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3 **Combining taxon-by-trait and taxon-by-site matrices for analysing trait patterns**  
4 **of macroinvertebrate communities: a rejoinder to Monaghan & Soares (2014)**

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25 Running head: How to combine taxa-by-trait and taxa-by-site matrices?

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28 Keywords: macroinvertebrates, trait-based analyses, weighting

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32 **Summary**

33

- 34 1. Monaghan & Soares (2014) suggested that combining traits with log-transformed  
35 abundance of taxa may cause anomalies in analyses of stream macroinvertebrate  
36 communities. While they addressed an important issue in stream ecology, here we  
37 present an opposite view. To identify the causes of these contrasting opinions, we  
38 carefully examined the examples provided by Monaghan & Soares (2014) and  
39 demonstrated how traits can be weighted by the presence, abundance and log-  
40 transformed abundance of the taxa in a meaningful way.
- 41 2. We found that Monaghan & Soares (2014), following other authors, use the term  
42 ‘weighting’ differently from classical papers of stream ecology. The general  
43 practice is to calculate the sum of trait values multiplied by the abundance of each  
44 taxon and divide it by the total invertebrate abundance to get a community-level  
45 trait value. In contrast, Monaghan & Soares (2014) did not perform the final  
46 division and consequently did not get a standardized community-level trait value.  
47 It follows that the term "weighting" is used with different meanings in stream  
48 ecology, and ecologists should keep these differences in mind.
- 49 3. We agree with Monaghan & Soares (2014) that the addition of log-transformed  
50 data is equivalent to multiplication on an arithmetic scale. However, we disagree  
51 that this provides an inconsistent scaling that confounds quantitative analyses.  
52 Using example data sets, we illustrate how trait-based data analysis can be  
53 performed in community ecology in a meaningful way.

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55

56 **Introduction**

57

58 In a recent paper, Heino, Schmera & Erős (2013) provided an overview of trait  
59 patterns of stream communities from a macroecological perspective. In this paper,  
60 reference was made to Gayraud *et al.* (2003), who showed that abundance-weighted  
61 traits were less powerful than presence-absence weighted traits in discriminating  
62 communities under different degrees of human impacts, and to Stutzner & Beche  
63 (2010), who therefore advocated the use of the second type of weighting for practical  
64 biomonitoring due to logistic constraints (e.g. sorting of qualitative samples makes

65 assessment programmes more cost effective). Heino *et al.* (2013) pointed out that  
66 these findings contradict the results of taxon-based analyses which suggest that, if  
67 taxa are weighted by their abundance, then communities are better separated in  
68 relation to environmental variation than when taxa are weighted only by their  
69 presence. Based on an overview of the literature, Heino *et al.* (2013) concluded that  
70 results of presence- and abundance-based analyses should be evaluated carefully  
71 when examining traits of organisms, because differences among studies can reflect  
72 both methodological (i.e. handling of data) and real ecological differences (see p.  
73 1549 in Heino *et al.*, 2013). More recently, Monaghan & Soares (2014) stated that (1)  
74 Heino *et al.* (2013) identified the weak explanatory power of abundance data as a  
75 major limitation of macroinvertebrate trait analysis and that (2) the log-transformation  
76 of abundance data may cause anomalies in trait-based analyses. We disagree with  
77 both conclusions, because (1) Heino *et al.* (2013) did not actually state this (see  
78 above) and because (2), in our view, log-transformation of abundance data in trait-  
79 based analyses can also be meaningful. To reveal the causes of these differing views,  
80 we go through the examples provided by Monaghan & Soares (2014) and examine  
81 how traits can be weighted by the presence, abundance and log-transformed  
82 abundance of the taxa. To do this, first we define the terminology used here, comment  
83 on the approach of Monaghan & Soares (2014) and show how this procedure should  
84 be performed.

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## 87 **Terminology**

88

89 Characterizing taxa by their traits is a challenging task in ecology. A research team  
90 (Stazner *et al.*, 1994) studying the Upper Rhone River in France made a seminal  
91 contribution by determining which traits can be used and how to examine  
92 macroinvertebrate communities from a 'functional' perspective. High heterogeneity  
93 of organisms, remarkable variation within taxa, gaps and uncertainties in our  
94 knowledge led them to adopt a "*fuzzy coding*" system (Chevenet *et al.*, 1994). Fuzzy  
95 coding is based on expert opinion and assigns an integer score, ranging from 0 (no  
96 affinity) to an arbitrary maximum, often 5 (high affinity), to express the relationship  
97 of a taxon to a particular trait. To improve comparability, scores are often

98 standardized by the total number of scores within a taxon (Usseglio-Polatera *et al.*,  
 99 2000, Bady *et al.*, 2005, Dolédec *et al.*, 2006).

100

101 Information on taxa characterized by different traits can be summarized by a taxon-  
 102 by-trait matrix (**A**) of size  $s \times t$ , in which rows represent taxa and columns correspond  
 103 to traits describing a well-defined character (e.g. the feeding habit) of  
 104 macroinvertebrates. More formally,

105

$$106 \quad \mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & & a_{1t} \\ a_{21} & a_{22} & & a_{2t} \\ & & a_{ij} & \\ a_{s1} & a_{s2} & & a_{st} \end{bmatrix}$$

107

108 Each entry in the matrix ( $a_{ij}$ ) is determined such that  $\sum_{j=1}^t a_{ij} = 1$  for every row  $i$  (see  
 109 Bady *et al.*, 2005). In a sense, the  $a_{ij}$  values express the relative importance of trait  $j$   
 110 for taxon  $i$  or, in other words, the relative affinity of taxon  $i$  to trait  $j$ . For example, if a  
 111 given taxon acts exclusively as a shredder, it cannot be a predator or anything else:  
 112 only one value can be 1, and all the others are zero (see also Chevenet *et al.*, 1994;  
 113 Dolédec *et al.*, 2006). Other possibilities, with  $0 < a_{ij} < 1$  are shown in the upper part  
 114 of Table 2 of Monaghan & Soares (2014). Note, however, that zero values are missing  
 115 from that matrix, which takes a mathematically correct form as given below:

116

$$117 \quad \mathbf{A} = \begin{bmatrix} 0.17 & 0 & 0.5 & 0.33 & 0 \\ 0 & 0.75 & 0.25 & 0 & 0 \\ 0 & 0.2 & 0 & 0.2 & 0.6 \\ 0 & 0.5 & 0 & 0 & 0.5 \\ 0 & 0.4 & 0.6 & 0 & 0 \\ 0 & 0 & 0 & 0.75 & 0.25 \\ 0.25 & 0 & 0.75 & 0 & 0 \end{bmatrix}$$

118

119 Sampling sites characterized by different taxa can be summarized by a taxon-by-site  
 120 matrix (**B**) of size  $s \times z$ , in which rows represent taxa, and the columns correspond to  
 121 sampling sites. Each entry  $b_{ik}$  in **B** is the abundance of taxon  $i$  at site  $k$ :

122

$$123 \quad \mathbf{B} = \begin{bmatrix} b_{11} & b_{12} & & b_{1z} \\ b_{21} & b_{22} & & b_{2z} \\ & & b_{jk} & \\ b_{s1} & b_{s2} & & b_{sz} \end{bmatrix}$$

124

125 The top left part of Table 2 in Monaghan & Soares (2014) displays such a matrix for  
126 seven taxa and a single site, so it is in fact a column vector, abbreviated as **b**:

127

$$128 \quad \mathbf{b} = \begin{bmatrix} 80 \\ 50 \\ 100 \\ 20 \\ 40 \\ 30 \\ 20 \end{bmatrix}$$

129

130

### 131 **Comments on the approach of Monaghan & Soares (2014)**

132

133 Monaghan & Soares (2014) write that, if standardized trait values are used "as a non-  
134 linear weighting for trait descriptions", then the "practitioners should be aware of the  
135 potential of inconsistent weighting". This causes some uncertainty, because the term  
136 'weighting' in Monaghan & Soares (2014) is different from that used by Charvet *et al.*  
137 *et al.* (2000) and Statzner *et al.* (2001). Charvet *et al.* (2000) and Statzner *et al.* (2001)  
138 summed the trait values multiplied by the abundance of each taxon and divided by the  
139 total invertebrate abundance to get a community-level trait value. In contrast,  
140 Monaghan & Soares (2014), following Archambault *et al.* (2005) and Larsen &  
141 Ormerod (2010), did not perform the final division, only summed the trait values  
142 multiplied by the abundance of each taxon. In other words, some studies weight by  
143 proportions, whereas others by abundance. It follows that the term "weighting" has  
144 been used with at least by two different meanings in stream ecology. Although the  
145 proper measurement of the overall community trait values is still debated (Ricotta &  
146 Moretti, 2011), we argue that, from a statistical point of view, *weighting* means that

147 some elements have more impact on the result than other elements in the same set.  
148 Stream ecologists should be aware of these two variants when comparing different  
149 results.  
150  
151 Monaghan & Soares (2014) exemplify this inconsistent weighing by back-  
152 transformation, and emphasise that the "addition of log-transformed data is equivalent  
153 to multiplication on an arithmetic scale" (point 2 in their Summary). From a  
154 mathematical point of view we agree, but note that accepting this point would exclude  
155 all kinds of nonlinear transformations (square-root, logarithmic, etc.) from the toolkit  
156 of statistical data analysis. Nonlinear transformations modify the original data  
157 structure in a meaningful way and, if properly used, facilitate significance tests by  
158 decreasing skewness in the data (Elliot, 1977; Podani, 2000; Legendre & Legendre,  
159 2012). Furthermore, community ecologists very often use log transformation for  
160 'equalization' purposes, that is, to balance the importance of variables (species, or  
161 other taxa) before cluster analysis or ordination. The result of this operation is that  
162 species with highly unequal abundance will be almost equally influential in  
163 multivariate analysis. Non-linear transformation can be conceived as a parameterized  
164 series, with raw abundances at one end and presence-absence data at the other (e.g.,  
165 Clymo transformation, see Podani, 2000). In any case, it is fundamental that all data  
166 are transformed by the same method *before* any other calculations are performed.  
167 Since the scale of data (i.e. linear vs. nonlinear) is selected completely arbitrarily by  
168 the investigator (pH is measured mostly at a nonlinear scale, while individuals are  
169 counted mostly at a linear scale), changing from a linear to a nonlinear scale, or *vice*  
170 *versa*, is acceptable. Finally, we do not know of any situation when back-  
171 transformation to the original scores would be necessary, so that the warning by  
172 Monaghan & Soares (2014) is mainly of academic significance.

173

174

### 175 **Combining traits with the abundance of a single taxon**

176

177 Monaghan & Soares (2014) state that "the abundance of taxa represented by  
178 respective trait classes is obtained by multiplying the total abundance by the  
179 proportionate frequencies of the traits". Moreover, their Table 1 exemplifies the  
180 situation and shows how this procedure runs with the feeding habits of larvae of the

181 trichopteran *Psychomyia* using raw abundance and standardized trait values (Table 1a  
182 of Monaghan & Soares, 2014), using the log-transformation of the product (their  
183 Table 1b,c), and using the log-transformed abundance multiplied by the standardized  
184 trait values (their Table 1d). However, Monaghan & Soares (2014) give no  
185 information about the purpose of these calculations.

186

187 Basically, this table is unnecessary because the utility and the effects of log  
188 transformation of community data is only apparent when several species appear in the  
189 data simultaneously. For a single species, we cannot see why these operations are  
190 important. Multiplying a standardized trait value by the raw abundance of that single  
191 species: (a) may be correct but not useful, because the standardized trait values of a  
192 species provide sufficient information about the (feeding habits of the) taxon anyway.  
193 Operations (b-c) are illogical (but are in fact identical) because the raw abundances  
194 are multiplied first by the relative frequency of the trait while log transformation is  
195 undertaken only subsequently. As mentioned previously, handling the data this way is  
196 not correct, because transformation should be done first in order to ensure consistency  
197 in subsequent analyses of the data.

198

### 199 **Combining traits with the abundance of several taxa in a community**

200

201 In examining communities, Monaghan & Soares (2014) suggest that "the relative  
202 abundance of trait classes at assemblage level is given by the sum of the trait  
203 frequencies of respective taxa. This can be calculated by multiplying the respective  
204 taxon abundances by their corresponding proportionate frequencies, followed by the  
205 summation of the resultant trait abundances" and exemplify this and some log-  
206 transformed versions in their Table 2. They write in their Summary that, compared to  
207 other versions, the summation of the product of log-transformed abundance and  
208 standardized trait values "represents a non-linear abundance weighting of trait  
209 frequencies, as opposed to an expression of trait abundance *per se*, because the  
210 addition of logarithmic data is equivalent to multiplication on an arithmetic scale,  
211 summing this abundance-weighted frequencies provides an inconsistent scaling of  
212 trait abundance that might confound quantitative comparison" .

213

214 This also risks confusion because Monaghan & Soares (2014) use the term "*relative*  
215 abundance of trait classes at assemblage level" (italics ours) to specify the trait value  
216 of the community (*community traits* in Heino *et al.*, 2013). It is not clear why is this  
217 “relative”, which generally means a comparison to some standard (or total). Another  
218 source of confusion, compared to the methodology of Charvet *et al.* (2000) and  
219 Statzner *et al.* (2000), is that the "relative abundance of trait classes at the assemblage  
220 level" is the sum of products ("This can only be calculated by multiplying the  
221 respective taxon abundances by their corresponding proportionate frequencies,  
222 followed by the summation of the resultant trait abundances") without dividing by the  
223 total abundance of invertebrates. In the following, we show how to combine  
224 standardized trait values of several taxa in a community to calculate community-level  
225 trait values, where weighting includes a division.

226

227 Let us start with the taxa-by-traits and taxa-by-sites matrices provided in Table 2 of  
228 Monaghan & Soares (2014) and, for the time being, focus only on the  
229 presence/absence of the taxa and on the trait "fine sediments". The data suggest that  
230 *Psychomyia* has a standardized trait value of 0.17 and *Agapetus* a value of 0.25. We  
231 should not forget, however, that the other five taxa (*Sericostoma*, *Polycentropus*,  
232 *Odontocerum*, *Drusus* and *Hydropsyche*) have a standardized trait value of 0 (no  
233 affinity for feeding on fine sediment). It follows that our community contains five  
234 taxa with a standardized trait value of 0, one taxon (*Psychomyia*) with a standardized  
235 trait value of 0.17 and one (*Agapetus*) with a value of 0.25. We can illustrate this on a  
236 frequency histogram where the horizontal axis shows the standardized trait values  
237 while the vertical axis corresponds to their frequencies (top-left subplot in Fig. 1). The  
238 mean of these values gives the community trait value (indicated by the arrow). How  
239 can we interpret this value? It tells us that if we consider only the presence of the taxa  
240 then the community is represented by an average value of 0.06 ( $= [0.17 + 0.25 + 5 \times$   
241  $0]/7$ ) with respect to the trait ‘feeding on fine sediment’. In other words, members of  
242 the community, based on the presence of taxa, have a low affinity for feeding on fine  
243 sediment. It follows that weighting by the presence of taxa means that we give more  
244 importance to those elements of the set (to those standardized trait values of the  
245 observed trait values), which are most frequently represented by the taxa in the  
246 community. Examining the issue from a taxonomic point of view, we give equal  
247 importance to each taxon. We can do the same procedure for the other traits (top row



248 in Fig. 1). We should note that it is really beneficial that our calculation is based on  
249 standardized trait values, because the sum of community trait values equals 1. It  
250 follows that we can clearly interpret which function is the most characteristic of the  
251 community (in this case, feeding as a scraper).

252

253 The logic of using abundances is similar to that used for presence: we count the  
254 frequencies of standardized trait values and use these frequencies to calculate  
255 community trait values. However, compared to using the *presence of the taxa*, here  
256 we use the numerical abundance of *individuals* to weight the standardized trait values.  
257 In other words, we have 240 individuals (50 *Sericostoma*, 100 *Polycentropus*, 20  
258 *Odontocerum*, 40 *Drusus* and 30 *Hydropsyche*) showing the standardized trait value 0  
259 for fine sediment trait, we have 80 individuals (80 *Psychomyia*) showing the  
260 standardized trait value 0.17 and, finally we have 20 individuals (20 *Agapetus*) with a  
261 standardized trait value 0.25 (bottom-left subplot in Fig. 1). The community trait  
262 value is the weighted average of these values, and the measure is usually known as  
263 community-weighted mean trait value (Ricotta & Moretti, 2011). In the example,  
264  $(80 \times 0.17 + 20 \times 0.25 + 240 \times 0) / 340 = 0.055$ . For the other four traits we obtain 0.245,  
265 0.269, 0.202 and 0.228. We should note that the sum of these community trait values  
266 is 1 and that the community trait value corresponds to the mean standardized trait  
267 value of an *individual* in the community. The latter calculations (without the final  
268 division) are missing from Table 2 of Monaghan & Soares (2014), although the  
269 weighted values are presented in row a (for example,  $80 \times 0.17 + 20 \times 0.25 = 18.6$ ).

270

271

## 272 **Log-transformation of abundance in trait-based community analyses**

273

274 What happens if we use the log-transformed abundance for weighting? First, we  
275 transform the abundance of each taxon. While the original column vector contains  
276 abundances of 80, 50, 100, 20, 40, 30 and 20, the log-transformed column vector  
277 contains abundances of  $\ln(80) = 4.38$ ,  $\ln(50) = 3.91$ ,  $\ln(100) = 4.61$ ,  $\ln(20) = 2.99$ ,  
278  $\ln(40) = 3.69$ ,  $\ln(30) = 3.40$  and  $\ln(20) = 2.99$ . Note that, in order to make the example  
279 comparable with Monaghan & Soares (2014), we use also  $\log_e$  transformation  
280 (abbreviated to  $\ln$ ), although we suggest that logarithms to the base of 10 are easier to  
281 interpret (see below.) We then use these log-transformed abundances, instead of the

282 original abundances. To calculate the frequency of standardized trait value of 0 for  
283 fine sediment, we should sum the frequency of taxa with 0 standardized trait value for  
284 fine sediment:  $\ln(50) + \ln(100) + \ln(20) + \ln(40) + \ln(30) = 18.6$ . The transformed  
285 abundance of standardized trait value of 0.17 for fine sediment is  $\ln(80) = 4.38$ .  
286 Finally, the transformed abundance of the standardized trait value 0.25 equals  $\ln(20)$   
287  $= 2.99$ . The community trait value can be calculated by the mean of these values  
288  $[(18.6 \times 0 + 4.38 \times 0.17 + 2.99 \times 0.25) / (18.6 + 4.38 + 2.99) = 0.057$ . We can  
289 calculate frequencies and community trait values for the other traits in similar manner.  
290 We should note that the sum of these community trait values equals to 1 (Fig. 1), as  
291 required (see also Usseglio-Polatera *et al.*, 2000; Dolédec *et al.*, 2006).

292

293 There are, however, some problems with such analyses. First, this transformation is  
294 not able to handle 0 abundances (in comparing multiple communities, some species  
295 may be absent from several sites) because  $\ln(0)$  is mathematically not interpretable.  
296 Therefore, we suggest the use of the  $\log(x+1)$  transformation, as used by Statzner *et*  
297 *al.* (2001), Gayraud *et al.* (2003), Dolédec *et al.* (2006) and also suggested by  
298 Monaghan & Soares (2014). Moreover, although the base of logarithms is again an  
299 arbitrary decision, we suggest a  $\log_{10}(x+1)$  transformation to facilitate interpretation.  
300 To demonstrate this, we start with the same taxon-by-trait matrix discussed before,  
301 but our community contains only three taxa: 1000 individuals of *Psychomyia*, 100  
302 individuals of *Sericostoma* and 10 individuals of *Polycentropus* (other four taxa are  
303 represented by 0 individuals). Here, we do not intend to discuss all details of the  
304 analyses, but only emphasize some important differences among different weighting  
305 schemes. Regarding the trait ‘scraper’, we have one taxon (*Polycentropus*) showing a  
306 standardized trait value of 0, one showing a value of 0.25 (*Sericostoma*), and one a  
307 value of 0.5 (*Psychomyia*). Consequently, the community trait value in presence-  
308 weighted analyses equals 0.25 (Fig. 2). In abundance-weighted analyses, however,  
309 there are 10 individuals with a standardized trait value of 0 (10 individuals of  
310 *Polycentropus*), 100 individuals with a standardized trait value of 0.25 (100  
311 individuals of *Sericostoma*) and 1000 individuals with a standardized trait value of 0.5  
312 (1000 individuals of *Psychomyia*). The community trait value equals to 0.47, very  
313 close to the value of *Psychomyia*, the dominant species in the assemblage. In log-  
314 transformed abundance-weighted analyses [using a  $\log_{10}(x+1)$  transformation], the  
315 standardized trait values 0, 0.25 and 0.5 will be multiplied by  $\log_{10}(11) = 1.04$ ,

316  $\log_{10}(101) = 2.00$  and  $\log_{10}(1001) = 3.00$ , respectively (Fig. 2). We should recognize  
317 that the  $\log_{10}$ -transformed abundance expresses the magnitude of the number of  
318 individuals in the decimal system. The community trait value will be lower than  
319 above, 0.33, reflecting our decision to give less weight to the most common species. It  
320 is interesting to see how the community trait value of the scraper trait increases as we  
321 change from presence-weighting, through log-transformed abundance weighting to  
322 abundance weighting (Fig. 2). All of these suggest that, although Monaghan & Soares  
323 (2014) are right in stating that the addition of log-transformed data is equivalent to  
324 multiplication on the arithmetic scale, the interpretation of logarithmic transformation  
325 is straightforward. Therefore, in our view, papers using log-transformed abundances  
326 in trait-based analyses provide not only a useful indicator of environmental conditions  
327 (as acknowledged by Monaghan & Soares, 2014), but also a clear and easily-  
328 interpretable mathematical procedure.

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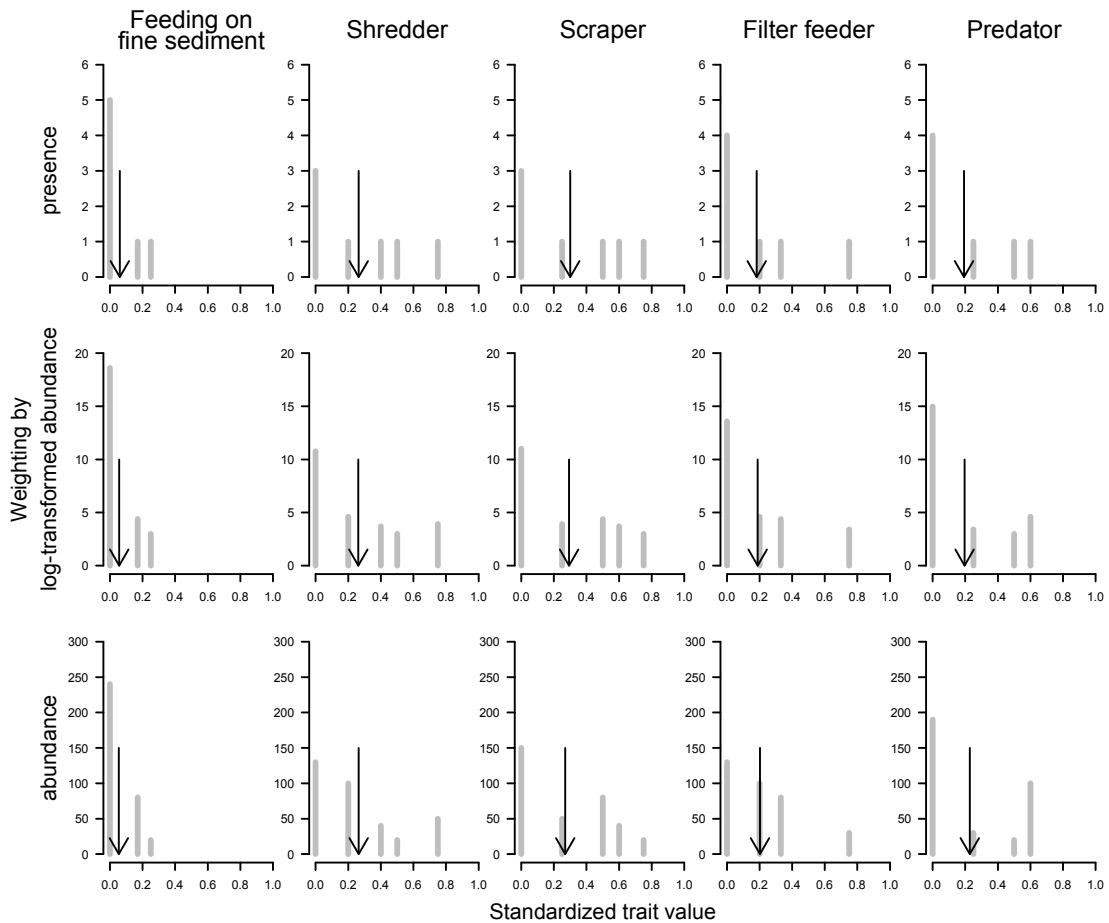
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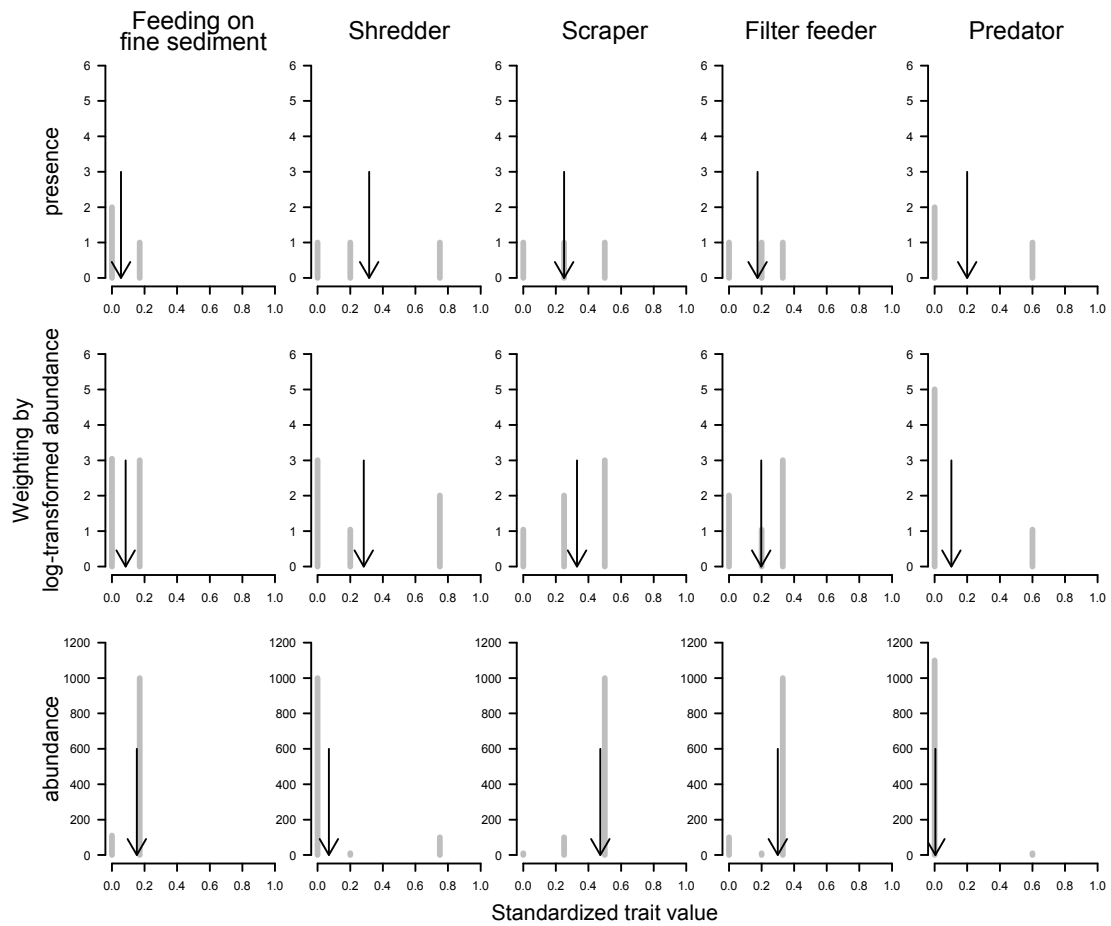
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386



388

389 Fig. 1: The frequency distribution of standardized trait values weighted by taxon  
 390 presence (top row), log-transformed abundance (middle row) and raw abundance  
 391 (lower row) for five different traits. Grey columns represent frequencies of  
 392 standardized trait values of the taxa, while arrows show the community-level trait  
 393 values. The community contains 80 individuals of *Psychomyia*, 50 individuals of  
 394 *Sericostoma*, 100 individuals of *Polycentropus*, 20 individuals of *Odontocerum*, 40  
 395 individuals of *Drusus*, 30 individuals of *Hydropsyche* and 20 individuals of *Agapetus*  
 396 as displayed in Table 2 of Monaghan & Soares (2014).

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Fig. 2: The frequency distribution of standardized trait values weighted by taxon

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presence (top row),  $\log_{10}$ -transformed abundance (middle row) and raw abundance

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(lower row) for five different traits. Grey columns represent frequencies of

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standardized trait values of the taxa, while arrows show the community-level trait

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values. The community contains 1000 individuals of *Psychomyia*, 100 individuals of

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*Sericostoma* and 10 individuals of *Polycentropus*.

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