

1 **Criteria and Tools for Determining Drainage Divide Stability**

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9

10 **Abstract**

11

12 Watersheds are the fundamental organizing units in landscapes and thus the controls on  
13 drainage divide location and mobility are an essential facet of landscape evolution. Additionally,  
14 many common topographic analyses fundamentally assume that river network topology and  
15 divide locations are largely static, allowing channel profile form to be interpreted in terms of  
16 spatio-temporal patterns of rock uplift rate relative to base level, climate, or rock properties.  
17 Recently however, it has been suggested that drainage divides are more mobile than previously  
18 thought and that divide mobility, and resulting changes in drainage area, could potentially  
19 confound interpretations of river profiles. Ultimately, reliable metrics are needed to diagnose  
20 the mobility of divides as part of routine landscape analyses. One such recently proposed  
21 metric is cross-divide contrasts in  $\chi$ , a proxy for steady-state channel elevation, but cross-divide  
22 contrasts in a number of topographic metrics show promise. Here we use a series of landscape  
23 evolution simulations in which we induce divide mobility under different conditions to test the  
24 utility of a suite of topographic metrics of divide mobility and for comparison with natural  
25 examples in the eastern Greater Caucasus Mountains, the Kars Volcanic Plateau, and the  
26 western San Bernadino Mountains. Specifically, we test cross-divide contrasts in mean gradient,  
27 mean local relief, channel bed elevation, and  $\chi$  all measured at, or averaged upstream of, a  
28 reference drainage area. Our results highlight that cross-divide contrasts in  $\chi$  only faithfully  
29 reflect current divide mobility when uplift, rock erodibility, climate, and catchment outlet

30 elevation are uniform across both river networks on either side of the divide, otherwise a  $\chi$ -  
31 anomaly only indicates a possible future divide instability. The other metrics appear to be more  
32 reliable representations of current divide motion, but in natural landscapes, only cross-divide  
33 contrasts in mean gradient and local relief appear to consistently provide useful information.  
34 Multiple divide metrics should be considered simultaneously and across-divide values of all  
35 metrics examined quantitatively as visual assessment is not sufficiently reliable in many cases.  
36 We provide a series of Matlab tools built using TopoToolbox to facilitate routine analysis.

37

## 38 **1. Introduction**

39 Drainage divides are fundamental organizing boundaries within landscapes. The extent  
40 to which the topologic form of divides, and thus river networks as a whole, are largely static  
41 (e.g., Bishop, 1995; Oberlander, 1985) or are dynamic features, changing rapidly through  
42 progressive divide migration and/or discrete capture events has recently become a topic of  
43 considerable interest and some debate (e.g. Whipple et al., 2017c; Willett et al., 2014).

44 Assessing whether a drainage divide is potentially mobile is important, not only for quantifying  
45 how landscape evolution is affected by the resulting changes in drainage area, but also because  
46 many of the topographic metrics we use to interpret climatic or tectonic change (e.g., Wobus et  
47 al., 2006) assume that drainage area has not changed significantly over the response timescale  
48 of a catchment (e.g., Howard, 1988; Kooi and Beaumont, 1996; Whipple, 2001). Violation of this  
49 static drainage area assumption at best complicates the interpretation of topographic metrics  
50 and at worst invalidates the inferences drawn from them (e.g. Whipple et al., 2017a, 2017b;  
51 Willett, 2017; Yang et al., 2015). While recent work suggests that under normal circumstances  
52 the rate of divide motion is slow compared to the rate of channel adjustment to drainage area  
53 change (Whipple et al., 2017c), the potential importance of drainage divide mobility suggests  
54 that assessments of divide stability should be a routine part of topographic analyses.

55 Metrics of the relative stability of drainage divides are not new, indeed Gilbert (1877)  
56 first proposed a means of assessing divide stability with his 'law of unequal declivities', positing  
57 that if a divide was asymmetrical, this would imply different erosion rates on either side of the  
58 divide. The resulting across-divide erosion rate contrast would force the divide to move toward

59 the side with lower slopes and erosion rates (Figure 1A). Recently, Willett et al (2014) proposed  
60 a new method of assessing divide stability through the use of  $\chi$ -maps.  $\chi$ , discussed in more  
61 detail in the following section, can be used as a proxy for steady-state channel elevation and  
62 thus this quantity should be nearly equal on either side of a stable divide. Maps of drainage  
63 networks colored by  $\chi$  can reveal  $\chi$ -anomalies across divides, where the  $\chi$  value at channel  
64 heads are higher on one side of a divide, suggesting that this divide is unstable and should  
65 move from lower to higher  $\chi$ . Barring complicating factors, divide migration would continue  
66 until the topology of the drainage network and drainage area distribution has changed such  
67 that the  $\chi$ -anomaly is removed. In a limited number of locations,  $\chi$ -anomalies appear  
68 coincident with an across-divide difference in average erosion rate, the underlying driver of  
69 divide motion (e.g., Beeson et al., 2017; Willett et al., 2014).

70  $\chi$ -maps are appealing as they are 1) relatively easy to calculate and 2) allow for a quick  
71 visual assessment of the stability of divides across a large area. There are, however, some  
72 challenges with their use and interpretation. Most significantly, the interpretation of  $\chi$ -  
73 anomalies typically assumes uniform uplift, rock erodibility, and climate (Willett et al., 2014)  
74 and thus in situations where any of those parameters vary, as is often the case in natural  
75 systems,  $\chi$ -anomalies can occur even when divides are stable (e.g. Whipple et al., 2017c). This  
76 led Whipple et al. (2017c) to propose a suite of alternative metrics of divide stability, largely an  
77 expansion of the ideas originally put forward by Gilbert (1877), including cross divide  
78 differences in channel elevation at a reference drainage area, mean headwater hillslope  
79 gradient, and mean headwater local relief. Whipple et al. (2017c) showed that for a simple  
80 synthetic landscape experiencing a non-uniform uplift rate, these alternative metrics were  
81 more consistent indicators of the current rate and direction of divide motion than across-divide  
82 differences in  $\chi$ . Here we expand upon that work by 1) developing a set of user friendly Matlab  
83 based tools to produce maps of these alternative metrics along with  $\chi$ -maps and to perform  
84 detailed analysis of multiple divide stability criteria, 2) applying these tools to two synthetic  
85 landscapes with non-uniform uplift and non-uniform lithology, 3) applying these metrics to  
86 three natural examples, and 4) comparing and contrasting the relative utility of these four  
87 different divide stability metrics.

88

## 89 **2. Metrics of Divide Stability**

90

### 91 **2.1. Theory and Limitations of Metrics**

92 Active motion of a drainage divide implies across-divide differences in erosion rates,  
93 thus many potential metrics of divide stability will essentially be topographic proxies for erosion  
94 rate. This was the basis for Gilbert's (1877) law of unequal declivities, which assumed that  
95 divides bounded by distinctly different gradients were unstable, with faster erosion on the  
96 steeper side progressively moving the divide towards the side with a gentler slope (Figure 1A).  
97 Since the time of Gilbert, empirical measures of erosion rate and comparison to various  
98 topographic metrics have suggested monotonic relationships at the catchment scale between  
99 erosion rates and normalized channel steepness (river slope normalized for drainage area) or  
100 local topographic relief (e.g., Harel et al., 2016; Kirby and Whipple, 2012; Lague, 2014) and at  
101 the hillslope scale between erosion rates and mean hillslope gradient, hillslope relief, and  
102 hilltop curvature (e.g., Hurst et al., 2013; Roering et al., 2007, 1999). Ultimately, divide motion  
103 is driven by differences in erosion rate at or in close proximity to the divide itself, so a metric  
104 like normalized channel steepness, which is only measurable away from the divide, may not be  
105 a viable proxy. Therefore, we choose to focus on gradient and relief. We do not consider  
106 hillslope curvature as accurate measurement of this quantity requires high resolution  
107 topographic data (e.g., Roering et al., 1999) and thus is not widely applicable to areas for which  
108 such data does not exist. Because mean gradients reach threshold values in steep landscapes  
109 and become insensitive to increases in erosion rate (e.g., Burbank et al., 1996), if gradients on  
110 both side of a divide are above  $\sim 0.7$ , then it is expected that the slope metric will no longer be  
111 sensitive to divide mobility. We also consider a third proxy, across-divide differences in channel  
112 elevation at a reference drainage area. Together we refer to these three metrics as the 'Gilbert  
113 metrics'. In detail, all three Gilbert metrics are intimately related because for a given divide, if a  
114 channel on one side has a steeper hillslope gradient, this generally implies both greater local  
115 relief and lower elevation of the channel at a reference drainage area as a simple geometrical  
116 consequence (Figure 1B).

117 We compare the Gilbert metrics to differences in the quantity ‘ $\chi$ ’ across a divide,  
118 (Willett et al., 2014). The derivation of and underlying rationale for the calculation of  $\chi$  is  
119 discussed in detail in several recent publications (e.g., Harkins et al., 2007; Mudd et al., 2014;  
120 Perron and Royden, 2013; Royden and Perron, 2013) so we provide only a brief treatment here.  
121 In practice,  $\chi$  is an integral quantity evaluated along a channel from the outlet ( $x_b$ ) to the  
122 position of interest ( $x$ ) with

$$123 \chi = \int_{x_b}^x \left( \frac{A_0}{A(x')} \right)^{\theta_{ref}} dx' \quad (1)$$

125 where  $A$  is upstream drainage area,  $A_0$  is a reference scaling area,  $\theta_{ref}$  is a reference concavity  
126 (Wobus et al., 2006), and  $x'$  is a dummy variable. A plot of channel elevation vs  $\chi$  for a stream  
127 that is equilibrated to a spatially constant uplift rate and erosional efficiency should be a  
128 straight line and under these circumstances  $\chi$  can be considered a proxy for steady-state  
129 channel elevation. If  $A_0$  is set to unity then the slope of the chi-z plot will equal the normalized  
130 channel steepness (Wobus et al., 2006) but is dimensionless. As described by Willett et al.  
131 (2014), differences in  $\chi$  at a reference drainage area across a drainage divide imply different  
132 steady-state channel elevations if uplift, climate, and rock properties are spatially uniform.  
133 Thus, the divide and drainage network topology are unstable if uplift, climate and rock  
134 properties are indeed uniform or will become unstable if current spatial differences in these  
135 properties are eliminated in future. This led to the proposition that maps of stream networks  
136 colored by  $\chi$  and the identification of  $\chi$ -anomalies across drainage divides could provide proxies  
137 for the stability of a drainage network.

139 In practice, interpretation of  $\chi$ -maps and  $\chi$ -anomalies has some challenges. When the  
140 assumption of spatially uniform rate of uplift (or base level fall) and erosional efficiency (set  
141 primarily by climate and rock properties) is violated,  $\chi$ -anomalies can develop and persist at  
142 stable divides (e.g., Whipple et al., 2017c). In addition  $\chi$ -values are sensitive to the choice of  $x_b$   
143 and thus the elevation of the catchment outlet defined for computation of  $\chi$ . Because  $\chi$  is an  
144 integrated quantity and calculated from the outlet to the headwaters of a stream network, a  $\chi$ -  
145 anomaly can result because of the choice of different outlet elevations for streams on either

146 side of the divide. Many potential problems with this can be avoided with careful analysis and  
147 treatment of data (e.g., Willett et al., 2014), such as ensuring that stream networks are  
148 complete and all drain to the desired outlet elevation, but in some instances the choice of the  
149 ‘correct’ outlet elevation is non-trivial. As an example, we consider the case of the Greater  
150 Caucasus Mountains and two principle drainage divides within this range, one between rivers  
151 draining to the Black Sea or Caspian Sea and one between rivers draining into the northern or  
152 southern forelands of this range (Figure 2). We consider three options for selecting outlet  
153 elevations and calculate  $\chi$  using (1) true base level, which varies between 0 m for rivers draining  
154 to the Black Sea and -27 m for rivers draining into the internally drained Caspian Sea, (2) a  
155 constant elevation of 550 m that roughly approximates the range-front of the Greater Caucasus  
156 in both the northern and southern forelands, or (3) a variable outlet elevation based on manual  
157 mapping of the apparent bedrock-alluvial transition at the range-front (Figure 2). If we choose  
158 to use true base level, this suggests the presence of a stark  $\chi$ -anomaly around all streams  
159 draining into the Black Sea and through the southern foreland (Figure 2A). This  $\chi$ -anomaly  
160 persists (but is more subtle) if we use the constant elevation of 550m, but the anomaly  
161 disappears when using the bedrock-alluvial transition as the outlet elevation (Figure 2B & C).  
162 Similarly, in the eastern Greater Caucasus,  $\chi$ -anomalies suggest that the main divide between  
163 northern and southern drainages is unstable, but is predicted to move either south using true  
164 base-level (Figure 2A) or north using constant elevation or the bedrock-alluvial transition  
165 (Figure 2B & C). This highlights that care must be exercised when choosing outlet elevations for  
166  $\chi$  analysis, but also that there may be non-unique answers depending on different, but still  
167 reasonable, choices of outlet elevation.

168

## 169 **2.2. Proposed Methodology for Use of Divide Metrics**

170 While there are some potential problems with the use of  $\chi$ -maps, they are appealing as  
171 a data exploration tool as they allow for quick assessment of the relative stability of a drainage  
172 divide and associated drainage network. Here we develop similar maps using the three  
173 “Gilbert” metrics described above. Mapping the elevation metric only requires coloring a  
174 drainage network by channel elevation. The elevation metric is interpreted the same as  $\chi$  in  $\chi$ -

175 maps: divides are expected to move from low to high values in the presence of an anomaly in  
176 channel elevations (Figure 1C & D).

177 For the local relief and gradient metrics, we are primarily concerned with average values  
178 of these properties near the divide, so a simple strategy of coloring stream networks by  
179 upstream running averages of either local relief or gradient is sufficient. These two metrics are  
180 more direct proxies for erosion and as such, divides are expected to move from high to low  
181 values (Figure 1C & D). For all four metrics, we are only concerned with the values at the  
182 channel heads, which are approximated by choosing a reference drainage area at which to  
183 evaluate the values, which we refer to as 'stream endpoints', so for all metrics (including  $\chi$ ) a  
184 full map of values along streams are not necessary, but provide for useful visuals.

185 Visual comparisons of contrasts in colors across a divide are useful for identifying  
186 potentially interesting patterns. However, the perception that a particular divide is unstable can  
187 be influenced by visual bias or choices of color scales. To interrogate this further one must  
188 assess the actual across-divide differences in the quantities of interest. Additionally, sometimes  
189 a single drainage divide may be heterogeneous so it is useful to segment a divide and analyze  
190 the stability of these sections individually. We visualize individual divide sections as histograms  
191 of values at all of the stream endpoints on either side of a divide (Figure 1D). In practice, this is  
192 useful to assess the degree of overlap or separation between values on either side of a divide.  
193 Along with the histograms, we calculate a mean, standard error of the mean, 95% bootstrap  
194 confidence interval, and standard deviation for the population of values on either side of a  
195 divide. In this study, we primarily use the conservative criteria that a divide is potentially stable  
196 according to a given metric if the mean of one side of the divide is within one standard  
197 deviation of the mean of the other side. These 'delta' values and their associated uncertainties  
198 can then be standardized so that positive and negative delta values of the different metrics  
199 indicate similar divide migration direction, providing an easy visual assessment of divide  
200 stability for individual divide segments (Figure 1E). The specific stability criteria we use is  
201 arbitrary, but it serves well to illustrate our main points. Ultimately determining the most  
202 suitable criteria requires comparisons of these types of data with empirical observations of

203 divide motion. More generally, we emphasize the importance of looking at the populations of  
204 values across a divide and choosing some consistent criteria for stability or instability.

205

### 206 **2.3. Tools for Evaluating Divide Stability**

207 To ease assessing divide stability, we developed a series of Matlab functions based upon  
208 TopoToolbox (Schwanghart and Scherler, 2014). These functions are designed to produce  
209 visually appealing and readily assessed maps of the metrics described in the previous section  
210 (Figure 1). Beyond facilitating rapid qualitative assessment, a primary goal was also to allow  
211 users to interrogate individual sections of divides more deeply as will be illustrated in the  
212 examples below. These functions are available via github  
213 (<http://github.com/amforte/DivideTools>) and all of the base plots and data for the subsequent  
214 figures and maps were generated with these codes. In the supplement, we provide a brief  
215 summary of the primary functions included in this repository and where appropriate, the  
216 rationale behind the workings of these functions.

217

## 218 **3. Principles of Metric Interpretation from Simulations**

219 We present two landscape evolution models as simple examples of the expected  
220 behavior of the different divide metrics and to form a basic set of rules for interpreting these  
221 metrics in concert. These simulations are explicitly designed to explore cases that violate the  
222 underlying assumptions of  $\chi$ -map analysis, specifically landscapes experiencing non-uniform  
223 uplift rate and/or spatially/temporally variable erosional efficiency. Both models were run in  
224 FastScape (Braun and Willett, 2013) and were 10 km wide by 5 km long with a grid spacing of 25  
225 meters. For both models, we track the average rate of divide motion at each time-step and  
226 compare that to across-divide differences in erosion rate, which is driving the divide motion,  
227 and the four proposed metrics, channel head elevation, mean upstream local relief, mean  
228 upstream gradient, and  $\chi$  (computed using equation 1).

229

### 230 **3.1. Asymmetric Uplift Simulation**



231           The first simulation matches the scenario previously presented in Whipple et al. (2017c).  
232 In this simulation, we induce divide motion by first imposing an asymmetric uplift rate that  
233 increases toward the top of the model on an initially steady-state landscape, thus driving the  
234 divide towards the north (top) side of the model. This uplift-rate gradient is imposed for 20 Myr  
235 after which we force the model to return to a spatially uniform uplift rate for another 20 Myr to  
236 allow the divide to return to its original position at the center of the model. As expected,  
237 across-divide differences in erosion rate are linearly correlated with divide migration rates  
238 (Figure 3A). Across divide differences in all of the Gilbert metrics show similar linear  
239 correlations with divide migration rate (Figure 3B-D). In contrast, during the asymmetric uplift  
240 phase,  $\chi$  is inversely correlated with divide migration rate, with the magnitude of the  $\chi$ -  
241 anomaly increasing as the divide approaches a stable position. Conversely, when the uniform  
242 uplift phase begins,  $\chi$ -anomalies correctly track divide migration rate (Figure 3E). Visualizing  
243 these across-divide differences as histograms of the values of the metrics at the reference  
244 drainage area provides an assessment of the variability even in this simple synthetic landscape  
245 and also highlights when the different metrics disagree (Figure 4).

246

### 247 **3.2. Dipping Hard Layer Simulation**

248           The second simulation has uniform uplift throughout the model run, but has a 500m  
249 thick layer, dipping at 35° to toward the top of the model that is more resistant to erosion than  
250 the rest of the landscape. The model is first run for a sufficient time to develop a steady state  
251 landscape with a single erosional efficiency. When the hard layer is first exposed, the divide  
252 begins to move south (e.g. 3.5 Myr in Figure 5). Once the hard layer reaches the divide, the  
253 divide begins to move north (e.g 6.0 & 8.0 Myr in Figure 5), until the hard layer is completely  
254 eroded at which time the divide again moves south toward the center of the model (e.g. 10.0  
255 Myr in Figure 5). Like the asymmetric uplift model, divide migration rate is roughly linearly  
256 correlated with across-divide differences in erosion rate and all three Gilbert metrics (Figure  
257 5A-D), whereas  $\chi$  has a more complicated relationship to divide migration rate (Figure 5E). Also,  
258 like the asymmetric uplift model, this complicated relationship between across-divide

259 differences in  $\chi$  and divide migration rate results in times when  $\chi$ -anomalies incorrectly predict  
260 the current direction of divide motion (Figure 6).

261

### 262 **3.3. Proposed Usage of Metrics**

263 Both simulations indicate that the Gilbert metrics are well correlated to both the current  
264 rate and direction of divide migration rate, and by extension the magnitude and sign of across-  
265 divide differences in erosion rates. However, it is important to note that the relationship  
266 between values of across-divide differences in any of these metrics and either the exact divide  
267 migration rate or across divide difference in erosion rate will depend on various factors  
268 including erosional efficiency, uplift rates, and the form of the erosion law (e.g. Whipple et al.,  
269 2017c). Thus, outside of application to models, the magnitudes of across divide differences in  
270 any of these metrics can only be reliably interpreted in terms of the direction of divide motion.  
271 With respect to  $\chi$ , the model results highlight the expected outcome that  $\chi$  only correctly  
272 predicts the current direction of divide motion when the uniform condition assumptions  
273 inherent in the interpretation of  $\chi$ -maps are met. What the models also highlight is that  $\chi$ -  
274 anomalies, and the divide motion implied by them, that exist or develop during non-uniform  
275 portions of the model runs indicate the predicted motion of the divide when and if the  
276 landscape returns to uniform conditions in the future. For example, at 2.0 Myr in the  
277 asymmetric uplift model, when the divide is moving north because of the gradient in uplift rate,  
278 the  $\chi$ -anomaly that progressively develops indicates that the divide will eventually move south  
279 when (or if) that uplift gradient is relaxed (Figure 3 & 4).

280 Generally, the results of these two simulations suggests that using  $\chi$ -maps in concert  
281 with one (or all) of the Gilbert metrics is ideal and further that if  $\chi$ -maps are used exclusively,  
282 the current stability of a drainage divide may be interpreted incorrectly. If the different metrics  
283 agree, this should indicate both the direction of current divide motion and that the uniformity  
284 assumption within  $\chi$  is met, or alternatively that the differences in uplift rate and erosional  
285 efficiency in the landscape either are small or counterbalance each other. If  $\chi$  disagrees with  
286 the other metrics, this likely suggests that the Gilbert metrics are indicative of current divide  
287 behavior and that  $\chi$  is (1) indicating potential future divide behavior should differences in uplift

288 rate and/or erosional efficiency be eliminated, and (2) may suggest there is sufficient variability  
289 in uplift rate and/or erosional efficiency to cause divergence in the metrics. With this as a rubric  
290 for interpreting across divide differences in these metrics, we now apply them to three field  
291 examples.

292

## 293 **4. Field Examples**

294

### 295 ***4.1. Eastern Greater Caucasus Mountains***

296 The Greater Caucasus Mountains are the main loci of active shortening within the  
297 central Arabia-Eurasia collision zone (e.g., Reilinger et al., 2006) and in the eastern Greater  
298 Caucasus (west of 45°E) are characterized by active thrust systems along both its northern and  
299 southern margins (Forte et al., 2014). This segment of the Greater Caucasus divide is notable as  
300 its location is consistently offset southward, sometimes by up to 40 km, with respect to the  
301 highest peaks of the range. Based on spatial patterns in normalized channel steepness and  
302 results of landscape evolution models, Forte et al., (2015) hypothesized that (1) the drainage  
303 divide location predates development of the topographic crest of the range, (2) the divide  
304 location is at least partially controlled by spatial gradients in uplift rate that reach a maximum  
305 near the divide, and (3) eventually the divide should move north as channels are generally  
306 steeper south of the divide.

307 For analysis purposes, we segment this drainage divide into 8 sections based on visual  
308 inspection of the four metrics and choose break points between portions of the divide that  
309 appear to display transitions in at least one of the criteria. Results for all metrics and associated  
310 river profiles for the eight divide segments are available in the supplement (Supplemental  
311 Figures 1-16). In practice, while the elevation metric was useful in the model results (e.g. Figure  
312 4 & 6), the results of the elevation metric are typically equivocal in natural settings we have  
313 examined due to large standard deviations (Figure 7). The elevation metric, and indeed all of  
314 the metrics, tend to indicate more divide mobility if the standard error of the mean is used to  
315 estimate uncertainty. Unsurprisingly, the mean upstream slope and mean upstream relief  
316 metrics are very similar, though the similarity of these metrics will depend on the chosen relief

317 radius (e.g., DiBiase et al., 2010). Thus, in comparing divide metrics along the length of the  
318 divide for this and subsequent examples, we focus our discussion on  $\chi$  and relief.

319 With the exception of two segments (GC7 & GC8, Figure 7C),  $\chi$  always predicts  
320 northward movement of the divide (using the 550m outlet elevation) whereas the relief metric  
321 suggests the divide is stable within uncertainty (using the standard deviation) except for two  
322 segments (GC3 and GC5, Figure 7C). The means of all metrics (except for GC7) agree in the  
323 direction of divide motion and applying a less restrictive uncertainty (e.g. standard error) shows  
324 more agreement between all metrics. As discussed earlier, the choice of outlet elevation for  $\chi$   
325 in the eastern Greater Caucasus (e.g. Figure 2) significantly influences predicted divide  
326 behavior, with  $\chi$  suggesting southward motion of the divide if 'true base level' is used for the  
327 outlet elevation (Figure 7D). There are no quantitative estimates of erosion rates on either side  
328 of the divide so we do not have a way to evaluate the 'right' answer in this setting, but  
329 depending on the uncertainty criteria used, this result is consistent with previous suggestions  
330 by Forte et al. (2015) that the divide is currently fixed but may eventually move northward  
331 depending on future circumstances or may already be moving northward. There are isolated  
332 south flowing drainages showing characteristic 'area-gain' signatures in  $\chi$ -normalized profiles  
333 (Willett et al., 2014) indicating past divide motion to the north (Figure 8), though these  
334 signatures are rare (Supplemental Figures 2, 4, 6, 8, 10, 12, 14, & 16).

335 In terms of diagnosing contributions to divide stability, there are no significant  
336 differences in either rock type (Forte et al., 2014) or mean annual precipitation (Forte et al.,  
337 2016) directly across the divide, suggesting that a change in erosional efficiency is unlikely as a  
338 driver. Thus, the simplest interpretation of these results is similar to that posited by Forte et al.  
339 (2015, 2014), that this indicates the presence of an uplift rate gradient that is 'holding' the  
340 divide in place and that the  $\chi$  metric is sensitive to this and indicating the expected reaction of  
341 the divide if or when this uplift rate gradient dissipates.

342

#### 343 **4.2. Kars Volcanic Plateau**

344 The Kars Volcanic Plateau is also part of the Arabia-Eurasia collision zone, but the  
345 tectonics and local geology are decidedly different than that of the Greater Caucasus to the

346 north. This portion of the collision zone has relatively low rates of active internal deformation  
347 (Reilinger et al., 2006), which occur primarily on normal and strike slip faults with some  
348 portions of the deformation related to local volcanic features (e.g., Dhont and Chorowicz, 2006;  
349 Koçyiğit et al., 2001). The Kars Plateau is part of the broader East Anatolian Plateau which lacks  
350 mantle lithosphere (Zor, 2008) after a slab detachment or delamination event at ~7-8 Ma  
351 (Keskin, 2003; Şengör et al., 2003). The average ~2 km high, roughly dome shaped plateau (e.g.,  
352 Şengör et al., 2003) is thought to have been produced by this delamination event through  
353 mantle upwelling (e.g., Göğüş and Pysklywec, 2008). The delamination is also thought to have  
354 driven extensive melting and the eruption of a package of nearly horizontal volcanic rocks  
355 ranging in composition from basalts to rhyolites with thicknesses of 100-1000 m that blanket  
356 much of the plateau region (Keskin et al., 1998; Pearce et al., 1990, Figure 8B).

357 We selected two distinct drainage divides within the Kars Volcanic Plateau, both of  
358 which lie near the edges of the volcanic deposits (Figure 9B). We segment the western drainage  
359 divide into 4 sections based again on visual inspection of divide metrics, but also to separate  
360 areas where there are major differences in the outlet locations on at least one side of the  
361 divide. Streams west of the KV1 & KV2 divides are part of the Choruh watershed and flow into  
362 the Black Sea and streams east of the KV1 & KV2 divides form the headwaters of the Kura river,  
363 the main axial drainage within the Kura Basin and which flows eastward between the Greater  
364 and Lesser Caucasus and eventually into the Caspian Sea. Streams east of KV3 are also part of  
365 the Choruh watershed, but streams west of KV3 flow southeast into the Ararat Basin and are  
366 tributaries of the Arax River that merges with the Kura River shortly before it empties into the  
367 Caspian. KV4 divides two different tributaries of the Arax river and KV5, on the eastern edge of  
368 the Kars Plateau, separates branches and tributaries of the Kura river (Figure 9).

369 For all the analyzed divides, the relief metric suggests they are stable using the standard  
370 deviation criterion and close to stable using the standard error whereas  $\chi$  consistently suggests  
371 that divides should move toward the center of the Kars Plateau (Figure 9D). Using the model  
372 results as a basis for interpretation suggests this is likely a case in which a contrast in either  
373 erosional efficiency or rock uplift perturbs the  $\chi$  metric (e.g., Figure 5 & 6). Specifically, in all  
374 cases, rivers within the plateau flow through significant portions of volcanic rocks, whereas

375 rivers more external to the plateau flow through less of the young volcanic sequence (Figure  
376 9B). This interpretation depends on the hypothesis that key volcanic units are more resistant to  
377 erosion, which has not been quantified in this region, but is consistent with the form of the  
378 topography (e.g. river profiles in Supplemental Figures 18, 20, 22, 24, & 26). It is also possible  
379 that differential uplift, specifically from dynamic topography, influences this pattern. For the  
380 case of the western divide separating the Kars Plateau from the Choruh watershed (KV1, KV2, &  
381 KV3), Forte et al. (2016) suggested that the topography of this region was primarily controlled  
382 by response to mantle upwelling (Zor, 2008) producing a gradient in uplift rate between the  
383 outlet and headwaters of the Choruh watershed. Ultimately, because we do not have  
384 quantitative estimates of divide mobility from catchment averaged or in-situ erosion rates, it is  
385 difficult to (1) independently know the stability of these divides or (2) link the stability of these  
386 divides to a particular cause, but it does suggest that environments such as this where the  
387 Gilbert and  $\chi$  metrics are in consistent disagreement represent important opportunities for  
388 empirically testing these metrics.

389

#### 390 **4.3. San Bernadino Mountains**

391 The San Bernadino Mountains are part of the Transverse Ranges in southern California  
392 and are bounded to the south by several strands of the San Andreas Fault (e.g., Spotila et al.,  
393 1998). The western portion of the range primarily consists of the Big Bear Plateau which is a  
394 high-elevation, low-relief region interpreted as a relict landscape developed in deeply  
395 weathered granite that is surrounded by steep escarpments on nearly every side (e.g., Blythe et  
396 al., 2000; Spotila et al., 2002, 1998). Thermochronologic data indicate that the Big Bear Plateau  
397 has been uplifted with respect to both the Mojave Desert to the north and Los Angeles Basin to  
398 the south. Prior work suggests that the escarpments surrounding the Big Bear Plateau are  
399 retreating inwards, gradually consuming the plateau (e.g., Binnie et al., 2008; Spotila et al.,  
400 2002).

401 We investigate a portion of the drainage divide that roughly defines much of the Big  
402 Bear Plateau and includes portions of both the southern and northern Big Bear escarpments as  
403 defined by Binnie et al. (2008) and segment this divide into 10 sections (Figure 10). We use

404 available cosmogenic erosion rates for this region (Binnie et al., 2008, 2007) and a relation  
405 between mean local relief within a 2.5 km radius and these catchment averaged erosion rates  
406 (Supplemental Figure 27) to produce a continuous map of erosion rate to compare to the  
407 various divide metrics (Figure 10). This result is broadly consistent with a similar map produced  
408 by Spotila et al (2002) based on low-temperature thermochronology and geologic constraints.  
409 In this region, we use a constant outlet elevation of 1100 m to calculate  $\chi$ . We choose this  
410 outlet elevation as this is approximately the effective base level to which drainages on the  
411 northern side of the San Bernadino Mountains grade, though this means a portion of drainages  
412 on the southern flank are excluded as the effective base level for these southern drainages is  
413 significantly lower in elevation.

414 Across all 10 divide segments,  $\chi$  and Gilbert metrics are both largely consistent with  
415 each other and what is predicted from the erosion rate map (Figure 10D). If we accept the  
416 erosion rate map as accurate and that contrasts in erosion rate across a divide from this map  
417 are unequivocal evidence of current or future divide motion, then despite agreement overall,  
418 there are examples of both  $\chi$  and relief failing to correctly identify divide instability given  
419 uncertainty in cross-divide differences (Figure 10D). In detail, SB2 and SB8 are cases where  $\chi$   
420 agrees with erosion rates (but not relief) and SB6 is a case where relief agrees with erosion  
421 rates (but not  $\chi$ ). There are possible interpretations of these deviations, but importantly, these  
422 are all cases where (1) a relatively small number of values are used to determine potential  
423 divide motion and (2) the determination of divide stability or mobility is dictated by how much  
424 overlap or separation in means and standard deviations are required to deem a divide stable or  
425 mobile, respectively. This highlights the utility of viewing divide metrics in forms like the  
426 histograms used here for evaluating confidence in a given determination and also suggests that  
427 there is likely a minimum segmentation length of divides below which the data is simply too  
428 noisy to make a clear determination (Figure 10D). In this case all of the divide metrics are  
429 broadly consistent with prior interpretations (e.g., Binnie et al., 2008; Spotila et al., 2002) of this  
430 region suggesting that portions of the divide along the southern and northern Big Bear  
431 escarpments are mobile and actively consuming the Big Bear Plateau (Figure 10). The only

432 stable portion of the divide appears to be between internal plateau streams and streams  
433 draining into Big Bear Lake (SB5 & SB6, Figure 10).

434 Using the model results as a means to interpret the divide metrics would suggest that all  
435 other divides are currently moving and that any spatial differences in erosional efficiency or  
436 uplift rate are absent or sufficiently small such that  $\chi$  is still a viable metric in this setting. This is  
437 consistent with known constraints from this region, specifically uniform uplift, simple bedrock  
438 geology, and unique relationships between erosion rate and mean channel steepness and  
439 erosion rate and mean hillslope gradient (e.g., Binnie et al., 2007; DiBiase et al., 2010).

440

## 441 **5. Discussion and Conclusions**

442 The results of both the simulations and field examples highlight differences in the utility  
443 of the considered metrics for assessing drainage network stability and further demonstrate that  
444 relying on any one metric is limiting. The Gilbert metrics are the best choice to assess the  
445 current status of the drainage network (i.e. are divides currently moving), whereas  $\chi$ -maps may  
446 be the best choice to assess whether a drainage network may reorganize in the future, though  
447 (1) the lack of a clear timescale that emerges from these measurements and (2) the reliance on  
448 a future and uncertain change in uplift and/or erosional efficiency gradients complicates this  
449 assessment. Among the Gilbert metrics, relief is likely the most reliable. Gradient also works  
450 well in the natural examples, but there are challenges related to the development of threshold  
451 slopes (e.g., Burbank et al., 1996) so care must be exercised when using mean gradients, and by  
452 extension local relief with small radii which mirror gradients (e.g., DiBiase et al., 2010),  
453 especially in high-relief landscapes. The elevation metric works well in model results, but in  
454 most natural cases always suggests stable divides within uncertainty using the standard  
455 deviation. We believe that there is value in considering all of the Gilbert metrics, but emphasize  
456 the importance of interrogating the results of the divide stability analysis.

457 Ultimately, using the Gilbert and  $\chi$  metrics in concert maximizes the information one  
458 can extract from the landscape with regard to drainage network stability. The strengths of the  
459 two classes of metric are not necessarily surprising. The top-down method of calculation for the  
460 Gilbert metrics means that they are largely only sensitive to changes in the hillslopes directly



461 near divides and thus represent a more ‘instantaneous’ view of the behavior of the divides. In  
462 contrast, the bottom-up method of calculation of  $\chi$ -values at divides means that they are  
463 sensitive to spatial variability in rock strength, climate, and tectonics throughout a catchment  
464 and thus represent a more integrated, ‘long-term’ view of possible influences on divide  
465 stability. These differences in scale also present different challenges in calculation. The Gilbert  
466 metrics, at least the mean upstream gradient and to a lesser extent, mean upstream local relief,  
467 have the potential to be sensitive to data resolution (e.g., Finlayson and Montgomery, 2003),  
468 where as because  $\chi$ -values only require drainage area measurements, these should be  
469 relatively insensitive to data resolution as long as flow routing algorithms are reasonably  
470 accurate. In contrast, the divide-scale of the Gilbert metrics make them entirely insensitive to  
471 any of the choice of outlet elevation issues that can potentially plague  $\chi$ -maps (e.g. Figure 2). It  
472 is also worth noting that none of the metrics are useful for explicitly illuminating past divide  
473 motion. All metrics in certain scenarios may be useful in this regard to the extent that current  
474 divide motion implies some prior history of divide motion, but because none of these metrics  
475 contain any time information, this assumption is hard to validate without independent  
476 evidence of past divide motion.

477         In addition to considering multiple metrics, more detailed analyses of differences in  
478 values across divides are necessary to fully assess divide stability. In many cases, visual  
479 differences in maps of either  $\chi$  or the Gilbert metrics seem to suggest a robust ‘anomaly’ across  
480 a divide, but the histogram of values or the uncertainty on delta values actually show significant  
481 amounts of overlap in values, e.g. divide GC2 which in map view seems to highlight an across  
482 divide difference in local relief (Figure 7A), but in detail has relatively similar values in local  
483 relief near channel heads (Figure 7D). A lingering issue is what constitutes suitable amounts of  
484 overlap in values across a divide to suggest that said divide is stable or unstable. We do not  
485 have any basis for suggesting that the criteria we primarily use (i.e. neither mean value is within  
486 one standard deviation of the other for a stable divide) is correct. Comparing predictions using  
487 the standard deviation and standard error highlights the importance of the stability criteria, as  
488 for example in the Greater Caucasus examples, using the standard deviation with the Gilbert  
489 metrics suggested mostly stable divides where as using the standard error suggests more

490 mobile divides. Generally, because standard deviations are larger than bootstrap confidence  
491 intervals which are in turn larger than standard errors, using standard deviations bias results  
492 towards stable divides (more possibility of overlap) and standard errors bias results towards  
493 mobile divides (less possibility of overlap) with bootstrap confidence intervals representing a  
494 middle ground. Choosing any estimation of uncertainty is reasonable, but we emphasize that at  
495 minimum workers should specify what criteria they are using to judge relative stability or  
496 mobility.

497         The software tools provided along with this work allow for relatively easy analysis of  
498 drainage divide stability and hopefully will aid the addition of this analysis to routine  
499 characterization of landscapes. However, this should always be done in concert with traditional  
500 landscape analyses. As described above, the presence of a  $\chi$ -anomaly along with absence of a  
501 Gilbert-anomaly at a divide indicates a spatial gradient in uplift rate, erosional efficiency, or  
502 both may exist in one or both sets of the catchments that define the divide, but it doesn't  
503 provide any information as to the nature of these gradients or their location. For this, maps of  
504 streams colored by normalized channel steepness or examining traditional longitudinal or  $\chi$ -  
505 transformed river profiles would provide more information. Thus, we primarily view these types  
506 of metrics as cursory data analysis tools to illuminate areas that necessitate deeper  
507 investigation.

508         Finally, fully testing the accuracy of different metrics of divide stability fundamentally  
509 requires comparing them to areas for which we have some constraints on erosion rates on  
510 either side of divides and thus direct information on the degree of divide mobility. Special  
511 attention should be paid to areas where the Gilbert and  $\chi$  metrics disagree, as understanding  
512 erosion rate contrasts (or the lack of contrasts) in these settings have the greatest potential to  
513 provide more general information on the utility of these metrics in different situations and thus  
514 contribute to determining the most reliable topographic expression of divide mobility.

515

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517

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523 thank Nicole Gasparini for helpful discussions regarding implementation of the tools.

524

## 525 **Figure Captions**

526

527 **Figure 1** – A) Schematic of Gilbert’s (1877) ‘Law of Unequal Declivities’, predicated on the idea  
528 that divides will move when erosion rates are not equal on either side of the divide and that  
529 this difference in erosion rate will likely be driven by differences in topographic gradient on  
530 either side of the divide. B) Reference drainage area used in all metrics for calculating across  
531 divide differences. C) Idealized form of maps of the four different divide metrics discussed in  
532 the main text in the case that they are all consistent and all indicative of divide motion to the  
533 left (Side 2). D) Corresponding plots of the distributions of values at minimum reference  
534 drainage areas. All metrics are predicated on the idea that the stable condition is nearly equal  
535 quantities on either side of the divide, however the prediction of motion direction based on  
536 across divide differences is different for the different metrics. For  $\chi$  and elevation metrics, the  
537 divide should move towards the side with higher values, whereas for relief and gradient, the  
538 divide should move towards the side with lower values. E) Comparison of delta values for all  
539 four metrics with propagated uncertainties normalized such that positive and negative delta  
540 values indicate the same direction of divide motion across all metrics. If any portion of the  
541 mean or its uncertainty overlaps with the stable divide line, then we assume the divide is  
542 stable. Bars are considering standard deviation as the uncertainty, shaded boxes the standard  
543 error. Though not shown, bootstrap confidence intervals would be intermediate.

544

545 **Figure 2** – Example from the Greater Caucasus Mountains of the complications of the sensitivity  
546 of  $\chi$  values to the choice of outlet elevation. Extent of maps are all the same, heavy black line is

547 divide between rivers flowing to the Caspian (base level -27m ASL) and Black (base level 0m  
548 ASL) Seas, thin black line is divide between rivers flowing north and south. Solid black arrows  
549 indicate general flow direction for streams on either side of divides, hollow arrows indicate  
550 interpreted direction of divide motion from  $\chi$ -anomalies. Maps show  $\chi$  calculated as a  
551 continuous grid with base level defined as A) true base level, i.e. calculated from river mouths  
552 at either the Black or Caspian Seas, B) a constant elevation of 550m ASL, which approximates  
553 the range-front for most of the Greater Caucasus, and C) an estimate of the bedrock-alluvial  
554 transition based on manual clipping of the DEM. All three outlet elevations are equally valid,  
555 but suggest markedly different stability for the main divides.

556

557 **Figure 3** – Selected results from the asymmetric uplift model. Left side are plots of average  
558 delta values of divide migration rate compared to A) erosion rate, B) channel head elevation, C)  
559 local 250m relief, D) gradient and E)  $\chi$ . Points are colored by the model time step with open  
560 circles during the asymmetric uplift phase and filled circles during the uniform uplift phase.  
561 Right side are maps of selected portions of the landscape during 5 key time frames and from  
562 top to bottom, colored by A) erosion rate, B) channel elevation, C) upstream mean relief, D)  
563 upstream mean gradient, and E)  $\chi$ . In top erosion rate panels, the rate and direction of divide  
564 motion is displayed as a vector, with divide motion towards the top of the page defined as  
565 positive.

566

567 **Figure 4** – Divide metric histograms for three key timesteps during the asymmetric uplift model:  
568 A) near peak divide migration rate during the tilt phase (2.0 Myr); B) stable divide near the end  
569 of the tilt phase (19.8 Myr); and C) near peak divide migration rate during the return phase  
570 (21.4 Myr).

571

572 **Figure 5** – Selected results from the dipping hard layer model, plot setup is nearly identical to  
573 Figure 3, except the top map panel is split between erosion rate on the left and rock strength  
574 on the right. The erosional efficiency,  $K$ , of the hard layer is 0.25 times the rest of the landscape.

575

576 **Figure 6** – Divide metric histograms for two key timesteps during the dipping hard layer model:  
577 A) near peak divide migration rate before the hard layer reaches the divide (3.5 Myr) and B)  
578 when the divide and the hard – soft contact are coincident (6.0 Myr).

579

580 **Figure 7** – Divide stability analysis of the southeastern Greater Caucasus drainage divide. For  
581 this analysis, we use a constant outlet elevation of 550m for calculating  $\chi$  (e.g. Figure 2B). A)  
582 Stream network colored by mean upstream relief superimposed on a continuous  $\chi$  grid draped  
583 over a hillshade. White line is the divide, black squares mark boundaries between divide  
584 segments and small inset shows nomenclature for the divide segments. Black box shows outline  
585 of Figure 8A. B) Hillshade colored by elevation of the same area for context. C) Standardized  
586 delta plot for the 8 segments along the divide. Bars are considering standard deviation as the  
587 uncertainty, shaded boxes the standard error.

588

589 **Figure 8** – Evidence of northward divide motion in the eastern Greater Caucasus. A)  
590 Topography near the divide with a possible capture highlighted, see Figure 7 for location within  
591 context. When identifying potential former captures from ‘area-gain’ signatures in  $\chi$ -  
592 normalized profiles, a viable former connection such as this is an essential observation given  
593 the extremely short time-scale of preservation for such signatures within topography (Whipple  
594 et al., 2017c). B)  $\chi$ -transformed profiles of the two drainages highlighted in 8A. Section of the  
595 south flowing drainage shows characteristic ‘area-gain’ signature below a possible captured  
596 reach. C) Longitudinal profile of the two drainages highlighted in 8A.

597

598 **Figure 9**– Kars volcanic plateau. For this analysis, we use a constant outlet elevation of 550m  
599 for calculating  $\chi$ . A) Stream network colored by mean upstream relief on top of continuous  $\chi$   
600 grid and hillshade. While lines indicate divides, black squares are boundaries between divide  
601 segments. Inset in top left shows labels for the divide segments. B) Simplified geologic map  
602 from Forte et al. (2016), area is same as in A, divides shown for reference. C) Hillshade colored  
603 by elevation for the Kars area with labels of features discussed in the main text. D) Standardized

604 delta plot for the 5 segments along the divide. Bars are considering standard deviation as the  
605 uncertainty, shaded boxes the standard error.

606

607 **Figure 10** – Western San Bernadino Mountains. A) Streams colored by mean upstream relief on  
608 top of a continuous  $\chi$  grid draped over a hillshade. White lines mark divides of interest, black  
609 squares show boundaries between divide segments. Inset in top left show names for divide  
610 segments. B) Interpolated erosion rate map based on cosmogenic erosion rate data from Binnie  
611 et al. (2008, 2007), see text and supplement for additional discussion. C) Hillshade colored by  
612 elevation of the western San Bernadino Mountains with labels for important features discussed  
613 in text. D) Standardized delta plot for the 10 segments along the divide. Bars are considering  
614 standard deviation as the uncertainty, shaded boxes the standard error.

615

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745

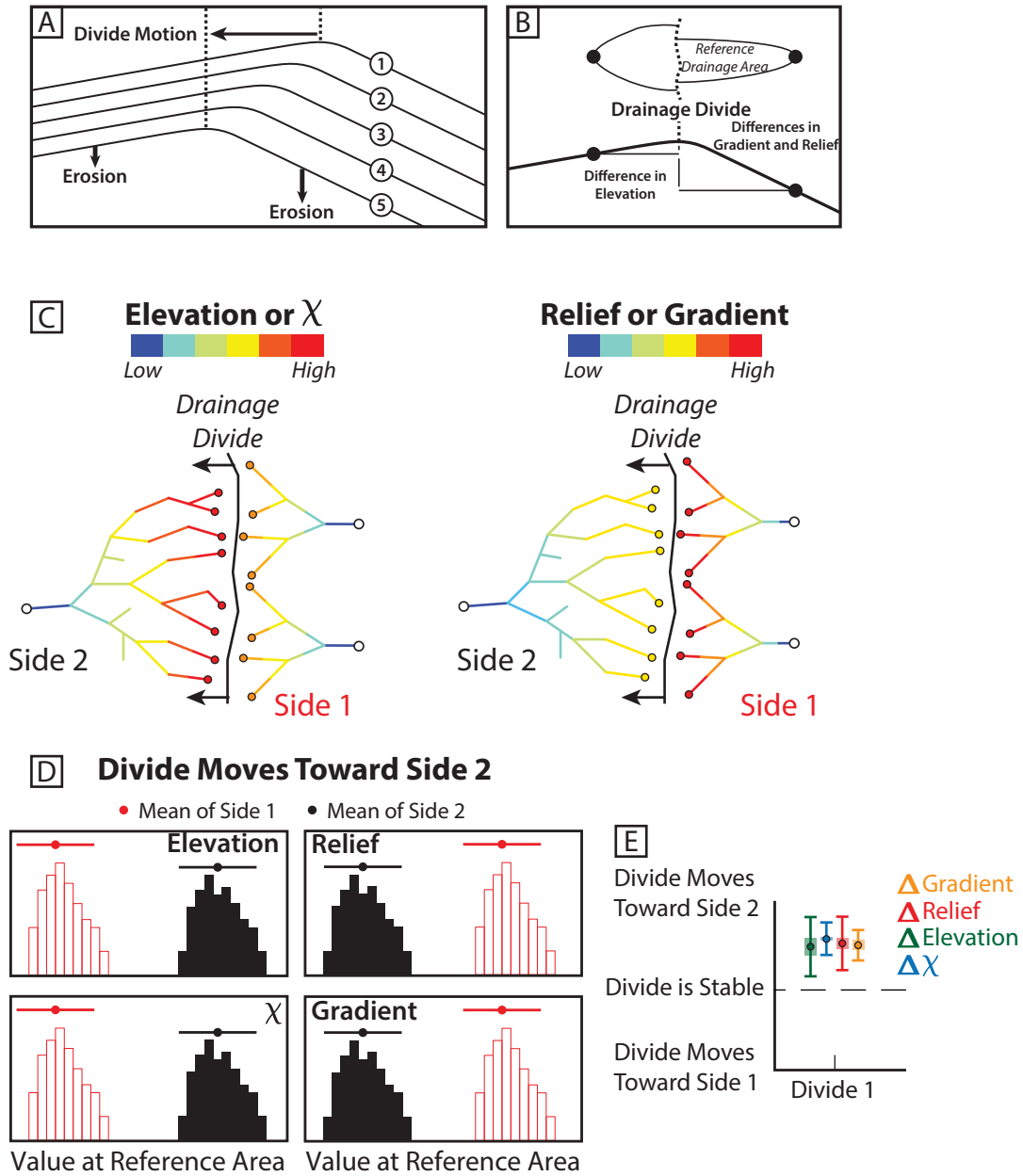


Figure 1

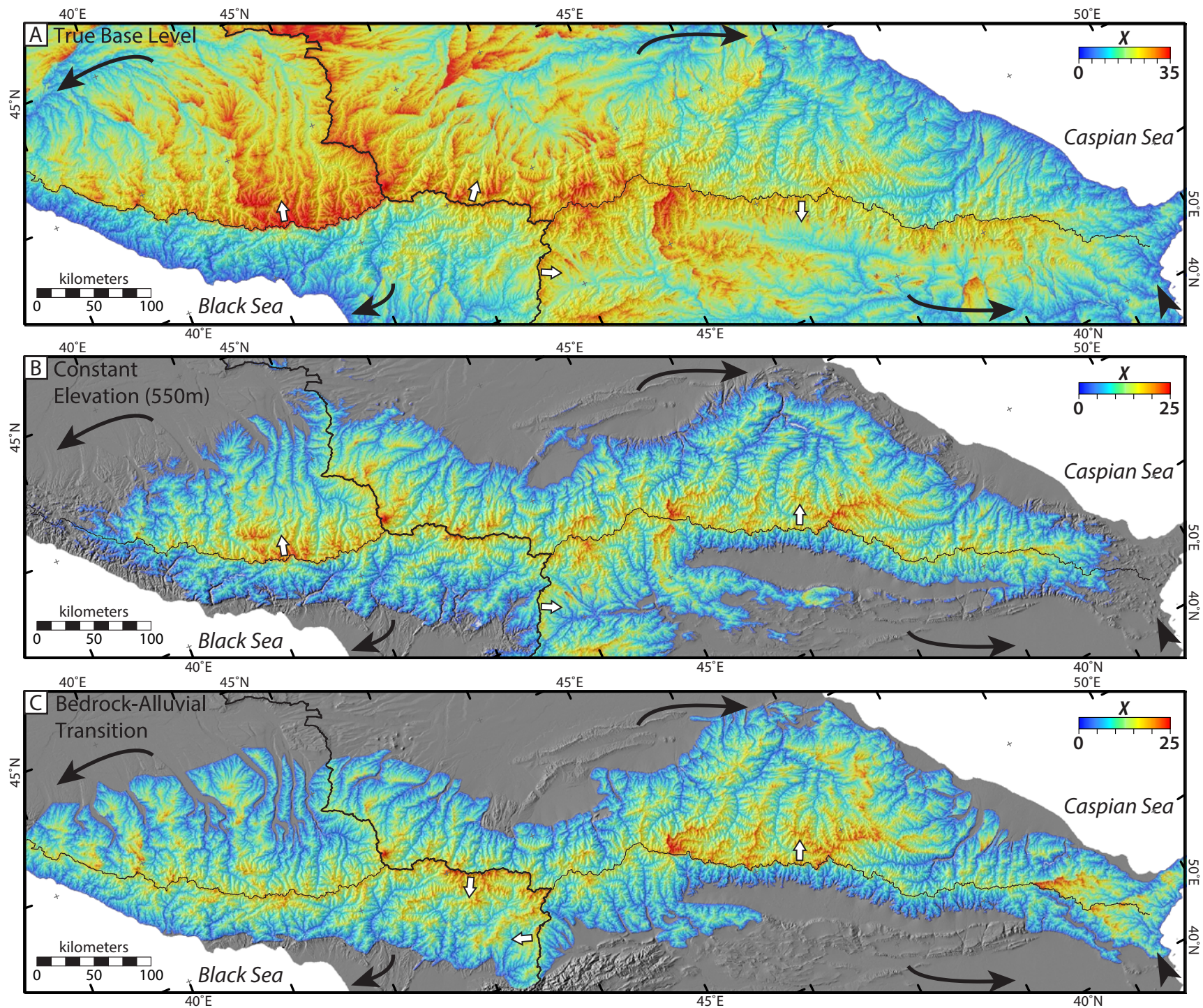


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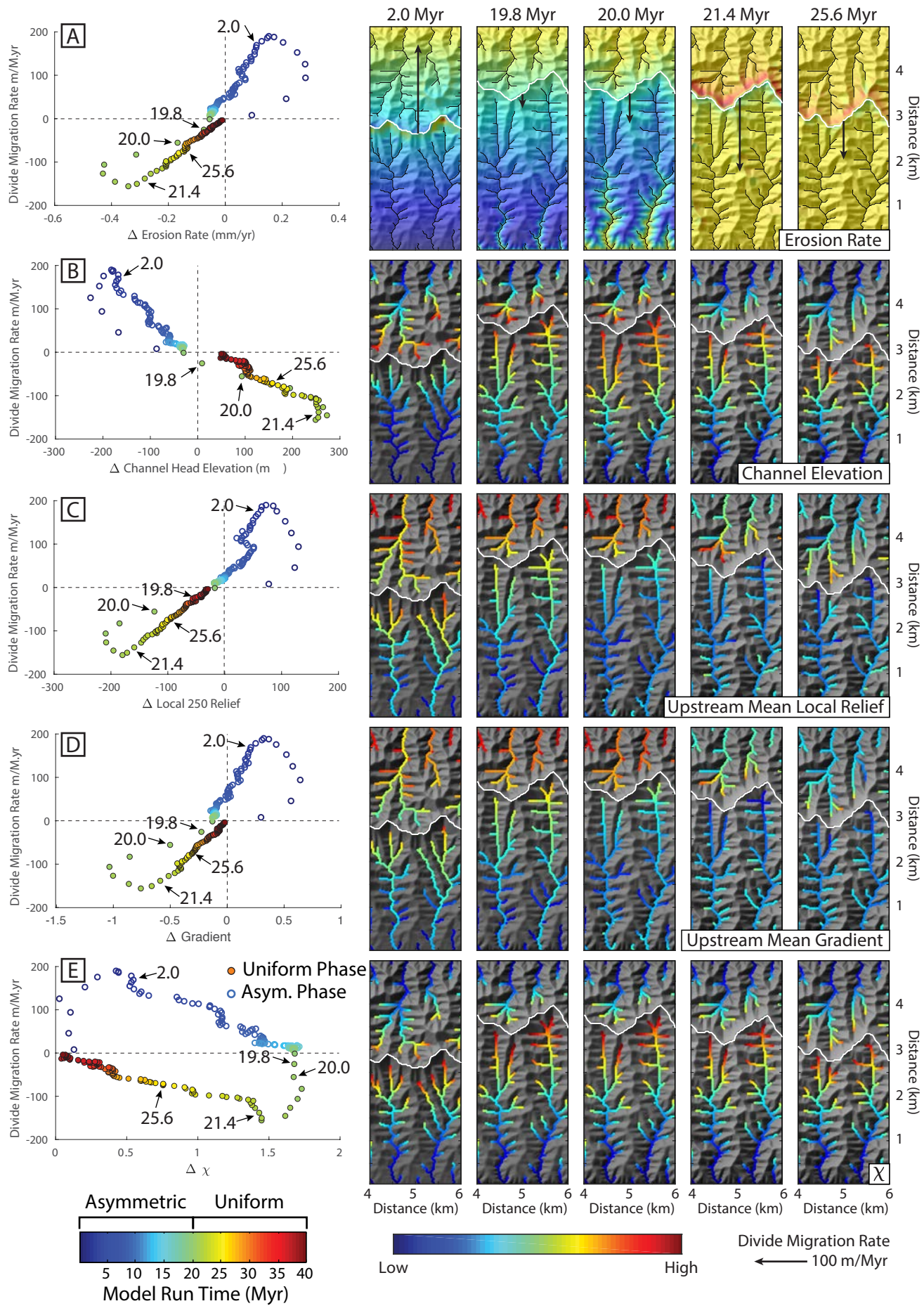


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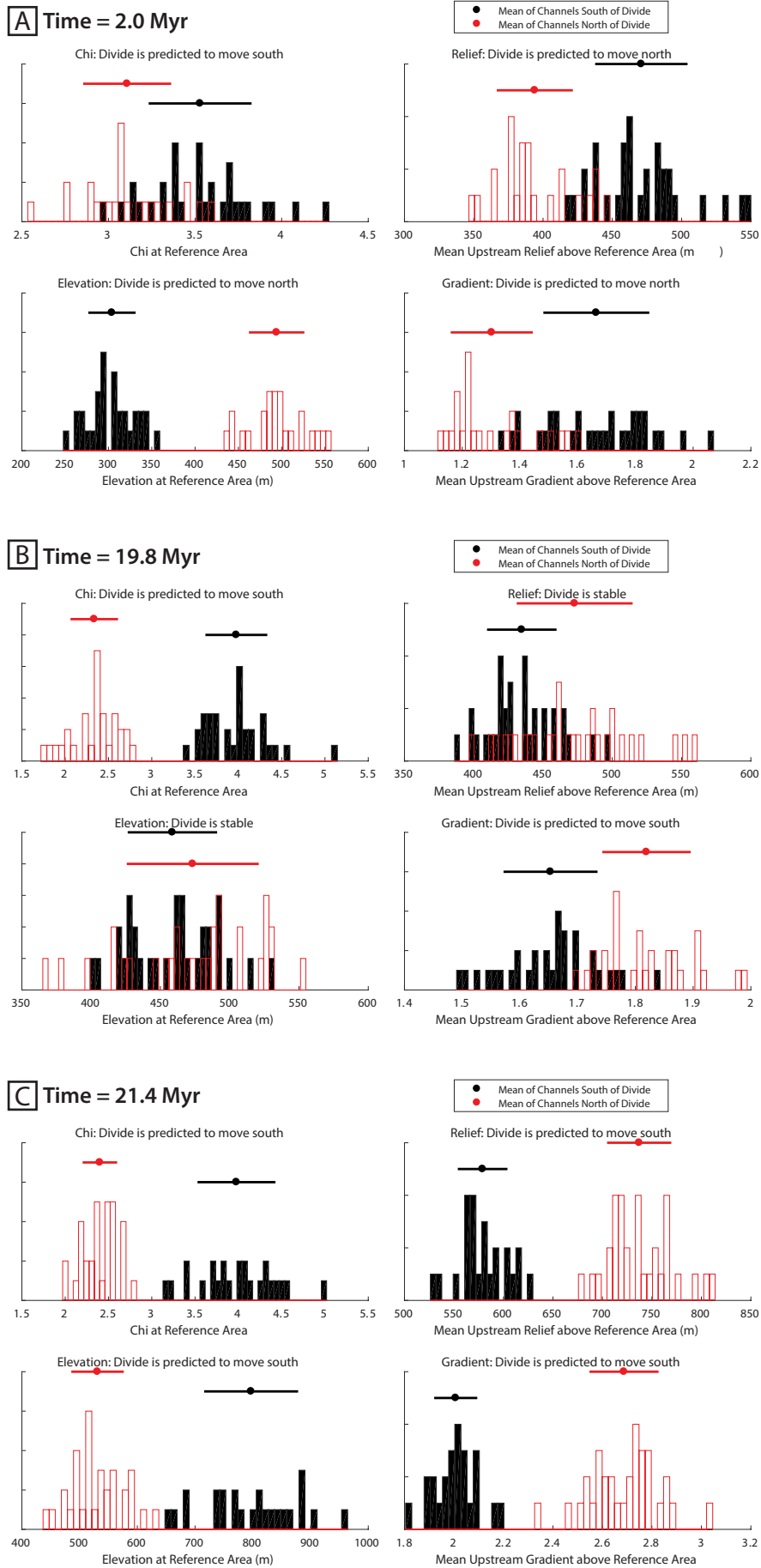


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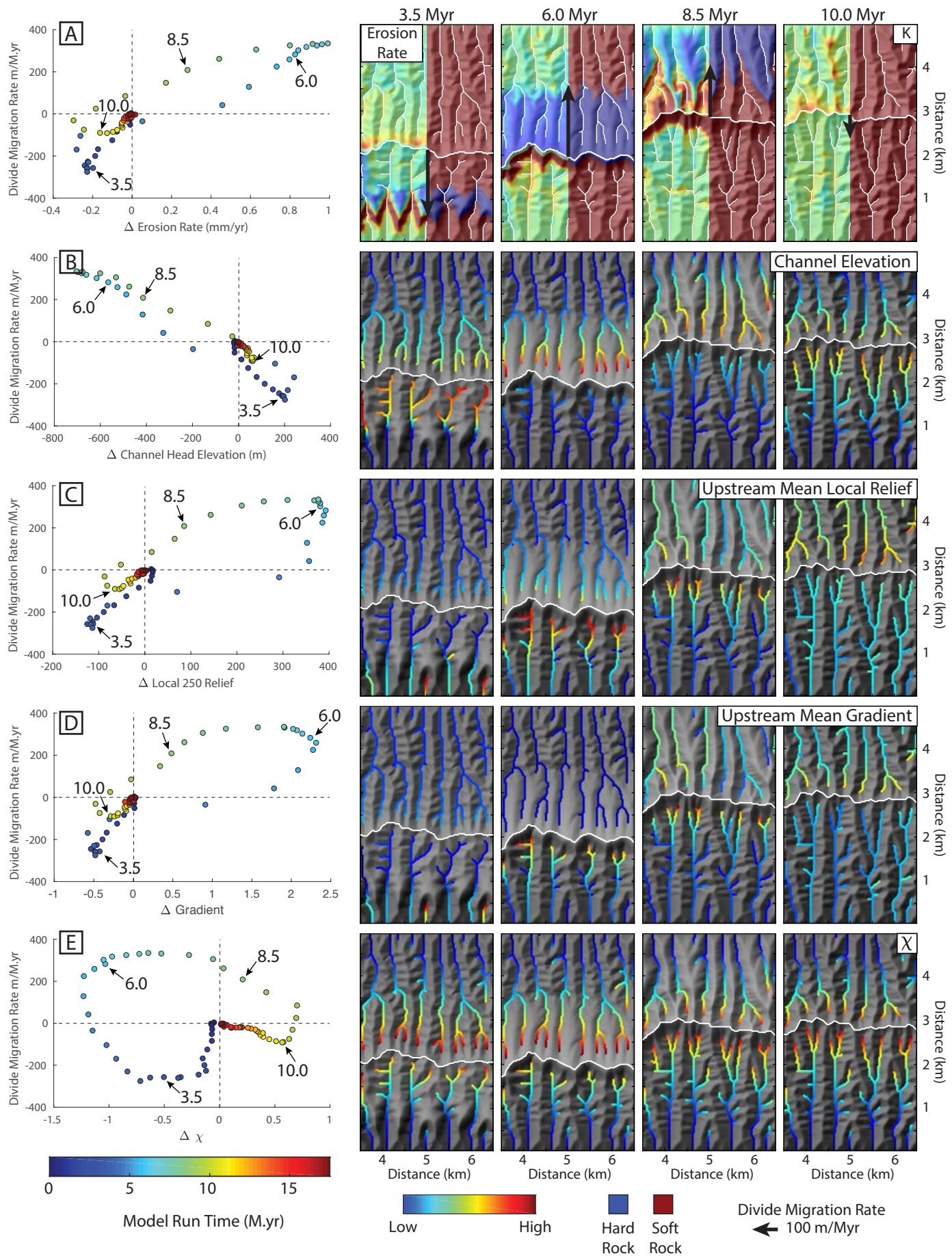


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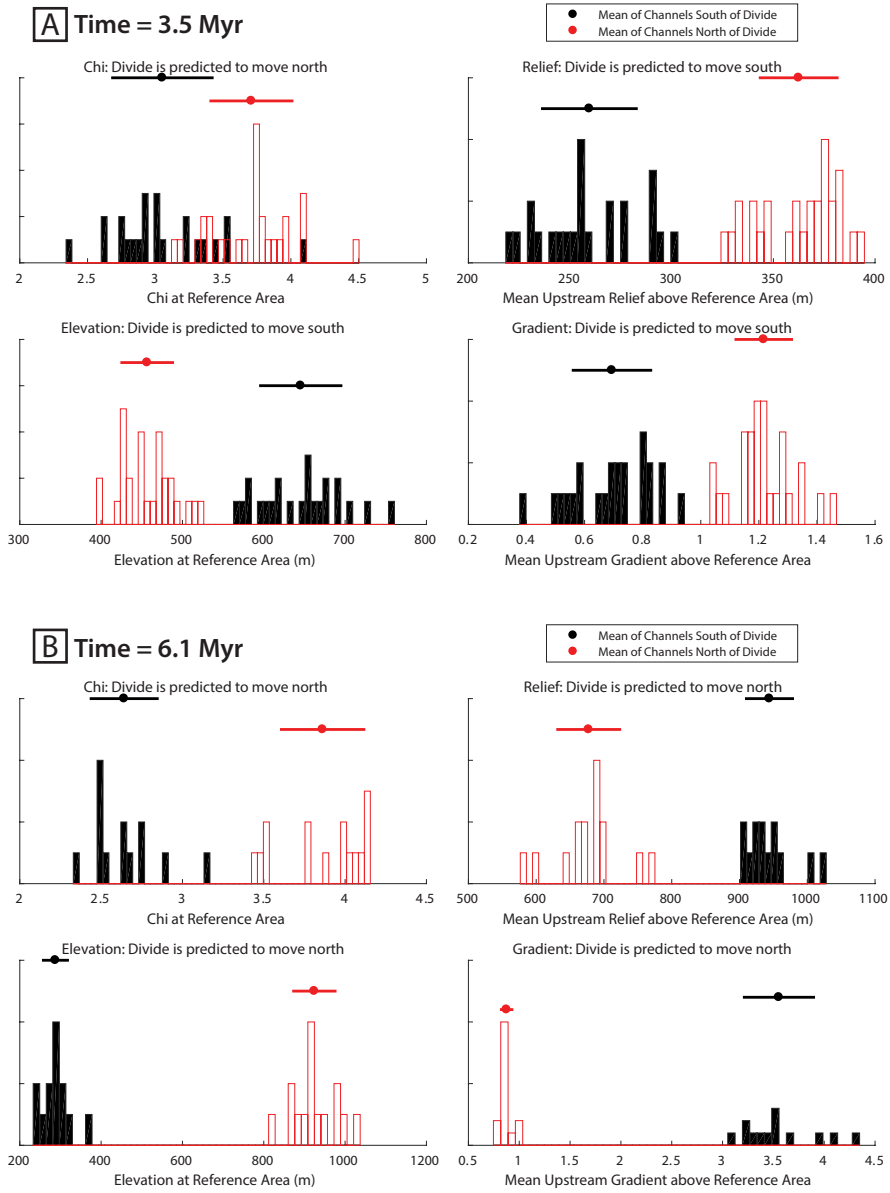


Figure 6



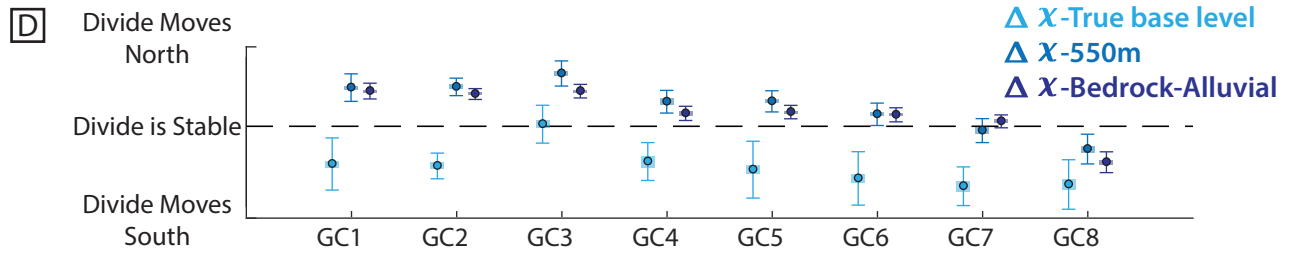
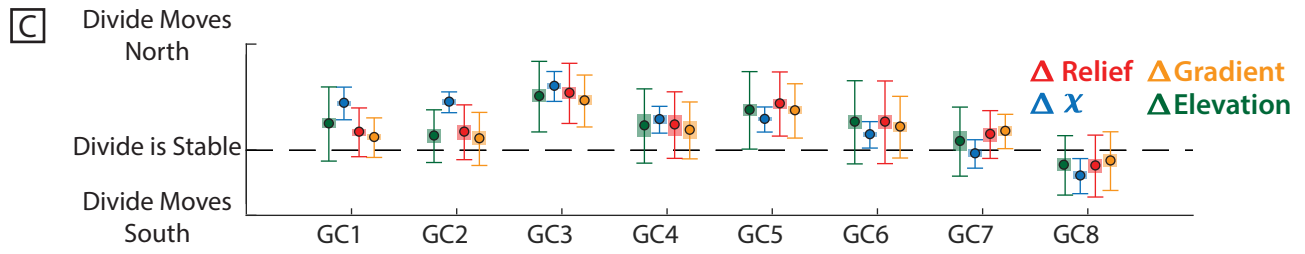
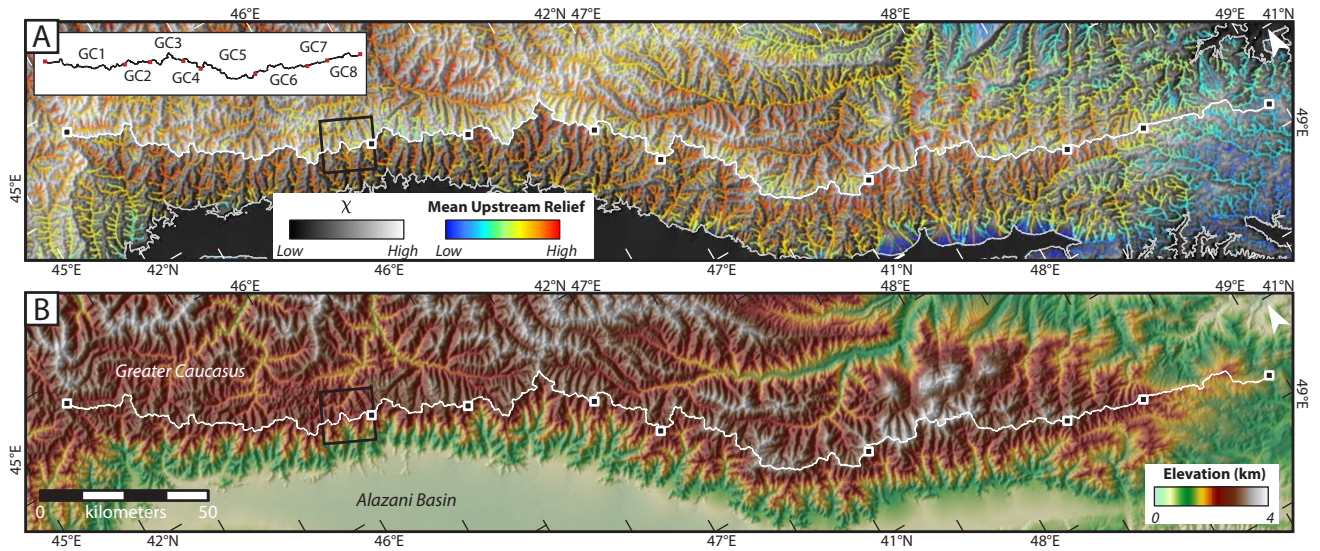


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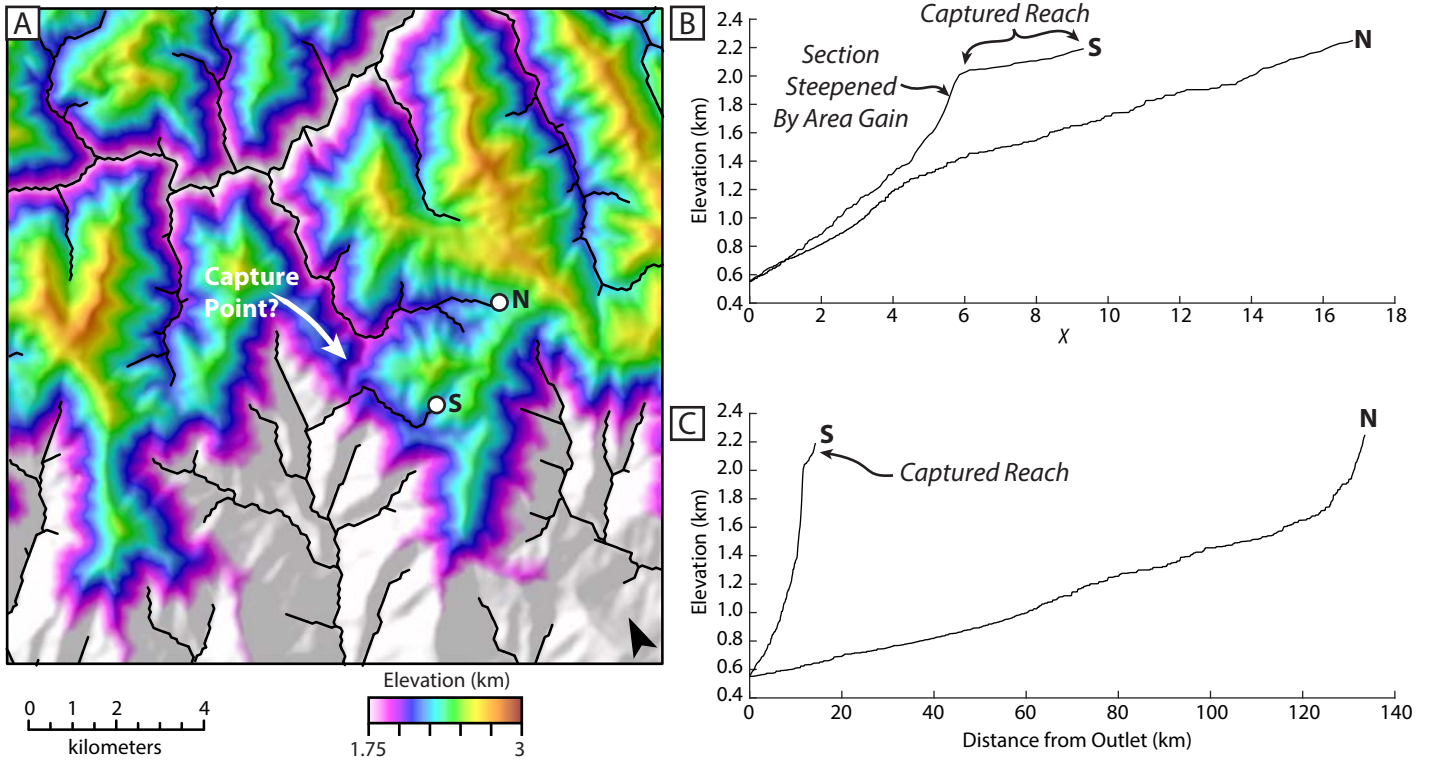


Figure 8

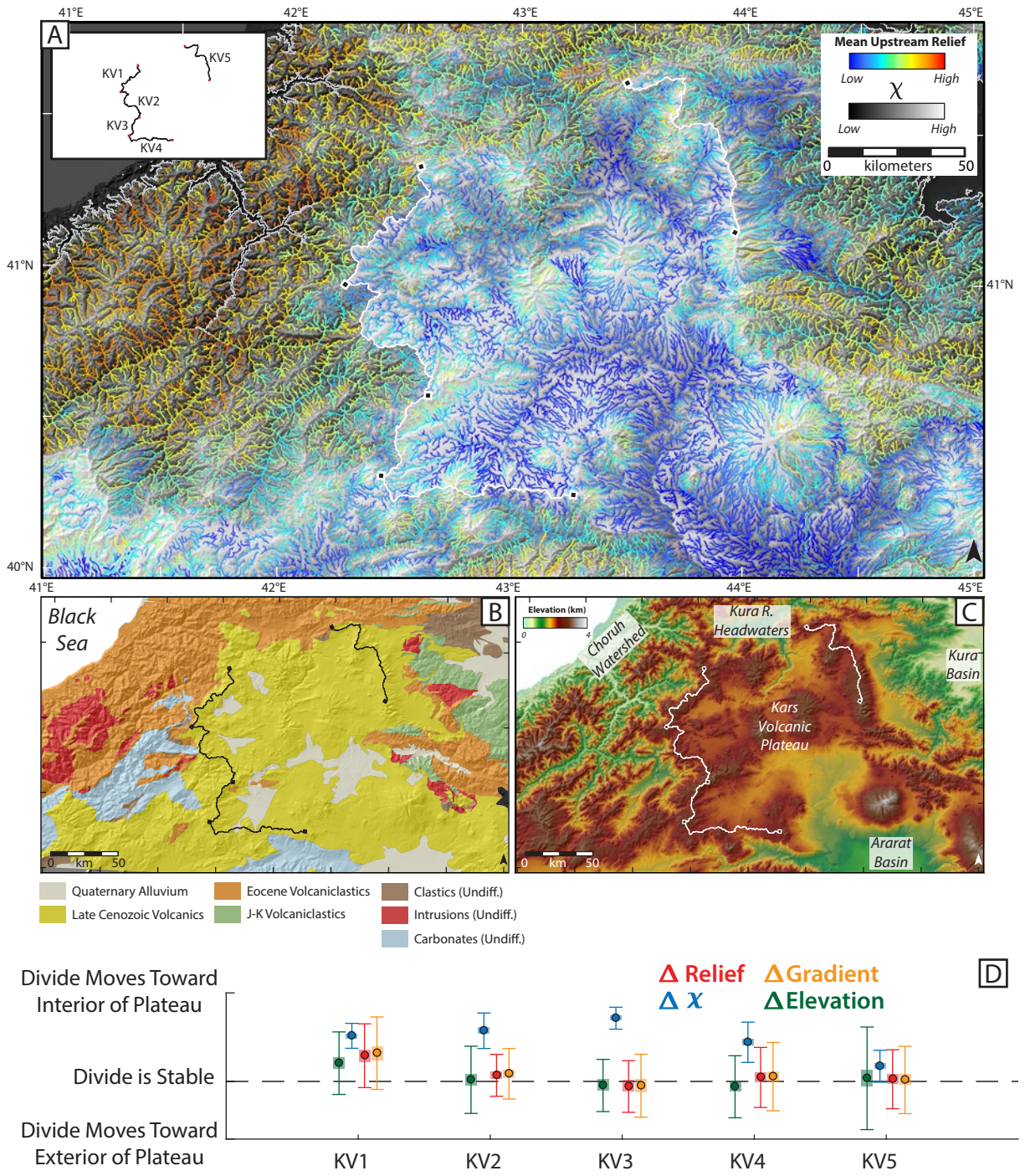


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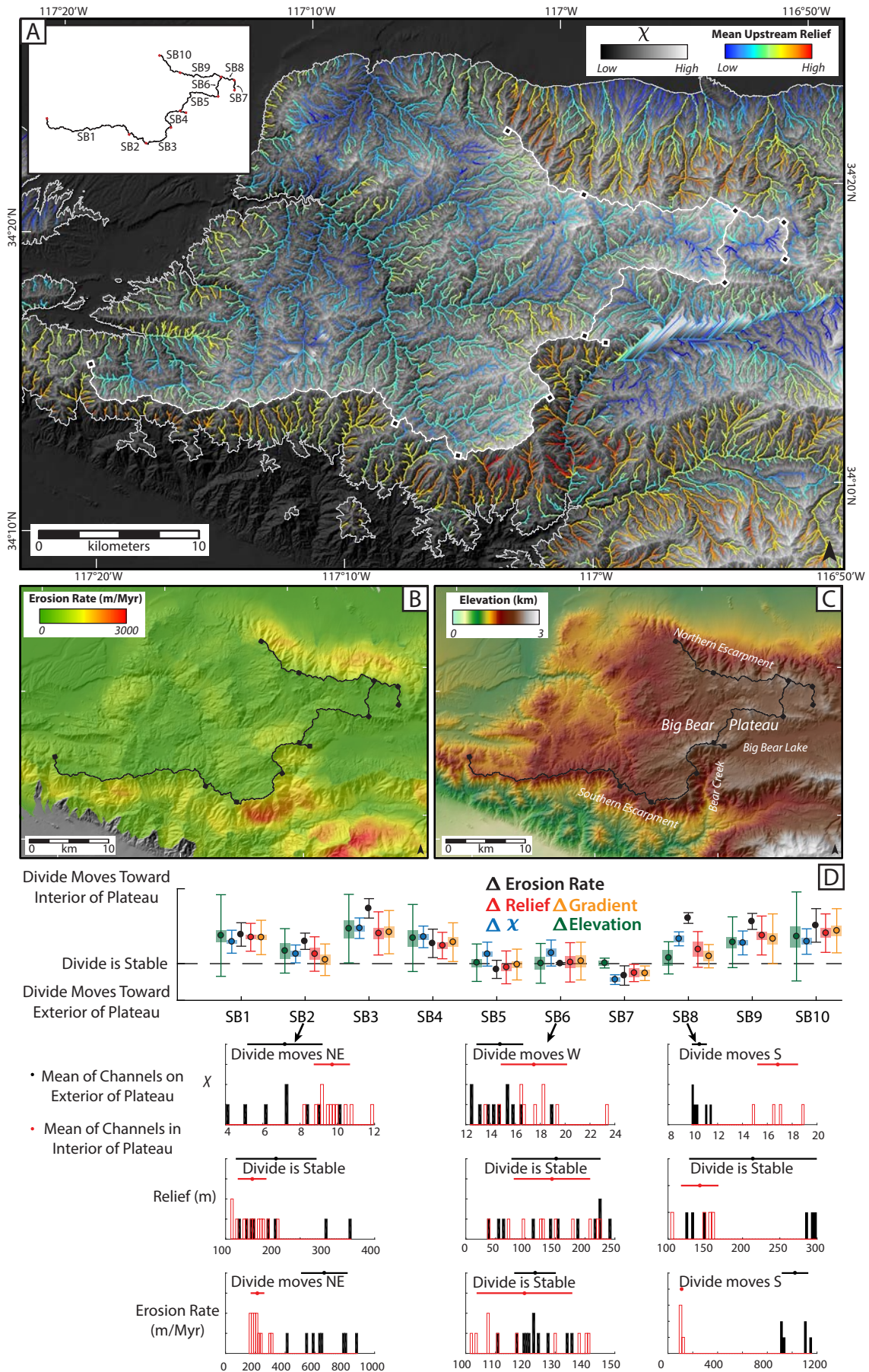


Figure 10

## Supplemental Text for ‘Criteria and Tools for Determining Drainage Divide Stability’ by Adam M. Forte & Kelin X. Whipple

### S1. Description of codes included in Github repository

**S1.1. *DivideStability*** – For a given area, this routine produces shapefiles (or alternatively rasters convertible to shapefiles within a GIS program) that includes a stream network with values for all four metrics ( $\chi$ , channel elevation, upstream mean relief, and upstream mean gradient) so that the user can produce maps of stream networks colored by these quantities (e.g. Figure 1). To aid in sensible color scaling, values for the ‘Gilbert metrics’ are normalized to vary between 0 and 1. Because the calculation of  $\chi$  is sensitive to the choice of outlet elevation (e.g. Figure 2), careful control of outlet elevation is essential for meaningful interpretations of  $\chi$ -anomalies. For this reason, this function allows the user to modify a stream network to remove portions of streams below a minimum elevation. This function also checks to make sure that stream outlets along the edge of the DEM meet the criteria defined and that all drainage basins are ‘complete’, i.e. that the summation of drainage area is accurate and is not influenced by tributaries that are cut off. Either of these cases could result in artificial  $\chi$ -anomalies, but generally should have no effect on the Gilbert metrics.

**S1.2. *ChiGrid*** – The *DivideStability* code calculates  $\chi$  along the stream network, but we find it useful to be able to visualize  $\chi$  and an additional metric simultaneously. This code calculates  $\chi$  at every pixel in the DEM so that colored stream networks can overlay this  $\chi$  raster. Similar to *DivideStability*, a minimum elevation can be specified for calculating the  $\chi$  raster and a check is performed to ensure included drainage basins are complete.

**S1.3. *AcrossDivide*** – Tool uses the output of *DivideStability* and allows users to select sections of a divide of interest to perform detailed analysis of divide sections (e.g. Figure 1D). This function provides users multiple ways of defining a divide of interest, but all of them generally function on the idea that the user (or the function itself in the case of automated detection schemes) defines divides of interest by selecting the drainages that define this divide. End results are plots of the distribution and means of the values at the reference drainage area of all four metrics on either side of the divide of interest (e.g. Figure 1) along with a prediction from each metric independently regarding whether and in which direction the divide should move. The prediction of a divide stability or mobility is made on the basis of a user selected assessment of uncertainty and whether the uncertainty of the distributions overlap with the means of the opposing side of the divide. If there is overlap, the divide is considered stable, and if there is no overlap, the divide is considered mobile. This is not meant as an absolute criteria, simply a quick first order assessment. The user can choose to use the standard deviation of the population (default), the standard error on the mean, or the 95% bootstrap confidence interval determined from a 1000 iteration resampling scheme. This function also produces a list of channel head coordinates and their respective values for the four metrics that define the divide of interest.

**S1.4. PlotDivideProfiles** – To understand the predicted behavior of a divide, it is often necessary to consider the longitudinal profiles of the rivers in question. This function plots  $\chi$ -elevation and distance-elevation plots for the streams used to define the divide. Various plotting options exist to allow the user to plot only specific channels and to color drainages by either gradient or relief to compare predictions of individual metrics.

**S1.5. AlongDividePlot** – If the user has defined multiple divide segments, this allows them to produce a plot similar to what is shown in the text (e.g. Figures 7C, 9D, or 10D). In detail, this function will produce three plots for each divide (made up of multiple segments), (1) a plot of divide segment means with uncertainties, (2) a plot of across-divide delta values with propagated uncertainties with true values (i.e. relief and gradient will have opposite signs from elevation and chi delta values if they are all consistent), and (3) a standardized plot of delta values to that ‘positive’ values indicate the same direction of divide motion for all metrics. Similar to *AcrossDivide*, the user can choose to use the standard deviation of the populations, standard error on the mean, or 95% bootstrap confidence interval as the uncertainty value.

## **S2. Captions for supplemental figures**

**Supplemental Figure 1** – Divide stability histograms for divide GC1.

**Supplemental Figure 2** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC1.

**Supplemental Figure 3** – Divide stability histograms for divide GC2.

**Supplemental Figure 4** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC2.

**Supplemental Figure 5** – Divide stability histograms for divide GC3.

**Supplemental Figure 6** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC3.

**Supplemental Figure 7** – Divide stability histograms for divide GC4.

**Supplemental Figure 8** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC4.

**Supplemental Figure 9** – Divide stability histograms for divide GC5.

**Supplemental Figure 10** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC5.

**Supplemental Figure 11** – Divide stability histograms for divide GC6.

**Supplemental Figure 12** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC6.

**Supplemental Figure 13** – Divide stability histograms for divide GC7.

**Supplemental Figure 14** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC7.

**Supplemental Figure 15** – Divide stability histograms for divide GC8.

**Supplemental Figure 16** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide GC8.

**Supplemental Figure 17** – Divide stability histograms for divide KV1.

**Supplemental Figure 18** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide KV1.

**Supplemental Figure 19** – Divide stability histograms for divide KV2.

**Supplemental Figure 20** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide KV2.

**Supplemental Figure 21** – Divide stability histograms for divide KV3.

**Supplemental Figure 22** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide KV3.

**Supplemental Figure 23** – Divide stability histograms for divide KV4.

**Supplemental Figure 24** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide KV4.

**Supplemental Figure 25** – Divide stability histograms for divide KV5.

**Supplemental Figure 26** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide KV5.

**Supplemental Figure 27** – Empirical relationship between Be10 erosion rate data and local 2.5 km relief in the San Bernadino Mountains used to produce the erosion rate map in Figure 10B.

**Supplemental Figure 28** – Divide stability histograms for divide SB1.

**Supplemental Figure 29** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB1.

**Supplemental Figure 30** – Divide stability histograms for divide SB2.

**Supplemental Figure 31** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB2.

**Supplemental Figure 32** – Divide stability histograms for divide SB3.

**Supplemental Figure 33** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB3.

**Supplemental Figure 34** – Divide stability histograms for divide SB4.

**Supplemental Figure 35** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB4.

**Supplemental Figure 36** – Divide stability histograms for divide SB5.

**Supplemental Figure 37** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB5.

**Supplemental Figure 38** – Divide stability histograms for divide SB6.

**Supplemental Figure 39** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB6.

**Supplemental Figure 40** – Divide stability histograms for divide SB7.

**Supplemental Figure 41** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB7.

**Supplemental Figure 42** – Divide stability histograms for divide SB8.

**Supplemental Figure 43** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB8.

**Supplemental Figure 44** – Divide stability histograms for divide SB9.

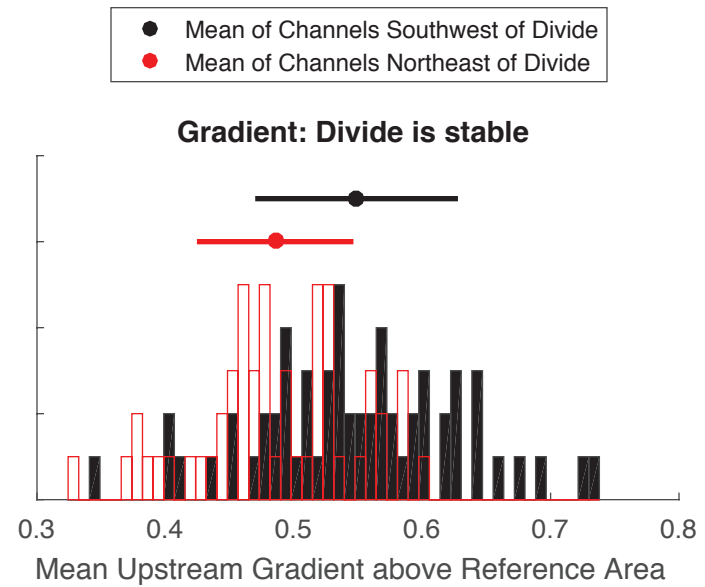
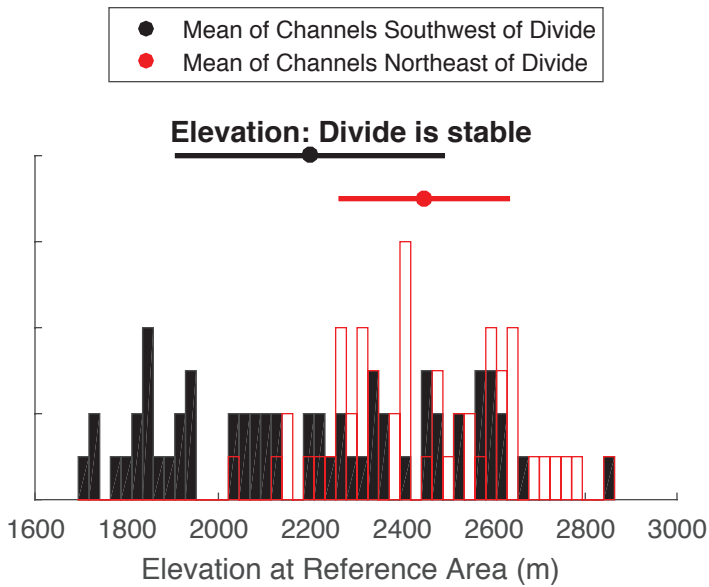
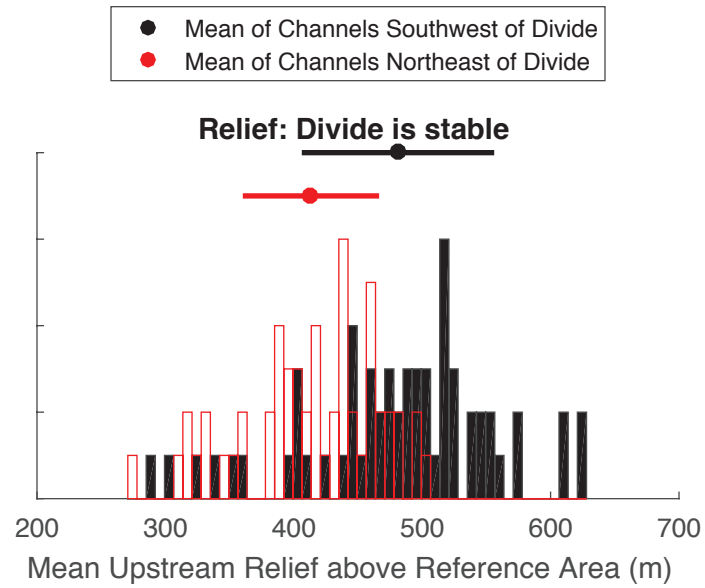
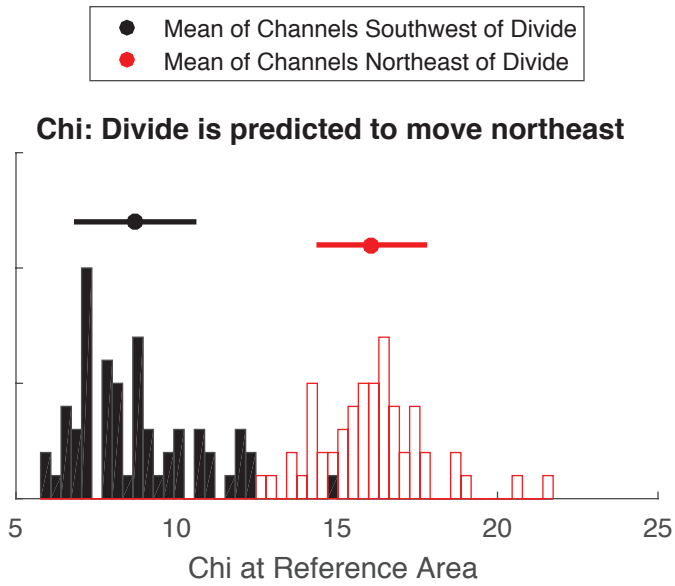


**Supplemental Figure 45** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB9.

**Supplemental Figure 46** – Divide stability histograms for divide SB10.

**Supplemental Figure 47** – Longitudinal and  $\chi$ -normalized profiles for rivers that define divide SB10.

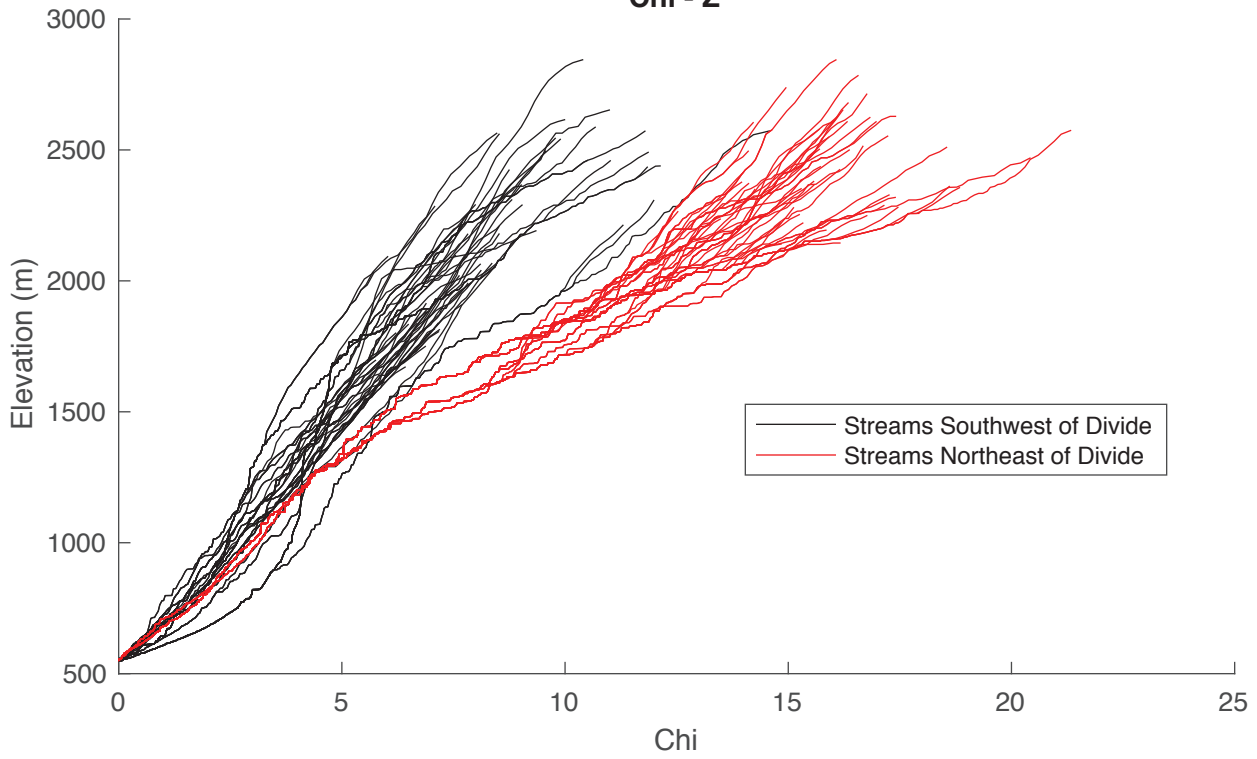
## Divide GC1



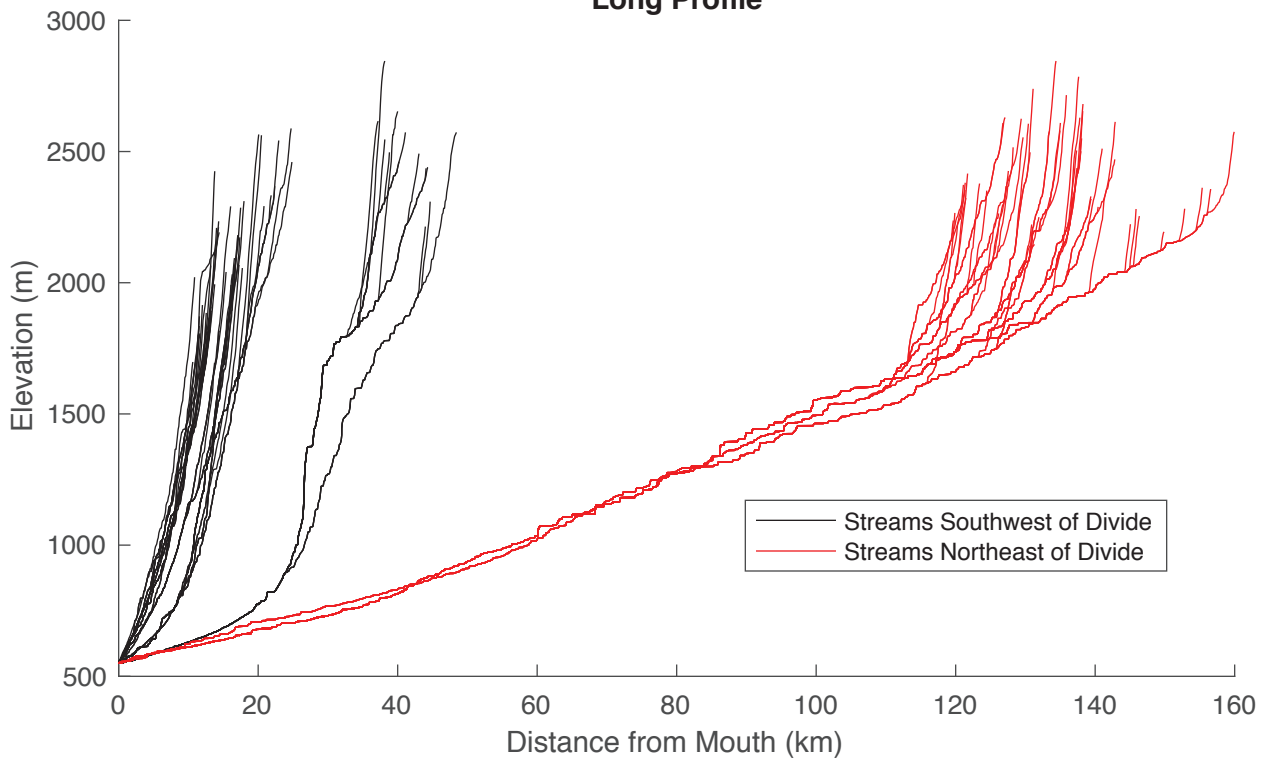
**Supplemental Figure 1**

### Divide GC1

Chi - Z

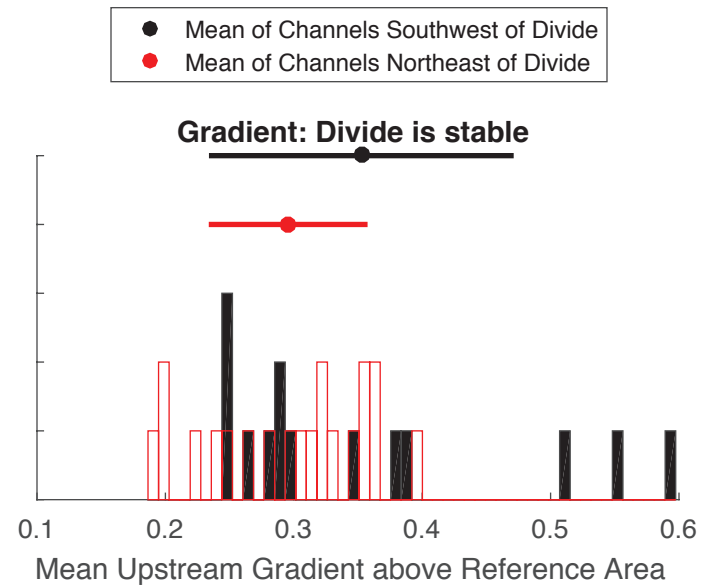
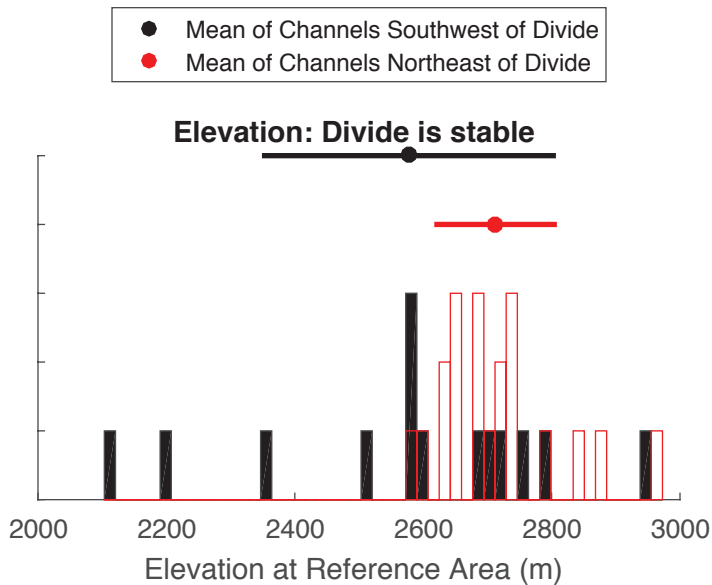
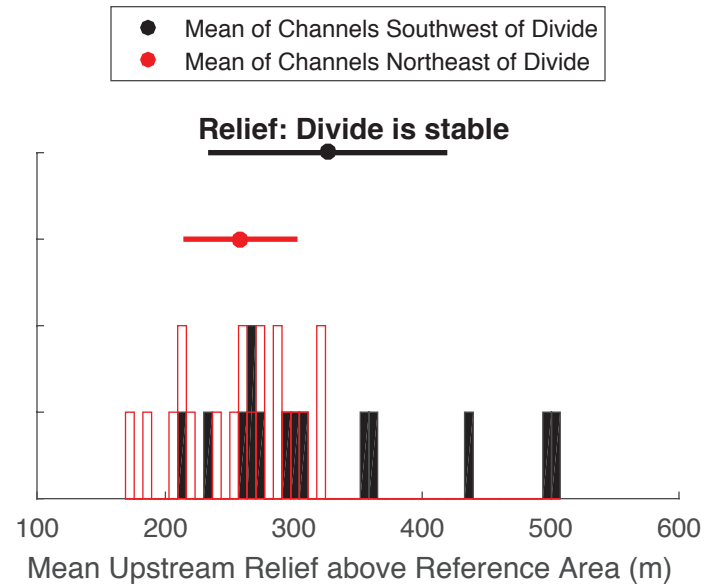
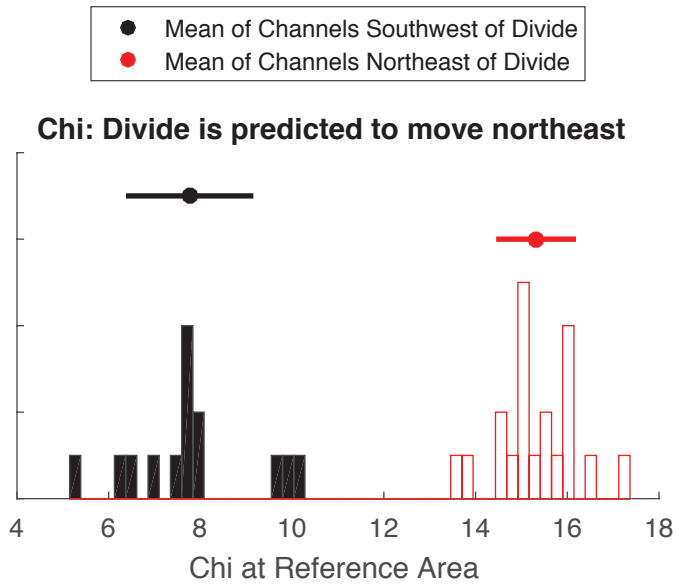


### Long Profile



**Supplemental Figure 2**

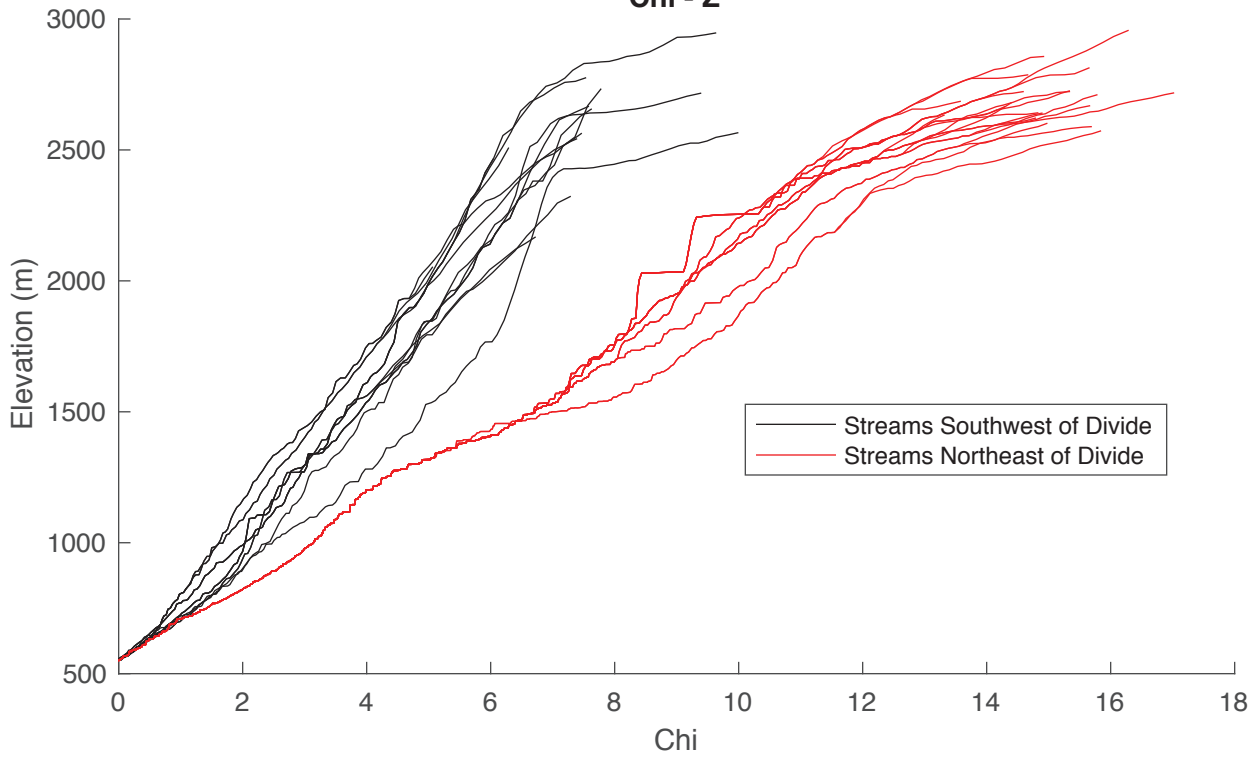
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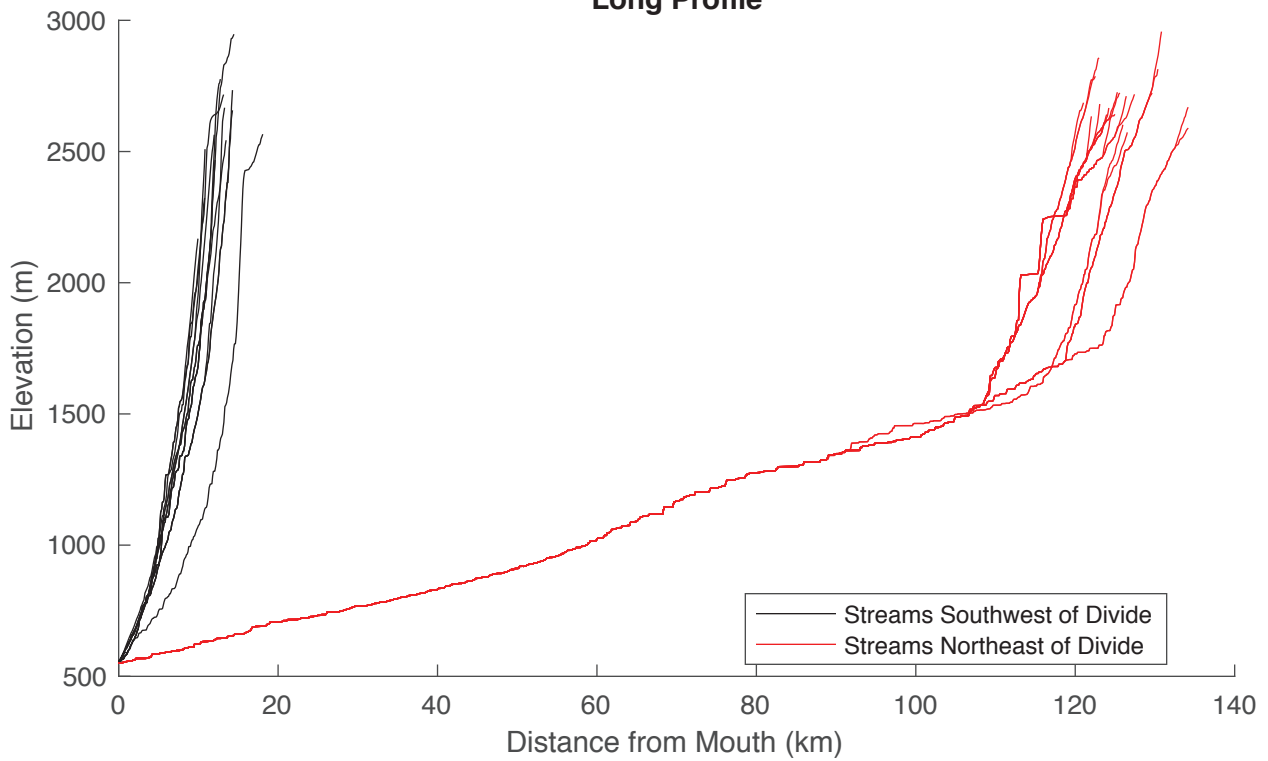
**Supplemental Figure 3**

## Divide GC2

Chi - Z

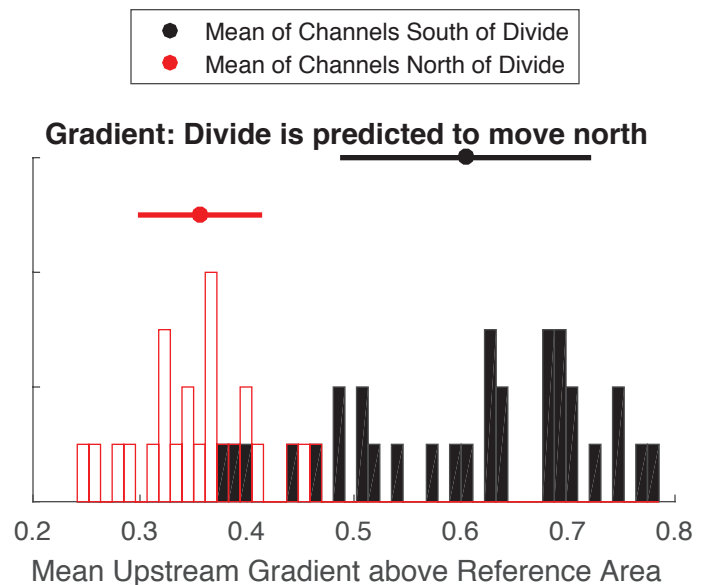
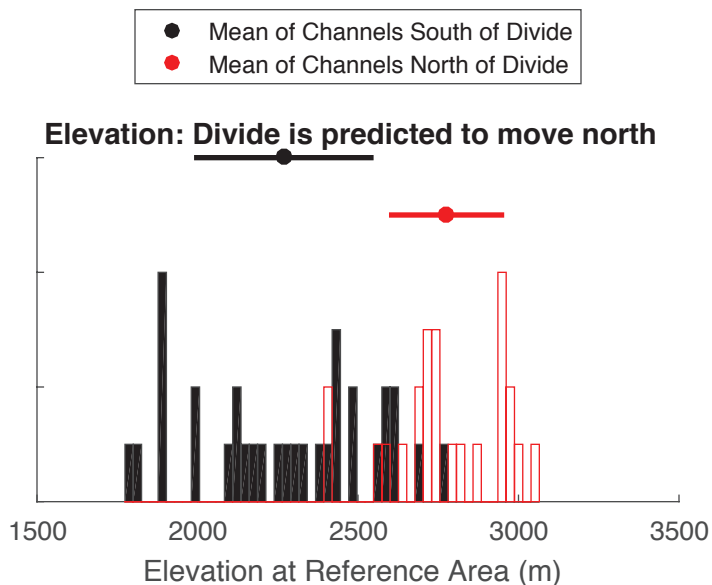
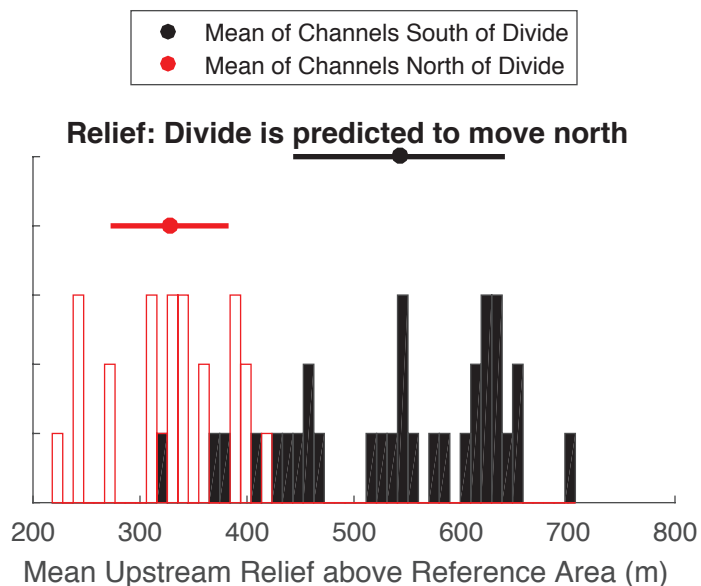
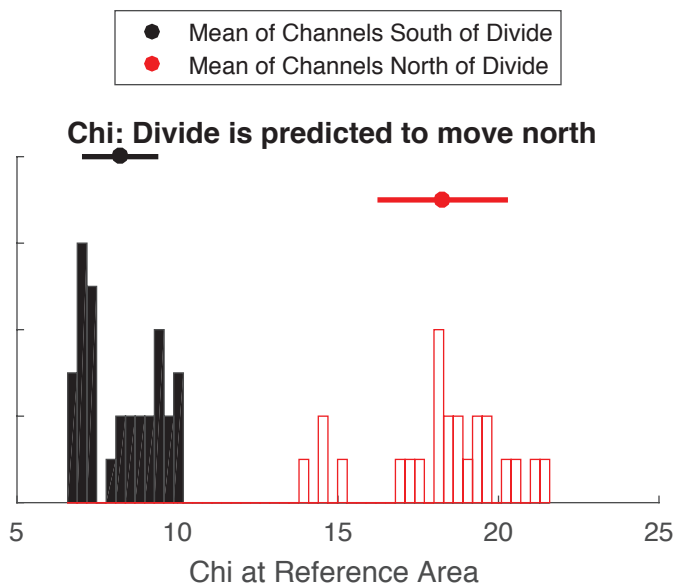


## Long Profile



**Supplemental Figure 4**

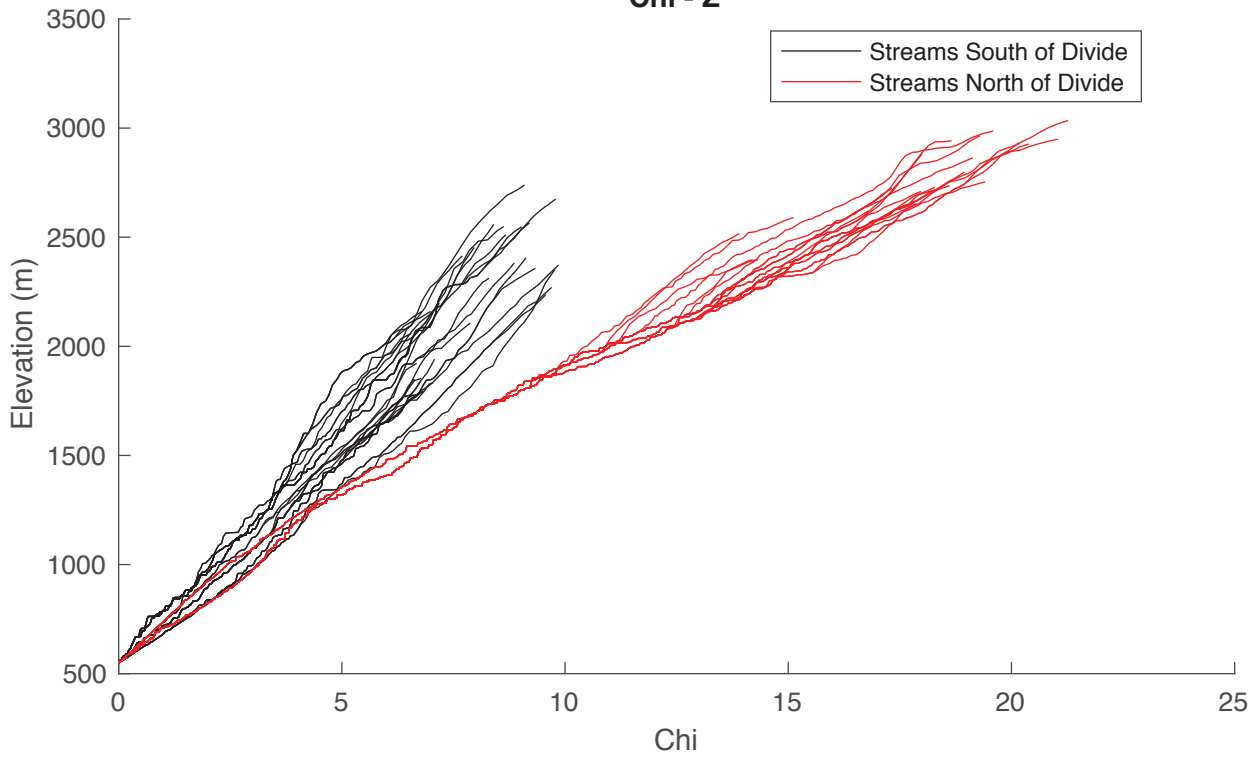
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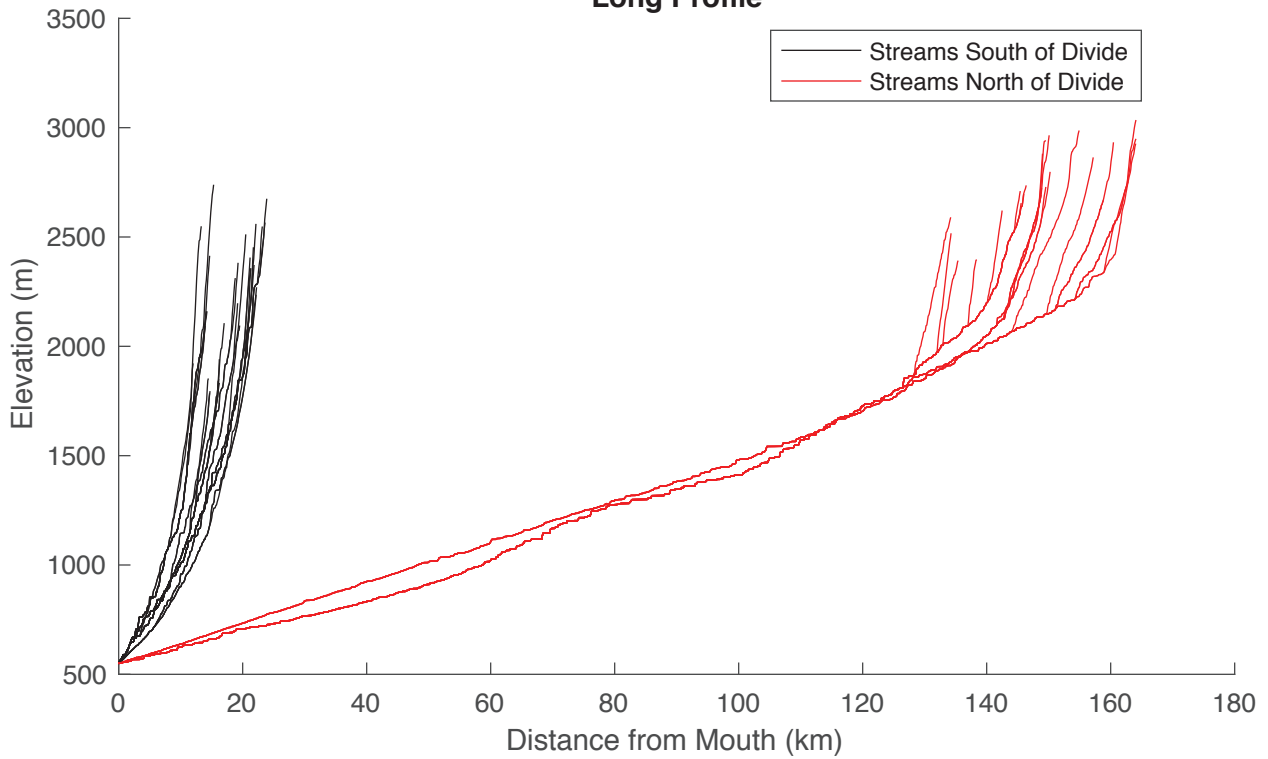
**Supplemental Figure 5**

### Divide GC3

#### Chi - Z

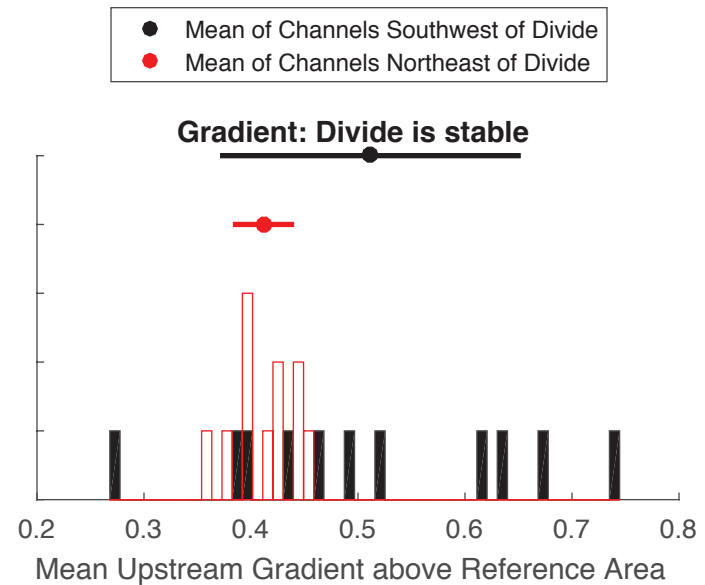
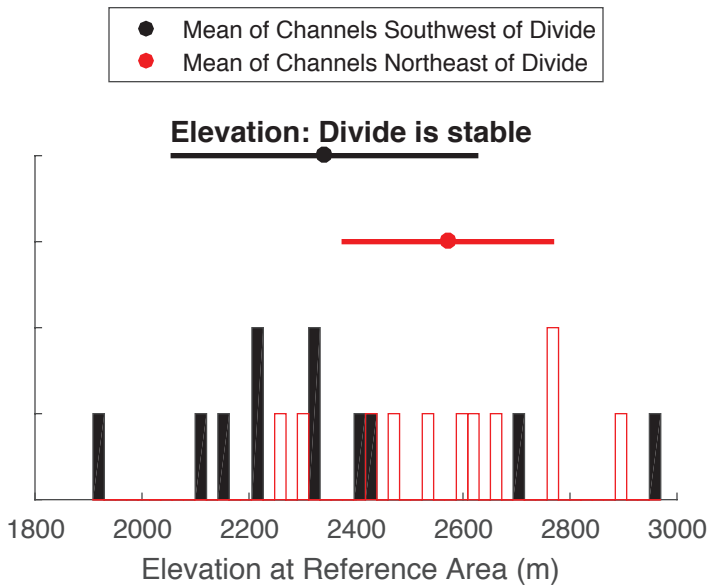
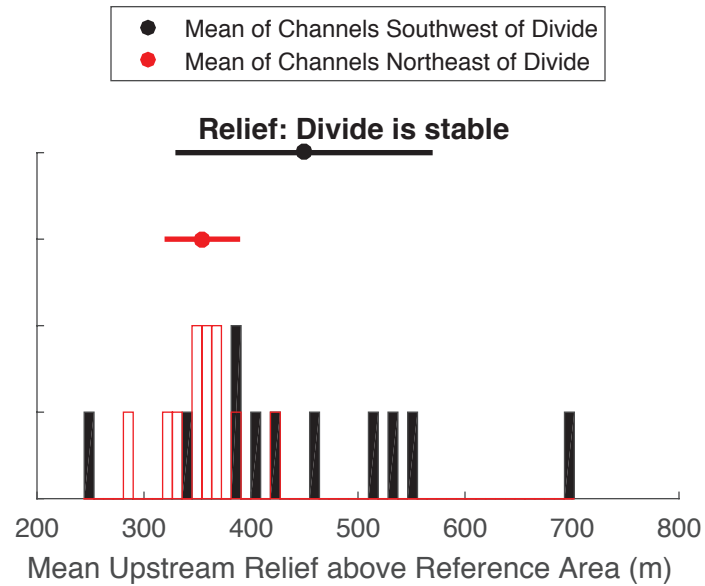
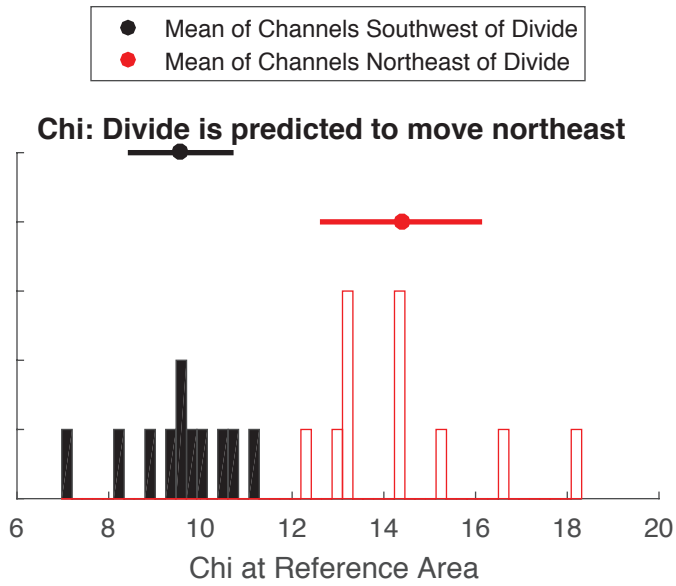


#### Long Profile



**Supplemental Figure 6**

## Divide GC4

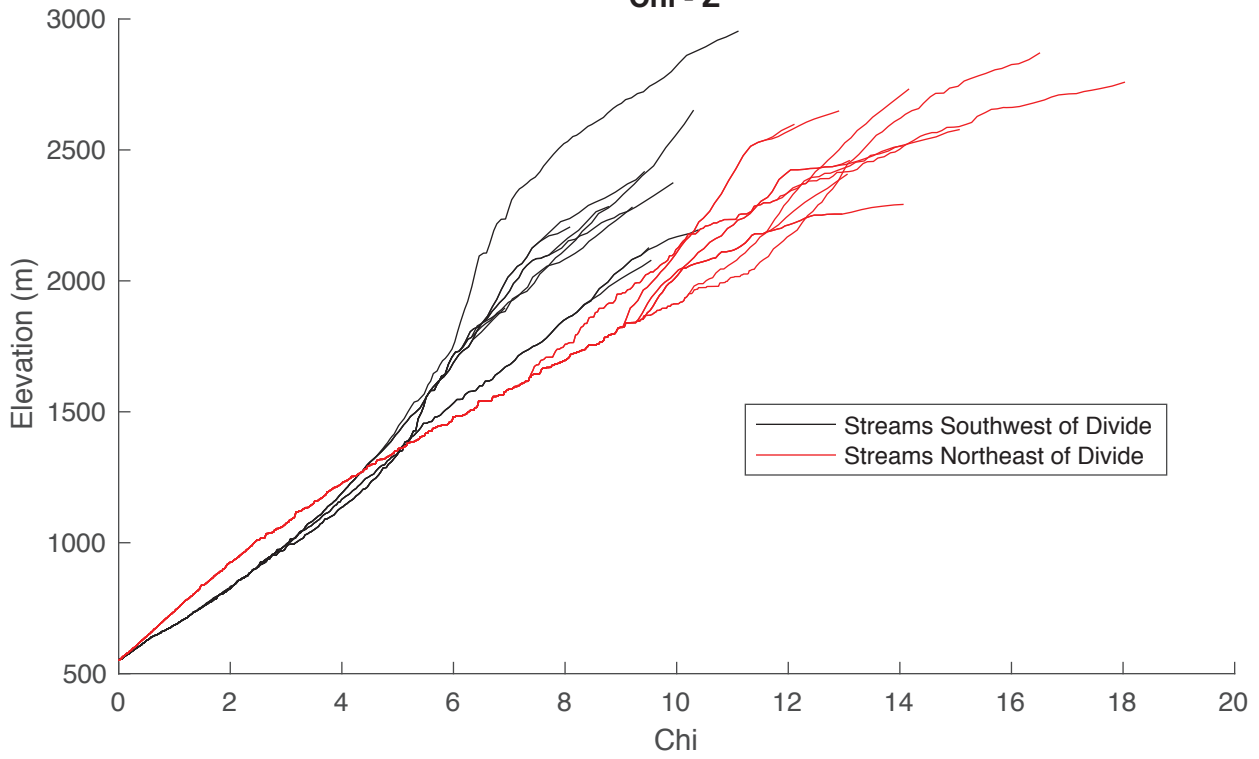


**Supplemental Figure 7**

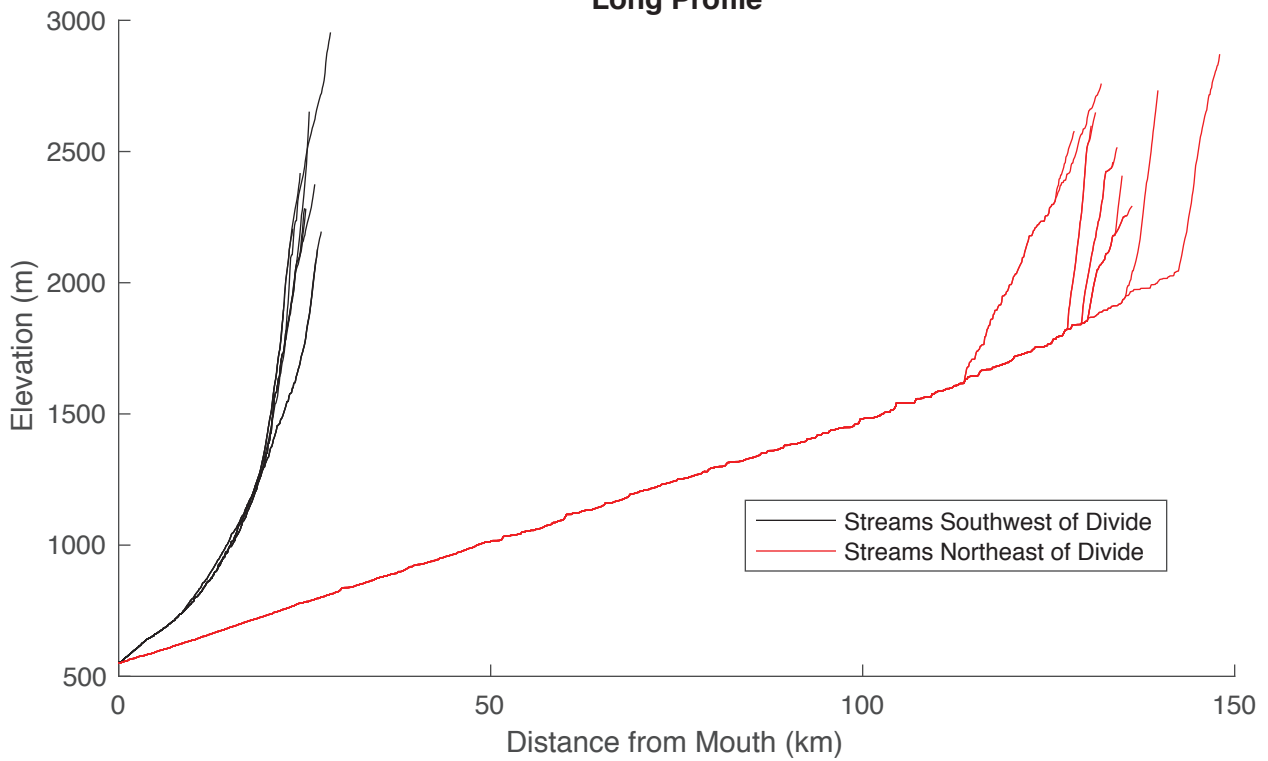


# Divide GC4

Chi - Z

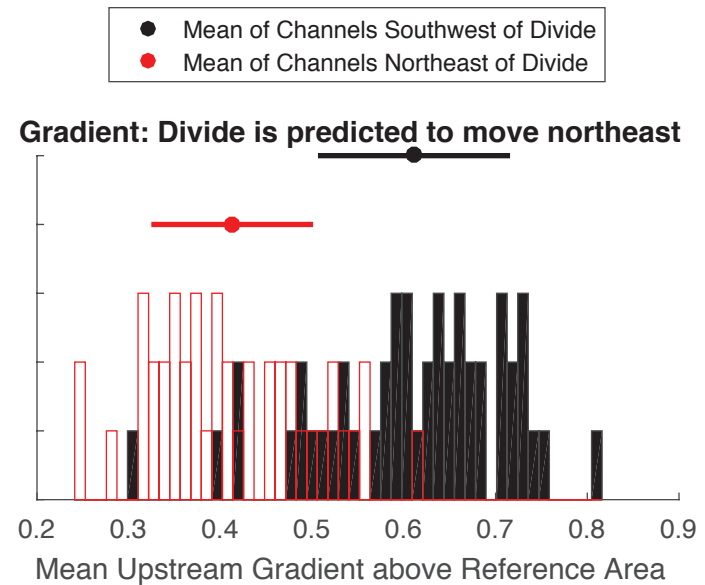
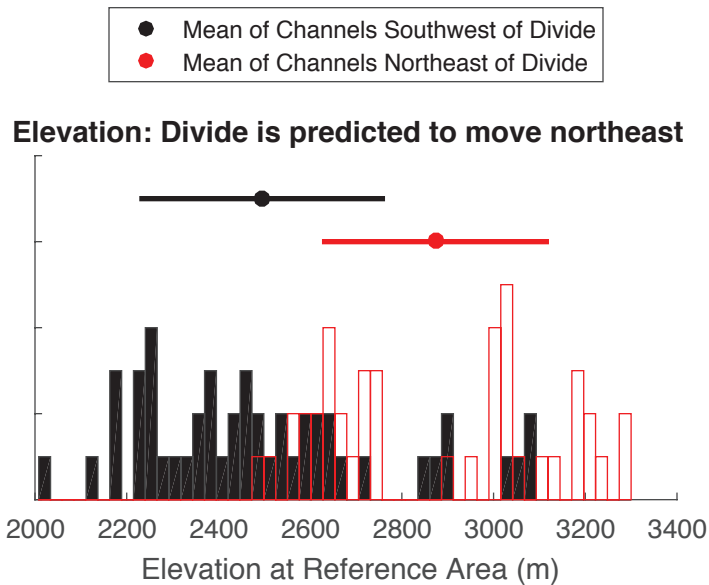
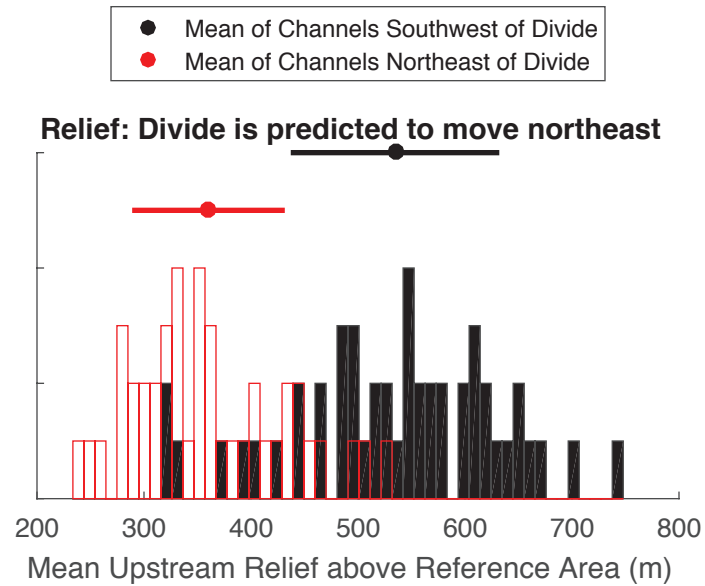
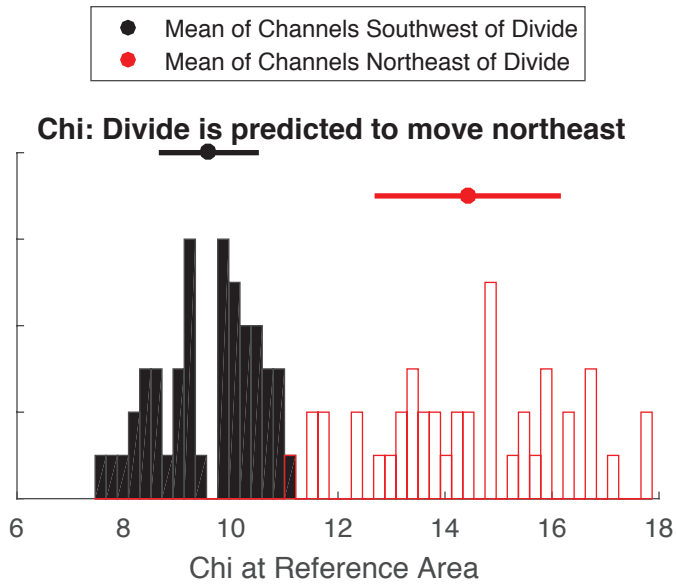


# Long Profile



**Supplemental Figure 8**

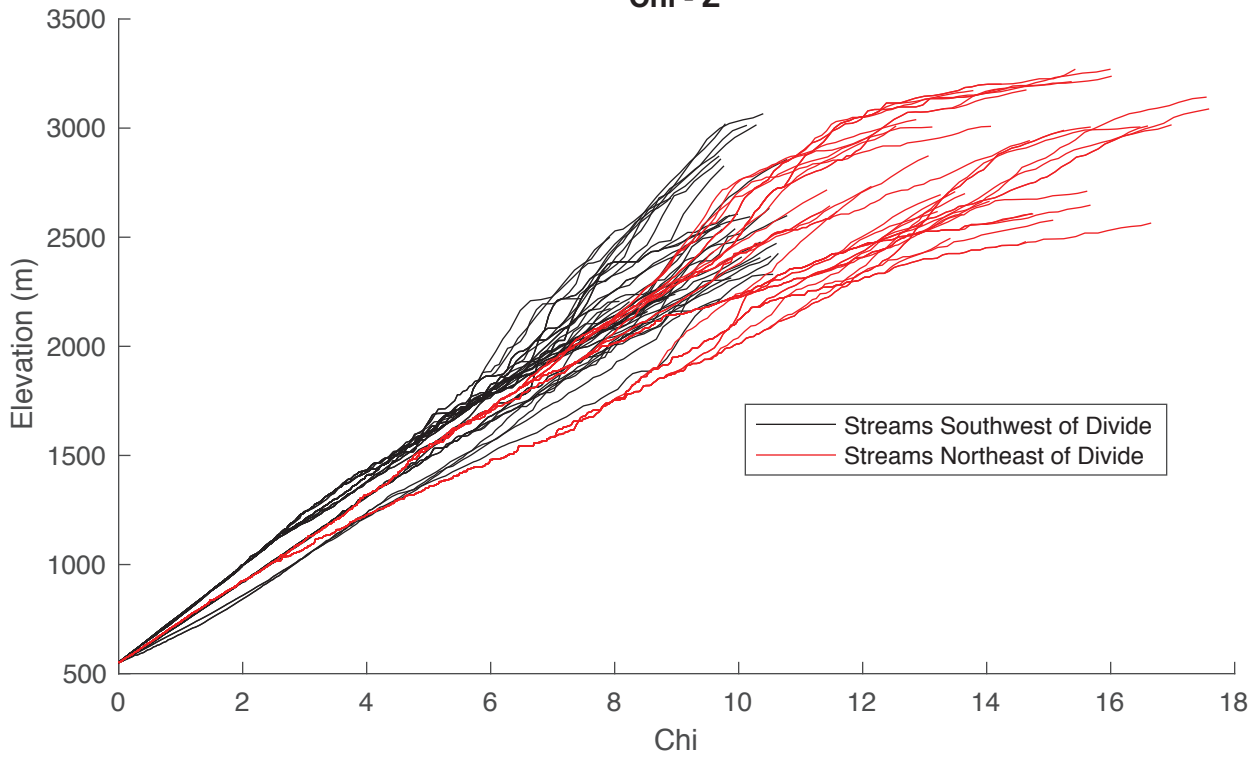
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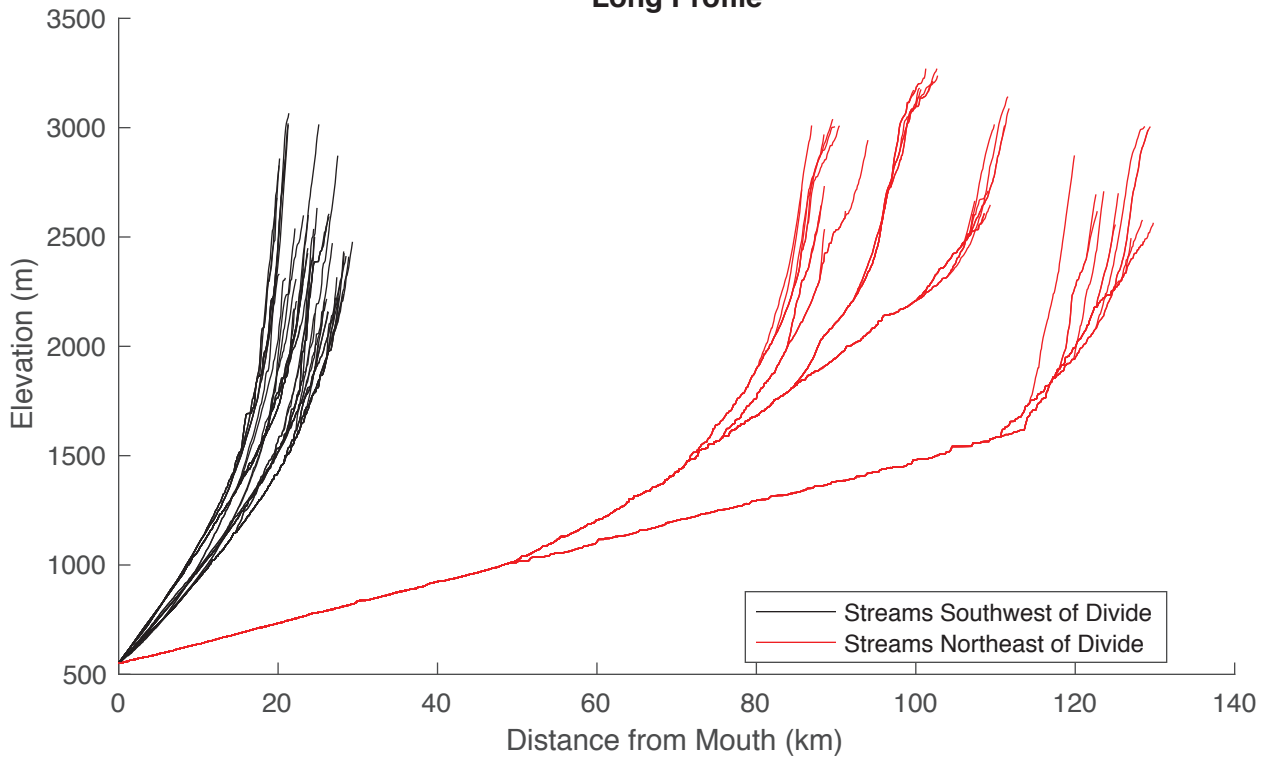
**Supplemental Figure 9**

# Divide GC5

## Chi - Z

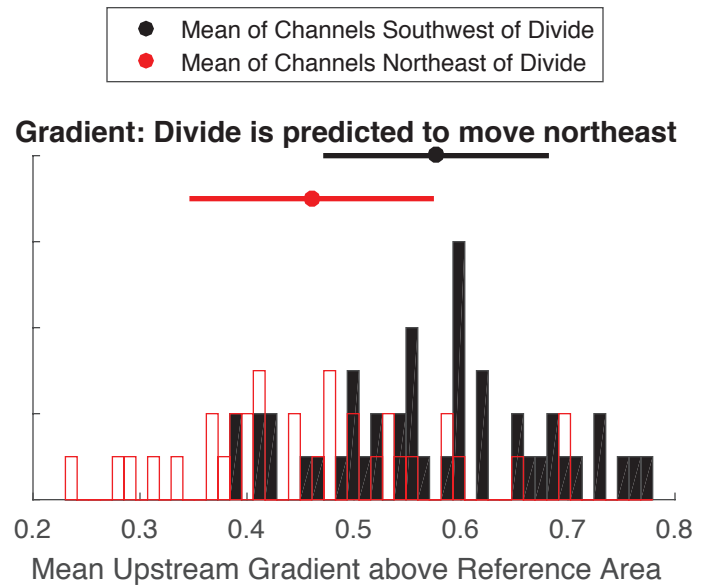
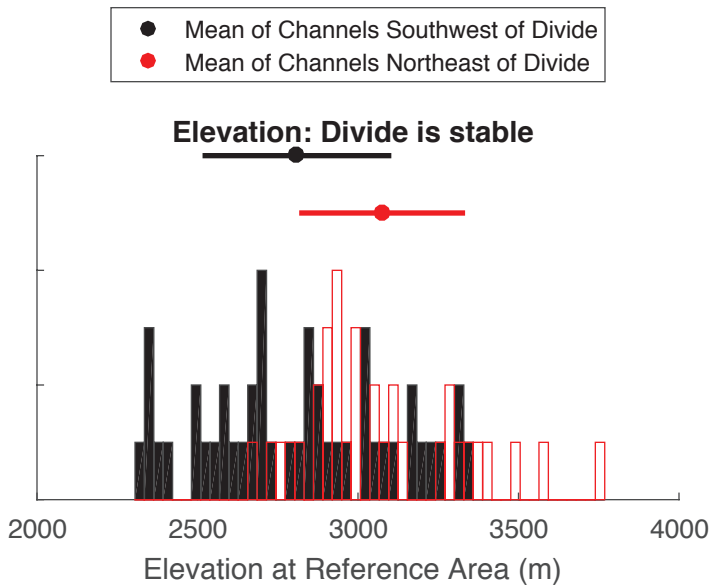
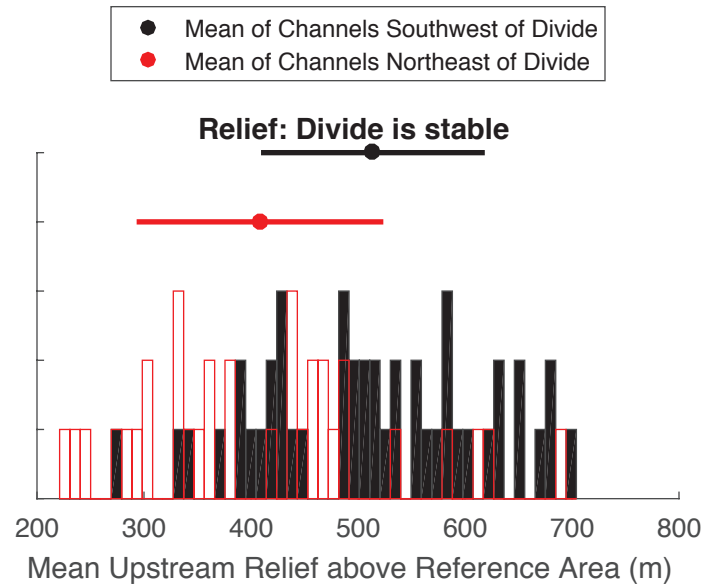
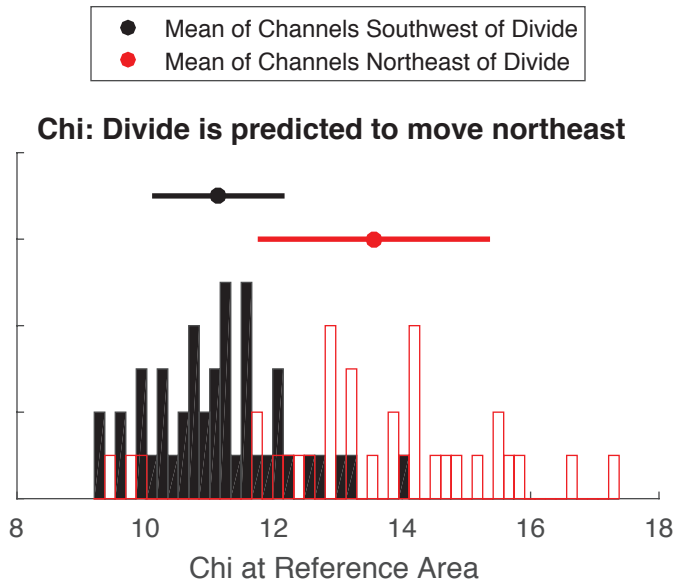


## Long Profile



**Supplemental Figure 10**

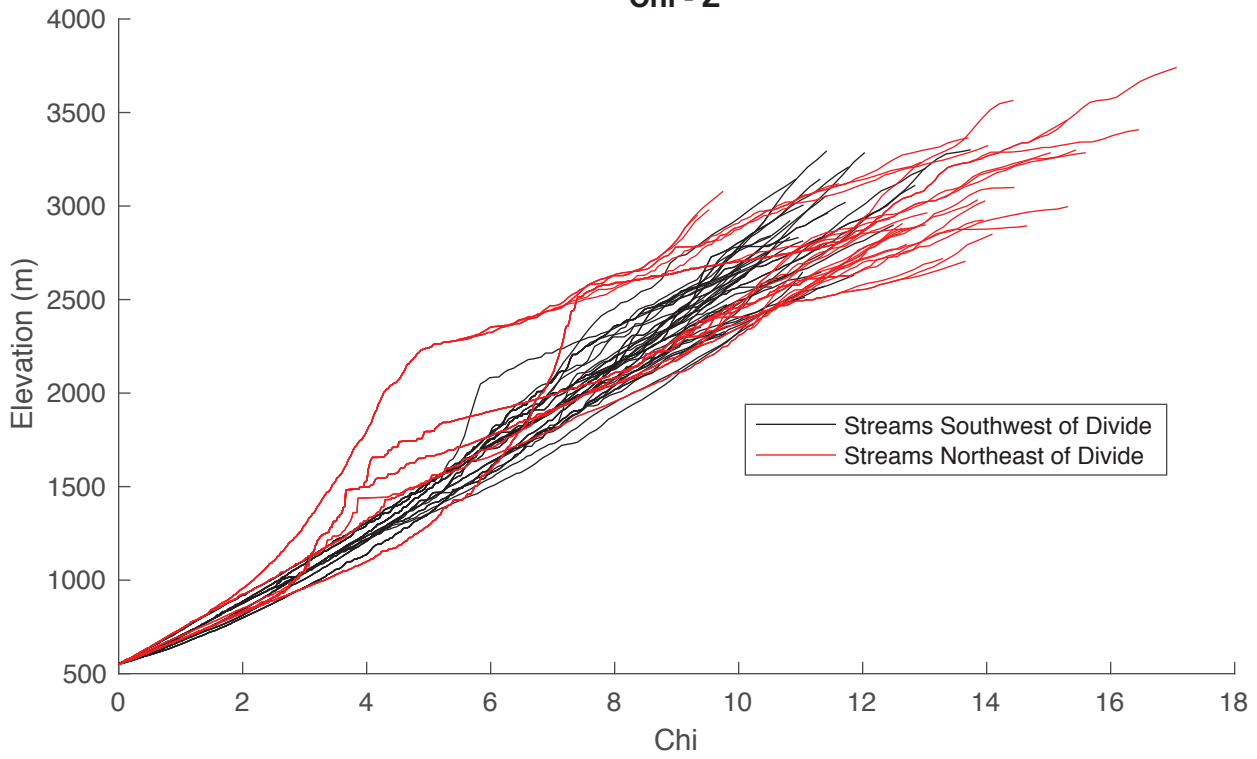
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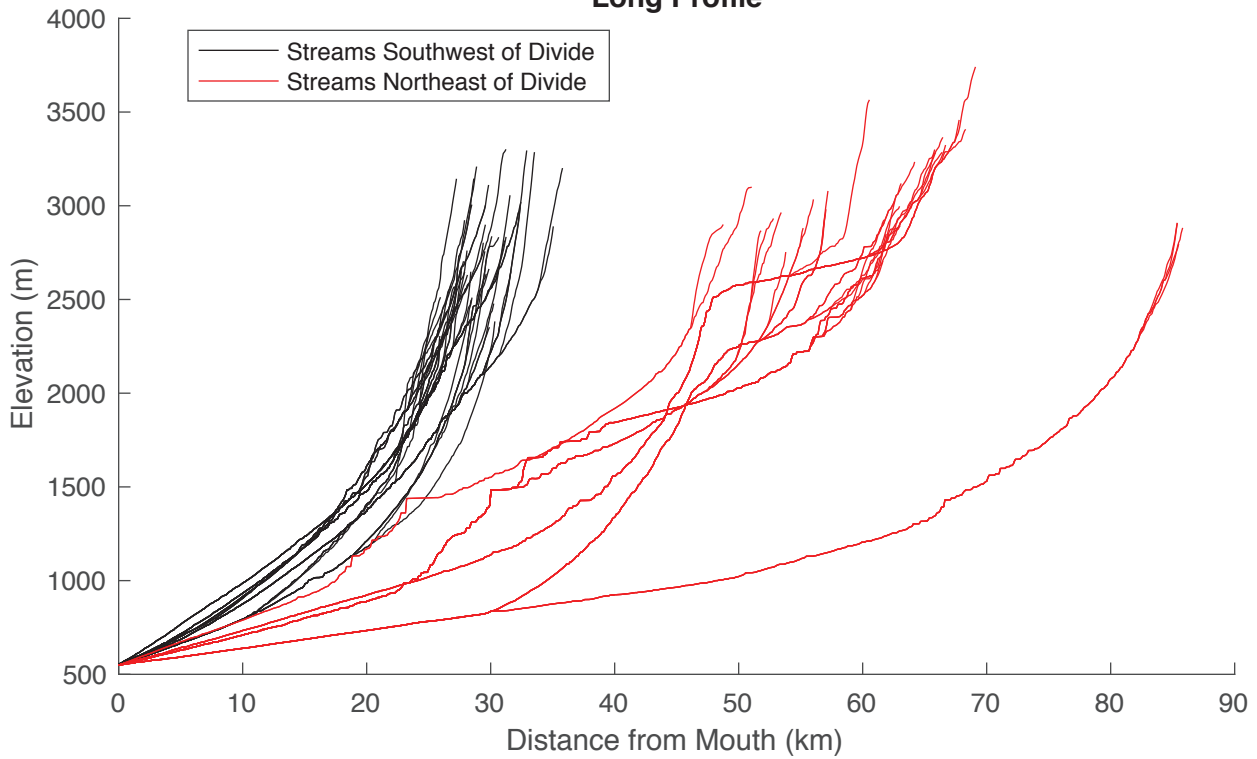
**Supplemental Figure 11**

# Divide GC6

Chi - Z

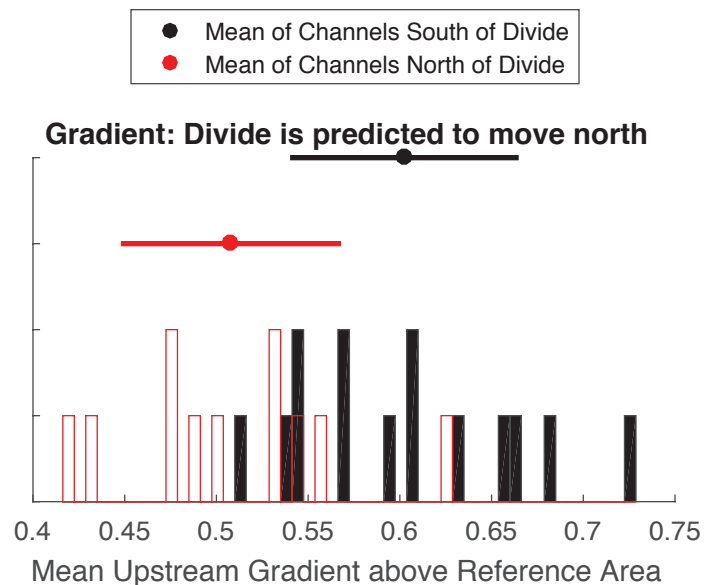
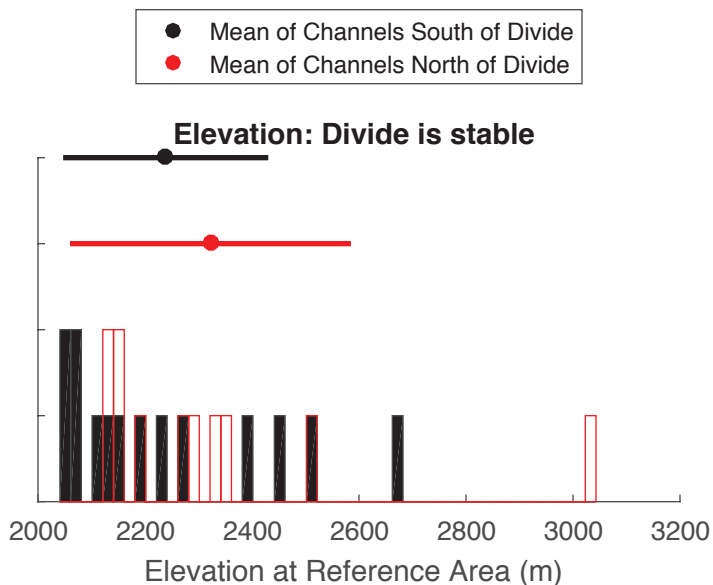
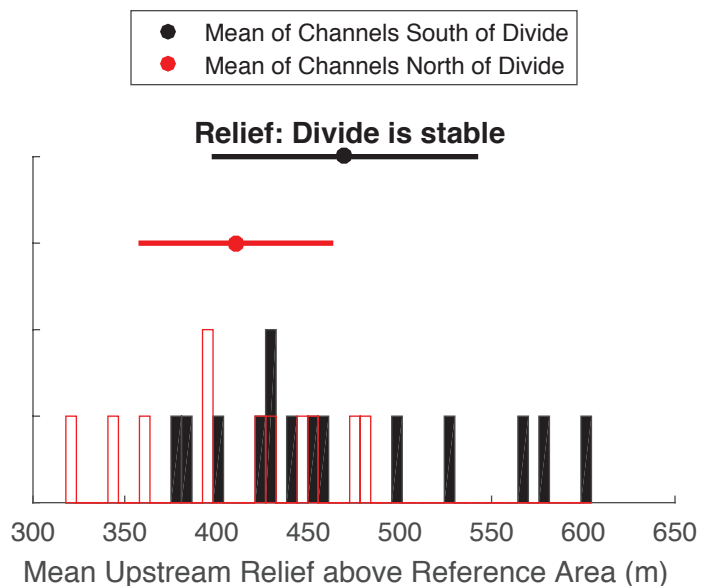
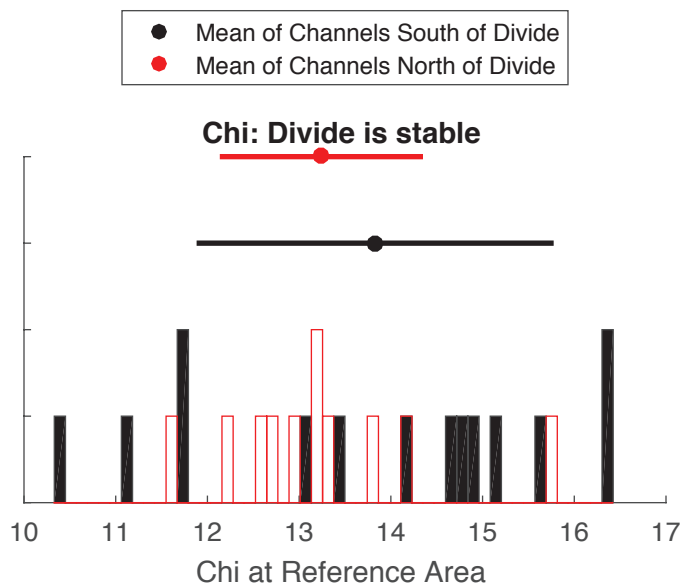


# Long Profile



Supplemental Figure 12

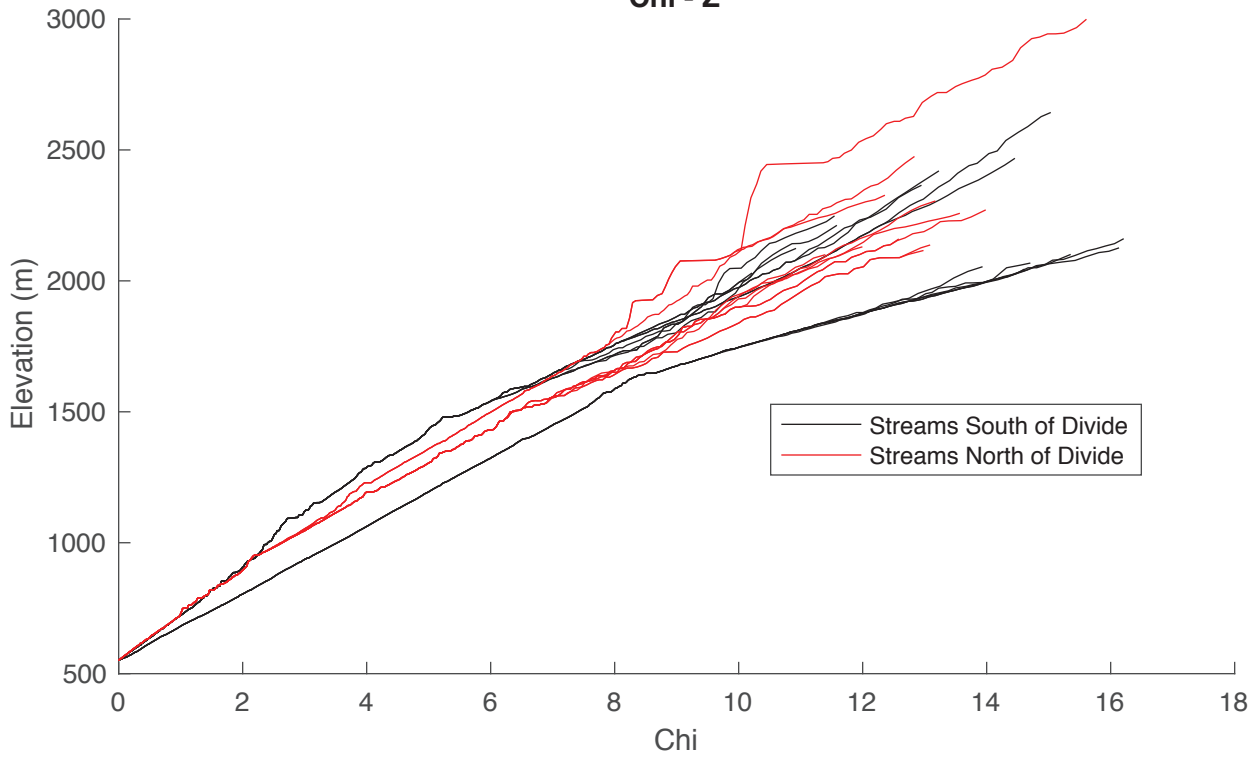
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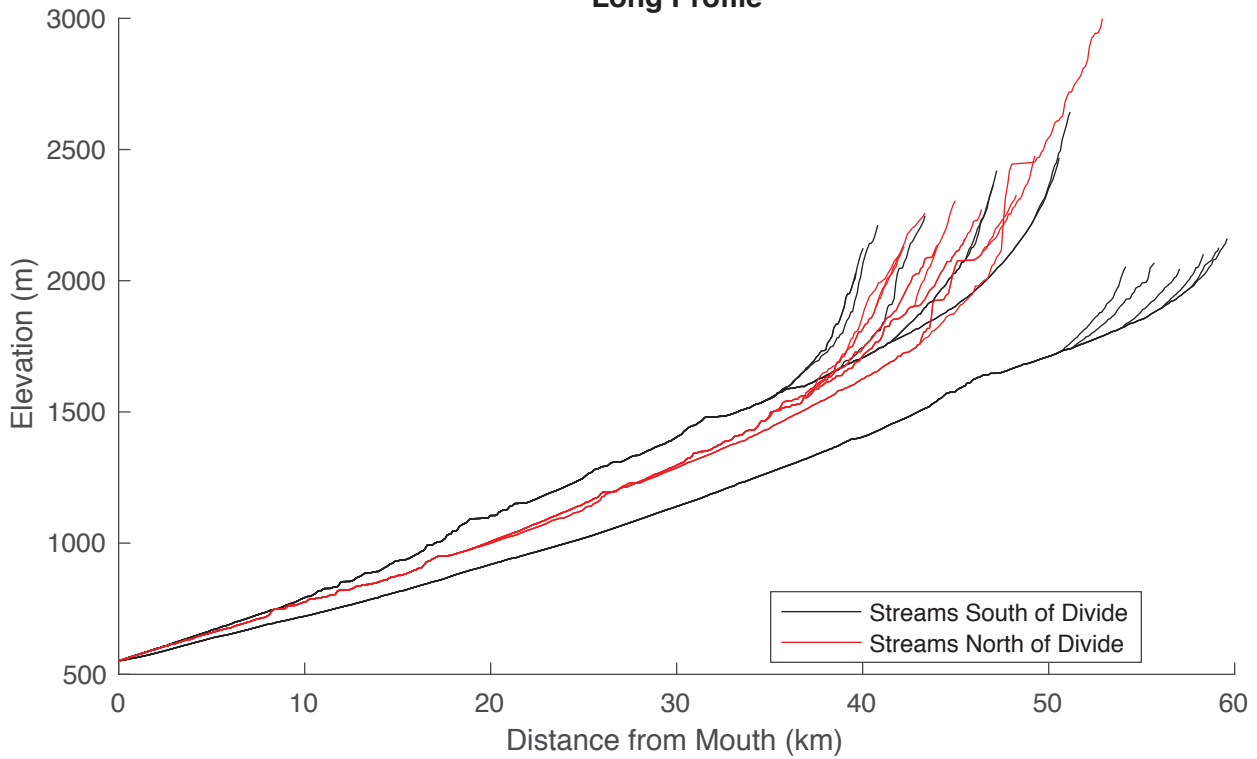
**Supplemental Figure 13**

# Divide GC7

## Chi - Z

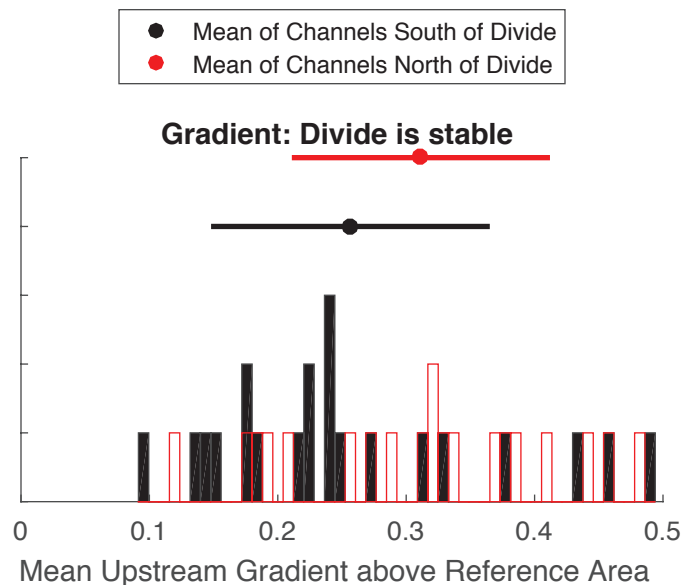
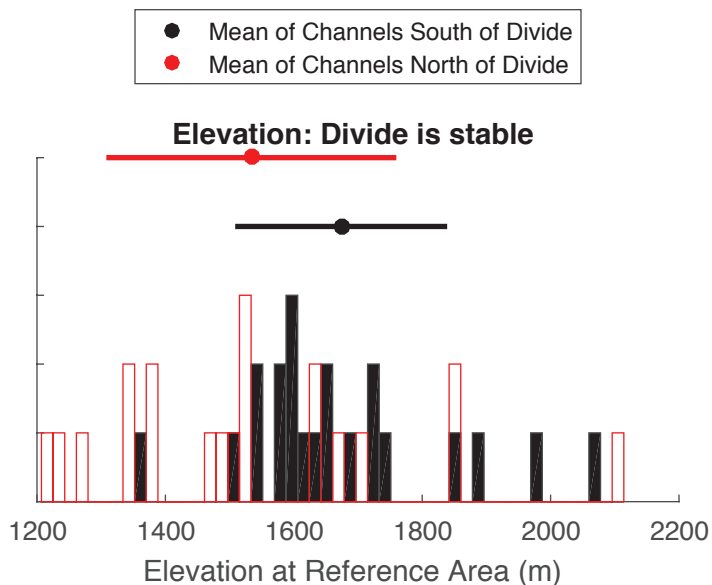
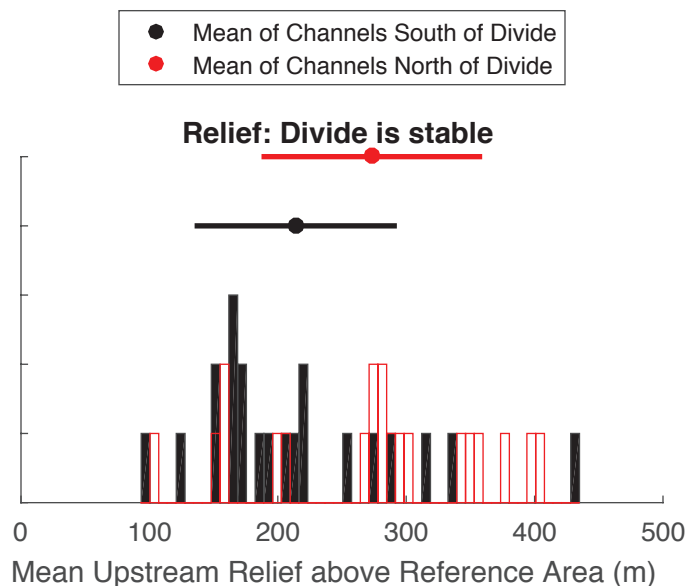
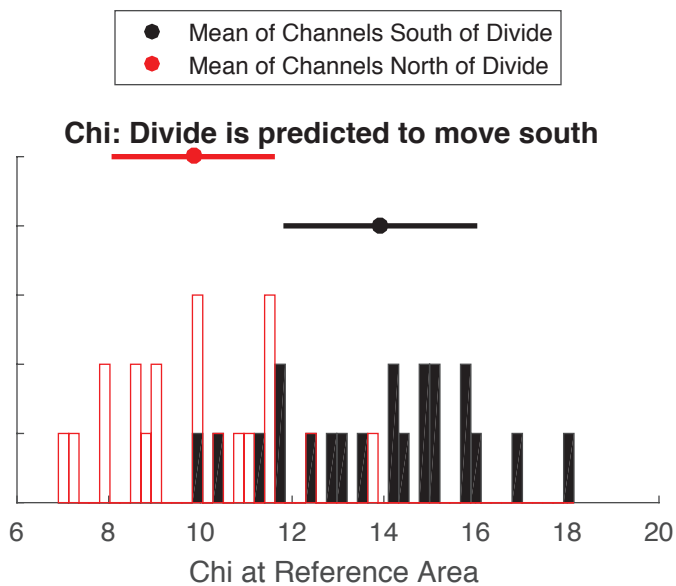


## Long Profile



**Supplemental Figure 14**

## Divide GC8

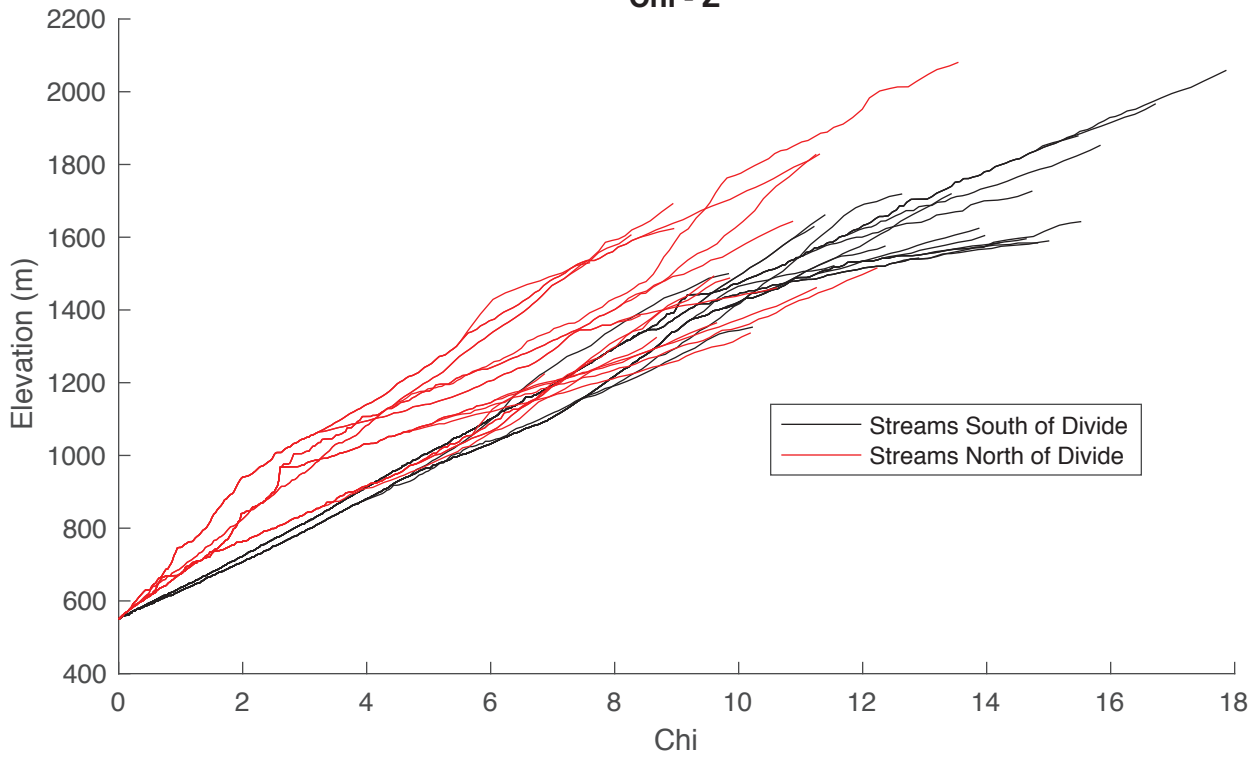


**Supplemental Figure 15**

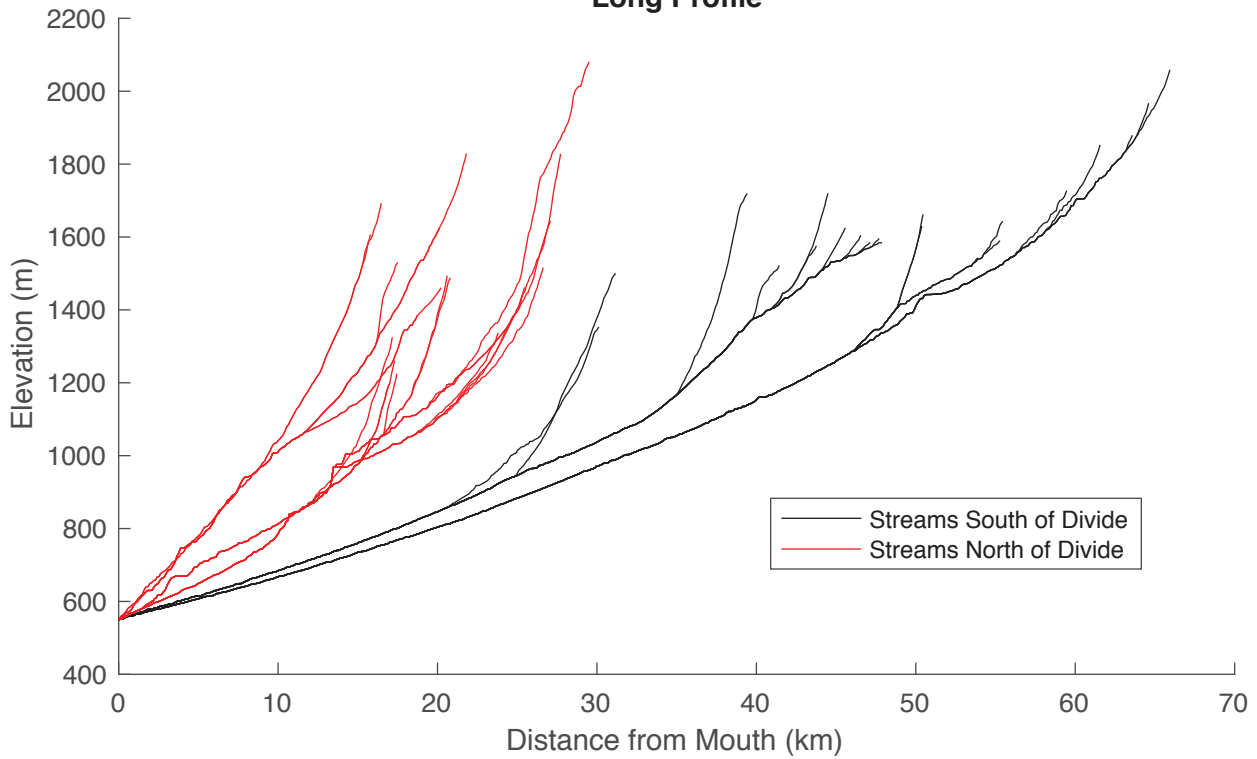


# Divide GC8

Chi - Z

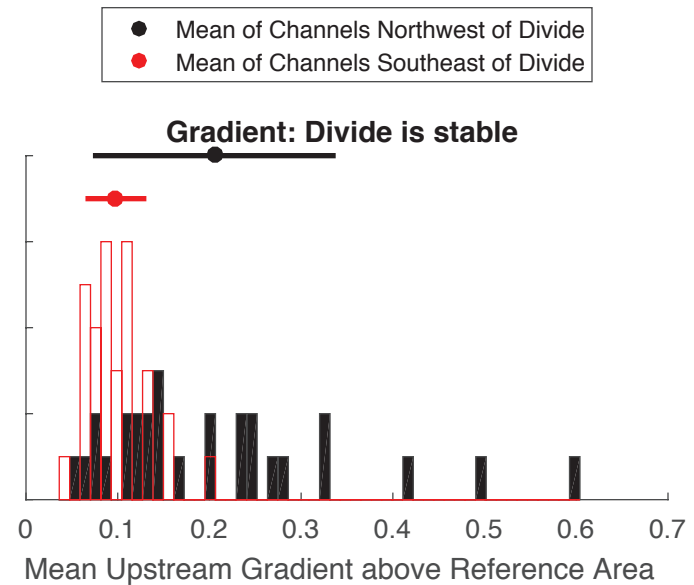
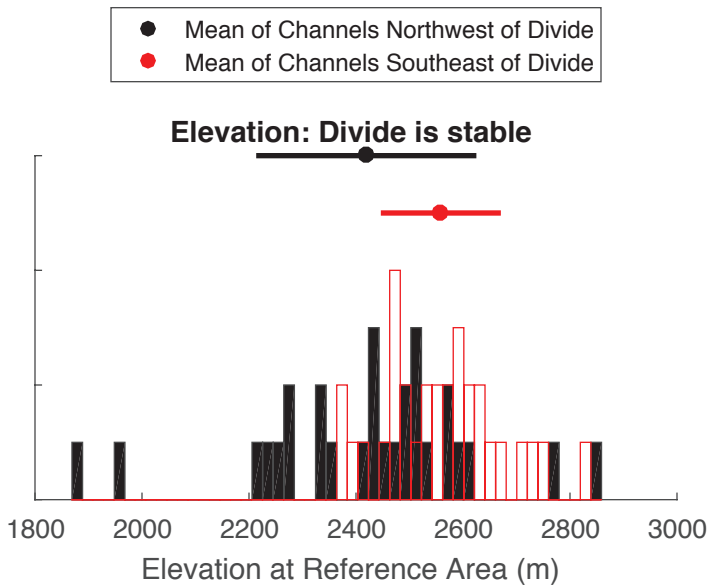
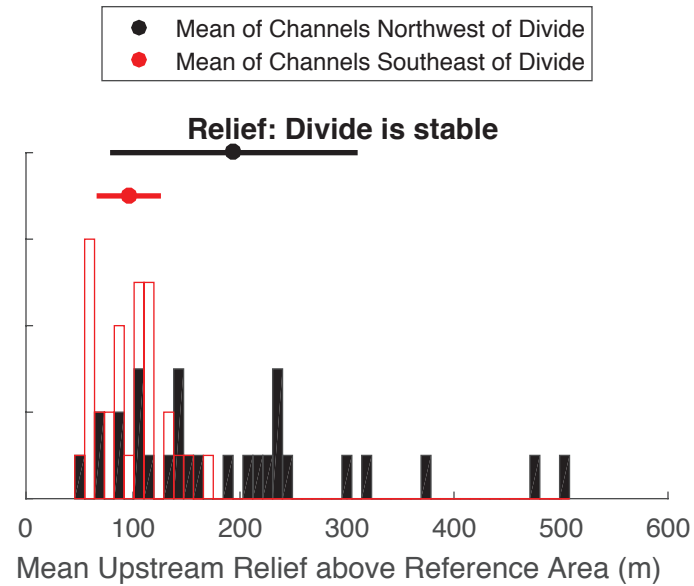
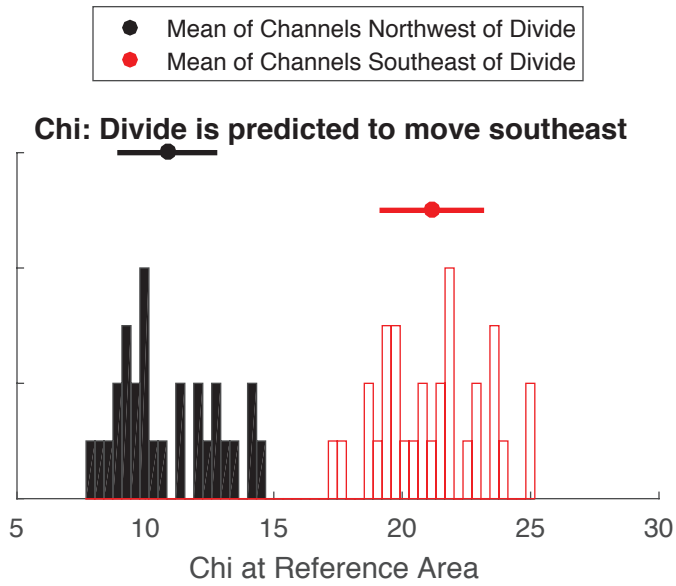


# Long Profile



Supplemental Figure 16

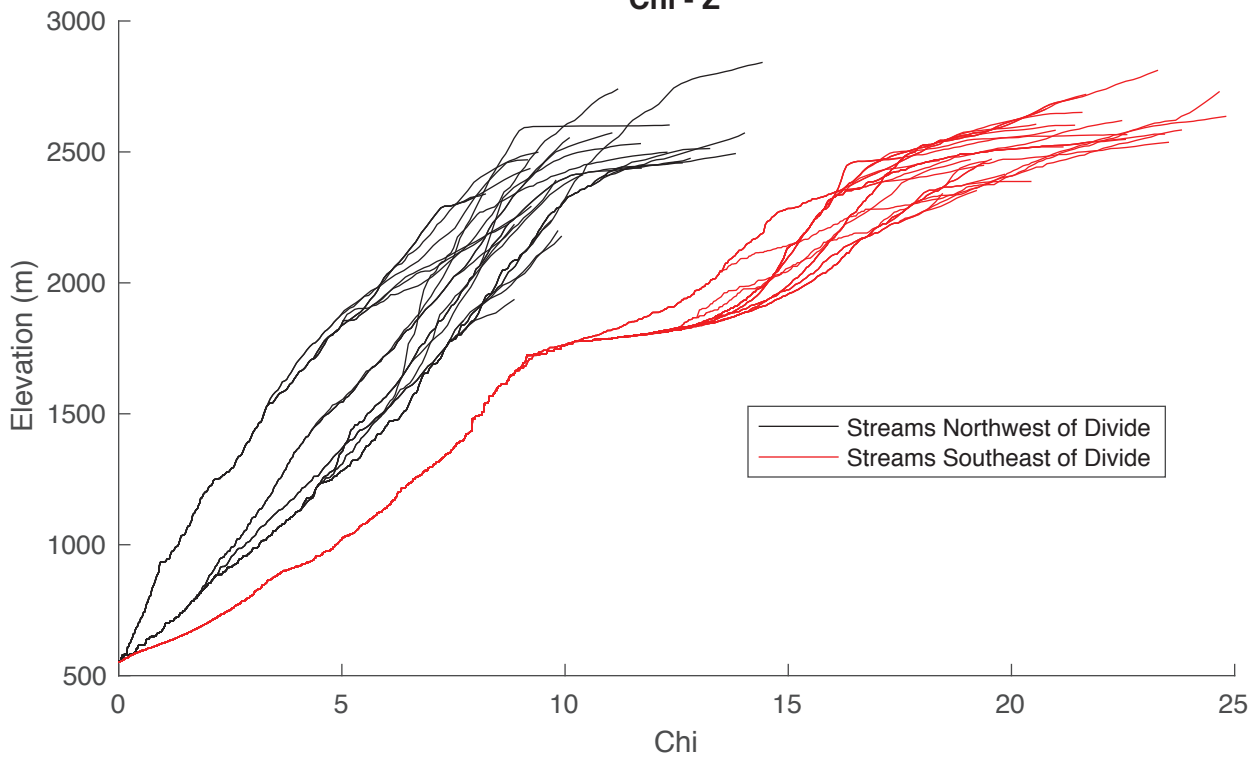
## Divide KV1



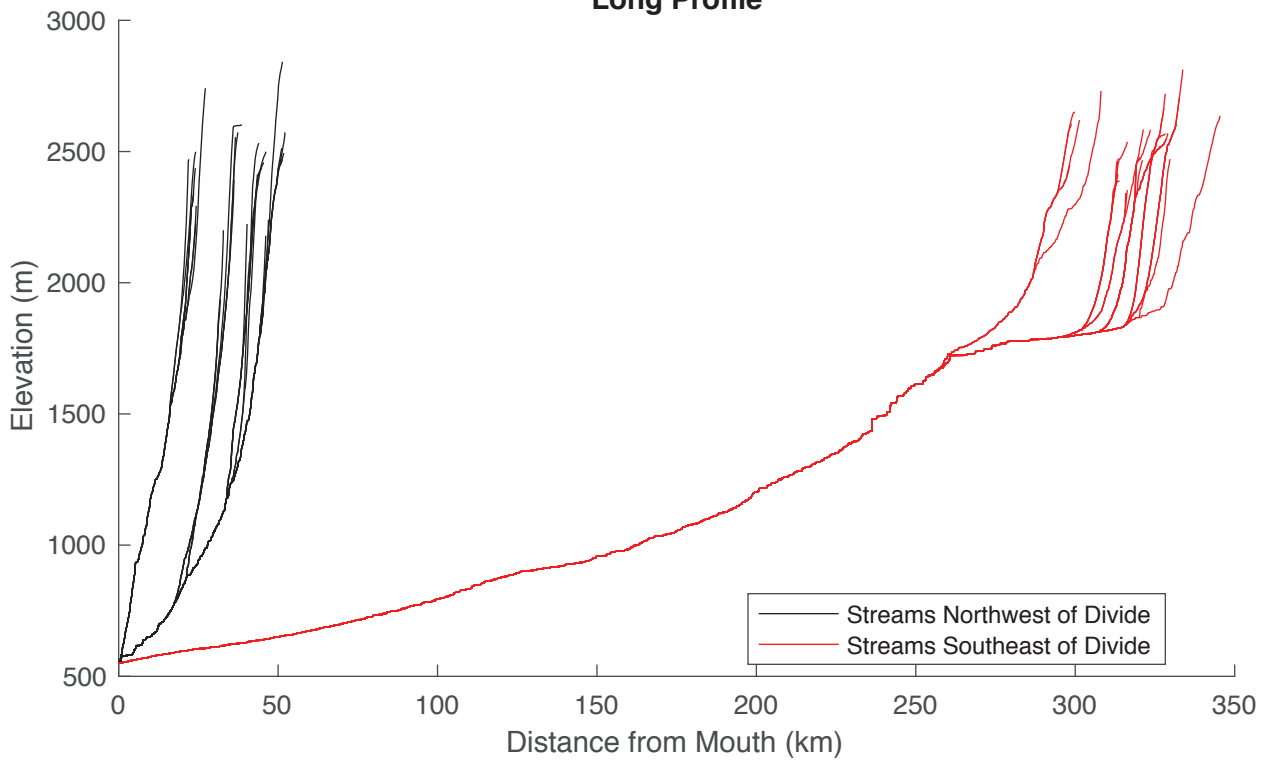
**Supplemental Figure 17**

### Divide KV1

Chi - Z

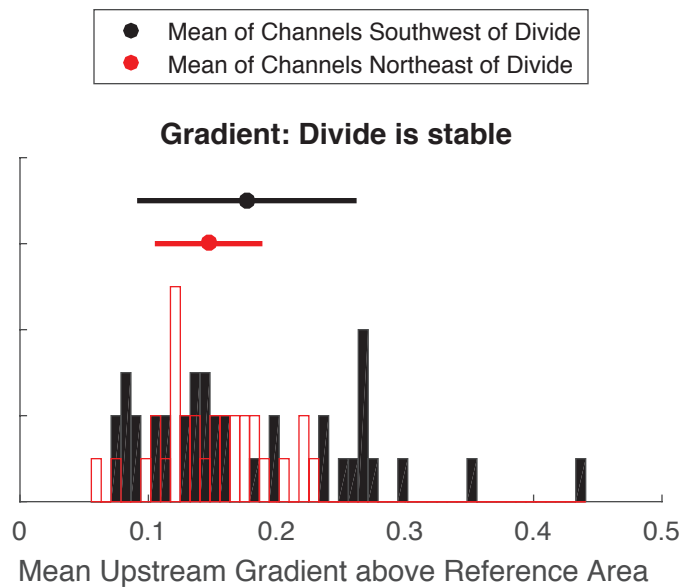
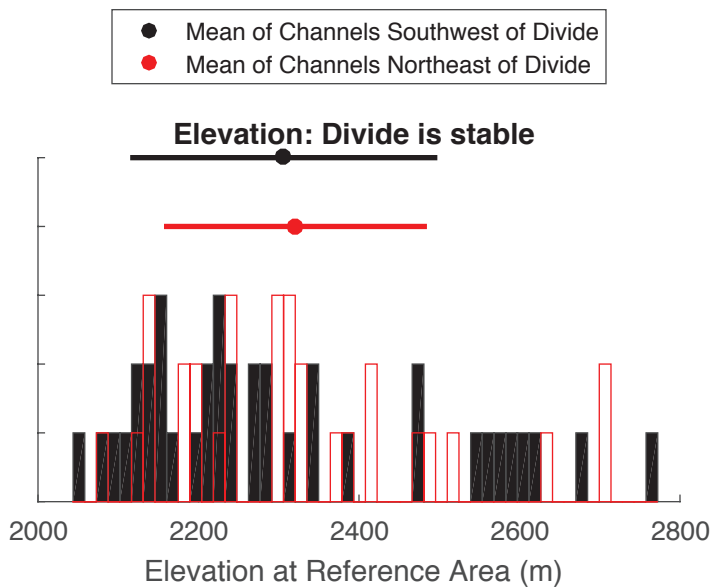
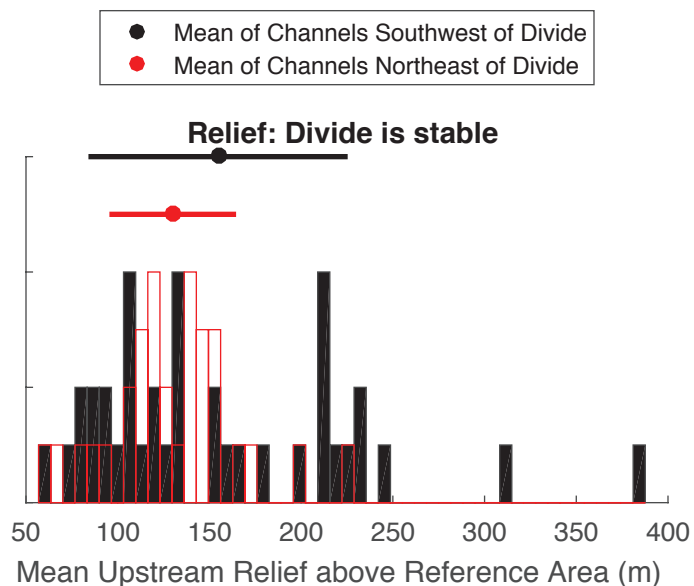
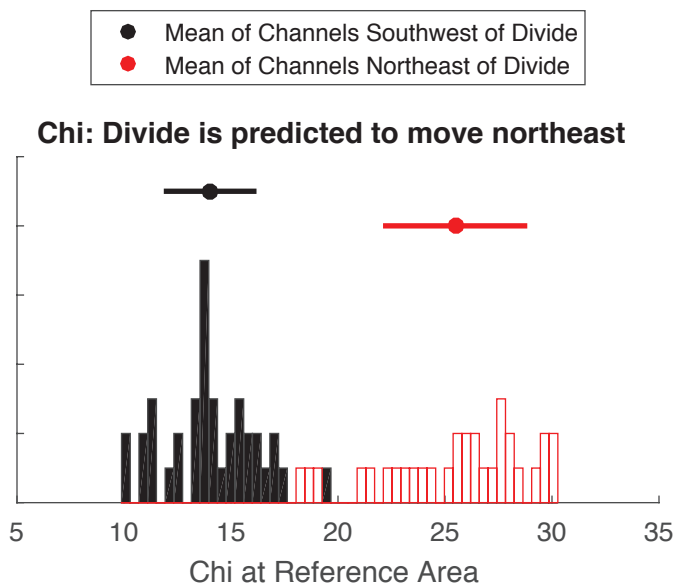


### Long Profile



**Supplemental Figure 18**

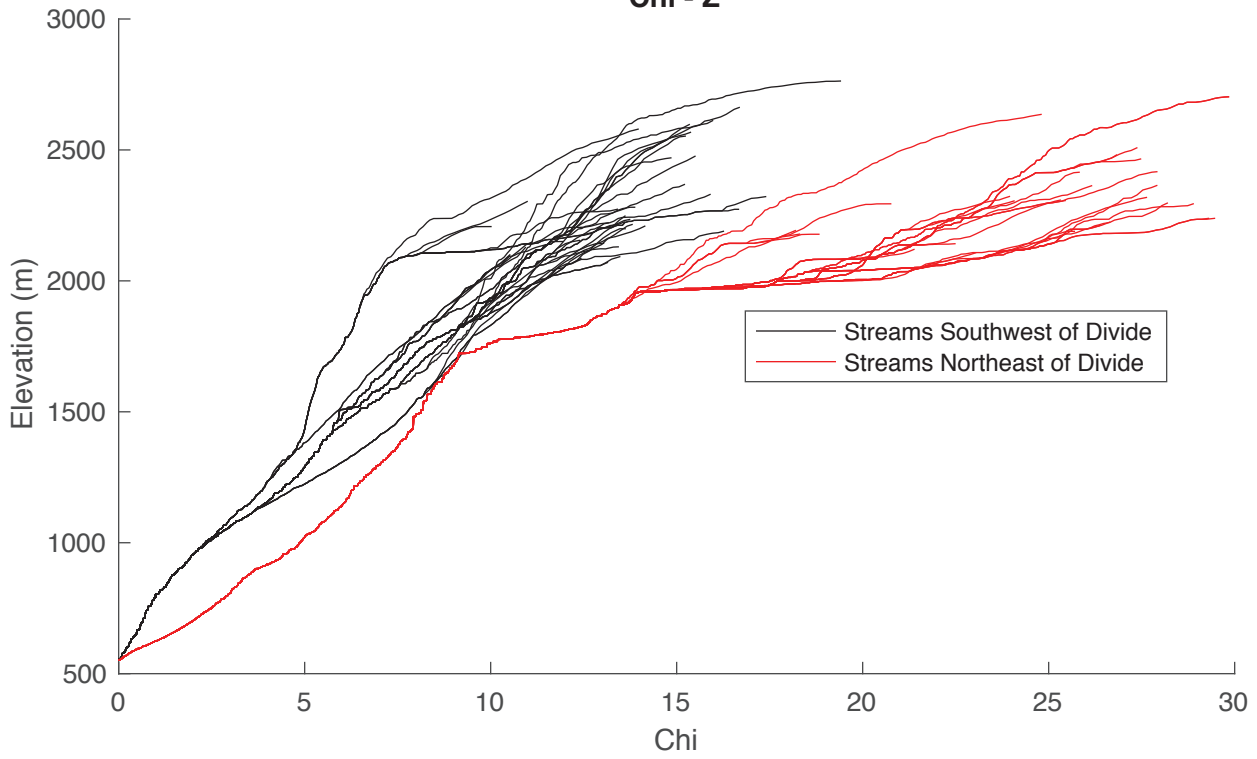
## Divide KV2



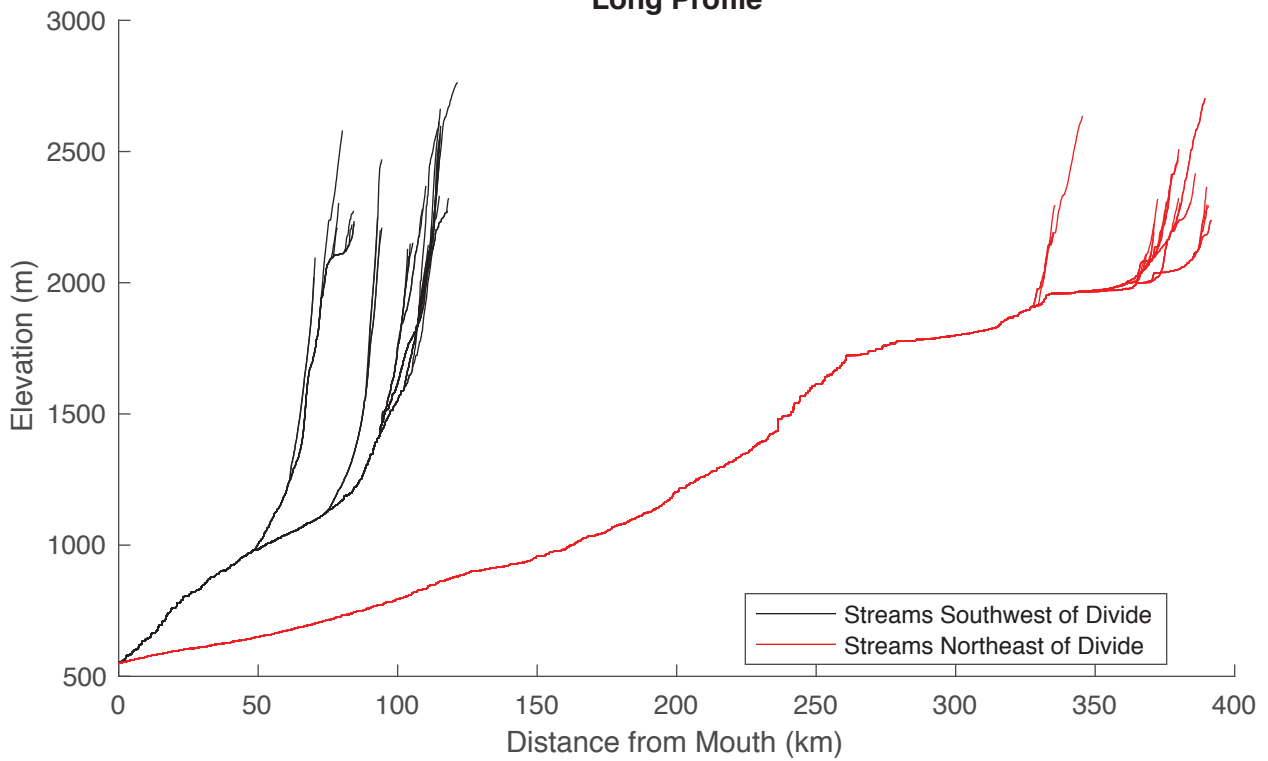
**Supplemental Figure 19**

## Divide KV2

Chi - Z

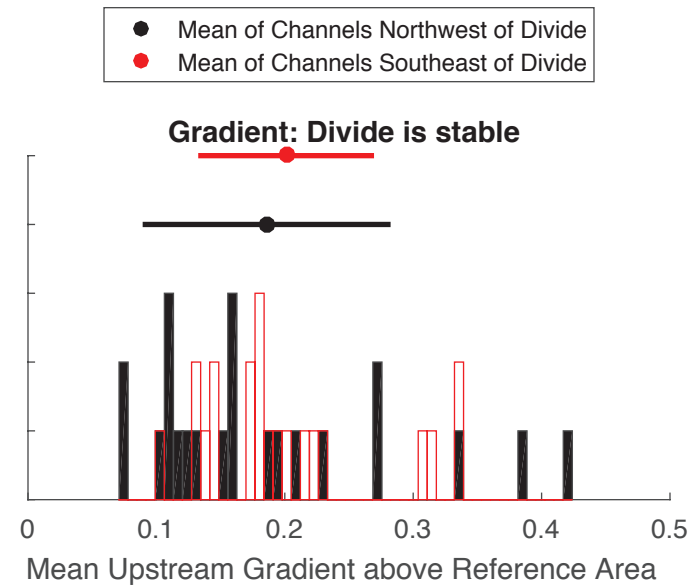
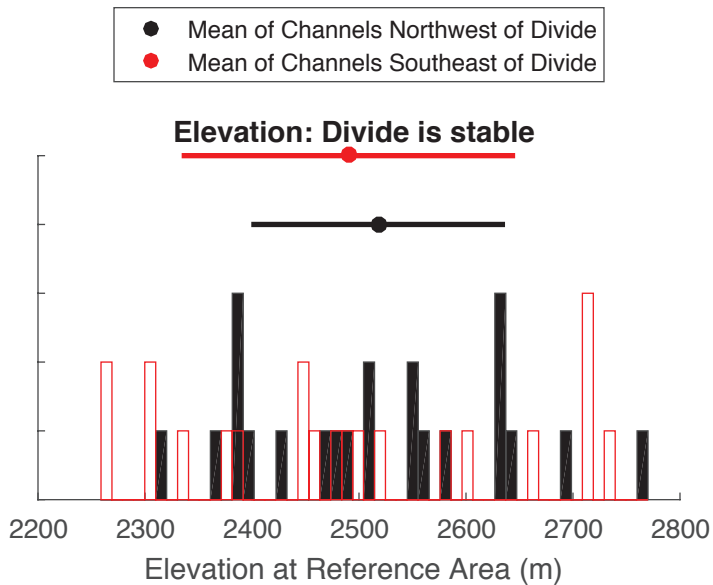
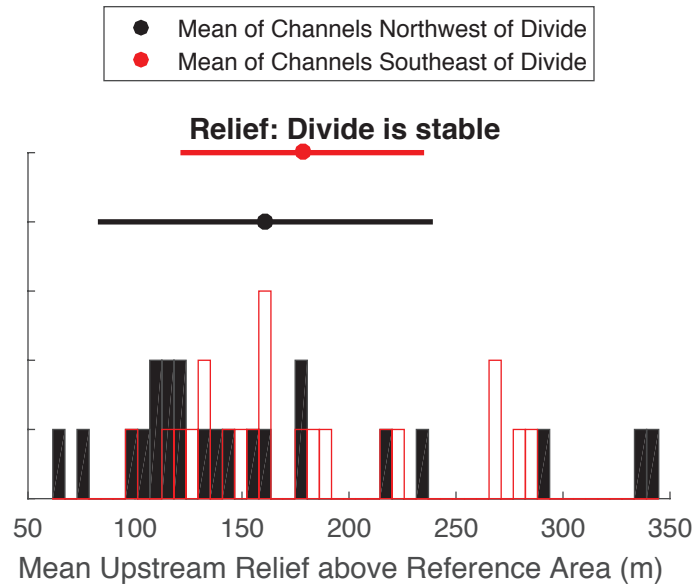
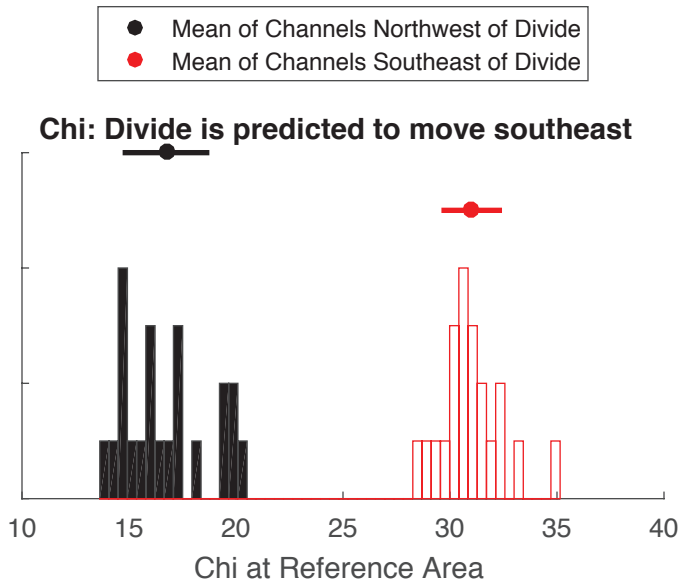


## Long Profile



**Supplemental Figure 20**

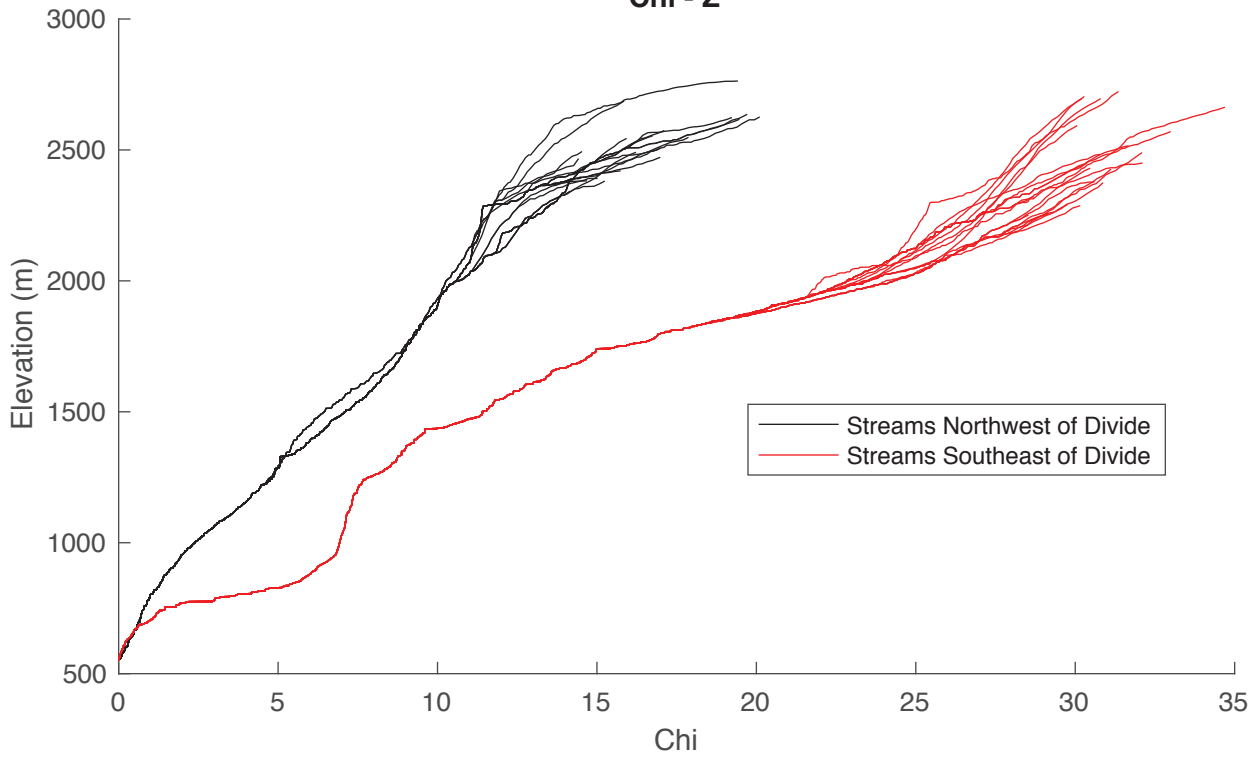
## Divide KV3



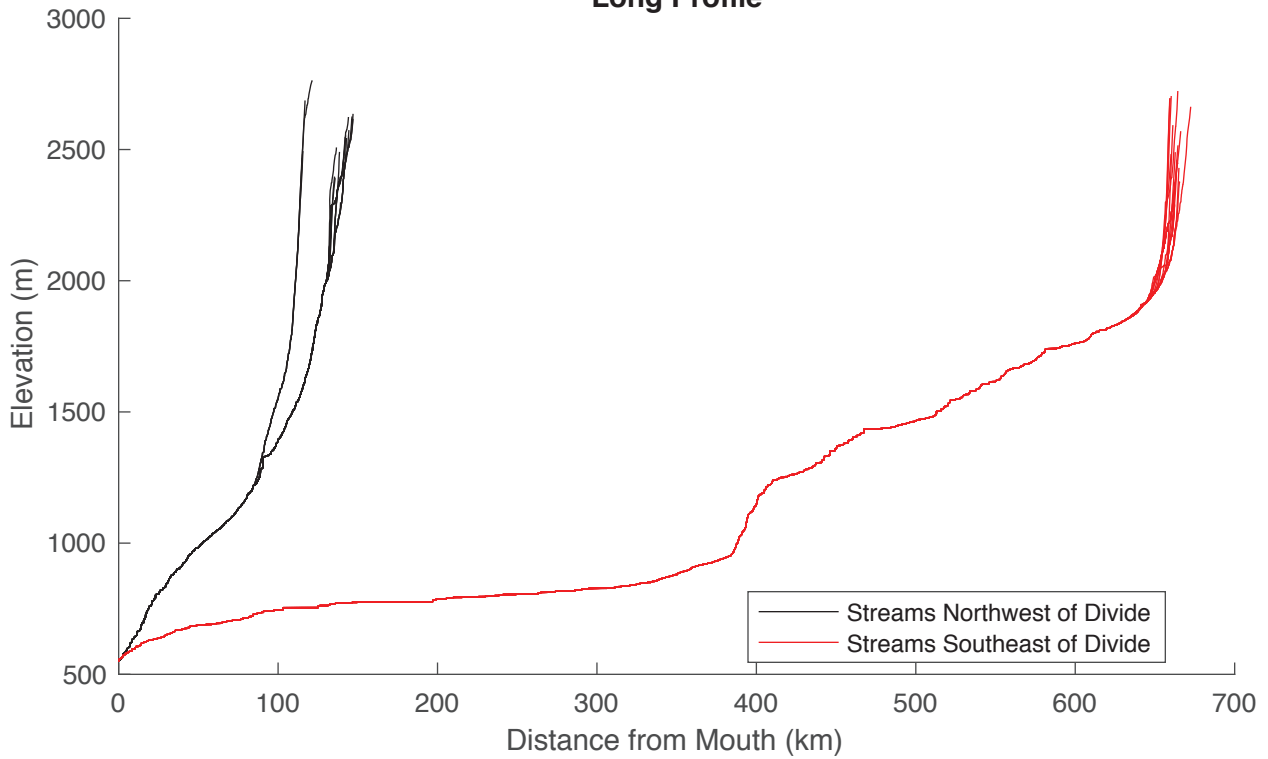
**Supplemental Figure 21**

### Divide KV3

Chi - Z

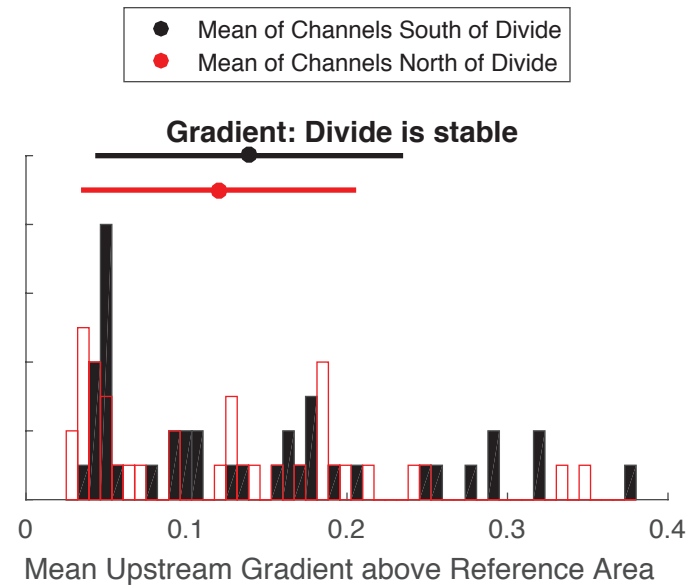
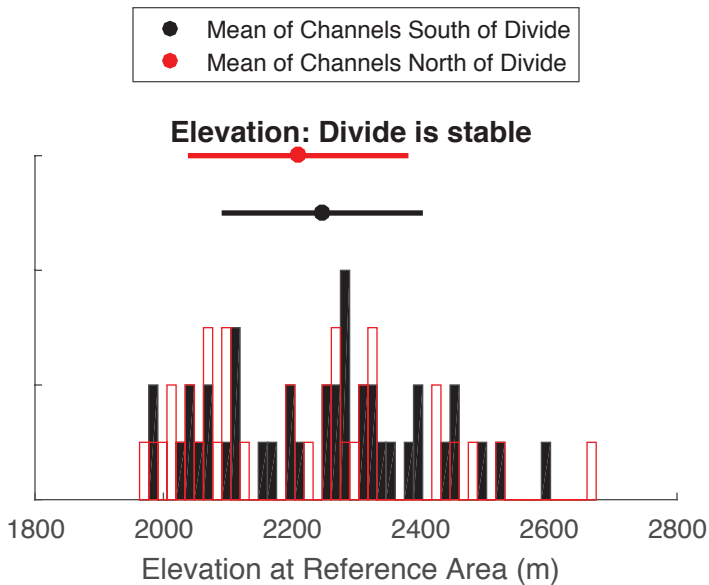
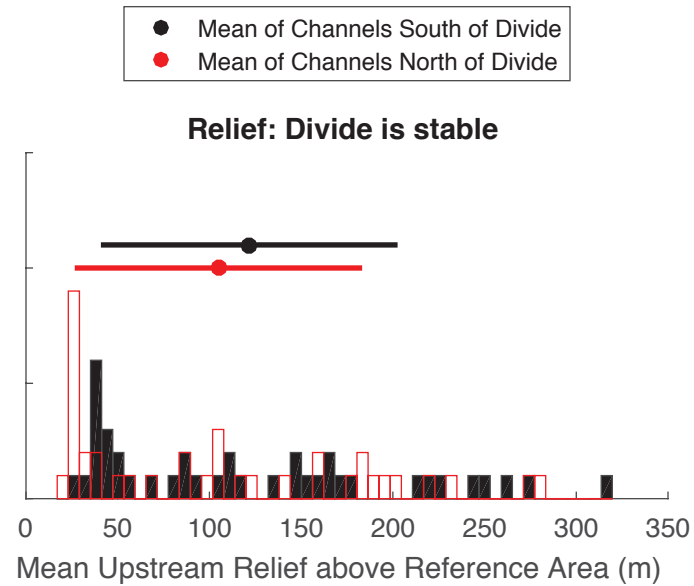
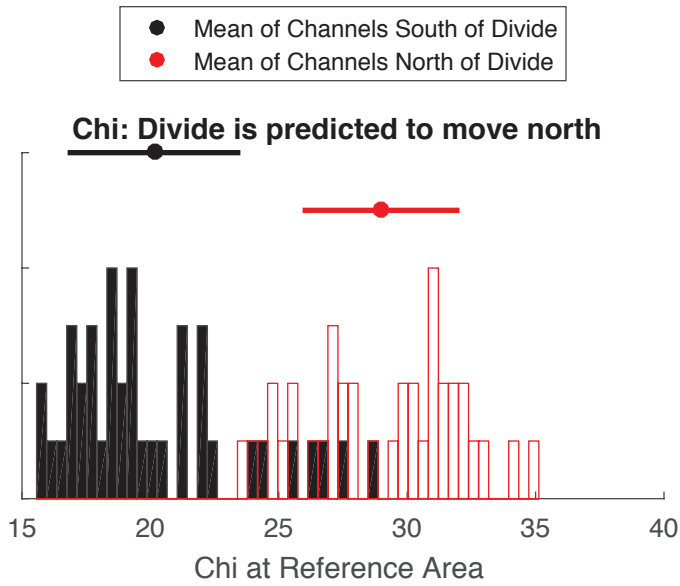


### Long Profile



**Supplemental Figure 22**

## Divide KV4

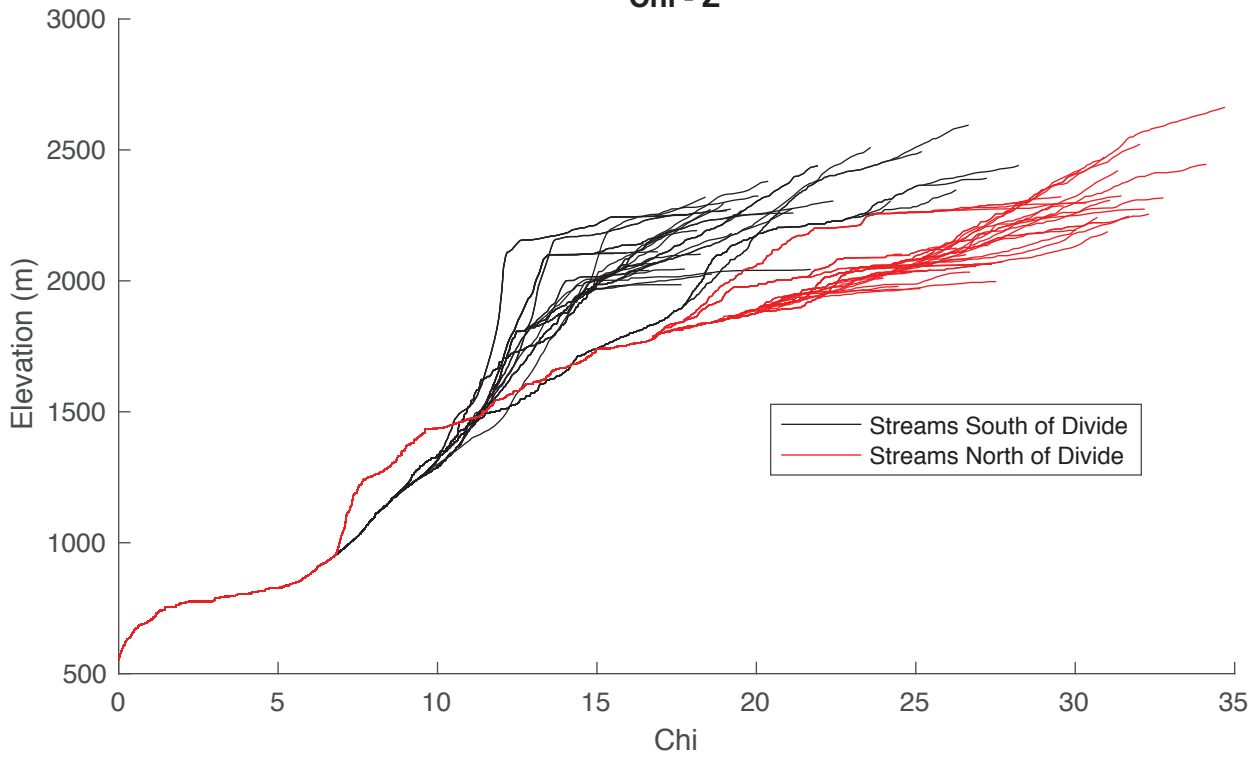


**Supplemental Figure 23**

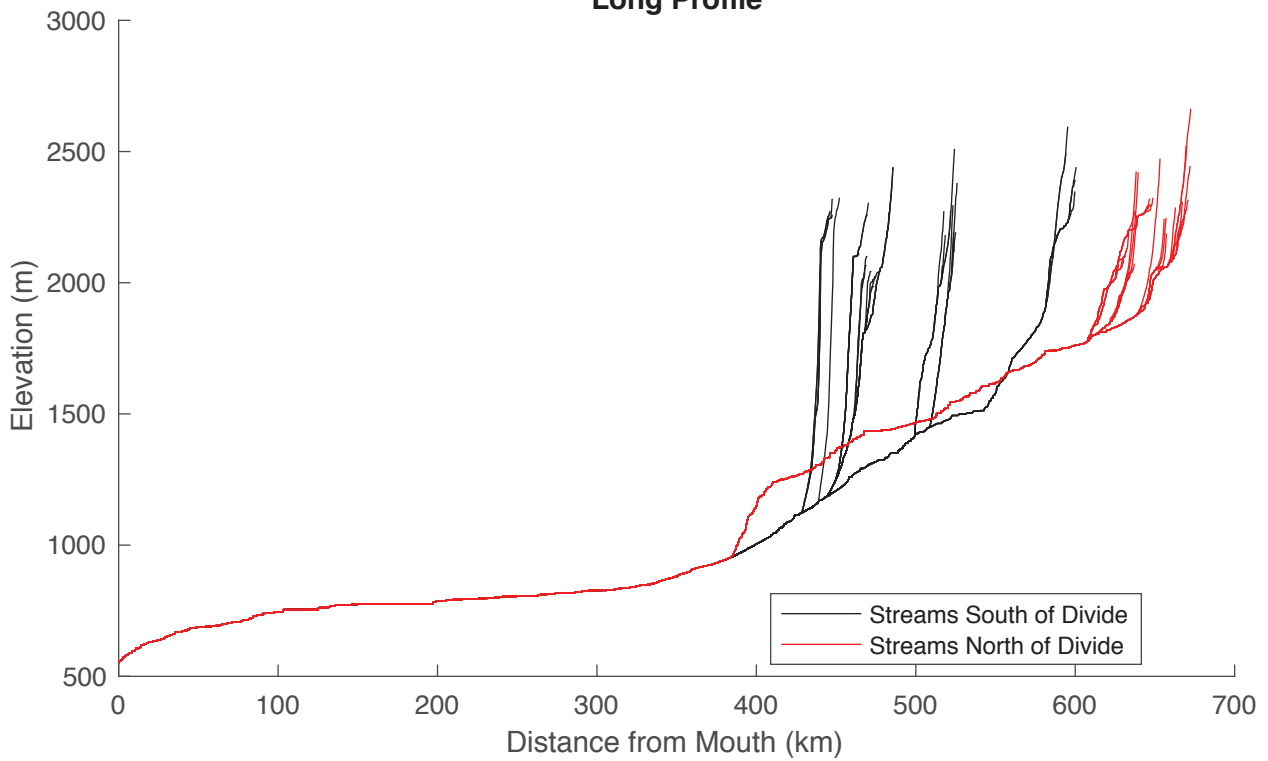


# Divide KV4

Chi - Z

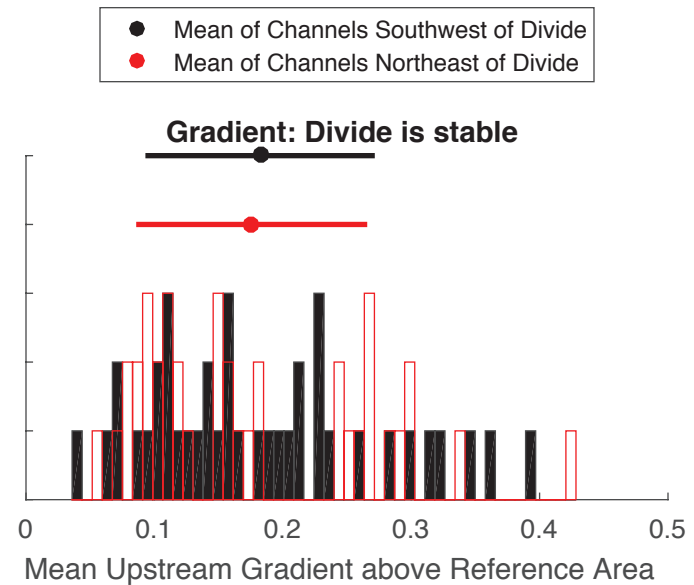
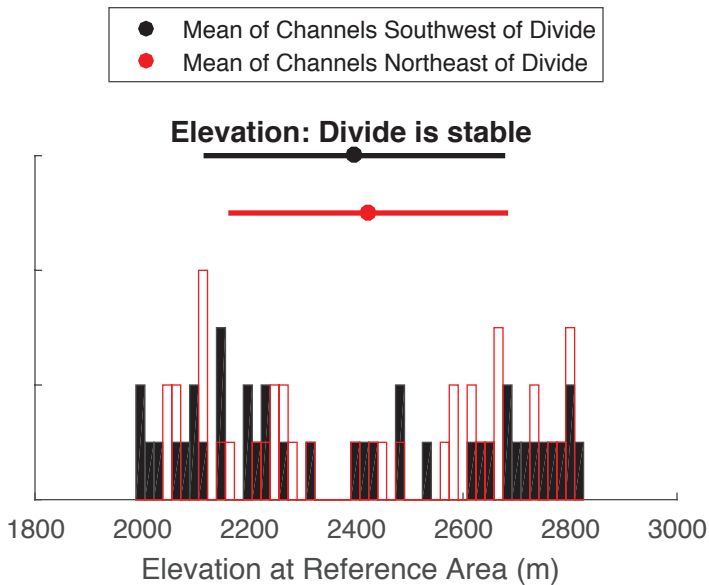
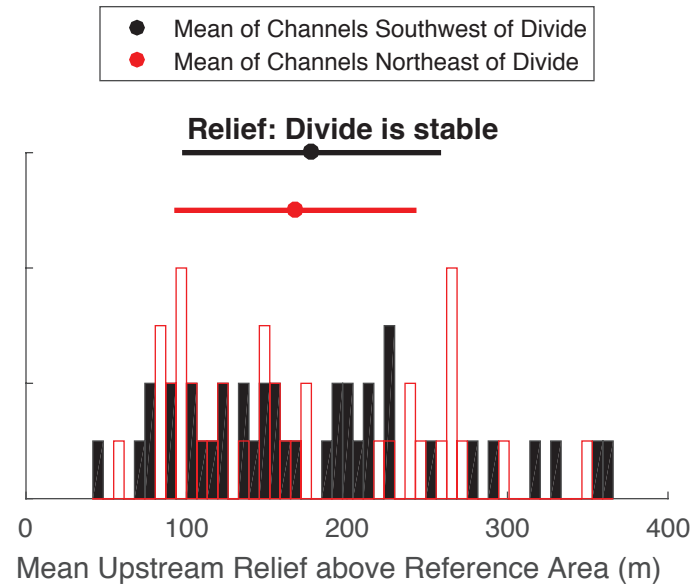
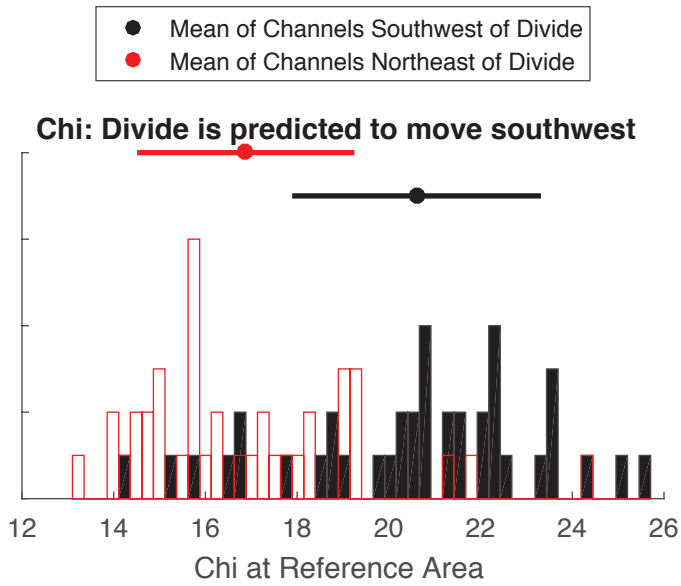


# Long Profile



Supplemental Figure 24

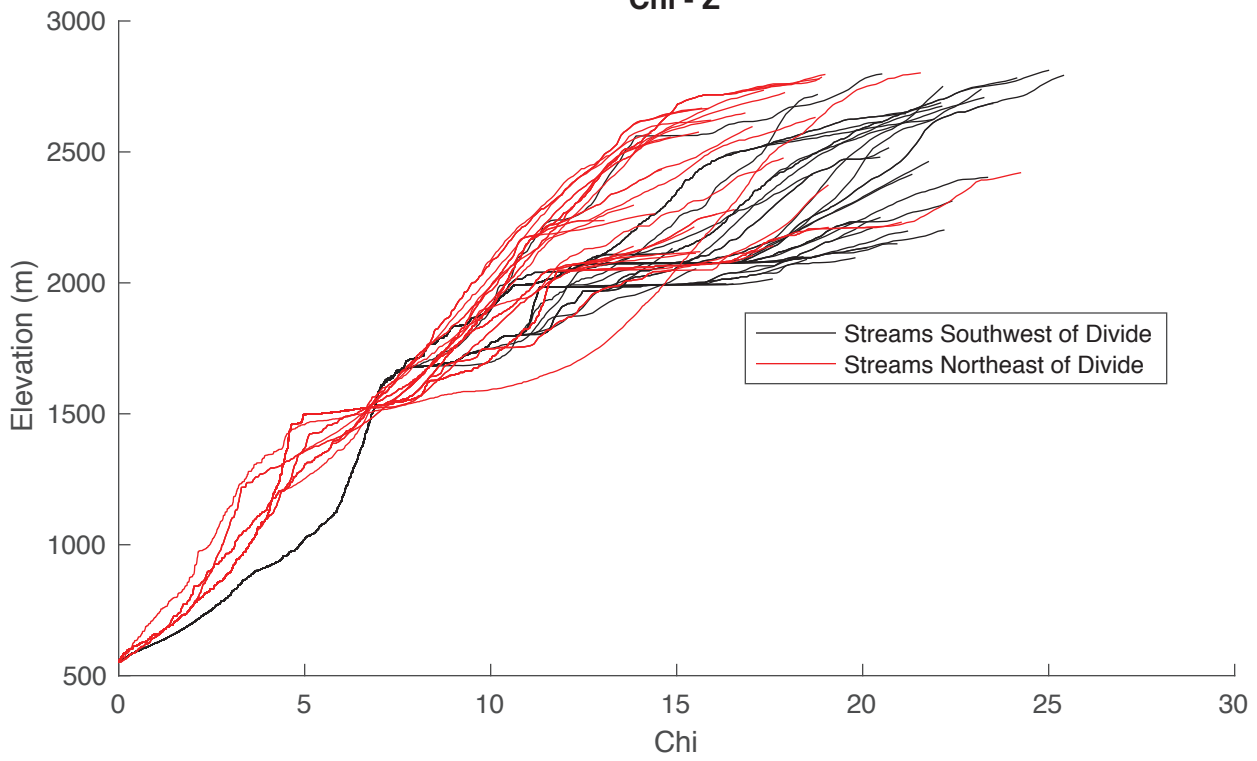
## Divide KV5



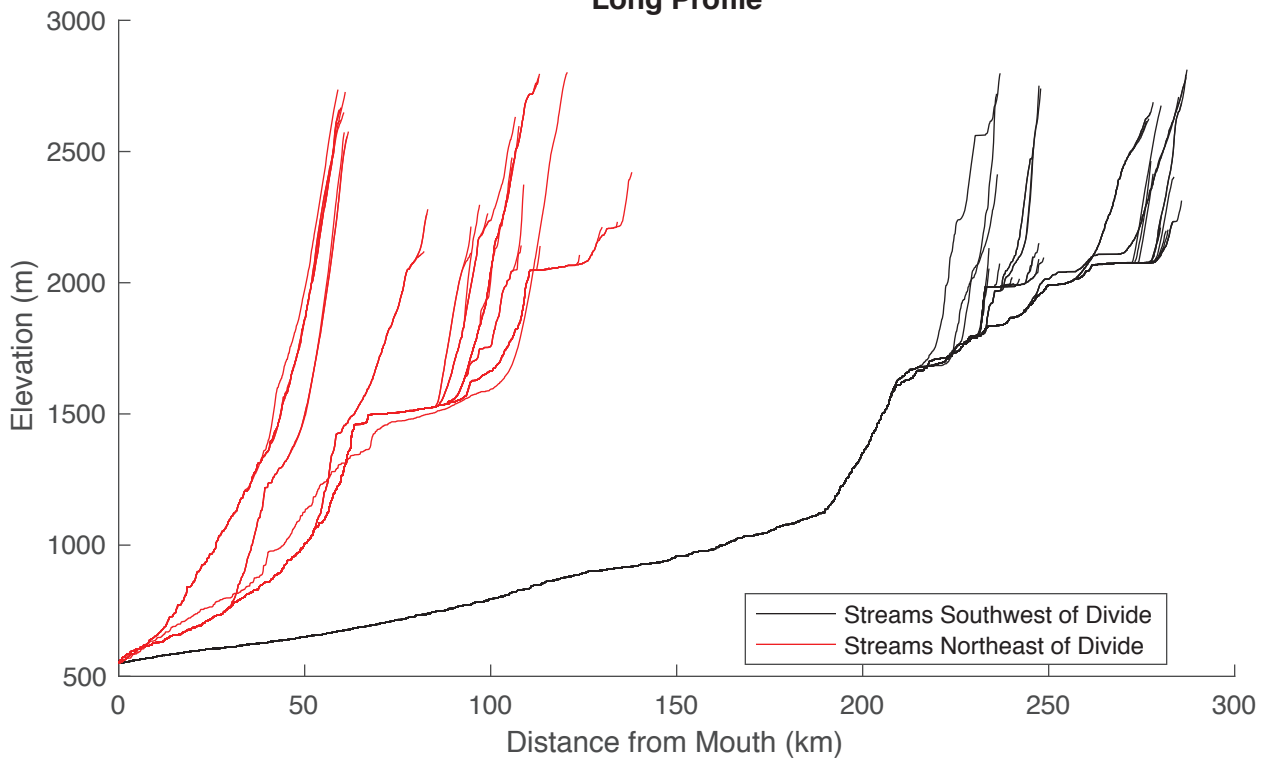
**Supplemental Figure 25**

# Divide KV5

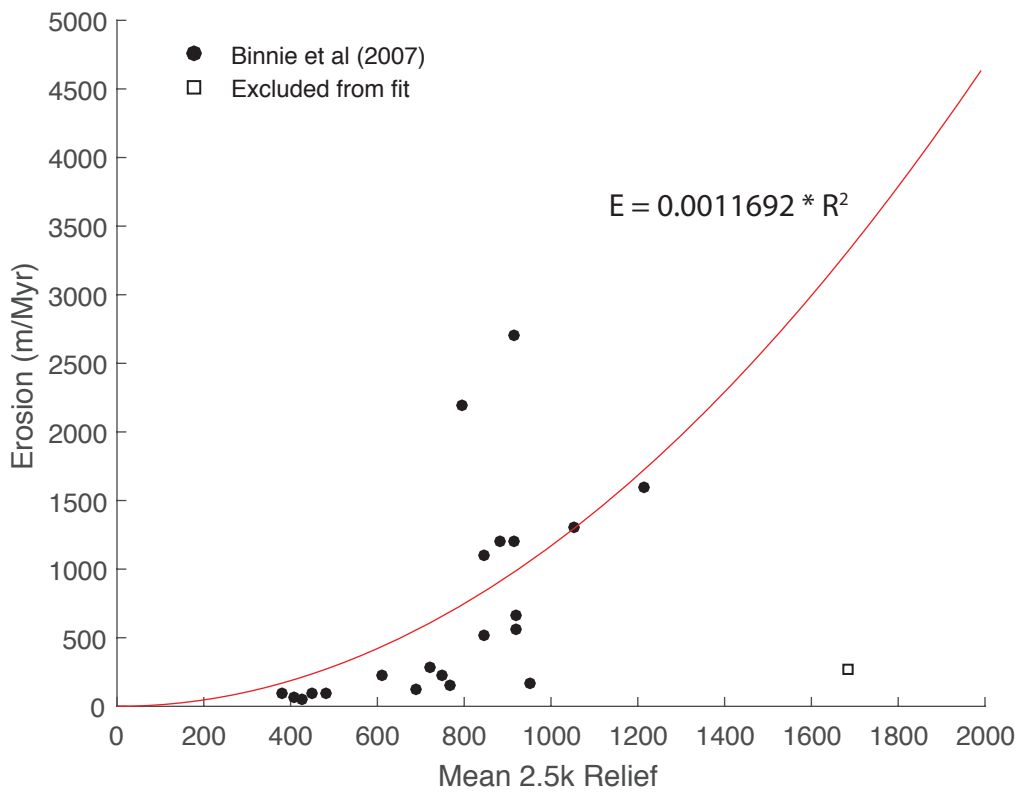
Chi - Z



# Long Profile

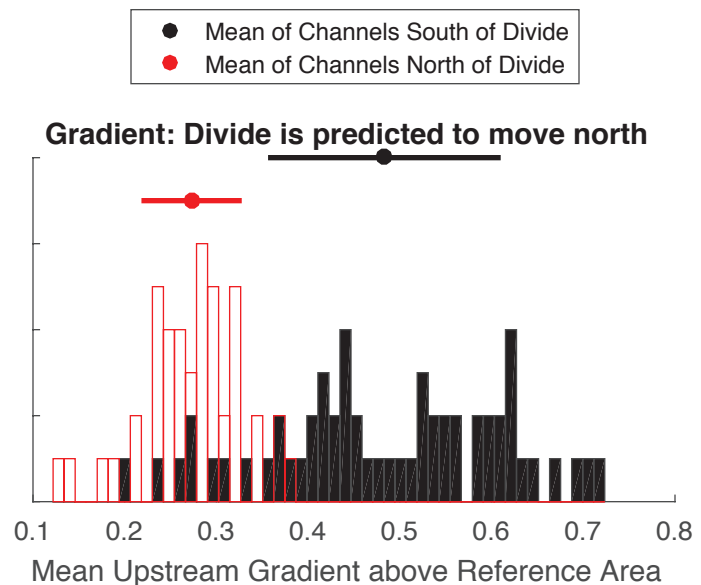
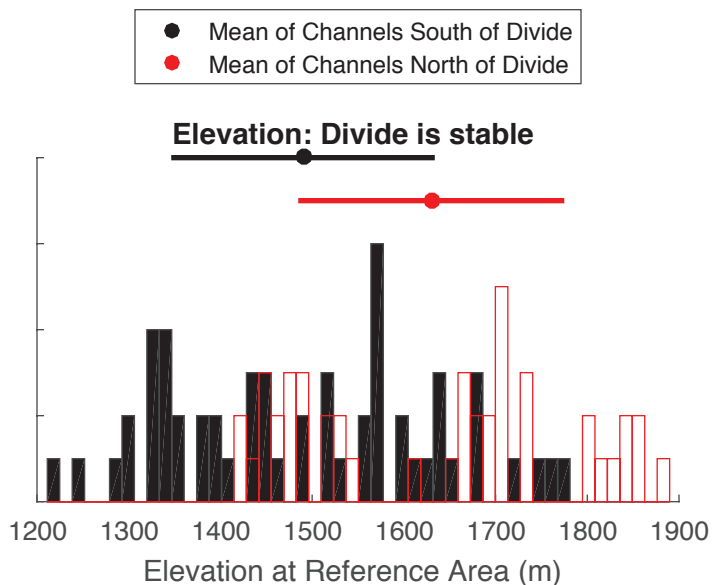
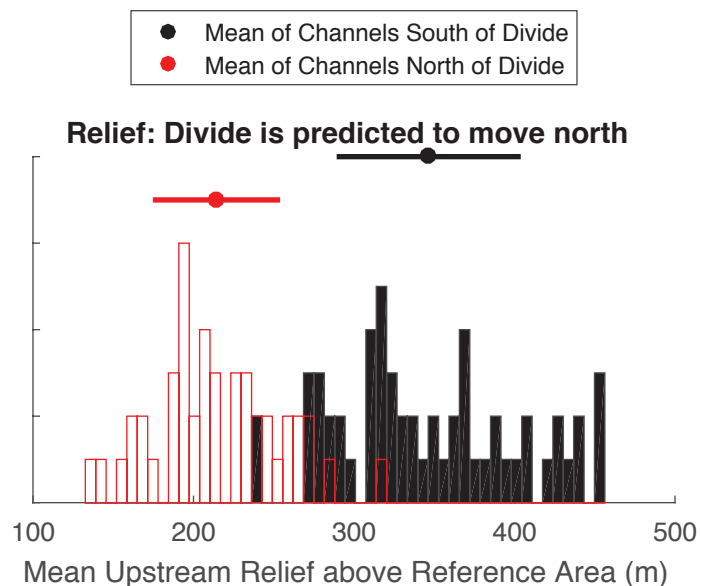
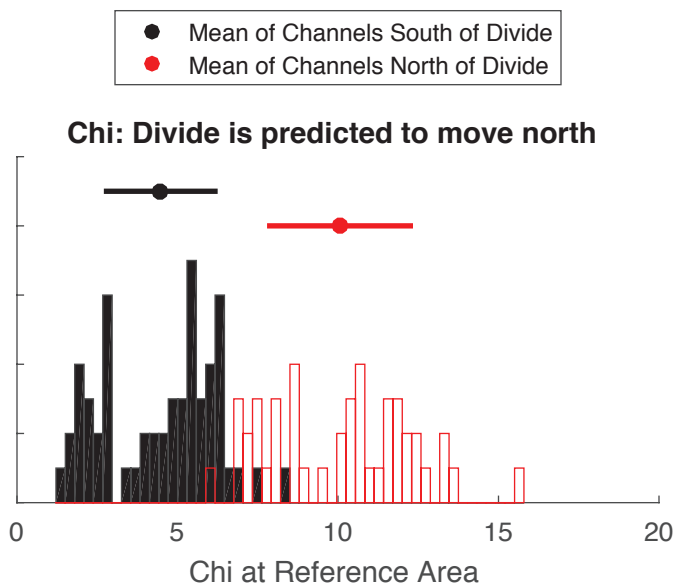


Supplemental Figure 26



**Supplemental Figure 27**

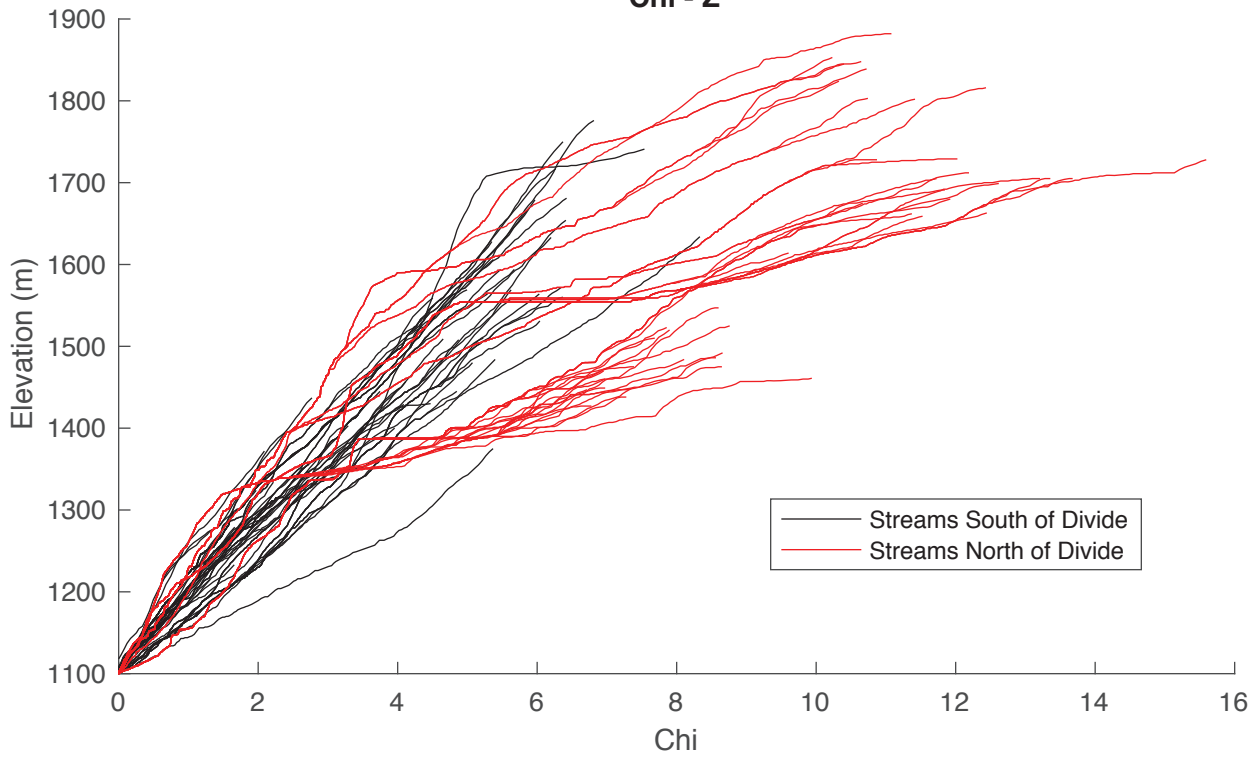
## Divide SB1



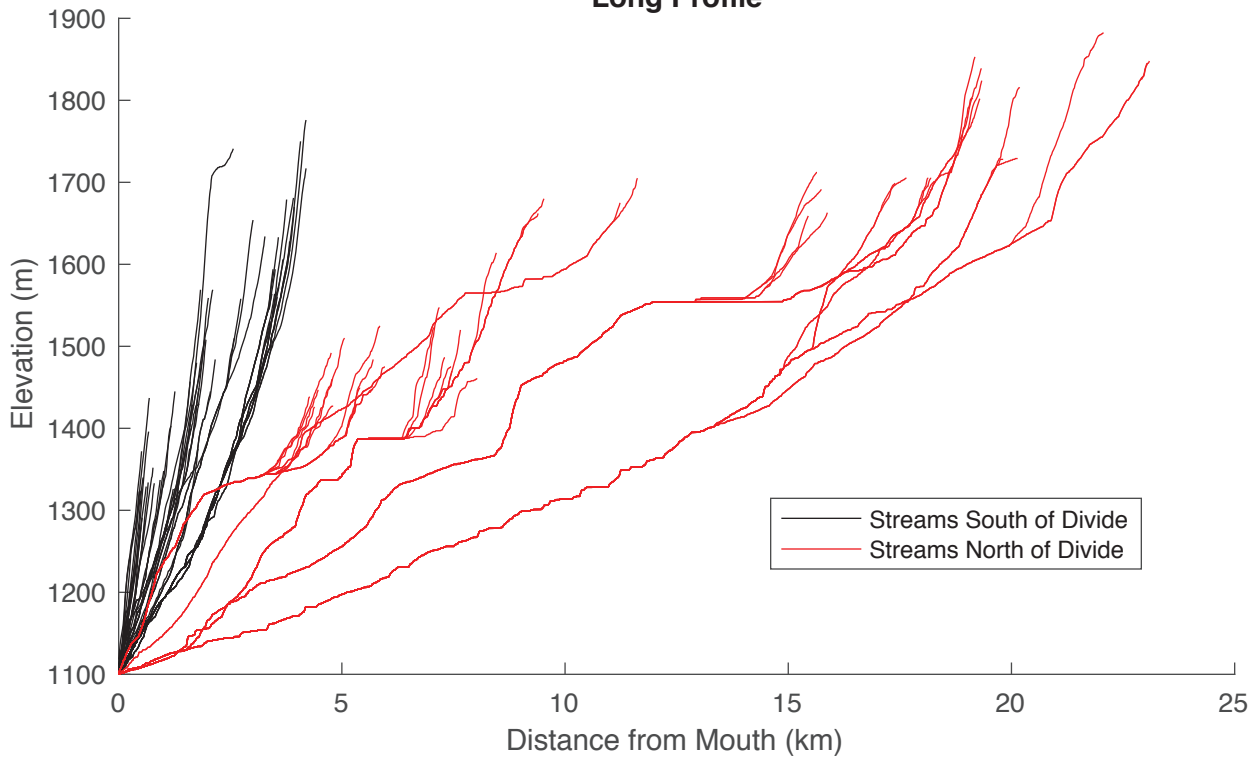
**Supplemental Figure 28**

### Divide SB1

Chi - Z



### Long Profile

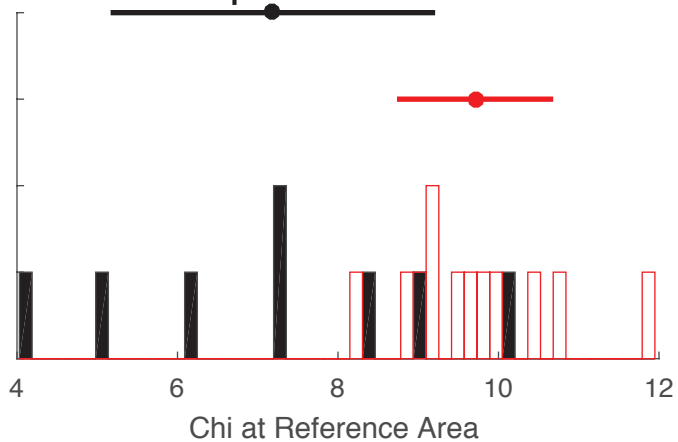


**Supplemental Figure 29**

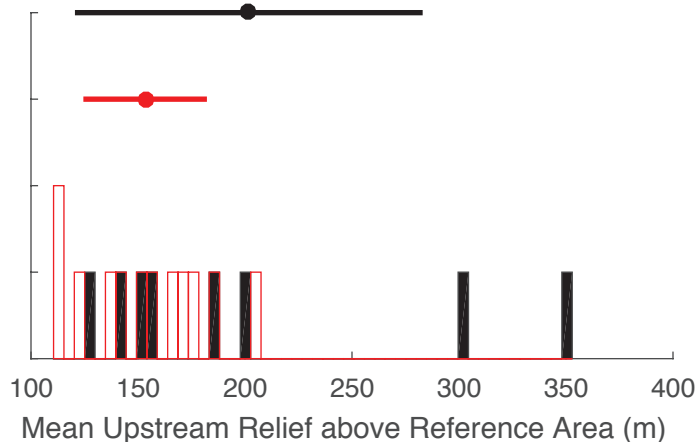
## Divide SB2



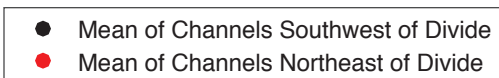
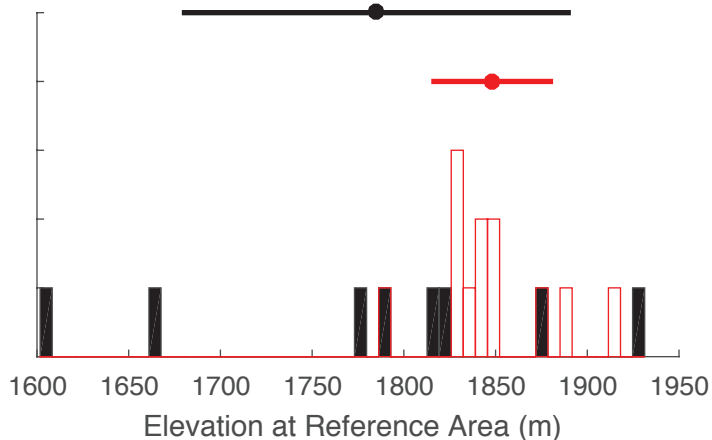
**Chi: Divide is predicted to move northeast**



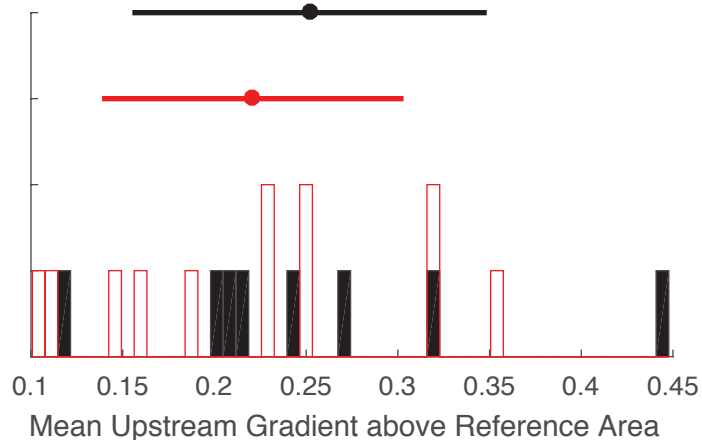
**Relief: Divide is stable**



**Elevation: Divide is stable**



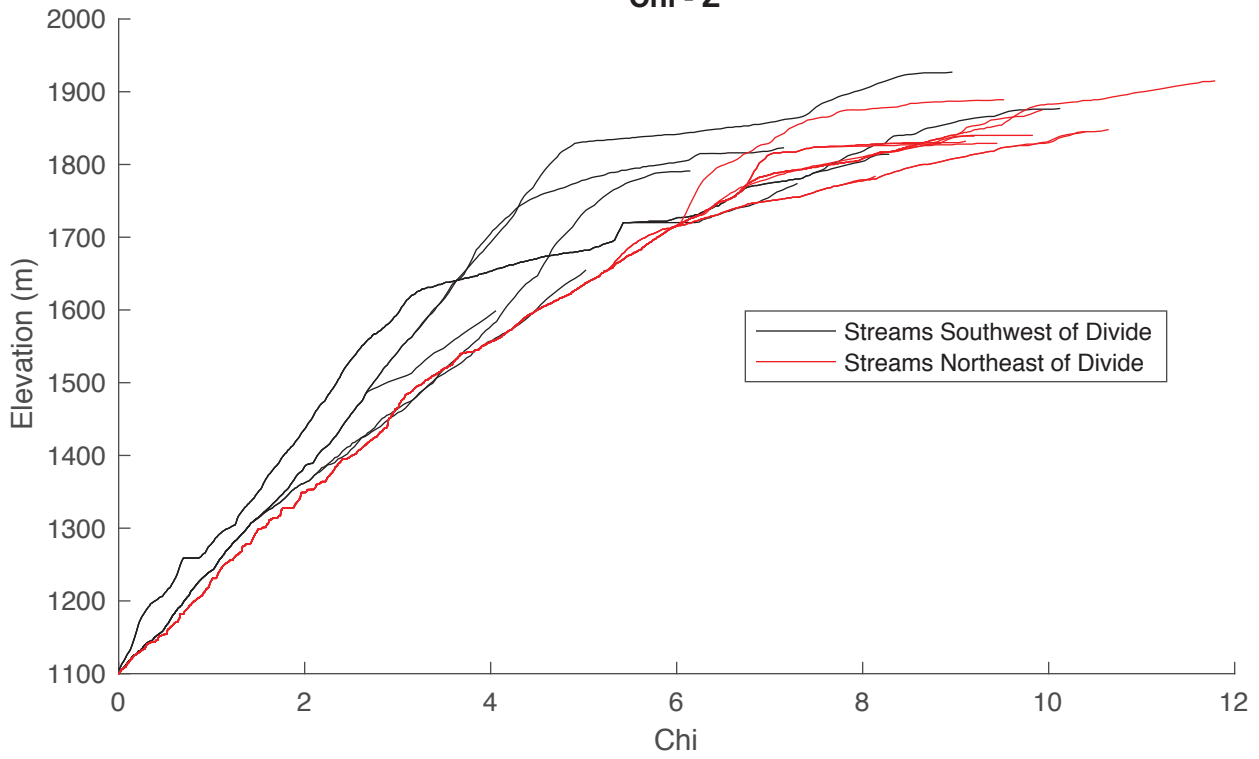
**Gradient: Divide is stable**



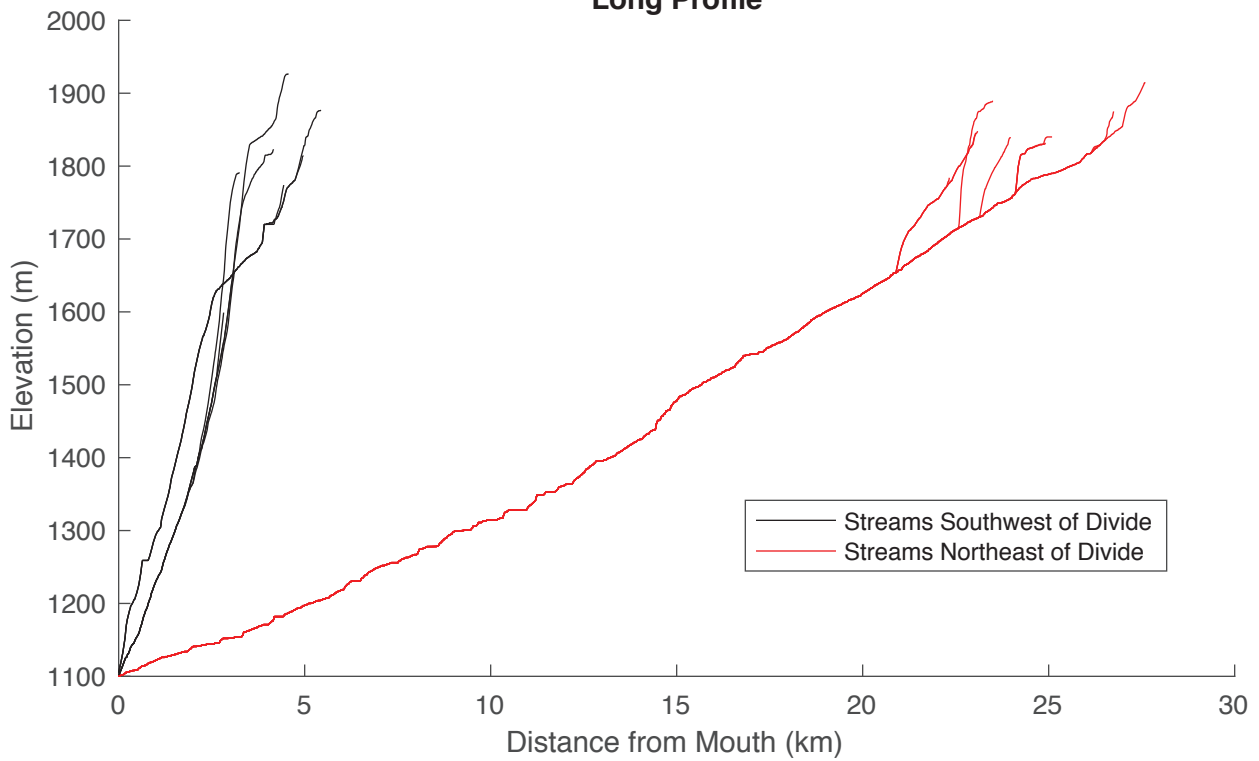
**Supplemental Figure 30**

## Divide SB2

Chi - Z



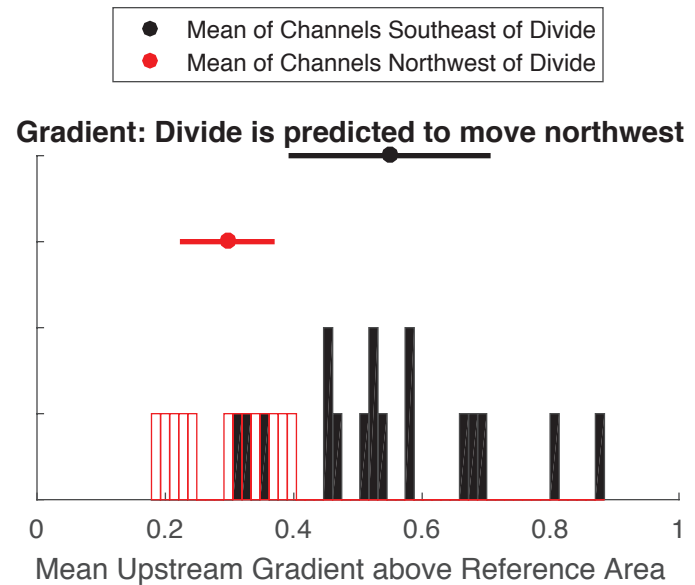
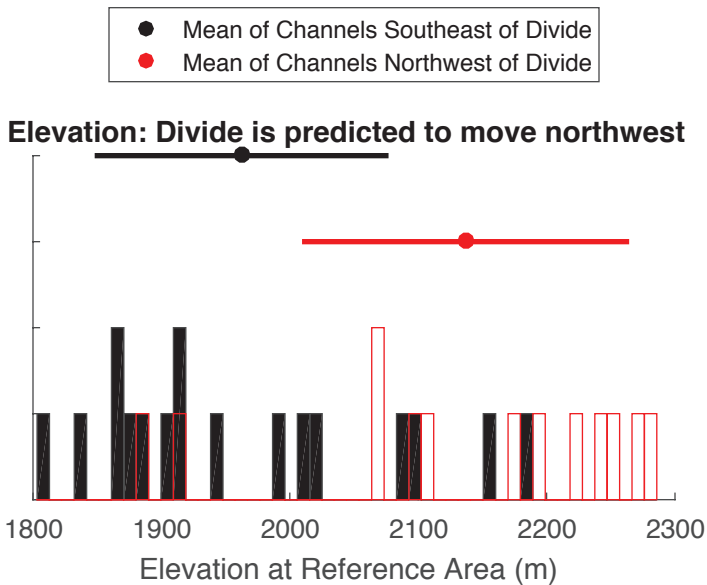
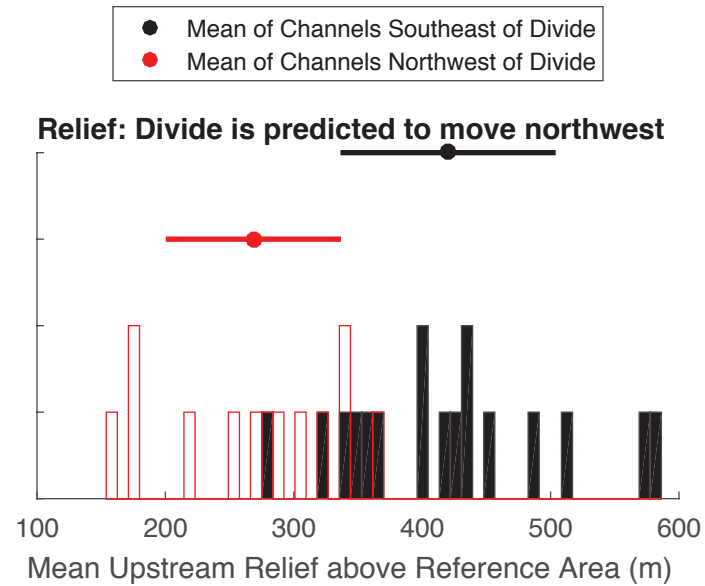
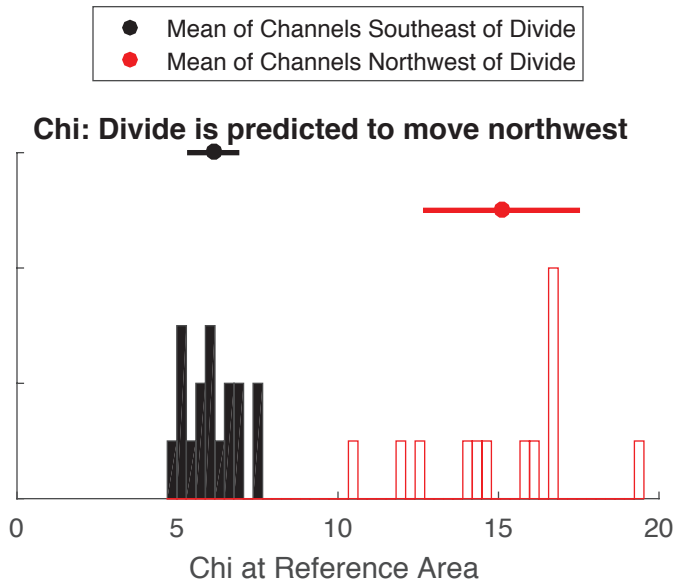
## Long Profile



Supplemental Figure 31



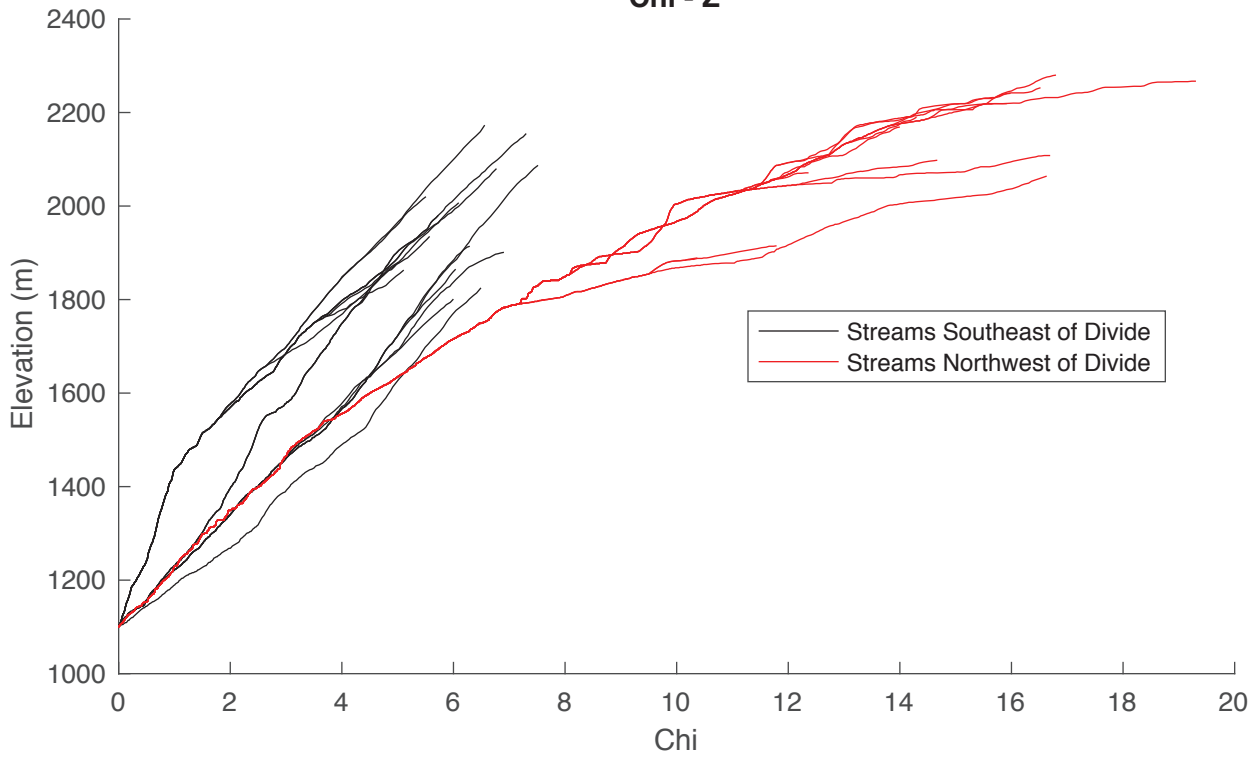
## Divide SB3



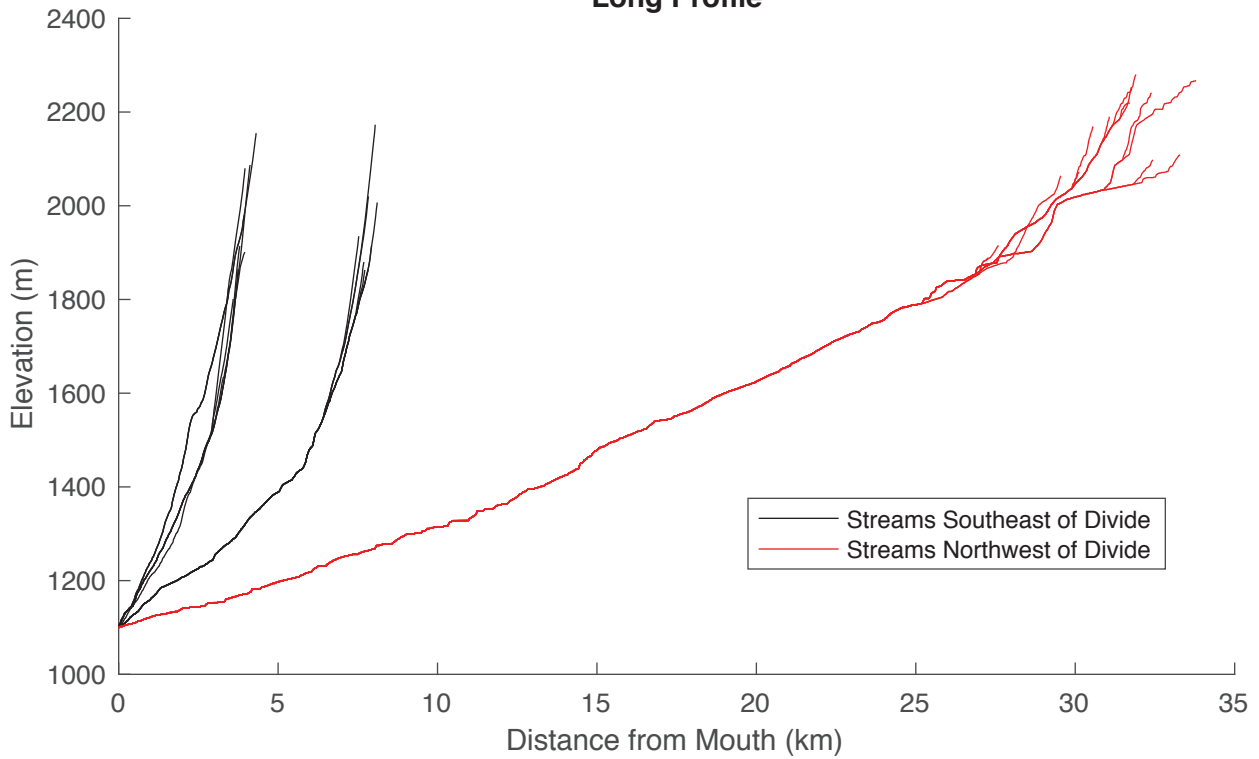
**Supplemental Figure 32**

### Divide SB3

Chi - Z

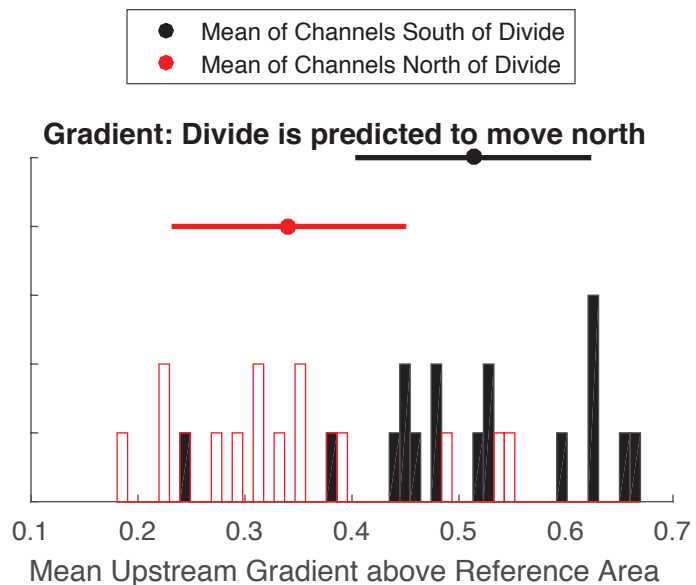
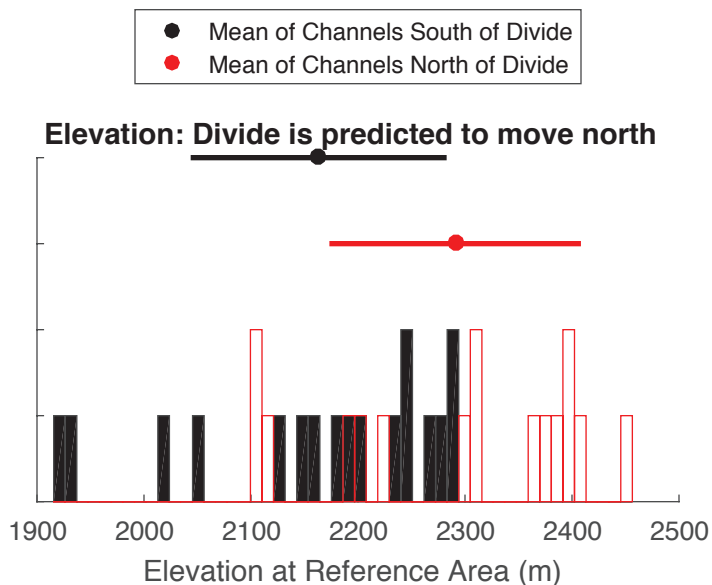
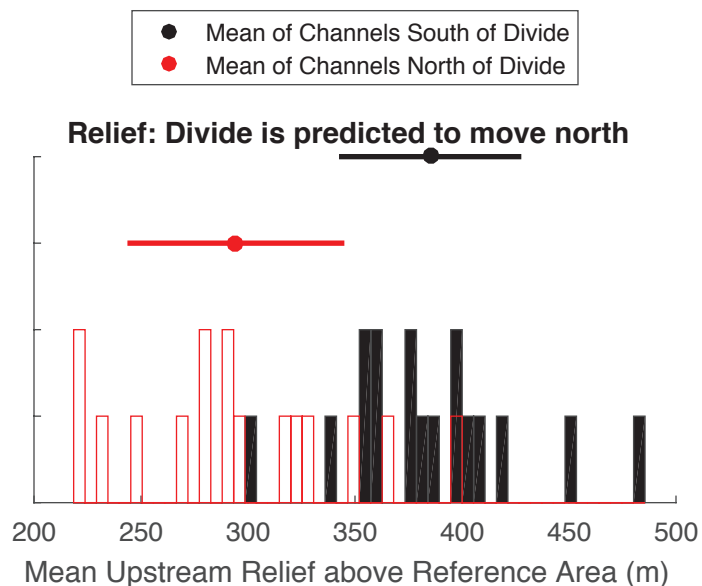
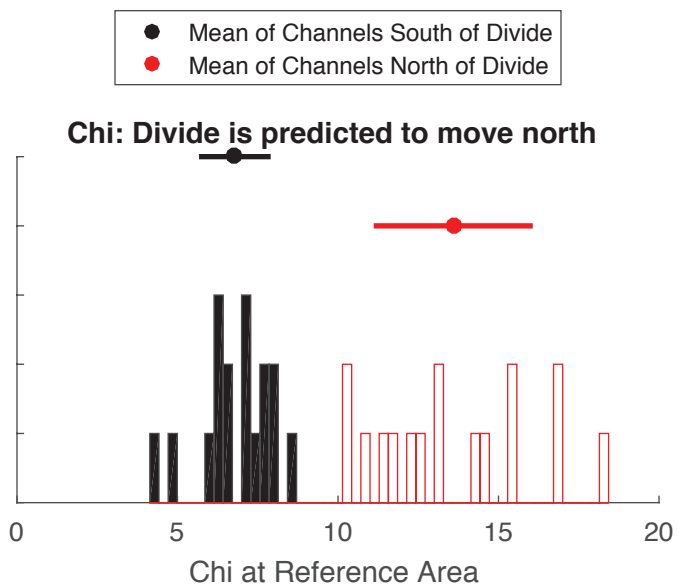


### Long Profile



Supplemental Figure 33

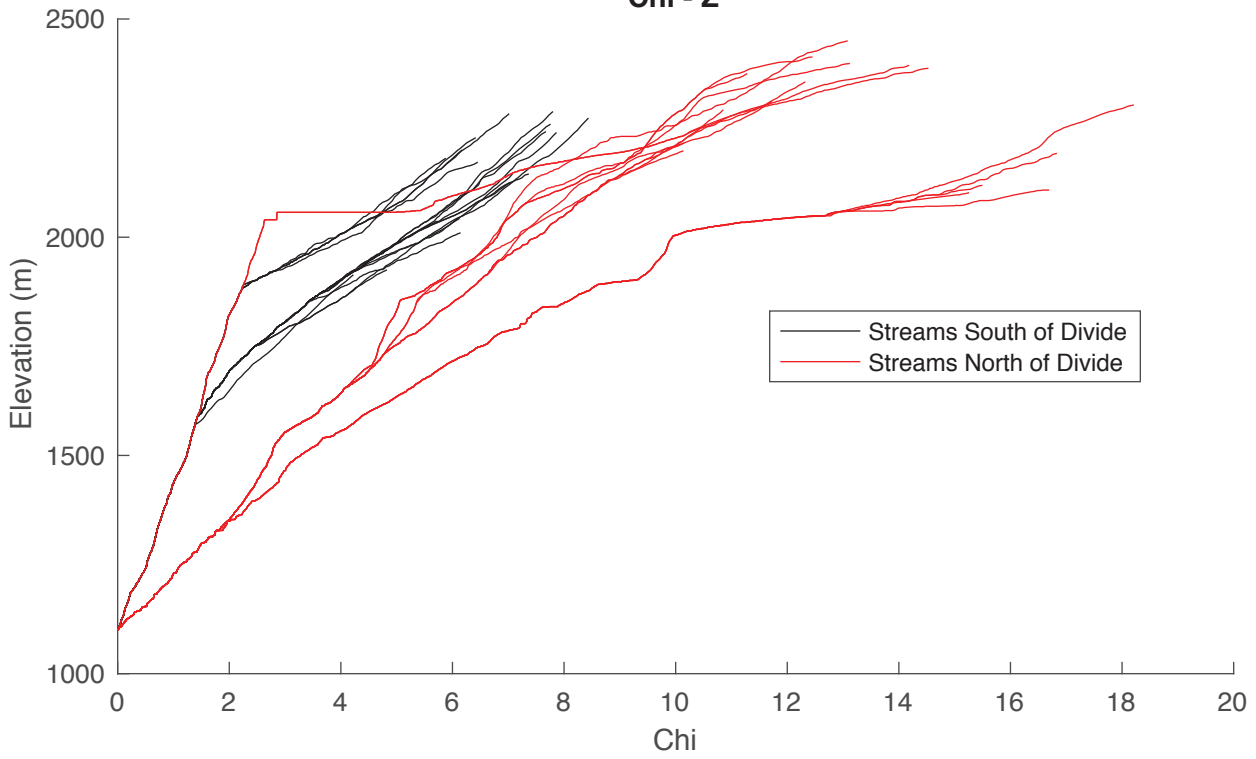
## Divide SB4



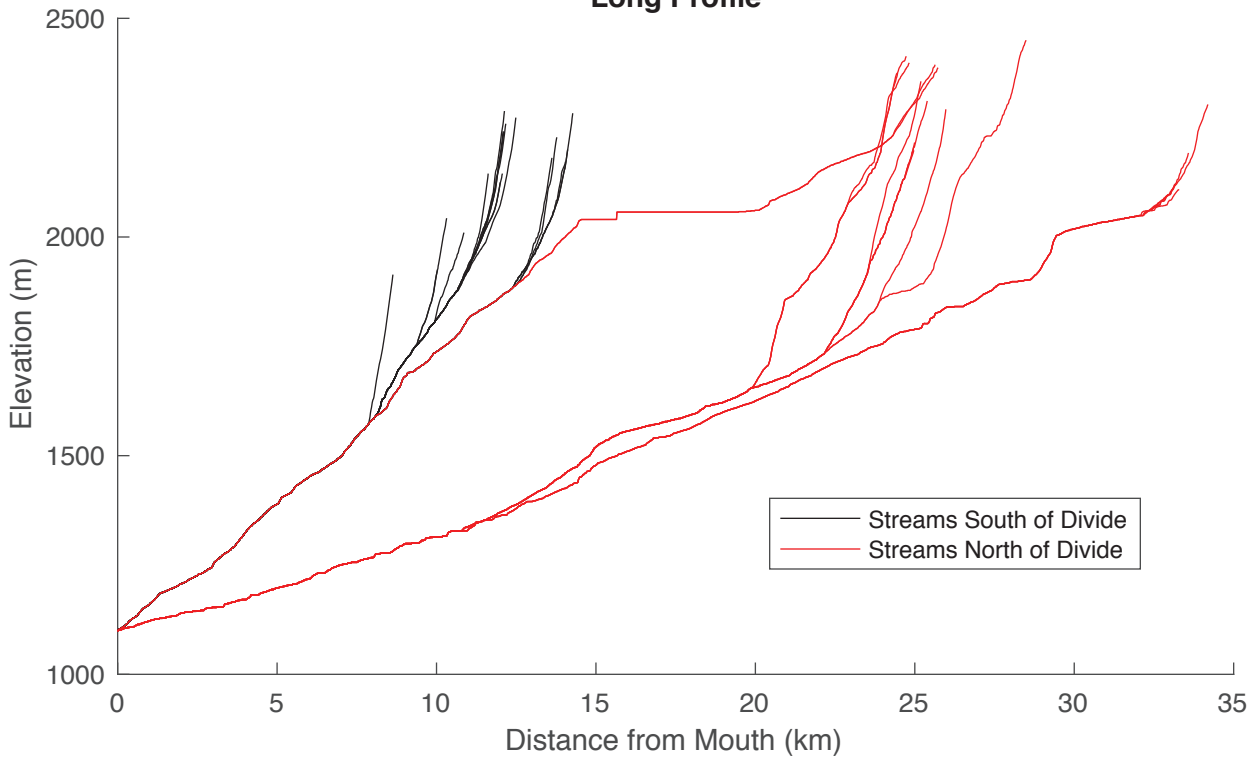
**Supplemental Figure 34**

### Divide SB4

Chi - Z

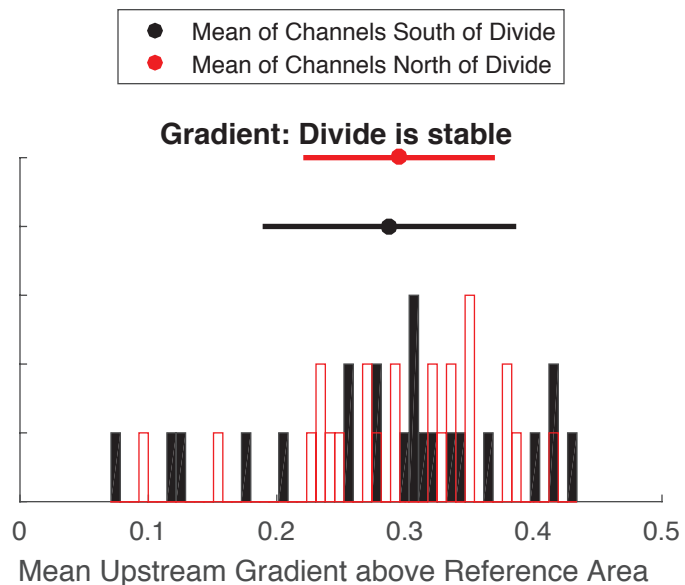
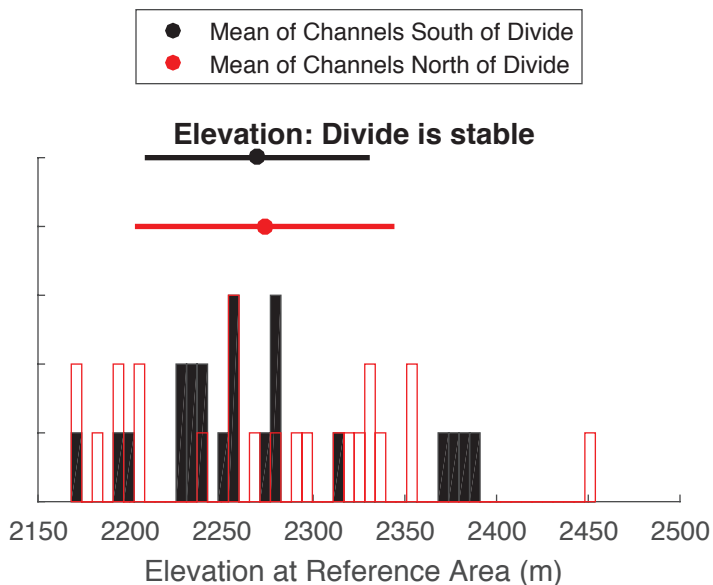
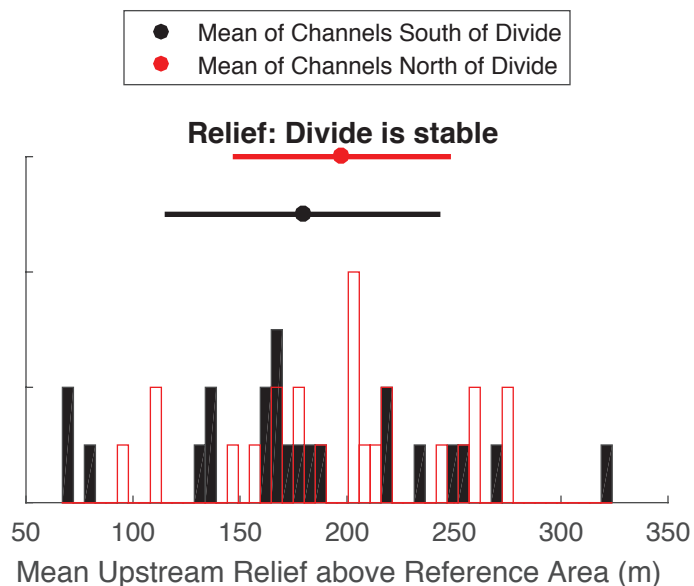
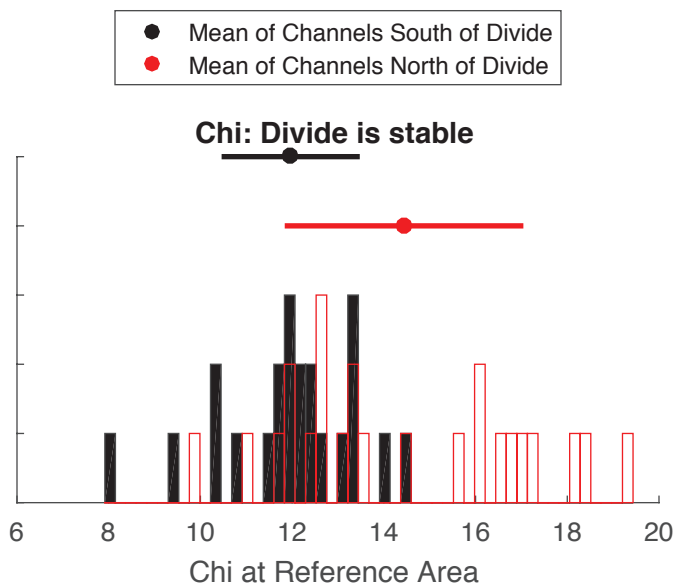


### Long Profile



**Supplemental Figure 35**

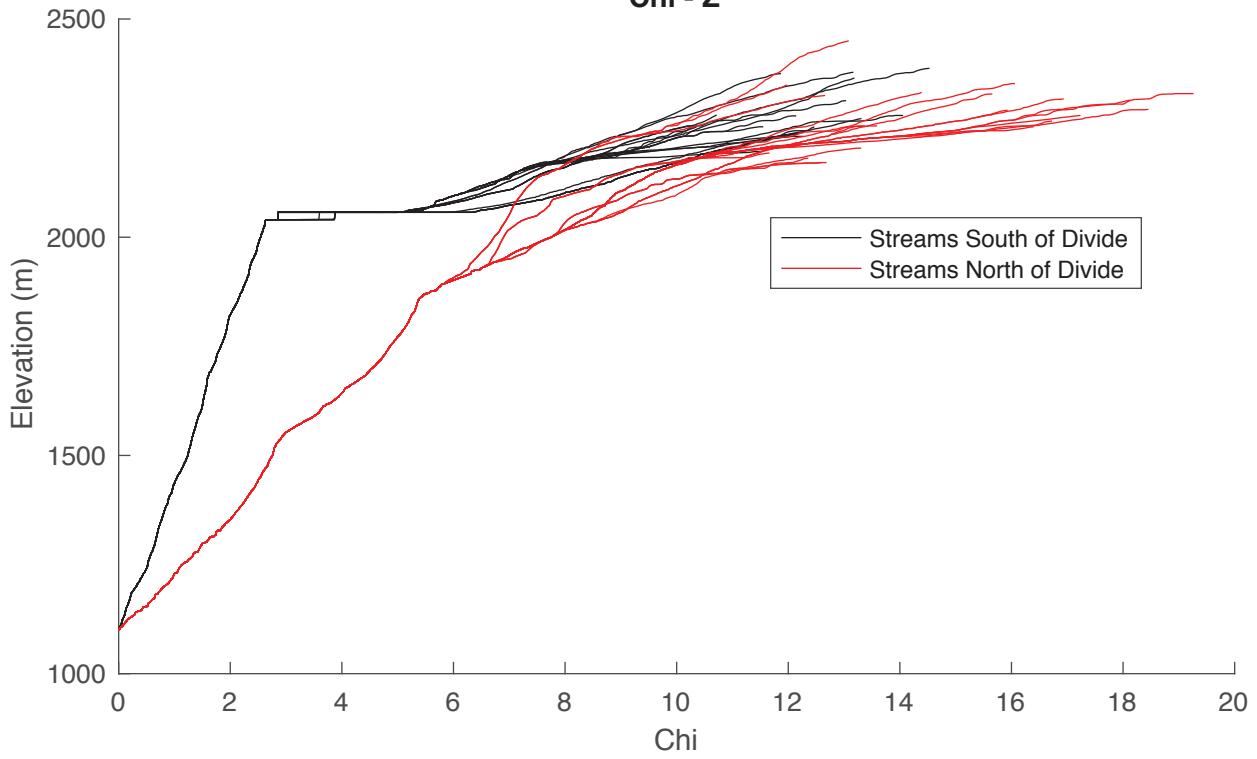
## Divide SB5



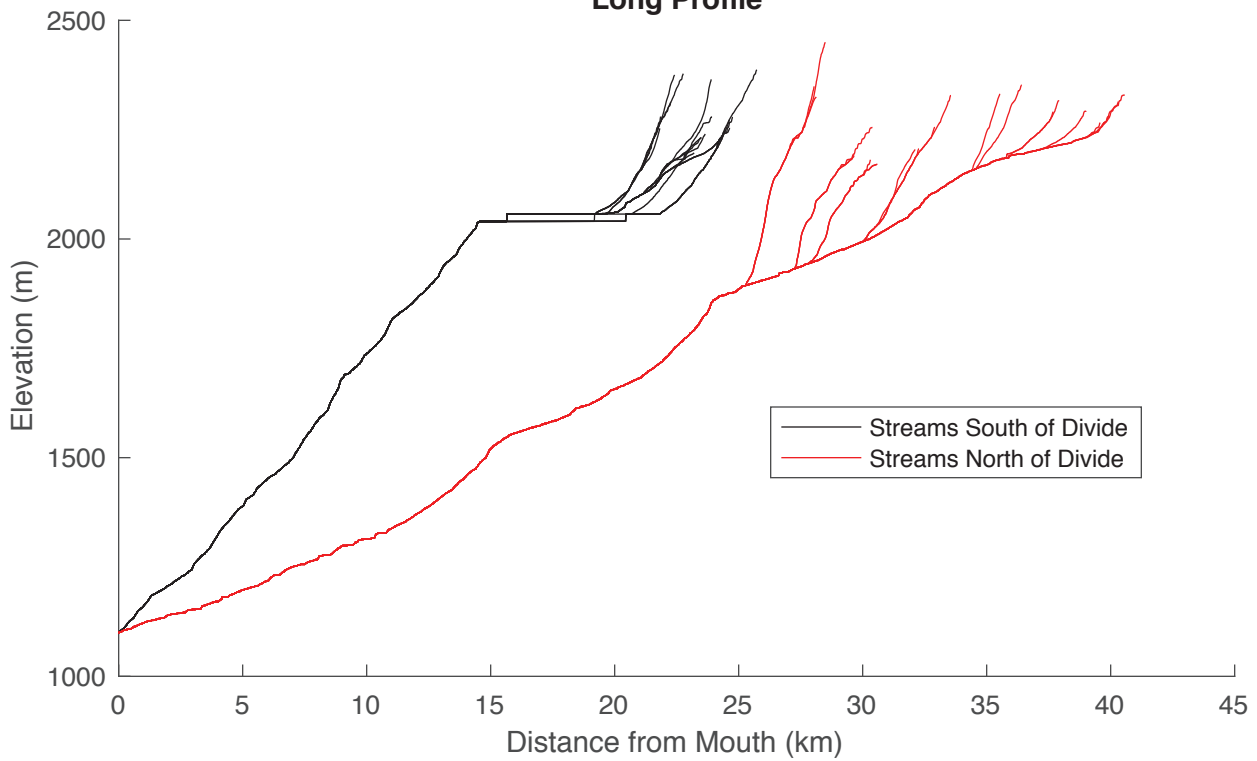
**Supplemental Figure 36**

# Divide SB5

Chi - Z

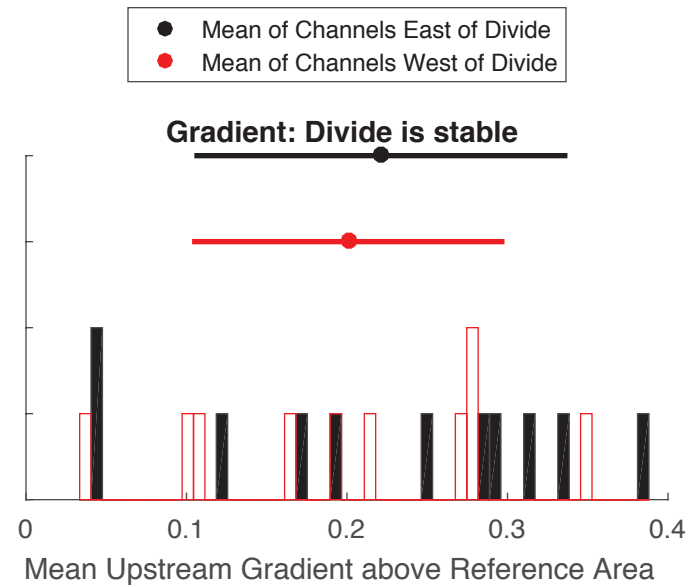
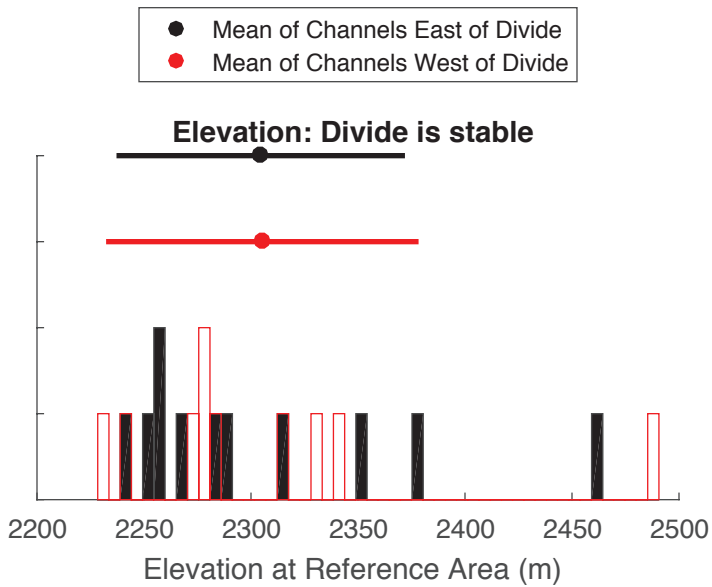
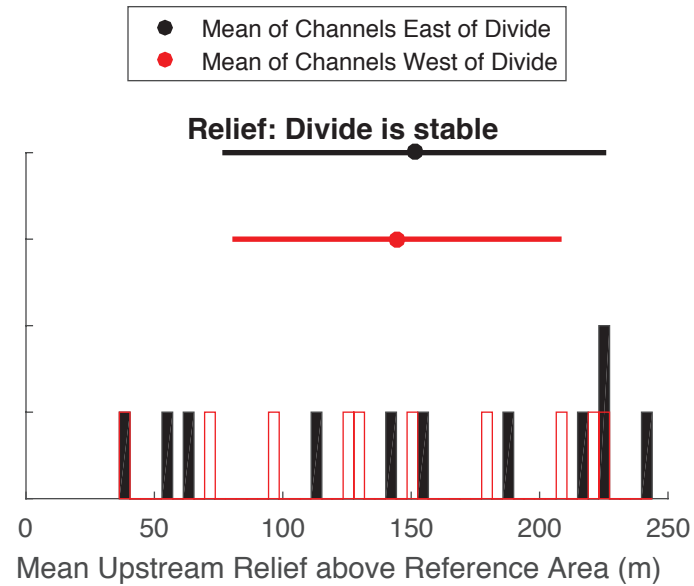
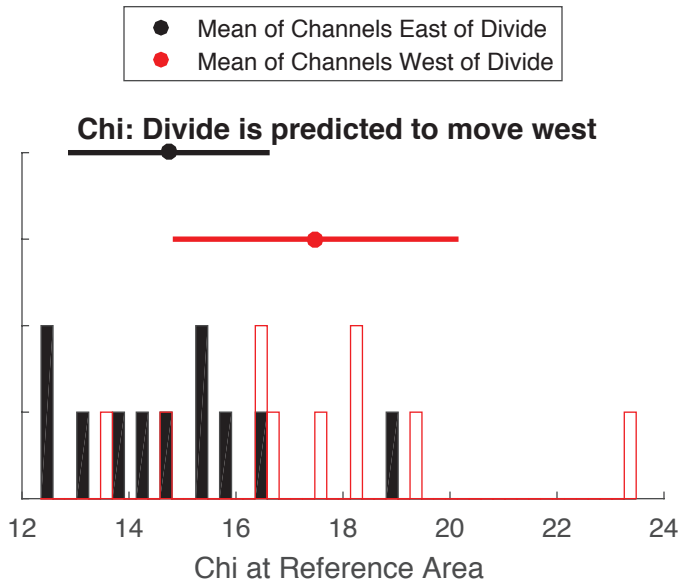


# Long Profile



Supplemental Figure 37

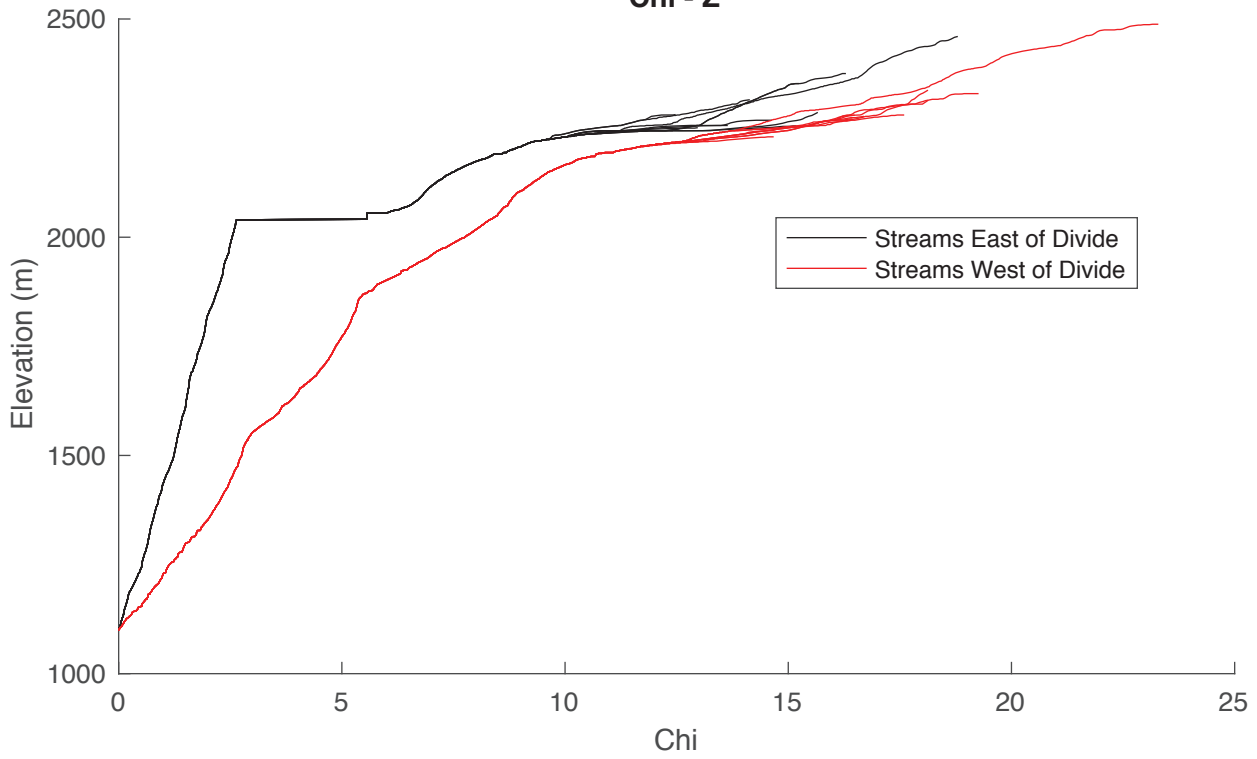
## Divide SB6



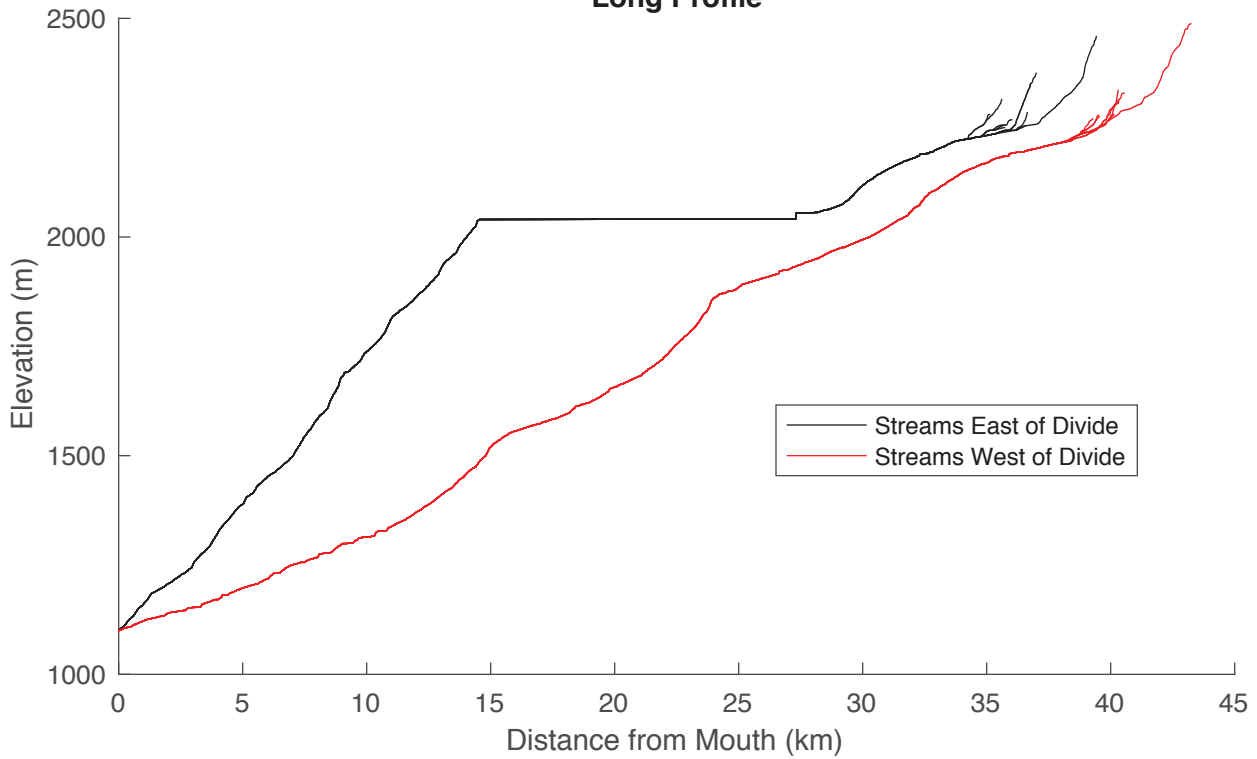
**Supplemental Figure 38**

### Divide SB6

Chi - Z



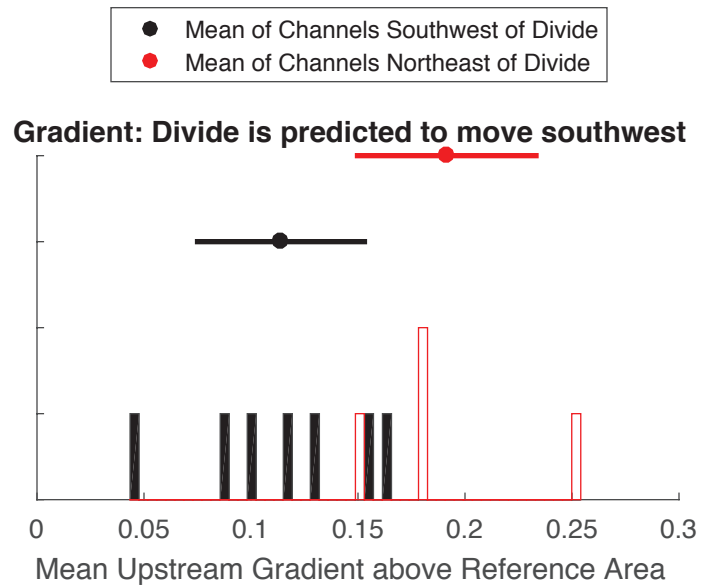
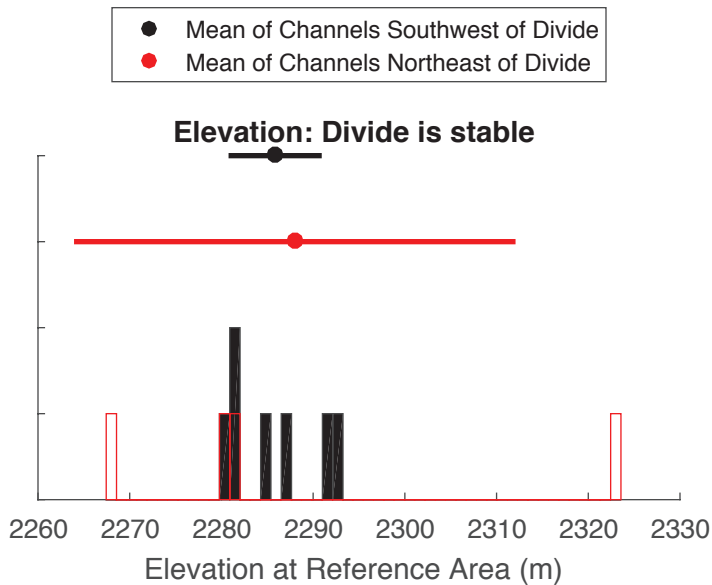
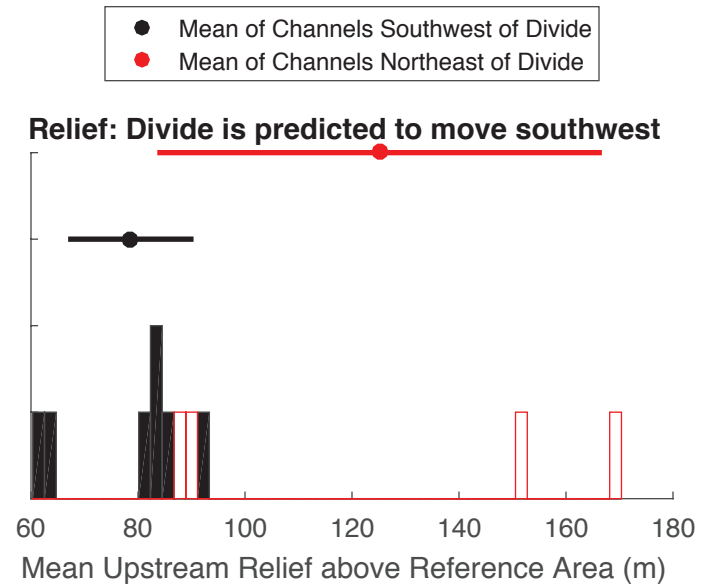
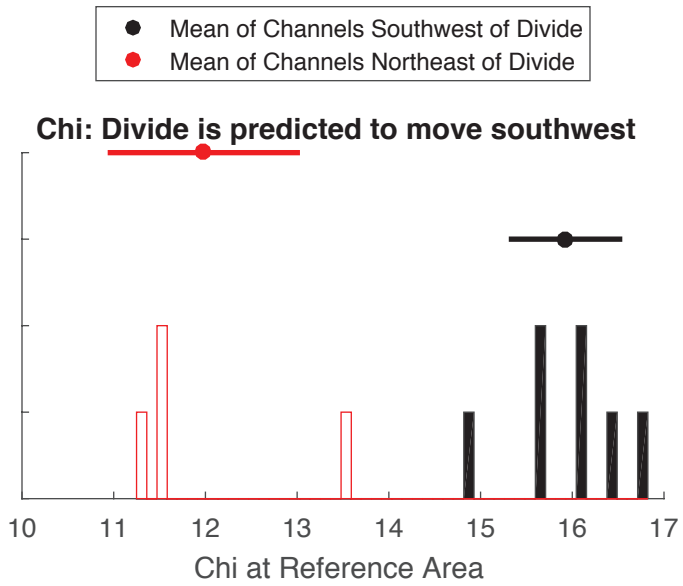
### Long Profile



Supplemental Figure 39



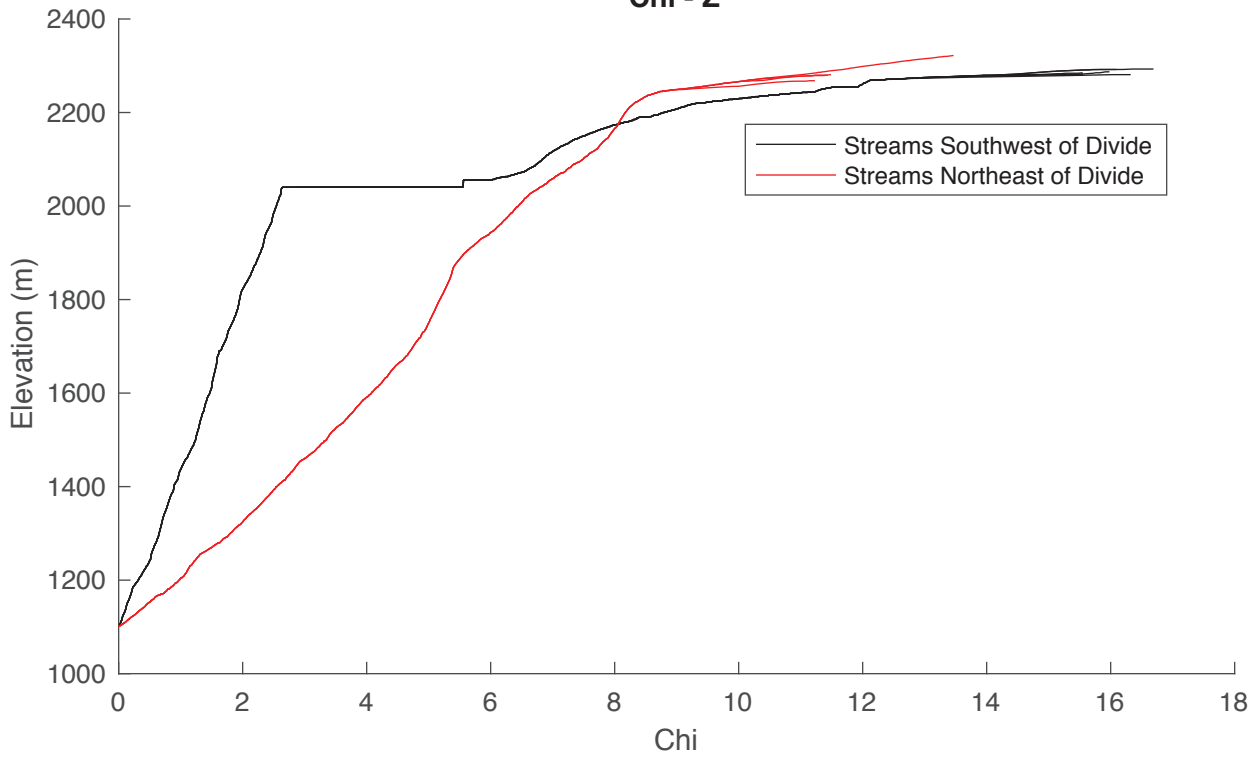
## Divide SB7



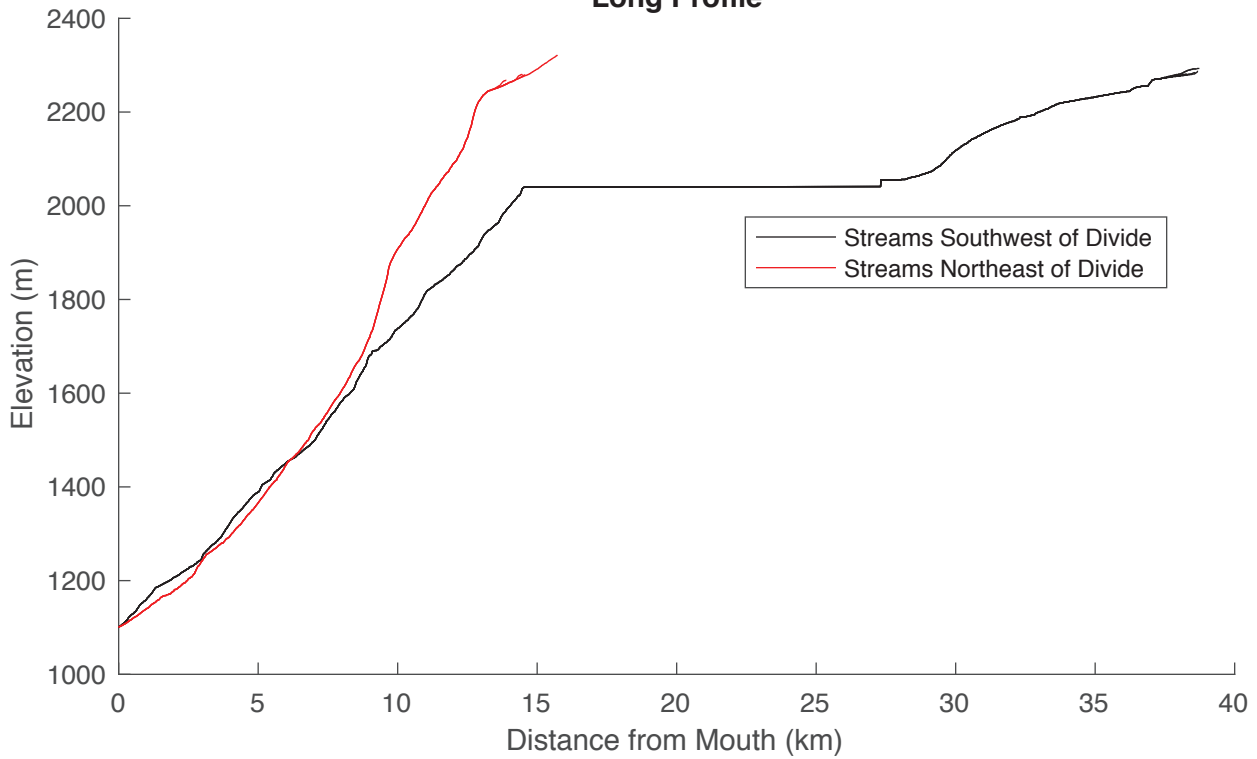
**Supplemental Figure 40**

# Divide SB7

## Chi - Z

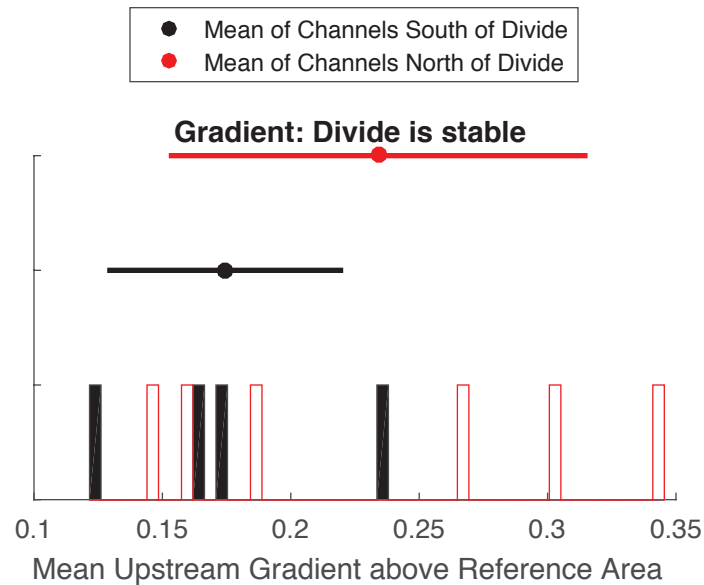
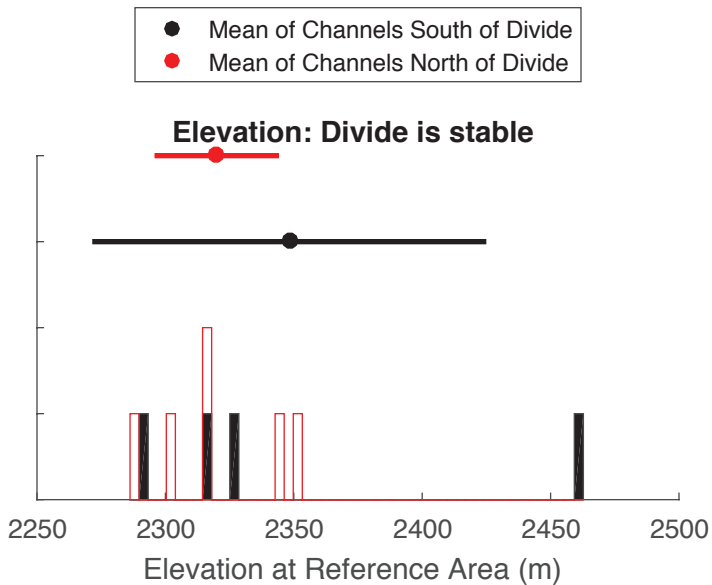
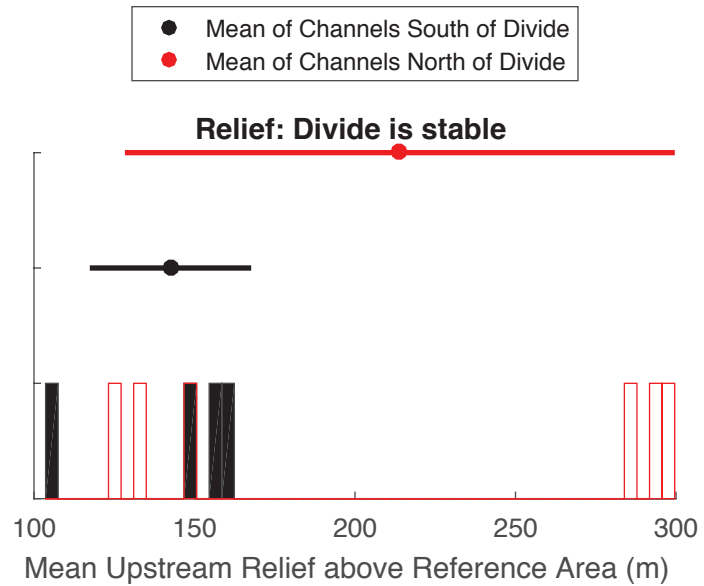
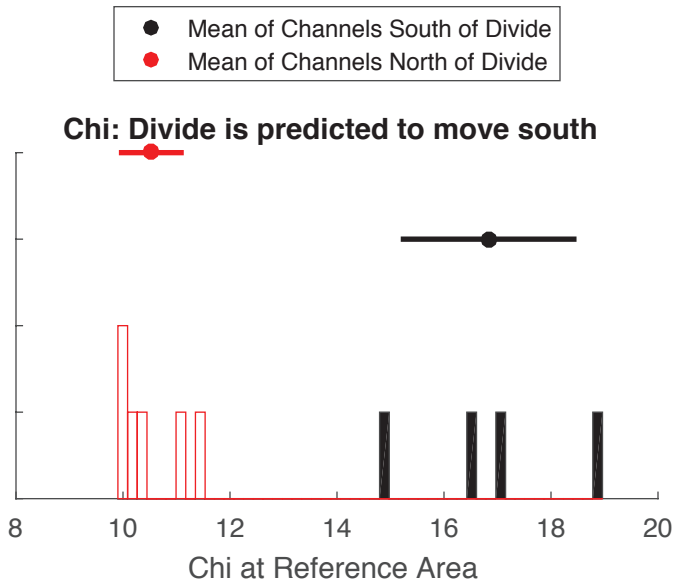


## Long Profile



Supplemental Figure 41

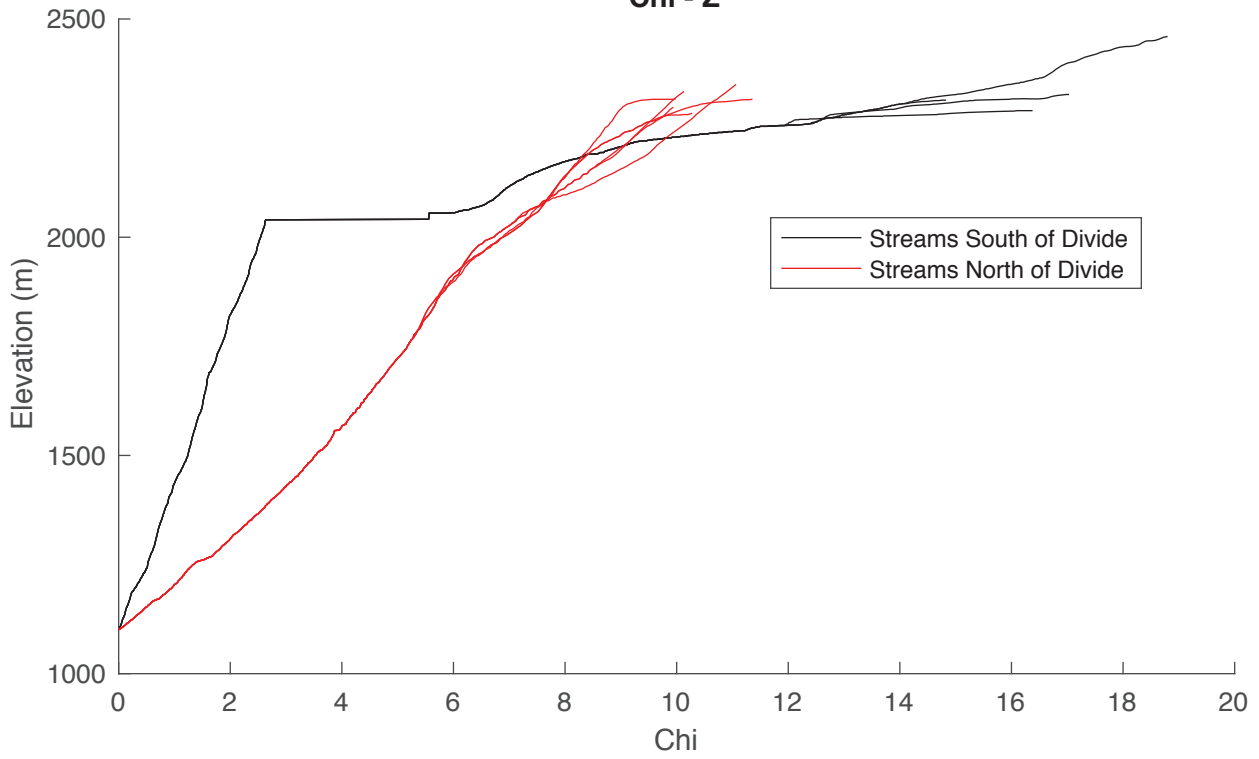
## Divide SB8



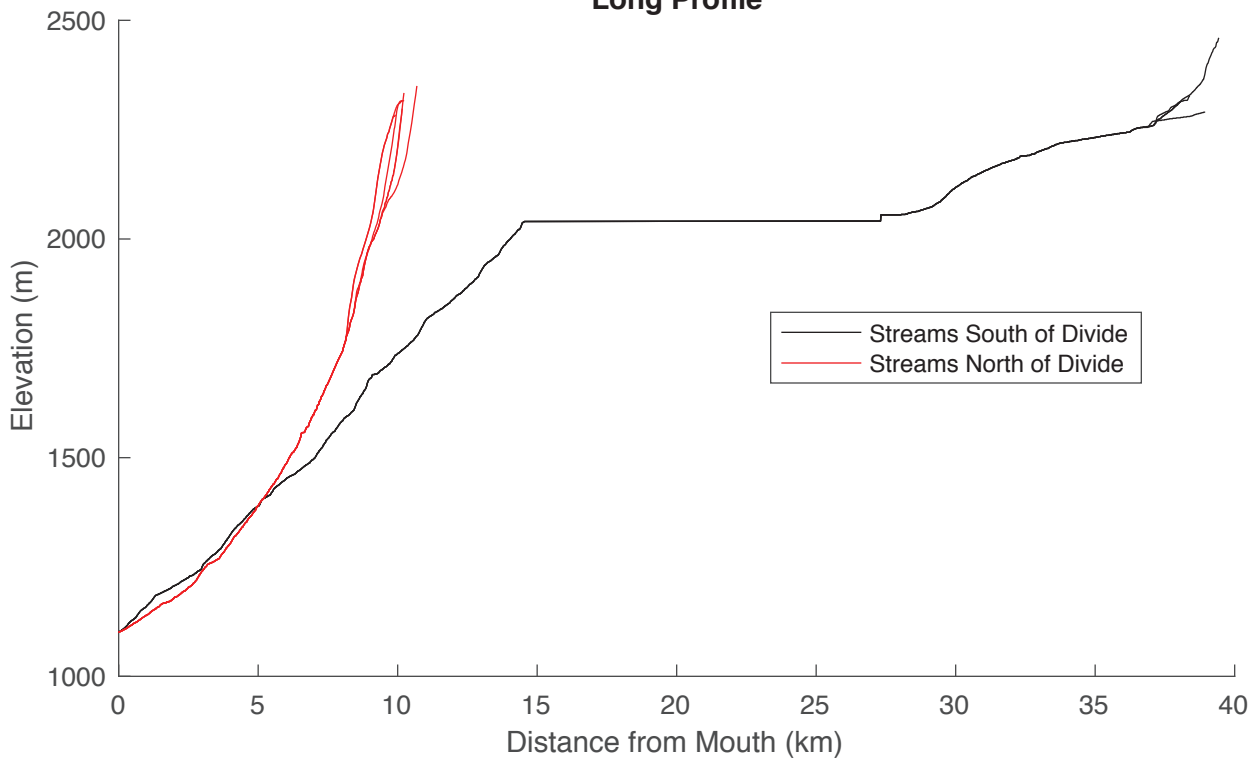
**Supplemental Figure 42**

### Divide SB8

Chi - Z

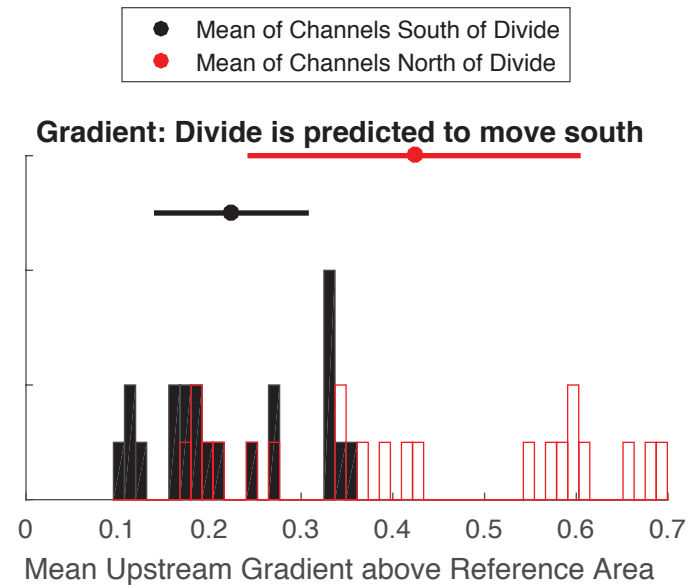
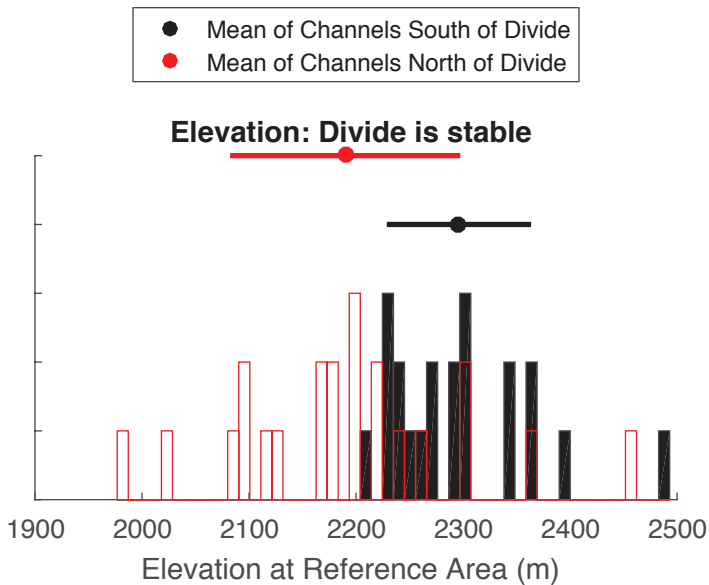
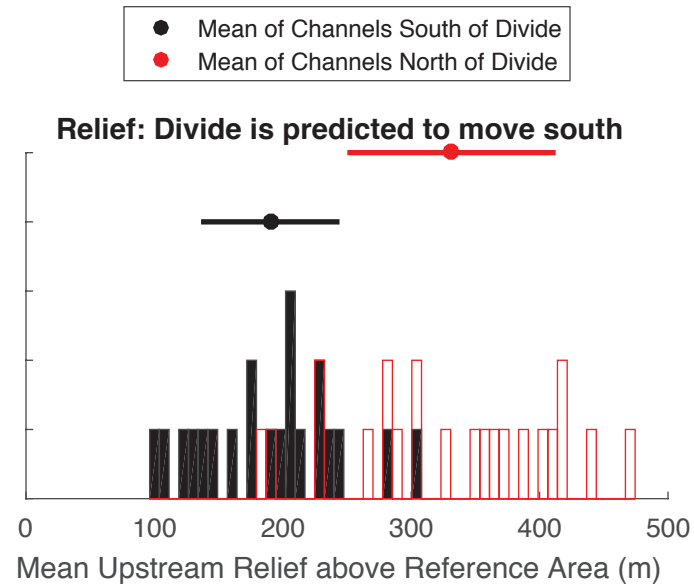
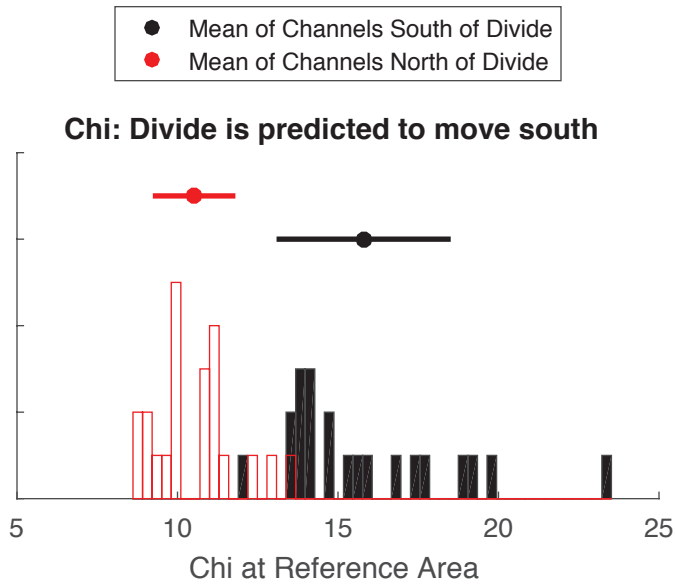


### Long Profile



Supplemental Figure 43

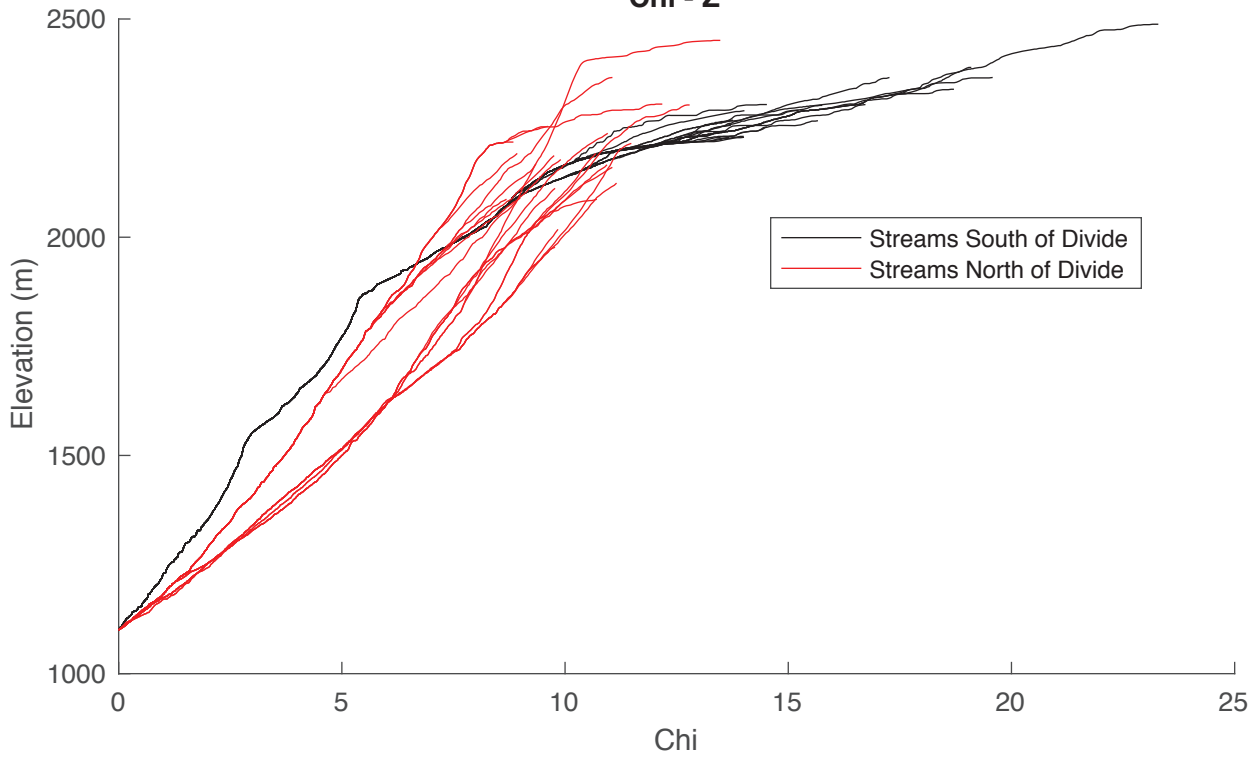
## Divide SB9



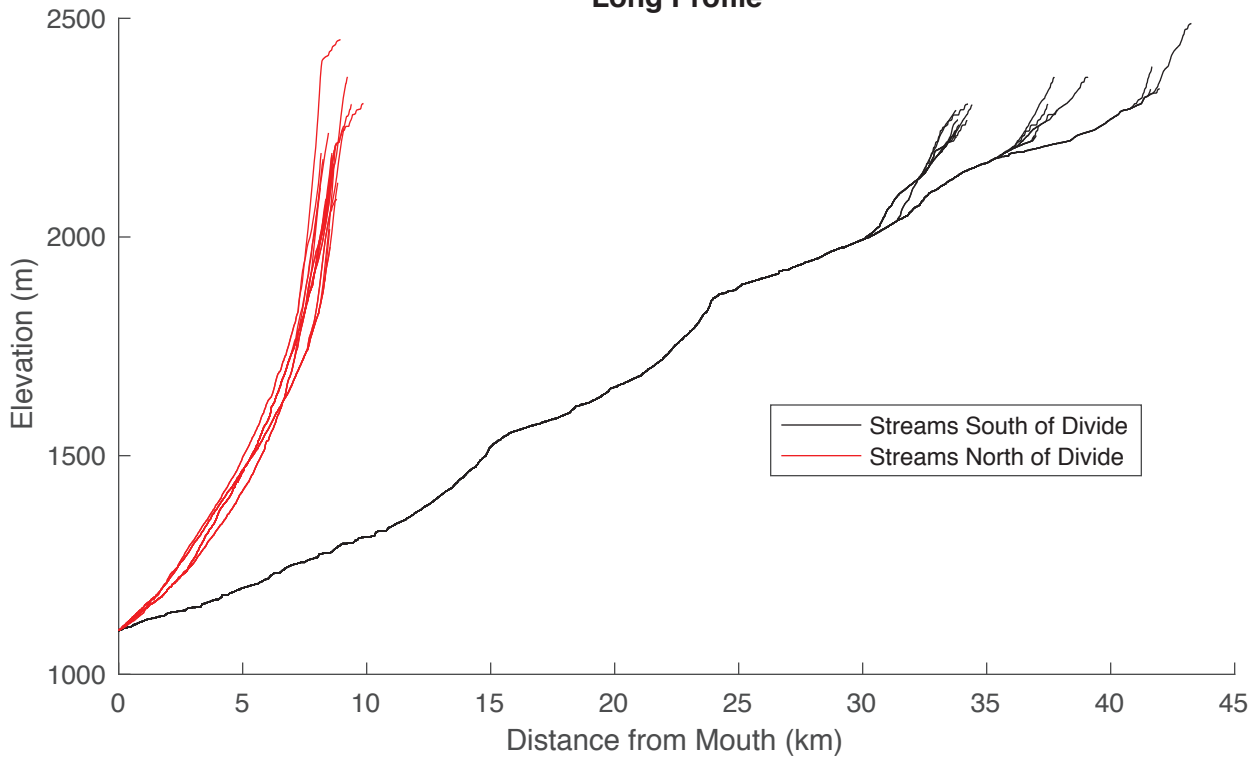
**Supplemental Figure 44**

### Divide SB9

Chi - Z

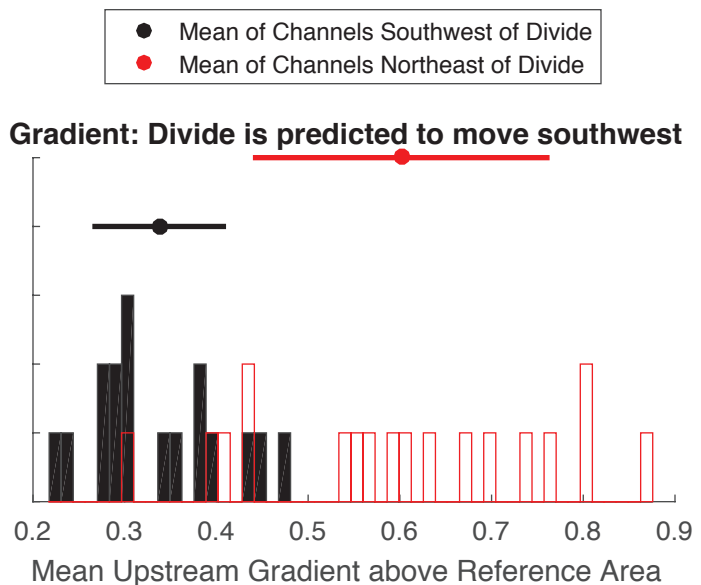
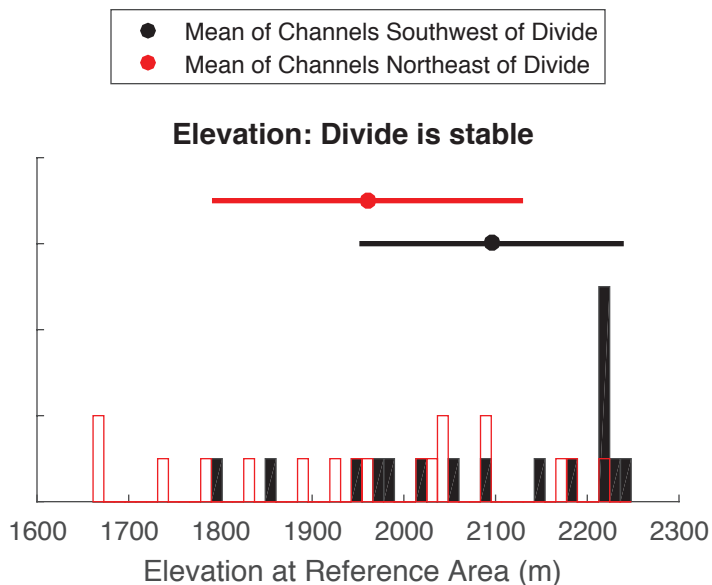
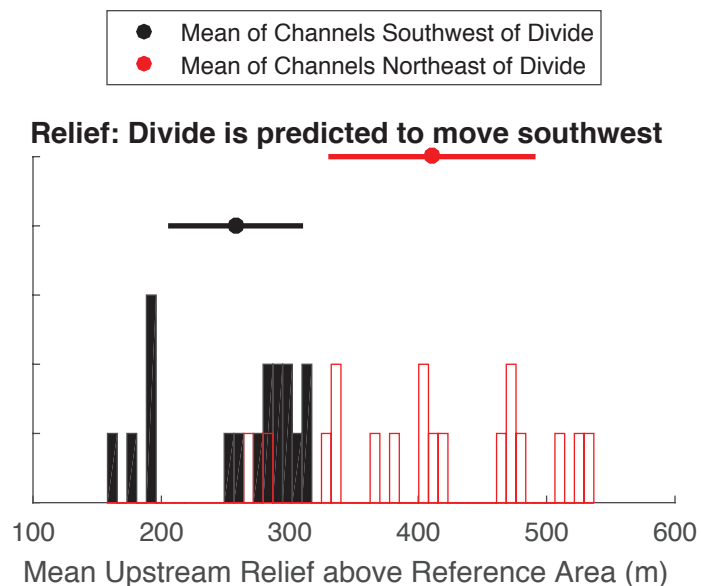
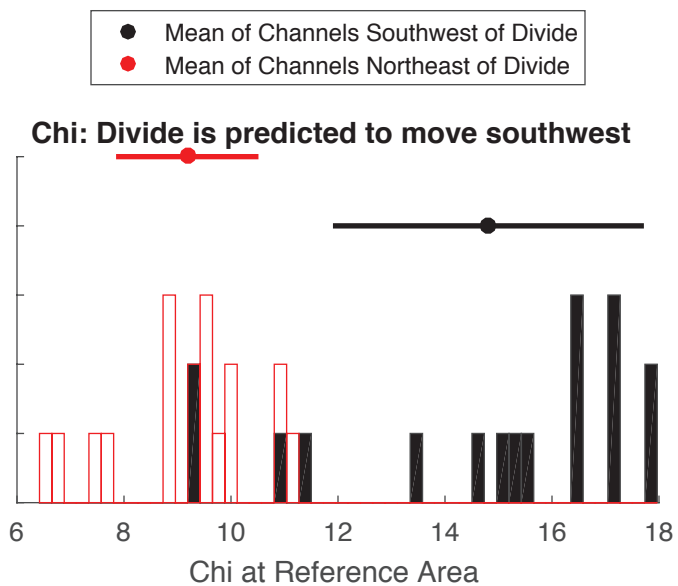


### Long Profile



Supplemental Figure 45

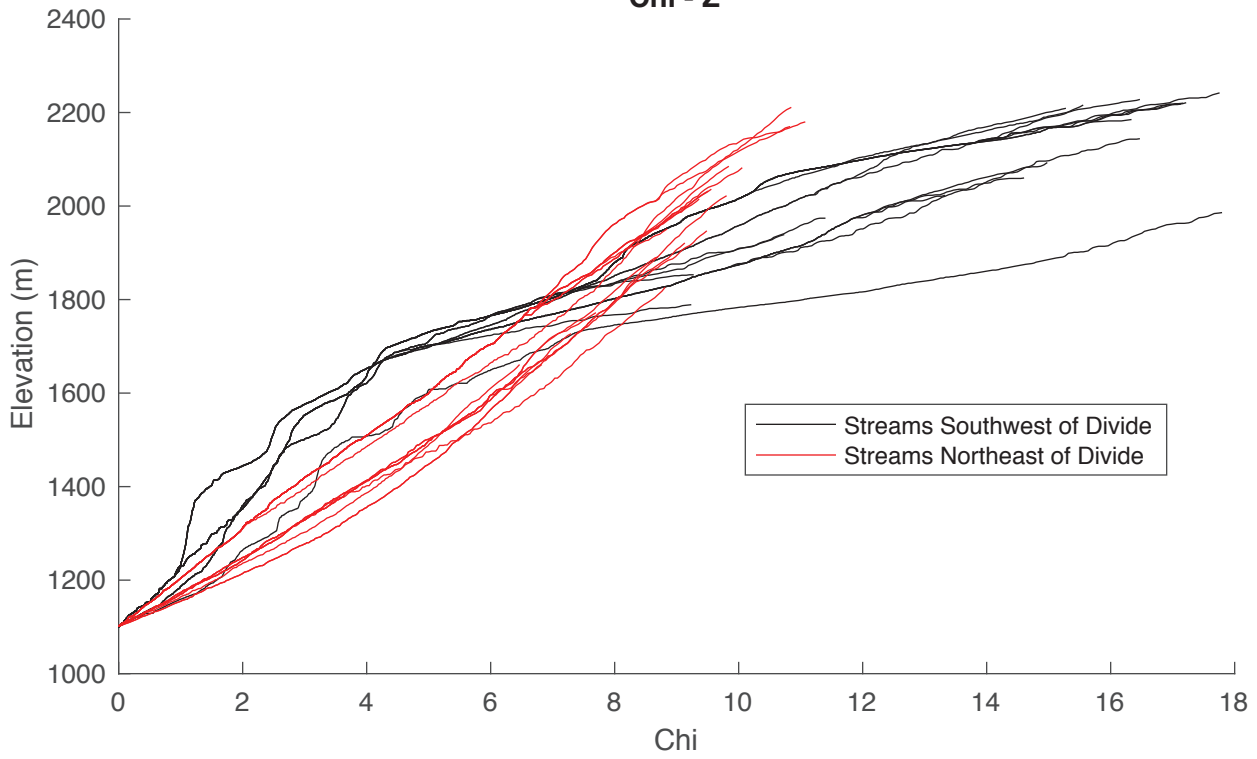
## Divide SB10



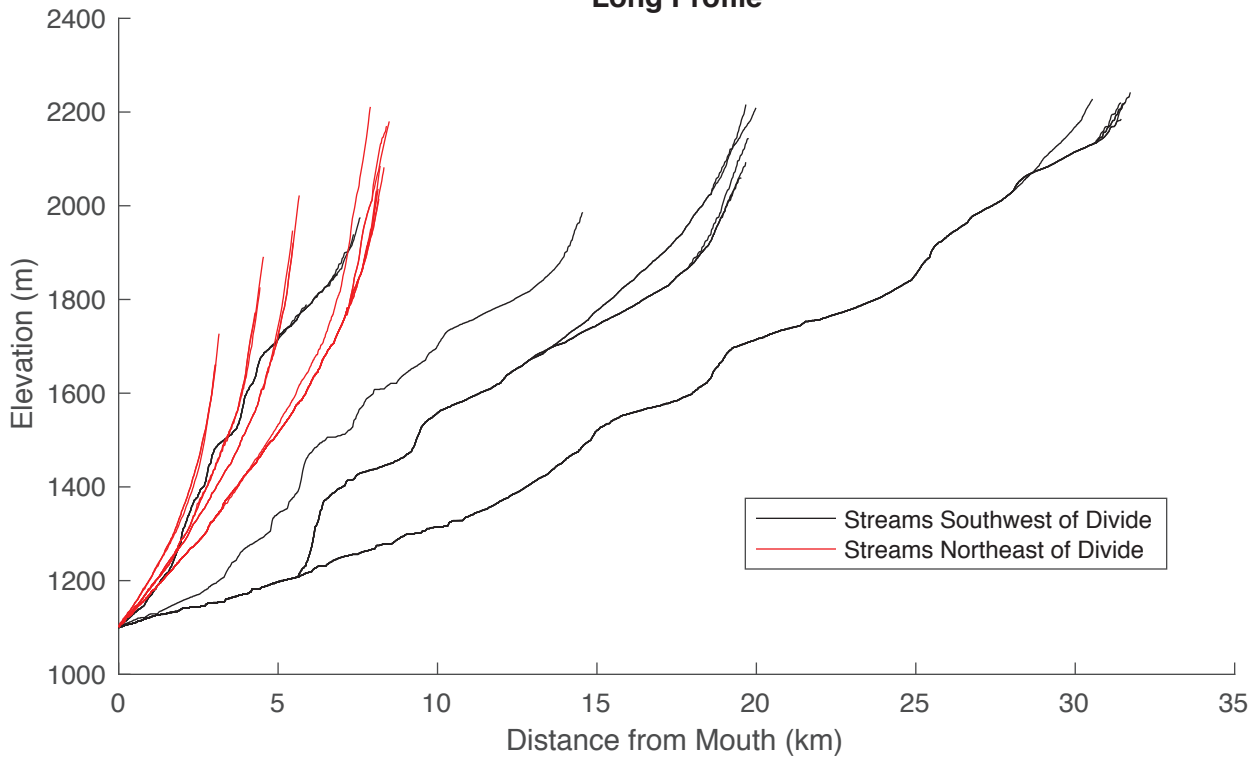
**Supplemental Figure 46**

# Divide SB10

## Chi - Z



## Long Profile



Supplemental Figure 47