows the surface latent heat flux, which is computed by the surface layer parameterization, after 180 hours of the CTRL simulation. This demonstrates that the latent heat flux of the MC-U0.2 and BG-U0.5 most resemble the latent heat flux of the CTRL simulation. Latent heat flux increase is larger in the MC simulations than the BG simulations, even though the wind speed is lower. The cause of the difference in latent heat flux between land-use categories is due to enhancements in the bulk enthalpy transfer coefficient, depicted by Figure 4. The greatest amount of latent heat flux is found in the simulations with the Pw profile. The Pw SMA profiles also have the largest gradient in latent heat flux, as expected with the gradient in the SMA profile. The enhanced soil moisture gradient allows dry air from outside the radius to be advected within the storm, enhancing the moisture gradient between the surface and lower atmosphere, further increasing the latent heat flux.

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4 Discussion and Conclusions

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Through the use of idealized simulations of tropical cyclones, the plausibility of the BOE was confirmed. The explanatory power of the BOE through the enhancements of latent heat flux ignores the reality of SMA gradients, which can produce more precipitation than the BOE itself in areas with a lower roughness length. Moreover, increases in friction enhance the precipitation produced, at the cost of hurricane intensity. Thus, it is proposed that the BOE should be evaluated among two different modes, precipitation enhancement and intensification/maintenance. The pattern of the sudden enhancement of RI due to heightened surface friction is consistent with Montgomery et al. (2010), and the reduction in the QSS is

167 consistent with Bryan (2013). Enhancements in precipitation are more likely in areas that have
168 more friction and weaker wind speeds. The effect of friction on precipitation suggests that
169 hurricane-related flooding is enhanced in urbanized areas, which is consistent with the study by
170 Zhang et al. (2018) on Hurricane Harvey.

171 Some caveats should be mentioned concerning this study. This was a modeling study in
172 an idealized environment, so the conditions may not perfectly align in observational studies. This
173 includes the expanse of one singular land use type, as well as the validity of parameterizations
174 used by the simulation, and the presence of environmental shear. Some of the assumptions used
175 also may reduce feedback mechanisms (such as evaporative cooling decreasing in surface
176 temperature) that could be an artefact of a stagnant tropical cyclone rather than a tropical cyclone
177 moving over an infinite expanse. Deactivating these settings also eliminates potentially relevant
178 feedback mechanisms and signals (Subramanian, 2016; Tang et al., 2019; Tang & Zhang, 2016).
179 This experiment also does not test changes to surface temperature or gradients in surface
180 temperature. Also, the method of replacing the land surface does not induce asymmetries typical
181 of a landfalling tropical cyclone. Despite these shortcomings, these simulations demonstrate the
182 importance of having accurate representation of soil moisture profile and surface features.

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