Small but mighty: headwaters are vital to stream network biodiversity at two levels of organization

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In this 8th article of the series, Debra S. Finn, Núria Bonada, Cesc Múrria, and Jane M. Hughes evaluate a combination of genetic and taxonomic diversity data from streams around the world and argue that small headwaters contribute substantially to regional-scale biodiversity via strong among-site variation across stream networks. Debra S. Finn is a Marie Curie International Fellow at the University of Birmingham where she studies climate change effects in glacier-influenced streams and focuses broadly on conservation ecology of headwaters that are particularly vulnerable to change. Núria Bonada is a tenure-track lecturer and member of the Freshwater Ecology and Management (FEM) research group at the University of Barcelona. Her research focuses on ecology and conservation of Mediterranean rivers and the large-scale patterns of their macroinvertebrate communities. Cesc Múrria is a post-doctoral researcher at Natural History Museum, London where he examines structure and distribution of diversity within species (population approach) and among species (communities approach) at macroecological scales. Jane M. Hughes is a Professor at the Australian Rivers Institute where she studies ecology and genetics of animal populations.

Small but mighty: headwaters are vital to stream network biodiversity at two levels of organization

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Abstract. Headwaters (stream orders 1–2) traditionally have been considered depauperate compared to mid-order streams (orders 3–4)—a conclusion that arises from a perception of streams as linear systems and emphasizes change in average α (local) diversity along streams. We hypothesized an opposite pattern for β (among-site) diversity and suggest that headwaters might account for a large degree of basin-scale biodiversity if considered within the more realistic framework of streams as branching networks. We assembled pre-existing biodiversity data from across the globe to test this hypothesis broadly at the population-genetic (mitochondrial haplotype diversity within species) and community (species/taxonomic diversity) levels, with a focus on macroinvertebrates. We standardized 18 (9 headwater and 9 mid-order) population-genetic and 16 (10 headwater and 6 mid-order) community-level ecoregional data sets from 5 global ecozones for robust comparisons of β-diversity estimates between the 2 stream-size categories.

At the population-genetic level, we applied measures of among-site variation commonly used at both population-genetic (FST and ΦST) and community (Sørensen’s dissimilarity with both presence/absence and abundance data) levels and developed a novel strategy to compare expected rates of loss of γ (regional) diversity as individual sites are eliminated sequentially from regions. At the community level, we limited analyses to Sørensen’s presence/absence measures. We found that Sørensen’s dissimilarity was significantly greater among headwaters than among mid-order streams at both population-genetic and community levels. We also showed that individual headwater reaches accounted for greater proportions of genetic γ diversity than did mid-order reaches. However, neither FST nor ΦST was significantly different between stream-size categories. These measures, which have been used traditionally for comparisons of population-genetic variation, measure proportions of total variation rather than solely among-site variation (i.e., they also are influenced by within-site variation). In contrast, Sørensen’s dissimilarity measures only...
Three decades ago, the influential River Continuum Concept (RCC; Vannote et al. 1980) modeled broad-scale spatial structure and function in streams in terms of gradual and predictable change from headwaters to mouth. One pattern predicted by the RCC was a unimodal distribution of biological diversity, with a peak in mid-order streams and lower values in headwaters and large rivers. Perhaps because small streams have been studied more extensively, the upper ½ of this predicted biodiversity pattern—an increase from headwaters to mid-orders—has been evaluated and supported in many stream types, particularly high-gradient mountain streams (Minshall et al. 1985, Ward 1986, Finn and Poff 2005, Sheldon and Warren 2009). These studies measured species or taxonomic diversity at the level of biological organization to which we refer henceforth as the community level. Of course, biodiversity includes multiple levels of biological organization (NRC 1999), and the pattern of increasing diversity with downstream distance from 1st to 5th-order streams also is apparent at the level of intraspecific genetic diversity (henceforth, the population-genetic level). In other words, for many common species distributed from headwaters to mid-order streams, genetic diversity within species increases from upstream to downstream (Shaw et al. 1994, Crispo et al. 2006, Hänfling and Weetman 2006, Watanabe et al. 2008).

We make the above observations under the linear model of streams embodied by the RCC. Therefore, this linear view considers headwaters to be relatively depauperate (at either biological level), contributing little to whole-stream biodiversity. However, streams are not linear. Streams are branching networks, and this alternative to linearity has been conceptualized in various ways during the past decade (Fagan 2002, Benda et al. 2004, Grant et al. 2007, Morrissey and de Kerckhove 2009). Fisher (1997) warned that the long-influential linear view of streams is “at best incomplete and at worst incorrect”. With respect to stream biodiversity distribution, we expect the best—that the linear view is incomplete (although potentially misleading). Biodiversity across any landscape (or riverscape) can be decomposed into 3 fundamental components: local (α) diversity, regional (γ) diversity, and some description of diversity turnover or variation from locality to locality within a region (β diversity) (Whittaker 1960, Anderson et al. 2011). Patterns of α diversity in streams potentially can be accommodated within the linear conceptualization (as above), but this view ignores β and γ components of diversity.

If we consider a branching-network view of streams, all 3 components should be essential to describe diversity distributions. A brief examination of a network reveals that small streams are much more numerous than large streams. Headwaters (1st- and 2nd-order streams) are thought to comprise >75% of total stream length in most basins (Leopold et al. 1964, Benda et al. 2004). Their abundance alone suggests that headwaters could contribute substantially to γ diversity in streams. Furthermore, the relative spatial isolation of headwaters in the tips of stream networks is expected to enhance β diversity at both community and population-genetic levels (Finn and Poff 2005, 2011, Muneeppeerakul et al. 2008, Hughes et al. 2009, Brown and Swan 2010). Headwaters also may exhibit greater among-stream differences in local habitat characteristics than do larger streams, a condition that is expected to promote β diversity (Lowe and Likens 2005, Meyer et al. 2007, Clarke et al. 2008).

A hypothesis of greater β diversity in headwaters than in mid-order streams has not been addressed directly, but some evidence supports the idea within limited biogeographical regions (Finn and Poff 2005, Hughes 2007). This pattern is opposite the pattern of α diversity expected along a linear gradient (Fig. 1), and it potentially exists at both community and population-genetic levels. If headwaters in general tend to be significantly more β-diverse than mid-order streams, a paradigm shift will become necessary in stream ecology, where headwaters have long been considered depauperate.

The outcome hypothesized in Fig. 1 also would have significant practical implications. Greater β diversity means that each locality (e.g., each branch in a stream network) contributes a larger proportion of γ diversity. A major implication of this pattern for
To quantify $\beta$ diversity. We also implemented a conservation-minded simulation that describes $\beta$ diversity as an expected average rate of erosion of $\gamma$ as local sites are lost from a region, and we compared these simulation results between the 2 stream-size categories.

**Methods**

**General approach**

For both population-genetic and community levels of diversity, we limited our analyses to the macroinvertebrate component of the stream biota. Invertebrates make up the bulk of macrofaunal diversity in streams, and they are well studied representatives of stream diversity. In many cases, the diversity patterns of invertebrates—particularly insects—are likely to represent the consequences of natural movement patterns, whereas other biota (e.g., fish) often include many introduced taxa. We also limited our analyses to nonamphidromous macroinvertebrates in an attempt to keep dispersal characteristics broadly similar among otherwise quite diverse taxa.

Across the globe, $\alpha$ and $\gamma$ diversity in streams can be expected to vary considerably with biogeographic histories and differing major environmental pressures (Vinson and Hawkins 2003, Clarke et al. 2008, Bonada et al. 2009). Cross-region comparisons of $\alpha$ or $\gamma$ are also troublesome because, at this scale, researchers typically must rely on data accumulated from several different sources that vary in methods, timing, intensity, level of taxonomic resolution, and spatial scales of resolution (Vinson and Hawkins 2003, Pecher et al. 2010). However, standardizing data sources to compare $\beta$ diversity among regions is more feasible. Because $\beta$ diversity in the general sense (Anderson et al. 2011) measures within-region variation, the main concern is to ensure similar sampling across all local sites within regions. Therefore, our strategy was to accumulate reliable data from headwaters or mid-order streams in many regions, standardize them, compare and evaluate the utility of some commonly applied $\beta$ diversity measures, and ask whether overall (among-region) means of these estimates differed between the 2 stream-size classes.

We evaluated biodiversity data at both community and population-genetic levels, but we chose to emphasize the population-genetic level. Genetic diversity is an important component of landscape-scale biodiversity (Crandon et al. 2000, Manel et al. 2003, Sork et al. 2010) and is a good choice here for 2 main reasons. First, $\beta$ diversity estimates generated from putatively neutral population-genetic data are expected to respond primarily to the effects of isolation among
local sites in a region. At the community level, assemblages probably respond to a combination of isolation and habitat heterogeneity among streams (Thompson and Townsend 2006, Heino and Mykrä 2008, Brown and Swan 2010). Effects measured at the population-genetic level will address the isolation hypothesis more directly, without a potential confounding effect of habitat differences. Second, an international industry standard (GenBank) exists for filing genetic-sequence data, which makes them easy to search and retrieve (Benson et al. 2005). Such a standard does not exist for community-level diversity data. Therefore, the population-genetic level provides practical feasibility that the community level is lacking to gather comparable data easily among many regions. These data can be sorted readily into highly specific regions of the genome, thereby allowing a further degree of data standardization (e.g., by characterizing diversity at homologous loci across multiple species and regions).

Data specifications: both levels of organization

To filter the myriad pre-existing data sets potentially useful for our project, we formulated a series of rules to identify the population-genetic and community-level data most appropriate to answer our overarching question. Essentially, these strict guidelines served the purpose of maximizing region-to-region and data set-to-data set comparability in terms of sample size, spatial extent, and quality. At the coarsest level, we created 3 priority classes for data selection. The highest priority was for data generated within our own laboratory groups or groups with which we had been directly associated as a student or researcher over the preceding 5 y. We did not require data in this priority class to have been published, but at a minimum they had to have been part of a graduate-level thesis. The 2nd priority class consisted of data generated by our close research colleagues, outside of our own laboratory groups. Any data considered in this or the following class we required to be published in a peer-reviewed journal. The 3rd priority class included data generated by any of our colleagues participating in population or community sessions at the 2009 North American Benthological Society meeting in Grand Rapids, Michigan.

After accumulating potential data sets from the prioritized sources, we applied more-detailed filters (Appendix 1). We wanted to adhere to a strict definition of headwaters and mid-order streams that was consistent across regions (see Clarke et al. 2008), so a key requirement was that geographic coordinates were available for each local collection site in a potential data set, and that high-resolution maps (≥1:25,000) were available to assign stream order accurately for each site. Following assignment of sites to stream orders, each regional data set was classified as either headwaters (1st–2nd order) or mid-order (3rd–4th order) or was split into 2 separate data sets according to stream size if sampling was sufficient (n ≥ 4 sampling sites/stream-size class) for both size categories. Any collection sites on stream orders >4 were removed from consideration. Henceforth, the term data set refers to a regional collection of samples, either all headwater or all mid-order in size, for which β diversity could be measured.

Another requirement imposed upon each data set was spatial independence of all sites (i.e., no 2 sites could be flow-connected; ver Hoef and Peterson 2010). After we sorted data sets into size classes, we culled any flow-connected sites, typically by randomly selecting one of the nonindependent sites for removal. In a few cases, culling was nonrandom to preserve the maximum number of sites in a data set. For example, if a 2nd-order collection site in a headwater data set was flow-connected to multiple 1st-order sites, the 2nd-order site was removed.

Increasing the geographic extent of sampling can increase β diversity estimates (Nekola and White 1999, Soininen et al. 2007), so we also imposed minimum and maximum spatial extents for each data set (Appendix 1), and we required that all sites in each data set were from a single terrestrial ecoregion (Olson et al. 2001). If a data set fell below the minimum size requirement, we removed it from further consideration. If a data set exceeded the maximum limit, we split it and assessed whether smaller subsets of sites could still meet the remaining requirements (Appendix 1). For example, data sets with sites from 2 ecoregions could be split by ecoregion and retained as 2 separate data sets. If a data set covered >50,000 km² surface area within an ecoregion, we culled it to achieve a smaller spatial extent while retaining the maximum number of sites.

Diversity estimates generated in our paper are likely to differ to some degree from those reported in original publications given the reorganization of the original data sets described above. We also imposed additional guidelines to standardize population-genetic and community data sets (described below).

Data specifications: population-genetic level

Population-genetic data for stream macroinvertebrate species has accumulated rapidly over the past decade, and a large proportion of these data, worldwide, has been in the form of sequence data
from the mitochondrial cytochrome c oxidase subunit I (COI) gene. To achieve robust cross-species and cross-region comparability while maintaining a reasonably large number of species and regions represented, we retained only population-genetic data sets composed of COI sequence data. Single-species collections for population-genetic analyses typically contain information from orders-of-magnitude fewer individuals than do community collections. In some lines of inquiry (e.g., phylogeography; Pauls et al. 2006) collection of just 1 or 2 individuals/sampling locality can be suitable. However, a robust measure of variability among sites (i.e., $\beta$ diversity), requires a reasonable estimate of $\alpha$ diversity for each site. Therefore, we required that enough individuals were sampled per site to provide an accurate representation of $\alpha$ diversity. We retained only those data sets with a mean of $n \geq 10$ individuals/site, with no site having $n < 7$. For COI, evidence for stream insects suggests that 7 to 10 individuals are sufficient to capture most of the genetic diversity present in a local population (Monaghan et al. 2009). In our study, sites with insufficient $n$ could be removed from some data sets and the smaller data set retained. Our final list for the population-genetic analyses consisted of 18 data sets: 9 headwaters and 9 mid-order (Table 1).

Analyses: population-genetic level

For each of the 18 data sets, we applied 4 different measures to characterize general $\beta$ diversity, including 2 traditionally used in population genetics ($F_{ST}$ and $\Phi_{ST}$) and 2 traditionally used in community ecology (Sørensen’s dissimilarity measures with either presence/absence or abundance information). We used the distribution of COI haplotypes as the raw input for each. The 2 categories provide 2 different perspectives on $\beta$ diversity.

Sørensen’s dissimilarity in its simplest form and with presence/absence data can be described as:

$$1 - 2W/(A + B)$$

where $W$ is the number of entities (haplotypes) shared between 2 samples, and $A$ and $B$ are the total number of haplotypes in each of the 2 samples. The maximum value is 1, which is achieved if the sites share no haplotypes. This metric can be converted easily for use with abundance data (McCune and Mefford 2006). We characterized presence/absence and abundance-based Sørensen’s dissimilarity for each data set by calculating the average value across all pairs of sites.

The metrics $F_{ST}$ and $\Phi_{ST}$ take a different approach to $\beta$ diversity, and both require abundance data because they evaluate frequency differences. A simple way to consider either $F_{ST}$ or $\Phi_{ST}$ for haplotype abundance data collected at 2 sites is as:

$$\Pi_{between} - \Pi_{within}/\Pi_{between}$$

where $\Pi$ represents the mean number of pairwise differences in haplotype (among individuals) either between or within the 2 sites. Like Sørensen’s dissimilarity, the maximum value of either $F_{ST}$ or $\Phi_{ST} = 1$. However, the maximum value is more difficult to reach in this case because $F_{ST}$ and $\Phi_{ST}$ are influenced by both among- and within-site variation (see also Jost 2008). Thus, to have an outcome of $F_{ST}$ or $\Phi_{ST} = 1$, not only must sites share no haplotypes (as with Sørensen’s), but also each site must have no polymorphism (i.e., each site must have just a single haplotype). The difference between $F_{ST}$ and $\Phi_{ST}$ is that the former accounts only for qualitative haplotype differences, whereas the latter incorporates information about evolutionary distance among haplotypes. For each of these 4 general measures of $\beta$ diversity, we used $t$-tests to compare means from headwater vs mid-order data sets.

We also were interested in the effect of local habitat loss on $\gamma$ diversity for each of the 2 stream-size categories. This issue is directly related to $\beta$ diversity because each local habitat is expected to contribute a larger proportion of $\gamma$ diversity in regions with greater $\beta$ diversity. We developed a novel approach to visualize the rate at which $\gamma$ would be lost if local habitats were removed sequentially from a region. We statistically compared these hypothetical loss rates for headwater vs mid-order data sets. We used regional haplotype richness as a measure of $\gamma$ diversity.

Achieving this objective involved direct comparisons of $\gamma$ diversity among diverse data sets, so it required 2 further standardization steps. First, we reduced all original COI sequence data to the length of the shortest segment in the original data sets (307 base pairs; Appendix 2). This step was necessary because longer sequences from the same genomic region yield more haplotypes identified/data set, rendering cross-species estimates of $\gamma$ richness incomparable. Second, we reduced all data sets to $n = 4$ sites, the minimum size (Appendix 1), by splitting data sets with $n > 4$ sites (Table 1, number of populations) into geographic quadrants and retaining from each quadrant the site with most individuals sampled. This step was necessary because more densely sampled regions are more likely to have reached the actual $\gamma$ value for the region, in a manner similar to reaching the asymptote in species-accumulation curves.
Equalization of the number of local sites in each region was the most effective way to generate comparable \( \gamma \)-diversity values across the data sets. For each of the reduced data sets, we simulated random, sequential loss of sites until a single site remained. We used linear regression to calculate the slope of the function relating number of sites remaining to total remaining regional richness (\( \gamma \) diversity). We ran this random-loss procedure 1000 to 10,000 times and recalculated the slope term each time. Henceforth, we refer to the average slope across all 1000 to 10,000 random-loss simulations for each data set as the slope of loss. The slope of loss allows visualization of the effects of local habitat loss on regional diversity, and its magnitude is an approximation of \( \beta \) diversity. We used a \( t \)-test to assess differences in slope of loss between headwater and mid-order data sets. We also tested for differences between mid-order and headwater Trichoptera data.

### Table 1. Population-genetic data sets used in our analyses. World Wildlife Fund (WWF) terrestrial ecoregions were taken from Olson et al. (2001). See Appendix 2 for index numbers. Number of populations is the number of local populations in our study (may be reduced from the number in the original studies to meet data-selection guidelines). Dispersal refers to the maximum capacity for terrestrial, among-stream dispersal. AA = Australasia, AT = Afrotropic, NA = Nearctic, NT = Neotropic, PA = Paleartic.

<table>
<thead>
<tr>
<th>Species</th>
<th>Order</th>
<th>WWF terrestrial ecoregion (ecozone)</th>
<th>No. populations</th>
<th>Dispersal</th>
<th>References</th>
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<tr>
<td><em>Elporia barnardi</em></td>
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<td>Montane fynbos and renosterveld (AT)</td>
<td>4</td>
<td>Fly</td>
<td>Wishart and Hughes 2003</td>
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<tr>
<td><em>Metacnephia coloradensis</em></td>
<td>Diptera</td>
<td>Colorado Rockies (alpine zone) (NA)</td>
<td>4</td>
<td>Fly</td>
<td>Finn and Adler 2006</td>
</tr>
<tr>
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<td>Colorado Rockies (alpine zone) (NA)</td>
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<td>Fly</td>
<td>Finn et al. 2006</td>
</tr>
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<td>Eastern Australian temperate forests (AA)</td>
<td>14</td>
<td>Fly</td>
<td>McLean et al. 2008</td>
</tr>
<tr>
<td><em>Abedus herberti</em></td>
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<td>24</td>
<td>Crawl</td>
<td>Finn et al. 2007</td>
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<tr>
<td><em>Allogamus uncatus</em></td>
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<td>Fly</td>
<td>Kubow et al. 2010</td>
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<td>Previšič et al. 2009</td>
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<td>Crawl</td>
<td>Cook et al. 2008</td>
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<td>Cook et al. 2007</td>
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<tr>
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<td>Fly</td>
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<td>Fly</td>
<td>Baker et al. 2003</td>
</tr>
<tr>
<td><em>Hydropsyche siltalai</em></td>
<td>Trichoptera</td>
<td>Northeastern Spain and southern France Mediterranean forests (PA)</td>
<td>7</td>
<td>Fly</td>
<td>Müría et al. 2010</td>
</tr>
</tbody>
</table>
sets separately because this group was the only major taxon represented by ≥2 species in both headwater (n = 4) and mid-order (n = 2) data sets.

**Data specifications and analyses: community level**

We assembled community-level data sets following **General data specifications** (above), and we included an additional requirement that data were sampled with consistent methods across all collection sites within a data set (Appendix 1). Consistent methods included samples collected within the same season(s), similar field collection methods (including types of microhabitat sampled), and the same taxonomic resolution applied for identification of specimens at all sites. We also ensured that no taxonomic redundancy existed in any data set (i.e., any taxa identified at multiple, potentially overlapping levels were either aggregated into higher taxonomic groupings or the less-resolved taxa were removed from analysis). Our final list for the community-level analyses consisted of 16 data sets: 10 headwater and 6 mid-order (Table 2).
We allowed the number of sites/data set to vary to some degree. However, we had particularly extensive data sets from 5 different ecoregions on the Iberian Peninsula. We reduced the more-extensive Iberian data sets by randomly retaining 1 site from each 4th-order basin so that these data sets were more comparable to data sets from the rest of the world (which included 4–10 sites). This procedure resulted in sample sizes of 7 to 10 sites/reduced Iberian data set (Table 2). Therefore, all community data sets had final \( n = 4 \) to 10 sites.

We calculated mean Sørensen’s dissimilarity (with presence/absence data only) for each community data set, and we used a \( t \)-test to assess whether means were different between headwater and mid-order data sets. Headwater and mid-order data sets from the same ecoregion were available for 4 ecoregions (1 Nearctic and 3 Palearctic; Table 2). For these data sets, we applied a paired \( t \)-test to test the hypothesis that \( \beta \) diversity (as Sørensen’s dissimilarity) shifted in a consistent direction from headwaters to mid-order streams within ecoregions. The Colorado Rockies ecoregion had 2 headwater data sets (Table 2), so we used their mean Sørensen’s dissimilarity as the headwater value for the paired \( t \)-test.

Results

Population-genetic level

The 18 data sets used for analyses at the population-genetic level spanned 5 terrestrial ecozones (Table 1). Australasia was represented in both headwater (\( n = 2 \)) and mid-order (\( n = 6 \)) groups, as were Nearctic (\( n = 3 \) and \( n = 1 \), respectively) and Palearctic (\( n = 3 \) and \( n = 1 \), respectively). The Afrotropic was represented by 1 headwater data set, and the Neotropic by 1 mid-order data set. All of the species represented had some capacity for terrestrial dispersal among streams. They generally could be classified into 1 of 2 dispersal subcategories, capable of flight or crawling (Table 1). Many (14 of 18) species were insects, only one of which lacked flight ability. The remaining 4 species were decapod crustaceans (1 crab and 3 shrimp), each capable of some degree of terrestrial crawling. The noninsects represented mid-order streams.

Comparisons of \( \beta \)-diversity measures between headwater and mid-order data sets had varying results (Fig. 2A–D). In general, the traditional population-genetic measures (\( F_{ST} \) and \( \Phi_{ST} \); Fig. 2A, B) revealed no difference and the traditional community measures
revealed significant differences between headwater and mid-order data sets (Fig. 2C, D). $F_{ST}$ ($p = 0.67$) and $\Phi_{ST}$ ($p = 0.24$) did not differ between headwater and mid-order data sets, although mean $\Phi_{ST}$ tended to indicate greater β diversity in headwaters. In contrast, β diversity measured as Sørensen’s dissimilarity for presence/absence and abundance data was significantly greater ($p = 0.004$ and $p = 0.007$, respectively) in headwaters than in mid-order streams.

We also found significant differences in the slope of loss of haplotypes between headwater and mid-order data sets (Fig. 3A, B). The slope of loss for headwater data sets ranged from 1.8 to 5.8 haplotypes/population lost (mean = 4.1, median = 4.3), and the slope of loss for the mid-order data sets ranged from 0.5 to 5.4 (mean = 1.8, median = 1.5) (Table 3). We obtained similar results when we restricted the analysis to include only Trichoptera (headwater mean = 3.7, mid-order mean = 1.2; $t$-test, $p = 0.005$). Mode of dispersal (crawl vs fly) had no effect on slope of loss among the species comprising our population-genetic data sets. The 4 mid-order, noninsect data sets had 4 of the 5 lowest slopes of loss (Table 3). We obtained similar results when we restricted the analysis to include only Trichoptera (headwater mean = 3.7, mid-order mean = 1.2; $t$-test, $p = 0.005$). 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Discussion

Outcomes at both levels of organization

We hypothesized decreasing $\beta$ diversity along a stream-size gradient from headwaters to mid-order streams, a pattern that is the inverse of that expected (and typically supported) for $\alpha$ diversity (Fig. 1). Results of our global-scale analysis of pre-existing diversity data at population-genetic and community levels supported our hypothesis when $\beta$ diversity was characterized strictly as among-site variation (in haplotypes or taxa) within a region. Typical measures used to characterize population-genetic structure ($F_{ST}$ and $\Phi_{ST}$), which incorporate both within- and among-site variation, did not reveal significant differences between headwater and mid-order streams, although $\Phi_{ST}$ tended to be higher in headwaters than in mid-order streams, a result suggesting increased evolutionary distance among headwater populations (see further discussion below).

All stream types have in common a branching network structure, so this outcome of greater $\beta$ diversity in headwaters in our global-scale study suggests that general rules could exist for predicting biodiversity distribution in streams (Rodriguez-Iturbe et al. 2009, Brown et al. 2011), potentially across multiple levels of biological organization. Hence, other major drivers like climate, flow regime, biogeographic history, or even assumed dispersal capacity appear to be outweighed by the simple influence of differences in position within a network (tips vs middle branches, in this case) on patterns of $\beta$ diversity.

Networks are unique spatial structures in which tips are always more isolated from one another than are interior branches. Therefore, spatial isolation probably is a key mechanism underlying the observed patterns of $\beta$ diversity. Isolation also might explain the greater effect size for mean Sørensen’s dissimilarity in the population-genetic data sets than in the community-level data sets (Fig. 2C, D, Fig. 4; note differences in $y$-axes). Spatial patterns of putatively neutral population-genetic data are expected to reflect primarily gene flow, which is strongly influenced by spatial isolation (Hughes et al. 2009). Community-level $\beta$ diversity also can be influenced by isolation (e.g., Thompson and Townsend 2006, Munepeerakul et al. 2008, Bonada et al. 2009, Brown and Swan 2010), but community structure also responds strongly to habitat drivers (i.e., species sorting; Brown and Swan 2010, Brown et al. 2011). Indeed, habitat heterogeneity could be a secondary mechanism to increase $\beta$ diversity in headwaters because among-stream heterogeneity might be greater in the tips than interior of stream networks (Gomi et al. 2002, Lowe and Likens 2005, Meyer et al. 2007). These effects should be detectable primarily at the community level. Various statistical approaches (e.g., Borcard et al. 1992, Cottenie 2005, Thompson

<table>
<thead>
<tr>
<th>Data set (as species)</th>
<th>$F_{ST}$</th>
<th>$\Phi_{ST}$</th>
<th>Mean Sørensen’s dissimilarity (+/−)</th>
<th>Mean Sørensen’s dissimilarity (abundance)</th>
<th>Slope of loss (no. haplotypes/population)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Headwater</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td><em>Elporia barnardi</em></td>
<td>0.53</td>
<td>0.86</td>
<td>1.00</td>
<td>1.00</td>
<td>4.3</td>
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<td>0.17</td>
<td>0.76</td>
<td>0.74</td>
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</tr>
<tr>
<td><em>Prosimulium neomacropyga</em></td>
<td>0.28</td>
<td>0.36</td>
<td>0.66</td>
<td>0.62</td>
<td>3.6</td>
</tr>
<tr>
<td><em>Bungona narilla</em></td>
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<td>0.22</td>
<td>0.67</td>
<td>0.74</td>
<td>4.3</td>
</tr>
<tr>
<td><em>Abedus herberti</em></td>
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<td>0.52</td>
<td>0.77</td>
<td>0.84</td>
<td>4.0</td>
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<td>0.89</td>
<td>0.85</td>
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<td>0.91</td>
<td>0.91</td>
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<tr>
<td><em>Drusus discolor</em></td>
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<td>1.8</td>
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<td>Mid-order</td>
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<td></td>
<td></td>
<td></td>
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<td><em>Epilobocera simatifrons</em></td>
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<td>0.32</td>
<td>0.57</td>
<td>0.66</td>
<td>1.5</td>
</tr>
<tr>
<td><em>Paratya australiensis</em> sp. 1</td>
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<td>0.66</td>
<td>0.70</td>
<td>0.72</td>
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<td>0.20</td>
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<td><em>Paratya australiensis</em> sp. 8</td>
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<td>0.53</td>
<td>0.59</td>
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<tr>
<td><em>Atalophlebia</em> sp. AV13 A</td>
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<td>0.00</td>
<td>0.71</td>
<td>0.64</td>
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</tr>
<tr>
<td><em>Atalophlebia</em> sp. AV13 D</td>
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<td>0.6</td>
<td>0.74</td>
<td>0.60</td>
<td>1.9</td>
</tr>
<tr>
<td><em>Baetis bicaudatus</em></td>
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<td>0.12</td>
<td>0.49</td>
<td>0.48</td>
<td>1.7</td>
</tr>
<tr>
<td><em>Cheumatopsyche</em> sp. AV1</td>
<td>0.02</td>
<td>0.03</td>
<td>0.46</td>
<td>0.52</td>
<td>1.8</td>
</tr>
<tr>
<td><em>Hydropsyche siltalai</em></td>
<td>0.10</td>
<td>0.07</td>
<td>0.16</td>
<td>0.22</td>
<td>0.5</td>
</tr>
</tbody>
</table>
and Townsend 2006) exist to disentangle the potentially interactive effects of habitat heterogeneity and spatial isolation.

The population-genetic level

Increased spatial isolation probably was the key mechanism behind our observation of increased $\beta$ diversity among headwater populations, and at its extreme, genetic isolation of populations can lead to allopatric speciation. Indeed, one of the headwater data sets we evaluated (Elporia barnardi; Wishart and Hughes 2003) attained the maximum values for both Sørensen’s measures (Table 3) because none of the independent populations that we evaluated shared any haplotypes. This pattern of maximum $\beta$ diversity might indicate incipient speciation in this blepharicerid occupying isolated headwater streams in South African mountain ranges. Seidel et al. (2009) set out to quantify population-genetic patterns of isolated Gammarus pecos populations in Chihuahuan desert headwaters of western North America and instead found cryptic species in each drainage they sampled. Similarly, Myers et al. (2001) found strong population-genetic structure among headwater caddisfly populations across the Great Basin of North America and recommended that different subbasins should be regarded as different management units. Thus, $\beta$ diversity is particularly pervasive among headwaters in harsh environments, such as deserts or high altitudes, which presumably impede among-stream gene flow (see also Schultheis et al. 2002, others in Table 1).

The differences between measures traditionally used in population-genetic and in community-level studies (Fig. 2A, B vs Fig. 2C, D) revealed the contrasting approaches used to quantify biological variation in the 2 disciplines. The population-genetic measures ($F_{ST}$ and $\Phi_{ST}$) did not show significant differences between headwaters and mid-order streams across the diverse data sets we evaluated. These measures account simultaneously for among- and within-site variation and, therefore, are more literally fixation indices rather than measures of among-population differentiation (Jost et al. 2010). Indeed, any putative measure of $\beta$ diversity that is dependent on $\alpha$ diversity to some degree is not robust for comparing regions with differing $\alpha$ diversity (Jost 2007). Alternatively, Sørensen’s dissimilarity measures only among-site variation and, therefore, can be compared directly among disparate regions. Thus, $F_{ST}$ and $\Phi_{ST}$ generally are not suitable for comparing $\beta$ diversity at either biological level of organization. Sørensen’s distances are more appropriate, and other related alternatives have been
developed specifically for genetic diversity data (e.g., Jost 2008).

From a conservation perspective, increased β diversity essentially means that individual localities each account for a greater proportion of regional-scale biodiversity. Hence, a prudent management choice might be to emphasize conservation in regions (or sections of stream networks) with high β diversity. Our slope-of-loss approach provided a way to visualize differences in the average contribution of local stream reaches to γ diversity between stream-size categories (Fig. 3A, B). The slope-of-loss values in our analysis can be interpreted as additive measures of β diversity (Lande 1996) because they are directly related to the difference between γ and mean α diversity. However, the key utility of the slope-of-loss approach lies in its ability to visualize the effect of sequential loss of local habitats. The results of the comparison of slope of loss between headwaters and mid-order data sets are qualitatively similar to the comparison of mean Sørensen’s dissimilarity measures. The results suggest that effective conservation action in headwaters probably would retain valuable sources of biodiversity at the scale of whole stream basins.

Information on genetic diversity for species occupying both headwaters and mid-order streams within the same ecoregion was missing from our analysis. An important next step in testing whether β diversity is consistently greater in headwaters than in mid-order reaches will be to design studies to compare populations of the same species in mid-order reaches and their headwaters. Tests of the hypothesis (Fig. 1) with individual species occurring from 1st to 4th order streams would provide an added element of control that is missing from our current analysis. Several of the species used to represent headwaters in our study probably were headwater specialists (e.g., Wishart and Hughes 2003, Finn et al. 2006, 2007, Múrria and Hughes 2008), whereas many of the mid-order representatives were more widespread species. Decreased dispersal is thought to be a beneficial trait for habitat specialists (Hughes et al. 2009), particularly those that specialize on temporally stable habitat types (Roff 1990), and these differences could provide some explanation for our observations of higher β diversity among headwater data sets. Indeed, headwaters probably inherently harbor more habitat specialists (e.g., Gomi et al. 2002, Meyer et al. 2007) than mid-order streams. If this is the case, population-genetic β diversity would be expected to be greater in headwaters than in mid-order streams for the combined reasons of increased habitat specialization and increased spatial isolation.

The community level

The community-level analyses were more limited than the intensive population-genetic analyses. The number of data sets from headwaters and mid-order streams was comparable to values of β diversity in both mid-order and headwater streams that were comparable to values of β diversity in the headwater (but not mid-order) data sets for the Colorado Rockies. In general, Iberian streams are quite diverse (Bonada et al. 2009), and high β diversity might occur in both headwaters and mid-order streams because of an atypical biogeographic history (Soininen et al. 2007, Valladolid et al. 2007).

Our results, although relatively restricted, call for more research within and among other types of ecoregions. Increased habitat specialization could be more effective in headwaters than in mid-order streams, and the volume of existing community data was daunting. However, our demonstration of an overall pattern of increased β diversity in headwaters vs in mid-order streams (Fig. 4) provides proof-of-concept, and the volume of existing community data suggests that further addressing our overarching question should be feasible in the near future.

Conclusions

Differences in stream size alone explained significant variation in β diversity at community and population-genetic levels in streams across the world, although additional variables, such as altitude, watershed topography, and habitat heterogeneity probably would further increase explanatory power. Therefore, stream network structure appears to influence population-genetic and community structure in similar ways, and future studies that evaluate both biological levels
simultaneously should be instructive. Evaluation of species- and genetic-diversity patterns across the same landscape provides increased biogeographic understanding beyond that achieved from either individual level in streams (Bonada et al. 2009, Sei et al. 2009, Finn and Poff 2011) and in other environments (e.g., Cleary et al. 2006, Evanno et al. 2009).

We intentionally excluded large streams (≥ 5th order) from our analysis for the simple reason that not enough consistently collected macroinvertebrate data exist for analysis of β diversity at these stream sizes. The linear framework of the RCC predicts a decrease in β diversity in large streams relative to mid-order streams, and some evidence supports this prediction (Minshall et al. 1985, Watanabe et al. 2008). However, in their natural state, large streams often produce extensive floodplains, which are complex habitat mosaics that typically support high biodiversity (Ward 1998, Ward et al. 1999, Arscott et al. 2005). These large river systems probably contribute substantially to regional-scale biodiversity in streams. Development of a consistent means to incorporate diversity information from large streams would be useful for testing ideas about broad-scale distribution of diversity throughout entire stream networks.

A major implication of our analyses of β-diversity patterns in small-to-mid-order streams is that aquatic ecologists need to change the way we think about the contribution of headwaters to whole-stream biodiversity. The linear view of streams implied that headwater streams were depauperate and, therefore, potentially more expendable than higher-order streams. A more realistic view of streams as networks demonstrates the opposite pattern. Moreover, recent modeling exercises suggest that headwaters might play an important role in actively sustaining biodiversity across many stream sizes (Morrisey and de Kerckhove 2009). These findings demand that headwaters acquire a more pressing conservation status because each small branch lost represents the loss of unique diversity in a river network.

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Literature Cited


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**Appendix 1. Guidelines for assembling population-genetic and community data sets. Refer to text for more details.**

**Guidelines for data sets**

General guidelines (either organizational level):
1. Stream order for each sample site can be determined from high-quality, high-resolution (at least 1:25,000) topographic maps
2. At least 4 collection sites consisting of either all headwater (1st–2nd order) or all mid-sized (3rd–4th order) stream samples
3. No 2 sites connected by flow
4. Minimum spatial extent: at least two 4th-order basins
5. Maximum spatial extent: 50,000 km², same terrestrial ecoregion

Guidelines specific to population-genetic data sets:
1. Individuals from each site sequenced at the mitochondrial cytochrome c oxidase subunit I (COI) gene
2. Mean $n \geq 10$ individuals sequenced per site; $n \geq 7$ individuals from any 1 site

Guidelines specific to community data sets:
1. Taxonomic richness of each site determined with comparable collection and identification methods

**Appendix 2. Addendum to Table 1: information on population-genetic data sets listed by species name. World Wildlife Fund (WWF) index number identifies the ecoregions listed by name in Table 1. COI = cytochrome c oxidase subunit I, bp = base pair.**

<table>
<thead>
<tr>
<th>Stream-size category</th>
<th>Species</th>
<th>Family/order</th>
<th>WWF index number</th>
<th>No. COI bp (original data sets)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Headwater</td>
<td><em>Elporia barnardi</em></td>
<td>Diptera/Blephariceridae</td>
<td>AT1203</td>
<td>641</td>
</tr>
<tr>
<td></td>
<td><em>Metacnephia coloradensis</em></td>
<td>Diptera/Simuliidae</td>
<td>NA0511</td>
<td>307</td>
</tr>
<tr>
<td></td>
<td><em>Prosimulium neomacropyga</em></td>
<td>Diptera/Simuliidae</td>
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<td>307</td>
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<tr>
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<tr>
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<tr>
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<tr>
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<td></td>
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<td>Trichoptera/Hydropsychidae</td>
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<td>614</td>
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</table>
**Appendix 3.** Addendum to Table 2: information on community-level data sets (in the same order as in Table 2). World Wildlife Fund (WWF) index number identifies the ecoregions listed in Table 2. Taxonomic resolution: highest possible = species–family level, as per original publications (with no redundancy). Total no. taxa = taxon richness across the sites included in our study and at the level of resolution listed.

<table>
<thead>
<tr>
<th>Stream size category</th>
<th>Community description</th>
<th>WWF index number</th>
<th>Taxonomic resolution</th>
<th>Total no. of taxa</th>
</tr>
</thead>
<tbody>
<tr>
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<td>Insects in riffles, high-flow season</td>
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