

1 **Cloud-induced stabilization of Greenland surface melt**

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5 **Key Points:**

- 6 • Clouds warm Greenland's surface in the accumulation zone on monthly, seasonal, and
7 climatological timescales
- 8 • In the ablation zone surface albedo decreases and drives net cloud radiative effects to-
9 ward cooling, leading to net cloud-induced cooling in the southern ablation zone
- 10 • As Greenland warms and darkens, clouds increasingly cool its surface, and reduce en-
11 ergy available for melt

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Abstract

Net cloud radiative effects (CREs) on Greenland's surface energy budget result from shortwave shading and longwave heating. We use tilt-adjusted automatic weather station data to show that where surface albedo is bright and close to cloud albedo, shading is suppressed and heating dominates. When and where albedo declines (e.g., due to temperature-induced snow metamorphism and/or melt), shading increases more rapidly than cloud greenhouse effects, leading to net cloud-induced cooling. Such cooling is common at lower albedos in the ablation zone, where the 0.57 albedo isopleth distinguishes areas of positive from negative CREs. During extensive surface melt across the Greenland ice sheet (GrIS) in 2012, clouds exerted anomalously strong cooling in the southern ablation zone, and only climatological-mean warming in the accumulation zone. Clouds reduced more than promoted surface melt during summer 2012. These results suggest that clouds will increasingly inhibit surface melt and reduce GrIS dimming in future melt seasons.

1 Introduction

Greenland's surface melt has accelerated in the past 30 years [Tedesco *et al.*, 2013; Velicogna and Wahr, 2013]. The triggered mass-loss and energy-related feedbacks have great potential to affect global climate in the long-term through, e.g., sea-level rise. Clouds strongly influence surface radiative fluxes in the Arctic due to its persistent cloudiness [Intrieri *et al.*, 2002; Morrison *et al.*, 2011; Cesana *et al.*, 2012; Shupe *et al.*, 2013], dry atmosphere, and highly reflective surface [Shupe and Intrieri, 2003]. Clouds may have exerted a profound influence on recent dramatic sea-ice loss and snow melt events [Kay *et al.*, 2008; Schweiger *et al.*, 2008; Bennartz *et al.*, 2013; Kapsch *et al.*, 2015; Letterly *et al.*, 2016; van Tricht *et al.*, 2016]. However, the dearth of coordinated surface-cloud observations in remote and harsh Arctic conditions [Liu *et al.*, 2004; Shupe *et al.*, 2013] has caused cloud radiative effect studies that relate cloud properties directly to the surface energy budget and thus surface melt to be case-dependent and inconclusive on climatological timescales. Recent progress [Wang *et al.*, 2016] in correcting tilt-induced biases in radiometric observations over snow covered surfaces allows us to study cloud effects on GrIS surface processes with unprecedented accuracy on climatological timescales.

Cloud radiative effects (CREs) on the surface energy budget result from a complex interaction between cloud properties (e.g., cloud fraction, cloud water path, and cloud droplet shape and size) and their environment (e.g., surface albedo, moisture profile, solar zenith angle) [Curry *et al.*, 1996; Shupe and Intrieri, 2003]. For example, Kay *et al.* [2008] attributes widespread Arctic melting in 2007 to reduced summer cloudiness, which allows more insolation to heat the surface. In contrast, Bennartz *et al.* [2013] argue that warming by shortwave-thin yet longwave-opaque clouds helped to trigger Greenland's widespread surface melt in July, 2012. Given the dynamic nature of snow-covered surfaces, establishing the climatological role of surface CREs on the Greenland Ice Sheet (GrIS) requires *in situ* observations with long coverage and high temporal resolution.

The 30+ automatic weather stations (AWSs) deployed since 1995 on the GrIS provide the best data to study surface CREs there. These AWSs, which measure hourly surface radiation and weather, are arranged from near glacier termini in the ablation zone to Summit atop the GrIS. We use AWS observations [Wang *et al.*, 2016], complemented by clear-sky simulations from a Column Radiation Model (CRM) [Zender, 1999], to estimate GrIS CREs from 2008–2013, to assess the leading factors that determine them, and to investigate their influence on the GrIS surface energy budget and thus surface melt. It follows from the definition of CREs and is true worldwide that surface albedo contrast with cloud albedo strongly affects the net result of cloud-induced shortwave (SW) cooling and longwave (LW) warming at the surface. For the GrIS in particular, AWS data show that as albedo decreases due to surplus surface energy, cloud-induced cooling eventually outweighs warming and acts to stabilize the warmest and darkest regions. This process embodies a negative feedback between CREs and surface warming with implications for current and future surface melt on the GrIS.

63 **2 Data and Methods**

64 Surface CRE measures the direct radiative effect of clouds on the surface energy bud-
 65 get. Indirect effects of clouds on surface temperature and humidity are not included. Positive
 66 CRE indicates that clouds warm the surface, and negative CRE indicates cloud-induced sur-
 67 face cooling. We define AWS CRE as the difference of AWS-measured all-sky radiative fluxes
 68 and CRM-simulated clear-sky radiative fluxes at the surface, assuming other meteorological
 69 conditions remain constant.

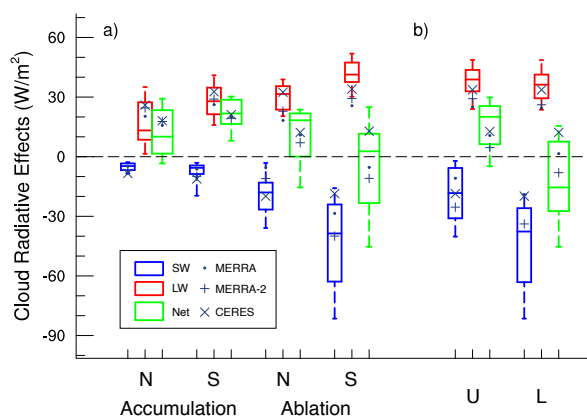
70 Most AWSs on GrIS are unattended for at least one year at a time, and their data must
 71 be interpreted with care. Of the total 25 long-term AWSs we use here, seven are from the Green-
 72 land Climate Network (GC-Net), mostly in the accumulation zone (elevation > 2000 m) [*Stef-*
 73 *fen et al.*, 1996]. Another 18 stations are from the Programme for Monitoring of the Green-
 74 land Ice Sheet (PROMICE), almost all of which are in the ablation zone (elevation < 2000 m)
 75 [*van As and Fausto*, 2011]. The PROMICE stations are usually arranged in pairs, with one sta-
 76 tion near the equilibrium line and the other well into the ablation zone [*van As and Fausto*, 2011].
 77 Stations from these two networks with less than 10 months of qualified data during the melt
 78 seasons from 2008 to 2013 are excluded. The main source of bias in shortwave measurement
 79 is station tilt caused by spatially heterogeneous snow melt, snow compaction, and glacier dy-
 80 namics [*van den Broeke et al.*, 2004; *van As*, 2011; *Stroeve et al.*, 2013; *Wang et al.*, 2016]. *Wang*
 81 *et al.* [2016] further describes our methods and details how we adjust insolation measured by
 82 tilted AWSs using the Retrospective, Iterative, Geometry-Based (RIGB) tilt-correction method,
 83 to remove per-station mean-absolute biases that average 18 W m^{-2} over the GrIS during melt
 84 seasons. “Dome-heating”—absorption of shortwave radiation and heating of pyrgeometer sil-
 85 icon windows—causes the largest biases in AWS-measured longwave surface fluxes [*van den*
 86 *Broeke et al.*, 2004]. It is impossible to estimate this bias accurately without knowing the win-
 87 dows temperature. In this study we remove the maximum window heat offset from the long-
 88 wave AWS measurements— 25 W m^{-2} out of 1000 W m^{-2} of insolation according to the sen-
 89 sor’s manual [*Kipp & Zonen*, 2003]— and potentially underestimate CREs.

90 Clear-sky surface radiative fluxes are simulated every 3 hours using a CRM [*Zender*, 1999]
 91 driven by atmospheric profiles from the second version of Modern-Era Retrospective Anal-
 92 ysis for Research and Applications (MERRA-2) [*Rienecker et al.*, 2011]. Results were eval-
 93 uated against high-quality reference radiometric observations from the Atmospheric Radiation
 94 Measurement (ARM) site at Barrow, Alaska, U.S.A [*Atmospheric Radiation Measurement (ARM)*
 95 *Climate Research Facility*, 1994]. Throughout the study period, the absolute shortwave sim-
 96 ulation discrepancy between ARM measurements and CRM simulations is less than 1 W m^{-2} .
 97 However, the longwave discrepancy is about 25 W m^{-2} , of which 11 W m^{-2} is explained by
 98 the 5 K difference in temperature profiles and the 1 kg m^{-2} difference in humidity profiles be-
 99 tween MERRA-2 and local radiosonde measurements. To work around the LW simulation bias,
 100 we employ the Stefan-Boltzmann equation and use the atmospheric emissivity from the av-
 101 erage of all the clear days in one month to estimate the longwave downwelling radiation un-
 102 der clear-sky conditions in that month. This adjustment reduces the mean LW discrepancy to
 103 about 5 W m^{-2} .

104 We also use monthly radiative fluxes (to estimate CREs independently of AWSs) and
 105 cloud macro-properties from the first and second version of the Modern-Era Retrospective Anal-
 106 ysis for Research and Applications (MERRA and MERRA-2) [*Rienecker et al.*, 2011], and the
 107 Clouds and the Earth’s Radiant Energy System (CERES) [*CERES Science Team*, 2015]. Monthly
 108 MERRA-2 radiative fluxes behave well. However, hourly MERRA-2 clear-sky insolation fluxes
 109 contain an inexplicable and artificial shift of at least one hour in local solar noon time on 10%
 110 of the days during our study period near some AWSs. Hence we rely on monthly rather than
 111 hourly data for MERRA-2-estimated CREs.

112 **3 Results**

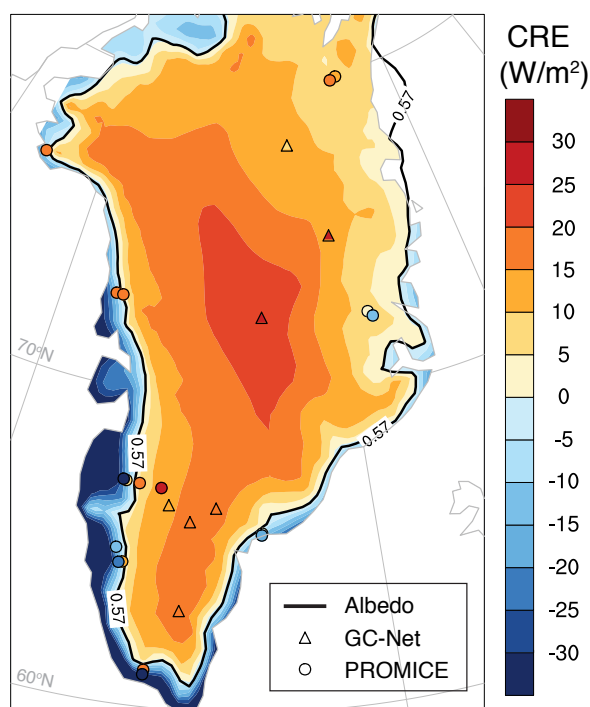
118 We estimate CRE separately for accumulation and ablation-zone stations in the north (N)
 119 and south (S) (separated by latitude = 70°N) during the melt season (May–Sept) for 2008–
 120 2013 (Fig. 1a). Net CRE remains positive in almost all months in the highly reflective accu-
 121 mulation zone, where the average surface warming by clouds is 21 W m⁻². CRE becomes neg-
 122 ative in a greater number of months in the less reflective ablation zone, especially in the south
 123 (-6 W m⁻²) and at low (L) elevations (-12 W m⁻²; Fig. 1b). In the accumulation zone, surface
 124 albedo is consistently high (0.81±0.04), from the northern stations to the southern stations.
 125 Where high albedo suppresses SW CRE, LW CRE determines net CRE. The decrease of SW
 126 CRE drives net CRE changes in the ablation zone with its lower and more variable albedo (0.57±0.19).
 127 LW CRE increases from north to south in the ablation zone, probably due to the extra water
 128 vapor availability. Lower elevation (L) stations have albedos 0.18 less than their upper (U) el-
 129 evation counterparts. SW CRE outweighs LW CRE in most months at low (L) stations, and
 130 produces net surface cooling. CREs estimated using MERRA (dots) and MERRA-2 (pluses)
 131 radiative fluxes generally agree with AWS CREs. CREs from CERES (crosses) are less con-
 132 sistent with the other three, probably due to the coarser CERES spatial resolution that does
 not resolve areas of heterogeneous albedo (Fig. S1).



113 **Figure 1.** Monthly CREs in a) accumulation zones (North and South) and ablation zones (North and
 114 South), and b) at upper (U) and lower (L) PROMICE stations. SW, LW and net AWS CREs are represented
 115 by blue, red, and green boxes, respectively. Whiskers show min and max, box lines show upper quartile,
 116 median, and lower quartile. CREs estimated from MERRA, MERRA-2, and CERES data are overlaid as dots,
 117 pluses and crosses, respectively.

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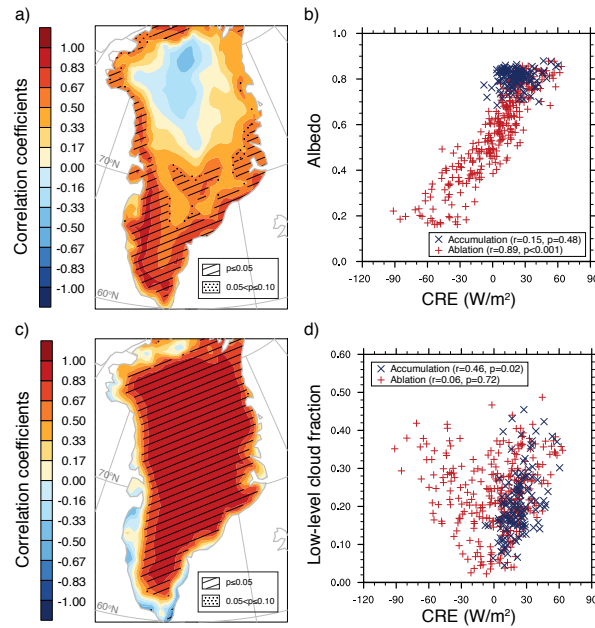
137 AWS CREs generally agree well with MERRA-2 CREs (Fig. 2). Their Root-Mean-Square-
 138 Difference (RMSD) is 18 W m⁻², comparable to the RMSD with MERRA (15 W m⁻²) and
 139 CERES (19 W m⁻²; Fig. S1a and c). The albedo = 0.57 isopleth distinctly separates, with 99%
 140 accuracy, areas of positive from negative MERRA-2 summertime CREs. Above this thresh-
 141 old, CRE is positive and dominated by LW effects; below the threshold, CRE is negative and
 142 dominated by SW effects. The threshold albedo shifts with the availability of insolation from
 143 0.63 in June to 0.50 in August. It drops to 0.25 in September, when clouds begin to exert net
 144 surface warming throughout the fall over the entire GrIS (except bare rock). The albedo thresh-
 145 old of 0.57 measured by AWSs also successfully separates stations of positive from negative
 146 AWS CRE.



134 **Figure 2.** AWS CREs (GC-Net in triangles and PROMICE in circles) averaged from May–August, 2008–
 135 2013, superimposed on MERRA-2 CREs (color contours). The isopleth where albedo = 0.57 is the solid black
 136 line.

151 Clouds reduce surface insolation and thus SW CREs intensify (become more negative)
 152 over lower albedo surfaces where cloud reflectance prevents greater absorption. This effect is
 153 most pronounced in the ablation zone. The temporal correlation between monthly surface albedo
 154 and MERRA-2 CREs from 2008–2013 is high and significant (Fig. 3a). Monthly AWS CREs
 155 in the ablation zone are also significantly ($r = 0.89$, $p < 0.001$) correlated with albedo (Fig. 3b).
 156 At high elevations, cloud shielding is limited by the comparable albedos of the surface and
 157 clouds. In these areas, cloud temperature, phase, and micro-properties are important to LW
 158 CRE and thus net CRE [Shupe and Intrieri, 2003]. Nevertheless, cloud fraction, and low-level
 159 cloud fraction in particular, is the primary factor controlling CRE temporal and spatial vari-
 160 ability. The temporal correlation between low-level cloud fraction and CRE is high and sig-
 161 nificant throughout the accumulation zone (Fig. 3c). This correlation is also significant ($r =$
 162 0.46 , $p = 0.02$) at AWSs (Fig. 3d). Low-level clouds are often present in strong tempera-
 163 ture inversions [Shupe and Intrieri, 2003], and can efficiently warm the relatively cold surface.
 164 The secondary factor controlling GrIS CRE variability in the accumulation zone is condensed
 165 water path, especially its liquid component. Clouds containing super-cooled water are ubiq-
 166 uitous and persistent in the Arctic [Morrison *et al.*, 2011; Cesana *et al.*, 2012; Bennartz *et al.*,
 167 2013]. Many small water droplets in mixed-phase Arctic clouds are radiatively more efficient
 168 than fewer large ice crystals [Shupe and Intrieri, 2003]. The temporal correlation between cloud
 169 water path and CREs is significant across a large area and at many stations in the accumula-
 170 tion zone (Fig. S2).

173 In the ablation zone, albedo change explains most seasonal and inter-annual CRE vari-
 174 ability (Fig. 4), with correlation coefficients of 0.90 and 0.96, respectively (Fig. S3). During
 175 minimum CRE months (July and August), clouds exert net cooling on the surface. CRE is low-

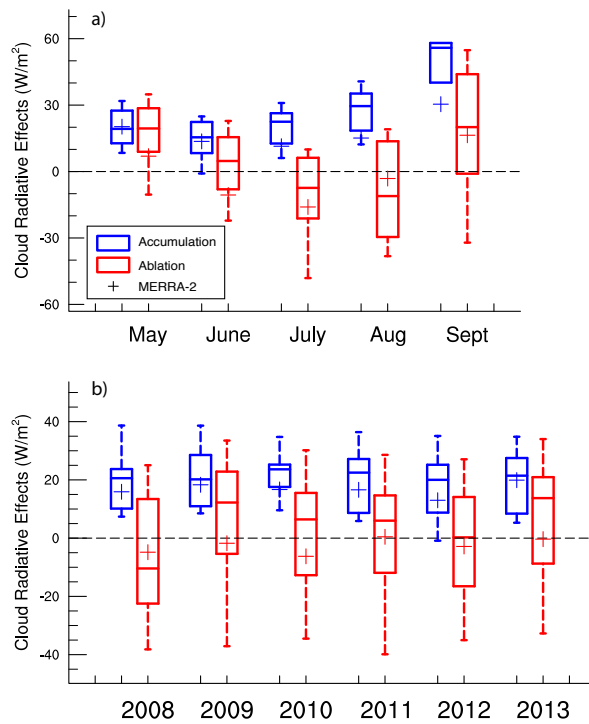


147 **Figure 3.** Temporal correlation between monthly albedo and CRE in a) MERRA-2 and b) AWS mea-
 148 surements, and correlation between low-level cloud fraction and CRE in c) MERRA-2 and d) AWS mea-
 149 surements. and those significant at the 5% level are shaded by lines in a) and c). In b) and d), CREs in the
 150 accumulation and ablation zones are represented by blue crosses and red pluses, respectively.

176 est in June in the accumulation zone, and increases afterwards until September, likely as a re-
 177 sult of changes in both low-level cloud fraction and liquid water path. However, the slightly
 178 decreasing inter-annual CRE trend in the accumulation zone is not explicable by concurrent
 179 changes in cloud macro-properties (i.e., cloud fraction, water/ice content, height and temper-
 180 ature). Longer observations are needed to confirm the significance of this trend, and reliable
 181 measurements of cloud micro-properties may help resolve this enigma. Examination of CREs
 182 from AWS and MERRA-2 for 2008–2013 reveals that CREs are unlikely to have significantly
 183 contributed to the unparalleled surface melt on GrIS in 2012 [Nghiem *et al.*, 2012]. In the
 184 ablation zone, the low summer albedo reduces CREs (Fig. 4b) which becomes negative (net cool-
 185 ing) in the southern ablation zone (Fig. S3). CREs in the accumulation zone in 2012 are close
 186 to the long-term mean. Our results do not support Benmartz *et al.* [2013] who argue, based on
 187 measurements at Summit, that low-level liquid clouds played a key role in July 2012 Green-
 188 land’s melt. We found no anomalous cloud-induced heating in the accumulation zone in July,
 189 2012. Moreover, at low elevations more susceptible to surface melt, clouds exerted net cool-
 190 ing.

191 4 Discussion

192 GrIS surface albedo has been decreasing [Box *et al.*, 2012; Polashenski *et al.*, 2015], and
 193 the trend is expected to continue as climate warms and subjects the surface to accelerated metaphor-
 194 phism [Flanner and Zender, 2006]. Our analysis of AWS and reanalysis data shows that this
 195 will enhance cloud SW cooling effects and thus diminish net CRE. Moreover, as the liquid
 196 water path increases in the warmer and wetter Arctic [Tedesco *et al.*, 2013; Screen and Sim-
 197 monds, 2010], the greenhouse effect of clouds will become saturated while the shading effect
 198 of clouds will continue to increase [Shupe and Intrieri, 2003]. It is therefore likely that sur-



171 **Figure 4.** Variabilities of CREs in a) seasons and b) long-term in the accumulation zone (blue) and in the
 172 ablation zone (red). Box and whiskers show AWS CREs and pluses show MERRA-2 CREs.

199 face cooling by clouds will outpace cloud-induced surface warming over larger areas and in
 200 longer periods on the GrIS. Cloud property-induced variability in CREs is currently small relative
 201 to surface albedo-induced variability over the GrIS. So long as this continues, enhanced
 202 SW CREs expected in the future will reduce surface energy available for melt and/or metaphor-
 203 phism, and will act as a negative feedback on surface melt.

204 5 Summary

205 Surface albedo largely determines the net outcome between cloud-induced surface warm-
 206 ing and cooling on the GrIS. Clouds warm the surface by about $21 W m^{-2}$ in the accumula-
 207 tion zone, and cool the surface by about $12 W m^{-2}$ in the lower ablation zone. During the 2008–
 208 2013 melt seasons, the isopleth of 0.57 albedo separates areas of positive and negative CREs.
 209 In the ablation zone, the intensity of the cloud cooling effect is highly positively correlated
 210 with albedo: lower albedos associated with warmer environments strengthen cloud-induced sur-
 211 face cooling. The effect of albedo on CRE is suppressed in the accumulation zone where the
 212 surface albedo is close to cloud albedo. The seasonal and inter-annual variabilities of CREs
 213 in the ablation zone are also highly correlated with albedo: the cooling effect peaks in the low-
 214 est albedo months (July and August); the low albedo in 2012 caused by massive surface melt
 215 drove net CREs toward cooling. This cooling effect of clouds on the dimming surface acts as
 216 a stabilizing mechanism for Greenland’s surface melt, and may help to reduce future surface
 217 melt.

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