Research Article

Noor Badshah and Haji Akbar*

Stability analysis of fractional order SEIR model for malaria disease in Khyber Pakhtunkhwa

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Abstract: We discussed stability analysis of susceptible-exposed-infectious-removed (SEIR) model for malaria disease through fractional order and check that malaria is epidemic or endemic in Khyber Pakhtunkhwa (Pakistan). We show that the model has two types of equilibrium points and check their stability through Routh-Hurwitz criterion. We find basic reproductive number using next-generation method. Finally, numerical simulations are also presented.

Keywords: SEIR model, fractional order, Routh-Hurwitz criterion, next-generation method

MSC 2020: 92B15, 92B05

1 Introduction

Malaria is an infectious disease having large economic and health impact on society. Although malaria disease is interrogated for several years but still it is a serious health problem in many countries of the world [1]. Malaria is an infectious disease caused by the plasmodium parasite and transmitted between humans through the bite of the female Anopheles mosquito. There are four types of plasmodium parasite which infect humans, namely, Plasmodium falciparum, Plasmodium vivax, Plasmodium malariae and Plasmodium ovale [2]. P. falciparum can cause serious complications and can be fatal if untreated. Pregnant women and children under the age of five are most infected from malaria because they have a lower immunity as compared to others in the community. According to WHO there are 243 million malaria cases across the world and 7.8 million of deaths occur every year. Therefore, we also check whether the malaria disease is epidemic or endemic in Khyber Pakhtunkhwa (KP, Pakistan).

In the early twentieth century, mathematical models play important role in the study of analyzing the spread of infectious disease [3]. Although there is a lot of work on integer order in epidemiological models [4,5], due to the effective nature of fractional differential equation (FDE) many models in science such as physics, fluid mechanics, mechanical system, and other fields of engineering and epidemiology have been successfully formulated and analyzed [6]. Up to now many fractional results have been presented which are very useful [7,8]. There are three basic definitions of FDE which are the Grunwald-Letnikov, the Riemann-Liouville, and the Caputo formula [9]. We used the Caputo formula due to its convenience for initial condition of the FDE. The Caputo fractional derivative of order \( \alpha \) for the function \( f(t) \) is expressed as:
where \( t > a \) and \( a, t, n \in \mathbb{R} \).

The rest of the paper is arranged as follows: in Section 2, the model derivation is given. The equilibrium points and basic reproductive number are given in Section 3. Stability of equilibrium points is analyzed in Section 4. In Section 5, numerical simulations of the model are given. In Section 6, conclusion is given.

2 Model formulation

In the following model [10], the total population is distributed into four classes with respect to disease. The classes consist of susceptible population \( S \) (the individuals who are able to catch the disease), exposed population \( E \) (the individuals who have been infected but eventually are not infectious), infectious population \( I \) (the individuals who are able to conduct the disease), and recovered population \( R \) (the individuals who are recovered from disease). Figure 1 shows the disease transmission of model.

The SEIR model can be expressed as a system of FDEs of order \( \alpha \) given as:

\[
\begin{align*}
\frac{d^\alpha S}{dt^\alpha} &= AN - \beta SI - dS, \\
\frac{d^\alpha E}{dt^\alpha} &= \beta SI - (\delta + d)E, \\
\frac{d^\alpha I}{dt^\alpha} &= \delta E - (\gamma + d)I, \\
\frac{d^\alpha R}{dt^\alpha} &= \gamma I - dR.
\end{align*}
\]  

(2)

\( S(0) = S_0, E(0) = E_0, I(0) = I_0, R(0) = R_0, N = S + E + I + R \in \mathbb{R}_+, \) and \( 0 < \alpha \leq 1 \).

In the SEIR model, \( A \) is the natural birth rate, \( \beta \) denotes the infectious rate, \( \delta \) denotes the incubation rate, \( \gamma \) is the recovery rate and \( d \) denotes the disease-related death. We take fractional order derivative instead of inter order derivative because fractional order derivative has a unique property called memory effect which does not exist in integer order derivative.

3 Equilibrium points and basic reproductive number of system (2)

To find the equilibrium points of the model consider:

\[
AN - \beta SI - dS = 0,
\]

(3)
\[ \beta SI - (\delta + d)E = 0, \quad (4) \]
\[ \delta E - (y + d)I = 0, \quad (5) \]
\[ yI - dR = 0. \quad (6) \]

Adding (3) and (4) we get
\[ AN - dS - (\delta + d)E = 0, \]
then
\[ S = \frac{AN - (\delta + d)E}{d}. \quad (7) \]

From (5) we get
\[ I = \frac{\delta E}{y + d}, \quad (8) \]
while from (6) we obtain that
\[ R = \frac{yI}{d}. \quad (9) \]

Now putting (7) and (8) in (4) it follows that
\[ \frac{E[\beta \delta AN - (\delta + d)E] - (\delta + d)]}{d(y + d)} = 0, \quad (10) \]
so either \( E = 0 \) or
\[ E = \frac{\beta \delta AN - d(y + d)(\delta + d)}{d(\delta + d)}. \quad (11) \]

For \( E = 0 \), (7) becomes
\[ S = \frac{AN}{d}. \quad (12) \]
While from (8) we obtain \( I = 0 \). Also from (9) we get \( R = 0 \). It follows that disease-free equilibrium points are given by
\[ X_0 = \left[ \frac{AN}{d}, 0, 0, 0 \right]. \quad (13) \]

From (11) there is another equilibrium with
\[ E^* = \frac{AN}{\delta + d} - \frac{(y + d)d}{\beta \delta}. \quad (14) \]
By substituting the value of \( E^* \) in (7) we obtain
\[ S^* = \frac{(\delta + d)(y + d)}{\beta \delta}. \quad (15) \]
Again using the value of \( E^* \) in (8) we obtain
\[ I^* = \frac{\delta AN}{(\delta + d)(y + d)} - \frac{d}{\beta}. \quad (16) \]
Finally, using the value of \( I^* \) in (9) we get
\[ R^* = \frac{y}{d} \left[ \frac{\delta AN}{(y + d)(\delta + d)} - \frac{d}{\beta} \right]. \quad (17) \]

So the endemic equilibrium point \( X^* = (S^*, E^*, I^*, R^*) \) is given by expressions (14)–(17).
To find basic reproductive number $R_0$, we apply the next-generation method [11]. Suppose there are $n$ disease compartments in the model (2). Let $X = (x_1, x_2, \ldots, x_n)$ where $i = 1, 2, 3, \ldots, n$ represent individuals in $i$th infected class. Further the rate of expression of new infections in $i$th compartment is denoted by $F_i$ and the rate of expression of disease progression in $i$th compartment is denoted by $V_i$. Then we linearize the $i$th infected compartment about disease free equilibrium for finding the spreading of disease in the population.

Next-generation matrix is made from partial derivative of $F_i$ and $V_i$ as:

$$ F = \frac{\partial F_i(x_0)}{\partial x_j} \quad \text{and} \quad V = \frac{\partial V_i(x_0)}{\partial x_j}, \quad i = 1, 2, 3, \ldots, n, $$

where $x_0$ is the disease-free equilibrium point. The $R_0$ will be found through the dominant eigenvalue of matrix $FV^{-1}$. Note that for system (2) we have two classes that spread the infection which are $E$ and $I$ given as:

$$ \frac{dE}{dt} = \beta SI - (\delta + d)E, $$

$$ \frac{dI}{dt} = \delta E - (y + d)I. $$

From (19), $F_1 = \beta SI$, $F_2 = 0$, $V_1 = (\delta + d)E$, $V_2 = (y + d)I - \delta E$, where

$$ F = \begin{bmatrix} \frac{\partial F_1}{\partial E} & \frac{\partial F_1}{\partial I} \\ \frac{\partial F_2}{\partial E} & \frac{\partial F_2}{\partial I} \end{bmatrix} = \begin{bmatrix} \frac{\partial (\beta SI)}{\partial E} & \frac{\partial (\beta SI)}{\partial I} \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & \beta S^* \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} \beta \left( \frac{AN}{d} \right) \\ 0 \end{bmatrix}, $$

and

$$ V = \begin{bmatrix} \frac{\partial (V_1)}{\partial E} & \frac{\partial (V_1)}{\partial I} \\ \frac{\partial (V_2)}{\partial E} & \frac{\partial (V_2)}{\partial I} \end{bmatrix} = \begin{bmatrix} \frac{\partial (\delta + d)E}{\partial E} & \frac{\partial (\delta + d)E}{\partial I} \\ \frac{\partial (y + d)I - \delta E}{\partial E} & \frac{\partial (y + d)I - \delta E}{\partial I} \end{bmatrix} \Rightarrow V = \begin{bmatrix} \delta + d & 0 \\ -\delta & y + d \end{bmatrix}. $$

Now inverse of matrix $V$ is given as:

$$ V^{-1} = \begin{bmatrix} 1 & 0 \\ \frac{\delta + d}{\delta} & \frac{1}{y + d} \end{bmatrix}, $$

so

$$ FV^{-1} = \begin{bmatrix} \frac{\beta AN \delta}{(\delta + d)(y + d)d} & \frac{\beta AN}{y + d}(d) \\ 0 & \frac{1}{0} \end{bmatrix}. $$

We find basic reproductive number $R_0$ by dominant eigenvalue or spectral radius of $(FV^{-1})$ that is $\rho(FV^{-1})$ which is defined as:

$$ \rho(FV^{-1}) = \begin{bmatrix} \frac{\beta AN \delta}{(\delta + d)(y + d)d} - \lambda & \frac{\beta AN}{y + d}(d) \\ 0 & -\lambda \end{bmatrix}, $$

this implies that either $\lambda = 0$ or $\lambda = \frac{\beta AN \delta}{(\delta + d)(y + d)d}$. Hence, the $R_0$ is given for system (2) as:

$$ R_0 = \frac{\beta AN \delta}{(\delta + d)(y + d)d}. $$
4 Stability of equilibrium points

In this section, we analyze the local stability of disease-free equilibrium point and endemic equilibrium point. The Jacobian matrix is calculated from equilibrium points as:

\[
\begin{bmatrix}
\frac{dS}{d\tau} & \frac{dS}{d\tau} & \frac{dS}{d\tau} & \frac{dS}{d\tau} \\
\frac{dE}{d\tau} & \frac{dE}{d\tau} & \frac{dE}{d\tau} & \frac{dE}{d\tau} \\
\frac{dI}{d\tau} & \frac{dI}{d\tau} & \frac{dI}{d\tau} & \frac{dI}{d\tau} \\
\frac{dR}{d\tau} & \frac{dR}{d\tau} & \frac{dR}{d\tau} & \frac{dR}{d\tau}
\end{bmatrix}
= \begin{bmatrix}
AN - \beta SI - dS \\
\beta SI - (\delta + d)E \\
\delta E - (\gamma + d)I \\
\gamma - dR
\end{bmatrix},
\]

...and evaluated equilibrium points to decide on the local stability which is directly determined from the eigenvalues \(\lambda\) as:

\[|J(X_0 \text{ or } X^*) - \lambda| = 0.\]  \hspace{1cm} (21)

The eigenvalues of equation (21) will decide that the SEIR model of fractional differential equation is either stable (i.e., all the eigenvalues contain negative real part) or unstable (i.e., at least one of the eigenvalues contains positive real part).

4.1 Disease-free equilibrium points

Using disease-free equilibrium points, i.e., \(X_0\left(\frac{AN}{d}, 0, 0, 0\right)\), the Jacobian matrix \(J(X_0)\) for system (2) is given as:

\[
J(X_0) = \begin{bmatrix}
-d & 0 & -\beta \left(\frac{AN}{d}\right) & 0 \\
0 & -(\delta + d) & \beta \left(\frac{AN}{d}\right) & 0 \\
0 & \delta & -\gamma & 0 \\
0 & 0 & \gamma & -d
\end{bmatrix},
\]

with eigenvalue \(\lambda\)

\[|J(X_0) - \lambda| = 0,\]  \hspace{1cm} (22)
The eigenvalue $\lambda$ is found from the following polynomial:

$$(-d - \lambda)(\lambda^3 + k_1\lambda^2 + k_2\lambda + k_3) = 0,$$

where $k_3 = y + 2d$, $k_2 = yd + d^2 + \frac{\delta \lambda \beta N}{d}$, and $k_3 = \delta + d - \beta AN d$.

From equation (23), one eigenvalue is $\lambda_1 = -d$ and the remaining three eigenvalues $\lambda_2$, $\lambda_3$, and $\lambda_4$ are obtained from equation $\lambda^3 + k_1\lambda^2 + k_2\lambda + k_3 = 0$. Routh-Hurwitz criterion is used for stability to check that eigenvalues have negative real parts, i.e., $k_1 > 0$, $k_2 > 0$, and $k_2 k_3 > k_3$.

Clearly, $k_1 = y + 2d > 0$, $k_3 = \delta + d - \beta AN d > 0$, and $k_2 k_3 - k_3 = \left[yd + d^2 + \frac{\delta \lambda \beta N}{d}\right] - \delta + d - \beta AN d > 0$.

Therefore by Routh-Hurwitz criterion the disease-free equilibrium point $X_0(\frac{AN}{d}, 0, 0, 0)$ is locally stable.

**4.2 Endemic equilibrium points**

Using endemic equilibrium points i.e., $X^* = (S^*, E^*, I^*, R^*)$ the Jacobian matrix $J(X_0)$ for system (2) is given as:

$$J(X^*) = \begin{bmatrix}
-d - \lambda & 0 & -\beta \left(\frac{AN}{d}\right) & 0 \\
0 & -(\delta + d) - \lambda & \beta \left(\frac{AN}{d}\right) & 0 \\
0 & \delta & -(y + d) - \lambda & 0 \\
0 & 0 & \gamma & -d - \lambda
\end{bmatrix}.$$  

with eigenvalues $\lambda$

$$J(X^*) - \lambda I = 0.$$

$$J(X^* - \lambda) = \begin{bmatrix}
-d - \lambda & 0 & -\beta \left(\frac{AN}{d}\right) & 0 \\
0 & -(\delta + d) - \lambda & \beta \left(\frac{AN}{d}\right) & 0 \\
0 & \delta & -(y + d) - \lambda & 0 \\
0 & 0 & \gamma & -d - \lambda
\end{bmatrix}.$$  

the eigenvalues $\lambda$ are found from the following polynomial:

$$(d + \lambda)^2(\lambda^2 + k_1\lambda + k_2) = 0,$$

so either $(d + \lambda)^2 = 0 \Rightarrow \lambda = -d$ or $\lambda^2 + k_1\lambda + k_2 = 0$, where $k_1 = y + \delta + 2d$ and $k_2 = \delta y + \delta d + d^2 + d(\delta + d)(yd) - y AN \beta + \frac{1}{d(\delta + d)(yd)}$. From equation (25) one eigenvalue is $\lambda_1 = -d$ and the other two eigenvalues $\lambda_2$ and $\lambda_3$ are obtained from equation $\lambda^2 + k_1\lambda + k_2 = 0$. Routh-Hurwitz criterion is used for stability to check that eigenvalues have negative real parts, i.e., $k_1 > 0$ and $k_2 > 0$. Clearly, $k_1 = y + \delta + 2d > 0$ and $k_2 = \delta y + \delta d + d^2 + d(\delta + d)(yd) - y AN \beta + \frac{1}{d(\delta + d)(yd)} > 0$, so that endemic equilibrium point $X^* = (S^*, E^*, I^*, R^*)$ is locally stable for system (2).
5 Numerical method and simulation

For the numerical solutions of a system of FDE, we use fde12.m code [12].

For numerical simulations, we observe from malaria data taken from different hospitals of Khyber Pakhtunkhwa that 80% of the individuals are susceptible, 5% of the individuals are exposed, and 15% of the individuals are infected, i.e., $S(0) = 0.80$, $E(0) = 0.05$, $I(0) = 0.15$. In the numerical simulations of the model (2), we assume that the natural birth $A$ and death rates $d$ are equal to 0.0086. The infectious rate $\beta = 0.3$, the incubation rate $\delta = 0.1764$, and the recovery rate $\gamma = 0.1$ [10].

Now in Figure 2, we take fractional order $\alpha = 0.80$ and observe that susceptible $S$ is equal to 35.63%, exposed $E$ is 3.31%, infected $I$ is 5.65% and recovered $R$ is equal to 55.54%. Next we take fractional order $\alpha = 0.90$ in Figure 3 and observe that susceptible $S$ is equal to 37.74%, exposed $E$ is 3.04%, infected $I$ is 5.01% and recovered $R$ is equal 54.76%. Now we take fractional order $\alpha = 0.98$ in Figure 4 and observe that susceptible $S$ is equal to 38.37%, exposed $E$ is 2.84%, infected $I$ is 4.56%, and recovered $R$ is equal to 54.18%.

![Figure 2: A solution to the SEIR model with fractional order $\alpha = 0.80$.](image1)

![Figure 3: A solution to the SEIR model with fractional order $\alpha = 0.90$.](image2)
Similarly in Figure 5, we take fractional order $\alpha = 1$ and observe that susceptible $S$ is equal to 38.81%, exposed $E$ is 2.84%, infected $I$ is 4.59%, and recovered $R$ is equal to 53.72%.

We compare all the above results of fractional order and notice that for fractional order $\alpha = 0.98$ infected population decreases from 15% to 4.56% and recovered population reaches from 0% to 54.18%. So that the fractional order $\alpha = 0.98$ gives good results than integer order. We also note from memory effect property that fractional order model behaves to particular point over a longer period of time.

6 Conclusion

We have formulated and analyzed the SEIR model for fractional order for malaria disease in KP. We observe that disease-free equilibrium points $X_0$ exist and are locally stable when $R_0 < 1$ and for $R_0 \geq 1$, and the endemic equilibrium points $X^*$ exist and are locally stable. We find basic reproductive number $R_0$ from the
SEIR model and its value by using malaria data of KP (Pakistan) for the year 2018. From the data the value of $R_0 = 2.6340$, which is greater than one which means that malaria is endemic in KP. We also give a numerical example and observe that for fractional order $\alpha = 0.98$, the SEIR model of fractional differential equation gives good result than integer order. In the future, I will work on the fractional order SEIR model for epidemiological disease transmission with control strategies.

**Conflict of interest:** Authors state no conflict of interest.

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