

1 Multi-modal albedo distributions in the ablation area of the 2 southwestern Greenland Ice Sheet

3
4 **S.E. Moustafa¹, A.K. Rennermalm¹, L.C. Smith², M.A. Miller³, J.R.
5 Mioduszewski¹, L.S. Koenig^{4,5,*}, M.G. Hom^{6,7}, and C.A. Shuman^{4,8}**

6 [1]{Department of Geography, Rutgers, The State University of New Jersey, 54 Joyce Kilmer
7 Avenue, Piscataway, NJ 08854-8045, USA }

8 [2]{Department of Geography, University of California, Los Angeles, 1255 Bunche Hall, P.
9 O. Box 951524, Los Angeles, CA 90095-1524, USA }

10 [3]{Department of Environmental Sciences, Rutgers, The State University of New Jersey, 14
11 College Farm Rd, New Brunswick, NJ 08901-8551, USA }

12 [4]{Cryospheric Sciences Laboratory, NASA Goddard Space Flight Center, 8800 Greenbelt
13 Road, Greenbelt, MD, 20771, USA }

14 [5]{Cooperative Institute for Research in Environmental Sciences, University of Colorado
15 Boulder, 216 UCB, Boulder, CO, 80309, USA }

16 [6]{Biospheric Sciences Laboratory, NASA Goddard Space Flight Center, 8800 Greenbelt
17 Road, Greenbelt, MD, 20771, USA }

18 [7]{Science Systems and Applications, Inc., 10210 Greenbelt Rd, Lanham, MD, 20706,
19 USA }

20 [8]{Joint Center for Earth Systems Technology (JCET), University of Maryland, Baltimore
21 County, 1000 Hilltop Circle, Baltimore, MD, 21250, USA }

22
23 Correspondence to: S.E. Moustafa (samiah.moustafa@rutgers.edu)

24 *L.S. Koenig is now at National Snow and Ice Data Center, University of Colorado, 1540 30th
25 Ave., Boulder CO 80303, USA

1 **Abstract**

2 Surface albedo is a key variable controlling solar radiation absorbed at the Greenland Ice
3 Sheet (GrIS) surface, and thus, meltwater production. Recent decline in surface albedo over
4 the GrIS has been linked to enhanced snow grain metamorphic rates, earlier snowmelt, and
5 amplified ice-albedo feedback from atmospheric warming. However, the importance of
6 distinct surface types on ablation area albedo and meltwater production is still relatively
7 unknown. In this study, we analyze albedo and ablation rates using in situ and remotely-
8 sensed data. Observations include: 1) a new high-quality in situ spectral albedo dataset
9 collected with an Analytical Spectral Devices (ASD) spectroradiometer measuring at 325–
10 1075 nm, along a 1.25 km transect during three days in June 2013; 2) broadband albedo at
11 two automatic weather stations; and 3) daily MODerate Resolution Imaging
12 Spectroradiometer (MODIS) albedo (MOD10A1) between 31 May – 30 August 2012 and
13 2013. We find that seasonal ablation area albedos in 2013 have a bimodal distribution, with
14 snow and ice facies characterizing the two peaks. Our results show that a shift from a
15 distribution dominated by high to low albedos corresponds to an observed melt rate percent
16 difference increase of 51.5% (between 10 – 14 July and 20 – 24 July, 2013). In contrast, melt
17 rate variability caused by albedo changes from pentad-to-pentad before and after this shift was
18 much lower, and varied between ~10-30% in the melting season. In 2012, a more complex
19 multimodal distribution emerges, reflecting a transition from light to dark-dominated surface,
20 as well as sensitivity to the so called ‘dark band’ region in southwest Greenland. In addition
21 to a darkening surface from ice crystal growth, our findings demonstrate that seasonal
22 changes in GrIS ablation area albedo are controlled by changes in the fractional coverage of
23 snow, bare ice, and impurity-rich surface types. Thus, seasonal variability in ablation area
24 albedo appears to be regulated primarily as a function of bare ice expansion at the expense of
25 snow, surface meltwater ponding, and melting of outcropped ice layers enriched with mineral
26 materials, enabling dust and impurities to accumulate. As climate change continues in the
27 Arctic region, understanding the seasonal evolution of ice sheet surface types in Greenland’s
28 ablation area is critical to improve projections of mass loss contributions to sea level rise.

29

30 **1 Introduction**

31 Surface albedo, defined as the bihemispherical reflectance integrated across the visible and
32 near-infrared wavelengths (Schaepman-Strub et al., 2006), is a key variable controlling

1 Greenland Ice Sheet (GrIS) surface melting. During the melt season, surface albedo
2 modulates absorbed solar radiation at the ice surface, and consequently, the surface energy
3 and mass balance of the ice sheet (Cuffey and Paterson, 2010). Over the last decade, an
4 observed decline in albedo has been linked to less summer snow cover, expansion of bare ice
5 area, and enhanced snow grain metamorphic rates from atmospheric warming, amplified by
6 the melt-albedo feedback (Box et al., 2012; Stroeve et al., 2013; Tedesco et al., 2011). This
7 positive feedback involves increased melting, and exposure of bare ice, impurities and
8 meltwater ponding, reducing surface albedo, thereby increasing solar radiation absorption,
9 and thus, accelerating melt further (Box et al., 2012; Tedesco et al., 2011).

10 The GrIS surface has a wide range of surface types with different albedos, including
11 snow, ice, dust and sediment-rich impurities, cryoconite holes, melt ponds, and streams. Yet,
12 the importance of these surface types on ablation area albedo, and thus, meltwater production
13 over the melt season is still relatively unresolved (Rennermalm et al., 2013). Current state-of-
14 the-art surface mass balance (SMB) models, such as Modèle Atmosphérique Régionale
15 (MAR) v3.2 and Regional Atmospheric Climate Model (RACMO2), consider some
16 variability in surface types by including the presence of meltwater ponding, snow, black
17 carbon concentrations on snow, and bare ice surfaces to characterize seasonal variations in
18 ablation area albedo (Alexander et al., 2014; Van Angelen et al., 2012). Furthermore,
19 RACMO2 is capable of utilizing realistic MODIS background albedo data (Van Angelen et
20 al., 2012), thereby representing the impact of surface types spatially aggregated to the MODIS
21 resolution. However, few studies have utilized these modeling tools to understand how the
22 distributions of surface types are changing ablation area albedo (e.g., Alexander et al. 2014).
23 This is increasingly important due to enhanced surface melt associated with anomalously
24 warm atmospheric circulation patterns in 2007-2012 that may become more frequent in the
25 future (Hall et al., 2013; Nghiem et al., 2012; Tedesco et al., 2013). Additionally, some
26 studies suggest that a new control of ice sheet albedo is the deposition and accumulation of
27 light-absorbing impurities advected from snow-free areas and forest fires outside of
28 Greenland (Dumont et al., 2014; Keegan et al., 2014).

29 The large-scale decline in albedo has been greatest in southwest Greenland (-0.04 to -
30 0.16 per decade trend in June and August, respectively; Stroeve et al., 2013). This is related to
31 stronger warming trends (2-4 °C in some regions; Hanna et al., 2014), early melt onset, a lack
32 of wintertime accumulation (van den Broeke et al., 2008), expansion of bare ice area (Tedesco

1 et al., 2011), high concentration of impurities (cryoconite, dust, and soot), melting of
2 outcropped ice layers enriched with mineral content (Wientjes and Oerlemans, 2010; Wientjes
3 et al., 2011), and enhanced meltwater production and runoff (e.g., Mernild et al., 2012).
4 Seasonal changes in the distribution of different surface types in southwest Greenland's
5 ablation area have considerable influence on the spatiotemporal variability of surface albedo
6 (Chandler et al., 2014; Knap and Oerlemans, 1996; Konzelmann and Braithwaite, 1995).
7 During the melt season, surface albedo decreases as cryoconite hole coverage increases
8 (Chandler et al., 2014), melt ponds and supraglacial rivers form efficient drainage networks
9 (Lampkin and VanderBerg, 2013; Kang and Smith, 2013; Smith et al., 2015), crevasses and
10 other types of roughness begin to form, and impurities accumulate from exposure of the
11 underlying ice surface (Wientjes and Oerlemans, 2010). Albedo in western Greenland's
12 ablation area averages around ~0.41 for the duration of the melt season (Wientjes et al.,
13 2011), but can vary between > 0.80 for fresh snow, to 0.30-0.60 for bare ice (Cuffey and
14 Patterson, 2010), and ~0.10 for cryoconite surfaces (Bøggild et al., 2010; Chandler et al.,
15 2014; Knap and Oerlemans, 1996). Furthermore, negative albedo trends since 2000 (Box et
16 al., 2012) are linked to an expansion of areas of ablation relative to accumulation facies.

17 Changes in surface albedo are typically characterized from the MODerate Resolution
18 Imaging Spectroradiometer (MODIS) and the Advanced Very High Resolution Radiometer
19 (AVHRR) satellite sensors (e.g., Chandler et al., 2014; Stroeve et al., 2013; Wang et al., 2012;
20 Wright et al., 2014) or modeled with regional climate models (RCMs) such as Regional
21 Atmospheric Climate Model (RACMO2; Van Meijgaard et al., 2008) and Modèle
22 Atmosphérique Régional (MAR; Fettweis, 2007). Remotely-sensed and modeled albedo has
23 been validated with ground measurements from dispersed Greenland Climate Network
24 Automatic Weather Stations (GC-Net AWS; Knap and Oerlemans, 1996; Steffen and Box,
25 2001). These comparisons reveal that satellite products provide reasonable albedo estimates
26 (Box et al., 2012; Stroeve et al., 2005, 2006, 2013), although discrepancies between different
27 MODIS albedo products have been identified (Alexander et al., 2014). Despite this, RCM
28 surface albedos remain represented in relatively simplistic terms, particularly in regions that
29 frequently experience prolonged bare ice exposure like southwest Greenland (Fettweis et al,
30 2011; Fitzgerald et al., 2012; Rae et al., 2012; Van Angelen et al., 2012). This is attributed to
31 a lack of surface roughness in the RCMs (Ettema et al., 2010), and relatively simplistic bare
32 ice and impurity albedo schemes (Alexander et al., 2014), resulting in large inter-model
33 differences in runoff (42% variance; Vernon et al., 2013), despite the existence of spatially

1 distributed ice albedo schemes and inclusion of black carbon contaminants on snow surfaces
2 (Van Angelen et al., 2012). Recent surface albedo observations and snow model simulations
3 of impurity-rich surfaces have been linked to enhanced ice sheet melt (Chandler et al., 2014;
4 Dumont et al., 2014; Keegan et al., 2014), suggesting that incorporating seasonal changes in
5 the albedo distribution of distinct surface types might improve accuracy of modeled meltwater
6 runoff and GrIS sea level rise contributions. These findings point to the importance of a
7 detailed assessment of high spectral, spatial, and temporal resolution albedo data to quantify
8 how different surface types control ablation area albedo, and therefore, melt.

9 In this study, we report the results of an assessment of ablation area albedo along the
10 southwestern GrIS for the 2012 and 2013 melt seasons. We use 1) a new high-quality in situ
11 spectral albedo dataset collected with an Analytical Spectral Devices (ASD)
12 spectroradiometer measuring at 325–1075 nm, along a 1.25 km transect during three days in
13 June 2013; 2) in situ broadband albedo at two automatic weather stations; and 3) daily
14 MODerate Resolution Imaging Spectroradiometer (MODIS) albedo (MOD10A1) product
15 (Hall et al., 2012) between 31 May – 30 August 2012 and 2013 to investigate how ice sheet
16 surface types influence surface albedo and ablation rates; and 4) summer seasonal changes in
17 surface type coverage reported in literature. First, we describe the collection of high-quality in
18 situ spectral albedo ground, automatic weather station broadband albedo, and ablation stake
19 measurements collected during early 2013 melt season along a fixed transect in the GrIS
20 ablation area. From the MODIS daily albedo data, we estimate seasonal changes in the albedo
21 distributions by using fractional area of surface types from a nearby site (1030 m a.s.l.;
22 reported by Chandler et al., 2014). These distributions were compared with seasonal changes
23 in computed albedo distributions derived from in situ albedo and ablation stake data. Thirdly,
24 the impact of changing albedo and surface type coverage on surface melt was quantified and
25 compared with transect ablation stake measurements. Finally, we compare these results with
26 2012 MOD10A1 data to better understand the overall frequency distribution, spatiotemporal
27 variability, and ablation rates associated with dominant surface types in southwest
28 Greenland’s ablation area. This study is the first high spatial, temporal, and spectral resolution
29 albedo dataset collected in the southwestern GrIS ablation area.

30

1 **2 Study site description**

2 The study site is located on the southwestern GrIS approximately 30 km northeast of
3 Kangerlussuaq, Greenland (Fig. 1). Albedo measurements were collected along a 1.25 km
4 transect situated between ~510 to 590 m a.s.l., well within the ablation area for this region
5 (mean equilibrium line altitude of 1553 m a.s.l.; van de Wal et al., 2012). Two meteorological
6 stations, referred to as Base and Top Met Stations, were installed near the transect end points
7 by Site E and A, respectively (Fig. 1) to derive independent measurements of in situ
8 broadband albedo (300-1100 nm), hereafter α_{base} and α_{top} . In addition, ablation stakes were
9 installed at five sites along the albedo transect and by the Base Met Station to measure ice
10 surface ablation rates. Ice sheet surface types examined included white ice, shallow
11 supraglacial streams, and dirty ice, where dirty ice was qualitatively distinguished from white
12 ice based on visible surface sediments. Visual assessment in the study area revealed that snow
13 had melted before mid-June and no snowfall events occurred between 8-26 June 2013. A few
14 small melt ponds ($< 1 \text{ km}^2$) were observed in the study area, but likely not in sufficient
15 quantity to explain discrepancies between in situ and MODIS albedo-derived estimates.

16

17 **3 Methods**

18 **3.1 Field spectroscopy measurements**

19 High spatial (~10 m posting), temporal (1-2 days), and spectral (1 nm) resolution spectral
20 albedo measurements, hereafter α_{ASD} , were measured at 325-1075 nm using an ASD
21 Fieldspec HandHeld 2 Spectroradiometer (PANalytical, formerly ASD Inc.), fitted with a
22 Remote Cosine Receptor (RCR) foreoptic. The ASD was mounted on a tripod at 0.4 m
23 distance, and with no foreoptic attached (i.e., bare fiber), had a 25° field-of-view,
24 corresponding to a spot size of ~0.18 m diameter on the surface.

25 Spectral albedos were measured along the transect starting at Site E and ending at Site
26 A on 16, 17, 19, 21, 24, and 25 June, 2013 between 1000 and 1800 local time (1200 – 2000
27 GMT). After rigorous quality control (see Appendix A), only transect observations made on
28 the 16, 19, and 25 June were used in analyses. Broadband α_{ASD} were calculated by averaging
29 albedo over its entire spectral range at each site along the transect. These measurements were
30 compared with MOD10A1 and meteorological station data, as described in section 3.3.

1

2 **3.2 Continuous broadband albedo measurements at meteorological stations**

3 Two AWSs, the Top and Base Met Stations, were installed at each end of the transect to
4 independently measure broadband (300-1100 nm) albedo from 8-26 June 2013, hereafter α_{base}
5 and α_{top} . See Appendix A for details on surface installation conditions and tilt uncertainty
6 estimates. This analysis suggests that α_{top} were compromised by surface roughness effects,
7 and thus, α_{base} alone is used for most analyses.

8 Daily average broadband albedo was computed using shortwave flux measured at
9 SZAs $< 70^\circ$ (Stroeve et al., 2005) to minimize the cosine response error inherent to the
10 pyranometers (uncertainty increases by $\pm 5\%$ for SZAs $> 70^\circ$; Onset Computer Corp., 2010).
11 Expected accuracy of α_{base} and α_{top} is $\pm 10\%$ based on the intrinsic accuracy and cosine
12 response error of the pyranometers. Additional sources of error not quantified here include
13 tower shadowing and surface roughness effects on measured surface albedo (e.g., Lhermitte
14 et al., 2014).

15

16 **3.3 MODIS albedo data**

17 Daily MODIS broadband albedo (300-3000 nm) was acquired from the MOD10A1 product
18 (Version 005) from NASA's Terra satellite (Hall et al., 2006; Klein and Stroeve, 2002). High-
19 quality flagged MOD10A1 albedo data (periods of high SZA and cloudiness were excluded;
20 Schaaf et al., 2011) from 31 May to 30 August 2012 and 2013 (when SZAs are minimized;
21 e.g., Box et al., 2012) were used in two analyses. First, MOD10A1 albedo for pixels
22 overlapping with our transect site (Fig. 1), hereafter $\alpha_{\text{MOD Pixel 1}}$ and $\alpha_{\text{MOD Pixel 2}}$, were
23 compared with observations as described below. Second, distributions of MOD10A1 albedo
24 were examined at three spatial extents as described in section 3.5.

25 Broadband $\alpha_{\text{MOD Pixel 1}}$ and $\alpha_{\text{MOD Pixel 2}}$ were compared with α_{ASD} and α_{base} . Direct
26 comparison of α_{ASD} , α_{base} , and α_{MOD} absolute values are not possible due to different
27 wavelength ranges, and α_{MOD} is expected to have lower values than the other two datasets.
28 However, relative comparisons of spatial and temporal patterns are reasonable, because the
29 α_{MOD} is dominated by the ASD visible and near-infrared (i.e., 325-1075 nm) wavelengths. In

1 a standard Top-of-Atmosphere solar irradiance reference spectrum, the 325-1075 nm range
2 comprises 80.52% of the total irradiance in the 300-3000 nm range. The dominance of
3 reflectance in the ASD visible and near-infrared wavelengths in determining broadband
4 albedo means that α_{MOD} can be used qualitatively to provide spatiotemporal context. High-
5 quality broadband (325-1075 nm) α_{ASD} data within pixels 1 and 2, hereafter $\alpha_{\text{ASD Pixel 1}}$ and
6 $\alpha_{\text{ASD Pixel 2}}$, were averaged together to indirectly validate $\alpha_{\text{MOD Pixel 1}}$ and $\alpha_{\text{MOD Pixel 2}}$ data, and to
7 facilitate comparison between in situ and remotely-sensed observations. While absolute
8 values will differ between the datasets, and issues of MODIS pixel separability may exist due
9 to off-nadir footprint effects (Dozier et al., 2008), the difference shouldn't change spatial and
10 temporal patterns.

11

12 **3.4 Ablation and albedo at dominant surface types**

13 Surface melting between 8 – 26 June was estimated using ablation stakes installed at the Base
14 Met Station, hereafter M_{base} , and at five sites across the albedo transect, hereafter M_{stakeXY} ,
15 where X denotes Sites A-E, and Y denotes surface type - white ice (W), dirty ice (D), or
16 shallow 5-10 cm deep streams (S) (Fig. 1). Bamboo poles were used as stakes (Hubbard and
17 Glasser, 2005), and ablation rates were recorded every 1-3 days by measuring the distance
18 between the bamboo pole top and ice sheet surface at cm-scale resolution.

19 α_{ASD} spectra were made within 30 m of ablation stakes to identify representative surface
20 type albedos. With the exception of Site D, all sites were relatively homogenous. At Site D,
21 the two surface types could be classified into distinct groupings: clean and dirty ice. Albedos
22 of clean ice at Sites A-C, and E, hereafter $\alpha_{\text{ASD_AW}}$, $\alpha_{\text{ASD_BW}}$, $\alpha_{\text{ASD_CW}}$, and $\alpha_{\text{ASD_EW}}$, were
23 estimated by averaging broadband α_{ASD} observations made within 30 m of stakes for each
24 transect date. At Site D, albedos of clean and dirty ice, hereafter $\alpha_{\text{ASD_DW}}$ and $\alpha_{\text{ASD_DD}}$, were
25 estimated from the histograms of α_{ASD} observations made within 30 m of stakes for each
26 transect date. At the M_{base} stake, no albedo observations were made. Instead, $\alpha_{\text{ASD_DD}}$ is
27 assumed to be representative of albedo at the Base Met Station, hereafter $\alpha_{\text{MET_base}}$. Stream
28 albedo, hereafter α_{stream} , was determined from occasional α_{ASD} measurements at various
29 shallow surface streams between 13-25 June. Cryoconite hole albedo, hereafter α_{cryo} , was
30 parameterized using published values (from Bøggild et al., 2010) of broadband albedo

1 averaged together for damp cryoconite material and cryoconite basin surface types under
2 clear-sky and overcast conditions.

3

4 **3.5 Melt season albedo distributions**

5 Two types of melt season albedo distributions were constructed: 1) computed distributions
6 based on broadband α_{ASD} for distinct surfaces and fractional surface coverage area from
7 Chandler et al. (2014); and 2) observed MODIS-derived distributions.

8 The computed distributions were constructed by assuming that the albedo distribution
9 for each distinct surface is represented by a normal distribution $N(\bar{x},s)$, with $\bar{x} = \overline{\alpha_{ASD}}$
10 representing surface type and standard deviation, s , different for each surface type. Four
11 distributions were constructed: clean ice $N(0.56, 0.07)$, dirty ice $N(0.19, 0.05)$, shallow
12 streams $N(0.23, 0.09)$, and cryoconite holes $N(0.10, 0.05)$. Relative surface coverage of these
13 four dominant surface types was derived at five distinct time periods (1 June, 19 June, 18
14 July, 28 July, and 5 August) over the 2012 melt season from Chandler et al. (2014; see Fig.
15 6a-g) to represent transient ice surface conditions, classified here as “early summer ice”,
16 “dirty ice exposure”, “melt”, “darkening ice”, and “late summer ice”, respectively (Table 3).
17 A composite distribution for each distinct time step was calculated as the weighted mean of
18 surface type distributions, where the weights were determined by their relative surface
19 coverage area. Since Chandler et al. (2014) data are from 2012, results were not directly
20 comparable with 2013 MOD10A1 data, but should capture melt season evolution.

21 To compare with the computed distributions, high-quality 2012 and 2013 MOD10A1
22 data were used to construct observed albedo distributions at three spatial extents (50x50,
23 100x100, and 150x150 pixel extents; Fig. 1). The spatial resolution of the original MOD10A1
24 data is 463 m at nadir (exact resolution varies with overpass time), corresponding to study
25 areas of 23.2, 46.3, and 69.5 km² for the three spatial extents, respectively. Using a kernel
26 smoothing density estimator, the average probability density distribution was computed at
27 0.01 albedo bin widths (range from 0.05 to 1). The seasonal average albedo distribution was
28 calculated at the three spatial extents, and five-day average albedo distributions and spatial
29 averages were calculated for the 100x100 pixel scale for 2012 and 2013 MOD10A1 data.

30

1 3.6 Identification of snowfall events

2 To identify possible snowfall events in our study area and MODIS spatial extents, hourly
3 precipitation and air temperature measurements collected by a meteorological station,
4 hereafter 660 Met Station, installed near the ice sheet edge at the proglacial and ice sheet
5 margin interface (Fig. 1), was examined. Near surface air temperature measurements from the
6 shorter Base Met Stations time series (available from 8 – 26 June 2013) were also examined
7 to estimate temperature differences between the proglacial and ice surfaces. To validate that
8 solid precipitation fell, NASA’s WorldViewer was utilized to browse daily MODIS
9 reflectance imagery (bands 7-2-1 and 3-6-7) to identify textural and brightness changes
10 related to precipitation events.

11

12 3.7 Computation of relative melt rates

13 To examine seasonal changes in MODIS albedo, and estimate the importance of distinct
14 surface types, relative surface melt rates were computed using the net shortwave solar
15 radiation equation, observed values of incoming solar radiation from the Base Met Station on
16 16, 19, and 25 June, and broadband albedo values for computed and observed distribution
17 methods. Net solar radiation (E_R) varies as a function of incoming solar radiation (E_S^\downarrow) and
18 albedo (α_s), where units of energy are represented as $W\ m^{-2}$:
19 $E_R = E_S^\downarrow(1 - \alpha_s)$ (1)

20 Melt rate, defined as the heat needed to melt snow/ice when near-surface temperatures are ≥ 0
21 $^\circ C$, was computed in units of $m\ s^{-1}$ (Cuffey and Patterson, 2010):

$$22\ M = (E_R * \Delta t)(L_f * \rho_w)^{-1} \quad (2)$$

23 where Δt is the time interval (s); L_f is latent heat of fusion ($3.34 \times 10^5\ J\ kg^{-1}$); and ρ_w is
24 density of water ($1000\ kg\ m^{-3}$). Since the meteorological station datasets lack surface energy
25 balance terms (i.e., net longwave radiation, sensible and latent heat fluxes) required to
26 compute the entire energy budget, calculating absolute melt rates was not possible. Instead,
27 the percent difference in estimated melt rates was computed for each distribution relative to
28 the early melt season ablation rates (mean of $4.40 \times 10^{-7}\ m\ s^{-1}$ for “early summer ice” computed
29 distribution; mean of $2.70 \times 10^{-7}\ m\ s^{-1}$ for 31 May – 4 June observed MODIS distribution).

30

1 4 Results

2 4.1 Spatiotemporal patterns in ablation area albedo

3 Spatial variability of broadband α_{ASD} along the transect follows a consistent pattern on all
4 three dates, averaging low values (0.50 ± 0.04) the first ~ 300 m, followed by increased
5 albedo, reaching a plateau of 0.64 ± 0.07 at ~ 600 m, and remaining nearly constant with the
6 exception of a dip to 0.44 ± 0.02 at ~ 900 m (Fig. 2a). While discrete α_{ASD} observations often
7 differ from the nearest observation made at another transect time due to slight day-to-day
8 changes in sample location (Fig. 2a), data averaged in 50 m bins covary spatially along the
9 transect gradient (Fig. 2b). The spatial variability of broadband α_{ASD} is considerable and
10 varies between a minimum of 0.14 (19 June) and a maximum of 0.75 (16 June; Table 1). The
11 high spatial variability in α_{ASD} over short distances is indicative of the heterogeneous surface
12 that characterizes the field site and surrounding ablation area, not necessarily captured in α_{base}
13 observations.

14 Temporal variability in daily average α_{base} follows a non-linear decline from 8-26 June
15 2013 starting at 0.49 and ending at 0.34 (Fig. 3). An increase in α_{base} of 0.11 between June 12
16 and 16 might be related to tilt errors, which influenced what part of the increasingly
17 heterogeneous surface the instruments were monitoring. Indeed, the net lowering of α_{base} by
18 0.15 between 8-26 June is confirmed and observed from June to mid-August for $\alpha_{MOD \text{ Pixel } 1}$
19 and $\alpha_{MOD \text{ Pixel } 2}$. $\alpha_{MOD \text{ Pixel } 1}$ and $\alpha_{MOD \text{ Pixel } 2}$ drop from values slightly above 0.5 in June to 0.24
20 and 0.37, respectively, around mid-August. In between these dates, sudden increases in
21 albedo could be caused by occasional snowfall events, where the difference in tundra and ice
22 sheet near surface air temperatures is ~ 3 °C (Fig. 4). A brief snowfall event on 28-29 June
23 (Fig. 4) raised MOD10A1 albedos from 0.31 to 0.53 between 27 June and 30 June,
24 respectively. July MOD10A1 albedos exhibited some temporal variability, but were generally
25 lower at the end than the start of the month. It is unclear if they were triggered by snowfall
26 events. While precipitation events occurred several times on the tundra in July, it is unknown
27 if these events extended to the ice sheet and if temperatures were sufficiently cold to trigger
28 snow rather than rain (Fig. 4). August MOD10A1 albedo increased from early to late in the
29 month with a snowfall event on ~ 18 August, triggering large increases in albedo to values
30 above 0.75. High-quality daily average broadband $\alpha_{ASD \text{ Pixel } 1}$ and $\alpha_{ASD \text{ Pixel } 2}$ data don't exhibit
31 the slight increase in α_{base} at the end of June (0.04 from 22-26 June), which may be reflected

1 by differences in footprint sizes, less temporal α_{ASD} sampling frequency, and α_{base} tilt errors.
2 Instead, $\alpha_{ASD \text{ Pixel } 1}$ and $\alpha_{ASD \text{ Pixel } 2}$ data exhibit a steady decline over the month of June, while
3 $\alpha_{MOD \text{ Pixel } 1}$ and $\alpha_{MOD \text{ Pixel } 2}$ data remain relatively constant over the same time period. Absolute
4 magnitudes among the three ground- and satellite-derived albedo products diverge due to
5 sensor, wavelength range and spatial resolution differences. However, all products have
6 higher albedo values in the first than the last observation in the month of June, prior to the 28-
7 29 June snowfall event.

8 Albedos of dirty and clean ice surfaces are distinctly different for each ablation stake
9 site (Table 2). Broadband α_{ASD} spectra made within 30 m of ablation stakes were individually
10 assessed to classify each surface type into two distinct groupings: clean and dirty ice (Fig. 5).
11 Only Site D had both dirty and clean ice surfaces. Manual inspection of individual spectra at
12 Site D confirm that samples with $\alpha_{ASD} < 0.4$ are qualitatively similar to typical spectra for wet
13 or debris rich ice as shown in Pope and Reese (2014), and distinctly different from α_{ASD}
14 above 0.4 (Fig. 5).

15

16 **4.2 Melt season albedo distributions**

17 4.2.1 2013 computed vs. observed distributions

18 Computed albedo frequencies using typical albedo values for four distinct surface types
19 (Table 3) and changing area fractions of these surfaces identified at a nearby site by Chandler
20 et al. (2014) reveal a bimodal distribution as the melt season progresses (Fig. 6). The relative
21 strength of the first and secondary modes change as the fractional area of darker surfaces
22 expands from “dirty ice exposure” to “melt” distributions and onwards. At the start of the
23 melt season, the abundance of lighter surfaces coincides with a higher probability of high
24 broadband α_{ASD} values. Here, snow and clean ice surfaces dominate and gradually degrade,
25 exposing the impurity-rich surface underneath. As darker surfaces progressively populate the
26 ablation area with the onset of the melt season, computed albedo distributions predict a
27 concomitant higher probability of lower albedo. Thus, there is an apparent dichotomy
28 between darker and lighter surfaces ‘competing’ to control the overall albedo distribution of
29 the ablation area. A transition towards a distribution biased towards lower albedo values is
30 due to darker surfaces shifting the overall distribution, and is confirmed by high-quality

1 broadband α_{ASD} distributions (Fig. 7). Relative melt rates increase sharply (by 25.7%) from
2 “dirty ice exposure” to “melt”, coinciding with a strengthening of the second, lower mode in
3 the computed albedo distribution (Fig. 8). Once the secondary mode is established, a smaller
4 increase in melt rates occurred as the mode strengthens between “melt” to “darkening ice”
5 and finally to “late summer ice” (6.7% and 9.1%, respectively).

6 Observed MOD10A1 albedo distributions at three spatial extents (Fig. 9) reveal that
7 the bimodal distributions (cf. Fig. 6) are manifested in reality at the 100x100 MODIS pixel
8 (px) extent (i.e. 46.3 km²). While the spatial extent of the MOD10A1 sample influences the
9 seasonal average albedo distribution, two distinct surface types - dark and light surfaces -
10 dominate the seasonal signal (Fig. 9). At the smallest spatial extent (50x50 px – i.e., 23.2
11 km²), lower albedos from darker surfaces of the lower ablation area control the density
12 distribution, while at the largest spatial extent (150x150 px – i.e., 69.5 km²), the probability
13 distribution is primarily influenced by higher albedos from lighter surfaces (e.g., snow) of the
14 upper ablation area. The central tendencies of each mode are ~0.46 and ~0.72, which is much
15 larger than in the computed distributions (~0.18 and ~0.56; cf. Fig. 6).

16 The bimodal distribution identified in the observed 100x100 px MODIS albedo
17 distribution in 2013 (Fig. 9) is the result of snow and ice surfaces characterizing the two
18 peaks, as each mode centers around typical values of snow and clean ice, respectively. As
19 such, the observed MODIS bimodal distribution is associated with a transition from ice to
20 snow, rather than a change from clean to dirty ice, which caused the two modes in the
21 computed distribution (Fig. 6). Indeed, analysis of 2013 meteorological observations reveal
22 that short term snowfall events that fell on top of the underlying ice can result in variations in
23 ablation area albedo (Fig. 3 and 10). In 2013, the bimodal distribution at the 100x100 px
24 spatial extent is likely the result of snow deposition or redistribution of blowing snow on top
25 of the ice surface (Fig. 9 and 10).

26 MOD10A1 albedo at the 100x100 px (i.e., 46.3 km²) spatial extent transitions from a
27 unimodal distribution with high albedo values at the start of the melt season (31 May – 4
28 June), to a bimodal-like distribution with intermediate albedo values at mid-melt season (20 –
29 24 June), and shifts abruptly to a new, unimodal distribution with low albedo values at peak
30 melt season (30 July – 3 August; Fig. 10). By assuming an unchanged radiation budget, the
31 relative impact of albedo distribution changes on melt rates was quantified. The abrupt shift
32 from a lighter- (high albedo) to darker-dominated (low albedo) surface corresponds to an

1 observed melt rate percent difference increase of 51.5% between 10 – 14 July and 20 – 24
2 July pentad average albedo distributions (Fig. 14). Before and after this shift, melt rates
3 changed much less from pentad-to-pentad, ranging between ~10-30%, with the exception of
4 the dramatic drop of 103.3% when the melt season ends in late August.

5 The bimodality seen in the 30 June – 4 July pentad (Fig. 10) coincides to a brief period
6 of higher MODIS albedo values (~0.6 – 0.7), indicative of snow. Identification of a snowfall
7 event on 28-29 June 2013 (Fig. 4), confirms the source of the bimodal distribution observed
8 in the 30 June – 4 July pentad (Fig. 10), corresponding to a brief “jump” in the probability
9 density distribution to higher albedos.

10

11 4.2.2 Differences between 2012 vs. 2013 observed albedo distributions

12 While the 2013 MODIS albedo bimodal distribution shown in Fig. 9 and 10 are a result of
13 snow and ice albedo, analysis of MODIS 2012 data reveal a more complex, multi-modal
14 albedo distribution (Fig 15). These distributions cannot be explained by the presence or
15 absence of snow and ice alone. The 2012 MODIS is characterized by generally lower albedo,
16 with six out of nine pentad albedo distributions ranging mostly between 0.2 and 0.5,
17 compared to three out of nine pentad albedo distributions in 2013 (cf. Fig. 10 and 11). These
18 low albedos are confirmed by the average seasonal MODIS 2012 albedo distributions, where
19 a higher probability of albedos are centered on ~0.35, compared to two peaks at ~0.45 and
20 ~0.7, in 2013 at the 100x100 px spatial extent (cf. Fig 12 and 16). The higher probability of
21 these very low albedos observed in 2012 are likely due to dust, sediment, and impurity-rich
22 ice in the so-called ‘dark-band’ region (Wientjes and Oerlemans, 2010). The identification
23 of this dark zone feature is presented in section 4.2.3.

24

25 4.2.3 2012 vs. 2013 spatial maps

26 The presence of the dark band region is confirmed by the diagonal band of very low albedo (<
27 ~0.35) in the 2012 MODIS seasonal average at the 100x100 px extent (Fig. 12). However, the
28 presence of the dark band region is not visible in 2013, where albedo gradually increases from
29 west to east (Fig. 13). The lack of the dark zone feature in 2013 is likely due to snow covering
30 the dark band for most of the season. Overall, 2012 exhibits substantially lower ablation area

1 albedos (Fig. 11), while 2013 reveals higher ablation area albedos in the MODIS spatial
2 averages (Fig. 10). The large inter- and intra-annual variability in MODIS ablation area
3 albedo may be indicative of the large spatial variability in surface types that characterize the
4 lower elevations of the ablation area. Alternatively, a larger distribution in cryoconite hole
5 coverage may have also contributed to low albedos (~ 0.25) observed in the 2012 MODIS
6 seasonal averages (Fig. 15).

7

8 **4.3 Relative melt rates**

9 Observed ablation rates, derived from stake readings, are typically higher for dark surfaces
10 (dirty ice and streams) than light surfaces (clean ice; Fig. 16). Clean ice surfaces have higher
11 broadband α_{ASD} values (mean of 0.57), corresponding to lower average ablation rates
12 ($5.38 \times 10^{-7} \text{ m s}^{-1}$). In contrast, dirty ice and stream surfaces have lower mean broadband α_{ASD}
13 values (0.24), corresponding to higher average ablation rates ($6.75 \times 10^{-7} \text{ m s}^{-1}$). The observed
14 mean difference between light and dark surface ablation rates is $1.37 \times 10^{-7} \text{ m s}^{-1}$. Melt rate
15 calculations (Eqn. 1 and 2) resulted in a lower average ablation rate for clean ice surfaces
16 ($4.24 \times 10^{-7} \text{ m s}^{-1}$) and a higher average ablation rate for dark ice surfaces ($7.56 \times 10^{-7} \text{ m s}^{-1}$),
17 corresponding to a mean difference of $3.33 \times 10^{-7} \text{ m s}^{-1}$. Differences between observed and
18 calculated melt rates could be due ablation stake measurement errors and simplification of
19 calculations (e.g., no consideration of longwave radiation or turbulent heat fluxes).
20 Regardless, in both cases relative melt rates between light and dark surfaces are considerably
21 different, and thus useful for investigating seasonal melt rate changes as described next.

22 The spread in observed clean ice broadband albedo values results in greater variability
23 in observed ablation rate estimates (Fig. 16). In contrast, minimal broadband albedo
24 variability is observed for dirty ice and stream surfaces. As such, grouping these two ice
25 surface types into a 'darker surface' type classification is justified. Few dirty ice albedo
26 measurements were sampled as compared to clean ice surfaces. Differences in ablation rates
27 for stream surfaces are due to a lack of albedo data. While ablation rates were measured at
28 several ablation stake stream sites, only occasional α_{ASD} measurements were collected over
29 these surfaces. Considerable spread in ablation rates for stream observations could be
30 explained by varying stream depth (Legleiter et al., 2014). The depth of these ice streams
31 determines the attenuation and scattering of radiant energy, thereby influencing the observed

1 albedo measurements. Sensible heat flux from the stream water, not accounted for in radiative
2 estimates, may also be a mechanism for increased melting.

3

4 **5 Discussion**

5 **5.1 The importance of surface types on observed and computed ablation area** 6 **albedo**

7 GrIS ablation area albedos are strongly influenced by the presence or absence of impurity-rich
8 debris on its surface. Clean ice and dust-covered, dirty ice have distinctly different albedos,
9 resulting in a left-skewed albedo distribution at mid- and end of June (Fig. 7). This pattern is
10 supported by computed and remotely-sensed albedo distributions, revealing that a multimodal
11 distribution develops seasonally. A modest melt or snowfall event can trigger a sudden switch
12 from a high to low albedo mode or vice versa, drastically changing ablation rates. These
13 findings suggest that a shift in dominant surface type from snow to bare ice, and clean ice to
14 impurity-rich surfaces is an important driver in abruptly increasing seasonal ice sheet melt
15 rates.

16 The first quality-controlled in situ ablation area albedo dataset collected along a 1.25
17 km transect during three days in June 2013 is presented. Albedo data collected during in situ
18 transect dates resemble an early summer ice surface classified in Chandler et al. (2014) and
19 Knap and Oerlemans (1996; Fig. 6). Here, remaining snow cover and superimposed ice
20 gradually melts, revealing underlying impurities and cryoconite holes. Visual assessment and
21 continuous monitoring in the field revealed that the ice surface along the transect was snow-
22 free from 8-26 June 2013. This period corresponds to a non-linear decrease in albedo (Fig. 3).
23 Accumulation of exposed below-surface impurities (Wientjes and Oerlemans, 2010), the
24 gradual erosion of snow patches in local depressions on the ice surface (van den Broeke et al.,
25 2011), as well as the activation and development of the hydrologic system and cryoconite
26 hole coverage (Chandler et al., 2014) may mitigate the rate of change in ablation area albedo.
27 Turbulent sensible heat fluxes from adjacent pro-glacial areas provide an additional
28 explanation for the non-linear decline in ground albedo measurements, serving to limit the
29 melt-albedo feedback's influence (van den Broeke et al., 2011).

30 Under the assumptions that distinct surface types follow a normal distribution, a
31 bimodal probability distribution preferentially develops as ablation area albedo decreases

1 rapidly over the melt season due to development of an efficient meltwater drainage system,
2 increase in cryoconite hole coverage, and accumulation of debris-rich sediments (Fig. 6). An
3 increase in debris-rich and stream surfaces over the melting season (Fig. 6) is likely
4 responsible for the enhanced frequency of low albedo values identified in the observed α_{ASD}
5 distribution from 16-25 June (Fig. 7). However, the observed changes at transect sites appear
6 to be more gradual than for the MODIS data (Fig. 10 and 15). This may be due to a lack of
7 snow cover influencing the local albedo distribution and a lower temporal sampling
8 frequency. The lack of a pronounced secondary mode with lower albedo values in the
9 observed left-skewed distributions (Fig. 7) compared to the modeled bimodal distribution
10 (Fig. 6) may be related to different melt season conditions (2012 vs. 2013), and corresponding
11 range of surface types captured along the transect, which undersamples dark surfaces (e.g.,
12 dirty ice and stream surfaces; Fig. 5). While, Chandler et al. (2014) surface types cover a
13 wider range of surface types, and thus, albedos.

14 Compared to reality, the computed distribution (Fig. 6) probably overemphasizes each
15 mode and does not account for darkening due to ice crystal growth over the melting season.
16 The observed albedo distributions reveal abrupt and variable shifts in the seasonal albedo
17 distribution (Fig. 10 and 11). At certain spatial extents, these albedo distributions transition
18 from a high- to low-dominated mode (Fig. 9), enabling enhanced melt rates (Fig. 8 and 14).
19 Alexander et al. (2014) also observed bimodal albedo distributions for Greenland's ablation
20 area by analyzing MAR and MODIS products between 2000-2013. Alexander et. al. (2014)
21 attributes the dominant modes to the presence of snow and ice (and firn). This is in agreement
22 with the analysis of the 2013 conditions, but disagrees with 2012 conditions. This discrepancy
23 could be due to the larger study area that includes areas unaffected by dust from deposition
24 and outcropped ice layers, and a thirteen-year averaging period suppressing outlier years like
25 2012 used in Alexander et al. (2014).

26 The bimodal albedo distribution and shift from a higher to a lower albedo mode
27 centering at values below 0.4 (Fig. 6, 10, and 11) indicate that a switch in dominant surface
28 type (i.e., from light to dark) during the melt season, and not solely grain size metamorphism,
29 are largely responsible for lowering albedo in snow-free ablation areas. Furthermore, results
30 from the MODIS data (Fig. 10 and 11) suggest that a transition from a light- to dark-
31 dominated surface is abrupt rather than gradual, likely associated with the addition and
32 removal of snow. The transition is more gradual in the left-skewed observed (Fig. 7) and

1 computed albedo distributions (Fig. 6), likely reflecting changes in impurity content and
2 different time stamps. Consistent with Chandler et al. (2014), the initial drop in MODIS
3 ablation area albedo is likely due to the transition from dry to wet, and patchy snow surfaces.
4 Successive lowering of albedo after snow melt is predominantly due to an increase ice crystal
5 size and possibly also by expansion of darker surface area coverage (e.g., cryoconite holes,
6 accumulation of impurities, and stream organization) and melting of dust-enriched ice layers.
7 These distributions correspond to percent differences (e.g., 51.5% between 10 – 14 July and
8 20 – 24 July pentads) in melt rate estimates that are substantial over the melt season (Fig. 8
9 and 14), and highlight the importance of considering the albedo of ablation area surface types.
10 The higher melt rates associated with darker surfaces (Fig. 16) may lead to lighter surfaces
11 becoming topographically prominent. In theory, this should enhance sensible heat transfer to
12 the lighter surfaces, increasing their ablation. Future studies should consider quantifying the
13 effects of surface roughness on ablation area albedo (e.g., Warren et al., 1998; Zhuravleva and
14 Kokhanovsky, 2011), and the possibility of enhanced ablation of light surfaces following
15 upon adjacent, dark surface ablation.

16 Recent studies have proposed scenarios of future atmospheric warming, where excess
17 deposition of light-absorbing impurities (Dumont et al., 2014) and black carbon from
18 increased forest fire frequency or incomplete fuel combustion (Keegan et al., 2014), will
19 promote accumulation of impurities, contributing to amplified surface melting. If these
20 findings turn out to be true, these effects will likely be exacerbated in southwest Greenland's
21 ablation area, where continued negative albedo trends (Stroeve et al., 2013), and increasingly
22 warmer average summer temperatures (Keegan et al., 2014), in conjunction with bare ice,
23 light-absorbing impurities, and cryoconite holes, are expected to dominate.

24

25 **5.2 Insights from 2012 and 2013 melt seasons' albedo distributions**

26 The spatial distribution of snow cover and background bare ice albedo is important for
27 understanding temporal changes in 2012 and 2013 MODIS albedo distributions (Fig. 12 and
28 13). Compared to 2013, snow melt in 2012 was more pronounced and reached higher
29 elevations (Tedesco et al., 2014), allowing the dark band feature to be exposed, resulting in a
30 lower seasonal albedo mode (Fig. 15).

1 The large albedo distribution changes from one MODIS pentad to another in 2012
2 (Fig. 11) is likely due to variability in meltwater ponding on the ice surface, and perhaps
3 deposition of wind-blown dust from tundra regions, and not necessarily increases in melted-
4 out debris from internal ice layers at such short timescales. However, exposure of dust and
5 sediment-rich ice surfaces probably caused the high probability of considerably low 2012
6 MODIS albedo values relative to 2013. This is expected since it was identified as an extreme
7 melt year with early onset snow melt (e.g., Nghiem et al., 2012; Tedesco et al., 2013; Fig. 11
8 and 15), while 2013 was a normal melt-year in the 1979-2013 context (Tedesco et al., 2014).
9 Given the coarse resolution of the MODIS pixel, it is likely that it averages out finer scale
10 details of distinct surface types (e.g., dirty ice and cryoconite hole surfaces) along the ice
11 sheet edge. It is hypothesized that higher spatial resolution satellite imagery may be able to
12 capture such regions closer to the ice sheet margin. We postulate that the area of these regions
13 may grow in size over the melting season as demonstrated on local scales by Chandler et al.
14 (2014) in situ observations.

15 The bimodal distribution observed in the 2013 MODIS data (Fig. 6) appears to be
16 governed by the relative extent of clean ice and snow surfaces. This aligns with findings from
17 current SMB models, as the majority of variability in the overall Greenland ablation area
18 albedo is driven by the deposition, change, and removal of snow (Alexander et al., 2014; Van
19 Angelen et al., 2012). However, 2012 MODIS albedo distributions cannot be explained by
20 transitions from snow to ice and vice versa. Instead, the 2012 MODIS albedo distributions
21 likely reflect abrupt shifts in ablation area albedo from the exposure of impurities on the ice
22 surface in the so-called “dark-band” region as well as ice crystal growth and expansion of
23 dirty ice areas, even with the presence of a few snowfall events. As such, dust and impurities
24 on Greenland’s ice sheet surface can influence surface albedo in the ablation area. The current
25 state of SMB models are capable of simulating albedo as a function of meltwater ponding
26 (Alexander et al., 2014) and impurities from atmospheric dust deposition on snow (Van
27 Angelen et al., 2012). The models might be improved by incorporating the melting out of dust
28 and sediments in outcropped ice layers, found in the dark band region.

29

30 **6 Conclusions**

31 A first high-quality in situ spectral albedo dataset collected along a fixed transect is presented
32 for southwest Greenland’s ablation area. Previous studies have attributed an increase in melt

1 season duration, less snowfall accumulation, enhanced snow grain metamorphism rates and
2 ice-albedo feedback as primary mechanisms for lowering ablation area albedo. Here, we
3 demonstrate an additional control on albedo in the ablation area, namely the distribution of
4 distinct surface types such as snow, clean ice, impurity-rich ice, melt ponds and streams, and
5 also examine their modulation on surface ablation. The spatial extent of each of these surface
6 types result in a multi-modal albedo distributions in the ablation area. Analysis of MODIS
7 data suggest that a multi-modal distribution and consequentially, a shift from light to dark-
8 dominated surfaces, and sensitivity to melting of outcropped ice layers, characterize seasonal
9 changes in Greenland's ablation area, and therefore, melt rates.

10 Continued atmospheric warming coinciding with a darkening ice surface will increase
11 the ice sheet surface meltwater production and runoff. Here, we show the importance of the
12 distribution of dirty ice surfaces, which are likely the result of accumulation of impurities
13 melted out from internal ice layers rather than contemporary deposition of atmospherically
14 transported dust. Future research should investigate the importance of surface accumulation of
15 impurities and if its surface area can change to significantly influence GrIS albedo and surface
16 ablation. Finer spatial resolution satellite imagery is needed to adequately characterize the
17 high spatial variability in surface types and their corresponding albedo in the ablation area of
18 the GrIS. Analysis of spatio-temporal variability in albedo using higher spatial resolution
19 imagery may be needed to adequately characterize surface types, particularly for dust and
20 sediment-rich surfaces, to improve our understanding of ablation area albedos' contribution to
21 GrIS mass loss.

22

23 **Acknowledgements**

24 S.E. Moustafa, A.K. Rennermalm, L.C. Smith and J.R. Mioduszewski were funded by NASA
25 grant NNX11AQ38G and NNX14AH93G. S.E. Moustafa was also funded by NASA Earth
26 and Space Science Fellowship Program NNX12AN98H. Additional funding was provided by
27 Rutgers University Faculty Research Grant. The authors would like to thank Drs. A. Pope and
28 M.S. Pelto as well as two anonymous reviewers for valuable feedback and commentary.

29

30 **References**

1 Alexander, P. M., Tedesco, M., Fettweis, X., van de Wal, R. S. W., Smeets, C. J. P. P. and
2 van den Broeke, M. R.: Assessing spatio-temporal variability and trends (2000–2013) of
3 modelled and measured Greenland ice sheet albedo, *The Cryosphere Discuss.*, 8(4), 3733–
4 3783, doi:10.5194/tcd-8-3733-2014, 2014.

5 Van Angelen, J. H., Lenaerts, J. T. M., Lhermitte, S., Fettweis, X., Kuipers Munneke, P., van
6 den Broeke, M. R., van Meijgaard, E. and Smeets, C. J. P. P.: Sensitivity of Greenland Ice
7 Sheet surface mass balance to surface albedo parameterization: a study with a regional climate
8 model, *The Cryosphere*, 6, 1175–1186, doi:10.5194/tc-6-1175-2012, 2012.

9 Bøggild, C. E., Brandt, R. E., Brown, K. J. and Warren, S. G.: The ablation zone in northeast
10 Greenland: ice types, albedos and impurities, *J. Glaciol.*, 56(195), 101–113,
11 doi:10.3189/002214310791190776, 2010.

12 Box, J. E., Fettweis, X., Stroeve, J. C., Tedesco, M., Hall, D. K., and Steffen, K.: Greenland
13 ice sheet albedo feedback: thermodynamics and atmospheric drivers, *The Cryosphere*, 6, 821-
14 839, doi:10.5194/tc-6-821-2012, 2012.

15 Van den Broeke, M., van As, D., Reijmer, C. and van de Wal, R.: Assessing and Improving
16 the Quality of Unattended Radiation Observations in Antarctica, *J. Atmos. Ocean. Technol.*,
17 21(9), 1417–1431, doi:10.1175/1520-0426(2004)021<1417:AAITQO>2.0.CO;2, 2004.

18 Van den Broeke, M., Smeets, P., Ettema, J., van der Veen, C., van de Wal, R., and
19 Oerlemans, J.: Partitioning of melt energy and meltwater fluxes in the ablation zone of the
20 west Greenland ice sheet, *The Cryosphere*, 2, 179-189, doi:10.5194/tc-2-179-2008, 2008.

21 Van den Broeke, M., van de Wal, R. and Smeets, P.: The seasonal cycle and interannual
22 variability of surface energy balance and melt in the ablation zone of the west Greenland ice
23 sheet, *The Cryosphere*, 5, 377-390, doi:10.5194/tc-5-377-2011, 2011.

24 Chandler, D. M., Alcock, J. D., Wadham, J. L., Mackie, S. L. and Telling, J.: Seasonal
25 changes of ice surface characteristics and productivity in the ablation zone of the Greenland
26 Ice Sheet, *The Cryosphere Discuss.*, 8, 1337–1382, doi:10.5194/tcd-8-1337-2014, 2014.

27 Cuffey, K. and Paterson, W. S. B.: *The Physics of Glaciers*, 4th Edn., Elsevier Inc., Burlington
28 and Oxford, 2010.

1 Dumont, M., Brun, E., Picard, G., Michou, M., Libois, Q., Petit, J., Geyer, M., Morin, S. and
2 Josse, B.: Contribution of light-absorbing impurities in snow to Greenland's darkening since
3 2009, *Nat. Geosci.*, 7(7), 509–512, doi:10.1038/ngeo2180, 2014.

4 Dumont, M., Gardelle, J., Sirguey, P., Guillot, a., Six, D., Rabatel, a. and Arnaud, Y.: Linking
5 glacier annual mass balance and glacier albedo retrieved from MODIS data, *The Cryosphere*,
6 6, 1527–1539, doi:10.5194/tc-6-1527-2012, 2012.

7 Ettema, J., van den Broeke, M. R., van Meijgaard, E., van de Berg, W. J., Box, J. E. and
8 Steffen, K.: Climate of the Greenland ice sheet using a high-resolution climate model – Part
9 1: Evaluation, *The Cryosphere*, 4, 511–527, doi:10.5194/tc-4-511-2010, 2010.

10 Fettweis, X.: Reconstruction of the 1979–2006 Greenland ice sheet surface mass balance
11 using the regional climate model MAR, *The Cryosphere*, 1, 21–40, doi:10.5194/tcd-1-123-
12 2007, 2007.

13 Fettweis, X., Tedesco, M., van den Broeke, M., and Ettema, J.: Melting trends over the
14 Greenland ice sheet (1958–2009) from spaceborne microwave data and regional climate
15 models, *The Cryosphere*, 5, 359–375, doi:10.5194/tc-5-359-2011, 2011.

16 Fitzgerald, P. W., Bamber, J. L., Ridley, J. K. and Rougier, J. C.: Exploration of parametric
17 uncertainty in a surface mass balance model applied to the Greenland ice sheet, *J. Geophys.*
18 *Res.*, 117(F1), F01021, doi:10.1029/2011JF002067, 2012.

19 Grenfell, T. C. and Perovich, D. K.: Seasonal and spatial evolution of albedo in a snow-ice-
20 land-ocean environment, *J. Geophys. Res.*, 109(C1), C01001, doi:10.1029/2003JC001866,
21 2004.

22 Hall, D.K., G.A. Riggs and V.V. Salomonson.: "MODIS Snow and Sea Ice Products," 2006:
23 *Earth Science Satellite Remote Sensing - Volume I: Science and Instruments*, J.J. Qu, W.
24 Gao, M. Kafatos, R.E. Murphy and V.V. Salomonson (eds.), Springer, New York, pp. 154-
25 181, 2006.

26 Hall, D. K., Comiso, J. C., DiGirolamo, N. E., Shuman, C. A., Box, J. E. and Koenig, L. S.:
27 Variability in the surface temperature and melt extent of the Greenland ice sheet from
28 MODIS, *Geophys. Res. Lett.*, 40(10), 2114–2120, doi:10.1002/grl.50240, 2013.

29 Hanna, E., Fettweis, X., Mernild, S. H., Cappelen, J., Ribergaard, M. H., Shuman, C. a.,
30 Steffen, K., Wood, L. and Mote, T. L.: Atmospheric and oceanic climate forcing of the

1 exceptional Greenland ice sheet surface melt in summer 2012, *Int. J. Climatol.*, 34(4), 1022–
2 1037, doi:10.1002/joc.3743, 2014.

3 Holben, N., Tanr, D., Smirnov, A., Eck, T. F., Slutsker, I., Newcomb, W. W., Schafer, J. S.,
4 Chatenet, B., Lavenu, F., Kaufman, J., Castle, J. Vande, Setzer, A., Markham, B., Clark, D.,
5 Halthore, R., Karneli, A., Neill, N. T. O., Pietras, C., Pinker, T., Voss, K. and Zibordi, G.: An
6 emerging ground-based aerosol climatology : Aerosol optical depth from AERONET, *J.*
7 *Geophys. Res.*, 106(D11), 12,067–12,097, 2001.

8 Iqbal, M.: Spectral and total sun radiance under cloudless skies. *Physical Climatology for*
9 *Solar and Wind Energy*, R. Guzzi and C. G. Justus, Eds., World Scientific, 196–242, 1988.

10 Keegan, K. M., Albert, M. R., McConnell, J. R. and Baker, I.: Climate change and forest fires
11 synergistically drive widespread melt events of the Greenland Ice Sheet., *Proc. Natl. Acad.*
12 *Sci. U. S. A.*, (14), 1–4, doi:10.1073/pnas.1405397111, 2014.

13 Klein, A. G., & Stroeve, J.: Development and validation of a snow albedo algorithm for the
14 MODIS instrument. *Annals of Glaciol.*, 34(1), 45-52, 2002.

15 Knap, W. H. and Oerlemans, J.: The surface albedo of the Greenland ice sheet: satellite
16 derived and in situ measurements in the Sendre Stromfjord area during the 1991 melt season,
17 *J. Glaciol.*, 42(141), 364–374, 1996.

18 Konzelmann, T. and Braithwaite, R. J.: Variations of ablation, albedo and energy balance at
19 the margin of the Greenland ice sheet, Kronprins Christian Land, eastern north Greenland, *J.*
20 *Glaciol.*, 41(137), 174–182, 1995.

21 Kuhn, M.: Anisotropic reflection from sastrugi fields. *Antarctic Journal of the United States*,
22 9, 123–125, 1974.

23 Lampkin, D. J. and VanderBerg, J.: Supraglacial melt channel networks in the Jakobshavn
24 Isbrae region during the 2007 melt season, *Hydrol. Process.*, doi:10.1002/hyp.10085, 2013.

25 Legleiter, C. J., Tedesco, M., Smith, L. C., Behar, a. E. and Overstreet, B. T.: Mapping the
26 bathymetry of supraglacial lakes and streams on the Greenland ice sheet using field
27 measurements and high-resolution satellite images, *The Cryosphere*, 8, 215–228,
28 doi:10.5194/tc-8-215-2014, 2014.

1 Van Meijgaard, E., van Uft, L. H., Van de Berg, W. J., Bosveld, F. C., Van den Hurk, B.,
2 Lenderink, G., and Siebesma, A. P.: The KNMI regional atmospheric climate model RACMO
3 version 2.1, Tech. Rep. 302, Royal Netherlands Meteorological Institute, De Bilt, 2008.

4 Mernild, S. H. and Liston, G. E.: Greenland Freshwater Runoff. Part II: Distribution and
5 Trends, 1960–2010, *J. Clim.*, 25(17), 6015–6035, doi:10.1175/JCLI-D-11-00592.1, 2012.

6 Nghiem, S. V., Hall, D. K., Mote, T. L., Tedesco, M., Albert, M. R., Keegan, K., Shuman, C.
7 A., DiGirolamo, N. E. and Neumann, G.: The extreme melt across the Greenland ice sheet in
8 2012, *Geophys. Res. Lett.*, 39(20), 1–6, doi:10.1029/2012GL053611, 2012.

9 Onset Computer Corp.: Silicon Pyranometer Smart Sensor (Part # S-LIB-M003), 1–6, 2010.

10 Rae, J. G. L., Aðalgeirsdóttir, G., Edwards, T. L., Fettweis, X., Gregory, J. M., Hewitt, H. T.,
11 Lowe, J. A., Lucas-Picher, P., Mottram, R. H., Payne, A. J., Ridley, J. K., Shannon, S. R., van
12 de Berg, W. J., van de Wal, R. S. W., and van den Broeke, M. R.: Greenland ice sheet surface
13 mass balance: evaluating simulations and making projections with regional climate models,
14 *The Cryosphere*, 6, 1275-1294, doi:10.5194/tc-6-1275-2012, 2012.

15 Rennermalm, A. K., Smith, L. C., Chu, V. W., Box, J. E., Forster, R. R., Van den Broeke, M.
16 R., Van As, D. and Moustafa, S. E.: Evidence of meltwater retention within the Greenland ice
17 sheet, *The Cryosphere*, 7, 1433–1445, doi:10.5194/tc-7-1433-2013, 2013.

18 Román, M. O., Schaaf, C. B., Lewis, P., Gao, F., Anderson, G. P., Privette, J. L., Strahler, A.
19 H., Woodcock, C. E. and Barnsley, M.: Assessing the coupling between surface albedo
20 derived from MODIS and the fraction of diffuse skylight over spatially-characterized
21 landscapes, *Remote Sens. Environ.*, 114(4), 738–760, doi:10.1016/j.rse.2009.11.014, 2010.

22 Schaaf, C. B., Liu, J., Gao, F., & Strahler, A. H.: Aqua and Terra MODIS albedo and
23 reflectance anisotropy products. In *Land Remote Sensing and Global Environmental Change*
24 (pp. 549-561). Springer New York, 2011.

25 Schaepman-Strub, G., Schaepman, M. E., Painter, T. H., Dangel, S. and Martonchik, J. V.:
26 Reflectance quantities in optical remote sensing—definitions and case studies, *Remote Sens.*
27 *Environ.*, 103(1), 27–42, doi:10.1016/j.rse.2006.03.002, 2006.

28 L. C. Smith, V. W. Chu, K. Yang, C. J. Gleason, L. H. Pitcher, A. K. Rennermalm, C. J.
29 Legleiter, A. E. Behard, B. T. Overstreet, S. E. Moustafa, M. Tedesco, R. R. Forster, A. L.
30 LeWinter, D. C. Finnegan, Y. Sheng, and J. Balog. Efficient meltwater drainage through

1 supraglacial streams and rivers on the southwest Greenland ice sheet. Proceedings of the
2 National Academy of Sciences, doi:10.1073/pnas.1413024112, 2015.

3 Steffen, K. and Box, J.: Surface climatology of the Greenland ice sheet: Greenland Climate
4 Network 1995-1999, *J. Geophys. Res.*, 106(D24), 33951–33964, 2001.

5 Stroeve, J., Box, J. E., Gao, F., Liang, S., Nolin, A. and Schaaf, C.: Accuracy assessment of
6 the MODIS 16-day albedo product for snow: comparisons with Greenland in situ
7 measurements, *Remote Sens. Environ.*, 94(1), 46–60, doi:10.1016/j.rse.2004.09.001, 2005.

8 Stroeve, J., Box, J. E., Wang, Z., Schaaf, C. and Barrett, A.: Re-evaluation of MODIS
9 MCD43 Greenland albedo accuracy and trends, *Remote Sens. Environ.*, 138, 199–214,
10 doi:10.1016/j.rse.2013.07.023, 2013.

11 Stroeve, J. C., Box, J. E. and Haran, T.: Evaluation of the MODIS (MOD10A1) daily snow
12 albedo product over the Greenland ice sheet, *Remote Sens. Environ.*, 105(2), 155–171,
13 doi:10.1016/j.rse.2006.06.009, 2006.

14 Tedesco, M., Fettweis, X., van den Broeke, M. R., van de Wal, R. S. W., Smeets, C. J. P. P.,
15 van de Berg, W. J., Serreze, M. C. and Box, J. E.: The role of albedo and accumulation in the
16 2010 melting record in Greenland, *Environ. Res. Lett.*, 6(1), 014005, doi:10.1088/1748-
17 9326/6/1/014005, 2011.

18 Tedesco, M., Fettweis, X., Mote, T., Wahr, J., Alexander, P., Box, J. E. and Wouters, B.:
19 Evidence and analysis of 2012 Greenland records from spaceborne observations, a regional
20 climate model and reanalysis data, *The Cryosphere*, 7, 615–630, doi:10.5194/tc-7-615-2013,
21 2013.

22 M. Tedesco, J. E. Box, J. Cappelen, X. Fettweis, T. Jensen, T. L. Mote, A. K. Rennermalm, L.
23 C. Smith, R. S. W. van de Wal, J. Wahr. The Arctic: Greenland Ice Sheet in “State of the
24 Climate 2013.” *Bulletin of the American Meteorological Society*, 95(7), S136-S137,
25 doi:10.1175/2014BAMSStateoftheClimate.1, 2014.

26 Van de Wal, R. S. W., Boot, W., Smeets, C. J. P. P., Snellen, H., van den Broeke, M. R. and
27 Oerlemans, J.: Twenty-one years of mass balance observations along the K-transect, West
28 Greenland, *Earth Syst. Sci. Data*, 4(1), 31–35, doi:10.5194/essd-4-31-2012, 2012.

29 Wang, Z., Schaaf, C. B., Chopping, M. J., Strahler, A. H., Wang, J., Román, M. O., Rocha, A.
30 V., Woodcock, C. E. and Shuai, Y.: Evaluation of Moderate-resolution Imaging

- 1 Spectroradiometer (MODIS) snow albedo product (MCD43A) over tundra, *Remote Sens.*
2 *Environ.*, 117, 264–280, doi:10.1016/j.rse.2011.10.002, 2012.
- 3 Wientjes, I. G. M. and Oerlemans, J.: An explanation for the dark region in the western melt
4 zone of the Greenland ice sheet, *The Cryosphere*, 4, 261–268, doi:10.5194/tc-4-261-2010,
5 2010.
- 6 Wientjes, I. G. M., Van de Wal, R. S. W., Reichert, G. J., Sluijs, A. and Oerlemans, J.: Dust
7 from the dark region in the western ablation zone of the Greenland ice sheet, *The Cryosphere*,
8 5, 589–601, doi:10.5194/tc-5-589-2011, 2011.
- 9 Wright, P., Bergin, M., Dibb, J., Lefer, B., Domine, F., Carman, T., Carmagnola, C., Dumont,
10 M., Courville, Z., Schaaf, C. and Wang, Z.: Comparing MODIS daily snow albedo to spectral
11 albedo field measurements in Central Greenland, *Remote Sens. Environ.*, 140, 118–129,
12 doi:10.1016/j.rse.2013.08.044, 2014.
- 13 Yang, K. and Smith, L. C.: Supraglacial Streams on the Greenland Ice Sheet Delineated From
14 Combined Spectral–Shape Information in High-Resolution Satellite Imagery, *IEEE Geosci.*
15 *Remote Sens. Lett.*, 10(4), 801–805, doi:10.1109/LGRS.2012.2224316, 2013.

1 Table 1. Descriptive statistics for high-quality albedo transects. SZA and CC listed for Base Met Station only. Brd is used to abbreviate
 2 broadband.

Transect Date	Start Time	End Time	Min Brd α_{ASD}	Max Brd α_{ASD}	Mean Brd α_{ASD}	Daily Average α_{base}	Daily Average α_{top}	Min SZA (°)	Max SZA (°)	Mean SZA (°)	Min CC	Max CC	Mean CC
16-Jun	10:32:33	11:53:57	0.260	0.754	0.550	0.404	0.636	45.615	50.454	47.828	0.135	0.176	0.157
19-Jun	10:39:30	11:35:59	0.141	0.730	0.532	0.316	0.541	46.449	49.925	48.093	0.045	0.084	0.065
25-Jun	10:20:29	11:11:00	0.210	0.670	0.490	0.333	0.525	47.963	51.525	49.677	0.119	0.138	0.125

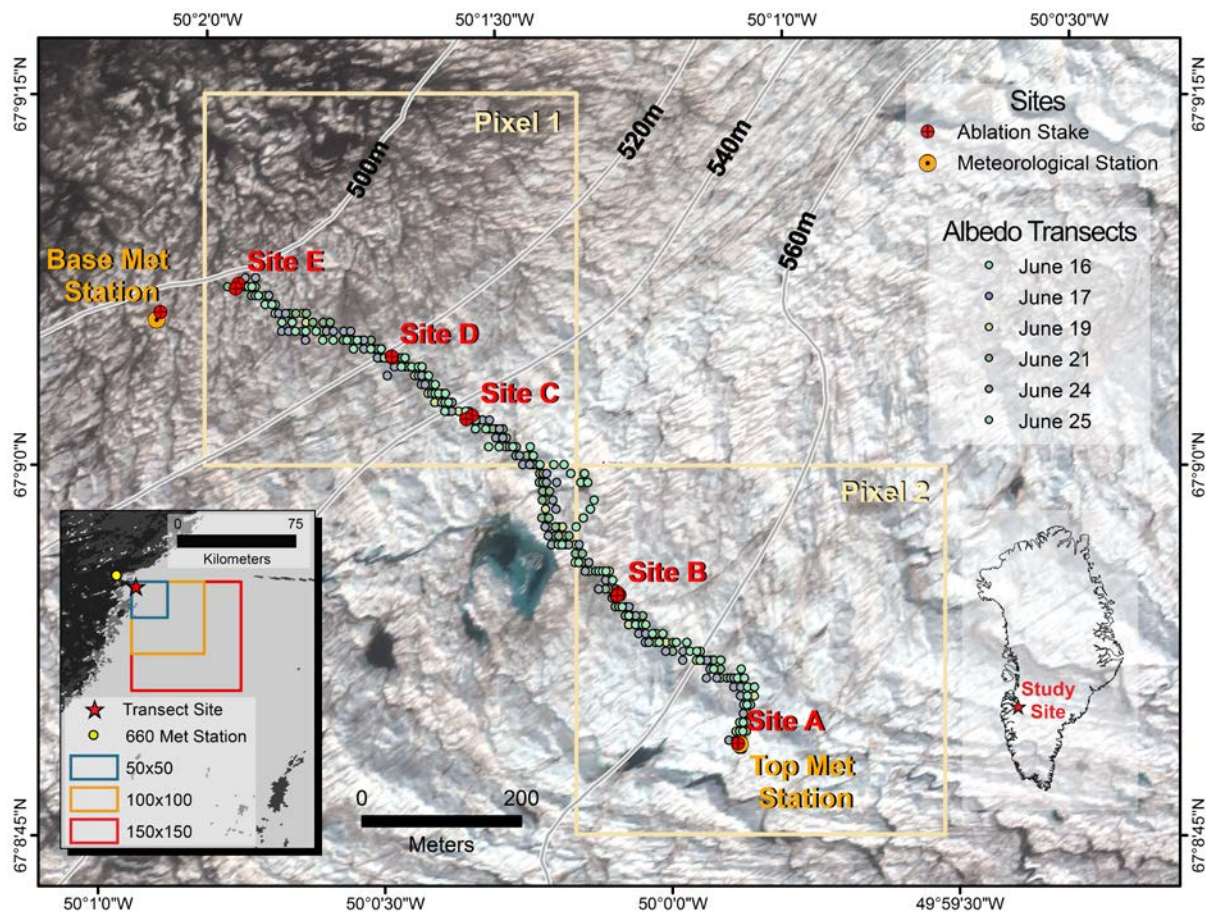
- 1 Table 2. Average broadband α_{ASD} within a 30 m radius of ablation stake sites and classified by
- 2 surface type.

Ablation stake sites	α_{ASD} site average	Clean surfaces	Dark surfaces
Site A	0.641	-	-
Site B	0.540	-	-
Site C	0.591	-	-
Site D	0.432	0.530	0.243
Site E	0.555	-	-

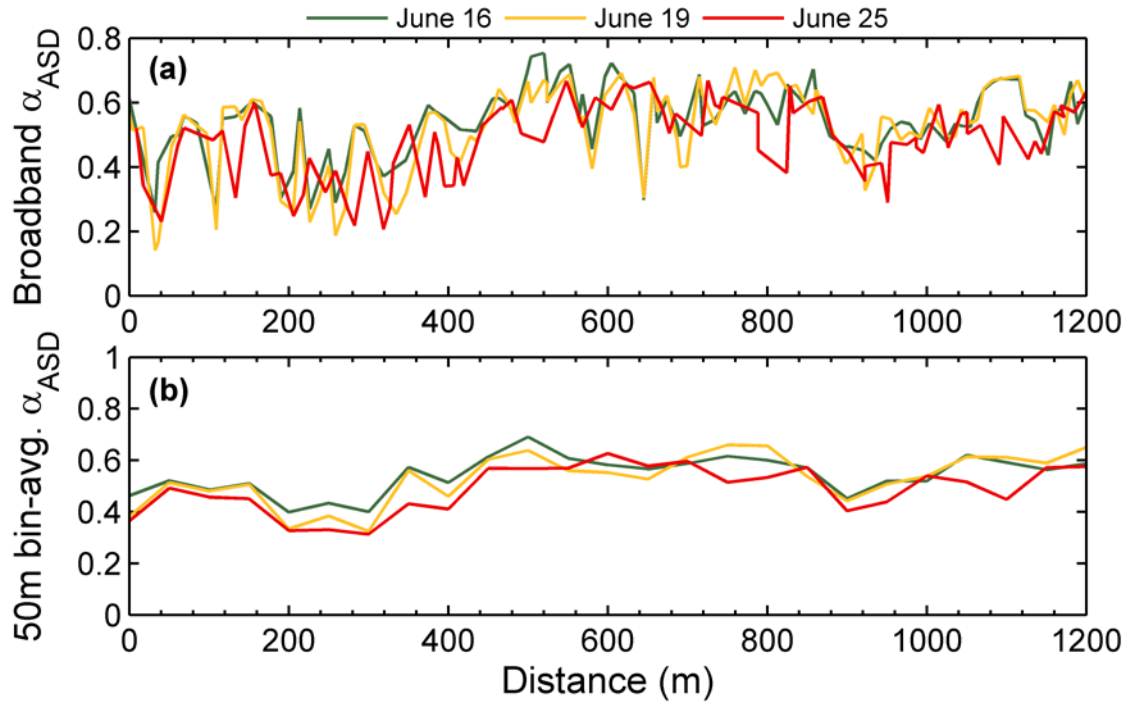
1 Table 3. Seasonal evolution (%) of four surface types at five distinct time steps approximated
 2 from Chandler et al. (2014).

Time steps	Classified names	Clean ice	Dirty ice	Streams	Cryoconite holes
1 June	Early summer ice	100	0	0	0
19 June	Dirty ice exposure	90	3	1	6
18 July	Melt	60	20	1	19
28 July	Darkening ice	50	30	3	17
5 August	Late summer ice	40	40	6	14

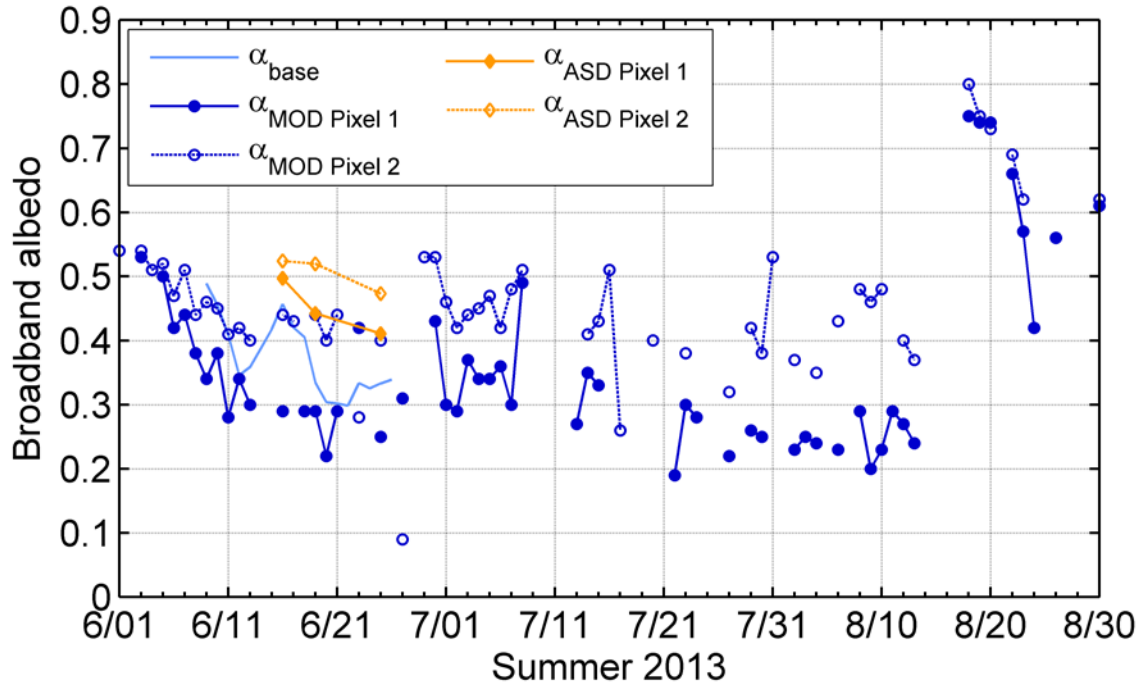
3



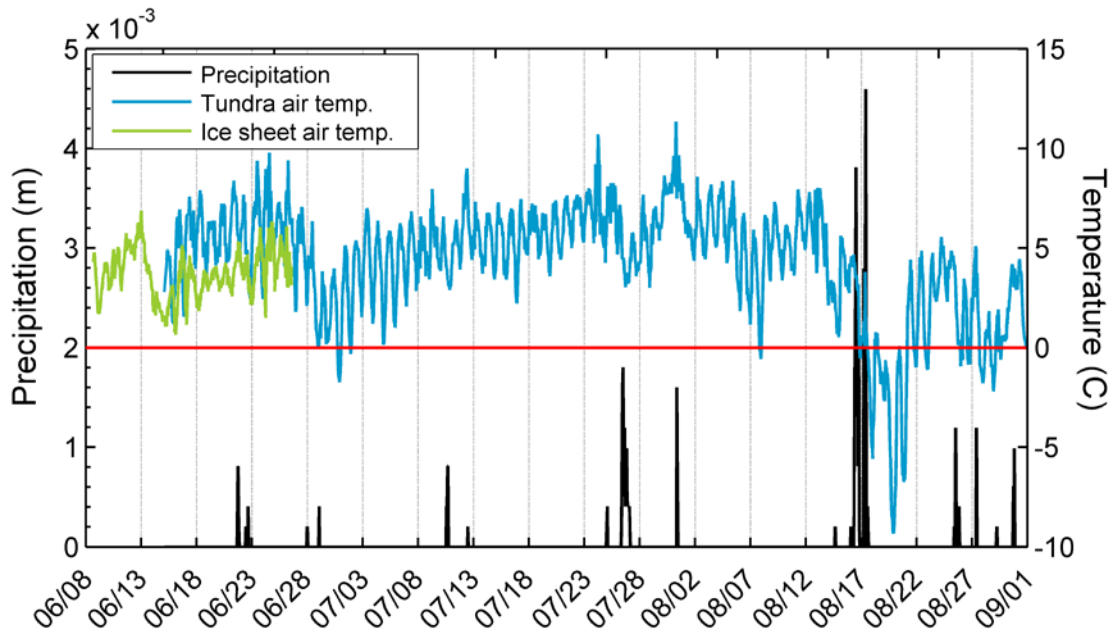
1
2 Figure 1. 23 June 2013 WorldView-2 true color image (bands 5, 3, and 2 RGB) of the study
3 site with elevation contours (m), MODIS pixel extents (yellow boxes), and location of the six
4 albedo transects, ablation stake, and meteorological station sites. Location of four MODIS
5 spatial extent regions overlaid on a 31 May 2013 MOD10A1 image (black box inset).



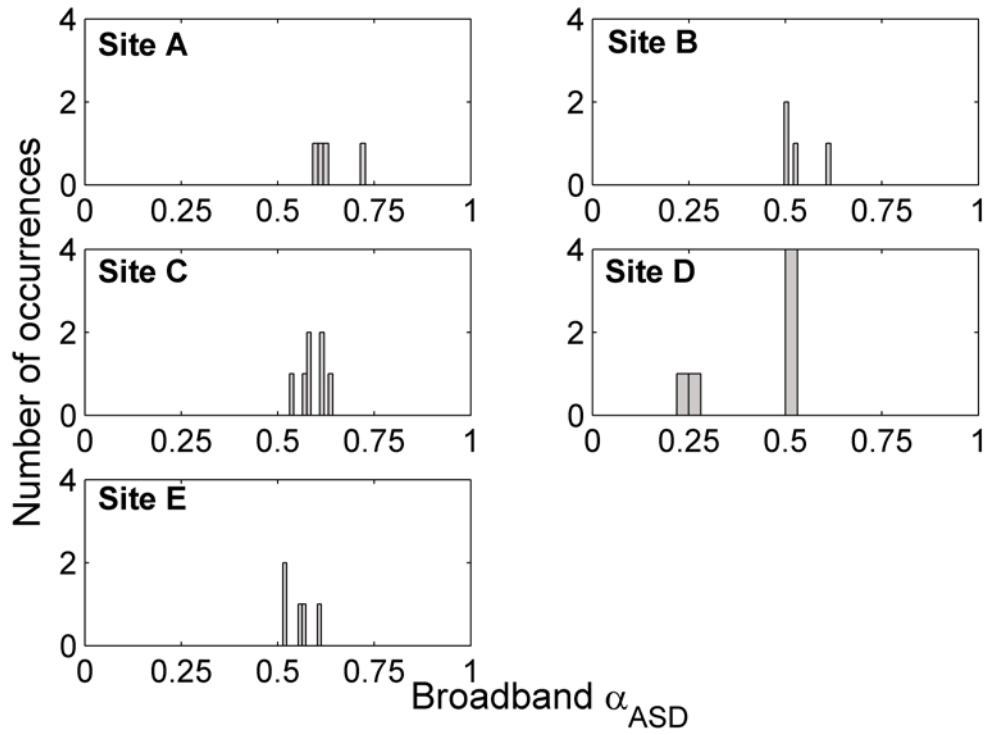
1
 2 Figure 2. High-quality broadband α_{ASD} observations on 16, 19 and 25 June (a), and broadband
 3 α_{ASD} averaged in 50 m bins (b) along the length of the transect starting near Site E (0 m) and
 4 ending near Site A (1200 m).



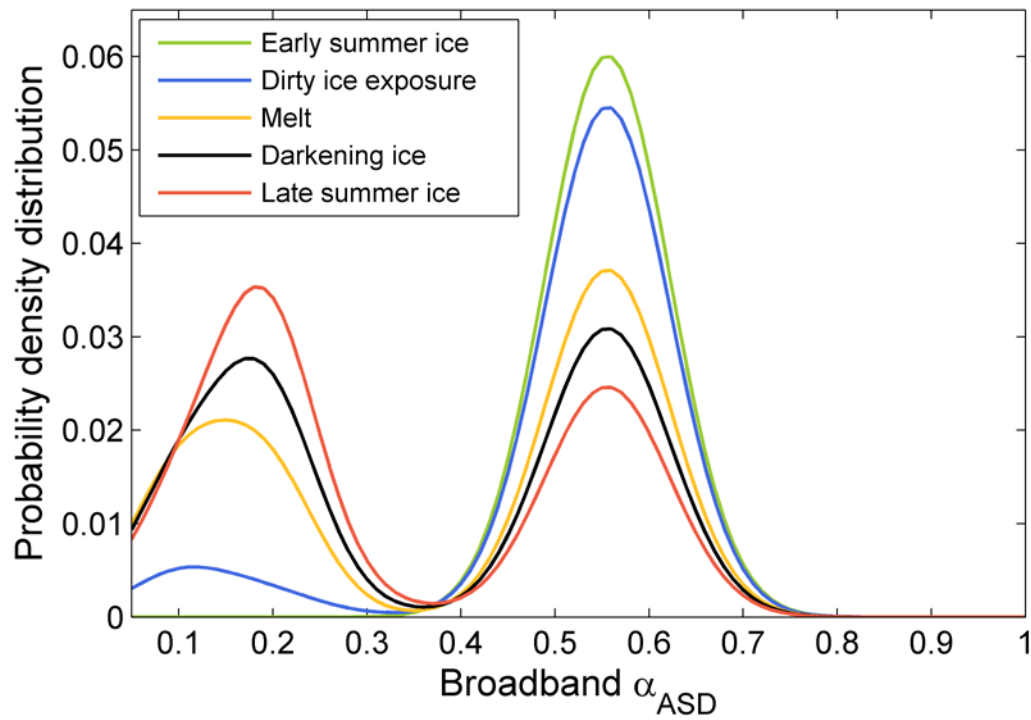
1
 2 Figure 3. High-quality daily average broadband $\alpha_{\text{ASD Pixel 1}}$ and $\alpha_{\text{ASD Pixel 2}}$, α_{base} (for SZA <
 3 70°), and $\alpha_{\text{MOD Pixel 1}}$ and $\alpha_{\text{MOD Pixel 2}}$ time series for the 2013 melt season. $\alpha_{\text{ASD Pixel 1}}$ and α_{ASD}
 4 $\alpha_{\text{MOD Pixel 2}}$ pixel-averaged values correspond to high-quality ASD transect dates 16, 19 and 25 June.



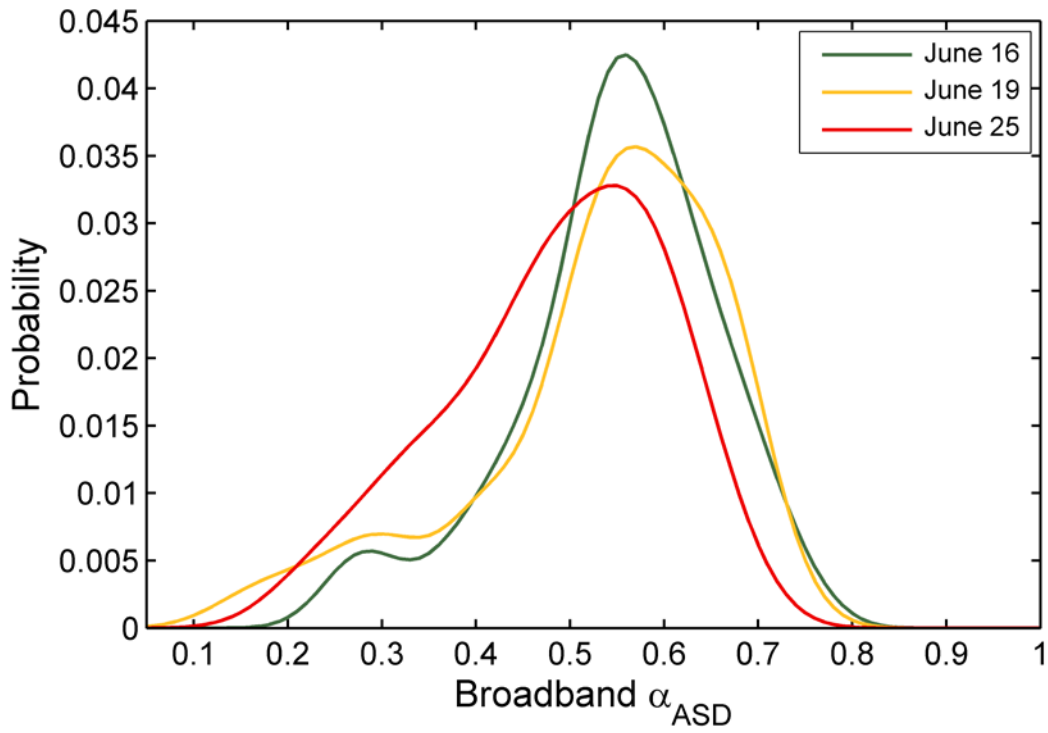
1
 2 Figure 4. Summer 2013 precipitation (left y-axis; black line) and near surface air temperature
 3 (right y-axis; blue line) time series collected from a meteorological station installed at the
 4 edge of the pro-glacial tundra environment. Base Met Station near surface air temperature
 5 time series is available from 8 – 26 June 2013 (green line). The zero degree line is in red.
 6



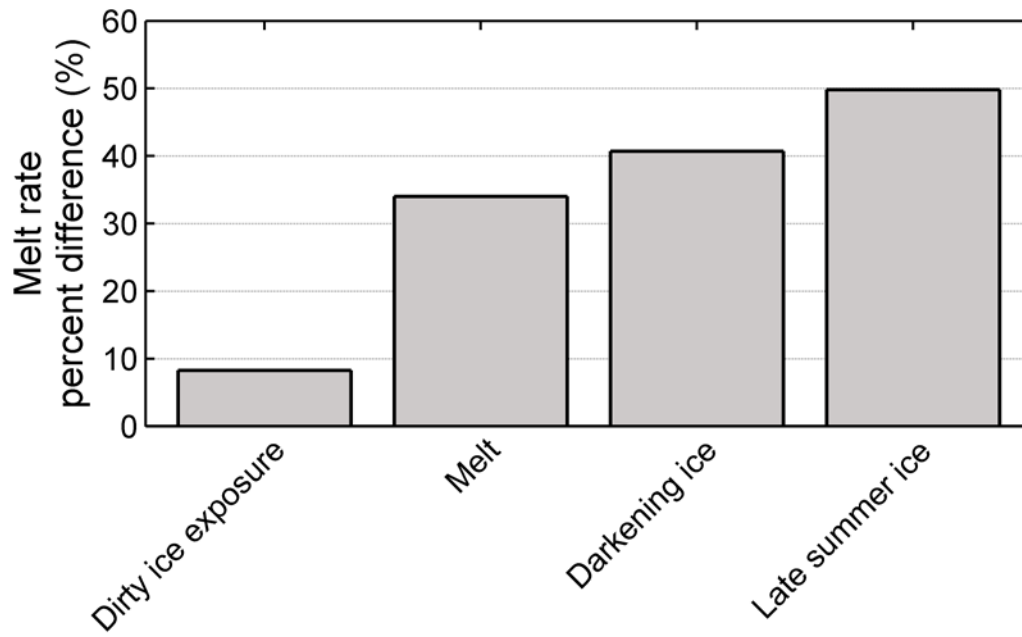
1
 2 Figure 5. Distribution of broadband α_{ASD} within 30 m radius of ablation stake sites.



1
 2 Figure 6. Computed albedo distribution for a nearby site of Chandler et al. (2014) simulated
 3 across the melt season based on observed broadband α_{ASD} values for dominant surface types,
 4 weighted by their relative surface area coverage. Each surface type is assumed to follow a
 5 normal distribution. Computed albedo distributions represent the sum of each surface type's
 6 probability distribution function.

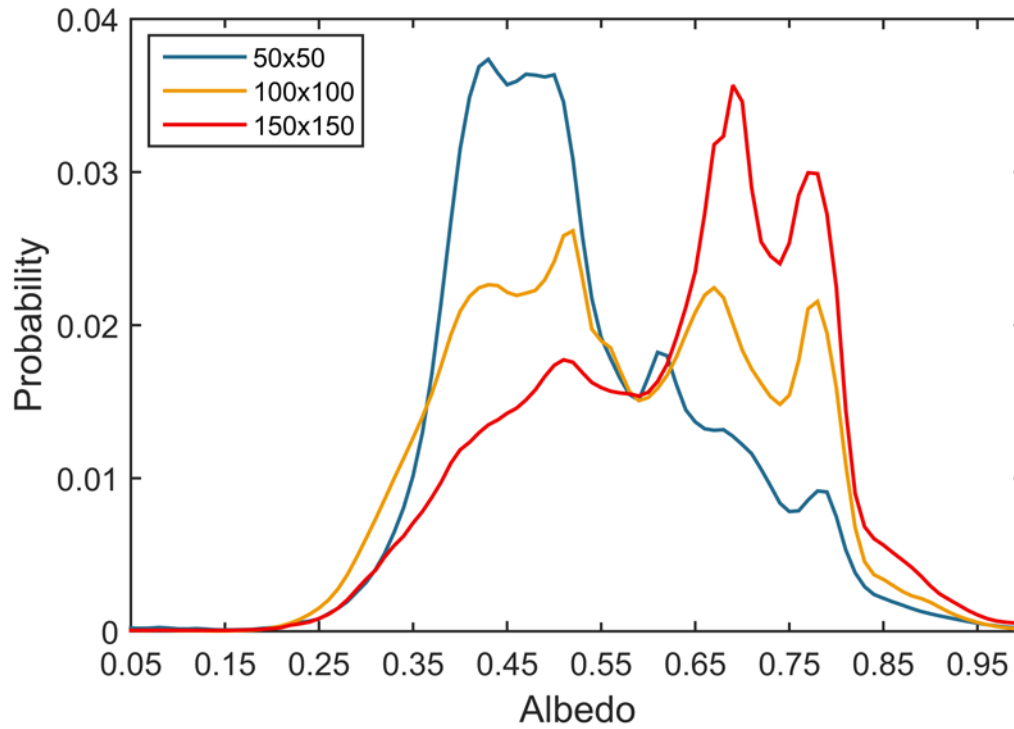


1
 2 Figure 7. Observed distributions of high-quality broadband α_{ASD} transects on June 16, 19, and
 3 25.

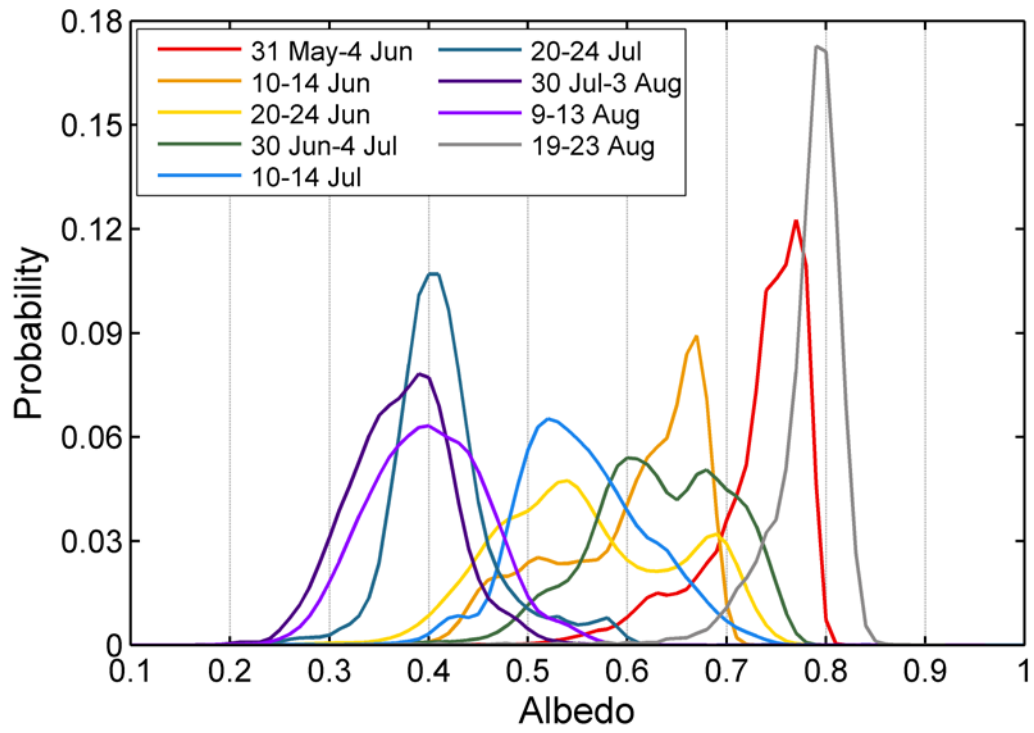


1

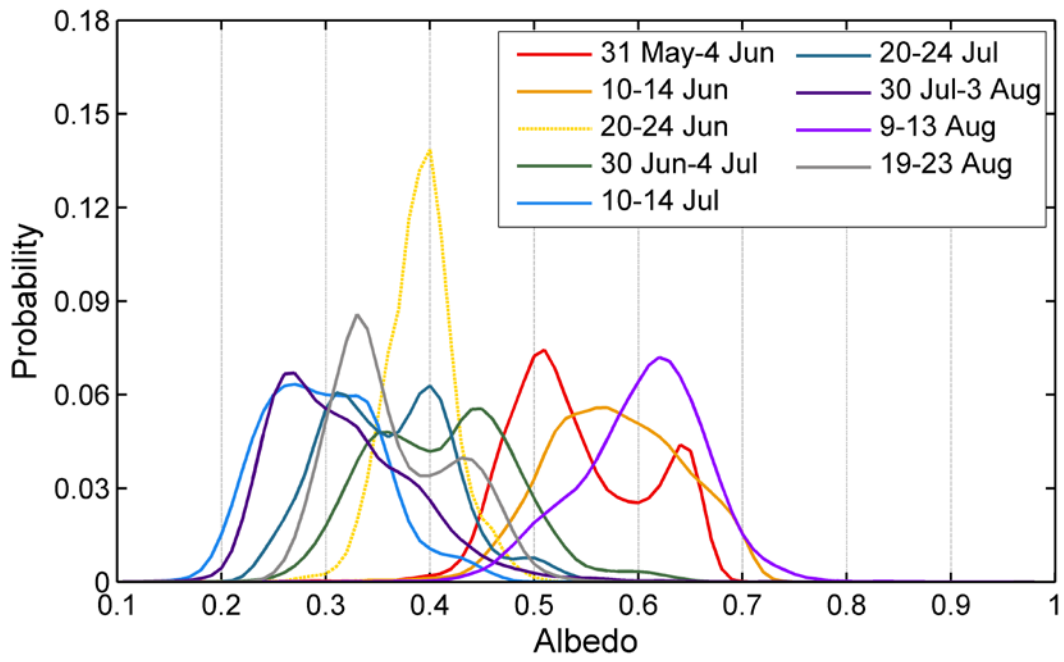
2 Figure 8. Percent difference in melt rate estimates for different albedo probability density
3 functions and averaged incoming solar radiation conditions at Base Met Station from 16, 19,
4 and 25 June, relative to 'early summer ice (1 June)' distribution.



1
 2 Figure 9. MOD10A1 2013 seasonal average albedo probability density distributions at three
 3 spatial extents, 50x50 MODIS pixels (px), 100x100 px, and 150x150 px, respectively. The
 4 bimodal distribution seen at the 100x100 px (46.3 km²) spatial extent is likely the result of
 5 almost equal area of snow and ice facies characterizing the two peaks. In contrast, the right
 6 and left skew distributions of 50x50 px and 150x150 px illustrate the dominance of ice and
 7 snow surfaces, respectively.

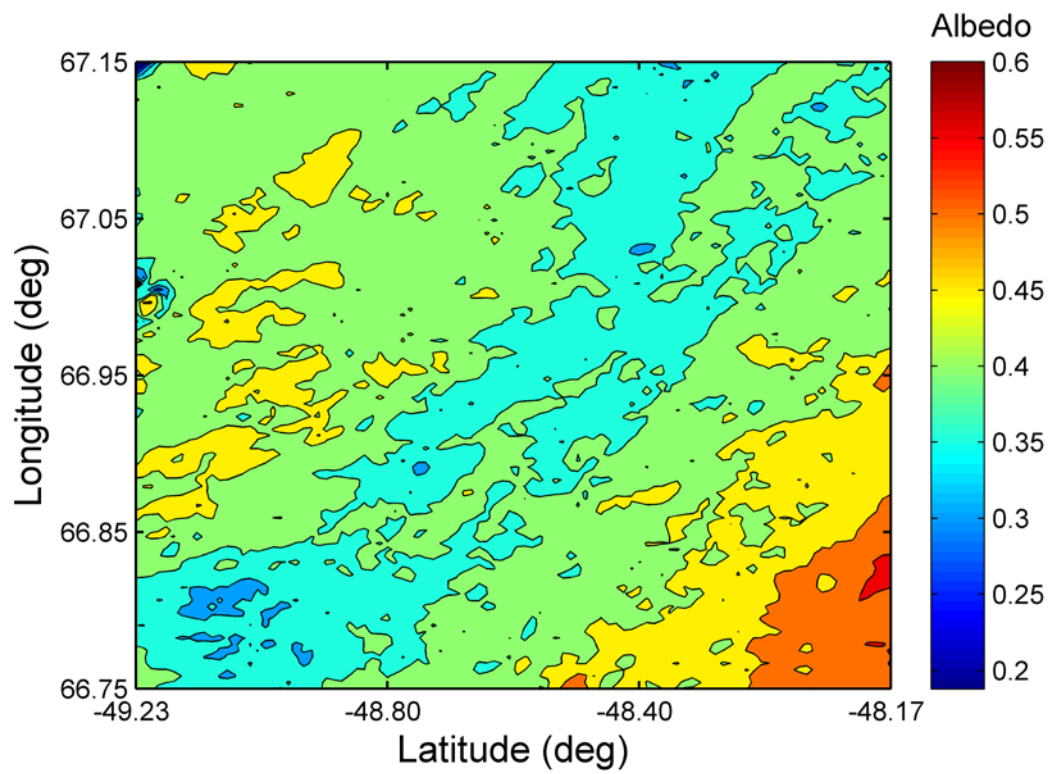


1
 2 Figure 10. 100x100 px pentad averages over the 2013 melt season. Every other pentad
 3 average line is plotted.

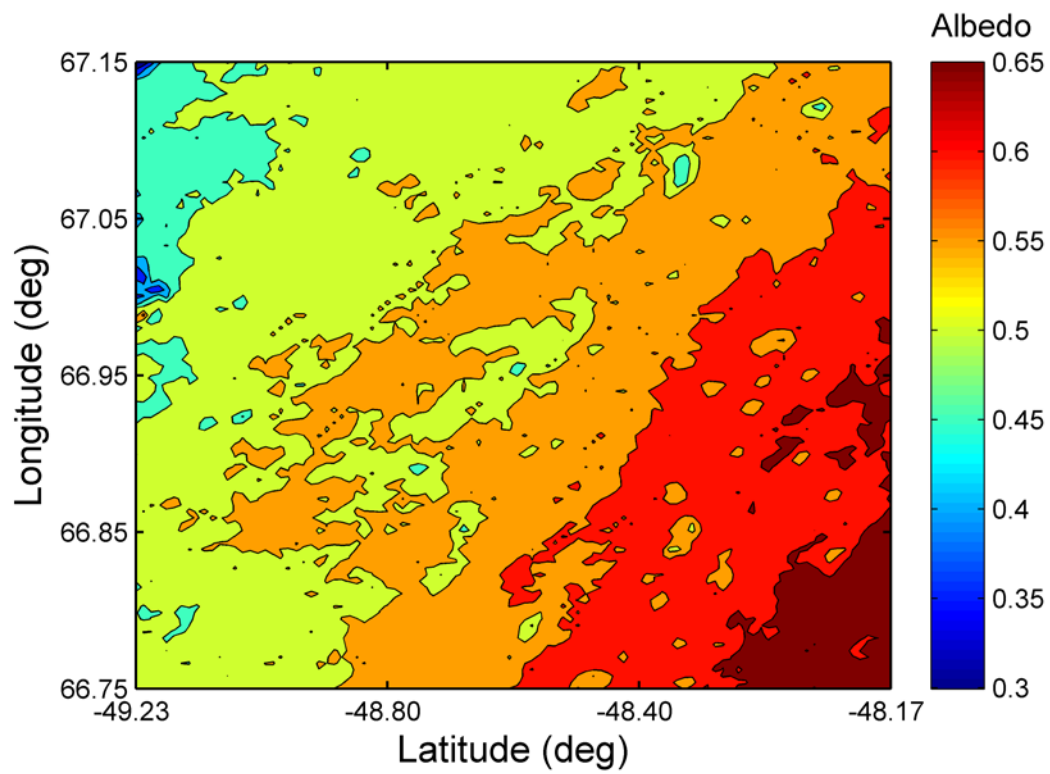


1

2 Figure 11. MODIS 100x100 px spatial extent pentad average albedo distributions for the 2012
 3 melt season. Note, the 20-24 June pentad (yellow stippled line) is most likely erroneous due
 4 to an outlier in the MODIS data on 21-22 June 2012.

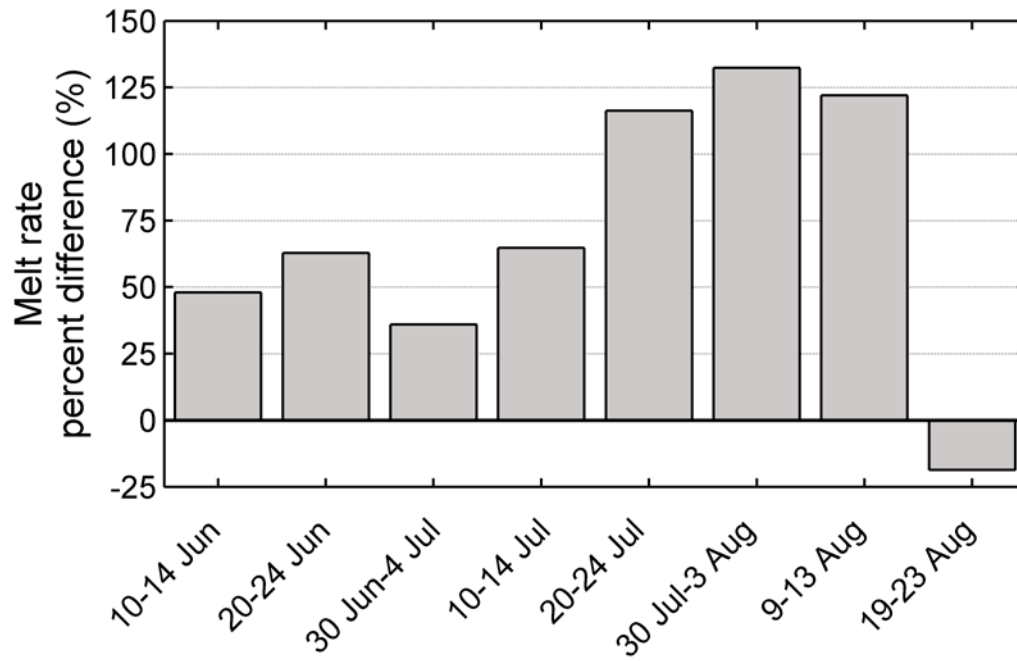


1
2 Figure 12. MODIS 2012 seasonal average for the 100x100 spatial extent. A region of dark
3 ice, known as the “dark band”, extends through our study area ($< \sim 0.35$, shown in cyan blue
4 and blue colors).

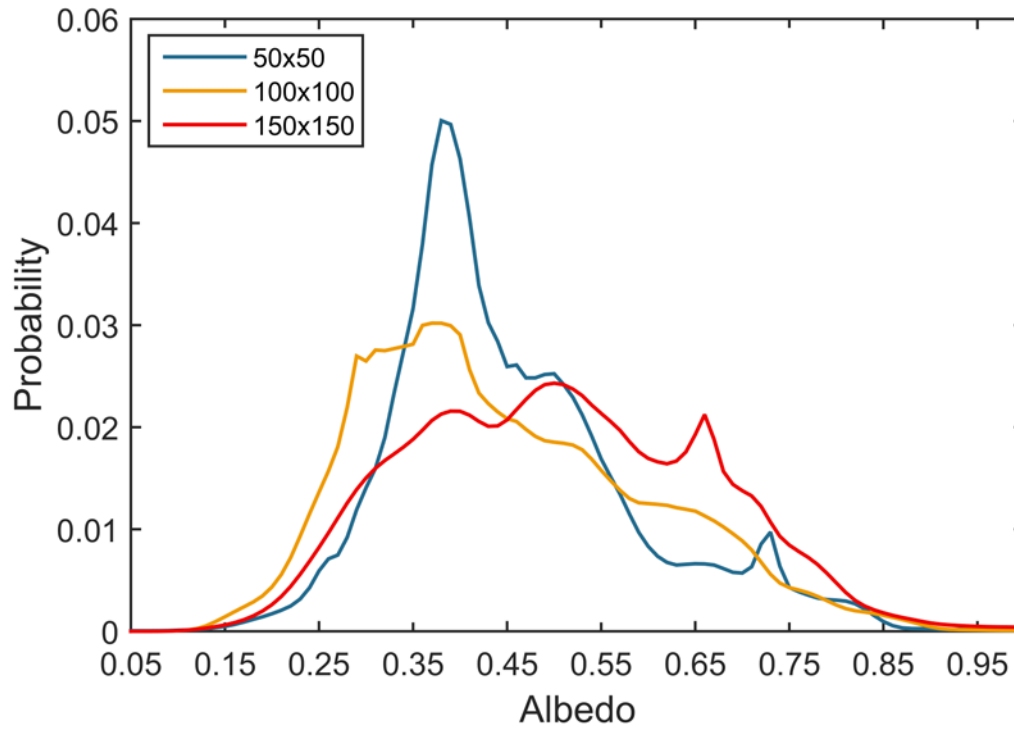


1

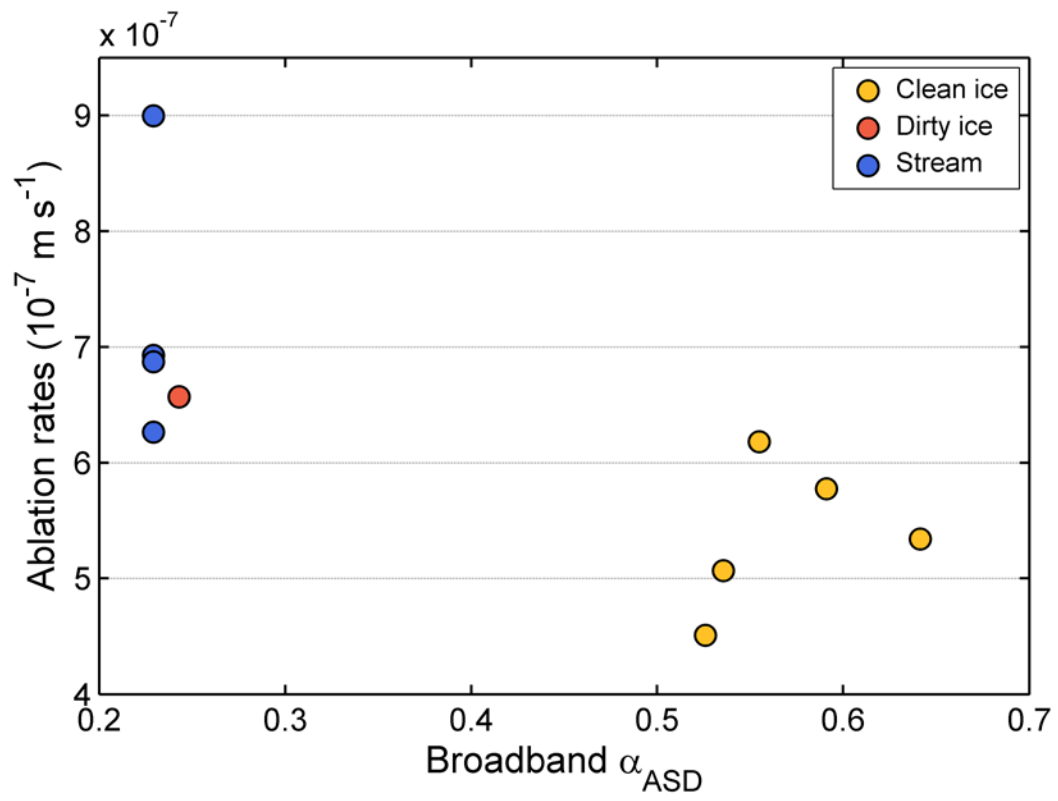
- 2 Figure 13. MODIS 2013 seasonal average for the 100x100 px spatial extent. Overall higher
3 MODIS albedo values are observed in 2013, without a “dark band” region surface expression.



1
 2 Figure 14. Percent difference in melt rate estimates for 100x100 px pentad albedo
 3 distributions for the 2013 melt season, relative to 31 May – 4 June pentad albedo distribution.
 4 Melt rates are calculated with identical radiation budget conditions to isolate the effect of
 5 albedo distribution changes.



1
 2 Figure 15. MODIS 2012 seasonal average albedo probability density distributions at three
 3 spatial extents. The MODIS 2012 seasonal average albedo probabilities for the 100x100 px
 4 and 150x150 px reveal a high probability of low albedo values (0.2 – 0.3). This is likely
 5 influenced by the expansion of the “dark band” region in these spatial extents.



1
2 Figure 16. Observed ablation rates and broadband α_{ASD} for different ice surface types.

1 **7 Appendix A**

2 **7.1 Field spectroscopy measurements**

3 At the start of each transect, the ASD was calibrated to current hemispherical atmospheric
4 conditions by orienting the RCR skyward, along a nadir-viewing angle. Subsequent
5 measurements were taken with the ASD rotated 180° to view the ice surface. Under changing
6 sky conditions, the instrument was recalibrated. Each transect consisted of ~100 sample
7 locations, roughly 10 m apart. Despite changing ice conditions rapidly deteriorating
8 temporary location markers, global positioning system (GPS) locations reveal that sample
9 sites in consecutive transects were gathered in close proximity (Fig. 1). While samples were
10 not taken from exactly the same sites preventing a point-by-point comparison, the transect
11 sample distribution and smoothed spatial patterns can be analyzed for change over time.
12 Sample sites along the transect were selected based on distance. If a spectrum site intersected
13 with a stream, melt pond, or cryoconite hole, the nearest ice surface was sampled instead. To
14 capture spectral albedo of different ice surface types, separate measurements of streams, dirty
15 ice, and white ice were collected. At each sample location, five consecutive spectra consisting
16 of 10 dark currents per scan and 10 white reference measurements were recorded and
17 averaged.

18 Apparent outliers were identified using the Spectral Analysis and Management System
19 software (SAMs) to identify outliers. Outliers were defined as physically unrealistic spectral
20 albedo values (> 1.0) and raw spectra that were significantly different to the other spectra
21 across the entire spectral range (visible and near-infrared wavelengths) taken for the same
22 sample. For 16 June, 20 spectra were deemed outliers (total spectra collected = 555); 19 June,
23 17 spectra were deemed outliers (total spectra collected = 560); and 25 June, 12 spectra were
24 deemed outliers (total spectra collected = 480). The outliers for these transect dates comprise
25 less than 4% of all spectra collected, and thus, likely had an insignificant impact on the final
26 albedo calculations. On 17 June, spectra with unrealistic > 1.0 values were collected, as will
27 be shown in section 7.2 that all data from this day were considered low-quality and removed
28 from the dataset.

29

30

1 7.2 Quality-control of α_{ASD} data

2 To ensure a high-quality α_{ASD} dataset, an impact assessment of variable cloud conditions (i.e.,
3 irregular lighting due to transient clouds) and high SZAs during late afternoon albedo transect
4 collections was made. Key et al. (2001) reported a 4-6% increase in albedo, on average, under
5 cloudy conditions. Albedo readings have also been reported as unreliable at SZAs beyond
6 70° , due to an increase in diffuse radiation reaching the ice surface (Schaaf et al., 2011;
7 Stroeve et al., 2005; Stroeve et al., 2013; Wang et al., 2012).

8 As a proxy for cloud cover, relative cloud cover, hereafter CC, was calculated every
9 second as the ratio of modeled clear-sky and observed incoming solar radiation similar to Box
10 (1997). ‘Clear-sky’ incoming shortwave fluxes at the surface were calculated with a solar
11 radiance model (Iqbal, 1988). Model inputs of water vapor content, surface pressure, aerosol
12 optical depth at 380 and 500 nm, and area optical thickness were estimated from the
13 Kangerlussuaq AEROSol Robotic NETwork (AERONET) station (Holben et al., 2001). SZA
14 was also modeled with the solar radiance model using latitude, longitude, time of day, and
15 day of year at the Base Met Station. α_{ASD} collected under high CC variability and SZAs
16 approaching extreme angles were subsequently removed. Filtering α_{ASD} data under these
17 criteria ensured the production of a high-quality dataset necessary for subsequent analysis.

18 Cloud cover and radiative conditions varied among transects (Fig. A1). The majority
19 of α_{ASD} measurements were made at small SZAs (~ 1030 - 1200 local time), except on 21 and
20 24 June, when observations were made in late afternoon (1530 - 1630 and 1640 - 1750 local
21 time, respectively). Incoming solar radiation fluxes exhibited considerable range of diurnal
22 variability (average $662 \pm 83 \text{ W m}^{-2}$). Outgoing solar radiation displayed similar range of
23 variability at lower magnitudes (average $239 \pm 18 \text{ W m}^{-2}$) during transect dates. Derived CC
24 reveals daily range of variability in cloud conditions roughly consistent with incoming solar
25 radiation observations, yet on average, remained low (~ 0.13) indicating that the majority of
26 the transect times were collected during nearly cloud-free conditions. During transect times,
27 half-hourly α_{base} changed linearly with SZA, yet remained fairly stable (Fig. A2a). Above 80°
28 SZA, half hourly α_{base} variability increased, confirming that 70° SZA was a suitable threshold
29 for daily average albedo calculations. Installation tilt, heterogeneous and changing surface
30 conditions likely contributed locally to “unstable” α_{base} observations at higher SZAs. A
31 hysteresis observed in α_{top} observations (data not shown) is attributed primarily to a low

1 installation height (0.5 m), but may also be partly due to changing surface conditions. These
2 effects can compromise the accurate representation of illumination and viewing geometries,
3 resulting in reduced albedo estimates at high SZAs (Kuhn, 1974; Wang et al., 2012; Dumont
4 et al., 2012). As such, Top Met Station measurements, and α_{base} at SZAs greater than 70° ,
5 were excluded for most analyses. Despite its limitations, α_{top} were used for α_{ASD} comparison
6 described below.

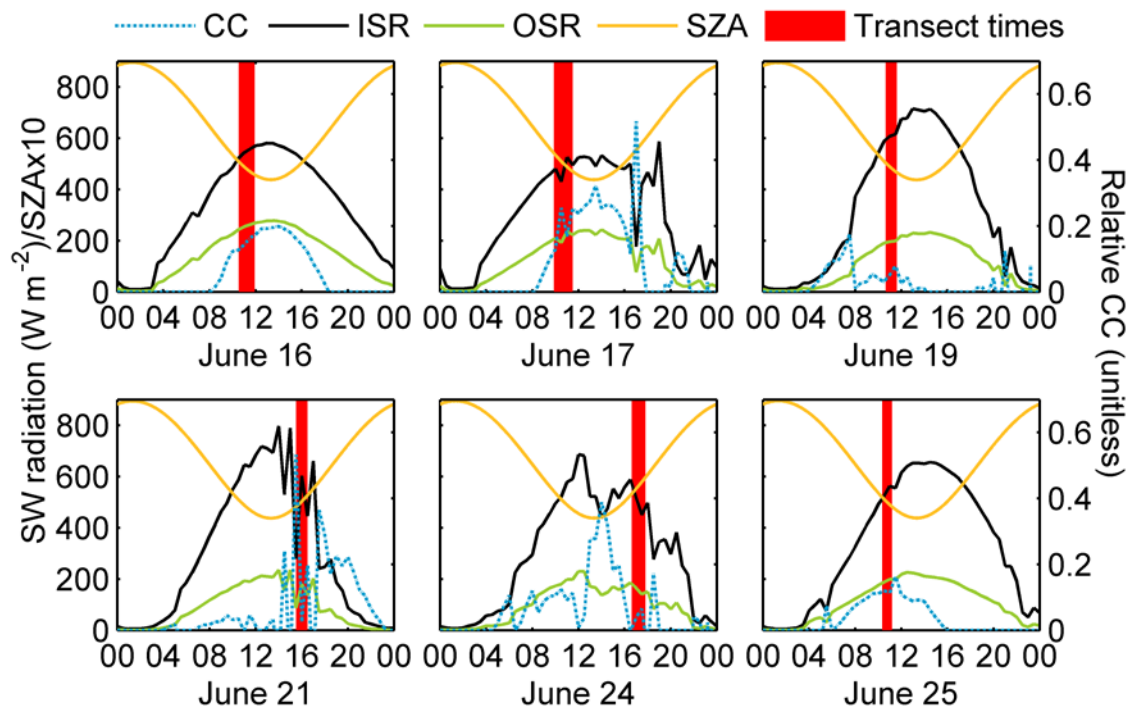
7 High range of CC variability, instead of consistently high CC, was found to be
8 responsible for saturating α_{ASD} readings on 17, 21, and 24 June (Fig. A2b). Continuous
9 recalibration of the ASD instrument on 17 and 24 June was inadequate to overcome variable
10 lighting conditions resulting in saturated α_{ASD} readings (> 1). During 21 June, α_{ASD} data did
11 not saturate despite variable sky conditions (0.01-0.52 CC range). Variable cloud conditions
12 on 17, 21, and 24 of June effectively increased the amount of downwelling longwave
13 radiation relative to shortwave radiation available at the surface, of which, the net effect
14 results in a larger portion of solar radiation available to be reflected by the ice surface
15 (Grenfell and Perovich, 2004; Román et al., 2010; Wang et al., 2012). This can translate to an
16 increase in spectral albedo estimates by ~ 0.06 over active melting ice surfaces (Grenfell and
17 Perovich, 2004).

18 By removing the majority of shortcomings and uncertainties identified in transect
19 radiative and surface conditions, a high-quality albedo dataset was produced. Optimal SZA,
20 CC, and radiative conditions were observed for 16, 19 and 25 June. α_{ASD} data collected on 17,
21 21, and 24 June were identified as low-quality based on their dependence on SZA, CC
22 variability, and issues with albedo saturation, and subsequently removed from further analysis
23 (Fig. A2). The first and last high-quality α_{ASD} measurements closest to the AWSs were
24 compared and reveal that they agree reasonably well with α_{base} and α_{top} data (Fig. A3). As
25 much as 62% of α_{ASD} variance is explained by α_{base} and α_{top} , and the linear regression model
26 slope between the two datasets is close to one ($\alpha_{\text{ASD}} = 0.26\alpha_{\text{MET}} + 0.41$, where α_{MET} is α_{base}
27 and α_{top} combined). The discrepancy is likely due to differences in exact sample locations and
28 instrumentation. Table 1 provides summary statistics related to high-quality α_{ASD} and transect
29 conditions.

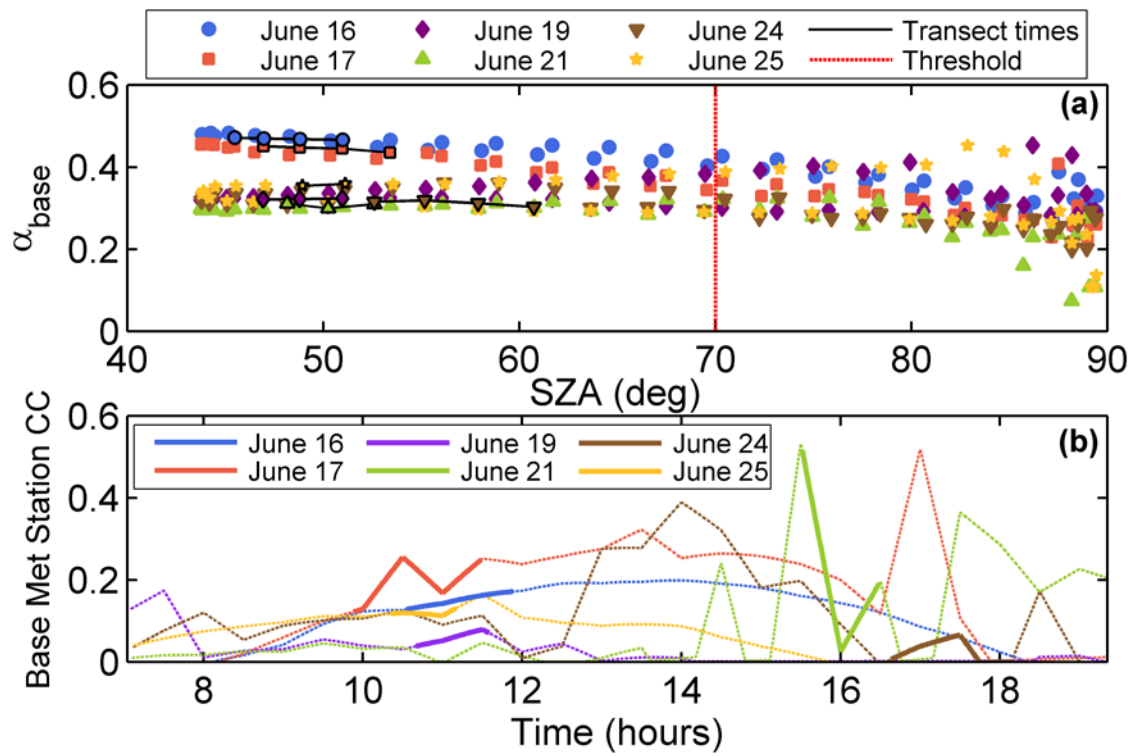
30

1 7.3 Installation of meteorological stations

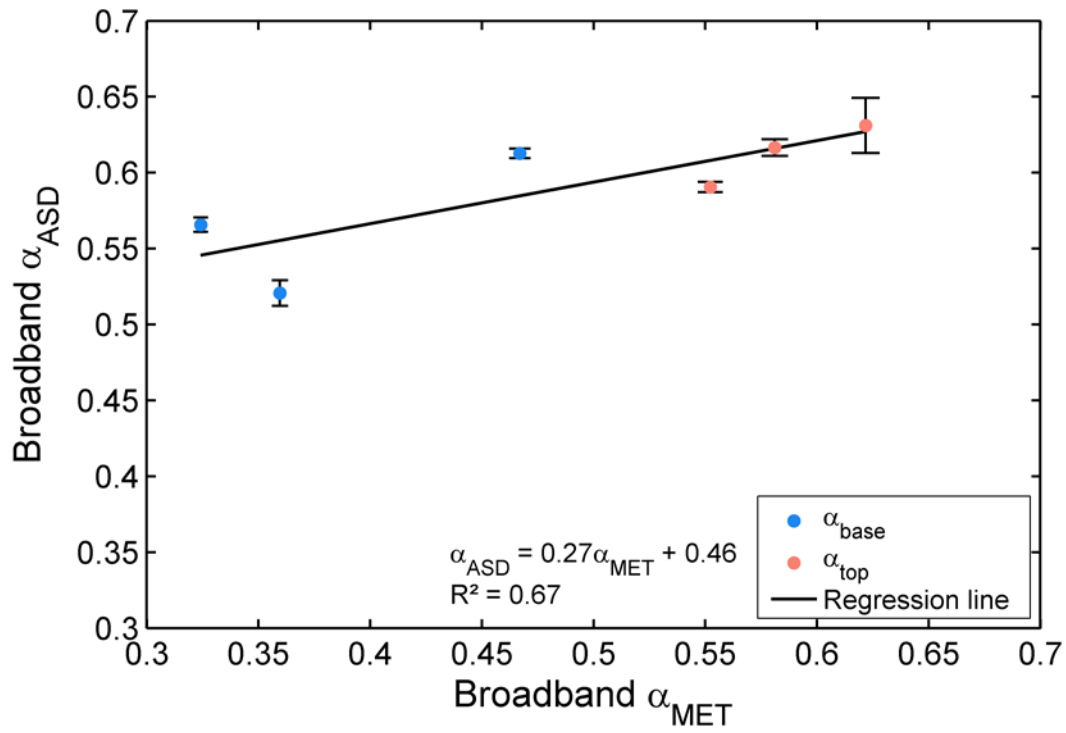
2 The Top Met Station was installed upon a homogenous clean ice surface, and the Base Met
3 Station was installed above a heterogeneous surface of mixed clean and dirty ice. Both
4 stations measured solar radiation fluxes every 0.5 h at 300-1100 nm, using S-LIB-M003
5 silicon pyranometers and a U30 data logger (Table 1A; $\pm 5\%$ or 10 W m^{-2} precision; Onset
6 Computer Corp., 2010) from 8-26 June. Sensors were attached to a pole drilled into the ice at
7 1.5 m above the surface, and were kept relatively constant at this height, but occasionally
8 tilted off-level. After a period of heavy melting, the Top Met Station was re-drilled and
9 installed at 0.5 m height and remained at this height as melting seized. A very large hysteresis
10 in α_{top} as a function of SZA suggests that the low installation height resulted in α_{top} errors due
11 to a disproportionately large influence of surface roughness on its measurements. Despite not
12 having observed tilt information for the AWSs, we use a theoretical tilt (for Fig. A2b) in Van
13 den Broeke et al. (2004) to provide a reasonable uncertainty range. Assuming a tilt of 1° on 18
14 January at Kohnen station, Antarctica (75°S , 0°) is associated with $\sim 15 \text{ W m}^{-2}$ offset in net
15 shortwave at noon local time. This is associated with an absolute error of 5% with a tilt of 1° .
16 Here, we assume double the uncertainty ($\pm 10\%$).
17



1
 2 Figure A1. Radiative conditions during transect dates at the Base Met Station, including
 3 incoming solar radiation (ISR, black line), outgoing solar radiation (OSR, green line; left y-
 4 axis), modeled relative cloud cover (CC, blue stippled line; right y-axis), and solar zenith
 5 angles (SZA, yellow line right axis). Red shaded regions show α_{ASD} data collection times.



1
 2 Figure A2. Half hourly broadband α_{base} (a) measurements as a function of SZA. Symbols and
 3 colors correspond to transect dates. Transect times correspond to the black line. A SZA
 4 threshold at 70° is represented by the red stipple line. (b) Relative CC determined at α_{base} as a
 5 function of time during transect dates. Symbols and colors correspond to transect dates.
 6 Transect times correspond to bold lines.
 7



1
 2 Figure A3. Broadband α_{base} (blue dots) and α_{top} (pink dots) vs. α_{ASD} and α_{MET} (i.e., both α_{base}
 3 and α_{top}) measurements fitted to a linear regression equation ($R^2 = 0.67$). The value of α_{ASD}
 4 error is based on the standard deviation of individual α_{ASD} measurements.

5

1 Table 1A. Meteorological station sites and associated variables.

Site	Latitude	Longitude	Elevation (m)	Start Date	End Date
Base Met Station	67.151629	50.027993	511.3	8-Jun	26-Jun
Top Met Station	67.146857	50.001186	586.0	14-Jun	26-Jun

2