

Chaos Control in a Discrete Ecological System

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ABSTRACT

In research [1], the authors investigate the dynamic behaviors of a discrete ecological system. The period-double bifurcations and chaos are found in the system. But no strategy is proposed to control the chaos. It is well known that chaos control is the first step of utilizing chaos. In this paper, a controller is designed to stabilize the chaotic orbits and enable them to be an ideal target one. After that, numerical simulations are presented to show the correctness of theoretical analysis.

Keywords: Chaos Control; Discrete Ecological System; Numerical Simulation

1. Introduction

Population dynamics in ecology are generally governed by discrete and continuous systems. In recent years, the study of discrete ecological systems has attracted extensive attentions [1-6]. This is because that some natural populations have non-overlapping generations, thus discrete models are more realistic than continuous ones to study these species. Another reason is that people always study population changes by one year (month, week or day). Such investigations are often required discrete models. Especially, using discrete models is more efficient for numerical simulations. Recently, Zhang and Li [1] studied the following discrete ecological model:

$$\begin{aligned}x_{n+1} &= x_n + \delta(1 - x_n + x_n y_n) \\ y_{n+1} &= y_n + \delta(a - bx_n)y_n^2\end{aligned}\quad (1)$$

where x_n, y_n denote the two ecological species' densities respectively in generation n ; δ is the integral step size. The more meaning of system (1) can refer to the reference [1,2]. It is shown that the system (1) generates period-double bifurcations and chaos. But the authors did not investigate the chaos control of the system.

It is well known that chaos control is the first step of utilizing chaos. The possibility of chaos control in biological systems has been stimulated by recent advances in the study of heart and brain tissue dynamics. Recently, some authors have investigated that such a method can be applied to population dynamics and even play a non-trivial evolutionary role in ecology [7-9]. In this paper, we design a proper controller to control the chaos of system (1).

2. Chaos Control

In this section, chaotic orbits to an unstable fixed point are stabilized by utilizing some control techniques. Firstly, we introduce the following lemma which is useful to establish our results

Lemma 1 [1]. *If $a > b$, then system (1) has an unique positive fixed point at $E(x^*, y^*)$, where $x^* = a/b$, $y^* = 1 - b/a$.*

Consider the following map which is the feedback is applied to system (1)

$$X_{n+1} = F(X_n, \mu_n) = \begin{pmatrix} x_n + \delta(1 - x_n + x_n y_n) + \mu_n \\ y_n + \delta(a - bx_n)y_n^2 \end{pmatrix} \quad (2)$$

where $X_n = (x_n, y_n)^T$, μ_n is control variable and satisfies $|\mu_n| < \varepsilon$, $\varepsilon \ll 1$. Evidently, map (2) degenerates to original system (1) only if $\mu_n = 0$. We select the feedback variable μ_n in the range $(-\varepsilon, \varepsilon)$, so that the orbit holds in the neighborhood of fixed point E as long as the control arises. The ergodic nature of the chaotic dynamics guarantees that the mode trajectory in the neighborhood of the wishful orbit $E(x^*, y^*)$. In the neighborhood of E , map (2) can be approximated by the following form:

$$\begin{pmatrix} x_{n+1} - x^* \\ y_{n+1} - y^* \end{pmatrix} \rightarrow A \begin{pmatrix} x_n - x^* \\ y_n - y^* \end{pmatrix} + B\mu_n \quad (3)$$

where A is the Jacobian matrix at E and B is a column vector, and they are given by:

$$A = \frac{\partial F}{\partial X_n} \Big|_{X_n=E, \mu_n=0} = \begin{pmatrix} 1 - \frac{\delta b}{a} & \frac{\delta a}{b} \\ -\delta b \left(1 - \frac{b}{a}\right)^2 & 1 \end{pmatrix},$$

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$$B = \frac{\partial F}{\partial \mu_n} \Big|_{X_n=E, \mu_n=0} = \begin{pmatrix} 1 \\ 0 \end{pmatrix}.$$

Let $X^* = (x^*, y^*)^T$ and suppose that μ_n is a linear function of X_n , which is expressed as $\mu_n = P^T(X_n - X^*)$, $P^T \in R^{1 \times 2}$. Substitute the result into (3), we get

$$\begin{pmatrix} x_{n+1} - x^* \\ y_{n+1} - y^* \end{pmatrix} \rightarrow (A - BP^T) \begin{pmatrix} x_n - x^* \\ y_n - y^* \end{pmatrix}.$$

According to the study [10], the fixed point E will be stable if the matrix $(A - BP^T)$ is asymptotically stable, that is to say, all its eigenvalues are less than 1 in modulus. Now, we make use of "pole placement technique" [11] to determine the specific values in $(A - BP^T)$. If system (1) is chaotic, we obtain

$$\left| 1 - \frac{\delta b}{2a} \right| > 1.$$

Then we choose $\sigma_1 = 0$, $\sigma_2 = 1 - \frac{\delta b}{2a}$

as the desired eigenvalues of the matrix $(A - BP^T)$. The controllability matrix

$$C = (B \quad AB) = \begin{pmatrix} 1 & 1 - \frac{\delta b}{a} \\ 0 & -\delta b \left(1 - \frac{b}{a}\right)^2 \end{pmatrix}$$

has two rank. Thus the solution to the pole placement problem is obtained as

$$P^T = (q_2 - p_2 \quad q_1 - p_1)Q^{-1},$$

where $Q = CW$, $W = \begin{pmatrix} p_1 & 1 \\ 1 & 0 \end{pmatrix}$,

p_1 and p_2 are the coefficients of characteristic polynomial of the matrix A , $\det(\lambda I - A) = \lambda^2 + p_1\lambda + p_2$,

and $p_1 = -2 + \frac{\delta a}{b}$, $p_2 = \frac{\delta^2(a-b)^2 - \delta b}{a}$;

q_1 and q_2 are the coefficients of characteristic polynomial of the matrix $(A - BP^T)$,

$$\det(\sigma I - (A - BP^T)) = \lambda^2 + q_1\lambda + q_2,$$

and $q_1 = \frac{\delta b}{2a} - 1$, $q_2 = 0$.

After calculations, we get

$$P^T = \begin{pmatrix} \frac{2a - \delta b}{2a} & \frac{a[a - \delta b + \delta^2(a-b)^2]}{\delta b(a-b)^2} + \frac{a(2a - \delta b)}{2\delta b(a-b)^2} \end{pmatrix} \quad (4)$$

Furthermore, the controller has the following form:

$$\mu_n = P^T (\varepsilon x_n \quad \varepsilon y_n)^T,$$

where $\varepsilon x_n = x_n - x^*$, $\varepsilon y_n = y_n - y^*$.

However, the above considerations only are fit for a local small neighbor of E . In view of the global situation, we can specify μ_n by making $\mu_n = 0$ if $|P^T(X_n - X^*)|$ is too large. This is because the range of μ_n is restrained by $|\mu_n| < \varepsilon$ and $\varepsilon \ll 1$. Thus, we limit the number value

$$|X_n - X^*| < \frac{\varepsilon}{|P^T|}.$$

Therefore, in practice we take μ_n as

$$\mu_n = \begin{cases} P^T (\varepsilon x_n \quad \varepsilon y_n)^T & \text{if } |X_n - X^*| < \frac{\varepsilon}{|P^T|}, \\ 0 & \text{if } |X_n - X^*| > \frac{\varepsilon}{|P^T|}. \end{cases} \quad (5)$$

According to the above analysis, we get the following result.

Theorem 1. If $|X_n - X^*| < \frac{\varepsilon}{|P^T|}$,

then the control variable $\mu_n = P^T (\varepsilon x_n \quad \varepsilon y_n)^T$ can stabilize chaotic trajectory of system (1) to the fixed point E , where P^T is given by Equation (4).

3. Numerical Simulations

In the section, we use density-time diagrams and phase portraits to confirm the above theoretical analysis.

Let $a = 2.21$, $b = 1.02$, $\delta = 0.9666$. At the condition, $|P^T|$ has the value 7.70732. According to Lemma 1, system (1) has and only has a positive fixed point $E(x^*, y^*) = (2.16667, 0.53846)$. We adopt

$$B_r(E) = \left\{ (x_n, y_n) \mid (x_n - x^*)^2 + (y_n - y^*)^2 \leq (0.001)^2 \right\}.$$

When ε is given the value 0.03 and 0.09, Theorem 1 is satisfied. Density-time diagram of ecological specie x_n is given by **Figure 1(a)**, which is characterized by switches between apparently regular and chaotic behaviors. Actually, it is intermittency, which is a basic characteristic of chaos. At the same parameters, phase portrait is illustrated by **Figure 1(b)**, which is a chaotic attractor. **Figure 2** is the chaos control diagrams corresponding to **Figure 1**. With the same parameters of **Figure 1**, system (1) is chaotic if $n < 800$ when $\varepsilon = 0.03$ (**Figures 2(a) and (b)**) according to the control strategy. Actually, **Figure 2(a)** is supertransient, which is used to denote an unusually long convergence to an attractor. **Figure 2(b)** is phase portrait corresponding to **Figure 2(a)**. When ε increases to 0.09, supertransient disappears and the system

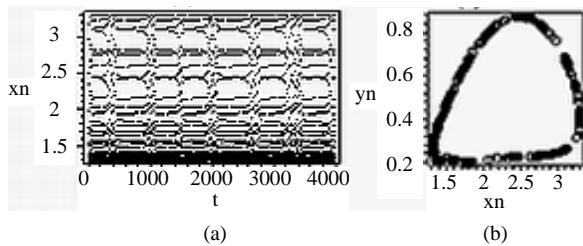


Figure 1. (a) Density-time diagram of x_n ; (b) phase portrait of x_n and y_n . Where the parameters given by $a = 2.21$, $b = 1.02$, $\delta = 0.9666$.

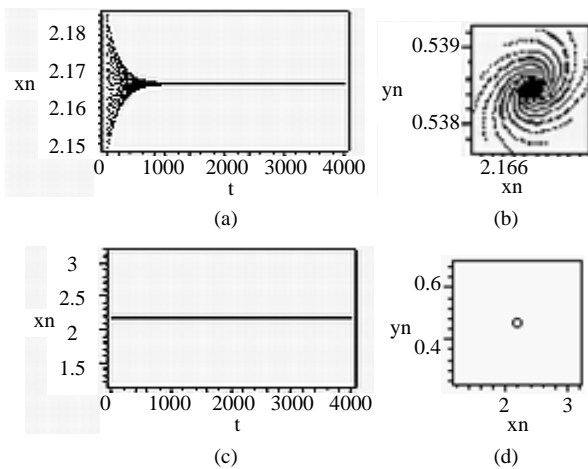


Figure 2. (a) Density-time controlled diagram of x_n with $\varepsilon = 0.03$; (b) controlled phase portrait with $\varepsilon = 0.03$; (c) Density-time controlled diagram of x_n with $\varepsilon = 0.09$; (d) controlled phase portrait with $\varepsilon = 0.09$. The other parameters given by $a = 2.21$, $b = 1.02$, $\delta = 0.9666$.

stabilizes to the fixed point $(2.16667, 0.53846)$, which is simulated by **Figures 2(c) and 2(d)**.

4. Conclusion

In this paper, we design a proper controller to control the chaos of system (1) which was firstly studied by Zhang and Li [6]. From the theoretical analysis, we concluded that the control variable $\mu_n = P^T (\varepsilon x_n \ \varepsilon y_n)^T$ can stabilize chaotic trajectory of system (1) to the fixed point $E(x^*, y^*)$ under the condition of

$$\left\| (X_n - X^*) \right\| < \frac{\varepsilon}{|P^T|},$$

where P^T is given by Equation (4). Then simulations are presented to show the correctness of theoretical analysis. **Figure 1** demonstrates system (1) is chaotic with parameters $a = 2.21$, $b = 1.02$, $\delta = 0.9666$. **Figure 2** indicates system (1) processes from supertransient to the fixed point when the control variable applied to the system.

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