

Abstract

The Greenland Ice Sheet discharges ice to the ocean through hundreds of outlet glaciers. Recent acceleration of Greenland outlet glaciers has been linked to both oceanic and atmospheric drivers. Here, we leverage temporally dense observations, regional climate model output, and newly developed time series analysis tools to assess the most important forcings causing ice flow variability at one of the largest Greenland outlet glaciers, Helheim Glacier, from 2009 to 2017. We find that ice speed correlates most strongly with catchment-integrated runoff at seasonal to interannual scales, while multi-annual flow variability correlates most strongly with multi-annual terminus variability. The multiple relevant time scales and the influence of subglacial topography on Helheim Glacier’s dynamics highlight different regimes that can inform modeling and forecasting of its future. Notably, our results suggest that the recent terminus history observed at Helheim is a response to, rather than the cause of, upstream changes.

Plain Language Summary

Hundreds of “outlet glaciers” transport ice from the Greenland Ice Sheet to the ocean. The flow of those outlet glaciers has sped up in recent decades, increasing the contribution of Greenland ice to global mean sea-level rise. Previous studies suggest that changes in both the ocean and the atmosphere could cause the observed speedup. In this study, we bring together satellite observations and information from a regional climate model with new software tools to determine which factors in an outlet glacier system are the most likely causes of velocity variation. Based on previous studies, we might expect ocean-driven changes to be the dominant cause of velocity variation for Helheim Glacier. Instead, we find that the most important factors differ at different time scales. Velocity changes over the course of a few months to a year respond to variations in the amount of meltwater in the glacier system, which depends on atmospheric conditions. Over longer time scales, velocity changes respond more strongly to the position of the glacier front, where it meets the ocean. The interaction of multiple factors across time scales highlights the importance of continued efforts to simulate future changes of large Greenland outlet glaciers in detail.

1 Introduction

In recent decades, several glaciers draining the Greenland Ice Sheet have accelerated, increasing their contribution to global mean sea-level rise (Rignot & Kanagaratnam, 2006; Rignot et al., 2011; Bevan et al., 2012). The observed acceleration of outlet glaciers and the ice sheet interior has been attributed to warmer ocean waters melting glacier fronts (Murray et al., 2010; Rignot et al., 2012) as well as increased surface melt (Joughin et al., 2008; Doyle et al., 2014). Numerical models and indirect observations indicate that increasing runoff could enhance solid ice loss by lubricating the glacier bed and warming the ice such that it deforms more readily (Reeh & Olesen, 1986; Krabill et al., 1999; Parizek & Alley, 2004; Phillips et al., 2010; Poinar et al., 2017). However, in situ observations of the Greenland Ice Sheet margin have found limited evidence for annual-scale acceleration of ice flow driven by increasing runoff (Stevens et al., 2016; Nienow et al., 2017). At marine outlets including Helheim Glacier, observations show that ice flow speed (and therefore mass discharge) correlates most strongly with iceberg calving activity rather than runoff (Howat et al., 2005; Joughin et al., 2008; Nettles et al., 2008; Kehrl et al., 2017; Vijay et al., 2019).

Helheim Glacier is currently the highest-flux outlet of the Greenland Ice Sheet (Mankoff et al., 2019). Its dynamics through the early 21st century showed pronounced variability, including episodes of multi-annual retreat and readvance (Howat et al., 2005, 2007; Bevan et al., 2012) and net mass gain while most Greenland outlet glaciers were losing mass (Howat et al., 2011). Sediment records from the past century suggest that Helheim

67 responds to atmospheric and oceanic variability on time scales of a few years (Andresen
68 et al., 2012), highlighting the importance of understanding its dynamics on sub-annual
69 to multi-annual time scales. The high ice flux through Helheim Glacier (Rignot et al.,
70 2004; Mankoff et al., 2019), its recent variability (Stearns & Hamilton, 2007; Howat et
71 al., 2005, 2007), and its sensitivity to short-term variation in climate forcings (Nick et
72 al., 2009; Andresen et al., 2012) motivate a quantitative comparison of hypothesized con-
73 trols on velocity variability.

74 Processes contributing to velocity variability operate at different time scales. For
75 example, fracture-driven changes in stress balance can be nearly instantaneous and prop-
76 agate rapidly, shaping velocity on the order of hours to days (Das et al., 2008; Nettles
77 et al., 2008; Cassotto et al., 2019), while changes in the subglacial drainage system may
78 take days to months (Meier et al., 1994; Kamb et al., 1994; Shepherd et al., 2009; Bartholomew
79 et al., 2010; Pimentel & Flowers, 2011) and response to changing upstream snow accu-
80 mulation can take many years (Weertman, 1958; Nye, 1960; van der Veen, 2001). Ob-
81 servations that permit a detailed understanding of one process – such as intensive field
82 study of a calving front – may not be sufficient to contextualize influences from processes
83 operating at other scales. Accounting for the relative influence of each process, for ex-
84 ample to develop accurate predictive models, requires synthesising observations and in-
85 ference across scales. Here, we apply the flexible time series analysis tools developed by
86 Riel et al. (2021) to publicly available velocity fields (Joughin et al., 2020) and correlate
87 the results with temporally dense climate model output (Van Meijgaard et al., 2008; Noël
88 et al., 2018) and terminus observations (Supplementary Material) to study the forcings
89 of and responses to velocity variability at Helheim over multiple temporal scales.

90 2 Methods

91 2.1 Inference framework

92 We investigate correlations between surface velocity and several factors hypothe-
93 sized to drive its variability at seasonal to multi-annual scales. Figure 1 shows the sys-
94 tem of causal connections we investigate here. Limited time-dependent data precludes
95 us from studying the effect of ice mélange, ocean temperature, and surface damage di-
96 rectly. Here, we assume that the primary effect of those three variables is on the rate of
97 calving, and we restrict the present study to the relationship between glacier terminus
98 position and surface ice velocity. We focus our analysis on time scales of months to years.
99 As such, we do not consider the flow response to individual calving events (Murray et
100 al., 2015) or tidal variation (de Juan et al., 2010; Voytenko et al., 2015), which have been
101 described elsewhere.

102 We investigate three factors varying in time (surface mass balance, runoff, and width-
103 averaged terminus position) and one varying in space (subglacial topography). To quan-
104 tify the strength of the temporal variables’ relationship with velocity, we compute their
105 cross-correlation as described below. We interpret the qualitative effect of local topog-
106 raphy on velocity variation by analysing spatial patterns in the cross-correlations com-
107 puted for the temporal variables.

108 2.2 Catchment data

109 We produce a one-dimensional time series for each catchment variable. We inte-
110 grate monthly surface mass balance and runoff derived from Noël et al. (2018) over the
111 Helheim Glacier catchment defined by Mankoff et al. (2020). The time series of calving
112 front position is a width-averaged distance from an upstream flux gate, identified from
113 satellite imagery (Supplementary Dataset S1) with variable temporal resolution. For the
114 present study of seasonal to multi-annual time scales, we apply a 10-day smoothing win-
115 dow to the terminus record. We trim all time series to the period for which data is avail-

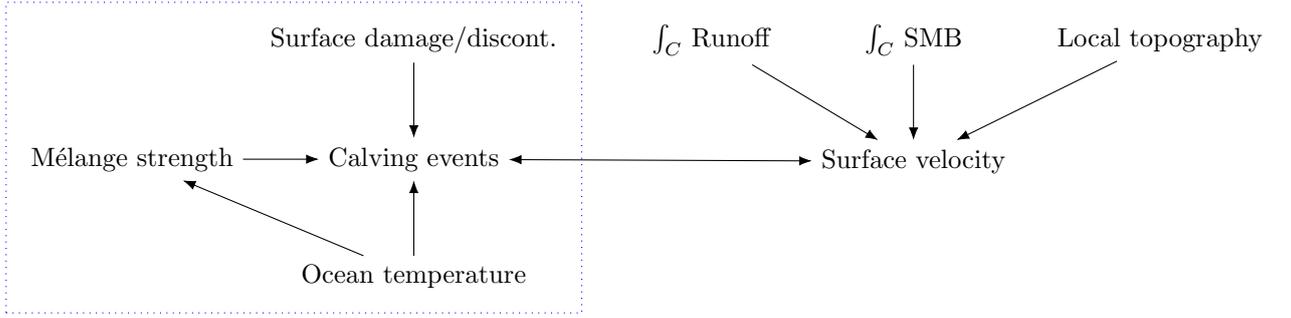


Figure 1. Causal relationships tested in this work. Surface mass balance (SMB) and runoff are catchment-integrated quantities (\int_C), both from (Noël et al., 2018). Surface velocity from (Joughin et al., 2020) is evaluated at each point. We use terminus position from satellite imagery (Section 2.2) as a proxy for all ocean-driven processes (blue dotted box)

116 able for all variables: 2009-2017. We interpolate a piecewise linear time-continuous func-
 117 tion for each time series using the Interp1d class of SciPy v1.4.1 (Virtanen et al., 2020).

118 **2.3 Producing temporally continuous velocity functions**

119 We use frequent observations and spline interpolation to produce time-continuous
 120 estimates of ice surface velocity. We stack all available InSAR-derived glacier site ve-
 121 locity observations from Joughin et al. (2020) and extract 1-dimensional time series of ve-
 122 locity at points spaced at 1 km intervals along a central flowline (as defined in Felikson
 123 et al., 2021). We define an upstream limit to our analysis by the area for which there
 124 are sufficient velocity observations to constrain a time-continuous fit. The selected points
 125 are shown in Figure 2A-B.

126 We then construct a continuous function that best fits the observed values at each
 127 point. Following Riel et al. (2021), we perform a regularized least squares regression that
 128 estimates the optimal linear combination of representative time functions (linear poly-
 129 nomials, B-splines, and integrated B-splines of pre-defined center times and scales) to
 130 fit the data at each point. The resulting function is an optimized superposition of lin-
 131 ear trend, seasonal variability, and secular change, which facilitates later decomposition
 132 into components of interest. Example observations and constructed continuous functions
 133 are shown in Figure 2C.

134 **2.4 Normalized cross-correlation**

135 Finally, we find and compare the cross-correlations describing ice speed response
 136 to each variable at each point. We sample each time-continuous function at regular in-
 137 tervals. Dickey-Fuller and KPSS tests applied using the Python package statsmodels v0.12.2
 138 (Seabold & Perktold, 2010) indicate that the raw time series are non-stationary — that
 139 is, their means and/or variances change over time, which can produce spurious results
 140 in cross-correlation analysis (Shumway & Stoffer, 2017). In sections 3.1-3.2, we enforce
 141 stationarity by differencing:

$$f_i = \hat{f}_i - \hat{f}_{i-1}, \tag{1}$$

142 where \hat{f}_i is the i^{th} point in the raw time series and f is the differenced time series. We
 143 elect not to difference the long-term-varying series tested in section 3.3, as doing so would
 144 remove the signal of interest.

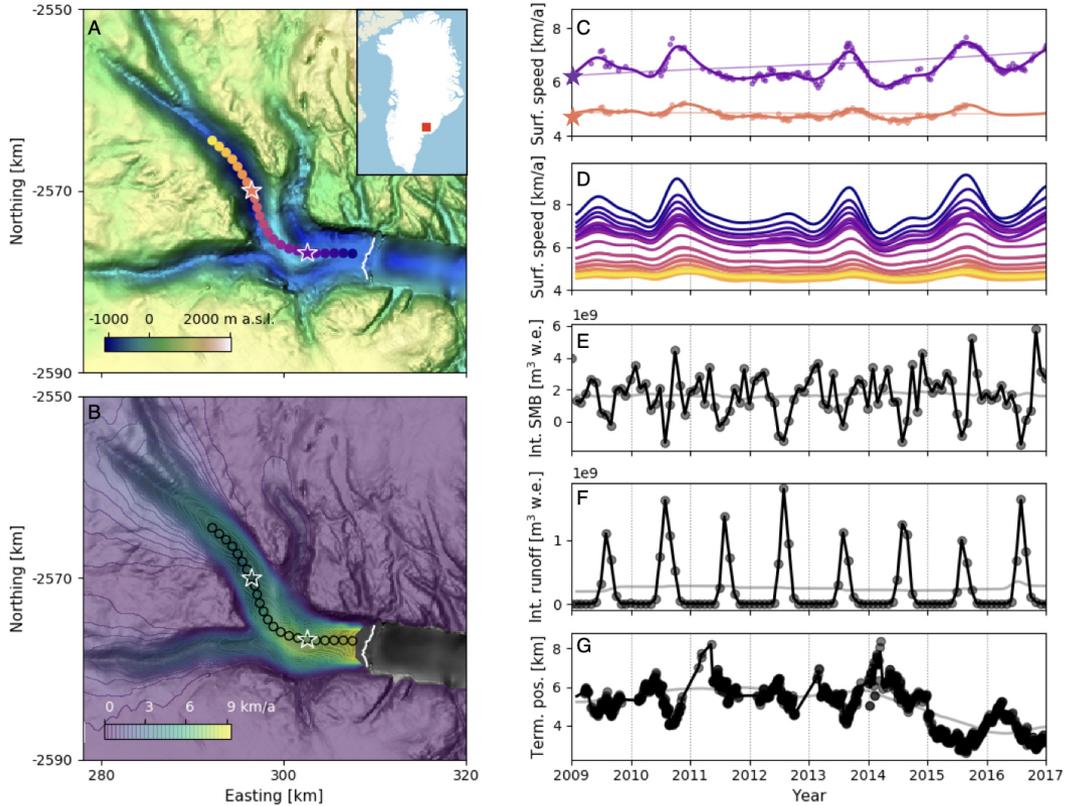


Figure 2. The physical setting of Helheim Glacier studied here. (A) Hillshade map of Helheim Glacier subglacial topography from Morlighem et al. (2017) with 2009 terminal edge from Joughin et al. (2020) in white, points along central flowline in bright colors, and inset map of Helheim Glacier location within Greenland; (B) Mean ice surface speed as of 2016 (ENVEO, 2017), with flowline points outlined; (C) Ice surface speed at two locations from Joughin et al. (2020) (points) and B-spline smooth approximation to each time series (curves); (D) B-spline continuous velocity functions for each point along the flowline in panel A, with curve color indicating which point is represented; (E) Catchment-integrated surface mass balance from RACMO; (F) Catchment-integrated runoff from RACMO (Noël et al., 2018); and (G) Width-averaged terminus position, relative to a fixed gate on the glacier (larger numbers indicate advance). In panels E-G, data from the original source is plotted as points, and dark lines show the values of 1d-interpolated functions used to determine signal cross-correlation. In panels C and E-G, light curves show the long-term-varying component of each signal.

145 We compute the normalized cross-correlation at lag k ,

$$XCorr(f, v)_k = \frac{1}{N} \sum_{i=1}^N \frac{f_{i+k} - \bar{f}}{\sigma(f)} \frac{v_i - \bar{v}}{\sigma(v)}, \quad (2)$$

146 for $k \in [-N, N]$, where ice speed v and variable f are each time series of length N , dif-
 147 ferenced as in Eqn. 1, with means (\bar{v}, \bar{f}) and standard deviations $(\sigma(v), \sigma(f))$. With this
 148 convention, a lag $k < 0$ refers to a cross-correlation with the velocity series offset back-
 149 ward in time; that is, strong cross-correlations at negative lag indicate that a change is
 150 observed first in the velocity signal and a similar change is observed later in the variable
 151 f signal. The normalized cross-correlation may take values between ± 1 , and a cross-correlation
 152 at lag k is statistically significant at the 95% confidence level if it exceeds $1.96/\sqrt{N-k}$.

153 Because we anticipate multiple influences on observed surface velocity (Figure 1),
 154 we do not expect the magnitude of correlations to be close to 1. Rather, we identify the
 155 largest-magnitude statistically significant correlations for each variable at each point, and
 156 we compare their relative strength. From the full time series (Section 3.1) and then from
 157 annual subsets (Section 3.2) and from series filtered to show only multi-annual variabil-
 158 ity (Section 3.3), we identify the largest magnitude of cross-correlation between the se-
 159 ries and the lag in days at which that extreme value occurs. We analyse both positive
 160 and negative lag times for terminus position, given the bidirectional causal relationship
 161 we expect for that variable (Figure 1). We restrict our analysis to positive lag values for
 162 surface mass balance and runoff. The quasi-periodic nature of those signals is likely to
 163 produce significant cross-correlations at negative lags, but there is no physical reason to
 164 expect feedbacks from velocity to mass balance or runoff at seasonal to multi-annual time
 165 scales.

166 3 Results

167 3.1 Seasonal to interannual velocity variability responds most strongly 168 to runoff

169 The normalized, single-differenced cross-correlations with ice surface speed are dis-
 170 tinct for each variable. The weakest cross-correlations along the flowline, on average, are
 171 with catchment-integrated surface mass balance (Fig. 3, left column). For that variable,
 172 the strongest negative correlation is -0.17 , found at the farthest downstream point. The
 173 strongest positive correlation is 0.13 , found 14 km upstream. Cross-correlation of ice sur-
 174 face speed with catchment-integrated runoff (Fig. 3, center column) is stronger. Its peak
 175 positive value is 0.25 , found near the terminus, and its strongest negative value is -0.20 ,
 176 found 19 km upstream. Terminus position (Fig. 3, right column) also shows compara-
 177 tively strong cross-correlations with velocity. The strongest correlation is -0.22 , found
 178 near the terminus, and the strongest positive correlation is 0.19 , found 14 km upstream
 179 from the terminus. However, the strongest cross-correlations are found at negative lags,
 180 suggesting that terminus position is responding to upstream velocity variation rather than
 181 vice versa.

182 At every point, the magnitude of strongest cross-correlation with velocity is larger
 183 for runoff than for surface mass balance, on average 1.2 times larger over the flowline.
 184 The cross-correlation between terminus position and velocity is similar in magnitude to
 185 that of runoff, but the former appears to be a response to velocity changes while the lat-
 186 ter leads velocity changes. We infer that runoff exerts the strongest control on seasonal
 187 to interannual ice surface velocity variability along the main trunk of Helheim Glacier.

188 3.2 No year in which terminus position is more important than runoff

189 Because Helheim Glacier is a complex system that changes over time, the multi-
 190 year bulk analysis of section 3.1 may not capture important interannual changes in the

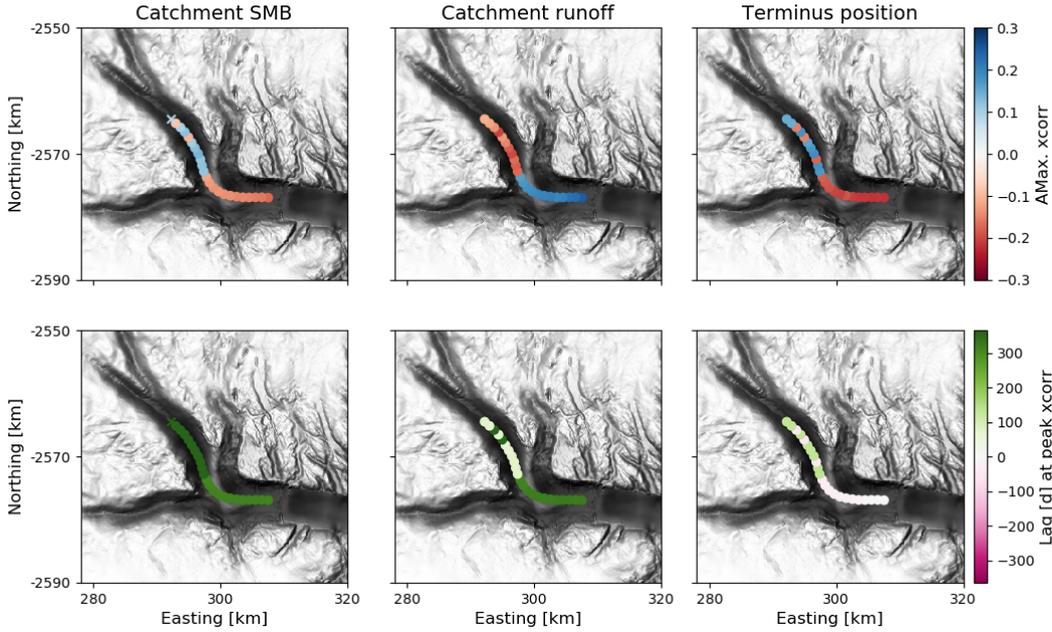


Figure 3. The cross-correlation of largest absolute value (“AMax. xcorr”) (top row) between ice surface speed and each variable (columns), and the lag in days (bottom row) at which that cross-correlation is found. Circles and crosses indicate values that are and are not significant at the 95% confidence level, respectively.

191 dominant sources of its velocity variability. To study year-to-year changes in more de-
 192 tail, we computed the cross-correlation between single-year subsets of the variables we
 193 studied in section 3.1. Cross-correlations of these single-year subsets are generally stronger
 194 than those found over the full time period. We present both positive and negative lags
 195 for all variables, as the signals are quasi-periodic.

196 The patterns of cross-correlation between single-year sections of the signals vary
 197 from year to year, as shown in Figure 4. For example, in 2009 the cross-correlation be-
 198 tween runoff and surface speed at a downstream point is strongest at a lag of around 90
 199 days, with a statistically significant minimum following at longer lag times. In 2010, 2012,
 200 and 2013, the strongest correlation between runoff and surface speed is in negative lag
 201 space, which we interpret as a response to past years’ runoff peaks. There are statisti-
 202 cally significant correlations between runoff and surface speed every year.

203 We find that the normalized cross-correlation of terminus position and ice surface
 204 speed is generally low. That correlation is statistically significant in only four of the eight
 205 years we analyse. The strongest of those correlations are in 2010 and 2013, where there
 206 is a significant correlation for small negative lags, indicating that terminus position changed
 207 in response to a velocity change. For every year we study, ice surface speed correlates
 208 much more strongly with catchment-integrated runoff than with terminus position.

209 3.3 Multi-annual velocity variability correlates with terminus position

210 We apply a boxcar filter with window of 2 years to the surface mass balance, runoff,
 211 and terminus position data to isolate long-term (multi-annual) from shorter-term vari-
 212 ability. We then extract the long-term-varying signal from the ice speed timeseries as
 213 discussed in Section 2.3. The isolated long-term components are shown as light curves
 214 in Figure 2. Finally, we recompute the cross-correlation for the filtered time series.

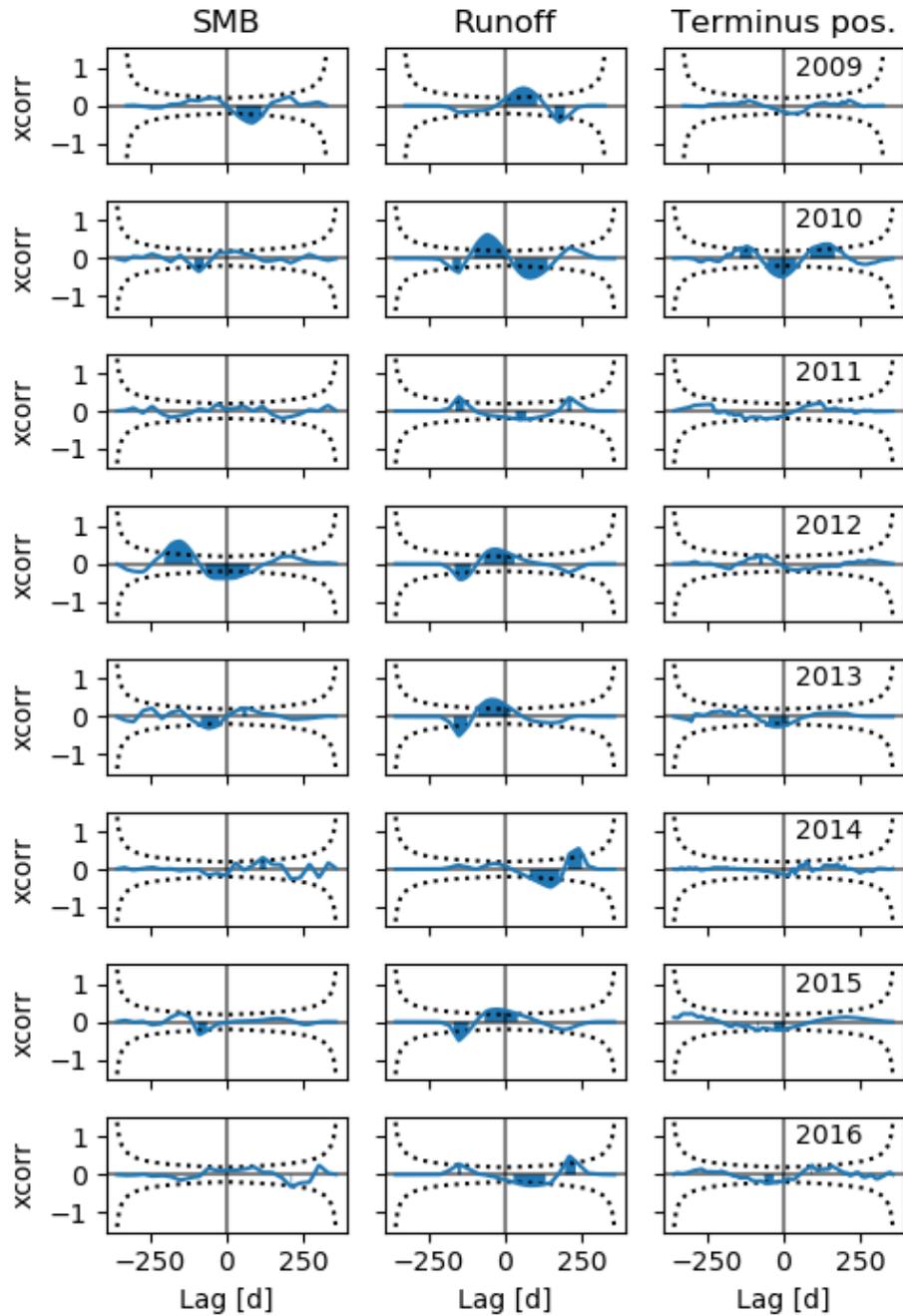


Figure 4. Annual patterns of cross-correlation between surface speed and system variables for (left) surface mass balance, (center) runoff, and (right) terminus position, sampled at a point 5 km upstream from the 2009 terminus. Dotted curves indicate 95% confidence intervals around $XCorr(f, v) = 0$; shading indicates statistically significant difference from zero.

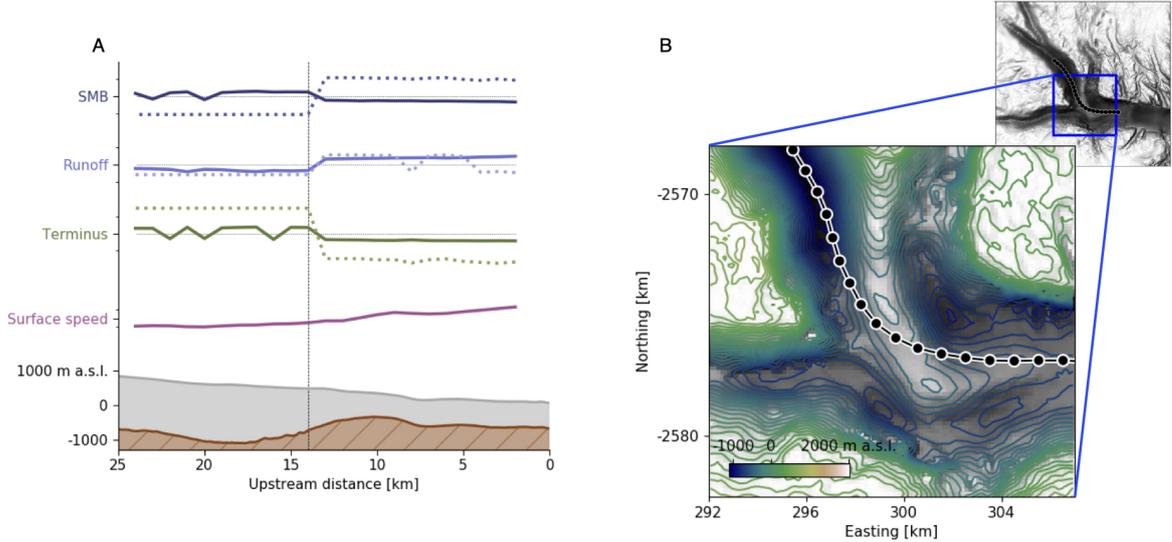


Figure 5. Influence of a subglacial ridge on Helheim Glacier dynamics. A) Ice speed cross-correlation with each variable tested, for each point along the flowline, vertically off-set for legibility. Variable labels coincide with zero cross-correlation and minor ticks indicate $XCorr(f, v) = \pm 0.5$. Solid lines are cross-correlations of the full signals (as reported in Figure 3 and Section 3.1). Dashed lines show results filtered to isolate long-term variability (as in Section 3.3 and Figure S1). Lower portion shows bed topography (brown), ice surface (grey), and mean surface speed (purple) along the flowline. Velocity ticks correspond to 4.5, 6.0, and 7.5 km a⁻¹. Vertical marker indicates position of sign changes in cross-correlation for multiple variables. B) Enlarged contour map of the Helheim Glacier trough around the bedrock bump. Background image is a black and white hillshade of the topography as in Figure 3; contours show intervals of approximately 60 meters elevation. Contour colormap and flowline points (black) are consistent with Figure 2A.

215 We see a strong correlation between the long-term-varying components of ice speed
 216 and terminus position. The correlation between these two component signals is much stronger
 217 than between the corresponding full signals (Figures 5 and S1), with values along the
 218 lower trunk averaging -0.8 , all for non-negative lags. A cross-correlation stronger than
 219 that for the full signals is also seen for long-term-varying surface mass balance, ranging
 220 from -0.54 to 0.54 . The correlation between long-term-varying components of ice speed
 221 and runoff is comparable to that between the full signals, ranging from -0.29 to 0.29 . We
 222 infer that terminus position variability is most important for Helheim Glacier’s dynam-
 223 ics at multi-annual time scales.

224 **3.4 Subglacial topography modulates velocity response to each variable**

225 The flowline we examine flows through a trough with a pronounced ridge in its subglacial
 226 topography. The ridge creates a steep along-flow thickness gradient as well as a
 227 lateral constriction (Figure 5). For all three variables, the flowline separates into two seg-
 228 ments with opposite sign of maximum cross-correlation. We find changes in sign of ab-
 229 solute maximum cross-correlation with velocity at 14 km upstream from the terminus—
 230 coincident with the upstream edge of the subglacial ridge (Figures 3, 5, and S1). We also
 231 find step changes in the lag at which the strongest cross-correlation occurs aligned with
 232 the ridge. The spatial pattern of cross-correlation is similar for both seasonal and multi-

233 annual signals (Sections 3.1-3.3). These patterns suggest that the dynamics of the up-
234 stream and downstream segments of the flowline are fundamentally different from one
235 another. We interpret that the bedrock ridge is an obstacle to the propagation of trav-
236 eling waves (Nye, 1960; Fowler, 1982; Weertman & Birchfield, 1983)—whether the waves
237 carry velocity variability, seasonal runoff input, or stress adjustment to terminus change
238 or surface mass balance.

239 4 Discussion

240 Our analysis illustrates that Helheim Glacier is a dynamic system with more than
241 one important control on its velocity. We find that seasonal-scale variations in ice sur-
242 face speed correlate more strongly with catchment-integrated runoff than with terminus
243 position, for the full period 2009-2017 (Figure 3) and for every year in it (Figure 4). At
244 the multi-annual scale we find stronger correlation with terminus position than with runoff
245 (Figure 5), consistent with the findings of Vijay et al. (2019), and in agreement with ear-
246 lier work relating ice velocity to ice thickness and glacier terminus position on Alaskan
247 tidewater glaciers (Meier & Post, 1987; O’Neel et al., 2005). Our results support pre-
248 vious findings that increasing meltwater supply can enhance seasonal speedups in ice flow,
249 but does not contribute to multi-annual acceleration (summarized in Nienow et al., 2017).
250 Our analysis also supports the assessment by Enderlin et al. (2018) that hypothesised
251 distinct variables driving different timescales of velocity variability at Columbia Glacier,
252 Alaska.

253 The correlation of runoff with seasonal-scale velocity variation described in Sec-
254 tion 3.1 is consistent with observations of land-terminating margins of the Greenland Ice
255 Sheet (Joughin et al., 2008) and some marine outlets on Greenland’s west coast (Sole
256 et al., 2011; Moon et al., 2015), as well as inference of surface-melt-induced acceleration
257 (Andersen et al., 2011) and dynamic thinning (Bevan et al., 2015) at Helheim. Although
258 our conclusions differ from Moon et al. (2014) and Vijay et al. (2019), who infer that ter-
259 minus changes are the strongest control on Helheim’s velocity, the observations presented
260 in those studies are consistent with our interpretation that seasonal variations in ice sur-
261 face speed are more strongly correlated with runoff than with terminus position change.
262 In agreement with Kehrl et al. (2017), we find that 2010 and 2013 are the years for which
263 ice surface speed at Helheim is most correlated with terminus position (Fig. 4). How-
264 ever, our quantitative analysis shows that, in all years, Helheim’s speed is more corre-
265 lated with runoff than terminus position.

266 The relative importance of each driver at Helheim Glacier likely does not trans-
267 late to other outlets or other time periods. For example, our findings at Helheim con-
268 trast those of King et al. (2020), who found that regionally aggregated trends in Green-
269 land Ice Sheet discharge correlated most strongly to glacier front position. Ice velocity
270 at Helheim may be unusually sensitive to catchment-integrated runoff because of the pres-
271 ence of a large firn aquifer that allows hydrofracturing of deep crevasses and enhances
272 deformational ice motion (Poinar et al., 2017; Miller et al., 2020). The lower trunk of
273 the glacier was also near flotation during the time period we study here (Kehrl et al.,
274 2017), which could render it especially sensitive to both changing basal water pressure
275 (runoff) and calving activity (Andersen et al., 2010; Cassotto et al., 2019). Finally, the
276 spatial pattern of our results highlights the role of unique subglacial topography in shap-
277 ing the dynamic response to forcing (Enderlin et al., 2013; Khan et al., 2014; Felikson
278 et al., 2017; Catania et al., 2018; Enderlin et al., 2018; Felikson et al., 2021). Although
279 our results may be unique to Helheim during the 2009-2017 period, our methods can be
280 applied to investigate any glacier with a sufficient observational record.

281 One explanation for the weak correlation between ice surface speed and terminus
282 position throughout our study period is that the sensitivity of surface speed to termi-
283 nus position is itself determined by the terminus position (Cassotto et al., 2019), and that

284 the terminus did not reach a hypothetical critical position during the time we observed.
285 From 2009-2014, the observed terminus positions oscillated around a steady mean po-
286 sition at approximately 6 km forward of our reference position; a period of multi-annual
287 retreat beginning in late 2014 reflects a multi-annual acceleration on the lower glacier
288 trunk beginning around the same time (Figure 2D and G). If the terminus had reached
289 a critical position that increased the sensitivity of surface speed to terminus change, we
290 would expect to see change in the correlation between those variables as terminus po-
291 sition changed over time. Instead we find that the annual cross-correlation between sur-
292 face speed and terminus position is no stronger in 2015 and 2016 than in previous years
293 (Figure 4).

294 A second explanation for the weak correlation between ice surface speed and ter-
295 minus position is that iceberg calving is episodic and discontinuous. Field observations
296 of Helheim Glacier at finer temporal scales than we study here have found that calving
297 activity was an important control on velocity at the timescale of minutes to hours (Nettles
298 et al., 2008; de Juan et al., 2010), and that runoff during the melt season contributes to
299 daily velocity increases (Andersen et al., 2010, 2011). Thus, even with our temporally
300 dense records—average 3 days between measurements—we may more realistically expect
301 to see responses to runoff than to iceberg calving. Further, we analyse a width-averaged
302 terminus position, which will not capture differing dynamic responses to iceberg calv-
303 ing at different points along the face. Extending our methodology to analyse the fine spa-
304 tial and temporal scales captured in field observations could provide a fuller picture of
305 the forcings driving velocity variability (building on Podrasky et al., 2012, for example).

306 In this work, we have assumed that terminus position evolves independently from
307 catchment-integrated runoff (Figure 1). This choice ignores the established connection
308 between calving rate and subglacial discharge at the terminus (Bartholomaeus et al., 2013;
309 Cook et al., 2014; Slater et al., 2015; Fried et al., 2018; van Dongen et al., 2020). Mod-
310 elling efforts suggest that the calving response to subglacial discharge depends on the
311 subglacial hydrologic system near the terminus, in particular whether melt is localized
312 to channels (Slater et al., 2015; Todd et al., 2018; van Dongen et al., 2020). Subglacial
313 discharge also affects calving through its influence on the vertical pattern of submarine
314 melt (Motyka et al., 2003; Jenkins, 2011; O’Leary & Christoffersen, 2013; Luckman et
315 al., 2015; Slater et al., 2015; Ma & Bassis, 2019). Recent observations have found no ev-
316 idence for a melt-induced enhancement of calving at Helheim Glacier, perhaps because
317 of its broad and deep terminus (Everett et al., 2021). However, we anticipate that ac-
318 counting for a connection between runoff and calving activity would further strengthen
319 the apparent role of runoff in setting Helheim Glacier surface velocity. As new observa-
320 tions of the near-terminus environment become available, future work may apply mul-
321 tivariate statistical methods to assess this prediction quantitatively.

322 Our results show that numerical ice flow modeling experiments will require mul-
323 tiple forcing mechanisms to capture the dynamics of Helheim Glacier. Several state-of-
324 the-art studies, including the standard experiments performed by several numerical mod-
325 els as part of the the Ice Sheet Modeling Intercomparison for the Coupled Model Inter-
326 comparison Project Phase 6 (“ISMIP6”, Nowicki et al., 2020), have used projections of
327 outlet glacier terminus positions to force Greenland Ice Sheet mass change simulations
328 (Choi et al., 2017; Morlighem et al., 2019). Our results show that this approach is a good
329 strategy for projections of multi-annual changes of glaciers like Helheim. However, if fu-
330 ture ice sheet modeling efforts seek to reproduce seasonal velocity changes, runoff forc-
331 ing must be included. The continued development of subglacial hydrology models (Pimentel
332 & Flowers, 2011; Werder et al., 2013) and efforts to couple them with ice dynamics mod-
333 els (Aschwanden et al., 2016; Brinkerhoff et al., 2021) are therefore vital to refining our
334 understanding of the future evolution of the Greenland Ice Sheet.

5 Conclusions

We have computed normalized cross-correlations between three catchment variables (surface mass balance, runoff, and terminus positions) and ice surface velocity of Helheim Glacier, revealing the dominant controls on velocity variability at multiple time scales. We find that velocity responds most strongly to catchment-integrated runoff at seasonal scale. At multi-annual scale, velocity variability shows stronger correlation with terminus position change. We find distinct patterns in correlation along upstream and downstream portions of the glacier trunk, separated by a subglacial ridge. The time scale separation of major sources of variability, and the role of underlying topography, are important considerations in designing numerical ice flow simulators to project the future evolution of large outlet glaciers.

Acknowledgments

LU designed the study, with input from DF and BM. DF gathered model data and contributed literature review. LS contributed a dense record of satellite-derived terminus positions. BR developed software used to construct time-continuous velocity functions. LU performed quantitative analysis and produced manuscript figures. LU and DF drafted the manuscript, and all authors contributed to editing and approving its final form.

Work toward this manuscript was supported by the National Aeronautics and Space Administration grant NNX16AJ90G and Heising Simons Foundation grant 2017-316.

All code used in this analysis is available via GitHub and archived on Zenodo. Construction of the time-continuous velocity functions: <https://doi.org/10.5281/zenodo.4474829>. Along-flowline data extraction and cross-correlation: <https://doi.org/10.5281/zenodo.4707999>. Data pre-processing and visualization: <https://doi.org/10.5281/zenodo.4707997>. Terminus position data is available as a supplement to this manuscript and [will be deposited] in the Dryad data repository.

The authors thank Brice P. Y. Noël for providing RACMO surface mass balance data and discussing its processing. LU thanks Jeremy Bassis for long-ago chats about inference on weighted directed graphs, which percolated into Figure 1.

References

- Andersen, M. L., Larsen, T. B., Nettles, M., Elosegui, P., van As, D., Hamilton, G. S., ... Dahl-Jensen, D. (2010). Spatial and temporal melt variability at Helheim Glacier, East Greenland, and its effect on ice dynamics. *Journal of Geophysical Research: Earth Surface*, *115*(F4). doi: 10.1029/2010JF001760
- Andersen, M. L., Nettles, M., Elosegui, P., Larsen, T. B., Hamilton, G. S., & Stearns, L. A. (2011). Quantitative estimates of velocity sensitivity to surface melt variations at a large Greenland outlet glacier. *Journal of Glaciology*, *57*(204), 609-620. doi: 10.3189/002214311797409785
- Andresen, C. S., Straneo, F., Ribergaard, M. H., Bjørk, A. A., Andersen, T. J., Kuipers, A., ... Ahlstrøm, A. P. (2012). Rapid response of Helheim Glacier in Greenland to climate variability over the past century. *Nature Geoscience*, *5*(1), 37-41. doi: 10.1038/ngeo1349
- Aschwanden, A., Fahnestock, M. A., & Truffer, M. (2016). Complex Greenland outlet glacier flow captured. *Nature Communications*, *7*, 10524 EP. doi: 10.1038/ncomms10524
- Bartholomäus, T. C., Larsen, C. F., & Oneel, S. (2013). Does calving matter? Evidence for significant submarine melt. *Earth and Planetary Science Letters*, *380*, 21-30. doi: 10.1016/j.epsl.2013.08.014
- Bartholomew, I., Nienow, P., Mair, D., Hubbard, A., King, M. A., & Sole, A. (2010). Seasonal evolution of subglacial drainage and acceleration in a Greenland

- 384 outlet glacier. *Nature Geoscience*, 3(6), 408–411. doi: 10.1038/ngeo863
- 385 Bevan, S. L., Luckman, A., Khan, S. A., & Murray, T. (2015). Seasonal dynamic
386 thinning at Helheim Glacier. *Earth and Planetary Science Letters*, 415, 47–53.
387 doi: 10.1016/j.epsl.2015.01.031
- 388 Bevan, S. L., Luckman, A. J., & Murray, T. (2012). Glacier dynamics over
389 the last quarter of a century at Helheim, Kangerdlugssuaq and 14 other
390 major Greenland outlet glaciers. *The Cryosphere*, 6(5), 923–937. doi:
391 10.5194/tc-6-923-2012
- 392 Brinkerhoff, D., Aschwanden, A., & Fahnestock, M. (2021). Constraining subglacial
393 processes from surface velocity observations using surrogate-based Bayesian
394 inference. *Journal of Glaciology*, 1-19. doi: 10.1017/jog.2020.112
- 395 Cassotto, R., Fahnestock, M., Amundson, J. M., Truffer, M., Boettcher, M. S.,
396 De La Peña, S., & Howat, I. (2019). Non-linear glacier response to calving
397 events, Jakobshavn Isbræ, Greenland. *Journal of Glaciology*, 65(249), 39-54.
398 doi: 10.1017/jog.2018.90
- 399 Catania, G. A., Stearns, L. A., Sutherland, D. A., Fried, M. J., Bartholomaeus, T. C.,
400 Morlighem, M., ... Nash, J. (2018). Geometric controls on tidewater glacier
401 retreat in central western Greenland. *Journal of Geophysical Research: Earth
402 Surface*, 123(8), 2024-2038. doi: 10.1029/2017JF004499
- 403 Choi, Y., Morlighem, M., Rignot, E., Mouginot, J., & Wood, M. (2017). Modeling
404 the response of Nioghalvfjærdsfjorden and Zachariae Isstrøm glaciers, Green-
405 land, to ocean forcing over the next century. *Geophysical Research Letters*,
406 44(21), 11,071–11,079. doi: 10.1002/2017GL075174
- 407 Cook, S., Rutt, I. C., Murray, T., Luckman, A., Zwinger, T., Selmes, N., ... James,
408 T. D. (2014). Modelling environmental influences on calving at Hel-
409 heim Glacier in eastern Greenland. *The Cryosphere*, 8(3), 827–841. doi:
410 10.5194/tc-8-827-2014
- 411 Das, S. B., Joughin, I., Behn, M. D., Howat, I. M., King, M. A., Lizarralde, D., &
412 Bhatia, M. P. (2008). Fracture propagation to the base of the Greenland Ice
413 Sheet during supraglacial lake drainage. *Science*, 320(5877), 778–781. doi:
414 10.1126/science.1153360
- 415 de Juan, J., Elósegui, P., Nettles, M., Larsen, T. B., Davis, J. L., Hamilton, G. S.,
416 ... Forsberg, R. (2010). Sudden increase in tidal response linked to calving
417 and acceleration at a large Greenland outlet glacier. *Geophysical Research
418 Letters*, 37(12). doi: 10.1029/2010GL043289
- 419 Doyle, S. H., Hubbard, A., Fitzpatrick, A. A. W., van As, D., Mikkelsen, A. B., Pet-
420 tersson, R., & Hubbard, B. (2014, 2021/02/26). Persistent flow acceleration
421 within the interior of the Greenland ice sheet. *Geophysical Research Letters*,
422 41(3), 899–905. doi: 10.1002/2013GL058933
- 423 Enderlin, E. M., Howat, I. M., & Vieli, A. (2013). High sensitivity of tidewater out-
424 let glacier dynamics to shape. *The Cryosphere*, 7(3), 1007–1015. doi: 10.5194/
425 tc-7-1007-2013
- 426 Enderlin, E. M., O’Neel, S., Bartholomaeus, T. C., & Joughin, I. (2018). Evolving
427 environmental and geometric controls on Columbia Glacier’s continued re-
428 treat. *Journal of Geophysical Research: Earth Surface*, 123, 1528 - 1545. doi:
429 10.1029/2017JF004541
- 430 ENVEO. (2017). *Greenland ice velocity map 2016/2017 from Sentinel-1 [ver-*
431 *sion 1.0]*. [http://products.esa-icesheets-cci.org/products/details/
432 greenland_ice_velocity_map_winter_2016_2017_v1_0.zip/](http://products.esa-icesheets-cci.org/products/details/greenland_ice_velocity_map_winter_2016_2017_v1_0.zip/).
- 433 Everett, A., Murray, T., Selmes, N., Holland, D., & Reeve, D. E. (2021). The
434 impacts of a subglacial discharge plume on calving, submarine melting
435 and mélange mass loss at Helheim Glacier, south east Greenland. *Jour-
436 nal of Geophysical Research: Earth Surface*, e2020JF005910. doi: 10.1029/
437 2020JF005910
- 438 Felikson, D., A. Catania, G., Bartholomaeus, T. C., Morlighem, M., & Noël,

- 439 B. P. Y. (2021). Steep glacier bed knickpoints mitigate inland thinning
440 in Greenland. *Geophysical Research Letters*, *48*(2), e2020GL090112. doi:
441 <https://doi.org/10.1029/2020GL090112>
- 442 Felikson, D., Bartholomaus, T. C., Catania, G. A., Korsgaard, N. J., Kjær, K. H.,
443 Morlighem, M., . . . Nash, J. D. (2017). Inland thinning on the Greenland ice
444 sheet controlled by outlet glacier geometry. *Nature Geoscience*, *10*(5), 366–369.
445 doi: 10.1038/ngeo2934
- 446 Fowler, A. C. (1982). Waves on glaciers. *Journal of Fluid Mechanics*, *120*, 283–321.
447 doi: 10.1017/S0022112082002778
- 448 Fried, M. J., Catania, G. A., Stearns, L. A., Sutherland, D. A., Bartholomaus,
449 T. C., Shroyer, E., & Nash, J. (2018). Reconciling drivers of seasonal ter-
450 minus advance and retreat at 13 Central West Greenland tidewater glaciers.
451 *Journal of Geophysical Research: Earth Surface*, *123*(7), 1590–1607. doi:
452 10.1029/2018JF004628
- 453 Howat, I. M., Ahn, Y., Joughin, I., van den Broeke, M. R., Lenaerts, J. T. M.,
454 & Smith, B. (2011). Mass balance of Greenland’s three largest outlet
455 glaciers, 2000–2010. *Geophysical Research Letters*, *38*(12), L12501. doi:
456 10.1029/2011GL047565
- 457 Howat, I. M., Joughin, I., & Scambos, T. A. (2007, 03). Rapid changes in ice dis-
458 charge from Greenland outlet glaciers. *Science*, *315*(5818), 1559. doi: 10.1126/
459 science.1138478
- 460 Howat, I. M., Joughin, I., Tulaczyk, S., & Gogineni, S. (2005). Rapid retreat and
461 acceleration of Helheim Glacier, east Greenland. *Geophysical Research Letters*,
462 *32*(22), L22502. doi: 10.1029/2005GL024737
- 463 Jenkins, A. (2011). Convection-driven melting near the grounding lines of ice shelves
464 and tidewater glaciers. *Journal of Physical Oceanography*, *41*(12), 2279–2294.
465 doi: 10.1175/JPO-D-11-03.1
- 466 Joughin, I., Das, S. B., King, M. A., Smith, B. E., Howat, I. M., & Moon, T. (2008).
467 Seasonal speedup along the western flank of the Greenland Ice Sheet. *Science*,
468 *320*(5877), 781–783. doi: 10.1126/science.1153288
- 469 Joughin, I., Howat, I. M., Smith, B., & Scambos, T. (2020). *MEaSUREs Green-*
470 *land Ice Velocity: Selected Glacier Site Velocity Maps from InSAR, Version 3.*
471 NASA National Snow and Ice Data Center Distributed Active Archive Center.
472 Boulder, Colorado USA. doi: 10.5067/YXMJRME5OUNC
- 473 Joughin, I., Smith, B., Howat, I. M., Scambos, T., & Moon, T. (2010). Greenland
474 flow variability from ice-sheet-wide velocity mapping. *Journal of Glaciology*,
475 *56*(197), 415–430. doi: 10.3189/002214310792447734
- 476 Kamb, B., Engelhardt, H., Fahnestock, M. A., Humphrey, N., Meier, M., & Stone,
477 D. (1994). Mechanical and hydrologic basis for the rapid motion of a large
478 tidewater glacier: 2. Interpretation. *Journal of Geophysical Research: Solid*
479 *Earth*, *99*(B8), 15231–15244. doi: 10.1029/94JB00467
- 480 Kehrl, L. M., Joughin, I., Shean, D. E., Floricioiu, D., & Krieger, L. (2017). Sea-
481 sonal and interannual variabilities in terminus position, glacier velocity, and
482 surface elevation at Helheim and Kangerlussuaq Glaciers from 2008 to 2016.
483 *Journal of Geophysical Research: Earth Surface*, *122*(9), 1635–1652. doi:
484 10.1002/2016JF004133
- 485 Khan, S. A., Kjeldsen, K. K., Kjær, K. H., Bevan, S., Luckman, A., Aschwanden,
486 A., . . . Fitzner, A. (2014). Glacier dynamics at Helheim and Kangerdlugssuaq
487 glaciers, southeast Greenland, since the Little Ice Age. *The Cryosphere*, *8*(4),
488 1497–1507. doi: 10.5194/tc-8-1497-2014
- 489 King, M. D., Howat, I. M., Candela, S. G., Noh, M. J., Jeong, S., Noël, B. P. Y., . . .
490 Negrete, A. (2020). Dynamic ice loss from the Greenland Ice Sheet driven by
491 sustained glacier retreat. *Communications Earth & Environment*, *1*(1), 1. doi:
492 10.1038/s43247-020-0001-2
- 493 Krabill, W., Frederick, E., Manizade, S., Martin, C., Sonntag, J., Swift, R., . . . Yun-

- 494 gel, J. (1999, 03). Rapid thinning of parts of the southern Greenland Ice Sheet.
 495 *Science*, 283(5407), 1522. doi: 10.1126/science.283.5407.1522
- 496 Luckman, A., Benn, D. I., Cottier, F., Bevan, S., Nilsen, F., & Inall, M. (2015).
 497 Calving rates at tidewater glaciers vary strongly with ocean temperature.
 498 *Nature Communications*, 6(1), 8566. doi: 10.1038/ncomms9566
- 499 Ma, Y., & Bassis, J. N. (2019). The effect of submarine melting on calving from
 500 marine terminating glaciers. *Journal of Geophysical Research: Earth Surface*,
 501 124(2), 334–346. doi: 10.1029/2018JF004820
- 502 Mankoff, K. D., Colgan, W., Solgaard, A., Karlsson, N. B., Ahlstrøm, A. P., van
 503 As, D., ... Fausto, R. S. (2019). Greenland Ice Sheet solid ice discharge
 504 from 1986 through 2017. *Earth System Science Data*, 11(2), 769–786. doi:
 505 10.5194/essd-11-769-2019
- 506 Mankoff, K. D., Noël, B., Fettweis, X., Ahlstrøm, A. P., Colgan, W., Kondo,
 507 K., ... Fausto, R. S. (2020). Greenland liquid water discharge from
 508 1958 through 2019. *Earth System Science Data*, 12(4), 2811–2841. doi:
 509 10.5194/essd-12-2811-2020
- 510 Meier, M. F., Lundstrom, S., Stone, D., Kamb, B., Engelhardt, H., Humphrey, N.,
 511 ... Walters, R. (1994). Mechanical and hydrologic basis for the rapid motion
 512 of a large tidewater glacier: 1. Observations. *Journal of Geophysical Research:*
 513 *Solid Earth*, 99(B8), 15219–15229. doi: 10.1029/94JB00237
- 514 Meier, M. F., & Post, A. (1987). Fast tidewater glaciers. *Journal of Geophysical Re-*
 515 *search: Solid Earth*, 92(B9), 9051–9058. doi: 10.1029/JB092iB09p09051
- 516 Miller, O., Solomon, D. K., Miège, C., Koenig, L., Forster, R., Schmerr, N., ... Mc-
 517 Connell, J. R. (2020). Hydrology of a perennial firn aquifer in Southeast
 518 Greenland: An overview driven by field data. *Water Resources Research*,
 519 56(8), e2019WR026348. doi: 10.1029/2019WR026348
- 520 Moon, T., Joughin, I., & Smith, B. (2015). Seasonal to multiyear variability of
 521 glacier surface velocity, terminus position, and sea ice/ice mélange in northwest
 522 Greenland. *Journal of Geophysical Research: Earth Surface*, 120(5), 818–833.
 523 doi: 10.1002/2015JF003494
- 524 Moon, T., Joughin, I., Smith, B., van den Broeke, M. R., van de Berg, W. J.,
 525 Noël, B., & Usher, M. (2014). Distinct patterns of seasonal Greenland
 526 glacier velocity. *Geophysical Research Letters*, 41(20), 7209–7216. doi:
 527 10.1002/2014GL061836
- 528 Morlighem, M., Williams, C. N., Rignot, E., An, L., Arndt, J. E., Bamber, J. L., ...
 529 Zinglensen, K. B. (2017). BedMachine v3: Complete bed topography and ocean
 530 bathymetry mapping of Greenland from multibeam echo sounding combined
 531 with mass conservation. *Geophysical Research Letters*, 44(21), 11,051–11,061.
 532 doi: 10.1002/2017GL074954
- 533 Morlighem, M., Wood, M., Seroussi, H., Choi, Y., & Rignot, E. (2019). Mod-
 534 eling the response of northwest Greenland to enhanced ocean thermal
 535 forcing and subglacial discharge. *The Cryosphere*, 13(2), 723–734. doi:
 536 10.5194/tc-13-723-2019
- 537 Motyka, R. J., Hunter, L., Echelmeyer, K. A., & Connor, C. (2003). Subma-
 538 rine melting at the terminus of a temperate tidewater glacier, LeConte
 539 Glacier, Alaska, U.S.A. *Annals of Glaciology*, 36, 57–65. doi: 10.3189/
 540 172756403781816374
- 541 Murray, T., Nettles, M., Selmes, N., Cathles, L. M., Burton, J. C., James, T. D.,
 542 ... Baugé, T. (2015, 07). Reverse glacier motion during iceberg calving
 543 and the cause of glacial earthquakes. *Science*, 349(6245), 305. doi:
 544 10.1126/science.aab0460
- 545 Murray, T., Scharrer, K., James, T. D., Dye, S. R., Hanna, E., Booth, A. D., ...
 546 Huybrechts, P. (2010). Ocean regulation hypothesis for glacier dynam-
 547 ics in southeast Greenland and implications for ice sheet mass changes.
 548 *Journal of Geophysical Research: Earth Surface*, 115(F3), F03026. doi:

- 549 10.1029/2009JF001522
- 550 Nettles, M., Larsen, T. B., Elósegui, P., Hamilton, G. S., Stearns, L. A., Ahlstrøm,
551 A. P., . . . Forsberg, R. (2008). Step-wise changes in glacier flow speed co-
552 incide with calving and glacial earthquakes at Helheim Glacier, Greenland.
553 *Geophysical Research Letters*, *35*(24). doi: 10.1029/2008GL036127
- 554 Nick, F. M., Vieli, A., Howat, I. M., & Joughin, I. (2009). Large-scale changes in
555 Greenland outlet glacier dynamics triggered at the terminus. *Nature Geo-*
556 *science*, *2*(2), 110–114. doi: 10.1038/ngeo394
- 557 Nienow, P. W., Sole, A. J., Slater, D. A., & Cowton, T. R. (2017). Recent
558 advances in our understanding of the role of meltwater in the Greenland
559 Ice Sheet system. *Current Climate Change Reports*, *3*(4), 330–344. doi:
560 10.1007/s40641-017-0083-9
- 561 Noël, B., van de Berg, W. J., van Wessem, J. M., van Meijgaard, E., van As,
562 D., Lenaerts, J. T. M., . . . van den Broeke, M. R. (2018). Modelling
563 the climate and surface mass balance of polar ice sheets using RACMO2
564 – Part 1: Greenland (1958–2016). *The Cryosphere*, *12*(3), 811–831. doi:
565 10.5194/tc-12-811-2018
- 566 Nowicki, S., Goelzer, H., Seroussi, H., Payne, A. J., Lipscomb, W. H., Abe-Ouchi,
567 A., . . . van de Wal, R. (2020). Experimental protocol for sea level projections
568 from ISMIP6 stand-alone ice sheet models. *The Cryosphere*, *14*(7), 2331–2368.
569 doi: 10.5194/tc-14-2331-2020
- 570 Nye, J. F. (1960). The response of glaciers and ice-sheets to seasonal and climatic
571 changes. *Proceedings of the Royal Society of London A: Mathematical, Physical*
572 *and Engineering Sciences*, *256*(1287), 559–584.
- 573 O’Leary, M., & Christoffersen, P. (2013). Calving on tidewater glaciers amplified by
574 submarine frontal melting. *The Cryosphere*, *7*(1), 119–128. doi: 10.5194/tc-7
575 -119-2013
- 576 O’Neel, S., Pfeffer, W. T., Krimmel, R., & Meier, M. (2005). Evolving force balance
577 at Columbia Glacier, Alaska, during its rapid retreat. *Journal of Geophysical*
578 *Research: Earth Surface*, *110*(F3), F03012. doi: 10.1029/2005JF000292
- 579 Parizek, B. R., & Alley, R. B. (2004). Implications of increased Greenland surface
580 melt under global-warming scenarios: ice-sheet simulations. *Quaternary Sci-*
581 *ence Reviews*, *23*(9), 1013–1027. doi: 10.1016/j.quascirev.2003.12.024
- 582 Phillips, T., Rajaram, H., & Steffen, K. (2010). Cryo-hydrologic warming: A poten-
583 tial mechanism for rapid thermal response of ice sheets. *Geophysical Research*
584 *Letters*, *37*(20). doi: 10.1029/2010GL044397
- 585 Pimentel, S., & Flowers, G. E. (2011). A numerical study of hydrologically driven
586 glacier dynamics and subglacial flooding. *Proceedings of the Royal Society A:*
587 *Mathematical, Physical and Engineering Sciences*, *467*(2126), 537–558. doi: 10
588 .1098/rspa.2010.0211
- 589 Podrasky, D., Truffer, M., Fahnestock, M., Amundson, J. M., Cassotto, R., &
590 Joughin, I. (2012). Outlet glacier response to forcing over hourly to interan-
591 nual timescales, Jakobshavn Isbræ, Greenland. *Journal of Glaciology*, *58*(212),
592 1212–1226. doi: 10.3189/2012JoG12J065
- 593 Poinar, K., Joughin, I., Lilien, D., Brucker, L., Kehrl, L., & Nowicki, S. (2017).
594 Drainage of Southeast Greenland firn aquifer water through crevasses to the
595 bed. *Frontiers in Earth Science*, *5*, 5. doi: 10.3389/feart.2017.00005
- 596 Reeh, N., & Olesen, O. B. (1986). Velocity measurements on Daugaard-Jensen
597 Gletscher, Scoresby Sund, East Greenland. *Annals of Glaciology*, *8*, 146–150.
598 doi: 10.3189/S0260305500001336
- 599 Riel, B., Minchew, B., & Joughin, I. (2021). Observing traveling waves in glaciers
600 with remote sensing: new flexible time series methods and application to
601 Sermeq Kujalleq (Jakobshavn Isbræ), Greenland. *The Cryosphere*, *15*(1),
602 407–429. doi: 10.5194/tc-15-407-2021
- 603 Rignot, E., Braaten, D., Gogineni, S. P., Krabill, W. B., & McConnell, J. R. (2004).

- 604 Rapid ice discharge from southeast Greenland glaciers. *Geophysical Research*
605 *Letters*, *31*(10). doi: 10.1029/2004GL019474
- 606 Rignot, E., Fenty, I., Menemenlis, D., & Xu, Y. (2012). Spreading of warm ocean
607 waters around Greenland as a possible cause for glacier acceleration. *Annals of*
608 *Glaciology*, *53*(60), 257-266. doi: 10.3189/2012AoG60A136
- 609 Rignot, E., & Kanagaratnam, P. (2006). Changes in the velocity structure
610 of the Greenland Ice Sheet. *Science*, *311*(5763), 986-990. doi: 10.1126/
611 science.1121381
- 612 Rignot, E., Velicogna, I., van den Broeke, M. R., Monaghan, A., & Lenaerts,
613 J. T. M. (2011). Acceleration of the contribution of the Greenland and Antarc-
614 tic ice sheets to sea level rise. *Geophysical Research Letters*, *38*(5), L05503.
615 doi: 10.1029/2011GL046583
- 616 Seabold, S., & Perktold, J. (2010). statsmodels: Econometric and statistical mod-
617 eling with python. In *9th python in science conference*. doi: 10.25080/Majora-
618 -92bf1922-011
- 619 Shepherd, A., Hubbard, A., Nienow, P., King, M., McMillan, M., & Joughin, I.
620 (2009). Greenland ice sheet motion coupled with daily melting in late summer.
621 *Geophysical Research Letters*, *36*(1). doi: 10.1029/2008GL035758
- 622 Shumway, R. H., & Stoffer, D. S. (2017). *Time series analysis and its applications*
623 (Fourth ed.). Cham, Switzerland: Springer.
- 624 Slater, D. A., Nienow, P. W., Cowton, T. R., Goldberg, D. N., & Sole, A. J.
625 (2015). Effect of near-terminus subglacial hydrology on tidewater glacier
626 submarine melt rates. *Geophysical Research Letters*, *42*(8), 2861–2868. doi:
627 10.1002/2014GL062494
- 628 Sole, A. J., Mair, D. W. F., Nienow, P. W., Bartholomew, I. D., King, M. A.,
629 Burke, M. J., & Joughin, I. (2011). Seasonal speedup of a Greenland marine-
630 terminating outlet glacier forced by surface melt-induced changes in subglacial
631 hydrology. *Journal of Geophysical Research: Earth Surface*, *116*(F3). doi:
632 10.1029/2010JF001948
- 633 Stearns, L. A., & Hamilton, G. S. (2007). Rapid volume loss from two East Green-
634 land outlet glaciers quantified using repeat stereo satellite imagery. *Geophysical*
635 *Research Letters*, *34*(5), L05503. doi: 10.1029/2006GL028982
- 636 Stevens, L. A., Behn, M. D., Das, S. B., Joughin, I., Noël, B. P. Y., van den Broeke,
637 M. R., & Herring, T. (2016). Greenland Ice Sheet flow response to runoff
638 variability. *Geophysical Research Letters*, *43*(21), 11,295–11,303. doi:
639 10.1002/2016GL070414
- 640 Todd, J., Christoffersen, P., Zwinger, T., Råback, P., Chauché, N., Benn, D., ...
641 Hubbard, A. (2018). A full-Stokes 3-D calving model applied to a large
642 Greenlandic glacier. *Journal of Geophysical Research: Earth Surface*, *123*(3),
643 410–432. doi: 10.1002/2017JF004349
- 644 van der Veen, C. J. (2001). Greenland ice sheet response to external forcing. *Journal*
645 *of Geophysical Research: Atmospheres*, *106*(D24), 34047–34058. doi: 10.1029/
646 2001JD900032
- 647 van Dongen, E. C. H., Åström, J. A., Jouvét, G., Todd, J., Benn, D. I., & Funk,
648 M. (2020). Numerical modeling shows increased fracturing due to melt-
649 undercutting prior to major calving at Bowdoin Glacier. *Frontiers in Earth*
650 *Science*, *8*, 253. doi: 10.3389/feart.2020.00253
- 651 Van Meijgaard, E., Van Ulft, L., Van de Berg, W., Bosveld, F., Van den Hurk, B.,
652 Lenderink, G., & Siebesma, A. (2008). *The KNMI regional atmospheric*
653 *climate model RACMO version 2.1*. Koninklijk Nederlands Meteorologisch
654 Instituut.
- 655 Vijay, S., Khan, S. A., Kusk, A., Solgaard, A. M., Moon, T., & Bjørk, A. A. (2019).
656 Resolving seasonal ice velocity of 45 Greenlandic glaciers with very high
657 temporal details. *Geophysical Research Letters*, *46*(3), 1485–1495. doi:
658 10.1029/2018GL081503

- 659 Virtanen, P., Gommers, R., Oliphant, T. E., Haberland, M., Reddy, T., Cournapeau,
660 D., . . . SciPy 1.0 Contributors (2020). SciPy 1.0: Fundamental Algorithms
661 for Scientific Computing in Python. *Nature Methods*, *17*, 261–272. doi:
662 10.1038/s41592-019-0686-2
- 663 Voytenko, D., Stern, A., Holland, D. M., Dixon, T. H., Christianson, K., & Walker,
664 R. T. (2015). Tidally driven ice speed variation at Helheim Glacier, Greenland,
665 observed with terrestrial radar interferometry. *Journal of Glaciology*, *61*(226),
666 301–308. doi: 10.3189/2015JoG14J173
- 667 Weertman, J. (1958). Traveling waves on glaciers. *IASH*, *47*, 162–168.
- 668 Weertman, J., & Birchfield, G. E. (1983). Basal water film, basal water pressure,
669 and velocity of traveling waves on glaciers. *Journal of Glaciology*, *29*(101), 20–
670 27.
- 671 Werder, M. A., Hewitt, I. J., Schoof, C. G., & Flowers, G. E. (2013). Modeling chan-
672 nelerized and distributed subglacial drainage in two dimensions. *Journal of Geo-*
673 *physical Research: Earth Surface*, *118*(4), 2140–2158. doi: 10.1002/jgrf.20146