A mathematical analysis of Hopf-bifurcation in a prey-predator model with nonlinear functional response

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Abstract

In this paper, our aim is mathematical analysis and numerical simulation of a prey-predator model to describe the effect of predation between prey and predator with nonlinear functional response. First, we develop results concerning the boundedness, the existence and uniqueness of the solution. Furthermore, the Lyapunov principle and the Routh–Hurwitz criterion are applied to study respectively the local and global stability results. We also establish the Hopf-bifurcation to show the existence of a branch of nontrivial periodic solutions. Finally, numerical simulations have been accomplished to validate our analytical findings.

MSC: 65L12; 65M20; 65N40

Keywords: Prey-predator system; Hopf-bifurcation; Global stability; Numerical simulations

1 Introduction

The study of the dynamics relationship of the prey-predator system has long been and will continue to be one of the dominant subjects in both ecology and mathematical ecology due to its universal existence and importance. In recent decades, mathematics has had a huge impact as a tool for modeling and understanding biological phenomena. Mathematical modeling of the population dynamics of a prey-predator system is an important objective of mathematical models in biology, which has attracted the attention of many researchers [1–4]. Many authors, such as Holling 1959 [5], Getz 1984, and Arditi and Ginzburg 1989 [6, 7], studied the prey-predator system with various functional responses. These different types of functional responses present a key element for understanding the dynamics of these populations. The main questions concerning population dynamics concern the effects of interaction in the regulation of natural populations, the reduction of their size, the reduction of their natural fluctuations, or the destabilization of the equilibria in oscillations of the states of the population [8–13]. The predator-prey relationship is important to maintain the balance between different animal species. Without predators, some prey species would force other species to disappear due to competition. Without prey, there

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would be no predators. The main feature of predation is therefore a direct impact of the predator on the prey population.

It is in this line of thought that we are interested here in the study of the dynamics of prey-predator populations with an alternative food resource for predators, meaning that the predator population can survive if there is no prey. Our objective is to understand what is the impact of predation on the dynamics of prey and predator species, in order to avoid any extinction of the two species.

Several authors have studied the prey-predator model with logistics growth in both species. Haque in [14] proposed a prey-predator model with logistic growth in both species and a linear functional response. The author assumed that the predator has logistic growth rate since it has sufficient resources for alternative foods; and it is argued that alternative food sources may have an important role in promoting the persistence of predator-prey systems. Guin in [15] studied a prey-predator model with logistic growth in both species and using ratio-dependent functional for predators.

Motivated by the above works, we consider the following predator-prey model [14]:

$$\begin{align*}
\frac{dN}{dt} &= r_1 \left(1 - \frac{N}{K_1}\right)N - bNP, \quad N(0) = N_0 > 0, \\
\frac{dP}{dt} &= r_2 \left(1 - \frac{P}{K_2}\right)P + ebNP, \quad P(0) = P_0 > 0,
\end{align*}$$

(1.1)

where

- $N(t)$ and $P(t)$ stand for the prey and predator density, respectively, at time $t$.
- $r_1, K_1, b, e$ are positive constants that stand for prey intrinsic growth rate, the prey carrying capacity of the environment, predation rate per unit of time, and conversion rate, respectively.
- The term $bNP$ models prey consumption due to predation.
- $r_2, K_2$ represent respectively the predator intrinsic growth rate and the predator carrying capacity of the environment.

Some similar models have appeared in the recent literature [14, 15]. We remark that the main new distinctive feature is the inclusion of Holling function response of type II. Thus, by incorporating Holling function response of type II, we describe the predation strategy. Indeed, many researchers suggested that Holling type II response is the characteristic of predators. It determines the stability and bifurcation dynamics of the model. Usually, the feeding rate of predator is saturated, so it is more realistic to consider prey dependence functional response. Our model differs from the one of [14], since in the latter the term of predation is linear. In fact, we consider Holling function response of type II defined by

$$\phi(x) = \frac{B\alpha_0 x}{1 + B\alpha_1 x},$$

where

- $\alpha_0$ and $\alpha_1$ represent respectively the search and capture time of the prey,
- $B$ is the predation rate per unit of time.

Indeed, physiological absorption capabilities of prey are limited, and even if a large number of prey is available, a predator will not be able to absorb prey numbers beyond this limit, it is more realistic to design a response function with a saturation effect with the density of prey. Thus, Holling function response of type II is more appropriate. In order to sustain the
coexistence of ecosystem species, it is very important to control some key demographic parameters.

The paper is organized as follows. In Sect. 2, we present the general mathematical model of the prey-predator system. Section 3 provides the mathematical analysis of the model established in Section 2. We perform some numerical simulations to support our main results in Section 4. A final discussion concludes the paper.

2 Mathematical model formulation

In this section, we proceed to the construction of the prey-predator model by taking into account the fact that the predator has an alternative source of food. Our main goal is to modify system (1.1) in order to describe the effect of predation on the prey. Our task here is to analyze the impact of predation on a predator-prey community [16–19].

The following hypotheses hold for our models:

H$_1$: prey populations follow a logistic growth in the absence of the predator;
H$_2$: functional response of the predator is Holling type II;
H$_3$: the predator has an alternative source of food.

The system modeling the evolution over time of prey and predators is given by

\[
\begin{align*}
\frac{dN}{dt} &= \psi(N) - \phi(N)P, \\
\frac{dP}{dt} &= g_0(P) + g_1(N, P),
\end{align*}
\]

(2.1)

where

- \( \psi, \phi, g_0, \) and \( g_1 \) are positive functions and \( C^\infty \);
- \( \psi(N), g_0(P) \) is a growth function of prey and predator population, respectively;
- \( \phi(N) \) is the amount of prey consumed by a predator per time unit;
- \( g_1(N, P) \) represents the rate of conversion of the prey into the predator.

The model presented here is general, and it is necessary to make choices, particularly for the functions \( g_0(P), g_1(N, P), \phi(N), \) and \( \psi(N). \) Then we make the following choices:

- \( \psi(N) = r_1(1 - \frac{N}{K_1})N \) represents the dynamics of the prey population governed by the logistic equation when there is no predator;
- \( g_0(P) = r_2(1 - \frac{P}{K_2})P \) represents the logistic growth of the predator population when there is no prey;
- \( \phi(N) = \frac{\delta_1 N}{1 + \delta_2 N} \) represents the functional response of the predator which is Holling type II;
- \( g_1(N, P) = \frac{\omega N P}{1 + \delta_3 N} \) represents the quantity of prey consumed by predators.

Consequently, we obtain the following nonlinear differential system defined by

\[
\begin{align*}
\frac{dN}{dt} &= r_1(1 - \frac{N}{K_1})N - \frac{\delta_1 N P}{1 + \delta_2 N}, \quad N(0) = N_0 > 0, \\
\frac{dP}{dt} &= r_2(1 - \frac{P}{K_2})P + \frac{\delta_3 N P}{1 + \delta_2 N}, \quad P(0) = P_0 > 0,
\end{align*}
\]

(2.2)

where

- \( r_1, r_2 > 0 \) are respectively the prey and predator growth rates;
- \( K_1, K_2 > 0 \) represent respectively the carrying capacity of the prey and the predator;
- \( \delta_1 \) and \( \delta_2 \) represent respectively predator search and satiety rates;
- \( e = \frac{\delta_3}{\delta_1} \) represents the conversion rate of prey biomass into predatory biomass, with \( 0 < e < 1. \)
• $\frac{\delta_1 NP}{1 + \delta_2 N}$ represents the quantity of prey taken by predators per unit of time;
• $\frac{\omega NP}{1 + \delta_2 N}$ represents the amount of prey consumed by predators per unit of time;
• $(\delta_1 - \omega)\frac{NP}{1 + \delta_2 N}$ is a residual term and represents the quantity of prey taken by predators and which did not contribute to the growth of predators.

Thus, we obtain the following interaction diagram: Fig. 1.

Using the above assumptions and according to Figure 1, at any time $t > 0$, the dynamics of the system can be represented by the following set of differential equations:

$$
\begin{align*}
G_1(N, P) &= \frac{dN}{dt} = r_1 \left(1 - N \frac{K_1}{r_1}\right) N - \frac{\delta_1 NP}{1 + \delta_2 N}, \quad N(0) = N_0 > 0, \\
G_2(N, P) &= \frac{dP}{dt} = r_2 \left(1 - \frac{P}{K_2}\right) P + \frac{\omega NP}{1 + \delta_2 N}, \quad P(0) = P_0 > 0.
\end{align*}
$$

(2.3)

3 Mathematical analysis

This section deals with mathematical analysis including the stability and the bifurcation analysis of system (2.3) [2, 8, 15, 20–22].

Then we rewrite model (2.3) in the following form:

$$
\dot{X}(t) = G(X(t)),
$$

where $X(t) = (N(t), P(t))^T$ and $G$ is defined on $\mathbb{R}^2$ by

$$
G(X) = \begin{pmatrix} G_1(N, P) \\ G_2(N, P) \end{pmatrix} = \begin{pmatrix} r_1 \left(1 - N \frac{K_1}{r_1}\right) N - \frac{\delta_1 NP}{1 + \delta_2 N} \\ r_2 \left(1 - \frac{P}{K_2}\right) P + \frac{\omega NP}{1 + \delta_2 N} \end{pmatrix}.
$$

The preliminary results concern the existence, positiveness, and boundedness of solutions of system (2.3).

3.1 Existence, positiveness, and boundedness of solutions

From the biological point of view, it is important to show the existence, positivity, and boundedness of the solution of system (2.3) [9, 19, 23, 24].

Proposition 1 System (2.3) admits a unique global solution $(N(t), P(t))$ defined on the interval $[0, T_{\text{max}}[$. Moreover, the set

$$
A := \{(N, P) \in \mathbb{R}_+^2 / 0 \leq N \leq K_1, 0 \leq P \leq K_p\} \quad \text{with} \ K_p = K_2 \left(1 + \frac{\omega}{r_2 \delta_2}\right)
$$

for
is positively invariant and absorbing for system (2.3).

Proof. Indeed,

• the theorem of Cauchy–Lipschitz [11] assures the existence and uniqueness of local solution of system (2.3) on \([0, T_{\text{max}}]\) given the regularity of the functions involved in the model.

• Now, let us show that the set \(A = \{(N, P) \in \mathbb{R}^2 / 0 \leq N \leq K_1, 0 \leq P \leq K_P\}\) is positively invariant and absorbing for system (2.3).

Let us show that

\[
A_0 = \{(N, P) \in \mathbb{R}^2 / 0 \leq N \leq K_1\} \quad \text{and} \quad B = \{(N, P) \in \mathbb{R}^2 / 0 \leq P \leq K_P\}
\]

are positively invariant and absorbing for system (2.3).

Let us prove that

\[
A_1 = \{(N, P) \in \mathbb{R}^2 / N \geq 0\} \quad \text{(3.1)}
\]

is positively invariant. Indeed, let \(f_1\) be the function defined on \(\mathbb{R}^2\) by \(f_1(N, P) = -N\). We have

\[
\nabla f_1(N, P) = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \neq 0_{\mathbb{R}^2} \quad \text{and} \quad \langle \nabla f_1 | G \rangle = 0 \times r_2 \left( 1 - \frac{P}{K_2} \right) P = 0.
\]

Thus, \(\langle \nabla f_1 | G \rangle \leq 0\) on \(\{(N, P) \in \mathbb{R}^2 / N = 0\}\), where \(\langle \rangle\) is the usual scalar product. Therefore, \(A_1\) is positively invariant.

Proceeding in the same way, with \(f_2(N, P) = -P\), we show that

\[
A_2 = \{(N, P) \in \mathbb{R}^2 / P \geq 0\} \quad \text{(3.2)}
\]

is positively invariant.

Let us show that

\[
A_3 = \{(N, P) \in \mathbb{R}^2 / N \leq K_1\}. \quad \text{(3.3)}
\]

Indeed, let \(f_3\) be defined on \(\mathbb{R}^2\) by \(f_3(N, P) = N - K_1\). We have

\[
\nabla f_3(N, P) = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \neq 0_{\mathbb{R}^2} \quad \text{and} \quad \langle \nabla f_3 | G \rangle = -\frac{\delta_1 K_1 P}{1 + \delta_2 K_1}.
\]

Thus, \(\langle \nabla f_3 | G \rangle \leq 0\) on \(\{(N, P) \in \mathbb{R}^2 / N = K_1\}\). Therefore, \(A_3\) is positively invariant.

According to (3.1) and (3.3), \(A_0\) is positively invariant.

Now, we aim to show that the set \(A_0\) is absorbing. The variable \(N\) satisfies the inequality

\[
\frac{dN}{dt} \leq r_1 \left( 1 - \frac{N}{K_1} \right) N,
\]
and by the principle of comparison, we deduce that \( \lim_{t \to +\infty} \sup N(t) \leq K_1 \). Hence, for \( \epsilon > 0 \), there exists \( T > 0 \) such that \( N(t) \leq \sup_{t \geq T} N(t) \leq K_1 + \epsilon \); as \( \epsilon \) is arbitrary, we deduce that \( A_0 \) is absorbing.

Now, we aim to show that \( P \leq K_p \). Indeed,

\[
\frac{dP}{dt} = r_2 \left( 1 - \frac{P}{K_2} \right) P + \frac{\omega NP}{1 + \delta_2 N} \quad \text{and} \quad \frac{e\delta_1 N}{1 + \delta_2 N} < \frac{e\delta_1}{\delta_2},
\]

thus we have the following differential inequality:

\[
\frac{dP}{dt} \leq \left( r_2 \left( 1 - \frac{P}{K_2} \right) + \frac{e\delta_1}{\delta_2} \right) P, \quad \forall t \geq t_0. \tag{3.4}
\]

According to the comparison principle, we deduce that

\[
\lim_{t \to +\infty} \sup (P(t)) \leq K_2 \left( 1 + \frac{\alpha}{r_2 \delta_2} \right) = K_p. \tag{3.5}
\]

According to (3.2) and (3.5), \( B \) is positively invariant and absorbing.

From the above result, the set defined by \( A = \{(N, P) \in \mathbb{R}_+^2 / 0 \leq N \leq K_1, 0 \leq P \leq K_p \} \) is positively invariant and absorbing.

To show the global existence of solutions, we must show that the solutions of the system are bounded. In the previous demonstration we have established that \( N \) and \( P \) are bounded. Thus, we can conclude that the solutions of system (2.3) exist globally.

For the study of system (2.3), we restrain a set defined by

\[
A = \{(N, P) \in \mathbb{R}_+^2 / 0 \leq N \leq K_1, 0 \leq P \leq K_p \}.
\]

### 3.2 Stability analysis of the equilibria

In this section, we analyze the local and global stability of different equilibrium.

#### 3.2.1 Trivial equilibrium points of the model

The trivial stationary states of system (2.3) are given in the following proposition [8, 11–13].

**Proposition 2** The equilibrium states are as follows:

(i) \( E_0 = (0, 0) \), the predators and prey are extinct. This equilibrium is always admissible.

(ii) \( E_1 = (0, K_2) \), the prey is extinct. This equilibrium is always admissible.

(iii) \( E_2 = (K_1, 0) \), the predator is extinct. This equilibrium is always admissible.

**Proof** Indeed, to get the equilibrium points, we solve the following system:

\[
\begin{aligned}
& r_1 \left( 1 - \frac{N}{K_1} \right) N - \frac{\delta_1 NP}{1 + \delta_2 N} = G_1(N, P) = 0, \\
& r_2 \left( 1 - \frac{P}{K_2} \right) P + \frac{\omega NP}{1 + \delta_2 N} = G_2(N, P) = 0.
\end{aligned} \tag{3.6}
\]

(i) We have \( G_1(0, 0) = G_2(0, 0) = 0 \). Thus, \( E_0 = (0, 0) \) is the trivial equilibrium point.
(ii) In the same way, \( G_1(0, K_2) = G_2(0, K_2) = 0 \). Then \( E_1 = (0, K_2) \) is an equilibrium point of system (2.3).

(iii) We also have \( G_1(K_1, 0) = G_2(K_1, 0) = 0 \). Then \( E_2 = (K_1, 0) \) is an equilibrium point of system (2.3).

The local stability analysis of a trivial equilibrium point is given by the following proposition.

**Proposition 3**

(i) \( E_0 \) and \( E_2 \) are always unstable.

(ii) \( E_1 \) is locally asymptotically stable if \( \delta_1 > \frac{r_1}{K_2} \), with extinction for the prey population and stability for the predator population. If \( \delta_1 < \frac{r_1}{K_2} \), \( E_1 \) is unstable with stability for the predator. In addition, if \( \delta_1 = \frac{r_1}{K_2} \), \( E_1 \) is a stable non-hyperbolic point.

**Proof**

Indeed, let us determine the eigenvalues of the Jacobian matrix associated with each equilibrium point \( E_i = 0, 1, 2 \). The Jacobian matrix of system (2.3) is

\[
DG(X) = \begin{pmatrix}
-r_1 & -\frac{\delta_1 K_1}{1 + \delta_2 N} \\
\frac{\delta_1 K_1}{1 + \delta_2 N} & \frac{\alpha P}{1 + \delta_2 N} \\
0 & \frac{\alpha P}{1 + \delta_2 N} - r_2
\end{pmatrix},
\]

where \( X(t) = (N(t), P(t))^T \).

(i) For \( E_0 = (0, 0) \), the associated Jacobian matrix is \( DG(E_0) = \begin{pmatrix} -r_1 & 0 \\ 0 & r_2 \end{pmatrix} \). The eigenvalues are \( r_1 > 0 \) and \( r_2 > 0 \). Then \( E_0 \) is always unstable. In this case, we have instability of the prey and the predator.

(ii) For \( E_2 = (K_1, 0) \), the Jacobian matrix of system (2.3) evaluated at \( E_2 \) is

\[
DG(E_2) = \begin{pmatrix}
-r_1 & -\frac{\delta_1 K_1}{1 + \delta_2 N} \\
\frac{\delta_1 K_1}{1 + \delta_2 N} & \frac{\alpha P}{1 + \delta_2 N} + r_2
\end{pmatrix}.
\]

The eigenvalues are \( \lambda_1 = -r_1 < 0 \) and \( \lambda_2 = r_2 + \frac{\alpha P}{1 + \delta_2 N} > 0 \). Then \( E_2 \) is unstable with stability for the prey population and instability for the predator population.

(iii) For \( E_1 = (0, K_2) \), the associated Jacobian matrix is

\[
DG(E_1) = \begin{pmatrix}
r_1(1 - \frac{\delta_1 K_2}{r_1}) & 0 \\
0 & -r_2
\end{pmatrix}.
\]

The associated characteristic polynomial is given by

\[
P_{DG(E_1)}(\lambda) = \left( r_1 \left( 1 - \frac{\delta_1 K_2}{r_1} \right) - \lambda \right)(-r_2 - \lambda).
\]

Then the eigenvalues of \( DG(E_1) \) are \( \lambda_1 = r_1(1 - \frac{\delta_1 K_2}{r_1}) \) and \( \lambda_2 = -r_2 < 0 \). If \( \delta_1 > \frac{r_1}{K_2} \), then \( \lambda_1 < 0 \), therefore the system is locally asymptotically stable with extinction for the prey population and stability for the predator population. If \( \delta_1 < \frac{r_1}{K_2} \), then \( \lambda_1 > 0 \), therefore \( E_1 \) is unstable, so we have stability for the predator population.

If \( \frac{\delta_1 K_2}{r_1} = 1 \), then the equilibrium \( E_1 \) is a stable non-hyperbolic point. Indeed, to study the stability of \( E_1 \), we will use the center manifold theorem [25].
The eigenvalues of $DG(E_1)$ are $\lambda_1 = 0$ and $\lambda_2 = -r_2$, and the eigenspace associated with those eigenvalues is

$$W^0 = \left\{ \begin{pmatrix} r_2 \\ \omega K_2 \\ 1 \end{pmatrix} \right\},$$

$$W^{-r_2} = \langle (0;1) \rangle.$$

According to the center manifold theorem [25], there exists a center manifold which is tangent to $W^0$ at the point $E_1$.

Denote $x = N$ and $y = P$. The center manifold in this case is given by

$$W^c = \left\{ y = h(x)/h(0) = K_2, h'(0) = \frac{\omega K_2}{r_2} \right\},$$

where $h(x) = h(0) + h'(0)x + h''(0)x^2 + \mathcal{O}(x^3)$ is analytical at the neighborhood of the origin. Denoting $a = \frac{1}{K_2}$ and $b = \frac{1}{K_1}$, we have

$$y = h(x)$$

$$\Leftrightarrow \dot{y} = \dot{h}(x)$$

$$\Leftrightarrow r_2 h(x)(1 - ah(x)) + \frac{\omega x h(x)}{1 + x \delta_2} - \left( r_1 x (1 - bx) - \frac{\delta_1 x h(x)}{1 + x \delta_2} \right) h'(x) = 0.$$

By plugging the function $h$ by its expression in the previous equation and by grouping, we get:

$$r_2 h(0) \left( 1 - ah(0) \right) + \left( r_2 \delta_2 h(0) \right) \left( 1 - ah(0) \right) + r_2 h'(0) \left( 1 - 2ah(0) \right) + \omega h(0)$$

$$+ h'(0) \left( \delta_1 h(0) - r_1 \right) x + \left( r_1 h'(0) \right) \left( b - \delta_2 \right) + \left( h''(0) \right) \left( \delta_1 - ar_2 \right)$$

$$- 2h''(0) \left( r_1 - \delta_1 h(0) \right) + r_2 \delta_2 h'(0) \left( 1 - 2ah(0) \right)$$

$$+ r_2 h''(0) \left( 1 - 2a \right) x^2 + \mathcal{O}(x^3) = 0. \quad (*)$$

Since $r_1 = \delta_1 K_2$, then $r_1 - \delta_1 K_2 = 0$, by using $(*)$, we deduce that

$$h'(0) = \frac{\omega h(0)}{r_2} = \frac{\omega K_2}{r_2},$$

$$h''(0) = \frac{r_1 \left( b_1 - \delta_2 \right) + h'(0) \left( \delta_1 - ar_2 \right) - r_2 \delta_2}{r_2}.$$

The equation reduced to the center manifold is

$$\dot{x} = r_1 x (1 - ax) - \frac{\delta_1 x h(0) + h'(0)x + h''(0)x^2}{1 + \delta_2 x} = f(x).$$

A study of the sign of $f$ in the neighborhood of 0 gives the following result:

(i) If $x < 0$, then $f(x) < 0$;

(ii) If $x > 0$, then $f(x) < 0$. So the equilibrium point $E_1 = (K,0)$ is stable in $A$. \qed
3.2.2 Coexistence equilibria point of the model

To determine the coexistence equilibrium \( E_2 = (N^*, P^*) \) of system (2.3), we solve the following system:

\[
\begin{aligned}
& r_1 (1 - \pi_N N - \frac{\delta_r P}{1 + \theta N}) = G_1(N, P) = 0, \\
& r_2 (1 - \frac{P}{K_2}) = G_2(N, P) = 0.
\end{aligned}
\]  

(3.7)

Assume that \( N^*, P^* > 0 \). Dividing \( G_1(N^*, P^*) \) by \( N^* \), we obtain

\[
P^* = \frac{r_1}{\delta_1} (1 + \delta_2 N^*) \left( 1 - \frac{N^*}{K_1} \right) \quad \text{if } N^* < K_1.
\]

By plugging \( P^* \) in \( G_1(N^*, P^*) \), we get the cubic equation in \( N^* \) as follows:

\[
(N^*)^3 + \theta_2 (N^*)^2 + \theta_1 N^* + \theta_0 = 0,
\]

(3.8)

where

\[
\begin{aligned}
\theta_0 &= \frac{K_1 (\delta_1 K_2 - r_1)}{r_1 \delta_2}, \\
\theta_1 &= \frac{r_1 r_2 + r_2 \delta_1 \delta_2 K_1 K_2 + \omega \delta_1 K_1 K_2 - 2 r_1 r_2 \delta_2 K_1}{r_1 r_2 \delta_2^2} = \frac{1}{\delta_2} \left( \frac{1}{\delta_2} - 2 K_1 \right) + \frac{\delta_1 K_1}{r_1 \delta_2} K_p, \\
\theta_2 &= \frac{2 r_1 r_2 \delta_2 - r_1 r_2 \delta_2^2 K_1}{r_1 r_2 \delta_2^2} = \frac{2}{\delta_2} - K_1.
\end{aligned}
\]

The following result gives the existence of a coexistence equilibrium point [8, 19, 22, 26, 27].

**Theorem 1** Set \( \Delta' = (K_1 + \frac{1}{\delta_2})^2 - \frac{3 \delta_1 K_1}{r_1 \delta_2} K_p, \kappa_0 = \frac{\delta_2 \Delta_1}{\delta_1 \Delta_2}, \) and \( \kappa_1 = \frac{\theta^0_2}{\theta_1} \).

1. **System (2.3)** has no feasible coexistence equilibria if either

   (i) \( r_1 < \delta_1 K_2 \) and \( \Delta' \leq 0 \);

   (ii) \( r_1 < \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 < 2, \Delta' > 0, \) and \( \kappa_1 > 1 \);

   (iii) \( r_1 < \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 < 2, \Delta' > 0, \) and \( \kappa_0 > 1 \).

2. **System (2.3)** has a unique feasible coexistence equilibrium \( E_3 \) if either

   (i) \( r_1 > \delta_1 K_2 \) and \( \Delta' \leq 0 \);

   (ii) \( r_1 > \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 < 2, \Delta' > 0, \) and \( \kappa_1 < 1 \);

   (iii) \( r_1 > \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 > 2, \Delta' > 0, \kappa_1 > 1, \) and \( \kappa_0 < 1 \);

   (iv) \( r_1 > \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 > 2, \Delta' > 0, \kappa_0 > 1, \) and \( \kappa_1 > 1 \).

3. **System (2.3)** has two distinct feasible coexistence equilibria if either

   (i) \( r_1 < \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 < 2, \Delta' > 0, \) and \( \kappa_1 < 1 \), with \( E_3 \ll E_1 \);

   (ii) \( r_1 < \delta_1 K_2, \frac{r_1 K_1}{\delta_2} (2 K_1 - \frac{1}{\delta_2}) > K_p, \delta_2 K_1 > 2, \Delta' > 0, \kappa_0 > 1, \) and \( \kappa_1 < 1 \), with \( E_3 \ll E_1 \).

4. **System (2.3)** has three distinct feasible coexistence equilibria if

\[
r_1 > \delta_1 K_2, \quad \frac{r_1}{\delta_1 K_1} (2 K_1 - \frac{1}{\delta_2}) > K_p, \quad \delta_2 K_1 > 2, \quad \Delta' > 0, \quad \kappa_0 > 1, \quad \text{and} \quad \kappa_1 < 1.
\]
Proof Indeed, consider the following cubic equation:

\[ L(X) = X^3 + \theta_2 X^2 + \theta_1 X + \theta_0 = 0. \quad (3.9) \]

We have

\[ L(0) = \theta_0 \]

and

\[ L(K_1) = \frac{\delta_1 K_1 K_2 (\omega K_1 + r_2 (1 + K_1))}{r_1 r_2 \delta_2^3} > 0, \]

\[ L(X_1) = \frac{r_2 \Delta_1 - \delta_1 \delta_2 \Delta_2}{r_1 r_2 \delta_1 \delta_2^3} \]

and

\[ L(X_2) = \frac{\delta_2 \Delta_1 - \Delta_2}{r_1 r_2 \delta_1 \delta_2^3}, \]

where

\[ \Delta_1 = 22 r_1 \delta_1 \delta_2 + 27 r_1 K_1 \delta_1 \delta_2 K_1 K_2 + 27 \delta_1^2 \delta_2 K_1 K_2 \]

\[ + 9 \delta_1 \delta_2^3 K_1^2 K_2 + 2 r_1 \delta_1 \delta_2 (1 + \delta_2 K_1)^2 \sqrt{\Delta'}, \]

\[ \Delta_2 = 5 r_1 r_2 \delta_2^3 K_1 + 87 r_1 r_2 K_1 + 18 r_1 r_2 \delta_2 K_1 + 6 K_1 K_2 (r_2 \delta_2 + \omega)(3 + \delta_1 \sqrt{\Delta'}), \]

\[ \Delta_1' = 2 r_1 r_2 \delta_1 K_1 + 9 \delta_1 r_2 \delta_2^2 K_1^2 K_2 + 12 r_1 r_2 \delta_1 \delta_2 K_1 + 6 \delta_1 K_1 K_2 (\omega + r_2 \delta_2) \sqrt{\Delta'}, \]

\[ \Delta_2' = 2 r_1 r_2 \delta_1 \delta_2 (1 + \delta_2 K_1) \sqrt{\Delta'} + 2 r_1 r_2 \delta_1 \delta_2^3 K_1 + 18 r_1 r_2 \delta_1 \delta_2^2 K_1^2 + 18 \delta_1 \delta_2 K_1 K_2 (\omega + r_2 \delta_2) \]

with \( \Delta' \) defined from

\[ L'(X) = 3 X^2 + 2 \theta_2 X + \theta_1 = 0 \quad (3.10) \]

as

\[ \Delta' = \delta_2^2 - 3 \theta_1 \]

\[ = \left( K_1 + \frac{1}{\delta_2} \right)^2 - \frac{3 \delta_1 K_1 K_2}{r_1 \delta_2} \left( 1 + \frac{\omega}{r_2 \delta_2} \right) \]

\[ = \left( K_1 + \frac{1}{\delta_2} \right)^2 - \frac{3 \delta_1 K_1}{r_1 \delta_2} K_p. \]

There are two cases:

1. If \( \frac{r_1 \delta_2}{\delta_1 K_1} (K_1 + \frac{1}{\delta_2})^2 \leq K_p \), then \( \Delta' \leq 0 \). Thus, \( L \) is increasing on \( [0, K_1] \).
   
   (a) According to \( r_1 < \delta_1 K_2 \), we have \( \theta_0 > 0 \) with \( L(0) \times L(K_1) > 0 \). Hence, \( L(X) > 0 \) \( \forall X \in [0, K_1] \). Thus, equation (3.9) has no real roots on \( [0, K_1] \) and there are no feasible coexistence equilibria for system (2.3).
(b) If \( r_1 > \delta_1 K_2 \), we have \( \theta_0 < 0 \) with \( L(0) \times L(K_1) < 0 \). Thus, equation (3.9) has a unique positive root. Then system (2.3) has a unique feasible coexistence equilibrium.

2 If \( \frac{r_1 K_1^2}{\delta_0 K_1} (K_1 + \frac{1}{\delta_2}) > K_p \), then \( \Delta' > 0 \). Therefore, equation (3.10) has two roots

\[
X_1 = \frac{-\theta_2 - \sqrt{\theta_2^2 - 3\theta_1}}{3} \quad \text{and} \quad X_2 = \frac{-\theta_2 + \sqrt{\theta_2^2 - 3\theta_1}}{3},
\]

where

\[
\begin{cases}
X_1 + X_2 = \frac{-2\theta_2}{3}, \\
X_1 X_2 = \frac{\theta_1}{3}.
\end{cases}
\]

(a) If \( \delta_2 K_1 < 2 \) and \( \frac{r_1 K_1^2}{\delta_0 K_1} (2K_1 - \frac{1}{\delta_2}) > K_p \), then we have respectively \( \theta_2 > 0 \) and \( \theta_1 < 0 \).

Thus, we obtain also \( X_1 < 0 < X_2 \).

Using the fact that \( \Delta' < (K_1 + \frac{1}{\delta_2})^2 \), we have

\[
X_2 - K_1 = \frac{-\theta_2 + \sqrt{\theta_2^2 - 3\theta_1}}{3} - K_1
\]

\[
\leq \frac{1}{3} \left( K_1 + \frac{1}{\delta_2} \right).
\]

So \( X_2 < K_1 \). Hence \( L \) is decreasing on \([0, X_2]\) and increasing on \([X_2, K_1]\).

- If \( r_l < \delta_1 K_2 \) and \( \kappa_1 > 1 \), then we have respectively \( \theta_0 > 0 \) and \( L(X_2) > 0 \), then \( L(X) = 0 \) has no root. Therefore, system (2.3) has no coexistence equilibria.
- If \( r_1 > \delta_1 K_2 \) and \( \kappa_1 < 1 \), then we get respectively \( \theta_0 < 0 \) and \( L(X_2) \leq 0 \). Thus, equation (3.9) has one positive root with \( \beta_2 = L(X_2) \) is a minimum. Therefore, system (2.3) has a unique coexistence equilibrium.
- If \( r_1 < \delta_1 K_2 \) and \( \kappa_1 > 1 \), then we have respectively \( \theta_0 > 0 \) and \( L(X_2) < 0 \) with \( L(K_1) > 0 > L(X_2) \). Thus, equation \( L(X) = 0 \) has two distinct real positive roots, one is \( N_1^* \) in \([0, X_2]\), and the other \( N_1^+ \) in \([X_2, K_1]\). Each root corresponds to a distinct feasible coexistence equilibrium. Therefore, system (2.3) has two coexistence equilibria \( E_2(N_1^-, P^*) \) and \( E_2(N_1^+, P^*) \) with \( N_1^- < N_1^+ \).

(b) If \( \delta_2 K_1 > 2 \) and \( \frac{r_1 K_1^2}{\delta_0 K_1} (2K_1 - \frac{1}{\delta_2}) > K_p \), then we have respectively \( \theta_2 < 0 \) and \( \theta_1 < 0 \).

Thus, we obtain \( X_1 < 0 < X_2 \) with \( X_2 < K_1 \). By a similar argument as previously, we obtain the same result.

(c) If \( \delta_2 K_1 > 2 \) and \( \frac{r_1 K_1^2}{\delta_0 K_1} (2K_1 - \frac{1}{\delta_2}) < K_p \), then we get respectively \( \theta_2 < 0 \) and \( \theta_1 > 0 \).

Thus, we obtain \( 0 < X_1 < X_2 < K_1 \). Consequently, \( L \) is increasing on \([0, X_1]\) and \([X_2, K_1]\) and decreasing on \([X_1, X_2]\).

- According to \( r_1 < \delta_1 K_2 \), we have \( \theta_0 > 0 \). If \( \kappa_0 > 1 \) and \( \kappa_1 > 1 \), then we have respectively \( L(X_1) > 0 \) and \( L(X_2) > 0 \). So equation (3.9) has no real positive roots, and there are no feasible coexistence equilibria.
- According to \( r_1 < \delta_1 K_2 \), and if \( \kappa_0 > 1 \) and \( \kappa_1 < 1 \), then we have respectively \( L(X_1) > 0 \) and \( L(X_2) < 0 \). Thus, \( L(X) = 0 \) has two distinct real positive roots, one is \( N_2^* \) in \([X_1, X_2]\), and the other \( N_2^+ \) in \([X_2, K_1]\). Therefore, system (2.3) has two coexistence equilibria \( E_1(N_2^-, P^*) \) and \( E_1(N_2^+, P^*) \) with \( N_2^- < N_2^+ \).
• According to \( r_1 > \delta_1 K_2 \), and if \( \kappa_0 < 1 \) and \( \kappa_1 < 1 \), we have respectively \( L(X_1) < 0 \) and \( L(X_2) < 0 \). Consequently, \( L(X) = 0 \) has one root.
• According to \( r_1 > \delta_1 K_2 \), and if \( \kappa_0 > 1 \) and \( \kappa_1 < 1 \), we have respectively \( L(X_1) > 0 \) and \( L(X_2) < 0 \). Thus, \( L(X) = 0 \) has three roots. Therefore, system (2.3) has three coexistence equilibria.
• According to \( r_1 > \delta_1 K_2 \), and if \( \kappa_0 < 1 \) and \( \kappa_1 > 1 \), we have respectively \( L(X_1) < 0 \) and \( L(X_2) > 0 \). Thus, \( L(X) = 0 \) has one root.

\[ \square \]

The local stability analysis of coexistence equilibrium is given by the following theorem [8, 11, 12].

**Theorem 2** If condition (2) of Theorem 1 is satisfied, and moreover the following condition holds:

\[
\frac{2r_1(1 + \delta_2 N^*)(K_1 - N^*)}{\delta_1 K_1 K_2} > 1,
\]

(3.12)

then the coexistence equilibrium \( E_3 = (N^*, P^*) \) is locally asymptotically stable.

**Proof** Indeed, the Jacobian matrix of system (2.3) evaluated at the point \( E_3 \) is given by

\[
DG(E_3) = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix},
\]

where

\[
A_{11} = \frac{r_1(K_1 - 2N^*)(1 + \delta_2 N^*) - r_1(K_1 - N^*)}{K_1(1 + \delta_2 N^*)},
\]

\[
A_{12} = \frac{-\delta_1 N^*}{(1 + \delta_2 N^*)},
\]

\[
A_{21} = \frac{\omega r_1(K_1 - N^*)}{\delta_1 K_1(1 + \delta_2 N^*)},
\]

\[
A_{22} = \frac{r_2(1 + \delta_2 N^*)(\delta_1 K_1 K_2 - 2r_1(1 + \delta_2 N^*)(K_1 - N^*) + \omega N^* \delta_1 K_1 K_2)}{\delta_1 K_1 K_2(1 + \delta_2 N^*)}.
\]

The characteristic polynomial is therefore \( P(\lambda) = \lambda^2 - B_1 \lambda + B_2 = 0 \), with

\[
B_1 = \text{tr}(DG(E_3)) = \frac{\delta_1 [M_3 K_2 (r_1 (M_2 - N^*) + r_2 K_1) - r_1 r_2 K_2] - 2r_1 r_2 M_2 M_4}{AM_3},
\]

\[
B_2 = \text{det}(DG(E_3)) = \frac{r_1 r_2 \delta_1 K_2 (\delta_1 K_1 K_2 M_3 - 2r_1 M_4) \times (\delta_2 N^*(N^* - K_1) - N^*) + \omega r_1 N^* M_2}{AK_1 M_3^2} + \frac{\omega r_1 N^* M_2}{K_1 M_3^2},
\]

where

\[
M_1 = K_1 + (\delta_2 K_1 - 2)N^*, \quad M_2 = K_1 - N^*, \quad M_3 = 1 + \delta_2 N^*,
\]

\[
M_4 = -\delta_2^2 (N^*)^3 + \delta_2 (\delta_2 K_1 - 2)(N^*)^2 + (2\delta_2 K_1 - 1)N^* + K_1 = (1 + \delta_2 N^*)^2 (K_1 - N^*),
\]

and \( \mathcal{A} = \delta_1 K_1 K_2 \).
By a simple calculation, we get $B_1 = A_{11} + A_{22}$ and $B_2 = A_{11}A_{22} - A_{12}A_{21}$. According to (3.12), we get $B_1 < 0$ and $B_2 > 0$. By applying the Routh–Hurwitz criterion, $E_3$ is locally asymptotically stable.

The following theorem gives the global stability [8, 9, 13, 19, 22, 28, 29].

**Theorem 3** If condition (2) of Theorem 1 is satisfied, then the coexistence equilibrium $E_3 = (N^*, P^*)$ is globally asymptotically stable in the following subset of $\mathbb{R}^2$:

$$B = \left\{ (N, P) \in \mathbb{R}^2 \mid N \geq N^*, P \geq P^*, N \geq \frac{1}{\delta_2} \left( \frac{K_1}{r_1(1 + \delta_2N^*)} - 1 \right) \right\}.$$

**Proof** Indeed, we construct a Lyapunov candidate function defined by

$$V(N, P) = h_1(N) + h_2(P),$$

with $h_1(N) = a_1 \int_{N^*}^{N} \frac{N - N^*}{\zeta} d\zeta$, $h_2(P) = a_2 \int_{P^*}^{P} \frac{P - P^*}{\zeta} d\zeta$, and $(a_1, a_2) \in \mathbb{R}_+^2$ to be determined. It is easy to see that $V(N^*, P^*) = 0$, and for all $(N, P) \neq (N^*, P^*)$, $V(N, P) > 0$. So $V$ is well defined.

The time derivative of $V(N, P)$ along the solutions of system (2.3) is

$$\dot{V}(N, P) = a_1 (N - N^*) \left[ r_1 \left( 1 - \frac{N}{K_1} \right) - \frac{\delta_1 P}{1 + \delta_2 N} \right] + a_2 (P - P^*) \left[ r_2 \left( 1 - \frac{P}{K_2} \right) + \frac{\omega N}{1 + \delta_2 N} \right].$$

After simplification, we can write

$$\dot{V}(N, P) = -\frac{a_2 r_2}{K_2} (P - P^*)^2 + \left( a_2 \omega - \delta_1 a_1 - a_1 \delta_2 N^* \right) \frac{(N - N^*)(P - P^*)}{(1 + \delta_2 N)(1 + \delta_2 N^*)} + (N - N^*)^2 \left[ \frac{a_1 r_1}{K_1} + \frac{a_1 \delta_1 \delta_2 P^*}{(1 + \delta_2 N^*)(1 + \delta_2 N)} \right].$$

(3.13)

Taking $a_1 = 1$ and $a_2 = \frac{\delta_1 M_0}{\omega}$, finally we obtain

$$\dot{V}(N, P) = (N - N^*)^2 \left[ \frac{r_2}{K_2} + \frac{\delta_1 \delta_2 P^*}{(1 + \delta_2 N^*)(1 + \delta_2 N)} \right] - \frac{r_2 (1 + \delta_2 N^*)}{K_2 \omega} (P - P^*)^2$$

for all $(N, P) \neq (N^*, P^*)$. The coefficients $(N - N^*)^2$ and $(P - P^*)^2$ are positive. By using the fact that $N \geq \frac{1}{\delta_2} \left( \frac{K_1}{r_1(1 + \delta_2N^*)} - 1 \right)$, we have $\dot{V}(N, P) < 0$.

In addition $\dot{V}(N, P) = 0$ if and only if $(N, P) = (N^*, P^*)$. By using LaSalle’s invariance principle, $E_3 = (N^*, P^*)$ is globally asymptotically stable on $B$.

The following proposition gives the necessary and sufficient conditions of stability in case there is more than one equilibrium point [13, 19, 22, 27, 30].

Let us define the quadratic function

$$\pi(N) = -2r_1 \delta_2 N^2 + 2r_1 (1)N + K_1 (2r_1 - \delta_1 K_2).$$

(3.14)
For \( N^* \in ]0, K_1[ \),
\[
\pi(N^*) > 0 \iff \frac{2r_1(1 + \delta_2 N^*)(K_1 - N^*)}{\delta_1 K_1 K_2} > 1.
\] (3.15)

**Theorem 4**
(i) According to condition 3, (i) of Theorem 1, and if \( \delta_2 K_1 < 1 \) and \( r_1 > \frac{\delta_1 K_2}{2} \), then the coexistence equilibrium \( E_3(N_1^*, P^*) \) is locally asymptotically stable and \( E_3'(N_1^*, P^*) \) is unstable.

(ii) According to condition 3, (ii) of Theorem 1, if \( \delta_1 < \frac{5r_1}{\delta_2} (\delta_2 K_1 - 1) \) and \( r_1 > \frac{\delta_1 K_2}{2} \), then the coexistence equilibrium \( E_3'_{(N_2^*, P^*)} \) is locally asymptotically stable and \( E_3'(N_2^*, P^*) \) is unstable.

**Proof** Indeed,
(i) Consider the function \( \pi(N) \) defined by (3.14), we have
\[
\pi'(N) = -4r_1 \delta_2 N + 2r_1 (\delta_2 K_1 - 1), \quad \forall N \in ]0, K_1[.
\] (3.16)

Also, we have \( \pi(0) = K_1 (2r_1 - \delta_1 K_2) > 0 \) if \( r_1 > \frac{\delta_1 K_2}{2} \), \( \pi(K_1) = -\delta_1 K_1 K_2 \). According to \( \delta_2 K_1 < 1 \), we obtain \( \pi'(N) < 0 \) \( \forall N \in ]0, K_1[ \). Consequently, there exists \( \delta \) such that \( \pi(\delta) = 0 \) with \( 0 < N_1^* < \delta < N_1^* < K_1 \). Thus, \( \pi(N_1^*) > 0 \) and \( \pi(N_1^*) < 0 \).

By using \( \pi(N_1^*) > 0 \), we obtain
\[
\frac{2r_1(1 + \delta_2 N_1^*)(K_1 - N_1^*)}{\delta_1 K_1 K_2} > 1.
\] (3.17)

Thus, inequality (3.17) verifies the condition of stability given by (3.15). As result, \( \text{tr}(E_3) < 0 \) and \( \det(E_3) > 0 \). Consequently, \( E_3(N_1^*, P^*) \) is locally asymptotically stable.

By using \( \pi(N_1^*) < 0 \), we obtain
\[
\frac{2r_1(1 + \delta_2 N_1^*)(K_1 - N_1^*)}{\delta_1 K_1 K_2} < 1.
\] (3.18)

Thus, inequality (3.18) does not check the condition of stability given by (3.15). Consequently, \( E_3'(N_1^*, P^*) \) is unstable.

(ii) By using equation (3.16), there exists \( N_0 \in ]0, K_1[ \) such that \( \pi'(N_0) = 0 \) with \( N_0 = \frac{5r_1}{\delta_2} \). Thus, \( \pi \) is increasing on \( ]0, N_0[ \) and decreasing on \( ]N_0, K_1[ \). By simple computation, we get \( \pi(N_0) = 5r_1 K_1 - (\frac{5r_1}{\delta_2} + 5r_1 K_1 K_2) \). By using \( \delta_2 K_1 > 2 \), we get \( \delta_1 < \frac{5r_1}{\delta_2} (\delta_2 K_1 - 1) \). According to \( r_1 > \frac{\delta_1 K_2}{2} \) and \( \delta_1 < \frac{5r_1}{\delta_2} (\delta_2 K_1 - 1) \), we get respectively \( \pi(0) > 0 \) and \( \pi(N_0) > 0 \). Consequently, there exists \( \delta_0 \) such that \( \pi(\delta_0) = 0 \) with \( 0 < N_2^* < \delta_0 < N_1^* < K_1 \). Thus, \( \pi(N_2^*) > 0 \) and \( \pi(N_2^*) < 0 \).

Consequently, \( E_3'(N_2^*, P^*) \) is locally asymptotically stable and \( E_3'(N_2^*, P^*) \) is unstable.

\[ \square \]

**3.3 Bifurcation analysis**
In this subsection, we define the conditions of Hopf-bifurcations and the critical values of Hopf bifurcations. Here, \( \delta_1 \) is taken as a bifurcation parameter \([10, 15, 31]\).
Theorem 5 If condition (2), (ii) of Theorem 1 is satisfied and if the following conditions are satisfied:

\[
\delta_1 < \frac{2r_1r_2M_4}{M_2K_2(r_1(M_2 - N^*) + r_2K_1) - r_1r_2K_2}, \quad (3.19)
\]

\[
B_3^1(\delta_1) \geq 4B_2(\delta_1), \quad (3.20)
\]

then a Hopf-bifurcation occurs at the value \( \delta_1 = \delta_{1c} \), where

\[
\delta_{1c} = \frac{2r_1r_2M_4}{M_2K_2(r_1(M_2 - N^*) + r_2K_1) - r_1r_2K_2}. \quad (3.21)
\]

Proof Indeed, assuming that \( \delta = 0 \), and separating real and complex parts, we obtain \( E \) is satisfied and according to the Routh–Hurwitz criterion, the equation corresponding to the Jacobian matrix \( DG \) of \( E \) is evaluated at \( E_3 = (N^*, P^*) \):

\[
x^2 - B_1(\delta_1)x + B_2(\delta_1) = 0, \quad (3.22)
\]

where

\[
B_1(\delta_1) = \text{tr}(DG(E_3)) = \frac{\delta_1 [M_2K_2(r_1(M_2 - N^*) + r_2K_1) - r_1r_2K_2] - 2r_1r_2M_2M_4}{AM_3},
\]

\[
B_2(\delta_1) = \det(DG(E_3)) = \frac{r_1r_2\delta_1K_2(r_1K_2M_3 - 2r_1M_4) \times (\delta_2N^*(K_1 - N^*) - N^*) + \omega r_1N^*M_2K_3}{AK_1M_3^2}. \quad (4.a)
\]

If conditions (3.20) and (3.21) are respectively satisfied, we have respectively \( B_1(\delta_1) = 0 \) and \( B_2(\delta_1) > 0 \), then the eigenvalues will be purely complex at \( \delta_1 = \delta_{1c} \) with

\[
x = B_1(\delta_1) + \frac{\sqrt{B_1^2(\delta_1) - 4B_2(\delta_1)}}{2} \quad \text{or} \quad x = B_1(\delta_1) - \frac{\sqrt{B_1^2(\delta_1) - 4B_2(\delta_1)}}{2}.
\]

Replacing \( x = x_1 + ix_2 \) into (3.22), we have

\[
(x_1^2 - x_2^2) - B_1(\delta_1)x_1 + B_2(\delta_1) + i(2x_1x_2 - B_1(\delta_1)x_2) = 0,
\]

and separating real and complex parts, we obtain

\[
x_1^2 - x_2^2 - B_1(\delta_1)x_1 + B_2(\delta_1) = 0, \quad (4.a)
\]

\[
2x_1x_2 - B_1(\delta_1)x_2 = 0. \quad (4.b)
\]

Now, we verify the transversality condition.

Considering \( \text{Re}(x) = 0 \) and differentiating (4.b) with respect \( \delta_1 \), we get

\[
\left( \frac{dx_1}{d\delta_1} \right)_{\delta_1 = \delta_{1c}} = \frac{r_1r_2K_1K_2M_3(K_2(1 - \delta_{1c}) + 2M_2M_4)}{2(\delta_{1c}K_1K_2M_3)^2} \neq 0.
\]

As result, system (2.3) admits a Hopf-bifurcation at \( \delta_1 = \delta_{1c} \) corresponding to \( E_3 \).

Remark 1 Since \( B_1(\delta_{1c}) = 0 \) and \( \left( \frac{dB_1}{d\delta_1} \right)_{\delta_1 = \delta_{1c}} = 2(\frac{dB_1}{d\delta_1})_{\delta_1 = \delta_{1c}} \neq 0 \), if condition (3.12) is satisfied and according to the Routh–Hurwitz criterion, \( E_3 \) is locally asymptotically sta-
Table 1 Parameter values used for the numerical simulation

| Parameters | Values | References |
|------------|--------|------------|
| $r_1$      | 0.1    | estimated  |
| $r_2$      | 1.8    | estimated  |
| $K_1$      | 10     | estimated  |
| $K_2$      | 1800   | estimated  |
| $\delta_1$| 0.01   | estimated  |
| $\delta_2$| 0.01   | [33]       |
| $\omega$  | 0.015  | [33]       |

Table 2 Parameter values used for the numerical simulation

| Parameters | Values | References |
|------------|--------|------------|
| $r_1$      | 1.8    | estimated  |
| $r_2$      | 0.01   | estimated  |
| $K_1$      | 200    | estimated  |
| $K_2$      | 30     | estimated  |
| $\delta_1$| 0.01   | [33]       |
| $\delta_2$| 0.029  | estimated  |
| $\omega$  | 0.0015 | estimated  |

In addition, for $\delta_1 = \delta_{1c}$, a Hopf-bifurcation occurs. For $\delta_1 > \delta_{1c}$, $E_3$ approaches a periodic solution.

4 Numerical experiments and biological explanations

In this section, we present a sequence of numerical simulations in order to support our mathematical results and to analyze the effect of predation on the dynamics of the two species. We use MATLAB technical computer software [8, 12, 32]. The values of the parameters are given in Tables 1 and 2.

4.1 Global behavior of system (2.3)

Here, we are interested in the predation effect on the dynamics of the two species in order to follow its impact over time. Figure 2 shows the behavior of system (2.3) around $E_1$ and the parameter values used are given in Table 1. We observe the stability of the predator population and the extinction of the prey population for the predation parameter $\delta_1 = 0.01 > 5.510^{-5}$. This result supports (ii) of Proposition 3. This result confirms that the predator population can survive even if the prey dies out.

Now, we examine the behavior of system (2.3) around the coexistence equilibrium. We take $N > 96$, and the parameter values used are given in Table 2. We observe that system (2.3) converges globally towards the coexisting equilibrium $E_3 = (70,354.1)$ (see Figure 3(a)–(b)–(c)). The existence of center (Figure 3(d)) confirms the existence and the global asymptotic stability of the coexisting equilibrium. It means that the prey population exists despite the predation. Thus, we talk about the phenomenon of subsistence. That illustrates the result of our Theorem 3.

If we increase the value of $\delta_1 = 0.033$ and keep the other parameters fixed in Table 2, from Figure 4, we observe that the equilibrium $E_3$ loses its stability. This result confirms Theorem 2. In next subsection Figure 5 shows the Hopf-bifurcation of system (2.3) around $E_3$ at $\delta_1 = \delta_{1c}$.

We continue our numerical simulations when the system admits two coexistence equilibria in order to look the behavior of the system around $E^-_3$ and $E^+_3$. By increasing the
Figure 2 Evolution of system (2.3) around $E_1 = (0, 1800)$

Figure 3 Global asymptotic stability of the coexisting equilibrium of system (2.3) around $E_3 = (70, 354.51)$
Figure 4 Local asymptotic stability of the coexisting equilibrium of system (2.3) corresponding to $\delta_1 = 0.033$

Figure 5 Dynamics of the trajectories showing the existence of limit cycle arising from the Hopf-bifurcation of system (2.3) around $E_3 = (N^*, P^*)$ with $\delta_1 = \delta_{1c} = 0.0636$
value $\delta_1$ to $\delta_1 = 0.023$, we observe that system (2.3) converges globally towards the coexisting equilibrium $E_{3} = (75, 400)$ (see Figure 6(a)–(b)–(c)). By increasing the parameter of predation $\delta_1$ to $\delta_1 = 0.044$, we observe the loss of stability of the coexistence equilibrium $E_{3}$ (see Figure 7). This is in accordance with the mathematical results established in Theorem 4.

At the same time, we observe the instability of the coexistence equilibrium $E_{3}^{+}$ showing the existence of a limit cycle illustrated by Figures 8 and 9.

4.2 Analysis of Hopf-bifurcation diagram

We continue our numerical analysis in this subsection to observe the dynamics behavior of the system by considering the predation parameter. Now, if we consider the critical value $\delta_{1c} = 0.0636$, Figure 5((c)–(d)) shows that the coexisting equilibrium $E_{3} = (N^*, P^*)$ is unstable, and we have a limit cycle arising from the Hopf-bifurcation. Theorem 5 then holds.

Remark 2 The biological interpretation of the Hopf-bifurcation is that the prey coexists with the predator, exhibiting oscillatory equilibrium behavior [10, 11]. Indeed, we observe that if the predation threshold $\delta_1 > \delta_{1c}$, we have periodic fluctuation of the prey and predator species: Figures 5(c) and 5(d) show the existence of a limit cycle resulting from the Hopf-bifurcation. This highlights an extinction of the population of prey (at risk) if predation exceeds a certain threshold.

5 Conclusion

The effect of predation in the dynamics of the prey-predator model plays an essential role in the equilibrium of the ecosystem, because it allows natural mechanisms of regulation of
species. It is for this reason that in this paper we proposed and analyzed a nonlinear mathematical model to describe the dynamics of the populations of prey and predators, taking
into account the effect of predation. The formulation of the model derived from an ODE system by considering Holling function response of type II to represent the interaction between the prey and the predator. The mathematical results allowed us first to establish the positivity of the solutions indicating the existence of the population, as well as the boundedness to explain the natural control of the growth due to the restriction of the resources. In addition, we established the conditions of existence of the coexistence equilibria. Under certain conditions of the predation rate, we were able to establish the local stability of the coexistence equilibrium. In order to show the long-term coexistence of prey and predator species, we established the global stability of the coexistence equilibrium via an appropriate Lyapunov function under certain conditions of the model parameters. Moreover, we have described the conditions of existence of the Hopf-bifurcation in order to analyze to what extent the trajectories will be influenced by changes in the predation rate.

Our numerical results gave interesting findings on the effect of predation on the dynamics of the prey-predator model and also allowed to validate our results established in the mathematical study. We have shown the dynamic behavior of our model under different values of the predation rate. Indeed, considering Fig. 2, under certain values of the predation rate, we note an extinction of the prey species and persistence of predators towards the carrying capacity. Staying in this same logic of variation of the predation rate and by considering the parameters fixed in Table 2, we obtain the global stability of coexistence equilibrium indicated by Figure 2(d); this also attests the results of Theorem 3. By increasing the value of \( \delta_1 \), we lost the stability indicated in Figure 4(d); this phenomenon confirms our mathematical results established in Theorem 2. If we exceed the critical threshold of predation \( \delta_{1c} \) found in Theorem 5, then we observe a periodic variation in the numbers of prey and predators indicated by Figures 5(a), (b), (c) and the existence of a limit cycle arising from the Hopf-bifurcation. In the light of these observations, we are led to conclude that the predation rate is a key parameter which governs our model and is useful for
understanding the dynamics of species of prey and predators in the natural environment, and plays a regulator role of species.

Despite the important findings on this dynamic, in order to deepen our study, we plan to extend this work, taking into account the presence of infectious diseases in both species in order to look at the impact of this disease on the dynamics of the two species.

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