

Effects of patch connectivity and heterogeneity on metacommunity structure of planktonic bacteria and viruses

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Abstract

Dispersal limitation is generally considered to have little influence on the spatial structure of
25 biodiversity in microbial metacommunities. This notion derives mainly from the analysis of
spatial patterns in the field, but experimental tests of dispersal limitation using natural
communities are rare for prokaryotes and, to our knowledge, non-existent for viruses. We
studied the effects of dispersal intensity (3 levels) and patch heterogeneity (2 levels) on the
structure of replicate experimental metacommunities of bacteria and viruses using outdoor
30 mesocosms with plankton communities from natural ponds and lakes. Low levels of dispersal
resulted in a decrease in the compositional differences (beta diversity) among communities
of both bacteria and viruses, but we found no effects of patch heterogeneity. The reductions
in beta diversity are unlikely to be a result of mass effects and only partly explained by
indirect dispersal-mediated interactions with phytoplankton and zooplankton. Our results
35 suggest that even a very limited exchange among local communities can alter the trajectory
of bacterial and viral communities at small temporal and spatial scales.

Keywords: viruses; bacteria; metacommunity; experiment; dispersal limitation; patch
heterogeneity

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Introduction

A metacommunity is defined as a group of local communities connected via dispersal of
45 interacting species (Leibold *et al.*, 2004). Metacommunity theory integrates the study of
interactions between organisms and their local biotic and abiotic environment with spatial
dynamics at the regional scale. The theory makes a variety of predictions about the spatial
structure of biodiversity that depend on patch connectivity, patch heterogeneity, and niche
differences among species. In neutral metacommunities, where organisms are assumed to
50 have identical niches and local habitat patches are the same, increased connectivity among
patches will counteract the effects of ecological drift and lead to metacommunity
homogenization, resulting in a reduction of compositional differences among communities in
different patches (i.e. 'beta diversity'). In niche-based metacommunities, patch quality is not
spatially homogenous, species occupy different niches and spatial patch heterogeneity can
55 enhance beta diversity (Chesson, 2000). However, in such cases, the impact of patch
heterogeneity on metacommunity structure will depend on the degree of dispersal among
communities. At very low rates of dispersal, species may not reach suitable habitat patches,
resulting in species-poor communities with vacant niches and a weak match between species
composition and environmental conditions. Higher dispersal rates may therefore increase
60 beta diversity in a heterogeneous landscape. At very high dispersal rates, mass effects can
lead to the predominance of regionally superior competitors and result in a decline of beta
diversity (Mouquet and Loreau, 2003). Dispersal-mediated ecological interactions can also
add an additional level of complexity because patch connectivity and heterogeneity may
differentially affect metacommunity structure of organism groups belonging to different but
65 interacting trophic levels (Verreydt *et al.*, 2012).

The prevailing view is that the geographic distribution and community composition of microbial organisms is largely unaffected by dispersal limitation, which is consistent with the idea that everything is everywhere, and the environment selects. The small size of microbes allows for wide dispersal and large population sizes make extinction unlikely at local scales (Fenchel and Finlay, 2004). Furthermore, some populations can recruit from dormant cells, allowing them to grow rapidly to high densities when suitable conditions arise (Jones and Lennon, 2010). Such characteristics allow local dynamics to play a dominant role in structuring microbial metacommunities and reduce the likelihood that dispersal limitation will influence spatial patterns of microbial diversity (Van der Gucht *et al.*, 2007; De Bie *et al.*, 2012). These predictions are consistent with the wide distribution patterns of many microbial taxa (Short and Suttle, 2002; Fenchel and Finlay, 2004; Short and Suttle, 2005) and strong associations between community composition and environmental gradients (Beisner *et al.*, 2006; De Bie *et al.*, 2012). However, studies covering a broad range of spatial scales, varying from widely separated extreme habitats (Papke *et al.*, 2003; Whitaker *et al.*, 2003), stream networks (Heino *et al.*, 2010), rock pool clusters (Langenheder and Ragnarsson, 2007), and individual salt marshes (Martiny *et al.*, 2011) suggest that microbial dispersal can be too slow to counteract community differentiation resulting from ecological drift. In light of such mixed evidence from field studies, there is a strong need for experimental tests of dispersal limitation in microbial metacommunities.

In the current study, we tested whether low levels of dispersal can influence the compositional changes of microbial metacommunities in response to spatial variation in environmental conditions. In a neutral scenario, we would expect no effects of dispersal or environmental heterogeneity on beta diversity. In a niche-based scenario, however, we would expect an interaction between dispersal and patch heterogeneity, whereby dispersal

90 enhances community differentiation in metacommunities with heterogeneous compared to
homogenous environments. To test these ideas, we performed an outdoor mesocosm
experiment using plankton from natural ponds and lakes in British Columbia (Canada). We
created replicate experimental metacommunities with different levels of patch heterogeneity
by adding different amounts of nutrients to mesocosms. We analyzed the changes in the
95 metacommunity structure of both bacteria and viruses in response to different levels of
dispersal intensity. We chose very low exchange rates of organisms among communities in
the dispersal treatments to ensure that dispersal effects on the trajectories of community
composition reflect demographic responses of dispersed phylotypes to environmental
conditions rather than merely being the result of metacommunity homogenization due to
100 mixing. We also tested if changes in metacommunity structure of bacteria and viruses that
we observe in response to the dispersal treatments can be attributed to direct dispersal
effects or to indirect interactions with other organisms (e.g. phytoplankton, zooplankton)
that may also be constrained by dispersal themselves (Verreydt *et al.*, 2012).

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Material and methods

Experimental design

'Metacommunities' consisted of pairs of plastic containers (280 L) outdoors that were
110 connected through the dispersal of planktonic organisms. The experimental design consisted
of the cross-factorial combination of 'patch heterogeneity' and 'dispersal intensity' (Figure
1). Patch heterogeneity was created through the addition of nutrients in either the same
(homogenous metacommunities) or different amounts (heterogeneous metacommunities)

between the two containers. Homogenous metacommunities (HOMO) received an identical,
115 low amount of nutrients only at the start of the experiment (NUT_L). Heterogeneous
metacommunities (HETERO) were created by the repeated addition of a high amount of
nutrients to one of the containers (NUT_H), while no nutrients were added to the other
(NUT_N). The dispersal treatment consisted of three levels of exchange rates of planktonic
organisms among containers: no dispersal (D0), low dispersal (DL) and high dispersal (DH)
120 rates. All multifactorial combinations of dispersal intensity and nutrient heterogeneity were
replicated four times, according to a randomized block design (Figure 1), with blocks
corresponding to the spatial grouping of the containers on the experimental terrain.

Microbial and zooplankton communities were introduced into containers at the
beginning of the experiment by adding water and sediments from lakes near Vancouver,
125 Kelowna, Squamish and Victoria, BC, Canada. Within blocks, all NUT_H- and half of the NUT_L
containers were inoculated with a water and sediment mixture of three lakes, whereas all
NUT_N and the remaining NUT_L containers received inoculations of three other lakes (see
Appendix 1 for details on the inoculation design).

The bottom of the containers was covered with coarse sand and filled with municipal
130 drinking water on 28 March 2008. Lake water and sediments were collected during April
2008. The lake water was added to the containers as soon as it was brought from the field.
Sediment samples were kept in the dark at 4°C until 200 g from each lake was added to each
container on 9 May 2008. On 16 May 2008, we added nutrients (nitrogen and phosphorus)
according to the Redfield ratio to the NUT_L (0.34 µM KH₂PO₄ and 5.48 µM NaNO₃ final
135 concentrations) and NUT_H containers (17 µM KH₂PO₄ and 274 µM NaNO₃ final
concentrations). Due to experimental problems, we had to restart the NUT_H containers on 16
June. The NUT_H containers were emptied, cleaned, refilled with tap water, and nutrients

were added again (same concentrations as initially). On 1 July, we re-inoculated these tanks with water from the corresponding NUT_L tanks with the aim of providing them with the same initial plankton community composition as in the NUT_L tanks. Nutrient additions were then repeated weekly with 3.4 μM KH₂PO₄ and 55 μM NaNO₃ to the NUT_H containers and 0.17 μM KH₂PO₄ and 2.74 μM NaNO₃ to the NUT_L containers, until 20 August 2008. The dispersal treatment was started on 13 July, and continued on a weekly basis until 19 September. We exchanged 0.025 L among DL-containers and 1L among DH-containers, corresponding to 0.009 and 0.35 % of the container volume, respectively. These volumes also contained zooplankton collected from 8 and 80L, respectively, with a Schindler-Patalas trap (mesh size: 64 μm). To keep the same level of physical disturbance in all treatments, the same actions were repeated for D0, but without exchanging organisms or water among containers. Containers were covered by a mosquito net (mesh size: 1 mm) during the entire experiment in order to prevent insects and birds from entering.

Sampling and sample analysis

We took samples of bacterial, viral, and zooplankton communities at two occasions, once just before the start of the dispersal treatment (10 - 11 July 2008) and once at the end of the experiment (29 September - 1 October). In addition, during April 2008, we sampled the microbial communities (bacteria and viruses) of all the lakes that we used as inocula for the experiment (the 'source lakes').

We estimated prokaryotic and viral abundances with flow cytometry. We determined phytoplankton chlorophyll *a* concentrations fluorometrically and measured dissolved organic carbon (DOC) by filtering a water sample through ashed GF/F filters (Whatman) which were then analyzed on a Shimadzu 5000 TOC analyzer. We assessed bacterial community

composition using terminal restriction fragment length polymorphism (T-RFLP; Moeseneder *et al.*, 1999) and viral community composition with a randomly-amplified polymorphic DNA polymerase chain reaction (RAPD-PCR; Winget and Wommack, 2008). Samples of crustacean zooplankton were counted and identified to the genus level using a stereo microscope. We refer to Appendix 2 for a more detailed description of sampling protocols and analysis procedures.

Data analysis

The central aim of our study was to investigate the response of beta diversity in bacterial and viral metacommunities to nutrient heterogeneity and dispersal intensity. For a metacommunity at a given sampling day, we quantified spatial beta diversity with the Jaccard dissimilarity index. Jaccard dissimilarity is the complement of Jaccard similarity, the latter equaling the ratio of the number of operational taxonomic units (OTUs) occurring in both patches of a metacommunity and the total number of OTUs observed in that metacommunity. The Jaccard dissimilarity thus reflects the degree to which communities differ in their observed OTU composition and varies from 0 (equal OTU composition) to 1 (no OTUs shared). We used two-way ANOVA to evaluate the interaction effect of nutrient heterogeneity and dispersal intensity on spatial metacommunity beta diversity. In case of significant effects, we applied paired t-tests to further explore differences among treatment levels.

Treatment effects on phytoplankton and zooplankton communities in our experiment may have indirectly affected the metacommunity structure of bacteria and viruses (Verreydt *et al.*, 2012). For example, heterogeneous nutrient addition may have caused more heterogeneous metacommunities of interacting species (e.g. zooplankton, phytoplankton,

viruses) that indirectly increased bacterial beta diversity relative to the homogenous nutrient addition regime. Similarly, the dispersal treatment may have homogenized the biotic environment, contributing to a stronger decrease in beta diversity of bacteria than would have been expected from the physical exchange of these organisms alone. To evaluate the importance of such indirect effects (Verreydt *et al.*, 2012), we measured the degree of biotic heterogeneity in metacommunities (i) by calculating ΔCHLA , as $\log_{10} (|\text{chla}_1 - \text{chla}_2| + 1)$, where chla_1 and chla_2 represent the chlorophyll *a* concentrations in two containers of the same metacommunity at 19 September 2008, and (ii) by calculating the Bray-Curtis distance using the zooplankton species abundance matrix (BC_{zoop}). We incorporated ΔCHLA and BC_{zoop} in general linear models explaining the bacterial and viral beta diversity in order to evaluate the direct effects of the experimentally manipulated dispersal and nutrient heterogeneity independently of any potential indirect effects.

We used two-way ANOVA to test for the effect of nutrient addition and dispersal intensity on abundances of prokaryotes, viruses, and zooplankton, and on concentrations of chlorophyll *a* and DOC. Similarly, we applied distance-based redundancy analysis (db-RDA; Appendix 4) to evaluate the effect of the experimental treatments on the community composition of the bacterial and viral communities.

With Mantel tests, we investigated pair wise associations in the community composition of bacteria, viruses and zooplankton. In these analyses, we used Jaccard dissimilarity matrices derived from the presence-abundance matrices of bacteria and viruses and the Bray-Curtis dissimilarity matrix calculated from the zooplankton species abundance matrix.

To evaluate the strength of the temporal dynamics of bacterial and viral community composition, we calculated temporal beta diversity for communities within each mesocosm

210 by calculating the Jaccard dissimilarity index on community data from the start and end of
the experiment (July and September).

Results

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Effects of dispersal and nutrient addition on metacommunity structure of bacteria and viruses

The dispersal treatment affected the beta diversity of both bacterial and viral metacommunities. At the end of September, beta diversity of bacteria was significantly lower
220 in DL- and DH- than in D0-treatments (Figure 2a, Table 1). Thirty percent of the observed OTU's were shared by both mesocosms in the DL-and DH-metacommunities, compared to 15% in D0-metacommunities (an average of, respectively, 3 and 1.5 OTUs on a total of c. 10 OTUs). Beta diversity of viruses was lower in the DH-treatment than in the D0-and DL-treatments (Figure 2b, Table 1): mesocosms of the DH-treatment shared 9.5% of the
225 observed OTUs, compared to 4% in D0- and DL-treatments (an average of, respectively, 1.6 and 0.8 OTUs on a total of c. 18 OTUs). Heterogeneity in nutrient additions had no effect on the beta diversity of bacteria or virus metacommunities (Figure 2; Table 1). Blocks significantly explained variation in bacterial and viral beta diversity (Table 1).

Abundances of prokaryotes and viruses were higher under conditions of strong
230 nutrient enrichment compared to low or no nutrient additions (Appendix 3). Redundancy analyses revealed effects of nutrient addition, dispersal and inoculation origin on the community composition of bacteria (Appendix 4). Viral community composition was affected by a dispersal x nutrient effect and an inoculation effect (Appendix 4).

235 **Effects of treatments on phytoplankton and zooplankton heterogeneity**

The experimental treatments also affected the heterogeneity of other biota in the metacommunities. Bray-Curtis distances calculated from zooplankton abundance data (BC_{zoop}) were lower in the DL- and DH- than in the D0-treatments and were higher in heterogeneous than in homogenous metacommunities at the end of the experiment (Figure 240 3a, Table 2; see also Appendix 3 for more details on treatment effects on population densities). We found no effects of dispersal intensity and heterogeneity in nutrient addition on $\Delta CHLA$ -values (Figure 3b, Table 2).

Assessment of indirect treatment effects through trophic interactions

245 BC_{zoop} was positively correlated with beta diversity in metacommunities of bacteria ($r=0.47$, $p=0.02$, Appendix 5) but not of viruses ($r=0.32$, $p=0.137$, Appendix 5). $\Delta CHLA$ was positively correlated with beta diversity of bacteria ($r=0.44$, $p=0.031$, Appendix 5), and, a similar but marginal trend was also observed for viruses ($r=0.36$, $p=0.058$, Appendix 5). According to GLM-analyses, the effect of the dispersal treatment on the beta diversity of bacteria and 250 viruses was significant even with the inclusion of $\Delta CHLA$ and BC_{zoop} in the model (Table 3). Beta diversity of bacteria did not contribute to the explanation of variation in the beta diversity of viruses or vice versa (Table 3).

Community concordance patterns

255 Mantel correlation analyses revealed no association between the bacterial and viral communities in either July or September, but a significant association of bacteria with

zooplankton community composition in both July ($r=0.015$, $p=0.001$) and September ($r=0.04$, $P=0.001$).

260 **Distribution of bacterial and viral OTU's across time and metacommunities**

Bacteria and viruses exhibited contrasting patterns of diversity across the array of mesocosms. The number of distinct OTUs observed for bacteria (N=88 OTUs) were much lower than for viruses (N=140 OTUs). The number of detected bacterial OTUs declined by 65% over two months (from 82 OTUs in July to 29 OTUs in September), leaving a nested
265 subset of the original species pool at the end of the experiment, with only 6 new OTUs found in September. On both dates several of the bacterial OTUs were widespread across the experiment, with more than half of the OTUs occurring in more than 5 mesocosms (Appendix 6). In contrast, the total number of detected viral OTUs increased by 30% (from 99 OTUs in July to 124 in September, with 40 new OTUs in September), with viral OTUs being more
270 sparsely distributed than those for bacteria. At both time points, <20% of the original viral OTUs occurred in five or more mesocosms, and no OTU occurred in more than 12 of the mesocosms (Appendix 6).

We observed strong temporal dynamics in bacterial and viral community composition within local patches, as indicated by low Mantel correlations between July and September
275 communities (bacteria: $r_{sp}=0.1$; $p=0.021$; viruses: non-significant). The average number of bacterial OTUs per local community was 21 in July and 6 in September, and temporal beta diversity amounted to 0.85. The average number of viral OTUs per container was 6.6 in July and 9.6 in September, and temporal beta diversity equaled 0.96. Despite strong temporal
280 dynamics, the bacterial communities within each patch shared a considerable proportion of OTUs with the original inocula from the source lakes. Across all sampled lakes 54 bacterial

OTUs were detected in total, and 29 and 17 of these were shared with July and September communities respectively. For these shared OTUs, there was a positive correlation between their frequency of occurrence in mesocosms during September and the source lakes ($r=0.49$; $P=0.007$). No such correlation was found for viruses, and of the 101 viral OTUs found in the source lakes, only 6 were detected in the experiment.

Discussion

Despite strong effects of nutrient additions on the abundance of prokaryotes, viruses, and zooplankton, we observed no effects of environmental heterogeneity on the metacommunity structure of bacteria and viruses. The fact that beta diversity in metacommunities with heterogeneous nutrient addition treatments showed no difference from metacommunities with homogenous patches is consistent with our predictions of a neutral scenario. However, multivariate analyses (db-RDA) demonstrated effects of nutrient additions on community composition, indicating that niche-environment interactions partly steered the compositional trajectories of the communities of bacteria and viruses. These niche-environment interactions, however, may not have been strong enough to affect compositional differentiation among patches at the metacommunity scale. Contrary to our predictions, dispersal did not enhance community differentiation along our established environmental gradient. Instead, the weekly exchange of very small volumes of water among containers reduced beta diversity of both bacterial and viral metacommunities. For example, the fraction of bacterial OTUs shared by patches within metacommunities was more than twice as high in those with dispersal as in isolated communities.

The observed reductions of beta diversity for both bacteria and viruses was
305 surprising, particularly given dispersal treatments that consisted of weekly exchanges ranging
from 0.009 to 0.35 % of mesocosm volumes in the low and high dispersal treatments,
respectively. From a purely neutral perspective, such low levels of dispersal relative to the
population sizes of the resident community would unlikely lead to metacommunity
homogenization due to mixing. Given the relatively high detection limits of our screening
310 methods, it is therefore unlikely that neutral dynamics have resulted in a convergence of
community profiles in the dispersal treatments. Metacommunity homogenization caused by
mass effects is even less likely, given that the sustenance of populations of mal-adapted
phylotypes would require exchange rates higher than in a neutral scenario. Indeed, previous
experiments (Jones and McMahon, 2009; Lindström and Ostman, 2011) and field surveys
315 (Logue and Lindström, 2010) have shown that dispersal rates need to be very high to
overwhelm the local dynamics of species sorting in bacterial communities. A more likely
explanation for the observed reduction in beta diversity is that the dispersal treatment has
alleviated dispersal limitation in the metacommunities and has thereby facilitated the
matching of community composition to environmental conditions in our experimental
320 mesocosms. This explanation makes sense if the environmental conditions in the mesocosms
generate a selection regime which favors only a subset of the original microbial communities.
The observation of higher numbers of shared OTUs in metacommunities with dispersal
suggests that local communities have enriched each other with such OTUs.

Experiments demonstrating dispersal limitation in natural microbial communities are
325 scarce, and to our knowledge none have examined the effect of dispersal on viral
communities. Lindström and Ostman (2011) studied the influence of dispersal treatments on
bacterial community composition of three lakes for a wide range of dispersal rates. They

found evidence of mass effects at very high dispersal rates, but no indications of dispersal limitation at low dispersal rates. In a series of incubation experiments with microbial communities from rainwater pools in a woodland, Bell (2010) found evidence for dispersal limitation, but only over short time scales (a few days). Langenheder *et al.* (2006) incubated bacterial communities of eight different aquatic habitats under identical environmental conditions and found no increase in the similarity among communities during the course of the experiment. Although their experiment did not involve a dispersal treatment, their result could have occurred because of dispersal limitation from source communities. Verreydt *et al.* (2012) showed stronger compositional shifts of bacterial communities in response to a nutrient gradient when dispersal was present, which was largely explained by dispersal-mediated trophic interactions, in which effects of dispersal on zooplankton and phytoplankton communities translated into apparent dispersal effects in bacterial communities, probably through modified grazing and nutrient recycling regimes (Langenheder and Jürgens, 2001; Degans *et al.*, 2002; Jürgens and Matz, 2002). Such dispersal-induced homogenization of zooplankton metacommunities may also have contributed to the observed reduction of bacterial beta diversity in our experiment. Indeed, we detected significant Mantel correlations between zooplankton and bacterial communities and a positive association between zooplankton and bacterial beta diversity. However, GLM-analysis showed a significant effect of the dispersal treatment on the beta diversity of bacteria and viruses, independent of zooplankton beta diversity, indicating that the dispersal treatment also exerted unique effects. Viruses are typically highly host-specific such that a decrease in bacterial community differentiation could lead to a reduction in the beta diversity of the viral metacommunities, but our results show no evidence for this. Mantel

correlation analyses revealed no association between bacterial and viral communities, and our GLM-analyses showed no effects of bacterial beta diversity on viral beta diversity.

Our results show that very low exchange rates have enabled microbial communities to affect each other's compositional trajectories. Although this seems to be at odds with the
355 idea that prokaryotes and viruses are ubiquitously distributed in nature, our results do not allow us to make statements about distribution patterns of OTUs in natural lakes at larger spatial scales, given the high detection limits of T-RFLP and RAPD and the possibility that our initial inoculums may have only represented a fraction of the microbial communities present in the lakes. Our results, however, do suggest that the compositional changes of bacterial and
360 viral communities in response to new environmental circumstances can be constrained by dispersal limitation, at least over time scales corresponding up to a few months.

A considerable share of the bacterial OTUs (35% in July and 49% in September) was also found in the source lakes, and in July we found a positive correlation between the frequency of occurrence of OTUs across containers in the experiment and the lakes supplying
365 the corresponding inocula. In contrast, viral communities shared fewer OTUs with the source lakes and we detected a large fraction of new viral OTUs in September that had not been observed during July or in the lake survey. The majority of viruses in freshwater environments are viruses infecting prokaryotes (phages) and these have been shown to exhibit strong host-specificity (Wommack and Colwell, 2000). The Bank model (Breitbart and
370 Rohwer, 2005), for example, suggests that the active viral community composition is strongly determined by the composition of the hosts. Our results provide no support for this prediction as responses of viral and bacterial composition to the experimental treatments differed qualitatively (e.g., response to nutrient addition in bacteria but not in viruses) and Mantel tests provided no evidence for overall compositional associations. Similarly, Shurin *et*

375 *al.* (2012) found that the response of viruses to experimental treatments of warming, fish
predators and eutrophication were distinct from those of bacteria. Weak associations
between bacteria and viruses can be expected if a large fraction of the viruses infect other
organisms, such as phytoplankton cells. In most cases, however, bacteriophages are much
more abundant than other viruses, because of the numerical dominance of bacteria in
380 plankton communities. Furthermore, viruses infecting other organisms than prokaryotes
tend to be considerably larger than bacteriophages. Our sampling procedures may have
partially excluded such viruses, given that samples were filtered over 0.22 μm filters and then
shock-frozen, which facilitates the break-up of large viruses. Patterns of association may also
be reduced if viral communities show low host-specificity (Sano *et al.*, 2004; Holmfeldt *et al.*,
385 2007). Prokaryotic hosts may also be infected by multiple viruses (Holmfeldt *et al.*, 2007) and,
given the high reproduction potential of viruses, viral communities can show rapid
compositional turnover rates leading to poor associations between bacterial and viral
communities. Finally, host-virus interactions may also have been obscured by limitations in
the resolution of our molecular screening techniques (e.g., Holmfeldt *et al.*, 2007).

390 Six months after the start of the experiment, we were still able to find signatures of
the original lake inoculations in the composition of bacterial communities, and in the beta
diversity of bacterial and viral metacommunities. This indicates persistent effects of the initial
distribution of taxa within the original source communities. Such lasting differences between
bacterial communities of different origin suggest dispersal limitation, although it should be
395 noted that our inoculations also included lake sediments. Inoculation effects may therefore
also have been consolidated by the presence of large reservoirs of bacterial spores (Jones
and Lennon, 2010) and persistent effects of sediments on water quality.

Although our results demonstrate that low levels of dispersal can affect the compositional trajectory of microbial communities, we found no interaction effects of environmental heterogeneity and dispersal on the beta diversity of microbial metacommunities. The lack of such an interaction may be the result of several specific features of our experiment. First, each of the patches in our experimental metacommunities was initially inoculated with communities from a mixture of three different locations. If compositional similarity among the source lakes was already low then increased nutrient heterogeneity may not have been able to decrease the similarity among these communities any further. Second, dispersal failed to enhance compositional differentiation in response to nutrient heterogeneity. Possibly, the initial inoculums were too species rich and functionally saturated so as to be constrained in their response to a nutrient enrichment gradient. Third, the 'regional' species pool in each community consisted of only two communities, which limits the potential for new functional types to establish in the communities. Therefore, the probability that dispersal will interact with patch heterogeneity in determining metacommunity structure of microbial organisms will likely increase when functionally poor communities are connected with a relatively rich pool of potential immigrants along pronounced environmental gradients. Fourth, we used a relatively coarse method for quantifying microbial diversity, because the detection limits of T-RFLP and RAPD are high and only allow detecting relatively abundant OTUs. This methodological limitation may have obscured responses of the less abundant OTUs to the experimental treatments. Overall, such limitations have also made it impossible to identify the original sources of individual OTUs and track their spatial dynamics throughout the experiment. We therefore see great potential in the combination of high-throughput sequencing technology with similarly designed experiments to study microbial metacommunity dynamics.

Conclusions

Very low dispersal rates among local communities within replicate metacommunities
425 resulted in a decrease in the compositional differences (beta diversity) in the
metacommunities of both bacteria and viruses. Given the relatively small numbers of
individuals that were exchanged, this response cannot be attributed to community
homogenization or to mass effects. Beta diversity of bacteria and to lesser extent viruses was
also associated with the heterogeneity in phytoplankton and zooplankton communities.
430 Nevertheless, the effects of the dispersal treatment could not be explained by indirect effects
of dispersal on these other trophic levels. While we found no evidence for interactive effects
between dispersal intensity and patch heterogeneity on metacommunity beta diversity, our
results demonstrate that a very limited exchange of organisms among local communities can
alter the species sorting trajectory of bacterial and viral communities at temporal spatial
435 scales of a few months.

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445 Supplementary information is available at The ISME Journal website.

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Figure legends

Figure 1 Schematic representation of the randomized block experimental design. This scheme represents one block consisting of six pairs of two containers ('metacommunities') connected by dispersal. The experimental treatments consist of the six cross-factorial combinations of two factors: 'nutrient heterogeneity' (two levels) and 'dispersal intensity' (three levels). Three metacommunities received identical amounts of nutrients at low concentrations (NUT_L) and are referred to as metacommunities with homogenous nutrient additions (HOMO). Heterogeneous metacommunities (HETERO) were created by repeatedly adding high amounts of nutrients to one of the containers (NUT_H) and by adding no nutrients to the other container (NUT_N). The factor dispersal consisted of three levels of exchange rates of planktonic organisms among containers, with no (D0), low (DL) and relatively high dispersal rates (DH). Blocks were replicated four times with identical heterogeneity and dispersal treatment combinations but inoculations from different lakes (see Appendix 1 for a more detailed explanation of inoculation assignment).

Figure 2 Beta diversity of bacterial (a) and viral (b) metacommunities in response to dispersal and nutrient addition treatments. Beta diversity was calculated as the Jaccard dissimilarity index among communities within metacommunities. D0: no dispersal, DL: low dispersal, DH: high dispersal, HETERO: heterogeneous metacommunities, HOMO: homogeneous metacommunity. Different letters at symbols indicate significant differences among dispersal levels. Error bars represent the standard deviation.

Figure 3 Biotic heterogeneity in metacommunities in response to dispersal and nutrient addition treatments. (a) Bray-Curtis distances among the pairs of zooplankton communities (BC_{zoop}), (b) Differences in chlorophyll a concentration (log-transformed absolute values) ($\Delta CHLA$). D0: no dispersal, DL: low dispersal, DH: high dispersal, HETERO: heterogeneous metacommunities, HOMO:

homogeneous metacommunity. Different letters at symbols indicate significant differences among dispersal levels. Error bars represent the standard deviation.

Table 1 Results of ANOVA-analyses testing for the effects of dispersal intensity and metacommunity type (heterogeneous versus homogeneous nutrient addition regime) on the beta diversity (Jaccard dissimilarity) of bacterial and viral metacommunities.

| Factor | Bacteria | | | | | Viruses | | | |
|-----------------------------|----------|-------|-------|------|--------------|---------|-------|------|--------------|
| | df | SS | MS | F | P | SS | MS | F | P |
| Dispersal intensity (Disp) | 2 | 0.211 | 0.106 | 6.62 | 0.009 | 0.0151 | 0.008 | 4.36 | 0.032 |
| Metacommunity type (Mctype) | 1 | 0.003 | 0.003 | 0.15 | 0.699 | 0.000 | 0.000 | 0.00 | 0.964 |
| Disp x Mctype | 2 | 0.033 | 0.017 | 1.05 | 0.375 | 0.005 | 0.002 | 1.35 | 0.298 |
| Blocks | 3 | 0.212 | 0.071 | 4.42 | 0.020 | 0.023 | 0.008 | 4.46 | 0.020 |
| Residuals | 15 | 0.240 | 0.016 | | | 0.026 | 0.002 | | |

Abbreviations: df, degrees of freedom; SS, sum of squares; MS, mean squares; P, significance level; Blocks, refers to the randomized block design.

Significant P-values are in bold.

Table 2 Results of ANOVA-analyses testing for the effects of dispersal intensity and metacommunity type (heterogeneous versus homogeneous nutrient addition regime) on the heterogeneity found for other trophic levels.

| Factor | df | BC _{zoop} | | | | ΔCHLA | | | |
|-----------------------------|----|--------------------|-------|--------|--------------|-------|-------|-------|--------------|
| | | SS | MS | F | P | SS | MS | F | P |
| Dispersal intensity (Disp) | 2 | 0.349 | 0.174 | 8.870 | 0.003 | 0.045 | 0.023 | 0.084 | 0.920 |
| Metacommunity type (Mctype) | 1 | 0.344 | 0.344 | 17.503 | 0.001 | 0.536 | 0.536 | 1.988 | 0.179 |
| Disp x Mctype | 2 | 0.010 | 0.005 | 0.246 | 0.785 | 0.200 | 0.100 | 0.370 | 0.697 |
| Blocks | 3 | 0.162 | 0.054 | 2.746 | 0.082 | 3.079 | 1.026 | 3.804 | 0.033 |
| Residuals | 14 | 0.275 | 0.020 | | | 4.047 | 0.270 | | |

Abbreviations: ΔCHLA, difference in chlorophyll a concentrations among mesocosms of the same metacommunity; BC_{zoop}, Bray-Curtis distance among the zooplankton communities of metacommunities; df, degrees of freedom; SS, sum of squares; MS, mean squares; P, significance level; Blocks, refers to the randomized block design.

Significant P-values are in bold.

Table 3 Results of GLM-analyses explaining variation of the beta diversity (Jaccard dissimilarity) in bacterial and viral metacommunities. Dispersal intensity and metacommunity type (heterogeneous versus homogeneous nutrient addition regime) represent the experimental treatments.

| | Bacteria | | | | | Viruses | | | |
|-----------------------------|----------|-------|-------|--------|--------------|---------|-------|-------|--------------|
| | df | SS | MS | F | P | SS | MS | F | P |
| Dispersal intensity (Disp) | 2 | 0,177 | 0,088 | 7,584 | 0,009 | 0,015 | 0,007 | 3,990 | 0,050 |
| Metacommunity type (Mctype) | 1 | 0,001 | 0,001 | 0,117 | 0,739 | 0,000 | 0,000 | 0,001 | 0,981 |
| Disp x Mctype | 2 | 0,055 | 0,028 | 2,360 | 0,140 | 0,006 | 0,003 | 1,508 | 0,264 |
| Δ CHLA | 1 | 0,160 | 0,160 | 13,738 | 0,003 | 0,010 | 0,010 | 5,501 | 0,039 |
| BCzoop | 1 | 0,073 | 0,073 | 6,295 | 0,029 | 0,005 | 0,005 | 2,625 | 0,133 |
| β_{vir} | 1 | 0,001 | 0,001 | 0,047 | 0,833 | | | | |
| β_{bact} | | | | | | 0,000 | 0,000 | 0,042 | 0,842 |
| Blocks | 3 | 0,064 | 0,021 | 1,818 | 0,202 | 0,013 | 0,004 | 2,328 | 0,131 |
| Residuals | 11 | 0,128 | 0,012 | | | 0,020 | 0,002 | | |

Abbreviations: Δ CHLA, difference in chlorophyll a concentrations among mesocosms of the same metacommunity; BC_{zoop} , Bray-Curtis distance among the zooplankton communities of metacommunities; df, degrees of freedom; SS, sum of squares; MS, mean squares; P, significance level; Blocks, refers to the randomized block design. We also included bacterial beta diversity (β_{bact}) as explanatory variable in the model for the beta diversity of viruses (β_{vir}) and vice versa. Significant P-values are in bold.

Figure 1

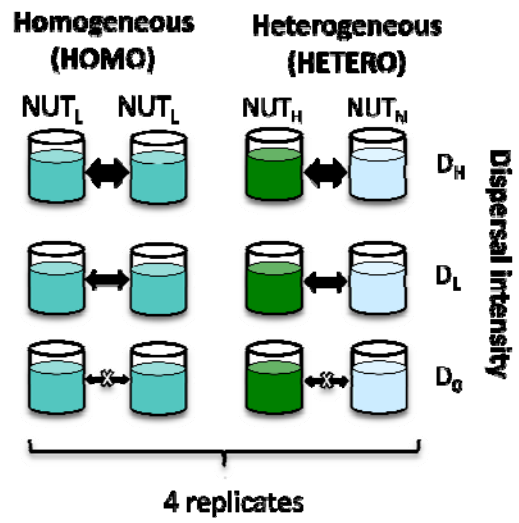


Figure 2

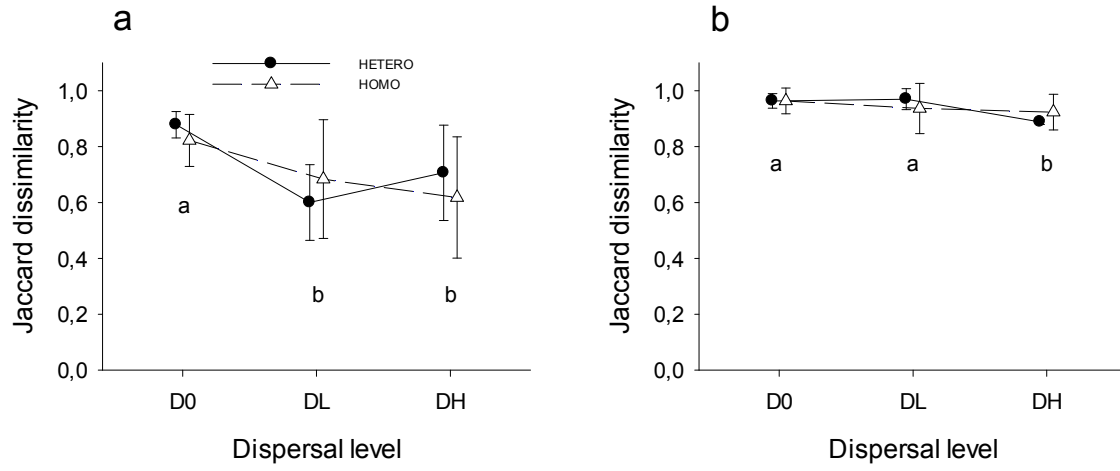


Figure 3

