

1 **Clumpy coexistence in phytoplankton: The role of functional similarity in community**
2 **assembly**

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18

19 **Abstract**

20 Emergent neutrality (EN) suggests that species must be sufficiently similar or sufficiently
21 different in their niches to avoid interspecific competition. Such a scenario results in a
22 multimodal distribution of species abundance along the niche axis (e.g., body size), namely
23 clumps. From this perspective, species within clumps should behave in a quasi-neutral state,
24 and their abundance will show stochastic fluctuations. Plankton is an excellent model system
25 for developing and testing ecological theories, especially those related to size structure and
26 species coexistence. We tested EN predictions using the phytoplankton community along the
27 course of a tropical river considering (i) body size structure, (ii) functional clustering of species
28 in terms of morphology-based functional groups (MBFG), and (iii) the functional similarity
29 among species with respect to their functional traits. Considering body size as the main niche
30 axis, two main clumps (clump I and II) were detected in different stretches of the river and
31 remained conspicuous through time. The clump I comprised medium-sized species mainly
32 from the MBFG IV, while the clump II included large-bodied species from the groups V and
33 VI. Pairwise differences in species biovolume correlated with species functional redundancy
34 when the whole species pool was considered, but not among species within the same clump.
35 Within-clump functional distinctiveness was positively correlated with species biovolume
36 considering both seasons, and also at the upper course. These results suggest that species within
37 clumps behave in a quasi-neutral state, but even minor shifts in trait composition may affect
38 their biovolume. In sum, our findings point that EN belongs to the plausible mechanisms
39 explaining community assembly in river ecosystems.

40 *Keywords:* species coexistence; emergent neutrality; functional distinctiveness; functional
41 redundancy

42

43 Introduction

44 Understanding the mechanisms promoting species coexistence and shaping community
45 structure has been a long-standing goal in community ecology. The former idea that the
46 number of coexisting species is limited by the number of growth-limiting resources or niche
47 dimensions (Gause 1936, Hardin 1960), and its derivative idea, “*the paradox of the plankton*”
48 (Hutchinson 1957), have been widely explained in terms of endogenous and exogenous
49 spatio-temporal mechanisms (Roy and Chattopadhyay 2007). Trait-based approaches are
50 useful to test this matter due to their potential to generalize patterns beyond species’ identity,
51 especially because traits influence the species’ ability to acquire resources and persist through
52 environmental changes (McGill et al. 2006, Díaz et al. 2013, 2016). Nonetheless, niche-based
53 theory proposes that the environment filters community composition through species’
54 ecological requirements, which can be perceived through species’ traits. However, at small
55 spatial scales, species interactions present a predominant role, as the environmental
56 heterogeneity is expected to be small (Götzenberger et al. 2012). In contrast, the more recent
57 neutral theory suggests that diversity results from random dispersal, speciation, and
58 extinction rates with no role of niche differences in species coexistence (Hubbell 2001). This
59 type of dynamics should then result in random distribution of functional traits along
60 environmental gradients (Kraft et al. 2008, Cornwell and Ackerly 2009).

61 More recently, it was shown that community organization is driven by eco-evolutionary
62 processes such as speciation and nutrient uptake kinetics resulting in groups comprising
63 different species with similar ecological requirements (Gravel et al. 2006, Hubbell 2006,
64 Scheffer and van Nes 2006). This finding led to the ‘emergent neutrality hypothesis’ (EN)
65 (Holt 2006) that has been supported by observational studies, e.g., for phytoplankton from
66 brackish waters (Segura et al. 2011), birds from the North of Mexico (Thibault et al. 2011),

67 and beetles at the global scale (Scheffer et al. 2015). EN suggests that species must be
68 sufficiently similar, and thus, behave neutrally, or different enough in their niches to avoid
69 competition. Such a scenario would result in species-rich aggregations or clumps along the
70 niche axis (Scheffer and van Nes 2006, Vergnon et al. 2009, Fort et al. 2010). Modelling
71 studies have shown that such predictions apply for both steady environmental conditions
72 (Fort et al. 2010), and also fluctuating resource conditions (Sakavara et al. 2018). Empirical
73 evidence about EN is still scarce, however (Scheffer et al. 2018).

74 The clumpy pattern arises from the species niche preferences, meaning that species within
75 clumps occupy same niche and their abundances are distributed stochastically (Scheffer et al.
76 2018). However, within a clump, trait differences may be important to species performance in
77 the niche space (McGill et al. 2006, Violle et al. 2007), and it is difficult to state whether
78 species behave neutrally within clumps (i.e., when the strength of interspecific interactions
79 equals the intraspecific interactions). Zooming in on the uniqueness of trait combinations of
80 species, i.e. functional distinctiveness, within clumps may advance our comprehension of
81 biotic interactions and move towards a measurable value of similarity at which species
82 coexistence is driven stochastically. Functional distinctiveness reflects the non-shared
83 functions among species within a given species pool (Violle et al. 2007), mirroring the
84 concept of functional redundancy (Pavoine et al. 2017). However, although two functionally
85 redundant species most likely show high similarity in trait combination, functional
86 distinctiveness is not directly linked to redundancy (Coux et al. 2016, Ricotta et al. 2016,
87 Violle et al. 2017). For example, two species could show similar functional distinctiveness,
88 i.e. the degree at which a species differs from all the others within the species pool
89 concerning their functional traits, and still not be similar in their trait composition at a
90 pairwise level (Coux et al. 2016). This suggests that both functional redundancy and
91 distinctiveness are good metrics to assess the role of trait combination in community

92 assembly. To this end, planktonic communities offer an excellent highly interesting model
93 for biodiversity theory testing due to their highly speciose communities and their known
94 relationship between morphological traits and function (Litchman and Klausmeier 2008,
95 Kruk and Segura 2012, Litchman et al. 2012).

96 Body size is considered as a master ecological trait and it is highly useful to characterize
97 species niche (Downing et al. 2014). In phytoplankton, the body size is related to physiology
98 and life-history (Litchman and Klausmeier 2008), photosynthetic processes (Marañón 2008),
99 nutrient uptake kinetics (Litchman et al. 2010) and other eco-evolutionary processes (Sauterey
100 et al. 2017). Despite the importance of body size, the use of a single trait as a proxy for the
101 niche may not evidence species differences generated by hidden niches and impair the
102 understanding of clumpy patterns (D'Andrea et al. 2018). The use of multiple traits emerges
103 as a powerful tool to disentangle plankton functional structure and evaluate competing
104 hypotheses (Reynolds et al. 2014, Chen et al. 2015, Bortolini and Bueno 2017). Morphology-
105 based functional groups (MBFG) classification of phytoplankton species (Kruk et al. 2010)
106 uses a is a multidimensional combination of morphological traits that cluster organisms into
107 seven groups with similar physiology and ecological responses (Kruk et al. 2011), potentially
108 overcoming the limitations of using a single trait dimension only. Assessing the uniqueness in
109 trait combinations of species within the same functional cluster (e.g., clumps, MBFGs) could
110 help to study the existence of functional equivalence (i.e., neutrality) among species. Overall,
111 the functional distance among species is a useful tool to compare species in a
112 multidimensional space by comparing species' functional differences, particularly because the
113 environment may filter different functional traits across space and time (Mouillot et al. 2013,
114 D'Andrea et al. 2020).

115 Riverine ecosystems are highly heterogeneous **andsystems** characterized by a continuous
116 water flow that affects the morphology, sedimentation patterns, dispersal, and phytoplankton
117 diversity (Reynolds **and** Descy 1996, Wetzel 2001). Several theories explain the **longitudinal**
118 **occurrence and abundance** of river phytoplankton communities, such as River Continuum
119 (Vannote et al. 1980) and Flood Pulse (Junk et al. 1989) concepts. However, an explicit study
120 of community size structure and phytoplankton species coexistence under EN in riverine
121 ecosystems is lacking. **For example, phytoplankton species should attain higher biomass at**
122 **the middle reaches or in the upper reaches of low-gradient stretches, and the continuous water**
123 **flow reduces the likelihood of biotic interactions (Reynolds et al. 1994). However, the**
124 **physiological and morphological characteristics of species within the local species pool may**
125 **result from eco-evolutionary processes that cluster species into limited functional groups that**
126 **are favored under specific environmental conditions (Scheffer et al. 2015). Here, we push**
127 **forward three hypotheses to be tested** in a tropical river by investigating phytoplankton
128 community size structure both seasonally and spatially:

129 **H₁ – There are peak aggregations of species abundance (i.e., clumps) along the body size**
130 **axis of phytoplankton in the river that remain constant across space and time as a result of**
131 **eco-evolutionary processes.**

132 **H₂ – Differences in species biovolume scales positively with functional redundancy at the**
133 **community-level but not at the clump level, because species within the same clump behave in**
134 **a quasi-neutral state. Hence, the dominance within clump varies stochastically between**
135 **redundant species because fitness differences would be negligible within clumps.**

136 **H₃ – There is a positive relationship between species abundance and species functional**
137 **distinctiveness within clumps. Though abundance fluctuates stochastically at pairwise level,**
138 **the number of species holding similar trait combinations may affect the likelihood of the**

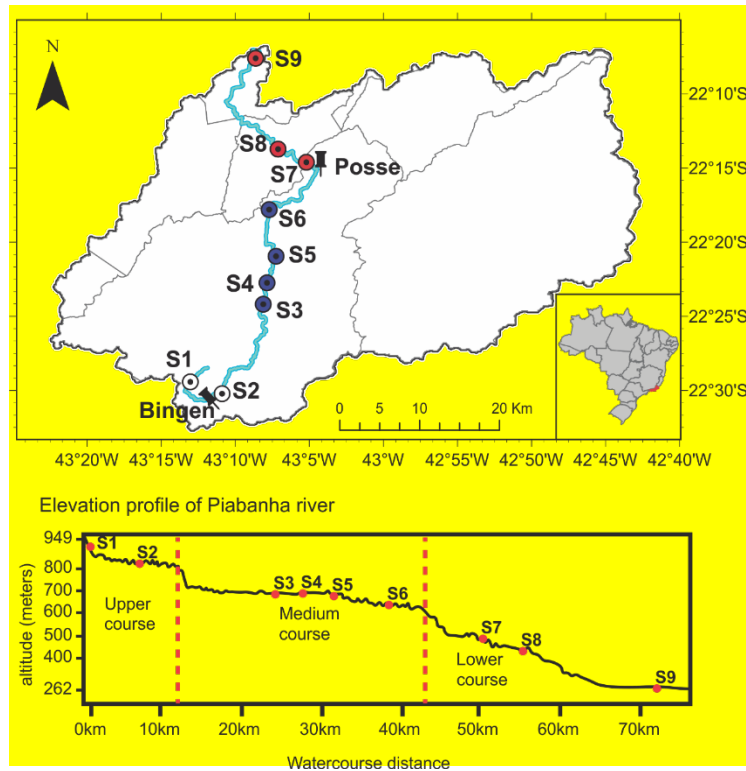
139 interactions within clumps. Therefore, species with the most distinct trait combinations with
140 respect to their clump peers are less likely to share same ecological requirements, and by
141 consequence, attain higher abundance.

142 **Methods**

143 *Study area*

144 Samples were taken monthly at nine stations along the Piabanha river between May 2012 and
145 April 2013. Piabanha river is in the tropical region of Brazil and has a drainage basin of
146 approximately 4500 km² (Figure 1). The headwater is on Petrópolis at 1546m altitude and
147 drains to the medium valley of Paraíba do Sul river crossing three cities and with agricultural
148 activities in their watershed. We set three river stretches (lower, medium and upper courses)
149 based on the location of steep slopes the river elevation profile (Figure 1). Data from two
150 meteorological stations, located close to the sampling sites, were used to measure rainfall. We
151 analyzed meteorological data up to three days before each sampling campaign. We classified
152 seasons as dry season (May - October) and a wet season (November – April) based on the
153 rainfall data.

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155

156 **Figure 1. Map of the study area.** The watershed area of the Piabanha river showing the
 157 river course (blue line) and the sampling sites colored according to river stretches (white
 158 circles = upper course, blue circles = medium course, red circles = lower course). The vertical
 159 dotted red line in the elevational profile figure indicates the locations of steep slopes used to
 160 define the boundaries of the river stretches.

161 *Sampling and sample analysis*

162 In the field, we measured temperature ($^{\circ}\text{C}$), dissolved oxygen (DO , mg L^{-1}) and turbidity by
 163 a multiparameter probe sonde (YSI model 600 QS). Water discharge (WD , $\text{m}^3 \text{s}^{-1}$) was
 164 measured with the SonTek RiverSurveyor – M9. Furthermore, water samples were taken and
 165 kept frozen (one or 2 weeks) until the laboratory analysis for ammonium ($\text{N}\cdot\text{NH}_4^+$, mg L^{-1}),
 166 nitrate ($\text{N}\cdot\text{NO}_3^-$, mgL^{-1}), total phosphorus (TP , mg L^{-1}) and soluble reactive phosphorus
 167 (SRP , mg L^{-1}). The water samples were filtered (except for total phosphorus analysis) using
 168 borosilicate filters (Whatman GF/C), and nutrient concentrations were measured following

169 (APHA 2005). A complete description of the spatial and temporal patterns of the
170 environmental variables measured in the Piabanha river can be found elsewhere (Graco-Roza
171 et al. 2020).

172 *Phytoplankton samples*

173 Subsurface samples of phytoplankton were collected with a bottle of 200 mL and fixed with
174 Lugol. In the laboratory, phytoplankton species were identified, and population densities
175 estimated under an inverted microscope (Olympus CKX41) (Utermöhl 1958). At least 100
176 individuals of the dominant species were counted in each sample (error < 20%, $p < 0.05$,
177 (Lund et al. 1958, Uhelinger 1964). Biovolumes ($\text{mm}^3 \text{L}^{-1}$) were estimated by multiplying
178 the density of each population by the average volume (V , μm^3), estimated from 20 random
179 chosen individuals (whenever was possible), using the appropriate geometrical forms
180 (Hillebrand et al. 1999, Sun and Liu 2003). In addition to the volume, we measured species'
181 surface area (S , μm^2), the maximum linear dimension (MDL, μm), and the presence of
182 aerotopes, mucilage, flagella, and siliceous exoskeleton. Hence, we used the volume and
183 surface area of the species to estimate the individual surface volume ratio (SV). Species were
184 then classified into Morphological Based Functional Groups (MBFG) according to (Kruk et
185 al. 2010).

186 *Traits-environment relationship*

187 We tested the relationship between functional traits and the environmental variables using an
188 RLQ analysis (Dolédec et al. 1996). The RLQ allows the visualization of the distribution of
189 species traits and their related ecological preferences by testing the co-inertia between the
190 matrices of environmental variables (R), species abundances (L), and species traits (Q). A
191 Monte Carlo permutation test with 999 permutations at $\alpha = 0.05$ was used to test the statistical

192 significance of the RLQ axes. Furthermore, a fourth-corner method was applied to test the
 193 statistical significance of all pairwise associations between species functional traits and
 194 environmental variables. We quantified the strength of the associations using Pearson
 195 correlation coefficient r_p , and the global statistics F-value (Legendre et al. 1997). The p-
 196 values were corrected with 49999 permutations of the sites and the species to correct type I
 197 errors, using false discovery rate method for adjusting p-values to control overall error rate
 198 (Dray et al. 2014). However, because RLQ does not provide a significance test to identify
 199 which combination of environmental variables acts on which combination of functional traits,
 200 and the fourth-corner analysis tests the functional traits one at a time without considering the
 201 covariation among traits and environmental variables, we used also a combined method of
 202 RLQ and fourth-corner analysis (Dray et al. 2014). The significance of the associations was
 203 tested by 99999 permutations of the sites and the species (Dray and Legendre 2008) to
 204 correct type I errors, using false discovery rate method for adjusting p values to control the
 205 overall error rate (Dray et al. 2014).

206 *Clumpy patterns*

207 To test for the existence of peak aggregations of species abundance along the body size axis
 208 of phytoplankton - H_1 , we analyzed the community structure in each season (dry and wet) and
 209 river stretches (upper, medium, and lower course). First, the individual volume of species was
 210 log-transformed (\log_2) and used as the main niche axis ($X = \log_2$ volume). Hence, we divided
 211 the niche axis into equally spaced segments (one segment per unit of \log_2 biovolume) and for
 212 each segment (j), we estimated the Shannon entropy (H) using the biovolume of the observed
 213 species (Fort et al. 2010, Segura et al. 2011). The entropy index was defined as:

$$214 \quad H_j = \sum_{i=1}^n p_i \log_2(p_i), \quad (1)$$

215 where p_i is the fraction of biovolume of species i in the community of n species. Finally, we

216 tested the significance of the entropy (H) by comparing the observed H against an expected
 217 uniform distribution under the null hypothesis of homogeneous H . For this, we created 1000
 218 communities by sampling the volume of species from a random uniform distribution bounded
 219 by observed individual volumes. Then, each species had a biovolume assigned to it, which
 220 was taken from a randomization of the observed biovolume matrix, keeping both the
 221 empirical species rank-abundance pattern and total biovolume in the sample. For each
 222 segment, the observed H was compared with the distributions of H generated under the null
 223 hypothesis, with significance defined according to standard 5% criterion (Fort et al. 2010;
 224 Segura et al. 2011).

225 *Functional redundancy*

226 To test whether differences in species biovolume scales positively with functional
 227 redundancy – H_2 , we first calculated the functional redundancy (F_{Red}) and the differences in
 228 biovolume between pairs of species using the whole community and using only the species
 229 from the clumps with significant entropy values. The functional redundancy was obtained by
 230 calculating Gower's general similarity coefficient [function *gowdis*, R package FD (Laliberté
 231 et al. 2014)] on the species functional traits. We used Gower's dissimilarity (Gd) because it
 232 can handle mixed variable types (continuous, binary, ordinal and categorical traits). Gd
 233 defines a distance value d_{jk} between two species as

$$234 \quad d_{jk} = \frac{1}{N} \sum_{i=1}^n \left| \frac{(x_{ij} - x_{ik})}{\max(x_i) - \min(x_i)} \right|, \quad (2)$$

235 where, N is the number of functional traits considered, x_{ij} the value of trait i for species j and
 236 x_{ik} the value of the trait i for species k . We thus tested H_2 by conducting Mantel tests
 237 [function *mantel*, R package *vegan* (Oksanen 2017)] with 1000 randomizations, whenever

238 possible, on F_{Red} and $\text{Diff}_{\text{BioV}}$ matrices using all species, and separately for the species of each
 239 significant clump.

240 *Functional distinctiveness* (F_{Dist})

241 To test whether there is a positive relationship between species relative abundance at clump-
 242 level and species functional distinctiveness – H_3 , we estimated the functional distinctiveness
 243 (F_{Dist}) as the distance of a species to the average trait position (centroid) in the
 244 multidimensional functional space for the set of species pertaining to each of the significant
 245 clumps using the equations proposed by Anderson (2006). First, we obtained species
 246 coordinates in the functional space by applying a Principal Coordinates Analysis (PCoA) in
 247 the species-by-traits data table using the Gower's dissimilarity (Gd). Hence, the distance from
 248 a species to its given centroid was calculated as

$$249 \quad z_{ij}^c = \sqrt{\Delta^2(u_{ij}^+, c_{i'j'}^+) - \Delta^2(u_{ij}^-, c_{i'j'}^-)}, \quad (3)$$

250 where Δ^2 is the squared Euclidean distance between u_{ij} , the principal coordinate for the j th
 251 point in the i th group, and c_i , the coordinate of the centroid for the i th group. The super-
 252 scripted '+' and '-' indicate the real and imaginary parts respectively (Equation 3, see
 253 Anderson 2006 for details). We did not weight the clump-centroid by species abundance and
 254 calculated F_{Dis} using all species pertaining to the significant clumps. Hence, to test H_3 , we
 255 modeled the relationship between species biovolume and F_{Dist} using linear models. We used
 256 biovolume as response variable, with F_{Dist} , Clump position, and seasonal or spatial categories
 257 as fixed factors. Because the aggregation of sites across space and time have a nested design
 258 (i.e. spatial aggregation comprises all months sampled, and temporal aggregations comprises
 259 all sites sampled), we modelled spatial and temporal categories separately.

260 Results

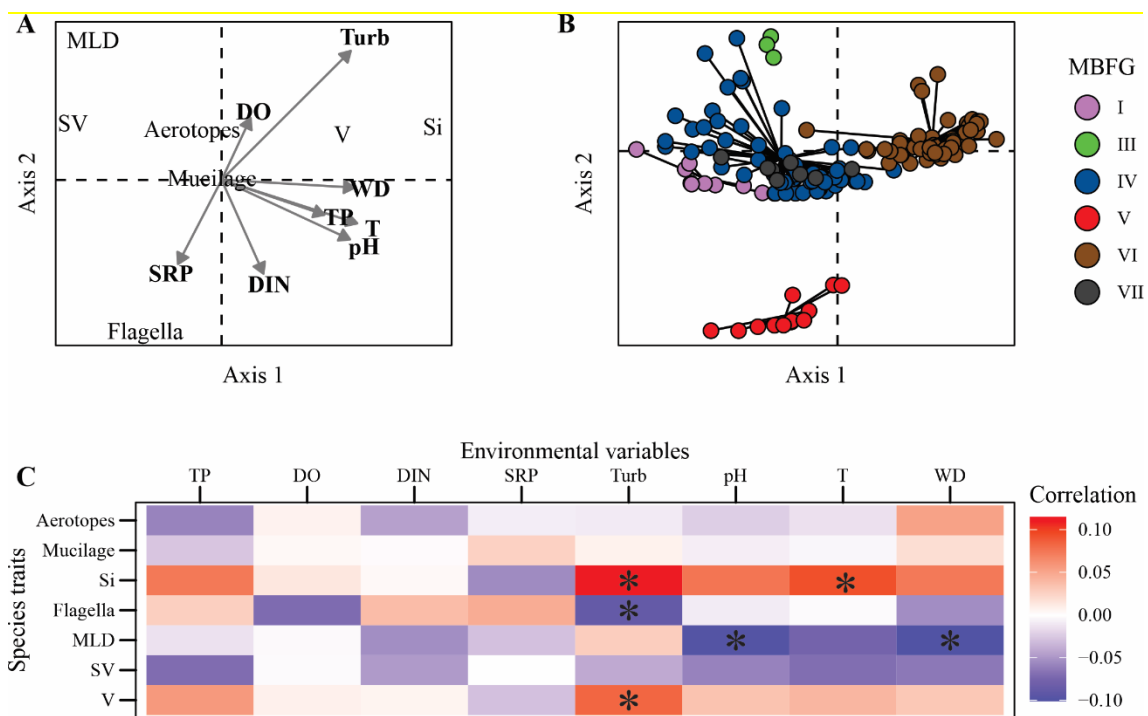
261 Our samples included 150 species that were classified in six (MBFG I, III, IV, V, VI, and
 262 VII) from the seven MBFGs based on their functional traits (Table 1). Species traits were
 263 significantly associated with environmental variables (RLQ analysis; Figure 3; $P < 0.05$).
 264 Individual volume (V), surface volume ratio (SV), maximum linear dimension (MLD), and
 265 siliceous exoskeleton (Si) correlated significantly with the first axis, where SV had the
 266 highest correlation ($r = -0.10$). Also, MLD and flagella correlated significantly with the
 267 second axis ($r = -0.07$, $r = -0.08$). For the environmental variables, water temperature (T) and
 268 water discharge (WD) correlated the strongest with the first axis ($r = 0.10$ for both variables),
 269 which also correlated significantly ($p < 0.05$) with total phosphorus (TP), dissolved oxygen
 270 (DO), turbidity (Turb), and pH. In addition, dissolved inorganic nitrogen (DIN), soluble
 271 reactive phosphorus (SRP), and turbidity (Turb) correlated significantly with the second axis,
 272 and Turb had the highest correlation ($r = 0.90$) (Figure 2A).

Table 1. Distribution of species among the morphological-based functional groups.

| MBFG | Number of species | Representative taxa |
|-------|-------------------|--|
| I | 9 | <i>Chroococcales</i> sp, <i>Chroococcus</i> sp. |
| III | 3 | <i>Limnothrix</i> sp. |
| IV | 60 | <i>Pseudoanabaena limnetica</i> , <i>Pseudoanabaena catenata</i> |
| V | 13 | <i>Euglena</i> sp. <i>Cryptomonas</i> sp. |
| VI | 57 | <i>Cymbella</i> sp., <i>Synedra</i> sp. |
| VII | 8 | <i>Dictyosphaerium</i> sp. |
| Total | 150 | |

273 The distribution of the species belonging to different MBFGs in the RLQ ordination (Figure
 274 2B) reflect their occurrence in relation to the environmental variables. The groups IV, V and
 275 VI included 87% of the total number of species. Species from the group IV included
 276 filamentous, colonial and unicellular species ranging from $21 \mu\text{m}^3$ to $8181 \mu\text{m}^3$ with no
 277 specialized morphological traits (e.g. flagella, Si). Group V comprised unicellular flagellated

278 species ranging in size from $31\mu\text{m}^3$ to $31864\mu\text{m}^3$, and group VI included unicellular and
 279 chain-forming species with siliceous exoskeletal body that ranged from $48\mu\text{m}^3$ to $19045\mu\text{m}^3$.
 280 According to the fourth-corner analysis, MLD was negatively correlated with pH ($r = -0.09$)
 281 and WD ($r = -0.10$), flagella was negatively correlated with Turb ($r = -0.08$), Si was
 282 positively correlated with Turb ($r = 0.11$) and WT ($r = 0.09$), whereas V was positively
 283 correlated with Turb ($r = 0.08$) (Figure 2C).



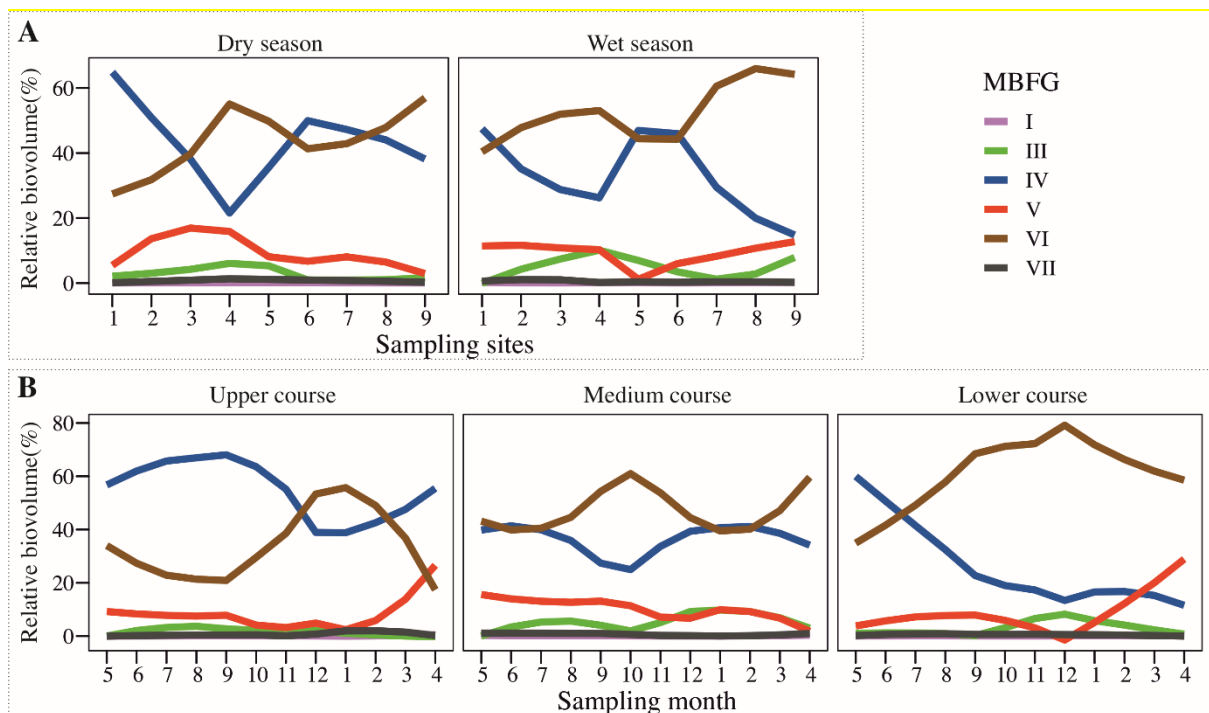
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285 **Figure 2. Results of the RLQ and fourth-corner analysis.** (A) The relationships between
 286 species traits and environmental variables. First and second axes summarized 64.9% and
 287 20.7% of variation, respectively. (B) The distribution of species in the functional space. Each
 288 point in the ordination plot within clusters represents the position of a species modelled
 289 according to its traits on RLQ axes 1 and 2. (C) The bivariate relationships between species
 290 traits and environmental variables. Functional traits are abbreviated as: Si = siliceous
 291 exoskeleton, MLD = maximum linear dimension, SV = surface volume ratio, V = individual
 292 volume. Environmental variables are abbreviated as: TP = total phosphorus, DO = dissolved

293 oxygen, DIN = dissolved inorganic nitrogen, SRP = soluble reactive phosphorus, Turb =
 294 turbidity, T = water temperature, WD = water Discharge.

295

296 The groups IV, V, and VI contributed to 95% of the mean biovolume of total phytoplankton
 297 in dry and wet seasons, and 96.6% across river stretches (Figure 3). The group VI had the
 298 highest contribution for the biovolume of almost all river stretches in both seasons, except for
 299 the upper course during dry season was dominated by the group IV (Figure 3). The entire list
 300 of species, functional traits, and assigned MBFG can be found in Table S1.



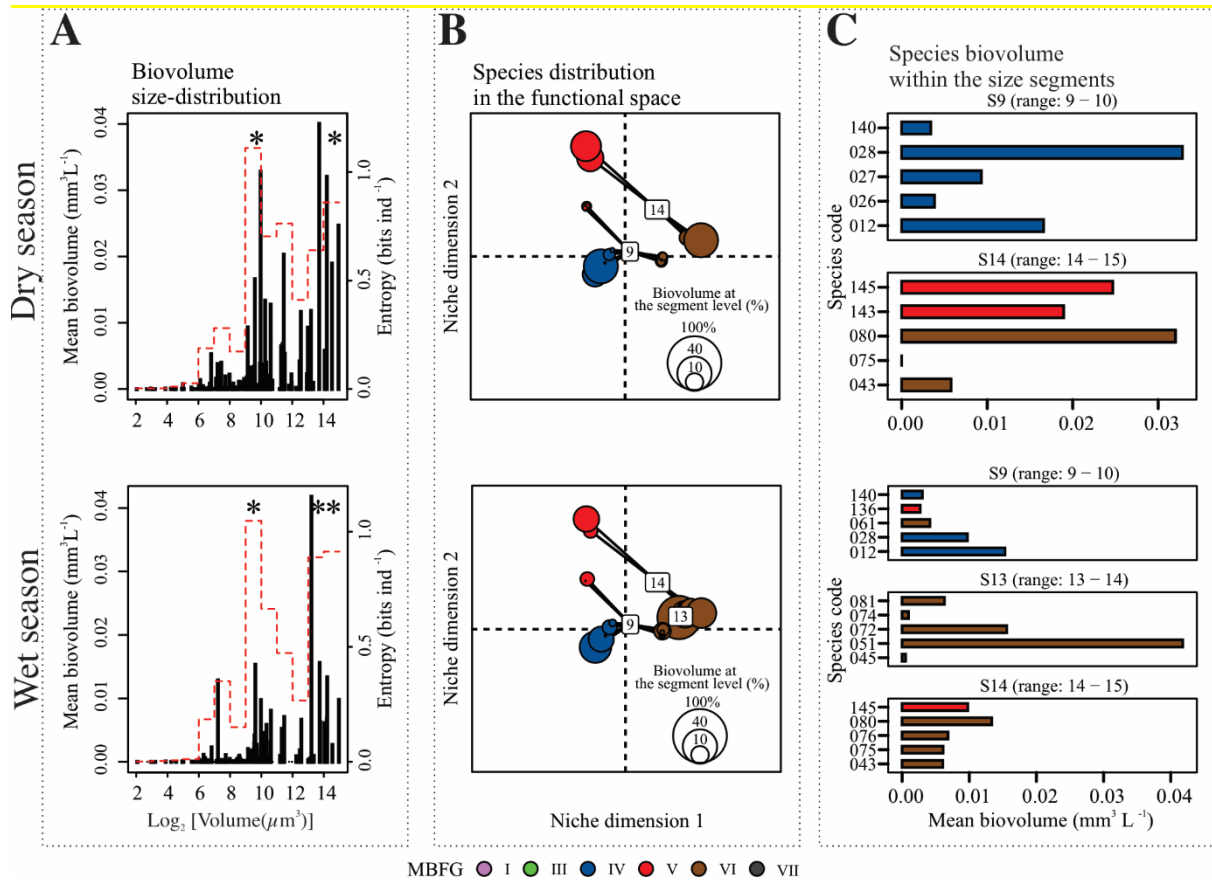
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Figure 3. Relative contribution of morphology-based functional groups (MBFG) to total biovolume observed along the seasons and river stretches of the Piabanha river between May-2012 and April-2013. Trend lines are loess curves fit of each MBFG.

302 Overall, species individual volume ranged from 4.19 μm^3 to 31864 μm^3 totaling 14 equally
 303 spaced segments (S) of volume along the niche axis. From the 14 segments, three of them

304 showed significant entropy values, specifically S9, S13, and S14. This resulted in a biovolume
305 aggregation (i.e. clumps) in two regions of the niche axis considering both temporal (Figure 4)
306 or spatial categories (Figure 5). The first clump (Clump I) comprised species mainly from the
307 group IV from S9 (e.g., *Pseudoanabaena limnetica*, *Pseudo anabaena catenata*), whereas the
308 second clump (Clump II) included species from groups V and VI (e.g., *Euglena* sp., *Cymbella*
309 sp.) at the segment S14. However, during the wet season, a few species from S13 also had
310 significant entropy values and therefore were included in the clump II (Fig 4B).

311 The mean biovolume of species within clumps differed across seasons, but the identity of the
312 most abundant species did not vary. *Pseudanabaena* sp. 4 (spp. 28) and *Pseudanabaena*
313 *catenata* (spp. 12) had the highest biovolumes of clump I and held the highest biovolumes of
314 the clump at both dry and wet seasons. Within the clump II, *Synedra* sp. (spp. 80) attained the
315 highest biovolume during the dry season, whereas *Cymbella* sp. (spp. 051) contributed the most
316 to clump biovolume in the wet season.



317

318 **Figure 4. Distribution of phytoplankton biovolume along the body size axis, the**

319 **ordination of species from the significant size segments (S) in the functional space, and**

320 **the mean biovolume of the five most abundant species of each significant size segment**

321 **during the dry and wet seasons of the Piabanha river, RJ. (A) Stem plots show size**

322 **distribution in the sampling sites of the river Piabanha. Each stem represents a species with its**

323 **body size (in log_2) plotted on the abscissa and the mean biovolume plotted on the ordinate. The**

324 **red dotted line indicates the entropy value of each size segment (i.e., one unit of log_2 volume),**

325 **and the asterisk highlights the significant entropy values tested through 1000 randomizations.**

326 **(B) The species of the corresponding significant size-segment are ordinated in the functional**

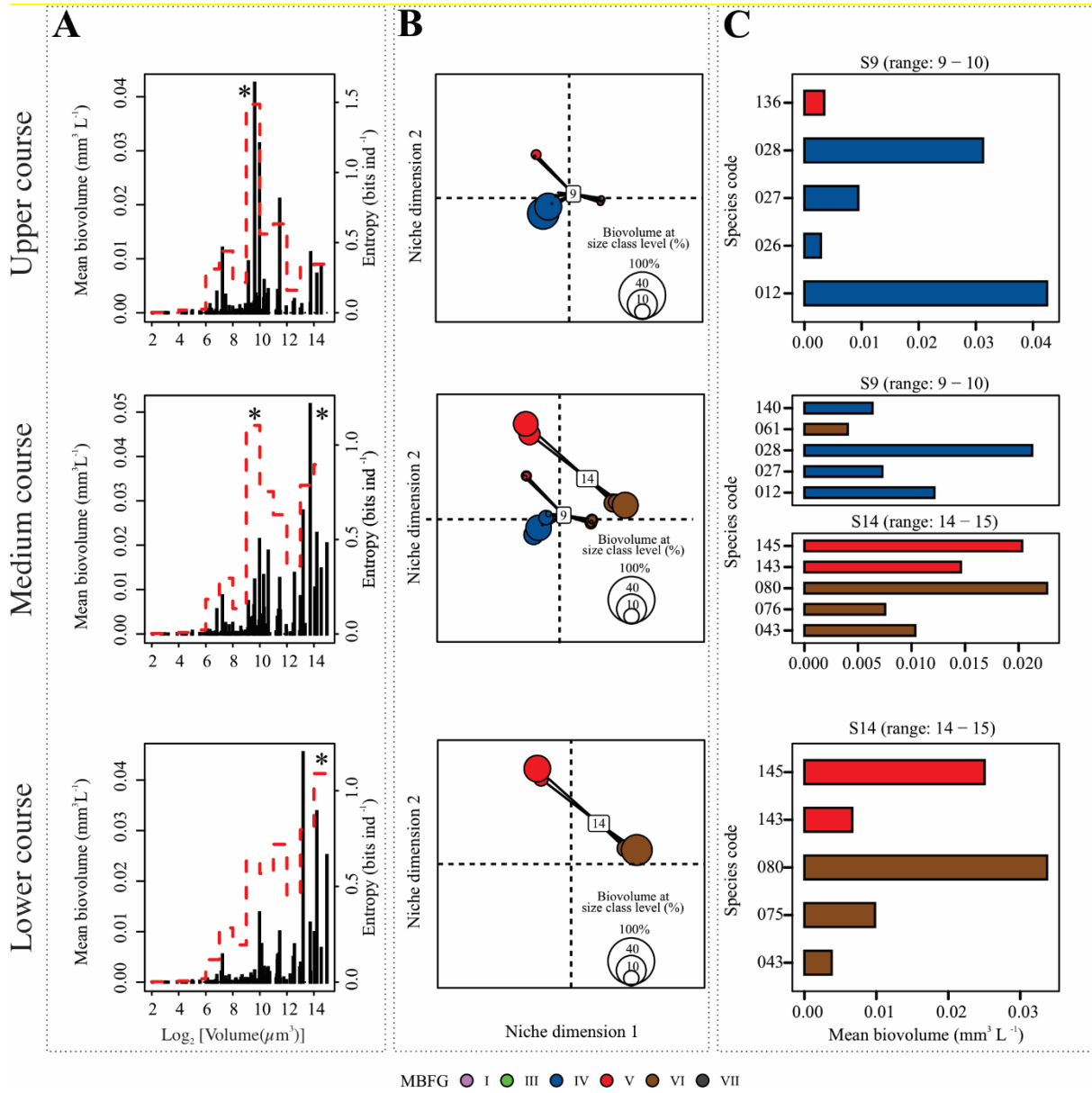
327 **space. The size of the circles represents the species contribution to the total biovolume of the**

328 **size segment. (C) Bar plots (right show the biovolume of the five most abundant species from**

329 **each significant size segment. Species are colored according to their morphology-based**

330 functional groups (MBFG). The code of each species can be found in the supplementary
331 material, Table S1.

332 Regarding the river stretches, *Pseudoanabaena* sp. 4 and *Pseudoanabaena catenata*
333 contributed the most to the biovolume of the upper course, and only the clump I had significant
334 entropy values for species from the S9 (Figure 5A). At the medium course, both the clump I
335 and II had significant entropy values, mainly due to species from the S9 with *Pseudoanabaena*
336 sp. 4 as the most abundant, and from the S14 with *Synedra* sp. as the most abundant (Figure
337 5B). At the lower course, the clump II had significant entropy values, specifically at the S14
338 where *Synedra* sp. contributed most to biovolume at segment level (Figure 5C).



340 **Figure 5. Distribution of phytoplankton biovolume along the body size axis, the**
 341 **ordination of species from the significant size segments (S) in the functional space, and**
 342 **the mean biovolume of the five most abundant species of each significant size segment at**
 343 **the upper, medium, and lower courses of the Piabanha river, RJ. (A) Stem plots show size**
 344 **distribution in the sampling sites of the river Piabanha. Each stem represents a species with its**
 345 **body size (in \log_2) plotted on the abscissa and the mean biovolume plotted on the ordinate. The**
 346 **red dotted line indicates the entropy value of each size segment (i.e., one unit of \log_2 volume),**
 347 **and the asterisk highlights the significant entropy values tested through 1000 randomizations.**

348 (B) The species of the corresponding significant size-segment are ordinated in the functional
 349 space. The size of the circles represents the species contribution to the total biovolume of the
 350 size segment. (C) Bar plots (right show the biovolume of the five most abundant species from
 351 each significant size segment. Species are colored according to their morphology-based
 352 functional groups (MBFG). The code of each species can be found in the supplementary
 353 material, Table S1.

354 Differences in biovolume among species were positively related to the functional redundancy
 355 at the whole-community level, but not at the clump-level irrespectively to the season or river
 356 stretch studied (Table 2). At the clump-level, we observed that species with higher
 357 distinctiveness attained the highest biovolume in all the seasons and river stretches (except for
 358 the lower course). The upscaling of biovolume with functional distinctiveness was observed
 359 for species from both clumps, at different seasons and river stretches, but was particularly
 360 different in the wet season, where species from clump II did not show any significant
 361 relationship with biovolume, and at the medium and lower courses, where there were no
 362 significant trends (Table 3).

363 **Table 2. Mantel correlation results.** Mantel correlation between the differences in species
 364 biovolume and functional redundancy for the overall set of species and significant clumps
 365 along the seasons (dry and wet) and river stretches (upper, medium and lower courses).

| Seasons and river stretches | Stratum | Mantel correlation |
|-----------------------------|----------|----------------------|
| <i>Dry season</i> | Overall | 0.180 (0.001) |
| | Clump I | 0.130 (0.120) |
| | Clump II | -0.172 (0.709) |
| <i>Wet Season</i> | Overall | 0.230 (0.001) |
| | Clump I | 0.065 (0.235) |
| | Clump II | -0.167 (0.705) |
| <i>Upper course</i> | | |

| | | |
|----------------------|----------|----------------------|
| | Overall | 0.165 (0.002) |
| | Clump I | 0.065 (0.242) |
| <i>Medium course</i> | Overall | 0.171 (0.001) |
| | Clump I | 0.059 (0.217) |
| | Clump II | -0.168 (0.633) |
| <i>Lower course</i> | Overall | 0.258 (0.001) |
| | Clump II | -0.167 (0.696) |

Note: Significant ($p < 0.05$) values are in bold and p-values are inside parentheses

Table 3. Linear model results. Regression parameters of the relationship between species biovolume and functional distinctiveness at the clump level.

| Dependent variable: Log_{10} Biovolume | | | | | |
|---|---------------------|---------------------|----------------------|---------------------|---------------------|
| | Biovolume | | | | |
| | Dry season | Wet season | Upper course | Medium course | Lower course |
| F_{Dist} | 14.925* (6.127) | 15.854* (6.015) | 24.440** (7.816) | 12.377 (6.085) | 7.267 (4.373) |
| Clump II | 4.676 (6.148) | 7.501* (3.175) | 9.619 (18.074) | 4.373 (5.169) | 2.067 (3.689) |
| $F_{\text{Dist}} \times$ Clump II | -8.458 (15.608) | -15.863 (7.734) | -21.301 (46.232) | -7.911 (13.329) | -1.651 (9.501) |
| Intercept | -8.979** (2.439) | -9.748** (2.496) | -12.951** (3.143) | -7.945** (2.418) | -6.165** (1.734) |
| Observations | 26 | 31 | 19 | 28 | 26 |
| R^2 | 0.420 | 0.437 | 0.457 | 0.374 | 0.579 |
| Adjusted R^2 | 0.341 | 0.374 | 0.348 | 0.296 | 0.522 |
| Residual | 0.756 | 0.663 | 0.823 | 0.755 | 0.523 |
| Std. Error | (df = 22) | (df = 27) | (df = 15) | (df = 24) | (df = 22) |
| F Statistic | 5.313** | 6.987** | 4.204* | 4.778** | 10.090** |

(df = 3; 22)

(df = 3; 27)

(df = 3;15)

(df = 3;24)

(df = 3; 22)

Note:

* p < 0.05 ** p < 0.01

366 Discussion

367 Present results showed that (i) the clumps in body size are a conspicuous feature of
 368 phytoplankton community structure in riverine systems across seasons and river stretches; (ii)
 369 species within clumps show random distribution of biovolume with respect to their pairwise
 370 functional differences, but functional redundancy is related to differences in biovolume at the
 371 whole-community level; and (iii) species biovolume generally scales positively with
 372 functional distinctiveness at the clump level. Note that, on the one hand, multimodal
 373 aggregation of species biovolume along body size axis only points to the integration of niche-
 374 based processes and neutrality driving community assembly (Vergnon et al. 2009), but it does
 375 not imply that EN is the only mechanism behind phytoplankton coexistence. On the other
 376 hand, our results do not support alternative hypothesis explanations such as pure neutrality
 377 (Hubbell 2001) or high dimensional hypothesis (Clark et al. 2007) because pure neutrality
 378 predicts a uniform distribution of species biovolume along the niche axis (Hubbell 2001), and
 379 the HDH does not predict any particular trait distribution (Vergnon et al. 2009, Ingram et al.
 380 2018).

381 One alternative theory that is likely to explain clumpy aggregations is the Holling's textural
 382 hypothesis (Holling 1992). The textural hypothesis explains multimodal species size
 383 distribution with environmental constraints, and although distribution of the clumps was
 384 spatially structured, our results do not support HDH as a whole. River stretches and seasons
 385 are radically different in river hydrology, nutrients and fluxes, but that was not reflected in
 386 phytoplankton size structure. Moreover, if landscape controls the body size pattern alone,
 387 differences in phylogenetics, biogeography, energeticsenergenetics, and community

388 interactions should not significantly affect such pattern (Allen et al. 2005). However, the
389 clumps in our data were in the similar size range and comprised species from the same
390 MBFG found in previous studies from different systems (Segura et al. 2011, 2013). More
391 importantly, there was a clear functional differentiation of species between clumps in terms
392 of MBFG (Figure 5-6), and the trait composition of the communities reflected adaptations to
393 survive in different environmental templates (Kruk and Segura 2012). For example, species
394 from the group IV dominated in low-flow waters with high nutrient input, while species from
395 group V and VI show different adaptations to survive in turbulent and mixed conditions
396 (Margalef 1978, Soares et al. 2007, Bortolini and Bueno 2017). Therefore, our findings
397 corroborate our H_1 on the existence of significant peak aggregations in species biovolume
398 along the body size axis. Such results agree with previous evidence of EN using
399 phytoplankton as the model community in coastal and estuarine environments (Segura et al.
400 2011, 2013), and shows that clumpy distributions are evident on the phytoplankton
401 communities of riverine ecosystems presenting fluctuations in environmental conditions. In
402 addition, our findings are in line with recent modeling results suggesting clumpy patterns
403 arise in environments subjected to resources fluctuation (Sakavara et al. 2018). It remains to
404 be tested the trade off between resources among competing species, which is the remaining
405 ingredient for the emergence of clumps.

406 We expanded the level of analysis from mere size structure towards the inclusion of new trait
407 dimensions by analyzing the functional similarity at MBFG classification level and functional
408 redundancy. We found that the species from the same clump represent a selection of all the
409 possibilities amongst MBGFs, reinforcing that body size is a good proxy for niche
410 preferences of the species (Blanckenhorn 2000). However, assessing MBFGs helped to
411 assign niche differences at a finer degree, taking into account that using only a single trait
412 could impair the patterns behind community assembly (D'Andrea et al. 2018). Species within

413 the same MBFG are expected to share ecological strategy (Kruk et al. 2010, Kruk and Segura
414 2012) and accordingly to EN, would also perform similarly (Scheffer et al. 2018). Still, the
415 importance of traits to equalize or stabilize niche differences are context-dependent and
416 assessing the functional multidimensional space should help to unveil community assembly
417 processes (Bonser 2006). Here, we found that size-related traits (volume, MLD) and the
418 presence of flagella or siliceous exoskeleton were the main drivers of phytoplankton
419 distribution. Not surprisingly, such traits are key determinants on the assignment of species to
420 the main MBFG observed here (Kruk et al. 2010). Furthermore, we found that fitness
421 differences were significant at the community level, but not at within-clumps level, even
422 considering multiple morphological traits. Still, the dominance of medium-sized species
423 (clump I) in the upper course and of large-bodied species (clump II) in the lower course
424 suggests that the coexistence of these MBFGs tends to generate clumpy distributions that are
425 eroded when the ecosystem properties favor the dominance of other MBFGs, but the
426 fluctuations in biovolume within clumps remains stochastic even when clumps hold a few
427 species belonging to different MBFGs. These findings corroborate our H2 - differences in
428 species biovolume scales positively with functional redundancy at the community level but
429 not at within-clumps level - and reinforces the predictions on the quasi-neutrality state of
430 species within clumps (Holt 2006).

431 We also outlined the role of functional similarity in community assembly by studying the
432 effects functional distinctiveness on species biovolume at clump level. We found that at the
433 clump I, species biovolume increased with functional distinctiveness, but this pattern was
434 weaker within clump II. Despite the species from group V and VI show lower growth rates
435 than the species from group IV (Kruk et al. 2010), their elongated shape provides advantage
436 under turbulent conditions (Reynolds et al. 1994) favoring these species under high flow
437 conditions on nutrient-rich environments (Irwin et al. 2006). Furthermore, nutrients are

438 hardly limiting phytoplankton growth especially in meso- or eutrophic rivers (Reynolds
439 2006), and large-bodied species are mostly randomly introduced from different habitats (e.g.
440 periphyton or epiphyton) (Wang et al. 2014, Descy et al. 2017), which brings some
441 uncertainty to the assumption of niche overlap due to size similarity. In fact, nutrients were
442 not relevant for structuring the phytoplankton community of the Piabanha river, and we also
443 observed a significantly different size-segment (S13) within the clump II during the high flow
444 periods. Thus, one cannot be sure if species within clump II are truly occupying same niche
445 or if they just were introduced to the plankton by random displacement. We suggest this may
446 be the mechanism leading to the weak relationship between functional distinctiveness and
447 biovolume within clump II.

448 Emergent neutrality results from eco-evolutionary processes that leads species selection
449 towards a limited number of functional groups (Scheffer and van Nes 2006). This implies that
450 the clumps observed here are not likely a result of competitive exclusion at the Piabanha
451 river, but a convergent evolution of competing species over time as previously suggested
452 (MacArthur and Levins 1967). Therefore, even when the competition rates are relaxed due to
453 sufficient nutrient supply, some other limiting factors that are not consumed by biotic
454 organisms such as heat energy or turbulence determine species biovolume, and even minor
455 shifts in species trait combination may lead to higher biovolume with respect to other species
456 occupying same niche. Our results showed that it is also possible to predict the biovolume of
457 species within clumps, but only when neutral processes (e.g., species displacement) are
458 relaxed and biotic interactions are more likely to occur. Therefore, our findings partially
459 agree with H₃ - there is a positive relationship between species abundance and species
460 functional distinctiveness within clumps, but the environmental conditions seem to play a key
461 role in the outcome.

462 In summary, we provided evidence of both neutral and niche perspective driving planktonic
463 community assembly and support the view that emergent neutrality is a likely mechanism to
464 explain species coexistence in an open and environmentally heterogeneous ecosystem. The
465 use of MBFG classification and functional space to describe species within clumps revealed
466 that under the same size range, species with a greater degree of redundancy alternate their
467 dominance in an unpredictable way. The significance and dominance of the clumps were
468 related to the environmental conditions, but the biovolume of species within the clumps was
469 better predicted by functional distinctiveness than by pairwise functional redundancy. This
470 addresses the difficulty to avoid the ghost of hidden niches (Barabás et al. 2013), and also
471 provides evidence from multiple angles that points to EN as a plausible mechanism in
472 shaping species coexistence.

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