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Extinctions in a Metapopulation with Nonlinear Dispersal Coupling

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Abstract: Major threats to biodiversity are climate change, habitat fragmentation (in particular, habitat loss), pollution, invasive species, over-exploitation, and epidemics. Over the last decades habitat fragmentation has been given special attention. Many factors are causing biological systems to extinct; therefore, many issues remain poorly understood. In particular, we would like to know more about the effect of the strength of inter-site coupling (e.g., it can represent the speed with which species migrate) on species extinction or persistence in a fragmented habitat consisting of sites with randomly varying properties. To address this problem we use theoretical methods from mathematical analysis, functional analysis, and numerical methods to study a conceptual single-species spatially-discrete system. We state some simple necessary conditions for persistence, prove that this dynamical system is monotone and we prove convergence to a steady-state. For a multi-patch system, we show that the increase of inter-site coupling leads to the formation of clusters – groups of populations whose sizes tend to align as coupling increases. We also introduce a simple one-parameter sufficient condition for a metapopulation to persist.

Keywords: metapopulation collapse; Allee effect; inter-patch coupling; pattern formation

MSC: 92D25



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1. Introduction

In recent decades, significant attention has been directed towards the factors and processes that lead to the survival or extinction of natural populations [1]. This focus has been spurred by ongoing global environmental changes, including the impact of global warming on populations and communities [2]. One specific consequence of global warming is the alteration of species ranges and the fragmentation of habitats. Additionally, habitat fragmentation can occur due to human activities such as forest logging and the construction of new roads. Both habitat fragmentation and general habitat loss have a noticeably detrimental impact on corresponding populations, often leading to species extinctions [3,4]. In fact, habitat fragmentation is widely recognized as the most significant threat to biodiversity on a global scale [5,6].

That is why it is important to understand population dynamics in a complex or fragmented habitat and there is indeed a large number empirical and theoretical studies addressing this issue [7–16]. The most widely used models of population dynamics in a fragmented habitat are metapopulation models [4,17–22]. In this framework, a fragmented habitat is viewed as a collection of separate sites, with subpopulations of a species residing in these sites. The subpopulations can be connected either through dispersal between sites or by a shared external factor with spatial correlations, such as weather fluctuations [23,24]. Metapopulation models can either be spatially implicit, where the state of the metapopulation is described by a single global variable, for example the fraction of occupied sites [4,17–19], or spatially explicit, where each site is characterized by its own

'local' variables, such as the size of a specific subpopulation [10,16,25], the probability of a patch being inhabited [26] etc. The former case sometimes can show results similar to lattice models [27,28] and network models [29,30], particularly when the relative locations of sites are explicitly considered.

Numerous studies have investigated the persistence or extinction of metapopulations in relation to habitat geometry [26,31,32], as well as environmental and demographic stochasticity [33,34]. However, there is a noticeable scarcity of research that specifically explores the impact of coupling strength between different sites on persistence or extinction, even though it may be implicitly accounted for through habitat geometry, where coupling strength generally diminishes with greater inter-site distance. Nevertheless, understanding the impact of coupling strength is critical, especially in light of evidence suggesting that inter-site coupling might be altered due to climate change [35,36].

It should be noted that the possibility of extinction depends on the type of density-dependence observed in local population growth. In deterministic models, in a closed system (i.e., without outward migration), populations with logistic growth cannot go extinct because the extinction state is unstable [37–39]. However, natural populations rarely conform to logistic growth patterns. Instead, growth rates often exhibit the Allee effect [40–43], which can be caused by many factors that are often present in real-life situations [44]. The presence of a strong Allee effect significantly alters population dynamics [40,42,43,45–48]. Notably, the extinction state becomes stable, thus allowing for the possibility of extinction within a closed population.

For a two-site system studied previously [49], it was demonstrated that, subject to certain limitations, an increase in coupling strength can potentially trigger a population outbreak, where the system transitions from a low-density steady state to a high-density one. Mathematically, this transition corresponds to a saddle-node bifurcation, in which the low-density steady state vanishes as a consequence of increased coupling. Although with the model proposed below we focus on extinction rather than outbreaks, it will be shown that the extinction may follow a sufficiently large increase in the coupling strength due to essentially the same mechanism as in [49].

This paper complements the research done in [50] with linear coupling. Here we consider two types of coupling: linear and quadratic. This work differs from the work done in [50] in the sense that we use analytic methods from mathematical analysis, nonlinear functional analysis, monotone dynamical systems theory etc. The methods are used to prove some sufficient conditions for metapopulation persistence. We also show that the solutions are bounded and analytic and we study the asymptotic behavior for some initial conditions. In the end we present a one parameter criterion for a system to persist and estimate the parameter.

2. Materials and Methods

The existence of a non-zero steady-state point for the case with logistic growth will be proved analytically.

For the Allee effect we simulate both types of coupling using the RK45 method, which is programmed in Python using `scipy.integrate.solve_ivp`. This method with standard settings is perfect for the model with quadratic coupling; for linear coupling we will change the settings, see this section below. The Euler method is not very efficient here because of its slow convergence to the solution. Also the use of higher order methods can be motivated by analyticity of solutions. We do not consider Runge-Kutta methods of higher order because it is not necessary for our tasks. The RK45 method has global error on the order of $O(h^5)$ [51].

We let $u_i(0) = \max k_j$ and change q with a step size of 0.5 from 0.5 to 20. It was checked in simulations that $t = 200$ was sufficiently large to ensure the system's convergence to its steady-state distribution. For linear case we had to set the value of related tolerance to an error $rtol = 10^{-6}$ instead of default $rtol = 10^{-3}$ to ensure the convergence for large q .

In the section “Two-Patch System” we assume that the solution to the Cauchy problem exists and is unique and continuously differentiable for $t \geq 0$, it will be proved in the section “Multi-Patch System”, which is written more formally and states all necessary proofs. The section “Two-Patch System” helps become better acquainted with the model in a simpler case.

3. Two-Patch System

In this section we consider the systems with a linear and quadratic coupling. The linear coupling between two populations u and v is written as $q(u - v)$ for some coefficient q . The quadratic coupling between two populations u and v is written as $q(u^2 - v^2)$ for some coefficient q .

The quadratic coupling is also called density-dependent dispersal. It is due to the fact that $u^2 - v^2 = (u + v)(u - v)$. So the strength of the coupling depends on the total population $u + v$.

Here we begin with a quadratic coupling as a continuation of the paper [50] with linear model. Then we list some additional properties for a linear coupling which can be analogously proven.

The dynamics of the two-patch system with a quadratic coupling is described by the following equations:

$$\frac{du_1}{dt} = f_1(u_1) + q(u_2^2 - u_1^2), \quad \frac{du_2}{dt} = f_2(u_2) + q(u_1^2 - u_2^2), \tag{1}$$

where f_1, f_2 are polynomials of the same form such that $f_1(0) = f_2(0) = f_1(k_1) = f_2(k_2) = 0$ for some positive real numbers k_1, k_2 . Here we are considering polynomials of the forms $f_i(u_i) = \alpha_i u_i (1 - \frac{u_i}{k_i})$ (logistic growth) and $f_i(u_i) = \alpha_i u_i (u_i - \beta_i) (1 - \frac{u_i}{k_i})$ (logistic growth with an Allee effect) with positive coefficients, where $\beta_i < k_i$.

The properties of the system (1) are determined by its steady states; in particular, a long-term persistence of the two subpopulations is only possible if there exists a stable ‘coexistence’ steady state, i.e., a positive solution of the following system:

$$f_1(u_1) + q(u_2^2 - u_1^2) = 0, \quad f_2(u_2) + q(u_1^2 - u_2^2) = 0. \tag{2}$$

From (2) we readily get:

$$f_1(u_1) + f_2(u_2) = 0. \tag{3}$$

If $q \neq 0$, the system (2) can be rewritten as

$$u_2^2 = u_1^2 - \frac{1}{q} f_1(u_1), \quad u_1^2 = u_2^2 - \frac{1}{q} f_2(u_2) \tag{4}$$

When $q \rightarrow \infty$, we get $u_1 = u_2$.

Let $u_1(0) = u_{01}, u_2(0) = u_{02}$. By \hat{u}_1, \hat{u}_2 further we will denote the steady state values for these initial conditions if they exist (in a sense that $u_1(t) \rightarrow \hat{u}_1, u_2(t) \rightarrow \hat{u}_2$, when $t \rightarrow \infty$).

If there are steady state values, then $f_1(\hat{u}_1) + f_2(\hat{u}_2) = 0$. So the Equation (3) is a necessary condition for a point (\hat{u}_1, \hat{u}_2) to be a steady state point.

We also define $\bar{u} = \frac{u_1 + u_2}{2}, \bar{u}_0 = \bar{u}(0), \frac{d\bar{u}}{dt} = \frac{f_1(u_1) + f_2(u_2)}{2}, Z(f) = \{u | f(u) = 0\}$.

The case $q = 0$ is not very interesting because the system (2) is simplified to $f(u_1) = 0, f(u_2) = 0$, hence, a point (\hat{u}_1, \hat{u}_2) is a steady state point iff $(\hat{u}_1, \hat{u}_2) \in Z(f_1) \times Z(f_2) = \{(u_1, u_2) | f(u_1) = 0, f(u_2) = 0\}$.

There is another trivial case which is covered by Lemma 1.

Lemma 1. *Let $Z(f) = \{u | f(u) = 0\}$. Then for any $\hat{u} \in Z(f_1) \cap Z(f_2)$ we have a steady state point (\hat{u}, \hat{u}) .*

Proof. We fix any $\hat{u} \in Z(f_1) \cap Z(f_2)$. Let $u_{01} = \hat{u}$, $u_{02} = \hat{u}$. Then $\frac{du_1}{dt}(0) = \frac{du_2}{dt}(0) = 0$, hence, the point (\hat{u}, \hat{u}) is a steady state point. \square

Now we consider more specific models: logistic growth and logistic growth with an Allee effect. For the logistic growth the case $k_1 = k_2$ is trivial, for other two cases it is enough to consider an example 1 below because another one becomes the example 1, if we change indexes.

Example 1. $f_i(u_i) = \alpha_i u_i (1 - \frac{u_i}{k_i})$, $i = 1, 2$, $k_1 > k_2$; $u_1(0) = k_2$, $u_2(0) = k_2$, $q \neq 0$. Then there is a non-zero steady state point.

Proof. We know that $\frac{du_1}{dt}|_{u_1=u_2 \neq 0} > 0$ for $u_1 = u_2 < k_1$, hence, for all $t > 0$ we have $u_1(t) > u_2(t)$. We note that $\frac{d\bar{u}}{dt} > 0$ when $u_1 \in [0; k_1]$, $u_2 \in [0; k_2]$. This means that the population cannot extinct because $\bar{u}(t) > k_2$ for all $t > 0$, in other words, we got that for all $t > 0$ $u_1(t) > u_2(t) > k_2$.

Let us now consider the behavior as $t \rightarrow \infty$ (Figure 1). We know that $\frac{du_1}{dt}(0) > 0$, the function $\frac{du_1}{dt}$ is continuous, hence, it is positive in some neighborhood. It is also clear that for $i = 1$ there exists $t_i > 0$ (t_i may be infinity) such that $\frac{du_1}{dt}(t_i) = 0$, otherwise there would be some constant $C_i > 0$ such that $\frac{du_1}{dt} > C_i$ that would lead to $u_1 \rightarrow \infty$ – it is a contradiction. So we have $\frac{du_1}{dt}(t_i) = 0$, and two cases: $t_i = \infty$, and $t_i \neq \infty$. If we have more than one point t_i we number the set $T = \{t_i \in \mathbb{R} \cup \{\infty\} | \frac{du_1}{dt}(t_i) = 0\}$ in such way that for any integer i between 2 and $card(T)$ we will have $t_{i-1} < t_i$ (the set T is no more than countable because $\frac{du_2}{dt} > 0$).

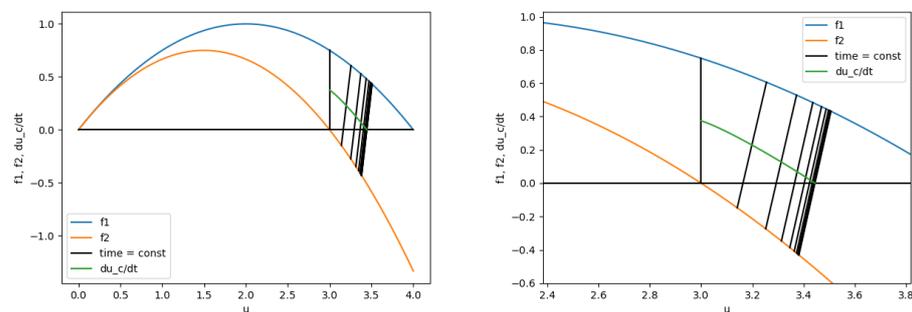


Figure 1. The functions u_1, u_2 are monotonically increasing, the function $\frac{d\bar{u}}{dt}$ is monotonically decreasing and has a limit 0, that leads to an existence of a non-zero steady-state point. The first coordinate of the ends of the lines is the value of u_1 and u_2 at the particular $time:time = const$. The second coordinate of the centres of the lines is $\frac{d\bar{u}}{dt}$.

Let us first assume that $t_i \neq \infty$. $\frac{du_2}{dt} > 0$ while $\frac{du_1}{dt} > 0$, nontrivial solution of an autonomous system cannot approach a fixed point in finite time, hence, we have

$$\frac{du_1}{dt}(t_i) = 0, \frac{du_2}{dt}(t_i) > 0,$$

hence, for some $\epsilon_i > 0$ for all $\epsilon \in (0; \epsilon_i]$ we have $\frac{du_1}{dt}(t_i + \epsilon) > 0$, $\frac{du_2}{dt}(t_i + \epsilon) > 0$. Repeating these actions again, if the function $\frac{du_1}{dt}$ has infinite number of $t_i \neq \infty$ such that $\frac{du_1}{dt}(t_i) = 0$ and $\frac{du_1}{dt} > 0$ in some deleted neighborhood, we get that for any $i \in \mathbb{N} \setminus \{1\}$ $t_{i-1} < t_i$ and

$$\frac{du_1}{dt}(t_i) = 0, \frac{du_2}{dt}(t_i) > 0.$$

It means that $\frac{d\bar{u}}{dt}(t) > 0$ for all $t \in [0; \infty)$. The derivative $\frac{d\bar{u}}{dt}(t)$ must approach zero, otherwise there would be a constant $C > 0$ such that $\frac{d\bar{u}}{dt}(t) > C$ for all $t \in [0; \infty]$, leading

to $u_1 \rightarrow \infty$ ($\infty \leftarrow \bar{u} < u_1$), which contradicts with $u_1 \leq k_1$. Therefore, $\frac{d\bar{u}}{dt}(\infty) = \frac{du_1}{dt}(\infty) = \frac{du_2}{dt}(\infty) = 0$ because all the derivatives were non-negative. It means that there is a non-zero steady-state point (\bar{u}_1, \bar{u}_2) . Moreover, we also proved that $\infty \in T$, and that $\frac{du_2}{dt} > 0$ while $\frac{du_1}{dt} \geq 0$.

We are left to prove that there is a non-zero steady-state point (\hat{u}_1, \hat{u}_2) in case of finite number of zeros of the derivative $\frac{du_1}{dt}$. If we let $t_0 = 0$ there is an index i such that $t_{i-1} < t_i = \infty$, $\varepsilon_i = \infty$, for all $\varepsilon \in (0; \infty)$ we have $\frac{du_1}{dt}(t_{i-1} + \varepsilon) > 0$, $\frac{du_2}{dt}(t_{i-1} + \varepsilon) > 0$. It means again that $\frac{d\bar{u}}{dt}(t) > 0$ for all $t \in [0; \infty)$, hence, $\frac{d\bar{u}}{dt}(\infty) = \frac{du_1}{dt}(\infty) = \frac{du_2}{dt}(\infty) = 0$, and there is a non-zero steady-state point (\hat{u}_1, \hat{u}_2) . \square

All stated above gives us a proof of a following theorem:

Theorem 1. *The system (1) with logistic growth functions has a non-zero steady state point.*

Remark. *Another proof of Theorem 1 is given in the Section 4.1 (Theorem 6).*

Example 2. $f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i})$, $i = 1, 2$, $k_1 > k_2, \beta_1 < k_2$; $u_1(0) = k_2, u_2(0) = k_2$. *There is a non-zero steady state point—the proof is identical to the proof in the Example 1.*

Further we will consider the system with $f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i})$, $i = 1, 2$, $k_1 > k_2, \beta_1 > k_2$.

Firstly, we note that if $q \leq \max_{u_2 \in (\beta_2; k_2)} \frac{f_2(u_2)}{u_2^2}$, then there is a steady state point $(\bar{u}_1; \bar{u}_2)$, $\bar{u}_1 < \bar{u}_2 \in (\beta_2; k_2]$. For $q = 0$ it is obvious. For $q > 0$, indeed, this means that there exists $u_{02} \in (\beta_2; k_2)$ (that will be the initial condition for u_2 , and 0 for u_1) such that $f_2(u_{02}) - qu_{02}^2 \geq 0$. $\frac{du_2}{dt} \geq \frac{du_2}{dt}|_{u_1=0} \geq 0$, $f_2(k_2) + q(u_1^2 - k_2^2) = q(u_1^2 - k_2^2)$, $\frac{du_1}{dt}|_{u_2=u_1 \leq k_2} < 0$, hence $u_1(t) < u_2(t) \leq k_2$ for all $t > 0$. The function u_2 as a monotone bounded continuously differentiable function has a limit $\hat{u}_2 \in [u_{02}; k_2]$ as $t \rightarrow \infty$. $\frac{du_1}{dt} > \frac{du_1}{dt}|_{u_2=const} > 0$ for all t , but $\frac{du_1}{dt}$ must approach 0, otherwise it contradicts with the condition $\frac{du_1}{dt}|_{u_2=u_1 \leq k_2} < 0$, hence, there is a limit $\hat{u}_1 \in (0, \hat{u}_2]$.

Now we consider a system with a linear coupling:

$$\frac{du_1}{dt} = f_1(u_1) + q(u_2 - u_1), \quad \frac{du_2}{dt} = f_2(u_2) + q(u_1 - u_2), \tag{5}$$

where f_1, f_2 are of the same types as in (1).

In this case Examples 1 and 2 will have the same proofs as in (1) because we used only monotone property of the functions and their transitional points, which are the same.

Further for the case $f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i})$, $i = 1, 2$, $k_1 > k_2, \beta_1 > k_2$ we in the same way get that if $q \leq \max_{u_2 \in (\beta_2; k_2)} \frac{f_2(u_2)}{u_2}$ then we have a non-zero steady state point.

Remark. *Local extrema of the functions f_i in the system with the Allee effect can be easily computed:*

$$f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i}) = -\frac{\alpha_i}{k_i} u_i^3 + \frac{\alpha_i(\beta_i + k_i)}{k_i} u_i^2 - \alpha_i \beta_i u_i$$

$$f'(u_i) = -\frac{3\alpha_i}{k_i} u_i^2 + \frac{2\alpha_i(\beta_i + k_i)}{k_i} u_i - \alpha_i \beta_i$$

$$u_{imax} = \frac{\beta_i + k_i + \sqrt{(\beta_i + k_i)^2 - 3\beta_i k_i}}{3}$$

$$u_{imin} = \frac{\beta_i + k_i - \sqrt{(\beta_i + k_i)^2 - 3\beta_i k_i}}{3}$$

That helps us to state the following theorem.

Theorem 2. Let $q \neq 0$. For both systems (1) and (5) there is a sufficient condition for the case with $f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i})$, $i = 1, 2$, $k_1 > k_2$, $\beta_1 > k_2$:

$$f_1(u_{1max}) + \min_{u_2 \in [\beta_1; u_{1max}]} f_2(u_2) \geq 0 \tag{6}$$

then we have a non-zero steady state (u_1^*, u_2^*) with $u_1^*, u_2^* \in [\beta_1; u_{1max}]$.

We note that the condition does not include the parameter q meaning that the system will have a non-zero steady-state point for all $q > 0$. Now let us prove this sufficient condition.

Proof. Let $Q_i = \max_{u_i \in [\beta_1; u_{1max}]} f_i(u_i)$, $R_i = \min_{u_i \in [\beta_1; u_{1max}]} f_i(u_i)$. Then we have $Q_1 \geq -R_2$.

We need to prove that there is always an intersection of curves l_1, l_2 on $[0; +\infty) \times [0; +\infty) \setminus \{(0, 0)\}$, where the curves are defined by the following implicit equations:

$$l_1 : f_1(u_{11}) + qd(u_{21}, u_{11}) = 0; \tag{7}$$

$$l_2 : f_2(u_{22}) + qd(u_{12}, u_{22}) = 0; \tag{8}$$

where $d(u_1, u_2) = u_1 - u_2$ for the case without the Allee effect and $d(u_1, u_2) = u_1^2 - u_2^2$ for the case with the Allee effect. This is equivalent to

$$\begin{aligned} qd(u_{11}, u_{21}) &= f_1(u_{11}); \\ qd(u_{12}, u_{22}) &= -f_2(u_{22}). \end{aligned}$$

Then on the set $[\beta_1; k_1] \times [\beta_1; k_1]$ we have

$$\begin{aligned} 0 &\leq \frac{R_1}{q} \leq d(u_{11}, u_{21}) \leq \frac{Q_1}{q}; \\ 0 &< \frac{-Q_2}{q} \leq d(u_{12}, u_{22}) \leq \frac{-R_2}{q}; \end{aligned}$$

moreover, d may take all values in between $\frac{R_1}{q}, \frac{Q_1}{q}$, and $\frac{-Q_2}{q}, \frac{-R_2}{q}$ respectively due to continuity of all functions. We have an inequality $Q_1 \geq -R_2$, hence, $0 \leq d(u_{12}, u_{22}) \leq \frac{Q_1}{q}$.

Therefore, letting $u_1 = u_{11} = u_{12}$ and getting $u_{21}(u_1)$ and $u_{22}(u_1)$ from (7) and (8) for $u_1 \in [\beta_1; k_1]$ (for the (7) we firstly get $u_1(u_{21})$ and then use the Cardano method (see [52], p. 135–140) to get the inverse of a cubic function), we finally get that we have a continuous function of one variable $g(u_1) = d(u_1, u_{21}(u_1)) - d(u_1, u_{22}(u_1))$ such that $g(\beta_1) = d(\beta_1, \beta_1) - d(\beta_1, u_{22}(\beta_1)) = -d(\beta_1, u_{22}(\beta_1)) < 0$ because $u_{22}(\beta_1) < \beta_1$, and for u_{1max} we have $g(u_{1max}) = d(u_{1max}, u_{21}(u_{1max})) - d(u_{1max}, u_{22}(u_{1max})) = \frac{1}{q}(f_1(u_{1max}) + f_2(u_{22}(u_{1max}))) \geq \frac{1}{q}(Q_1 + R_2) \geq 0$ because $u_{22}(u_{1max}) \in [k_2; u_{1max}]$, hence, $f_2(u_{22}(u_{1max})) > R_2$. Therefore, if $g(u_{1max}) > 0$ then, by the intermediate value theorem [53], there is a point $u_1^* \in (\beta_1; u_{1max})$ or if $g(u_{1max}) = 0$ then there is a point $u_1^* = u_{1max}$ such that $d(u_1^*, u_{21}(u_1^*)) = d(u_1^*, u_{22}(u_1^*))$ and $(u_1^*, u_{21}(u_1^*))$ is a point on a curve l_1 and $(u_1, u_{22}(u_1^*))$ is a point on a curve l_2 . But that means that $u_{21}(u_1^*) = u_{22}(u_1^*) =: u_2^*$. So there is a point $(u_1^*, u_2^*) \in [\beta_1; u_{1max}] \times [\beta_1; u_{1max}]$ that lies on both curves. \square

4. Multi-Patch System

4.1. Existence and Uniqueness, Steady-State Points

Let \mathbb{N}_+ be the set of positive integers, $N \in \mathbb{N}_+$, \mathbb{R}^n (for $n \in \mathbb{N}_+$, $n \leq N$) be a n -dimensional (topological) vector space over real numbers with a standard euclidean

topology (with a norm $\|\cdot\|$ defined for $x = (x_1, \dots, x_n) \in \mathbb{R}^n$ as $\|x\| = \sqrt{\sum_{i=1}^n x_i^2}$). We define $\mathbb{R}_+ = [0, \infty)$, $\overline{1, N} = \{1, \dots, N\}$. We can define a ball with a center at $x \in \mathbb{R}^n$ and with a radius $r \in \mathbb{R}_+$: $B(x, r) = \{\xi \in \mathbb{R}^n \mid \|\xi - x\| < r\}$.

Here we will consider a system of N equations:

$$\frac{du_i}{dt} = f_i(u_i) + \sum_{j=1}^N q_{ij}d(u_j, u_i) = F_i(u_1, \dots, u_N), \quad i \in \overline{1, N},$$

where all $f_i(u_i) = \alpha_i u_i(1 - \frac{u_i}{k_i})$, $\alpha_i > 0$, $k_i > 0$ or all $f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i})$, $\alpha_i > 0$, $k_i > \beta_i > 0$, $d(u_j, u_i) = u_j - u_i$ or $d(u_j, u_i) = u_j^2 - u_i^2$, for all i, j we have $q_{ij} \geq 0$.

Or in a shorter form:

$$\frac{du}{dt} = F(u). \tag{9}$$

Let $\bar{u} = \frac{1}{N}(u_1 + \dots + u_N)$.

Further we sometimes will use a notation $u(t, u_0)$ for the solution of a Cauchy problem

$$\frac{du}{dt} = F(u), \quad u(0) = u_0. \tag{10}$$

Now we prove the boundedness of solutions with initial conditions $u_0 \in [-a, \max k_j + a]^N$ for some $a \geq 0$ and get some important corollaries from that. We will need a parallelepiped $[-a, \max k_j + a]^N$ with $a \neq 0$ in the next section.

Lemma 2. *Let u be a solution for the system (9), let $a \in \mathbb{R}$ be a non-negative constant. If for some $t_0 \in \mathbb{R}$, $i_0 \in \overline{1, N}$ we have $u_{i_0}(t_0) = -a$ or $u_{i_0}(t_0) = \max k_j + a$ and $u_j(t_0) \in [-a, \max k_j + a]$ for all $j \neq i_0$ then for the case $u_{i_0}(t_0) = -a$ we have $\frac{du_{i_0}}{dt}(t_0) \geq 0$ and for the case $u_{i_0}(t_0) = \max k_j + a$ we have $\frac{du_{i_0}}{dt}(t_0) \leq 0$.*

Proof. (1) $u_{i_0}(t_0) = -a$, hence, $f(u_{i_0}(t_0)) \geq 0$ and for all $j \in \overline{1, N}$ $u_{i_0}(t_0) \leq u_j(t_0) \in [-a, \max k_j + a]$; therefore, $\frac{du_{i_0}}{dt}(t_0) = f(u_{i_0}(t_0)) + \sum_{j=1}^N q_{i_0j}d(u_j, u_{i_0}) \geq 0$.

2) $u_{i_0}(t_0) = \max k_j + a$, hence, $f(u_{i_0}(t_0)) \leq 0$, for all $j \in \overline{1, N}$ $u_{i_0}(t_0) \geq u_j(t_0) \in [-a, \max k_j + a]$ and $\frac{du_{i_0}}{dt}(t_0) = f(u_{i_0}(t_0)) + \sum_{j=1}^N q_{i_0j}d(u_j, u_{i_0}) \leq 0$. \square

Lemma 3. *Let $\frac{dx}{dt} = g(x)$ be an autonomous system, $g : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a continuously differentiable mapping, $n \in \mathbb{N}_+$. For any solution x of the system we define a set $O(x) = \{\xi \mid \text{there is } t_0 \in \mathbb{R} : x(t_0) = \xi\}$, which is called the orbit of a solution x . Then for any two solutions x_1, x_2 we either have $O(x_1) \cap O(x_2) = \emptyset$ or $O(x_1) = O(x_2)$.*

Proof. Let x_1, x_2 be solutions of the system and $O(x_1) \cap O(x_2) \neq \emptyset$. Then there are t_1, t_2 such that $x_1(t_1) = x_2(t_2)$. Let $x(t) = x_2(t + (t_2 - t_1))$. We have $\frac{dx}{dt}(t) = \frac{dx_2}{dt}(t + (t_2 - t_1)) = g(x_2(t + (t_2 - t_1))) = g(x(t))$, hence, it is a solution. Moreover, $x(t_1) = x_2(t_2) = x_1(t_1)$. Hence, $x_2(t + (t_2 - t_1)) = x(t) = x_1(t)$ by the Picard-Lindelöf theorem (see [54], p. 86) and $O(x_1) = O(x_2)$. \square

Theorem 3. *Let u be a solution for the system (9), and $u(0) = u_0 \in [-a; \max k_j + a]^N$ for some $a \geq 0$. Then we have $-a \leq u_i(t) \leq \max k_j + a$ for all $i \in \overline{1, N}$, for all $t \geq 0$.*

Proof. Let us assume that it is true not for all $t \geq 0$, $S = \{\tau \geq 0 \mid \text{for all } 0 \leq t \leq \tau - a \leq u_i(t) \leq \max k_j + a \text{ for all } i \in \overline{1, N}\}$, $t_0 = \max S$. By the definition of t_0 we have the

inequality $-a \leq u_i(t_0) \leq \max k_j + a$ for all i , but for every neighborhood of the time t_0 we have some points at which the inequality is not true.

All functions $u_i, i \in \overline{1, N}$ are differential (in particular, continuous) because u is a solution, hence (see [55], p. 61]):

$$u_i(t_0 + h) - u_i(t_0) = \left(\frac{du_i}{dt}(t_0) + \phi_i(h)\right)h, \lim_{h \rightarrow 0} \phi_i(h) = 0.$$

For i such that $\frac{du_i}{dt}(t_0) \neq 0$ we can choose $h_i > 0$ such that for all $h : 0 < h < h_i$ the sign of a number $\frac{du_i}{dt}(t_0) + \phi_i(h)$ is the same as the sign of a number $\frac{du_i}{dt}(t_0)$.

For i such that $u_i(t_0) \in (-a; \max k_j + a)$ we can choose $h_i > 0$ such that for all $h : 0 < h < h_i$ we have $u_i(t_0 + h) \in (-a; \max k_j + a)$ due to continuity of the function u_i .

Let $H = \min h_j$. For i such that $u_i(t_0) \in (-a; \max k_j + a)$ for all $h : 0 < h < H$ we have $u_i(t_0 + h) \in (-a; \max k_j + a)$. By the lemma 2 for i such that $u_i(t_0) = -a$ we have $\frac{du_i}{dt}(t_0) \geq 0$ and for i such that $u_i(t_0) = \max k_j + a$ we have $\frac{du_i}{dt}(t_0) \leq 0$. Hence, if $\frac{du_i}{dt}(t_0) > 0$ then $u_i(t_0 + h) - u_i(t_0) > 0$, if $\frac{du_i}{dt}(t_0) < 0$ then $u_i(t_0 + h) - u_i(t_0) < 0$; therefore, for all $i \in \overline{1, N}$, for all $h : 0 < h < H$ we have $-a \leq u_i(t_0 + h) \leq \max k_j + a$ – it is a contradiction.

Now we consider a case when $\frac{du_i}{dt}(t_0) = 0$. From the system (9) we have $\frac{du_i}{dt}(t_0) = f_i(u_i(t_0)) + \sum_{j=1}^N q_{ij}d(u_j(t_0), u_i(t_0))$. If $u_i(t_0) = -a$ then we have $d(u_j(t_0), u_i(t_0)) \geq 0$, hence, $f_i(u_i(t_0)) = 0$. If $u_i(t_0) = \max k_j + a$ then we have $d(u_j(t_0), u_i(t_0)) \leq 0$, hence, $f_i(u_i(t_0)) = 0$. This is only possible when $a = 0$.

So now we have $\frac{du_i}{dt}(t_0) = \sum_{j=1}^N q_{ij}d(u_j(t_0), u_i(t_0)) = 0$. Since for all j we have $q_{ij} \geq 0$ and $d(u_j(t_0), u_i(t_0))$ are all of the same sign, we have that for all j it is true that $q_{ij}d(u_j(t_0), u_i(t_0)) = 0$, so either $q_{ij} = 0$ or $u_j(t_0) = u_i(t_0)$, and again by the Lemma 2 for all j such that $q_{ij} \neq 0$ we get that the sign of a number $\frac{du_j}{dt}(t_0)$ is the same as the sign of a number $\frac{du_i}{dt}(t_0)$.

$$\frac{d^2u_i}{dt^2}(t_0) = \sum_{j \neq i} q_{ij} \frac{du_j}{dt}(t_0) \text{ for linear coupling,}$$

$$\frac{d^2u_i}{dt^2}(t_0) = 2 \sum_{j \neq i} q_{ij} u_j(t_0) \frac{du_j}{dt}(t_0) \text{ for quadratic coupling.}$$

Let $J_i = \{j | q_{ij} \neq 0\}$.

(1) $u_i(t_0) = 0$. Then if there is $j_0 \in J_i$ such that $\frac{du_{j_0}}{dt}(t_0) > 0$ then $\frac{d^2u_i}{dt^2}(t_0) > 0$, hence, the function u_i has a strict local maximum at a point t_0 [53] – it is a contradiction. In the other case $\frac{d^2u_i}{dt^2}(t_0) = 0$. So for all $j \in J_i$ we have $\frac{d^2u_j}{dt^2}(t_0) = 0$ (in particular, $f_j(u_j)(t_0) = 0$) and $\frac{d^2u_i}{dt^2}(t_0) = 0$. We consider an autonomous system

$$\frac{du_i}{dt} = f_i(u_i) + \sum_{j \in J_i} q_{ij}d(u_j, u_i),$$

$$\frac{du_j}{dt} = f_j(u_j) + \sum_{l \in J_j} q_{lj}d(u_l, u_i), j \in J_i.$$

By the Lemma 3 any two of its orbits are either disjoint or coinciding. The system has a steady-state solution $\hat{u}(t) = (0, \dots, 0)$, but we also have $u_i(t_0) = 0, u_j(t_0) = 0$ for all $j \in J_i$. Therefore, for all $t \in \mathbb{R}$ we have $u_i(t) = 0, u_j(t) = 0$ for all $j \in J_i$ – it is a contradiction.

(2) $u_i(t_0) = \max k_j$. Then if there is $j_0 \in J_i$ such that $\frac{du_{j_0}}{dt}(t_0) < 0$ then $\frac{d^2u_i}{dt^2}(t_0) < 0$, hence, the function u_i has a strict local minimum at a point t_0 [53]—it is a contradiction.

In the other case $\frac{d^2 u_i}{dt^2}(t_0) = 0$. So for all $j \in J_i$ we have $\frac{d^2 u_j}{dt^2}(t_0) = 0$ (in particular, $f_j(u_j)(t_0) = 0$) and $\frac{d^2 u_i}{dt^2}(t_0) = 0$. We consider an autonomous system

$$\begin{aligned} \frac{du_i}{dt} &= f_i(u_i) + \sum_{j \in J_i} q_{ij} d(u_j, u_i), \\ \frac{du_j}{dt} &= f_j(u_j) + \sum_{l \in J_j} q_{lj} d(u_l, u_i), \quad j \in J_i. \end{aligned}$$

By the Lemma 3 any two of its orbits are either disjoint or coinciding. The system has a steady-state solution $\hat{u}(t) = (\max k_j, \dots, \max k_j)$, but we also have $u_i(t_0) = \max k_j$, $u_j(t_0) = \max k_j$ for all $j \in J_i$. Therefore, for all $t \in \mathbb{R}$ we have $u_i(t) = \max k_j$, $u_j(t) = \max k_j$ for all $j \in J_i$ – it is a contradiction. \square

Corollary. For $u(0) = u_0 \in [-a; \max k_j + a]^N$ we have $-a \leq \bar{u} \leq \frac{k_1 + \dots + k_N}{N} \leq \max k_j + a$.

Corollary (Picard–Lindelöf theorem). There is $b < \infty$ such that the solution $u(\cdot, u_0) : [b, \infty) \rightarrow [-a; \max k_j + a]^N$ to the Cauchy problem (10) with an initial condition $u(0) = u_0 \in [-a; \max k_j + a]^N$ for some $a \geq 0$ exists and is unique. Moreover, the dynamical system $u(t, \xi) : [b, \infty) \times [-a; \max k_j + a]^N \rightarrow [-a; \max k_j + a]^N$ is continuous.

Proof. The function of the right part of the system and its derivative are bounded on a compact set $[-a; \max k_j + a]^N$ (see [55], p. 33). Then, informally speaking, we can get local solutions (Picard-Lindelöf theorem) [54] with the same parameters and then cover the set $[b, \infty) \times [-a; \max k_j + a]^N$ with the parallelepipeds (of the same “sizes”) from the theorem.

Let us write out a more detailed proof. We write the family of solutions $u(t, \xi)$ as a sum $u(t, \xi) = \xi + v(t, \xi)$. Then we have an equivalent Cauchy problem

$$\frac{\partial v}{\partial t}(t, \xi) = F(\xi + v(t, \xi)) = G(t, \xi, v), \quad v(0, \xi) = 0. \tag{11}$$

Indeed, let $u(t, \xi)$ be a solution to the Cauchy problem (10). Then $0 = u(0, \xi) - \xi = v(0, \xi)$ and $\frac{\partial v}{\partial t}(t, \xi) = \frac{\partial(u - \xi)}{\partial t}(t, \xi) = \frac{\partial u}{\partial t}(t, \xi) = F(u(t, \xi)) = F(\xi + v(t, \xi))$.

Now let $v(t, \xi)$ be a solution to the Cauchy problem (11). Then $u(0, \xi) = \xi + v(0, \xi) = \xi$ and $\frac{\partial u}{\partial t}(t, \xi) = \frac{\partial(\xi + v)}{\partial t}(t, \xi) = \frac{\partial v}{\partial t}(t, \xi) = F(\xi + v(t, \xi)) = F(u(t, \xi))$.

Now we consider an equivalent integral equation

$$v(t, \xi) = \int_0^t G(\tau, \xi, v(\tau, \xi)) d\tau. \tag{12}$$

Indeed, let $v(t, \xi)$ be a solution to the problem (11). Then $v(t, \xi) = v(t, \xi) - v(0, \xi) = \int_0^t \frac{\partial v}{\partial t}(\tau, \xi) d\tau = \int_0^t G(\tau, \xi, v(\tau, \xi)) d\tau$ (by the fundamental theorem of calculus, see [53]).

Now let $v(t, \xi)$ be a solution to the integral Equation (12). Then $v(0, \xi) = 0$ and $\frac{\partial v}{\partial t}(t, \xi) = \frac{\partial}{\partial t} \int_0^t G(\tau, \xi, v(\tau, \xi)) d\tau = G(t, \xi, v(t, \xi))$ [53]. In particular, the solution $v(\cdot, \xi)$ is differentiable.

Further we prove that the solution to the integral Equation (12) exists and is unique.

On a compact set $\Omega = [-d, d] \times [-a - c, \max k_j + a + c]^{2N} \supset [-d, d] \times [-a, \max k_j + a]^{2N}$ for some $c, d \in (0, \infty)$ the function G is bounded by some constant K . $|G(t, \xi, v_1) - G(t, \xi, v_2)| = |F(\xi + v_1(t, \xi)) - F(\xi + v_2(t, \xi))| \leq (\sup_{v \in [-a - c, \max k_j + a + c]^N} \|F'(v)\|) * \|v_1 - v_2\|$. Let $M = \sup_{v \in [-a - c, \max k_j + a + c]^N} \|F'(v)\|$.

We fix $\zeta \in [-a, \max k_j + a]^N$.

We choose $d > 0$ such that

- (1) $(t, \zeta', v) \in \Omega$, if $|t| \leq d, \|\zeta' - \zeta\| \leq d$ and $\|v\| \leq Kd$;
- (2) $Md < 1$.

Let C^* be a space of continuous functions defined on a “rectangle” $R = \{(t, \zeta') \mid |t| \leq d, \|\zeta' - \zeta\| \leq d\}$ such that $\rho(v, 0) \leq Kd$ where ρ is a metric on this space defined as $\rho(v^{(1)}, v^{(2)}) = \max_{(t, \zeta')} |v^{(1)}(t, \zeta') - v^{(2)}(t, \zeta')|$ (the maximum of the continuous function $v^{(1)} - v^{(2)}$ is correctly defined because the set R is compact). The space C^* is a complete metric space as a closed subset of a complete metric space of all continuous functions on R .

We consider another integral equation

$$\psi(t, \zeta') = \int_0^t G(\tau, \zeta', \phi(\tau, \zeta')) d\tau =: (A\phi)(t, \zeta'), \quad (t, \zeta') \in R, \quad \phi \in C^*,$$

which defines an operator A such that $\psi = A\phi$. Now we prove that $A : C^* \rightarrow C^*$ is a contraction mapping ([54], p. 82) from the complete metric space C^* to itself and use the contraction mapping theorem [54] to show that there is a unique fixed point $u \in C^*$ such that $u = Au$.

For $\phi \in C^*$ and $(t, \zeta') \in R$ we have

$$|\psi(t, \zeta')| = \left| \int_0^t G(t, \zeta', \phi(\tau, \zeta')) d\tau \right| \leq Kd.$$

Hence, $\rho(\psi, 0) \leq Kd$ and $\psi \in C^*$. That means that $A(C^*) \subset C^*$. Moreover, for $\phi_1, \phi_2 \in C^*$ and ψ_1, ψ_2 such that $\psi_1 = A\phi_1, \psi_2 = A\phi_2$ we have $\rho(\psi_1, \psi_2) \leq \int_0^d \max_{(t, \zeta')} |G(t, \zeta', \phi_1(\tau, \zeta')) - G(t, \zeta', \phi_2(\tau, \zeta'))| d\tau \leq Md\rho(\phi_1, \phi_2)$. Since $Md < 1$, the operator A is a contraction mapping.

So we have a contraction mapping of a complete metric space to itself. Then by the contraction mapping theorem there exists a unique solution $v \in C^*$ to the equation $v = Av$. So, due to the arbitrariness of $\zeta \in [-a, \max k_j + a]^N$, for all $\zeta \in [-a, \max k_j + a]^N$ there exists a unique solution $v(t, \zeta)$ for $|t| \leq d$ which is continuously differentiable in t and continuous in ζ .

Now we consider the following sequences of solutions to the problem (11): $\{v^{(m)}(t, \zeta^{(m)})\}_{m=0}^\infty$, where $\zeta^{(m)} = \zeta^{(m-1)} + v^{(m-1)}(d, \zeta^{(m-1)})$ for $m \in \mathbb{N}_+$, $\zeta^{(0)} = \zeta \in [-a, \max k_j + a]^N$. We note that $\zeta^{(m)} = \zeta + \sum_{0 \leq i < m} v^{(i)}(d, \zeta^{(i)})$.

We define a mapping $v(t, \zeta) = \zeta^{(m)} - \zeta + v^{(m)}(t - md, \zeta^{(m)})$ for $t \in [md, (m + 1)d)$ for some m . It is a continuous mapping in t by the definition of a sequence.

For $t \in [m_0d, (m_0 + 1)d)$ for some m_0 we have

$$\forall m \in \mathbb{N}_+ : m \leq m_0 \forall \varepsilon_m > 0 \exists \varepsilon_0 > 0 : \zeta' : \|\zeta' - \zeta\| < \varepsilon_0 \Rightarrow \|\zeta'^{(m)} - \zeta^{(m)}\| < \varepsilon_m$$

by continuity of all mappings. Hence, $v(t, \zeta)$ is a continuous mapping in ζ .

$v(t, \zeta)$ is a unique solution to the problem (11) in $(m_0d, (m_0 + 1)d)$ by the definition of a sequence. On the boundary we have:

$$\begin{aligned} \frac{\partial v}{\partial t}(md, \zeta) &= \frac{\partial v^{(m)}}{\partial t}(0, \zeta^{(m)}) = G(0, \zeta^{(m)}, 0) = F(\zeta^{(m)}) \\ &= F(\zeta + \zeta^{(m)} - \zeta) = F(\zeta + v(md, \zeta)) = G(md, \zeta, v(md, \zeta)). \end{aligned}$$

Moreover, the derivative $\frac{\partial v}{\partial t}$ is continuous with respect to t on the boundary:

$$\begin{aligned} \lim_{t \rightarrow md-0} \frac{\partial v}{\partial t}(t, \xi) &= \lim_{t \rightarrow md-0} \frac{\partial v^{(m-1)}}{\partial t}(t - (m - 1)d, \xi^{(m-1)}) \\ &= \lim_{t \rightarrow md-0} F(\xi^{(m-1)} + v^{(m-1)}(t - (m - 1)d, \xi^{(m-1)})) = F(\xi^{(m)}). \quad \square \end{aligned}$$

Remark (to the Cauchy problem (10)). *Let the conditions of the previous corollary be true. Then $u(t_2, u(t_1, u_0)) = u(t_2 + t_1, u_0)$ for all $t_1, t_2 \in [\frac{b}{2}, \infty)$ ($b < 0$, see the previous corollary).*

Proof. Let $u^{(1)}(t) = u(t, u(t_1, u_0))$, $u^{(2)}(t) = u(t + t_1, u_0)$. Then $u^{(1)}(0) = u(0, u(t_1, u_0)) = u(t_1, u_0)$ and $u^{(2)}(0) = u(t_1, u_0)$. But by the assumption the solution to the Cauchy problem (10) is unique, hence, $u^{(1)}(t) = u^{(2)}(t)$ for all $t \in [\frac{b}{2}, \infty)$. \square

Theorem 4. *There is $b < 0$ such that the solution $u \in C^\infty((b, \infty); [-a; \max k_j + a]^N)$ to the Cauchy problem (10) with an initial condition $u(0) = u_0 \in [-a; \max k_j + a]^N$ for some $a \geq 0$ exists and is unique and analytic for all $t \in (b, \infty)$ (its Taylor series at every point of the interval (b, ∞) converge uniformly to the mapping u in some neighborhood of that point; see [53], p. 219).*

Proof. All the functions F_i are analytic (their Taylor series converge because the functions F_i are polynomials), hence, F is an analytic vector field. Then by the Cauchy-Kovalevskaya theorem [56] we have a solution for any initial condition $v_0 \in [-a; \max k_j + a]^N$ which is analytic on some open interval $J(v_0)$, containing zero.

Let $J(u_0)$ be the maximal interval of convergence of the Taylor series of the solution $u(t, u_0)$, and let us assume that $S_J = \sup J(u_0) < \infty$. From the previous remark we have that $u(t, u(S_J, u_0)) = u(t + S_J, u_0)$ and from the previous part of this proof we have that the solution $u(t, u(S_J, u_0))$ is analytic on some open interval $J(u(S_J, u_0))$, containing zero. But that means that $\frac{d^n u(0, u(S_J, u_0))}{dt^n} = \frac{d^n u(S_J, u_0)}{dt^n}$ and there is a neighborhood $U \subset J(u(S_J, u_0))$ of zero such that for all $t \in U$

$$u(t + S_J, u_0) = \sum_{n=0}^{\infty} \frac{d^n u(S_J, u_0)}{n!} ((t + S_J) - S_J)^n = \sum_{n=0}^{\infty} \frac{d^n u(S_J, u_0)}{n!} t^n.$$

—it is a contradiction. Hence, the solution is analytic for all $t > 0$. In particular, the solution is smooth. \square

Theorem 5. *If $q_{ij} = q$ for all i, j and $q \leq \max_i \max_{u_i \in (\beta_i; k_i)} \frac{f_i(u_i)}{u_i^2}$ for quadratic coupling or $q \leq \max_i \max_{u_i \in (\beta_i; k_i)} \frac{f_i(u_i)}{u_i}$ for linear coupling, then there is a non-zero steady-state point.*

Proof. The proof is the same as in the case of the two-patch system. \square

Theorem 6. *The system with logistic growth always has a non-zero steady-state point.*

To prove the theorem we have to prove a lemma about approximation of a steady-state point by periodic points.

Lemma 4. *For a dynamical system $u(t, \xi)$ induced by the problem (10) let M be a compact set such that for all $\xi \in M$ for all $t \geq 0$ we have $u(t, \xi) \in M$, let $\{\xi_n\}_{n=1}^{\infty} \in M$ be a sequence of periodic points where each point ξ_n has a period $T_n > 0$ and there are limits $\lim_{t \rightarrow \infty} \xi_n = \xi_0$, $\lim_{t \rightarrow \infty} T_n = 0$. Then the point ξ_0 is a steady-state (fixed) point of the dynamical system $u(t, \xi)$.*

Proof. We prove the lemma by contradiction: we assume that the point ξ_0 is not a steady-state point, meaning that there is $t_0 > 0$ such that $u(t_0, \xi_0) \neq \xi_0$. Let $\gamma = \|u(t_0, \xi_0) - \xi_0\|$.

Then the balls $B(\xi_0, \frac{\gamma}{4})$ and $B(u(\tau, \xi_0), \frac{\gamma}{4})$ do not intersect. Let us choose T such that $0 < T < t_0$ and $\|u(t, \xi_0) - \xi_0\| < \frac{\gamma}{8}$ for $0 \leq t \leq T$. By continuity of $u(t, \xi)$ there is $\delta > 0$ such that choosing any ψ such that $\|\psi - \xi_0\| < \delta$ implies that $\|u(T, \psi) - u(T, \xi_0)\| < \frac{\gamma}{8}$ for $0 < t < T$. In particular, we notice that if $\|\psi - \xi_0\| < \delta$ then $\|u(t, \psi) - \xi_0\| = \|u(t, \psi) - u(t, \xi_0) + u(t, \xi_0) - \xi_0\| \leq \|u(t, \psi) - u(t, \xi_0)\| + \|u(t, \xi_0) - \xi_0\| \leq \frac{\gamma}{8} + \frac{\gamma}{8} = \frac{\gamma}{4}$ for all t such that $0 \leq t \leq T$.

There is $N_0 \in \mathbb{N}_+$ such that for all $n > N_0$ we have $T_n < T$ and $\|\xi_n - \xi_0\| < \delta$. Hence, $\|u(t, \xi_n) - \xi_0\| < \frac{\gamma}{4}$ for $0 \leq t \leq T_n < T$. And as the orbit $O(\xi_n)$ is periodic of period T_n , we have $\|u(t, \xi_n) - \xi_0\| < \frac{\gamma}{4}$ for all $t \in \mathbb{R}$. But this contradicts with the fact that $\|u(t, \xi_n) - u(t, \xi_0)\| < \frac{\gamma}{4}$ because the last two statements mean that $u(t, \xi_n) \in B(\xi_0, \frac{\gamma}{4})$ and $u(t, \xi_n) \in B(u(\tau, \xi_0), \frac{\gamma}{4})$ and from the assumption we know that $B(\xi_0, \frac{\gamma}{4}) \cap B(u(\tau, \xi_0), \frac{\gamma}{4}) = \emptyset$. \square

Proof of Theorem 6. Firstly we note that if for some t_0 we have $0 < \bar{u}(t_0) < \frac{\min k_j}{N}$ then $\frac{d\bar{u}}{dt}(t_0) > 0$ because $u_i(t_0) \in [0; k_i]$, $i \in \overline{1, N}$ and at least one of the populations is greater than zero at the time t_0 . But that means that the metapopulation cannot extinct.

Let us consider a family of mappings $\Pi_t u_0 = u(t, u_0)$ for any $u_0 \in [\frac{\min k_j}{2N}, \max k_j]^N$, where $\Pi_t : [\frac{\min k_j}{2N}, \max k_j]^N \rightarrow [\frac{\min k_j}{2N}, \max k_j]^N$.

To apply the Brouwer fixed-point theorem [57–59] we need the set $[\frac{\min k_j}{2N}, \max k_j]^N$ to be compact and convex, which is obviously true, and the mapping $\Pi_t : [\frac{\min k_j}{2N}, \max k_j]^N \rightarrow [\frac{\min k_j}{2N}, \max k_j]^N$ to be continuous. The statement “all mappings Π_t are continuous” means that

$$\forall t > 0 \forall v_0 \in [\frac{\min k_j}{2N}, \max k_j]^N \forall \varepsilon > 0 \exists \delta > 0 : \forall v \|v - v_0\| < \delta \Rightarrow \|u(t, v) - u(t, v_0)\| < \varepsilon$$

or equivalently that means the continuous dependence on initial conditions. But that is true due to the Picard-Lindelöf theorem, hence, all mappings Π_t are continuous.

Let $\{T_n\}_{n=1}^\infty \subset \mathbb{R}_+$ be a monotone sequence such that there is a limit $\lim_{n \rightarrow \infty} T_n = 0$. And by the Brouwer fixed-point theorem for every $n \in \mathbb{N}_+$ there is a fixed point $\xi_n \in [\frac{\min k_j}{2N}, \max k_j]^N$ of the mapping Π_{T_n} . So we have $\Pi_{T_n}(\xi_n) = u(T_n, \xi_n) = \xi_n$. The sequence $\{\xi_n\}_{n=1}^\infty \subset [\frac{\min k_j}{2N}, \max k_j]^N$ is bounded, hence, there is a subsequence $\{\xi_{n_m}\}_{m=1}^\infty \subset [\frac{\min k_j}{2N}, \max k_j]^N$ such that there is a limit $\lim_{m \rightarrow \infty} \xi_{n_m} = \xi_0 \in [\frac{\min k_j}{2N}, \max k_j]^N$. Then by the Lemma 4 we conclude that the point ξ_0 is a steady-state point. \square

4.2. Solutions as a Monotone Dynamical System

From the previous section we know that the dynamical system $u(t, u_0)$, defined by the Cauchy problem (10), is bounded in \mathbb{R}_+^N for $u_0 \in [-a, \max k_j + a]^N$ for some $a \geq 0$ in the sense that for all $t \geq 0$ each component of a vector $u(t, u_0)$ is bounded by $-a$ and $\max k_j + a$ in \mathbb{R}_+ . The dynamical system $u(t, u_0)$ is continuous. It is analytical in the first variable t .

In this section we prove that the dynamical system $u(t, u_0)$ is strongly-monotone; moreover, we prove that it is asymptotically stable (as $t \rightarrow \infty$) for some initial conditions, that are important for us, for example, in computer simulations. Here the asymptotical stability means the convergence to some steady-state point.

On a topological vector space \mathbb{R}^N from the previous section we define non-strict partial orders \leq and $<$ and a strict partial order \ll by the following rules:

$$\begin{aligned} x, y \in \mathbb{R}^N, x \leq y \text{ iff for all } i \in \overline{1, N} \ x_i \leq y_i; \\ x, y \in \mathbb{R}^N, x < y \text{ iff for all } i \in \overline{1, N} \ x_i \leq y_i \text{ and } x \neq y; \\ x, y \in \mathbb{R}^N, x \ll y \text{ iff for all } i \in \overline{1, N} \ x_i < y_i. \end{aligned}$$

Remark. Let $x, y \in \mathbb{R}^N$. If $x \ll y$ then there are neighborhoods U and V of x and y respectively, such that for all $u \in U, v \in V$ we have $u \leq v$ (We will denote it as $U \leq V$).

Proof. By definition, $x \ll y$ means for all $i \in \overline{1, N}$ we have $x_i < y_i$. Then for all $i \in \overline{1, N}$ for all $u_i \in (x_i - \frac{y_i - x_i}{2}, x_i + \frac{y_i - x_i}{2})$ and $v_i \in (y_i - \frac{y_i - x_i}{2}, y_i + \frac{y_i - x_i}{2})$ we have $u_i \leq v_i$, hence, $u = (u_1, \dots, u_N) \leq v = (v_1, \dots, v_N)$.

So we can choose $U = (x_1 - \frac{y_1 - x_1}{2}, x_1 + \frac{y_1 - x_1}{2}) \times \dots \times (x_N - \frac{y_N - x_N}{2}, x_N + \frac{y_N - x_N}{2})$, $V = (y_1 - \frac{y_1 - x_1}{2}, y_1 + \frac{y_1 - x_1}{2}) \times \dots \times (y_N - \frac{y_N - x_N}{2}, y_N + \frac{y_N - x_N}{2})$. \square

Theorem 7. Let $a \geq 0$. Let $u^{(m)}$ be a solution for an initial value problem $u^{(m)}(0) = u_0^{(m)} \in [-a; \max k_j + a]^N$, $m = 1, 2$. If we have $u^{(1)}(0) \ll u^{(2)}(0)$ then for all $t \geq 0$ we have $u^{(1)}(t) \ll u^{(2)}(t)$.

Proof. Due to continuity of the solutions the inequality $u^{(1)}(t) \ll u^{(2)}(t)$ is true for t in some neighborhood of 0. Let us prove the rest of the statement by contradiction: we suppose that there is $t_0 > 0$ and there are indexes i_1, \dots, i_{r_0} ($r_0 \in \mathbb{N}_+$) such that we have $u_{i_r}^{(1)}(t_0) = u_{i_r}^{(2)}(t_0)$ for all $r \in \overline{1, r_0}$ and t_0 is such that for all $t < t_0$ we have $u^{(1)}(t) \ll u^{(2)}(t)$.

We fix $i_0 \in \{i_j | j \in \overline{1, r_1}\}$. From the system (9) we have

$$\frac{du_{i_0}^{(m)}}{dt} = f_{i_0}(u_{i_0}^{(m)}) + \sum_{j=1}^N q_{i_0j} d(u_j^{(m)}, u_{i_0}^{(m)}), \quad m = 1, 2.$$

For the following Cauchy problems

$$\frac{du_{i_0}^{(m)}}{dt} = f_{i_0}(u_{i_0}^{(m)}), \quad u_{i_0}^{(1)}(0) \ll u_{i_0}^{(2)}(0), \quad m = 1, 2, \tag{13}$$

we would have $u_{i_0}^{(1)}(t) < u_{i_0}^{(2)}(t)$ for all t due to uniqueness of the solution, Theorem 4.

Then we note that

$$\sum_{j=1}^N q_{i_0j} d(u_j^{(1)}(t_0), u_{i_0}^{(1)}(t_0)) = \sum_{j=1}^N q_{i_0j} d(u_j^{(1)}(t_0), u_{i_0}^{(2)}(t_0)) < \sum_{j=1}^N q_{i_0j} d(u_j^{(2)}(t_0), u_{i_0}^{(2)}(t_0))$$

if $\sum_{j=1, j \neq i_0}^N q_{i_0j}^2 \neq 0$.

So if $\sum_{j=1, j \neq i_0}^N q_{i_0j}^2 \neq 0$ then for all t in some neighborhood of t_0 we have $\frac{du_{i_0}^{(1)}}{dt}(t) < \frac{du_{i_0}^{(2)}}{dt}(t)$, hence, $\frac{d(u_{i_0}^{(1)} - u_{i_0}^{(2)})}{dt}(t) < 0$ and $(u_{i_0}^{(1)} - u_{i_0}^{(2)})(t) < 0$, in particular, $u_{i_0}^{(1)}(t_0) \neq u_{i_0}^{(2)}(t_0)$ —it is a contradiction.

If $\sum_{j=1, j \neq i_0}^N q_{i_0j}^2 = 0$ then the functions $u_{i_0}^{(1)}$ and $u_{i_0}^{(2)}$ are the solutions to the Cauchy problems (13), hence, we have $u_{i_0}^{(1)}(t) < u_{i_0}^{(2)}(t)$ for all t —it is a contradiction. \square

Corollary 1. Let $u^{(m)}$ be a solution for an initial value problem $u^{(m)}(0) = u_0^{(m)} \in [0; \max k_j]^N$, $m = 1, 2$. If we have $u^{(1)}(0) < u^{(2)}(0)$ then for all $t \geq 0$ we have $u^{(1)}(t) < u^{(2)}(t)$.

Proof. Let us choose neighborhoods U_1 and U_2 of $u^{(1)}(0)$ and $u^{(2)}(0)$ respectively which does not intersect. We can choose $v^{(1)}(0) \in U_1$ and $v^{(2)}(0) \in U_2$ in such a way that $v^{(1)}(0) \ll v^{(2)}(0)$. Then for all $t \geq 0$ we have $v^{(1)}(t) \ll v^{(2)}(t)$, it means that there are some neighborhoods $V_1(t)$ of $v^{(1)}(t)$ and $V_2(t)$ of $v^{(2)}(t)$ such that $V_1(t) \leq V_2(t)$ for all $t \geq 0$.

We fix $t_0 \geq 0$. The dynamical system $u(t, \zeta)$ is continuous with respect to the second variable ζ , when $\zeta \in [-a, \max k_j + a]$, $a > 0$. Hence, there is $\varepsilon_0 > 0$ such that for all $\varepsilon \in (0, \varepsilon_0]$ such that $V_1(t_0) = \{x \in \mathbb{R}^N \mid \|x - v^{(1)}(t_0)\| < \varepsilon\}$ and $V_2(t_0) = \{x \in \mathbb{R}^N \mid \|x - v^{(1)}(t_0)\| < \varepsilon\}$ there is $\delta > 0$ such that for all $\zeta_1 \in U_1 = \{x \in \mathbb{R}^N \mid \|x - u^{(1)}(0)\| < \delta\}$ and $\zeta_2 \in U_2 = \{x \in \mathbb{R}^N \mid \|x - u^{(2)}(0)\| < \delta\}$ we have $u(t_0, \zeta_1) \in V_1(t_0)$ and $u(t_0, \zeta_2) \in V_2(t_0)$, in particular, $u^{(1)}(t_0) \in V_1(t_0)$ and $u^{(2)}(t_0) \in V_2(t_0)$, but $V_1(t_0) \leq V_2(t_0)$, hence, $u^{(1)}(t_0) \leq u^{(2)}(t_0)$. Due to the arbitrariness of $t_0 \geq 0$ we conclude that for all $t \geq 0$ we have $u^{(1)}(t) \leq u^{(2)}(t)$. \square

Remark. Here we used the fact that for sufficiently small $\varepsilon_0 > 0$ the initial values lie in some open set containing $[0; \max k_j]^N$ in which the solution exists and is unique.

Corollary 2. Let $u^{(m)}$ be a solution for an initial value problem $u^{(m)}(0) = u_0^{(m)} \in [0; \max k_j]^N$, $m = 1, 2$. If we have $u^{(1)}(0) \leq u^{(2)}(0)$ then for all $t \geq 0$ we have $u^{(1)}(t) \leq u^{(2)}(t)$.

Proof. If $u^{(1)}(0) = u^{(2)}(0)$ then it is obviously true. The case $u^{(1)}(0) \neq u^{(2)}(0)$ follows from the previous corollary. \square

Corollary 3. Let $U \subset \mathbb{R}^N$. We define a set $u(t, U) = \{v \in \mathbb{R}^N \mid \text{there is } u_0 \in U : v = u(t, u_0)\}$. Then the dynamical system $u(t, \zeta)$ is strongly order-preserving, meaning that for $u^{(1)}(0) < u^{(2)}(0)$ there are neighborhoods U_1 and U_2 respectively such that for all $t \geq 0$ $u(t, U_1) \leq u(t, U_2)$.

Proof. The proof is done in the proof of Corollary 1. \square

Corollary 4. If for two solutions $u^{(1)}(t), u^{(2)}(t)$ we have $u_i^{(1)}(t_0) R u_i^{(2)}(t_0)$ for $R \in \{\leq, <, \ll\}$ and some $t_0 \in \mathbb{R}$ then $u_i^{(1)}(t) R u_i^{(2)}(t)$ for all $t \geq t_0$.

Proof. Let $v^{(m)}(t) = u^{(m)}(t + t_0)$, $m = 1, 2$. Then $\frac{dv^{(m)}}{dt}(t) = \frac{du^{(m)}}{dt}(t + t_0) = F(u^{(m)}(t + t_0)) = F(v^{(m)}(t))$, $v^{(m)}(0) = u^{(m)}(t_0)$, $m = 1, 2$. For all i we have $v_i^{(1)}(0) = u_i^{(1)}(t_0) R u_i^{(2)}(t_0) = v_i^{(2)}(0)$. \square

Theorem 8. Let the function of two variables $u(t, u_0)$ represent the solution of the Cauchy problem (10). Then setting for all $i \in \overline{1, N}$ $u_{0i} = \max k_j$ there is a limit $\lim_{t \rightarrow \infty} u(t, u_0) = \hat{u}$ which is a steady-state of the system (9), $\lim_{t \rightarrow \infty} \frac{du}{dt}(t, u_0) = 0$. Moreover, for all $e \in E$ (the set of all equilibrium points) we have $\hat{u} \geq e$.

Proof. $u(t, u_0) \in [0, \max k_j]^N$ for all $t \geq 0$, hence, there is $t_0 > 0$ such that for all $T \in (0, t_0)$ we have $u(T, u_0) \leq u_0$. Hence, there is a limit $\lim_{t \rightarrow \infty} u(t, u_0) = \hat{u}$; see [60], p. 248, Theorem 1.4 (Convergence Criterion).

For all $v_0 \in [0, \max k_j]^N$ we have $v_0 \leq u_0$, hence, $u(t, v_0) \leq u(t, u_0)$ for all $t \geq 0$. For $v_0 \in E$ we have $v_0 \leq u(t, u_0)$, and as $t \rightarrow \infty$ we have $v_0 \leq \hat{u}$. \square

Theorem 9. If for the Cauchy problem (10) there is at least one point $u_0 > 0$ such that $\frac{du}{dt}(0) \gg 0$ then there is a non-zero steady-state point \hat{u} such that $\lim_{t \rightarrow \infty} u(t, u_0) = \hat{u}$.

Proof. The function $\frac{du}{dt}$ is continuous as a derivative of a solution to the problem (10), hence, there is $T > 0$ such that $\frac{du}{dt}(t) \gg 0$ for all $t \in [0; T]$, hence, $u(T, u_0) \gg u_0$, in particular, $u(T, u_0) > u_0$. But by the corollary 3 the dynamical system u is strongly order-preserving, hence, there is a limit $\lim_{t \rightarrow \infty} u(t, u_0) = \hat{u} > u_0 > 0$ ([60], Theorem 1.4). \square

5. Computer Simulations

Here we will consider a system of N equations representing a chain of populations:

$$\begin{aligned} \frac{du_1}{dt} &= f_1(u_1) + qd(u_2, u_1), \\ \frac{du_i}{dt} &= f_i(u_i) + qd(u_{i-1}, u_i) + qd(u_{i+1}, u_i), \quad i \in \overline{2, N-1}, \\ \frac{du_N}{dt} &= f_N(u_N) + qd(u_{N-1}, u_N), \end{aligned}$$

where $f_i(u_i) = \alpha_i u_i(u_i - \beta_i)(1 - \frac{u_i}{k_i})$ and $d(y, x) = y - x$ or $d(y, x) = y^2 - x^2$.

In this section we focus on finding one global parameter $p(\beta, k)$ which somewhat characterize the system for all q . Here we will let $\alpha_i = 1$ for all i . We consider $\{k_i\}$ to be uniformly distributed on interval $[k_{min}, k_{max}]$, $\{\beta_i\}$ to be uniformly distributed, where each β_i is uniformly distributed on interval $[0; k_i]$, $\{k_i\}$ and $\{\beta_i\}$ are independent. So $\{k_i\}$ and $\{\beta_i\}$ can be defined by the following formulas:

$$\begin{aligned} k_i &= k_{min} + (k_{max} - k_{min})\phi, \\ \beta_i &= k_i\psi, \quad i \in \overline{1, N}, \end{aligned}$$

where ϕ, ψ are two independent random variables uniformly distributed on $[0, 1]$.

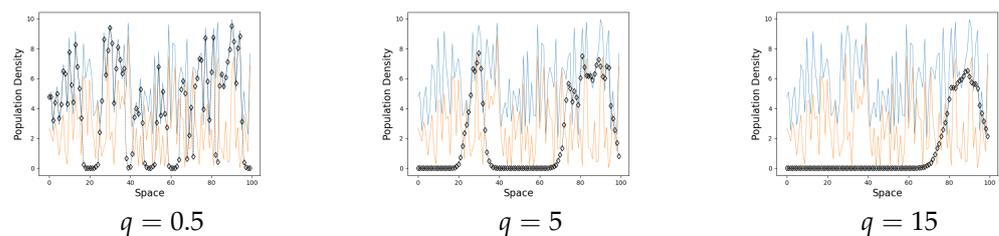
Let $p = \bar{k} - 2\bar{\beta} = \frac{1}{N}((k_1 + \dots + k_N) - 2(\beta_1 + \dots + \beta_N))$. By the weak law of large numbers [61] we have $p \approx E(k) - 2E(\beta) = \frac{k_{max} + k_{min}}{2} - 2(k_{min}E(\psi) + (k_{max} - k_{min})E(\phi)E(\psi)) = \frac{k_{max} + k_{min}}{2} - k_{min} - \frac{k_{max} - k_{min}}{2} = 0$. Here we will show that slightly changing p around 0 leads to bifurcation in most of the systems, in particular, there is a “small” constant $p^* > 0$ such that if $p > p^*$ then we can guarantee the persistence. Analytically the constant is still unknown, but here we try to find it approximately using examples.

An optimal value for N is 100, for this N the parameter p is not too large, not too small. We simulate both types of coupling using the RK45 method, which is programmed in Python using `scipy.integrate.solve_ivp`. We let $u_i(0) = \max k_j$ and change q with a step size of 0.5 from 0.5 to 20. It was checked in simulations that $t = 200$ was sufficiently large to ensure the system’s convergence to its steady-state distribution, for linear case we had to set the value of related tolerance to an error $rtol = 10^{-6}$ instead of default $rtol = 10^{-3}$ to ensure the convergence for large q .

For the quadratic coupling we have 5 test trials then we generate 5 random values of k and β in a predetermined range of p . From the data we conclude that the constant $p_2 \approx 0.52$ and $p_2 > 0.514$.

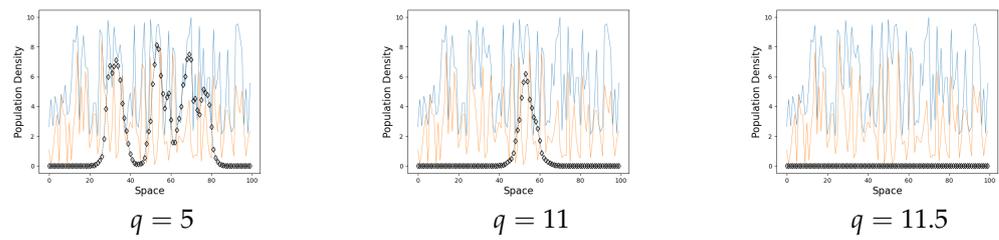
For linear coupling we run the simulation on the same data then add other trials in a predetermined range of p . For linear coupling we have $p_1 \approx 0.0164$.

We focus on the asymptotical steady-state behaviour of the system and hence show only the final metapopulation distribution. Figure 2 shows the examples of persistence and extinction.



(a) Persistence, linear, $p = 0.01655$.

Figure 2. Cont.



(b) Extinction, linear, $p = 0.0164$.

Figure 2. On the graphs the black line represents the population on each site when $t = 200$. The blue line is a carrying capacity of each site. The orange line is an Allee threshold of each site. The steady state is generally near the corresponding value of k for a small q , it can drop to 0 on a rare occasion. An increase in the coupling strength q eventually leads to the formation of clusters. The populations of the same cluster tend to align as q increases.

Below are Table 1 with cases which demonstrated persistence and Table 2 with extinct cases for $q = 20$ with their last q which gave the persistence, we also may show the distribution for a smaller parameter q . We note that for a better precision in a linear case we have to consider larger qs or more examples because in a quadratic model the absolute value of a coupling term grows faster. Here for the sake of uniformity we have chosen the second option. We begin both tables with the quadratic model as it is simpler in these ranges. We skip some of the examples.

Table 1. Persistence cases. The letter in the index in the column " $k^{(m)}, \beta^{(m)}$ " represents the dataset we use (L for linear, Q for quadratic), the number represents the iteration, the test dataset is marked by just a number.

Case No.	Model	q	Which $k^{(m)}, \beta^{(m)}$	\bar{k}	$\bar{\beta}$	p
1	Quadratic	20	7Q	6.0829	2.7824	0.5180
2	Quadratic	20	8Q	6.3336	2.9024	0.5287
3	Quadratic	20	9Q	6.1039	2.7939	0.5161
4	Linear	5	1	5.9931	2.7475	0.4981
5	Linear	20	1	5.9931	2.7475	0.4981
6	Linear	5	5	6.1951	3.2314	-0.2678
7	Linear	20	5	6.1951	3.2314	-0.2678
8	Linear	5	6L	5.8518	2.9176	0.0166
9	Linear	20	6L	5.8518	2.9176	0.0166
10	Linear	5	9L	5.9594	2.9715	0.0163
11	Linear	20	9L	5.9594	2.9715	0.0163
12	Linear	20	10L	6.1707	3.0772	0.0162
13	Linear	20	11L	5.9465	2.9655	0.0155
14	Linear	5	13L	5.9663	2.9750	0.0163
15	Linear	20	13L	5.9663	2.9750	0.0163
16	Linear	5	14L	5.7967	2.8902	0.0164
17	Linear	20	14L	5.7967	2.8902	0.0164
18	Linear	20	15L	5.8942	2.9389	0.0165
19	Linear	5	17L	6.1019	3.0428	0.0163
20	Linear	20	17L	6.1019	3.0428	0.0163

Table 2. Extinction cases. The letter in the index in the column "Which $k^{(m)}, \beta^{(m)}$ " represents the dataset we use (L for linear, Q for quadratic), the number represents the iteration, the test dataset is marked by just a number. NI marks trivial cases that are not interesting.

Case No.	Model	q	Which $k^{(m)}, \beta^{(m)}$	\bar{k}	$\bar{\beta}$	p
21	Quadratic	5	1	5.9931	2.7475	0.4981
22	Quadratic	11.5	1	5.9931	2.7475	0.4981
23 (NI)	Quadratic	1.5	2	5.8322	3.2905	-0.7489
24 (NI)	Quadratic	0.5	3	5.9883	3.0180	-0.0477
25 (NI)	Quadratic	1	3	5.9883	3.0180	-0.0477
26 (NI)	Quadratic	3.5	4	6.4793	3.2230	0.0333
27	Quadratic	0.5	5	6.1951	3.2314	-0.2678
28	Quadratic	4.5	5	6.1951	3.2314	-0.2678
29	Quadratic	4.5	6Q	6.0280	2.7574	0.5132
30	Quadratic	7.5	10Q	6.0660	2.7756	0.5147
31	Linear	9.5	2	5.8322	3.2905	-0.7489
32	Linear	5	3	5.9883	3.0180	-0.0477
33	Linear	5	4	6.4792	3.2230	0.0333
34	Linear	13	4	6.4792	3.2230	0.0333
35 (NI)	Linear	13	7L	5.4677	2.7261	0.0156
36	Linear	10	8L	5.9101	2.9469	0.0163
37	Linear	12	8L	5.9101	2.9469	0.0163
38 (NI)	Linear	8.5	12L	5.9783	2.9812	0.0159
39	Linear	5	16L	5.6687	2.8261	0.0164
40	Linear	11	16L	5.6687	2.8261	0.0164

Now we show Figure 3 corresponding to Table 1 and Figure 4 corresponding to Table 2.

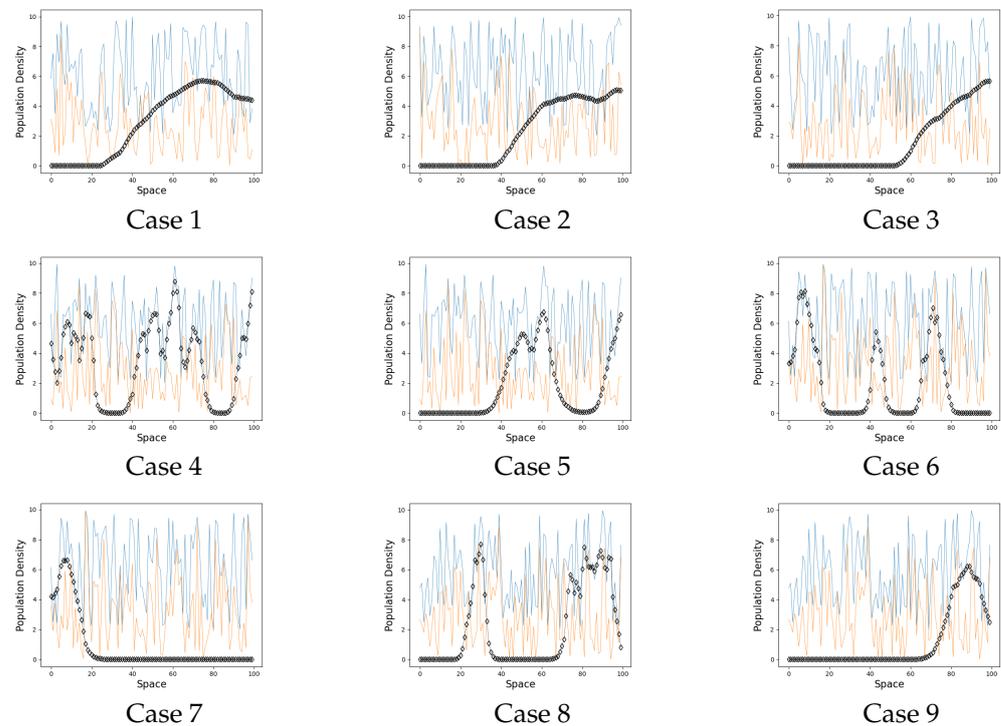


Figure 3. Cont.

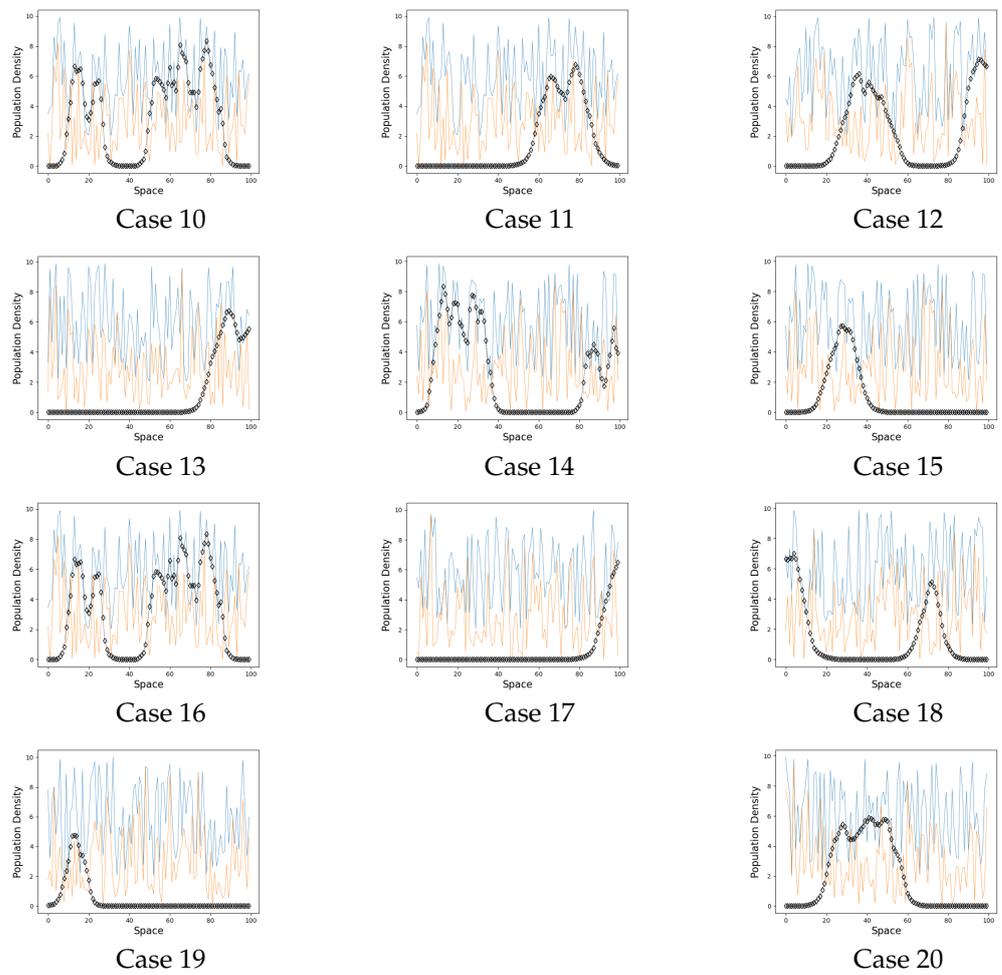


Figure 3. Persistence cases. On the graphs the black line represents the population on each site when $t = 200$. The blue line is a carrying capacity of each site. The orange line is an Allee threshold of each site.

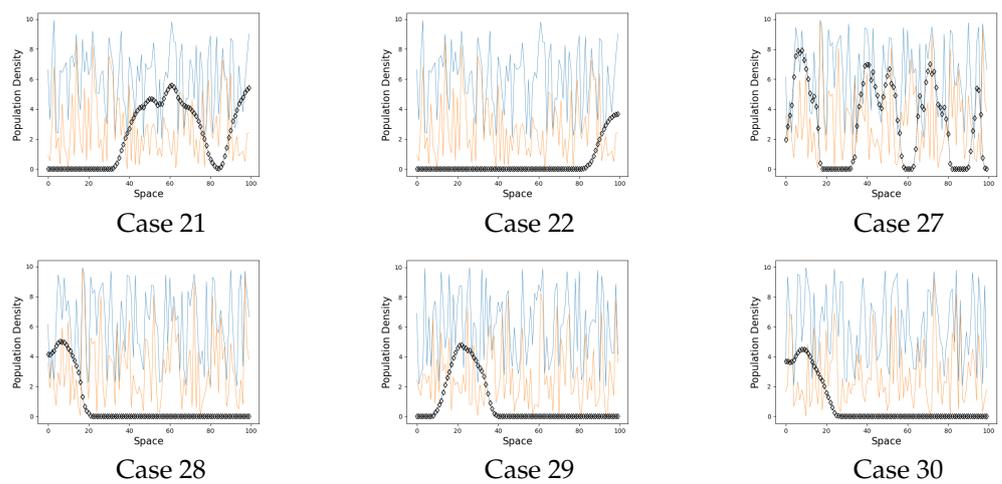


Figure 4. Cont.

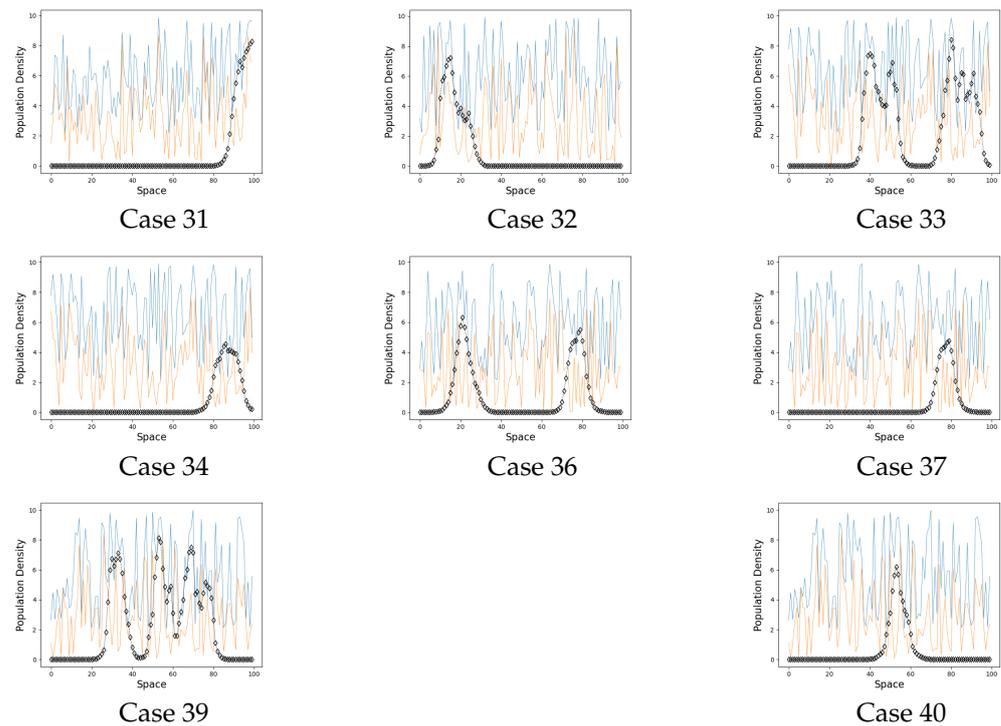


Figure 4. Extinction cases. On the graphs the black line represents the population on each site when $t = 200$. The blue line is a carrying capacity of each site. The orange line is an Allee threshold of each site.

6. Discussion and Concluding Remarks

Nature has many complex and fragmented environments and there are still many open theoretical problems [11,12,15,22,32,46,62]; conditions resulting in population collapse and species extinction in a fragmented habitat have long been a focus of the metapopulation theory. Previous research has identified specific factors, such as habitat geometry and demographic/environmental stochasticity, which can contribute to metapopulation collapse under certain conditions [31–34]. This study aims to contribute to this ongoing discourse by presenting another factor that could potentially result in metapopulation extinction. We investigate a system of arbitrary connected populations; we are primarily concerned with the conditions which correspond to persistence and extinction.

We first considered a baseline two-patch metapopulation. We continued the research done in [50] giving more sufficient conditions which can be subdivided into a condition on a system type (systems without Allee effect), a condition on extrema of growth functions f_i , conditions on q . Then we considered an arbitrary multi-patch system and showed that some of the conditions on q can be extended on the multi-patch system. We showed that the solution to the Cauchy problem exists and is unique, analytic and bounded. We showed that the model belongs to the class of so called monotone dynamical systems, which is very common in mathematical biology [60], and got some important corollaries from that, including another sufficient condition.

We then considered a 1D random metapopulation: a string of patches coupled by a short-distance dispersal (i.e., where each patch is coupled to its immediate neighbours) where the carrying capacity and the Allee threshold of the local population growth is a random function of space and stated a one-parameter sufficient condition. Computer simulations were supported by theoretical results. In particular, Theorem 8 basically tells us that we indeed converge to some steady-state point in Section 4. From the numerical results it can be seen that an increase in coupling may either lead to metapopulation collapse and global species extinction or to the formation of ‘persistence clusters’ (groups of patches where the subpopulations persist) separated by large stretches of empty space where the

subpopulations go extinct. We emphasize that the persistence clusters are completely self-organized, as our model does not include any long-distance correlations. A slight change in the vector α causes a slight change in the boundary p^* of the parameter $p = \bar{k} - 2\beta$, so this sufficient condition is also applicable to more general systems where $\alpha \neq (1, \dots, 1)^T$.

Thus, the study of this conceptual model can be considered complete. This paper continues the study done in the paper [50] of the mechanism that may lead to, on one hand, metapopulation extinction or, on the other hand, pattern formation through creating persistence clusters. Although the model used in this paper is very simple, it may give a rise to some important ecological interpretations and stimulate further study. Real ecosystems are usually much more complex: there can be multiple mechanisms; moreover, they can turn on and off independently from each other under specific conditions. A single-species model is typically only applicable on certain timescales [63]. Therefore, it is worth considering more complex models to reveal whether there is similar mechanism as in this model. Despite useful insights from previous work [15,16,19], this issue remains controversial. For example adding other species with some interaction laws to the model may cause the appearance of periodic and chaotic solutions. Coupling different habitats may greatly change the dynamics leading to appearance of new mechanisms or to synchronization of mechanisms between the habitats. All these issues should be studied further in future research.

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