

## **Research** Article

# Spatiotemporal Evolution Characteristics of Time-Delay Ecological Competition Systems with Food-Limited and Diffusion

Feilong Wang<sup>[b]</sup>,<sup>1</sup> Min Xiao<sup>[b]</sup>,<sup>1</sup> Zhengxin Wang<sup>[b]</sup>,<sup>2</sup> Jing Zhao<sup>[b]</sup>,<sup>1</sup> Gong Chen<sup>[b]</sup>,<sup>1</sup> and Jinde Cao<sup>[b]</sup><sup>3</sup>

<sup>1</sup>College of Automation & College of Artificial Intelligence, Nanjing University of Posts and Telecommunications, Nanjing 210023, China

<sup>2</sup>School of Science, Nanjing University of Posts and Telecommunications, Nanjing 210023, China

<sup>3</sup>School of Mathematics, Southeast University, Nanjing 210096, China, and Yonsei Frontier Lab, Yonsei University, Seoul 03722, Republic of Korea

Correspondence should be addressed to Min Xiao; candymanxm2003@aliyun.com

Received 6 April 2022; Accepted 6 July 2022; Published 3 August 2022

Academic Editor: Sergey Dashkovskiy

Copyright © 2022 Feilong Wang et al. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

In this paper, we put forward a time-delay ecological competition system with food restriction and diffusion terms under Neumann boundary conditions. For the case without delay, the conditions for local asymptotic stability and Turing instability are constructed. For the case with delay, the existence of Hopf bifurcation is demonstrated by analyzing the root distribution of the corresponding characteristic equations. Furthermore, by using the normal form theory and the center manifold reduction of partial functional differential equations, explicit formulas are obtained to determine the direction of bifurcations and the stability of bifurcating periodic solutions. Finally, some simulation examples are provided to substantiate our analysis.

## 1. Introduction

Based on the diversity of biological populations and ecosystems, various ecological competition systems have been established and widely studied (see [1-5]). In ecological competition systems, the interaction between populations is usually reflected by functional response functions. Results have shown that the predation relationship between populations greatly affects the dynamic behavior of predatorprey systems (see [6-8]). In fact, with the development of the economy, humans will harvest biological populations and develop related biological resources to obtain economic benefits, such as fisheries, forestry, and wildlife management (see [9–11]). In recent years, many scholars have introduced the harvesting terms into biological systems to study the system modeling and related dynamic characteristics [12–15]. May et al. [16] established the following model to analyse the interaction between populations under different harvesting strategies:

$$\begin{cases} \frac{dU}{dt} = r_1 U \left( 1 - \frac{U}{K} \right) - \alpha U V - H_1, \\ \\ \frac{dV}{dt} = r_2 V \left( 1 - \frac{V}{\beta U} \right) - H_2, \end{cases}$$
(1)

where U and V represent the prey and predator densities, respectively.  $r_1$  and  $r_2$  indicate the intrinsic growth rates. K represents the internal growth limit of the population without predators.  $\alpha$  stands for the maximum predation coefficient of predator.  $\beta$  is the quality standard for measuring prey as food.  $H_1$  and  $H_2$  represent the impact of human harvesting for predator and prey populations. On this basis, scholars have studied the relevant dynamic characteristics of different harvesting terms in system (1). In [17], the authors studied the case of  $H_1 = r_1h_1u$  and  $H_2 = r_2h_2v$  in system (1), that is, the constant effort harvesting to both prey and predator population. Particularly, they analyzed the maximum sustainable yield of another population under the condition of limiting the harvest of one population. In [18], the authors discussed the case of  $H_1 = h_1$ and  $H_2 = r_2 h_2 v$  for system (1), namely constant yield harvesting for prey and constant effort harvesting for the predator population. At the same time, the relevant dynamic characteristics of the system were analyzed and the effects of different management strategies on the stability of the system were compared. Moreover, in [19, 20], the authors considered the constant yield harvesting of the prey population in system (1), that is,  $H_1 = h_1$  and  $H_2 = 0$ . They obtained the conditions of Bogdanov-Takens bifurcation and supercritical/subcritical Hopf bifurcation of codimension 1.

Based on the research of the appeal literature, it is found that only the impact of linear capture is taken into account. However, from the perspective of biology and economics, the linear harvesting term cannot accurately reflect the real social activities (harvesting population and developing biological resources) of humans and their impact on the predator-prey system. Thus, the nonlinear harvesting terms have been introduced to model the dynamics of predatorprey systems in recent years (see [21–23]). In [24], a realistic harvesting functional form was proposed as follows:

$$H(t) = \frac{qEU(t)}{1+aE},$$
(2)

where q represents the catchability coefficient and E stands for the harvesting effort. The conditions for Hopf bifurcation were derived and a nonlinear state feedback controller was designed to control the Hopf bifurcation.

Consider the logistic equation

$$\frac{\mathrm{d}N}{\mathrm{d}t} = rN\left(1 - \frac{N}{C}\right). \tag{3}$$

Here, r is the intrinsic growth rate and C represents the carrying capacity. Meanwhile, we can see from (3) that the predicted relation of the specific growth rate, dN/N dt, to mass, N, is a straight line. However, when considering a measure of the portion of available limiting factors not yet utilized by the population, the linear average growth rate cannot accurately describe the growth trend of the population. Research shows the growth and development of organisms depend on the availability and utilization of food in the living environment, which implies that different degrees of food supply rates will affect the stable age composition of the population and then have an influence on the average growth rate of the population. Thus, when growth limitations are based on the proportion of available resources not utilized, Smith [25] established a food-limited growth function

$$\frac{\mathrm{d}M}{\mathrm{d}t} = \frac{rM\left(K-M\right)}{K+\left(r/c\right)M},\tag{4}$$

where c is positive constant and represents the rate of replacement of mass in the population at saturation. It follows from (4) that the predicted relation of the specific growth rate, dM/Mdt, to mass, M, is not a straight line but a concave curve.

Considering the growth limit is based on the proportion of unused available resources and setting  $H_1 = qEu/1 + fE$ ,  $H_2 = 0$ , then system (1) becomes

$$\begin{cases} \frac{\mathrm{d}u}{\mathrm{d}t} = \frac{r_1 u \left(K - u\right)}{K + \alpha u} - F\left(u, v\right) - \frac{qEu}{1 + fE},\\ \frac{\mathrm{d}v}{\mathrm{d}t} = r_2 v \left(1 - \frac{v}{p + bu}\right), \end{cases}$$
(5)

where F(u, v) represents the functional response function between populations. *p* represents that when the density of prey population is low, the predator population can switch to other prey for capture. Then, we will analyse the cases on  $F(u, v) = \beta uv$  and  $F(u, v) = e_1 uv/e_2 + u$  in the discussion that follows, respectively.

In nature, the survival and development of a population often depend on the amount of food and the living space available. Importantly, the greater the population density, the higher the requirements for the living environment. At the same time, the acquisition of food mainly depends on the living environment of the population, which shows that the change of the population's living space affects the survival and development of the population to a great extent. Therefore, the population will instinctively migrate and diffuse in space to seek a more suitable environment for survival and development. It is necessary to consider the influence of the diffusion effect on population dynamics in predator-prey systems. Mathematically, the nonlinear system with diffusion will show complex dynamic properties [26-29]. In the reaction-diffusion system proposed by Turing [30, 31], the spatial heterogeneity caused by the internal reaction-diffusion characteristics of the system results in the loss of system symmetry and makes the system self-organize to produce some spatial patterns. The process of pattern formation is called Turing instability (Turing bifurcation). The symmetry of the system is broken, leading to the formation of Turing patterns. Therefore, we call this phenomenon "Turing instability caused by diffusive reaction" [32].

On the other hand, time delay has become a factor that cannot be ignored in many biological dynamic systems. A large number of studies have revealed that time delay has an important impact on the dynamic characteristics of biological systems and it is common in predator-prey systems, mainly including mature time delay, capture time delay, and pregnancy time delay [33-35]. Local stability of the system means that if the initial state is adjacent to the equilibrium state, the system will not vibrate, and its state trajectory will eventually fall to the equilibrium state. In particular, Hopf bifurcation is a dynamic bifurcation phenomenon, which shows that when the parameters change near the critical value, the stability of the equilibrium point will change and periodic solutions will be generated in its small neighborhood. Meanwhile, it is found that time delay as a bifurcation parameter can induce Hopf bifurcation [29, 36, 37]. Therefore, in this paper, we consider the pregnancy delay of the predator population and analyse the dynamic characteristics of ecological competition systems.

Based on the discussion above, we introduce the diffusion effect of the population [26–29] and the pregnancy delay of the predator population [33–35] into system (5) to explore its impacts on the dynamic characteristics of the ecological competition system, which can be described by

$$\begin{cases} \frac{\partial u(t,x)}{\partial t} = D_1 \Delta u + \frac{r_1 u(K-u)}{K+\alpha u} - F(u,v) - \frac{qEu}{1+fE}, \\ \frac{\partial v(t,x)}{\partial t} = D_2 \Delta v + r_2 v \left(1 - \frac{v}{p+bu(t-\tau)}\right), \end{cases}$$
(6)

with Neumann boundary conditions and initial conditions

$$u(t, x) = u_0(t, x) \ge 0, v(t, x)$$
  
=  $v_0(t, x) \ge 0, (t, x) \in [-\tau, 0] \times \overline{\Omega},$   
$$\frac{\partial u(t, x)}{\partial \eta} = \frac{\partial v(t, x)}{\partial \eta}$$
  
=  $0, t > 0, \quad x \in \partial \Omega,$  (7)

Here, u(t, x), v(t, x) stand for the population density of prey and predator at a spatial location x and time t, respectively.  $D_1, D_2 > 0$  represent the diffusion coefficients associated to u and v, respectively.  $\Delta$  denotes the Laplacian operator in  $\mathbb{R}^n$ . Suppose  $\Omega = (0, \pi)$  is a bounded domain with a smooth boundary  $\partial \Omega$ .

Lemma 1. For any solution of system (6) without delay,

$$\limsup_{t \to \infty} u(t, x) \leq \frac{r_1 K (1 + fE) - qEK}{qE\alpha + r_1 (1 + fE)},$$

$$\limsup_{t \to \infty} v(t, x) \leq p + \frac{b(r_1 K (1 + fE) - qEK)}{qE\alpha + r_1 (1 + fE)}.$$
(8)

*Proof.* Let (u(t, x), v(t, x)) be a nonnegative solution of system (6) without delay. Note that the functional response F(u, v) > 0. Then,

$$\begin{split} \frac{\partial u}{\partial t} &\leq D_1 \Delta u + \frac{r_1 u \left(K - u\right)}{K + \alpha u} - \frac{qEu}{1 + fE}, \ x \in (0, \pi), \quad t > 0, \\ u_x(t, 0) &= u_x(t, \pi) \\ &= 0, \ t > 0, \\ \frac{\partial v}{\partial t} &\leq D_2 \Delta v + r_2 v \left(1 - \frac{v}{p + bu}\right), \ x \in (0, \pi), \quad t > 0, \\ v_x(t, 0) &= v_x(t, \pi) \\ &= 0, \ t > 0. \end{split}$$

$$(9)$$

We can estimate the upper limits of u(t, x) and v(t, x) due to the standard comparison principle:

$$u(t,x) \le \frac{r_1 K (1 + fE) - qEK}{qE\alpha + r_1 (1 + fE)},$$

$$v(t,x) \le p + \frac{b(r_1 K (1 + fE) - qEK)}{qE\alpha + r_1 (1 + fE)}.$$
(10)

In other words, for arbitrary  $\varepsilon_1 > 0$ ,  $\varepsilon_2 > 0$ , there exists positive constants  $t_1, t_2$  such that

$$u(t,x) \le \frac{r_1 K \left(1 + fE\right) - qEK}{qE\alpha + r_1 \left(1 + fE\right)} + \varepsilon_1,\tag{11}$$

for  $t \ge t_1$ ,  $x \in \Omega$ , and

$$v(t,x) \le p + \frac{b(r_1 K (1 + fE) - qEK)}{qE\alpha + r_1 (1 + fE)} + \varepsilon_2,$$
(12)

for  $t \ge t_2$ ,  $x \in \Omega$ . So, the conclusion follows immediately.

Compared with the models proposed in [1-5], model (6) is a new measure of the population with a nonlinear average growth rate based on food restriction. At the same time, we consider the reaction-diffusion factors for system (6) to study their influence on the dynamic behaviors of systems. In addition, on the basis of references [16-20], we introduce the nonlinear harvesting term of the harvesting-effort coefficient *E* into system (6) and study the stability and related dynamic characteristics of the system under Holling I and Holling II functional response functions. Importantly, the addition of pregnancy delay can more accurately reflect the evolution of the population and make the system show more complex dynamic characteristics than the model without delay, which is also a widely concerned direction in the research of biological systems.

The main contributions of this paper can be stated as follows. In Section 2, the effects of diffusion on the dynamic behavior of the systems without time delay are investigated and some conditions for system stability and Turing instability are determined. It is found that the appropriate diffusion coefficients will lead to Turing instability. In Section 3, we analyse the stability of equilibrium and Hopf bifurcation in the predator-prey system with time delay as the bifurcation parameter. The condition for Hopf bifurcation is constructed and the expression of the bifurcation threshold is given. In Section 4, we calculated the direction of the bifurcations to get more information about the bifurcations. Section 5 uses some numerical examples to verify the correctness of the previous derivation. Section 6 gives the conclusion of this paper. 

## 2. Equilibrium Stability and Turing Instability Analysis

Assume that the predator-prey relationship in system (6) satisfies Holling type I functional response, that is  $F(u, v) = \beta uv$ , then we make the following nondimensional scaling transformation [22]:

Then system (10) has a positive equilibrium 
$$E^* = (u^*, v^*)$$
, where  $v^* = \theta(p + u^*)/b$  and  $u^*$  satisfies the following quadratic equation:

$$A_0 u^2 + A_1 u + A_2 = 0, (17)$$

where

$$A_{0} = \alpha n > 0,$$

$$A_{1} = 1 + \alpha n p + n + \alpha S > 0,$$

$$A_{2} = S + n p - 1,$$

$$n = \frac{\theta}{b},$$

$$S = \frac{hE}{(1 + fE)}.$$
(18)

**Lemma 2.** For equation (12), we come to the following results:

If A<sub>2</sub> > 0, then equation (12) has no positive roots.
 If A<sub>2</sub> < 0, then equation (12) has a unique positive root u<sup>\*</sup> = -A<sub>1</sub> + √A<sub>1</sub><sup>2</sup> - 4A<sub>0</sub>A<sub>2</sub>/2A<sub>0</sub>.

Hence, when  $A_2 < 0$ , system (10) has a unique positive equilibrium  $E^* = (u^*, v^*)$ , where  $v^* = \theta(p + u^*)/b$ .

Consider system (10) without diffusion

$$\begin{cases} \frac{\mathrm{d}u(t)}{\mathrm{d}t} = \frac{u(1-u)}{1+\alpha u} - uv - \frac{hEu}{1+fE},\\ \frac{\mathrm{d}v(t)}{\mathrm{d}t} = v \left[\theta - \frac{bv}{p+u}\right]. \end{cases}$$
(19)

The Jacobian matrix of (19) at  $E^*$  is  $\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}$ , where  $a_{11} = \frac{1 - 2u^* - \alpha u^{*2}}{(1 + \alpha u^*)^2} - v^* - \frac{hE}{1 + fE},$ 

$$a_{12} = -u^* < 0,$$

$$a_{21} = \frac{bv^{*2}}{(p+u^*)^2} > 0,$$

$$a_{22} = \theta - \frac{2bv^*}{p+u^*} = -\theta < 0.$$
(20)

**Lemma 3.** When  $A_2 < 0$ , then we can deduce  $a_{11} < 0$ .

Proof. Notice that

$$\widetilde{t} = r_1 t,$$

$$\widetilde{D}_1 = \frac{D_1}{r_1},$$

$$\widetilde{D}_2 = \frac{D_2}{r_2},$$

$$h = \frac{q}{r_1},$$

$$\theta = \frac{r_2}{r_1},$$
(13)

$$\widetilde{b} = \frac{r_2}{b\beta K},$$

$$\widetilde{\rho}$$

 $\tilde{u} = \frac{u}{K},$ 

 $\tilde{v} = \frac{\beta v}{r_1},$ 

$$\tilde{p} = \frac{p}{bK},$$

Dropping the tildes, system (6) can be rewritten by

$$\begin{cases} \frac{\partial u(t,x)}{\partial t} = D_1 \Delta u + \frac{u(1-u)}{1+\alpha u} - uv - \frac{hEu}{1+fE}, \\ \frac{\partial v(t,x)}{\partial t} = D_2 \Delta v + v \left[ \theta - \frac{bv}{p+u(t-\tau)} \right]. \end{cases}$$
(14)

Consider the case of no time delay in system (9), namely,  $\tau = 0$ , then system (9) becomes

$$\begin{cases} \frac{\partial u(t,x)}{\partial t} = D_1 \Delta u + \frac{u(1-u)}{1+\alpha u} - uv - \frac{hEu}{1+fE}, \\ \frac{\partial v(t,x)}{\partial t} = D_2 \Delta v + v \left[ \theta - \frac{bv}{p+u} \right]. \end{cases}$$
(15)

Considering the practical significance of the ecosystem, we are interested in the coexisting equilibrium. In order to obtain the positive equilibrium of system (10), let

$$\begin{cases} \frac{u(1-u)}{1+\alpha u} - uv - \frac{hEu}{1+fE} = 0, \\ v\left[\theta - \frac{bv}{p+u}\right] = 0. \end{cases}$$
(16)

$$a_{11}(1 + \alpha u^{*})^{2} = 1 - 2u^{*} - 2u^{*2} - v^{*}(1 + \alpha u^{*})^{2} - S(1 + \alpha u^{*})^{2}$$

$$= 1 - 2u^{*} - 2u^{*2} - n(c + u^{*})(1 + 2\alpha u^{*} + \alpha^{2}u^{*2})$$

$$- S(1 + 2\alpha u^{*} + \alpha^{2}u^{*2})$$

$$= -\alpha \left[A_{0}u^{*3} + A_{1}u^{*2} + nu^{*2} + \frac{1}{\alpha}(2A_{1} - n)u^{*} - \frac{A_{2}}{\alpha}\right]$$

$$= -\alpha \left[-A_{2}u^{*} + \frac{A_{0}u^{*2}}{\alpha} + \frac{2A_{1}u^{*}}{\alpha} - \frac{nu^{*}}{\alpha} - \frac{A_{2}}{\alpha}\right]$$

$$= -\alpha \left[-A_{2}u^{*} + \frac{A_{0}u^{*2} + A_{1}u^{*}}{\alpha} + \frac{A_{1}u^{*}}{\alpha} - \frac{nu^{*}}{\alpha} - \frac{A_{2}}{\alpha}\right]$$

$$= -\alpha \left[-A_{2}u^{*} + \frac{A_{1}u^{*}}{\alpha} - \frac{nu^{*}}{\alpha} - \frac{2A_{2}}{\alpha}\right]$$

$$= -\alpha \left[(1 + \frac{1}{\alpha})u^{*} - \frac{2A_{2}}{\alpha}\right].$$

$$X = \left\{v \in W^{2,2}(0, \pi): \dot{v}(0) = \dot{v}(\pi) = 0\right\}, \quad (24)$$

Thus, we can obtain  $a_{11}(1 + \alpha u^*)^2 < 0$  when  $A_2 < 0$ . As  $(1 + \alpha u^*)^2 > 0$ , then we can deduce  $a_{11} < 0$ .

Then from the Jacobian matrix of (13), we get the characteristic equation

$$\lambda^2 - B_1 \lambda + B_2 = 0, \qquad (22)$$

where

$$B_1 = a_{11} + a_{22}, \tag{23}$$

$$B_2 = a_{11}a_{22} - a_{12}a_{21}.$$

**Theorem 1.** If  $A_2 < 0$ , then  $E^*$  of system (13) is locally asymptotically stable.

*Proof.* By Lemma 3, we know  $a_{11} < 0$  when  $A_2 < 0$ . Together with  $a_{12} < 0$ ,  $a_{21} > 0$  and  $a_{22} < 0$ , we can get  $B_1 < 0$  and  $B_2 > 0$ . Thus, by the Routh-Hurwitz criterion,  $E^*$  of system (13) is locally asymptotically stable.

It follows from [38] that the Laplacian operator  $-\Delta$  has the eigenvalue  $k^2$  ( $k \in \{0, 1, 2, ...\}$ ) under the homogeneous Neumann boundary condition. And the corresponding eigenfunctions are  $\beta_k^1 = (\iota_k, 0)^T, \beta_k^2 = (0, \iota_k)^T$ , where  $\iota_k = \cos(kx)$  and  $\{\beta_k^1, \beta_k^2\}_{k=0}^{\infty}$  construct a basis of the phase space X and X is defined by with the inner product  $\langle \cdot, \cdot \rangle$ . Thus, for system (10), the characteristic equation at  $E^*$  is

$$\lambda^2 + C_{1k}\lambda + C_{2k} = 0, (25)$$

where

$$C_{1k} = (D_1 + D_2)k^2 - a_{11} - a_{22},$$

$$C_{2k} = D_1 D_2 k^4 - (a_{11}D_2 + a_{22}D_1)k^2 + a_{11}a_{22} - a_{12}a_{21}.$$
Hence,
$$\lambda_1 + \lambda_2 = -C_{1k} = -(D_1 + D_2)k^2 + a_{11} + a_{22},$$

$$\lambda_1 \lambda_2 = C_{2k} = D_1 D_2 k^4 - (a_{11}D_2 + a_{22}D_1)k^2 + a_{11}a_{22} - a_{12}a_{21}.$$
(26)

$$\lambda_1 \lambda_2 = C_{2k} = D_1 D_2 k^4 - (a_{11} D_2 + a_{22} D_1) k^2 + a_{11} a_{22} - a_{12} a_{21}.$$
(27)

**Theorem 2.** If  $A_2 < 0$ , then  $E^*$  of system (10) is asymptotically stable.

*Proof.* It follows from Lemma 3 that  $a_{11} < 0$  when  $A_2 < 0$ . Notice that  $a_{12} < 0$ ,  $a_{21} > 0$  and  $a_{22} < 0$ . For equation (14), we obtain

$$\lambda_{1} + \lambda_{2} = -C_{1k} = -(D_{1} + D_{2})k^{2} + a_{11} + a_{22} < 0,$$
  

$$\lambda_{1}\lambda_{2} = C_{2k} = D_{1}D_{2}k^{4} - (a_{11}D_{2} + a_{22}D_{1})k^{2} + a_{11}a_{22} - a_{12}a_{21} > 0, \ k \in N_{0}.$$
(28)

(24)

Thus, all roots of (25) have negative real parts for  $k \in N_0$ , which implies that  $E^*$  of system (10) is asymptotically stable.

*Remark* 1. Assume that there are no time delays and  $F(u, v) = \beta uv$  for system (6). It can be seen from Theorems 1 and 2 that when  $A_2 < 0$ , the introduction of diffusion terms does not change the stability of  $E^*$ , which means that Turing instability does not occur.

Next, we consider the predator-prey relationship among populations in system (6) satisfying Holling type II functional response, that is  $F(u, v) = e_1 uv/e_2 + u$ , where  $e_1$  represents the maximum per capita reduction rate of the prey population and  $e_2$  stands for the average saturation rate [3]. Then we make the following nondimensional scaling transformation [22]:

$$\begin{split} \widetilde{u} &= \frac{u}{K}, \\ \widetilde{v} &= \frac{e_1 v}{r_1}, \\ \widetilde{t} &= r_1 t, \\ \widetilde{D_1} &= \frac{D_1}{r_1}, \\ \widetilde{D_2} &= \frac{D_2}{r_2}, \\ h &= \frac{q}{r_1}, \\ \theta &= \frac{r_2}{r_1}, \\ \widetilde{b} &= \frac{r_2}{be_1 K}, \\ \widetilde{p} &= \frac{p}{bK}, \\ \widetilde{e_1} &= \frac{1}{K}, \end{split}$$

(29)

 $\tilde{e_2} = \frac{e_2}{K}.$ 

Dropping the tildes, then system (6) without time delay turns into

$$\begin{cases} \frac{\partial u(t,x)}{\partial t} = D_1 \Delta u + \frac{u(1-u)}{1+\alpha u} - \frac{e_1 u v}{e_2 + u} - \frac{hEu}{1+fE}, \\ \frac{\partial v(t,x)}{\partial t} = D_2 \Delta v + v \left[ \theta - \frac{bv}{p+u} \right]. \end{cases}$$
(30)

System (15) has a positive equilibrium  $E^* = (u^*, v^*)$ , where  $v^* = \theta(p + u^*)/b$  and  $u^*$  satisfies the following quadratic equation:

$$Y_0 u^2 + Y_1 u + Y_2 = 0, (31)$$

where

$$Y_{0} = \alpha m + \alpha S + 1 > 0,$$
  

$$Y_{1} = \alpha m p + m + \alpha e_{2}S + S + e_{2} - 1,$$
  

$$Y_{2} = e_{2}S + m p - e_{2},$$
  

$$m = e_{1} \frac{\theta}{b}.$$
  
(32)

**Lemma 4.** For equation (16), we have the following results:

(1) If  $Y_1 < 0$ ,  $Y_1^2 - 4Y_0Y_2 > 0$  and  $Y_2 > 0$ , then equation (16) has positive roots

$$u_{1,2} = \frac{-Y_1 \pm \sqrt{Y_1^2 - 4Y_0 Y_2}}{2Y_0}.$$
 (33)

- (2) If  $Y_1 < 0$  and  $Y_1^2 4Y_0Y_2 = 0$ , then equation (16) has a unique positive root  $u_0 = -Y_1/2Y_0$ .
- (3) If  $Y_2 < 0$ , then equation (16) has a unique positive root  $u^* = -Y_1 + \sqrt{Y_1^2 - 4Y_0Y_2}/2Y_0.$

**Theorem 3.** For system (15), we come to the following results:

- (1) If  $Y_1 < 0$ ,  $Y_1^2 4Y_0Y_2 > 0$  and  $Y_2 > 0$ , then system (15) has two positive equilibrium points  $E_1(u_1, v_1)$  and  $E_2(u_2, v_2)$ , where  $v_{1,2} = \theta(p + u_{1,2})/b$ .
- (2) If  $Y_1 < 0$  and  $Y_1^2 4Y_0Y_2 = 0$ , then system (15) has a unique positive equilibrium point  $E_0 = (u_0, v_0)$ , where  $v_0 = \theta(p + u_0)/b$ .
- (3) If  $Y_2 < 0$ , then system (15) has a unique positive equilibrium point  $E^* = (u^*, v^*)$ , where  $v^* = \theta(p + u^*)/b$ .

The linearization of system (15) at  $E^*$  can be represented by

$$\begin{pmatrix} \frac{\partial u}{\partial t} \\ \frac{\partial v}{\partial t} \end{pmatrix} = D\Delta \begin{pmatrix} u(t,x) \\ v(t,x) \end{pmatrix} + J \begin{pmatrix} u(t,x) \\ v(t,x) \end{pmatrix}, \quad (34)$$

where  $D = \text{diag}(D_1, D_2), J = \begin{pmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{pmatrix}$ , and

$$b_{11} = \frac{1 - 2u^* - \alpha u^{*2}}{(1 + \alpha u^*)^2} - \frac{e_1 e_2 v^*}{(e_2 + u^*)^2} - \frac{hE}{1 + fE},$$
  

$$b_{12} = -\frac{e_1 u^*}{e_2 + u^*},$$
  

$$b_{21} = \frac{bv^{*2}}{(p + u^*)^2},$$
(35)

 $b_{22} = \theta - \frac{2bv^*}{p+u^*}.$ 

Thus, the characteristic equation is

$$\lambda^2 + C_{3k}\lambda + C_{4k} = 0, (36)$$

where

$$C_{3k} = (D_1 + D_2)k^2 - b_{11} - b_{22},$$

$$C_{4k} = D_1 D_2 k^4 - (b_{11} D_2 + b_{22} D_1)k^2 + b_{11} b_{22} - b_{12} b_{21}.$$
(37)

Obviously, for (36), we have

$$\lambda_{1} + \lambda_{2} = -C_{3k}$$

$$= -(D_{1} + D_{2})k^{2} + b_{11} + b_{22},$$

$$\lambda_{1}\lambda_{2} = C_{4k}$$

$$= D_{1}D_{2}k^{4} - (b_{11}D_{2} + b_{22}D_{1})k^{2} + b_{11}b_{22} - b_{12}b_{21}.$$
(38)

We make the following assumptions:

$$(H1)b_{11} + b_{22} < 0,$$
  

$$(H2)b_{11}b_{22} - b_{12}b_{21} > 0,$$
  

$$(H3)b_{11}D_2 + b_{22}D_1 < 0.$$
  
(39)

**Theorem 4.** If (H1) - (H3) hold and  $Y_2 < 0$ , then system (15) is locally asymptotically stable at  $E^*$ .

*Proof.* If (H1) - (H3) and  $Y_2 < 0$  hold, for equation (17), we have

$$\lambda_{1} + \lambda_{2} = -C_{3k}$$

$$= -(D_{1} + D_{2})k^{2} + b_{11} + b_{22} < 0,$$

$$\lambda_{1}\lambda_{2} = C_{4k}$$

$$= D_{1}D_{2}k^{4} - (b_{11}D_{2} + b_{22}D_{1})k^{2} + b_{11}b_{22} - b_{12}b_{21} > 0.$$
(40)

Thus, all roots of (36) have negative real parts. By the Routh-Hurwitz criterion, system (15) is locally asymptotically stable at  $E^*$ .

Then we can describe system (15) without diffusion term by the following equations:

$$\begin{cases} \frac{\mathrm{d}u(t)}{\partial t} = \frac{u(1-u)}{1+\alpha u} - \frac{e_1 u v}{e_2 + u} - \frac{hEu}{1+fE},\\ \frac{\mathrm{d}v(t)}{\partial t} = v \left[\theta - \frac{bv}{p+u}\right]. \end{cases}$$
(41)

The characteristic equation of system (18) at  $E^*$  is

$$\lambda^2 - C_3 \lambda + C_4 = 0, \qquad (42)$$

where

I ot

$$C_{3} = b_{11} + b_{22},$$

$$C_{4} = b_{11}b_{22} - b_{12}b_{21}.$$
(43)

**Lemma 5.** If (H1), (H2), and  $Y_2 < 0$  are satisfied, then  $E^*$  of system (18) is locally asymptotically stable.

$$\Phi_1(\varepsilon) = D_1 D_2 \varepsilon^4 - (b_{11} D_2 + b_{22} D_1) \varepsilon^2 + b_{11} b_{22} - b_{12} b_{21}.$$
(44)

We make the following assumptions:

$$(H4)b_{11}D_2 + b_{22}D_1 > 0,$$

$$(H5)b_{11}D_2 + b_{22}D_1 - 2\sqrt{D_1D_2(b_{11}b_{22} - b_{12}b_{21})} > 0.$$
(45)

**Lemma 6.** If (H2), (H4), and (H5) hold, then  $\Phi_1(\varepsilon) = 0$  has two positive roots  $\varepsilon_1, \varepsilon_2$ .

**Theorem 5.** Assume that  $Y_2 < 0$  and (H1), (H2), (H4), and (H5) hold. Then, diffusion-driven instability occurs for system (15) at  $E^*$  if there exists a  $k \in N_0$  such that  $C_{4k} < 0$  for  $0 < \varepsilon_1 < k < \varepsilon_2$ .

*Proof.* It follows from Lemmas 5 and 6 that system (18) is stable at  $E^*$  and  $\Phi_1(\varepsilon) = 0$  has two positive roots  $\varepsilon_1, \varepsilon_2$  when  $Y_2 < 0$  and (H1), (H2), (H4), and (H5) hold. Then for the spatial system (15), the corresponding characteristic equation (17) has an eigenvalue with positive real part if there exists a  $k \in N_0$  such that  $C_{4k} < 0$  for  $0 < \varepsilon_1 < k < \varepsilon_2$ . This implies the spatial system (15) is unstable at  $E^* = (u^*, v^*)$ , that is, the diffusion-driven instability occurs.

*Remark 2.* Assume that there are no time delays and  $F(u, v) = e_1 uv/e_2 + u$  for system (6). It can be seen from Theorem 5 that the introduction of diffusion terms may change the stability of  $E^*$ , resulting in Turing instability.

#### 3. Hopf Bifurcation Analysis

In this section, we consider the effect of time delay on the dynamics of the system and get the conditions for Hopf bifurcation of system (9).

Linearizing system (9) at  $E^* = (u^*, v^*)$  [34–36, 39], we have

$$\begin{pmatrix} \frac{\partial u}{\partial t} \\ \frac{\partial v}{\partial t} \end{pmatrix} = D\Delta \begin{pmatrix} u(t,x) \\ v(t,x) \end{pmatrix} + J_1 \begin{pmatrix} u(t,x) \\ v(t,x) \end{pmatrix} + J_2 \begin{pmatrix} u(t-\tau,x) \\ v(t-\tau,x) \end{pmatrix},$$
(46)

where  $D = \text{diag}(D_1, D_2)$ ,  $J_1 = \begin{pmatrix} a_{11} & a_{12} \\ 0 & a_{22} \end{pmatrix}$ ,  $J_2 = \begin{pmatrix} 0 & 0 \\ a_{21} & 0 \end{pmatrix}$ . Thus, the characteristic equation is

$$\det\left(\lambda I - D_k - J_1 - J_2 e^{-\lambda \tau}\right) = 0, \tag{47}$$

where *I* stands for  $2 \times 2$  identity matrix and  $D_k = -k^2 \operatorname{diag}(D_1, D_2), k \in N_0$ . (47) is equivalent to

$$\lambda^{2} + m_{1k}\lambda + m_{2k} + m_{3}e^{-\lambda\tau} = 0, \qquad (48)$$

where

$$m_{1k} = (D_1 + D_2)k^2 - a_{11} - a_{22},$$
  

$$m_{2k} = D_1 D_2 k^4 - (a_{11} D_2 + a_{22} D_1)k^2 + a_{11} a_{22},$$
  

$$m_3 = -a_{12} a_{21}.$$
(49)

*Remark 3.* It should be pointed out that this paper adopts the linearization method [29, 36, 37, 39,40] to deal with the dynamics analysis of system (9), including the local stability, Turing instability, and Hopf bifurcation. It is common knowledge that Lyapunov's second method is important to the stability theory of dynamical systems and control theory. However, this method is not suitable for investigating the dynamics of the ecological competitive system with delay and diffusion proposed in this paper. The Lyapunov stability criterion can only give a sufficient condition for the stability of a system. In this paper, not only the condition of the local stability is established, but also the boundary of stability (the onset of Hopf bifurcation) is determined.

**Lemma 7.** If  $A_2 < 0$ , then all roots of equation (21) have negative real parts when  $\tau = 0$ .

When  $\tau > 0$ , assume  $\lambda = \pm i\omega$  ( $\omega > 0$ ) is a pair of pure imaginary roots of (48),  $\omega$  satisfies the following equation:

$$-\omega^2 + im_{1k}\omega + m_{2k} + m_3 e^{-i\omega\tau} = 0.$$
 (50)

By separating real and imaginary parts, we obtain

$$\begin{cases} -\omega^2 + m_{2k} = -m_3 \cos(\omega \tau), \\ m_{1k}\omega = m_3 \sin(\omega \tau). \end{cases}$$
(51)

Add the squares of both sides of equation (23) to get

$$A^{4} + P_{k}\omega^{2} + Q_{k} = 0,$$
 (52)

where

$$P_{k} = m_{1k}^{2} - 2m_{2k} = (a_{11} - D_{1}k^{2})^{2} + (a_{22} - D_{2}k^{2})^{2} > 0,$$
  

$$Q_{k} = m_{2k}^{2} - m_{3}^{2} = (m_{2k} + m_{3})(m_{2k} - m_{3}).$$
(53)

Let  $z = \omega^2$ , (52) can be converted to

ώ

$$z^2 + P_k z + Q_k = 0. (54)$$

Then we make the following assumptions:

$$(H6)a_{11}a_{22} + a_{12}a_{21} > 0,$$
  
(H7) $a_{11}a_{22} + a_{12}a_{21} < 0.$  (55)

**Theorem 6.** If (H6) and  $A_2 < 0$  hold, then system (9) is locally asymptotically stable at  $E^*$  for all  $\tau \ge 0$ .

*Proof.* By Theorem 2, we know  $m_{2k} + m_3 = C_{2k} > 0$  when  $A_2 < 0$ . If (H6) is satisfied, we obtain

$$m_{2k} - m_3 = D_1 D_2 k^4 - (a_{11} D_2 + a_{22} D_1) k^2 + a_{11} a_{22} + a_{12} a_{21} > 0.$$
(56)

Thus, (54) has no positive roots, which implies (48) has no purely imaginary roots. By Lemma 7, we deduce that all roots of (48) have negative real parts for all  $\tau \ge 0$ , that is, system (9) is locally asymptotically stable at  $E^* = (u^*, v^*)$  for all  $\tau \ge 0$ .

Let

$$\Phi_2(\varepsilon) = D_1 D_2 \varepsilon^4 - (a_{11} D_2 + a_{22} D_1) \varepsilon^2 + a_{11} a_{22} + a_{12} a_{21}.$$
(57)

It follows that when (*H7*) and  $A_2 < 0$  hold,  $\Phi_2(\varepsilon) = 0$  has a unique positive root  $\varepsilon_0$  having the following form:

$$\varepsilon_{0} = \sqrt{\frac{\left(a_{11}D_{2} + a_{22}D_{1}\right) + \sqrt{\left(a_{11}D_{2} + a_{22}D_{1}\right)^{2} - 4D_{1}D_{2}\left(a_{11}a_{22} + a_{12}a_{21}\right)}{2D_{1}D_{2}}}.$$
(58)

Denote

$$k^* = \begin{cases} [\varepsilon_0], \ \varepsilon_0 \notin N, \\ \varepsilon_0 - 1, \ \varepsilon_0 \in N. \end{cases}$$
(59)

**Lemma 8.** If (H7) and  $A_2 < 0$  are satisfied, equation (21) has a pair of purely imaginary roots  $\pm i\omega_k (0 \le k \le k^*)$  at  $\tau = \tau_k^j$ , and

#### Complexity

$$\tau_{k}^{j} = \tau_{k}^{0} + \frac{2j\pi}{\omega_{k}}, \ j \in N_{0},$$
  
$$\tau_{k}^{0} = \frac{1}{\omega_{k}} \arccos \frac{\omega_{k}^{2} - m_{2k}}{m_{3}},$$
  
$$\sqrt{2m_{2k} - m_{2k}^{2} + \sqrt{\left(m_{2k}^{2} - 2m_{2k}\right)^{2} - 4\left(m_{2k}^{2} - m_{2}^{2}\right)}}$$
(60)

$$\omega_{k} = \sqrt{\frac{2m_{2k} - m_{1k}^{2} + \sqrt{(m_{1k}^{2} - 2m_{2k})^{2} - 4(m_{2k}^{2} - m_{3}^{2})}{2}}$$

**Theorem 7.** If (H7) and 
$$A_2 < 0$$
 hold, we have

$$\tau_{k^*}^j \ge \tau_{k^*-1}^j \ge \tau_{k^*-2}^j \cdots \tau_1^j \ge \tau_0^j.$$
(61)

Proof. From equation (28), we obtain

$$\omega_{k}^{2} = \frac{2m_{2k} - m_{1k}^{2} + \sqrt{\left(m_{1k}^{2} - 2m_{2k}\right)^{2} - 4\left(m_{2k}^{2} - m_{3}^{2}\right)}}{2}$$

$$= \frac{2}{\left(m_{1k}^{2} - 2m_{2k}/m_{3}^{2} - m_{2k}^{2}\right) + \sqrt{\left(\left(m_{1k}^{2} - 2m_{2k}\right)^{2}/\left(m_{3}^{2} - m_{2k}^{2}\right)^{2}\right) + \left(4/m_{3}^{2} - m_{2k}^{2}\right)}}$$
(62)

By Lemma 3, we know that  $a_{11} < 0$  when  $A_2 < 0$ . Together with  $a_{12} < 0$ ,  $a_{21} > 0$  and  $a_{22} < 0$ , we can get that  $a_{11}D_2 + a_{22}D_1 < 0$ ,  $m_{2k}^2 - 2m_{2k}$  is increasing and  $m_3^2 - m_{2k}^2$  is decreasing in  $k \in [0, k^*]$  when (*H7*) hold. Thus,

$$\omega_{k^*} \le \omega_{k^*-1} \le \omega_{k^*-2} \le \cdots \le \omega_1 \le \omega_0. \tag{63}$$

Notice that  $m_3 = -a_{12}a_{21} > 0$  and  $m_{2k}$  is increasing in  $k \in [0, k^*]$  with  $a_{11}D_2 + a_{22}D_1 < 0$ . We can deduce that  $\omega_k^2 - \frac{m_{2k}}{m_3}$  is decreasing in  $k \in [0, k^*]$ . Hence,  $\tau_k^j = 1/\omega_k \arccos \omega_k^2 - \frac{m_{2k}}{m_3} + \frac{2j\pi}{\omega_k}$  is increasing in  $k \in [0, k^*]$ . As  $\tau_k^{j+1} > \tau_k^j$ , we obtain

$$\tau_0^0 = \min\{\tau_k^j\}, \ 0 \le k \le k^*, \quad j \in N_0.$$
(64)

Let  $\lambda(\tau) = \rho(\tau) + i\sigma(\tau)$  be the root of (48) satisfying  $\rho(\tau_k^j) = 0$  and  $\sigma(\tau_k^j) = \omega_k$ .

**Theorem 8.** *For*  $k \in \{0, 1, \dots, k^*\}$ *, we have* 

$$\operatorname{Re}\left[\left(\frac{d\lambda}{d\tau}\right)\right]\Big|_{\tau=\tau_{k}^{j}} > 0.$$
(65)

*Proof.* By derivation of equation (21) with respect to the delay  $\tau$ , we have

$$\left(2\lambda + m_{1k} - m_3 \tau e^{-\lambda \tau}\right) \frac{\mathrm{d}\lambda}{\mathrm{d}\tau} = \lambda m_3 e^{-\lambda \tau},\tag{66}$$

then

$$\left(\frac{\mathrm{d}\lambda}{\mathrm{d}\tau}\right)^{-1} = \frac{(2\lambda + m_{1k})e^{\lambda\tau} - m_3\tau}{m_3\lambda}.$$
 (67)

It follows from (50) and (51) that

$$Re\left(\frac{d\lambda}{d\tau}\right)^{-1}\Big|_{\tau=\tau_{k}^{j}} = Re\left[\frac{(2i\omega_{k}+m_{1k})e^{i\omega_{k}\tau_{k}^{j}}-m_{3}\tau_{k}^{j}}{i\omega_{k}m_{3}}\right]$$
$$= \frac{2\omega_{k}\cos(\omega_{k}\tau_{k}^{j})+m_{1k}\sin(\omega_{k}\tau_{k}^{j})}{\omega_{k}m_{3}}$$
$$= \frac{2\omega_{k}^{2}+m_{1k}^{2}-2m_{2k}}{m_{3}^{2}}.$$
(68)

Noticing 
$$m_{1k}^2 - 2m_{2k} = P_k > 0$$
, we obtain  

$$\operatorname{Re}\left[\left(\frac{d\lambda}{d\tau}\right)^{-1}\right]\Big|_{\tau=\tau_k^j} = \frac{2\omega_k^2 + m_{1k}^2 - 2m_{2k}}{m_3^2} = \frac{2\omega_k^2 + P_k}{m_3^2} > 0.$$
(69)

Therefore, the transversality condition holds.  $\Box$ 

**Theorem 9.** Suppose that (H7) and  $A_2 < 0$  hold. The following statements are valid:

- (1) System (9) is locally asymptotically stable at  $E^*$  when  $\tau \in [0, \tau_0^0]$  and unstable for  $\tau > \tau_0^0$ .
- (2) System (9) undergoes a Hopf bifurcations at  $E^*$  when  $\tau = \tau_0^0$ .

#### 4. Direction and Stability of Hopf Bifurcation

The previous analysis has shown that system (9) admits a series of periodic solutions bifurcating from the trivial uniform steady state  $E^*$  at some critical values. Then, in this section, we are concerned with the direction of Hopf bifurcations and the stability of bifurcating periodic solution.

Denote  $\tau_k^j$  ( $j \in N_0, 0 \le k \le k^*$ ) by  $\tilde{\tau}$  and introduce the new parameter  $\sigma = \tau - \tilde{\tau}$ . Normalizing the delay  $\tau$  by the timing-scaling  $t \longrightarrow t/\tau$ , then system (9) can be transformed in the phase space  $\vartheta^* = C([-1, 0], X)$ 

Complexity

$$\dot{V} = \tilde{\tau} D \Delta V + L(\tilde{\tau}) (V_t) + G(V_t, \sigma),$$
(70)

where  $V = (u(\cdot, t))v(\cdot, t)^T$ , L:  $\vartheta^* \longrightarrow \mathbb{R}^2$  and G:  $\vartheta^* \times \mathbb{R}^+$  $\longrightarrow \mathbb{R}^2$  are given, respectively, by

$$L(\tilde{\tau})(\zeta) = \tilde{\tau} \begin{pmatrix} a_{11}\zeta_1(0) + a_{12}\zeta_2(0) \\ a_{21}\zeta_1(-1) + a_{22}\zeta_2(0) \end{pmatrix},$$
(71)

$$G(\zeta, \sigma) = \sigma D\Delta \zeta + L(\sigma)(\zeta) + g(\zeta, \sigma), \tag{72}$$

with 
$$g(\zeta, \sigma) = (\tilde{\tau} + \sigma)G_1(\zeta, \sigma)G_2(\zeta, \sigma)^T$$
, and

$$G_{1}(\zeta,\sigma) = \frac{(\zeta_{1}(0) + u^{*})(1 - \zeta_{1}(0) - u^{*})}{1 + \alpha(\zeta_{1}(0) + u^{*})} - (\zeta_{1}(0) + u^{*})(\zeta_{2}(0) + v^{*}) - \frac{h(\zeta_{1}(0) + u^{*})}{1 + fE} - a_{11}\zeta_{1}(0) - a_{12}\zeta_{2}(0),$$

$$G_{2}(\zeta,\sigma) = \theta(\zeta_{2}(0) + v^{*}) - \frac{b(\zeta_{2}(0) + v^{*})^{2}}{p + \zeta_{1}(-1) + u^{*}} - a_{21}\zeta_{1}(-1) - a_{22}\zeta_{2}(0),$$
(73)

for  $\zeta = (\zeta_1 \zeta_2)^T \in \vartheta^*$ .

From the previous discussions, we know that  $\Lambda_k = \{i\omega_k \tilde{\tau}, -i\omega_k \tilde{\tau}\}$  are a pair of simple purely imaginary eigenvalues of the linear equation

$$V(t) = -\tilde{\tau} D\Delta V(t) + L(\tilde{\tau}) (V_t).$$
(74)

By the Riesz representation theorem, there exists a  $2 \times 2$  matrix function  $\eta(\delta, \tau, k)(-1 \le \delta \le 0)$  such that

$$-\tilde{\tau}Dk^{2}\zeta(0) + L(\tilde{\tau})(\zeta) = \int_{-1}^{0} [d\eta(\delta,\tilde{\tau},k)]\zeta(\delta).$$
(75)

Actually, we can take

$$\eta(\delta,\tilde{\tau},k) = \begin{cases} \tilde{\tau} \begin{pmatrix} a_{11} - D_1 k^2 & a_{12} \\ 0 & a_{22} - D_2 k^2 \end{pmatrix}, \, \delta = 0, \\ 0, \, \delta \in (-1,0), \\ -\tilde{\tau} \begin{pmatrix} 0 & 0 \\ a_{21} & 0 \end{pmatrix}, \, \delta = -1. \end{cases}$$
(76)

Let  $Q(\tilde{\tau})$  denote the infinitesimal generator of the semigroup and  $Q^*$  be the formal adjoint of  $Q(\tilde{\tau})$  under the bilinear pairing

$$\begin{aligned} (\varsigma(\delta),\xi(s)) &= \varsigma(0)\xi(0) - \int_{-1}^{0} \int_{\varepsilon=0}^{s} \varsigma(\varepsilon-s)d\eta(s)\xi(\varepsilon)\,d\varepsilon \\ &= \varsigma(0)\xi(0) + \tilde{\tau} \int_{-1}^{0} \varsigma(s+1) \begin{pmatrix} 0 & 0 \\ a_{21} & 0 \end{pmatrix} \xi(s)\,ds\,, \end{aligned}$$
(77)

for  $\xi \in C$ ,  $\varsigma \in C^* = C([0, 1], R^2)$ . Then, we can give the definition of  $Q(\tilde{\tau})$  is

$$Q(\tilde{\tau}) = \tilde{\tau} \begin{pmatrix} a_{11} - D_1 k^2 & a_{12} \\ a_{21} e^{-\lambda} & a_{22} - D_2 k^2 \end{pmatrix}.$$
 (78)

From the above analysis, we get that  $Q(\tilde{\tau})$  has a pair of simple purely imaginary eigenvalues  $\pm i\omega_k \tilde{\tau}$ . Supposing that  $p_1(\delta) = e^{i\omega_k \tau \delta} (1M)^T, q_1(s) = (1N)e^{-i\omega_k \tau s}$ , then we can calculate the eigenfunction  $p_1(\delta)$  of  $Q(\tilde{\tau})$  corresponding to  $i\omega_k \tilde{\tau}$  and the eigenfunction  $q_1(s)$  of  $Q^*$  corresponding to  $-i\omega_k \tilde{\tau}$ . From (78), we derive

$$(\lambda I - Q(\tilde{\tau}))p_1(0) = 0|_{\lambda = i\omega_k \tilde{\tau}},$$
(79)

$$(\lambda I - \mathbf{Q}^*) q_1 (\mathbf{0})^T = \mathbf{0}|_{\lambda = -i\omega_k \widetilde{\tau}^*}$$
(80)

Then, we obtain

$$M = \frac{i\omega_{k} + D_{1}k^{2} - a_{11}}{a_{12}},$$

$$N = -\frac{a_{12}}{a_{22} + i\omega_{k} - D_{2}k^{2}}.$$
Let  $\varsigma = (\varsigma_{1}, \varsigma_{2}), \xi = (\xi_{1}, \xi_{2})^{T}$ , where
$$\varsigma_{1}(\delta) = \frac{p_{1}(\delta) + \overline{p_{1}(\delta)}}{2}$$

$$= \begin{pmatrix} Re\left(e^{i\omega_{k}\tau\delta}\right) \\ Re\left(Me^{i\omega_{k}\tau\delta}\right) \end{pmatrix},$$

$$\varsigma_{2}(\delta) = \frac{p_{1}(\delta) - \overline{p_{1}(\delta)}}{2i}$$

$$= \begin{pmatrix} Im\left(e^{i\omega_{k}\tau\delta}\right) \\ Im\left(Me^{i\omega_{k}\tau\delta}\right) \end{pmatrix},$$
(82)

for  $\delta \in (-1, 0)$ , and

$$\xi_{1}(s) = \frac{q_{1}(s) + \overline{q_{1}(s)}}{2}$$

$$= \begin{pmatrix} Re(e^{-i\omega_{k}\tau^{*}s}) \\ Re(Ne^{-i\omega_{k}\delta\tau s}) \end{pmatrix},$$

$$\xi_{2}(s) = \frac{q_{1}(s) - \overline{q_{1}(s)}}{2i}$$

$$= \begin{pmatrix} Im(e^{-i\omega_{k}\tau^{*}s}) \\ Im(Ne^{-i\omega_{k}\tilde{\tau}s}) \end{pmatrix},$$
(83)

for  $s \in (0, 1)$ . Let  $a_{1k} = a_{11} - D_1 k^2$ ,  $a_{2k} = a_{22} - D_2 k^2$ , then by (77), we can calculate

$$\begin{split} &(\xi_{1},\varsigma_{1}) = 1 + \frac{a_{1k}a_{2k}}{a_{2k}^{2} + \omega_{k}^{2}} + \frac{\tilde{\tau}B_{3}a_{21}\sin(\omega_{k}\tilde{\tau})}{2} - \tilde{\tau}B_{4}a_{21}\left(\frac{\cos(\omega_{k}\tilde{\tau})}{2} + \frac{\sin(\omega_{k}\tilde{\tau})}{2\omega_{k}\tilde{\tau}}\right), \\ &(\xi_{1},\varsigma_{2}) = -\frac{a_{2k}\omega_{k}}{a_{2k}^{2} + \omega_{k}^{2}} + \tilde{\tau}B_{3}a_{21}\left(\frac{\cos(\omega_{k}\tilde{\tau})}{2} - \frac{\sin(\omega_{k}\tau^{*})}{2\omega_{k}\tau^{*}}\right) + \frac{\tilde{\tau}B_{4}a_{21}\sin(\omega_{k}\tilde{\tau})}{2}, \\ &(\xi_{1},\varsigma_{1}) = -\frac{a_{1k}\omega_{k}}{a_{2k}^{2} + \omega_{k}^{2}} + \tilde{\tau}B_{3}a_{21}\left(\frac{\cos(\omega_{k}\tilde{\tau})}{2} + \frac{\sin(\omega_{k}\tilde{\tau})}{2\omega_{k}\tilde{\tau}}\right) + \frac{\tilde{\tau}B_{4}a_{21}\sin(\omega_{k}\tilde{\tau})}{2}, \\ &(\xi_{1},\varsigma_{2}) = \frac{\omega_{k}^{2}}{a_{2k}^{2} + \omega_{k}^{2}} - \frac{\tilde{\tau}B_{3}a_{21}\sin(\omega_{k}\tilde{\tau})}{2} + \tilde{\tau}B_{4}a_{21}\left(\frac{\cos(\omega_{k}\tilde{\tau})}{2} - \frac{\sin(\omega_{k}\tilde{\tau})}{2\omega_{k}\tilde{\tau}}\right), \end{split}$$

where

$$B_{3} = \frac{a_{12}\omega_{k}}{a_{2k}^{2} + \omega_{k}^{2}},$$

$$B_{4} = \frac{a_{12}a_{2k}}{a_{2k}^{2} + \omega_{k}^{2}}.$$
(85)

Next, we define  $(\xi, \varsigma) = (\xi_i, \varsigma_j) (i, j = 1, 2)$  and construct a new basis

$$\Phi = \left(\Phi_1, \Phi_2\right)^T = \left(\xi, \varsigma\right)^{-1} \xi.$$
(86)

Then, we obtain  $(\Phi, \varsigma) = I_2$ . Denote  $f_k = (\beta_k^1, \beta_k^2)$  and  $a \cdot f_k = a_1 \beta_k^1 + a_2 \beta_k^2$ , where  $a = (a_1, a_2)^T \in \vartheta^*$ . Thus, the center space of the (74) is given by  $P_{CN\vartheta^*}$  having the following form:

$$P_{CN\psi} = \varsigma \left( \Phi, \langle \psi, f_k \rangle \right) \cdot f_k, \ \psi \in \vartheta^*, \tag{87}$$

where

$$\langle \psi, f_k \rangle = \left( \langle \psi, \beta_k^1 \rangle, \langle \psi, \beta_k^2 \rangle \right)^T,$$

$$\langle a, b \rangle = \frac{\int_0^{\pi} \left( a_1 \overline{b_1} + a_2 \overline{b_2} \right) dx}{\pi},$$
(88)
for  $a = \left( a_1 \ a_2 \right)^T, b = \left( b_1 \ b_2 \right)^T.$ 

Let  $Q(\tilde{\tau})$  be the infinitesimal generator induced by the solution of (74), then system (9) can be rewritten by

$$\dot{V}(t) = Q(\tilde{\tau})V_t + Y_0 G(V_t, \sigma),$$
(89)

where

$$Y_{0}(\delta) = \begin{cases} 0, -1 \le \delta < 0, \\ I, \ \delta = 0. \end{cases}$$
(90)

From the decomposition  $\vartheta^* = P_{CN\vartheta^*} \oplus P_{S\vartheta^*}$  and (38), where  $P_{S\vartheta^*}$  denotes the complement subspace of  $P_{CN\vartheta^*}$  in  $\vartheta^*$ , the solutions of system (39) can be written as

$$V(t) = \varsigma \left(\frac{\gamma_1}{\gamma_2}\right) \cdot f_k + h(\gamma_1, \gamma_2, \sigma), \tag{91}$$

where

$$\begin{pmatrix} \gamma_1 \\ \gamma_2 \end{pmatrix} = (\Phi, \langle V_t, f_k \rangle).$$
(92)

The bifurcation direction of this part is based on  $\sigma = 0$ , so the solution of system (31) on the center manifold is given by

$$V_t^* = \varsigma \left(\frac{\gamma_1}{\gamma_2}\right) \cdot f_k + h(\gamma_1, \gamma_2, 0).$$
(93)

Let  $x = \gamma_1 - i\gamma_2$  and notice  $p_1 = \varsigma_1 + i\varsigma_2$ , then (93) becomes

$$V_{t}^{*} = \frac{1}{2} \varsigma \begin{pmatrix} x + \overline{x} \\ i(x - \overline{x}) \end{pmatrix} \cdot f_{k} + W(x, \overline{x})$$

$$= \frac{1}{2} \left( p_{1}x + \overline{p}_{1}\overline{x} \right) \cdot f_{k} + W(x, \overline{x}),$$
(94)

where

$$W(x,\overline{x}) = h\left(\frac{x+\overline{x}}{2}, -\frac{x-\overline{x}}{2i}, 0\right).$$
(95)

By [40], it is easy to know x satisfies

$$\dot{x} = i\omega_k \tilde{\tau} x + g(x, \bar{x}) \tag{96}$$

with

$$g(x, \overline{x}) = (\Phi_1(0) - i\Phi_2(0)) < G(V_t^*, 0), \ f_k >.$$
(97)

By Taylor expansion of  $W(x, \overline{x})$  and  $g(x, \overline{x})$ , we get

$$W(x, \overline{x}) = w_{20} \frac{x^2}{2} + w_{11} x \overline{x} + w_{02} \frac{\overline{x}^2}{2} + \cdots,$$

$$g(x, \overline{x}) = g_{20} \frac{x^2}{2} + g_{11} x \overline{x} + g_{02} \frac{\overline{x}^2}{2} + g_{21} \frac{x^2 \overline{x}}{2} + \cdots.$$
(98)

From (94) and (98), we obtain

$$u_{t}(0) = \frac{1}{2} (x, \overline{x})\iota_{k} + w_{20}^{(1)}(0) \frac{x^{2}}{2} + w_{11}^{(1)}(0)x\overline{x} + w_{02}^{(1)}(0)\frac{\overline{x}^{2}}{2} + \cdots,$$

$$v_{t}(0) = \frac{1}{2} (Mx + \overline{M}\overline{x})\iota_{k} + w_{20}^{(2)}(0)\frac{x^{2}}{2} + w_{11}^{(2)}(0)x\overline{x} + w_{02}^{(2)}(0)\frac{\overline{x}^{2}}{2} + \cdots,$$

$$u_{t}(-1) = \frac{1}{2} \left( xe^{-i\omega_{k}\overline{\tau}} + \overline{x}e^{-i\omega_{k}\overline{\tau}} \right)\iota_{k} + w_{20}^{(1)}(-1)\frac{x^{2}}{2} + w_{11}^{(1)}(-1)x\overline{x} + w_{02}^{(1)}(-1)\frac{\overline{x}^{2}}{2} + \cdots,$$

$$(99)$$

$$v_{t}(-1) = \frac{1}{2} \left( Mxe^{-i\omega_{k}\overline{\tau}} + M\overline{x}e^{-i\omega_{k}\overline{\tau}} \right)\iota_{k} + w_{20}^{(2)}(-1)\frac{x^{2}}{2} + w_{11}^{(2)}(-1)x\overline{x} + w_{02}^{(2)}(-1)\frac{\overline{x}^{2}}{2} + \cdots,$$

$$(1) = \frac{1}{2} \left( Mxe^{-i\omega_{k}\overline{\tau}} + M\overline{x}e^{-i\omega_{k}\overline{\tau}} \right)\iota_{k} + w_{20}^{(2)}(-1)\frac{x^{2}}{2} + w_{11}^{(2)}(-1)x\overline{x} + w_{02}^{(2)}(-1)\frac{\overline{x}^{2}}{2} + \cdots,$$

$$\overline{G}_{1}(V_{t}^{*},0) = \frac{\overline{G}_{1}}{\overline{\tau}}$$

$$= \begin{bmatrix} \frac{f_{uu}}{2}u_{t}^{2}(0) + f_{uv}u_{t}(0)v_{t}(0) + \frac{f_{vv}}{2}v_{t}^{2}(0) + \frac{f_{uuu}}{6}u_{t}^{3}(0) \\ + \frac{f_{uuv}}{2}u_{t}^{2}(0)v_{t}(0) + \frac{f_{uvv}}{2}v_{t}^{2}(0)u_{t}(0) + \frac{f_{vvv}}{6}v_{t}^{3}(0) + \cdots \end{bmatrix},$$

$$\overline{G}_{2}(V_{t}^{*},0) = \frac{\overline{G}_{2}}{\overline{\tau}}$$

$$= \begin{bmatrix} \frac{g_{uu}}{2}u_{t}^{2}(-1) + g_{uv}u_{t}(-1)v_{t}(0) + \frac{g_{vv}}{2}v_{t}^{2}(0) + \frac{g_{uuu}}{6}u_{t}^{3}(-1) \\ + \frac{g_{uuv}}{2}u_{t}^{2}(-1)v_{t}(0) + \frac{g_{uvv}}{2}v_{t}^{2}(0)u_{t}(-1) + \frac{g_{vvv}}{6}v_{t}^{3}(0) + \cdots \end{bmatrix},$$
(100)

where

Complexity

$$f_{uu} = \frac{2 + 2\alpha}{(1 + \alpha u^{*})^{3}},$$

$$f_{uuu} = \frac{6\alpha + 6\alpha^{2}}{(1 + \alpha u^{*})^{4}},$$

$$f_{uv} = -1,$$

$$f_{vv} = f_{vvv} = f_{uvv} = f_{uuv} = 0,$$

$$g_{uu} = \frac{2b((v^{*})^{2}}{(p + u^{*})^{3}},$$

$$g_{uv} = \frac{2bv^{*}}{(p + u^{*})^{2}},$$

$$g_{uuu} = \frac{6b(v^{*})^{2}}{(p + u^{*})^{4}},$$

$$g_{uuv} = \frac{4bv^{*}}{(p + u^{*})^{3}},$$

$$g_{uvv} = \frac{2b}{(p + u^{*})^{2}},$$

$$g_{uvv} = \frac{2b}{(p + u^{*})^{2}},$$

$$g_{uvv} = 0.$$
(101)

Thus, we can get

$$\overline{G}_{1}(V_{t}^{*},0) = \left(\frac{x^{2}}{2}\Psi_{20} + x\overline{x}\Psi_{11} + \frac{\overline{x}^{2}}{2}\overline{\Psi}_{20}\right) l_{k}^{2} + \frac{x^{2}\overline{x}}{2} \left(\Psi_{1}l_{k} + \Psi_{2}l_{k}^{3}\right) \dots,$$

$$\overline{G}_{2}(V_{t}^{*},0) = \left(\frac{x^{2}}{2}\nu_{20} + x\overline{x}\nu_{11} + \frac{\overline{x}^{2}}{2}\overline{\nu}_{20}\right) l_{k}^{2} + \frac{x^{2}\overline{x}}{2} \left(\nu_{1}l_{k} + \nu_{2}l_{k}^{3}\right) \dots,$$
(102)

$$\langle G(V_t^*, 0), f_k \rangle = \tilde{\tau} \langle \overline{G}(V_t^*, 0), f_k \rangle$$

$$= \tilde{\tau} (\overline{G}_1(V_t^*, 0)\beta_k^1 + \overline{G}_2(V_t^*, 0)\beta_k^2)$$

$$= \frac{x^2}{2} \tilde{\tau} \begin{pmatrix} \Psi_{20} \\ v_{20} \end{pmatrix} \Gamma + x\overline{x}\tilde{\tau} \begin{pmatrix} \Psi_{11} \\ v_{11} \end{pmatrix} \Gamma$$

$$+ \frac{\overline{x}^2}{2} \tilde{\tau} \begin{pmatrix} \overline{\Psi}_{20} \\ \overline{v}_{20} \end{pmatrix} \Gamma + \frac{x^2\overline{x}}{2} \tilde{\tau} \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} + \cdots,$$

$$(103)$$

with

where

$$\mu_{1} = \frac{1}{\pi} \int_{0}^{\pi} \left( \Psi_{1} \iota_{k}^{2}(x) + \Psi_{2} \iota_{k}^{4}(x) \right) dx, \qquad (104)$$
$$\mu_{2} = \frac{1}{\pi} \int_{0}^{\pi} \left( \nu_{1} \iota_{k}^{2}(x) + \nu_{2} \iota_{k}^{4}(x) \right) dx,$$

 $\Gamma = \frac{1}{\pi} \int_0^{\pi} \iota_k^3(x) dx,$ 

$$\begin{split} \Psi_{20} &= \frac{1}{4} \Big( f_{uu} + 2M f_{uv} + M^2 f_{vv} \Big), \\ \Psi_{11} &= \frac{1}{4} \Big( f_{uu} + M f_{uv} + \overline{M} f_{uv} + M \overline{M} f_{vv} \Big), \\ \Psi_{1} &= w_{20}^{(1)}(0) \bigg( \frac{f_{uu} + \overline{M} f_{uv}}{2} \bigg) + w_{11}^{(1)}(0) \big( f_{uu} + M f_{uv} \big) \\ &+ w_{20}^{(2)}(0) \bigg( \frac{f_{uv} + \overline{M} f_{vv}}{2} \bigg) + w_{11}^{(2)}(0) \big( f_{uv} + M f_{vv} \big), \end{split}$$
(105)  
$$\Psi_{2} &= \frac{1}{8} f_{uuu} + (2M + \overline{M}) f_{uuv} + \big( M^2 + 2M \overline{M} \big) f_{uvv} + M^2 \overline{M} f_{vvv}, \end{split}$$

and,

$$\begin{split} \nu_{20} &= \frac{1}{4} \left( g_{uu} e^{-2i\omega_{k}\widetilde{\tau}} + 2Mg_{uv} e^{-i\omega_{k}\widetilde{\tau}} + M^{2}g_{vv} \right), \\ \nu_{11} &= \frac{1}{4} \left( g_{uu} + \left( Me^{i\omega_{k}\widetilde{\tau}} + \overline{M}e^{-i\omega_{k}\widetilde{\tau}} \right) g_{uv} + M\overline{M}g_{vv} \right), \\ \nu_{1} &= w_{20}^{(1)} (-1) \left( \frac{g_{uu}e^{i\omega_{k}\widetilde{\tau}} + \overline{M}g_{uv}}{2} \right) + w_{11}^{(1)} (-1) \left( g_{uu}e^{-i\omega_{k}\widetilde{\tau}} + Mg_{uv} \right) \\ &+ w_{20}^{(2)} (0) \left( \frac{e^{i\omega_{k}\widetilde{\tau}}g_{uv} + \overline{M}g_{vv}}{2} \right) + w_{11}^{(2)} (0) \left( g_{uv}e^{-i\omega_{k}\widetilde{\tau}} + Mg_{vv} \right), \\ \nu_{2} &= \frac{1}{8}g_{uuu}e^{-i\omega_{k}\widetilde{\tau}} + \frac{1}{8} \left( 2M + \overline{M}e^{-2i\omega_{k}\widetilde{\tau}} \right) g_{uuv} + \frac{1}{8}M^{2}\overline{M}g_{vvv} \\ &+ \frac{1}{8} \left( M^{2}e^{i\omega_{k}\widetilde{\tau}} + 2M\overline{M}e^{-i\omega_{k}\widetilde{\tau}} \right) g_{uvv}. \end{split}$$
(106)

Denote

$$\Phi_1(0) - i\Phi_2(0) = (\chi_1\chi_2), \tag{107}$$

and notice that

$$\frac{1}{\pi} \int_{0}^{\pi} t_{k}^{3}(x) dx = \frac{1}{\pi} \int_{0}^{\pi} \cos^{3}(kx) dx$$

$$= 0, \ k = 1, 2, 3 \cdots$$
(108)

Then we deduce

 $= \frac{x^2}{2} \left( \Psi_{20} \chi_1 + \nu_{20} \chi_2 \right) \Gamma \tilde{\tau} + x \overline{x} \left( \Psi_{11} \chi_1 + \nu_{11} \chi_2 \right) \Gamma \tilde{\tau}$  $+ \frac{\overline{x}^2}{2} \left( \overline{\Psi}_{20} \chi_1 + \overline{\nu}_{20} \chi_2 \right) \Gamma \tilde{\tau} + \frac{x^2 \overline{x}}{2} \left( \mu_1 \chi_1 + \mu_2 \chi_2 \right) \tilde{\tau} + \cdots$ (109)

Thus, by equations (43), (45), and (46), we have  $g_{20} = g_{02} = g_{11} = 0$  for k = 1, 2, ... If k = 0, we get

$$g_{20} = (\Psi_{20}\chi_1 + \nu_{20}\chi_2)\tilde{\tau}, g_{11} = (\Psi_{11}\chi_1 + \nu_{11}\chi_2)\tilde{\tau},$$
(110)  
$$g_{02} = (\overline{\Psi}_{20}\chi_1 + \overline{\nu}_{20}\chi_2)\tilde{\tau}.$$

$$g_{21} = (\mu_1 \chi_1 + \mu_2 \chi_2) \tilde{\tau}.$$
 (111)

So far, we have given the expressions of  $g_{20}, g_{11}, g_{02}$ . It follows from (98) that

Also, for  $k = N_0$ , we can get

$$W(x, \bar{x}) = w_{20}x\dot{x} + w_{11}(z \,\bar{x} + x\bar{x}) + w_{02}\bar{x}\bar{x} + \cdots,$$
(112)

and,

$$Q(\tilde{\tau})W(x,\overline{x}) = Q(\tilde{\tau})w_{20}\frac{x^2}{2} + Q(\tilde{\tau})w_{11}x\overline{x} + Q(\tilde{\tau})w_{02}\frac{\overline{x}^2}{2} + \cdots$$
(113)

By [40], we know  $W(x, \overline{x})$  satisfies

$$W = Q(\tilde{\tau})W + Z(x, \overline{x}), \tag{114}$$

where

$$Z(x, \overline{x}) = Z_{20} \frac{x^2}{2} + Z_{11} x \overline{x} + Z_{02} \frac{\overline{x}^2}{2} + \cdots$$
  
=  $Y_0(\delta) G(V_t^*, 0) - \varsigma(\Phi, \langle Y_0(\delta) G(V_t^*, 0), f_k \rangle) \cdot f_k.$  (115)

Substituting (96) into the derivative of (98) and comparing the coefficients with (114) and (115), we have

$$[2i\omega_k \tilde{\tau} - \mathbf{Q}(\tilde{\tau})] w_{20}(\delta) = \mathbf{Z}_{20}(\delta), \qquad (116)$$

$$Q(\tilde{\tau})w_{11}(\delta) = -Z_{11}(\delta).$$
(117)

From (115), we have that for  $\theta \in [-1, 0)$ ,  $Z(x, \overline{x}, \delta) = -\zeta(\delta)\Phi(0) < G(V_t^*, 0), f_k > \cdot f_k$ 

$$= -\left(\frac{p_1(\delta) + \overline{p_1(\delta)}}{2}, \frac{p_1(\delta) - \overline{p_1(\delta)}}{2i}\right) \left(\Phi_1(0), \Phi_2(0)\right)^T$$
$$\cdot < G\left(V_t^*, 0\right), f_k > \cdot f_k$$
$$= -\frac{1}{2} \left[g_{20}p_1(\delta) + \overline{g}_{02}\overline{p_1(\delta)}\right] \frac{x^2}{2} \cdot f_k$$
$$-\frac{1}{2} \left[g_{11}p_1(\delta) + \overline{g}_{11}\overline{p_1(\delta)}\right] x \overline{x} \cdot f_k + \cdots.$$
(118)

Hence, for 
$$\delta \in [-1, 0)$$
,

 $Z_{20}(\delta) = \begin{cases} 0, & k \in N, \\ -\frac{1}{2} \left[ g_{20} p_1(\delta) + \overline{g}_{02} \overline{p_1(\delta)} \right] \cdot f_k, & k = 0, \\ 0, & k \in N, \end{cases}$ (119)  $Z_{20}(\delta) = \begin{cases} 0, & k \in N, \\ -\frac{1}{2} \left[ g_{20} p_1(\delta) + \overline{g}_{02} \overline{p_1(\delta)} \right] \cdot f_k, & k \in N, \end{cases}$ 

$$\begin{bmatrix} -\frac{1}{2} \left[ g_{20} p_1(\delta) + \overline{g}_{11} \overline{p_1(\delta)} \right] \cdot f_k, \quad k = 0, \\ Z(x, \overline{x})(0) = G(V_t^*, 0) - \varsigma(\Phi, \langle G(V_t^*, 0), f_k \rangle) \cdot f_k, \quad (120) \end{bmatrix}$$

where

$$Z_{20}(0) = \begin{cases} \tilde{\tau} \begin{pmatrix} \Psi_{20} \\ \nu_{20} \end{pmatrix} \cos^2 kxk \in N, \\ \tilde{\tau} \begin{pmatrix} \Psi_{20} \\ \nu_{20} \end{pmatrix} - \frac{1}{2} \left[ g_{20} p_1(0) + \overline{g}_{02} \overline{p_1(0)} \right] \cdot f_k k = 0, \end{cases}$$

$$Z_{11}(0) = \begin{cases} \tilde{\tau} \begin{pmatrix} \Psi_{11} \\ \nu_{11} \end{pmatrix} \cos^2 kxk \in N, \\ \tilde{\tau} \begin{pmatrix} \Psi_{11} \\ \nu_{11} \end{pmatrix} - \frac{1}{2} \left[ g_{11} p_1(0) + \overline{g}_{11} \overline{p_1(0)} \right] \cdot f_k k = 0. \end{cases}$$
(121)

Thus, we can get from (78) and (116) that

$$\dot{w}_{20} (\delta) = 2i\omega_k \tilde{\tau} w_{20} (\delta) + \frac{1}{2} \left[ g_{20} p_1 (\delta) + \overline{g}_{02} \overline{p_1 (\delta)} \right] \cdot f_k,$$
  
$$\dot{w}_{11} (\delta) = -\frac{i}{2\omega_k \tilde{\tau}} \left[ g_{11} p_1 (\delta) + \overline{g}_{11} \overline{p_1 (\delta)} \right] f_k,$$
(122)

for  $-1 \le \delta < 0$ . That is,

$$w_{20}(\delta) = \frac{i}{2\omega_k \tilde{\tau}} \left[ g_{20} p_1(\delta) + \frac{\overline{g}_{02}}{3} \overline{p_1(\delta)} \right] \cdot f_k + E_1 e^{2i\omega_k \tilde{\tau} \delta},$$
  
$$w_{11}(\delta) = \frac{i}{2\omega_k \tilde{\tau}} \left[ \overline{g}_{11} \overline{p_1(\delta)} - g_{11} p_1(\delta) \right] \cdot f_k + E_2,$$
  
(123)

where  $-1 \le \delta < 0$  and  $E_1, E_2 \in R^2$  are constant vectors. From the above discussion and (116), we have

$$\begin{cases} 2i\omega_k E_1 - Q(\tilde{\tau}) - L(\tilde{\tau})E_1 e^{2i\omega_k \tilde{\tau}} = \tilde{\tau} \begin{pmatrix} \Psi_{20} \\ \nu_{20} \end{pmatrix} \cos^2 kx, \\ -Q(\tilde{\tau})E_2 - L(\tilde{\tau})E_2 = \tilde{\tau} \begin{pmatrix} \Psi_{11} \\ \nu_{11} \end{pmatrix} \cos^2 kx. \end{cases}$$
(124)

Then, we obtain

$$E_{1} = \tilde{\tau} \begin{pmatrix} 2i\omega_{k} + D_{1}k^{2} - a_{11} & -a_{12} \\ -a_{21}e^{-2i\omega_{k}\tilde{\tau}} & 2i\omega_{k} + D_{2}k^{2} - a_{22} \end{pmatrix}^{-1} \begin{pmatrix} \Psi_{20} \\ \nu_{20} \end{pmatrix} l_{k}^{2},$$

$$E_{2} = \tilde{\tau} \begin{pmatrix} D_{1}k^{2} - a_{11} & -a_{12} \\ -a_{21} & D_{2}k^{2} - a_{22} \end{pmatrix}^{-1} \begin{pmatrix} \Psi_{11} \\ \nu_{11} \end{pmatrix} l_{k}^{2}.$$
(125)

Therefore, each  $g_{20}$ ,  $g_{11}$ ,  $g_{02}$ ,  $g_{21}$  can be determined and we can evaluate the following values determining the direction and stability of bifurcating periodic orbits:

$$\begin{cases} C_{1}(0) = \frac{i}{2\omega_{k}\tilde{\tau}} \left(g_{11}g_{20} - 2|g_{11}|^{2} - \frac{1}{3}|g_{02}|^{2}\right) + \frac{1}{2}g_{21}, \\ \kappa_{2} = -\frac{\operatorname{Re}[C_{1}(0)]}{\operatorname{Re}[\lambda'(\tilde{\tau})]}, \\ \beta_{2} = 2\operatorname{Re}[C_{1}(0)], \\ T_{2} = -\frac{\operatorname{Im}[C_{1}(0)] + \gamma_{2}\operatorname{Im}[\lambda'(\tilde{\tau})]}{\omega_{k}\tilde{\tau}}. \end{cases}$$
(126)

**Theorem 10.** For any critical value  $\tau_k^j$ , we have

- When κ<sub>2</sub> > 0 (κ<sub>2</sub> < 0), the Hopf bifurcation is supercritical (subcritical), that is, the bifurcating periodic solutions exist for τ > τ<sup>j</sup><sub>k</sub>(τ < τ<sup>j</sup><sub>k</sub>).
- (2) When  $\beta_2 > 0$  ( $\beta_2 < 0$ ), the bifurcating periodic solutions on the center manifold are orbitally asymptotically unstable (stable).
- (3) When  $T_2 > 0$  ( $T_2 < 0$ ), the period of bifurcating periodic solutions increases (decreases).

#### 5. Numerical Simulations

In the following, we will carry out numerical simulations of three impacts of diffusion, time delay, and harvesting effort to illustrate our theoretical findings in the previous sections.

5.1. The Impact of Diffusion. We confirm that system (13) has a unique equilibrium  $E^* = (0.1877, 0.5158)$  with the parameters  $\alpha = 0.6, b = 0.4, h = 0.375, f = 0.5, E = 0.8, p = 0.5, \theta = 0.3$ . By a simple verification, we can see that  $A_2 < 0$  is satisfied. By Theorem 1, we deduce that system (13) is locally asymptotically stable at  $E^*$ . Meanwhile, it is revealed from Theorem 2 that system (10) is still stable at  $E^*$ , which implies Turing instability will not occur in system (10) (see Figures 1–3).



FIGURE 1: The graph of each node and space x in system (10) when time t = 100 and  $D_1, D_2 > 0$ .



FIGURE 2: Spatiotemporal evolution behaviors of prey and predator of system (10) with  $D_1 = 0.2$ ,  $D_2 = 2$  and the initial condition  $u(0, x) = 0.187 + 0.1 \cos x$ ,  $v(0, x) = 0.515 + 0.3 \cos x$ .



FIGURE 3: Spatiotemporal evolution behaviors of prey and predator of system (10) with  $D_1 = 0.5$ ,  $D_2 = 2$  and the initial condition  $u(0, x) = 0.187 + 0.1 \cos x$ ,  $v(0, x) = 0.515 + 0.4 \cos x$ .

Next, we consider system (15) and set  $\alpha = 0.5$ ,  $e_1 = 0.2$ ,  $e_2 = 0.32$ , h = 0.3, f = 0.6, E = 0.6, p = 0.52, b = 0.15,  $\theta = 0.36$ . When  $D_1 = D_2 = 0$ , system (15) can be transformed into the nondiffusion system (18). By calculation, we get  $E^* = (0.1196, 1.5350)$ , and (H1), (H2) and  $Y_2 < 0$  are satisfied. Thus, from Lemma 5, system (18) is asymptotically stable at  $E^* (u^*, v^*) = (0.1196, 1.5350)$ . Then we set  $D_1 = 0.1$ ,  $D_1 = 0.8$  and other variables are the same as above.

Clearly, (H1) - (H3) and  $Y_2 < 0$  are reached. It follows from Theorem 4 that system (15) is still stable at  $E^*(u^*, v^*) =$ (0.1196, 1.5350) (see Figure 4). When  $D_1 = 0.02$ ,  $D_2 = 5$ and other variables are the same as above. By a simple calculation, we can see that  $Y_2 < 0$ , (H1), (H2), (H4), and (H5) are satisfied. By Theorem 5, the diffusion-driven instability occurs for system (15) at  $E^*(u^*, v^*)$ , which is shown in Figures 5 and 6.



FIGURE 4: Spatiotemporal evolution behaviors of prey and predator of system (15) with  $D_1 = 0.1, D_2 = 0.8$  and the initial condition  $u(0, x) = 0.119 + 0.1 \cos x, v(0, x) = 1.535 + 0.5 \cos x$ .



FIGURE 5: Spatiotemporal evolution behaviors of prey and predator of system (15) with  $D_1 = 0.02$ ,  $D_2 = 5$  and the initial condition  $u(0, x) = 0.119 + 0.1 \cos x$ ,  $v(0, x) = 1.535 + 0.5 \cos x$ .



FIGURE 6: Turing patterns of prey and predator when Turing instability occurs.



FIGURE 7: The evolution process of node u(t) and v(t) in system (9) when  $D_1 = 0.1, D_2 = 0.1, \tau = 20$ .



FIGURE 8: The evolution process of node u(t) and v(t) in system (9) when  $D_1 = 0.1, D_2 = 0.1, \tau = 50$ .



FIGURE 9: Phase diagram of system (9) at  $E^*(u^*, v^*) = (0.1549, 0.6823)$  when  $D_1 = 0.1, D_2 = 0.1$  and  $\tau = 23 < \tau_0^0 = 26.0037$ .



FIGURE 10: The evolution process of node u(t) and v(t) in system (9) when  $D_1 = 0.1$ ,  $D_2 = 0.1$  and  $\tau = 23 < \tau_0^0 = 26.0037$ .



FIGURE 11: Phase diagram of system (9) at  $E^*(u^*, v^*) = (0.1549, 0.6823)$  when  $D_1 = 0.1, D_2 = 0.1$  and  $\tau = 26.2 > \tau_0^0 = 26.0037$ .



FIGURE 12: The evolution process of node u(t) and v(t) in system (9) when  $D_1 = 0.1, D_2 = 0.1$  and  $\tau = 26.2 > \tau_0^0 = 26.0037$ .



FIGURE 13: Stability region exploring the dynamics of system (9) in  $(E, \tau_0^0)$  parameter space.

5.2. The Impact of Time Delay. For system (9), we take  $\alpha = 0.7, b = 0.4, h = 0.25, f = 0.5, E = 0.8, p = 0.45, \theta = 0.6, D_1 = 0.1, D_2 = 0.1$ . By calculation, we get  $E^*(u^*, v^*) = (0.05813, 0.7622)$ . In the meantime,  $A_2 < 0$  and (H6) are satisfied. By Theorem 6, system (9) is asymptotically stable at  $E^*(u^*, v^*) = (0.05813, 0.7622)$  for all  $\tau \ge 0$ , as displayed in Figures 7 and 8.

Next, we set  $\alpha = 0.6, b = 0.4, h = 0.134, f = 0.8, E = 1.5, p = 0.3, \theta = 0.6, D_1 = 0.1, D_2 = 0.1$  for system (9), which has a unique positive equilibrium  $E^*(u^*, v^*) = (0.1549, 0.6823)$ . It is verified that  $A_2 < 0$  and (*H*7) are satisfied. Meanwhile, we get  $\omega_0 = 0.0975, \tau_0^0 = 26.0037, \lambda'(\tau_0^0) = 0.021 - 0.0307i$ , implying that the cross-section condition holds. By Theorem 9, system (9) is locally asymptotically stable at  $E^*(u^*, v^*) = (0.1549, 0.6823)$  when  $\tau \in [0, \tau_0^0)$  (see Figures 9 and 10) and the Hopf bifurcation occurs when  $\tau$  crosses the critical value  $\tau_0^0$  (see Figures 11 and 12).

Furthermore, from formula (125), we obtain  $C_1(0) = -8.36984 - 2.3355i$ ,  $\kappa_2 = 398.56381 > 0$ ,  $\beta_2 = -16.73968 < 0$ ,  $T_2 = 2.3552 > 0$ . It follows from Theorem 10 that the direction of the bifurcation in system (9) at  $\tau_0^0$  is forward, the bifurcating period solutions are stable, and the period of bifurcating periodic solutions is increasing.

5.3. The Impact of Harvesting Effort. Considering the impact of harvesting efforts, we fix  $\alpha = 0.6, b = 0.4, h = 0.134, f = 0.8, p = 0.3, \theta = 0.6, D_1 = 0.1, D_2 = 0.1$  and let *E* vary in

[0.5, 5]. The stability and instability regions for system (9) are depicted by mapping the nonlinear harvesting *E* to the critical value  $\tau_0^0$  in Figure 13. Obviously, the critical value  $\tau_0^0$  of bifurcation increases with the increase of *E*, that is, the stable region of system (9) is expanded with the increase of *E*.

#### 6. Conclusion

We investigate the spatiotemporal dynamic evolution of a time-delay ecological competition system with food restriction and diffusion terms under Neumann boundary conditions. The conditions of asymptotic stability and Turing instability at the positive equilibrium of delay-free systems are obtained under different functional response functions. Compared with the classical population growth model described by the logistic equation, the model with food restriction studied in this paper is more in line with actual biological competition systems. The results show that the diffusion phenomenon caused by the change of population position in space will seriously affect the stability of biological competition systems, eventually resulting in the appearance of Turing instability. Then, by selecting the time delay as the bifurcation parameter, we reveal that the delay can cause very complex dynamic phenomena. When the delay is less than the bifurcation critical value, the system maintains asymptotic stability at the positive equilibrium point, while the system becomes unstable and produces a Hopf bifurcation when the delay is greater than the bifurcation critical value. Meanwhile, by using the central manifold method, we derive the conditions for determining the bifurcation direction and the stability of the bifurcation periodic solution. The harvesting effort has a major influence on the stability and Hopf bifurcation. As the harvesting effort increases in the appropriate range, the bifurcation critical value increases; that is, the stable region of the system is expanded. Therefore, we can conclude that the appropriate harvesting of biological populations and the rational development of biological resources can not only meet human economic needs but also play a beneficial role in biological systems.

#### **Data Availability**

The data in relation to the findings of this study are available upon request.

#### **Conflicts of Interest**

The authors declare that they have no conflicts of interest.

#### Acknowledgments

This work was supported by the National Natural Science Foundation of China (Nos. 62073172 and 61573194), the Open Research Project of the State Key Laboratory of Industrial Control Technology of Zhejiang University (No. ICT2022B43), and the Postgraduate Research and Practice Innovation Program of Jiangsu Province (No. KYCX21\_0754)

#### References

- H. Liu, Y. Ye, and Y. Wei, "Pattern formation in a reactiondiffusion predator-prey model with weak Allee effect and delay," *Complexity*, vol. 2019, Article ID 6282958, 2019.
- [2] T. Huang, H. Zhang, and X. Cong, "Exploring spatiotemporal complexity of a predator-prey system with migration and diffusion by a three-chain coupled map lattice," *Complexity*, vol. 2019, Article ID 3148323, 2019.
- [3] C. Arancibia-Ibarra and J. Flores, "Dynamics of a Leslie-Gower predator-prey model with Holling type II functional response, Allee effect and a generalist predator," *Mathematics and Computers in Simulation*, vol. 188, pp. 1–22, 2021.
- [4] B. Ghanbari, H. Günerhan, and H. M. Srivastava, "An application of the Atangana-Baleanu fractional derivative in mathematical biology: a three-species predator-prey model," *Chaos, Solitons & Fractals*, vol. 138, Article ID 109910, 2020.
- [5] S. Wang, H. Tang, and Z. Ma, "Hopf bifurcation of a multipledelayed predator-prey system with habitat complexity," *Mathematics and Computers in Simulation*, vol. 180, pp. 1–23, 2021.
- [6] P. Yang and Y. Wang, "Hopf-Zero bifurcation in an agedependent predator-prey system with Monod-Haldane functional response comprising strong Allee effect," *Journal of Differential Equations*, vol. 269, no. 11, pp. 9583–9618, 2020.
- [7] D. Luo, "Steady state for a predator-prey cross-diffusion system with the Beddington-DeAngelis and Tanner functional response, Bound," *Value Probl*, vol. 2021, no. 1, pp. 1–11, 2021.

- [8] X. Feng, C. Li, H. Sun, and Y. Wang, "Global bifurcation structure of a predator-prey system with a spatial degeneracy and B-D functional response," *Complexity*, vol. 2021, Article ID 9970255, 10 pages, 2021.
- [9] W. Liu, C. Fu, and B. Chen, "Hopf bifurcation for a predatorprey biological economic system with Holling type II functional response," *Journal of the Franklin Institute*, vol. 348, no. 6, pp. 1114–1127, 2011.
- [10] J. Datta, D. Jana, and R. K. Upadhyay, "Bifurcation and bioeconomic analysis of a prey-generalist predator model with Holling type IV functional response and nonlinear age-selective prey harvesting," *Chaos, Solitons & Fractals*, vol. 122, pp. 229–235, 2019.
- [11] D. Xiao, "Dynamics and bifurcations on a class of population model with seasonal constant-yield harvesting," *Discrete and Continuous Dynamical Systems - Series B*, vol. 21, no. 2, pp. 699–719, 2015.
- [12] Z. Shang, Y. Qiao, L. Duan, and J. Miao, "Bifurcation analysis in a predator-prey system with an increasing functional response and constant-yield prey harvesting," *Mathematics and Computers in Simulation*, vol. 190, pp. 976–1002, 2021.
- [13] M. G. Mortuja, M. K. Chaube, and S. Kumar, "Dynamic analysis of a predator-prey system with nonlinear prey harvesting and square root functional response," *Chaos, Solitons* & *Fractals*, vol. 148, Article ID 111071, 2021.
- [14] F. Souna, A. Lakmeche, and S. Djilali, "Spatiotemporal patterns in a diffusive predator-prey model with protection zone and predator harvesting," *Chaos, Solitons & Fractals*, vol. 140, Article ID 110180, 2020.
- [15] W. Q. Zuo, Z. P. Ma, and Z. B. Cheng, "Spatiotemporal dynamics induced by michaelis-menten type prey harvesting in a diffusive leslie-gower predator-prey model," *International Journal of Bifurcation and Chaos*, vol. 30, no. 14, Article ID 2050204, 2020.
- [16] R. M. May, J. R. Beddington, C. W. Clark, S. J. Holt, and R. M. Laws, "Management of multispecies fisheries," *Science*, vol. 205, no. 4403, pp. 267–277, 1979.
- [17] J. R. Beddington and R. M. May, "Maximum sustainable yields in systems subject to harvesting at more than one trophic level," *Mathematical Biosciences*, vol. 51, no. 3-4, pp. 261–281, 1980.
- [18] J. R. Beddington and J. G. Cooke, "Harvesting from a preypredator complex," *Ecological Modelling*, vol. 14, no. 3-4, pp. 155–177, 1982.
- [19] C. Zhu and K. Lan, "Phase portraits, Hopf bifurcations and limit cycles of Leslie-Gower predator-prey systems with harvesting rates," *Discrete & Continuous Dynamical Systems -B*, vol. 14, no. 1, pp. 289–306, 2010.
- [20] Y. J. Gong and J. C. Huang, "Bogdanov-Takens bifurcation in a Leslie-Gower predator-prey model with prey harvesting," *Acta Mathematicae Applicatae Sinica, English Series*, vol. 30, no. 1, pp. 239–244, 2014.
- [21] D. Hu and H. Cao, "Stability and bifurcation analysis in a predator-prey system with Michaelis-Menten type predator harvesting," *Nonlinear Analysis: Real World Applications*, vol. 33, pp. 58–82, 2017.
- [22] R. Wu, M. Chen, B. Liu, and L. Chen, "Hopf bifurcation and Turing instability in a predator-prey model with Michaelis-Menten functional response," *Nonlinear Dynamics*, vol. 91, no. 3, pp. 2033–2047, 2018.
- [23] X. Y. Meng and Y. Q. Wu, "Bifurcation analysis in a singular Beddington-DeAngelis predator-prey model with two delays and nonlinear predator harvesting," *Mathematical Biosciences* and Engineering, vol. 16, no. 4, pp. 2668–2696, 2019.

- [24] M. Li, B. Chen, and G. Wang, "Hopf bifurcation and nonlinear state feedback control for a modified lotka-volterra differential algebraic predator-prey system," in *Proceedings of the Fifth International Conference on Intelligent Control and Information Processing*, pp. 233–238, IEEE, Dalian, China, August 2014.
- [25] F. E. Smith, "Population dynamics in Daphnia magna and a new model for population growth," *Ecology*, vol. 44, no. 4, pp. 651–663, 1963.
- [26] L. Ma and D. Wei, "Hopf bifurcation of a delayed reactiondiffusion model with advection term," *Nonlinear Analysis*, vol. 212, Article ID 112455, 2021.
- [27] Q. Zhao, S. Liu, and X. Niu, "Turing instability and Hopf bifurcation for a diffusion-plankton system with cell size," *International Journal of Computer Mathematics*, vol. 98, no. 3, pp. 480–501, 2021.
- [28] S. Fu, X. He, L. Zhang, and Z. Wen, "Turing patterns and spatiotemporal patterns in a tritrophic food chain model with diffusion," *Nonlinear Analysis: Real World Applications*, vol. 59, Article ID 103260, 2021.
- [29] R. Yang and C. Zhang, "Dynamics in a diffusive modified Leslie-Gower predator-prey model with time delay and prey harvesting," *Nonlinear Dynamics*, vol. 87, no. 2, pp. 863–878, 2017.
- [30] A. M. Turing, "The chemical basis of morphogenesis," Bulletin of Mathematical Biology, vol. 52, no. 1-2, pp. 153–197, 1990.
- [31] A. Hunding and P. Graae Sørensen, "Size adaptation of Turing prepatterns," *Journal of Mathematical Biology*, vol. 26, no. 1, pp. 27–39, 1988.
- [32] R. J. Han and B. X. Dai, "Spatiotemporal pattern formation and selection induced by nonlinear cross-diffusion in a toxicphytoplankton-zooplankton model with Allee effect," *Nonlinear Analysis: Real World Applications*, vol. 45, pp. 822–853, 2019.
- [33] X. Jiang, X. Chen, T. Huang, and H. Yan, "Bifurcation and control for a predator-prey system with two delays," *IEEE Transactions on Circuits and Systems II: Express Briefs*, vol. 68, no. 1, pp. 376–380, 2021.
- [34] D. Hu, Y. Li, M. Liu, and Y. Bai, "Stability and Hopf bifurcation for a delayed predator-prey model with stage structure for prey and Ivlev-type functional response," *Nonlinear Dynamics*, vol. 99, no. 4, pp. 3323–3350, 2020.
- [35] A. Kumar and B. Dubey, "Dynamics of prey-predator model with strong and weak Allee effect in the prey with gestation delay," *Nonlinear Analysis Modelling and Control*, vol. 25, no. 3, pp. 417–442, 2020.
- [36] T. Wang, Y. Wang, and Z. Cheng, "Stability and Hopf bifurcation analysis of a general tri-diagonal BAM neural network with delays," *Neural Processing Letters*, vol. 53, no. 6, pp. 4571–4592, 2021.
- [37] G. Sun and B. Dai, "Stability and bifurcation of a delayed diffusive predator-prey system with food-limited and nonlinear harvesting," *Mathematical Biosciences and Engineering*, vol. 17, no. 4, pp. 3520–3552, 2020.
- [38] J. D. Murray, Mathematical Biology II: Spatial Models and Biomedical Applications, Springer, NY, USA, 2001.
- [39] B. B. Tao, M. Xiao, and W. X. Zheng, "Dynamics analysis and design for a bidirectional super-ring-shaped neural network with neurons and multiple delays," *IEEE Transactions on Neural Networks and Learning Systems*, vol. 99, pp. 1–15, 2020.
- [40] X. P. Yan, "Stability and Hopf bifurcation for a delayed preypredator system with diffusion effects," *Applied Mathematics* and Computation, vol. 192, no. 2, pp. 552–566, 2007.