Quantifying volumetric scattering bias in ICESat-2 and Operation IceBridge altimetry over snow-covered surfaces

Fair Zachary¹, Flanner Mark², Neumann Thomas A³, Vuyovich Carrie³, Smith Benjamin E.⁴, and Schneider Adam Michael⁵

¹Goddard Space Flight Center ²University of Michigan-Ann Arbor ³NASA Goddard Space Flight Center ⁴Un. Washington ⁵University of California, Irvine

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Abstract

The Ice, Cloud, and Land Elevation Satellite-2 (ICESat-2) mission has collected global surface elevation measurements for over three years. ICESat-2 carries the Advanced Topographic Laser Altimeter (ATLAS) instrument, which emits laser light at 532 nm, and ice and snow absorb weakly at this wavelength. Previous modeling studies found that melting snow could induce significant bias to altimetry signals, but there is no formal assessment on ICESat-2 acquisitions during the Northern Hemisphere melting season. In this work, we performed two case studies over the Greenland Ice Sheet to quantify volumetric scattering in ICESat-2 signals over snow. Elevation data from ICESat-2 was compared to Airborne Topographic Mapper (ATM) data to quantify bias. We used snow optical grain sizes derived from ATM and the Next Generation Airborne Visible/Infrared Imaging Spectrometer (AVIRIS-NG) to attribute altimetry bias to snowpack properties. For the first case study, the mean optical grain sizes were $340\pm65 \ \mu m$ (AVIRIS-NG) and $670\pm420 \ \mu m$ (ATM), which corresponded with a mean altimetry bias of $4.81\pm1.76 \ cm$ in ATM. We observed larger grain sizes for the second case study, with a mean grain size of $910\pm381 \ \mu m$ and biases of $6.42\pm1.77 \ cm$ (ICESat-2) and $9.82\pm0.97 \ cm$ (ATM). Although these altimetry biases are within the accuracy requirements of the ICESat-2 mission, we cannot rule out more significant errors over coarse-grained snow, particularly during the Northern Hemisphere melting season.

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Zachary Fair¹, Mark Flanner², Tom Neumann¹, Carrie Vuyovich¹, Benjamin Smith³, Adam Schneider⁴

6	¹ NASA Goddard Space Flight Center, Greenbelt, MD, USA
7	² Department of Climate and Space Sciences and Engineering, University of Michigan, Ann Arbor, MI,
8	USA
9	³ Applied Physics Laboratory, University of Washington, Seattle, WA, USA
10	⁴ Department of Earth System Science, University of California, Irvine, CA, USA

Key Points:

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12	•	The effects of snow density and optical grain size on ICESat-2 and Operation Ice-
13		Bridge lidar signals were characterized.
14	•	Both altimeters experienced centimeter-level bias that is linked to the optical grain
15		size of snow over the Greenland Ice Sheet.
16	•	A strong agreement between ICESat-2 and modeled bias suggests that retrieved

ATL03 photons are sensitive to subsurface snow properties.

Corresponding author: Zachary Fair, zachary.fair@nasa.gov

18 Abstract

The Ice, Cloud, and Land Elevation Satellite-2 (ICESat-2) mission has collected global 19 surface elevation measurements for over three years. ICESat-2 carries the Advanced To-20 pographic Laser Altimeter (ATLAS) instrument, which emits laser light at 532 nm, and 21 ice and snow absorb weakly at this wavelength. Previous modeling studies found that 22 melting snow could induce significant bias to altimetry signals, but there is no formal 23 assessment on ICESat-2 acquisitions during the Northern Hemisphere melting season. 24 In this work, we performed two case studies over the Greenland Ice Sheet to quantify 25 volumetric scattering in ICESat-2 signals over snow. Elevation data from ICESat-2 was 26 compared to Airborne Topographic Mapper (ATM) data to quantify bias. We used snow 27 optical grain sizes derived from ATM and the Next Generation Airborne Visible/Infrared 28 Imaging Spectrometer (AVIRIS-NG) to attribute altimetry bias to snowpack properties. 29 For the first case study, the mean optical grain sizes were $340\pm65 \ \mu m$ (AVIRIS-NG) and 30 $670\pm420 \ \mu \text{m}$ (ATM), which corresponded with a mean altimetry bias of $4.81\pm1.76 \ \text{cm}$ 31 in ATM. We observed larger grain sizes for the second case study, with a mean grain size 32 of $910\pm381 \ \mu m$ and biases of $6.42\pm1.77 \ cm$ (ICESat-2) and $9.82\pm0.97 \ cm$ (ATM). Al-33 though these altimetry biases are within the accuracy requirements of the ICESat-2 mis-34 sion, we cannot rule out more significant errors over coarse-grained snow, particularly 35 during the Northern Hemisphere melting season. 36

³⁷ Plain Language Summary

The Ice, Cloud, and Land Elevation Satellite-2 (ICESat-2) mission has been used 38 to measure changes in land ice, vegetation cover, and sea ice, and there is growing in-30 terest to use ICESat-2 for snow science. ICESat-2 uses a lidar that operates at a green 40 wavelength, through which it estimates the elevation of Earth's surface based on the time 41 it takes for the laser to travel between the satellite and the surface. Snow weakly absorbs 42 green light and can increase the travel time of the laser, which would introduce errors 43 in elevation measurements. In this study, we used ICESat-2 and airborne lidar and spec-44 trometer data to (i) identify errors in the ICESat-2 data and (ii) link the errors to snow 45 properties over the Greenland Ice Sheet. We found that ICESat-2 errors have a link with 46 snow grain size and density. The errors are within the accuracy requirements of ICESat-47 2, but more significant errors may be possible during the Northern Hemisphere melting 48 season. 49

50 1 Introduction

The Ice, Cloud, and Land Elevation Satellite-2 (ICESat-2) was launched in Septem-51 ber 2018 to perform measurements of surface height over glaciers and ice sheets (Markus 52 et al., 2017). Since then, ICESat-2 data products have been developed to estimate the 53 surface height of land ice, vegetation canopies, and sea ice (Smith et al., 2019; Kwok et 54 al., 2019; Neuenschwander & Pitts, 2019). The sole onboard instrument, the Advanced 55 Topographic Laser Altimeter System (ATLAS), emits laser light at 532 nm and produces 56 high spatial resolution data (12 m footprint diameter and 10 kHz pulse repetition fre-57 quency) and a required accuracy of 0.4 cm yr^{-1} for ice sheet annual elevation change (Markus 58 et al., 2017). Recent comparisons with ground-based data have shown that the ATLAS 59 laser has a measured accuracy of <4 cm over ice sheet interiors (Brunt et al., 2021). 60

There is growing interest in the snow science community to use ICESat-2 to derive snow depth over remote locations through comparison of snow-on and snow-off elevations. A complication with past snow studies is that forested and mountainous environments have significant seasonal snow, yet these regions are subject to elevation uncertainty in ground-based and airborne surveys. In recent years, digital elevation models (DEMs) from lidar have become common data sets for snow depth estimates (Deems et al., 2013), though current lidar acquisitions are limited to airborne and ground-based

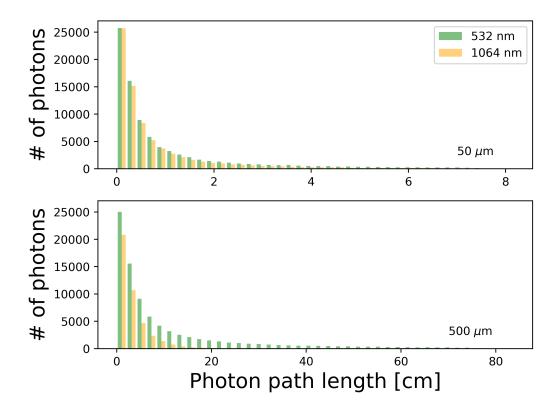


Figure 1. Histograms of 100000 photon path lengths traveled through a simulated, semiinfinite snowpack. The snow is assumed to be clean (i.e., no impurities) with ρ_s =400 kg m⁻³ and r_{eff} =50 μ m (top), r_{eff} =500 μ m (bottom). The range of path lengths differs between histograms to highlight the effects of volumetric scattering.

surveys. There are currently no documented efforts to measure deep snow in forests and
 mountains using spaceborne instrumentation (Bormann et al., 2018), so ICESat-2 has
 the potential to support snow studies through inter-seasonal measurements of terrain height.

A possible complication is that a laser shot from ICESat-2 may experience mul-71 tiple scattering events within a snow layer before returning to the detector (Perovich, 72 2007) due to weak absorption of visible light (Warren & Wiscombe, 1980). This phenomenon, 73 which we refer to as "volumetric scattering", is greatest in clean, coarse-grained snow, 74 where the increased path length between individual snow particles will introduce a de-75 lay time in the returned laser pulse. The optical grain size of snow, a quantity used to 76 represent snow grains as simplified shapes, is strongly linked to photon path length. Fig-77 ure 1 shows the modeled effects of volumetric scattering by spherical snow particles at 78 common lidar wavelengths. The path length traveled by photons within a snowpack is 79 similar between 532 nm and 1064 nm at small optical grain sizes, but the path lengths 80 at 532 nm increase with grain size. Near-infrared snow reflectance is low in snow with 81 an optical grain size of 500 μ m, so fewer 1064 nm photons propagate through the snow-82 pack. Snow impurities may attenuate the ICESat-2 signal and reduce volumetric scat-83 tering bias, though impurity content has significant variability at small spatial and tem-84 poral scales (Flanner et al., 2007; Skiles et al., 2017). 85

Previous studies have assessed the potential impacts of snow on lidar measurements at 532 nm. Harding et al. (2011) found that return waveforms from an airborne 532 nm lidar experienced significant pulse broadening over snow, resulting in range biases on the

Instrument	Dataset	Wavelengths	Case Study	Application
ICESat-2	ATL03	532 nm	CS2	Altimetry
ATM	ILNSAW1B ILNIRW1B	$\begin{array}{c} 532 \ \mathrm{nm} \\ 1064 \ \mathrm{nm} \end{array}$	CS1, CS2	Altimetry, Snow grain size
AVIRIS-NG	L2 Reflectance	$380\text{-}2510~\mathrm{nm}$	CS1	Snow grain size

 Table 1.
 Data Summary

order of a few centimeters. A modeling study by Kerekes et al. (2012) found that centimeter-89 level biases occurred most frequently when the optical grain size of snow was 500 μ m or 90 more, and the amplitude of received waveforms was low relative to fine-grained snow re-91 turns. Smith et al. (2018) simulated ICESat-2 measurements over a snow-covered sur-92 face using a suite of surface height estimation techniques. The authors concluded that 93 elevation biases may exceed 0.45 m over regions of clean, coarse-grained snow if the cur-94 rent ICESat-2 height estimation scheme is used for retrieved photons, though biases could 95 decrease if other techniques are used or if snow impurities (i.e., black carbon) are present. 96

At the time of writing, the ICESat-2 mission has collected over 3 years of altime-97 try measurements over high-latitude regions, yet there have been no documented efforts 98 to quantify volumetric scattering biases over snow. As part of an extensive validation 99 effort, Operation IceBridge (OIB) launched a series of flights over Greenland late in the 100 2019 melt season. The flights collected elevation measurements using the Airborne To-101 pographic Mapper (ATM), a lidar that operated at 532 nm and 1064 nm during the 2019 102 flights. Near-coincident flights were performed with the Next Generation Airborne Vis-103 ible/Infrared Imaging Spectrometer (AVIRIS-NG) to retrieve hyperspectral reflectance 104 and the optical grain size of snow. 105

Here we perform two case studies to assess bias in ICESat-2 and ATM altimetry 106 measurements due to volumetric scattering in snow. Optical grain sizes derived from AVIRIS-107 NG reflectance data and ATM waveforms serve as input to a Monte Carlo ray tracing 108 model to simulate altimetry bias over the Greenland ablation zone. In parallel, surface 109 heights derived from ICESat-2 and the ATM 532 nm beam are compared to the ATM 110 1064 nm beam, which we assume also measures the surface, to estimate observed bias. 111 The findings presented here will serve as a benchmark for an ICESat-2 bias correction 112 algorithm over snow-covered surfaces. 113

¹¹⁴ 2 Data Description and Case Study Locations

2.1 Case Study Locations

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We performed two case studies over the Greenland ablation zone, which we refer to as CS1 and CS2 in the remainder of the paper. The locations of the study regions are shown in Figure 2, and Table 1 outlines the data sets used for each.

The first case study (CS1), performed for September 6, 2019, is located at coordinates 75.316-75.438°N, 56.528-56.778°W. This date and location correspond with a significant overlap between ATM and AVIRIS-NG flights, with ~40 km of OIB flight data overlapping with AVIRIS-NG surveys. The ice surface features many crevasses and refreezing supraglacial lakes during this time of year, several of which were observed by ATM and AVIRIS-NG. The lakes are characterized by anomalously high optical grain

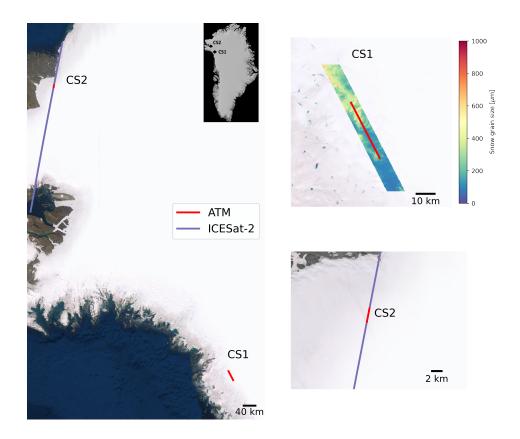


Figure 2. Landsat-8 imagery showing the location of the two case studies over the Greenland Ice Sheet. The red line given for CS1 (top right) is the path flown by both ATM and AVIRIS-NG, and the false color overlay is the snow grain sizes observed by AVIRIS-NG. The red line shown for CS2 (bottom right) represents the region where ATM and the central strong ground track of ICESat-2 (GT2L) intersect.

sizes in the AVIRIS-NG data, whereas the ATM beams exhibit a greater degree of noise over crevassed ice. These features are small relative to the size of the instrument swaths, so we applied a moving mean filter with a window size of 500 samples (30 m) to mitigate noise. There were no significant overlaps with ICESat-2 data over this region, so we used CS1 as a proof-of-concept to demonstrate green light penetration in snow.

The second case study (CS2) was performed for September 4, 2019 at coordinates 130 78.783-78.807°N, 66.066-66.090°W. Across this region, ATM followed an ICESat-2 over-131 pass for 20 minutes, closely matching with the central ICESat-2 ground tracks (GT2L/GT2R). 132 133 The ATM products used in this study (Section 2.2.1) have a swath width smaller than the distance between GT2L and GT2R, so the aircraft overlapped the two beams in al-134 ternating segments. Local topography can have a first-order impact on our analysis (Wang 135 et al., 2019), where ~ 10 m of separation between ATM and an ICESat-2 beam may lead 136 to significant differences in elevation estimates, particularly over rough terrain. We there-137 fore limited our analysis to regions where ATM footprints were within the ICESat-2 foot-138 print, or a maximum distance of 12 m (Magruder et al., 2021). This restriction minimized 139 errors due to data separation, but it also limited the analysis to a 2.5 km region over the 140 ice sheet interior (CS2 in Figure 2, green line). 141

142 2.2 Altimetry Data

143 **2.2.1** ATM

ATM is an altimetric lidar that has been used for high-latitude elevation measure-144 ments since 1993 (Brock et al., 2002; Krabill et al., 2002). In recent years, it has been 145 used to validate ICES at-2 surface height estimates over sea ice and the $88^{\circ}S$ transect of 146 Antarctica (Kwok et al., 2020; Brunt et al., 2021) as part of Operation IceBridge. The 147 instrument suite is composed of two laser altimeters that feature off-nadir scan angles 148 2.5° and 15° , which correspond to swath widths of 40 m and 245 m at typical flight al-149 titudes. The 2.5° "narrow swath" altimeter is a dual-color laser that operates at 532 nm 150 (green) and 1064 nm (near-infrared) simultaneously. The near-infrared laser has a foot-151 print diameter of 0.91 m, or 40% larger than the 532 nm beam (0.64 m). 152

Here, we used two Level-1B Narrow-Swath data products: the Elevation and Re-153 turn Strength with Waveforms (ILNSAW1B) and the Near-Infrared Waveforms (ILNIR1B). 154 Both data products include information about the transmitted and received waveforms, 155 including the amplitude and width of each waveform and the corresponding aircraft-surface 156 range estimates. The ranges are derived using the centroid (median) time of the trans-157 mitted and received pulses, and these ranges are compared to the WGS84 ellipsoid to 158 estimate surface elevation (Brock et al., 2002). Brunt, Neumann, and Larsen (2019) found 159 that the 532 nm laser agrees well with ground-based measurements over the $88^{\circ}S$ tran-160 sect of Antarctica, with a mean uncertainty of ~ 8.5 cm. 161

162 2.2.2 ICESat-2

ICESat-2 is a polar orbiting satellite with an operational altitude of 500 km and a 91-day repeat cycle. A single ATLAS 532 nm laser pulse is split into six beams that are configured in pairs, with a 90 m separation between beams within a pair and a 3.3 km separation between pairs (Neumann et al., 2019). The beams are named according to their ground track from left to right: GT1L/R, GT2L/R, and GT3L/R. At its operational altitude, ICESat-2 has a surface footprint of 12 m, which allows for significant overlap with its 0.7 m along-track resolution.

The ATLAS product used here is the ATL03 Global Geolocated Photon Data V005 (Neumann et al., 2020), which consists of latitude, longitude, and surface elevation for received photons. Each tagged photon is also classified as either signal or solar background, based on a statistical confidence algorithm (Neumann et al., 2019). Although noisier than

other ICESat-2 data sets, we selected ATL03 to better capture the scattering experienced 174 by individual photons. The number of photon returns was high over the Greenland Ice 175 Sheet, so we only considered high confidence photons. The OIB flights over CS2 over-176 lapped with the central beams in alternating segments. We selected a 2 km extent where 177 OIB flew inline with the central strong beam (Figure 2, bottom right) to ensure that 178 we received a high rate of photons across the study area. Comparisons with ground-based 179 measurements over the $88^{\circ}S$ transect show a mean uncertainty of ~ 4 cm for ATL03 (Brunt 180 et al., 2021). 181

182 We briefly consider the impacts of volumetric scattering on ATL06, the Land Ice Height Product V005 (Smith et al., 2019). The ATL06 algorithm aggregates geolocated 183 ATL03 photons into 40 m segments, from which a mean surface height is derived. ATL06 184 segment values are posted every 20 m, yielding a 50% overlap between consecutive seg-185 ments. Brunt, Neumann, and Smith (2019) found that ATL03 photon-based heights are 186 generally a few centimeters higher than those from ATL06 segments due to differences 187 in the processing algorithms. We only consider high-confidence ATL03 photons with ATM 188 data in close proximity, so additional errors relative to ATL06 (which considers photons 189 of low, medium, and high confidence and corrects for several instrument effects) are ex-190 pected. 191

¹⁹² 2.3 Snow Grain Size Data

¹⁹³ 2.3.1 AVIRIS-NG

The Next Generation Airborne Visible/Infrared Imaging Spectrometer is an air-194 borne hyperspectral imager that has been used to retrieve surface radiances since 1986 195 (Gao et al., 1993; Green et al., 1998). Originally operating at 10 nm spectral resolution, 196 the instrument now observes the Earth's surface between 380 and 2510 nm at a spec-197 tral resolution of 5 nm. Surface reflectances are derived from the radiances by applying 198 an atmospheric correction and orthorectification. Reflectances from AVIRIS-NG gener-199 ally have an accuracy within 2-5% (Thompson et al., 2019). The spectrometer has been 200 used for a suite of applications since its inception, including vegetation mapping, trace 201 gas identification, and retrieval of snow grain size (Kokaly et al., 2003; Thorpe et al., 2016; 202 Nolin & Dozier, 2000). 203

We used AVIRIS-NG reflectances for CS1 to derive the optical grain size of snow for comparison against the altimetry data. An inversion algorithm derived by (Nolin & 205 Dozier, 2000) was used to relate changes in the ice absorption feature at 1.03 μ m to changes 206 in optical grain size. In short, the algorithm compares AVIRIS-NG reflectances to those 207 derived from a radiative transfer model to show that optical grain size increases as the 208 near-infrared snow reflectance decreases. The snow is assumed to be composed of spher-209 ical ice particles, and snow impurities are assumed to have a negligible impact on the 210 retrievals. Although impurity content is assumed to be negligible over the regions of in-211 terest, we recognize that impurities such as ice algae or cryoconite may impact retrievals 212 over the Greenland ablation zone (Cook et al., 2020). Optical grain sizes derived from 213 this algorithm have a stated uncertainty of 50 μ m (Nolin & Dozier, 2000; Fair et al., 2022). 214

2.3.2 ATM

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AVIRIS-NG grain sizes were unavailable over CS2, so we instead used an algorithm that infers grain size from ATM data. Over snow, subsurface scattering affects green ATM waveforms by reducing the maximum amplitude and increasing the width of the received pulse (Smith et al., 2018). The algorithm exploits this occurrence to compare waveforms from the ILNSAW1B product (Section 2.2.1) to an idealized waveform with no subsurface scattering. The grain size is then estimated based on differences in amplitude and pulse width. The model waveform is derived assuming that the snow has no impurities and has a density of 400 kg m⁻³. While the former assumption is reasonable over this
region of the Greenland Ice Sheet (Flanner et al., 2007), the assumed snow density is higher
than typical values (Fausto et al., 2018; Schaller et al., 2016). The algorithm is also more
sensitive to subsurface snow properties, so grain sizes derived by ATM are generally higher
than those found using AVIRIS-NG (Section 4).

228 3 Methods

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3.1 Monte Carlo Modeling

We first estimated altimetry bias using a combination of optical grain size data and 230 Monte Carlo modeling. The model fires photons into a simulated semi-infinite snowpack 231 and records their total path length until they are absorbed or leave the medium (Schneider 232 et al., 2019). The snowpack has user-prescribed optical grain sizes and density, and it 233 is configured to have spherical ice particles with negligible impurity content. The model 234 has additional inputs for solar zenith and particle surface roughness, but we assumed that 235 (i) the snow particles were smooth and (ii) the solar zenith angle was equal to the mean 236 solar geometry observed at the time of flight for ATM. The snowpack was also assumed 237 to have a uniform optical grain size. 238

The Monte Carlo model was used to benchmark lidar delay time within a snow-239 pack. Simulations were conducted for different permutations of photon wavelength, snow 240 density, and optical grain size. The simulations launched 10^5 photons into a snowpack 241 at wavelengths 532 nm and 1064 nm to emulate the ATM dual-colored laser interact-242 ing with a snow-covered surface. We performed these simulations for grain sizes 50-1500 243 μm at 50 μm resolution. We then applied spline curve fitting to improve the resolution 244 to 1 μ m. Snow density was configured to be consistent with observations by Fausto et 245 al. (2018), i.e. 315 kg m⁻³. We obtained the path length traveled by the photons that 246 escaped from the top of the snowpack, and for each wavelength the median path length 247 of escaped photons was calculated to replicate the reference photon technique employed 248 by ICESat-2 (Neumann et al., 2020). The median path length was treated as the sur-249 face height offset relative to an unbiased measurement. If we treat the 532 nm path lengths 250 as biased surface height measurements (L_{532}) and the 1064 nm path lengths as ideal mea-251 surements (L_{1064}) , then the modeled bias estimate ΔL is simply: 252

$\Delta L = L_{532} - L_{1064} \tag{(1)}$	(1))
— <i>E E</i> 352 <i>E</i> 1004	<u>۲</u> .	J

In this configuration, a positive ΔL implies that 532 nm photons traveled a greater 254 distance within the snowpack, which would suggest a negative bias in the final surface 255 height estimate. Conversely, the 1064 nm path length (surface height) will be biased high 256 (low) if there is a negative ΔL . Modeled biases were placed into lookup tables depend-257 ing on the density used in the simulation. The result was six lookup tables that each had 258 1500 bias estimates as a function of optical grain size, given $\rho_s = [100, 200, 300, 315, 400, 500]$ 259 kg m⁻³. The biases in these lookup tables are the errors we would expect if grain size were 260 the only factor impacting ICESat-2 and ATM observations. 261

3.2 Observed Bias

We look for bias in the altimetry data by comparing 532 nm elevation estimates with those from the ATM 1064 nm beam. The ATM beams periodically did not record laser pulses, so we applied a co-registration algorithm to match data samples from both beams. Because the beams fire simultaneously, the algorithm co-registers shots between beams by using the time stamps recorded for each laser pulse. The co-registered shots were then filtered to match with the central strong beam of ICESat-2 (see Section 2.3). For the first case study, rough terrain caused 17-21% signal loss in the ATM beams. Smoother ice was present for the second case study, so signal loss was lower (4-5%). Co-registered
 ICESat-2 and ATM elevations were used to approximate observed bias using Equation
 1.

The accuracy of the ATM beams relative to each other has not been documented, 273 so we performed a bias assessment of the two beams in the absence of snow. Operation 274 IceBridge was flown from Qaanaaq Air Base (formerly Thule) in September 2019, and 275 each flight included an overpass of the aircraft ramp for calibration purposes. We selected 276 the ramp overpass from September 6 (Track 1730, Figure 4a), and Equation 1 was used 277 278 to estimate bias over a dark, flat surface. Figure 4b shows the differences between the green and NIR beams. Comparisons over the ramp consistently feature negative bias, 279 implying that NIR ranges surface heights are lower than those of the green beam. The 280 bias has a nearly Gaussian distribution between -8 cm and 0 cm, with a slightly larger 281 distribution toward less negative values (Figure 4c). Repeat flights over the ramp on 282 different dates yielded similar results (Studinger, 2022). The median bias was -3.85 cm, 283 and we applied this value as a correction factor to all comparisons between the ATM beams 284 in the featured case studies. 285

To attribute altimetry bias to optical grain size, we co-registered ICESat-2 and cor-286 rected ATM laser pulses with AVIRIS-NG or ATM grain size estimates. We mapped each 287 segment of grain size data with an estimate of modeled bias by matching grain sizes with 288 the closest values found in each lookup table. In other words, each segment of co-registered 289 data had six modeled bias estimates for each snowpack density given in Section 3.1. The 290 observed biases were compared to matched model biases at these densities. If the ob-291 servations agreed with at least one of the modeled results, then we could conclude that 292 (i) the altimetry biases are linked to the optical grain size of snow and (ii) the bias is con-293 sistent with one of the given snow densities. 294

295 4 Results

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4.1 Case Study 1

Model-derived results of altimetry bias have a strong dependence on snow optical 297 grain size and density, as seen in Figure 3. At smaller grain sizes, bias is less sensitive 298 to changes in snow density, particularly at grain sizes below 400 μ m. Larger grain sizes 299 exhibit greater dependence on snow density, especially when $\rho_s \leq 200$ kg m⁻³. The largest 300 modeled biases occur for $\rho_s = 100 \text{ kg m}^{-3}$, up to a maximum of 37.84 cm at the largest 301 grain sizes. At densities $\rho_s = 315{-}500$ kg m⁻³ biases of 7.55-11.88 cm are more likely to 302 occur. The bias asymptotically approaches zero with decreasing grain size at all densi-303 ties, implying that little altimetry bias should be expected over fine-grained snow. 304

The AVIRIS-NG optical grain sizes co-registered with ATM are shown in Figure 305 5. The southern reaches of CS1 are characterized by grain sizes of $\leq 200 \ \mu m$ that typ-306 ically increase near crevassed terrain or near melt ponds. In the northern portions of the 307 flight track, grain sizes increase to 300-400 μ m. This increase corresponds with a gen-308 eral decrease in surface elevation (Figure 6), with lower elevations implying warmer tem-309 peratures, greater melt, and faster snow metamorphism. Subsurface scattering on the 310 order of 1-10 m is evident throughout the study area (green dots in Figure 6), indicat-311 ing the presence of heavy crevassing. Although ATM waveform-fitted grain sizes exhibit 312 similar trends to those from AVIRIS-NG, the derived values are much larger, with a mean 313 grain size of $653\pm422 \ \mu m$ for ATM and $338\pm65 \ \mu m$ for AVIRIS-NG. ATM waveform-314 fitted grain sizes are derived as a function of received waveform amplitude and width, 315 so we speculate in Section 5.2 that the estimated grain sizes are values obtained from 316 subsurface snow. 317

Figure 7 shows that the small grain sizes at the start of CS1 correspond to negligible model bias and uncertainty due to snow density. In regions with larger grain sizes,

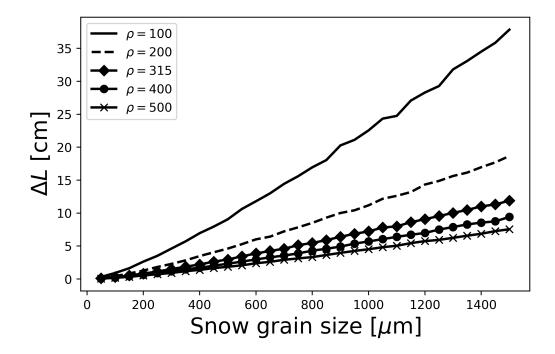


Figure 3. Modeled altimetry bias derived using median path lengths estimated from a Monte Carlo model (Schneider et al., 2019). Bias is given as a function of snow optical grain size and snowpack density. The snow density $\rho_s = 315 \text{ kg m}^{-3}$ is used to represent the average snow density reported over Greenland by Fausto et al. (2018) ($\rho_s = 315 \text{ kg m}^{-3}$). The simulated bias at $\rho_s = 300 \text{ kg m}^{-3}$ is nearly identical to that of $\rho_s = 315 \text{ kg m}^{-3}$ and was thus omitted.

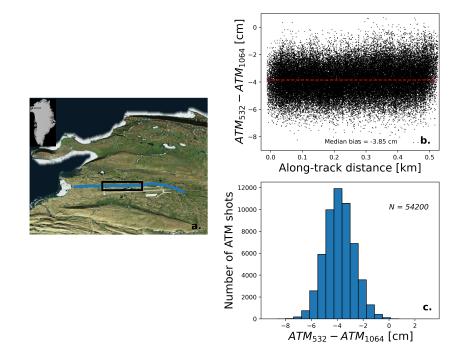


Figure 4. (a) Location of the aircraft ramp used to assess the bias between the ATM 532 nm and 1064 nm beams. The black box highlights the ramp overpass. (b) Along-track scatter plot of the 532 nm bias relative to 1064 nm measurements. Negative values indicate lower surface heights measured by the 1064 nm beam. The red dashed line depicts the median bias observed across the overpass. (c) Bias distribution between the ATM beams, as derived from ramp overpass measurements.

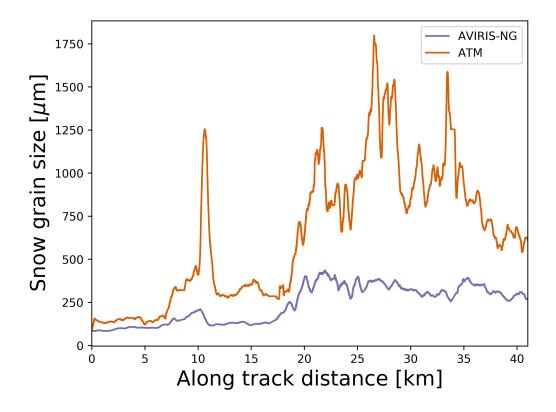


Figure 5. Snow optical grain sizes derived along-track from AVIRIS-NG (purple) and ATM waveforms (orange) for CS1.

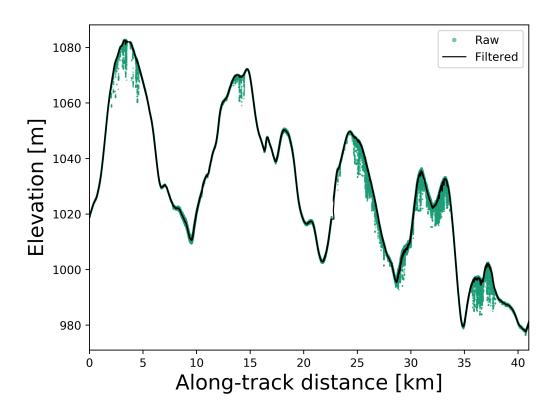


Figure 6. Surface heights for CS1, as given by ILNSAWL1B ("Raw", green dots). A moving mean filter (black line) was applied to remove features significantly below the surface.

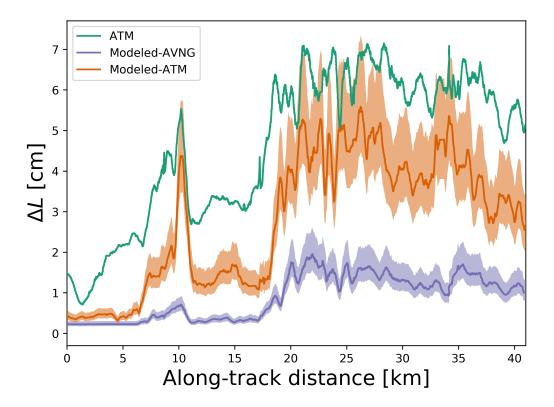


Figure 7. Observed ATM green-NIR range differences (green) as compared to modeled estimates using optical grain sizes from AVIRIS-NG reflectances (purple) and ATM waveform fitting (orange). The solid lines for the modeled estimates represents $\rho_s = 400 \text{ kg m}^{-3}$, whereas the shading is the uncertainty to due changes in snow density, given $\rho_s = 315-500 \text{ kg m}^{-3}$.

the bias increases, with the full extent dependent on the optical grain sizes used as model 320 input. The lower grain sizes of AVIRIS-NG correspond with a maximum bias of 1.95 ± 0.39 321 cm, whereas bias peaks at 5.58 ± 1.12 cm with ATM waveform-fitted grain sizes. The green-322 NIR path length differences generally show agreement with the model when the ATM 323 grain size algorithm is used, with the ATM-derived model estimates accounting for 71%324 of the observed bias. The best agreement between the model and the observations is in 325 regions of larger optical grain size. The model underestimates bias relative to the ob-326 servations at smaller grain sizes, suggesting that (a) the observations may agree better 327 with snow densities of $\rho_s = 200-250$ kg m⁻³, or (b) other factors are influencing the bias 328 in this region. 329

330 4.2 Case Study 2

The optical grain sizes and along-track surface heights for CS2 are given in Fig-331 ures 8 and 9. The region features gently sloped terrain that decreases in surface height 332 over the track. At the large scale, co-registered ATM 532 nm and ICESat-2 data show 333 general agreement in surface height trends. The ICESat-2 data has slighly larger vari-334 ability among individual photons which may be attributed to the inherent noisiness of 335 the ATL03 product. The mean bias between the ATM 532 nm heights and ICESat-2 heights 336 is ~ 1 cm, with a precision of 10 cm. The grain sizes are comparable to those derived over 337 CS1, with a mean value of $910\pm381 \ \mu m$. The variability in grain size is larger than in 338 CS1, though this is likely due to the smaller spatial scale. A surface feature at 1.3 km 339

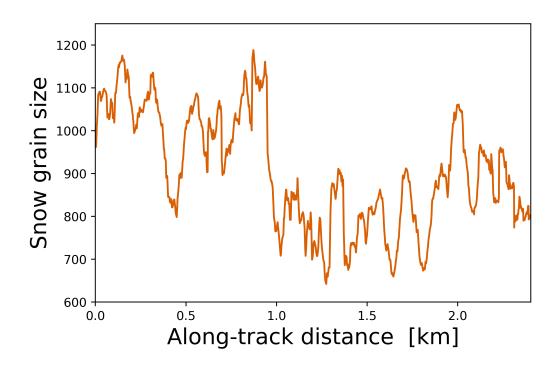


Figure 8. Snow optical grain sizes for CS2, as derived from 532 nm ATM waveforms.

along-track corresponds with a localized increase in grain size and greater disagreement
 between the altimeters. The ICESat-2 observations have a greater spread at this feature
 that is not replicated in ATM, suggesting the presence of a shallow melt pond or crevasse
 that was undetected by ATM pulses co-registered with ICESat-2.

Despite the variability in optical grain size, the modeled bias over CS2 peaks at 344 6.7 cm at the start of the track before decreasing to ~ 5.5 cm as grain size decreases. The 345 mean modeled bias across the region is 4.93 ± 1.89 cm. Between the two altimeters, ICESat-346 2 bias trends show the closest agreement to modeled estimates. The ICESat-2 biases peak 347 at 17.64 cm at the start of the track before reducing to 6.42 ± 1.77 cm. The ATM green-348 NIR range differences show weaker agreement with the model than in CS_1 , with a mean 349 bias of 9.82 ± 0.97 cm. However, ATM trends resemble those seen in the model, imply-350 ing that optical grain size still has an influence on ATM signals. Overall, the model ac-351 counts for 66% (ATM) and 95% (ICESat-2) of the observed bias. The rough surface fea-352 ture at 1.3 km produces the greatest disagreement between the altimeters and the model, 353 with ICESat-2 showing the greatest agreement with the 1064 nm beam and ATM 532 354 nm having the weakest. We speculate that ATM 532 nm and ICESat-2 illuminated dif-355 ferent components of the surface feature, as the model results suggest that optical grain 356 size contributed little to the biases over this region. 357

358 5 Discussion

359

5.1 Modeled and Observed Bias

The relationship between optical grain size and ICESat-2 bias is a function of the path length of signal photons incident upon a snow surface. When photons interact with snow, there are two potential outcomes: reflection after subsurface scattering or absorption by the snowpack. The first outcome is more frequent for ICESat-2 and the ATM 532 nm beam over coarse-grained snow, and it is responsible for the largest biases over

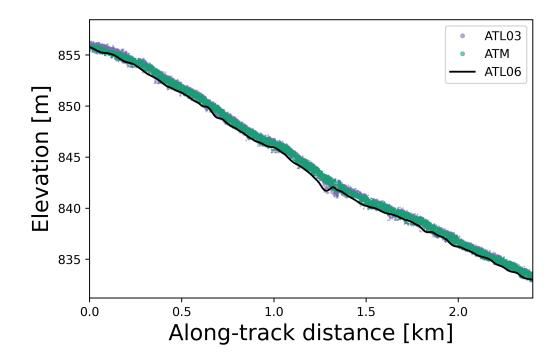


Figure 9. Along-track surface heights from ATL03 (light blue), the ATM 532 nm beam (green), and ATL06 (black). The plot only includes spot measurements where ATM was within 12 m of ICESat-2 footprints.

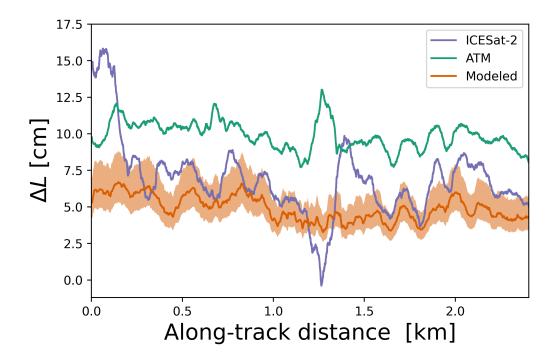


Figure 10. Observed ICESat-2 (purple) and ATM (green) elevation bias, compared to modeled biases using ATM grain sizes at $\rho_s = 315-500$ kg m⁻³ (orange).

the Greenland Ice Sheet. The second outcome occurs with the ATM 1064 nm beam when grain size is large, due to the low reflectance of aged snow in this part of the electromagnetic spectrum. Snow absorption reduces the occurrence of multiple scattering in NIR signals, so bias will be low over snow surfaces. The presence of snow impurities, such as black carbon and ice algae, may increase the probability of absorption for 532 nm signals and reduce bias (Smith et al., 2018), but further research is needed to confirm this hypothesis.

Model-derived results of altimetry bias have a strong dependence on optical grain 372 373 size and density. As seen in Figure 3, the dependence on snow density is minor at small grain sizes, particularly at snow densities expected over Greenland. The modeled bias 374 asymptotically approaches zero when grain size is small, which implies that the bias be-375 tween two altimeters should be negligible when excluding other factors. Although dis-376 crepancies in bias become more evident between snow densities at larger grain sizes, we 377 note that lower densities (i.e., $\rho_s = 100{\text{-}}200 \text{ kg m}^{-3}$) only occur for fresh snow. The case 378 studies presented here take place at the end of the Greenland melt season, where snow 379 densities of $\rho_s = 300-500$ kg m⁻³ and larger grain sizes are more common, as supported 380 by the observations of Schaller et al. (2016) and Fausto et al. (2018). 381

382 5.2 Sources of Uncertainty

In CS1, we found that optical grain size retrievals differed substantially in mag-383 nitude between AVIRIS-NG and ATM. The exact cause may be related to the respec-384 tive retrieval methods. Grain sizes from AVIRIS-NG are derived from the near-infrared 385 surface reflectance of a location, so they are more sensitive to changes in surface snow 386 properties. The ATM algorithm estimates grain size through received waveform pulses. 387 which contain backscatter from below the snow surface if volumetric scattering is sig-388 nificant. The differences between the two algorithms are evident in Figure 5, which can 389 be separated into fine-grained and coarse-grained regions, which correspond to sections 390 0-17 km and 17-42 km along the CS1 transect, respectively. In the fine-grained region, 391 the pore spacing between snow grains is small, so ATM beam penetration beyond the 392 surface layer will be minimal, and the derived optical grain size will be smaller as a con-393 sequence. Volumetric scattering becomes more significant at larger optical grain sizes, 394 as is reflected in the coarse-grained region of Figure 5. If the surface grain sizes are large, 395 then the ATM beam is more likely to penetrate the subsurface, where grain sizes may 396 be larger than those observed by AVIRIS-NG at the surface. Although ICESat-2 was not 397 considered in CS1, the strong agreement between ICESat-2 and modeled bias in Fig-398 ure 10 indicates that retrieved ATL03 photon path lengths are more sensitive to sub-399 surface grain sizes over aged or melting snow. 400

The data sets used here each have different approaches to estimating surface height, 401 which may influence the biases given in Figures 7 and 10. Both ATM beams estimate 402 surface height from the centroid of received waveforms, and a signal strength threshold 403 is applied to filter noise. Although most background noise is removed with the thresh-404 old, sufficiently coarse snow or rough terrain may broaden waveforms and shift the cen-405 troid by nanoseconds, or centimeters in height change. ICESat-2 and the Monte Carlo 406 model use similar approaches by estimating the median surface height (ICESat-2) or travel 407 time (model) for photon aggregates. Subsurface scattering increases the distribution of 408 photon delay times, therefore also increasing uncertainty and bias in ATL03 and the model. 409 Thus, the differences between ATM and the model may be partly explained by these dif-410 ferent approaches in signal processing. The two bias estimates show better agreement 411 in the coarse-grained region of Figure 7, which may indicate that the model is neglect-412 ing snowpack features that impact the ATM signal. A model that allows for more com-413 plex scenarios, such as a rough surface layer or layer-dependent optical grain sizes, could 414 help to answer these questions, though we leave the development of such a model to a 415 future study. 416

As noted above, the ATL06 algorithm aggregates ATL03 photons, applies several 417 corrections, and produces 40 m segment heights that are 3 cm lower than metrics based 418 solely on ATL03 photon heights. Consequently, ATL06 biases in the presence of volu-419 metric scattering should be 3 cm larger than the ATL03-based biases reported here. The 420 impact of these biases should be greatest when comparing times of year with relatively 421 little and relatively significant volumetric scattering, for example summer vs. winter sur-422 face heights. However, given the magnitude of seasonal elevation change in the ablation 423 zone of Greenland, it may be difficult to isolate the magnitude of volumetric-scattering-424 based biases from the height change due to seasonal melt and accumulation. 425

5.3 Implications for Snow Studies

This study was performed to assess ICESat-2 measurements over snow. There has 427 been increasing interest in using ICESat-2 to derive spaceborne measurements of snow 428 depth (Bormann et al., 2018). Currently, digital elevation models from lidar are com-429 monly used to estimate snow depth (Deems et al., 2013), though current lidar applica-430 tions are restricted to airborne and ground-based surveys. There is a critical need to mea-431 sure deep snow in forests and mountains using spaceborne instrumentation (Bormann 432 et al., 2018; National Academies of Sciences, Engineering, and Medicine, 2018), and progress 433 is this area is an objective of the NASA-sponsored Snow Experiment (SnowEx). ICESat-2 has the potential to support SnowEx objectives through interseasonal measurements 435 of terrain height. 436

Field campaigns conducted for the SnowEx mission require measurements from the 437 mid-latitude melting season, when the optical grain size of snow is largest. The results in Figure 10 indicate that ICESat-2 measurements over melting snow should be accu-439 rate to within ~ 10 cm, assuming that the snow has compacted prior to melt. In contrast, 440 ATM shows low bias for CS_1 , where the grain size is smaller. Although we were unable 441 to consider ICESat-2 for CS1, the close agreement between ICESat-2 and ATM (Fig-442 ure 10) implies that ICESat-2 would experience minimal bias over locations with fresh, 443 fine-grained snow. Hence, utilizing ICESat-2 for accurate measurements of snow depth 444 is possible, though melting or aged snow may introduce bias and uncertainty in uncor-445 rected ATL03 measurements. Higher-level products, such as ATL06 or ATL08, may re-446 duce noise from the ATL03 photon cloud, but they will retain biased snow surface heights 447 if subsurface scattering is unaccounted for, particularly over shallow snow. 448

6 Conclusions

In this study, we used altimetry data from ICESat-2 and ATM to quantify volu-450 metric scattering bias in snow. A fusion of airborne optical grain size retrievals and Monte 451 Carlo modeling was used to predict altimetry bias over the western Greenland ablation 452 zone at the end of the melt season. ICESat-2 and the green ATM beam were compared 453 to the near-infrared ATM beam to estimate observed bias. Our results suggest a pos-454 itive relationship between the optical grain size of snow and altimetry bias over two case 455 studies. The modeled results show that snowpack density is an important driver for vol-456 umetric scattering, though actual biases in the study locations remained consistent with 457 densities of $\sim 315-500$ kg m⁻³. Although bias in both altimeters was generally within 10 458 cm, we cannot rule out more significant biases near the peak of the Northern Hemisphere 459 melting season, when snow grain coarsening will enhance volumetric scattering at all snow 460 densities. 461

The results in CS1 indicate that retrieved snow grain size is dependent on the instrument used. Grain sizes from AVIRIS-NG appear to originate from the snow surface, whereas ATM retrieves larger grain sizes within the snow subsurface. When combined with the Monte Carlo model, both data sets adequately reproduce trends in volumetric scattering bias, though the magnitude of the observed bias is not fully captured. Although used here, both instruments have infrequent coverage over mid-latitude field sites,
and ATM is not expected to collect data in the near future. Other sources of effective
grain size, such as MODIS or Sentinel-3 (Painter et al., 2009; Mei et al., 2021), will therefore be needed for future volumetric scattering assessments over snow.

Further research is needed to identify altimetry biases in the presence of snowpack 471 impurities or rough topography. The full impact of dust and black carbon on altimetry 472 signals is not known, so there is a need for accurate airborne and satellite retrievals of 473 surface impurity content. Similarly, a correction for rough or sloped terrain is needed, 474 given that both factors increase height uncertainties for both ICESat-2 and ATM. To address these problems among others, a follow-up study is in preparation that validates 476 the results in this paper over mid-latitude snow. The SnowEx mission is conducting air-477 borne lidar surveys for its 2023 Alaska campaign, several of which are expected to have 478 significant overlap with ICESat-2 tracks. The flights will overpass coastal and forested 479 regions of Alaska, so we anticipate a more rigorous analysis of ICESat-2 over multiple 480 terrain types. The expected result is a bias correction algorithm that ideally will be ap-481 plicable to all snow surfaces. 482

483 7 Open Research

The ICESat-2 ATL03 data may be found at NSIDC (Neumann et al., 2021). The ATM 532 nm and 1064 nm data is provided by Studinger and Manizade (2020b) and Studinger and Manizade (2020a), respectively. The AVIRIS-NG optical grain size data was obtained through correspondence with John Chapman (john.w.chapman@jpl.nasa.gov) and Winston Olson-Duvall (winston.olson-duvall@jpl.nasa.gov). The AVIRIS-NG reflectances used to derive grain size may be obtained from Chapman and Olson-Duvall (2019).

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