

1 **Using satellite observations to evaluate model**
2 **microphysical representation of Arctic mixed-phase**
3 **clouds**

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8 **Key Points:**

- 9 • CAM6-Oslo and CAM6 capture the vertical structure of Arctic mixed-phase clouds,
10 with supercooled liquid cloud tops overlying icy interiors.
11 • Removing an erroneous limit on heterogeneous and secondary ice nucleation pro-
12 cesses in CAM6 reduces supercooled liquid water in Arctic clouds.
13 • Parametrizations of mixed-phase processes control the longwave cloud feedback
14 parameter by modifying how winter and spring clouds respond to warming.

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Abstract

Mixed-phase clouds play an important role in determining Arctic warming, but are highly parametrized in models and difficult to constrain with observations. We use two satellite-derived cloud phase metrics to investigate the vertical structure of Arctic clouds in two global climate models that use the Community Atmosphere Model version 6 (CAM6) atmospheric component. We report a model error limiting ice nucleation, produce a set of Arctic-constrained model runs by adjusting model microphysical variables to match the cloud phase metrics, and evaluate cloud feedbacks for all simulations. Models in this small ensemble uniformly overestimate total cloud fraction in the summer, but have variable representation of cloud fraction and phase in the winter and spring. By relating modelled cloud phase metrics and changes in low-level liquid cloud amount under warming to longwave cloud feedback, we show that mixed-phase processes mediate the Arctic climate by modifying how wintertime and springtime clouds respond to warming.

Plain Language Summary

[Clouds are important regulators of warming in the Arctic. The thermodynamic phase of a cloud affects its lifetime and transparency to incoming and outgoing radiation. As a result, transitions from ice to liquid in a warming climate change the influence of clouds on surface temperature. At temperatures between -37 °C and 0 °C, both ice and supercooled liquid water may exist simultaneously in a cloud layer. Global climate models struggle to capture cloud phase in this temperature range because it depends on both cloud temperature and aerosol properties. This study investigates how the fraction of supercooled liquid water changes vertically in Arctic clouds, comparing liquid-rich cloud tops with their icy interiors. We describe a significant model error that limits the formation of new ice crystals. We also find that global climate models reproduce observations, and that a range of model parameters produce results consistent with observations. Changes in cloud fraction resulting from these adjustments mostly occur in the winter and spring, and cause the models to trap longwave radiation differently. The results of this study highlight the need to capture seasonal changes in cloud phase and amount in order to successfully predict future changes to the Arctic climate.]

1 Introduction

Uncertainties in cloud and aerosol radiative effects are a principal contributor to climate model uncertainty, and remain so despite decades of model development (Boucher et al., 2013). These uncertainties arise from the difficulty of representing aerosol-cloud interactions and other key physical processes at the typical resolutions of global climate models (GCMs). Evaluations of available models from the Coupled Model Intercomparison Project Phase 6 (CMIP6) (Eyring et al., 2016) indicate that changes in climate sensitivity relative to CMIP5 (Taylor et al., 2012) are mostly due to changes in cloud representation, specifically for extratropical low-level clouds (Zelinka et al., 2020). Using observations to reevaluate the representation of these clouds in the latest generation of GCMs is a vital part of testing the validity of these new predictions.

In the Arctic, clouds mediate climate change through interactions with land and sea ice, and impacts on surface radiative fluxes (H. Morrison et al., 2012). As the thermodynamic phase of Arctic clouds shifts from ice to liquid in response to warming, the radiative effect they exert on the surface changes (Mitchell et al., 1989). This cloud phase feedback depends on cloud optical thickness and lifetime changes. In the Arctic, observations indicate that liquid and ice clouds exert very different radiative forcings on the surface (Shupe & Intrieri, 2004; Cesana et al., 2012), highlighting the need for models to capture cloud phase in order to produce a realistic surface energy budget.

63 At temperatures between approximately $-37\text{ }^{\circ}\text{C}$ and $0\text{ }^{\circ}\text{C}$, cloud ice forms via het-
64 erogeneous nucleation processes that are dependent on temperature, in-cloud vapor pres-
65 sure, and the presence of ice nucleating particles (INPs) (Korolev, 2007). Cloud ice and
66 water can coexist as mixed-phase clouds in this regime. The fraction of supercooled liq-
67 uid water in a mixed-phase cloud layer can be referred to as the supercooled liquid frac-
68 tion (SLF) (Komurcu et al., 2014). Observations show that Arctic mixed-phase clouds
69 are both common and long-lived (Matus & L’Ecuyer, 2017; H. Morrison et al., 2012),
70 due in part to their vertical structure in which INP-limited liquid cloud tops are sepa-
71 rated from glaciated interiors, preventing ice from quickly depleting cloud water and al-
72 lowing clouds to persist for several days (Hobbs & Rangno, 1998). Through this effect
73 on cloud lifetime and opacity, cloudtop phase mediates the resulting long- and shortwave
74 cloud feedbacks.

75 Model representation of mixed-phase clouds relies on uncertain parameters. Sev-
76 eral studies of mixed-phase clouds in Version 5 of the Community Atmosphere Model
77 (CAM5) have found that the Wegener–Bergeron–Findeisen (WBF) process time scale
78 has the largest role in determining liquid cloud fraction (Tan & Storelvmo, 2016; McIl-
79 hattan et al., 2017; Huang et al., 2021), with Huang et al. (2021) finding that reducing
80 the WBF time scale was primarily important in INP-limited environments. High-resolution
81 modelling studies of Arctic mixed-phase clouds also indicate that cloud phase is highly
82 sensitive to ice formation mechanisms and the availability of INPs (Jiang et al., 2000;
83 Fridlind et al., 2007; Fu et al., 2019). Because INP concentrations and the strength of
84 the WBF process specifically impact mixed-phase clouds and have large reasonable ranges,
85 they make appropriate tuning parameters for GCM experiments (Mauritsen et al., 2012).

86 Observations of cloud fraction and phase obtained from the Cloud-Aerosol Lidar
87 with Orthogonal Polarization (CALIOP) sensor aboard the CALIPSO platform provide
88 a strong observational constraint for assessing cloud representation in GCMs (Winker
89 et al., 2009). Cloud phase is especially important, as observations of cloud fraction alone
90 may hide compensating phase biases with large radiative impacts (Cesana & Chepfer,
91 2012). Using CALIOP cloud phase products, Cesana et al. (2012) found that radiatively
92 clear (opaque) atmospheric states in the Arctic were characterized by the absence (pres-
93 ence) of liquid cloud. This result supported the findings of Shupe and Intrieri (2004) and
94 demonstrated that satellite retrievals of cloud phase can be effectively used to study the
95 Arctic surface energy budget.

96 Comparing CAM6 to cloud phase observations is enabled by the use of definition-
97 and scale-aware cloud phase variables produced by the Cloud Feedback Model Intercom-
98 parison Project (CFMIP) Observational Simulator Package: Version 2 (COSP2) (Swales
99 et al., 2018). With new heterogeneous ice nucleation (Hoose et al., 2008) and stratiform
100 cloud microphysics (H. Morrison & Gettelman, 2008) schemes in CAM6 causing signif-
101 icant changes to precipitation and cloud fraction over the Greenland Ice Sheet (Lenaerts
102 et al., 2020), further investigation of cloud representation over the entire Arctic region
103 is well motivated.

104 Recent studies of Arctic feedbacks highlight the importance of the temperature and
105 albedo feedbacks (Pithan & Mauritsen, 2014), noting the lack of a cloud response to sum-
106 mer sea ice loss (A. L. Morrison et al., 2018). While longwave cloud feedbacks are be-
107 lieved to play a secondary role in Arctic warming, large uncertainties associated with mixed-
108 phase processes have yet to be evaluated.

109 In the Arctic, both CMIP5 models and reanalysis data products struggle to repro-
110 duce observed cloud phase and optical depth (Lenaerts et al., 2017). Because many Arc-
111 tic feedbacks are non-linear, this inability to capture the mean state influences projected
112 cloud feedbacks (Goosse et al., 2018). Targeted modeling experiments which capture the
113 observed mean state are valuable tools for disentangling the roles of different Arctic feed-
114 back mechanisms and evaluating model parameterizations (Kay et al., 2016). Tan and

115 Storelvmo (2019) found that minimizing global cloud phase biases in the CESM1 model
 116 yielded a broad range of cloud microphysical variables and Arctic Amplification factors.
 117 This study, however, did not examine whether the global model adjustments created a
 118 reasonable representation of cloud phase in the Arctic itself or distinguish between re-
 119 mote and local drivers of Arctic feedbacks (Feldl et al., 2020). We address this concern
 120 by focusing our model adjustments and analysis on the Arctic and assessing model per-
 121 formance with additional observational constraints. Atmosphere-only simulations iso-
 122 late how the microphysical representation of mixed-phase clouds impacts Arctic warm-
 123 ing. Whereas the fully-coupled simulations of Tan and Storelvmo (2019) investigated the
 124 Arctic impact of a global SLF correction, this work specifically investigates how and why
 125 local cloud microphysics affect the Arctic.

126 2 Methods

127 2.1 Cloud Phase Metrics

128 To investigate the vertical structure of mixed-phase clouds, we filter by overlying
 129 cloud optical thickness (COT) to produce two SLF metrics. We obtain one metric (here-
 130 after: cloud-top SLF) by selecting only the highest layer of observed mixed-phase clouds
 131 after discarding the uppermost layers with $COT < 0.3$ in order to avoid including optically-
 132 thin cirrus clouds. Another metric (hereafter: cloud-bulk SLF) is obtained by selecting
 133 all cloud layers retrieved by CALIOP with overlying COT less than 3.0. Validation of
 134 the cloud bulk metric, as well as methodology for calculating the observational and mod-
 135 elled SLF metrics are described in the Supplementary Material. SLF is calculated on isotherms
 136 from -40°C to 0°C , with a 5°C increment. Measurements of cloud phase were retrieved
 137 from NASA’s CALIOP instrument (Winker et al., 2009) for a four year observational
 138 period from 1 June 2009 through 31 May 2013.

139 2.2 Additional Satellite Products

140 To conduct further model evaluation of cloud fraction and radiative fluxes, we com-
 141 pare against the GCM-Oriented CALIPSO Cloud Product (GOCCP) Version 3 (Chepfer
 142 et al., 2010) and Clouds and the Earth’s Radiant Energy System Energy Balanced and
 143 Filled (CERES-EBAF) Ed4.0 datasets (Kato et al., 01 Jun. 2018). The GOCCP data
 144 product (82°S - 82°N) separates total cloud fraction by phase, and is produced specif-
 145 ically for comparison with the COSP satellite simulator. From CERES-EBAF, we use
 146 computed surface long- and shortwave cloud radiative effect (CRE) values and surface
 147 all-sky downwelling fluxes.

148 2.3 Modeling Simulations

149 We present atmosphere-only runs of the Nordic Earth System Model Version 2 (NorESM2)
 150 (Seland et al., 2020) and the Community Earth System Model Version 2 (CESM2) (Danabasoglu
 151 et al., 2020). Both models have 32 vertical levels and are run at $1.9^{\circ} \times 2.5^{\circ}$ horizontal
 152 resolution. We use identical model components in both GCMs to isolate the impact of
 153 differences between the atmospheric modules. While both models use CAM6, CESM2
 154 uses the MAM4 aerosol scheme (Liu et al., 2016), and NorESM2 uses the OsloAero5.3
 155 (Kirkevåg et al., 2018) aerosol scheme while also parametrizing mid- and high-level ice
 156 clouds differently. Runs of NorESM2 and CESM2 are subsequently referred to as CAM6-
 157 Oslo and CAM6. All modelled data represent averages over the same 4-year period from
 158 which SLF values were calculated following a 3-month model windup to allow the atmo-
 159 sphere to adjust to microphysics changes. To reduce variability in meteorology between
 160 runs, we nudge horizontal winds and surface pressure to ERA-Interim reanalysis data
 161 for the observational period (Dee et al., 2011). Because sea ice concentrations are pre-
 162 scribed, these simulations do not explore cloud-sea ice feedbacks.

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2.4 Model Modifications

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INP availability is an important limiting factor in cloud glaciation at mixed-phase temperatures. In CAM6 and CAM6-Oslo, the in-cloud ice number concentration cannot exceed the calculated concentration of available ice nuclei. Neither heterogeneous nor secondary nucleation processes contribute to this INP limit, preventing them from nucleating ice crystals. Heterogeneous nucleation processes are still able to increase ice crystal mass, however, and can artificially inflate ice crystal size and increase sedimentation. This model error has been shared with model developers and identified as an issue to be resolved in future releases of CAM (personal communications, A. Gettelman, 2021) (Gettelman, 2021), and one global correction significantly alters the climate sensitivity of CESM2 (Zhu et al., 2021).

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To assess the importance of this model mechanism on cloud properties and ice number concentration and size, we disable the ice number limit at mixed-phase temperatures ($-37^{\circ}\text{C} < T < 0^{\circ}\text{C}$) in CAM6-Oslo, restoring heterogeneous ice production as well as secondary ice production through the Hallet-Mossop process, and producing an additional model variation we label as CAM6-OsloIce. To isolate the impact of heterogeneous nucleation, we also cap the ice number tendency variable from secondary ice production in CAM6-OsloIce to avoid strong secondary production in the absence of the ice number limit, describing this implementation in the Supplementary Material. To focus on Arctic clouds, these changes are made only in the Arctic Circle (latitude $> 66^{\circ}\text{N}$). Whereas mixed-phase clouds in CAM6 are strongly (and potentially unrealistically) INP-limited by the ice number limit, CAM6-OsloIce serves as an alternate ensemble end-member representing the assumption that transport and entrainment bring in fresh INPs to replace those that nucleated ice in the previous model time step.

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Using nudged 12-month simulations, we iteratively modify the time scale of the WBF process and the number of aerosols active as INPs in the base models (CAM6-Oslo, CAM6, and CAM6-OsloIce) to minimize root-mean-square error with the SLF metrics. Methodology for parameter tuning ranges and validation of ice crystal concentrations is described in the Supplementary Material. This approach produced four “fitted” models that agree with observations. Table 1 summarizes the three base models and four “fitted” models presented in this work, as well as six ancillary simulations in which single parameters were tuned to the “fitted” values. When the ice limit is in place, large INP multipliers increase ice crystal size and decrease the ice number concentration (CAM6 Fit 4 vs CAM6, CAM6-Oslo vs CAM6-Oslo(1,10)), demonstrating non-physical behavior caused by the model error. Conversely, runs without an ice number limit have smaller ice crystals and higher concentrations than those with the limit in place. Ice crystal size and concentration variables in the constrained runs (Table 1) fall near observed ranges from the M-PACE experiment (Prezzi et al., 2009), preventing us from discarding any of the simulations.

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2.5 Radiative Feedback Calculations

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We use surface radiative kernels from Soden et al. (2008) to calculate long- and short-wave cloud feedback parameters. We repeat each simulation from Table 1 with prescribed sea surface temperatures increased by 4K globally to create perturbed runs for the radiative feedback calculations. Because we run atmosphere-only simulations and modify models only poleward of 66°N , feedback parameters are calculated with respect to the temperature change in the Arctic rather than the global mean.

Run name	Model	Ice Number Limit (Secondary Ice Limit)	WBF Multiplier	INP Multiplier	Average Ice Radius at 860 hPa (μm)	Ice Concentration at 860 hPa (m^{-3})
CAM6-Oslo	NorESM2	Yes (No)	1.0	1.0	151	4120
CAM6	CESM2	Yes (No)	1.0	1.0	165	5550
CAM6-OsloIce	NorESM2	No (Yes)	1.0	1.0	132	15670
CAM6-Oslo Fit 1	NorESM2	Yes (No)	1.25	10.0	163	3870
CAM6-OsloIce Fit 2	NorESM2	No (Yes)	0.5	0.05	124	5410
CAM6-OsloIce Fit 3	NorESM2	No (Yes)	0.2	0.1	112	8600
CAM6 Fit 4	CESM2	Yes (No)	1.0	100	209	5060
CAM6-Olso(1.25,1)	NorESM2	Yes (No)	1.25	1.0	156	3950
CAM6-Olso(1,10)	NorESM2	Yes (No)	1.0	10.0	160	4020
CAM6-OsloIce(0.2,1)	NorESM2	No (Yes)	0.2	1.0	112	23460
CAM6-OsloIce(0.5,1)	NorESM2	No (Yes)	0.5	1.0	123	16540
CAM6-OsloIce(1,0.05)	NorESM2	No (Yes)	1.0	0.05	134	4250
CAM6-OsloIce(1,0.1)	NorESM2	No (Yes)	1.0	0.1	134	4430

Table 1. Model run descriptions. Prenni et al. (2009) reported an average INP concentration of 700m^{-3} and a maximum INP concentration 6000m^{-3} from the M-PACE experiment.

208 3 Results

209 3.1 SLF Metrics

210 Figure 1(a) shows the SLF metrics from CALIOP observations and the base mod-
211 els. In the CALIOP retrievals, cloud-top SLF is greater than cloud-bulk SLF values for
212 all isotherms between -35°C and -10°C . At -20°C where this difference is the most pro-
213 nounced, cloud-top SLF exceeds the cloud-bulk value by nearly a factor of three. All mod-
214 els reproduce the structure of icier cloud interiors, but with varying degrees of quanti-
215 tative agreement with CALIOP: CAM6-Oslo shows strong agreement across both met-
216 rics, CAM6 overestimates SLF in cloud tops, and CAM6-OsloIce underestimates SLF
217 along both the cloud-top and cloud-bulk SLF metrics. The poor performance of CAM6-
218 OsloIce results from a high ice number concentration that allows liquid water to be quickly
219 depleted. Differences in SLF between CAM6-Oslo and CAM6-OsloIce mostly occur at
220 cold isotherms where secondary ice processes are not active, showing that freed hetero-
221 geneous nucleation processes are responsible for the cloud changes between these mod-
222 els and that environments with few INPs are necessary for maintaining liquid cloud tops
223 below 20°C . Figure 1(b) shows SLF metrics for CALIOP and the fitted models.

224 3.2 Evaluation against CALIOP-GOCCP and CERES-EBAF data prod- 225 ucts

226 Monthly averages of cloud fraction by CALIOP phase designation allow us to iden-
227 tify seasonal trends and biases (Figure 2). We find that fitting to the SLF metrics brings
228 CAM6-Oslo and CAM6-OsloIce models into good agreement with each other, indicat-
229 ing that the effect of removing the limit on ice number can be compensated for with the
230 adjustment of the WBF and INP parameters. In the summer and early fall, the total
231 cloud fraction and the liquid and ice components are consistent across all models, with
232 an overestimation of liquid and total cloud fraction during June and July. Differences
233 between models emerge in the winter and spring months. CAM6-OsloIce, CAM6-Oslo
234 Fit 1, CAM6-Oslo Fit 2, and CAM6 Fit 4 all produce insufficient total cloud fraction dur-
235 ing the winter, while CAM6 produces excess total cloud fraction. All models fail to cap-
236 ture the total cloud fraction in the spring, mostly due to insufficient ice cloud fraction.

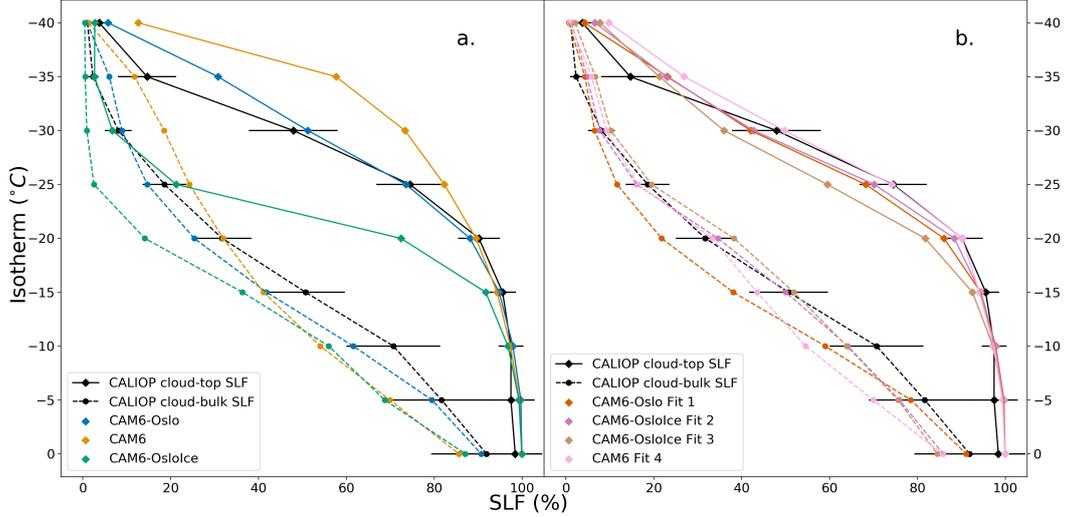


Figure 1. Supercooled liquid fraction by isotherm for cloud-top and cloud-bulk metrics for (a) base models and (b) fitted models. Error bars on CALIOP SLF values correspond to one standard deviation. All values represent an area-weighted average from 66°-82°N.

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Finally, a positive liquid cloud bias in CAM6 persists throughout the year and is especially pronounced during the winter.

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Annual model biases in Arctic-averaged cloud fraction and CRE with respect to CALIOP-GOCCP and CERES-EBAF (Table S1) follow the results of Figure 2. Notable compensating biases in cloud fraction by phase are present, with CAM6 producing excess liquid cloud and insufficient ice cloud, and CAM6-OsloIce producing excess ice cloud and insufficient liquid cloud. CAM6 Fit 4 shares the ice cloud bias of CAM6 despite having good agreement with the observed SLF metrics because positive biases in mid- and high-level ice clouds are unaffected by the model adjustments. Polar projections of model cloud phase biases (Figure S2) show the spatial features of model cloud phase biases.

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All simulations have negative shortwave CRE biases between -2.9 and -5.1 Wm^{-2} in the annual mean due to the shared positive cloud fraction bias in the summer. Downwelling shortwave surface flux and CRE biases (Fig. 3(a) and (b)) strongly resemble each other, confirming that clouds are responsible for the shortwave biases. Excessive summer cloudiness is largely independent of mixed-phase processes, since low-level liquid clouds make up roughly two thirds of the total cloud fraction during the summer months in our simulations (Fig. S3). Because the shortwave CRE is similar between model runs, changes in mixed-phase microphysics impact the Arctic mainly through the longwave CRE in the winter and spring.

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Like the shortwave, the downwelling longwave surface flux and CRE biases (Fig. 3(c) and (d)) are also highly similar. There is strong seasonal variation in the longwave biases, with excess downward flux from clouds in the summer and insufficient downward flux in the winter. The positive summer biases occur when all models produce excess cloud fraction, but the negative wintertime biases occur even in the models that capture both cloud fraction and phase well. CAM6 is the only model to capture the downward flux despite overproducing winter cloud fraction, suggesting biases in cloud height and emission temperature across all simulations. While passive sensors and their corresponding satellite simulators are poorly suited to constrain this behavior, cloud height and opacity variables recently incorporated into the COSP2 Lidar simulator will allow this win-

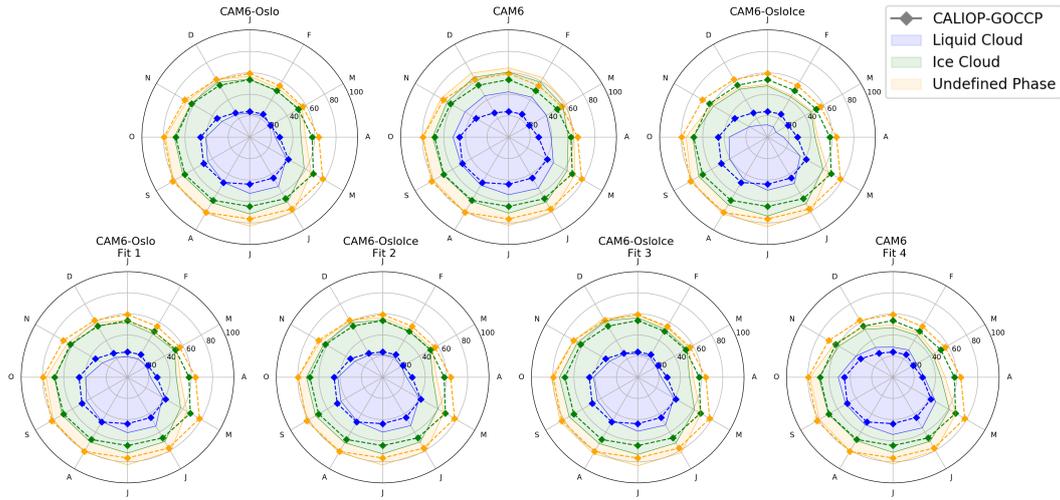


Figure 2. Comparison of monthly cloud fraction by phase. Model values are produced by the COSP CALIPSO satellite simulator and observations are from the CALIPSO-GOCCP cloud phase products. The different cloud phase components are stacked on top of each other so that the outermost contour gives the total cloud fraction and compensating phase biases can be easily visualized.

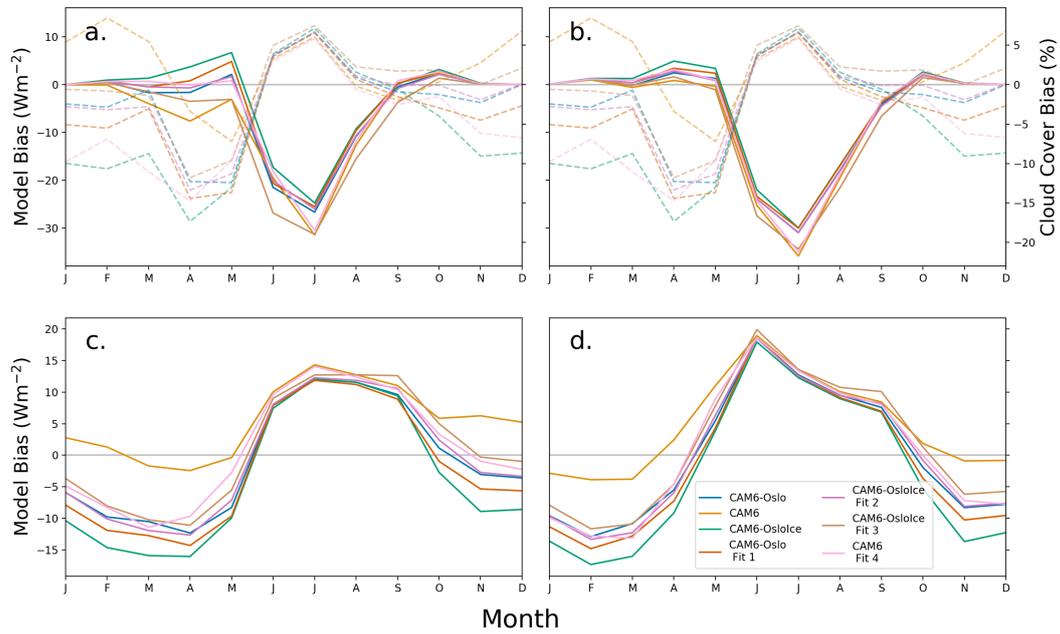


Figure 3. Monthly values for: (a) Model bias in shortwave downwelling flux at the surface (solid) and total cloud fraction (dashed), (b) Model bias in surface shortwave cloud radiative effect (solid) and total cloud fraction (dashed), (c) Model bias in longwave downwelling flux at the surface, (d) Model bias in surface longwave cloud radiative effect.

266 tertime bias to be investigated in future versions of CAM (Guzman et al., 2017; A. L. Mor-
 267 rison et al., 2019).

268 3.3 Cloud Radiative Feedbacks

269 Computing cloud radiative feedbacks allows us to assess the relative importance
 270 of the long- and shortwave cloud feedback processes and to investigate their dependence
 271 on the present-day cloud state and cloud microphysical properties. Figure 4(a) shows
 272 the long- and shortwave cloud feedback parameters and the net cloud feedback for each
 273 model simulation.

274 Differences in present day cloud fraction and phase are greatest in the winter and
 275 spring, suggesting that mixed-phase processes impact cloud feedbacks most during this
 276 time. We investigate how low-level (cloud-top pressure > 680 mb) liquid cloud fraction
 277 changes between the present day and +4K warming simulations (Figure 4(b)) correlate
 278 with the longwave cloud feedback parameter, finding that the average change in low-level
 279 liquid cloud fraction from November through April is strongly correlated with the long-
 280 wave feedback ($R^2 = 0.92$) (Figure 4(c)). Individual correlations by month indicate that
 281 this pattern is consistent from November through April, whereas no correlation is found
 282 from May through October (Figure S4).

283 To investigate why different parametrizations give rise to these feedbacks, we pro-
 284 pose that the slope of the SLF curves in Fig. 1 should determine how quickly liquid clouds
 285 replace ice clouds during warming. Since the SLF metrics change little in our +4K sim-
 286 ulations (Fig. S5), cloud phase changes occur because atmospheric warming shifts clouds
 287 to warmer isotherms where liquid replaces ice. The SLF metrics effectively describe the
 288 sensitivity of cloud phase to warming, potentially offering a predictor of longwave cloud
 289 feedback. We approximate this sensitivity as the change in SLF between the -10C and
 290 -30C isotherms and correlate this value with the LW cloud feedback (Fig. 4(d)). All but
 291 one model (CAM6 Fit 4) show agreement with the predicted relationship. We hypoth-
 292 esize that large ice crystal sizes in CAM6 Fit 4 (Table 1) depress ice cloud lifetimes and
 293 lead to a much larger increase in low-level liquid cloud amount under warming, but leave
 294 this question open for future work. Excluding CAM6 Fit 4 from our analysis (Fig. 4(d)
 295 grey lines and text) demonstrates that the SLF metric can effectively predict the LW cloud
 296 feedback. These results demonstrate that different parametrizations of mixed-phase pro-
 297 cesses impact Arctic warming by modifying how low-level liquid clouds increase in re-
 298 sponse to warming.

299 4 Discussion and Conclusion

300 We find large differences in thermodynamic phase between cloud tops and interi-
 301 ors in satellite observations of Arctic mixed-phase clouds, consistent with previous ground-
 302 based measurements. CAM6-Oslo captures this vertical phase structure better than CAM6,
 303 suggesting that model aerosol schemes and high cloud parameterizations play an impor-
 304 tant role in determining cloud phase. We evaluate a significant model error that prevents
 305 heterogeneous and secondary nucleation processes from creating new ice crystals and find
 306 that cloud water is significantly reduced when these nucleation processes are able to op-
 307 erate freely.

308 All models produce insufficient cloud fraction in the spring and excess cloud frac-
 309 tion in the summer. The summer bias dominates the shortwave impact, leading to a net
 310 negative annual shortwave flux bias. Longwave flux biases are strongly seasonal, with
 311 positive summer biases explained by excess summer cloud fraction and negative winter
 312 biases likely resulting from low-biased cloud emission temperatures. The greatest vari-
 313 ation between models occurs in the winter and spring, and changes in low-level liquid
 314 cloud fraction under warming during these seasons are strongly correlated with the long-

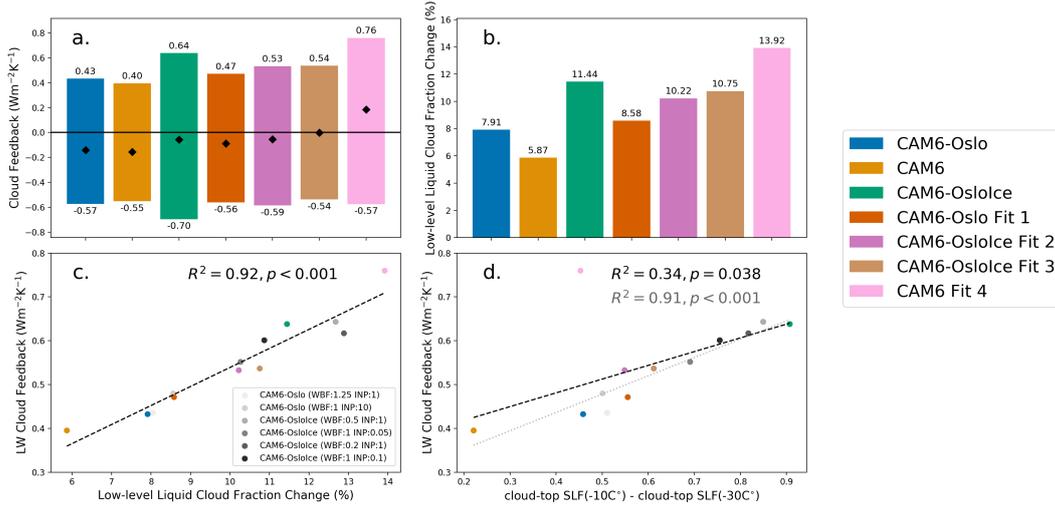


Figure 4. a. Arctic-averaged longwave and shortwave cloud feedback. Diamonds denote the net cloud feedback. Kernel calculations do not incorporate surface albedo changes with mean state when calculating shortwave cloud feedback and tend to overestimate the shortwave cooling effect of clouds at high latitudes. b. Arctic-averaged change in low-level liquid cloud fraction between present-day and +4K simulations. c. Longwave cloud feedback as a function of the mean change in low-level liquid cloud fraction from November to April. d. Longwave cloud feedback as a function of the difference in cloud-top SLF between the -10°C and -30°C isotherms. In panel d, grey text and best fit lines represent analysis excluding CAM6 Fit 4.

315 wave cloud feedback parameter. Our results indicate that the parametrization of mixed-
 316 phase microphysics influences the Arctic climate by modifying how winter and spring
 317 clouds respond to warming, and that cloud phase observations may constrain this rela-
 318 tionship. Positive longwave cloud feedback associated with winter cloud fraction increases
 319 was also observed in fully-coupled simulations of CESM1 (A. L. Morrison et al., 2019),
 320 raising to question the relative contributions of sea ice loss and simple surface warming
 321 to these winter feedbacks.

322 Our results demonstrate the need to capture local cloud phase processes in order
 323 to understand how mixed-phase cloud processes impact Arctic warming. Future stud-
 324 ies should use multiple atmospheric components, and use fully-coupled models to deter-
 325 mine whether the proposed constraint is valid in a dynamic climate system.

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336 CALIOP and CERES-EBAF data are available online at the NASA Langley At-
 337 mospheric Sciences Data Center website (<https://asdc.larc.nasa.gov/>). The CALIOP GOCCP

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observational data set can be downloaded from <https://climserv.ipsl.polytechnique.fr/cfmip-obs/>. The ERA-Interim reanalysis data can be downloaded freely (<https://www.ecmwf.int/en/forecasts/datasets/datasets/era-interim>).

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