

1 Dispersal-mediated trophic interactions can generate apparent patterns of dispersal limitation in
2 aquatic metacommunities

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43 **Statement of authorship**

44 The experiment was designed by SD, LDM, DV, ED and WV. The mesocosm experiment was
45 carried out by DV, ED and MJV, sample analysis was performed by DV, MJV and KVDG.

46 Troubleshooting and discussing data analysis was mainly done by SD, DV, LDM and WV. Data

47 analysis was mainly done by DV and SD, with contributions of PV and KVDG. The manuscript

48 was written by DV, SD and LDM, with editing by PV, ED and KVDG.

49 **Abstract**

50 Dispersal is a major organizing force in metacommunities, which may facilitate compositional
51 responses of local communities to environmental change and affect ecosystem function. Organism
52 groups differ widely in their dispersal abilities and their communities are therefore expected to
53 have different adaptive abilities. In mesocosms, we studied the simultaneous compositional
54 response of three plankton communities (zoo-, phyto- and bacterioplankton) to a primary
55 productivity gradient and evaluated how this response was mediated by dispersal intensity.
56 Dispersal enhanced responses in all three planktonic groups, which also affected ecosystem
57 functioning. Yet, variation partitioning analyses indicated that responses in phytoplankton and
58 bacterial communities were not only controlled by dispersal directly, but also indirectly through
59 complex trophic interactions. Our results indicate that metacommunity patterns emerging from
60 dispersal can cascade through the food web and generate patterns of apparent dispersal limitation
61 in organisms at other trophic levels.

62

63

64 **Introduction**

65 Dispersal is a major structuring force in metacommunities, also affecting the composition,
66 diversity and functioning of local communities (Loreau *et al.* 2003; Matthiessen & Hillebrand
67 2006; Duffy 2009; Hillebrand & Matthiessen 2009; Howeth & Leibold 2010). The qualitative and
68 quantitative effects of dispersal are, however, difficult to predict and depend on many factors, such
69 as the intensity of dispersal, the functional traits of the organisms, and the occurrence of complex
70 ecological interactions. Dispersal can lead to increased local species richness because it allows
71 new species to enter communities and compensate for local extinctions (Loreau & Mouquet 1999).
72 Dispersal can also facilitate a rapid and efficient compositional response of local communities to
73 changing environmental conditions through the process of species sorting. At the metacommunity
74 level, this process can lead to enhanced differentiation among communities and an increase in beta
75 diversity (Cottenie & De Meester 2004; Leibold & Norberg 2004). Conversely, dispersal can also
76 increase the probability of invasions by superior competitors, predators or parasites potentially
77 causing extinctions of local populations or entire metapopulations. In the case of mass effects, high
78 dispersal rates may result in the presence of species in unsuitable habitat patches, and lead to a
79 homogenization of metacommunities (Mouquet & Loreau 2003).

80 Lack of dispersal, on the other hand, may impede species to reach potentially suitable
81 habitat patches (Loreau *et al.* 2003; Ozinga *et al.* 2005; McCauley 2006; Pärtel & Zobel 2007). At
82 the local scale, this reduces the capacity of resident communities to track environmental change,
83 which may have a profound impact on the performance of entire functional groups or trophic
84 levels and as such affect ecosystem functioning. Although empirical evidence remains scarce,
85 there are a limited number of recent experimental studies that have demonstrated such effects.
86 Naeslund & Norberg (2006) found stronger responses of zooplankton communities to a change in
87 basal productivity if the communities at the start of the experiment represented the entire regional

88 species pool instead of just local pools. They also found that the communities resulting from the
89 regional species pool treatment exerted stronger top-down control on phytoplankton than the
90 communities resulting from local species pools. Howeth & Leibold (2008, 2010) showed that
91 dispersal can affect ecosystem stability and dampen trophic cascade effects in plankton
92 communities that are subject to temporal fluctuations in the density of a top predator.

93 Metacommunity structure varies widely among organism groups (Beisner *et al.* 2006). One
94 potential reason for this is that dispersal capabilities of organisms vary greatly (Bohonak &
95 Jenkins 2003; Jenkins *et al.* 2007). Microbial organisms, for example, are supposed to have very
96 high dispersal rates (Finlay & Clarke 1999; Finlay 2002), in contrast to passively dispersing
97 macroscopic organisms of which dispersal rates tend to decrease with increasing body size
98 (Jenkins *et al.* 2007; Shurin *et al.* 2009). An additional level of complexity may be generated by
99 indirect mechanisms, such as dispersal-mediated trophic interactions. Indeed, the metacommunity
100 structure of a group of organisms that is strongly affected by trophic interactions can also reflect
101 patterns resulting from spatial dynamics at other trophic levels. In such case, an association
102 between connectivity patterns and the metacommunity structure of an organism group does not
103 unequivocally reflect dispersal limitation or mass effects (Staddon *et al.* 2010).

104 With this study, we wanted to explore the direct and indirect effects of dispersal on the
105 composition and functioning of communities representing different trophic levels within a
106 metacommunity context. We used freshwater plankton in mesocosms as a model system and
107 performed an experiment combining varying degrees of dispersal with an important environmental
108 gradient (i.e. a gradient in primary productivity) according to an orthogonal design. The main
109 objectives of our study were to (1) study the extent to which dispersal can mediate the response of
110 plankton organisms to a change in basal productivity, (2) compare these responses among three
111 functional groups of plankton organisms that differ widely in body size, life strategy and expected

112 dispersal rates (i.e. zoo-, phyto- and bacterioplankton), (3) evaluate the consequences of dispersal
113 on food web structure and a crucial ecosystem function (i.e. zooplankton grazing on
114 phytoplankton), and (4) evaluate whether the response of the functional groups to the dispersal
115 gradient is caused directly by dispersal itself or rather indirectly by changes in trophic interactions
116 that are caused by dispersal-mediated community shifts at other trophic levels. In the absence of
117 positive size-selective predators, grazing by large-bodied cladoceran zooplankton can strongly
118 affect community composition of phytoplankton and bacterioplankton (Jürgens 1994; Lampert
119 2006). We therefore expect that dispersal-enhanced differentiation of zooplankton communities
120 along a productivity gradient may generate patterns of community differentiation at lower trophic
121 levels that appear to be caused by dispersal limitation but in fact are generated by indirect
122 dispersal-mediated trophic interactions.

123

124 **Material and methods**

125 *Experimental design*

126 Using mesocosms (n = 96), we studied the simultaneous interactive effects of nutrient addition and
127 dispersal on communities of planktonic bacteria, phytoplankton and zooplankton. For this, we first
128 collected plankton from 16 lakes (LAKE ID) representing a broad gradient in trophic state and
129 limnological characteristics. The plankton of each of these lakes was used to inoculate six
130 mesocosms per lake (6 x 16 = 96 mesocosms) at the start of the experiment (see Appendix S1 in
131 Supporting Information). In each of these sets of six mesocosms, we created two levels of nutrient
132 addition and three levels of dispersal intensity, so that the entire experimental set-up accorded to a
133 cross factorial randomized block design (with LAKE ID as blocks). With the dispersal treatment,
134 we tried to achieve a broad range of dispersal intensities among our experimental containers,
135 ranging from no dispersal to strong dispersal. In each block, the no dispersal (NDISP) and low
136 dispersal (LDISP) mesocosms were inoculated with plankton originating from one single lake. In
137 contrast, the high dispersal (HDISP) mesocosms were initially inoculated with a plankton mixture
138 from all 16 lakes (see below for details). During the entire experiment, we tried to prevent any
139 input of organisms from other mesocosms into the NDISP-mesocosms. The communities in these
140 mesocosms thus consisted solely of species collected from one individual lake, although some
141 airborne exchange of phytoplankton and bacteria could probably not be entirely excluded. For the
142 other two dispersal levels, we achieved dispersal by manually exchanging water among
143 mesocosms. For this, we collected water from all mesocosms of the respective dispersal level
144 (n=32) and redistributed the pooled volume in equal parts over the same mesocosms again. In this
145 way, a level of low dispersal (LDISP) was achieved through the exchange of 40 mL, whereas the
146 highest dispersal (HDISP) was generated by exchanging 2L volumes. We initially applied the
147 dispersal treatment on a weekly basis but switched to a biweekly treatment from day 59 on until

148 the end of the experiment. The NDISP-treatment was meant to represent a metacommunity
149 without dispersal among habitat patches. The LDISP-treatment was designed to represent a
150 situation where locally abundant species can disperse in low numbers among habitat patches, with
151 sporadic exchange of locally rare species. The HDISP-treatment represents metacommunities with
152 relatively high exchange rates among local communities, where each species has historically had
153 the occasion to enter each habitat patch, but where current dispersal rates are not high enough to
154 cause mass-effects (Michels *et al.* 2001; Howeth & Leibold 2008).

155 At day 1 of the experiment (23 May 2006), we filled plastic containers (volume: 200L) with a
156 mixture of 120 L distilled water and 60 L tap water. The nutrient treatments were established on
157 days 4 and 5 through addition of phosphate (KH_2PO_4) and nitrogen (NaNO_3). Initial nutrient
158 additions were equivalent to $1000\mu\text{g P L}^{-1}$ and $16000\mu\text{g N L}^{-1}$ in the high nutrient (HNUT)
159 containers and $10\mu\text{g P L}^{-1}$ and $160\mu\text{g N L}^{-1}$ in the low nutrient (LNUT) containers. Because
160 earlier mesocosm experiments have indicated that nutrient gradients can decline with time, we
161 continued with a weekly addition of a tenth of these concentrations throughout the experiment.

162 We added phyto- and bacterioplankton to the containers on the 5th day of the experiment. For
163 this, we collected a 30 L volume of lake water and filtered it twice (mesh sizes: 100 μm and 50
164 μm , respectively) to remove zooplankton. NDISP and LDISP mesocosms were all inoculated with
165 an inoculum originating from one individual lake, whereas the inoculum of HDISP mesocosms
166 consisted of a mixture of all 16 lakes (each experimental container received an equal amount of
167 chlorophyll a and for the HDISP mesocosms there was an equal representation of lakes in terms of
168 chlorophyll a). From the moment we observed a consistent difference in phytoplankton biomass
169 between LNUT and HNUT mesocosms (day 32), we inoculated the zooplankton. Total
170 zooplankton biomass was the same in all inocula. Similar as with the phytoplankton inoculation,
171 we inoculated NDISP and LDISP mesocosms with inocula from individual lakes, whereas HDISP

172 containers received an inoculum for 80 % consisting of the respective lake and for 20 % consisting
173 of a mixture of all 16 lakes. Throughout the experiment, mesocosms were covered by mosquito
174 netting to prevent contamination by macro-invertebrates. The experiment was ended at day 87.

175

176 *Sampling and sample analysis*

177 We measured chlorophyll a on a weekly basis with a fluorometer (Trilogy Laboratory
178 Fluorometer, Turner Designs). Near the end of the experiment, we sampled the zoo-, phyto-, and
179 bacterioplankton communities. At day 77, we sampled phytoplankton with a 250 mL jar
180 approximately 10 cm below the water surface. The phytoplankton samples were preserved with a
181 mixture of Lugol's solution, formaldehyde and sodium thiosulfate (Sherr & Sherr 1993) and
182 counted using an inverted microscope to the genus level. *Desmodesmus* was a dominant
183 phytoplankton taxon in some treatments of the experiment. *Desmodesmus* colony size has been
184 shown to be a defense against zooplankton grazing and may therefore serve as an indicator for the
185 prevailing zooplankton grazing regime (Vanormelingen *et al.* 2009). We therefore characterized
186 the size distribution of *Desmodesmus* by counting the number of cells per colony in each sample.
187 At day 79, we sampled the bacterioplankton. Samples were filtered over a 0.22 µm filter and
188 stored at -80°C for later analysis with denaturing gradient gel electrophoresis (DGGE). DGGE
189 analysis followed Van der Gucht *et al.* (2007); details are given in Appendix S2 in Supporting
190 Information. In short, DNA was extracted directly using the bead-beating method concomitant
191 with phenol extraction and ethanol precipitation and purified on a Wizard column. A small rDNA
192 fragment was amplified with primers specific to the domain Bacteria (357F-GC-clamp and 518R).
193 PCR products were analyzed on a 35 to 70% denaturant DGGE gel, and DGGE gels were stained
194 with Sybr Gold solution. The 96 samples were analyzed on 12 parallel DGGE-gels, which were
195 aligned with Bionumerics 5.10 (Applied Maths BVBA, Kortrijk, Belgium) using three standard

196 lanes (known mixtures of DNA from 9 clones from a clone library) on each gel. A matrix was
197 compiled based upon the relative contribution of individual bands to the total band signal in each
198 lane, with bands corresponding to OTU's (Operational Taxonomic Units). Zooplankton samples
199 were taken at day 86 and 87 of the experiment. Two samples were taken in each container with a
200 Schindler Patalas (volume: 12 L; mesh size: 30 μ m) and preserved with acid lugol solution. One
201 sample was used to measure zooplankton dry weight. These samples were weighed after drying at
202 100°C during 24h. The other sample was used for the assessment of species composition and
203 population densities. A minimum of 300 individuals were counted. Taxa were identified to species
204 level for cladocerans using Flössner (2000); for copepods we made a distinction between
205 cyclopoids and calanoids.

206

207 *Grazing experiment*

208 Zooplankton grazing is widely acknowledged as a key characteristic of the aquatic food web that
209 determines patterns of energy and material flows and underwater light climate, and may mediate
210 regime shifts in ponds and shallow lakes (Scheffer 1998). To evaluate its importance as potential
211 driving force underlying phyto- and bacterioplankton community responses to the experimental
212 treatments, we performed assays to assess zooplankton grazing pressure on phytoplankton at day
213 81. In each mesocosm, we incubated 2 bottles (250 mL) with mesocosm water, one with ambient
214 zooplankton densities and one without zooplankton (water filtered over 64 μ m mesh). The bottles
215 were incubated at the bottom of the mesocosms and gently shaken twice a day in order to keep the
216 phytoplankton in suspension. At day 1 and day 11 of the experiment, we measured the chlorophyll
217 a concentrations with a fluorometer. Per bottle, we calculated the change in chlorophyll a using the
218 formula: $(\ln[\text{chl a}]_{t11} - \ln[\text{chl a}]_{t0})/\text{time}$ and used the difference between the treatments as a measure
219 of zooplankton grazing pressure.

220

221 *Data analysis*

222 We applied mixed model ANOVA to evaluate the impact of our experimental treatments on
223 chlorophyll a concentration, *Desmodesmus* colony size, total zooplankton biomass (dry weight)
224 and *in situ* zooplankton grazing pressure. In these analyses, we specified LAKE ID as a random
225 factor. For chlorophyll a, we analyzed the time weighted averages so as to give more weight to
226 data that are collected later during the experiment. These averages were calculated for each
227 mesocosm by multiplying each chlorophyll *a* value with the time that had passed since the start of the
228 experiment (expressed in numbers of days). The sum of these values was then divided by the total
229 duration of the experiment. For *Desmodesmus* colony size, we analyzed the weighted average of
230 cell number. Significant effects were explored with Tukey HSD post hoc tests.

231 We tested the effects of the experimental treatments on the composition of zooplankton,
232 phytoplankton and bacterioplankton communities with redundancy analysis (RDA). In these
233 analyses, we followed a two-step approach. First, we evaluated the overall effects of the
234 experimental treatments and their potential interactions on each of the communities separately
235 (Lepš & Šmilauer 2003). Second, we performed variation partitioning analyses (Peres-Neto *et al.*
236 2006) on more elaborate RDA-models to explore the extent to which indirect trophic interactions
237 can explain apparent dispersal effects in phytoplankton and bacterioplankton. We constructed
238 RDA models for these groups in each of the nutrient addition levels separately with the dispersal
239 treatment and the biomass of the principal zooplankton grazer, *Daphnia magna*, as explanatory
240 variables. We also included summary variables of phytoplankton community composition as
241 explanatory variables in the RDA model of bacterioplankton because phytoplankton composition
242 can be a determining factor for bacteria and may itself be directly affected by dispersal or
243 indirectly by dispersal mediated zooplankton grazing. With variation partitioning, we assessed the

244 unique contribution of the dispersal treatment (conditional effect) as well as its degree of
245 collinearity with *Daphnia magna* density (in the phyto- and bacterioplankton models) and
246 phytoplankton community composition (in the bacterioplankton models). The summary variables
247 for phytoplankton community composition were extracted prior to the RDA analyses through
248 principal components analysis of the phytoplankton data (i.e. the four sample scores vectors with
249 the highest eigen values; prior analyses indicated that these four vectors all had a unique and
250 significant contribution to variation in the bacterioplankton community). All community data were
251 Hellinger transformed prior to analysis (Legendre & Gallagher 2001). Associations of *Daphnia*
252 *magna* densities with *in situ* measured grazing pressure and the weighted average of *Desmodesmus*
253 colony size were tested with Spearman rank correlation. All statistical analyses were performed in
254 R (v2.10.1, R Development Core Team 2008), using the *rda* and *varpart* functions of the *vegan*
255 library (Peres-Neto *et al.* 2006; Oksanen *et al.* 2010). Adjusted R^2 values were calculated on
256 residuals after partialling out the effect of LAKE ID. The significance of model components was
257 tested through 9999 random permutations.

258 **Results**259 *Zooplankton*

260 There was a significant nutrient x dispersal treatment interaction effect on zooplankton biomass
261 (dry weight) (Table 1). Zooplankton biomass was higher in the containers with high than low
262 nutrient addition, but only at high levels of dispersal (Figure 1A). Similarly, the RDA analyses
263 indicated significant main effects and an interaction effect of the nutrient addition and dispersal
264 treatments on zooplankton community composition (Table 2, Figure 2A). Separate analyses for
265 each of the dispersal levels showed that the response strength of the zooplankton community to the
266 nutrient addition treatment increased with increasing dispersal, explaining 9, 20 and 59 % of the
267 zooplankton community variation in the no, low and high dispersal treatment, respectively (Table
268 3). Most species responded negatively to high nutrient addition (Figure 2A). The nutrient by
269 dispersal interaction could almost entirely be attributed to the specific response of the large
270 cladoceran *Daphnia magna*. The absolute and relative abundance of this species was especially
271 high in high-nutrient cattle tanks subjected to high dispersal (see Appendix S3). Zooplankton
272 species richness was negatively affected by nutrient addition and positively by dispersal (see
273 Appendix S4).

274

275 *Phytoplankton*

276 High nutrient addition resulted in a strong increase of chlorophyll a concentrations throughout the
277 course of the experiment (Figure 1B; Table 1). According to RDA analysis, the nutrient addition
278 and dispersal treatments affected phytoplankton community composition and there was also an
279 interaction between both factors (Figure 2B; Table 2). Overall, nutrient addition resulted in a
280 strong increase in the contribution of *Desmodesmus* species, while containers with low nutrient
281 addition tended to be mainly characterized by a variety of other phytoplankton taxa (Figure 2B).

282 Analyses for each of the dispersal treatments separately showed an increase of the impact of
283 nutrient addition with increasing dispersal intensity (13, 11 and 22 % of the total community
284 variation explained by nutrient addition in the no, low and high dispersal treatment, respectively)
285 (Table 3). When dispersal was tested separately for each of the nutrient addition levels, there was
286 only a significant effect in containers with high nutrient addition (see Table S1). In containers with
287 high nutrient addition, colony size of the *Desmodesmus* morphs decreased with dispersal (Figure
288 1C): large colonial morphs were most abundant in the absence of dispersal, whereas unicellular
289 morphs were mainly associated with high levels of dispersal. This was also confirmed by an
290 ANOVA on the weighted average of colony cell number ($F(2,30) = 64.28$, $p < 0.001$).
291 Phytoplankton richness was positively affected by dispersal but only at low nutrient addition levels
292 (see Appendix S4).

293

294 *Bacterioplankton*

295 Nutrient addition and dispersal treatments affected the DGGE profiles of the bacterioplankton
296 communities (Table 2; Figure 2C). The compositional response to the nutrient addition increased
297 with increasing dispersal (nutrient x dispersal interaction): the contribution of the nutrient addition
298 to the community variation increased from 9, over 19 to 22 % in the no, low and high dispersal
299 treatment, respectively (Table 3). Dispersal was equally important under low and high nutrient
300 addition (see Table S1). The number of OTUs was significantly higher under high than low
301 nutrient addition levels, but only in the high dispersal treatment (see Appendix S4).

302

303 *Indirect dispersal-mediated intertrophic interactions*

304 Zooplankton grazing rates on the phytoplankton community, as measured by the in situ grazing
305 experiments, increased with dispersal intensity but only at high nutrient levels (Figure 1D, Table

306 1). Overall, grazing pressure was positively correlated with *D. magna* density (Spearman rank
307 correlation: $n = 96$, $r_s = 0.477$, $p < 0.001$). This correlation was especially strong in the mesocosms
308 with high nutrient levels ($n = 48$, $r_s = 0.704$, $p < 0.001$), but insignificant in containers with low
309 levels of nutrients ($n = 48$, $r_s = 0.0688$, $p = 0.642$).

310 Phytoplankton community composition was significantly associated with the population
311 density of *D. magna* and dispersal intensity but only under high nutrient conditions (Figure 3;
312 Table S2). Under these conditions, the dispersal treatment and *D. magna* jointly explained
313 approximately 35% of the total phytoplankton community variation. A large fraction of the
314 variation explained by *D. magna* proved highly collinear with dispersal (15%), whereas the
315 conditional effect of *D. magna* was relatively small (4%). At high nutrient concentrations,
316 weighted average *Desmodesmus* colony size was negatively correlated to *D. magna* density ($n =$
317 48 , $r_s = -0.546$, $p < 0.001$; Figure S1). We detected very similar patterns when using the estimated
318 zooplankton grazing rate and zooplankton dry biomass as explanatory variables in these analyses
319 (results not shown).

320 The dispersal gradient, *D. magna* density and phytoplankton community composition jointly
321 explained 17% of the compositional variation in the bacterioplankton communities under high
322 nutrient conditions (Figure 3; Table S2). The marginal effects of each of the three categories of
323 variables were highly significant (explaining 12, 7 and 14% of bacterial community variation,
324 respectively). Variation partitioning showed that relatively large fractions of community variation
325 were explained by joint effects, mainly of all three variable categories together (5%) or effects
326 shared by the dispersal treatment and phytoplankton community composition (6%). In contrast, the
327 conditional effects of the explanatory variables were relatively low and statistically insignificant.
328 The effect of *D. magna* was almost entirely collinear with dispersal and phytoplankton community

329 composition. Under low nutrient conditions, dispersal contributed most to bacterioplankton

330 community variation (11%).

331

332 **Discussion**

333 We observed highly significant interaction effects between the dispersal and nutrient addition
334 treatments for the zoo-, phyto- and bacterioplankton communities. In each of these functional
335 groups, the interaction effects reflected increased strength in community compositional shifts to
336 nutrient addition with increasing rates of dispersal. These results show that the degree of
337 connectivity among habitat patches within a metacommunity can profoundly affect the
338 composition of the constituent communities by facilitating species sorting. In addition, the results
339 of the variation partitioning analyses indicate that responses observed for specific communities (*in*
340 *casu* phytoplankton and bacterioplankton) are not uniquely caused by the physical exchange of
341 members of those communities but also indirectly by changed trophic interactions that result from
342 the impact of dispersal at other trophic levels (*in casu* zooplankton).

343 For zooplankton, we observed enhanced compositional responses to the productivity gradient
344 already at low exchange rates. A shift from no to low dispersal (i.e. from no to the weekly
345 exchange of 0.02% of container volumes) more than doubled the community response strength to
346 the nutrient addition treatment (see Table 3). This response was still considerably enhanced upon
347 further increase in connectivity, as the nutrient addition treatment contributed to a total of 59% of
348 the variation in zooplankton community composition at maximal dispersal rates. These dispersal-
349 mediated responses also affected ecosystem functioning: zooplankton grazing rates measured *in*
350 *situ* were higher at high than low nutrient conditions, but only at the highest levels of dispersal.
351 This effect mainly seemed to result from the response of one single zooplankton key stone species,
352 i.e. *Daphnia magna*. Although most zooplankton species responded negatively to nutrient
353 addition, *D. magna* performed very well under high nutrient conditions. *D. magna* was detected in
354 only a limited number of experimental communities in the absence of dispersal. With increasing
355 dispersal rates, the species expanded within the metacommunity and became dominant in

356 containers with high nutrient levels. The species was so influential that the response of its
357 populations to the experimental treatments constituted most of the response patterns of total
358 zooplankton community biomass. Consequently, zooplankton grazing rates also correlated
359 strongly with *D. magna* population density. In conclusion, our results demonstrate the potential
360 importance of dispersal-related spatial dynamics for ecosystem functioning, in agreement with
361 predictions of theoretical metacommunity models (Loreau *et al.* 2003; Gonzalez & Loreau 2009).

362 We observed effects of the dispersal treatment on the compositional response of the phyto- and
363 bacterioplankton communities to nutrient enrichment. Although such effects can be considered
364 evidence for dispersal limitation, our results suggest that they may also have resulted from indirect
365 effects, such as dispersal-mediated changes in trophic interactions. Indeed, as discussed above,
366 increased dispersal rates resulted in a strong increase of zooplankton grazing pressure at high
367 nutrient levels. Variables related to this gradient (i.e. *D. magna* population density, zooplankton
368 biomass, grazing rates) explained a substantial part of the variation in phytoplankton community
369 composition at high nutrient levels and most (79%) of this explained variation was collinear with
370 the dispersal treatment. The quality of the shifts in phytoplankton characteristics also strongly
371 suggests grazing intensity as a strong community structuring factor. Along with a gradient of
372 increasing dispersal intensity and increasing population densities of *D. magna*, average
373 *Desmodesmus* colony cell number decreased and large colonial and spined *Desmodesmus* morphs
374 were replaced by spineless, unicellular or smaller colonial morphs. Given that spine and colony
375 formation are well-known defenses against zooplankton grazing, this response appears contra-
376 intuitive. However, it can be well understood in the light of a trade-off between grazing resistance
377 and grazing tolerance (Agrawal 1998; Chase *et al.* 2000). While colony formation may indeed
378 decrease the vulnerability of phytoplankton cells to grazing by small and intermediate sized
379 zooplankton, such defense is largely ineffective against grazing by zooplankton with a very wide

380 food particle size range such as *D. magna* (Matveev *et al.* 2000; Mayeli *et al.* 2004; Sarnelle
381 2005). The phytoplankton community composition shift observed by us in response to the
382 dispersal and grazing gradients thus represents a shift towards smaller colonies and single cells
383 that may be more vulnerable to predation but that are better able to compensate for mortality losses
384 through faster population growth (Agrawal 1998). Our results thus suggest that the dispersal
385 treatment affected the phytoplankton community in an indirect way through an intensification of
386 top-down control by zooplankton. *D. magna* thrived under high nutrient addition, and release from
387 dispersal limitation allowed this species to successfully spread through the metacommunity,
388 developing dense populations in containers with high nutrients, which then resulted in major
389 quantitative and qualitative changes in the phytoplankton communities compared to mesocosms
390 where no dispersal was applied. It should be noted, however, that our statistical analysis suggests
391 that such indirect dispersal induced trophic interactions can only account for about half of the
392 phytoplankton community variation that was caused by the dispersal treatment, which indicates
393 that phytoplankton communities were also directly affected by the dispersal treatment.

394 Similar to our observations for the phytoplankton communities, effects of the dispersal
395 treatment on bacterial communities were not only due to increased exchange rates of bacteria
396 among mesocosms, but may also have been shaped by several indirect mechanisms. According to
397 our variation partitioning results, a large part (65 %) of the total explained bacterioplankton
398 variation was explained by variation among phytoplankton communities that was also collinear
399 with the dispersal treatment. Phytoplankton community composition can affect bacterioplankton
400 composition through competition for nutrients (Cherif & Loreau 2007; Daufresne *et al.* 2008) or
401 via the composition of DOC that it excretes (Giroldo *et al.* 2007). Dispersal may thus have
402 affected the composition of bacteria indirectly by its direct and indirect effects on the composition
403 of phytoplankton communities. An important subfraction of the variation in bacterial community

404 composition was also collinear with *D. magna*. Zooplankton can affect bacterioplankton directly
405 through selective grazing (Zollner *et al.* 2003; Hambright *et al.* 2007) or indirectly by structuring
406 bacterivore communities (e.g. ciliates, flagellates; Jürgens 1994; Jürgens & Stolpe 1995; Lampert
407 2006). The strong collinearity in a large fraction of the variation in zooplankton composition,
408 phytoplankton composition and dispersal suggests that the dispersal treatment affected an
409 important part of the bacterioplankton variation via a cascade of effects: increased dispersal rates
410 affected zooplankton community composition and grazing rates, which induced changes in the
411 phytoplankton community composition, which then affected bacterioplankton community
412 composition.

413 While our analysis suggests that there is a direct effect of dispersal limitation on phytoplankton
414 species composition in our experiment, we have no unequivocal evidence for dispersal limitation
415 of bacteria under high nutrient conditions, because the effects of the dispersal treatment were
416 strongly reduced and became insignificant when *D. magna* and phytoplankton community
417 composition were accounted for. There were, however, indications for dispersal limitation of
418 bacteria at low nutrient levels. Dispersal limitation is generally assumed to be of minor importance
419 in determining the composition of micro-organisms, because taxa are supposed to be omnipresent,
420 especially at small spatial scales (Finlay & Clarke 1999; Finlay 2002). Our data, nevertheless, also
421 suggest that dispersal limitation in micro-organisms can matter at the spatial and temporal scales
422 of our experiment. Indeed, we found indications for enhanced species sorting with increasing
423 dispersal rates by physical exchange of medium and organisms for phytoplankton and bacteria
424 after correction for indirect effects caused by dispersal limitation of zooplankton. This suggests
425 that not all taxa were present in the original inocula or at least not in sufficient densities to have an
426 impact during the course of our experiment. The fact that our results indicate more direct dispersal
427 limitation effects in the low than in the high nutrient treatment may reflect that existent

428 communities can intrinsically more easily adapt to high than to low nutrient conditions, or
429 alternatively may reflect that nutrient rich habitats are more abundant in the study region so that
430 these tend to be more omnipresent than species that are pre-adapted to low nutrient conditions.

431 The results of our study have important implications for the interpretation of metacommunity
432 patterns observed in the field. An increasing number of studies (Cottenie 2005; Beisner *et al.*
433 2006; Van der Gucht *et al.* 2007; Pandit *et al.* 2009; Declerck *et al.* 2011) have tried to link
434 patterns of community variation with existing metacommunity paradigms, by partitioning
435 community variation into spatial and environmental components, using direct gradient ordination
436 techniques. Our results show that spatial structures observed in such studies on natural
437 metacommunities may not only be caused by dispersal limitation or mass effects, but may also
438 emerge as the result of strong ecological interactions with other groups of organisms, which are
439 themselves impacted by dispersal limitation or mass effects. A spatial signal can emerge from an
440 insufficient sampling of relevant variation in environmental conditions (Langenheder &
441 Ragnarsson 2007). If variation in the environment is not measured and spatially structured, it will
442 result in a spatial signal. Our results indicate that species composition at other trophic levels is an
443 important environmental factor to be taken into account when determining the response of a given
444 group of organisms to environmental and spatial factors. Failing to incorporate this information
445 may introduce a significant spatial signal, even when the focal group of organisms is not dispersal
446 limited and mainly the subject of species sorting.

447

448 Conclusions

449 Our study provides evidence that increased dispersal rates within a metacommunity can strongly
450 mediate the compositional response of zoo-, phyto- and bacterioplankton communities to a
451 gradient in primary productivity, and that this strong response can be largely generated by the

452 response of a limited number of keystone species at higher trophic levels (such as *Daphnia magna*
453 in our experiment) that are dispersal limited and strongly affect the intensity of trophic
454 interactions. We show that key-stone species mediated impacts cause major changes in ecosystem
455 functioning (grazing pressure) and lead to apparent dispersal effects at lower trophic levels
456 (phytoplankton and bacterioplankton). Our results suggest that these indirect effects of
457 metacommunity structure may strongly impact species composition in local communities and even
458 ecosystem functioning, and should be taken into consideration in metacommunity analyses.
459

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469

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- 583
- 584

585 **Supporting Information**

586 **The following Supporting Information is available for this article:**

587 **Appendix S1** *Supplementary information on the composition of zooplankton and phytoplankton*
588 *communities in the source ponds.*

589 **Appendix S2** *Supplementary information on the DGGE analysis of the bacterioplankton*
590 *communities.*

591 **Appendix S3** *Supplementary information on the zooplankton and phytoplankton species responses*
592 *to the experimental treatments.*

593 **Appendix S4** *Supplementary information on the effects of the experimental treatments on the*
594 *taxon richness of zooplankton, phytoplankton and bacterioplankton.*

595 **Table S1** *RDA analysis per nutrient addition level.*

596 **Table S2** *Variation partitioning results quantifying the marginal and conditional effects on the*
597 *phyto- and bacterioplankton communities.*

598 **Table S3** *Abbreviations of phytoplankton and zooplankton species in the PCA biplot*

599 **Figure S1** *Averages Desmodesmus colony cell number along the Daphnia magna density gradient.*

600

601 Additional Supporting Information may be found in the online version of this article.

602

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606

607 Table 1. ANOVA results testing for the effect of nutrient addition (NUT) and dispersal (DISP)
 608 treatments and their interaction on zooplankton dry weight, phytoplankton chlorophyll a and
 609 grazing rate of the zooplankton on the phytoplankton. Grazing rates were measured by
 610 experiments *in situ*. LAKE ID refers to the origin of inoculation samples and was specified as a
 611 random block factor in the models.

612

	d.f.	SS	MS	<i>F</i>	<i>P</i> -value
614 Zooplankton dry weight					
615 LAKE ID	15	22.30	1.49	0.91	0.585
616 NUT	1	11.02	11.02	7.66	0.014
617 DISP	2	18.91	9.46	5.51	0.0092
618 NUT x DISP	2	17.88	8.94	5.9	0.0069
619 Phytoplankton chlorophyll a					
620 LAKE ID	15	0.50	0.034	1.39	0.280
621 NUT	1	13.52	13.52	641	< 0.001
622 DISP	2	0.06	0.030	1.90	0.167
623 NUT x DISP	2	0.08	0.039	3.07	0.0611
624 Zooplankton grazing rate					
625 LAKE ID	15	0.035	0.0021	2.52	0.281
626 NUT	1	0.017	0.017	14.89	0.0016
627 DISP	2	0.026	0.013	8.34	0.0013
628 NUT x DISP	2	0.013	0.0065	3.44	0.0451

629

630 Table 2. RDA results, testing for the effect of nutrient addition (NUT) and dispersal (DISP)
 631 treatments and their interaction on the variation in the zoo-, phyto-, and bacterioplankton
 632 community composition in the experimental containers. LAKE ID refers to the origin of
 633 inoculation samples and was specified as a random block factor in the models.

634

635	Variable	Covariables	R ²	F	P-value
636	Zooplankton				
637	NUT	DISP	0.19	26.64	< 0.001
638		LAKE ID			
639	DISP	NUT	0.079	5.64	< 0.001
640		LAKE ID			
641	NUT x DISP	NUT	0.057	4.43	< 0.001
642		DISP			
643		LAKE ID			
644	Phytoplankton				
645	NUT	DISP	0.091	10.29	< 0.001
646		LAKE ID			
647	DISP	NUT	0.061	3.41	0.002
648		LAKE ID			
649	NUT x DISP	NUT	0.052	3.066	0.002
650		DISP			
651		LAKE ID			
652	Bacterioplankton				
653	NUT	DISP	0.11	11.72	< 0.001

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654		LAKE ID			
655	DISP	NUT	0.05	2.73	< 0.001
656		LAKE ID			
657	NUT x DISP	NUT	0.046	2.65	< 0.001
658		DISP			
659		LAKE ID			
660					

661 Table 3. RDA-analysis results, testing for the effect of nutrient addition (NUT) within each of the
 662 dispersal treatments for zoo-, phyto-, and bacterioplankton communities in the experimental
 663 containers. In all these analyses, we specified LAKE ID as covariable to adjust for differences
 664 among lake origin of inoculation samples.

665

666		R ²	<i>F</i>	<i>P</i> -value
667	Zooplankton			
668	No Dispersal	0.085	3.64	0.008
669	Low Dispersal	0.20	8.90	< 0.001
670	High Dispersal	0.59	44.008	< 0.001
671	Phytoplankton			
672	No Dispersal	0.13	4.82	< 0.001
673	Low Dispersal	0.11	3.55	0.017
674	High Dispersal	0.22	8.033	< 0.001
675	Bacterioplankton			
676	No Dispersal	0.086	2.86	< 0.001
677	Low Dispersal	0.19	7.73	< 0.001
678	High Dispersal	0.22	7.83	< 0.001

679

680 **Figure legends**

681

682 Figure 1: Zooplankton biomass (A), phytoplankton chlorophyll a (B), *Desmodesmus* colony size
 683 (C) and *in situ* measured zooplankton grazing pressure (D) for each of the multifactorial
 684 combinations of experimental treatments. Chlorophyll a data represent time weighted averages;
 685 *Desmodemus* colony size is expressed as the mean colony cell number weighted by the relative
 686 abundance of colony size classes (only data given for high nutrient levels); the percentages in (C)
 687 indicate the contribution of *Desmodesmus* to total phytoplankton biomass. White triangles and
 688 black circles indicate containers of the low and high nutrient addition, respectively. Error bars
 689 denote twice the standard error of the mean. Characters indicate significant post hoc differences
 690 ($p < 0.05$).

691

692 Figure 2: Biplot of principal component analysis representing the response of zooplankton (A),
 693 phytoplankton (B), and bacterioplankton (C) community composition to the experimental
 694 treatments. Centroids (filled circles) indicate the average location of communities belonging to the
 695 same multifactorial treatment combinations of nutrient addition (HNUT = high nutrient content,
 696 LNUT = low nutrient content) and dispersal (NDISP = no dispersal, LDISP = low dispersal, and
 697 HDISP = high dispersal). For phyto- and bacterioplankton only species are shown for which a
 698 minimum of 5% of the variation can be explained by the treatments. In (B), *D.magna* is plotted as
 699 a supplementary variable; abbreviations of species are given in supplementary Table S3. In (C),
 700 each OTU is labeled by a code in which the number is referring to a specific band location on the
 701 DGGE-gel. Data were Hellinger transformed prior to analysis.

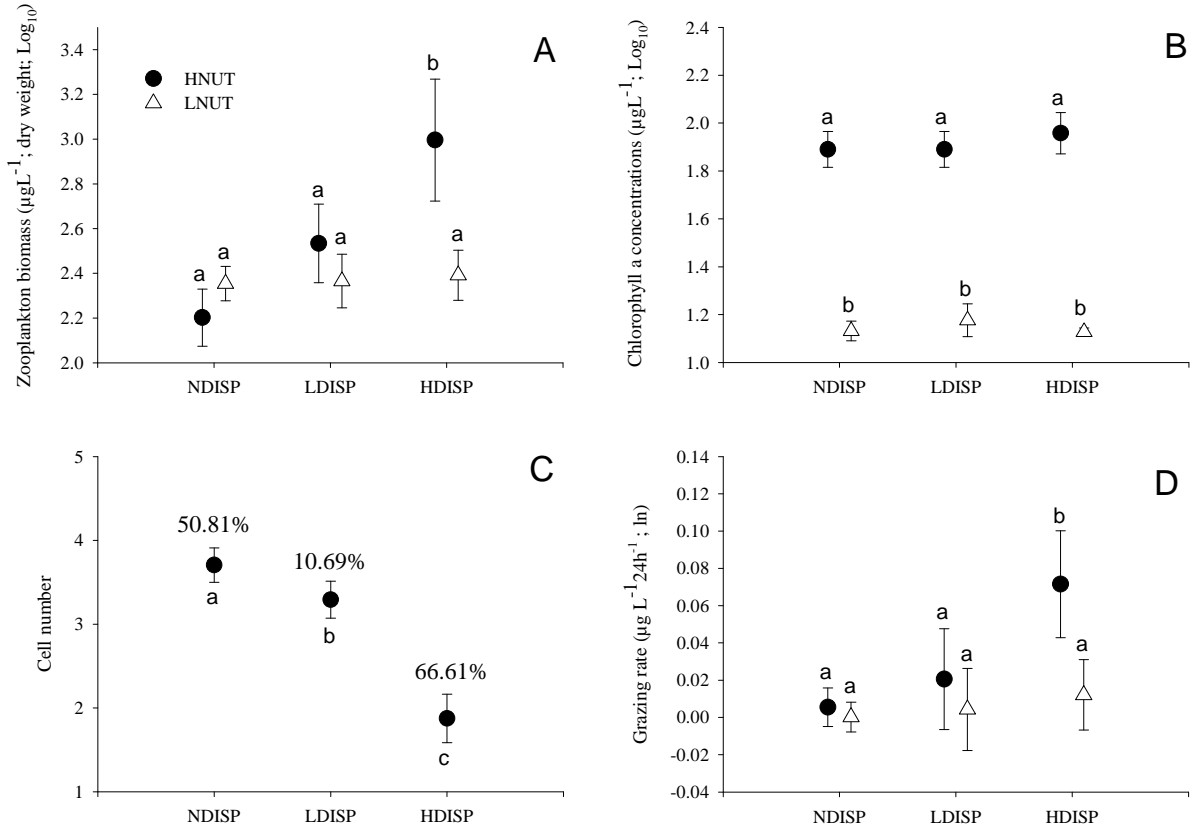
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703 Figure 3: Venn diagrams presenting the results of variation partitioning analyses performed on the
704 phytoplankton and bacterial community data at low and high nutrient levels, separately. For the
705 phytoplankton communities, the diagrams represent the unique and shared contributions of the
706 dispersal treatment and *Daphnia magna* population densities. For the bacterial communities, the
707 diagrams represent the unique and shared contributions of the dispersal treatment, *Daphnia magna*
708 densities and phytoplankton community composition. Figures outside the diagrams represent the
709 R^2 of the marginal effects of the factors tested. Figures within the diagrams represent the R^2 of the
710 conditional effects. R^2 -values express the percentage of total variance explained. Asterisks denote
711 significance level: *: $P < 0.05$; **: $P < 0.01$; ***: $P < 0.001$; ^{NS}: not significant.

712

713

714 Figure 1.



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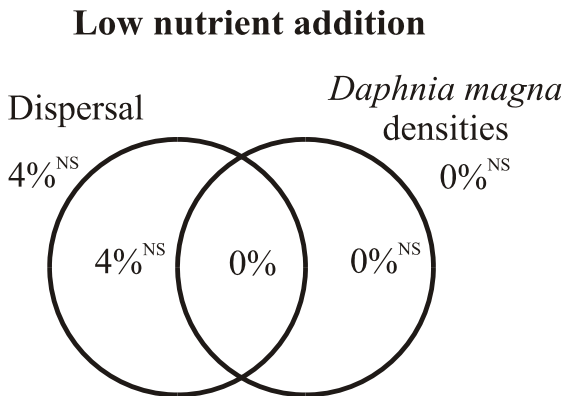
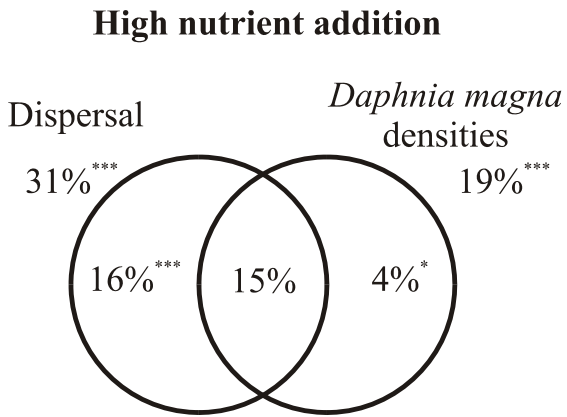
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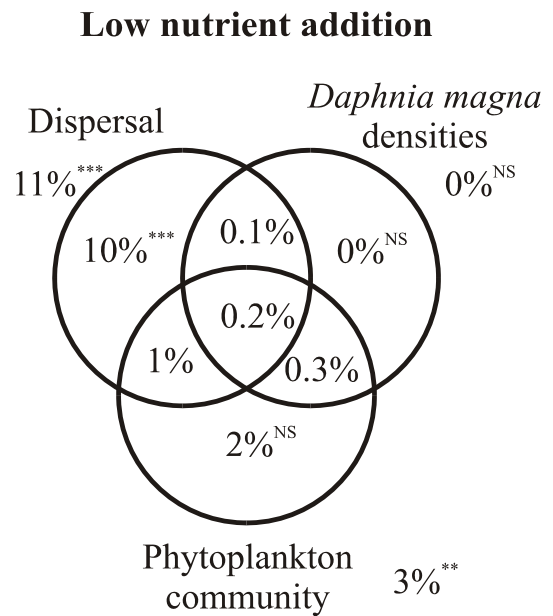
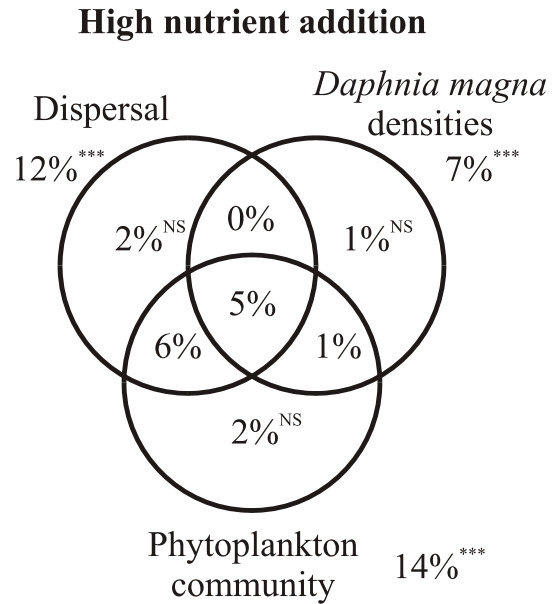
722 Figure 3.

723

Phytoplankton community



Bacterioplankton community



724

725

726