

1 **The role of topographic variability in river channel classification**

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24

25 **Abstract**

26 To date, subreach-scale variations in flow width and bed elevation have rarely been included in
27 channel classifications. Variability in topographic features of rivers, however, in conjunction with
28 sediment supply and discharge produces a mosaic of channel forms that provides unique habitats
29 for sensitive aquatic species. In this study we investigated the utility of topographic variability
30 attributes (TVAs) in distinguishing channel types and dominant channel formation and
31 maintenance processes in montane and lowland streams of the Sacramento River basin, California
32 USA. A stratified random survey of 161 stream sites was performed to ensure balanced sampling
33 across groups of stream reaches with expected similar geomorphic settings. For each site surveyed,
34 width and depth variability were measured at baseflow and bankfull stages, and then incorporated
35 in a channel classification framework alongside traditional reach-averaged geomorphic attributes
36 (e.g., channel slope, width-to-depth, confinement, and dominant substrate) to evaluate the
37 significance of TVAs in differentiating channel types. In contrast to more traditional attributes
38 such as slope and contributing area, which are often touted as the key indicators of hydrogeo
39 morphic processes, bankfull width variance emerged as a first-order attribute for distinguishing
40 channel types. A total of nine channel types were distinguished for the Sacramento Basin
41 consisting of both previously identified and new channel types. These results indicate that
42 incorporating TVAs in channel classification provides a quantitative basis for interpreting
43 nonuniform as well as uniform geomorphic processes, which can improve our ability to distinguish
44 linked channel forms and processes of geomorphic and ecological significance.

45 46 **Keywords**

47 Channel classification, river topography, nonuniform, channel form

48 **Introduction**

49 Building on the classic premise of Davis (1909), Thornbury (1954) stated that geomorphic
50 processes create a characteristic assemblage of landforms. Through judicious use of inverse
51 reasoning, investigation of landforms can provide an understanding of linked geomorphic
52 processes. Over the past century, studies have shown that ecological structure and function of
53 rivers are strongly influenced by channel type (e.g., Hack and Goodlett, 1960; Smith et al., 1995;
54 Vannote et al., 1980). As a result of these strong foundations, channel classification has come to
55 the forefront of river science and management as a central feature of methods for understanding,
56 protecting, and restoring rivers in North America (Rosgen 1994; Kondolf 1995; Buffington and
57 Montgomery 2013), Europe (e.g., González del Tánago and García de Jalón 2004; Orr et al.
58 2008), Australia (Brierley and Fryirs 2005), and South Africa (Rowntree and Wadeson 1998).
59 Channel classification is of critical importance today for river management, because

60 anthropogenic changes to flow regimes (Molles et al. 1998; Mailligan and Nislow 2005),
61 sediment regimes (Graf 1980; Pitlick and Van Steeter 1998; Wohl et al. 2015), and the physical
62 structure of rivers (Price et al. 2012) have led to widespread degradation of river ecosystems
63 worldwide (Dynesius and Nilsson 1994; Arthington 2012).

64 Reach-scale geomorphic settings [e.g., pool-riffle, step-pool (Montgomery and Buffington
65 1997)] distinguished by attributes related to channel form and sediment transport and supply
66 have been shown to influence ecosystem dynamics and biological diversity (Montgomery and
67 Bolton 2003; Biggs et al. 2005; Meitzen et al. 2013; Milner et al. 2015), highlighting channel
68 reach classification as a critical step in river ecosystem management. Geomorphic attributes used
69 in channel classification are often chosen to describe relevant, persistent reach-scale
70 characteristics that influence hydraulics and sediment dynamics and in turn aquatic and riparian
71 ecosystem functioning (Birkeland 1996; Hupp and Osterkamp 1996; Merritt and Wohl 2003;
72 Kasprak et al. 2016). Considerable recent efforts have been invested in developing geomorphic
73 attributes for river characterization, particularly in Europe through the implementation of the
74 Water Framework Directive (e.g., Raven et al. 1998; Orr et al. 2008; Sear et al. 2009; Polvi et al.
75 2014). Common attributes considered include uniform metrics such as reach-averaged channel
76 slope, width-to-depth ratio, entrenchment ratio, valley confinement, sinuosity, stream power, and
77 dominant channel substrate (Church 1992; Rosgen 1994; Montgomery and Buffington 1997;
78 Knighton 1999; Brierley and Fryirs 2005; Kasprak et al. 2016).

79 However, nonuniform mechanisms not well characterized or indicated by reach-averaged
80 uniform metrics have been identified as primary drivers of channel formation and maintenance in
81 many channel settings (Lane and Carlson 1953; Dietrich and Smith 1983; Thompson 1986;
82 Paustian et al. 1992; Wohl and Thompson 2000; Makaske 2001; Powell et al. 2005; Wilcox and
83 Wohl 2006; White et al. 2010). For example, subreach-scale flow convergence routing has been
84 shown to control riffle-pool formation and maintenance and the locations of sediment deposition
85 and bar instability (MacWilliams et al. 2006). In meandering and alternate bar morphologies,
86 nonuniformity is maintained primarily by the alternating converging and diverging secondary
87 transverse flow cells in and between bends, respectively, which help to maintain sediment
88 routing through the inside of meander bends (Thompson 1986).

89 Topographic variability attributes (TVAs), defined here as any measure of subreach-scale
90 variability [i.e., departures from average conditions in channel bed elevation, bankfull width,

91 curvature, and floodplain width], are closely tied to nonuniform channel processes and likely
92 offer more appropriate metrics for characterizing and comparing dominant channel processes and
93 habitat dynamics than their far more common uniform counterparts used in many channel
94 morphologies. For example, measures of subreach-scale channel width and depth variance are
95 expected to capture the frequency and magnitude distribution of flow expansions and
96 contractions associated with flow convergence routing under a dynamic flow regime
97 (MacWilliams et al. 2006). Furthermore, high within-reach topographic variability is often
98 associated with heterogeneous habitat units available across a wider range of discharges that can
99 support a variety of native biota and ecological functions (Murray et al. 2006; Scown et al.
100 2016), promoting high biodiversity (Poff and Ward 1990; Townsend and Hildrew 1994; Fausch
101 et al. 2002) and ecological resilience (Elmqvist et al. 2003; McCluney et al. 2014).

102 Channel topographic variability exists naturally and is part of a dynamic equilibrium with
103 other channel variables. At the valley scale, there are nested layers of topographic variability,
104 including variations in the width of hillsides, terraces and floodplains along a corridor (e.g.,
105 Gangodagamage et al. 2007; White et al. 2010). When a flow of a set magnitude moves through
106 a layered topographic boundary, it engages one or more of these controls and a specific scale of
107 topographic steering is initiated. That specific type of steering then drives subreach variability in
108 the hydraulic flow field that focuses erosion and deposition locally (Strom et al. 2016). For a
109 dynamic flow regime, topographic steering changes with flow and this results in a diversity of
110 stage-dependent hydraulic patch behaviors (Scown et al. 2016; Strom et al. 2016), each with a
111 different capability to promote erosion or deposition (Brown and Pasternack 2014; Grams et al.
112 2013).

113 As a result of these factors, rivers exhibit complex patterns of topographic change processes
114 that promote strong longitudinal variation in width and depth (Wyrick and Pasternack 2015).
115 Variability itself is expected to differ between reaches, because many geomorphic processes
116 control aspects of variability, such as flow convergence, avulsion, turbulence-driven scour, and
117 meander bend cut-off. One might conjecture that variability is indicated by reach-scale
118 homogenous metrics like specific stream power, and thus not needed to define channel classes,
119 but if the processes that control channel form are governed by variability, then the reverse should
120 be taken as the dominant conjecture: reach-scale homogenous metrics are the outcome of the
121 interplay between channel variability and flow, not the controls on it.

122 In spite of the established geomorphic (Thompson 1986; MacWilliams et al. 2006; White et
123 al. 2010; Gostner et al. 2013a,b; Brown et al. 2014; 2015) and ecological (Murray et al. 2006;
124 Scown et al. 2016; Elmqvist et al. 2003; McCluney et al. 2014) significance of subreach-scale
125 topographic variability, very few existing channel classifications consider TVAs. While the
126 Rosgen (1994) and Montgomery and Buffington (1997) classifications both consider the spacing
127 of individual channel-unit types along a reach (e.g., non-dimensional pool spacing measured in
128 channel widths) in their suite of geomorphic attributes, no direct measure of channel width or
129 depth variability is included. The limited consideration of TVAs in past channel classifications
130 may be due to the preference by practitioners to conduct rapid field surveys (sometimes at only
131 one cross-section per reach) in order to maximize the number of channel reaches surveyed in lieu
132 of performing more in-depth surveys across fewer reaches (Buffington and Montgomery 2013)
133 given resource limitations. With the emergence of meter-scale remote sensing of rivers, datasets
134 that support computing and analyzing TVAs will become more available, accurate, and useful
135 (Gleason and Wang 2015; Gonzalez and Pasternack 2015). There has already been significant
136 progress on the use of high resolution aerial imagery from drones to map river characteristics
137 (e.g., Lejot et al. 2007; Rivas Casado et al. 2015, 2016).

138 A few exceptions include Trainor and Church (2003) and Jaeger (2015). Trainor and Church
139 (2003) included channel depth and width variability as key geomorphic attributes in a channel
140 comparison study, but the focus on quantifying dissimilarity between channel reach pairs
141 precluded an evaluation of the relative significance of individual attributes for distinguishing
142 channel types. Jaeger (2015) considered the standard deviation of channel bed elevation (a
143 measure of depth variability) in their classification of headwater streams. However, the set-up of
144 the study as an analysis of the geomorphic significance of mountaintop mining again precluded
145 any evaluation of attribute significance. This major gap in the channel classification literature
146 indicates a need to test the value of incorporating TVAs into the suite of potentially significant
147 geomorphic attributes distinguishing ecologically relevant channel types. This must be done
148 before we can even begin to evaluate the geomorphic or ecological significance of these
149 emerging attributes compared to the more traditional reach-averaged attributes described above.

150 The purpose of this study was to investigate how TVAs can be incorporated in a channel
151 classification framework to improve the utility of morphological analysis to distinguish dominant
152 channel processes and habitat dynamics along channel networks in varied landscapes. The

153 specific study objectives were to test the use of TVAs in (i) distinguishing channel types across a
154 landscape and (ii) characterizing dominant channel processes of interest. The utility and
155 ecological implications of incorporating TVAs in a channel classification of montane and
156 lowland streams of a Mediterranean basin are then discussed and evaluated in the context of the
157 existing body of channel classification literature and current understanding of landscape form –
158 process linkages.

159 **Methodology**

160 The Rosgen channel classification (Level II, Rosgen 1994), arguably the most commonly
161 used channel classification system in North America and globally (Kasprak et al. 2016), was
162 adopted and expanded on in this study to facilitate ease of application of the proposed methods in
163 future channel classifications. The Rosgen channel classification is a stream-reach taxonomy that
164 classifies channel types using field-collected geomorphic attributes (e.g., slope, entrenchment
165 ratio, width-to-depth ratio, sinuosity, and median grain size). In an effort to support the
166 incorporation of TVAs into field-based mapping for channel classification given the common
167 constraint of resource limitations, the Rosgen channel classification procedure was extended in
168 three ways: (1) the channel network was binned into hydro-geomorphically similar groups prior
169 to field data collection using a stratified analysis of hydrologic and topographic data in a
170 Geographic Information System (GIS); (2) four TVAs consisting of within-reach low flow and
171 bankfull width and depth variance were measured in the field in addition to the traditional
172 geomorphic attributes considered by Rosgen (1994); and (3) a heuristic refinement procedure
173 was used to distinguish the most parsimonious set of physically interpretable channel types
174 instead of associating the field-observed channel types with known Rosgen classes.

176 *Study area*

177 The study was conducted in the Sacramento Basin of California, USA, encompassing the
178 largest river in the State of California by discharge (producing ~ 30% of California's surface
179 water runoff) and the second largest U.S. river draining into the Pacific Ocean (after the
180 Columbia River) (Carter and Resh 2005). This 70,000-km² basin lies between the Sierra Nevada
181 and Cascade Range to the east and the Coast Range and Klamath Mountains to the west. From its

182 headwaters in the volcanic plateau of northern California (Upper Sacramento, McCloud, and Pit
183 Rivers), the Sacramento River flows south for 715 km before reaching the Sacramento–San
184 Joaquin River Delta and San Francisco Bay. The river has many small to moderate-sized
185 tributaries (e.g., Clear, Cottonwood, Cow, Battle, Antelope, Mill, Deer, Stony, Big Chico, and
186 Butte Creek) and two large tributaries, the Feather River and the American River. The basin
187 primarily exhibits a Mediterranean climate with cold, wet winters (Oct - Apr) and warm, dry
188 summers (May - Sep) (Leung et al. 2003).

189 The basin's diverse physiographic settings range from the glacially-carved Sierra Nevada
190 mountains to lowland marshes and agricultural lands, with a total relief of about 4,300 m (USGS
191 2011). The Sacramento Basin is split into three overlying physiographic provinces: the Pacific
192 Border, the Cascade-Sierra Mountains, and the Basin and Range provinces (Fenneman and
193 Johnson 1946) (Figure 1). These provinces exhibit distinct landscape units (*sensu* Brierley and
194 Fryirs 2005) based on differential tectonic uplift, lithology, and climate (CGS 2002) and are
195 therefore expected to account for major differences in geomorphic processes and resulting
196 channel morphologies (Schmitt et al. 2007; Trainor and Church 2003). For instance, the Basin
197 and Range province consists primarily of a thick accumulation of lava flows and tuff beds,
198 supporting low slope meandering streams and large marshlands with low sediment transport
199 capacity. The Cascade-Sierra Mountains province consists of a massive tilted fault block; the
200 western slope descends in a series of undulating low-relief upland surfaces punctuated by deeply
201 incised river canyons, driving high sediment transport rates (Stock et al. 2005). The Pacific
202 Border province delineates an alluvial basin that acts as a depositional trough (CGS 2002).
203 Relationships between contributing area and channel bed composition are expected to vary
204 significantly between these provinces based on major differences in sediment regimes.

205 California's legacy of intensive and widespread hydrologic and geomorphic alteration for
206 water supply, flood control, land use change, hydropower, and mining has left the Sacramento
207 Basin's river ecosystems severely degraded (Healey et al. 2008; Hanak et al. 2011). The basin
208 simultaneously supports 2.8 million people and numerous federally endangered and threatened
209 aquatic species [e.g., winter-run Chinook salmon (*oncorhynchus tshawytscha*), Sacramento
210 splittail (*pogonichthys macrolepidotus*)] (Lindley et al. 2007; Moyle et al. 2011). Most of the
211 Sacramento Basin valley is intensively cultivated, with over 8,100 km² of irrigated agriculture.
212 Major reservoirs in the basin include Lake Shasta (5.6 km³, upper Sacramento, McCloud and Pit

213 Rivers), Lake Oroville (4.4 km³, Feather River), Lake Folsom (1.2 km³, American River), and
214 New Bullards Bar Reservoir (1.2 km³, Yuba River). In light of systemic anthropogenic alteration
215 promoting channel homogenization and simplification (Arnold et al 1982; Booth and Jackson
216 1997; Walsh et al. 2005), one might expect that topographic variability would be suppressed.
217 Therefore, if TVAs prove important here in the characterization of in-channel habitat dynamics,
218 then they are likely even more important in undisturbed settings in which topographic variability
219 is expected to be greater and thus influence habitat dynamics across a larger range of TVAs.

220 This study was constrained to one hydrologic regime found within the Sacramento Basin to
221 help isolate factors that cause diverse hydrological and geomorphic effects. An existing regional
222 hydrologic classification of California (Lane et al. 2016) was used to identify stream reaches
223 exhibiting the low-volume snowmelt and rain (LSR) regime. The LSR hydrologic regime was
224 chosen as it captures the transition from the montane snowmelt-driven to lowland rain-driven
225 flow regime and has the largest spatial footprint of hydrologic regimes in the Sacramento Basin
226 (47%); stream reaches in this hydrologic regime are expected to exhibit high geomorphic
227 variability.

228 *Channel network stratification*

229 Given the large study domain with about 100,000 reaches and limited resources, the process
230 of observing representative sites requires selecting a relatively small number of samples
231 compared to the scope of the system. If sites were selected at random, then the odds are that
232 different geomorphic settings would be observed in proportion to their frequency of occurrence,
233 and that would bias the assessment of classification, especially if too few sites of rare yet
234 important classes were sampled. Therefore, instead of random sampling, a stratified random
235 approach was used to obtain an equal effort strategy mindful of process-based controls on river
236 organization. Stratified random sampling and related variants using equal effort in each stratum
237 have not been widely applied in channel classification studies to date to capture reach-scale
238 geomorphic heterogeneity, but are well known in field ecology (Johnson 1980; Miller and
239 Ambrose 2000; Manly and Alberto 2014; CHaMP 2016) and hydrology (Thomas and Lewis
240 1995; Yang and Woo 1999). Three landscape characteristics accounting for geologic structure,
241 sediment availability, and sediment transport capacity were obtained from GIS data and analyses
242 as described below and used to stratify the Sacramento Basin channel network into 15 subgroups
243 or strata of potentially distinct reach-scale geomorphic characteristics.

244 Geologic structure (i.e., tectonic uplift and lithology), derived from the overlying
245 physiographic provinces (Fenneman and Johnson 1946; CGS 2002) (Figure 1), was used in
246 conjunction with sediment availability and transport capacity to distinguish 15 geomorphic
247 strata. Sediment supply and transport capacity were represented using contributing area to a
248 reach (A_c) and the channel bed slope of a reach (S). These were obtained through analysis of the
249 National Hydrography Dataset (HUC 1802) (USGS 2013) in conjunction with a 10-m digital
250 elevation model (DEM) of the study area (USGS 2009). A_c is a common topographically-derived
251 surrogate for channel-forming discharge (e.g., Hack 1957; Schumm et al. 1984; Rosgen 1994)
252 and S is consistently used in classifications to characterize local flow energy dissipation (e.g.,
253 Rosgen 1994; Montgomery and Buffington 1997; Gartner et al. 2015). The combination of the
254 two variables is also prominent in hydrogeomorphic classification, as it is often conjectured that
255 channel bed morphology arises as a function of reach-scale shear stress and/or specific stream
256 power, which are determined by both unit discharge and channel slope (Flores et al. 2006).
257 Indices combining A_c and S as a measure of stream power (Lane 1957; Leopold and Wolman
258 1957; Sklar and Dietrich 1998) and have been used to distinguish braided from meandering
259 rivers (Carson 1984), to identify thresholds for channel incision (Schumm et al. 1984) and
260 sediment transport capacity (Bledsoe et al. 2002), and in reach-scale channel classification (e.g.,
261 Schmitt et al. 2007).

262

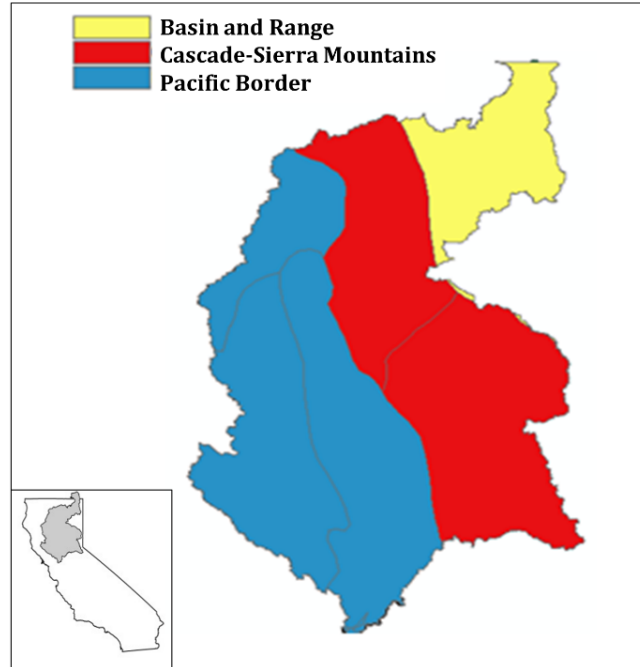
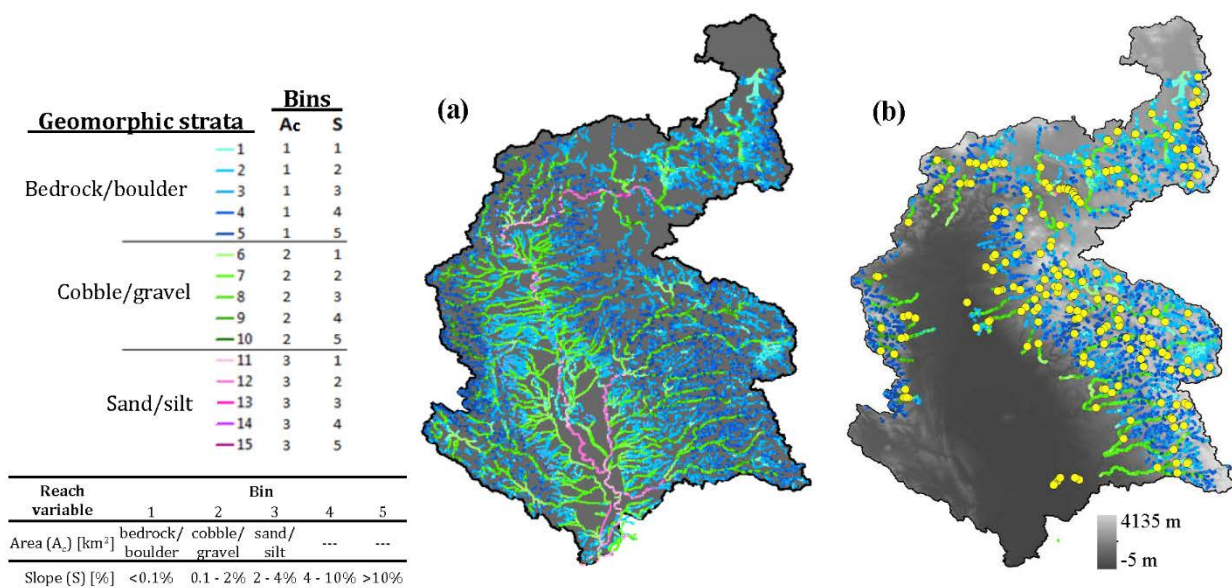


Figure 1. Sacramento Basin physiographic provinces used to refine contributing area (A_c) thresholds for channel network stratification.

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268 The channel network was derived from the 10-m DEM and dissected into equidistant
269 segments of 250 m length; S and A_c were subsequently derived from the DEM for each segment.
270 Within each physiographic province, channel segments were binned according to GIS-derived S
271 and A_c thresholds to aid with sampling – the results of the study are not sensitive to the exact
272 number of bins or thresholds between bins, as long as the procedure aids with sampling the
273 diversity in the system with equal effort. Five S bins were considered based on Rosgen's (1994)
274 channel classification thresholds for ease of comparison: $< 0.1\%$, $0.1 - 2\%$, $2 - 4\%$, $4 - 10\%$, and
275 $> 10\%$. Three A_c bins were established based on estimated A_c threshold transitions for prevalent
276 sediment sizes: (1) bedrock/boulder, (2) cobble/gravel, and (3) sand/silt. The A_c thresholds
277 assigned to distinguish channel bed composition classes were unique for each of the three
278 physiographic provinces within the Sacramento Basin. This decision was based on the expected
279 differences in A_c required to transition from boulder- to cobble- and from gravel- to sand-
280 dominated channels arising from large-scale differences in geology, topography, and climate
281 driving distinct sediment regimes. The physiographic provinces provide bounds on what
282 channels are potentially comparable in terms of relations between drainage area and discharge,

283 sediment supply, and substrate size (Montgomery and Buffington 1993). Within each province,
 284 A_c bin thresholds were estimated based on identified channel composition transition locations
 285 reported in available literature combined with expert knowledge relating A_c and sediment
 286 composition in the region (e.g., Montgomery and Buffington 1993; Gasparini et al. 2004) (Table
 287 1). Fifteen geomorphic strata were then distinguished as all possible combinations of
 288 topographically-derived A_c and S bins (Figure 2, top-left), and each stream segment in the
 289 channel network was assigned to a stratum based on its particular GIS-based A_c and S values
 290 (Figure 2a).



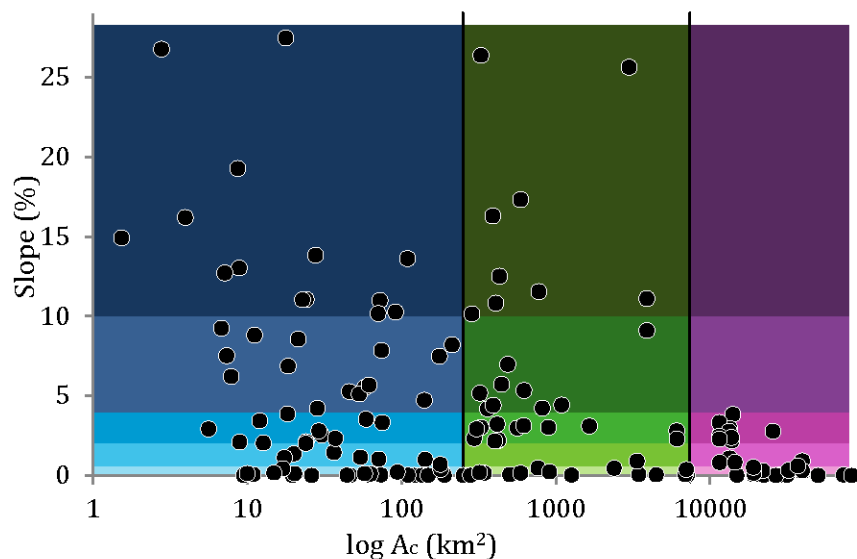
291
 292 Figure 2. Map of geomorphic strata (a) across the Sacramento Basin and (b) across the low-volume
 293 snowmelt and rain (LSR) reaches of the Sacramento Basin. Yellow dots indicate the randomly chosen
 294 field survey locations across the 15 strata. The geomorphic strata are defined in the top-left table based on
 295 the combination of contributing area (A_c) and slope (S) bins, which are derived based on thresholds stated
 296 in the bottom-left table and Table 1.
 297

298 Table 1. Contributing area (A_c) thresholds for channel composition across Sacramento Basin
 299 physiographic provinces (see Figure 1 for map of physiographic provinces).

Physiographic Province	Contributing Area Threshold (km^2)	
	Bedrock/boulder to cobble/gravel	cobble/gravel to sand/silt
Pacific Border	50	5,000
Cascade-Sierra Mountains	300	9,000
Basin and Range	300	10,000

300

301 Of the 15 geomorphic strata distinguished across the Sacramento Basin by A_c and S
302 combinations, 13 strata were exhibited by LSR reaches, indicating that LSR-dominated
303 hydrologic regimes were 87% representative of the full range of geomorphic variability in the
304 Sacramento Basin as expressed with binned combinations of A_c and S . The two geomorphic
305 strata not found within LSR reaches consisted of the combinations of the highest A_c bin and 4-
306 10% or >10% slope bins. Based on reach accessibility and expected variability of geomorphic
307 attributes, 10 to 12 field surveys were performed within each of the 13 geomorphic strata
308 exhibited by LSR reaches for a total of 161 field survey reaches representing a large range of A_c
309 – S combinations (Figure 3). Note that DEM-derived S was not used further in this study, as it is
310 not highly accurate at representing reach-scale channel slope.



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312 Figure 3. The stratified random field survey locations ($n=161$) represent a large range of GIS-based
313 reach slopes (S) and contributing areas (A_c). Colors and shading indicate the distinct S and A_c bins that
314 correspond to the geomorphic strata listed in Fig. 3 based on the Cascade – Sierra Mountains
315 physiographic province A_c thresholds in Table 1.

316

317 *Data-driven geomorphic channel classification*

318 *Field surveys.* Geomorphic field surveys were performed for each study reach identified
319 through the stratified random sampling scheme described above. Surveys of 64 reaches were
320 conducted by the authors' crew and data from another 97 reaches were obtained from the Surface

321 Water Ambient Monitoring Program (SWAMP) of the California State Water Resources Control
322 Board. Both field campaigns used the same sampling protocols, outlined in Ode (2007) and
323 briefly summarized below. Depending on whether the average wetted channel width was less
324 than or greater than 10 m, a stream reach was surveyed over a length of 150 or 250 m,
325 respectively (Ode 2007), corresponding to 10 – 100 bankfull widths. Eleven evenly spaced cross-
326 sectional transects were surveyed along each stream reach to quantify variability in 22
327 geomorphic attributes listed in Table 2 (Ode 2007). These decisions were intended to balance
328 geomorphic (Grant et al. 1990; Montgomery and Buffington 1997) and ecological (Frissell et al.
329 1986) relevance with the practical time and resource limitations of field surveying. The choice of
330 reach length and transect spacing also enabled incorporation of the existing SWAMP
331 geomorphic dataset for the study region that uses the same values. Channel morphology and
332 reach characteristics for the 161 surveyed reaches were measured using a surveying level and
333 stadia rod (Topcon AT-B, 0.01m). Longitudinal streambed profiles were surveyed at consecutive
334 transects along the thalweg for the entire length of the reach. Wolman pebble counts (Wolman
335 1954) of 110 pebbles were performed at each reach such that ten pebbles were randomly selected
336 from each of eleven transects to balance sampling precision and effort across a range of sediment
337 material variability assuming normally distributed sediment size (Edwards and Glysson 1999;
338 Bunte and Abt 2001).

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Table 2. Reach-scale geomorphic and topographic variability attributes considered in channel classification

<i>Geomorphic Attribute</i>	<i>Code</i>	<i>Description</i>	<i>Units</i>
wetted depth	\bar{d}	average across 11 transects; 0 if dry channel	m
wetted width	\bar{w}	average across 11 transects; 0 if dry channel	m
wetted width-to-depth	$\frac{\bar{w}}{\bar{d}}$	ratio of channel width to depth	--
wetted depth-to-D50	$\frac{\bar{d}}{\bar{D}_{50}}$	low water roughness; channel depth standardized by median grain size	--
bankfull depth	\bar{d}_{BF}	average across 11 transects	m
bankfull width	\bar{w}_{BF}	average across 11 transects	m
bankfull width-to-depth	$\frac{\bar{w}_{BF}}{\bar{d}_{BF}}$	ratio of bankfull width to depth	--
bankfull depth-to-D50	$\frac{\bar{d}_{BF}}{\bar{D}_{50}}$	roughness; bankfull depth standardized by median grain size	--
entrenchment ratio	$\overline{e.ratio}$	floodprone width / average bankfull width; floodprone width manually estimated from high resolution aerial imagery (<1m)	--
shear stress	\overline{shear}	depth-slope product approximation	Pa
shields stress	$\overline{shields}$	non-dimensionalization of shear stress (Shields 1936)	--
contributing area	\bar{A}_c	drainage area to downstream end of reach	km ²
slope	\overline{slope}	average water surface slope over 11 transects	%
sinuosity	\overline{sin}	straightline distance/actual channel distance along ~2000m of channel	--
sediment distribution variance	CV_{sed}	variance of transect sediment distribution (n=10) across 11 transects	--
D50	\bar{D}_{50}	median grain size across reach (n=110)	mm
D84	\bar{D}_{84}	84th percentile grain size across reach (n=110)	mm
Dmax	\bar{D}_{max}	maximum grain size across reach (n=110)	mm
† wetted depth variance	CV_d	std/mean across 11 transects; 0 if no water in channel	--
† wetted width variance	CV_w	std/mean across 11 transects; 0 if no water in channel	--
† bankfull depth variance	$CV_{d,BF}$	std/mean across 11 transects	--
† bankfull width variance	$CV_{w,BF}$	std/mean across 11 transects	--

† topographic variability attributes (TVAs)

343

344 *Reach-scale geomorphic attributes.* Twenty-two geomorphic attributes (Table 2) were
 345 chosen to describe relevant, persistent reach-scale geomorphic characteristics that influence
 346 hydraulics and sediment dynamics and in turn aquatic and riparian ecosystem functioning
 347 (Birkeland 1996; Hupp and Osterkamp 1996; Merrit and Wohl 2003). The field-measured and
 348 computed attributes included traditional reach-averaged diagnostic variables [e.g., slope (\overline{slope}),
 349 contributing area (A_c), sinuosity (\overline{sin}), entrenchment ($\overline{e.ratio}$), shear stress (\overline{shear}), relative
 350 roughness ($\overline{d.D_{50}}$), sediment composition (i.e., \bar{D}_{50} , \bar{D}_{84} , and \bar{D}_{max}) and base flow and bankfull

351 depth (\bar{d}), width (\bar{w}), and width-to-depth ratio ($\overline{w \cdot d_{BF}}$] as well as four TVAs capturing within-
352 reach variability in base flow and bankfull channel width (CV_w) and bed elevation (CV_d) (Table
353 2).

354 Reach-scale estimates of geomorphic attributes were computed from field surveys by
355 averaging values across the eleven surveyed cross-sections within each reach. Entrenchment was
356 calculated as flood-prone width divided by bankfull width (Rosgen 1994), where flood-prone
357 width was measured manually from sub-meter resolution aerial imagery. Sinuosity was
358 calculated as the linear valley distance divided by the actual channel distance along 2 km of
359 channel straddling the field site (Elliott et al. 2009). The coefficient of variation (CV) of base
360 flow and bankfull width and depth was calculated among the eleven cross-sections of each
361 survey reach as a measure of within-reach variability. CV is a nondimensional measure of
362 standard deviation that provides a useful but not exclusive metric of variability (Schneider 1994)
363 that is commonly used in spatial analysis of ecological patterns (Rossi et al. 1992; Simonson et
364 al. 1994; Gubala et al. 1996; Palmer et al. 1997; Thoms 2006; Gostner et al. 2013a). A list of
365 geomorphic attributes considered and their methods of measurement or calculation is provided in
366 Table 2. When possible, these attributes were made non-dimensional for application in a range of
367 physiographic and climatic settings (Parker 1979; Parker et al. 2003). Given the dual aims of
368 adapting the Rosgen classification to incorporate TVAs and comparability with existing field
369 data for the study region, the present study omitted several potentially significant metrics [e.g.,
370 channel vegetation, bank material, dominant flow types (Raven et al. 1998), and stream power
371 (Knighton 1999; Orr et al. 2008)] that could be considered in future studies.

372 *Statistical analyses.* The geomorphic attributes (Table 2) were initially re-scaled to range
373 from 0 to 1 and examined for correlation to identify and remove highly correlated attributes
374 (Pearson's correlation coefficient > 0.8) to meet the assumption of lack of multicollinearity. Five
375 of the original 22 attributes were highly correlated (\bar{d} , \bar{w} , $\overline{d \cdot D_{50}}$, $\overline{D_{50}}$, CV_{sed}), reducing the dataset
376 to 17 geomorphic attributes (Table 2).

377 A hierarchical clustering analysis using Ward's algorithm (Ward 1963; Murtagh and
378 Legendre 2013) was used to examine the clustering structure of the uncorrelated, standardized
379 geomorphic attributes describing the 161 study reaches. The dataset also was analyzed by *k*-
380 means cluster analysis stipulating 2 to 15 (*k*) clusters that maximize the between-group variation
381 (Hartigan and Wong 1979; Kaufman and Rousseeuw 1990). Slope breaks in the *k*-means scree

382 plot of the within-group sum of squares for each clustering solution were interpreted as numbers
383 of clusters at which information content of the clustering process changed. Scree plot slope
384 breaks and the Davies-Bouldin internal clustering index (DBI=0.91) indicated that 12 clusters
385 created distinct groups of study reaches, similar to the hierarchical clustering results.

386 A combination of univariate and multivariate statistical methods was then applied to (i)
387 examine the strength of variables for distinguishing identified channel types, (ii) test the
388 hypothesis that channel types exhibit significantly different values of geomorphic attributes, (iii)
389 examine the potential range of values for variables of interest between channel types, and (iv)
390 validate the basis of the channel classification by predicting the channel type using geomorphic
391 attributes. These statistical methods included nonmetric multidimensional scaling (NMDS)
392 (Clarke 1993), one-way analysis of variance (ANOVA) with Tukey's honestly significant
393 differences (HSD) test, nonparametric permutational multivariate analysis of variance
394 (PerMANOVA) (Anderson 2001), and classification and regression trees (CART) (Breiman et
395 al. 1984; De'ath and Fabricius 2000).

396 An exploratory NMDS analysis (Clarke 1993; Oksanen 2011) of the surveyed reaches based
397 on the uncorrelated geomorphic attributes was performed to visually represent the structure of
398 the multivariate dataset and evaluate the relative significance and correlation of attributes.
399 NMDS is common in ecological studies, including those identifying differences in biological
400 communities based on geomorphic variables (e.g., Walters et al. 2003; Virtanen et al. 2010) and
401 is increasingly included in dedicated geomorphic studies (e.g., Merriam et al. 2011; Sutfin et al.
402 2014; Varanka et al. 2014; Jaeger 2015). Histograms of each geomorphic attribute were also
403 used to evaluate the density distributions of attribute values across the survey reaches and lend
404 insight into the multivariate clustering structure.

405 Individual one-way ANOVAs were conducted to compare geomorphic attribute means
406 between channel types. A post-hoc Tukey's HSD test at the 95% confidence level indicated the
407 best attributes for distinguishing between channel types. A PerMANOVA analysis (Anderson
408 2001) [Euclidean distance, 9999 permutations (Oksanen 2011)] was performed to test the
409 hypothesis that the channel types distinguished through clustering analysis exhibit significant
410 differences ($p < 0.01$) in geomorphic attributes.

411 Toward the primary goal of the study, CART (Breiman et al. 1984) was then used to identify
412 the most explanatory geomorphic attributes distinguishing channel types and their threshold

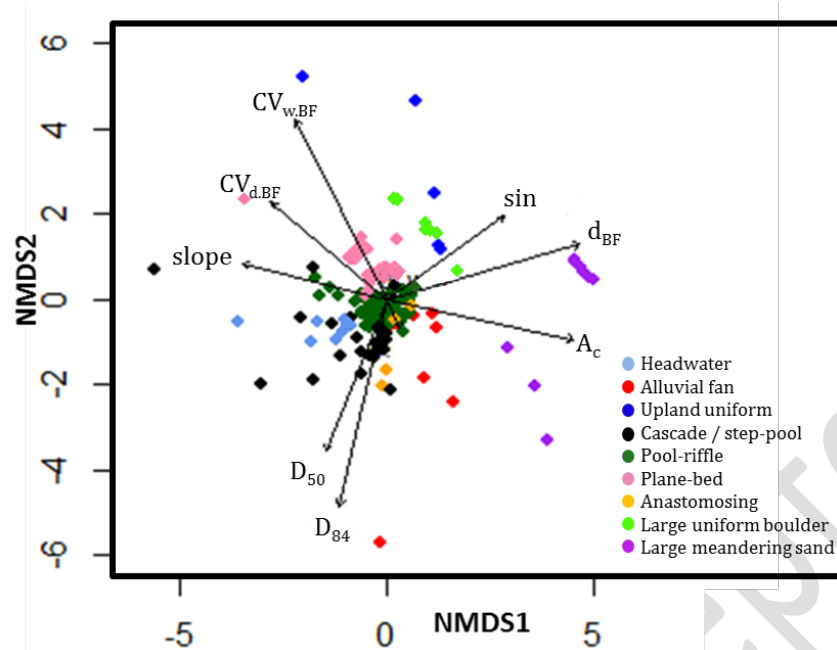
413 values. CART yields a binary decision tree where the response variable (study reach) is
414 partitioned into groups (channel types) with minimized within-group variance (based on ten-fold
415 cross-validation, Therneau et al. 2010) and increasing purity (based on the Gini index, De'ath
416 and Fabricus 2000).

417 *Heuristic refinement of inductive clustering solution.* The final number of clusters
418 distinguished was determined heuristically based on a combination of statistical analysis
419 interpretation and physical understanding of the region. First, potential splitting solutions were
420 identified based on the structure of the hierarchical clustering and the shape of the scree-plots
421 from the non-hierarchical k -means clustering. Each potential splitting solution was assessed
422 iteratively from largest to smallest splitting distance (based on Ward's hierarchical clustering).
423 Heuristic (dis)aggregation of clusters was subsequently performed based on the physical
424 distinction and interpretability of the resulting clusters with the objective of minimizing the final
425 number of physically interpretable channel types. For instance, if a particular splitting solution
426 distinguished only some empirical clusters to a level of reasonable physical interpretability, the
427 remaining clusters would be iteratively disaggregated based on the next potential splitting
428 solutions until the minimal number of physically meaningful clusters was identified.

429 **Results**

430 *Relative significance of geomorphic attributes*

431 The two-dimensional NMDS ordination illustrated the significance of TVAs and the relative
432 roles of geomorphic attributes in structuring the multivariate dataset. The NMDS minimized
433 mean stress at 0.08 for 161 study reaches (Figure 4); stress values of < 0.1 are considered to be a
434 good ordination with little risk of drawing false inferences (McCune and Grace 2002). NMDS
435 indicated that the first axis (NMDS1) is dominated by $CV_{d.BF}$, $CV_{w.BF}$, \overline{slope} , and A_c , while the
436 second axis (NMDS2) is dominated by cross-sectional geomorphic attributes (e.g., \overline{D}_{84} , \overline{D}_{50} ,
437 $\overline{d_{BF}}$, \overline{D}_{50} , \overline{w} , $\overline{d_{BF}}$) as well as $CV_{w.BF}$. As these axes represent gradients of maximum variation,
438 dominant attributes on each axis control the structure of the multivariate dataset.

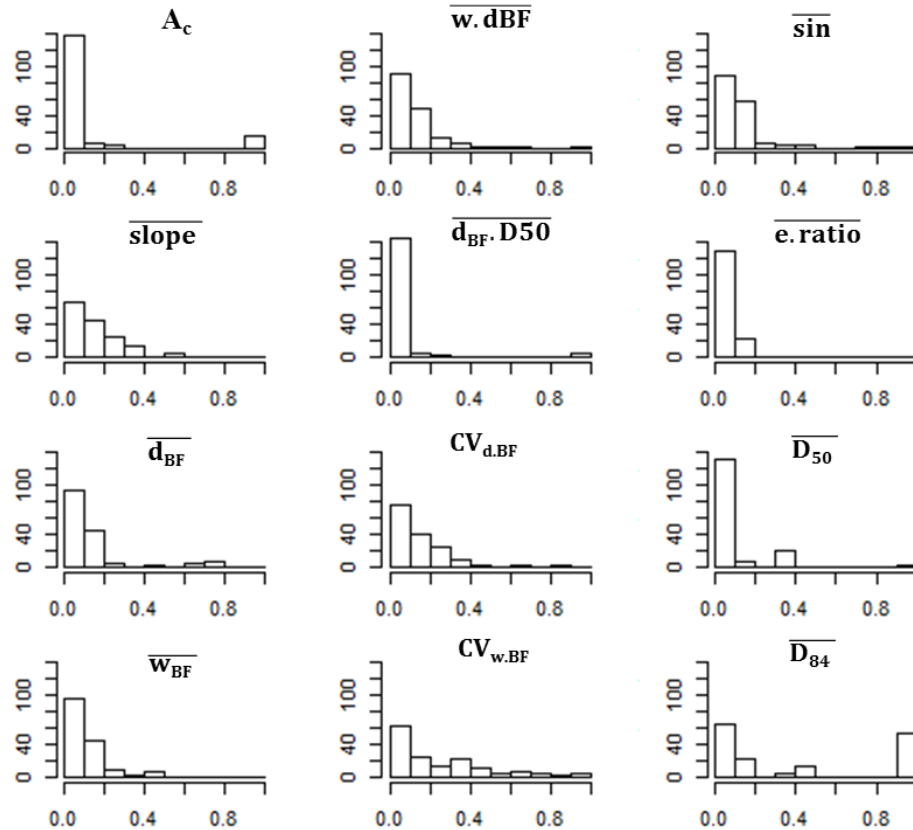


439

440 Figure 4. Nonmetric dimensional scaling (NMDS) for the first two axes with the eight most
 441 significant geomorphic attributes shown. Vectors of attributes are plotted based on the strength of their
 442 correlation to the axis (e.g. longer vectors are more strongly correlated to an axis).
 443

444

445 Histograms of rescaled geomorphic attributes lend insight into how the density distributions
 446 of geomorphic attribute values control the multivariate data structure (Figure 5). If an attribute is
 447 normally distributed with a predominance of its values within a narrow band of its full range for
 448 most study reaches, then that attribute will likely yield a single grouping, so it cannot explain
 449 differences between those reaches; it may instead distinguish the few statistical outlier reaches.
 450 In contrast, an attribute with a more uniform distribution will tend to produce more, equally
 451 weighted groupings and thus be a dominant factor explaining differences among many reaches.
 452 Upon visual assessment of the geomorphic attribute distributions, most attributes exhibited
 453 highly skewed distributions towards lower values (e.g., \overline{sin} , $\overline{e.ratio}$, and $\overline{w_{BF}}$). In contrast, the
 454 TVAs ($CV_{d.BF}$ and $CV_{w.BF}$) and \overline{slope} exhibited more uniform distributions, helping to explain
 their dominant roles in structuring the multivariate dataset.



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Figure 5. Histograms of geomorphic attributes (re-scaled from 0 to 1) across the 161 study reaches illustrate the distribution of each attribute. In contrast to the exponential distributions exhibited by most attributes, the TVAs ($CV_{d.BF}$ and $CV_{w.BF}$) and slope exhibit more uniform distributions.

459

Distinguishing channel types

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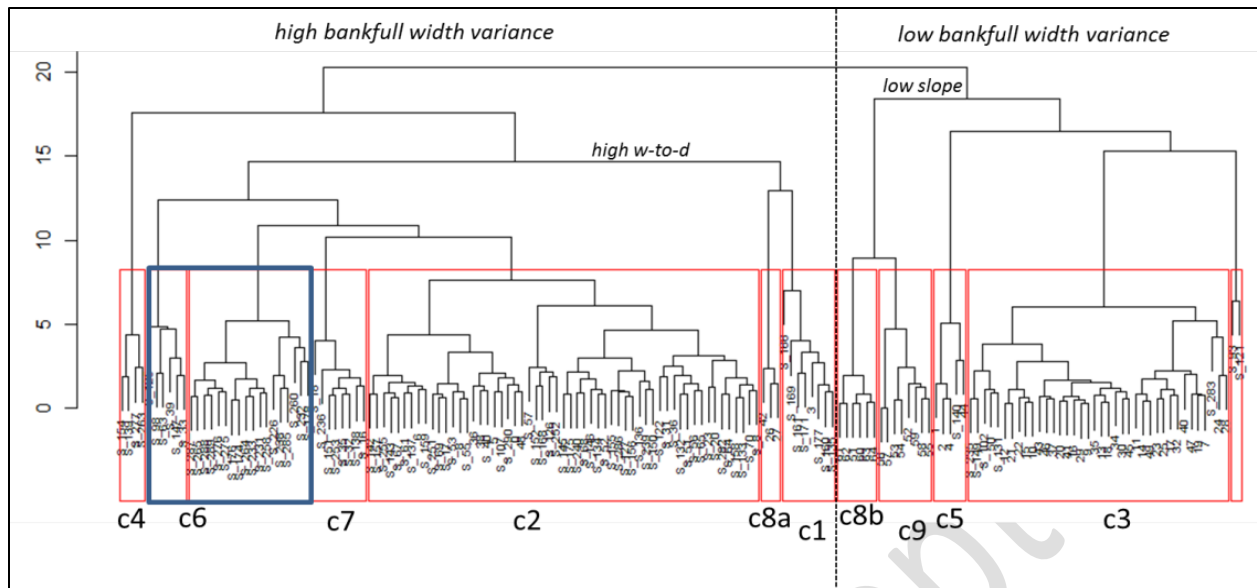
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Agglomerative hierarchical clustering with Ward's linkage (Ward 1963; Murtagh and Legendre 2013) illustrated the clustering structure of the 161 study reaches across the re-scaled uncorrelated geomorphic attributes (Figure 6). The first split occurs at a distance of 20, distinguishing reaches of high ($\sim 0.2 - 1.7$) and low ($\sim 0 - 0.2$) bankfull width variance. Splitting groups at a distance of eight distinguished 12 groups that were then reduced to nine physically meaningful groups by applying the heuristic clustering refinement procedures explained in Section 2.3.4. The nine resulting groups represented physically distinct channel types containing between 4 and 57 study reaches each (average of 18 reaches).



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Figure 6. Hierarchical clustering of study reaches using Ward's method showing twelve distinct groups (boxed in red) representing 9 physically distinct channel types following heuristic refinement.

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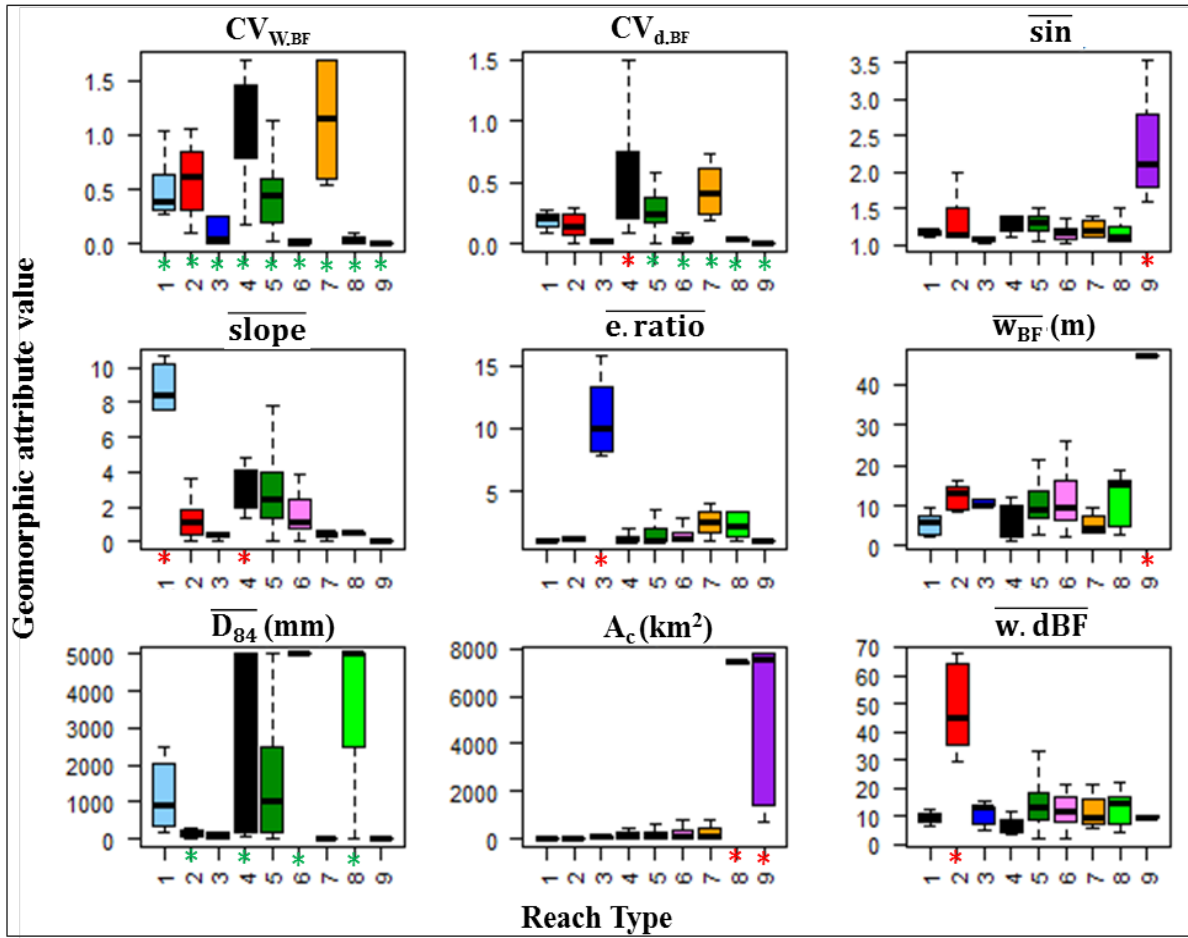
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Individual one-way ANOVA results indicated that group means of 12 of 17 geomorphic attributes varied significantly between the nine channel types ($p < 0.05$) (all attributes except \bar{w} , \bar{d} , \bar{D}_{50} , \bar{D}_{max} , and $\overline{shields}$) (Table 3). Multiple comparisons of group means of each attribute using Tukey's HSD post-hoc test at the 95% confidence level indicated particularly significant channel types for specific attributes (Figure 7). For example, $\overline{w} \cdot \bar{d}_{BF}$ is significantly higher for type 2 reaches than all other channel types. Conversely, $CV_{w,BF}$ differs significantly between channel types 4 and 7 and channel types 6, 8, and 9 while there is no significant difference in the attribute within those groups. Box-and-whisker plots illustrate relative differences in geomorphic attributes within and across the nine identified channel types (Figure 7). Finally, a map of the spatial distribution of classified channel types across LSR-dominated reaches in the Sacramento Basin is provided in Figure 8.



* indicates significantly different from 1+ other reach types (95% confidence level)

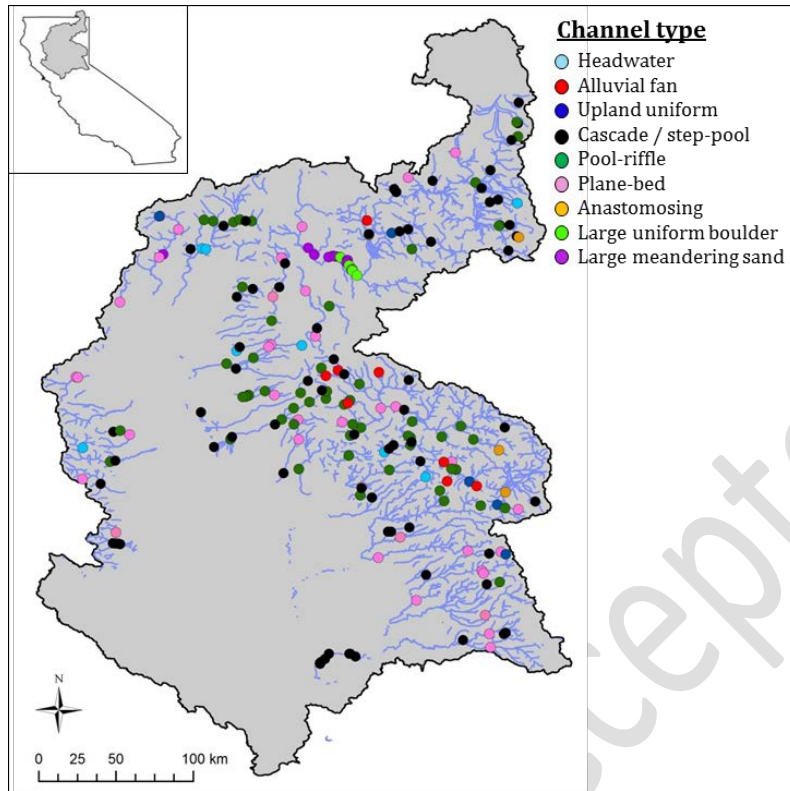
* indicates significantly different from all other reach types based on Tukey's HSD test (95% confidence level)

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Figure 7. . Box-and-whisker plots of geomorphic and topographic variability attributes across the nine identified channel types.



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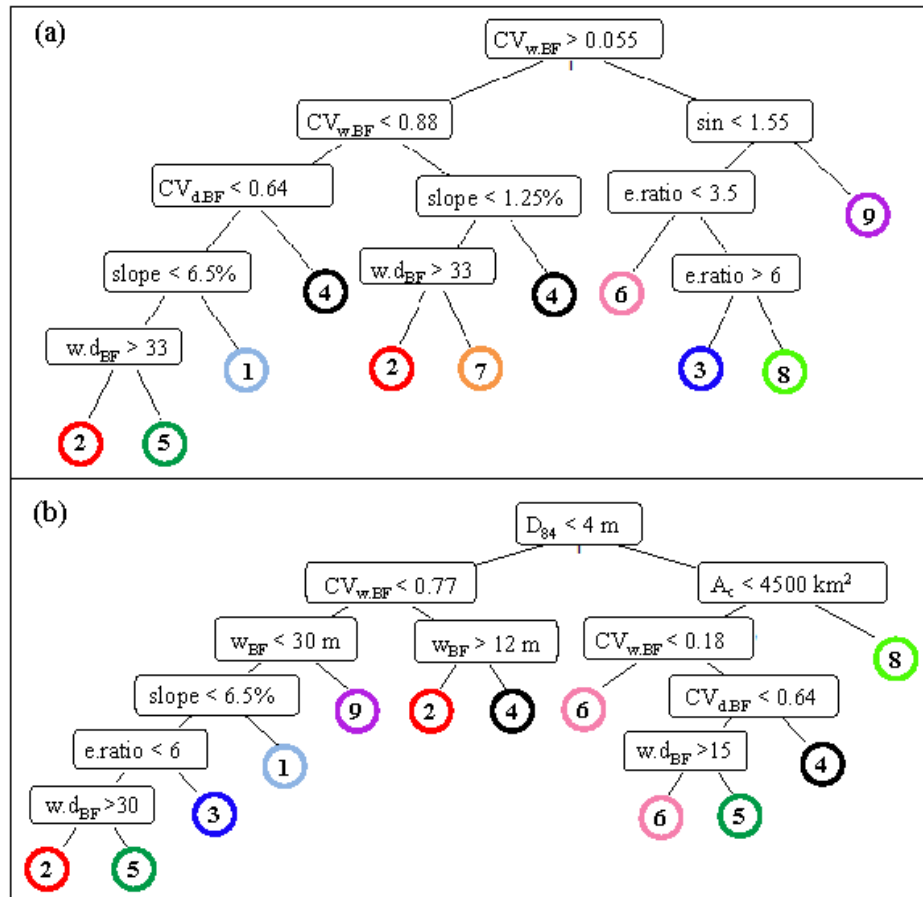
488 Figure 8. Map of the spatial distribution of field sites in the hydrological regime investigated and
 489 their classified channel types across LSR reaches (light blue lines) of the Sacramento Basin.

490

491 Table 3. ANOVA results show that mean geomorphic attribute values differ between the nine channel
 492 types. Statistically significant attributes ($p < 0.05$) are indicated in bold.

Geomorphic attribute	Mean Square	F	p-value
$\overline{A_c}$	334.59	106.28	0.00
$\overline{d_{BF} D_{50}}$	121.09	26.96	0.00
$CV_{w,BF}$	0.25	19.90	0.00
\overline{slope}	37.06	18.63	0.00
$\overline{w.d_{BF}}$	76.26	15.98	0.00
$CV_{d,BF}$	0.24	15.90	0.00
$\overline{d_{BF}}$	59.50	12.20	0.00
$\overline{e.ratio}$	20.43	10.27	0.00
$\overline{w_{BF}}$	42.36	8.50	0.00
\overline{sin}	28.36	5.59	0.02
$\overline{D_{84}}$	9.86	4.96	0.03
\overline{shear}	9.28	4.66	0.03
Dmax	17.66	3.43	0.07
shields	0.74	0.14	0.71

493 Multivariate analyses revealed that the data-driven channel types identified exhibit
494 significantly different geomorphic settings and identified the geomorphic attribute ranges across
495 each channel type in the study basin. PerMANOVA results indicated that multivariate mean
496 geomorphic setting is not equal for all nine channel types ($p=0.0001$; F-statistic=13), allowing
497 for the rejection of the null hypothesis that channel types were identical. The CART analysis
498 identified the most explanatory geomorphic attributes distinguishing channel types and their
499 threshold values, providing potential ranges of attribute values expected for each channel type
500 (Figure 9). The classification tree model determined the relative strength of non-dimensional
501 variables to be as follows: $CV_{w,BF}$, \overline{sin} , \overline{slope} , $\overline{e.ratio}$, $CV_{d,BF}$, $\overline{w.d_{BF}}$. This indicates that two of
502 the six explanatory attributes identified by the model were TVAs (i.e., $CV_{w,BF}$, $CV_{d,BF}$), while
503 slope played a lesser role. The non-dimensional classification tree correctly classified 85% of
504 survey reaches based on their reach-averaged geomorphic attribute values (Figure 9a).
505 Alternatively, 93% of reaches could be correctly classified by the classification tree considering
506 all attributes (Figure 9b). When both dimensional and non-dimensional attributes were
507 considered ($n=17$, Table 2), \overline{D}_{84} , A_c , and $\overline{w_{BF}}$ emerged as additional significant attributes for
508 distinguishing channel types. Separate classification tree models using only the author's field
509 sites ($n=64$) and using both the author's and SWAMP field sites ($n=161$) both identified $CV_{w,BF}$,
510 \overline{sin} , and \overline{slope} as the three primary attributes distinguishing channel types, emphasizing their
511 persistent significance independent of individual field sites. Furthermore, $CV_{w,BF}$ emerged as a
512 dominant attribute above traditional Rosgen (1994) geomorphic attributes in both models.



513

514 Figure 9. CART classification trees considering (a) non-dimensional and (b) all geomorphic
 515 attributes, indicating primary attributes and their threshold values distinguishing channel types.

516 Geomorphic and topographic variability attributes are defined in Table 2 and circled numbers refer to
 517 channel types as defined in Table 3.

518 *Physical interpretation of channel types*

519 Physical interpretation of the above statistical analyses (summarized in Table 4) was used in
 520 combination with expert evaluation and existing channel classification literature to name the nine
 521 channel types based on their valley setting and distinguishing channel attributes (this
 522 nomenclature is used for the remainder of this study): 1. *confined headwater small boulder*
 523 *cascade*, 2. *partly-confined expansion pool - wide bar*, 3. *unconfined upland plateau large*
 524 *uniform*, 4. *confined cascade/step-pool*, 5. *partly-confined pool-riffle*, 6. *partly-confined large*
 525 *uniform*, 7. *unconfined anastomosing plateau small pool-riffle*, 8. *unconfined large uniform*
 526 *boulder*, and 9. *unconfined large meandering sand* (Figure 10, Table 4).

1. CONFINED HEADWATER
SMALL BOULDER-CASCADE



2. PARTLY-CONFINED EXPANSION
POOL - WIDE BAR



3. UNCONFINED UPLAND PLATEAU
LARGE UNIFORM



4. CONFINED
CASCADE/STEP-POOL



5. PARTLY-CONFINED
POOL-RIFFLE



6. PARTLY-CONFINED
LARGE UNIFORM



7. UNCONFINED
ANASTOMOSING PLATEAU
SMALL POOL - RIFFLE



8. UNCONFINED
LARGE UNIFORM BOULDER



9. UNCONFINED
LARGE MEAN DERING SAND BED



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Figure 10. Example images of channel types distinguished by the classification from field and Google Earth© imagery.

Corrected, 2021

Table 4. Descriptive name, literature analogs, key channel form characteristics, and physical process interpretation of identified channel types are provided.

Channel number	Descriptive name		Literature analog	Morphological characterization				Physical process interpretation
	Valley setting	Channel type		Channel slope and planform	Cross-sectional attributes	Bed material	TVAs	
1	Confined headwater	Small boulder-cascade‡	Type A†; Bedrock or Cascade‡; Steep headwater*	Very steep, straight	Low w-to-d, highly entrenched	Poorly sorted boulder-dominated	Moderate w and d variance	High stream power combined with variable topography drive high sediment transport and high subreach-scale variability in scour and fill (Powell et al. 2005)
2	Partly-confined expansion	Pool-wide bar	Moderate gradient alluvial fan channel (Paustian et al. 1992)	Variable slope, high sinuosity	Wide and shallow, entrenched	Poorly sorted pebble- to cobble-sized	High w variance, moderate d variance; + covariance	Lateral flow divergence drives rapid deposition of unsorted alluvial sediment (Paustian 1992)
3	Unconfined upland plateau	Large uniform		Low slope, straight	Large channel dimensions, low entrenchment	Homogenous pebble- to cobble-sized	Low variability	Low energy depositional valley; uniform topography drives sediment transport as uniform sheet (Miller and Burnett 2008)
4	Confined	Cascade / Step-pool	Cascade or Step-pool‡; Type G1-2 or A†; Gorge*	Steep, moderate sinuosity	Low w-to-d, entrenched	Boulder-dominated	Extremely high variability; - covariance	High topographic variability drives complex subreach-scale flow resistance dynamics and generates turbulence (Wohl and Thompson 2000; Wilcox and Wohl 2006)
5	Partly-confined	Pool - riffle	Pool-Riffle ‡; Type C†	Mid- to high- slope, moderate sinuosity	Moderate w-to-d, moderate entrenchment	Gravel- to cobble-sized	High d variance, moderate w variance; + covariance	Channel constraint by valley and floodplain topographic controls drives localized vertical and lateral flow convergence; convective accelerations reinforce non uniform flow convergence (Dietrich and Smith 1983)
6	Partly-confined	Large uniform	Plane-bed ‡; Type B†	Mid slope, straight	Moderate w-to-d, moderate entrenchment	Cobble- to boulder-sized	Low variability	No topographic steering controls on deposition or erosion; uniform topography drives sediment transport as uniform sheet (Lane and Carlson 1953; Miller and Burnett 2008)
7	Unconfined anastomosing plateau	Small pool-riffle	Type DA†; Type E†	Low slope, moderately sinuous	Small channel dimensions, low entrenchment	Poorly sorted pebble- to cobble-sized	High variability; +/- covariance	Anastomosing channels formed by avulsion driven by rapid channel aggradation (Makaske 2001). High channel depth variance and poorly sorted sediment may be indicative of rapid, heterogeneous channel deposition triggering avulsion
8	Unconfined	Large uniform boulder		Low slope, moderate sinuosity	Large channel dimensions, entrenched	Cobble- to boulder-sized	Low variability	Underlying bedrock geology constrains formation of meander bends and determines sediment supply and composition
9	Unconfined	Large meandering sand bed	Labile channel (Church 2006); Meandering sand (Lane 1957)	Low slope, high sinuosity	Very large channel dimensions, highly entrenched	Sand	Low variability	Meanders maintained by secondary transverse flow cells that drive sediment routing through inside of bends (Thompson 1986); "live bed" sediment transport (Henderson 1963)

† Rosgen (1994)

‡ Montgomery and Buffington (1997)

*Brierley and Fryirs (2005)

533 The order of the identified channel types represents an idealized upstream to downstream
534 progression in the landscape from montane to lowland streams, however some channel types are
535 less predictable along such a progression (e.g., *partly-confined expansion pool - wide bar*,
536 *unconfined upland plateau large uniform*). Four of the identified channel types (i.e., 2, 3, 6, and
537 8) were not commonly identified by previous classifications. The geomorphic characteristics of
538 each channel type are described below, organized and interpreted with respect to presumed
539 dominant channel processes and related to TVAs where applicable.

540 The *confined headwater small boulder-cascade* channel type (1) (*sensu* Sullivan 1986;
541 Montgomery and Buffington 1997; Hassan et al. 2005) is characterized by the highest slopes and
542 lowest A_c of any channel type. These channels exhibit high entrenchment, low width-to-depth,
543 low sinuosity, and a boulder-dominated bed. High stream power combined with variable
544 topography drive high sediment transport and high subreach-scale variability in scour and fill
545 (Powell et al. 2005) indicated by high $CV_{d,BF}$. The *confined cascade/step-pool* channel type (4) is
546 distinguished from the *boulder - cascade* by slightly lower slopes and larger A_c , as well as
547 slightly increased channel dimensions and a reduction in $\overline{w \cdot d_{BF}}$ and dominant sediment size.
548 These changes are indicative of a downstream progression from hillslope- to channel-dominated
549 processes. *Cascade/step-pool* channels are also characterized by the highest $CV_{d,BF}$ and $CV_{w,BF}$ of
550 any channel type and generally negatively covarying bed and width undulations, indicating
551 complex subreach-scale flow resistance dynamics. Flow resistance in these channels is
552 hypothesized to be generated by the form drag of constricting step-forming roughness features
553 and by tumbling flow regimes in which critical or supercritical flow over narrow step crests
554 plunges into wider pools, abruptly decreasing velocity and generating substantial turbulence
555 (Peterson and Mohanty 1960; Montgomery and Buffington 1997; Wohl and Thompson 2000;
556 Wilcox and Wohl 2006; Wyrick and Pasternack 2008).

557 The *partly-confined pool-riffle* channel type (5) exhibits the next highest slopes and shear
558 stress and slightly larger A_c than the *cascade/step-pool* channel. *Pool-riffle* channels are
559 constrained by valley and floodplain topographic controls and characterized by positively
560 covarying bed and width undulations that generate subreach-scale width and depth constrictions
561 and expansions (indicated by high $CV_{w,BF}$ and $CV_{d,BF}$) which drive localized flow convergence.
562 Topographically-driven convective accelerations have been shown to reinforce these nonuniform
563 convergent and divergent flow patterns, and thus pool-riffle morphogenesis (Dietrich and Smith

1983; Dietrich and Whiting 1989; Nelson and Smith 1989). The *pool-riffle* channel type is morphologically similar in many regards to the *partly-confined large uniform* channel type (6) except for significantly higher topographic variability and smaller sediment composition. This is interpreted as a difference in sediment transport mechanisms. In *pool-riffle* channels, topographic variability has been shown to control sediment transport through mechanisms such as topographic steering (Whiting and Dietrich 1991; MacWilliams et al. 2006), flow convergence (MacWilliams et al. 2006; Sawyer et al. 2010), and recirculating eddies (Lisle 1986; Rathburn and Wohl 2003; Woodsmith and Hassan 2005; Thompson and Wohl 2009). Alternatively, in *large uniform* channels largely devoid of any organized or rhythmic bedforms, at the time of transport the whole bed is expected to move as a conveyor belt (Lane and Carlson 1953; Montgomery and Buffington 1997). As there are no topographic steering controls on where deposition or erosion takes place in *large uniform* channels, the presumed result is maintenance of uniform width and depth with energy dissipation dominated by grain and bank roughness (Montgomery and Buffington 1997). The well-armored bed indicated by the large \bar{D}_{50} and \bar{D}_{84} suggest relative channel stability and a supply limited sediment transport regime (Dietrich et al. 1989).

Partly-confined expansion pool - wide bar channels (2) generally occur at abrupt valley widenings and exhibit very high $\overline{w \cdot d_{BF}}$ and heterogeneous sediment composition (CV_{sed}). Alluvial fans develop by the accumulation of sediment where a channel exits an upland drainage area (Drew 1873). These lower-gradient Type 2 channels running through alluvial fan style valley expansions likely have limited transport capacity due to reduced stream power and lateral flow divergence, driving rapid deposition of unsorted alluvial sediment (Paustian et al. 1992). These channels are distinguished by pool- wide bar morphology in which positively covarying bed and width variability combine with mobile sediment and limited lateral confinement to generate extremely wide, entrenched bars between constricted troughs.

The *unconfined upland plateau large uniform* channel type (3) exhibits very low entrenchment due to moderate-sized channels bordered by vast floodplains. The laterally unconfined upland plateau valleys through which these channels run are low-energy (low slope and A_c) depositional environments in which sediment supply is presumed to exceed transport capacity (Nagel et al. 2014). The uniform topography, low sinuosity, and homogenous sediment composition are indicative of uniform geomorphic processes [e.g., sediment transport as a

595 uniform sheet (Miller and Burnett 2008)]. The *unconfined anastomosing plateau small pool -*
596 *riffle* channel type (7), also characterized by low entrenchment and a laterally unconfined valley
597 setting, is distinguished from the large uniform channel type by much smaller channel
598 dimensions and higher topographic variability and sinuosity. Similar to *partly-confined pool-*
599 *riffle* channels, these channels are expected to maintain nonuniform morphology through
600 nonuniform mechanisms such as topographic steering, flow convergence, and eddy recirculation.
601 At the valley scale, these channels appear to connect to create multi-thread channels that diverge
602 and converge around vegetated, rarely inundated islands cut from the floodplain (Knighton and
603 Nanson 1993). The high channel depth variability that distinguishes this channel type from the
604 *upland valley uniform* channel may be indicative of past avulsion triggered by rapid,
605 heterogeneous channel deposition (Makaske 2001).

606 Finally, *unconfined large uniform boulder* (8) and *large meandering sand bed* channels (9)
607 are characterized by very large A_c , large channel dimensions, low slopes, high sinuosity, and
608 very low width and depth variability. *Large uniform boulder bed* channels are distinguished by
609 boulder-dominated beds and lower bankfull depths, while the *large meandering sand bed*
610 channels are sand-dominated and exhibit extremely high sinuosity and entrenchment typical of
611 meandering morphologies (Hickin 1974). These differences likely indicate a difference in
612 underlying geology and sediment supply constraining the formation of meanders by lateral
613 migration and influencing channel bed composition. The *large meandering sand* channel type
614 distinguished in this study appears similar to the meandering sand bed channel described by Lane
615 (1957) and the labile channel distinguished by Church (2006). Meanders are hypothesized to be
616 maintained primarily by the alternating converging and diverging secondary transverse flow cells
617 in and between bends, respectively, which help to maintain sediment routing through the inside
618 of meander bends (Thompson 1986). Mobile bedforms provide the primary hydraulic resistance
619 in these channels (Kennedy 1975), driving “live bed” sediment transport (Henderson 1963).

620 **Discussion**

621 *Lessons learned from channel classification modifications*

622 *Channel network stratification.* The initial GIS-based stratification of the channel network
623 based on catchment DEM-derived S and A_c proved effective at distinguishing underrepresented

624 geomorphic settings in the landscape that would likely otherwise have been overlooked. While
625 some channel types (e.g., *pool-riffle*, *plane-bed*, *cascade/step-pool*) spanned many S- A_c bins,
626 indicating their limited dependence on S or A_c , others were almost exclusively found in one bin
627 (e.g., *pool - wide bar*, *large uniform boulder*, *large meandering sand*). Bins with the largest
628 representation across the landscape unsurprisingly captured the largest number of channel types.
629 Bins 2, 3, and 4 (Figure 2) represented 28, 16, and 20% of the channel network in the study
630 domain and contained 7, 6, and 5 channel types, respectively, compared with 3 channel types per
631 bin on average. Geomorphic bins 1 – 5 with the smallest A_c accounted for 78% of LSR-
632 dominated reaches in the Sacramento Basin while bins 11 – 13 with the largest A_c accounted for
633 less than one percent of the study domain combined. However, field sites classified as *large*
634 *uniform boulder* and *large meandering sand* channels fell almost exclusively in bins 11 – 13,
635 emphasizing the value of stratified sampling for revealing naturally underrepresented channel
636 types. Slope bins were more evenly distributed, but very low (<0.1%) and very high (>10%)
637 slopes each accounted for less than 10% of the study domain. The identification of low slope
638 dominated channel types by the classification (e.g., *anastomosing*, *large uniform boulder*, and
639 *large meandering sand*) highlights the value of stratified sampling as these channel types would
640 likely not have been sampled sufficiently to distinguish distinct classes in a uniform random
641 sampling scheme given their limited representation in the basin.

642 The stratified sampling scheme enabled a large proportion of the full range of geomorphic
643 variability present in the study domain to be captured by the field sites. For example, bankfull
644 channel width across all surveyed sites ranged from 1.1 to 98.8 m. The smallest and largest
645 channels evident in the system from visual inspection are 0.8 and 100 m, respectively, indicating
646 that the sampling scheme captured 98% of the total range of bankfull widths. Similarly, the
647 sampling scheme captured 78% of the total range of A_c and 65% of the total range of S. The
648 maximum A_c for a surveyed site was 7,760 km² while the maximum A_c of any reach in the LSR
649 channel network was closer to 10,000 km². The maximum surveyed S of 14.3% was
650 substantially less than the estimated 22% maximum reach S. Overall, these results indicate that,
651 while not entirely representative, stratifying field data collection by GIS-based landscape
652 characteristics accounting for geologic structure, sediment availability, and sediment transport
653 capacity enabled the resulting field sites to capture a large range of geomorphic variability.
654 Splitting the channel network into further bins with more refined A_c and S requirements could

655 increase the proportion of the total range of geomorphic variability captured by field surveys.
656 Alternatively, stratifying the network across other GIS-based characteristics such as bankfull
657 width or adjusting the A_c and S thresholds for bin membership could potentially improve results.

658 *Heuristic refinement of classification results.* The nine channel types identified in this study
659 capture a diverse range of reach-scale geomorphic settings including channel types previously
660 identified by existing channel typologies and new, thus far unidentified, channel types. These
661 findings emphasize the value of the *a posteriori* heuristic refinement of inductive classification
662 results by suggesting that the resulting channel types retain a physical basis (deductive
663 component) but are capable of capturing the unique context of the landscape under study
664 (inductive component).

665 Identified channel types with strong analogs in the classification literature highlight the
666 physical basis of the classification results achieved after heuristic classification refinement. For
667 example, cascade channels as defined by Montgomery and Buffington (1997) generally occur on
668 steep slopes, are narrowly confined by valley walls, and are characterized by longitudinally and
669 laterally disorganized bed material typically consisting of cobbles and boulders. This channel
670 type corresponds strongly to our identified *confined cascade/step-pool* channel, characterized by
671 valley-confined channels with steep slopes, low width-to-depth, high bankfull width and depth
672 variance, and cobble/boulder dominated sediment. Montgomery and Buffington (1993)'s plane-
673 bed channel type refers to mid-slope planar gravel- and cobble- bed channels generally lacking
674 discrete bars or in-channel features. This channel type is similar to our *partly-confined large*
675 *uniform* channel, characterized by a moderate slope, cobble-dominated bed, and very low
676 bankfull width and depth variance (indicating absence of bars and planar longitudinal
677 morphology).

678 Some identified channel types have no analog in the Montgomery and Buffington
679 classification designed for the mountains of the Pacific Northwest of the US, particularly those
680 channel types associated with non-mountain environments. In these cases (e.g., *unconfined*
681 *anastomosing plateau small pool-riffle*), the more descriptive Rosgen (1994) channel types may
682 provide a better analog (Table 4). Alternatively, the *large meandering sand bed* (9) channel type,
683 while not present in the Montgomery and Buffington (1993) or Rosgen (1994) channel
684 classifications, has been distinguished in numerous other channel classification frameworks (e.g.,
685 Lane 1957; Schumm 1963; Church 2006). The *partly-confined expansion pool – wide bar*

686 channel type seems to only have an analog in the moderate gradient alluvial fan channel as
687 described by Paustian et al. (1992). This similarity of our results with the process-based channel
688 types distinguished by Paustian et al. (1992) indicates that the classification framework as
689 applied in this study is similarly capable of revealing distinct associations between channel
690 morphology and processes.

691 Channel types with no clear analog in the literature were also identified (e.g., *unconfined*
692 *upland plateau large uniform*, *unconfined large uniform boulder*), suggesting that the addition of
693 TVAs to the classification framework combined with channel network stratification and heuristic
694 refinement enabled the resulting channel classification to reveal the unique context of the
695 landscape under study. For instance, *upland plateau large uniform* channels were distinguished
696 from *anastomosing plateau small pool-riffle* channels primarily on the basis of topographic
697 variability. Distinct geomorphic channel formation and maintenance processes and associated
698 ecosystem functions were thus revealed from otherwise similar channel types and valley settings
699 based on differences in subreach-scale topographic variability.

700 *Value of topographic variability attributes*

701 *Distinguishing channel types.* With respect to the first study objective, TVAs were found to
702 play a major role in distinguishing channel types across the landscape. Numerous univariate and
703 multivariate statistical analyses all identified bankfull width and depth variability as first-order
704 predictors of geomorphic channel type. Even though S and A_c - frequently identified as dominant
705 variables controlling channel form and geomorphic processes (Leopold and Maddock 1953;
706 Dunne and Leopold 1978; Dietrich et al. 1992; Montgomery and Buffington 1997; Church 2002)
707 - were used to stratify the channel network prior to random sampling, they were not identified as
708 the primary attributes distinguishing geomorphic channel types, though they were significant
709 attributes in CART. The hierarchical clustering structure (Figure 6) and classification tree
710 (Figure 9) both identified $CV_{w.BF}$ as the primary splitting variable distinguishing channel types
711 for LSR streams of the Sacramento Basin.

712 Unlike most geomorphic attributes, which had overlapping value ranges across all but one
713 channel type (e.g., $\overline{w.d_{BF}}$, $\overline{e.ratio}$, \overline{sn} , \overline{shear}), $CV_{w.BF}$ and $CV_{d.BF}$ exhibited more uniform
714 density distributions (Figure 5) and expressed a continuum of value ranges across all nine
715 channel types (Figure 7). Thus, TVAs were found to be very important because they show that
716 some rivers have substantial channel bed and width variability and some do not– it is the

717 variability in the variability that makes them powerful classifiers compared to A_c and many other
718 reach-average metrics. For example, the channel classification distinguished four channel types
719 with very low, one with moderate, and four with high topographic variability. Of the highly
720 variable channel types, two exhibited primarily positive width and depth covariance, one
721 exhibited primarily negative covariance, and one exhibited a mixture of both.

722 It may be possible that the significance of TVAs in this study is influenced by the specific
723 positioning or frequency of cross-sections along each study reach. Topographic variability is
724 often structured with quasi-periodic undulations, so how sample locations align with those
725 structures is very important and probably should not be left to chance when designing
726 observation protocols. Future studies with more cross-sections per reach or using near-census
727 channel width measurements based on high-resolution remote sensing data would reduce the
728 likelihood that the variability being measured is a function of the cross-section locations.
729 However, the statistically distinct clustering solution and physical interpretability of results
730 indicate that the significance of TVAs in the channel classification is fundamentally based on
731 differences in subreach-scale channel forms and processes.

732 Furthermore, study results indicate that the history of land use and anthropogenic alterations
733 in the Sacramento Basin are not artificially inflating the importance of TVAs in the landscape. If
734 any reaches with small degrees of variability stood out given the simplified nature (e.g., dredged
735 and straightened) of many parts of the basin, one would expect to see a highly skewed
736 distribution of TVA values towards low variability. However, the uniform distributions exhibited
737 by $CV_{w,BF}$ and $CV_{d,BF}$ (Figure 5) negate this hypothesis, indicating instead a large, relatively
738 evenly distributed range of width and depth variability across the landscape.

739
740 *Characterizing dominant channel processes.* With respect to the second study objective,
741 TVAs were found to be extremely useful for characterizing dominant channel processes that
742 have been reported extensively in the literature but which have been neglected from quantitative
743 classification studies prior to this. Most studies only consider processes in terms of reach-average
744 erosive potential, sometimes relative to sediment supply. They have no basis for describing
745 channel types in terms of the actual specific processes that occur in reaches, such as knickpoint
746 migration, bank erosion, and island formation. By incorporating TVAs in a channel classification
747 framework, we were able to characterize and distinguish the type and magnitude of topographic

748 variability within reaches. In doing so, this study provided a quantitative basis for interpreting
749 the resultant classes in terms of a diversity of mechanisms for fluvial landform formation and
750 maintenance that rely on both nonuniform and uniform channel morphology (Lane and Carlson
751 1953; Dietrich and Smith 1983; Thompson 1986; Paustian et al. 1992; Wohl and Thompson
752 2000; Makaske 2001; Powell et al. 2005; Wilcox and Wohl 2006; White et al. 2010). As
753 hypothesized, TVAs - closely tied to nonuniform processes - improved the ability to characterize
754 and compare dominant channel processes in many channel types. For example, differences in
755 TVAs and their covariance as distinguished by the channel classification appeared to be
756 indicative of different sediment transport mechanisms in partly-confined pool – riffle and large
757 uniform channels. Similarly, the high channel depth variance distinguishing unconfined plateau
758 small pool-riffle channels from large uniform channels supported the interpretation of the
759 dominant channel forming process as avulsion and the dominant channel maintenance processes
760 as topographic steering, flow convergence, and eddy recirculation in spite of very similar valley
761 settings and traditional geomorphic attributes (e.g., \overline{slope} , $\overline{w.d_{BF}}$, $\overline{e.ratio}$, $\overline{D_{84}}$). Alternatively,
762 unconfined large uniform boulder and meandering sand bed channel types were differentiated on
763 the basis of underlying geology rather than TVAs.

764 *Ecological implications.* The spatial variability or lack thereof of channel morphology and
765 associated geomorphic processes as distinguished by TVAs has important ecological
766 implications. For example, differences in spatial patterns of hyporheic exchange (Kasahara and
767 Wondzell 2003; Tonina and Buffington 2009) drive differences in local biogeochemistry (Poole
768 et al. 2008) and habitat dynamics (Geist 2000). Channels with high subreach topographic
769 variability and associated heterogeneous sediment scour and deposition (e.g., our *pool-riffle* and
770 *cascade/step-pool* channels) may exhibit highly localized hyporheic exchange (Kasahara and
771 Wondzell 2003; Poole et al. 2006, 2008), creating local nutrient hotspots associated with algae or
772 macrophyte growth (Fisher et al. 1998) and preferential spawning habitat (Geist 2000). In
773 contrast, the uniform flow and sediment transport processes exhibited by very low topographic
774 variability (e.g., *upland valley uniform* channels) are associated with long hyporheic flow paths
775 that modify the reach's mean daily temperature (Poole et al. 2008) and biogeochemistry (Findlay
776 1995) from average channel conditions, in turn affecting habitat quality (Poole et al. 2008;
777 Tonina and Buffington 2009) and salmonid population structure (e.g., Burnett et al. 2003)
778 throughout the reach. Unconfined uniform channels with the propensity for these long hyporheic

779 flow paths have also been shown to provide low-velocity refugia for biota during periods of high
780 flow (e.g., Wenger et al. 2011) and support wider riparian zones (Polvi et al. 2011).

781 Incorporating TVAs in channel classification is also expected to inform river restoration
782 efforts. For example, riparian species richness has been shown to increase with subreach-scale
783 bed elevation variability (Pollock et al. 1998), suggesting that characterizing TVAs in addition to
784 more traditional geomorphic attributes may help predict the impact of disturbances on the biotic
785 community across the channel network. Targeting high variability channel types (e.g.,
786 *cascade/step-pool*, *pool-riffle*) for riparian restoration efforts may increase the likelihood of
787 success by increasing the range of hydrogeomorphic and thus ecological responses to
788 disturbance. Alternatively, channel change associated with channel unit to reach scale (e.g., 10 –
789 100 channel widths) changes in TVAs may indicate changes in flow regimes, sediment regimes,
790 or land use (Montgomery and Bolton 2003), indicating critical locations for larger-scale
791 restoration efforts. For example, the conversion of fully forested riparian zones to grasslands has
792 been associated with a significant reduction in within-reach width variability (Jackson et al.
793 2014). By identifying channels with rapidly changing $CV_{BF,W}$, practitioners may more easily
794 define management objectives and prioritize restoration activities. Characteristic TVA values of
795 ecologically functional reaches could provide practitioners with a baseline level of channel and
796 floodplain variability to incorporate into restoration efforts for degraded reaches.

797 *Future research*

798 With the aim of characterizing dominant process regimes of distinct channel types as
799 differentiated by TVAs, we speculated as to the physical processes associated with each
800 identified channel type. We suggest direct measurement of these hypothesized dominant
801 subreach-scale processes and their co-occurrence with distinct TVA settings as an important
802 direction for future work. For instance, measurement of hydraulic flow fields, hyporheic
803 exchange, or sediment transport rates across channel types would bolster physical understanding
804 of the differences in processes regimes between distinct TVA settings.

805 With the emergence of meter-scale remote sensing of rivers, datasets that support computing
806 and analyzing TVAs will become more available, accurate, and useful (Gleason and Wang 2015;
807 Gonzalez and Pasternack 2015). In the meantime, by considering TVAs in addition to more
808 traditional channel classification attributes, we hope to encourage future research into how a
809 stream reach is influenced by its surrounding landscape at various scales based on hierarchical

810 topographic variability relationships. This could enable the application of increasingly available
811 larger-scale topographic datasets to distinguishing differences in multi-scale process controls on
812 channel morphology and predicting reach-scale geomorphic settings. Further understanding of
813 relationships between TVAs and multi-scale geomorphic processes is critical to developing
814 insight into sediment transport and formative processes in these diverse channel types.

815 **Conclusion**

816 This study found that measures of subreach-scale topographic variability provided improved
817 information on river geomorphic landforms and processes in channel networks of varied
818 landscapes. When incorporated in a channel classification framework among a suite of more
819 traditional geomorphic attributes, TVAs improved the ability to distinguish dominant channel
820 types and associated geomorphic processes in low-volume snowmelt and rain dominated streams
821 of a Mediterranean region. Bankfull width variance was identified as the primary attribute
822 distinguishing channel types over common attributes such as channel slope, width-to-depth ratio,
823 confinement, sinuosity, and dominant substrate. The nine channel types distinguished for the
824 Sacramento Basin included both channel types with strong analogs in existing geomorphic
825 literature and novel channel types. By reenvisioning channel classification through the
826 incorporation of TVAs, distinct channel landforms and processes were revealed from otherwise
827 similar geomorphic settings with limited additional resource requirements. Results indicate that
828 incorporating TVAs in channel classification may improve river restoration efforts by revealing
829 ecologically-significant differences in channel form and function.

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833

834

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