

1 **Impact of Global Warming on U.S. Summertime Mesoscale Convective**
2 **Systems: A Simple Lagrangian Parcel Model Perspective**

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9 ABSTRACT: Mesoscale convective systems (MCSs) are the dominant rainfall producer over the
10 U.S. during the warm season, causing natural disasters and severe weather every year. Global cli-
11 mate models have large uncertainty in projecting precipitation changes in the future climate. Here
12 a simple Lagrangian parcel model is used to investigate the impact of global warming on MCS
13 initiation and growth. The single-column parcel model projects a mean precipitation decrease over
14 the central U.S. and an increase to its east, in agreement with the CMIP5 model projection. It also
15 highlights the crucial role of current climate mean state model bias in exaggerating the change in
16 future mean precipitation projection by 25%. As for convective population, the model captures the
17 decreased occurrence frequency of weak to moderate convection and increased frequency of strong
18 convection due to the increased CAPE and CIN, in agreement with convection-permitting regional
19 simulations. Novel parameterizations of gust front propagation speed and subsidence strength are
20 developed as guided by cloud-resolving simulations. The multi-column parcel model employing
21 those parameterizations captures readily the cold pool-induced upscale growth feature. It simulates
22 smaller mesoscale clusters over the central U.S. under global warming due to gust front slowdown
23 and subsidence strength enhancement, which is further attributed to land aridity-induced weak-
24 ening of initial accumulated precipitation and strengthening of updraft speed, respectively. That
25 said, mesoscale clusters could become bigger under more favorable conditions in future climate,
26 including boundary layer moistening, convection life time lengthening, and cold pool mechanical
27 lifting enhancement, which require further investigations to improve mechanistic understanding of
28 future MCS changes.

29 1. Introduction

30 A mesoscale convective system (MCS) is a complex of thunderstorms that is typically organized
31 over hundreds of kilometers and persists for several hours or more (??). MCSs are ubiquitous in
32 tropics, subtropics, and midlatitudes, having a significant impact on local weather and hydrologic
33 cycle through massive release of rainfall (????????). In particular, long-lived and intense MCSs
34 are frequently initiated to the east of the Rocky Mountains and propagate eastward afterwards,
35 accounting for more than 50% of warm season precipitation in the Great Plains (?). Due to its
36 higher intensity and larger area coverage than non-MCS, MCS rainfall produces more surface
37 runoff than non-MCS rainfall (?) and contributes to the majority of slow-rising and hybrid floods
38 during the warm season in the U.S. (?). Besides, the dynamics of MCSs in the central U.S. show
39 significant interactions with eastward propagating large-scale environments (??) as well as soil
40 moisture (?). The crucial role of MCSs in influencing atmospheric conditions over the U.S. provides
41 a key benchmark for testing the skills of the present-day numerical models for weather forecast
42 and climate change projection. It is hypothesized that the long-standing warm and dry bias in the
43 summertime central U.S. is related to the failure of coarse-resolution general circulation models
44 (GCMs) in simulating MCSs and the associated precipitation (?????). In contrast, recent studies
45 based on the multiscale modeling framework (??) and convection-permitting regional simulations
46 (??) show some encouraging progress in capturing realistic features of MCSs and the associated
47 precipitation. That said, a better understanding of the underlying mechanisms throughout the
48 whole MCS life cycle is still lacking.

49 The crucial role of MCSs in the water cycle underscores the necessity of investigating its future
50 changes under global warming. Thanks to the improved modeling and computing capabilities,
51 convection-permitting regional simulations have now become a useful and practical tool to investi-
52 gate the behaviors of MCSs and precipitation response in future climate. For example, ? undertook
53 two 13-year high-resolution climate change simulations at convection-permitting resolution, one
54 of which is driven by initial and boundary conditions from current climate and the other is forced
55 with warmer and moister perturbed boundary conditions in the pseudo-global warming (PGW)
56 framework. The present-day convection permitting regional simulation was validated against ob-
57 servational datasets and the PGW experiments were proved to be useful for demonstrating a new
58 mechanism for warm-season precipitation response to global warming (?), investigating the future

59 intensification of hourly precipitation extremes (?), and simulating the convective precipitation
60 diurnal cycle in both current and future climate (?). In particular, ? showed that weak to moderate
61 convection will decrease and strong convection will increase in frequency in a future climate,
62 which was further attributed to the increasing convective available potential energy (CAPE) and
63 convective inhibition (CIN) downstream of the Rockies in a future climate. ? further concluded
64 that intense summertime MCS frequency will more than triple in North America and the combined
65 effect of increasing maximum precipitation rates and heavy precipitation area contributes to 80%
66 increases in the total MCS precipitation volume. These studies indicate the significant impact of
67 global warming on both convective initiation and MCS genesis, the two early stages of MCS life
68 cycle. However, due to the heavy computational expense, using convection-permitting regional
69 simulations is not an affordable way to discuss the impact of global warming on MCS initiation and
70 growth under various favorable/unfavorable conditions and complex mesoscale forcing, let alone
71 isolating key mechanisms response for its behavior change in future climate. Furthermore, even
72 convection permitting modeling faces challenges in realistically simulating summertime MCSs in
73 the central U.S. (?), motivating the need to further investigate how summertime MCSs may change
74 with global warming.

75 To theoretically understand how summertime MCS would change under global warming, it
76 is important to isolate the key mechanisms of change related to convective initiation and MCS
77 genesis, respectively. Previous studies used single-column models to discuss thermodynamic
78 and dynamic constraints for convective initiation (????). The upscale growth feature during
79 MCS genesis resembles the convective self-aggregation phenomenon that scattered small-scale
80 convection is aggregated to several isolated mesoscale clusters in radiative-convective equilibrium
81 simulations. The underlying mechanisms for convective self-aggregation are attributed to the
82 interactions between clouds, moisture, radiation, surface fluxes, and circulation (??????), boundary
83 layer diabatic processes (?), and convection-gravity wave interactions (???). In particular, it has
84 been shown that cold pools resulting from downdrafts and rain evaporation help trigger new
85 convection in the vicinity of existing convective storms, leading to convective organization (??).
86 Novel tracking algorithms were designed to track cold pool gust fronts, confirming that cold
87 pools can influence the initiation of new convective cells (???). High-resolution cloud-resolving
88 simulations further highlight the stronger convection triggering effects by cold pool collision than

89 that by isolated cold pools (??). ? conceptualized cold pool interactions into a parameter-free
90 mathematical model and successfully developed a simple model for convective self-aggregation
91 (?). That said, a simple unified framework for modeling both convective initiation and MCS genesis
92 and isolating their corresponding bare-bone mechanisms is still lacking.

93 Recently, ? developed a simple Lagrangian parcel model for capturing the basic features of
94 summertime convective initiation over the central U.S. and understanding the salient mechanisms
95 necessary for reproducing the upscale growth feature during MCS genesis. To achieve that, a
96 simple single-column Lagrangian parcel model is first developed by considering conservation laws
97 of mass, momentum, and energy, condensation and precipitation, as well as entrainment process
98 for a rising parcel. This single-column model was useful for investigating the collective effects of
99 boundary layer moistening and dynamical lifting on convective initiation, studying the diurnal cycle
100 of convective initiation and identifying the differences between cases with/without MCS initiation,
101 and exploring the impact of global warming on convective initiation and theoretically predicting the
102 convection population under climate change. Then this single-column model was further extended
103 to a multi-column model that includes an array of single-column models aligned in the east–west
104 direction and incorporates idealized cold pool interaction mechanisms. The multi-column model
105 captures readily the cold pool–induced upscale growth feature during MCS genesis from initially
106 scattered convection in a few hours. It also highlights the crucial role of lifting effects due to cold
107 pool collision and spreading, subsidence effect, and gust front propagation speed in controlling the
108 final size of mesoscale clusters and cold pool regions. Nevertheless, all these controlling factors
109 were fixed as some constant value for simplicity but should vary in a more realistic way when
110 assessing the impact of global warming.

111 The overall goal of this study is to investigate the impact of global warming on the convective
112 initiation and genesis of summertime MCSs over the U.S. based on this simple Lagrangian parcel
113 model and highlight the key underlying mechanisms responsible for the MCS behavior changes
114 during the convective initiation and MCS genesis stages of MCS life cycle.

115 All our discussion in this study is based on the simple Lagrangian parcel model (?), includ-
116 ing a single-column model for convective initiation and a multi-column model for MCS genesis,
117 but additionally with some modifications. Specifically, the single-column model is modified by
118 using a better parameterization of entrainment process, while the multi-column model is further

119 improved by incorporating realistic parameterizations of gust front propagation speed and subsi-
120 dence strength guided by cloud-resolving simulations. After specifying environmental profiles and
121 initial conditions, the parcel behavior in the single-column model is used to assess the favorability
122 for convective initiation, while the collective behavior of parcels in the multi-column model is
123 interpreted as the upscale growth feature during MCS genesis. To examine the impact of global
124 warming, we choose to use the PGW approach (???) where the future climate is approximated by
125 the current climate plus a climate perturbation derived from GCM ensemble projection of mean
126 climate change. To assess the impact of global warming on convective initiation, we run the
127 single-column model multiple times for each 1deg-by-1deg grid over the U.S. driven by all hourly
128 environmental profiles during a 14-year summertime period, and investigate the global warming
129 driven changes in terms of climatological mean precipitation, its sensitivity to current climate mean
130 state model bias, and convective population. To assess the impact of global warming on MCS
131 genesis, we run the multi-column model with the large-scale environmental profile averaged over
132 the central U.S. and investigate the global warming driven changes in mesoscale cluster size, cold
133 pool area, and convective cell number.

134 The results highlight the key mechanisms responsible for the MCS behavior changes during
135 the convective initiation and MCS genesis stages in the future climate. By using the single-
136 column model, we first investigate the future climatological mean precipitation changes under
137 global warming and compare with the CMIP5 model projection. Then we discuss the changes
138 in convective population as well as their dependence on geographic location. The simple model
139 provides an analog testbed to illustrate the impact of biases in current climate in predicting future
140 mean precipitation in the GCM. The multi-column model is then used to discuss the potential
141 changes of MCS final state caused by land aridity, boundary layer moistening, convection lifetime
142 lengthening and cold pool mechanical lifting enhancement under global warming.

143 The rest of the paper is organized as follows. Section 2 summarizes the single- and multi-column
144 models and particularly the modifications, input datasets, and experimental setup. Section 3 uses
145 the single-column model to assess the impact of global warming on convective initiation in terms
146 of climatological mean precipitation, their sensitivity to current climate mean state model bias,
147 and convective population. Section 4 uses the multi-column model to assess the impact of global
148 warming on MCS genesis, emphasizing the dependence of MCS genesis on gust front propagation

149 speed and subsidence strength, and illustrating the changes in MCS genesis due to land aridity
150 and other favorable conditions under global warming. The paper concludes with a discussion in
151 section 5.

152 **2. Methodology**

153 In this section, we first briefly summarize the derivation details and basic properties of the single-
154 column and multi-column Lagrangian parcel models, which was developed by ? to understand
155 the initiation of summertime MCSs over the central U.S. Then we describe the parameterization
156 development for gust front propagation speed and subsidence strength under the guidance of two
157 sets of cloud-resolving simulations. Lastly, we document all the information about input data for
158 current and future environmental profiles as well as experiment setup details.

159 *a. Single-column Lagrangian parcel model*

160 The single-column Lagrangian parcel model describes the ascent trajectory of a buoyancy-
161 driven parcel from the surface to the upper troposphere, mimicking the initiation of small-scale
162 individual convection cells during the early stage of MCS initiation (?). The model considers
163 several key physical processes during the ascent motion of the parcel, including the imbalance
164 between buoyancy, gravity, and momentum drag, the phase change between water vapor and liquid
165 water, and the environmental mass exchange through entrainment. The readers are referred to ?
166 for detailed governing equations (Table 1 of ?), key parameters and microphysical closure (Table
167 2 of ?), and explanations (Section 2a of ?).

168 Compared with ?, we further improve the parameterization of entrainment process by allowing
169 its dependence on updraft speed (see ϵ in Table 1). Specifically, besides its dependence on parcel
170 volume, the strength of entrainment rate ϵ is modified to be linearly proportional to the updraft
171 speed w^+ so that entrainment process is significant only if the updraft speed of the parcel is strong
172 and positive (?). Without this modification, the entrainment rate of the parcel is always significant,
173 even in the initial and final stages when the updraft speed is low. Consequently, a parcel that is
174 initialized with a less favorable condition for convection can still rise to upper levels due to the
175 entrainment of environmental moisture and ends up with a large mass in the equilibrium state. Such

176 an unrealistic scenario was not considered in ? where parcels are typically initialized with favorable
 177 conditions for convection, but becomes possible in this study with general initial conditions.

178 TABLE 1. Cloud microphysical closure and parameterizations of gust front speed and subsidence strength in
 179 the simple Lagrangian parcel model. Only modified or new parameters are listed here where the parenthesis
 180 indicates the original value or expressions in ?. The remaining parameters in the model are the same as ?. The
 181 notation w^+ means $w^+ = w$ if $w \geq 0$ and 0 otherwise.

Category	Symbol	Value	Description
Cloud microphysical closure	ϵ	$\epsilon = \frac{1}{\tau_{en}} \rho \left(\frac{V^*}{V} \right)^{1/3} \frac{w^+}{w^*}$	mass entrainment rate
		$(\epsilon = \frac{1}{\tau_{en}} \rho \left(\frac{V^*}{V} \right)^{1/3})$	
Parameterization of gust front speed	w^*	10 m/s	reference updraft speed
	s_{gf}	5 m/s or parameterized (5 m/s)	propagation speed of gust fronts
	α	10^{-5} m/s/kg	ratio coefficient between gust front speed and accumulated surface precipitation
	β	2.0×10^{-3} m/s ²	decaying rate of gust front speed in time
	U_0	4.5 m/s	baseline gust front speed
	δU	5.0 m/s	gust front boost speed
	T_1, T_2, T_3	2.5, 5.0, 10.0 min	predefined timing since the onset of surface precipitation
	ξ	0.2	fraction of surface precipitation loss due to rain evaporation
Parameterization of subsidence strength	V_d	10 m/s	raindrop falling speed
	σ	3.5	ratio coefficient of subsidence below 2 km and that at 11 km
	w_{up}^*	1.9 m/s or parameterized (1.9 m/s)	averaged updraft speed

182 In each run, a parcel is first initiated at the surface with its initial water vapor mass fraction larger
 183 than that of the environmental air by δq_v and its initial temperature the same as the environment. For
 184 the sake of simplicity, we choose not to discuss the scenarios with initial temperature perturbations,
 185 although such scenarios also happen in reality. Then the parcel gets lifted adiabatically by a certain
 186 distance, δz . After that, the parcel is released freely in the troposphere and its dynamics is
 187 completely governed by the model and model parameter values. Here the two parameters δz
 188 and δq_v represent the dynamical lifting and boundary layer moistening effects respectively, both of
 189 which should have a significant impact on the parcel dynamics. Besides, background environmental
 190 profiles including temperature and moisture are prescribed for current and future climate, serving
 191 as a controlling factor in determining the behavior of the parcel.

192 *b. Multi-column Lagrangian parcel model*

193 1) MODEL DESCRIPTION

194 The multi-column Lagrangian parcel model is based on the previous single-column one and
195 derived for modeling the collective behaviors of multiple parcels, particularly the upscale growth
196 feature during the MCS genesis (?). The multi-column model considers 1000 single-column
197 Lagrangian parcel models that are aligned in the east-west direction. The multi-column model
198 is further augmented by incorporating the cold pool driven interaction mechanisms, including i)
199 the lifting effect by gust front spreading, ii) the lifting effect by cold pool collision, and iii) the
200 clear-sky subsidence effect produced by active convection. The readers are referred to ? for more
201 details of the model setup.

202 ? concluded that both gust front propagation speed and subsidence strength are the controlling
203 factors for determining the final state of the mesoscale clusters. In the following subsections
204 2) and 3), we develop more realistic parameterizations for both factors under the guidance of
205 cloud-resolving simulations discussed next.

206 2) CLOUD-RESOLVING SIMULATIONS FOR GUIDING PARAMETERIZATION DEVELOPMENT

207 Cloud-resolving simulations are useful tools for investigating convection organization and cold
208 pool dynamics (??). Two sets of 2-D nonrotating cloud-resolving simulations based on the System
209 for Atmospheric Modeling (SAM) (?), version 6.11.1, are used here as a guidance for parameterizing
210 gust front propagation speed and subsidence strength. In general, the SAM has been used widely to
211 simulate convection in idealized domain geometry (????). All simulations use the Smolarkiewicz's
212 MPDATA advection scheme with monotonic corrector, the 1.5-order closure (prognostic SGS
213 turbulent kinetic energy) subgrid-scale scheme, the CAM3 radiation package as well as the single-
214 moment microphysics scheme (?). The surface of the domain is set as land with intermediate soil
215 wetness (0.5) and typical roughness length (0.1) for the central U.S. Both interactive surface flux
216 and radiation (shortwave and longwave) schemes are switched on to allow convection-moisture-
217 surface flux-radiation interaction, which is crucial for convective self-aggregation phenomenon
218 (?). The domain has periodic boundary conditions in zonal direction and a 27-km vertical extent.
219 The initial background sounding is set at the state of rest and zonal winds are nudged towards zero

220 with a time scale of 1 day. A sponge layer is added in the upper one-third of the model domain to
221 damp gravity waves.

222 The first set of 25 SAM simulations is for parameterizing gust front propagation speed. To
223 generate a cold pool from an individual convection cell, we initiate a warm bubble in the middle of
224 the domain at height 0.5 km with 4-km horizontal radius and 1-km vertical radius (part of the warm
225 bubble is below the surface). The initial temperature perturbation of the warm bubble is fixed at 5
226 K, while its initial water vapor perturbation is varied every 0.5 g/kg from 7 g/kg to 9 g/kg. The
227 initial temperature and moisture sounding is representative of 14-year (2004-2017) climatological
228 mean summertime environments over the central U.S. from ERA5. To test the sensitivity of cold
229 pools to the environment, we further vary the background moisture sounding by multiplying a
230 constant (0.9, 0.95, 1.0, 1.05, 1.1). This 2-D mesoscale domain has 102.4-km zonal extent in a
231 100-m horizontal grid spacing. All simulations are run for 2 hours with a 3-sec time step. To better
232 resolve the cold pool dynamics in the lower troposphere, these simulations use a vertical grid with
233 30-m grid spacing below 1 km, 60-m grid spacing between 1 km and 2 km, 100-m grid spacing
234 between 2 km and 5 km.

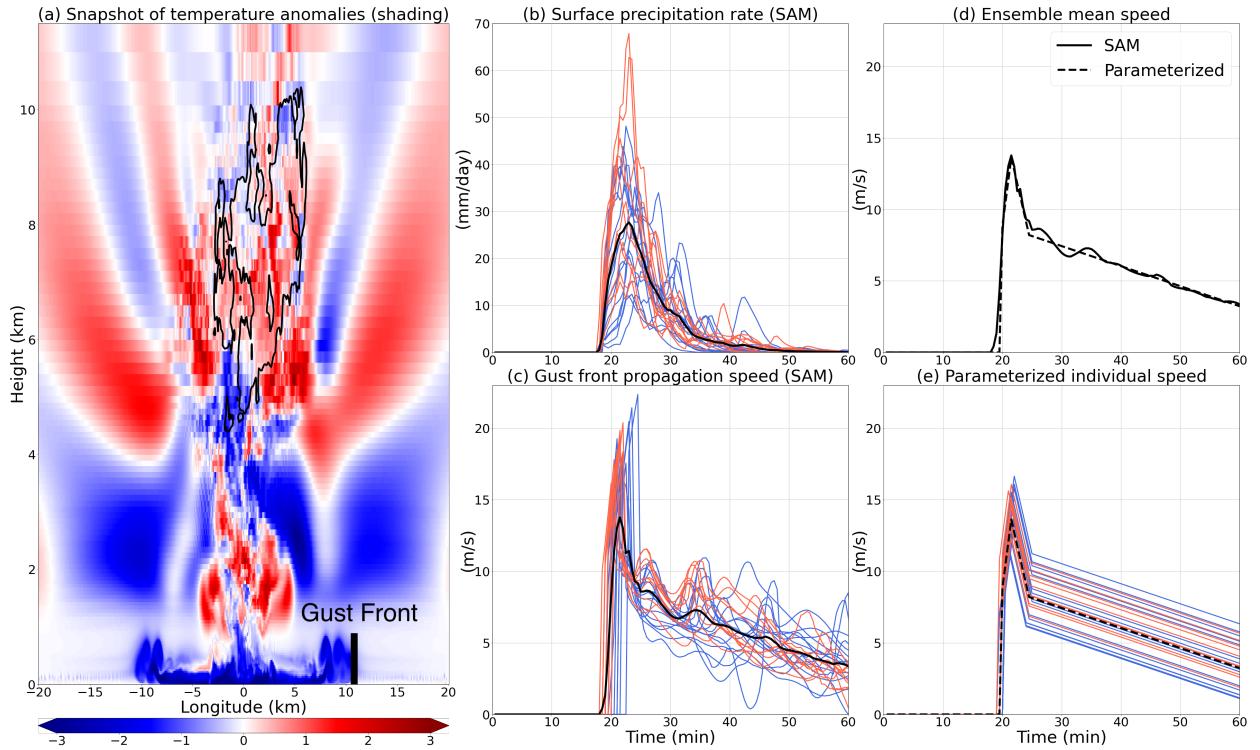
235 The second set of 11 SAM simulations is for parameterizing subsidence strength. To promote
236 convective self-aggregation, we initiate some white noise in the temperature field near the surface
237 to trigger convection and run the model till it reaches the radiative-convective equilibrium (RCE)
238 state. The initial temperature and moisture sounding is representative of climatological mean
239 summertime environments over the central U.S. To test the sensitivity of self-aggregation to the
240 environment, we further vary the background moisture sounding by multiplying a constant (0.5,
241 0.6,0.7...1.5). This 2-D synoptic domain has 1024-km zonal extent in a 2-km horizontal grid
242 spacing. All simulations are run for 50 days with a 5-sec time step. It turns out that these
243 simulations typically reach the RCE state in 15 days.

244 3) PARAMETERIZATION DEVELOPMENT AND SENSITIVITY

245 It is worth noting that both gust front propagation speed s_{gf} and the averaged updraft speed w_{up}^*
246 in subsidence strength are set as constant ($s_{gf} = 5 m/s$, $w_{up}^* = 1.9 m/s$) in the multi-column model
247 of ? . Such a simplification assumption may not be realistic, considering the facts that the gust
248 front propagation speed typically slows down as the cold pool expands and the subsidence strength

249 should be related to the updraft speed over the neighboring convective regions (??). For gust
250 front propagation speed, we run a set of 25 SAM warm bubble simulations where the background
251 moisture profile and warm bubble initial moisture anomaly are perturbed about the reference value
252 in the control run ($1.0 \times Q_v$, background moisture profile and 8 g/kg initial moisture anomaly).
253 Then we diagnostically calculate the surface precipitation rate and gust front propagation speed
254 in each simulation, which is used as a guidance for developing a new parameterization. Similar
255 steps are taken to develop the parameterization of subsidence strength, based on a set of 11 SAM
256 simulations in the RCE state. Details about these two sets of SAM simulations can be found in
257 Sec. 2c.

269 Fig. 1a shows a snapshot of temperature anomalies from the background sounding in the
270 control run at 40 mins. In general, the initial warm bubble rises gradually towards the upper
271 levels since initiation and releases large amounts of rainfall after reaching saturation. Due to rain
272 evaporation, the resulting cold and dense air in the downdraft hits the surface and forms the cold
273 pool that expands in both directions. The updraft motion in the gust front indicates its lifting
274 effects, potentially triggering new convection in the neighboring regions (?). The mean surface
275 precipitation over the whole domain in panel (b) starts to increase at 18 mins, reaches its peak at 25
276 mins, and then decay gradually afterwards, following the dissipation of the warm bubble initiated
277 convection. The gust front propagation speed in individual simulations shows large differences,
278 which is generally faster in wet cases (red curves) and slower in dry cases (blue curves). Such
279 a positive relationship between background moisture profile and gust front propagation speed is
280 quite apparent in the first 25 mins (Fig. 1c). After 25 mins, the surface precipitation exhibits an
281 oscillatory pattern with a few peaks, due to the fluctuation of gust front speeds as they move to the
282 neighboring area while affected by turbulent mixing. As for the mean gust front propagation speed
283 in panel (c), it is initialized at time 20 mins, a few minutes later than the surface precipitation in
284 panel (b), while it reaches its peak at time 23 mins, a few minutes earlier than surface precipitation.
285 After 23 mins, the propagation speed decays dramatically in the first 5 mins and then slows down
286 gradually afterwards. When compared with the mean speed, the gust front propagation speed in
287 individual simulations exhibits much larger value in its peak and more oscillatory patterns in the
288 decaying stage. Such oscillatory gust front propagation speed can be induced by a few factors, such
289 as the unsteady downdraft mass flux and the complex interactions among cold pools, surface fluxes,



258 FIG. 1. Parameterization of gust front propagation speed based on 25 SAM warm bubble simulations. Panel
 259 (a) shows a snapshot of the SAM simulation with $1.0 \times Q_v$ background moisture profile and 8 g/kg warm bubble
 260 moisture perturbation at 40 mins. The color shading shows temperature anomalies from background sounding
 261 and the single contour is for non-precipitating condensate (water+ice) at 0.2 g/kg . The black vertical line near
 262 the surface indicates the location of the eastward-moving gust front determined by the algorithm. Panel (b)
 263 shows the time series of surface precipitation rate from 25 SAM simulations with different moisture background
 264 profiles and warm bubble moisture perturbations (blue curves for drier cases and red curves for more moist
 265 cases). The bold black curve shows their mean. Panel (c) is similar to panel (b) but for gust front propagation
 266 speed determined by the algorithm. Panel (d) shows the time series of gust front propagation speed from (solid)
 267 SAM simulation ensemble mean, (dashed) parameterization. Panel (e) is similar to panel (d) but for individual
 268 SAM simulations.

290 and convection-induced turbulent flows. The reason is because different simulations have different
 291 initiation timing of gust front propagation. Moreover, the dry cases (blue curves) have much later
 292 initiation timing, faster peak speed and slower mean speed in the decaying stage. Dry cases take
 293 longer time to form the cold pool near the surface. Meanwhile, the first wave of downdraft cold

294 air induces more abrupt acceleration effects to the cold pool edge and fast peak propagation speed.
 295 Due to the small surface precipitation, downdraft that serves as driving force can not be maintained
 296 as the cold pool expands, resulting in slower mean gust front propagation speed.

297 As guided by the cloud-resolving simulations in panels (b) and (c), the gust front propagation
 298 speed consists of two parts, one of which is for the short-time speed boost in the first few minutes
 299 after cold pool initiation and the other is the linearly decaying trend throughout the whole duration.
 300 Thus the detailed parameterization is formulated as follows,

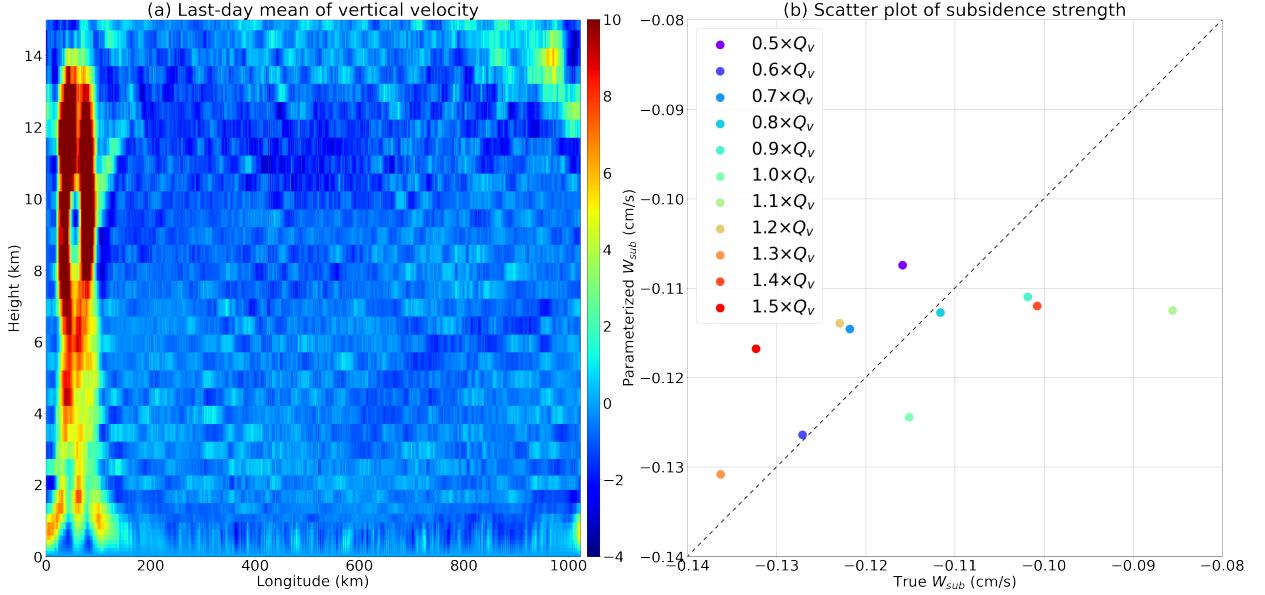
$$s = s_1 + s_2, \quad (1)$$

$$s_1 = \begin{cases} 0 & \text{if } t < T_1 \\ \frac{t-T_1}{T_2-T_1} \delta U & \text{if } T_1 \leq t < T_2 \\ \frac{T_3-t}{T_3-T_2} \delta U & \text{if } T_2 \leq t < T_3 \\ 0 & \text{if } t \geq T_3 \end{cases} \quad (2)$$

$$s_2 = \begin{cases} 0 & \text{if } t < T_1 \\ \alpha P(T_1) + U_0 - \beta(t - T_1) & \text{if } t \geq T_1 \end{cases} \quad (3)$$

301 where $P(t)$ represents the accumulated surface precipitation, $\alpha, \beta, U_0, \delta U$ are tunable parameters
 302 and coefficients whose value is determined through the fitting method. T_1, T_2, T_3 are the predefined
 303 timing for the gust front initiation, the peak of gust front propagation speed, and the end of the first
 304 decaying stage, respectively. Fig. 1d shows the close match between the parameterization and the
 305 mean propagation speed from panel (c), reflecting the skill of the parameterization in capturing
 306 the complex evolution of cold pool expansion dynamics. Furthermore, the parameterization also
 307 captures readily the spread of gust front propagation speed in all cases in panel (e), where the
 308 majority of wet cases have a faster speed above the mean and that of dry cases is slower.

316 As for the subsidence, we first diagnostically calculate its strength in the 11 SAM RCE simulations
 317 and use the results as guidance for developing the parameterization. To filter out high-frequency
 318 gravity wave signals, the vertical velocity field is first filtered by taking daily mean and spatial
 319 smoothing over a 20-km moving window. Fig. 2a shows the last-day mean of vertical velocity
 320 from the reference SAM RCE simulation ($1.0 \times Q_v$ environmental moisture profile), which features
 321 a single convective cluster to the left side of the domain and compensating subsidence prevailing in



309 FIG. 2. Parameterization of subsidence strength based on 11 SAM RCE simulations. Panel (a) shows the last-
 310 day mean vertical velocity (unit: cm/s) from the SAM simulation with $1.0 \times Q_V$ background moisture profile.
 311 The vertical velocity is further smoothed zonally by using the 1-D uniform filter over a 20-km window. Panel (b)
 312 shows the scatter plot of subsidence strength between SAM simulations and the parameterization. The overlaid
 313 dashed line is the 1:1 line. Dots in different colors correspond to the SAM simulations with different background
 314 moisture profiles as shown in the legend. For each dot, its x- and y-axis values are based on the last 20-day
 315 average of daily mean simulation output.

322 the remaining area. The updraft motion reaches the upper troposphere with its maximum near 11
 323 km . The subsidence strength in these cloud-resolving simulations is then calculated by averaging
 324 of vertical velocity at lower levels below 2 km over the suppressed convection regions, which are
 325 defined as columns with no updraft speed at height 11 km . It is worth noting that 2 km is the cloud
 326 base level in the simple model.

327 The parameterization of subsidence strength is based on the conservation law of mass so that
 328 updraft and downdraft mass flux should be balanced at all vertical levels within the model domain.
 329 Specifically, we have $N_{conv} \bar{w}_{up} + N_{sup} w_{sub} = 0$ at the level 11 km , where $\bar{w}_{up} = \frac{1}{N_{conv}} \sum_{active} w_{up}$ is
 330 the mean updraft speed averaged over N_{conv} active convection columns and subsidence is assumed
 331 to be uniform over N_{sup} suppressed columns. Furthermore, if we assume the relative strength of
 332 subsidence at the lowest 2 km and 11 km is σ , we can obtain the parameterization of subsidence

333 strength at the lowest 2 km as follows,

$$w_{para} \approx \sigma w_{sup} = -\sigma \bar{w}_{up} \frac{N_{conv}}{N_{tot} - N_{conv}}, \quad (4)$$

334 where N_{tot} and N_{conv} denote the number of total and active convection columns, respectively,
 335 satisfying the constraint $N_{conv} + N_{sup} = N_{tot}$. Fig. 2b shows the scatter plots of parameterized and
 336 simulated subsidence strength in all 11 SAM RCE simulations with most dots located near the
 337 $y = x$ line, reflecting the skill of the simple parameterization in capturing the subsidence strength in
 338 cloud-resolving simulations in the presence of complex physical processes. The coefficient value
 339 σ is chosen accordingly through the fitting method. Surprisingly, the green outlier ($1.1 \times Q_v$)
 340 corresponds to a case with dryness close to the default one ($1.0 \times Q_v$), which may be worth further
 341 investigation.

342 Before implementing the parameterization of gust front propagation speed in Eqs. 1-2 and
 343 subsidence strength in Eq. 4, we need to take extra steps to obtain surface precipitation and updraft
 344 speed that are used in the parameterizations based on the multi-column model output. First, the
 345 multi-column model currently does not include the surface precipitation, but only the instantaneous
 346 precipitation rate at the height z_p of the parcel. Here we diagnostically calculate the accumulated
 347 surface precipitation by assuming a constant speed of falling raindrops, $V_d = 10 \text{ m/s}$ (?) and fixed
 348 fraction of rain evaporation $\xi = 0.2$ (?) so that only $1 - \xi$ of precipitation amount from the parcel
 349 will reach the surface at a time lag $\frac{z_p}{V_d}$. Besides, the updraft speed over convective columns at
 350 height 11 km is not available in the model as it only resolves the updraft speed of the parcel at a
 351 specific level at each time step. Here we drop the constraint of 11 km and take the average of all
 352 parcel updraft speed over convective columns to approximate the mean updraft speed. Thus Eq. 4
 353 reads as follows,

$$w_{para} = -\sigma w_{up}^* \frac{N_{conv}}{N_{tot} - N_{conv}}. \quad (5)$$

354 where $w_{up}^* = \frac{1}{N_{conv}} \sum_i w_{up}(i)$ is the averaged updraft speed of parcels over convective columns. This
 355 approximation is based on the fact that these parcels in a large number are vertically distributed
 356 uniformly so that their mean updraft is proportional to that at a certain level. ? assumed a constant
 357 value of w_{up}^* and concluded that its magnitude determines the final state of mesoscale clusters in the
 358 multi-column model. Considering the fact that the convective column number N_{conv} is coupled to

359 the parcel dynamics, we recognize the value of w_{up}^* as a key parameter to determine the subsidence
360 strength.

361 *c. Input data for environmental profiles*

362 Two datasets are used in this study to assess the environmental profiles in current and future
363 climate. To discuss the initiation of summertime MCSs over the U.S., we focus on the 14-year
364 (2004-2017) summer-time (June-August) data over a domain slightly larger than the contiguous U.S.
365 ($20^\circ - 55^\circ\text{N}$, $130^\circ - 65^\circ\text{W}$). We choose this 14-year period to be consistent with ?. The first dataset is
366 from the fifth generation ECMWF reanalysis (ERA5) hourly data product (?) and downloaded from
367 the Climate Data Store (CDS) website (<https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-pressure-levels?tab=form>). The original $0.25^\circ \times 0.25^\circ$ reso-
368 lution data is further coarse-grained onto the $1^\circ \times 1^\circ$ grid spacing, the latter of which serves as local
369 environmental profiles for the Lagrangian parcel model. The climatological mean environmental
370 profiles are also used as initial sounding for those SAM simulations as mentioned above. It has
371 been shown that ERA5 successfully reproduces severe local storm environments with strong spa-
372 tiotemporal correlations and low biases, especially over the Great Plains (?). These hourly ERA5
373 profiles of temperature and specific humidity are used as input in both single- and multi-column
374 models to simulate the current climate. The second dataset is the monthly temperature and specific
375 humidity anomalies between future climate (2056-2099; RCP8.5) and current climate (1962-2005;
376 historical) based on 37 CMIP5 models (?). This dataset is used to assess how changes of the envi-
377 ronmental thermodynamic profiles under global warming may influence the convective initiation
378 and genesis of summertime MCSs over the U.S.

380 *d. Experiment setup details*

381 The Lagrangian parcel model describes the rising trajectory of a buoyancy-driven parcel, pro-
382 viding a useful tool in assessing the convective activity response to the input environmental profile.
383 For future climate under global warming, we use the PGW approach where the future hourly en-
384 vironmental profile is approximated by the ERA5 hourly environmental profile in current climate
385 plus a climate perturbation derived from GCM ensemble projection of mean climate change. In this
386 study, we conduct 6 sets of experiments in total based on the single- and multi-column models. The

392 TABLE 2. Setup details for all the experiments conducted in this study. From left to right, the columns are
393 for (#) experiment index, (Model) single- or multi-column Lagrangian parcel model, (Environmental profile)
394 the temperature and moisture profiles used to drive the model, (Simulation Count) total number of simulations,
395 (Figure) figures for the corresponding results, (Description) the purpose of the experiment. The “future climate
396 perturbation” is calculated as the mean state difference between future climate (2056-2099; RCP8.5) and current
397 climate (1962-2005; historical) based on 37 CMIP5 model projections. The “current climate model bias”
398 is calculated as the summer mean state difference between CMIP5 models and ERA5 data for the period
399 (1996–2005). The U.S. domain is 130°W–65°W, 20°N–55°N, and the central U.S. domain is 100°W–90°W,
400 35°N–45°N. All multi-column model simulations have four ensemble members with only differences in random
401 initiation conditions.

#	Model	Environmental profile	Simulation Count	Figure	Description
1	single-column	current profile = 2004–2017 JJA ERA5 hourly data over the U.S., Future profile = current profile + future climate perturbation	$2 \times (24 \times 92 \times 14) \times (65 \times 35) \approx 1.4 \times 10^8$	Fig. 3g-h, 5, 6, 7a	impact of global warming on convective initiation
2	single-column	current profile = 2004–2017 JJA ERA5 hourly data over the U.S. + current climate model bias, future profile = current profile + current climate model bias + future climate perturbation	$2 \times (24 \times 92 \times 14) \times (65 \times 35) \approx 1.4 \times 10^8$	Fig. 4d-f	impact of current climate mean state model bias on future projection
3	multi-column	climatological mean (2004–2017 JJA) ERA5 data averaged over central U.S., moisture profile is multiplied by 1.0, 1.1, 1.2	$2 \times 3 \times 4 = 24$	Fig. 8	effects of parameterized gust front propagation speed and subsidence strength on convection aggregation
4	multi-column	current profile = specific ERA5 hourly data averaged over central U.S., future profile = current profile + future climate perturbation	$5 \times 4 = 20$	Fig. 9	impact of global warming on MCS genesis in four different global warming scenarios
5	multi-column	current profile = specific ERA5 hourly data averaged over central U.S., future profile = current profile + future climate perturbation	$3 \times 2 \times 4 = 24$	Fig. 10	additional three cases with different environmental profiles for sensitivity experiments

387 setup details for all the experiments are summarized in Table 2. In all single-column experiments
388 (#1, #2), the initial condition $\delta q = 0 \text{ g/kg}$ and $\delta z = 2.5 \text{ km}$ are used throughout the US domain in
389 both current and future climate scenarios. In all multi-column experiments (#3, #4, #5), the initial
390 condition $\delta q = 3 \text{ g/kg}$ and $\delta z = 0\text{--}2.5 \text{ km}$ along with the central U.S. environmental profiles are
391 used in both current and future climate scenarios.

402 Experiment 1 consists of two sets of single-column model simulations (one for current climate
403 and the other for future climate) and is used to assess the impact of global warming on convective
404 initiation. For each 1deg-by-1deg grid over the U.S. domain (130°W–65°W, 20°N–55°N), the
405 single-column model is run multiple times and driven by ERA5 hourly environmental profiles

406 (full pressure-level temperature and specific humidity) during the 14-year summertime period
407 (2004–2017 JJA). That is, a total of $2 \times 24 \times 92 \times 14 = 61824$ simulations are conducted for each
408 1deg-by-1deg grid. The environmental profiles used to drive the model reads as follows,

409 Current profile = ERA5 hourly data

410 Future profile = ERA5 hourly data + future climate perturbation

411 where the future climate perturbation (full pressure-level temperature and specific humidity) is
412 the mean state JJA difference between future climate (2056-2099; RCP8.5) and current climate
413 (1962-2005; historical) based on 37 CMIP5 model projection. To reconcile the inconsistency in
414 horizontal and vertical resolutions between ERA5 and CMIP5, we first interpolate the CMIP5 data
415 onto the same grids as the ERA5 through the spline interpolation method.

416 Experiment 2 is similar to experiment 1 except that both current and future profiles are modified
417 by adding a current climate model bias,

418 Current profile = ERA5 hourly data + current climate model bias

419 Future profile = ERA5 hourly data + current climate model bias + future climate perturbation

420 where the current climate model bias is calculated as the mean state difference between CMIP5
421 models and ERA5 data for the period (1996–2005 JJA). Here we choose this 10-year period that
422 does not overlap with the 14-year period (2004–2017 JJA) to match the CMIP5 historical runs that
423 only cover the period 1962–2005.

424 Experiment 3 is used to assess the effects of parameterized gust front propagation speed and sub-
425 sidence strength based on the multi-column model, driven by the climatological mean (2004–2017
426 JJA) ERA5 environmental profiles averaged over central U.S. (100°W – 90°W , 35°N – 45°N). We
427 considered 3 cases with their moisture environmental profiles multiplied by 1.0, 1.1, 1.2, respec-
428 tively. For each case, 4 ensemble simulations with only difference in initial random conditions are
429 conducted for the sake of robustness.

430 Experiment 4 consists of 5 sets of multi-column model simulations for 5 different scenarios,
431 including the current climate scenario and 4 different global warming scenarios for future climate.
432 The current environmental profile used to drive the model is based on a specific ERA5 hourly data
433 averaged over the central U.S. at 5 UTC, July 17, 2013. The future environmental profile is equal
434 to the current environmental profile plus the future climate perturbation as mentioned above.

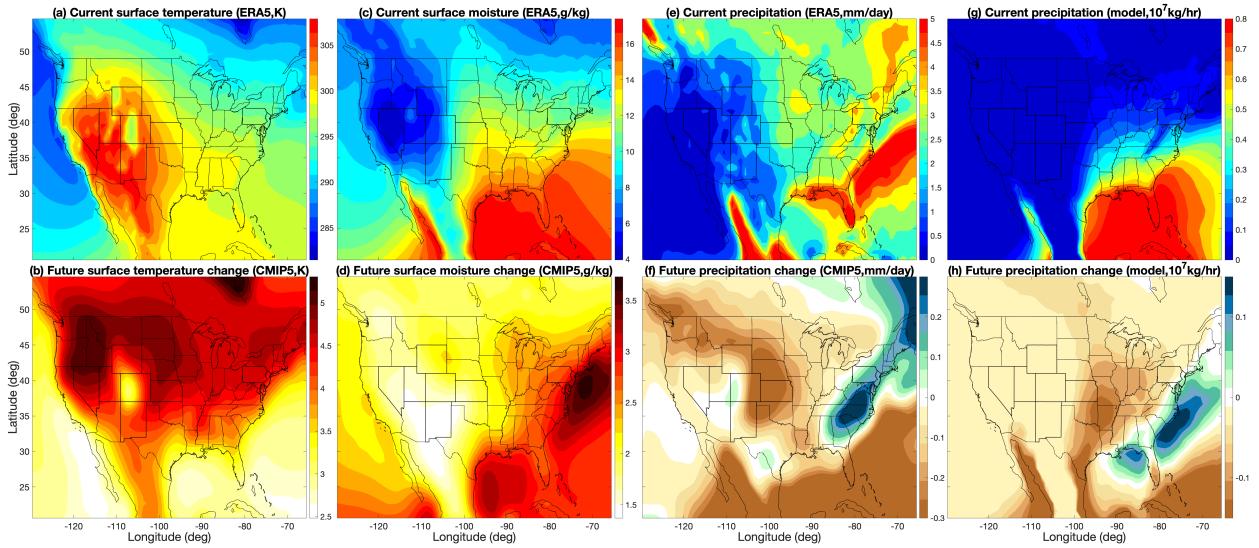
435 Experiment 5 is similar to experiment 4, except that their environmental profiles are selected for
436 different times of the day and different dates over the central U.S., including i) 18 UTC, June 2,
437 2014, ii) 4 UTC, August 28, 2017, iii) 19 UTC, June 30, 2016. This experiment is used to assess
438 the sensitivity of model results to environmental profiles.

439 **3. Impact of global warming on convective initiation: results from the single-column model**

440 Convective initiation is the early stage of MCS initiation where scattered convection are trig-
441 gered due to either dynamical lifting or thermodynamic favorable conditions. ? showed that
442 the single-column Lagrangian parcel model captures readily the basic features of a rising parcel,
443 demonstrating the collective effects of boundary layer moistening and dynamical lifting in trig-
444 gering convective initiation and reproducing successfully its early afternoon peak over the central
445 U.S. in the observation. In general, the single-column model describes the response of convective
446 initiation to the background thermodynamic fields in temperature and moisture. Here we use the
447 single-column model to study the convective initiation stage in general, no matter whether it leads
448 to MCS genesis eventually. The goal of this section is to investigate the skills of this model in
449 capturing climatological mean precipitation and convective population in current climate and its
450 ability to reproduce the future mean precipitation and convective population projected by the CMIP
451 models. As motivated by the common climate model biases in simulating precipitation and surface
452 temperature in the central U.S., we also discuss the impact of current climate mean state model
453 bias in predicting future climatological mean precipitation. The initial moisture of the parcel is set
454 to be the same as the environmental air ($\delta q_v = 0$) and dynamical lifting distance δz is 2.5 km for all
455 simulations in this section. The choice of these initial conditions are used to mimic the favorable
456 conditions for convection due to dynamical lifting effects δz from eastward-moving large-scale
457 environments and subsynoptic perturbations (?????). The moisture anomaly δq_v is set to zero to
458 mimic the local thermodynamic favorable condition for convection for the sake of simplicity.

459 *a. Changes in climatological mean precipitation*

469 Fig. 3 shows the climatological mean spatial distributions of summertime surface temperature,
470 moisture, and precipitation in the current climate and the future change under global warming.
471 In the observed current climate, surface temperature exhibits a clear east-west contrast with its



460 FIG. 3. Comparisons of climatological mean summertime precipitation and its future change between the
 461 single-column model results and 37 CMIP5 model projections. Panel (a) shows climatological mean surface
 462 temperature based on 2004-2017 June-July-August (JJA) ERA5 data, while panel (b) show future mean surface
 463 temperature change between future climate (2056-2099; RCP8.5) and current climate (1962-2005; historical)
 464 based on 37 CMIP5 models. Panels (c,d) and (e,f) are similar to panels (a,b) but for surface moisture and
 465 precipitation, respectively. Panel (g,h) are similar to panel (e,f) but based on the single-column model, where
 466 the value (color shading) indicates the accumulated precipitation amount within the 1-h simulation duration for
 467 each $1deg \times 1deg$ grid. The pattern correlation between panels (e) and (g) is 0.49, and that between panels (f)
 468 and (h) is 0.57.

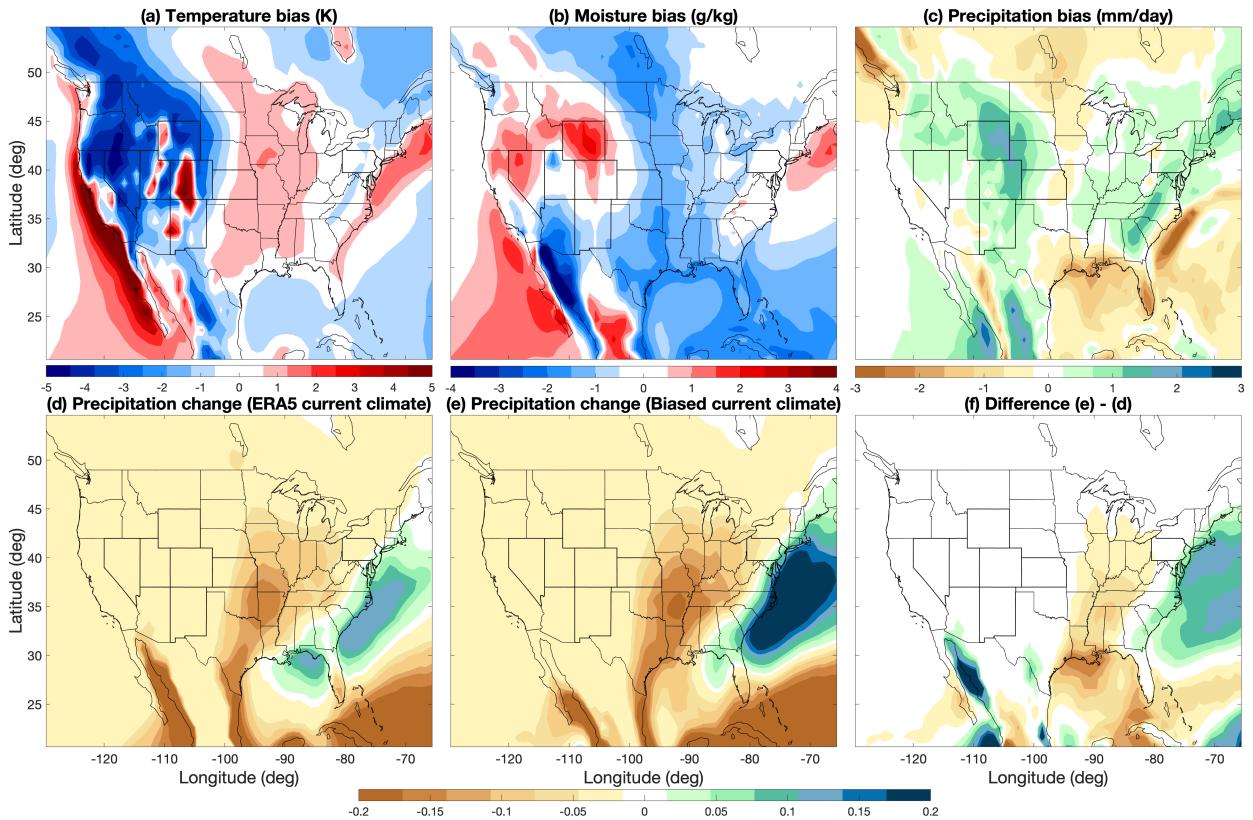
472 maximum over the western and central U.S. in panel (a), while surface moisture peaks to the south
 473 over the southeastern U.S., the Gulf of Mexico, the Atlantic ocean, and the Pacific coast of Mexico
 474 in panel (b). The future surface temperature increase projected by CMIP5 models is manifested
 475 by large warming anomalies over North America with its maximum over the Pacific Northwest in
 476 panel (b). The future surface moisture increase is much weaker over land than ocean, consistent
 477 with the land aridity effect (?), as shown in panel (d).

478 By incorporating the hourly 1deg-by-1deg ERA5 temperature and moisture as background en-
 479 vironmental profiles, the single-column model estimates theoretically the climatological mean
 480 precipitation amount in panel (g) with its maximum over the southeastern U.S., the Atlantic ocean,
 481 and Pacific coast of Mexico. This spatial pattern matches well with that based on the ERA re-

482 analysis data in panel (e). The underestimated precipitation over the northeastern U.S. and Canada
483 is presumably due to the absence of ice-phase processes in the simple model. By adding the
484 future temperature and specific humidity mean changes from the CMIP5 projections to the ERA5
485 hourly environmental profiles, the single-column model projects precipitation decreases over the
486 central U.S., Pacific coast of Mexico, and the Atlantic ocean, and precipitation increases along
487 the eastern coast of the U.S. in panel (h), resembling those projected by the CMIP5 models in
488 panel (f). Overall, these similar spatial patterns between the single-column model results, ERA5
489 reanalysis data, and CMIP5 model projections demonstrate the skills of the simple theoretical
490 model in capturing the climatological mean precipitation pattern and highlight the crucial role
491 of environmental thermodynamic fields in controlling the mean state of convection and its future
492 changes. That said, there also exist several differences among these fields, including the large moist
493 bias in the subtropical ocean and a generally dry bias in most of the inland region in panel (g),
494 and the eastward displacement of the moistening response over the ocean areas in panel (h). This
495 is partly due to the missing dynamical fields in the simple-column model. For example, vertical
496 wind shear is an important controlling factor modulating the convective storms at midlatitudes.
497 Although the spatial pattern of current precipitation and future precipitation change provide us
498 many interesting insights, it is inappropriate to directly compare the absolute value of precipitation
499 amount in the same unit between panels (e,f) and (g,h). Notably, a larger/smaller initial parcel
500 volume (currently $10^9 m^3$) would result in larger/smaller precipitation amount, but not change the
501 spatial patterns in panels (g,h). Also, the rain falling out from a plume may not be equal to the
502 rainfall amount in a GCM grid box. To distinguish these differences, we leave the precipitation
503 unit (mm/day) in panels (e,f) to be different from that (kg/hr) in panels (g,h).

504 *b. Current climate mean state model bias*

510 The state-of-the-art climate models exhibit significant biases in simulating the mean state of the
511 atmosphere as shown in Fig. 4a-c. Compared with ERA5 reanalysis data, the CMIP5 models have
512 significant warm biases over the central U.S. and along the Pacific coast of North America, and
513 cold biases over the Pacific northwest in panel (a). Besides, the surface moisture in the CMIP5
514 simulations in panel (b) is underestimated in the southern U.S., the Atlantic ocean and the Sierra
515 Madre Occidental, but overestimated in the Pacific northwest, the Pacific coast and central Mexico.



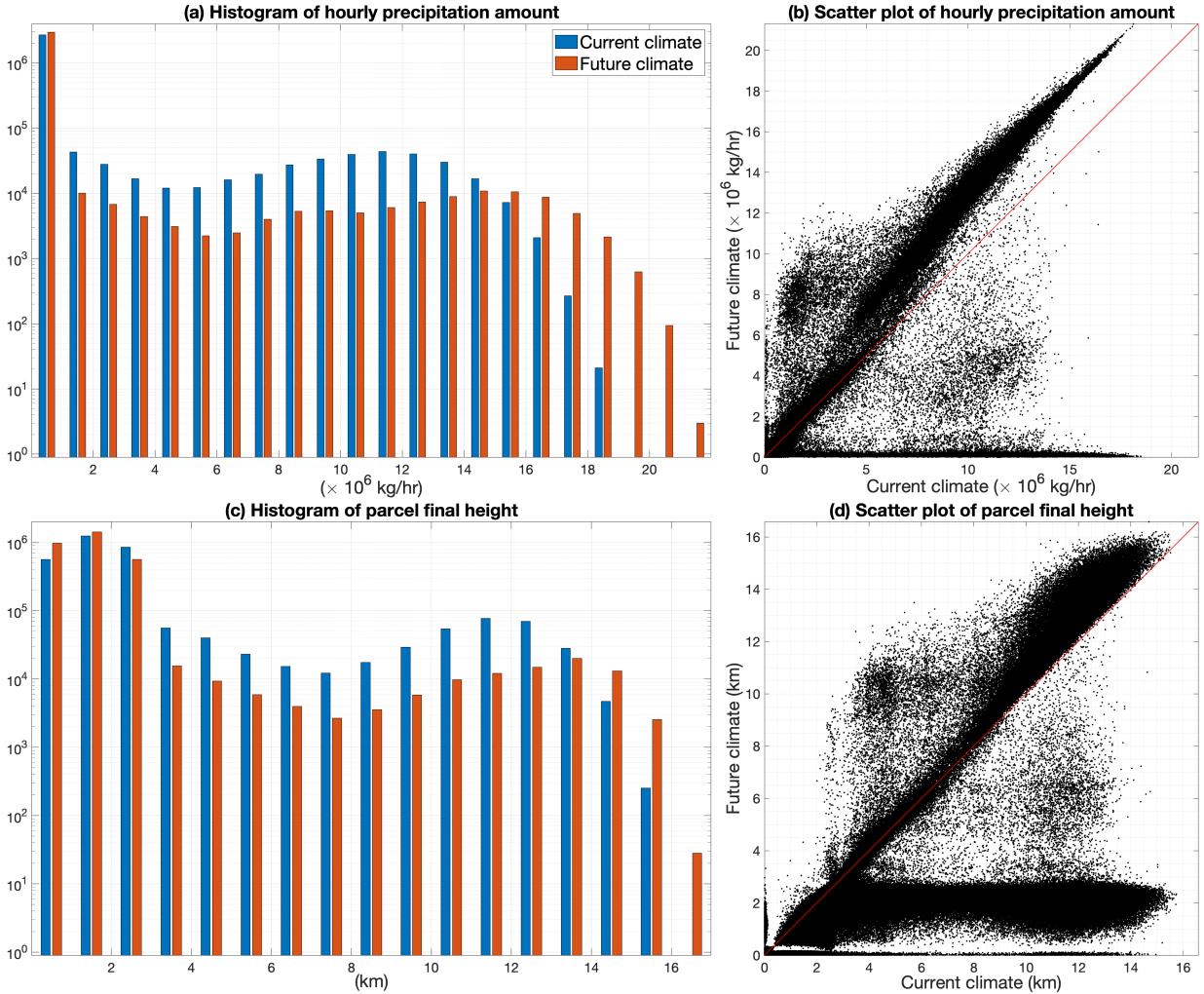
505 FIG. 4. Biases of climatological mean fields of (a) surface temperature, (b) surface moisture, and (c) precipita-
 506 tion in CMIP5 model relative to the ERA5 data for the 10-year period (1996–2005). Panel (d) shows the future
 507 mean precipitation change (unit: 10^7 kg/hr) predicted by the single-column model (the same as Fig. 6h). Panel
 508 (e) is similar to panel (d) except that each hourly current environmental profile is modified by adding the mean
 509 state biases in temperature and moisture (panel a,b). Panel (e) shows the difference between panels (e) and (d).
 516 Furthermore, slight dry biases are found over the central U.S., while wet biases are seen in the Pacific
 517 northwest and western Mexico. It is important to determine how these biases in current climate
 518 mean state could influence the model projection for future precipitation change. Fig. 4d-f shows
 519 such an assessment of climate model projection bias based on the single-column model. Compared
 520 with panel (d) which is identical to panel (f) of Fig. 3, the future precipitation change based on the
 521 biased current climate environmental profiles in panel (e) overestimated the precipitation decrease
 522 over the central U.S. and precipitation increase to the east. Their difference map in panel (f) reaches
 523 its magnitude as large as 25% of that in panel (d), reflecting the impact of the large warm and
 524 dry biases over the central U.S. in amplifying the projected changes in precipitation. Such a result

525 also indicates the nonlinear nature of thermodynamic environmental profiles as a controlling factor
526 in determining convection frequency and intensity. Besides the surface field biases, it is worth
527 mentioning that CMIP5 biases at higher levels of the free troposphere may also affect results by
528 modulating atmospheric instability (?).

529 *c. Changes in convective population*

530 Based on the simple model output driven by different environmental profiles, we proceed to study
531 the changes in convective population between current climate (ERA5 data) and future climate (PGW
532 approach). Below convective population is categorized in terms of total precipitation amount or
533 final height of the parcel.

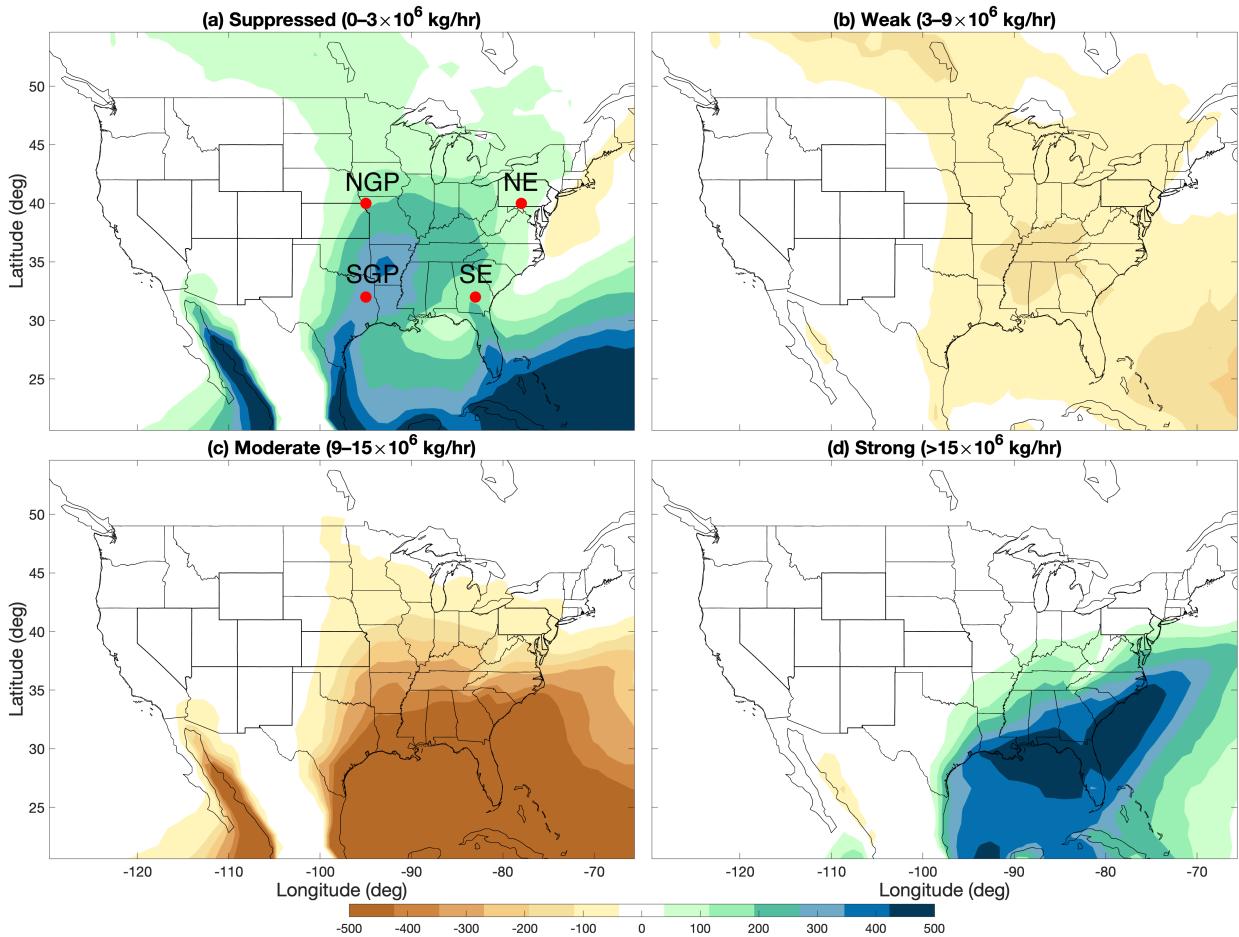
534 Fig. 5 shows the change of convective population over the central U.S. under global warming,
535 as predicted by the single-column model. Here we define different precipitation events in terms
536 of their precipitation amounts, including suppressed convective events ($0-3 \times 10^6 \text{ kg/hr}$), weak
537 precipitation events ($3-9 \times 10^6 \text{ kg/hr}$), moderate precipitation events ($9-15 \times 10^6 \text{ kg/hr}$), and
538 strong precipitation events $> 15 \times 10^6 \text{ kg/hr}$. It is worth noting that the suppressed convective
539 events typically correspond to a scenario where the parcel does not rise to upper levels and only
540 releases negligible precipitation amount. As shown by panel (a), the convective population based
541 on precipitation amount features a bimodal distribution with its first peak at zero and the second
542 peak near $12 \times 10^6 \text{ kg/hr}$. For larger precipitation events, their frequency drops dramatically as the
543 precipitation amount increases beyond $12 \times 10^6 \text{ kg/hr}$. In contrast, the convective population in
544 future climate has reduced frequency for events with precipitation amount less than $14 \times 10^6 \text{ kg/hr}$
545 but increased frequency for those events larger than that. This result is consistent with convection-
546 permitting regional climate simulations of ? that weak to moderate convection will decrease
547 and strong convection will increase in frequency in the future climate. The scatter plot in panel
548 (b) indicates that in future climate, the majority of moderate and strong convection is further
549 enhanced by $2 \times 10^6 \text{ kg/hr}$, and a fraction of weak convection is enhanced more than $2 \times 10^6 \text{ kg/hr}$.
550 Meanwhile, a fraction of weak to moderate to strong convection is reduced or even suppressed,
551 presumably due to increased convective inhibition. A parallel analysis based on the parcel final
552 height in panels (c) shows similar results with reduced weak and moderate convection and increased
553 strong convection. The scatter plot in panel (d) confirms that the majority of deep convection with
554
555
556
557



534 FIG. 5. Convective population shifting over the central U.S. ($35^\circ - 45^\circ\text{N}$, $100^\circ - 90^\circ\text{W}$) due to global warming,
 535 predicted by the single-column model. Panels (a-b) are the histogram and scatter plot for hourly precipitation
 536 amount, while panels (c-d) are for final height of the parcel. The overlaid red lines in panels (b,d) indicate the
 537 scenario when precipitation/final height in current and future climate are the same.

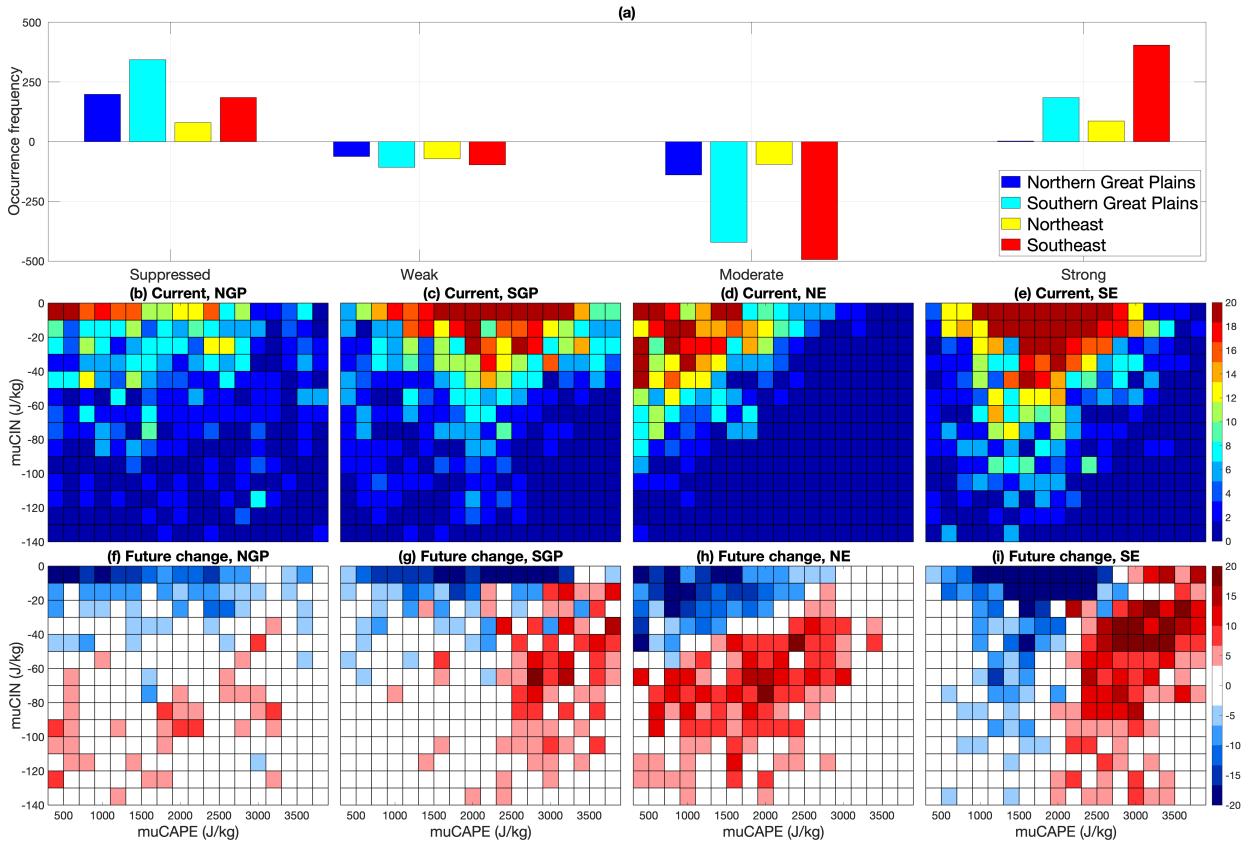
558 final height more than 10 km is enhanced the most, while convection suppression occurs throughout
 559 all convection categories.

565 To further examine the change of convection population, Fig. 6 shows the spatial distributions
 566 of frequency changes in all six convection categories in terms of precipitation amount. Overall,
 567 the single-column model predicts reduced weak and moderate convection and increased strong
 568 convection over the North America, consistent with the high-resolution convection-permitting



560 FIG. 6. Difference maps of occurrence of JJA precipitation events between future and current climate predicted
 561 by the single-column model. These panels correspond to six different precipitation amount categories, including
 562 (a) suppressed convective events ($0-3 \times 10^6 \text{ kg/hr}$), (b) weak precipitation events ($3-9 \times 10^6 \text{ kg/hr}$), (c) moderate
 563 precipitation events ($9-15 \times 10^6 \text{ kg/hr}$), (d) strong precipitation events ($> 15 \times 10^6 \text{ kg/hr}$). The yellow circles
 564 in panel (a) indicate the four representative grid points we selected for detailed analysis in Fig. 7.

569 regional climate simulations of ?. In details, the suppressed convective ? category in panel (a) has
 570 increased frequency over the eastern and central U.S., the Atlantic ocean, and the Pacific coast
 571 of Mexico, reflecting the reduced total convection frequency there under the global warming.
 572 The weak convection categories in panels (b,c) have reduced frequency over the eastern U.S.,
 573 the Atlantic, and the southern Canada. The moderate convection categories in panels (d,e) have
 574 reduced frequency mostly over the southeastern U.S. and the Atlantic ocean, while strong convection
 575 category in panel (f) exhibits increased frequency over the similar regions. These results reflect



578 FIG. 7. Panel (a) shows the histogram for 14-year mean JJA occurrence changes of different precipitation
 579 events over four 1deg-by-1deg regions (indicated by the yellow circles in Fig. 6a), predicted by the single-
 580 column model. Panel (b) shows the 2D histogram for most unstable CAPE (muCAPE) and most unstable CIN
 581 (muCIN) associated with the 2004 JJA hourly environmental profiles over the Northern Great Plains in current
 582 climate, while panel (f) shows its future change under the warming. Panels (c,g), (d,h),(e,i) are similar to panels
 583 (b,f) but for Southern Great Plains, Northeast, Southeast regions.

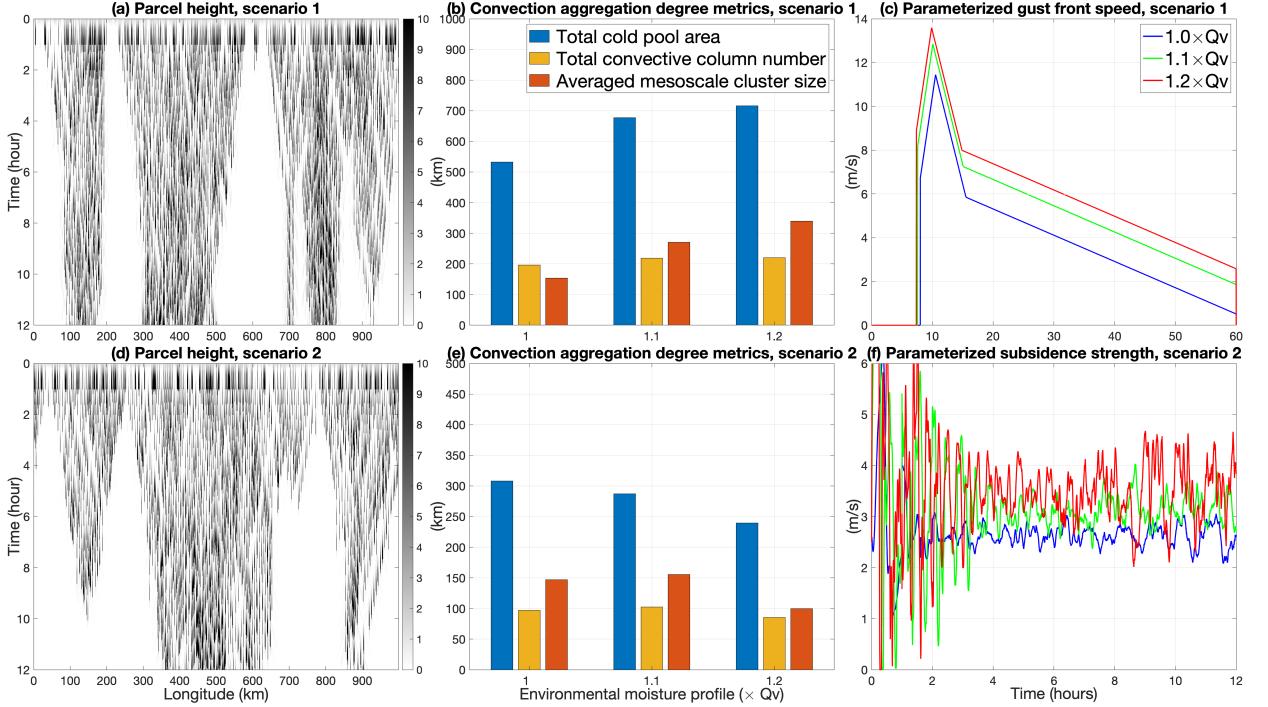
576 the usefulness of the single-column model to infer the future change of convection population and
 577 highlights the different response of convective population to the global warming.

584 To understand the essential reason for the spatial distribution of mean precipitation changes in
 585 Fig. 6, we select four 1deg-by-1deg grid points over the Northern Great Plains (NGP), Southern
 586 Great Plains (SGP), Northeast (NE), and Southeast (SE), respectively, for detailed analysis. Fig. 7a
 587 shows the histogram for JJA mean occurrence changes of different precipitation events over these
 588 four regions, which is generally characterized by decreased weak and moderate precipitation events,

589 increased strong precipitation events, and increased suppressed convective events. The increased
590 occurrence frequency of suppressed convective events reflects the decreased total occurrence
591 frequency of convective events. In particular, the NGP region is featured by slightly less occurrence
592 of weak and moderate precipitation events but almost no changes in strong precipitation events,
593 leading to a mean precipitation decrease as seen in Fig. 3h. In contrast, the SE region is featured
594 by much less occurrence of weak and moderate precipitation events and more occurrence of strong
595 precipitation events, leading to a mean precipitation increase. A further analysis based on these
596 hourly environmental profiles in panels (b-i) indicates increased most unstable CAPE (muCAPE)
597 over the SGP, NE, and SE regions and increased most unstable CIN (muCIN) over all four regions,
598 promoting strong convection and suppressing weak and moderate convection respectively. In
599 particular, the NGP region is dominated by small muCAPE and large muCIN events, while the SE
600 region has more occurrence of large muCAPE events. Due to expensive computation, here we only
601 calculate the muCAPE and CIN in the year 2004 as an example.

602 **4. Impact of global warming on MCS genesis: results from the multi-column model**

603 ? further developed a multi-column model by including an array of single-column models
604 aligned in the east-west direction and incorporating idealized cold pool interaction mechanisms.
605 The multi-column model captures readily the cold pool-induced upscale growth feature in MCS
606 genesis. The goal of this section is to investigate the impact of global warming on the final
607 state of mesoscale clusters in the multi-column model. To achieve that, we first develop the
608 parameterization of gust front propagation speed and subsidence strength under the guidance of
609 cloud-resolving simulations in Section 2b. Then we demonstrate how these two factors control the
610 upscale growth of mesoscale clusters in the presence of different background moisture profiles.
611 Finally, we use this improved version of the multi-column model to investigate the change of MCS
612 genesis under global warming in various possible scenarios. The sensitivity of model results to
613 environmental profiles is also discussed. The initial moisture anomaly δq_v of all parcels is set as
614 $3g/kg$, and their dynamical lifting distance δz is uniformly distributed in equal probability between
615 0 and 2.5 km for all simulations in this section. In contrast to the initial condition $\delta q_v = 0$ for
616 convective initiation in the last section, we consider a positive the moisture anomaly δq_v here to
617 mimic the thermodynamic favorable condition for MCS genesis, resulting from the accumulation



621 FIG. 8. Multi-column model simulations in two different scenarios, (a-c) with parameterization of gust front
 622 propagation speed and fixed subsidence strength, (d-f) with fixed gust front propagation speed and parameteriza-
 623 tion of subsidence strength. In either of these two scenarios, three cases with the same environmental temperature
 624 profile but different environmental moisture profiles (multiplied by 1.0, 1.1, 1.2, respectively) are considered.
 625 Panels (a,d) show the Hovmöller diagrams for parcel height (unit: km) in specific ensemble simulations with
 626 $1.0 \times Q_v$ environmental moisture profile. Panels (b,e) show the bar diagrams for three convection aggregation
 627 degree metrics in the three cases with different moisture profile. Panel (c) shows the parameterized gust front
 628 propagation speed in the first scenario, while panel (f) shows the time series of parameterized updraft speed for
 629 subsidence strength parameterization in the second scenario.

618 of moisture around the edges of cold pools (?). The distribution of δz introduces randomness in
 619 the initial condition, mimicking the scattered convection before the MCS genesis.

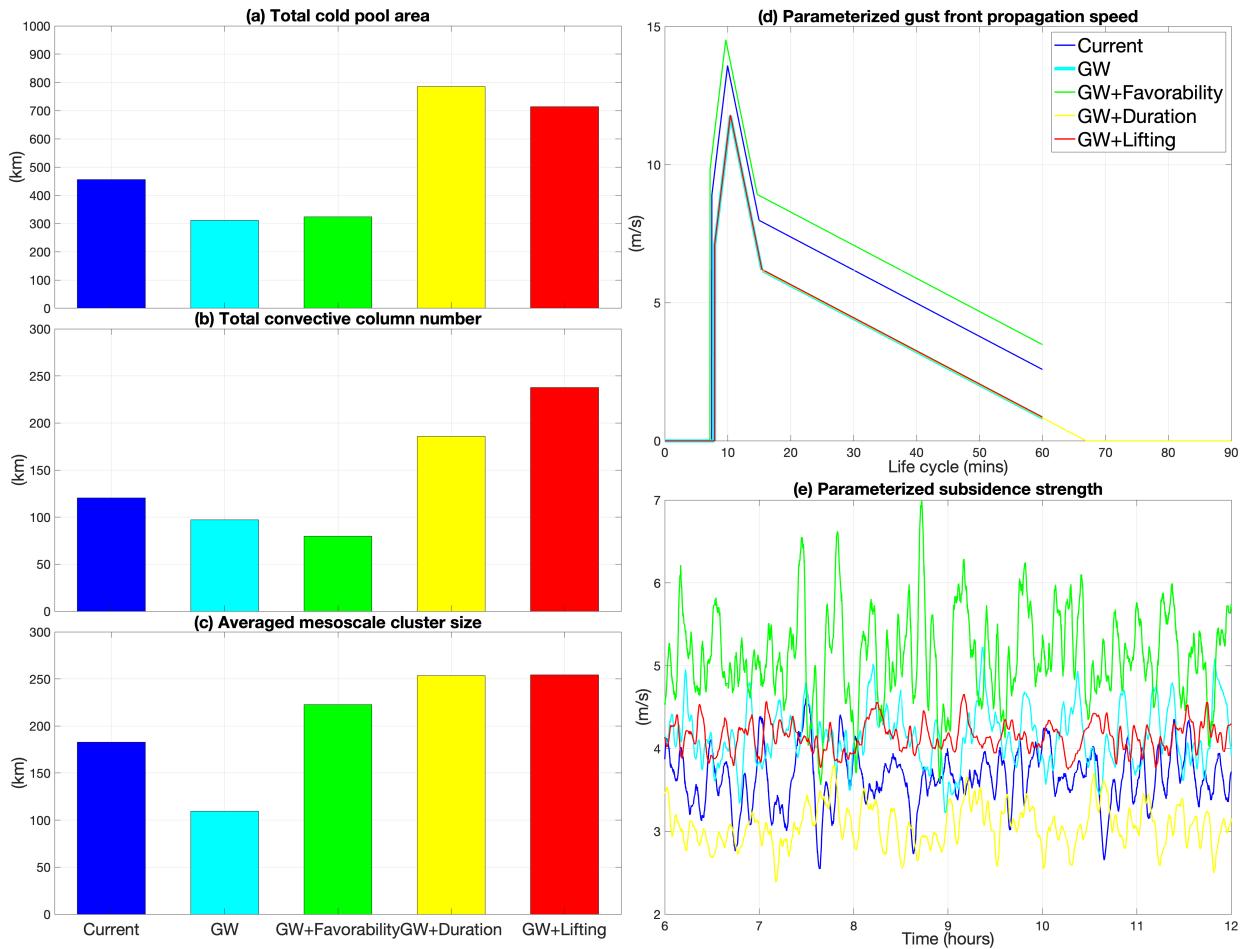
620 *a. Dependence of MCS genesis on gust front propagation speed and subsidence strength*

630 Fig. 8a shows the Hovmöller diagram for parcel height from the multi-column model with the
 631 parameterized gust front propagation speed and fixed subsidence strength. Overall, all parcels
 632 are initiated from a random distribution and gradually aggregate into mesoscale clusters after a

633 few hours, resembling the upscale growth features during MCS genesis (?). By increasing the
634 background moisture profile from $1.0 \times Q_v$ to $1.2 \times Q_v$, both total cold pool area and averaged
635 mesoscale cluster size increase in panel (b), while total convective column number stays the same.
636 A further investigation in panel (c) shows that this is due to the increasing gust front propagation
637 speed in the parameterization. In general, the wetter background profile provides more favorable
638 conditions for convection, promoting larger accumulated surface rainfall. The resulting faster
639 propagation speed in Eqs. 1-2 means the gust front propagates further away from the convection,
640 expanding the cold pool area as well as mesoscale cluster size. On the other hand, the faster
641 propagating gust front spends less time and introduces less lifting effects at each column as it
642 passes by, offsetting the effects of cold pool expansion. Similar upscale growth feature is also
643 seen in Fig. 8d with the fixed gust front propagation speed and parameterized subsidence strength.
644 By increasing the background moisture profile from $1.0 \times Q_v$ to $1.2 \times Q_v$, all three aggregation
645 degree metrics decrease in panel (e). The slight increase of total convective column number and
646 averaged mesoscale cluster size in the $1.1 \times Q_v$ case is presumably related to the uncertainty from
647 random initial conditions in the 4-ensemble member simulations. A further investigation in panel
648 (f) shows that this is mainly due to the stronger subsidence strength in the wetter cases so that more
649 convection is suppressed. This result highlights the crucial role of gust front propagation speed and
650 subsidence strength in determining the final state of mesoscale cluster in the multi-column model.

651 *b. Impact of global warming on MCS genesis in various scenarios*

652 To assess the impact of global warming, we compare the multi-column model output by switching
653 the environmental profiles from current to future climate, turning on both the gust front propagation
654 speed and subsidence strength parameterizations. Besides current climate (Current) and future
655 climate (GW), we also consider another three scenarios in the future climate, including one with
656 more favorable conditions for convection (GW+Favorability), one with long convection duration
657 (GW+Duration), and one with stronger cold pool lifting effect (GW+Lifting). To implement them in
658 the multi-column model, we increase the initial moisture anomaly from 3 g/kg to 9 g/kg in scenario
659 GW+Favorability, extend the parcel lifetime from 1 hr to 1.5 hr in scenario (GW+Duration), and
660 double the lifting distance by both cold pool collision and spreading in scenario (GW+Lifting).



661 FIG. 9. Multi-column model simulations with both parameterizations of gust front propagation speed and
 662 subsidence strength in five different scenarios, including one for current climate and the other four for future
 663 climate under global warming. Panels (a-c) shows the bar diagrams for three convection aggregation metrics
 664 respectively, including (a) total cold pool area, (b) total convective column number, (c) averaged mesoscale
 665 cluster size. Panel (d) shows the parameterized gust front propagation speed in these five scenarios, while panel
 666 (e) shows the time series of parameterized updraft speed for subsidence strength parameterization.

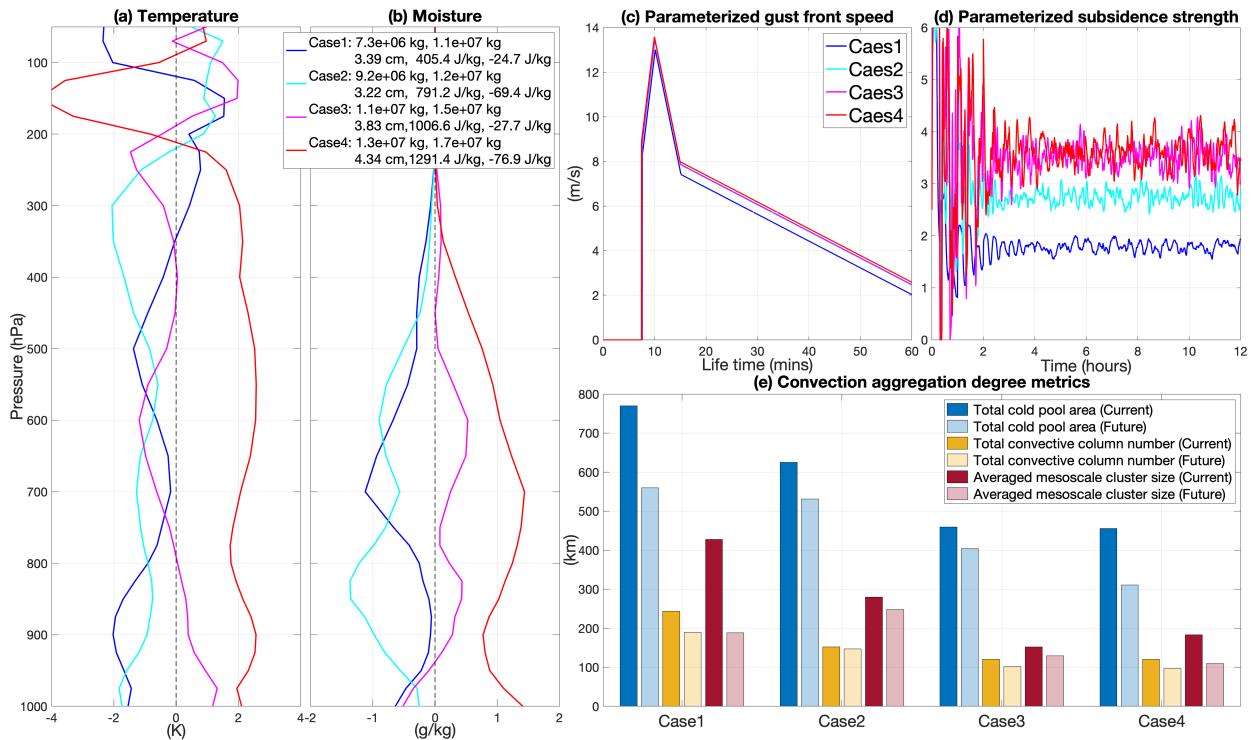
667 Fig. 9a-c summarizes the aggregation degree of convection in these 5 scenarios, and Fig. 9d-e
 668 show the corresponding gust front propagation speed and subsidence strength. When compared
 669 with the scenario Current (blue), all three aggregation degrees including total cold pool area, total
 670 convective column number, and averaged mesoscale cluster size decrease in the scenario GW
 671 (cyan). It turns out that the parameterized gust front speed is reduced and subsidence strength is
 672 enhanced in the future climate, both of which provide unfavorable conditions for MCS genesis (?).

673 This change is attributed to the increasing CAPE and CIN under global warming (?). Due to land
674 aridity and reduced relative humidity under global warming, the increasing CIN leads to reduced
675 initial buoyancy for the parcel so that the initial accumulated surface precipitation is smaller. Also,
676 the gust front propagation speed largely depends on accumulated surface precipitation (see Eqs.
677 1-2), resulting in slower speed in future climate. Note that the total precipitation amounts in both
678 the Current and GW scenarios are shown in Fig. 10b. On the other hand, the subsidence strength
679 is determined by the averaged updraft speed (see Eq. 5). Due to the larger CAPE in future climate,
680 the updraft speed of parcel typically becomes faster in the later stage of its life cycle.

681 Here we also investigate three other possible scenarios under global warming. While on average
682 parcels over the land become drier due to the increased land aridity (reduced relative humidity)
683 under the global warming, regional changes in circulation may still provide favorable conditions
684 for convection. For example, ? found that the frequency of Great Plains low-level jet related
685 environments favorable for MCS initiation will increase by 65% under a high emission scenario,
686 which could increase dynamical lifting by wind convergence in the exit region of the low-level
687 jet. The scenario GW+Favorability (green) mimics more favorable conditions for convection,
688 which may occur in environments strongly influenced by atmospheric moisture transported from
689 the ocean or under favorable soil moisture conditions and enhanced evapotranspiration from the
690 surface (?????). Fig. 9a-c indicates that both total cold pool area and convective column number
691 decrease while averaged mesoscale cluster size increases, resulting from the competing effects of
692 the faster gust front propagation speed in panel (d) and stronger subsidence strength in panel (e). The
693 scenario GW+Duration (yellow) mimics longer life time of individual convection, presumably due
694 to the larger fraction of stratiform clouds and slower detrainment process under global warming.
695 It turns out that all three aggregation degree metrics increase in this scenario, partly due to
696 weaker subsidence strength in panel (e). Meanwhile, the longer duration of convection allows
697 further propagation distance of gust fronts, favoring cold pool expansion and triggering more new
698 convection in the neighboring regions. Lastly, the scenario GW+Lifting (red) mimics stronger
699 mechanical lifting effects from cold pool expansion and collision, presumably due to more intense
700 precipitating events under the global warming. This stronger lifting effect leads to an increase in
701 all three aggregation degree metrics in Fig. 9a-c, although both gust front propagation speed and
702 subsidence strength are the same as the scenario GW. In this model, the GW+Lifting scenario is

703 achieved by increasing the magnitude of both lifting effects σ_{iso} and σ_{col} . In reality, the actual
704 lifting effect depends on virtual temperature gradient at the head of the gust front. For example,
705 a bigger initial moisture anomaly for the parcel would produce higher precipitation intensity,
706 creating stronger cold pool gust fronts as well as their lifting effects. When compared with the
707 scenario GW, all the other three GW scenarios shows tremendous increase in averaged mesoscale
708 cluster size, and the scenarios GW+Duration and GW+Lifting also show significant increase in
709 total cold pool area and total convective column number. Overall, these three additional global
710 warming scenarios demonstrate different outcomes compared to the scenario GW (e.g., increased
711 rather than decreased mesoscale cluster size with warming), highlighting those factors (convective
712 environments, convection duration, and cold pool lifting effects) in determining the final state of
713 mesoscale clusters in the multi-column model.

722 It is worth noting that the environmental profiles used in the previous multi-column model
723 simulations is only representative of the mean central U.S. environment at one specific hour. Fig.
724 10 shows the sensitivity of these multi-column model simulations to the environmental profiles
725 in assessing the impact of global warming, based on four different cases (Case 4 is the one used
726 in the previous simulations). They are selected from environments where future precipitation is
727 higher than current precipitation but at different intensities over the central U.S. The most unstable
728 (MU) CAPE of these four cases ranges from 405.4 J/kg to 1291.4 J/kg , representing typical
729 environmental conditions for weak and strong convection. Besides, the total precipitation amount
730 in current climate (see the first number in Fig. 10b legend) is the largest in Case 4 and the smallest
731 in Case 1, and all the four cases show a total precipitation amount increase under global warming
732 (see the second number in Fig. 10b legend). The corresponding temperature profiles in panel
733 (a) and moisture profile in panel (b) show a warmer and wetter free troposphere in Case 4 than
734 Case 1. Panels (c-d) indicate that Case 4 has slightly faster gust front propagation speed (increase
735 by 0.5 m/s) but much stronger subsidence strength (double) than Case 1. All four cases show a
736 decrease in the three aggregation degree metrics in panel (e), confirming the previous conclusion in
737 the scenario GW. Meanwhile, a direct comparison among these cases show that mesoscale cluster
738 become smaller in an environment with more favorable conditions for convection, consistent with
739 the previous result that stronger subsidence strength tends to suppress MCS genesis (see Fig. 8d-f).



714 FIG. 10. Multi-column model simulations with both parameterizations of gust front propagation speed and
 715 subsidence strength. Four cases with different ERA5 hourly environmental profiles over the central U.S. are
 716 considered, where Case 4 is the same as that in Fig. 9. Panels (a,b) show the departure of environmental
 717 temperature and moisture in each case from the mean of all four cases. The 5 numbers in panel (b) legend
 718 indicates total precipitation amount (based on simple model) in current and future climate, total precipitable
 719 water, CAPE, and CIN. Panel (c) shows the parameterized gust front propagation speed in these four cases, while
 720 panel (d) shows the time series of parameterized updraft speed for subsidence strength parameterization. Panel
 721 (e) shows the bar diagrams for three aggregation degree metrics in these four cases.

740 5. Discussion and conclusion

741 This study investigates the impact of global warming on the initiation of summertime MCSs
 742 over the U.S., based on a simple Lagrangian parcel model of ?. The key findings in this study and
 743 their linkage with other studies in the literature are summarized and organized for each key finding
 744 below.

745 *Single-column model for studying convective initiation* As driven by hourly local environmental
 746 profiles from current and future climate, the single-column model projects successfully a mean

747 precipitation decrease over the central U.S. and an increase over the southeastern U.S. in Fig.
748 3, consistent with the 37 CMIP5 model projection. In particular, the model predicts a mean
749 precipitation increase over the eastern U.S., reflecting a more favorable condition for convection
750 over that region under the global warming. Fig. 7a indicates that the mean precipitation increase
751 over the Northeast region results from increasing frequency of intense precipitation events, which
752 can be intense summertime MCSs and landing tropical cyclones in reality. A further investigation
753 based on this single-column model shows that model bias in simulating the current climate not only
754 increases the magnitude of mean precipitation response but also leads to its eastward shift (Fig. 4).
755 This result points out the major source of uncertainty for GCM future climate projection and urges
756 the need to further narrow down the GCM uncertainty in simulating the current climate. Besides
757 the projected changes in total precipitation, the model also projects a convective population shift
758 with reduced weak to moderate convection and increased strong convection in Fig. 5, suggesting
759 a potential frequency increase in natural disaster and severe weather under global warming. It
760 also shows that such a convective population shifts results from enhanced moderate and strong
761 convection as well as convection suppression in all convection categories in Fig. 6, due to increased
762 CAPE and CIN under global warming (see Fig. 7). Such suppressed summertime precipitation over
763 the central U.S. is also seen in high-resolution (4-km) convection-permitting regional WRF model
764 simulations (?), which employ the state-of-the-art microphysics scheme, planetary boundary layer,
765 radiative transfer model, land-surface model as well as realistic initial and boundary conditions. The
766 close match between the single-column model, GCMs, and cloud-resolving regional simulations
767 reflects the skill of the simple theoretical model in capturing the leading-order mean precipitation
768 response to global warming. This convective population shift is consistent with the high-resolution
769 convection-permitting regional simulations (?), reflecting the usefulness of the model in assessing
770 convective population response to climate change.

771 *Parameterization development guided by cloud-resolving simulations ?* recognized the key role
772 of gust front propagation speed and subsidence strength in controlling the final state of mesoscale
773 clusters in the multi-column model. To improve their representation in the model, we rely on two
774 sets of cloud-resolving simulations as guidance and develop the corresponding parameterizations
775 based on direct diagnostic analysis and physical law reasoning. Specifically, the parameterization
776 of gust front propagation speed in Eqs. 1-2 consists of two parts, one of which mimics the abrupt

777 speed boost as the first wave of cold air reaches the surface and the other mimics the gradual
778 slowdown as the cold pool expands. It turns out that the initial accumulated surface precipitation is
779 the key quantity to determine the overall gust front propagation speed in Fig. 1. This is consistent
780 with the physical understanding that cold pool spreading is driven by downdraft mass flux which
781 is correlated with precipitation mass flux. On the other hand, the parameterization of subsidence
782 strength in Eq. 4 is developed based on the conservation law of mass, in agreement with the
783 cloud-resolving simulations in Fig. 2. These two parameterizations are devoted to represent the
784 cold-pool interaction mechanisms, providing a prototype parameterization for incorporating cold
785 pool dynamics in the coarse-resolution GCMs. Similar efforts to track the gust fronts were also
786 reported in previous studies (??).

787 *Multi-column model for studying MCS genesis* After employing the new parameterization of
788 gust front propagation speed and subsidence strength, we use the improved multi-column model
789 to study the impact of global warming on MCS genesis. It turns out that the environmental
790 temperature and moisture perturbations due to global warming leads to a decrease in the degree
791 of aggregation, which is further attributed to the land aridity induced gust front slowdown and
792 subsidence strength enhancement in Fig. 9. The result here isolates the key response of cold pool
793 interaction mechanisms to global warming through the gust front speed and subsidence strength.
794 It is possible that other environmental factors also play a role in determining the final state of
795 mesoscale clusters in future climate. Thus we further investigate three additional global warming
796 scenarios with more initial parcel moisture anomaly, longer convection lifetime, and stronger cold
797 pool lifting effect. The result confirms the size increase of mesoscale clusters under these global
798 warming scenarios, and illustrates the underlying mechanisms based on the multi-column model.
799 These results explain the potential MCS genesis change under the global warming and point out
800 several important factors for controlling the upscale growth of MCS. The mechanistic processes
801 underpinning the MCS changes under global warming may inform analysis of convection permitting
802 climate change simulations for advancing insights on how convection aggregation may respond to
803 global warming and its role in cloud-radiation feedback. It is worth noting that the high-resolution
804 convection-permitting regional simulations with full physics projects a 20%-70% increase area in
805 heavy rainfall (?). In addition, the warm and dry bias that widely exists in current climate models

806 and convection-permitting models (????) may further increase the model uncertainty in projecting
807 the MCS changes under global warming.

808 Besides reproducing those key features in the reanalysis and CMIP5 projections, this study based
809 on the simple modeling framework also provides some additional knowledge and benefits. First,
810 the simple model itself provides an efficient tool to evaluate environmental profiles for favoring
811 convection in both current and future climate. Because of the simplicity, the simple model allows for
812 physical interpretations to better understand how changes in the environmental profiles influence
813 convection under global warming. Secondly, it emphasizes the crucial role of thermodynamic
814 environmental fields (temperature and moisture) in determining mean convective activity state,
815 highlighting the necessity for GCMs to simulate a better thermodynamic fields in future climate
816 projection. Thirdly, it quantifies the effects of dynamical lifting and boundary layer moistening in
817 triggering convection, which should be a source of uncertainty in current GCM projection. Lastly,
818 it supports the dominant role of cold pool dynamics in controlling MCS genesis and its possible
819 changes under the global warming, providing insights for developing model parameterization. This
820 modeling framework can be elaborated and extended in several ways. For example, atmospheric
821 gravity waves are important dynamics that communicate across columns producing organized
822 convection (???). It is interesting to incorporate the effects of gravity waves into the model and
823 investigate how it impacts the MCS genesis under the global warming. Besides, it is also important
824 to further extend the model to 3-D so that it accounts for more realistic convection organization
825 during the convective initiation and MCS genesis stages of the MCS life cycle. The model itself
826 should be a useful tool for investigating the impact of global warming on MCS convection in both
827 tropics and midlatitudes.

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832 *Data availability statement.* The ERA5 data can be downloaded from the Climate
833 Data Store (CDS) website ([https://cds.climate.copernicus.eu/cdsapp#!/dataset/
834 reanalysis-era5-pressure-levels?tab=overview](https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-pressure-levels?tab=overview)). The data from 37 CMIP5 models and
835 all model output in this study are available on application to the corresponding author.