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Hierarchically nested river landform sequences. Part 2: Bankfull channel morphodynamics governed by valley nesting structure

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3

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5

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12

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18

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22 **Abstract**

23 River corridors exhibit landforms nested within landforms repeatedly down spatial
24 scales. In Pasternack *et al.* (2018), a new, scale-independent, hierarchical river
25 classification was developed that uses five landform types to map the domains of a
26 single fluvial process— flow convergence routing— at each of 3-5 spatial scales. Given
27 those methods, this study investigated the details of how flow convergence routing
28 organizes nested landform sequences. The method involved analyzing landform
29 abundance, sequencing, and hierarchical nesting along the 35-km gravel/cobble lower
30 Yuba River in California. Independent testing of flow convergence routing found that
31 hydraulic patterns at every flow matched the essential predictions from classification,
32 substantiating the process-morphology link. River width and bed elevation sequences
33 exhibit large, nonrandom, and linked oscillations structured to preferentially yield wide
34 bars and constricted pools at base flow and bankfull flow. At a flow of 8.44 times
35 bankfull, there is still an abundance of wide bar and constricted pool landforms, but
36 larger topographic drivers also yield an abundance of nozzle and oversized landforms.
37 The nested structure of flow convergence routing landforms reveals that baseflow and
38 bankfull landforms are nested together within specific floodprone valley landform types,
39 and these landform types control channel morphodynamics during moderate to large
40 floods. As a result, this study calls into question the prevailing theory that the bankfull
41 channel of a gravel/cobble river is controlled by in-channel, bankfull, and/or small flood
42 flows. Such flows may initiate sediment transport, but they are too small to control
43 landform organization in a gravel/cobble river with topographic complexity.

44

45 **Introduction**

46

47 Study motivation

48

49 For several decades Earth and environmental scientists have conceived of the
50 landscape and its rivers as consisting of hierarchically nested objects (Woldenberg;
51 1969; Frissell *et al.*, 1986; Hunsaker and Levine, 1995; Imhol *et al.*, 1996; Brierley and
52 Fryirs, 2000; Hay *et al.*, 2001). This conception means the landscape consists of
53 discrete, discernable features that are organized by size, with a small number of larger
54 objects containing an exponentially larger number of smaller objects, repeated down
55 spatial scales until the continuum assumption breaks down (Horton, 1945). Specifically,
56 terrestrial continent objects consist of catchment objects, which in turn consist of
57 subcatchment objects. Subcatchment nesting continues down scales until the scale of
58 hillside and river segment objects (Rodriguez-Iturbe and Rinaldo, 1997). Foregoing
59 hillsides, river segment objects then consist of reach objects, which in turn consist of
60 morphological unit objects, which in turn consist of hydraulic unit objects, which in turn
61 consist of surficial roughness objects (Thomson *et al.*, 2001).

62 Object-based hierarchical conceptualization allowed for the development of
63 independent questions, methods, and results at each scale (Pasternack, 2011) as well
64 as corresponding management solutions (Beechie *et al.*, 2010). The dominant scientific
65 paradigm of empirical, field-based research at each scale involved representative
66 sampling with a very small number of samples, because research was data-limited
67 (Brennan *et al.*, 2002; Smith and Jones, 2008). In some cases, tests were done to

68 ensure that results were not impacted by insufficient sampling (e.g., Angermeier and
69 Smogor, 1995). Such testing has yielded mixed results, especially for physical studies
70 (e.g., Thomson *et al.*, 2004; Gonzalez and Pasternack, 2015). Most often, no such
71 testing was possible in the absence of a population census to test against. Many
72 scientific ideas and practical applications therefore make assumptions about spatial
73 scaling that are largely untested. As a result of the lack of commensurate data and
74 results, it is extremely difficult to synthesize a universal scientific conceptualization
75 based on empirical research that works across all scales.

76 In contrast to the object-oriented hierarchical nesting paradigm of data-limited
77 settings, data-rich systems are predominantly analyzed using signal processing
78 methodologies (Priestley, 1981) that deconstruct data series in time or space (or both)
79 to find patterns at each scale (Pawlowicz *et al.*, 2002) and ones that transcend scales
80 (Barenblatt and Monin, 1979; Rodriguez-Iturbe and Rinaldo, 1997). These analyses use
81 the finest resolution data support. They have unified questions, methods, and results
82 that work across all scales to provide a coherent, universal conceptualization. For
83 Earth's surface data, this approach has been feasible using remotely sensed raster data
84 (Kumar and Fofoula-Georgiou, 1997; Jakubauskas *et al.*, 2002), but pixel resolution
85 has been too coarse (~ 30-100 m) for fluvial geomorphology. Topographic data at that
86 scale has poor vertical accuracy (Neeson *et al.*, 2008).

87 Mapping of the Earth's surface can now achieve a near census (1-m) sampling of
88 the population of elevation (Westoby *et al.*, 2012), LiDAR intensity (Mandlbürger *et al.*,
89 2015), and electromagnetic multi-spectral properties (Legleiter *et al.*, 2009). There
90 remain finer levels of continuum detail that ground-based technology addresses over

91 small areas (Brasington *et al.*, 2012), and will eventually span at the landscape scale.
92 Nevertheless, the 1-m, near-census scale of data acquisition is capable of
93 fundamentally transforming analysis of hierarchically nested landscapes. Studies
94 pursuing this for science (Legleiter, 2014; Pasternack and Wyrick 2016; Brown and
95 Pasternack, 2017) and management (Brown *et al.*, 2014; Pasternack and Brown, 2016)
96 are emerging.

98 Study purpose

99
100 The overall goal of this study was to reenvision the notion of hierarchical nesting in
101 rivers and reveal a new understanding of river patterning. Prior to this study, object-
102 oriented river classifications used unique typologies at each scale that are
103 incommensurate with those at other scales (see citations in first sentence of this article).
104 In Pasternack *et al.* (2018), we proposed a new, continuum-based, scale-independent
105 approach to classifying landforms with respect to a single morphodynamic mechanism
106 that can occur at many fluvial scales. The approach is amenable to signal processing
107 analyses that enable the same typology to be employed over the same wide range of
108 scales that the mechanism spans. This capability provides a unified theory of fluvial
109 process-morphology linkages for any one process. We chose the mechanism of flow
110 convergence routing as the illustrative mechanism to focus on (see Pasternack *et al.*
111 (2018) for background literature, classification scheme, and data analysis methods).

112 In this article, we apply the classification and analysis framework to spatial series of
113 topographic data from a 35-km gravel/cobble bed river corridor to reveal the abundance,

114 sequencing, and nesting of scale-independent landforms relevant for understand flow
115 convergence routing. There are four study objectives- three to analyze the landforms
116 and one to validate the velocity pattern assumed by the underlying classification theory.
117 For each objective, there are three to five specific, tractable questions (Table 1). Some
118 results from this application support existing concepts about fluvial geomorphology,
119 while others present significant evidence against prevailing wisdom; hence this article is
120 not merely descriptive but tests fundamental scientific ideas about rivers.

121

122 **Study area**

123

124 Geographic Setting

125

126 The Yuba catchment in California drains 3480 km² of Dry Summer Subtropical
127 mountains down to the confluence with the Feather River (Figure 1). Like many
128 mountain catchments, this one experienced anthropogenic impacts, notably hydraulic
129 gold mining (Gilbert, 1917), timber harvesting, and flow regulation. Englebright Dam
130 was built in 1940 to trap nearly all sediment and thereby promote downstream
131 geomorphic recovery, which continues to proceed more than 70 years later (Carley *et*
132 *al.*, 2012). Daguerre Point Dam (DPD) is an 8-m high irrigation diversion structure
133 located at river kilometre (RKM) 17.8 that creates a slope break and partial sediment
134 barrier.

135 The 37.1-km LYR segment (Figure 1) is a single-thread channel (~ 20 emergent
136 bars/islands at bankfull) with low sinuosity, high width-to-depth ratio, slight to no

137 entrenchment, and eight distinct geomorphic reaches (Wyrick and Pasternack, 2012).
138 The river segment has a mean bed slope of 0.185% and a mean surface substrate
139 diameter of 97 mm (i.e., small cobble). As a comparison to other rivers, the LYR is
140 classified as a C3 channel by the Rosgen (1994) Stream Type classification method
141 when applied to the segment and as transitional straight-meandering by the flow
142 instability method. This study investigated the six alluvial geomorphic reaches as one
143 segment and by reach (Figure 1).

144 The LYR corridor has natural canyon and valley walls in the first 9 RKM below
145 Englebright Dam, and there are major artificial constraints on corridor width. The river
146 corridor is confined in a steep-walled bedrock canyon for the upper 3.1 RKM, then
147 transitions first into a wider bedrock valley with some meandering through Timbuctoo
148 Bend (RKM 28.3-34.0), then into a wide, alluvial valley downstream to the mouth.
149 During the late 19th to mid 20th century, gravel and gold miners dredged and re-
150 arranged the topography of the LYR creating high and wide berms of dredger mine
151 tailings that isolate the modern river from the ~ 40-km² of extremely disturbed landscape
152 (Yuba Goldfields), which is still actively mined. Upstream of the Yuba Goldfields there
153 are two major artificial mine-tailing berms within Timbuctoo Bend. Downstream of the
154 Yuba Goldfields river corridor width is constrained by agricultural land use and
155 engineered levees. All of these forced geographic controls on width and width
156 undulation drive geomorphic responses in bed elevation and its downstream undulation
157 in turn (Brown and Pasternack, 2014, 2017). Such links are further investigated in this
158 study.

159

160 Hydrogeomorphic Regime

161

162 This study investigates landform patterns associated with flows spanning 0.14 to
163 8.44 times bankfull discharge (Q_{bf}), which equals 19.82 to 1195 m^3/s . Regulated LYR
164 base flows are commonly between ~ 14 and 23 m^3/s , with a flow of 19.82 m^3/s serving
165 as the negotiated minimum release from Englebright Dam during all but the driest years.
166 Different locations along the river exhibit spillage out of the channel into low-lying
167 peripheral swales and onto lateral and point bars at flows from ~ 84.95 to 141.6 m^3/s .
168 When water stage rises to 141.6 m^3/s , relatively flat active bar tops become inundated
169 and the wetted extents line up with the base of willows along steeper banks flanking the
170 channel. Based on these and other field indicators, 141.6 m^3/s represents Q_{bf} adjusted
171 to the modern regulated flow regime since 1970. This flow has $\sim 82\%$ annual
172 exceedence probability. By a flow of 198.2 m^3/s , banks are all submerged and water is
173 spilling out to various degrees onto the floodplain. The modern floodplain is considered
174 fully inundated when the discharge reaches 597.5 m^3/s , so this is the water surface area
175 referred to herein as the “floodway”. Above this flow, alluvial terraces, bedrock outcrops,
176 training berms, and soil-mantled hillsides become inundated. A flow of 1195 m^3/s yields
177 a depth twice that of bankfull discharge (Wyrick and Pasternack, 2012), which by
178 definition fills the floodprone area, as defined by Rosgen (1994).

179

180 **Methods**

181

182 Data used in this study consisted of a DEM, geomorphic reach breaks, water surface
183 area polygons, and depth-average velocity rasters earned through years of fieldwork,
184 quality assurance procedures, and mechanistic numerical modeling. They were
185 thoroughly vetted and published in peer reviewed technical reports and journal articles.

186 River corridor topography and bathymetry were collected for the meter-resolution
187 DEM using a combination of airborne LiDAR, ground-based surveying, and boat-based
188 sonar. Each method involved its own internal performance tests and yielded different
189 point densities (complete details in Pasternack *et al.*, 2014; Strom *et al.*, 2016). For
190 example, within and beyond the 24.92 m³/s water surface area, point density
191 downstream of TBR was 59 and 554 pts/100 m², respectively.

192 Water surface area polygons from a published meter-scale 2D hydrodynamic model
193 (solved with the United States Bureau of Reclamation SRH-2D algorithm) were
194 available for 28 flows ranging from 8.50 to 3126 m³/s (Abu-Aly *et al.*, 2013; Pasternack
195 *et al.*, 2014). This study focused on evaluating spatial series of detrended, standardized,
196 cross-sectionally averaged bed elevation (Z_s), standardized cross-sectional top width
197 (W_s), $W_s \cdot Z_s$, and landforms identification codes at five representative flows– 19.82,
198 141.6, 283.2, 597.5, and 1195 m³/s– whose significance was explained in the study
199 area section, except 283.2 m³/s, which is simply $2 \cdot Q_{bf}$. In addition to the five flows
200 previously listed, velocity data for four other flows (17.63, 28.32, 2390, and 3126 m³/s)
201 were used to improve the detail of the velocity-discharge hydraulic geometry relation
202 and span larger floods when addressing objective two. For objective four, landform
203 nesting was investigated at three scales, the perennial base flow channel ($0.14 \cdot Q_{bf}$), the
204 bankfull channel (Q_{bf}), and the floodprone valley floor ($8.44 \cdot Q_{bf}$).

205 The bisecting centerline of the water surface area at each flow was obtained using
206 ArcGIS® version 10.3. Centerlines were stationed with a spacing of 3% of mean bankfull
207 channel width. In this study, the bankfull width was ~ 100 m (Wyrick and Pasternack,
208 2012) and spacing had to be done in American customary units (10 ft), so in metric units
209 the spacing was 3.048 m.

210 Pasternack (2011) provided workflows for obtaining water surface area polygons
211 and velocity magnitude rasters from SRH-2D outputs. Many 2D model validation tests
212 were done for an order of magnitude range of flow from ~ 14 to 170 m³/s (Barker, 2011).
213 Water surface elevation, depth, velocity magnitude, and velocity direction model
214 performance was on par or better than accepted scientific norms. Median unsigned
215 velocity magnitude error was 16%, which is less than commonly reported.

216 Data analysis methods (Table 1) were explained in Pasternack et al. (2018). Each
217 analysis was implemented using ArcGIS® 10.3 for geospatial processing and Microsoft
218 Excel® for statistical analysis. Analyses by discharge were performed on each
219 geomorphic reach and the whole river segment to compare and contrast reach-
220 dependent hierarchical controls on landform organization.

221

222 **Results**

223

224 Structure of topographic heterogeneity

225

226 Every analysis performed in this study provided a strong corroboration building on
227 the previous study of TBR by Brown and Pasternack (2017). Specifically, the LYR is

Table 1. Scientific analysis framework for this study applied to whole river segment and each geomorphic reach.

Objectives (O#) and their questions	Test variables	Analysis
(O1) Analyze stage-dependent structure of fluvial topographic deviation from central tendency using longitudinal series of standardized width (Ws) and detrended, standardized bed elevation (Zs)		
(1a) What percent of the river has topographic variations greater than 0.5 and one standard deviations away from the mean?	Abs(Zs), Abs(Ws)	percent of values > 1
(1b) Is longitudinal topographic structure random?	series of Zs, Ws	Wald-Wolfowitz runs tests
(1c) Are width and bed elevation series correlated, as one indicator of coherent organization?	series of Zs, Ws	Pearson's product-moment correlation for Ws and Zs
(1d) Is the specific longitudinal structure of the river's morphology consistent with a dominant role for flow convergence routing?	series of Ws·Zs	mean(Ws·Zs); percent of values > 0
(O2) Do landforms classified in this system exhibit the specific stage-dependent differences in velocity (V) expected by the flow convergence routing mechanism?		
(2a) Does oversized have lower velocity than normal channel and does the latter have lower velocity than nozzle?	0.91-m V raster from 2D hydrodynamic model	mean(V) and V95* among raster cells in landform
(2b) Are constricted pool landforms low velocity at base flow and high velocity at flood flow?	0.91-m V raster from 2D hydrodynamic model	mean(V) and V95* among raster cells in landform
(2c) Are wide bar landforms high velocity at base flow and low velocity at flood flow?	0.91-m V raster from 2D hydrodynamic model	mean(V) and V95* among raster cells in landform
(O3) Analyze relative abundance and longitudinal sequencing of landforms by reach and discharge.		
(3a) What is the relative abundance of each landform for the whole river for each flow?	series of landform IDs	count and compare
(3b) How do geomorphic reaches compare in landform composition?	series of landform IDs	count and compare
(3c) How does landform abundance change with flow?	series of landform IDs	count and compare
(3d) What is the longitudinal sequencing of landforms?	series of landform IDs	count times each unit followed another
(3e) How does longitudinal sequencing change with flow?	series of landform IDs	count times each unit followed another
(O4) what is the stage-dependent, nested structure of landforms classified by their flow convergence routing potential?		
(4a) What are top five most abundant nested permutations?	nested series of landform IDs	permutation abundance analysis
(4b) For each landform at the floodprone scale, what are the top five most abundance nested permutations?	nested series of landform IDs	permutation abundance analysis
(4c) For each landform at the bankfull scale, what are the top five most abundant nested permutations?	nested series of landform IDs	permutation abundance analysis
(4d) For each landform at the bankfull scale, what are the top three most abundant floodprone landform hosts?	nested series of landform IDs	permutation abundance analysis
*V95 is the 95th percentile value of velocity among all 0.91-m pixels in the area of any one landform type		

228 primarily defined by its variability, not its central tendency. Stations along the river rarely
229 exhibited Zs or Ws values of zero, and instead exhibited wide swings (Figure 2). The
230 Wald-Wolfowitz runs test showed that Zs and Ws series were nonrandom at all flows
231 tested ($p < 10^{-6}$).

232 At nearly all flows, more than half the river's bed length had Zs values > 0.5 standard
233 deviations away from the mean value, and almost two-thirds of it had Ws values beyond
234 that criterion (Table 2). Approximately a quarter of the river's bed length had Zs and Ws
235 values more than one standard deviation high and low. MR and TBR had the most
236 extreme bed undulations, while DPDR and DCR had the most uniform beds (Figure 2).

237 Because width was standardized at the segment scale, the mean Ws for each reach
238 (Table 2c) was free to deviate from the overall central tendency. Some reaches (DCR
239 and DPDR) were especially wide, while others were especially narrow (MR and TBR).
240 The two reaches with the most uniform Zs (DPDR and DCR) were also the widest on
241 average. DCR was wider than average at all flows and was widest at 1-2 times Q_{bf} ,
242 whereas DPDR was an average width for in-channel flows, but abruptly widened a lot
243 after $2 \cdot Q_{bf}$ (Figure 3). Greater widths are explained by the presence of a secondary
244 anastomosing channel and excellent floodplain connectivity in that reach.

245 Moving from reach-average width to Ws variability, MR was not only narrow with
246 extreme Zs variations, but it also had the most extreme Ws variations (Table 2d,e). PBR
247 had a normal abundance of extreme widths for in-channel flows, but its overbank flows
248 were unusually constricted. This result is explained by the presence of historically
249 created artificial terraces of dredged coarse sediment. DCR showed a similarly

Table 2. Metrics for topographic variability. Dark and light shading indicate high and low values, respectively.

flow (xQbf)*	Segment	Geomorphic reaches**						
		MR	HR	DPDR	DCR	PBR	TBR	
A) % Abs(Zs)>0.5								
0.14	59	67	57	41	52	60	72	
1	58	69	58	36	39	61	75	
2	57	70	60	35	30	59	75	
4.22	53	73	63	36	16	45	67	
8.44	47	87	21	34	25	58	58	
(B) % Abs(Zs)>1								
0.14	29	36	20	10	23	30	50	
1	28	38	25	5	15	29	49	
2	28	52	26	5	8	25	47	
4.22	25	54	25	7	5	15	39	
8.44	21	70	3	5	6	19	31	
(C) Ws mean								
0.14	0.00	-0.03	-0.09	-0.17	0.49	-0.15	0.16	
1	0.00	-0.63	0.20	0.14	0.80	0.00	-0.34	
2	0.00	-0.81	0.26	0.29	0.94	0.00	-0.50	
4.22	0.00	-0.97	0.12	1.29	0.52	-0.01	-0.74	
8.44	0.00	-0.84	0.01	1.28	0.32	0.18	-0.82	
(D) % Abs(Ws)>0.5								
0.14	62	65	54	63	66	66	63	
1	63	81	68	43	64	63	60	
2	66	86	70	49	70	60	65	
4.22	69	85	73	94	58	40	70	
8.44	65	82	61	94	26	44	81	
(E) % Abs(Ws)>1								
0.14	28	27	17	25	38	34	33	
1	29	37	35	15	39	28	24	
2	32	57	44	16	48	14	20	
4.22	39	71	37	75	14	11	35	
8.44	40	61	38	74	21	15	36	

*Flow values are given in multiples of bankfull discharge

**Reach names as given in Figure 1.

250 significant decline in width variability with flow, but that did not happen until after $2 \cdot Q_{bf}$.
251 DPDR showed a significant increase in extreme widths above $2 \cdot Q_{bf}$.

252 For in-channel flows, although the abundance of Ws extremes was equal to that of
253 Zs extremes, the higher flow rose above bankfull, the more Ws deviated from its central
254 tendency (Table 2d,e; Figure 3). The segment as a whole secularly decreased its Zs
255 variability and secularly increased its Ws variability with increasing discharge, but no
256 reach does either on its own. MR was the only reach that increased its Zs and Ws
257 variability with discharge.

258 Given that Zs and Ws showed significant, nonrandom variations along the LYR at all
259 flows tested, the next analysis tested whether those were linked. The landform
260 classification developed in this study does not need them to be, but the interpretation of
261 flow-dependent hydrogeomorphic processes depends on how they relate. All Pearson's
262 product-moment correlation coefficient (r) values were positive correlations statistically
263 significant at the 99% confidence level (Figure 4). That means that when a water
264 surface area is narrow (low Ws), it tends to be deep (low Zs); when it is wide (high Ws),
265 it tends to be shallow (high Zs). That result is consistent with an interpretation that the
266 river is primarily organized into constricted pools, normal channels, and wide bars.
267 Eleven out of 35 cases (31%) had a r-value > 0.7 , which means Zs and Ws variations
268 were directly linked to a significant degree. No reach always had the highest r-value
269 across all flows, but HR and TBR had the highest average of r-values among flows,
270 while DCR had the lowest. Among flows, the highest r-value for the river segment
271 existed for Q_{bf} , with a close second by $2 \cdot Q_{bf}$. Meanwhile, the highest and lowest flows
272 had the lowest and second lowest r-values, respectively.

273 The last test pushed further to aid interpretation of flow convergence routing by
274 assessing the sign and magnitude of $Ws \cdot Zs$, also known as the “geomorphic
275 covariance” (Brown and Pasternack, 2014, 2017). The mean value of $Ws \cdot Zs$ was above
276 zero for all flows (Table 3a) and the series of $Ws \cdot Zs$ consisted of both ++ and --
277 combinations, indicating a predominance of self-sustainable constricted pool and wide
278 bar units, capable of rejuvenating themselves by way of the flow convergence routing
279 mechanism. Further, the vast majority of $Ws \cdot Zs$ values were positive for all flows
280 (Figure 5). Most interestingly, Q_{bf} had the highest mean $Ws \cdot Zs$ value and the highest
281 percent of station $Ws \cdot Zs$ values above zero (Table 3b). The lowest mean and percent
282 positive $Ws \cdot Zs$ values occurred for the largest flood investigated (Figure 5). All reaches,
283 except MR and HR exhibited a decrease in $Ws \cdot Zs$ above Q_{bf} . HR also exhibited a
284 decrease in $Ws \cdot Zs$, but it did not begin until after $2 \cdot Q_{bf}$. MR had a unique large increase
285 in mean $Ws \cdot Zs$ with discharge, which is consistent with its narrowing and increased bed
286 and width undulations with discharge. Finally, correlation between the $Ws \cdot Zs$ values for
287 a given flow and the sequentially higher flows went down as discharge increased, with
288 the highest correlation for Q_{bf} versus $2 \cdot Q_{bf}$, similar to results from Brown and
289 Pasternack (2017).

290

291 Landform-stratified velocity

292

293 Landform classification was applied to the $Ws \cdot Zs$ longitudinal series (colors in Figure
294 5). Prior to analyzing those results, 2D model velocity rasters were tested for the
295 expected relative differences in velocity among landform types required of the flow

Table 3. Results of Ws-Zs analysis. Dark and light shading indicate highest and lowest values, respectively.

flow (xQbf)	Segment	MR	HR	DPDR	DCR	PBR	TBR
(A) mean Ws-Zs							
0.14	0.46	0.26	0.40	0.33	0.48	0.51	0.76
1	0.62	0.59	0.77	0.31	0.45	0.67	0.78
2	0.55	0.66	0.80	0.26	0.24	0.51	0.65
4.22	0.40	0.92	0.56	0.07	0.09	0.29	0.35
8.44	0.32	1.24	0.17	0.12	0.20	0.15	0.24
(B) percent of Ws-Zs >0							
0.14	70	61	75	74	71	73	64
1	77	63	86	82	65	84	72
2	77	65	86	77	63	89	67
4.22	66	73	76	48	65	77	49
8.44	62	74	60	57	68	70	46

296 convergence routing morphodynamic mechanism. The results largely matched
297 expectations, corroborating the theoretical underpinnings of this framework. Both
298 landform-averaged velocity and the 95th percentile of raster cell velocities in each
299 landform show the same relationships (Figure 6). For all discharges, oversized had
300 lower mean velocity and 95th percentile of velocity than normal channel, which in turn
301 had lower values than nozzle. The values were most differentiated for the lowest and
302 highest discharges tested and closest at $2 \cdot Q_{bf}$. Meanwhile, constricted pool and wide
303 bar showed the expected velocity divergence, with the former having a steep rate of
304 increase in velocity with increasing discharge and the latter having a gentle one (Figure
305 6). At the lowest discharge, constricted pool velocity was lower than wide bar velocity
306 for both velocity metrics. The velocity reversal between these two landforms occurred
307 between the lowest and second lowest discharges, but constricted pool had a relatively
308 high velocity even at low discharge. Beginning at Q_{bf} and for all higher flows, constricted
309 pool had significantly higher velocity than wide bar for both metrics. In essence, normal
310 channel, constricted pool, and wide bar all had similar cross-sectional areas at low
311 flows, so they had similar velocities. For overbank flows, their cross-sectional areas
312 dramatically diverged, causing the associated change in flow convergence routing.

313 The most interesting velocity results came from comparing wide bar versus
314 oversized and constricted pool versus nozzle. Velocity reversals occurred in which
315 nozzle and oversized had the most extreme velocities for in-channel flow, but for
316 overbank flows constricted pool and wide bar had them at some flows (Figure 6). These
317 results are consistent with the theoretical expectation that constricted pools exhibit flow

318 convergence routing during floods, while wide bars have low velocities and thus may
319 receive the sediment scoured out of pools.

320 Yet it is also evident that the decision tree is not yielding the purest theoretical
321 outcome wherein nozzle should universally have the highest velocity and oversized the
322 lowest. Three sample velocity maps illustrate why this outcome is occurring (Figure 7).
323 The maps are briefly described as results, and then a mechanistic explanation is
324 provided in the discussion section. For in-channel flows, 2D model hydraulics closely
325 conform with the expectations of the landform classification, with longitudinal variation in
326 velocity dominating over lateral variation. For example, nozzles are the fastest and
327 oversized slowest (Figure 7a). However, during flood flows, the flow field exhibits strong
328 lateral gradients that can match or confound expectations. For example, at one MR site
329 the constricted pool definitely has a uniformly high velocity for the majority of its width,
330 and the constricted pool's velocity is higher than that in the upstream wide bar (Figure
331 7b). However, the wide bar has a strong lateral gradient with a core of high velocity in
332 the bankfull channel and a range of lower velocities across the whole inundated bar
333 complex. Thus, on average the wide bar is lower velocity than the constricted pool and
334 its 95th percentile of velocity is lower than that of the constricted pool, conforming to flow
335 convergence routing theory, even if the average state does not convey the whole story
336 of the hydraulic mechanism at play.

337 However, the situation at the highest flows is where the complexity of natural 2D flow
338 fields defies the purest form of a cross-sectionally averaged interpretation of flow
339 convergence routing. For example, during the floodprone flow of $8.44 \cdot Q_{bf}$, constricted
340 pool again showed a large fraction of its cross-sectional areas as having very high

341 velocity, whereas the nozzle upstream of it was comparatively wider (though still narrow
342 compared to segment-wide flow width) and had a slower central velocity core
343 surrounded by a wide area of low velocity (Figure 7c). This velocity pattern is a classic
344 example of a modest effective flow width. Thus, constricted pool had a higher effective
345 flow width and higher cross-sectional velocity than nozzle, which is explained in the
346 discussion section.

347

348 Landform abundance and sequencing

349

350 Even though much of the LYR exhibited a positive correlation between W_s and Z_s ,
351 there were locations with a negative correlation, yielding a diversity of landforms when
352 viewed from the lens of flow convergence routing. Among all flows and considering the
353 whole segment, normal channel was the most abundant morphology and oversized was
354 the least abundant (Table 4; Figure 5). Further, wide bar and constricted pool were
355 present in a similar medium abundance, with slightly more constricted pool. Oversized
356 and nozzle were in a similar low abundance, with slightly more nozzle. At the segment
357 scale, there was no trend in the composition of morphologies as flow increased. The
358 second highest flow was different, with more of all types at the expense of normal
359 channel, but at the highest flow the typical composition had returned.

360 Considering differences between reaches that were common among all flows, DPDR
361 and DCR, which had the most uniform bed elevations, also had the most length of
362 normal channel (Table 4). MR and TBR, which had the most undulating bed elevations,
363 had the least normal channel. MR (the narrowest) had the most nozzle and constricted

Table 4. Analysis of landform composition as a function of flow.

flow (xQbf)	% of XS locations					flow (xQbf)	% of XS locations				
	O*	CP*	NC*	WB*	NZ*		O*	CP*	NC*	WB*	NZ*
(A) segment						(E) DCR					
0.14	2.2	16	61	16	4.7	0.14	1.7	17	60	20	1.9
1	0.6	21	58	18	3.4	1	4.2	8.0	66	21	0.0
2	0.8	21	58	17	3.6	2	5.3	4.3	69	22	0.0
4.22	8.3	25	36	20	9.7	4.22	8.6	5.6	67	19	0.0
8.44	3.7	16	62	11	6.3	8.44	1.4	0.0	85	13	0.0
(B) MR						(F) PBR					
0.14	5.4	13	55	14	12	0.14	1.1	21	55	18	5.2
1	0.00	34	42	7	16	1	0.4	21	56	23	0.0
2	0.00	38	33	8	20	2	0.3	20	62	17	0.0
4.22	0.00	55	13	10	22	4.22	1.1	22	58	19	0.4
8.44	0.8	46	20	17	16	8.44	5.4	12	74	8.0	0.3
(C) HR						(G) TBR					
0.14	0.2	16	69	14	0.5	0.14	5.9	17	49	21	7.5
1	0.3	21	55	24	0.04	1	0.00	27	54	13	6.4
2	0.5	20	54	25	0.5	2	0.00	30	55	11	4.2
4.22	8.6	27	29	32	3.0	4.22	0.00	37	28	1.9	33
8.44	0.00	13	74	9.0	4.8	8.44	0.00	24	58	1.2	16
(D) DPDR											
0.14	0.1	14	74	10	1.9						
1	0.3	7.6	79	13	0.0						
2	0.6	8.0	78	13	0.1						
4.22	36.0	0.0	26	38	0.0						
8.44	16.1	0.1	59	25	0.0						

*O=oversized, CP=constricted pool,
 NC=normal channel, WB=wide bar,
 NZ=nozzle

364 pool. Wide bar was relatively abundant in all reaches, except MR and TBR. DPDR had
365 the most oversized channel.

366 Differences between reaches were also present as a function of discharge (Table 4).
367 MR and TBR showed a decrease in normal channel as discharge increased, and this
368 was offset by an increase in constricted pool. TBR also showed a significant decrease
369 in wide bar offset by an increase in nozzle, with a significant increase in nozzle above
370 $2 \cdot Q_{bf}$. This result shows the effect of artificial mine-tailing berms and natural valley
371 hillsides in TBR activating as firm constrictions. HR showed a significant increase in
372 wide bar from $0.14 \cdot Q_{bf}$ to $4.22 \cdot Q_{bf}$, but then at the highest flood, wide bars became
373 normal channel. DPDR exhibited a unique flow dependence in which oversized and
374 wide bar became significantly more abundant after $2 \cdot Q_{bf}$, which was explained by the
375 presence of a secondary anastomosing channel to the north of its perennially inundated
376 main channel. DCR had a typical abundance of constricted pool at base flow, but that
377 declined with discharge and went to zero for the largest flood. PBR was the most
378 uniform in its composition as discharge increased.

379 Beyond landform composition, landform sequencing was also analyzed to
380 understand how flow convergence routing was structured in the river. In theory, an ideal
381 river with flow convergence routing at any scale would have a sequence that alternates
382 between wide bar and constricted pool, which would necessitate some length of normal
383 channel in between to make the transition. However, this ideal was not expected,
384 because most rivers have forcing elements that also induce nozzles and oversized unit,
385 so the question involved ascertaining whether the percentage of transitions between
386 wide bar and constricted pool were higher than would occur by random chance alone.

387 Across all flows, all units predominantly transitioned to normal channel, because any
388 time there is a zero-crossing for $W_s \cdot Z_s$, then that means the presence of normal
389 channel. Though it was plausible that a wide bar might follow a constricted pool (and
390 vice versa) or that a nozzle might follow an oversized channel (and vice versa), the
391 results showed that this almost never happened— at least not with the highly
392 conservative thresholds for normal channel used in this study.

393 When normal channel was excluded from sequencing analysis, then the results
394 supported the presence and importance of flow convergence routing in maintaining
395 landform differentiation. The number of times that constricted pool was followed by wide
396 bar, or vice versa, was highest for Q_{bf} and $2 \cdot Q_{bf}$, but then decreased as flow went up or
397 down from those flows (Table 5). The percents at Q_{bf} (55 and 69%) were significantly
398 higher than expected at random given three possible transitions (33%), which is a
399 strong indicator of preference. Conversely, they were significantly lower at the two
400 largest floods tested (9 and 19%), which is a strong indicator of avoidance. At all flows,
401 oversized channel was predominantly followed by wide bar, while nozzle was
402 predominantly followed by constricted pool. This sequencing reflects the geometric
403 condition that when a reach is universally narrow at a given flow, then its sub-reach-
404 scale landforms alternate between nozzle and constricted pool. Conversely, when a
405 reach is universally wide at a given flow, then its sub-reach-scale landforms alternate
406 between oversized and wide bar. This discharge-dependent result was especially
407 dominant at the two highest floods tested; it is the first indication of hierarchical nesting
408 in the study— narrow large landforms tend to have narrow small landforms nested in
409 them, while wide large landforms tend to have wide small landforms nested in them.

Table 5. Longitudinal sequencing of landforms for the whole river, excluding normal channel units. Shading indicates highest values.

Starting unit	% of times unit followed starting unit			
	O*	CP*	WB*	NZ*
(A) 0.14·Qbf				
O	--	33	67	0
CP	31	--	43	26
WB	15	45	--	39
NZ	5	64	32	--
(B) Qbf				
O	--	30	70	0
CP	0	--	69	31
WB	32	55	--	12.9
NZ	0	85	15	--
(C) 2·Qbf				
O	--	15	85	0
CP	6	--	66	28
WB	31	57	--	11
NZ	0	77	23	--
(D) 4.22·Qbf				
O	--	13.3	87	0
CP	5	--	19	76
WB	67	10	--	24
NZ	0	81	19	--
(E) 8.44·Qbf				
O	--	17	83	0
CP	12.1	--	9	79
WB	67	19	--	14
NZ	0	90	10	--

*O=oversized, CP=constricted pool, WB=wide bar, NZ=nozzle

410 Further, given that Zs cannot control floodprone width in reaches with artificial or strong
411 natural constrictions, then it also shows that Zs is more of a response variable than Ws.

412

413 Landform nesting

414

415 Eighty-five permutations of the 125 nesting possibilities had at least one occurrence,
416 while the remaining 40 had no occurrence. The most common permutation by far was
417 the strictly defined normal channel across all flows, which occurred for 16% of stations.
418 Thus, while most abundant, it was still not particularly common. The next four most
419 common permutations were normal channel at $0.14 \cdot Q_{bf}$ and Q_{bf} nested within a wide
420 bar at $8.44 \cdot Q_{bf}$, the same nested within a constricted pool at $8.44 \cdot Q_{bf}$, the same nested
421 within oversized channel at $8.44 \cdot Q_{bf}$, and wide bar at $0.14 \cdot Q_{bf}$ and Q_{bf} nested within
422 normal channel at $8.44 \cdot Q_{bf}$. These results show that nesting permutation frequencies
423 mimic landform abundance; because normal channel is the most abundant landform at
424 all flows (Figure 5), then there is simply a higher probability of its nesting permutations
425 also being most abundant. This result is further revealed by looking at the top five
426 permutations of base flow and bankfull flow landforms within each of the five floodprone
427 landform types (Table 6). The presence of normal channel participating in nesting with
428 another unit was the top permutation in each case, and several of the other top
429 permutations involve normal channel at some level.

430 It is usually possible to handle the problem of widely different landform abundances
431 by normalizing an analysis by abundance (e.g., Wyrick and Pasternack, 2014), but in
432 the case of permutation analysis that is not possible. Instead, the question was posed in

Table 6. Top five permutations of hierarchical nesting of flow convergence routing landforms within the five floodprone landform types.

Landform ID*			count	% of river
8.44·Qbf	Qbf	0.14·Qbf		
(A) Nested within floodprone nozzle				
NZ	NC	NC	379	3.2
NZ	CP	NC	213	1.8
NZ	CP	CP	125	1.1
NZ	NC	WB	123	1.0
NZ	WB	WB	118	1.0
(B) Nested within floodprone wide bar				
WB	NC	NC	1010	8.6
WB	NC	CP	173	1.5
WB	WB	NC	165	1.4
WB	NC	WB	156	1.3
WB	CP	NC	154	1.3
(C) Nested within floodprone normal channel				
NC	NC	NC	1924	16
NC	WB	WB	481	4.1
NC	NC	WB	438	3.7
NC	WB	NC	424	3.6
NC	CP	NC	422	3.6
(D) Nested within floodprone constricted pool				
CP	NC	NC	845	7.2
CP	CP	NC	480	4.1
CP	CP	CP	284	2.4
CP	NC	WB	201	1.7
CP	NC	CP	109	0.9
(E) Nested within floodprone oversized				
O	NC	NC	592	5.0
O	WB	NC	169	1.4
O	NC	CP	146	1.2
O	NC	WB	83	0.7
O	CP	CP	73	0.6

*O=oversized, CP=constricted pool, NC=normal channel, WB=wide bar, NZ=nozzle

433 a different way, which was to ask what the other bankfull landforms are nested in and
434 what is nested within them? This time, the top three permutations were tallied.

435 Because classic velocity reversal theory anticipates a two-stage flow convergence
436 routing mechanism, then the expectation is that wide bar and nozzle landforms acting
437 as riffles at base flow should be nested within wide bar bankfull landforms. Further,
438 oversized and constricted pool base flow landforms should be nested within constricted
439 pool bankfull landforms. This expectation was largely met (Table 7a-e; Figures 5 and 8).
440 In fact, in each case, the situation in which the same landform type was nested within
441 itself occurred within the top 2 out of 5 possible permutations every time. Further, wide
442 bar had nozzle nested in it and constricted pool had oversized within it, meeting
443 expectations. Bankfull wide bar did not have oversized or constricted pool base flow
444 landforms preferentially nested in them, nor did bankfull constricted pool have wide bar
445 or nozzle preferentially nested. Figure 9a,b illustrate a ~ 2.5-km section of the river in
446 which four out of five base flow nozzles are nested within bankfull wide bars.

447 Considering bankfull nesting, bankfull nozzles exhibited nozzles and wide bars
448 nested in them, which affects the interpretation significantly. If the base flow channel
449 were set by topographic steering at bankfull flow per classic theory, then one would
450 expect to see scoured, deep (negative Zs) units nested within bankfull nozzle, assuming
451 the bed material is equally erodible in all unit types. The reason is that by definition and
452 as affirmed in the velocity results in this study, nozzle has a high sediment transport
453 capacity and thus should promote scour of the things inside of it. Using more traditional
454 terminology, the expected pattern would be to have a bankfull riffle with one or more
455 constricted base flow chutes nested within. Instead, bankfull nozzles had shallow

Table 7. Top three permutations of hierarchical nesting of bankfull landforms, either within (A-E) or beyond (F-J) them.

Qbf	0.14·Qbf	count	% of river	8.44·Qbf	Qbf	count	% of river
(A) within bankfull nozzle				(F) hosting bankfull nozzle			
NZ	NZ	228	1.9	NC	NZ	139	1.2
NZ	NC	152	1.3	NZ	NZ	132	1.1
NZ	WB	46	0.4	WB	NZ	103	0.9
(B) within bankfull wide bar				(G) hosting bankfull wide bar			
WB	NC	864	7.3	NC	WB	1021	8.7
WB	WB	845	7.2	WB	WB	327	2.8
WB	NZ	211	1.8	O	WB	316	2.7
(C) within bankfull normal channel				(H) hosting bankfull normal channel			
NC	NC	4750	40.4	NC	NC	2759	23.5
NC	WB	1001	8.5	WB	NC	1350	11.5
NC	CP	772	6.6	CP	NC	1258	10.7
(D) within bankfull constricted pool				(I) hosting bankfull constricted pool			
CP	NC	1305	11.1	NC	CP	853	7.3
CP	CP	1008	8.6	CP	CP	841	7.2
CP	O	99	0.8	NZ	CP	345	2.9
(E) within bankfull oversized				(J) hosting bankfull oversized			
O	NC	50	0.4	NC	O	31	0.3
O	O	15	0.1	O	O	26	0.2
O	WB	7	0.1	WB	O	16	0.1

*O=oversized, CP=constricted pool, NC=normal channel, WB=wide bar, NZ=nozzle

456 (positive, high Zs) base flow units (Figure 10a,b), meaning that both the bankfull and
457 base flow features were likely formed at the same time, and thus driven by higher flood
458 mechanisms, not by bankfull flow mechanisms. Similarly, bankfull constricted pool units
459 had deep units nested within them (Figures 9 and 10), also suggesting that they were
460 formed together at higher flow through scour mechanisms. Oversized had both
461 oversized and wide bar nested within, so that means that they were formed conditionally
462 on what higher flow units were doing.

463 In light of the indications that both bankfull and base flow units were likely formed at
464 the same time, an analysis was done to see what each bankfull unit type was nested
465 within at the floodprone flow (Table 7f-j). For every bankfull landform, the case where
466 that landform type is nested into the same type at the floodprone scale was in the top
467 two most frequent permutations. This outcome is illustrated by the several of the units
468 shown in Figure 10 and is a strong indication that bankfull landforms are locked in to the
469 same unit type at the next scale up, which means that the next scale up controls them,
470 or both scales are controlled by an even larger scale dynamic. Bankfull wide bar
471 occurred predominantly in normal channel floodprone landforms and secondarily over
472 floodprone wide bar, but also in oversized floodprone landforms. The section shown in
473 Figure 9 illustrates these nesting scenarios. Bankfull wide bars are thought to be
474 depositional features, and thus it makes sense that they are preferentially nested in
475 places with low to average velocity at higher discharges. Yet, that means that the
476 common interpretation of a simple two-stage riffle-pool self-maintenance mechanism
477 may in fact be influenced by or completely dominated by flow convergence routing and
478 sediment deposition on wide bars during much larger floods $> 8 \cdot Q_{bf}$. Conversely,

479 bankfull constricted pool occurred in floodprone landforms that had high velocity, which
480 were constricted pool and secondarily nozzle, as well as in normal channel floodprone
481 landforms. Bankfull normal channel was primarily hosted in normal channel floodprone
482 landforms as well as wide bar and constricted pool floodprone landforms. They were
483 most rare in floodprone nozzles. Oversized bankfull landforms occurred in normal
484 channel and oversized floodprone landforms most, and then wide bar floodprone
485 landforms.

486 Bankfull nozzle occurred in nearly equal occurrence in normal and nozzle floodprone
487 landforms, and then secondarily in wide bar ones. This result means that nozzle was
488 uniformly present over all three spatial scales as one major type for nozzle (Figure 10),
489 and these appear to be locations with highly resistant beds or channel dimensions
490 tightly constrained by human intervention at all scales, otherwise how could they persist
491 against the highest velocities present along the river. Bankfull nozzles nested within
492 wide bar floodprone landforms suggest a bar-chute morphology in which sediment
493 deposited during a large flood is then dissected with chutes to form bankfull flow
494 pathways, likely on the falling limb of the flood. This result is similar to the finding of
495 bankfull wide bars hosting base flow nozzles, again suggesting a smaller scale type of
496 bar-chute complex in such locations.

497

498 **Discussion**

499

500 Traditionally, geomorphologists viewed riffle-pool self-maintenance as largely due to
501 a relocation in peak velocity from over riffles at low flow to over pools during high flow

502 (Keller, 1971). Subsequent work generalized the driving force from velocity to shear
503 stress and Shields stress to account for bed material differences (Cao *et al.*, 2003;
504 Jackson *et al.*, 2015), but this retained the fundamental two-stage conceptualization in
505 which material eroded out of pools (and that transported from upstream) deposits on
506 riffles. Milan *et al.* (2001) reported that shear stress reversals may occur at different
507 river stages during a flood hydrograph for up to just over Q_{bf} , while Sawyer *et al.* (2010)
508 reported multiple velocity reversals during a flood of $7.63 \cdot Q_{bf}$ at one pool-riffle-run
509 sequence.

510 This study looks at the widest flow range to date and with many novel tests. It also
511 used a highly dynamic river undergoing significant erosion and deposition as its testbed
512 (Carley *et al.*, 2012). Study results call the classic two-stage mechanism into question.
513 Specifically, on the lower Yuba River, results show that in-channel and bankfull channel
514 morphologies are not just controlled by flows at those scales, but to a large degree are
515 set by morphodynamics induced by floods interacting with multiple scales of
516 topographic heterogeneity within the floodprone area. This conclusion is not
517 unprecedented. Sawyer *et al.* (2010) was the first to look across a modest range of
518 flood flows and find a three-stage shift in the location of peak velocity. Most recently,
519 Strom *et al.* (2016) showed that flow convergence routing occurs with a diversity of
520 hydraulic patch behaviors across many flows and that each type of morphological unit
521 exhibits a unique velocity versus discharge relation. The extent to which these results
522 apply elsewhere will require more studies of hierarchical landform nesting to find out.

523

524 Hierarchical topographic complexity

525

526 This study looked at an alluvial river in its transition from the foothills to the wide
527 valley floor and found wide variations in detrended bed elevation and width. Further,
528 those variations were largely organized, with high Zs and Ws coinciding and low Zs and
529 Ws coinciding. That means that the river is primarily organized into wide bars alternating
530 with constricted pools at flows from zero to $2 \cdot Q_{bf}$, and thus two-stage flow convergence
531 routing is most likely to be present.

532 Yet the structure of topographic complexity does not end at the channel banks. Wide
533 swings in Ws and Zs were found at all flows. For moderate to large floods, these higher
534 frequency variations were on top of large coherent constrictions and expansions
535 stemming from natural and anthropogenic constraints on valley floor morphology. That
536 means that width is adjustable to a degree, but that forced width constraints drive more
537 adjustments in Zs than Ws.

538

539 Velocity patterns confirmed

540

541 To a large degree, 2D modeling of river velocity corroborated the morphodynamic
542 theory underlying the landform classification proposed and implemented in this study.
543 That is important, because it means that people can move forward with assessing the
544 functionality of flow convergence routing in their rivers based on topographic analysis of
545 a meter-resolution DEM, without having to do 2D modeling. Even though 2D modeling is
546 gaining popularity and is useful when assessing river conditions and test river

547 engineering designs (Pasternack and Brown, 2013), mechanistic tools are needed for
548 river analysis at the project planning stage, long before designs are ready for 2D
549 modeling. The framework used in this study only requires a DEM (Gore and Pasternack,
550 2016).

551 Still, rivers exhibit complex lateral velocity fields, especially during floods, and thus
552 nozzles may not have the highest velocity and oversized the lowest (Figures 6 and 7).
553 First, the landform classification is based on thresholds and therefore does not consider
554 exact cross-sectional areas. To be classified as a nozzle in this system, all that has to
555 happen is that a cross section's absolute value of $Ws \cdot Zs$ has to be greater than 0.5 and
556 the cross section has to be narrower and shallower than average. However, it is
557 plausible that a constricted pool might be deeper than average, but so
558 disproportionately narrow that its cross section is smaller than that of a nozzle. This
559 geometry is exactly what happened at the site shown in Figure 7c. Similarly, a wide bar
560 might be shallower than average, yet its cross-section is so vast that its cross-sectional
561 area is greater than that of oversized. In these cases, even if velocity had a uniform
562 distribution across the water surface area, nozzle would not be fastest and oversized
563 would not be slowest.

564 Second, rivers exhibit strong lateral variation in velocity in response to topographic
565 complexity. Abrupt expansions and constrictions can cause hydraulic jetting with a
566 narrow effective width conveying most flow flanked by one or two peripheral
567 recirculations (Thompson *et al.*, 1996; Clifford, 1993). Alternately, a river may exhibit
568 large, laterally discrete inundation zones with significantly different depths and velocities
569 during floods, and thus a cross-sectional average interpretation may not fit well.

570 Finally, this classification does not account for landform differences in bed and water
571 surface slopes that drive differences in landform-stratified velocity. At the reach scale,
572 LYR has a relatively narrow slope range. However, at the landform scale it is possible
573 that there are significant differences that have been neglected. This geometry is unlikely
574 in this study because constricted pools are definitely not steeper sloped than nozzles,
575 but it is easily imaginable for a mountain river segment. Yet even then, the purpose of
576 the method in this study was to isolate an individual process, in this case the process
577 associated with variations in width and depth that drive flow convergence routing. Slope
578 patterns could be captured with a second functional classification and then merged with
579 this one, as envisioned in the conceptualization in the study purpose section.

580

581 Nesting reveals new understanding

582

583 Among the diverse results in this study, one particularly novel outcome is that the
584 focus on the structure of topographic heterogeneity revealed that base flow and bankfull
585 channel landforms are in fact organized together as dictated by the topographic steering
586 of the landforms at the floodprone scale (Table 7). Specifically, bankfull wide bars and
587 constricted pools are preferentially nested in floodprone wide bars and constricted
588 pools, respectively. The expansion or constriction of the floodprone region controls
589 erosion and deposition of the bankfull channel during modest floods. It is also important
590 that during in-channel and bankfull flows on the LYR, 2D modeling predicts that gravel
591 and cobble substrates are largely immobile (except at knickpoints), so the notion that a
592 two-stage self-maintenance mechanism can be in play is refuted in this case. These

593 model findings are consistent with our extensive field observations of the river during
594 such flows over the last 15 years. Some lateral migration of locally over-steepened,
595 noncohesive banks has been observed in TBR at bankfull or lower flows, but the
596 volumes are comparatively small. For other rivers with much smaller substrate sizes
597 and significantly higher Shields stresses, a two-stage mechanism may be plausible.
598 Nevertheless, on the LYR base flow and bankfull channel geometries, especially their
599 structured nonuniformity, are set during floods. Recent topographic change detection
600 and analysis studies (Wyrick and Pasternack, 2015; Pasternack and Wyrick, 2016)
601 showed that during the epoch from 1999 to 2008, a series of high-magnitude, long-
602 duration floods, including one instantaneous peak of ~ 23 times bankfull discharge,
603 drove rejuvenation of the diversity of in-channel morphological units at the subwidth
604 spatial scale. Overall, multiple studies looking at hydraulics, topographic structure, and
605 morphodynamics now provide a coherent conceptualization of the importance of
606 hierarchical topographic structure in controlling how the LYR functions.

607

608 **Conclusions**

609

610 This study answered 16 scientific questions organized under four objectives. At the
611 highest level this study showed that it is possible to analyze the hierarchical nesting of
612 landforms in a data-rich setting to provide a scale-independent typology that yields a
613 unified conceptualization of morphodynamics over a wide range of scales. Given that
614 capability, this study found that base flow and bankfull channel landforms are most likely
615 structured not by their own flow-dependent interactions, but by an overarching role for

616 valley-scale topographic steering of large floods that occur roughly every decade or less
617 frequently. This finding has important implications for professional practices in river
618 management and engineering, because practitioners must now look beyond the bankfull
619 channel to obtain self-sustainable riverine landforms within the channel.

620

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622

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632

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786

787 Table Captions

788 Table 1. Scientific analysis framework for this study applied to whole river segment and
789 each geomorphic reach.

790 Table 2. Metrics for topographic variability. Dark and light shading indicate highest and
791 lowest values, respectively.

792 Table 3. Results of $W_s \cdot Z_s$ analysis. Dark and light shading indicate highest and lowest
793 values, respectively.

794 Table 4. Analysis of landform composition as a function of flow.

795 Table 5. Longitudinal sequencing of landforms for the whole river, excluding normal
796 channel units.

797 Table 6. Top five permutations of hierarchical nesting of flow convergence routing
798 landforms within the five floodprone landform types.

799 Table 7. Top three permutations of hierarchical nesting of bankfull landforms, either
800 within (A-E) or beyond (F-J) them.

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803 Figure Captions

804

805 Figure 1. Location map of the Lower Yuba River (LYR) showing the geomorphic
806 reaches, the key factor indicating a reach break, gaging stations, and other
807 features of interest superimposed over the water surface area map for a flow of
808 $1194.97 \text{ m}^3/\text{s}$. The reach acronyms stand for Marysville Reach (MR), Hallwood
809 Reach (HR), Daguerre Point Dam Reach (DPDR), Dry Creek Reach (DCR),
810 Parks Bar Reach (PBR), Timbuctoo Bend Reach (TBR), Narrows Reach (NR),
811 and Englebright Dam Reach (EDR).

812 Figure 2. Longitudinal series of Zs and Ws.

813 Figure 3. Changes in Ws series with increasing discharge from (A) Q_{bf} to (B) $4.22 \cdot Q_{bf}$ to
814 (C) $8.44 \cdot Q_{bf}$.

815 Figure 4. Pearson's product-moment correlation values for Ws and Zs, stratified by
816 reach and flow.

817 Figure 5. Series of Ws·Zs for three flows with colors representing landform type.

818 Figure 6. Landform-stratified 2D-model velocity results, (A) mean and (B) 95th
819 percentile.

820 Figure 7. Examples illustrating (A) simple and (B), (C) complex velocity patterns at the
821 indicated flows.

822 Figure 8. Idealized two-level nested channel illustrating a typical nesting scenario.

823 Figure 9. Sample map showing the scale-independent landforms at a location where (A)
824 base flow nozzles (N, red patches) are nested within (B) bankfull wide bar units
825 (WB, orange patches), which in turn are nested within bankfull oversized (O,
826 black patches) and mixed normal channel-wide bar units (NC/WB, grey-orange
827 patches). Flow is from right to left.

828 Figure 10. Sample map showing the scale-independent landforms at a location where
829 (A) base flow nozzles are nested within (B) bankfull nozzles, but the bankfull
830 nozzles are nested in (C) three different floodprone landforms- constricted pool,
831 nozzle, and normal channel. Legend is the same as the previous figure. Flow is
832 from right to left.



















