

# Density-dependent dispersal and relative dispersal affect the stability of predator-prey metacommunities

C. Céline Hauzy, Mathias Gauduchon, Florence D. Hulot, Michel Loreau

#### ▶ To cite this version:

C. Céline Hauzy, Mathias Gauduchon, Florence D. Hulot, Michel Loreau. Density-dependent dispersal and relative dispersal affect the stability of predator-prey metacommunities. Journal of Theoretical Biology, 2010, 266 (3), pp.458. 10.1016/j.jtbi.2010.07.008. hal-00616256

HAL Id: hal-00616256

https://hal.science/hal-00616256

Submitted on 21 Aug 2011

**HAL** is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

## Author's Accepted Manuscript

Density-dependent dispersal and relative dispersal affect the stability of predator-prey metacommunities

Céline Hauzy, Mathias Gauduchon, Florence D. Hulot, Michel Loreau

PII: S0022-5193(10)00357-7

DOI: doi:10.1016/j.jtbi.2010.07.008

Reference: YJTBI 6070

To appear in: Journal of Theoretical Biology

Received date: 29 September 2009

Revised date: 12 July 2010 Accepted date: 12 July 2010



www.elsevier.com/locate/yjtbi

Cite this article as: Céline Hauzy, Mathias Gauduchon, Florence D. Hulot and Michel Loreau, Density-dependent dispersal and relative dispersal affect the stability of predator-prey metacommunities, *Journal of Theoretical Biology*, doi:10.1016/j.jtbi.2010.07.008

This is a PDF file of an unedited manuscript that has been accepted for publication. As a service to our customers we are providing this early version of the manuscript. The manuscript will undergo copyediting, typesetting, and review of the resulting galley proof before it is published in its final citable form. Please note that during the production process errors may be discovered which could affect the content, and all legal disclaimers that apply to the journal pertain.

### Density-dependent dispersal and relative dispersal affect

the stability of	predator-prey	metacommunities
------------------	---------------	-----------------

3

2

1

#### Authors and their affiliations:

Céline Hauzy a,b,c, Mathias Gauduchon d,e, Florence D. Hulot a,f & Michel Loreau g

5 6 7

8

9

10

11

12

13

14

15

16

17

- a. UMR7618, Laboratoire Bioemco, Ecole Normale Supérieure, 46 rue d'Ulm, F-75230 Paris Cedex 05, France
- b. UFR Sciences de la Vie, Université Pierre et Marie Curie Paris 6, 4 place Jussieu, 75252 Paris cedex 05, France
  - c. IFM Theory and Modelling, Linköping University, 581 83 Linköping, Sweden
- d. UMR7625, Laboratoire d'Ecologie et d'Evolution, Université Pierre et Marie Curie, 7 quai Saint Bernard, F-75252 Paris Cedex 05
  - e. UMR6117, Laboratoire de Microbiologie, Géochimie et Ecologie Marines, Centre d'Océanologie de Marseille, campus de Luminy, case 901, 13288 Marseille cedex 09
- f. Université Paris XI, UMR 8079, ESE, Bât. 362, 91405 Orsay, France
- g. Department of Biology, McGill University, 1205 avenue Docteur Penfield, Montreal, Quebec H3A 1B1, Canada

18 19 20

#### Corresponding Author: Céline Hauzy (celha@ifm.liu.se)

- 21 Present address: IFM Theory and Modelling, Linköping University, 581 83 Linköping,
- Sweden 22
- 23 Phone: 00 46 (0)13 28 26 84 24 Fax: 00 46 (0)13 28 26 11

2526

#### Authors e-mail:

- 27 Mathias Gauduchon: mathias.gauduchon@univmed.fr
- Florence D. Hulot: florence.hulot@u-psud.fr
  Michel Loreau: michel.loreau@mcgill.ca

30 31

#### Abstract:

- Although density-dependent dispersal and relative dispersal (the difference in dispersal rates
- between species) have been documented in natural systems, their effects on the stability of
- metacommunities are poorly understood. Here we investigate the effects of intra- and
- interspecific density-dependent dispersal on the regional stability in a predator-prey
- metacommunity model. We show that, when the dynamics of the populations reach
- equilibrium, the stability of the metacommunity is not affected by density-dependent
- dispersal. However, the regional stability, measured as the regional variability or the
- 39 persistence, can be modified by density-dependent dispersal when local populations
- 40 fluctuate over time. Moreover these effects depend on the relative dispersal of the predator
- and the prey. Regional stability is modified through changes in spatial synchrony.
- 42 Interspecific density-dependent dispersal always desynchronises local dynamics, whereas
- intraspecific density-dependent dispersal may either synchronise or desynchronise it
- depending on dispersal rates, Moreover, intra- and interspecific density-dependent dispersal
- 45 strengthen the top-down control of the prey by the predator at intermediate dispersal rates.
- As a consequence the regional stability of the metacommunity is increased at intermediate
- dispersal rates. Our results show that density-dependent dispersal and relative dispersal of
- species are keys to understanding the response of ecosystems to fragmentation.

49 50

#### Keywords:

dispersal behaviours, spatial synchrony, top-down control, regional variability, persistence

#### INTRODUCTION

3

4

5

6

7

8

9

10

11

12

13

14

15

16

17

18

19

1

Understanding the mechanisms that govern the stability of metacommunities is a key challenge to understand the response of ecosystems to fragmentation. Metacommunities are regional sets of local communities linked by dispersal, i.e., by the movement of individuals from a patch (emigration) to another (immigration) (Leibold et al. 2004). Both local and regional processes affect the stability of metacommunities. For instance, in predator-prey metacommunities, local processes such as local density regulation mechanisms drive local predator-prey population fluctuations, which may lead to local extinctions when communities are isolated (Gause 1934; Lotka 1925; Rosenzweig and MacArthur 1963; Volterra 1926). But dispersal, which is a regional process, may allow regional persistence of the prey and the predator (Huffaker 1958) because it allows recolonization of vacant patches (Levins 1970) or rescue effects (Brown and Kodric-Brown 1977). This positive effect of dispersal on the regional stability of predator-prey metacommunities requires that dynamics are spatially asynchronous. If local populations fluctuate strongly and synchronously, the whole metapopulation may go extinct because of simultaneous extinction of local populations. But if fluctuations of local densities are asynchronous, recolonizations or rescue effects can occur and the persistence and stability of the metacommunity is improved (Comins and Blatt 1974; Hassell et al. 1991; Jansen 2001).

20

21

22

23

24

25

26

But dispersal and synchrony between communities are not independent processes.

On the one hand, dispersal may increase spatial synchrony because exchanges of individuals between communities tend to reduce differences in population sizes, especially when dispersal is high (Bjornstad *et al.* 1999). On the other hand, weak dispersal may also synchronise dynamics in predator-prey metacommunities with nonlinear density regulation (Bjornstad 2000; Jansen 1999). Thus, both dispersal and local density regulation can

synchronise community dynamics. Since there is empirical evidence for asynchronous
spatial dynamics in predator-prey metacommunities (Holyoak and Lawler 1996a; Holyoak
and Lawler 1996b; Janssen et al. 1997; Nachman 1991; van de Klashorst et al. 1992), some
mechanisms maintain persistent differences between local densities and counteract the
synchronising effect of dispersal. These mechanisms have been explored using a number of
models, and involve either stochastic or deterministic processes. Asynchrony between
patches of the same quality can be due to independent stochastic local factors (Taylor 1990),
such as demographic stochasticity in small populations (Nachman 1987a) and stochasticity
of dispersal between patches or of environmental variations in population parameters
(Crowley 1981; Reeve 1988; Reeve 1990). These models show that an intermediate
dispersal intensity of the prey and the predator insures metacommunity persistence
(Crowley 1981; Nachman 1987b; Nachman 1991; Reeve 1988; Zeigler 1977 reviewed in
Holyoak and Lawler 1996b). Dispersal must be sufficiently high to allow recolonization of
extinct patches, but sufficiently low to maintain the desynchronising effect of stochastic
local processes. Spatial asynchrony can also emerge from the interaction between local
dynamics and dispersal (Jansen 1995). In deterministic predator-prey metacommunities with
nonlinear density regulation, spatial asynchrony between patches arises when dispersal rates
are either low or intermediate. In these models, population dynamics are synchronous when
dispersal is high, and temporal variation in regional densities is decreased when dispersal
rates are decreased.
In this paper, we investigate the influence of two key aspects of dispersal behaviours
on the stability of predator-prey metacommunities: the density dependence in dispersal, and

on the stability of predator-prey metacommunities: the density dependence in dispersal, and the ratio of predator and prey dispersal rates, which we name "relative dispersal". Density-dependent dispersal, i.e. dispersal changes in response to changes of species densities in the donor patch, has been observed in several experimental systems (see for instance Bernstein

1984; French and Travis 2001; Hauzy et al. 2007). It implies a direct effect of local
processes (population dynamics) on regional processes (dispersal). A species' population
density can affect its own dispersal (intraspecific density-dependent dispersal) or the
dispersal of another species (interspecific density-dependent dispersal). Recent models
including density-dependent dispersal in predator-prey metacommunities yielded contrasted
results. McCann et al. (2000) suggested that local predator population outbreaks do not
depend on interspecific density-dependent dispersal. By contrast, Li et al. (2005) showed
that interspecific density-dependent dispersal can reduce spatial synchrony in both the prey
and the predator. Most of these studies, however, have assumed that prey and predator
dispersal rates are equal. But empirical data suggest that dispersal abilities of prey and
predator can be very different. The large range of predator-prey body-size ratios observed in
real food webs (Brose et al. 2006) and the allometric relationship between body size and the
scale of species' movements (Rooney et al. 2008) or dispersal abilities (Jenkins et al. 2007)
suggest that the relative dispersal rate of the predator compared to the prey can vary within a
large range. In some ecosystems, scale of movement is strongly positively related to trophic
position, suggesting a higher dispersal rate for the predator than for the prey. In other
systems, predators can even be less mobile than their prey, suggesting a higher dispersal rate
for the prey than for the predator. Differences in the dispersal abilities of the prey and the
predator can have important consequences, and yet have received little attention so far.

Here we develop a deterministic metacommunity model of a prey and a predator that includes density-dependent dispersal. We compare the effects of both intra- and interspecific density-dependent dispersal on the dynamics and the stability of the prey and the predator at the regional scale with those of constant dispersal. We study the effects of density-dependent dispersal in two contrasting cases of relative dispersal: either predator dispersal is distinctly higher than prey dispersal or prey dispersal is distinctly higher than

1	predator disp	persal. These	comparisons	allow us to	assess the	effects of	density-de	penden

- dispersal and relative dispersal on the relationship between dispersal and the stability of
- 3 predator-prey metacommunities.

4

5

#### THE MODEL

2

3

4

5

6

7

8

9

10

11

12

1

We consider a patchy landscape consisting of several localities. Each locality is occupied by a local predator-prey community. Within each community, local dynamics follow a Rosenzweig-MacArthur predator-prey model (Rosenzweig and MacArthur 1963): prey growth is logistic, the predator has a Holling type II functional response and its mortality rate is constant. Local communities are linked by dispersal of both species from one patch to another. A collection of such communities constitutes a predator-prey metacommunity. For the sake of simplicity, and following Jansen (1995, 2001), we study a 2-patch predator-prey metacommunity in which the two patches are identical. The dynamics of the prey and the predator in the two patches (1 and 2) are given by a system of ordinary differential equations:

13

13
$$\begin{cases}
\frac{dN_{1}}{dt} = r N_{1} \left( 1 - \frac{N_{1}}{K} \right) - \frac{b N_{1} P_{1}}{c + N_{1}} - N_{1} D_{n} (N_{1}, P_{1}) + N_{2} D_{n} (N_{2}, P_{2}) \\
\frac{dP_{1}}{dt} = \frac{b N_{1} P_{1}}{c + N_{1}} - m P_{1} - P_{1} D_{p} (N_{1}, P_{1}) + P_{2} D_{p} (N_{2}, P_{2}) \\
\frac{dN_{2}}{dt} = r N_{2} \left( 1 - \frac{N_{2}}{K} \right) - \frac{b N_{2} P_{2}}{c + N_{2}} - N_{2} D_{n} (N_{2}, P_{2}) + N_{1} D_{n} (N_{1}, P_{1}) \\
\frac{dP_{2}}{dt} = \frac{b N_{2} P_{2}}{c + N_{2}} - m P_{2} - P_{2} D_{p} (N_{2}, P_{2}) + P_{1} D_{p} (N_{1}, P_{1})
\end{cases}$$
15

15

16

17

18

19

20

where  $N_i$  and  $P_i$  denote the population densities of the prey and the predator, respectively, in patch i = 1, 2. As patches are identical, species parameters are the same in both patches. rand K denote the growth rate and carrying capacity of the prey; b and c are the saturation value and the half-saturation constant of the predator functional response; m is the predator mortality rate. Conversion efficiency from prey to predator has been set to 1. This does not

constitute an ecological assumption but arises from a simple mathematical rescaling of  $N_i$  or

 $2 P_i$ .

3

1

The last two terms of equations describe dispersal from and to each patch. We make 4 5 the ecological hypothesis that migration is not costly; as a consequence, the emigration term 6 in one patch corresponds exactly to the immigration term in the other. This assumption fits 7 in well with dispersal through corridors that allow fast moves with limited mortality risks. Dispersal rates are determined locally: the *per capita* dispersal rates from patch *i* of the prey 8  $D_n(N_i, P_i)$  and of the predator  $D_n(N_i, P_i)$  are functions of species densities in the donor patch 9 10 i only. The dispersal rate of each species may depend on its own local density, which we 11 refer to as "intraspecific density-dependent dispersal", and on the local density of the other 12 species, which we refer to as "interspecific density-dependent dispersal" (Hauzy et al. 13 2007). Our mathematical analysis deals with the general model, but our numerical 14 simulations investigate the separate effects of intraspecific and interspecific density 15 dependence in dispersal. The dispersal rates then simplify into  $D_n(N_i)$  or  $D_n(P_i)$  and  $D_p(P_i)$ 16 or  $D_p(N_i)$ . In numerical simulations, density dependence in dispersal is modelled through 17 Hill functions, which have been chosen for their convenient plasticity: the per capita

19 
$$D_n(N_i) = d_n \frac{N_i^{x_{nn}}}{S_{nn}^{x_{nn}} + N_i^{x_{nn}}} \quad \text{or} \quad D_n(P_i) = d_n \frac{P_i^{x_{np}}}{S_{np}^{x_{np}} + P_i^{x_{np}}}$$

and the *per capita* dispersal rate of the predators writes

dispersal rate from patch i of the prey writes

21 
$$D_p(P_i) = d_p \frac{P_i^{x_{pp}}}{S_{pp}^{x_{pp}} + P_i^{x_{pp}}} \quad \text{or} \quad D_p(N_i) = d_p \frac{N_i^{x_{pn}}}{S_{pn}^{x_{pn}} + N_i^{x_{pn}}}.$$

22

23

24

18

These formulations imply several assumptions. Species have a maximal *per capita* dispersal rate  $d_n$  for the prey, and  $d_p$  for the predator (fig. 1). These parameters can reflect

1	the ability of the species to disperse or the distance between patches. The half-saturation
2	parameters $S_{\alpha\beta}$ are the densities of species $\beta$ (prey: $\beta$ stands for "n"; or predators: $\beta$ stands
3	for "p") for which the <i>per capita</i> dispersal rates $D_{\alpha}$ of species $\alpha$ ( $\alpha$ = "n", "p") is half of the
4	maximal per capita dispersal rate $d_{\alpha}$ . Parameters $x_{\alpha\beta}$ determine the shape of density
5	dependence in dispersal. When $x_{\alpha\beta} = 0$ , dispersal is independent of density and occurs at a
6	constant <i>per capita</i> rate $d_{\alpha}/2$ ; when $x_{\alpha\beta} > 0$ , the dispersal rates increase with density;
7	whereas when $x_{\alpha\beta} < 0$ , they decrease with density. The absolute value of $x_{\alpha\beta}$ determines the
8	steepness of density dependence: when $ x_{\alpha\beta}  = 1$ , dispersal rates are concave functions of
9	density up to their saturation level (cyrtoid shape), whereas they are sigmoidal when $ x_{\alpha\beta}  > 1$
10	(the inflexion point of the dispersal function then occurs at the threshold density $S_{\alpha\beta}$ ); high
11	values of $ x_{\alpha\beta} $ model step-like sigmoid curves, whereas low values model smoother
12	variations in dispersal with population density.
13	
14	In this paper, we explore density-dependent dispersal behaviours that mimic those
15	observed in several experiments (Bernstein 1984; French and Travis 2001; Hauzy et al.
16	2007), i.e., prey dispersal increases with prey and predator densities (positive intra- and
17	interspecific density-dependent dispersal in the prey), and predator dispersal increases with
18	predator density (positive intraspecific density-dependent dispersal in the predator) and
19	decreases with prey density (negative interspecific density-dependent dispersal in the
20	predator). In terms of model parameters, this implies $x_{nn} > 0$ , $x_{np} > 0$ , $x_{pp} > 0$ and $x_{pn} < 0$ . We
21	conduct both a mathematical and a numerical analysis of the model.
22	
23	

$^{\circ}$	
Z	

3	Stability of the symmetric equilibrium
4	
5	We first focus on equilibriums of the model, and more specifically on the symmetric
6	equilibrium, i.e., that for which densities are equal in patches 1 and 2 ( $N_1=N_2$ and $P_1=P_2$ ).
7	We analyse the stability of this equilibrium using the linearization method developed by
8	Jansen (1994). The original method considers a predator-prey metacommunity model with
9	constant dispersal. We extend the method to general forms of density-dependent dispersal
10	and apply it to our model (Appendix A).
11	
12	The stability of the symmetrical equilibrium depends only on the local parameters $r$ ,
13	b, c, m and $K$ . The equilibrium is stable if and only if the prey carrying capacity $K$ is below a
14	threshold value: $K < c(b+m)/(b-m)$ (Appendix A). For higher values of the prey carrying
15	capacity, it is always unstable and a stable limit cycle appears. These results for density-
16	dependent dispersal generalize those obtained for constant dispersal by Jansen (1995) and
17	correspond to the well-known paradox of enrichment in the non-spatial MacArthur-
18	Rosenzweig model. Neither prey nor predator dispersal affects the stability of the symmetric
19	equilibrium and this conclusion is unaffected by density dependence in dispersal. A
20	metacommunity exhibits the same sudden destabilisation of its dynamics when $K$ increases
21	beyond its threshold whether dispersal is constant or density dependent.
22	
23	Regional densities at the symmetric equilibrium

Next, we analyse the effect of density-dependent dispersal on the total (regional)
population densities of the prey and the predator at the symmetric equilibrium. From
equations (1), we see that emigration from patch $i$ and immigration to patch $i$ cancel each
other out for each species, so that the regional densities of the prey and the predator are
independent of dispersal. Thus, prey and predator densities at the symmetric equilibrium are
unaffected as well by dispersal (Jansen 1995), whether dispersal is constant or density
dependent. When the metacommunity has reached equilibrium, local population densities
are the same as in isolated communities. Asymptotic population densities at the symmetric
equilibrium depend only on local processes. In our two-patch predator-prey
metacommunity, when $K$ is below the instability threshold, the asymptotic regional densities
are simply equal to the values they would have in an isolated patch.
Accepted manus

#### NUMERICAL ANALYSIS

$\mathbf{a}$
Z
_

When K > c(b+m)/(b-m), There is no stable equilibrium any longer, and the prey and the predator show persistent oscillations through time. We used numerical simulations to investigate the effect of density-dependent dispersal on prey and predator asymptotic dynamics at the scale of the metacommunity.

*Methods* 

We studied the various types of dispersal separately, *i.e.*, we studied constant dispersal in both species, intraspecific density-dependent dispersal in either the prey or the predator, and interspecific density-dependent dispersal in either the prey or the predator. For each of these five types of dispersal, we also studied two contrasting relative dispersal rates, one in which the maximal dispersal rate of the predator,  $d_p$ , is lower than that of the prey,  $d_n$  ( $d_p = 0.01 \ d_n$ ) and one in which the converse is true ( $d_p = 100 \ d_n$ ). Maximal dispersal rates varied between  $10^{-4}$  and  $10^2$  when  $d_n > d_p$ , and between  $10^{-6}$  and  $10^0$  when  $d_p > d_n$ . Within these ranges, we took 1000 sample values of each parameter (keeping the ratio  $d_n/d_p$  constant), uniformly distributed on a log scale. The parameters of local interactions were the following: r=10, K=25, b=11, c=1, m=10. For the density dependence in dispersal, we focused the numerical analysis on sigmoidal shapes ( $|x_{ap}|=4$ ,  $\alpha=n,p$ ;  $\beta=n,p$ ) and on values for the half-saturation parameters  $S_{ap}$  set close to the mean densities observed when populations are isolated:  $S_{nn}=S_{pn}=18$  and  $S_{np}=S_{pp}=2$ .

The system of ordinary differential equations (1) was numerically simulated by means of C programs using the GNU Scientific Library. Numerical integrations used the embedded Runge-Kutta Prince-Dormand method. Based on preliminary inspection of numerous simulations, we estimated a time delay before population dynamics entered its

1

2

3

4

5

6

7

8

9

10

11

12

13

14

15

16

17

18

19

20

21

22

23

24

25

26

asymptotic regime. This "asymptotic time" ( $t_{asympt}$ ) was much wider for low dispersal rates; it was then to 2 x 10<sup>4</sup> time units for high dispersal rates (Min ( $d_n$ ,  $d_p$ ) > 10<sup>-4</sup>) and 2 x 10<sup>5</sup> time units for low dispersal rates (Min ( $d_n$ ,  $d_p$ ) < 10<sup>-4</sup>). Our results were obtained from population dynamics observed between  $t_{asympt}$  and  $t_{asympt}$  + 10<sup>4</sup>, except for the first Lyapunov exponent (see below), which was calculated on dynamics between  $t_{asympt}$  + 10<sup>4</sup> and  $t_{asympt}$  + 2.10<sup>4</sup>.

All the variables and statistics considered in this study are related to the asymptotic dynamical regime and are computed on the attractors reached by population dynamics. Therefore, we needed a method to detect and follow the system's attractors as some control parameter (here, the maximal dispersal rates) is varied within a study interval. Except for catastrophic changes in the configuration of attractors, smooth variations in control parameters generate smooth "continuation branches", which correspond to the curves (or pieces of curves) shown on figures 2, 3, 4, 5, S3 and S4. Our model is complicated by the possibility of multiple coexisting attractors and chaotic regimes. Instead of using techniques based on numerical continuation of attractors (Kuznetsov 2004), which proved inefficient and intractable, we used a "sieve" technique of exploration, followed by the ex post facto reconstruction of "continuation branches". For each parameter value, the initial densities of the prey and the predator in the two patches were randomly sampled within ranges [0;20] for the prey and [0:10] for the predator as these ranges were commonly observed in the numerical simulations. This random sampling allowed us to explore the various basins of attraction of the possibly multiple attractors, and hence to detect the main attractors (i.e., those with a wide enough basin of attraction) by means of simulations runs from these initial population densities. The drawback of this method is that only one initial condition is simulated for each set of parameter values, so that in the case of multiple coexisting attractors, only one of them is reached. But, provided that parameter values are sampled densely enough in the study interval (we used a fine 1000-values sampling grid for the

maximal dispersal rates), the other attractors are very likely to be detected for neighboring
values of the parameter. This eventually results in a slightly scattered general portrait of the
various "continuation branches". The portrait "gaps" are then filled out by linking pairs of
detected attractors, provided that parameter values are not too distant and the attractors are
close enough in the state space. Attractor closeness was measured by means of the
Hausdorff distance in $\mathfrak{R}^4$ .
Analyses of dynamics were performed under the Matlab environment (R2008a). We
studied two aspects of metacommunity dynamics: (1) the nature of the attractor, and (2) the
degree of spatial synchrony. First, we determined whether population fluctuations were
periodic or chaotic by numerically estimating the first Lyapunov exponent. Computation of
the Lyapunov exponent in fast oscillating models like ours gives easily rise to an important
computational error accumulation. For the sake of numerical accuracy, the first Lyapunov
exponent was estimated on a Poincaré section of the orbit, as the exponential increase of
deviations from the iterations of the Poincaré map (Wolf et al. 1985). In each simulation, we
computed the Poincaré section that intersects the trajectory orthogonally at the asymptotic
time $t_{asympt}$ . We checked that the $10^4$ -time-unit interval we used to compute the Lyapunov
exponent provided satisfactory converge of the estimate. A positive value of the first
Lyapunov exponent reveals a chaotic attractor, while a negative value indicates a stable
equilibrium on the Poincaré map, which corresponds to a stable limit cycle for the
population dynamics in $\Re^4$ . Second, to quantify the effects of dispersal behaviours on
spatial synchrony in the asymptotic regime, we used the Pearson moment correlation
coefficient between local densities in the two patches. As an alternative synchrony estimate,
the correlation of the per capita population growth rates is recommended when population
dynamics exhibits long-term transient trends, as often encountered in field or experimental
measurements (Bjornstad et al. 1999). But as we studied the asymptotic regime here,
Pearson correlation between local densities is appropriate. The higher spatial synchrony, the

closer to 1 the correlation coefficient. We investigated spatial s	synchrony for both the prey
and the predator.	

Because many definitions of stability exist in the literature (see e.g. a review in Loreau *et al.* 2002), the stability of the metacommunity can be defined in several ways. Two stability properties are regional persistence and the temporal variability of regional densities. Because our model is purely deterministic and does not allow for extinctions, persistence cannot be estimated directly. Persistence is here assessed indirectly by means of the minimal regional densities reached in the asymptotic regime; the lower these minimal densities, the higher the vulnerability of species to stochastic extinction, and the lower the probability of persistence. Results are presented in the supplementary material (supplementary material, fig. S3 and S4). We also measured the temporal variability of regional densities by their coefficient of variation (CV). To disentangle the effects of the temporal mean of regional densities and of their standard deviation, however, we present both the temporal mean and the CV of regional densities. Stability increases when the CV decreases.

In order to investigate the robustness of our results with regard to changes in dispersal parameters, we used a factorial design in which the independent variables were the shape of the density dependence of the dispersal rates ( $|x_{\alpha\beta}|$ ), their half-saturation parameters ( $S_{\alpha\beta}$ ), and their maximal values for both the prey and the predator ( $d_n$  and  $d_p$ ). We compared two different shapes, cyrtoid ( $|x_{\alpha\beta}|=1$ ) and sigmoid ( $|x_{\alpha\beta}|=4$ ). We varied the half-saturation parameters  $S_{\alpha\beta}$  by 20% above and below the mean densities when population are isolated ( $S_{nn}=S_{pn}=18$  and  $S_{np}=S_{pp}=2$ ). We also varied the maximal dispersal rates of the prey ( $d_n$ ) and the predator ( $d_p$ ) between  $10^{-6}$  and  $10^2$ . Within this range, we took a sample of 100 values for each parameter  $d_n$  and  $d_p$  (and up to 200 values for the reference parameter combination  $|x_{\alpha\beta}|=4$ ,  $S_{nn}=S_{pn}=18$  and  $S_{np}=S_{pp}=2$ ), uniformly distributed on a log scale (see supplementary material, fig. S1 and S2).

1	

#### Metacommunity dynamics

We first focus on two descriptors of metacommunity dynamics, the nature of the dynamics (chaotic or periodic), and the spatial synchrony between patches. The first Lyapunov exponent characterises the nature of dynamics of the whole community. The dynamics of the two species cannot be separated; they are either both periodic or both chaotic. Spatial synchrony was calculated for each species separately, but proved to be completely similar between the prey and the predator in all simulations. Therefore we present below results for spatial synchrony of the prey only. For the sake of simplicity, we present results for two contrasted relative dispersal rates but these results are robust to wide variations in relative dispersal rates (supplementary material, fig. S1 and S2). The different shapes  $(x_{\alpha\beta})$  and half-saturation values  $(S_{\alpha\beta})$  of density dependence give qualitatively the same results for each type of dispersal, except for intraspecific density-dependent dispersal in the prey, which is sensitive to  $|x_{\alpha\beta}|$  (supplementary material, fig. S2). Below, we present the results for  $|x_{\alpha\beta}|=4$  and mention when they are different for  $|x_{\alpha\beta}|=1$ .

At high dispersal rates, population dynamics are synchronous and periodic for all types of dispersal behaviours (fig. 2). The metacommunity model is then equivalent to a non-spatial predator-prey model. In contrast, at low and intermediate dispersal rates, synchrony depends both on the relative dispersal rate of the prey and the predator and on the type of dispersal.

When dispersal is density-independent (constant) and higher in the prey than in the predator (fig. 2A), spatial synchrony decreases monotonically with dispersal decrease, through both desynchronisation of periodic dynamics and the emergence of chaotic dynamics at intermediate dispersal rate. This decrease is rather low and the minimum value

1	of synchrony is still high. In contrast, when predator dispersal is higher than prey dispersal
2	(fig. 2D), the dynamics can remain perfectly synchronous for all dispersal rates or become
3	very asynchronous, provided that the dispersal rates are not too high. Indeed, for low and
4	intermediate dispersal rates, the metacommunity dynamics shows two coexisting periodic
5	attractors, the first one characterized by a perfect spatial synchrony while the second one
6	shows a strong asynchrony between patches. The synchrony of the later decreases with
7	increasing dispersal rates until the attractor disappears, letting the synchronous attractor
8	alone. This coexistence of synchronous and asynchronous attractors has been observed by
9	Jansen (1995). No chaotic dynamics is observed for this relative dispersal rate.
10	Density-dependent dispersal in the prey (fig. 2B, 2E) affects the dynamics only when
11	prey dispersal is higher than predator dispersal $(d_n > d_p)$ (fig. 2B). When predator dispersal
12	is higher than prey dispersal $(d_p > d_n)$ (fig. 2E), the dynamics is very similar to the case
13	where dispersal is constant (fig. 2D). When $d_n > d_p$ , intra- and interspecific density-
14	dependent dispersal in the prey have different effects on spatial synchrony (fig. 2B).
15	Intraspecific density-dependent dispersal synchronises local dynamics when dynamics are
16	periodic (at low and intermediate dispersal rates), whereas spatial synchrony is decreased
17	when dynamics are chaotic. The presence of the synchronising effect of intraspecific
18	density-dependent dispersal does not depend on the shape of density dependence in
19	dispersal ( $ x_{\alpha\beta} =4$ : fig. 2B; $ x_{\alpha\beta} =1$ : Appendix B, fig. 5A), but the presence of chaotic
20	dynamics does. When $ x_{\alpha\beta} =4$ chaotic dynamics arise for a wider range of dispersal rates
21	than when dispersal is constant (fig. 2A, 2B) whereas when $ x_{\alpha\beta} =1$ chaotic dynamics
22	disappear (Appendix B, fig. 5A). On the contrary, when prey dispersal is subject to
23	interspecific density dependence, population dynamics are strongly desynchronised whether
24	the dynamics be periodic or chaotic (fig. 2B).
25	Whatever the relative dispersal rates, intraspecific density-dependent dispersal in the
26	predator does not affect spatial synchrony (fig. 2C, 2F). The effect of density-dependence in

predator dispersal shows up when it is interspecific and when predator dispersal is higher
than prey dispersal (fig. 2F). Indeed, the bistable regime observed for constant dispersal
disappears through the loss of the synchronous limit cycle. For low and intermediate
dispersal rates, the remaining limit cycle shows strong asynchronous dynamics. The
dynamics abruptly become synchronous at intermediate dispersal rate and the transition is
accompanied with the emergence of a new chaotic window. Thus, for interspecific density
dependent dispersal in the predator, spatial synchrony dependence upon dispersal rate
simplifies into a global increasing relationship.

#### Regional densities

The temporal mean of regional densities of the prey and the predator are affected by dispersal rates in the same way for all types of dispersal behaviour and of relative dispersal. For the same range of dispersal rates, prey regional density decreases (fig. 3) while predator regional density increases (fig. 4) in a correlated way. This suggests that dispersal modifies the strength of the impact of predator populations on prey populations at the regional scale. First, changes in prey and predator regional densities occur for dispersal rate in chaotic windows and for dispersal rate close to chaotic window but at which dynamics are periodic (fig. 3A, 3B, 3C, 3F; fig. 4A, 4B, 4C, 4F). Second, these changes occur for asynchronous limit cycle which exists when predator dispersal is higher than prey's one (fig. 3D, 3E, 3F; fig. 4D, 4E, 4F). While the asynchronous limit cycle deeply differs from the synchronous one in terms of spatial synchrony, it shows identical mean regional densities at low dispersal rates. But when dispersal rates increase, prey regional density slowly decreases while predator regional density increases. These changes in mean regional densities are not accompanied with chaotic dynamics.

Moreover, these mean regional densities can be quantitatively affected by dispersal	
behaviours and relative dispersal: the mean regional density of the prey decreases and the	
mean regional density of the predator increases more strongly when $d_n \ge d_p$ and there is	
intraspecific density-dependent dispersal for the prey (only when $ x_{\alpha\beta} =4$ : fig. 3B; fig. 4B;	
$ x_{\alpha\beta} =1$ : Appendix B, fig. 5B,C) or interspecific density-dependent dispersal for the prey (fig.	g.
3B; fig. 4B), or when $d_p > d_n$ and there is interspecific density-dependent dispersal for the	
predator (fig. 3F; fig. 4F). These changes commonly go hand in hand with the emergence of	of
new chaotic regimes (fig. 3F; fig. 4F). As a consequence, density-dependent dispersal and	
relative dispersal can increase the strength of the impact of predator populations on prey	
populations at the regional scale, and thus increase the control of the predator on the prey.	

### Regional variability

The regional variability of prey and predator populations, as measured by the CV of their regional densities, may be strongly affected by density-dependent dispersal through both spatial synchrony and predator control on the prey. We should expect low regional variability when local dynamics are asynchronous and high regional variability when local dynamics are synchronous. Predator control on the prey should also affect the regional variability through changes in mean regional densities: this control should decrease predator regional variability and should increase prey regional variability.

When dispersal is constant, the regional variability of the prey and the predator is modified by their relative dispersal (fig. 3A, 3D; fig. 4A, 4D). When prey dispersal is higher than predator dispersal ( $d_n > d_p$ ) (fig. 3A; fig. 4A), the regional variability of the prey and of the predator at low dispersal rates is slightly lower than their regional variability at high dispersal rate. This is caused by desynchronisation of dynamics at low dispersal rates for

both species (fig. 2A). But at intermediate dispersal rate, the regional variabilities of the
prey and the predator vary in different way. Two mechanisms act in opposite direction for
the prey and in the same direction for the predator. The increase in asynchrony tends to
lower regional variability of both the prey and the predator, whereas predator control
increases regional variability of the prey and decreases the predator's. As a result, the
regional variability of the prey increases slightly while the predator's decreases strongly.
When predator dispersal is higher than prey dispersal $(d_p > d_n)$ (fig. 3D; fig. 4D), the
regional variability of the prey and the predator can decrease when dispersal rates decrease.
Indeed, at intermediate and low dispersal rates, two limit cycles coexist: the synchronous
one has high regional variability and the asynchronous one shows low regional variability.
Moreover, regional variability of the prey and the predator on this asynchronous limit cycle
vary with increasing dispersal rates. The increase in asynchrony tends to lower regional
variability of both the prey and the predator, whereas predator control increases regional
variability of the prey and decrease the predator's. As a result, the regional variability of the
prey increases slightly while the predator's decreases strongly. Thus, the effect of dispersal
on a species regional variability depends on its trophic position.
When dispersal rates are high, as in a well-mixed system, the regional variabilities of
the prey and the predator are not affected by the type of dispersal behaviour and by the
relative dispersal. But when dispersal rates decrease, density-dependent dispersal can affect
regional variabilities of both species, depending on their relative dispersal rates (fig. 3B, 3C,
3E, 3F; fig. 4B, 4C, 4E, 4F).
Density-dependent dispersal in the prey can modify regional variability when prey
dispersal is higher than predator dispersal $(d_n > d_p)$ (fig. 3B; fig. 4B). Intraspecific density-
dependent dispersal in the prey can strongly modify regional variability of both species by
comparison to constant dispersal. At intermediate dispersal rates, prey and predator regional
variabilities vary in opposite way when $ x_{\alpha\beta} =4$ , because of the increase of spatial synchrony

(fig. 2B) and of predator control on the prey (fig. 3B; fig. 4B). Prey regional variability
increases or decreases compared to constant dispersal, depending on the relative effects of
these two mechanisms that act in opposite direction. On the contrary, these two mechanisms
both decrease predator regional variability. At low dispersal rates, intraspecific density-
dependent dispersal in the prey increases regional variability of both species because of
spatial synchrony of periodic dynamics when $ x_{\alpha\beta} =4$ and when $ x_{\alpha\beta} =1$ (Appendix B, fig. 5B,
C). The effects of interspecific density-dependent dispersal in the prey (fig. 3B; fig. 4B) on
the relationship between dispersal rates and regional variability are qualitatively the same as
under constant dispersal (fig. 3A; fig. 4A), but they are stronger. Indeed, prey and predator
regional variabilities are strongly lower at low dispersal rates than at high dispersal rates
because of strong spatial asynchrony at low dispersal rates. The transition between high and
low dispersal rates occurs with increase of spatial asynchrony and of predator control on the
prey. As a consequence, prey regional variability is increased whereas predator regional
variability is decreased compared to constant dispersal.
Only interspecific density-dependent dispersal in the predator modifies regional
variability when predator dispersal is higher than prey dispersal $(d_p > d_n)$ (fig. 3C, 3F; fig.
4C, 4F). Intraspecific density-dependent dispersal in the predator does not affect prey and
predator regional variability (fig. 3C, 3F; fig. 4C, 4F), which is consistent with the results on
spatial dynamics and regional densities (fig. 2-4C, 2-4F). When predator dispersal depends
on prey density, the synchronous limit cycle disappears at low and intermediate dispersal
rates, leaving only the asynchronous limit cycle. As a consequence, prey and predator
regional variabilities are strongly lower at low dispersal rates than at high dispersal rates,
whatever the initial densities are. Moreover, between low and high dispersal rates, prey
regional variability results of a balance between spatial synchrony and predator control on
the prey, whereas these two mechanisms act in same way to decrease predator regional
variability.

#### DISCUSSION

$^{\circ}$	
/	
_	

Our results when dispersal rates of the prey and the predator are constant, such as the potential coexistence of several attractors, are consistent with those obtained by Jansen (1995, 2001). They further show that density-dependent dispersal in both the prey and the predator and their relative dispersal rates may affect the nature of dynamics (periodic vs chaotic) and the spatial synchrony and stability of species when the symmetric equilibrium of the metacommunity is unstable and populations fluctuate over time.

## Dispersal behaviours and spatial synchrony

One important factor that affects metacommunity stability is the spatial synchrony of the population dynamics of the various species. Since a species is more vulnerable to extinction when all its population densities in a metacommunity are low at the same time, spatial synchrony decreases the probability of persistence of that species at the regional scale (Briggs and Hoopes 2004). Spatial synchrony in predator-prey metacommunities is a balance between synchronising and desynchronising processes. Processes that desynchronise dynamics include different forms of stochasticity (Taylor 1990), but also nonlinear trophic interactions (Bjornstad 2000; de Roos *et al.* 1991; Jansen 1999). Local dynamics may be synchronised by environmental forcing (the "Moran" effect), high dispersal rates (Bjornstad *et al.* 1999, Vasseur and Fox 2009) and nonlinear trophic interactions when dispersal rates are low, (Bjornstad 2000; de Roos *et al.* 1991; Jansen 1999).

Our results show that density-dependent dispersal behaviours can affect spatial synchrony in metacommunities at low and intermediate dispersal rates. In particular, interspecific density-dependent dispersal can strongly desynchronise dynamics. In a patchy

Rosenzweig-MacArthur model, Li et al. (2005) observed the same trend when prey
migration was negatively correlated to the gradient of predator density or when predator
migration followed the gradient of prey density. However, their study was focused on equal
dispersal rates in the prey and the predator. Here, we show that this effect depends on the
relative dispersal of the two species: interspecific density-dependent dispersal in the prey
has a strong impact when prey dispersal is higher than predator dispersal, whereas
interspecific density-dependent in the predator has a strong impact when predator dispersal
is higher than prey dispersal.
According Briggs and Hoopes (2004), spatial asynchrony in predator-prey
metacommunities is increased when local population dynamics are uncoupled from
immigration, i.e. when a local population receive more individuals when its density is low
than when its density is high. Because prey and predator densities do not fluctuate
synchronously within patches (predator abundance peaks after prey abundance), the
dependence of a species' dispersal on the density of another species (interspecific density-
dependent dispersal) could promote their uncoupling and increase spatial asynchrony.
We fund a stabilising effect of interspecific density-dependent dispersal in a patchy
Rosenzweig-MacArthur model with identical patches in agreement with Li et al.'s results
(2005) obtained in a similar model. By contrast, in patchy models with consistent
differences between patches in parameter values and local dynamics that obey either the
Lotka-Volterra model (Murdoch et al. 1992) or a variant with a type-II functional response
(Ives 1992), the aggregation of the predator in patches of high prey density can be either
synchronising or desynchronising. The desynchronising effect could come either from the
absence of density-dependence in the prey growth rate (e.g. logistic growth), which is
stabilising, or from the differences between patches.
Our results show also that the synchrony of spatial dynamics is affected by
intraspecific density-dependent dispersal in the prey. This dispersal behaviour can fully

synchronise population dynamics at low dispersal rates, but can also desynchronise local
dynamics at intermediate dispersal rates when the shape of density-dependence in dispersal
is sigmoidal. Synchronising and desynchronising effects have been observed also in single-
species metapopulation models (Ylikarjula et al. 2000). In our metacommunity model, we
observed this effect for the prey, but not for the predator. We show that the effects of
intraspecific density-dependent dispersal in the prey depend on relative dispersal: these
effects are strong when prey dispersal is higher than predator dispersal. Thus, density-
dependent dispersal and relative dispersal have interacting effects and play an important part
in regulating spatial synchrony in metacommunities.
The interaction between density-dependent dispersal and relative dispersal suggests
that density-dependent dispersal behaviours can have different effects on stability in
different types of predator-prey systems. Predators often disperse more than their prey
(Rooney et al. 2008). In such systems, our results suggest that interspecific density-
dependent dispersal in the predator can modify stability whereas density-dependent
dispersal in the prey does not. But prey sometimes disperse more than their predators
(Rooney et al. 2008). In such situations, density-dependent dispersal in the prey can have a
strong effect on the stability of the metacommunity: interspecific density-dependent
dispersal can increase spatial synchrony, whereas intraspecific density-dependent dispersal
can either decrease or increase it depending on dispersal rates.
Dispersal behaviours and top-down control
Dispersal venaviours and top-aown control
The extent to which populations are regulated by predation ("top-down" control) or
by the availability of resources ("bottom-up" control) has been intensely debated in ecology
(e.g. Hairston et al. 1960; Murdoch 1966; Oksanen et al. 1981). Most ecologists now agree

that both types of control are present in most ecosystems (Shurin et al. 2002; Borer et al.

1	2005) and that their relative strength depends on local factors such as functional diversity
2	(Hulot et al. 2000) and coevolution between prey and predators (Loeuille and Loreau 2004).
3	Spatial processes can also modify the strength of these controls. A recent model predicts
4	that nutrient inputs and immigration should generally increase the strength of trophic
5	cascades in open ecosystems (Leroux and Loreau 2008).
6	Moreover, predator-prey metacommunity models reveal that dispersal between local
7	communities can also modify the strength of top-down control. Stochastic predator-prey
8	metacommunity models have shown that the relative abundances of the prey and the
9	predator vary with dispersal rate: the mean of prey abundance decreases and the mean of
10	predator abundance increases at intermediate dispersal rates (e.g. Reeve 1988; Zeigler
11	1977). We fund this result in our deterministic predator-prey model. It is based on a type-II
12	functional response, which implies that the predator controls the prey at equilibrium. When
13	prey and predator local densities fluctuate over time, we fund that the temporal mean of
14	prey density decreases while the temporal mean of predator density increases at
15	intermediate dispersal rates. These variations are observed both at the regional scale (fig. 3-
16	4) and at the local scale (results not shown). Thus, results of our deterministic model and of
17	stochastic models suggest that intermediate dispersal rates increase the average strength of
18	the control of the prey by the predator in predator-prey metacommunities. Moreover our
19	results show that intra- and interspecific density-dependent dispersal in the prey.reinforces
20	the increase in the strength of predator control on the prey at intermediate dispersal rates.
21	The effect of dispersal on the strength of top-down control has some experimental
22	support. In aquatic predator-prey metacommunities, the mean density of the predator
23	(Didinium nasutum) was highest at intermediate dispersal rates whereas the mean density of
24	the prey (Colpidium striatum) was decreased by dispersal (Holyoak and Lawler 1996b). In
25	trophically more complex metacommunities, dispersal between mesocosms increased
26	predator biomass, decreased herbivore biomass and increased primary producer biomass

(Chase et al. 2010). This suggests that the trophic cascade was stronger in mesocosms
connected by dispersal than in isolated mesocosms. Thus, theory and experiments concur to
suggest that dispersal and other spatial flows across ecosystems (Leroux and Loreau 2008)
generally tend to increase the strength of top-down control food-web metacommunities.
Dispersal behaviours and the stability of metacommunities
Our results show that the synchrony of the prey and the predator are strongly related
in metacommunities. Indeed, spatial synchrony is affected in the same way in the prey and
in the predator by density-dependent dispersal and relative dispersal. In particular, density-
dependent dispersal in a species can modify not only its own spatial synchrony, as in
metapopulation models, but also the spatial synchrony of the other species. For instance,
interspecific density-dependent dispersal in the prey can increase both its spatial asynchrony
and that of the predator.
Metacommunity stability depends not only on spatial asynchrony but also on the
strength of prey control by the predator: as discussed above, density-dependent dispersal
increases predator control on the prey at intermediate dispersal rates. This generates a
decrease in regional variability of the predator, and an increase in that of the prey. Regional
persistence shows the same trends: the regional persistence of the predator is increased
while that of the prey is decreased, although the latter effect is smaller than for regional
variability (See supplementary material, fig. S3 and S4). Thus, the stability of the prey and
that of the predator display opposite responses to density-dependent dispersal because of the
increase in the average strength of predator control on the prey.
Although the regional stability (as measured by both persistence and variability) of
the predator is maximal at intermediate dispersal rates, it is nevertheless lower than that of

the prey. Thus, top-down control stabilises the less stable species (the predator) and

destabilises the more stable species (the prey). This suggests that the metacommunity as a
whole may be stabilised at intermediate dispersal. This conclusion is consistent with the
general finding of models including different forms of stochasticity that the persistence of
predator-prey metacommunities is enhanced at intermediate dispersal rates (Crowley 1981;
Reeve 1988; Zeigler 1977). The mechanism that is generally thought to generate the hump-
shaped relationship between the stability of predator-prey metacommunities and dispersal is
the balance between the positive effect of low dispersal (allowing recolonization and rescue
effects) and the negative effect of high dispersal (synchronisation). Here we suggest that
another mechanism could stabilise predator-prey metacommunities at intermediate dispersal
rates: the modification of the strength of top-down control, which increases the stability of
the most unstable species (the predator).
Our numerical results show a hump-shaped relationship between predator stability
and dispersal for both constant and density-dependent dispersal. But they suggest that the
hump can become more or less marked when dispersal is density dependent. For instance,
the hump is stronger when the prey has intraspecific density-dependent dispersal because
the latter destabilises the metacommunity at low dispersal rates (spatial synchronisation) and
increases the stability at intermediate dispersal rates (strengthening of predator control on
the prey). Although interspecific density-dependent dispersal in the prey stabilises the
metacommunity at low dispersal (desynchronisation of local dynamics), the increase in the
strength of predator control is sufficient to maintain the hump-shaped relationship between
predator stability and dispersal.
Thus, density-dependent dispersal can modify the stability of predator-prey
metacommunities by modifying both the spatial synchrony of local populations and the
strength of prey control by the predator. We hope that these results will stimulate further
work on the effects of dispersal on top-down control and the stability of food-web
metacommunities.

Ecosystem fragmentation and conservation management

Ecosystem fragmentation leads to decreased species dispersal between patches. Therefore studying the relationship between dispersal and metacommunity stability is of particular importance to understand the response of metacommunities to fragmentation. We showed that this response may depend on the existence of density-dependent dispersal in species and on their relative dispersal. If fragmentation decreases dispersal of only one species in the metacommunity because of its specific sensitivity to changes in environmental quality between patches, relative dispersal will be modified. In the case of constant prey dispersal and interspecific density-dependent dispersal for the predator, for instance, our model predicts that decreased predator dispersal only may lead to destabilisation of the metacommunity. Thus our work emphasizes the importance of density-dependent dispersal and relative dispersal of species in understanding the response of ecosystems to fragmentation.

This also suggests that enhancing the dispersal of one species through e.g. corridors may not be the best way to prevent its own extinction. For instance, if prey dispersal is constant and predator dispersal is lower and depends on prey density, the increase in prey dispersal by corridors will not increase its stability. In contrast, an increase in predator dispersal may stabilise both the prey and the predator if the ratio between the dispersal rates of the two species is reversed. Thus, our results emphasize the need to integrate conservation policies at the scale of communities and metacommunities, instead of considering them merely at the scale of populations and metapopulations.

-	ററ	NI	$\boldsymbol{c}$	1	ICI	Ĺ
						 ٦

Our work shows that density-dependent dispersal, which generates interactions between local and regional dynamics, and relative dispersal of the prey and the predator are key to understand the dynamics and stability of metacommunities. Density-dependent dispersal behaviours and relative dispersal deserve more theoretical and experimental explorations to increase our ability to predict the effects of habitat fragmentation on the stability of ecosystems.

#### **ACKNOWLEDGEMENTS**

We thank Vincent A. A. Jansen, Bernard Cazelles, Silvia de Monte, E. Peter Münger for stimulating discussions and Maurice Sabelis, Louis-Félix Bersier and two anonymous reviewers for providing comments and suggestions on the manuscript. Michel Loreau was supported by a Discovery Grant from the Natural Sciences and Engineering Research Council of Canada and by a team research project from the Quebec Nature and Technologies Research Fund. Céline Hauzy was supported by a grant from the Chancellerie des Universités de Paris.

#### APPENDIX A: STABILITY ANALYSIS OF SYMMETRICAL EQUILIBRIUM

- 2 Linearization of two-patch model with density-dependent dispersal
- 3

- We used the method proposed by Jansen (1994) and generalised it to density-
- dependent dispersal. System of equations (1), and more generally a two-patch model of
- 6 predator-prey dynamics, can be written in vectorial form into:

$$\begin{cases}
\frac{dX}{dt} = F(X) - G(X) + G(Y) \\
\frac{dY}{dt} = F(Y) - G(Y) + G(X)
\end{cases} (2)$$

- 8 where  $X = \begin{pmatrix} N_1 \\ P_1 \end{pmatrix}$  and  $Y = \begin{pmatrix} N_2 \\ P_2 \end{pmatrix}$  are the vectors of population densities of the prey
- 9 (N) and of the predator (P) in patches 1 and 2. The local dynamics of the prey and the
- predator are described by the functions  $F(X) = \begin{pmatrix} f_N(X) \\ f_P(X) \end{pmatrix}$  and  $F(Y) = \begin{pmatrix} f_N(Y) \\ f_P(Y) \end{pmatrix}$ . The
- dispersal dynamics of both species in patch 1 and 2 are described respectively by the

12 functions 
$$G(X) = \begin{pmatrix} g_N(X) \\ g_P(X) \end{pmatrix}$$
 and  $G(Y) = \begin{pmatrix} g_N(Y) \\ g_P(Y) \end{pmatrix}$ .

- The symmetric equilibrium of (2), noted  $(X^*, Y^*)$  is those for which X(t) = Y(t) for all
- *t*. In order to facilitate the analysis of symmetric equilibrium, we introduce
- $\Sigma = \frac{1}{2}(X + Y) = \begin{pmatrix} \Sigma_N \\ \Sigma_P \end{pmatrix}, \text{ the vector of average densities of each species over the two patches,}$
- and  $\Delta = X \Sigma = \Sigma Y = \frac{1}{2}(X Y) = \begin{pmatrix} \Delta_N \\ \Delta_P \end{pmatrix}$ , the vector of differences between densities in a
- patch and average densities in the metacommunity.
- 18 System (2) can then be recast into:

1 
$$\begin{cases} \frac{d\Sigma}{dt} = \frac{1}{2} \left[ F(\Sigma + \Delta) + F(\Sigma - \Delta) \right] \\ \frac{d\Delta}{dt} = \frac{1}{2} \left[ F(\Sigma + \Delta) - F(\Sigma - \Delta) \right] - G(\Sigma + \Delta) + G(\Sigma - \Delta) \end{cases}$$
(3)

- We are interested in the equilibrium  $(\Sigma^*, \Delta^*)$  of (3) and more particularly to the
- symmetric equilibria ( $\Sigma^*$ ,  $\Delta^*=0$ ). Note that for  $\Delta=0$ , the dynamics of the mean densities,  $\Sigma$ ,
- 4 is exactly the same as the local dynamics of densities in the absence of migration. In
- 5 particular, the equilibrium values of  $\Sigma$  are the equilibrium values of the local dynamics
- 6 determined by F. In order to study the stability near a symmetric equilibrium, the system (3)
- 7 is linearized around  $\Sigma = \Sigma^*$ ,  $\Delta = 0$ :

8 
$$\begin{cases} \frac{d\Sigma}{dt} = \boldsymbol{J}_{F}(\Sigma^{*}) \times (\Sigma - \Sigma^{*}) + o(\|\Sigma - \Sigma^{*}\| + \|\Delta\|) \\ \frac{d\Delta}{dt} = \left[\boldsymbol{J}_{F}(\Sigma^{*}) - 2\boldsymbol{J}_{G}(\Sigma^{*})\right] \times \Delta + o(\|\Sigma - \Sigma^{*}\| + \|\Delta\|) \end{cases}$$
(4.1)

where 
$$J_F(\Sigma^*) = \begin{pmatrix} \frac{\partial f_N}{\partial N} & \frac{\partial f_N}{\partial P} \\ \frac{\partial f_P}{\partial N} & \frac{\partial f_P}{\partial P} \end{pmatrix}_{\substack{N = \Sigma_N^* \\ P = \Sigma_P^*}}$$
 and  $J_G(\Sigma^*) = \begin{pmatrix} \frac{\partial g_N}{\partial N} & \frac{\partial g_N}{\partial P} \\ \frac{\partial g_P}{\partial N} & \frac{\partial g_P}{\partial P} \end{pmatrix}_{\substack{N = \Sigma_N^* \\ P = \Sigma_P^* }}$  are the

- Jacobian matrices of F and G.
- The stability of  $(\Sigma = \Sigma^*, \Delta = 0)$  is determined by the eigenvalues of matrices  $J_F(\Sigma^*)$
- and  $J_F(\Sigma^*) 2 J_G(\Sigma^*)$ . The equilibrium  $(\Sigma = \Sigma^*, \Delta = 0)$  of system (4), or the symmetrical
- equilibrium ( $X^*$ ,  $Y^*$ ) of system (2), is stable if both matrices  $J_F(\Sigma^*)$  and
- $J_F(\Sigma^*) 2 J_G(\Sigma^*)$  do not have eigenvalues with positive real part. Since these matrices are
- of dimension 2, this is equivalent to:

$$\begin{cases}
Tr(J_{F}(\Sigma^{*})) < 0 \\
\det(J_{F}(\Sigma^{*})) > 0
\end{cases}$$
and
$$\begin{cases}
Tr(J_{F}(\Sigma^{*}) - 2 J_{G}(\Sigma^{*})) < 0 \\
\det(J_{F}(\Sigma^{*}) - 2 J_{G}(\Sigma^{*})) > 0
\end{cases}$$

Application to the two-patch Rosenzweig – MacArthur model

2

- In order to determine the effects of density-dependent dispersal on the stability of the symmetric equilibrium of (1), we analyse the signs of the trace and of the determinant of the matrices  $J_F(\Sigma^*)$  and  $J_F(\Sigma^*) 2 J_G(\Sigma^*)$ , where F and G correspond to our specific model.
- 6 In our specific model, F and G write:

$$F(N,P) = \begin{pmatrix} rN\left(1 - \frac{N}{K}\right) - \frac{bNP}{c+N} \\ \frac{bNP}{c+N} - mP \end{pmatrix} \text{ and } G(N,P) = \begin{pmatrix} ND_N(N,P) \\ PD_P(N,P) \end{pmatrix}$$

- 8 , where  $D_N$  and  $D_P$  are the *per capita* dispersal rate of the prey and the predator. We
- 9 allow any form of density-dependant dispersal rates, provided that prey dispersal increases
- with prey density ( $\partial D_N/\partial N>0$ ) and predator density ( $\partial D_N/\partial P>0$ ) and that predator
- dispersal increases with predator density ( $\partial D_P/\partial P > 0$ ) and decreases with prey density
- 12  $(\partial D_P/\partial N < 0)$ . Thus, the sign structure of the Jacobian matrix of F and G at  $\Sigma^*$  are:

13 
$$J_F(\Sigma^*) = \begin{pmatrix} * & - \\ + & 0 \end{pmatrix} \text{ and } J_G(\Sigma^*) = \begin{pmatrix} + & + \\ - & + \end{pmatrix}$$

- 14 , where \* is determined by the sign of K c(b+m)/(b-m). The determinant of
- 15  $J_F(\Sigma^*)$  is positive and its trace is equal to the \* term. The \*term is negative, and thus  $\Sigma^*$  is a
- stable equilibrium of (4.1) if and only if K < c(b+m)/(b-m). At the transition, both
- eigenvalues real parts vanish, which indicates a (here supercritical) Hopf bifurcation. When
- 18 K passes the threshold value c(b+m)/(b-m), the equilibrium becomes unstable and
- dynamics converges to a stable limit cycle.
- Given \* term is negative, the sign structure of  $J_G$  implies that  $J_F(\Sigma^*) 2 J_G(\Sigma^*)$
- has a stable sign structure: when K < c(b+m)/(b-m), the trace of  $J_F(\Sigma^*) 2J_G(\Sigma^*)$  is

1	negative and its determinant is positive. As a consequence, when $K < c(b+m)/(b-m)$ , the
2	solution $\Delta^* = 0$ of (4.2) is a stable equilibrium.
3	Thus, the stability condition for the non-spatial Rosenzweig - MacArthur model
4	carries over to the symmetric equilibrium of the spatial model, whatever the dispersal
5	behaviours.
6	
7	APPENDIX B: SENSITIVITY TO SHAPE OF DENSITY DEPENDENCE IN
8	DISPERSAL
9	(fig. 5)
11	REFERENCES
12	Bernstein, C. 1984. Prey and predator emigration responses in the acarine system
13	Tetranychus urticae-Phytoseiulus persimilis. Oecologia 61:134-142.
14	Bjornstad, O. N. 2000. Cycles and synchrony: two historical 'experiments' and one
15	experience. Journal of Animal Ecology 69:869-873.
16	Bjornstad, O. N., R. A. Ims, and X. Lambin. 1999. Spatial population dynamics: analyzing
17	patterns and processes of population synchrony. Trends in Ecology & Evolution
18	14:427-432.
19	Borer, E. T., E. W. Seabloom, J. B. Shurin, K. E. Anderson, C. A. Blanchette, B. Broitman,
20	S. D. Cooper, and B. S. Halpern. 2005. What determines the strength of a trophic
21	cascade? Ecology 86:528-537.
22	Briggs, C. J., and M. F. Hoopes. 2004. Stabilizing effects in spatial parasitoid-host and
23	predator-prey models: a review. Theoretical population biology 65:299-315.
24	Brose, U., T. Jonsson, E. L. Berlow, et al. 2006. Consumer-resource body-size relationships
25	in natural food webs. Ecology 87:2411-2417.

1	Brown, J. H., and A. Kodric-Brown. 1977. Turnover Rates in Insular Biogeography: Effect
2	of immigration on extinction. Ecology 58:445-449.
3	Chase, J. M., A. A. Burgett, and E. G. Biro. 2010. Habitat isolation moderates the strength
4	of top-down control in experimental pond food webs. Ecology 91:637-643.
5	Comins, H. N., and D. W. E. Blatt. 1974. Prey-Predator Models in Spatially Heterogeneous
6	Environments. Journal of Theoretical Biology 48:75-83.
7	Crowley, P. H. 1981. Dispersal and the stability of predator-prey interactions. The American
8	Naturalist 118:673-701.
9	de Roos, A. M., E. McCauley, and W. G. Wilson. 1991. Mobility versus density-limited
10	predator-prey dynamics on different spatial scales. Proceedings of the royal society
11	of London series B-Biological Sciences 246:117-122.
12	French, D. R., and J. M. J. Travis. 2001. Density-dependent dispersal in host-parasitoid
13	assemblages. Oikos 95:125-135.
14	Gause, G. F. 1934, The Struggle for Existence. Baltimore, Williams & Wilkins Co.
15	Hairston, N. G. J., F. E. Smith, and L. B. Slobodkin. 1960. Community structure, population
16	control, and competition. American Naturalist XCIV: 421-425.
17	Hassell, M. P., H. N. Comins, and R. M. May. 1991. Spatial structure and chaos in insect
18	population dynamics. Nature 353: 255-258.
19	Hauzy, C., F. D. Hulot, A. Gins, and M. Loreau. 2007. Intra- and interspecific density-
20	dependent dispersal in an aquatic prey-predator system. Journal of Animal Ecology
21	76: 552-558.
22	Holyoak, M., and S. P. Lawler. 1996a. Persistence of an extinction-prone predator-prey
23	interaction through metapopulation dynamics. Ecology 77: 1867-1879.
24	—. 1996b. The role of dispersal in predator-prey metapopulation dynamics. Journal of
25	Animal Ecology 65: 640-652.

1	Hulot, F. D., G. Lacroix, F. O. Lescher-Moutoue, and M. Loreau. 2000. Functional diversity
2	governs ecosystem response to nutrient enrichment. Nature 405: 340-344.
3	Huffaker, C. B. 1958. Experimental studies on predation: dispersion factors and predator-
4	prey oscillations. Hilgardia 27: 343-383.
5	Jansen, V. A. A. 1994. The dynamics of two diffusively coupled, identical Lotka-Volterra
6	patches Theoretical Aspects of Metapopulation Dynamics. Leiden, The Netherlands,
7	Ph.D Thesis Leiden Univ.
8	—. 1995. Regulation of predator-prey systems through spatial interactions: a possible
9	solution to the paradox of enrichment. Oikos 74: 384-390.
10	—. 1999. Phase locking: another cause of synchronicity in predator-prey systems. Trends in
11	Ecology & Evolution 14: 278-279.
12	—. 2001. The Dynamics of Two Diffusively Coupled Predator-Prey Populations.
13	Theoretical Population Biology 59: 119-131.
14	Janssen, A., E. van Gool, R. Lingeman, J. Jacas, and G. van de Klashorst. 1997.
15	Metapopulation dynamics of a persisting predator-prey system in the laboratory:
16	Time series analysis. Experimental and Applied Acarology 21: 415-430.
17	Jenkins, D. G., C. R. Brescacin, C. V. Duxbury, et al. 2007. Does size matter for dispersal
18	distance? Global Ecology & Biogeography 16: 415-425.
19	Kuznetsov, Y. 2004. Elements of Applied Bifurcation Theory. Springer, Berlin.
20	Leibold, M. A., M. Holyoak, N. Mouquet, P. Amarasekare, J. M. Chase, M. F. Hoopes, R.
21	D. Holt et al. 2004. The metacommunity concept: a framework for multi-scale
22	community ecology. Ecology Letters 7: 601-613.
23	Leroux, S. J., and M. Loreau. 2008. Subsidy hypothesis and strength of trophic cascades
24	across ecosystems. Ecology Letters 11: 1147-1156.

I	Levins, R. 1970. Extinction. Some mathematical questions in biology (ed. M.
2	Gerstenhaber), Vol. 2, pp. 75-108. American Mathematical Society, Providence,
3	Rhode Island.
4	Li, Z. Z., M. Gao, C. Hui, X. Z. Han, and H. H. Shi. 2005. Impact of predator pursuit and
5	prey evasion on synchrony and spatial patterns in metapopulation. Ecological
6	Modelling 185: 245-254.
7	Loeuille, N., and M. Loreau. 2004. Nutrient enrichment and food chains: can evolution
8	buffer top-down control? Theoretical population biology 65: 285-298.
9	Loreau, M., A. Downing, M. Emmerson, A. Gonzalez, J. Hughes, P. Inchausti, J. Joshi et al
10	2002. A new look at the relationship between diversity and stability in M. Loreau,
11	Naeem, S. & Inchausti, P., ed. Biodiversity and ecosystem functioning: synthesis
12	and perspectives. Oxford, Oxford University Press.
13	Lotka, A. J. 1925, Elements of Physical Biology. Baltimore, Williams & Wilkins,
14	Baltimore, MD.
15	McCann, K., A. Hastings, S. Harrison, and W. Wilson. 2000. Population Outbreaks in a
16	Discrete World. Theoretical Population Biology 57: 97-108.
17	Murdoch, W. W. 1966. "Community Structure, Population Control, and Competition"-A
18	Critique. The American Naturalist 100: 219-226.
19	Nachman, G. 1987a. Systems Analysis of Acarine Predator-Prey Interactions. I. A
20	Stochastic Simulation Model of Spatial Processes. The Journal of Animal Ecology
21	56: 247-265.
22	—. 1987b. Systems Analysis of Acarine Predator-Prey Interactions. II. The Role of Spatial
23	Processes in System Stability. The Journal of Animal Ecology 56: 267-281.
24	—. 1991. An acarine predator-prey metapopulation system inhabiting greenhouse
25	cucumbers. Biological Journal of the Linnean Society 42: 285-303.

1	Oksanen, L., S. D. Fretwell, J. Arruda, and P. Niemela. 1981. Exploitation Ecosystems in
2	Gradients of Primary Productivity. The American Naturalist 118: 240-261.
3	Reeve, J. D. 1988. Environmental variability, migration, and persistence in host-parasitoid
4	systems. The American Naturalist 132: 810-836.
5	—. 1990. Stability, Variability, and Persistence in Host-Parasitoid Systems. Ecology 71:
6	422-426.
7	Rooney, N., K. S. McCann, and J. C. Moore. 2008. A landscape theory for food web
8	architecture. Ecology Letters 11: 867-881.
9	Rosenzweig, M. L., and R. H. MacArthur. 1963. Graphical representation and stability
10	conditions of predator-prey interactions. American Naturalist XCVII: 209-223.
11	Shurin, J. B., E. T. Borer, E. W. Seabloom, K. Anderson, C. A. Blanchette, B. Broitman, S
12	D. Cooper, and B. S. Halpern. 2002. A cross-ecosystem comparison of the strength
13	of trophic cascades. Ecology Letters 5: 785-791.
14	Taylor, A. D. 1990. Metapopulations, Dispersal, and Predator-Prey Dynamics: An
15	Overview. Ecology 71: 429-433.
16	van de Klashorst, G., J. Les Readshaw, M. W. Sabelis, and R. Lingeman. 1992. A
17	demonstration of asynchronous local cycles in an acarine predator-prey system.
18	Experimental and Applied Acarology 14: 185-199.
19	Vasseur DA, Fox JW. 2009. Phase-locking and environmental fluctuations generate
20	synchrony in a predator-prey community. Nature 460: 1007-1010
21	Volterra, V. 1926. Variations and fluctuations in the numbers of individuals in animal
22	species living together. Animal Ecology (ed. R.N. Chapman), pp. 409-448.
23	McGraw-Hill, New York.
24	Wolf, A., J. B. Swift, H. L. Swinney, and J. A. Vastano. 1985. Determining Lyapunov
25	Exponents from a Time-Series. Physica D 16: 285-317.

1	Ylikarjula, J., S. Alaja, J. Laakso, and D. Tesar. 2000. Effects of Patch Number and
2	Dispersal Patterns on Population Dynamics and Synchrony. Journal of Theoretical
3	Biology 207: 377-387.
4	Zeigler, B. P. 1977. Persistence and patchiness of predator-prey systems induced by discrete
5	event population exchange mechanisms. Journal of Theoretical Biology 67: 687-713
6	

#### 2 3 Figure 1. Per capita dispersal rate $D_{\alpha}$ of species $\alpha$ ( $\alpha = n$ or $\alpha = p$ ) in response to changes in the density of species $\beta$ ( $\beta = n$ or $\beta = p$ ). $S_{\alpha\beta}$ is the half-saturation parameters. Density-4 5 independent ( $x_{\alpha\beta}$ =0: solid line) dispersal; positive (black dashed line) and negative (grey 6 dashed line) density-dependent dispersal; cyrtoid ( $|x_{\alpha\beta}|=1$ : thin dashed lines) and sigmoid 7 $(|x_{\alpha\beta}|=4$ : thick dashed lines) shape of density dependence in dispersal. 8 9 Figure 2. Dynamics and spatial synchrony of the prey in predator-prey metacommunities as 10 functions of its maximal dispersal rate $(d_n)$ . Top diagrams illustrate the various cases studied: N and P represent prey and predator densities; dispersal of each species is 11 12 represented by a wide white arrow; positive or negative effects of densities on dispersal are 13 represented by thin solid arrows with a + or a - sign. Figures in the first column (A and D) 14 present results when dispersal is constant in both species. In the second column (B and E), 15 we present results for (intra- and interspecific) density-dependent dispersal in the prey. In 16 the third column (C and F), we present results for (intra- and interspecific) density-17 dependent dispersal in the predator. We present results for $d_n = 100 d_p (d_n > d_p)$ and for $d_p =$ 18 $100 d_n (d_p > d_n)$ in the first (A,B and C), and in the second (D, E and F) row, respectively. 19 Markers on curves (except for A and D) indicate what type of density dependence the curve 20 refers to: intraspecific (dots) or interspecific (triangles). The nature of dynamics obtained 21 using the Lyapunov exponent is represented on curves: chaotic dynamics (with colour 22 thickening) or periodic dynamics (without colour thickening). The specific colours used 23 (gray, cyan and red) as well as the position of the thickening (above or under the curve) only 24 help, in addition to curve markers, to distinguish between the different curves and do not 25 have any further meaning. Notice on figures D, E, F $(d_p > d_n)$ the two coexisting attractors 26 (synchronous and asynchronous) for some dispersal rate values. Parameters values are the 27 following. Constant dispersal: $x_{\alpha\beta}$ =0; positive density-dependent dispersal: $x_{\alpha\beta}$ =4 28 $(\alpha\beta=nn,np,pp)$ ; negative density-dependent dispersal: $x_{pn}=-4$ . Other parameters: $S_{nn}=18$ , $S_{np}=2$ , $S_{pp}=2$ , $S_{pn}=18$ ; local interactions: r=10, K=25, b=11, c=1, m=10. 29 30 31 Figure 3.Temporal mean of regional density (upper sub-graph) and coefficient of variation 32 (CV, lower sub-graph) of the prey in predator-prey metacommunities as functions of its 33 maximal dispersal rate $(d_n)$ . See legend of figure 2.

34

1

**LEGENDS OF FIGURES** 

maximal dispersal rate of the prey $(d_n)$ . See legend of figure 2.  Figure 5. The effect of intraspecific density-dependent dispersal in the prey when $x_{nn}=1$ . The spatial synchrony (A), the temporal mean of regional densities (upper sub graph) and the coefficient of variation (CV, lower sub graph) of the prey (B) and of the predator (C) as functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 d_p$ $(d_n > d_p)$ . The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more	1	Figure 4. Temporal mean of regional density (upper sub-graph) and coefficient of variation
Figure 5. The effect of intraspecific density-dependent dispersal in the prey when $x_{nn}=1$ . The spatial synchrony (A), the temporal mean of regional densities (upper sub graph) and the coefficient of variation (CV, lower sub graph) of the prey (B) and of the predator (C) as functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 \ d_p$ ( $d_n > d_p$ ). The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	2	(CV, lower sub-graph) of the predator in predator-prey metacommunities as functions of the
Figure 5. The effect of intraspecific density-dependent dispersal in the prey when $x_{nm}=1$ . The spatial synchrony (A), the temporal mean of regional densities (upper sub graph) and the coefficient of variation (CV, lower sub graph) of the prey (B) and of the predator (C) as functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 d_p$ ( $d_n > d_p$ ). The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nm} =1$ .	3	maximal dispersal rate of the prey $(d_n)$ . See legend of figure 2.
spatial synchrony (A), the temporal mean of regional densities (upper sub graph) and the coefficient of variation (CV, lower sub graph) of the prey (B) and of the predator (C) as functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 d_p$ ( $d_n > d_p$ ). The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	4	
coefficient of variation (CV, lower sub graph) of the prey (B) and of the predator (C) as functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 d_p$ ( $d_n > d_p$ ). The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	5	Figure 5. The effect of intraspecific density-dependent dispersal in the prey when $x_{nn}=1$ . The
functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 d_p$ $(d_n > d_p)$ . The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	6	spatial synchrony (A), the temporal mean of regional densities (upper sub graph) and the
$(d_n > d_p)$ . The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3 and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	7	coefficient of variation (CV, lower sub graph) of the prey (B) and of the predator (C) as
and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	8	functions of the maximal dispersal rate of the prey $(d_n)$ . We present results for $d_n = 100 d_p$
and that two attractors can now coexist at intermediate dispersal rates. Parameters values are the same as in figure 2, except $ x_{nn} =1$ .	9	$(d_n > d_p)$ . The curve coding is the same as in figure 2. Notice that, compared to figures 2, 3
the same as in figure 2, except $ x_{nn} =1$ .	10	and 4, intraspecific density-dependent dispersal does not induce chaotic dynamics any more
	11	and that two attractors can now coexist at intermediate dispersal rates. Parameters values are
Accepted manuscrift	12	the same as in figure 2, except $ x_{nn} =1$ .
		Accepted manual









