Numerical Study of a Seivr Epidemic Model among Infants with Vaccination and Temporary Immune Protection

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Abstract

In this work, the numerical study of a deterministic mathematical model of an SEIVR (Susceptible-Exposed-Infected-Vaccinated-Recovered) epidemic model among infants was carried out. This model incorporates a temporary immune recovery class which involves subsequent dose vaccination for the infants. Hypothetical values were chosen for the parameters to test the validity of the mathematical model. The parameter with the greatest impact on the model was computed using the eigenvalue elasticity and sensitivity analyses and it was found that the parameter of the rate at which the vaccine wanes in the infants (ω) has the greatest impact on the mathematical model.

Keywords: SEIVR epidemic model; Eigen value elasticity analysis; Eigen value sensitivity analysis; Vaccination; Temporary immune recovered class

Introduction

Many infectious diseases among infants do transmit either through horizontal or vertical modes [1]. These infectious diseases among the infants include polio, measles, diphtheria, influenza, cholera and so on. Many authors have studied a variety of infectious diseases that do transmit both vertical and horizontal modes and gave a comprehensive survey of the formulation and the mathematical analysis of compartmental models that also incorporates vertical transmission [2].

At present, vaccination is a commonly used method for controlling the spread of infectious diseases among the infants between the ages of one month and one year [3]. But, in fact, for many infectious diseases, the immunity which is acquired either by preventive vaccine or by natural immunity may wane with time which may lead to subsequent dose vaccination strategy and at determining changes in qualitative behavior that could result from such a control measure [4].

Dan Long and Zhongyi established and analyzed a deterministic mathematical model in their paper [2]. However, the integration of the recovered infant class to the population was not incorporated into the mathematical model they established. In this paper, it is intended to determine the parameter that has the greatest impact on the formulated mathematical model which incorporates the temporary immune recovered infant class into the model. With this mathematical model, it will enable the public health workers to take into consideration on how to reduce the spread of the infectious diseases among the infants by encouraging and enlightening the parents to allow their children to be given the vaccine regularly and subsequently.

Model description

The SEIVR mathematical model is partitioned into compartments of susceptible infants class (S), the exposed infant’s class (E), the infectious infants class (I), the vaccinated infants class (V) and the recovered infants class (R). The immunized compartment changes due to the coming in of the immunized infants into the population where we assume that a proportion of λ of the incoming infants are immunized against the infectious diseases. This compartment reduces due to the expiration of duration of vaccine efficacy at the rate ω and also by natural death at the rate of µ. The susceptible population increases due to the coming in of the infants from the immunized compartment as a result of the expiration of the duration of vaccine efficacy at the rate ω. The susceptible population also reduces due to the natural death rate µ and infection with contact rate of infection β. The population dynamics of the exposed infants class at the latent period grows with the incidence rate of βSI(1+αI). This class reduces by natural death rate µ and occasional breakdown of the exposed infants at the latent period into infectious class at the rate of µ. Also, the population dynamics of the infectious class grows with the past information of the infected given by a while this class reduces by the natural death rate µ and successful cure of the infectious diseases at the rate of τ by given the infected immunization. The vaccinated class denoted by V but because the efficacy of the vaccine is not one hundred percent such that it can wane at the rate of ω but the recovery class increases with temporary immunity at the rate of δ which are transferred back to the susceptible class and decreases by natural death rate µ. The schematic description of the model is shown in the Figure 1.

The mathematical model

The SEIVR model is expressed as the system of nonlinear initial value problem given in the form

\[ \frac{dS}{dt} = -\beta SI(t)(1+\alpha I(t)) - (\mu + \eta)S(t), S(0) = S_0 \]  
\[ \frac{dE}{dt} = \beta SI(t)(1+\alpha I(t)) + \eta S(t) - (\mu + \sigma)E(t), E(0) = E_0 \]  

Keywords: SEIVR epidemic model; Eigen value elasticity analysis; Eigen value sensitivity analysis; Vaccination; Temporary immune recovered class

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Using the eigenvalue elasticity and sensitivity analyses, the parameter with the greatest impact on the mathematical model [2,5-8] can be determined. With the hypothetical values of the parameters given in the Table 1, the parameter that has the greatest impact can be obtained. The eigenvalue sensitivity analysis with respect to a parameter is defined as the partial derivative of the eigenvalue with respect to that parameter. The eigenvalue sensitivity $S_i$ (i = 1, ..., N and N is the dimension of the state vector) with respect to the jth parameter of the system $p_j$ is given in the form [8,9]

$$S_i(p_j) = \Delta \lambda_i \frac{\partial \lambda_i}{\partial p_j} = \frac{\partial \lambda_i}{\partial p_j} = \frac{d I_i^J}{d p_j} \tau_j$$

and the eigenvalue elasticity with respect to a parameter is defined as the partial derivative of the eigenvalue with respect to that parameter normalized for the size of the parameter and the size of the eigenvalue. This could also be described as the product of the eigenvalue sensitivity and the ratio of the parameter to the eigenvalue. Thus, it is given by the expression [5,10]

$$E_i(p_j) = \Lambda \lim_{p_j \to 0} \frac{\Delta \lambda_i}{\Delta p_j} = (\frac{\partial \lambda_i}{\partial p_j} \tau_j)^{-1} \frac{\partial \lambda_i}{\partial p_j} = \frac{d J_i^J}{d p_j} \tau_j$$

With equations (6) and (7), the eigenvalue elasticity and sensitivity analyses with respect to the parameters in the mathematical model can be computed using the left eigenvectors ($I_i$) and the right eigenvectors ($J_i$) with the partial derivatives of the linearized Jacobian matrix ($J$) of the mathematical model (1)-(5) with respect to the parameters ($p_j$) (j=1, 2, ..., 11) [2]. Because the Jacobian matrix ($J$) and the $\frac{d J_i^J}{d p_j}$ of the model (1)-(5) can often be easily determined symbolically and the eigenvalues can be computed for particular parameter values using the hypothetical values in Table 1 above, both eigenvalue elasticity and sensitivity with respect to a parameter can be computed at the endemic equilibrium state.

At the endemic equilibrium state (in the presence of the infectious diseases), the Jacobian matrix $J$ of the mathematical model (1)- (5) is given in the form

$$J = \begin{pmatrix}
-\beta I (s+I) (s+I) & 0 & 0 & 0 & 0
\beta I I + I & -\beta I - \mu I & 0 & 0 & 0
0 & -\mu I & -\omega & 0 & 0
0 & 0 & -\mu I & 0 & 0
0 & 0 & 0 & -\mu I & 0
\end{pmatrix}
$$

while

$$I^* = \frac{(\mu + \sigma)(\rho \mu + \mu + \tau) I}{\sigma}
$$

$$V^* = \frac{P A + \tau I^*}{\mu + \sigma}
$$

and

$$R^* = \frac{\omega (P A + \tau I^*)}{(\mu + \delta)(\mu + \omega)}
$$

Table 1: The interpretation of the parameters with their hypothetical values.

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Definitions</th>
<th>Values</th>
</tr>
</thead>
<tbody>
<tr>
<td>$\Lambda$</td>
<td>recruitment rate of the infants into the susceptible class</td>
<td>100</td>
</tr>
<tr>
<td>$A$</td>
<td>number of infants with the infectious diseases</td>
<td>1500</td>
</tr>
<tr>
<td>$P$</td>
<td>fraction of the recruited infants who are vaccinated</td>
<td>0.8</td>
</tr>
<tr>
<td>$\mu$</td>
<td>the mortality rate</td>
<td>0.2</td>
</tr>
<tr>
<td>$\tau$</td>
<td>rate at which infected infants are treated with vaccines</td>
<td>0.35</td>
</tr>
<tr>
<td>$\beta$</td>
<td>transmission coefficient</td>
<td>0.5</td>
</tr>
<tr>
<td>$\omega$</td>
<td>rate at which the vaccine wanes</td>
<td>0.0182</td>
</tr>
<tr>
<td>$\sigma$</td>
<td>rate at which the exposed infants become infectious</td>
<td>0.01</td>
</tr>
<tr>
<td>$\alpha$</td>
<td>rate of re-infection</td>
<td>0.05</td>
</tr>
<tr>
<td>$\eta$</td>
<td>rate at which the susceptible infants are exposed to the diseases</td>
<td>365/3</td>
</tr>
<tr>
<td>$S_0$</td>
<td>initial value of the susceptible infants</td>
<td>0.3982</td>
</tr>
<tr>
<td>$E_0$</td>
<td>initial value of the infected infants</td>
<td>0.3086</td>
</tr>
<tr>
<td>$I_0$</td>
<td>initial value of the vaccinated infants</td>
<td>0.2131</td>
</tr>
<tr>
<td>$V_0$</td>
<td>initial value of the temporary recovered infants</td>
<td>0.0502</td>
</tr>
<tr>
<td>$\tau_0$</td>
<td>initial value of the exposed infants</td>
<td>0.0299</td>
</tr>
</tbody>
</table>

In which $S(t)$, $E(t)$, $I(t)$, $V(t)$ and $R(t)$ represent the population of susceptible infant class, exposed infant class but not yet infected, the infected infant class, the vaccinated infant class and the temporary immune recovered infant class, respectively. The parameters in the mathematical model are positive and the Table 1 above provides the definitions for the model parameters. The model assumes a varying population of $N(t)$ so that

$$N(t) = S(t) + E(t) + I(t) + V(t) + R(t)$$

and it is given in the form

$$N(t) = K + Ce$$

for

$$K = \frac{A + \Lambda}{\mu}$$

Parameter with the greatest impact

The interpretation of the parameters with their hypothetical values.

$$\frac{d S}{dt} = \sigma E(t) - \rho \mu I(t) - (\mu + \tau) I(t) - (\mu + \alpha) E(t)$$

$$\frac{d E}{dt} = \rho \mu I(t) - (\mu + \tau) E(t) - (\mu + \alpha) E(t)$$

$$\frac{d I}{dt} = \rho \mu I(t) - (\mu + \omega) I(t)$$

$$\frac{d R}{dt} = \omega I(t)$$

Figure 1: Diagrammatic representation of an SEIVR model.
Table 2: Table of eigenvalue elasticity and sensitivity analyses of the parameters.

<table>
<thead>
<tr>
<th>Parameters</th>
<th>Eigenvalue Elasticity</th>
<th>Eigenvalue Sensitivity</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \mu )</td>
<td>-1.0000</td>
<td>-0.7462</td>
</tr>
<tr>
<td>( \tau )</td>
<td>1.4097 \times 10^{-3}</td>
<td>2.0455 \times 10^{-7}</td>
</tr>
<tr>
<td>( \beta )</td>
<td>-3.5027 \times 10^{-1}</td>
<td>-7.2606 \times 10^{-9}</td>
</tr>
<tr>
<td>( \omega )</td>
<td>3.1757 \times 10^{-1}</td>
<td>2.3961 \times 10^{-9}</td>
</tr>
<tr>
<td>( \sigma )</td>
<td>-0.0315</td>
<td>-1.3059 \times 10^{-9}</td>
</tr>
<tr>
<td>( \alpha )</td>
<td>-1.4595 \times 10^{-7}</td>
<td>-7.2606 \times 10^{-11}</td>
</tr>
<tr>
<td>( \delta )</td>
<td>1.1537 \times 10^{-9}</td>
<td>2.3914 \times 10^{-10}</td>
</tr>
<tr>
<td>( \rho )</td>
<td>-1.0002</td>
<td>-5.0450</td>
</tr>
<tr>
<td></td>
<td>0.0011</td>
<td>1.4868 \times 10^{-5}</td>
</tr>
</tbody>
</table>

While \( I^* \) is the root of the equation \( a(I^*)^2 + b(I^*) + cI^* + d = 0 \) for

\[
a = \alpha \mu \sigma (\mu + \delta)(\mu + \omega) + \sigma \omega \mu \delta - \alpha \mu (\mu + \delta)(\mu + \omega)(\mu + \sigma)(\mu + \tau),
\]

\[
b = \alpha \sigma \mu \delta (\mu + \delta)(\mu + \omega) + \beta \delta \mu \sigma \omega (\mu + \sigma)(\mu + \tau) - \beta \mu (\mu + \delta)(\mu + \sigma)(\mu + \tau),
\]

\[
c = \alpha \mu (\mu + \delta)(\mu + \omega) + \beta \mu (\mu + \delta)(\mu + \omega) + \sigma \omega \mu \delta (\mu + \sigma)(\mu + \tau) - \sigma \mu (\mu + \delta)(\mu + \sigma)(\mu + \tau)
\]

and

\[d = \sigma \omega \mu \delta (\mu + \delta)(\mu + \omega)\]

Substituting the hypothetical values in the Table 1 into equations (8) to (12), computer program was written for the evaluation of the values of the eigenvalue elasticity and sensitivity of the mathematical model [7]. The results obtained from the programming are shown in the Table 2 above. From the Table 2, it can be seen that the parameter \( \omega \) which is defined as the rate at which the vaccine wanes in the infants has the largest positive values of both the eigenvalue elasticity and sensitivity analyses. This means that the parameter \( \omega \) has the greatest impact on the formulated mathematical model (1)-(5). This is a notification for the public health workers to note by making sure that the temperature at which the vaccines are kept from the country they were imported from, should be maintained before given to the infants so that the vaccines can be effective. With the maintenance of the optimum temperature for refrigerated vaccines which is between \(+2^\circ C\) and \(+8^\circ C\) [9], the prevention of the spread of the infectious diseases among the infants will be reduced or eventually eradicated from the community.

**Conclusion**

The eigenvalue elasticity and sensitivity analyses on the formulated mathematical model (1)-(5) show that the parameter \( \omega \) has the greatest impact on the mathematical model. Therefore, for effective control of the infectious diseases among the infants, subsequent dose vaccination should be given to the infants because of the low or poor maintenance of the optimum temperature of the vaccines [5]. Vaccines are sensitive biological products which may become less effective or even destroyed, when exposed to temperatures outside the recommended range [9]. Therefore, there is a need to ensure that an effective product of the vaccine is being used to reduce the spread of the infectious diseases among the infants. Vaccine failures caused by administration of compromised vaccine may result in the re-emergence or occurrence of vaccine preventable diseases [10].

**References**