

MBE, 20(11): 19584–19616. DOI: 10.3934/mbe.2023868 Received: 26 July 2023 Revised: 16 October 2023 Accepted: 16 October 2023 Published: 25 October 2023

http://www.aimspress.com/journal/mbe

Research article

More complex dynamics in a discrete prey-predator model with the Allee effect in prey

Mianjian Ruan¹, Xianyi Li^{2,*}and Bo Sun³

- ¹ Institute of Applied Mathematics, Zhejiang University of Water Resources and Electric Power, Hangzhou 310018, China
- ² Department of Big Data Science, School of Science, Zhejiang University of Science and Technology, Hangzhou 310023, China
- ³ School of Statistics and Mathematics, Central University of Finance and Economics, Beijing 102206, China
- * Correspondence: Email: mathxyli@zust.edu.cn.

Abstract: In this paper, we revisit a discrete prey-predator model with the Allee effect in prey to find its more complex dynamical properties. After pointing out and correcting those known errors for the local stability of the unique positive fixed point E_* , unlike previous studies in which the author only considered the codim 1 Neimark-Sacker bifurcation at the fixed point E_* , we focus on deriving many new bifurcation results, namely, the codim 1 transcritical bifurcation at the trivial fixed point E_1 , the codim 1 transcritical and period-doubling bifurcations at the boundary fixed point E_2 , the codim 1 period-doubling bifurcation and the codim 2 1:2 resonance bifurcation at the positive fixed point E_* . The obtained theoretical results are also further illustrated via numerical simulations. Some new dynamics are numerically found. Our new results clearly demonstrate that the occurrence of 1:2 resonance bifurcation confirms that this system is strongly unstable, indicating that the predator and the prey will increase rapidly and breakout suddenly.

Keywords: prey-predator model; Allee effect; transcritical bifurcation; period-doubling bifurcation; 1:2 resonance bifurcation

1. Introduction

Prey-predator interaction among populations is a kind of universal phenomena, which may lead to a stable coexistence and population cycles. Some population problems, such as the persistence, the stability of steady states, the bifurcation, etc. have been research focal points. How to deal with these problems is the main question. An effective method is to adopt reasonable assumptions and establish appropriate mathematical models to analyze such problems. So, mathematical modeling is a tool to deal with these problems, as it can give some insight into the corresponding model dynamics. Among the mathematical models, the Lotka-Volterra model is a famous prey-predator one, and it was independently constructed by Lotka [1] and Volterra [2], respectively. These mathematical models have two different kinds of types: continuous-time models and discrete-time models. Over the last several decades, an increasing number of studies have investigated discrete-time prey-predator models. There are two reasons for this. One reason is that discrete models are sometimes more appropriate as a tool to describe the evolution rule for the population than continuous models when the population size is small or when the generation of the population does not overlap. The other reason is that some discrete models have more diverse dynamical behaviors than the corresponding continuous models. For example, it is well known that the single-species discrete-time logistic model has very complex dynamical behavior, ranging from period-doubling bifurcation to chaotic behavior [3], whereas the solutions of the corresponding continuous model are monotonic.

To date, the investigations of bifurcation problems for discrete prey-predator systems have already received much attention, especially after Smith [4] considered the Lotka–Volterra type discrete-time prey-predator model. For this, refer to [5–11].

Focusing on the deeper research, people have started to consider an important ecological mechanism that describes realistic interactions among species in the ecosystem. For the readers' convenience, we present some existing works [12-16]. Li et al. considered the interactions of time delay and spatial diffusion to induce the periodic oscillation of a vegetation system [12]. Chowdhury et al. deduced canard cycles, relaxation oscillations and pattern formation for a slow-fast ratio-dependent predatorprey system [13]. Sun et al. investigated the impacts of climate change on vegetation patterns [14]. Chowdhury et al. studied eco-evolutionary cyclic dominance among predators, prey and parasites [15]. The dynamic behavior of a plant-water model with spatial diffusion has been analyzed by Sun et al. [16]. But one must remember that this work was first considered by the well known ecologist W. C. Allee and later named the Allee effect [17-22]. The Allee effect, classified into two kinds of types, i.e., the strong Allee effect and weak Allee effect, states that a population that is low in number is affected by a positive relationship between the population growth rate and its density. In the process of the population growing, a critical density, called the Allee threshold, will occur and measure the extinction of the population. The Allee effect is used to study the extinction scenario of the population. When the average growth rate of the population at low density is negative, one says that there is a strong Allee effect whereas a weak Allee effect corresponds to when the average growth rate of the population at zero density is positive. For more investigations of the Allee threshold under the strong/weak Allee effect, see also, for example, [17–24] and the references cited therein. In fact, most of the research is on the codim 1 bifurcation problem of such systems. Therefore, our main aim in this paper is to consider the codim 2 bifurcation problem of a discrete prey-predator model with the Allee effect in prey.

Recall that Khan [25] considered the local stability and Neimark-Sacker bifurcation of the following discrete-time prey-predator model

$$\begin{cases} X_{t+1} = aX_t(1 - X_t) - X_tY_t, \\ Y_{t+1} = \frac{1}{\beta}X_tY_t, \end{cases}$$
(1.1)

where X_t and Y_t denote the numbers of prey and predator respectively, and the parameters *a* and β are positive real numbers.

A number of changes in the model (1.1) have been presented by researchers to include Hollingtype functional responses and density-dependent prey growth. Another way of modification of the Lotka–Volterra model can be considered by introducing the Allee effect which can more realistically model the prey–predator interaction.

In [26], the author considered the modification of the model (1.1) by introducing the Allee effect to the prey population as follows

$$\begin{cases} X_{t+1} = aX_t(1 - X_t) - X_t Y_t(\frac{X_t}{X_t + m}), \\ Y_{t+1} = \frac{1}{\beta} X_t Y_t, \end{cases}$$
(1.2)

where $\frac{X_t}{X_t+m}$ is the term for the Allee effect, m > 0 can be defined as the Allee effect coefficient [12], and the parameters *a* and β are the growth rates of the prey and predator, respectively.

Considering the biological meanings of system (1.2), its nonnegative fixed points were derived in [26], and they are $E_1 = (0,0)$, $E_2 = (\frac{a-1}{a}, 0)$ if a > 1 and $E_* = (\beta, \frac{(\beta+m)(a(1-\beta)-1)}{\beta})$ if $0 < \beta < 1 - \frac{1}{a}$.

Regarding the stability analysis of E_1 , E_2 and E_* , some results have been obtained in [26]. For the readers' convenience, we present them in Appendix A.

On the one hand, however, based on those results presented in Appendix A, which are mainly for the type and the stability of the positive fixed point E_* , we find that there are some errors, which can be shown as follows.

Counterexample 1.1. Consider system (1.2) with the parameters

$$a = 3, \quad \beta = 0.6, \quad m = 0.1.$$

Obviously, $0 < \beta < 1 - \frac{1}{a}$, which means that E_* exists. Again, $\frac{1}{1-2\beta-m} < a < \frac{3\beta+5m}{3\beta^2+m\beta+m-\beta}$, which means that (*iii.*2) in Appendix A holds. According to (*iii.*2), E_* should be a source. But, by some computations, one finds that

$$\lambda_1 = 0.8831, \quad \lambda_2 = -0.7117,$$

which shows that the fixed point E_* is a saddle, not a source.

Counterexample 1.2. Consider system (1.2) with the parameters

$$a = 3$$
, $\beta = 0.1$, $m = 0.4$.

Evidently, $0 < \beta < 1 - \frac{1}{a}$, which means that E_* exists. $a > \frac{1}{1-2\beta-m}$ is also true. According to the condition (*iii*.3) in Appendix A, E_* should be a saddle. But one can easily derive the following:

$$\lambda_1 = 0.17 + i, \quad \lambda_2 = 0.17 - i,$$

which indicates that E_* is an unstable focus, not a saddle.

These counterexamples show that the stability of E_* is worthy of further studies.

On the other hand, the author of [26] only studied codim 1 Neimark-Sacker bifurcation in E_* . A question naturally arises: are there other bifurcations in E_* , especially, 1:2 resonance bifurcation? Additionally, are there bifurcations that occur in the fixed points E_1 and E_2 ?

These problems have not yet been considered in any known studies. This has motivated us to investigate these problems in this paper. We not only give precise conditions for the stability of E_* in the whole parametric space in which it exists, which serves to correct those known errors, but, also, and what is more important, we consider the complex bifurcation problems of (1.2), mainly focusing on its codim 2 resonance 1:2 bifurcation.

The rest of the paper is organized as follows. In Section 2, we give the complete stability results for E_* for the entire parametric space in which it exists. In Section 3, we respectively choose a and β as bifurcation parameters to discuss its bifurcation problems at the fixed points E_1 and E_2 , including transcritical bifurcation and period-doubling bifurcation. For the fixed point E_* , we not only consider its codim 1 period-doubling bifurcation, but also the 1:2 resonance bifurcation of codim 2 while varying the two parameters a and m. In Section 4, we present some numerical simulations to illustrate our theoretical analysis results. Finally, we draw some conclusions and discussions in Section 5.

2. Stability of fixed point E_*

In this section, we revisit the stability of the fixed point E_* of system (1.2). In order to study the stability and local bifurcation of fixed points of a general 2D system, the definitions of the topological types of a fixed point and a lemma for determining the stability of a fixed point are essential, and they are presented in Appendix B. The main results in this section are as follows.

Theorem 2.1. When $0 < \beta < 1 - \frac{1}{a}$, the unique positive fixed point E_* exists. Its dynamical properties are completely formulated in Table 1, where $R_1 = \frac{3\beta^2 + (m-1)\beta + m}{5m + 3\beta}$ and $R_2 = 1 - 2\beta - m$.

Proof. The characteristic equation of $J(E_*)$ is

$$F(\lambda) = \lambda^2 + B\lambda + C = 0, \qquad (2.1)$$

where $B = \frac{-2\beta - 3m + am + a\beta^2}{m + \beta}$, $C = \frac{a\beta - 2a\beta^2 - a\beta m + m}{m + \beta}$. Now, we will analyze the stability of the unique positive fixed point E_* . Notice that F(1) > 0 always holds when $0 < \beta < 1 - \frac{1}{a}$. Obviously,

$$F(-1) = \frac{5m + 3\beta - 3a\beta^2 - am - am\beta + a\beta}{m + \beta},$$
(2.2)

$$F(-1) > (=, <)0 \iff a(3\beta^{2} + (m-1)\beta + m) < (=, >)5m + 3\beta$$
$$\iff \frac{1}{a} > (=, <)\frac{3\beta^{2} + (m-1)\beta + m}{5m + 3\beta} \stackrel{\scriptscriptstyle \triangle}{=} R_{1},$$
$$C - 1 > (=, <)0 \iff a(1 - 2\beta - m) - 1 > (=, <)0$$
$$\iff \frac{1}{a} < (=, >)1 - 2\beta - m \stackrel{\scriptscriptstyle \triangle}{=} R_{2}$$
$$(2.3)$$

and

$$R_{1} > (=, <)R_{2} \iff 5m^{2} + (14\beta - 4)m + 9\beta^{2} - 4\beta > (=, <)0$$

$$\iff (5m + 9\beta - 4)(m + \beta) > (=, <)0$$

$$\iff m > (=, <)\frac{4 - 9\beta}{5}.$$

$$(2.4)$$

Mathematical Biosciences and Engineering

	Conditions		Eigenvalues	Properties
$0 < \beta < \frac{4}{9}$	$m < \frac{4-9\beta}{5}$	$0 < \frac{1}{a} < R_1$	$ \lambda_1 < 1, \lambda_2 > 1$	Saddle
		$\frac{1}{a} = R_1$	$\lambda_1 = -1, \lambda_2 \neq 1$	Non- hyperbolic
		$R_1 < \frac{1}{a} < R_2$	$ \lambda_{1,2} > 1$	Source
		$\frac{1}{a} = R_2$	$ \lambda_i = 1, \lambda_1 = \overline{\lambda}_2$	Non- hyperbolic
		$R_2 < \frac{1}{a} < 1 - \beta$	$ \lambda_{1,2} < 1$	Sink
	$m = \frac{4-9\beta}{5}$	$0 < \frac{1}{a} < R_2$	$ \lambda_1 < 1, \lambda_2 > 1$	Saddle
		$\frac{1}{a} = R_2$	$\lambda_{1,2} = -1$	Non- hyperbolic
		$R_2 < \frac{1}{a} < 1 - \beta$	$ \lambda_{1,2} < 1$	Sink
	$m > \frac{4-9\beta}{5}$	$0 < \frac{1}{a} < R_1$	$ \lambda_1 < 1, \lambda_2 > 1$	Saddle
		$\frac{1}{a} = R_1$	$\lambda_1 = -1, \lambda_2 \neq 1$	Non- hyperbolic
		$R_1 < \frac{1}{a} < 1 - \beta$	$ \lambda_{1,2} < 1$	Sink
$\frac{4}{9} \leq \beta < \frac{2}{3}$	<i>m</i> > 0	$0 < \frac{1}{a} < R_1$	$ \lambda_1 < 1, \lambda_2 > 1$	Saddle
		$\frac{1}{a} = R_1$	$\lambda_1 = -1, \lambda_2 \neq 1$	Non- hyperbolic
		$R_1 < \frac{1}{a} < 1 - \beta$	$ \lambda_{1,2} < 1$	Sink
$\tfrac{2}{3} \leq \beta < 1 - \tfrac{1}{a}$	m > 0	$0 < \frac{1}{a} < 1 - \beta$	$ \lambda_1 < 1, \lambda_2 > \overline{1}$	Saddle

Table 1. Properties of the positive fixed point E_* .

If $1 < a \le 2$, then $(0, 1 - \frac{1}{a}) \subset (0, \frac{1}{2})$; if 2 < a, then $(0, \frac{1}{2}) \subset (0, 1 - \frac{1}{a})$. Now for a > 2, we separate $\beta \in (0, 1 - \frac{1}{a})$ into two cases: $\beta \in (0, \frac{1}{2})$ and $\beta \in [\frac{1}{2}, 1 - \frac{1}{a})$.

Case I: $\frac{1}{2} \leq \beta < 1 - \frac{1}{a}$. Then, $\frac{4-9\beta}{5} < 0 < m$. So, it follows from (2.4) that $R_1 > R_2$, and $1 - 2\beta - m < 0 < \frac{1}{a}$. So C < 1. Notice the following condition:

$$max\{1-2\beta-m, \frac{3\beta^2+(m-1)\beta+m}{5m+3\beta}\} < (=,>)1-\beta$$
$$\iff \frac{3\beta^2+(m-1)\beta+m}{5m+3\beta} < (=,>)1-\beta$$
$$\iff \beta < (=,>)\frac{2}{3}.$$

Hence, one can further divide $\beta \in [\frac{1}{2}, 1 - \frac{1}{a})$ into two subcases $\beta \in [\frac{1}{2}, \frac{2}{3})$ and $\beta \in [\frac{2}{3}, 1 - \frac{1}{a})$ if possible. Subcase I: $\frac{1}{2} \leq \beta < \frac{2}{3}$. Then the following derivations hold true.

- 1) $0 < \frac{1}{a} < R_1$. It follows from (2.3) that F(-1) < 0. According to Lemma 5.2(i.3), the eigenvalues are $|\lambda_1| < 1$ and $|\lambda_2| > 1$. So, Definition 5.1(3) tells us that E_* is a saddle.
- 2) $\frac{1}{a} = R_1$. One can see from (2.3) that F(-1) = 0 and $C \neq 1$. Lemma 5.2(i.2) yields that the eigenvalues are $\lambda_1 = -1$ and $\lambda_2 \neq 1$. By Definition 5.1(4), E_* is non-hyperbolic.
- 3) $R_1 < \frac{1}{a} < 1 \beta$. Then from (2.3) we know that F(-1) > 0 and C < 1. According to Lemma 5.2(i.1), the eigenvalues satisfy that $|\lambda_{1,2}| < 1$. From Definition 5.1(1), E_* is a sink.

Subcase II: $\frac{2}{3} \le \beta < 1 - \frac{1}{a}$. Then $0 < \frac{1}{a} < 1 - \beta < R_1$. Using (2.3), Lemma 5.2(i.3) and Definition 5.1(3) in the same way, it is easy to derive that E_* is a saddle.

Case II: $0 < \beta < \frac{1}{2}$.

Then, we consider two subcases: I: $0 < \beta < \frac{4}{9}$ and II: $\frac{4}{9} \le \beta < \frac{1}{2}$. First consider Subcase I: $0 < \beta < \frac{4}{9}$. We divide $m \in (0, +\infty)$ into three cases:

$$1.0 < m < \frac{4-9\beta}{5}, \quad 2.m = \frac{4-9\beta}{5}, \quad 3.m > \frac{4-9\beta}{5}.$$

Correspondingly, $R_1 < (=, >)R_2$. For the first case that $0 < m < \frac{4-9\beta}{5}$, $R_1 < R_2$. Then consider $\frac{1}{a} \in$ $(0, 1 - \beta)$ for the following five cases:

- 1) $0 < \frac{1}{a} < R_1 \Longrightarrow F(-1) < 0 \Longrightarrow |\lambda_1| < 1, |\lambda_2| > 1 \Longrightarrow E_*$ is a saddle; 2) $\frac{1}{a} = R_1 \Longrightarrow F(-1) = 0, C \neq 1 \Longrightarrow \lambda_1 = -1, \lambda_2 \neq -1 \Longrightarrow E_*$ is non-hyperbolic (flip bifurcation) may occur);
- 3) $R_1 < \frac{1}{a} < R_2 \implies F(-1) > 0, C > 1 \implies |\lambda_{1,2}| > 1 \implies E_*$ is a source; 4) $\frac{1}{a} = R_2 \implies C = 1, F(-1) > 0 \implies |\lambda_{1,2}| = 1, \lambda_1 = \overline{\lambda_2} \implies E_*$ is non-hyperbolic (Neimark-Sacker bifurcation possibly occurs);
- 5) $R_2 < \frac{1}{a} < 1 \beta \Longrightarrow C < 1, F(-1) > 0 \Longrightarrow |\lambda_{1,2}| < 1 \Longrightarrow E_*$ is a sink.

For the second case of $m = \frac{4-9\beta}{5}(R_1 = R_2)$ and the third case that $m > \frac{4-9\beta}{5}(R_1 > R_2), \frac{1}{a} \in (0, 1 - \beta)$

can be similarly divided. For simplification, these cases can be shown in the following formats.

$$0 < \beta < \frac{4}{9} \begin{cases} 0 < m < \frac{4-9\beta}{5} \\ m = \frac{4-9\beta}{5} \end{cases} \begin{cases} 0 < \frac{1}{a} < R_1, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_1, & E_* \text{ is non - hyperbolic;} \\ R_1 < \frac{1}{a} < R_2, & E_* \text{ is a source;} \\ \frac{1}{a} = R_2, & E_* \text{ is non - hyperbolic;} \\ R_2 < \frac{1}{a} < 1 - \beta, & E_* \text{ is a sink.} \end{cases}$$

$$m = \frac{4-9\beta}{5} \begin{cases} 0 < \frac{1}{a} < R_2, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_2, & E_* \text{ is non - hyperbolic;} \\ R_2 < \frac{1}{a} < 1 - \beta, & E_* \text{ is a saddle;} \end{cases}$$

$$m > \frac{4-9\beta}{5} \begin{cases} 0 < \frac{1}{a} < R_2, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_2, & E_* \text{ is non - hyperbolic;} \\ R_2 < \frac{1}{a} < 1 - \beta, & E_* \text{ is a saddle;} \end{cases}$$

$$m > \frac{4-9\beta}{5} \begin{cases} 0 < \frac{1}{a} < R_2, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_2, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_2, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_1, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_1, & E_* \text{ is non - hyperbolic;} \\ R_1 < \frac{1}{a} < 1 - \beta, & E_* \text{ is a sink.} \end{cases}$$

Next consider Subcase II: $\frac{4}{9} \le \beta < \frac{1}{2}$. Then one can get that $R_1 > R_2$. Similar to the analysis for Subcase I, one can derive the following:

$$\frac{4}{9} \leq \beta < \frac{1}{2} \begin{cases} 0 < \frac{1}{a} < R_2, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_2, & E_* \text{ is a saddle;} \\ R_2 < \frac{1}{a} < R_1, & E_* \text{ is a saddle;} \\ \frac{1}{a} = R_1, & E_* \text{ is non - hyperbolic;} \\ R_1 < \frac{1}{a} < 1 - \beta, E_* \text{ is a sink;} \end{cases}$$

this may be reduced to the following:

$$\frac{4}{9} \leq \beta < \frac{1}{2} \begin{cases} 0 < \frac{1}{a} < R_1, & E_* \text{ is a saddle}; \\ \frac{1}{a} = R_1, & E_* \text{ is non - hyperbolic}; \\ R_1 < \frac{1}{a} < 1 - \beta, E_* \text{ is a sink.} \end{cases}$$

This case is the same as the case that $\frac{1}{2} \le \beta \le \frac{2}{3}$. So, these two cases can be combined into one case. Thus, we have summarized all of the results in Table 1.

3. Bifurcation analysis

In this section, we respectively consider the local bifurcation problems of system (1.2) at the fixed points E_1 , E_2 and E_* , including codimension one (codim 1) bifurcation and codimension two (codim 2) bifurcation.

3.1. Codimension one bifurcation

3.1.1. Bifurcation at the trivial fixed point $E_1(0, 0)$

For the bifurcation at the fixed point $E_1(0,0)$, one has the following result.

Theorem 3.1. As the parameter a passes through the critical value $a_0 = 1$, system (1.2) undergoes a transcritical bifurcation at the fixed point $E_1(0, 0)$.

Proof. Given a small perturbation a^* of the parameter a around a_0 , namely, $a^* = a - a_0$, with $0 < |a^*| \ll 1$, system (1.2) is changed into the following form:

$$\begin{cases} x_{t+1} = (1+a^*)x_t(1-x_t) - \frac{x_t^2 y_t}{m+x_t}, \\ y_{t+1} = \frac{1}{\beta}x_t y_t. \end{cases}$$
(3.1)

Set $a_{t+1}^* = a_t^* = a^*$; then, (3.1) becomes

$$\begin{cases} x_{t+1} = (1 + a_t^*) x_t (1 - x_t) - \frac{x_t^2 y_t}{m + x_t}, \\ y_{t+1} = \frac{1}{\beta} x_t y_t, \\ a_{t+1}^* = a_t^*. \end{cases}$$
(3.2)

Applying Taylor expansion to (3.2) at $(x_t, y_t, a_t^*)=(0,0,0)$ yields

$$\begin{pmatrix} x_t \\ y_t \\ a_t^* \end{pmatrix} \longrightarrow \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_t \\ y_t \\ a_t^* \end{pmatrix} + \begin{pmatrix} g_1(x_t, y_t, a_t^*) + o(\rho_1^3) \\ g_2(x_t, y_t, a_t^*) + o(\rho_1^3) \\ 0 \end{pmatrix},$$
(3.3)

where $\rho_1 = \sqrt{x_t^2 + y_t^2 + (a_t^*)^2}$,

$$g_1(x_t, y_t, a_t^*) = x_t a_t^* - x_t^2 a_t^* - x_t^2 - \frac{x_t^2 y_t}{m},$$

$$g_2(x_t, y_t, a_t^*) = \frac{x_t y_t}{\beta}.$$

Assume the following for the center manifold

$$y_t = h(x_t, a_t^*) = a_{20}x_t^2 + a_{11}x_ta_t^* + a_{02}a_t^{*2} + o(\rho_2^2),$$

where $\rho_2 = \sqrt{x_t^2 + {a_t^*}^2}$; then, from

$$y_{t+1} = h(x_{t+1}, a_{t+1}^*) = g_2(x_t, h(x_t, a_t^*), a_t^*) + o(\rho_2^2),$$

$$y_{t+1} = a_{20}x_{t+1}^2 + a_{11}x_{t+1}a_t^* + a_{02}a_t^{*2} + o(\rho_2^2),$$

$$x_{t+1} = x_t + g_1(x_t, h(x_t, a_t^*), a_t^*) + o(\rho_2^2),$$

comparing the corresponding coefficients of terms with the same order as in the above center manifold equation, it is easy to derive that

$$a_{20} = a_{11} = a_{02} = 0.$$

Mathematical Biosciences and Engineering

So, system (3.3), restricted to the center manifold, is given by

$$\begin{aligned} x_{t+1} &= f_1(x_t, a_t^*) := x_t + g_1(x_t, h(x_t, a_t^*), a_t^*) + o(\rho_2^3) \\ &= x_t + x_t a_t^* - x_t^2 a_t^* - x_t^2 + o(\rho_2^3). \end{aligned}$$

Therefore, the following results are derived:

$$\begin{aligned} f_1(x_t, a_t^*)|_{(0,0)} &= 0, \frac{\partial f_1}{\partial x_t}\Big|_{(0,0)} = 1, \frac{\partial f_1}{\partial a_t^*}\Big|_{(0,0)} = 0, \\ \frac{\partial^2 f_1}{\partial x_t \partial a_t^*}\Big|_{(0,0)} &= 1 > 0, \frac{\partial^2 f_1}{\partial x_t^2}\Big|_{(0,0)} = -2 < 0. \end{aligned}$$

According to (21.1.43)–(21.1.46) in [27, pp 507], all conditions hold for a transcritical bifurcation to occur; hence, system (1.2) undergoes a transcritical bifurcation at the fixed point E_1 .

3.1.2. Bifurcation at the boundary fixed point $E_2 = (\frac{a-1}{a}, 0)$

The fixed point E_2 always exists when a > 1. One can see from [26] that E_2 is a non-hyperbolic fixed point when a = 3 or $\beta = \frac{a-1}{a}$. As soon as the parameter a or β goes through the corresponding critical values, the dimensional numbers for the stable manifold and the unstable manifold of the fixed point E_2 vary. So, a bifurcation probably occurs. Now, the considered parameter values are divided into the following three cases:

Case I: $a = 3, \beta \neq \frac{2}{3}$; Case II: $a \neq 3, \beta = \frac{a-1}{a}$; Case III: $a = 3, \beta = \frac{2}{3}$.

First, we consider Case I: $a = 3, \beta \neq \frac{2}{3}$, namely, the parameters $(a, \beta) \in \Omega_1 = \{(a, \beta) \in R^2_+ | \beta \neq \frac{2}{3}, a > 1\}$. Then, the following result is obtained.

Theorem 3.2. Assume the parameters $(a,\beta) \in \Omega_1 = \{(a,\beta) \in R^2_+ | \beta \neq \frac{2}{3}, a > 1\}$, and let $a_0 = 3$; then, system (1.2) undergoes a period-doubling bifurcation at E_2 when the parameter a goes through the critical value a_0 .

Proof. Let $l_t = X_t - \frac{a-1}{a}$ and $m_t = Y_t - 0$, which transforms $E_2 = (\frac{a-1}{a}, 0)$ to the origin O(0, 0) and system (1.2) into the following:

$$\begin{cases} l_{t+1} = l_t (2 - a - al_t) - (\frac{a - 1}{a} + l_t) m_t (\frac{a - 1 + al_t}{am + a - 1 + al_t}), \\ m_{t+1} = \frac{1}{\beta} (l_t + \frac{a - 1}{a}) m_t. \end{cases}$$
(3.4)

Give a small perturbation a^* of the parameter a around a_0 , namely, $a^* = a - a_0$, with $0 < |a^*| \ll 1$; then, system (3.4) is changed into the following:

$$\begin{cases} l_{t+1} = -(1+a^*)l_t - (3+a^*)l_t^2 - (\frac{2+a^*}{3+a^*} + l_t)\frac{2+a^*+(3+a^*)l_t}{(3+a^*)(m+l_t)+2+a^*}m_t, \\ m_{t+1} = \frac{1}{\beta}(l_t + \frac{2+a^*}{3+a^*})m_t. \end{cases}$$
(3.5)

Mathematical Biosciences and Engineering

Set $a_{t+1}^* = a_t^* = a^*$ to change (3.5) into the following:

$$\begin{cases} l_{t+1} = -(1+a_t^*)l_t - (3+a_t^*)l_t^2 - (\frac{2+a_t^*}{3+a_t^*} + l_t)(\frac{2+a_t^*+(3+a_t^*)l_t}{(3+a_t^*)(m+l_t)+2+a_t^*})m_t, \\ m_{t+1} = \frac{1}{\beta}(l_t + \frac{2+a_t^*}{3+a_t^*})m_t, \\ a_{t+1}^* = a_t^*. \end{cases}$$
(3.6)

Applying a Taylor expansion to (3.6) at $(l_t, m_t, a_t^*)=(0,0,0)$ yields

$$\begin{pmatrix} l_t \\ m_t \\ a_t^* \end{pmatrix} \longrightarrow \begin{pmatrix} -1 & \frac{-4}{3(3m+2)} & 0 \\ 0 & \frac{2}{3\beta} & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} l_t \\ m_t \\ a_t^* \end{pmatrix} + \begin{pmatrix} g_1(l_t, m_t, a_t^*) + o(\rho_1^3) \\ g_2(l_t, m_t, a_t^*) + o(\rho_1^3) \\ 0 \end{pmatrix},$$
(3.7)

where $\rho_1 = \sqrt{l_t^2 + m_t^2 + (a_t^*)^2}$,

$$g_{1}(l_{t}, m_{t}, a_{t}^{*}) = -3l_{t}^{2} - \frac{(12m+4)}{(3m+2)^{2}}l_{t}m_{t} - l_{t}a_{t}^{*} - \frac{(12m+4)}{9(3m+2)^{2}}m_{t}a_{t}^{*}$$
$$- \frac{27m^{2}}{(3m+2)^{3}}l_{t}^{2}m_{t} - l_{t}^{2}a_{t}^{*} - \frac{6m^{2}}{(3m+2)^{3}}l_{t}m_{t}a_{t}^{*}$$
$$+ \frac{27m^{2} + 36m + 8}{27(3m+2)^{3}}m_{t}(a_{t}^{*})^{2},$$
$$g_{2}(l_{t}, m_{t}, a_{t}^{*}) = \frac{1}{\beta}l_{t}m_{t} + \frac{1}{9\beta}m_{t}a_{t}^{*} - \frac{1}{27\beta}m_{t}(a_{t}^{*})^{2}.$$

It is easy to derive the three eigenvalues of the matrix $A = \begin{pmatrix} -1 & \frac{-4}{3(3m+2)} & 0\\ 0 & \frac{2}{3\beta} & 0\\ 0 & 0 & 1 \end{pmatrix}$ to be $\lambda_1 = -1$, $\lambda_2 = \frac{2}{3\beta}$

and $\lambda_3 = 1$ with the corresponding eigenvectors $\xi_1 = \begin{pmatrix} 1\\0\\0 \end{pmatrix}$, $\xi_2 = \begin{pmatrix} \frac{3\beta}{2+3\beta}\\ -\frac{3(3m+2)}{4}\\0 \end{pmatrix}$ and $\xi_3 = \begin{pmatrix} 0\\0\\1 \end{pmatrix}$. $\beta \neq \frac{2}{3}$ implies that $|\lambda_2| \neq 1$.

Set $T = (\xi_1, \xi_2, \xi_3)$, namely,

$$T = \begin{pmatrix} 1 & \frac{3\beta}{2+3\beta} & 0\\ 0 & \frac{-3(3m+2)}{4} & 0\\ 0 & 0 & 1 \end{pmatrix}; \text{ then, } T^{-1} = \begin{pmatrix} 1 & \frac{4\beta}{(3\beta+2)(3m+2)} & 0\\ 0 & \frac{-4}{3(3m+2)} & 0\\ 0 & 0 & 1 \end{pmatrix}.$$

The transformation $\begin{pmatrix} l_t \\ m_t \\ a_t^* \end{pmatrix} = T \begin{pmatrix} u_t \\ v_t \\ \sigma_t \end{pmatrix}$ changes system (3.7) into the following: $\begin{pmatrix} u_{t+1} \\ v_{t+1} \\ \sigma_{t+1} \end{pmatrix} = \begin{pmatrix} -1 & 0 & 0 \\ 0 & \frac{2}{3\beta} & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_t \\ v_t \\ \sigma_t \end{pmatrix} + \begin{pmatrix} g_3(u_t, v_t, \sigma_t) + o(\rho_2^3) \\ g_4(u_t, v_t, \sigma_t) + o(\rho_2^3) \\ 0 \end{pmatrix},$ (3.8)

Mathematical Biosciences and Engineering

where $\rho_2 = \sqrt{u_t^2 + v_t^2 + \sigma_t^2}$,

$$g_{3}(u_{t}, v_{t}, \sigma_{t}) = g_{1}(u_{t} + \frac{3\beta}{3\beta + 2}v_{t}, \frac{-3(3m + 2)}{4}v_{t}, \sigma_{t}) + \frac{4\beta}{(3\beta + 2)(3m + 2)}g_{2}(u_{t} + \frac{3\beta}{3\beta + 2}v_{t}, \frac{-3(3m + 2)}{4}v_{t}, \sigma_{t}),$$

$$g_{4}(u_{t}, v_{t}, \sigma_{t}) = \frac{-4}{3(3m + 2)}g_{2}(u_{t} + \frac{3\beta}{3\beta + 2}v_{t}, \frac{-3(3m + 2)}{4}v_{t}, \sigma_{t}).$$

Assume the following for the center manifold [28]:

$$v_t = h(u_t, \sigma_t) = c_{20}u_t^2 + c_{11}u_t\sigma_t + c_{02}\sigma_t^2 + o(\rho_3^2),$$

where $\rho_3 = \sqrt{u_t^2 + \sigma_t^2}$; then, according to the following relations:

$$\begin{aligned} v_{t+1} &= h(v_{t+1}, \sigma_{t+1}) = \frac{2}{3\beta} v_t + g_4(u_t, h(u_t, \sigma_t), \sigma_t) + o(\rho_3^2), \\ v_{t+1} &= c_{20} u_{t+1}^2 + c_{11} u_{t+1} \sigma_t + c_{02} \sigma_t^2 + o(\rho_3^2), \\ u_{t+1} &= -u_t + g_3(u_t, h(u_t, \sigma_t), \sigma_t) + o(\rho_3^2), \end{aligned}$$

and by comparing the corresponding coefficients of terms with the same order as in the above center manifold equation, one derives that

$$c_{20} = c_{11} = c_{02} = 0.$$

So, system (3.8) restricted to the center manifold is given by

$$u_{t+1} = f_1(u_t, \sigma_t) := -u_t + g_3(u_t, h(u_t, \sigma_t), \sigma_t) + o(\rho_3^3)$$

= $-u_t - 3u_t^2 - u_t \sigma_t - u_t^2 \sigma_t + o(\rho_3^3).$

Thus, one has

$$f_1^2(u_t, \sigma_t) = f_1(f_1(u_t, \sigma_t), \sigma_t) = u_t + 2u_t\sigma_t - 18u_t^3 - 3u_t^2\sigma_t + u_t\sigma_t^2 + o(\rho_3^3).$$

Therefore, the following results are derived:

$$f_1(0,0) = 0, \frac{\partial f_1}{\partial u_t}|_{(0,0)} = -1, \frac{\partial f_1^2}{\partial \sigma_t}|_{(0,0)} = 0, \frac{\partial^2 f_1^2}{\partial u_t^2}|_{(0,0)} = 0,$$
$$\frac{\partial^2 f_1^2}{\partial u_t \partial \sigma_t}|_{(0,0)} = 2 \neq 0, \frac{\partial^3 f_2^2}{\partial u_t^3}|_{(0,0)} = -108 \neq 0,$$

which, according to (21.2.17)–(21.2.22) in [27, pp 516], satisfy all of the conditions for a period-doubling bifurcation to occur. So, system (1.2) undergoes a period-doubling bifurcation at E_2 . Again,

$$\frac{-\frac{\partial^3 f_1^2}{\partial u_t^3}|_{(0,0)}}{\frac{\partial^2 f_1^2}{\partial u_t \partial \sigma_t}|_{(0,0)}} = 54 > 0.$$

Mathematical Biosciences and Engineering

So, the period-two orbit bifurcated from E_2 lies to the right of $a_0 = 3$.

Of course, one can also compute the following two quantities, which respectively, are the transversal condition and non-degenerate condition for judging the occurrence and direction of a period-doubling bifurcation (see [27,28]):

$$\begin{split} \alpha_1 &= \Big(\frac{\partial^2 f_1}{\partial u_t \partial \sigma_t} + \frac{1}{2} \frac{\partial f_1}{\partial \sigma_t} \frac{\partial^2 f_1}{\partial u_t^2}\Big)\Big|_{(0,0)},\\ \alpha_2 &= \Big(\frac{1}{6} \frac{\partial^3 f_1}{\partial u_t^3} + \Big(\frac{1}{2} \frac{\partial^2 f_1}{\partial u_t^2}\Big)^2\Big)\Big|_{(0,0)}. \end{split}$$

It is easy to get that $\alpha_1 = -1$ and $\alpha_2 = 9$. Because $\alpha_1 \neq 0$ and $\alpha_2 > 0$, a period-doubling bifurcation occurs in system (1.2) at E_2 and the period-two orbit bifurcating from E_2 is stable. The proof is complete.

Next, we study Case II: $a \neq 3$, $\beta = \frac{a-1}{a}$. We can see that $|\lambda_1| \neq 1$ and $\lambda_2 = -1$ when $a \neq 3$ and $\beta = \frac{a-1}{a}$. Thus, the following result can be derived.

Theorem 3.3. Assume the parameters $(a,\beta) \in \Omega_2 = \{(a,\beta) \in R^2_+ | a \in (1,3) \cup (3,+\infty)\}$, and set $\beta_0 = \frac{a-1}{a}$. If the parameter β goes through the critical value β_0 , then system (1.2) at E_2 undergoes a transcritical bifurcation.

Proof. Shifting $E_2 = (\frac{a-1}{a}, 0)$ to the origin O(0, 0), applying a small perturbation β^* of the parameter β with $0 < |\beta^*| \ll 1$ and letting $\beta^*_{t+1} = \beta^*_t = \beta^*$, system (3.4) is changed into the following form:

$$\begin{cases} l_{t+1} = -(a-2)l_t - al_t^2 - (\frac{a-1}{a} + l_t)(\frac{a-1+al_t}{am+a-1+al_t})m_t, \\ m_{t+1} = \frac{1}{\beta_t^* + \frac{a-1}{a}}(l_t + \frac{a-1}{a})m_t, \\ \beta_{t+1}^* = \beta_t^*. \end{cases}$$
(3.9)

Applying a Taylor expansion to (3.9) at $(l_t, m_t, \beta_t^*) = (0, 0, 0)$ produces

$$\begin{pmatrix} l_t \\ m_t \\ \beta_t^* \end{pmatrix} \longrightarrow \begin{pmatrix} 2-a & \frac{-(a-1)^2}{a(ma+a-1)} & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} l_t \\ m_t \\ \beta_t^* \end{pmatrix} + \begin{pmatrix} g_5(l_t, m_t, \beta_t^*) + o(\rho_4^3) \\ g_6(l_t, m_t, \beta_t^*) + o(\rho_4^3) \\ 0 \end{pmatrix},$$
(3.10)

where $\rho_4 = \sqrt{l_t^2 + m_t^2 + {\beta_t^*}^2}$,

$$g_{5}(l_{t}, m_{t}, \beta_{t}^{*}) = -al_{t}^{2} - \frac{(2am + a - 1)(a - 1)}{(am + a - 1)^{2}}l_{t}m_{t} - \frac{a^{3}m^{2}}{(am + a - 1)^{3}}l_{t}^{2}m_{t},$$

$$g_{6}(l_{t}, m_{t}, \beta_{t}^{*}) = \frac{a}{a - 1}l_{t}m_{t} - \frac{a}{a - 1}m_{t}\beta_{t}^{*} - (\frac{a}{a - 1})^{2}l_{t}m_{t}\beta_{t}^{*} + (\frac{a}{a - 1})^{2}m_{t}(\beta_{t}^{*})^{2}.$$

It is not difficult to derive the three eigenvalues of the matrix

$$A = \begin{pmatrix} 2-a & \frac{-(a-1)^2}{a(ma+a-1)} & 0\\ 0 & 1 & 0\\ 0 & 0 & 1 \end{pmatrix} \text{ to be } \lambda_1 = 2-a \text{ and } \lambda_{2,3} = 1, \text{ with the corresponding eigenvectors } \xi_1 = \begin{pmatrix} 1\\ 0\\ 0 \end{pmatrix},$$
$$\xi_2 = \begin{pmatrix} \frac{1}{1-a}\\ \frac{a(ma+a-1)}{(a-1)^2}\\ 0 \end{pmatrix} \text{ and } \xi_3 = \begin{pmatrix} 0\\ 0\\ 1 \end{pmatrix}. \text{ The condition } a \in (1,3) \cup (3,+\infty) \text{ shows that } |\lambda_1| \neq 1.$$

Mathematical Biosciences and Engineering

Take
$$T = \begin{pmatrix} 1 & \frac{1}{1-a} & 0\\ 0 & \frac{a(ma+a-1)}{(a-1)^2} & 0\\ 0 & 0 & 1 \end{pmatrix}$$
; then, $T^{-1} = \begin{pmatrix} 1 & \frac{(a-1)}{a(ma+a-1)} & 0\\ 0 & \frac{(a-1)^2}{a(ma+a-1)} & 0\\ 0 & 0 & 1 \end{pmatrix}$.
Using the transformation
$$\begin{pmatrix} l_t\\ \beta_t^* \end{pmatrix} = T \begin{pmatrix} u_t\\ v_t\\ \sigma_t \end{pmatrix},$$

system (3.9) is changed into the following:

$$\begin{pmatrix} u_{t+1} \\ v_{t+1} \\ \sigma_{t+1} \end{pmatrix} = \begin{pmatrix} 2-a & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_t \\ v_t \\ \sigma_t \end{pmatrix} + \begin{pmatrix} g_7(u_t, v_t, \sigma_t) + o(\rho_5^3) \\ g_8(u_t, v_t, \sigma_t) + o(\rho_5^3) \\ 0 \end{pmatrix},$$
(3.11)

where $\rho_5 = \sqrt{u_t^2 + v_t^2 + \sigma_t^2}$,

$$g_{7}(u_{t}, v_{t}, \sigma_{t}) = g_{5}(u_{t} + \frac{1}{1-a}v_{t}, \frac{a(ma+a-1)}{(a-1)^{2}}v_{t}, \sigma_{t}) \\ + \frac{(a-1)}{a(ma+a-1)}g_{6}(u_{t} + \frac{1}{1+a}v_{t}, \frac{a(ma+a-1)}{(a-1)^{2}}v_{t}, \sigma_{t}), \\ g_{8}(u_{t}, v_{t}, \sigma_{t}) = \frac{(a-1)^{2}}{a(ma+a-1)}g_{6}(u_{t} + \frac{1}{1+a}v_{t}, \frac{a(ma+a-1)}{(a-1)^{2}}v_{t}, \sigma_{t}).$$

Assume the following for this center manifold:

$$u_t = h(v_t, \sigma_t) := a_{20}v_t^2 + a_{11}v_t\sigma_t + a_{02}\sigma_t^2 + o(\rho_6^2),$$

where $\rho_6 = \sqrt{v_t^2 + \sigma_t^2}$, which must satisfy

$$u_{t+1} = h(u_{t+1}, \sigma_{t+1}) = (2 - a)h(v_t, \sigma_t) + g_7(h(v_t, \sigma_t), v_t, \sigma_t) + o(\rho_6^2),$$

$$u_{t+1} = a_{20}v_{t+1}^2 + a_{11}v_{t+1}\sigma_t + a_{02}\sigma_t^2 + o(\rho_6^2),$$

$$v_{t+1} = v_t + g_8(h(v_t, \sigma_t), v_t, \sigma_t) + o(\rho_6^2).$$

Comparing the corresponding coefficients of terms to the same type in the above equations produces

$$a_{20} = \frac{a(a-1)(2am+a-1)-a^2}{(a-1)^4}, a_{11} = \frac{-a^2(ma+a-1)}{(a-1)^4}, a_{02} = 0.$$

That is to say,

$$u_t = h(v_t, \sigma_t) = \frac{a(a-1)(2am+a-1) - a^2}{(a-1)^4} v_t^2 - \frac{a^2(ma+a-1)}{(a-1)^4} v_t \sigma_t + o(\rho_6^2).$$

Thus, one has

$$v_{t+1} = f_2(v_t, \sigma_t) := v_t + g_8(h(v_t, \sigma_t), v_t, \sigma_t) + o(\rho_6^2)$$

Mathematical Biosciences and Engineering

$$=v_t + \frac{a}{(a-1)^2}v_t^2 - \frac{a^2(ma+a-1)}{(a-1)^3}v_t\sigma_t + o(\rho_6^2).$$

Hence, the following equations may be obtained:

$$\begin{split} f_2(v_t, \sigma_t)|_{(0,0)} &= 0, \frac{\partial f_2}{\partial v_t} \Big|_{(0,0)} = 1, \frac{\partial f_2}{\partial \sigma_t} \Big|_{(0,0)} = 0, \\ \frac{\partial^2 f_2}{\partial v_t \partial \sigma_t} \Big|_{(0,0)} &= -\frac{a^2(ma+a-1)}{(a-1)^3} < 0, \frac{\partial^2 f_2}{\partial v_t^2} \Big|_{(0,0)} = \frac{2a}{(a-1)^2} > 0. \end{split}$$

According to (21.1.43)–(21.1.46) in [27, pp 507], all conditions are true for a transcritical bifurcation to occur; hence, system (1.2) undergoes a transcritical bifurcation at the fixed point E_2 .

Finally, for the bifurcation of system (1.2) at E_2 in Case III: $a = 3, \beta = \frac{2}{3}$, which implies that $\lambda_1 = -1$ and $\lambda_2 = 1$, a fold-flip bifurcation may occur there. This is left as our future work. **Remark.** For the fold-flip bifurcation, refer also to [29,30].

3.1.3. Bifurcation analysis at the positive fixed point $E_* = (x^*, y^*)$

From Theorem 2.1 and Table 1, it is clear that F(1) > 0, F(-1) = 0 and $B \neq 2$ when $0 < \beta < \frac{2}{3}$, $m \notin \{\frac{\beta(1-3\beta)}{\beta+1}, \frac{\beta(1-3\beta)}{\beta-3}, \frac{4}{5} - \frac{9}{5}\beta\}$ and

$$a_0 = \frac{1}{R_1} = \frac{3\beta + 5m}{3\beta^2 + \beta m + m - \beta}.$$
 (3.12)

At this time, the eigenvalues are $\lambda_1 = -1$ and $\lambda_2 = \frac{6\beta^2 + 4\beta m - m - 3\beta}{3\beta^2 + \beta m + m - \beta}$, with $|\lambda_2| \neq 1$. Moreover, when the parameter *a* goes through the critical value a_0 , the dimensional numbers for the stable manifold and the unstable manifold of the fixed point E_* vary. Hence, a bifurcation is likely to occur. In what follows, we consider the bifurcation problem of system (1.2) in this case.

Notice that the parameters a, β and m belong to the set

$$\Theta = \{(a,\beta,m) \in R^3_+ | 0 < \beta < \frac{2}{3}, m \notin \{\frac{\beta(1-3\beta)}{\beta+1}, \frac{\beta(1-3\beta)}{\beta-3}, \frac{4}{5} - \frac{9}{5}\beta\}\}.$$
(3.13)

Take $l_t = X_t - x^* = X_t - \beta$ and $m_t = Y_t - y^* = Y_t - \frac{(\beta + m)(a(1 - \beta) - 1)}{\beta}$; then, system (1.2) is changed into the following form:

$$\begin{cases} l_{t+1} = (l_t + \beta)[a(1 - l_t - \beta) - (m_t + \frac{(\beta + m)(a(1 - \beta) - 1)}{\beta})\frac{l_t + \beta}{m + l_t + \beta}] - \beta, \\ m_{t+1} = \frac{m_t}{\beta}(l_t + \beta) + \frac{l_t(\beta + m)(a(1 - \beta) - 1)}{\beta^2}. \end{cases}$$
(3.14)

Take a small perturbation a^* of the parameter a around a_0 , i.e., $a^* = a - a_0$, with $0 < |a^*| \ll 1$; then,

system (3.14) becomes as follows:

$$\begin{cases} l_{t+1} = (l_t + \beta)[(a^* + \frac{3\beta + 5m}{3\beta^2 + \beta m + m - \beta})(1 - l_t - \beta) \\ -(m_t + \frac{(\beta + m)((a^* + \frac{3\beta + 5m}{\beta})(1 - \beta) - 1)}{\beta}) \frac{l_t + \beta}{m + l_t + \beta}] - \beta, \qquad (3.15) \\ m_{t+1} = \frac{m_t}{\beta}(l_t + \beta) + \frac{l_t(\beta + m)((a^* + \frac{3\beta + 5m}{3\beta^2 + \beta m + m - \beta})(1 - \beta) - 1)}{\beta^2}. \end{cases}$$

Supposing that $a_{t+1}^* = a_t^* = a^*$, system (3.15) can be regarded as follows:

$$\begin{cases} l_{t+1} = (l_t + \beta)[(a_t^* + \frac{3\beta + 5m}{3\beta^2 + \beta m + m - \beta})(1 - l_t - \beta) \\ -(m_t + \frac{(\beta + m)((a_t^* + \frac{3\beta + 5m}{3\beta^2 + \beta m + m - \beta})^{(1 - \beta) - 1})}{\beta})\frac{l_t + \beta}{m + l_t + \beta}] - \beta, \\ m_{t+1} = \frac{m_t}{\beta}(l_t + \beta) + \frac{l_t(\beta + m)((a_t^* + \frac{3\beta + 5m}{3\beta^2 + \beta m + m - \beta})(1 - \beta) - 1)}{\beta^2}, \end{cases}$$
(3.16)

Applying a Taylor expansion to (3.16) at $(l_t, m_t, a_t^*) = (0, 0, 0)$ yields the following:

$$\begin{cases} l_{t+1} = \frac{2m\beta - 3m - \beta}{3\beta^2 + m\beta + m - \beta} l_t - \frac{\beta^2}{m + \beta} m_t + G_1(l_t, m_t, a_t^*) + o(r_1^3), \\ m_{t+1} = \frac{-2(m + \beta)^2(3\beta - 2)}{\beta^2(3\beta^2 + m\beta + m - \beta)} l_t + m_t + G_2(l_t, m_t, a_t^*) + o(r_1^3), \\ a_{t+1}^* = a_t^*, \end{cases}$$
(3.17)

where $r_1 = \sqrt{l_t^2 + m_t^2 + {a_t^*}^2}$,

$$G_{1}(l_{t}, m_{t}, a_{t}^{*}) = a_{200}l_{t}^{2} + a_{110}l_{t}m_{t} + a_{101}l_{t}a_{t}^{*} + a_{020}m_{t}^{2} + a_{011}m_{t}a_{t}^{*} + a_{002}(a^{*})^{2} + a_{300}l_{t}^{3} + a_{210}l_{t}^{2}m_{t} + a_{201}l_{t}^{2}a^{*} + a_{120}l_{t}m_{t}^{2} + a_{111}l_{t}m_{t}a_{t}^{*} + a_{102}l_{t}(a_{t}^{*})^{2} + a_{030}m_{t}^{3} + a_{021}m_{t}^{2}a_{t}^{*} + a_{012}m_{t}(a_{t}^{*})^{2} + a_{003}(a_{t}^{*})^{3}, G_{2}(l_{t}, m_{t}, a_{t}^{*}) = b_{200}l_{t}^{2} + b_{110}l_{t}m_{t} + b_{101}l_{t}a_{t}^{*} + b_{020}m_{t}^{2} + b_{011}m_{t}a_{t}^{*} + b_{002}(a^{*})^{2} + b_{300}l_{t}^{3} + b_{210}l_{t}^{2}m_{t} + b_{201}l_{t}^{2}a^{*} + b_{120}l_{t}m_{t}^{2} + b_{111}l_{t}m_{t}a_{t}^{*} + b_{102}l_{t}(a_{t}^{*})^{2} + b_{030}m_{t}^{3} + b_{021}m_{t}^{2}a_{t}^{*} + b_{012}m_{t}(a_{t}^{*})^{2} + b_{003}(a_{t}^{*})^{3},$$

 $\begin{aligned} a_{200} &= \frac{1}{3\beta^2 + m\beta + m - \beta} \left(\frac{2m^2(3\beta - 2)}{\beta(m + \beta)} - (3\beta + 5m) \right), a_{110} &= -\frac{\beta^2 + 2m\beta}{(m + \beta)^2}, \\ a_{101} &= -\frac{\beta^2 + m}{\beta + m}, a_{020} = a_{011} = a_{002} = 0, \end{aligned}$

Mathematical Biosciences and Engineering

 $\overline{a_{300} = \frac{-2m^2(3\beta-2)}{\beta(m+\beta)^2(3\beta^2+m\beta+m-\beta)}, a_{210} = -\frac{m^2}{(m+\beta)^3}, a_{201} = -\frac{\beta^3+2m\beta^2+m^2}{\beta(\beta+m)^2}, a_{120} = a_{111} = a_{102} = a_{030} = a_{021} = a_{012} = a_{003} = 0,}$ $b_{200} = 0, b_{110} = \frac{1}{\beta}, b_{101} = -\frac{(m+\beta)(\beta-1)}{\beta^2}, b_{020} = b_{011} = b_{002} = b_{300} = 0, b_{210} = b_{201} = b_{120} = b_{111} = b_{102} = b_{030} = b_{021} = b_{012} = b_{003} = 0.$ It is easy to see that the three eigenvalues of the matrix

$$A = \begin{pmatrix} \frac{2m\beta - 3m - \beta}{(3\beta^2 + m\beta + m - \beta)} & \frac{-\beta^2}{m + \beta} & 0\\ \frac{-2(m + \beta)^2(3\beta - 2)}{\beta^2(3\beta^2 + m\beta + m - \beta)} & 1 & 0\\ 0 & 0 & 1 \end{pmatrix} \text{ are } \lambda_1 = -1 ,$$

 $\lambda_2 = \frac{6\beta^2 + 4\beta m - m - 3\beta}{3\beta^2 + \beta m + m - \beta}$ and $\lambda_3 = 1$, with the corresponding eigenvectors $\xi_1 =$

$$\begin{pmatrix} \frac{\beta^2(3\beta^2+m\beta+m-\beta)}{(m+\beta)^2(3\beta-2)}\\ 1\\ 0 \end{pmatrix},$$

$$\xi_2 = \begin{pmatrix} -\frac{\beta^2}{2(m+\beta)} \\ 1 \\ 0 \end{pmatrix} \text{ and } \xi_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}.$$

Notice that $m \neq \frac{4}{5} - \frac{9}{5}\beta$. So $|\lambda_2| \neq 1$.

Set
$$T = (\xi_1, \xi_2, \xi_3)$$
, i.e., $T = \begin{pmatrix} \frac{\beta^2 (3\beta^2 + m\beta + m-\beta)}{(m+\beta)^2 (3\beta-2)} & -\frac{\beta^2}{2(m+\beta)} & 0\\ 1 & 1 & 0\\ 0 & 0 & 1 \end{pmatrix}$;

then

$$T^{-1} = \begin{pmatrix} \frac{2(m+\beta)^2(3\beta-2)}{\beta^2(9\beta^2+5m\beta-4\beta)} & \frac{(3\beta-2)(m+\beta)}{9\beta^2+5m\beta-4\beta} & 0\\ \frac{-2(m+\beta)^2(3\beta-2)}{\beta^2(9\beta^2+5m\beta-4\beta)} & \frac{2(3\beta^2+m\beta+m-3\beta)}{9\beta^2+5m\beta-4\beta} & 0\\ 0 & 0 & 1 \end{pmatrix}.$$

Next, we consider the following translation:

$$\begin{pmatrix} l_t \\ m_t \\ a_t^* \end{pmatrix} = T \begin{pmatrix} u_t \\ v_t \\ \sigma_t \end{pmatrix}.$$

Then system (3.17) is translated as follows:

$$\begin{pmatrix} u_t \\ v_t \\ \sigma_t \end{pmatrix} \longrightarrow \begin{pmatrix} -1 & 0 & 0 \\ 0 & \frac{6\beta^2 + 4m\beta - 3\beta - m}{3\beta^2 + m\beta + m - \beta} & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_t \\ v_t \\ \sigma_t \end{pmatrix} + \begin{pmatrix} G_3(u_t, v_t, \sigma_t) + o(r_2^3) \\ G_4(u_t, v_t, \sigma_t) + o(r_2^3) \\ 0 \end{pmatrix},$$
(3.18)

where $r_2 = \sqrt{u_t^2 + v_t^2 + \sigma_t^2}$,

$$\begin{pmatrix} G_3(u_t, v_t, \sigma_t) \\ G_4(u_t, v_t, \sigma_t) \\ 0 \end{pmatrix} = T^{-1} \begin{pmatrix} G_1(\frac{\beta^2(3\beta^2 + m\beta + m - \beta)}{(m+\beta)^2(3\beta - 2)}u_t - \frac{\beta^2}{2(m+\beta)}v_t, u_t + v_t, \sigma_t) \\ G_2(\frac{\beta^2(3\beta^2 + m\beta + m - \beta)}{(m+\beta)^2(3\beta - 2)}u_t - \frac{\beta^2}{2(m+\beta)}v_t, u_t + v_t, \sigma_t) \\ 0 \end{pmatrix}$$

Mathematical Biosciences and Engineering

Assume for the center manifold that $v_t = m_{20}u_t^2 + m_{11}u_t\sigma_t + m_{02}\sigma_t^2 + o(r_3^2)$, where $r_3 = \sqrt{u_t^2 + \sigma_t^2}$. It is easy to derive the following:

$$m_{20} = \frac{-2(3\beta^2 + m\beta + m - \beta)^2(9\beta^3 + 21m\beta^2 + m^2 - 3\beta^2 + 6m^2\beta)}{(m + \beta)^4(3\beta - 2)(9\beta + 5m - 4)},$$

$$m_{11} = \frac{-2(3\beta^2 + m\beta + m - \beta)^2(2\beta^2 - m\beta^2 + 3m - 3m\beta - \beta)}{\beta^2(9\beta + 5m - 4)^2(m + \beta)(3\beta - 2)},$$

$$m_{02} = 0.$$

Hence, system (3.18) restricted to the center manifold is given by $u_{t+1} =: f_3(u_t, \sigma_t) = -u_t + a_{11}u_t\sigma_t + a_{20}u_t^2 + a_{30}u_t^3 + a_{21}u_t^2\sigma_t + a_{12}u_t\sigma_t^2 + o(r_3^3)$, where

$$a_{11} = -\frac{(3\beta^2 + m\beta - 3m - \beta)(3\beta^2 + m\beta + m - \beta)}{\beta(m + \beta)(9\beta + 5m - 4)},$$

$$a_{20} = \frac{(3\beta^2 + m\beta + m - \beta)(-9\beta^3 - 28m\beta^2 - 7m^2\beta + 8m\beta - 2m^2 + 2\beta^2)}{(m + \beta)^3(9\beta + 5m - 4)(3\beta - 2)},$$

$$a_{30} = \frac{-(3\beta^2 + m\beta + m - \beta)^2}{(m + \beta)^3(9\beta + 5m - 4)(3\beta - 2)} [(4m^2 - 4m\beta + 9\beta^3 + 24m\beta^2 - m^2\beta)]$$

$$\begin{split} a_{30} &= \frac{(5\beta - 1)(3\beta - 2)(9\beta + 5m - 4)}{(m + \beta)^5(3\beta - 2)(9\beta + 5m - 4)} [(4m^2 - 4m\beta + 9\beta^3 + 24m\beta^2 - m^2\beta) \\ &\times (\frac{2m + \beta}{9\beta + 5m - 4} - \frac{2m^2\beta}{m + \beta} + \frac{\beta^2(3\beta + 5m)}{3\beta - 2} + \frac{3\beta^2 + m\beta + m - \beta}{(3\beta - 2)(9\beta + 5m - 4)}) \\ &- \frac{2m^2\beta(3\beta + m)}{m + \beta}], \\ a_{21} &= \frac{-(3\beta^2 + m\beta + m - \beta)^3}{\beta(9\beta + 5m - 4)(m + \beta)^3} [\frac{2m + \beta}{9\beta + 5m - 4} - \frac{2m^2\beta}{m + \beta} + \frac{\beta^2(3\beta + 5m)}{3\beta - 2} \\ &+ \frac{3\beta^2 + m\beta + m - \beta}{(3\beta - 2)(9\beta + 5m - 4)}] + \frac{8m^2\beta + 3\beta^3 + m^2\beta - 4m\beta - 4m^2}{\beta(m + \beta)^3(9\beta + 5m - 4)^2} \\ &= [\frac{(\beta - 1)(3\beta^2 + m\beta + m - \beta)}{3\beta - 2} - (m + \beta^2)] \\ &- \frac{2(\beta^3 + 2m\beta^2 + m^2)(3\beta^2 + m\beta + m - \beta)^2}{(9\beta + 5m - 4)(3\beta - 2)(m + \beta)^4}, \\ a_{12} &= \frac{(3\beta - 2)(3\beta^2 + m\beta + m - \beta)}{2\beta(9\beta + 5m - 4)}. \end{split}$$

Next we calculate the following quantities to judge the occurrence of a period-doubling bifurcation. One can derive the following:

$$\begin{split} f_3^2(u_t,\sigma_t) &= u_t - 2a_{11}u_t\sigma_t - 2a_{30}u_t^3 + (a_{11}^2 - 2a_{12})u_t\sigma_t^2 + a_{11}a_2u_t^2\sigma_t + o(r_3^3), \\ f_3(0,0) &= 0, \frac{\partial f_3}{\partial u_t}|_{(0,0)} = -1, \frac{\partial f_3^2}{\partial \sigma_t}|_{(0,0)} = 0, \frac{\partial^2 f_3^2}{\partial u_t^2}|_{(0,0)} = 0, \\ \beta_1 &= \frac{\partial^2 f_3^2}{\partial u_t\partial\sigma_t}|_{(0,0)} = -2a_{11}, \beta_2 = \frac{\partial^3 f_3^2}{\partial u_t^3}|_{(0,0)} = -12a_{30}. \end{split}$$

Mathematical Biosciences and Engineering

Notice that $\beta_1 = \frac{2(3\beta^2 + m\beta - 3m - \beta)(3\beta^2 + m\beta + m - \beta)}{\beta(m + \beta)(9\beta + 5m - 4)}$. Under the condition that $3\beta^2 + m\beta - 3m - \beta \neq 0$ 0 and $3\beta^2 + m\beta + m - \beta \neq 0$, one can see that $\beta_1 \neq 0$. At the same time,

$$\beta_2 \neq 0 \iff a_{30} \neq 0$$
,

or, equivalently,

$$(4m^{2} - 4m\beta + 9\beta^{3} + 24m\beta^{2} - m^{2}\beta)(\frac{2m + \beta}{9\beta + 5m - 4} - \frac{2m^{2}\beta}{m + \beta}) + \frac{\beta^{2}(3\beta + 5m)}{3\beta - 2} + \frac{3\beta^{2} + m\beta + m - \beta}{(3\beta - 2)(9\beta + 5m - 4)}) - \frac{2m^{2}\beta(3\beta + m)}{(m + \beta)} \neq 0.$$
(3.19)

So, according to [27, (21.2.17)–(21.2.22), pp 516], all conditions hold for a period-doubling bifurcation to occur. Accordingly, system (1.2) undergoes a period-doubling bifurcation at E_* .

Of course, the transversal condition and non-degenerate condition for judging the occurrence and direction of a period-doubling bifurcation are also given by the following two quantities:

$$\begin{split} \gamma_1 &= \Big(\frac{\partial^2 f_3}{\partial u_t \partial \sigma_t} + \frac{1}{2} \frac{\partial f_3}{\partial \sigma_t} \frac{\partial^2 f_3}{\partial u_t^2}\Big)\Big|_{(0,0)},\\ \gamma_2 &= \Big(\frac{1}{6} \frac{\partial^3 f_3}{\partial u_t^3} + \Big(\frac{1}{2} \frac{\partial^2 f_3}{\partial u_t^2}\Big)^2\Big)\Big|_{(0,0)}. \end{split}$$

It is easy to derive that $\gamma_1 = a_{11} \neq 0$, so system (1.2) undergoes a period-doubling bifurcation at E_* . Again,

$$\begin{split} \gamma_2 = & a_{30} + a_{20}^2 = \frac{(3\beta^2 + m\beta + m - \beta)^2}{(m + \beta)^4 (9\beta + 5m - 4)^2} [\frac{m^2 - 4m\beta - \beta^2}{m + \beta} - \frac{2\beta(3\beta + 5m)}{3\beta - 2}]^2 \\ & - \frac{(3\beta^2 + m\beta + m - \beta)^2}{(m + \beta)^5 (3\beta - 2)(9\beta + 5m - 4)} [(4m^2 - 4m\beta + 9\beta^3 + 24m\beta^2 - m^2\beta) \\ & (\frac{2m + \beta}{9\beta + 5m - 4} - \frac{2m^2\beta}{m + \beta} + \frac{\beta^2(3\beta + 5m)}{3\beta - 2} + \frac{3\beta^2 + m\beta + m - \beta}{(3\beta - 2)(9\beta + 5m - 4)}) \\ & - \frac{2m^2\beta(3\beta + m)}{(m + \beta)}]. \end{split}$$

In order to determine whether the period-doubling orbits that bifurcate from E_* are stable, let

$$\begin{split} K = & \frac{1}{(9\beta + 5m - 4)^2} \Big[\frac{m^2 - 4m\beta - \beta^2}{m + \beta} - \frac{2\beta(3\beta + 5m)}{3\beta - 2} \Big]^2 \\ & - \frac{1}{(m + \beta)(3\beta - 2)(9\beta + 5m - 4)} \Big[(4m^2 - 4m\beta + 9\beta^3 + 24m\beta^2 - m^2\beta) \\ & (\frac{2m + \beta}{9\beta + 5m - 4} - \frac{2m^2\beta}{m + \beta} + \frac{\beta^2(3\beta + 5m)}{3\beta - 2} + \frac{3\beta^2 + m\beta + m - \beta}{(3\beta - 2)(9\beta + 5m - 4)} \Big) \end{split}$$

Mathematical Biosciences and Engineering

19602

$$-\frac{2m^2\beta(3\beta+m)}{(m+\beta)}].$$
(3.20)

If K > 0 (< 0), the period-doubling orbits that bifurcate from E_* are stable (unstable). Based on the above analysis, we have the following consequences for the period-doubling bifurcation of system (1.2) at E_* .

Theorem 3.4. Let the symbols a_0 , Θ and K respectively be defined as in (3.12), (3.13) and (3.20). Assume the parameter vector $(a, \beta, m) \in \Theta$. Then system (1.2) undergoes a period-doubling bifurcation at E_* when the parameter a passes through the critical value a_0 . If K > 0 (K < 0), the period-doubling bifurcation is supercritical (subcritical), as well as stable (unstable).

3.2. Codimension two bifurcation

When a = 3 and $\beta = \frac{2}{3}$, the eigenvalues of system (1.2) at E_2 are $\lambda_1 = -\lambda_2 = 1$. Then a fold-flip bifurcation probably occurs. Even a limit cycle of a closed invariant curve or a more complicated invariant set exists nearby. We will further study this case in the future.

In this subsection, we mainly consider codimension two bifurcation of system (1.2) at E_* in the parameter space $\{(a,\beta,m)|0 < \beta < \frac{4}{9}, m > 0, a > 0\}$ when the two parameters *a* and *m* vary. When $m = m_0 := \frac{4}{5} - \frac{9}{5}\beta$, $a = a_0 := \frac{5}{1-\beta}$ and $\lambda_{1,2} = -1$, then a 1:2 strong resonance bifurcation may occur there (see Section 9.5.3 in [30], Section 3.2 in [31] and Section 3 in [32,33]). The following result may be derived.

Theorem 3.5. Consider the parameter vector $(a,\beta,m) \in \{(a,\beta,m) \in R^3_+ | 0 < \beta < \frac{4}{9}, \beta \neq \frac{2175}{637} - \frac{\sqrt{40978029}}{1911}$ and $\beta \neq \frac{\sqrt{38187398689}}{39358} - \frac{17511}{3578}\}$. If the parameters (a,m) vary in the neighborhood of (a_0,m_0) , then system (1.2) undergoes a codimension two bifurcation with 1:2 resonance at the fixed point E_* . Moreover, around E_* system (1.2) admits a pitchfork bifurcation and a non-degenerate Neimark-Sacker bifurcation.

Proof. In order to transfer the fixed point E_* to the origin point (0,0), let $l_t = x_t - \beta$ and $m_t = y_t - \frac{(\beta + m)(a(1 - \beta) - 1)}{\beta}$; then, system (1.2) becomes

$$\binom{l_{t+1}}{m_{t+1}} = \begin{pmatrix} 2 - a - \frac{\beta}{m+\beta} [a(\beta-1)+1] & \frac{-\beta^2}{m+\beta} \\ \frac{-(m+\beta)[a(\beta-1)+1]}{\beta^2} & 1 \end{pmatrix} \binom{l_t}{m_t} + \binom{G_5(l_t,m_t) + o(r_3^2)}{G_6(l_t,m_t) + o(r_3^2)},$$
(3.21)

where $r_3 = \sqrt{l_t^2 + m_t^2}$, $G_5(l_t, m_t) = (\frac{m^2[a(\beta - 1) + 1]}{\beta(m + \beta)^2} - a)l_t^2 - \frac{\beta^2 + 2m\beta}{(m + \beta)^2}l_tm_t$, $G_6(l_t, m_t) = \frac{1}{\beta}l_tm_t$. Denote

$$A(\alpha) = \left(\begin{array}{cc} 2 - a - \frac{\beta}{m+\beta} [a(\beta-1)+1] & \frac{-\beta^2}{m+\beta} \\ -(m+\beta) [a(\beta-1)+1] & \\ \hline \beta^2 & 1 \end{array} \right),$$

Mathematical Biosciences and Engineering

where $\alpha = (a, m)$. Then

$$A_0 = A(\alpha_0)|_{\alpha_0 = (a_0, m_0)} = \left(\begin{array}{cc} -3 & \frac{5\beta^2}{4(\beta - 1)} \\ \frac{16(1 - \beta)}{5\beta^2} & 1 \end{array} \right).$$

The eigenvector and the generalized eigenvector of A_0 corresponding to the eigenvalue of -1 are respectively $q_0 = \begin{pmatrix} 1 \\ 8(\beta-1)/(5\beta^2) \end{pmatrix}$ and $q_1 = \begin{pmatrix} 1 \\ 12(\beta-1)/(5\beta^2) \end{pmatrix}$. At the same time, the eigenvector and the generalized eigenvector of A_0^T corresponding to the eigenvalue of -1 are $p_1 = \begin{pmatrix} -2 \\ 5\beta^2/(4(\beta-1)) \end{pmatrix}$ and $p_0 = \begin{pmatrix} 3 \\ 5\beta^2/(-4(\beta-1)) \end{pmatrix}$, respectively. These four vectors satisfy the following equalities:

$$A_0 q_0 = -q_0, \quad A_0 q_1 = -q_1 + q_0, \quad A_0^T p_1 = -p_1, \quad A_0^T p_0 = -p_0 + p_1, \langle q_0, p_0 \rangle = \langle q_1, p_1 \rangle = 1, \quad \langle q_1, p_0 \rangle = \langle q_0, p_1 \rangle = 0,$$

where $\langle . \rangle$ stands for the standard scalar product in \mathbb{R}^2 .

Make an invertible linear transformation:

$$\binom{l_t}{m_t} = u_n q_0 + v_n q_1 = \left(\frac{1}{\frac{8(\beta - 1)}{5\beta^2}} - \frac{12(\beta - 1)}{5\beta^2}\right) \binom{u_n}{v_n},$$
(3.22)

and denote $x = \begin{pmatrix} l_t \\ m_t \end{pmatrix}$; then, the new coordinates (u_n, v_n) can be computed explicitly as follows:

$$\begin{cases} u_n = \langle x, p_0 \rangle, \\ v_n = \langle x, p_1 \rangle. \end{cases}$$
(3.23)

In the coordinates (u_n, v_n) , system (3.21) takes the following form:

$$\begin{pmatrix} u_{n+1} \\ v_{n+1} \end{pmatrix} = \begin{pmatrix} -1 + a(\alpha) & 1 + b(\alpha) \\ c(\alpha) & -1 + d(\alpha) \end{pmatrix} \begin{pmatrix} u_n \\ v_n \end{pmatrix} + \begin{pmatrix} f_1(u_n, v_n, \alpha) + o(r_4^3) \\ f_2(u_n, v_n, \alpha) + o(r_4^3) \end{pmatrix},$$
(3.24)

where $r_4 = \sqrt{u_n^2 + v_n^2 + |\alpha|^2}$,

$$f_{1}(u_{n}, v_{n}, \alpha) = \langle F(u_{n}q_{0} + v_{n}q_{1}, \alpha), p_{0} \rangle = a_{20}u_{n}^{2} + a_{11}u_{n}v_{n} + a_{02}v_{n}^{2},$$

$$f_{2}(u_{n}, v_{n}, \alpha) = \langle F(u_{n}q_{0} + v_{n}q_{1}, \alpha), p_{1} \rangle = b_{20}u_{n}^{2} + b_{11}u_{n}v_{n} + b_{02}v_{n}^{2},$$

$$F(x, \alpha) = \left(\frac{m^{2}[a(\beta - 1) + 1]}{\beta(m + \beta)^{2}} - a)l_{t}^{2} - \frac{\beta^{2} + 2m\beta}{(m + \beta)^{2}}l_{t}m_{t}\right);$$
(3.25)

Mathematical Biosciences and Engineering

also,

$$a_{20} = \frac{15m^2[a(\beta-1)+1] - 24(\beta+2m)(\beta-1)}{5\beta(m+\beta)^2} - \frac{2}{\beta} - 3a,$$

$$a_{11} = \frac{6m^2[a(\beta-1)+1] - 12(\beta+2m)(\beta-1)}{\beta(m+\beta)^2} - \frac{5}{\beta} - 6a,$$

$$a_{02} = \frac{15m^2[a(\beta-1)+1] - 36(\beta+2m)(\beta-1)}{5\beta(m+\beta)^2} - \frac{3}{\beta} - 3a,$$

$$b_{20} = \frac{-10m^2[a(\beta-1)+1] + 16(\beta+2m)(\beta-1)}{5\beta(m+\beta)^2} + \frac{2}{\beta} + 2a,$$

$$b_{11} = \frac{-4m^2[a(\beta-1)+1] + 8(\beta+2m)(\beta-1)}{5\beta(m+\beta)^2} + \frac{5}{\beta} + 4a,$$

$$b_{02} = \frac{-10m^2[a(\beta-1)+1] + 24(\beta+2m)(\beta-1)}{5\beta(m+\beta)^2} + \frac{3}{\beta} + 2a,$$

and

$$\begin{aligned} a(\alpha) &= \langle [A(\alpha) - A_0] q_0, p_0 \rangle, \\ &= [\frac{5(m+\beta)}{4(\beta-1)} + \frac{3m}{m+\beta}] [a(\beta-1)+1] - \frac{24(\beta-1)}{5(m+\beta)} - 3\beta a + 2, \\ b(\alpha) &= \langle [A(\alpha) - A_0] q_1, p_0 \rangle, \\ &= [\frac{5(m+\beta)}{4(\beta-1)} + \frac{3m}{m+\beta}] [a(\beta-1)+1] - \frac{36(\beta-1)}{5(m+\beta)} - 3\beta a - 1, \\ c(\alpha) &= \langle [A(\alpha) - A_0] q_0, p_1 \rangle, \\ &= -[\frac{5(m+\beta)}{4(\beta-1)} + \frac{2m}{m+\beta}] [a(\beta-1)+1] + \frac{16(\beta-1)}{5(m+\beta)} + 2\beta a, \\ d(\alpha) &= \langle [A(\alpha) - A_0] q_1, p_1 \rangle, \\ &= -[\frac{5(m+\beta)}{4(\beta-1)} + \frac{2m}{m+\beta}] [a(\beta-1)+1] + \frac{24(\beta-1)}{5(m+\beta)} + 2\beta a + 2. \end{aligned}$$

Clearly, $a(\alpha_0) = b(\alpha_0) = c(\alpha_0) = d(\alpha_0) = 0$. Denoting the matrix

$$B(\alpha) = \begin{pmatrix} 1 + b(\alpha) & 0 \\ -a(\alpha) & 1 \end{pmatrix}$$

with the nonsingular linear coordinate transformation

$$\begin{pmatrix} u_n \\ v_n \end{pmatrix} = B(\alpha) \begin{pmatrix} x_n \\ y_n \end{pmatrix}, \tag{3.26}$$

system (3.24) can be written as

$$\begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ \varepsilon(\alpha) & -1 + \sigma(\alpha) \end{pmatrix} \begin{pmatrix} x_n \\ y_n \end{pmatrix} + \begin{pmatrix} G \\ H \end{pmatrix},$$
(3.27)

where $\varepsilon(\alpha) = d(\alpha) + b(\alpha)c(\alpha) - a(\alpha)d(\alpha), \ \sigma(\alpha) = a(\alpha) + d(\alpha),$

$$\begin{pmatrix} G \\ H \end{pmatrix} = B^{-1}(\alpha) \begin{pmatrix} f_1((1+b(\alpha))x_n, -a(\alpha)x_n + y_n, \alpha) + o(r_4^3) \\ f_2((1+b(\alpha))x_n, -a(\alpha)x_n + y_n, \alpha) + o(r_4^2) \end{pmatrix}.$$
(3.28)

Mathematical Biosciences and Engineering

19605

Set

$$\beta = \begin{pmatrix} \beta_1 \\ \beta_2 \end{pmatrix} = \begin{pmatrix} \varepsilon(\alpha) \\ \sigma(\alpha) \end{pmatrix}; \tag{3.29}$$

then $\beta_1(\alpha_0) = \beta_2(\alpha_0) = 0$. System (3.27) can be rewritten in the following form:

$$\begin{pmatrix} x_{n+1} \\ y_{n+1} \end{pmatrix} = \begin{pmatrix} -1 & 1 \\ \beta_1 & -1 + \beta_2 \end{pmatrix} \begin{pmatrix} x_n \\ y_n \end{pmatrix} + \begin{pmatrix} G(x_n, y_n, \alpha) \\ H(x_n, y_n, \alpha) \end{pmatrix}.$$
(3.30)

Expand G and H into Taylor series with respect to x_n and y_n to get

$$\begin{split} G(x_n, y_n, \alpha) &= g_{20} x_n^2 + g_{11} x_n y_n + g_{02} y_n^2 + o(r_5^2), \\ H(x_n, y_n, \alpha) &= h_{20} x_n^2 + h_{11} x_n y_n + h_{02} y_n^2 + o(r_5^2), \end{split}$$

where $r_5 = \sqrt{x_n^2 + y_n^2}$,

 $g_{20} = a_{20}(1+b(\alpha)) - a_{11}a(\alpha) + \frac{a^{2}(\alpha)a_{02}}{1+b(\alpha)}, g_{11} = a_{11} - \frac{2a(\alpha)a_{02}}{1+b(\alpha)}, g_{02} = \frac{a_{02}}{1+b(\alpha)}, h_{20} = (1+b(\alpha))^{2}b_{20} + (a_{20} - b_{11})(1+b(\alpha))a(\alpha) + (b_{02} - a_{11})a^{2}(\alpha) + \frac{a^{3}(\alpha)a_{02}}{1+b(\alpha)}, h_{11} = a_{11}a(\alpha) - \frac{a^{2}(\alpha)a_{02}}{1+b(\alpha)} - 2a(\alpha)b_{02} + b_{11}(1+b(\alpha)),$ $h_{02} = b_{02} + \frac{a(\alpha)a_{02}}{1+b(\alpha)}$ and $g_{jk} = h_{jk} = 0$ for all nonnegative integers j, k and j + k = 3. Given the transformation

$$\begin{cases} x_n = u_n + \sum_{2 \le j+k \le 3} \phi_{jk}(\beta) u_n^J v_n^k, \\ y_n = v_n + \sum_{2 \le j+k \le 3} \psi_{jk}(\beta) u_n^j v_n^k, \end{cases}$$
(3.31)

with $\phi_{03} = \psi_{03}, \phi_{20} = \frac{1}{2}g_{20} + \frac{1}{4}h_{20}, \phi_{11} = \frac{1}{2}g_{20} + \frac{1}{2} + g_{11} + \frac{1}{2}h_{20} + \frac{1}{4}h_{11}, \phi_{02} = \frac{1}{4}g_{11} + \frac{1}{2}g_{02} + \frac{1}{8}h_{20} + \frac{1}{4}h_{11} + \frac{1}{4}h_{02},$ $\psi_{20} = \frac{1}{2}h_{20}, \psi_{11} = \frac{1}{2}h_{20} + \frac{1}{2}h_{11}, \psi_{02} = \frac{1}{4}h_{11} + \frac{1}{2}h_{02}, \phi_{30} = \phi_{21} = \phi_{12} = 0 \text{ and } \psi_{30} = \psi_{21} = \psi_{12} = 0, \text{ we can}$ obtain the following 1:2 resonance normal form:

$$\binom{u_{n+1}}{v_{n+1}} = \binom{-1}{\beta_1} \frac{1}{-1+\beta_2} \binom{u_n}{v_n} + \binom{0}{C(\beta)u_n^3 + D(\beta)u_n^2v_n} + o(r_6^3),$$
(3.32)

where $r_6 = \sqrt{u_n^2 + v_n^2}$,

$$\begin{split} C(\beta_1(\alpha_0),\beta_2(\alpha_0)) =& h_{30}(0) + g_{20}(0)h_{20}(0) + \frac{1}{2}h_{20}^2(0) + \frac{1}{2}h_{20}(0)h_{11}(0), \\ D(\beta_1(\alpha_0),\beta_2(\alpha_0)) =& h_{21}(0) + 3g_{30}(0) + \frac{1}{2}g_{20}(0)h_{11}(0) + \frac{5}{4}h_{20}(0)h_{11}(0) + h_{20}(0)h_{02}(0) \\ &\quad + 3g_{20}^2(0) + \frac{5}{2}g_{20}(0)h_{20}(0) + \frac{5}{2}g_{11}(0)h_{20}(0) + h_{20}^2(0) + \frac{1}{2}h_{11}^2(0), \\ g_{20}(0) =& \frac{26161\beta^2 + 46900\beta - 2456}{52\beta(1-\beta)^2}, g_{11}(0) = \frac{2629\beta^2 - 3505\beta - 236}{52\beta(1-\beta)^2}, \\ h_{20}(0) =& \frac{-1911\beta^2 + 13050\beta - 836}{2\beta(1-\beta)^2}, h_{11}(0) = \frac{36722\beta^2 - 47858\beta - 908}{52\beta(1-\beta)^2}, \\ h_{02}(0) =& \frac{-1828\beta^2 + 1486\beta + 92}{52\beta(1-\beta)^2}. \end{split}$$

Thus

 $C(\beta_1(\alpha_0),\beta_2(\alpha_0)) = \frac{(19679\beta^2 + 192621\beta - 13778)(-1911\beta^2 + 13050\beta - 836)}{1942344}$ $104\beta^2(1-\beta)^4$

Mathematical Biosciences and Engineering

 $D(\beta_1(\alpha_0), \beta_2(\alpha_0)) = \frac{630616082\beta^4 + 19939696470\beta^3 + 49520266820\beta^2 - 5731104734\beta + 1730000030820}{(52\beta(1-\beta)^2)^2}.$ By Theorem 9.3 in [29], for all sufficiently small $|\beta|$, the second iteration of (3.32) can be represented in the following form:

$$\begin{pmatrix} u \\ v \end{pmatrix} \mapsto \varphi_{\beta}^{1}(u, v) + o(r_{7}^{3}), \tag{3.33}$$

where $r_7 = \sqrt{u^2 + v^2}$ and $\varphi_{\beta}^1(u, v)$ is the time-one flow of a planar system, which is smoothly equivalent to

$$\begin{pmatrix} \dot{u}_1 \\ \dot{v}_1 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ \gamma_1(\beta) & \gamma_2(\beta) \end{pmatrix} \begin{pmatrix} u_1 \\ v_1 \end{pmatrix} + \begin{pmatrix} 0 \\ C_1(\beta)u_1^3 + D_1(\beta)u_1^2v_1 \end{pmatrix},$$
(3.34)

where

$$\gamma_1(\beta) = 4\beta_1 + o(|\beta_1|), \gamma_2(\beta) = -2\beta_1 - 2\beta_2 + o(|\beta_1|)$$

and

$$C_1(\beta(\alpha_0)) = 4C(\beta(\alpha_0)), D_1(\beta(\alpha_0)) = -2D(\beta(\alpha_0)) - 6C(\beta(\alpha_0)).$$

Assume that $C_1(\beta(\alpha_0)) \neq 0$ and $D_1(\beta(\alpha_0)) \neq 0$. These conditions can be expressed in terms of the normal form coefficients:

$$C(\beta_1(\alpha_0), \beta_2(\alpha_0)) \neq 0, \tag{3.35}$$

$$D(\beta_1(\alpha_0), \beta_2(\alpha_0)) + 3C(\beta_1(\alpha_0), \beta_2(\alpha_0)) \neq 0.$$
(3.36)

Notice that $C(\beta_1(\alpha_0), \beta_2(\alpha_0)) \neq 0$ is equivalent to

$$19679\beta^2 + 192621\beta - 13778 \neq 0$$

and

$$-1911\beta^2 + 13050\beta - 836 \neq 0.$$

Under the condition that $0 < \beta < \frac{4}{9}$, $19679\beta^2 + 192621\beta - 13778 \neq 0 \leftrightarrow \beta \neq \beta_1^* =: \frac{\sqrt{38187398689}}{39358} - \frac{17511}{3578}$ and $-1911\beta^2 + 13050\beta - 836 \neq 0 \leftrightarrow \beta \neq \beta_2^* =: \frac{2175}{637} - \frac{\sqrt{40978029}}{1911}$). Assume the following:

$$y = -2302696300\beta^4 + 11259249552\beta^3 + 246359675612\beta^2 - 32316161102\beta + 173089846664.$$

Then, (3.36) holding is equivalent to $y \neq 0$. It follows from the following Figure 1 that y > 0 for $0 < \beta < \frac{4}{9}$, which means that $D(\beta_1(\alpha_0), \beta_2(\alpha_0)) + 3C(\beta_1(\alpha_0), \beta_2(\alpha_0)) > 0$. Therefore, when $\beta \neq \beta_1^*$ and $\beta \neq \beta_2^*$, the non-degeneracy conditions $C_1(\beta_1(\alpha_0), \beta_2(\alpha_0)) \neq 0$ and $D_1(\beta_1(\alpha_0), \beta_2(\alpha_0)) \neq 0$ are satisfied; so, the fixed point E_* of system (1.2) is a 1:2 resonance bifurcation point.

By scaling the variables, parameters and time in (3.34), we obtain the following system:

$$\begin{cases} \dot{\zeta}_1 = \zeta_2, \\ \dot{\zeta}_2 = \varepsilon_1 \zeta_1 + \varepsilon_2 \zeta_2 + s \zeta_1^3 - \zeta_1^2 \zeta_2, \end{cases}$$
(3.37)

where $s = signC(\beta_1(\alpha_0), \beta_2(\alpha_0)) = \pm 1$. If s = 1, then, for the approximating system (3.37), the following holds:

Mathematical Biosciences and Engineering



Figure 1. The plot of the function y- β .

(i) there is a pitchfork bifurcation which occurs on the curve

$$PF = \{(\varepsilon_1, \varepsilon_2) : \varepsilon_1 = 0\};$$

(ii) there is a non-degenerate Neimark-Sacker bifurcation which occurs on the curve $NS = \{(\varepsilon_1, \varepsilon_2) : \varepsilon_2 = 0, \varepsilon_1 < 0\};$

(iii) there is a heteroclinic bifurcation which occurs on

 $HL = \{(\varepsilon_1, \varepsilon_2) : \varepsilon_2 = -\frac{1}{5}\varepsilon_1 + o(\varepsilon_1), \varepsilon_1 < 0\}.$



Figure 2. (a) Bifurcation diagram for the approximating system (3.37) for s = 1; (b) heteroclinic cycle; (c) phase portraits for system (3.37) under different cases.

The bifurcation diagram for (3.37) for s = 1 is presented in Figure 2(a). Because of the Z_2 -symmetry, the heteroclinic orbits connecting the two saddles appear simultaneously, forming a

heteroclinic cycle upon crossing the curve HL (see Figure 2(b)). Moreover, the corresponding phase portraits of the contents of Figure 2(a) are showed in Figure 2(c). For the details, refer to [30]. **Remark.** If system (1.2) undergoes a 1:2 resonance bifurcation near the positive fixed point E_* , system (1.2) is strongly unstable. The predator and the prey will increase rapidly and breakout suddenly. For this, refer also to [26].

4. Numerical simulation

In this section, to illustrate our theoretical results and reveal some new dynamical behaviors in system (1.2), we present the bifurcation diagrams, phase portraits and maximal Lyapunov exponents that were derived for specific parameter values by using Matlab software. First, for the fixed point



Figure 3. Bifurcation of system (1.2) in (*a*, *x*)-plane and maximal Lyapunov exponents for $m = 0.5, \beta = 1.6$.

 $E_2(\frac{a-1}{a}, 0)$, we chose the parameters to be $a \in (2.8, 4)$, m = 0.5 and $\beta = 1.6$, and the initial values as $(x_0, y_0) = (0.1, 0.1)$. Because the bifurcation diagram for the (a, x)-plane is similar to that for the (a, y)-plane, we will only show the former. Figure 3(a) shows that the fixed point E_2 is stable for a < 3, and that the period-doubling bifurcation occurs at a = 3 (the multipliers are $\lambda_1 = -1$ and $|\lambda_2| \neq 1$). Meanwhile E_2 becomes unstable when a > 3, which is in accordance with the result in Theorem 3.2. Moreover, a chaotic set emerges with the increase of the value of the parameter a. Figure 3(b) shows the spectrum of the maximum Lyapunov exponents with respect to the parameter $a \in (2.8, 4)$ when m = 0.5and $\beta = 1.6$; it also shows that the maximal Lyapunov exponents are positive, that is to say, chaos may occur. Figure 4(a) shows that periods of 2, 4, 8, etc., will occur, which means that chaos will occur (period-doubling bifurcation to chaos). However, chaos is a ubiquitous nonlinear phenomenon that leads to undesirablity. So controlling chaos to make dynamic behavior predictable is essential. How can one control chaos? The main methods of chaos control are state feedback, pole-placement and hybrid control. For this, refer to [34,35]. From Figure 4(b),(c), we observe that, when the value of the growth rate a of prey is close to 3.63 or 3.74, the population density distribution is sparse. Meanwhile, the prey population may become extinct when the parameter a increases to 4 according to Figure 4(d). The phase portrait is given in Figure 5. From Figure 5(a), (b), the predator population is gradually decreasing with a small number representing two species coexistence. And, when the parameter a =2.8 or a = 3.5, the prey population is near 0.643 or 0.42 (0.85) respectively. In Figure 5(c), the predator population is decreasing to extinction and there is only intraspecific competition for prey. The density

Mathematical Biosciences and Engineering



Figure 4. Bifurcation of system (1.2) in (a, x)-plane versus *a* for $m = 0.5, \beta = 1.6$.



Figure 5. Phase portraits of system (1.2) in (x, y)-plane versus *a* for m = 0.5, $\beta = 1.6$.

of the prey population ranges from 0 to 1. The excessive growth rate of the prey first reduces the density of the predator population until extinction; it then causes them to compete among themselves until they reach balance. We set the parameters as m = 0.5 and a = 3.5 and varied the value of the parameter b to get Figures 6. One finds that the prey population and predator population coexist at $\beta = 0.5$, and that the two species are periodic and coexist on two closed orbits at $\beta = 0.6$. Then the predator population decreases and the prey population gradually becomes stable.



Figure 6. Phase portraits of system (1.2) in (x, y)-plane versus β for m = 0.5, a = 3.5.

Next, one can see that two eigenvalues of system (1.2) are $\lambda_1 = 1$ and $\lambda_2 = -1$ when a = 3 and $\beta = \frac{2}{3}$. We call the fixed point $E_2(\frac{2}{3}, 0)$ a fold-flip bifurcation point. The bifurcation diagram for system (1.2) is plotted in Figure 7. As we can see, when the parameter a = 3 and the parameter β varies from 0.5 to 0.95, the prey population keeps increasing until it is split into two parts at the parameter $\beta = \frac{2}{3}$; meanwhile, the predator population gradually decreases to extinction. For the codim 2 bifurcation diagram and corresponding details, refer also to [29].

Finally, we applied the parameters $a \in (3.6, 4.15), m = 0.8$ and $\beta = 0.5$ and the initial values $(x_0, y_0) = (0.2, 3.1)$ for the fixed point E_* . We constructed Figure 8(a) and determined the existence of period-doubling bifurcation when $a = a_0 = 3.79$. Figure 8(b) represents the maximum Lyapunov exponents under the conditions of $a \in (3.6, 4.15), m = 0.8$ and $\beta = 0.5$; it indicates that chaos probably occurs in the system when the parameter *a* is beyond 4.15. According to the result in Theorem 3.4, one has K = -2.6398 < 0; thus, the period-doubling bifurcation is subcritical and the system is unstable. From Figure 9, the phase portraits demonstrate that the prey population and predator population can coexist at a = 3.6, 3.8. When the parameter *a* increases continuously, the predator population will decrease to extinction. Thus, the period-doubling orbit that is bifurcated from E_* is unstable, which



Figure 7. Bifurcation of system (1.2) in (β, x) -plane (a) and (β, y) -plane (b) for m = 0.5, a =3.



Figure 8. Bifurcation of system (1.2) in (a, x)-plane and maximal Lyapunov exponents for $m = 0.8, \beta = 0.5.$



Figure 9. Phase portraits of system (1.2) in (*x*, *y*)-plane versus *a* for m = 0.8, $\beta = 0.5$.

Mathematical Biosciences and Engineering

will break the population balance.

5. Discussion and conclusions

In this paper, we revisit a discrete-time prey-predator model with the Allee effect in prey. Besides correcting the errors for the local stability at the fixed point E_* , we have mainly studied the bifurcation problems of the fixed points $E_1(0,0)$, $E_2(\frac{a-1}{a},0)$ and $E_*(\beta, \frac{(\beta+m)(a(1-\beta)-1)}{\beta})$ of system (1.2) to find some complex bifurcation results, which are ecologically important. Especially, for the fixed point E_* , we observed the existence of the codimension one period-doubling bifurcation and Neimark-Sacker bifurcation and codimension two 1:2 strong resonance bifurcation. The 1:2 strong resonance bifurcation observed to occur at the fixed point E_* implies that the discrete system may undergo pitchfork bifurcation, Neimark-Sacker bifurcation and heteroclinic bifurcation near E_* , and this will lead to a potential dramatic variation of the system dynamics. The conditions for these bifurcations to occur have been given, and interesting dynamical properties have also been obtained through numerical simulations. Our results clearly demonstrate the dynamics of the system: as the parameter β increases, the two species change from the coexistence on two closed orbits to gradual extinction of the predator population; at the same time, the value of the parameter a controls the prey and predator population. Moreover, by taking appropriate value of Allee effect, 1:2 strong resonance bifurcation at fixed point E_* will occur, which indicates that the two species will increase rapidly and breakout suddenly [29]. However, the fold-flip bifurcation at E_2 and chaos control that have not been solved in this paper are still interesting problems, and they will be considered in our future work.

Use of AI tools declaration

The authors declare that they have not used artificial intelligence tools in the creation of this article.

Acknowledgments

This work was partly supported by the National Natural Science Foundation of China (61473340), Distinguished Professor Foundation of Qianjiang Scholars in Zhejiang Province (F703108L02) and Natural Science Foundation of Zhejiang University of Science and Technology (F701108G14).

Conflict of interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

References

1. A. J. Lotka, Undamped oscillations derived from the law of mass action, *J. Am. Chem. Soc.*, **42** (1920), 1595–1599.

- 2. V. Volterra, Fluctuations in the abundance of species considered Mathematically, *Nature*, **118** (1926), 558–560. http://dx.doi.org/10.1038/118558a0
- 3. C. Robinson, *Dynamical Systems: Stability, Symbolic dynamics, and Chaos*, CRC Press, London, 1995.
- 4. J. M. Smith, Mathematical Ideas in Biology, Cambridge University Press, Cambridge, 1968.
- 5. L. H. A. Monteiro, P. N. Mustaro, Hero's journey in bifurcation diagram, *Commun, Nonl. Sci. Numer. Simul.*, **17** (2012), 2233–2236. https://doi.org/10.1016/j.cnsns.2011.09.035
- H. J. Viljoen, J. E. Gatica, H. Vladimir, Bifurcation analysis of chemically driven convection, *Chem. Eng. Sci.*, 45 (1990), 503–517. https://doi.org/10.1016/0009-2509(90)87037-S
- 7. C. Wang, X. Li, Further investigations into the stability and bifurcation of a discrete predator–prey model, *J. Math. Anal. Appl.*, **422** (2015), 920–939. https://doi.org/10.1016/j.jmaa.2014.08.058
- 8. S. Ruan, D. Xiao, Global analysis in a predator–prey system with nonmononotonic functional response, *SIAM. J. Appl. Math.*, **61** (2001), 1445–1472. https://doi.org/10.1137/S0036139999361896
- 9. O. P. Misra, P. Sinha, C. Singh, Stability and bifurcation analysis of a preypredator model with age based predation, *Appl. Math. Model.*, **37** (2013), 6519–6529. https://doi.org/10.1016/j.apm.2013.01.036
- 10. S. Li, Z. Xiong, Bifurcation analysis of a predator-prey system with sex-structure and sexual favoritism, *Adv. Differ. Equation*, **219** (2013). https://doi.org/10.1186/1687-1847-2013-219
- D. Sen, S. Ghorai, S. Sharma, M. Banerjee, Allee effect in prey's growth reduces the dynamical complexity in prey-predator model with generalist predator, *Appl. Math. Model.*, **91** (2021), 768– 790. https://doi.org/10.1016/j.apm.2020.09.046
- 12. J. Li, G. Q. Sun, Z. Jin, Interactions of time delay and spatial diffusion induce the periodic oscillation of the vegetation system, *DCDS-B*, **27** (2022), 2147–2172, https://doi.org/10.3934/dcdsb.2021127
- P. R. Chowdhury, M. Banerjee, S. Petrovskii, Canards, relaxation oscillations, and pattern formation in a slow-fast ratio-dependent predator-prey system, *Appl. Math. Model.*, **109** (2022), 519–535. https://doi.org/10.1016/j.apm.2022.04.022
- G. Q. Sun, L. Li, J. Li, C. Liu, Y. Wu, S. Gao, et. al, Impacts of climate change on vegetation pattern: Mathematical modeling and data analysis, *Phys. Life Rev.*, 43 (2022), 239– 270. https://doi.org/10.1016/j.plrev.2022.09.005
- S. N. Chowdhury, J. Banerjee, M. Perc, D. Ghosh, Eco-evolutionary cyclic dominance among predators, prey, and parasites, *J. Theor. Bio.*, 564 (2023), 111446. https://doi.org/10.1016/j.jtbi.2023.111446
- 16. G. Q. Sun, H. Zhang, Y. Song, L. Li, Z. Jin, Dynamic analysis of a plant-water model with spatial diffusion, *JDE*, **329** (2022), 395–430. https://doi.org/10.1016/j.jde.2022.05.009
- 17. Y. Kang, A. A. Yakubu, Weak Allee effects and species coexistence, *Nonl. Anal.*, **12** (2011), 3329–3345. https://doi.org/10.1016/j.nonrwa.2011.05.031
- 18. D. S. Boukal, L. Berec, Single-species models of the Allee effect: extinction boundaries, sex ratios and mate encounters, *J. Theor. Biol.*, **218** (2002), 375–394. https://doi.org/10.1006/jtbi.2002.3084

- 19. C. M. Taylor, A. Hastings, Allee effects in biological invasions, *Ecol. Letters*, **8** (2005), 895–908. https://doi.org/10.1111/j.1461-0248.2005.00787.x
- R. Sophia, J. Jang, Allee effects in a discrete-time host-parasitoid model, J. Diff. Equation Appl., 12 (2006), 165–181. https://doi.org/10.1080/10236190500539238
- H. Liu, K. Zhang, Y. Ye, Y. Wei, M. Ma, Dynamic complexity and bifurcation analysis of a host-parasitoid model with Allee effect and Holling type III functional response, *Adv. Differ. Equation*, **507** (2019). https://doi.org/10.1186/s13662-019-2430-8
- 22. E. Knipling, Possibilities of insect control or eradication through the use of sexually sterile males, *J. Econ. Entomol*, **48** (1955), 459–462. https://doi.org/10.1093/jee/48.4.459
- 23. B. B. Lamont, P. G. Klinkhamer, E. Witkowski, Population fragmentation may reduce fertility to zero in banksia Goodii–a demonstration of the Allee effect, *Oecologia*, **94** (1993), 446–450. https://doi.org/10.1007/BF00317122
- 24. T. Perälä, A. Kuparinen, Detection of Allee effects in marine fishes: analytical biases generated by data availability and model selection, *Proc. R. Soc. B*, **284** (2017), 20171284. http://dx.doi.org/10.1098/rspb.2017.1284
- 25. A. Q. Khan, Neimark–Sacker bifurcation of a two-dimensional discrete-time predator–prey model, *Springer Plus 5*, **126** (2016). https://doi.org/10.1186/s40064-015-1618-y
- F. Kangalgil, Neimark–Sacker bifurcation and stability analysis of a discrete-time prey–predator model with Allee effect in prey, *Adv. Differ. Equation*, **92** (2019). https://doi.org/10.1186/s13662-019-2039-y
- 27. S. Winggins, *Introduction to Applied Nonlinear Dynamical Systems and Chaos*, Springer-Verlag, New York, 2003.
- 28. J. Carr, Application of Center Manifold Theorem, Springer-Verlag, New York, 1981.
- Q. Chen, Z. Teng, F. Wang, Fold-flip and strong resonance bifurcations of a discrete-time mosquito model, *Chaos Solitons Fractals*, **144** (2021), 110704. https://doi.org/10.1016/j.chaos.2021.110704
- 30. Y. A. Kuznetsov, *Elements of Applied Bifurcation Theory*, Springer-Verlag, New York, 1998.
- L. G. Yuan, Q. G. Yang, Bifurcation invariant curve and hybrid control in a discrete-time predator-prey system, *Appl. Math. Model*, **39** (2015), 2345–2362. https://doi.org/10.1016/j.apm.2014.10.040
- 32. B. Li, Z. M. He, Bifurcations and chaos in a two-dimensional discrete Hindmarsh–Rose model. *Nonlinear Dyn.*, **76** (2014), 697–715. https://doi.org/10.1007/s11071-013-1161-8
- 33. B. Li, Z. M. He, Yang, 1:2 and 1:4 resonances in a two-dimensional discrete Hindmarsh-Rose model, *Nonlinear Dyn.*, **79** (2015), 705–720. https://doi.org/10.1007/s11071-014-1696-3
- A. Singh, V. S. Sharma, Bifurcations and chaos control in a discrete-time prey-predator model with Holling type-II functional response and prey refuge, *J. Comp. Appl. Mathe.*, 418 (2023), 114666. https://doi.org/10.1016/j.cam.2022.114666
- S. Akhtar, R. Ahmed, M. Batool, N. A. Shah, J. D. Chung, Stability, bifurcation and chaos control of a discretized Leslie prey-predator model, *Chaos Solitons Fractals*, 152 (2021), 111345. https://doi.org/10.1016/j.chaos.2021.111345

Appendix A. Known results for the stability of fixed points

For the fixed point $E_1(0,0)$, the following statements are true.

(*i*.1) $E_1(0, 0)$ is a sink if a < 1;

(*i*.2) $E_1(0, 0)$ is a saddle if a > 1;

(*i*.3) $E_1(0, 0)$ is non-hyperbolic if a = 1.

Assume that a > 1. For the fixed point $E_2(\frac{a-1}{a}, 0)$, the following statements are true.

(*ii*.1) $E_2(\frac{a-1}{a}, 0)$ is a sink if $1 < a < min\{\frac{1}{1-\beta}, 3\}$ and $0 < \beta < 1$;

(*ii*.2) $E_2(\frac{a-1}{a}, 0)$ is a saddle if $(\frac{1}{1-\beta} < a < 3$ and $0 < \beta < \frac{2}{3})$ or $(3 < a < \frac{1}{1-\beta} \text{ and } \frac{2}{3} < \beta < 1)$;

(*ii*.3) $E_2(\frac{a-1}{a}, 0)$ is a source if $a > max\{3, \frac{1}{1-\beta}\}$ and $0 < \beta < 1$;

(*ii.*4) $E_2(\frac{a-1}{a}, 0)$ is non-hyperbolic if a = 3 or $(a = \frac{1}{1-\beta}, 0 < \beta < 1 \text{ and } \beta \neq \frac{2}{3})$.

Assume that $a > \frac{1}{1-\beta}$ and $\beta < 1$; then, for the unique positive fixed point E_* , the following statements are true.

(*iii*.1) E_* is a sink if the following condition holds:

$$\frac{1}{1-\beta} < a < \min\{\frac{1}{1-2\beta-m}, \frac{3\beta+5m}{3\beta^2+m\beta+m-\beta}\};$$

(*iii*.2) E_* is a source if the following condition holds:

$$\frac{1}{1-2\beta-m} < a < \frac{3\beta+5m}{3\beta^2+m\beta+m-\beta};$$

(*iii*.3) E_* is a saddle if the following condition holds:

$$a > \frac{1}{1 - 2\beta - m};$$

(*iii*.4) E_* is a non-hyperbolic if the following condition holds:

$$a = \frac{3\beta + 5m}{3\beta^2 + m\beta + m - \beta} \text{ and } m \neq \frac{4}{5} - \frac{9}{5}\beta;$$

(*iii*.5) E_* is a non-hyperbolic with the conjugate complex roots if the following condition holds:

$$a = \frac{1}{1 - 2\beta - m}$$
 and $m < \frac{4}{5} - \frac{9}{5}\beta$.

Appendix B. Key definition and lemma

Definition 5.1. Let E(x, y) be a fixed piont of the system (1.2) with multipliers λ_1 and λ_2 .

- 1) If $|\lambda_1| < 1$ and $|\lambda_2| < 1$, a fixed point E(x, y) is called a sink; thus, a sink is locally asymptotically stable.
- 2) If $|\lambda_1| > 1$ and $|\lambda_2| > 1$, a fixed point E(x, y) is called a source; thus, a source is locally asymptotically unstable.
- 3) If $|\lambda_1| < 1$ and $|\lambda_2| > 1$ (or $|\lambda_1| > 1$ and $|\lambda_2| < 1$), a fixed point E(x, y) is called a saddle.
- 4) If either $|\lambda_1| = 1$ or $|\lambda_2| = 1$, a fixed point E(x, y) is called non-hyperbolic.

Mathematical Biosciences and Engineering

Lemma 5.2. Let $F(\lambda) = \lambda^2 + B\lambda + C$, where B and C are two real constants. Suppose that λ_1 and λ_2 are two roots of $F(\lambda) = 0$. Then the following statements hold.



© 2023 the Author(s), licensee AIMS Press. This is an open access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/4.0)