1 Natural Marine Cloud Brightening in the Southern Ocean
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3 Gerald G. Mace[1], Sally Benson[1], Ruhi Humphries[2,3], Mathew Peter Gombert[1], Elizabeth
4 Sterner[1]
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6 [1]Department of Atmospheric Sciences, University of Utah, Salt Lake City, Utah
7 [2]Climate Science Centre, CSIRO Oceans and Atmosphere, Melbourne, Australia
8 [3]Australian Antarctic Program Partnership, Institute for Marine and Antarctic Studies,
9 University of Tasmania, Hobart,Tasmania, Australia
10
11 Corresponding Author Information:
12 Gerald "Jay" Mace, Professor
13 Department of Atmospheric Sciences, University of Utah
14 135 South 1460 East Rm 819 (819 WBB)
15 Salt Lake City, Utah, 84112-0110
16 Cell Phone: 801 201 7944
17 Office Phone: 801 585 9489
18 Email: jay.mace@utah.edu
19 Fax: 801 860 0381
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Abstract: The number of cloud droplets per unit volume (N_d) is a fundamentally important property of marine boundary layer (MBL) liquid clouds that, at constant liquid water path, exerts considerable controls on albedo. Past work has shown that regional N_d has direct correlation to marine primary productivity (PP) because of the role of seasonally varying biogenically-derived precursor gasses in modulating secondary aerosol properties. These linkages are thought to be observable over the high latitude oceans where strong seasonal variability in aerosol and meteorology covary in mostly pristine marine environments. Here, we examine N_d variability derived from five years of MODIS level 2 derived cloud properties in a broad region of the summertime Eastern Southern Ocean and adjacent marginal seas.   We demonstrate latitudinal, longitudinal, and temporal gradients in N_d that are strongly correlated with the passage of air masses over regions of high PP waters that are mostly concentrated along the Antarctic Shelf poleward of 60°S. We find that the albedo of MBL clouds in the latitudes south of 60°S is significantly higher than similar LWP clouds north of this latitude.

Short Summary: The number cloud droplets per unit volume is a significantly important property of clouds that controls their reflective properties. Computer models of the Earth's atmosphere and climate have low skill at predicting the reflective properties of Southern Ocean clouds. Here we investigate the properties of those clouds using satellite data and find that the cloud droplet number in the Southern Ocean is related to the oceanic phytoplankton abundance near Antarctica and cause clouds there to be significantly brighter than clouds futher north.

1. Introduction

The cloud and precipitation properties of the Southern Ocean (SO) have received considerable attention since Trenberth and Fasullo (2010) identified a high bias in surface-absorbed solar energy there (McFarquhar et al., 2020). This bias has been traced to erroneously small Marine Boundary Layer (MBL) cloud cover in simulations of the Southern Ocean climate (Bodas-Salcedo, et al., 2016; Naud et al., 2016). The actual SO cloud climatology and associated albedo are dominated by geometrically thin MBL clouds (Mace et al., 2010; Mace et al., 2020, 2021). Because the predominant shallow boundary layer clouds rarely precipitate (Huang et al., 2016), they are sensitive to cloud condensation nuclei (CCN) concentrations (Twohy and Anderson, 2008; Petters and Kreidenweis, 2007).

In the SO, the CCN seasonal cycle (Ayers and Gras, 1991; Vallina et al. 2006; Gras and Keywood, 2017) is reflected in basin-wide cloud property variations (Krüger and Graßl, 2011). McCoy et al. (2015) and Mace and Avey (2017) also found that MODIS- and A-Train-derived cloud properties over the SO demonstrate a similar seasonal cycle in cloud droplet number concentration (N_d) as for CCN. The basin wide variability in CCN and cloud albedo have been shown to be correlated with marine primary productivity (PP – defined as the net organic matter, mostly produced by phytoplankton, that is suspended in the ocean; Vallina et al., 2006; Krüger and Graßl,2011; McCoy et al., 2015). McCoy et al. (2020) argue that the SO can be viewed as an analog of the preindustrial Earth. As such and given the large natural seasonal variability in CCN

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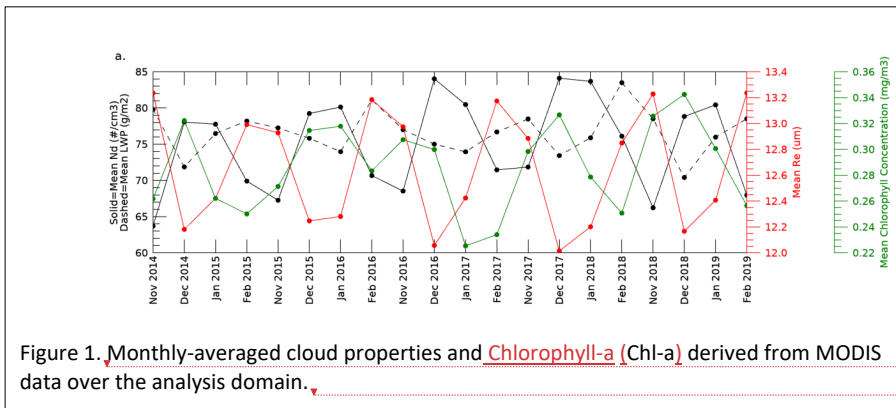**Deleted:** , respectively

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76 and clouds, the SO is a natural laboratory to understand the processes that contribute
77 to simulated aerosol-related indirect forcing variability in climate models (Carslaw et al.
78 2013).
79
80 CCN and cloud droplet N_d in the SO are higher in Summer when significant latitudinal
81 gradients have been documented in the SO Australasian sector (Humphries et al.,
82 2021). Using time of flight aerosol chemical speciation monitor (ACSM) and ion

Figure 1. Monthly-averaged cloud properties and Chlorophyll-a (Chl-a) derived from MODIS data over the analysis domain.

83 concentrations from filter samples, Humphries et al., (2021) analyzed the covariance of
84 aerosol chemistry, CCN at 0.5% supersaturation, and Condensation Nuclei (CN) larger
85 than 10 nm collected aboard Australian research vessels during the 2018 Austral
86 Summer (McFarquhar et al., 2021). While sulfates were a major compositional
87 component of aerosol at all latitudes during summer these compounds were in higher
88 fractional abundance poleward of 65°S where overall CCN numbers were higher by
89 ~50%. Chloride derived from sea salt was dominant in the region equatorward of 65°S
90 but was mostly absent south of 65°S.   The ratio of CCN to CN at 0.5% supersaturation
91 increased considerably south of 65°S suggesting unique aerosol chemistry compared to
92 the open ocean. Humphries et al. (2021) also discusses how this compositional
93 boundary in aerosol chemistry is often very distinct in the East Antarctic waters between
94 60°S and 65°S. Following Humphries et al. we will refer to this belt as the Atmosphere
95 Compositional Front of Antarctica (ACFA).   Humphries et al. (2021) conclude that
96 aerosol, newly condensed from gas phase sulfur species such as from the oxidation of
97 dimethylsulfide (DMS), are an important component of the high latitude CCN. These
98 products of phytoplankton physiology are released into the atmosphere from the highly
99 productive waters from ~60°S to the Antarctic – a region well known for a vast marine
100 food web (Deppler and Davidson, 2017; Behrenfeld et al., 2016).
101
102 Mace et al. (2021a) derived N_d and other cloud microphysical properties from non-
103 precipitating stratocumulus clouds using shipborne remote sensing data. They found
104 that stratiform clouds poleward of the ACFA had significantly higher N_d than
105 equatorward. One particular case took place when the Icebreaker Aurora Australis was
106 at the Davis Antarctic station just east of Prydz Bay (~77°E) between 1 and 5 January

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113 2018 and featured nearly continuous high Nd clouds (> 150 cm^{-3}) that occurred in a
114 southerly flow passing over the ship that had trajectories from the Antarctic Continent.
115 Similarly, Twohy et al., (2021) report that the highest sulfur-based concentrations of
116 aerosol in the free troposphere north of 60°S observed from research aircraft in
117 Summer 2018 had occurred in airmasses that had originated recently from over the
118 Antarctic continent. See also Shaw et al. (1988) for an early examination of the role of
119 biogenic sulfate in modulating summertime aerosol along coastal Antarctica. Shaw et al.
120 (2007) expands on this idea as does Korhonen et al., (2008).
121
122
123
124 2. Results
125
126 See Appendix A for methods and definitions. Approximately 40,000 1° latitude by 2°
127 longitude MBL cloud scenes per month meet our criteria for liquid phase non
128 precipitating clouds in the analysis domain. This number varies by ~25% in a seasonal
129 cycle that is due mostly to our solar zenith angle criteria. A seasonal cycle is evident in
130 the monthly-averaged cloud properties. LWP and r_e have seasonal minima in the
131 months of December and January. Due to an $r_e^{-5/2}$ dependence, N_d is of opposite

Figure 2. a) N_d frequency distribution from the cloud scenes in the analysis domain during
the 5-years of summer months analyzed. Vertical lines are defined in the inset. b) The
latitudinal distribution of the cloud scenes that compose the high and low N_d quartiles. c)
the distributions of liquid water path for the high and low Nd quartiles, d) the distribution of
normalized CERES solar albedo of the high and low Nd quartiles. The normalization
procedure is described in the appendix. The colors of the histograms in paneles b, c, and d,
are as described in the inset of panel a.

140 phase with r_e and correlated with it at -0.93. The seasonal variability in LWP (r_e) is on
141 the order of 7% (4%) and is small in comparison to Nd (~25%). τ and r_e are derived
142 from the visible and near infrared reflectances with the MODIS level 2 retrieval algorithm
143 (Nakajima and King, 1990). LWP is, then, derived from a relationship similar to $\tau =$
144 $\frac{3}{2\rho_w}\frac{LWP}{r_e}$ (Stephens, 1978). It is reasonable to consider whether seasonal variations in N_d,
145 perhaps linked to CCN, might be associated with variability in LWP. We find that LWP
146 decreases as N_d increases with a correlation coefficient in the monthly means of Figure
147 1 of -0.60.
148
149 In four of the five years, we see by inspection of Figure 1 that Chl-a leads changes in N_d
150 by approximately 1 month. The correlation coefficient of Nd and Chl-a increases from
151 0.27 to 0.66 when Nd is lagged from 0 to 1 month in the Figure 1 time series although
152 this result should be interpreted with caution given the break between February and
153 November in the time series. These results are broadly similar to those presented by
154 McCoy et al., (2015) and Mace and Avey (2017). McCoy et al. (2015) link N_d variations
155 to PP using regression analysis of MODIS derived N_d against a biogeochemical
156 parameterization of biogenic sulfate and organic mass fraction (See also Lana et al.,
157 2012).
158
159 We find a broad distribution of scene-averaged N_d (Figure 2a) with median, lower and
160 upper quartile values of 66 cm^{-3}, 42 cm^{-3} and 101 cm^{-3} respectively. Henceforth, we
161 focus our analysis on the groups of scenes that are less than and greater than the
162 upper and lower quartiles. The high and low N_d scenes have distinct latitudinal
163 occurrence distributions (Figure 2b) with low N_d scenes peaking broadly at 48°S while
164 the high N_d scenes demonstrate a modal occurrence near 64°S. Overall, the N_d gradient
165 implied by Figure 2 is correlated with the latitudinal distribution of imager-derived Chl-a
166 (i.e. Deppler and Davidson, 2017). The seasonally averaged N_d gradient is also
167 discussed in McCoy et al., (2020). Differentiating seasonally varying properties north
168 and south of the ACFA (not shown), we find a clear differentiation in r_e and N_d with
169 smaller r_e south of the ACFA (mean r_e~11um, N_d ~100) compared to north (mean
170 r_e~13um, N_d ~67 cm^{-3}). LWP is slightly larger by ~7% south of the ACFA. Both
171 regions have a distinct seasonal cycle in cloud properties shown in Figure 1 although
172 the southern latitudes have larger interannual variability likely owing to variations in
173 annual sea ice extent and melt. The LWP distribution of the high N_d quartile is
174 significantly shifted to lower values compared to the low Nd quartile LWP distribution
175 (Figure 2c). This finding is in accordance with the observational and theoretical work
176 presented in Glassmeier et al., (2021) who argue that closed cell stratocumulus that
177 dominate the clouds examined here have increased entrainment drying under higher N_d
178 conditions. Figure 2c and 2d illustrate that even though the high N_d quartile scenes tend
179 to have lower LWP, their solar albedo (A) tends to be significantly higher than the low
180 N_d quartile scenes illustrating the influence of cloud microphysics on the radiative
181 forcing of these different cloud populations.
182
183 The high N_d scenes occur predominantly poleward of the ACFA (Figure 3). Interestingly
184 we find that the latitudinal gradient weakens slightly west of 90°E with a broad region of
185 higher N_d occurrence in the vicinity of the Kerguelen Rise where PP is higher (Cavagna

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215 et al., 2015).   Establishing causality between regions of high PP and cloud properties
216 is challenging (i.e. Meskhidze and Nenes, 2006; Miller and Yuter, 2008). While we find
217 seasonal associations over broad regions here, the chain of causality between
218 phytoplankton and clouds is not immediate or even necessarily direct because the
219 chemical processes take time to evolve and can move along chemical pathways that
220 have divergent outcomes (Woodhouse et al., 2013). To increase cloud N_d, new CCN
221 must be formed. Formation of new CCN can occur when sulfur compounds emitted
222 from the ocean surface nucleate after oxidation in the presence of sunlight. This
223 process of new particle formation occurs in the absence of other aerosol and often
224 requires mixing of the gaseous compounds from the boundary layer into the low-aerosol
225 free-troposphere where the newly formed aerosol can be transported widely (Shaw,
226 2007; Korhonen et al., 2008). Other chemical pathways are possible such as deposition
227 of sulfate compounds onto primary sea salt particles that modify the chemical properties
228 of existing CCN rather than nucleating new CCN (Fossum et al., 2020) or even removal
229 of sulfur compounds via aqueous phase oxidation in clouds (Woodhouse et al., 2013).
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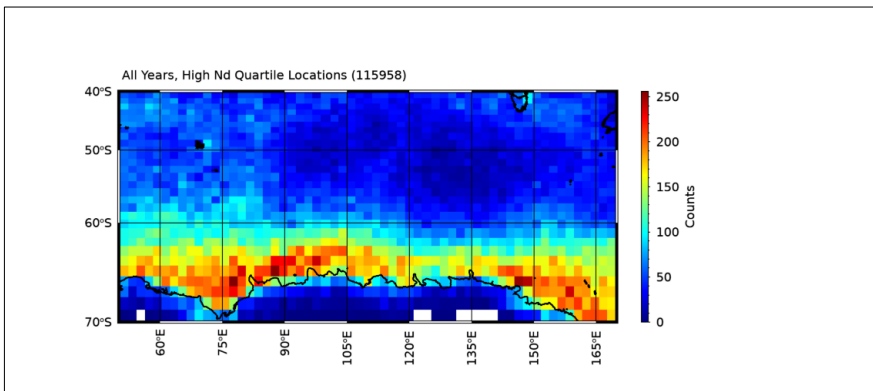All Years, High Nd Quartile Locations (115958)

Figure 3. Geographic distribution of the high N_d quartile cloud scenes. Number in
parentheses show the total of number cloud scenes from the 5-year summer data set.

231 Given the foregoing discussion, it seems reasonable that an airmass that is producing
232 clouds with certain features could be interacting with an aerosol population that has
233 evolved over periods of days (Brechtel et al., 1998).   In addition, natural cloud
234 processes such as collision and coalescence of drops tend to cause N_d to decrease
235 while precipitation efficiently scavenges CCN, thereby lowering CCN concentration and
236 even modifying their composition and size through aqueous processing (Hoppel et al.,
237 1986). With larger r_e north of the ACFA, the collision-coalescence process is likely more
238 active (Freud and Rosenfeld, 2012) and could explain the latitudinal difference in
239 adiabaticity (see methods) found in in situ data.   For instance, Kang et al. (2022)
240 analyzed data collected from Macquarie Island (54.6°S, 158.9°E) and found that, not
241 only were most clouds drizzling, but that precipitation as light as 0.01 mm hr^{-1} could
242 reduce N_d by ~50%. Therefore, a cloud field should be considered as the product of

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244 both local dynamics and thermodynamics primarily with modulation by a local
245 population of CCN. To examine the role of airmass history, we calculate the 5-day back
246 trajectories using the Hybrid Single-Particle Lagrangian Integrated Trajectory (HYSPLIT;

247 Stein et al., 2015) model using the Global Data Assimilation System (GDAS; Kamitsu,
248 1989) as input.   The parcel's endpoint is the central latitude and longitude of the cloud
249 scene and the location and model output are stored hourly.
250
251 South of the ACFA, the histories of the populations tend to be statistically different

252 (Figure 4). The low N_d clouds are more likely to be observed in airmasses that have
253 trajectories that originated in the open ocean region to the north of the ACFA. High N_d
254 scenes rarely evolve in airmasses that originate in the open ocean to the north of the
255 ACFA. The likelihood is that an airmass that has produced a high N_d cloud scene south

256 of the ACFA latitude has spent most of the previous 5 days over latitudes south of the
257 ACFA. North of the ACFA, the latitude distributions during the months of November and
258 February (not shown) are essentially identical for the high and low Nd quartiles.
259 However, for December and January, we find that the high N_d clouds observed north of
260 the ACFA have an increased likelihood of trajectories emanating from south of the
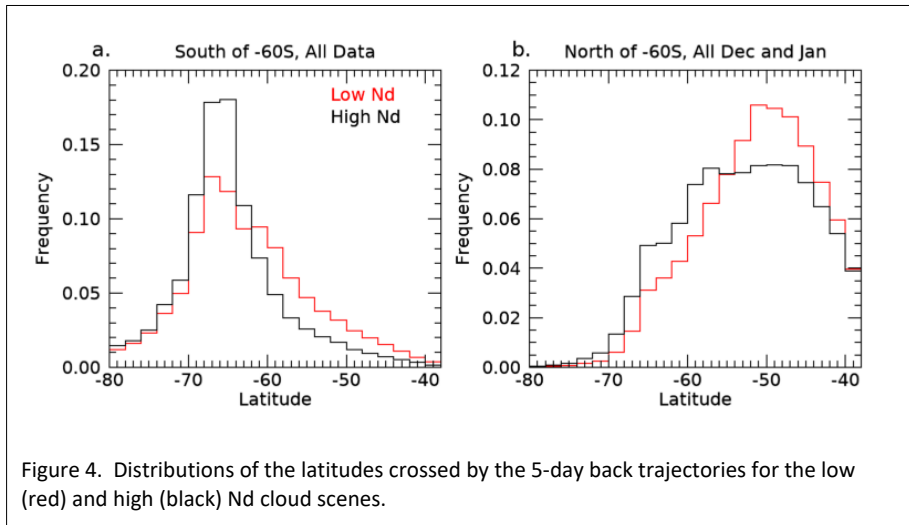261 ACFA during the 5-days prior to the MODIS observation.
262
263   3.   Discussion and Conclusions
264
265 Using MODIS level 2 cloud property retrievals and the technique developed in
266 Grosvenor et al. (2018; hereafter G18) to estimate N_d, we examine the latitudinal and

267 seasonal cycles of non-precipitating liquid-phase clouds in the Australasian sector of the
268 Summertime Southern Ocean. The r_e and N_d have distinctive differences north and
269 south of the ACFA but demonstrate similar seasonal cycles. We infer that the spatial
270 and temporal variability in cloud N_d, and r_e are at least partially a function of the
271 geographic and temporal variability in CCN that, in turn, is related to the seasonality of
272 primary sources such as sea salt and the latitudinal variability in marine PP.   The

273 highest N_d clouds tend to be overwhelmingly found along the East Antarctic coastal
274 waters south of the ACFA.
275

Figure 4. Distributions of the latitudes crossed by the 5-day back trajectories for the low (red) and high (black) Nd cloud scenes.

281 Because aerosol precursor gasses like DMS often require trajectories through the free
282 troposphere to nucleate new particles that then take time to reach CCN sizes
283 (Korohonen et al., 2008; Shaw et al. 2007), we examine the back trajectories of the
284 airmasses observed with high and low N_d south of the ACFA and find significant
285 differences. Low N_d cloud scenes are more likely to have arrived south of the ACFA
286 from northerly trajectories that would have transported low CCN air dominated by sea
287 salt. The high N_d cloud scenes are more likely to have trajectories that have remained
288 adjacent to or had passed over the Antarctic continent. North of the ACFA, while the
289 trajectory statistics for the high and low N_d quartiles in November and February are
290 nearly identical, during December and January the high N_d clouds scenes tend to have
291 an increased likelihood of arriving north of the ACFA from southerly trajectories,
292 suggesting that high CCN airmasses are being transported northward especially during
293 December and January.
294
295 Given that the main difference between the source regions north and south of the ACFA
296 is the magnitude of the marine PP, and given previous analyses of CCN compositional
297 sensitivity to marine biological factors (e.g. Humphries e al., 2021; Vallina et al., 2006;
298 Lana et al., 2012; McCoy et al, 2015), we conclude that the biological source of sulfate
299 precursor gasses and the slackening of surface winds with latitude during Summer
300 plays a dominating role in controlling the latitudinal gradients in the properties of weakly
301 precipitating MBL cloud fields over the Southern Ocean. Figure 5 summarizes our
302 findings by presenting composite seasonal cycles of MBL cloud scenes north and south
303 of 60°S. The LWP in both latitudinal bands go through a weak seasonal cycle. The
304 significant contrast in optical depth between the northern and southern bands is, we
305 infer, mostly caused by the latitudinal contrast in N_d. Based on available evidence, we
306 conclude that the differences in r_e in MODIS retrievals are causally linked to oceanic PP

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316 gradients that drive CCN, and thereby N_d, to be higher over the southern region.   This
317 sensitivity, in turn, plays a significant role in modulating the regional albedo (A) and,
318 thereby, influences the input of sunlight to the surface ocean. We note that the
319 seasonal cycle in A is different between the northern and southern latitude domains (a
320 topic for future work), however, at all times A of the southern domain is higher than that
321 of the northern domain.   However, we should be careful not to overstate this case.
322 Cloud processes that consume N_d and modify CCN (i.e. precipitation and cloud
323 processing) also play a role in modulating cloud N_d and therefore regional A (Kang et
324 al., 2022; McCoy et al., 2020). The airmass history and source region, while apparently
325 important, are among many factors involved.
326
327 Since the magnitude of PP is significantly lower north of the ACFA throughout the
328 summer season, a similar seasonal cycle in N_d and r_e suggests that CCN derived from
329 DMS oxidation of precursor gasses emitted primarily from Antarctic coastal waters
330 perhaps seeds much of the rest of the Southern Ocean with biogenic sulfate aerosol as
331 observed in recent airborne observations (Twohy et al., 2021).   The northerly transport
332 of these high sulfate airmasses out of the Antarctic coastal waters (Figure 4b) and
333 southerly transport of low sulfate air masses into the Antarctic coastal region near the
334 surface (Figure 4a) have been reported by Humphries et al. (2016, 2021) and Shaw
335 (1988) and also observed in the free troposphere with recent research aircraft
336 measurements (Twohy et al. 2021).
337
338 Our ability to identify natural marine cloud brightening (Latham et al., 2008) due to
339 aerosol-cloud coupling is a direct result of the absence of other anthropogenic and
340 continental influences in the pristine SO.   As argued by McCoy et al. (2020), it seems
341 clear that in several important ways, the Southern Ocean is the last vestige of the
342 preindustrial atmosphere allowing us to constrain processes that remain important to
343 our understanding of the climate today (Carslaw et al., 2013).

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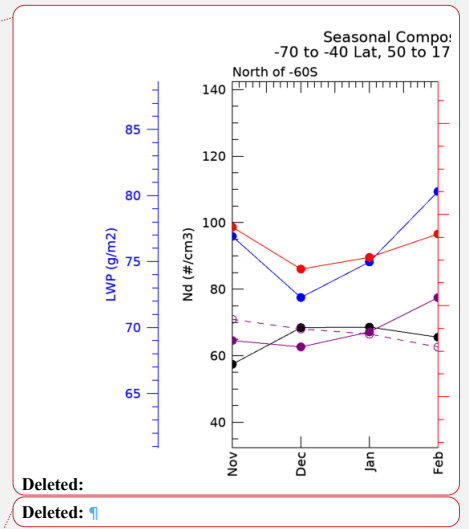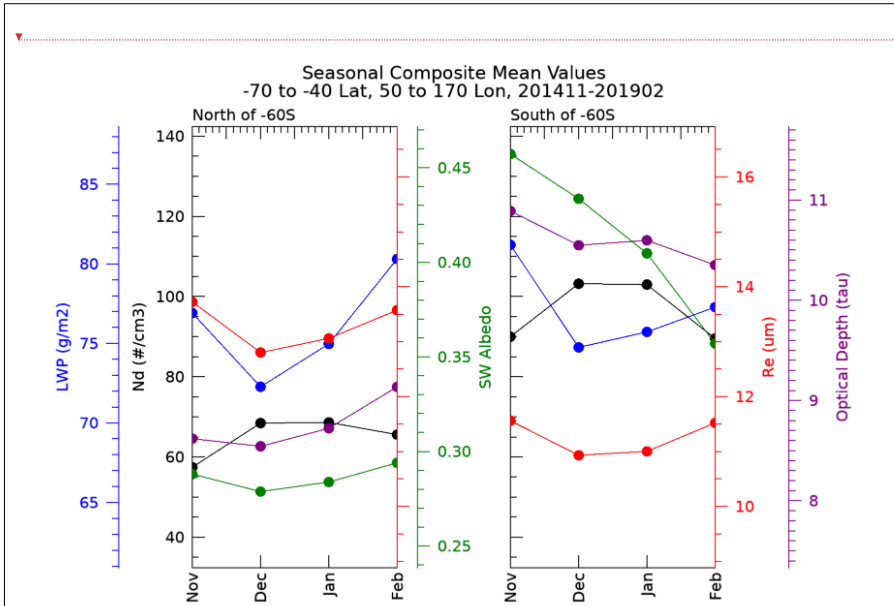
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Figure 5. Composite seasonal cycle of cloud properties. Each data point is comprised of the monthly mean of cloud scenes in the analysis domain compiled over the 5-year period in the figure title. The effective radius (Re) and the optical depth (solid purple curve) are taken directly from MODIS L2 granules. The liquid water path (LWP) and cloud droplet number (Nd) are derived as described in the text. The solar (SW) albedo is derived from CERES data and normalized to a solar zenith angle of 45° as described in the Appendix.

348   Appendix A. Methods
349
350   We use MODIS imager-derived Level-2 retrievals (Platnick et al., 2015) of effective
351   radius (r_e) and optical depth (τ) from five summer periods (2014-2019) collected
352   between the latitudes of 45°S and 76°S and longitudes of 40°E and 170°E to focus
353   roughly on where the ships and aircraft sampled in Summer 2017-18. We calculate N_d
354   using the method derived and evaluated in G18:
355

$$N_d = \frac{\sqrt{5}}{2\pi\kappa}\left(\frac{f_{ad}c_w\tau}{Q_{ext}\rho_w r_e^5}\right)^{1/2} \tag{A1}$$

357
358   where ρ_w is the density of liquid water (1 g cm^{-3}), f_{ad} is an adiabaticity assumption, c_w is
359   the vertical derivative of the adiabatic liquid water content, Q_{ext} is the extinction efficiency
360   that is typically assumed to be 2 for cloud droplets, and κ is the cubed ratio of r_e to r_v. As

364 noted by G18, N_d depends on $r_e^{-5/2}$ which implies that the sensitivity or the rate of change
365 of N_d to retrieved r_e goes as the -7/2 exponent. Any biases in r_e, then would significantly
366 bias N_d. G18 provide a thorough evaluation of the sources of uncertainty in N_d due to
367 assumptions of adiabaticity, scene heterogeneity, etc., and conclude that N_d derived
368 using equation 1 applied to MODIS cloud retrievals has an overall uncertainty of ~80%.

370 The most uncertain quantity in the assumptions used in Equation A1 is f_{ad} since the cloud
371 vertical structure is not constrained by MODIS measurements. Using cloud thickness from
372 ship-based cloud radar and lidar along with retrieved LWP from collocated microwave
373 radiometer (Mace et al., 2021a), we estimate the value of f_{ad} in nonprecipitating
374 stratocumulus observed during the summer of 2018 (Mcfarquhar et al., 2021). We find
375 that the mean and standard deviation of f_{ad} north of the ACFA is 0.66 and 0.48,
376 respectively. South of the ACFA, the mean and standard deviation of f_{ad} is 0.93 and 0.60,
377 respectively. For the calculations of Nd in equation A1, we use a constant value for f_{ad} of
378 0.8. N_d is proportional to the square root of f_{ad}, therefore, $\frac{\partial ln N_d}{\partial \ln f_{ad}} = \frac{1}{2}$ and a fractional
379 variation in f_{ad} on the order of 0.5 would imply an uncertainty in N_d of 0.25. Furthermore,
380 we would expect in regions with f_{ad} higher (lower) than 0.8 the Nd would be biased low
381 (high). As we show, the regions with higher N_d tend be in the south and lower N_d in the
382 north counter to these expected biases. Additionally in this study, we will be examining
383 differences in spatially-averaged N_d that are greater than a factor of 2.   These results
384 imply that bias and random error due to uncertainty in f_{ad} is unlikely to significantly
385 influence the qualitative findings of this study.

387 Another source of systematic bias could be from the quantity κ that can be shown to be
388 a function of the variance of the droplet size distribution and is assumed to be a constant
389 at 0.7. G18 discusses this issue in some detail and concludes that there may be
390 systematic biases on the order of 12% that could be a function of N_d in pristine conditions.
391 While this quantity can be investigated with data collected in situ, no such data exists in
392 stratocumulus clouds south of the ACFA. Therefore, we recognize a potential source of
393 bias due to κ that is likely much smaller than the systematic latitudinal differences we find.

395 Given the uncertainties in N_d at the pixel level, we implement a filtering and averaging
396 scheme to focus on liquid phase, weakly precipitating cloud scenes. We define a scene
397 as a 1° latitude by 2° longitude domain where pixels are of liquid-phase, weakly
398 precipitating clouds if the cloud liquid water path (LWP) < 300 g m^{-2} and the cloud phase
399 is identified as liquid. We require that the sensor and solar zenith angles (θ) at that
400 pixel are less than 30° and 60°, respectively. The maximum θ requirement is motivated
401 by the findings of Grosvenor and Wood (2014) who find that systematic errors in MODIS
402 retrievals increase significantly for θ>60°. The θ requirement causes us to focus on the
403 months from November through February. We require at least 1000 1-km resolution
404 pixels with these characteristics to exist within a scene (typical number >10000). In
405 addition, we require that no more than 10% of the pixels have a cloud top temperature
406 less than -20°C to ensure the absence of ice phase hydrometeors. Cloud properties
407 within a scene are averaged.
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Collocated cloud albedos (A) of the MBL cloud scenes are analyzed. A is derived from the Clouds and the Earth's Radiant Energy System (CERES) Energy Balanced and Filled (EBAF) version 4.0 (Loeb et al, 2018) data collected using instruments on board Aqua and Terra. The albedo is derived by dividing the upwelling shortwave flux at the top of the atmosphere (TOA) by the downwelling shortwave flux at TOA. Because A has a solar zenith angle dependence, (Minnis et al. 1998), we normalize all albedo values to θ=45° (approximately the mean value of θ for the analysis domain and months analyzed) with an empirical method using theoretically calculated A (A) as a function of latitude presented in Minnis et al. (1998 – their figure 7). The normalization is implemented by first approximating the latitudinal dependence of A for various cloud optical depths (τ) using the following regression equation: $A = 0.51 - 0.43\mu_0^{1/2} + 0.17\ln\tau$ where $\mu_0 = \cos\theta$. A approximates the variation of A with latitude within 15% at τ=8. The fit decreases in accuracy at higher and lower τ increasing to an uncertainty of ~30% for τ=2 and τ=32 (these values of τ (2, 8, 32) are those presented in Minnis et al., 1998, Figure 7). The averaged τ of the MBL cloud scenes in our analysis is approximately between 9 and 11 (Figure 5) so we expect that A is typically a reasonable approximation of A. The normalization of all A to $\theta = 45°$ is accomplished by multiplying the CERES A by the ratio $\frac{A(\mu_0(\theta=45),\tau)}{A(\mu_0,\tau)}$ where τ is from the MODIS cloud scene. The magnitude of the ratio applied to the data ranges from 0.85 at higher latitudes to 1.2 at lower latitudes with an average near 1.

Author Contributions: GM led the overall conception, data analysis of the study and interpretation of the results. SB was responsible for implementing data analysis code and generation of figures. RH provided background on aerosol and provided insight regrading various aspects of the study. MPG and ES assisted GM in the study design and implementation.

Competing Interests: The authors declare no conflict of interest.

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