A population changes model of spatial distributed two-species consocietum is considered. Species react as a "predator-prey" system. Each species occupies two states. Migrations lead to the problem of a net reproduction maximization. The basic model and its three reflexive modifications are considered.

Keywords: population dynamics, purposeful migration, spatially distributed population.

Models of spatially distributed populations and/or communities are still a matter of challenge for the students working in population biology, ecology, environmental sciences and mathematical modelling. The key problem in the present investigation is to construct an adequate model that describes models and predicts the impact of the spatial structure on a community dynamics, as well as the migration processes. Adequate modelling of a spatial transfer of a being is the basic difficulty here. Currently, the basic methodology addressing the problem is based on the implementation of partial differential equation of "reaction + kinetics" type.

Such an approach has a serious discrepancy. To be valid, the models strongly require that the beings move over space randomly and spontaneously (aimlessly). This constraint is never met in nature; even microorganisms control their spatial redistribution [1, 2, 3].

Previously, there was proposed an approach to model the dynamics of a community with respect to spatial effects based on the (micro)evolutionary principle [3, 4, 5]. In brief, the principle forces beings to migrate in the manner improving their existence. An improvement of existence of beings is the key question here; the answer is given by the net reproduction function [6, 7, 8, 5] $k(\rho, \vec{r})$. Here $\rho = \rho(\vec{r})$ is the (local) density of a population, and $\vec{r}$ is a point in space. Obviously, the population density $\rho$ depends on the point at space.

A net reproduction function results from two effects: the first is reproduction, and the second is inheritance [6, 7, 8]. If these two effects take place, then an equation of the dynamics of a biological entity has the following form:

$$\dot{\rho}(\vec{r}, t) = \rho(\vec{r}, t) \cdot k(\rho(\vec{r}, t)) \quad \text{or} \quad \rho_{t+1}(\vec{r}, t) = \rho_t(\vec{r}, t) \cdot k(\rho_t(\vec{r}, t)),$$

(1)

for discrete time. Here $k(\rho)$ ($k(N)$, respectively) is the net reproduction function. This function must be bounded from above. The equations (1) are the equations with inheritance. A comprehensive theory of such equations, in the most general case, including the investigation of three types of evolutionary stability can be found in [5, 6, 7, 8]. The most general result is that the net reproduction function $k(\rho)$ must meet the extreme principle, for any free evolving biological community. The function $k(\rho)$ is maximal for those species (entities) that sustained during the evolution; $k(\rho) = 0$ in the case of continuous time, and $k(\rho) = 1$ for discrete time.
Finally, the basic issue of our model of spatially dependent dynamics of a community is that any migration must not be a random walk, but the result of the growth of a net reproduction.

Further, we shall consider a model in discrete time and discrete space sites (called stations). A study of continuous model leads to severe technical problems, so we shall start with the discrete case.

1. A Model of Two-Species Community

We shall study the dynamics of a community consisting of two species; they are supposed to be “prey ÷ predator” related. It means that one species exists due to external resources, but the other one lives due to the beings of the former species. It is also supposed that both species occupy two stations* and migration means a transfer of beings (of any species) from station to station. Any other movements (inevitable in a real situation) are neglected and supposed to have no effect on the community dynamics.

The dynamics of (isolated) subcommunity occupying a station is supposed to follow the discrete analogue of classic Lotka-Volterra equation, if no migration takes place:

\[
\begin{align*}
N_{t+1} &= N_t \cdot (a - bN_t - fX_t), \\
M_{t+1} &= M_t \cdot (c - dM_t - gY_t), \\
X_{t+1} &= X_t \cdot (\varepsilon fN_t - hX_t), \\
Y_{t+1} &= Y_t \cdot (\varepsilon gM_t - kY_t).
\end{align*}
\]  

(2)

Here \(N_t, X_t\) are the abundances of prey and predator, respectively, at the first station, and \(M_t, Y_t\) are the analogous variables at the second station. The parameters \(a\) and \(c\) determine the fertility of the prey population, in the relevant stations; the parameters \(b\) and \(d\) describe the density-dependent self-regulation of this population, in the relevant station. The parameters \(h\) and \(k\) describe the similar density-dependent regulation at predator subpopulations. The parameters \(f\) and \(g\) describe, in the general form, an efficiency of the interaction of the beings of these two species, including the success of hunting, the success of escape, etc., in corresponding stations. Finally, \(\varepsilon\) represents the efficiency of conversion of the prey biomass into the predator biomass.

1.1. A Basic Model of Migration

The parameter \(p, 0 \leq p \leq 1\) reflects the mobility of prey beings; similar, \(q, 0 \leq q \leq 1\) is the mobility of predator beings. These parameters are the transfer cost and might be interpreted as a probability of the successful migration from one station to the other; success here means that no damage for further reproduction has taken place. Migration from station A to station B starts, if living conditions “there” are better, than “here”, with respect to the transfer cost:

\[
\begin{align*}
(a - bN_t - fX_t) &< p \cdot (c - dM_t - gY_t), \\
(\varepsilon fN_t - hX_t) &< q \cdot (\varepsilon gM_t - kY_t),
\end{align*}
\]  

(3)

for prey and predator beings, respectively. It should be stressed, that the migration act is executed independently by each being, while the model considers it as a population event. The backward migration conditions are defined similarly:

\[
\begin{align*}
p \cdot (a - bN_t - fX_t) &> (c - dM_t - gY_t), \\
q \cdot (\varepsilon fN_t - hX_t) &> (\varepsilon gM_t - kY_t).
\end{align*}
\]  

(4)

* These are the sites considered together with the environmental conditions.
A migration act runs at each time moment \( t \), independently for both species. If none of the inequalities (3, 4) is fulfilled, then no migration takes place, at the given time moment \( t \). Prey migration flux \( \Delta \) (predator migration flux \( \Theta \), respectively) must equalize the inequalities (3, 4):

\[
(a - b(N_t - \Delta) - fX_t) = p \cdot (c - d(M_t + p\Delta) - gY_t),
\]
\[
(\varepsilon f N_t - h(X_t - \Theta)) = q \cdot (\varepsilon g M_t - k(Y_t + q\Theta))
\]

in the case (3), or

\[
p \cdot (a - b(N_t + p\Delta) - fX_t) = (c - d(M_t - \Delta) - gY_t),
\]
\[
q \cdot (\varepsilon f N_t - h(X_t + q\Theta)) = (\varepsilon g M_t - k(Y_t - \Theta))
\]

in the case (4). Then, \( \Delta \) (\( \Theta \), respectively) is equal to

\[
\Delta = \frac{pc - a + bN - pdM + f\bar{X} - pg\bar{Y}}{b + p^2d}, \quad \Theta = \frac{hX + \varepsilon gM - \varepsilon fN - qkY}{h + q^2k}
\]

for migration from station \( A \) to station \( B \), and

\[
\Delta = \frac{pa - c + dM - pbN + gY - pf\bar{X}}{d + p^2b}, \quad \Theta = \frac{kY + \varepsilon gN - \varepsilon gM - qhX}{k + q^2h}
\]

for the backward migration.

Finally, let’s the outline how the basic model (2–6) works. For each time moment \( t \), a direction and the migration fluxes \( \Delta \) and \( \Theta \), respectively) are determined. Then, the species redistribute themselves according to the Eqs. (6). Then, the abundances of the next generation \( \{N_{t+1}, X_{t+1}; M_{t+1}, Y_{t+1}\} \) are determined, according to (2), with the relevant abundances of the current generation \( \{\bar{N}_t, \bar{X}_t; \bar{M}_t, \bar{Y}_t\} \) defined by (5). If no migration takes place at the current time moment \( t \), the stage with species redistribution is omitted.

1.2. Reflexive Behaviour

Reflection in behaviour means the ability of a being to foresee and/or predict the behaviour of an opponent, in a competitive behavioural act. An implementation of reflexive behavioural strategy by animals is well known. Without going into further discussion of psychological or ethological aspects of the implementation of such strategies, let us concentrate on a simple model revealing their dynamic of effects.

The basic model (2–6) does not exhibit any reflexive behaviour. An introduction of such a behaviour into the basic model may only be concerned with a spatial redistribution. In other words, a reflection of the optimal migration behaviour means that a being is able to “foresee” the migration behaviour of the competitive species. With respect to it, one may assume the following patterns of the reflection in the behaviour of the species: (i) preys reflect predators; (ii) predators reflect preys, and, finally, (iii) both species reflect each other.

Thus, within the framework of our model, a reflection means that the species manifesting a reflection in the behaviour, detects the migration conditions and chooses the migration flux according to the abundances of a competing species, that would be produced due to its migration, not the current ones. In the case (i) the formula for \( \Theta \) will remain the same, but the formula for \( \Delta \) must be replaced by

\[
\Delta = \begin{cases} 
\frac{pc - a + bN - pdM + f\bar{X} - pg\bar{Y}}{b + p^2d} & \text{or} \\
\frac{pa - c + dM - pbN + g\bar{Y} - pf\bar{X}}{d + p^2b} & 
\end{cases}
\]
in depending on the migration direction. Here \( \tilde{X} \) and \( \tilde{Y} \) are determined according to (5).

Reciprocally, \( \tilde{N} \) and \( \tilde{M} \) are determined according to (5), in the case (ii), but the migration flux \( \Theta \) of the predator will be determined by

\[
\Theta = \begin{cases} 
\frac{hX + \varepsilon qg\tilde{M} - \varepsilon f\tilde{N} - qkY}{h + q^2k} & \text{or} \\
\frac{kY + \varepsilon qf\tilde{N} - \varepsilon g\tilde{M} - qhX}{k + q^2h},
\end{cases}
\]

depending on the migration direction.

Finally, if both species reciprocally reflect the behaviour of each other, then the basic model should be changed into the following one. On the first stage, both species determine the migration fluxes according to basic model (5, 6). Then, they redefine the migration fluxes (and migration direction, as well as the fact of migration) so that each species changes the current abundances of the competitive beings for those that could be produced due to a migration rule determined by the basic model. So, they redefine the fluxes, redistribute themselves between the stations, and reproduce.

2. Results and Discussion

The purpose of this paper is to figure out the sets of the parameters providing an evolutionary advantage to a bearer of some (reflexive, or not) spatial distribution strategy. Evolutionary advantage here is understood as an excess of the total abundance of some species realizing reflexive strategy, in comparison to the same species in the case of realization of regular (non-reflexive) strategy.

- #1 \( a = 3.1, c = 1.49999, b = 0.00098, d = 0.00099, h = 0.00052, k = 0.0005, f = g = 0.00542, \varepsilon = 0.099, p = q = 0.99. \)
- #2 \( a = 2.5, c = 1.49999, b = 0.00098, d = 0.00099, h = k = 0.0005, f = g = 0.0054, \varepsilon = 0.099, p = q = 0.99. \)
- #3 \( a = c = 1.5, b = d = 0.0001, h = 0.0001, k = 0.0005, f = g = 0.0059, \varepsilon = 0.05, p = q = 0.99. \)
- #4 \( a = c = 1.5, b = d = 0.0001, h = k = 0.00001, f = 0.0059, g = 0.0049, \varepsilon = 0.05, p = q = 0.99. \)
- #5 \( a = 1.7, c = 1.5, b = d = 0.0001, h = k = 0.000001, f = g = 0.005, \varepsilon = 0.05, p = q = 0.99. \)
- #6 \( a = 2.2, c = 2.1, b = d = 0.0001, h = k = 0.000001, f = g = 0.005, \varepsilon = 0.05, p = q = 0.99. \)

First of all, it should be said, that the basic model (2–6) exhibits a great diversity of the limiting regimes. It may be a steady state (in both stations, for both species), limit cycles of various length, and a complex irregular behaviour looking like a dynamical chaos. All these peculiar regimes may be met in combinations, with respect to a station and/or of a species. In general, a decrease of transfer cost \( p \) and \( q \) yields a simplification of an observed regime. Both the basic model (2–6) and its versions implementing various reflexive strategies of spatial distribution yield an expansion of the area of permissible parameter values, and the area of the phase space (i.e., abundance figures). All these issues are very interesting, from the point of view of the study of the models of optimally migrating communities, but they fall beyond the scope of our research.
Table 1. Comparison of various strategies of space distribution. $S$ — type of strategy: 1 — basic model, 2 — model i, 3 — model ii, 4 — model iii.

<table>
<thead>
<tr>
<th></th>
<th>$N$</th>
<th>$X$</th>
<th>$M$</th>
<th>$Y$</th>
<th></th>
<th>$S$</th>
<th>$N$</th>
<th>$X$</th>
<th>$M$</th>
<th>$Y$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>2089</td>
<td>5</td>
<td>4539</td>
<td>6</td>
<td>1</td>
<td>1513.5</td>
<td>1.4</td>
<td>1952.5</td>
<td>56.2</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>1930</td>
<td>20.5</td>
<td>2368</td>
<td>50</td>
<td>2</td>
<td>1438.8</td>
<td>4.1</td>
<td>2244.6</td>
<td>55.6</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>1692.9</td>
<td>43.7</td>
<td>1849.9</td>
<td>41.7</td>
<td>3</td>
<td>1524.5</td>
<td>0.47</td>
<td>1938.6</td>
<td>56.7</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>2142</td>
<td>0.1</td>
<td>5040</td>
<td>0.1</td>
<td>4</td>
<td>1530</td>
<td>0.01</td>
<td>5047</td>
<td>0.01</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>$N$</th>
<th>$X$</th>
<th>$M$</th>
<th>$Y$</th>
<th></th>
<th>$S$</th>
<th>$N$</th>
<th>$X$</th>
<th>$M$</th>
<th>$Y$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>3697</td>
<td>18</td>
<td>3698</td>
<td>18</td>
<td>1</td>
<td>3542</td>
<td>20.9</td>
<td>4073</td>
<td>14.3</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>4999</td>
<td>0.01</td>
<td>4999</td>
<td>0.01</td>
<td>2</td>
<td>4997</td>
<td>0.01</td>
<td>4998</td>
<td>0.01</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>3873</td>
<td>6.4</td>
<td>3869</td>
<td>6.5</td>
<td>3</td>
<td>3619</td>
<td>11.2</td>
<td>3739</td>
<td>8</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>4189</td>
<td>13.8</td>
<td>4190</td>
<td>13.8</td>
<td>4</td>
<td>4325</td>
<td>12.5</td>
<td>4384</td>
<td>11.5</td>
<td></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>$N$</th>
<th>$X$</th>
<th>$M$</th>
<th>$Y$</th>
<th></th>
<th>$S$</th>
<th>$N$</th>
<th>$X$</th>
<th>$M$</th>
<th>$Y$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>6994</td>
<td>0.05</td>
<td>4996</td>
<td>0.01</td>
<td>1</td>
<td>5.4</td>
<td>0.1</td>
<td>4.4</td>
<td>0.04</td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>5688</td>
<td>32.1</td>
<td>4122</td>
<td>11.6</td>
<td>2</td>
<td>8406</td>
<td>80.5</td>
<td>7172</td>
<td>69.3</td>
<td></td>
</tr>
<tr>
<td>3</td>
<td>4664</td>
<td>19.5</td>
<td>2961</td>
<td>13.9</td>
<td>3</td>
<td>3.6</td>
<td>0.07</td>
<td>3.6</td>
<td>0.1</td>
<td></td>
</tr>
<tr>
<td>4</td>
<td>5412</td>
<td>30.5</td>
<td>4086</td>
<td>18.7</td>
<td>4</td>
<td>8404</td>
<td>80.5</td>
<td>7171</td>
<td>69.3</td>
<td></td>
</tr>
</tbody>
</table>

What we did, was a comparative study of those four models (basic model and three versions with reflexive behaviour) from the point of view of the evolutionary advantage. In fact, we tried various combinations of the parameters, similar for all four models, in order to identify the model that yields the highest total abundance of the species.

To answer this question, we have carried out a series of computational experiments. We have calculated the abundance of each species, in each station, for four models with the same parameter set. Then, the abundances of prey subpopulation (of predator subpopulation, in turn) have been added. Table 1 shows the results of this comparison. The parameters yielding the observed regimes are shown below the Table.

It is evident that there exist parameters sets yielding an evolutionary advantage for various types of space distribution strategies. Yet, we did not study carefully the peculiarities of the limit regimes relevant to each evolutionary advantageous situation. Meanwhile, one may expect that the reflexive strategies provide an advantage for rather regular limit regimes, while the non-reflexive strategy of space distribution is advantageous for chaotic-like, complicated limit regimes.

The authors were supported by Grant NSh-3431.2008.9 from the President of Russian Federation.

References


