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Analysis of a Vaccination Model For Carrier Dependent Population With A Saturated Incidence Rate

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Abstract- This study to proposed and analyzed a vaccination model for carrier dependent population with a saturated incidence rate. The equilibrium's for the model are found and their stability investigated. Gives the numerical examples in support of results

Keywords: carrier dependent, vaccination, environmental discharge, infectious diseases, carrying capacity.

I. INTRODUCTION

Infectious diseases are spread by carriers which are present in the environment [3]. Carriers are individuals who are able to transmit the disease but do not show any symptoms. Infectious diseases are also known as transmissible disease or communicable disease. Infectious disease can be caused by bacteria, viruses, fungi or parasites. Some infectious diseases can be passed from person to person and some are transmitted by bites from insects or animals. Infectious diseases are also transmitted by contaminated food or water. The spread of such diseases is very much dependent on the carrier population, the density of which increases due to environmental factors such as temperature, humidity, rain, vegetation, etc [13, 15]. The per capita growth rate and the modified carrying capacity of carrier population are taken to be functions of human population density and assumed to increase as the human population density increases [10,11]. Many infectious diseases are prevented by vaccination. The modeling and analysis of infectious disease have been done by many researchers [2, 4, 6, 8, 12, 14]. Infectious diseases model with population dependent death rate and logistic population growth studied by Greenhalth [1] and Gao et al. [7] . Zhou and Hethcote [5] introduced the various kinds of demographics for infectious diseases. Ghosh et al. [9] studied the spread of carrier dependent infectious diseases with environmental effects using variable carrier population.

Naresh et al. [16] considered a following vaccination model for carrier dependent infectious diseases with environmental effects:

$$\frac{dX}{dt} = A - \beta XY - \lambda XC + \upsilon Y - \varphi X - dX$$

$$\frac{dY}{dt} = \beta XY + \lambda XC - (\upsilon + \alpha + d)Y + \gamma VY + \upsilon_1 VC$$

$$\frac{dV}{dt} = \varphi X - \gamma VY - \upsilon_1 VC - dV$$

$$\frac{dN}{dt} = A - dN - \alpha Y$$

$$\frac{dC}{dt} = s(N)C - \frac{s_0 C^2}{L(N)} - s_1 C$$
(*)

with

$$X + Y + V = N$$

$$X(0) = X_0 > 0, Y(0) = Y_0 \ge 0, V(0) = V_0 \ge 0$$

$$N(0) = N_0 > 0, C(0) = C_0 \ge 0$$

where N(t) be the total human population at any time t, which is divided into three subclasses: the susceptible X(t), the infectives Y(t), and the vaccinated individuals V(t). C(t) represent the density of carrier population, which is governed by a generalized logistic model. It is further assumed that the susceptible are vaccinated at a constant rate and some of them may again become infected while coming in contact with infectives or with carriers due to inefficacy of vaccines. β and λ are transmission coefficients due to infectives and carrier population respectively. The parameters ϕ, υ and d represent vaccination coverage, therapeutic treatment coverage and natural deaths respectively, α is the disease related death constant, γ and υ_1 denote the transmission coefficient of vaccinated individuals due to interaction with infectives and carrier population respectively. However, the rate with which vaccinated persons become infected is very small as compared to the rate with which susceptible get infected i.e. $\gamma \leq \beta$ and $\upsilon_1 \leq \lambda$. The constant s_1 is the death rate coefficient of carriers due to natural factors as well as by control measures. Here, s(N) denotes growth rate per capita of the carrier population are balanced, then it may tend to zero. Similarly, L(N) is the modified carrying capacity of the carrier population. It has been pointed out in the introduction, that as the human population increases, the effects of human population related factors enhance the changes of growth of carrier population.

1.1 The model

In this study, consider a saturated incidence rate $\frac{\beta XY}{1+aY}$, where *a* is the psychological effect rate. The model dynamics is governed by following system of nonlinear ordinary differential equations:

$$\frac{dX}{dt} = A - \frac{\beta XY}{1 + aY} - \lambda XC + \nu Y - \varphi X - dX$$

$$\frac{dY}{dt} = \frac{\beta XY}{1 + ay} + \lambda XC - (\nu + \alpha + d)Y + \frac{\gamma VY}{1 + aY} + \nu_1 VC$$

$$\frac{dV}{dt} = \varphi X - \frac{\gamma VY}{1 + aY} - \nu_1 VC - dV$$

$$\frac{dN}{dt} = A - dN - \alpha Y$$

$$\frac{dC}{dt} = s(N)C - \frac{s_0 C^2}{L(N)} - s_1 C$$

$$X + Y + V = N$$

$$X(0) = X_0 > 0, Y(0) = Y_0 \ge 0, V(0) = V_0 \ge 0$$
(1)

$$X(0) = X_0 > 0, T(0) = T_0 \ge 0, V(0) = V_0 \ge 0$$
$$N(0) = N_0 > 0, C(0) = C_0 \ge 0$$

where *a* is the psychological effect rate and rest of the parameters are describes in above model (*). Thus, in the model (1), s(N) and L(N) are taken to be functions of total human population instead of infective population. Assume that the growth rate per capita increase as the human population density increase, we have

(2)

(3)

$$s(0) = s_0$$
 and $s'(N) \ge 0$,

where s_0 is the value of s(N) at N = 0 and ()' denotes the derivatives of the function with respect to its argument. We also assume that the modified carrying capacity increase with human population density, so that

$$L(0) = L_0 > 0$$
 and $L'(N) \ge 0$,

where L_0 is the value of L(N) when N = 0 From equation (1), (2) and (3), we see that even if human population

related factors are absent, carrier population density increase in its natural environment and it tends to $L_0\left(1-\frac{s_1}{s_0}\right)$ which may

become zero if $s_1 \rightarrow s_0$. In the model, all the dependent variables and parameters are assumed to be non negative.

II. EQUILIBRIUM ANALYSIS

$$\frac{dY}{dt} = \frac{\beta Y(N-Y-V)}{1+aY} + \lambda C(N-Y-V) - (\upsilon + \alpha + d)Y + \frac{\gamma VY}{1+aY} + \upsilon_1 VC,$$

$$\frac{dV}{dt} = \varphi(N-Y-V) - \frac{\gamma VY}{1+aY} - \upsilon_1 VC - dV,$$

$$\frac{dN}{dt} = A - dN - \alpha Y,$$

$$\frac{dC}{dt} = s(N)C - \frac{s_0 C^2}{L(N)} - s_1 C$$
(4)

There exist following three non negative equilibria of the system (4)

1. Disease free equilibrium $E_0 = \left(0, \frac{\varphi A}{d(d+\varphi)}, \frac{A}{d}, 0\right)$ exists, without any condition. The existence of E_0 is obvious.

2. Carrier free equilibrium $E_1 = (\bar{Y}, \bar{V}, \bar{N}, 0)$.

This equilibrium may be obtain by solving the following algebraic equations,

$$\beta(N-Y) - (\upsilon + \alpha + d)(1 + aY) - (\beta - \gamma)V = 0$$
(5)

$$Y = \frac{A - dN}{\alpha} \tag{6}$$

$$V = \frac{\varphi \left[1 + a \left(\frac{A - dN}{\alpha} \right) \right] \left[(\alpha + d)N - A \right]}{\alpha (\varphi + d) \left[1 + a \left(\frac{A - dN}{\alpha} \right) \right] + \gamma (A - dN)}$$
(7)

Using equation (6) and (7) in equation (5), we get an algebraic equation in single variable N. i.e. F(N) = 0 where F(N) = 0 is given by the following equation.

$$F(N) = [\beta N - (\nu + \alpha + d)(1 + aY)] - (\beta - \gamma)V - \beta Y = 0$$
(8)

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From equation (6) and (7), we note that \overline{Y} and \overline{V} will be positive only when F(N) = 0 has a root in the interval $\left(\frac{A}{\alpha+d}, \frac{A}{d}\right)$. From equation (8) it is easy to observe that, $F\left(\frac{A}{\alpha+d}\right) < 0$ and $F\left(\frac{A}{d}\right) > 0$, if $\frac{A(\beta d + \gamma \phi)}{d(\phi+d)(\nu+\alpha+d)} > 1$, thus there exists a root \overline{N} of F(N) = 0 in $\frac{A}{\alpha+d} < N < \frac{A}{d}$. Also F'(N) > 0 in $\frac{A}{\alpha+d} < N < \frac{A}{d}$. hence there exists a unique positive root \overline{N} given by F(N) = 0. Knowing the value of \overline{N} , we can compute the values of \overline{Y} and \overline{V} from equations (6) and (7), respectively.

Thus there exists a unique carrier free equilibrium $E_1(Y,V,N,0)$, provided the condition $R(\varphi) = \frac{A(\beta d + \gamma \varphi)}{d(\phi + d)(v + \alpha + d)} > 1$ is satisfied.

2.1. Endemic equilibrium

The endemic equilibrium, $E_2 = (Y^*, V^*, N^*, C^*)$

The endemic equilibrium is given by the solution of the following set of algebraic equations,

$$(\beta Y + \lambda c(1 + aY)(N - Y - V) + (\gamma Y + \upsilon_1 C(1 + aY)V) - (\upsilon + \alpha + d)Y(1 + aY) = 0$$
(9)

$$Y = \frac{A - dN}{\alpha} \tag{10}$$

$$V = \frac{\varphi((\alpha + d)N - A)(1 + aY)}{\alpha[(\varphi + d + \nu_1 C)(1 + aY)] + \gamma(A - dN)}$$
(11)

$$C = \frac{L(N)[s(N) - s_1]}{s_0}$$
(12)

We may reduce equation (9) in a single variable N i.e. F(N) by using equations (10),(11) and (12), where

$$F(N) = (\beta Y + \lambda c(1 + aY)(N - Y - V) + (\gamma Y + \upsilon_1 C(1 + aY)V) - (\upsilon + \alpha + d)Y(1 + aY)$$
(13)

It is clear from (13) that $F(\frac{A}{\alpha+d}) < 0$ and $F\left(\frac{A}{d}\right) > 0$. this implies that there exists a root N of F(N)=0 in

 $\frac{A}{\alpha+d} < N < \frac{A}{d}$ Also F'(N) > 0, provided $[\alpha v_1 NC' - \gamma A] > 0$. Hence there exist a unique positive root N^* given by F(N)=0 in $\frac{A}{\alpha+d} < N < \frac{A}{d}$. Knowing the value of N^* , the values of Y^*, V^* and C^* can be computed from equations (10),(11) and (12), respectively. Thus, the equilibrium E_2 exists provided $s(N^*) - s_1 > 0$ and F'(N) > 0. From equation (13) it is easy to note that $F(\bar{N}) > 0$ keeping in mind the above analysis of equation (13), we have $\bar{N} > N^*$. From equation (12), it may be noted that $dC^* / dN^* > 0$ in view of equation (2) and (3). Thus, the equilibrium value of carrier population

density increases with increase in the equilibrium value of human population.

III. STABILITY ANALYSIS

Now we analyze the stability of equilibria E_o, E_1 and E_2 . The local stability results of these equilibria are stated in the following theorem:

Theorem 3.1. The equilibrium E_0 is unstable whenever E_1 or E_2 exists, E_1 is unstable whenever E_2 exists and the equilibrium E_0 and E_0 are locally asymptotically stable provided the following conditions are satisfied.

$$\alpha \left[\frac{(\beta - \gamma)Y^{*}}{(1 + aY^{*})} + (\lambda - \upsilon_{1})C^{*} \right]^{2} < \frac{1}{4} b_{1} b_{2}^{2} b_{3} \min \left\{ \frac{b_{1}}{4b_{3} \left(\varphi + \frac{\gamma V^{*}}{(1 + Y^{*})} \right)^{2}}, \frac{d}{3(\varphi)}, \frac{k_{3} s_{0} C^{*}}{3 \upsilon_{1}^{2} V^{*2} L(N^{*})} \right\}$$
(14)

$$3\alpha L^{2} (N^{*}) \left[\lambda (N^{*} - Y^{*} - V^{*}) + \upsilon_{1} V^{*} \right]^{2} \left[s'(N^{*}) + \frac{s_{0} C^{*} L'(N^{*})}{L^{2} (N^{*})} \right]^{2} < \frac{8}{27} b_{1} b_{3} ds_{0}^{2}$$
(15)

$$b_{1} = \frac{\lambda N^{*} C^{*}}{Y^{*}} - \frac{(\lambda - \upsilon_{1}) V^{*} C^{*}}{Y^{*}} + \frac{\beta}{(1 + aY^{*})^{2}}$$

$$b_{2} = \varphi + d + \frac{\gamma Y^{*}}{(1 + aY^{*})} + \upsilon_{1} C^{*}$$

$$b_{3} = \frac{\beta Y^{*}}{(1 + aY^{*})} + \lambda C^{*}$$

Proof: the general variational matrix M for the system is given as follows:

$$M = \begin{pmatrix} m_{11} & m_{12} & m_{13} & m_{14} \\ -\phi + \frac{(1+aY)\gamma V + \gamma V Y a}{(1+aY)^2} & -\phi - \frac{\gamma Y}{(1+aY)} - \upsilon_1 C - d & \phi & -\upsilon_1 V \\ -\alpha & 0 & -d & 0 \\ 0 & 0 & s'(N)C + \frac{s_0 C^2}{L^2(N)} L'(N) & s(N) - s_1 - \frac{2s_0 C}{L(N)} \end{pmatrix}$$

where

$$\begin{split} m_{11} &= (1+aY)(\beta N - 2\beta Y - \beta V + \gamma V) - (\beta YN - \beta Y^2 - \beta VY + \gamma VY)a - \lambda C - (\upsilon + \alpha + d) \\ m_{12} &= -\frac{\beta Y}{1+aY} - \lambda C + \frac{\gamma Y}{1+aY} + \upsilon_1 C \\ m_{13} &= \frac{\beta Y}{1+aY} + \lambda C \\ m_{14} &= \lambda (N - Y - V) + \upsilon_1 V \end{split}$$

At E_0

$$M_{0} = \begin{pmatrix} \frac{A}{d(\phi+d)} (\beta d + \gamma \phi) - (\upsilon + \alpha + d) & 0 & 0 & \frac{\lambda A d + \upsilon_{1} \phi A}{d(\phi+d)} \\ -\phi + \frac{\gamma \phi A}{d(\phi+d)} & -(\phi+d) & \phi & \frac{-\upsilon_{1} \phi A}{d(\phi+d)} \\ -\alpha & 0 & -d & 0 \\ 0 & 0 & 0 & s(\frac{A}{d}) - s_{1} \end{pmatrix}$$
$$\psi_{1} = s \left(\frac{A}{d}\right) - s_{1}$$
$$\psi_{2} = \frac{A}{d(\phi+d)} (\beta d + \gamma \phi) - (\upsilon + \alpha + d)$$
$$\psi_{3} = -d$$
$$\psi_{4} = -(\phi+d)$$

Since all the model parameters are assumed to be non negative, it follows that $\psi_3.\psi_4 < 0$. Thus, the stability of E_0 will depend on the sign of ψ_1 and ψ_2 .

We define a threshold parameter $R(\varphi) = \frac{A(\beta d + \gamma \varphi)}{d(\varphi + d)(\upsilon + \alpha + d)}$

The disease free equilibrium is locally asymptotically stable if $R(\phi) < 1$ and $s\left(\frac{A}{d}\right) - s_1 < 0$. Since s(N) is increasing function of N, so $\psi_1 < 0$, this always implies that $s(N) - s_1 < 0$. biologically $s\left(\frac{A}{d}\right) - s_1 < 0$ implies the absence of carrier population. Thus DFE may be stable only in the absence of carrier population because otherwise disease still persists even if there is no direct interaction of susceptible with infectives. Thus, E_0 is unstable if either $R(\phi) > 1$ or $s\left(\frac{A}{d}\right) - s_1 > 0$, keeping in mind that the necessary condition for the existence of E_1 is $R(\phi) > 1$ and for the existence of E_2 is $s(N^*) - s_1 > 0$. Thus, E_0 is unstable whenever E_1 or E_2 exists, similarly one of the eigen values of the variational matrix is $s(\overline{N}) - s_1$. Now keeping in mind that $\overline{N} > N^*$, we have that E_1 is unstable whenever E_2 is exists.

From equation (5), (6) and (7), we get ϕ as a function of Y alone

$$\phi(Y) = \frac{\left[\beta Y - \frac{\beta(A - \alpha Y)}{d} + (\nu + \alpha + d)(1 + aY)\right] [d(1 + aY) + \gamma Y]}{(1 + aY) \left[-(\nu + \alpha + d)(1 + aY) + \frac{\gamma(A - \alpha Y)}{d} - \gamma Y\right]}$$
(16)

We note that,

$$\phi(0) = d \left[\frac{\frac{\beta A}{d(\nu + \alpha + d)} - 1}{1 - \frac{\gamma A}{d}(\nu + \alpha + d)} \right]$$
$$= d \left[\frac{R_0 - 1}{1 - R_1} \right]$$

The expression for $\phi(0)$ is the threshold vaccination rate given by the term ϕc_1 . We have already shown the uniqueness of \overline{Y} for $R(\phi) > 1$ in section 3, Now here we show the bifurcation analysis for the disease free equilibrium E_0 when $R(\phi) = 1$. We note that for $R(\phi) = 1$, one eigenvalue of M_0 is 0, whereas other eigen values are negative provided $s(A/d) - s_1 < 0$.

Local Stability of endemic equilibrium

$$U_{1} = \frac{1}{2} (k_{0}y^{2} + k_{1}v^{2} + k_{2}n^{2} + k_{3}c^{2})$$

$$Y = Y^{*} + y$$

$$V = V^{*} + v$$

$$N = N^{*} + n$$

$$C = C^{*} + c$$

$$\frac{dU_{1}}{dt} = -k_{0} \left[\frac{\beta}{(1 + aY^{*})^{2}} + \frac{\lambda N^{*}C^{*}}{Y^{*}} + \frac{(\lambda - v_{1})V^{*}C^{*}}{Y^{*}} \right] y^{2} - k_{1} \left(\varphi + d + \frac{\gamma Y^{*}}{1 + aY^{*}} + v_{1}C^{*} \right) v^{2}$$

$$-k_{2}dn^{2} - \frac{k_{3}s_{0}C^{*}}{L(N^{*})}c^{2} - \left[k_{0} \left\{ \frac{(\beta - \gamma)Y^{*}}{1 + aY^{*}} + (\lambda - v_{1})C^{*} \right\} + k_{1} \left(\varphi + \frac{\gamma V^{*}}{(1 + aY^{*})^{2}} \right) \right] vy$$

$$+k_{1}\varphi vn - k_{1}v_{1}V^{*}vc + k_{0} \left[\left(\frac{\beta Y^{*}}{1 + aY^{*}} + \lambda C^{*} \right) - \alpha k_{2} \right] ny + k_{0}[\lambda(N^{*} - Y^{*} - V^{*}) + v_{1}V^{*}]yc$$

$$+k_{3} \left[s'(N^{*})C^{*} + \frac{s_{0}C^{*2}}{L^{2}(N^{*})}L'(N^{*}) \right] nc$$

$$(17)$$

dt

 $\underline{dU_1}$ will be negative under the following conditions

$$k_{0} \left[(\beta - \gamma) \frac{Y^{*}}{1 + aY^{*}} + (\lambda - \upsilon_{1})C^{*} \right]^{2} < k_{1} \left(\varphi + d + \frac{\gamma Y^{*}}{1 + aY^{*}} + \upsilon_{1}C^{*} \right) \left[\frac{\lambda N^{*}C^{*}}{Y^{*}} - \frac{(\lambda - \upsilon_{1})V^{*}C^{*}}{Y^{*}} + \frac{\beta}{(1 + aY^{*})^{2}} \right]$$
$$k_{1} \left(\varphi + \frac{\gamma V^{*}}{(1 + aY^{*})^{2}} \right)^{2} < k_{0} \left(\varphi + d + \frac{\gamma Y^{*}}{1 + aY^{*}} + \upsilon_{1}C^{*} \right) \left[\frac{\beta}{(1 + aY^{*})^{2}} + \frac{\lambda N^{*}C^{*}}{Y^{*}} - \frac{(\lambda - \upsilon_{1})V^{*}C^{*}}{Y^{*}} \right]$$

$$\begin{split} \left[k_0 \left(\frac{\beta Y^*}{1 + a Y^*} + \lambda C^* \right) - \alpha k_2 \right]^2 &< \frac{2k_0 k_2 d}{3} \left[\frac{\lambda N^* C^*}{Y^*} - \frac{(\lambda - \upsilon_1) V^* C^*}{Y^*} + \frac{\beta}{(1 + a Y^*)^2} \right] \\ k_0 \left[\lambda (N^* - Y^* - V^*) + \upsilon_1 V^* \right]^2 &< \frac{2k_3}{3} \frac{s_o C^*}{L(N^*)} \left[\frac{\lambda N^* C^*}{Y^*} - \frac{(\lambda - \upsilon_1) V^* C^*}{Y^*} + \frac{\beta}{(1 + a Y^*)^2} \right] \\ k_1 \varphi^2 &< \frac{k_2}{3} d \left(\varphi + d + \frac{\gamma Y^*}{1 + a Y^*} + \upsilon_1 C^* \right) \\ k_1 v^2 V^{*2} &< \frac{k_3}{3} \frac{s_0 C^*}{L(N^*)} \left(\varphi + d + \frac{\gamma Y^*}{1 + a Y^*} + \upsilon_1 C^* \right) \\ k_3 \left[s'(N) C^* + \frac{s_0 C^{*2}}{L^2(N^*)} L'(N^*) \right]^2 &< \frac{4k_2 ds_0 C^*}{9L(N^*)} \end{split}$$

We choose k_1 and k_3 such that

$$\frac{4\alpha}{b_{1}b_{2}b_{3}} \left[\frac{(\beta - \gamma)Y^{*}}{1 + aY^{*}} + (\lambda - \upsilon_{1})C^{*} \right]^{2} < k_{1} < b_{2} \min \left\{ \frac{b_{1}}{4b_{3} \left(\varphi + \frac{\gamma V^{*}}{(1 + aY^{*})^{2}}\right)}, \frac{d}{3\varphi}, \frac{k_{3}s_{0}C^{*}}{3\upsilon_{1}^{2}V^{*2}L(N^{*})} \right\} \right\}$$

$$\frac{3\alpha L(N^{*})}{2s_{0}C^{*}b_{1}b_{3}} \left[\lambda(N^{*} - Y^{*} - V^{*}) + \upsilon_{1}V^{*} \right]^{2} < k_{3} < \frac{4ds_{0}C^{*}}{9L(N^{*})\left[s'(N^{*})C^{*} + \frac{s_{0}C^{*2}}{L^{2}(N^{*})}L'(N^{*})\right]}$$

$$b_{1} = \frac{\lambda N^{*}C^{*}}{Y^{*}} - \frac{((\lambda - \upsilon_{1})V^{*}C^{*}}{Y^{*}} + \frac{\beta}{(1 + aY^{*})^{2}}$$

$$b_{2} = \varphi + d + \frac{\gamma Y^{*}}{1 + aY^{*}} + \upsilon_{1}C^{*}$$

$$b_{3} = \frac{\beta Y^{*}}{1 + aY^{*}} + \lambda C^{*}$$

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Hence $\frac{dU_1}{dt}$ is a negative definite under the conditions as stated in the statement of the theorem, showing that E_2 is locally asymptotically stable.

To study the non linear asymptotic stability of endemic equilibrium E_2 , we require the bounds of dependent variables. For this, we state the following lemma giving the region of attraction, with out proof.

Lemma 3.2. the region of attraction for the system is given by,

$$\Omega = \left\{ (Y, N, V, C) : 0 \le Y \le N \le \frac{A}{d}, 0 \le V \le \frac{\varphi A}{d(\varphi + d)}, 0 \le C \le C_m \right\}$$

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Which attracts all the solutions initiating in the positive orthant, where $C_m = \frac{L(A/d)}{s_0} [s(A/d) - s_1]$

Theorem 3.3. In addition to assumptions (2) and (3), let s(N) and L(N) satisfy $0 \le s'(N) \le p$ and $0 \le L'(N) \le q$ for some positive constant p and q in Ω , then E_2 is nonlinearly asymptotically stable in Ω provided the following inequalities are satisfied:

$$\alpha N^* \left[\frac{\beta - \gamma}{N^*} + \frac{\lambda C_m}{Y^*} \right] < \frac{1}{5} d\beta, \tag{18}$$

$$\alpha N^{*}L^{2}(N^{*})\left[\lambda(N^{*}-Y^{*}-V^{*})+\upsilon_{1}V^{*}\right]^{2}\left[p+\frac{s_{0}qC_{m}}{L_{0}^{2}}\right]^{2} < \frac{4}{45}\beta ds_{0}^{2}Y^{*2},$$

$$N^{*}\left[\frac{\beta-\gamma}{N^{*}}+\frac{(\lambda-\upsilon_{1})C_{m}}{Y^{*}}\right]^{2} < \frac{1}{5}\beta\left(\varphi+d+\frac{\gamma Y^{*}}{N^{*}}\right)^{2} \times$$

$$\min\left\{\frac{\beta}{5\left(\varphi+\frac{\gamma\varphi A}{d(\varphi+d)N^{*}}\right)^{2}}, \frac{d}{4\left(\varphi+\frac{\gamma A}{dN^{*}}\right)^{2}}, \frac{m_{3}s_{0}}{3\nu_{1}^{2}V^{*2}L(N^{*})}\right\}$$

$$(20)$$

it is clear from (19) that in the absence of human related factors, i.e. p = q = 0, the inequality is automatically satisfied. This implies that human population related factors, conductive to the growth of carrier population, have a destabilizing effect on the system. Here we also note that due to presence of a vaccinated class, a condition (20) is required for the nonlinear stability which further destabilizes the system.

Proof.

$$-\left\{\frac{m_{o}\beta}{5(1+aY)N^{*}}(Y-Y^{*})^{2} + \frac{m_{3}s_{0}}{3L(N)}(C-C^{*})^{2} - m_{0}\frac{\lambda(N^{*}-Y^{*}-V^{*}) + \upsilon_{1}V^{*}}{Y^{*}}(Y-Y^{*})(C-C^{*})\right\}$$

$$-\left\{\frac{m_{1}}{4}\left(\varphi + d + \frac{\gamma Y^{*}}{1+aY^{*}}\right)(V-V^{*})^{2} + \frac{m_{2}d}{4}(N-N^{*})^{2} - m_{1}\left(\varphi + \frac{\gamma VY}{(1+aY)N^{*}}\right)(V-V^{*})(N-N^{*})\right\}$$

$$-\left\{\frac{m_{1}}{4}\left(\varphi + d + \frac{\gamma Y^{*}}{1+aY^{*}}\right)(V-V^{*})^{2} + \frac{m_{3}s_{0}}{3L(N)}(C-C^{*})^{2} + m_{1}\upsilon_{1}V^{*}(V-V^{*})(C-C^{*})\right\}$$

$$-\left\{\frac{m_{2}d}{4}(N-N^{*})^{2} + \frac{m_{3}s_{0}}{3L(N^{*})}(C-C^{*})^{2} - m_{3}\left(f(N) + s_{0}Cg(N)\right)(C-C^{*})(N-N^{*})\right\}$$

where f(N) and g(N) are defined as follows

$$f(N) = \begin{cases} \frac{s(N) - s(N^*)}{N - N^*}, & N \neq N^* \\ \frac{ds}{dN}, & N = N^* \end{cases}$$
$$g(N) = \begin{cases} \frac{L(N) - L(N^*)}{(N - N^*)L(N)L(N^*)}, & N \neq N^* \\ \frac{1}{L^2(N^*)} \frac{dL}{dN}, & N = N^* \end{cases}$$

Then by using assumptions of the theorem and the mean value theorem ,we have ,

$$\begin{split} |f(N)| &\le p \text{ and } |g(N)| \le \frac{q}{L_0^2} \\ \text{After choosing } m_0 &= 1, m_2 = \frac{1}{\alpha} \text{ we choose } m_1 \text{ and } m_3 \text{ such that:} \\ &\frac{N^* L(N^*) [\lambda (N - Y - V) + \upsilon_1 V^*]^2}{s_0 \beta Y^{*2}} < m_3 < \frac{4}{45} \frac{ds_0}{L(N^*)} \frac{1}{[p + \frac{s_0 q C_m}{L_0^2}]^2} \end{split}$$

The stability conditions can then be easily obtained, as given in the statement of the theorem. Thus negative definite under the conditions (18)-(20). Hence proof.

The above theorem implies that under appropriate conditions, if the carrier population density increases, then the number of infectives in human population also increases leading to fast spread of carrier dependent infectious diseases.

IV. VACCINE INDUCED REPRODUCTION NUMBER

We define

$$R(\varphi) = \frac{A(\beta d + \gamma \varphi)}{d(\varphi + d)(\upsilon + \alpha + d)} = \frac{A\beta}{d(\upsilon + \alpha + d)} \left[1 - \frac{(\beta - \gamma)\varphi}{\beta(\varphi + d)} \right]$$

where

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$$R'(\varphi) = -\frac{A(\beta - \gamma)d}{d(\varphi + d)^2(\upsilon + \alpha + d)} < 0 \quad (\text{since } \beta \ge \gamma)$$

Thus $R(\varphi)$ is a decreasing function in $\varphi \ge 0$. This indicates the impact of vaccination in reducing the vaccine induced reproduction number. Moreover, in the absence of vaccination i.e.,

$$\varphi = 0$$
, $R(\varphi) = \frac{A\beta}{d(\upsilon + \alpha + d)} = R_0$

From the definition of $R(\varphi)$ and R_0 , it is clear that the introduction of vaccination implies $R(\varphi) \le R_0$ and consequently, if $R_0 < 1$ then $R(\varphi) < 1$ when $\varphi > 0$. Thus E_0 is locally asymptotically stable as long as $R(\varphi)$ is less than one.

V. NUMERICAL SIMULATION

Let us take the parameters A = 100, $\beta = 0.001$, d = 0.1, $\varphi = 0.8$, v = 1.8, $\alpha = 1.9$, $\lambda = 2$,

 $\gamma = 0.001, v_1 = 2$, then we obtain $R(\varphi) = 0.263 < 1$. In this case disease dies out

Again take the parameters A = 100, $\beta = 0.001$, d = 0.0166, $\varphi = 0.65$, v = 0.9, $\alpha = 0.8$, $\lambda = 2$,

 $\gamma = 0.001, v_1 = 2$, then we obtain $R(\varphi) = 3.4962$. In this case disease persist

VI. CONCLUSIONS

In this paper, we have analyzed a vaccination model for carrier dependent population with a saturated incidence rate. We have shown that there are three non-negative equilibrium, namely disease free (DFE), carrier free (CFE) and the endemic equilibrium. The stability analysis has shown that the disease free and endemic equilibriums are locally asymptotically stable under certain condition by using Routh Hurwitz criterion.

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