

# 1 **An integral approach to bedrock river profile analysis**

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8

## 9 **Abstract**

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11 Bedrock river profiles are often interpreted with the aid of slope-area analysis, but noisy

12 topographic data make such interpretations challenging. We present an alternative

13 approach based on an integration of the steady-state form of the stream power equation.

14 The main component of this approach is a transformation of the horizontal coordinate

15 that converts a steady-state river profile into a straight line with a slope that is simply

16 related to the ratio of the uplift rate to the erodibility. The transformed profiles, called chi

17 plots, have other useful properties, including co-linearity of steady-state tributaries with

18 their main stem and the ease of identifying transient erosional signals. We illustrate these

19 applications with analyses of river profiles extracted from digital topographic datasets.

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## 24 **Introduction**

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26 Bedrock rivers record information about a landscape's bedrock lithology, tectonic  
27 context, and climate history. It has become common practice to use bedrock river profiles  
28 to test for steady-state topography, infer deformation history, and calibrate erosion  
29 models (see reviews by Whipple, 2004, and Wobus et al., 2006). The most widely used  
30 models of bedrock river incision express the erosion rate in terms of channel slope and  
31 drainage area, which makes them easy to apply to topographic measurements and  
32 incorporate into landscape evolution models. We focus on the stream power equation:

$$\frac{\partial z}{\partial t} = U(x,t) - K(x,t)A(x,t)^m \left| \frac{\partial z}{\partial x} \right|^n \quad (1)$$

33 where  $z$  is elevation,  $t$  is time,  $x$  is horizontal upstream distance,  $U$  is the rate of rock  
34 uplift relative to a reference elevation,  $K$  is an erodibility coefficient,  $A$  is drainage area,  
35 and  $m$  and  $n$  are constants. Although equation (1) is commonly referred to as the stream  
36 power equation, it can be derived from the assumption that erosion rate scales with either  
37 stream power per unit area of the bed (Seidl and Dietrich, 1992; Howard et al., 1994) or  
38 bed shear stress (Howard and Kerby, 1983).

39

40 If the stream power equation is used to describe the evolution of a river profile, a  
41 common analytical approach is to assume a topographic steady state ( $\partial z/\partial t = 0$ ) with  
42 uniform  $U$  and  $K$  and solve equation (1) for the channel slope:

$$\left| \frac{dz}{dx} \right| = \left( \frac{U}{K} \right)^{\frac{1}{n}} A(x)^{\frac{m}{n}} \quad (2)$$

43 Equation (2) predicts a power-law relationship between slope and drainage area. If such a  
44 power law is observed for a given profile, it supports the steady state assumption, and the  
45 exponent and coefficient of a best-fit power law can be used to infer  $m/n$  and  $(U/K)^{1/n}$ ,  
46 respectively. Alternatively, deviations from a power law slope-area relationship may be  
47 evidence of transient evolution of the river profile, variations in bedrock erodibility, or  
48 transitions to other dominant erosion and transport mechanisms (Whipple and Tucker,  
49 1999; Tucker and Whipple, 2002; Stock et al., 2005).

50

51 Slope-area analysis has been widely applied to the study of bedrock river profiles (e.g.,  
52 Flint, 1974; Tarboton et al., 1989; Wobus et al., 2006), but it suffers from significant  
53 limitations. Topographic data are subject to errors and uncertainty and are typically noisy.  
54 Estimates of slope obtained by differentiating a noisy elevation surface are even noisier.  
55 This typically causes considerable scatter in slope-area plots, which makes it challenging  
56 to identify a power-law trend with adequate certainty. Perhaps more concerning is the  
57 possibility that the scatter may obscure deviations from a simple power law that could  
58 indicate a change in process, a transient signal, or a failure of the stream power model.

59 Another limitation of slope-area analysis is that the slope measured in a coarsely sampled  
60 topographic map may differ from the reach slope relevant to flow dynamics.

61

62 Strategies have been proposed to cope with some of these problems. In the common case  
63 of digital elevation maps (DEMs) that contain stair-step artefacts associated with the  
64 original contour source maps, for example, sampling at a regular and carefully selected  
65 elevation interval can extract the approximate points where the stream profile crosses the

66 original contours (Wobus et al., 2006). This method requires care, however, and at best it  
67 reproduces the slopes that correspond to the original contours, which may be inaccurate.  
68 Measuring the slope over elevation intervals that correspond to long horizontal distances  
69 can compound the problem of measuring an average slope that differs from the local  
70 slope that drives flow. Furthermore, as Wobus et al. (2006) note, the contour sampling  
71 approach cannot distinguish between artefacts associated with the DEM generation  
72 procedure and real topographic features. Other common techniques for reducing noise  
73 and uncertainty in slope-area analyses include smoothing the river profile and logarithmic  
74 binning of slope measurements. Some of these approaches have been shown to yield  
75 good results (Wobus et al., 2006), but all introduce biases that are difficult to evaluate  
76 without field surveys.

77

78 In this paper, we propose a more robust method that alleviates many of these problems by  
79 avoiding measurements of channel slope. Our method uses elevation instead of slope as  
80 the dependent variable, and a spatial integral of drainage area as the independent variable.  
81 This approach has additional advantages that include the simultaneous use of main stem  
82 and tributaries to calibrate the stream power law, the ease of comparing profiles with  
83 different uplift rates, erosion parameters, or spatial scales, and clearer identification of  
84 transient signals. We present examples that demonstrate these advantages.

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89 **Transformation of river profiles**

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91 *Change of horizontal coordinate*

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93 Our procedure is based on a change of the horizontal spatial coordinate of a river  
94 longitudinal profile. Separating variables in equation (2), assuming for generality that  $U$   
95 and  $K$  may be spatially variable, and integrating yields

$$\int dz = \int \left( \frac{U(x)}{K(x)A(x)^m} \right)^{\frac{1}{n}} dx \quad (3)$$

96 Performing the integration in the upstream direction from a base level  $x_b$  to an  
97 observation point  $x$  yields an equation for the elevation profile:

$$z(x) = z(x_b) + \int_{x_b}^x \left( \frac{U(x)}{K(x)A(x)^m} \right)^{\frac{1}{n}} dx \quad (4)$$

98 There is no special significance associated with the choice of  $x_b$ ; it is merely the  
99 downstream end of the portion of the profile being analysed. The integration can also be  
100 performed in the downstream direction, but it is best to use the upstream direction for  
101 reasons that will become apparent below.

102

103 Equation (4) applies to cases in which the profile is in steady state, but is spatially  
104 heterogeneous (if, for example, the profile crosses an active fault or spans different rock  
105 types, or if precipitation rate varies over the drainage basin). In the case of spatially  
106 invariant uplift rate and erodibility, the equation for the profile reduces to a simpler form,

$$z(x) = z(x_b) + \left(\frac{U}{K}\right)^{\frac{1}{n}} \int_{x_b}^x \frac{dx}{A(x)^{\frac{m}{n}}} \quad (5)$$

107 To create transformed river profiles with units of length on both axes, it is convenient to  
 108 introduce a reference drainage area,  $A_0$ , such that the coefficient and integrand in the  
 109 trailing term are dimensionless,

$$z(x) = z(x_b) + \left(\frac{U}{KA_0^m}\right)^{\frac{1}{n}} \chi \quad (6a)$$

110 with

$$\chi = \int_{x_b}^x \left(\frac{A_0}{A(x)}\right)^{\frac{m}{n}} dx \quad (6b)$$

111 Equation (6) has the form of a line in which the dependent variable is  $z$  and the  
 112 independent variable is the integral quantity  $\chi$ , which has units of distance. The  $z$ -  
 113 intercept of the line is the elevation at  $x_b$ , and the dimensionless slope is  $(U/K)^{1/n}/A_0^{m/n}$ .

114 We refer to a plot of  $z$  vs.  $\chi$  for a river profile as a “chi plot.”

115

116 The use of this coordinate transformation to linearize river profiles was originally  
 117 proposed by Royden et al. (2000), and has subsequently been used to determine stream  
 118 power parameters (Sorby and England, 2004; Harkins et al., 2007; Whipple et al., 2007).  
 119 In this paper, we expand on this approach and explore additional applications of chi plots.  
 120 As we show in the examples below, a chi plot can be useful even if  $U$  and  $K$  are spatially  
 121 variable, or if the profile is not in steady state. The coordinate  $\chi$  in equation (6) is also  
 122 similar to the dimensionless horizontal coordinate  $\chi$  in the analysis of Royden and Perron

123 (2012), which can be referred to for a more theoretical treatment of the stream power  
124 equation.

125

### 126 *Measuring $\chi$*

127

128 It is usually not possible to evaluate the integral quantity  $\chi$  in equation (6) analytically,  
129 but given a series of upslope drainage areas measured at discrete values of  $x$  along a  
130 stream profile, it is straightforward to approximate the value of  $\chi$  at each point using the  
131 trapezoid rule or another suitable approximation. If the points along the profile are spaced  
132 at approximately equal intervals, the simplest approach is to calculate the cumulative sum  
133 of  $[A_0/A(x)]^{m/n}$  along the profile in the upstream direction and multiply by the average  
134 distance between adjacent points. (Using the average distance avoids the “quantization”  
135 effect introduced by a steepest descent path through gridded data, in which point-to-point  
136 distances can only have values of  $\delta$  or  $\delta\sqrt{2}$ , where  $\delta$  is the grid resolution.) If  $\delta$  varies  
137 significantly along the profile, or varies systematically with  $x$ , it is preferable to calculate  
138 the cumulative sum of  $[A_0/A(x)]^{m/n} \delta(x)$ . If desired,  $\delta(x)$  can be smoothed with a moving  
139 average before performing the summation.

140

141 In most cases, the value of  $m/n$  required to compute  $\chi$  will be unknown. In the next  
142 section, we illustrate a procedure for finding  $m/n$  that improves on conventional slope-  
143 area analysis.

144

### 145 **Examples**

146

147 *Identifying steady-state profiles*

148

149 The preceding analysis predicts that a steady-state bedrock river profile will have a linear  
150 chi plot. To demonstrate how the coordinate transformation can be used to identify a  
151 steady state river profile, we analysed the longitudinal profile of Cooskie Creek (Fig. 1a),  
152 one of several bedrock rivers in the Mendocino Triple Junction (MTJ) region of northern  
153 California studied previously by Merritts and Vincent (1989) and Snyder et al. (2000,  
154 2003a,b). We determined upstream distance, elevation, and drainage area along the  
155 profile by applying a steepest descent algorithm to a DEM with 10 m grid spacing. For a  
156 range of  $m/n$  values ranging from 0 to 1, we calculated  $\chi$  in equation (6), performed a  
157 linear least-squares regression of elevation against  $\chi$ , and recorded the  $R^2$  value as a  
158 measure of goodness of fit. A plot of  $R^2$  against  $m/n$  (Fig. 1b) has a well-defined  
159 maximum at  $m/n = 0.36$ , implying that this is the best-fitting value. We then transformed  
160 the longitudinal profile according to equation (6) with  $m/n = 0.36$  and  $A_0 = 1 \text{ km}^2$ . The  
161 resulting chi plot (Fig. 1c) shows that the transformed profile closely follows a linear  
162 trend, suggesting that the profile is nearly in steady state. The slope of the regression line  
163 is 0.12, which, combined with an uplift rate of 3.5 mm/yr inferred from uplifted marine  
164 terraces (Merritts and Bull, 1989), implies an erodibility  $K = 0.0002 \text{ m}^{0.28}/\text{yr}$  for  $n = 1$ .  
165 Note that the stair-step features in the longitudinal profile (Fig. 1a), which would produce  
166 considerable scatter in a slope-area plot, do not interfere with the regression analysis, and  
167 introduce only minor deviations from the linear trend in the chi plot of the transformed  
168 profile (Fig. 1c).



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170

171 *Using tributaries to estimate stream power parameters*

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173 A useful property of the coordinate transformation is that it scales points with similar  
174 elevations to similar values of  $\chi$ , even if those points have different drainage areas. This  
175 implies that tributaries that are in steady state and that have the same uplift rate and  
176 erosion parameters as the main stem should be co-linear with the main stem in a chi plot.  
177 The co-linearity of tributaries and main stem provides a second, independent constraint  
178 on  $m/n$ : in theory, the correct value of  $m/n$  should both linearize all the profiles and  
179 collapse the tributaries and main stem to a single line. This highlights one reason for  
180 performing the integration in equation (3) in the upstream direction: tributaries have the  
181 same elevation as the main stem at their downstream ends, but not at their upstream ends.

182

183 Fig. 2 illustrates this principle with an analysis of Rush Run in the Allegheny Plateau of  
184 northern West Virginia. We extracted profiles of the main stem and nine tributaries of  
185 Rush Run from a DEM with 3 m grid spacing (Fig. 2a). The tributary longitudinal  
186 profiles differ from one another and from the main stem profile (Fig. 2b). Transforming  
187 the profiles with three different values of  $m/n$  (Fig. 2c-e) demonstrates how the best  
188 choice of  $m/n$  collapses the tributaries and main stem on a chi plot (Fig. 2d); for other  
189 values of  $m/n$ , the tributaries have systematically higher (Fig. 2c) or lower (Fig. 2e)  
190 elevations than the main stem in transformed coordinates.

191

192 In practice, the value of  $m/n$  that best collapses the tributaries and main stem is not  
193 always the value that maximizes the linearity of each individual profile. In the case of  
194 Rush Run, the value of  $m/n = 0.65$  that best collapses the tributaries makes the main stem  
195 slightly concave down in transformed coordinates (Fig. 2d). Provided there are no  
196 systematic differences in erodibility or precipitation rates between the main stem and  
197 tributaries, this minor discrepancy may be an indication that the drainage basin is slightly  
198 out of equilibrium. Alternatively, it could be an indication that the mechanics of channel  
199 incision are not completely described by the stream power equation. This example  
200 illustrates how the comparison of transformed tributary and main stem profiles can  
201 provide a perspective on drainage basin evolution that would be difficult to attain with  
202 slope-area analysis.

203

#### 204 *Comparisons among profiles*

205

206 Another common application of river profile analysis is to identify topographic  
207 differences among rivers that are thought to experience different uplift or precipitation  
208 rates or that have eroded through different rock types (e.g., Kirby and Whipple, 2001;  
209 Kirby et al., 2003). These effects are modelled by the uplift rate  $U$  and the erodibility  
210 coefficient  $K$ . The coefficient of the power law in equation (2), which includes the ratio  
211 of these two parameters, is often referred to as a steepness index, because, all else being  
212 equal, the steady-state relief of the river profile is higher when  $U/K$  is larger. The  
213 steepness index is usually determined from the intercept of a linear fit to log-transformed  
214 slope and area data. The uncertainty in this intercept can be substantial due to scatter in

215 the slope-area data (Harkins et al., 2007). In our coordinate transformation, the steepness  
216 index is simply the slope of the transformed profile,  $dz/d\chi$ , which provides a means of  
217 estimating  $U/K$  that is less subject to uncertainty (Royden et al., 2000; Sorby and  
218 England, 2004; Harkins et al., 2007; Whipple et al., 2007) as well as an intuitive visual  
219 assessment of differences among profiles.

220

221 To illustrate this point, we analysed 18 of the profiles from the MTJ region studied by  
222 Snyder et al. (2000) (Fig. 3a). The profiles span an inferred increase in uplift rate  
223 northward along the coast from roughly 0.5 mm/yr to roughly 4 mm/yr associated with  
224 the passage of the Mendocino Triple Junction (Fig. 3c; Merritts and Bull, 1989; Merritts  
225 and Vincent, 1989; Merritts, 1996). The topographic data and procedures were the same  
226 as in the Cooskie Creek example. We determined the best-fitting value of  $m/n$  for each  
227 profile with the approach in Fig. 1b, and found a mean  $m/n$  of  $0.46 \pm 0.11$  (s.d.).

228

229

230 When comparing the steepness of transformed profiles, it is important to use the same  
231 values of  $A_0$  and  $m/n$  to calculate  $\chi$ . We therefore transformed all the profiles using  $A_0 = 1$   
232  $\text{km}^2$  and  $m/n = 0.46$  (Fig. 1b). The goodness of linear fits to the profiles using this mean  
233 value of  $m/n$  (average  $R^2$  of 0.992) is nearly as good as when using the best-fitting  $m/n$   
234 for each profile (average  $R^2$  of 0.995). (Note that these measures of  $R^2$  are inflated by  
235 serially correlated residuals – see Discussion section – but the comparison of their  
236 relative values is valid.) With the profiles' concavity largely removed by the  
237 transformation, the effect of uplift rate on profile steepness (the slopes of the profiles in

238 Fig. 3b) is very apparent, whereas a very careful slope-area analysis is required to resolve  
239 the steepness difference due to the noise in the elevation data (compare to Fig. 4 of  
240 Wobus et al. (2006)).

241

242 The analysis in Fig. 3 also supports the conclusion of Snyder et al. (2000) that the  
243 difference in steepness between the profiles in the zones of fast uplift (red profiles in Fig.  
244 3b) and slower uplift (blue profiles in Fig. 3b) is less than expected if only uplift rate  
245 differs between these two zones. The dimensionless slope of the transformed profiles is  
246  $0.21 \pm 0.06$  (mean  $\pm$  s.d.) for those inferred to be experiencing uplift rates of 3 to 4 mm/yr  
247 and  $0.13 \pm 0.01$  for those inferred to be experiencing uplift rates of 0.5 mm/yr, a slope  
248 ratio of only  $1.62 \pm 0.48$  for a six- to eight-fold difference in uplift rate. If these uplift  
249 rates are correct, and if  $n$  is less than  $\sim 2$ , as is typically inferred (Howard and Kerby,  
250 1983; Seidl and Dietrich, 1992; Seidl et al., 1994; Rosenbloom and Anderson, 1994;  
251 Stock and Montgomery, 1999; Whipple et al., 2000; van der Beek and Bishop, 2003),  
252 there must be other differences among the profiles that affect the steepness. Given the  
253 inferred uniformity of the lithology in the MTJ region (Snyder et al., 2003a, and  
254 references therein), one possible explanation is that increased rainfall and associated  
255 changes in weathering and erosion mechanisms have elevated the erodibility,  $K$ , in the  
256 zone of faster uplift and higher relief (Snyder et al., 2000, 2003a,b).

257

258 The importance of variables other than uplift rate is most apparent in the chi plots of  
259 Fourmile and Cooskie Creeks (orange profiles in Fig. 3b). These rivers have only slightly  
260 slower inferred uplift rates than the red profiles in Fig. 3b, but they have much gentler

261 slopes. In fact, their slopes are comparable to those of the blue profiles in the slower  
262 uplift zone. A possible explanation for this discrepancy is that local structural  
263 deformation has rendered the bedrock more easily erodible in the Cooskie Shear Zone.  
264 Whatever the reason for the reduced effect of uplift rate on profile steepness, this  
265 example from the MTJ region demonstrates the ease of comparisons between  
266 transformed river profiles believed to be in steady state with respect to different erosion  
267 parameters or rates of tectonic forcing.

268

269

#### 270 *Transient signals*

271

272 Even if a river is not in a topographic steady state, a chi plot of its longitudinal profile can  
273 be useful. Just as transformed tributaries plot co-linearly with a transformed main stem,  
274 transient signals with a common origin, propagating upstream through different channels,  
275 plot in the same location in transformed coordinates ( $\chi$  and  $z$ ). Whipple and Tucker  
276 (1999) noted that transient signals in river profiles governed by the stream power  
277 equation propagate vertically at a constant rate, and exploited this property to calculate  
278 timescales for transient adjustment of profiles in response to a step change in  $K$  or  $U$ . The  
279 transformation presented in this paper removes the effect of drainage area, and therefore  
280 shifts the transient signals in the profiles to the same horizontal position ( $\chi$ ), provided  
281 that  $K$  and  $U$  are uniform.

282

283 We illustrate this property with an example from the Big Tujunga drainage basin in the  
284 San Gabriel Mountains of southern California (Fig. 4a), where Wobus et al. (2006)  
285 observed an apparent transient signal in multiple tributaries within the basin.  
286 Longitudinal profiles (Fig. 4b) reveal steep reaches in the main stem and some tributaries  
287 at elevations of roughly 900-1000 m but different streamwise positions. Differences in  
288 drainage basin size and shape make it difficult to compare the profiles and determine if  
289 and how these features are related. Transforming the profiles using  $m/n = 0.4$ , the value  
290 that best collapses the tributaries to the main stem and linearizes the profiles, clarifies the  
291 situation (Fig. 4c). The steep sections of the transformed profiles plot in nearly the same  
292 location. The transformed profiles also have a systematically steeper slope downstream of  
293 this knick point than upstream, suggesting an increase in uplift rate, the preferred  
294 interpretation of Wobus et al. (2006), or a reduction in erodibility. It is difficult to tell  
295 whether the knick point is stationary or migrating (Royden and Perron, 2012), but the  
296 lack of an obvious fault or lithologic contact suggests that it may be a transient signal that  
297 originated downstream of the confluence of the analysed profiles and has propagated  
298 upstream to varying extents.

299

300 The horizontal overlap of the steep sections in the chi plot in Fig. 3c is compelling, but it  
301 is not perfect. The residual offsets may have arisen from spatial variability in channel  
302 incision processes, precipitation, or bedrock erodibility. This difference, which is not  
303 obvious in the original longitudinal profiles (Fig. 3b) and would probably not be apparent  
304 in a slope-area plot, highlights the sensitivity of the coordinate transformation technique.

305

306

## 307 **Discussion**

308

### 309 *Advantages of the integral approach to river profile analysis*

310

311 The approach described in this paper has several advantages over slope-area analysis.

312 The most significant advantage is that it obviates the need to calculate slope from noisy  
313 topographic data. This makes it possible to perform useful analyses with elevation data  
314 that would ordinarily be avoided. The landscape in Fig. 2, for example, has sufficiently  
315 low relief that even elevation data derived from laser altimetry contains enough noise to  
316 frustrate a slope-area analysis, but the transformed profiles are relatively easy to interpret.

317

318 The reduced scatter relative to slope-area plots provides better constraints on stream  
319 power parameters estimated from topographic data. In addition, a chi plot can potentially  
320 provide an independent constraint on both  $m/n$  and  $U/K$ , because the profile fits are  
321 constrained two ways: by the requirement to linearize individual profiles (Fig. 1), and by  
322 the requirement to align tributaries with the main stem (Fig. 2). Although steady state  
323 tributary and main stem channels should also be co-linear on a logarithmic slope-area  
324 plot, they typically have different drainage areas, and therefore do not usually overlap. In  
325 contrast, the integral method produces transformed longitudinal profiles with overlapping  
326 chi coordinates, making it easier to visually assess the match between tributaries and  
327 main stem.

328

329 Removing the effect of drainage area through this coordinate transformation makes it  
330 possible to compare river profiles independent of their spatial scale. This is useful both  
331 for comparing different drainage basins (Fig. 3) and for comparing channels within a  
332 drainage basin (Fig. 4). Transient erosional features, such as knick points, that originated  
333 from a common source should plot at the same value of  $\chi$  in all affected channels (Fig.  
334 4). Transient features are also easier to identify in a chi plot because it is easy to see  
335 departures from a linear trend with relatively little noise. Similarly, transformed profiles  
336 should accentuate transitions from bedrock channels to other process zones within the  
337 fluvial network, such as channels in which elevation changes are dominated by alluvial  
338 sediment transport or colluvial processes and debris flows (Whipple and Tucker, 1999;  
339 Tucker and Whipple, 2002; Stock et al., 2005).

340

341 Finally, the coordinate transformation presented here is compatible with the analytical  
342 solutions of Royden and Perron (2012), which aid in understanding the transient  
343 evolution of river profiles governed by the stream power equation. As noted above, the  
344 integral quantity  $\chi$  in equation (6) is similar to the dimensionless horizontal coordinate  $\chi$   
345 used by Royden and Perron (2012) to derive analytical solutions for profiles adjusting to  
346 spatial and temporal changes in uplift rate, erodibility, or precipitation. (For uniform  $K$ ,  
347 as is assumed in this paper, it is linearly proportional to their  $\chi$ .) River profiles  
348 transformed according to equation (6) can easily be compared with these solutions to  
349 investigate possible scenarios of transient profile evolution.

350

351 *Disadvantages of the integral approach*



352

353 The main disadvantage of the integral approach is that the coordinate transformation  
354 requires knowledge of  $m/n$ , which is usually not known *a priori*. However, we have  
355 demonstrated a simple iterative approach for finding the best-fitting value of  $m/n$  that is  
356 easy to implement (Fig. 1b). Moreover, the dependence of the transformation on  $m/n$   
357 provides an additional constraint on  $m/n$ , the co-linearity of main stem and tributaries,  
358 which is not available in slope-area analysis.

359

360 Another drawback of the integral method is that chi plots, like slope-area plots, do not  
361 account for variations on, or inadequacy of, the stream power/shear stress model.

362 Multiple studies have found that effects not included in equation (1), including erosion  
363 thresholds (e.g., Snyder et al., 2003b; DiBiase and Whipple, 2011), discharge variability  
364 (e.g., Snyder et al., 2003b; Lague et al., 2005; DiBiase and Whipple, 2011) and abrasion  
365 and cover by sediment (e.g., Whipple and Tucker, 2002; Turowski et al. 2007), can  
366 influence the longitudinal profiles of bedrock rivers. It may be possible to use an integral  
367 approach to derive definitions of  $\chi$  for channel incision models that include these effects,  
368 but such an analysis is beyond the scope of this paper.

369

370 The form of the integral method presented here can, however, help to identify profiles  
371 that are not adequately described by equation (1), because their chi plots should be non-  
372 linear. Given the larger uncertainties in slope-area analyses, it is possible that some  
373 profiles have incorrectly been identified as steady state, or otherwise consistent with the  
374 stream power equation, with deviations from the model prediction concealed by the

375 scatter in the slope-area data. The integral method, which is less susceptible to noise in  
376 elevation data, is a more sensitive tool for identifying such deviations.

377

378 *Evaluating uncertainty*

379

380 The transformation of river longitudinal profiles into linear profiles with little scatter  
381 raises the question of how to estimate the uncertainty in stream power parameters  
382 determined from topographic data. The most obvious approach is to use the uncertainties  
383 obtained by fitting a model to an individual profile. Slope-area analysis of steady-state  
384 profiles is appealing from this standpoint, because a least-squares linear regression of  
385 log-transformed slope-area data provides an easy way of estimating the uncertainty in  
386  $m/n$  (the slope of the regression line) and  $(U/K)^{1/n}$  (the intercept). However, the resulting  
387 uncertainties mostly describe how precisely one can measure slope, not how precisely the  
388 parameters are known for a given landscape.

389

390 The integral method presented in this paper makes this distinction more apparent. For  
391 example, when the profile of Cooskie Creek in Fig. 1a is transformed with the best-fitting  
392 value of  $m/n$ , there is a very small uncertainty (0.2% standard error) in the slope of the  
393 best linear fit (Fig. 1c), but this small uncertainty surely overestimates the precision with  
394 which  $(U/K)^{1/n}$  can be measured for the bedrock rivers of the King Range. Statistically,  
395 this uncertainty in the slope of the regression line is also an underestimate because the  
396 transformed profile is a continuous curve, and therefore the residuals of the linear fit are  
397 serially correlated. This property of the data does not bias the regression coefficients,

398  $z(x_b)$  and  $(U/K)^{1/n}/A_0^{m/n}$ , but it does lead to underestimates of their uncertainties. Thus, if a  
399 chi plot is used to estimate the uncertainty of stream power parameters by fitting a line to  
400 a single river profile, a procedure for regression with autocorrelated residuals must be  
401 used (e.g., Kirchner, 2001).

402

403 There are better ways to estimate uncertainty in stream power parameters. One, which  
404 can be applied to either slope-area analysis or the integral method, is to make multiple  
405 independent measurements of different river profiles. This was the approach used to  
406 estimate the uncertainty in steepness within each uplift zone in the MTJ region example.  
407 The standard errors of the mean steepness among profiles within the fast uplift zone  
408 (8.6%) and the slower uplift zone (3.4%) are considerably larger than the standard errors  
409 of steepness for individual profiles. If it is possible to measure multiple profiles that are  
410 believed to be geologically similar, this approach provides estimates of uncertainty that  
411 are more meaningful than the uncertainty in the fit to any one profile. Alternatively, if  
412 only one drainage basin is analysed, the integral method provides a new means of  
413 estimating the uncertainty in  $m/n$ : comparing the value that best linearizes the main stem  
414 profile (Fig. 1c) with the value that maximizes the co-linearity of the main stem with its  
415 tributaries (Fig. 2c-e).

416

417

## 418 **Conclusions**

419

420 We have described a simple procedure that makes bedrock river profiles easier to  
421 interpret than in slope-area analysis. The procedure eliminates the need to measure  
422 channel slope from noisy topographic data, linearizes steady-state profiles, makes steady-  
423 state tributaries co-linear with their main stem, and collapses transient erosional signals  
424 with a common origin. The procedure is well suited to analysing both steady state and  
425 transient profiles, and is useful for interpreting the lithology, tectonic histories, and  
426 climate histories of river profiles, even from coarse or imprecise topographic data.

427

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429

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435

436

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563

## 564 **Figure Captions**

565

566 **Figure 1.** Profile analysis of Cooskie Creek in the northern California King Range, USA. (a) Longitudinal  
567 profile of the bedrock section of the Creek, as determined by Snyder et al. (2000), extracted from the 1/3  
568 arcsecond (approximately 10 m) U. S. National Elevation Dataset using a steepest descent algorithm. (b)  $R^2$   
569 statistic as a function of  $m/n$  for least-squares regression based on equation (6). The maximum value of  $R^2$ ,  
570 which corresponds to the best linear fit, occurs at  $m/n = 0.36$ . (c) Chi plot of the longitudinal profile (black  
571 line), transformed according to equation (6) with  $A_0 = 1 \text{ km}^2$ , compared with the regression line for  $m/n =$   
572  $0.36$  (gray line). If the stream power equation is valid and the uplift rate  $U$  and erodibility coefficient  $K$  are  
573 spatially uniform, the slope of the regression line is  $(U/K)^{1/n}/A_0^{m/n}$ .

574

575 **Figure 2.** Rush Run drainage basin in the Allegheny Plateau of northern West Virginia, USA. (a) Shaded  
576 relief map with black line tracing the main stem and gray lines tracing nine tributaries. Digital elevation  
577 data are from the 1/9 arcsecond (approximately 3 m) U.S. National Elevation Dataset. UTM zone 17 N. (b)  
578 Longitudinal profiles of the main stem (black line) and tributaries (gray lines). (c-e) Chi plots of  
579 longitudinal profiles, transformed according to equation (6), using  $A_0 = 10 \text{ km}^2$  and (c)  $m/n = 0.55$ , (d)  $m/n$   
580  $= 0.65$ , (e)  $m/n = 0.75$ .

581

582 **Figure 3.** River profiles in the Mendocino Triple Junction region of northern California, USA. (a) Shaded  
583 relief map showing locations of bedrock sections of the channels, as determined by Snyder et al. (2000),  
584 extracted from the 1/3 arcsecond (approximately 10 m) National Elevation Dataset. Blue profiles have  
585 slower uplift rates, red profiles have faster uplift rates, and orange profiles have faster uplift rates but are  
586 located in the Cooskie Shear Zone. (b) Chi plot of longitudinal profiles transformed according to equation  
587 (6), using  $A_0 = 1 \text{ km}^2$  and  $m/n = 0.46$ , the mean of the best-fitting values for all the profiles. Profiles have  
588 been shifted so that their downstream ends are evenly spaced along the horizontal axis. Elevation is  
589 measured relative to the downstream end of the bedrock section of each profile. (c) Uplift rate at the

590 location of each drainage basin, inferred from dating of marine terraces (Merritts and Bull, 1989; Merritts  
591 and Vincent, 1989; Snyder et al., 2000).

592

593 **Figure 4.** Big Tujunga drainage basin in the San Gabriel Mountains of California, USA. (a) Shaded relief  
594 map with black line tracing the main stem and gray lines tracing seven tributaries. Digital elevation data are  
595 from the 1/3 arcsecond (approximately 10 m) U.S. National Elevation Dataset. UTM zone 11 N. (b)  
596 Longitudinal profiles of the main stem (black line) and tributaries (gray lines). The gap in the main stem is  
597 the location of Big Tujunga Dam and Reservoir. (c) Chi plot of longitudinal profiles, transformed according  
598 to equation (6), using  $A_0 = 10 \text{ km}^2$  and  $m/n = 0.4$ , illustrating the approximate co-linearity of the tributaries  
599 and main stem despite the fact that the profiles do not appear to be in steady state with respect to uniform  
600 erodibility and uplift. Two straight dashed segments with different slopes are shown for comparison.







