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

2
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7 8 9 **Abstract**

10
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.

24 **Introduction**

25

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.

85

86

87

88

89 **Transformation of river profiles**

90

91 *Change of horizontal coordinate*

92

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 χ , 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. χ 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 χ in equation (6) is also
 122 similar to the dimensionless horizontal coordinate χ 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 χ*

127

128 It is usually not possible to evaluate the integral quantity χ 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 χ 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 δ or $\delta\sqrt{2}$, where δ is the grid resolution.) If δ 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 χ 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 χ in equation (6), performed a
157 linear least-squares regression of elevation against χ , 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).

169

170

171 *Using tributaries to estimate stream power parameters*

172

173 A useful property of the coordinate transformation is that it scales points with similar
174 elevations to similar values of χ , 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 ± 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 χ . We therefore transformed all the profiles using $A_0 = 1$
232 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 ± 0.06 (mean \pm s.d.) for those inferred to be experiencing uplift rates of 3 to 4 mm/yr
247 and 0.13 ± 0.01 for those inferred to be experiencing uplift rates of 0.5 mm/yr, a slope
248 ratio of only 1.62 ± 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 ~ 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 (χ 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 (χ), 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 χ 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 χ in equation (6) is similar to the dimensionless horizontal coordinate χ
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 χ .) 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 χ 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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438

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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.







