Research article

Stability and bifurcation analysis of a diffusive modified Leslie-Gower prey-predator model with prey infection and Beddington DeAngelis functional response

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\section*{ABSTRACT}

In this paper, we present and analyze a spatio-temporal eco-epidemiological model of a prey predator system where prey population is infected with a disease. The prey population is divided into two categories, susceptible and infected. The susceptible prey is assumed to grow logistically in the absence of disease and predation. The predator population follows the modified Leslie-Gower dynamics and predates both the susceptible and infected prey population with Beddington-DeAngelis and Holling type II functional responses, respectively. The boundedness of solutions, existence and stability conditions of the biologically feasible equilibrium points of the system both in the absence and presence of diffusion are discussed. It is found that the disease can be eradicated if the rate of transmission of the disease is less than the death rate of the infected prey. The system undergoes a transcritical and pitchfork bifurcation at the Disease Free Equilibrium Point when the prey infection rate crosses a certain threshold value. Hopf bifurcation analysis is also carried out in the absence of diffusion, which shows the existence of periodic solution of the system around the Disease Free Equilibrium Point and the Endemic Equilibrium Point when the ratio of the rate of intrinsic growth rate of predator to prey crosses a certain threshold value. The system remains locally asymptotically stable in the presence of diffusion around the disease free equilibrium point once it is locally asymptotically stable in the absence of diffusion. The Analytical results show that the effect of diffusion can be managed by appropriately choosing conditions on the parameters of the local interaction of the system. Numerical simulations are carried out to validate our analytical findings.

\section*{1. Introduction}

It is known that infectious diseases can affect ecological systems and regulate population density. Thus, studying the influence of epidemiological factors on the dynamics of prey-predator interactions plays a crucial role for better understanding of the eco-system. Because of these, mathematical modelling of epidemics has become a very important subject of research after the seminal model of Kermack-McKendric [1] on SIRS systems. Anderson and May [2] were the pioneers for investigating the invasion, persistence, and spread of infectious diseases by formulating an eco-epidemiological prey-predator model. At recent times, many researchers have proposed and studied epidemic and eco-epidemiological models [3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15]. For example, Shaikh et al. [6] have investigated the dynamics of an eco-epidemiological system with disease in competitive prey species. They have considered hyperbolic mortality of predators and Holling type II functional response and showed the local and global stability of the feasible equilibria and the existence of Hopf bifurcation at both the endemic and Disease Free Equilibrium Points.

On the other hand, in reality, prey and predator species, both infected and healthy ones, are in-homogeneously distributed in different ecological space at a given time and interact with other organisms present within their spatial domain. This consideration involves diffusion process which can be quite intricate as different concentration levels of prey and predator cause different population movements [16]. Thus, this movement or diffusion process must be incorporated in temporal eco-epidemiological models that do not represent space explicitly. Thus, the resulting eco-epidemiological models are represented by reaction-diffusion equations.

The spatio-temporal dynamics of a prey-predator system with disease has been investigated by many researchers [17, 18, 19, 20, 21, 22, 23]. Ko et al. [17] have considered a ratio-dependent prey-predator sys-
tem with infection in the prey population and studied the asymptotic behavior of constant solutions, whereas R.K. Upadhyaya and P. Roy [24] developed a reaction-diffusion eco-epidemiological model of prey-predator interaction and found out the occurrence of temporal chaos at a fixed point in space. Raw et al. [18] have studied the dynamical complexities and formation of pattern in a spatial eco-epidemiological prey-predator system with prey infection and harvesting. They investigated conditions for the existence of Turing instability. R.K. Upadhyaya et al. [21] designed a spatial model to study a damaged diffusive eco-epidemiological system of Tilapia and Pelican populations in Salton Sea, California, USA. They observed the existence of Hopf bifurcation and investigated conditions for the Turing instability of the system. Li et al. [20] have considered an eco-epidemiological prey-predator system with infection in predator population and delay, and found the parameter ranges for the occurrence of Turing patterns.

Leslie and Gower [25] introduced a predator prey model where the predator grows logistically with carrying capacity depending on the availability of variable resource, the number of prey. The Leslie-Gower model is a ratio dependent model which has a singularity at the origin. However, when the predator is provided with an alternative food apart from its favored food, it gives rise to modified Leslie-Gower predator dynamics. Now, the model is mathematically free from any singularity and well behaved.

In recent years, more attention has been paid to the study of the dynamics of an eco-epidemiological Leslie-Gower predator-prey interactions [10, 19, 26, 27, 28, 29, 30, 31, 32]. However, most of the works focus on the non-spatial dynamics of an eco-epidemiological Leslie-Gower prey predator dynamics.

The novelty of this paper is the consideration of prey infection, with nonlinear incidence rate, mixed type of functional response: Beddington-DeAngelis type functional response and Holling type II functional response for the susceptible and infected prey population, respectively, and the modified Leslie-Gower predator dynamics. Thus, the main aim of this paper is to study the spatio-temporal dynamics of a diffusive eco-epidemiological predator-prey system with prey infection, Beddington-DeAngelis type functional response and the modified Leslie-Gower type predator dynamics.

The organization of this paper is as follows: in section 2, Mathematical Model Formulation is discussed. Section 3 deals with the analysis of the temporal system: existence and boundedness of solutions, the local and global stability analysis of the biologically feasible equilibrium points and bifurcation analysis of the system (3). Section 4 is devoted to the analysis of the spatio-temporal system: persistence properties of solutions and local stability of equilibrium points of the system (2) are discussed. Numerical simulation results are presented in section 5. Lastly, conclusions are given in section 6.

2. Mathematical model formulation

Let $N(X, T)$ and $W(X, T)$ represent the total prey population densities and the predator population density, respectively at time $T$ and position $X$ in a habitat $\Omega \subset R^2$, and the prey population is infected with a disease. We took the following assumptions to formulate our eco-epidemiological model.

1. In the presence of disease, the prey population is divided into two groups: Susceptible prey $U$ and infected prey $V$. Therefore the total prey population is $N = U + V$.

2. The susceptible prey is capable of reproducing and hence grows logistically with carrying capacity $K$ and intrinsic growth rate $r_1$. The infected population does not recover and the disease is not genetically inherited. The infected prey is removed by death at a natural rate $d$.

3. We assume that the disease transmission follows the non-linear incidence rate $\frac{d(UV)}{2(r_1 + r_2)T}$. In this incidence rate the number of effective contacts between infective and susceptible individuals is assumed to saturate at high infective levels due to crowding of infective individuals.

4. Predator predate both susceptible and infected prey following the Beddington-DeAngelis functional response and Holling type-II functional response, respectively. The Beddington-DeAngelis functional response is used to capture the mutual interference of predators. Whereas, Holling type II functional response is used because the infected preys are relatively accessible for predation, as they are weak to escape from the predator.

5. The predator dynamics follow the modified Leslie-Gower dynamics with intrinsic growth rate $r_1$, carrying capacity proportional to the density of the susceptible and infected prey populations.

6. The susceptible prey, infected prey and predator population moves in the habitat with constant diffusion coefficients $D_U$, $D_V$ and $D_W$, respectively.

Based on the above assumptions and parameters (Table 1), the spatio-temporal eco-epidemiological model is given by the following set of reaction diffusion equations.

$$
\begin{align*}
U_T - D_U \Delta U &= r_1 \left(1 - \frac{U}{K}\right) U - \frac{aUV}{1 + bV} - \frac{cUW}{B + U + \omega W}, \\
V_T - D_V \Delta V &= \frac{aUV}{1 + bV} - \frac{AWV}{1 + AhV} - dV, \\
W_T - D_W \Delta W &= r_2 \left(1 - \frac{W}{s_1 + s_2U + s_3V}\right) W, \\
U(t, X, 0) &= U^0(X), U^0(X) \geq 0, \\
V(t, X, 0) &= V^0(X), V^0(X) \geq 0, \\
W(t, X, 0) &= W^0(X), W^0(X) \geq 0,
\end{align*}
$$

Table 1. Biological meaning of parameters.

| Parameters | Biological meaning |
|------------|--------------------|
| $r_1$      | The intrinsic growth rate of Susceptible prey, |
| $K$        | Environmental carrying capacity of prey, |
| $a$        | Infection rate of prey, |
| $b$        | Measure of Inhibition of prey, |
| $c$        | Predation rate of Predator on susceptible prey |
| $B$        | Saturation constant |
| $\omega$   | Predator interference |
| $A$        | Search rate |
| $h$        | Handling time |
| $d$        | Natural death rate of infected prey |
| $r_2$      | Maximum per capita growth rate of the predator |
| $s_1$      | Residual loss in predator population due to severe scarcity of its favorite food |
| $s_2$      | Conversion factor of susceptible prey into predator |
| $s_3$      | Conversion factor of infected prey into predator |
| $D_U$      | Diffusion coefficient of susceptible prey |
| $D_V$      | Diffusion coefficient of infected prey |
| $D_W$      | Diffusion coefficient of predator |

where $\Omega \subset R^2$ is a bounded region with smooth boundary $\partial \Omega$, and all the parameters in the model are assumed to be positive. $v$ is the outward unit normal vector to the boundary $\partial \Omega$. The admissible initial data $U_0(X), V_0(X)$ and $W_0(X)$ are continuous functions on $\Omega$. The homogeneous Neumann boundary condition means that the system (1) is self-contained and has no population flux across the boundary $\partial \Omega$.

Introduce the following non-dimensional variables and parameters so as to reduce the number of parameters of the system (1):

$$
\begin{align*}
&u = \frac{U}{K}, v = \frac{V}{K}, w = \frac{W}{K}, \tau = \frac{r_1 T}{D_U}, x = X, \\
&\theta = \frac{r_2}{r_1}, \delta = \frac{d}{r_1}, \\
&\alpha = \frac{aK}{r_1}, \beta = \frac{B}{K}, \gamma = \frac{c}{r_1}, \eta = \frac{\omega}{r_1}
\end{align*}
$$

where $\Omega \subset R^2$ is a bounded region with smooth boundary $\partial \Omega$, and all the parameters in the model are assumed to be positive. $v$ is the outward unit normal vector to the boundary $\partial \Omega$. The admissible initial data $U_0(X), V_0(X)$ and $W_0(X)$ are continuous functions on $\Omega$. The homogeneous Neumann boundary condition means that the system (1) is self-contained and has no population flux across the boundary $\partial \Omega$. Introduce the following non-dimensional variables and parameters so as to reduce the number of parameters of the system (1):
The system (1) is transformed to the following non-dimensional system.

\[
\begin{align*}
    u_t - \Delta u &= (1-u)u - \frac{aw}{1 + \kappa u} - \frac{yw}{\beta + u + aw}, \\
    v_t - \Delta v &= \frac{aw}{1 + \kappa u} - \theta u - \delta v, \\
    w_t - \Delta w &= \eta \left( 1 - \frac{w}{s_1 + s_2 w + s_3 v} \right) w, \\
    u_0 &= v_0 = w_0 = 0,
\end{align*}
\]

\[u(x,0) = u_0(x) \geq 0, v(x,0) = v_0(x) \geq 0, w(x,0) = w_0(x) \geq 0.\]

(2)

3. Analysis of the temporal system

It is ecologically and epidemiologically reasonable to study the local dynamics of a predator-prey system before proceeding to the study of its spatio-temporal dynamics. Thus the temporal dynamics of the system (1), which involves only the interaction terms, is

\[
\begin{align*}
    \frac{du}{dt} &= (1-u)u - \frac{aw}{1 + \kappa u} - \frac{yw}{\beta + u + aw}, \\
    \frac{dv}{dt} &= \frac{aw}{1 + \kappa u} - \theta u - \delta v, \\
    \frac{dw}{dt} &= \eta \left( 1 - \frac{w}{s_1 + s_2 w + s_3 v} \right) w, \\
    u(0) &= u_0 \geq 0, v(0) = v_0 \geq 0, w(0) = w_0 \geq 0.
\end{align*}
\]

Let us define

\[
\begin{align*}
    G_1(u,v,w) &= (1-u)u - \frac{aw}{1 + \kappa u} - \frac{yw}{\beta + u + aw}, \\
    G_2(u,v,w) &= \frac{aw}{1 + \kappa u} - \theta u - \delta v, \\
    G_3(u,v,w) &= \eta \left( 1 - \frac{w}{s_1 + s_2 w + s_3 v} \right) w.
\end{align*}
\]

(3)

3.1. Positive invariance and boundedness

Theorem 3.1. All solutions of the system (3) with positive initial conditions exist and remain positive.

Proof. Let \((u(t),v(t),w(t))\) be a solution of the system (3). One can easily show that the functions \(G_1, G_2,\) and \(G_3\) are continuous functions and locally Lipschitzian on \(\mathbb{R}_+^3\). Therefore, the solution of the system (3) with positive initial condition exists and is unique. Moreover, it can be shown that these solutions exist for all \(t > 0\) and stay positive.

Theorem 3.2. All solutions of the system (3) which initiate in \(\mathbb{R}_+^3\) are uniformly bounded in the region

\[
\Theta = \{(u,v,w) \in \mathbb{R}_+^3 : u \leq 1 + \epsilon_1, v(t) \leq \frac{a}{\delta k} + \epsilon_2, w(t) \leq s_1 + s_2 + s_3 \frac{a}{\delta k} + \epsilon_3, \forall t, \epsilon_1, \epsilon_2, \epsilon_3 > 0\}.
\]

Proof. Let \((\hat{u}(t),\hat{v}(t),\hat{w}(t))\) be any solution of the system (3) with positive initial conditions. Since \(\frac{du}{dt} \leq u(1-u)\), by the comparison principle we have \(\lim_{t \to \infty} u(t) \leq 1\). Thus there exists \(t_2 > 1\) such that \(u(t) \leq 1 + \epsilon_1\) for some arbitrarily small \(\epsilon_1 > 0\).

From the second equation of the system (3), we have

\[
\frac{dv}{dt} + \delta v \leq \frac{aw}{1 + \kappa u} \leq \frac{a(1 + \epsilon_1)}{\kappa}.
\]

Applying Gronwall’s inequality [33], we have

\[0 < v(t) < \frac{a(1 + \epsilon_1)}{\delta k} \left( 1 - e^{-\delta t} \right) + v(0)e^{-\delta t}.\]

Since \(\epsilon_1 > 0\) is arbitrarily small and for letting \(t \to \infty\), we have \(\lim_{t \to \infty} v(t) \leq \frac{a}{\delta k} + \epsilon_2\). Thus there exists \(t_2 > 1\) such that \(v(t) \leq \frac{a}{\delta k} + \epsilon_2\) for some arbitrarily small \(\epsilon_2 > 0\).

From the third equation of the system (3), we have

\[
\frac{dw}{dt} \leq \eta \left( s_1 + s_2 (1 + \epsilon_1) + s_3 \left( \frac{a}{\delta k} + \epsilon_2 \right) - w \right) \frac{1}{s_1 + s_2 (1 + \epsilon_1) + s_3 \left( \frac{a}{\delta k} + \epsilon_2 \right)}.
\]

By the comparison principle, we get

\[\lim_{t \to \infty} w(t) \leq \frac{a}{\delta k} + \epsilon_2\].

Since \(\epsilon_1 > 0\) and \(\epsilon_2 > 0\) are arbitrarily small, we have \(\lim_{t \to \infty} w(t) \leq s_1 + s_2 + s_3 \frac{a}{\delta k} + \epsilon_3\). Thus there exists \(t_3 > 1\) such that \(w(t) \leq s_1 + s_2 + s_3 \frac{a}{\delta k} + \epsilon_3\) for some arbitrarily small \(\epsilon_3 > 0\).

Hence all solutions of the system (3), starting in \(\mathbb{R}_+^3\) are uniformly bounded for all \(t \geq 0\) and eventually confined in the region

\[
\Theta = \{(u,v,w) \in \mathbb{R}_+^3 : u \leq 1 + \epsilon_1, v(t) \leq \frac{a}{\delta k} + \epsilon_2, w(t) \leq s_1 + s_2 + s_3 \frac{a}{\delta k} + \epsilon_3, \forall t, \epsilon_1, \epsilon_2, \epsilon_3 > 0\}.
\]

3.2. Extinction criteria

Theorem 3.3. The disease will be removed from the system (3) if \(\alpha < \delta\).

Proof. From the second equation of the system (3), we have

\[
\frac{dv}{dt} \leq (\frac{aw}{1 + \kappa u} - \delta) v \leq (\alpha - \delta) v.
\]

And hence,

\[
\frac{dv}{dt} + (\alpha - \delta) v \leq 0.
\]

Applying Gronwall’s inequality [33], we have \(0 \leq v(t) < v(0)e^{\alpha - \delta t}\). Thus, for \(t \to \infty\), we have \(0 \leq v(t) \leq 0\), if \(\alpha < \delta\). Therefore, \(v(t) \to 0\) as \(t \to \infty\) if \(\alpha < \delta\). Hence, the theorem.

3.3. Equilibrium points and reproduction number

3.3.1. Equilibrium points

The temporal system (3) has the following six biologically feasible equilibrium points.

1. The Extinction Equilibrium Point \(E_0 = (0,0,0)\), which exists always.
2. The Infection and Predator Free Equilibrium Point \(E_1 = (1,0,0)\), which exists always.
3. The Prey Free Equilibrium Point \(E_2 = (0,0,1)\), which exists always.
4. The Predator Free Equilibrium Point \(E_3 = (u^*, v^*, 0)\), where \(u^* = \frac{\delta k (1 + s_2 + s_3)}{\alpha}\) and \(v^*\) is the unique positive root of the quadratic equation

\[
\delta^2 v^2 + (\alpha^2 - a\delta k + 2\delta^2 k)v - \delta(a - \delta) = 0.
\]

Equation (5) will have a unique positive root if \(a > \delta\). Thus, \(E_3\) exists if \(a > \delta\).

5. The Disease Free Equilibrium Point \(E_4 = (\hat{u}, 0, \hat{w})\), where

\[
\hat{u} = \frac{-B_1 + \sqrt{B_1^2 + 4(1 + s_2 w)C_1}}{2(1 + s_2 w)}; \hat{w} = s_1 + s_2 \hat{u}
\]

(6)

with

\[
B_1 = \beta + s_2 y + \omega(s_1 + s_2) - 1; C_1 = \beta - s_1(y - \omega).
\]

It can be observed that \(\hat{u}\) is positive if \(C_1 > 0\). Thus, \(E_4\) exists if

\[
\beta > s_1(y - \omega).
\]
6. The Endemic Equilibrium Point $E_3 = (\bar{u}, \bar{v}, \bar{w})$, where
\[
\bar{u} = \frac{1 + \kappa \delta + \theta_3 + (s_3 \theta + \sigma \delta) \delta}{a - s_3 \theta + (a \sigma - s_3 \theta_3) \delta},
\]
\[
\bar{v} = \frac{1 + \kappa \delta + \theta_3 + (s_3 \theta + \sigma \delta) \delta}{a - s_3 \theta + (a \sigma - s_3 \theta_3) \delta},
\]
and $\delta$ is the unique positive root of the polynomial
\[
A_i \delta^3 + A_i \delta^2 + A_i \delta + A_i = 0.
\]
The coefficients $A_i$ ($i = 1, 2, 3, 4, 5$) are given in Supplementary Appendix.

3.3.2. Reproduction number
The basic reproduction number $R_0$ is obtained by using the next generation matrix method [34]. The temporal system (3) has one infected state, $v$, and two uninfected states, $u$ and $w$. Therefore, the flux of newly infected is $F$, where $F = \frac{aw}{kv+1}$; other entering and leaving fluxes is $V$, where $V = \frac{aw}{kv+1} \delta v$.
The Next generation matrix is then $K = FV^{-1}$, where
\[
F = \left( \frac{\partial F}{\partial u} \right)_{E_3 = (0,0,0)} \text{ and } V = \left( \frac{\partial V}{\partial v} \right)_{E_3 = (0,0,0)}.
\]
The reproduction number $R_0$ is the spectral radius of the generation matrix $K$, which can be given as [34]
\[
R_0 = \rho(K) = \frac{aw}{\delta + \theta(s_1 + s_2 \bar{u})}.
\]

3.4. Local stability analysis
In this subsection, the local dynamics of the system (3) around the biologically feasible equilibrium points is investigated. The Jacobian matrix of the system (3) at any arbitrary point $(u, v, w)$ is given as
\[
J(u, v, w) = \begin{pmatrix}
  a_{11} & a_{12} & a_{13} \\
  a_{21} & a_{22} & a_{23} \\
  a_{31} & a_{32} & a_{33}
\end{pmatrix},
\]
where
\[
a_{11} = 1 - 2u - \frac{aw}{kv + 1} - \frac{\theta w (\beta + a \omega)}{(\beta + a \omega)^2},
\]
\[
a_{12} = -\frac{aw}{(kv + 1)^2}, \quad a_{13} = -\frac{aw \delta (\beta + u)}{(kv + 1)^2},
\]
\[
a_{21} = \frac{aw}{kv + 1}, \quad a_{22} = -\frac{\theta v}{(kv + 1)^2} - \frac{aw}{(kv + 1)^2} \delta v,
\]
\[
a_{23} = -\frac{\theta v}{(kv + 1)^2}, \quad a_{31} = \frac{\eta s \mu v^2}{(v_1 + s \mu + s \nu)^2},
\]
\[
a_{32} = -\frac{\eta s \mu v^2}{(v_1 + s \mu + s \nu)^2}, \quad a_{33} = \frac{\eta (v_1 + s \mu + s \nu)}{(v_1 + s \mu + s \nu)^2}.
\]

Theorem 3.4. The Extinction Equilibrium Point $E_0 = (0, 0, 0)$ and the Infected Prey and Predator Free Equilibrium Points $E_1 = (1, 0, 0)$ are unstable saddle points.

Proof. The eigenvalues of the Jacobean matrices $J(E_0)$ and $J(E_1)$ are $-1, -\delta, \eta$ and $-1, -\alpha, -\delta, -\eta$, respectively. Therefore, both the equilibrium points $E_0$ and $E_1$ are unstable saddle equilibrium points as $J$ has a positive eigenvalue at the respective equilibrium points.

Theorem 3.5. The Prey Free Equilibrium Point $E_2 = (0, 0, s_1)$ is locally asymptotically stable provided $\beta < s_1(\gamma - \omega)$. Otherwise, it is unstable.

Proof. The eigenvalues of the Jacobean matrix $J(E_2)$ are $-\eta, -1, -\frac{\eta}{s_1}$, $-\delta - s_1 \beta$. Thus, all eigenvalues will be negative provided $1 - \frac{\eta}{s_1}$ is negative. Therefore, $E_2$ will be locally asymptotically stable provided $\beta < s_1(\gamma - \omega)$. However, if $\beta > s_1(\gamma - \omega)$, then $E_2$ becomes unstable.

Theorem 3.6. The Predator Free Equilibrium Point $E_3 = (a \bar{u}, a \bar{v}, 0)$ is unstable.

Proof. One of the eigenvalues of the Jacobean matrix $J(E_3)$ is $\eta > 0$. Therefore, the equilibrium point $E_3$ is unstable.

Theorem 3.7. The Disease Free Equilibrium Point $E_4 = (\bar{u}, 0, 0)$ is locally asymptotically stable if
\[
R_0 < 1, \quad \eta > a_{11},
\]
where
\[
a_{11} = \frac{1}{(\beta + \bar{u} + a \omega \bar{u})^2} - 1\bar{u}.
\]

Proof. The characteristic equation of the Jacobean matrix $J(E_4)$ is
\[
a_{22} - \lambda(\lambda^2 + (\eta - a_1)\lambda - (a_1 + s_2 a_13)\eta) = 0,
\]
where
\[
a_{22} = -\frac{\eta(\beta + \bar{u})}{(\beta + \bar{u} + a \omega \bar{u})^2}, \quad a_{22} = \delta(s_1 + s_2 \theta \delta)(R_0 - 1).
\]
The roots of the characteristics equation (13) are
\[
\lambda_{1,2} = \frac{s_1 + s_2 \theta \delta}{2}, \quad \lambda_3 = \delta(s_1 + s_2 \theta \delta)(R_0 - 1).
\]
It is clear that $\lambda_3$ becomes negative for $R_0 < 1$ and $\lambda_{1,2}$ will have negative real part if $a_{11} - \eta < 0$ and $a_{11} + s_2 a_{13} < 0$. Now,
\[
a_{11} + s_2 a_{13} = -\frac{\bar{u}(1 + s_2 \omega)^2 \bar{u}^2 + B \bar{u} + C}{(\beta + \bar{u} + a \omega \bar{u})^2}.
\]
$B = 2(\beta + s_1 \omega)(1 + s_1 \omega), \quad C = (\beta + s_1 \omega)^2 + \gamma (s_2 \beta - s_1)$.
Since $2(1 + s_2 \omega)\bar{u} > 1 + s_2 \omega - (\beta + s_2 \gamma + s_1 \omega)$ and $\beta - s_1(\gamma - \omega) > 0$ (cf. (6) and (7)), we have
\[
B \bar{u} + C > (1 + s_2 \omega)(\beta - s_1(\gamma - \omega)) > 0.
\]
This implies $a_{11} + s_2 a_{13} < 0$.
Therefore, the Disease free Equilibrium Point $E_4$ will be locally asymptotically stable if $R_0 < 1$ and $\eta > a_{11}$ (ie. condition (12) holds). Hence the theorem.

Theorem 3.8. The Endemic Equilibrium Point $E_3 = (\bar{u}, \bar{v}, \bar{w})$ is locally asymptotically stable if
\[
\phi_2 > 0, \quad \phi_0 > 0, \quad \phi_1 \phi_2 - \phi_0 > 0,
\]
where
\[
\phi_2 = -a_{11} - a_{22} - \eta,
\]
\[
\phi_1 = a_{11} a_{22} - a_{21} a_{12} - \eta(s_2 a_{13} + s_1 a_{23} + a_{11} + a_{12}),
\]
\[
\phi_0 = \eta(a_{11} a_{22} - a_{12} a_{21} + s_1(a_{11} a_{23} - a_{13} a_{21}) + s_2(a_{12} a_{22} - a_{12} a_{23})),
\]
and $\phi_j$ ($i = 1, j = 1, 2, 3$) are the entries of the Jacobean matrix $J(E_3)$ which are given as
\[
a_{12} = \frac{\eta \bar{u} (\beta + \bar{u} + a \omega \bar{u})^2}{(\beta + \bar{u} + a \omega \bar{u})^2} - 1, \quad a_{13} = -\frac{a \bar{u}}{(kv + 1)^2},
\]
\[
a_{21} = -\frac{\eta (\beta + \bar{u} + a \omega \bar{u})^2}{(\beta + \bar{u} + a \omega \bar{u})^2}, \quad a_{22} = -\frac{\eta (\beta + \bar{u} + a \omega \bar{u})^2}{(\beta + \bar{u} + a \omega \bar{u})^2},
\]
\[
a_{23} = -\frac{\eta (\beta + \bar{u} + a \omega \bar{u})^2}{(\beta + \bar{u} + a \omega \bar{u})^2}.
\]
Proof. The characteristic equation of the Jacobian matrix $J(E_3)$ is
\[ \lambda^3 + \phi_2 \lambda^2 + \phi_1 \lambda + \phi_0 = 0. \]  \hfill (17)
According to Routh-Hurwitz criteria, all the roots of the characteristics equation (17) have negative real parts if and only if $\phi_2 > 0$, $\phi_0$ and $\phi_1 \phi_2 - \phi_0 > 0$.
Therefore, the Endemic Equilibrium Point $E_3$ is locally asymptotically stable provided the condition (16) is satisfied. Hence the result. \hfill \Box

3.5. Global stability

In this subsection the global stability of the Disease Free Equilibrium Point and the Endemic Equilibrium Point is investigated.

**Theorem 3.9.** The Disease Free Equilibrium Point $E_3$ is globally asymptotically stable if
\[ a < \delta, \quad s_3 \left( 1 + \frac{\sigma \alpha}{\delta} \right) < s_1 \theta. \]  \hfill (18)

**Proof.** Consider a Lyapunov function
\[ S(u, v, w) = \left( \frac{1}{\delta} (u - \hat{u} - \hat{u} \ln \frac{u}{\hat{u}}) \right) + \left( \frac{1}{\delta} (v - \hat{v} - \hat{v} \ln \frac{v}{\hat{v}}) \right). \]  \hfill (19)
Differentiating equation (19) with respect to time $t$ along the solutions of the temporal system (3) yields
\[
\frac{dS}{dt} = A(u - \hat{u})^2 + C (w - \hat{w})^2 + B(u - \hat{u})(w - \hat{w}) + Lv
\]
where
\[ A = \frac{x_{11} \hat{u}}{x_{12} x_2}, \quad B = \frac{s_2}{x_3}, \quad C = -\frac{1}{x_1}, \quad L = \frac{s_3 (w - \hat{w})}{x_3} + \frac{a \hat{u}}{1 + \kappa \hat{u} - \delta - \frac{\theta w}{1 + \sigma v}} \]
Since, $C < 0$, $\frac{dS}{dt}$ will be negative if $4AC - B^2 > 0$ and $L < 0$.

Now, under condition (18), we have
\[ 4AC - B^2 = 2x_{11} \hat{u} + s_2 (1 + 2\gamma) \hat{u} - \left( \frac{4}{x_3} \right)^2 \left( \frac{x_3}{x_3^2} \right)^2 + \left( \frac{\beta + \hat{u}}{x_3} \right)^2 > 0 \]
and
\[ L \leq -\delta \left( \frac{s_3}{s_1} - \frac{\theta}{1 + \kappa \hat{u}} \right) \]
\[ = a - \delta \left( \frac{s_3 (1 + \frac{\sigma \alpha}{\delta})}{s_1 (1 + \frac{\sigma \alpha}{\delta})} \right) \frac{w}{\hat{w}} < 0. \]

Therefore, by Lyapunov theorem, the Disease Free Equilibrium Point $E_3$ is globally asymptotically stable if condition (18) holds. Hence the theorem. \hfill \Box

**Theorem 3.10.** The Endemic Equilibrium Point $E_3$ is globally asymptotically stable if
\[ l_{11} > 0, \quad l_{22} > 0, \quad l_{13}^2 < l_{11}l_{22}, \quad l_{13}^2 < l_{12}l_{33}, \quad l_{23}^2 < l_{22}l_{33}, \]  \hfill (20)
where
\[ l_{11} = 1 - \frac{\gamma \hat{w}}{x_1}, \quad l_{12} = \frac{a \hat{u}}{2(1 + \kappa \hat{u})}, \quad l_{13} = \frac{\sigma \alpha \hat{v}}{x_2}, \quad l_{23} = \frac{\theta \sigma \hat{u}}{x_3} \]
\[ x_1 = (\beta + \hat{u} + \hat{u}) \left( \frac{\beta + \hat{u} + \hat{w}}{x_2} \right), \quad x_2 = (1 + \kappa \hat{u})(1 + \kappa \hat{v}), \quad x_3 = (1 + \sigma \hat{v})(1 + \sigma \hat{w}). \]

**Proof.** Consider a Lyapunov function
\[ S(u, v, w) = \left( u - \hat{u} - \hat{u} \ln \frac{u}{\hat{u}} \right) + \left( v - \hat{v} - \hat{v} \ln \frac{v}{\hat{v}} \right) \]
\[ + \left( w - \hat{w} - \hat{w} \ln \frac{w}{\hat{w}} \right). \]  \hfill (21)
Differentiating equation (21) with respect to time $t$ along the solutions of the temporal system (3) results
\[ \frac{dS}{dt} = -l_{11} (u - \hat{u})^2 - l_{22} (v - \hat{v})^2 + l_{12} (u - \hat{u})(v - \hat{v}) \]
\[ - l_{13} (w - \hat{w})^2 + l_{13} (u - \hat{u})(w - \hat{w}) + l_{23} (v - \hat{v})(w - \hat{w}) \]
\[ \Rightarrow P^T M P. \]
where
\[ P = (u - \hat{u}, v - \hat{v}, w - \hat{w}), \quad M = \begin{bmatrix} -l_{11} & l_{12} & l_{13} \\ l_{12} & -l_{22} & l_{23} \\ l_{13} & l_{23} & -l_{33} \end{bmatrix}. \]
Now, $\frac{dS}{dt}$ is negative if and only if the matrix $M$ is negative definite. The sufficient conditions for the matrix $M$ to be negative definite are
\[ l_{11} > 0, \quad l_{22} > 0, \quad l_{13} > 0, \quad l_{11} < l_{12}l_{22}, \quad l_{12} < l_{11}l_{33}, \quad l_{23} < l_{22}l_{33}. \]
Since, $l_{13} > 0$, $M$ is negative definite if condition (20) holds. Thus, $\frac{dS}{dt}$ becomes negative if condition (20) holds.

Therefore, by Lyapunov theorem, the Endemic Equilibrium Point $E_3$ is globally asymptotically stable if condition (20) holds. Hence the theorem. \hfill \Box

3.6. Bifurcation analysis

In this subsection the local bifurcation at the Disease Free Equilibrium Point and the Endemic Equilibrium Points is discussed with the help of Sotomayor Theorem [35].

**Theorem 3.11.** When the bifurcation parameter $a$ passes through the critical value $a^* = \frac{\delta l_{11} + \delta l_{12}l_{22}}{\delta l_{12}}$ (i.e. $R_0 = 1$), the temporal system (3) at the Disease Free Equilibrium Point $E_3 = (\hat{u}, \hat{v}, \hat{w})$ has

1. no saddle-node bifurcation,
2. a transcritical bifurcation if $ag \neq \theta h + k \delta (s_1 + s_2 \theta h) - \theta s \hat{w},$
3. a pitchfork bifurcation if $ag = \theta h + k \delta (s_1 + s_2 \theta h) - \theta s \hat{w},$ \( \sigma \neq \kappa \) and $\sigma \hat{w} \neq \hat{h},$

where
\[ g = \frac{\delta (s_1 + s_2 \theta h) - s_3 a_{13}}{a_{11} + s_2 a_{13}}, \quad \hat{h} = \frac{s_3 a_{13} + s_3 a_{11}}{a_{11} + s_2 a_{13}}, \quad a_{11} = \frac{\gamma \hat{w} (\beta + \hat{u} + \hat{w})}{(\beta + \hat{u} + \hat{w})^2} - \hat{u}, \quad a_{13} = -\frac{\gamma \hat{w} (\beta + \hat{u})}{(\beta + \hat{u} + \hat{w})^2}. \]
Theorem 3.12. The temporal system (3) undergoes Hopf bifurcation around the Disease Free Equilibrium Point $E_1 = (\hat{u}, 0, \hat{w})$, when the bifurcation parameter $\eta$ crosses the critical value $\eta_{cr} = \alpha_{11}$, where $\alpha_{11} = \frac{\tilde{a}_{13}}{(\tilde{a}_{11} + \tilde{a}_{12})}$. 

Proof. From the eigenvalues of the Jacobian matrix $J(E_1)$, which are given in (15), we can see that $\lambda_1$ is real, $\lambda_2$, $\lambda_3$ are purely imaginary if and only if there is a critical value of $\eta = \eta_{cr}$. Thus, at $\eta = \eta_{cr}$, we have $\lambda_1 = i \sqrt{p_2}$, $\lambda_2 = -i \sqrt{p_2}$, $\lambda_3 = \delta(s_1 + s_2\hat{w})(R_0 - 1)$, where $p_2 = -(\alpha_{11} + s_2\alpha_{13})$ and $\alpha_{11} = \frac{\tilde{a}_{13}}{(\tilde{a}_{11} + \tilde{a}_{12})}$. 

Now, differentiating equation (13) with respect to $\eta$ gives 

$$\frac{d\lambda}{d\eta} \bigg|_{\eta=\eta_{cr}} = \left[ \frac{\lambda p_1 + p_2}{p_1^2 + 4p_2} \right] \frac{1}{\eta_{cr}} + \left[ \frac{2p_1 \sqrt{p_2}}{p_1^2 + 4p_2} \right] \frac{1}{\eta_{cr}} = 0.5 \neq 0,$$

where $p_1 = -(\alpha_{11} + s_2\alpha_{13})$. 

This implies $\text{Re} \left( \frac{d\lambda}{d\eta} \right) \bigg|_{\eta=\eta_{cr}} = 0.5 \neq 0$. 

Therefore, the temporal system (3) undergoes a Hopf bifurcation around the Disease Free Equilibrium Point at a certain critical value of the parameter $\eta = \eta_{cr}$. □

Theorem 3.13. The temporal system (3) will experience a Hopf bifurcation around the Endemic Equilibrium Point $E_2$ when the bifurcation parameter passes the critical value $\eta = \eta_{cr}$ if the following conditions hold.

(i) $\phi_2(\eta) > 0$ and $\phi_1(\eta) > 0$ at $\eta = \eta_{cr}$
(ii) $H(\eta) := \phi_1(\eta)\phi_2(\eta) - \phi_0(\eta) = 0$ at $\eta = \eta_{cr}$
(iii) $\text{Re} \left( \frac{d\lambda}{d\eta} \right) \bigg|_{\eta=\eta_{cr}} = 0$, 

where $\phi_2, \phi_1, \phi_0$ are as defined in theorem (3.8).

Proof. From conditions (i) and (ii), it follows that the temporal system (3) will have one negative root and two purely imaginary roots. Thus, for $\eta = \eta_{cr}$, the characteristic equation of the Jacobian matrix $J(E_2)$ given in (17), must be written as $(\lambda^2 + \phi_1(\lambda + \phi_2) = 0$, and gives the three roots: $\lambda_3 = i\sqrt{\epsilon_1}$, $\lambda_2 = -i\sqrt{\epsilon_1}$ and $\lambda_3 = -\phi_2$. 

For all values of the bifurcation parameter $\eta$, the eigenvalues $\lambda_{1,2}$ are of the form $\lambda_1 = \phi_2(\eta) + i\phi_1(\eta)$ and $\lambda_2 = \phi_2(\eta) - i\phi_1(\eta)$, in which $\phi_1(\eta)$ and $\phi_2(\eta)$ are real. Substituting $\lambda = \phi_1(\eta) + i\phi_2(\eta)$ in to the characteristics equation (17) gives

$$(\eta_1(\eta) + i\eta_2(\eta))^3 + \phi_2(\eta)(\eta_1(\eta) + i\eta_2(\eta))^2$$

$$+ \phi_1(\eta)(\eta_1(\eta) + i\eta_2(\eta)) + \phi_1(\eta) = 0.$$ (22)

Differentiating equation (22) with respect to the bifurcation parameter $\eta$ and separating real and imaginary parts gives

$$A(\eta)\eta_1(\eta) - B(\eta)\eta_2(\eta) + C(\eta) = 0,$$ (23)

$$B(\eta)\eta_2(\eta) + A(\eta)\eta_1(\eta) + D(\eta) = 0,$$ (24)

where

$$A(\eta) = \phi_1(\eta) + 2\phi_2(\eta)\eta_1(\eta) + 3\phi_1(\eta)(\eta_1(\eta) - \eta_2^2(\eta)),$$

$$B(\eta) = 2\eta_2(\eta) + 3\phi_1(\eta)\eta_2(\eta),$$

$$C(\eta) = (\eta_1^2(\eta) - \eta_2^2(\eta))\phi_1(\eta) + \phi_1(\eta)\phi_1(\eta) + \phi_1(\eta),$$

$$D(\eta) = 2(\eta_1(\eta)\phi_2(\eta) + \phi_1(\eta)\eta_2(\eta)).$$

Solving the simultaneous equations (23) and (24) for $\eta_1$ gives

$$\text{Re} \left( \frac{d\lambda}{d\eta} \right) \bigg|_{\eta=\eta_{cr}} = \left[ \frac{A(\eta)\eta_1(\eta) + B(\eta)\eta_2(\eta)}{A(\eta)^2 + B(\eta)^2} \right] \bigg|_{\eta=\eta_{cr}}$$

$$= \left[ \frac{\phi_1(\eta)\phi_1(\eta) + \phi_2(\eta)\phi_1(\eta) + \phi_2(\eta)}{2(\phi_2^2(\eta) + \phi_1^2(\eta))} \right] \bigg|_{\eta=\eta_{cr}}.$$
Thus, for any given \( \epsilon > 0 \) there exists \( \tau_1 > \tau_0 \), such that \( v(x,t) \leq \frac{w}{k} + \epsilon_1 \), for \( (x,t) \in \Omega \times [\tau_1, \infty) \). As a result, for \( (x,t) \in \Omega \times [\tau_1, \infty) \), the equation of \( u \) satisfies

\[
\begin{align*}
u_t - D_1 \Delta u &\leq \eta \left( \frac{1 - \frac{w}{\kappa k} - \epsilon}{s_1 (1 + \gamma)} \right) u \\
&\leq \eta \left( \frac{s_1 (1 + \gamma)}{s_1 (1 + \gamma)} - \frac{\frac{w}{\kappa k} + \epsilon}{s_1 (1 + \gamma) - \epsilon} \right) u.
\end{align*}
\]

Since \( \epsilon > 0 \) and \( \epsilon_1 > 0 \) are arbitrary, Lemma 4.1 yields

\[
\lim_{t \to \infty} \sup \max_{\Omega} u(\cdot, t) \leq \frac{a s_1}{\delta k}.
\]

\[\Box\]

\textbf{Theorem 4.3.} The system (2) is persistent if for any nonnegative initial data \((u(x,0), v(x,0), w(x,0)) \) with \((u(x,0), v(x,0), w(x,0)) \neq (0,0,0)\), there exist positive constants \( \sigma_0, \sigma_1 \) and \( \sigma_2 \) such that the solutions \((u(x,t), v(x,t), w(x,t))\) of (2) satisfy

\[
\liminf_{t \to \infty} \min_{\Omega} u(\cdot, t) \geq \sigma_0, \quad \liminf_{t \to \infty} \min_{\Omega} v(\cdot, t) \geq \sigma_1, \quad \liminf_{t \to \infty} \max_{\Omega} u(\cdot, t) \geq \sigma_2.
\]

\textbf{Definition 4.1.} The system (2) is said to be persistent if for any nonnegative initial data \((u(x,0), v(x,0), w(x,0)) \neq (0,0,0)\), there exist positive constants \( \sigma_0, \sigma_1 \) and \( \sigma_2 \) such that the solutions \((u(x,t), v(x,t), w(x,t))\) of (2) satisfy

\[
\liminf_{t \to \infty} \min_{\Omega} u(\cdot, t) \geq \sigma_0, \quad \liminf_{t \to \infty} \min_{\Omega} v(\cdot, t) \geq \sigma_1, \quad \liminf_{t \to \infty} \max_{\Omega} u(\cdot, t) \geq \sigma_2.
\]
Since $\epsilon$ and $e^* \epsilon$ are arbitrary, we have

$$\liminf_{t \to \infty} \min \frac{u_i(t)}{u_i} \geq s_1 + s_2 l_i + s_3 l_i = l_i > 0.$$  \hfill (30)

Hence, from (28), (29) and (30), it follows that the system (2) is persistent.

4.2. Local stability

In this subsection, the stability of the constant steady states of the spatio-temporal system (2) is discussed. It is easy to see that the constant steady states of the spatio-temporal system (2) are the six biologically feasible equilibrium points of the temporal system (3).

Now, denote $\mathbf{u} = (u(x, y, t), v(x, y, t), w(x, y, t))^T$ and $G(\mathbf{u}) = (G_1(\mathbf{u}), G_2(\mathbf{u}), G_3(\mathbf{u}))^T$. Let $0 = \mu_0 < \mu_1 < \mu_2 < \mu_3 < \ldots$ be the eigenvalues of the operator $-\Delta$ on $\Omega$ under the homogeneous Neumann boundary condition. $\Omega(\mu_i)$ is the eigenspace corresponding to the eigenvalue $\mu_i$, $X_j := \{e_i \in \mathbb{R}^l : c \in \mathbb{R}^l\}$, where $\{\psi_i\}$ are orthonormal basis of $X_j$ for $j = 1, 2, 3, \ldots ; dim(Q), X := \{u \in (u, v, w) \in \Omega(1) \mathbb{R}^3 \} \cap 0 = 0$ on $\partial \Omega$, and so $\mathbf{x} = \mathbf{x}_\infty$, where $\mathbf{x}_\infty = \sum_{j=1}^{\infty} \mathbf{x}_j$.

Linearization of the system (2) at $E_n$ $(n = 0(1)5)$ yields

$$\mathbf{u} = \mathbf{L}u, \quad \mathbf{L} = D + J(E_n), \quad D = diag(1, D_2, D_3),$$

where $J(E_n)$ is the Jacobian Matrix which is defined in section 3.4. The eigenvalues $\lambda_i$, $i \geq 1$, is an eigenvalue of $L$ on $X_i$ if and only if it is an eigenvalue of the matrix $L_i = -\mu_i D + J(E_n)$.

Theorem 4.4. The equilibrium points $E_0$, $E_1$ and $E_2$ are unstable.

Proof. For $i = 0$, the operator $L_i$ at $E_0$, $E_1$, and $E_2$ has a common positive eigenvalue $\eta > 0$. This shows that the equilibrium points $E_0$, $E_1$ and $E_2$ are unstable.

Theorem 4.5. The equilibrium point $E_2 = (0, 0, s_3)$ is uniformly asymptotically stable if $\beta + s_3 \omega < s_2 \gamma$.

Proof. The eigenvalues of the operator $L_i$ at $E_2$ are $\lambda_{1/1} = -\delta - s_2 \theta - D_2 \mu_1$, $\lambda_{2/2} = -\eta - D_1 \mu_1$, and $\lambda_{3} = 1 - \frac{\eta}{s_2 \gamma}$. Hence, all the eigenvalues will be negative if $\beta + s_3 \omega < s_2 \gamma$. Thus, there exist some positive numbers $\rho_i$ such that $Re\{\lambda_{1/1}\}, Re\{\lambda_{2/2}\}, Re\{\lambda_{3}\} \leq -\rho_i$. Let $\rho = \min\{\rho_i\}$. Then, $\rho > 0$ and $Re\{\lambda_{1/1}\}, Re\{\lambda_{2/2}\}, Re\{\lambda_{3}\} \leq -\rho$. Consequently, the spectrum of $L$ lies in $\{Re \leq -\rho\}$. Thus, Theorem 5.1.1. of Dan Henry (p. 98) [37] concludes the uniform asymptotically stability of $E_2$.

Remark 4.1. If $\beta + s_3 \omega < s_2 \gamma$ then the temporal stability of equilibrium point $E_2$ ensures the uniform stability of the spatio-temporal system (2) in the vicinity of $E_2$. That is, diffusion do not have an effect on the stability of the locally asymptotically stable equilibrium point $E_2$.

Theorem 4.6. The Disease Free Equilibrium Point $E_4 = (\check{u}, 0, \check{v})$ is uniformly asymptotically stable if

$$R_0 < 1, a_{11} < 0.$$ \hfill (31)

Proof. The characteristics equation of the operator $L_i$ at $E_4$ is

$$\lambda = (\lambda - (s_1 + s_2 \check{u})(R_0 - 1) - \mu_i D_3)(\lambda^2 + TR_1 \lambda + det_1) = 0,$$ \hfill (32)

where

$$TR_1 = (a_{11} - \eta) - (1 + D_3)\eta,$$

det_1 = -\eta(a_{11} + s_1 a_{11}) + (\eta - a_{11} D_3)\eta + D_3 \mu_i^2$$

and $a_{11}$ and $a_{13}$ are as in Theorem 3.7.

Hence, under condition (31) and the fact that $a_{11} + s_2 a_{23} < 0$ (cf. Theorem 3.7), we can see that $TR_1 < 0, det_1 > 0$ and $(\delta(s_1 + s_2 \check{u})(R_0 - 1) - \mu_i D_3) < 0$ for all $i \geq 0$. From the Routh-Hurwitz criterion it follows that, for each $i \geq 0$, all the three roots $\lambda_{1/1}$, $\lambda_{2/2}$ and $\lambda_{3}$ have negative real parts. Thus, there exist some positive numbers $\rho_i$ such that $Re\{\lambda_{1/1}\}, Re\{\lambda_{2/2}\}, Re\{\lambda_{3}\} \leq -\rho_i$.

Let $\rho = \min\{\rho_i\}$. Then, $\rho > 0$ and $Re\{\lambda_{1/1}\}, Re\{\lambda_{2/2}\}, Re\{\lambda_{3}\} \leq -\rho$. Consequently, the spectrum of $L$ lies in $\{Re \leq -\rho\}$. Thus, Theorem 5.1.1. of Dan Henry (p. 98) [37] concludes the uniform asymptotically stability of $E_4$.

Theorem 4.7. The Endemic Equilibrium Equilibrium $E_5 = (\check{u}, \check{v}, \check{w})$ is uniformly asymptotically stable if

$$a_{11} < 0, a_{22} < 0,$$ \hfill (33)

where $a_{11}, a_{12}$ and $a_{22}$ are as defined in Theorem 3.8.

Proof. The characteristics equation of $L_i$ at the Endemic Equilibrium Point $E_5$ is given by

$$\lambda^3 + P_2 \lambda^2 + P_1 \lambda + P_0 = 0,$$ \hfill (34)

where $P_2 = (1 + D_2 + D_3)\mu_1 + \phi_2$, $P_1 = (D_2 + D_3)(1 + D_3)\mu_1 - (a_{22}(1 + D_3) + a_{11}(D_2 + D_3))\eta_i$, \(\eta(1 + D_2)\mu_1 + \phi_1$, $P_0 = D_2 D_3 \mu_1^2 + (\eta D_2 - (a_{22} + a_{12} D_2) D_3) \mu_2^2 - a_{12} a_{22} D_3 D_4 \mu_1$, $a_{12} a_{22} D_3 + (a_{22} a_{33} - a_{22} (a_{11} + a_{13} D_2) D_3) \mu_1 + \phi_0$ and $a_{33} (r = 1, 3; s = 1, 2, 3), \phi_2, \phi_3, \phi_4, \phi_5$ and $\phi_0$ are as defined in Theorem 3.8.

Algebraic manipulations and simplifications give

$$P_2 |\phi_2| - \phi_0 = M_1 \mu_1^2 + M_2 \mu_1^3 + M_3 \mu_1^2 + M_4,$$ \hfill (35)

where $M_1 = (1 + D_2)(1 + D_3) D_2 + D_3$, $M_2 = -a_{11}(D_2 + D_3) (2 + D_3) - a_{22}(1 + D_3)(1 + 2 D_2 + D_3)$, $M_3 = -a_{33}(1 + D_3) - a_{33} a_{23} (D_2 + D_3) - a_{12} a_{23} (1 + D_2)$, $M_4 = (a_{12} a_{22} a_{33} a_{23} - a_{12} a_{32} + (a_{13} + a_{12} D_2) \eta)$, $\phi_0 = (a_{12} a_{22} + a_{12} a_{23} + a_{13} (a_{23} + a_{12} \eta)) \eta$, $\phi_0 = (-a_{12} + a_{22} + a_{23} \eta)^2 \eta$.

Now, under condition (33) and the fact that $a_{12} > 0, a_{13} < 0, a_{22} > 0, a_{23} < 0$ and

$$a_{23} + a_{12} \eta = \delta + s_2 \theta + (s_3 \eta + \sigma \delta \eta) \eta < 0,$$ \hfill (36)

we have $\phi_2 > 0, \phi_1 > 0, \phi_0 > 0$ and $M_4 (n = 1, 2, 3, 4) > 0$. This implies $\phi_2 > 0, \phi_1 > 0, \phi_0 > 0, \phi_1, \phi_2 - \phi_0 > 0, \forall i \geq 0$.

Hence, the Routh-Hurwitz criterion implies that, for each $i \geq 0$, all the three roots $\lambda_{1/1}$, $\lambda_{2/2}$ and $\lambda_{3}$ have negative real parts. Thus, there exist
some positive numbers \( \rho_i \) such that \( \text{Re}(\lambda_1), \text{Re}(\lambda_2), \text{Re}(\lambda_3) \leq -\rho_i \forall i \geq 0 \). Let \( \rho = \min(\rho_i) \), Then, \( \rho > 0 \) and \( \text{Re}(\lambda_1), \text{Re}(\lambda_2), \text{Re}(\lambda_3) \leq -\rho \forall i \geq 0 \). Consequently, the spectrum of \( \mathcal{L} \) lies in \( \{ \text{Re} \leq -\rho \} \).

Therefore, by Theorem 5.1.1. of Dan Henry (p. 98) [37], the Endemic Equilibrium Point is uniform asymptotically stable. \( \Box \)

5. Numerical simulation

In this section, we present some numerical simulation results of the temporal system (3) and the spatio-temporal system (2) to support our analytical findings stated in the previous sections. The numerical simulations are performed with the help of MATLAB-R2014a, Mathematica-11 and MatCont-6p1 software packages.

5.1. Temporal system

In this subsection, we consider the following two sets of parametric values.

\[
\begin{align*}
\alpha &= 0.03, \gamma = 1.5, \theta = 0.6, \delta = 0.1, \sigma = 1, \kappa = 1, \\
\beta_1 &= 0.1, s_2 = 0.5, s_3 = 0.4, \beta = 0.15, \omega = 0.2, \\
\alpha &= 0.8, \gamma = 0.6, \theta = 0.6, \delta = 0.1, \sigma = 1, \kappa = 0.1, \\
\beta_1 &= 0.1, s_2 = 0.2, s_3 = 0.4, \beta = 0.1, \omega = 0.1.
\end{align*}
\]

(35) \hspace{1cm} (36)

For the data set (35), the Disease Free Equilibrium Point \( E_0 = (0.23954, 0.021977) \) exists. For \( \eta = 0.25 \), the temporal system (3) undergoes a transcritical bifurcation around \( E_0 \) at \( \alpha = 0.96794813 \) as shown in Fig. 1. Whereas, for the data set (35), the temporal system (3) undergoes a Hopf bifurcation about \( E_0 \) when the parameter \( \eta \) crosses the critical value \( \eta = \eta_{cr} = 0.18067445 \) as shown in Fig. 2.

From the Hopf bifurcation diagram (cf. Fig. 2), we can infer that the temporal system (3) exchanges stability when the bifurcation parameter \( \eta \) crosses its threshold value \( \eta_{cr} = 0.18067445 \). The local stability analysis also shows that, for the parametric values as in (35), the Disease Free Equilibrium Point \( E_0 \) will be locally asymptotically stable for \( \eta > \eta_{cr} = 0.18067445 \) and unstable for \( \eta < \eta_{cr} = 0.18067445 \). The existence of Hopf bifurcation ensures the existence of periodic solution, leading to the existence of a limit cycle, for \( \eta < \eta_{cr} = 0.18067445 \). Fig. 3 shows the local stability of the temporal system (3) for the parametric values as in (35) and \( \eta = 0.25 \). Moreover, the temporal system (3) is globally asymptotically stable around \( E_0 \) as shown in Fig. 4.

Fig. 5 shows the existence of periodic solution of the system (3) around the Disease Free Equilibrium Point \( E_0 \) for the parametric values as in (35) and \( \eta = 0.1 \).

For the data set (36), the Endemic Equilibrium Point \( E_1 = (0.303278, 0.357855, 0.303798) \) exists. For \( \eta = \eta_{cr} = 0.32517 \), the temporal system (3) undergoes a Hopf bifurcation around \( E_1 \) (cf. Fig. 6).
Moreover, the $\eta < \eta_{cr}$ for the parametric values as in (35) and $\eta = 0.1$. (a) The time series solution (b) The phase portrait showing the existence of a limit cycle.

The Hopf bifurcation diagram (cf. Fig. 6) of the temporal system (3) around the Endemic Equilibrium Point $E_i$ shows the existence of exchange of stability of the temporal system (3) around $E_i$ when the bifurcation parameter $\eta$ passes its threshold value $\eta = \eta_{cr} = 0.32517$. From the local stability analysis of the temporal system (2), one can see that, for the parametric values as in (36), the Endemic Equilibrium Point $E_i$ is locally asymptotically stable for $\eta > \eta_{cr} = 0.32517$ and unstable for $\eta < \eta_{cr} = 0.32517$. The existence of Hopf bifurcation ensures the existence of periodic solution, leading to the existence of a limit cycle, for $\eta < \eta_{cr} = 0.32517$.

The bifurcation diagram (cf. Fig. 7 and Fig. 8) of the temporal system (3) around the Endemic Equilibrium Point $E_i$ shows the existence of exchange of stability of the temporal system (3) around $E_i$ when the bifurcation parameter $\alpha$ passes its threshold value $\alpha = \alpha_{cr} = 0.7474$. Moreover, it can be seen that the Endemic Equilibrium Point disappears when the value of $\alpha$ is less 0.312. In this case, a trans-critical bifurcation occurs at the Endemic Equilibrium Point. From the local stability analysis of the temporal system (3), one can see that, for the parametric values as in (36) and $\eta = 0.2$, the Endemic Equilibrium Point $E_i$ is locally asymptotically stable for $0.312 < \alpha < \alpha_{cr} = 0.7474$ and unstable for $\alpha > \alpha_{cr} = 0.7474$. The existence of Hopf bifurcation ensures the existence of periodic solution, leading to the existence of a limit cycle, for $\alpha > \alpha_{cr} = 0.7474$.

Fig. 9 shows the local stability of the temporal system (3) for the parametric values as in (36) and $\eta = 0.6$. Moreover, the temporal system (3) is globally asymptotically stable around $E_i$ as shown in Fig. 10. Fig. 11 shows the existence of periodic solution of the system (3) around the Endemic Equilibrium Point $E_i$ for the parametric values as in (36) and $\eta = 0.2$.

Fig. 12 shows the stability of the temporal system (3) around the Endemic Equilibrium Point for the parametric values as in (36) except $\alpha = 0.7$ and $\eta = 0.2$. Thus, from Figs. 11 and 12, we can observe that a decrease in the amount of prey infection leads to damping of the oscillation and results in the stability of the temporal system (3) around the Endemic Equilibrium Point.

5.2. Diffusive system

In this subsection, numerical simulation results of the stability of the spatio-temporal system (2) around the Disease Free Equilibrium Point and the Endemic Equilibrium Point are presented. Fig. 13 shows that the spatio-temporal system (2) is locally asymptotically stable around the Disease Free Equilibrium Point $E_d = (0.368648, 0.0, 0.284324)$ for the parametric values as in (35) except $\beta = 0.25$, $\eta = 0.25$, $D_2 = 0.01$ and $D_1 = 10$. Fig. 13(a), (b) and (c) represent the time series solution of the spatio-temporal system (2) around
Fig. 9. Stability behavior of the temporal system (3) around \( E_i \) for the parametric values as in (36) and \( \eta = 0.6 \). (a) The times series solution (b) The phase portrait showing the local asymptotically stability of \( E_i \).

Fig. 10. The phase portrait of the temporal system (3) showing the global asymptotic stability of the system (3) around the point \( E_i \) for the parametric values as in (36) and \( \eta = 0.6 \).

the Disease Free Equilibrium Point \( E_i \) at spatial locations \( x = 500, 2000 \) and \( x = 4000 \), respectively. Fig. 13(d) represents the spatial distribution of the species at \( t = 600 \).

Fig. 11. Dynamical behavior of the temporal system (3) around \( E_i \) for the parametric values as in (36) and \( \eta = 0.2 \). (a) The time series solution (b) The phase portrait showing the existence of a limit cycle.

Fig. 12. Stability behavior of the temporal system (3) around \( E_i \) for the parametric values as in (36) except \( \alpha = 0.7 \) and \( \eta = 0.2 \). (a) The times series solution (b) The phase portrait showing the local asymptotically stability of \( E_i \).
Fig. 13. Dynamical behavior of the spatio-temporal system (2) around $E_4$ for the parametric values as in (35) except $\beta = 0.25$, $\eta = 0.25$, $D_2 = 0.01$ and $D_3 = 10$. (a) Time series solution at $x = 500$ (b) Time series solution at $x = 2000$ (c) Time series solution at $x = 4000$ (d) Spatial distribution at time $t = 600$.

Fig. 14. Dynamical behavior of the spatio-temporal system (2) around $E_5$ for the parametric values as in (36) except $\kappa = 0.9$, $\eta = 0.325$, $D_2 = 0.01$ and $D_3 = 10$. (a) Time series solution at $x = 500$ (b) Time series solution at $x = 2000$ (c) Time series solution at $x = 4000$ (d) Spatial distribution at time $t = 600$. 
value, the Endemic Equilibrium Point loses its stability and Hopf bifurcations occurs. The three population exhibits an oscillatory behavior. Wang et al. [38] points out that the increase of the infectious rate can lead to the lost of stability. Thus, our results are inline with the results of Wang et al. [38].

The main novelty between our work and other recent works is the inclusion of prey infection, with nonlinear incidence rate, mixed type of functional response for the susceptible and infected prey population, and the modified Leslie-Gower predator dynamics. These additional ecological components enrich the dynamics of the system and make the system more realistic than the existing models.

Both analytical and numerical simulation results show the complex and rich dynamics of the system under consideration. The future work can be carried out by incorporating the horizontal and vertical disease transmission to the predator population with ecological factors like refuge, additional food and delay and the formation of spatial and spatio-temporal patterns.

6. Conclusions

In this paper, a spatio-temporal eco-epidemiological model with Beddington-DeAngelis functional response and the modified Leslie-Gower type predator dynamics under homogeneous Newman boundary condition is considered. The prey population is assumed to be infected with a disease and the disease spread in the system according to the nonlinear incidence rate.

It is observed that the temporal system (3) has six biologically feasible equilibrium points: $E_0$, $E_1$, $E_2$, $E_3$ and $E_4$. It is also seen that the six biologically feasible equilibrium points of the temporal system (3) are also the constant equilibrium points of the spatio-temporal system (2). The equilibrium points $E_0$, $E_1$, and $E_2$ are unstable both in the presence and absence of diffusion. The prey free equilibrium point is locally asymptotically stable if and only if $\beta < \alpha (\gamma - \omega)$, in the presence and absence of diffusion. The local and global stability conditions for the Disease Free Equilibrium Point and Endemic Equilibrium Point of the temporal system (3) are obtained. Moreover, the local stability conditions for the Disease Free Equilibrium Point and Endemic Equilibrium Point of the spatio-temporal system (2) are established.

From the results of Theorem 3.7 and Theorem 4.6, we can conclude that the presence of diffusion will not have an effect on the dynamics of the system (3) around the Disease Free Equilibrium Point as long as the reproduction number is less than unity and the first entry of the corresponding Jacobian matrix is negative.

The infected prey will be extinct if the prey infection rate is less than the death rate of the infected prey (cf. Theorem 3.3). The stability of the Disease Free Equilibrium Point implies that total extinction of the species is not possible and hence the introduction of infected prey into the system may act as a biological control. The Bifurcation analysis of the temporal system (3) shows that the temporal system (3) undergoes a transcritical, pitchfork and Hopf bifurcations under certain conditions.

Numerical simulations are performed to support the analytical results. The numerical simulation results show the existence of transcritical and Hopf bifurcation of the temporal system (3) around the Disease Free Equilibrium Point (cf. Fig. 1 and Fig. 2), and the existence of Hopf bifurcation (cf. Fig. 6 and Fig. 7) and trans-critical bifurcation (cf. Fig. 7) at the Endemic Equilibrium Point. The emergence of chaotic pattern for the system (2) is shown in Fig. 15.

We conclude that the prey infection rate, $\alpha$, has both stabilizing and destabilizing effect on the Endemic Equilibrium Point. When it is less than its critical value $\alpha_c$, the susceptible prey, infected prey and predator will coexist and approaches to the Endemic Equilibrium Point. However, when the prey infection rate passes through some critical

\[
\nu(x, 0) = 0.362293
\]

\[
\omega(x, 0) = 0.308113 + 10^{-5}(x - 1200)(x - 2800).
\]

Declarations

Author contribution statement

Dawit Melese and Shiferaw Feyissa: Conceived and designed the experiments; Performed the experiments; Analyzed and interpreted the data; Contributed materials, analysis tools or data; Wrote the paper.

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