The dynamics of HIV infection model with logistic growth and infected cells in eclipse phase

Sanaa Harroudi\textsuperscript{a} and Karam Allali\textsuperscript{b}

Department of Mathematics, FSTM, Laboratory of Mathematics & Applications, University Hassan II-Casablanca, PO Box 146, Mohammedia, Morocco

Abstract. In this paper, we study a mathematical model of human immunodeficiency virus dynamics with logistic growth and infected cells in eclipse phase. This model describes the interactions between uninfected CD4\textsuperscript{+} T cells, infected CD4\textsuperscript{+} T cells in latent stage, productively infected CD4\textsuperscript{+} T cells and free virus. The positivity and boundedness of solutions for non negative initial data are proved. The stability of disease-free equilibrium and endemic equilibrium are rigorously established. Numerical simulations are also provided to give a more complete representation of the system dynamics.

Keywords: HIV infection, Logistic growth, Stability, Viral dynamics.

1 Introduction

Human immunodeficiency virus (HIV) is a pathogen which causes the well known acquired immunodeficiency syndrome (AIDS). HIV continues to be a major global public health issue, having claimed more than 35 million lives so far; in 2016 more than 1 million people died from HIV-related causes globally [1]. Without treatment of the HIV infection, HIV advances in stages getting worse over time [2]. However, the most powerful antiretrovirals cannot completely eliminate the virus because it remains dormant in some cells [3]. In the last decades, many mathematical models have proved to be useful for describing and understanding the dynamics of HIV infection by considering three classes of CD4\textsuperscript{+} T cells: uninfected cells, infected cells in latent phase, and productively infected cells [4–6]. In recent years, much works extend the basic model by including a logistic growth term that describes the growth rate of healthy CD4\textsuperscript{+} T cells [7,8], since the proliferation rate of T-cells is density-dependent with the rate of proliferation slowing as the T-cell count gets high [7]. In this study, we extend the recent work [9] by considering a logistic growth rate instead of constant one and the model that we consider is the following

$$
\begin{align*}
\dot{x} &= rx(t)(1 - \frac{T(t)}{T_m}) - d_1 x(t) - \frac{k_1 x(t) v(t)}{x(t) + v(t)}, \\
\dot{y} &= \frac{k_1 x(t) v(t)}{x(t) + v(t)} - (d_2 + k_2) y(t), \\
\dot{s} &= k_2 y(t) - d_3 s(t), \\
\dot{v} &= a s(t) - d_4 v(t),
\end{align*}
$$

With $T(t) = x(t) + y(t) + s(t)$.

The initial data are

$$
x(0) = x_0 \geq 0, y(0) = y_0 \geq 0, s(0) = s_0 \geq 0, \ v(0) = v_0 \geq 0,$$

$$
T_m \geq T_0 = x_0 + y_0 + s_0 > 0.
$$

In this model, $x(t)$, $y(t)$, $s(t)$, $v(t)$ represent the concentration of uninfected CD4\textsuperscript{+} T cells, infected CD4\textsuperscript{+} T cells in latent stage, productively infected CD4\textsuperscript{+} T cells and free virus (HIV), respectively. The uninfected CD4\textsuperscript{+} T cells grow at a rate $r$, die at a rate $d_1$ and become infected by free virus at a rate $k_1$. $T_m$ is the carrying capacity of the T-cell population. Infected CD4\textsuperscript{+} T cells in latent stage are produced at a rate $k_1$, die at a rate $d_2$ and become productively infected cells at a rate $k_2$. Productively infected CD4\textsuperscript{+} T cells are produced at a rate $k_2$ and die at a rate $d_3$. Free virus (HIV) is produced from infected cells at a rate $a$ and die at a rate $d_4$.

The paper is organized as follows. The next section is devoted to the analysis of the model which contain the existence, positivity, boundedness of solutions and the stability analysis. Section 3, deals with some numerical simulations of the model. Finally, we conclude in the last section.

2 Analysis of the model

2.1 Steady states

There exist two steady states of the studied model: The infection-free equilibrium $E_1 = \left(t(\frac{r(\alpha - d_2)}{d_1}), 0, 0, 0\right)$ which represent the disease free equilibrium and correspond to the maximal level of healthy CD4\textsuperscript{+} T cells. $E^* = \left(x^*, y^*, s^*, v^*\right)$ is a state of persistent, chronic HIV infection. Explicitly, $E^*$ requires

$$
x^* = \frac{T_m d_3 d_4 (d_2 + k_2)(R^* - R_0)}{r(d_1 + k_2)(R_0 - 1) + a k_2},
$$
2.2 Positivity and boundedness of solutions

For the problem dealt with cell population evolution, the cell densities should remain non-negative and bounded. In this subsection, we will establish the positivity and boundedness of solutions of the model (1). First of all, for biological reasons, the parameters $x_0$, $y_0$, $s_0$ and $v_0$ must be larger than or equal to 0. Hence, we have the following result.

**Proposition 21** The solutions of the problem (1) exist. Moreover, they are bounded and non-negative for all $t > 0$.

**Proof** Notice that system (1) is locally lipschitzian at $t = 0$. Hence the solution of this system exists and is unique on $[0, b)$ for some $b > 0$. Observe that if $x(0) = 0$, then $x \equiv 0$ for all $t > 0$. Thus, we assume below that $x(0) > 0$. We also have the following:

$$\dot{y} = \frac{a k_1 k_2}{x + v} y \geq 0, \quad \dot{s} = k_2 y \geq 0 \quad \text{and} \quad \dot{v} = a k_2 s \geq 0.$$

This shows that $x(0) = x_0 > 0, y(0) = y_0 \geq 0, s(0) = s_0 \geq 0$ and $v(0) = v_0 \geq 0$ for all $t \in [0, b)$.

On the other hand, for the boundedness of the solutions, we have the following:

$$\frac{dT(t)}{dt} = rx(t)(1 - \frac{T(t)}{T_m}) - d_1 x(t) - d_2 y(t) - d_3 s(t),$$

since

$$T(t) \leq T_m \quad \text{and} \quad x(t) \leq T(t),$$

we obtain

$$\frac{dT(t)}{dt} \leq rT(t)(1 - \frac{T(t)}{T_m}) \leq rT(t),$$

thus

$$T(t) \leq T_0 e^{rt},$$

with $T_0 = x_0 + y_0 + s_0$.

We conclude that $T(t)$ is bounded, which means also that $x$, $y$ and $s$ are bounded.

From the last equation of (1), we have

$$v(t) \leq v(0)e^{-dt_1} + \int_0^t s(t)e^{t_1 - t_1}d\xi,$$

therefore

$$v(t) \leq v(0) + \frac{a}{d_4} \|s\|_{\infty} (1 - e^{-dt_1}).$$

Since $(1 - e^{-dt_1}) \leq 1$, we conclude that $v$ is bounded.

2.3 The stability analysis

First, the jacobian matrix of the system (1) is given by

$$J = \begin{pmatrix}
    r(1 - \frac{2x_0+y_0}{T_m}) & -\frac{a k_1^2}{(1+y_0)} & -\frac{a x_0}{T_m} \\
    k_1^2 & -(d_2 + k_2) & 0 \\
    0 & k_2 & -d_3 & 0 \\
    0 & 0 & a & -d_4
\end{pmatrix}$$

A straightforward calculation gives the following expression for the basic reproductive number in the model (1):

$$R_0 = \frac{a k_1 k_2}{d_3 d_4 (d_2 + k_2)}.$$

2.3.1 Stability of the infection-free equilibrium point $E_f$

Here, we will analyze locally asymptotical stability and globally asymptotical stability of the disease-free equilibrium $E_f$.

**Proposition 22** The free equilibrium point $E_f$ is locally asymptotically stable when $R_0 < 1$ and unstable when $R_0 > 1$.

**Proof** The Jacobian matrix at $E_f$ is given by

$$J_{E_f} = \begin{pmatrix}
    -(r - d_1) & -(r - d_1) & -(r - d_1) & -k_1 \\
    0 & -(d_2 + k_2) & 0 & k_1 \\
    0 & k_2 & -d_3 & 0 \\
    0 & 0 & a & -d_4
\end{pmatrix}$$

The characteristic polynomial of $J_{E_f}$ is

$$P_{E_f}(\xi) = (\xi + (r - d_1))[(\xi^2 + a_1 \xi^2 + a_2 \xi + a_3)],$$

where

$$a_1 = d_2 + d_3 + d_4 + k_2,$$

$$a_2 = d_3 d_4 + (d_3 + d_4)(d_2 + k_2),$$

$$a_3 = d_3 d_4 (d_2 + k_2) - ak_1 k_2.$$

While $\xi_1 = -(r - d_1)$ is a negative eigenvalue, the other three eigenvalues are given by the solution of the following cubic equation,

$$\xi^3 + a_1 \xi^2 + a_2 \xi + a_3 = 0.$$

It is clear that, $a_1 > 0, a_2 > 0$ and $a_1 a_2 - a_1 = d_3 d_4 (d_2 + d_4) + (d_2 + k_2)(d_3 + d_4)(d_2 + d_4) + ak_1 k_2 > 0$. If $R_0 < 1$, then $a_3 > 0$. From the Routh-Hurwitz Theorem given in [7], all roots of this equation have negative real parts. Then $E_f$ is locally asymptotically stable when $R_0 < 1$.

**Proposition 23** The free equilibrium point $E_f$ is globally asymptotically stable when $R_0 < 1$, provided the initial data satisfies.
Proof From positivity of solutions, we have that \( s \) and \( v \) satisfy the differential inequality

\[
d\frac{ds}{dt} \leq k_2y(t) - d_3s(t),
\]

\[
d\frac{dv}{dt} \leq as(t) - d_4v(t).
\]

It is clear from (1) that if \((x(t), v(t)) \to (0, 0)\) then \(x(t) \to T_m\) and \(y(t) \to 0\). We can apply Theorem 3.2 of [10] and conclude that \((x(t), v(t)) \to (0, 0)\) as \(t \to \infty\) when \(R_0 < 1\). Thus, \(E_f\) is globally asymptotically stable.

2.3.2 Stability of the endemic equilibrium point \(E^*\)

In this part, we discuss the local stability of the endemic infection equilibrium point \(E^*\).

Proposition 24 If \(R^* < R_0\) or \(R_0 < 1\), the point \(E^*\) does not exist. The endemic equilibrium point \(E^*\) is locally asymptotically stable when \(R_0 > 1\) and \(R^* > R_0\).

Proof From the expression of \(E^*\) we observe that this point exists when \(1 \leq R_0 \leq R^*\) and it becomes \(E_f\) when \(R_0 = 1\). We assume that \(1 \leq R_0 \leq R^*\).

The Jacobian matrix at the endemic equilibrium point \(E^*\) is given by

\[
J_{E^*} = \begin{pmatrix}
& -\frac{r(1 - \frac{1}{2}xv + \frac{1}{2}r^2s)}{xv} - d_1 - \frac{k_1}{v} - \frac{k_2}{y} - \frac{k_3}{s} \\
& \frac{r}{y} - d_2 - k_2 - d_3 - \frac{k_1}{v} - \frac{k_0}{y} \\
& 0 \\
& 0 \\
& -\frac{k_0}{y} - \frac{k_1}{v} - \frac{k_2}{y} - d_4
\end{pmatrix}
\]

(3)

The characteristic polynomial of \(J_{E^*}\) is

\[ P_{E^*}(\xi) = \xi^4 + b_1\xi^3 + b_2\xi^2 + b_3\xi + b_4. \]

From the Routh-Hurwitz theorem applied to the fourth order polynomial, the eigenvalues of the jacobian matrix (3) have negative real parts since we have \(b_1b_2 > b_3\) and \(b_1b_2b_3 > b_4^2 + b_2^2b_4\). Consequently, we obtain the asymptotic local stability of the endemic point \(E^*\).

3 Numerical simulations

In order to carried out the numerical simulations, we have used the Euler finite-difference scheme method. The parameter values or ranges used are presented in Table 1.

Figure 1 shows the behavior of the infection during the first 800 days of observation. We clearly see that the curves converge to the disease-free steady state \(E_f = (805, 0, 0, 0)\).

We assume that \(R_0 = 5776\), which corresponds to the stability of \(E_f\). Figure 2 shows the evolution of the infection during the first 800 days of observation. We use a same set of parameter values as those in Table 1, but we vary the value of \(r\). For \(r = 1\) all the curves converge to the endemic equilibrium \(E^* = (443.237, 152.462, 290.353, 1675.439)\), \(R_0 = 4.78 > 1\) and \(R^* = 10.427\) which prove the stability of this endemic equilibrium. For \(r = 2\) all the curves also converge to the endemic equilibrium \(E^* = (357.85, 147.39, 278.75, 1530.28)\).

Table 1. Parameters, their symbols and default values used in HIV literature.

| Parameters | Value | References |
|------------|-------|------------|
| \(r\)      | 0.03–3 day\(^{-1}\) | [7]         |
| \(T_m\)    | 1500 mm\(^{-3}\) | [7]         |
| \(d_1\)    | 0.0139 days\(^{-1}\) | [9]         |
| \(k_1\)    | 2.5 \times 10\(^{-4}\)-0.5 mm\(^{-3}\)day\(^{-1}\) | [9]         |
| \(d_2\)    | 0.0495 days\(^{-1}\) | [9]         |
| \(k_2\)    | 1.1 days\(^{-1}\) | [9]         |
| \(d_3\)    | 0.5776 days\(^{-1}\) | [9]         |
| \(a\)      | 2–1250 virion cell\(^{-1}\) | [9]         |
| \(d_4\)    | 0.3466–2.4 day\(^{-1}\) | [9]         |

Fig. 1. Behavior of the infection dynamics for \(r = 0.03\), \(T_m = 1500\), \(k_0 = 0.0027\), \(d_1 = 0.0139\), \(d_2 = 0.0495\), \(k_1 = 1.1\), \(d_3 = 0.5776\), \(a = 5\) and \(d_4 = 2.4\), which correspond to the stability of the free-equilibrium point \(E_f\).
4 Conclusion

In this work, we have studied a mathematical model of HIV dynamics with logistic function that describes the interactions between HIV and CD4+ T cells. The considered model includes four differential equations describing the interaction between the uninfected CD4+ T cells, infected CD4+ T cells in latent stage, productively infected CD4+ T cells and free virus (HIV). First, the existence, positivity and boundedness of solutions are proved. Next, both the local stability and global stability of the disease-free equilibrium and local stability of the endemic equilibria are established. Some numerical simulations are performed in order to confirm the theoretical results concerning the equilibrium stability.

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