

1 **Compounding seasonal variations in outlet glacier**
2 **dynamics revealed by high-resolution observations**

3 **Enze Zhang¹, Ginny Catania^{1,2}, Ben Smith³, Denis Felikson⁴, Beata Csatho⁵,**
4 **and Daniel T. Trugman⁶**

5 ¹The University of Texas at Austin, Institute of Geophysics, TX, USA

6 ²The University of Texas at Austin, Department of Geological Sciences, TX, USA

7 ³Polar Science Center, Applied Physics Laboratory, University of Washington, Seattle, WA, USA

8 ⁴Cryospheric Sciences Laboratory, NASA Goddard Space Flight Center, Greenbelt, MD, USA

9 ⁵Department of Geological Sciences, University at Buffalo, Buffalo, NY, USA

10 ⁶Nevada Seismological Laboratory, Nevada Geosciences, University of Nevada, Reno, NV, USA

11 **Key Points:**

- 12 • Simple model can be used to attribute a portion of seasonal velocity variability
13 caused by terminus position change.
- 14 • Seasonal velocity variability is complex and results from multiple compounding
15 processes.
- 16 • Seasonal velocities are more sensitive to surface slope changes than uniform changes
17 in elevation.

Corresponding author: Enze Zhang, zhangenze@link.cuhk.edu.hk

Abstract

Understanding seasonality in outlet glacier dynamics reveals insight into long-term retreat and acceleration. Leveraging recent high-resolution satellite data, we examine changes in surface elevation, velocity, and terminus position for five glaciers in Central Western Greenland over the past ~ 6 years. We employ an approach that examines the stress at the ice-ocean terminus and models the expected response in upstream velocity caused by the observed terminus changes. The model shows that some glaciers' seasonal velocity changes can be largely explained by terminus changes, while others can be compounded by multiple processes. Additionally, we test the sensitivity of the results by including seasonally varying and artificially modified surface topography. We find surface slope changes impact velocity response to terminus changes more than spatially uniform changes in along-flow elevation. Our approach provides a scalable framework to comprehend the compounded nature of glacier seasonal velocity variations across the Greenland Ice Sheet outlet glaciers.

Plain Language Summary

Understanding seasonal changes in glaciers is crucial for studying long-term trends. To capture glacier seasonality in detail, we combine detailed data on glacier speed and terminus movement at sub-weekly to daily intervals, along with seasonal surface topography data. We use a model that reveals how much glacier speed changes in response to terminus variations. For some glaciers, our model shows that seasonal glacier speed change is completely driven by terminus change. For other glaciers, we find that seasonal velocity changes can be influenced by runoff and seasonal changes in the drainage system beneath the glacier, in addition to terminus position change. Additional tests suggest that change in the surface slope of a glacier has a stronger impact on the sensitivity of seasonal speed changes to terminus changes than uniform changes in glacier surface topography; flattening of the glacier surface results in less sensitivity of the surface velocity to terminus changes. Our approach provides a framework that can be applied to the entire Greenland Ice Sheet to reveal the complexity of glacier seasonality.

1 Introduction

The Greenland Ice Sheet (GrIS) is currently the largest land ice contributor to present-day rising sea level (IPCC, 2022) with an acceleration in mass loss over the past few decades primarily attributed to ice discharge through outlet glaciers (Shepherd et al., 2012; Enderlin et al., 2014; van den Broeke et al., 2016). This acceleration underscores the importance of comprehending the intricate mechanisms that govern glacier dynamics. Despite the prevalence of glacier acceleration in Greenland, there exists notable spatio-temporal variability in glacier velocity change at a range of time scales (Moon et al., 2012, 2020), which is likely influenced by local conditions, like topography, and regionally by environmental factors. For example, runoff and ocean thermal forcing can influence velocity by changing basal friction (Ultee et al., 2022), subaqueous melt rates (Holland et al., 2008), and through terminus fluctuations (Howat et al., 2008; King et al., 2020; Wood et al., 2021). Numerical simulations also suggest a non-linear feedback between terminus changing rate and ice discharge (Sergienko, 2022). At the seasonal scale, many of these processes are synchronized, making it difficult to understand the cause and effect behind seasonal glacier acceleration. Despite this, recent advances in the temporal frequency of satellite measurements provide an opportunity to examine the factors that force glacier dynamic change over multiple epochs (Kehrl et al., 2017). Moreover, numerical simulations suggest that glacier seasonal changes can induce systematic bias in mass loss estimates at the multi-decadal time scale (Felixson et al., 2022). Thus, delving into the patterns of glacier seasonality is instrumental in unveiling and simulating the pivotal factors that control glacier dynamics at longer time scales and into the future.

67 Glaciers across Greenland exhibit discernible seasonal changes in terminus posi-
68 tion (Goliber et al., 2022; Zhang et al., 2023) and surface velocity (Moon et al., 2015;
69 Joughin et al., 2008), but the mechanisms behind such changes are varied. Studies have
70 suggested that seasonal glacier retreat due to summertime air temperature increases causes
71 reduced contact with bed and/or fjord walls, which along with increased net force at the
72 calving cliff, leading to glacier acceleration (Howat et al., 2005; Joughin et al., 2012). Oth-
73 ers have indicated that glacier acceleration can be due to seasonal changes in basal lu-
74 brication related to changes in the subglacial hydrological system (Davison et al., 2020;
75 Stevens et al., 2022; Werder et al., 2013; Andrews et al., 2014), which can cause com-
76 plex responses from glacier velocities as subglacial conduits grow more efficient (Bartholomew
77 et al., 2010; Andrews et al., 2014; Vijay et al., 2019). Subglacial hydrology can also be
78 influenced by remnant meltwater (Iken & Truffer, 1997) that is stored englacially (Abe
79 & Furuya, 2015) or in basal crevasses (Harper et al., 2010) and through exfiltrated ground-
80 water (Robel et al., 2023), which can leak out of the subglacial system over time, includ-
81 ing in winter (Rennermalm et al., 2013).

82 Previous efforts have classified glacier seasonal velocity variations into types based
83 on the observed timing of velocity change and their correlation to runoff and terminus
84 change (Moon et al., 2014; Vijay et al., 2019, 2021). Three categories have been gener-
85 ally described in this literature; 1) positive correlation of glacier velocity to glacier ter-
86 minus retreat; 2) positive correlation of glacier velocity to summer runoff, and; 3) glacier
87 velocity that slows in late summer and speeds up in winter. Recently, Solgaard et al. (2022)
88 applied a machine learning approach to analyze velocity time series data across Green-
89 land, revealing similar seasonal patterns as those first identified by Moon et al. (2014).
90 However, most of these studies used data with limited temporal sampling resulting in
91 classifications that are based on identifying a single process that influences velocity. Poinar
92 (2023) applied principal component analysis to decompose the seasonality of glaciers in
93 Sermilik Fjord from an observational perspective. They emphasized the importance of
94 extracting velocity patterns across the entire glacier, not just a single point. Furthermore,
95 they classified four glaciers in Sermilik Fjord by quantifying the prevalence of multiple
96 glacier types at a single glacier. Here, we re-examine glacier seasonality using high-frequency
97 terminus (Zhang et al., 2023), velocity (Gardner et al., 2023), and surface elevation change
98 observations. We interpret these observations with an analytical model of velocity re-
99 sponse to terminus position change (Joughin et al., 2012). By comparing model results
100 with observed velocity time series we find that glaciers are more typically influenced by
101 an interplay of multiple processes rather than a single process, and that glacier veloc-
102 ity behavior can transition between different modes seasonally.

103 2 Study regions

104 We investigate five glaciers in central-west Greenland (Figure 1): Rink Isbrae (RNK),
105 Sermeq Avannarleq (AVA), Sermeq Kujalleq (KUJ), Kangilernata Sermia (KAN), and
106 Eqip Sermia (EQP) over the time period 2015-2021. This time span is specifically cho-
107 sen to take advantage of the increased sample frequency available in both velocity and
108 terminus position data due to the launch of Sentinel-1/2 in 2014. These five glaciers are
109 selected because 1) they exhibit regular seasonal changes in both terminus position and
110 velocity (Catania et al., 2018; Fried et al., 2018) with minimal long-term variations over
111 our study period and; 2) they exhibit a range of sub-seasonal behavior in both the ter-
112 minus and velocity variability. For example, all glaciers advance in winter and retreat
113 in summer and yet their seasonal velocity behavior differs over time and space. EQP and
114 KUJ speed up during summertime terminus retreat while AVA and KAN slow down dur-
115 ing summertime terminus retreat (Fried et al., 2018). AVA, KUJ, KAN, and EQP are
116 located close to one another, suggesting that they likely experience the same regional cli-
117 mate forcing. We also examine RNK, which is further north than these four glaciers be-
118 cause it has a deep grounding line, in contrast to the shallower grounding lines of the

119 other four glaciers to the south, and a partially floating terminus that permits large, buoy-
120 ant flexure-style calving events driving glacier-wide step changes in the terminus posi-
121 tion (Medrzycka et al., 2016; Fried et al., 2018). Medrzycka et al. (2016) investigated the
122 calving styles of RNK by using time-Lapse photos and found that the northern part of
123 the RNK terminus undergoes small calving events, while the southern part experiences
124 larger events driven in response to buoyant flexure. The authors suggested that these
125 buoyancy conditions exert primary control on terminus behavior, implying that the north-
126 ern part of RNK is lightly grounded, while the southern part is floating.

127 **3 Data**

128 We use dense velocity time series data generated using auto-RIFT (Gardner et al.,
129 2018) and provided by the NASA MEaSURES ITS_LIVE project (Gardner et al., 2023).
130 ITS_LIVE combines velocity products derived from Landsat-8, Sentinel-1, and Sentinel-
131 2 producing a near-daily temporal resolution since 2014. For each glacier, we use mul-
132 tiple flowlines across the glacier from Felikson et al. (2021) to extract velocities at mul-
133 tiple points along each flowline (red and blue points in Figure 1). We then average these
134 velocities across all flowlines at each cross-section to produce mean (across-flow) veloc-
135 ity time series from downstream to upstream for each glacier. The flowlines of AVA pre-
136 dominantly converge on the western side of the basin because AVA is formed by the con-
137 fluence of two upstream tributaries, and we focus on the main tributaries with higher
138 velocities on the western side. RNK has eight flowlines because half of the terminus re-
139 gion of this glacier is floating (Medrzycka et al., 2016) and we want to examine the ve-
140 locity variations of the floating and grounded ice separately. We identify floating ice based
141 on the flattening of the surface elevation along flowlines towards the terminus and we
142 take the mean of the velocities on floating ice and grounded ice separately (Figure S1).
143 Note that this separation serves as a theoretical experiment designed to assess the po-
144 tential impact of surface slope on simulated velocity. We refrain from employing bed el-
145 evation data to ascertain the floating condition for RNK because of the reliance of the
146 bed data on mass conservation and the assumption that the glacier is grounded (Morlighem
147 et al., 2017). Terminus position data come from AutoTerm (Zhang et al., 2023), a ma-
148 chine learning pipeline that automatically produces terminus traces with an average sam-
149 pling frequency of 10 per month since 2014. We derive a time series of terminus changes
150 by calculating the sequential area changes between termini, accumulating these over time,
151 and then normalizing this by a static glacier width of 4.6 km for RNK, 6 km for AVA,
152 5 km for KUJ, 4 km for KAN, and 3 km for EQP.

153 We generate surface elevation data through a novel fusion of ICESat-2 data with
154 DigitalGlobe high-resolution digital elevation models (DEM), termed “DG-IS2-DEM”
155 producing four DEMs per year since Fall 2018. The algorithms that generate the DG-
156 IS2-DEMs are described in the Supplementary Information. We also use ArcticDEM (Porter
157 et al., 2022) as supplementary elevation data in locations where the DG-IS2-DEMs do
158 not extend to the most advanced terminus position found in AutoTerm. To determine
159 ice thickness, we subtract surface elevation data from bed elevation data from BedMa-
160 chineV5 (Morlighem et al., 2022), which assimilates seafloor bathymetry and ice thick-
161 ness data through a mass conservation approach (Morlighem et al., 2017). We extract
162 the surface and bed elevation profiles along each flowline individually. We use GSFC-
163 FDMv1.2.1 simulations of the surface mass balance (Medley et al., 2022) to produce a
164 runoff time series with a five-day sampling frequency. We use runoff as a proxy for the
165 start and end of the melt season.

166 **4 Terminus-Driven model**

167 Force balance methods can be used to understand the dynamic evolution of glaciers
168 through examination of the balance of stresses on them (Veen et al., 2011; Carnahan et
169 al., 2022). However, the force balance method requires double derivative of surface ve-

170 locity data, which can result in large uncertainties when using satellite products that have
 171 lower accuracy including ITS.LIVE. Therefore, we adopt a different approach that em-
 172 ploys a modified force balance termed the “terminus-driven model” described by Joughin
 173 et al. (2012). This model explicitly considers the influence of the dynamic changes at the
 174 glacier terminus on upstream velocity. By using the terminus-driven model and near daily
 175 velocity data, we are able to isolate the contribution of sub-annual terminus variations
 176 to the observed variations in the velocity time series. The terminus-driven model is a 1-
 177 D model along the flow direction, assuming both an ice mélange free condition and a con-
 178 sistent glacier geometry over time. Additionally, since we use a consistent bed elevation
 179 product in our model, we assume that all the glaciers are grounded and that the eleva-
 180 tion does not change over time. The terminus-driven model focuses on the driving stress
 181 (τ_d) expressed as

$$\tau_d = -\rho_i \times gH \times \frac{\partial h}{\partial x} \quad (1)$$

182 where g is the gravitational acceleration, H is ice thickness, h is ice surface elevation,
 183 and ρ_i is the density of ice (910 km/m³). In addition, the terminus driven model exam-
 184 ines an additional force due to the presence of the free calving face. This latter force is
 185 determined by the height above the fjord surface at the calving front and the density of
 186 seawater. The difference between these two forces at the terminus is expressed as

$$F = \frac{1}{2} \times \rho_i gH^2 - \frac{1}{2} \times \rho_w g(H - h)^2 \quad (2)$$

187 where ρ_w is the density of seawater (1028 km/m³). Here, we call F the “frontal force”
 188 following the naming convention found in Joughin et al. (2012), although it has units of
 189 N/m . The force balance at the terminus requires the frontal force to be balanced up-
 190 stream by the longitudinal stress, which redistributes much of the frontal force to the
 191 margins and bed of the glacier upstream. We term the longitudinal stress that originates
 192 from the frontal force as $\tau_F(X)$, which pulls the glacier and enhances the original driv-
 193 ing stress (e.g., $\tau_d + \tau_F(X)$).

194 The integration of $\tau_F(X)$ along the flowline equals F and is assumed to linearly
 195 decrease upstream of the terminus to zero at the stress coupling length following Joughin
 196 et al. (2012):

$$\tau_F(X) = 2 \times \frac{F}{\lambda} \times \left(1 - \frac{X}{\lambda}\right) \quad (3)$$

197 where λ is the stress coupling length, and X is the distance between terminus and the
 198 point where we simulate velocity. The terminus variations cause changes in the geom-
 199 etry of the free calving, consequently influencing the frontal force (F). These changes
 200 in the frontal force, subsequently, lead to modifications in the enhanced driving stress
 201 ($\tau_d + \tau_F(X)$) in the upstream region.

202 In the lamellar flow model or the shallow ice approximation, when basal sliding is
 203 zero, the surface velocity has a linear relationship with the cube of driving stress ($n=3$)
 204 (Van der Veen, 2013):

$$V = \frac{1}{2} AH \tau_d^3 \quad (4)$$

205 where A is a constant from Glen’s flow law. Such a linear relationship also applies when
 206 the driving stress is mainly balanced by lateral drag (Van der Veen, 2013):

$$V = \frac{1}{2} A \left(\frac{\tau_d}{H}\right)^3 W^4 \quad (5)$$

207 where W is half of the glacier width. Based on the above two models and following the
 208 method designed by Joughin et al. (2012), we assume a linear relationship between ve-
 209 locity and the cube of the enhanced driving stress: $\tau_d + \tau_F(X)$, and the predicted ve-
 210 locity from terminus changes is thus given by:

$$\frac{V(X, t)}{V_0} = \left(\frac{\tau_d + \tau_F(X)}{\tau_d + \tau_{F_0}}\right)^3 \quad (6)$$

211 where V_0 is a reference velocity at the same location as velocity observations and τ_{F_0} is
 212 the $\tau_F(X)$ corresponding to the reference velocity. For each year, we use the minimum
 213 velocity observation as the reference velocity (following Joughin et al. (2012) and cal-
 214 culate τ_{F_0} based on the terminus position nearest the date of the reference velocity. Us-
 215 ing Eqn. 6, we can simulate a velocity time series at each observation point for each glacier
 216 (Figure 1).

217 We vary stress coupling lengths for each glacier and choose the one that produces
 218 the lowest mean difference between observations and simulated velocity (Table S1). The
 219 mean difference is determined by:

$$\frac{\text{abs}(\text{model} - \text{observation})}{\text{model}} \times 100\% \quad (7)$$

220 For each flowline, we extract the geometry profile and compute a simulated velocity and
 221 then average the simulated velocity across all flowlines in a manner consistent with ob-
 222 served velocity. Subsequently, we compare these averaged simulated velocities to observed
 223 velocities.

224 Although the terminus-driven model was initially designed by Joughin et al. (2012)
 225 to assume invariant geometry in its operation, we analytically examine the impact of sea-
 226 sonal variations in surface elevation on velocity simulations. Specifically, we leverage the
 227 new time-varying DG-IS2-DEM and periodically update the elevation profiles each quar-
 228 ter from Fall 2018, maintaining profile consistency within each quarter. We produce a
 229 simulated velocity for all glaciers with and without time-varying surface elevation in or-
 230 der to evaluate the impact of seasonally-varying surface elevation change on velocity. For
 231 the fixed geometry simulations, we choose a time step from DG-IS2-DEM with an ex-
 232 tent that aligns best with the position of the terminus when it is most advanced. This
 233 provides the most complete elevation profile across the terminus region. For EQP, KAN,
 234 and AVA we choose the October 2019 DG-IS2-DEM and for KUJ, we use the April 2019
 235 DG-IS2-DEM time step. For RNK, we use additional elevation data from ArcticDEM
 236 (Porter et al., 2022) for the fixed geometry case, as the DG-IS2-DEM does not cover the
 237 most advanced terminus position for this glacier.

238 5 Results

239 5.1 Comparison between velocity simulation and observations

240 We compare the simulated velocity time series with velocities from satellite obser-
 241 vations to determine whether seasonal velocity variations are influenced primarily by ter-
 242 minus change, co-influenced by other factors, or entirely independent of terminus change.
 243 Overall, we find that the time-series velocity observations from 2015 are well-described
 244 by the terminus-driven model for the grounded portions of RNK, KUJ, and EQP but
 245 not for KAN and AVA (Figures 2-6). For RNK, KUJ, and EQP, seasonal changes in glacier
 246 speed align well with terminus variations. This is supported by the coincident timing of
 247 the end of terminus retreat and the peak summertime velocity (vertical black lines in Fig-
 248 ure 2, 4, and 6), even in instances when retreat continues beyond the end of the melt sea-
 249 son (Figure 6). For these glaciers, the mean misfit between simulated and observed ve-
 250 locities over all years are 4.6% for KUJ, 6.2% for EQP, and 6.8% for RNK (Table S1),
 251 with correlations of 0.84, 0.67, and 0.56, respectively (Figure 7).

252 For AVA and KAN, we find that simulated velocities differ substantially from the
 253 observed velocities. For AVA, simulated velocity fluctuations are relatively small in mag-
 254 nitude compared to observations (Figure 3), while at KAN, the simulated and observed
 255 velocities are out of phase but of the same magnitude (Figure 5).

256

5.2 Compounded velocity processes

257

258

259

260

261

262

263

264

265

266

267

268

269

Although the terminus-driven model adequately resolves seasonal variability in velocity for RNK, KUJ, and EQP, there are observed sub-seasonal velocity changes that are not explained by the terminus-driven model alone. For all glaciers but KUJ, we observe additional pulses in velocity (acceleration and deceleration) in the middle of the melt season, a phenomenon not captured by the terminus-driven model (black dashed boxes in Figure 2,3,5, and 6). Especially for EQP, the acceleration induced by terminus changes, along with melt-season pulses, collectively form a bimodal velocity response. While these are predominant, they do not as obviously across all years for all glaciers. For example, melt-season pulses are strongly visible for every year in the record for AVA (Figure 2) but they are only obviously visible from 2015-2019 for KAN (Figure 5). For 2020 and 2021 the change in velocity in the melt season is less prominent. Similar for EQP for 2020 and 2021 the melt season pulses are less obvious (Figure 6). Melt season pulses for RNK are even more sporadic (Figure 2).

270

271

272

273

274

275

276

277

278

In addition to melt-season velocity pulses, we find additional sub-seasonal pulses on RNK that coincide with large calving events. RNK experiences much larger calving events than the other glaciers and these create large (~ 1 km) step changes in the terminus position. Calving-related pulses in velocity are only predicted to impact velocity noticeably for the grounded portion of RNK (Figure 2d). While we observe sub-seasonal velocity pulses that are coincident with some of these predicted events (blue dashed box in Figures 2), they have a magnitude that is muted compared to those predicted by the terminus-driven model. Further, there are many more predicted velocity pulses from large calving events than are visible in the observed velocity.

279

280

281

282

283

284

285

286

287

288

289

290

291

While both KAN and AVA experience summertime terminus retreat and wintertime terminus advance similar to the other three glaciers, their velocity response is poorly predicted by the terminus-driven model. For these two glaciers, we observe accelerations during winter (during terminus advance) that plateau before the onset of the melt season in the following year, and early melt season accelerations with the annual maximum velocity reached in the middle of the melt season (black dashed boxed in Figures 3 and 5). The terminus-driven model does not capture wintertime acceleration because across all glaciers the terminus is advancing in winter. For KAN, the model predicts slight deceleration in winter (Figure 5). For AVA, there is no significant seasonality in the simulated velocity likely because the scale of seasonal terminus advance and retreat for this glacier is small (Figure 3) and the surface elevation is flat in frontal region (Figure 8). For reference the averaged seasonal range in terminus position is 144 meters for AVA, while EQP is 224 meters, KAN is 390 meters, and KUJ is 417 meters.

292

5.3 Experiments with seasonally varying surface elevation

293

294

295

296

297

298

299

300

301

302

303

304

305

306

We investigate the influence of changing surface topography by comparing the velocity simulated using a fixed geometry against velocity simulated using a seasonally varying surface elevation from 2018-2022. We find minimal differences between these results for all glaciers (black versus red lines in Figure 3–6). To investigate this further, we consider only KUJ as an example and probe the terminus-driven model via two experiments; 1) we artificially shift the entire elevation profile vertically by ± 10 -20 meters and; 2) we alter the surface slope by $\pm 2\%$ within the 2 km-frontal region. The results suggest that terminus-driven velocities are relatively insensitive to spatially uniform along-flow changes in surface elevation, but are highly sensitive to changes in surface slope (Figure 9). This result is important for providing context for interpreting the results for RNK, which has a flat, floating portion of the terminus. We find that while the seasonal variations in velocity are similar on the floating and grounded portions of RNK, the simulated velocities in the floating portion are much lower magnitude and lack strong seasonality (Figure 2e).

6 Discussion

6.1 Potential explanations of the compounding seasonality

Using high-temporal-resolution observations and a terminus-driven model to simulate velocity variations from terminus change, we investigate sub-seasonal velocity changes for GrIS outlet glaciers and find that glacier velocity responds to multiple compounding processes. The seasonal velocity changes of three glaciers (KUJ, EQP, and the grounded portion of RNK) can largely be attributed to seasonal terminus variation, particularly for KUJ, which has a velocity that is almost entirely driven by the terminus fluctuations. However, four out of our five study glaciers experience additional processes that drive changes in velocity. EQP, RNK, AVA, and KAN all experience occasional sub-seasonal peaks in velocity that are coincident with the middle of the melt season, AVA and KAN exhibit wintertime speedup that occurs when their termini are advancing, and RNK experiences short-time pulses in velocity throughout the record.

We hypothesize that the peaks in the middle of the melt season observed for EQP, RNK, AVA, and KAN (black dashed squares in Figure 2,3,5, and 6) result from runoff-driven acceleration and subsequent evolution of the subglacial drainage system (Moon et al., 2014; Vijay et al., 2019). Early in the melt season, the subglacial drainage system is inefficient (Andrews et al., 2014), thus as meltwater availability begins to increase (marked by increasing runoff in early summer), subglacial water pressures increase enhancing basal sliding by reducing friction between the ice and the bed (Bartholomew et al., 2010; Bartholomew et al., 2008). As the melt season progresses, the drainage system channelizes becoming more efficient (Andrews et al., 2014; Schoof, 2010) and available meltwater decreases, producing a reduction in glacier speed. Beyond the melt season, the impact of terminus retreat on seasonal velocities can become more pronounced. For example, EQP typically has a melt season that ends in October, but the terminus continues to retreat until December/January (Figure 6). This produces a wintertime peak in velocity that is coincident with the most retreated terminus of EQP and is distinct from the melt-season peak.

To further explore the velocity increases that occur in summer we examine the along-flow variability in velocity to determine how far upstream velocity changes occur (Figure 10). We extract along-flow velocity profiles from monthly velocity mosaics provided by the the Greenland Ice sheet Mapping Project (Joughin, 2023). We use this dataset only for extracting velocity profiles, specifically to achieve better spatial consistency. To confirm that runoff drives summertime speed up for EQP, we compare the along-flow spatial pattern of velocity change that occurs in the summer melt season of 2017 (Apr 2017 - Sep 2017; Figure 10a) and the subsequent time period after runoff has ceased, when the velocity is primarily influenced by terminus changes (Oct 2017 - Mar 2018; Figure 10b). We quantify the range of upstream velocity at a distance of 10 km upstream of the terminus, which is 25 times the terminus thickness to ensure we are several longitudinal coupling lengths upstream of the terminus. We find that runoff-driven acceleration is noticeable in the velocity further upstream than during the period of terminus-driven velocity change. For example, when the runoff is large, EQP experiences a range in upstream velocity that is 64% of the velocity range observed at the terminus (Figure 10a). Conversely, in the winter when runoff is absent and the terminus alone is changing, EQP experiences a range in upstream velocity inland that is just 12% of what is observed at the terminus (Figure 10b). The rapid decline in speed with distance from the terminus is expected when a glacier is terminus-driven because of the reduction in the terminus force with distance from the terminus (Joughin et al., 2012). Conversely, elevated inland velocities are typical for melt-driven acceleration (Sundal et al., 2011) because meltwater percolates throughout the entire ablation zone (Andrews et al., 2014), which extends about 700 km inland of the terminus for EQP (Noël et al., 2019).

358 While AVA and KAN experience mid-summer velocity pulses similar to EQP, they
 359 do not exhibit any terminus-driven seasonal acceleration (Figures 3 and 5) and instead
 360 accelerate in winter. This suggests a decoupling between velocity changes and terminus
 361 change for these glaciers. We examine wintertime acceleration similar to above by de-
 362 termining the along-flow pattern of velocity change (Figure 10c, and d). For both glaciers
 363 we find significant inland acceleration. For KAN, the range in upstream velocity dur-
 364 ing winter is 72% of the range in frontal velocity and for AVA, the range in upstream
 365 velocity is 39% of the range of frontal velocity, which is larger than terminus-driven up-
 366 stream velocity range that was observed for EQP (Figure 10c and d versus b). We hy-
 367 pothesize that the elevated range of KAN and AVA’s upstream velocities in winter sug-
 368 gests that winter acceleration is due to enhanced extensive basal slip, which can be caused
 369 by several different processes. During the onset of winter, refreezing of percolating melt-
 370 water (Boon & Sharp, 2003) and viscous deformation over subglacial conduits (Vieli et
 371 al., 2004; Bartholomaeus et al., 2011) can obstruct the drainage system. Consequently,
 372 water becomes trapped within an inefficient drainage network, leading to increased wa-
 373 ter pressure and winter acceleration (Vijay et al., 2019). There are three possible sources
 374 of water at the ice-bed interface during winter: 1) remnants of summer meltwater (Iken
 375 & Truffer, 1997), 2) englacial water stored by basal crevasses that do not reach the sur-
 376 face (Abe & Furuya, 2015; Harper et al., 2010), and/or 3) sustained exfiltration of un-
 377 derground water caused by rapid unloading in melt season (Robel et al., 2023). The win-
 378 ter acceleration phase ends when the melt season begins to supply additional water to
 379 the subglacial system, which further increases basal water pressures causing summertime
 380 pulses in speed forcing the glacier to reach maximum speeds in summer.

381 RNK is a glacier that experiences three distinct modes of velocity variations includ-
 382 ing 1) seasonal terminus-driven velocity change; 2) occasional runoff-driven velocity change
 383 and; 3) frequent, small-magnitude velocity change that appears to be linked to large calv-
 384 ing events. Large calving events have been documented to cause step-like acceleration
 385 for Helheim Glacier (Nettles et al., 2008; de Juan et al., 2010) and, like Helheim, RNK
 386 experiences calving via buoyant flexure causing glacier-wide step-retreat of the termi-
 387 nus position (Fried et al., 2018; Medrzycka et al., 2016). Calving-related velocity pulses
 388 at RNK are significantly muted compared to those predicted by the terminus-driven model.
 389 In part, this may be due to the lower sampling frequency of terminus change. Prior to
 390 2017, our terminus record of RNK contains just six termini per month while after 2017,
 391 there are up to fourteen termini per month. Indeed, we observe more correlation between
 392 the magnitude of velocity pulses related to calving events between the observed and sim-
 393 ulated velocities after 2017. In addition to reduced sampling frequency, the floating por-
 394 tion of RNK with its flat surface topography produces a much weaker simulated veloc-
 395 ity pulse in response to calving events than is seen on grounded ice, where surface to-
 396 pography is much steeper. Thus, the floating ice with its flat surface topography damp-
 397 ens the impact of calving on surface speed.

398 The floating portion of the RNK terminus produces simulated velocities that have
 399 a much lower magnitude than observed (Figure 2), indicating that velocity change on
 400 this part of RNK is not driven by terminus change through longitudinal stress coupling.
 401 Terminus change may still have a strong, but more indirect impact on the velocity change
 402 on the floating portion via lateral stresses originating on the adjacent grounded ice. In
 403 contrast, the observed velocities in the floating portion of RNK show just as much vari-
 404 ation over time as we observe in the grounded portion. We attribute this discrepancy
 405 to the fact that the terminus-driven model captures longitudinal stresses but not lateral
 406 stresses. Thus, the seasonal variations in the observed velocity over the floating region
 407 might be driven by the nearby velocities on the grounded ice through lateral stress, which
 408 is not captured by the terminus-driven model. The impact of flotation may also be ob-
 409 served on the grounded ice. For example, in 2017 and 2018 RNK underwent a large multi-
 410 year advance (~ 1000 meters), after which the seasonal variations in both simulated and
 411 observed velocity in the grounded ice are reduced in amplitude compared to other years.

412 We speculate that as the glacier advanced, its original grounded front became floating
 413 and the surface flattened, which caused the velocity to be less sensitive to seasonal ter-
 414 minus variation.

415 **6.2 Impact of observed seasonal elevation changes**

416 The availability of seasonally-resolved elevation change allows us to investigate the
 417 degree to which velocity is sensitive to changing surface elevation. We find that seasonal
 418 elevation change for EQP, KAN, KUJ, and AVA is relatively uniform along flow (Fig-
 419 ure 8), and as a result, they do not significantly alter terminus-driven velocity (black lines
 420 in Figures 3–6). This aligns with our experimental results that suggest that vertical shifts
 421 in elevation have a limited contribution to velocity seasonality (Figure 9a). The exper-
 422 imental results also suggest steepening surface elevation will cause stronger velocity re-
 423 sponses (Figure 9b), which agrees with our results for RNK that simulated velocity is
 424 comparable with observations along the steep grounded flowlines but nearly absent on
 425 flat floating flowlines (Figure 2).

426 **6.3 Analysis of model sensitivity on velocity positions and derived stress** 427 **coupling lengths**

428 We further test the sensitivity of the position of the velocity points to the results
 429 by including more data points in both upstream and downstream regions (blue points
 430 in Figure 1). For RNK, KUJ, and EQP, differences exist between upstream and down-
 431 stream measurements (Figures S2, S3, S5, S7). The spatial changes for AVA and KAN
 432 in both velocity observations and simulations are subtle (Figures S4 and S6). However,
 433 the temporal pattern of velocity remains relatively consistent across the entire water-
 434 shed, with only the magnitude of velocity decreasing from downstream to upstream. In
 435 the ground portion of RNK, both observed and simulated velocity exhibit reduced sea-
 436 sonality from downstream to upstream, while also displaying a clear calving-induced ac-
 437 celeration. While most regions of the floating portion exhibit no discernible seasonality
 438 in the simulations, some areas demonstrate moderate seasonality (Figures S3b and S3c).
 439 The moderate seasonality of these areas can be attributed to their relatively small driv-
 440 ing stress; when the driving stress is reduced, the impact of $\tau_F(X)$ on velocity becomes
 441 more pronounced.

442 Our simulated velocities match well with the observed ones for KUJ and EQP over-
 443 all. However, the simulated velocity for KUJ overestimates the seasonality in 2020 in up-
 444 stream regions (Figures S5e and f) and underestimates the seasonality in 2021 in down-
 445 stream regions (Figure S5). This suggests that environmental factors, aside from termi-
 446 nus change, may influence the seasonality of velocity for KUJ, and that this influence
 447 varies spatiotemporally. In 2020, seasonal elevation change in EQP diminish the simu-
 448 lated velocity’s seasonality more prominently in the upstream region, better aligning with
 449 observed velocity than in the downstream region (Figures S7d and e). However, the terminus-
 450 driven model with varying geometry overestimates the seasonal velocity in the downstream
 451 region. We hypothesize that seasonal elevation change acts to dampen the seasonal ve-
 452 locity signal in the upstream region, potentially propagating downstream. However, since
 453 the terminus-driven model only considers the geometry at a given location, it thus fails
 454 to capture such propagation.

455 The stress coupling length is 25km for RNK, 10km for EQP, and 20km for KUJ
 456 (Table S1). KAN and AVA’s seasonal velocities are not driven by terminus variation, so
 457 we do not include the estimation of stress coupling length for these glaciers. Enderlin
 458 et al. (2016) estimated the stress coupling length by using an empirical method, suggest-
 459 ing that the stress coupling length should be approximately four times the glacier thick-
 460 ness, which is smaller than our estimation. Our estimation of SCL might be uncertain

461 since we assume a constant stress coupling length along the profile and the terminus-driven
 462 model is a simplified 1-D model.

463 7 Conclusion

464 We apply a terminus-driven model to elucidate the seasonal and sub-seasonal ve-
 465 locity changes for five glaciers in Central West Greenland. The comparison between sim-
 466 ulated and observed velocity suggests that glacier velocity change is driven by the in-
 467 terplay of multiple processes: terminus variations, runoff changes, evolution of the sub-
 468 glacial drainage system, and calving. Most glaciers exhibit more than one of these ve-
 469 locity variations and thus are a compounded signal. Notably, the observed seasonal el-
 470 evation changes appear to have limited influence on simulated velocities largely because
 471 the seasonal elevation signal is dominated by shifts in elevation and not changes in sur-
 472 face slope. Our experiments indicate that changes in surface slope have a stronger im-
 473 pact on the response of velocity to terminus changes than uniform changes in elevation.
 474 Our study provides a framework that can be applied to all outlet glaciers around the Green-
 475 land Ice Sheet to reveal the compounded nature of each glacier’s seasonal velocity change.
 476 Moreover, the same framework could be applied to investigate the long-term changes in
 477 glacier dynamics with adequate historical data. By systematically discerning the com-
 478 monalities and disparities among glaciers with distinct glaciological settings, our approach
 479 has the potential to shed light on diverging controls on outlet glaciers.

480 Acknowledgments

481 We acknowledge funding for this work from NASA (Grant NNH20ZDA001N-ICESAT2)
 482 and the Institute for Geophysics Postdoctoral Fellowship at the Jackson School to E. Zhang.
 483 We acknowledge the National Snow and Ice Data Center QGreenland package (Moon
 484 et al., 2023). We acknowledge DEMs provided by the Polar Geospatial Center under NSF-
 485 OPP awards 1043681, 1559691, 1542736, 1810976, and 2129685.

486 References

- 487 Abe, T., & Furuya, M. (2015). Winter speed-up of quiescent surge-type glaciers
 488 in yukon, canada. *The Cryosphere*, *9*(3), 1183–1190. doi: 10.5194/tc-9-1183
 489 -2015
- 490 Andrews, L. C., Catania, G. A., Hoffman, M. J., Gulley, J. D., Lüthi, M. P., Ryser,
 491 C., . . . Neumann, T. A. (2014). Direct observations of evolving subglacial
 492 drainage beneath the greenland ice sheet. *Nature*, *514*(7520), 80–83. doi:
 493 10.1038/nature13796
- 494 Bartholomäus, T. C., Anderson, R. S., & Anderson, S. P. (2008). Response of glacier
 495 basal motion to transient water storage. *Nature Geoscience*, *1*(1), 33–37. doi:
 496 10.1038/ngeo.2007.52
- 497 Bartholomäus, T. C., Anderson, R. S., & Anderson, S. P. (2011). Growth and
 498 collapse of the distributed subglacial hydrologic system of kennicott glacier,
 499 alaska, usa, and its effects on basal motion. *Journal of Glaciology*, *57*(206),
 500 985–1002. doi: 10.3189/002214311798843269
- 501 Bartholomew, I., Nienow, P., Mair, D., Hubbard, A., King, M. A., & Sole, A. (2010).
 502 Seasonal evolution of subglacial drainage and acceleration in a greenland outlet
 503 glacier. *Nature Geoscience*, *3*(6), 408–411. doi: 10.1038/ngeo863
- 504 Boon, S., & Sharp, M. (2003). The role of hydrologically-driven ice fracture in
 505 drainage system evolution on an arctic glacier. *Geophysical Research Letters*,
 506 *30*(18). doi: <https://doi.org/10.1029/2003GL018034>
- 507 Carnahan, E., Catania, G., & Bartholomäus, T. C. (2022). Observed mechanism for
 508 sustained glacier retreat and acceleration in response to ocean warming around
 509 Greenland. *The Cryosphere*, *16*(10), 4305–4317. doi: 10.5194/tc-16-4305-2022

- 510 Catania, G. A., Stearns, L. A., Sutherland, D. A., Fried, M. J., Bartholomaus, T. C.,
 511 Morlighem, M., . . . Nash, J. (2018). Geometric controls on tidewater glacier
 512 retreat in central western Greenland. *Journal of Geophysical Research: Earth*
 513 *Surface*, *123*(8), 2024–2038. doi: 10.1029/2017JF004499
- 514 Davison, B. J., Sole, A. J., Cowton, T. R., Lea, J. M., Slater, D. A., Fahrner, D.,
 515 & Nienow, P. W. (2020). Subglacial drainage evolution modulates seasonal
 516 ice flow variability of three tidewater glaciers in southwest greenland. *Jour-*
 517 *nal of Geophysical Research: Earth Surface*, *125*(9), e2019JF005492. doi:
 518 <https://doi.org/10.1029/2019JF005492>
- 519 de Juan, J., Elósegui, P., Nettles, M., Larsen, T. B., Davis, J. L., Hamilton, G. S.,
 520 . . . Forsberg, R. (2010). Sudden increase in tidal response linked to calving
 521 and acceleration at a large greenland outlet glacier. *Geophysical Research*
 522 *Letters*, *37*(12), L12501. doi: <https://doi.org/10.1029/2010GL043289>
- 523 Enderlin, E. M., Hamilton, G. S., O’Neel, S., Bartholomaus, T. C., Morlighem, M.,
 524 & Holt, J. W. (2016). An empirical approach for estimating stress-coupling
 525 lengths for marine-terminating glaciers. *Frontiers in Earth Science*, *4*, 104.
 526 doi: 10.3389/feart.2016.00104
- 527 Enderlin, E. M., Howat, I. M., Jeong, S., Noh, M.-J., van Angelen, J. H., & van den
 528 Broeke, M. R. (2014). An improved mass budget for the Greenland ice sheet.
 529 *Geophysical Research Letters*, *41*(3), 866–872. doi: 10.1002/2013GL059010
- 530 Felikson, D., A. Catania, G., Bartholomaus, T. C., Morlighem, M., & Noël,
 531 B. P. Y. (2021). Steep glacier bed knickpoints mitigate inland thinning
 532 in greenland. *Geophysical Research Letters*, *48*(2), e2020GL090112. doi:
 533 <https://doi.org/10.1029/2020GL090112>
- 534 Felikson, D., Nowicki, S., Nias, I., Morlighem, M., & Seroussi, H. (2022). Sea-
 535 sonal tidewater glacier terminus oscillations bias multi-decadal projections of
 536 ice mass change. *Journal of Geophysical Research: Earth Surface*, *127*(2),
 537 e2021JF006249. doi: <https://doi.org/10.1029/2021JF006249>
- 538 Fried, M. J., Catania, G. A., Stearns, L. A., Sutherland, D. A., Bartholomaus,
 539 T. C., Shroyer, E., & Nash, J. (2018). Reconciling Drivers of Seasonal Termi-
 540 nus Advance and Retreat at 13 Central West Greenland Tidewater Glaciers.
 541 *Journal of Geophysical Research: Earth Surface*, *123*(7), 1590–1607. doi:
 542 10.1029/2018JF004628
- 543 Gardner, A. S., Fahnestock, M. A., & Scambos, T. A. (2023). *ITS_LIVE regional*
 544 *glacier and ice sheet surface velocities: Version 1*. National Snow and Ice Data
 545 Center. Retrieved from <https://doi:10.5067/6II6VW8LLWJ7>
- 546 Gardner, A. S., Moholdt, G., Scambos, T., Fahnestock, M., Ligtenberg, S., van den
 547 Broeke, M., & Nilsson, J. (2018). Increased west antarctic and unchanged east
 548 antarctic ice discharge over the last 7 years. *The Cryosphere*, *12*(2), 521–547.
 549 doi: 10.5194/tc-12-521-2018
- 550 Goliber, S., Black, T., Catania, G., Lea, J. M., Olsen, H., Cheng, D., . . . Zhang, E.
 551 (2022). Termpicks: a century of greenland glacier terminus data for use in sci-
 552 entific and machine learning applications. *The Cryosphere*, *16*(8), 3215–3233.
 553 doi: 10.5194/tc-16-3215-2022
- 554 Harper, J. T., Bradford, J. H., Humphrey, N. F., & Meierbachtol, T. W. (2010).
 555 Vertical extension of the subglacial drainage system into basal crevasses. *Nat-*
 556 *ure*, *467*(7315), 579–582. doi: 10.1038/nature09398
- 557 Holland, D. M., Thomas, R. H., De Young, B., Ribergaard, M. H., & Lyberth, B.
 558 (2008). Acceleration of Jakobshavn Isbræ triggered by warm subsurface ocean
 559 waters. *Nature geoscience*, *1*(10), 659–664. doi: 10.1038/ngeo316
- 560 Howat, I. M., Joughin, I., Fahnestock, M., Smith, B. E., & Scambos, T. A. (2008).
 561 Synchronous retreat and acceleration of southeast Greenland outlet glaciers
 562 2000–06: ice dynamics and coupling to climate. *Journal of Glaciology*,
 563 *54*(187), 646–660. doi: 10.3189/002214308786570908
- 564 Howat, I. M., Joughin, I., Tulaczyk, S., & Gogineni, S. (2005). Rapid retreat and

- 565 acceleration of Helheim Glacier, east Greenland. *Geophysical Research Letters*,
 566 *32*(22), L22502. doi: 10.1029/2005GL024737
- 567 Iken, A., & Truffer, M. (1997). The relationship between subglacial water pressure
 568 and velocity of findelengletscher, switzerland, during its advance and retreat.
 569 *Journal of Glaciology*, *43*(144), 328–338. doi: 10.3189/S0022143000003282
- 570 IPCC. (2022). *Climate change 2022: Mitigation of climate change* (P. Shukla et al.,
 571 Eds.). Cambridge, UK and New York, NY, USA: Cambridge University Press.
 572 doi: 10.1017/9781009157926
- 573 Joughin, I. (2023). *Measures greenland monthly ice sheet velocity mosaics from*
 574 *sar and landsat, version 5*. NASA National Snow and Ice Data Center Dis-
 575 tributed Active Archive Center. Retrieved from [https://nsidc.org/data/](https://nsidc.org/data/NSIDC-0731/versions/5)
 576 [NSIDC-0731/versions/5](https://nsidc.org/data/NSIDC-0731/versions/5) doi: 10.5067/EGKZX6FXXM4P
- 577 Joughin, I., Das, S. B., King, M. A., Smith, B. E., Howat, I. M., & Moon, T. (2008).
 578 Seasonal Speedup Along the Western Flank of the Greenland Ice Sheet. *Sci-*
 579 *ence*, *320*(5877), 781–783. doi: 10.1126/science.1153288
- 580 Joughin, I., Smith, B. E., Howat, I. M., Floricioiu, D., Alley, R. B., Truffer, M., &
 581 Fahnestock, M. (2012). Seasonal to decadal scale variations in the surface
 582 velocity of Jakobshavn Isbrae, Greenland: Observation and model-based anal-
 583 ysis. *Journal of Geophysical Research: Earth Surface*, *117*(F2), F02030. doi:
 584 10.1029/2011JF002110
- 585 Kehrl, L. M., Joughin, I., Shean, D. E., Floricioiu, D., & Krieger, L. (2017). Sea-
 586 sonal and interannual variabilities in terminus position, glacier velocity, and
 587 surface elevation at Helheim and Kangerlussuaq Glaciers from 2008 to 2016.
 588 *Journal of Geophysical Research: Earth Surface*, *122*(9), 1635–1652. doi:
 589 10.1002/2016JF004133
- 590 King, M. D., Howat, I. M., Candela, S. G., Noh, M. J., Jeong, S., Noël, B. P., ...
 591 Negrete, A. (2020). Dynamic ice loss from the greenland ice sheet driven by
 592 sustained glacier retreat. *Communications Earth & Environment*, *1*(1), 1. doi:
 593 10.1038/s43247-020-0001-2
- 594 Medley, B., Neumann, T. A., Zwally, H. J., Smith, B. E., & Stevens, C. M. (2022).
 595 Simulations of firn processes over the greenland and antarctic ice sheets: 1980–
 596 2021. *The Cryosphere*, *16*(10), 3971–4011. doi: 10.5194/tc-16-3971-2022
- 597 Medrzycka, D., Benn, D. I., Box, J. E., Copland, L., & Balog, J. (2016). Calving be-
 598 havior at rink isbræ, west greenland, from time-lapse photos. *Arctic, Antarctic,*
 599 *and Alpine Research*, *48*(2), 263–277. doi: 10.1657/AAAR0015-059
- 600 Moon, T., Fisher, M., Stafford, T., & Thurber, A. (2023). *Qgreenland (v3) [dataset]*.
 601 National Snow and Ice Data Center. Retrieved from [https://10.5281/zenodo](https://10.5281/zenodo.8326507)
 602 [.8326507](https://10.5281/zenodo.8326507)
- 603 Moon, T., Gardner, A., Csatho, B., Parmuzin, I., & Fahnestock, M. (2020). Rapid
 604 Reconfiguration of the Greenland Ice Sheet Coastal Margin. *Journal of Geo-*
 605 *physical Research: Earth Surface*, *125*(11). doi: 10.1029/2020jf005585
- 606 Moon, T., Joughin, I., & Smith, B. (2015). Seasonal to multiyear variability of
 607 glacier surface velocity, terminus position, and sea ice/ice mélange in northwest
 608 Greenland. *Journal of Geophysical Research: Earth Surface*, *120*(5), 818–833.
 609 doi: 10.1002/2015JF003494
- 610 Moon, T., Joughin, I., Smith, B., & Howat, I. (2012). 21st-Century evolution of
 611 Greenland outlet glacier velocities. *Science*, *336*(6081), 576–578. doi: 10.1126/
 612 science.1219985
- 613 Moon, T., Joughin, I., Smith, B., van den Broeke, M. R., van de Berg, W. J.,
 614 Noël, B., & Usher, M. (2014). Distinct patterns of seasonal Greenland
 615 glacier velocity. *Geophysical Research Letters*, *41*(20), 7209–7216. doi:
 616 10.1002/2014GL061836
- 617 Morlighem, M., Williams, C. N., Rignot, E., An, L., Arndt, J. E., Bamber, J. L.,
 618 ... Zinglarsen, K. B. (2017). BedMachine v3: Complete bed topography
 619 and ocean bathymetry mapping of Greenland from multibeam echo sound-

- ing combined with mass conservation. *Geophysical Research Letters*, *44*(21), 11,051–11,061. doi: 10.1002/2017GL074954
- Morlighem, M., Williams, C. N., Rignot, E., An, L., Arndt, J. E., Bamber, J. L., ... Zinglensen, K. B. (2022). *Icebridge bedmachine greenland, version 5 [data set]*. Boulder, Colorado USA. NASA National Snow and Ice Data Center Distributed Active Archive Center. Retrieved from <https://doi.org/10.5067/GMEVBWFLWA7X.DateAccessed08-30-2023>.
- Nettles, M., Larsen, T. B., Elósegui, P., Hamilton, G. S., Stearns, L. A., Ahlstrøm, A. P., ... Forsberg, R. (2008). Step-wise changes in glacier flow speed coincide with calving and glacial earthquakes at helheim glacier, greenland. *Geophysical Research Letters*, *35*(24), L24503. doi: <https://doi.org/10.1029/2008GL036127>
- Noël, B., van de Berg, W. J., Lhermitte, S., & van den Broeke, M. R. (2019). Rapid ablation zone expansion amplifies north greenland mass loss. *Science Advances*, *5*(9), eaaw0123. doi: 10.1126/sciadv.aaw0123
- Poinar, K. (2023). Seasonal flow types of glaciers in sermilik fjord, greenland, over 2016–2021. *Journal of Geophysical Research: Earth Surface*, *128*(7), e2022JF006901. doi: <https://doi.org/10.1029/2022JF006901>
- Porter, C., Morin, P., Howat, I. M., Noh, M.-J., Bates, B., Peterman, K., ... Bojesen, M. (2022). *ArcticDEM - Strips, Version 4.1*. Retrieved 2023, from <https://doi.org/10.7910/DVN/C98DVS>
- Rennermalm, A. K., Smith, L. C., Chu, V. W., Box, J. E., Forster, R. R., Van den Broeke, M. R., ... Moustafa, S. E. (2013). Evidence of meltwater retention within the greenland ice sheet. *The Cryosphere*, *7*(5), 1433–1445. doi: 10.5194/tc-7-1433-2013
- Robel, A. A., Sim, S. J., Meyer, C., Siegfried, M. R., & Gustafson, C. D. (2023). Contemporary ice sheet thinning drives subglacial groundwater exfiltration with potential feedbacks on glacier flow. *Science Advances*, *9*(33), eadh3693. doi: 10.1126/sciadv.adh3693
- Schoof, C. G. (2010, 12). Ice-sheet acceleration driven by melt supply variability. *Nature*, *468*(7325), 803–806. doi: 10.1038/nature09618
- Sergienko, O. V. (2022). Marine outlet glacier dynamics, steady states and steady-state stability. *Journal of Glaciology*, *68*(271), 946–960. doi: 10.1017/jog.2022.13
- Shepherd, A., Ivins, E. R., A, G., Barletta, V. R., Bentley, M. J., Bettadpur, S., ... Zwally, H. J. (2012). A reconciled estimate of ice-sheet mass balance. *Science*, *338*(6111), 1183–1189. doi: 10.1126/science.1228102
- Solgaard, A. M., Rapp, D., Noël, B. P. Y., & Hvidberg, C. S. (2022). Seasonal patterns of greenland ice velocity from sentinel-1 sar data linked to runoff. *Geophysical Research Letters*, *49*(24), e2022GL100343. doi: <https://doi.org/10.1029/2022GL100343>
- Stevens, L. A., Nettles, M., Davis, J. L., Creyts, T. T., Kingslake, J., Hewitt, I. J., & Stubblefield, A. (2022). Tidewater-glacier response to supraglacial lake drainage. *Nature Communications*, *13*(1), 6065. doi: 10.1038/s41467-022-33763-2
- Sundal, A. V., Shepherd, A., Nienow, P., Hanna, E., Palmer, S., & Huybrechts, P. (2011). Melt-induced speed-up of greenland ice sheet offset by efficient subglacial drainage. *Nature*, *469*(7331), 521–524. doi: 10.1038/nature09740
- Ultee, L., Felikson, D., Minchew, B., Stearns, L. A., & Riel, B. (2022). Helheim glacier ice velocity variability responds to runoff and terminus position change at different timescales. *Nature Communications*, *13*(1), 6022. doi: 10.1038/s41467-022-33292-y
- van den Broeke, M. R., Enderlin, E. M., Howat, I. M., Kuipers Munneke, P., Noël, B. P. Y., van de Berg, W. J., ... Wouters, B. (2016). On the recent contribution of the greenland ice sheet to sea level change. *The Cryosphere*, *10*(5),

- 1933–1946. doi: 10.5194/tc-10-1933-2016
- 675 Van der Veen, C. J. (2013). *Fundamentals of glacier dynamics* (Second Edition ed.).
 676 New York: CRC press.
- 677 Veen, C. J. v. d., Plummer, J., & Stearns, L. A. (2011). Controls on the recent
 678 speed-up of Jakobshavn Isbrae, West Greenland. *Journal of Glaciology*,
 679 *57*(204), 770 – 782.
- 680 Vieli, A., Jania, J., Blatter, H., & Funk, M. (2004). Short-term velocity varia-
 681 tions on hansbreen, a tidewater glacier in spitsbergen. *Journal of Glaciology*,
 682 *50*(170), 389–398. doi: 10.3189/172756504781829963
- 683 Vijay, S., Khan, S. A., Kusk, A., Solgaard, A. M., Moon, T., & Bjørk, A. A.
 684 (2019). Resolving seasonal ice velocity of 45 greenlandic glaciers with very
 685 high temporal details. *Geophysical Research Letters*, *46*(3), 1485–1495. doi:
 686 <https://doi.org/10.1029/2018GL081503>
- 687 Vijay, S., King, M. D., Howat, I. M., Solgaard, A. M., Khan, S. A., & Noël, B.
 688 (2021). Greenland ice-sheet wide glacier classification based on two distinct
 689 seasonal ice velocity behaviors. *Journal of Glaciology*, *67*(266), 1241–1248. doi:
 690 10.1017/jog.2021.89
- 691 Werder, M. A., Hewitt, I. J., Schoof, C. G., & Flowers, G. E. (2013). Modeling chan-
 692 nelerized and distributed subglacial drainage in two dimensions. *Journal of Geo-*
 693 *physical Research: Earth Surface*, *118*(4), 2140–2158. doi: [https://doi.org/10](https://doi.org/10.1002/jgrf.20146)
 694 [.1002/jgrf.20146](https://doi.org/10.1002/jgrf.20146)
- 695 Wood, M., Rignot, E., Fenty, I., An, L., Bjørk, A., van den Broeke, M., ... others
 696 (2021). Ocean forcing drives glacier retreat in greenland. *Science Advances*,
 697 *7*(1), eaba7282. doi: 10.1126/sciadv.aba7282
- 698 Zhang, E., Catania, G., & Trugman, D. T. (2023). Autoterm: an automated pipeline
 699 for glacier terminus extraction using machine learning and a “big data” repos-
 700 itory of greenland glacier termini. *The Cryosphere*, *17*(8), 3485–3503. doi:
 701 10.5194/tc-17-3485-2023
 702

703 Open Research Section

704 Velocity data can be freely downloaded at <https://its-live.jpl.nasa.gov/> (Gardner
 705 et al., 2023). ArcticDEMs are available at <https://livingatlas2.arcgis.com/arcticdemexplorer/>.
 706 The code, terminus data, flowlines, and DG-IS2-DEMs can be downloaded at [https://](https://zenodo.org/record/8428196)
 707 zenodo.org/record/8428196.

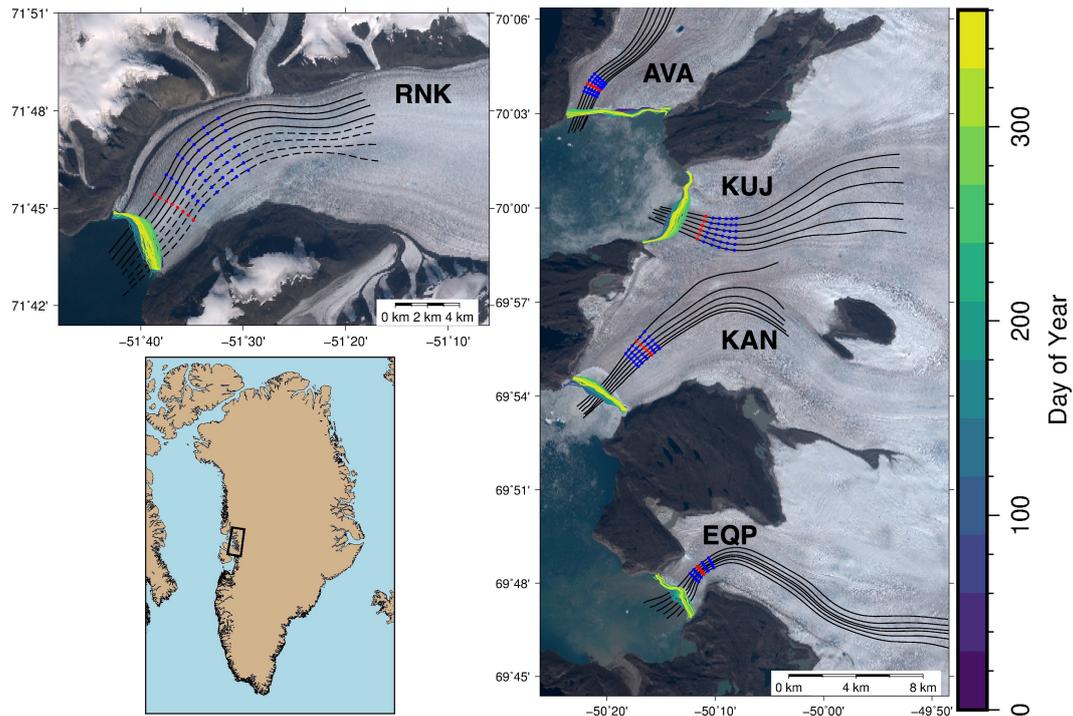


Figure 1. Glacier and data locations. Glaciers examined in this study include Rink Isbrae (RNK), Sermeq Avannarleq (AVA), Sermeq Kujalleq (KUJ), Kangilernata Sermia (KAN), and Eqip Sermia (EQP). Black solid flowlines represent grounded portions of the glaciers while dashed flowlines represent floating portions. The terminus traces in 2018 from AutoTerm are colored by date. Colored points on flowlines are the locations where we obtain velocity time series from ITS_LIVE. The results section shows the results of the red points. Blue points are for testing the sensitivity of the position of velocity to the results.

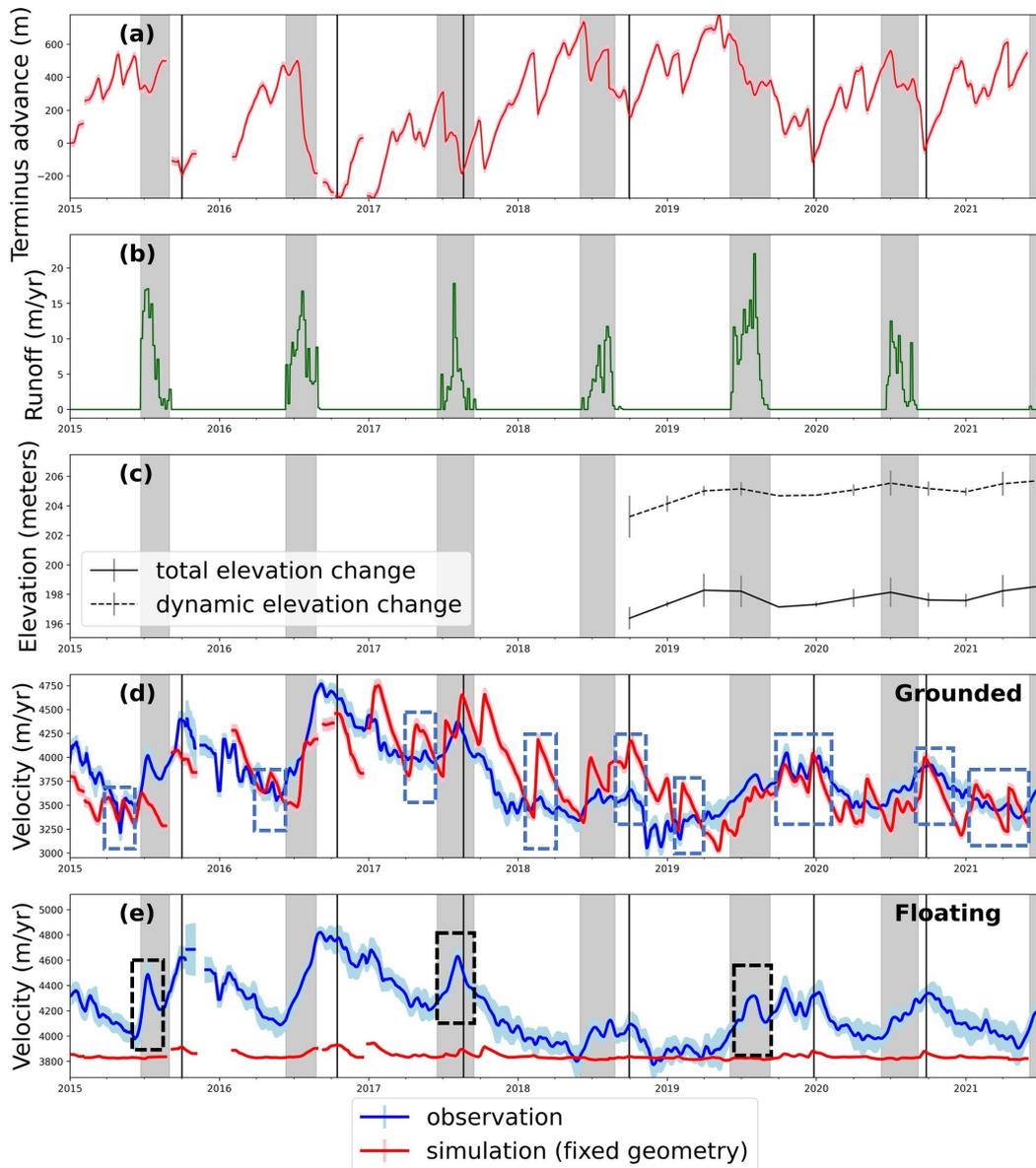


Figure 2. The full record of RNK. (a) is the terminus variation with pink shading representing uncertainty derived from AutoTerm. (b) Displays the runoff time series. (c) Illustrates surface elevation changes. (d) Compares simulated and observed velocity for grounded flowlines. (e) Extends the comparison to floating flowlines (as in d). The dashed boxes show the accelerations caused by runoff.

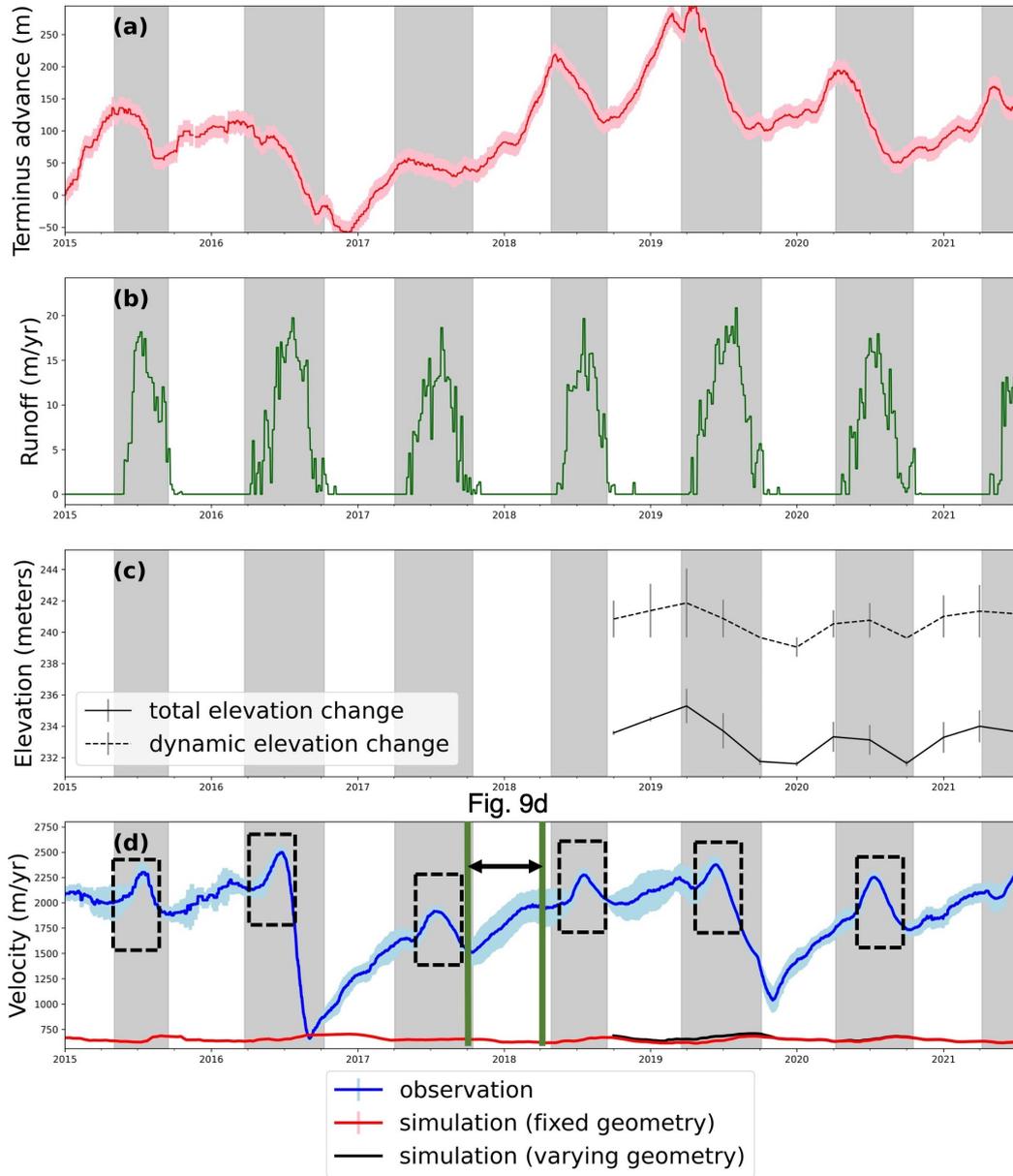


Figure 3. The full record of AVA. The figure follows the same design as Figure 2. The vertical green lines indicate the study period in Figure 10d.

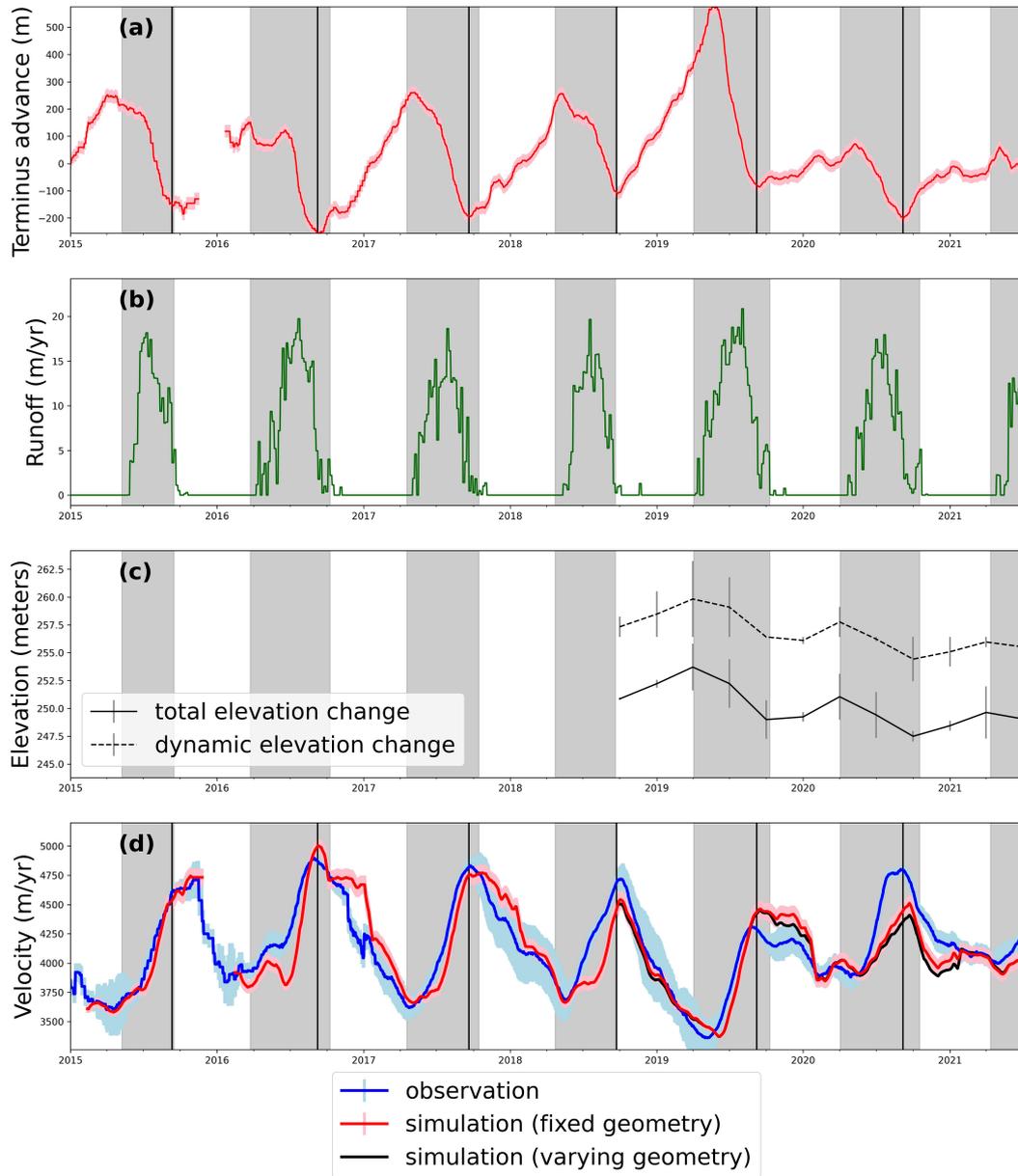


Figure 4. The full record of KUJ. The figure follows the same design as Figure 2.

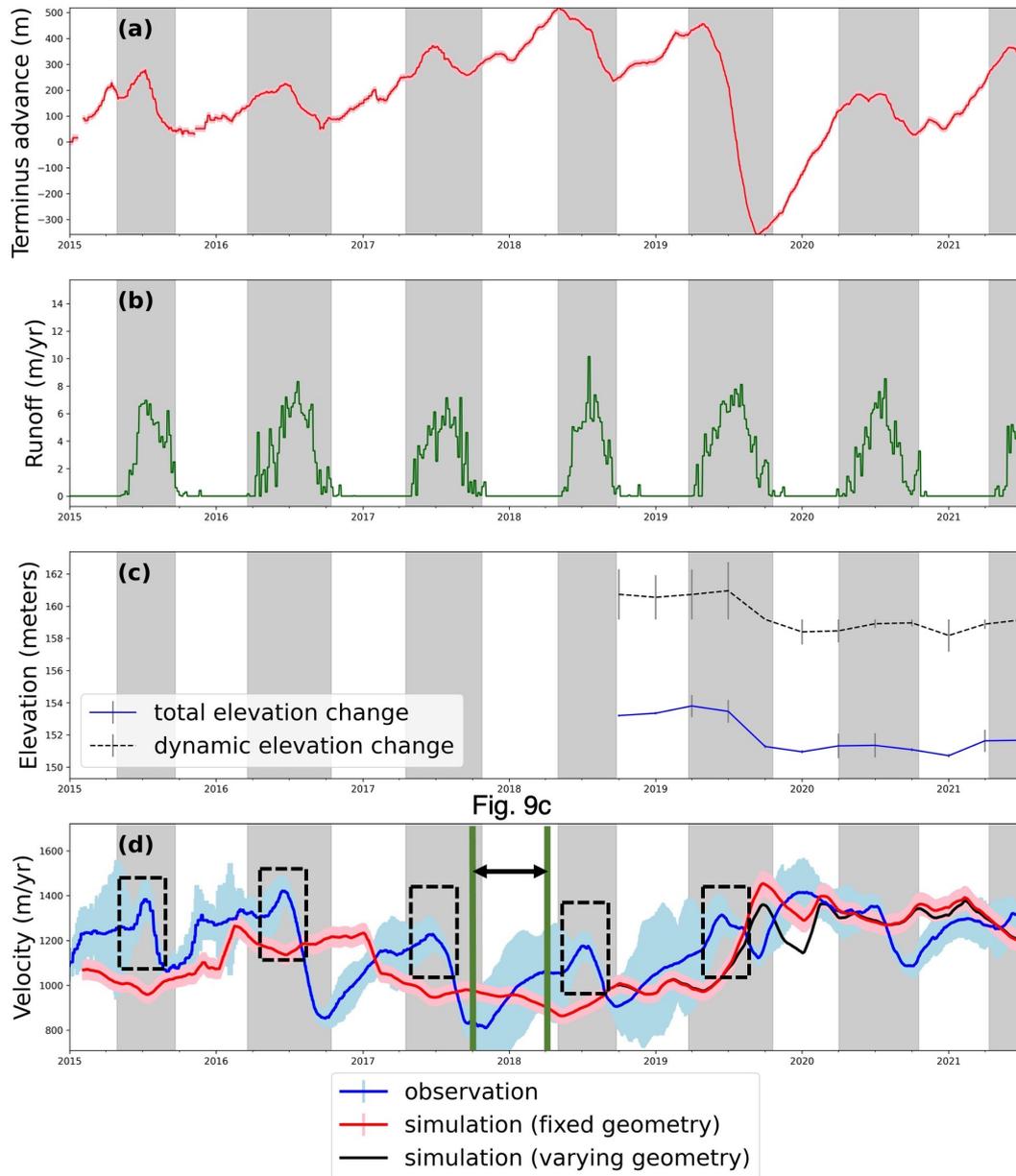


Figure 5. The full record of KAN. The figure follows the same design as Figure 2. The vertical green lines indicate the study period in Figure 10c.

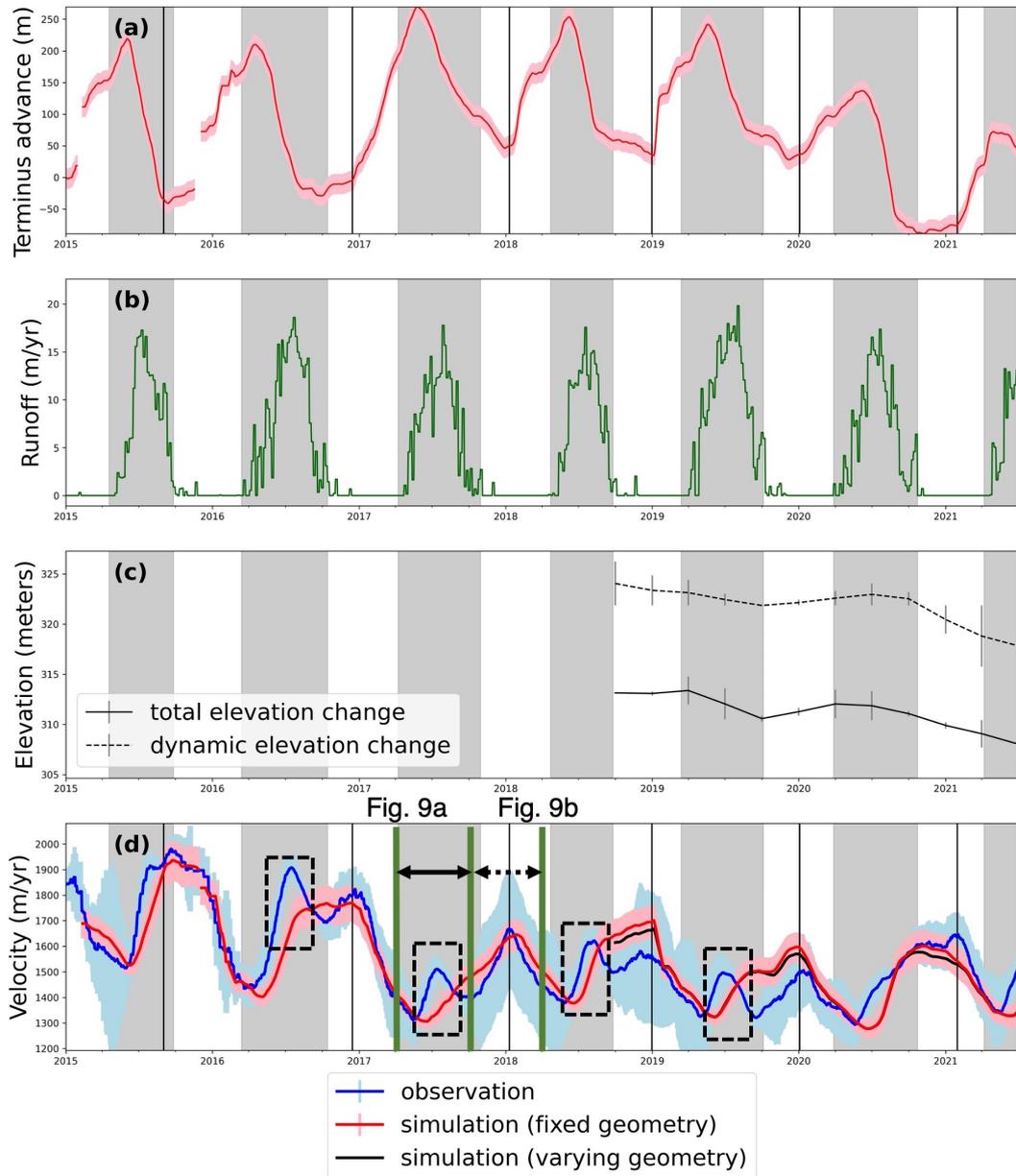


Figure 6. The full record of EQP. The figure follows the same design as Figure 2. The vertical green lines indicate the study period in Figure 10a and 10b.

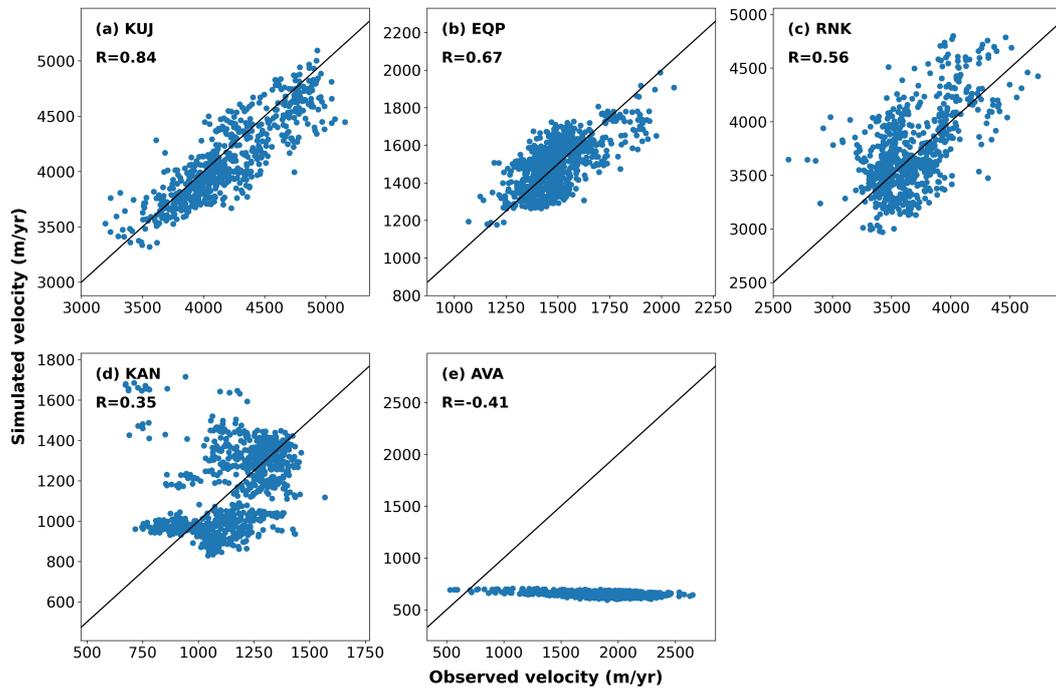


Figure 7. Scatter plot showing the comparison between simulated and observed velocity for KUJ, EQP, RNK, KAN, and AVA. The 1:1 line is as shown for each.

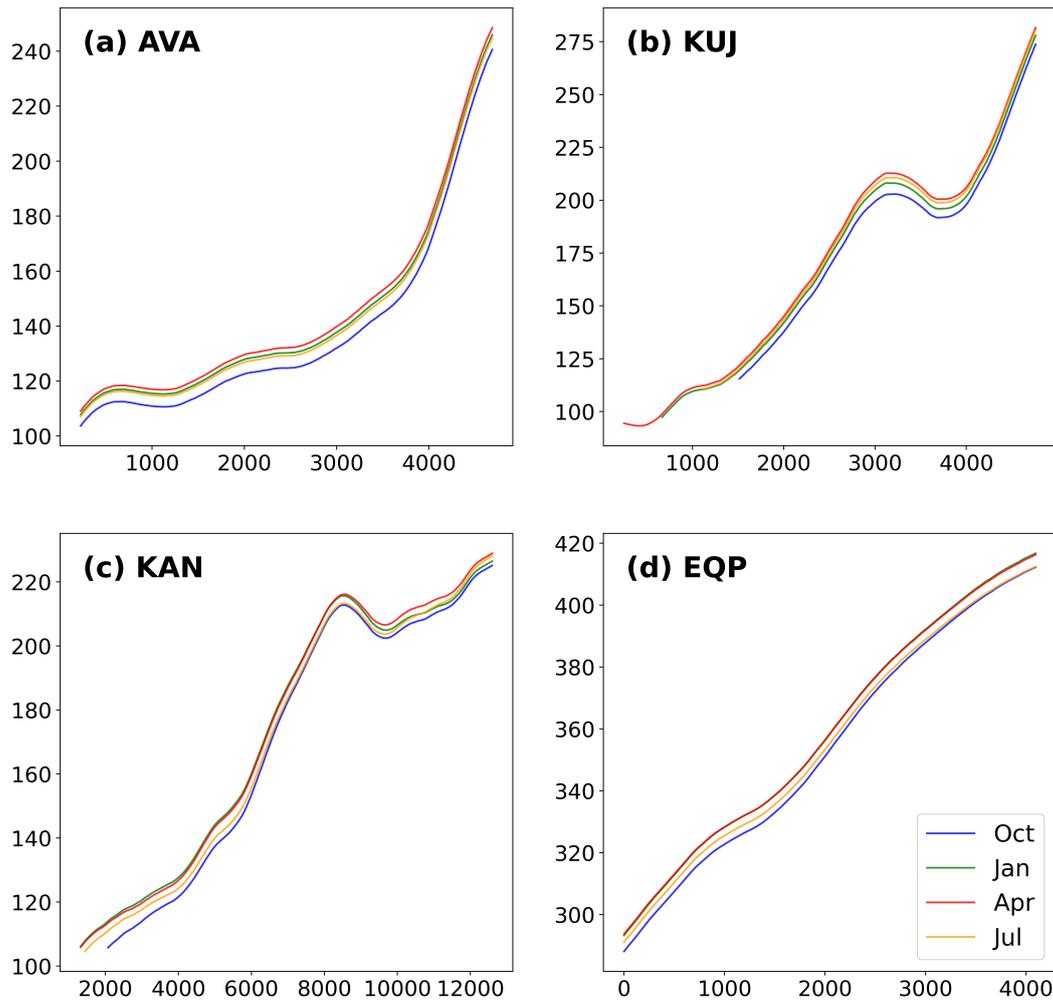


Figure 8. Surface elevation profiles in 2019 for AVA, KUJ, KAN, and EQP.

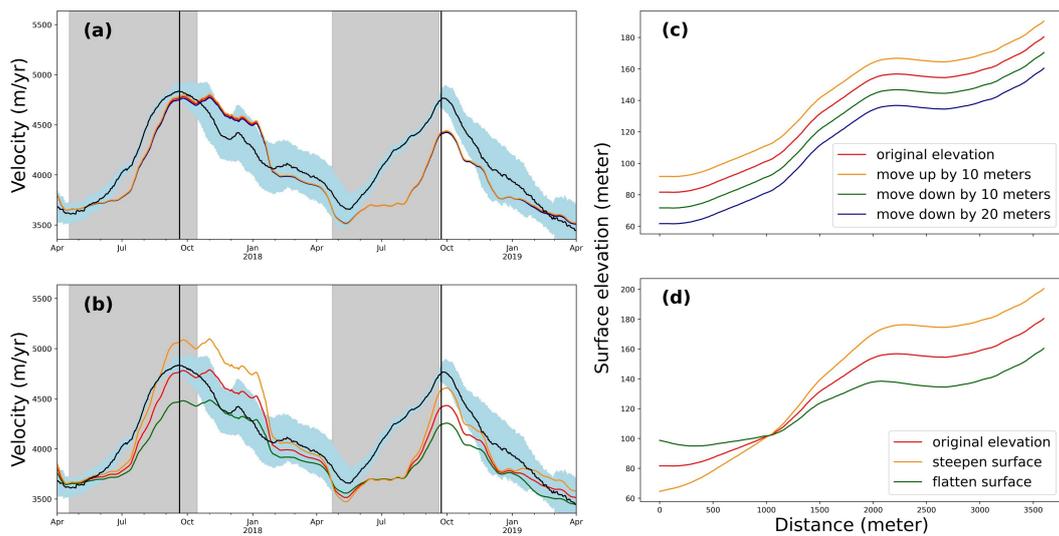


Figure 9. Experiment results using the artificially modified surface elevations. The results of (a) correspond to (c), and the results of (b) correspond to (d).

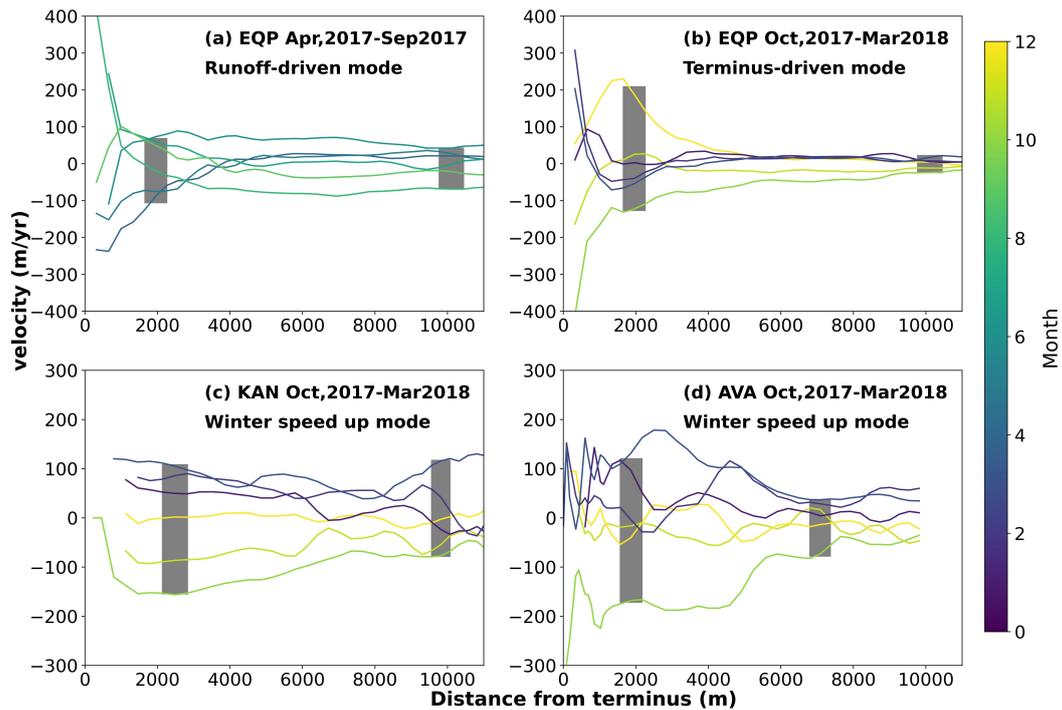


Figure 10. Velocity profiles over time for EQP, KAN, and AVA. The average velocity profile has been subtracted for a better display of changes over time. The original velocity profiles are shown in Figure S8. The shaded areas indicate regions where we obtain velocity variations in the frontal and upstream sections. (a) Velocity profiles of EQP during the melt season. (b) Velocity profiles of EQP after the melt season, during which velocity is primarily influenced by terminus changes. (c) Velocity profile of KAN during winter and early melt season. (d) Velocity profile of AVA during winter and early melt season.