



1 Why do GCMs overestimate the aerosol cloud lifetime effect? A comparison of CAM5 2 and a CRM Cheng Zhou<sup>1</sup>, Joyce E. Penner<sup>1</sup> 3 4 (1){University of Michigan, Ann Arbor, MI, USA} 5 6 Corresponding author: C. Zhou (zhouc@umich.edu). 7 8 9 Abstract 10 Observation-based studies have shown that the aerosol cloud lifetime effect or the

11 increase of cloud liquid water (LWP) with increased aerosol loading may have been overestimated in climate models. Here, we simulate shallow warm clouds on 05/27/2011 at 12 13 the Southern Great Plains (SGP) measurement site established by Department of Energy's 14 Atmospheric Radiation Measurement (ARM) Program using a single column version of a 15 global climate model (CAM5.3) and a cloud resolving model (CRM). The LWP simulated by 16 CAM increases substantially with aerosol loading while that in the CRM does not. The 17 increase of LWP in CAM is caused by a large decrease of the autoconversion rate when cloud droplet number increases. In the CRM, the autoconversion rate is also reduced, but this is 18 19 offset or even outweighed by the increased evaporation of cloud droplets near cloud top, resulting in an overall decrease in LWP. Our results suggest that climate models need to 20 21 include the dependence of cloud top growth and the evaporation/condensation process on 22 cloud droplet number concentrations.

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# 24 **1. Introduction**

25 Traditionally aerosols have been thought to lengthen cloud lifetime (Albrecht, 1989) by 26 increasing droplet number and reducing droplet size thereby delaying and reducing the 27 formation of rain in clouds. These longer lived clouds would then increase cloud cover and 28 reflect more sunlight. Yet observational evidence for these lifetime effects is limited and 29 contradictory (Boucher et al. 2013). Observations of ship tracks show that marine boundary-30 layer clouds polluted by aerosol particles show that the liquid water path (LWP) can either 31 increase or decrease depending on factors like mesoscale cloud cellular structures, dryness of 32 the free troposphere and boundary layer depth (Christensen and Stephens 2011; Chen et al., 33 2012, 2015). Results from large-eddy simulations (LES) and cloud resolving models (CRM)





34 show the response of cloud water to aerosols is complicated by competing effects like 35 reduced precipitation formation efficiency in clouds and enhanced evaporation at cloud top or 36 in the downdraft regions of cloud edges (Ackerman et al. 2004; Xue and Feingold, 2006; Tao 37 et al., 2012). Since CRMs and LES models resolve clouds, have more complete physics and 38 depend less on subgrid parameterizations than general circulations models (GCMs), they are 39 often used together with field measurements to evaluate and improve parameterizations of 40 clouds and radiation used in climate models. Several previous studies have compared single 41 column models, which are essentially an isolated column of a GCM, and cloud resolving models (Moncreiff et al. 1997; Ghan et al., 2000; Xu et al., 2002; Xie et al., 2002; Xie et al., 42 43 2005). Lee and Penner (2010) extended these types of comparisons to the response of the two 44 models (CAM and a CRM) to increases in aerosols in marine stratocumulus. Both models 45 found that LWP increased but the effect from increased condensation dominated in the CRM 46 while the effect from decreased autoconversion dominated in CAM. Wang et al. (2012) used 47 satellite observations of the precipitation frequency susceptibility together with model 48 simulations to constrain cloud lifetime effects in GCMs. They show that GCMs tend to 49 overestimate the precipitation frequency susceptibility of marine clouds. Since the LWP 50 increase as a result of increased cloud condensation nuclei concentrations is highly correlated with precipitation frequency susceptibility in climate models, they surmise that the LWP 51 52 increase is too high and show that this overestimation could be "fixed" by reducing the 53 dependence of the autoconversion rate on cloud droplet number in the models.

54 In this study, we simulated continental shallow warm clouds observed on 05/27/2011 at 55 the Southern Great Plains (SGP) measurement site established by Department of Energy's 56 Atmospheric Radiation Measurement (ARM) Program using the single column version of a 57 global climate model (CAM5.3) and a cloud resolving model and explored plausible causes 58 for the differences in the response of these two models to increases in aerosols. Here we 59 specifically identify that the cloud top growth and turbulence mixing parameterizations 60 within CAM require improvement, rather than only the autoconversion rate. Section 2 61 describes the models and set-up. Section 3 presents results followed by conclusions and a 62 discussion in section 4.

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#### 2. Description of models and set-up

We used the Goddard Cumulus Ensemble model (GCE) with recent improvements (Tao et al. 2014) and the single column version of Community Atmosphere Model (CAM, version 5.3) which is the atmospheric component of the Community Earth System Model (CESM,





68 version 1.2.2). Readers are referred to Neale et al. (2012) for more model details of CAM. 69 Here we briefly summarize the two most critical parameterizations for warm stratus clouds in 70 CAM: cloud microphysics and cloud macrophysics. The cloud microphysics (version MG1.5) 71 is a two-moment scheme (Morrison et al. 2005, Morrison and Gettelman 2008) which 72 predicts the number concentrations and mixing ratios of cloud droplets. The source term for 73 the cloud droplets in warm clouds only includes the activation of cloud condensation nuclei 74 while the sink terms include the instantaneous evaporation of falling cloud droplets into the 75 clear portions of grids beneath clouds, autoconversion of cloud droplets to form rain, and 76 accretion of cloud droplets by rain. The first two sink terms (instantaneous evaporation of 77 falling cloud droplets and autoconversion) depend on the aerosol number concentration since 78 the terminal falling speed of cloud droplets is related to cloud droplet size and the autoconversion rate is inversely proportional to cloud droplet number ( $\sim N_c^{-1.79}$  where  $N_c$  is 79 80 the in-cloud cloud droplet number). The last sink term (accretion) does not depend on the 81 cloud droplet number (Khairoutdinov and Kogan 2000). The conversion of water vapor to 82 cloud condensate is computed by the cloud macrophysics parameterization which also 83 predicts the cloud fraction in each grid as well as the horizontal and vertical overlapping 84 structures of clouds. Following Smith (1990), the liquid fraction of stratus clouds in CAM5 is 85 derived from an assumed triangular distribution of total relative humidity (i.e. the sum of 86 water vapor and liquid cloud water). The net conversion rate of water vapor to stratus 87 condensate is diagnosed using saturation equilibrium conditions: (1) the RH over the water 88 within the liquid stratus is always 100%, and (2) no liquid stratus droplets exist in the clear 89 portion of the grid.

90 The Goddard Cumulus Ensemble model (GCE) is a CRM that has been developed and 91 improved at the NASA Goddard Space Flight Center (GSFC). Its development and main 92 features were published in Tao and Simpson (1993) and Tao et al. (2003) and recent 93 improvements and applications were presented in (Tao et al. 2014). The GCE model used in 94 the present paper uses the double moment version of the Colorado State University Regional 95 Atmospheric Modeling System (RAMS) bulk microphysics scheme (Saleeby and Cotton, 96 2004) which assumes a gamma-shaped particle size distribution for three species of liquid 97 (small and large cloud droplets and rain). The small cloud droplets range from 2 to 40 98 microns in diameter, and the large cloud droplets range from 40 to 80 microns. Collection of 99 cloud droplets is simulated using stochastic collection equation solutions, facilitated by bin-





emulating look-up tables. Readers are referred to Lee et al. (2009) and Tao et al. (2014) formore detailed descriptions of the model physics.

102 CAM has 30 vertical layers and a variable vertical resolution which depends on the surface pressure and the vertical temperature profile. In the case studied in this paper the 103 104 vertical resolution is roughly 100 meters near the surface and stretches to about 300 m at 2 105 km decreasing to 1 km at 10 km. The time step is 20 minutes. GCE has 128 grids in the two 106 horizontal directions and 144 vertical layers. The horizontal resolution is 50 m, so the domain 107 size is 6.4 km  $\times$  6.4 km. GCE also uses a stretched vertical resolution that varies from about 108 30 m near the surface to about 90 m at 2 km and further to ~200 m at 10 km. The time step of the GCE model is 1 second. Both models use the same initial conditions (surface 109 110 pressure/temperature, vertical temperature/water vapor/wind profiles), boundary conditions 111 (surface sensible/latent heat fluxes, surface pressure/temperature). Advective tendencies of 112 temperature and moisture (both vertically and horizontally) are specified based on an 113 objective variational analysis approach (Xie et al. 2014) fit to the Midlatitude Continental 114 Convective Clouds Experiment (MC3E) campaign observations which were conducted from April to June 2011 near the DOE ARM Southern Great Plains (SGP) site. The analyzed 115 advective tendencies cover the period from April 22<sup>nd</sup> to June 21<sup>th</sup>, 2011. Middle to deep 116 convective clouds were observed in most cloudy days. For this study, May 27<sup>th</sup>, 2011, was 117 118 selected because middle and high clouds were absent during a low cloud period observed 119 near noon. The vertical wind/temperature/moisture/cloud fraction profiles, surface 120 latent/sensible heat fluxes, and advective tendencies of temperature and moisture are shown 121 in Fig S1. Low clouds occurred from  $\sim$ 1 km to  $\sim$  2 km near the top of the boundary layer and 122 were strongly modulated by the advective tendencies of temperature and moisture. Positive 123 moisture flux and negative temperature flux were observed during the growing stage of the 124 clouds while negative moisture flux and positive temperature flux were observed during the 125 decaying stage. Both models are initialized at 00:00 local time and run for 18 hours.

126 To study the effect of aerosols on clouds, we scaled the aerosol vertical profiles in both models by increasing the surface aerosol number concentrations from 250 cm<sup>-3</sup> to 4000 cm<sup>-3</sup>. 127 128 GCE uses a prescribed aerosol profile which decreases linearly from its surface concentration to 100 cm<sup>-3</sup> at an altitude of 14 km and above. The activation of aerosols to cloud droplets is 129 130 based on the grid resolved vertical updraft velocity, temperature, and aerosol number and size 131 from a look-up table constructed from results of a Lagrangian parcel model (Saleeby and 132 Cotton, 2004). For CAM, we extracted the averaged aerosol profile in May at this location 133 from a 5-year run of CAM5 using the MAM3 aerosol module and scaled the aerosol profile





134 based on the surface aerosol number concentrations (see Fig. S2 for profiles of aerosol 135 number concentrations used in the two models). The activation of aerosols into cloud droplets 136 in CAM is diagnosed as a function of the modeled subgrid-scale updraft velocity and aerosol 137 compositions/sizes/numbers (Abdul-Razzak and Ghan 2000). Even though we set the total 138 surface aerosol number concentrations the same in the two models, the aerosol composition, 139 size, and number at cloud level, and the nucleation schemes are inherently different. However, 140 since this paper focuses on a sensitivity study which is aimed at revealing the different cloud physical representations in the two models that lead to opposite responses of the simulated 141 LWP to increasing aerosol number concentrations that cover a wide range (250 cm<sup>-3</sup> to 4000 142 cm<sup>-3</sup>) rather than quantifying the changes of the LWP, these differences are not critical to the 143 144 conclusions of the paper. To better isolate differences in the aerosol indirect effect in the two 145 models, we also turned off the aerosol direct radiative effect.

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#### 147 **3. Results**

Figure 1a shows the observed cloud fractions from the early morning to the late afternoon 148 on May 27<sup>th</sup>, 2011 at the SGP site, while Figures 1b and 1c show the simulated mean cloud 149 water content from the two models assuming a surface aerosol number concentration of 500 150 151  $cm^{-3}$ . Compared to the observations, the simulated clouds from both models begin later in the 152 day and have a smaller vertical coverage. But the models compare relatively well to each 153 other which suggests that differences between the models and the observations may largely 154 be caused by the possible errors/uncertainties associated with the derived initial conditions or 155 advective tendencies. Nevertheless, we can see that the GCE model captures the observed 156 growth of the clouds with height while CAM does not. A detailed analysis of the GCE (next 157 paragraph) shows that the clouds could be loosely classified as stratocumulus which occur 158 near the top of the planetary boundary layer (PBL) and are mainly driven by long wave radiative cooling offset by short wave radiative heating. This is corroborated by CAM's 159 160 result which shows all simulated clouds are stratus clouds and no convective clouds are able 161 to form above the PBL. The advective tendencies of heat and moisture also strongly modulate 162 the clouds. For example, the positive moisture tendency before 14:00 hours leads to slightly 163 larger in-cloud water vapor mixing ratio than that below the clouds (more details will be 164 presented in the discussion of Figure 2). Figure 1d and 1e show the domain averaged liquid 165 water path (LWP) from the two models for five different surface aerosol number concentrations (250, 500, 1000, 2000 and 4000 cm<sup>-3</sup>). Both models underestimate the LWP 166 167 during the day, similar to their underestimation of cloud cover. GCE shows relatively small





168 changes in the LWP when using different surface aerosol numbers. The LWP slightly 169 increases with the increasing aerosol number before ~14:00 but starts to decrease with the 170 increasing aerosol number when the clouds start to decay after around 14:00. On the other 171 hand, the LWP from CAM increases substantially and consistently with increasing aerosol 172 number and matches the observed LWP better when the surface aerosol number is equal to 4000 cm<sup>-3</sup>. As noted earlier, due to uncertainties associated with the derived forcing data as 173 well as uncertainties in the models, this should not be interpreted as proof that CAM 174 175 represents the physics better. Figure 1f and 1g show the precipitation rates from the two 176 models. The precipitation rate from CAM consistently decreases with increasing aerosol 177 number and is nearly suppressed after 13:00. The change is most prominent when the aerosol number is increased from 250 to 500 cm<sup>-3</sup>. The precipitation rates from GCE are overall very 178 179 small with maximum values less than 0.08 mm day<sup>-1</sup>. The change in precipitation for GCE 180 with increasing aerosol numbers is a little more complex. During the growing phase of the 181 clouds, as in CAM, the precipitation rate decreases. But during the decaying phase, the 182 precipitation rate actually increases even though the LWP decreases.

183 Figures 2a-2c show the domain averaged potential temperatures ( $\theta$ ), total water specific humidity  $(q_t)$  and cloud water content  $(q_c)$  at three times (13:00, 14:00 and 15:00) from the 184 case with surface aerosol numbers equal to 250 cm<sup>-3</sup> (dash-dotted curves) and 1000 cm<sup>-3</sup> 185 (solid curves), respectively.  $q_t$  is the sum of  $q_c$ , rain and water vapor mixing ratios, which is 186 187 an invariant within the PBL for stable non-precipitating well-mixed stratocumulus.  $\theta$  and  $q_t$ 188 from the two cases almost overlap except near the cloud top at 14:00 and 15:00. Fig. 2a 189 shows the growth of the PBL. At 13:00 the clouds do not completely reside within the PBL as 190 the top of the PBL is at about 1.2 km which is lower than the cloud top height (~1.5 km) 191 shown in Fig. 2c. Fig. 2b shows that  $q_t$  in the top half of the cloud (from ~1.2-1.5 km) is 192 larger than  $q_t$  in the bottom half of clouds (from ~1-1.2 km) and  $q_t$  below the clouds at 13:00. 193 This suggests that the top half of the clouds are not fully coupled with the surface and the 194 cloud water in the top half of the clouds is strongly affected by the horizontally advected 195 positive moisture flux. At 14:00 and 15:00, the advected moisture flux becomes negative and 196 the PBL is high enough that the clouds reside fully within the top of the PBL and possess the 197 characteristics of well-mixed stratocumulus. The domain averaged long-wave cooling rate at the cloud top height is about 2 K hr<sup>-1</sup> and is offset by a short-wave heating of about 0.5 K hr<sup>-1</sup>. 198 199 Fig. 2c shows that the cloud top is a little higher for the higher aerosol case, but the maximum 200 values of  $q_c$  are smaller. A closer look at  $\theta$  in Fig 2a also shows that the top of the PBL which





201 is near 1.5 km is higher and colder in the higher aerosol number case. These differences of 202  $q_c$  and  $\theta$  between the two cases are clearer in an enlarged portion of Fig 2a and 2b shown in 203 Fig. S3. The potential temperature in the sub-cloud layer at 14:00 and 15:00 is also slightly 204 higher (about 0.005 K) for higher aerosols. Figs. 2d to 2i show the time-averaged profiles of 205 q<sub>c</sub> and the net result of condensation and evaporation (Conden-Evap) during two 1-hour 206 intervals (Fig. 2d-f for 13:00 to 14:00 and Fig. 2g-i for 14:00 to 15:00) representing the 207 growing and decaying phases of the cloud, respectively. Figures 2e and 2h show that a net 208 evaporation occurs just below the cloud base and near the cloud top. The largest net 209 condensation is located near the cloud base. The most obvious change between the growing 210 phase and decaying phase of the cloud is the increased evaporation near the cloud top, 211 especially for the high aerosol number case (see the changes from blue curve to the red curve 212 at around 1.5 km from Fig. 2e and Fig. 2h). Choosing (Conden – Evap)/ $q_c$  as a measure of the inverse of the characteristic evaporation time of cloud droplets, Figures 2f and 2i show 213 that it increases substantially from 300  $hr^{-1}$  to about 600  $hr^{-1}$  (an evaporation time of ~6 214 215 seconds) near the cloud top for the higher aerosol number case.

216 Figure 3 shows the LWP and the column integrated LWP source and sink terms from the low and high aerosol cases (250 and 1000 cm<sup>-3</sup>). The source term for LWP only includes the 217 218 net condensation term (Conden - Evap) while the loss terms include autoconversion and 219 accretion. Since CAM includes a separate autoconversion and accretion terms while GCE 220 does not, we combined autoconversion and accretion as one term (Auto+Accre) for easier 221 comparison. As shown in Fig. 1, when we increase the aerosol numbers from 250 to 1000 cm<sup>-</sup> 222 <sup>3</sup>, the LWP increase is relatively small in GCE and substantially larger in CAM. Both models 223 show decreased Auto+Accre which acts to increase the LWP. This is expected as increased 224 aerosol numbers increase the cloud droplet number which decreases the autoconversion rate. 225 But CAM shows much larger changes, especially before 13:00 hours. This is likely due to the 226 fact that the two models use different cloud droplet activation schemes as well as schemes to 227 parameterize the autoconversion and accretion processes. Moreover, in GCE, the decreased 228 autoconversion is largely offset or even outweighed by increased evaporation. As shown in 229 Fig. 2e and 2h the increased evaporation occurs near cloud top. The increased evaporation 230 near the cloud top and the higher PBL suggests that higher aerosol number concentrations 231 lead to smaller cloud droplet sizes and enhanced evaporation at the cloud top which can then 232 decrease the temperature slope near the cloud top and promote the sinking of entrained air 233 into the cloud layer, a point made previously by Bretherton et al. (2007). This evaporation-





234 entrainment feedback mechanism was also observed in small cumulus clouds (Small et al. 235 2009). Before  $\sim 14:00$ , the effect from the decreased autoconversion rates outweighs the 236 effect from increased evaporation so that the LWP shows a slight increase. But as the cloud starts to decay after ~14:00, the PBL keeps growing and the enhanced 237 238 evaporation/entrainment rates accelerate the decaying process. Thus the LWP decreases 239 faster and eventually a smaller LWP results over the decaying period for the high aerosol 240 case. In the CAM model, the change of the net condensation term (Conden - Evap) is smaller 241 than that in the CRM model. Since the simulated cloud top remains unchanged between 242 12:00 and 15:00 hours, the drying effect seen in the CRM due to enhanced entrainment of 243 overlying dry air is not present. This is likely due to the fact that the moist turbulence scheme 244 in CAM does not depend on the cloud droplet number/size and the condensation and 245 evaporation in the CAM's macrophysics scheme is not linked to the cloud droplet number or 246 size. Even though the instantaneous evaporation of falling cloud droplets into the clear 247 portions of grids beneath clouds in the microphysics scheme is related to the cloud droplet 248 number, it is about one order of magnitude smaller than the net condensation term in the 249 macrophysics scheme. Consequently the net condensation and evaporation is less sensitive to 250 the change in aerosol number and the effect from the decreased autoconversion rate 251 dominates the condensate loss, leading to an increase of the LWP.

252 To confirm that the effect from enhanced entrainment at the cloud top is the critical 253 reason for the reduced LWP change in GCE, we ran a sensitivity test to reduce the cloud top 254 mixing by increasing the grid spacing from 50 m to 100 km. With this larger grid spacing, we 255 greatly reduced the overshooting at the cloud top by reducing the maximum vertical speed in 256 the updrafts from meters per second to a few centimeters per second. As a result, the 257 enhanced entrainment effect was reduced and the microphysical effect from the reduced 258 autoconversion rate dominated. Figure 4 shows that the LWP from GCE decreases by about 259 5% for the dx=50 m case while it increases by about 12% for the dx=100 km case when the surface aerosol number is increased from 250 cm<sup>-3</sup> to 4000 cm<sup>-3</sup>. We also ran two more tests 260 261 to explore whether the LWP sensitivity in CAM could match that in the GCE. In the default 262 set-up of CAM, the autoconversion rate is inversely proportional to cloud droplet number  $(\sim N_c^{-1.79}$  where N<sub>c</sub> is the in-cloud cloud droplet number). We ran two cases, auto06 and 263 auto00, each with a reduced dependence of the autoconversion rate on the cloud droplet 264 number. In case auto06, the autoconversion rate is proportional to  $N_c^{-0.60}$  and in case auto00, 265 the autoconversion rate does not depend on the cloud droplet number. The autoconversion 266





267 rate is scaled in both cases to produce the same rate as that from the default case at a droplet number concentration of 100 cm<sup>-3</sup>. As shown in Fig. 4, the LWP from the default case is 268 more than doubled when the surface aerosol number is increased from  $250 \text{ cm}^{-3}$  to  $4000 \text{ cm}^{-3}$ 269 270 while the LWP from auto06 only increases by  $\sim$ 50% and the LWP from case auto00 remains 271 almost unchanged. These results suggest that the dependence of the autoconversion rate on 272 the cloud droplet number can play a determining role on the simulated LWP consistent with 273 the findings of precipitation frequency susceptibility in Wang et al. (2012). However, this 274 adjustment is unable to simulate decreases in LWP seen in the GCE model.

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## 276 4. Conclusion and Discussion

We simulated shallow warm clouds on May 27th, 2011 at the DOE ARM SGP site with a 277 cloud resolving model (Goddard Cumulus Ensemble model) and a single column model 278 279 (CAM) using the same initial/boundary conditions and advected moisture/heat tendencies 280 derived from the MC3E campaign data. The liquid water path (LWP) simulated by CAM shows a large dependence on the aerosol loading and is more than doubled when the surface 281 aerosol number is increased from 250 cm<sup>-3</sup> to 4000 cm<sup>-3</sup> while the LWP simulated by the 282 CRM decreases by ~5%. The high sensitivity of LWP on aerosol loading in CAM can be 283 284 reduced by reducing the dependence of the autoconversion rate on the cloud droplet number 285 concentration, but is unable to reproduce the decrease in LWP seen in the CRM. Whereas 286 Wang et al. (2012) concluded that this term in GCM models can be tuned to fit observations 287 of the precipitation frequency susceptibility, we find that the poor representation of 288 entrainment and droplet evaporation in CAM model may be the fundamental cause of 289 differences with the more complete CRM. While in the CRM a reduced autoconversion rate 290 is also observed with increased aerosol loading, it is offset or even outweighed by the 291 increased evaporation of cloud droplets near the cloud top. The increased evaporation cools 292 the cloud top, reduces the temperature lapse rate and thus increases the entrainment of drier 293 air above the cloud top and accelerates the decaying process of the clouds. Reduced LWP 294 through enhanced entrainment with increased aerosol number has also been reported in 295 previous literature using large eddy simulations (e.g., Ackerman 2004, Bretherton et al. 2007, 296 Seifert et al. 2015). One unique aspect of the present paper is that the response of the LWP 297 over the lifetime of the cloud is negative in the CRM while it is positive in the CAM model 298 for the same forcing conditions. One critical deficiency of CAM for this case is that the effect 299 from increased mixing of drier air from above the cloud layer through enhanced entrainment 300 caused by increased aerosol numbers is missing. First, CAM is not able to simulate the





301 growth of the cloud top due to its coarse vertical resolution. However, even if the CAM 302 vertical resolution were high enough to capture the growth of the cloud top, since the moist 303 turbulence scheme and the evaporation of cloud condensate in the cloud macrophysics 304 parameterization at the cloud top are not related to the cloud droplet number, aerosol number 305 will not have a direct impact on the cloud top mixing or the LWP.

306 Our CRM model results demonstrate that the relative importance of the decreased 307 autoconversion rate effect and the enhanced entrainment effect from increased aerosol 308 numbers can change based on environmental conditions as manifested in different stages 309 during the cloud lifecycle. Thus, one may need to distinguish the cloud stage when studying 310 the aerosol lifetime effect either with a model or from observations.

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- **Figure 1.** Observed cloud fractions on May 27<sup>th</sup>, 2011 at the SGP site (a); domain
- 405 averaged cloud water content from the GCE model (b) and the single column version of
- 406 CAM (c) for the case assuming a surface aerosol number of  $500 \text{ cm}^{-3}$ ; liquid water path
- 407 and surface precipitation rates from GCE (d, f) and CAM (e, g) with varying surface
- 408 aerosol number concentrations.







409 Figure 2. (a-c) Domain averaged potential temperatures ( $\theta$ ), total water specific humidity 410  $(q_t)$  and cloud water content  $(q_c)$  at three times (13:00, 14:00 and 15:00) from two GCE 411 cases with surface aerosol numbers equal to 250 cm<sup>-3</sup> (dash-dotted curves) and 1000 cm<sup>-3</sup> 412 (solid curves). (d-f) Averaged profiles of  $\boldsymbol{q}_{c}$  , net results of condensation and evaporation 413 (Conden-Evap), and (Conden-Evap)/ $q_c$  for the 1-hour interval from 13:00 to 14:00 from 414 the two CRM cases with surface aerosol numbers equal to 250 cm<sup>-3</sup> (blue dash-dotted 415 curves) and 1000 cm<sup>-3</sup> (solid red curves). (g-i) Same as (d-f) except for the 1-hour interval 416 417 from 14:00 to 15:00.







418 Local time (hour)
419 Figure 3. LWP and the column integrated LWP source and sink terms from the case with
420 surface aerosol number concentration equal to 250 cm<sup>-3</sup> (thick lines) and 1000 cm<sup>-3</sup> (thin
421 lines) for (a) GCE and (b) CAM.

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Figure 4. Normalized LWP as a function of surface aerosol concentration in CAM (red
curves) and GCE (blue curves). A case for CAM using an autoconversion rate proportional
to N<sub>d</sub><sup>-0.6</sup> (CAM, auto06) as well as a case in which autoconversion is independent of N<sub>d</sub>
(CAM, auto00) is shown. The GCE model was run with a horizontal grid resolution of 50
m (default case) and 100 km.