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### Assessing the Effects of Fjord Geometry on Greenland Tidewater Glacier Stability

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Abstract:	<p>Tidewater glaciers advance and retreat in unpredictable ways, complicating the task of forecasting the evolution of individual glaciers and the overall Greenland Ice Sheet, much of which is drained by tidewater glaciers. Past observational research has identified a set of processes collectively known as the Tidewater Glacier Cycle (TGC) to describe tidewater glacier evolution in four stages: the advancing stage, the extended stage, the retreating stage and the retreated stage. Once glacier retreat is initiated, the TGC is thought to depend largely on the glacier's calving rate, which is controlled by fjord geometry. However, there has been little modeling or systematic observational work on the topic.</p> <p>Measuring calving rates directly is challenging and thus we developed an averaged von Mises stress state at the glacier terminus as a calving rate proxy that can be estimated from surface velocities, ice thickness, a terminus position, and subglacial topography. We then analyzed 44 tidewater glaciers in Greenland and assessed the glaciers current state in the TGC. Of the 44 glaciers, we find that fjord geometry is causing instability in 10 cases, vs. stability in 7, with 11 in rapid retreat and 16</p>

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# Assessing the Effects of Fjord Geometry on Greenland Tidewater Glacier Stability

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**ABSTRACT.** Tidewater glaciers advance and retreat in unpredictable ways, complicating the task of forecasting the evolution of individual glaciers and the overall Greenland Ice Sheet, much of which is drained by tidewater glaciers. Past observational research has identified a set of processes collectively known as the *Tidewater Glacier Cycle* (TGC) to describe tidewater glacier evolution in four stages: the *advancing stage*, the *extended stage*, the *retreating stage* and the *retreated stage*. Once glacier retreat is initiated, the TGC is thought to depend largely on the glacier's calving rate, which is controlled by fjord geometry. However, there has been little modeling or systematic observational work on the topic.

Measuring calving rates directly is challenging and thus we developed an averaged von Mises stress state at the glacier terminus as a calving rate proxy that can be estimated from surface velocities, ice thickness, a terminus position, and subglacial topography. We then analyzed 44 tidewater glaciers in Greenland and assessed the glaciers current state in the TGC. Of the 44 glaciers, we find that fjord geometry is causing instability in 10 cases, vs. stability in 7, with 11 in rapid retreat and 16 have been historically stable.

## 23 1. INTRODUCTION

24 It has long been known that tidewater glaciers advance and retreat out of sync with land terminating glaciers  
25 and external ocean and climate forcing (Post, 1975; Meier and Post, 1987; Pfeffer, 2003). This observation  
26 has led to the formulation of the *Tidewater Glacier Cycle* (TGC) (Pfeffer, 2007; Pollard and DeConto,  
27 2009; Brinkerhoff and others, 2017; Post and others, 2011), a combination of processes that proceed in  
28 four archetypal phases, as described in Brinkerhoff and others (2017). In the advancing stage, development  
29 and advection of a shoal at the front reduces calving and submarine melting, causing glacier thickening  
30 and advance. Eventually the glacier enters an extended phase, in which accumulation and ablation are  
31 in balance and further advance is halted. A glacier enters the retreating phase when it can no longer  
32 maintain sufficient thickness to remain grounded on the shoal and the glacier retreats into progressively  
33 deeper water; at which point dramatic unstable rapid retreat takes place. Retreat ends when the terminus  
34 approaches a position that reduces the calving in the absence of sedimentation; possibly at a *pinning point*  
35 (temporary narrowing of the fjord), or else the terminus effectively re-grounds on bedrock. Multi-decadal  
36 modeling studies with sedimentation are able to reproduce this cycle (Nick and others, 2007; Amundson,  
37 2016; Brinkerhoff and others, 2017) even in the absence of variations in climate (Brinkerhoff and others,  
38 2017).

39 Typical timescales and rates of advance / retreat for tidewater glaciers is an area of active research.  
40 Catania and others (2018) report rates of retreat for Greenland tidewater glaciers of up to  $500 \text{ m a}^{-1}$ .  
41 Brinkerhoff and others (2017) shows a simulated retreat phase lasting about 100 years and advance phase  
42 about 1000 years with the terminus advancing or retreating about 5 km in that time ( $50 \text{ m a}^{-1}$  retreat and  
43  $5 \text{ m a}^{-1}$  advance). Carlson and others (2017) reports the Columbia Glacier in Alaska has retreated approx.  
44 20 km in 30 years ( $667 \text{ m a}^{-1}$ ). Pearce and others (2022) reports an advance rate of Kangiata Nunaata  
45 Sermia of  $115 \text{ m a}^{-1}$  during the Little Ice Age (12th and 13th centuries CE). In general, the retreat phase  
46 of the TGC is thought to happen on a decadal or centennial timescale, and the advance phase is about an  
47 order of magnitude slower.

48 In this observational study, we focus on detecting glaciers beginning, sustaining or finishing the retreat  
49 phase using the ITS-LIVE dataset from 1985 through 2018, or 33 years (Gardner and others, 2019). We use  
50 *annual* terminus positions, thereby sidestepping the complex issue of seasonal variability. Our timeseries  
51 are not long enough to investigate the advance phase. Sedimentation is an important part of the TGC at



52 the century time scale (Brinkerhoff and others, 2017); but one we can safely ignore in this study covering  
 53 rapid retreat of numerous glaciers over a few decades.

54 Interestingly, the TGC suggests that tidewater glacier termini can only remain in stable equilibrium,  
 55 neither advancing or retreating, at places where further advance would cause a negative feedback: at fjord  
 56 mouths, at *pinning points* (temporary narrowing of the fjord), and at other places involving change in the  
 57 fjord width (Mercer, 1961).

58 Greenland's outlet glaciers are currently at diverse stages in the tidewater glacier cycle. For example,  
 59 after advancing 800 m from 1973 to 2000, Sermeq Silarleq in central west Greenland retreated 5 km from  
 60 2000 through 2019; but just 47 km to the south, Store Glacier has remained remarkably stable during the  
 61 same time frame (Cheng and others, 2021b).

62 Many glaciers around Greenland have been retreating since 2000 (Murray and others, 2015), suggesting  
 63 that conditions required for TGC advance are currently rare. While the TGC has been used as a post-  
 64 hoc explanation for advancing (McNeil and others, 2021) and retreating glaciers, it has not been used to  
 65 systematically investigate how a glacier's current behavior reflects the TGC phases. Quantifying where a  
 66 glacier currently falls in the TGC could help to understand how it might respond to future environmental  
 67 change.

### 68 1.1. Basics of Advance and Retreat

69 Wood and others (2021) present a model across Greenland in which the position  $L$  of a tidewater glacier's  
 70 grounded terminus is a result of four competing processes causing advance or retreat: advection of ice  
 71 downstream ( $q_f$ ) leads to terminus advance, whereas frontal melt ( $q_m$ ), calving ( $q_c$ ) and thinning-induced  
 72 retreat ( $q_s$ ) (Felikson and others, 2017) lead to terminus retreat. Adopting the convention that positive  
 73 sign means advancing terminus for all advance/retreat rates (units  $\text{m s}^{-1}$ ), mass balance at the ice front  
 74 requires:

$$\Delta L = L - L_0 = \int_{t_0}^t (q_f + q_m + q_c + q_s) dt, \quad (1)$$

75 where  $L$  is the current terminus position,  $L_0$  is the terminus position at a reference time  $t_0$  and  $t$  is the  
 76 current time. The supplement from Wood and others (2021) shows that  $q_s$  is at least an order of magnitude  
 77 smaller than  $q_c$  for the glaciers in this study, and we can therefore ignore it. Wood and others (2021) observe  
 78 or model all the terms of Equation 1 *except* for  $q_c$ .

79 Wood and others (2021) use a parameterization for frontal melt, similar to Slater and others (2019), in  
80 which subglacial discharge and thermal forcing are both derived from an ensemble of *MITgcm* runs (Rignot  
81 and others, 2016; Xu and others, 2013): because ocean grid cells are too large to resolve fjords, model-  
82 based ocean temperatures at the mouth of each fjord are translated into temperatures inside the fjord at the  
83 calving front.<sup>1</sup> Thinning-induced retreat  $q_s$  is calculated using a simple, geometrically derived relationship  
84 for grounding line migration rate as a function of surface elevation change (Thomas and Bentley, 1978).

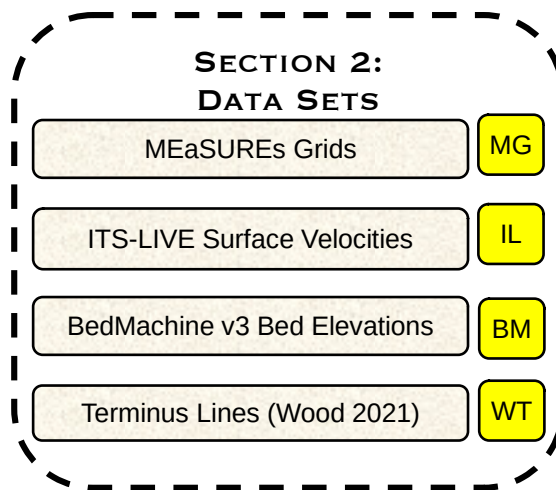
85 Disregarding frontal melt for a moment, stability may be investigated systematically by evaluating the  
86 relationship between terminus position and calving rate  $q_c$  for each glacier. If a currently stable glacier is  
87 about to enter the retreat phase of the TGC, then  $q_c$  would be expected to increase as the glacier begins to  
88 retreat, potentially leading to runaway retreat; whereas if the glacier is stabilizing,  $q_c$  would be expected  
89 to decrease as the glacier retreats, causing retreat to slow. One could identify glaciers that are about to  
90 enter or exit the rapid retreat phase of the TGC by investigating the correlation between observed changes  
91 in  $q_c$  versus observed changes in terminus position.

92 Measuring or modeling all four components of Equation 1 individually would allow us to do just  
93 that. However, such observations are challenging because of limitations in observational capabilities and  
94 uncertainties in process models. While calving rates can be measured with localized systems (Taylor and  
95 others, 2022; Walter and others, 2020), earth observing satellites do not provide high enough resolution in  
96 time to apply these techniques over a wide region. For this reason, a proxy  $\bar{\sigma}_T$ , representing relative levels  
97 of expected  $q_c$  and computable from remote sensing data, is used to evaluate glacier stability instead of the  
98 calving rate  $q_c$ . This proxy relies on the *von Mises* calving law (von Mises, 1913; Morlighem and others,  
99 2016) as a simple but reasonable model for tidewater glaciers (Section 4.1).

100 Even with a reliable proxy for  $q_c$ , efforts to find a relationship between  $q_c$  and observed retreat  $\Delta L$  will  
101 fail: due to warming in the ocean around Greenland, frontal melt has recently become the dominant process  
102 driving retreat (Slater and others, 2019; Wood and others, 2021), with calving playing a secondary role.  
103 Most glaciers are retreating, and we cannot immediately conclude that observed retreat is due to tidewater  
104 glacier instability. Some glaciers continue to retreat even as they move into shallower water, for example  
105 Lille Glacier (Figure 15). In order to observe a correlation between glacier retreat and calving rate, it is  
106 first necessary to estimate and remove the amount of retreat caused by changes in frontal melt rate  $q_m$ .  
107 We use the model of Slater and others (2019) for that task.

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<sup>1</sup>See section titled *Thermal forcing* in Wood and others (2021), page 8 of 10.



**Fig. 1.** Data sets used in this study. Each data set is represented by a yellow tag, used in Figure 2. See Section 2 for further details.

## 108 1.2. Methodology

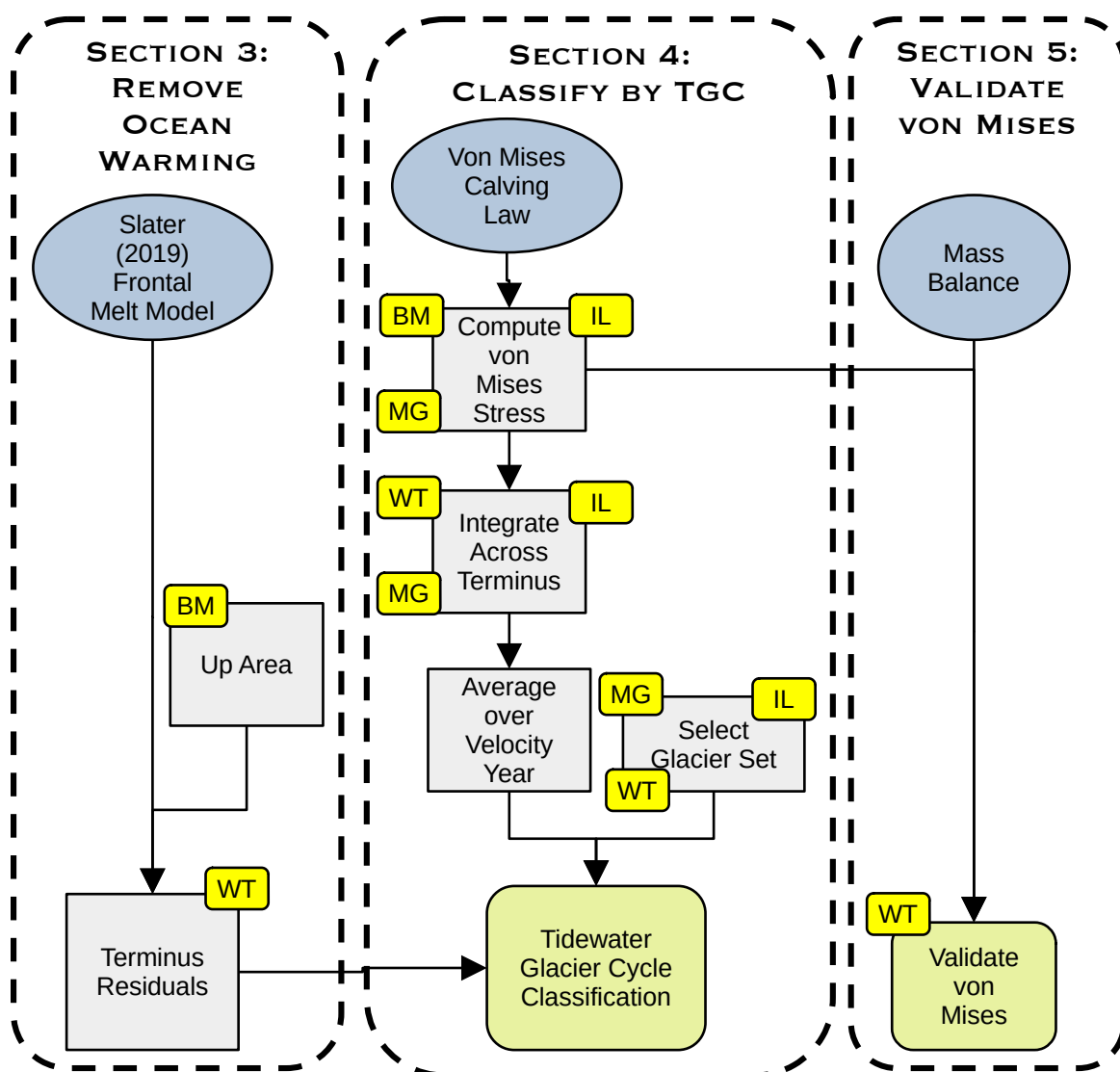
109 We develop an averaged proxy  $\bar{\sigma}_T$  for the calving rate  $q_c$ , which can be computed from readily-available  
 110 observations of surface velocities, ice thickness, a terminus line and subglacial topography. The proxy is  
 111 derived from the *von Mises* calving law (von Mises, 1913; Morlighem and others, 2016), which this study  
 112 shows in Section 5 can be tuned to be consistent with observations.

113 Values of  $\bar{\sigma}_T$  and  $L$  in the recent history of the glacier are regressed against each other, allowing diagnosis  
 114 of the glacier's stability and current stage in the TGC. In this study we use *annual* averages of  $L$ , thereby  
 115 sidestepping issues of seasonal melting driven by submarine discharge; and we also remove the dominant  
 116 effects of ocean warming. If  $\bar{\sigma}_T$  is found to *decrease* as the glacier retreats, then the fjord geometry at that  
 117 point in space is destabilizing, providing evidence that the glacier may be entering the retreat phase of the  
 118 TGC; whereas if  $\bar{\sigma}_T$  *increases* the opposite is true and the fjord geometry is now stabilizing the glacier,  
 119 suggesting the glacier may be finishing the retreat phase of the TGC and moving up onto land.

120 If  $\bar{\sigma}_T$  remains the same as the glacier retreats, the glacier may be retreating through a section of the  
 121 fjord with nearly constant cross sectional width and depth, typical for glaciers in the middle of rapid  
 122 retreat. Finally, some glaciers have remained stable in recent years and it is not possible to tell strictly  
 123 from observations whether  $\bar{\sigma}_T$  would go up or down if the terminus were to retreat.

## 124 1.3. Paper Overview

125 The methods in this paper involve an existing set of grid definitions, four datasets, three theoretical models,  
 126 and 8 interconnected computational / statistical procedures to produce two final results: a classification of



**Fig. 2.** Models and methods used in this paper: blue ovals are theoretical models, grey rectangles are methods; and green rounded rectangles are methods that produce an end result of this study. Arrows represent dependencies, for example *Up Area* values (Section 3.2) are required to produce *Terminus Residuals*. Section 3 presents the frontal melt model by Slater and others (2019) driven by ocean warming, and uses it to remove effects of ocean warming from terminus data, resulting in terminus residuals. Section 4 introduces the von Mises Calving Law and derives  $\bar{\sigma}_T$ , a proxy for calving rate, which it regresses against terminus residuals to provide diagnostics on tidewater glaciers. Section 5 uses the data to show why the von Mises calving law is a reasonable model.

127 glaciers based on their current stage in the TGC, and validation that the von Mises calving criterion  
 128 is a reasonable model. The models and methods required to attain these conclusions have complex  
 129 interdependencies, illustrated in an organizational chart in Figure 2 and supported by datasets in Figure 1.  
 130 Because of these interdependencies, there is no single obvious way to order them. We have ordered sections

Introduction	
$\sigma_T$	Average von Mises stress across a glacier terminus $T$
$\bar{\sigma}_T$	Average of $\sigma_T$ for one terminus $T$ over different years' velocity fields
$\bar{L}$	Generic scalar terminus position
$q_f$	Rate of advection of ice downstream (positive number)
$q_c$	Rate of calving (negative number)
$q_m$	Rate of frontal melt (negative number)
$q_s$	Rate of thinning-induced retreat (negative number)
Model of Slater and others (2019)	
$Q$	Subglacial discharge due to surface meltwater runoff and basal melt
TF	Thermal forcing in the fjord
$l_s$	Observed scalar terminus position
$l_p$	Modeled scalar terminus position
$l_p = \kappa q_m + \beta$	Empirical predictive relationship for $l_p$ based on $q_m = Q^{0.4}TF$
Measurement of Terminus	
$A_T$	Area of fjord upstream of terminus $T$
$A_T = wl_s + b$	Empirical relationship between $A_T$ and $l_s$
$l_\epsilon = l_p - l_w$	Terminus residual of observations vs. Slater and others (2019) model
$\vec{n}$	Unit normal to terminus line
Von Mises Model	
$\tilde{\sigma}$	Scalar von Mises tensile stress
$\vec{i}$	Vertically averaged horizontal surface velocity of a glacier
$\dot{\epsilon}$	Strain rate tensor
$\dot{\epsilon}$	Scalar tensile strain rate (Morlighem and others, 2016)
$\dot{\epsilon}_1, \dot{\epsilon}_2$	Eigenvalues of $\dot{\epsilon}$
$\sigma$	Deviatoric (shear stress) tensor where $\dot{\epsilon} = \tilde{A}\sigma^n$
$\tilde{A}$	Temperature-dependant rate factor with units $s^{-1}Pa^{-n}$
$\tilde{B} = \tilde{A}^{-1/n}$	Ice hardness (Greve and Blatter, 2009)
$n$	Constitutive relation exponent typically assumed to be 3 (Behn and others, 2021)
$\sigma_{max}$	Ice yield strength (kPa)
Calving vs. Retreat	
$\Delta l_\epsilon = \nu \Delta \bar{\sigma}_T$	Empirical relationship between $l_\epsilon$ and $\bar{\sigma}_T$ , disregarding the y-intercept.
$R^2$	Amount of variance in $l_\epsilon$ explained by model
$p$	Goodness of fit for model
Validation of von Mises Calving	
$L - L_0$	Change in terminus position between reference time $t_0$ and current the time $t$
$dL/dt$	Rate of terminus advance or retreat
Line Integrals on a Grid	
$A, B$	Two adjacent gridcells covered in ice
$u, v$	$x$ and $y$ components of horizontal vertically averaged ice velocity
$m$	A mask identifying gridcells with west-to-east (or south-to-north) flux

**Fig. 3.** Symbols used in this paper, organized by section where they are introduced.

131 such that later material builds on earlier material; however, the reader may wish to read the sections in a  
 132 different order, depending on their interests.

133 Figure 3 lists all symbols used in this paper. Based on Figure 2, the rest of this paper is organized as  
 134 follows: Section 2 presents the four datasets used in this study, which are used by the methods in future  
 135 sections. Section 3 presents the frontal melt model of Slater and others (2019), and describes its use to  
 136 remove the effects of frontal melt from the terminus data of Wood and others (2021), resulting in *terminus*  
 137 *residuals*. Section 4 presents the von Mises calving law (von Mises, 1913; Morlighem and others, 2016;  
 138 Choi and others, 2018); and presents a series of steps that, when combined with the terminus residuals of  
 139 Section 3, allow for classification of glaciers within the TGC. Section 6, along with the Supplement, shares

140 results. Section 7 and Section 8 synthesize and discuss the results further. Finally, Appendix A documents  
141 a novel numerical technique, developed in conjunction with this study, that may be applicable for other  
142 projects.

## 143 2. DATA SETS

144 We used the following grids and data sets. The reader may wish to skip to Section 3 and refer back to this  
145 section as needed.

146 **Local MEaSURES Grids:** Local grids (resolution and extent), as defined by the *MEaSURES*  
147 *Greenland Ice Velocity* dataset NSIDC-481 (Joughin and others, 2010, 2020), which we re-use for our  
148 analysis.

149 **ITS-LIVE Surface Velocities:** Annual average surface velocities (advection rate) from 1985 through  
150 2018, necessary to compute the von Mises stress, were obtained from the ITS-LIVE dataset (Gardner  
151 and others, 2019) and regridded to the local MEaSURES grids.

152 **BedMachine v3 Subglacial Topography:** Subglacial topography required for this computation was  
153 provided by BedMachine v3 (Morlighem and others, 2017) and regridded to the local MEaSURES grids.

154 **Terminus Lines:** The scalar terminus positions of Slater and others (2019) are cross-referenced against  
155 information obtained from the two-dimensional terminus lines of Wood and others (2021).

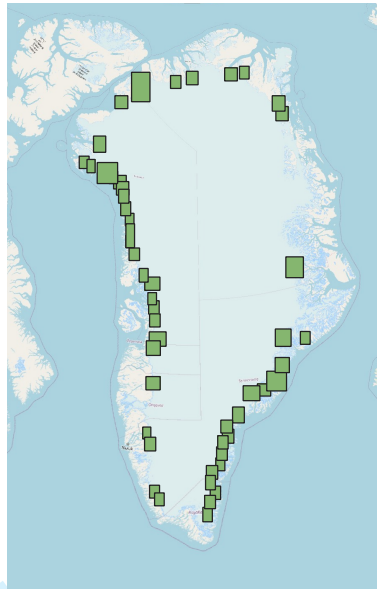
156 **Modeled Frontal Melt:** The models of Slater and others (2019) are used to predict frontal melt based  
157 on ocean warming.

158 **Selection of Glacier Set:** We used glaciers present in the datasets provided by *three* previous studies:  
159 Slater and others (2019), Wood and others (2021), Gardner and others (2019); and falling within grids  
160 from Joughin and others (2010).

161 Those datasets are further examined below.

### 162 2.1. Local MEaSURES Grids

163 The *MEaSURES Greenland Ice Velocity* dataset NSIDC-481 (Joughin and others, 2010, 2020) has already  
164 been constructed to cover many Greenland glaciers, with local high-resolution grids defined in areas with  
165 glacier activity (Figure 4). Regridding the other datasets (below) to these local grids allows for detailed



**Fig. 4.** Local high-resolution grids (green rectangles) defined by the MEaSUREs dataset, NSIDC-0481.

166 study of individual glaciers while omitting most of the interior of the ice sheet. They also allow for cross-  
167 referencing with other datasets and studies that also use the same grids.

168 Each glacier in our analysis was identified as falling on a single local grid from NSIDC-481 (a *MEaSUREs*  
169 *grid*). For glaciers located on more than one MEaSUREs grid, the most appropriate grid for that glacier  
170 was determined by hand based on the distance from the center of the grid for each glacier. Glaciers that  
171 did not fall within a MEaSUREs grid were removed from the selection.

## 172 2.2. ITS-LIVE Surface Velocities

173 Annual average surface velocities (advection rate) from 1985 through 2018, necessary to compute the von  
174 Mises stress, were obtained from the ITS-LIVE dataset (Gardner and others, 2019), and regridded to the  
175 local MEaSUREs grids. An annually averaged dataset was used to avoid complexities of seasonality, surges  
176 and short-term variability; however, use of a higher temporal resolution dataset for surface velocities might  
177 produce improved statistical precision. Mouginit and others (2017) also provide surface velocity datasets  
178 derived from satellite InSAR or optical images, which might be useful in similar future studies, for example  
179 in Antarctica.

## 180 2.3. BedMachine v3 Subglacial Topography

181 Subglacial topography required for this computation was provided by BedMachine v3 (Morlighem  
182 and others, 2017) and regridded to the local MEaSUREs grids. BedMachine was chosen as our best



183 understanding of the bed underneath the Greenland Ice Sheet, given available data and models. Knowledge  
184 of the bed is typically least certain near each glacier terminus.

## 185 2.4. Terminus Lines

186 Terminus positions may be computed from satellite images by tracing the terminus, either manually (Wood  
187 and others, 2021) or via machine learning (Cheng and others, 2021a; Goliber and others, 2022), with newer  
188 machine-learning approaches greatly expanding the quantity of available terminus traces. Wood and others  
189 (2021) provide one of the two main theoretical models for this study (Section 1.1), and we re-use terminus  
190 traces from it to maintain compatibility.

191 Slater and others (2019) provide the other main theoretical model for this study, but it represents terminus  
192 positions as a one dimensional scalar distance up the fjord. This implicitly assumes a specific model of a  
193 fjord as a long narrow channel with length much greater its approximately uniform width, which is not  
194 always reasonable. There is no simple automated way to delineate center lines, and some fjords have  
195 complex geometry not well described by a simple one dimensional model.

196 Therefore, spatial analysis in this study is conducted on a full two dimensional map of the fjord.  
197 For compatibility, the scalar terminus positions of Slater and others (2019) are cross-referenced against  
198 information obtained from the two-dimensional terminus lines of Wood and others (2021).

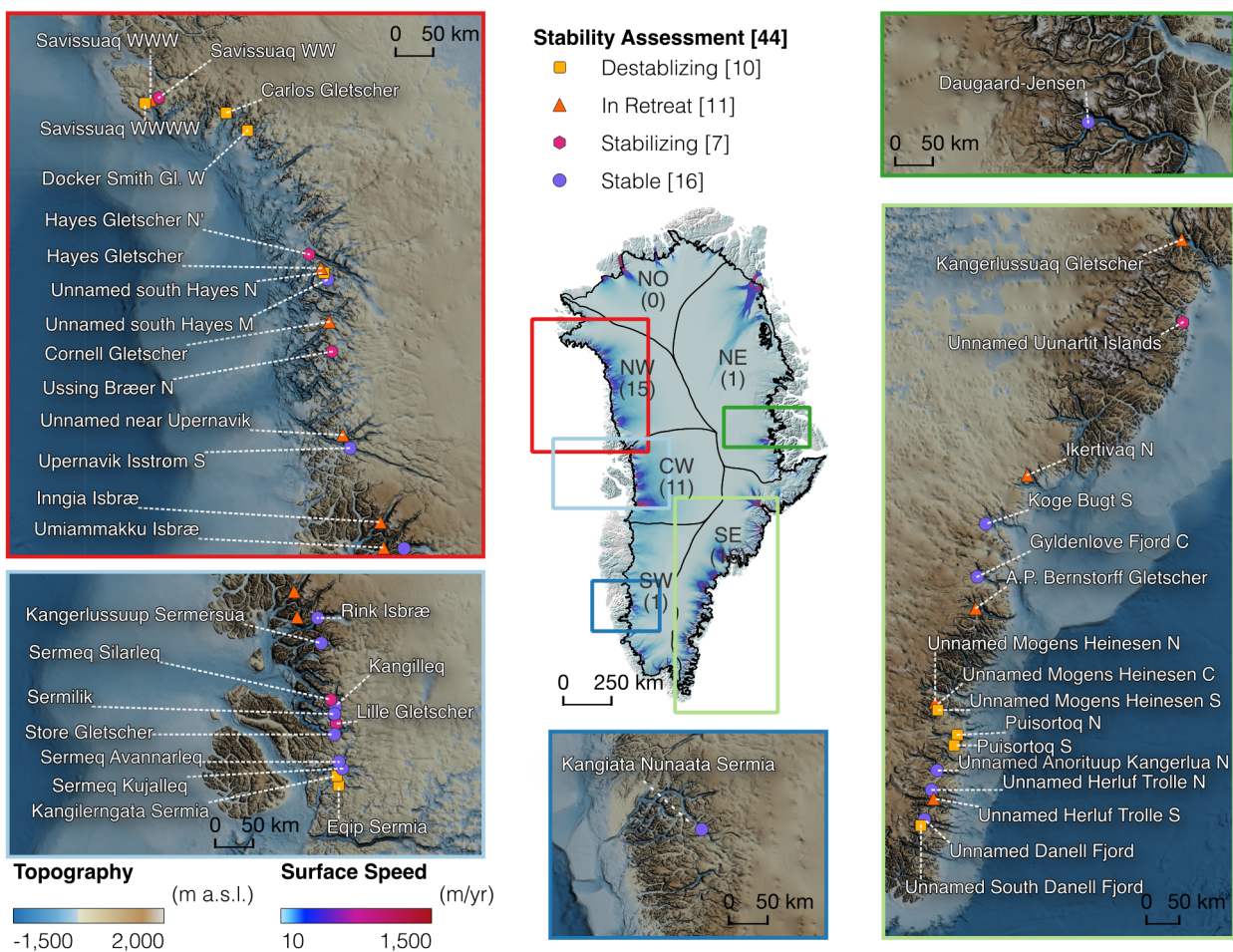
## 199 2.5. Modeled Frontal Melt

200 We use two data sets from Slater and others (2019): annual scalar terminus positions  $l_s$  and annual frontal  
201 melt rate  $q_m$ , modeled as  $q_m = Q^{0.4}TF$ , where  $Q$  is subglacial discharge due to surface meltwater runoff and  
202 basal melt and  $TF$  is the thermal forcing in the fjord. These values are based on model ocean outputs of  
203 *MITgcm* (Adcroft and others, 2004). The model of Slater and others (2019) may significantly underestimate  
204 submarine melting (Sutherland and others, 2019; Catania and others, 2020; Jackson and others, 2022); but  
205 at first order we do not expect that to affect results because frontal melt is empirically calibrated to  
206 observations (section 3.1).

## 207 2.6. Selection of Glacier Set

208 As described above, this study uses models from Slater and others (2019), data from Wood and others  
209 (2021) and Gardner and others (2019), and grids from Joughin and others (2010). Therefore glaciers need  
210 to be present in *all four* datasets, resulting in a set of 44 glaciers available for the study as shown in the  
211 results (Fig. 5). Although this procedure reduces the number of glaciers for analysis, it maximizes the





**Fig. 5.** Location and stability assessment of the 44 Greenland tidewater glaciers in this study. Of the 44 glaciers, 9 glaciers are stable, 7 are stabilizing, 13 are destabilizing, and 15 are in retreat. Subglacial topography is from BedMachine v3 (Morlighem and others, 2014) and surface speeds from ITS-LIVE (Gardner and others, 2019).

212 ability to compare and cross-reference results with previous studies. Geographic representation of glaciers,  
 213 classifying by regions as defined by Wood and others (2021), is: central-west Greenland (11 of 14 total  
 214 tidewater glaciers), north-east (1 of 14), north-west (15 of 64), south-east (16 of 56), and south-west (1 of  
 215 12). This study has no geographic representation in the central-east (35 total) or north (12 total)  
 216 of Greenland.

217 We note that the data sets we used do not include many glaciers in the south-west of Greenland.

### 218 3. FRONTAL MELT

#### 219 3.1. Frontal Melt Model

220 Slater and others (2019) state that warming oceans are currently the primary driver of tidewater glacier  
 221 retreat in Greenland. Based on data, they provide a glacier-by-glacier relationship between the change of  
 222 the scalar terminus position  $l_p$  and frontal melt rate  $q_m$ , i.e. they empirically derive  $\kappa$  and  $\beta$ , based on data  
 223 averaged over 5-year intervals such that:

$$l_p = \kappa q_m + \beta. \quad (2)$$

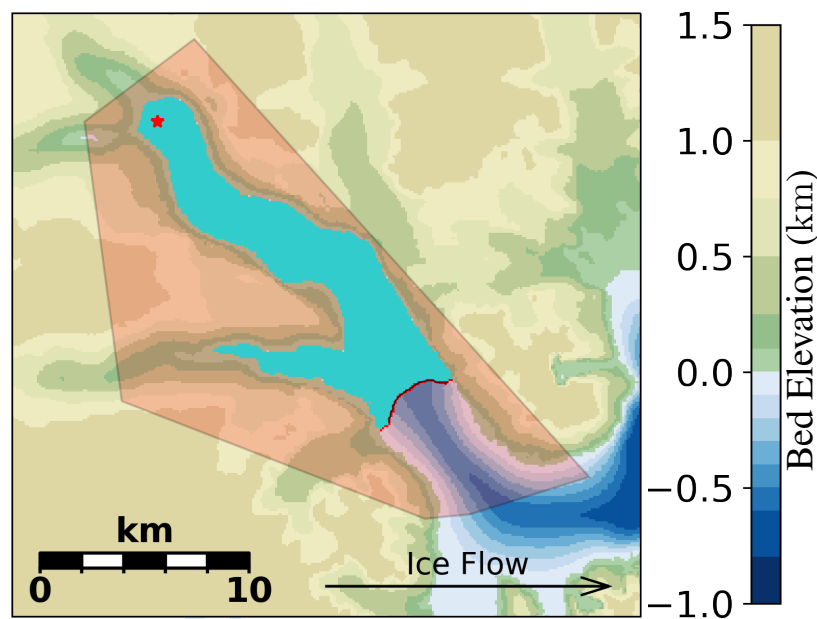
224 This relationship is based on *the process of ice front undercutting / frontal melt only*, modeled because  
 225 it cannot be observed directly via remote sensing. The value  $q_m = Q^{0.4}TF$  from Slater and others (2019)  
 226 represents the ocean heat available to drive melting. As a proxy for subglacial discharge  $Q$ , Slater and others  
 227 (2019) used surface meltwater runoff estimated by the regional climate model RACMO2 (Noël and others,  
 228 2018); and for TF, they used the monthly EN4 dataset from the Hadley Center consisting of observed  
 229 subsurface ocean temperature and salinity profiles (Good and others, 2013).

230 The linear model of Slater and others (2019) incorporates frontal melt from ocean warming but ignores  
 231 the calving effects due to glacier geometry. Glaciers close to each other will experience similar changes in  
 232 ocean temperature, but different fjord geometry could cause them to behave differently in spite of similar  
 233 ocean forcing. Therefore, the model can predict advance or retreat of glaciers as a whole within a region  
 234 due to ocean warming, but cannot predict the behavior of individual glaciers, which also depends on fjord  
 235 geometry.

236 In recent years, ocean warming has become the dominant process causing glaciers in this study to retreat  
 237 (Slater and others, 2019; Wood and others, 2021). In order to study the secondary effect of fjord geometry,  
 238 the effects of the dominant process must first be removed from the data. We use the model of Slater and  
 239 others (2019) to estimate the amount of retreat caused by ocean warming and subtract that out of the total  
 240 retreat, leaving a *terminus residual* (Section 3.3) in which retreat due to calving is the dominant process.

#### 241 3.2. Up Area

242 Spatial analysis in this study is conducted on a full two dimensional map of the fjord. In place of a scalar  
 243 terminus position  $L$ , the scalar *up area*  $A_T$  is used, defined as the entire ice-covered area upstream of the  
 244 glacier terminus  $T$  for which the basal topography is below sea level. This avoids assumptions about fjords,  
 245 their linear geometry or center lines.



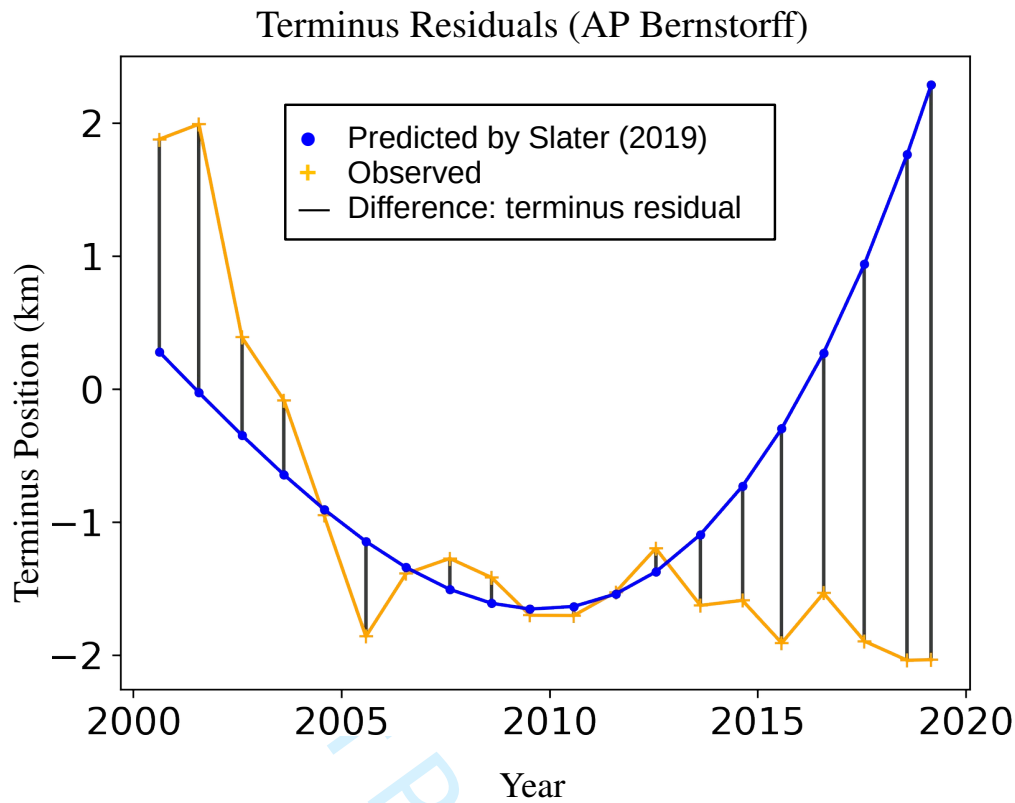
**Fig. 6.** Aerial map of AP Bernstorff Glacier in Southeast Greenland, with terminus as of year 2005. Digitized terminus datasets typically come in vector format (black line on top of red gridcells), which is rasterized (red grid cells). To help the computer determine the extent of the fjord, we drew a rough polygon around the fjord by hand (red shaded area), and identified a point (red star) that is upstream of *all* expected termini used in this study. Based on these inputs and bathymetry from BedMachine, the computer was able to delineate the extent of the fjord (green) as those gridcells that are below sea level and reachable from the identified point via flood fill.

246 The up area is calculated as follows (Figure 6). Using GIS software, a polygon is drawn around the  
 247 fjord by hand, and a single point identified in the upper reaches of the fjord (the *up point*). The fjord  
 248 is determined in raster form by identifying gridcells within the polygon with bed below sea level. The  
 249 terminus line is extended to the full width of the fjord and rasterized, to produce the set of gridcells *on* the  
 250 terminus. A raster *flood fill* algorithm is then used, starting from the up point, to identify all the grid cells  
 251 of the fjord that are upstream of the terminus. The *up area* is computed by summing the areas of these  
 252 upstream grid cells.

253 Because of the way polygons are drawn,  $A_T$  does not include far-upstream areas of some fjords. Therefore,  
 254 *up area* may be used for relative comparison between termini, but not as an absolute measure of how much  
 255 fjord “remains” ice covered before a glacier becomes land terminating.

### 256 3.3. Terminus Residuals

257 We examine the relationship between fjord geometry and glacier retreat due to calving rates, an effect  
 258 that Slater and others (2019) determined to be secondary to ocean warming. In order to see this effect in



**Fig. 7.** Computation of the *terminus residual* for AP Bernstorff glacier. **Blue dots:** terminus positions as predicted by thermal forcing model from Slater and others (2019). Annual predictions are available because annual thermal forcing estimates are available; however, note that the Slater model coefficients are determined based on regressions involving 5-year averaged data. **Orange pluses:** terminus positions based on *up area* calculated from termini in Wood and others (2021) and calibrated to terminus positions from Slater and others (2019). **Black lines:** The terminus residual is the difference between the two predictions. The increasingly negative terminus residual means the glacier is retreating faster than Slater and others (2019) would predict based on thermal forcing alone, indicating a destabilizing influence of fjord geometry. The *Fjord Map* for this glacier (Figure 17) confirms that runaway retreat is well underway.

259 the data, it is essential to remove the dominant effect of ocean warming. This takes place in two steps:  
 260 calibration and computation.

#### 261 *Calibration and Computation*

262 This study describes observed terminus state based on data from Wood and others (2021) using *up area*  
 263  $A_T$  (Section 3.2); whereas Slater and others (2019) describes observed terminus state as a linear position  
 264  $l_s$  along a center line. Assuming a constant fjord width  $w$ , there is a linear relationship between the two:

$$A_T = wl_s + b. \quad (3)$$

265 We determine  $w$  and  $b$  empirically via linear regression, where  $b$  is an arbitrary constant that depends  
 266 on zero points chosen for  $l_s$  and  $A_T$ . These coefficients are then used to convert observed *up area*  $A_T$  to  
 267 observed  $l_w$ , an effective terminus position calibrated to the same scale and offset used in Slater and others  
 268 (2019):

$$l_w = (A_T - b)/w. \quad (4)$$

269 Using Equations 2 and 4, the predicted terminus position  $l_p$  (Slater and others, 2019) *based solely on*  
 270 *observed advection vs increases in frontal melt due to ocean warming* may be compared to the observed  
 271 terminus position  $l_w$ , which is based on all processes affecting terminus position (Equation 1). We compute  
 272 the *terminus residual*  $l_\epsilon$  (Figure 7) as:

$$l_\epsilon = l_p - l_w = \kappa q_m + \beta - \frac{A_T - b}{w}. \quad (5)$$

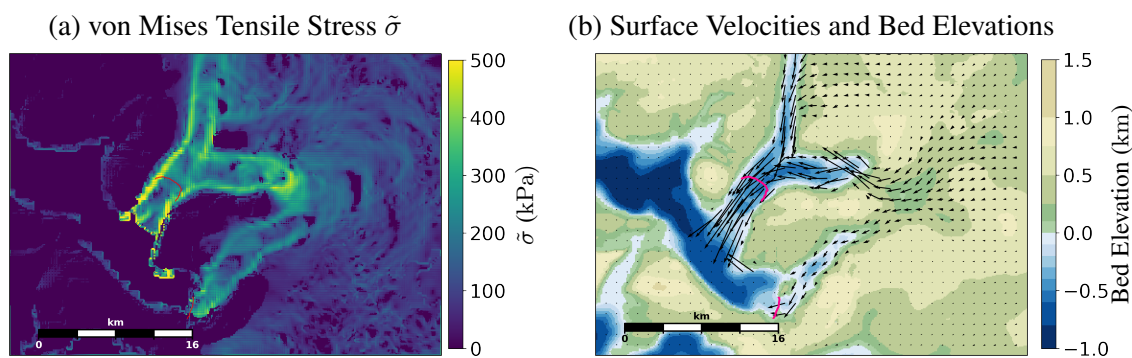
273  $l_\epsilon$  will be affected by all processes *except* advection and frontal melt: calving ( $q_c$ ) and thinning-induced  
 274 retreat ( $q_s$ ). The supplement from Wood and others (2021) shows that  $q_s$  is at least an order of magnitude  
 275 smaller than  $q_c$  for these glaciers. Therefore to first order,  $l_\epsilon$  is an estimate of advance or retreat due to  
 276 decreases or increases in calving.

277 Observations show that some glaciers have been stable, for example *Rink Isbræ* and *Sermeq Avannarleq*  
 278 (Figure 16). Even though they have been stable overall, their termini have still advanced or retreated by  
 279 up to 600 m over the study period; where total retreat is computed based on a least squares fit through the  
 280 annual terminus locations. Therefore, total retreat of less than 600 m over the study period of 1980–2020 is  
 281 considered not significant because that is within the common range of natural variability for stable glaciers  
 282 in this study, and the terminus has had little opportunity to “sample” different portions of the fjord. Note  
 283 that most glaciers in our study that are retreating today only began to do so around the year 2000 (Murray  
 284 and others, 2015).

#### 285 4. CLASSIFICATION BY TGC STAGE

286 The current stage in the TGC for an individual glacier may be evaluated by computing the calving proxy  
 287  $\bar{\sigma}_T$  and regressing it against the terminus residuals (Section 3.3). Derivation of  $\bar{\sigma}_T$ , based on the von Mises  
 288 Calving Law, follows below.





**Fig. 8.** (a) von Mises tensile stress  $\tilde{\sigma}$  shown for Kangilleq and Sermeq Silarleq, as computed by the *Parallel Ice Sheet Model* (PISM), based on a sample velocity field from 2018. (b) Ice velocity vectors and sample terminus (red line), used in conjunction with  $\tilde{\sigma}$  to obtain calving proxy  $\sigma_T$ . Ice velocities downstream of the terminus do not reflect grounded ice, they could be an ice shelf or ice melange.

#### 289 4.1. von Mises Calving Law

290 The von Mises Calving Law (von Mises, 1913; Morlighem and others, 2016) predicts a glacier will calve  
 291 when the tensile von Mises stress  $\tilde{\sigma}$  at the terminus exceeds the ice's yield strength  $\sigma_{max}$ . The calving rate  
 292  $q_c$  is given by Morlighem and others (2016):

$$q_c = \frac{\tilde{\sigma}}{\sigma_{max}} \|\vec{u}\|, \quad (6)$$

293 where  $\vec{u}$  is the vertically-averaged horizontal velocity. We assume plug flow near the calving front (Bassis  
 294 and Ultee, 2019; Greve and Blatter, 2009), making the vertically averaged velocity equal to surface velocity.

#### 295 4.2. Tensile von Mises Stress

296 The von Mises Calving Law requires computation of the *tensile von Mises stress*. In continuum mechanics,  
 297 the strain rate tensor  $\dot{\epsilon}$  may be computed from the gradient of the velocity  $\vec{u}$  as:

$$\dot{\epsilon} = \frac{1}{2} (\nabla \vec{u} + \nabla \vec{u}^T), \quad (7)$$

298 where  $\nabla \vec{u}^T$  is the transpose of the rank 2 tensor  $\nabla \vec{u}$ ; see Gibbs and Wilson (1901) page 404 Eq 3, also  
 299 Cajori (1928) volume II page 135. The scalar *tensile strain rate*  $\bar{\epsilon}$  (Morlighem and others, 2016) is defined  
 300 as:

$$\bar{\epsilon}^2 = \frac{1}{2} (\max(0, \dot{\epsilon}_1)^2 + \max(0, \dot{\epsilon}_2)^2), \quad (8)$$

301 where  $\dot{\epsilon}_1$  and  $\dot{\epsilon}_2$  are the eigenvalues of  $\dot{\epsilon}$ . Glen's Flow Law, the constitutive relation used to model ice  
 302 deformation and flow, relates the *strain rate* tensor  $\dot{\epsilon}$  to the *deviatoric* or *shear stress* tensor  $\sigma$ :

$$\dot{\epsilon} = \tilde{A}\sigma^n, \quad (9)$$

303 where  $\tilde{A}$  is the temperature-dependant rate factor with units  $\text{s}^{-1}\text{Pa}^{-n}$ , and  $n$  is typically assumed to be  
 304 3 (Behn and others, 2021). In this case (Morlighem and others, 2016), Glen's Flow Law is used with the  
 305 scalar tensile strain rate  $\bar{\dot{\epsilon}}$ , and solved for the scalar tensile von Mises stress  $\tilde{\sigma}$ , to obtain:

$$\tilde{\sigma} = \sqrt{3}\tilde{B}\bar{\dot{\epsilon}}^n, \quad (10)$$

306 where  $\tilde{B} = \tilde{A}^{-1/n}$  is the ice hardness (Greve and Blatter, 2009).

307 Figure 8 shows the von Mises stress computed on a grid for one velocity field. Disregarding processes  
 308 other than calving for now, the von Mises calving law predicts that advancing glaciers will have  $\tilde{\sigma} < \sigma_{max}$ ,  
 309 and retreating glaciers will have  $\tilde{\sigma} > \sigma_{max}$ . As a catch-all parameter,  $\sigma_{max}$  accounts not just for ice cliff  
 310 properties and fjord geometry but all factors affecting calving, for example frozen melange in the fjord  
 311 (Schlemm and Levermann, 2020).

### 312 4.3. Computing von Mises Stress

313 For each surface velocity map, we use the Parallel Ice Sheet Model (PISM; Khroulev and the PISM  
 314 Authors, 2022) to compute the tensile von Mises stress  $\tilde{\sigma}$  for a given ITS-LIVE velocity field, using the  
 315 PISM default constant ice hardness of  $\tilde{B} = 68\,082\text{ Pa s}^{1/3}$ .

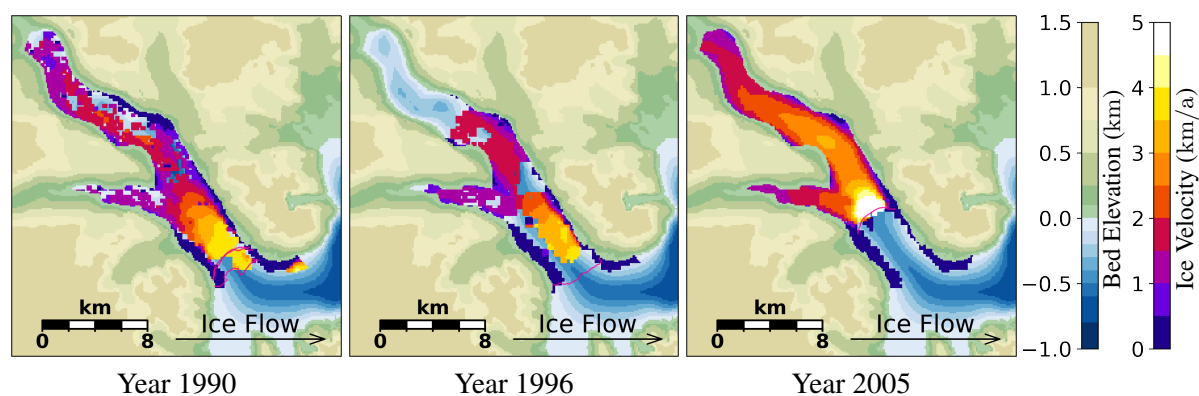
### 316 4.4. Integrating Across the Terminus

317 To obtain a single von Mises stress number for a glacier, the von Mises map computed in Section 4.3 is  
 318 integrated across the glacier's terminus. The value  $\sigma_T$  is defined as the average von Mises stress across the  
 319 glacier terminus for an entire terminus line  $T$  of arbitrary shape:

$$\sigma_T = \frac{\int_T (\tilde{\sigma}\vec{u} \cdot \vec{n})ds}{\int_T (\vec{u} \cdot \vec{n})ds}, \quad (11)$$

320 where  $\vec{n}$  is the unit normal and  $ds$  is used for the line integral along  $L$ , using a rasterized terminus and a  
 321 raster-based formulation for the line integral (Appendix A).

322 This definition of  $\sigma_T$  is robust to missing velocity data near the edges of fast glacier flow and near the  
 323 terminus, a common situation when using InSAR data. If  $\vec{u}$  is missing at some points along  $L$ , then the



**Fig. 9.** Aerial map of AP Bernstorff Glacier in Southeast Greenland showing incomplete data for ice velocities that happen in some cases. Annual average InSAR-derived ice velocity data within the fjord are overlaid on bedmap elevation and fjord bathymetry. Ice velocity data are not shown outside the fjord, where bedrock is above sea level. Terminus measurements within the year are shown in red, with three termini available in year 1990, and just one each in 1996 and 2005. Velocity data coverage is sometimes incomplete, especially close to the terminus or near the margins of the glacial trough. Line integrals in this study disregard any portion of the terminus with missing data. Although the equation for  $\sigma_T$  is robust to missing data at the terminus, it can still fail for lack of data, as in 1996.

324 line integrals in the numerator and denominator will both be missing at the same points along  $L$ , and  
 325 will therefore avoid biasing the result to first order. In this way,  $\sigma_T$  is normalized by the amount of data  
 326 that can be measured (Figure 9). Because of missing data near the margins, the value of  $\sigma_T$  depends more  
 327 heavily on what is happening in the center of the fjord.

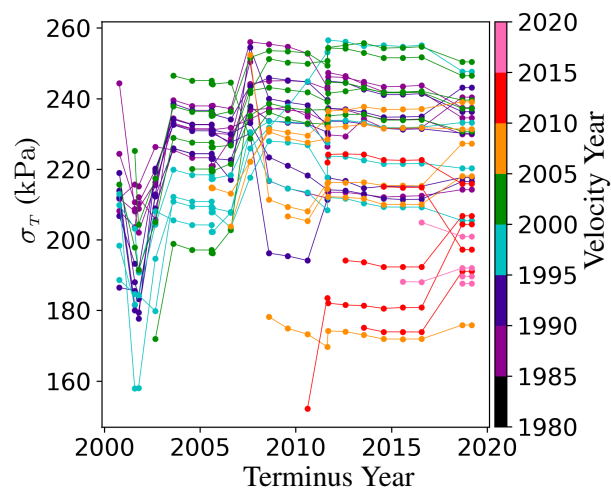
328 Annual terminus lines from Wood and others (2021), manually digitized from LANDSAT 5, 7 and 8  
 329 imagery, were rasterized on the MEaSUREs grids and used for this analysis. By reusing data from Wood  
 330 and others (2021), this study maximizes the ability to compare results with other recent work; however, it  
 331 is also limited to glaciers included in that study.

332 In theory a one dimensional calving rate  $q_c$  can be estimated directly by using  $\sigma_T$  for  $\tilde{\sigma}$  in Equation 6.  
 333 However, errors in estimating  $\sigma_{max}$  lead to large uncertainties in the actual value of  $q_c$ , which is not needed  
 334 anyway. Instead, the von Mises calving law suggests that  $\sigma_T$  on average should be proportional to calving  
 335 rate  $q_c$ . Even without knowing the coefficient of proportionality, this allows  $\sigma_T$  to be used as a proxy for  $q_c$   
 336 without ever having to explicitly determine  $\sigma_{max}$ .

#### 337 4.5. Averaging over Velocity Year

338 Change in surface velocity, not terminus position, is the dominant driver of annual variation in  $\sigma_T$   
 339 (Figure 10). To single out the effect of the *position of the terminus in the fjord* rather than surface velocity,





**Fig. 10.** The calving proxy  $\sigma_T$  value computed for one Glacier (Hayes N); plotted by *velocity year* (year of the velocity field used) and *terminus year* (year of the terminus used), where the velocity year is always less than the terminus year. Although  $\sigma_T$  varies due to the position of the terminus, the largest variation usually occurs due to changes in the overall ice velocity field: some years a glacier may be moving faster, whereas other years it may be moving more slowly.  $\sigma_T$  is averaged across velocity fields of different years to obtain a single value  $\bar{\sigma}_T$  for each year's terminus.

340  $\sigma_T$  is computed using multiple velocity fields for each terminus, even if the terminus and velocity field are  
 341 from different years. The result is then averaged to create  $\bar{\sigma}_T$ . For this procedure to work, there must be  
 342 ice at the terminus so that  $\sigma_T$  can be computed; which for retreating glaciers means the velocity field must  
 343 pre-date the terminus position.

344 Most glaciers in this study were relatively stable until around the year 2000, after which they began to  
 345 retreat en masse (Murray and others, 2015). Due to limited availability of data and the need for surface  
 346 velocities to pre-date terminus positions, only the post-2000 period of retreat is studied. Therefore, only  
 347 terminus/velocity pairs were used in which the terminus year was 2001 or later, and the velocity year was  
 348 older than the terminus year.

#### 349 4.6. Effect of InSAR Data Quality

350 This study uses many approaches to be robust in the face of missing InSAR data. It uses a robust integration  
 351 technique along the terminus (Section 4.4), and then it uses an averaging technique analogous to *stacking*  
 352 (Section 4.5), a well established technique in seismology to improve the signal-to-noise ratio of data. Each  
 353 terminus line is used to integrate all the available velocity fields older than it, thereby decreasing the effect  
 354 of a poor velocity field on any single year. Figure 10 shows each terminus behaving similarly no matter

355 which velocity field it is applied to, thereby adding confidence that poor quality velocity fields with missing  
 356 data are not overwhelming the signal. Finally, termini are only applied to *older* velocity fields. Because  
 357 most glaciers are retreating, this means that the newer terminus will typically be somewhat upstream of  
 358 the end of an older velocity field and will likely be sampling an improved portion of that velocity field.

#### 359 4.7. Analysis of $\bar{\sigma}_T$ and Terminus Residual

360 The residual  $l_\epsilon$  represents the amount of terminus advance / retreat that is *not explained* by thermal  
 361 forcing alone (Section 3.3). With  $l_\epsilon$  and  $\bar{\sigma}_T$  it is now possible to evaluate whether the calving rate proxy  
 362  $\bar{\sigma}_T$  increases or decreases as the glacier retreats.  $l_\epsilon$  and  $\bar{\sigma}_T$  are regressed against each other with a p-value  
 363 significance threshold of 0.21 (see Section ??):

$$\Delta l_\epsilon = \nu \Delta \bar{\sigma}_T, \quad (12)$$

364 where  $\nu$  is the regression coefficient indicating whether fjord geometry causes  $\bar{\sigma}_T$  to increase or decrease as  
 365 the glacier retreats.

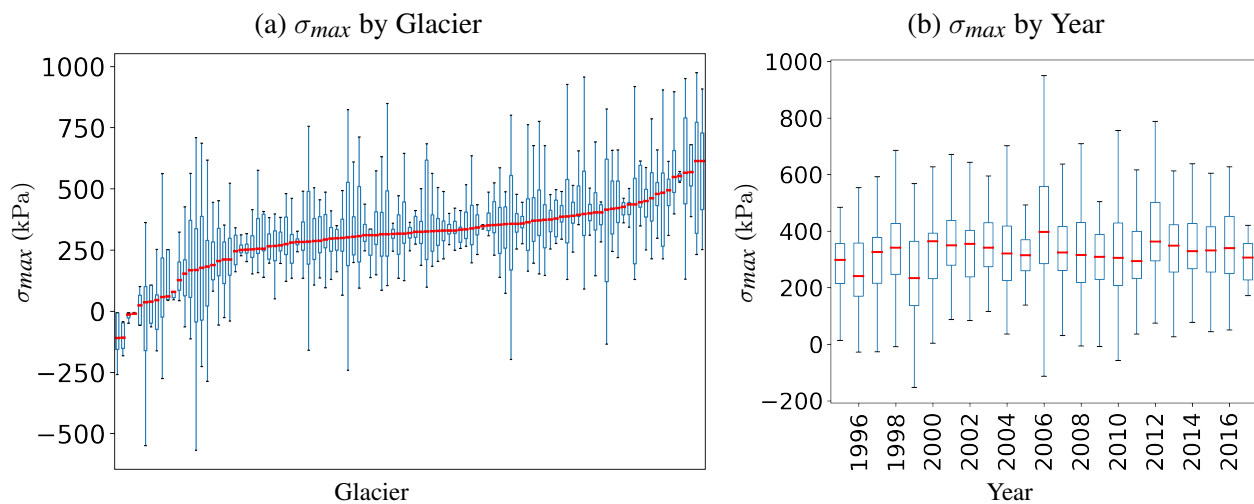
366 If a glacier is susceptible to rapid retreat and has just begun to retreat — as many may be at this  
 367 point — then the TGC predicts  $\nu$  should be negative. That is, stress at the terminus  $\bar{\sigma}_T$  increases as the  
 368 glacier begins to retreat, causing a positive feedback that could lead to instability. Such a glacier could well  
 369 continue to retreat, even if frontal melt rate were to stabilize or decrease.

370 If on the other hand a glacier is in a stable configuration, then  $\nu$  will be positive, meaning  $\bar{\sigma}_T$  *decreases* as  
 371 the glacier retreats. Such glaciers could be retreating in spite of their geometric stability due to the primary  
 372 forcing from warming oceans; however, at this time the fjord geometry is helping stabilize the glacier and  
 373 prevent runaway retreat.

374 If a glacier has already begun rapid retreat and is currently retreating through an area with little variation  
 375 in fjord cross-sectional geometry, then  $\bar{\sigma}_T$  will be about constant, even as  $l_\epsilon$  changes. There is no relationship  
 376 between  $\bar{\sigma}_T$  and the terminus residual  $l_\epsilon$ , and hence the p-value value for  $\nu$  will be high. Lack of statistical  
 377 significance is correlated with glaciers already in rapid retreat, as was confirmed in our results. Note that  
 378 in principle, lack of predictive power of the Slater regression must also be considered as a possible cause.

### 379 5. VALIDATION OF VON MISES CALVING LAW

380 The von Mises Calving Law as a model may be validated by applying it to the data of Wood and others  
 381 (2021) and evaluating the result for coherence and consistency. Wood and others (2021) measure or model



**Fig. 11.** Implied  $\sigma_{max}$  parameter obtained by fitting  $\sigma_T$  computed using same-year velocity and terminus measurements, to calving rate obtained by residuals of other quantities from Wood and others (2021) (Equation 13), and grouped by either glacier or year. The red line is the median, the box extends to the edge of the first and third quartile, the whiskers extend to the furthest data point in the first and third quartiles, and outliers are not shown. (a)  $\sigma_{max}$  grouped by glacier. For most glaciers,  $\sigma_{max}$  lies in the range 250 kPa to 350 kPa, with some outliers. Occasional negative values of  $\sigma_{max}$  are non-physical and caused by issues with Wood et al data:  $\sigma_T$  is always positive. Consistent value across most glaciers supports von Mises calving law as a reasonable model. (b)  $\sigma_{max}$  across all glaciers grouped by year. Consistent year-to-year stability supports von Mises calving law as a reasonable model.

382 all terms of the mass balance equation (Equation 1) except for the calving rate  $q_c$  — which is computed as  
 383 a residual between observed terminus location  $L$  and the integrated effect of all other fluxes: ice advection  
 384 ( $q_f = ||\vec{u}||$ ), frontal melt ( $q_m$ ) and thinning-induced retreat ( $q_s$ ). Although  $\sigma_{max}$  does not need to be  
 385 computed for this study, it may still be determined from the Wood and others (2021) data and the definition  
 386 of the von Mises calving law (Equation 6):

$$\sigma_{max} = \frac{\sigma_T}{q_c} \frac{dL}{dt}, \quad (13)$$

387 where  $dL/dt$  is the rate of terminus advance or retreat.

388 Using this formula,  $\sigma_{max}$  was estimated based on all available terminus lines of Wood and others (2021),  
 389 using the velocity field from the same year as each terminus line. Figure 11(a) shows the result grouped  
 390 by glacier. In this plot,  $\sigma_{max}$  displays a two-tailed cumulative distribution function. This is to be expected  
 391 for a value like  $\sigma_{max}$  that is thought to be affected by a number of glacier-specific parameters such as ice  
 392 shelves, melange characteristics, etc; and would therefore be expected to converge on a normal distribution.  
 393 Figure 11(b) shows that the average of  $\sigma_{max}$  across all glaciers does not change much from year to year.

394 These results are reasonable and coherent, even though  $q_c$  is a residual, and therefore incorporates all errors  
395 and biases from the various datasets and models used by Wood and others (2021). Overall estimated value  
396 for  $\sigma_{max}$  is  $300 \pm 100$  kPa.

397 In some cases,  $\sigma_{max}$  is estimated to be negative. That is a limitation of the Wood et al dataset and  
398 the nature of  $q_c$  as a residual, since  $\sigma_T$  used in Equation 13 by definition is always positive. The overall  
399 consistency of  $\sigma_{max}$  suggests that the von Mises calving law is a useful model for Greenland tidewater  
400 glaciers; and that residual values of  $q_c$  (Wood and others, 2021) are not overwhelmed by noise, in spite of  
401 the multi-step process used to compute them.

402 Overall, our results (Figure 11) support the von Mises criterion as a reasonable calving model for  
403 Greenland tidewater glaciers.

## 404 6. RESULTS AND DISCUSSION

405 Of the 44 glaciers analyzed, 10 were determined to have a destabilizing fjord geometry (the glacier is calving  
406 more as it retreats), 7 a stabilizing fjord geometry (the glacier is calving less as it retreats), 16 were found  
407 to be stable so far (their termini haven't moved much in the dataset) and 11 to have already entered the  
408 rapid retreat phase of the TGC (Figure 12 and 5). Each category is analyzed further below.

### 409 6.1. Destabilizing Fjord Geometry

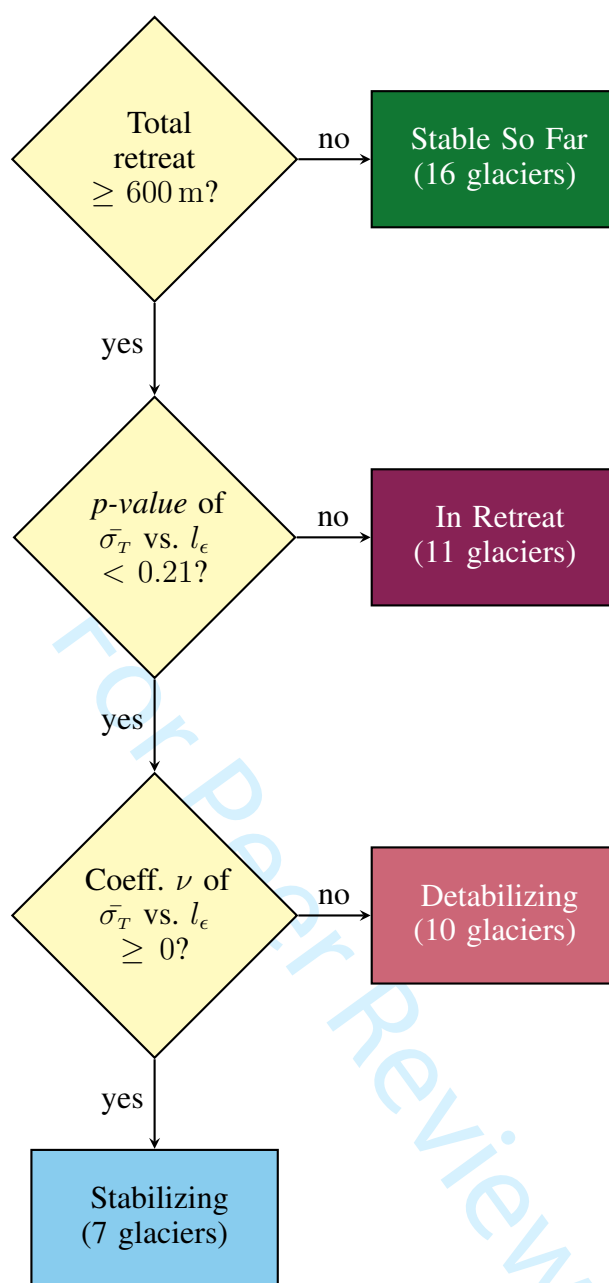
410 Some glaciers have a regression coefficient  $\nu < 0$  (negative slope of line in column (b)), suggesting that  
411 they are entering the rapid retreat stage of the TGC. Their termini have retreated substantially (more  
412 than 600 m) since 2000; and they have retreated faster than thermal forcing would have predicted.

413 *Puisortoq N* and *Puisortoq S* (Figure 14) in Southeast Greenland are both canonical examples of retreat  
414 that has continued due to fjord geometry in spite of recent decreases in thermal forcing, suggesting that  
415 the retreat has become self-sustaining. *Carlos Glacier* on the west coast offers a similar story.

416 Some glaciers show episodic retreat; for example, *Eqip Sermia* on the west coast. In this case, the episodic  
417 retreat is correlated to changes in thermal forcing, although it could also be due to pinning points.

### 418 6.2. Stabilizing Fjord Geometry

419 Some glaciers have a stabilizing fjord geometry. A small number is to be expected: to be in this category,  
420 a glacier must have stable fjord geometry but still be retreating anyway due to frontal melt, a condition  
421 that would happen near the end of the rapid retreat phase.

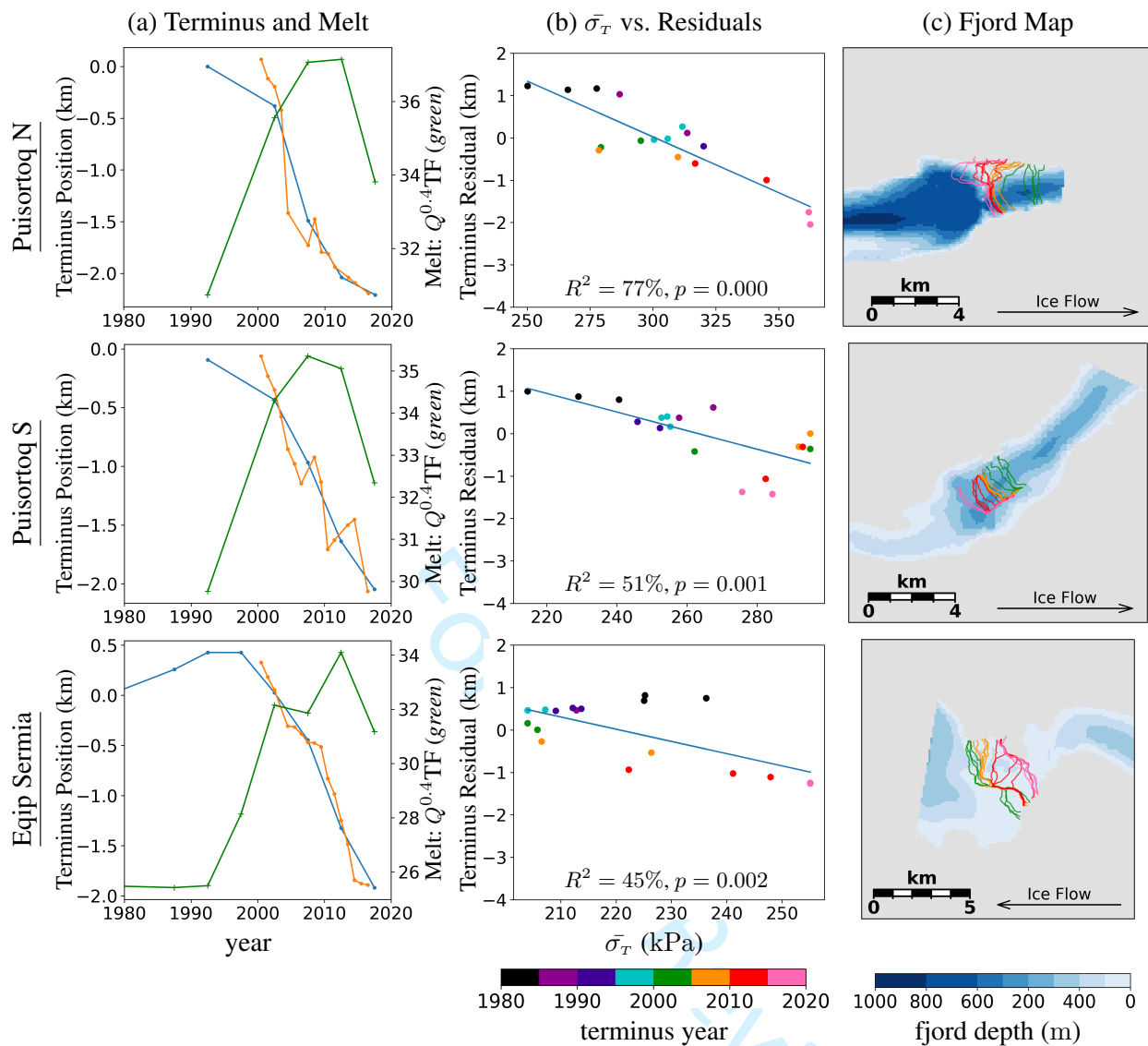


**Fig. 12.** Glacier categorization flowchart. Glaciers that have moved less than 600m over the study period are considered stable so far. Otherwise, a regression between the calving proxy  $\bar{\sigma}_T$  and the terminus residual  $l_\epsilon$  is performed. If that regression lacks significance at  $p$ -value of 0.21, then the glacier is considered to already be in rapid retreat. Otherwise, the sign of the regression coefficient  $\nu$  distinguishes between destabilizing geometry (negative sign) vs. stabilizing geometry (positive sign).

422 *Lille Glacier* (Figure 15) is a good example of this, as the terminus retreats into a narrow section at  
 423 the head of the fjord. *Ussing Braeër N* may also fall into this category, although its geometry is more  
 424 complex. This increasing stability and slowing down of retreat is the ultimate fate for many tidewater

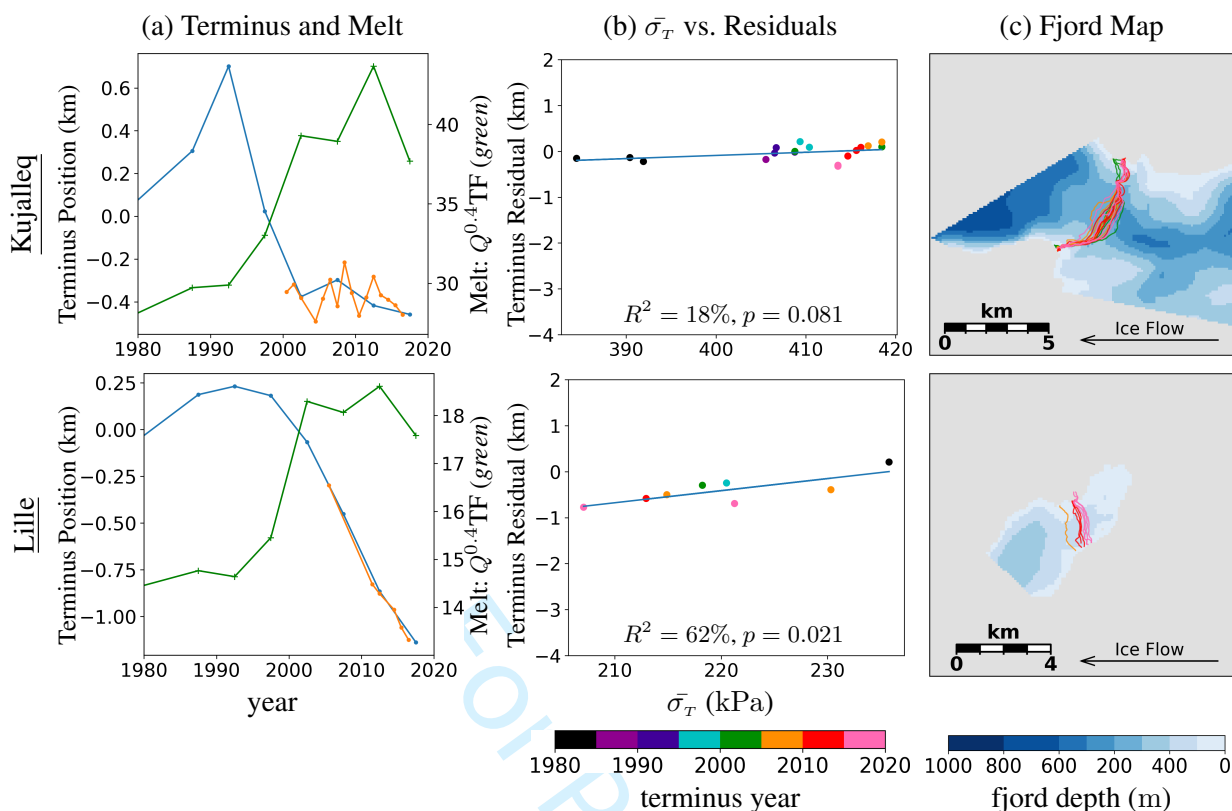
ID	Name	Latitude	Retreat (m)	$\nu$	p-value	Mean $\bar{\sigma}_T$ (kPa)
<b>Destablizing</b>						
143	Danell S	E-60.8	-998	-9e-06	0.01	212
85	Puisortoq S	E-61.9	-1762	-2e-05	0.00	264
55	Puisortoq N	E-62.1	-2370	-3e-05	0.00	305
23	Mogens Heinesen S	E-62.4	-2782	-1e-05	0.08	378
90	Eqip Sermia	W-69.8	-2260	-3e-05	0.00	223
52	Kangilernata	W-69.9	-3098	-2e-05	0.21	229
96	Hayes N	W-74.9	-1202	-2e-05	0.01	221
150	Savissuaq WWWW	W-76.2	-3059	-1e-05	0.06	192
89	Docker Smith W	W-76.3	-940	-1e-05	0.07	219
118	Carlos	W-76.4	-806	-7e-06	0.00	212
<b>Stabilizing</b>						
60	Mogens Heinesen C	E-62.5	-1321	2e-05	0.14	275
10	Uunartit	E-67.4	-4316	5e-05	0.18	322
134	Lille	W-70.5	-814	3e-05	0.02	220
25	Sermeq Silarleq	W-70.8	-4743	3e-05	0.00	304
106	Ussing Braeer N	W-73.9	-688	2e-05	0.00	262
116	Hayes North	W-75.0	-1028	9e-06	0.00	234
192	Savissuaq WW	W-76.3	-1229	8e-06	0.15	167
<b>Stable</b>						
108	Danell	E-60.9	283	-1e-05	0.10	273
43	Herluf Trolle N	E-61.3	-373	-5e-06	0.03	313
33	Anorituup Kangerlua N	E-61.6	452	1e-06	0.95	386
7	Gyldenlove N	E-64.3	-211	3e-06	0.01	265
14	Koge Bugt S	E-65.0	-283	2e-06	0.57	318
8	Daugaard Jensen	E-71.9	-434	3e-05	0.26	258
36	Kangiata Nunaata	W-64.3	72	-5e-05	0.30	438
13	Kujalleq	W-70.0	-41	7e-06	0.08	408
53	Sermeq Avannarleq	W-70.1	-132	3e-08	0.99	249
6	Store	W-70.4	69	2e-06	0.01	298
119	Sermilik	W-70.6	-14	-2e-07	0.12	223
70	Kangilleq	W-70.7	-24	4e-08	0.85	288
47	Kangerlussuup	W-71.5	-33	-7e-06	0.81	263
5	Rink Isbrae	W-71.7	-139	4e-07	0.81	256
22	Upernavik Isstrom S	W-72.8	-591	-1e-05	0.05	257
24	Hayes M	W-74.8	-20	-5e-06	0.36	301
<b>In Retreat</b>						
45	Herluf Trolle S	E-61.2	-2359	5e-06	0.31	348
31	Mogens Heinesen N	E-62.5	-3202	-4e-06	0.82	307
12	AP Bernstorff	E-63.8	-2929	3e-05	0.24	263
29	Ikertivaq N	E-65.6	-652	5e-06	0.35	313
2	Kangerlussuaq	E-68.6	-1688	-6e-05	0.28	299
68	Ummiammakku	W-71.7	-5579	-2e-06	0.90	238
88	Inngia	W-72.0	-8038	-7e-05	0.45	156
98	Akullikassaap E	W-73.0	-1915	9e-06	0.78	120
130	Cornell N	W-74.3	-818	-8e-06	0.52	130
37	Hayes NN	W-74.9	-2141	-1e-05	0.33	194
171	Savissuaq WWW	W-76.2	-1663	2e-06	0.37	165

**Fig. 13.** Summary of results per glacier. Glaciers are grouped by their final categorization (*Destablizing*, *Stabilizing*, *Stable* or *In Retreat*). Columns are **ID**: ID of glacier as found in Rignot and Mouginot (2012); **Name**: Name of glacier as found in Wood and others (2021); **Latitude**: Latitude (degrees North) of glacier combined with indication of its location on the east (*E*) or west (*W*) side of Greenland; **Retreat**: Total amount of retreat (m) over the study period (negative for retreat, positive for advance);  $\nu$ : Relationship between terminus residuals and  $\bar{\sigma}_T$  (Section 4.7); **p-value**: Level of statistical significance of  $\nu$  (Section 4.7); **Mean  $\bar{\sigma}_T$** : Mean value of  $\bar{\sigma}_T$  for this glacier's terminus across *all* years.



**Fig. 14.** Analysis of glaciers that destabilize upon retreat. (a) 5-year Slater relative terminus (blue) and melt (green crosses) used in Slater regressions; and annual Wood relative terminus (orange). Slater (blue) and Wood (orange) relative termini should be similar because they measure the same physical quantity. Predictions from the Slater thermal forcing model are not shown. (b) Regression of calving proxy  $\bar{\sigma}_T$  vs. relative terminus residuals as per Slater; (c) Reference map of fjord. Although thermal forcing has decreased since 2015, retreat has continued. Based on fjord geometry and recent decreases in retreat rate, Puisortoq N and Eqip Sermia might stabilize soon; however, that is speculation because the terminus has not yet had a chance to “see” these potential pinning points, and thermal forcing could cause continued retreat in any case.

425 glaciers because fjords must become shallower close to their head, or narrower at a pinning point. In the  
 426 past, glaciers may have come to rest long-term at pinning points, but continually rising ocean temperatures  
 427 make that less likely in the future. In this study we observe many glaciers slowing their retreat at pinning

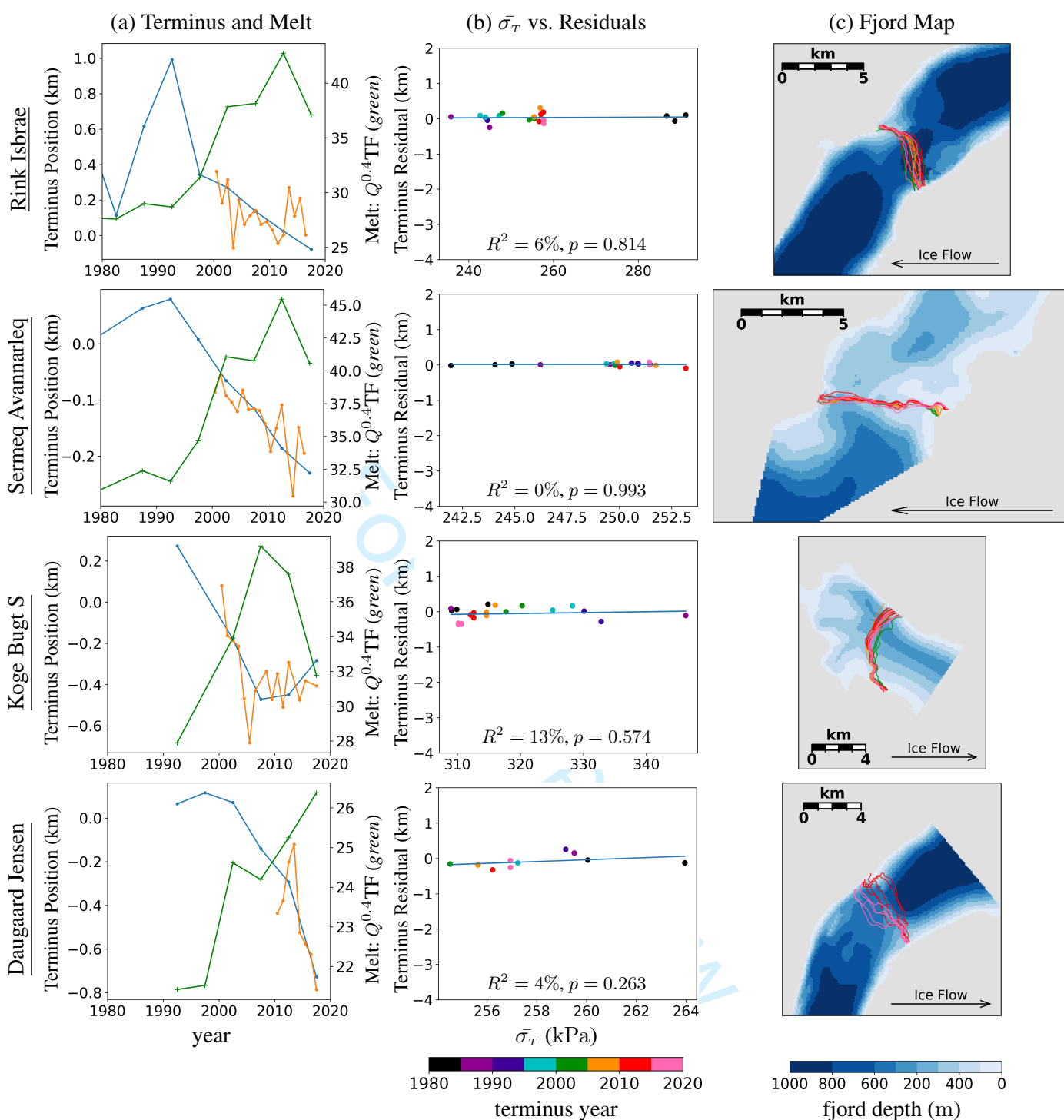


**Fig. 15.** Analysis of glaciers that stabilize upon retreat. (a) 5-year Slater relative terminus (blue) and melt (green crosses) used in Slater regressions; and annual Wood relative terminus (orange). Slater (blue) and Wood (orange) relative termini should be similar because they measure the same physical quantity. Predictions from the Slater thermal forcing model are not shown. (b) Regression of calving proxy  $\bar{\sigma}_T$  vs. relative terminus residuals as per Slater; (c) Reference map of fjord. Gyldenlove N shows two distinct clusters of points — one cluster with unusually low  $\bar{\sigma}_T$  and no correlation to residuals, and one cluster with more typical  $\bar{\sigma}_T$  values and negative correlation, which would indicate a more typical destabilization upon retreat. The methodology might be confounded here by the complex terminus geometry. Kujalleq’s terminus has not moved enough to adequately sample changes in fjord geometry. And from the map, it Lille now terminates near the head of the fjord, where water becomes more shallow with further retreat.

428 points; but we see no evidence of tidewater glaciers stabilizing anywhere but on land once they have begun  
 429 rapid retreat.

430 Some glaciers, for example *Hayes North*, have complex geometry and are a poor “fit” for a linear  
 431 regression.





**Fig. 16.** Analysis of glaciers for which a least square fit of terminus position has retreated less than 600 m over the study period; and due to lack of sampling from terminus movement, were statistically insignificant. (a) 5-year Slater relative terminus (blue) and melt (green crosses) used in Slater regressions; and annual Wood relative terminus (orange). Slater (blue) and Wood (orange) relative termini should be similar because they measure the same physical quantity. Predictions from the Slater thermal forcing model are not shown. (b) Regression of calving proxy  $\bar{\sigma}_T$  vs. relative terminus residuals as per Slater; (c) Reference map of fjord.

### 432 6.3. Currently Stable

433 Some glaciers have been stable during the study period, with termini that moved on average less than  
434 600 m: the methods of this study revealed no new information about them, beyond their already known  
435 recent stability. Because the terminus stayed relatively stationary, no statistically significant relationship  
436 was found between terminus residual and  $\bar{\sigma}_T$  (Figure 16). The complete list of glaciers in this class is  
437 *Anorituup Kangerlua N*, *Daugaard Jensen*, *Hayes M*, *Kangerlussup*, *Kangiata Nunaata*, *Kangilleq*, *Koge*  
438 *Bugt S*, *Rink Isbrae* and *Sermeq Avannarleq*.

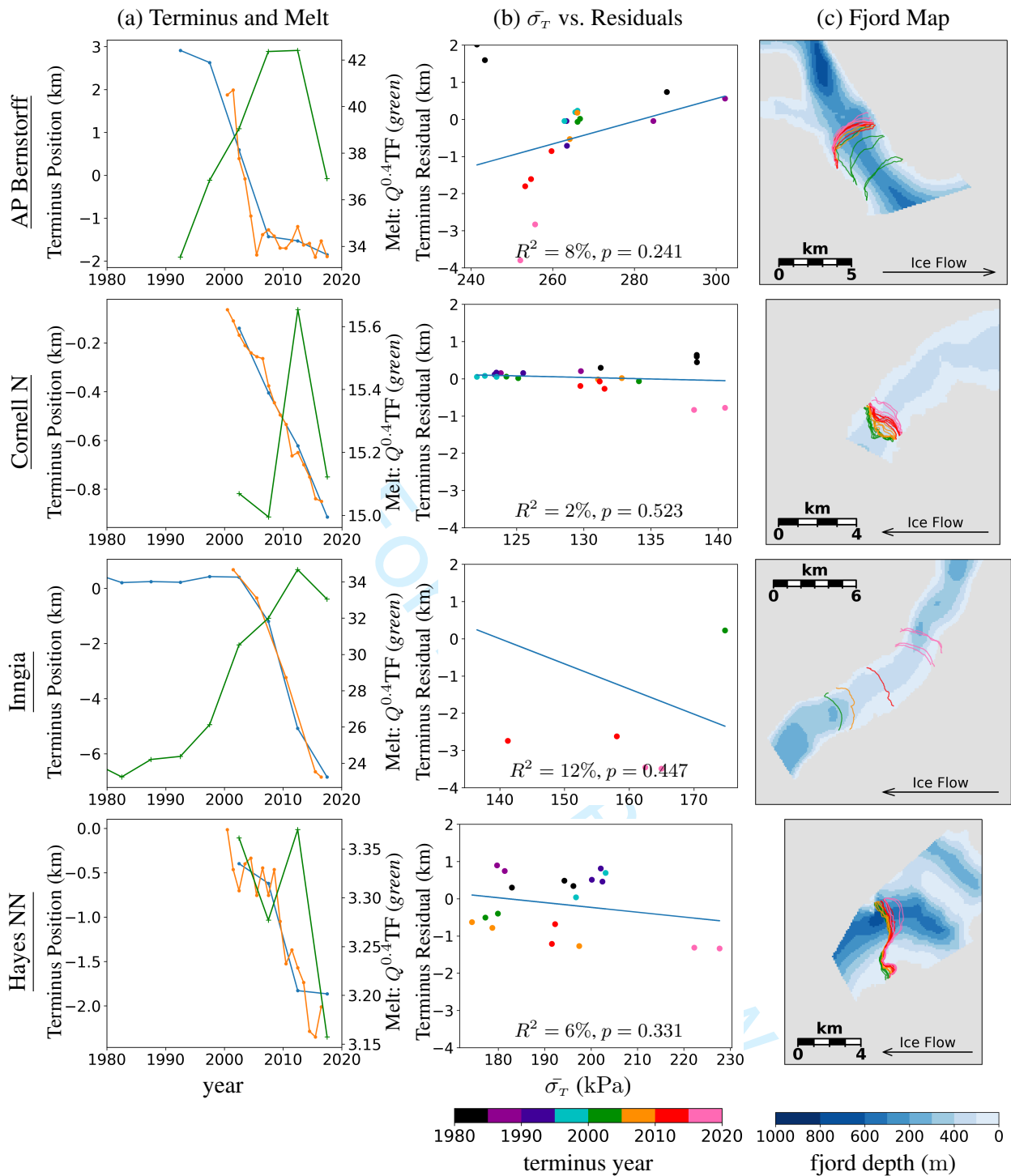
439 To account for noisy data, the threshold for retreat was determined based on the slope of the *least squares*  
440 *fit* line through the terminus positions of Wood and others (2021) since the year 2000. This criterion can  
441 yield surprising results in some cases: for example, *Daugaard Jensen*, a glacier that was considered to be  
442 “stable” by the criteria of this study. The glacier was historically considered stable, with concern it could  
443 soon destabilize (Bevan and others, 2012). Examination of the data suggests it advanced a modest 700 m  
444 before 2013, and since then has retreated almost 1 km. The advance and retreat periods cancel out in  
445 this study, but only by chance. *Daugaard Jensen* is no longer, in fact a stable glacier — it is currently  
446 retreating. This kind of mischaracterization can happen for any glacier that changes its behavior over the  
447 study period.

### 448 6.4. In Retreat

449 Finally, there are the glaciers for which no statistically significant relationship could be found between  
450 terminus residual and  $\bar{\sigma}_T$ , and which have retreated at least 600 m over the study period (Figure 17).

451 As above, some glaciers show transition between regimes, confounding a single linear model. For example,  
452 *AP Bernstorff* retreated rapidly until the year 2005, after which it has remained stable — in spite of changes  
453 in thermal forcing both up and down. This is apparently caused by a shallowing of the fjord at the current  
454 terminus location. *Herluf Trolle S* and *Mogens Heinesen C* are other examples. *Ummiammakku* retreated  
455 rapidly until 2010, at which it stabilized on a pinning point. It is classified as *in retreat* by the systematic  
456 methods of this study because most of the data show it retreating: if it has truly stopped retreating, there  
457 have not yet been enough years of stability to statistically “overwhelm” the previous years of retreat.  
458 Improvements to the methodology that weight recent behavior more strongly might be able to overcome  
459 this kind of limitation.

460 *Inngia*, *Kangerlussuaq*, *Mogens Heinesen N* and *Savissuaq WWW* are retreating steadily through a  
461 uniform portion of the fjord, likely driven by thermal forcing and already “knocked” off their stable terminal



**Fig. 17.** Analysis of glaciers lacking a statistically significant relationship between  $\bar{\sigma}_T$  and relative terminus residual, but with more than 600 m of retreat. (a) 5-year Slater relative terminus (blue) and melt (green crosses) used in Slater regressions; and annual Wood relative terminus (orange). Slater (blue) and Wood (orange) relative termini should be similar because they measure the same physical quantity. Predictions from the Slater thermal forcing model are not shown. (b) Regression of calving proxy  $\bar{\sigma}_T$  vs. relative terminus residuals as per Slater; (c) Reference map of fjord. A simple linear regression is confounded for *AP Bernstorff*, which underwent a regime transition around 2005. *Cornell N* has not retreated enough to produce statistically significant results. *Inngia* is in rapid retreat along a homogenous section of the fjord. *Hayes NN* exhibits complex geometry that is hard to describe or classify, with multiple adjacent glaciers and calving fronts.

462 moraine before the start of this study. Fjord geometry does not affect retreat at this point in time because  
463 of this uniformity, which results in a “cloud” of points and lack of statistical significance when regressing for  
464  $\nu$  (the slope of the line in column (b)). However, they all show negative  $\nu$  at less than statistical significance,  
465 suggesting that small variations in fjord geometry *are* affecting terminus position as the TGC hypothesis  
466 would suggest.

467 Some glacier termini inhabit broad regions of grounded ice without well defined fjords, often fed by  
468 multiple glaciers upstream. In these cases, neither Slater’s thermal forcing model nor the Tidewater Glacier  
469 Cycle seems to show much predictive power; *Hayes NN* and *Uunartit*, for example. Some glaciers *do* exist  
470 in well defined fjords but the terminus is close to a branching or merging point, for example *Savissuaq*  
471 *WW*.

472 Other glaciers simply lack data sufficient to build statistically meaningful results: for example, Wood and  
473 others (2021) provide only two terminus positions for *Akullikassaap E* and three for *Anorituup Kangerlua*  
474 *N*.

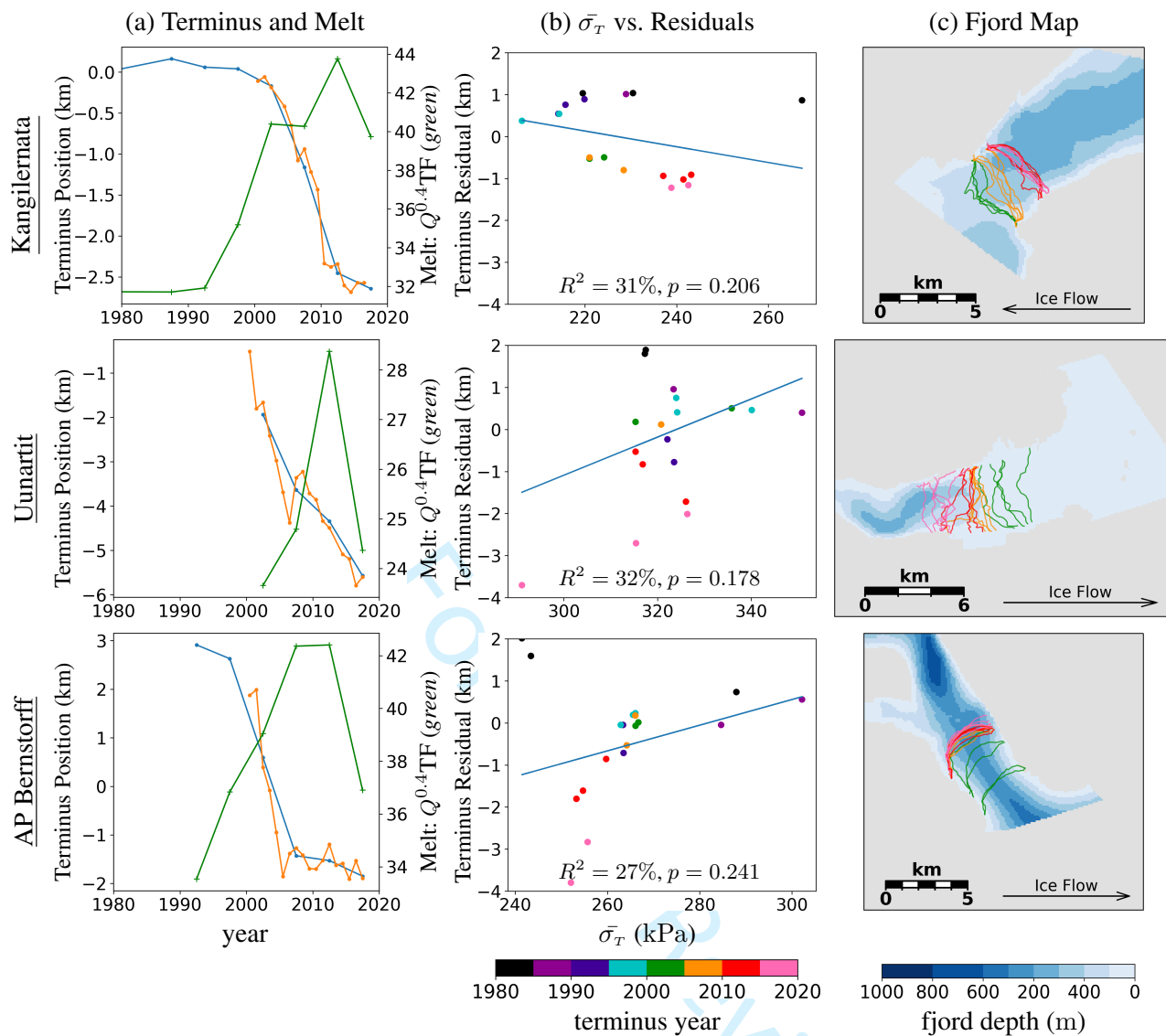
## 475 6.5. Edge Cases

476 The choice of the threshold at 0.21 to separate glaciers in rapid retreat from ones that are stabilizing  
477 / destabilizing (Figure 12) is somewhat arbitrary. Some glaciers show clear and consistent behavior and  
478 have small p-values, for example *Puisortoq N* (Figure 14) and *Lille* (Figure 15). Other glaciers show large  
479 p-values indicating no effect of changes in fjord geometry on continued retreat (Figure 17). However, it is  
480 harder to classify the behavior of glaciers with p-values close 0.21.

481 Figure 18 shows the glaciers of highest p-value in each of the *destabilizing* and *stabilizing* categories,  
482 and the glacier of lowest p-value in the *in retreat* category (*AP Bernstorff*). All three of these glaciers are  
483 correctly classified, but are also edge cases for their categories, as evidenced by their marginal p-values.

484 *Kangilernata* is in fact destabilizing; but it is retreating off an unusually broad shoal more than 2 km  
485 wide, creating a situation, similar to that of ongoing rapid retreat, in which the fjord cross section does  
486 not change much even as the glacier has retreated more than 2 km.

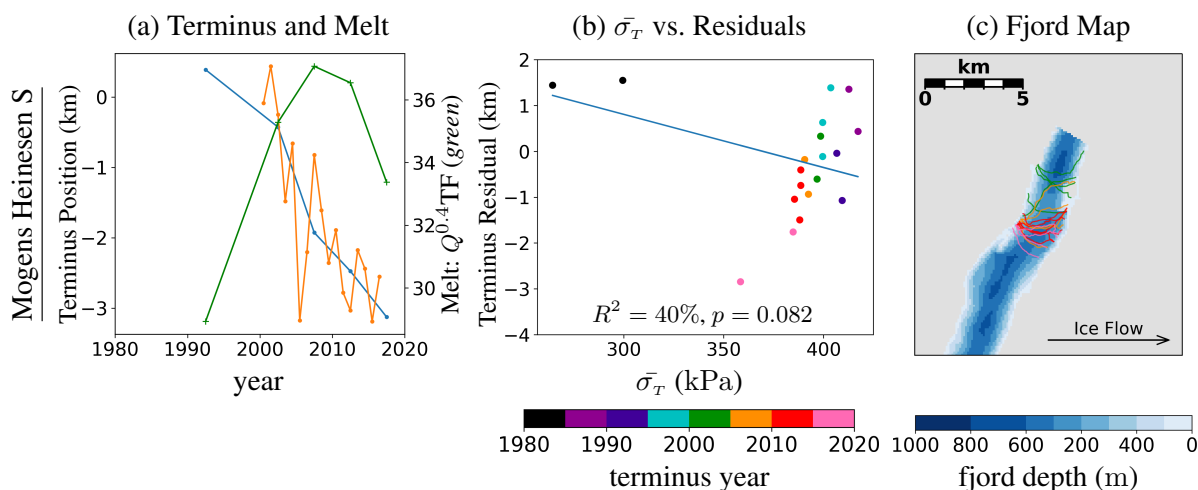
487 *Uunartit* is in fact stabilizing as it reaches the end of the fjord; but the scenario is confounded because  
488 *this* fjord gets deeper even as it narrows, thereby reducing for now the amount that calving decreases  
489 as it retreats.



**Fig. 18.** Edge case glaciers within each category: *Kangilernata* and *Uunartit* are destabilizing and stabilizing, respectively, and have the highest  $p$ -values in their classifications. *AP Bernstorff* is in retreat and has the lowest  $p$ -value in its classification.

490 *AP Bernstorff* is in fact in the midst of rapid retreat; however, retreat has recently slowed down as it  
 491 has reached a shallower section of the fjord, resulting in overall more stabilizing behavior. This shows  
 492 how changes in behavior over the study period can confound the methods of this study.

493 Full results and classifications are provided in the Supplemental material, allowing the reader to compare  
 494 other glaciers to these edge cases and to evaluate the potential effect of other  $p$ -value cutoffs. Although  
 495 the cutoff value  $p = 0.21$  provided accurate classifications in this case, we do not expect  $p = 0.21$  to be  
 496 fundamental to this method. The data sources in this study came with large and often unquantifiable



**Fig. 19.** *Mogens Heinesen S*, which is mis-classified due to two outlier points in the regression of  $\bar{\sigma}_T$  vs. Residuals (column b).

497 uncertainties. If those were to be reduced in a future study, we would expect a smaller p-value cutoff to be  
 498 appropriate.

## 499 6.6. Outliers

500 In some cases, outliers can cause the regression to mis-classify. For example, *Mogens Heinesen S* (Figure 19)  
 501 is classified as *stabilizing*, but the regression data suggest it is actually *destabilizing*, except for two outlier  
 502 data points from the 1980's.

## 503 7. CONCLUSIONS

504 This study quantitatively identifies glaciers for which fjord geometry is either adding to or detracting from  
 505 terminus stability, based on the calving proxy  $\bar{\sigma}_T$ ; and qualitatively matches to expectations based on a  
 506 visualization of fjord geometry. We draw the following conclusions:

507 This study provides quantitative support for Tidewater Glacier Cycle as a useful model for understanding  
 508 the behavior of tidewater glaciers in Greenland, showing general agreement with that model.

509 This study confirms the general assertions of Wood and others (2021) and Slater and others (2019),  
 510 concluding that increased frontal melt due to ocean warming since the year 2000 is currently the  
 511 dominant process driving tidewater glacier retreat in Greenland today. This dominant effect must  
 512 be removed from the data in order to study calving dynamics and rapid retreat controlled by fjord  
 513 geometry. Frontal melt has only recently become the dominant mechanism for tidewater glacier retreat

514 in Greenland. In fact, early work on the tidewater glaciers focus on calving as the primary mechanism  
515 of retreat, and does not address ocean warming (Post, 1975; Meier and Post, 1987).

516 Although statistically significant in many cases, the linear relationship between ocean thermal forcing  
517 and tidewater glacier retreat as developed by Slater and others (2019) should be used with caution  
518 because it does not account for the calving effects of fjord geometry inherent in the TGC. In the  
519 limit, the linear relationship would suggest tidewater glaciers behave like land terminating glaciers,  
520 advancing and retreating in lockstep with climate, which runs contrary to our understanding of the  
521 TGC (Pfeffer, 2007). For this reason, we suggest caution in using Slater and others (2019) to generate  
522 future extrapolated boundary conditions for a GCM, as was proposed in that study.

523 In spite of increasing frontal melt, not all Greenland glaciers are retreating. One can hypothesize this is  
524 due to exceptionally stabilizing fjord geometry, but the methods of this study are unable to confirm or  
525 deny such a hypothesis. Speculation on the future of currently stable glaciers might best be accomplished  
526 through modeling studies based on the measured bed geometry and idealized thermal forcings and frontal  
527 melt.

528 A number of glaciers confound the methods presented here. Some lack statistical significance for glaciers  
529 with complex bed geometries or ill-defined fjords. Some transition between regimes over time — either  
530 increasing or decreasing retreat rate quickly as in a surge-type glacier. These issues are problems in  
531 the current analysis, which is based on simple linear regressions with an assumption of stationarity.  
532 However, the satellite era for glaciers is short and overall lack of data could render useless efforts to use  
533 more powerful machine learning techniques, which would require large datasets.

534 In spite of the complexity, the main thrust of the Tidewater Glacier Cycle is consistently supported  
535 by the evidence in this systematic study of glaciers. Glaciers that retreat faster than thermal forcing  
536 models would predict have increasing  $\bar{\sigma}_T$  with retreat; and in these cases, the terminus is observed to  
537 be retreating through a section of the fjord that is widening or deepening, thereby generally confirming  
538 the TGC. Whereas glaciers with less retreat than thermal forcing would show decreasing  $\bar{\sigma}_T$ , which can  
539 often be verified by observing pinning points, confirming the TGC as well.



540 Based on data from Wood and others (2021) and estimates of  $\sigma_{max}$  from that data, this study supports  
541 the von Mises criterion as a reasonable calving model for Greenland tidewater glaciers (Figure 11), with  
542  $\sigma_{max} \approx 300 \pm 100\text{kPa}$ .

## 543 8. FUTURE WORK

544 The present study offers encouraging preliminary results that could be improved in many ways: more  
545 glaciers in the study, more data per glaciers, more advanced machine learning techniques, and more  
546 predictor variables. Lack of data over only 20 years is a persistent issues limiting the statistical techniques  
547 available.

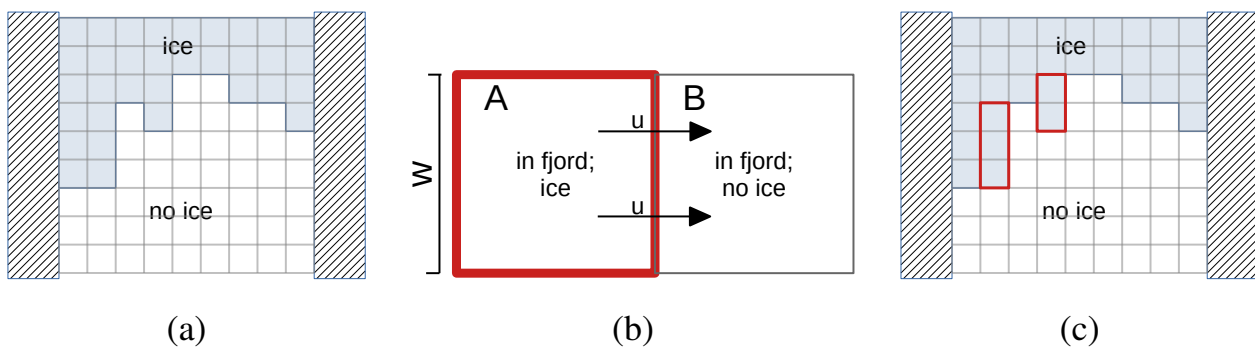
548 Although glaciers are examined systematically in this study, only 44 of the over 200 Greenland tidewater  
549 outlet glaciers (Fahrner and others, 2021) were included, a consequence of relying on multiple previous  
550 studies for data. The limiting factor was the requirement that glaciers appear in *both* the Wood and  
551 Slater datasets, and also on a MEaSURES grid. Although Wood and others (2021) provide data on the  
552 different factors driving terminus retreat, ultimately the only portion of that dataset used was the terminus  
553 lines. Recent efforts have produced abundant terminus traces through machine learning techniques (Cheng  
554 and others, 2021a), which could in principle allow these methods to be run on more data and a larger  
555 set of glaciers. It might be possible to reduce noise by “stacking” results from high frequency terminus  
556 measurements from different seasons within one year. Similarly, it might be possible to use more than one  
557 velocity field per year.

558 Improving the data analysis of this study is another avenue for future research. The current study uses  
559 two sequential linear regressions: first the regression of Slater and others (2019), and then a regression of  
560  $\bar{\sigma}_T$  on terminus residuals. More typically, multiple linear regression would be used here.

561 The use of sub-annual termini could add data for more robust statistics, but would also introduce more  
562 natural seasonal variability in terminus position, which would have to be accounted for; there is no obvious  
563 way to take an “average” of multiple terminus lines. Recent efforts have produced abundant terminus traces  
564 through machine learning techniques such as automated deep learning (Cheng and others, 2021a), which  
565 would in principle enable a larger number of glaciers for a study like this.

566 Beyond that, the volume of data permitting, advanced machine learning techniques could be applied  
567 to predict terminus position based on a range of predictor variables: subsurface runoff ( $Q$ ), ocean/fjord  
568 thermal forcing ( $TF$ ),  $\bar{\sigma}_T$ , air temperature, and other climate drivers (Fahrner and others, 2021). These  
569 methods might be a reasonable way to use high-frequency (sub-annual) terminus lines and velocity fields.





**Fig. 20.** Illustration of line integrals on a grid. (a) Schematic of gridded ice mask, in which the terminus boundary follows grid cell boundaries; hatched areas are the edge of the fjord. (b) Grid cell *A* has west-to-east flux flowing into grid cell *B*, based on the  $u$  component of the vector field. Such cells are identified by the rule that *A* must be in the fjord and ice covered; whereas *B* must be in the fjord and ice-free. (c) Illustration of grid cells, in red, that have west-to-east flux across the gridded terminus.

570 The observational methods in this study rely on large amounts of data and are only applicable for the  
 571 satellite era. For the study of tidewater glacier behavior in the past or future, modelling based on dynamic ice  
 572 models such as PISM (PISM; Khroulev and the PISM Authors, 2022) bounded by *Bed Machine* (Morlighem  
 573 and others, 2017) would be more appropriate. Although observations from before the satellite era are too  
 574 sparse to use for the methods in this study, they would be invaluable in calibrating and validating physics  
 575 based models, opening a window into the past.

576 Although this study focuses on Greenland only, it does not rely on any properties specific to Greenland;  
 577 and given appropriate datasets, we believe its methods can be generalized to tidewater glaciers worldwide.  
 578 Given appropriate data, these methods could help provide a stability assessment for tidewater glaciers in  
 579 other regions such as Alaska and Antarctica.

## 580 APPENDIX A. LINE INTEGRALS ON A GRID

581 Computation of the *up area* (Section 3.2) provides a gridded *ice mask*, which is used to determine whether  
 582 each grid cells is ice-covered or ice-free. A vector field is also provided on the same grid (components  $u$   
 583 and  $v$ ); for example, representing surface velocity. The method presented here allows line integration of  
 584 the vector field across the terminus directly on the gridded representation, without having to convert the  
 585 terminus to a set of line segments.

586 The key observation is that in a gridded environment, *the boundary of the ice sheet follows gridcell*  
 587 *boundaries*, like a “Manhattan” street grid (Figure 20a) because flux of a constant vector field across a

588 line depends only on its endpoints. Integrating a vector field across this boundary will produce a result  
589 approximately equal to integration of the same vector field across a more physically realistic boundary,  
590 which is approximated here in gridded form. Note that the gridded form is “native” to this study, which  
591 identifies the *up area* in gridded form. Therefore, flux across the gridded terminus can be broken into four  
592 parts, which can be summed together for total flux: flux west-to-east, flux east-to-west, flux north-to-south  
593 and flux south-to-north. Without loss of generality, we focus on the west-to-east part of flux.

594 Suppose a grid cell *A* on the terminus with ice flowing west-to-east has been identified (Figure 20b). The  
595 west-to-east flux from grid cell *A* to *B* is exactly the *u* component of the vector field times the length *w*  
596 of the side of the grid cell through which flux is flowing. The *v* component of the vector field contributes  
597 zero here because it is orthogonal to the boundary being integrated across.

598 Grid cells with west-to-east flux are easily identified: they are exactly those that are contained in the  
599 fjord and are ice-covered; and lie just west of another grid cell also contained in the fjord but with no ice.  
600 A mask *m* for such gridcells can be computed using 2-D array operations of shifting and logical AND, in  
601 which *m* is 1 for grid cells with west-to-east flux, and 0 otherwise. Therefore, the total west-to-east flux  
602 for the entire terminus is found by computing *muw* over each grid cell, and then summing over the entire  
603 gridded domain. A similar procedure is used to compute the other three parts of the total flux.

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## 722 SUPPLEMENTS

723 Code and data for this study are available at the NSF arctic data center under  
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725 Supplement S1 provides results for all glaciers are in this study, sorted by category (Figure 12).

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