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# Complex Interaction of Dendritic Connectivity and Hierarchical Patch Size on Biodiversity in River-Like Landscapes

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**ABSTRACT:** Habitat fragmentation and land use changes are causing major biodiversity losses. Connectivity of the landscape or environmental conditions alone can shape biodiversity patterns. In nature, however, local habitat characteristics are often intrinsically linked to a specific connectivity. Such a link is evident in riverine ecosystems, where hierarchical dendritic structures command related scaling on habitat capacity. We experimentally disentangled the effect of local habitat capacity (i.e., the patch size) and dendritic connectivity on biodiversity in aquatic microcosm metacommunities by suitably arranging patch sizes within river-like networks. Overall, more connected communities that occupy a central position in the network exhibited higher species richness, irrespective of patch size arrangement. High regional evenness in community composition was found only in landscapes preserving geomorphological scaling properties of patch sizes. In these landscapes, some of the rarer species sustained regionally more abundant populations better tracking their own niche requirements compared to landscapes with homogeneous patch size or landscapes with spatially uncorrelated patch size. Our analysis suggests that altering the natural link between dendritic connectivity and patch size strongly affects community composition and population persistence at multiple scales. The experimental results are demonstrating a principle that can be tested in theoretical metacommunity models and eventually be projected to real riverine ecosystems.

**Keywords:** dendritic ecological networks, riverine ecosystems, community assembly, directional dispersal, experimental microcosms, spatial heterogeneity.

## Introduction

Biodiversity has been strongly affected by humans over the last decades. In many systems, diversity has been declining at all levels, with major consequences on ecosystem

functioning and services (Vörösmarty et al. 2010). Anthropogenic alterations of the natural environment, such as land use changes and habitat fragmentation, directly threaten species (Fahrig 2003; Poff et al. 2007; Franzen and Nilsson 2010; Gonzalez et al. 2011; Perkin and Gido 2012). One of the main challenges in community ecology and conservation biology is the understanding of the interactions between local and regional factors controlling population demography and community composition (Sheldon 1968; Chesson 2000; Hubbell 2001; Holyoak et al. 2005) and to eventually predict community dynamics (Fagan 2002; Economo and Keitt 2008).

Recent theoretical, experimental, and comparative work suggests that dispersal constrained by specific habitat structures is a major determinant of the observed diversity patterns at both species and genetic level (Fagan 2002; Muneepeerakul et al. 2008; Clarke et al. 2008; Morrissey and de Kerckhove 2009; Brown and Swan 2010; Finn et al. 2011; Carrara et al. 2012). Up to now, however, two major aspects have been neglected by most theoretical and experimental studies. First, landscape connectivity was generally considered independent of local environmental factors, such as habitat quality, patch size, environmental disturbances, and intra- and interspecific competition. While there are indeed cases for which this simplification is appropriate, such as forests (Hubbell 2001), island archipelagos (MacArthur and Wilson 1963), or natural ponds (Altermatt et al. 2008; De Bie et al. 2012), it does not represent many natural landscapes, such as fluvial and mountainous ecosystems, where local properties of the habitat and connectivity are intrinsically linked (Sheldon 1968; Benda et al. 2004; Lowe et al. 2006). Second, past studies generally adopted constant dispersal rates, symmetric kernels, and simplified landscape attributes (Warren 1996; Gonzalez et al. 1998; Mouquet and Loreau 2003; Matthiessen and Hillebrand 2006; Altermatt et al. 2011b;

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Chisholm et al. 2011). Traditionally, many studies in stream ecology, influenced by the river continuum concept (Vannote et al. 1980) have considered linear conceptual models to analyze drainage basins. Such simplified linear environmental matrices, however, may not completely capture biodiversity patterns within dendritic ecosystems (Fagan 2002; Grant et al. 2007; Brown and Swan 2010; Finn et al. 2011; Carrara et al. 2012).

In riverine ecosystems, landscape-forming discharges are related to total contributing drainage area, a by-product of spatial aggregation, depth, and width of the active river cross section (Leopold et al. 1964; Rodriguez-Iturbe and Rinaldo 1997; Benda et al. 2004). The river network not only provides suitable ecological corridors for individuals to disperse (Fagan 2002; Rodriguez-Iturbe et al. 2009) but often dictates the availability of microhabitats that species may eventually exploit (Cardinale 2011). Habitat capacity (i.e., river width/depth, reflecting patch size in rivers) scales with contributing area, dispersal is often biased downstream, and the distribution and intensity of disturbances are intrinsically linked with the position along the network through abrupt changes at confluences (Benda et al. 2004). Consequently, spatial correlations emerge between local properties and regional network descriptors in dendritic environments, where the hierarchical spatial organization of environmental heterogeneity is a fundamental driver of local species richness and community composition (Fernandes et al. 2004; Muneepeerakul et al. 2007). Riverine ecosystems, which are among the most threatened ecosystems on Earth (Vörösmarty et al. 2010), are thus a prominent natural system in which disentangling the effect of local environmental conditions and connectivity of the landscape on diversity is needed (Lowe et al. 2006; Lake et al. 2007; Gonzalez et al. 2011; Grant et al. 2012). For example, habitat capacity and interannual streamflow variability are changed in rivers undergoing hydropower development or cross-basin connections (Poff et al. 2007; Grant et al. 2012; Ziv et al. 2012). Furthermore, land-use and agricultural practices in many countries are affecting riparian zones, foreseing buffer zones on a fixed distance from the river bank only, irrespective of the spatial position within the river network (Gassner 2006). Theoretical, empirical, and comparative studies have suggested that the degradation of riparian vegetation structure and alteration of connectivity between the patches in the habitat mosaic may significantly reduce stream diversity at different trophic levels (Urban et al. 2006; Vörösmarty et al. 2010; Grant et al. 2012; Perkin and Gido 2012; Ziv et al. 2012). However, most studies on ongoing habitat change and legal regulations regarding riverine landscapes are not considering the intrinsic link of habitat capacity and network position (Lake et al. 2007).

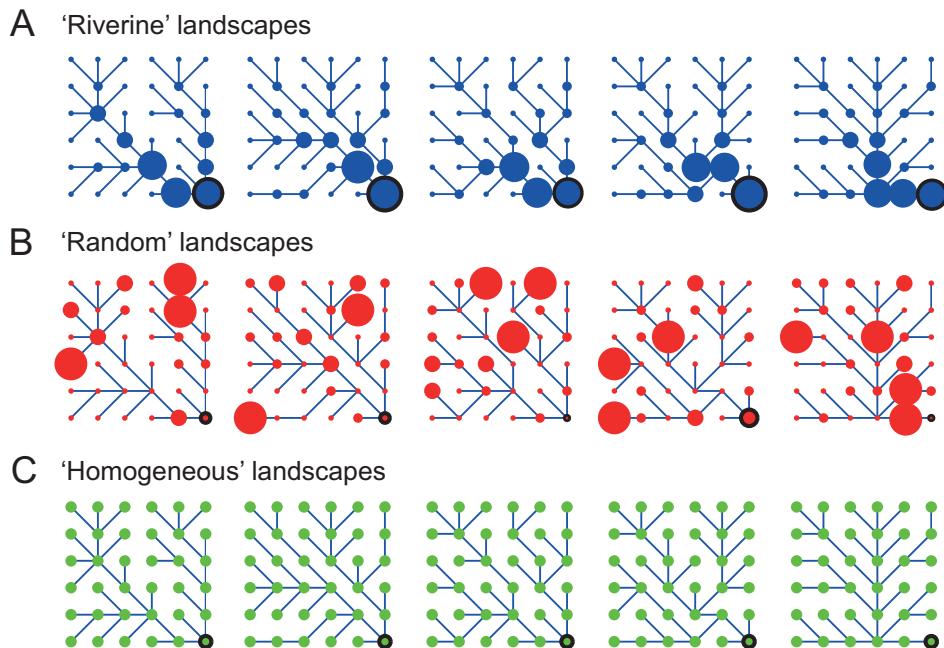
Here, we experimentally singled out the interaction of

dendritic connectivity and local habitat capacity (by modulating patch size) on the diversity of microorganisms in dendritic metacommunities, which were mimicking network structure and patch connectivity of natural river networks. Specifically, we singled out the individual influence of connectivity and habitat capacity on microbial diversity by using three different configurations of patch sizes (Riverine, Random, and Homogeneous), connected following a river network geometry (fig. 1; “Material and Methods”). In Riverine landscapes, local habitat capacity correlates with position along the network and distance to the outlet (fig. A1, table A1; figs. A1–A5 and tables A1–A5 available online). Larger downstream communities receive more immigrants from upstream communities, eventually having a combined positive effect on biodiversity. In the Homogeneous and Random landscapes, motivated by ongoing riverine habitat modifications, local habitat capacity (i.e., the patch size) does not preserve the geomorphological scaling observed in natural river systems (Leopold et al. 1964; Rodriguez-Iturbe and Rinaldo 1997). By measuring species’ persistence and species’ density we followed diversity patterns in terms of  $\alpha$ -,  $\beta$ -, and  $\gamma$ -diversity (local species richness, among-community dissimilarity, and regional species richness) and community evenness in the above landscape configurations. Aquatic microcosms as employed here and in several other studies (e.g., Fukami and Morin 2003; Cadotte et al. 2006; Haddad et al. 2008; Carrara et al. 2012; Giometto et al. 2013) offer a useful bridge between theoretical models and comparative field studies, to test for general macroecological principles (Holyoak and Lawler 2005; Livingston et al. 2012). Findings from such laboratory experiments, even if not directly comparable to natural systems, may cast light on important underlying mechanisms that steer metacommunity dynamics in river systems.

## Material and Methods

### *The Riverine Landscapes*

Each metacommunity (MC) consisted of 36 local communities grown in culture well-plates connected by dispersal. Communities, composed of nine protist and one rotifer species (see “Aquatic Communities”), were connected according to five different river network geometries (fig. 1). We used dendritic network landscapes derived from five different space-filling optimal channel networks (OCNs; Rinaldo et al. 1992; Rodriguez-Iturbe et al. 1992) known to reproduce the scaling properties observed in real river systems (Rinaldo et al. 2006). An appropriate coarse-graining procedure was enforced to reduce a complex construct to an equivalent  $6 \times 6$ -patch network, preserving the characteristics of the original three-dimensional basin



**Figure 1:** Spatial configuration of dendritic networks and corresponding patch sizes in the microcosm experiment. *A*, Riverine landscapes (blue) preserved the observed scaling properties of real river basins; *B*, Random landscapes (red) had the exact values of volumes as in the Riverine landscapes, randomly distributed across the networks; *C*, in Homogeneous landscapes (green) the total volume of the whole metacommunity was equally distributed to each 36 local communities. Patch size (size of the circle) is scaled to the actual medium volume. We had five unique river-like (dendritic) networks (columns; dispersal to neighboring communities followed the respective network structure, with a downstream bias in directionality toward the “outlet” community (black circled dot).

(for details, see app. A; apps. A–C available online). To have independent replication on the level of the networks, we have specifically chosen five different realization of river network configurations.

We relate the habitat capacity of an MC to physical properties that affect persistence of species (Reche et al. 2005). The largest organism to be sustained by an ecosystem is known to depend on habitat size (see, e.g., Banavar et al. 2007), termed “habitat capacity.” Therefore, habitat capacity conceptualizes our ability to rank different landscapes in terms of their capacity to support viable populations. We implemented three different treatments of patch size configurations in each of the five OCN landscapes (fig. 1): (i) a “Riverine” landscape, in which the volume  $v_i$  (i.e., the patch size) of the local community (LC)  $i$  preserves the scaling law observed in real river systems:  $v_i \propto A_i^{1/2}$ . Area  $A_i$  is the drainage area of the LC  $i$ , defined as the sum over all the volumes  $v_j$  draining in that particular point; (ii) a “Random” landscape, in which the exact values of patch volumes  $v_i$  as in the Riverine landscape were randomly distributed across the network; and (iii) a “Homogeneous” landscape where the total volume of the whole MC was equal to the other two treatments but each LC had a constant average value  $\bar{v}_{\text{hom}} =$

3.6 mL (Riverine = Riv, Random = Ran, Homogeneous = Hom). We binned values of  $v_i$  based on their original drainage area, in four size categories (2, 3.5, 6, and 12 mL). In order to test species coexistence in isolation, we had 72 communities of “isolation” treatment, in which the patch sizes were equal to the first two replicates of the Riv configuration of the main experiment but without dispersal. Environmental conditions, dictated in the well plates by the ratio of surface area to volume, are changing between communities with different patch size, favoring different sets of species at a time (app. B, fig. B1; figs. B1, B2 available online).

#### Aquatic Communities

We kept LCs in multiwell culture plates containing a variable medium volume (2–12 mL), in a climatized room at 21°C under constant fluorescent light. Protozoan Pellets (Carolina Biological Supply) and soil provided nutrients for bacteria (*Bacillus cereus*, *Bacillus subtilis*, and *Serratia marcescens*), which are consumed by protists. Each LC within an MC initially contained nine protist species (*Blepharisma* sp., *Chilomonas* sp., *Colpidium* sp., *Euglena gracilis*, *Euplotes aediculatus*, *Paramecium aurelia*, *Para-*

*mecium bursaria*, *Spirostomum* sp., and *Tetrahymena* sp.) and one rotifer species (*Cephalodella* sp., in the following included when speaking about protist; see app. B for species details). On day 0, we added  $K_s/10$  individuals of each species, except for *E. gracilis* ( $K_{Eug}/100$  individuals), which naturally occurs at higher densities. Species' specific carrying capacities  $K_s$  and intrinsic growth rates  $r_s$  were measured in pure cultures (Altermatt et al. 2011a; Carrara et al. 2012).

### *The Dispersal Events*

After the onset of the experiment, a dispersal event occurred every 3 days, in total eight times. Each time, half of the individuals of each community emigrated. Emigration happened by transfer of the individuals in the medium in well-mixed conditions and was thus density independent. The dispersers were manually transferred from every single LC to the nearest neighbors along the network, with absorbing boundary conditions. Emigration and subsequent immigration happened simultaneously across the landscape in well-mixed conditions. This specific type of density-independent (diffusive) dispersal, chosen to avoid long-tailed dispersal events, imposes equal per capita dispersal rates for all species. By applying eight one-step dispersal events, individuals from any population may disperse at most the average maximum size of our networks. In the dispersal procedure, we imposed a bias in directionality; that is, the probability for an individual to be dispersed into a downstream direction was double that of dispersing upstream. We introduced into the system a dispersal-induced mortality by discarding half of the emigrating individuals (equal to 25% of LC). We enforced the landscapes to maintain the initial spatial volume configuration by adding fresh medium to LC with a negative balance after the dispersal or by disregarding immigrating individuals in exceedance (see app. A for a detailed description of the method). Parallel to the experiment, we developed a suite of metacommunity models, including a purely diffusive model, maintaining the dispersal characteristics of the experiment and thus complementing and extending our experimental findings (app. A).

### *Biodiversity Patterns*

On day 27, 3 days after the last dispersal-disturbance treatment, we measured species richness and species abundance (on a logarithmic scale) for each LC. This time interval corresponds to  $\approx 10\text{--}100$  generations, depending on the species' intrinsic growth rate, and the dynamics occurs over ecologically significant timescales (Carrara et al. 2012). We screened the entire culture plate under a stereo microscope (Olympus SZX16) to avoid false absences of

the rarer species. We then estimated species density following a standard procedure (Haddad et al. 2008; Altermatt et al. 2011b) by direct microscope observation of 0.5 mL of medium. We obtained the number of species present in the MC ( $\gamma$ -diversity) and in every single LC ( $\alpha$ -diversity) and the spatial distribution of  $\alpha$ -diversity. For  $\beta$ -diversity, we considered Jaccard's dissimilarity index,  $\beta_J$ , based on species presence-absence, and an index based on species abundances,  $\beta_A$ . Index  $\beta_J$  is defined as  $1 - S_{ij}/(S_i + S_j - S_{ij})$ , where  $S_{ij}$  is the number of species present in both local communities  $i$  and  $j$ , whereas  $S_i$  is the total number of species in LC  $i$ . Index  $\beta_A$  is defined as  $(\sum_{k=1}^S w_k |x'_{ik} - x'_{jk}|)/(\sum_{k=1}^S w_k) - \beta_J$ , where  $x_{jk}$  is the abundance of species  $k$  in the LC  $j$ ,  $x'_{jk} = \log x_{jk} + 1$ , whenever  $x_{jk} \neq 0$ ,  $w_k$  is a weight to take into account joint absences, and  $S = 10$  is the species pool. Index  $\beta_J$  only looks at compositional differences in species identities, whereas  $\beta_A$  studies the compositional differences in abundances, and is not bounded between 0 and 1. Their sum gives the modified Gower's index,  $\beta_{MG}$  (Anderson et al. 2006). Thereby we could discriminate the effects of species richness on  $\beta$ -diversity. We measured local community evenness (LC evenness), described by  $E_i = 1 - 2/\pi \cdot \arctan [\sum_{s=1}^S (\log x_{is} - \sum_{t=1}^S \log x_{it}/S)^2/S]$ . This defines an index bounded between 0 and 1, independent of species richness (Smith and Wilson 1996). Regional evenness was calculated from species' populations in the whole network.

### *Statistical Analysis*

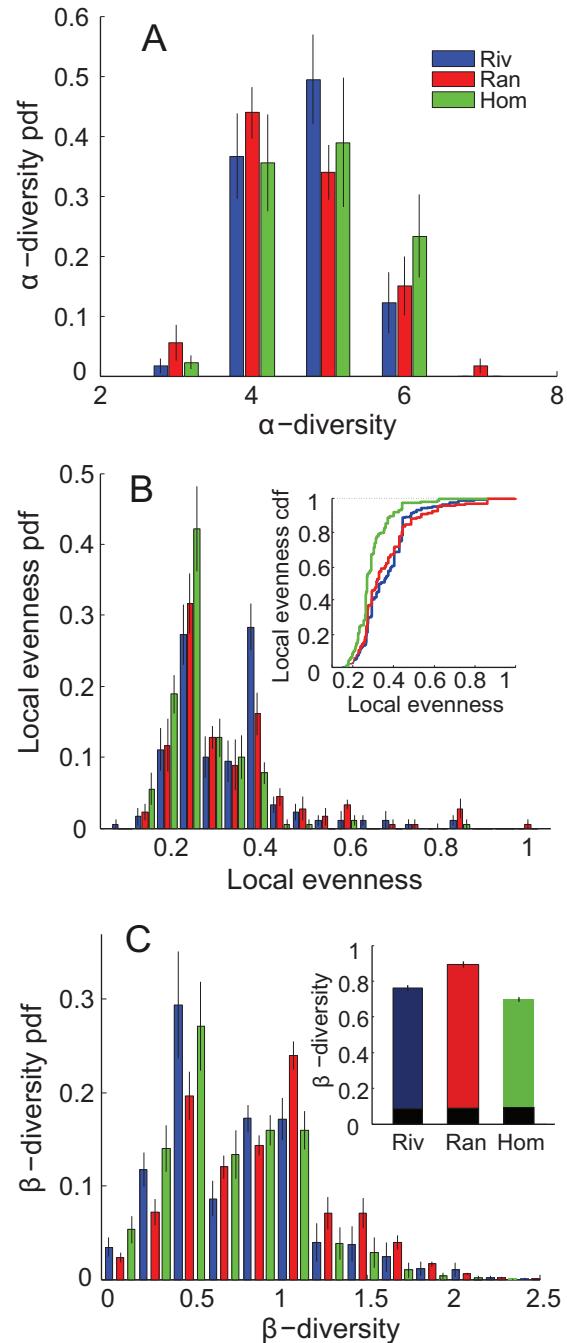
We used degree of connectivity  $d$  and patch size  $v$  as local network descriptors. Term  $d_i$  is defined as the number of connected neighboring nodes to the LC  $i$ , whereas  $v_i$  represent its habitat capacity. We used distance to the outlet  $o$ , drainage area  $A$ , and the ecological diameter  $l$  as regional network descriptors. Term  $o_i$  is calculated as the shortest path connecting  $i$  to the outlet community. The specific spatial arrangement of patch sizes upstream of LC  $i$  defines  $A_p$ . The average distance of  $i$  from all other communities in the river network defines  $l_i = \langle d_{ij} \rangle_p$ , where  $d_{ij}$  represents the shortest topological distance between  $i$  and  $j$ . Thus,  $l_i$  is the inverse of the closeness centrality of community  $i$  (Newman 2010) and in riverine landscapes defines network positioning (fig. A1; table A1).

We performed an ANCOVA with configuration treatments as categorical fixed effects, OCNs as random effect, and the above-described network descriptors (continuous variables) as fixed effects. Normality and constancy of variance in  $\alpha$ -diversity and LC evenness over covariates were verified with Bartlett's test. A parallel analysis performed with a generalized linear mixed model (GLMM) framework was conducted, giving qualitatively consistent results.

However, as model assumptions on error structure were better fulfilled by the ANCOVA than by a GLMM with Poisson error structure, we restricted ourselves to the former. Models were hierarchically simplified in a stepwise algorithm, starting with the full model and removing non-significant terms, with highest-level interactions first. We analyzed the overall effect of the configuration treatment on  $\alpha$ -diversity and LC evenness. We then used individual ANCOVAs within each landscape configuration to disentangle the effect of individual network properties on biodiversity indicators for each landscape configuration separately. We performed Kolmogorov-Smirnov tests (K-S tests) on the cumulative density functions of  $\alpha$ - and  $\beta$ -diversity and LC evenness of the three landscapes configurations. Mantel tests were used to study how  $\beta$ -diversity was influenced by topological distance, differences in environmental conditions and differences in network position in individual landscapes (Anderson et al. 2011). Differences in environmental conditions and network positioning were captured by respectively taking, for two focal communities  $i$  and  $j$ , the absolute differences in patch sizes  $|v_i - v_j|$  and ecological diameters  $|l_i - l_j|$ . The notation  $\langle \cdot \rangle$  in the main text means a spatial average over the nodes, while  $\bar{\cdot}$  represents an average over the five landscape replicates.

## Results

The spatial distribution of hierarchical patch size significantly affected community composition in dendritic environments (for all full ANCOVA results, see tables A2–A5; data are deposited in the Dryad Digital Repository: <http://dx.doi.org/10.5061/dryad.15np2>; Carrara et al. 2013). The  $\alpha$ -diversity and LC evenness were significantly different across the three landscape configurations, Riv, Ran, and Hom (ANCOVA  $\alpha$ -diversity:  $F_{2,4} = 7.99$ ,  $P = .0006$ , fig. 2A; ANCOVA LC evenness:  $F_{2,4} = 4.65$ ,  $P = .01$ , fig. 2B). Regionally, at the landscape scale,  $\gamma$ -diversity and mean  $\alpha$ -diversity did not vary significantly across the three different landscape configurations ( $\bar{\gamma}_{\text{Riv}} = 5.8$ ,  $\bar{\gamma}_{\text{Ran}} = 6.2$ ,  $\bar{\gamma}_{\text{Hom}} = 6$ ,  $\langle \bar{\alpha} \rangle_{\text{Riv}} = 4.72$ ,  $\langle \bar{\alpha} \rangle_{\text{Ran}} = 4.63$ ,  $\langle \bar{\alpha} \rangle_{\text{Hom}} = 4.83$ ;  $P$  values of all pairwise K-S tests on different configuration treatments were  $P > .1$ ). The LC evenness was higher for the configuration treatments in which spatial heterogeneity was introduced by varying patch sizes, that is, Riv (K-S test,  $P < 10^{-5}$ ,  $K = 0.33$ ) and Ran landscapes (K-S test,  $P < 10^{-5}$ ,  $K = 0.26$ ) compared to Hom landscapes (inset, fig. 2B). No significant difference was detected between Riv and Ran landscapes (K-S test,  $P = .45$ ,  $K = 0.09$ ). The  $\beta$ -diversity (modified Gower's index  $\beta_{\text{MG}}$ ) was higher in the two heterogeneous landscape configurations (Riv, Ran) compared to the Hom landscapes (K-S tests, both  $P < 10^{-5}$ ; fig. 2C). The Ran land-



**Figure 2:** Probability density function (pdf) for  $\alpha$ -diversity (A), local evenness (B), and  $\beta$ -diversity in Riverine (Riv; blue), Random (Ran; red), and Homogeneous (Hom; green) configurations over the five replicated dendritic landscapes. The insets give the cumulative density function of local evenness (B) and the mean  $\pm$  SEM of  $\beta$ -diversity (C), decomposed in an abundance-based index ( $\beta_A$ , upper colored-coded part), and presence-absence-based index ( $\beta_B$ , black, bottom part).

scapes showed the highest values of  $\beta_{MG}$  (K-S test,  $P < 10^{-5}$ ,  $K = 0.16$ ), but considering only  $\beta_p$ , no significant differences between the three configurations were found (*inset*, fig. 2C).

In the isolation treatment, species richness increased with increasing patch size. Larger protist species (especially *Paramecium aurelia* but also *Paramecium bursaria*) preferentially occupied patches with larger sizes (figs. B1, B2). The  $\alpha$ -diversity was significantly affected by the degree of connectivity ( $F_{1,4} = 77.5$ ,  $P < 10^{-5}$ ; fig. 3A) and by patch size ( $F_{1,4} = 15.8$ ,  $P = .001$ ; fig. 3B). A significant interaction was observed between connectivity and patch size on  $\alpha$ -diversity ( $F_{1,4} = 12.05$ ,  $P = .0006$ ), whereas distance to the outlet, ecological diameter, and drainage area had no overall significant effect on  $\alpha$ -diversity. LC evenness was significantly affected by drainage area only ( $F_{1,4} = 11.49$ ,  $P = .005$ ; fig. 3I).

#### $\alpha$ -Diversity Patterns

Because of the significant interaction between configuration treatment (i.e., Riv, Ran, and Hom) and individual network descriptors (tables A2, A3), three separate ANCOVA analyses were needed to address within-landscape relationships (one for each landscape configuration; table A4). In the following, we compare the three analyses (Riv, Ran, and Hom) separately for each network descriptor. The  $\alpha$ -diversity significantly increased with increasing degree of connectivity ( $P_{Riv} = .002$ ,  $P_{Ran} < 10^{-5}$ ,  $P_{Hom} = .005$ ), although with different slopes ( $\bar{m}_{Riv} = 0.26$ ;  $\bar{m}_{Ran} = 0.28$ ;  $\bar{m}_{Hom} = 0.40$ ; fig. 3A). Patch size significantly determined  $\alpha$ -diversity in Ran landscapes ( $P_{Ran} = .0002$ ; fig. 3B), where it also interacted with connectivity ( $P_{Ran} = .016$ ). Distance to the outlet significantly affected  $\alpha$ -diversity in Riv and Hom landscapes ( $P_{Riv} = .041$ ;  $P_{Hom} = .021$ ), where  $\alpha$ -diversity decreased with increasing distance to the outlet ( $\bar{m}_{Riv} = -0.22$ ,  $\bar{m}_{Hom} = -0.18$ ; fig. 3C). No dependence of  $\alpha$ -diversity on distance to the outlet was observed in the Ran landscapes ( $\bar{m}_{Ran} = -0.022$ ; fig. 3C). Drainage area  $A$  and ecological diameter  $l$  (fig. A1), two regional network descriptors ("Material and Methods"), did not significantly affect  $\alpha$ -diversity (fig. 3D, 3E).

#### Evenness Patterns

In Riv landscapes, LC evenness depended significantly on patch size ( $P_{Riv} = .004$ ; fig. 3G). There was also a significant interaction of patch size and distance to the outlet in determining local evenness ( $P_{Riv} = .02$ ; table A5). Instead, LC evenness in Ran and Hom landscapes did not show any significant dependence on any of the network descriptors (fig. 3F–3L; table A5). Importantly, the de-

pendence of LC evenness on patch size in Riv landscapes at the local scale contributed to shape significantly higher evenness values at the regional scale, compared to values in Ran landscapes (paired *t*-test,  $t_4 = 2.95$ ,  $P = .041$ ; fig. 4A) or in Hom landscapes (paired *t*-test,  $t_4 = 3.36$ ,  $P = .028$ ; fig. 4A). A higher variability in terms of coefficient of variation was detected in Riv landscapes across the five replicates (*inset*, fig. 4B). In Riv landscapes, species that presented low numbers of individuals in pure cultures, with higher body sizes and lower intrinsic growth rates, persisted at higher densities (figs. 4B, A3, B2).

#### $\beta$ -Diversity Patterns

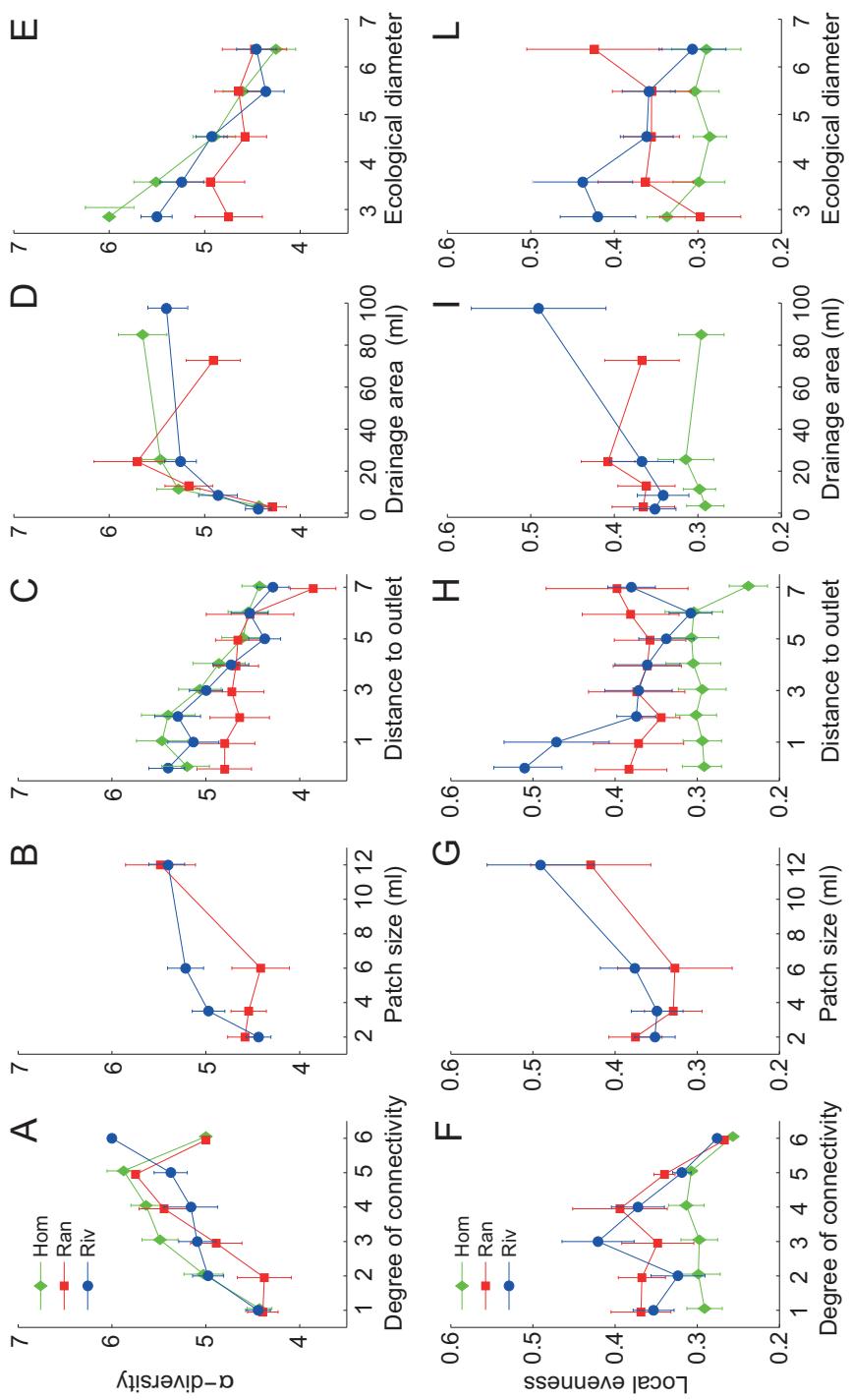
In Riv landscapes,  $\beta_{MG}$  increased with increasing pairwise topological distance between community pairs. No such pattern was found in the two other landscapes, and  $\beta$ -diversity showed a flat behavior (fig. 5A). Differences in community composition did not significantly depend on topological distance: Mantel tests revealed that there was no spatial dependence of  $\beta$ -diversity in any of the three configurations (figs. 5A, C1; table C1; fig. C1 and table C1 available online). Changes in patch size, instead, significantly affected  $\beta$ -diversity in Riv and Ran landscapes (figs. 5B, C1; table C1). Changes in ecological diameter, capturing the difference in network positioning (fig. A1) had a significant effect on  $\beta$ -diversity only in Riv and Hom landscapes (figs. 5C, C1; table C1,  $\bar{m}_{Riv} = 0.74$ ,  $\bar{m}_{Ran} = 0.16$ ,  $\bar{m}_{Hom} = 0.59$ ).

#### Discussion

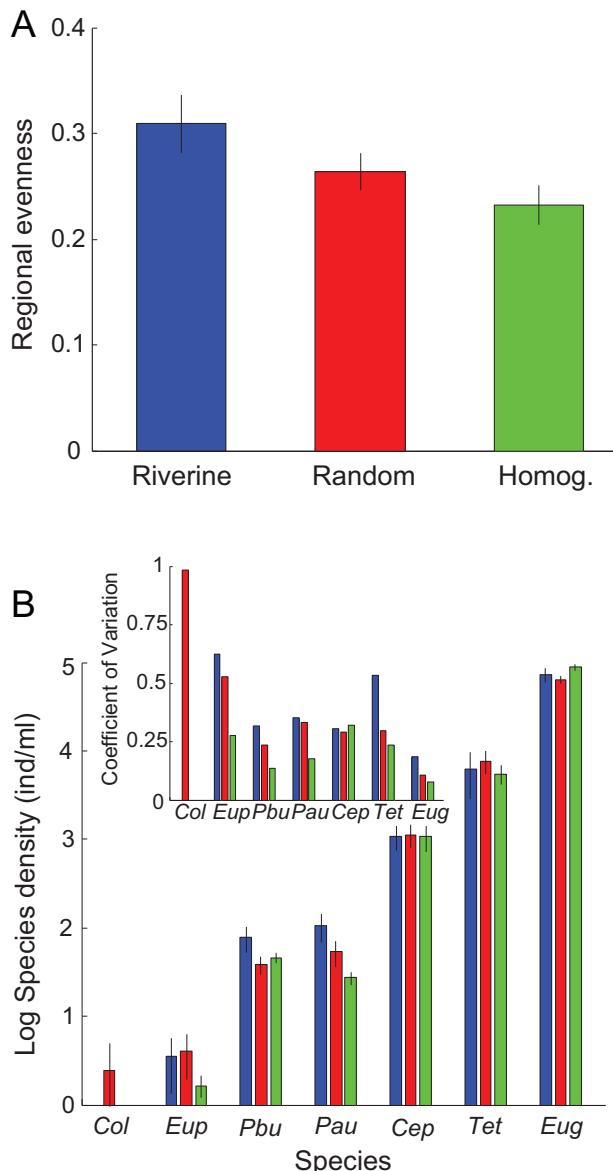
Our aquatic microcosm experiment disentangled the interaction between dendritic connectivity and hierarchical patch size on biodiversity of communities and showed that biodiversity patterns in river-like metacommunities significantly depend on the spatial covariance between dendritic connectivity, patch size, and position along the network.

#### $\alpha$ -Diversity Patterns

Highly connected nodes, irrespective of the local environmental factors such as patch size, disturbances, and interspecific competition among species, sustained higher levels of local species richness compared to more peripheral communities (Finn et al. 2011; Carrara et al. 2012; Perkin and Gido 2012). Such a correlation is predicted by network theory (Newman 2010), where individuals on a random walk will be found in community  $i$  proportionally to the degree of connectivity of  $i$  on a connected network. In all three landscape configurations (fig. 1),  $\alpha$ -diversity increased with increasing connectivity. Similar results are



**Figure 3:**  $\alpha$ -diversity (top row) and local evenness (bottom row) in dendritic landscapes for the three different configuration treatments (Riverine [Riv] landscapes in blue, Random [Ran] landscapes in red, and Homogeneous [Hom] landscapes in green), depending on local (connectivity, patch size) and regional network descriptors (distance to outlets, drainage area, ecological diameter).  $A, F, \alpha$ -diversity and local evenness as a function of degree of connectivity;  $C, H, G$ , as a function of the patch size;  $D, I, L$ , as a function of ecological diameter. Symbols represent the mean  $\pm$  SEM of the experimental data over the five experimental replicates.

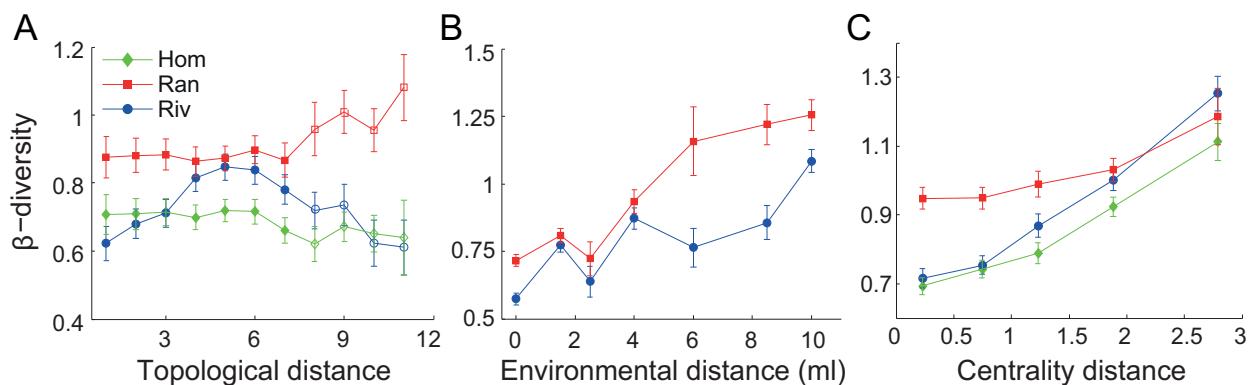


**Figure 4:** Community composition at metacommunity scale in Riverine (Riv; blue), Random (Ren; red), and Homogeneous (Hom; green) configurations. A, Regional evenness was significantly higher in Riv landscapes compared to Ran and Hom landscapes (average  $\pm$  SEM over the five experimental replicates). B, Species' density of the seven species detected at the end of the experiment for the three landscape configurations, sorted in increasing order of abundances (log scale). Values represent the mean  $\pm$  SEM across the five replicated landscapes. Inset, coefficient of variation for the same species over the five replicates (Col = *Colpidium* sp., Eup = *Euploites aediculatus*, Pau = *Paramecium aurelia*, Pbu = *Paramecium bursaria*, Cep = *Cephalodella* sp., Tet = *Tetrahymena* sp., and Eug = *Euglena gracilis*).

obtained from a neutral model employed in spatially explicit networks (Economou and Keitt 2010). Neutral models focus on the spatial structure and dispersal limitation constrained by the specific landscape connectivity (Economou and Keitt 2008; Muneepeerakul et al. 2008; Economou and Keitt 2010), suggesting that dispersal in our experiment was a driving factor of  $\alpha$ -diversity.

In Riv landscapes, a consistent increase in  $\alpha$ -diversity was observed for larger patch sizes. Overall, in hierarchical Riv environments patch size is not significant because of the spatial covariance between patch size with the other network descriptors (table A1), which better describe the variation in local species richness. Only in Ran landscapes patch size, controlling formation, and composition of microhabitats (app. B), significantly affected  $\alpha$ -diversities. Our results show that species richness in dendritic landscapes with spatially uncorrelated patch size was determined largely by local drivers (i.e., degree of connectivity and patch size). The significant interaction between degree of connectivity and patch size in Ran landscapes (table A4) is suggesting that in intermediate patches (patch size 3.5 and 6 mL), or in linear branches (connectivity  $d = 2$ ), the detrimental effects of disturbances (i.e., mortality due to emigration) on local species richness are enhanced (fig. 3A, 3B). In Ran landscapes, the largest patches sustained higher abundances and thereby maintained a high  $\alpha$ -diversity irrespective of position (fig. 3B), as found in aquatic bacterial communities in mountain lakes (Reche et al. 2005). Previous protist studies identified species' specific responses to disturbances (Haddad et al. 2008; Carrara et al. 2012), and found that a high intrinsic growth rate is the most important factor in promoting a species' ability to survive a disturbance (Haddad et al. 2008). Such model systems cover substantial biological complexity in terms of species interactions and trophic levels that cannot be entirely captured by any model (app. B). Species with low reproductive number are prone to suffer more from environmental disturbances and require larger and well-connected habitats to persist (Staddon et al. 2010; Perkin and Gido 2012). In our experiment, *Euploites aediculatus*, *Paramecium aurelia*, and *Paramecium bursaria* have lower intrinsic growth rates and lower carrying capacities in pure cultures (Altermatt et al. 2011a; Carrara et al. 2012). Because absolute numbers of individuals are lower for these species, demographic stochasticity plays an important role in affecting populations dynamics. In Ran landscapes, the negative effect of disturbances related to emigration on the survival of species with low reproductive rates was aggravated by the alteration of the natural hierarchical patch size configuration (see also app. A).

Regional network descriptors, that is, drainage area and ecological diameter (fig. A1), provide a synthesis of the above dynamics (fig. 3D, 3E). In Riv and Hom landscapes,



**Figure 5:**  $\beta$ -diversity as a function of the topological (A), environmental (B), and centrality distance (C) between local community pairs, in Riverine (Riv), Random (Ran), and Homogeneous (Hom) configurations of patch size. Symbols represent the mean  $\pm$  SEM of the  $\beta_{MG}$ -diversity data over the five experimental replicates. Open symbols in panel (A) indicate the points where the statistical significance is lower (topological distance  $\geq 8$ ; fig. C1, available online).

$\alpha$ -diversity increased with increasing contributing drainage area. In Ran landscapes, we instead observed a local peak at intermediate values of total contributing area with a decline in  $\alpha$ -diversity for the largest values (fig. 3D). A likely explanation for this pattern is that in Ran metacommunities high-capacity patches are displaced also in peripheral zones of the network (fig. 1B), thereby acting as local sources of immigrants for the neighboring communities in the sub-basin. Through mass effects, the displaced high-capacity patches maintained species populations in the less favorable small or peripheral habitats (fig. 3B). In general, such mechanism of population survival is important in determining metapopulation dynamics (Holyoak et al. 2005). For example, in a plant pollinator system exposed to habitat fragmentation, it was found that high-quality patches with large size were essential for bees persistence (Franzen and Nilsson 2010). It is a characteristic of natural riverine landscapes that high-capacity communities are placed at network position that have a high closeness centrality (Fernandes et al. 2004; fig. A1, table A1). These populations are thus strongly affecting the overall MC dynamics (Muneepeerakul et al. 2007; fig. 3E). Here, we experimentally demonstrated that breaking the natural link between patch size and connectivity at local scales destroys the regional pattern in species richness observed in Riv and Hom landscapes. In Ran configuration, local species richness did not increase while approaching the outlet community (fig. 3C), despite the converging character of the network structure and downstream-biased dispersal (Muneepeerakul et al. 2008). This suggests that in rivers the positioning of high-capacity patches has a higher significance for biodiversity (Lowe et al. 2006; Grant et al. 2007) than in other types of landscapes, such as ponds, islands, or lakes (De Bie et al. 2012). Even though

habitat capacity and interannual streamflow variability, which are altered in rivers undergoing hydropower development (Poff et al. 2007; Grant et al. 2012; Ziv et al. 2012), may not exactly correspond to our Ran landscapes, we here provide proof of principles for how altering the hierarchical patch size configuration in dendritic systems may impact on important biodiversity patterns of aquatic microbial communities.

#### Community Composition

Locally (i.e., at the patch scale), variation in patch size configuration in dendritic networks altered community composition. Theory suggests that species that are better competitors in a particular environmental condition eventually spread along the system, impeding other species' growth and thus exposing them to higher extinction levels (Hillebrand et al. 2008; Cardinale 2011). Accordingly, in our experiment we observed a predominance of *Euglena gracilis* (best competitor of our species pool) in Hom landscapes, whereas populations of larger species (*E. aediculatus*, *P. aurelia*, and *P. bursaria*) were reduced compared to heterogeneous configurations (fig. 4B). We found that habitat heterogeneity in Riv and Ran landscapes promoted both local species evenness (fig. 2B) and persistence of  $\beta$ -diversity (fig. 2C), compared to Hom landscapes. This gives a causal, experimental proof of principles on how homogenization of habitat size along river networks can affect diversity (Lowe et al. 2006).

At regional spatial scales, the consequences of the spatial configuration of patch sizes on community composition in terms of degree of dominance and species turnover were subtle. Only in Riv landscapes, LC evenness increased consistently with increasing patch size and decreasing distance

to the outlet (fig. 3G, 3H). This highlights the structuring power of hierarchy (fig. 3I, 3L). Spatial environmental autocorrelation resulted in higher levels of regional evenness in Riv ecosystems (fig. 4A), by increasing the population size of some of the rarer species and at the same time decreasing the total biomass of the more abundant species (figs. 4B, B2). Possibly, rare species with lower growth rates were able to track their specific niche requirements more efficiently in Riv landscapes (Cardinale 2011). Thereby, they grew to higher population densities compared to the Ran landscapes where the spatial autocorrelation was disentangled (see app. A, figs. A2–A5, where a stochastic metacommunity model is presented, generalizing this finding). We demonstrated how alterations of river-like landscapes may have strong effects on MC dynamics and impact important regional diversity and evenness properties (Hillebrand et al. 2008). We also found that the modified Gower's  $\beta_{MG}$  was mostly driven by its abundance component  $\beta_A$  (inset, fig. 2C), indicating that empirical and theoretical studies in community ecology and conservation biology have to consider species abundances.

The  $\beta$ -diversity increased in Riv landscapes with increasing topological distance along the network (fig. 5A). Such a pattern is commonly observed in comparative studies on riverine diversity (Muneepeerakul et al. 2008; Brown and Swan 2010). No spatial correlation of LC similarity was found in Ran and Hom landscapes (fig. 5A), suggesting that a combination of patch size and network position is needed to reproduce this pattern. The flat behavior of  $\beta$ -diversity against topological distance reflected the prominent role of local environmental conditions in structuring communities through species sorting and competition. Dispersal limitation alone cannot reproduce such an effect on community differentiation (Brown and Swan 2010; Astorga et al. 2012). We argue that priority effects, reflecting colonization history, did not play an important role in community assembly as species were present everywhere at the start of the experiment. Randomizing patch size, that is, altering the hierarchical riverine structure, is an analog of fragmenting the landscapes (see effects on  $\alpha$ -diversity: fig. 3A, 3B, 3D). At the same time it is opening up more diversified spatiotemporal niches (Chesson 2000) and producing more distinct species compositions. Interestingly,  $\beta$ -diversity depended on centrality in the Hom landscapes with fixed habitat capacity. This strongly indicates that dendritic connectivity per se shapes both  $\alpha$ - and  $\beta$ -diversity (Carrara et al. 2012). Community-composition turnover along centrality gradient was maximized for Riv landscapes, and hierarchical patch size distribution enhanced the turnover provided by dendritic connectivity itself (fig. 5C; table C1).

Our previous work, which was done with a similar

model system, focused only on effect of connectivity (Carrara et al. 2012). This study implemented a more realistic realization of natural rivers with varied patch size and biased dispersal kernel (Fagan 2002; Grant et al. 2007; Muneepeerakul 2007). We disentangled for the first time the effects of the intrinsic link of network position and habitat patch size on diversity and community evenness and also looked at effects on common versus rare species by altering riverine structure. We indeed provided evidence of the interaction between species traits and population responses to spatiotemporal gradients of local environmental conditions in spatially structured habitats. For example, fast population growth, allowing rapid population responses to a more unpredictable environment, might favor species with higher intrinsic growth rates and vice versa. Our protist species are naturally co-occurring in freshwater habitats and cover a wide range of intrinsic growth rates, body sizes, and other important biological traits, such as dispersal ability (Altermatt et al. 2011a; Carrara et al. 2012). We stress that our experiments, as in similar model systems (Cadotte et al. 2006; Haddad et al. 2008; Livingston et al. 2012), are conceptualized versions of natural ecosystems and do not allow direct extrapolation of our results to natural rivers. However, they enhance our understanding of complex systems in nature, where multiple processes are interacting on different scales (Holyoak and Lawler 2005). Dispersal rate, dispersal mode, and the strength of directionality are important factors in determining community patterns in theoretical models (Mouquet and Loreau 2003; Morrissey and de Kerckhove 2009) and natural communities (Clarke et al. 2008). In analogy to our experiments, recent comparative studies (Brown and Swan 2010; Astorga et al. 2012; De Bie et al. 2012; Heino 2013) showed that different dispersal abilities, controlled by body size and dispersal mode, determine a change in the response of community similarity to environmental variation and geographic distance. This suggests that the relative importance of the two structuring forces may depend on the group of organisms and the spatial scale. In riverine ecosystems, the river network itself provides at the same time the primary habitat for the species and suitable ecological corridors for individuals to disperse (Fagan 2002; Grant et al. 2007; Rodriguez-Iturbe et al. 2009), resulting in a close match between the physical and the ecological scales (Grant et al. 2007). This correspondence is recognized as important for the ecosystems' resilience at different levels of ecological complexity (Gonzalez et al. 2011). For a variety of species living in natural riverine systems, out-of-network movements are likely to occur, leading to intercatchment dispersal (Clarke et al. 2008; Brown et al. 2010). In macroinvertebrates, active dispersers with a terrestrial stage should track environmental heterogeneity better than passive dispersers with

only an aquatic stage (Heino 2013). Moreover, the strength of directionality in river systems might be much stronger for passive dispersers, as bacteria and protists, compared to macroinvertebrates, amphibians, or fishes (Astorga et al. 2012; De Bie et al. 2012). For example, a neutral metacommunity model showed that a symmetric dispersal kernel suitably described fishes' biodiversity patterns in the Mississippi-Missouri river system (Muneepeerakul et al. 2008). Thus, when comparing or extrapolating our results to natural systems, one needs to carefully assume that taxon specific aspect of dispersal are fulfilled and patterns and processes may not scale directly across all species and landscapes sizes. In our experiment, adopting a diffusive downstream-biased dispersal between isolated habitat patches and at discrete-time intervals ("Material and Methods"), different dispersal strategies could not naturally arise. Competition-colonization trade-offs were documented in protist studies that were adopting similar species (Cadotte et al. 2006; but see Haddad et al. 2008). Such mechanism of species coexistence, together with the storage effect (Chesson 2000), may interact with the spatial structure to shape diversity and ecosystem productivity, as tested in bacterial metacommunities (Livingston et al. 2012). As the system was continuously perturbed away from stationarity by dispersal and emigration mortality, we were observing a transient state. The duration of transient dynamics in our networks may depend nontrivially on the different patch sizes, which are sustaining different population sizes. System relaxation time to equilibrium could be investigated by implementing a metacommunity model with salient features of our experiment, but it goes beyond the scope of the present study.

### Conclusions

Understanding the effects of fragmentation and patch size distribution on communities is crucial, especially in highly diverse riverine systems. Hierarchical riverine habitats, characterized by a natural spatiotemporal heterogeneity, sustain higher levels of diversity (Muneepeerakul et al. 2007; Carrara et al. 2012). Because dispersal is constrained by the network pathway, the river network may become a trap for species when the dendritic system is exposed to habitat fragmentation and path size alterations. Protecting highly connected communities could help to avoid extinctions of species with low reproductive rates. These species are prone to suffer more from environmental disturbances and require larger and well-connected habitats to persist. By preserving the natural hierarchy of spatiotemporal heterogeneity along river networks, fast growing species and weak competitors alike are better able to persist. Our results not only causally demonstrate general ecological principles, but also give insights for developing the-

oretical metacommunity models in dendritic environments and for future empirical studies focusing on riverine ecosystems.

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Connectivity and habitat size rule community composition in riverine landscapes. Some of the protist species used in the experiment (from the top, counterclockwise): *Euglena gracilis* (green), *Colpidium* sp. (white), *Blepharisma* sp. (pink), *Spirostomum* sp. (black and white), *Blepharisma* (again in pink), *Cephalodella* sp. (black and white), and in the center, *Paramecium bursaria* (green). Photo credit: Francesco Carrara and Regula Illi.

## Appendix A from F. Carrara et al., “Complex Interaction of Dendritic Connectivity and Hierarchical Patch Size on Biodiversity in River-Like Landscapes”

(Am. Nat., vol. 183, no. 1, p. 13)

### Dendritic Metacommunities

#### The Riverine Landscapes

Optimal channel networks (OCNs) are based on the educated guess that configurations occurring in nature are those that minimize a functional describing the dissipated energy. OCNs are exactly related to the stationary solutions of the basic landscape evolution equation to leading order in the small gradient approximation. In particular, any configuration that minimizes total energy dissipation, within the framework of general dynamical rules (Rinaldo et al. 2006), corresponds, through a slope-discharge relation, to an elevation field that is a stationary solution of the basic landscape evolution equation, including channel erosion and deposition, hillslope mass wasting, and tectonic uplift. Loopless network configurations characterized by minimum energy dissipation are obtained by selecting the configuration, say  $s$ , that minimizes  $E(s) = \sum_j A_j^\gamma$  where  $j$  spans the spatial domain, and where  $A_i = \sum_{j \in \mathcal{A}_i} v_j + v_i$  is the drainage area of the node  $i$ . The expression  $\mathcal{A}_i := \{j | W_{j,i} = 1\}$  defines the ensemble of network points  $j$  upstream of  $i$  (Leopold et al. 1964; Rodriguez-Iturbe and Rinaldo 1997). Each network is thus identified uniquely, in a spanning loopless tree, by its connectivity matrix  $W$ . The slope-area exponent  $\gamma$  is exactly  $\gamma = 1/2$  in the small gradient approximation (typically  $\gamma < 1$ ; see Rinaldo et al. 2006). The global minimum (the ground state) of the functional  $E$  is exactly characterized by known mean field exponents. Thus, any stationary solution of the landscape evolution equation must locally satisfy the slope-area relationship. The flux at a point  $i$  has the direction of the maximum gradient of the elevation field (the direction toward the lowest among all nearest neighbors to  $i$ ) and the three-dimensional drainage basin can be reconstructed. Identifying the flux in a point  $i$  with the total contributing area  $A_i$  drained in that point, the field of landscape-forming fluxes corresponding to a given oriented spanning graph is reconstructed. From the fluxes, a field of elevation can be defined using the above relation with the local slope. A coarse-graining procedure, preserving the mean elevation and thus total energy dissipation of the space-filling OCNs (Rinaldo et al. 2006), was enforced to reduce a complex construct to an equivalent  $6 \times 6$ -patch network (fig. A1). Each landscape replicates is a local, dynamically accessible minimum of total energy dissipation and thus a tree (Rinaldo et al. 2006).

In river networks, several empirical relations exist that relate the size of a channel reach to that of the landscape-forming discharge at-a-station within a catchment (Leopold et al. 1964; Rodriguez-Iturbe and Rinaldo 1997). Such relations hinge on broadly verified assumptions: (i) the landscape-forming discharge is linearly proportional to the catchment area  $A$ . The assumption is reasonable for catchments not exceeding hundreds of square kilometers and/or where the driving climatic processes can be regarded as substantially constant in space; (ii) the width, depth and velocity at-a-station scale with total contributing area  $A$  to exponents, namely,  $a$ ,  $b$ , and  $c$  constrained by  $a + b + c = 1$  (Leopold et al. 1964). Typically, the morphology of the cross section is defined by  $a$  and  $b$ , such that  $a + b \sim (0.8 \div 0.9)$ .

#### Methods: The Dispersal Events

Dispersal was accomplished by manual transfer of medium from every single local community (LC) to its nearest neighbors, with downstream bias in directionality (nonisotropic dispersal). We applied the following dispersal procedure:

- Thoroughly mixing the medium in every LC.
- Selecting 50% of medium from each LC for emigration.
- Disregarding 50% of emigrating medium (25% of the original volume, fixed disturbance rate  $d_E = 0.5$ ).
- Transferring the remaining medium to its nearest neighbors along the network on a “mirror” landscape of 36 empty plates (see also Carrara et al. 2012, fig. S1). The downstream migration was always twice as great as the upstream migration. We chose absorbing boundary conditions, in which dispersal from headwaters/outlet was only two-thirds/one-third, respectively, of the medium dispersed by confluences.
- Collecting and mixing dispersed medium in mirror landscapes.
- Retransferring to the “real” landscape from the mirror landscapes, eventually disregarding the exceedance medium in order to maintain a constant community volume over time (see eq. [A1] below).

- Filling up every LC to its original volume with fresh medium to maintain the distribution of initial volume in each treatment (volume conservation).

In this manner the dispersal happened simultaneously, imposing equal per capita dispersal rates for all different species in well-mixed conditions, and we avoided long-tailed dispersal events.

Given two communities  $i$  and  $j$ , the probability  $P_{ij}$  to disperse from  $i$  to  $j$  is  $P_{ij} = P_{\text{down}}$  if  $i \rightarrow j$ ;  $P_{ij} = P_{\text{up}}/d_{\text{up}}(i)$  if  $i \leftarrow j$ ;  $P_{ij} = 0$  if  $i \nleftrightarrow j$ , where  $d_{\text{up}}(i)$  is the number of upstream communities connected to  $i$ ;  $P_{\text{up}} = 1/3$ ; and  $P_{\text{down}} = 2/3$ . We enforced the system to maintain the initial spatial volume configuration, by adding new medium in each location,  $V_i^{\text{new}}$ , or by disregarding a fraction  $m_{ji}$  of immigrating individuals from community  $j$  in exceedance, before retransferring from the mirror to the real landscape (see procedure above). Solving a balance equation for every LC  $i$ , imposing the volume  $v_i(t+1) = v_i(t)$  after every dispersal events, one can derive the following equation for new medium  $V_i^{\text{new}}$  or for mortality  $m_{ji}$ :

$$\Theta(\xi_i)V_i^{\text{new}} = \Delta t \sum_j [\Theta(-\xi_i)P_{ji}v_jm_{ji}] - \xi_i, \quad (\text{A1})$$

where  $\Theta$  is the Heaviside step function,  $\xi_i = v_i - (1 - d_E)\sum_j P_{ji}v_j$ ,  $d_E = 0.5$  is the fixed disturbance rate, and  $P_{ji}$  is given above.

## Metacommunity Models

*Communities with passive diffusive dispersal.* Figure A2 shows the final relative concentration of initial medium after the eight dispersal treatments, for a passive diffusion over the network (no reaction at the nodes, abundance dropped to  $\sim 10\%$  of the initial abundances). The riverine networks, by definition, are suffering a slightly lower disturbance rate due to emigration mortality, that is,  $m_{ji}$  in eq. (1) are minimized due to the spatial autocorrelation of hierarchical patch sizes. The highest disturbance rate is found in random networks with a difference of 10%–20% between Riv and Ran. Such experimental assumptions will be relaxed in next section, by suitably adjusting the dispersal rates in order to equalized the disturbance rates  $m_{ji}$  among the three treatments. We note that an overall disturbance rates of 25% (our disturbance rate) in the disturbance rates does not affect such communities, as already experimentally shown by previous studies (Altermatt et al. 2011a). In fact, the recovery from disturbances is much faster than the time between disturbances. Only at very high disturbance rates (90%–98%) the differences in disturbance rates become effective (Altermatt et al. 2011b). Thus, as our disturbances were well below that threshold, the disturbance difference could not explain the observed patterns, and species-interactions over the intradisturbance time periods in differently sized patches were driving the observed dynamics. This is also supported by our nonreactive particle dispersion model: disturbance affected the abundance of species, but the relative frequency across treatments was the same for all species. Thus, the suggested higher loss in random is indeed observable (and to be expected) in the particle model (fig. A3, left), but very different from our experimental work (fig. 4). We find that there is no effect on the regional community composition across the three treatments when only conservative, noninteracting particles are considered (fig. A3, right). Thus, the observed effects of the interaction between network structure and patch size position is indeed nontrivial and not just to be expected from a diffusion-only model. Specifically, the key results on the rarer versus more common species, and different “niche use” of species in these three landscape treatments can not be driven or explained by a conservative particle model.

*Communities with reaction and diffusion.* We extended the generality of our experimental results by running a metacommunity model, with the same dispersal procedure and network structures as in the experiment. We simulated biological communities of multiple species living in a patchy environment of varied patch size and competing for the same resources with intrinsic growth rates and carrying capacities similar to the empirically known species-specific traits, measured in Carrara et al. 2012. We generalized a stochastic formulation of the logistic process (the one-step “birth and death process” with space/food limitation to multiple species. Each individual of species  $s$  has a natural death rate  $d_s$  and a probability  $b_s$  per unit time to reproduce. To insure that the Markov property holds,  $d_s$  and  $b_s$  are assumed to be fixed and independent of the age/size of the individual. Moreover, competition gives rise to an additional death rate  $\alpha(\vec{n}(i))/V$ , which depends on individuals present in community  $i$ . We considered a pure interspecific competition for space framework to provide a null hypothesis for the effects of dispersal-habitat configuration interaction, not considering competitive hierarchy among species. The competition term reflects the fraction of space occupied by an individual of species  $s$  compared to that of individuals of species  $r$ , that is,  $K_s/K_r$ . A metacommunity is composed of 36 nodes  $\mathcal{N} = (1, 2, \dots, i, \dots, A = 36)$ , with the same spatial configuration of patch sizes as in the experiment (Riverine, Random, Homogeneous). The transition probabilities for birth, death and migration of an individual of the  $r$ th species within

community  $i$  (with  $\vec{n}(i) = (n_1, n_2, \dots, n_r, \dots, n_I)$  individuals in species pool  $\mathcal{P} = (1, 2, \dots, r, \dots, I)$ , respectively) read

$$T(\vec{n}(i) + \vec{e}_{ri} | \vec{n}(i)) = b_r(i)n_r(i) + l_r \sum_{j=1}^A [P_{ji}m_{ji}n_r(j)], \quad (\text{A2})$$

$$T(\vec{n}(i) - \vec{e}_{ri} | \vec{n}(i)) = (d_r(i) + l_r(i))n_r(i) + \frac{1}{v_i}(b_r(i) - d_r(i))n_r(i) \left( \sum_{s \neq r}^I \frac{n_s(i)}{K_s} + \frac{n_r(i) - 1}{K_r} \right), \quad (\text{A3})$$

where  $\vec{e}_{ri}$  is a unit vector whose only  $r$ th component in community  $i$  is not 0, and  $l_r$  is the dispersal rate of species  $r$ , kept constant in the network. The multivariate master equation for the whole metacommunity is given by

$$\begin{aligned} \frac{\partial}{\partial t} p(\vec{n}, t) = & \sum_j \sum_r \{ T(\vec{n}(j) | \vec{n}(j) + \vec{e}_{rj}) p(\vec{n}(j) + \vec{e}_{rj}, t) \\ & + T(\vec{n}(j) | \vec{n}(j) - \vec{e}_{rj}) p(\vec{n}(j) - \vec{e}_{rj}, t) \\ & - [T(\vec{n}(j) + \vec{e}_{rj} | \vec{n}(j)) + T(\vec{n}(j) - \vec{e}_{rj} | \vec{n}(j))] p(\vec{n}(j), t) \}. \end{aligned} \quad (\text{A4})$$

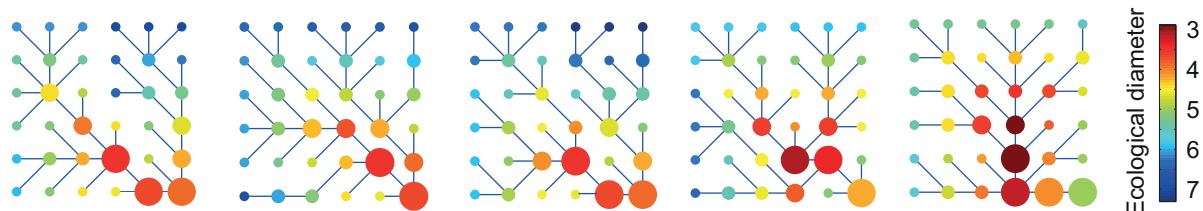
Due to the limited LC volume ( $v_i \leq 12$  mL) and the fact that the species' carrying capacity in some cases is small (fewer than 100 individuals  $\text{mL}^{-1}$  of medium), fluctuations around the macroscopic solutions may not be negligible. Thus, we performed numerical simulations employing the Gillespie algorithm, which allowed us to produce time series that exactly recover the solution of the multivariate master equation in equation (A4) with transition probabilities in equations (A2) and (A3). As we can not assume "well-mixed" conditions for individuals of all species, the nearest-neighbor dispersal along the network is also simulated in a stochastic procedure, following the experimental procedure, with discrete temporal steps. At each dispersal time step (every 3 days), we randomly distributed individuals in 100 cells, and then we randomly chose 50 cells to be dispersed to LCs nearest neighbors (emigration rate). From the emigrating pool, we disregarded the individuals from 25 randomly selected cells (emigration mortality):

$$\vec{n}(i, t+1) = (1 - \Delta l_i)\vec{n}(i, t) + F \sum_{j=1}^A (\Delta l_j P_{ji}m_{ji}\vec{n}(j, t)) + \vec{b}(i, \vec{n}(i, t)) \cdot \vec{n}(i, t) - \vec{d}(i, \vec{n}(i, t)) \cdot \vec{n}(i, t), \quad (\text{A5})$$

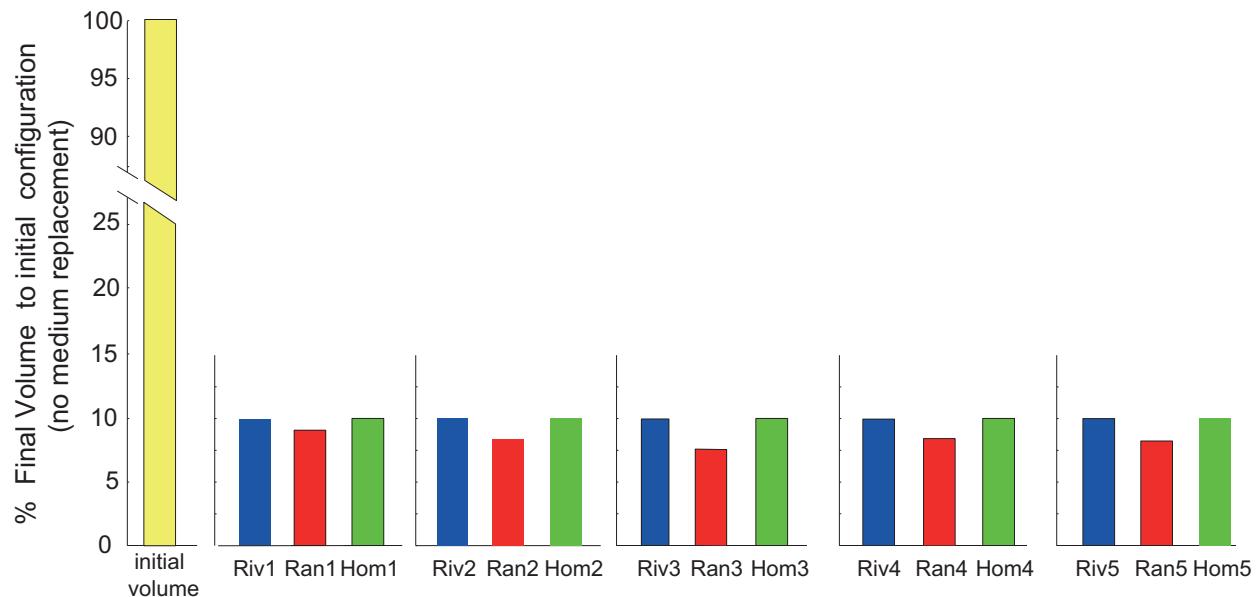
where  $\Delta l_i$  is the fraction of individuals to be dispersed at each time step from community  $i$ , which depend on the node  $i$ ,  $F = 1 - d_E$  is the survival probability during emigration, and  $m_{ji}$  is the mortality rate imposed to keep the volume configuration constant over time, given in equation [A1]. We simulated "neutral" and "nonneutral" metacommunities, composed of 35 species, with equal carrying capacity,  $K = 350$  individuals  $\text{mL}^{-1}$ . Species in neutral communities shared the same intrinsic growth rate,  $r_s = 1$ . Nonneutral metacommunities had species with a variable intrinsic growth rates, in the range  $[0.1, 3.5] \text{ day}^{-1}$ , with a 0.1 step. We observed the system over 72 days, a time window that covers three times the experimental duration. In the model, we equalized the disturbance rate among the three different spatial configurations by varying the dispersal rate (node-specific dispersal rate instead of node-specific disturbance rate) in order to have the same amount of disturbance at the metacommunity level. We found that regional evenness was not impacted in neutral metacommunities by the spatial configuration of community sizes. In nonneutral metacommunities, regional evenness was indeed higher in Riverine landscapes, with hierarchical patch size distribution, and the effect was consistent over time (figs. A4, A5). In Riverine landscapes, species with lower intrinsic growth rate occurred at higher abundances compared to Random and Homogeneous landscapes, with a similar trend to that highlighted in the main experiment, where much more biological complexity is captured (fig. 4; see also app. B). Species with a low intrinsic growth rate, which are suffering most by the disturbances due to emigration mortality, are possibly tracking more efficiently the environmental variation in autocorrelated Riverine environments, compared to landscapes where the environmental variation in patch size was disentangled (Random) or kept equal across the landscape (Homogeneous).

## Statistical Analysis

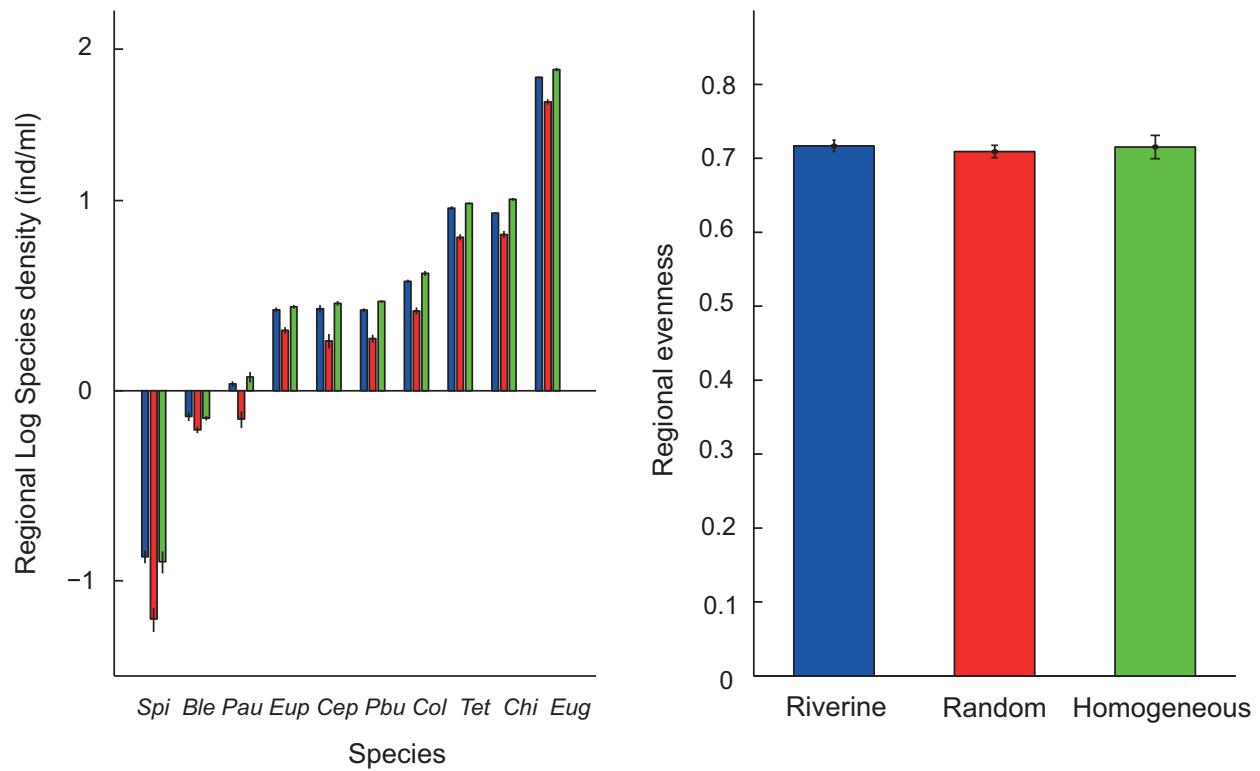
In our analysis, we used degree of connectivity  $d$  and patch size  $v$  as local network descriptors. Terms  $d_i$  and  $v_i$  are respectively the number of connected neighboring nodes and the habitat capacity of local community LC  $i$  (size of the circle). We used distance to the outlet  $o$ , drainage area  $A$ , and the ecological diameter  $l$  as regional network descriptors. Term  $d$  varies between 1 and 6, term  $o$  varies between 0 and 7 (in units of topological distance),  $v$  falls in four size categories (2, 3.5, 6, and 12 mL), and  $A$  varies over the range of 2 (headwater communities) to 136 mL (outlet of the river network). Term  $l$  varies in the range from  $l_{\min} = 2.7$  (central LC) to  $l_{\max} = 7.1$  (peripheral LC). We performed a correlation among these network descriptors (table A1).



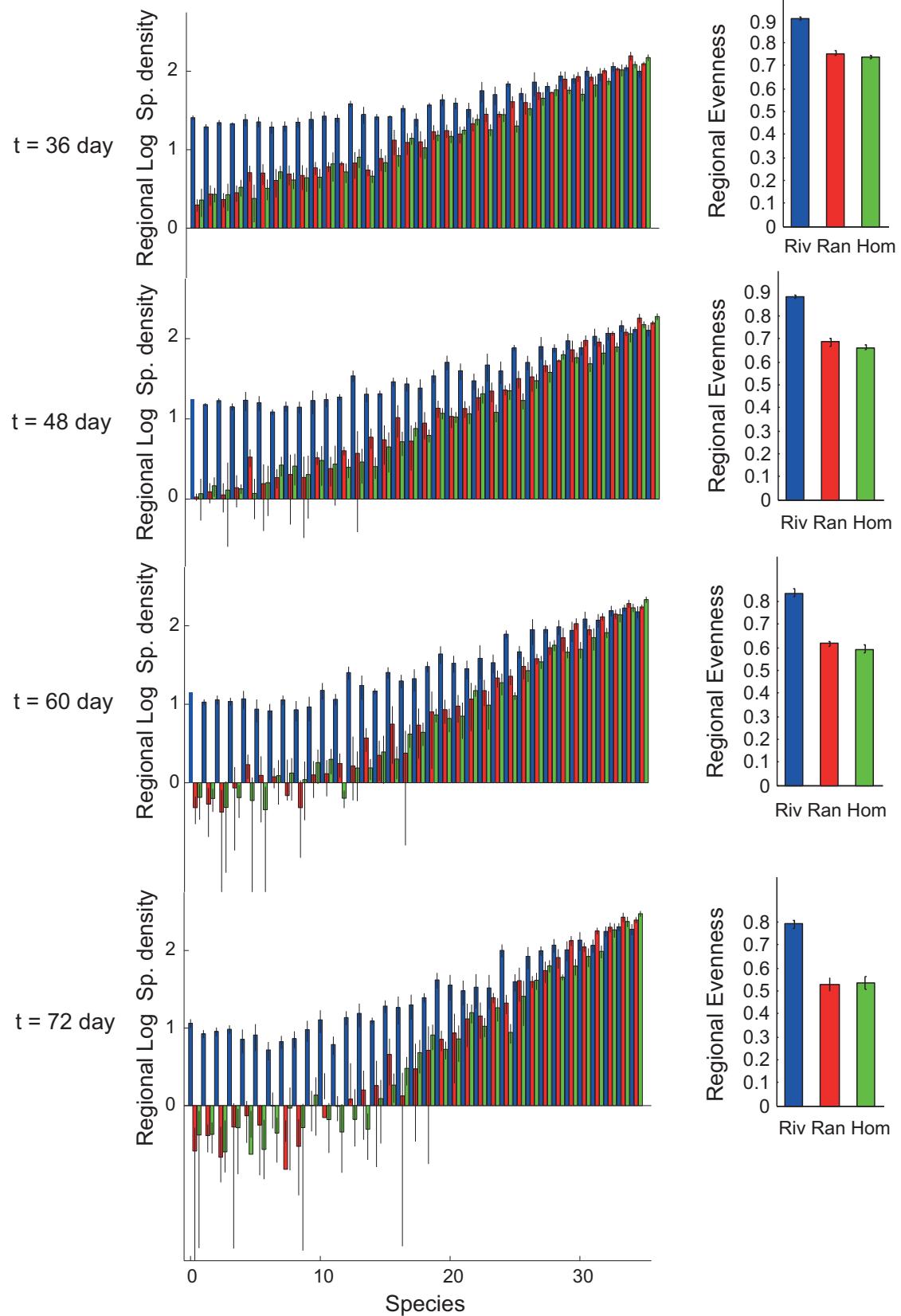
**Figure A1:** Spatial distribution of ecological diameter (color coded) for the five different river-network replicates in riverine landscapes. Ecological diameter is defined as the average distance of a node from all the other nodes in the network. Closeness centrality, in network theory, is simply the inverse of ecological diameter.



**Figure A2:** Hypothetical final volume of medium (or abundances of individuals in the case of nonreacting particles) relative to initial volume conditions (in yellow) after eight dispersal treatments, without replacing medium (i.e., not maintaining volume conservation). Dispersal was simulated as in the experiment but without replacing medium. The volume/abundances drop to  $\sim 10\%$  of the initial volume due to the migration mortality we implemented into the dispersal procedure. Differences in final volume/abundances among the three landscapes are within 20% in the five experimental replicates (Riverine, Random, and Homogeneous; blue, red, and green, respectively).

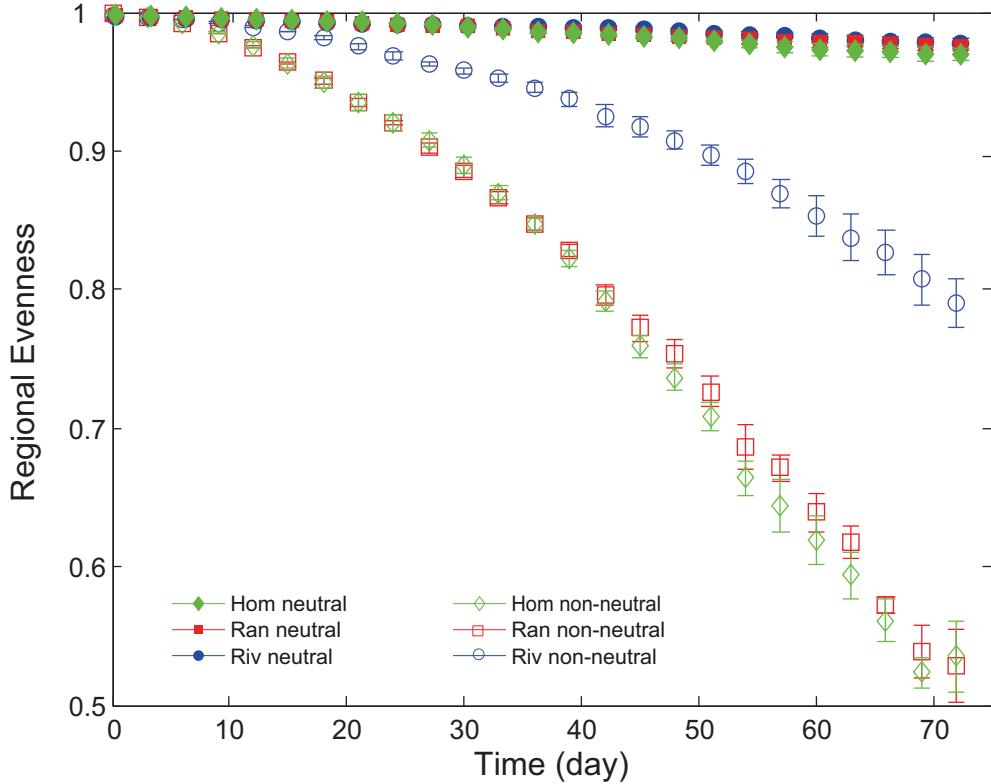


**Figure A3:** Results from stochastic metacommunity simulations adopting no-growth particles. Species abundance (*left*) and community evenness (*right*) were simulated using nonreactive, diffusive dispersal along the network structure only. In the initial setup of the model, species started with identical abundance as in the experiment (reflecting species-specific carrying capacities), and dispersal-disturbance rates were identical as in the experiment and across the three treatments (Riverine, Random, and Homogeneous; blue, red, and green, respectively).



**Figure A4:** Results from stochastic metacommunity simulations: species abundances (*left*) and regional evenness (*right*), averaged over

the metacommunity and over the five different realizations of the five river networks adopted in the main experiment (Riverine [Riv], Random [Ran], Homogeneous [Hom]; blue, red, and green, respectively). Nonneutral communities consisted of 35 species with varying growth rates over the range  $[0.1–3.5]$  day $^{-1}$  and the same carrying capacities  $K = 350$  individuals mL $^{-1}$ . Species are ordered from left to right with increasing growth rate, which is reflecting directly in biomass production, as the interaction is only for space. Initial conditions were the same across the three landscapes, with each species starting with the same number of individuals. The Riverine landscapes maintain higher abundances of the rarer species over time, compared to Random and Homogeneous landscapes. The more common species on the right part of the graph, with higher intrinsic growth rate, were instead equally abundant in the three landscapes. This results into a higher regional evenness, reflecting the results found in the main experiment, where some of the rarer species had higher abundances in Riverine landscapes (fig. 4).



**Figure A5:** Stochastic metacommunity simulations with neutral and nonneutral dynamic: regional evenness as a function of time, for the three spatial configurations adopted in the main experiment (Riverine [Riv], Random [Ran], Homogeneous [Hom]; blue, red, and green, respectively). Filled symbols show regional evenness for metacommunities with 35 neutral species (same intrinsic growth rate  $r = 1$  and same carrying capacities  $K = 350$  individuals mL $^{-1}$ ). Open symbols represent regional evenness for nonneutral metacommunities of 35 species with varying growth rates over the range  $[0.1–3.5]$  day $^{-1}$  and the same carrying capacities  $K = 350$  individuals mL $^{-1}$ . The neutral metacommunities maintain the same regional evenness (as diffusive particles, without reaction at the nodes; fig. A2). Interestingly, the nonneutral metacommunities show differentiations between the configuration treatments, with Riverine having higher evenness compared to Random and Homogeneous landscapes, an effect that we found also in the experiment. The decline in regional evenness reflects the increasing abundance imbalance of rare versus common species over time (see fig. A4).

**Table A1.** Correlations among network descriptors

Landscape, variable	$d$	$v$	$o$	$A$
Riverine:				
$v$		.63*		
$o$		-.36*	-.63*	
$A$		.48*	.94*	-.63*
$l$		-.57*	-.62*	.55*
Random:				
$v$		-.03		

**Table A1** (*Continued*)

Landscape, variable	<i>d</i>	<i>v</i>	<i>o</i>	<i>A</i>
<i>o</i>	-.36*	.07		
<i>A</i>	.53*	.09	-.6	
<i>l</i>	-.57*	.13	.55*	-.54*
Homogeneous:				
<i>v</i>	...			
<i>o</i>	-.36*	...		
<i>A</i>	.54*	...	-.64*	
<i>l</i>	-.57*		.55*	-.57*

Note: Correlation matrix (Pearson correlation) among network descriptors (*d* = degree of connectivity; *v* = patch size; *o* = distance to outlet; *A* = drainage area; *l* = ecological diameter) for the three network configurations. In Random landscapes, patch size loses correlation with the other network descriptors, compared with Riverine landscapes. Patch size is constant by definition in Homogeneous landscapes.

\**P*<.01.

**Table A2.** Results of the ANCOVA testing for effects of network descriptors on  $\alpha$ -diversity

Variable	df	<i>F</i>	<i>P</i>
<i>R</i>	4	.11	.977
<i>N</i>	2	7.99	>.001
<i>d</i>	1	77.50	>.001
<i>v</i>	1	15.80	.001
<i>o</i>	1	4.92	.088
<i>d</i> × <i>o</i>	1	12.05	>.001
<i>N</i> × <i>d</i>	2	6.30	.002
<i>N</i> × <i>o</i>	2	8.03	>.001
<i>R</i> × <i>N</i>	8	6.89	>.001
<i>R</i> × <i>d</i>	4	1.40	.232
<i>R</i> × <i>v</i>	4	3.23	.012
<i>R</i> × <i>o</i>	4	1.69	.151
Error	505		

Note: Replicated landscapes (*R*, reflecting independent network configurations) were specified as random effects, whereas network configurations (*N*) were categorical fixed effects. The analysis was made taking into account the network descriptors in table A1 (introduced in “Material and Methods”). Variables *l* (ecological diameter) and *A* (drainage area) were eliminated from the model because of nonsignificant effects on  $\alpha$ -diversity.

**Table A3.** Results of the ANCOVA testing for effects of network descriptors on local community evenness

Variable	df	<i>F</i>	<i>P</i>
<i>R</i>	4	1.00	.405
<i>N</i>	2	4.77	.009
<i>v</i>	1	3.62	.071
<i>A</i>	1	11.49	.005
<i>l</i>	1	.10	.769
<i>N</i> × <i>l</i>	2	4.14	.016
<i>R</i> × <i>N</i>	8	3.20	.002
<i>R</i> × <i>v</i>	4	.24	.917
<i>R</i> × <i>A</i>	4	.19	.946
<i>R</i> × <i>l</i>	4	1.12	.347
Error	508		

Note: The analysis was made taking into account the network descriptors in table A1 (introduced in “Material and Methods”), replicates (*R*) and network configurations (*N*) factors. The model was simplified in a stepwise procedure.

**Table A4.** Results of the ANCOVA testing for  $\alpha$ -diversity patterns

Landscape, variable	df	F	P
Riverine:			
R	4	1.53	.196
d	1	44.26	.002
v	1	.42	.55
o	1	8.75	.041
R $\times$ d	4	.54	.706
R $\times$ v	4	1.12	.347
R $\times$ o	4	3.09	.017
Error	160		
Random:			
R	4	.99	.417
d	1	32.56	>.001
v	1	17.78	>.001
o	1	.49	.518
d $\times$ v	1	5.86	.016
R $\times$ d	4	1.05	.381
R $\times$ v	4	.85	.496
R $\times$ o	4	1.44	.222
Error	159		
Homogeneous:			
R	4	.55	.156
d	1	30.59	.005
o	1	12.61	.021
R $\times$ d	4	3.38	.011
R $\times$ o	4	.57	.681
Error	165		

Note: Three separate ANCOVA analyses were performed for each landscape configuration. Landscape replicates (R) were specified as random effects. The analysis was made taking into account the network descriptors in table A1.

**Table A5.** Results of the ANCOVA testing for local community evenness patterns

Landscape, variable	df	F	P
Riverine:			
R	4	1.12	.349
v	1	10.08	.004
o	1	.37	.55
v × o	1	5.22	.024
R × v	4	1.23	.298
R × o	4	.92	.453
Error	164		
Random:			
R	4	1.52	.197
v	1	2.87	.152
o	1	2.96	.135
l	1	3.16	.138
R × v	4	.21	.932
R × o	4	1.92	.110
R × l	4	2.24	.068
Error	160		
Homogeneous:			
R	4	2.11	.081
d	1	.001	.964
o	1	1.80	.224
l	1	.89	.385
A	1	.74	.433
R × d	4	.89	.469
R × o	4	1.42	.231
R × l	4	1.41	.232
R × A	4	.56	.691
Error	155		

Note: Three separate ANCOVA analyses were performed for each landscape configuration. Landscape replicates (R) were specified as random factors. The analysis was made taking into account the network descriptors in table A1.

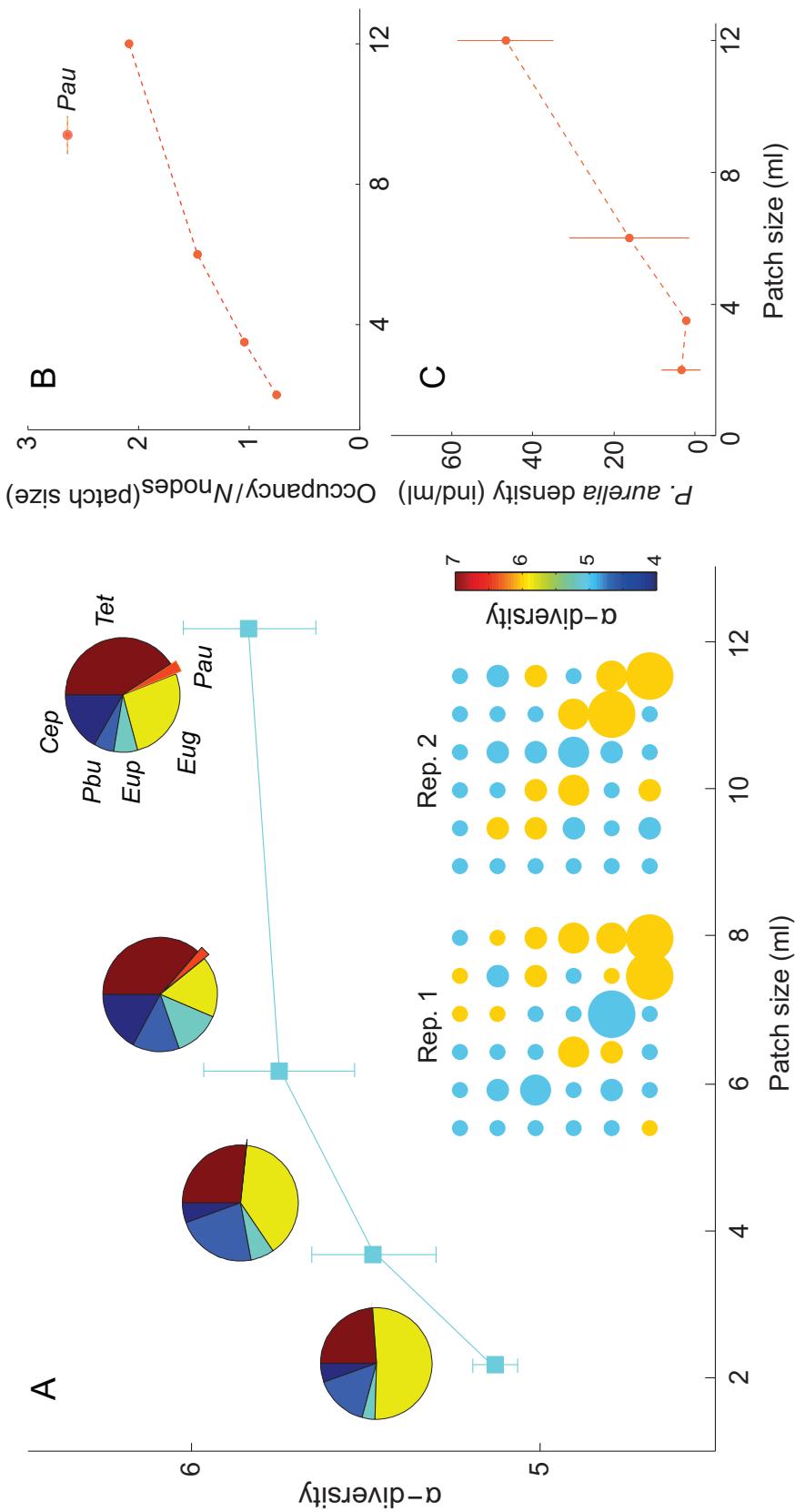
## **Appendix B from F. Carrara et al., “Complex Interaction of Dendritic Connectivity and Hierarchical Patch Size on Biodiversity in River-Like Landscapes”**

**(Am. Nat., vol. 183, no. 1, p. 13)**

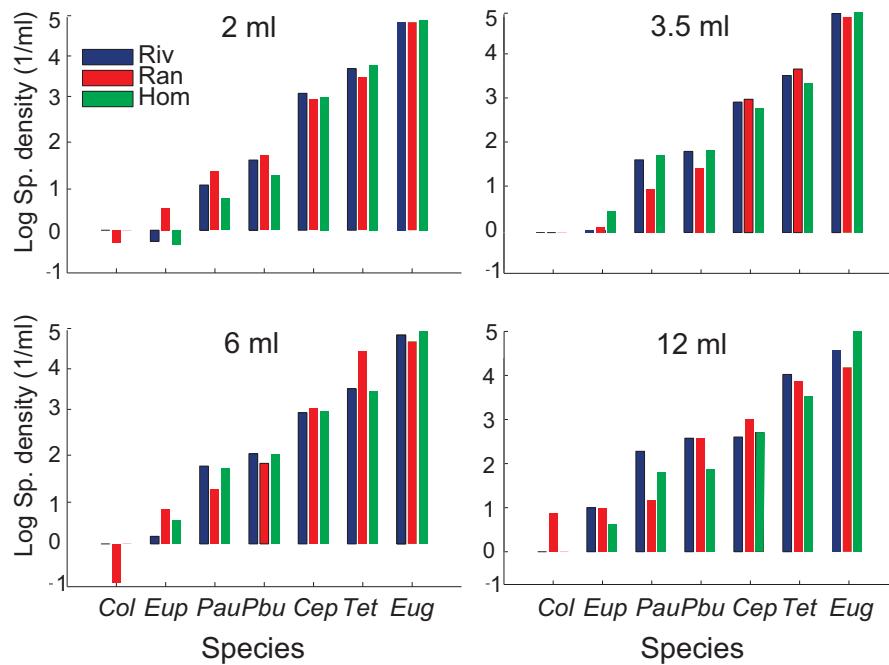
### **Local Communities and Patch Size Effects**

Local communities were located in multiwell culture plates containing a solution of sterilized local spring water, 1.6 g L<sup>-1</sup> of soil and 0.45 g L<sup>-1</sup> of Protozoan Pellets (Carolina Biological Supply). In our species pool, *Blepharisma* sp., *Chilomonas* sp., and *Tetrahymena* sp. were supplied by Carolina Biological Supply, while all other species were originally isolated from a natural pond (McGrady-Steed et al. 1997). All species are bacterivores, whereas *Euglena gracilis*, *Euplotes aediculatus*, and *Paramecium bursaria* can also photosynthesize. Furthermore, *Blepharisma* sp., *Cephalodella*, *E. aediculatus*, and *Spirostomum* sp. may not feed only on bacteria but can also predate on smaller flagellates, which are always present and remain unidentified.

We ran two replicates (72 local communities) without emigration events to test species coexistence in isolation and dependence of community composition on patch size. Patch sizes were the same as adopted for the first two replicates of the Riverine configuration treatment of the main experiment (fig. B1). Environmental conditions, dictated in the well plates by the ratio of surface area to volume, are changing between communities with different patch size, favoring different sets of species at a time. In figure B1, pie charts in inset of A and panels B and C of emphasize the increasing in density of *Paramecium aurelia* with increasing patch size.



**Figure B1:** “Isolation” treatment to test the effect solely of habitat capacity (patch size). *A*,  $\alpha$ -diversity as a function of patch size for the two replicates in the isolation treatment; inset, pie charts show the average proportion of each species in the corresponding patch size. Cep = *Cephalodella* sp. (rotifer), Eug = *Euglena gracilis*, Eup = *Euplotes adiculatus*, Pau = *Paramecium aurelia*, Pbu = *Paramecium bursaria*, and Tet = *Tetrahymena* sp. (see relative increasing of *P. aurelia*, in orange, with increasing patch size). *B*, Distribution of occupancy relative to community patch size for *P. aurelia*, which generally survived in bigger patch sizes, with higher densities (*C*).



**Figure B2:** Species' density, sorted in increasing order of abundances (log scale) for the seven species detected at the final stage of the experiment for the three patch size configurations (Riverine in blue, Random in red, and Homogeneous in green), averaged over similar community patch sizes and over the five replicates. Col = *Colpidium* sp., whereas the other names correspond to species as in figure B1. Homogeneous communities were considered in the comparison by averaging over the same nodes as in Riverine landscapes. Only in Random landscapes did we observe *Colpidium* sp., a weak competitor but a fast grower.

## Literature Cited Only In Appendix B

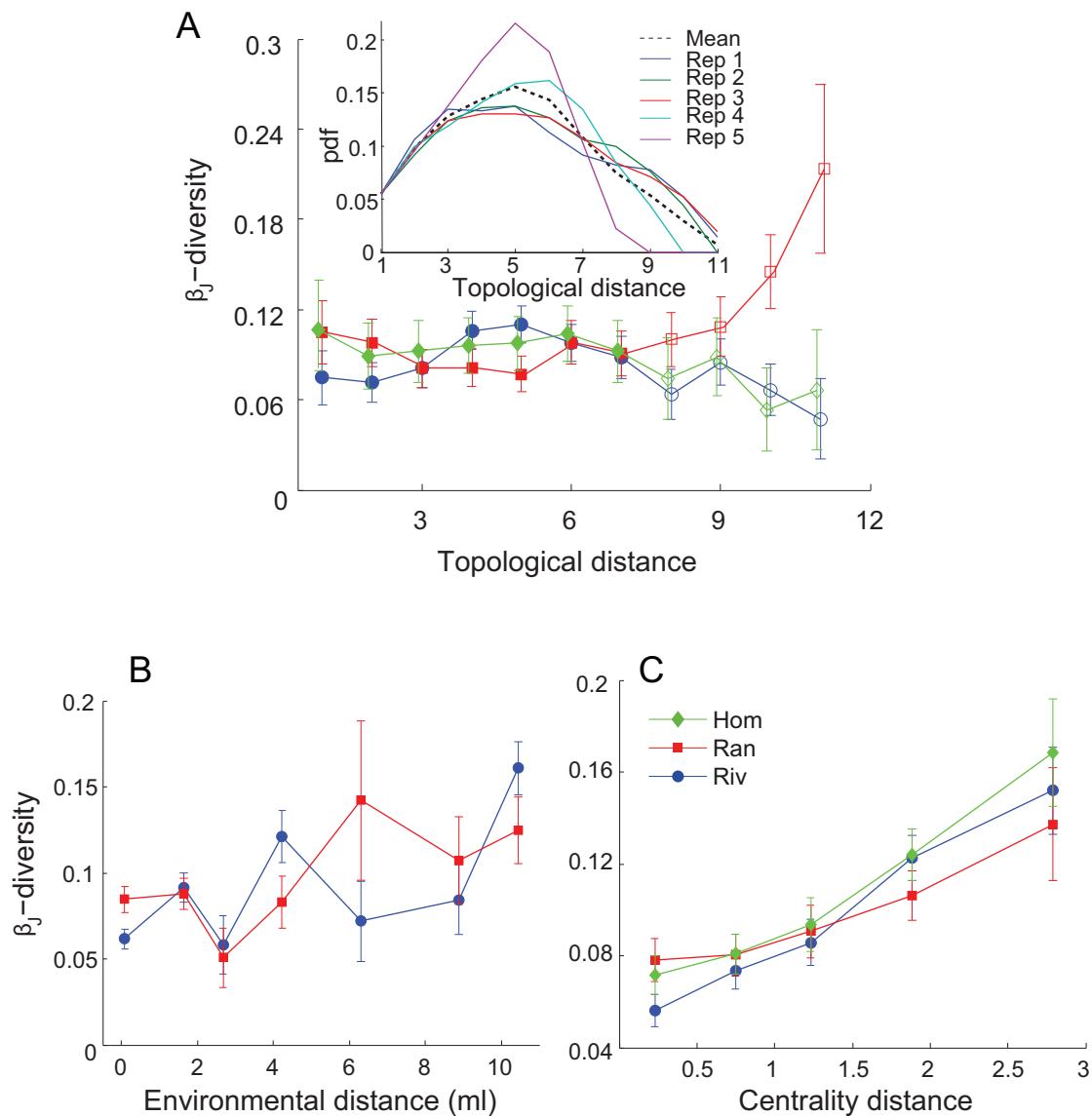
McGrady-Steed, J., P. M. Harris, and P. J. Morin. 1997. Biodiversity regulates ecosystem predictability. Nature 390:162–165.

## Appendix C from F. Carrara et al., “Complex Interaction of Dendritic Connectivity and Hierarchical Patch Size on Biodiversity in River-Like Landscapes”

(Am. Nat., vol. 183, no. 1, p. 13)

### $\beta$ -diversity Patterns

We performed Mantel tests (999 permutations, Kendall correlation) in order to get statistical analysis on  $\beta$ -diversity. The explanatory variables adopted in our analysis were the topological distance between community pairs (real distance along the network), the environmental distance, calculated as the difference in patch size between community pairs, and the centrality distance, calculated as the difference between the ecological diameters between community pairs (see “Material and Methods”). Figures 5 and C1 and table C1 show the dependance of  $\beta$ -diversity on these distances.



**Figure C1:**  $\beta$ -diversity as a function of the topological (A), environmental (B), and centrality (C) distance between local community

**Appendix C from F. Carrara et al., Biodiversity in River-Like Landscapes**

pairs, in Riverine (blue), Random (red), and Homogeneous (green) configurations of patch size. Symbols represent the mean  $\pm$  SEM of the  $\beta_j$ -diversity data over the five experimental replicates. Inset in A shows the probability density function (pdf) of topological distance for the five different river-network replicates adopted in the experiment. Open symbols in A and in figure 5A represent community pairs where topological distance is  $\geq 8$ . At these distances, pdf starts to decline and statistical power too.

**Table C1.** Mantel tests on  $\beta$ -diversity

Landscape, replicate	Topology	Environment	Centrality
	<i>r</i>	<i>r</i>	<i>r</i>
<b>Riverine:</b>			
1	.028	.22**	.21**
2	.002	.2**	.38**
3	.12**	.09	.14**
4	-.005	.32**	.1*
5	-.03	.32**	.29**
<b>Random:</b>			
1	.13**	.18*	.034
2	-.002	.13	.034
3	.05	.19**	.082
4	.04	.37**	.034
5	-.05	.17*	.11*
<b>Homogeneous:</b>			
1	.026	...	.09*
2	-.09	...	.11**
3	.023	...	.01**
4	-.11	...	.21**
5	-.05	...	.33**

Note: Environmental distance is always 0 by definition in Homogeneous landscapes (fixed patch size).

\* $P < .05$ .

\*\* $P < .01$ .