

1 **Characterizing geologic and climatic controls on rockfall hazards**  
2 **using an inventory and integrated kinematic and runout model with**  
3 **an integrated kinematic analysis and runout model: Skagway,**  
4 **Alaska, USA**

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11 **Abstract**

12 Rockfall is common in steep terrain and poses a hazard to nearby communities. While rockfall triggering mechanisms are  
13 highly variable and difficult to quantify, the susceptibility of rock slopes to planar, wedge, or toppling failure can be readily  
14 assessed using kinematic analysis. As such, valley slopes with favourable joint orientations exhibit high rockfall  
15 susceptibility although the potential for rockfall runout to impact infrastructure and public safety depends on the morphology  
16 of downslope terrain. Integrating rockfall susceptibility and runout models with maps of talus deposits accumulated from  
17 past rockfall events is an effective combination of tools to inform mitigation but can be difficult to accomplish across  
18 extensive areas. Here, we combine these methods with a a-historic rockfall inventory spanning 2005 to 2022 to assess  
19 geologic and climate controls on rockfall hazard-activity in the steep and forested postglacial valleys proximal to Skagway,  
20 Alaska, where recent rockfall activity has imperilled public safety, infrastructure, and tourism. The inventory reveals  
21 rockfall activity throughout the year with peak activity in early spring that coincides with a rapid rise in minimum daily  
22 temperatures. Our field investigations identified two steeply dipping orthogonal joint sets that favour toppling failure along  
23 NW-facing hillslopes in the lower Skagway River valley as well as the NW-facing valleys that bound nearby Dyea Bay and  
24 Nahku Bay. We used new and existing lidar data and 405-300 field-derived joint orientations to inform a kinematic toppling  
25 failure model that identifies likely zones of rock toppling that identifies. The predicted source zones are positioned upslope of  
26 abundant talus slopes that we mapped from field observations and lidar analyses. Along the prominent ridgeline on the  
27 eastern margin of Skagway, we used We coupled the source zones with RAMMS:Rockfall to model-simulate nearly  
28 200,000-197,800 rockfall runout events for four scenarios that account for with varying variations in elast-block size and  
29 ground cover. The runout predictions highlight distinct zones of low and high rockfall -propagation hazard-susceptibility  
30 along the ridgeline that result from changes in hillslope morphology is negatively correlated with hillslope roughness which

31 ~~results from set by~~ the combined influence of joint orientations ~~that generate bedrock benches~~ and the ~~spatial~~ pattern of  
32 glacial erosion. High-hazard segments of the ridgeline exhibit distinct bedrock ~~escarpments-cliffs~~ and slope-spanning talus  
33 slopes that result from the accumulation of rockfall activity over millennia. ~~Taken together, our findings illustrate past~~  
34 ~~controls on rockfall location and timing. Our findings reveal controls on past and future rockfall activity and can be used to~~  
35 inform mitigative measures.

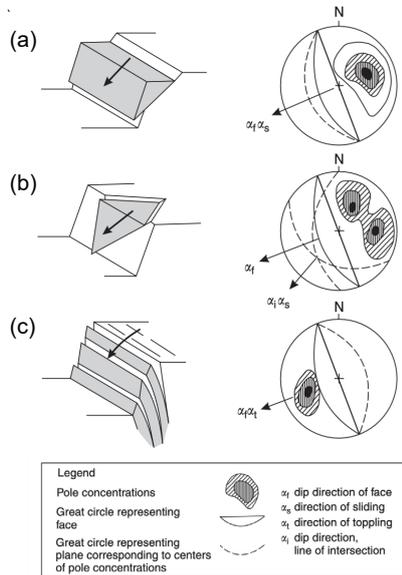
## 36 1 Introduction

37 In steep, rocky landscapes, the detachment and downslope movement of discrete rock fragments (~~i.e., rockfall~~) can occur  
38 frequently and poses a significant hazard to proximal communities and infrastructure (Hungar et al., 2014). Rockfall activity  
39 has been attributed to an array of highly disparate conditioning and triggering processes, including precipitation, frost  
40 weathering, insolation, seismic activity, and slope modification (Collins and Stock, 2016; Rosser and Massey, 2022). In  
41 Yosemite Valley, ~~California~~, for example, rockfall triggers include rainfall events, snow melt, and freeze-thaw action that  
42 can increase pressure along joints (Stock et al., 2011; Wieczorek and Jäger, 1996). Furthermore, rockfalls in Yosemite  
43 Valley may also be triggered on warm summer days by cyclic solar heating, which can propagate exfoliation fractures and  
44 lead to detachment (Collins and Stock, 2016). ~~In the European Alps, warming and permafrost degradation have been~~  
45 ~~implicated in significant increases in rockfall activity~~ (Allen and Huggel, 2013; Bajni et al., 2021; Bodin et al., 2015;  
46 Paranunzio et al., 2015, 2016). Despite significant progress in characterizing ~~these and other~~ rockfall triggering mechanisms,  
47 prediction of rockfall timing and location has limited ability to inform warning systems (Rosser and Massey, 2022).  
48 Precursory rock deformation can signal future activity (Abellán et al., 2010; Rosser et al., 2007; Royán et al., 2014), but  
49 current methods to quantify precursor deformation across extensive areas composed of steep, high-relief surfaces are limited  
50 and oftentimes rockfall occurs without prior deformation (Abellán et al., 2011). As a result, identifying rock slopes with the  
51 propensity to generate rockfall, often referred to as rockfall source areas, is a key first step in mitigating rockfall hazards.

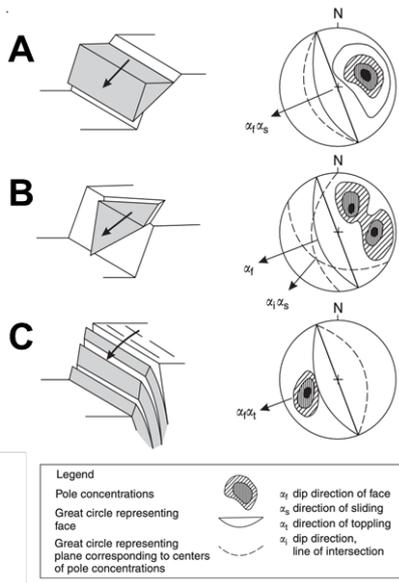
52  
53 A wide array of methods has been proposed to assess the extent to which hillslopes are prone to rockfall activity. Delineating  
54 potential source zones (e.g., Loye et al., 2009) can be accomplished from direct observation of past events, which assumes  
55 the location of past detachments coincides with the ~~likely~~ location of future rockfall activity (Luckman, 1976; Matsuoka and  
56 Sakai, 1999; Rapp, 1960; Whalley, 1984). Source zones can also be inferred from distinctive evidence such as talus ~~slope~~  
57 and scree deposits that have accumulated below cliff faces (Borella et al., 2019; Frattini et al., 2008; Moore et al., 2009;  
58 Stock and Collins, 2014). Coupled assessment of rock slope morphology and the properties of potential source zones is  
59 another common approach that uses digital elevation models (DEMs) for assessing source zones over large areas (Frattini et  
60 al., 2008; Guzzetti et al., 2003; Messenzehl et al., 2017; Samodra et al., 2016), while more data-intensive and physically-  
61 based deformation models can be used for slope-scale analyses (Matasci et al., 2018). In ~~the~~ absence of rock structure data,  
62 some studies (e.g., Guerriero et al., 2024) have applied morphologic criteria (e.g., slope and curvature thresholds) to DEMs

63 to identify anomalous rock slope protrusions that are likely to experience rockfall events (Aksoy and Ercanoglu, 2006;  
 64 Frattini et al., 2008; Guzzetti et al., 2003; Marquínez et al., 2003; Sarro et al., 2024). Taken together, these studies suggest  
 65 that the approach to rockfall characterization is highly dependent on the availability of data, expansiveness of the study area,  
 66 and the desired level of confidence.

67  
 68 Because rockfall is typically localized along bedding planes, fractures, or joints, collectively referred to as discontinuities,  
 69 susceptibility can also be evaluated by determining the geometry of these planes of weakness with respect to the slope and  
 70 orientation of rock slopes. Kinematic analysis identifies blocks that can experience instability according to sliding, toppling,  
 71 or wedge failure criteria as determined by the geometry of rock slopes and discontinuities (Fig. 1) (Bovis and Evans, 1996;  
 72 Wyllie and Mah, 2004). Combining high-resolution DEMs with rock structure data can inform kinematic analyses and  
 73 determine the relative likelihood of different failure modes across expansive-extensive study areas (Grant et al., 2016;  
 74 Günther, 2003; Kundu et al., 2023; Meentemeyer and Moody, 2000; Stock and Collins, 2014). Recent applications of  
 75 kinematic analysis leverage lidar or photogrammetry to extract bedrock discontinuity data and test kinematic failure criteria  
 76 on complex slope geometries, like overhanging rock quarries (Fanos and Pradhan, 2018; Gigli et al., 2022). Increasingly, the  
 77 acquisition of discontinuity data is accomplished using automated analysis of point cloud data acquired from terrestrial laser  
 78 scanning (TLS) (Matasci et al., 2018) or lidar or photogrammetry acquired from uncrewed aerial systems (UAS) (Utlu et al.,  
 79 2023). These approaches are powerful but can be challenging to implement across large areas characterized by steep, rocky



80 forested slopes where the details of rock structure are often obscured by vegetation. ~~As such, traditional, field-based means~~  
 81 ~~of bedrock structural characterization continue to be relevant.~~



83 Figure 1. Schematic ~~and corresponding stereonet representations representation~~ of common rock slope failure modes: (a) planar  
 84 sliding, (b) wedge, and (c) toppling. On the ~~right of each diagram are stereonet plots, where the the~~ rock slope face orientation  
 85 is represented by a solid black line, and poles to planar discontinuities that meet conditions for failure are shown in shaded  
 86 contours. Dashed lines represent planes of these discontinuities. Modified from Wyllie and Mah (2004).

87  
 88 Initially, rock fragments move via creep, sliding, toppling, or falling before traveling downslope by following ballistic paths,  
 89 sliding, and rolling across rocky or talus slopes until stopping when sufficient energy dissipation has occurred via impacts  
 90 and/or friction (Caviezel et al., 2021). Computational rockfall runout models that account for the physics of these rockfall  
 91 processes can estimate the trajectories and kinetic energy of falling rocks to determine potential downslope impacts (Dorren  
 92 et al., 2007; Lu et al., 2021; Scheidl et al., 2020). Many DEM-based tools exist with a range of parameterization options to  
 93 perform physically-based rockfall simulations and predict the path of rocks over complex terrain and across variable land  
 94 cover (Lu et al., 2021; Moos et al., 2021). These models have been successfully employed to mitigate rockfall hazards with  
 95 diversion and attenuation structures, development setbacks, signage, or other means, in a variety of settings, including mines  
 96 and national parks (Guerriero et al., 2024; Klimeš et al., 2024; Stock and Collins, 2014; Stoffel et al., 2024). As a result, the

97 ~~risk associated with rockfall activity~~~~such that rockfall risk~~ can be reduced even though accurate prediction of triggering  
98 events remains elusive (Rosser and Massey, 2022).

99  
100 Although rockfall activity occurs in a wide range of geologic and climatic settings, it is particularly commonplace in post-  
101 glacial landscapes owing to glacial erosion that alters near-surface stresses, fracture density, topographic variations from  
102 glacial erosion, and changes in environmental conditions that occur in the wake of retreating glaciers (Ballantyne, 2002;  
103 Leith et al., 2014). In particular, the spatial pattern of glacial erosion can ~~follow-align with~~ the fabric ~~or-trend~~ of bedrock  
104 discontinuities and set up failure-prone conditions across extensive areas. As such, relatively small changes in the orientation  
105 and geometry of glacial valleys relative to the orientation of discontinuities can result in significant and systematic variations  
106 in rockfall susceptibility. ~~Although rockfall runout susceptibility is often simulated, the terrain characteristics that govern~~  
107 ~~why and where rockfall events tend to propagate long distances and pose risks downslope are seldom quantified. As a result,~~  
108 ~~geologic and geomorphic controls on rockfall hazard potential are typically not well understood, which prevents us from~~  
109 ~~addressing basic research and applied hypotheses such as how the magnitude and frequency of events accumulate and~~  
110 ~~contribute to mountain-scale erosion and how post-glacial slope morphology affects the efficacy of mitigation efforts~~  
111 ~~Although it has been implied that the cumulative impact of small but frequent rockfalls in post-glacial settings can match~~  
112 ~~that associated with large-scale but less frequent catastrophic or progressive rock slope failure, data are currently unavailable~~  
113 ~~to rigorously test this notion~~ (Barlow et al., 2012; Corominas et al., 2014; Hales and Roering, 2007; Hungr et al., 1999;  
114 Moore et al., 2009; Rosser and Massey, 2022).

115  
116 Rockfall activity is common across much of Southeast Alaska but has been particularly acute in the Municipality of  
117 Skagway, which is situated in a narrow, glacially carved valley herein referred to as the ‘Skagway River valley’ and hosts  
118 vigorous cruise ship tourism from late spring to early fall. Indigenous knowledge of avalanches in the area has been  
119 established (Thornton, 2010) and ~~western~~ documentation of rockfall activity in Skagway ~~by colonial settlers~~ began in the late  
120 1800s when gold prospecting fuelled the establishment of the town. Following decades of sporadic activity, several large  
121 rockfall events in summer 2022 impacted cruise docks along Skagway Harbor and generated renewed concern about the  
122 extent and scope of rockfall hazards in the area. In particular, the extent and timing of past rockfall activity is not well known  
123 and the susceptibility of rockfall initiation and runout in the area, and particularly along a ~5km long ridgeline that abuts the  
124 harbour, town centre, and railroad, has not been characterized. ~~This study area affords an opportunity to explore novel~~  
125 ~~approaches to combining kinematic and runout models to assess how glacial erosion and bedrock properties modulate the~~  
126 ~~spatial pattern of rockfall hazard.~~ In this contribution, we summarize ~~a rockfall inventory~~ and geologic data that  
127 reflects the distribution and timing of past rockfall activity, document rock structure data from field observations, and  
128 synthesize new and existing lidar data from airborne and UAS platforms to inform ~~an integrated~~ kinematic analysis  
129 susceptibility and dynamic runout model ~~for Skagway and the surrounding area.~~ Our findings establish the pervasive imprint  
130 of rockfall activity along slopes oriented to promote toppling failure ~~and demonstrates how slope morphology determines~~

131 rockfall propagation susceptibility. We highlight how the pattern of glacial erosion resulted in substantial rockfall erosion  
132 and cliff retreat along favourably oriented slopes while unfavourably oriented slopes experienced minimal modification since  
133 glacial retreat. Our integrated mapping and modelling approach~~coupled modeling reveals~~ reveal high variability in potential  
134 impacts which can inform mitigation efforts~~suggests that postglacial feedbacks between rock structure and hillslope~~  
135 evolution can generate strong variability in potential rockfall impacts and thus inform current mitigation efforts and  
136 anticipate future needs. -

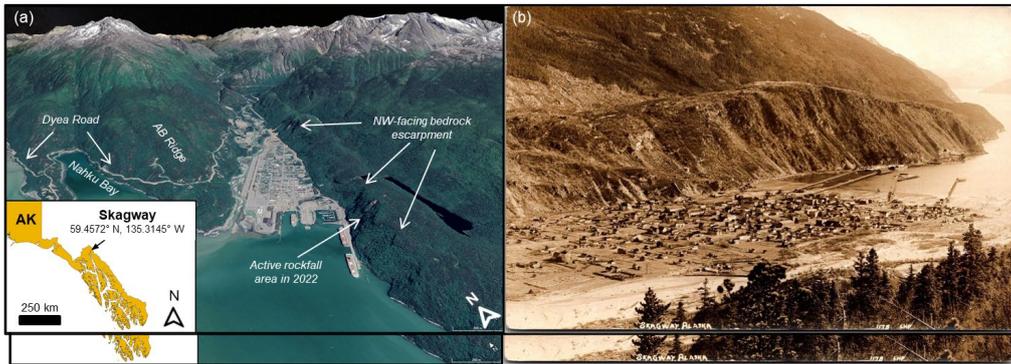
## 137 2 Study area: Skagway, SE Alaska

### 138 2.1. Geology and glacial history

139 Near the northern extent of the Alaska Panhandle, Skagway is situated in the Taiya Inlet atop deltaic and fluvial deposits  
140 near the outlet of a deep fjord (Fig. 2). The surrounding terrain is steep and rugged, composed of Tertiary granodiorite of the  
141 Coast Range Batholith, a belt of plutonic and metamorphic rocks that extends to northern Washington (Yehle and Lemke,  
142 1972). Deformation in southeastern Alaska and southwest Yukon is governed by the subduction and translation of  
143 the Pacific-Yakutat plates relative to the North American plate in the St. Elias region (Biegel et al., 2024). The Eastern  
144 Denali Fault and the Chatham strait fault lineaments, both strike-slip fault systems, meet just south of Skagway (Choi et al.,  
145 2021). Deformation associated with these structures appears to impart a significant influence on the orientation of glacial  
146 valleys, as fjords in the area tend to be linear, striking north and northeasterly (Yehle and Lemke, 1972). A dam  
147 reconnaissance study focused on West Creek, a drainage just 9 km northwest of Skagway, identified three joint sets in the  
148 granodiorite bedrock (Fig. 3e), including two abundant sets with northeast-strike and vertical or steep dips to the south, and  
149 one less abundant set with northwest-strike and a consistent vertical dip (Callahan and Wayland, 1965). That study ~~authors~~  
150 also noted the coincident orientation of topographic lineations in the area and the strike of joints interpreted to be splays from  
151 the Chatham strait lineament, which is related to the nearly 3,000-km long Denali fault system. Spacing between the joints is  
152 variable, ranging from 1 to 4 meters, and the joints do not exhibit slickensides or cataclastic fabric (Fig. 3a-c-b). Sheeting  
153 joints in granodiorite observed near the tops of glaciated ridges are slightly curved or irregular, tend to parallel the ground  
154 surface, and spaced from 1 to 2 meters apart at the surface. Regional studies of glacial history imply that the most recent  
155 episode of major glacial retreat and valley exposure in Skagway occurred 10 to 12kya (Baichtal et al., 2021; Menounos et al.,  
156 2017) and icefields persist today in nearby inland valleys. The steep slopes around Skagway are generally devoid of glacial  
157 till owing to post-retreat erosion and deposition in valley floors in the form of alluvial fans and colluvial deposits.

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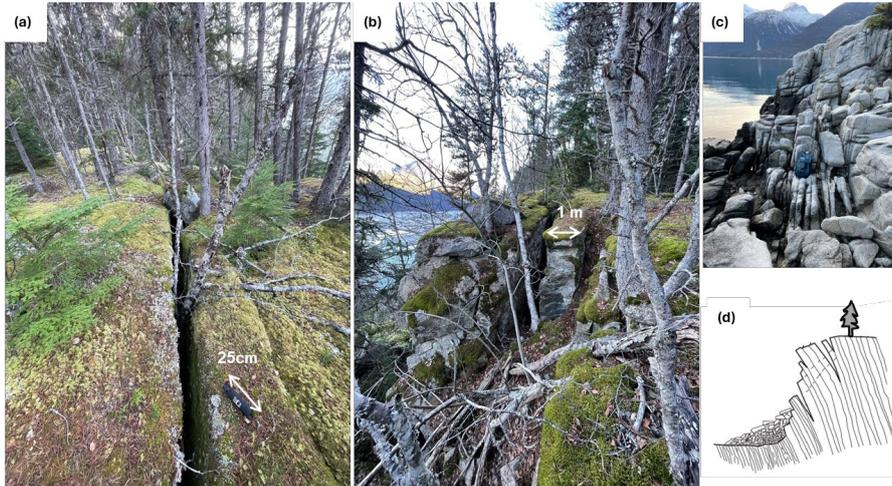
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158  
159  
160 Figure 2. Study area. (a) Oblique view looking N-NE into Skagway River valley, Alaska. Note active rockfall along the NW-  
161 facing ridgeline and escarpment above the harbour and cruise ship docks. Use of Google Earth ©2025 permitted for non-  
162 ~~commercial~~ commercial use. (b) Oblique view from AB Ridge looking SE across Skagway towards the NW-facing ridgeline  
163 and escarpment. This undated and uncredited historic image (Wright et al., 2021) postdates the late 1890s construction of the  
164 current Skagway City Hall and Museum (photo provided by A. Beierly). Local steep cliff faces with fresh bedrock exposure  
165 signify active rockfall. Note the lack of vegetation and the sharp bedrock escarpment along the crest of the ridgeline  
166 and the abundance of active talus slopes that connect to the harbour.

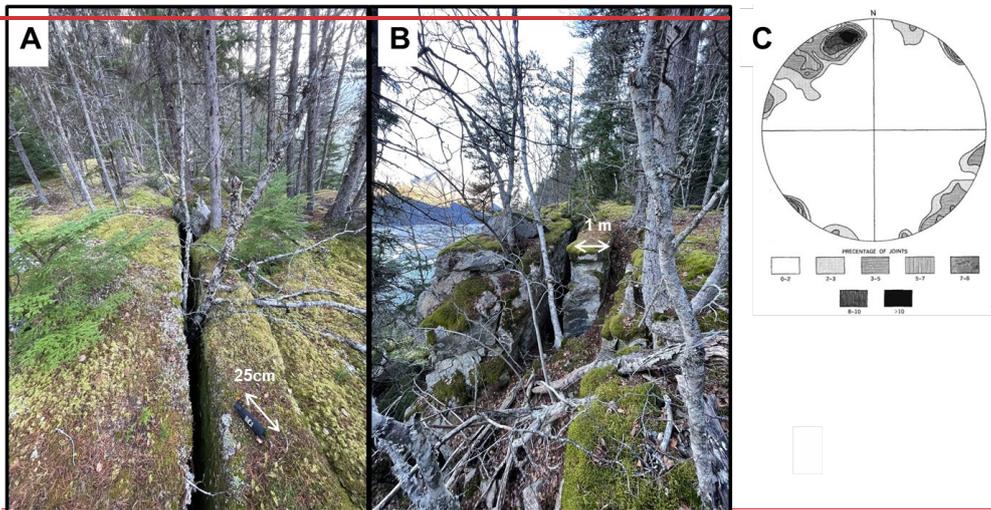
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170 valleys. The steep slopes around Skagway are generally devoid of glacial till owing to post-retreat erosion and deposition in  
171 valley floors in the form of alluvial fans and colluvial deposits.

172



174  
175 Figure 3. Images (a.) and (b) of open joints and incipient toppling along the bedrock escarpment on the NW-facing ridgeline  
176 above Skagway Harbor. Equal-area, lower hemisphere stereonet with 1% area contours of 60 joints measured near Dyea, west  
177 of Skagway. (c) Image of joint orientations along shore of Nahku Bay. (d) Schematic of toppling prone hillside with opening  
178 joints and talus accumulation modified from Wyllie and Mah (2004). Photos a, b, and c in a and b by Ian Madin-Wachino and  
179 Josh Roering. Modified from Callahan and Wayland, 1965. See Fig. 4 for location.

## 180 **2.2. Climate and vegetation**

181  
182 Currently, Skagway experiences a subarctic maritime climate, characterized by cool summers and cold, snowy winters. C  
183 and current average summer temperatures range from 10 to 21°C, with occasional rainfall, and winter temperatures range  
184 from -12 to -1°C, with heavy snowfall, and freezing conditions (Western Regional Climate Center, 2025). High winds that  
185 funnel up the Taiya Inlet are common and for decades it was reported that Skagway's name originated from Lingít Hlingít  
186 words describing north winds (Thornton, 2010) although recent research concludes that the name derives from a contraction  
187 of 'Áa Wushigagu Ye', which translates as "the Place with Solid Core Trees" (X. Twitchell, pers. comm, 2024). Compared  
188 to other areas in SE Alaska, Skagway receives relatively low annual precipitation (mean annual precipitation is 1.1m) owing  
189 to the rain shadow imposed by the bounding coastal ranges to the south (Daly et al., 2018) (Western Regional Climate  
190 Center, 2025). Atmospheric rivers account for nearly 70% of annual rainfall in Skagway and intense precipitation associated  
191 with these phenomena occur with highest frequency and intensity from August to October (Nash et al., 2024). Skagway  
192 hosts a coastal rainforest of spruce (Picea sitchensis), pine (Pinus contorta), and cedar (Chamaecyparis nootkatensis) trees  
193 with dense underbrush at low elevations (under 1,000m above sea level) and high-alpine tundra above tree line (Harris and  
194 Farr, 1974). Historical photographs and descriptions suggest that forests covering the slopes perched above Skagway Harbor  
195 and township were disturbed timber through timber harvest and burning in the early 1900s (Wright et al., 2021).

## 197 **2.3. Historic rockfall activity**

198 Early Western descriptions of rockfall activity in Skagway tend to focus on impacts to the harbour and railroad, including a  
199 series of events in 1901 that recorded burial of the tracks near the approach to the wharf (The Daily Alaskan, 1901). The  
200 location of this event coincides with the steep rocky slopes above Skagway Harbor, which have generated numerous  
201 rockfalls since that account (Fig. 2). A study of geologic hazards in Skagway noted the striking linearity of N- and NE-  
202 oriented fjords and valleys and s and identified abundant actively eroding bedrock escarpments on a NW-facing ridgeline  
203 that runs along the eastern margin of Skagway that coincides that coincide with a zone of historic rockfall activity adjacent  
204 to the township and harbour (Yehle and Lemke, 1972). Downslope of these escarpments are colluvial deposits, consisting of  
205 landslide deposits, including talus from historic rockfall events. The abundance and extent of these deposits implies  
206 significant slope adjustment and retreat since glacial retreat and the relative activity of the deposits is based-reflected by  
207 the relative abundance or absence of vegetation cover (Ruffner and Abrams, 1998; van Steijn, 2002). Across the valley on

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208 the western side of Skagway, these talus deposits are much less prevalent, and the Yehle and Lemke (1975) maps do not  
209 indicate the presence of erosional escarpments.

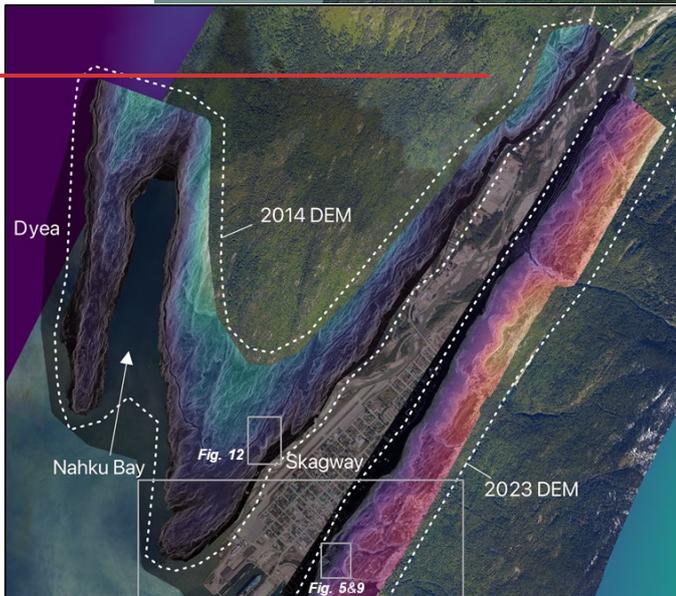
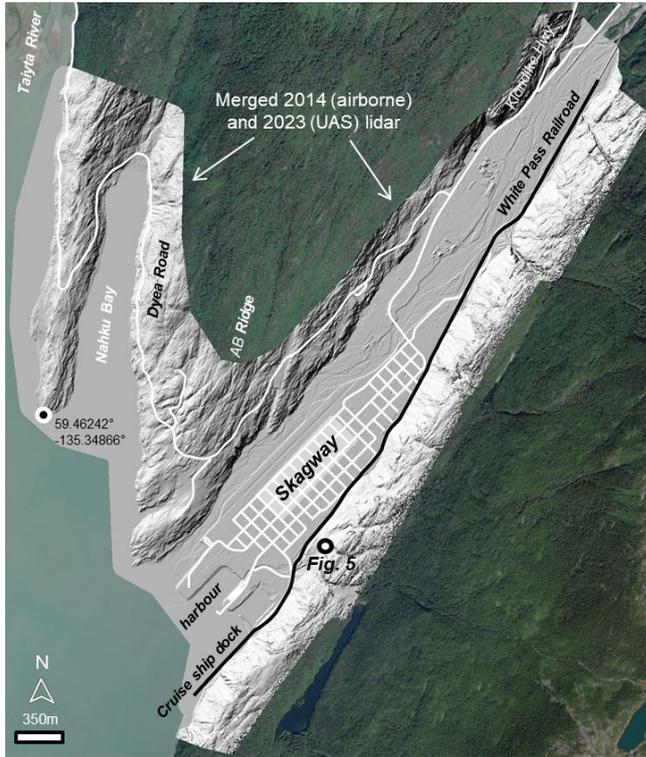
210  
211 Rockfall hazard mitigation in the area is currently focused on the active rockfall source areas above the cruise ship dock in  
212 Skagway Harbor. On June 23, 2022, rocks detached from the eastern ridgeline and impacted the cruise ship dock where  
213 pedestrian traffic is frequent (Munson, 2022b). Two more rockfall events originating in the rocky slopes above the harbour  
214 followed in rapid succession on August 3 and 5, 2022 (Munson, 2022a).

215  
216 ~~Rockfall hazard mitigation in the area is currently focused on the active rockfall source areas above the cruise ship dock in~~  
217 ~~Skagway Harbor~~. Engineering efforts have been completed, which include wrapping rock mesh covers over source areas,  
218 installing attenuator nets to block falling rocks, and scaling loose rocks (Brennan and Whistler, 2022). Instrumentation has  
219 been installed to monitor the source area, including extensometers installed at the top of the slope, which show that  
220 movement in the slide mass has increased from 4 cm per year as of 2022 (Brennan and Whistler, 2022).  
221 Notably, the small section of rocky slopes above Skagway’s cruise ship dock, ~~where the engineered mitigation and~~  
222 ~~monitoring efforts are focused~~, is a small fraction of the roughly 5-km long stretch of ridgeline that borders the eastern  
223 margin of Skagway. Rocky escarpments and talus deposits have been noted along the entire ridge (Yehle and Lemke, 1972),  
224 including the “Cemetery Slide”, a rockfall source area and runout zone stripped of vegetation by falling debris, which is  
225 similar to the active source areas above the cruise ship dock. Although these zones of localized activity are well known, ~~the~~  
226 forest cover obscures the geologic and topographic signature of past rockfall activity along the remainder of the ridge such  
227 that the pattern of relative susceptibility and runout remains ambiguous.

## 228 **3 Methods**

### 229 **3.1. Overview**

230 Motivated by renewed rockfall activity, this study seeks to identify areas susceptible to rockfall initiation and runout within  
231 the steep, post-glacial valleys around Skagway. Our analysis extends to the west of Skagway along Dyea Road to the Taiya  
232 River, which hosts the Lingit Tlingit settlement of Dyea and the Chilkoot Trailhead, ~~which is~~. Dyea Road is a well-travelled  
233 corridor that also provides the opportunity to test our methodology across a wider range of topographic and structural  
234 configurations (Fig. 4). ~~The components of our analysis include a historical rockfall inventory, synthesis of new and existing~~  
235 ~~lidar data, geomorphic and structural mapping, kinematic analysis of rockfall susceptibility, and rockfall runout modelling.~~



237 Figure 4: Location map of Skagway River valley, Nahku Bay, and Dyea Road showing a merged hillshade image of the 2014  
238 airborne survey and 2023 UAS lidar bare earth data DEM acquired for this study. White lines denote roads and the thick black  
239 line denotes a section of White Pass Railroad and cruise ship dock used for our rockfall runoff susceptibility analyses. The  
240 2023 UAS lidar data was acquired along the ridge southeast of the railroad while the 2014 airborne lidar accounted for the  
241 remainder of the study area imagery shown here. The dashed lines demarcate the areas analyzed in this study. Gray boxes  
242 denote the extent of other figures. Location of Figure 5 is shown by black/white filled dot. Background image from USGS  
243 NAIP (National Agriculture Imagery Program).

### 244

### 245 3.2 Rockfall inventory

246 To investigate controls on rockfall timing, we searched newspaper articles (primarily the Skagway News) and public  
247 announcements that describe the location and timing of rockfall events since August 26, 2017, which marks the beginning of  
248 rockfall mitigation efforts along the Skagway Harbor. In addition, we accessed data generated for rockfall events between  
249 2005 and 2022 from the GeoEvent Slope Stability Database generated by the Alaska Department of Transportation & Public  
250 Facilities (AKDOT&PF) Geotechnical Asset Management Program (Thompson, 2017). This database features  
251 includes details from AKDOT&PF maintenance and operations reports, including the location, date, event type (e.g., debris  
252 flows, rockfalls, landslides, snow avalanches, flooding), relative magnitude, and cost, of geologic slope processes events that  
253 impact the Alaska highway system. Notably, this database does not include information for activity along the White Pass  
254 Railroad that abuts the rockfall-prone ridgeline along the eastern margin of Skagway. Nonetheless, it includes 866 rockfall  
255 event reports along segments of Skagway-Dyea Road and Klondike Highway within our study area (Figure 4) that enabled  
256 us to assess climatic controls on rockfall activity. We aggregated rockfall events according to Julian day to investigate  
257 seasonal trends in activity over the 17-yr record. For climate context and comparison, we accessed Skagway Airport (PAGY)  
258 mean daily rainfall and mean daily maximum and minimum temperature data for 2005 to 2025 (Horel et al., 2002).

### 259 3.3 High-resolution topography: Airborne and UAS lidar

260 Our analysis used two sources of lidar data to inform geomorphic and bedrock mapping, kinematic analysis, and runoff  
261 modelling. One We used a 2014 airborne lidar dataset DEM acquired in 2014 with an average ground classified point  
262 density of 4.6 m<sup>-2</sup> is available from acquired by the Alaska Department of Geological and Geophysical Surveys (DGGs Staff,  
263 2013) and which includes low-elevation terrain in the Skagway River valley, Nahku Bay, and part of Dyea valley  
264 (Macpherson et al., 2014). Because this lidar acquisition does not span a significant portion of Skagway's rockfall-prone east  
265 ridge that abuts the township and harbour, we conducted a UAS lidar survey in 2023 (Fig. 4). We acquired lidar data across  
266 the 2.3 km<sup>2</sup> area with peak elevations of 220 m near the cruise ship docks and 320 m near the northern extent of the ridgeline  
267 (Roering et al., 2025). For the acquisition, the NSF RAPID facility used a Trinity F90+ fixed wing drone with a Qube 240  
268 lidar payload to fly ~120m above ground with 90% coverage overlap. The surveyed area was slightly abbreviated due to a  
269 patch of extremely steep terrain where the UAS could not be flown safely at distances sufficiently close to the ground

270 surface to acquire data. The UAS survey produced a point cloud containing 650 million total points with 200 million ground  
271 classified points, giving an average ground point density of 43 m<sup>-2</sup>. We used the ground points to create a 1 m DEM using  
272 Cloud Compare (version 2.12.3) and combined it with the 2014 airborne lidar data to provide a seamless-continuous lidar  
273 coverage for our analyses (Fig. 4).

### 274 **3.4 Geomorphic mapping**

275 We used field observations, historic photographs, and slope thresholds and surface texture from the combined lidar DEM to  
276 identify and map talus deposits that reflect the accumulation of rockfall deposits. These talus deposits include both forested  
277 and exposed-unforested occurrences. To define the characteristic slope angles associated with active talus slopes we  
278 measured slope angles from 20° to 45° that coincide with mapped talus deposits along the eastern ridge. Slopes steeper than  
279 45° tend to correspond with bedrock cliffs and outcrops, whereas slopes gentler than 20° often reflect relatively uneroded  
280 bedrock surfaces or deposits from fluvial or other mass wasting processes. The relatively smooth texture of accumulated  
281 rockfall deposits identified with hillshade and slopeshade layers (Burns and Madin, 2009) was also used to identify active  
282 talus slopes as well as bedrock cliffs or outcrops that constitute a rockfall source area. Talus deposits occur on a wide range  
283 of scales and for this analysis we focused on mapping patches of talus with area >100m<sup>2</sup> to ensure accuracy and highlight  
284 zones of significant activity (Fig. 55). Specifically, we identified terrain that exhibits the relevant slope values as well as  
285 surface textures consistent with talus deposits and then mapped polygons that enclose these zones when their areas exceeded  
286 100m<sup>2</sup>. The polygons were often separated by bedrock outcrops, gullies, and anthropogenic features.  
287

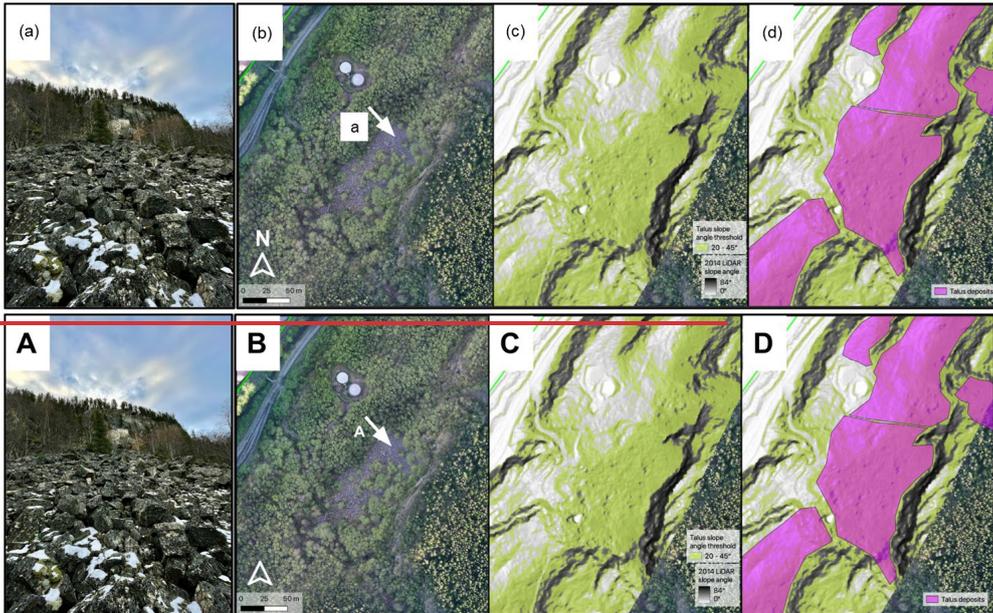


Figure 5S: Field images and lidar mapping of talus mantled slopes. See Fig. 4 for location. (a) View looking SE along an exposed talus deposit with boulder sized talus blocks at the base of a 40 m cliff known as Kirmse's Cliff (lat: 59.45251, long: -135.31147) on far left of image, (b) Aerial imagery from 2023 UAS survey showing the talus deposit and bedrock cliff (center-left), white arrow represents shows view direction for photograph location and view direction, (c) Terrain within the slope angle range of 20°-45° colored in yellowlight green, and (d) map of talus polygons (pink) generated from lidar slope shade and slope angle maps and orthoimagery. Hillshades in c and d derived from DGGs-2023 UAS lidar data. Background image from USGS NAIP (National Agriculture Imagery Program).

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### 3.5 Bedrock structure and discontinuities

To quantify the geometry of joints and discontinuities that contribute to rockfall initiation, we collected 405 structural measurements (specifically, the strike and dip of planes that define the interfaces) at a range of distinct locations nearly 50 locations within the study area and combined them to generate a comprehensive discontinuity dataset (Kundu et al., 2023). The steep, forested, and uneven terrain surrounding Skagway limits access to outcrops and the traditional 'scanline' method

(Priest and Hudson, 1981) was untenable. Furthermore, the significant forest cover and expansiveness of the study area precluded the use of ground-based laser or SfM (structure from motion) methods for acquiring structural information. Instead, we traversed the base of outcrops as an approximation of scanlines and measured the orientation of planar surfaces expressed in the outcrop with FieldMoveClino, a digital compass-clinometer smartphone app (Oliinyk et al., 2020), which enabled rapid and accurate data acquisition. Our measurements were acquired across a wide range of locations in the study area ~~in order to~~ characterize spatial variations in discontinuity orientations. In particular, we visited outcrops along the eastern, ~~southern~~, and western ~~sides-corners~~ of the ~~Skagway-River-valley study area as well as outcrops in secondary valleys perpendicular to these ridgelines~~ to ensure that the full range of relevant joint orientations ~~were was~~ represented (Terzaghi, 1965). ~~We identified joint clusters by converting the strike and dip of discontinuity planes to poles on stereonet plots and used contouring to separate different clusters~~ (Wyllie and Mah, 2004).

### 3.6 Kinematic analysis of rockfall susceptibility

To assess the spatial pattern of rockfall susceptibility across the study area, we adopted a kinematic analysis approach ~~with our discontinuity data~~ and applied criteria for planar sliding and toppling failure within our combined lidar DEM. This approach foregoes the mechanical complexity and extensive parameterization of more sophisticated analyses (e.g., Matasci et al., 2018) in order to generate results that span an extensive area. ~~Kinematic analysis requires discontinuity data that is locally representative. Traditionally, discontinuity measurements are taken in the field, whereas many modern applications extract discontinuity orientations from high-resolution point clouds of the rock slope face (Utlu et al., 2023). Field measurements remain a reliable and relevant means to capture joint orientations (Kundu et al., 2023), especially in locales where terrestrial laser scans are not feasible due to hazardous terrain or where slope faces are obscured by vegetation.~~ As described below, the remarkable consistency of discontinuity orientations across the study area inspired this approach and therefore we invoked the entire distribution of measured joint orientations at all locations to perform the kinematic analysis. Specifically, we applied the stability criteria for each failure mode at each pixel in the ~~combined-merged~~ lidar DEM by estimating the fraction of the ~~337-405~~ joint orientations that are predicted to be unstable given the topographic aspect and slope angle of that pixel. The failure criteria for planar and toppling failure are defined following Wyllie & Mah (2004) and described below. Essentially, this approach is equivalent to locally performing a stereonet analysis of rock slope failure across our study area and aggregating the results to identify potential rockfall failure modes ~~as well as~~ areas of high relative susceptibility. For the analysis, we used a friction angle of 40°, consistent with measured values for jointed granodiorite similar to Skagway's lithology (Alejano et al., 2019). ~~To assess the accuracy For validation~~ of our modelled rockfall susceptibility maps, we ~~visually~~ compared our predictions to the location of recent rockfall events in Skagway, as well as mapped ~~bedrock cliffs and talus deposits~~, which serve as a proxy for prior rockfall activity (Loye et al., 2009; Stock and Collins, 2014). ~~In addition, we compared the downslope direction of talus deposits with the aspect of topple-prone slopes to assess their relative correspondence.~~

335  
336 Planar slides occur when the inclination of a bedrock slab exceeds the friction angle and it slides along a planar  
337 discontinuity. Toppling failures occur when discontinuities steeply dipping into the rock slope face form slabs or columns of  
338 rock that rotate forward along a fixed base (Fig. 1). Two types of toppling failures can occur that are influenced by the  
339 strength of the rock mass and the geometry of discontinuities. Flexural toppling, where slabs of rock bend forward until they  
340 break in flexure, is typical in shale and slate where orthogonal jointing is not well developed. Block toppling is common in  
341 crystalline bedrock with orthogonal joint sets, where two steeply dipping joint sets form the sides of blocks, and a third set of  
342 low angle, widely spaced joints form a basal failure plane. The active rockfall source area situated above Skagway's cruise  
343 ship dock has been described as a progressive toppling failure with a stair-stepped basal feature (Brennan and Whistler,  
344 2022). Topple failures observed in the field are consistent with this description, which is described by the block toppling  
345 failure mechanism. As a result, our analysis focuses on block toppling and planar sliding given that although we also account  
346 for planar sliding given that sporadic sliding was observed in the field. F field observations and geotechnical reports do not  
347 identify wedge failures as a prevalent mechanism of rock slope instability. The challenge of implementing multiple  
348 discontinuity orientations required for wedge analyses across our entire study area also encouraged us to focus on block  
349 toppling and planar sliding a potential mechanism and we opted not to include it in our analyses.

350  
351 The criteria for planar, wedge, or toppling failure is based on the orientation of discontinuities and their orientation relative  
352 to the rock slope face (Wyllie and Mah, 2004). The dip direction of the discontinuity ( $\alpha_d$ ) and rock slope ( $\alpha_f$ ) is given by  $\alpha_d$   
353 and  $\alpha_f$ , respectively, and the dip angle of the discontinuity and rock slope, both relative to horizontal, is given by  $\psi_d$  and  $\psi_f$ ,  
354 respectively. The friction angle of the joint interfaces is given by  $\phi$ .

355  
356 Accordingly, a rock slope is susceptible to planar sliding failure along a discontinuity if the following three conditions are  
357 simultaneously met:

$$|\alpha_d - \alpha_f| < 20^\circ ; \psi_d < \psi_f ; \psi_d > \phi \quad (1)$$

359 The first condition requires that the interfaces are aligned in a sufficiently similar orientation (in this case within  $20^\circ$ ) while  
360 the second ensures that the discontinuity dip is shallower than the angle of the bedrock slope such that the discontinuity  
361 intersects or 'daylights'. The third and final condition demands that the interface slope angle exceeds the angle of friction. All  
362 three conditions must be met at a given location for planar sliding to be deemed likely.

363  
364 For block toppling failure, the following two conditions must be met simultaneously:

$$\alpha_f - 20^\circ < (\alpha_d \pm 180^\circ) < \alpha_f + 20^\circ ; (90^\circ - \psi_f) + \phi < \psi_d \quad (2)$$

366 The first condition asserts that the discontinuity must dip into the rock slope face and be parallel, or nearly parallel (e.g., within  
367 20°), to the dip direction of the slope face. The second condition indicates that the discontinuity dip must exceed the friction  
368 angle allowing for interlayer slip between the blocks. In our model, the maximum allowable dip direction deviation is  $\pm 20^\circ$   
369 for both planar slide and block toppling failure. Although this value ~~is sometimes was often~~ chosen to be  $\pm 10^\circ$  for block  
370 toppling we expanded the constraint to  $\pm 20^\circ$ , ~~consistent with recent contributions~~ (Cruden, 1989; Gigli et al., 2022).

371

372 ~~Kinematic analysis requires discontinuity data that is locally representative. Traditionally, discontinuity measurements are~~  
373 ~~taken in the field, whereas many modern applications extract discontinuity orientations from high-resolution point clouds of~~  
374 ~~the rock slope face (Utlu et al., 2023). Field measurements remain a reliable and relevant means to capture joint orientations~~  
375 ~~(Kundu et al., 2023), especially in locales where terrestrial laser scans are not feasible due to hazardous terrain or where~~  
376 ~~slope faces are obscured by vegetation.~~

377

378 Using equations 1 and 2, we estimated the number of joints in our field-derived dataset ( $n=405337$ ) that are predicted to  
379 exhibit planar and toppling failure, respectively, for each pixel in our ~~combined-merged~~ lidar DEM. ~~Also, by simultaneously~~  
380 ~~accounting for both modes of potential failure at a given location it is possible to assess the combined susceptibility to~~  
381 ~~toppling and planar sliding although we do not formally account for this possibility in our analysis.~~ Our maps of planar and  
382 toppling failure are then calculated as the percentage of joints that meet the conditions required for failure conditions ~~at each~~  
383 ~~pixel in our study area.~~ For example, a toppling failure index value of ~~0.2134~~ for a given pixel in our DEM indicates that  
384 ~~231%~~ of the joints in our joint dataset satisfy the two conditions in equation 2. This approach provides a description of  
385 relative rockfall initiation susceptibility across our study area.

### 386 3.7 Rockfall runout modelling with RAMMS

387 To model potential runout paths associated with rock slopes that have high rockfall susceptibility, as determined by our  
388 kinematic analyses, we used the 3D rockfall simulation software RAMMS:Rockfall (<https://ramms.ch/ramms-rockfall/>) to  
389 represent the sliding, bouncing, and rolling motion of ~~block/rock-elasts as well as the internal, gravitational, and contact~~  
390 ~~forces of a rigid body that can translate and rotate.~~ This model accounts for the energy balance of falling rocks and has been  
391 used extensively at a range of scales for both applied and fundamental research studies (e.g., Caviezel et al., 2021; Kleinn et  
392 al., 2024; Leine et al., 2014; Lu et al., 2019; Ringenbach et al., 2023). The primary inputs required for the RAMMS:Rockfall  
393 model include digital elevation data (i.e., DEM), the location of rockfall source areas, specification of ground cover, and the  
394 shape and size of the falling blocks.

395

396 We used our ~~combined-merged~~ lidar DEM, which has a 1x1 m pixels and spans the area shown in Fig. 4. To identify  
397 potential source areas, we identified pixels in our maps of toppling susceptibility with values greater than 5%. Pixels with  
398  $>5\%$  toppling susceptibility demarcate rocky cliffs situated above our mapped talus deposits as well as areas of recent

399 rockfall activity in the Skagway Harbor (Fig. 5). The pixels with a >5% high toppling susceptibility were converted to  
400 polygons and only polygons with area greater than ~~25-25~~  $m^2$  were retained ~~in order to~~ eliminate local high-relief features  
401 like boulders and trees that can perpetrate the signature of rockfall source zones. Rockfall source points for RAMMS  
402 modelling were randomly distributed across the ~~resulting~~ polygon ~~coverages~~ with a density of ~~0.02-04~~  $m^2$  and a minimum  
403 point spacing of 5 m. This methodology follows convention used in other studies (Lu et al., 2021) and yielded ~~nearly~~  
404 ~~5,0004,945~~ rockfall initiation points for ~~each of our RAMMS-simulations-simulation scenarios~~.

405  
406 The slopes east of Skagway are heavily forested, except in locations where falling debris has stripped vegetation, such as the  
407 cruise ship dock and at the northern extent of the eastern ridge bordering Skagway. In RAMMS, we represented forested  
408 areas as spruce alpine forests and trees were simulated in these areas using the 'dense forest' category in RAMMS, which is  
409 defined by a stem density of 600 trees per hectare with a mean diameter of 30 cm. These parameters were chosen based on  
410 our field observations and the typical density of mixed red alder coniferous stands in Southeast Alaska (Poage et al., 2007).  
411 The effect of trees in the RAMMS rockfall runout module is to attenuate energy and reduce velocity, thus constituting a  
412 significant impact on hazard potential. ~~We mapped the distribution of trees by visually assessing the USGS NAIP imagery as~~  
413 ~~well as our UAS orthoimagery~~. The ~~very~~-small amount of terrain (less than 5% of the study area) that exists outside these  
414 forested areas was set to the fine talus ~~cover~~ category in RAMMS, which reflects negligible ground cover atop relatively  
415 fine-grained talus deposits ~~consistent with our field observations~~. We performed simulations with ~~two land cover scenarios:~~  
416 ~~1) and without dense forest forest cover and 2) fine talus cover, assigned across the entire simulation area in order to~~ assess  
417 the potential role of timber harvest and fire on rockfall runout and provide a conservative assessment of the hazard extent.

418  
419 We estimated representative block size and shape by measuring ~~the major axes of randomly selected~~ blocks in exposed talus  
420 piles (Hales and Roering, 2007). To estimate ~~representative~~ block sizes, the intermediate axis of 74 blocks was measured  
421 from ~~field measurements and~~ high-resolution point clouds combined with orthoimagery. We determined the mean  
422 intermediate axis size to be 1.08 m while the 50<sup>th</sup> percentile was 0.66 m and the 95<sup>th</sup> percentile was 3.56 m. Blocks observed  
423 in the field are typically tabular and the ~~characteristic shape-geometry~~ of blocks was estimated by measuring the long,  
424 intermediate, and short axes of 10 blocks with sufficient exposure to allow measurement using our UAS-derived point  
425 clouds. ~~The block dimensions were measured and w~~We applied the ~~resulting-characteristic~~ aspect ratio to the 50<sup>th</sup> percentile  
426 and 95<sup>th</sup> percentile intermediate axis values from our ~~block-size~~ measurements. The dimensions of the resulting rocks used in  
427 the simulation were 0.94 m x 0.66 m x 0.37 m (medium, 50<sup>th</sup> percentile), and 5.10 m x 3.56 m x 1.99 m (large, 95<sup>th</sup>  
428 percentile), which represent moderate and large ~~elast-block~~ sizes, respectively. Given the tendency for large ~~elasts-blocks~~  
429 to travel longer distances, our 95<sup>th</sup> percentile blocks are intended to reflect the conservative (or long runout) scenario (Statham,  
430 1976). ~~For these simulations, we assumed rock density of 2,700 kg m<sup>-3</sup>~~.

432 To simulate the range of potential rockfall impacts, we focused on four scenarios for RAMMS simulation with the following  
433 parameterizations: 1) 50<sup>th</sup> percentile ~~elasts-block~~ with ~~fine no-forest~~ talus cover, 2) 50<sup>th</sup> percentile ~~elasts-block~~ with dense  
434 forest, 3) 95<sup>th</sup> percentile ~~elasts-blocks~~ with ~~fine no-forest~~ talus cover, and 4) 50<sup>th</sup> percentile ~~elasts-block~~ with dense forest. To  
435 account for the stochastic nature of rockfall release, we used RAMMS to select 1 of 10 randomly chosen ~~roek-elast-block~~  
436 orientations to be released at each source point, yielding ~~nearly 50,000~~ 49,450 individual rockfall runout paths ~~in-for~~ each of  
437 the four ~~simulationsscenarios~~. Our results presented here focus on the two end-member scenarios (50<sup>th</sup> percentile with dense  
438 forest cover and 95<sup>th</sup> percentile with fine-grained talus over) to illustrate the range of rockfall behaviour.

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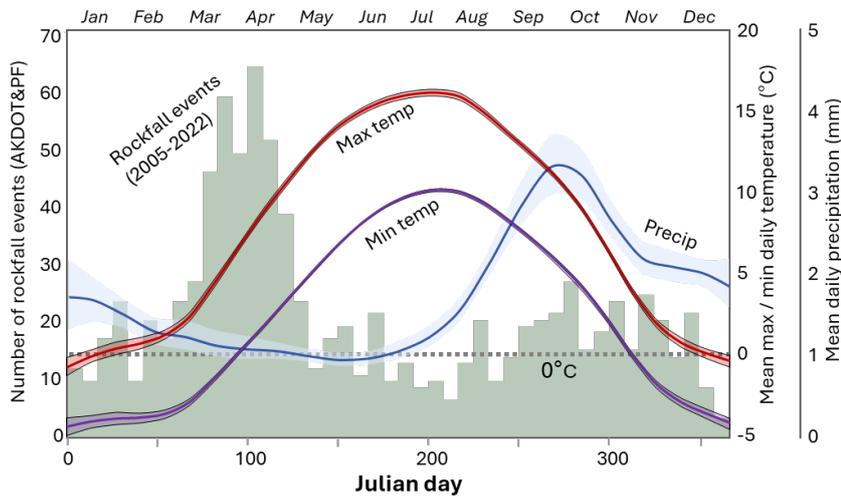
440 For our runout analysis, we focused on the eastern NW-facing ridgeline of the Skagway River valley ~~to assess potential~~ given  
441 ~~the need to assess~~ impacts to the harbour and township. ~~Because each simulated runout event results in an individual~~  
442 ~~rockfall path, it can be difficult to effectively visualize the cumulative pattern of predicted runout~~ We used an array of  
443 ~~analyses to investigate rockfall runout patterns and topographic controls on the runout behaviour.~~ First, ~~t~~o identify terrain  
444 with high likelihood of rockfall runout we used RAMMS to ~~count~~ calculate the number of rockfall events that traversed each  
445 pixel in our ~~model~~ domain for each of the ~~four~~ scenarios. As such, the cumulative number of rockfall passages at each pixel  
446 accounts for both the abundance of upslope source areas as well as the tendency for topography to steer or direct rockfall  
447 into particular pathways. ~~Second, In addition,~~ we used RAMMS to create a “digital” boundary (or barrier) coincident with  
448 the railroad tracks along the base of the ridgeline to record ~~the number and kinetic energy of modelled~~ rockfalls that bypass  
449 the boundary and ~~have the potential to~~ impact the harbour and township. ~~Third, to estimate how topographic properties~~  
450 ~~moderate rockfall runout, we calculated local slope and terrain roughness across the simulation domain. Roughness was~~  
451 ~~calculated as the standard deviation of local slope values within a 5-m radius window. This approach has been successfully~~  
452 ~~used to differentiate the kinematics and age of landslide deposits~~ (Booth et al., 2017; LaHusen et al., 2020) ~~and in this case,~~  
453 ~~we choose the 5-m scale to be consistent with the scale of blocks, fractures, and bedrock landforms. Finally, we generated a~~  
454 ~~5.4-km long transect coincident with the railroad within our simulation domain to quantify along-ridge variations in rockfall~~  
455 ~~passages, upslope source areas, slope angle, and terrain roughness.~~

## 456 4 Results

### 457 4.1 Rockfall inventory

458 ~~Descriptions of rockfall events before 2005 can be found in newspaper articles dating back over a century, although we focus~~  
459 ~~on recent events in this contribution. We identified 11 reported rockfall events reported in the Skagway News since 2017~~  
460 ~~(Table 1). These reports tended to reflect sporadic rockfalls along Dyea Road as well as activity in 2022 abutting Skagway~~  
461 ~~Harbor on the eastern ridgeline. Many reports are recorded in the police blotter section of the Skagway News, but the timing~~  
462 ~~and location of these events is sometimes unclear and those cases were not included in our rockfall inventory. The~~  
463 AKDOT & PF GeoEvent database includes ~~536 866~~ reports of rockfall-related maintenance and operation activities ~~in our~~

464 study area along the 10.3-km long segment of Skagway-Dyea and Klondike Hwy within our study area between 2005 and  
 465 2022 (Fig. 6.7). Notably, this database does not include events along the eastern ridgeline in Skagway because those events  
 466 do not impact the state highway system. Rather, the vast majority (>415) of the AKDOT reports originate from the NW-  
 467 facing sections of Dyea Road west of Skagway. Among those events, 7 resulted in road closures that lasted 3 days or longer  
 468 in March 2012, October 2012, January 2014, March 2015, February 2016, September 2016, and December 2020. This  
 469 implies an average rockfall frequency of 50 events per year, although this value is an overestimate because some of the  
 470 reports are redundant given that multiple days are reported in response to a single event. Among the events, 7 resulted in  
 471 road closures that lasted 3 days or longer in March 2012, October 2012, January 2014, March 2015, February 2016,  
 472 September 2016, and December 2020.



474 Figure 6. Analysis of seasonal variation in rockfall activity and climate variables from the AKDOT&PF geoevent database  
 475 (2005-2022) and daily mean precipitation and daily maximum and minimum temperature from the Skagway Airport  
 476 (PAGY) for the period 2005-2025. The number of rockfall events (green bars) exhibits seasonal variations that don't  
 477 obviously align with precipitation (blue) although maximum temperature (red) and minimum temperature (purple) increase  
 478 significantly during the typical period of rockfall activity in March and April. The horizontal dashed grey line represents  
 479 freezing. The confidence intervals for the climate data represent standard error of mean values over the period of record.

481 Table 1: Rockfall events reported in the Skagway News since 2017

Event date	Event time (local)	Pre-event 24-hr rainfall total (inches)	Location
9-Oct-2023	n/a	0.63	East Dyea Valley
7-Oct-2023	n/a	1.46	Eas Nahku Bay
18-May-2023	6:17pm	0	North Slide
29-Sept-2022	3:57pm	0.50	East Nahku Bay
5-Aug-2022	n/a	0.04	North Slide
3-Aug-2022	5:00pm	0.02	North Slide
23-Jun-2022	7:30am	Trace	South Slide
2-Dec-2020	12:23pm	3.24	East Nahku Bay
5-Sept-2017	5:30am	0.76	North Slide
5-Sept-2017	3:00am	1.05	North Slide
26-Aug-2017	6:30am	0.55	North Slide

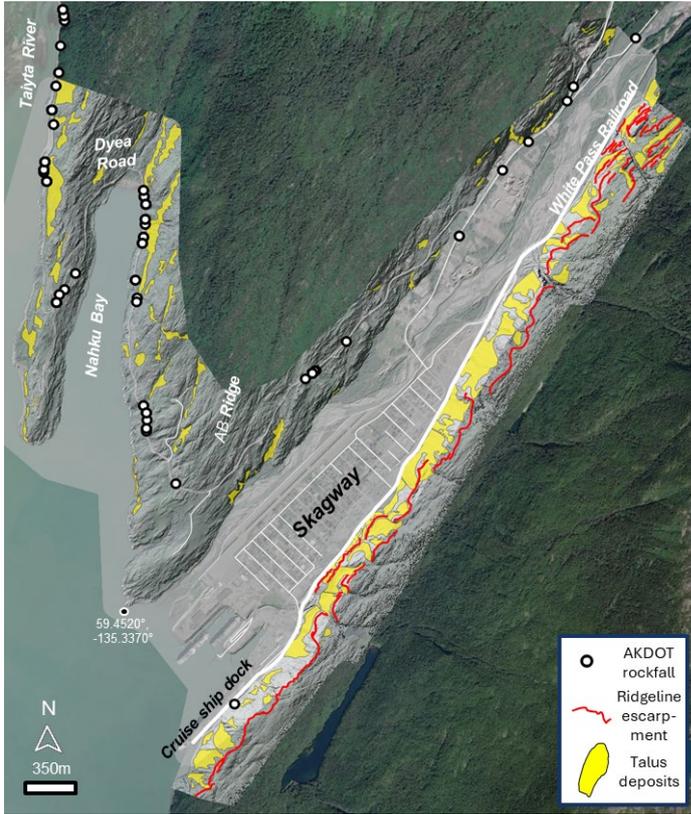
482 According to our analysis, rockfall events peak in late March and April, occurring at a pace that is 4x greater than the  
483 remainder of the year (Fig. 6). This substantial uptick in activity corresponds with a seasonal increase in daily temperatures,  
484 particularly when the minimum daily temperature exceeds zero. This implies that thawing of frozen rock slopes promotes the  
485 destabilization of blocks through the loss of tensile strength and increases in pore pressure from melting ice and snow. In the  
486 fall, increased daily rainfall associated with atmospheric rivers with a detectable but small increase in activity. The mid-  
487 summer period of warm temperatures and dry conditions corresponds with the least active period.  
488 Rockfall events reported in the newspaper and AKDOT database occur sporadically throughout the year with most activity  
489 in the summer and fall months. Some notable high impact rockfall events correspond with intense rainfall events although  
490 most events do not coincide with an obvious climatic trigger. Rockfall activity on including December 2, 2020, was when  
491 widespread events were preceded by over 3 inches of rainfall in the previous 24 hours. By contrast, a May 2023 event, which  
492 we observed in the field, initiated on a clear, sunny day with no precipitation in the 24 hours leading up to the event. In  
493 addition, the August 2022 events that impacted Skagway Harbor were not preceded by notable rainfall. Thus, while  
494 precipitation plays a role in initiating some rockfall events in Skagway other triggers maybe relevant making prediction  
495 difficult.

#### 496 4.2 Geomorphic mapping

497 The morphology of glaciated valleys around Skagway is variable owing to bedrock structure, differential glacial erosion, and  
498 post-glacial landscape evolution (Fig. 4). These factors generate systematic variations in the abundance of bedrock cliffs that  
499 serve as rockfall source areas as well as long, steep slopes that facilitate long rockfall runout. On the east side of the lower  
500 Skagway River valley, our lidar DEM and field observations reveal a distinct northwest-facing bedrock escarpment that runs  
501 parallel just below the crest of the ridgeline (Fig. 7+6). This feature is particularly distinct above the cruise ship dock,  
502 where the ridgeline has high relief compared with sections to the north that abut the township. A similar high relief  
503 escarpment also emerges on the same ridgeline near the northern extent of the township. Below these bedrock escarpments,  
504 we observe abundant talus deposits that extend continuously downslope to the base of the ridgeline and the valley floor (Fig.  
505 76). An undated historic photograph which postdates the 1899 construction of the current Museum and City Hall depicts the

506 eastern ridgeline in a state of minimal forest cover such that the escarpment and talus slopes above the harbour are clearly  
507 visible (Fig. 2b). These observations imply substantial post-glacial erosion through lateral (southeastward) retreat of the  
508 ridgeline as talus slopes convey bedrock downslope creating long and relatively smooth pathways for rockfall runoff. In  
509 ~~Figure 2b~~On that image, the ridgeline slopes just north (on the left side of the image) exhibits gentler slope angles ( $<20^\circ$ ) and  
510 a benchy morphology which implies less extensive post-glacial erosion and slope modification via rockfalls (~~Fig. 2b~~). Atop  
511 the ridge~~line~~ and east of the escarpment, the ridgeline contains abundant evidence of unmodified glacial erosion features  
512 (Figure 4).

513  
514 By contrast, ~~t~~the west side of the ~~lower~~ Skagway River valley has a very different morphology, in that it lacks a distinct  
515 escarpment and instead exhibits consistent and gradual slopes that imply minimal post-glacial modification. Locally, we  
516 observe vertical cliffs along the Skagway River. ~~Otherwise but otherwise~~, the ~~topography on the~~ west side of the ~~lower~~  
517 Skagway River valley ~~primarily~~ consists of gentle rock slopes that form prominent ridges ~~that and~~ parallel the strike of the  
518 valley. We observe a similar pattern of bedrock escarpments and talus deposits along ~~the NW-facing~~northwest-facing slopes  
519 of two parallel ridgelines between Skagway and ~~Dyea~~the Taiya River (Fig. 67).



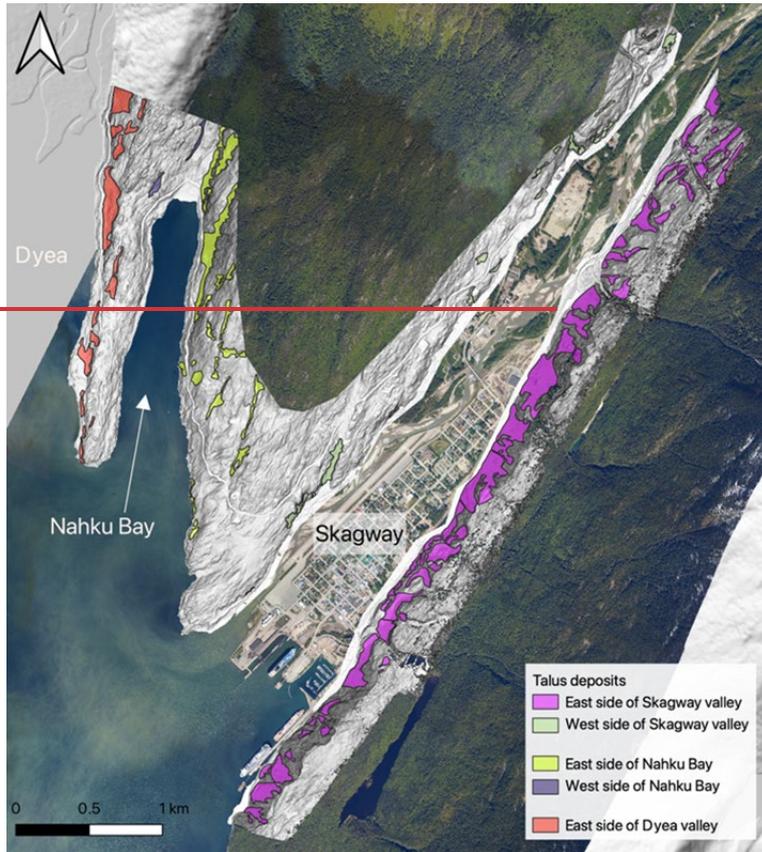


Figure 67: Distribution of talus deposits (yellow polygons), AKDOT&PF rockfall geocvents (2005-2022) (white/black dots), and Eastern ridge bedrock escarpment (red lines) in study area. Polygons identified by their location: Skagway River valley, Nahku Bay, and Dyea valley. Note the abundance of talus deposits on the NW-facing ridgeline in Skagway and on NW-facing slopes in Nahku Bay and near along Dyea Road. Hillshade derived from DGGS lidar data. Background image from USGS NAIP (National Agriculture Imagery Program). Merged lidar hillshade image has 50% transparency.

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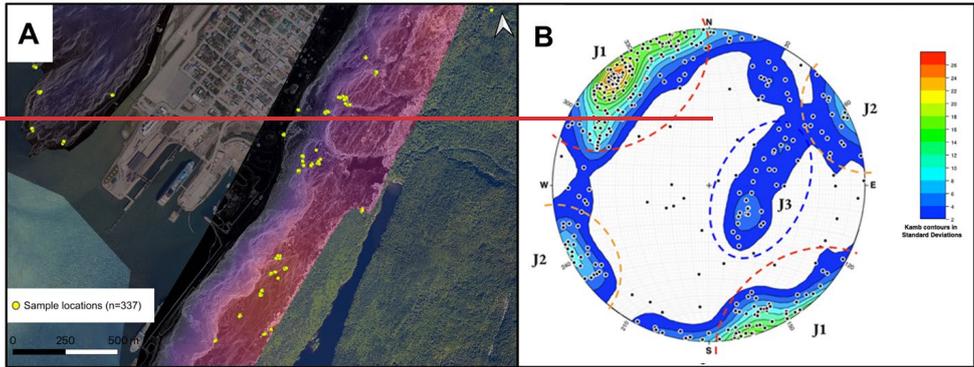
531 Our lidar- and field-derived mapping revealed abundant talus slopes on northwest-facing ridgelines across our study area.  
532 ~~More~~ More generally, talus deposits compose 12% of the 7.2 km<sup>2</sup> total mapped area (Fig. 7). Along the eastern ridgeline, we  
533 identified ~~abundant~~ 56 talus slopes with an average area of 9,290 m<sup>2</sup>, which collectively make up 250% of that ridgeline area.  
534 On the west side of the ~~lower~~ Skagway River valley, the southeast-facing ridgeline hosts ~~20 talus deposits, and these talus~~  
535 ~~slopes~~ sparse talus cover that constitutes only 3% of the 1.67 km<sup>2</sup> mapped area on that side of the valley. Similarly, talus  
536 deposits on the southeast-facing ridgelines bordering Nahku Bay are much less abundant (1%) and smaller in area than on the  
537 northwest-facing ridgeline that abuts the bay (11%). In summary, talus slopes are significantly more abundant on NW-facing  
538 slopes across our study area and their location tends to correspond with AKDOT&PF geoevents along Dyea Road (Fig. 7).

#### 539 4.3 Bedrock structure and discontinuities

540 To characterize the geometry of discontinuities with the potential to generate rockfalls, we collected ~~337~~ 405 joint  
541 orientations from ~~4836~~ granodiorite outcrops across the study area (~~Supplemental material~~) (Fig. 8a, supplemental materials)  
542 and plotted the data as poles to planes ~~on an equal-area stereonet on an equal area stereonet~~ to identify dense clusters of poles  
543 which were then grouped into joint sets (Fig. ~~78b-c~~). ~~Three~~ Two dominant joint sets were identified in this survey, ~~two~~  
544 ~~steeply dipping orthogonal sets (J1 & J2), which are steeply dipping and approximately orthogonal (Fig. 8c), and one set that~~  
545 ~~dips gently to the west (J3).~~ The steeply dipping joints are of relevance, as they tend to be conducive to toppling, which is  
546 the most observed failure mode in Skagway. The most densely defined joint set is J1, which parallels the strike of the lower  
547 Skagway River valley and the eastern ridgeline and predominantly dips to the southeast. The less densely defined joint set  
548 (J2), is approximately orthogonal to J1, has near vertical dips, and strikes northwest.

549  
550 We subsampled the joint orientation data in three geographic zones at the corners of our study area in order to assess the  
551 consistency of the rock structure. Joint orientations from the western portion of the study area are remarkably similar to those  
552 observed in the eastern and southern sections of our study area (Fig. 8b-d). This finding supports our use of the same  
553 bedrock structure data for kinematic analysis across the entire study area. The third, gently dipping joint set (J3), are  
554 interpreted as sheeting joints, typical in plutonic rocks, which have significantly higher curvature at the outcrop scale than  
555 the steeply dipping orthogonal joints resulting in an elongate field of poles on the stereonet.

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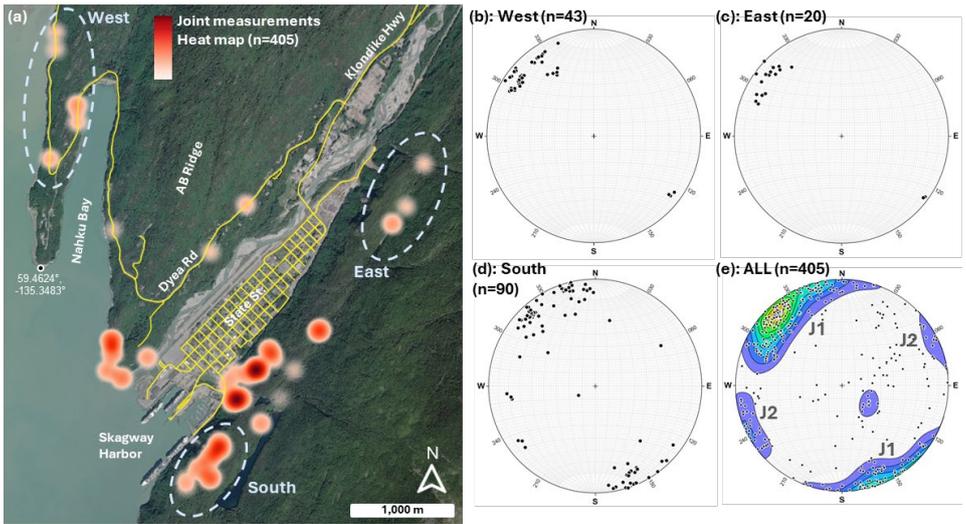
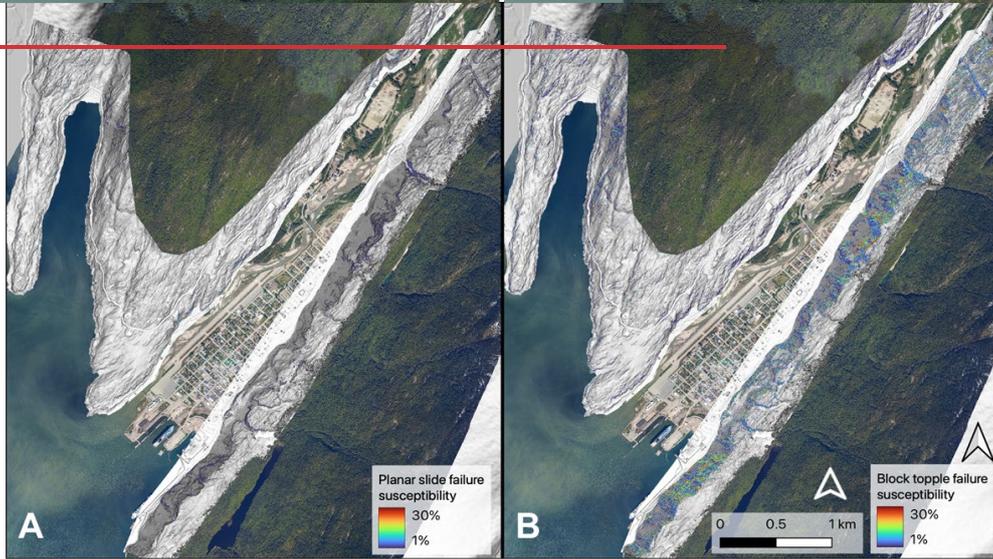
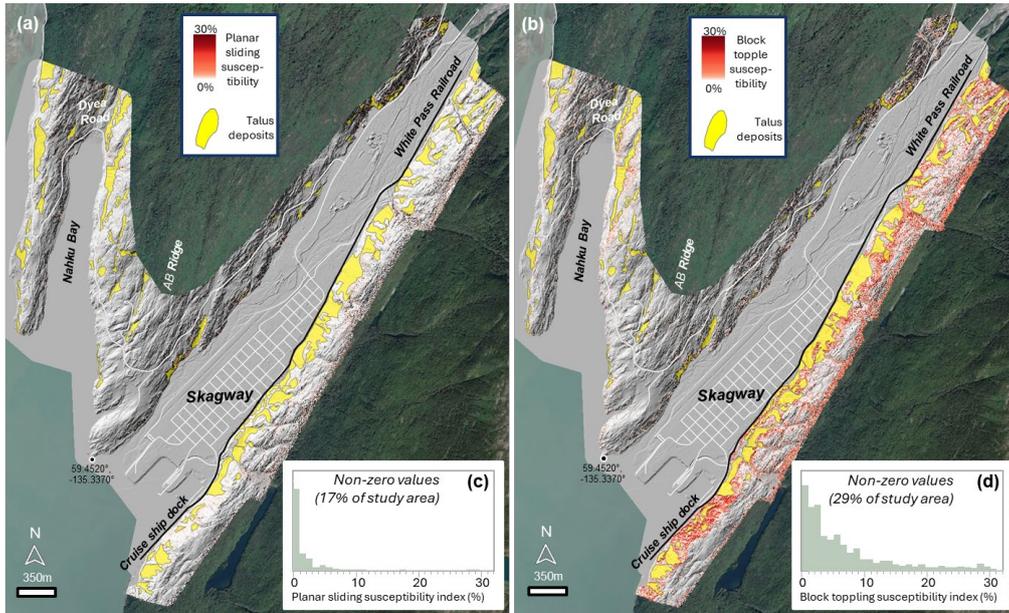


Figure 78: Structural Joint survey data. (a) Location of 337-405 structural measurements from 36-48 outcrop locations distributed across Skagway Valley and the AB ridgeline separating Skagway from Nahku Bay with heat map showing relative abundance and dashed ellipses showing subsampled data shown in panels b-d. See Fig. 4 for location. (b-c) Lower hemisphere equal area stereonet plots of joint measurements visualized as poles to planes for West (b), East (c), South (d), and ALL joint data (e). The plot in e shows (n=337) with Kamb contours to highlight dense clusters of poles (warm colors reflect high density and cool colors convey low density) and, which were grouped into three joint sets (J1, J2, and J3) two dominant joint sets (J1 and J2) are labeled. Background image in (a) from USGS NAIP (National Agriculture Imagery Program).

#### 567 4.4 Kinematic analysis of rockfall susceptibility

568 ~~Finally~~By combining equations 1 and 2 with our joint orientation data (Fig. 8) and merged lidar dataset (Fig. 4), we mapped  
 569 ~~susceptibility to planar sliding and block toppling, respectively, across our study area (Fig. 9a,b).~~ Terrain susceptible to  
 570 ~~observe small areas of terrain with planar failure is sparse and 83% of the terrain has zero values of susceptibility such that~~  
 571 ~~none of the measured joint orientations satisfy the planar sliding criteria. Furthermore, the terrain with non-zero planar~~  
 572 ~~sliding susceptibility values has very low values with a small fraction exceeding 5% (Fig. 9c)~~susceptibility according to  
 573 ~~equation 1 and in these patches, less than 5% of the joints are predicted to promote planar sliding. By contrast, our analysis~~  
 574 ~~demonstrates that 29% of the terrain has non-zero values of block toppling susceptibility and a substantial fraction of the~~  
 575 ~~non-zero values exceed 5% (Fig. 9d). This result emerges because t~~The abundance of consistently oriented joints with  
 576 vertical or sub-vertical dips ~~suggests that rock promotes block toppling is the dominant rockfall mechanism in Skagway,~~  
 577 which is supported by field observations. Furthermore, given the consistent orientation of sub-vertical joints ~~across our study~~  
 578 ~~area (Fig. 8), the primary control on rockfall susceptibility to toppling is the orientation and inclination of rock faces~~  
 579 ~~according to equation 2 (equation 2). On~~Along the ~~crest of the eastern ridgeline in~~ridge that borders the ~~eastern lower~~  
 580 ~~S~~margin of Skagway River valley, for example, the glacially flattened bench at the ~~ridge crest top of the slope~~ is not steep  
 581 enough ( $<20^\circ$ ) to meet topographic conditions for either failure mode (Fig. 9a,b) (Fig. 5). ~~As one moves~~Moving to the west  
 582 side of the crest, however, the gentle ridgetop abruptly transitions to the steep escarpment where overhanging, cliffy bedrock  
 583 slopes are ~~perched above extensive talus deposits observed (Fig. 7). At this position on the ridgeline, in these locations,~~ our  
 584 analyses show ~~that large patches of terrain have that~~ a substantial portion ( $>20\%$ ) of ~~discontinuities observed joints that~~  
 585 promote toppling failure ~~according to equation 2 (see deep red colours in Fig. 9b)8). At downslope locations (i.e., between~~  
 586 ~~the ridgeline and valley floor)~~Furthermore downslope, we observe ~~rock slopes continue to exhibit~~ patchy zones of ~~high~~  
 587 toppling susceptibility, many of which are in close proximity to Skagway township and harbour. Along the ~~southeast-facing~~  
 588 rock slopes ~~on that border~~ the west ~~side side~~ of Skagway, our analyses ~~reveal show fewer~~ and smaller patches of ~~terrain with~~  
 589 ~~toppling terrain susceptibility ble to toppling with approximating less than 5-10%, indicating low susceptibility of the~~  
 590 ~~discontinuities predicted to be unstable. This asymmetric susceptibility pattern continues. We observe a similar pattern~~ along  
 591 the sub-parallel ridges west of Skagway along Dyea Road with ~~west- and~~ northwest-facing ridgelines exhibiting abundant  
 592 patches of terrain with high propensity for toppling failure ~~(Fig. 8) and east- and southeast-facing slopes showing negligible~~  
 593 ~~toppling susceptibility (Fig. 9b). Finally, we observe small areas of terrain with planar failure susceptibility according to~~  
 594 ~~equation 1 and in these patches, less than 5% of the joints are predicted to promote planar sliding. More generally, by~~  
 595 combining our talus ~~slope deposit maps~~ with our ~~toppling failure susceptibility map~~le zones, we note a strong  
 596 correspondence such that zones with  $>5\%$  toppling failure ~~commonly tend to~~ occur just upslope of talus-mantled slopes (Fig.  
 597 9b9). Furthermore, our maps show that zones with values of toppling susceptibility less than 5% tend to be more isolated and

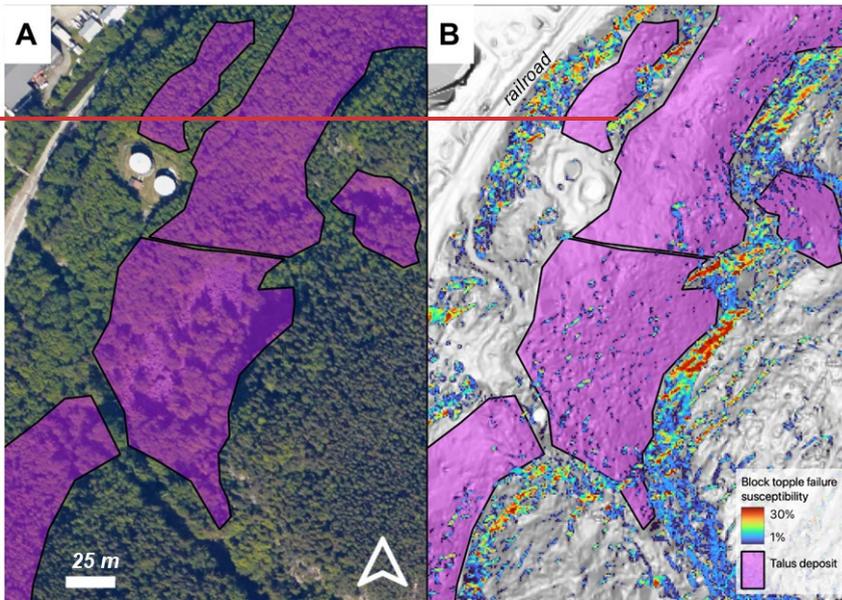
598 less clearly connected with talus deposits. As a result, the 5% toppling susceptibility index value serves as an effective  
599 threshold for defining rockfall source areas.



600

601 Figure 89: Maps of (a) rockfall-planar sliding and (b) block toppling susceptibility indices using combined 2014  
602 airborne and 2023 UAS lidar datasets and joint orientation data. Histograms of (c) planar sliding and (d) block toppling  
603 susceptibility indices for non-zero values. (a) Susceptibility to planar slide failure according to equation 1, and (b)  
604 susceptibility to block toppling according to equation 2. Both indices The mapped and plotted values are estimated as the  
605 percentage of field-surveyed joints deemed unstable at each pixel according to equations 1 (planar sliding) and 2 (block  
606 toppling). Yellow polygons denote talus deposits. Note the abundance of toppling failure zones on NW-facing rock slopes  
607 that border Skagway and Nahku Bay as well as the correspondence between toppling susceptibility and talus deposits.  
608 Hillshade imagery in a and b derived from DGGS combined 2014 and 2023 lidar data. Background image from USGS  
609 NAIP (National Agriculture Imagery Program).

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612  
613 Figure 9: Map of talus deposits and block toppling susceptibility index (percentage of joints deemed unstable in each pixel  
614 according to equation 2). See Fig. 4 for location. (a) Orthoimage of Kirmse's cliff and the partially forested talus deposit at the  
615 base of the slope, (b) Block toppling susceptibility index. Note the abundance of susceptible toppling areas perched above the

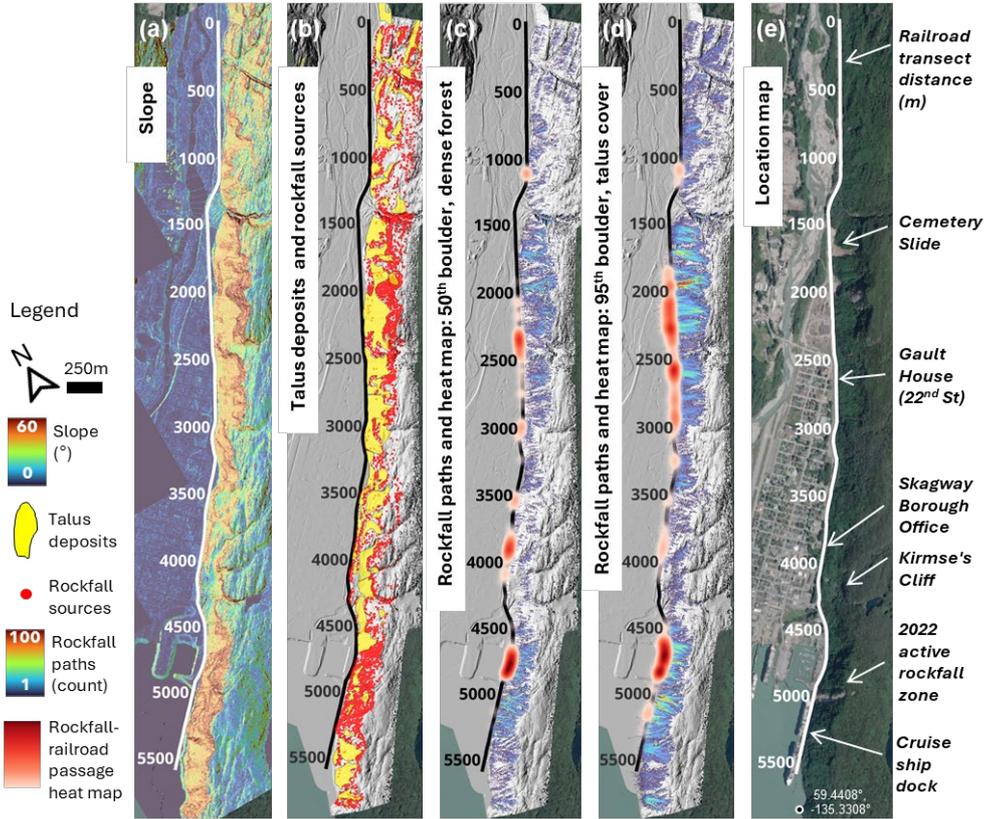
616 talus deposits. Background image in (a) from USGS NAIP (National Agriculture Imagery Program). Hillshade in (b) derived  
617 from DGGS lidar data.

#### 619 4.5 Rockfall runoff modelling using RAMMS

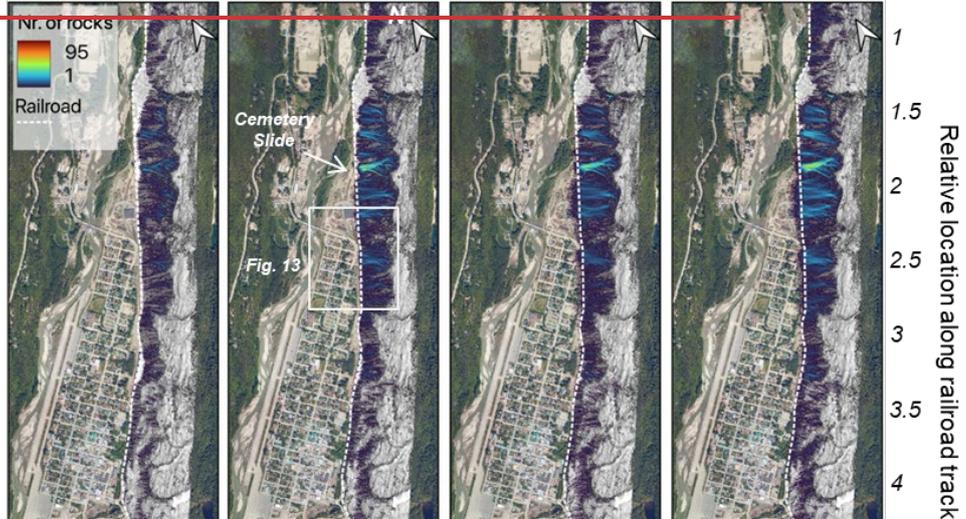
620 ~~Using~~By using zones of high toppling susceptibility (>5%) as source areas for rockfall initiation, we ~~used~~ applied RAMMS  
621 to model the runoff of ~~nearly 49,450~~50,000 rockfall events for each of the four scenarios that account for differences in ~~elast~~  
622 block size and ground cover. We applied the runoff model along the ridge bordering the eastern margin of Skagway and  
623 used the 5.4-km long section of the White Pass Railroad extended along the cruise ship dock to demarcate rockfall passages  
624 with potential impact to the harbour and township (Fig. 10a-e). The ridge is steeper in the southern (4500 to 5500 m on the  
625 railroad transect) and central (1500 to 3000 on the transect) sections where extensive bedrock cliffs protrude above talus  
626 deposits (Fig. 10a). These cliffs are identified as zones of high toppling susceptibility and thus designated as rockfall source  
627 areas (Fig. 10b). Our simulations showing the ~~total~~ number of rockfall events ~~traversing that pass through~~ each pixel  
628 reveal distinct zones subject to high rockfall susceptibility as well as extensive downslope transport (Fig. 10c,d). For both the  
629 50<sup>th</sup> block/dense forest and 95<sup>th</sup> block/fine talus scenarios shown here, in each of the four scenarios, the southern end of the  
630 ridgeline above the harbour (between 4 and 5 km 4500 and 5500 on our railroad-based the transect) exhibits ~~~10~~>10-specific  
631 chutes or paths of ~~likely-efficient~~ rockfall runoff whereby initiation near the escarpment results in the concentration of  
632 rockfall runoff along these paths and conveyance to the cruise ship docks and/or harbour (Fig. 10). Both scenarios exhibit a  
633 zone of events that pass across the railroad as depicted by the heat map (Fig. 10c,d). This zone coincides with the area of high-  
634 relief terrain and extensive escarpment development. In contrast, the central portion of the ridgeline (between 2.8 and  
635 4 km between 3000 and 4500 m on the transect) immediately adjacent to much of the township exhibits patchy and less  
636 frequent rockfall activity along gentle and irregularly oriented bedrock steps and benches. Further north, our simulations  
637 again reveal abundant long rockfall runoff paths at the northern extent of the township with a ~~concentration~~ concentrated  
638 activity near the Cemetery Slide area (between 1 and 2 km 500 and 3000 m on the transect) with a high concentration of  
639 blocks predicted to pass across the railroad (Fig. 10d). Finally, at the northernmost extent of our simulation domain (from  
640 near 1 km 0 to 1500 m on the transect), the results show patchy and short rockfall transport paths coincident with gentle,  
641 benchy topography. Taken together, these simulations demonstrate a large fraction of the ridge has similar morphology and  
642 modelled rockfall runoff behaviour as the zone of recent activity above the cruise ship dock. Although the relative pattern of  
643 predicted runoff described here is consistent across the four scenarios, our simulations of 95<sup>th</sup> percentile size clasts and no  
644 forest cover result in the most abundant passage of rockfall events to the valley floor with high potential for impacting  
645 infrastructure and imperilling public safety (Fig. 10d).

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Medium rock (50th percentile) dense forest    Medium rock (50th percentile) no forest    Large rock (95th percentile) dense forest    Large rock (95th percentile) no forest



648 Figure 10: Map of topography and rockfall models along the ridgeline that abuts Skagway township and Harbor. (a) Map of  
649 slope angle using merged 2014 airborne and 2023 UAS lidar. (b) Map of rockfall sources (red points) and talus deposits (yellow  
650 polygons). (c, d) Maps of modeled rockfall runout paths along eastern ridgeline in Skagway Valley conducted for 4-2 scenarios  
651 with variable block size and land cover. (panels a-d). The number of rockfall events that traverse each cell pixel is  
652 indicated by the color ramp with warm colors reflecting frequent rockfall passage. Heat map (red shades) denotes relative  
653 density of modeled rockfalls that pass over the railroad. Note the abundance of long runout events in the SW section bordering  
654 the harbour and near the northern extent of the township. The dashed white line denotes the railroad track used to document  
655 rockfall runout and kinetic energy in Fig. 11 for relative position along the tracks. The white box in (c) denotes the area shown  
656 in Fig. 13, site of an historic rockfall event that impacted the township. (e) Satellite image from USGS NAIP (National  
657 Agriculture Imagery Program) with annotated locations indicated. All maps include railroad transect demarcations. Hillshade  
658 images in (b-d) derived from DGGS merged lidar datasets. Background image from USGS NAIP (National  
659 Agriculture Imagery Program).

660  
661 To characterize the location and potential impact of simulated rockfall events that reach the valley floor, we assess controls on  
662 rockfall propagation susceptibility along the railroad transect, we tallied the number and kinetic energy of simulated  
663 rockfall events that traverse the railroad track and quantified topographic properties of the ridge (Fig. 11). Although the  
664 railroad-to-ridge distance (or ridge width) is relatively consistent along the transect, the density of rockfall source areas  
665 varies substantially (Fig. 11a). The southern section of the ridge (4500 to 5400m) has a concentration of toppling source  
666 areas, which is ~4x higher than slopes in the central section (2500 to 4000m) of our transect. Another zone of high source  
667 density occurs at ~2000m on the transect and values to the northern extent are low to moderate in magnitude. Average slope  
668 angle along the ridge varies substantially with high values (>35°) in the southern (4500 to 5400m) and north-central (1500 to  
669 2500m) sections and low values in the south-central section (3000 to 4500m) (Fig. 11b). Average values of slope roughness  
670 (estimated as the standard deviation of slope estimated for local windows with 5m radius) are relatively high in the southern  
671 section of the transect and low in the central section before increasing monotonically to the north (Fig. 11b). Notably, our  
672 analysis shows that slope angle, rockfall source density, and roughness exhibit distinctive patterns along the ridge enabling  
673 us to distinguish their influence on the modelled runout pattern.

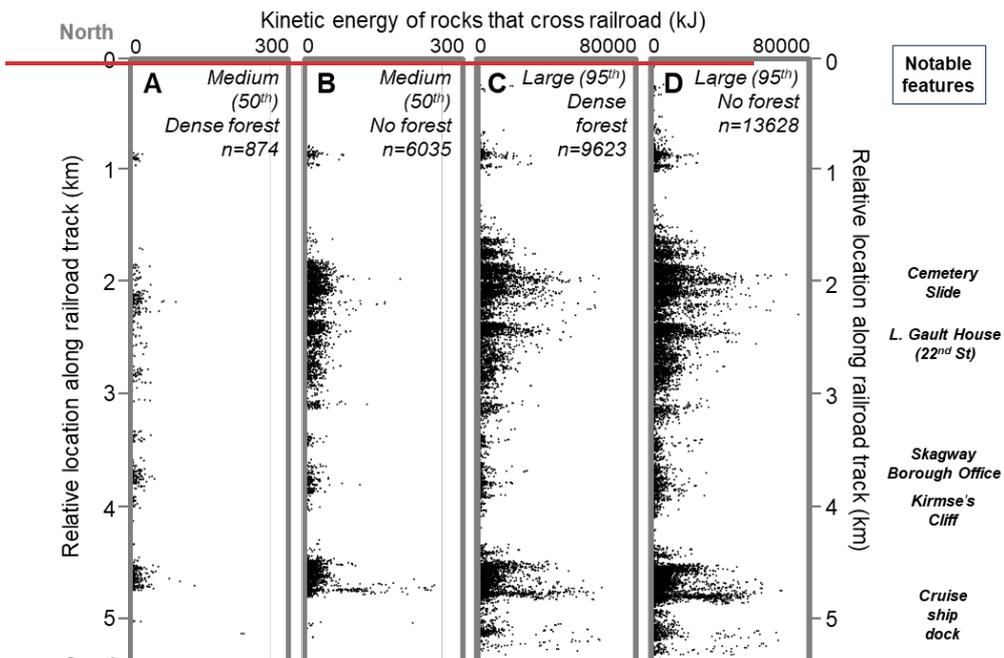
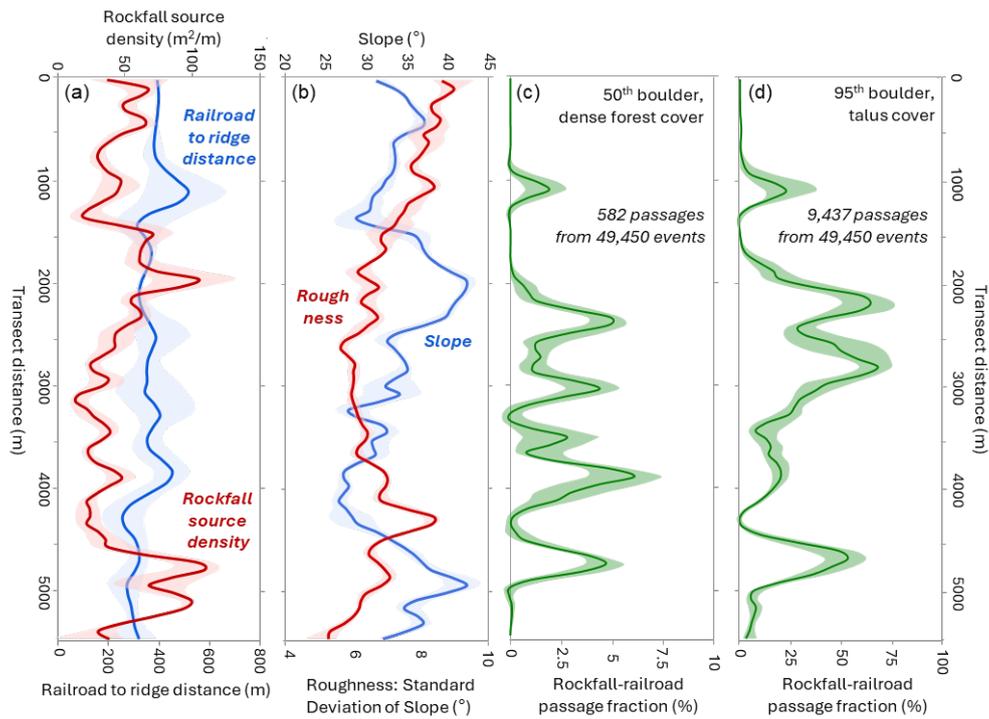
674  
675 Although the relative pattern of simulated runout is consistent for the two scenarios shown here (Fig. 10c,d), our transect  
676 analysis shows that only 1.2% of the simulated events for the 50<sup>th</sup> block/dense forest scenario result in passage across the  
677 railroad whereas 19.1% of the 95<sup>th</sup> block/fine talus scenario events travel across the railroad (Fig. 11c,d). In several sections,  
678 the passage fraction for the 95<sup>th</sup> block/fine talus scenario exceeds 50%, implying that the destabilization of a large block will  
679 likely result in downslope impacts (Fig 11d). By contrast, the passage fractions for the 50<sup>th</sup> block/dense forest scenario rarely  
680 exceed 4% such that unstable blocks have a low probability of impacting the valley (Fig 11c). This disparity reveals the  
681 profound influence of block size and land cover on rockfall propagation susceptibility. Furthermore, the modelled runout  
682 patterns reveal how slope morphology controls rockfall propagation. In both scenarios, sections of the ridge with high  
683 passage fractions correspond with low roughness values and the zone of high roughness at 4000 to 4500m on the transect

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684 corresponds with negligible rockfall passage (Fig 11b). In each of these scenarios, the north-to-south alternating pattern of  
685 high-low rockfall runout hazard is reflected in the number and kinetic energy of elasts that travel to the base of the ridgeline.  
686 Given the 49,450 simulated rockfall events in each of the four scenarios, <2% and 12% rocks crossed the railroad for the  
687 dense forest and no forest scenarios with the 50<sup>th</sup> percentile elast size, respectively. By contrast, using the 95<sup>th</sup> percentile elast  
688 size resulted in nearly 20% and 28% rockfall passage of the railroad in the dense forest and no forest scenarios, respectively.  
689 These results highlight the importance of dense forest and elast size in determining risk of rockfall runout. By contrast, our  
690 analyses show that mean slope angle and rockfall source density do not strongly correlate with the pattern of rockfall passage  
691 across the railroad.



694 Figure 11: Plots of rockfall model variables and topographic properties along the railroad transect that bisects Skagway and  
695 the rockfall prone ridgeline (see Figure 10 for location). (a) Profile of rockfall source density ( $\text{m}^2 \text{m}^{-1}$ ), which depicts the  
696 upslope area of toppling-prone terrain (normalized by distance along transect) calculated with equation 2 for terrain between  
697 the railroad and the ridge crest, and plot of the horizontal distance between the railroad and ridge crest. (b) Profile of average  
698 slope and roughness (estimated as standard deviation of local slope within a 5-m radius window) for terrain between railroad  
699 and the ridge crest. (c) Profile of modeled rockfall-railroad passage percentage for the scenario with 50<sup>th</sup> percentile block size  
700 and dense forest cover, and (d) Profile of modeled rockfall-railroad passage percentage for the scenario with 95<sup>th</sup> percentile  
701 block size and fine talus land cover. Shaded intervals convey standard deviation of values in all plots.

702 Kinetic energy of simulated rocks that pass the railroad tracks (see Fig. 10 for location) for 4 scenarios with variable elast size  
703 and land cover (panels a-d). Note the difference in scale of kinetic energy between panels (a,b) and (c,d). The number of rocks  
704 that pass across the track is highest for the scenario with the largest elast size and no forest cover. Notable locations are  
705 annotated along the right margin of the plots.

## 707 5 Discussion

708 Although rainfall has been responsible for triggering some notable rockfall events in Skagway, but a  
709 substantial seasonal increase in the majority of rockfall events documented in the AKDOT&PF inventory occurred in late  
710 March and April. This period corresponds to rapid increases in temperatures, particularly the minimum daily temperature,  
711 which implies that thawing of ice, snow, and rock mass may contribute to increased activity (Fig. 6). are not preceded by  
712 heavy precipitation. This observation is consistent with many studies that have documented thermal-driven rockfall activity  
713 in alpine environments in Europe and elsewhere (Draebing et al., 2022; Gruber et al., 2004; Krautblatter et al., 2013;  
714 Matsuoka and Sakai, 1999); the proposed mechanisms include active layer thaw, ice-filled fracture destabilization, and  
715 thermal stress cycles, with peak activity typically occurring during spring and early summer months when temperatures rise  
716 above freezing (Draebing et al., 2022; Gruber et al., 2004; Krautblatter et al., 2013; Matsuoka and Sakai, 1999; Mayer et al.,  
717 2024). Given that the low-elevation rock slopes around Skagway are too warm to support permafrost, ice-filled fracture  
718 destabilization may be a relevant mechanism. Field-based studies have noted cleft ice in rockfall scars which implies that ice  
719 segregation processes during cold periods followed by warming and thaw-induced weakening can destabilize fractured rock  
720 masses (Huggel et al., 2012; Ravelin et al., 2013). Dendrochronological studies in Switzerland reveal that 88% of rockfall  
721 scars occur between mid-October and end of May, with peak activity in April when ice lenses formed from meltwater slowly  
722 infiltrate into fractures (Stoffel et al., 2005). These studies highlight the critical role of spring thaw in destabilizing  
723 previously frozen rock masses which may be relevant to the timing of rockfall activity in Skagway. Nonetheless, the specific  
724 timing of seasonal activity remains challenging to predict, motivating efforts to determine the spatial pattern of rockfall  
725 initiation and propagation susceptibility to mitigate hazard.

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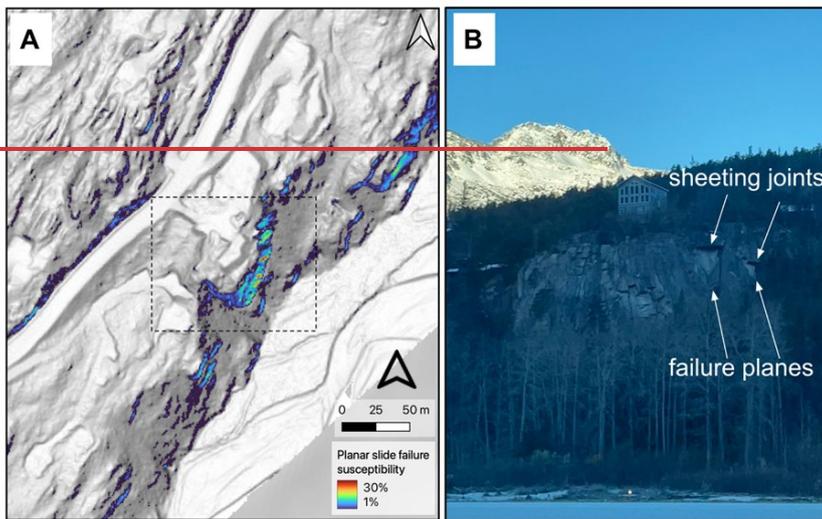
727 ~~Therefore, accurately predicting the timing of rockfall events based on precipitation metrics, like rainfall initiation thresholds~~  
728 ~~used to estimate the likelihood of landslide initiation in Sitka, AK (Patton et al., 2023), is not likely to be successful. Instead,~~  
729 ~~estimating the spatial pattern of rockfall susceptibility is a useful approach to mitigating rockfall hazard.~~ Taken together, our  
730 analyses of rockfall block toppling susceptibility and talus deposition maps demonstrate that rockfall source areas and runout  
731 paths tend to be located/situated on northwest-facing rock slopes in the lower Skagway River valley, Nahku Bay, and the  
732 lower Taiya River valley. Our joint dataset, suggests that this indicating a strong topographic control on rockfall activity  
733 owing/owes to the coincident regularity/alignment of glacial valleys and joint orientations in the area/s, both of which are  
734 remarkably consistent in the area.

735  
736 The joints measured in this study are consistent with the steeply dipping, orthogonal joint sets measured in a nearby (~15km  
737 northwest of Skagway) structural survey in West Creek on a tributary to the Taiya River (Callahan and Wayland, 1965) and  
738 described in a geotechnical assessment of the rock slopes above Skagway's cruise ship dock (Brennan and Whistler, 2022).  
739 These hHigh angle joints are likely formed by the tensile component of shearing from the nearby Chatham Strait fault and  
740 Eastern Denali fault systems, ~~while sheeting joints reflect the combined influence of far field tectonic stresses and~~  
741 ~~topographic stress that arise from landscape curvature (Martel, 2006, 2017).~~ Rock slopes with orthogonal jointing are often  
742 predisposed to block toppling failure particularly where steep orthogonal joints form the sides of toppling blocks and low  
743 angle joints, ~~like the sheeting joints observed in this study, s~~ act as the basal failure plane (Wyllie and Mah, 2004). ~~Because~~  
744 ~~the densest cluster of joints we observed dips steeply to the southeast, rock slopes facing northwest tend to form anti-dip~~  
745 ~~slopes which are conducive to toppling failure. This is supported by larger and more abundant talus deposits on the east sides~~  
746 ~~of the lower Skagway River valley and Nahku bay, confirming that rockfall occurs preferentially on northwest-facing slopes~~  
747 ~~(Fig. 6).~~

748  
749  
750  
751 The results of our kinematic analysis demonstrate that rock slopes in the steep rugged terrain surrounding Skagway are ~~more~~  
752 highly susceptible to block toppling ~~failure/failure than planar slide failure~~. The preference for toppling failure is due to the  
753 steep inclination of joints, which are ~~generally steep and~~ conducive to toppling failure as their near verticality ~~may tends to~~  
754 preclude ~~them-them~~ from daylighting in rock slope faces, a necessary condition for planar sliding ~~or wedge~~ failure (Figure  
755 1). The abundance of talus deposits situated just downslope of zones with high toppling susceptibility (Fig. 9b) is consistent  
756 with other studies (Guerriero et al., 2024) and supports the utility of our modelling approach. More advanced treatments that  
757 explicitly account for individual trees, land cover features, and source block geometry would improve the local accuracy of  
758 our simulations although the regional-scale results presented here are robust and helpful for informing mitigation efforts. On  
759 the west side of the lower Skagway River valley, an isolated zone of steep vertical cliffs is susceptible to planar slide failure

760 as recent planar failures are evident, which release from sheeting joints and slide on joints steeply dipping southeast (Fig.  
761 12).

762 Consistent with previous studies (e.g., Loye et al., 2009), our r

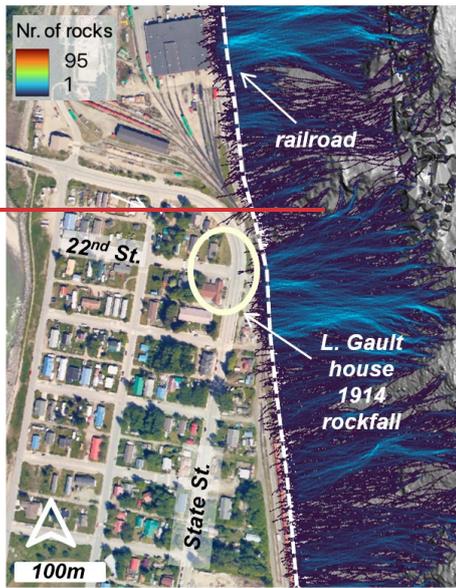


764  
765 Figure 12: Planar sliding susceptibility along the western ridgeline in Skagway River Valley. Location depicted in Fig. 4. (a)  
766 Map of planar failure susceptibility index using equation 1. (b) Image of SE-facing ridgeline in lower Skagway River valley.  
767 Note high values on SE-facing steep bedrock cliffs that coincide with field evidence for sheeting joints and planar failures.  
768 Hillshades derived from DGGS lidar data.

769  
770 Rockfall susceptibility maps are consistent with the failure mode and location of rock slope deformation observed in the  
771 field. Just on the uphill side of the prominent escarpment along the eastern ridgeline in the lower Skagway River valley (Fig.  
772 7), we observed numerous instances of detached parallel slabs of rock separated by open tension cracks that extend  
773 downward for several meters (Fig. 3). These blocks appear to be experiencing the early phases of toppling failure as  
774 vegetation and other environmental perturbations processes can contribute to crack widening (Pawlik et al., 2016; Rempel et  
775 al., 2016). These observations also indicate that our predicted zones of toppling failure along active bedrock escarpments are  
776 likely to continue propagating southeast with the active bedrock escarpment to the southeast providing additional blocks  
777 available to initiate rockfall. More generally, these observations suggest that erosion and SE-southeast-oriented lateral

778 migration of the eastern-ridgeline has been substantial since the glacial retreat 10 to 12 kya. Approximate rReconstruction of  
779 the ridgeline (Moore et al., 2009) to its immediate post-glacial geometry (Moore et al., 2009) implies that 10 to 100 meters  
780 of lateral erosion has occurred during the Holocene. The linear character of the escarpment and pervasive indicators of active  
781 toppling also imply parallel retreat over geomorphic timescales, which occurs given the favourable combination of dip and  
782 slope angle as well as erosion rate (Imaizumi et al., 2015). As such, On-going rockfall activity along this ridgeline suggests  
783 that this the unravelling and lateral retreat of the of the escarpment will likely continue until the interface encounters a  
784 different configuration of slope and structural geometry. continue.

785  
786 Importantly As our simulations demonstrate, extensive rockfall activity and lateral divide migration does not occur along the  
787 entire ridgeline. Rather, our results show large sections of the 5-km long ridgeline with patchy and localized talus deposits  
788 and benchy bedrock landforms that arises because of variations in the depth and pattern of glacial erosion. On Kirmse's Cliff  
789 (at 4200m 4 km on the railroad transect, Fig. 10e), for example, talus deposits extend upslope from the valley floor to vertical  
790 cliff faces that terminate at the ridgeline (Fig. 5). In this area, rockfall-toppling susceptibility is relatively high, but  
791 concentrated in a relatively small area of vertical rock slope, and a forested talus deposit and benchy zone sits between the  
792 cliff and infrastructure in Skagway the valley floor (Fig. 10b9). By contrast, several areas along the escarpment exhibit high  
793 potential for rockfall runout to reach the valley. At the cruise ship dock (4500m on the transect-5 to 5 km), continuous,  
794 rockfall-prone slopes above the cruise ship dock span from the escarpment to the base of the slope, resulting in a large area  
795 with high rockfall propagation susceptibility in close proximity to harbour infrastructure. Similarly, from 1.5 to 2.5 km 1500  
796 to 2000m on our railroad transect, a high-relief escarpment with steep vertical bedrock outcrops near the crest exhibits a  
797 continuous steep slope to the valley floor (Fig. 10a). In this area, which is the source area for the Cemetery Slide, rockfall  
798 susceptibility and long runout potential are high and we observed frequent spalling and rolling rocks traversing the talus  
799 deposits during our field work (Fig. 10c.db). In addition In fact, The Daily Alaskan reported a 1914 rockfall event just south  
800 of the Cemetery Slide at 2.5 km 2500m along the transect which coincides with our simulations of high rockfall runout  
801 potential (Figs. 10c.d0 & 13). During that event, L. Gault noted “an avalanche of earth and stone” that “leaped the railroad  
802 track...pounding against the fence.” Gault also noted that the railroad track served as a “safety barrier” because “the force of  
803 the descent had been so much lessened.” Further afield in the NE-northeast section of Nahku Bay, another zone of high  
804 rockfall susceptibility is revealed by our analysis. Steep rocky slopes rise to an elevation of 250 m with close proximity to  
805 Dyea Road (Fig. 7). This area lacks a well-developed runout path, but it does coincide with source areas for two dozens of  
806 events in the AKDOT rockfall inventory that impacted Dyea Road. Most generally, a rigorous characterization of rockfall  
807 risk in Skagway requires estimation of occupation and usage of structures and transportation corridors coupled with our  
808 results (e.g., Michoud et al., 2012).



810  
 811 Figure 13: Detailed map of modeled rockfall runout (Medium rock (50<sup>th</sup> percentile), no forest) with location depicted shown  
 812 by white box on Fig. 10b. Note the location of rockfall near the intersection of 22<sup>nd</sup> and State Street described by L. Gault in  
 813 the March 26, 1914, edition of The Daily Alaskan. That location corresponds with a region of likely rockfall runout identified  
 814 by our coupled initiation-runout simulations. Hillshade derived from DGGS lidar data. Background image from USGS NAIP  
 815 (National Agriculture Imagery Program).

816  
 817 Our analyses reveal distinctive zones of high rockfall initiation and runout (or propagation) susceptibility hazard in the  
 818 Skagway region that result from the combination of glacial erosion that sets the morphology of bedrock slopes and the  
 819 orientation of joint sets that determine the geometry of potential bedrock failures. Most likely, the orientation and extent of  
 820 glacial erosion is not independent of the joints (Hooyer et al., 2012; Krabbendam and Glasser, 2011). Rather, we suspect that  
 821 the nearly parallel orientation of joints and the N-NE trending ridgeline along the eastern margin of the lower Skagway River  
 822 valley arise due to glacial erosion tracking discontinuities that facilitate erosion more readily than undeformed bedrock.  
 823 Localized zones of the prominent ridgeline that abuts Skagway River valley exhibit abundant historic rockfall activity and  
 824 over long timescales this activity is likely to continue. As such, This long-term evolution implies feedbacks between rock  
 825 structure and bedrock erosion processes from glaciers and rockfalls that support strong gradients in rockfall activity. In  
 826 assessing how these factors moderate the current hazard context, our findings critically depend on the combination of

827 kinematic modelling and runout susceptibility to provide key constraints for identifying areas at risk to inform mitigation  
828 efforts. ~~Most generally~~ Additional work is needed, however, to conduct a rigorous characterization of rockfall risk in  
829 Skagway requires that includes estimation of occupation and usage of structures and transportation corridors to be coupled  
830 with our results (e.g., Michoud et al., 2012).

## 831 **6 Conclusions**

832 The integrated kinematic analysis and the runout modelling presented here matches well with field observations and provides  
833 the basis for a sophisticated approach in rockfall hazard assessment. ~~Rock slope failure is a major driver of landscape~~  
834 evolution in Skagway since glacial retreat. The steeply dipping orthogonal joints in Skagway's rock slopes are highly  
835 conductive to block toppling failure which has resulted in highly asymmetric rockfall susceptibility. Our results highlight a  
836 distinct contrast in the morphology of Skagway River valley's eastern and western ridgelines, which is primarily determined  
837 by the predisposition of northwest-facing slopes to block toppling failure along a densely fractured joint set that dips steeply  
838 to the southeast. This is consistent with geomorphic evidence of extensive rockfall activity, where talus deposits are more  
839 abundant and larger on the eastern side of the lower Skagway River valley than the west. This structural control on rockfall  
840 activity is also reflected evident in the east and west sides of Nahku Bay, situated to the west of Skagway. Historic rockfall  
841 records Our analysis of a rockfall inventory indicates a similar preference for initiating on northwest-facing slopes and reveal  
842 a seasonal peak in rockfall activity in late March and April that likely results from thawing of ice-filled joints and reduction  
843 of rock mass strength. Our simulations of rockfall runout show areas with high rockfall susceptibility that may merit further  
844 investigation and mitigation: 1) the ridgeline and slopes above the cruise ship dock, 2) the high-relief topography at the north  
845 end of Skagway (which includes the Cemetery Slide), and 3) the eastern ridgeline bordering Nahku Bay. Recent rockfall  
846 source areas located in these regions are characterized by steep northwest-facing slopes that are highly susceptible to block  
847 toppling failure and reach >200 m in elevation which facilitates long runout paths. Our results also demonstrate that slope  
848 roughness over ~10m scales serves as a better predictor of rockfall propagation that slope angle and the density of rockfall  
849 source areas. Our integrated modelling approach and field assessment provides a simple but robust approach to assess  
850 rockfall hazard.

## 851 **7 Data availability**

852 All topographic data used for analyses are cited within the text. The structural data are available as Supplemental material.

853 **8 Author contribution**

854 IW and JR conceived and designed the study and IW and JR executed the study, RC provided on-site context, knowledge  
855 and guidance, AP contributed to study design and fieldwork, IW and JR prepared the manuscript with contributions from all  
856 other co-authors.

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865

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