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16 **Abstract**

17 Fluvially-carved bedrock valleys are ubiquitous landscape features. Vertical incision into
18 underlying bedrock generates valley relief, whereas lateral migration of the river channel widens
19 the valley floor as the river erodes the valley walls. The relative efficacy of these processes,
20 which can be modulated by precipitation and water discharge, sediment supply, lithology, and
21 uplift rates is recorded by the relationship between valley width, channel width, and valley relief
22 at a cross-section. Numerous studies have interpreted differences in bedrock valley morphology
23 to be evidence for variability in tectonics, climate, or geology, or as evidence of landscape
24 disequilibrium. However, expected downstream covariation of these parameters under steady-
25 state conditions has not been yet thoroughly explored. This contribution develops a 1-D profile
26 model to describe expected downstream coevolution of channel width, valley width, and valley
27 relief, described holistically using a valley confinement index. Model predictions agree well with
28 data from seven catchments in the King Range, CA spanning a gradient in tectonic uplift and
29 precipitation. Systematic variations in valley morphology correspond with differences in the
30 Erosion number between catchments, which summarizes the combined effect of multiple factors
31 that influence fluvial bedrock incision. These results suggest that longitudinal patterns in bedrock
32 valley morphology provide a promising, holistic metric to interpret the relative influence of
33 tectonics, climate, and lithology on bedrock valley formation in detachment-limited landscapes.

34

35 **Plain Language Summary**

36 Rivers continuously sculpt the bedrock valleys in which they reside via erosion. Valleys are
37 created through two main processes: vertical erosion into bedrock, which forms the valley relief,

38 and lateral migration of the river across the valley floor, widening the valley as it erodes the
39 valley walls. The relative influence of these processes results in a wide range of valley forms,
40 from narrow, plunging canyons to expansive valleys that are many times wider than the channels
41 that have carved them. Past research has linked differences in valley shapes to tectonics, climate,
42 or geology. However, to our knowledge, there is no expectation for how valley shape,
43 specifically the relationship between river channel width, valley width, and valley relief, changes
44 downstream within a river basin. We develop a simple model to describe downstream changes in
45 valley shape under a range of scenarios capturing differences in uplift, precipitation, and rock
46 strength. We find that model results align well with actual data from river basins with varying
47 tectonic uplift and precipitation. We propose that analysis of downstream patterns in valley shape
48 offers a comprehensive lens through which the combined influence of tectonics, climate, and
49 geology affects river erosion and the evolution of bedrock landscapes.

50 **1 Introduction**

51 Fluvially-carved bedrock valleys are ubiquitous landscape features across Earth's surface
52 (Montgomery, 2002). Valleys form and evolve through a combination of lateral and vertical
53 fluvial erosion processes. Vertical incision of the channel into the underlying bedrock generates
54 catchment or valley relief (Brocard & van der Beek, 2006; Stock et al., 2005; Willett, 2010),
55 whereas lateral migration of the river widens a valley as the river interacts with and erodes the
56 valley walls (e.g. Bufe et al., 2016, 2017; Langston & Tucker, 2018; Malatesta et al., 2017). The
57 relative efficacy of these processes, which can be modulated by precipitation and water
58 discharge, sediment supply, lithology, and tectonic uplift, is thus embedded in bedrock valley
59 morphology - specifically in the relationship between valley width, channel width, and valley
60 relief (e.g. Limaye and Lamb, 2014). Because these aspects of valley morphology are linked

61 through fluvial processes, they are not independent descriptions of landscape form. Rather,
62 channel width, valley width, and valley relief coevolve, reflecting the long-term partitioning of
63 vertical incision and lateral channel migration and underlying geomorphic process links.

64 The longitudinal elevation profile of the main stem river channel within a watershed is
65 commonly used to infer variability in tectonic uplift, rock erodibility, and climatic forcing
66 between watersheds, as well as to identify landscape disequilibrium within individual catchments
67 (Frankel et al., 2007; Kirby & Whipple, 2012; Snyder, 2000; Wobus et al., 2006). While the river
68 channel elevation profile is most often used, previous studies have also explored the control of
69 climate and tectonics on channel width, valley width, and valley relief. Rivers in more rapidly
70 uplifting landscapes tend to be relatively narrow for a given water discharge, this narrowing
71 helps to increase the boundary shear stress active on the channel bed, allowing erosion to keep
72 pace with uplift (e.g. Finnegan et al., 2005). More recently, variations in valley width between
73 valley cross-sections has also been shown to reflect differences in climatic forcing (Hancock &
74 Anderson, 2002; He et al., 2026; Mackin, 1937; Van Appledorn et al., 2019; Wegmann &
75 Pazzaglia, 2002), uplift (Bufe et al., 2016; Clubb et al., 2023), sediment supply (Brocard & van
76 der Beek, 2006; Finnegan & Balco, 2013; Langston & Robertson, 2023; Schanz et al., 2019;
77 Tofelde et al., 2022), and rock strength (Brocard & van der Beek, 2006; Bursztyn et al., 2015;
78 Keen-Zebert et al., 2017; Langston & Temme, 2019; Ortega-Becerril et al., 2018; Schanz &
79 Montgomery, 2016; Wohl, 2008; Zondervan et al., 2024). Recent work has further emphasized
80 that downstream changes in valley width reflect the cumulative balance between lateral erosion
81 and vertical incision (Clubb et al., 2022; Snyder et al., 2003; Tomkin et al., 2003; Turowski et
82 al., 2024). In this view, the rate of downstream valley widening has been interpreted has been
83 interpreted to reflect catchment-scale differences in tectonic uplift rates, sediment supply, and

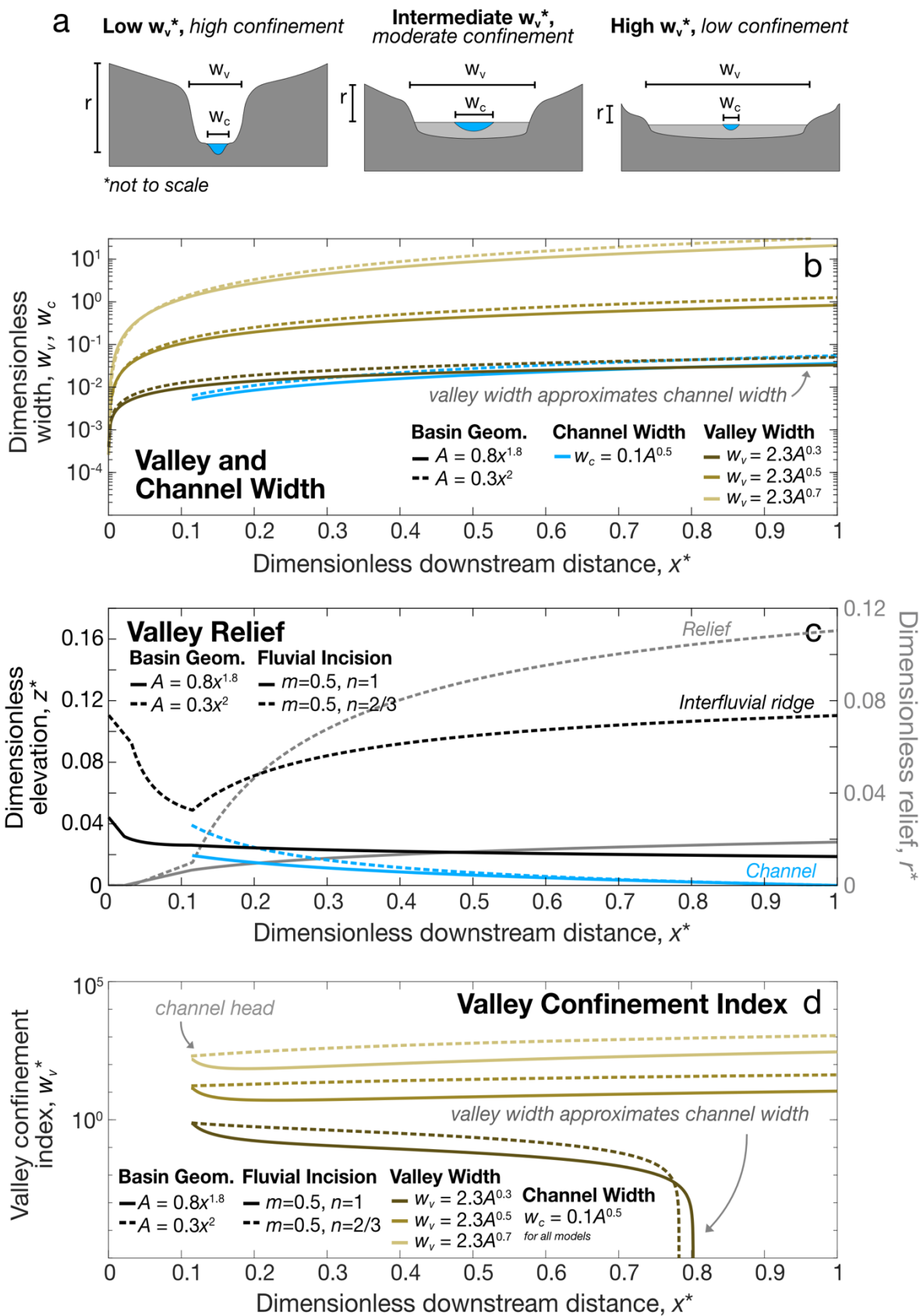
84 rock strength (Beeson et al., 2018; Clubb et al., 2023; May et al., 2013; Schanz & Montgomery,
85 2016). While these interpretations are largely empirical, a recent physics-based model for valley
86 widening hypothesized that the rate of downstream widening may be controlled by the ratio of
87 lateral sediment transport capacity relative to uplift (Turowski et al., 2024). Deviations from
88 these expected downstream patterns in valley width have also been used as evidence for ongoing
89 drainage basin reorganization and landscape disequilibrium (e.g. Harel et al., 2022). Valley relief
90 also generally increases in areas with increased uplift due to enhanced bedrock river incision
91 (Finnegan et al., 2008; Montgomery & Brandon, 2002). While each of these geomorphic features
92 has been cited individually as an expression of climate or tectonics at the landscape scale - the
93 covariation of channel width, valley width, and valley relief, and spatial structure of this
94 covariation remains less thoroughly explored.

95 Limaye and Lamb (2014) demonstrated that a process-based fluvial erosion model
96 incorporating both lateral and vertical erosion processes could produce a range of relationships
97 between bedrock valley width, channel width, and valley relief at individual cross-sections. In
98 their model, valley morphology emerges from the coevolution of channel lateral migration and
99 vertical incision, mediated by variations in bank erodibility driven by sediment cover, together
100 control the resulting valley form. This process-based perspective highlights that valley
101 morphology is not defined by any single metric but instead reflects the coupled adjustment of
102 channel and valley geometry through time. Limaye and Lamb (2014) quantified resultant
103 differences in bedrock valley morphology using a dimensionless bedrock valley confinement
104 index, w_v^* , defined as

105

$$106 \quad w_v^* = \frac{w_v - w_c}{r}, (1)$$

107 where the numerator, $w_v - w_c$ is relative valley width, where w_v is valley width, w_c is channel
108 width, and r is valley relief. Larger values of w_v^* are associated with low confinement, or wide
109 valleys with low relief and smaller values of w_v^* are associated with narrow valleys with high
110 relief, or highly confined valleys (Fig. 1A). Previous studies have used a similar metric, valley
111 aspect ratio, w_v/r , to quantify valley shape (e.g. Langston & Temme, 2019). However, valley
112 confinement index, w_v^* explicitly captures the fraction of the valley floor occupied by the active
113 channel. This is particularly informative in situations where relative valley width ($w_v - w_c$),
114 varies across a catchment or in scenarios where channel width and valley width do not change in
115 proportion with one another downstream (Fig. 1B).



117 **Figure 1.** Valley confinement index, w_v^* , and its components. A) Schematic illustration of
118 bedrock valley morphology for valleys with low, intermediate, and high w_v^* ; B) Dimensionless
119 channel width, w_c , (blue) and valley width, w_v (tan shades), w_c and w_v are normalized by their
120 respective maximum values at x^* for schematic purposes only. A range of downstream valley-
121 widening rates for two distinct Hack's Law scaling relationships (solid and dashed lines); C)
122 Channel (blue) and interfluvial ridge (black) elevation profiles and associated downstream relief
123 profiles (gray) for two distinct combinations of Hack's Law and the stream power incision model
124 (solid and dashed lines) and an erosion number, $N_e=0.3$. Similar to B) relief has been normalized
125 by x^* for schematic purposes only. D) Valley confinement index, w_v^* , profiles derived from
126 profiles in B and C.

127

128 Robust attribution of the relative influence of lateral and vertical erosion processes, and
129 their associated drivers, on catchment-scale differences in bedrock valley morphology requires
130 an underpinning steady state expectation, or null hypothesis, for each combination of driving
131 variables. This contribution develops a relatively simple expectation for downstream patterns in
132 bedrock valley confinement index by combining empirical channel width- and valley width-
133 drainage area scaling relationships with a 1-D profile model for interfluvial ridge height from
134 Willett (2010) derived from the stream power incision model (SPIM) (See Equation 10 in
135 Willett, 2010). Modeled profiles of valley confinement index with downstream distance are
136 compared to data from seven catchments in the King Range, CA spanning a gradient in tectonic
137 uplift and precipitation (Merritts & Bull, 1989; Snyder et al., 2003). This comparison
138 demonstrates that first-order differences in bedrock valley morphology, characterized by
139 downstream profiles of w_v^* , can predictably record differences in climatic and tectonic forcing.

140 These results provide a new 1-D modeling framework that can be applied to more definitively
141 attribute catchment-scale patterns in bedrock valley morphology to specific drivers such as
142 tectonics, climate, and erodibility, and for identifying disequilibrium conditions and ongoing
143 landscape adjustment within a river valley.

144 **2 Methods**

145 *2.1 Model Framework for Downstream Valley Confinement Profiles*

146 The 1D profile model for steady state bedrock valley confinement index, w_v^* , is
147 developed by combining empirical scaling relationships for channel width, w_c , (Leopold &
148 Maddock, 1953) and valley width, w_v , (Beeson et al., 2018; Clubb et al., 2022; Langston &
149 Tucker, 2018) with expected profiles in valley relief, r , (Willett, 2010) derived from the stream
150 power incision model (SPIM). As such, this formulation represents the baseline detachment-
151 limited assumption, providing a reference case that assumes limited effects of sediment cover on
152 catchment-scale patterns in valley morphology.

153 Channel width, w_c , robustly scales with increasing downstream discharge, Q , (e.g.
154 Ferguson, 1986; Gleason, 2015; Leopold & Maddock, 1953; Métivier et al., 2017; Phillips et al.,
155 2022), or with drainage area, A , assuming Q scales linearly with increasing A (e.g. Dunne and
156 Leopold, 1978; Whipple, 2004). Substantial empirical work demonstrates that downstream
157 width-discharge scaling is well-defined by a power-law with the form $w_c = k_c Q^b$, or

$$158 \quad w_c = k_c A^b, \quad (2)$$

159 where $b \sim 0.4-0.5$ for alluvial rivers (Fig 1B; Anderson et al., 2004; Leopold & Maddock,
160 1953; Moody & Troutman, 2002; Whitbread et al., 2015), though more recent studies show a
161 wider range of scaling relationships across individual watersheds (Allmendinger et al., 2005;
162 Anderson et al., 2004; Kostynick et al., 2026; Masteller et al., n.d.; Phillips et al., 2024). Similar

163 channel width-drainage area (w_c - A) scaling exponents have been measured for bedrock rivers,
 164 where $b \sim 0.3$ - 0.5 (Whipple, 2004; Wohl & David, 2008) and may vary with rock resistance
 165 (Montgomery & Gran, 2001; Wohl, 2004) and uplift rate, though the latter has not been shown
 166 consistently (Pazzaglia & Brandon, 2001; Tomkin et al., 2003).

167 Because bedrock valleys are formed by the rivers that incise them, it is not surprising that
 168 downstream changes in valley width can similarly be expressed as a function of discharge, or
 169 more commonly, drainage area, as $w_v = k_v Q^c$, or

$$170 \quad w_v = k_v A^c, \quad (3)$$

171 where, in most cases, $c \sim 0.2$ to 0.6 , bracketing the observed values for channel width-
 172 drainage area scaling exponents (Fig 1B; Beeson et al., 2018; Clubb et al., 2022; Langston &
 173 Temme, 2019; Snyder et al., 2003; Tomkin et al., 2003; see Turowski et al., 2024 for a
 174 summary). The valley width-drainage area (w_v - A) scaling exponent c has been shown to vary
 175 with rock resistance (e.g. Langston and Temme, 2019), sediment supply (e.g. Beeson et al.,
 176 2018), or uplift rate (e.g. Clubb et al., 2023). Landscape evolution models that incorporate lateral
 177 erosion have produced a similar range of downstream valley widening exponents to these
 178 empirical studies, with c ranging from 0.3 to 0.7 (Langston and Tucker, 2018).

179 While there is general overlap between channel and valley width-drainage area (w_c - A and
 180 w_v - A , respectively) scaling relationships, differences between downstream widening rates,
 181 primarily captured by differences in the scaling exponents, c and b , can result in continuously
 182 increasing or decreasing relative valley width (w_v - w_c) within a bedrock valley. Relative valley
 183 width will increase downstream if valley-widening outpaces channel-widening ($c > b$). Whereas,
 184 if downstream channel-widening outpaces valley-widening ($c < b$), relative valley width will
 185 decrease downstream as the river channel continues to occupy a larger and larger fraction of the

186 valley floor. To our knowledge, how downstream valley and channel widening covary – as
 187 reflected in their respective width-drainage area scaling relationships – and what controls these
 188 relative rates remain largely unexplored.

189 Generation of bedrock valley relief is driven by ongoing fluvial incision in actively
 190 uplifting orogens. In detachment-limited cases, the steady-state channel longitudinal profile is
 191 well-described using the stream-power incision model (SPIM), where bedrock incision can be
 192 calculated as

$$193 \quad E = KA^m S^n, \quad (4)$$

194 where K is rock erodibility, which can include climatic effects (e.g. Ferrier et al., 2013),
 195 A is catchment drainage area, S is local slope, and m and n are empirical scaling exponents
 196 (Howard, 1994; Whipple, 2004; Whipple & Tucker, 1999). Embedded within this formulation is
 197 the implicit assumption that channel width follows the downstream hydraulic geometry
 198 relationship described by Equation 2, such that width increases systematically with drainage
 199 area, typically with $b \approx 0.5$. Under steady state conditions, bedrock incision, E , can be directly
 200 related to tectonic uplift rate, U , such that bedrock incision rates increase with more rapid uplift
 201 (Duvall, 2004; Kirby & Whipple, 2012; Wobus et al., 2006). Willet (2010) modeled interfluvial
 202 ridge height for this steady state case using a combination of SPIM and Hack's Law, which
 203 accounts for the structure of the river channel network within a catchment with the form

$$204 \quad A = k_r x^h, \quad (5)$$

205 where x is downstream channel length from the divide, and k_r and h are empirical
 206 constants (Hack, 1957; Montgomery & Dietrich, 1992; Rigon et al., 1996). Due to the nonlinear
 207 nature of both SPIM and Hack's Law, modeled interfluvial ridge profiles can take a variety of
 208 forms, from concave ridge profiles mirroring the channel itself to convex ridge profiles with

209 enhanced increases in catchment relief downstream (Fig. 1C; See Equation 10 in Willet, 2010).
 210 To account for the controls of tectonic uplift (U), precipitation (P), and rock strength (K), Willet
 211 (2010) employs a dimensionless “Erosion number”, N_e , where

$$212 \quad N_e = \left(\frac{U}{x_d^{hm} K P^m k_r^m} \right)^{1/n} \quad (6)$$

213 and x_d is the total length of the river (Willett, 1999). N_e ultimately acts as a scaling factor for
 214 interfluvial ridge height, where higher N_e values increase interfluvial ridge heights relative to the
 215 channel elevation profile, enhancing relief (calculated as the elevation difference between the
 216 channel and the ridgeline) throughout a catchment. An advantage of N_e in a modeling context is
 217 that it summarizes the combined effect of multiple factors that influence fluvial bedrock incision
 218 rates, allowing for a general prediction of interfluvial ridge height, while also allows for
 219 individual drivers to be treated separately.

220 Downstream profiles of bedrock valley confinement index, w_v^* (Eq. 1), are calculated
 221 from channel width, valley width, and valley relief, where each is individually calculated as a
 222 function of drainage area, A . Bedrock valley confinement index is reprojected to downstream
 223 distance using Hack’s Law and nondimensionalized by total stream length to yield profiles in
 224 dimensionless distance, x^* (Fig. 1D). Because w_v^* combines multiple nonlinear relationships,
 225 downstream patterns are expected to be similarly nonlinear. Although potentially more complex
 226 than individual morphologic metrics, these elements are likely covary, as relief generation,
 227 valley widening, and channel widening are all governed by fluvial erosion processes (Limaye
 228 and Lamb, 2014). As such, downstream patterns in w_v^* provide a useful framework for evaluating
 229 the relative efficacy of lateral and vertical erosion towards a more holistic description of
 230 downstream evolution of valley morphology.

231 To explore the range of downstream patterns in bedrock valley confinement that may
232 arise in steady state, detachment-limited landscapes, this contribution generates w_v^* profiles for a
233 combination of valley width-drainage area scaling exponents, Hack's Law relationships, SPIM
234 exponents m and n , and Erosion number values. To reduce the number of free parameters in this
235 initial exploration, the model assumes a common drainage area scaling for channel width, $w_c =$
236 $0.1A^{0.5}$. Valley width is calculated as $w_v = 2.3A^c$ (Langston & Tucker, 2018). Although the
237 coefficients used in these relationships can vary widely (e.g. Turowksi et al., 2024), constant
238 values are adopted here to isolate first-order downstream trends. This contribution first considers
239 two end-member scenarios, (1) valley widening outpaces channel widening ($c=0.7$, $b=0.5$) and
240 (2) channel widening outpaces valley widening, ($c=0.3$, $b=0.5$) (Fig. 2). It is worth noting here
241 that even valley and channel widening exponents are equal ($c=0.5$, $b=0.5$), the larger coefficient
242 in the valley-width relationship still produces downstream increases in relative valley width,
243 yielding patterns comparable to the second end-member case (Fig. 1B, Supplementary
244 Information). Catchment relief profiles are then calculated using two combinations of Hack's
245 Law and SPIM exponents following Willet (2010): producing one concave ridge profile ($A =$
246 $0.8x^{1.8}$, $m=0.5$, and $n=1$) and one convex ridge profile ($A = 0.3x^2$, $m=0.5$, and $n=2/3$) (Fig.
247 1C). Finally, these components are combined to calculate w_v^* profiles across Erosion numbers
248 spanning five orders of magnitude, $N_e=10^{-5}$ to 3.

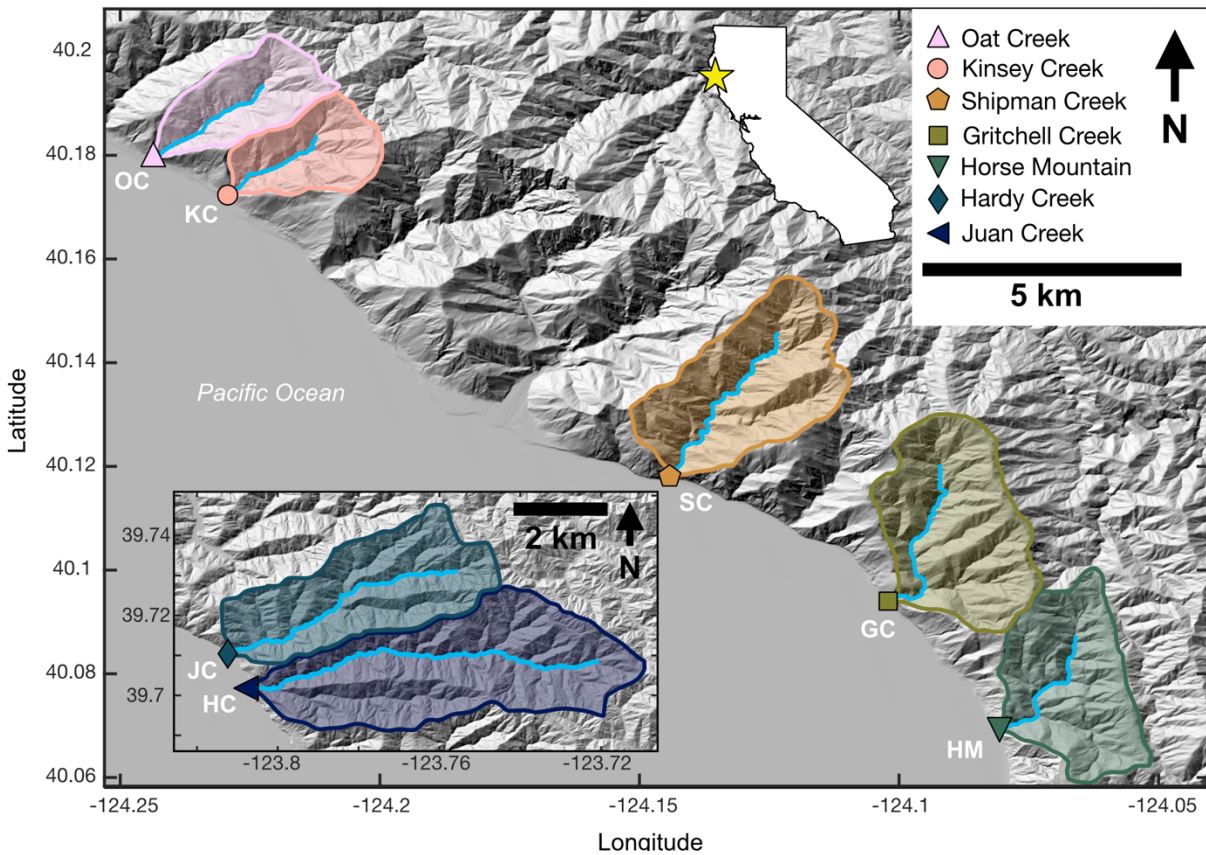
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250 *2.2 Constraining bedrock valley confinement for catchments in the King Range, CA*

251 To assess the performance of both the w_v^* metric and the modeling framework developed
252 here, modeled w_v^* profiles to downstream patterns in bedrock valley confinement for seven
253 catchments in the King Range in northern California (Fig. 2). Profiles in w_v^* for the King Range

254 catchments are derived using field-derived valley and channel width-drainage area scaling-
255 relationships (Snyder, 2000; Snyder et al., 2003; Supplementary Information) and bedrock valley
256 relief profiles extracted from Digital Elevation Model (DEM) topography. These catchments are
257 selected because prior work suggests that channel slopes are close to steady state and are well
258 described by detachment-limited incision when erosion thresholds and discharge variability are
259 taken into account (Snyder et al., 2000; 2003). Further, there is a wealth of existing precipitation,
260 uplift, geologic, and geomorphic data available for the region (e.g. Merritts & Bull, 1989;
261 Snyder, 2000; Snyder et al., 2003; Whipple, 2004).

262 Downstream w_v^* profiles are calculated using previously published valley and channel
263 width-drainage area scaling relationships from Snyder et al., (2000). Contributing drainage area
264 is extracted along the main stem channel of each river from 1-m resolution USGS 3DEP DEMs
265 (US Geological Survey) using Topotoolbox, a Matlab-based topographic analysis package
266 (Schwanghart & Scherler, 2014). Expected downstream valley and channel widths are then
267 computed from the reported drainage area scaling relationships. Catchment relief is calculated by
268 extracting catchment ridgelines and projecting ridge elevations onto the trunk channel using the
269 shortest Euclidean distance between the ridge and channel network. Relief at each point along
270 the long profile is defined as the difference between local maximum ridge elevation and trunk
271 channel elevation. To facilitate comparison across catchments, all profiles are normalized by
272 total mainstem length. Resulting downstream profiles in valley width, channel width, and relief
273 are then combined to calculate w_v^* as a function of dimensionless downstream distance, x^* .



274

275 **Figure 2. Map of King Range catchments used in this study with southern catchments,**
 276 **Juan Creek and Hardy Creek, displayed in the inset. Colors reflect a gradient in calculated**
 277 **Erosion number, where pinks indicate high N_e and blues indicate low N_e .**

278 The extensive prior work in the King Range, allows for the calculation of catchment-
 279 specific Erosion numbers using Equation 6 (Merritts & Bull, 1989; Merritts, 1996; Snyder et al.,
 280 2003, summarized in Supplementary Information). The study catchments span a tectonic uplift
 281 gradient from ~ 0.5 mm/year in the southern catchments to ~ 4 mm/year in the northern
 282 catchments. Orographic precipitation similarly increases northward, with mean annual
 283 precipitation ranging from ~ 1.5 to 3 m/yr. All catchments are underlain by similarly erodible
 284 Franciscan mélangé lithology. Catchment-specific erodibility values reported by Snyder et al.
 285 (2003), which range from $K = 3.7 \times 10^{-5}$ to $9.6 \times 10^{-4} \text{ m}^{0.14}/\text{yr}$, assuming $m = 0.43$ and $n = 1$.

286 Catchment-specific Hack's Law parameters are derived using TopoToolbox and used to
 287 calculate N_e (see Supplementary Information). Calculated Erosion numbers decrease
 288 systematically from north to south, ranging from $N_e=5.07 \times 10^{-2}$ for Oat Creek to a minimum
 289 $N_e=1.02 \times 10^{-4}$ for Hardy Creek (Table 1). Although precipitation and erodibility vary across the
 290 region, spatial variability in N_e primarily reflects the strong north-south gradient in tectonic
 291 uplift.

292 **Table 1.** Calculated Erosion Numbers, N_e , for King Range study catchments

Catchment Name (North to South)	Erosion Number, N_e
Oat Creek	5.07×10^{-2}
Kinsey Creek	5.00×10^{-2}
Shipman Creek	1.95×10^{-2}
Gritchell Creek	1.52×10^{-2}
Horse Mountain	6.31×10^{-3}
Hardy Creek	1.29×10^{-4}
Juan Creek	2.89×10^{-4}

293

294 *2.3 Metrics for model-data comparison of bedrock valley confinement*

295 For comparison with the King Range catchments, the downstream channel-width scaling
 296 exponent in the profile model is modified from $b = 0.5$ to $b = 0.35$ to better reflect the average
 297 value reported by Snyder et al. (2003). We then apply a range of valley width-drainage area
 298 scaling exponents, $c = 0.2-0.5$, again consistent with the range reported originally by Snyder et
 299 al., (2003). Hack's Law relationships and SPIM parameters are retained from the initial end-
 300 member models. Erosion numbers are varied from $N_e=10^{-6}$ to 10.

301 It is worth acknowledging that while the model could be calibrated individually for each
 302 catchment, the aim of this contribution is to evaluate the general behavior and utility of the
 303 simplified w_v^* framework rather than reproduce observations exactly. Catchment-specific

304 calibration would require simultaneous adjustment of numerous empirical scaling coefficients
305 and exponents, obscuring the relative influence of downstream valley widening, channel
306 widening, and relief generation on valley form. Because general parameter values are used for
307 aspects of the modeling (e.g. valley and channel width-drainage area scaling coefficients)
308 modeled w_v^* values are not expected to match observations directly. Rather, comparison presented
309 here focuses on whether downstream trends in bedrock valley confinement and the relationship
310 to N_e is reproduced consistently across both profile models and King Range data.

311 Comparison between modeled and field-derived w_v^* profiles focuses on two distinct
312 metrics: 1) a normalized median w_v^* and 2) a normalized residual confinement index (*NRCI*). For
313 both the King Range data and the suite of modeled profiles, w_v^* is normalized by the maximum
314 value in each dataset to facilitate comparison. Because N_e directly scales relief generation,
315 increasing N_e decreases median w_v^* and produces more confined valleys. For a given N_e , median
316 w_v^* is also influenced by differences between valley- and channel-widening rates, whether driven
317 by scaling exponents or coefficients. In both cases, larger disparities between valley and channel
318 widening increase median w_v^* , indicating reduced valley confinement. Importantly, the simplified
319 framework developed here treats relief generation, valley widening, and channel widening
320 independently, likely producing a broader range of confinement states than would occur if these
321 processes were tightly coupled. This approach intentionally explores the broadest possible
322 parameter space describing downstream patterns in valley confinement. As such, comparison of
323 these results with more mechanistic approaches (e.g. Limaye and Lamb, 2014; Turowski et al.,
324 2024) may help to provide further insight regarding the relative strength of the coupling between
325 these processes under differing climatic, tectonic, or geologic forcing.

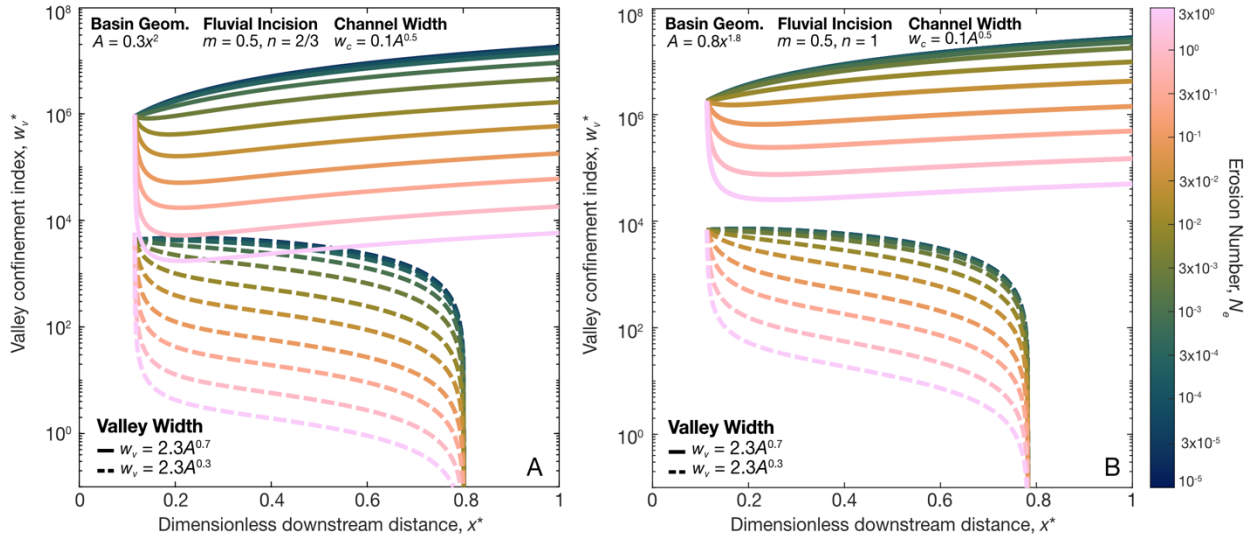
326 The normalized residual confinement index (*NRCI*) metric is introduced to characterize
327 overall differences in the shape of w_v^* profiles. This metric is adapted from the normalized
328 concavity index (*NCI*), developed for river longitudinal profiles (described in Chen et al., 2019).
329 Following this approach, a straight line to each log-transformed w_v^* profile from $x^*=0.1$ and 0.9
330 to minimize edge effects related to channel initiation in the headwaters or as the channels exit the
331 King Range and reach the coastline. The integrated difference between the log-transformed w_v^*
332 profile and straight line fit is then calculated, representing the residual valley confinement for the
333 whole of the w_v^* profile (Fig. 5B insets, Supplementary Information). Values are then normalized
334 by the maximum value within the modeled and field-derived datasets prior to comparison to
335 produce a final *NRCI* metric. Negative *NRCI* values indicate concave-up w_v^* profiles, whereas
336 positive *NRCI* values indicate convex profiles. High relief catchments associated with efficient
337 fluvial incision and high N_e , are expected to exhibit concave w_v^* profiles (e.g. Fig. 1D) and
338 negative *NRCI* values. In contrast, low relief catchments with wide valleys, associated with more
339 efficient lateral erosion and low N_e , should exhibit convex w_v^* profiles (e.g. Fig. 1D) and positive
340 *NRCI* values.

341 **3 Results**

342 *3.1 Modeled downstream trends in bedrock valley confinement*

343 Modeled w_v^* profiles are primarily controlled by the balance between valley widening
344 and channel widening, captured by the relative magnitudes of the w_v -*A* and w_c -*A* scaling
345 exponents (*c* and *b*), and by the Erosion number, N_e (Fig. 3). While variations in Hack's Law
346 and SPIM parameters produce slight differences in w_v^* values and profile shapes, these effects
347 are secondary (Fig. 3). Both end-member cases, show an initial downstream decrease in w_v^* ,

348 reflecting increasing valley confinement near channel heads when N_e exceeds $\sim 10^{-3}$. This
 349 upstream behavior arises because fluvial incision rapidly generates relief as channels initiate
 350 while drainage area, and by extension, channel width and valley width, remain low. As a result,
 351 upstream patterns in w_v^* are predominately set by the shape of the ridge profile (e.g. Fig. 1C).



352 **Figure 3. A) Downstream profiles of valley confinement index, w_v^* for a case where**
 353 **downstream valley widening outpaces downstream channel widening (solid line) and where**
 354 **downstream channel widening outpaced valley widening (dashed line) for a range of N_e**
 355 **values (line color). B) Same as A, but with Hack's Law scaling, $A = 0.8L^{1.8}$, and stream**
 356 **power incision parameters, $m=0.5$ and $n=1$. Colors denoting modeled Erosion Numbers, N_e ,**
 357 **use the *batlow* color ramp (Crameri, 2023).**
 358

359

360 For the first end member case, where downstream channel widening outpaces valley
 361 widening ($c=0.3$, $b=0.5$) generally decreases downstream as the river progressively occupies a
 362 larger and larger fraction of the valley floor. This behavior is consistent with landscapes where

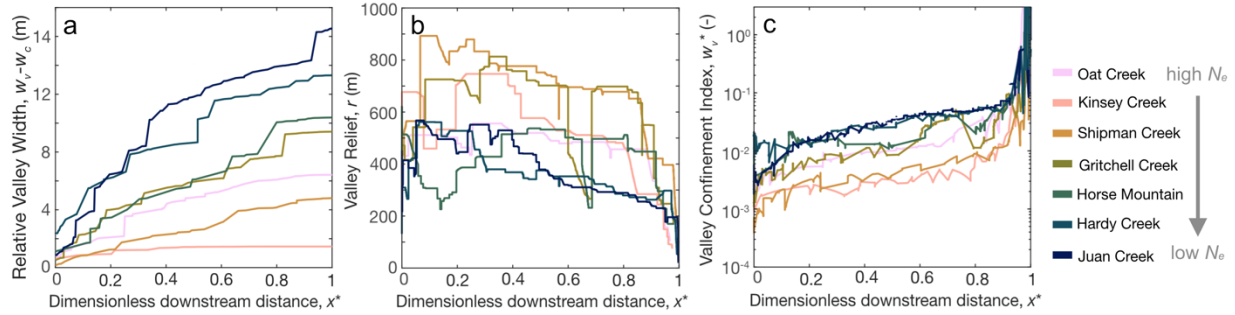
363 relatively efficient vertical incision dominates over lateral migration of the channel and beveling
364 of the valley floor, limiting downstream valley widening despite increasing discharge. In the
365 downstream limit, the channel eventually occupies the full valley floor, producing fully confined
366 conditions where $w_v^* = 0$ (Fig. 3). In contrast, for the second end-member case, where valley
367 widening outpaces channel widening ($c=0.7$, $b=0.5$) the channel progressively takes up a smaller
368 and smaller fraction of the valley floor (Figure 2). As such, w_v^* generally increases downstream
369 as the relative valley width increases. This end member is consistent with landscapes where
370 lateral erosion is efficient relative to fluvial incision, such as in instances where abundant valley-
371 bottom sediment cover acts to shield the channel floor from erosion and promote lateral channel
372 migration (DeLisle & Yanites, 2023; Turowski, 2018). Under these conditions, downstream
373 patterns in w_v^* predominately governed by relative valley width rather than relief generation and
374 ridgeline topography.

375 Across both end-member cases, larger Erosion numbers, N_e , result in lower w_v^* and
376 therefore more highly confined valleys (Fig. 3). This is because higher N_e reflects conditions that
377 should produce more efficient vertical incision as predicted by SPIM, such as rapid tectonic
378 uplift, weak bedrock, or increased precipitation. This enhanced erosional efficiency,
379 summarized by high N_e , results in greater catchment relief and reduced w_v^* . High- N_e landscapes
380 also tend to produce concave-up w_v^* profiles. Lower N_e values, reflective of landscapes with
381 slower uplift, more resistant rocks, or more arid climates, result in high w_v^* , reflecting less
382 confined valleys. Under these conditions, lateral erosion and valley widening overwhelm the
383 topographic signature of relief such that w_v^* profiles become increasingly convex downstream,
384 mirroring downstream increases in relative valley width (Fig. 1B). Taken together, these results
385 suggest that under detachment-limited conditions, catchment-scale differences in w_v^* reflect

386 differences in the relative efficacy of vertical incision and lateral erosion as a function of variable
387 uplift, erodibility, and precipitation, as summarized by the Erosion number, N_e .

388 *3.2 Downstream valley confinement across King Range catchments*

389 Across the seven King Range catchments, relative valley width systematically increases
390 with decreasing Erosion Number, N_e (Figure 4A, $R^2=0.69$, Supplementary Information). This
391 trend reflects downstream differences between valley- and channel-widening rates, expressed
392 through the relative magnitudes of the w_v - A and w_c - A scaling exponents ($c - b$). The difference
393 between these widening exponents is itself inversely proportional to N_e ($R^2=0.41$, see
394 supplement). Six of the seven catchments exhibit positive $c - b$ values, indicating that valley
395 widening outpaces channel widening downstream. The maximum difference ($c - b = 0.13$)
396 occurs in Hardy Creek, the catchment with the lowest Erosion number ($N_e = 1.02 \times 10^{-4}$). In
397 contrast, Kinsey Creek is the only catchment with a negative exponent difference ($c - b = -0.06$).
398 Kinsey creek also has the second largest Erosion number ($N_e = 0.05$). These differences strongly
399 influence downstream patterns in relative valley width. For example, Juan Creek ($N_e =$
400 2.89×10^{-4} , $c - b = 0.13$) exhibits nearly a 20-fold increase in relative valley width from
401 headwaters to outlet, whereas relative valley width in Kinsey Creek increases by only a factor of
402 two over the same downstream distance. Despite its negative $c - b$ value, valley-widening still
403 outpaces channel-widening in the Kinsey Creek catchment due to differences in the coefficients
404 of the previously reported scaling relationships (Snyder et al., 2003; Figure 4A, Supplementary
405 Information).



406
 407 **Figure 4. A) Downstream profiles of relative valley width calculated from w_v-A and w_c-A**
 408 **scaling reported in Snyder et al., (2003); B) Downstream profiles of catchment relief; C)**
 409 **Downstream profiles in valley confinement index, w_v^* derived from profiles in A and B.**

410

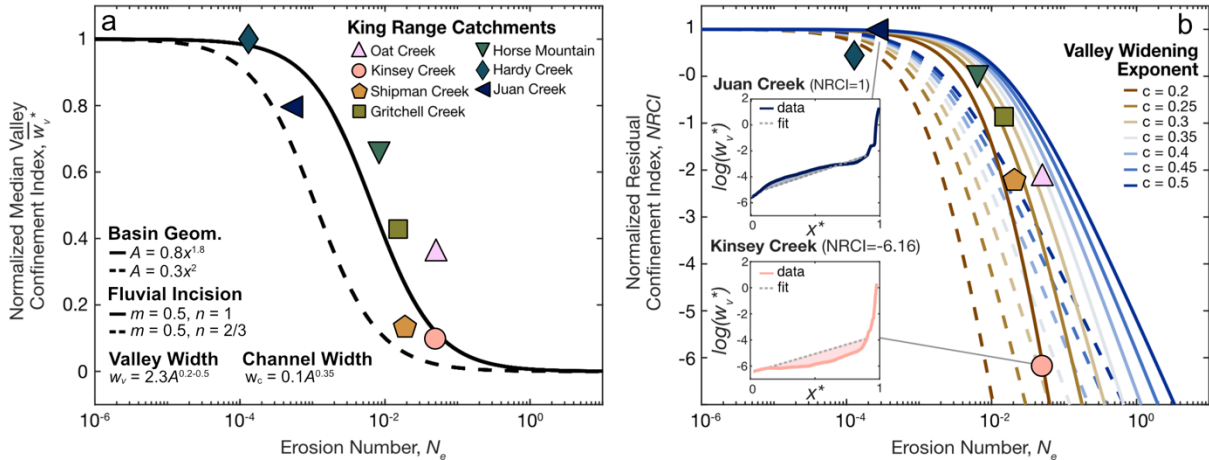
411 Catchment relief across the King range catchments similarly varies as a function of N_e
 412 (Fig. 4B), such that catchments with larger N_e generally also exhibit higher catchment relief after
 413 accounting for differences in the length of the main stem channel in each watershed ($R^2=0.60$,
 414 Supplementary Information). Kinsey Creek, the catchment with the smallest relative valley width
 415 is also the catchment with the maximum relative catchment relief, where $r = 746.5\text{m}$ ($x_d = 3.3$
 416 km, $N_e = 0.05$). Juan Creek, the catchment with the largest relative valley width, also exhibits the
 417 lowest relative catchment relief, $r = 567.3\text{ m}$ ($x_d = 11\text{ km}$, $N_e = 2.89 \times 10^{-4}$).

418 These observed variations in relative valley width (Fig. 4A) and valley relief (Fig. 4B)
 419 combine to produce systematic differences in w_v^* profiles across the study catchments (Fig. 4C).
 420 All seven catchments are characterized by relatively narrow valleys and high relief, resulting in
 421 generally low confinement index values ($w_v^* = 10^{-3} - 10^{-1}$). Despite the relative narrow range of w_v^* ,
 422 the resulting downstream profiles span a continuum from concave to convex (Fig. 4C) consistent
 423 with the modeled behaviors described in Section 3.1 (Fig. 3). In all catchments, w_v^* increases
 424 downstream, consistent with the end-member case in which valley widening outpaces channel
 425 widening. This outcome is informed by differences in both the empirically-derived coefficients

426 and exponents of the channel width and valley width-drainage area scaling relationships reported
427 by Snyder et al., (2003).

428 *3.3 Model-data comparison of bedrock valley confinement profiles*

429 Normalized median w_v^* values for the King Range decrease as a function of increasing
430 N_e ($R^2=0.59$, Fig. 5A), indicating that valleys become progressively more confined with
431 increasing N_e . This decrease is well-captured by the modeled bedrock valley confinement
432 profiles for the generalized Hack's Law scaling relationship $A=0.8L^{1.8}$ and SPIM exponents
433 $m=0.5$, $n=1$, with a mean absolute error of MAE=0.13 between modeled and field-derived
434 profiles. This parameterization is also broadly consistent with the best-fit SPIM exponents for the
435 seven King Range catchments ($m=0.43$ and $n=1$; Snyder et al., 2003). The alternative
436 parameterization ($A = 0.3L^2$, $m = 0.5$, $n = 2/3$) systematically underpredicts normalized
437 median w_v^* , although the overall downstream trend with increasing N_e remains similar between
438 the two model configurations. For all tested valley-width scaling exponents ($c = 0.2-0.5$),
439 normalized median w_v^* profiles collapse (Fig. 5A). This indicates that while the choice of
440 exponent does affect the catchment-average valley confinement, it does not influence the relative
441 dependence of median w_v^* on N_e for a given parameter set after normalization. This is not
442 necessarily surprising, since w_c-A scaling and w_v-A scaling are independent of N_e .



443
 444 **Figure 5. A) Normalized median w_v^* , with varying N_e using two combinations of**
 445 **generalized basin geometry and stream power incision parameters (dashed and solid lines),**
 446 **w_v - A scaling exponents, $c=0.2-0.5$ (curves collapse), and channel width-drainage area**
 447 **scaling exponent, $b=0.35$, the average value for the King Range catchments. King Range**
 448 **values plotted as diamonds. B) Normalized residual confinement index, $NRCI$ with varying**
 449 **N_e for the same scaling relationships used in A. King Range values plotted as diamonds.**
 450 **Insets show two w_v^* profiles for two King Range catchments to illustrate differences in**
 451 **concavity informing their respective $NRCI$ values (all catchment profiles included in**
 452 **Supplementary Information).**

453

454 Normalized residual confinement index ($NRCI$) for the King Range catchments also
 455 decreases systematically as a function of increasing Erosion number as valley confinement
 456 profiles transition from convex to concave ($R^2 = 0.73$, Fig. 3, 4C, 5B). Four catchments exhibit
 457 negative $NRCI$ values, indicative of concave confinement w_v^* profiles. Kinsey Creek displays
 458 the minimum $NRCI$ value ($NRCI = -6.16$; Fig. 5B inset), consistent with its high N_e , narrow
 459 valleys, and large normalized relief described in Section 3.2. Taken together, these results
 460 indicate that the valley morphology of Kinsey Creek is likely governed by vertical incision

461 processes, such that the w_v^* profile is strongly influenced by downstream patterns in catchment
462 relief, and that these downstream patterns in relief are consistent with the formulation for
463 interfluvial ridge height from Willet (2010). In contrast, three of the King Range catchments
464 have positive *NRCI* values, indicative of convex w_v^* profiles. Juan Creek has the maximum
465 convexity and resultant $NRCI = 1$, consistent with low N_e and observations of relatively wide
466 valleys and low relative relief, as described in Section 3.2. Given this, we suggest that the valley
467 morphology of Juan Creek is dominated by lateral erosion processes, at least compared to the
468 other King Range catchments evaluated here, and that the downstream patterns in w_v^* are
469 primarily set by relative valley width rather than ridgeline topography. These contrasting end
470 members suggest that the shape of w_v^* profiles across King Range catchments reflects a gradient
471 in bedrock valley form more strongly dominated by incision-driven relief growth to those more
472 strongly influenced by downstream valley widening.

473 The modeled profiles reproduce these observed *NRCI* trends across the King Range well,
474 particularly for the parameterization $A = 0.8L^{1.8}$, $m = 0.5$, and $n = 1$ (Fig. 5B). However, unlike
475 normalized median w_v^* , the dependence of *NRCI* on Erosion number is distinct for each valley
476 width-drainage area scaling exponent, c . For a given N_e , relatively slow downstream valley
477 widening rates, reflected by low values of c , produce increasingly negative *NRCI* and more
478 concave profiles, reflecting relatively inefficient lateral erosion compared to vertical incision.
479 Larger c values result in greater *NRCI* and more convex w_v^* profiles as downstream valley-
480 widening overwhelms relief generation via incision. Overall, $c=0.3$ provides the closest match to
481 the relative *NRCI* values observed across the King Range catchments (MAE = 0.98). However,
482 the absolute errors between each catchment and modeled w_v^* profiles are minimized for different
483 values of c (Supplementary Information). For example, the shape of the Kinsey Creek profile as

484 described by *NRCI*, is best described by the minimum model c value, $c=0.2$, and a difference in
485 valley and channel-drainage area scaling exponents of $c-b = -0.15$, consistent with the observed
486 $c-b = -0.06$. In contrast, Juan Creek, where reported $c-b=0.13$, is best described by the maximum
487 modeled c value, $c=0.5$, yielding a best-fit modeled $c-b=0.15$. Oat Creek, a catchment where
488 reported $c-b=0.06$, is best described by modeled profiles with $c=0.35$ ($c-b=0$). It is worth noting
489 here that exact agreement between observed and best-fit modeled c values is not expected
490 because the modeled profiles assume constant valley and channel width-drainage area scaling
491 coefficients. This assumption is not consistent with the observed scaling relationships reported
492 by Snyder et al., (2003) where both the coefficients and exponents of the derived drainage area
493 scaling relationships vary between catchments. Although additional calibration could likely
494 improve catchment-specific fits, the simplified framework captures the primary downstream
495 trends in bedrock valley morphology observed across the King Range catchments.

496 **5 Discussion and conclusions**

497 This contribution represents the first description of a steady-state expectation for
498 downstream patterns in bedrock valley morphology, as described by the downstream covariation
499 in valley width, channel width, and catchment relief, or the bedrock valley confinement index,
500 w_v^* . The systematic differences observed in both the magnitude and shape of w_v^* profiles
501 ultimately reflect the relative efficacy of fluvial incision to lateral erosion. Within the framework
502 developed here, relief generation is governed entirely by detachment-limited bedrock incision,
503 which embeds the influence of climate, tectonics, and lithology through the application of the
504 SPIM (Willet, 2000). These combined influences are encapsulated in the Erosion number, N_e .
505 Comparison with the King Range catchments demonstrates that the simplified framework
506 developed here reproduces first-order variations in observed valley morphology across a gradient

507 in uplift and precipitation, suggesting that w_v^* profiles capture meaningful differences in valley
508 form linked to relative differences in incisional efficiency as described by SPIM and Ne.

509 In the model-data comparison presented here, modeled interfluvial ridge heights assume
510 that the only process that influences patterns in catchment relief is fluvial incision. For
511 detachment-limited landscapes, this assumption is generally appropriate (Whipple & Tucker,
512 1999; Willett, 2010) and here, is reinforced by the good agreement observed between modeled
513 w_v^* profiles and w_v^* for the King Range catchments (Fig. 3,4). However, in steep landscapes,
514 active landsliding may also act to limit catchment relief (e.g. Larsen & Montgomery, 2012),
515 modifying the relationship between fluvial incision and valley relief with knock-on
516 consequences for w_v^* . Further, this enhanced hillslope sediment supply may result in valley floor
517 aggradation, further modifying the covariation between valley width and catchment relief as the
518 efficacy of fluvial incision is reduced and the valley becomes transport-limited (Clubb et al.,
519 2023; Shobe et al., 2018; Tofelde et al., 2022). However, even in transport-limited settings,
520 channel width, active valley width, and valley relief should still be expected to covary because
521 they remain dynamically linked through sediment transport and channel adjustment processes.
522 Indeed, previous studies have shown strong correlations between terrace height and valley width
523 across multiple cycles of aggradation and incision (Tofelde et al., 2022). Nevertheless, further
524 exploration is needed to assess the suitability of the framework developed here to these cases,
525 specifically the treatment of valley relief. At present, Willet (2010) provides the only general
526 formulation for expected downstream patterns in catchment relief.

527 It is worth emphasizing that the influence of climate, tectonics, and lithology on relative
528 valley width is not explicitly accounted for in the valley and channel width-drainage area scaling
529 relationships used here. While empirical evidence suggests that both valley and channel width-

530 drainage area scaling may be impacted by differences in climate, tectonics, and rock strength
531 (Bufe et al., 2016; Clubb et al., 2023; Kostynick, Phillips, et al., 2026; Langston & Temme,
532 2019; Masteller et al., n.d.; Phillips et al., 2024; Tofelde et al., 2022), there is not yet a general
533 framework capable of attributing differences in downstream scaling relationships to specific
534 controls. This is likely because there remain remarkably few catchment-specific studies that
535 constrain these scaling relationships, limiting cross-catchment comparisons. Perhaps more
536 fundamentally, Snyder et al. remains one of the few studies to empirically constrain both
537 channel-width and valley-width scaling relationships within the same set of catchments. We
538 suggest that exploration of the variation of catchment specific scaling relationships for both
539 valley and channel width, with Erosion number, N_e , may be a fruitful step forward towards
540 explicitly accounting for these effects. Further, the application of process-based erosion models
541 that combine vertical incision and lateral migration (e.g. Limaye and Lamb, 2014; Langston and
542 Tucker, 2018; Turowski et al, 2024) may help identify the mechanisms driving systematic
543 variations in downstream valley morphology. Ultimately, integrating detailed catchment-scale
544 observations with transport-based frameworks for channel adjustment offers a path toward more
545 physically grounded predictions of how river valleys respond to changing climate, uplift,
546 sediment supply, and other perturbations.

547 The newly-developed modeling framework presented captures differences in bedrock
548 valley morphology consistent with previous observations at individual valley cross-sections.
549 Modeled decreases in overall valley confinement with increasing Erosion number are consistent
550 with observations that valleys are typically relatively narrow in rapidly uplifting landscapes
551 (Bufe et al., 2016, 2017) or in settings with less erodible bedrock (Langston & Temme, 2019;
552 Schanz & Montgomery, 2016). These results are also consistent with process-based modeling

553 efforts (Limaye & Lamb, 2014). Beyond qualitative agreement with previous studies, the
554 framework developed here captures systematic differences in valley morphology between King
555 Range catchments spanning gradients in uplift and precipitation (Fig. 5, Snyder et al., 2003).

556 This contribution expands upon previous work by moving beyond a single cross section
557 and instead considering the spatial covariation of valley width, channel width, and valley relief
558 through a watershed from headwaters to outlet. The shape of confinement profiles varies
559 systematically with Erosion number, providing a quantitative framework for distinguishing
560 incision-dominated landscapes (concave profiles; negative *NRCI*) from those more strongly
561 influenced by lateral valley widening (convex profiles; positive *NRCI*). It is worth noting that the
562 model parameter space explored in this contribution is intentionally broad, aiming to capture the
563 widest range of possible downstream trajectories in bedrock valley confinement. Further, the
564 treatments of valley width, channel width, and catchment relief are fully independent within the
565 preliminary exploration presented here. However, the tendency for the King Range catchments to
566 cluster around a relatively limited range of modeled trajectories suggests that downstream valley
567 morphology exhibits some degree of self-organization across catchments. This behavior is
568 perhaps not surprising, as both valley width and catchment relief are generated through fluvial
569 erosion, whereas the efficacy of fluvial erosion itself is fundamentally governed by water
570 discharge, which is in turn embedded within downstream patterns in channel width. In this
571 sense, downstream patterns in valley confinement reflect how channel width, valley width, and
572 valley relief coevolve as a function of both catchment hydrology and erosion processes.

573 Beyond these catchment scale patterns, localized shifts in the relationship between
574 elements of valley confinement index (relative valley width, relief) across a catchment may
575 signal coincident shifts in the relative efficacy of vertical and lateral erosion processes that act to

576 set valley architecture locally and to identify downstream transitions with dominant valley-
577 forming processes. This behavior is perhaps most clearly illustrated in Figure 3 for profiles with
578 $c = 0.7$ and high N_e . In these cases, modeled confinement initially decreases in steep headwaters
579 where the onset of fluvial incision rapidly generates catchment relief. Farther downstream,
580 confinement reaches a minimum value and then begins to increase as downstream increases in
581 valley width exceed downstream increases in relief. This transition suggests that lateral erosion
582 and valley widening progressively exert a greater influence on valley morphology downstream.
583 Along-river shifts in confinement trajectories may also reflect downstream changes in climatic,
584 tectonic, or lithologic forcing. Gradual downstream changes may arise from orographic
585 precipitation gradients (Forte & Rossi, 2024), whereas more punctuated transitions may reflect
586 rivers traversing zones of enhanced uplift (Bufe et al., 2016) or more resistant bedrock (Langston
587 & Temme, 2019).

588 Along-river variations in the threshold for sediment motion may provide an additional
589 mechanism governing downstream transitions in bedrock valley confinement because sediment
590 transport ultimately sets the relative efficacy of fluvial erosion. Sediment entrainment thresholds
591 vary systematically with channel slope (Lamb et al., 2008; Prancevic & Lamb, 2015), the timing
592 and magnitude of floods (Kostynick, et al., 2026; Masteller et al., 2019, 2025; Ockelford et al.,
593 2019), riverbed structure and grain size distribution (Hodge et al., 2013; Kirchner et al., 1990;
594 Masteller & Finnegan, 2017; Whitfield et al., 2025), and sediment supply (Dietrich et al., 1989;
595 Hassan et al., 2020; Johnson, 2016). Because bedrock incision, and lateral channel migration are
596 fundamentally governed by sediment transport, downstream variations in sediment mobility (
597 Masteller et al., n.d.; Mueller et al., 2005) may systematically alter the relative efficacy of
598 vertical incision and lateral erosion across a catchment. For example, enhanced thresholds for

599 motion in steep headwaters may suppress lateral channel mobility while maintaining efficient
600 vertical incision through impact-driven erosion. In contrast, downstream reductions in
601 entrainment thresholds may promote more persistent sediment mobility, enhancing lateral
602 channel migration and shifting valley evolution toward widening-dominated behavior. As such,
603 downstream confinement patterns may also reflect spatial variations in the efficiency with which
604 floods mobilize sediment and erode bedrock. Future work should focus on more explicitly
605 linking the relative efficacy of vertical and lateral erosion processes within a catchment and
606 downstream transitions in valley-forming processes under steady state conditions.

607 Recent advances in data availability and analysis tools position the community to
608 evaluate these relationships more systematically. The rapid expansion of publicly available,
609 high-resolution topographic datasets, combined with emerging tools for automated extraction of
610 channel width (e.g. Kostynick et al., 2026), valley width (e.g. Clubb et al., 2022), and valley
611 relief (e.g. Scherler & Schwanghart, 2020), now enables consistent, large-scale characterization
612 of river channel and valley geometry across diverse landscapes. Such analyses would not only
613 help constrain empirical drainage area scaling relationships for both channels and valleys but
614 also provide a pathway to more directly evaluate the controls on the covariation between these
615 elements of valley form, including the roles of climate, tectonics, and lithology. Future efforts in
616 these areas could allow for a more rigorous assessment of how natural systems populate the
617 model domain developed here, offering a means to test whether observed combinations of
618 scaling exponents align with steady-state expectations and to identify systematic deviations that
619 may reflect differences in process regime or transient landscape adjustment.

620 This study develops and tests a 1-D profile model capturing the downstream covariation
621 in valley width, channel width, and valley relief for detachment-limited landscapes under steady

622 state conditions. The modeling framework developed here combines scaling relationships for
 623 channel width and valley width with downstream trends in valley relief generated via fluvial
 624 incision to produce longitudinal profiles in bedrock valley confinement, w_v^* , for a range of
 625 erosional efficiencies, summarized by the Erosion number, N_e , which captures the combined
 626 effects of tectonics, climate, and lithology on fluvial incision and relief generation. We find
 627 systematic variations in modeled w_v^* with increasing N_e that are in good agreement with bedrock
 628 valley confinement profiles from catchments in the King Range, CA (Snyder et al., 2003). We
 629 suggest that the framework presented here serves as a robust steady-state expectation, or null
 630 hypothesis, that can be used to compare differences in bedrock valley morphology more directly
 631 across settings and to attribute observed differences in valley morphology to specific drivers.
 632

Table 2. Parameter Values and Units

Basic Variables

x (m)	Downstream distance from channel head
x_d (m)	Maximum downstream distance
x^*	Dimensionless downstream distance
A (m^2)	drainage area
Q (m^3/s)	water discharge
w_c (m)	channel width
w_v (m)	valley width
r (m)	valley relief
w_v^*	bedrock valley confinement index
S	channel gradient
E (m/yr)	channel incision rate
U (m/yr)	rock uplift rate
P (m/yr)	precipitation rate
K (1-2m/yr, where m is a drainage area-scaling exponent)	Coefficient of erosion
N_e	Erosion Number

Scaling parameters

k_c (m^{1-2b})	channel width-drainage area coefficient
b	channel width-drainage area exponent

k_v (m^{1-2c})	valley width-drainage area coefficient
c	valley width-drainage area exponent
k_r ($m^{2-1/h}$)	Hack's law scaling coefficient
h	Hack's law scaling exponent
m	SPIM drainage area exponent
n	SPIM slope exponent

633

634 **Acknowledgments**

635 The author thanks Jedidiah Dale and Rachel Glade for helpful discussions.

636

637 **Availability Statement**

638 Topographic data used in this study are available from the U.S. Geological Survey, National Geospatial Program.

639 USGS 3DEP lidar data was downloaded from OpenTopography (<https://www.opentopography.org/>). All

640 figures utilize perceptually uniform *scientific colourmaps*. All data used to create the figures referenced in

641 the text are included as part of the supplementary information. Supplemental code is available at

642 [https://zenodo.org/records/20061114?preview=1&token=eyJhbGciOiJIUzUxMiJ9.eyJpZCI6Ijc3NTkxNTVhLTJkNWItNGRlNy04Y2M2LTNmNTZiYzE4NjA4MmIiImRhdGEiOnt9LCJyYW5kb20iOiI1MTMyZTk1NjExZTk3YzE4NTdmNDUyYjdiNTg5NDJkNCJ9.oO_MeRk9DpqAHSUSBThTHSpw79Tfz5ndRT62oO](https://zenodo.org/records/20061114?preview=1&token=eyJhbGciOiJIUzUxMiJ9.eyJpZCI6Ijc3NTkxNTVhLTJkNWItNGRlNy04Y2M2LTNmNTZiYzE4NjA4MmIiImRhdGEiOnt9LCJyYW5kb20iOiI1MTMyZTk1NjExZTk3YzE4NTdmNDUyYjdiNTg5NDJkNCJ9.oO_MeRk9DpqAHSUSBThTHSpw79Tfz5ndRT62oOwflXsERe4uAdxubZv9JCL76YLHyOEldKqvlqc5hNQfAvliw)

643 [wflXsERe4uAdxubZv9JCL76YLHyOEldKqvlqc5hNQfAvliw](https://zenodo.org/records/20061114?preview=1&token=eyJhbGciOiJIUzUxMiJ9.eyJpZCI6Ijc3NTkxNTVhLTJkNWItNGRlNy04Y2M2LTNmNTZiYzE4NjA4MmIiImRhdGEiOnt9LCJyYW5kb20iOiI1MTMyZTk1NjExZTk3YzE4NTdmNDUyYjdiNTg5NDJkNCJ9.oO_MeRk9DpqAHSUSBThTHSpw79Tfz5ndRT62oOwflXsERe4uAdxubZv9JCL76YLHyOEldKqvlqc5hNQfAvliw)

644 [. A note to editors and reviewers: *The final*](https://zenodo.org/records/20061114?preview=1&token=eyJhbGciOiJIUzUxMiJ9.eyJpZCI6Ijc3NTkxNTVhLTJkNWItNGRlNy04Y2M2LTNmNTZiYzE4NjA4MmIiImRhdGEiOnt9LCJyYW5kb20iOiI1MTMyZTk1NjExZTk3YzE4NTdmNDUyYjdiNTg5NDJkNCJ9.oO_MeRk9DpqAHSUSBThTHSpw79Tfz5ndRT62oOwflXsERe4uAdxubZv9JCL76YLHyOEldKqvlqc5hNQfAvliw)

645 *formally published reposotpru will be cited here with a DOI following review. The delay is*

646 *to enable edits to the dataset if methodological changes are suggested during the review*

647 *process that would result in material changes to the data.*

648

649 **Conflict of Interest Disclosure**

650 The author declares no conflicts of interests.

651

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Supporting Information for

Downstream patterns in bedrock valley morphology encode climatic and tectonic forcing

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Additional Supporting Information (Files uploaded separately)

DS1. Excel spreadsheet of supplementary datasets, full model output, and data-model comparison

Introduction

The supplemental information provides supporting details regarding:

- T1: Details regarding the calculation of interfluvial ridge heights and erosion numbers used in this study
- Fig. S1: c - b versus erosion number for King Range catchments
- Fig. S2: Modeled median wv^* values and normalized median wv^* values for varying valley widening coefficient, k_v , and constant valley widening exponent, c
- Fig. S3: Modeled median wv^* values and normalized median wv^* values for constant valley widening coefficient, k_v , and variable valley widening exponent, c (Fig. 4a)
- Fig. S4: wv^* profiles and associated calculations of RCI and NRCI for all King Range catchments.

Text S1.

Modeling interfluvial ridge height

Downstream profiles of interfluvial ridge height are calculated following the analytical framework developed by Willett (2010), which combines the stream power incision model (SPIM) with Hack's Law to derive expected steady-state ridge elevations as a function of downstream distance. Under steady-state conditions, bedrock incision rate equals rock uplift rate ($E = U$), such that the channel longitudinal profile can be solved analytically from the SPIM. The non-dimensional channel elevation profile is first obtained by integrating the steady-state form of the SPIM (Equation 7 in Willett, 2010), yielding a solution for dimensionless channel elevation z^* as a function of dimensionless downstream position x^* (Equation 8 in Willett, 2010). Interfluvial ridge height emerges from recognizing that tributary channels, which drain perpendicular to the main stem, follow the same SPIM solution when scaled by their own length in the cross-stream (y) direction using Hack's Law. Substituting tributary length y for downstream distance x , and scaling by x_d , yields the total relief for a tributary channel — equivalent to the interfluvial ridge height at position x^* , given by Equation 10 of Willett (2010), as follows:

$$z_d^* = y_c^* \tan \theta_c + y_d^{*(1-hm/n)} N_e \left(\frac{1}{1-hm/n} \right) \left(1 - \left(\frac{y_c^*}{y_d^*} \right)^{(1-hm/n)} \right), \frac{hm}{n} \neq 1$$
$$z_d^* = y_c^* \tan \theta_c - N_e \ln \left(\frac{y_c^*}{y_d^*} \right), \frac{hm}{n} = 1$$

where y_c^* and y_d^* are the dimensionless channel head position and tributary length respectively, θ_c is the channel head gradient, and N_e is the Erosion number (Equation 6 in Willett, 2010; Equation 6 in this study). The shape of the resulting ridge profile is governed by the exponent hm/n , such that when $hm/n < 1$ the ridge profile is concave, whereas when $hm/n > 1$ the profile becomes convex, producing greater relief at intermediate distances from the divide. The Erosion number N_e scales the overall magnitude of ridge height, such that higher N_e amplifies relief throughout the catchment, while lower N_e suppresses it, producing the relatively open, low-relief valleys characteristic of slowly uplifting or more resistant landscapes.

Erosion Number versus Uplift-Erosion Number

Similar to the Erosion Number, the Uplift-Erosion Number presented in (Whipple & Tucker, 1999) can summarize relative differences in uplift, precipitation, and erodibility. The Uplift-Erosion number also includes a vertical length scale, typically some measure of relief, to account for vertical differences in scale between catchments in addition to differences in stream length. We also calculated the Uplift-Erosion number for the King Range study catchments using the maximum catchment relief. We found that while there were differences in N_e and the Uplift-Erosion number due to the introduction of an additional scalar, that the Erosion Number and the Uplift-Erosion

number remained strongly correlated across the study catchments. Thus, to be consistent with the formulation for interfluvial ridge height in Willet (2010), we use the calculated Erosion numbers in this study. Calculated Uplift-Erosion numbers are included in the supplementary Excel spreadsheet.

Influence of precipitation

There is an orographically-driven precipitation gradient across the study sites, with $P \sim 0.5$ m/yr from the two southernmost catchments and $P \sim 3$ m/yr in the five northern catchments. It is worth noting here that the orographic precipitation effect described above may also be embedded in catchment-specific erodibility (Ferrier et al., 2013 among others). Indeed, Snyder attributes differences in their derived K values to this effect. Given this, precipitation effects can be incorporated into N_e both directly in P and indirectly in K , which may further enhance differences between catchments. Using a median value for precipitation to calculate N_e across all catchments resulted in an average percent difference of about 2% in calculated erosion numbers when compared to use of a spatially variable P . Given this, we conclude that while there is some imprint of orographic precipitation effects on erodibility, in this case N_e is not strongly influenced by doubly accounting for this effect in both K and P . Given this, we use a spatially variable P when calculating N_e in this study.

Determination of Hack's Law

We found systematic decreases in Topotoolbox-derived Hack's Law exponents with increasing N_e . In addition, we determine best-fit coefficients for Hack's Law scaling relationships with fixed exponents of $h=1.8$ and $h=2$ to more directly compare to the generalized models that we developed. These relationships resulted in slightly different Erosion Numbers for the King Range catchments but did not fundamentally change the results. Given this, we selected the Erosion Numbers calculated from our Topotoolbox-derived Hack's Law values because these resulted in the best fit to data. We report the Erosion Numbers calculated using best-fit coefficients for fixed exponents of $h=1.8$ and $h=2$ in the supplementary Excel spreadsheet.

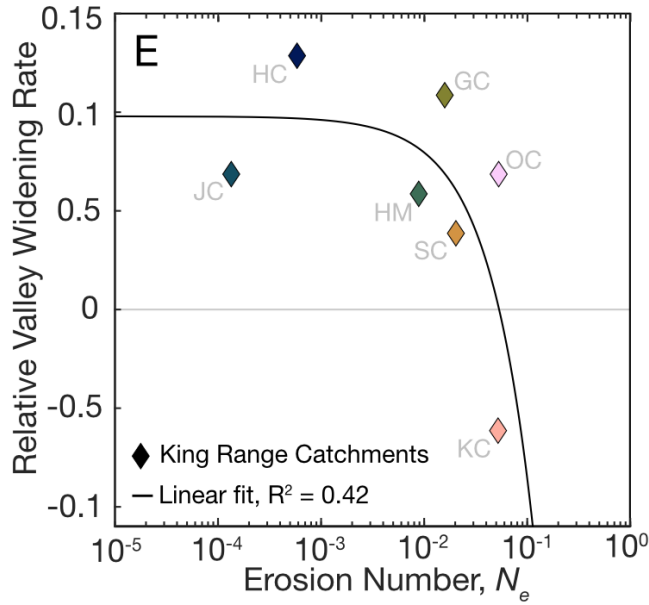


Figure S1. Difference between wv -A and wc -A scaling exponents, c and b , with calculated erosion number for King Range Catchments (Snyder et al., 2003).

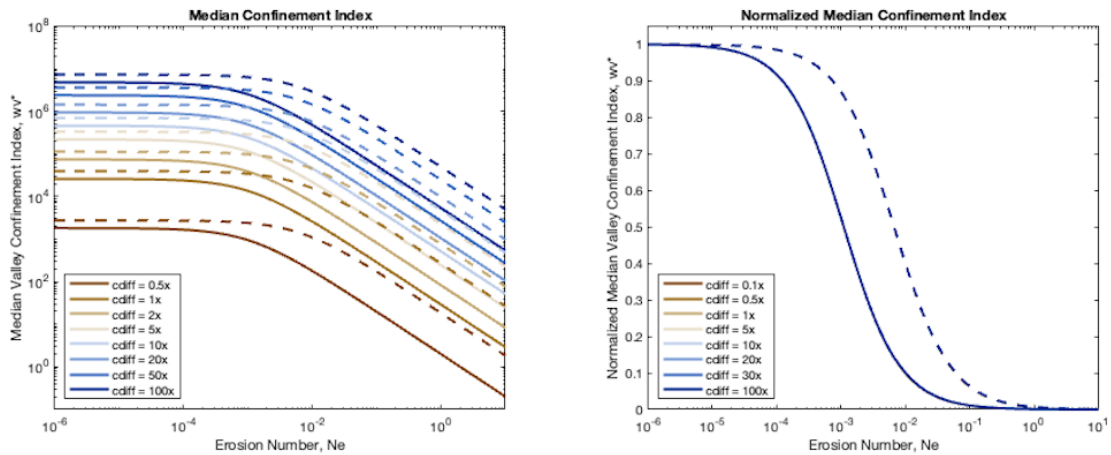


Figure S2. Modeled median wv^* values and normalized median wv^* values for varying valley widening coefficient, k_v , and constant valley widening exponent, $c=0.5$, which is equivalent to specified channel widening exponent, $b=0.5$. The range of k_v coefficients explored varied from 0.5 times the channel widening coefficient, k_w , to 100 times the channel widening coefficient, k_w . Dashed and solid lines relate to two combinations of

Hack's Law Scaling and SPIM parameterizations, $A=0.3L^2$, $m=0.5$, and $n=2/3$ (**solid**) and $A=0.8L^{1.8}$, $m=0.5$, and $n=1$ (**dashed**).

d

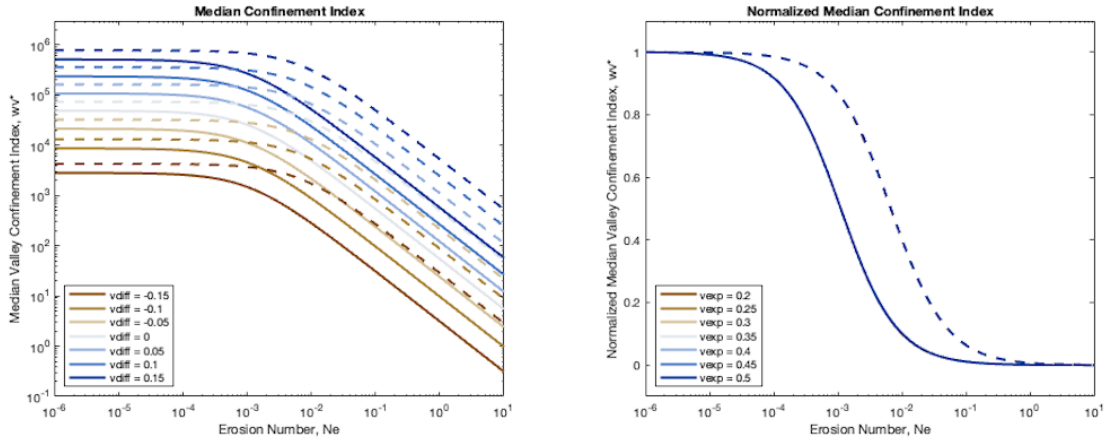


Figure S3. Modeled median wv^* values and normalized median wv^* values for constant valley widening coefficient, $k_v=2.3$ (where $k_w=0.1$), and varying valley widening exponent, $c=0.2-0.5$, compared to specified channel widening exponent, $b=0.5$. Data are consistent with Fig. 4 from the main text, but demonstrate the offset in median wv^* prior to normalizing datasets for comparison to King Range data. Dashed and solid lines relate to two combinations of Hack's Law Scaling and SPIM parameterizations, $A=0.3L^2$, $m=0.5$, and $n=2/3$ (**solid**) and $A=0.8L^{1.8}$, $m=0.5$, and $n=1$ (**dashed**).

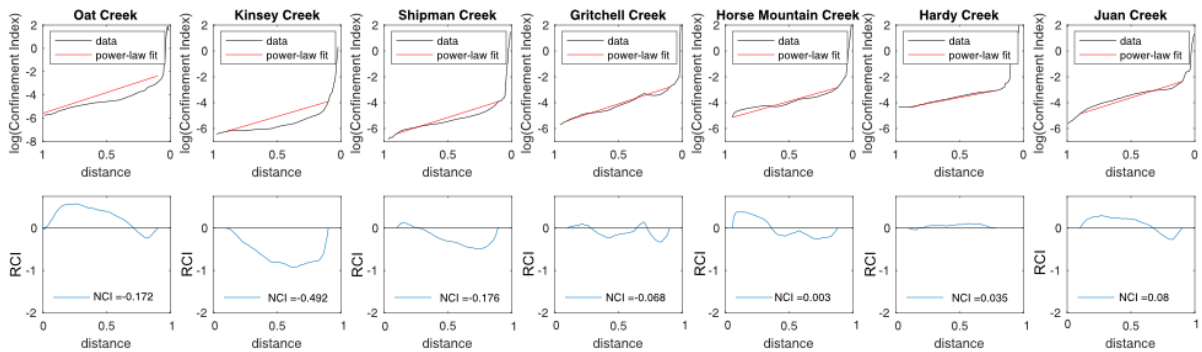


Figure S4. wv^* profiles for all King Range catchments and associated residual confinement (RCI) values and NRCI values.

Data Set S1. Excel spreadsheet summarizing all data used to produce figures in the main text, full model output, and associated model-data comparison

