Dynamical of Ratio-Dependent Eco-epidemical Model with Prey Refuge

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ABSTRACT
This paper discusses the dynamic analysis of three species in the eco-epidemiology model by considering the ratio-dependent function and prey refuge. The prey refuge is applied under the fact that infected prey has protection instincts that allow it to reduce predation risk. Here, we get the boundedness and three equilibrium points where are existence under certain conditions. In the model, three equilibrium points are locally asymptotically stable and one of the equilibrium points is globally asymptotically stable. We find that the system undergoes Hopf bifurcation around the interior equilibrium point by choosing prey refuge as a bifurcation parameter. We also find a condition for uniform persistence. Finally, several simulations of numerical are performed not only to illustrate the analytical results but also to illustrate the effect of the prey refuge.

Keywords: eco-epidemiology model; global stable; Hopf bifurcation; local stable; persistence

INTRODUCTION
One of the natural phenomena that described the interaction between one species and another individual is the prey-predator interaction. This interaction depends on whether the effects are profitable or detrimental. In the real world, prey-predator interaction also can be influenced by infectious diseases. These diseases can affect population size in the predator-prey interaction. Since then, the combination of epidemiological and ecological becomes important issues that are often discussed by many researchers. Mathematical studies have considered this issue into an eco-epidemiology model that contains the class of susceptible and infective populations.

Currently, several studies have focused on the spread of disease in prey only, e.g. [1]–[3]. It is well known that predators prefer to capture infected prey because they are easier to catch than susceptible prey. However, the predator can become infected after eating them. Therefore, several researchers are interested to investigate the spread of disease not only in prey but also in predators, e.g. [4],[5]. Moreover, some studies have reviewed the spread of disease in both populations, e.g. [6]. Base on several experiments, the spread of infectious disease becomes an important factor to know the regulation of population density [7].

In this paper, we focus on the situation where predators can eat infected prey only. It appropriates to the fact that infected prey tends to change its behavior. Infected prey shall live in an area that is accessible to predators [8]. Moreover, infected prey is less agile than healthy prey and can be predated by predators easily [1]. Hudson et al. [9] have
observed that predators selectively capture heavily infected red grouse.

Predators that consumed infected prey populations are described by the response functions. It is well known that response functions are an important component in the eco-epidemiology model. Generally, several researchers use Holling types as a response function in their model. According to [10], Holling types are divide into three types, namely Holling type I, Holling type II, and Holling type III. Holling type I means that the consumption rate of predator increases linearly with the density of prey but it achieves a constant value if predators are surfeit, e.g. [6][5]. Holling type II means that the consumption rate of predators increases if the density of prey is low, e.g. [1]–[4]. Meanwhile, Holling type III means that the consumption rate of predator increases when the density of prey is large but it decreases when the density of prey is low, e.g. [11]. In Holling type III, predators easily switch to eat one prey to another or they focus to eat prey in a location where it is most abundant [12].

The response function for the Holling type depends on the density of prey. This is unrealistic because it ignores the effects of predator interference. Base on the experiment, the density of predators can influence the consumption rate. In the modeling, the consumption rate that depended on the density of prey can’t describe the dynamics behavior when the density of predator influences the system [12]. Currently, several researchers have considered the density of both populations as a ratio-dependent function. This function depends on the ratio of prey to predator density [13]. According to [14], the ratio-dependent model is a more reasonable dynamic than the previous model.

One of the phenomena that reduce predation risk is prey refuge. It can avoid the extinction of prey and influence the stability of the dynamic behavior [15]. According to [3], the prey refuge that is incorporated in the model is divided into two types, namely the refuge for a constant-number of prey and the refuge for a constant-proportion of prey. It is well known that the refuge for a constant-number of prey has a stronger stabilizing effect than the refuge for a constant-proportion of prey [14]. Therefore, the model is more realistic by incorporating prey refuge and it gives an accessible factor to the predator.

In this study, we modify the eco-epidemiology model from [1] by changing Holling type II into a ratio-dependent function. Here, we also observe the effect of prey refuge in the system. Further, this article presents the results in the form of model analysis. Moreover, it is well observed that there are Hopf bifurcations around the positive equilibrium point and the condition for uniform persistence. Finally, several simulations are performed to illustrate the analytical results.

METHODS

We use several methods to modify the eco-epidemiology model from [1]. The method is presented as follows.
1. Reviewing and studying the eco-epidemiology model from previous literature.
2. Modifying the eco-epidemiology model by changing the Holling II type into the ratio-dependent function.
3. Investigating the boundedness, equilibrium points, and dynamical behavior in the modified model.
4. Investigating Hopf bifurcation and persistence in the modified model.
5. Performing numerical simulation by using the 5th-order predictor-corrector method as a numerical method to support the analytical results.
RESULTS AND DISCUSSION

The Mathematical Model

In this paper, the eco-epidemiology model consists of three populations, namely susceptible prey, infected prey, and predator. Let $X_0(t), X_i(t),$ and $Y(t)$ is respectively defined as the density of susceptible prey, infected prey, and predator at the time $t$. According to [1], the general eco-epidemiology model is presented as follows.

$$\begin{align*}
\dot{X}_S &= R - \beta X_S X_i - \delta X_S, \\
\dot{X}_I &= \beta X_S X_i - f(X_i, Y) Y - \eta X_I, \\
\dot{Y} &= e f(X_i, Y) Y - \gamma Y,
\end{align*}$$

(1)

with $X_S \geq 0, X_I \geq 0, Y \geq 0$.

The first equation of system (1) expresses that in the absence of disease, the prey population grows by following the equation as below.

$$\dot{X}_S = R - \delta X_S,$$

where $R$ is expressed as the level of recruitment in prey populations such as immigrants and new-born and $\delta$ is defined as the natural death rate of susceptible prey. Here, the growth of the prey population is affected by factors such as disease. The spread of disease is denoted by bilinear incidence rate $\beta X_S X_i$ with $\beta$ is the transmission rate. We assume that the prey is not infected because of inherited disease but other sources. Moreover, the disease spreads on susceptible prey only. We also assume that the infected prey populations do not recover nor reproduce.

The second equation of system (1) describes the development of the infected prey population. They will be erased by the natural death rate $\eta$ and the predation of the predator. Here, predation is denoted by the response function $f(X_i, Y)$. We assume that infected prey can hide. Hiding behavior gives protection for infected prey which can protect from predation. The protection of infected prey is denoted by constant $m$. Predators can capture infected prey by following a ratio-dependent function as in [14].

$$f(X_i, Y) = \begin{cases} 0, & \text{if } 0 \leq X_i \leq m, \\ \frac{a(X_i - m)}{X_i - m + \xi Y}, & \text{if } X_i > m, \end{cases}$$

with $a$ is expressed as the predation rate and $\xi$ is defined as half capturing saturation constant. According to [1], if the density of infected prey populations is below the constant $m$, then predators cannot eat them and will die exponentially. Meanwhile, if the density of infected prey populations is above the constant $m$, then predators can eat them. Thus, when $X_i > m$, then system (1) shall become as follows.

$$\begin{align*}
\dot{X}_S &= R - \beta X_S X_i - \delta X_S, \\
\dot{X}_I &= \beta X_S X_i - \frac{a(X_i - m)Y}{X_i - m + \xi Y} - \eta X_I, \\
\dot{Y} &= \frac{ae(X_i - m)Y}{X_i - m + \xi Y} - \gamma Y,
\end{align*}$$

(2)

with $X_S(0) \geq 0, X_I(0) \geq 0, Y(0) \geq 0$.

The last equation represents the behavior of the predator population. Predators only consume the infected prey and don’t consume the susceptible prey. When the infected prey population is absent, then the predator experiences a natural death rate $\gamma$. Here, it is assumed that disease does not spread from infected prey to predator. In the model, all parameters are positive values. The meaning of parameter and their units are summarized in Table 1.
Table 1. Units and the meaning of parameters in system (2)

| Parameters | Biological meaning | Units          |
|------------|--------------------|---------------|
| $X_S$      | The number of susceptible prey | Number        |
| $X_I$      | The number of infected prey      | Number        |
| $Y$        | The number of predators         | Number        |
| $R$        | The level of recruitment in prey | time^{-1}    |
| $\beta$    | The level of infection of disease | mass^{-1} time^{-1} |
| $\delta$   | The level of natural mortality in susceptible prey | time^{-1}    |
| $\eta$     | The level of natural mortality in infected prey | time^{-1}    |
| $\gamma$   | The level of natural mortality in predators | time^{-1}    |
| $\xi$      | Half saturation constant        | Number        |
| $a$        | The level of predation in predators | mass^{-1} time^{-1} |
| $e$        | The level of alteration from prey into predators | time^{-1}    |
| $m$        | The measure of prey in the refuge | Number        |

The Boundedness

To show the biological validity, we shall prove the boundedness of the system (2) as follows.

**Theorem 1.** All solutions of the system (2) are uniformly bounded.

**Proof:**
We define $W = X_S + X_I + \frac{1}{e}Y$. By differentiating $W$ to $t$, we get

$$\frac{dW}{dt} = \frac{dX_S}{dt} + \frac{dX_I}{dt} + \frac{1}{e} \frac{dY}{dt}.$$  

By substituting system (3), we get,

$$\frac{dW}{dt} = R - (\delta X_S + \eta X_I + \gamma Y).$$  

Next, we choose $q = \min\{\delta, \eta, \gamma\}$. Thus, we get

$$\frac{dW}{dt} \leq R - qW,$$

By using the theory of differential equation, we get

$$W(t) \leq \frac{R}{q} + C e^{-qt},$$

where $C$ is the arbitrary positive constant. For $t \to \infty$, we get

$$\lim_{t \to \infty} \sup W(t) \leq \frac{R}{q}.$$

Thus, all solutions of system (2) enter into $\Omega = \{(X_S, X_I, Y) \in \mathbb{R}_+^3 : W(t) \leq \frac{R}{q}\}$.

The Equilibrium Points

By setting $X_S = X_I = Y = 0$, we get the possible equilibrium points as below.

1. Axial equilibrium $E_0 \left(\frac{R}{\delta}, 0, 0\right)$ always existent.
2. Planar equilibrium $E_1\left(\frac{\eta}{\beta}, \frac{\beta R - \delta \eta}{\eta \beta}, 0\right)$ exists when $\beta R > \delta \eta$.

3. Interior equilibrium $E^\ast(X^\ast_S, X^\ast_I, Y^\ast)$, where $X^\ast_S = \frac{R}{\beta X^\ast_I + \delta}$, $Y^\ast = \frac{(ae - \gamma)(X^\ast_I - m)}{\gamma \xi}$, and $X^\ast_I$ is the positive root of the quadratic equation (3) as follows.

$$A_1(X^\ast_I)^2 + A_2X^\ast_I + A_3 = 0,$$

with

$$A_1 = -\beta (ae - \gamma + e\eta \xi),$$

$$A_2 = \beta m (ae - \gamma) - \delta (ae - \gamma + e\eta \xi) + eR\beta \xi,$$

$$A_3 = \delta m (ae - \gamma).$$

This point $E^\ast$ exists when it satisfies the condition $ae > \gamma$ and $X^\ast_I > m$. It is clear to show that $A_1 < 0$ and $A_3 > 0$. Thus, the determinant of the equation (3) is $D = (A_2)^2 - A_1A_3 \geq 0$. Therefore, to get the explicit form of the root of the equation (3), we have to check the following two cases:

a. For $D = 0$, the equation (3) has a twin positive root where $X^\ast_I = -\frac{A_2}{2A_1}$ with $A_2 > 0$.

b. For $D > 0$, the probability that equation (3) has positive roots is as follows.

- If $A_2 > 0$, then the equation (3) has two positive roots where $X^\ast_{I,2} = \frac{-A_2 \pm \sqrt{D}}{2A_1}$.

- If $A_2 < 0$, then the equation (3) has a single positive root where $X^\ast_I = \frac{-A_2 - \sqrt{D}}{2A_1}$.

Dynamical Behaviour

To investigate the stability in system (2), we have to determine the eigenvalues of the Jacobian matrix. Here, we identify the Jacobian matrix at $E(X_S, X_I, Y)$ as follows.

$$J(E) = \begin{bmatrix} -\beta X_I - \delta & -\beta X_S \frac{a e Y}{(X_I - m + \xi Y)^2} - \eta & 0 \\ \beta X_I & -\frac{a e Y}{(X_I - m + \xi Y)^2} & -\frac{a (X_I - m)^2}{(X_I - m + \xi Y)^2} \\ 0 & \frac{a e Y}{(X_I - m + \xi Y)^2} & -\frac{ae (X_I - m)^2}{(X_I - m + \xi Y)^2} - \gamma \end{bmatrix}.$$  

(4)

To check the stability of $E_0 \left(\frac{R}{\delta}, 0, 0\right)$, we get the Jacobian matrix by replacing $E(X_S, X_I, Y)$ in the equation (4) with $E_0 \left(\frac{R}{\delta}, 0, 0\right)$. Hence, we obtain the eigenvalues of the Jacobian matrix, namely $\lambda_1 = -\delta, \lambda_2 = \frac{\beta R}{\delta} - \eta$, and $\lambda_3 = ae - \gamma$. The point $E_0$ is locally asymptotically stable when $ae < \gamma$ and $\beta R < \delta \eta$.

To investigate the stability of $E_1 \left(\frac{\eta}{\beta}, \frac{\beta R - \delta \eta}{\eta \beta}, 0\right)$, we have to identify the Jacobian matrix by replacing $E(X_S, X_I, Y)$ in the equation (4) with $E_1 \left(\frac{\eta}{\beta}, \frac{\beta R - \delta \eta}{\eta \beta}, 0\right)$. Thus, we get the eigenvalues $\lambda_1 = ae - \gamma$ and other eigenvalues are the roots of the quadratic equation $\lambda^2 + \varphi_1 \lambda + \varphi_2 = 0$ with $\varphi_1 = \frac{\beta R - \delta \eta}{\eta} + \delta$ and $\varphi_2 = \beta R - \delta \eta$. By using the Routh-Hurwitz criteria, the eigenvalues have negative real roots when $\beta R > \delta \eta$. Hence, the point $E_1$ is locally asymptotically stable when $ae < \gamma$ and $\beta R > \delta \eta$. From the above discussion, we get the following theorem as follows.

**Theorem 2.** The axial equilibrium $E_0$ is locally asymptotically stable when $ae < \gamma$ and $\beta R < \delta \eta$ and the planar equilibrium $E_1$ is locally asymptotically stable when $ae < \gamma$ and $\beta R > \delta \eta$.

Theorem 2 means that if the level of natural mortality in predator is lower than a certain value and the level of infection is lower than a certain value, then the point $E_0$
becomes stable. Meanwhile, if the level of infection is greater than a certain value, then the point \( E_1 \) becomes stable.

Now, we shall investigate the stability of \( E^*(X_S^*, X_I^*, Y^*) \). By replacing \( E(X_S, X_I, Y) \) in the equation (3) with \( E^*(X_S^*, X_I^*, Y^*) \), we get the Jacobian matrix where \( u_{ij} \) is the entry of matrix with row \( i \) and column \( j \) as follows.

\[
J(E^*) = \begin{bmatrix}
    u_{11} & u_{12} & 0 \\
    u_{21} & u_{22} & u_{23} \\
    0 & u_{32} & u_{33}
\end{bmatrix},
\]

with

\[
\begin{align*}
    u_{11} &= -\beta X_I^* - \delta; \\
    u_{12} &= -\beta X_S^*; \\
    u_{21} &= \beta X_I^*; \\
    u_{22} &= \frac{\alpha \xi (Y^*)^2}{(X_I^* - m + \xi Y^*)^2} - \eta; \\
    u_{23} &= -\frac{a (X_I^* - m)^2}{(X_I^* - m + \xi Y^*)^2}; \\
    u_{32} &= \frac{ae \xi (Y^*)^2}{(X_I^* - m + \xi Y^*)^2}; \\
    u_{33} &= -\frac{a e Y^* (X_I^* - m)}{(X_I^* - m + \xi Y^*)^2}.
\end{align*}
\]

Hence, we obtain the characteristic equation of \( E^* \), namely

\[
\lambda^3 + \mu_1 \lambda^2 + \mu_2 \lambda + \mu_3 = 0 \tag{5}
\]

with

\[
\begin{align*}
    \mu_1 &= -(u_{11} + u_{22} + u_{33}), \\
    \mu_2 &= u_{11} u_{22} + u_{11} u_{33} + u_{22} u_{33} - u_{12} u_{21} - u_{23} u_{32}, \\
    \mu_3 &= -u_{11} u_{22} u_{33} + u_{11} u_{23} u_{32} + u_{12} u_{21} u_{33},
\end{align*}
\]

By using the Routh-Hurwitz criteria, the eigenvalues have negative real roots when \( \mu_1 > 0, \mu_3 > 0, \mu_1 \mu_2 > \mu_3 \). Thus, the point \( E^* \) is locally asymptotically stable when \( \mu_1 > 0, \mu_3 > 0, \mu_1 \mu_2 > \mu_3 \). Therefore, we get the following theorem.

**Theorem 3.** The interior equilibrium \( E^* \) is locally asymptotically stable when it satisfies the condition \( \mu_1 > 0, \mu_3 > 0, \mu_1 \mu_2 > \mu_3 \).

In the next theorem, we shall prove that the point \( E_1 \) is globally asymptotically stable under a certain condition.

**Theorem 4.** The point \( E_1 \) is globally asymptotically stable in the \( X_S - X_I \) plane.

**Proof:**

By applying the Dulac function as \( H(X_S, X_I) = \frac{1}{X_I} \), we have

\[
H(X_S, X_I) = \frac{1}{X_I},
\]

\[
h_1(X_S, X_I) = R - \beta X_S X_I - \delta X_I,
\]

\[
h_2(X_S, X_I) = \beta X_S X_I - \eta X_I,
\]

where \( H(X_S, X_I) > 0 \) in the \( X_S - X_I \) plane. Thus, we get

\[
\Delta(X_S, X_I) = \frac{\partial}{\partial X_S} (h_1 H) + \frac{\partial}{\partial X_I} (h_2 H) = -\beta - \frac{\delta}{X_I} < 0.
\]

Base on Bendixson-Dulac criteria, there is no limit cycle in the \( X_S - X_I \) plane. Thus, the point \( E_1 \) is globally asymptotically stable in the \( X_S - X_I \) plane.

**Hopf Bifurcation**

In this section, we shall investigate Hopf bifurcation around \( E^*(X_S^*, X_I^*, Y^*) \). Hopf bifurcation guarantees that all solutions of system (2) enter a limit cycle around \( E^*(X_S^*, X_I^*, Y^*) \). Here, we choose the constant \( m \) as a bifurcation parameter. Hopf bifurcation around \( E^*(X_S^*, X_I^*, Y^*) \) is presented in the following theorem.
\textbf{Theorem 5.} The system (2) undergoes a Hopf bifurcation around $E^*(X^*_S, X^*_I, Y^*)$ when $m$ passes through a critical value $m = m_c$.

Proof:
We consider equation (5) as the characteristic equation at $E^*$. Next, we choose $m = m_c$ such that $\mu_1\mu_2 = \mu_3$ where $\mu_1, \mu_2, \mu_3 > 0$. Therefore, we get
\[
(\lambda^2 + \mu_2)(\lambda + \mu_1) = 0
\]
with the roots are $\lambda_{1,2} = \pm i\sqrt{\mu_2}$ and $\lambda_3 = -\mu_1$.

For all $m$, the roots become $\lambda_{1,2} = v_1(m) \pm iv_2(m)$ and $\lambda_3 = -\mu_1(m)$ where $v_1(m)$ and $v_2(m)$ are real. Next, we shall prove the transversality condition as follows.
\[
\frac{d(\text{Re} \lambda_j(m))}{dm} \bigg|_{m=m_c} \neq 0, \quad j = 1,2.
\]

By substituting $\lambda_1 = v_1(m) + iv_2(m)$ into equation (6) and differentiating to $m$, we obtain
\[
K\dot{v}_1 -Lv_2 + M = 0,
Lv_1 + K\dot{v}_2 + N = 0,
\]
where $K = 3(v_1^2 - v_2^2) + \mu_2 + 2v_1\mu_1, L = 6v_1 v_2 + 2v_2\mu_1, M = \mu_1(v_1^2 - v_2^2) + v_1\mu_2 + \mu_3$, and $N = 2v_1 v_2\mu_1 + v_2\mu_2$. Next, we solve equation (7). Thus, we have
\[
\dot{v}_1 = -\frac{KM + LN}{K^2 + L^2}.
\]
Since $KM + LN \neq 0$ and $\mu_1 \neq 0$, we obtain
\[
\frac{d(\text{Re} \lambda_j(m))}{dm} \bigg|_{m=m_c} = -\frac{KM + LN}{K^2 + L^2} \neq 0, \quad j = 1,2,
\]
and $\lambda_3(m_c) = -\mu_1(m_c) < 0$. Thus, the system (2) occurs Hopf bifurcation when $m$ passes through a critical value $m = m_c$.

\textbf{Persistence}

To show that all species are present and are not extinct in the future time, we shall prove that system (2) is uniform persistence.

\textbf{Theorem 6.} Let the assumption of Theorem 4 holds. If the inequalities $ae > \gamma$ and $\beta R > \delta \eta$ hold, then the system (2) is uniform persistence.

Proof:
We consider average Lyapunov function $\sigma(X) = X^*_S X^*_I r_2 Y^* r_3$ with $r_1, r_2, r_3 > 0$. Here, $\sigma(X)$ is nonnegative $C^1$ in $\mathbb{R}^3_+$. Thus, we have
\[
\vartheta(X) = \frac{1}{\sigma} \frac{d\sigma}{dt} = r_1 \frac{\dot{X}_S}{X_S} + r_2 \frac{\dot{X}_I}{X_I} + r_3 \frac{\dot{Y}}{Y},
\]
\[
= r_1 \left( \frac{R}{X_S} - \beta X_I - \delta \right) + r_2 \left( \beta X_S - \frac{a(X_I - m)Y}{X_I(X_I - m + \xi Y) - \eta} \right) + r_3 \left( \frac{ae(X_I - m)}{X_I - m + \xi Y - \gamma} \right).
\]

In the system, the point $E_0$ is the only equilibrium point which is no limit cycle around the equilibrium point. Therefore, Theorem 4 holds. Hence, it is enough to prove that $\vartheta(X) > 0$ for all equilibrium point $X \in bd \mathbb{R}^3_+$. Thus, we get
\[
\vartheta(E_0) = r_2 \left( \frac{RR - \delta \eta}{\delta} \right) + r_3 (ae - \gamma) > 0,
\]
\[
\vartheta(E_1) = r_3 (ae - \gamma) > 0.
\]
We note that if the inequalities $ae > \gamma$ and $\beta R > \delta \eta$ hold, then $\vartheta(E_0) > 0$ holds. Meanwhile, if the inequalities $ae > \gamma$ holds, then $\vartheta(E_1) > 0$ holds. Therefore, Theorem 6 expresses the instability of the point $E_0$ and $E_1$.

**Numerical Solutions**

The analytical results obtained are incomplete without numerical investigation. In this section, we present the numerical solution by using the 5th-order predictor-corrector method at $\Delta t = 0.01$. Here, we will give four simulations to verify our analytical results and also to demonstrate the effect of the prey refuge. We choose several parameters as in (8). Their units are given as in Table 1.

$$R = 2, \beta = 1, \delta = 1, \eta = 0.5, \gamma = 0.5, \xi = 1, a = 2, e = 0.75, m = 0.5. \quad (8)$$

**Simulation 1.** It confirms that system (2) has an equilibrium $E^* (1.0685, 0.8717, 0.7434)$. On simulation 1, Theorem 3 and Theorem 6 are satisfied. We observe that all solutions converge to the point $E^*$, see figure 1(a). Thus, the point is locally asymptotically stable, see figure 1(b), and the system (2) is uniform persistence.

**Simulation 2.** We replace $m = 0.5$ into $m = 0.0002$. Here, we investigate that system (2) has an equilibrium point $E^* (1.8305, 0.0926, 0.1848)$. On simulation 2, Theorem 3 is not satisfied but Theorem 6 is satisfied. Therefore, system (2) is uniform persistence but the equilibrium point $E^*$ is unstable, see figure 2.

![Figure 1](image1.png)  
(a)  
(b)  

**Figure 1.** The Dynamics of System (2) with $m = 0.5$ and Other Parameters as in (8)

![Figure 2](image2.png)  
(a)  
(b)  

**Figure 2.** This Figure Shows The Instability of System (2)
**Simulation 3.** To see Hopf bifurcation in the system (2), we choose \( m = m_c = 0.0004 \) as a bifurcation parameter. Thus, we identify that the equilibrium point of system (2) is \( E^*(1.8277,0.0943,0.1878) \). If we choose \( m = 0.005 > m_c = 0.0004 \), then all solutions of the system (2) converge to \( E^*(1.7795,0.1239,0.2378) \), see figure 3(a). Meanwhile, figure 3(b) shows the phase portrait of the system (2) with \( m = 0.005 \) which means that the point \( E^* \) is stable. Furthermore, if we choose \( m = 0.0001 < m_c = 0.0004 \), then the point \( E^*(1.8319,0.0918,0.1833) \) is unstable, see figure 4(a). Meanwhile, the phase portrait in the system (2) with \( m = 0.0001 \) that is presented in figure 4(b) means that the solution of the system (2) enter a limit cycle around \( E^* \). Thus, Theorem 5 is satisfied.

**Simulation 4.** To see the effect of prey refuge in the system (2), we use several values of prey refuge, namely \( m_1 = 0.1, m_2 = 0.65, \) and \( m_3 = 1.3 \). Figure (5) shows the time graph of the system (2) by using \( m \) makes different values. Here, we obtain that all populations exist no matter how large \( m \) with \( m < X_I \). This prey refuge creates the system (2) to become stable rapidly and no extinction occurs. Here, it is worthy to attention that when we choose the constant \( m \) with \( m < X_I \), then the measure of prey refuge doesn’t lead to predator extinction.

**Figure 3.** The Dynamics of System (2) with \( m = 0.005 \) and Other Parameters as in (8)

**Figure 4.** The Dynamics of System (2) with \( m = 0.0001 \) and Other Parameters as in (8)
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CONCLUSIONS

In this study, we have observed three species in the eco-epidemiology model with the ratio-dependent function incorporating prey refuge. We obtain three equilibrium points where all points, i.e. axial equilibrium, planar equilibrium, and interior equilibrium, are locally asymptotically stable under certain conditions. Moreover, the planar equilibrium point in the system (2) is globally asymptotically stable. Next, we find that Hopf bifurcation occurs around the interior equilibrium by choosing a bifurcation parameter in the constant \( m \). When \( m > m_c = 0.0004 \), the system (2) is stable. However, when \( m < m_c = 0.0004 \), the system (2) is unstable. Furthermore, we also find a condition for uniform persistence. If the level of natural mortality in predators is lower than a certain value and the level of infection is greater than a certain value, then all species exist in the future time. Next, by applying the prey refuge, all populations exist no matter how large \( m \) with \( m < X_I \). We conclude that system (2) is stable faster and there is no extinction.

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