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1 **Effects of model resolution and parameterizations on the simulations of clouds, precipitation,**  
2 **and their interactions with aerosols**

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26 **Abstract**

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28 This study investigates the effects of model resolution and microphysics parameterizations on the  
29 uncertainties or errors in the simulations of clouds, precipitation, and their interactions with aerosols  
30 using the Global Forecast System (GFS) model as one of the representative numerical weather  
31 prediction (NWP) models. For this investigation, we used the GFS model results and compare them  
32 with those from the cloud-system resolving model (CSRM) simulations as benchmark simulations that  
33 adopt high resolutions and full-fledged microphysical processes. These simulations were evaluated  
34 against observations and this evaluation demonstrated that the CSRM simulations can function as  
35 benchmark simulations. Substantially lower updrafts and associated cloud variables (e.g., cloud mass  
36 and condensation) were simulated by the GFS model compared to those simulated by the CSRM. This  
37 is mainly due to the coarse resolutions in the GFS model. This indicates that the parameterizations that  
38 represent sub-grid processes in the GFS model do not work well and thus need to be improved. Results  
39 here also indicate that the use of the coarse resolution in the GFS model lowers the sensitivity of  
40 updrafts and cloud variables to increasing aerosol concentrations compared to the CSRM simulations.  
41 The parameterization of the saturation process plays an important role in the sensitivity of cloud  
42 variables to aerosol concentrations while the parameterization of the sedimentation process has a  
43 substantial impact on how cloud variables are distributed vertically. The variation in cloud variables  
44 with resolution is much greater and contributes to the discrepancy between the GFS and CSRM  
45 simulations to a much greater degree than what happens with varying microphysics parameterizations.



## 46 **1. Introduction**

47

48 The treatment of clouds and precipitation and their interactions with aerosols in the NWP models is  
49 likely a major source of errors in the simulations of the water and energy cycles (Sundqvist et al., 1989;  
50 Randall et al., 2006; Seifert et al., 2012). The NWP community has recognized that the accurate  
51 representation of clouds, precipitation, and cloud-aerosol-precipitation interactions (CAPI) is important  
52 for the improvement of the NWP models and thus, some of these models have started to improve the  
53 representation by considering CAPI (Morcrette et al., 2011; Sudhakar et al., 2016).

54 CAPI may not have a substantial impact on the total precipitation amount but they do affect the  
55 temporal and spatial variabilities of precipitation (Li et al., 2011; van den Heever et al., 2011; Seifert et  
56 al., 2012; Lee and Feingold, 2013; Fan et al., 2013; Lee et al., 2014), whose importance increases as the  
57 temporal/spatial scales of forecast decrease. The distribution of extreme precipitation events such as  
58 droughts and floods, closely linked to the spatiotemporal variability, has important social and economic  
59 implications.

60 In recent years, resolutions in the NWP models have increased to the point that the traditional  
61 cumulus parameterization schemes may no longer work properly. Motivated by this, scale-aware  
62 cumulus parameterization schemes (e.g., Bogenschutz and Krueger, 2013; Thayer-Calder et al., 2015;  
63 Griffin and Larson, 2016) are being implemented into these models of different resolutions for better  
64 representation of clouds and precipitation.



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65 The uncertainties or errors in the simulations of clouds, precipitation, and CAPI in the NWP  
66 models may be incurred both from microphysics parameterizations and from model resolutions. The  
67 implementation of the cloud microphysics such as the two-moment (e.g. Morrison and Gettelman, 2008;  
68 Morrison et al., 2009) and scale-aware schemes are intended to reduce these uncertainties. It is  
69 important to first understand and quantify the uncertainties associated with the two-moment scheme and  
70 how model resolutions create the uncertainties, as well as the relative significance between the  
71 uncertainties associated with the two-moment scheme and those created by the resolutions. These  
72 understanding and quantification can provide us with a guideline on how to represent microphysics in  
73 the two-moment schemes and sub-grid processes in the scale-aware schemes for the efficient reduction  
74 of the uncertainties in the NWP models. Note that the representation of sub-grid processes requires  
75 information on the contribution of resolution to the uncertainties and, in this study, we focus on the two-  
76 moment scheme developed by Morrison and Gettelman (2008) and Morrison et al. (2009), which is  
77 referred to as the MG scheme, henceforth.

78 Fan et al. (2012) and Khain et al. (2015) have shown that the parameterizations of three key  
79 microphysical processes (i.e., saturation, collection, and sedimentation) in microphysical schemes act as  
80 a main source of errors in the simulation of clouds, precipitation, and CAPI. We try to identify and  
81 quantify the errors or the uncertainties through comparisons between simulations with  
82 parameterizations of the three key processes in the MG scheme and the CSRMs simulations with full-  
83 fledged microphysical processes. Regarding the understanding of the uncertainties arising from the  
84 choice of resolution, we also perform comparisons between the high-resolution CSRMs simulations and



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85 the low-resolution simulations, and do additional comparisons with the GFS simulations. This helps  
86 gain an understanding of how the microphysical representation and coarse resolution in the GFS model  
87 (as compared to those in the CSRМ) contribute to the uncertainties in the GFS simulations of clouds  
88 and precipitation by accounting for CAPI. Here, the CSRМ simulations act as benchmark simulations  
89 by representing microphysical processes with high-level sophistication and by resolving cloud-scale  
90 physical and dynamic processes with high resolutions.

91

## 92 **2. Models**

93

### 94 **2.1 The CSRМ**

95

96 The Advanced Research Weather Research and Forecasting (ARW) model, a non-hydrostatic  
97 compressible model, is the CSRМ selected for use in this study. A fifth-order monotonic advection  
98 scheme is used for the advection of cloud variables (Wang et al., 2009). The ARW model considers  
99 radiation processes by adopting the Rapid Radiation Transfer Model for General Circulation Models  
100 (RRTMG) (Fouquart and Bonnel, 1980; Mlawer et al., 1997). The RRTMG considers the effects of  
101 aerosols on the effective sizes of hydrometeors and the associated changes in radiation.

102 For an assessment of the uncertainties in the MG scheme, which is a type of a bulk scheme, we  
103 need to use microphysics schemes that are much more sophisticated than the MG scheme. Through  
104 extensive comparisons between various types of bin schemes and bulk schemes, Fan et al. (2012) and



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105 Khain et al. (2015) have concluded that the use of bin schemes or bin-bulk schemes is desirable for  
106 reasonable simulations of clouds, precipitation, and their interactions with aerosols. This is because  
107 these schemes do not use a saturation adjustment, a mass-weight mean terminal velocity, or constant  
108 collection efficiencies that have been used in bulk schemes. Instead, bin schemes use predicted  
109 supersaturation levels, and terminal velocities and collection efficiencies that vary with the sizes of  
110 hydrometeors. Based on the work by Fan et al. (2012) and Khain et al. (2015), this study considers bin  
111 schemes to be a full-fledged microphysics schemes against which the uncertainties in the MG scheme  
112 can be assessed. Hence, a bin scheme is adopted in the CSRM used here.

113 The bin scheme adopted by the CSRM is based on the Hebrew University Cloud Model described  
114 by Khain and Lynn (2009). The bin scheme solves a system of kinetic equations for the size distribution  
115 functions of water drops, ice crystals (plate, columnar and branch types), snow aggregates, graupel and  
116 hail, as well as cloud condensation nuclei. Each size distribution is represented by 33 mass-doubling  
117 bins, i.e., the mass of a particle  $m_k$  in the  $k^{\text{th}}$  bin is  $m_k = 2m_{k-1}$ .

118 It is assumed that aerosol particles are composed of ammonium sulfate. The aerosol size  
119 distribution is calculated prognostically with sinks and sources, which include advection, droplet  
120 nucleation, and aerosol regeneration from droplet evaporation (Fan et al., 2009). Aerosol activation is  
121 calculated according to the Köhler theory, i.e., aerosol particles with radii exceeding the critical value at  
122 a grid point are activated to become droplets based on predicted supersaturation, and the corresponding  
123 bins of the aerosol spectra are emptied. After activation, the aerosol mass is transported within  
124 hydrometeors by collision-coalescence and removed from the atmosphere once hydrometeors that



125 contain aerosols reach the surface. Aerosol particles return to the atmosphere upon evaporation or the  
126 sublimation of hydrometeors that contain them.

127

## 128 2.2 The GFS model

129

130 The GFS model is a global NWP model that is run by the National Oceanic and Atmospheric  
131 Administration (NOAA). The GFS model has 64 vertical sigma-pressure hybrid layers and a T382 (~ 35  
132 km) horizontal resolution. Output fields for a forecast generated at 3-hour intervals (i.e. at 03, 06, 09, 12,  
133 15, 18, 21, 24 universal coordinated time, or Z), starting from the control time of 00Z, are used for this  
134 study.

135 The GFS model posts parameters for 21 vertically different layers. From the surface (1000 hPa) to  
136 the 900-hPa level, the vertical resolution is 25 hPa. Less than 900 hPa, there are 16 levels at a 50-hPa  
137 resolution up to 100 hPa. The cloud phase is determined by the mean temperature ( $T_c$ ) of a cloud layer  
138 which is defined as the average of temperatures at the top and bottom of a cloud layer. If  $T_c$  is less than  
139 258.16 K, the cloud layer is an ice cloud; otherwise, it is a water cloud.

140 A prognostic condensate scheme by Moorthi et al. (2001) has been used to parameterize clouds in  
141 the GFS model. In this scheme, cloud mass, one of the representative cloud variables, evolves by  
142 considering the cloud-mass advection, diffusion and conversion to precipitation, and the diagnosed sub-  
143 grid and grid-scale condensation and evaporation. The grid-scale condensation and evaporation are  
144 calculated based on Sundqvist et al. (1989) and Zhao and Carr (1997), while the sub-grid condensation



145 and evaporation are calculated based on a cumulus parameterization that adopts the mass-flux  
146 approach. This cumulus parameterization was developed by Moorthi et al. (2001) based on a simplified  
147 Arakawa-Schubert scheme (Pan and Wu, 1995).

### 149 **3. The cases**

#### 151 **3.1 The Seoul case**

152  
153 A mesoscale convective system (MCS) was observed over Seoul, Korea (37.57°N, 126.97°E; 0900 local  
154 solar time (LST) 26 July 2011–0900 LST 27 July 2011). This case, referred to as the Seoul case,  
155 involved heavy rainfall with a maximum precipitation rate of  $\sim 150 \text{ mm h}^{-1}$ . This heavy rainfall caused  
156 flash floods and landslides on a mountain at the southern flank of the city, leading to the deaths of 60  
157 people.

158 At 0900 LST July 26th 2011, favorable synoptic-scale features for the development of heavy  
159 rainfall over Seoul were observed. The western Pacific subtropical high (WPSH) was located over the  
160 southeast of Korea and Japan, and there was a low-pressure trough over north China (Figure 1a). Low-  
161 level jets between the flank of the WPSH and the low-pressure system brought warm, moist air from the  
162 Yellow Sea to the Korean Peninsula (Figure 1b). Transport of warm and moist air by the southwesterly  
163 low-level jet is an important condition for the development of heavy rainfall events over Seoul (Hwang  
164 and Lee, 1993; Sun and Lee, 2002).





### 3.2 The Houston case

An MCS was observed over Houston, Texas (29.42°N, 94.45°W; 0700 LST 18 July 2013–0400 LST 19 July 2013). The Houston case involved moderate rainfall with a maximum precipitation rate of ~50 mm h<sup>-1</sup>.

At 0500 LST, two hours before the initiation of convection, the low-level wind in and around Houston was southerly (Figure 1c), favoring the transport of water vapor from the Gulf of Mexico to the Houston area. Associated with this, the environmental convective available potential energy (CAPE) (Figure 1d) in and around Houston along the coastline was high (as represented by red areas in Figure 1d). The high CAPE provided a favorable condition for the development of the MCS.

## 4. Simulations

### 4.1 The CSRM simulations

Using the ARW model and its bin scheme, a three-dimensional CSRM simulation of the observed MCS was performed over the MCS period for each of the cases.

Initial and boundary conditions for the control run are derived from the National Centers for Environmental Prediction GFS final (FNL) analysis. All experiments employ a prognostic surface skin



185 temperature scheme (Zeng and Beljaars, 2005) and a revised roughness length formulation (Donelan  
186 et al., 2004).

187 The control run for each of the cases consists of a domain with a Lambert conformal map  
188 projection. The domain is marked by the rectangle for the Seoul case in Figure 2a and the domain for  
189 the Houston case is shown in Figure 2b. While the control run for the Seoul case is referred to as “the  
190 control-Seoul run”, the control run for the Houston case is referred to as “the control-Houston run”,  
191 henceforth. The domain for the Seoul (Houston) case covers the Seoul (Houston) area and to resolve  
192 cloud-scale processes, a 500-m horizontal resolution is applied to the domain. The domain has 41  
193 vertical layers with resolutions ranging from 70 m near the surface to 800 m at the model top (~50 hPa).  
194 Note that the cumulus parameterization scheme is not used in this domain where convective rainfall  
195 generation is assumed to be explicitly resolved. Based on observations, the aerosol concentration at the  
196 surface at the first time step is set at 5500 (1500)  $\text{cm}^{-3}$  for the Seoul (Houston) case. Above the top of  
197 the planetary boundary layer (PBL) around 2 km, the aerosol concentration reduces exponentially.

198 To examine and isolate CAPI, i.e., the effect of increasing the loading of aerosols on clouds and  
199 precipitation, the control run is repeated with the aerosol concentration at the first time step reduced by  
200 a factor of 10. This factor is based on observations showing that that reduction in aerosol loading  
201 between polluted days and clean days is generally tenfold over Seoul and Houston (Lance et al., 2009;  
202 Kim et al., 2014). This simulation is referred to as the low-aerosol-Seoul run for the Seoul case and the  
203 low-aerosol-Houston run for the Houston case. Since the control-Seoul run and the control-Houston run  
204 involve higher aerosol concentrations than the low-aerosol-Seoul run and the low-aerosol-Houston run,



205 respectively, for naming purposes, the control-Seoul run and the control-Houston run are also  
206 referred to as the high-aerosol-Seoul run and the high-aerosol-Houston run, respectively.

207 In addition to the simulations described above, more simulations were performed to fulfill the goals  
208 of this study (Table 1). Details of those simulations are given in the following sections.

## 210 4.2 The GFS simulations

211  
212 Note that the GFS produces the forecast data over the globe and for this study, we use the data  
213 only during the MCS period and only at grid points in the domain for each case. We collect GFS data in  
214 the domain and then average the data over those grid points at each of the GFS time steps for each of  
215 the cases. For the comparison between the GFS and CSRM simulations at specific time steps over the  
216 MCS period, these averaged data are compared to the CSRM simulations during the period. In case the  
217 time and domain-averaged GFS data are compared to the CSRM counterparts, these averaged data are  
218 averaged again over the MCS period and compared to their CSRM counterparts.

## 220 5. Results

### 222 5.1 Test on the effects of resolution on the simulations of clouds, 223 precipitation, and CAPI

224



### 5.1.1 Cloud liquid content (CLC) and cloud ice content (CIC)

To test the effects of resolution on the simulations of clouds, precipitation, and their interactions with aerosols, we repeat the standard CSRMs at the 500-m resolution (i.e., the high-aerosol-Seoul run, the low-aerosol-Seoul run, the high-aerosol-Houston run, and the low-aerosol-Houston run) by using 15- and 35-km resolutions instead. These resolutions are similar to those generally adopted by current NWP models (e.g., the GFS model) and thus comparisons between these repeated simulations and the CSRMs can evaluate how coarse resolutions adopted by the NWP models affect the simulations of clouds, precipitation, and their interactions with aerosols. The repeated simulations at the 15-km resolution are referred to as the high-aerosol-15-Seoul run, the low-aerosol-15-Seoul run, the high-aerosol-15-Houston run, and the low-aerosol-15-Houston run, while the repeated simulations at the 35-km resolution are referred to as the high-aerosol-35-Seoul run, the low-aerosol-35-Seoul run, the high-aerosol-35-Houston run, and the low-aerosol-35-Houston run. In this study, simulations whose name includes “high-aerosol” represent the polluted scenario, while those whose name includes “low-aerosol” represent the clean scenario. In the following, we describe results from the standard and repeated simulations. For the Houston case, no clouds form at the 35-km resolution, so the description of results is only done for results at the 15-km resolution.

Figures 3a and 3b show the vertical distributions of the time- and domain-averaged CLC in the simulations for the Seoul case and the Houston case, respectively. Figures 4a and 4b show the vertical distributions of the time- and domain-averaged CIC in the simulations for the Seoul case and the



245 Houston case, respectively. There are increases in the cloud mass (represented by CLC and CIC) with  
246 increasing aerosol concentration between the polluted scenario and the clean scenario not only for both  
247 the Seoul and Houston cases but also at all resolutions considered. There are substantial decreases in the  
248 cloud mass at the 15- and 35-km resolutions compared to the cloud mass in the simulations at the 500-  
249 m resolution. In addition, increases in the cloud mass with increasing aerosol concentration reduce  
250 substantially as the resolution coarsens. At the 500-m resolution, on average, there is about a ~30–50%  
251 increase in cloud mass, while at the 15- or 35-km resolutions, there is only a ~2–5% increase in cloud  
252 mass in both cases.

253 For both the Seoul and Houston cases, comparisons between the cloud mass produced by the  
254 GFS simulations and that produced by the ARW simulations show that the GFS-simulated cloud mass  
255 is similar to that in the ARW simulations at the 15- and 35-km resolutions. However, the GFS-  
256 simulated cloud mass is much smaller than that in the ARW simulations at the 500-m resolution, i.e.,  
257 the CSRMs simulations. This suggests that the coarse resolutions used in the GFS simulations are an  
258 important cause of the differences in cloud mass between the CSRMs simulations and the GFS  
259 simulations.

260

### 261 **5.1.2 Liquid water path (LWP) and ice water path (IWP)**

262

263 Figures 5a and 5b show the time series of the domain-averaged LWP and IWP for the Seoul case while  
264 Figures 6a and 6b show the same for the Houston case. Note that LWP and IWP are the vertical



265 integrals of CLC and CIC, respectively. Consequently, the same behavior as that of CLC and CIC is  
266 seen, namely, there are increases in LWP and IWP with increasing aerosol concentrations between the  
267 polluted and clean scenarios at all resolutions, while there are decreases in LWP and IWP with the use  
268 of the 15- and 35-km resolutions compared to using the 500-m resolution. Also, the sensitivity of LWP  
269 and IWP to increasing aerosol concentrations reduces significantly as the resolution coarsens.

270 In Figures 5 and 6, satellite-observed LWP and IWP for both cases follow reasonably well their  
271 CSRМ-simulated counterparts for the polluted scenario. This shows that the CSRМ simulations  
272 perform well and can thus represent benchmark simulations. The GFS-produced LWP and IWP are  
273 similar to those in the ARW simulations at the 15- and 35-km resolutions and are much smaller in  
274 magnitude than those from the CSRМ simulations and observations. Hence, the discrepancy in LWP  
275 and IWP between the GFS and CSRМ simulations or that between the GFS simulations and  
276 observations is closely linked to the coarse resolution adopted by the GFS simulations. Taking the  
277 CSRМ simulations as benchmark simulations, we see that GFS simulations underestimate the cloud  
278 mass compared to observations mainly due to the coarse resolution adopted by the GFS model.

279 Among the ARW simulations, the sensitivity of the cloud mass to increasing aerosol  
280 concentrations reduces considerably with coarsening resolution. CSRМ simulations are benchmark  
281 simulations so the sensitivity in the CSRМ simulations is the benchmark sensitivity. Note that the GFS  
282 simulation results and the ARW simulations at coarse resolutions of 15 and 35 km are similar. Their  
283 sensitivities are thus also likely similar, i.e., the sensitivity of the cloud mass to increasing aerosol



284 concentrations in the GFS simulation is likely to be underestimated compared to the benchmark  
285 sensitivity of the CSRM simulations.

### 287 **5.1.3 Updrafts, condensation, and deposition**

288  
289 To understand the response of the cloud mass to increasing aerosol concentrations, and the variation in  
290 the cloud mass and its response to increasing aerosol concentrations with varying resolutions as shown  
291 in Figures 3, 4, 5, and 6, we calculate updraft mass fluxes since these fluxes control supersaturation that  
292 in turn controls condensation and deposition as key determination factors for the cloud mass. We also  
293 obtain condensation and deposition rates. The vertical distributions of time- and domain-averaged  
294 updraft mass fluxes, condensation rates, and deposition rates for the Seoul and Houston cases are shown  
295 in Figures 7, 8, and 9, respectively.

296 As seen for the cloud mass, updraft mass fluxes, and condensation and deposition rates increase  
297 with increasing aerosol concentrations between the polluted scenario and the clean scenario at all  
298 resolutions and for all cases considered. Aerosol-induced percentage increases in updraft mass fluxes,  
299 and deposition and condensation rates at the 500-m resolution are approximately one order of  
300 magnitude greater than those at the 15-km and 35-km resolutions. Stated differently, the sensitivity of  
301 updraft mass fluxes to increasing aerosol concentrations reduces substantially with coarsening  
302 resolution and due to this, the sensitivity of deposition and condensation rates, and thus the cloud mass,  
303 to increasing aerosol concentrations also reduces substantially with coarsening resolution.



304 Similar to the situation with the cloud mass, the GFS-produced updraft mass fluxes are much smaller  
305 than those produced by the ARW simulations at the 500-m resolution (or the CSRMs simulations) and  
306 similar to those produced by the ARW simulations at the 15- and 35-km resolutions (Figure 7). Hence,  
307 the discrepancy in updraft mass fluxes between the GFS simulations and the CSRMs simulations is  
308 closely linked to the discrepancy in resolutions between these two types of simulations. The  
309 underestimation of the updraft mass fluxes in the GFS simulations is mainly due to the coarse resolution  
310 adopted by the GFS model. Taking the sensitivity of updraft mass fluxes to increasing aerosol  
311 concentrations in the CSRMs simulations as the benchmark sensitivity, the GFS simulations likely also  
312 underestimate the sensitivity.

313 Figure 10 shows the frequency distribution of updrafts over the updraft speed, which is normalized  
314 over the domain and the simulation period. We first calculate the frequency over the domain at each  
315 time step and in each discretized updraft bin. The frequency in each bin and at each time step is then  
316 divided by the total number of grid points in the whole domain. The normalized frequency at each time  
317 step is summed over all of the time steps in each updraft bin. This sum is divided by the total number of  
318 time steps as the final step in the normalization process. With coarsening resolution, the normalized  
319 frequency of weak updrafts with speeds less than  $\sim 2 \text{ m s}^{-1}$  increases for both scenarios in both cases.  
320 However, the normalized frequency of strong updrafts with speeds greater than  $\sim 2 \text{ m s}^{-1}$  reduces with  
321 coarsening resolution. The frequency shift from high-level updraft speeds to low-level speeds leads to a  
322 reduction in mean updrafts with coarsening resolution for both scenarios in both cases.





323 The updraft frequency is greater in the polluted scenario than in the clean scenario at all  
324 resolutions and for all cases. The overall difference in the frequency between the scenarios reduces with  
325 coarsening resolution. This is associated with the reduction in the sensitivity of the averaged updrafts to  
326 increasing aerosol concentrations with coarsening resolution. In particular, the difference in the  
327 frequency for weak updrafts (speeds less than  $\sim 2 \text{ m s}^{-1}$ ) between the scenarios does not vary much with  
328 coarsening resolution. On average, the percentage difference for weak updrafts is less than 2–3% at all  
329 resolutions. However, the difference for strong updrafts varies significantly with varying resolution.  
330 The mean difference for strong updrafts varies from  $\sim 30$ – $60\%$  for the 500-m resolution to less than  $\sim 5$ –  
331  $6\%$  for the 15- and 35-km resolutions. Numerous studies (e.g., Khain et al., 2005; Seifert and Beheng,  
332 2006; Tao et al., 2007, 2012; van den Heever and Cotton, 2007; Storer et al., 2010; Lee et al., 2013,  
333 2017) have shown that aerosol-induced invigoration of convection through aerosol-induced increases in  
334 freezing or aerosol-induced intensification of gust fronts is the main mechanism behind aerosol-induced  
335 increases in updraft mass fluxes or the intensity of updrafts. Based on this, analyses of the updraft  
336 frequency here suggest that strong updrafts are more sensitive to aerosol-induced invigoration than  
337 weak updrafts. The variation in the sensitivity of the averaged updrafts to increasing aerosol  
338 concentrations at varying resolutions is associated more with the variation of the response of strong  
339 updrafts to aerosol-induced invigoration at varying resolutions than with that of weak updrafts. Another  
340 point to make here is that the frequency of weak updrafts is overestimated while that of strong updrafts  
341 is underestimated at coarse resolutions compared to the frequencies in the fine-resolution CSRM  
342 simulations.



#### 5.1.4 Evaporation and precipitation distributions

Aerosol-induced increases in evaporation and associated cooling affect downdrafts, and changes in downdrafts in turn affect gust fronts. Aerosol-induced changes in the intensity of gust fronts affect the organization of cloud systems, which is characterized by cloud-cell spatiotemporal distributions. In general, aerosol-induced greater increases in evaporation result in aerosol-induced greater changes in the intensity of gust fronts and in cloud system organization (Tao et al., 2007, 2012; van den Heever and Cotton, 2007; Storer et al., 2010; Lee et al., 2013, 2017).

Considering that individual cloud cells act as individual sources of precipitation, aerosol-induced changes in the cloud system organization can alter precipitation spatiotemporal distributions, which play an important role in hydrological circulations. It is thus important to examine how the response of evaporation to increasing aerosol concentrations varies with varying resolution, i.e., to see how coarse resolutions affect the quality of simulations of aerosol effects on hydrological circulations. Motivated by this, evaporation rates are obtained and are shown in Figure 11.

As seen in the above-described variables, evaporation rates increase as the aerosol concentration increases and the sensitivity of the evaporation rate to increasing aerosol concentrations reduces with coarsening resolution among the ARW simulations. This suggests that the sensitivities of the cloud system organization and precipitation distributions to increasing aerosol concentrations likely also reduce with coarsening resolution, as reported in previous studies (e.g., Tao et al., 2007, 2012; van den



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363 Heever and Cotton, 2007; Storer et al., 2010; Lee et al., 2013, 2017). This is confirmed by the  
364 distribution of normalized precipitation frequency over precipitation rates shown in Figure 12. Similar  
365 to the normalization for the updraft frequency, we first calculate the frequency of surface precipitation  
366 rates at each time step and in each discretized precipitation rate bin. The frequency in each bin and at  
367 each time step is then divided by the total number of grid points at the surface. The normalized  
368 frequency at each time step is summed over all of the time steps. This sum is divided by the total  
369 number of time steps as the final step in the normalization process. Figure 12 shows that due to the  
370 reduction in the sensitivity of evaporative cooling to increasing the aerosol concentration as the  
371 resolution coarsens, differences in the distribution of precipitation frequency between the polluted  
372 scenario and the clean scenario reduce substantially as the resolution coarsens. Taking the 500-m  
373 resolution CSRM simulations as benchmark simulations, this suggests that the coarse-resolution GFS  
374 simulations likely underestimate the sensitivity of evaporative cooling, cloud system organization, and  
375 precipitation distributions to increasing aerosol concentrations.

## 377 **5.2 Test on the effects of microphysics parameterizations on the simulations of clouds,** 378 **precipitation, and CAPI**

379  
380 As mentioned previously, among microphysical processes, saturation, sedimentation, and collection  
381 processes are those whose parameterizations are a main cause of errors in the simulation of clouds,  
382 precipitation, and CAPI. Motivated by this, we focus on these three microphysical processes for testing



383 the effects of microphysics parameterizations on the simulations of clouds, precipitation, and CAPI.

384 As a preliminary step to this test, we first focus on the effects of microphysics parameterizations on the  
385 simulation of the cloud mass, which plays a key role in cloud radiative properties and precipitation.  
386 Based on Figures 3 and 4, we focus on the CLC, which accounts for the bulk of the total cloud mass.

387 Figure 13 shows the vertical distributions of the time- and domain-averaged CLC. In Figure 13a,  
388 solid red and black lines represent the high-aerosol-Seoul run and the low-aerosol-Seoul run,  
389 respectively, while in Figure 13b, those lines represent the high-aerosol-Houston run and the low-  
390 aerosol-Houston run, respectively. Note that these runs shown in the figure are performed using the bin  
391 scheme and the 500-m resolution. These simulations were repeated with the Morrison two-moment  
392 scheme. These repeated simulations using the MG scheme, referred to as the high-aerosol-MG-Seoul  
393 run, the low-aerosol-MG-Seoul run, the high-aerosol-MG-Houston run and the low-aerosol-MG-  
394 Houston run, are represented by solid yellow and green lines in Figure 13. Between the high-aerosol  
395 and low-aerosol runs using the MG scheme for the two cases, there is an increase in CLC with  
396 increasing aerosol concentration. However, this increase is much smaller than that between the high-  
397 aerosol and low-aerosol runs using the bin scheme for the two cases. In addition, there is a significant  
398 difference in the shape of the vertical profile of CLC between the simulations with the MG scheme and  
399 those with the bin scheme for both cases. Here, the shape is represented by the peak value of CLC and  
400 the altitude of the peak value in the vertical profile. The peak value is higher in the simulations with the  
401 bin scheme than in the simulations with the MG scheme for each of the polluted and clean scenarios.  
402 The altitude of the peak value is lower in the simulations with the bin scheme than in the simulations



403 with the MG scheme. For the Seoul (Houston) case, the altitude is ~2 (3) km in the simulations with  
404 the bin scheme, while it is ~5 km in those with the MG scheme.

405 We next test how the parameterization of saturation processes affects the simulations by  
406 comparing the supersaturation prediction in the bin scheme to the saturation adjustment in the MG  
407 scheme. To do this, the simulations with the bin scheme are repeated after replacing the supersaturation  
408 prediction in the bin scheme with the saturation adjustment in the MG scheme. These repeated  
409 simulations are referred to as the high-aerosol-sat-Seoul run, the low-aerosol-sat-Seoul run, the high-  
410 aerosol-sat-Houston run, and the low-aerosol-sat-Houston run. The high-aerosol-sat-Seoul run and the  
411 low-aerosol-sat-Seoul run for the Seoul case and the high-aerosol-sat-Houston run and the low-aerosol-  
412 sat-Houston run for the Houston case are represented by dashed lines in Figure 13. As in the other  
413 simulations, there is an increase in CLC with increasing aerosol concentrations between the high-  
414 aerosol-sat and the low-aerosol-sat runs for the two cases. However, this increase is much smaller than  
415 that between the high-aerosol and low-aerosol runs for the two cases, but is similar to that between the  
416 high-aerosol-MG and low-aerosol-MG runs for the two cases. This suggests that the sensitivity of the  
417 CLC to increasing aerosol concentrations is affected by the parameterization of the saturation process  
418 and that the use of the saturation adjustment reduces the sensitivity compared to using the  
419 supersaturation prediction.

420 The high-aerosol-sat-Seoul run, the low-aerosol-sat-Seoul run, the high-aerosol-sat-Houston run,  
421 and the low-aerosol-sat-Houston run are repeated by replacing the bin-scheme sedimentation with the  
422 sedimentation from the MG scheme as a way of testing the effects of the parameterization of



423 sedimentation on the simulations. These repeated runs are referred to as the high-aerosol-sed-Seoul  
424 run, the low-aerosol-sed-Seoul run, the high-aerosol-sed-Houston run, and the low-aerosol-sed-Houston  
425 run. These runs are identical to the high-aerosol-Seoul run, the low-aerosol-Seoul run, the high-aerosol-  
426 Houston run and the low-aerosol-Houston run, respectively, except for the parameterization of the  
427 saturation and sedimentation processes. As mentioned previously, terminal velocities vary as  
428 hydrometeor sizes vary in the bin scheme, while the MG scheme adopts mass-weight mean terminal  
429 velocities for the calculation of the sedimentation process.

430 The vertical distributions of the CLC in the high-aerosol-sed-Seoul run, the low-aerosol-sed-Seoul  
431 run, the high-aerosol-sed-Houston run, and the low-aerosol-sed-Houston run are represented by dashed  
432 lines in Figure 14. Comparisons between the pair of high-aerosol-sed and low-aerosol-sed runs for the  
433 two cases and the pair of high-aerosol-MG and low-aerosol-MG runs for the two cases show that not  
434 only the increases in the CLC with increasing aerosol concentrations but also the shapes of the vertical  
435 distribution of the CLC in the high-aerosol-sed and low-aerosol-sed runs for the two cases are similar to  
436 those in the high-aerosol-MG and low-aerosol-MG runs for the two cases. This demonstrates that  
437 differences in the shape of the vertical profile of CLC between the bin-scheme simulations and the MG-  
438 scheme simulations are not explained by differences in the representation of the saturation process  
439 alone. This also demonstrates that the representation of the sedimentation process plays an important  
440 role in generating the differences in the shape of the vertical profile of CLC.

441 In Figure 14, we still see differences in the vertical profiles of CLC between the high-aerosol-sed-  
442 Seoul and high-aerosol-MG-Seoul runs, and between the low-aerosol-sed-Seoul and low-aerosol-MG-



443 Seoul runs, as well as between the high-aerosol-sed-Houston and high-aerosol-MG-Houston runs,  
444 and between the low-aerosol-sed-Houston and low-aerosol-MG-Houston runs. To understand the cause  
445 of these differences, the high-aerosol-sed-Seoul run, the low-aerosol-sed-Seoul run, the high-aerosol-  
446 sed-Houston run, and the low-aerosol-sed-Houston run are repeated again with the MG-scheme  
447 collection process. These repeated runs are referred to as the high-aerosol-col-Seoul run, the low-  
448 aerosol-col-Seoul run, the high-aerosol-col-Houston run, and the low-aerosol-col-Houston run. These  
449 runs are identical to the high-aerosol-Seoul run, the low-aerosol run-Seoul, the high-aerosol-Houston  
450 run, and the low-aerosol-Houston run, respectively, except for the parameterization of the saturation,  
451 sedimentation, and collection processes. As mentioned previously, collection efficiencies vary as  
452 hydrometeor sizes vary in the bin scheme, while the MG scheme uses constant collection efficiencies.

453 As seen in Figure 15, the remaining differences between the high-aerosol-col-Seoul and high-  
454 aerosol-MG-Seoul runs and between the low-aerosol-col-Seoul and low-aerosol-MG-Seoul runs, as  
455 well as between the high-aerosol-col-Houston and high-aerosol-MG-Houston runs, and between the  
456 low-aerosol-col-Houston and low-aerosol-MG-Houston runs nearly disappear. This demonstrates with  
457 fairly good confidence that differences between the high-aerosol-Seoul run (the high-aerosol-Houston  
458 run) and the high-aerosol-MG-Seoul run (the high-aerosol-MG-Houston run) or between the low-  
459 aerosol-Seoul run (the low-aerosol-Houston run) and the low-aerosol-MG-Seoul run (the low-aerosol-  
460 MG-Houston run) are explained by differences in the parameterizations of the saturation,  
461 sedimentation, and collection processes between the bin scheme and the MG scheme.

462



### 5.3 Relative importance of resolution and parameterizations

Comparisons between ARW simulations with different resolutions and those with different microphysics parameterizations as shown in Figures 3 and 13 demonstrate that the variation in cloud variables is much greater with respect to the variation in resolution than with the variation in microphysics parameterizations. For example, comparisons between Figures 3 and 13 show that the variation in the time- and domain-averaged cloud mass is  $\sim 2$ – $4$  times greater as the resolution varies than when the microphysics parameterizations varies. These comparisons also show that the variation in cloud variables with varying resolutions explains the discrepancy between GFS simulations and CSRMs simulations and between GFS simulations and observations much better than the variation in microphysics parameterizations. As a first step toward reducing the first-order errors in the GFS simulations, we first need to focus on the reduction in errors that are associated with the use of coarse resolutions in the GFS model.

## 6. Summary and Discussion

This study examines the uncertainties in the simulations of clouds, precipitation, and CAPI in the NWP models. Here, we focus on those uncertainties that are created by the microphysics parameterizations and by the model resolution chosen. In particular, for the examination of the uncertainties associated





482 with microphysics parameterizations, we investigate the contributions of the parameterizations of  
483 three key microphysical processes, i.e., saturation, collection, and sedimentation, to the uncertainties.

484 As a way of examining the uncertainties created by the microphysics parameterizations, we  
485 compare the MG scheme (a representative bulk scheme) to the bin scheme, which acts as a benchmark  
486 scheme. The vertical distribution of the cloud mass simulated by the MG scheme deviates substantially  
487 from that simulated by the bin scheme. In particular, there is a substantial discrepancy in the peak value  
488 of the distribution and the altitude of the peak value between the schemes. Also, there is a substantial  
489 discrepancy between the schemes in the sensitivity of the cloud mass to increasing aerosol  
490 concentrations.

491 The discrepancy in the sensitivity is closely linked to the discrepancy in the parameterization of the  
492 saturation processes between the schemes. The use of the saturation adjustment in the bulk scheme  
493 reduces the sensitivity by a factor of  $\sim 2$  compared to the use of the supersaturation prediction in the bin  
494 scheme. The discrepancy in the peak value and its altitude between the schemes is strongly linked to the  
495 parameterization of sedimentation in the schemes. The use of identical parameterizations of saturation  
496 and sedimentation makes the sensitivity and the peak value and its altitude similar between the schemes,  
497 although there still remains a slight difference in the magnitude of the cloud mass. This remaining  
498 difference is explained by the discrepancy in the parameterization of the collection process. When the  
499 two schemes use identical parameterizations of the saturation, sedimentation, and collection processes,  
500 the sensitivity and the peak value and its altitude become nearly identical between the two schemes.  
501 This confirms that differences in the parameterizations of the three key processes (i.e., saturation,



502 sedimentation, and collection) are the main cause of the differences in the simulations of clouds  
503 between the schemes as indicated by Fan et al. (2012) and Khain et al. (2015).

504 By selecting the simulations with the bin scheme as benchmark simulations, we see that the use  
505 of the saturation adjustment, as done in most current NWP models, can lead to an underestimation of  
506 the sensitivity of the cloud mass to increasing aerosol concentrations. Fan et al. (2012) and Khain et al.  
507 (2015) have also shown that the sensitivity of the cloud mass to increasing aerosol concentrations is  
508 lower in the bulk scheme than in the bin scheme. This study shows that the lower sensitivity in the bulk  
509 scheme is closely linked to the use of the saturation adjustment in the bulk scheme.

510 It is well known that the shape of the vertical profile of the cloud mass (i.e., the peak value of the  
511 cloud mass and its altitude) or how cloud mass is distributed in the vertical domain has substantial  
512 implications for cloud radiative forcing and precipitation processes. This study demonstrates that the  
513 different parameterizations of the sedimentation process between the schemes lead to different shapes  
514 of the cloud-mass profiles and thus different cloud radiative forcings and precipitation processes. The  
515 use of a mass-weight mean terminal velocity for sedimentation as done in the bulk schemes can lead to  
516 misleading shapes, cloud radiative forcings, and precipitation processes compared to those in the  
517 benchmark bin-scheme simulations where terminal velocities vary as hydrometeor sizes vary.

518 NWP models (e.g., the GFS model) adopt coarse resolutions. This study shows that the use of  
519 coarse resolutions can cause an underestimation of the updraft intensity and thus supersaturation, which  
520 leads to an underestimation of the cloud mass. Also, the use of coarse resolutions likely results in the



521 underestimation of the sensitivity of updrafts and cloud mass and that of evaporation, cloud system  
522 organization, and precipitation distributions to increasing aerosol concentrations.

523 Through the examination of the sensitivity of the results to the resolution chosen, we find that  
524 updrafts, associated other cloud variables, and their sensitivity to increasing aerosol concentrations are  
525 strongly controlled by small-scale updrafts. When they are resolved with the use of high-resolution  
526 models, there are high-level averaged updrafts, associated variables, and their sensitivity but when they  
527 are not resolved in low-resolution models, there are low-level averaged updrafts, associated variables,  
528 and their sensitivity. This means that small-scale updrafts not resolved with coarse resolutions play an  
529 important role in the simulation of the correct magnitude of updrafts, associated variables, and their  
530 sensitivity to increasing aerosol concentrations.

531 The frequency distributions of updrafts simulated in this study show that the frequency of weak  
532 updrafts is overestimated while that of strong updrafts is underestimated in the simulations with coarse  
533 resolutions compared to those in the CSRMs simulations. Hence, the updraft speed shifts toward lower  
534 values with coarsening resolution. The difference in the frequency between the polluted and clean  
535 scenarios reduces substantially, particularly for strong updrafts, with coarsening resolution. This is why  
536 the sensitivity of updrafts and associated cloud variables to increasing aerosol concentrations reduces  
537 with coarsening resolution. We see that not resolving small-scale updrafts results in the underestimation  
538 of strong updrafts and the overestimation of weak updrafts for both scenarios and in the reduced  
539 difference in strong updrafts between the scenarios.



540 The GFS simulations use the so-called sub-grid parameterizations (e.g., cumulus  
541 parameterizations) that represent sub-grid updrafts and associated variables, while the ARW  
542 simulations at the 500-m resolution (i.e., the CSRSM simulations) do not use these sub-grid  
543 parameterizations based on consideration that the CSRSM simulations resolve sub-grid processes. Thus,  
544 the CSRSM simulations (that prove to act as benchmark simulations through comparisons to  
545 observations) are able to evaluate the sub-grid parameterizations in the GFS model. The sub-grid  
546 parameterizations are designed to correct errors that are caused by the use of coarse resolutions in the  
547 GFS model. However, comparisons between the GFS simulations and the ARW simulations at different  
548 resolutions indicate that despite the presence of sub-grid parameterizations in the GFS model, the errors  
549 or differences in the updraft intensity and associated cloud variables between the GFS simulations and  
550 the CSRSM simulations exist due to resolutions. Hence, sub-grid parameterizations need to be improved  
551 to better represent sub-grid processes. To this end, results here indicate that sub-grid parameterizations  
552 (e.g., scale-aware cumulus schemes) which are being implemented into the NWP models (e.g., the GFS  
553 model) should be able to compensate for the over- and under-estimation of weak updrafts and strong  
554 updrafts, respectively, due to coarse resolutions.

555 Comparisons between GFS simulations and ARW simulations also indicate that it is likely that  
556 the GFS model underestimates the sensitivity of updrafts and associated cloud variables to increasing  
557 aerosol concentrations. In general, parameterizations that represent sub-grid updrafts and other  
558 associated variables do not have pathways through which increasing aerosol concentrations affect  
559 updrafts and associated cloud variables. However, recent studies by Lim et al. (2014), Thayer-Calder et



560 al. (2015), and Griffin and Larson (2016) have attempted to consider interactions among  
561 microphysical processes, their variations with varying aerosol concentrations, and sub-grid dynamic  
562 (e.g., updrafts and downdrafts) and thermodynamic (e.g., temperature) variables in those  
563 parameterizations. These efforts should focus on countering the variation in the sensitivity of updrafts,  
564 in particular strong updrafts and thus that of cloud variables, cloud system organization, and  
565 precipitation distributions to increasing aerosol concentrations with coarsening resolution. While the  
566 pattern of the sensitivity and its variation shown in this study provides valuable information useful for  
567 aiding these efforts, results may be different for different cloud types and environments, given the  
568 strong dependence of aerosol-cloud interactions on cloud type and environmental conditions. So to aid  
569 the efforts in a generalized way, future studies with more cases that involve various types of aerosol-  
570 cloud interactions are needed.

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719 **Tables**

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721 Table 1. Description of the simulations.

| Simulations                 | Case    | Aerosol number concentration at the surface (cm <sup>-3</sup> ) | Microphysics scheme | Resolution | Saturation                 | Sedimentation            | Collection            |
|-----------------------------|---------|---|---------------------|------------|----------------------------|--------------------------|-----------------------|
| High-aerosol-Seoul run      | Seoul   | 5500  | Bin                 | 500 m      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-Seoul run       | Seoul   | 550   | Bin                 | 500 m      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| High-aerosol-Houston run    | Houston | 1500  | Bin                 | 500 m      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-Houston run     | Houston | 150   | Bin                 | 500 m      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| High-aerosol-15-Seoul run   | Seoul   | 5500  | Bin                 | 15 km      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-15-Seoul run    | Seoul   | 550   | Bin                 | 15 km      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| High-aerosol-15-Houston run | Houston | 1500  | Bin                 | 15 km      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-15-Houston      | Houston | 150   | Bin                 | 15 km      | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |



| run                         |         |      |     |       |                            |                          |                       |
|-----------------------------|---------|------|-----|-------|----------------------------|--------------------------|-----------------------|
| High-aerosol-35-Seoul run   | Seoul   | 5500 | Bin | 35 km | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-35-Seoul run    | Seoul   | 550  | Bin | 35 km | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| High-aerosol-35-Houston run | Houston | 1500 | Bin | 35 km | Supersaturation prediction | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-35-Houston run  | Houston | 150  | Bin | 35 km | Supersaturation prediction | MG-scheme sedimentation  | MG-scheme collection  |
| High-aerosol-MG-Seoul run   | Seoul   | 5500 | MG  | 500 m | Saturation adjustment      | MG-scheme sedimentation  | MG-scheme collection  |
| Low-aerosol-MG-Seoul run    | Seoul   | 550  | MG  | 500 m | Saturation adjustment      | MG-scheme sedimentation  | MG-scheme collection  |
| High-aerosol-MG-Houston run | Houston | 1500 | MG  | 500 m | Saturation adjustment      | MG-scheme sedimentation  | MG-scheme collection  |
| Low-aerosol-MG-Houston run  | Houston | 150  | MG  | 500 m | Saturation adjustment      | MG-scheme sedimentation  | MG-scheme collection  |
| High-aerosol-sat-Seoul run  | Seoul   | 5500 | Bin | 500 m | Saturation adjustment      | Bin-scheme sedimentation | Bin-scheme collection |
| Low-aerosol-sat-Seoul run   | Seoul   | 550  | Bin | 500 m | Saturation adjustment      | Bin-scheme sedimentation | Bin-scheme collection |
| High-aerosol-               | Houston | 1500 | Bin | 500 m | Saturation                 | Bin-scheme               | Bin-scheme            |



|                                     |         |      |     |       |                          |                             |                          |
|-------------------------------------|---------|------|-----|-------|--------------------------|-----------------------------|--------------------------|
| sat-Houston<br>run                  |         |      |     |       | adjustment               | sedimentation               | collection               |
| Low-aerosol-<br>sat-Houston<br>run  | Houston | 150  | Bin | 500 m | Saturation<br>adjustment | Bin-scheme<br>sedimentation | Bin-scheme<br>collection |
| High-aerosol-<br>sed-Seoul run      | Seoul   | 5500 | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | Bin-scheme<br>collection |
| Low-aerosol-<br>sed-Seoul run       | Seoul   | 550  | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | Bin-scheme<br>collection |
| High-aerosol-<br>sed-Houston<br>run | Houston | 1500 | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | Bin-scheme<br>collection |
| Low-aerosol-<br>sed-Houston<br>run  | Houston | 150  | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | Bin-scheme<br>collection |
| High-aerosol-<br>col-Seoul run      | Seoul   | 5500 | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | MG-scheme<br>collection  |
| Low-aerosol-<br>col-Seoul run       | Seoul   | 550  | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | MG-scheme<br>collection  |
| High-aerosol-<br>col-Houston<br>run | Houston | 1500 | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | MG-scheme<br>collection  |
| Low-aerosol-<br>col-Houston<br>run  | Houston | 150  | Bin | 500 m | Saturation<br>adjustment | MG-scheme<br>sedimentation  | MG-scheme<br>collection  |



723 **FIGURE CAPTIONS**

724  
725 Figure 1. (a) Sea-level pressure (hPa) and (b) 850 hPa wind ( $\text{m s}^{-1}$ ; arrows), geopotential height (m;  
726 contours) and equivalent potential temperature (K; shaded) at 0900 LST July 26<sup>th</sup> 2011 over the  
727 northeast Asia. The rectangles in the Korean Peninsula in panels (a) and (b) mark the center of Seoul. (c)  
728 Sea-level pressure (hPa;shaded) and wind at 10 m above sea level ( $\text{m s}^{-1}$ ; barbs) and (d) convective  
729 available potential energy ( $\text{J kg}^{-1}$ ) at 0500 LST 18 July 2013 in and around Houston. The rectangles in  
730 panels (c) and (d) mark the center of Houston.

731  
732 Figure 2. (a) The domain (marked by the rectangle) used in simulations for the Seoul case. The small  
733 white circle marks the center of Seoul. (b) The domain used in simulations for the Houston case. The  
734 small white circle marks the center of Houston.

735  
736 Figure 3. Vertical distributions of the time- and domain-averaged cloud liquid content (CLC) for (a) the  
737 Seoul case and (b) the Houston case. Solid lines represent simulations at the 500-m resolution, while  
738 dashed lines represent those at the 15-km resolution. Dotted lines represent simulations at the 35-km  
739 resolution and blue lines represent GFS-simulated CLC.

740  
741 Figure 4. Same as Figure 3, but for cloud ice content (CIC).

742  
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744 Figure 5. Time series of the domain-averaged (a) liquid water path (LWP) and (b) ice water path  
745 (IWP) for the Seoul case. Solid lines represent simulations at the 500-m resolution, while dashed and  
746 dotted lines represent those at 15-km and 35-km resolutions, respectively. Blue lines represent GFS-  
747 simulated LWP and IWP and green lines represent observed LWP and IWP.

748  
749 Figure 6. Same as Figure 5, but for the Houston case.

750  
751 Figure 7. Vertical distributions of the time- and domain-averaged updraft mass fluxes for (a) the Seoul  
752 case and (b) the Houston case. Solid lines represent simulations at the 500-m resolution, while dashed  
753 lines represent those at the 15-km resolution. Dotted lines represent simulations at the 35-km resolution  
754 and blue lines represent GFS-simulated updraft mass fluxes.

755  
756 Figure 8. Vertical distributions of the time- and domain-averaged condensation rates for (a) the Seoul  
757 case and (b) the Houston case. Solid lines represent simulations at the 500-m resolution, while dashed  
758 lines represent those at the 15-km resolution. Dotted lines represent simulations at the 35-km resolution.

759  
760 Figure 9. Same as Figure 8, but for deposition rates.

761



762 Figure 10. Distributions of normalized updraft frequency over updraft speeds for (a) the Seoul case  
763 and (b) the Houston case. Solid lines represent simulations at the 500-m resolution, while dashed lines  
764 represent those at the 15-km resolution. Dotted lines represent simulations at the 35-km resolution.

765  
766 Figure 11. Same as Figure 8, but for evaporation rates.

767  
768 Figure 12. Distributions of normalized precipitation frequency over precipitation rates for (a) the Seoul  
769 case and (b) the Houston case. Solid lines represent simulations at the 500-m resolution, while dashed  
770 lines represent those at the 15-km resolution. Dotted lines represent simulations at the 35-km resolution.

771  
772 Figure 13. Vertical distributions of the time- and domain-averaged cloud liquid content (CLC) for (a)  
773 the Seoul case and (b) the Houston case. Solid red and black lines represent simulations with the bin  
774 scheme and at the 500-m resolution, while dashed red and black lines represent the bin-scheme  
775 simulations with the saturation adjustment. Solid yellow and green lines represent simulations with the  
776 MG scheme.

777  
778 Figure 14. Vertical distributions of the time- and domain-averaged cloud liquid content (CLC) for (a)  
779 the Seoul case and (b) the Houston case. Solid red and black lines represent simulations with the bin  
780 scheme and at the 500-m resolution, while dashed red and black lines represent the bin-scheme



781 simulations with the saturation adjustment and the MG scheme sedimentation process. Solid yellow  
782 and green lines represent simulations with the MG scheme.

783

784 Figure 15. Vertical distributions of the time- and domain-averaged cloud liquid content (CLC) for (a)  
785 the Seoul case and (b) the Houston case. Solid red and black lines represent simulations with the bin  
786 scheme and at the 500-m resolution, while dashed red and black lines represent the bin-scheme  
787 simulations with the saturation adjustment and the MG scheme sedimentation and collection processes.  
788 Solid yellow and green lines represent simulations with the MG scheme.

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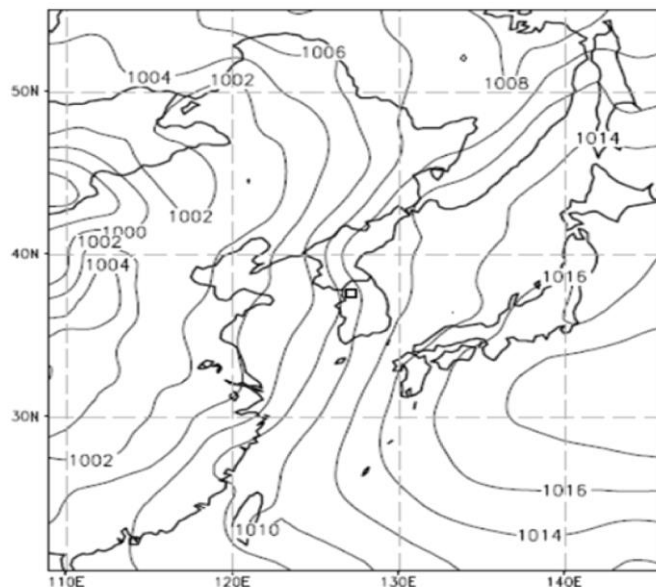
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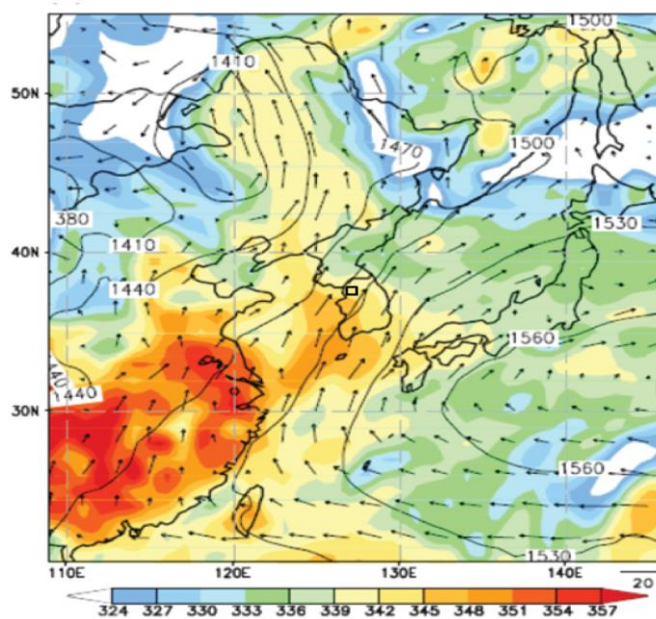




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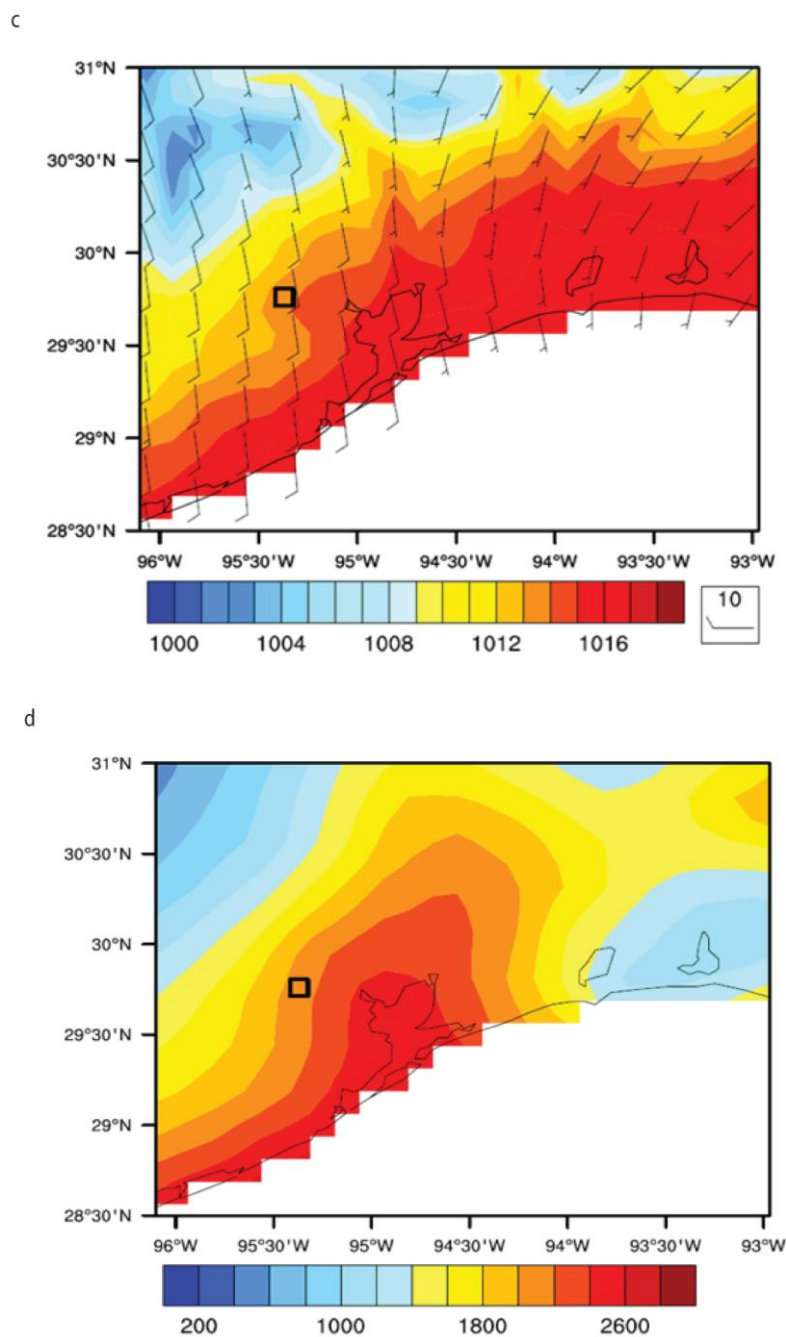
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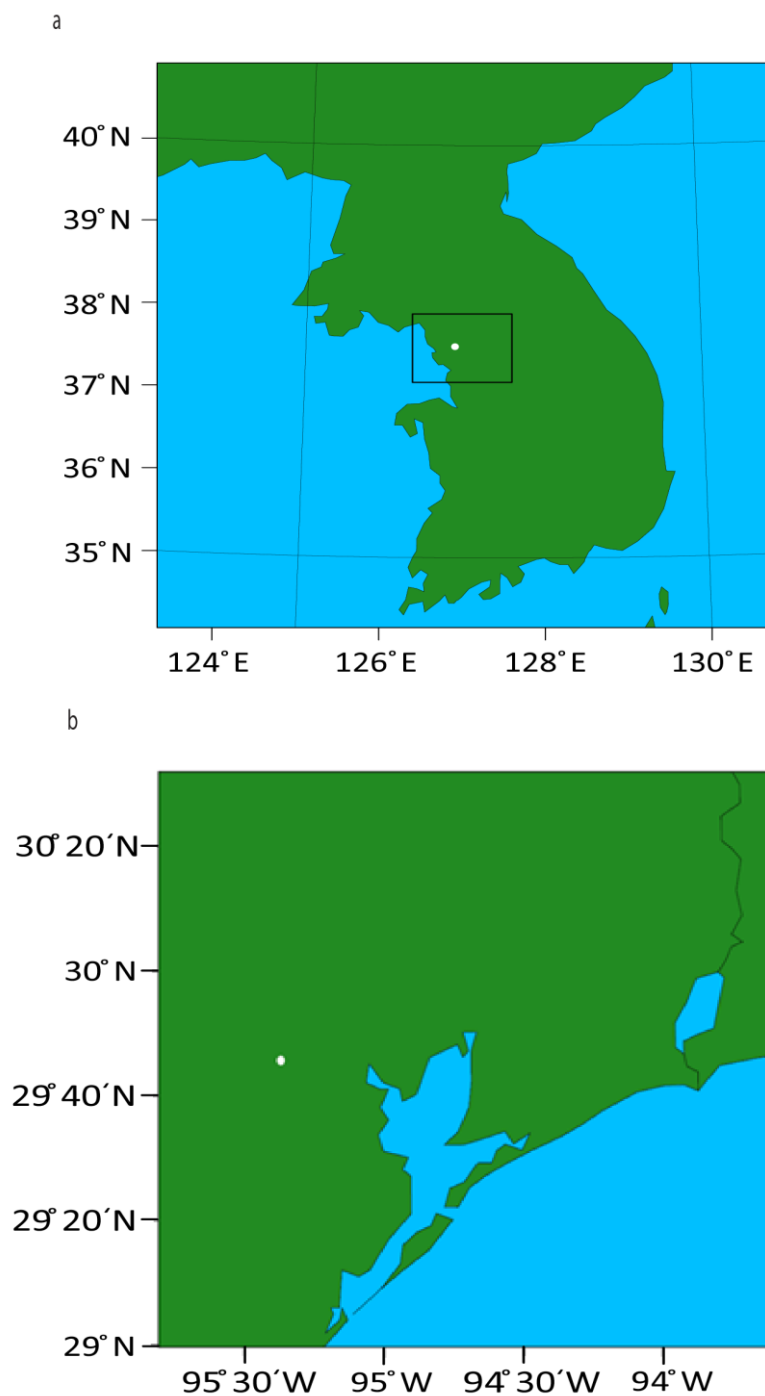
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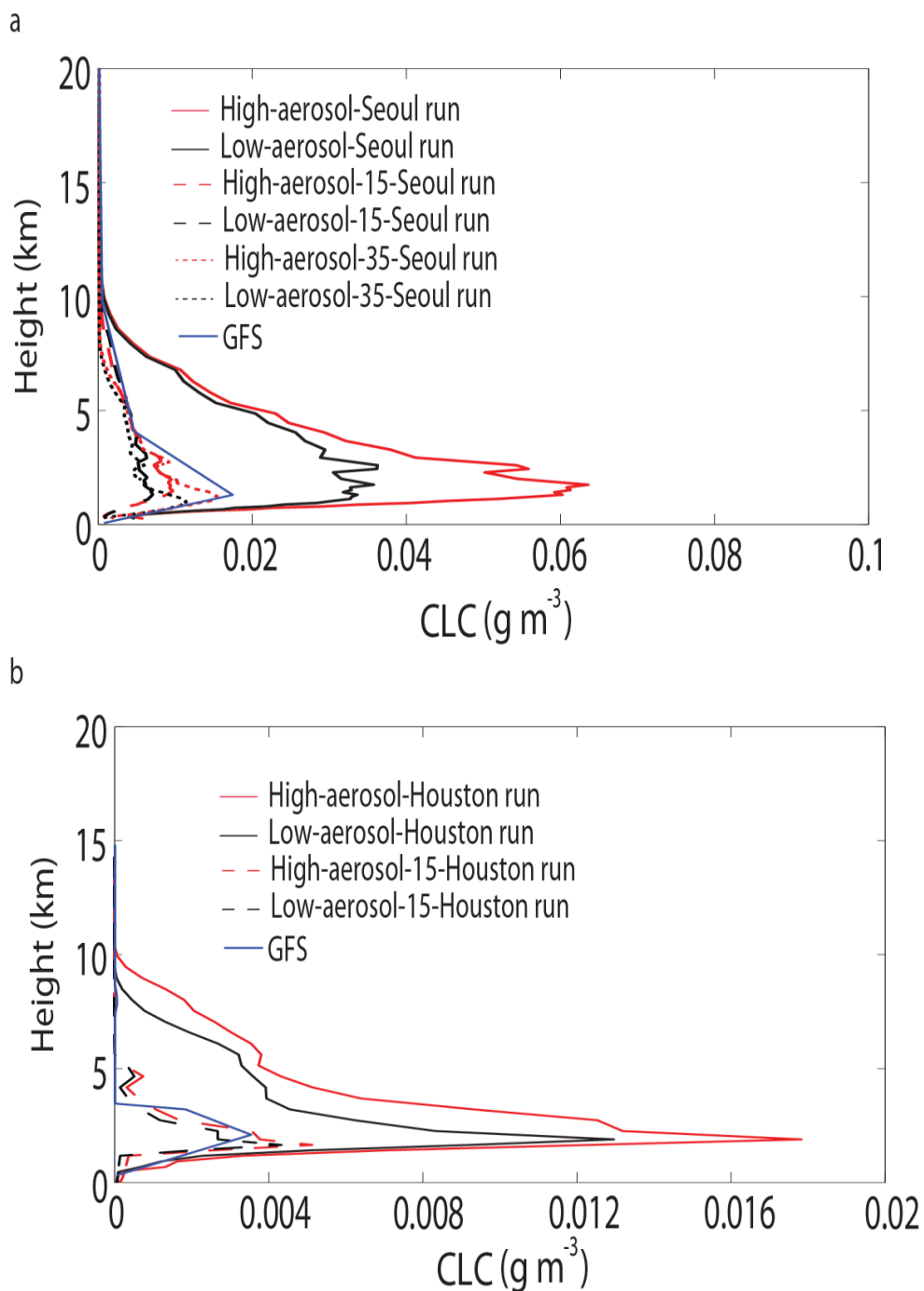
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**Figure 1**



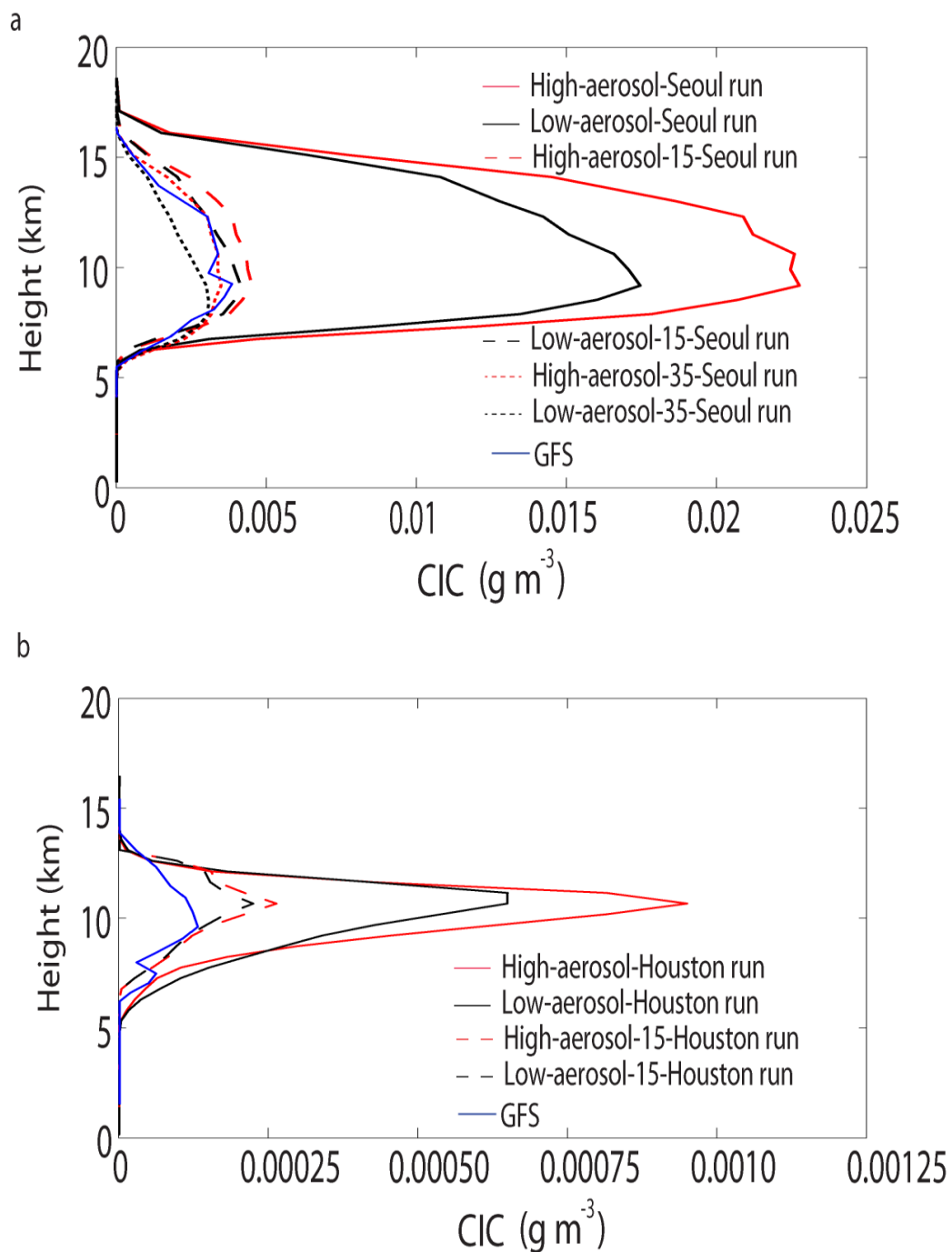
**Figure 2**



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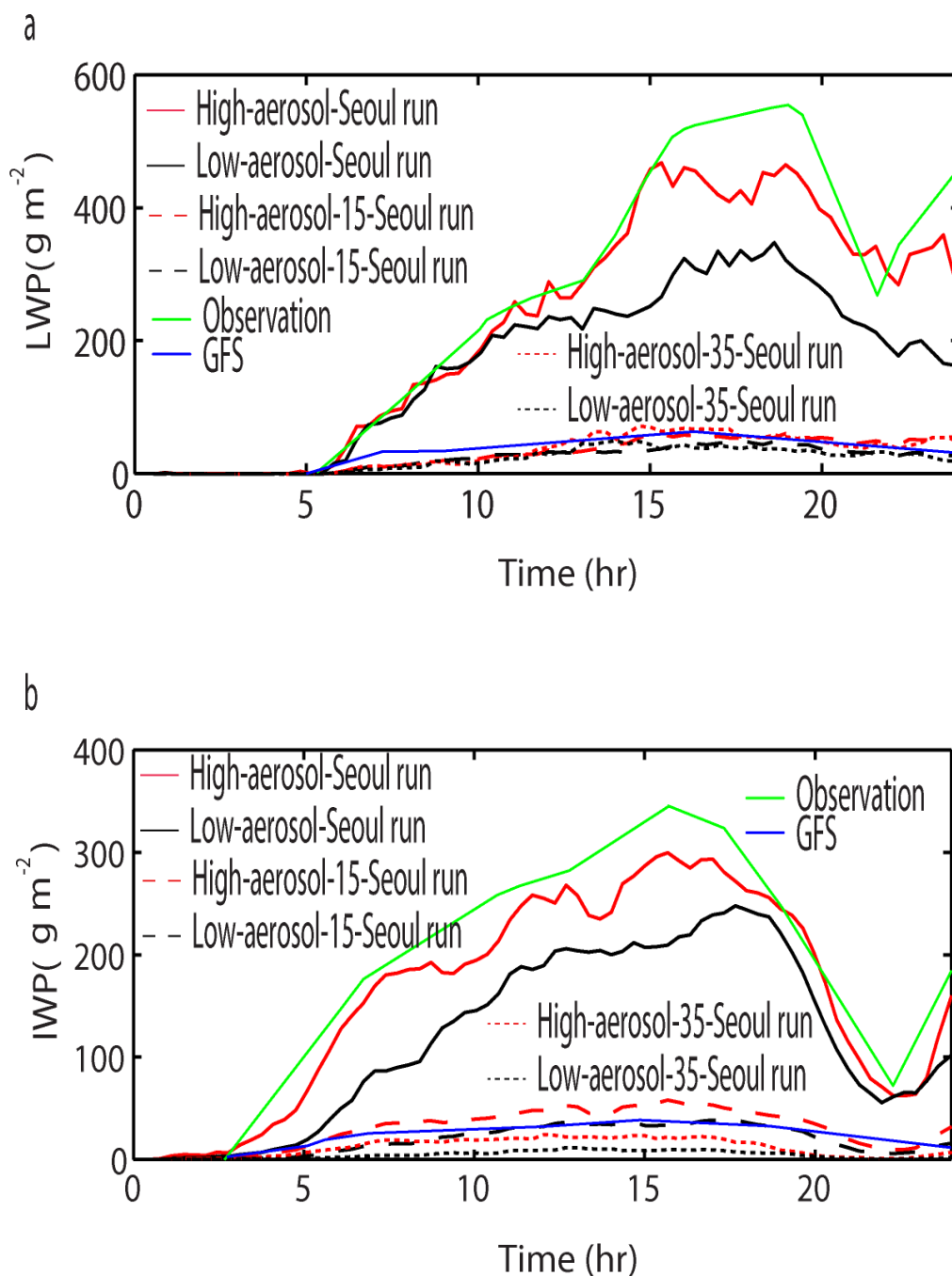
**Figure 3**



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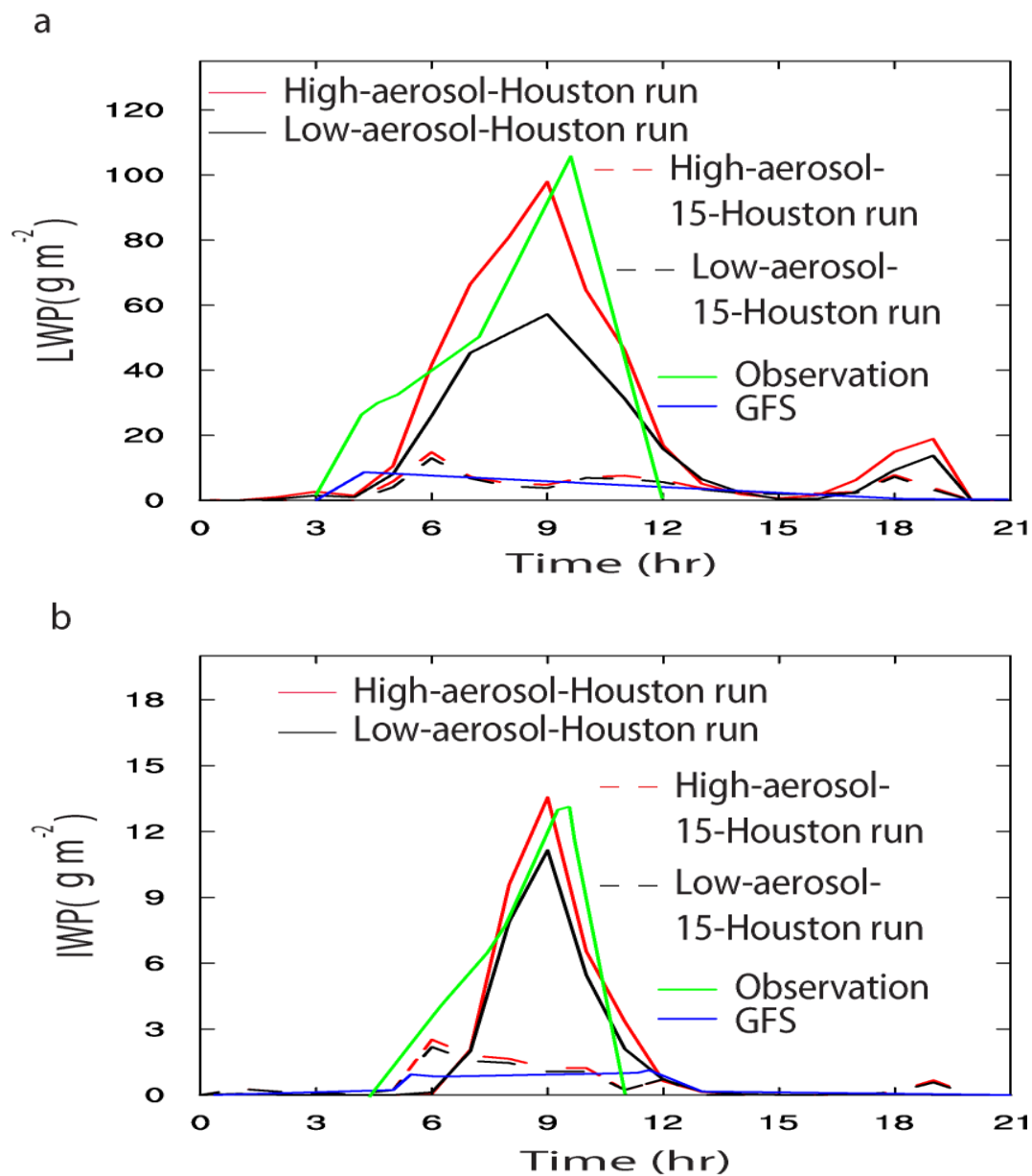
**Figure 4**



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**Figure 5**



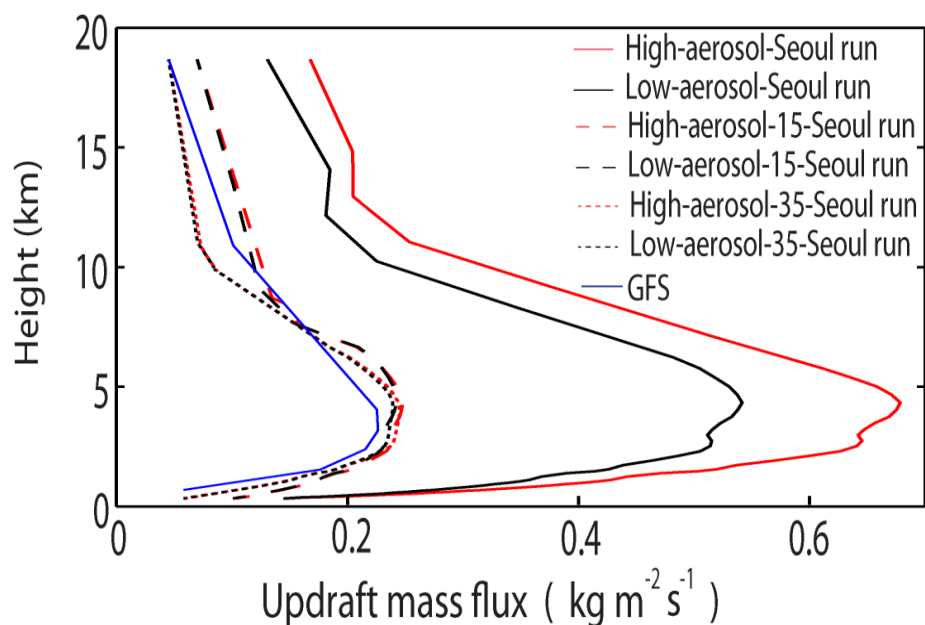
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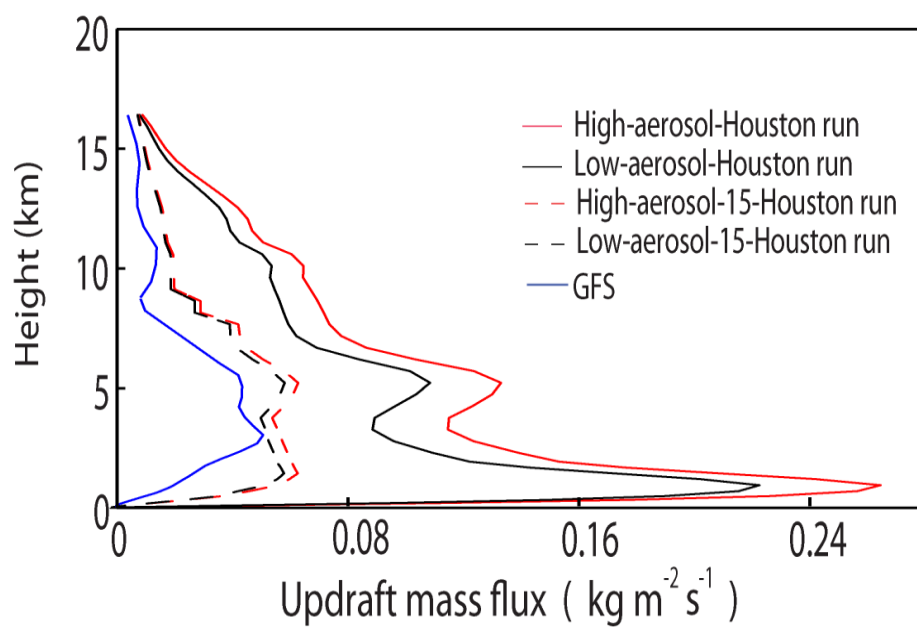
**Figure 6**



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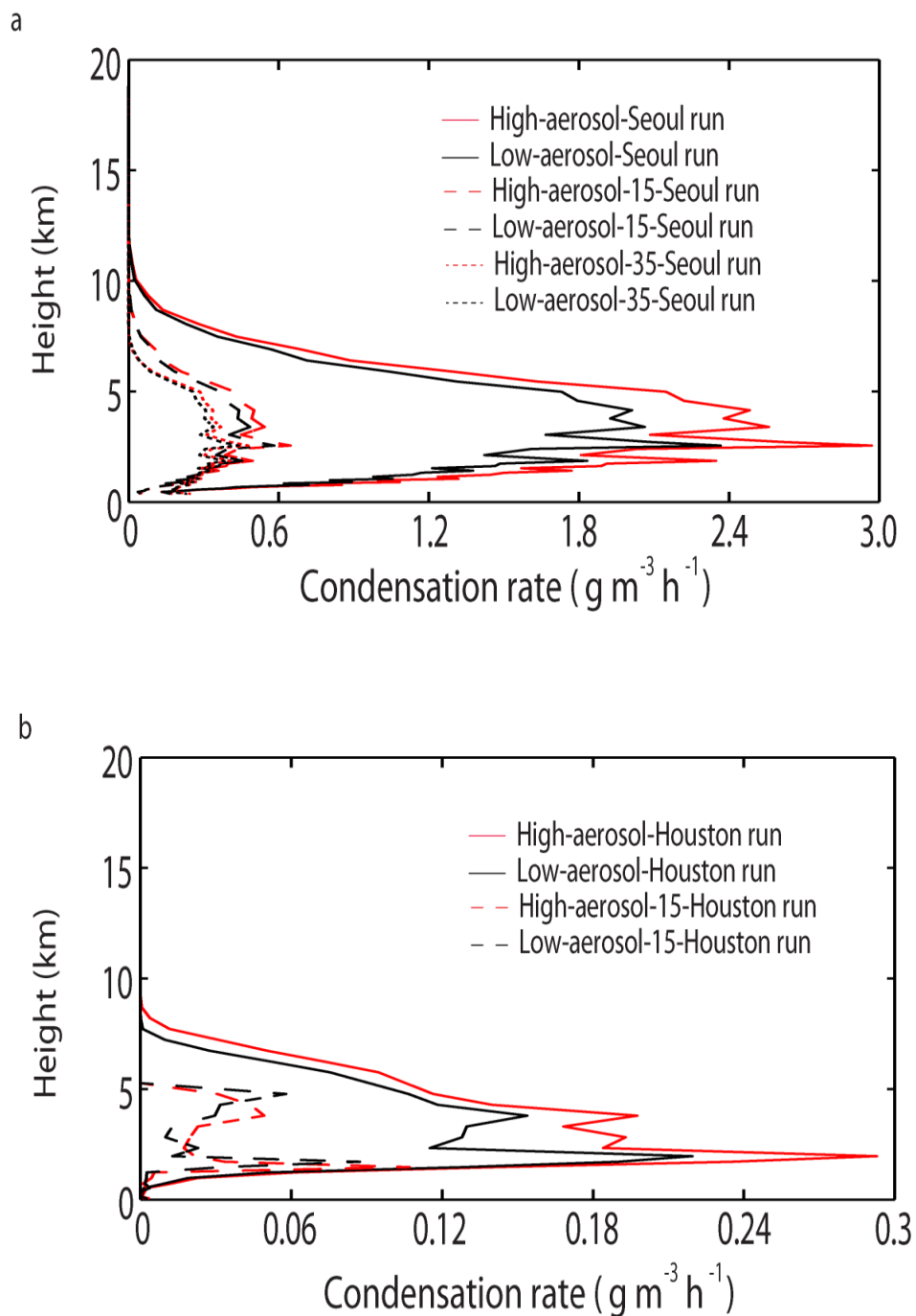


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**Figure 7**

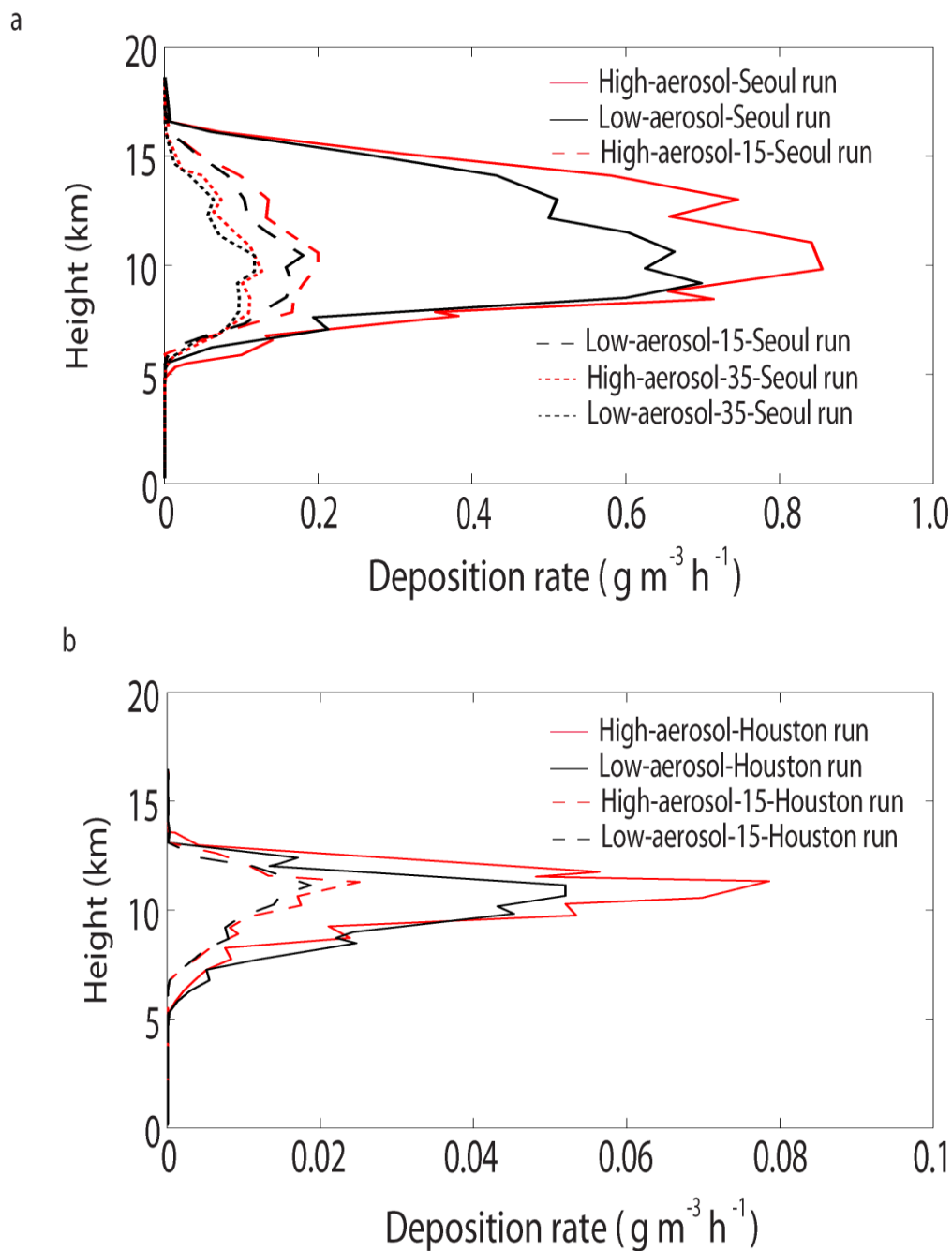




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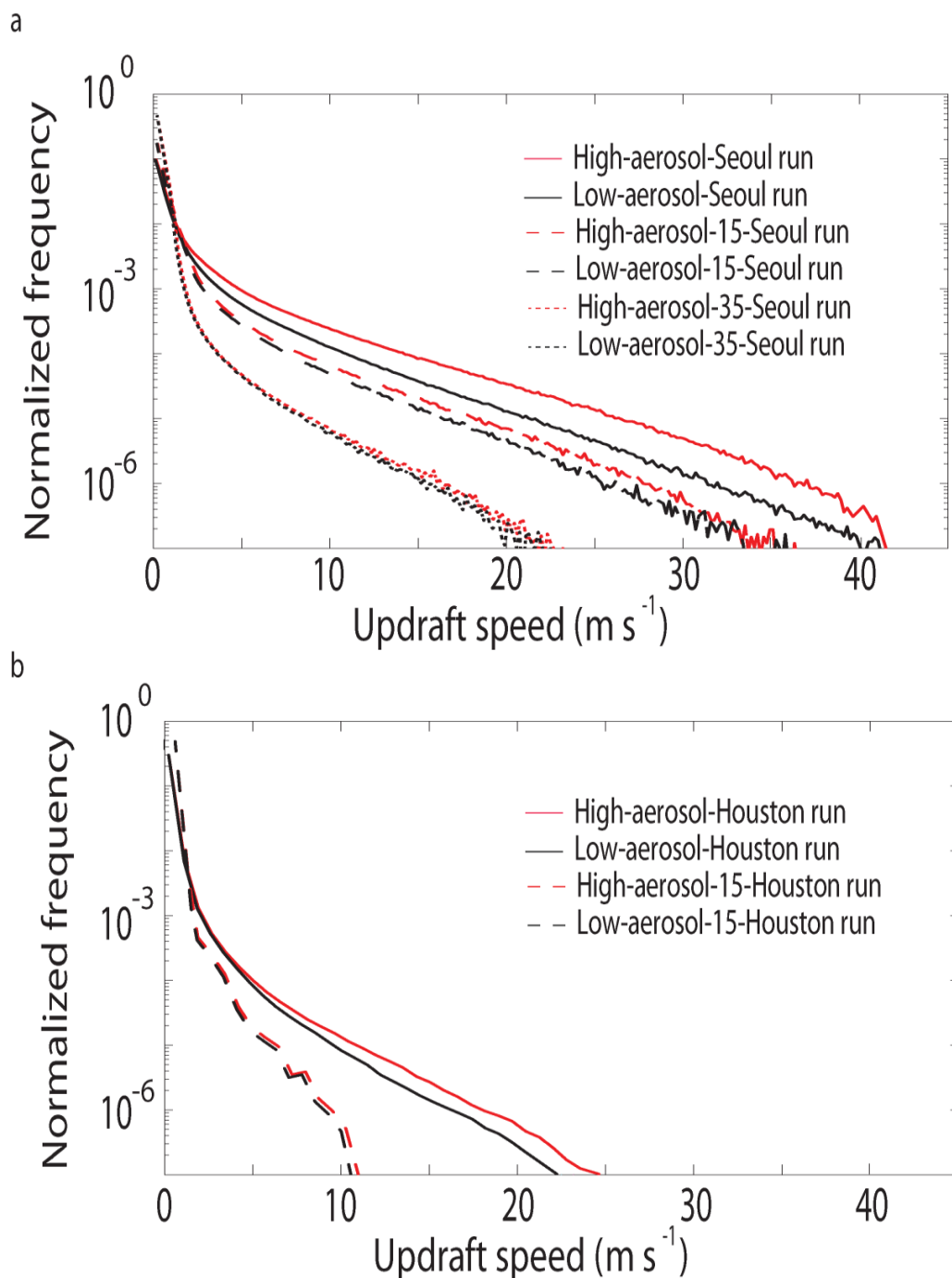
**Figure 8**



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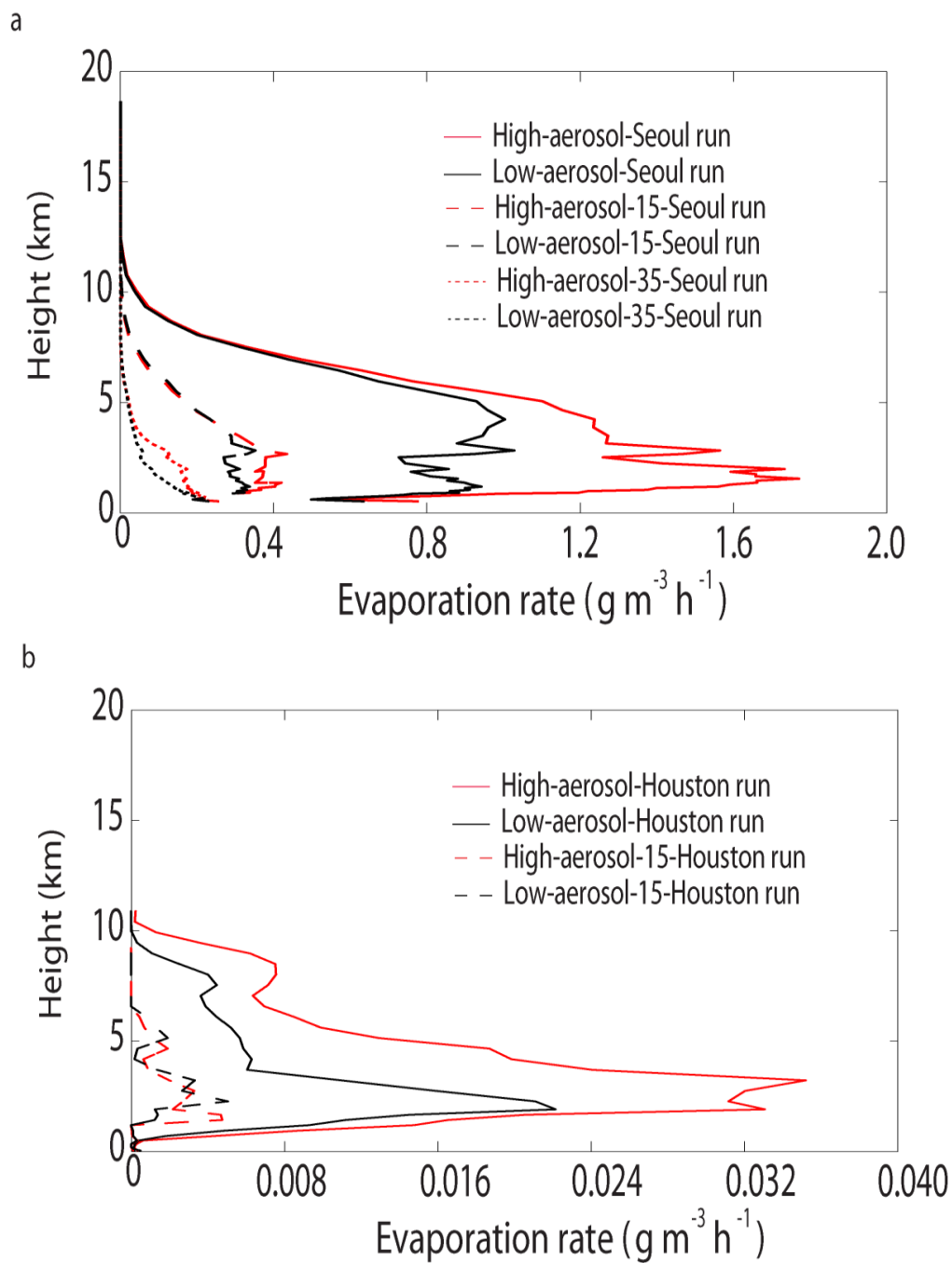
**Figure 9**



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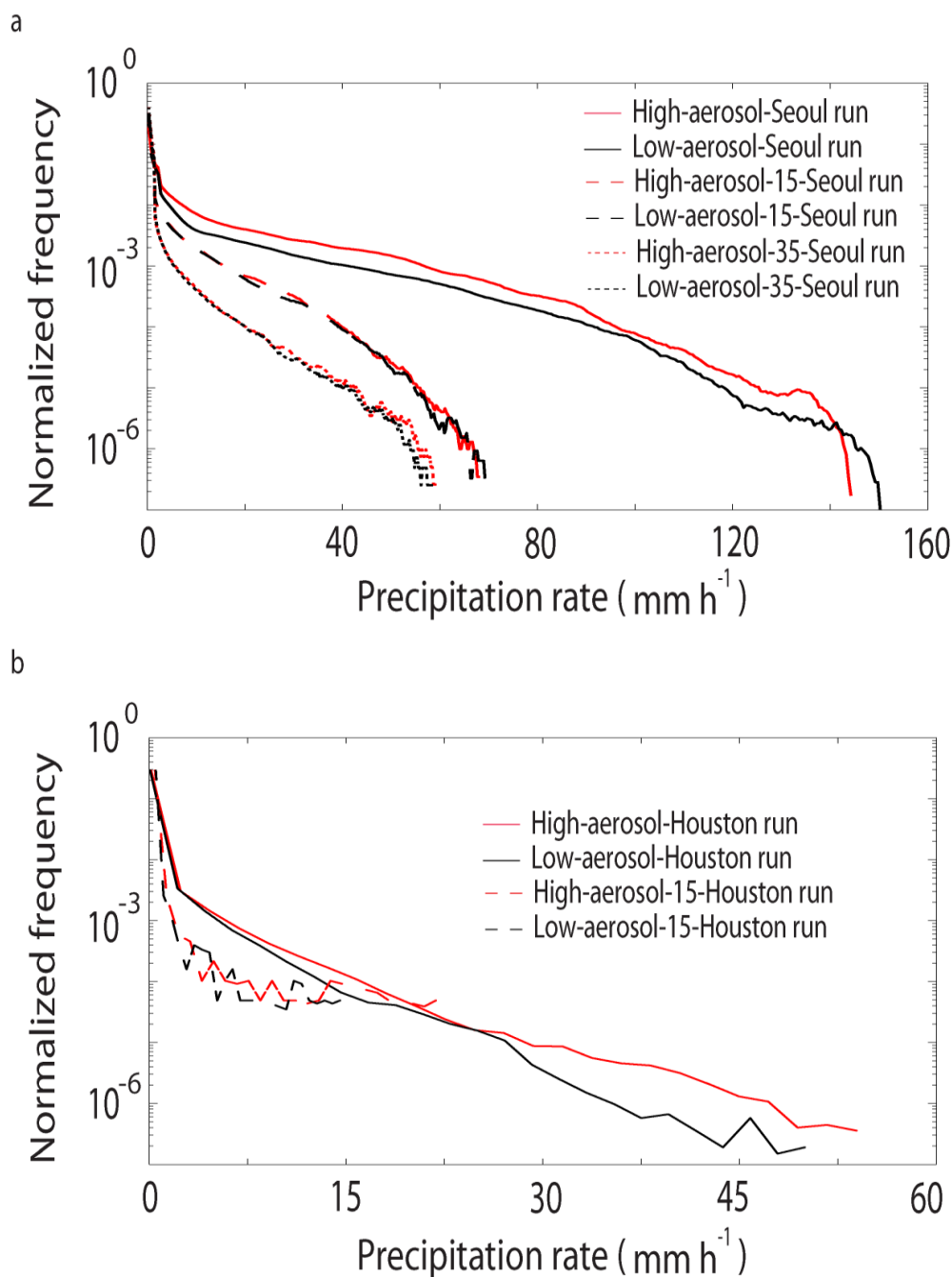
**Figure 10**



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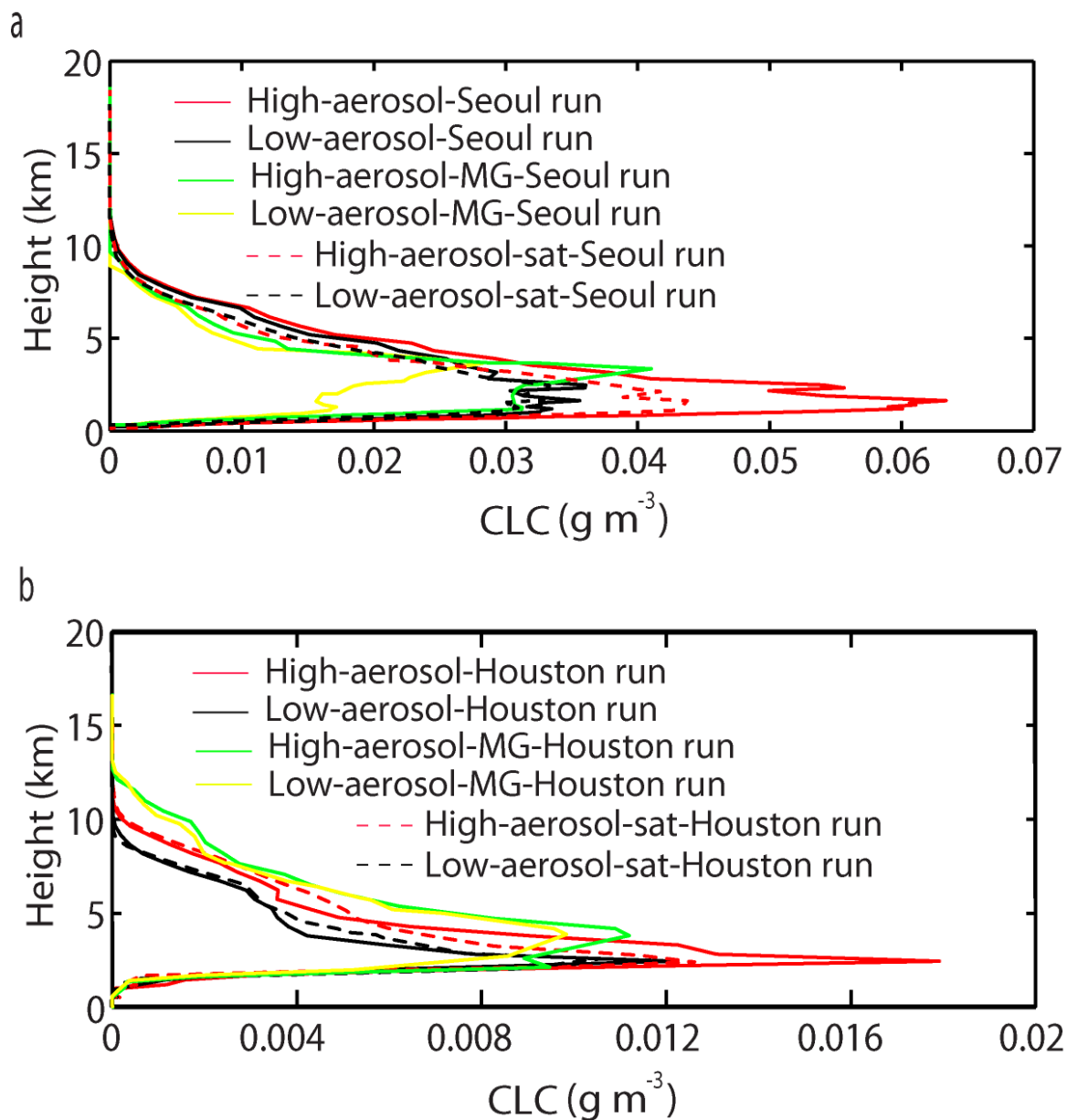
**Figure 11**



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**Figure 12**

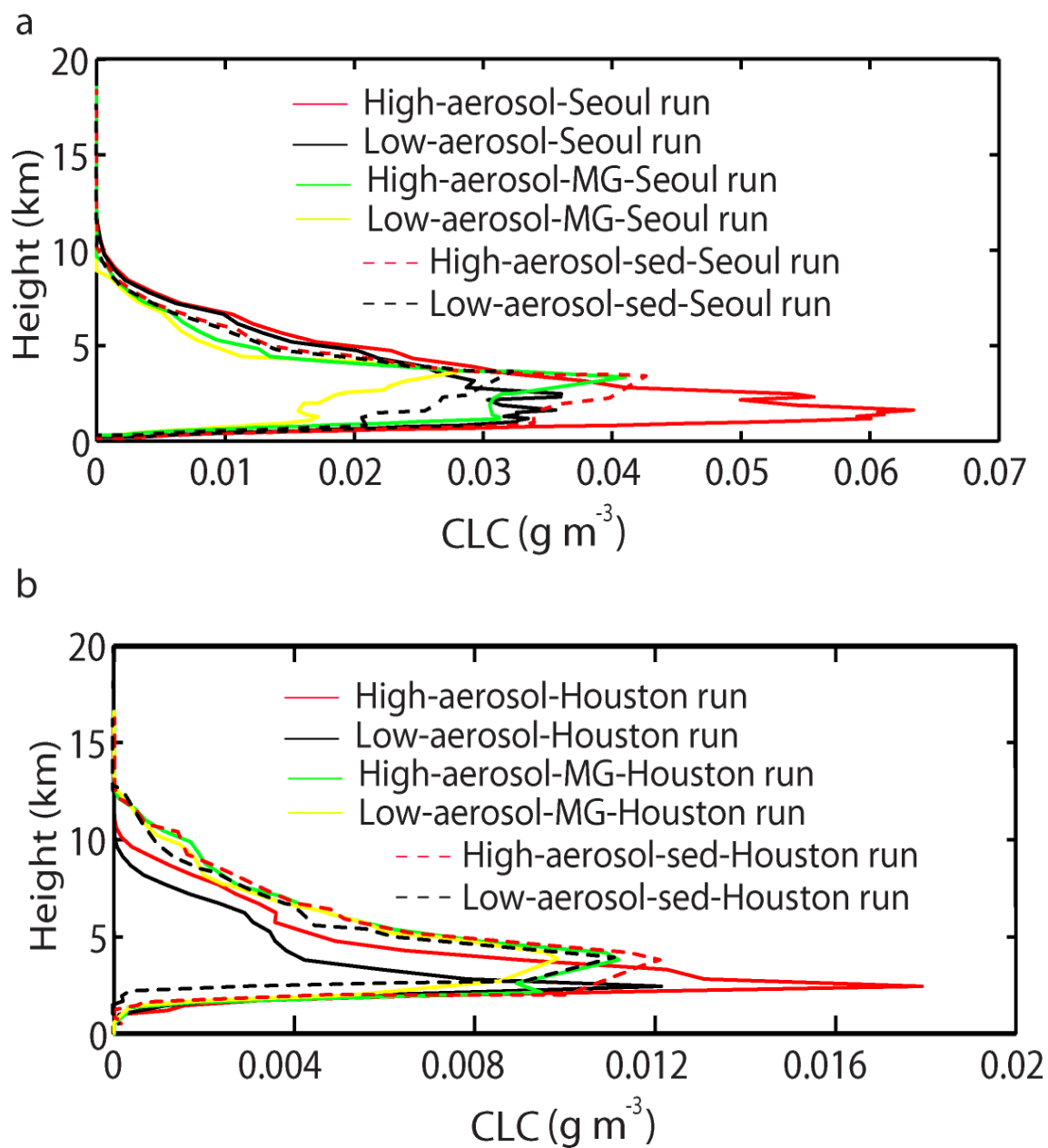


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**Figure 13**

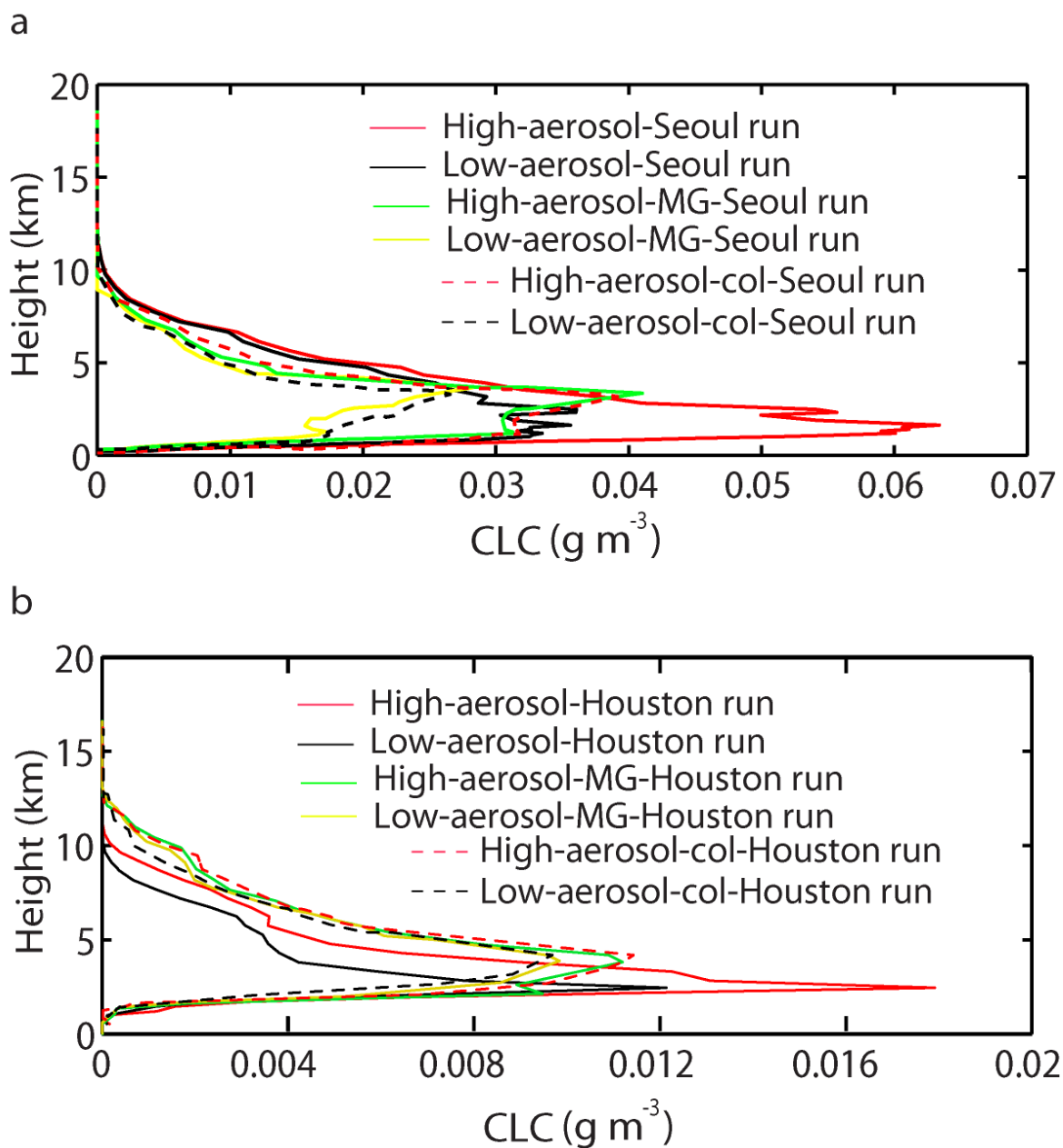


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**Figure 14**



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**Figure 15**