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1 Revealing the natural complexity of fluvial morphology through 2D hydrodynamic
2 delineation of river landforms

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17

18 **Abstract**

19

20 Fluvial landforms at the morphological-unit scale (~ 1-10 channel widths) are
21 typically delineated and mapped either by breaking up the one-dimensional longitudinal
22 profile with no accounting of lateral variations or by manually classifying surface water
23 patterns and two-dimensional areal extents *in situ* or with aerial imagery. Mapping
24 errors arise from user subjectivity, varying surface water patterns when the same area
25 is observed at different discharges and viewpoints, and difficulty in creating a complete
26 map with no gaps or overlaps in delineated polygons. This study presents a new theory
27 for delineating and mapping channel landforms at the morphological-unit scale that
28 eliminates in-field subjective decision making, adds full transparency for map users, and
29 enables future systemic alterations without having to remap in the field. Delineation is
30 accomplished through a few basic steps. First, near-census topographic and
31 bathymetric data are used in a two-dimensional hydrodynamic model to create meter-
32 scale depth and velocity rasters for a representative base flow. Second, expert
33 judgment and local knowledge determine the number and nomenclature of landform
34 types as well as the range of base flow depth and velocity over each type. This step
35 does require subjectivity, but it is transparent and adjustable at any time. Third, the
36 hydraulic landform classification is applied to hydraulic rasters to quickly, completely,
37 and objectively map the planform pattern of laterally explicit landforms. Application of
38 this theory will reveal the true natural complexity, yet systematic organization, of
39 channel morphology.

40

41 *Keywords:* morphological units; fluvial landforms; fluvial geomorphology; 2D modeling

42 1. Introduction

43 Geomorphic analyses involve mapping the shape of landforms to describe their
44 spatial patterns, observing landforms over time to record their changes, exploring the
45 drivers and mechanisms of landform change, and evaluating the responses of
46 biological, chemical, and hydrological processes to morphologic changes. A common
47 practice in fluvial geomorphology involves focusing on specific spatial scales at which
48 landforms have characteristic features (Grant et al., 1990; Rosgen, 1996; Thomson et
49 al., 2001). These scales are often thought of as dimensionless (i.e., exhibiting similarity
50 of forms and processes among systems of different absolute size) and proportional to
51 channel width (W), with common names such as catchment (entire watershed scale),
52 subcatchment, segment ($\sim 10^3$ - $10^4 W$), reach ($\sim 10^2$ - $10^3 W$), morphological (alternately
53 channel or geomorphic) unit ($\sim 10^0$ - $10^1 W$), and hydraulic unit ($\sim 10^{-1}$ - $10^0 W$) (Frissell et
54 al., 1986; Grant et al., 1990; Bisson et al., 1996; McDowell, 2001). This study presents a
55 new theory and methodology for delineating and mapping channel landforms at the
56 morphological-unit scale that eliminates in-field subjective decision making, adds full
57 transparency for map users, and enables future systemic alterations without having to
58 remap in the field.

59

60 1.1. MU definition

61 There are several terms for discernible units of channel morphology at the ~ 1 - $10 W$
62 scale, such as *channel unit* (e.g., Grant et al., 1990; Bisson et al., 1996), *channel*
63 *geomorphic unit* (e.g., Hawkins et al., 1993), *geomorphic unit* (e.g., Thomson et al.,
64 2001), *morphological unit* (e.g., Wadeson, 1994; Moir and Pasternack, 2008), and

65 *physical biotope* (e.g., Newson and Newson, 2000). The terms that begin with *channel*
66 preclude their usage for overbank landforms, which therefore can be more specific than
67 desired when considering the river corridor as a continuum. *Biotope* imposes a
68 biological requirement that may not be applicable or necessary for geomorphic analysis.
69 *Geomorphic unit* is a likely term, but is broadly used across all spatial scales and is not
70 limited to landform geometry. The term of choice for this study is *morphological unit*
71 (MU), which provides an appropriately descriptive term for topographic forms within the
72 river corridor that represent distinct form-process associations.

73 River topography is a continuous form, so to an extent the idea of breaking it down
74 into discrete units may seem artificial and arbitrary (Kondolf, 1995). However, we have
75 long understood that different landscape elements are responsible for different physical
76 processes and biological functions, so it is worthwhile to explore MUs in more detail and
77 with more objectivity than has been attempted before.

78 At the scale of ~ 1-10 W, MUs are conjectured to be a basic unit for understanding
79 physical processes and assessing instream habitat considering that ecohydraulic
80 variables such as depth, velocity, shear stress, and substrate are closely controlled by
81 the shape and structure of the landform over which they occur (Whiting and Dietrich,
82 1991; Newson and Newson, 2000; Thompson, 2006; Sawyer et al., 2010). The MU
83 scale therefore provides a relatively high degree of resolution of analysis that balances
84 scientific detail with the potential for segment-scale application (Padmore et al., 1998).
85 Since many studies have subjectively defined MUs and/or habitat types and then used
86 those classifications to make important geomorphic and ecological observations, river
87 scientists obviously find this spatial-scale delineation a valuable tool.

88 Notably an MU is not a habitat or biotic object or concept. Habitat at the mesoscale
89 is defined as the interdependent set of the ecohydraulic variables over an MU that
90 attracts organisms to reside there for a significant part of the day (e.g., Beisel et al.,
91 1998; Parasiewicz, 2007). The MUs can be revealed by their overlying hydraulics (see
92 section 1.2), but they are not an assemblage of flow-dependent hydraulic conditions;
93 thus, they do not change their spatial pattern as discharge changes (excluding scour
94 and fill). The MUs constitute a classification of the *landforms* that create key
95 environmental requirements of an aquatic community.

96 A key advancement for MU mapping is the trend toward performing spatially explicit,
97 detailed, planform mapping. Traditional sampling of rivers with a small number of cross
98 sections suffers from many problems (Pasternack and Senter, 2011), including biased
99 preconceptions as to which locations are more important, stable, accessible, or
100 representative. A census is a complete accounting of a population; but when
101 considering a continuous spatial variable like topography, there are ever-finer scales of
102 variability precluding a true census. Pasternack (2011) coined the term *near-census* to
103 refer to comprehensive, spatially explicit observation of the landscape emphasizing the
104 ~ 1-m scale as the basic building block for characterizing geomorphic processes and
105 ecological functions.

106

107 *1.2. Hydraulic MU delineation*

108 The morphology of channel beds and banks impacts overlying flow hydraulics (e.g.,
109 Whiting and Dietrich, 1991; Clifford and French, 1998; MacWilliams et al., 2006;
110 Pasternack et al., 2006), so hydraulics can in turn be used as a proxy for identifying

111 underlying MUs. In fact, most recent methods for delineating MUs are based on
112 categorizing a suite of local hydraulic combinations between fast or slow velocity with
113 deep or shallow depths (e.g., Hawkins et al., 1993; Borsanyi et al., 2004; Zimmer and
114 Power, 2006; Hauer et al., 2009; Klaar et al., 2009). The main differences among these
115 methods arise from how they determine local hydraulics, at what spatial area to apply
116 them, and how to locate MU boundaries. Most of these studies focused on qualitative
117 observations of surface flow patterns, surface water slope, and/or localized point
118 measurements or estimations of depth and velocity. A similar pattern of hierarchical
119 decisions about the use of flow hydraulics has emerged. Typically, the user decides in
120 the field whether some area exhibits fast or slow velocity, then whether the water
121 column is deep or shallow, which then leads to a mesohabitat unit description. The user
122 also subjectively delineates the unit boundaries. Mapping subjectivity is accepted in
123 peer review for lack of any objective methodology.

124 This subjective MU delineation method, however, has several deficiencies. First, a
125 field observer at ground level may have limited and insufficient vantage points to
126 observe the necessary hydraulics. Second, decisions may be improperly influenced by
127 conditions at time of measurement. Hydraulic thresholds for MU boundaries are often
128 not visible to the human eye. Third, visual qualitative observation of the magnitudes of
129 depth and velocity suffers from the same types of problems reported for visual substrate
130 and other classifications (Jowett, 1993; Marcus et al., 1995; Bunte and Abt, 2001;
131 Faustini and Kaufmann, 2007) in that: (i) individual observers may visually distinguish
132 areas with dramatically different hydraulics, but are unlikely to visually distinguish less
133 dramatic differences, (ii) individual observers tend to overvalue large magnitudes, (iii)

134 individual observers looking at the same magnitude but in different surrounding contexts
135 may experience optical illusions that lead to mischaracterizing the same magnitudes as
136 different, and (iv) different observers may look at the same location and report different
137 magnitudes. Fourth, the subjective delineation of spatial patterns will suffer from the
138 same types of problems as enumerated for estimating magnitudes, yielding unreliable,
139 nonrepeatable interpretations. Fifth, spatial patterns are commonly mapped as polygons
140 with a handheld GPS (with real-time or post-processed differential positions) whose
141 nominal precision is submeter, but whose true accuracy when operated while moving
142 (lacking repeated counts at each vertex) is often unchecked and actually poor (~ 2-5 m).
143 The accuracy of GPS polygon delineation is poor enough that lines may cross over and
144 thus requires that individual polygons be corrected later. Finally, field-delineated
145 polygons are not snapped, leaving gaps and overlaps that are difficult to reconcile.

146 Several researchers have enhanced the hydraulic delineation of MUs through the
147 use of digital elevation models (DEMs), which serve to reduce field subjectivity and
148 allow for repeatability of morphologic delineation methods. Near-census topographic
149 and bathymetric data collected using light detection and ranging (LiDAR) and
150 echosounding provide more robust data sets and quantitative terrain and hydraulic
151 metrics for mapping MUs. For example, Milne and Sear (1997) used ArcGIS to detrend
152 river DEMs based on cross-sectional surveys of several upland rivers and then used
153 elevational variations of the bed surfaces to differentiate between pools and riffles, i.e.,
154 using depth as the sole MU indicator. However, depth is an inadequate indicator when
155 used alone because it cannot distinguish between two landforms with the same depth
156 but significantly different bed slopes and bed roughnesses that yield different velocities

157 and shear stresses. Moir and Pasternack (2008) mapped a suite of laterally and
158 longitudinally explicit MUs based on expert judgment of hydraulics and substrate using
159 site visits as well as a 1-m resolution topographic map to hand-delineate MU polygons
160 in ArcGIS. Hauer et al. (2009) combined LiDAR and terrestrial survey data to create a
161 DEM of a gravel-bed river and simulated a range of discharges using a two-dimensional
162 (2D) hydrodynamic model. They then used an algorithm to map six types of
163 mesohabitat regions within this range of discharges based on binned values of velocity,
164 depth, and shear stress. Their study provides a template for repeatability but is focused
165 on flow-dependent mesohabitats, not MUs.

166

167 *1.3. Topographic MU delineation*

168 Ideally, both MUs and flow-dependent mesohabitats should be delineated objectively
169 without spatial interpretation by observers. Some studies have argued that the way to
170 achieve this is to forgo flow-based indicators and only use terrain. O'Neill and Abrahams
171 (1984) determined riffle crests and pool troughs using the one-dimensional channel-bed
172 longitudinal profile. They argued that any method involving depth and velocity would be
173 inherently dependent on discharge, and therefore proposed the use of variances along
174 the topographic slope for MU delineation. However, this method is not without
175 subjectivity either. The geomorphic community generally accepts longitudinal profiles
176 without questioning the process of obtaining them, but in fact this method of one-
177 dimensional MU mapping involves a process of picking and sampling pathways that
178 includes several opaque assumptions and choices lacking objectivity, transparency, or
179 justification. Most importantly, the geometry of a channel's fastest, deepest pathway and

180 the means by which geomorphologists locate it are both flow-dependent; so too is the
181 geometry of the centerline of the wetted area. In theory, a pathway connecting the
182 deepest points down a channel might not be flow dependent, but its location in the field
183 is difficult to identify without being influenced by the hydraulics on the day of
184 observation. Yet the longitudinal profile and subsequent MU delineation are determined
185 for a single discharge, which is often arbitrarily the one that happens to occur on the day
186 of summer field work, which in turn is usually chosen to be a wadable low flow or picked
187 for other logistic reasons instead of scientific ones. Further, the choice of variance
188 thresholds is subjective as to how much topographic high or low makes a riffle or pool
189 (as admitted in O'Neill and Abrahams (1984)), but once the flow, profile, and criteria are
190 set, the mapping algorithm to locate riffles and pools is objective. Additionally, there are
191 more MUs than just riffles and pools, and most rivers contain significant lateral and
192 oblique terrain variability that cannot accurately be captured by cross sections and
193 profiles (e.g., Borsanyi et al., 2004; Milan et al., 2010).

194 The greatest degree of objectivity and accuracy could be achieved if near-census
195 river corridor DEMs were objectively analyzed to delineate a full suite of individual MUs.
196 Terrain segmentation based on topographic slopes, aspects, and/or curvatures is not a
197 new concept in geomorphology (e.g., Waters, 1958; Brandli, 1996). In fact, MU-scale
198 mapping can be considered similar to the elementary forms concept proposed by Minar
199 and Evans (2008), in which units were defined by third-order slope equations. However,
200 analytical terrain segmentation does not include flow direction, and flow direction at any
201 given point in a river is not necessarily parallel to the downvalley channel slope. Hence,
202 the use of these topographic equations to delineate MUs could result in incorrectly

203 assigning an MU classification based on the assumption of linear, longitudinal flow.
204 Another significant hurdle in topographic delineation of MUs arises from the fact that
205 different MUs may have very similar nondimensional geometry, but they have subtly or
206 significantly different dimensional scales. Depending on the native resolution of the data
207 and different approaches to detrending and filtering, MUs may be revealed or obscured,
208 necessitating subjective interpretation and manipulation. New technologies may solve
209 these challenges in the future, but a strong basis still exists for taking advantage of the
210 innate connection between landforms and their overflowing hydraulics as an objective
211 method to map MUs.

212

213 *1.4. Study objectives*

214 Pure topographic analysis of landforms remains problematic, and past efforts at
215 hydraulic-based MU delineation have either lacked enough detail to capture meaningful
216 variations or emphasized mesohabitat instead of MU delineation. However, the previous
217 studies have provided templates and guidelines for this next logical step in creating a
218 complete MU coverage map. In this study, an objective map of MUs was found to be
219 obtainable from two inputs: (i) spatial grids of depth and velocity at a low discharge
220 (when topography is the primary control on hydraulics) estimated using a 2D
221 hydrodynamic model, and (ii) an expert-specified MU classification scheme using depth
222 and velocity values. With these inputs, one can objectively classify any location in a
223 river into an MU type and then identify coherent MUs as adjacent aggregates of
224 individually classified points. This study introduces a theory and methodology that
225 removes much of the subjectivity in mapping river MUs and presents the concepts and

226 justifications for spatially explicit delineation of MUs aided by 2D hydrodynamic
227 modeling. Near-census data and model results enable representation of all areas of the
228 wetted channel with equal emphasis and objectivity, and as such will yield unambiguous
229 and comprehensive MU results.

230

231 **2. MU mapping methodology**

232 A six-step procedure for mapping river MUs was developed in this study (Fig. 1),
233 with basic steps outlined in this paragraph followed by detailed information for each step
234 in the following subsections. First, obtain near-census topographic and bathymetric data
235 of the river corridor of interest and produce a DEM. Second, use expert judgment and
236 local knowledge (perhaps guided from observations during data collection) to
237 predetermine the number and nomenclature of MU types to be mapped, and then
238 estimate the range of each hydraulic variable for each MU type. Codify hydraulic
239 thresholds into an algorithm for classifying individual raster cells. Third, use
240 hydrogeomorphic processes and/or ecologic functions to determine an appropriate low
241 flow regime at which to identify the MUs. Fourth, develop, run, and validate a 2D
242 hydrodynamic model at the flow of relevance for MU delineation of the inundation zone
243 to be mapped. Fifth, create rasters of the key delineation variables (presently taken to
244 be depth and velocity, but future developments could also draw on Froude number,
245 Shields stress, or other derivative variables) consistent with the resolution of the 2D
246 model. Sixth, apply the MU delineation algorithm to obtain a preliminary MU map.
247 Finally, we recommend a review of the map to evaluate if the predetermined MU types
248 and thresholds yield meaningful patterns. Tests exist that can be used to evaluate the

249 spatial organization of MUs (Pasternack and Senter, 2011; Wyrick and Pasternack,
250 2012), but there is also a risk of circularity if the existence or nonexistence of coherent
251 MU spatial organization is used to modify the MU algorithm, and in turn, the same tests
252 are used to subsequently demonstrate the existence of spatial organization. Overall, the
253 proposed methodology eliminates subjectivity in assessing the magnitude of hydraulics
254 and the resulting spatial pattern, leaving the choice of number and nomenclature of MU
255 types as well as the ranges of joint depth and velocity magnitudes for each MU type as
256 the only subjective aspects. Any remaining subjectivity may be considered as a flaw, yet
257 so far no existing method is devoid of subjectivity. This new approach represents a
258 significant step forward in using 2D modeling results to develop objective criteria for
259 understanding the underlying landforms within a river corridor.

260

261 *2.1. Channel topography*

262 Looking beyond the era of fluvial geomorphology based on cross sections, a near-
263 census river corridor digital terrain model is the most important input for diverse
264 geomorphic, engineering, and ecological applications, including MU delineation
265 (Wheaton et al., 2004; Pasternack, 2011). Near-census data sets are obtained at
266 reasonable cost and are increasingly available for free (e.g.,
267 <http://www.opentopography.org>). The preferred methods at this time are airborne LiDAR
268 mapping of the terrestrial river corridor (Lane and Chandler, 2003; Hilldale and Raff,
269 2007) and boat-based echosounding of the subaqueous riverbed (single- or multi-beam
270 depending on depth). These methods typically have high point densities (~ 0.5 to 3
271 points per m²). Where remote methods are ineffective (e.g., shallow, wadable,

272 submerged areas; submerged areas with excessive bubbles; and terrestrial forests with
273 inadequate canopy openings), a combination of Real-Time Kinematic Global Positioning
274 System (RTK GPS) and Total Station (TS) surveys are recommended. Spatial sampling
275 may aim for maximal point density commensurate with channel type (Brasington et al.,
276 2000; Valle and Pasternack, 2006), emphasize key features and slope breaks (e.g.,
277 Bouwes et al., 2011), or do both (e.g., Pasternack, 2011). Each method and
278 interpolation scheme has unique, inherent uncertainties that need to be assessed and
279 reported (Milan et al., 2011) in order to provide full disclosure of steps taken to apply
280 high standards for quality of data used for all other analyses.

281

282 *2.2. Discharge selection*

283 A choice that must be made is the discharge to use in the 2D model for an accurate
284 MU delineation. Such a choice is inherent in almost every MU delineation method,
285 including those only analyzing the topography of the thalweg profile or wetted area
286 centerline, but this choice is often hidden and denied. If the flow is too low, especially for
287 a channel with gently sloping banks, then too little of the channel will have identifiable
288 hydraulics. If the flow becomes too high, then the momentum of the water will increase
289 enough that some topographic controls will be effectively drowned out (Pasternack et
290 al., 2006; Wyrick and Pasternack, 2008), and the resulting hydraulics will have
291 decreased spatial variation. The inherent self-maintenance of most channels results in a
292 morphology that is at quasi-equilibrium for all but flood flows, but manifests most clearly
293 at the low flows (Langbein and Leopold, 1962).

294 The choice of which low flow to use in the model is not that sensitive and is aided by
295 available discharge records along with flow indicators of hydrogeomorphic processes
296 and/or ecological functions. One option is to rely on a hydrological process, such as
297 base flow. Base flow is generally defined as the average annual low flow discharge that
298 exists for some measurable extended time period (i.e., not an instantaneous
299 measurement). Another option is to reference against a flow responsible for channel
300 maintenance, such as bankfull discharge. Based on experience thus far, a flow of $\sim 1/10$
301 to $1/5$ of bankfull discharge is recommended. A third option is to identify key low flows
302 for ecological functions such as anadromous salmonid migration or spawning. Finally,
303 an iterative process with sensitivity analyses may be used to compare and contrast
304 alternatives and quantify uncertainty (Wyrick and Pasternack, 2012).

305 The hydraulics over an MU change with discharge, but it is important to keep in mind
306 that the landform is what is being mapped with this method. Therefore, selection of the
307 'ideal' discharge to model is ultimately less important because for any selected
308 discharge a particular MU will have a specific depth-velocity combination (see section
309 2.4 for more details) that must be recognized and implemented into the methodology. In
310 other words, use of a lower or higher flow for MU mapping yields virtually no difference
311 in MUs because the hydraulic thresholds are adjusted down or up, respectively (Wyrick
312 and Pasternack, 2012). The resilience of MU delineations across discharges by
313 adjusting hydraulic thresholds is key evidence that this methodology is revealing
314 underlying landforms that are independent of discharge.

315

316 *2.3. 2D hydrodynamic modeling*

317 Two-dimensional (depth-averaged) hydrodynamic models have existed for decades
318 and are increasingly used to study a variety of hydrogeomorphic processes (Bates et
319 al., 1992; Leclerc et al., 1995; Miller and Cluer, 1998; Cao et al., 2003; Brown and
320 Pasternack, 2008; Sawyer et al., 2010) and to perform quantitative habitat assessments
321 (e.g., Leclerc et al., 1995; Elkins et al., 2007; Moir and Pasternack, 2010; Bouwes et al.,
322 2011). Notably, these previous studies were generally limited to short river areas, ~ 50
323 to 2000 m of channel length. While such distances may be adequate to reveal local
324 processes and test site-scale project designs, it is not adequate for comprehensive
325 instream flow analysis of a river segment (i.e., 10^3 - 10^4 W). As mapping and modeling
326 technology has progressed, 2D modeling is emerging as a preferred tool for near-
327 census river analysis. A recent textbook on 2D modeling presents the requisite inputs,
328 methods, and some applications of simulation outputs for fluvial geomorphology and
329 habitat assessment (Pasternack, 2011). The selection of a specific algorithm is not
330 important for the MU methodology reported in this study, as long as the model can
331 discern the hydraulic phenomena present in the study segment.

332 Results from any 2D model need to be converted to raster format for use with this
333 methodology. The output from a 2D model is often a point-based text or binary file with
334 point coordinates and the values of hydraulic variables at those coordinates. Depending
335 on the 2D model procedure used and point density, the user should select an
336 appropriate method (e.g., Delaunay triangulation, kriging, or nearest neighbor) for
337 converting the point results to a raster (Moore et al., 1991; Pasternack, 2011).

338

339 2.4. MU classification

340 Given a 2D model simulation of the spatial pattern of base-flow depth and velocity in
341 a river, the key step in MU delineation involves assigning each point's joint velocity and
342 depth combination to an MU type. To do this, one must already have a basic knowledge
343 of which MU types are relevant to the study region and what range of hydraulics are
344 likely to be associated with each MU type for the selected flow. This knowledge can
345 come from the literature on channel types and MUs, past regional studies, and/or
346 experience with the study area. Ideally, experts with different fluvial educations and
347 experiences would reach a consensus as to what fluvial landforms are potentially
348 present at the ~ 1-10 W scale. A strength of this new theory and methodology is that it
349 forces this key step into public discourse with transparency, whereas traditional
350 methods rely on experts to make these decisions *in situ* on the river with no chance for
351 future adjustment or adequate explanation.

352 A spectrum of MU and/or mesohabitat terminology and definitions exists that can
353 guide users in assessing what is relevant and meaningful for MU delineation for a new
354 study region (e.g., Grant et al., 1990; Hawkins et al., 1993; Brierly and Fryirs, 2000;
355 Milan et al., 2010). Existing terminologies have qualitative definitions that are generally
356 consistent throughout geomorphic literature, so quantitative delineations of MUs should
357 be appropriately grounded to these broadly accepted qualitative definitions. For
358 example, countless articles have been published assessing forms and processes of the
359 MU types known as *pool* (i.e., topographic low with deep, slow, subcritical hydraulics)
360 and *riffle* (i.e., topographic high with shallow, fast, near-critical hydraulics). Classically,
361 some fluvial geomorphologists only recognized pools and riffles, especially when relying

362 on a longitudinal profile for MU delineation. In the last ~ 20 years, however, a growing
363 number of studies have defined an increasingly large number of MU or mesohabitat
364 types. For example, McCain et al. (1990) listed 22 in-channel habitat types. Hawkins et
365 al. (1993) identified 18 channel unit types (seven fast water units and eleven slow water
366 ones). Brierly and Fryirs (2000) catalogued 12 different types of bank-attached
367 morphological units alone. Brown (1997) described 17 different types of floodplain
368 features. Although these diverse schemes have received limited objective scrutiny or
369 comparison, their application in river management has yielded significant statistical
370 associations with physical variables and biological observations. The purpose of this
371 study is not to question or justify any specific number of MUs for any particular purpose,
372 but instead to present a method for mapping diverse landforms as objectively as
373 possible for those who already accept that such diversity exists.

374 For an example of how to use classic definitions to classify MUs, consider some
375 common in-channel morphological units such as: pool, riffle, glide, and run (see more
376 comprehensive compilations in Grant et al. (1990), Newson and Newson (2000), or
377 Milan et al. (2010)). Descriptive and relevant definitions of each can be gleaned from
378 literature (Table 1). From these qualitative definitions, they can be sorted by ranges of
379 associated hydraulics relevant to the applied river system as a starting point. Typically
380 the topographic endmembers, pool and riffle, are succinctly defined for low flow
381 hydraulics as *deep and slow* and *shallow and fast*, respectively. Glides tend to be
382 defined as shallow and slow, while runs tend to be defined as deep and fast. Therefore,
383 a simple four-type classification can be created with the subjective choice being exactly
384 which depth and velocity values to use as hydraulic thresholds. From there, the number

385 of MU types and their respective hydraulic thresholds can be tailored to be more
386 specific to the river of interest and may include additional MU types with their own
387 depth–velocity ranges, such as chute, cascade, riffle transition, etc. The nomenclature
388 is less important than the ability to identify coherent landforms that exhibit similar
389 hydraulics.

390 Once the number and definitions of MU types are set, the next step is to assign
391 quantitative depth and velocity thresholds to delineate them at the relevant discharge of
392 the 2D model run. For those who prefer considering rivers as a continuum rather than
393 an assemblage of discrete MUs, the threshold uncertainty may optionally be addressed
394 using a fuzzy inference system in which a lower probability of being in an MU is
395 assigned on the basis of higher proximity to a threshold (e.g., Legleiter and Goodchild,
396 2005). Such a fuzzy inference system could also be used to cope with the effect of
397 uncertainty in 2D model estimation accuracy on MU designation.

398 Initial estimates of hydraulic thresholds for MU delineation come from the literature
399 on channel types and MUs, past regional studies, and experience with the study area.
400 Like the numerical thresholds in any landform classification, these thresholds are
401 arguably subjectively chosen, but the resulting map is objective because neither the
402 spatial pattern nor assignment of MU types to points is subjective. Further, this scheme
403 means that the subjective aspects still inherent to the methodology are fully transparent
404 and adjustable, whereas the suite of individual field-based choices cannot be fully
405 explained nor adjusted later without a high degree of uncertainty.

406 By way of comparison, this approach of assigning thresholds has some similarity to
407 supervised cluster analysis used to classify and interpret spatial patterns (e.g., Johnson,

408 1967; Maxwell et al., 2002). In that method, the number of units and an initial estimate
409 (seed) of the middle of each unit's hydraulic domain (strictly speaking, the centroid of
410 each n -dimensional phase-space unit where depth and velocity constitute a 2D phase
411 space) are selected by the user. All points are then assigned to the nearest seed and
412 the centroid of points in each unit is computed to yield an adjusted middle. This process
413 is repeated until the centroid value no longer changes. Then the points are assigned to
414 each unit centroid and boundaries delineating final units are inferred based on the point
415 classification.

416 The use of supervised cluster analysis for MU delineation suffers from two significant
417 drawbacks relative to the method developed in this study. First, the outcome would be
418 an array of units based on clusters that were biased by design to have a large
419 abundance and density of points, which is different from having units based on the
420 uniqueness of the joint distribution of individual, disparate hydraulics values associated
421 with underlying landforms. Second, the number of raster cells in an MU delineation
422 would be proportional to the total area of each MU type in the study domain, and this
423 would impact MU delineation. For example, one MU type with a lot of cells might draw
424 the attention of multiple seeds and be subdivided unnecessarily; whereas a rare MU
425 type with a distinct joint distribution of depth and velocity may be real and meaningful,
426 but might end up subsumed into one or more other clusters and not be revealed
427 because of low numbers of points. In other words, sub-MU scale features may possibly
428 yield complex joint distributions of depth and velocity distributions that are meaningful at
429 the smaller hydraulic unit scale but are not similarly appropriate for MU-scale landform
430 delineation. This is an example where the topographic detail of near-census data could

431 confound proper landform mapping. Classifying by boundary values instead of centroids
432 guarantees that the hydraulic domain of each MU matches a distinct signature of each
433 landform.

434 This methodology was implemented on a lowland, gravel–cobble river (Wyrick and
435 Pasternack, 2012) and an upland, cobble–boulder river (Pasternack and Senter, 2011).
436 In each case the set of MUs was unique to the landscape setting on the basis of local
437 knowledge and geomorphic theory (Fig. 2). For more details on the river settings, MU
438 type definitions, and classification selection choices, please refer to the referenced
439 reports. These examples should not be adopted in future studies without mindful
440 consideration of their suitability in each case, but are presented here simply as visual
441 representations of hydraulic classification.

442

443 *2.5. MU mapping*

444 Given 2D model rasters for depth and velocity, a specified number of MU types, and
445 the hydraulic threshold values for each MU type, the last step is to objectively map
446 individual MUs. This is accomplished using a computer program, such as the raster
447 calculator in ArcGIS Spatial Analyst (see workflow in Pasternack, 2011). This
448 calculation automatically assigns each raster pixel to a particular MU type based on its
449 discrete values of depth and velocity. All contiguous pixels with the same classification
450 coalesce into a single MU polygon, thus providing automatic spatial delineation for the
451 river. As an additional step, one may choose to limit an individual MU to a minimum size
452 on the basis of spatial coherence testing, for which procedures already exist (e.g.,

453 Wheaton et al., 2010; Carley et al., 2012). Alternatively, size discrimination could be
454 applied after the fact for any subsequent MU analyses.

455 After inspection of the initial MU map, the number of MU types and their hydraulic
456 thresholds may be individually manipulated based on expert knowledge to assess the
457 sensitivity of the map to different choices. A visual review of the MU map will reveal
458 qualitative patterns that can be assessed for realism. The MUs should be organized in a
459 somewhat expected manner — in the longitudinal direction and within contiguous
460 nondirectional clusters. For example, the geometrically steepest and flattest landforms
461 would be expected to be separated by some transitionally sloped landforms. If
462 submeter-scale color aerial imagery is available, then one could visually compare the
463 MU map to the imagery to check the delineation of easily observed MUs. No formal
464 evaluation or improvement procedure exists at this time, but it would be feasible to run
465 optimization tests to determine the scheme that yields the MU map with the most
466 significant spatial organization metrics. In the meantime, user judgment based on local
467 experience, expert group consensus, and independent peer review are the best aids to
468 final selection, just as they are for traditional approaches to MU mapping.

469 Following the example application of hydraulic thresholds (Fig. 2), examples of how
470 the depth and velocity rasters can be used in concert with the hydraulic thresholds to
471 create detailed MU maps are illustrated for a lowland, gravel–cobble river (Fig. 3) and
472 an upland, cobble–boulder river (Fig. 4). While detailed analyses of these MU maps are
473 beyond the scope of this article, some basic results can be evaluated to highlight the
474 methodology's veracity and relevance. First, the maps show that the suite of landforms
475 completely covers the wetted area of the selected base flow, i.e., no overlaps or gaps

476 exist within the mesh of polygons as might occur with field-delineated maps. Second,
477 MU shapes are highly irregular, as might be expected from intricacies of channel
478 morphology. Such detail is generally difficult to replicate with hand drawings (e.g., Milne
479 and Sear, 1997; Borsanyi et al., 2004; Moir and Pasternack, 2008). Third, the channels
480 exhibit high lateral variability at any given cross section that may be lost in field
481 delineations. This point is particularly important for identifying slender units along the
482 margin that may not dominate any given cross section but are valuable for habitat
483 studies. Lastly, these example sites are only small sections of complete longitudinal
484 coverage maps that extend for ~ 37 km (Fig. 3) and ~ 12 km (Fig. 4), lengths that would
485 be onerous to hand-map at this resolution. These examples are provided not to highlight
486 specific hydraulic thresholds and MU combinations, but rather as templates as to how
487 the mapping process, tailored for any river system, would look.

488

489 **3. Applications**

490 Morphological unit maps provide insight into the geomorphic structure of the river
491 corridor. The generation of basic map statistics may also provide feedback and
492 refinement for the mapping process described in the previous sections. More
493 importantly, analyses of MUs can be used to address fundamental questions about the
494 structure and function of river landforms. Previous literature on landform delineation
495 (e.g., Grant et al., 1990; Hauer et al., 2009) provided four broad groups of MU analysis
496 metrics: abundance and diversity, longitudinal distribution and spacing, lateral
497 variability, and nondirectional adjacency. Because this article emphasizes theoretical
498 developments for MU mapping, these MU statistics were not developed herein for a

499 case study but have been applied with success to two diverse rivers thus far
500 (Pasternack and Senter, 2011; Wyrick and Pasternack, 2012). However, some example
501 scientific questions that could be addressed with a detailed MU map may include
502 whether (i) MUs are organized in a nonrandom, coherent spatial structure, (ii) a river
503 exhibits significant lateral variability, or (iii) MUs are organized across multiple spatial
504 scales.

505 A detailed MU map also provides a basis for stratification of ecohydraulic data sets
506 (e.g., Abu-Aly et al., in press). An example scientific question might be to determine
507 whether the rates of change for hydraulic variables as discharge increases can be
508 isolated for a particular MU type to determine locations of possible velocity reversals.
509 Or, in other words, at what discharge will pool units exhibit higher average velocities
510 than riffle units, if at all? With an MU map, the locations of various lifestage habitats can
511 be linked to the geomorphic variables. One could determine whether a relationship
512 exists between MU type (and/or size, number, location, etc.) and areas of salmonid
513 spawning or rearing. The applications of an accurate, detailed MU map are only
514 bounded by data set availability and users' imaginations.

515

516 **4. Conclusions**

517 Mesoscale fluvial landforms have been described as *fundamental building blocks of*
518 *rivers* (Brierly and Fryers, 2000) and have been inserted as important links within
519 channel classification hierarchies (e.g., Frissell et al., 1986; Newson and Newson,
520 2000). Thus, improved identification and delineation of these morphological units are
521 vital to the progress of river science. The methodology presented in this study increases

522 the level of objectivity in the mapping procedure and provides a basis for streamlined,
523 repeatable, and rigorous classification within any river system.

524 This study presented several key advances to the science of river morphology
525 delineation. First, an MU is a flow-independent, structural landform; and identification of
526 the landform's morphology is important for defining the MU. Second, 2D hydrodynamic
527 results were used as a basis for identifying and delineating MUs, which provide the
528 means to create a continuous map in the context of any spatial scale. Third, the ability
529 to manipulate the delineation procedure digitally allows for a repeatable and more
530 objective methodology of MU mapping. Fourth, the robustness of the methodology is
531 such that imprecision on which low flow discharge to use in the procedure does not add
532 uncertainty to the final MU maps. Lastly, digital delineation can return results that are
533 scaled to pixel sizes smaller than what field methods produce, therefore creating maps
534 that are more detailed and ultimately more accurate than large scale averaging.

535

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552

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Table 1

Descriptions of MUs common to gravel-bed rivers; descriptions of depth and velocity refer to those typically created by the landforms during low flows

MU	Description at low flow
pool	Topographic low in the channel that exhibits high depth and low velocity, and low water surface slope. This unit covers both 'forced pool' and 'pool'. A forced pool is one that is typically along the periphery of the channel and is 'over-deepened' from local convective acceleration and scour during floods that is often associated with static structures such as wood, boulders, and bedrock outcrops. A pool is not formed by a forcing obstruction. The distinction between forced pool and pool cannot be made automatically within GIS.
riffle	Topographic high that exhibits shallow depths, moderate to high velocities, rough water surface texture, and steep water surface slope. Riffles are generally associated with the crest and backslope of a transverse bar (e.g., Knighton, 1998).
run	An area that exhibits moderate to high velocities, high depths, and moderate water surface slope. Runs typically occur in straight sections that exhibit a moderate water surface texture and tend not to be located over transverse bars.
glide	An area that exhibits low to moderate velocities and depths and low water surface slope. Glides commonly occur along the periphery of channels and flanking pools and can also exist in straight sections of low bed slope.

742

743

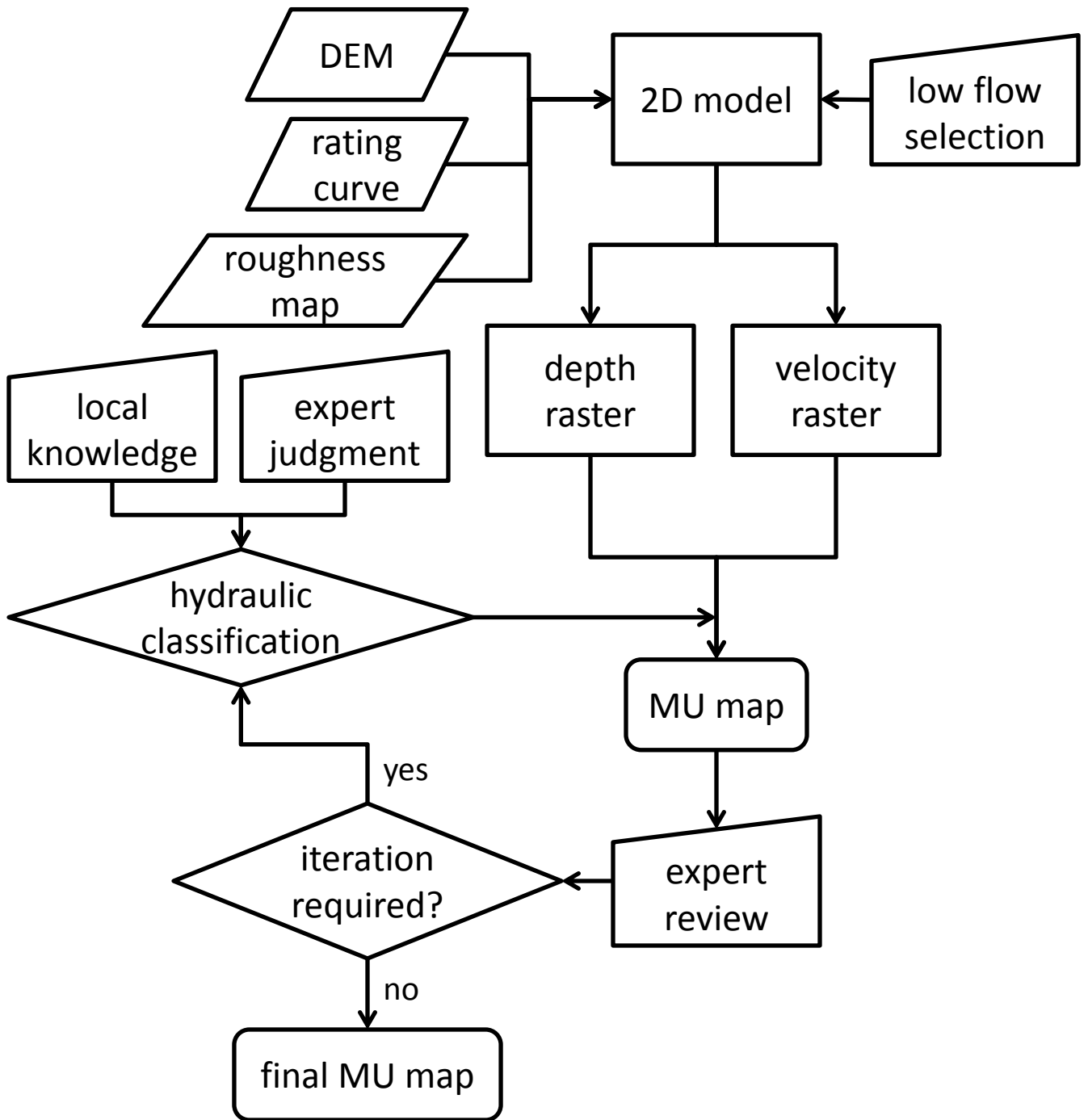
744 **Figure Caption**

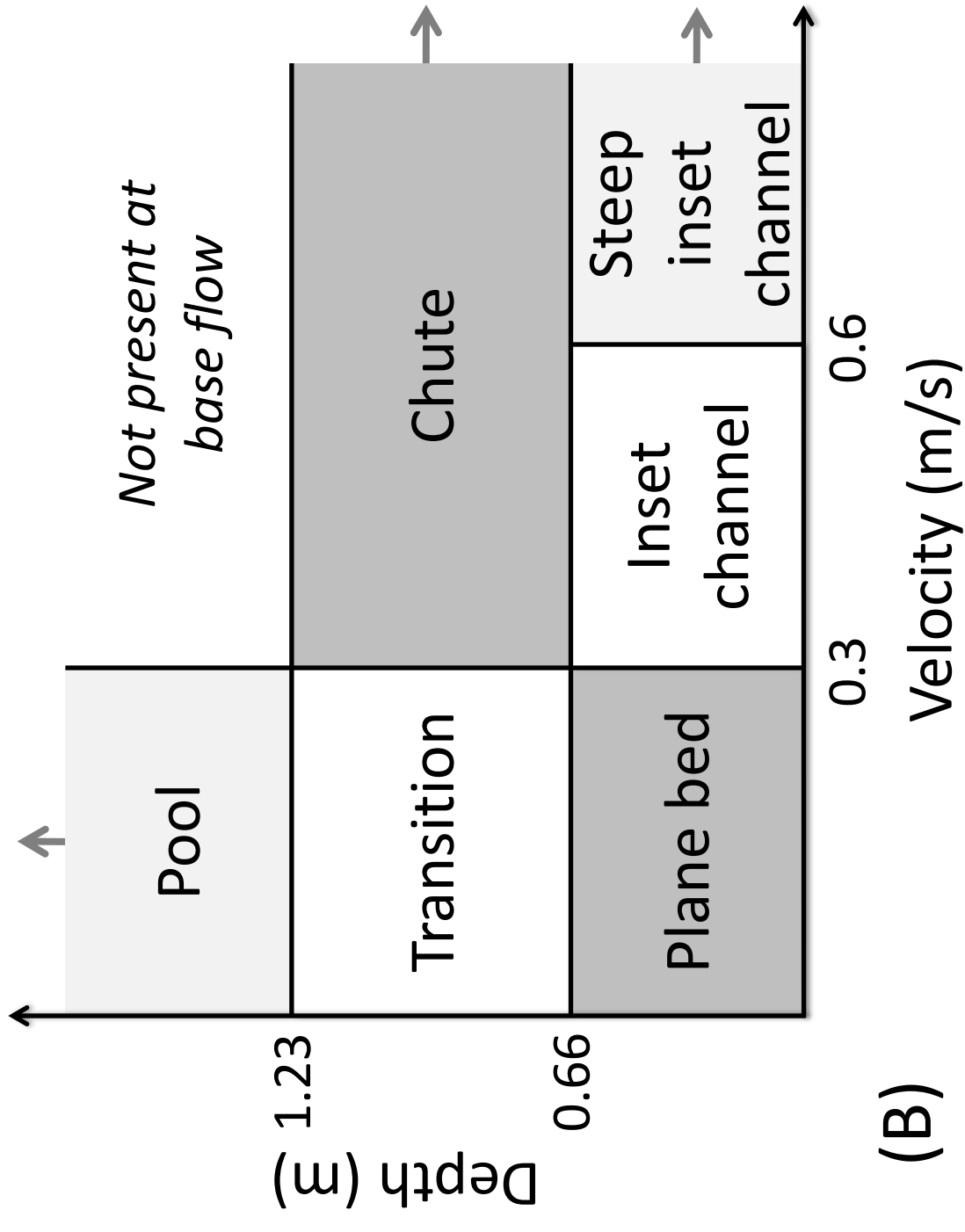
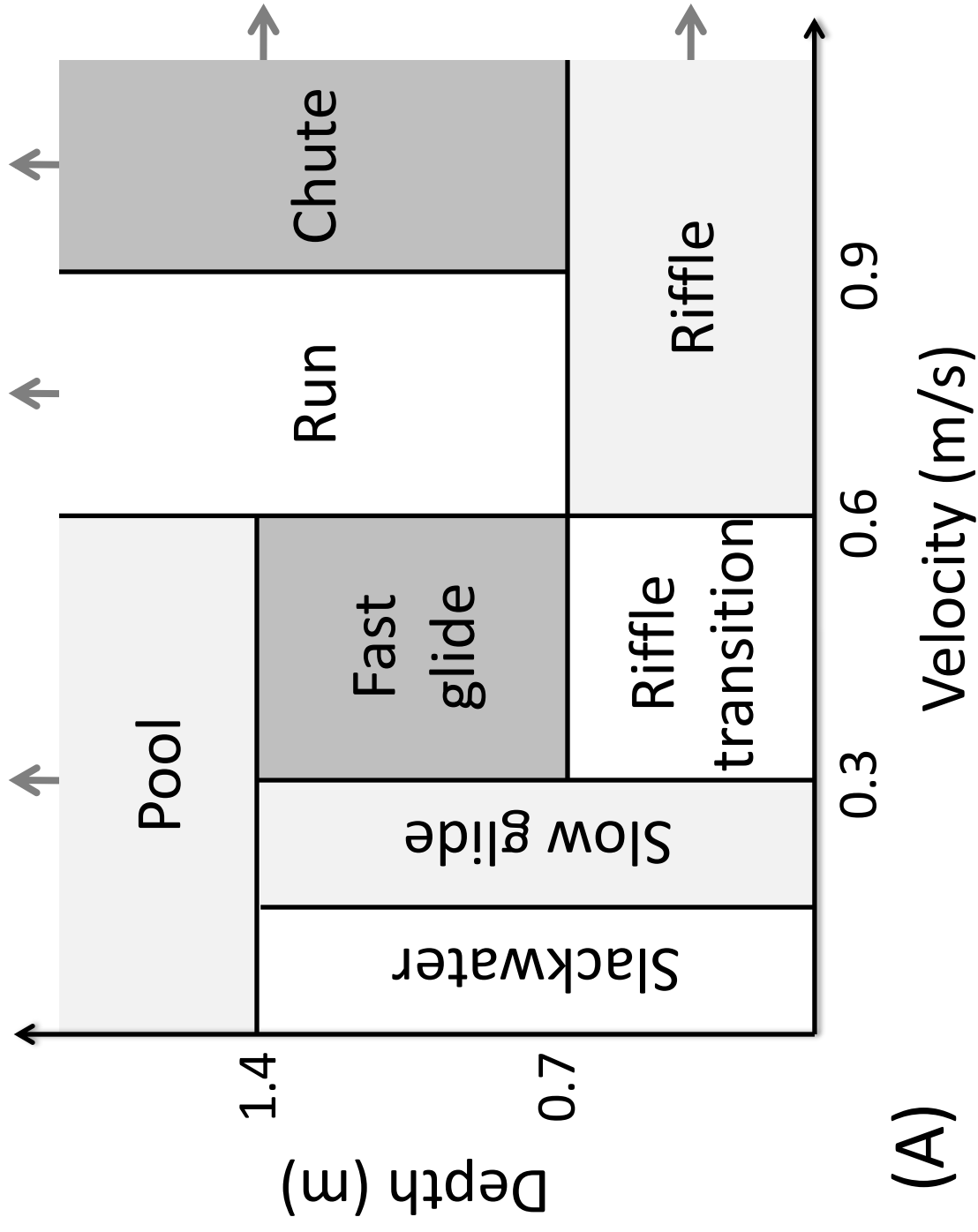
745 Fig. 1. Flowchart of MU delineation procedure. Parallelograms represent prepared data
746 input; trapezoids represent manual input; diamonds represent decisions.

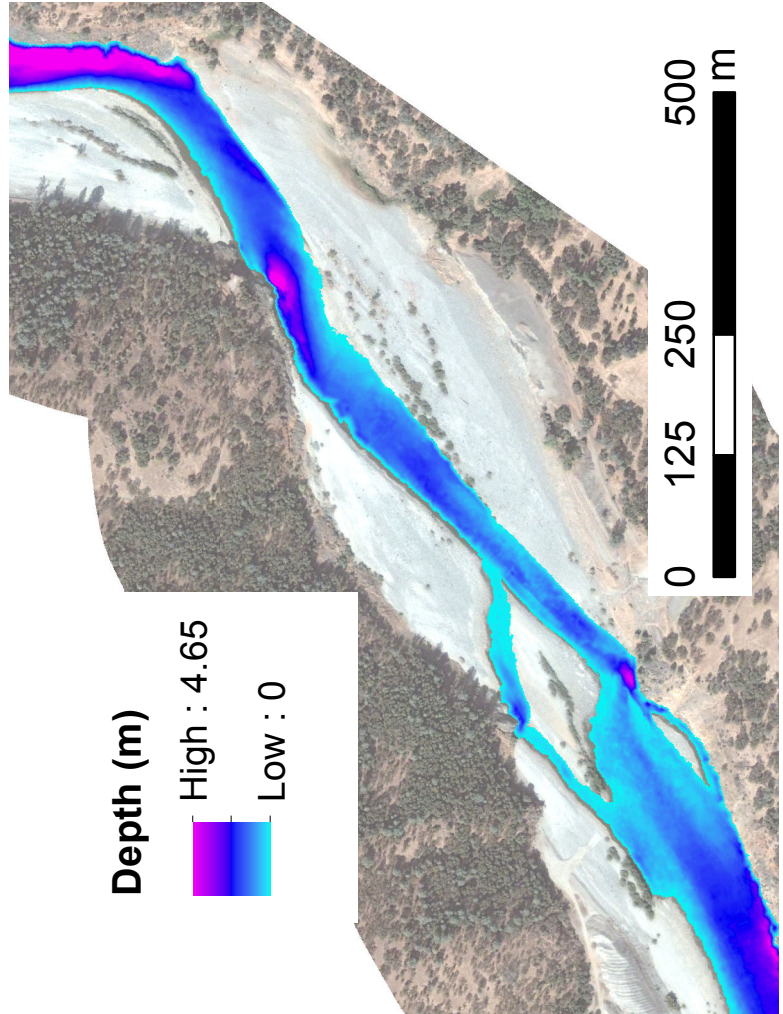
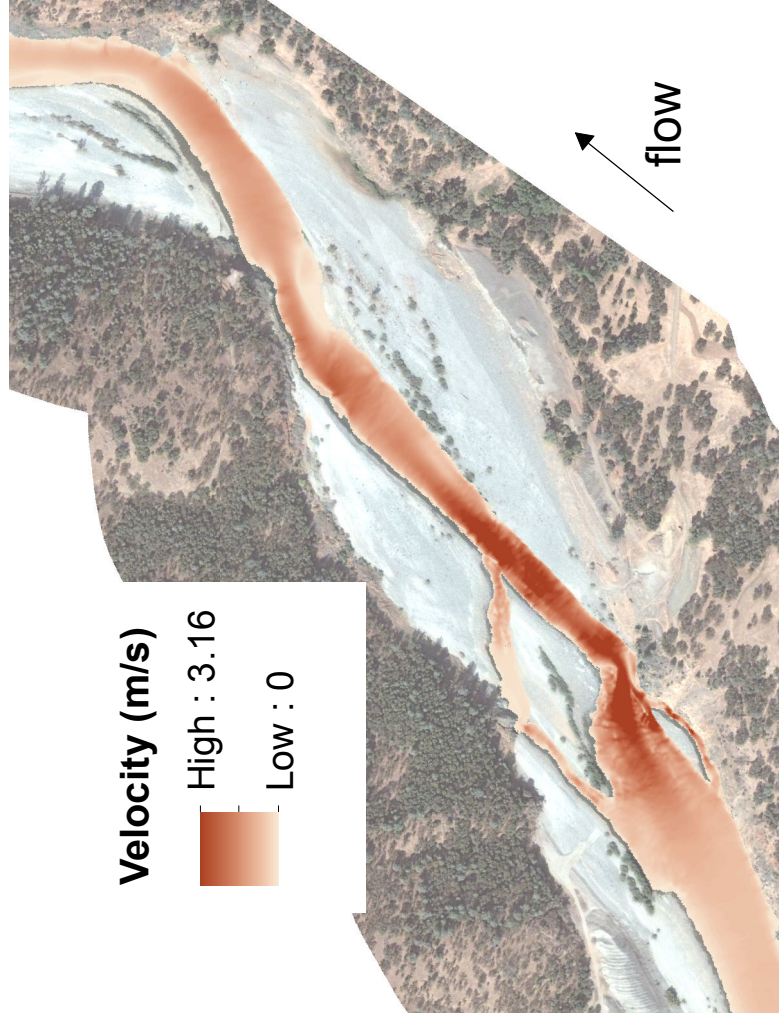
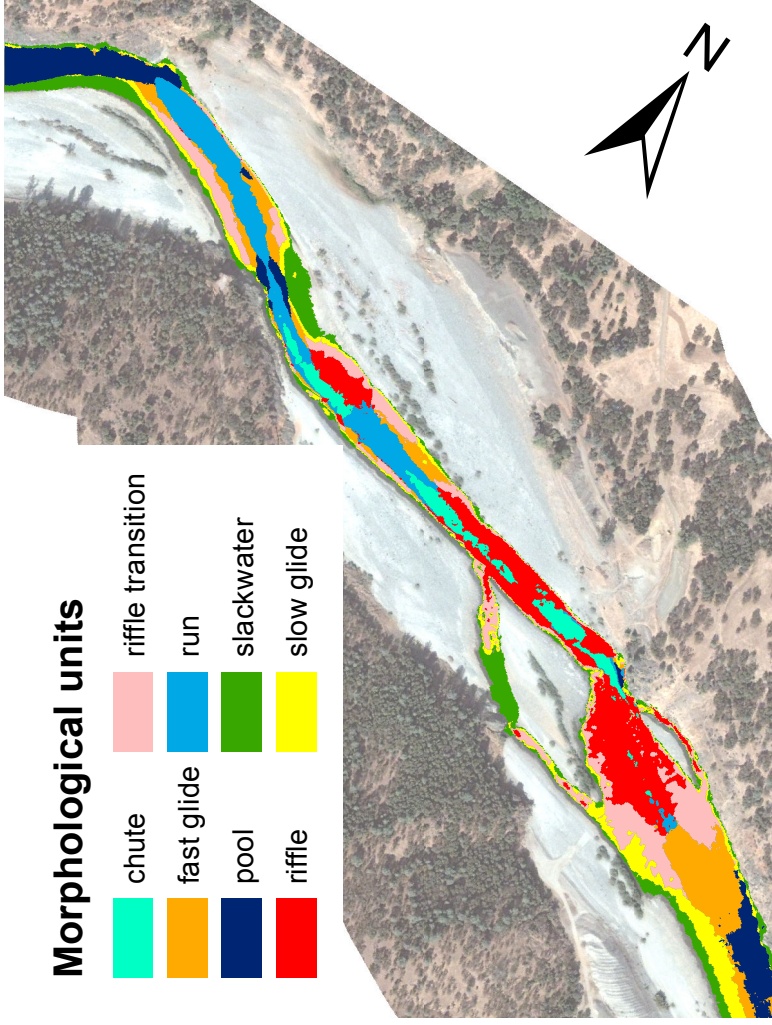
747 Fig. 2. Example MU types and hydraulic thresholds for two river morphologies. (A)
748 Lowland, gravel–cobble river at a low flow of $24.92 \text{ m}^3/\text{s}$ (Wyrick and Pasternack,
749 2012); (B) upland, cobble–boulder river at low flow of $2.577 \text{ m}^3/\text{s}$ (Pasternack
750 and Senter, 2011).

751 Fig. 3. Example MU delineation procedure for a lowland gravel–cobble river (Wyrick and
752 Pasternack, 2012).

753 Fig. 4. Example MU delineation procedure for an upland cobble–boulder river
754 (Pasternack and Senter, 2011).







0 12.5 25 50
m



- MU name
- Chute
 - Inset channel
 - Planebed
 - Pool
 - Steep inset channel
 - Step
 - Transition

- Depth (m)
- 0.01 - 0.5
 - 0.51 - 1
 - 1.01 - 1.5
 - 1.51 - 2
 - 2.01 - 2.5
 - 2.51 - 3
 - 3.01 - 3.5
 - 3.51 - 4
 - 4.01 - 4.5

- Velocity (m/s)
- 0.01 - 0.2
 - 0.21 - 0.4
 - 0.41 - 0.6
 - 0.61 - 0.8
 - 0.81 - 1
 - 1.01 - 1.2
 - 1.21 - 1.4
 - 1.41 - 1.6
 - 1.61 - 1.8

