

1 **Meta-community processes supersede leaf subsidy effects on lake aquatic communities**

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16

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19

## 20 Abstract

21 There is a growing consensus that cross-ecosystem fluxes significantly influence the structure  
22 and stability of aquatic communities. What remains unclear, however, is the relative importance  
23 and potential interactions of those fluxes to/with other well-studied factors affecting aquatic  
24 community structure such as local environmental parameters or regional-scale influences related  
25 to connectivity or habitat size. In this study we compare those different hypotheses by  
26 investigating their individual and combined influence on the functional structure of littoral  
27 benthic invertebrates' communities in 7 temperate Canadian lakes (total of 23 sites) located in a  
28 semi-protected area. At each site, we sampled macro-invertebrates, we measured C, N, P content  
29 of both benthic litter and terrestrial leaves, and we recorded several standard local (pH, dissolved  
30 oxygen, turbidity, conductivity, temperature) and regional -scale (perimeter to area ratio, lake  
31 connectivity) parameters. We then used a combination of model selection and ordination  
32 approaches to disentangle the relative importance of local, regional, and cross-ecosystem factors  
33 and their interactions. When considered alone, we found that local factors related to water quality  
34 performed best at explaining variations macro-invertebrate community structure. When in  
35 combinations, models with both water quality and regional scale variables related to connectivity  
36 performed best. However, some functional groups like predator abundances were well correlated  
37 to C:N and N:P ratio in sediments (of terrestrial origin). Overall, those results suggest that if  
38 stoichiometry (i.e. nutritional quality ) of particulate terrestrial subsidies can have some effects  
39 on lake benthic communities, local and regional factors still dominate in explaining invertebrate  
40 biodiversity patterns, at least in lakes without steep human land-use gradients. Closing  
41 knowledge gaps on cross-ecosystem subsidy effects is important to improve our understanding of  
42 how communities and functioning may change under human and/or climate driven changes in  
43 watersheds.

## 44 Introduction

45           It is becoming increasingly apparent that ecological systems, even distant ones, are linked  
46 to one another through a wide range of fluxes (Gounand et al. 2018; McCann et al. 2021; Polis,  
47 Anderson, and Holt 1997). From atmospheric deposition of particles, organic and inorganic  
48 nutrient diffusion to long range animal migrations, these exchanges of energy, matter and  
49 organisms mean that changes in a single system may propagate to adjacent systems (Gravel et al.  
50 2010; Harvey, Gounand, Little, et al. 2017; McCann et al. 2021). Whether reciprocal,  
51 unidirectional, regular, asynchronous or asymmetric, these energy transfers can be major drivers  
52 of ecosystem function for the receiving systems (Gounand et al. 2020; Marcarelli et al. 2011;  
53 Richardson et Sato 2015; Yang et al. 2008). The meta-ecosystem framework is increasingly  
54 adopted by ecologists for its capacity to bridge important gaps in our understanding of the  
55 regulation of biodiversity and ecosystem function in space (Gounand, Harvey, Little, et al. 2017;  
56 Richardson et Sato 2015). However, many challenges remain in the application of those  
57 frameworks (Gounand, Harvey, Little, et al. 2017). Among those challenges, it is still unclear  
58 how important those cross-ecosystem exchanges are relative to other well-studied local and  
59 regional drivers of community structure. Past studies on meta-communities have focused on the  
60 relative importance of local (abiotic factors) versus regional (dispersal constraints) drivers  
61 among same type habitats (i.e., aquatic to aquatic, terrestrial to terrestrial, Gounand et al., 2017),  
62 while cross-ecosystem subsidy research has focused solely on isolating the effect of specific  
63 sources of subsidy among different habitat types (e.g., aquatic-terrestrial, for a review see  
64 Richardson & Sato, 2015). However, the ecological significance of cross-ecosystem subsidy is  
65 likely contingent on local and regional conditions (Collins et al. 2016; Zhang et Richardson

66 2011). Thus, developing an integrative understanding of when cross-ecosystem subsidies are  
67 ecologically relevant and by how much is essential to eventually predict how alterations to those  
68 pulse dynamics (via human-induced perturbations such as harvesting, mining or climate change -  
69 (Harvey, Gounand, Ward, et al. 2017) will propagate across ecosystem boundaries.

70         Freshwater ecosystems such as lakes are especially relevant to address those questions  
71 given the extensive amount of research previously done on the relative importance of local  
72 environmental filtering and regional scale factors in driving biodiversity patterns (for a review  
73 see Heino et al., 2021). Interestingly, despite past pioneer studies showing that coarse terrestrial  
74 carbon sources (leaf falls) could constitute a significant part of lake metabolism (France et Peters  
75 1995; Gasith et Hosier 1976) and trophic energetic pathways (Cole et al. 2006; Scharnweber et  
76 al. 2014), a lot of the work on cross-ecosystem subsidy (excluding dissolved organic carbon, see  
77 Prairie, 2008) have focused on river ecosystems (Richardson et Sato 2015). Thus, there is a clear  
78 gap in our understanding of how sensitive lakes are to particulate cross-ecosystem subsidy  
79 (compared to rivers) and of how important those cross-ecosystem fluxes are relative to local and  
80 regional drivers. This knowledge is necessary to develop an accurate “meta-system” view of lake  
81 ecosystems (*sensu* Heino et al., 2021). Known for their sensibility to local abiotic factors,  
82 freshwater benthic macro-invertebrates are often used as proxies of water quality as their  
83 diversity and richness is linked to pH, dissolved oxygen saturation and other physico-chemical  
84 processes (Heino 2005; Jacobsen 2008). These organisms are also in close contact with terrestrial  
85 particulate organic matter (e.g. leaf litter) in aquatic systems and are key to the decomposition,  
86 recycling and integration of terrestrial carbon (Wetzel 2001), making them a suitable focal group  
87 for investigating terrestrial to aquatic linkages.

88           From an applied perspective, human activities such as logging or farming are modifying  
89 the exports of matter and energy from the terrestrial system to the aquatic system (Canham et al.  
90 2004; Little et al. 2018) and we know that these subsidies can modify community stability  
91 and productivity in streams (Wallace et al. 1997). Freshwater systems receive large quantities of  
92 low-quality subsidies from their terrestrial matrix, as dissolved organic carbon (DOC) and (to a  
93 lesser extent) as fallen leaves when in forested area. This important import of carbon drives up  
94 microbial production and ecosystem respiration rate (Marcarelli et al. 2011). Beyond purely  
95 quantitative effects, it has been proposed that the stoichiometry (i.e. nutritional value) of  
96 subsidies may also be an important factor in community structure and ecosystem functions  
97 (Bartels et al. 2012; Marcarelli et al. 2011; Sitters et al. 2015). While terrestrial carbon subsidies  
98 tend to push freshwater systems toward heterotrophy, it is unclear if the stoichiometry (or  
99 nutritive value) of allochthonous materials varies enough naturally to drive difference in  
100 community structure. It is likely that context-dependency in subsidy effects is caused by  
101 interactions occurring among factors acting at different scales (water quality and/or subsidy  
102 quality and/or spatial variables).

103           With this study, we aimed to untangle the effects of leaf subsidy quality (i.e.  
104 stoichiometric ratio of C, N and P) on community structure of lake littoral benthic invertebrates,  
105 its relative importance and its potential interactions with other factors such as environmental  
106 parameters, spatial connectivity, or habitat size. To achieve this, we compared models based on  
107 the competing hypotheses that A) subsidy quality drives community structure (classic ‘cross-  
108 ecosystem subsidy’ perspective), B) water quality drives community structure (local drivers), C)  
109 spatial parameters (regional drivers) drive community structure. We tested those competing  
110 views alone and in combination. Subsidy quality was characterized as sediment stock and leaf

111 subsidy C:N and N:P ratios; water quality metrics included pH, Oxido-reduction potential  
112 (ORP), specific conductance, turbidity, and dissolved oxygen (DO); spatial parameters  
113 considered were perimeter to area ratio as a proxy of land to water linkages and lake connectivity  
114 within the watershed. Thus, our study integrates both cross-ecosystem subsidy and meta-  
115 community perspectives, which are usually studied independently (Gounand, Harvey, Little, et  
116 al. 2017) into a more explicit meta-ecosystem view. Based on the limited evidence available on  
117 the effect of particulate subsidy in lake ecosystems (France et Peters 1995; Gasith et Hosier  
118 1976; Scharnweber et al. 2014), we expected that local and regional drivers would dominate, but  
119 that the inclusion of subsidy stoichiometry would significantly enhance the models and that  
120 higher nutritional value would lead to higher richness and diversity.

121

## 122 Methods

### 123 **Site Location and Selection:**

124 The sampling for this project was done at the University of Montreal's research station in  
125 the meridional Laurentians mountains, approximately 75 kilometers north-west of Montreal.  
126 Sampling was done in 7 lakes with low human influence (mostly from research projects), during  
127 the spring (May) and summer (July) of 2021. During each campaign, 23 sites were sampled with  
128 20 sites repeating in both sampling campaigns. Three sites sampled in spring were abandoned  
129 and replaced because of logistical difficulties. Prior to sampling, the area and perimeter of all  
130 lakes were measured using satellite imagery and Google Earth Pro (Google, LLC). To determine  
131 sampling effort, lakes were binned together based on their perimeter. Four size categories were

132 generated. Smallest lakes were sampled at 2 different locations with larger lakes being sampled  
133 up to 5 times. Perimeter was chosen over area as a measure of size due to its closer relation to  
134 terrestrial subsidies and littoral habitats.

135 **Sampling:**

136         Sampling of invertebrates was based on the Ontario Benthos Biomonitoring Network's  
137 protocol (Jones et al. 2007). At every site, invertebrates were sampled in three 1.25m transects  
138 perpendicular to shoreline, using a rectangle 500 micrometer mesh hand-held net with a  
139 collection cup. Material captured by the net was transferred into sieves (2mm, 1mm and 0.5mm)  
140 and abundantly rinsed with water. Material caught by the wider mesh sieve (>2mm) was sorted  
141 on site using a white tray to increase contrast. Sorted invertebrates were preserved in 95%  
142 ethanol and stored in cool and dark containers. The remaining coarse material (i.e. coarse  
143 sediments sorted from the larger invertebrates) was kept in sealed plastic bags for stoichiometry  
144 analyses, some of these samples were sub-sampled as material volume was too big. The  
145 remainder of the net samples (0.5mm < 2mm, unsorted) were preserved in 95% ethanol, in a 1L  
146 container to be sorted later.

147         For each site, measures of water quality (i.e. pH, turbidity, conductivity, temperature  
148 ORP and dissolved oxygen) were taken using a YSI ProDSS handheld multiparameter probe  
149 (YSI Inc/Xylem Inc.). Three measures were taken at every site and averaged.

150         Each site was revisited in the Fall, after leaf fall, to sample fallen leaves. Three 0.25m<sup>2</sup>  
151 quadrats were deposited in the water upon arrival on site, on top of the collection transects, and  
152 leaves within the quadrats were collected and stored in sealed plastic bags. Only recognizable  
153 leaves were collected as we presumed these were only recently fallen onto the sites.

154 **Processing samples:**

155           The fine material (< 2mm) that was collected by the net was sorted using a stereoscopic  
156 microscope. The volume of material varied a lot ranging from around 200mL to over 1L. The  
157 sorting was limited to 150mL (randomly collected from the 1L container) for each site, to keep  
158 the task manageable while reducing size bias in the sample. Identification of invertebrates was  
159 done using *An Introduction to the Aquatic insects of North America* (Merritt, Cummins, et Berg  
160 2009) and *Guide d'identification des principaux macroinvertébrés benthiques d'eaux douces du*  
161 *Québec* (Moisan 2010), as well as online resources such as [www.macroinvertebrates.org](http://www.macroinvertebrates.org) for  
162 images and additional information on groups. Individuals were identified at the family level,  
163 except for Annelids that were identified at the phylum. The resulting dataset includes about 19  
164 000 organisms divided into 79 taxonomic groups. We attributed each taxonomic group to one of  
165 seven functional groups (i.e. filterers, gatherers, macrophyte piercers, parasites, predators,  
166 scrapers and shredders) based on Merritt, Cummins and Berg 's grouping from *An Introduction*  
167 *to the Aquatic insects of North America*, with Annelids being ignored because of the higher  
168 degree of uncertainty concerning their identification.

169 **Stoichiometry:**

170           To measure spatial variation in the quality of terrestrial subsidy, carbon, nitrogen and  
171 phosphorus content were extracted from i) coarse material sampled with the collection net  
172 (littoral sediment) and ii) leaves collected in the Fall. The coarse material sampled with the net  
173 served as a proxy of terrestrial (CN ratios all higher than 15; Ishiwatari & Uzaki, 1987; Prah et  
174 al., 1994) resource stocks in the littoral zone, while freshly fallen leaves represented the yearly  
175 pulse of terrestrial resources.



176 Coarse (>2mm) material sorted on site was shredded before being freeze-dried and  
177 ground to a fine powder. Collected leaves were identified to the genus level (when possible) and  
178 dried in an oven at 40 Celsius for 12 hours. Dried leaves were then weighed before being  
179 grinded.

180 Carbon and Nitrogen content were measured by gas chromatography and electro-thermal  
181 detection using a Thermo Fisons EA1108 elemental analyser (Thermo Fisher Scientific, U.S.A).

182 Concentration of phosphorus was measured using a colorimetry method (Murphy et Riley  
183 1962) on a solution of digested material (Wilson, Braithwaite, et Brian Pyatt 2005). Analysis was  
184 conducted using Environment Canada's TH-2 Standard for phosphorus, with an average  
185 extraction efficiency of  $88.6\% \pm 0.73\%$ .

## 186 **Data analysis:**

### 187 *db-RDA Models*

188 To untangle the effects of subsidy stoichiometry on spatial variations in community  
189 structure, we built several multivariate models to compare their fit. Explanatory variables were  
190 split into four categories. Each category was based on a generic hypothesis concerning  
191 determination of community structure. The first category included variables relating to the  
192 stoichiometry of the collected material (first hypothesis; cross-ecosystem subsidy), it included:  
193 C:N ratio of sediments, N:P ratio of sediments, C:N ratio of leaf subsidy, N:P ratio of leaf  
194 subsidy. The second category included variables relating to water quality (second main  
195 hypothesis; local parameters) captured by the multiparameter probe, it included : specific  
196 conductance ( $\mu\text{S}/\text{cm}$ ), turbidity (FNU), pH, oxidation-reduction potential (ORP, mV), optical

197 dissolved oxygen (ODO, mg/L). Category 3 (perimeter to area ratio) and 4 (lake degree  
198 connectivity) represented our third main hypothesis (spatial drivers) but they were split for  
199 conceptual clarity because they represent different spatial processes. Perimeter to area ratio is a  
200 proxy for lateral connectivity with the terrestrial matrix and relates to inputs of dissolved organic  
201 carbon from the watershed. Lake degree connectivity represents longitudinal connectivity among  
202 lakes and thus represents limitations in the exchange of organisms (dispersal) and matter. All  
203 possible combinations of one, two, three or all four categories were tested (n=15). All models  
204 included sampling season (Spring or Summer) to factor in the effect of time on community  
205 variations. All 15 models were compared using adjusted-R<sup>2</sup>. Unlike with linear models, an  
206 Akaike Information Criterion score cannot be calculated on multivariate models. However,  
207 similarly to AIC, adjusted-R<sup>2</sup> imposes a penalty on each variable addition. Models were  
208 constructed using *vegan*'s *capscale()* function (Oksanen et al. 2022). A Hellinger transformation  
209 was applied to community data and explanatory variables were scaled and centered around  
210 means to make variables of different scales and units comparable. We used a bray-curtis  
211 dissimilarity to emphasize changes in community composition and relative abundances  
212 (Anderson et al. 2011). Ordination of these db-RDA models were done with weighted-average  
213 scores because these scores are less sensitive to environmental noises (Oksanen 2020) and  
214 scaling 2 was used to illustrate correlation between variables/vectors (Legendre and Legendre  
215 2012).

### 216 *Linear Models*

217           Using the same methodology to construct models, we compared 15 linear models of  
218 species richness and Shannon-Wiener diversity index based on their AIC scores. We included a  
219 null linear model to this comparison.

## 220 *Simple Regressions*

221 To further investigate apparent correlations shown by the db-RDA ordination, we  
222 constructed simple linear regressions of vectors that had similar direction (angles between  
223 vectors reflect correlation in ordination scaling 2 (Legendre & Legendre, 2012)).

224

## 225 **Results**

226 Our primary objective was to disentangle the relative influence of meta-ecosystem (i.e.  
227 cross-ecosystem subsidy) and meta-community (i.e. local and regional) drivers on community  
228 structure and biodiversity. The results from our multivariate model comparison indicated that the  
229 model including all parameters was the best and most parsimonious at explaining variations in  
230 community structure (adj.-R<sup>2</sup>= 0.644, see Table 1). When looking at one-way models, the model  
231 including only variables related to local water quality performed the best (adj.-R<sup>2</sup> = 0.330, which  
232 corresponds to more than half of the explained variance by the full model). Among the different  
233 model comparisons, any models combining local water quality and regional connectivity  
234 variables performed better than models excluding either (Table 1). Models with terrestrial  
235 subsidy attributes explained less community variation than other models, except when in  
236 combination with water quality and connectivity metrics. Our results showed evidence of  
237 interactions among parameter sets. For instance, the adj.-R<sup>2</sup> of the two-way models including  
238 water quality and terrestrial subsidy stoichiometric variables did not correspond to the sum of the  
239 corresponding one-way models, suggesting that the effect of terrestrial subsidy varies depending  
240 on local water quality attributes (Table 1).

241

242 Figure 1 presents the ordination for the best model, combining all variables. Overall, a  
243 centroid-to-centroid permutation analysis revealed that lakes differed significantly in functional  
244 group composition ( $F=4.2196$ ,  $p\text{-value}=0.0001$ ), but had similar intra-group variance (no  
245 difference in beta-diversity,  $F=1.1592$ ,  $p\text{-value}=0.3494$ ). A marginal test of significance on the  
246 db-RDA (at  $p=0.05$ ) identified perimeter to area ratio, lake connectivity, specific conductance,  
247 turbidity and sediments N:P ratio to be the most important variables to the model. The ordination  
248 showed several apparent correlations between functional groups abundance and explanatory  
249 variables that we further tested with linear regression analyses. Among these, scrapers abundance  
250 increased with increasing specific conductance ( $R^2=0.48$ , Figure 2); predator abundance declined  
251 with higher C:N ratios ( $R^2=0.29$ , Figure 2, middle panel); and gatherer abundance declined with  
252 increasing pH ( $R^2=0.24$ , Figure 2 bottom panel). Phosphorus concentrations were excluded from  
253 the db-RDA analysis due to collinearity concerns, but nonetheless it had significant influences on  
254 some functional group abundances. Notably its concentration in sediments was positively  
255 correlated to predator abundance ( $R^2=0.44$ ) and shredder abundance ( $R^2=0.12$ ). Interestingly its  
256 concentration in sediments (positive relation,  $R^2=0.3$ ) and leaves (negative relation,  $R^2=0.11$ ) had  
257 opposite effects on gatherer abundance (see Appendix S1, Figure S1).

258

259 All linear models of macro-invertebrate species richness (range = 6 to 26 species) scored  
260 similarly or above the null model. (as expressed by AICc, see Table 1). Therefore, none of the  
261 considered variables seemed to have any effect on species richness. On the other hand, variation  
262 of diversity indexes (range = 0 :4.37, the latter being the theoretical maximum since 79  
263 taxonomic groups were considered) was best explained by a two-way model. The best model  
264 combined water quality related factors and lateral connectivity (perimeter to area ratio). Adding

265 stoichiometry of terrestrial subsidies did not enhance the model fit. Analysis of variance showed  
266 perimeter to area ratio, turbidity and pH be the significant terms.

267

## 268 Discussion

269 Our main objective was to disentangle the relative influence of cross-ecosystem subsidy  
270 and meta-community related drivers on lake macroinvertebrates community structure and  
271 biodiversity. Comparison of multivariate models showed that, when considered separately, local  
272 water quality best predicted macro-invertebrate community structure. Nutritional quality of  
273 subsidies and spatial variables, on their own, explained significantly less variation. However,  
274 adding among-lake connectivity to water quality led to an important gain in explanatory power  
275 (adj.-R<sup>2</sup> of 0.470 vs. 0.330), suggesting that longitudinal connectivity can interact with local  
276 water quality in driving benthic community structure. The addition of both spatial variables  
277 (longitudinal and lateral connectivity) leads to an even greater gain in explanatory power (adj.-  
278 R<sup>2</sup>=0.533). The inclusion of terrestrial material nutritional quality in the models did increase  
279 predictive power marginally (adj.-R<sup>2</sup>=0.354 vs. adj.-R<sup>2</sup>=0.330). Nutritional quality of leaf  
280 subsidy does seem to affect species abundances and community structure to some degree, but the  
281 increase in predictive power is especially important when also including longitudinal  
282 connectivity (adj.-R<sup>2</sup>=0.567). The complete model, including all variables, scored the highest  
283 adjusted-R<sup>2</sup> (0.644). Marginal testing of its variables indicates that among-lake connectivity  
284 (longitudinal connectivity) and lake perimeter to area ratio (lateral connectivity) were among the  
285 most important variables. These regional factors scored lowest when considered alone, but this  
286 result underlines the interacting influence of spatial drivers on local ones and highlights the

287 importance of considering space in studying communities (and meta-communities), and  
288 especially their interacting influence.

289         The ordination of the best model showed interesting correlation between specific  
290 functional group abundances and variables relating to nutritional quality, local and regional  
291 drivers. Specifically, the abundance of scrapers was correlated with water conductance (simple  
292 linear regression with a  $R^2$  of 0.48, see Figure 2), which is probably related to the presence of  
293 calcium (Ca, among other ions) necessary to snails (the most abundant scraper organisms) for  
294 their shells. The abundance of predators was negatively correlated to the C:N and N:P ratio of  
295 sediments (a proxy of subsidy accumulation) ( $R^2 = 0.29$  and  $0.25$  respectively, see Figure 2), but  
296 positively correlated to phosphorus concentration ( $R^2=0.44$ , see Appendix). This is interesting  
297 because predators do not feed directly off this material, therefore this supports the idea that  
298 terrestrial subsidy quality can drive bottom-up trophic cascades (Bartels et al. 2012; Polis et  
299 Strong 1996). Results also show that gatherer abundances were negatively associated with pH  
300 ( $R^2=0.24$ , see Figure 2) and concentration of phosphorus in leaves (fresh subsidy,  $R^2=0.11$ , see  
301 Appendix), but positively associated with phosphorus concentration in sediments ( $R^2=0.3$ , see  
302 Appendix). This result could suggest that fresh material does not follow the same energy  
303 pathway as sediment material. It would be in line with a recent study showing that initial  
304 phosphorus excretion in leaf subsidy mainly feeds the plankton loop (Hirama et al. 2022).  
305 Shredder abundances were negatively associated with phosphorus content of sediments  
306 ( $R^2=0.12$ , see Appendix). Overall, our results suggest that subsidy nutritional quality influences  
307 the structure of the community by affecting the abundance of specific functional groups. Our  
308 study was done in a relatively homogeneous protected area representing a sort of baseline  
309 variation and effects of leaf subsidy on communities. Those influences on specific functional

310 groups suggest that perturbations leading to abrupt changes in subsidy quality could potentially  
311 lead to more noticeable changes in community structure.

312         The comparison of linear models of diversity index also shows that water quality and  
313 spatial parameters seem of great importance. The best fit (as suggested by AICc) was obtained  
314 with the model combining water quality metrics with lateral connectivity. Contrary to our initial  
315 hypothesis, we found no relation to species richness and therefore variation in diversity index  
316 score seems mostly driven by changes in evenness. As with the multivariate models, the  
317 exclusion of local water quality metric significantly lowers model fit. Although regional drivers  
318 and to some extent terrestrial subsidy nutritional quality can modify community structure, these  
319 results suggest that local drivers are essential to the determination of community.

320         Our results indicate that natural variation of leaf subsidies stoichiometry is not big enough  
321 to drive major differences in macroinvertebrate community structure, in lakes. While terrestrial  
322 subsidy nutritional quality did appear to impact community structure as it was correlated to some  
323 groups' abundance and increased predictive power of db-RDA models of community structure,  
324 the effect of local water quality and spatial connectivity (lateral or longitudinal) supersedes the  
325 influence of terrestrial particulate organic matter stoichiometry. Possibly, insects' dispersion and  
326 colonization occur independently from ecosystems subsidy dynamics, and therefore terrestrial  
327 leaf nutritional quality would influence communities only after their initial establishment from  
328 egg deposition. Hence it appears to modulate groups' abundance and community structure mostly  
329 through evenness. All sites were adjacent to forested shorelines and within watersheds  
330 dominated by forest cover. Therefore, none of the variation is linked to changes in land-usage.  
331 Introduction or exclusion of new tree species through species distribution shifts northwards  
332 (Chen et al. 2011) or human intervention may lead to greater variation in riparian plant

333 communities. Whether this would lead to greater variation in benthic communities would need to  
334 be investigated, but the results showed here suggest it could be the case. This also has  
335 implications for restoration projects targeting riparian habitats, as it suggests the choice of plants  
336 used to revegetate will impact functional groups of benthic communities adjacent to these sites.  
337 Overall, our results illustrate well the need for more empirical tests of the meta-ecosystem  
338 concept and the potential context-dependency of cross-ecosystem subsidy effects on community  
339 structure and functioning. This is essential to developing an integrative biodiversity and  
340 ecosystem function management framework at landscape extent.

341

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## 347 Authors contributions

348 GEH and EH designed the research project, GEH conducted the field work, conducted data  
349 processing and statistical analyses with support from EH and wrote the first draft of the  
350 manuscript. Both authors contributed to further manuscript revisions.

351



## 352 Conflict of interest statement

353 Authors do not have any conflict of interest to declare.

354

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489

490 Tables

491 **Table 1: Comparison of multivariate and linear models.** Comparison is based either on their Adjusted-R<sup>2</sup> or their second order  
 492 Akaike Information Criterion Score. For all three model types, the better model's score is framed by a black box.

Models	Comparative Criteria	0.Null	1.Terrestrial	2.Water Quality	3.Lateral Connectivity	4.Longitudinal Connectivity	5.Terrestrial + Water	6.Terrestrial + Lat. Co	7.Terrestrial + Lon. Co.	8.Water + Lat. Co.	9.Water + Lon. Co.	10.Lat. Con. + Lon. Co.	11.Terr. + Water + Lat. Co.	12.Terr. + Water + Lon. Co.	13.Terr. + Lat. Co. +Lon. Co.	14.Water + Lat. Co. +Lon. Co.	15.Complete
db-RDA, Funct.Gps.	R <sup>2</sup> -ajd.	NA	0.140	0.330	0.059	0.123	0.354	0.161	0.304	0.345	0.470	0.179	0.385	0.567	0.351	0.533	0.644
Linear, Richness	AICc	243.0	247.4	248.2	244.5	245.4	259.4	250.4	250.4	250.9	251.2	246.5	263.4	263.5	253.1	254.3	267.9
Linear, Diversity	AICc	30.8	30.2	28.6	29.4	33.7	31.8	32.3	32.2	23.7	26.6	31.7	30.1	27.7	35.3	26.6	31.7

493

## 494 Figure Captions

495 **Figure 1: Ordination of sites community structure, db-RDA model (Adjusted-R<sup>2</sup>=0.644).**

496 This ordination uses weighted-average scores and scaling 2. It shows sites distanced by their  
497 community composition and related to 12 explanatory variables (complete model). Sites were  
498 clustered in 7 groups showing the seven lakes that were sampled for this project. Sites were  
499 sampled for invertebrates in spring 2021 (n=20) and again in summer 2021 (n=23), each point  
500 representing a single invertebrate sample (n=43).

501

502 **Figure 2: Simple regressions showing linear relations suggested from vectors association of**

503 **the db-RDA ordination.** Top panel shows a strong linear relationship between scraper  
504 abundance and specific conductance. Middle panel shows strong linear relationships between  
505 predator abundance and nutritional quality of sediments (terrestrial subsidy stock). In red,  
506 regression and R<sup>2</sup> for C:N ratio. In blue, regression and R<sup>2</sup> for N:P ratio. Bottom panel shows a  
507 good linear relationship of gatherer abundance with pH.

508

509 **Figure 3: Simple linear regressions of Shannon-Wiener diversity index in relation with the**

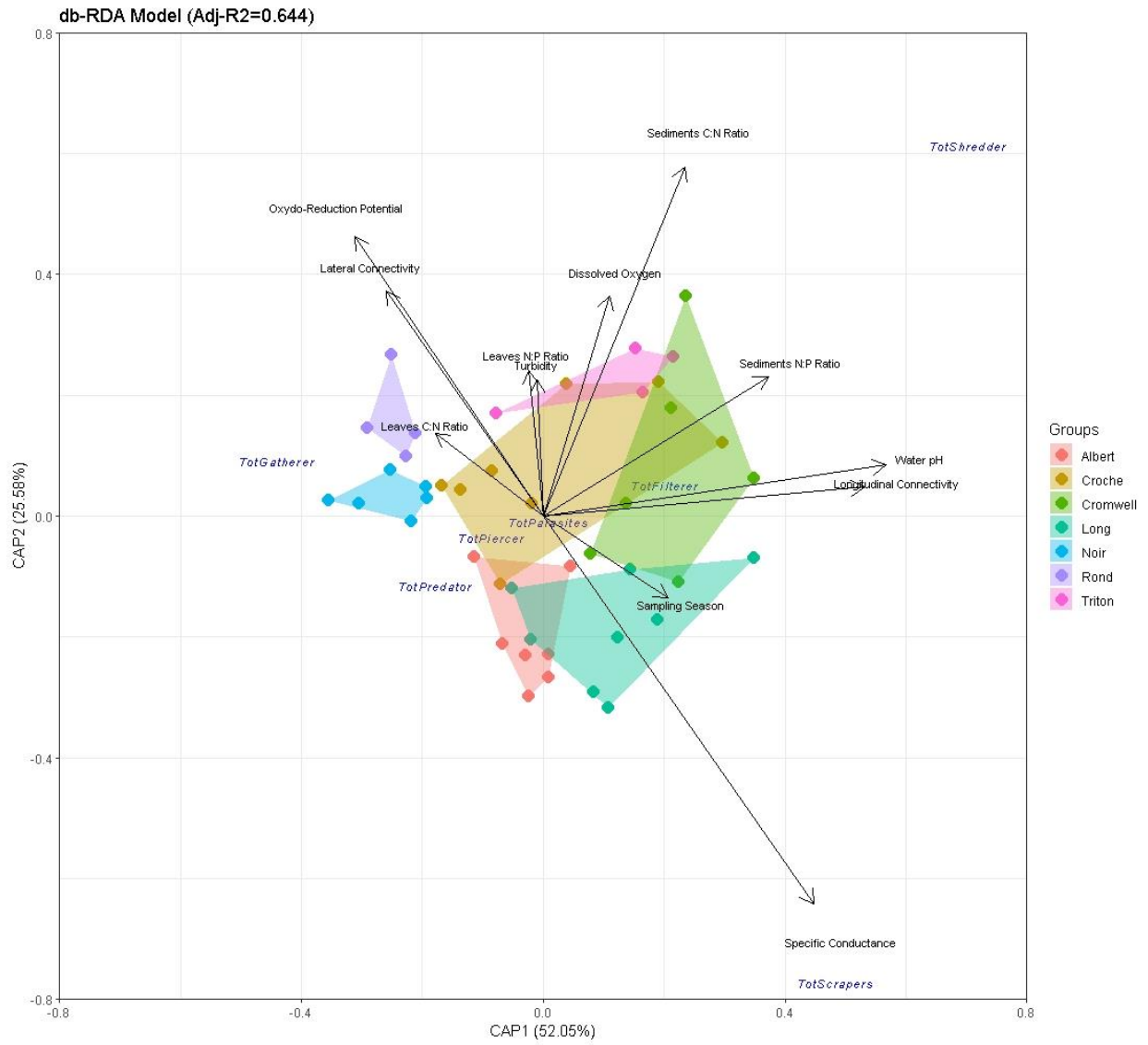
510 **significant variables from the linear model scoring the lowest AIC.** Top graph shows the  
511 relation of the diversity index with pH (water quality). Middle panel shows the relation between  
512 diversity index and turbidity (water quality). Bottom panel shows the relation between diversity  
513 index score and lateral connectivity (perimeter to area ratio).

514



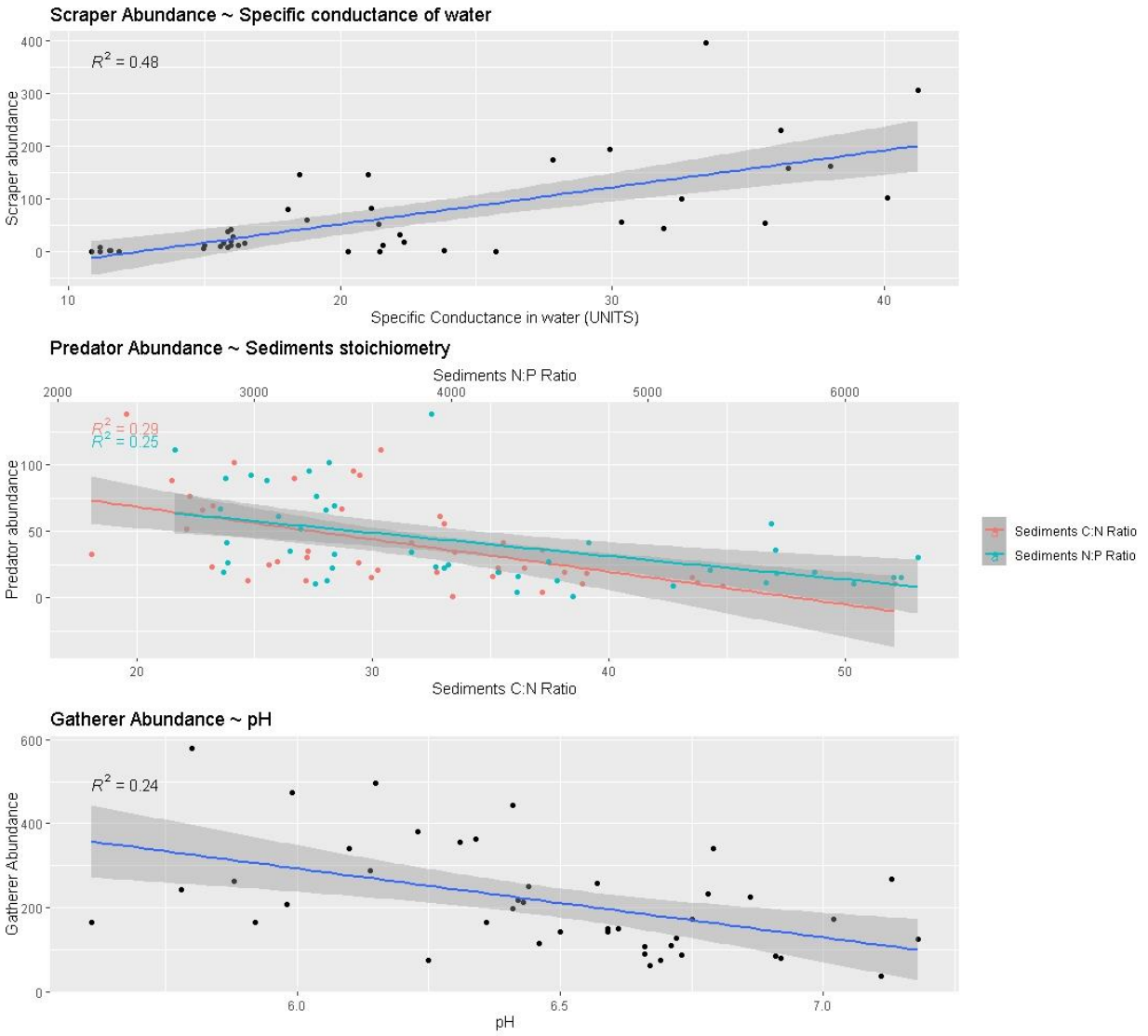
515 Figures

516 Figure 1:



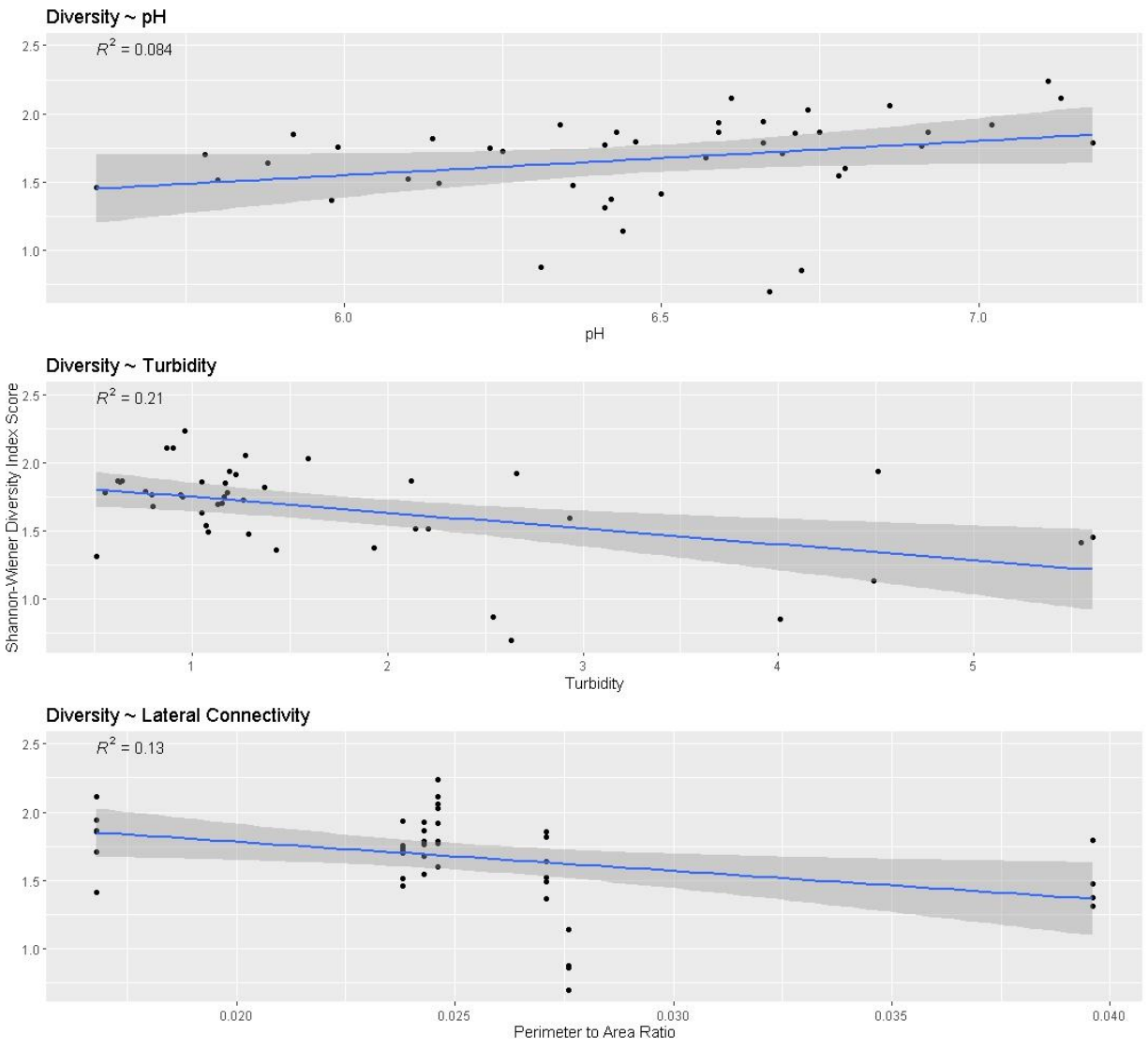
517  
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519 Figure 2:



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522 Figure 3:



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