

Hallett-Mossop rime splintering dims the Southern Ocean: New insight from global cloud-resolving simulations

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Abstract

In clouds containing both liquid and ice that have temperatures between -3°C and -8°C , liquid droplets collide with large ice crystals, freeze, and shatter, producing a plethora of small ice splinters. This process, known as Hallett-Mossop rime splintering, can cause clouds to reflect less sunlight and to have shorter lifetimes. Here, we use a novel suite of five global cloud-resolving models, which break up the Earth's atmosphere into columns with 2-4 km horizontal edges, to show that this microscale process has global implications. Simulations that include Hallett-Mossop rime splintering have reduced cumulus cloud cover over the Southern Ocean and reflect 12 Wm^{-2} less sunlight back to space over the same region, better matching satellite observed radiative fluxes. We evaluate simulated clouds using high-resolution visible images from the Himawari satellite, and radar reflectivities and two-dimensional images of cloud particles from the SOCRATES aircraft campaign. Cumulus clouds from simulations with Hallett-Mossop rime splintering included have more realistic cloud morphology, cloud vertical structure and ice crystal properties. We show that Hallett-Mossop rime splintering is an important control on cumulus cloud cover and cloud radiative effects over the Southern Ocean, and that including it in simulations improves model performance. We also demonstrate the key role that global cloud-resolving models can play in detangling the effects of clouds on Earth's climate across scales, making it possible to translate the behavior of tiny cloud particles (10^{-8} m^2) to their impact on the radiative budget of the massive Southern Ocean basin (10^{14} m^2).

1 **Hallett-Mossop rime splintering dims the Southern**
2 **Ocean: New insight from global cloud-resolving**
3 **simulations**

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8 **Key Points:**

- 9 • Including secondary ice production in simulations reduces summertime Southern
10 Ocean cloud shortwave forcing by 12 Wm^{-2}
- 11 • Cumulus cloud fraction is highly sensitive to the inclusion of secondary ice pro-
12 duction
- 13 • Global cloud-resolving simulations are invaluable for investigating climate impacts
14 of small-scale cloud processes

Abstract

In clouds containing both liquid and ice that have temperatures between -3°C and -8°C , liquid droplets collide with large ice crystals, freeze, and shatter, producing a plethora of small ice splinters. This process, known as Hallett-Mossop rime splintering, can cause clouds to reflect less sunlight and to have shorter lifetimes. Here, we use a novel suite of five global cloud-resolving models, which break up the Earth's atmosphere into columns with 2-4 km horizontal edges, to show that this microscale process has global implications. Simulations that include Hallett-Mossop rime splintering have reduced cumulus cloud cover over the Southern Ocean and reflect 12 Wm^{-2} less sunlight back to space over the same region, better matching satellite observed radiative fluxes. We evaluate simulated clouds using high-resolution visible images from the Himawari satellite, and radar reflectivities and two-dimensional images of cloud particles from the SOCRATES aircraft campaign. Cumulus clouds from simulations with Hallett-Mossop rime splintering have more realistic cloud morphology, cloud vertical structure and ice crystal properties. We show that Hallett-Mossop rime splintering is an important control on cumulus cloud cover and CREs over the Southern Ocean, and that including it in simulations improves model performance. We also demonstrate the key role that global cloud-resolving models can play in detangling the effects of clouds on Earth's climate across scales, making it possible to translate the behavior of tiny cloud particles (10^{-8} m^2) to their impact on the radiative budget of the massive Southern Ocean basin (10^{14} m^2).

Plain Language Summary

When clouds contain both liquid water and ice, liquid and frozen cloud particles compete with each other for water molecules. Frozen particles are larger than liquid particles so they fall faster under the influence of gravity. If frozen particles win the competition, they will efficiently remove water molecules from the cloud as they fall to the surface as snow. With too few water molecules, the cloud cannot persist and it dissipates. Here, we examine a set of five simulations that represent the entire atmosphere as a set of 42 million columns, each with 74 vertical levels. The five simulations use different combinations of formulas to control the rate at which frozen particles are produced within clouds. We find that the simulations which allow liquid particles to produce small frozen particles as they freeze and shatter, a process known as Hallett-Mossop rime splintering, can swing the competition towards ice within cumulus clouds over the Southern Ocean.

47 The causes the edges of the cumulus clouds to dissipate, which makes them narrower.
48 We find that these narrower clouds look more like the clouds that exist in the real at-
49 mosphere and so we conclude that Hallett-Mossop rime splintering should be included
50 in simulations.

51 **1 Introduction**

52 In mixed-phase clouds, liquid and frozen particles compete with each other for wa-
53 ter vapor (Bergeron, 1928). Glaciation occurs when frozen particles out-compete liquid
54 particles and the cloud condensed mass goes from predominantly liquid to predominantly
55 frozen. Glaciation can alter cloud optical properties (Sun & Shine, 1994), increase pre-
56 cipitation, and reduce cloud lifetime (Rogers & Yau, 1996).

57 Ice crystals in clouds form via three different pathways: on ice nucleating particles
58 (INPs), through the spontaneous freezing of water droplets at temperatures below -38°C ,
59 and via ice-liquid or ice-ice interactions (Pruppacher et al., 2010). The first two path-
60 ways are known as heterogeneous nucleation and homogeneous nucleation, respectively,
61 and, together, they make up primary ice production. The third process is known as sec-
62 ondary ice production or ice multiplication. Above -38°C , the number of ice crystals that
63 can be formed via primary ice production is capped by the number of INPs present in
64 the atmosphere. Brewer and Palmer (1949) speculated that secondary ice production “may
65 permit a water cloud to change to an ice cloud even though the number of ice-forming
66 nuclei initially present is inadequate.” In the seven decades that followed, numerous field
67 campaigns have observed ice crystal concentrations which are orders of magnitude higher
68 than INP concentrations (Field et al., 2016), attesting to the ubiquity of secondary ice
69 production in mixed-phase clouds, and confirming that speculation.

70 Evidence of secondary ice production has been observed at all latitudes (Koenig,
71 1963; Hobbs & Rangno, 1985; Rangno & Hobbs, 2001; Heymsfield & Willis, 2014; Tay-
72 lor et al., 2016; Huang et al., 2017; Ladino et al., 2017) but here we focus on low clouds
73 over the vast Southern Ocean, which control the albedo of the Southern Hemisphere (Vonder Haar
74 & Suomi, 1971) and are thus critically important for global climate. These clouds may
75 become brighter or more long-lived as the climate warms, if increased atmospheric tem-
76 peratures cause them to produce less ice and retain more supercooled water (Mitchell
77 et al., 1989). Constraining the magnitude of this negative climate feedback has remained

78 elusive due to an incomplete understanding of how ice particles form within Southern
79 Ocean mixed-phase clouds. Recently, several studies have pointed to a strong influence
80 of secondary ice production in these clouds. Indicators of secondary ice production within
81 Southern Ocean cumuli have been identified in two different in-situ datasets (Huang et
82 al., 2017; Scott, 2019), while two modelling studies have identified secondary ice produc-
83 tion as an important and underappreciated control on the Antarctic radiative budget via
84 the modulation of coastal stratus cloud properties (Young et al., 2019; Sotiropoulou et
85 al., 2021). However, to our knowledge, no study has quantified the global radiative im-
86 pact of secondary ice production in Southern Ocean clouds, which is a necessary step
87 in constraining global cloud-climate feedbacks. Here, we use hindcasts made with a suite
88 of global cloud-resolving models to show that secondary ice production is a strong con-
89 trol on Southern Ocean cloud albedo and is the largest source of inter-model variabil-
90 ity in Southern Ocean cloud radiative effects and model performance across our set of
91 simulations. We use in-situ aircraft observations and satellite data to show that more
92 realistic cloud morphologies, cloud microphysics and cloud radiative effects are found in
93 simulations that include secondary ice production.

94 **2 Datasets**

95 **2.1 Global Cloud-Resolving Simulations**

96 Our simulations are run using the global version of the System for Atmospheric Mod-
97 elling, or GSAM (Stevens et al., 2019). GSAM is anelastic and includes a comprehen-
98 sive land-surface model and a mixed-layer ocean. The simulations analyzed here are run
99 with CAM3 radiation (Collins et al., 2006) and a grid spacing of 4 km at the equator
100 and 2-3 km over the Southern Ocean. They have 4608 x 9216 horizontal grid cells and
101 74 vertical levels. Five-day simulations are run with five different microphysics schemes
102 but otherwise identical model setups. To allow the model time to spin up, we have ex-
103 cluded the first day of the simulation in all of the analyses shown in this study. Simu-
104 lated temperature and horizontal winds are initialized with and nudged to ERA5 reanal-
105 ysis with a timescale of 24 hours, tightly constraining the synoptic dynamics, as in Gettelman
106 et al. (2020) and Zhou et al. (2021). This has two key advantages. The first is that the
107 model output can faithfully be compared with coincident real-world observations. The
108 second is that it makes certain that differences in the model output between different
109 simulations necessarily arise from differences in the model microphysics.

110 We show animations of albedo from two GSAM simulations in Appendix A to give
111 a general sense of how clouds are represented in the simulations. GSAM simulations are
112 run from 0 UTC on February 16th to 0 UTC on February 21st 2018, which overlaps three
113 research flights from the Southern Ocean Clouds, Radiation, Aerosol Transport Exper-
114 imental Study (SOCRATES) (McFarquhar et al., 2020). Throughout this study, we eval-
115 uate GSAM using observations from two of these research flights, as described in Sec-
116 tion 2.2. In Appendix B, we show that concentrations of frozen particles imaged by the
117 Two-Dimensional Stereo Probe (2D-S) (Atlas et al., 2021) during SOCRATES are much
118 larger than typical INP concentrations from sea spray aerosol to show that secondary
119 ice production is active in SOCRATES-sampled clouds.

120 GSAM simulations are run with a suite of microphysics schemes that span the range
121 of complexity exhibited by bulk schemes. A useful proxy for the complexity and cost-
122 liness of bulk microphysics schemes is the number of prognostic variables that they use.
123 Table 1 lists the five different microphysics schemes used, the number of prognostic vari-
124 ables they use, and the modes of primary and secondary ice production that are active
125 within each scheme, in their implementations in GSAM. Note that the one-moment SAM
126 microphysics scheme, SAM1MOM, (Khairoutdinov & Randall, 2003) does not include
127 primary or secondary ice production because condensed cloud and precipitation mass
128 is partitioned into liquid and ice based on temperature. Therefore, it will not be men-
129 tioned below when the primary and secondary ice formation mechanisms in the other
130 schemes are described. All simulations are run with a fixed cloud liquid droplet num-
131 ber concentration of 100 cm^{-3} .

132 The clouds analyzed in this study occur at temperatures above -38°C so hetero-
133 geneous ice nucleation accounts for all primary ice production. As shown in Table 1, het-
134 erogeneous ice nucleation is represented differently in the five different microphysics schemes
135 used here. Thompson (Thompson et al., 2008), P3 (Morrison & Milbrandt, 2015), and
136 M2005 (Morrison et al., 2005) allow deposition nucleation (also referred to as deposition/condensation
137 nucleation), and immersion freezing for both cloud drops and raindrops. Immersion freez-
138 ing is parameterized following Bigg (1953) and produces far fewer particles than depo-
139 sition freezing in the clouds examined here. In all three implementations, the concen-
140 tration of INPs for deposition freezing is prescribed as a function of temperature follow-
141 ing the Cooper curve (Cooper, 1986, Figure B1). In M2005 and Thompson, deposition
142 nucleation occurs in two situations: 1) when ice supersaturation exceeds a fixed thresh-

Table 1. Characteristics of the five microphysics schemes used here

Microphysics Scheme	Prognostic Variables	Heterogeneous Nucleation	Secondary ice production
SAM1MOM	2	None	None
Thompson	7	Deposition	Hallett-Mossop rime splintering
P3	8	Deposition, Immersion, Raindrop Freezing	None
M2005	10	Contact, Deposition, Immersion, Raindrop Freezing	Hallett-Mossop rime splintering in high clouds
M2005 MOD	10	Contact, Deposition, Immersion, Raindrop Freezing	Hallett-Mossop rime splintering

old (8% in M2005, 25% in Thompson) or 2) in air that is saturated with respect to liquid and colder than -12°C . In P3, it occurs when the temperature is below -15°C and ice supersaturation is above 5%. M2005 also includes contact freezing at temperatures below -4°C , for which the concentration of INPs is prescribed as a function of temperature following the Meyers curve (Meyers et al., 1992, Figure B1). The Meyers curve prescribes higher concentrations of INPs than the Cooper curve for the same temperatures, and contact freezing operates in a wider range of atmospheric conditions. For these reasons, contact freezing dominates the primary ice production in M2005, and M2005 has much stronger primary ice production than either Thompson or P3.

The Thompson and M2005 schemes include parameterizations of Hallett-Mossop rime splintering (HMRS), a type of secondary ice production that occurs at temperatures between -3°C and -8°C (Hallett & Mossop, 1974). HMRS involves large frozen particles, small droplets and large droplets and can be conceptualized as a two-step process (Field et al., 2016). In the first step, small droplets freeze onto large frozen particles and largely retain their shapes to create an icy shell with narrow protrusions. In the second step, large droplets freeze and shatter when they come into contact with those small pro-

159 trusions. HMRS is parameterized in M2005 and Thompson as a single-step process; at
 160 each microphysics time step, the bulk mass of supercooled water rimed onto large frozen
 161 particles is computed as a function of droplet mass, and the size and number of the large
 162 frozen particles. The number of ice splinters produced is then parameterized as a func-
 163 tion of the rimed mass and the temperature. The number of splinters produced per unit
 164 of rimed mass maximizes at -5°C and decays towards -3°C and -8°C .

165 While M2005 and Thompson permit HMRS over the same range of temperatures,
 166 M2005 has additional, stricter requirements for when HMRS can occur. As described
 167 in Atlas et al. (2020), M2005 requires either that the droplet mass is greater than 0.5
 168 g kg^{-1} or that the rain mass is greater than 0.1 g kg^{-1} , and that either graupel, snow
 169 or ice mass exceeds 0.1 g kg^{-1} . While virtually all mixed-phase clouds within the Hallett-
 170 Mossop temperature range will satisfy the conditions for HMRS in the Thompson scheme,
 171 low clouds rarely satisfy the stricter conditions for HMRS in M2005. Thus, the fifth sim-
 172 ulation that we run is a modified version of M2005, which we refer to as ‘M2005 MOD’,
 173 with all mass thresholds removed from the HMRS parameterization so that the process
 174 can occur in low clouds. Throughout this study, we refer to M2005 as a simulation with-
 175 out HMRS because we focus primarily on low clouds but that characterization is not ac-
 176 curate for all cloud regimes.

177 **2.2 Satellite and Aircraft Observations**

178 We use CERES level 3 data (Doelling et al., 2013; NASA/LARC/SD/ASDC, 2017),
 179 obtained from the NASA Langley Research Center Atmospheric Science Data Center,
 180 to constrain global CREs. This dataset has $1^{\circ} \times 1^{\circ}$ horizontal resolution and hourly tem-
 181 poral resolution. We use high-resolution observations from Himawari (Smith & Minnis,
 182 2020) to qualitatively compare cloud morphology between GSAM and the real world.
 183 Retrievals of broadband shortwave albedo from Himawari from the spatial and tempo-
 184 ral ranges of interest were performed by NASA Langley’s Satellite Cloud and Radiation
 185 Property retrieval System (SatCORPS). This data has 0.5 to 2 km horizontal resolution
 186 and either 10-minute or 30-minute time resolution depending on the time of day.

187 We use in-situ airborne observations and remote sensing data collected by the NSF/NCAR
 188 Gulfstream-V HIAPER (High-Performance Instrumented Airborne Platform for Envi-
 189 ronmental Research) during SOCRATES to evaluate representations of the boundary

190 layer and low clouds in GSAM. GSAM simulations overlap three research flights from
 191 SOCRATES: 1) RF11 on February 17th, 2018, which sampled open-cell cumulus between
 192 -9°C and 0°C , 2) RF12 on February 17th-18th, 2018, which sampled stratocumulus be-
 193 tween -7° and -3° , and 3) RF13, on on February 19th-20th, 2018, which sampled stra-
 194 tocumulus between -3°C and 2°C . Since RF13 did not sample clouds within the Hallett-
 195 Mossop temperature range, we only compare GSAM output with observations from RF11
 196 and RF12. Simulated boundary layer thermodynamics are evaluated with in-situ tem-
 197 perature and water vapor measurements (EOL, 2019), and simulated cloud microphysics
 198 are evaluated with single-particle phase classifications of particles with projected areas
 199 $\geq 2500 \mu\text{m}^2$ that were imaged by the Two-Dimensional Stereo Probe (2D-S) (Atlas et
 200 al., 2021), particle concentrations from the Cloud Droplet Probe (CDP) (EOL, 2019),
 201 and radar reflectivities from the W-band HIAPER cloud radar (EOL, 2018).

202 **3 Results**

203 **3.1 Comparisons with satellite**

204 Figure 1 compares simulated global cloud radiative effects (CREs) at top of atmo-
 205 sphere from all five GSAM simulations with retrieved CREs from CERES. We coarsen
 206 the GSAM output to $1^{\circ} \times 1^{\circ}$ in order to compare with CERES. Here and throughout
 207 the study, red lines indicate simulations with HMRS and blue lines indicate those with-
 208 out. The orange box highlights the Southern Ocean region. In the tropics, longwave ra-
 209 diation dominates the variance and the bias in simulated CREs. In the Southern Ocean,
 210 shortwave radiation dominates the biases. All simulations have too much reflected short-
 211 wave radiation over the Southern Ocean, opposite to the bias historically exhibited by
 212 most climate models (Trenberth & Fasullo, 2010). GSAM simulations with HMRS have
 213 an average SW CRE of 7.5 W m^{-2} over the Southern Ocean ($42^{\circ}\text{S} - 60^{\circ}\text{S}$), compared
 214 to 19.4 W m^{-2} for simulations without HMRS. Thus, simulations with HMRS have a
 215 reduced SW bias by about 12 Wm^{-2} . The blue and red solid lines represent M2005 and
 216 M2005 MOD, respectively, and we can be certain that the differences between those two
 217 simulations are solely due to the activation of HMRS in low clouds.

218 We compare visible albedo between the Himawari satellite and GSAM to investi-
 219 gate if cloud cover and cloud brightness can explain the trends seen in the radiative bi-
 220 ases (Figure 2). Throughout this study, figures with grey borders show comparisons that

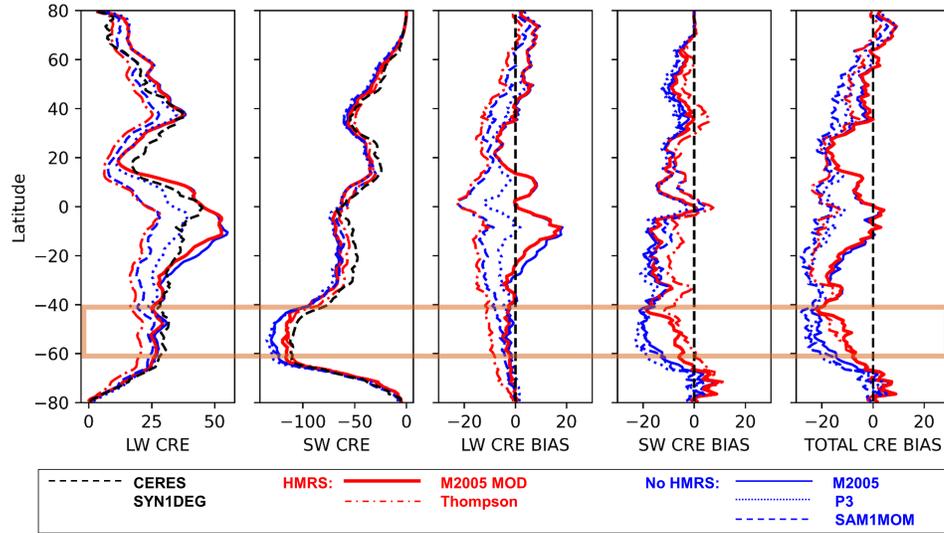


Figure 1. Zonal average longwave and shortwave cloud radiative effects (CREs) at top of atmosphere are shown for the five GSAM simulations and for CERES SYN1DEG in the first and second columns, respectively, averaged over the four-day period from 0 UTC on February 17th to 0 UTC on February 21st, 2018. Biases in longwave, shortwave, and total simulated CREs compared to CERES (GSAM-SYN1DEG) are shown in the third, fourth and fifth columns, respectively. The orange box highlights the latitudinal range of interest (42° - 60° S). Simulations with Hallett-Mossop rime splintering included (“HMRS”) are in red, and simulations without it (“No HMRS”) are in blue.

221 are limited to the SOCRATES study region, because they evaluate GSAM using Himawari
222 and/or in-situ observations. In Figure 2, our analysis is restricted to the region for which
223 high resolution Himawari data is available (40° - 68° S, 130° - 165° E). The snapshots shown
224 in Figures 2a and 2b are coincident with SOCRATES flights RF11 and RF12, respec-
225 tively, and the SOCRATES flight tracks are overlaid on the albedo maps in red. RF11
226 sampled open cell cumulus while RF12 sampled the stratocumulus cloud deck south of
227 the open cell region. Simulations with HMRS are in the top row and simulations with-
228 out it are in the bottom row.

229 Low stratus clouds and high/frontal clouds are too bright in all simulations in both
230 snapshots. This may account for the overall bright bias seen in all simulations. Simu-
231 lations without HMRS are also too bright in the open cell regions, due to the larger cloudy
232 area surrounding each cumulus cloud center. In other words, all simulations have a sim-
233 ilar number of cumulus clouds, but simulations without HMRS have more horizontal cloud
234 cover associated with each cumulus cloud. Simulations with HMRS are dimmer in the
235 open cell regions due to smaller cumulus cloud cover and bear a closer resemblance to
236 the snapshots from the Himawari satellite.

237 This qualitative comparison suggests the cumulus regime accounts for the smaller
238 shortwave CREs and better model performance exhibited by simulations with HMRS
239 (Figure 2). However, if there is variance in the brightness of the stratiform clouds be-
240 tween the different simulations, it may be difficult to detect by eye from the albedo maps.
241 In order to verify that cumulus clouds are responsible for the discrepancy in shortwave
242 CREs across the models, we create a “cloud mask” to classify model output into four
243 different cloud regimes: low cloud fraction (including clear sky), high clouds, low stra-
244 tus clouds, and low cumulus clouds. We use the output of the M2005 MOD simulation
245 to create the cloud mask, as discussed in Appendix C. Because all five GSAM simula-
246 tions are nudged to the same reanalysis, they typically simulate the same cloud morpholo-
247 gies in the same locations, and the mask based on M2005 MOD can be faithfully applied
248 to all simulations. In the following, locations where the simulated cloud morphology dis-
249 agrees with observations will be noted. However, as our focus is on the impact of changes
250 in simulated cloud due to differing microphysical parameterizations, it is valuable to use
251 a mask that behaves consistently across the simulations.

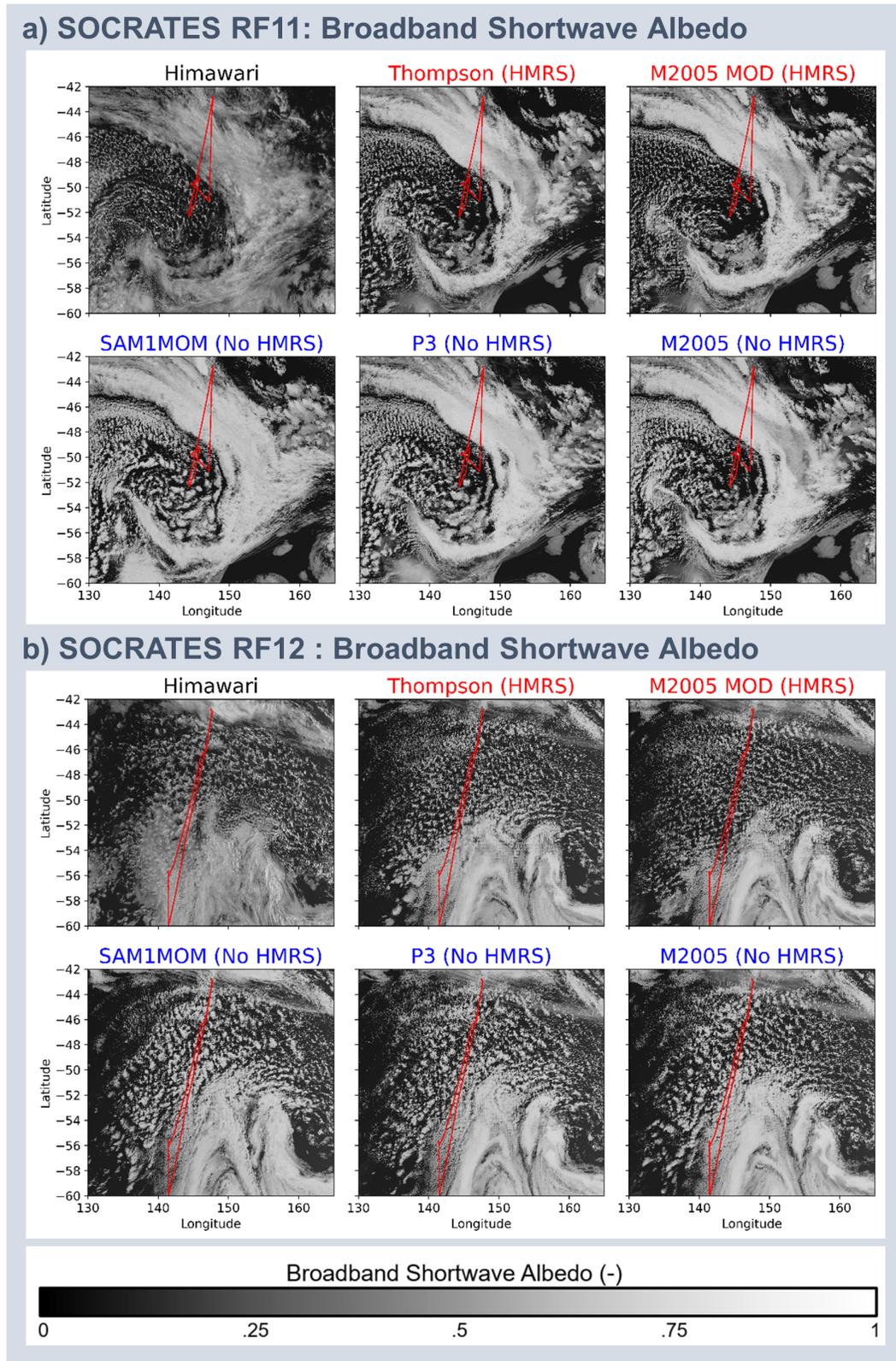


Figure 2. Coincident snapshots of visible cloud albedo are shown for the Himawari satellite and the five GSAM simulations for the SOCRATES sampling region. The snapshot in (a) at 3 UTC on February 17th 2018 coincides with SOCRATES RF11 which sampled open cell cumulus and the snapshot in (b) at 6 UTC on February 18th 2018 coincides with SOCRATES RF12 which sampled stratocumulus. Red lines indicate SOCRATES flight tracks. Simulations with

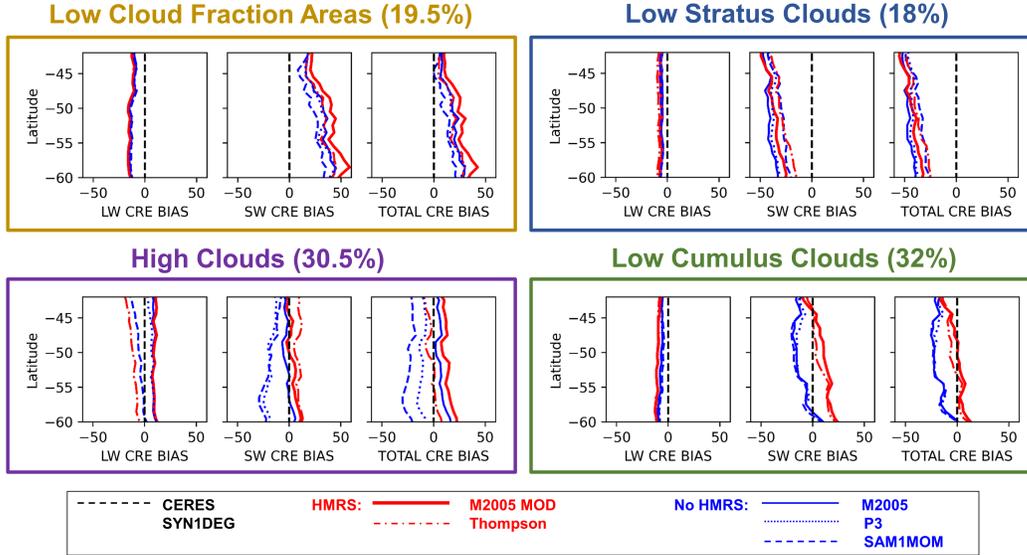


Figure 3. Zonal average biases in longwave, shortwave, and total simulated CREs compared to CERES (GSAM-SYN1DEG) are shown for the latitudinal range of interest (40° - 60° S), for the four cloud regimes averaged over the four-day period from 0 UTC on February 17th to 0 UTC on February 21st, 2018. Simulations with Hallett-Mossop rime splintering included (“HMRS”) are in red, and simulations without it (“No HMRS”) are in blue.

252 We compare CREs between CERES and GSAM separately for the four different
 253 cloud regimes in Figure 3. Low cloud fraction areas are too dim in all simulations be-
 254 cause these regions contains clouds that have broken up prematurely in the GSAM sim-
 255 ulations. For example, in the lower right corner of the albedo maps in Figure 2a, the five
 256 GSAM models simulate mainly clear sky but Himawari shows a robust stratiform cloud
 257 deck.

258 Low stratus regions are too bright in all simulations and the magnitude of the bias
 259 is similar for simulations with and without HMRS. Differences in high clouds account
 260 for most of the variance in longwave CREs. There is also a large spread in the shortwave
 261 CREs for high clouds, with SAM1MOM and P3 exhibiting substantial bright biases. How-
 262 ever, these bright biases cannot be explained by the exclusion of HMRS because M2005
 263 does not exhibit the same bias. We will evaluate the representation of high clouds in GSAM
 264 more thoroughly in a future study.

265 Low cumulus clouds are the only cloud regime for which the shortwave CREs ex-
266 hibit an unambiguous dependence on the inclusion of HMRS. Low cumulus clouds from
267 simulations with HMRS have smaller biases in the total CRE, due in part to compen-
268 sating biases in the shortwave and longwave CREs. The examples in Figure 2 show that
269 simulations with HMRS simulate cumulus cloud morphologies that agree well with Hi-
270 mawari, but cumulus cloud regimes from those simulations reflect less shortwave radi-
271 ative than what was observed. This is because there are regions where stratocumulus decks
272 have broken up into open cell cumulus prematurely in the simulations and, in those re-
273 gions, we are comparing dimmer simulated cumulus clouds with brighter observed stra-
274 tocumulus clouds. An example of such a region can be seen in Figure 2b, around 140°E
275 and between 52°S and 54°S. In that region, all five simulations produce open cell cumu-
276 lus but Himawari shows that a stratocumulus was present in reality. Because the cloud
277 mask is based on the output of M2005 MOD, this area is classified as low cumulus.

278 Thus far, we have shown that simulations with HMRS simulate dimmer Southern
279 Ocean clouds and agree better with satellite observations from Himawari and CERES.
280 Furthermore, the inclusion of HMRS in simulations primarily affects shortwave CREs
281 by reducing the cumulus cloud fraction. At this point, it is reasonable to ask whether
282 these improvements are related to increased physical realism of the simulated clouds or
283 caused by offsetting errors. In other words, have simulated cumulus clouds improved for
284 the right reasons? In the following section, we address this question by using in-situ ob-
285 servations to determine if simulations with HMRS have more realistic cloud microphysics.

286 **3.2 Comparisons with aircraft data**

287 Throughout this section, we compare GSAM output with in situ and remote sens-
288 ing data from two SOCRATES flights, RF11 and RF12. To evaluate simulated micro-
289 physics, liquid and frozen simulated particle size distributions (PSDs) are evaluated with
290 SOCRATES-observed PSDs from the CDP and 2D-S instruments (Figure 4). The two
291 SOCRATES flights shown in Figure 2, RF11 and RF12, both sampled clouds within the
292 HMRS temperature range. RF12 sampled a stratocumulus cloud deck that fully over-
293 lapped the HMRS temperature range, collecting 1142 seconds of particle observations
294 at all heights within the cloud. RF11 collected 176 seconds of particle observations, mainly
295 at a single height within a cumulus cloud field. Thus, we use RF12 for this analysis. The
296 simulations are sampled along the flight track, meaning that nearest neighbors in space

297 and time are extracted from the model output for every 10 seconds of in-situ sampling,
298 and averaged together. Although CREs of stratiform cloud regions are not sensitive to
299 the inclusion of HMRS in these simulations (Figure 3), we hypothesize that the inclu-
300 sion of HMRS affects cloud microphysical properties of stratiform clouds in largely the
301 same way that it affects cumulus clouds. Atlas et al. (2020) developed an LES case study
302 based on RF12 and found that simulated PSDs from M2005 MOD had more large frozen
303 particles and agreed better with SOCRATES observations than M2005. That result is
304 replicated here, as M2005 MOD has concentrations of large frozen particles that are three
305 orders of magnitude higher than in M2005 and has the best agreement with SOCRATES
306 PSDs out of the five microphysics schemes (Figure 4).

307 Thompson has more large frozen particles than any of the simulations except M2005
308 MOD and partially replicates the observed large frozen particle mode. The three sim-
309 ulations without HMRS do not have enough large frozen particles. We note that SAM1MOM
310 does not assume size distributions for cloud droplets and cloud ice so those two hydrom-
311 eteor classes are excluded from the SAM1MOM PSDs. However, since most of the large
312 frozen particles are contained in the snow and graupel classes, we do not expect the ab-
313 sence of cloud ice to affect our conclusions. Thus, the inclusion of HMRS in the simu-
314 lations improves simulated cloud microphysics in stratiform clouds within the HMRS tem-
315 perature range. Although in situ sampling of PSDs during RF11 was not statistically
316 representative, remote sensing from the HIAPER cloud radar gathered a larger sample
317 of data that can be used to evaluate the simulated cloud and precipitation features in
318 the cumulus regime.

319 We compute synthetic radar reflectivities for the five GSAM simulations using a
320 modified version of QUICKBEAM (Haynes et al., 2007). QUICKBEAM is Fortran-based
321 software that uses microphysics information to estimate radar reflectivities. It has been
322 implemented to run online with M2005 and Thompson microphysics within the SAM LES
323 (Atlas et al., 2020). Here, we develop a modified version of QUICKBEAM written in Python
324 which runs offline and is compatible with all five microphysics schemes used here. The
325 PSDs specified within the microphysics schemes and shown in Figure 4 are used to com-
326 pute radar reflectivities. Thus, radar reflectivities computed for SAM1MOM do not in-
327 clude contributions from cloud droplets and cloud ice. The original QUICKBEAM soft-
328 ware accounts for attenuation due to gases and hydrometeors along the radar path for
329 a space-based or ground-based radar, but is not included here because the changing po-

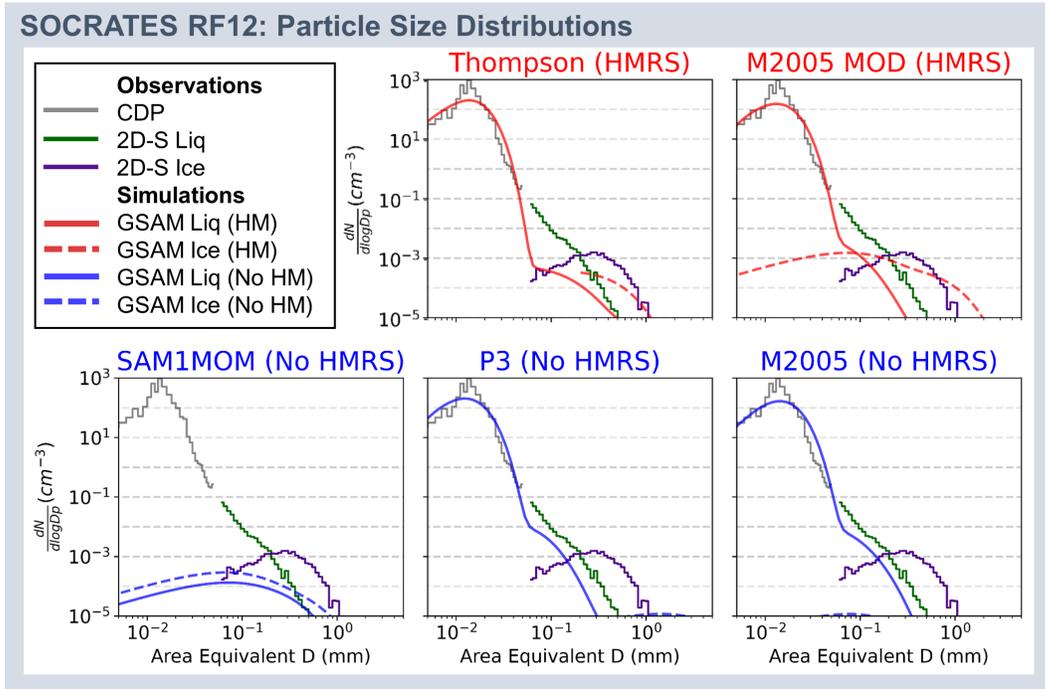


Figure 4. Liquid (solid lines) and frozen (dashed lines) simulated PSDs are compared with observed PSDs from SOCRATES. Grey lines are observed droplet concentrations from the CDP, green lines are observed large droplet concentrations from the 2D-S, and purple lines are observed frozen particle concentrations from the 2D-S. Simulations with Hallett-Mossop rime splintering included (“HMRS”) are in red in the top row, and simulations without it (“No HMRS”) are in blue in the bottom row. Simulations have been sampled along the flight track.

330 sition of the aircraft complicates the calculation of attenuation and because attenuation
331 is usually small for low clouds and for an airborne radar which profiles clouds at a close
332 range. However, we caution that high radar reflectivities (> 5 dBZ) may be overestimated
333 due to unaccounted for hydrometeor attenuation.

334 Figure 5 compares radar reflectivities and cumulus cloud texture between GSAM
335 output and observations from Himawari and the HIAPER radar. The observed cumu-
336 lus and the simulated cumulus from the simulations with HMRS are dominated by nar-
337 row banded structures (second column of Figure 5), resembling the “gravel” pattern from
338 Stevens et al. (2020). Simulated cumulus from the simulations without HMRS feature
339 wider and puffier cumulus clouds (fourth column of Figure 5), resembling the “flowers”
340 pattern from Bony et al. (2020). The synthetic reflectivities from all simulations are dom-
341 inated by columns, which span the height from the sea surface to the cloud tops, indi-
342 cating the presence of precipitation; these are associated with cumulus cloud centers. The
343 simulations without HMRS have more nonprecipitating “interstitial” cloud with low re-
344 flectivities, indicating the absence of large particles, between the columns. The intersti-
345 tial cloud contributes to the footprint of the simulated cumulus clouds in the albedo maps.

346 To test whether the HMRS or no-HMRS simulations better match the radar ob-
347 servations, we compute a vertical profile of “reflectivity fraction”, which is the fraction
348 of radar reflectivities that are greater than -50 dBZ, as a function of height. The reflec-
349 tivity fraction is sensitive to both cloud and precipitation but it can be used a proxy for
350 cloud fraction in the upper part of the boundary layer. We use model output from the
351 area shown in the albedo maps in Figure 5 (49 - 52.5° S, 144 - 146.5° E), and from 3, 4 and
352 5 UTC on February 17th 2018, to compute the reflectivity fraction for the simulations.
353 The observed reflectivity fraction is based on the reflectivities shown in the upper left
354 panel in Figure 5.

355 Figure 6a shows observed and simulated profiles of temperature, specific humid-
356 ity and reflectivity fraction. Profiles of temperature and humidity verify that the sim-
357 ulations are accurately representing the thermodynamics of the boundary layer. The ob-
358 servations and the two simulations with HMRS have peaks in reflectivity fraction at about
359 1 km, which is below the Hallett-Mossop temperature range. The three simulations with-
360 out HMRS have higher reflectivity fractions throughout the Hallett-Mossop temperature
361 range. Furthermore, they have peaks in reflectivity fraction that are either partially or

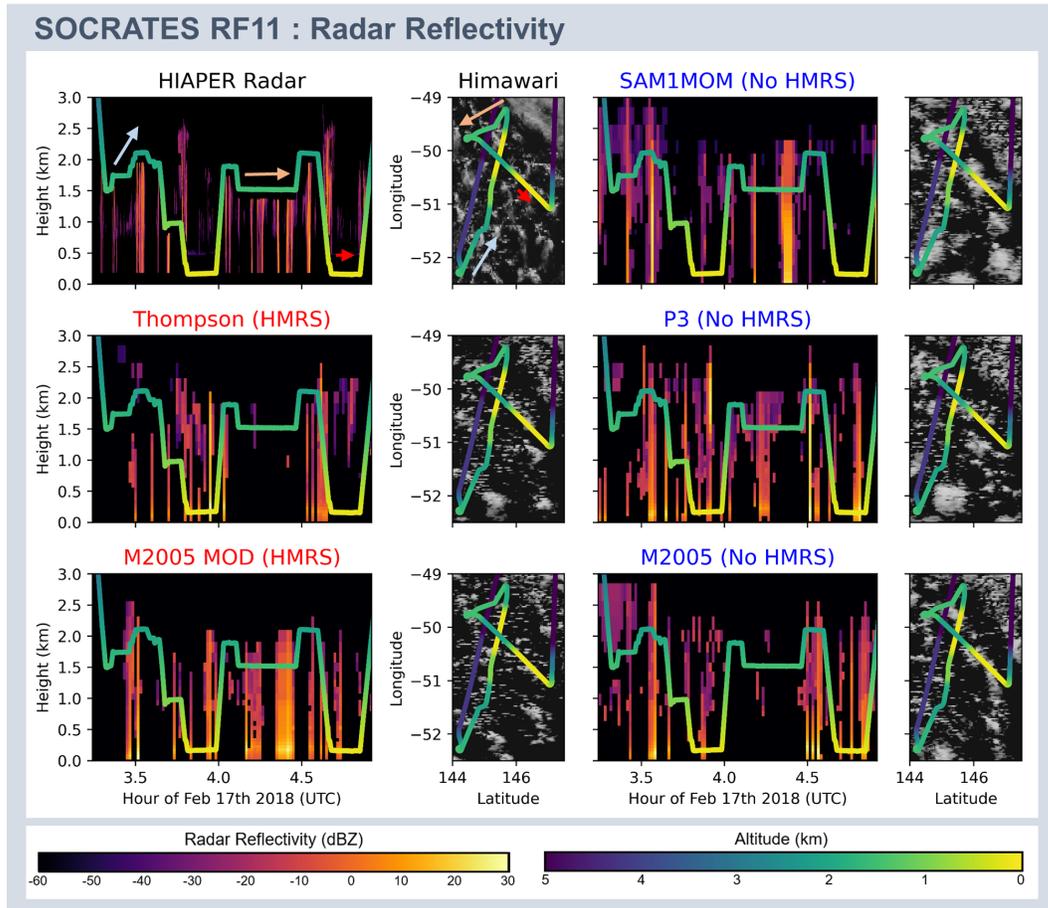


Figure 5. Radar reflectivity and visible albedo maps are shown for the observations (top left) and the five GSAM simulations. Flight tracks, color-coded by aircraft altitude, are shown on all plots. The arrows on the two plots in the upper left corner show how to align the two perspectives. Simulations with Hallett-Mossop rime splintering included (“HM”) are in the left column (below the observations), and simulations without it (“No HMRS”) are in the right column.

362 fully within the Hallett-Mossop temperature range. Their higher reflectivity fractions
363 are due to the presence of interstitial clouds. The lower observed reflectivity fraction in-
364 dicates that little interstitial cloud was present in the real atmosphere, in agreement with
365 the simulations with HMRS.

366 We showed in Section 3.1 that simulations with HMRS have more realistic short-
367 wave CREs over the Southern Ocean because they have less cumulus cloud cover. In this
368 section, we showed that these simulations also have more realistic cloud microphysics.
369 They simulate a larger number of frozen particles, in agreement with in-situ observations,
370 and less interstitial cloud, in agreement with the HIAPER radar. These comparisons also
371 illuminated how HMRS reduces cumulus cloud fraction and are summarized in Figure 6b.

372 In all simulations, updrafts occur preferentially in moist columns within the bound-
373 ary layer and form cumulus cloud centers. Interstitial cloud can detrain into the drier
374 air between the cumulus cloud centers. The synthetic reflectivities show that the inter-
375 stitial clouds are thin and lack large particles, making them more susceptible to glacia-
376 tion than the cumulus cloud centers. In the simulations with HMRS, ice formation is ef-
377 ficient and ice crystals out-compete liquid particles within the detrained interstitial cloud,
378 glaciating it and drastically reducing its lifetime. In the simulations without HMRS, ice
379 formation is suppressed and the interstitial clouds persist, reflecting excessive sunlight
380 back to space over the Southern Ocean.

381 4 Conclusions

382 We have analyzed supercooled boundary layer clouds over the Southern Ocean in
383 a unique suite of five meteorologically-nudged global cloud-resolving simulations of a 5-
384 day period during Feb. 2018 during the SOCRATES field study. The simulations dif-
385 fer only in their cloud microphysics parameterizations, which include several widely-used
386 schemes. We compared them with satellite, in-situ and radar observations. Our key find-
387 ing is that, in simulations that include Hallett-Mossop rime splintering (a form of sec-
388 ondary ice production), shallow cumulus clouds glaciates over the Southern Ocean, de-
389 creasing the surrounding cover of detrained “interstitial” cloud and reducing the regionally-
390 averaged shortwave cloud radiative effect biases by 12 Wm^{-2} . Simulations including sec-
391 ondary ice production have more frozen particles in low clouds, consistent with in-situ

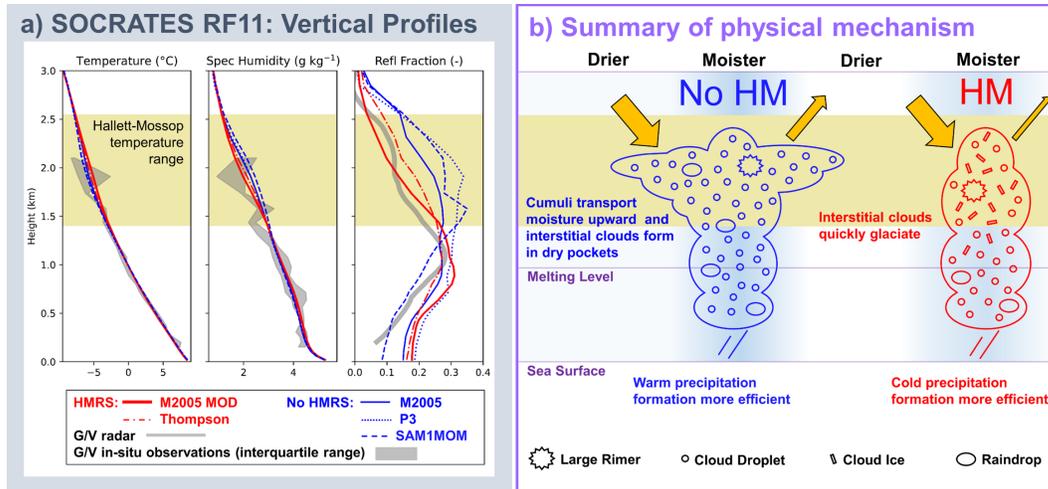


Figure 6. a) Profiles of temperature, specific humidity, and reflectivity fraction are shown, from left to right. Simulations with Hallett-Mossop rime splintering included (“HMRS”) are in red and simulations without it (“No HMRS”) are in blue. b) Schematic showing how HMRS reduces cumulus cloud fraction.

392 observations, and their reduced overall albedo over the Southern Ocean agrees better with
 393 satellite observations.

394 Our study illuminates the importance of secondary ice production to Earth’s cli-
 395 mate. It exemplifies how global cloud-resolving simulations, which can simulate super-
 396 cooled cloud fields much more realistically than conventional coarse-grid climate mod-
 397 els, are uniquely suited for evaluating and testing the global or regional climate impacts
 398 of small-scale microphysical processes. Large eddy simulations, which have very fine grids
 399 but only cover small computational domains, are also useful for investigating the detailed
 400 microphysics in boundary-layer clouds like those that have complex small-scale internal
 401 circulations. However, because they simulate only a small portion of the Earth, they can-
 402 not be used to comprehensively simulate interactions between clouds and meteorology
 403 to allow a reliable quantification of global or regional cloud radiative effects.

404 Other types of secondary ice production may be important in the real atmosphere.
 405 Recently, Luke et al. (2021) argued that freezing fragmentation (also referred to as droplet
 406 shattering), is stronger than rime splintering within the Hallett-Mossop temperature range
 407 in Arctic clouds. Sotiropoulou et al. (2020) combined Hallett-Mossop rime splintering
 408 with another secondary ice production process, mechanical breakup during ice-ice col-

409 lisions, to explain observed ice crystal concentrations in Antarctic clouds. The glacia-
410 tion mechanism we outline here is not specific to Hallett-Mossop rime splintering; any
411 secondary ice process that produces frozen particles at a sufficient rate would glaci-
412 ate the thinner parts of cumulus clouds. However, the detailed mechanism and temperature
413 range are important to parameterizing secondary ice production and quantifying its ra-
414 diative impacts.

415 Intriguingly, ice crystal concentrations were also observed to be greater than INP
416 concentrations in SOCRATES-sampled clouds at temperatures down to -25°C , where
417 Hallett-Mossop rime splintering is not active, indicating that other secondary ice pro-
418 cesses are active at colder temperatures (Appendix B). We acknowledge that our sim-
419 ulations do not account for secondary ice production outside of the Hallett-Mossop tem-
420 perature range, and we hope that future studies illuminate how secondary ice produc-
421 tion influences Southern Ocean clouds at temperatures below -8°C .

422 We have not altered the formulation of primary ice production in the various mi-
423 crophysics schemes used in our simulations to account for low INP concentrations ob-
424 served over the Southern Ocean. As discussed in Appendix B, primary ice production
425 operates much more strongly in some schemes than in others, and the assumed INP con-
426 centrations are much larger than observed over the Southern Ocean. However, across our
427 suite of simulations, the clouds are far less sensitive to differences in primary ice produc-
428 tion between the microphysics schemes than they are to the inclusion of secondary ice
429 production. This conclusion differs from Vergara-Temprado et al. (2018), who found that
430 Southern Ocean clouds are highly sensitive to the representation of primary ice produc-
431 tion.

432 In addition to the bright bias demonstrated by simulations without secondary ice
433 production in Southern Ocean cumulus clouds, all simulations demonstrate a bright bias
434 in Southern Ocean stratiform clouds. We hope that future studies can investigate its cause
435 and propose effective solutions.

436 In light of its substantial effect on Earth's climate, we encourage climate model-
437 ers to account for secondary ice production in supercooled clouds such as those preva-
438 lent over the Southern Ocean and constrain those models with the increasing number
439 of observational analyses becoming available. In particular, we recommend removing the
440 mass thresholds from the parameterization of Hallett-Mossop rime splintering in the M2005

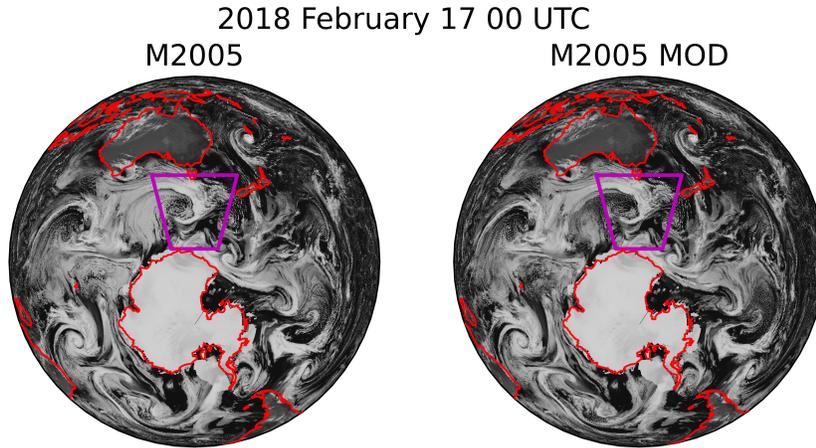


Figure A1. Animations of albedo from the M2005 and M2005 MOD simulations for the last four days of the simulation. The purple box highlights the SOCRATES study region.

441 microphysics scheme. We note that global (and boundary-forced regional) cloud-resolving
 442 models are an attractive testbed to capture the interplay of large scale meteorology and
 443 small scale cloud processes and we encourage their use in future investigations of cloud-
 444 climate interactions.

445 **Appendix A Animations of albedo from GSAM**

446 We show animations of visible albedo from two GSAM simulations, M2005 and M2005
 447 MOD, in Figure A1. Animations show the last four days of the simulation (February 17th
 448 to February 20th). GSAM resolves clouds on a wide range of scales and simulates real-
 449 istic cloud morphologies. M2005 has more cloud cover than M2005 MOD within the cu-
 450 mulus cloud regions.

451 **Appendix B Evidence for Secondary Ice Production during SOCRATES**

452 Figure B1 compares concentrations of ice crystals from SOCRATES with typical
 453 concentrations of INPs from sea spray aerosol for the range of temperatures sampled dur-
 454 ing SOCRATES. Marine aerosols are the predominant source of INPs over the South-
 455 ern Ocean (McCluskey et al., 2019), so we expect that concentrations of INPs during the
 456 period of study are within the range of marine INP concentrations that have been re-
 457 ported in the literature. Ice crystal concentrations are estimated from 2D-S imaged par-

458 ticles that have been classified as frozen (Atlas et al., 2021), and INP concentrations are
459 sourced from two studies. DeMott et al. (2016) includes INP concentrations from sea spray
460 generated in wave channel experiments as well as from ambient marine boundary layer
461 air sampled in the Carribean, the Pacific Ocean and the Arctic. McCluskey et al. (2018)
462 quantified INPs within the Southern Ocean boundary layer using filter samples collected
463 as part of The Clouds, Aerosols, Precipitation, Radiation, and atmospherIc Composi-
464 tion Over the southeRN ocean (CAPRICORN) campaign. We show data from DeMott
465 et al. (2016) using a grey shaded region and we show data from McCluskey et al. (2018)
466 using green stars, and we note that this representation is similar to Figure 2 in McCluskey
467 et al. (2018). We show ice crystal concentrations from the two SOCRATES flights that
468 we analyze throughout this study, which overlap the time period of the simulations.

469 Between -25°C and -20°C , median ice crystal concentrations are at the upper limit
470 of reported marine INP concentrations and mean ice crystal concentrations are greater
471 than reported marine INP concentrations. Between -20°C and -15°C , median and mean
472 ice concentrations are one to two orders of magnitude higher than the upper limit of re-
473 ported marine INP concentrations. These temperature ranges do not overlap the Hallett-
474 Mossop temperature range, implying that other types of secondary ice production are
475 occurring. There was not enough in-cloud data from these two flights in the range of -
476 15°C to -10°C to include in this analysis, and there is little data on INP concentrations
477 at temperatures above -10°C .

478 The large discrepancies between ice crystal concentrations and typical marine INP
479 concentrations suggest that secondary ice production is the dominant mechanism of ice
480 crystal formation in this subset of the SOCRATES dataset.

481 We also show two curves commonly used to prescribe INP concentrations for pri-
482 mary ice production in Figure B1. The Meyers curve (Meyers et al., 1992) is used for
483 contact nucleation, which is active in the M2005 scheme. The Cooper curve (Cooper,
484 1986) is used for deposition nucleation, which is active in all microphysics schemes used
485 here except SAM1MOM. Both curves drastically overestimate the concentration of INPs
486 over the Southern Ocean. If secondary ice production is excluded, and primary ice pro-
487 duction is implemented using one or both of these curves, then the number of ice crys-
488 tals will be overestimated at the coldest temperatures examined here ($T < -20^{\circ}\text{C}$) and
489 underestimated at temperature above -10°C .

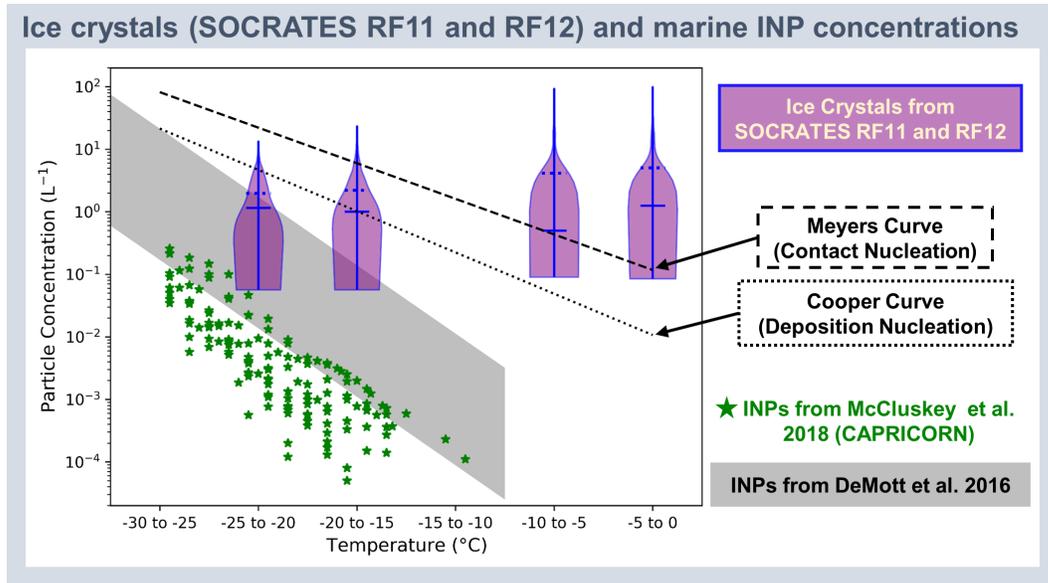


Figure B1. Purple and orange violin plots show distributions of ice crystal and INP concentrations, respectively, for 5° temperature brackets. Solid horizontal lines show median concentrations and dashed horizontal lines shown mean concentrations. Flights RF11 and RF12 from the SOCRATES dataset are used. The black lines show two different estimates of INP concentrations, modelled as exponential functions of temperature. Dashed and dotted lines show the Meyers curve, used for contact nucleation, and the Cooper curve, used for deposition/condensation nucleation, respectively.

Appendix C Cloud regimes

We categorize output from the five GSAM simulations and satellite data from CERES into four different cloud regimes using cloud fraction and cloud top height from the M2005 MOD simulation. We do not use satellite observations to develop the cloud mask because both CERES and Himawari report cloud fractions near 1.0 in both open cell cumulus and stratiform cloud regions, so it cannot be used to distinguish these two regimes.

In order to avoid removing the majority of the data from our comparisons, we use GSAM output to develop the cloud mask. Because all five GSAM simulations are nudged by the same reanalysis, they typically simulate the same cloud morphologies in the same locations. Hence we can use the output from one simulation to develop a cloud mask that can be applied to all five simulations.

M2005 MOD and Thompson have the highest contrast in cloud fraction between stratus and cumulus regions (Figure 2), which make them good candidates for developing the cloud mask. However, Thompson stores most of its frozen hydrometeor mass in the snow class, and precipitation is not taken into account in the calculation of cloud top height in GSAM so Thompson often has a low bias in the cloud top height. Thus, we base the cloud mask on M2005 MOD.

We break the model output into $1^\circ \times 1^\circ$ boxes (to compare with the CERES data, which is gridded at this resolution) and compute the average cloud fraction and cloud top height for each box. We use the decision tree in Figure C1 to classify each box into one of four regimes. We have hourly output from GSAM and we produce cloud masks for each hour. Examples of the resulting cloud mask for times during RF11 and RF12 are shown in Figure C2, overlaid on the albedo maps shown in Figure 2 as a visual check on the cloud mask. Note that we expect the best correspondence for M2005 MOD, which was used to derive the mask.

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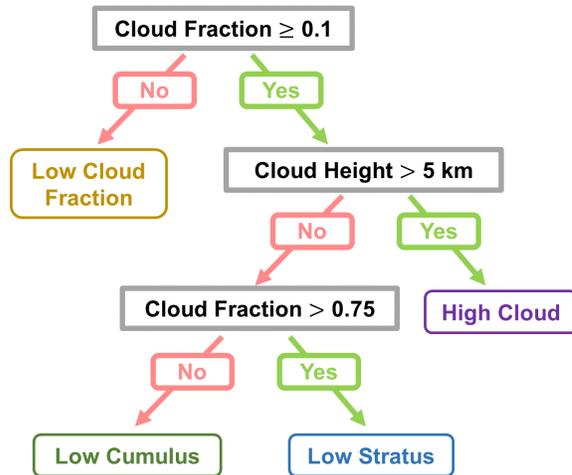


Figure C1. Decision tree for classifying $1^\circ \times 1^\circ$ boxes into four different cloud regimes based on model output.

520 all the SOCRATES scientists and NCAR EOL for collecting the data and helping us in-
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 529 sharing his discussion of M2005 and P3 schemes.

530 Data availability: CERES products used here can be downloaded at [https://ceres.larc](https://ceres.larc.nasa.gov/data/)
 531 [.nasa.gov/data/](https://ceres.larc.nasa.gov/data/). Himawari products and all SOCRATES measurements used here can
 532 be downloaded at https://data.eol.ucar.edu/master_lists/generated/socrates/.
 533 GSAM model output cannot be made available due to the experimental nature of the
 534 simulations and the large storage space required. Our modified QUICKBEAM code can
 535 be accessed at <https://github.com/cloudatlas-on-github/QUICKBEAM-python>.

536 References

537 Atlas, R. L., Bretherton, C. S., Blossey, P. N., Gettelman, A., Bardeen, C.,

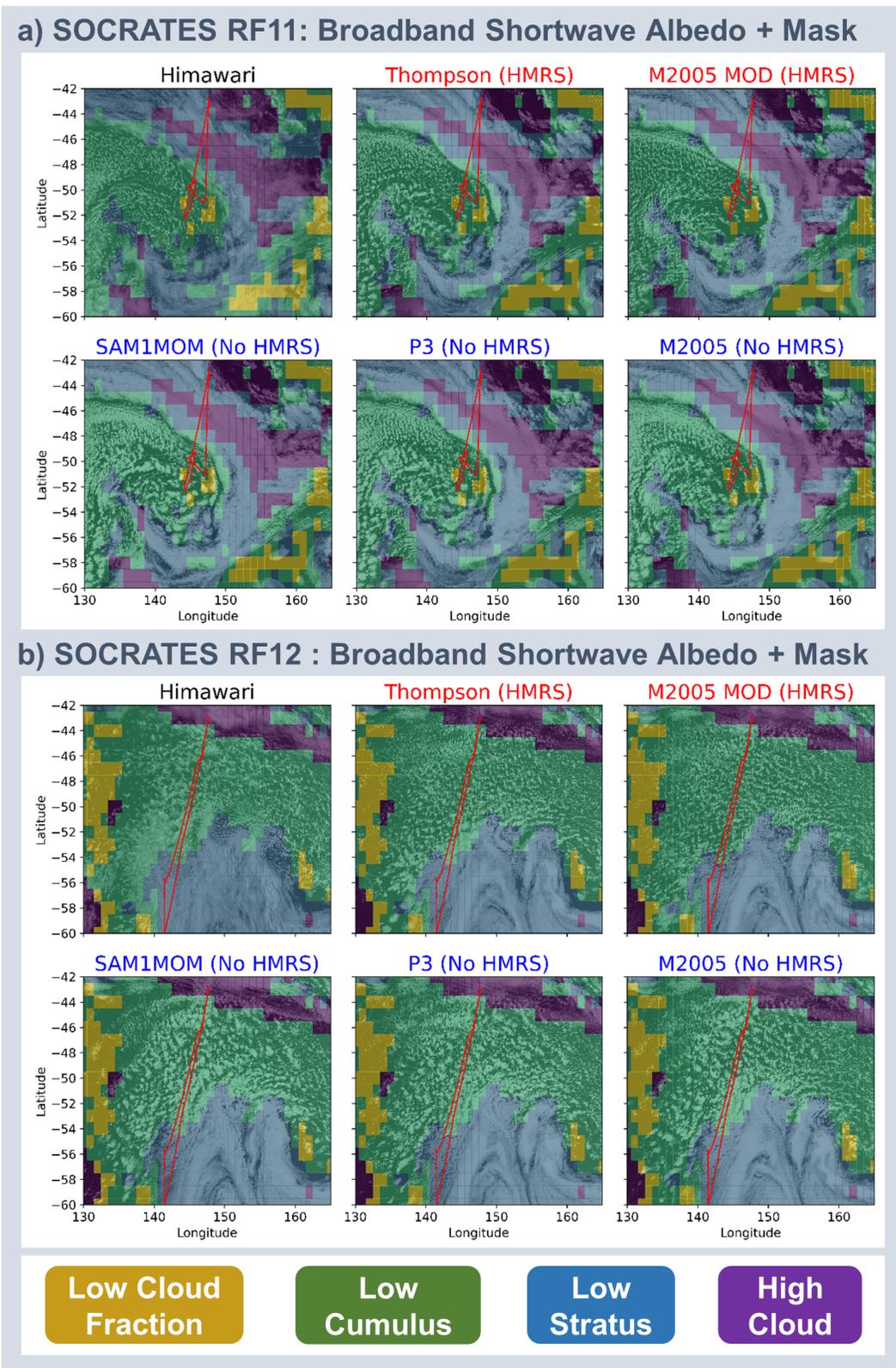


Figure C2. Cloud mask overlaid on the snapshots of visible albedo shown in Figure 2.

- 538 Lin, P., & Ming, Y. (2020). How well do large-eddy simulations and
 539 global climate models represent observed boundary layer structures and
 540 low clouds over the summertime southern ocean? [Journal Article]. *Jour-*
 541 *nal of Advances in Modeling Earth Systems*, *12*(11), e2020MS002205. doi:
 542 <https://doi.org/10.1029/2020MS002205>
- 543 Atlas, R. L., Mohrmann, J., Finlon, J., Lu, J., Hsiao, I., Wood, R., & Diao, M.
 544 (2021). The University of Washington Ice-Liquid Discriminator (UWILD)
 545 improves single particle phase classifications of hydrometeors within Southern
 546 Ocean clouds using machine learning [preprint]. *Atmospheric Measurement*
 547 *Techniques Discussions, in review*, e2020EA001241. doi: [https://doi.org/](https://doi.org/10.5194/amt-2021-123)
 548 [10.5194/amt-2021-123](https://doi.org/10.5194/amt-2021-123)
- 549 Bergeron, T. (1928). ber die dreidimensional verknüpfende wetteranalyse [Journal Ar-
 550 ticle]. *Geophys. Norv.*, *5*(6), 1-111.
- 551 Bigg, E. K. (1953). The formation of atmospheric ice crystals by the freezing of
 552 droplets [Journal Article]. *Quarterly Journal of the Royal Meteorological Soci-*
 553 *ety*, *79*(342), 510-519. doi: <https://doi.org/10.1002/qj.49707934207>
- 554 Bony, S., Schulz, H., Vial, J., & Stevens, B. (2020). Sugar, gravel, fish, and flowers:
 555 Dependence of mesoscale patterns of trade-wind clouds on environmental con-
 556 ditions [Journal Article]. *Geophysical Research Letters*, *47*(7), e2019GL085988.
 557 doi: <https://doi.org/10.1029/2019GL085988>
- 558 Brewer, A. W., & Palmer, H. P. (1949). Condensation processes at low tempera-
 559 tures, and the production of new sublimation nuclei by the splintering of ice
 560 [Journal Article]. *Nature*, *164*(4164), 312-313. doi: 10.1038/164312a0
- 561 Collins, W. D., Rasch, P. J., Boville, B. A., Hack, J. J., McCaa, J. R., Williamson,
 562 D. L., ... Zhang, M. (2006). The formulation and atmospheric simula-
 563 tion of the community atmosphere model version 3 (cam3) [Journal Ar-
 564 ticle]. *Journal of Climate*, *19*(11), 2144-2161. Retrieved from [https://](https://journals.ametsoc.org/view/journals/clim/19/11/jcli3760.1.xml)
 565 journals.ametsoc.org/view/journals/clim/19/11/jcli3760.1.xml doi:
 566 [10.1175/jcli3760.1](https://doi.org/10.1175/jcli3760.1)
- 567 Cooper, W. A. (1986). Ice initiation in natural clouds [Journal Article]. *Meteorologi-*
 568 *cal Monographs*, *21*(43), 29-32. doi: 10.1175/0065-9401-21.43.29
- 569 DeMott, P. J., Hill, T. C. J., McCluskey, C. S., Prather, K. A., Collins, D. B., Sul-
 570 livan, R. C., ... Franc, G. D. (2016). Sea spray aerosol as a unique source of

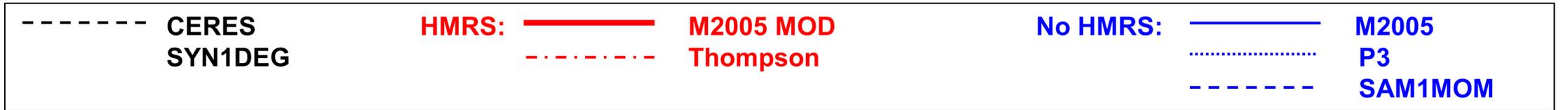
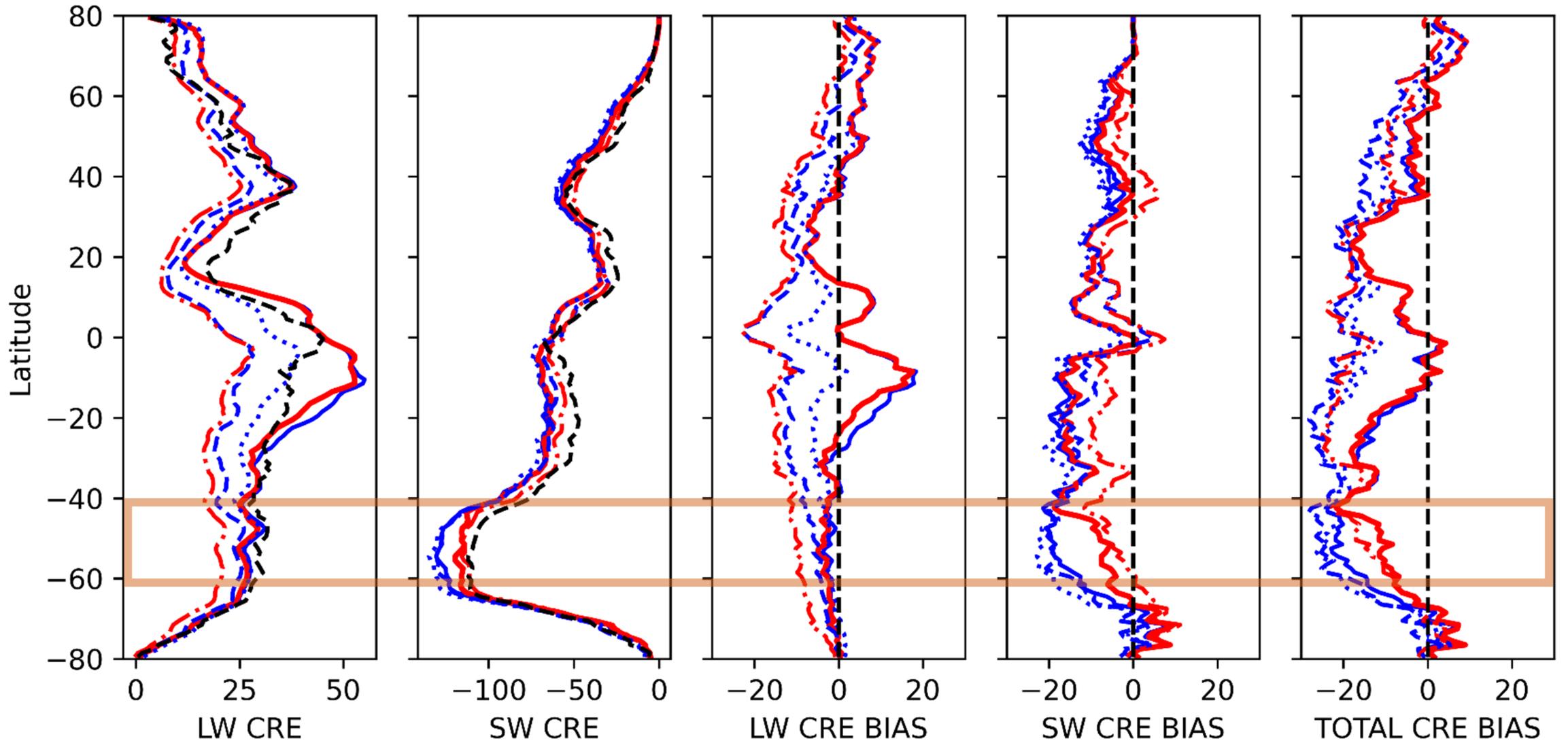
- 571 ice nucleating particles [Journal Article]. *Proceedings of the National Academy*
 572 *of Sciences*, 113(21), 5797-5803. Retrieved from [https://www.pnas.org/](https://www.pnas.org/content/pnas/113/21/5797.full.pdf)
 573 [content/pnas/113/21/5797.full.pdf](https://www.pnas.org/content/pnas/113/21/5797.full.pdf) doi: 10.1073/pnas.1514034112
- 574 Doelling, D. R., Loeb, N. G., Keyes, D. F., Nordeen, M. L., Morstad, D., Nguyen,
 575 C., ... Sun, M. (2013). Geostationary enhanced temporal interpolation for
 576 ceres flux products [Journal Article]. *Journal of Atmospheric and Oceanic*
 577 *Technology*, 30(6), 1072-1090. doi: 10.1175/jtech-d-12-00136.1
- 578 EOL. (2018). *Near hcr radar time series data. version 1.0.* UCAR/NCAR -
 579 Earth Observing Laboratory. Retrieved from [https://data.eol.ucar.edu/](https://data.eol.ucar.edu/dataset/552.022)
 580 [dataset/552.022](https://data.eol.ucar.edu/dataset/552.022) doi: 10.5065/D6D7998S
- 581 EOL. (2019). *Low rate (lrt - 1 sps) navigation, state parameter, and microphysics*
 582 *flight-level data. version 1.3.* UCAR/NCAR EOL. Retrieved from [https://](https://data.eol.ucar.edu/dataset/552.002)
 583 data.eol.ucar.edu/dataset/552.002 doi: 10.5065/D6M32TM9
- 584 Field, P. R., Lawson, R. P., Brown, P. R. A., Lloyd, G., Westbrook, C., Moisseev,
 585 D., ... Sullivan, S. (2016). Chapter 7. secondary ice production - cur-
 586 rent state of the science and recommendations for the future [Journal
 587 Article]. *Meteorological Monographs*, 58, D-16-0014.1. doi: 10.1175/
 588 [AMSMONOGRAPHS-D-16-0014.1](https://doi.org/10.1175/AMSMONOGRAPHS-D-16-0014.1)
- 589 Gettelman, A., Bardeen, C. G., McCluskey, C. S., Jrvinen, E., Stith, J., Bretherton,
 590 C., ... Wu, W. (2020). Simulating observations of southern ocean clouds
 591 and implications for climate [Journal Article]. *Journal of Geophysical Re-*
 592 *search: Atmospheres*, 125(21), e2020JD032619. doi: [https://doi.org/10.1029/](https://doi.org/10.1029/2020JD032619)
 593 [2020JD032619](https://doi.org/10.1029/2020JD032619)
- 594 Hallett, J., & Mossop, S. C. (1974). Production of secondary ice particles during
 595 riming process [Journal Article]. *Nature*, 249(5452), 26-28. doi: 10.1038/
 596 [249026a0](https://doi.org/10.1038/249026a0)
- 597 Haynes, J. M., Marchand, R. T., Luo, Z., Bodas-Salcedo, A., & Stephens, G. L.
 598 (2007). A multipurpose radar simulation package: Quickbeam [Journal Arti-
 599 cle]. *Bulletin of the American Meteorological Society*, 88(11), 1723-1728. doi:
 600 [10.1175/bams-88-11-1723](https://doi.org/10.1175/bams-88-11-1723)
- 601 Heymsfield, A., & Willis, P. (2014). Cloud conditions favoring secondary ice parti-
 602 cle production in tropical maritime convection [Journal Article]. *Journal of the*
 603 *Atmospheric Sciences*, 71(12), 4500-4526. doi: 10.1175/jas-d-14-0093.1

- 604 Hobbs, P. V., & Rangno, A. L. (1985). Ice particle concentrations in clouds [Journal
605 Article]. *Journal of Atmospheric Sciences*, *42*(23), 2523-2549. doi: 10.1175/
606 1520-0469(1985)042(2523:ipci)2.0.co;2
- 607 Huang, Y., Chubb, T., Baumgardner, D., Dehoog, M., Siems, S. T., & Manton,
608 M. J. (2017). Evidence for secondary ice production in southern ocean open
609 cellular convection [Journal Article]. *Quarterly Journal of the Royal Meteorolo-*
610 *gical Society*, *143*(704), 1685-1703. doi: 10.1002/qj.3041
- 611 Khairoutdinov, M. F., & Randall, D. A. (2003). Cloud resolving modeling of the
612 arm summer 1997 iop: Model formulation, results, uncertainties, and sensitiv-
613 ities [Journal Article]. *Journal of the Atmospheric Sciences*, *60*(4), 607-625.
614 doi: 10.1175/1520-0469(2003)060(0607:crmota)2.0.co;2
- 615 Koenig, L. R. (1963). The glaciating behavior of small cumulonimbus clouds [Jour-
616 nal Article]. *Journal of Atmospheric Sciences*, *20*(1), 29-47. doi: 10.1175/1520
617 -0469(1963)020(0029:tgbosc)2.0.co;2
- 618 Ladino, L. A., Korolev, A., Heckman, I., Wolde, M., Fridlind, A. M., & Ackerman,
619 A. S. (2017). On the role of ice-nucleating aerosol in the formation of ice parti-
620 cles in tropical mesoscale convective systems [Journal Article]. *Geophysical Re-*
621 *search Letters*, *44*(3), 1574-1582. doi: <https://doi.org/10.1002/2016GL072455>
- 622 Luke, E. P., Yang, F., Kollias, P., Vogelmann, A. M., & Maahn, M. (2021). New
623 insights into ice multiplication using remote-sensing observations of slightly
624 supercooled mixed-phase clouds in the arctic [Journal Article]. *Proceed-*
625 *ings of the National Academy of Sciences*, *118*(13), e2021387118. doi:
626 10.1073/pnas.2021387118
- 627 McCluskey, C. S., DeMott, P. J., Ma, P.-L., & Burrows, S. M. (2019). Nu-
628 merical representations of marine ice-nucleating particles in remote ma-
629 rine environments evaluated against observations [Journal Article]. *Geo-*
630 *physical Research Letters*, *46*(13), 7838-7847. Retrieved from [https://](https://agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/2018GL081861)
631 agupubs.onlinelibrary.wiley.com/doi/abs/10.1029/2018GL081861 doi:
632 <https://doi.org/10.1029/2018GL081861>
- 633 McCluskey, C. S., Hill, T. C. J., Humphries, R. S., Rauker, A. M., Moreau, S.,
634 Stratton, P. G., ... DeMott, P. J. (2018). Observations of ice nucleating
635 particles over southern ocean waters [Journal Article]. *Geophysical Research*
636 *Letters*, *45*(21), 11,989-11,997. doi: <https://doi.org/10.1029/2018GL079981>

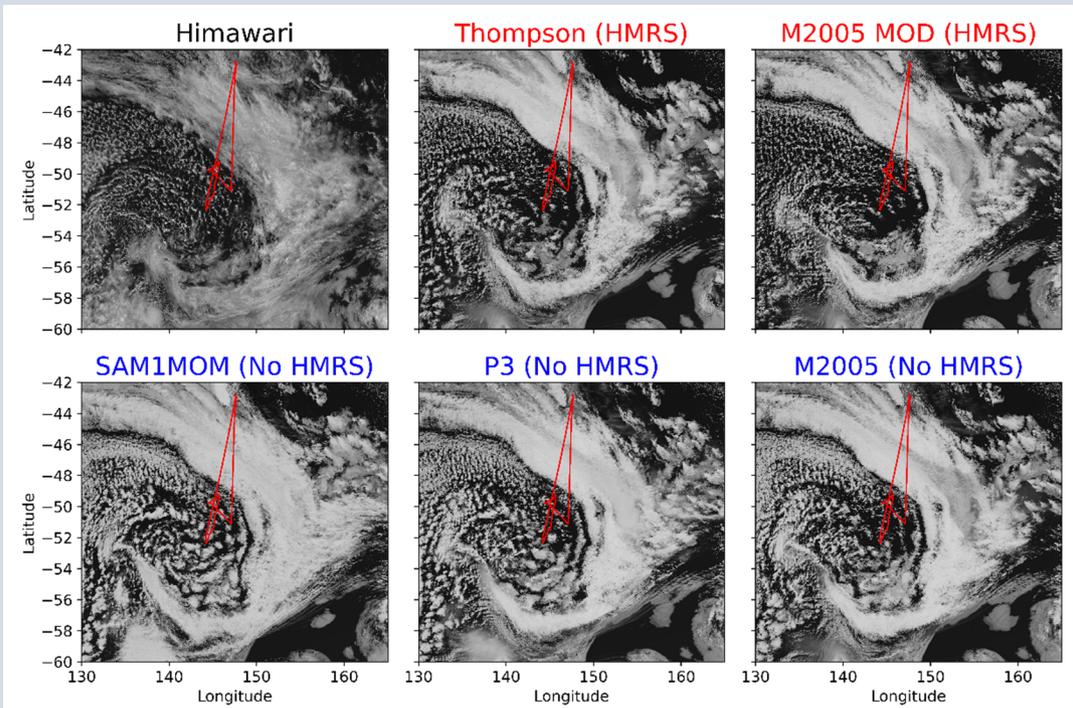
- 637 McFarquhar, G., Bretherton, C., Marchand, R., Protat, A., Demott, P., Alexander,
 638 S., . . . McDonald, A. (2020). Observations of clouds, aerosols, precipita-
 639 tion, and surface radiation over the southern ocean: An overview of capricorn,
 640 marcus, micre and socrates [Journal Article]. *Bulletin of the American Meteoro-*
 641 *logical Studies*.
- 642 Meyers, M. P., DeMott, P. J., & Cotton, W. R. (1992). New primary ice-
 643 nucleation parameterizations in an explicit cloud model [Journal Article].
 644 *Journal of Applied Meteorology and Climatology*, 31(7), 708-721. doi:
 645 10.1175/1520-0450(1992)031<0708:npinpi>2.0.co;2
- 646 Mitchell, J. F. B., Senior, C. A., & Ingram, W. J. (1989). CO₂ and climate: a miss-
 647 ing feedback? [Journal Article]. *Nature*, 341(6238), 132-134. doi: 10.1038/
 648 341132a0
- 649 Morrison, H., Curry, J. A., & Khvorostyanov, V. I. (2005). A new double-moment
 650 microphysics parameterization for application in cloud and climate models.
 651 part i: Description [Journal Article]. *Journal of the Atmospheric Sciences*,
 652 62(6), 1665-1677. doi: 10.1175/jas3446.1
- 653 Morrison, H., & Milbrandt, J. A. (2015). Parameterization of cloud microphysics
 654 based on the prediction of bulk ice particle properties. part i: Scheme descrip-
 655 tion and idealized tests [Journal Article]. *Journal of the Atmospheric Sciences*,
 656 72(1), 287-311. doi: 10.1175/jas-d-14-0065.1
- 657 NASA/LARC/SD/ASDC. (2017). *Ceres and geo-enhanced toa, within-*
 658 *atmosphere and surface fluxes, clouds and aerosols 1-hourly terra edition4a*.
 659 NASA Langley Atmospheric Science Data Center DAAC. Retrieved from
 660 <https://doi.org/10.5067/TERRA+AQUA/CERES/SYN1DEG-1HOURL3.004A> doi:
 661 10.5067/TERRA+AQUA/CERES/SYN1DEG-1HOURL3.004A
- 662 Pruppacher, H. R., Klett, J. D., & SpringerLink. (2010). *Microphysics of clouds and*
 663 *precipitation* (1. ed.) [Book]. Dordrecht: Springer Netherlands.
- 664 Rangno, A. L., & Hobbs, P. V. (2001). Ice particles in stratiform clouds in the
 665 arctic and possible mechanisms for the production of high ice concentrations
 666 [Journal Article]. *Journal of Geophysical Research-Atmospheres*, 106(D14),
 667 15065-15075. doi: 10.1029/2000jd900286
- 668 Rogers, R. R., & Yau, M. K. (1996). *A short course in cloud physics* (Third edition.
 669 ed.) [Book]. Burlington, Massachusetts: Butterworth-Heinemann.

- 670 Scott, E. L. (2019). *The influence of primary nucleation and rime splintering on ice*
671 *number concentrations in southern ocean cumuli* (Thesis).
- 672 Smith, W., & Minnis, P. (2020). *Nasa satcorps himawari cloud retrieval data.*
673 *version 2.2.* UCAR/NCAR - Earth Observing Laboratory. Retrieved from
674 <https://data.eol.ucar.edu/dataset/552.027> doi: 10.5065/D6CC0ZFJ
- 675 Sotiropoulou, G., Sullivan, S., Savre, J., Lloyd, G., Lachlan-Cope, T., Ekman,
676 A. M. L., & Nenes, A. (2020). The impact of secondary ice production on
677 arctic stratocumulus [Journal Article]. *Atmos. Chem. Phys.*, *20*(3), 1301-1316.
678 doi: 10.5194/acp-20-1301-2020
- 679 Sotiropoulou, G., Vignon, E., Young, G., Morrison, H., O'Shea, S. J., Lachlan-Cope,
680 T., ... Nenes, A. (2021). Secondary ice production in summer clouds over the
681 antarctic coast: an underappreciated process in atmospheric models [Journal
682 Article]. *Atmos. Chem. Phys.*, *21*(2), 755-771. doi: 10.5194/acp-21-755-2021
- 683 Stevens, B., Bony, S., Brogniez, H., Hentgen, L., Hohenegger, C., Kiemle, C., ...
684 Zuidema, P. (2020). Sugar, gravel, fish and flowers: Mesoscale cloud patterns
685 in the trade winds [Journal Article]. *Quarterly Journal of the Royal Meteorolo-*
686 *gical Society*, *146*(726), 141-152. doi: <https://doi.org/10.1002/qj.3662>
- 687 Stevens, B., Satoh, M., Auger, L., Biercamp, J., Bretherton, C. S., Chen, X., ...
688 Zhou, L. (2019). Dyamond: the dynamics of the atmospheric general circula-
689 tion modeled on non-hydrostatic domains [Journal Article]. *Progress in Earth*
690 *and Planetary Science*, *6*(1), 61. doi: 10.1186/s40645-019-0304-z
- 691 Sun, Z., & Shine, K. P. (1994). Studies of the radiative properties of ice and mixed-
692 phase clouds [Journal Article]. *Quart. J. Roy. Meteor. Soc.*, *120*, 111-137. doi:
693 10.1002/qj.49712051508
- 694 Taylor, J. W., Choulaton, T. W., Blyth, A. M., Liu, Z., Bower, K. N., Crosier, J.,
695 ... Brown, P. R. A. (2016). Observations of cloud microphysics and ice forma-
696 tion during cope [Journal Article]. *Atmos. Chem. Phys.*, *16*(2), 799-826. doi:
697 10.5194/acp-16-799-2016
- 698 Thompson, G., Field, P. R., Rasmussen, R. M., & Hall, W. D. (2008). Explicit fore-
699 casts of winter precipitation using an improved bulk microphysics scheme. part
700 ii: Implementation of a new snow parameterization [Journal Article]. *Monthly*
701 *Weather Review*, *136*(12), 5095-5115. doi: 10.1175/2008mwr2387.1
- 702 Trenberth, K. E., & Fasullo, J. T. (2010). Simulation of present-day and twenty-

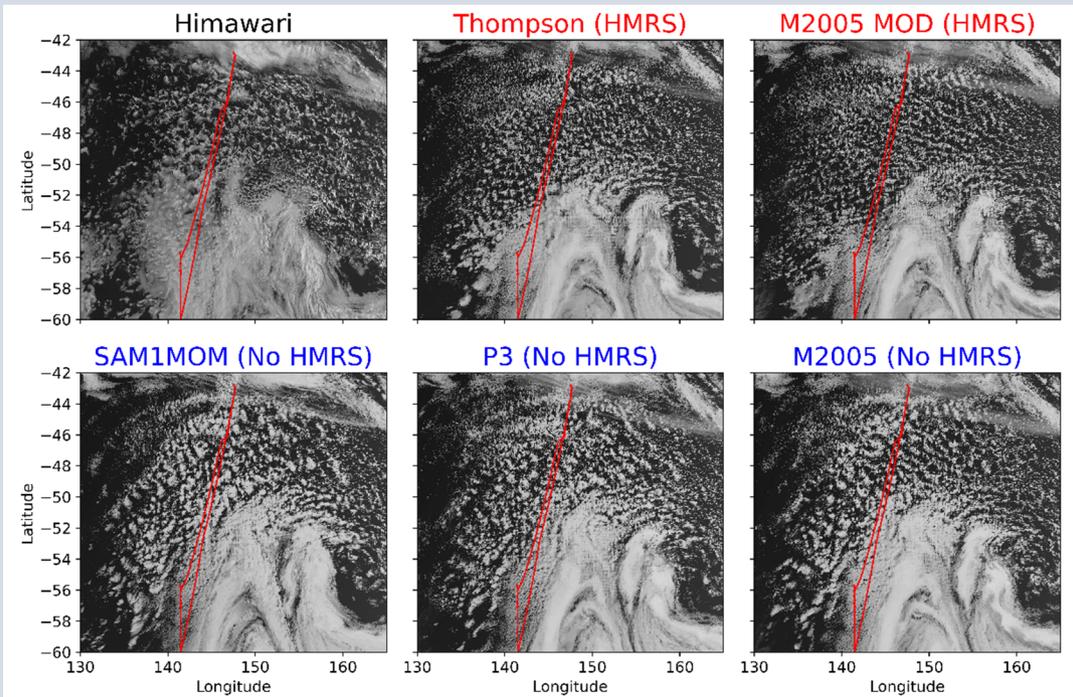
- 703 first-century energy budgets of the southern oceans [Journal Article]. *Journal*
 704 *of Climate*, 23(2), 440-454. doi: 10.1175/2009jcli3152.1
- 705 Vergara-Temprado, J., Miltenberger, A. K., Furtado, K., Grosvenor, D. P., Ship-
 706 way, B. J., Hill, A. A., . . . Carslaw, K. S. (2018). Strong control of southern
 707 ocean cloud reflectivity by ice-nucleating particles [Journal Article]. *Pro-*
 708 *ceedings of the National Academy of Sciences*, 115(11), 2687-2692. doi:
 709 10.1073/pnas.1721627115
- 710 Vonder Haar, T. H., & Suomi, V. E. (1971). Measurements of the earth's radi-
 711 ation budget from satellites during a five-year period. part i: Extended time
 712 and space means [Journal Article]. *Journal of Atmospheric Sciences*, 28(3),
 713 305-314. doi: 10.1175/1520-0469(1971)028<0305:moterb>2.0.co;2
- 714 Young, G., Lachlan-Cope, T., O'Shea, S. J., Dearden, C., Listowski, C., Bower,
 715 K. N., . . . Gallagher, M. W. (2019). Radiative effects of secondary ice en-
 716 hancement in coastal antarctic clouds [Journal Article]. *Geophysical Research*
 717 *Letters*, 46(4), 2312-2321. doi: 10.1029/2018gl080551
- 718 Zhou, X., Atlas, R., McCoy, I. L., Bretherton, C. S., Bardeen, C., Gettelman,
 719 A., . . . Ming, Y. (2021). Evaluation of cloud and precipitation sim-
 720 ulations in cam6 and am4 using observations over the southern ocean
 721 [Journal Article]. *Earth and Space Science*, 8(2), e2020EA001241. doi:
 722 <https://doi.org/10.1029/2020EA001241>



a) SOCRATES RF11: Broadband Shortwave Albedo



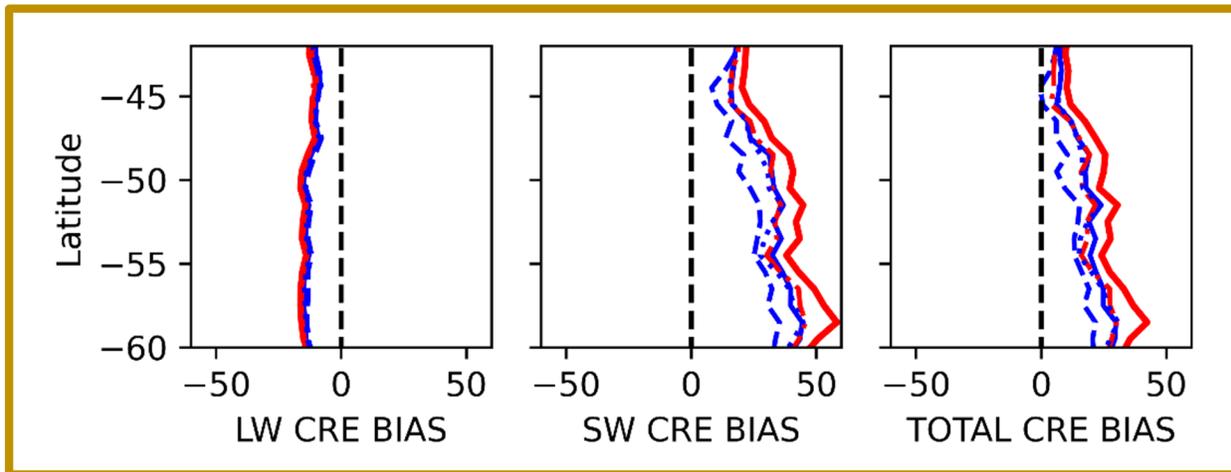
b) SOCRATES RF12 : Broadband Shortwave Albedo



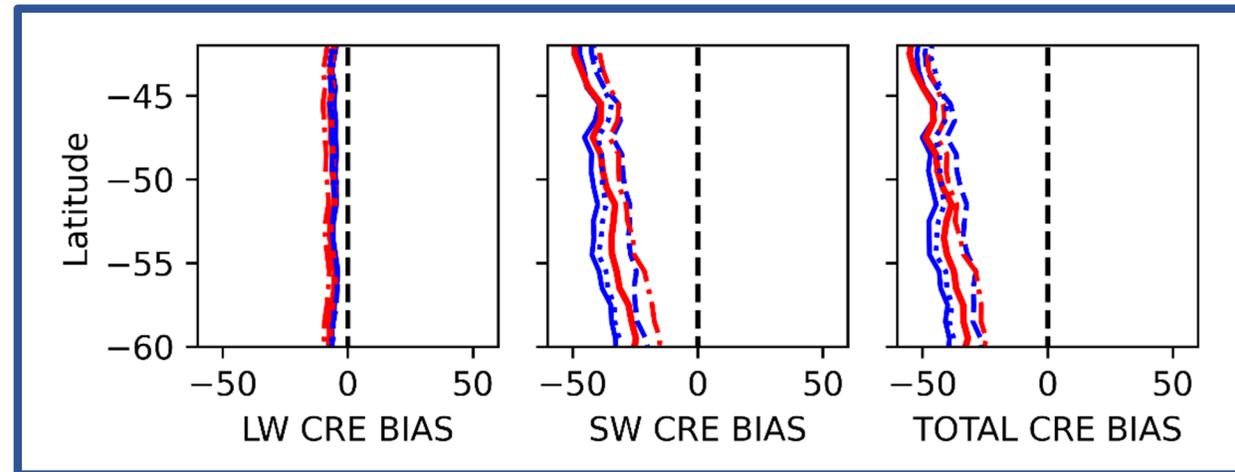
Broadband Shortwave Albedo (-)



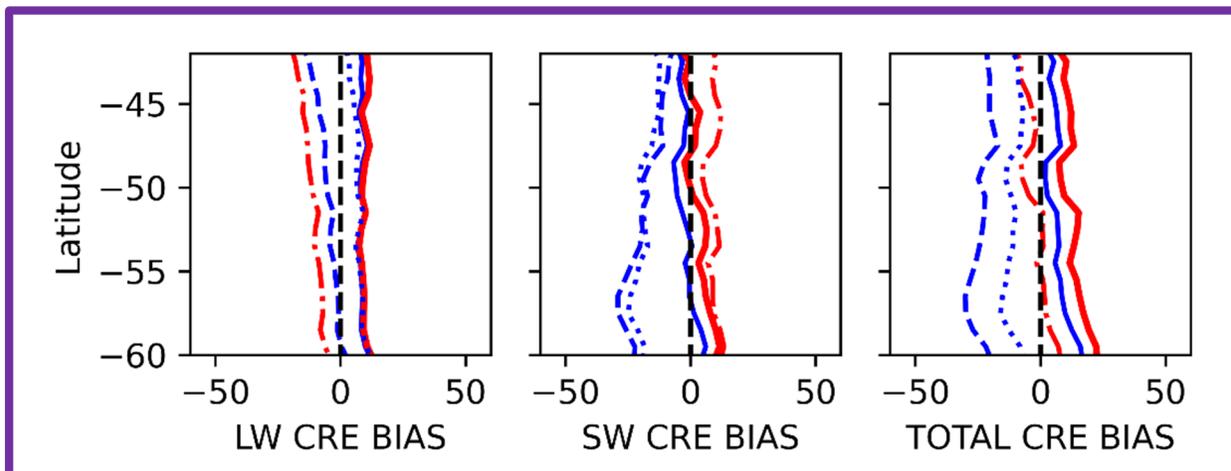
Low Cloud Fraction Areas (19.5%)



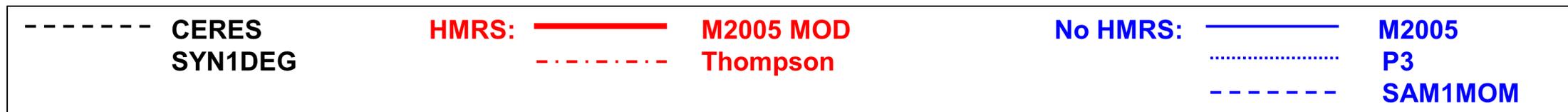
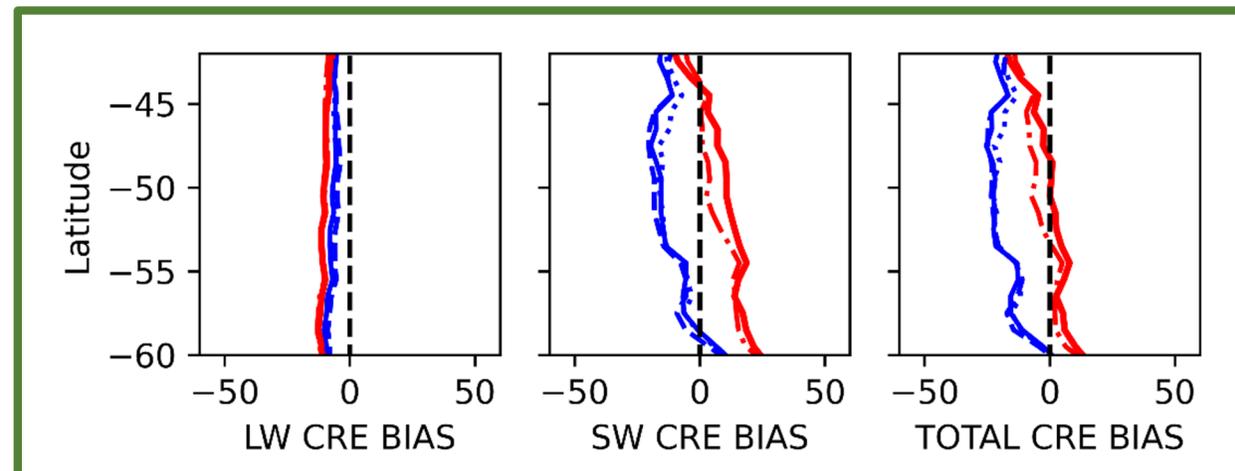
Low Stratus Clouds (18%)



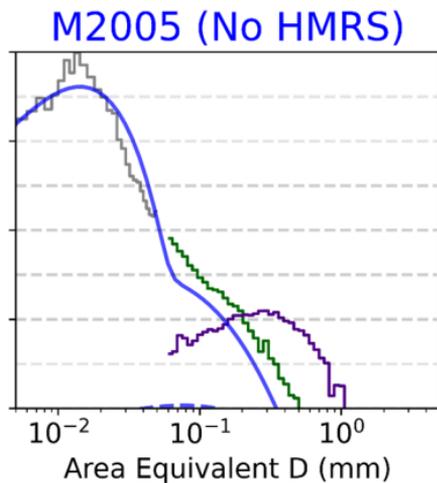
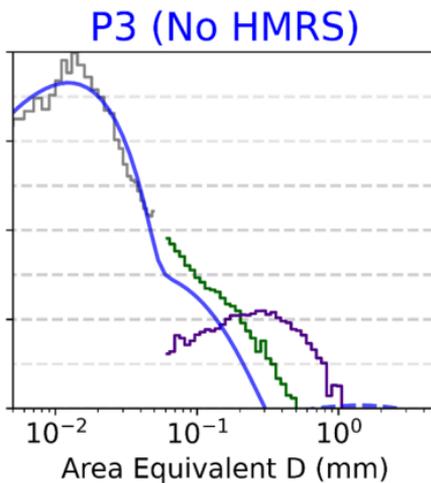
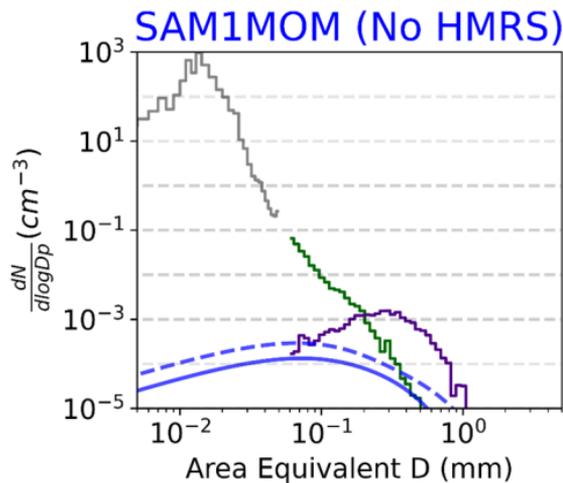
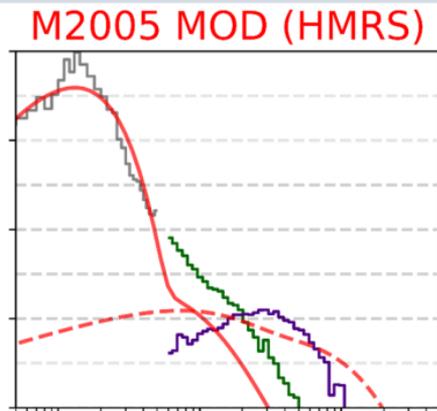
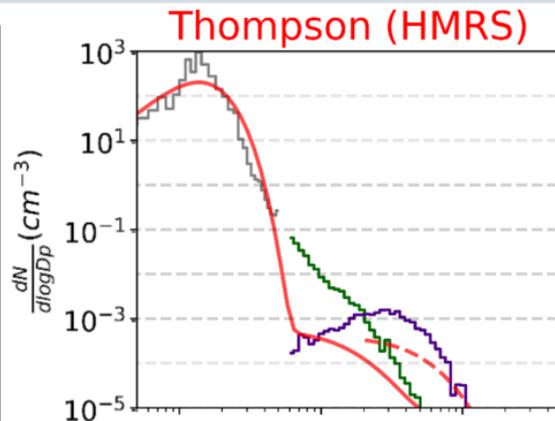
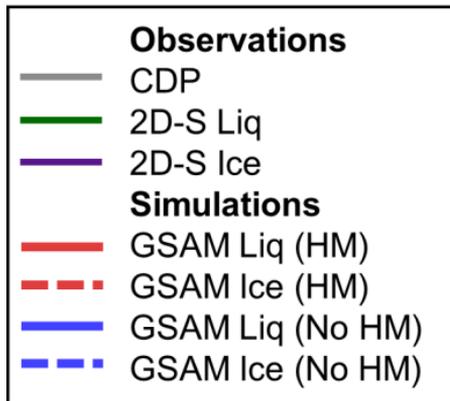
High Clouds (30.5%)



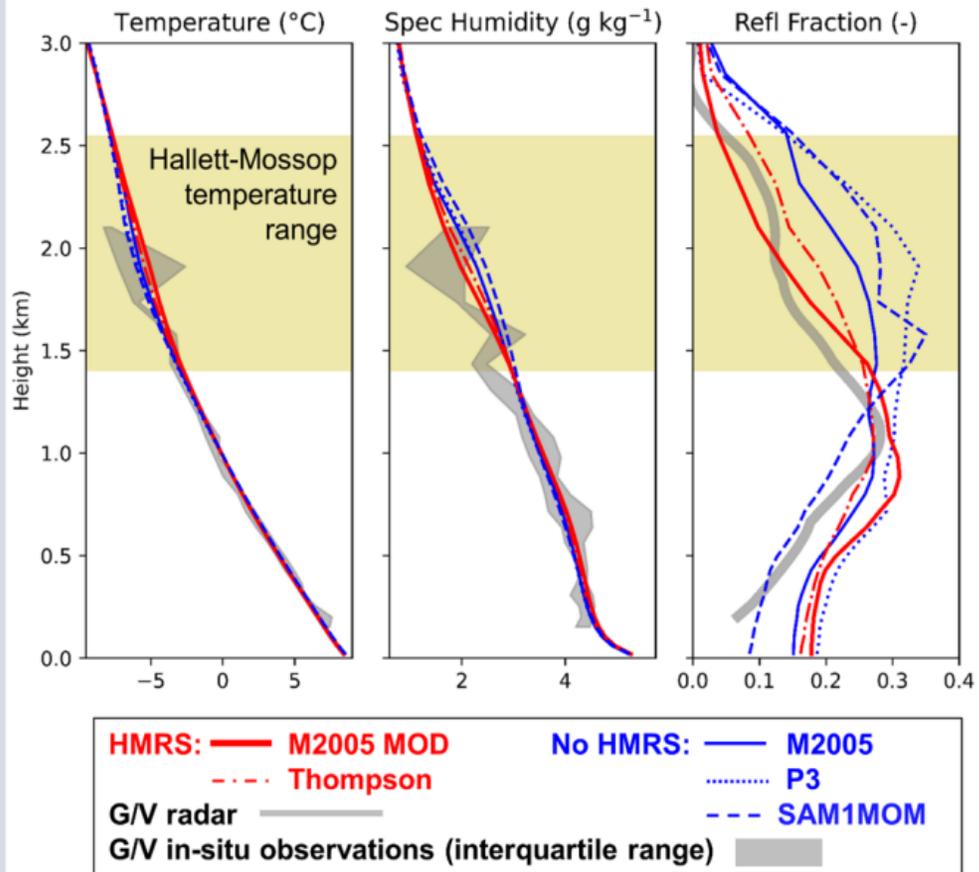
Low Cumulus Clouds (32%)



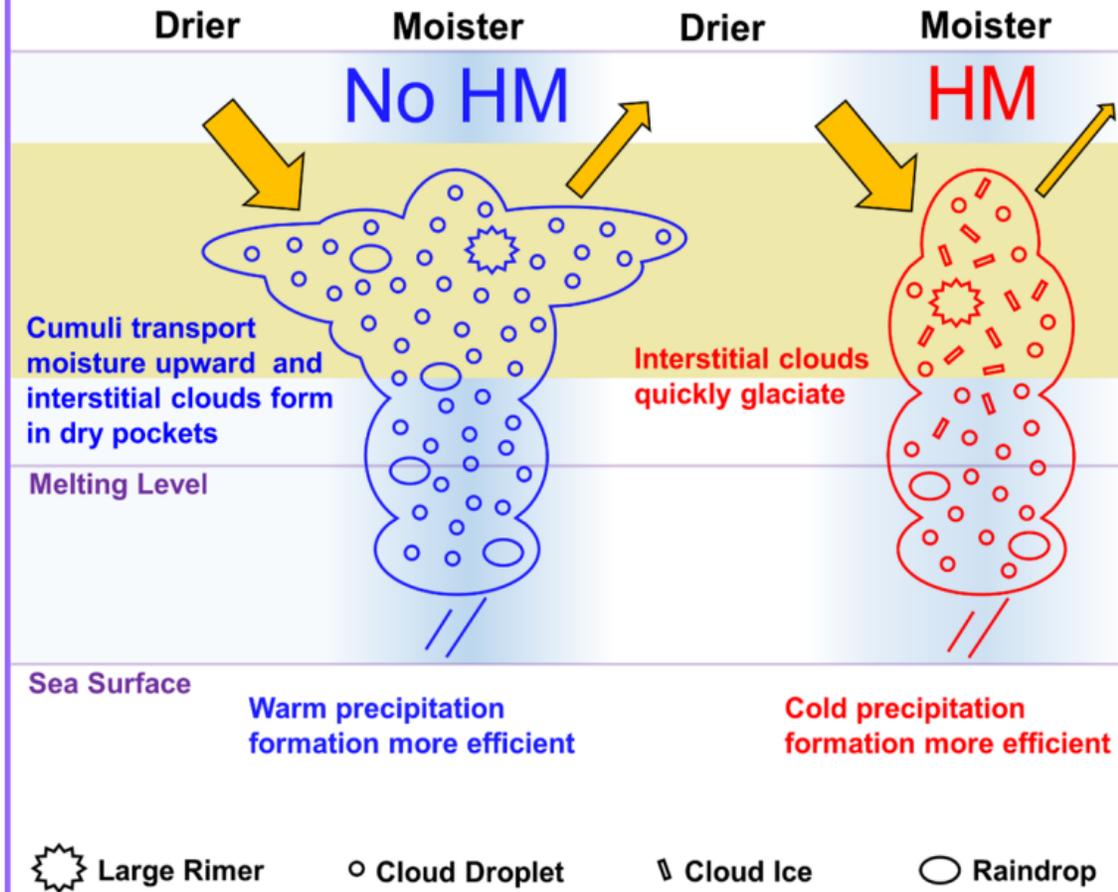
SOCRATES RF12: Particle Size Distributions



a) SOCRATES RF11: Vertical Profiles

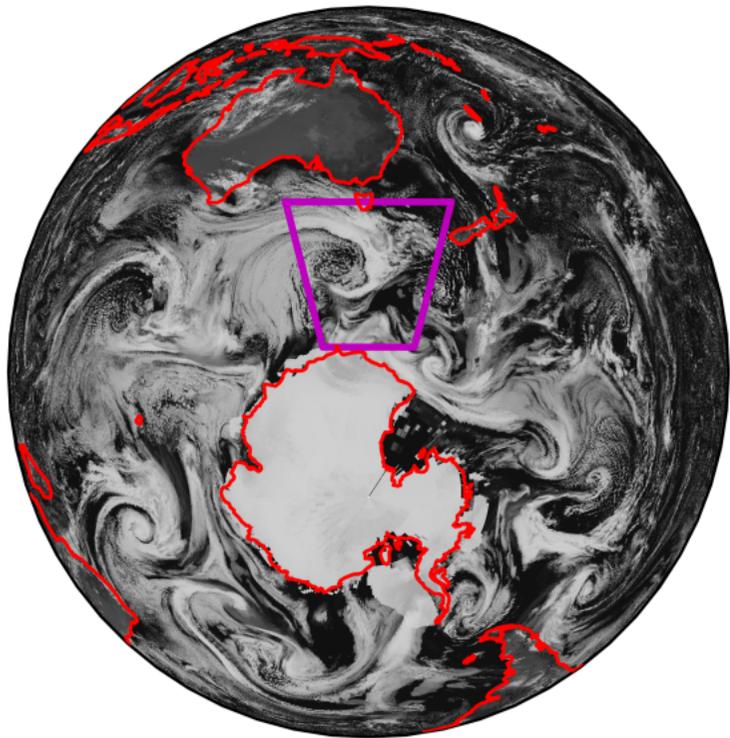


b) Summary of physical mechanism

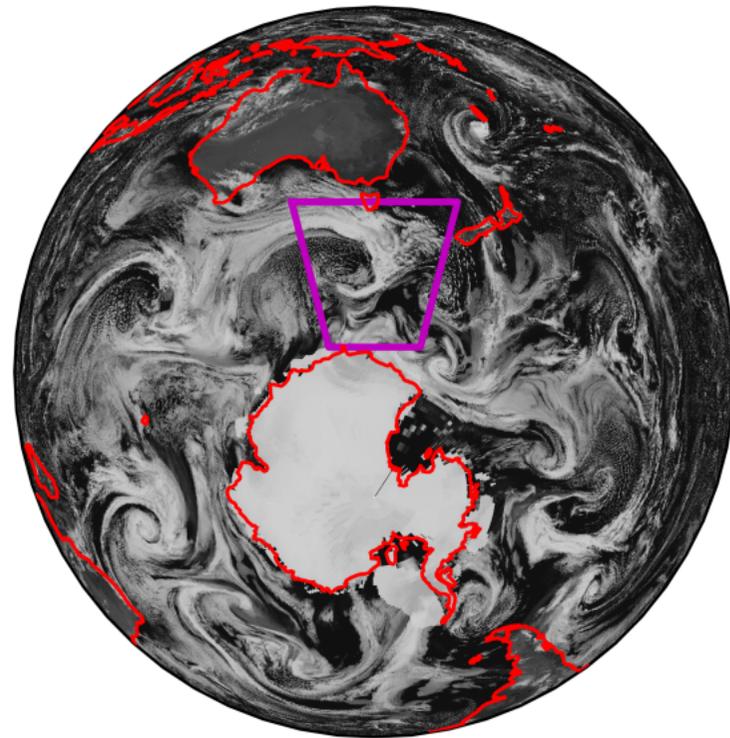


2018 February 17 00 UTC

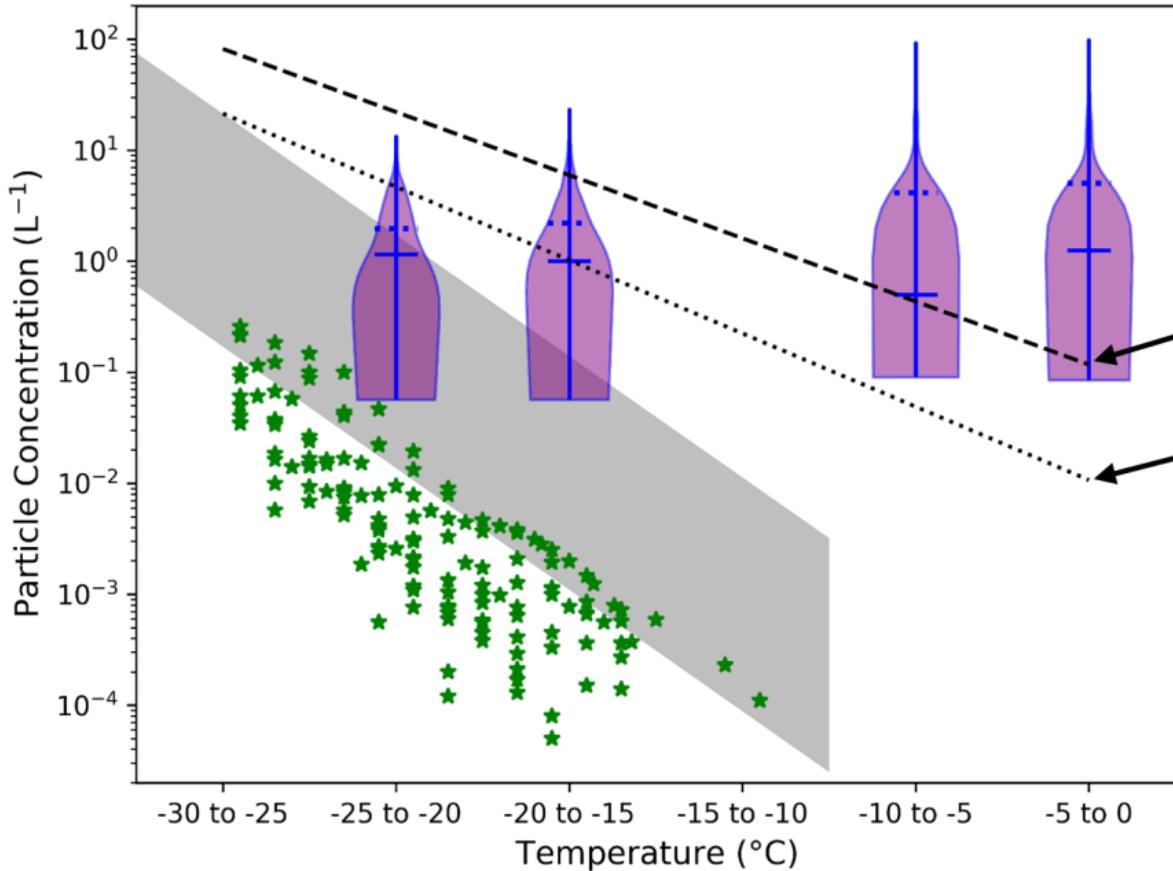
M2005



M2005 MOD



Ice crystals (SOCRATES RF11 and RF12) and marine INP concentrations



Ice Crystals from SOCRATES RF11 and RF12

Meyers Curve (Contact Nucleation)

Cooper Curve (Deposition Nucleation)

★ INPs from McCluskey et al. 2018 (CAPRICORN)

INPs from DeMott et al. 2016

Cloud Fraction ≥ 0.1

No

Low Cloud Fraction

Yes

Cloud Height > 5 km

No

Cloud Fraction > 0.75

Yes

High Cloud

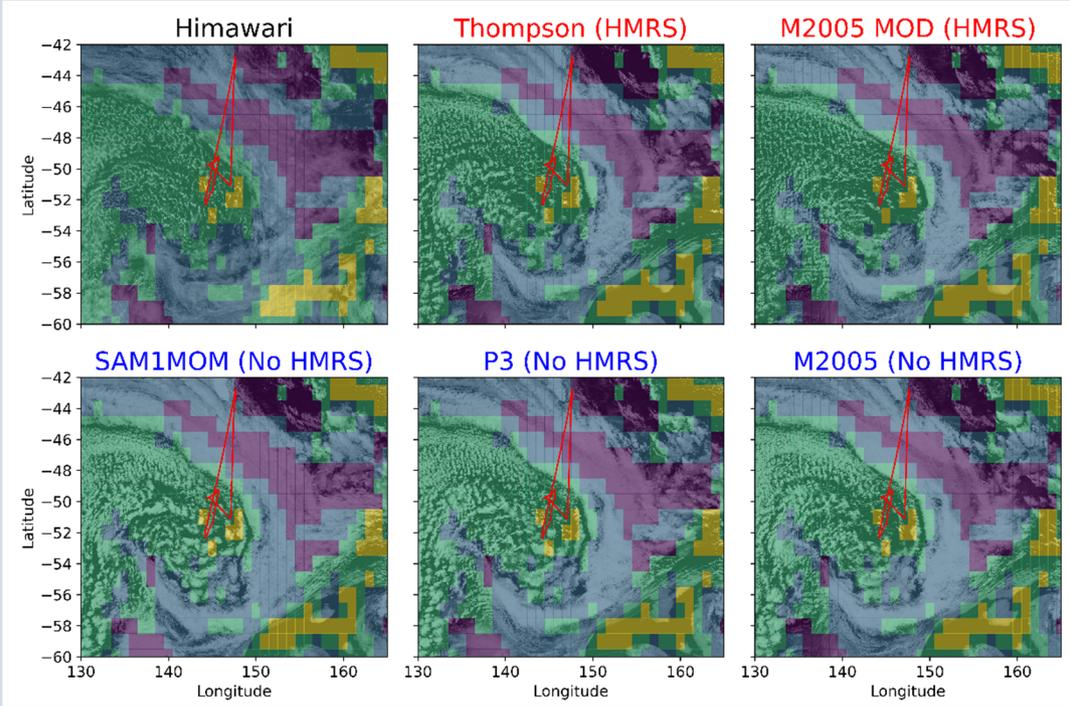
No

Low Cumulus

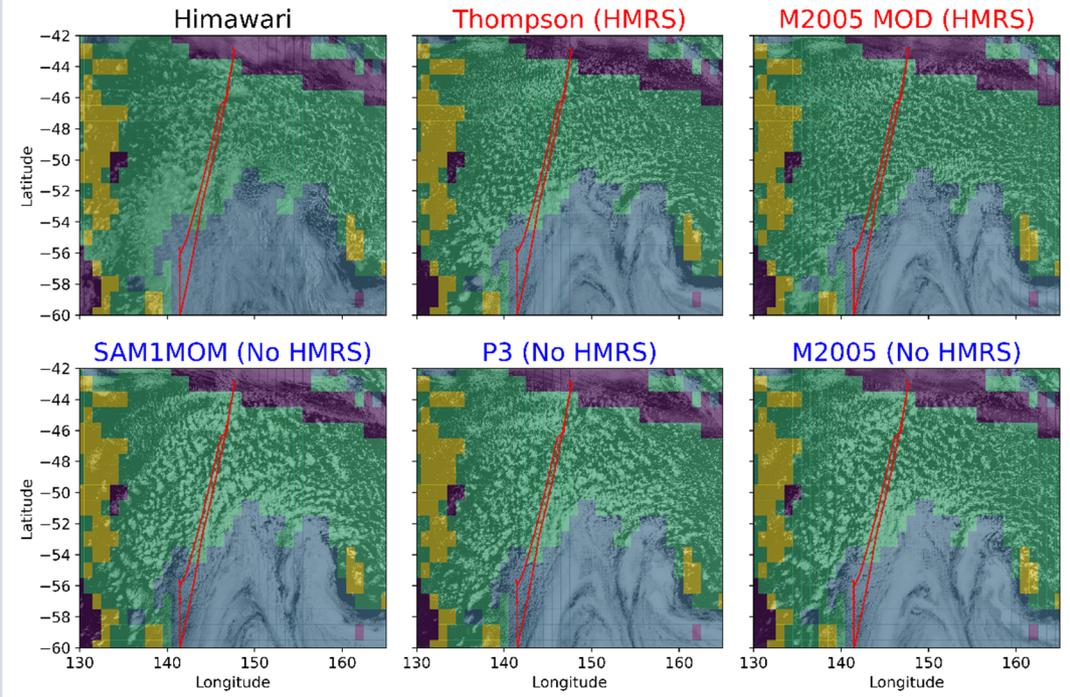
Yes

Low Stratus

a) SOCRATES RF11: Broadband Shortwave Albedo + Mask



b) SOCRATES RF12 : Broadband Shortwave Albedo + Mask



Low Cloud Fraction

Low Cumulus

Low Stratus

High Cloud