

Geographical variation in the trait- based assembly patterns of multitrophic invertebrate communities

Srivastava Diane S. ^{1,*}, Macdonald A. Andrew M. ^{2,3,4}, Pillar Valério D. ⁵, Kratina Pavel ⁶, Debastiani Vanderlei J. ⁵, Guzman Laura Melissa ^{1,7}, Trzcinski Mark kurtis ⁸, Dézerald Olivier ^{9,10}, Barberis Ignacio M. ¹¹, De Omena Paula M. ^{12,13}, Romero Gustavo Q. ¹², Ospina-bautista Fabiola ^{14,15}, Marino Nicholas A. C. ^{16,17}, Leroy Céline ^{18,19}, Farjalla Vinicius F. ¹⁶, Richardson Barbara A. ^{20,21}, Gonçalves Ana Z. ²², Corbara Bruno ²³, Petermann Jana S. ²⁴, Richardson Michael J. ^{20,21}, Melnychuk Michael C. ²⁵, Jocqué Merlijn ²⁶, Ngai Jacqueline T. ¹, Talaga Stanislas ^{27,28}, Piccoli Gustavo C. O. ²⁹, Montero Guillermo ¹¹, Kirby Kathryn R. ⁸, Starzomski Brian M. ³⁰, Céréghino Régis ²

¹ Department of Zoology & Biodiversity Research Centre University of British Columbia Vancouver BC ,Canada

² Laboratoire Ecologie Fonctionnelle et Environnement CNRS, Université Toulouse 3 Paul Sabatier Toulouse ,France

³ Centre for the Synthesis and Analysis of Biodiversity (CESAB-FRB) Montpellier, France

⁴ The Synthesis and Analysis of Biodiversity (CESAB-FRB) Aix-en-Provence,France

⁵ Department of Ecology and Graduate Program in Ecology, Universidade Federal do Rio Grande Porto Alegre, Brazil

⁶ School of Biological and Behavioural Sciences Queen Mary University of London London ,UK

⁷ Department of Biological Sciences Simon Fraser University Burnaby BC ,Canada

⁸ Department of Forest and Conservation Sciences University of British Columbia Vancouver BC, Canada

⁹ EcoFoG, Ecologie des Forêts de Guyane, CNRS UMR 8172 Kourou ,France

¹⁰ ESE, Ecology and Ecosystems Health, INRAE Rennes, France

¹¹ Facultad de Ciencias Agrarias Instituto de Investigaciones en Ciencias Agrarias de Rosario, IICAR-CONICET-UNR, Universidad Nacional de Rosario Zavalla ,Argentina

¹² Laboratory of Multitrophic Interactions and Biodiversity, Department of Animal Biology Institute of Biology, University of Campinas Campinas, Brazil

¹³ Institute of Biological Sciences Federal University of Pará Belém ,Brazil

¹⁴ Department of Biological Sciences Andes University Bogotá, Colombia

¹⁵ Departamento de Ciencias Biológicas Universidad de Caldas Caldas ,Colombia

¹⁶ Departamento de Ecologia Instituto de Biologia, Universidade Federal do Rio de Janeiro Rio de Janeiro, Brazil

¹⁷ Programa de Pós-Graduação em Ecologia Universidade Federal do Rio de Janeiro Rio de Janeiro, Brazil

¹⁸ AMAP, Univ. Montpellier, CIRAD, CNRS, INRAE, IRD Montpellier, France

¹⁹ ECOFOG, CIRAD, CNRS, INRAE, AgroParisTech, Université de Guyane, Université des Antilles Kourou, France

²⁰ Edinburgh, UK

²¹ Luquillo LTER, Institute for Tropical Ecosystem Studies, University of Puerto Rico San Juan ,Puerto Rico

²² Department of Botany Biosciences Institute, University of São Paulo São Paulo ,Brazil

²³ Laboratoire Microorganismes, Génome et Environnement Université Clermont Auvergne Aubière, France

²⁴ Department of Biosciences University of Salzburg Salzburg ,Austria

²⁵ School of Aquatic and Fishery Sciences University of Washington DC ,USA

²⁶ Aquatic and Terrestrial Ecology Royal Belgian Institute of Natural Sciences Brussels ,Belgium

²⁷ Institut Pasteur de la Guyane Unité d'Entomologie Médicale Cayenne ,France

²⁸ MIVEGEC, Univ. Montpellier, CNRS, IRD Montpellier, France

²⁹ Department of Zoology and Botany University of São Paulo State São José do Rio Preto, Brazil

³⁰ School of Environmental Studies University of Victoria Victoria BC, Canada

* Corresponding author : Diane S. Srivastava, email address : srivast@zoology.ubc.ca

Abstract :

1. It has been argued that the mechanisms structuring ecological communities may be more generalizable when based on traits than on species identities. If so, patterns in the assembly of community-level traits along environmental gradients should be similar in different places in the world. Alternatively, geographical change in the species pool and regional variation in climate might result in site-specific relationships between community traits and local environments. These competing hypotheses are particularly untested for animal communities.

2. Here we test the geographical constancy of trait-based assembly patterns using a widespread multi-trophic community: aquatic macroinvertebrates within bromeliads. We used data on 615 invertebrate taxa from 1,656 bromeliads in 26 field sites from Mexico to Argentina. We summarized invertebrate traits with four orthogonal axes, and used these trait axes to examine trait convergence and divergence assembly patterns along three environmental gradients: detrital biomass and water volume in bromeliads, and canopy cover over bromeliads.

3. We found no overall signal of trait-based assembly patterns along any of the environmental gradients. However, individual sites did show trait convergence along detrital and water gradients, and we built predictive models to explore these site differences.

4. Sites that showed trait convergence along detrital gradients were all north of the Northern Andes. This geographical pattern may be related to phylogeographical differences in bromeliad morphology. Bromeliads with low detritus were dominated by detritivorous collectors and filter feeders, where those with high detritus had more sclerotized and predatory invertebrates.

5. Sites that showed the strongest trait convergence along gradients in bromeliad water were in regions with seasonal precipitation. In such sites, bromeliads with low water were dominated by soft-bodied, benthic invertebrates with simple life cycles. In less seasonal sites, traits associated with short-term desiccation resistance, such as hard exoskeletons, were more important.

6. In summary, we show that there are strong geographical effects on the trait-based assembly patterns of this invertebrate community, driven by the biogeography of their foundational plant species as well as by regional climate. We suggest that inclusion of biogeography and climate in trait-based community ecology could help make it a truly general theory.

Keywords : bromeliad invertebrates, functional biogeography, habitat filtering, trait-based ecology

76 **Introduction**

77 Traits define the ways in which organisms interact with their surrounding environment and
78 other species. As such, traits provide mechanistic links between species and the niche processes that
79 may structure ecological communities (McGill, Enquist, Weiher, & Westoby, 2006). This presents the
80 enticing possibility that ecological concepts based on traits may be more generalizable than those
81 based on biological species. Shipley et al. (2016) argue that a “foundational claim” of trait-based
82 ecology is that communities should show similar trait-based assembly patterns along environmental
83 gradients, even if species composition differs among sites, asking: “Why, given the foundational
84 importance of such patterns, do we have so few examples of generalizable and quantitative trait-based
85 environmental gradients?”. Certainly there are some relevant examples, including mesic grasslands
86 where leaf traits change predictably with fire frequency (Forrestel, Donoghue, & Smith, 2014), and
87 tropical forests where foliar chemistry changes consistently over elevation (Asner & Martin, 2016),
88
89 However, such geographically repeated patterns may not be the general rule for three reasons. First,
90 these patterns require that the trait states favoured at different points along the environmental gradient
91 must be present in all species pools, yet the traits of species pools may be constrained by dispersal
92 (limited, for example, by habitat fragmentation: Zambrano et al., 2019) and micro-and macro-
93 evolution (Denelle, Violle, & Munoz, 2019). For example, the trait space occupied by spider and
94 beetle communities in native forests of the Azores archipelago is determined by the rate that exotic
95 species colonise islands (Whittaker et al., 2014). Second, there may be multiple trait solutions to the
96 challenges posed by environments (Warming, 1909), and the particular trait solution exhibited by a
97 community may be constrained by the taxonomic composition of its species pool (Peet, 1978; Pillar &
98 Orłóci, 1993). For example, the traits that desert plant communities use to survive aridity depend on
99 which families are represented in the species pool (Peet, 1978). Third, large-scale bioclimatic factors
100 determine how local environmental gradients filter traits. An example here is the shift in leaf nutrients
101 over elevational gradients, which is accentuated in warmer regions (Midolo, De Frenne, Hölzel, &
102 Wellstein, 2019). In short, the geographic consistency of trait-environment coupling in communities

103 could be influenced by both bioclimatic and biogeographic constraints (Barnagaud et al., 2019;
104 Violle, Reich, Pacala, Enquist, & Kattge, 2014).
105
106 It has been argued that a mechanistic understanding of community trait patterns requires the
107 underlying environmental drivers to be explicitly included in analyses (Pillar & Orłóci, 1993; Shipley
108 et al., 2016). Traits integrated at the community level can change in two potential ways along
109 environmental gradients (Figure 1). A trait-convergence assembly pattern (TCAP: Pillar, Duarte,
110 Sosinski, & Joner, 2009) occurs when communities at similar positions on an environmental gradient
111 converge on a common suite of traits. Such a pattern is often assumed to be underlain by strong
112 habitat filtering at intra- or interspecific levels. A trait-divergence assembly pattern (TDAP: Pillar et
113 al., 2009) occurs when trait dispersion within communities changes over the environmental gradient.
114 For example, species coexistence in competitive communities may require sufficient dispersion of
115 traits to minimize niche overlap, and the strength of the required niche differentiation may change
116 systematically with environmental context (but see Mayfield & Levine, 2010). Trait divergence
117 patterns can also be generated when either the strength of habitat filtering or environmental
118 heterogeneity changes systematically over the gradient in mean environmental conditions (Carlucci,
119 Streit, Duarte, & Pillar, 2012). Communities can simultaneously show trait convergence and
120 divergence along environmental gradients, although different traits may be involved in each process
121 (Pillar et al., 2009).

122
123 To date, studies that have examined the geographic consistency of community trait patterns have
124 focused largely on trait convergence (but see Bruelheide et al., 2018). However, a complete
125 understanding of community trait-based assembly requires examining both trait convergence and
126 divergence along similar environmental gradients in multiple regions of the world. Furthermore, the
127 majority of studies that have quantified trait-based assembly patterns have been conducted in plant
128 communities (e.g. Carlucci et al., 2012; Mendes et al., 2016). Animal communities may differ from
129 plant communities in trait-based assembly patterns for several reasons. It has been argued that animals
130 show less phenotypic plasticity than plants because animals have determinate growth and complete

131 organogenesis at the embryonic stage, as opposed to the indeterminate growth and lifelong production
132 of new organs (e.g. leaves) in plants (Borges, 2008). If so, we would expect weaker trait-assembly
133 patterns in animal than plant communities, at least when analyses considered intraspecific trait
134 variation. Alternatively, the behavioural traits of animals can be remarkably plastic and individualistic
135 which, when coupled with the ability of many animals to actively move, can lead to strong
136 associations of individual-level traits with environments (Dahirel, Dierick, De Cock, & Bonte, 2017).
137 Animal communities may also include a broader range of species interactions than plant communities,
138 such as predator-prey interactions in multi-trophic communities, complicating any theoretical
139 expectation that trait dispersion within communities aids coexistence. Although pairwise trophic
140 interactions can be predicted from traits (Brousseau, Gravel, & Handa, 2018), scaling these pairwise
141 patterns to community-level trait assembly patterns still remains elusive (Wong, Guénard, & Lewis,
142 2019). Finally, animal communities may not only experience environmental gradients directly, but
143 also indirectly via the traits of the plants that animals consume or use as foundational habitat (Abgrall
144 et al., 2017; Ohgushi, Craig, & Price, 2007; Pakeman & Stockan, 2014).

145

146 Here we examine the geographic consistency of trait convergence and divergence assembly patterns
147 along local environmental gradients, using a community found throughout the Neotropics: the
148 freshwater macroinvertebrate community living in water impounded by bromeliad plants (Figure 1).
149 The aquatic food webs within bromeliads are fuelled by decaying detritus and algae and dominated by
150 macroinvertebrates, especially insect larvae. These invertebrates include detritivores which shred and
151 scrape detritus and its biofilm, collectors and filter feeders of fine particulate organic matter and algae,
152 and intermediate and top predators (Céréghino et al., 2018; Diane S. Srivastava et al., 2004). We test
153 two hypotheses: (H1) bromeliad invertebrate communities are structured by similar niche processes in
154 all sites (“Multi-site” panel in Figure 1), or alternatively (H2) sites differ in the strength or drivers of
155 trait-based assembly patterns. If H1 is true, we would expect (a) geographically general patterns in
156 trait-based assembly over local environmental gradients, and (b) these patterns to be driven by the
157 same traits in every site. If H2 is true, we would expect site differences in trait-based assembly
158 patterns to be related to climate (temperature and precipitation) or biogeography (species pool and

159 dispersal barriers). Alternatively, differences in the sampling of sites (in terms of number or average
160 size of bromeliads) could obscure a geographically general pattern.

161

162 **Materials and Methods**

163 *Field sampling.* We compiled data on the aquatic macroinvertebrates in tank bromeliads previously
164 sampled in 26 different sites (Figure 1; Table S1) throughout the natural distribution of the tank
165 bromeliads (Bromeliaceae family). We obtained research permits for each of these field sites where
166 required (Table S1b); in no field site was approval from an animal ethics board required. Field sites
167 were distinct from each other in space, elevation and the species composition of invertebrate
168 communities (Supplemental Material). For every bromeliad, all water and detritus contained in the
169 plant were removed, either by dissecting the plant or by pipetting. The water and detritus were
170 examined for aquatic macroinvertebrates in small size-fractionated aliquots in white trays.
171 Macroinvertebrates were identified to morphospecies in the field, and subsequently to the lowest
172 possible taxonomic level. The detritus was oven-dried and weighed to determine dry mass.
173 Bromeliads were sampled across a range of habitats, from exposed *restinga* (coastal sand-based shrub
174 habitat) in Brazil to cloud forests on Caribbean mountaintops to rainforests in Central America. As no
175 bromeliad genus was found in all field sites, we sampled the most common genera in sites. As this
176 difference in bromeliad taxonomy between sites can lead to site differences in mean plant size, we
177 explicitly test whether mean size drives site differences in trait-based assembly patterns in a *post hoc*
178 test described later. In sum, the dataset consists of information on environmental attributes of
179 bromeliads (n = 1656), macroinvertebrate morphospecies (n = 615), macroinvertebrate traits (n = 64)
180 and field sites (n = 26).

181

182 *Environmental matrix.* Our analysis included three environmental characteristics of each sampled
183 bromeliad: (1) the dry mass of all detritus in the bromeliad (hereafter “detritus”, measured in grams);
184 (2) the volume of standing water in each bromeliad on the day of sampling (hereafter “water”,
185 measured in mL); (3) the openness of the canopy above the bromeliad (hereafter “canopy”, a binary
186 variable with 1 = open canopy and 0 = closed canopy). These variables were chosen because previous

187 site-specific research had established that they were important environmental drivers of community
188 structure and function (Montero, Feruglio, & Barberis, 2010; Petermann et al., 2015; Richardson,
189 1999; Romero, Piccoli, De Omena, & Gonçalves-Souza, 2016), and because there was enough
190 coverage within and amongst field sites to enable robust analyses (Figure S1). A final consideration
191 was that the three environmental gradients were not collinear (pairwise Pearson correlations, $r = 0.31$
192 to 0.36). For example, we did not include bromeliad water-holding capacity, even though it is known
193 to be an important driver of community composition, because it was tightly correlated ($r = 0.80$) with
194 the volume of water on the day of sampling and we had much higher data coverage of the latter. Our
195 environmental matrix (E , *sensu* Pillar et al. 2009, see below) consists of the three environmental
196 variables describing each of the sampled bromeliads.

197

198 *Community biomass matrix.* We defined as our community all macroinvertebrates found in
199 bromeliads that were macroscopic and either strictly aquatic or semi-aquatic. We organized the
200 abundance data using the R package *fwdata* (developed by A.A.M.M,
201 <https://github.com/SrivastavaLab/fwdata>). We then converted abundances to biomass by multiplying
202 abundance by the estimated per capita biomass of each morphospecies, based on taxon-specific
203 allometric relationships, using the *hellometry* R package (provided courtesy of P. Rogy,
204 <https://github.com/pierreroy/hellometry>). We used biomass rather than abundance to weight traits as
205 many large-bodied invertebrates in bromeliads (e.g. damselflies, tabanids, and sometimes tipulids) are
206 known to have strong consumptive effects on other species (Amundrud et al., 2019; Petermann et al.,
207 2015) but have such low abundance as to be essentially invisible in abundance-weighted traits. Our W
208 matrix (*sensu* Pillar et al. 2009, see below) describes the biomass of each morphospecies of
209 macroinvertebrate as a proportion of the total macroinvertebrate biomass within each sampled
210 bromeliad.

211

212 *Trait matrix.* Our analysis considers interspecific, but not intraspecific, differences in traits. In
213 Céréghino et al. (2018), each bromeliad invertebrate morphospecies was scored in terms of twelve
214 traits: aquatic developmental stage, body form, maximum body size, cohort production interval,

215 dispersal mode, food, feeding group, locomotion, morphological defence, reproduction mode,
216 resistance forms and respiration mode. Each trait was represented by several modalities or categories
217 (e.g. the modalities for the trait “dispersal mode” were passive and active), and the affinity of the taxa
218 for each modality was fuzzy coded. In total, the 12 traits were represented by 64 modalities.
219 Céréghino et al. (2018) reduced these 64 trait modalities to four orthogonal axes using PCA. Since
220 then, there have been modest updates to the trait data: some trait scores were improved, the number of
221 missing values reduced, a few microscopic or terrestrial species were removed, and the taxonomic
222 resolution of some species identifications was improved. We therefore reran the PCA analysis on the
223 updated trait matrix, using the same R script as in Céréghino et al. (2018), and used the first four axes
224 in our current study. Our *B* matrix (*sensu* Pillar et al. 2009, see below) describes the morphospecies of
225 macroinvertebrates in terms of the four PCA trait axes.

226

227 *Site information.* We collated information on biogeographic, bioclimatic, and sampling characteristics
228 of each field site in order to better contextualize differences among sites. Biogeographic
229 characteristics included position north and north west versus south and southeast of the Northern
230 Andes (simplified hereafter as north versus south of Andes; Figure 1). The Northern Andes are known
231 to be a dispersal barrier for both bromeliads (Givnish et al., 2011) and bromeliad invertebrates
232 (Amundrud, Videla, & Srivastava, 2018). Although the Southern and Central Andes could be a
233 potential barrier between the west coast and center of South America, we have no bromeliad data
234 from the former and so do not analyse these mountain ranges. We examined species pool richness,
235 estimated with Chao’s method (Chao, 1987; implemented in the vegan R package), to test if richer
236 sites had stronger trait-environment matching. From the WorldClim database (Fick & Hijmans, 2017),
237 we extracted site-specific estimates of four bioclimatic variables that a previous study (Guzman et al.,
238 2020) found to underlie spatial variation in bromeliad macroinvertebrate traits: mean diurnal range in
239 temperature (BC2), temperature annual seasonality (BC4), precipitation annual seasonality (BC15)
240 and precipitation of the driest quarter (BC17). WorldClim data was extracted at the 1km² scale; when
241 field sites exceeded 1 km² in size we averaged data over the relevant 1 km² pixels. Finally, we
242 examined sampling characteristics of each site, including the number of bromeliads sampled (which

243 affects the power of tests) and the site mean of the focal environmental gradient (in case trait-
244 environment relationships differ with site position on gradient).

245

246 *Analysis of trait-based community assembly patterns*

247 Our analysis of trait-based community assembly patterns is based on the methodology developed by
248 Pillar et al. (2009) and implemented in the R package SYNCSEA (Debastiani & Pillar, 2012). A trait-
249 convergence assembly pattern (TCAP) can be visualized as the multivariate correlation between the
250 average trait values of the community with the environment experienced by that community (Fig. 1).
251 The former is represented in the community-weighted trait mean (a new matrix, T), calculated by
252 multiplication of W (bromeliads by morphospecies in our data) and B (morphospecies by trait axes);
253 the latter is the environmental (E) matrix. TCAP is therefore simply the T - E correlation. Individual
254 taxa will differ in trait values from the mean of their community (i.e. be dispersed), and the trait-
255 divergence assembly pattern (TDAP) describes how this dispersion itself correlates with the
256 environmental gradient (Figure 1). Calculating TDAP requires first creating a new matrix X such that
257 the X - E correlation contains both TCAP and TDAP, and then partialling out the T - E correlation (i.e.
258 TCAP) to reveal TDAP (Pillar et al., 2009). In technical terms, the X matrix is created by weighting
259 the species in matrix W by their degree of belonging to fuzzy sets that describe similarities between
260 the species in traits (Duarte, Debastiani, Freitas, & Pillar, 2016; Pillar et al., 2009).

261

262 The significance of the matrix correlations that describe TCAP and TDAP must be tested through
263 permutations to ensure that traits, rather than the species that they are associated with, drive the
264 correlation. The SYNCSEA method compares the observed correlation coefficient to those obtained
265 after repeatedly permuting row vectors of the B matrix. This permutation breaks up the association
266 between species and their traits while maintaining trait correlation structure, and is an appropriate null
267 model for trait-based assembly patterns (Pillar et al., 2009). In order to incorporate our field site
268 structure into the TDAP and TCAP analysis, we added a “strata” field to the SYNCSEA R package,
269 which directs the permutations to be entirely within field sites so as to preserve the integrity of the
270 species pool in each site.

271

272 All matrix correlations were based on Procrustes analysis (Peres-Neto & Jackson, 2001), which
273 describes the concordance between two superimposed matrices after optimizing their resizing,
274 reflection and rotation, with 0 = no concordance and 1 = perfect concordance. We chose Procrustes
275 analysis over the commonly used Mantel test, as it is more powerful (Peres-Neto & Jackson, 2001)
276 and less prone to spurious correlations (Dutilleul, Stockwell, Frigon, & Legendre, 2000).

277

278 *Analyses of across- and between-site patterns*

279 To test if trait-based assembly patterns are universal (hypothesis H1) or not (H2), we first conducted a
280 global test of either TCAP- and TDAP-environment associations, that is, considering all sites together
281 but restricting permutations within species pools. We included all four trait axes to capture the entire
282 multivariate trait space of the community. As sites differed substantially in their coverage of
283 environmental variables, we examined each environmental gradient separately. We log-transformed
284 both detritus and water data, as this improved correlations with TCAP and TDAP. Within each site,
285 we centered environmental gradients (i.e. subtracted the site mean, after any log transformation) to
286 ensure that our multi-site analysis only captured within-site correlations.

287

288 We further examined support for H1 by repeating the analysis (a) at the site level to determine if
289 many sites contributed to the overall pattern, and (b) for each trait axis individually, to determine if
290 the same specific trait axes drove associations in all sites. As the inclusion of uninformative traits
291 dilutes tests of trait-environment relationships, it is recommended to consider which traits (or in our
292 case, trait axes) optimize TCAP and TDAP as part of analyses (Debastiani & Pillar, 2012; Pillar,
293 Sabatini, Jandt, Camiz, & Bruelheide, 2021).

294

295 Where site differences in TCAP-and TDAP-environment associations existed, we tested hypothesis
296 H2 by using linear models to relate site differences in the strength of these associations to sampling
297 (sample size, mean bromeliad environment), bioclimatic (BC2, BC4, BC15, BC17) and biogeographic
298 (species pool size, location north and south of Northern Andes) characteristics of sites. We established

299 the significance of explanatory variables using ANOVA based on type II sums-of-squares to prevent
300 collinearity from influencing results. We only constructed linear models for gradients represented by
301 sufficient sites, namely gradients in detrital and water content. We visualized patterns in TCAP by
302 plotting community-weighted trait means for specific trait axes against the environmental gradient. As
303 TDAP patterns were largely not significant, we do not present visualizations.

304

305

306 **Results**

307 Our updated PCA of the functional traits of bromeliad invertebrate taxa (Figure S2, Table S2)
308 can be interpreted in terms of the first four axes (which together represented 45.1% of the total
309 inertia). The first axis can broadly be interpreted as separating flattened, sclerotized taxa, many of
310 which are predators, from primary consumers that feed on algae, fine particulates and
311 microorganisms. The second axis separates siphon or spiracle-respiring insects found in the water
312 column from integument-respiring taxa that live in the benthos, often for their entire life cycle. The
313 third axis separates taxa with dorsal plates or sclerotized spines from those with few morphological
314 defences and simple, benthic life cycles. The fourth axis separates invertebrate taxa that spend their
315 entire life cycle in bromeliads from insects, which generally only spend their larval stage in
316 bromeliads. In general then, trait axes 1, 2, 3 and 4 can be simplified to trophic, habitat, defence and
317 life cycle dimensions, respectively, as previously reported in Céréghino et al. (2018). We used these
318 four orthogonal trait axes in all subsequent analyses of trait-assembly patterns (i.e. TCAP and TDAP).
319
320 Using these four trait axes, we then explored overall patterns in trait assembly. Considering all field
321 sites together and all trait axes together, there were no overall trait-based assembly patterns (neither
322 TCAP nor TDAP) along any of the environmental gradients (Figure 2). This contradicts hypothesis
323 1a. Site-level analyses of TDAP also support the regional analysis: hardly any sites (≤ 1 site per
324 gradient) had significant dispersion patterns along environmental gradients when all trait axes were
325 considered together (Figure 2b). However, site-level analyses of TCAP reveal a more nuanced pattern.
326 Although no sites had significant TCAP over canopy cover gradients, several sites exhibited
327 significant TCAP over bromeliad detrital or water gradients (five and six sites respectively: Figure
328 2a). When analyses are restricted to individual trait axes, TCAP are revealed in further sites, although
329 the optimal trait axis sometimes differs between sites (contrary to hypothesis 1b). As a whole, the
330 above analyses therefore show that there are not geographically general patterns in trait convergence
331 or divergence driven by common traits.

332

333 Given that there is substantial difference between sites in the strength of TCAP patterns along
334 gradients in bromeliad detritus and water, we next tested whether these differences can be attributed
335 (hypothesis H2) to climate, biogeography, or site differences in sampling. In regression analyses, sites
336 north of the Northern Andes were more likely to have significant TCAP over detrital gradients (Table
337 S3). Indeed, all five sites where we detected trait convergence over gradients in detritus were located
338 in the Caribbean or Central America (Figure 2a). To confirm this effect of geography on TCAP, we
339 separated our sites into those north vs. south of the Northern Andes and re-ran all matrix analyses.
340 There was now an overall significant TCAP ($p=0.04$, $\rho = 0.18$) driven by trait axis 1 ($p=0.04$, ρ
341 $=0.25$) for the subset of sites north of the Andes, but not the subset of sites south of the Andes ($p =$
342 0.62 , $\rho=0.10$). Specifically, as detritus increased in northern sites, invertebrate communities shifted
343 from those dominated by detritivorous collectors and filter feeders to those dominated by sclerotized
344 and predatory invertebrates (Figure 3). By contrast, site differences in bioclimatic variables or
345 sampling characteristics (sample size, site mean of log detritus) did not affect either the significance
346 or strength of TCAP correlations in our regression analyses (Table S3).

347

348 Gradients in the amount of water in bromeliads were associated with convergence in community traits
349 in six sites. In four of the sites, this TCAP was driven by trait axis 3 (defence: Figure 4a), whereas in
350 the two other sites either trait axis 1 (trophic) or 2 (habitat) were more important and inverse patterns
351 with trait axis 3 were observed (Figure 4a). The sites with significant water-based TCAP were
352 geographically scattered, unlike detrital-based TCAP results. Although site differences in the strength
353 of the TCAP were not explained by biogeographic or sampling attributes of sites, they could be
354 related to bioclimatic differences between sites (Table S3). Specifically, the strength of the TCAP
355 correlations was greatest in sites with more seasonal precipitation (bioclimatic variable BC15),
356 supporting H2, although not every site with seasonal precipitation had strong TCAP correlations
357 (Figure 4b).

358

359

360 **Discussion**

361 In general, we found little evidence for geographically consistent patterns in trait assembly.
362 Overall patterns in TCAP and TDAP were not significant along any of our local environmental
363 gradients, contrary to hypothesis H1. Supporting our alternative hypothesis H2, we were able to
364 relate differences between sites in the strength of trait assembly patterns or dominant trait axis to
365 biogeography (position relative to northern extent of Andes) and climate (precipitation seasonality).
366 Our analysis of geographic variance in trait assembly patterns joins only a few other geographically-
367 extensive studies of animal communities along environmental gradients, including those of fish
368 (Lamouroux, Poff, & Angermeier, 2002; McLean et al., 2021), ants (Bishop et al., 2016; Gibb et al.,
369 2018), bees (Moretti, De Bello, Roberts, & Potts, 2009) and birds (Barnagaud et al., 2019; Matthews
370 et al., 2015). These studies show that geography can have a range of effects on community trait
371 filtering by local environments, from minor effects of geographic location (McLean et al., 2021) to
372 dominant effects of biogeography (Barnagaud et al., 2019) and bioclimatic context (Moretti et al.,
373 2009).

374

375 In our study, local gradients in amounts of detritus were correlated with a shift in community-
376 weighted traits in a number of sites, especially those north of the Northern Andes. In such sites, trait
377 axis 1 (trophic) often underlies this TCAP-detritus relationship. For example, when we examined trait
378 assembly patterns individually for each trait axis, only trait axis 1 resulted in significant TCAP
379 patterns for most (five out of these eight) northern sites. This shift in trait axis 1 with increasing
380 bromeliad detritus represents a community shift from dipteran detritivores to hard-bodied predators
381 like dytiscid beetles and damselflies. There are several potential reasons for this shift in trophic traits.
382 The first reason is a type of trophic sampling effect. Invertebrate abundance usually correlates
383 positively with detrital quantity in tropical freshwaters (Benstead, Road, & El, 2010), and this is also
384 true for the bromeliad fauna (Richardson, 1999; Diane S. Srivastava, Trzcinski, Richardson, &
385 Gilbert, 2008). As predators have lower regional population sizes than detritivores, even random
386 assembly would lead to the ratio of predator to detritivore taxa increasing with the amount of detritus
387 (Diane S. Srivastava et al., 2008). A second possible reason is the energetic inefficiencies in trophic

388 transfer (Lindeman, 1942), leading to bottom-up limitation of the occurrence of large-bodied
389 predators when detritus is limiting (Wallace, Eggert, Meyer, & Webster, 1999). This hypothesis has
390 been tested by measuring the growth rate of predatory damselflies following relocation to bromeliads
391 with less detritus, and was not supported (Diane S. Srivastava, Ware, Ngai, Starzomski, & Amundrud,
392 2020). However, energetic limitation of other predatory taxa in this system remains unexplored. A
393 third possible reason is the often high covariance of bromeliad detritus with bromeliad capacity within
394 habitats (Diane S. Srivastava et al., 2008), coupled with the restriction of some large, long-lived
395 predators — such as the damselfly *Mecistogaster modesta* in Central America bromeliads — to high
396 capacity bromeliads that rarely dry out (Amundrud & Srivastava, 2015; Diane S. Srivastava et al.,
397 2020). Notably, both sampling effects and covariance with capacity are expected to influence trophic
398 structure most when the detrital gradient includes small bromeliads where the stochastic effects of
399 sampling or risk of drought are highest. Such bromeliads characterize the rainforests of Central
400 America and the Caribbean, and are in Tillandsioideae genera such as *Guzmania*, *Tillandsia*, *Vriesea*
401 and *Werauhia*. By contrast, phytotelm bromeliads south of the Andes are more often high-capacity
402 CAM species in Bromelioideae (Fig. S1, Table S1). This distinction reflects the historical
403 biogeography of bromeliads. The Tillandsioideae subfamily colonized the areas north of the Andes ca
404 8.7-14.2 million years ago, coinciding with rapid uplifting of the Northern Andes, whereas the
405 Bromelioideae subfamily evolved more recently, ca 5.5 million years ago, and radiated in the areas
406 south and east of the Andes (Givnish et al., 2011). Thus, the biogeography of the host plant may
407 contribute to the geographic signal in invertebrate trait-based assembly patterns.

408

409 The strength of TCAP over gradients in bromeliad water also varied substantially between sites, with
410 the strongest TCAP in sites with high seasonality in precipitation. A similar bromeliad volume by
411 precipitation seasonality interaction was reported for invertebrate traits by Guzman et al. (2020) using
412 a different analytical method, pointing to the robustness of this conclusion. In sites with strongly
413 seasonal precipitation, bromeliads are likely to dry out completely during the dry season, posing
414 substantial challenges to their aquatic fauna (Céréghino et al., 2020; Dézerald, Céréghino, Corbara,
415 Dejean, & Leroy, 2015). Therefore, we would expect the strength of filtering by bromeliad water

416 volume to be greater in such seasonal sites. In seasonal sites, trait axis 3 (defence) was often
417 important: bromeliads with low water volume were dominated by small, soft-bodied benthic
418 invertebrates with simple life cycles, like leeches and oligochaete worms, whereas high water volume
419 plants included sclerotized, surface-swimming insects like predacious diving beetles (Dytiscidae) and
420 riffle bugs (Veliidae). This suggests that in seasonal sites, invertebrate communities persist even at
421 low water through population resistance (simple life cycles in benthos), similar to the conclusion from
422 a study that experimentally excluded rainfall from bromeliads for up to three months (Bonhomme et
423 al., 2021). By contrast, in regions with less seasonal rainfall, TCAP was underlain by a variety of
424 different trait axes and showed a reverse pattern on trait axis 3. Under less seasonal rainfall, short-
425 term desiccation resistance of organisms (i.e. LD₅₀) may be more important for surviving fluctuating
426 water levels (Amundrud & Srivastava, 2015). Desiccation LD₅₀ is well predicted by traits like cuticle
427 thickness and body mass (Céréghino et al., 2020) that fall at the negative end of trait axis 1.
428 Invertebrate communities in low seasonality sites were also often positioned lower on trait axis 3, a
429 pattern that may be related to the absence of odonates from the species pool of the three Caribbean
430 islands we studied. In general, our findings complement studies from other freshwater systems that
431 show strong filtering effects of drought on the traits of aquatic invertebrate communities (Aspin et al.,
432 2019; Datry et al., 2014).

433

434 Given that the amount of detritus and water in bromeliads influences the trait composition of
435 invertebrates, at least in some sites, we can then ask: what underlies this variation between bromeliads
436 in water and detritus? Both bromeliad morphology (i.e. size and shape) and canopy cover are
437 important determinants of bromeliad detritus and water (Farjalla et al., 2016; Zotz, Leja, Aguilar-
438 Cruz, & Einzmann, 2020); however, canopy cover was not directly found to be a driver of trait-based
439 assembly patterns. This suggests a large role for the bromeliad plant itself in mediating the trait-based
440 assembly of the invertebrates it contains. Bromeliads are a foundation species for the invertebrate
441 community we study, much in the same way that kelp, corals and seagrass form critical structural
442 habitat for marine species or dominant tree species provide key microhabitats for forest species
443 (Ellison, 2019). Foundation species can have profound effects on the species that rely on them, even

444 though these interactions are non-trophic in nature (Ellison, 2019; Kéfi et al., 2012). Such interactions
445 between animal traits and plant traits have largely been overlooked in tests of trait assembly patterns
446 (but see Abgrall et al., 2017; Pakeman & Stockan, 2014). Future studies could incorporate
447 intraspecific trait variation of both the bromeliads and the invertebrates in such analyses of animal-
448 plant interactions. Certainly, one limitation of the current study was that trait data was restricted to the
449 species level or higher.

450

451 In summary, we show here that, while trait-assembly patterns can exist at the local level, these
452 patterns are not often constant over large geographic areas. There are several reasons for this
453 geographic variation. First, the effect of the environmental gradients on ecological communities may
454 depend on the regional climate context: hydrologic dynamics may be qualitatively different in regions
455 with strong seasonal variation in precipitation (Boersma, Bogan, Henrichs, & Lytle, 2014). Second,
456 trait assembly patterns may be affected by the historical biogeography of species, including both the
457 invertebrates and the bromeliads. While biogeographic effects on community trait patterns have been
458 shown before in terms of the species pool of the focal animal community (Barnagaud et al., 2019;
459 Gorczynski et al., 2021; Whittaker et al., 2014), here we show biogeographic effects also operate via
460 the plants that form the structural habitat for the animal communities. This suggests that the future
461 development of animal functional ecology should not occur in isolation from plant functional ecology,
462 given the myriad of ways that plants and animals influence each other via their functional traits. The
463 geographic contingency that we found in trait-assembly patterns does not necessarily invalidate the
464 potential of developing a mechanistic trait-based theory of community ecology applicable over large
465 regions of the globe. However, our results suggest that such a theory must include processes operating
466 at larger spatial and temporal scales, such as the effects of dispersal limitation and evolutionary
467 history (Violle et al., 2014), as well as often overlooked effects of non-trophic interactions between
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497 contributed to conceptualization, data collection and manuscript writing. AAMM led software
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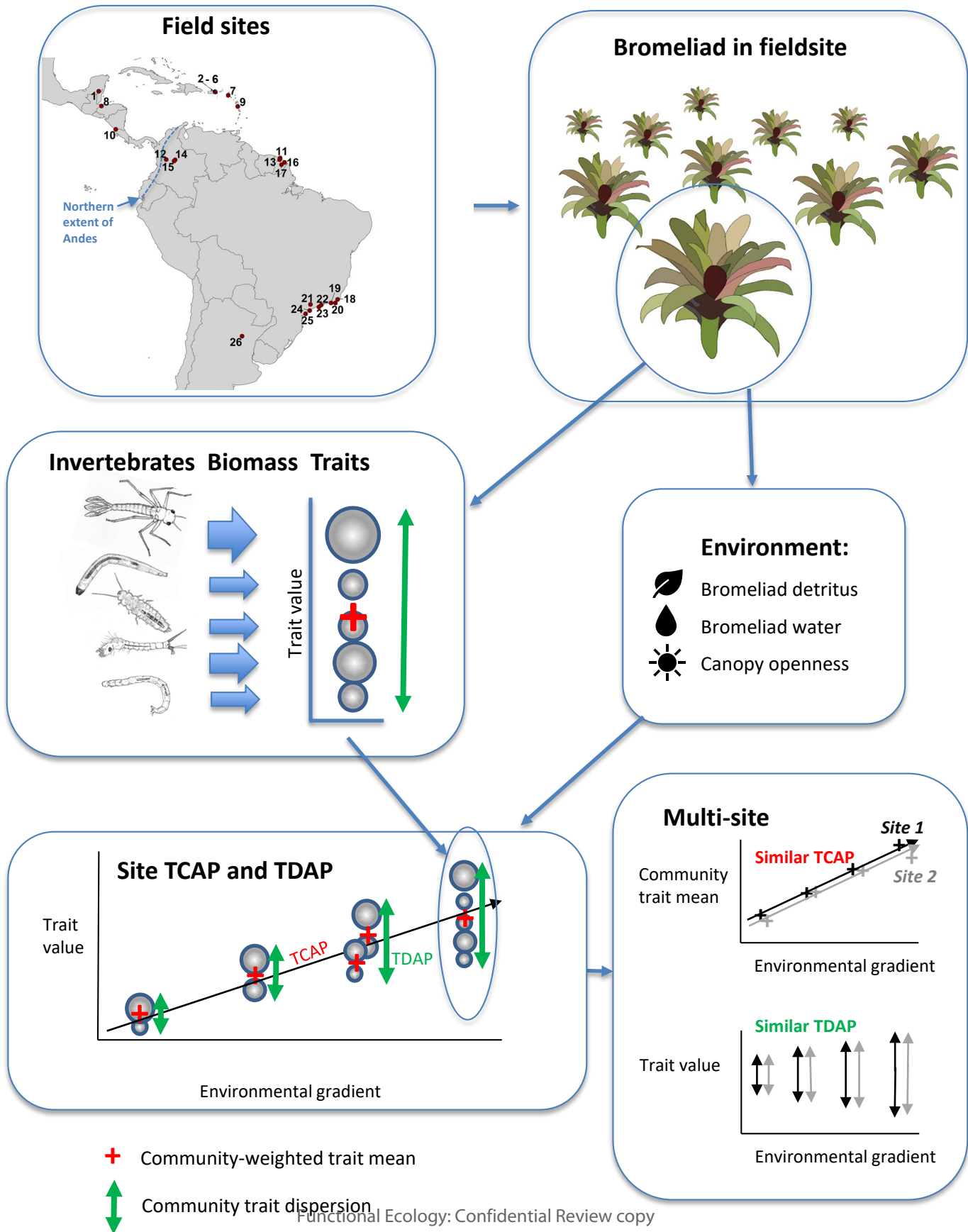
712 **Figure legends**

713 **Figure 1.** Study design for testing the geographic consistency of trait assembly patterns. Bromeliads
714 were sampled in 26 sites throughout the Neotropics (“Field sites” panel). In each site, bromeliads
715 were sampled (“Bromeliad in field site” panel) for aquatic macroinvertebrates along environmental
716 gradients (“Environment” panel). For each bromeliad, the traits of each invertebrate taxon were
717 weighted by the taxon’s total biomass (size of spheres and arrows in “Invertebrates Biomass Traits”
718 panel) to determine the community weighted mean trait value (red cross) and the dispersion of the
719 traits around this mean value (green arrow). A trait-based convergence assembly pattern (TCAP)
720 occurred when the community-weighted mean changed consistently over environmental gradients; a
721 trait-based divergence assembly pattern (TDAP) when dispersion changed consistently over
722 environmental gradients (“Site TCAP and TDAP” panel). Note that TDAP includes both intertaxa
723 variation in traits as well as variation in trait means between communities at similar points on the
724 environmental gradient. Finally, TCAP and TDAP were compared among sites to determine if trait
725 assembly patterns were geographically similar (“Multi-site” panel) or geographically different (not
726 illustrated).

727 **Figure 2.** (A) Trait-based convergence assembly patterns (TCAP) of bromeliad macroinvertebrate
728 communities along three local environmental gradients: detrital biomass (“detritus”, on a log scale) or
729 standing water volume in bromeliads (“water”, on a log scale), and openness of the canopy above
730 bromeliads (“canopy”). Sites are ordered from north to south, with red arrows indicating the northern
731 extent of the Andes. Sites with no data for a particular environmental gradient are indicated with
732 white space. The strength of TCAP along each environmental gradient is evaluated with Procrustes
733 correlations, and is based either on the four trait axes combined (bar graph) or assessed individually
734 (symbols; where multiple trait axes are significant, the order of symbols progresses left to right from
735 strongest correlation to weakest). (B) Trait convergence assembly patterns (TDAP) of bromeliad
736 macroinvertebrate communities along local environmental gradients, labelled and ordered as in (A).

737 **Figure 3.** The community weighted mean value of trait axis 1 (associated with trophic position) in
738 each bromeliad is plotted as a function of the mass (log grams, without scaling or centering) of
739 detritus in the bromeliad for each site, allowing a visualization of this particular TCAP-environmental
740 correlation. Trendlines are shown simply to aid visualization of site-specific patterns; trendlines
741 representing TCAP that were found by SYNCSEA to be significant for trait axis 1 are shown with solid
742 lines, otherwise the lines are dashed. Sites are divided into those north (sites 1-10) and south (sites 11-
743 26) of the northern extent of the Andes.

744 **Figure 4.** (a) The community weighted mean value of trait axis 3 (associated with defence traits) in
745 each bromeliad is plotted as a function of the volume (log ml, without scaling or centering) of water in
746 the bromeliad for each site, allowing a visualization of this particular TCAP-environmental
747 correlation. Trendlines are shown simply to aid visualization of site-specific patterns; trendlines
748 representing TCAP that were found by SYNCSEA to be significant for trait axis 3 are shown with solid
749 lines, otherwise the lines are dashed. Sites are divided into those with low ($\text{bio15} \leq 40$) and high
750 ($\text{bio15} > 40$) seasonality in precipitation as indicated in (b). (b) When considering all four trait axes,
751 the strength of TCAP correlations increases with seasonality in precipitation, as represented by the
752 bioclimatic variable bio15, with weak TCAP in low seasonality sites and often stronger TCAP in high
753 seasonality sites. The predicted effect of seasonality was extracted as the partial residuals from a
754 multiple regression model, conditioned on the mean effects of other site-level explanatory variables
755 (Table S3). The shaded area represents the 95% CI around the predicted effect.



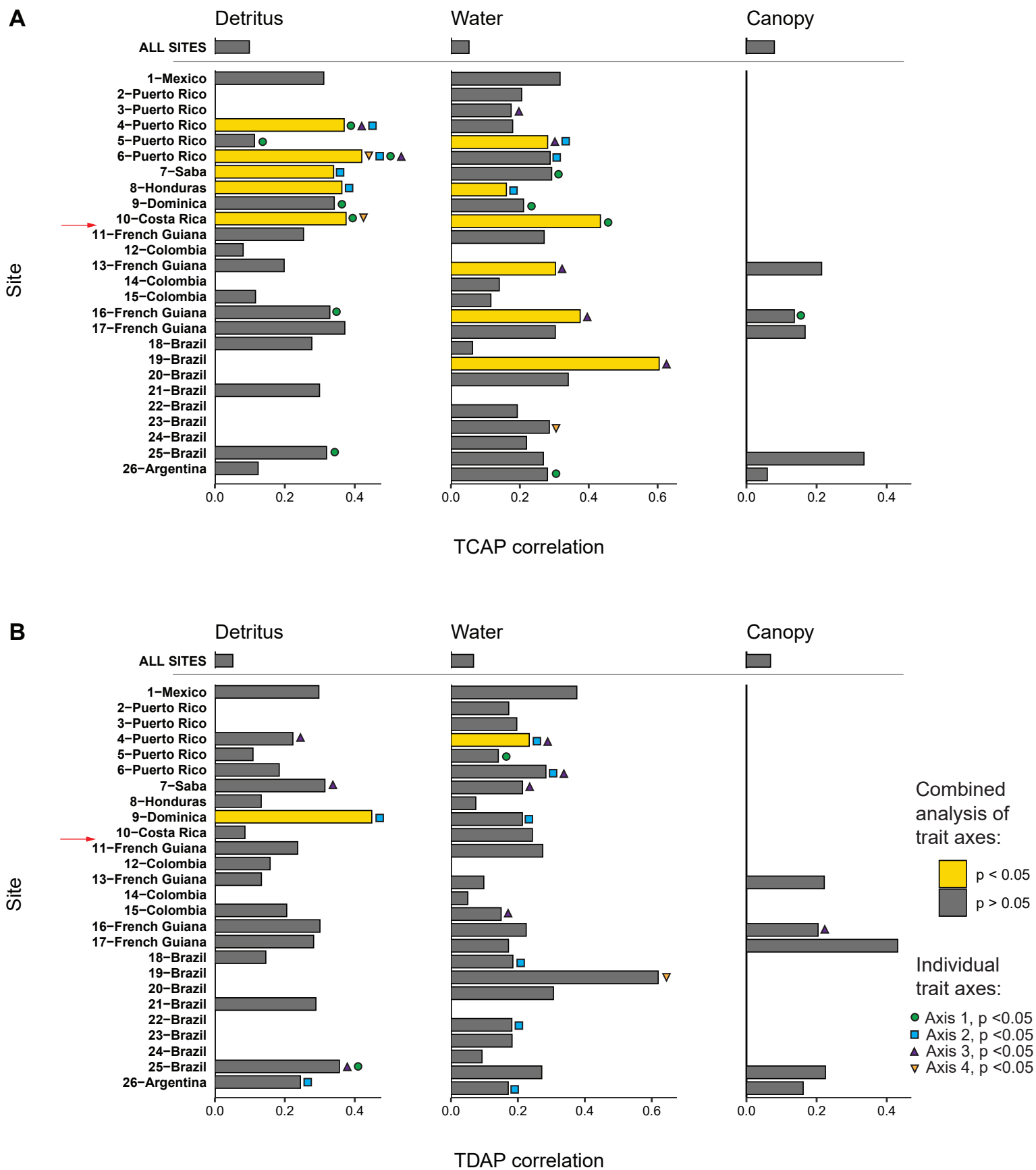


Figure 2

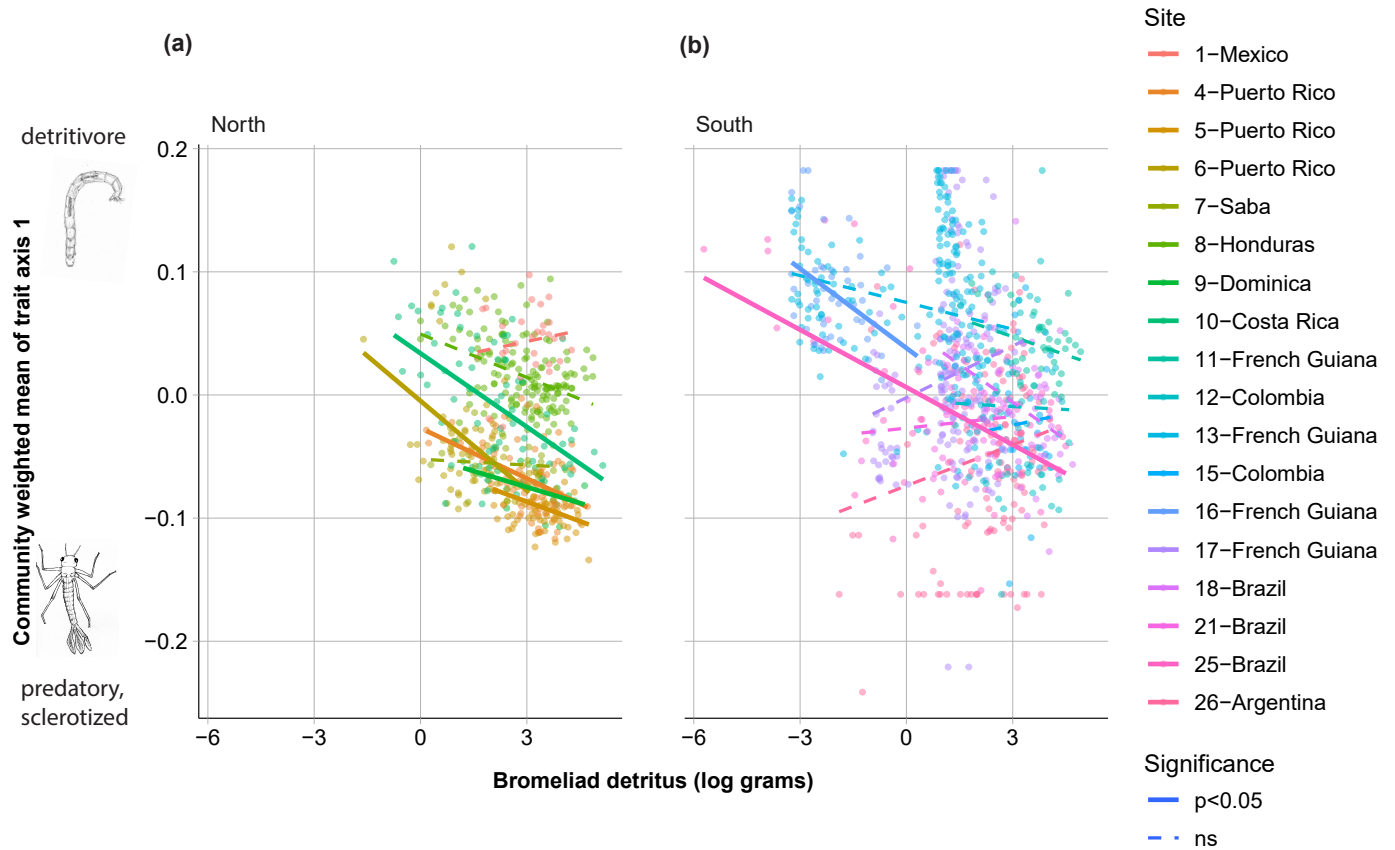
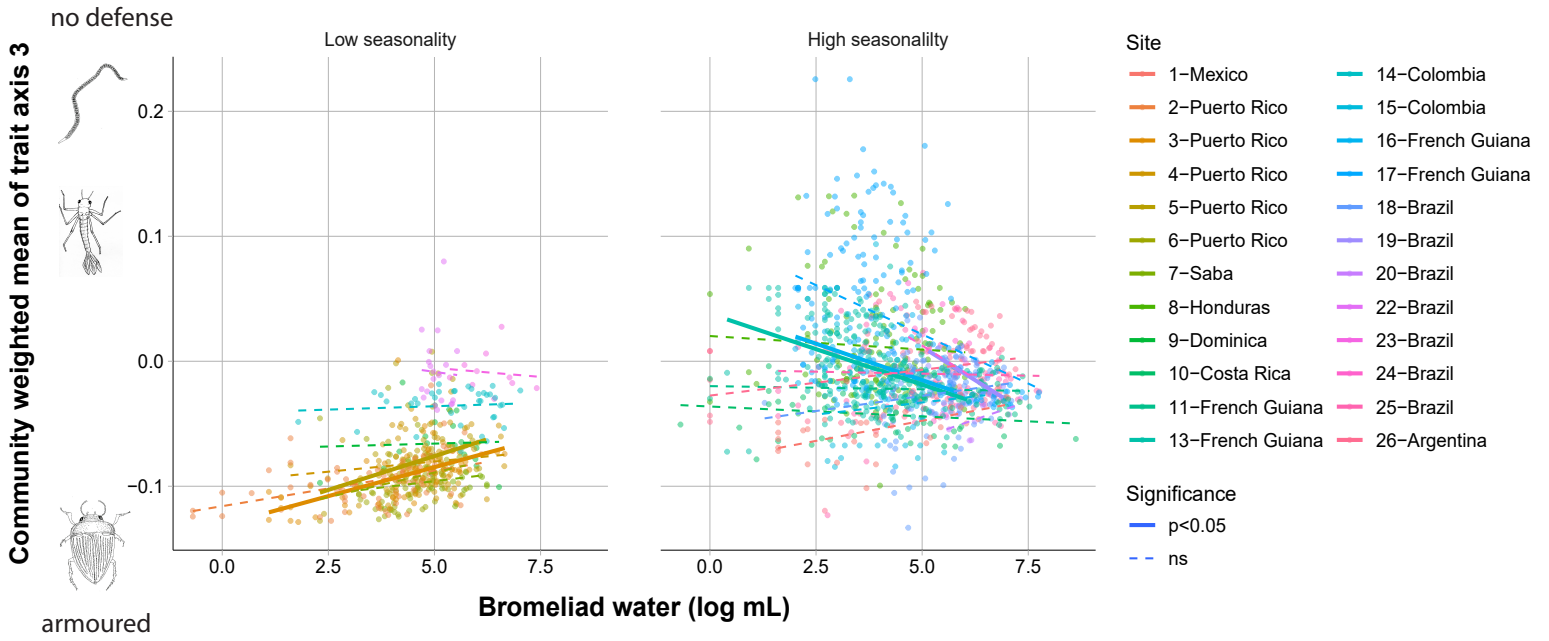


Figure 3

(a)



(b)

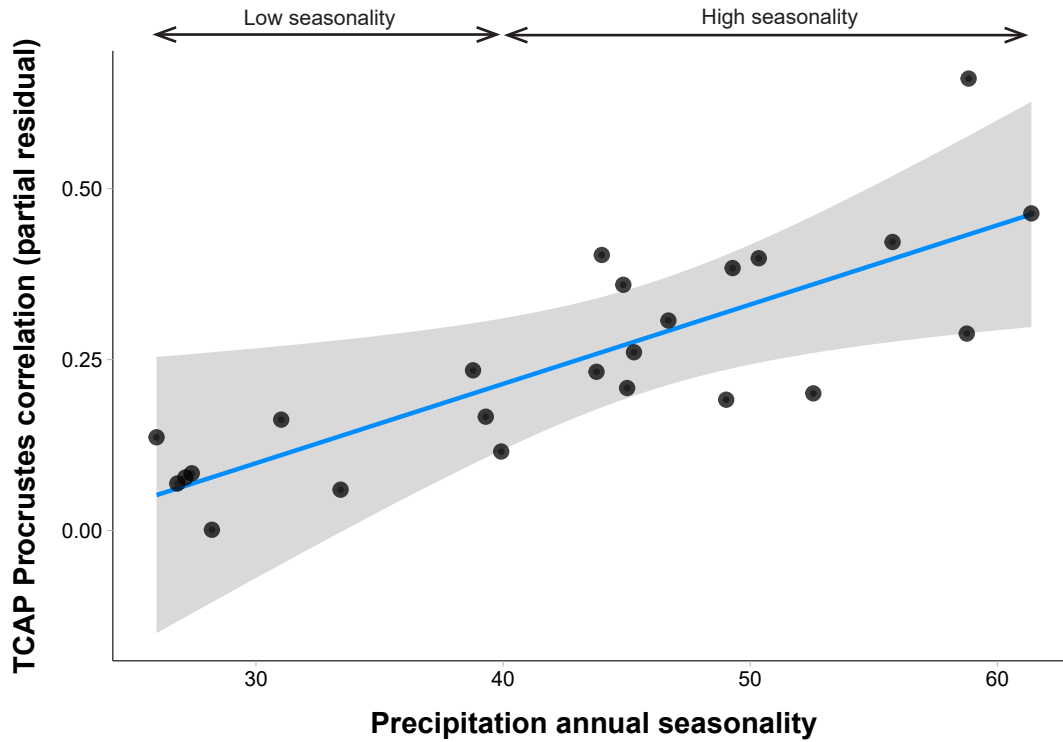


Figure 4

Geographical variation in the trait-based assembly patterns of multitrophic invertebrate communities

Diane S. Srivastava, A. Andrew M. MacDonald, Valério D. Pillar, Pavel Kratina, Vanderlei J. Debastiani, Laura Melissa Guzman, M. Kurtis Trzcinski, Olivier Dézerald, Ignacio M. Barberis, Paula M. de Omena, Gustavo Q. Romero, Fabiola Ospina Bautista, Nicholas A. C. Marino, Céline Leroy, Vinicius F. Farjalla, Barbara A. Richardson, Ana Z. Gonçalves, Bruno Corbara, Jana S. Petermann, Michael J. Richardson, Michael C. Melnychuk, Merlijn Jocqué, Jacqueline T. Ngai, Stanislas Talaga, Gustavo C. O. Piccoli, Guillermo Montero, Kathryn R. Kirby, Brian M. Starzomski, Régis Céréghino

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Figure S2. A replotting of Figure 3 from Céréghino et al. 2018.

Appendix S1

Supplemental methods. We defined a distinct “site” as one that was distinct in terms of: space, elevation and taxonomic identity of its species pool. All bromeliads surveyed within a site must occur in an area <3.5 km in linear dimension and within a 300 m range of elevation. The invertebrate species pool should also be >50% dissimilar (Jaccard dissimilarity, based on presence-absence) between adjacent sites. For example, bromeliads surveyed along a transect covering an elevation gradient (Sonadora transect, Puerto Rico) are assigned to a low elevation (400-700m) and high elevation (750-1000m) site based on differences in elevation and dissimilarity in the invertebrate species.

Table S1. (a) Site locations and characteristics

Site code	Field site name and location (latitude, longitude)	Sample size (no. bromeliads)	Species pool diversity (Chao estimate \pm SE)	Bromeliad genera (B = Bromelioideae subfamily, T = Tillandsioideae)
1 Mexico	Kohunlich, Quintana Roo (18.427, -88.804)	29	18.1 \pm 0.4	<i>Aechmea bracteata</i> (B)
2 Puerto Rico	Luquillo Experimental Forest, Sonadora transect (400-700m) (18.321, -65.817)	67	36.9 \pm 6.4	<i>Guzmania</i> spp.(T)
3 Puerto Rico	Luquillo Experimental Forest, Sonadora transect (750-1000m) (18.315, -65.798)	58	37.9 \pm 6.4	<i>Guzmania</i> and <i>Werauhia</i> spp.(T)
4 Puerto Rico	El Verde Biological Station, Tabonuco forest (<600m) (18.321, -65.817)	79	70.8 \pm 23	<i>Guzmania lingulata</i> , <i>G. berteroniana</i> (T)
5 Puerto Rico	El Verde Biological Station, Palo Colorado forest (600-900m) (18.296, -65.798)	70	53.2 \pm 6.8	<i>Guzmania lingulata</i> , <i>G. berteroniana</i> (T)
6 Puerto Rico	Pico del Este, Dwarf forest (900-1050m) (18.265, -65.760)	70	27.5 \pm 2.9	<i>Guzmania berteroniana</i> , <i>Werauhia sintenisii</i> (T)
7 Saba	Saba 650-840m (17.638, -63.235)	20	23.8 \pm 5.0	<i>Tillandsia utriculata</i> , <i>Guzmania plumieri</i> , <i>Vriesea antillana</i> , <i>Werauhia ringens</i> (T)
8 Honduras	Cusuco National Park (15.542, -88.264)	155	39.5 \pm 7.1	<i>Tillandsia guatemalensis</i> (T)

9	Dominica	Dominica 800-830m (15.362, -61.325)	20	54.4 ± 37.6	<i>Guzmania plumieri</i> , <i>G. megastachya</i> (T)
10	Costa Rica	Pitilla (10.983, -85.433)	73	50.4 ± 10.1	<i>Guzmania</i> and <i>Werauhia</i> spp. (T)
11	French Guiana	Sinnamary (4.083, -52.683)	39	24.9 ± 3.7	<i>Aechmea aquilega</i> (B)
12	Colombia	Rio Blanco (5.047, -74.579)	51	30.8 ± 11.4	<i>Guzmania multiflora</i> (T)
13	French Guiana	PetitSaut (4.736, -52.925)	199	59.0 ± 10.1	<i>Vriesea</i> sp (T), <i>Aechmea mertensii</i> (B)
14	Colombia	Sisga (5.017, -73.700)	37	21.5 ± 1.3	<i>Tillandsia turneri</i> (T)
15	Colombia	Guasca (4.767, -74.167)	35	28.1 ± 16.7	<i>Tillandsia turneri</i> , <i>T. complanata</i> (T)
16	French Guiana	Kaw (4.083, -52.683)	47	23.1 ± 4.8	<i>Aechmea mertensii</i> (B)
17	French Guiana	Nouragues (4.083, -52.683)	171	64.6 ± 6.8	<i>Catopsis berteroniana</i> , <i>Guzmania lingulata</i> , , <i>Vriesea pleiosticha</i> , <i>Vriesea</i> sp. (T), <i>Aechmea melinonii</i> , <i>A. aquilega</i> (B)
18	Brazil	Macaé (-22.216, -41.507)	133	65.5 ± 4.2	<i>Aechmea nudicaulis</i> , <i>A. lingulata</i> , <i>Neoregelia cruenta</i> (B), <i>Vriesea neoglutinosa</i> (T)
19	Brazil	Marica (-22.947, -42.888)	10	32.1 ± 9.2	<i>Neoregelia</i> sp. (B)
20	Brazil	Arraial do Cabo (-22.951, -42.043)	10	15.4 ± 1.2	<i>Neoregelia</i> sp. (B)
21	Brazil	Serra do Japi (-23.233, -46.967)	29	37.7 ± 7.3	<i>Vriesea carinata</i> (T)

22 Brazil	Picinguaba (-23.339, -44.781)	20	44.4 ± 11.9	<i>Vriesea procera</i> (T)
23 Brazil	Ilha Bella (-23.738, -45.291)	10	31.2 ± 6.8	<i>Neoregelia</i> sp. (B)
24 Brazil	Jureia (-24.431, -47.109)	20	94.4 ± 18.1	<i>Vriesea procera</i> (T)
25 Brazil	Cardoso (-25.084, -47.940)	66	119.5 ± 18.7	<i>Quesnelia arvensis</i> (B)
26 Argentina	Las Gamas (-29.433, -60.467)	137	57.3 ± 30.9	<i>Aechmea distichantha</i> (B)

Table S1. (b) Sample years and research permit information. NA = not applicable.

Site code	Years sampled	Permit organization	Permit number	Permit holder
1 Mexico	2011	No permit required at the time of sampling	NA	NA
2 Puerto Rico	2004	No permit required at time of sampling for Long Term Ecological Research (LTER) research areas of Luquillo Experimental Forest (LEF)	NA	NA
3 Puerto Rico	2004	No permit required at time of sampling for LTER research areas of LEF	NA	NA
4 Puerto Rico	1993,1994, 1996,1997, 2010	No permit required at time of sampling for LTER research areas of LEF	NA	NA
5 Puerto Rico	1993,1994, 1996,1997, 2010	No permit required at time of sampling for LTER research areas of LEF	NA	NA

6 Puerto Rico	1993,1994, 1996, 2010	No permit required at time of sampling for LTER research areas of LEF	NA	NA
7 Saba	2009	Executive Council of the Island Territory of Saba, Netherlands Antilles	1218/09	Barbara A. Richardson
8 Honduras	2006, 2007	La Administración Forestal del Estado – Corporación Hondureña de Desarrollo Forestal (AFE- COHDEFOR)	#7691	Merlijn Jocque, Jose Nuñez Miño
9 Dominica	2002	Dominica Dept. of Forestry and Wildlife	Permit not numbered	Barbara A. Richardson
10 Costa Rica	1997, 2000, 2002, 2004, 2010	1997-2004:Ministerio del Ambiente y Energia (MINAE); 2010: Ministerio del Ambiente, Energia y Telecomunicaciones (MINAET), Costa Rica	163-97-OFAU, 243-2000-OFAU, 296-2002-OFAU, ACG-PI-023-2004, ACG-PI-028-2010	Diane S. Srivastava 1997-2004, Jana S. Petermann 2010
11 French Guiana	2011	No permit required at the time of sampling	NA	NA
12 Colombia	2012, 2014	No permit required at the time of sampling	NA	NA
13 French Guiana	2014	No permit required at the time of sampling	NA	NA
14 Colombia	2000	No permit required at the time of sampling for private farms	NA	NA
15 Colombia	2001	No permit required at the time of sampling for private farms	NA	NA
16 French Guiana	2008	No permit required at the time of sampling	NA	NA
17 French Guiana	2006	No permit required at the time of sampling	NA	NA

18 Brazil	2008, 2015	Ministério do Meio Ambiente (MMA), Instituto Chico Mendes de Conservação da Biodiversidade (ICMBio) Sistema de Autorização e Informação em Biodiversidade(SISBIO)	2008 permit not numbered, 2015 permit: 47164-1	Vinicius Fortes Farjalla, L. Melissa Guzman Uribe
19 Brazil	2015	MMA-ICMBio - SISBIO	47164-1	L. Melissa Guzman Uribe
20 Brazil	2015	MMA-ICMBio - SISBIO	47164-1	L. Melissa Guzman Uribe
21 Brazil	2011	MMA-ICMBio - SISBIO	23689-1	Gustavo Quevedo Romero
22 Brazil	2011	Secretaria do Meio Ambiente (SMA) and MMA	12.429/2011 and 29738-3	Ana Zangirolame Gonçalves
23 Brazil	2015	MMA-ICMBio - SISBIO	47164-1	L. Melissa Guzman Uribe
24 Brazil	2013	SMA and MMA	12.429/2011 and 29738-3	Ana Zangirolame Gonçalves
25 Brazil	2008, 2011	MMA-ICMBio – SISBIO, Instituto Forestal	23689-1, COTEC N° 804/2011 D85/07 M	Gustavo Quevedo Romero
26 Argentina	2004-2005, 2010, 2012, 2013	"Specific Collaboration Agreement" between the Faculty of Agricultural Sciences of the National University of Rosario and the Ministry of Production of the Province of Santa Fe, Argentina	Not numbered	Guillermo Montero

Table S2. Correlation of trait modalities with the first four axes of a PCA analysis of 64 traits of bromeliad macroinvertebrates. Only traits with strong negative or positive correlations with the axis are listed, with the correlation coefficient in parentheses; a full list of the 64 traits is provided in Table 1 of Cereghino et al. (2018). The 64 traits in the PCA include those that characterize aquatic stage (traits starting with “AQ”), body form (“BF”), cohort production interval (“CPI”), dispersal mode (“DM”), food (“FD”), feeding group (“FG”), locomotion (“LO”), morphological defence (“MD”), reproduction (“RE”), resistance form (“RF”), respiration mode (“RM”), as well as traits on maximum body size.

	Trait Axis 1	Trait Axis 2	Trait Axis 3	Trait Axis 4
% inertia	15.1	12.5	9.9	7.6
Interpretation	Trophic: predators -> detritivores	Habitat: pelagic -> benthic	Defence: armoured -> none	Life cycle: simple -> complex
Traits negatively correlated with axis ($r \leq -0.5$)	RF4: No resting stage (-0.67) BF2: Flat ovoid body (-0.57) MD6: Sclerotized exoskeleton (-0.54) LO1: Flier (-0.50)	RM4: Siphon/spiracle respiration (-0.69) DM2: Active dispersal (-0.65)	MD4: Sclerotized spines (0.75) LO2: Surface swimmer (0.63) MD5: Dorsal plates (0.59) RE2: Isolated eggs, free (0.52)	LO3: Full water swimmer (0.60) AS4: Adults aquatic (0.59) BF1: Flat elongate (0.58)
Traits positively correlated with axis ($r \geq 0.5$)	FD4: Consumes living microphytes (0.75) AS3: Nymphs aquatic (0.73) RF3: Diapause or dormancy (0.70) CP1: Cohort production interval < 21 days (0.69) FG1: Deposit feeder (0.61) MD3: Hairs on body (0.59) FD1: Consumes micro-organisms	LO7: tube builder (0.81) MD8: case or tube defence (0.81) RE8: Asexual reproduction (0.72) LO6: Interstitial use (0.72) RM1: Integument respiration (0.72) FG3: Scraper (0.64) DM1: Passive dispersal (0.54)	No correlations less than -0.42	FD3: Consumes dead plant litter (-0.67) FG2: Shredder (-0.51)

	(0.58) FG4: Filter-feeder (0.52) FD2: Consumes detritus <1 mm (0.51)			
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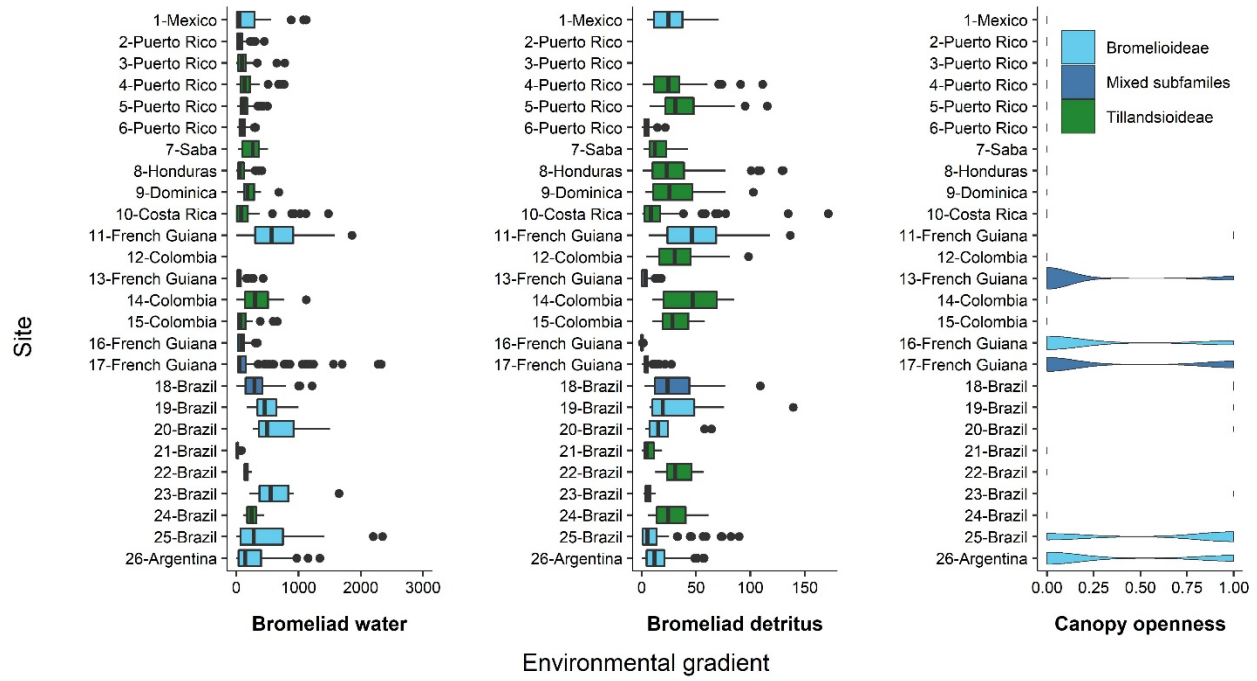
Table S3: Regression analyses of potential drivers of site variation in Procrustes correlations between TCAP along gradients in either bromeliad detritus or bromeliad water volume. For each environmental gradient, we built models that explain either the strength (“size of correlation coefficient”) or significance (“log(p value of correlation)”) of the TCAP Procrustes correlation. Bioclimatic variables include BC2 (mean diurnal range temperature), BC4 (temperature seasonality), BC15 (precipitation seasonality), and BC17 (precipitation of driest quarter). The site mean in the environmental gradient refers to either the site mean of log detritus per bromeliad (units = log grams) as an explanatory variable for Procrustes correlation for detritus-based TCAP, or the site mean of log water volume per bromeliad (units = log ml) as an explanatory variable for Procrustes correlation for water-based TCAP. Significance (“p”) of potential drivers is evaluated using type II ANOVAs (likelihood ratio = “LR”) which conservatively evaluate the partial effects of explanatory variables. * = significant variables (p<0.05)

	Procrustes correlation for detritus-based TCAP				Procrustes correlation for water-based TCAP			
	Size of correlation coefficient		Log(p value of correlation)		Size of correlation coefficient		Log(p value of correlation)	
	LR χ^2	p	LR χ^2	p	LR χ^2	p	LR χ^2	p
Sample size	0.00	0.96	0.16	0.69	0.00	0.99	0.66	0.42
Site mean in the environmental gradient	1.02	0.31	0.47	0.49	0.34	0.56	0.04	0.85
BC2	0.66	0.42	0.13	0.72	1.75	0.19	0.61	0.44
BC4	0.19	0.67	0.00	0.99	2.00	0.16	0.29	0.59
BC15	1.03	0.31	0.05	0.83	5.92	0.015*	1.45	0.23
BC17	0.28	0.60	0.02	0.89	3.18	0.07	0.62	0.43
Species pool (Chao)	0.04	0.83	0.00	0.99	1.46	0.22	0.04	0.84
North or south of Andes (binary)	6.20	0.01*	6.20	0.01*	1.44	0.23	0.29	0.59

Figure S1. Distribution of bromeliads in each site along the studied environmental gradients in either (a) a linear scale, in the case of bromeliad water and bromeliad detritus, or (b) log transformed scale. TCAP and TDAP analyses were conducted with a linear scale for canopy cover and a log scale for bromeliad water and detritus. Canopy cover is binary data, and so its distribution is displayed with a violin plot, where the shape represents the estimated density in either open (1) or closed (0) canopy. Bromeliad water and detritus are continuous data, and so their distribution is displayed with box plots, with the box bounded by the first and third quartiles, the median indicated with the central line in the box, and whiskers and dots indicating estimated range and potential outliers respectively.

Figure S2. A replotting of Figure 3 from Céréghino et al. 2018 (doi: 10.1111/1365-2435.13141) using updated matrices. Only traits with an absolute correlation ≥ 0.5 with one of the PCA axes are included in the biplot, and the alphanumeric trait codes for these traits are provided in Table S2. Species positions are shown on biplots, and the taxonomic affinity of species is indicated by different symbols.

(a)



(b)

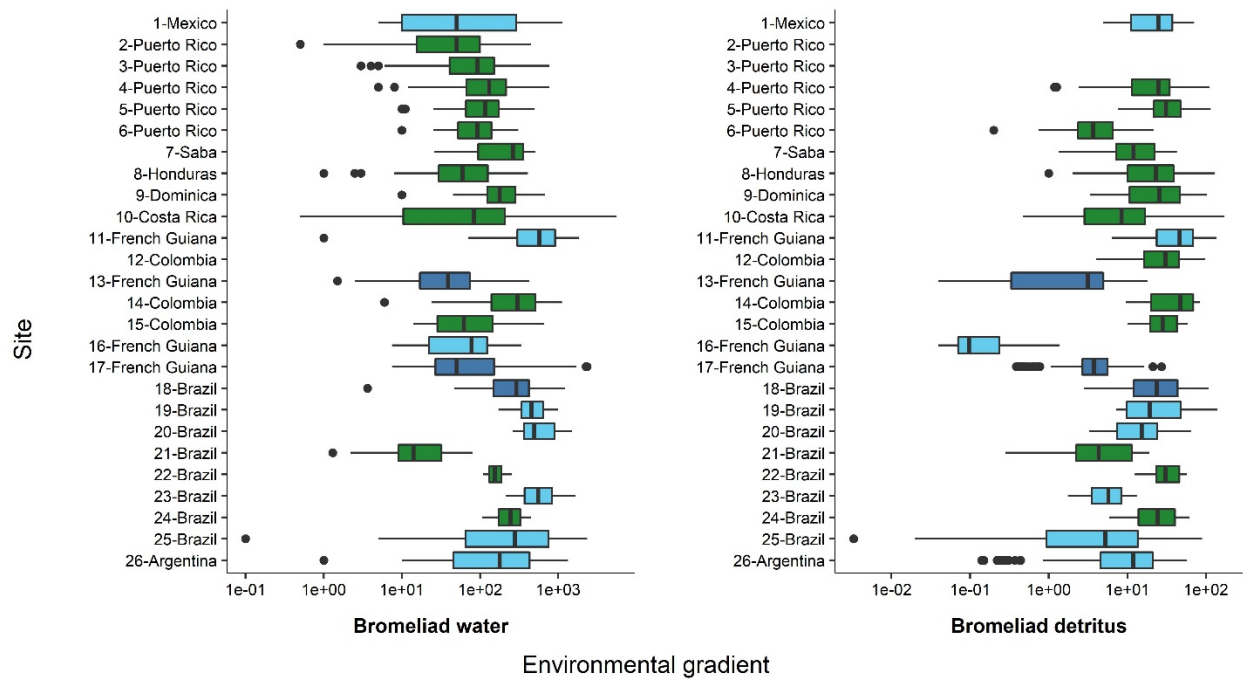


Figure S1.

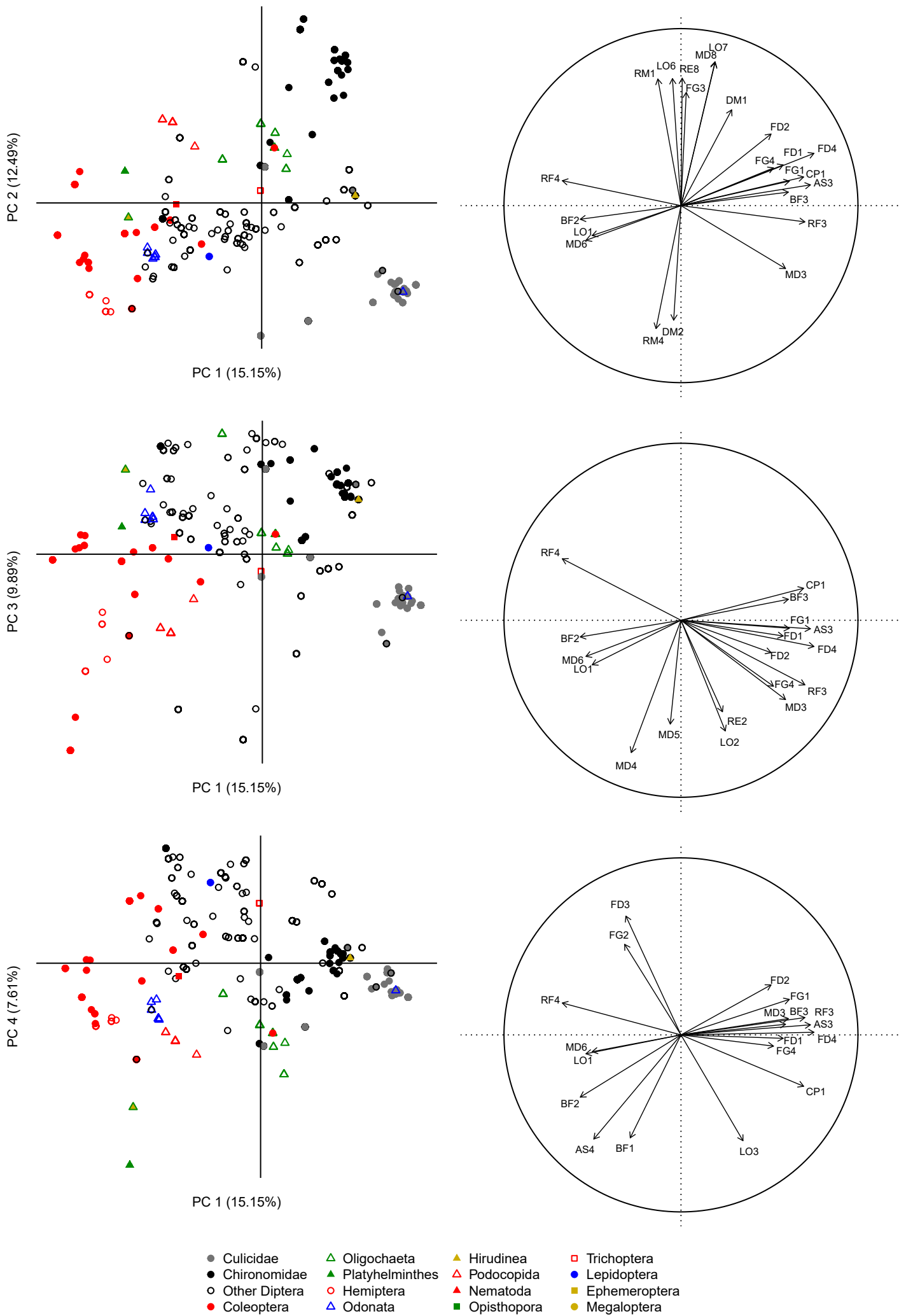


Figure S2