More than skin deep: sea surface temperature as a means of inferring Atlantic Water variability on the southeast Greenland continental shelf near Helheim Glacier

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Abstract

Outlet glaciers account for almost half of the Greenland Ice Sheet's mass loss since 1990. Warming subsurface Atlantic Water (AW) has been implicated in much of that loss, particularly along Greenland's southeastern coast. However, oceanographic observations are sparse prior to the last decade, making it difficult to diagnose changes in AW properties reaching the glaciers. Here, we investigate the use of sea surface temperatures (SST) to quantify ocean temperature variability on the continental shelf near Sermilik Fjord and Helheim Glacier. We find that after removing the short-term, atmospheric-driven variability in non-winter months, regional SSTs provide a reliable upper ocean temperature record. In the trough region near Sermilik Fjord, the adjusted SSTs correlate well with moored ocean measurements of the water entering the fjord at depth and driving glacier melting. Using this relationship, we reconstruct the AW variability on the shelf dating back to 2000, eight years before the first mooring deployments. Seasonally, AW reaches close to the fjord's mouth in fall and winter and further offshore in spring. Interannually, the AW temperatures in the trough do not always track properties in the source waters of the Irminger Current. Instead, the properties of the waters found at the fjord mouth depend on both variations in the source AW and, also, in the Polar Water that flows into the region from the Arctic Ocean. Satellite-derived SSTs, although dependent on local oceanography, have the potential to improve understanding around previously unanswered glacier-ocean questions in areas surrounding Greenland and Antarctica.

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Key Points:

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11	•	Sea surface temperatures measure upper ocean temperatures after variability tied
12		to the atmosphere is removed
13	•	Once adjusted for air temperature, Shelf Trough sea surface temperatures infer
14		deep Atlantic Water temperatures near Sermilik Fjord
15	•	Dilution of Atlantic Water as it intrudes onto the continental shelf modulates nearshore
16		temperatures

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17 Abstract

Outlet glaciers account for almost half of the Greenland Ice Sheet's mass loss since 1990. 18 Warming subsurface Atlantic Water (AW) has been implicated in much of that loss, par-19 ticularly along Greenland's southeastern coast. However, oceanographic observations are 20 sparse prior to the last decade, making it difficult to diagnose changes in AW proper-21 ties reaching the glaciers. Here, we investigate the use of sea surface temperatures (SST) 22 to quantify ocean temperature variability on the continental shelf near Sermilik Fjord 23 and Helheim Glacier. We find that after removing the short-term, atmospheric-driven 24 variability in non-winter months, regional SSTs provide a reliable upper ocean temper-25 ature record. In the trough region near Sermilik Fjord, the adjusted SSTs correlate well 26 with moored ocean measurements of the water entering the fjord at depth and driving 27 glacier melting. Using this relationship, we reconstruct the AW variability on the shelf 28 dating back to 2000, eight years before the first mooring deployments. Seasonally, AW 29 reaches close to the fjord's mouth in fall and winter and further offshore in spring. In-30 terannually, the AW temperatures in the trough do not always track properties in the 31 source waters of the Irminger Current. Instead, the properties of the waters found at the 32 fjord mouth depend on both variations in the source AW and, also, in the Polar Water 33 that flows into the region from the Arctic Ocean. Satellite-derived SSTs, although de-34 pendent on local oceanography, have the potential to improve understanding around pre-35 viously unanswered glacier-ocean questions in areas surrounding Greenland and Antarc-36 tica. 37

³⁸ Plain Language Summary

Greenland ice contributes one-quarter of global sea level rise each year and almost 39 half of that loss comes from glaciers at its periphery. Warming ocean waters may cause 40 much of that loss. Measurements made by ocean instruments serve as the predominant 41 method for studying the oceans around Greenland, but few observations exist prior to 42 the last decade. In this work, we investigate the use of sea surface temperatures acquired 43 by satellites to assess ocean temperature changes through time. We explore their use near 44 the southeastern Greenland coast, where warm water circulates from the North Atlantic 45 Ocean onto the continental shelf and eventually reaches Helheim Glacier, Greenland's 46 fifth largest glacier. Through a comparison with ocean instruments, we find that sea sur-47 face temperatures serve as a good indicator of upper ocean temperatures in this region 48

once proper corrections are applied. With these records, we find that the dilution of warm waters as they circulate from the North Atlantic changes over time and governs the temperature of the water that eventually reaches Helheim, which was previously unknown. Our work shows that sea surface temperatures can provide new insight into the ocean changes that may have impacted glacier retreat before ocean instruments were deployed.

54 1 Introduction

The Greenland Ice Sheet and its surrounding oceans have changed rapidly as a re-55 sult of shifting climate conditions in recent decades (Shepherd et al., 2012; The IMBIE 56 Team, 2019). Since the late 1990's, many of Greenland's tidewater glaciers have expe-57 rienced periods of substantial thinning and retreat, interspersed with periods of greater 58 stability and partial re-advance (Howat et al., 2008; Moon et al., 2012). Almost half of 59 the ice sheet mass loss occurs at marine-terminating outlet glaciers (Rignot & Kanagarat-60 nam, 2006; van den Broeke et al., 2009; Enderlin et al., 2014; The IMBIE Team, 2019) 61 and changes in total ice sheet discharge appear to be related to shifts in outlet glacier 62 frontal position (King et al., 2018). Enhanced submarine melting driven by ocean warm-63 ing has been implicated in many recent glacier front retreat events (Walsh et al., 2012; 64 Straneo & Heimbach, 2013; Millan et al., 2018), such as at Jakobshavn Isbræ (Holland 65 et al., 2008), Zachariae Isstrom (Mouginot et al., 2015), Kangerdlugssuag (Christoffersen 66 et al., 2011; Inall et al., 2014; Bevan et al., 2019), and Helheim Glacier (Howat et al., 67 2008). These glaciers alone accounted for more than 40% of Greenland's excess discharge, 68 as opposed to surface runoff, between 2000 and 2012 (Enderlin et al., 2014). However, 69 the changes in ocean circulation leading to these glacier retreat events is generally weakly 70 characterized. Changes in ocean temperature and volume transport near Greenland's 71 tidewater systems were mostly unmonitored during many earlier events. 72

Sermilik Fjord abuts Helheim Glacier, one of Greenland's largest glaciers (Enderlin 73 et al., 2014). The region is among the best instrumented and well understood glacier-74 ocean systems (Straneo et al., 2016), making it an ideal area to investigate the extent 75 to which sea surface temperature variability may be used to infer ocean variability in the 76 vicinity of an outlet glacier, where oceanographic thermal characteristics can be of great 77 significance. Sermilik experiences highly variable ocean circulation and heat transport 78 within the fjord and on the continental shelf (Straneo et al., 2010; Sutherland et al., 2013; 79 Jackson et al., 2014). Heat is primarily delivered by relatively warm, saline Atlantic Wa-80

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ter (AW; $\sim 2.0-5.2^{\circ}$ C, >150-250 m) from the Irminger Current (IC) offshore of the con-81 tinental shelf break (Straneo et al., 2010; Jackson et al., 2014). The IC carries AW equa-82 torwards at the surface and extending down to depths greater than 500 m (Rudels et al., 83 2002; Johannessen et al., 2011; Våge et al., 2011; Andresen et al., 2012). Along the in-84 ner shelf, the East Greenland Coastal Current (EGCC) is a low salinity wedge perched 85 atop deeper AW (Bacon et al., 2002; Sutherland & Pickart, 2008). Above 150-250 m, the 86 upper layer carries cold and fresh Polar Water (PW; $<4^{\circ}C$) exported out of the Arctic 87 and the northeastern Greenland fjords, including Sermilik (Bacon et al., 2002; Suther-88 land & Pickart, 2008; Harden et al., 2014). The EGCC flows south, 20-30 km wide, hug-89 ging the Greenland coastline (Sutherland & Pickart, 2008). Transport within the EGCC 90 varies seasonally with the greatest freshwater transport in December and generally higher 91 transport in winter and spring coinciding with its speedup and deepening (Sutherland 92 & Pickart, 2008; Bacon et al., 2014; Harden et al., 2014; Le Bras et al., 2018). IC and 93 EGCC variability on the shelf can influence water properties in Sermilik Fjord and, there-94 fore the glacier front; however, the oceanographic studies in the region are mostly recent. 95 Past ocean variability on the shelf and within the fjord is largely unknown before the record 96 that began in 2008 (Straneo et al., 2016), years after thinning and retreat occurred at 97 Helheim Glacier (Howat et al., 2005; Luckman et al., 2006). 98

While the EGCC and IC have been relatively well studied, the interactions between 99 them in the region outside of Sermilik Fjord, and the temporal variability in those in-100 teractions, are poorly understood as a result of limited spatial and temporal measure-101 ment coverage. From summer shipboard surveys and longer-term mooring deployments, 102 research in this area has suggested that seasonality in EGCC current width, depth, and 103 transport along the shelf is controlled by alongshore winds (Sutherland & Pickart, 2008; 104 Harden et al., 2014; Le Bras et al., 2018). The PW layer across the shelf thickens through 105 winter and spring (Straneo et al., 2010; Bacon et al., 2014), likely caused by increasing 106 freshwater transport from the Arctic (Harden et al., 2014). Warm AW (2.0-5.2°C, >150-107 250 m; Jackson et al., 2014) encroaches onto most of the shelf below PW throughout the 108 year, often via troughs that cut across the continental shelf and into the fjords at depth 109 (Rudels et al., 2002; Sutherland et al., 2013). The AW near the coast is cooler than at 110 the shelf break likely as a result of surface cooling and mixing with PW (Straneo et al., 111 2012). Aside from this general AW inflow at depth, AW can flow onto the shelf as a full-112 depth layer via occasional AW intrusions within the troughs or seasonally-varying in-113

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flow across portions of the shelf (Sutherland et al., 2013; Harden et al., 2014). This sea-114 sonal inflow intensifies in the fall and is associated with a narrower EGCC banked up 115 against the coast (Harden et al., 2014). Straneo and Heimbach (2013) have posited that 116 more frequent AW intrusions may lead to warmer waters on the inner shelf, and greater 117 influence of AW on the shelf links to increased glacier calving activity (Andresen et al., 118 2012). All past work in this region is limited spatially or temporally relative to the scale 119 of the overall study region, which extends ~ 200 km between the coast and continental 120 shelf break. As a result, little is known about the variability of the AW across much of 121 the shelf and its influence on inner shelf and fjord water temperatures. However, it is 122 clear that AW inflow strongly influences heat transport onto the shelf (Sutherland et al., 123 2013; Harden et al., 2014) and into the fjord (Straneo et al., 2010; Jackson et al., 2014). 124

The availability of sea surface temperature (SST) records from the period prior to 125 the speed up of many Greenland glaciers that occurred in the early 2000s raises the pos-126 sibility of inferring oceanic variability at Greenland's glacial margins through proxies that 127 are built on SSTs. Several recent studies have attempted to define this relation largely 128 through correlations of glacier activity and SST variability with mixed results. Warm 129 SSTs and sea ice variability have been correlated with glacier front changes (Howat et 130 al., 2008, 2010; Johannessen et al., 2011; Andresen et al., 2012; Schild & Hamilton, 2013; 131 Khan et al., 2014), but important questions remain about the extent to which SSTs around 132 Greenland provide information about the subsurface water column where the AW, which 133 influences submarine melting of the larger glaciers, resides. Important progress on that 134 front has been made by Sutherland et al. (2013) who found that, for the period from 2004-135 2010, summertime (JJA) MODerate resolution Imaging Spectroradiometer (MODIS) SSTs 136 on the continental shelf near Sermilik Fjord correlated closely with tagged-seal temper-137 ature measurements at 50 m depth, with diminishing correlation at deeper depths un-138 til decoupling below 250 m. However, a broader treatment is necessary in order to con-139 strain temporal variations in ocean temperatures, which is crucial for discerning ice-ocean 140 interactions in glacier retreat events back through time. SSTs have been hitherto un-141 derutilized in glacier change analysis and may provide observations that complement tem-142 poral and spatial gaps of *in situ* measurements. 143

Here, we investigate the use of SSTs as a means of assessing the variability of the subsurface waters that enter Greenland's glacial fjords and melt glaciers at depth (Straneo et al., 2012). Differently from other studies we make use of oceanographic subsurface data

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to investigate the correlation between surface properties in different regions on the shelf, 147 in the vicinity of Sermilik Fjord, and those observed at depth at the mouth of the fjord. 148 We derive a proxy for subsurface AW temperatures using SSTs adjusted for local air tem-149 peratures to produce the first AW record near Sermilik dating back to 2000. We also track 150 AW intrusion variability along the trough leading to Sermilik Fjord to give insight into 151 the drivers of subsurface AW temperature changes on the shelf. While our findings are 152 specific to the oceanography of this region, our analysis demonstrates that SSTs provide 153 novel insight into ocean variability and hold promise for addressing long-standing glacier-154 ocean questions around both ice sheets. 155

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2 Data and Methods

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2.1 Satellite-Derived Sea Surface Temperatures

In contrast to the spatially and temporally limited in situ data, SSTs acquired from 158 satellites offer an untapped and potentially illuminating resource for tracking ocean tem-159 perature and extent of AW inflow onto the continental shelf. To help reconstruct ocean 160 variability here, we use MODIS-derived SSTs, which provide the temperatures of the ocean 161 skin (upper few μ m; SST_{skin}; see Figure S1). The accessibility of MODIS SST products, 162 the instrument's moderate spatial resolution, and its ~ 15 scans per day by each of the 163 two satellites on which it flies (Aqua and Terra) provides extensive spatiotemporal cov-164 erage suitable for our objectives. In this work we use the MODIS Aqua and Terra Daily 165 Global Level 3 4-km Mapped Thermal daytime and nighttime SST R2014.0 products (qual-166 ity level 0 and 1) derived from the 11 and 12 μ m thermal infrared (IR) channels 31 and 167 32, respectively (Kilpatrick et al., 2015; Ocean Biology Processing Group, 2014a, 2014b). 168 The data we use span the period beginning Feb 24, 2000 for Terra and July 4, 2002 for 169 Aqua, and ending Dec 31, 2018 for both satellites. The retrieval error for the SSTs is 170 $\sim 0.4^{\circ}$ C (Kilpatrick et al., 2015). We reference the four MODIS SST products hereafter 171 based on their division by satellite and time of day: Terra daytime (T-D), Terra night-172 time (T-N), Aqua day (A-D), and Aqua night (A-N). 173

Before extracting SSTs from each of the four products, we account for cloud and sea ice contamination that may occur because the MODIS SST processing pipelines are not optimized for polar climates (Kilpatrick et al., 2019; Jia, 2019, see Supplementary Information). Arctic SSTs can have cloud contamination (Kilpatrick et al., 2019) that

can introduce noise by shifting SST retrievals toward an artificially cold measurement 178 (Ackerman et al., 1998). To reduce these effects, we apply a mask for clouds and ensure 179 further robustness by applying spatial and temporal averaging for each sampling region 180 and across multiple SST products, as described below. Sea ice contamination can also 181 lead to a cool bias in the SST retrievals, and we find that the daytime SST products, 182 especially in Aqua, contain systematically more sea ice contamination (Figure S2a). To 183 reduce sea ice misclassification and these inter-product differences, we apply a separate 184 sea ice mask, created from MODIS and passive microwave sea ice products, to the four 185 daily SST products (see Supplementary Information). 186

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2.2 Sampling Regions

To investigate SST spatial and temporal variability near Sermilik Fjord, we exam-188 ine three regions: the IC, EGCC, and the Shelf Trough (ShTr) region (Figure 1). We chose 189 the IC and EGCC sampling regions based on the observations from Rudels et al. (2002) 190 and Sutherland and Pickart (2008) that characterize the locations of the currents, re-191 spectively, and chose the boxes' sizes to include relatively homogenous SSTs based on 192 the SST climatology from the region. We use IC and EGCC regions as indicative of AW 193 and PW end members, respectively, because these are the primary water masses at the 194 surface in the respective boxes (Rudels et al., 2002; Sutherland et al., 2013). In addition, 195 we define a 'ShTr region' over the trough leading to Sermilik Fjord, where AW flows onto 196 the shelf and mixes with the EGCC (Sutherland et al., 2013). Results for the three re-197 gions are not sensitive to small changes in the box locations and size. 198

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2.3 Seasonal and Diurnal Biases

Seasonal differences in instrument scan coverage between the Level 3 MODIS SST 200 products must be accounted for before they can be used to investigate ocean variabil-201 ity in polar regions. Each MODIS instrument acquires 12-18 scans of our study region 202 each day. During the summer solstice, few are classified as nighttime and most scans are 203 binned into the daytime product (Figure S2b). The opposite is true during the winter 204 solstice. As there is a far higher likelihood of getting at least one sea ice- and cloud-free 205 measurement during a day with more scans, this disparity in scan coverage between sea-206 sons means the day products (daily, 8-day, monthly, annual) are skewed to summer mea-207 surements and night toward winter. As a result of these differences, creating a robust 208



Figure 1. 2000-2016 mean nighttime MODIS SST of the Ammassalik region around Sermilik fjord. Solid black arrows show the location of the Irminger Current (IC) and dashed show the East Greenland Coastal Current (EGCC), which mix across the Shelf Trough (ShTr) region. Boxes indicate areas over which SST is averaged for each region, the blue star shows the trough mooring, and the white circle marks Helheim Glacier. Bathymetry from BedMachine v3 is in thin black lines at 300, 400, 500 m, and every 500 m thereafter (Morlighem et al., 2017).

²⁰⁹ and continuous record of SSTs that is representative of all seasons requires creating a

²¹⁰ composite by combining day- and nighttime SSTs for each region.

The day and night products carry inherent biases based on diurnal differences in 211 the SST_{skin} , which we remove before combining the datasets into a composite. Diurnal 212 biases cause a decoupling between the ocean skin and underlying water (Price et al., 1986; 213 Donlon et al., 2002; Minnett, 2003). They are expected as a result of differences in di-214 urnal thermocline and skin temperature effects on the SST_{skin} between day- and night-215 time, which also vary seasonally (e.g. Sverdrup et al., 1942; Koizumi, 1956; Eastwood 216 et al., 2011). We calculate the diurnal bias based on differences between the day and night 217 products for each region separately. To determine the biases, we first take the mean of 218 the pixels in the sampling boxes and produce daily time series from 2000-2018 for each 219 of the four masked daily products (Figure 1). We average the A-D and T-D products 220 together to produce a daytime average for each region. We do the same for the two night-221 time products. From these records, we produce a day- and nighttime climatology for each 222 of the sampling regions using monthly means across the entire 19-year record (Figure 223 2). We calculate standard error for the climatologies between the 19 years of monthly 224 data. We define the seasonally-varying diurnal bias as the systematic warm bias in the 225 day products in comparison to the night records across the climatologies $(0.28^{\circ}, 0.39^{\circ},$ 226 0.46° C for the IC, ShTr, and available EGCC time period, respectively; see Supplemen-227 tary Information). We use the seasonally-varying diurnal biases for each sampling loca-228 tion – which result from wind speed and solar radiative forcing (Kawai & Wada, 2007) 229 - and subtract them from the day product. 230

With the diurnal bias removed from the daytime records so that they are equiv-231 alent to the nighttime, we assume that all four records represent the bulk SST temper-232 ature (Figure S1; Sutherland et al., 2013) and can be combined into a composite record. 233 This assumption for nighttime SST_{skin} is consistent with Minnett (2003) and used by 234 others (e.g. Kilpatrick et al., 2015; Jia, 2019). We hereafter refer to this bulk temper-235 ature as the SST, although the measurement should also still contain a slight and con-236 stant cool skin bias to bulk temperatures as a result of heat flux to the atmosphere ($\sim 0.17^{\circ}$ C; 237 Donlon et al., 2002). We average the nighttime and corrected daytime records to pro-238 duce the composite daily SST time series for each region (Figure 2b). From these, we 239 produce weekly and monthly mean SST time series that we use for the rest of our anal-240 241 yses.

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Figure 2. Monthly sea surface temperature climatologies for the 2000-2018 period for the three regions. The IC (yellow), ShTr (green), and EGCC (purple) records are shown for the averaged daytime products (dotted) and nighttime products (dashed). Solid lines show the climatology of the composite after the diurnal bias has been subtracted from the daytime record. Standard error for the nineteen years is shown as shading for each.

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2.4 Air and Ocean Temperature Records

To determine the extent to which variability in satellite-derived *SST* can be used to reconstruct upper ocean temperature outside Sermilik Fjord, we consider two factors that influence *SST* variability. The *SST* depends heavily on the depth, magnitude, and history of thermal gradients and stratification at the surface (Donlon et al., 2002, see Figure S1). These properties are controlled by solar heating, heat exchange with the atmosphere, and heat exchange with deeper waters (e.g. Donlon et al., 2002; Minnett, 2003). SST will therefore be impacted by and potentially covary with both atmospheric and
ocean mixed layer temperature changes (e.g. Frankignoul & Hasselmann, 1977; Jaswal
et al., 2012).

We use European Center Medium-Range Weather Forecasts (ECMWF) ERA-5 operational reanalysis dataset (Hersbach et al., 2020; Copernicus Climate Change Service, 2017) to assess the relationships between *SST* and air temperatures in each region. From ERA-5, we use the 2-m air temperature measured at 6-hourly time-steps on a 0.5°x 0.5° grid. Air temperatures vary significantly across our study region and are, thus, averaged for each of the *SST* sampling areas, separately.

To determine subsurface water temperature variability on the inner-shelf below the 258 EGCC, we use data from a mooring deployed multiple times between August 24, 2009 259 - August 18, 2013 on the continental shelf within the trough that leads to Sermilik Fjord 260 (Jackson et al., 2014; Harden et al., 2014; Jackson & Straneo, 2016). From the moor-261 ing, we use the temperatures recorded by one instrument each year, either a Microcat 262 SBE37SM or XR 420 RBR sensor, deployed between 264 and 305 m. These tempera-263 tures provide a time varying record of subsurface AW that is known to flow into Sermi-264 lik Fjord (Straneo et al., 2011; Jackson & Straneo, 2016). 265

266 **3 Results**

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3.1 Seasonal and Interannual

The seasonal SST records for the three sampling regions are similar but have off-268 sets and different amplitude ranges (Figure 2). Across the 19-year record, the IC is warmest 269 on average $(6.5^{\circ}C)$ and the EGCC coolest $(2.3^{\circ}C)$, with ShTr temperatures between them 270 $(4.3^{\circ}C)$. Similarly, the IC has the largest seasonal range $(5.3^{\circ}C)$, with the ShTr and EGCC 271 having progressively smaller ranges $(4.3^{\circ}C \text{ and } 3.3^{\circ}C, \text{ respectively})$. Across the entire 272 record, the seasonal cycle dominates the interannual variability for the three regions, es-273 pecially for the IC and ShTr (Figure 3). The interannual variability of the EGCC is slightly 274 more prominent because the EGCC has a smaller seasonal signal. 275

Seasonally, all three regions experience peak temperatures in August, while the timing of the minimum occurs at slightly different times (Figure 2). The IC experiences a March *SST* minimum and a more sinusoidal seasonal cycle consistent with seasonal ra-

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Figure 3. Composite and adjusted *SST* records for each sampling area. Monthly mean (medium) and daily (thin) *SST* composite records for the IC (yellow), ShTr (green), and EGCC (purple) boxes are shown. The *SST* adjusted for air temperature is shown as thick lines (see text for description).

- diative forcing. The ShTr and EGCC have a slight cooling trend in winter and spring
- with a steep transition into warming after reaching minimum temperatures in April/May
- $_{281}$ (ShTr) and May (EGCC). The EGCC minimum lags the ShTr by ~ 20 days. The late
- transition from winter cooling into warming near the coast is consistent with the influ-
- ence of sea ice (Hastings, 1960), freshwater runoff (Sutherland et al., 2009), and other
- water-stratifying processes inshore (e.g., weaker winds; Oltmanns et al., 2014; Moore et
- al., 2015).

3.2 Dependence on Air Temperature

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To determine the controls on SST variability near Sermilik Fjord, we compare the 287 SST with ERA-5 air temperature records (Figure 4, S3). We use ordinary least squares 288 (OLS) regression to examine the linear relationships between air temperature and SST289 in each of the regions separately (Seabold & Perktold, 2010). Through continuous heat 290 exchange, surface air temperature and upper ocean co-variability can have time scales 291 of hours (e.g., diurnal solar heating and turbulence) to a few days (e.g., inertial mixed 292 layer currents; Garwood, 1979; Donlon et al., 2002). We use weekly averages, rather than 293 daily, in the regression to account for both timescales. 294



Figure 4. The 'fjord mouth' subsurface water temperature proxy (ShTr SST^{adj}) compared to heat sources. Monthly Shelf Trough SST (green) and SST adjusted for air temperature (SST^{adj} ; orange) - which we identify as a fjord mouth subsurface proxy - are compared to mooring water temperatures from 290 m (black) and air temperatures (blue). The root mean square error (RMSE) between SST^{adj} and mooring temperatures are given. Standard deviations for the ShTr SST^{adj} and mooring temperatures are shown as orange shading and error bars, respectively. Winter (purple) months are shown. Not shown, air temperature reaches -1° to $-6^{\circ}C$ each winter.

Using an OLS regression model, we find that weekly SST in all regions are strongly 295 correlated to ERA-5 air temperature records in summer, but that this relationship does 296 not hold in winter (Figure 4, S3). For example, ShTr SST has a strong linear relation-297 ship $(r^2=0.48)$ with weekly air temperatures in summer (JJAS; slope=0.71\pm0.04; Fig-298 ure S3b), but the relationship becomes weak or insignificant $(r^2=0.03)$ during winter months 299 (DJFM; slope= 0.07 ± 0.02). For the remaining months (Apr, May, Oct, Nov), SST shows 300 a weaker correlation with air temperature (slope= 0.40 ± 0.04 ; r²=0.23). Our findings are 301 consistent with a shallower ML or strongly stratified surface ocean in non-winter months, 302 which results in a more closely coupled air-sea temperature response than in winter (Chang, 303 1993). It is also consistent with previous work that found a weaker coupling between SST304 and air temperature in winter than in summer around Greenland (Singh et al., 2005, 2006). 305 This general relationship holds for all three sampling regions (Figure S3), but the mag-306 nitudes and significance of the relationships in some months differ slightly (Table 1). This 307 is expected based on stratification and heating differences between the regions (e.g. Suther-308 land & Pickart, 2008), which affect ocean skin temperatures and air-sea interactions (Garwood, 309 1979). 310

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3.3 SST on the Shelf and Mooring

We also investigate the connection between ShTr SST and the subsurface moor-312 ing temperature using the OLS regression model. The ShTr box is a region where AW 313 inflow can extend all the way to the surface (Sutherland et al., 2013). In the regression, 314 we use monthly averages as the most appropriate timescale for comparisons between SST315 and mooring temperatures to account for potential sub-monthly lag times. We find that 316 ShTr SST correlates strongly with subsurface water temperatures in wintertime only (Fig-317 ure 4). Specifically, monthly trough mooring temperatures (290 m) have a significant lin-318 ear relationship with ShTr SST in winter (slope= 1.18 ± 0.16 , $r^2=0.79$), but not in sum-319 mer months ($slope=0.23\pm0.09$, $r^2=0.18$) (Figure S4a). Markedly, the strong relationship 320 between ShTr SST and subsurface waters occurs in the months when SST shows little 321 linkage to air temperature and the region receives little solar insolation. 322

We attribute this relationship to the fact that the upper ocean water masses in the ShTr box are linked with those found subsurface at the mooring location further downstream. This is consistent with the results of Sutherland et al. (2013), who showed that full depth AW intrusions occur in the ShTr region and that AW are found subsurface

at the mooring. Conversely, surface waters at the mooring location are indicative of PW 327 properties, consistent with the stratification described by Harden et al. (2014). Based 328 on observed velocities of $0.1-0.6 \text{ ms}^{-1}$ (Harden et al., 2014) and wind-driven velocity shifts 329 on synoptic timescales (Jackson et al., 2014), we expect a temperature lag for water trans-330 port between the middle of the ShTr box and the mooring site (~ 80 km) that may range 331 from a day to more than a week – supporting our choice of focusing on monthly vari-332 ability. The stronger wintertime mooring/SST relationship is consistent with deepen-333 ing of the IC wintertime mixed layer as a result of air-sea forcing (Våge et al., 2011; de 334 Jong et al., 2018). 335

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3.4 Upper Ocean Temperatures and a Fjord Mouth Subsurface Water Temperature Proxy

Since SST is significantly correlated with air temperature in non-winter months, consistent with a stronger near-surface stratification, we removed the portion of SST variability related to air temperature to obtain a better indicator of upper layer ocean temperatures. To do this, we build a multivariate linear model that expresses daily (t) SSTfor each region as the combination of a portion that covaries with air (T^{air}) and one that covaries with upper ocean (T^{ocean}) temperatures:

$$SST_R(t) = A_R^m T_R^{air}(t) + B_R T_R^{ocean}(t) + C_R$$
(1)

where A^m is the proportionality coefficient for the relationship between SST and air tem-344 perature that varies by month (indicated by the superscript m), B is a constant propor-345 tionality coefficient with upper ocean temperatures, and C is a skin bias (expected to 346 be similar to the ~ -0.17 global average; Donlon et al., 2002). We assume that this re-347 lationship holds for each of the regions with coefficients that are region dependent and 348 indicated by the subscript R. Physically, $A^m T^{air}(t)$ represents the variability resulting 349 from air-sea interactions that is a function of the heat exchange between the near-surface 350 ocean layers (dependent on layer thickness), the short and longer term flux of latent and 351 sensible heat through the air-sea interface (dependent on air temperature, ocean tem-352 perature, wind speed, and humidity), short- and long-wave radiation through the ocean 353 surface, salinity effects, and horizontal advection (Kraus & Turner, 1967; Denman, 1973; 354 Frankignoul & Hasselmann, 1977). Previous climate modeling work has estimated the 355 linkage between air temperature and SST, sometimes using a simple bias correction (Schulz 356

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- et al., 1997) or through more complex relationships that include humidity, and wind speed (Konda et al., 1996; Gautier et al., 1998; Jones et al., 1999; Singh et al., 2006). The re-
- $_{359}$ lationship between air temperatures and SST is complex, but as we show in Section 4.2
- and below, it can be approximated in our study regions as a simple statistical linear re-
- lationship. For each region (R), we further define an adjusted SST (SST^{adj}) :

$$SST_R^{adj}(t) = SST_R(t) - A_R^m T_R^{air}(t)$$
⁽²⁾

- where $A^m T^{air}(t)$ is subtracted to remove the SST variability tied to the atmosphere.
- We calculate A^m for each region, and each month, using an OLS regression model with
- $_{364}$ a monthly interaction term that finds the slope relationships between monthly SST and
- air temperatures (Table 1). For months with statistically insignificant slope relationships
- (p ≤ 0.05), we apply A=0; therefore, SST^{adj} is equivalent to SST for some winter months
- 367 (DJF).

Table 1. A^m parameters parameters calculated by the Ordinary Least Squares Regression models for the ShTr, IC, and EGCC. Number of measurements (N) and R^2 provided for all months of each model. Intercept and insignificant parameters not used for corrections.

Month	ShTr	IC	EGCC
	N = 965	N = 980	N = 776
	$R^2 = 0.71$	$R^2 = 0.91$	$R^2 = 0.60$
Jan	$0.02{\pm}0.04$	$0.02{\pm}0.03$	-0.09±0.04
Feb	$0.03 {\pm} 0.04$	-0.01 ± 0.03	-0.07 ± 0.05
Mar	$0.12{\pm}0.03$	-0.05 ± 0.03	-0.07 ± 0.05
Apr	$0.21{\pm}0.06$	$0.01 {\pm} 0.02$	-0.13 ± 0.09
May	$0.02{\pm}0.07$	$0.21{\pm}0.02$	$\textbf{-0.19}{\pm}\textbf{0.08}$
Jun	$0.45{\pm}0.04$	$0.41{\pm}0.01$	$0.12{\pm}0.03$
Jul	$0.63{\pm}0.02$	$0.53{\pm}0.01$	$0.40{\pm}0.02$
Aug	$0.58{\pm}0.02$	$0.54{\pm}0.01$	$0.47{\pm}0.02$
Sep	$0.52{\pm}0.02$	$0.50{\pm}0.01$	$0.36{\pm}0.02$
Oct	$0.55{\pm}0.03$	$0.44{\pm}0.01$	$0.24{\pm}0.04$
Nov	$0.57{\pm}0.06$	$0.31{\pm}0.02$	$-0.10 {\pm} 0.06$
Dec	$0.06{\pm}0.05$	$0.15{\pm}0.02$	$\textbf{-0.15}{\pm}\textbf{0.04}$

^{*a*}Bold indicates significant parameters ($p \le 0.05$).

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- The SST^{adj} for the three regions have a wide range of seasonal and interannual vari-
- ability that we interpret as representing the upper ocean temperature variability (Fig-
- ure S1). The EGCC and IC monthly and interannual variability (Figure 5) is consistent
- with ranges in the upper ocean mooring temperatures described by Harden et al. (2014)
- and de Jong et al. (2018), respectively. Unlike for the absolute SST, the EGCC and ShTr

- SST^{adj} exhibit a larger variance (0.51° and 0.45°C, respectively) than the IC (0.20°C).
- ³⁷⁴ This is consistent with synoptic and seasonal upper ocean temperature swings associ-
- ated with seasonal heating cycles, cold meltwater influx, and variable AW inflow inshore
- ³⁷⁶ (Straneo et al., 2010; Harden et al., 2014). SST^{adj} temperature ranges are smaller than
- $_{377}$ those of the absolute SST (Figure 3), which is consistent with differences between up-
- per ocean temperatures and bulk *SST* (see Figure S1; Chang, 1993).



Figure 5. Monthly SST^{adj} records for 2000-2018 for the Irminger Current (yellow), Shelf Trough (orange), and East Greenland Coastal Current (purple). Thin lines are monthly SST while thick lines represent 24 month low-pass Butterworth filtered records for each. The trough mooring temperatures (gray) are shown for comparison. The ShTr SST^{adj} record is a proxy for the trough mooring temperatures.

³⁷⁹ Using a second-order low-pass digital Butterworth Filter (Virtanen et al., 2019), ³⁸⁰ with a 24-month cutoff frequency, we further examine the longer-term SST^{adj} variabil-³⁸¹ ity for the three different regions. While we find that the upper ocean layer was warmest ³⁸² in the early 2000's in all three regions, their variability differed in subsequent years. Specif-³⁸³ ically, the IC remained warm from 2005 to 2008, while the ShTr and EGCC experienced ³⁸⁴ a general cooling that was more pronounced for the ShTr than the EGCC. Furthermore, the ShTr box also continuously warmed after 2012, whereas the IC exhibited long-term cooling, consistent with the deepening of convection in the Irminger Sea and generalized cooling of the subpolar gyre during this period (de Jong & Steur, 2016; de Jong et al., 2018). In general, warmer years in the ShTr record were consistent with IC temperatures, while they more closely resembled EGCC temperatures in the coldest years. Although the upper ocean temperature records for the three regions differed substantially, ShTr SST^{adj} correlated more with the EGCC (r²=0.31, p<0.001) than the IC (r²=0.10, p=0.13).

Given that the water column in the ShTr region can be relatively homogenous from 392 surface to depth as a result of full-depth AW layers flowing onto the shelf along the trough 393 (Sutherland et al., 2013; Harden et al., 2014), we investigate the extent to which ShTr 394 SST^{adj} can be used as a proxy for the subsurface water temperatures at depth at the 395 mooring location near Sermilik Fjord mouth, year-round (see Figure 1 for mooring lo-396 cation). We find that the ocean mooring temperatures show a linear relationship with 397 ShTr SST^{adj} (slope=0.98±0.14, r²=0.51; Figure S4b) that is similar to the wintertime 398 relationship (slope= 1.18 ± 0.16) found using the full SST. Non-winter ShTr SST^{adj} re-399 sembles the uncorrected wintertime measurements in comparison to mooring temper-400 atures (Figure 4; slope= 0.93 ± 0.18 , r²=0.46). We find a strong correlation (r²=0.69) be-401 tween ShTr SST^{adj} and mooring temperatures that is also stronger for wintertime mea-402 surements (winter RMSE=0.58°C, summer RMSE=0.70°C, total RMSE=0.67°C). 403

In addition to the correspondence between the mooring data and the adjusted SST404 in the ShTr region, the fact that the ShTr SST^{adj} is warmest in November and Decem-405 ber and coolest from March to May (Figure 6b) is consistent with subsurface temper-406 atures observed on the shelf between 2004 and 2010 using tagged-seals (Straneo et al., 407 2010). We also find good agreement between ShTr SST^{adj} measurements and shipboard 408 hydrographic surveys within the trough taken each August from 2009-2013 (Harden et 409 al., 2014). Thus, we conclude that the ShTr SST^{adj} derived here is a good proxy for the 410 subsurface ocean temperatures at the fjord mouth and hence of the waters that feed Ser-411 milik Fjord at depth and reach the base of Helheim Glacier. 412

413

3.5 AW Encroachment onto the Shelf

As shown in Figure 4 and 5, temperature changes in the ShTr region reflect the combined influence of IC and EGCC temperature variability. Here, we investigate changes

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in the AW intrusions onto the continental shelf by examining the occurrence of warm 416 temperatures along the trough that supplies AW to the 'ShTr' region and eventually, at 417 depth, to Sermilik Fjord (Figure 6). Specifically, we consider a transect of thirteen 14 418 x 14 km boxes along the trough crossing the continental shelf and leading to Sermilik 419 Fjord. Within each of the boxes we subtract the daily IC SST^{adj} temperature from the 420 SST^{adj} in the box to create a "trough anomaly" (Figure 6a). The trough anomaly thus 421 indicates how different the box SST^{adj} is from that of the IC. By doing this, we remove 422 any interannual variability in magnitude due to changes in the IC temperature itself as 423 opposed to more or less AW intruding onto the shelf. A less negative anomaly means that 424 trough waters are almost as warm as those offshore (IC region) while a more negative 425 anomaly means that the trough is considerably colder than the IC. Any box covered in 426 sea ice is assumed to be at the freezing temperature of seawater $(-1.8^{\circ}C)$. To determine 427 thresholds for quantifying when AW temperatures are present at the surface within the 428 trough, we compare anomalies found for all of the pixels within the IC and EGCC boxes, 429 which represent AW and PW end members, respectively. More than 99% of IC pixels 430 have anomalies above -1.5° C, while anomalies within the EGCC tend to be more neg-431 ative (Figure S5). Using this distinction, we consider a box to have AW at the surface 432 when weekly trough anomalies are greater than -1.5° C (Figure 6b). We determine the 433 seasonal climatology and annual mean for the location of the -1.5° C contour along the 434 transect, removing the weeks where cloud cover obscures part of the transect and makes 435 identification of the -1.5°C contour uncertain. Our results are not sensitive to slight vari-436 ations in threshold choices. 437

Temperature anomalies along the trough exhibit substantial variability on synop-438 tic and seasonal timescales (Figure 6). From the climatology, we show that waters with 439 properties similar to those in the IC box extend shoreward (location of box 3) in fall and 440 early winter (OND) but are found offshore (location of box 8) in spring (AM; Figure 6b). 441 We also find a high degree of variability on weekly timescales (Figure 6c). We interpret 442 instances of small amplitude anomalies in the trough temperatures to be associated with 443 AW inflow onto the shelf. Our observations indicate that AW intrudes deeper along the 444 trough in late fall, bringing surface AW closer to the fjord's mouth, while it remains fur-445 ther offshore in spring. This seasonality, in turn, is consistent with the seasonal variabil-446 ity in the ShTr SST^{adj} (Figure 6b) and with the findings of Sutherland et al. (2013). In-447 trusion of AW further along the trough coincides with warmer seasonal ShTr SST^{adj} tem-448



Figure 6. Variability in the trough anomaly along the trough leading to Sermilik Fjord. (a) Map of the 19-year climatological anomaly of SST^{adj} overlain by the thirteen – 14 x 14 km sampling boxes representing a transect from the fjord [1] out to the continental slope [13]. Bathymetry contours from BedMachine v3 (Morlighem et al., 2017). (b) Weekly 2000-2018 climatology of the trough anomaly from (a). The transect from box [1] on the top to [13] at the bottom spans the y-axis and time along the x-axis. The average weekly (thin line) and smoothed (thick line) location of the -1.5°C contour for the trough anomaly (white) is shown in comparison to the same for the ShTr SST^{adj} (orange) with smoothing using a 20-week Butterworth filter. Shading for each shows the standard error for the 19 years. (c) Weekly record for the trough anomaly from 2000-2018. Axes are similar to (b) and the annual means for the -1.5°C trough anomaly location and ShTr SST^{adj} are shown.

peratures ($r^2=0.85$, p<0.001), although there is a ~2-week lag between the minimum ShTr SST^{adj} and when AW is furthest offshore.

Interannually, we find that AW spreads furthest inshore in 2003 and 2014-2018, while it is most consistently offshore in 2006-2008 and 2011-2012 (Figure 6c). 2003 and 2018 experienced almost no seasonality, with AW extending along the entire trough (in shore of box 3) nearly year-round. Conversely, PW extended up to the continental shelf-break for more than half of the year in 2007 and 2008 with little to no AW on the shelf dur-

- ing that time. The extent of AW intrusion strongly correlates with the ShTr SST^{adj} (r²=0.76,
- p<0.001), although AW intrudes furthest inshore after 2014, which does not coincide with
- similarly extreme ShTr temperatures. The earlier periods of stronger intrusion, 2003 and
- ⁴⁵⁹ 2010, are consistent with times of anomalously warm IC, however, the later period is not
- (Figure 4). Explicitly, this means that the variability in heat content of the upper Irminger
- 461 Sea region (IC box) is not indicative of the extent of AW intrusions onto the shelf. Even
- 462 as the Irminger Sea has been cooling, in recent years, the AW is intruding deeper onto
- the shelf and presumably influencing the waters flowing into Sermilik Fjord at depth.
- $_{464}$ 4 Discussion

465

4.1 Reliable Application of Sea Surface Temperatures

While MODIS SSTs provide an under-utilized source of significant insight to oceanic 466 heat transfer to glaciers, significant challenges have slowed widespread application and 467 interpretation of this data trove. Frequent cloud and sea ice cover lead to few measure-468 ments of the ocean surface despite the on-average 15 scans acquired per day, and weak-469 nesses in the built-in SST masking protocols mean that the boxed SST products can of-470 ten have cloud or ice contamination and tendency toward cold biases around Greenland 471 (Szczodrak et al., 2014; Jia, 2019). Regionally, SSTs are influenced by the relatively fresh 472 PW and meltwater found near the coast and in some locations provide little informa-473 tion on the deeper AW temperatures on the continental shelf (Sutherland et al., 2013). 474 Our work, however, shows that with adjustments for diurnal and seasonal variability, SSTs 475 can provide a reliable measure of upper ocean temperatures, and, in the case of Sermi-476 lik Fjord, provide a measure of the AW temperatures entering a glacial fjord at depth. 477

In order to use MODIS skin SSTs in the polar regions as a measure of upper ocean 478 temperatures, additional processing and consideration needs to be given to the Level 3 479 R2014 SST products provided by the NASA Ocean Biology Processing Group (Jia, 2019). 480 Sea ice must be directly masked (Figure S2a), seasonal skewing by the daytime and night-481 time products in polar regions must be accounted for (Figure S1b), and diurnal biases 482 between day- and nighttime products corrected (Figure 2; Minnett, 2003; Kilpatrick et 483 al., 2015). We find that differences between the daytime and nighttime products can be 484 large here and daytime products skew observations towards the summer season when air 485

temperature changes dominate the SST signal. We also show that SST variability driven 486 by air-sea interactions in non-winter months can be removed to obtain an adjusted SST487 that is more closely linked with upper ocean temperature (Figure 4). This daytime/summertime 488 bias is important because most previous research using MODIS SSTs around Greenland 489 use daytime products (Howat et al., 2008, 2010; Schild & Hamilton, 2013; Sutherland 490 et al., 2013; Inall et al., 2014), and any research using summertime SSTs (e.g., Murray 491 et al., 2010; Andresen et al., 2012) is likely to be measuring an SST signal strongly tied 492 to atmospheric temperature, rather than the upper ocean. 493

For the ShTr region only, the resultant SST^{adj} is also found to be representative 494 of temperatures observed at 290 m at a mooring near the mouth of Sermilik Fjord, year-495 round. We attribute this link to the fact that the AW flowing into the trough are the 496 same which enter the fjord at depth, 80 km downstream, beneath the EGCC, consistent 497 with earlier studies. Our observations are consistent with the findings of Sutherland et 498 al. (2013) who find that 'uncorrected' summertime SST in the ShTr region do not sig-499 nificantly correlate with deeper trough water temperatures – where AW is primarily found 500 - making uncorrected SST unreliable for monitoring them. Once the higher frequency 501 imprinted atmospheric variability in non-winter months has been removed, however, the 502 relationship between these adjusted ShTr SST and subsurface water temperatures be-503 comes significant, albeit with more uncertainty than the respective wintertime relation-504 ship (winter RMSE=0.58°C, non-winter RMSE=0.71°C; Figure 4, Figure S4b). For this 505 reason, non-wintertime ShTr SST^{adj} can serve as a useful proxy for tracking subsurface 506 water temperatures flowing into the fjord as long as the higher uncertainty associated 507 with them does not exceed the variability in the ocean temperature signal. This would 508 likely make the non-wintertime subsurface AW estimates inadequate for locations that 509 experience less than a few degrees of water temperature variability. 510

While ShTr SST^{adj} serve as an estimate for the AW temperatures flowing into the 511 fjord, the connection to deeper waters varies across the region based on differences in strat-512 ification, mixing patterns, and water masses present, therefore it is unclear to what ex-513 tent SST^{adj} can serve as a proxy for subsurface waters in other locations. Many stud-514 ies have shown large horizontal changes in the properties of the water column between 515 the coast and offshore of the continental break near Sermilik Fjord (e.g., Rudels et al., 516 2002; Sutherland & Pickart, 2008; Harden et al., 2014). With a strong pycnocline be-517 tween PW and AW serving as a barrier between the surface and subsurface waters along 518

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the coast (Straneo et al., 2010; Harden et al., 2014), SST^{adj} over the EGCC are indica-519 tive of surface PW, and have a much weaker or insignificant connection to the AW flow-520 ing below (Table 1). This holds for the EGCC except when intrusions of water from the 521 IC mix horizontally into it, which is not uncommon (Sutherland & Pickart, 2008). Sea-522 sonal changes in the stratification and mixing that stem from changes in freshwater in-523 put (e.g., sea ice; Stroh et al., 2015), wind speeds, and solar heating (Donlon et al., 2002; 524 Minnett, 2003) will also impact the correlation between SST^{adj} and subsurface waters. 525 Therefore, the choice of SST sampling location heavily impacts what information can 526 be ascertained and the oceanography of each location must be well understood to use 527 SSTs reliably. 528

529

4.2 Historical Subsurface Water Temperatures and Implications

Our results provide two key insights about the Sermilik fjord/shelf system. First, 530 AW temperatures offshore within the IC are not necessarily indicative of coastal AW tem-531 peratures, which feed Sermilik Fjord. Instead, by the time it reaches the ShTr region, 532 AW is much colder than the IC box, indicating dilution as it crosses the continental shelf 533 (Figure 5). Second, we find that warmer waters intruded further inshore in the early 2000's 534 until early 2005 (consistent with the sediment-based reconstruction of Andresen et al. 535 (2012)), which generally corresponds to changes in discharge patterns at Helheim Glacier 536 (King et al., 2018). These combined observations indicate that the variability in AW tem-537 peratures found at depth nearshore result from an interplay of AW intruding onto the 538 shelf and EGCC water - and that the relative fraction of these vary in time. These find-539 ings also highlight that satellite-derived SSTs can provide previously unobserved con-540 text for spatially or temporally limited field measurements. 541

Variability observed in the Shelf Trough cannot be explained by IC and EGCC vari-542 ability taken separately - which represent the AW and PW end members - but is a time-543 varying combination of the two (Figure 5). Notably, we find that the ShTr SST^{adj} did 544 not always correlate with warmer AW in the IC from which the trough water is derived. 545 The ShTr SST^{adj} instead warms when our analyses show that waters with properties 546 similar to the IC intrude further onto the shelf (Figure 6). This linkage is most notable 547 in the vears when the ShTr SST^{adj} cooled while IC temperatures remained warm from 548 2005-2009 and after 2012 (Figure 6c). We infer that the varying dilution of AW as it crosses 549 the continental shelf controls the ShTr SST^{adj} . These findings suggest that while there 550

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is a direct connection between the North Atlantic Ocean and Sermilik Fjord (Straneo
et al., 2010; Andresen et al., 2012), the cooling of AW as it crosses the continental shelf
varies interannually, making offshore IC temperatures a poor indicator for the waters entering the fjord.

Within the 19-year record, months with the smallest differences between the ShTr 555 and IC SST^{adj} (Figure 5) are indicative of reduced AW dilution as it crosses the con-556 tinental shelf (Straneo et al., 2010; Moore et al., 2014). If we take ShTr temperatures 557 to be representative of the AW temperature entering the fjord, this suggests that the wa-558 ters flowing into the fjord at depth were similar to those in the IC in the early 2000's, 559 briefly in late 2009 to 2010, and in 2014-2018. These were also the years that exhibited 560 the least change in ShTr SST^{adj} across the shelf and, thus, when AW intruded furthest 561 onto the continental shelf (Figure 6c). We hypothesize that years with a more extended 562 intrusion of AW and warmer trough temperatures may also correspond with higher vol-563 ume transport, but that analysis is outside of the purview of this study. 564

The ShTr SST^{adj} record indicates that fluctuations in subsurface AW temperature 565 and intrusion coincide with some of the variability in discharge rates previously found 566 at Helheim Glacier for the same time period, but this relationship is not straightforward 567 (Figure 4b). AW spread inshore more consistently and the ShTr SST^{adj} was increasing 568 to their warmest values during the early 2000's when Helheim Glacier experienced ice 569 front retreat (Howat et al., 2005), thinning (Stearns & Hamilton, 2007), and heightened 570 discharge rates (Howat et al., 2007; King et al., 2018). The glacier also decelerated and 571 re-advanced from 2006-2008 (Howat et al., 2007; Schild & Hamilton, 2013) when the ShTr 572 SST^{adj} was the coldest on record, though notably offshore AW temperatures had not 573 measurably changed. In 2010, on the other hand, the ShTr SST^{adj} was relatively warm 574 and AW further intruded, although Helheim did not experience substantial increases in 575 ice discharge (King et al., 2018), which may have been driven by a host of other envi-576 ronmental factors influencing glacier discharge rates (e.g., air temperature, glacier con-577 figuration, mélange rigidity; e.g., Joughin et al., 2012; Carr et al., 2013). While more work 578 must be done to investigate mechanisms and the nature of these linkages, our work sup-579 ports the notion that warmer waters flowing into the fjord from the shelf trough may have 580 played a role in the glacier variability, especially in the early 2000's (e.g., Howat et al., 581 2008; Millan et al., 2018). 582

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The historical context we have constructed using both the ShTr SST^{adj} and trough 583 anomaly record likely have broader applications for understanding the shelf surface and 584 subsurface water temperatures. These waters directly feed the fjord of many deeply grounded 585 outlet glaciers in southeastern Greenland that may share similar AW sources and regional 586 forcings (Straneo et al., 2010; Harden et al., 2014; Jackson et al., 2014; Sutherland et al., 587 2014; Millan et al., 2018), including three of the largest contributors of ice discharge in 588 Greenland, Helheim Glacier, Kangerdlugssuaq Glacier, and Køge Bugt (Enderlin et al., 589 2014; King et al., 2018). The applicability of SSTs to specific ice-ocean questions, how-590 ever, has vast spatial and temporal variability, varies with the SST product used (i.e. day-591 time or nighttime), and depends heavily on the specific oceanography of the location be-592 ing explored. 593

594 5 Conclusions

We produce upper ocean temperature records for three regions on the continen-595 tal shelf near Sermilik Fjord using a composite of the MODIS Level 3 daytime and night-596 time SST R2014.0 products (Figure 2). We find that SST in the study regions has a monthly-597 varying linear relationship with air temperature that once adjusted for, produces a record 598 indicative of the upper ocean. The adjusted SST from our Shelf Trough region then have 599 a strong linear relationship with subsurface water temperatures from a mooring located 600 near Sermilik Fjord mouth at 290 m (Harden et al., 2014; Jackson et al., 2014), albeit 601 with higher uncertainty in the summer months (winter RMSE=0.58°C, summer RMSE=0.70°C). 602 This relationship confirms that AW in the Shelf Trough region is linked with subsurface 603 water - which ultimately continues inshore beneath the EGCC to the mooring location 604 where it feeds the fjord. Our records indicate that upper ocean temperatures in all three 605 regions, and at depth in the case of the Shelf Trough, were warmest in the early 2000's 606 when Helheim experienced rapid retreat, supporting previous ideas that ocean warm-607 ing played a role in the retreat. 608

Comparison of the upper ocean temperature variability in the three regions show that while there is a direct connection between the North Atlantic Ocean and the bathymetric trough leading to Sermilik Fjord (Straneo et al., 2010), the dilution of AW as it flows across the shelf from the IC varies substantially over long timescales. The extent to which AW intrudes onto the shelf correlates strongly with inferred subsurface AW temperatures on the inner-shelf indicating that this intrusion plays a key role in setting the

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properties of the heat-carrying waters that flow into Sermilik Fjord at depth. Inferences cannot be directly made between North Atlantic warming and AW changes on the continental shelf near Sermilik Fjord. These findings have important implications for models which seek to resolve ocean temperatures and transport paths within the region.

With proper consideration of the physical processes driving the measurements, SSTs are a relatively untapped tool that show promise in applications to a vast range of polar oceanography and glaciology questions. Further work will continue to expand contextual understanding around the Greenland and Antarctic ice sheets both where longstanding field measurements have been acquired and where none exist.

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