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

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Abstract

Emergent neutrality (EN) suggests that species must be sufficiently similar or sufficiently

different in their niches to avoid interspecific competition. Such a scenario results in a

multimodal distribution of species abundance along the niche axis (*e.g.*, body size), namely

clumps. From this perspective, species within clumps should behave in a quasi-neutral state,

and their abundance will show stochastic fluctuations. Plankton is an excellent model system

for developing and testing ecological theories, especially those related to size structure and

species coexistence. We tested EN predictions using the phytoplankton community along the

course of a tropical river considering (i) body size structure, (ii) functional clustering of species

in terms of morphology-based functional groups (MBFG), and (iii) the functional similarity

among species with respect to their functional traits. Considering body size as the main niche

axis, two main clumps (clump I and II) were detected in different stretches of the river and

remained conspicuous through time. The clump I comprised medium-sized species mainly

from the MBFG IV, while the clump II included large-bodied species from the groups V and

VI. Pairwise differences in species biovolume correlated with species functional redundancy

when the whole species pool was considered, but not among species within the same clump.

Within-clump functional distinctiveness was positively correlated with species biovolume

considering both seasons, and also at the upper course. These results suggest that species within

clumps behave in a quasi-neutral state, but even minor shifts in trait composition may affect

their biovolume. In sum, our findings point that EN belongs to the plausible mechanisms

explaining community assembly in river ecosystems.

Keywords: species coexistence; emergent neutrality; functional distinctiveness; functional

redundancy

Introduction

Understanding the mechanisms promoting species coexistence and shaping community

structure has been a long-standing goal in community ecology. The former idea that the

number of coexisting species is limited by the number of growth-limiting resources or niche

dimensions (Gause 1936, Hardin 1960), and its derivate idea, *"the paradox of the plankton"*

(Hutchinson 1957), have been widely explained in terms of endogenous and exogenous

spatio-temporal mechanisms (Roy and Chattopadhyay 2007). Trait-based approaches are

useful to test this matter due to their potential to generalize patterns beyond species' identity,

especially because traits influence the species' ability to acquire resources and persist through

environmental changes (McGill et al. 2006, Díaz et al. 2013, 2016). Nonetheless, niche-based

theory proposes that the environment filters community composition through species'

ecological requirements, which can be perceived through species' traits. However, at small

spatial scales, species interactions present a predominant role, as the environmental

heterogeneity is expected to be small (Götzenberger et al. 2012). In contrast, the more recent

neutral theory suggests that diversity results from random dispersal, speciation, and

extinction rates with no role of niche differences in species coexistence (Hubbell 2001). This

type of dynamics should then result in random distribution of functional traits along

environmental gradients (Kraft et al. 2008, Cornwell and Ackerly 2009).

More recently, it was shown that community organization is driven by eco-evolutionary

processes such as speciation and nutrient uptake kinetics resulting in groups comprising

different species with similar ecological requirements(Gravel et al. 2006, Hubbell 2006,

Scheffer and van Nes 2006). This finding led to the 'emergent neutrality hypothesis' (EN)

(Holt 2006) that has been supported by observational studies, e.g., for phytoplankton from

brackish waters (Segura et al. 2011), birds from the North of Mexico (Thibault et al. 2011),

and beetles at the global scale (Scheffer et al. 2015). EN suggests that species must be

sufficiently similar, and thus, behave neutrally, or different enough in their niches to avoid

competition. Such a scenario would result in species-rich aggregations or clumps along the

niche axis (Scheffer and van Nes 2006, Vergnon et al. 2009, Fort et al. 2010). Modelling

studies have shown that such predictions apply for both steady environmental conditions

(Fort et al. 2010), and also fluctuating resource conditions (Sakavara et al. 2018). Empirical

evidence about EN is still scarce, however (Scheffer et al. 2018).

The clumpy pattern arises from the species niche preferences, meaning that species within

clumps occupy same niche and their abundances are distributed stochastically (Scheffer et al.

2018). However, within a clump, trait differences may be important to species performance in

the niche space (McGill et al. 2006, Violle et al. 2007), and it is difficult to state whether

species behave neutrally within clumps (i.e., when the strength of interspecific interactions

equals the intraspecific interactions). Zooming in on the uniqueness of trait combinations of

species, i.e. functional distinctiveness, within clumps may advance our comprehension of

biotic interactions and move towards a measurable value of similarity at which species

coexistence is driven stochastically. Functional distinctiveness reflects the non-shared

functions among species within a given species pool (Violle et al. 2007), mirroring the

concept of functional redundancy (Pavoine et al. 2017). However, although two functionally

redundant species most likely show high similarity in trait combination, functional

distinctiveness is not directly linked to redundancy, and such metrics are often (Coux et al.

2016, Ricotta et al. 2016, Violle et al. 2017). For example, two species could show similar

functional distinctiveness, i.e. the degree at which a species differs from all the others within

the species pool concerning their functional traits, and still not be similar in their trait

composition at a pairwise level (Coux et al. 2016). This suggests that both functional

redundancy and distinctiveness are good metrics to assess the role of trait combination in

community assembly. To this end, planktonic communities offer an excellenta highly

interesting model for biodiversity theory testing due to their highly speciose communities and

their known relationship between morphological traits and function (Litchman and

Klausmeier 2008, Kruk and Segura 2012, Litchman et al. 2012).

Body size is considered as a master ecological trait and it is highly useful to characterize

species niche (Downing et al. 2014). In phytoplankton, the body size is related to physiology

and life-history (Litchman and Klausmeier 2008), photosynthetic processes (Marañón 2008),

nutrient uptake kinetics (Litchman et al. 2010)and other eco-evolutionary processes (Sauterey

et al. 2017). Despite the importance of body size, the use of a single trait as a proxy for the

niche may not evidence species differences generated by hidden niches and impair the

understanding of clumpy patterns (D'Andrea et al. 2018). The use of multiple traits emerges

as a powerful tool to disentangle plankton functional structure and evaluate competing

hypotheses (Reynolds et al. 2014, Chen et al. 2015, Bortolini and Bueno 2017). Morphology-

based functional groups (MBFG) classification of phytoplankton species (Kruk et al. 2010)

uses a is a multidimensional combination of morphological traits that cluster organisms into

seven groups with similar physiology and ecological responses (Kruk et al. 2011), potentially

overcoming the limitations of using a single trait dimension only. Assessing the uniqueness in

trait combinations of species within the same functional cluster (e.g., clumps, MBFGs) could

help to study the existence of functional equivalence (i.e., neutrality) among species. Overall,

the functional distance among species is a useful tool to compare species in a

multidimensional space by comparing species' functional differences, particularly because the

environment may filter different functional traits across space and time (Mouillot et al. 2013,

D'Andrea et al. 2020).

Riverine ecosystems are highly heterogeneous andsystems characterized by a continuous

water flow that affects the morphology, sedimentation patterns, dispersal, and phytoplankton

diversity (Reynolds and Descy 1996, Wetzel 2001). Several theories explain the longitudinal

occurrence and abundance of river phytoplankton communities, such as River Continuum

(Vannote et al. 1980) and Flood Pulse (Junk et al. 1989) concepts. However, an explicit study

of community size structure and phytoplankton species coexistence under EN in riverine

ecosystems is lacking. For example, phytoplankton species should attain higher biomass at

the middle reaches or in the upper reaches of low-gradient stretches, and the continuous water

flow reduces the likelihood of biotic interactions (Reynolds et al. 1994). However, the

physiological and morphological characteristics of species within the local species pool may

result from eco-evolutionary processes that cluster species into limited functional groups that

are favored under specific environmental conditions (Scheffer et al. 2015). Here, we push

forward three hypotheses to be tested in a tropical river by investigating phytoplankton

community size structure both seasonally and spatially:

H_1 – There are peak aggregations of species abundance (i.e., clumps) along the body size

axis of phytoplankton in the river that remain constant across space and time as a result of

eco-evolutionary processes.

H_2 – Differences in species biovolume scales positively with functional redundancy at the

community-level but not at the clump level, because species within the same clump behave in

a quasi-neutral state. Hence, the dominance within clump varies stochastically between

redundant species because fitness differences would be negligible within clumps.

H_3 – There is a positive relationship between species abundance and species functional

distinctiveness within clumps. Though abundance fluctuates stochastically at pairwise level,

the number of species holding similar trait combinations may affect the likelihood of the

interactions within clumps. Therefore, species with the most distinct trait combinations with

respect to their clump peers are less likely to share same ecological requirements, and by

consequence, attain higher abundance.

Methods

Study area

Samples were taken monthly at nine stations along the Piabanha river between May 2012 and

April 2013. Piabanha river is in the tropical region of Brazil and has a drainage basin of

approximately 4500 km² (Figure 1). The headwater is on Petrópolis at 1546m altitude and

drains to the medium valley of Paraíba do Sul river crossing three cities and with agricultural

activities in their watershed. We set three river stretches (lower, medium and upper courses)

based on the location of steep slopes the river elevation profile (Figure 1). Data from two

meteorological stations, located close to the sampling sites, were used to measure rainfall. We

analyzed meteorological data up to three days before each sampling campaign. We classified

seasons as dry season (May - October) and a wet season (November – April) based on the

rainfall data.

Figure 1. Map of the study area. The watershed area of the Piabanha river showing the

river course (blue line) and the sampling sites colored according to river stretches (white

circles = upper course, blue circles = medium course, red circles = lower course). The vertical

dotted red line in the elevational profile figure indicates the locations of steep slopes used to

define the boundaries of the river stretches.

Sampling and sample analysis

In the field, we measured temperature (°C), dissolved oxygen (DO, mg L^{-1}) and turbidity by

a multiparameter probe sonde (YSI model 600 QS). Water discharge (WD, m^3 s^{-1}) was

measured with the SonTek RiverSurveyor – M9. Furthermore, water samples were taken and

kept frozen (one or 2 weeks) until the laboratory analysis for ammonium (N·NH$_4^+$, mg L^{-1}),

nitrate (N·NO$_3^-$, mgL^{-1}), total phosphorus (TP, mg L^{-1}) and soluble reactive phosphorus

(SRP, mg L^{-1}). The water samples were filtered (except for total phosphorus analysis) using

borosilicate filters (Whatman GF/C), and nutrient concentrations were measured following

(APHA 2005). A complete description of the spatial and temporal patterns of the

environmental variables measured in the Piabanha river can be found elsewhere (Graco-Roza

et al. 2020).

Phytoplankton samples

Subsurface samples of phytoplankton were collected with a bottle of 200 mL and fixed with

Lugol. In the laboratory, phytoplankton species were identified, and population densities

estimated under an inverted microscope (Olympus CKX41) (Utermöhl 1958). At least 100

individuals of the dominant species were counted in each sample (error < 20%, $p < 0.05$,

(Lund et al. 1958, Uhelingher 1964). Biovolumes (mm^3 L^{-1}) were estimated by multiplying

the density of each population by the average volume (V, μm^3), estimated from 20 random

chosen individuals (whenever was possible), using the appropriate geometrical forms

(Hillebrand et al. 1999, Sun and Liu 2003). In addition to the volume, we measured species'

surface area (S, μm^2), the maximum linear dimension (MDL, μm), and the presence of

aerotopes, mucilage, flagella, and siliceous exoskeleton. Hence, we used the volume and

surface area of the species to estimate the individual surface volume ratio (SV). Species were

then classified into Morphological Based Functional Groups (MBFG) according to (Kruk et

al. 2010).

Traits-environment relationship

We tested the relationship between functional traits and the environmental variables using an

RLQ analysis (Dolédec et al. 1996). The RLQ allows the visualization of the distribution of

species traits and their related ecological preferences by testing the co-inertia between the

matrices of environmental variables (R), species abundances (L), and species traits (Q). A

Monte Carlo permutation test with 999 permutations at $\alpha = 0.05$ was used to test the statistical

significance of the RLQ axes. Furthermore, a fourth-corner method was applied to test the

statistical significance of all pairwise associations between species functional traits and

environmental variables. We quantified the strength of the associations using Pearson

correlation coefficient r_p, and the global statistics F-value (Legendre et al. 1997). The p-

values were corrected with 49999 permutations of the sites and the species to correct type I

errors, using false discovery rate method for adjusting p-values to control overall error rate

(Dray et al. 2014). However, because RLQ does not provide a significance test to identify

which combination of environmental variables acts on which combination of functional traits,

and the fourth-corner analysis tests the functional traits one at a time without considering the

covariation among traits and environmental variables, we used also a combined method of

RLQ and fourth-corner analysis (Dray et al. 2014). The significance of the associations was

tested by 99999 permutations of the sites and the species (Dray and Legendre 2008) to

correct type I errors, using false discovery rate method for adjusting p values to control the

overall error rate (Dray et al. 2014).

Clumpy patterns

To test for the existence of peak aggregations of species abundance along the body size axis

of phytoplankton - **H₁**, we analyzed the community structure in each season (dry and wet) and

river stretches (upper, medium, and lower course). First, the individual volume of species was

log-transformed ($\log_2$) and used as the main niche axis (X= $\log_2$ volume). Hence, we divided

the niche axis into equally spaced segments (one segment per unit of $\log_2$ biovolume) and for

each segment (j), we estimated the Shannon entropy (H) using the biovolume of the observed

species (Fort et al. 2010, Segura et al. 2011). The entropy index was defined as:

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

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

tested the significance of the entropy (*H*) by comparing the observed *H* against an expected

uniform distribution under the null hypothesis of homogeneous *H*. For this, we created 1000

communities by sampling the volume of species from a random uniform distribution bounded

by observed individual volumes. Then, each species had a biovolume assigned to it, which

was taken from a randomization of the observed biovolume matrix, keeping both the

empirical species rank-abundance pattern and total biovolume in the sample. For each

segment, the observed *H* was compared with the distributions of *H* generated under the null

hypothesis, with significance defined according to standard 5% criterion (Fort et al. 2010;

Segura et al. 2011).

Functional redundancy

To test whether differences in species biovolume scales positively with functional

redundancy – **H₂**, we first calculated the functional redundancy (F$_{Red}$) and the differences in

biovolume between pairs of species using the whole community and using only the species

from the clumps with significant entropy values. The functional redundancy was obtained by

calculating Gower's general similarity coefficient [function *gowdis*, R package FD (Laliberté

et al. 2014)] on the species functional traits. We used Gower's dissimilarity (*Gd*) because it

can handle mixed variable types (continuous, binary, ordinal and categorical traits). *Gd*

defines a distance value d_{jk} between two species as

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

where, *N* is the number of functional traits considered, x_{ij} the value of trait *i* for species *j* and

x_{ik} the value of the trait *i* for species *k*. We thus tested **H₂** by conducting Mantel tests

[function *mantel*, R package vegan (Oksanen 2017)] with 1000 randomizations, whenever

possible, on F_{Red} and $Diff_{Biov}$ matrices using all species, and separately for the species of each

significant clump.

Functional distinctiveness (F_{Dist})

To test whether there is a positive relationship between species relative abundance at clump-

level and species functional distinctiveness – **H₃**, we estimated the functional distinctiveness

(F_{Dist}) as the distance of a species to the average trait position (centroid) in the

multidimensional functional space for the set of species pertaining to each of the significant

clumps using the equations proposed by Anderson (2006). First, we obtained species

coordinates in the functional space by applying a Principal Coordinates Analysis (PCoA) in

the species-by-traits data table using the Gower's dissimilarity (*Gd*). Hence, the distance from

a species to its given centroid was calculated as

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

where Δ^2 is the squared Euclidean distance between u_{ij}, the principal coordinate for the *j*th

point in the *i*th group, and c_i, the coordinate of the centroid for the *i*th group. The super-

scripted '+' and '-' indicate the real and imaginary parts respectively (Equation 3, see

Anderson 2006 for details). We did not weight the clump-centroid by species abundance and

calculated F_{Dis} using all species pertaining to the significant clumps. Hence, to test **H₃**, we

modeled the relationship between species biovolume and F_{Dist} using linear models. We used

biovolume as response variable, with F_{Dist}, Clump position, and seasonal or spatial categories

as fixed factors. Because the aggregation of sites across space and time have a nested design

(i.e. spatial aggregation comprises all months sampled, and temporal aggregations comprises

all sites sampled), we modelled spatial and temporal categories separately.

Results

Our samples included 150 species that were classified in six (MBFG I, III, IV, V, VI, and

VII) from the seven MBFGs based on their functional traits (Table 1). Species traits were

significantly associated with environmental variables (RLQ analysis; Figure 3; $P < 0.05$).

Individual volume (V), surface volume ratio (SV), maximum linear dimension (MLD), and

siliceous exoskeleton (Si) correlated significantly with the first axis, where SV had the

highest correlation ($r = -0.10$). Also, MLD and flagella correlated significantly with the

second axis ($r = -0.07$, $r = -0.08$). For the environmental variables, water temperature (T) and

water discharge (WD) correlated the strongest with the first axis ($r = 0.10$ for both variables),

which also correlated significantly ($p < 0.05$) with total phosphorus (TP), dissolved oxygen

(DO), turbidity (Turb), and pH. In addition, dissolved inorganic nitrogen (DIN), soluble

reactive phosphorus (SRP), and turbidity (Turb) correlated significantly with the second axis,

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	*Chroococcales* sp, *Chroococcus* sp.
III	3	*Limnothrix* sp.
IV	60	*Pseudoanabaena limnetica*, *Pseudoanabaena catenata*
V	13	*Euglena* sp. *Cryptomonas* sp.
VI	57	*Cymbella* sp., *Synedra* sp.
VII	8	*Dictyosphaerium* sp.
Total	150	

The distribution of the species belonging to different MBFGs in the RLQ ordination (Figure

2B) reflect their occurrence in relation to the environmental variables. The groups IV, V and

VI included 87% of the total number of species. Species from the group IV included

filamentous, colonial and unicellular species ranging from 21 µm³ to 8181 µm³ with no

specialized morphological traits (e.g. flagella, Si). Group V comprised unicellular flagellated

species ranging in size from 31μm³ to 31864 μm³, and group VI included unicellular and

chain-forming species with siliceous exoskeletal body that ranged from 48μm³ to 19045μm³.

According to the fourth-corner analysis, MLD was negatively correlated with pH (r = -0.09)

and WD (r = -0.10), flagella was negatively correlated with Turb (r = -0.08), Si was

positively correlated with Turb (r = 0.11) and WT (r = 0.09), whereas V was positively

correlated with Turb (r = 0.08) (Figure 2C).

Figure 2. Results of the RLQ and fourth-corner analysis. (A) The relationships between

species traits and environmental variables. First and second axes summarized 64.9% and

20.7% of variation, respectively. (B) The distribution of species in the functional space. Each

point in the ordination plot within clusters represents the position of a species modelled

according to its traits on RLQ axes 1 and 2. (C) The bivariate relationships between species

traits and environmental variables. Functional traits are abbreviated as: Si = siliceous

exoskeleton, MLD = maximum linear dimension, SV = surface volume ratio, V = individual

volume. Environmental variables are abbreviated as: TP = total phosphorus, DO = dissolved

oxygen, DIN = dissolved inorganic nitrogen, SRP = soluble reactive phosphorus, Turb =

turbidity, T = water temperature, WD = water Discharge.

The groups IV, V, and VI contributed to 95% of the mean biovolume of total phytoplankton

in dry and wet seasons, and 96.6% across river stretches (Figure 3). The group VI had the

highest contribution for the biovolume of almost all river stretches in both seasons, except for

the upper course during dry season was dominated by the group IV (Figure 3). The entire list

of species, functional traits, and assigned MBFG can be found in Table S1.

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.

Overall, species individual volume ranged from 4.19 μm³ to 31864 μm³ totaling 14 equally

spaced segments (S) of volume along the niche axis. From the 14 segments, three of them

showed significant entropy values, specifically S9, S13, and S14. This resulted in a biovolume

aggregation (i.e. clumps) in two regions of the niche axis considering both temporal (Figure 4)

or spatial categories (Figure 5). The first clump (Clump I) comprised species mainly from the

group IV from S9 (e.g., *Pseudoanabaena limnetica*, *Pseudo anabaena catenata*), whereas the

second clump (Clump II) included species from groups V and VI (e.g., *Euglena* sp., *Cymbella*

sp.) at the segment S14. However, during the wet season, a few species from S13 also had

significant entropy values and therefore were included in the clump II (Fig 4B).

The mean biovolume of species within clumps differed across seasons, but the identity of the

most abundant species did not vary. *Pseudanabaena* sp. 4 (spp. 28) and *Pseudanabaena*

catenata (spp. 12) had the highest biovolumes of clump I and held the highest biovolumes of

the clump at both dry and wet seasons. Within the clump II, *Synedra* sp. (spp. 80) attained the

highest biovolume during the dry season, whereas *Cymbella* sp. (spp. 051) contributed the most

to clump biovolume in the wet season.

Figure 4. Distribution of phytoplankton biovolume along the body size axis, the ordination of species from the significant size segments (S) in the functional space, and the mean biovolume of the five most abundant species of each significant size segment during the dry and wet seasons of the Piabanha river, RJ. (A) Stem plots show size distribution in the sampling sites of the river Piabanha. Each stem represents a species with its body size (in $\log_2$) plotted on the abscissa and the mean biovolume plotted on the ordinate. The red dotted line indicates the entropy value of each size segment (i.e., one unit of $\log_2$ volume), and the asterisk highlights the significant entropy values tested through 1000 randomizations. (B) The species of the corresponding significant size-segment are ordinated in the functional space. The size of the circles represents the species contribution to the total biovolume of the size segment. (C) Bar plots (right show the biovolume of the five most abundant species from each significant size segment. Species are colored according to their morphology-based

functional groups (MBFG). The code of each species can be found in the supplementary

material, Table S1.

Regarding the river stretches, *Pseudoanabaena* sp. 4 and *Pseudoanabaena catenata*

contributed the most to the biovolume of the upper course, and only the clump I had significant

entropy values for species from the S9 (Figure 5A). At the medium course, both the clump I

and II had significant entropy values, mainly due to species from the S9 with *Pseudoanabaena*

sp. 4 as the most abundant, and from the S14 with *Synedra* sp. as the most abundant (Figure

5B). At the lower course, the clump II had significant entropy values, specifically at the S14

where *Synedra* sp. contributed most to biovolume at segment level (Figure 5C).

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

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

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

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

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

functional groups (MBFG). The code of each species can be found in the supplementary

material, Table S1.

Differences in biovolume among species were positively related to the functional redundancy

at the whole-community level, but not at the clump-level irrespectively to the season or river

stretch studied (Table 2). At the clump-level, we observed that species with higher

distinctiveness attained the highest biovolume in all the seasons and river stretches (except for

the lower course). The upscaling of biovolume with functional distinctiveness was observed

for species from both clumps, at different seasons and river stretches, but was particularly

different in the wet season, where species from clump II did not show any significant

relationship with biovolume, and at the medium and lower courses, where there were no

significant trends (Table 3).

Table 2. Mantel correlation results. Mantel correlation between the differences in species

biovolume and functional redundancy for the overall set of species and significant clumps

along the seasons (dry and wet) and river stretches (upper, medium and lower courses).

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

	Overall	**0.165 (0.002)**
	Clump I	0.065 (0.242)
Medium course		
	Overall	**0.171 (0.001)**
	Clump I	0.059 (0.217)
	Clump II	-0.168 (0.633)
Lower course		
	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[*]	15.854[*]	24.440[**]	12.377	7.267
	(6.127)	(6.015)	(7.816)	(6.085)	(4.373)
Clump II	4.676	7.501[*]	9.619	4.373	2.067
	(6.148)	(3.175)	(18.074)	(5.169)	(3.689)
$F_{Dist} \times$ Clump II	-8.458	-15.863	-21.301	-7.911	-1.651
	(15.608)	(7.734)	(46.232)	(13.329)	(9.501)
Intercept	-8.979[**]	-9.748[**]	-12.951[**]	-7.945[**]	-6.165[**]
	(2.439)	(2.496)	(3.143)	(2.418)	(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 Std. Error	0.756 (df = 22)	0.663 (df = 27)	0.823 (df = 15)	0.755 (df = 24)	0.523 (df = 22)
F Statistic	5.313[**]	6.987[**]	4.204[*]	4.778[**]	10.090[**]

$$(df = 3; 22) \quad (df = 3; 27) \quad (df = 3;15) \quad (df = 3;24) \quad (df = 3; 22)$$

Note: *p < 0.05 **p < 0.01

Discussion

Present results showed that (i) the clumps in body size are a conspicuous feature of

phytoplankton community structure in riverine systems across seasons and river stretches; (ii)

species within clumps show random distribution of biovolume with respect to their pairwise

functional differences, but functional redundancy is related to differences in biovolume at the

whole-community level; and (iii) species biovolume generally scales positively with

functional distinctiveness at the clump level. Note that, on the one hand, multimodal

aggregation of species biovolume along body size axis only points to the integration of niche-

based processes and neutrality driving community assembly (Vergnon et al. 2009), but it does

not imply that EN is the only mechanism behind phytoplankton coexistence. On the other

hand, our results do not support alternative hypothesisexplanations such as pure neutrality

(Hubbell 2001) or high dimensional hypothesis (Clark et al. 2007) because pure neutrality

predicts a uniform distribution of species biovolume along the niche axis (Hubbell 2001), and

the HDH does not predict any particular trait distribution (Vergnon et al. 2009, Ingram et al.

2018).

One alternative theory that is likely to explain clumpy aggregations is the Holling's textural

hypothesis (Holling 1992). The textural hypothesis explains multimodal species size

distribution with environmental constraints, and although distribution of the clumps was

spatially structured, our results do not support HDH as a whole. River stretches and seasons

are radically different in river hydrology, nutrients and fluxes, but that was not reflected in

phytoplankton size structure. Moreover, if landscape controls the body size pattern alone,

differences in phylogenetics, biogeography, energeticsenergenitics, and community

interactions should not significantly affect such pattern (Allen et al. 2005). However, the

clumps in our data were in the similar size range and comprised species from the same

MBFG found in previous studies from different systems (Segura et al. 2011, 2013). More

importantly, there was a clear functional differentiation of species between clumps in terms

of MBFG (Figure 5-6), and the trait composition of the communities reflected adaptations to

survive in different environmental templates (Kruk and Segura 2012). For example, species

from the group IV dominated in low-flow waters with high nutrient input, while species from

group V and VI show different adaptations to survive in turbulent and mixed conditions

(Margalef 1978, Soares et al. 2007, Bortolini and Bueno 2017). Therefore, our findings

corroborate our H_1 on the existence of significant peak aggregations in species biovolume

along the body size axis. Such results agree with previous evidence of EN using

phytoplankton as the model community in coastal and estuarine environments (Segura et al.

2011, 2013), and shows that clumpy distributions are evident on the phytoplankton

communities of riverine ecosystems presenting fluctuations in environmental conditions. In

addition, our findings are in line with recent modeling results suggesting clumpy patterns

arise in environments subjected to resources fluctuation (Sakavara et al. 2018). It remains to

be tested the trade off between resources among competing species, which is the remaining

ingredient for the emergence of clumps.

We expanded the level of analysis from mere size structure towards the inclusion of new trait

dimensions by analyzing the functional similarity at MBFG classification level and functional

redundancy. We found that the species from the same clump represent a selection of all the

possibilities amongst MBGFs, reinforcing that body size is a good proxy for niche

preferences of the species (Blanckenhorn 2000). However, assessing MBFGs helped to

assign niche differences at a finer degree, taking into account that using only a single trait

could impair the patterns behind community assembly (D'Andrea et al. 2018). Species within

the same MBFG are expected to share ecological strategy (Kruk et al. 2010, Kruk and Segura

2012) and accordingly to EN, would also perform similarly (Scheffer et al. 2018). Still, the

importance of traits to equalize or stabilize niche differences are context-dependent and

assessing the functional multidimensional space should help to unveil community assembly

processes (Bonser 2006). Here, we found that size-related traits (volume, MLD) and the

presence of flagella or siliceous exoskeleton were the main drivers of phytoplankton

distribution. Not surprisingly, such traits are key determinants on the assignment of species to

the main MBFG observed here (Kruk et al. 2010). Furthermore, we found that fitness

differences were significant at the community level, but not at within-clumps level, even

considering multiple morphological traits. Still, the dominance of medium-sized species

(clump I) in the upper course and of large-bodied species (clump II) in the lower course

suggests that the coexistence of these MBFGs tends to generate clumpy distributions that are

eroded when the ecosystem properties favor the dominance of other MBFGs, but the

fluctuations in biovolume within clumps remains stochastic even when clumps hold a few

species belonging to different MBFGs. These findings corroborate our **H2** - differences in

species biovolume scales positively with functional redundancy at the community level but

not at within-clumps level - and reinforces the predictions on the quasi-neutrality state of

species within clumps (Holt 2006).

We also outlined the role of functional similarity in community assembly by studying the

effects functional distinctiveness on species biovolume at clump level. We found that at the

clump I, species biovolume increased with functional distinctiveness, but this pattern was

weaker within clump II. Despite the species from group V and VI show lower growth rates

than the species from group IV (Kruk et al. 2010), their elongated shape provides advantage

under turbulent conditions (Reynolds et al. 1994) favoring these species under high flow

conditions on nutrient-rich environments (Irwin et al. 2006). Furthermore, nutrients are

hardly limiting phytoplankton growth especially in meso- or eutrophic rivers (Reynolds

2006), and large-bodied species are mostly randomly introduced from different habitats (e.g.

periphyton or epiphyton) (Wang et al. 2014, Descy et al. 2017), which brings some

uncertainty to the assumption of niche overlap due to size similarity. In fact, nutrients were

not relevant for structuring the phytoplankton community of the Piabanha river, and we also

observed a significantly different size-segment (S13) within the clump II during the high flow

periods. Thus, one cannot be sure if species within clump II are truly occupying same niche

or if they just were introduced to the plankton by random displacement. We suggest this may

be the mechanism leading to the weak relationship between functional distinctiveness and

biovolume within clump II.

Emergent neutrality results from eco-evolutionary processes that leads species selection

towards a limited number of functional groups (Scheffer and van Nes 2006). This implies that

the clumps observed here are not likely a result of competitive exclusion at the Piabanha

river, but a convergent evolution of competing species over time as previously suggested

(Macarthur and Levins 1967). Therefore, even when the competition rates are relaxed due to

sufficient nutrient supply, some other limiting factors that are not consumed by biotic

organisms such as heat energy or turbulence determine species biovolume, and even minor

shifts in species trait combination may lead to higher biovolume with respect to other species

occupying same niche. Our results showed that it is also possible to predict the biovolume of

species within clumps, but only when neutral processes (e.g., species displacement) are

relaxed and biotic interactions are more likely to occur. Therefore, our findings partially

agree with H_3 - there is a positive relationship between species abundance and species

functional distinctiveness within clumps, but the environmental conditions seem to play a key

role in the outcome.

In summary, we provided evidence of both neutral and niche perspective driving planktonic

community assembly and support the view that emergent neutrality is a likely mechanism to

explain species coexistence in an open and environmentally heterogeneous ecosystem. The

use of MBFG classification and functional space to describe species within clumps revealed

that under the same size range, species with a greater degree of redundancy alternate their

dominance in an unpredictable way. The significance and dominance of the clumps were

related to the environmental conditions, but the biovolume of species within the clumps was

better predicted by functional distinctiveness than by pairwise functional redundancy. This

addresses the difficulty to avoid the ghost of hidden niches (Barabás et al. 2013), and also

provides evidence from multiple angles that points to EN as a plausible mechanism in

shaping species coexistence.

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