



1 **The influence of multiple groups of biological ice nucleating particles on microphysical**
2 **properties of mixed-phase clouds observed during MC3E**

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31 **Abstract:**

32 A new empirical parameterization (EP) for multiple groups of primary biological aerosol
33 particles (PBAPs) is implemented in the aerosol cloud model (AC) to investigate their roles
34 as ice-nucleating particles (INPs). The EP describes the heterogeneous ice nucleation by (1)
35 fungal spores, (2) bacteria, (3) pollen, (4) detritus of plants, animals, and viruses, and (5)
36 algae. Each group includes fragments from the originally emitted particles. A high-resolution
37 simulation of a midlatitude mesoscale squall line by AC is validated against airborne and
38 ground observations.

39

40 Sensitivity tests are carried out by varying the initial vertical profiles of the loadings of
41 individual PBAP groups. The resulting changes in warm and ice microphysical parameters
42 are investigated. Overall, PBAPs have little effect on the ice phase, especially in the
43 convective region. In the stratiform region, increasing the initial PBAP loadings by a factor of
44 100 resulted in less than 60% change in ice number concentrations. The total ice
45 concentration is mostly controlled by various mechanisms of secondary ice production (SIP).
46 However, when SIP is artificially prohibited in sensitivity tests, increasing the PBAP loading
47 by a factor of 100 has no significant effect on the ice phase. Further sensitivity tests revealed
48 that PBAPs have little effect on surface precipitation as well as on shortwave and longwave
49 flux.

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54 **1. Introduction**

55 In most climate models, the largest source of uncertainty for estimating the total
56 anthropogenic forcing is associated with cloud-aerosol interactions (Forster et al., 2007).
57 Atmospheric aerosol particles can act as cloud condensation nuclei (CCN) and a few of them
58 act as ice-nucleating particles (INPs), thereby influencing the microphysical properties of
59 clouds and, depending on the cloud-type, potentially affecting precipitation formation. This
60 control of average cloud particle sizes regulates the radiative properties of layer clouds
61 produced in convection possibly influencing the atmospheric radiation budget. Various
62 sources of aerosol particles, including dust/metallic, marine aerosols, anthropogenic
63 carbonaceous emissions, and primary biological aerosol particles (PBAPs), contribute to the
64 observed INPs.

65

66 The formation of most precipitation globally is associated with the ice phase in cold clouds
67 (Heymsfield and Field 2015). In particular, mixed-phase clouds are vital for the global
68 climate. In a multimodel simulation study, Tsushima et al. (2006) showed that the doubling
69 of CO₂ concentrations caused the changes in the distribution of cloud-water in the mixed-
70 phase clouds in a climate simulation to be significant. Mixed-phase clouds play an important
71 role (about -3.4 Wm^{-2}) in the global net cloud radiative forcing of the present-day climate
72 (e.g., Matus and L'Ecuyer 2017).

73

74 PBAPs are solid particles containing insoluble material of biological origin and are emitted
75 from the Earth's surface. They are highly active in initiating ice as INPs and include bacteria,
76 fungal spores, pollen, algae, lichens, archaea, viruses, and biological fragments (e.g., leaf
77 litters, insects) and molecules (e.g., proteins, polysaccharides, lipids). Considering the onset



78 temperature of freezing, some ice nucleation active fungi and bacteria (especially
79 *Pseudomonas syringae*) are among the most active INPs present in the atmosphere (Després
80 et al. 2012; Hoose and Möhler 2012). The potential impact of PBAP INPs on cloud
81 microphysical characteristics has been recognized for many years. However, this topic
82 remains a subject of debate. Some previous modeling studies have shown that on a global
83 scale PBAPs have only a limited influence on clouds and precipitation (Hoose et al. 2010;
84 Sesartic et al. 2012, 2013; Spracklen and Heald 2014). On a global scale, the percentage
85 contribution of PBAPs to the immersion freezing is predicted to be much smaller (0.6%) as
86 compared to dust (87%) and soot (12%) (Hoose et al. 2010).

87

88 Many studies have used cloud models to highlight the potential impact of PBAP INPs on
89 cloud microphysics and precipitation (e.g., Levin et al. 1987; Grützun et al. 2008; Phillips et
90 al. 2009; Hummel et al. 2018). For example, the mesoscale aerosol-cloud model by Phillips et
91 al. (2009) had a 3-D domain of about 100 km in width, and many cloud types were present in
92 the mesoscale convective system that was simulated. Their simulations revealed that the
93 cloud cover, domain radiative fluxes, and surface precipitation rate were significantly altered
94 by boosting organic aerosols representing PBAPs. According to Hummel et al. (2018) in
95 shallow mixed-phase clouds (i.e., altostratus) when the cloud top temperature is below -15°C,
96 PBAPs have the potential to influence the cloud ice phase and produce ice crystals in the
97 absence of other INPs.

98

99 The quest for insights about the broader atmospheric role of PBAP INPs for cloud
100 microphysical properties and precipitation is hampered by the limited availability of
101 observations both of their ice nucleation activities for various species and their aerosol



102 distributions in the real atmosphere. More generally, there is incomplete knowledge about the
103 chemical identity of the key INPs, whether biological or otherwise (Murray et al. 2012). In
104 many global and regional models, the ice nucleation activity of bioaerosols is represented
105 either empirically or theoretically based on laboratory measurements of specific biological
106 species of PBAPs that are assumed as representative candidates (e.g., *Pseudomonas*
107 *syringae*). This assumption of representativeness introduces uncertainties that would be
108 expected to impact the model results, potentially introducing a bias into the estimation of
109 effects of bioaerosols on clouds.

110

111 In addition to primary ice nucleation, ice formation in clouds can occur because of processes
112 generating new particles from pre-existing ice, and these are known as Secondary Ice
113 Production (SIP) mechanisms. SIP can have a considerable impact on cloud micro- and
114 macro-physical properties such as precipitation rate, glaciation time, cloud lifetime, and cloud
115 electrification by modifying the order of magnitude of ice particle concentrations (e.g., Blyth
116 and Latham (1993); Crawford et al., 2012; Lawson et al., 2015; Phillips et al., 2017b, 2018,
117 2020; Phillips and Patade, 2021; Sotiropoulou et al. 2021a,b). This in turn can influence the
118 global hydrological cycle and climate. For example, Zhao and Liu (2021) demonstrated using
119 a global climate model that SIP dominates ice formation in moderately cold clouds and has a
120 significant influence on their liquid and ice water paths. They showed that including three
121 SIP mechanisms in the model simulated global annual average liquid water path decreases by
122 -15 g m^{-2} (-22%) and the ice water path increases by 9 g m^{-2} (23%), resulting in better
123 agreement with observations. Accounting for SIP in their model results in a change in the
124 global annual average net cloud radiative forcing by about 1 W m^{-2} . Although this is a small
125 fraction of the total cloud radiative forcing globally, this flux change underlines the
126 ubiquitous role of SIP on cloud properties on the large scale.



127

128 However, in many cloud models, the representations of these SIP mechanisms are uncertain.
129 A few secondary ice formation processes (e.g., Hallet-Mossop, 1974) have been suggested to
130 be active in the temperature range where active PBAP INPs exhibit strong ice nucleation
131 activity. The INPs of biological origin such as bacteria are highly active in the temperature
132 range of the Hallet-Mossop (HM) process (-3 to -8°C) as compared with non-biological INPs
133 (Möhler et al. 2008; Patade et al., 2021). At temperatures warmer than -15°C, some of the
134 PBAPs generated by biologically active landscapes can promote ice formation and crystal
135 growth in clouds (Morris et al., 2014).

136

137 In the USA, about 18% of the total landmass is used as cropland, farmland, and agricultural
138 activities. These are major sources of biological particles in the atmosphere. Biogenic
139 particles released from crops, either pre- and post-harvest, have previously been shown to
140 serve as INPs (in Colorado and Nebraska, (Garcia et al. 2012). Huffman et al. (2013) found
141 that airborne biological particles increase significantly in concentration, by an order of
142 magnitude or more, during rainfall in a forest in the western US and that bioaerosols are well
143 correlated with INPs. Prenni et al. (2013) observed a similar increase in concentrations of
144 ground-level INPs during rain at a forested site in Colorado, which was associated with
145 increased biological particles. If these potential INPs are detrained from the convective
146 outflow of a cell at mid-levels, then they may be entrained into other clouds aloft, influencing
147 the microphysical properties of that subsequent storm. Convective clouds can efficiently
148 transport lower tropospheric aerosol particles into the upper troposphere where they can
149 affect the cloud properties (Cui and Carslaw, 2006)

150



151 The current study aims to simulate realistic concentrations of multiple groups of PBAP INPs,
152 including bacterial and fungal particles, to investigate their interactions with convective
153 clouds observed during the Midlatitude Continental Convective Clouds Experiment field
154 campaign (MC3E), (Jensen et al. 2016), in the USA. In view of the literature noted above
155 about effects from PBAP INPs, there is a need for more detailed analyses of their role in
156 altering cloud microphysical properties and precipitation because the realistic treatment of IN
157 activity for major PBAP groups was not available, prior to our empirical scheme (Patade et
158 al. 2021). Hitherto, laboratory measurements of isolated biological species (e.g.,
159 *Pseudomonas syringae*, *Cladosporium sp*) have been the basis for attempts to simulate
160 biological ice nucleation in clouds, but the representativeness of the choice of such species
161 has been a longstanding issue. It is not known which biological species of INA PBAPs
162 contribute the most to biological ice nucleation. Consequently, there is a need for a new
163 approach oriented toward laboratory measurements of natural biological INPs sampled from
164 the real atmosphere, thus optimizing the representativeness of the data for studies of natural
165 clouds.

166

167 In this paper, such an approach is followed to investigate the effect on cloud properties from
168 various major groups of PBAP. We incorporated a recent empirical parameterization for
169 various PBAP groups by Patade et al. (2021, henceforth PT21) into our 3D aerosol-cloud
170 model (AC). PT21 created an empirical formulation resolving the ice nucleation of each
171 group of PBAPs including 1) fungal spores and their fragments, 2) bacteria and their
172 fragments, 3) pollen and their fragments, 4) detritus of plants, animals, and viruses, and 5)
173 algae.

174



175 The paper is structured as follows. In the next two sections, the methodology for the model
176 setup and a description of the observed storm simulated. The subsequent section shows
177 results comparing simulated cloud properties with observations to validate the model. There
178 follows the analysis of sensitivity tests concerning the loading of PBAPs. In this article, we
179 also examine the relative importance of various secondary ice processes in their role in
180 mediating the PBAP effects on cloud microphysical properties, given the weakness of PBAP
181 effects on cloud microphysical properties.

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184 **2. Description of field campaign of observations**

185 *2.1 Selected case of a deep convective system*

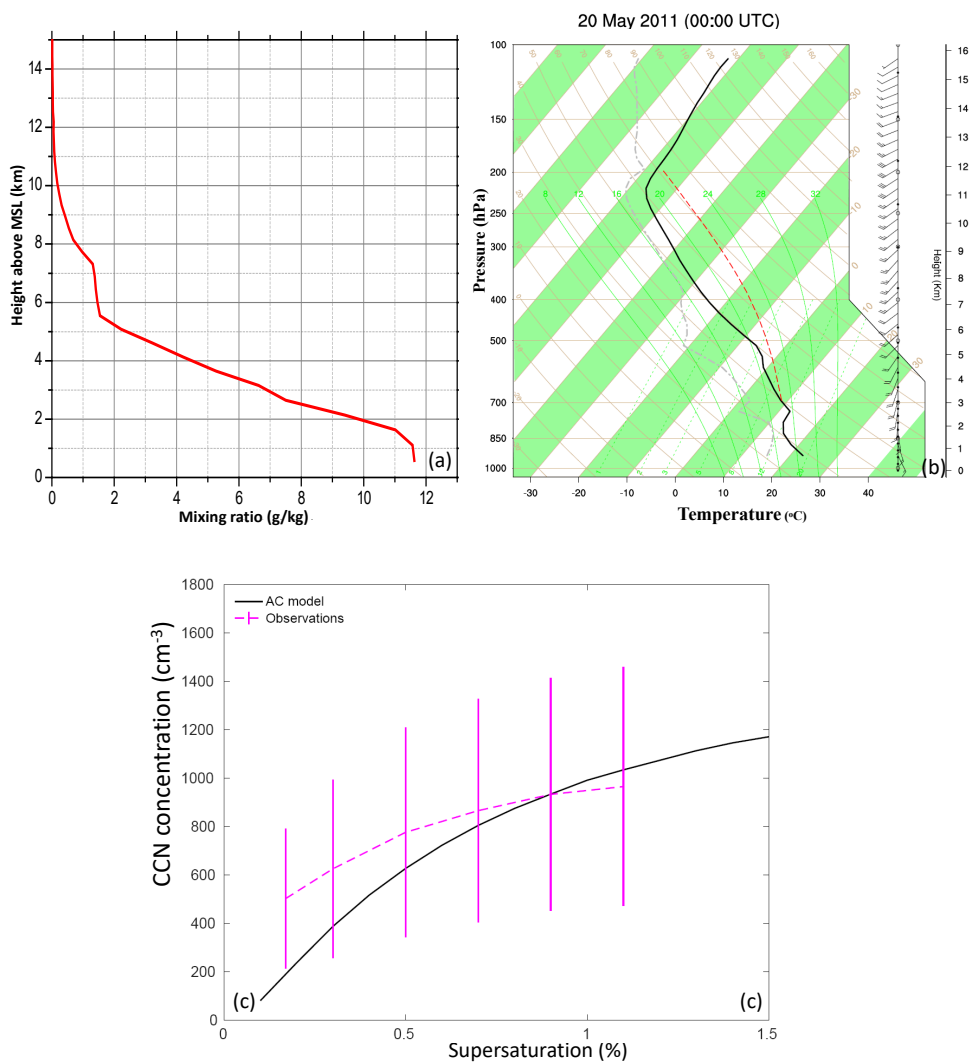
186 In the current study, we simulated a squall line that occurred on 20 May 2011 MC3E
187 (Jensen et al. 2016). The MC3E campaign took place from 22 April through to 6 June 2011
188 and was centred at the Atmospheric Radiation Measurement (ARM) Southern Great Plains
189 (SGP) Central Facility (CF), (36.6°N, 97.5°W) in north-central Oklahoma. Jensen et al.
190 (2016) describe the squall line as a "golden event" of the MC3E campaign in view of the
191 robust *in situ* sampling of extensive stratiform outflow from deep convection on this day. The
192 surface meteorological analysis on 20 May indicated a southerly flow at the surface, which
193 provided enough moisture from the Gulf of Mexico to trigger convection. Deep convection,
194 organized in the form of a squall line, passed over the measurement site between 1030 and
195 1100 UTC, resulting in convective precipitation. It was followed by widespread stratiform
196 precipitation that was well observed by both airborne and ground-based measurements.

197



198 The skew- T plot from the radiosonde sounding conducted on 20 May 2011, at (00 UTC) is
199 shown in Figure 1. The skew- T plot shows that the surface-based Convective Available
200 Potential Energy (CAPE) for this case was 2355 J kg^{-1} , and the Lifting Condensation Level
201 (LCL) was located at 840 hPa. The temperature at LCL, which is generally at the same height
202 as the convective cloud base, was 14°C . The estimated amount of vapour in the entire depth
203 of the troposphere corresponded to precipitable water of 3 cm.

204



205



206 **Figure 1.** (a) Vertical profile of water vapor mixing ratio on 20 May 2011 at 00 UTC. (b)
207 The *skew T* plot from May 20, 2011, sounding. The air temperature is represented by the
208 solid black line, while the dew point temperature is represented by a dashed grey line. The
209 moist adiabat is represented by a dotted red line. (c) The CCN spectrum from AC for a
210 simulated squall line case on May 20, 2011, for an environment 500 meters above MSL. The
211 predicted CCN spectrum is compared to the observed CCN spectrum at the SGP CF (300 m
212 above MSL).

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214

215 2.2 Aircraft Observations

216 The *in situ* microphysical observations used in this study were obtained from a University of
217 North Dakota Citation II aircraft. The aircraft collected observations of cloud microphysical
218 parameters from cloud base (1.8 km above MSL) to a maximum altitude of 7.5 kilometres
219 above MSL. The MC3E campaign collected extensive airborne measurements of aerosols and
220 cloud microphysical properties over north-central Oklahoma. A detailed description of the
221 scientific objectives of the MC3E program, including the field experiment strategy, airborne
222 and ground-based instrumentation, is given in the paper by Jensen et al. (2016). This section
223 summarizes the instrumentation used in the current study.

224 **Table 1:** Details of aircraft instruments used in this study.

Instrument	Measurement	Typical range
Cloud imaging probe (CIP) by Droplet Measurement Technologies (DMT)	Size distribution of cloud and precipitation particles	0.025–1.5 mm (0.2–1 mm for model validation in the current study)
2D cloud imaging probe (2D-C)	Size distribution of cloud and precipitation particles	0.03–1.0 mm (0.2–1 mm for model validation in the current study)
Cloud droplet probe (CDP)	Cloud droplet spectra	2–50 μm
High-volume precipitation spectrometer, version 3 (HVPS-3)	Precipitation particle spectra	0.15–19.2 mm



King hot-wire liquid water content (LWC) probe	Cloud liquid water	0.01–5 g m ⁻³
Temperature probe	Ambient air temperature	–
Static pressure sensor	Ambient air pressure	–

225

226

227 The Citation aircraft was equipped with a standard suite of meteorological instruments, which
228 provided high-resolution measurements of temperature, pressure, and humidity. In addition, it
229 carried microphysical probes for cloud and precipitation, and liquid water content, as listed in
230 Table 1.

231

232 Particle size distributions (PSDs) from cloud to precipitation particle sizes were measured
233 with various probes, including a 2D Cloud Imaging Probe (2D-C), a Cloud Imaging Probe
234 (CIP), and a High-Volume Precipitation Spectrometer Probe (HVPS). The 2D-C and CIP
235 probe data were processed objectively using the algorithm developed at the National Center
236 for Atmospheric Research (NCAR) to mitigate artifacts produced by shattering on the probes'
237 leading edges (Field et al. 2006). The 2D-C probe was equipped with Korolev anti-shattering
238 tips (Korolev et al., 2011), while the CIP did not have anti-shattering tips. The size
239 distribution of cloud drops with diameters from 2 to 50 μm was measured using a Cloud
240 Droplet Probe (CDP). A King hot-wire liquid water content (LWC) probe measured the
241 LWC. Vertical velocity is derived from air motion sensing systems available on the research
242 aircraft.

243

244 *2.3 Ground-based measurements*



245 A comprehensive instrumentation suite deployed at the ARM-SGP central facility provided
246 continuous measurements of atmospheric gases, aerosols, clouds, and local meteorological
247 conditions (e.g., wind, temperature, precipitation, and atmospheric profiles). A cloud
248 condensation nuclei (CCN) counter (DMT) measured the CCN number concentration at
249 seven supersaturation values with a temporal resolution of 1 hour. Surface precipitation was
250 measured with 16 rain gauge pairs placed within a 6-kilometer radius of the SGP CF.

251

252 During the MC3E campaign, the extended facility deployed at CF measured the spatial
253 variability of surface fluxes of heat, moisture, and momentum. A radiosonde array of 6 sites,
254 covering an area of 300 km × 300 km, was designed to capture the large-scale variability of
255 the atmospheric state. Radiosonde observations (Vaisala RS92-SGP) were conducted four
256 times daily at around 05:30, 11:30, 16:30, and 22:30 UTC, providing vertical profiles of
257 atmospheric state variables (pressure, temperature, humidity, and winds) of the environment
258 surrounding the ARM SGP site. When aircraft operations were planned based on forecasted
259 convective conditions, the sounding frequency was increased to eight times per day.

260

261 In addition to airborne observations, the ARM heterogeneous radar network was used to
262 conduct unique radar observations during the MC3E campaign. The information about
263 various radar assets during MC3E is given by Jensen et al. (2016). The surface precipitation
264 used for model validation in this study is a radar-based precipitation estimate as described by
265 Giangrande et al. 2014. They used radar observations from the C-band and X-band scanning
266 ARM precipitation radars (C-band Scanning ARM Precipitation Radar and X-band Scanning
267 ARM Precipitation Radars, respectively) to estimate rainfall to within 100 km of the ARM



268 facility in Lamont, Oklahoma. Their radar-based rainfall retrievals were in good agreement
269 with observations with absolute bias less than 0.5 mm for accumulations less than 20 mm.

270

271 The Interagency Monitoring of Protected Visual Environments (IMPROVE) network stations
272 close to the location of airborne observations provided ground-level measurements of various
273 chemical species. These included carbonaceous compounds (black and organic carbon), salt,
274 ammonium sulfate, and dust. The measurements of these aerosol species from various
275 IMPROVE sites, including Ellis (36.08°N, 99.93°W), Stilwell (35.75°N, 94.66°W), and
276 Wichita Mountains (34.73 °N, 98.71°W) sites in Oklahoma, were averaged to provide inputs
277 to AC. Initial concentrations for the aerosol species of AC (10 species) including sulfate, sea
278 salt, dust, black carbon, soluble organic, and insoluble organic (five groups of bioaerosols)
279 were derived from the Goddard Chemistry Aerosol Radiation and Transport (GOCART)
280 model (Chin et al. 2000). Profiles of aerosol mass concentration were scaled at all levels to
281 match simultaneous measurements by IMPROVE. The prescribed mass mixing ratios of
282 aerosol species in AC-based IMPROVE observations are enlisted in Table 2.

283 **Table 2:** The mass mixing ratio of aerosol species based IMPROVE observations which are
284 used as input to AC.

Aerosol species	Mass mixing ratio ($\mu\text{g}/\text{m}^3$)
(NH ₄) ₂ SO ₄	0.56
Dust	0.18
Sea salt	0.021
Black carbon	0.093
Soluble organic carbon (80 % of TOC)	0.45
Insoluble organic carbon (20 % of TOC)	0.18
PBAPs (50% of Insoluble organic carbon)	FNG=0.036; BCT=0.012; PLN=0.028;



	DTS=0.016; ALG=0.000022
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287

288 **3. Methodology**

289 *3.1 Model description*

290 The ‘*aerosol-cloud model*’ (AC) used in this study is a cloud-resolving model (CRM) with a
291 hybrid spectral bin/two-moment bulk microphysics, interactive radiation, and semi-
292 prognostic aerosol schemes (Phillips et al. 2017a, 2020). The model predicts the mass and
293 number concentrations for five types of hydrometeors: cloud liquid, cloud ice (or “crystals”),
294 rain, graupel/hail, and snow. The mixing ratios of the total number and mass of all particles in
295 each microphysical species are treated as model prognostic variables. AC treats all known
296 microphysical processes such as droplet nucleation, ice initiation through primary and
297 secondary processes, and growth processes such as deposition/sublimation of ice particles,
298 condensation/evaporation of drops, freezing/melting, as well as coagulation by collisions
299 between various hydrometeor types. Both cloud-base and in-cloud activation of aerosols to
300 form cloud-droplets are treated explicitly, with the predicted in-cloud supersaturation
301 resolved on the model grid being used to activate aerosols aloft. Bin-resolved size
302 distributions of each aerosol species are predicted for the interstitial and immersed
303 components of each aerosol species. Extra prognostic variables track the number of aerosols
304 in each aerosol species that have been lost by INP and CCN activation.

305

306 Secondary ice formation is represented by four types of fragmentation:



307 • breakup in ice–ice collisions (Phillips et al. 2017a, b),
308 • Hallett and Mossop (1974), rime splintering,
309 • fragmentation of freezing rain/drizzle by modes 1 and 2 (Phillips et al. 2018);
310 • and sublimation breakup (Deshmukh et al. 2021).

311 The empirical parameterization (EP) (Phillips et al. 2013) of heterogeneous ice nucleation
312 treats all known modes of ice formation (deposition mode, condensation-/immersion-
313 freezing, inside-out and outside-in contact-freezing) in terms of dependencies on the loading,
314 size, and chemistry of multiple aerosol species. In the previous version of the EP, there were
315 four species of INP aerosol. One of these was PBAP INPs. However, that version of the EP
316 did not resolve the individual types of PBAP IN, which exhibit a wide range of ice-nucleating
317 abilities. A recent study PT21 provided an empirical formulation for multiple groups of
318 PBAP INPs based on field observations over central Amazon. In this study, we modified AC
319 by implementing the recent empirical parameterization of PBAP INPs by PT21. A summary
320 of their formulation is provided in section 3.2.

321

322 Cloud processes and rainfall formation have been detected using different radar variables,
323 such as specific differential phase K_{DP} . Moisseev et al. (2015), for example, noted an increase
324 in observed K_{DP} because of aggregation. In addition, a few studies have hypothesized
325 evidence of SIP via K_{DP} (e.g., Sinclair et al. 2016; Kumjian and Lombardo 2017; Carlin et al.
326 2021). In this study, we attempted to detect secondary ice formation signatures by
327 implementing K_{DP} estimations into AC. Based on Ryzhkov et al. (2011), K_{DP} values were
328 estimated for various hydrometeor types, including cloud drops, raindrops, cloud ice, snow,
329 and graupel (their equations 22, 23, 24, 26, and 29). The scattering amplitudes were
330 calculated using the Rayleigh approximation. The K_{DP} estimations are made for 0° elevation



331 angle and S-band (radar wavelength of 11 cm). The equivalent volume diameter of the given
332 hydrometeor was used for all calculations.

333

334 *3.2 Empirical formulation for PBAP INPs:*

335 The empirical formulation by PT21 for multiple groups of PBAPs including:- 1) fungal
336 spores (FNG), 2) bacteria (BCT), 3) pollen (PLN), 4) viral particles, plant/animal detritus
337 (DTS), 5) algae (ALG) and their respective fragments are implemented in AC. This
338 formulation is based primarily on field observations over the central Amazon rainforest, with
339 empirically derived dependencies on the surface area of each group (except algae) and it
340 applies to the particles with diameters greater than 0.1 μm . The concentrations of algal
341 particles at the Amazon Tall Tower Observatory (ATTO) site were much smaller than our
342 detection threshold, so we could not use a similar empirical treatment for ALG. The frozen
343 fraction for the algal particles (Diatom cell, *T. pseudomonas*) available in the literature is
344 used to estimate INPs from ALG (Wilson et al. 2015). PT21 elaborate further.

345

346 *3.3 Model setup*

347 AC was driven by initial and evolving boundary data for meteorological conditions. The
348 model simulations were carried out for a three-dimension domain of 80 km x 80 km domain
349 with horizontal grid spacings of 2 km. In vertical, the model resolution was 0.5 km, and the
350 model top was located at about 16 km. The lateral boundary conditions are doubly periodic
351 on all sides of the domain. The large-scale forcing condition used for the simulation was
352 derived using the constrained variational analysis (Xie et al. 2014; Jensen et al. 2016). The



353 initial time of the simulations was at 1200 UTC on 19 May 2011 and all simulations were
354 performed for 48 hours at a time step of 10 seconds.

355

356 The GOCART model (Chin et al. 2000) was used to initialize the seven chemical species
357 associated with the EP. The data from the three IMPROVE sites mentioned above (Section
358 2.3) was used to rescale the mass concentration profiles at all levels so that they match the
359 measurements. Table 2 lists the mass mixing ratios of various aerosol species after the
360 corrections. The relative contribution of insoluble and soluble organic carbon to the total
361 organic carbon (TOC) was assumed to be 20% and 80%, respectively by assuming a water-
362 soluble fraction of 80% for carbonaceous aerosol (Phillips et al. 2017b). Due to a lack of
363 observational data, it was parsimoniously assumed that 50% of the insoluble organics were
364 biological in origin. The total PBAP loading was prescribed partly based on observations of
365 insoluble organics and partly based on the assumed fraction. PT21's observations were used
366 to calculate the relative contribution of various PBAP groups to insoluble organics. The
367 parameters for the shape of PSD of each PBAP group (modal mean diameters, standard
368 deviation ratios, and relative numbers in various modes) are prescribed based on observations
369 from Amazon (PT21).

370

371 From these prescribed loadings of aerosol species, AC predicts their size distribution and
372 hence the CCN activity spectrum. Using the initial sounding and aerosol profile, AC can
373 predict the in-cloud size distribution of aerosols in each species as well as in-cloud
374 supersaturation. To validate this prediction, Figure 1c shows the predicted CCN spectrum
375 compared with observations from the CCN counter at the surface at the SGP site. During 19-
376 20 May, the measured number concentration of active CCN at the SGP CF ranged from 400



377 to 3000 cm^{-3} at 1% supersaturation (Fridlind et al. 2017). The measurements were made on
378 20 May before the start of the rain in clear air. The normalized CCN number concentrations
379 at 1% supersaturation from observations and AC are $\sim 1000 \text{ cm}^{-3}$ and $\sim 940 \text{ cm}^{-3}$,
380 respectively. Given the wide range of observed CCN concentrations at each supersaturation,
381 the predicted and observed CCN activity spectra are in good agreement.

382

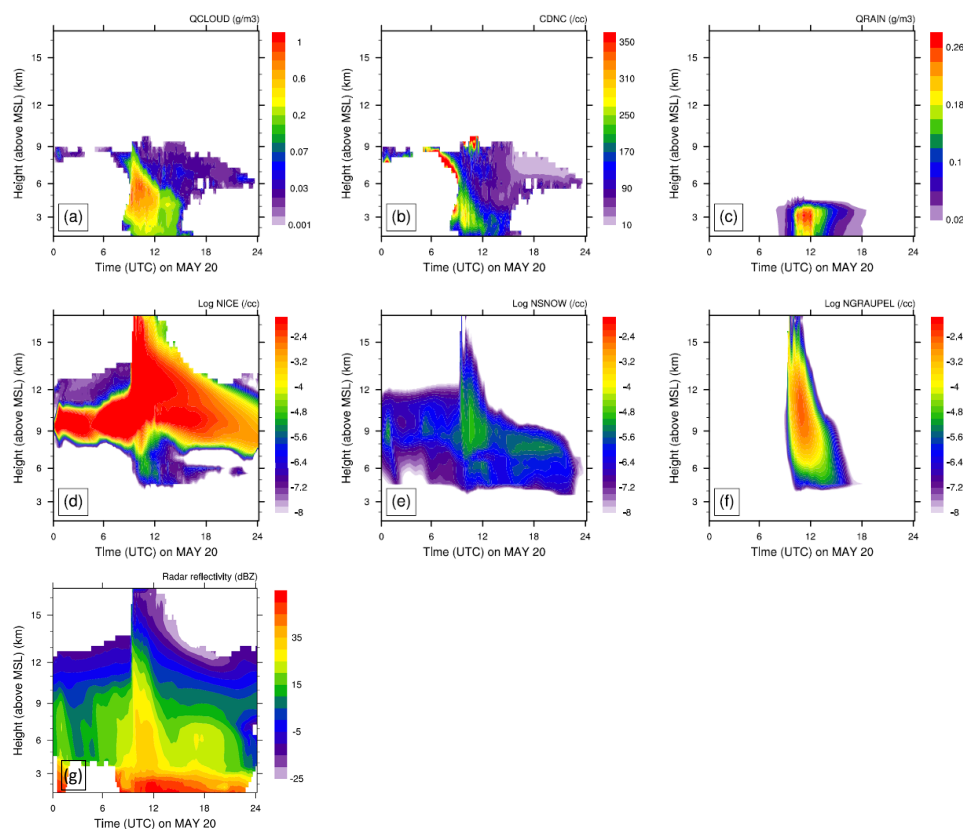
383

384 **4. Results from control simulation and model validation**

385 *4.1 Overview of the control simulation*

386 An intense north-to-south oriented squall line moved over the ARM SGP CF on May 20,
387 2011, from 1100 to 1400 UTC (Sec. 2.1). The new version of AC simulated this case, after
388 implementing the empirical formulation by PT21 for multiple groups of PBAP INPs
389 ('control' simulation) (Sec. 3).

390



391

392 **Figure 2:** Time-height contours of domain averaged a) cloud water mixing ratio (QCLOUD);
393 b) cloud droplet number concentration (CDNC); c) rainwater mixing ratio (QRAIN); d)
394 number concentration of cloud ice (NICE); e) number concentration of snow (NSNOW); f)
395 number concentration of graupel (NGRAUPEL). Due to a wide range of values, the log
396 values number concentrations are plotted. The surface height is ~ 500 m. The averaging was
397 done for cloud points with $LWC > 0.001\text{g m}^{-3}$ or $TWC > 10^{-6}\text{g m}^{-3}$. Also shown is the time-
398 height evolution of domain averaged (g) radar reflectivity.

399

400 Figure 2 shows the time-height evolution of various liquid and ice microphysical parameters
401 derived from the control simulation conditionally averaged over cloudy regions. The
402 maximum of the average cloud droplet number concentration was around 250 cm^{-3} . The
403 liquid water content was typically less than 0.5 g m^{-3} . The freezing level (0°C) was around
404 4.1 km above MSL. The deep convection began around 10 UTC, followed by intense



405 precipitation around 11 UTC, and reached its peak around 12 UTC. The time-height
406 evolution of cloud ice, snow, and graupel number concentrations shows maxima shortly
407 before 12 UTC, which coincides with the time of peak precipitation. This suggests that the
408 ice phase was important in precipitation formation.

409

410 The time-height map of simulated radar reflectivity during 20 May, unconditionally averaged
411 over the whole domain is shown in Figure 2g. It shows the well-defined squall line passing
412 over the domain from 1100 to 1500 UTC. The maximum of this domain-wide simulated
413 reflectivity is around 40 dBZ (Fig. 2d) when deep convection was happening. The
414 instantaneous maximum of reflectivity at any grid-point (not shown) was about 50 dBZ. At
415 other times, the average reflectivity was typical of the stratiform cloud of about 15 dBZ. The
416 cloud top height of the squall line decreases after 1400 UTC.

417

418

419 *4.2. Model validation against coincident observations of the storm*

420 The extended stratiform region of the squall line while in the vicinity of the SGP CF was
421 sampled by the Citation aircraft equipped with a full suite of cloud microphysical
422 instrumentation (Sec. 2). The aircraft started sampling the stratiform precipitation region at
423 around 1300 UTC and continued the observations at sub-freezing temperatures from 1335 to
424 1515 UTC. Occasionally, the aircraft encountered weak convective updrafts (< 6 m/s). The
425 aircraft actively avoided convection that was more vigorous than that. In this section, we
426 validate various microphysical and dynamical quantities from the control simulation against
427 aircraft and ground measurements.



428

429 Figure 3 compares the aircraft observations against predicted microphysical quantities, with
430 both the predictions and observations identically averaged, conditionally over convective ($6 >$
431 $|w| > 1$ m/s) and stratiform regions $|w| < 1$ m/s). The simulated LWC decreases exponentially
432 with height above cloud base. There is considerable scatter in observed LWC at each level.
433 The various degrees of dilution of sampled parts of the cloud can cause these variations in
434 LWC at a given altitude. The maximum simulated LWC of 0.5 gm^{-3} was observed in the
435 convective region at temperatures warmer than -5°C . Overall, the means of observed LWC
436 are in good agreement with the model results for convective as well as stratiform regions.

437

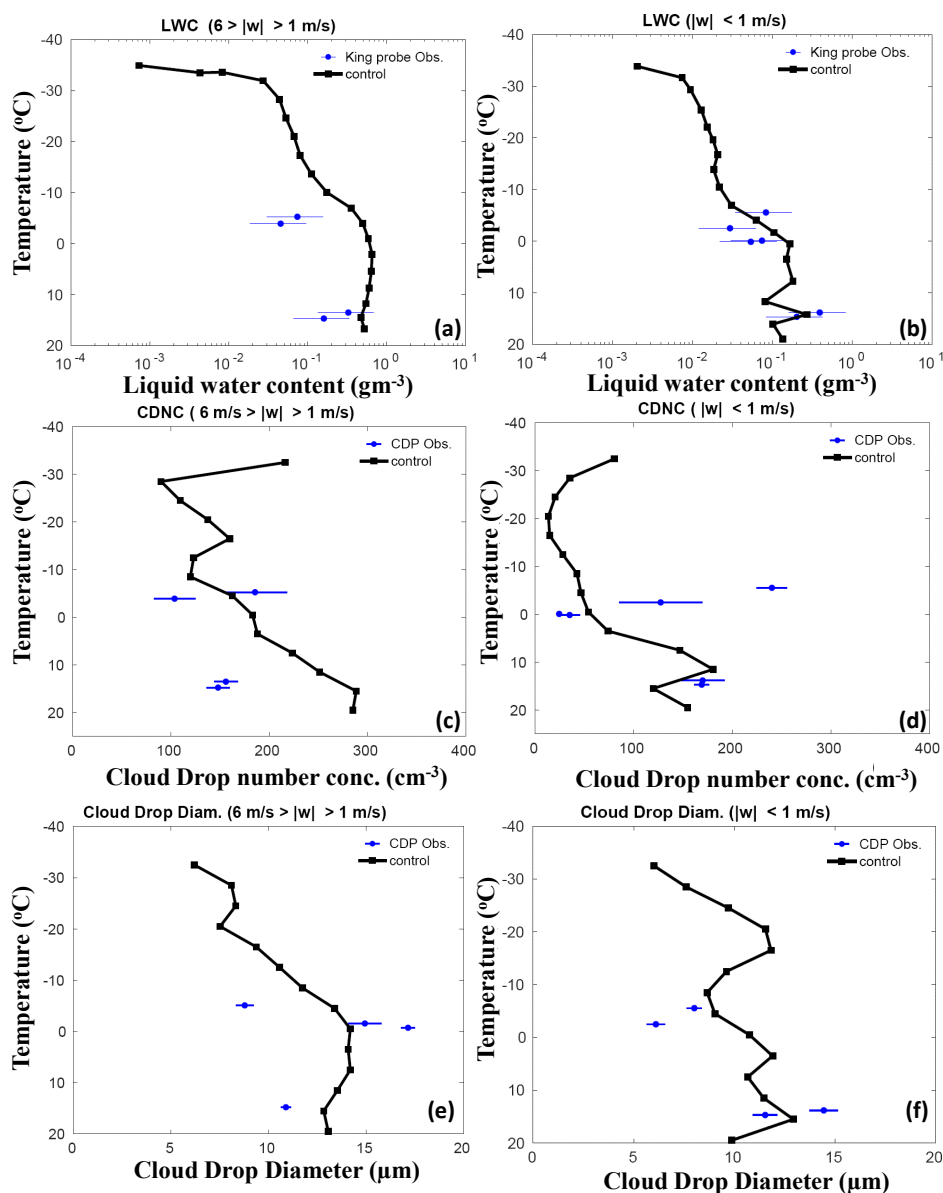
438 The vertical profiles of simulated and observed Cloud Drop Number Concentration (CDNC)
439 (Fig. 3c and 3d) showed that CDNC was lower than 300 cm^{-3} . Overall, the mean values of
440 CDNC simulated for convective and stratiform regions are in good agreement with
441 observations. The observed and simulated mean diameter of cloud-droplets varied between 6
442 to $15 \mu\text{m}$ over height (Figure 3e and 3f). Overall, the predictions of average CDNC and cloud
443 droplet diameter, in both convective and stratiform regions, show good agreement with
444 observations.

445

446



447



448

449 **Figure 3:** Comparison of the control simulations by AC with aircraft observations, for liquid
450 water content conditionally averaged over (a) convective ($6 \text{ m/s} > |w| > 1 \text{ m/s}$) and (b)
451 stratiform ($|w| < 1 \text{ m/s}$) regions; cloud drop number concentration over (c) convective and (d)
452 stratiform regions; average size of cloud droplets ($< 20 \mu\text{m}$) conditionally averaged over (e)
453 convective and (f) stratiform regions. All the vertical profiles shown here are averaged for the
454 whole domain.



455

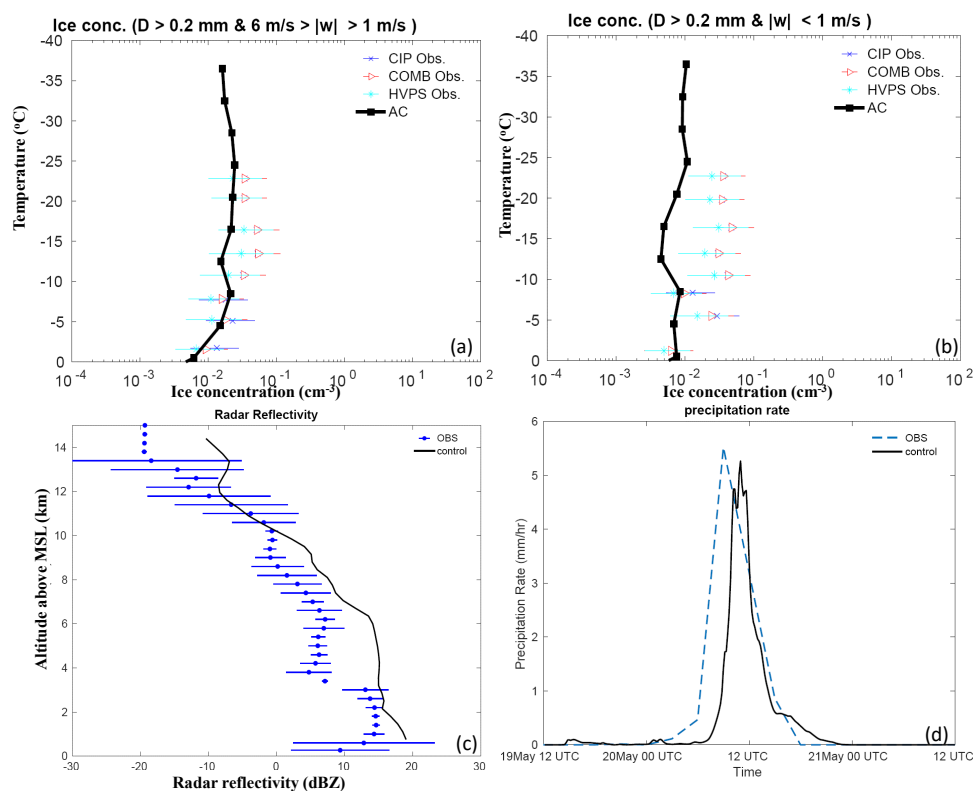
456 The ice particle number concentration from observations and the control simulation is also
457 compared as shown in Figures 4a and 4b for convective and stratiform regions, respectively.
458 Observations show that the concentration of ice particles gradually increases as the
459 temperature decreases, as predicted. The maximum ice number concentration from the
460 aircraft observations (with $D > 200 \mu\text{m}$) is $\sim 60 \text{ L}^{-1}$ around $-15 \text{ }^\circ\text{C}$. The number
461 concentrations of ice particles from the model with a diameter larger than $200 \mu\text{m}$ (cloud ice,
462 snow, graupel) (denoted by ‘NT200’) is compared with observed ice number concentrations
463 from CIP as well as from combined probes including 2DC, CIP, and HVPS. Ice particles
464 smaller than $200 \mu\text{m}$ from these imaging probes were excluded due to difficulties in
465 measuring small ice particles (Korolev et al., 1991) as well as to avoid contamination from
466 large cloud droplets. Good agreement to within 50% at most levels, was found between the
467 model simulated (NT200) and that observed for the convective region.

468

469 In the stratiform region, at most levels, model values of NT200 have the same order of
470 magnitude as observations. However, between about the -10 and -16°C levels, the stratiform
471 NT200 values are about two-thirds of an order of magnitude lower than the observations. In
472 similar simulations of the 20 May case, Fan et al. (2015) and Fridlind et al. (2017) also
473 showed underestimation of measured ice number concentrations. Compared to observations,
474 their simulations showed half an order of magnitude bias in ice crystal number concentration.
475 Comparatively, our model predicted ice number concentrations were in better agreement with
476 observations. As mentioned in section 2.2, imaging probe data is prone to shattering, and
477 various corrections were used to rectify it. However, there are currently no ways to determine
478 how many undetected artifacts remain after shattering corrections have been applied. Such



479 uncertainties in measured ice number concentration could result in such bias in observed and
480 simulated ice number concentrations.



481

482

483 **Figure 4:** Comparison of the control simulations by AC with aircraft observations, for ice
484 number concentration of all particles > 0.2 mm in the maximum dimension of all
485 microphysical species (cloud ice, graupel/hail, snow), averaged over (a) convective (6 m/s >
486 |w| > 1 m/s) and (b) stratiform (|w| < 1 m/s) regions. (c) The vertical profile of simulated
487 radar reflectivity conditionally averaged over all regions of significant reflectivity (> -20
488 dBZ) at each level is compared with observations from ground-based radars. The temperature
489 corresponding to each altitude is mentioned on the right axes; (d) predicted precipitation rate
490 (mm/hr) compared with ground observations at the SGP CF. All the vertical profiles shown
491 here are averaged for the whole domain.

492

493



494 In Figure 4c, the radar reflectivity from vertically pointing Ka-band ARM zenith radar is
495 compared with the mean profile from model simulations. This illustrates figure illustrates that
496 simulated reflectivity profiles below roughly 3 km and above 8 km MSL altitudes are in good
497 agreement with observations. Between 3 and 8 km MSL (temperatures of 2 and -30°C), the
498 bias in reflectivity from model simulations and observations is about 7-8 dBZ. Thus, the
499 simulated reflectivity is substantially higher than observed, particularly at levels where the
500 aircraft sampled the clouds. Fridlind et al. (2017), as well as Fan et al. (2015), noticed similar
501 overestimations of reflectivity within stratiform outflow of the squall line case on 20 May.
502 They attributed the reflectivity biases to significantly larger ice particles in the simulations
503 than observed.

504

505 Figure 4d compares the time series of precipitation rate from the control simulation with the
506 radar-based precipitation estimates. In both, control simulation and observations, a maximum
507 precipitation rate of about 5 mm/hr was noticed, with an error in the prediction of less than
508 5%. In comparison to observations, the simulated squall line arrives 1-2 hours later. The lack
509 of resolution of the 3D turbulence in the PBL and uncertainties associated with the 3D
510 structure of initial and boundary conditions can all have an independent impact on the
511 simulated rainfall structure, resulting in a delayed peak. Nonetheless, good agreement
512 between predicted and retrieved precipitation rates is observed.

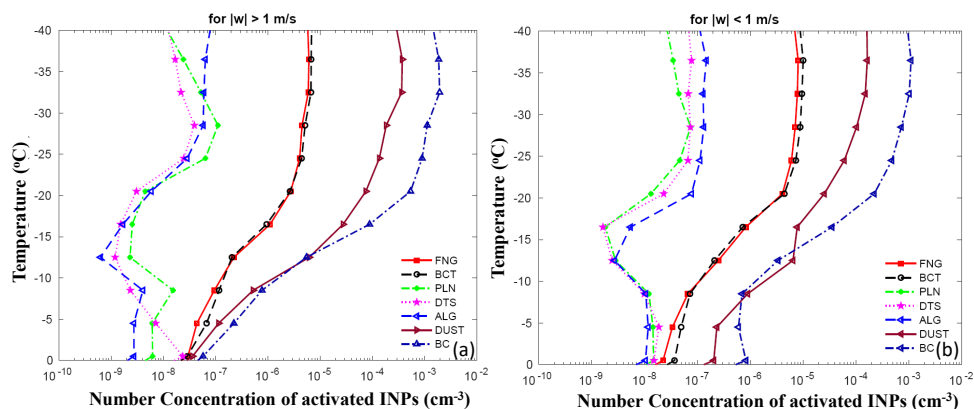
513

514 *4.3 Analysis of simulation with ice particle budgets and tagging tracers*

515 The activated PBAP INPs from the budget of the control run are shown in Figure 5 for the
516 convective and stratiform regions. Overall, bacterial, and fungal particles dominate the



517 biological INP concentration in the simulated cloud. For example, at -20°C the activated INPs
518 from bacteria and fungi are higher than the other three groups of PBAP INPs (detritus, pollen,
519 algal) by two orders of magnitude in both convective as well as stratiform regions. At that
520 level in convective regions, the maximum of the average concentration of simulated active
521 PBAP INPs is about 0.003 L^{-1} , which is two orders of magnitude less than the maximum
522 total for all active INPs (about 0.3 L^{-1}) in the whole simulation. In addition to the PBAP
523 INPs, Figure 5 also shows the activated INPs from dust and black carbon. Overall, the
524 predicted total INP concentration is dominated by black carbon and dust. At -10°C , the
525 Activated INPs from dust and black carbon differ by an order of magnitude from the total
526 PBAP INPs in convection.



527

528 **Figure 5:** The activated number concentration INPs from various PBAP groups along with
529 dust (DUST) and black carbon (BC) at various temperatures for (a) convective and (b)
530 stratiform regions. All the vertical profiles shown here are averaged for the whole domain.

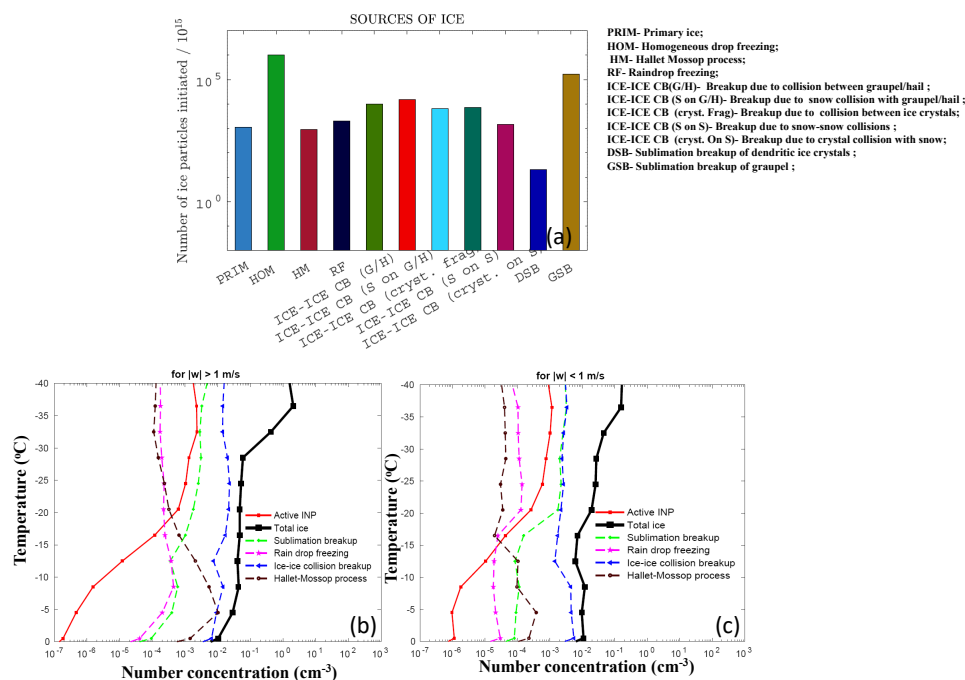
531

532

533 The formation of ice in a cloud is a result of several primary and secondary processes. It is
534 important to understand the relative importance of these processes in precipitation formation.



535 To that end, in Figure 6a a budget is provided for the total numbers of all ice particles
 536 initiated from various sources throughout the 3D domain of the entire simulation.
 537



538
 539

540 **Figure 6:** (a) Ice crystal budget for simulated MC3E case. The number of ice crystals
 541 produced by various mechanisms (as shown in the legend box) per 10^{15} particles is shown.
 542 The number concentration of ice particles from the model is shown for the whole range. Also
 543 shown is the number concentration of ice particles produced by various SIP mechanisms such
 544 as sublimation breakup, raindrop freezing, ice-ice collision breakup, and the Hallett-Mossop
 545 process as a function of temperature, averaged conditionally over only (b) convective and (c)
 546 stratiform regions. The convective and stratiform regions were identified based on criteria $|w|$
 547 > 1 and $|w| < 1$, respectively. The total ice number concentration and concentration of active
 548 INP are also displayed. All the vertical profiles shown here are averaged for the whole
 549 domain.

550
 551



552 Primary ice formation by homogeneous freezing of supercooled cloud droplets dominates the
553 total ice budget. Among all SIP mechanisms, breakup caused by collisions between various
554 ice particles is the most important in determining total ice number concentration. The ice
555 production by sublimation breakup of graupel is slightly lower than homogeneous freezing.
556 However, the contribution of ice production via sublimation breakup of dendritic ice crystals
557 is negligible.

558

559 Figure 6b depicts the dependency on the ambient temperature of components of ice
560 concentration from various SIP mechanisms, as well as active INPs and total ice number
561 concentration. Each source of ice displayed is tracked with “tagging tracers” throughout the
562 simulation. Overall, at temperatures warmer than -30°C , the total ice number concentration is
563 1-2 orders of magnitude higher than the concentration of active INPs, highlighting the
564 importance of SIP mechanisms in ice formation. At -25°C , breakup in ice-ice collisions
565 contributes 50% and 10% of the total ice concentration for the convective and stratiform
566 regions, respectively. At the same temperature, in both convective and stratiform regions,
567 sublimation breakup and raindrop freezing contribute 7% and 0.5 %, respectively.

568

569 Secondary ice formation via the HM process of rime-splintering contributes significantly to
570 ice production at temperatures warmer than about -15°C (Fig. 6b), enhancing the ice
571 concentration beyond the primary ice. In the convective region, the contribution of this
572 process to the total ice concentration is maximum around -5°C (about 30%). The simulated
573 cloud droplet diameter is mostly smaller than $15\ \mu\text{m}$. It is smaller than the cloud droplet size
574 required for the HM process to occur. AC represents the observed dependency of rime-
575 splintering on the concentration of droplets $> 24\ \mu\text{m}$. The fluctuations in CCN concentration



576 in the boundary layer can result in larger cloud droplet diameters in some parts of the cloud,
577 favouring secondary ice formation via the HM process.

578

579 **5. Results from sensitivity tests about influence from PBAP**

580 To quantify the effect of multiple types of PBAPs on cloud properties, sensitivity tests were
581 performed by modifying the control simulation and comparing the perturbed simulations with
582 it.

583

584 *5.1 All PBAPs*

585 Simulations were performed by eliminating all PBAPs from the control (*'no-pbap'* case) and
586 by multiplying their initial loadings at all levels by factors of 10 and 100 (*'high-pbap'* and
587 *'very high-pbap'* cases) respectively. Comparison with the control simulation reveals the
588 overall effect from both the CCN and IN activities of all bioaerosols combined. These factors
589 are justified by considering the variations in PBAP concentrations in the range of about 0.1 to
590 30 L^{-1} over North American forests (Huffman et al. 2013). An additional simulation was
591 conducted with a 1000-fold increase in initial PBAP loading (*'ultra high-pbap'*) to
592 investigate if these unrealistically high concentrations of PBAPs could affect the ice phase in
593 a purely hypothetical scenario.

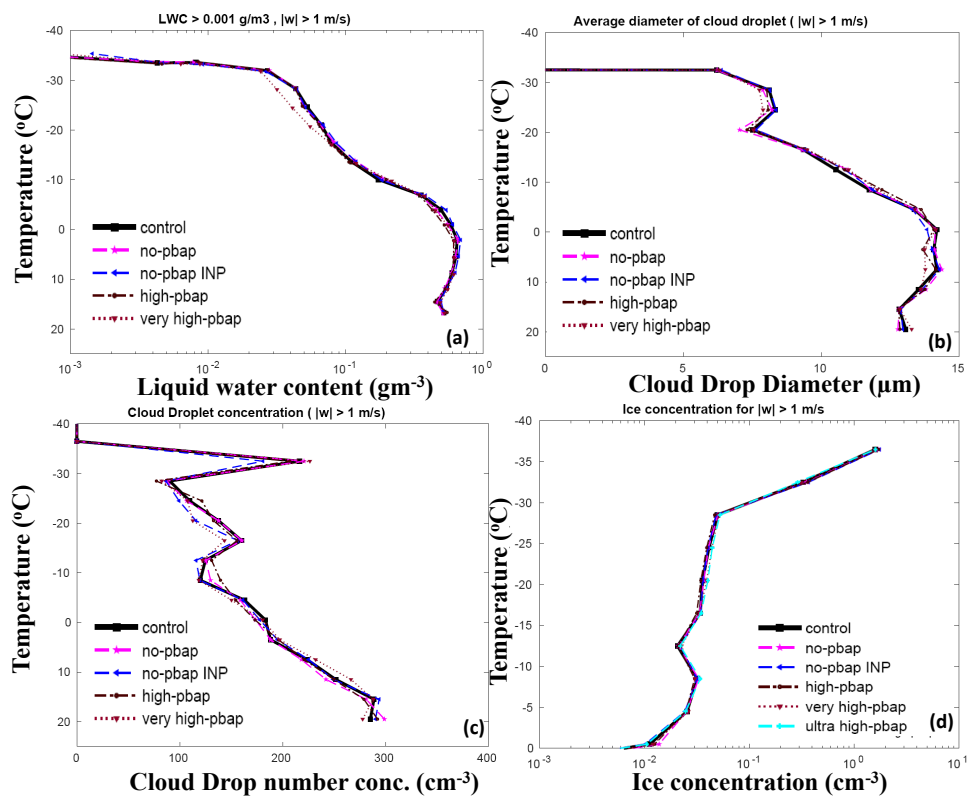
594

595 Additional simulations were performed by removing treatment of biological IN activity in the
596 EP (*'no-pbap INP'* case) relative to the control run. Comparison of both additional
597 simulations against the corresponding simulations with the full change in the PBAP loadings
598 (no-pbap and high-pbap cases) reveals the separate roles of the INP and CCN activities for



599 the changes in biological material. Apart from these changes in PBAPs, the perturbed
600 simulations are identical to the control run.

601



602

603 **Figure 7:** The temperature dependence of the (a) liquid water content, the (b) cloud droplet
604 number, (c) the cloud droplet diameter, and the (d) total ice number concentration for
605 ‘control’ simulation and various sensitivity runs involving a change in total PBAP number
606 concentrations for in the convective region. The averaging conditions are mentioned at the
607 top of each figure. The ice number concentration from the ultra high-pbap is also shown in
608 panel d. All the vertical profiles shown here are averaged for the whole domain.

609

610

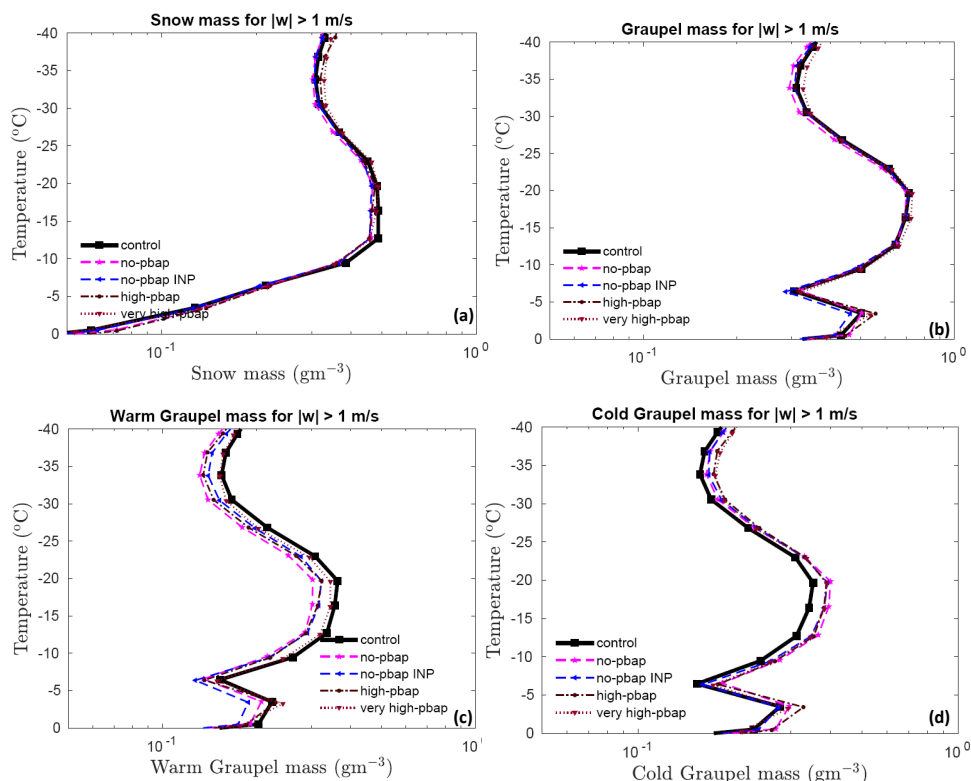
611 Figure 7 reveals the effects of all bioaerosols on cloud properties in the convective region

612 ($|w| > 1$ m/s). Overall, changes in cloud microphysical properties including liquid water



613 content, cloud droplet size, cloud drop number concentration, ice number concentration are
614 less sensitive to the changes in PBAPs for the convective part of the simulated clouds than
615 the stratiform part (Figure 9). The maximum change in simulated LWC is noted in the very-
616 high pbap simulation at -25°C and is less than 50%. The cloud droplet number and cloud
617 drop diameter in the perturbed simulations does not differ much ($< 5\%$) from the control run.
618 In the convective region, changes in ice crystal number concentration due to changes in
619 PBAPs are negligible. This includes the extreme changes in bioaerosol loading (ultra high-
620 pbap case).

621



622

623 **Figure 8:** The temperature dependence of total (a) snow and (b) graupel mass for the
624 convective region. Also shown are the (c) warm and (d) cold components of the graupel
625 mass. All the vertical profiles shown here are averaged for the whole domain.



626

627

628

629 Figure 8 shows the sensitivity of snow and graupel mass with respect to the changes in all
630 PBAPs for the convective region. The figure also includes the sensitivities for warm and cold
631 graupel mass to changes in PBAPs (Figures 8c and 8d). Warm graupel generation is linked to
632 raindrop formation by collision coalescence, which then freezes to form graupel. Cold
633 graupel is by the riming of snow through the ice crystal process. Overall, the changes in snow
634 and graupel mass are smaller than 15% indicating their low dependence on PBAPs. The
635 changes in PBAPs affected warm and cold components of graupel mass with a maximum
636 change of 30% in the no-pbap run.

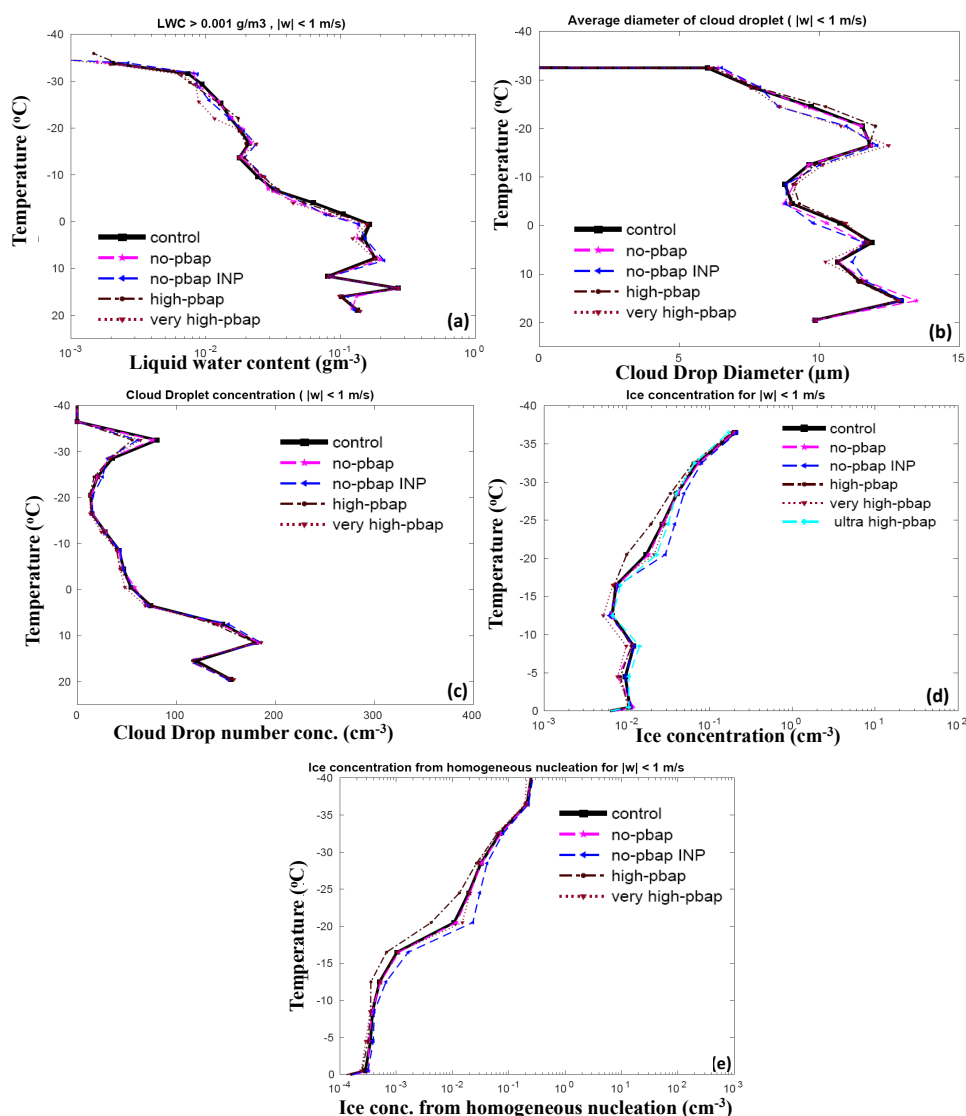
637

638 Figure 9 shows the corresponding effects in the stratiform region ($|w| < 1$ m/s) from all
639 bioaerosols. The changes in warm microphysical properties as a result of changes in PBAP
640 loadings are very small. In this part of the cloud, the ice-microphysical parameters are
641 comparatively more sensitive to the changes in PBAPs than in the convective region. The no-
642 pbap INP case predicted a 60% higher ice number concentration than the control run at -
643 20°C. This could be associated with enhanced ice formation through homogeneous nucleation
644 since the cloud droplet number concentration at this level was slightly higher compared to the
645 control run. At this temperature level, the changes in ice number concentrations are
646 maximum. At the same temperature, the simulated ice crystal number concentration in the
647 high-pbap condition is 40% lower than in the control run. Even with large increases in
648 bioaerosol loading in the ultra high-pbap, changes in simulated ice number concentrations are
649 less than 20% of the control run.

650



651 Figure 9e shows that at the colder temperatures ($T < -20^{\circ}\text{C}$) homogeneous freezing aloft with
652 downwelling of homogeneously nucleated ice dominates the ice number in all simulations.
653 The sensitivity of ice number concentrations to PBAP loading at temperatures colder than
654



655

656 **Figure 9:** The temperature dependence of (a) the liquid water content, (b) the cloud droplet
657 number, (c) the cloud droplet diameter, and the (d) total ice number concentration for

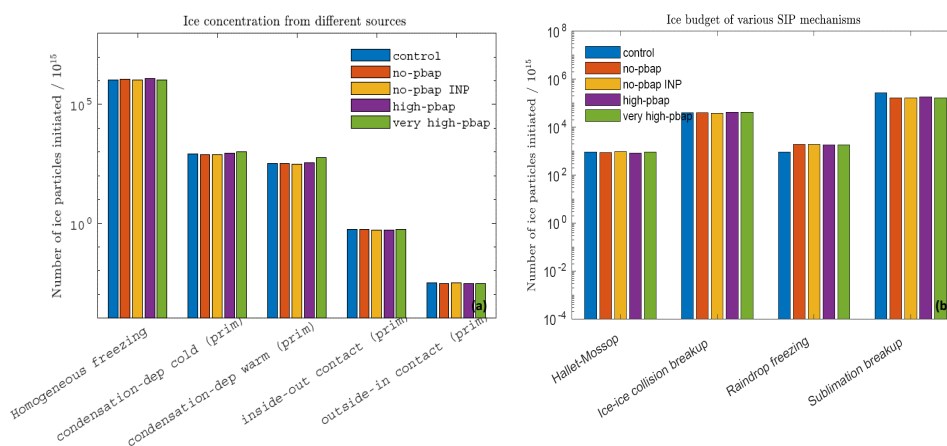


658 ‘control’ simulation and various sensitivity runs involving a change in total PBAP number
659 concentrations for in the stratiform region. Also shown is the temperature dependence of (e)
660 ice concentration from homogeneous freezing. The averaging conditions are mentioned at the
661 top of each figure. The ice number concentration from ultra high-pbap is also shown in panel
662 d. All the vertical profiles shown here are averaged for the whole domain.

663

664 -20°C is mostly associated with homogeneously nucleated ice downwelled in the mixed-
665 phase from above. At temperatures warmer than -15°C the predicted ice number
666 concentrations are less sensitive to the changes in bioaerosol loading. The biggest change
667 was predicted for the very high-pbap case, which had a 20% lower ice number concentration
668 than the control run. The changes in snow and graupel mass (including warm and cold
669 components) were lower than 10% (not shown here).

670



671

672 **Figure 10:** The number of ice crystals produced during the whole storm by (a) primary ice
673 nucleation mechanisms and homogeneous freezing as well as (b) various SIP mechanisms (as
674 shown in the legend box) per 10^{15} particles is shown for various sensitivity runs.

675

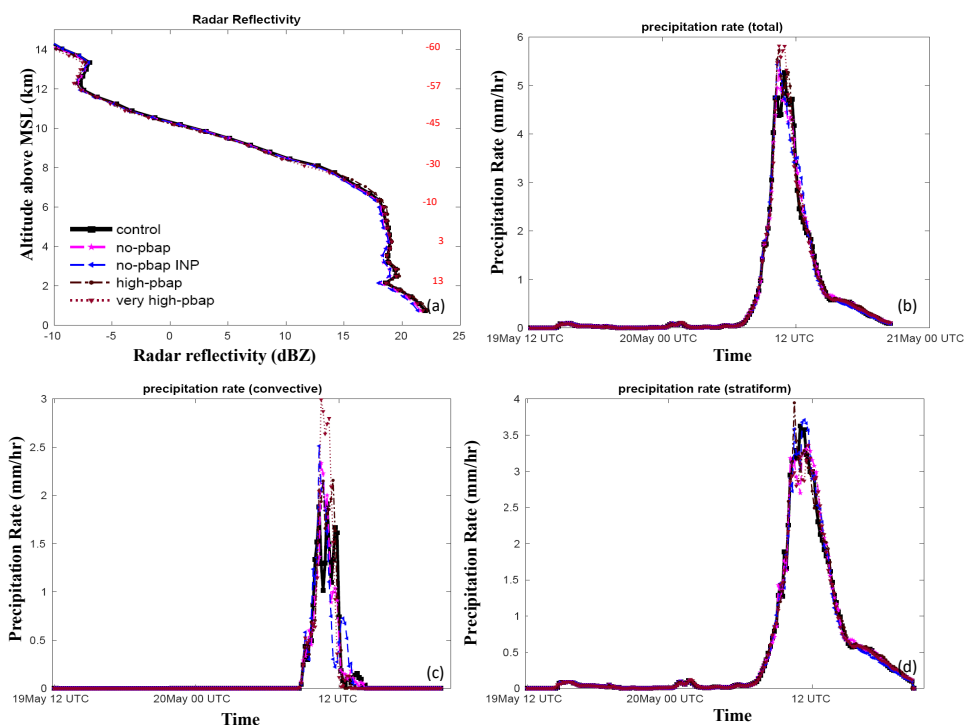
676 Figure 10 shows the number of ice particles generated by homogeneous nucleation, various
677 mechanisms of primary nucleation (10a), and secondary ice production (10b) per 10^{15} ice
678 particles for the entire storm. Homogeneous freezing dominates the ice production among the



679 three broad types of ice formation mechanisms (heterogeneous and homogeneous ice
680 nucleation, SIP). The maximum changes in ice nucleated through the primary ice mechanism
681 are noticed for the very high-pbap case and can be attributed to the 100-fold increase in all
682 PBAP loading. The high-pbap simulation predicted a 15% higher number of homogeneously
683 nucleated ice than the control run. The very high-pbap cases predicted about 80% more
684 primary ice crystals formed at temperatures warmer than -30°C . At temperatures colder than -
685 30°C , this case predicted 20% more primary ice crystals than the control run.

686

687 Figure 10b shows that among SIP mechanisms, the contributions of ice-ice collision breakup
688 and sublimation breakup are higher by an order of magnitude than the HM process and
689 raindrop fragmentation. However, the budget analysis (not shown in the plot) showed that
690 about 75% of the fragments associated with sublimation breakup are prone to evaporation,
691 making ice-ice collision breakup a major SIP mechanism. The estimated ice enhancement
692 ratio, which is a ratio between the number concentrations of total ice (excluding
693 homogeneous nucleation) and primary ice, is two orders of magnitude which indicates the
694 importance of SIP mechanisms. The budget analysis shows that overall, the perturbations in
695 bioaerosols resulted in very small changes (with maximum change $< 50\%$) in ice generated
696 by SIP mechanisms.



697

698 **Figure 11:** The vertical profiles of (a) radar reflectivity are shown for simulations involving
699 changes in PBAPs. (b) The temporal evolution of the total surface precipitation rate averaged
700 over the domain is also shown. The time series of surface precipitation rate averaged over the
701 domain is also shown separately for (c) convective and (d) stratiform regions. All the vertical
702 profiles shown here are averaged for the whole domain.

703

704

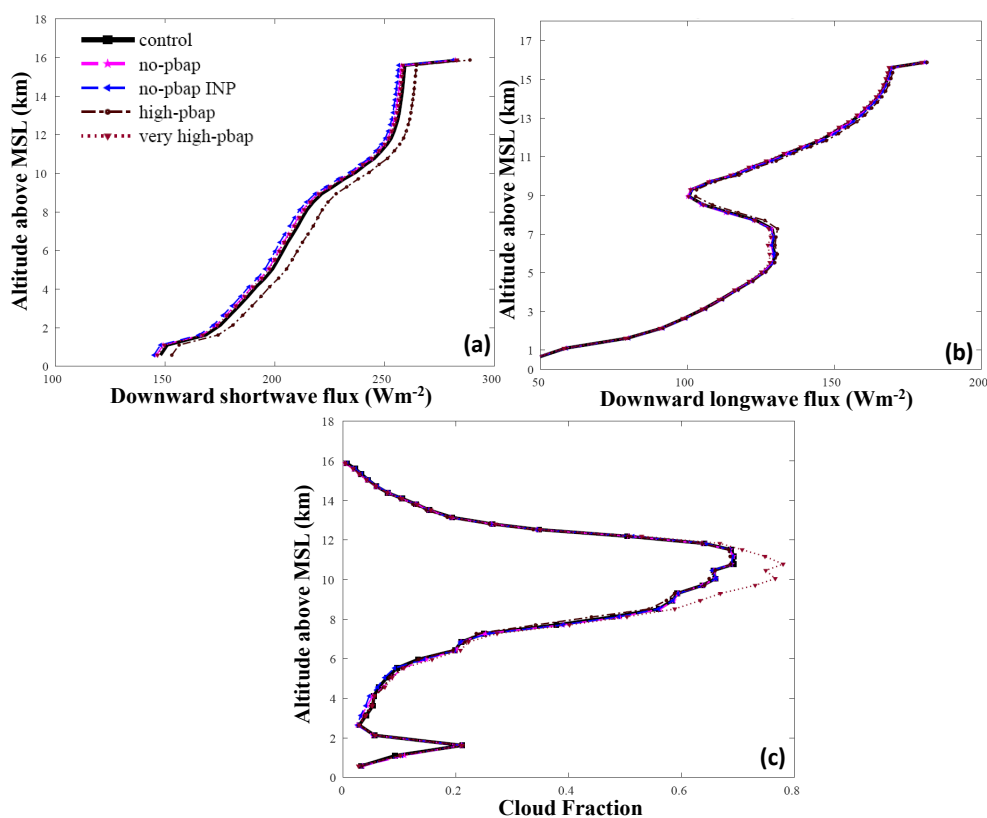
705 Figure 11a shows the effects of PBAPs on the simulated radar reflectivity for the whole
706 storm. When compared to the control run, there is no significant difference in the simulated
707 radar reflectivity of the perturbed simulations. Figure 11b depicts the sensitivity of the total
708 surface precipitation rate averaged over the domain to the changes in total PBAPs. Overall,
709 PBAPs have a minimal effect on surface precipitation, and it is most effective during the
710 period when precipitation is at its highest. The peak in surface precipitation rate is boosted by
711 about 10% in the very high-pbap cases compared to the control run. In remaining perturbed
712 simulations, changes in surface precipitation rate are less than 5% when compared with the



713 control run. The contribution from the stratiform component of rain is higher in the total
714 amount of rain (90%) as compared to the convective rain (remaining 10%) (see Fig. 11c and
715 11d). Convective rainfall is more sensitive to the changes in PBAPs than stratiform rainfall.
716 The increase in PBAPs by 100-fold results in a 50% higher peak of convective rainfall rate as
717 compared to the control run.

718

719



720

721 **Figure 12:** The domain averaged vertical profiles of downward components of (a) shortwave
722 flux, (b) longwave flux, and (c) cloud fraction for various sensitivity experiments. The data
723 shown here is an unconditional average over the whole duration and domain of each
724 simulation. All the vertical profiles shown here are averaged for the whole domain.

725

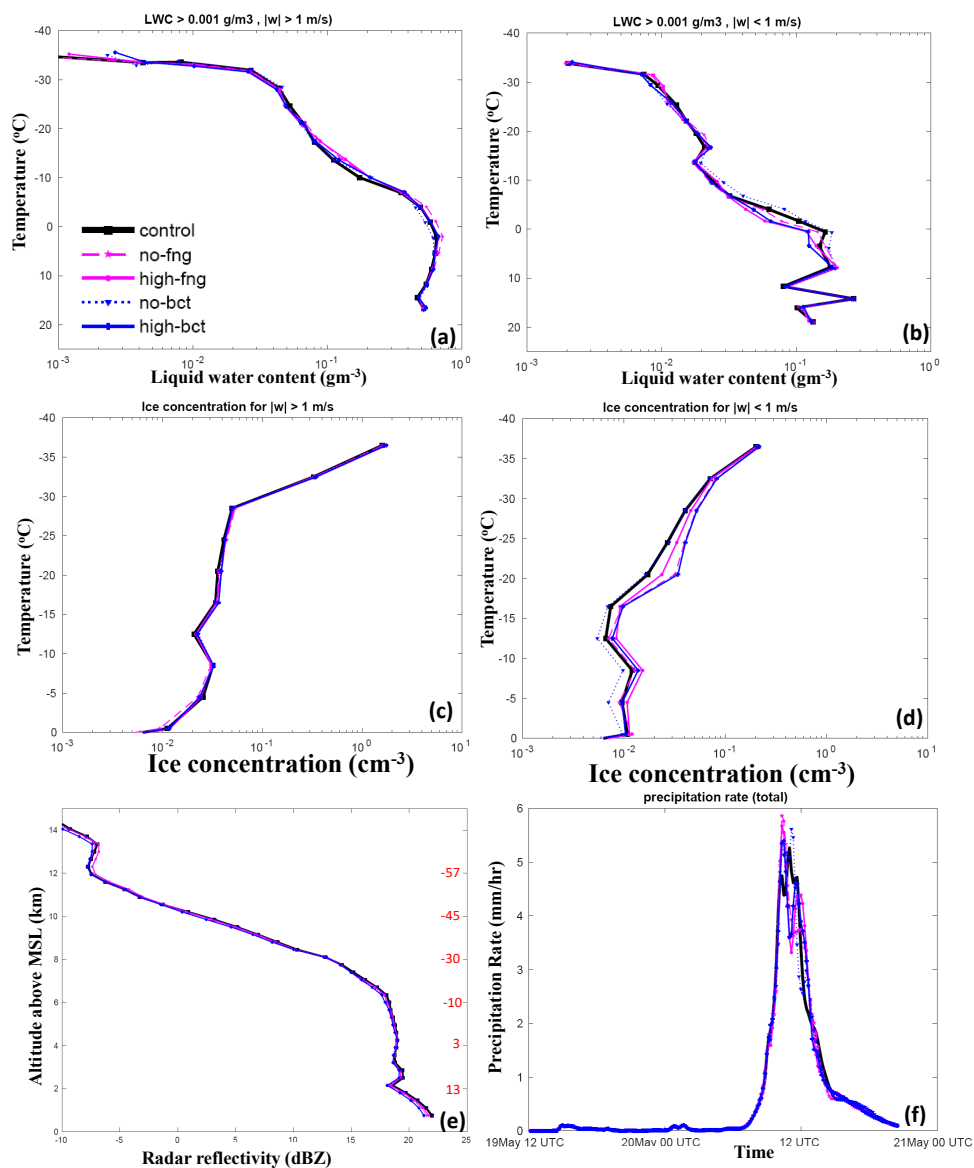


726 Figure 12 shows the domain averaged vertical profiles of shortwave, longwave fluxes, and
727 cloud fractions for the different sensitivity tests considered here. Among all the sensitivity
728 runs, only the high-pbap case showed a noticeable effect on shortwave flux, which was 3%
729 higher than the control run. The variations in longwave fluxes were less than 1%. The vertical
730 profiles of cloud fraction show that a 100-fold increase in total PBAP results in a 10% higher
731 cloud fraction between 8 and 12 km. The cloud fraction in other sensitivity runs was less
732 sensitive to the changes in PBAP loadings.

733

734 *5.2 Fungi and bacteria*

735 To investigate the role of various PBAP groups on cloud properties, model simulations were
736 carried out by eliminating most IN active PBAP groups including FNG and BCT from the
737 control ('*no-fng*' and '*no-bct*' cases) run and by multiplying their initial loadings at all levels
738 by a factor of 10 ('*high-fng*' and '*high-bct*' cases). The resulting sensitivity experiments are
739 then compared to the control run to understand how these PBAP groups may affect cloud
740 properties.



741

742 **Figure 13:** Temperature dependence of the liquid water content in (a) convective and (b)
 743 stratiform region for ‘control’ simulation, and various sensitivity runs involving changes in
 744 most active PBAP INPs including fungi and bacteria. Also shows is the total ice number
 745 concentration for (c) convective and (d) stratiform regions. The averaging conditions are
 746 mentioned at the top of each figure. The same simulation also shows (e) the vertical profiles
 747 of radar reflectivity. The temporal evolution of the (f) total surface precipitation rate averaged
 748 over the domain is also shown. All the vertical profiles shown here are averaged for the
 749 whole domain.

750



751 Figure 13 shows the effects of these two PBAP groups on cloud parameters including LWC,
752 ice number concentration as well as radar reflectivity, and surface precipitation for the
753 convective region and stratiform regions. Overall, the changes in LWC due to perturbations
754 in fungi and bacteria are very small for both convective ($< 10\%$) and stratiform regions ($<$
755 30%). In convective regions, the perturbation in fungi and bacteria does not affect the
756 predicted ice number concentrations with a maximum change of 5% with respect to the
757 control run. Comparatively, in the stratiform region, the changes in ice number
758 concentrations are more sensitive to the changes in the initial loading of fungi and bacteria.
759 At temperatures warmer than -30°C , the high-bct case predicted a 30% higher ice particle
760 number concentration than the no-bct case. For fungal particles, the transition from no- to
761 high- conditions resulted in 10% higher ice number concentration at temperatures warmer
762 than -15°C . At temperatures colder than -15°C , such transition resulted in a decrease in ice
763 number concentration by 5% .

764

765 The vertical profiles of the simulated radar reflectivity and surface precipitation rate for
766 various sensitivity experiments involving fungi and bacteria are shown in Figures 13e and
767 13b respectively. The maximum differences in radar reflectivity between the no-bct and high-
768 bct as well as no-fng and high-fng cases are smaller than 0.5 dBZ. Overall, the changes in
769 fungal and bacterial concentrations showed little difference in the surface precipitation rate
770 with respect to the control run ($< 10\%$).

771

772 **6. Results from sensitivity tests about secondary ice production**

773 The microphysical processes of AC include several SIP mechanisms such as breakup in ice-
774 ice collisions, raindrop freezing, sublimation breakup, etc. as mentioned in Section 3.1.
775 Various sensitivity experiments were conducted to evaluate the role of SIP mechanisms in



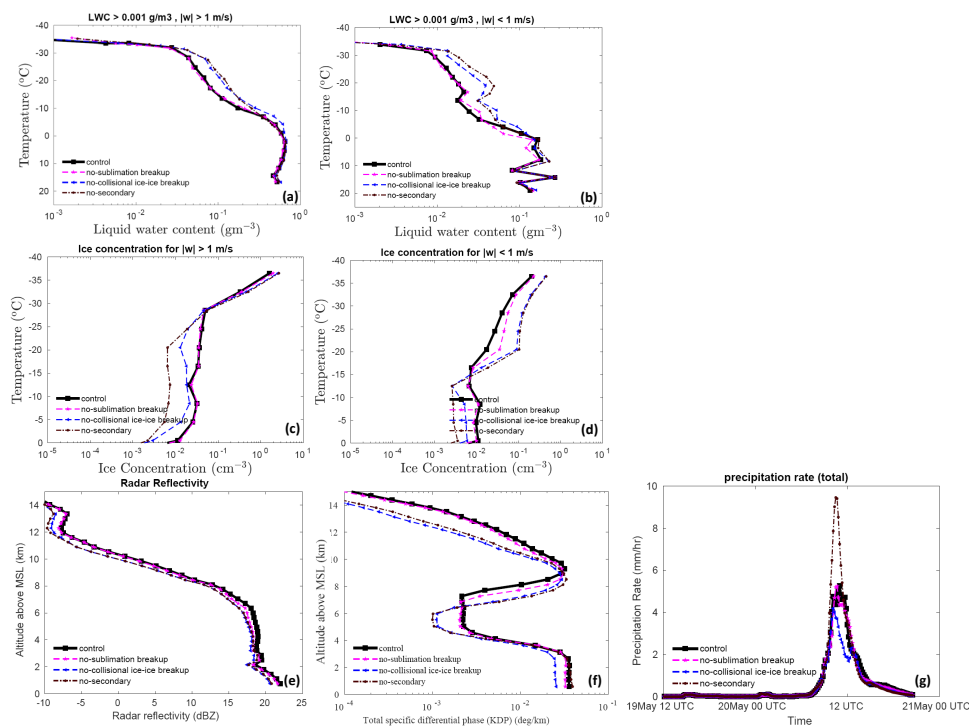
776 determining micro- and macrophysical parameters of the clouds. The SIP mechanisms,
777 including sublimation breakup and breakup in ice-ice collisions, were turned off in the
778 control run in the 'no-sublimation breakup' and 'no-collisional ice-ice breakup' simulations.
779 In the 'no-secondary' case, no SIP mechanisms were active.

780

781 The results from these sensitivity experiments are shown in Figure 14 for the convective
782 region of the simulated cloud. In the no-secondary and no-collisional ice-ice breakup case,
783 the LWC (Figure 14a) between -5 to -30°C is higher than in the control run for both in the
784 convective (~50%) and stratiform region (~80%). Lower ice number concentrations due to
785 the absence of SIP mechanisms may reduce the rate of conversion of liquid to ice via mixed-
786 phase processes, resulting in a higher LWC. The absence of all SIP mechanisms in the
787 mixed-phase region results in a larger cloud droplet diameter (between 0.5 to 1 µm) (not
788 shown here).

789

790 Comparing the no-SIP and control cases, the effect from the inclusion of SIP mechanisms is
791 to increase the average ice concentration by up to half an order of magnitude at temperatures



792

793 **Figure 14:** Temperature dependence of the liquid water content in (a) the convective and
 794 (b) the stratiform region for ‘control’ simulation and various sensitivity runs involving SIPs.
 795 The ice number concentration is also shown for the (c) convective and (d) stratiform regions.
 796 The averaging conditions are mentioned at the top of each figure. The vertical profiles of (e)
 797 radar reflectivity, (f) total specific differential phase are also shown for the same simulations.
 798 (g) The temporal evolution of the total surface precipitation rate averaged over the domain is
 799 also shown. All the vertical profiles shown here are averaged for the whole domain.

800

801 warmer than -25°C in the stratiform region. In the convective part, the absence of any SIP
 802 increased ice number concentration by half an order of magnitude at temperatures warmer
 803 than -15°C . At temperatures colder than this, the absence of SIP mechanisms resulted in
 804 higher ice number concentrations by a similar magnitude. The changes in ice number
 805 concentration in the no-collisional ice-ice breakup case are comparable with the no-secondary
 806 case. Compared to break up in ice-ice collisions, sublimation breakup has a lower impact (<
 807 40%) on the total ice number concentration in both convective and stratiform regions.



808

809 The changes in simulated radar reflectivity, total specific differential phase, and surface
810 precipitation rate with SIP mechanisms are shown in Figures 14e, 14f, and 14g, respectively
811 for the whole storm. Overall, the simulated radar reflectivity was 1 dBZ lower in the no-SIP
812 and no-collisional ice-ice breakup case than in the control run and can be attributed to the
813 overall increase in ice number concentration in the control run. The no-sublimation case
814 predicted slightly higher reflectivity than the control run. The absence of all SIPs resulted in
815 about a 100% decrease in the K_{DP} at a temperature colder than -40°C . Between -10°C to -
816 30°C , the absence of no-collisional breakup and no-secondary resulted in higher K_{DP} (half an
817 order of magnitude) values than the control run. The absence of all SIP mechanisms results in
818 a higher surface precipitation rate (75%) during the peak rainfall hour, which occurs around
819 11.30 UTC compared to the control run.

820

821

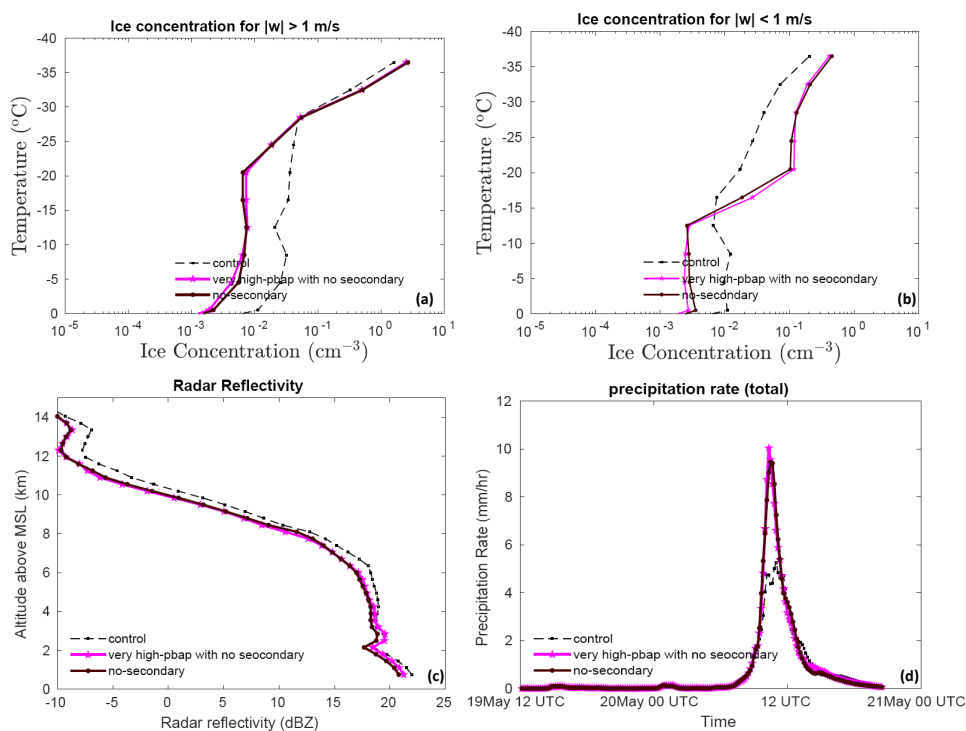
822 **7. Results about the influence of PBAP in the absence of SIP mechanisms**

823 Most SIP mechanisms are highly active at temperatures above -15°C . The majority of PBAPs
824 showed high ice nucleation activity occurs in this part of the cloud. Most of the ice
825 concentration in this part of the cloud is determined by various SIP mechanisms. Thus, the
826 SIP mechanisms may influence the role of PBAPs in altering cloud microphysical properties.
827 To investigate this aspect, an additional simulation was performed by eliminating all
828 secondary ice processes from the control run and multiplying the initial loading of all PBAP
829 groups by a factor of 100 (the '*very high-pbap with no SIP*' case). The results of this
830 simulation are then compared to the no-SIP case as shown in Figures 15.

831



832 In the absence of any SIP mechanisms, the 100-fold increase in bioaerosols resulted in
833 minimal effect on ice number concentration. Overall, the increase in bioaerosol loading by
834 100-fold resulted in less than 40% change in ice number concentration. This indicates that the
835 ice produced by various SIP mechanisms does not alter the effect of bioaerosols on-ice
836 number concentration in the simulated clouds. The changes in simulated radar reflectivity due
837 to a 100-fold increase in bioaerosols are negligible ($< 0.5\%$) (Figure 15c). The predicted
838 surface precipitation rate difference between very high-pbap with no-secondary and no-
839 secondary cases was lower than 5%.



840

841 **Figure 15:** The temperature dependence of ice number concentration for the control, very
842 high-pbap with no SIP and no-SIP simulations averaged for (a) convective and (b) stratiform
843 regions. The (c) vertical profile of radar reflectivity and the temporal evolution of (d) surface
844 precipitation rate is shown for the entire simulation. All the vertical profiles shown here are
845 averaged for the whole domain. All the vertical profiles shown here are averaged for the
846 whole domain.

847



848

849 **8. Summary and Conclusions**

850

851 A framework describing the ice nucleation activity of five major groups of PBAPs including
852 fungal spores, bacteria, pollen, viral particles, plant/animal detritus, algae, and their
853 respective fragments was provided by PT21. The ice nucleation activity of these major PBAP
854 groups in the EP was based on samples from the real atmosphere. The present study
855 implements this EP in AC and investigates the role of these five PBAP groups as INPs in
856 deep convective clouds. The high-resolution (2 km horizontally) simulations over a
857 mesoscale 3D domain (80 km wide) using AC elucidate the impact of these PBAP groups on
858 the cloud properties. A series of sensitivity experiments were conducted to test the impact of
859 PBAP groups on cloud properties.

860

861 A mid-latitude squall line that occurred on 20 May 2011 during MC3E over the US Southern
862 Great Plains is simulated with the model. The simulated number concentration of ice particles
863 showed good agreement (to within about 10%) with aircraft observations for the convective
864 clouds within the mesoscale system. The model predicted an ice number concentration
865 slightly lower than the aircraft observation in the stratiform region by up to a factor of 3
866 between -10 and -16°C but agreed better at all other levels. This factor of 3 is similar to the
867 uncertainty in the measurements due to various biases (e.g., Field et al. 2006), though it may
868 also be related to the radar reflectivity there being too low (by about 8 dBZ). Nevertheless, all
869 other simulated cloud microphysical parameters, radar reflectivity, and surface precipitation
870 rate were in good agreement with aircraft and ground-based observations. This confirms that



871 AC adequately reproduced the warm and cold microphysical processes of the simulated
872 squall line case.

873

874 Based on the sensitivity experiments, it can be concluded that PBAP INPs have only a limited
875 effect on the average state of the ice phase of the simulated clouds of this mesoscale
876 convective system. Any perturbation in the PBAP concentration resulted in changes in ice
877 number concentration by < 5% convective region and by < 60% in the stratiform region with
878 respect to the control run. Most of the changes in ice number concentration associated with
879 changes in PBAPs are controlled by their effect on homogeneous nucleation.

880

881 The weakness of the simulated impact from realistic PBAP fluctuations is explicable mostly
882 in terms of the low contribution from biological ice nucleation compared to non-biological
883 INPs to overall ice initiation. In terms of ice nucleation efficiency and onset temperatures,
884 each PBAP group has different ice nucleation properties. Based on vertical profiles of active
885 INPs (Figures 5), the overall contribution of activated INPs from all PBAP groups to the total
886 active INPs was ~1%. At -15°C, temperature, the predicted number of active INPs from dust
887 and black carbon were one order higher than PBAP INPs. At -30°C, the predicted INPs from
888 dust and black carbon were higher by one and two orders of magnitude, respectively, than
889 PBAP INPs. Overall, this resulted in low sensitivity of the average ice phase to the changes in
890 bioaerosol loading.

891

892 The ice production in the simulated cloud system at levels in the mixed-phase region (0 to -36
893 °C) is largely controlled by various SIP mechanisms of which the most important is the
894 breakup in ice-ice collisions. In the mixed-phase region, the total ice number concentration is



895 1-2 orders of magnitude higher than that of active INPs. The ice production associated with
896 SIP mechanisms is insensitive to the initial PBAP loading because SIP causes positive
897 feedback of ice multiplication with ice fragments growing to become precipitation size
898 particles that then fragment again. The explosive growth of ice concentrations occurs until
899 limited by other factors. In the absence of any SIP mechanisms in the model, the effect of
900 PBAP on the ice phase is again not significant. This indicates that SIP mechanisms were not
901 the main reason for the limited effect of PBAPs on the ice phase. A previous study by Phillips
902 et al. (2009) also noted an effect (by up to 4%) on surface shortwave and TOA longwave
903 radiation flux because of changes in PBAP number concentration. In our study, the changes
904 in PBAP loading caused smaller changes in simulated shortwave and longwave fluxes (<
905 3%).

906

907 Our results showed that surface precipitation is not significantly affected by changes
908 in PBAPs. A tenfold increase in all PBAPs resulted in less than 10% change in peak values of
909 surface precipitation rate. Using mesoscale model simulations, Phillips et al. (2009) reported
910 a 10% increase in accumulated surface precipitation associated with deep convective clouds
911 due to a 100-fold increase in biological particles. The perturbations in most active PBAPs
912 such as fungi and bacteria showed little change in surface precipitation.

913

914 Our results showed that the perturbations in PBAPs have only a limited effect (< 50% in
915 number and < 15% in mass) on the ice phase as well as on precipitation. This is consistent
916 with the findings of Hummel *et al.* (2018). Based on ensemble simulations of the regional
917 atmospheric model for Europe, they showed that the average ice crystal concentration by
918 biological INPs is not statistically significant, implying that PBAPs play only a minor role in



919 altering the cloud ice phase. Sesartic et al. (2012, 2013) showed that including fungi and
920 bacteria in the global climate model leads to minor changes ($< 0.5\%$) in ice water path, total
921 cloud cover, and total precipitation on a global scale.

922

923 To conclude, our simulations involving various changes in SIP indicated that peak surface
924 precipitation rate is highly sensitive ($\sim 75\%$) to the inclusion of SIP mechanisms. Eliminating
925 all SIP mechanisms from the model results in a higher and much narrower peak in surface
926 precipitation rate. However, even in the absence of SIP mechanisms, an increase in PBAP
927 number still causes only a minor increase in precipitation because the contribution of active
928 PBAP INPs to the total IN activity was only 1%.

929

930

931 **Code and data availability:** Data and the code for the empirical formulation of PBAPs are
932 available on request by contacting the corresponding author.

933 **Competing interests:** The authors declare no conflict of interest

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