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## **Does Type, Quantity, and Location of Habitat Matter for Fish Diversity in a Great Plains Riverscape?**

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11

### **Abstract**

12 Fisheries professionals frequently measure habitat type and amount, but less often measure the  
13 importance of where those habitats are located and in what combinations. We address this  
14 challenge by testing whether the individual and combined type, quantity, and location of habitat  
15 affects fish diversity in the upper Neosho River basin, Kansas, as a different approach to  
16 measuring habitat heterogeneity. Habitat type mattered in that species richness increased in areas  
17 of higher riffle density. Furthermore, variation within habitat type also influenced fish diversity;  
18 specifically, shallower riffles had more species of fish. The spatial arrangement (i.e.,  
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20 impact of neighbor habitats) influenced fish diversity patterns in that riffle–run and riffle–glide  
21 pairings altered riffle habitat characteristics. The study illustrates a useful approach by measuring  
22 the type, amount, and arrangement of habitats to assess fish populations and could be adapted to  
23 other stream ecosystems.

## 24 **INTRODUCTION**

25 The call to apply landscape approaches to aquatic systems (Wiens 2002; Datry et al.  
26 2016) underscores the need for researchers to address how patterns of organismal and resource  
27 heterogeneity affect basic and applied problems in natural and impacted aquatic ecosystems. For  
28 streams, collecting habitat and taxonomic data at multiple spatial scales are important (Fausch et  
29 al. 2002). Specifically, many fisheries professionals are interested in collecting, analyzing, and  
30 interpreting ecological data at larger geographic extents and would benefit from a better  
31 understanding of aquatic landscapes. Nontraditional examinations of fish distribution data within  
32 aquatic landscapes (e.g., innovative or modified metrics) that address key questions relevant to  
33 aquatic systems to solve persistent fisheries problems are needed. Our purpose was to measure  
34 patterns of heterogeneity and assess habitat arrangement on fish diversity to fill gaps in applied  
35 problem solving for aquatic landscapes.

36 Lotic ecosystems are defined by the spatial and temporal hierarchy of biological and  
37 physical characteristics along the network (Vannote et al. 1980; Frissell et al. 1986; Montgomery  
38 1999). Strong associations have been found between fish assemblages (e.g., individual species,  
39 richness, trophic structure) and large-scale features that include position within stream network  
40 (Smith and Kraft 2005), catchment area (Osborne and Wiley 1992; Newall and Magnuson 1999),  
41 catchment shape (Chiu et al. 2020), and landuse (Allan et al. 1997; Wenger et al. 2008). Other  
42 have found that local environmental factors (e.g., instream habitat, water quality) are just as  
43 important at predicting fish assemblage structure (Meador and Goldstein 2003; Rowe et al. 2009;  
44 Hitchman et al. 2018a). While empirical investigations have been mixed, it is likely that multiple  
45 factors operate at various scales to structure stream fish assemblages (Fausch et al. 2002; Walters  
46 et al. 2009). For example, both landscape-scale (e.g., watershed landcover) and local-scale (e.g.,  
47 substrate composition, woody debris) factors were important to explaining stream fish  
48 assemblage structure in central Texas river basins (Pease et al. 2011). These along with many  
49 other studies exploring fish assemblage structure at local and basin-wide scales have yielded  
50 important insights. However, a need exists to perceive streams not as lines on a map, individual

51 sampling points, or longitudinal gradients, but as spatially continuous, dynamic mosaics (Fausch  
52 et al. 2002).

53 While work has been done examining fish assemblage structure at local and basin-wide  
54 scales, much less has been done at intermediate scales (e.g.,  $\sim 10^2$ – $10^5$  m). Processes operating at  
55 these scales are often critical life-history events for many fishes (Fausch et al. 2002). To  
56 conceptualize streams as longitudinal and latitudinal mosaics, aquatic researchers need to gain an  
57 understanding of patterns and processes operating at intermediate scales. Spatial habitat  
58 heterogeneity (i.e., the uneven distribution of habitats of various sizes, shapes, amount, and  
59 arrangement) is an ecologically important characteristic of all ecosystems and, as in terrestrial  
60 systems, patterns of physical habitat heterogeneity can vary in aquatic systems with spatial scale.  
61 At smaller scales, discrete stream habitats exist based upon physical habitat structure (Cheek et  
62 al. 2016). At larger scales, a mosaic of habitats along the stream channel can create complex  
63 physical patterns to which various landscape metrics might be applied (Hitchman et al. 2018a).  
64 Therefore, metrics for aquatic landscapes are needed that measure habitat type, quantity, shape  
65 and arrangement (Figure 1-I). While many studies have been used to examine the relationship  
66 between fish and habitat, to our knowledge, this is the first that seeks to understand how the  
67 arrangement of habitats influences fish diversity.

68 Our goal was to develop an integrated approach that measures heterogeneity with the  
69 spatial arrangement of habitats and relate to fish diversity. We ask four specific research  
70 questions related to habitat type (variation across and within), quantity (size, density), shape, and  
71 spatial arrangement (Figure 1-I). First, we tested if specific combinations of physical variables  
72 (i.e., stream width, depth, flow velocity) created discrete habitat types that together created  
73 different spatial arrangements within each study reach (Figure 1-II, Q1a–b). Second, we tested  
74 how variation *within* discrete habitat types influenced fish diversity (Figure 1-II, Q2). Even if  
75 discrete types of habitat exist, within-habitat variation in physical conditions can still influence  
76 fish diversity patterns. Third, we asked if fish diversity is correlated to three commonly used  
77 terrestrial landscape metrics: habitat size, shape, and density (Figure 1- II, Q3 a–c). Fourth, we  
78 asked if the arrangement of a habitat with neighboring habitats influenced environmental  
79 variables and fish diversity (e.g., does a riffle below a pool function differently than a riffle  
80 below a run; Figure 1-II, Q4 a–c).

81

## 82 MATERIALS AND METHODS

### 83 Study Area

84 Our study was conducted within the upper Neosho River Basin (UNRB), Kansas, along  
85 the Neosho and Cottonwood rivers, 2 fifth-order rivers (Figure 2). The UNRB drains  
86 approximately 7,770 km<sup>2</sup> upstream of the John Redmond Reservoir. The basin is bounded on the  
87 north by Council Grove Reservoir and on the west by Marion Reservoir. The rivers within the  
88 study area lie upon limestone and shale bedrock underneath alluvium (Jurack and Perry 2005).  
89 Landuse is primarily agricultural (corn, wheat, soybean) with narrow riparian zones adjacent to  
90 crop fields and streams that are primarily cottonwood forests. Stream width ranged from  
91 approximately 22m to 60m. The highest elevations within our study site were 371.5m above  
92 mean sea level within the Neosho River (Figure 2, Site 1) and 353.9 m above mean sea level  
93 within the Cottonwood River (Figure 2, Site 10). The lowest elevation within our study site  
94 occurred at Site 7 (324.9m; Figure 2, Site 7) downstream of the confluence of the Neosho and  
95 Cottonwood rivers. Mean annual discharge was 8.72 m<sup>3</sup>/s (SE ± 0.94, United States Geological  
96 Survey [USGS] gage 07179730, 1963–2013) for the Neosho River and 24.55 m<sup>3</sup>/s (SE ± 2.19,  
97 USGS gage 07182250, 1963–2013) for the Cottonwood River.

98 We mapped habitat within 10 sites to measure habitat heterogeneity (Figure 2). Each site  
99 was 3 km in length from upstream to downstream, within which we measured habitat type (pool,  
100 riffle, run, glide), variation within habitat type, and the number, size, and arrangement of habitats  
101 for each type. Sampling occurred during baseflow conditions (13.0–19.0 m<sup>3</sup>/s; USGS gage  
102 07182250).

103 The basin contains three larger reservoir dams (listed previously) and six smaller low-  
104 head dams (Figure 2). One low-head dam was not sampled due to lack of landowner permission.  
105 Elsewhere we reported that no direct and consistent statistical differences existed in number of  
106 habitats per type between dammed and undammed sample sites, thus we included all sites in  
107 each analysis (Hitchman et al. 2018a, 2018b).

108

### 109 Field Sampling

#### 110 *Habitat Types*

111 From June–August 2013, we identified and mapped habitats (N=310) within each site  
112 based on agreement by two independent observers, using an objective series of stream channel

113 morphology, surface flow, depth, and sediment composition criteria (McCain et al. 1990; Harvey  
114 and Clifford 2009; Supplemental Material, Appendix 1). Habitats were classified as pool, riffle,  
115 run, or glide (see Supplemental Material, Appendix 1 for description). We quantified the spatial  
116 location of each habitat using trackplots at 5-second intervals and waypoints at the upper and  
117 lower boundary from a handheld Garmin GPSmap76Cx (Garmin International, Olathe, Kansas).  
118 Trackplots and waypoints for each habitat were imported into ArcMap v. 10.2 (Esri, Redlands,  
119 California) where they were digitized into polygons and converted to raster format.

120

### 121 *Habitat Characteristics*

122 Environmental variables (i.e., wetted stream width, water depth, and flow velocity) were  
123 measured within five habitats for each type (pool, riffle, run, glide) at each sample site. At sites  
124 where less than five habitats for a particular type existed, all habitats of that type were sampled.  
125 Wetted stream width was recorded using a Nikon 8398 range finder (<1 m accuracy, range 3–  
126 200 m) at the midpoint of each habitat. Cross-stream transects of five equally spaced points were  
127 used to collect depth and water velocity measurements. Flow velocity (60% of the depth) was  
128 measured with a Marsh–McBirney (Loveland, Colorado) Model 2000 flowmeter.

129

### 130 *Fish Diversity Sampling*

131 We captured fish using a two-person mini-Missouri trawl, which has been used to capture  
132 both small and large-bodied benthic fish (e.g., Fencel et al. 2017; Hitchman et al. 2018a). The  
133 mini-Missouri trawl caught the same or more numbers of species as a seine net and backpack  
134 electrofisher in a controlled gear experiment (Fencel et al. 2017). Fish were collected using a 2.4-  
135 m mini-Missouri trawl (Herzog et al. 2009) with 35-mm outer mesh and 3.2-mm inner mesh  
136 pulled from upstream to downstream for 30 m. Fish were sampled in 134 of the 310 habitats  
137 mapped (43.2%). Fish were counted, identified to species, then returned alive to the stream.  
138 Additional details on all methods are provided elsewhere (Hitchman et al. 2018a, 2018b).

139

### 140 **Data Analysis**

141 *Question 1: Do discrete habitat types exist (a) and does the arrangement of habitats differ across*  
142 *sites (b)?*

143 We used non-metric multidimensional scaling to evaluate if habitat types formed discrete  
144 groups based on the environmental variables (stream width, water depth, and flow velocity). For  
145 the non-metric multidimensional scaling, we used a Bray–Curtis distance matrix. Habitat types  
146 were factors, environmental variables were response variables. Confidence ellipses (95%) were  
147 plotted along with each habitat sample ( $N = 138$ ) to visualize the relative location of each habitat  
148 type. We quantitatively tested separation among habitat types (pool, glide, run, riffle) with an  
149 analysis of similarity (ANOSIM). A  $p$  value  $< 0.05$  indicated that the environmental variables  
150 were quantitatively discrete among habitat types.

151 Next, we used a series of longitudinal profile plots to illustrate variation in the diversity,  
152 sequence, and size of habitats within and across each of our ten 3-km sites. To create each  
153 longitudinal profile, the length (m) of each habitat was plotted sequentially from upstream to  
154 downstream (X axis). The area of each (ha) was plotted on the Y axis. Habitat length and area  
155 were calculated from raster files for each site using ArcMap v. 10.2. These longitudinal profiles  
156 visually illustrated habitat diversity. To quantify differences in habitat diversity, we calculated  
157 the number of habitats and overall habitat heterogeneity (Shannon's Diversity Index) for each  
158 site. We then used a chi-square test with 2,000 Monte Carlo simulations to assess if habitat  
159 heterogeneity varied significantly across sample sites. Note, all statistical analyses outlined here  
160 and below were run using R v.4.1 (R Core Team 2013).

161

162 *Question 2: Does within-habitat type variation influence fish diversity?*

163 To examine if continuous variation in environmental variables (width, depth, flow  
164 velocity) within-a habitat type (pool, riffle, run, glide) affected fish species diversity, we ran  
165 multiple linear regression models for each environmental variable (X axis) relative to species  
166 richness (Y axis) within each habitat type. Because we were examining the influence of habitat  
167 arrangement on fish communities (not comparing diversity metrics), we chose a single, common  
168 measure of diversity (species richness). Species richness is commonly used diversity metric in  
169 the ecological literature and in this study provides similar results to other diversity metrics (e.g.,  
170 species evenness, beta diversity; Supplemental Material, Appendix 2).

171

172 *Question 3: Do landscape ecology metrics predict riverscape-scale fish diversity?*

173 We chose three commonly used landscape ecology metrics that describe habitat to  
174 examine their relationship with fish species richness. For each habitat type (pool, riffle, run, and  
175 glide), we calculated mean habitat size (size), habitat perimeter–area ratio (shape), and habitat  
176 density (arrangement) using an ArcGIS platform map that was converted from a polygon-based  
177 feature file to a raster format and inputted into FRAGSTATS 4.1 (McGarigal et al. 2012). Mean  
178 habitat size was calculated as the average area (km<sup>2</sup>) of each habitat within a particular habitat  
179 type for each site. Perimeter–area ratio was calculated as the ratio of habitat perimeter to the  
180 habitat area. Habitat density was calculated as the total number of habitats per 100 ha. The  
181 metrics were chosen because they could be calculated from our dataset and have been shown to  
182 be ecologically meaningful in terrestrial ecology studies (Supplemental Material, Appendix 1).

183 We then used an inference–theoretic approach to select the top multiple regression  
184 models (AIC<sub>c</sub>; Burnham and Anderson 2002;  $\Delta$ AIC<sub>c</sub> scores < 2; MuMin package; Bartoń 2013)  
185 that related predictor variables (habitat size, habitat shape, habitat density for all four habitat  
186 types) to the species richness response variable. Each independent variable was standardized so  
187 the mean equals zero. For parsimony, we limited each candidate model to one or two predictor  
188 variables. We eliminated collinear models (variance inflation factor >10 or condition index >30)  
189 and models with pairs of correlated ( $r > 0.70$ ) predictor variables.

190

191 *Question 4: Do adjacent habitats influence habitat-specific characteristics and fish diversity?*

192 Neighboring habitats may affect fish community patterns. We tested the effect of  
193 adjacent habitat types (upstream or downstream) on the environmental characteristics (e.g.,  
194 depth, water velocity) of the target habitat type. For example, we investigated how  
195 environmental variables within pool habitat (target) would change when a riffle was present  
196 (adjacent habitat) both upstream and downstream. To test this neighbor effect, we used a  
197 multivariate analysis of variance (MANOVA) in which all combinations of adjacent habitat both  
198 upstream and downstream ( $N = 24$ ) were the treatments and environmental variables (stream  
199 width, depth, and flow velocity) for each target habitat (pool, riffle, run, and glide) were the  
200 response variables.

201 Next, we used path analysis to evaluate direct and indirect effects of neighboring habitats  
202 on environmental variables and species richness within a target habitat. Path analysis calculates

203 path coefficients (standardized partial regression coefficients) that quantify direct and indirect  
204 effects. Direct effects are the coefficients between two variables connected by a path. Indirect  
205 effects are effects mediated through another variable (e.g., habitat). Path analysis was performed  
206 for all combinations of target habitats (pool, riffle, run, and glide) and adjacent habitats both  
207 upstream and downstream (library lavaan with function sem in R; Rosseel 2012). Standardized  
208 path coefficients (standardized  $\beta$ ) indicated the strength of relationships and  $R^2$  measured the  
209 amount of variation explained by specific sets of variables. For both the MANOVA and the path  
210 analysis, we ran all combinations of target and neighbor habitats, but we only present significant  
211 relationships.

212

## 213 **RESULTS**

214 *Question 1a: Do discrete habitat types exist?*

215 Based on width, depth, and flow velocity, all habitat categories (riffle, run, pool, and  
216 glide) were quantitatively distinct (ANOSIM Global  $R = 0.29$ ,  $p = 0.001$ ; Figure 4). No overlap  
217 existed among 95% confidence ellipses, indicating clear separation among habitat types.

218 Although physically distinct, run and glide habitats had environmental conditions that were  
219 intermediate between pool and riffle habitat types (Figure 3).

220

221 *Question 1b: Does the arrangement of habitats differ across sites?*

222 Habitat arrangement differed across our sample sites (Figure 4). Specifically, the number  
223 and type of habitats varied across sites ( $\chi^2 = 70.42$ ,  $p < 0.001$ ; Figure 4). The total number of  
224 habitats per sample site ranged from 17 (Figure 4j, Site 10) to 59 (Figure 4b, Site 2). Pool, riffle,  
225 and run habitats were present across all sites, but glides were irregularly distributed and were  
226 most commonly associated with riffles and pools (i.e., adjacent to pools 68–74%; adjacent to  
227 riffles 19–21%). Overall, site-specific habitat diversity ranged from 0.56 (Figure 4g) to 1.15  
228 (Figure 4a, b). As specific examples, Sites 1 and 2 were characterized as having a high number  
229 of total habitats (40 and 59, respectively) and high habitat diversity (1.15 and 1.15, respectively;  
230 Figure 4a, b), whereas Sites 7 and 9 had fewer total habitats (19 and 24, respectively) and low  
231 habitat diversity (0.56 and 0.63, respectively; Figure 4g, i).

232



233 *Question 2: Does within-habitat variation influence fish diversity?*

234 Variation in species richness existed within select habitat types. Within 143 unique  
235 habitats, we sampled a total of 7,791 fish representing 35 species, across 7 families, at 10 sites  
236 within the upper Neosho River sub-drainage. No significant relationships existed between fish  
237 species richness and within-habitat variation in environmental variables (width, depth, flow, flow  
238 velocity) for pools, glides, and runs (data not shown). However, for riffles, shallower ( $F_{1,33} =$   
239  $18.82$ ;  $p = 0.001$ ; Figure 5a) and slower ( $F_{1,33} = 10.46$ ;  $p = 0.02$ ; Figure 5b) riffles had more fish  
240 species.

241  
242 *Question 3: Do landscape ecology metrics predict riverscape-scale fish diversity?*

243 For three commonly used landscape ecology metrics (*habitat size, habitat shape, habitat*  
244 *density for four habitat types*), riffle *density* had the greatest influence on fish species richness  
245 (Table 1). The four top models ( $\Delta AIC_c < 2$ ) were significant ( $p = 0.004$ – $0.007$ ) and had  $R^2$  values  
246 of  $0.69$ – $0.75$  (Table 1). In these top models, *riffle density* had a high and positive model-  
247 averaged standardized slope ( $\beta = 3.14$ ), a high variable importance ( $VI = 1.0$ ) and explained  
248  $54.3$ – $84.8\%$  (average  $74\%$ ) of the variation in fish richness. Thus, sites with many riffles had a  
249 higher number of fish species.

250 As predicted, some habitat size metrics were also positively related to fish species  
251 richness in that individual models and sites with larger mean habitat sizes of riffles, glides, and  
252 runs had more fish species, with larger areas having a higher species richness (mean *riffle area*  
253 [ $\beta = 2.0$ ;  $VI = 0.25$ ;  $23.4\%$  of variance]; mean *glide area*. [ $\beta = 1.9$ ;  $VI = 0.25$ ;  $45.7\%$  of  
254 variance]; mean *run area* [ $\beta = 2.2$ ;  $VI = 0.25$ ;  $15.3\%$  of variance]; Table 1). Contrary to  
255 predictions from terrestrial systems that higher edge area would increase diversity, riffle  
256 Perimeter–area ( $\beta = -1.9$ ;  $VI = 0.25$ ;  $19.7\%$  of variance) was negatively related to fish species  
257 richness (Table 1).

258  
259 *Question 4: Do adjacent habitats influence habitat-specific characteristics and fish diversity?*

260 Neighboring habitats altered environmental characteristics of target habitats, and, in some  
261 cases, species richness. Of the 24 target–neighbor habitat combinations tested, we observed three  
262 neighbor-altered habitat modifications for environmental conditions only (Figure 6; MANOVA).  
263 First, flow velocity was higher in pools (target) directly above runs (neighbor;  $F = 6.37$ ;  $p < 0.01$ ;

264 Figure 6a). Second, runs (target) were deeper when downstream of pools (neighbor;  $F = 3.07$ ;  $p$   
265  $< 0.1$ ; Figure 6b). Third, glides (target) were significantly faster upstream of riffles (neighbor;  $F$   
266  $= 6.00$ ;  $p < 0.01$ ; Figure 6c).

267 We observed three other significant neighbor–target habitat combinations that altered fish  
268 species richness within the target habitat as well as environmental variables (Figure 7; path  
269 analysis). In all three examples, species richness was *inversely* related to habitat depth. First,  
270 depth was *shallower* in riffles (target) located immediately downstream of glides (neighbor;  $R^2 =$   
271  $0.11$ ;  $p < 0.05$ ; Figure 7a, left and middle columns), which led to an *increase* in species richness  
272 ( $R^2 = 0.26$ ;  $p < 0.05$ ; Figure 7a, middle and right columns). Second, riffles (target) were *deeper*  
273 when downstream of runs (neighbor;  $R^2 = 0.12$ ;  $p < 0.05$ ; Figure 7b, left and middle columns),  
274 which led to *lower* species richness ( $R^2 = 0.31$ ;  $p < 0.05$ ; Figure 7b, middle and right columns).  
275 Third, riffles (target) were significantly *shallower* and *narrower* when upstream of glides  
276 (neighbor;  $R^2 = 0.16$ ;  $p < 0.05$ ; Figure 7c, left and middle columns), which led to an *increase* in  
277 species richness ( $R^2 = 0.26$ ;  $p < 0.05$ ; Figure 7c, middle and right columns).

278

## 279 DISCUSSION

280 Understanding riverscape patterns and drivers is important for fisheries professionals  
281 (Fausch et al. 2002; Wiens 2002; Allan 2004). Managers increasingly voice their need to make  
282 science-based conservation decisions beyond a single site and time. Scaling up data from the site  
283 level to entire watersheds is difficult. We provide an approach that captures the heterogenous  
284 nature of riverine ecosystems and how the arrangement and amount of habitat influences fish  
285 diversity. Our research provided three take-home messages, reviewed in detail below that can  
286 help advance an aquatic landscape approach for fish diversity patterns in a Great Plains  
287 riverscape.

288 First, a gap that needs to be bridged between fisheries research and conservation are  
289 studies conducted at intermediate scales. Generally, aquatic ecologists use sampling units that  
290 are often small (stream reaches  $< 200$  m), with data collected over disjunct locations that can,  
291 at times, span large spatial scales (e.g., watersheds). Fisheries managers, on the other hand, are  
292 often tasked with “scaling up” these data to areas that span geopolitical boundaries. Often what  
293 is missing are data collected at intermediate scales. Stream fish populations and assemblages  
294 often carry out important stages in their life history at intermediate scales (between 1–100k)

295 and respond to changes in habitat (Fausch et al. 2002). At a small scale, habitats can be  
296 described by their type, size, shape, and location along the landscape. However, at larger scales  
297 other patterns become important including repeating patterns of small-scale heterogeneity, the  
298 presence of rare but influential features, and the role of type, size, and spatial arrangement and  
299 location. Few of these complex patterns of heterogeneity are presently described  
300 comprehensively or tested systematically in aquatic systems (or in landscape ecology metrics  
301 applied to aquatic systems), making a synthesis of larger scale heterogeneity difficult. Therefore,  
302 sampling regimes that are able to capture the composition and configuration of habitats along a  
303 riverscape, as we describe in this pilot study, provide a better understanding of aquatic  
304 landscapes that can advance solutions to persistent management problems (e.g., threatened  
305 species management, watershed restoration, sustainability, stewardship, and managing impacts  
306 such as instream barriers).

307 Our research identified both established and novel patterns of heterogeneity. We found  
308 stream habitats to be functionally distinct units and were delineated as pool, riffle, run, or glide  
309 based upon stream channel morphology, depth, surface flow, and sediment composition.  
310 Ecologists have long recognized that different habitat types within a local stream reach often  
311 harbor different fish communities (Taylor 2000). Treating different habitat types as separate,  
312 self-contained entities is logistically convenient and ecologically useful (Taylor 2000; Hitchman  
313 et al. 2018a). However, at a larger scale, connected habitats allowed for the quantification of  
314 various heterogeneity metrics that can go beyond simply connecting stream fishes to a particular  
315 individual habitat. For example, Hitchman et al. (2018a) found fish diversity to significantly  
316 increase with overall habitat heterogeneity in Great Plains riverscapes. Linking the distribution  
317 and arrangement of stream habitats creates mosaics across the riverscape to which heterogeneity  
318 metrics can be applied to detect underlying ecological patterns for both common and uncommon  
319 habitat types. As we have shown here, how habitats are arranged and the role of various  
320 measures of habitat heterogeneity on stream fish diversity provide important information for  
321 environmental professionals working to maintain and restore habitats at larger spatial scales.

322 Second, landscape ecology metrics were found to be useful in describing fish–habitat  
323 relationships in streams. We found riffle density to be a consistently important predictor of  
324 stream fish diversity. Riffles have unique ecological characteristics and can act as keystone  
325 habitats in that they disproportionately increase fish diversity across Great Plains riverscapes

326 (Hitchman et al. 2018a). If management goals are to conserve overall native biodiversity (at least  
327 in study systems similar to ours), then increasing the density of riffles by constructing artificial  
328 riffles may be an effective conservation strategy. Artificial riffles are used in stream restoration  
329 to increase biodiversity (Edwards et al. 1984; Favata et al. 2018) and provide habitat for  
330 threatened fish species (Fuselier and Edds 1994). To increase riffle density, artificial riffles could  
331 be constructed to meet certain minimum habitat density thresholds that increase overall habitat  
332 heterogeneity. While artificial riffles have been shown to be effective, the context of the entire  
333 mosaic of habitats needs to be considered in order to identify where and how many riffles need  
334 to be constructed to optimize management efforts. Our approach can help resource managers  
335 identify critical habitats, set minimum habitat density requirements, and apply successful spatial  
336 arrangements for restoration.

337 Third, the spatial arrangement of habitats (e.g., placement of habitats and the identity of  
338 neighbors) matters. While characteristics of neighboring habitats have been shown to influence  
339 species diversity in terrestrial ecosystems (Glass and Floyd 2015), examinations in aquatic  
340 ecosystems is limited. Our finding that neighbor habitats can alter characteristics of target  
341 habitats in aquatic ecosystems is novel and can explain natural variability that is especially  
342 common in lotic field studies. Target habitats often took on characteristics of flow velocity and  
343 depth found in neighboring habitats. For example, flow velocity within pools and glides  
344 increased when these were adjacent to fast-flowing habitats (e.g., riffles and runs). Relative to  
345 the effect of neighbor habitats on fish diversity, riffles were shallower when adjacent to glides,  
346 both upstream and downstream, and had increased fish diversity. Conversely, riffles were deeper  
347 when located downstream of run habitats and had decreased fish diversity. Past studies have  
348 found preferences among fish with flow-sensitive habitat types and in particular, shallow or slow  
349 flowing riffles (Gelwick 1990; Mathews 1990; Aadland 1993). These neighbor effects on fish  
350 diversity in riffle target habitats were also consistent with our within-habitat analysis, in which  
351 we found that species richness was higher in shallower, slower riffles. This demonstrates the  
352 importance of habitat arrangement along aquatic landscapes and may be considered for  
353 prescriptive solutions (e.g., placement of artificial habitats during restoration).

354 Our goal was to show that the type, quantity, and location of stream habitat influences  
355 fish diversity. The density of riffle habitats was an important predictor for fish diversity in low-  
356 gradient prairie streams and can be used by resource managers to plan habitat alterations that

357 benefit watershed restoration and threatened species management. Once an understanding of how  
358 the amount and arrangement of habitats influences fish distributions and diversity, the next step  
359 would be to scale up from habitat level sampling to larger spatial extents. While our primary  
360 focus was on maintaining native fish diversity, this approach could potentially be used for other  
361 important metrics (e.g., abundance, functional guilds). For example, if conservation of a  
362 threatened or endangered species is a priority, our methods could be adapted to identify critical  
363 habitat and the influence neighboring habitats have on the abundance of the species of concern.  
364 Furthermore, managers could prescribe minimum habitat density requirements that could be  
365 evaluated using emerging technology (e.g., geographic information systems, drones). Our  
366 approach can be used to identify similar critical habitats in other systems and to explore how  
367 habitat arrangement may influence other ecological processes (e.g., trophic food webs, energy  
368 flow, predator–prey dynamics).

369

## 370 **ACKNOWLEDGMENTS**

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Table 1. Multiple regression models that examine relationships between species richness (response) and landscape metrics (predictor) such as habitat size, habitat shape, habitat density for all four habitat types. Bolded denotes significant standardized  $\beta$  (slopes) at  $\alpha = 0.05$  and () displays standard errors. df = degrees of freedom. The model average and variable importance are listed at the base of the table. Model averaged  $\beta$  was calculated as the mean slope for each predictor across models. Variable importance was calculated as the proportion of candidate models for which each predictor was included.

<b>Riffle density</b>		<b>Riffle area</b>		<b>Glide area</b>		<b>Riffle Perimeter-ratio</b>		<b>Run area</b>		Model	<i>p</i>	AICc	Delta	Weight
( <b><math>\beta</math></b> )	% Variation Explained	( <b><math>\beta</math></b> )	% Variation Explained	( <b><math>\beta</math></b> )	% Variation Explained	( <b><math>\beta</math></b> )	% Variation Explained	( <b><math>\beta</math></b> )	% Variation Explained	R2				
<b>3.2</b>		<b>2.0</b>												
( <b>0.6</b> )	76.64	( <b>0.6</b> )	23.36							0.80	0.00	53.20	-	0.30
<b>2.2</b>				<b>1.9</b>										
( <b>0.7</b> )	54.32			( <b>0.7</b> )	45.68					0.70	0.00	53.70	0.40	0.20
<b>3.3</b>						<b>-1.9</b>								
( <b>0.7</b> )	80.27					( <b>0.7</b> )	19.73			0.70	0.01	54.30	1.10	0.20
<b>4.0</b>								<b>2.2</b>						
( <b>0.9</b> )	84.75							( <b>0.9</b> )	15.25	0.70	0.01	55.10	1.90	0.10

										<b>Model Average</b>	
3.1		2.0		1.9		-1.91		2.2			
(0.7)	74.00	(0.6)	23.36	(0.7)	45.68	(0.69)	19.73	(0.9)	15.25		
1.00		0.25		0.25		0.25		0.25			

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487 **Figure Captions**

488 Figure 1. Conceptual diagram outlining (I) types of larger scale heterogeneity and (II) our  
489 specific research questions related to larger-scale heterogeneity and fish biodiversity.

490 Figure 2. Map of study area representing ten 3-km study sites located along the Neosho and  
491 Cottonwood rivers, Kansas. Black dots represent low-head dam sites and gray dots  
492 undammed sections of the rivers. The city of Emporia is represented by a star. Note:  
493 habitat–fish diversity relationships are not explained by the presence of low-head  
494 dams.

495 Figure 3. Non-metric multidimensional scaling (NMDS) biplot (Stress = 0.03) for stream  
496 habitats at ten 3-km study sites located along the Neosho and Cottonwood rivers,  
497 Kansas. Points represent each habitat sample ( $N = 138$ ). Colored ellipses indicate 95%  
498 standard error confidence ellipses for the mean. Analysis of similarity indicates  
499 significant separation among each habitat type (ANOSIM Global  $R = 0.30$ ,  $p = 0.001$ ).

500 Figure 4. Longitudinal profiles for each of ten 3-km study sites along the Neosho and  
501 Cottonwood rivers, Kansas. Represented are mean habitat area, total number of  
502 habitats ( $N$ ), and overall habitat diversity calculated using Shannon's Diversity Index  
503 ( $H'$ ). Habitat types are shown in different colors.

504 Figure 5. Linear regressions between species richness and (a) stream depth and (b) stream flow  
505 velocity within all riffle habitats sampled in the upper Neosho River Basin.

506 Figure 6. Boxplots showing significant relationships of within habitat characteristics relative to  
507 neighbor (adjacent) habitats: (a) illustrates how flow velocity within pool habitats is  
508 influenced by downstream neighbor habitats, (b) illustrates how depth within run  
509 habitats is influenced by upstream neighbor habitats, and (c) illustrates how flow  
510 velocity within glide habitats is influenced by downstream neighbor habitats. The box  
511 represents the second and third quartiles, whiskers show the first and fourth quartiles,  
512 and heavy horizontal lines depict the median.

513 Figure 7. Path analyses investigating direct and indirect relationships for species richness within  
514 (a) riffle habitats that were located immediately downstream of glides, (b) riffle  
515 habitats that were located immediately downstream of runs, and (c) riffle habitats that  
516 were located immediately upstream of glides. We only show significant relationships  
517 at  $\alpha = 0.05$ . Solid lines represent positive relationships and dashed lines represent

518 negative relationships. The standardized slope ( $\beta$ ), coefficient of determination ( $R^2$ ),  
519 and significance ( $p$ ) are shown for each variable pair (i.e., over each connecting line).

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**What, How Much, & Where**

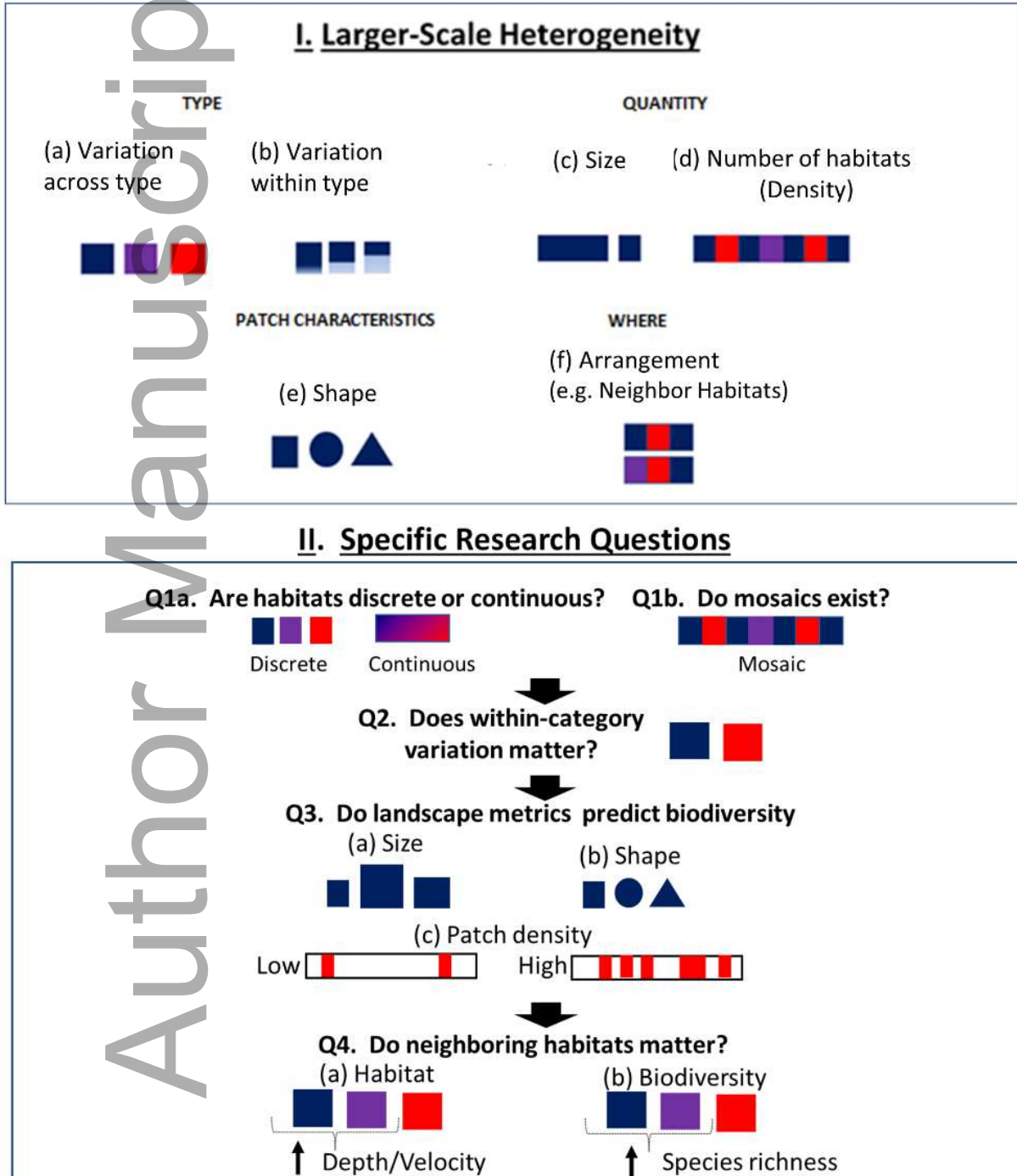


Figure 1

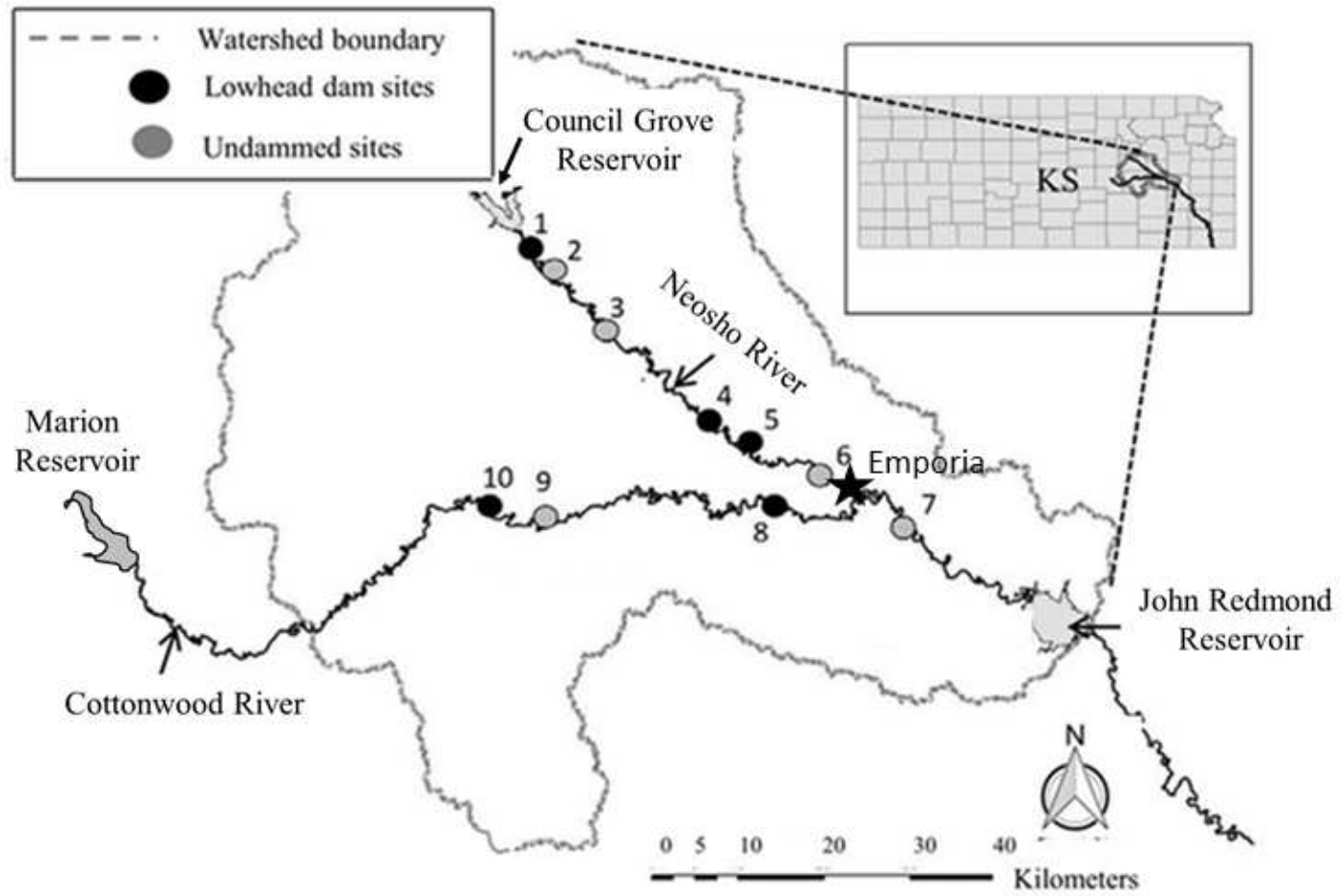


Figure 2

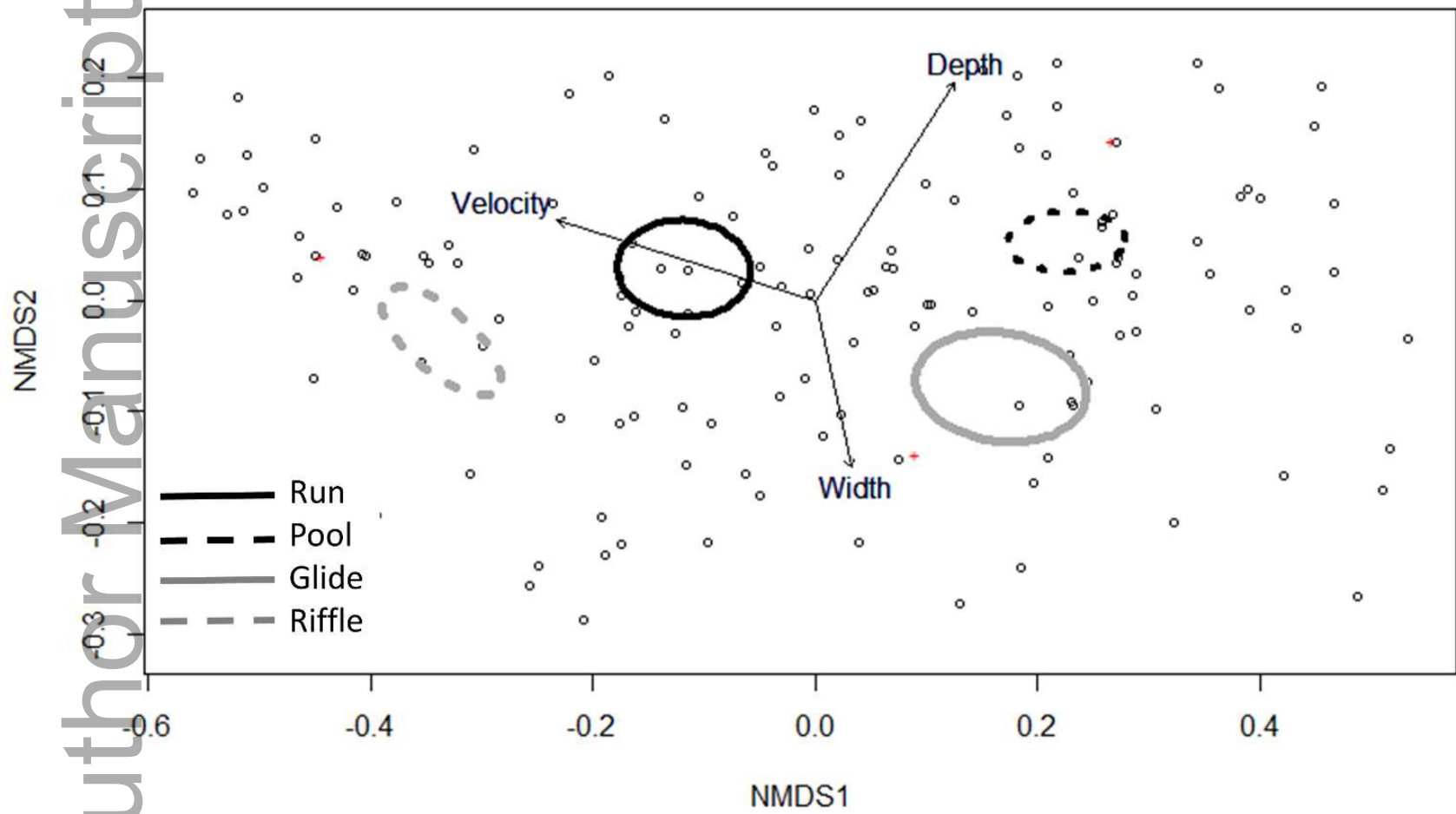


Figure 3



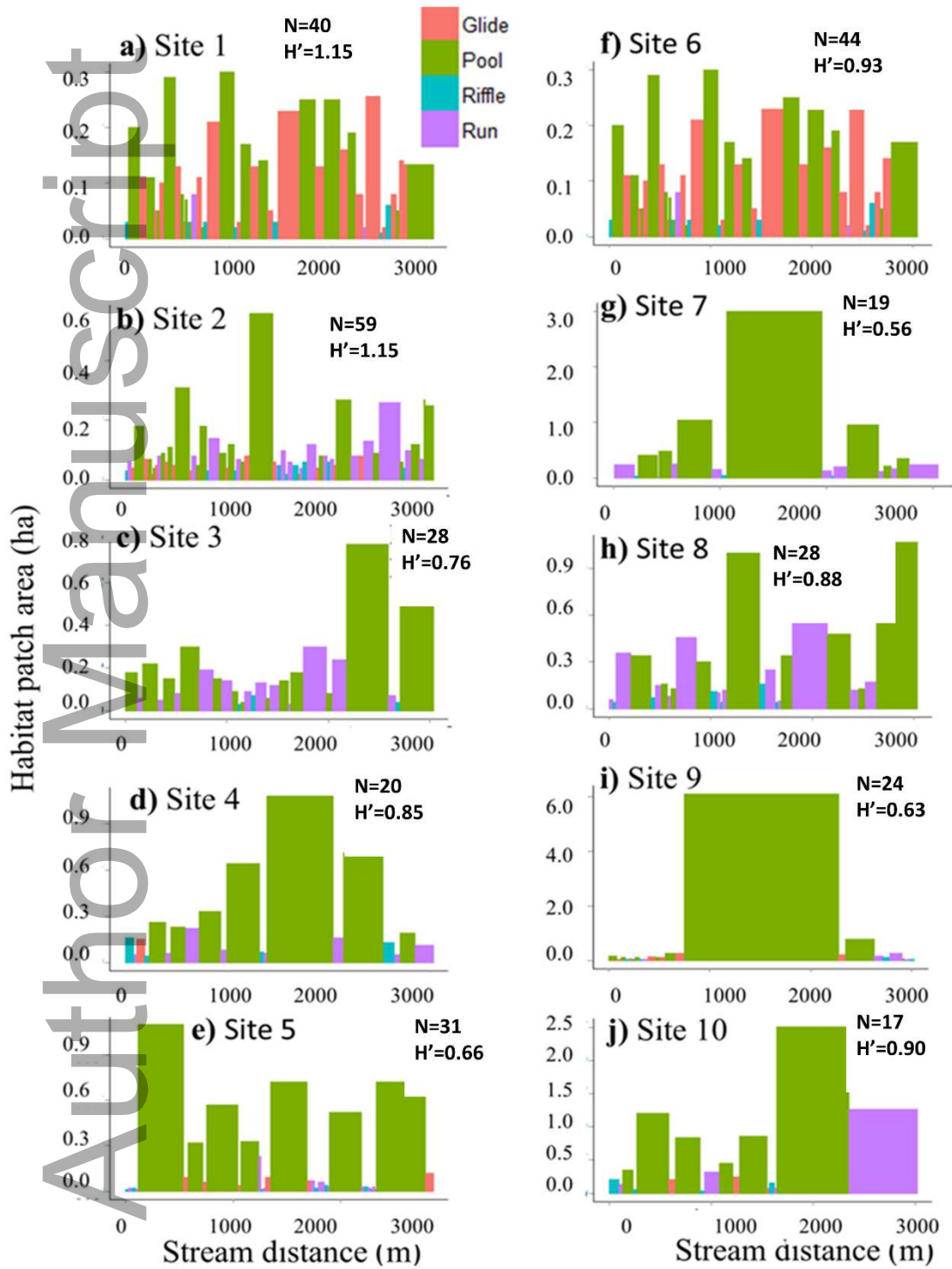


Figure 4

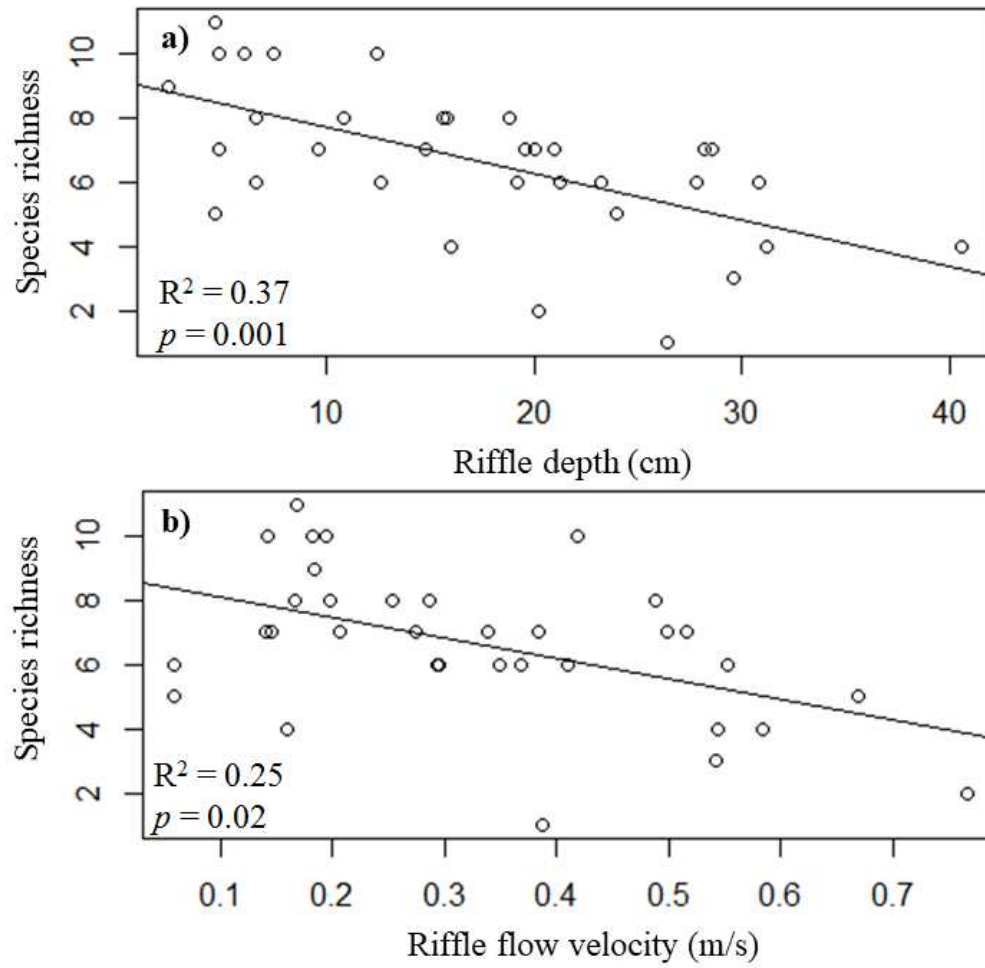


Figure 5

**Neighbor-Induced Effects on Environmental Changes**

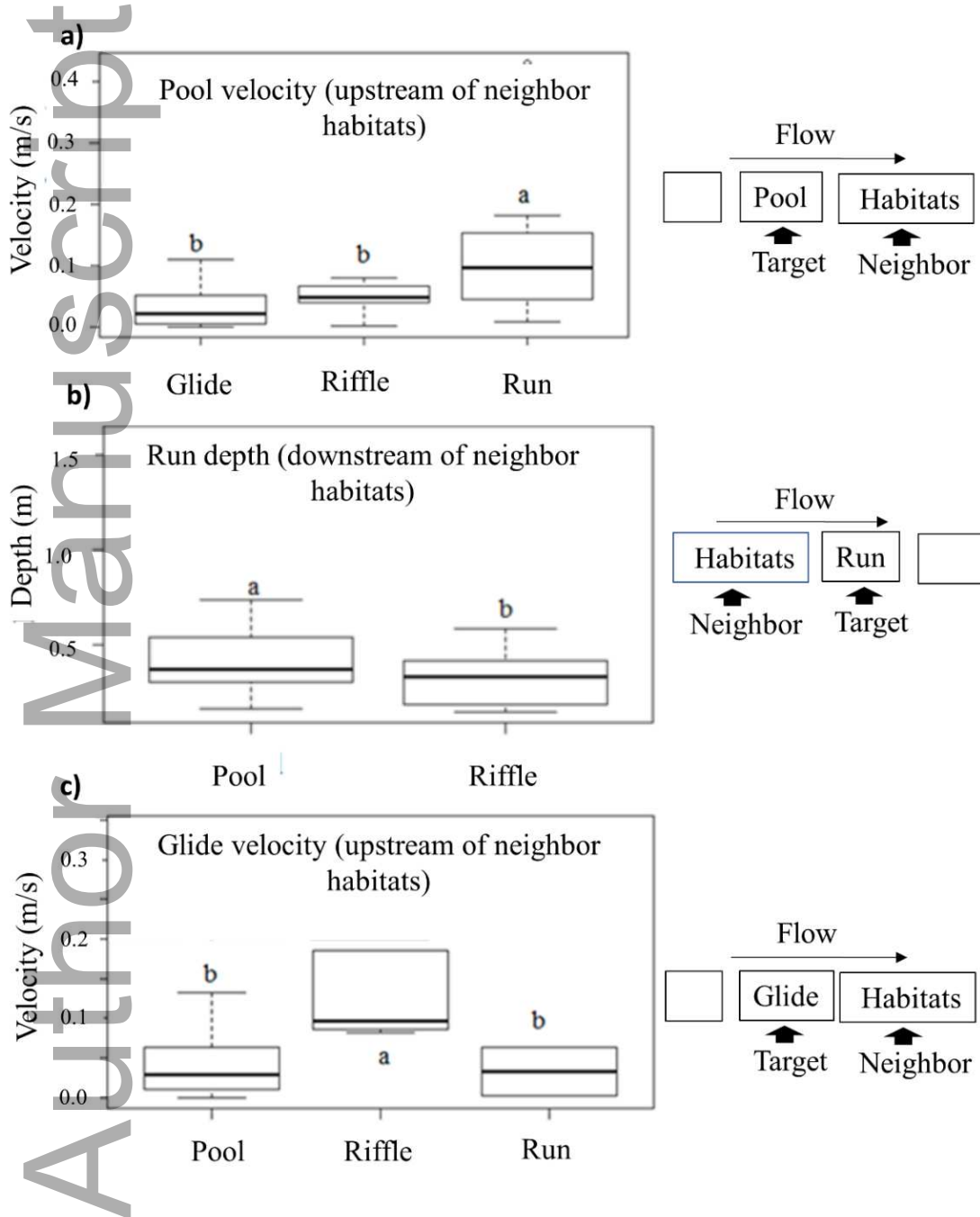


Figure 6

**Neighbor-Induced Effects on Environmental Changes and Biodiversity**

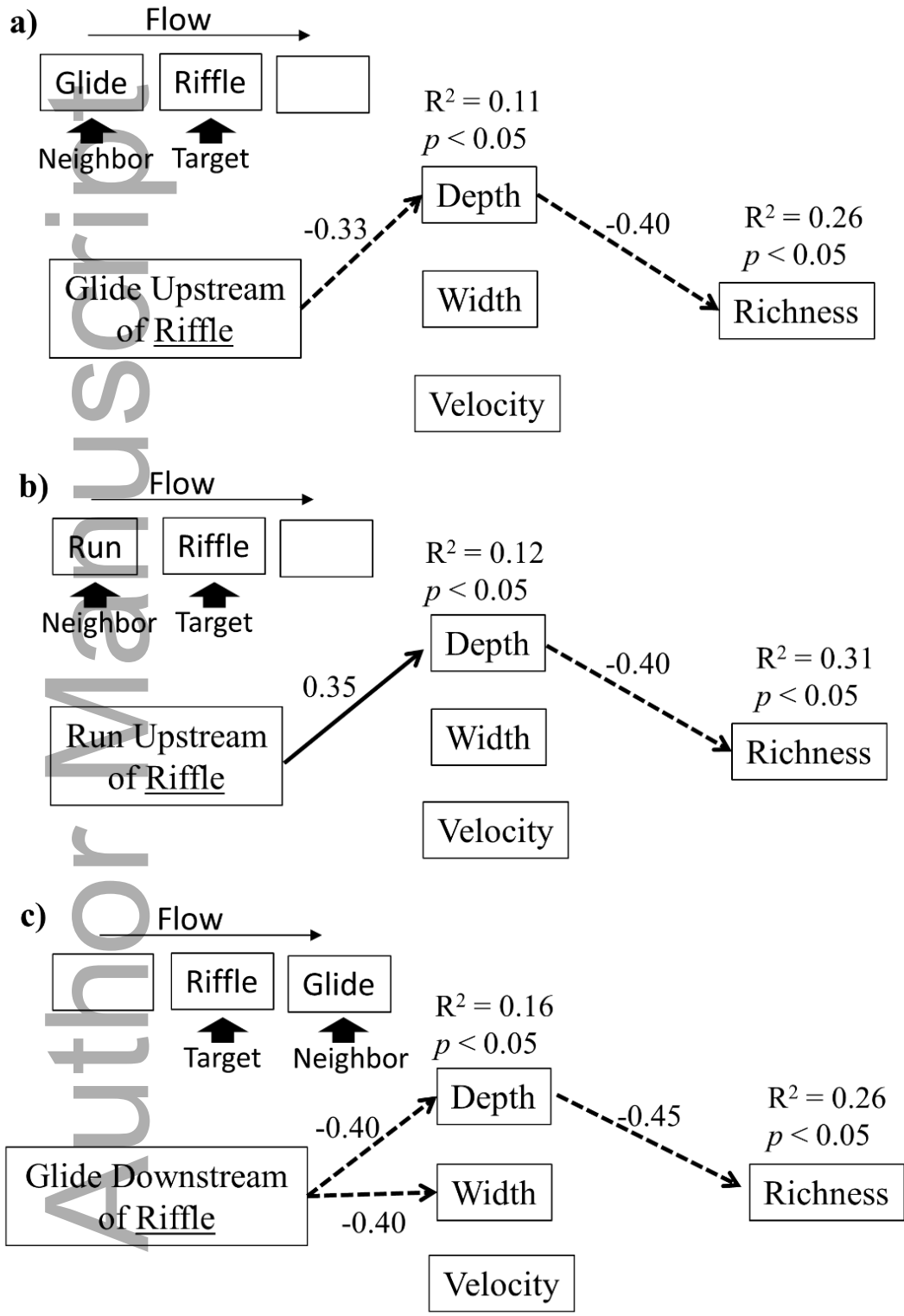


Figure 7

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