Modeling the Adaptive Immunity and Both Modes of Transmission in HIV Infection

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Abstract: Human immunodeficiency virus (HIV) is a retrovirus that causes HIV infection and over time acquired immunodeficiency syndrome (AIDS). It can be spread and transmitted through two fundamental modes, one by virus-to-cell infection, and the other by direct cell-to-cell transmission. In this paper, we propose a new mathematical model that incorporates both modes of transmission and takes into account the role of the adaptive immune response in HIV infection. We first show that the proposed model is mathematically and biologically well posed. Moreover, we prove that the dynamical behavior of the model is fully determined by five threshold parameters. Furthermore, numerical simulations are presented to confirm our theoretical results.

Keywords: HIV infection; immunity; Lyapunov functional; global stability

1. Introduction

Viruses are very small infectious agents that need to penetrate inside a cell of their host to replicate and multiply. Several viruses attack the human body, such as influenza virus, human immunodeficiency virus (HIV), hepatitis B virus (HBV), hepatitis C virus (HCV), Ebola virus, Zika virus, and so on. Viral infections caused by these viruses represent a major global health problem by causing the mortality of millions of people and the expenditure of enormous amounts of money in health care and disease control. In this study, we are interested in the viral infection caused by HIV. The World Health Organization (WHO) estimates that 36.7 million people were living with HIV at the end of 2016, and more than 1 million people died from HIV-related causes in 2016 [1]. In Morocco, the number of people living with HIV is estimated at 28,740, and 1097 people died from AIDS in 2014, while the cumulative number of HIV/AIDS cases reported since the beginning of the epidemic is 10,017 [2].

Many mathematical models have been developed to better understand the dynamics of HIV infection. One of the earliest of these models was presented by Perelson et al. [3] in 1996. This model is given by the following system:

\[
\begin{align*}
\dot{T}(t) &= \lambda - dT(t) - \beta_1 T(t)V(t), \\
\dot{I}(t) &= \beta_1 T(t)V(t) - aI(t), \\
\dot{V}(t) &= kI(t) - \mu V(t), \\
\end{align*}
\]

where \( T(t) \), \( I(t) \), and \( V(t) \) are the concentrations of healthy CD4+ T cells, infected cells, and free virus at time \( t \), respectively. Healthy cells are produced at rate \( \lambda \), die at rate \( d \), and become infected by free virus at rate \( \beta_1 \). The parameter \( a \) is the death rate of infected cells. Free virus is produced by an infected cell at rate \( k \) and is removed at rate \( \mu \).
In the system given by Equation (1), the cell infection is instantaneous and is caused only by contact with free virus. In reality, there are two kinds of delays: one in cell infection, and the other in virus production. In addition, HIV can spread by two fundamental modes, one by virus-to-cell infection, and the other by direct cell-to-cell transmission. For these above reasons, Lai and Zou [4] improved the model of Perelson et al. [3] by incorporating the two modes of transmission and infinite distributed delay in cell infection. They obtained the following model:

$$\begin{align*}
T(t) &= \lambda - dT(t) - \beta_1 T(t)V(t) - \beta_2 T(t)I(t), \\
I(t) &= \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} [\beta_1 T(t-\tau)V(t-\tau) + \beta_2 T(t-\tau)I(t-\tau)] d\tau - aI(t), \\
V(t) &= kI(t) - \mu V(t),
\end{align*}$$  \hspace{1cm} (2)

where $\beta_2 T(t)I(t)$ denotes the rate for a target cell to contact with an infected cell. It is assumed that the virus or infected cell contacts an uninfected target cell at time $t - \tau$ and the cell becomes infected at time $t$, where $\tau$ is a random variable taken from a probability distribution $f_1(\tau)$. The term $e^{-\alpha_1 \tau}$ represents the probability of surviving from time $t - \tau$ to time $t$, where $\alpha_1$ is the death rate for infected but not yet virus-producing cells. The other parameters have the same biological meaning as in Equation (1).

On the other hand, the adaptive immune responses of cytotoxic T lymphocytes (CTLs) and antibodies play an important role in the control of HIV infection. The first immune response exerted by CTL cells is called the cellular immunity. However, the second immune response mediated by antibodies is called the humoral immunity. In the literature, several authors are interested in modeling the role of these arms of immunity in viral infections. In 2016, Wang et al. [5] improved the model given by Equation (2) by considering the role of cellular immune response. In the same year, Elaiw et al. [6] improved the model given by Equation (2) by considering only the role of humoral immune response and infinite distributed delay in virus production. In 2017, Lin et al. [7] improved the models of Wang and Zou [8] and Murase et al. [9] by incorporating both modes of transmission, intracellular delay and humoral immunity.

The aim of this work is to improve and generalize all the above models by proposing a new mathematical model that takes into account the role of the adaptive immune response in HIV infection and incorporates both modes of transmission. To this end, the next section deals with the presentation of our model and some properties of solutions, such as positivity and boundedness. In Section 2, we derive the threshold parameters of our model and discuss the existence of equilibria. The global stability of equilibria is investigated in Section 3. Some numerical simulations of our main results are presented in Section 4. The mathematical and biological conclusions are given in Section 5.

2. Presentation of the Model

To model the role of the adaptive immune response in HIV infection with both virus-to-cell infection and cell-to-cell transmission, we propose the following model:

$$\begin{align*}
\dot{T}(t) &= \lambda - dT(t) - \beta_1 T(t)V(t) - \beta_2 T(t)I(t), \\
\dot{I}(t) &= \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} [\beta_1 T(t-\tau)V(t-\tau) + \beta_2 T(t-\tau)I(t-\tau)] d\tau - aI(t) + \mu I(t) - Z(t), \\
\dot{V}(t) &= kI(t) - \mu V(t), \\
\dot{Z}(t) &= cI(t)Z(t) - bZ(t),
\end{align*}$$  \hspace{1cm} (3)

where $W(t)$ and $Z(t)$ are the concentrations of antibodies and CTL cells at time $t$, respectively. Free HIV particles are neutralized by the antibodies at rate $qV(t)W(t)$. However, the infected cells are killed by CTL cells at rate $pI(t)Z(t)$. Antibodies develop in response to free virus at rate $gV(t)W(t)$, and CTL cells expand in response to viral antigens derived from infected cells at rate $cI(t)Z(t)$. The parameters $h$ and $b$ are, respectively, the death rates of antibodies and CTL cells. Further, we assume that the time necessary for the newly produced virions to become mature and infectious is a random
variable with a probability distribution \( f_2(\tau) \). The term \( e^{-\alpha_2 \tau} \) denotes the probability of surviving the immature virions during the delay period, where \( \frac{1}{\alpha_2} \) is the average lifetime of an immature virus. Therefore, the integral \( \int_0^\infty f_2(\tau)e^{-\alpha_2 \tau} I(t - \tau) d\tau \) describes the mature viral particles produced at time \( t \).

The other variables and parameters are defined as those in the systems given by Equations (1) and (2).

In this section, we first investigate the nonnegativity and boundedness of solutions under the following nonnegative initial conditions:

\[
\begin{align*}
T(\theta) &= \phi_1(\theta) \geq 0, \quad I(\theta) = \phi_2(\theta) \geq 0, \quad V(\theta) = \phi_3(\theta) \geq 0, \\
W(\theta) &= \phi_4(\theta) \geq 0, \quad Z(\theta) = \phi_5(\theta) \geq 0, \quad \theta \in (-\infty, 0].
\end{align*}
\]

We define the Banach space for the fading memory type as follows:

\[
C_a = \left\{ \varphi \in C((-\infty, 0], \mathbb{R}_+^5) : \varphi(\theta)e^{at} \text{ is uniformly continuous} \right\},
\]
on \((-\infty, 0]\) and \( \|\varphi\| = \sup_{\theta \leq 0} |\varphi(\theta)e^{at}| < \infty \),

where \( a \) is a positive constant and \( \mathbb{R}_+^5 = \{ (x_1, ..., x_5) : x_i \geq 0, \ i = 1, ..., 5 \} \).

**Theorem 1.** For any initial condition \( \phi = (\phi_1, \phi_2, \phi_3, \phi_4, \phi_5) \in C_a \) satisfying Equation (4), the model given by Equation (3) has a unique solution on \([0, +\infty)\). Furthermore, this solution is nonnegative and bounded for all \( t \geq 0 \).

**Proof.** By the fundamental theory of functional differential equations \([10–12]\), the model given by Equation (3) with initial condition \( \phi \in C_a \) has a unique local solution on \([0, t_{\max})\), where \( t_{\max} \) is the maximal existence time for the solution of Equation (3).

First, we prove that \( T(t) > 0 \) for all \( t \in [0, t_{\max}) \). In fact, supposing the contrary, we let \( t_1 > 0 \) be the first time such that \( T(t_1) = 0 \) and \( \dot{T}(t_1) \leq 0 \). By the first equation of the model given by Equation (3), we have \( T(t_1) = \lambda > 0 \), which is a contradiction. Thus, \( T(t) > 0 \) for all \( t \in [0, t_{\max}) \).

By Equation (3), we have

\[
\begin{align*}
I(t) &= \phi_2(0)e^{-\int_0^t (a + pZ(s))ds} + \int_0^t e^{-\int_0^s (a + pZ(s))ds} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} [\beta_1 I(\xi - \tau)V(\xi - \tau) \\
&\quad + \beta_2 T(\xi - \tau)I(\xi - \tau)]d\tau d\xi, \\
V(t) &= \phi_3(0)e^{-\int_0^t (\mu + qW(s))ds} + k \int_0^t e^{-\int_0^s (\mu + qW(s))ds} \int_0^h f_2(\tau)e^{-\alpha_2 \tau} I(\xi - \tau)d\tau d\xi, \\
W(t) &= \phi_4(0)e^{\int_0^t (cI(s) - b)ds}, \\
Z(t) &= \phi_5(0)e^{\int_0^t (eI(s) - b)ds},
\end{align*}
\]
which implies that \( I(t), V(t), W(t), \) and \( Z(t) \) are nonnegative for all \( t \in [0, t_{\max}) \).

Next, we show the boundedness of each solution. From the first equation of the model given by Equation (3), we have \( \dot{T}(t) \leq \lambda - dT(t) \), which implies that

\[
\limsup_{t \to +\infty} T(t) \leq \frac{\lambda}{d}.
\]

Then \( T(t) \) is bounded. Let

\[
G_1(t) = I(t) + \frac{p}{c}Z(t) + \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} T(t - \tau)d\tau.
\]
Because $T(t)$ is bounded and $\int_{0}^{\infty} f_1(\tau)d\tau = 1$, the integral in $G_1(t)$ is well defined and differentiable with respect to $t$. Hence,
\[
\frac{dG_1(t)}{dt} = \lambda \int_{0}^{\infty} f_1(\tau)e^{-\alpha_1 \tau}d\tau - d \int_{0}^{\infty} f_1(\tau)e^{-\alpha_1 T(t-\tau)}d\tau - aI(t) - \frac{pb}{c}Z(t)
\leq \lambda \eta_1 - \delta_1 G_1(t),
\]
where $\delta_1 = \min\{a, b, d\}$ and
\[
\eta_i = \int_{0}^{\infty} f_i(\tau)e^{-\alpha_i \tau}d\tau, \quad i = 1, 2.
\]

Thus, $G_1(t) \leq M := \max\{G_1(0), \frac{\lambda \eta_1}{\delta_1}\}$, which implies that $I(t)$ and $Z(t)$ are bounded. It remains to prove that $V(t)$ and $W(t)$ are bounded. To this end, we consider
\[
G_2(t) = V(t) + \frac{d}{s} W(t);
\]
then
\[
\frac{dG_2(t)}{dt} = k \int_{0}^{\infty} f_2(\tau)e^{-\alpha_2 \tau}I(t-\tau)d\tau - \mu V(t) - \frac{hd}{s} W(t)
\leq kM \eta_2 - \delta_2 G_2(t),
\]
where $\delta_2 = \min\{\mu, h\}$. Similarly to the above, we deduce that $V(t)$ and $W(t)$ are also bounded. We have proved that all variables of Equation (3) are bounded, which implies that $t_{\text{max}} = +\infty$ and that the solution exists globally. \qed

If in addition to Equation (4), we suppose that $\phi_i(0) > 0$ for all $i = 1, ..., 5$; then we obtain the following remark.

**Remark 1.** If $\phi = (\phi_1, \phi_2, \phi_3, \phi_4, \phi_5) \in C_0$ satisfies Equation (4) with $\phi_i(0) > 0$, then each solution of Equation (3) with initial condition $\phi$ remains positive and bounded for all $t \geq 0$.

Next, we derive threshold numbers and identify biological equilibria for the model given by Equation (3). Clearly, Equation (3) always has an infection-free equilibrium of the form $E_0(T_0, 0, 0, 0, 0)$, where $T_0 = \frac{\lambda}{d}$. Therefore, the basic reproduction number of Equation (3) can be defined as
\[
R_0 = \frac{\beta_1 k \lambda \eta_1 \eta_2 + \beta_2 \lambda \mu \eta_1}{d \alpha \mu}.
\]

As in [13], $R_0$ can be rewritten as $R_0 = R_{01} + R_{02}$, where $R_{01} = \frac{\beta_1 k \lambda \eta_1 \eta_2}{d \alpha \mu}$ is the basic reproduction number corresponding to the virus-to-cell infection mode, and $R_{02} = \frac{\beta_2 \lambda \eta_1}{d \alpha}$ is the basic reproduction number corresponding to the cell-to-cell transmission mode.

When $R_0 > 1$, Equation (3) has another infection equilibrium without immunity, $E_1(T_1, I_1, V_1, 0, 0)$, where
\[
T_1 = \frac{\lambda}{dR_0}, \quad I_1 = \frac{d\mu (R_0 - 1)}{\beta_1 k \eta_2 + \beta_2 \mu} \quad \text{and} \quad V_1 = \frac{kd \eta_2 (R_0 - 1)}{\beta_1 k \eta_2 + \beta_2 \mu}.
\]
If both humoral and cellular immune responses have not been established, we have $gV_1 - h \leq 0$ and $cI_1 - b \leq 0$. Thus, we define the reproduction number for humoral immunity:

$$R^H_1 = \frac{gV_1}{h} = \frac{gkd\eta_2(R_0 - 1)}{h(\beta_1k\eta_2 + \beta_2\mu)},$$

and the reproduction number for cellular immunity:

$$R^Z_1 = \frac{cI_1}{b} = \frac{cd\mu(R_0 - 1)}{b(\beta_1k\eta_2 + \beta_2\mu)}.$$  

Hence, $gV_1 - h \leq 0$ and $cI_1 - b \leq 0$ are equivalent to $R^H_1 \leq 1$ and $R^Z_1 \leq 1$, respectively.

When $R^H_1 > 1$, Equation (3) has an infection equilibrium with only humoral immunity, $E_2(T_2, I_2, V_2, W_2, 0)$, where

$$T_2 = \frac{aI_2}{\eta_1(\beta_1V_2 + \beta_2I_2)}, \quad I_2 = -B + \frac{\sqrt{B^2 + 4AC}}{2A}, \quad V_2 = \frac{h}{g}, \quad W_2 = \frac{\mu}{q} \left(\frac{\eta_2l_2}{\mu V_2} - 1\right),$$

with $A = ag\beta_2, B = a(dg + h\beta_1) - \lambda\eta_1\beta_2g$, and $C = \lambda h\eta_1\beta_1$.

It is not hard to see that $T_2, I_2$ and $V_2$ are positive. It remains to check that $W_2$ is positive. Because $R^H_1 > 1$ and

$$\frac{\mu V_2}{k^2\eta_2}(A\mu V_2 + B) - C = \frac{\mu ahV_2}{k^2\eta_2} \left(1 - R^H_1\right),$$

we easily deduce that $W_2 > 0$. If cellular immunity has not been established, we have $cI_2 - b \leq 0$. For this, we define the reproduction number for cellular immunity in competition as

$$R^Z_2 = \frac{cI_2}{b},$$

which implies that $cI_2 - b \leq 0$ is equivalent to $R^Z_2 \leq 1$.

When $R^Z_2 > 1$, Equation (3) has an infection equilibrium with only cellular immunity, $E_3(T_3, I_3, V_3, 0, Z_3)$, where

$$T_3 = \frac{\lambda c\mu}{dc\mu + b(k\beta_1\eta_2 + \mu\beta_2)}, \quad I_3 = \frac{b}{c}, \quad V_3 = \frac{k\eta_2}{c\mu}, \quad Z_3 = \frac{a}{p} \left(\frac{R_0}{1 + \frac{ab}{\lambda c\eta_1} - R_0} - 1\right).$$

We have that $T_3, I_3$, and $V_3$ are positive. It suffices to check that $Z_3$ is positive. Because $R^Z_1 > 1$ and

$$\frac{R_0}{1 + \frac{ab}{\lambda c\eta_1}R_0} - 1 = \frac{b(k\beta_1\eta_2 + \mu\beta_2)}{dc\mu(1 + \frac{ab}{\lambda c\eta_1}R_0)} \left(\frac{R^Z_1}{R_0} - 1\right),$$

we deduce that $Z_3 > 0$. If humoral immunity has not been established, we have $gV_3 - h \leq 0$. In this case, we define the reproduction number for humoral immunity in competition as

$$R^H_3 = \frac{gV_3}{h},$$

which implies that $gV_3 - h \leq 0$ is equivalent to $R^H_3 \leq 1$.

When $R^Z_2 > 1$ and $R^H_3 > 1$, Equation (3) has an infection equilibrium with both cellular and humoral immune responses, $E_4(T_4, I_4, V_4, W_4, Z_4)$, where

$$T_4 = \frac{\lambda gc}{dgc + \beta_1hc + \beta_2bg}, \quad I_4 = \frac{b}{c}, \quad V_4 = \frac{h}{g}, \quad W_4 = \frac{\mu}{q} \left(R^H_3 - 1\right), \quad Z_4 = \frac{a}{p} \left(\frac{\lambda c\eta_1(\beta_1hc + \beta_2bg)}{ab(dgc + \beta_1hc + \beta_2bg)} - 1\right).$$
We have $T_4 > 0$, $I_4 > 0$, $V_4 > 0$, and $W_4 > 0$ (as $R^W_3 > 1$). It remains to check that $Z_4 > 1$. On the other hand, $R^Z_2 > 1$ is equivalent to $C > \frac{b}{c^2}(bA + cB)$. By a simple computation, we have

$$C - \frac{b}{c^2}(bA + cB) = \frac{pb}{c^2}(dgc + \beta_1 hc + \beta_2 bg)Z_4.$$

Summarizing the above discussions, we obtain the following theorem.

**Theorem 2.**

(i) If $R_0 \leq 1$, then Equation (3) always has one infection-free equilibrium, $E_0(T_0, 0, 0, 0, 0)$, where $T_0 = \frac{\lambda}{d}$.  

(ii) If $R_0 > 1$, then Equation (3) has an infection equilibrium without immunity, $E_1(T_1, I_1, V_1, 0, 0)$, where

$$T_1 = \frac{\lambda}{dR^0_0}, \quad I_1 = \frac{d\mu(R_0 - 1)}{\beta_1 k\eta_2 + \beta_2 \mu} \quad \text{and} \quad V_1 = \frac{kh\eta_2(R_0 - 1)}{\beta_1 k\eta_2 + \beta_2 \mu}.$$

(iii) If $R^W_1 > 1$, then Equation (3) has an infection equilibrium with only humoral immunity, $E_2(T_2, I_2, V_2, W_2, 0)$, where

$$T_2 = \frac{a I_2}{\eta_1(\beta_1 V_2 + \beta_2 I_2)}, \quad I_2 = -\frac{B + \sqrt{B^2 - 4AC}}{2A}, \quad V_2 = \frac{h}{\eta} \quad \text{and} \quad W_2 = \frac{\mu}{q} \left( \frac{k\eta_2 I_2}{\eta V_2} - 1 \right).$$

(iv) If $R^Z_1 > 1$, then Equation (3) has an infection equilibrium with only cellular immunity, $E_3(T_3, I_3, V_3, 0, Z_3)$, where

$$T_3 = \frac{\lambda c\mu}{d\mu + b(k\beta_1 \eta_2 + \mu \beta_2)}, \quad I_3 = \frac{b}{c} \quad \text{and} \quad Z_3 = \frac{a}{p} \left( \frac{R_0}{1 + \frac{ab}{\lambda \eta_1}} - 1 \right).$$

(v) If $R^Z_2 > 1$ and $R^W_3 > 1$, then Equation (3) has an infection equilibrium with both cellular and humoral immune responses, $E_4(T_4, I_4, V_4, W_4, Z_4)$, where

$$T_4 = \frac{\lambda gc}{dgc + \beta_1 hc + \beta_2 bg}, \quad I_4 = \frac{b}{c}, \quad V_4 = \frac{h}{\eta} \quad \text{and} \quad W_4 = \frac{\mu}{q} (R^W_3 - 1)$$

and

$$Z_4 = \frac{a}{p} \left( \frac{\lambda c \eta_1(\beta_1 hc + \beta_2 bg)}{a b (dgc + \beta_1 hc + \beta_2 bg)} - 1 \right).$$

It is important to note that

$$R^W_3 = \frac{R^W_1}{R^Z_1} = \frac{g k b \eta_2}{h \mu c}, \quad W_2 = \frac{\mu}{q} (R^Z_2 R^W_3 - 1) \quad \text{and} \quad R^W_3 > \frac{1}{R^Z_2}.$$  

(12)

3. Global Stability

In this section, we investigate the global stability of the five equilibria of Equation (3) by constructing appropriate Lyapunov functionals. We first analyze the global stability of the infection-free equilibrium.

**Theorem 3.** The infection-free equilibrium $E_0$ is globally asymptotically stable when $R_0 \leq 1$. 

Proof. To study the global stability of $E_0$, we consider a Lyapunov functional defined as follows:

$$L_0(t) = T_0 \Phi\left(\frac{T(t)}{T_0}\right) + \frac{1}{\eta_1} \int_{0}^{t} f_1(s)e^{-\alpha_1 s} ds + \frac{\beta_1 T_0}{\mu} V(t) + \frac{q \beta_1 T_0}{\mu} W(t) + \frac{p}{\rho \eta_1} Z(t)$$

$$+ \frac{1}{\eta_1} \int_{0}^{t} f_1(s)e^{-\alpha_1 s} \int_{s}^{t} |\beta_1 T(s) V(s) + \beta_2 T(s) I(s)| ds d\tau$$

$$+ \frac{k \beta_1 T_0}{\mu} \int_{0}^{t} f_2(s)e^{-\alpha_2 s} \int_{s}^{t} I(s) ds d\tau,$$

where $\Phi(x) = x - 1 - \ln x$ for $x > 0$. It is not hard to see that $\Phi(x) \geq 0$ for all $x \in (0, +\infty)$. Hence, the functional $L_0$ is nonnegative.

In order to simplify the presentation, we use the following notations: $\Psi = \Psi(t)$ and $\Psi_T = \Psi(t - T)$ for any $\Psi \in \{T, I, V, W, Z\}$. Calculating the time derivative of $L_0$ along the positive solution of Equation (3), we obtain

$$\frac{dL_0}{dt}(t) = \left(1 - \frac{T_0}{T}\right) T + \frac{1}{\eta_1} I + \frac{\beta_1 T_0}{\mu} \Phi + \frac{q \beta_1 T_0}{\mu} W + \frac{p}{\rho \eta_1} Z$$

$$+ \frac{1}{\eta_1} \int_{0}^{t} f_1(s)e^{-\alpha_1 s} \int_{s}^{t} |\beta_1 TV + \beta_2 TI| ds d\tau$$

$$+ \frac{k \beta_1 T_0}{\mu} \int_{0}^{t} f_2(s)e^{-\alpha_2 s} (I - I_T) ds d\tau$$

$$= \frac{d}{dt}(T - T_0)^2 + \frac{a}{\eta_1} (R_0 - 1) I - \frac{hq \beta_1 T_0}{\mu} W - \frac{p}{\rho \eta_1} Z.$$

If follows from $R_0 \leq 1$ that $\frac{dL_0}{dt} \leq 0$. It is straightforward to show that the largest invariant set in \{(T, I, V, W, Z) | \frac{dL_0}{dt} = 0 \} is \{E_0\}. By LaSalle’s invariance principle [14], the infection-free equilibrium $E_0$ is globally asymptotically stable when $R_0 \leq 1$. $\square$

When $R_0 > 1$, Equation (3) has four infection steady states $E_i, 1 \leq i \leq 4$. The following theorem characterizes the global stability of these steady states.

Theorem 4. Assume $R_0 > 1$.

(i) The infection equilibrium without immunity $E_1$ is globally asymptotically stable if $R_1^W \leq 1$ and $R_2^Z \leq 1$.

(ii) The infection equilibrium with only humoral immunity $E_2$ is globally asymptotically stable if $R_1^W > 1$ and $R_2^Z \leq 1$.

(iii) The infection equilibrium with only cellular immunity $E_3$ is globally asymptotically stable if $R_2^Z > 1$ and $R_3^W \leq 1$.

(iv) The infection equilibrium with both cellular and humoral immune responses $E_4$ is globally asymptotically stable if $R_2^Z > 1$ and $R_3^W > 1$.

Proof. For (i), consider the following Lyapunov functional:

$$L_1(t) = T_1 \Phi\left(\frac{T(t)}{T_1}\right) + \frac{1}{\eta_1} I_1 \Phi\left(\frac{I(t)}{I_1}\right) + \frac{\beta_1 T_1 V_1}{k \eta_2 I_1} V_1 \Phi\left(\frac{V(t)}{V_1}\right) + \frac{q \beta_1 T_1}{\mu} W(t) + \frac{p}{\rho \eta_1} Z(t)$$

$$+ \frac{\beta_1 T_1 V_1}{\eta_1} \int_{0}^{t} f_1(s)e^{-\alpha_1 s} \int_{s}^{t} \Phi\left(\frac{T(s)V(s)}{T_1 V_1}\right) ds d\tau$$

$$+ \frac{\beta_2 T_1 I_1}{\eta_1} \int_{0}^{t} f_1(s)e^{-\alpha_1 s} \int_{s}^{t} T(s) I(s) \Phi\left(\frac{I(s)}{I_1}\right) ds d\tau$$

$$+ \frac{\beta_1 T_1 V_1}{\eta_2} \int_{0}^{t} f_2(s)e^{-\alpha_2 s} \int_{s}^{t} I(s) \Phi\left(\frac{I(s)}{I_1}\right) ds d\tau.$$
Then
\[
\frac{dL_1}{dt} \big|_{(3)} = (1 - \frac{T_1}{T})^2 + \frac{1}{\eta_1}(1 - \frac{I_1}{I})l + \frac{\beta_1 T_1 V_1}{k_1 \eta_1 I_1} (1 - \frac{V_1}{V}) \Phi(\frac{T V}{T V}) - \Phi(\frac{T V}{T V}) + \frac{q \beta_1 T_1}{s \mu} W + \frac{p}{c \eta_1} Z
\]
\[+ \frac{\beta_1 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} (\Phi(\frac{T V}{T V}) - \Phi(\frac{T V}{T V})) d\tau
\]
\[+ \frac{\beta_2 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} (\Phi(\frac{T V}{T V}) - \Phi(\frac{T V}{T V})) d\tau
\]
\[+ \frac{\beta_1 T_1 V_1}{\eta_2} \int_0^\infty f_2(\tau)e^{-\alpha_2 \tau} (\Phi(\frac{V V}{V V}) - \Phi(\frac{V V}{V V})) d\tau.
\]

By \( \lambda = dT_1 + \beta_1 T_1 V_1 + \beta_2 T_1 I_1 = dT_1 + \frac{q}{\eta_1} I_1 \) and \( k \eta_1 I_1 = \mu V_1 \), we obtain
\[
\frac{dL_1}{dt} \big|_{(3)} = - \frac{d}{T} (T - T_1)^2 + \frac{h q \beta_1 T_1}{s \mu} (R_1^W - 1) W + \frac{p b}{c \eta_1} (R_1^Z - 1) Z
\]
\[+ \frac{\beta_1 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} \left[ 3 - \frac{T_1}{T} - \frac{T V I_1}{T V I_1} + \ln \left( \frac{T V}{T V} \right) \right] d\tau
\]
\[+ \frac{\beta_2 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} \left[ 2 - \frac{T_1}{T} - \frac{T I_1}{T I_1} + \ln \left( \frac{T V}{T V} \right) \right] d\tau
\]
\[+ \frac{\beta_1 T_1 V_1}{\eta_2} \int_0^\infty f_2(\tau)e^{-\alpha_2 \tau} \left[ V_1 I_1 - \ln \left( \frac{I_1}{I_1} \right) \right] d\tau.
\]

Thus,
\[
\frac{dL_1}{dt} \big|_{(3)} = - \frac{d}{T} (T - T_1)^2 + \frac{h q \beta_1 T_1}{s \mu} (R_1^W - 1) W + \frac{p b}{c \eta_1} (R_1^Z - 1) Z
\]
\[- \frac{\beta_1 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} \left[ \Phi(\frac{T V}{T V}) + \Phi(\frac{T V}{T V}) \right] d\tau
\]
\[- \frac{\beta_2 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1 \tau} \left[ \Phi(\frac{T V}{T V}) + \Phi(\frac{T V}{T V}) \right] d\tau
\]
\[- \frac{\beta_1 T_1 V_1}{\eta_2} \int_0^\infty f_2(\tau)e^{-\alpha_2 \tau} \Phi(\frac{V V}{V V}) d\tau.
\]

Because \( \Phi(x) \geq 0 \) for \( x > 0 \), \( R_1^W \leq 1 \), and \( R_1^Z \leq 1 \), we have \( \frac{dL_1}{dt} \big|_{(3)} \leq 0 \) with equality if and only if \( T = T_1, I = I_1, V = V_1, W = 0 \), and \( Z = 0 \). It follows from LaSalle’s invariance principle that \( E_1 \) is globally asymptotically stable.

For (ii), consider the following Lyapunov functional:
\[
L_2(t) = T_2 \Phi \left( \frac{T(t)}{T_2} \right) + \frac{1}{\eta_2} I_2 \Phi \left( \frac{I(t)}{I_2} \right) + \frac{\beta_1 T_2 V_2}{k_2 \eta_2 I_2} V_2 \Phi \left( \frac{V(t)}{V_2} \right)
\]
\[+ \frac{q \beta_1 T_2 V_2}{k_2 \eta_2 I_2} W_2 \Phi \left( \frac{W(t)}{W_2} \right) + \frac{p}{c \eta_1} Z(t)
\]
\[+ \frac{\beta_1 T_2 V_2}{\eta_2} \int_0^t f_1(\tau)e^{-\alpha_1 \tau} \int_{t-\tau}^t \frac{T(s)V(s)}{T_2 V_2} ds d\tau
\]
\[+ \frac{\beta_2 T_2 I_2}{\eta_1} \int_0^t f_1(\tau)e^{-\alpha_1 \tau} \int_{t-\tau}^t \frac{T(s)I(s)}{T_2 I_2} ds d\tau
\]
\[+ \frac{\beta_1 T_2 V_2}{\eta_2} \int_0^t f_2(\tau)e^{-\alpha_2 \tau} \int_{t-\tau}^t \frac{I(s)}{I_2} ds d\tau.
\]
Thus, \[
\frac{dL_2}{dt} \bigg|_{(3)} = (1 - \frac{T_2}{T}) T + \frac{1}{\eta_1} (1 - \frac{I_2}{I}) I + \beta_1 T_2 V_2 \frac{(1 - V_2)}{V} \]
\[
+ \frac{q \beta_1 T_2 V_2}{g k \eta_2 I_2} (1 - \frac{W_2}{W}) W + \frac{p}{c \eta_1} \dot{Z} \\
+ \frac{\beta_1 T_1 V_1}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left( \Phi(\frac{TV}{T_1 V_1}) - \Phi(\frac{T_1 V_1}{TV}) \right) d\tau \\
+ \frac{\beta_2 T_1 I_1}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left( \Phi(\frac{T I}{T_1 I_1}) - \Phi(\frac{T_1 I_1}{TI}) \right) d\tau \\
+ \frac{\beta_1 T_1 V_1}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_1 \tau} \left( \Phi(\frac{I}{I_1}) - \Phi(\frac{I_1}{I}) \right) d\tau.
\]

By \( \lambda = dT_2 + \beta_1 T_2 V_2 + \beta_2 T_2 I_2 = dT_2 + \frac{a}{Q_1} I_2, \) \( V_2 = \frac{h}{g}, \) and \( k \eta_2 I_2 = \mu V_2 + q W_2 V_2, \) we obtain
\[
\frac{dL_2}{dt} \bigg|_{(3)} = -\frac{d}{T} (T - T_2)^2 + \frac{p b}{c \eta_1} (R_2^Z - 1) Z \\
+ \frac{\beta_1 T_2 V_2}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ 3 - \frac{T_2}{T} - \frac{T_1 V_1 I_2}{T_2 V_2 I} + \ln \left( \frac{T_1 V_1}{TV} \right) \right] d\tau \\
+ \frac{\beta_2 T_2 I_1}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ 2 - \frac{T_2}{T} - \frac{T_1 I_1}{T_2 I} + \ln \left( \frac{T_1 I_1}{TI} \right) \right] d\tau \\
- \frac{\beta_1 T_2 V_2}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_2 \tau} \left[ \frac{V_2 I_2}{V_2 I} - \ln \left( \frac{I_2}{T} \right) \right] d\tau.
\]

Thus,
\[
\frac{dL_2}{dt} \bigg|_{(3)} = -\frac{d}{T} (T - T_2)^2 + \frac{p b}{c \eta_1} (R_2^Z - 1) Z \\
- \frac{\beta_1 T_2 V_2}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ \Phi(\frac{T_2}{T}) + \Phi(\frac{T_1 V_1 I_2}{T_2 V_2 I}) \right] d\tau \\
- \frac{\beta_2 T_2 I_1}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ \Phi(\frac{T_2}{T}) + \Phi(\frac{T_1 I_1}{T_2 I}) \right] d\tau \\
- \frac{\beta_1 T_2 V_2}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_2 \tau} \Phi(\frac{V_2 I_2}{V_2 I}) d\tau.
\]

Because \( R_2^Z \leq 1 \) and \( \Phi(x) \geq 0 \) for \( x > 0 \), we have \( \frac{dL_2}{dt} \bigg|_{(3)} \leq 0 \) with equality if and only if \( T = T_2, \) \( I = I_2, \) and \( V = V_2. \) Then \( I = 0 \) and \( V = 0, \) which leads to \( Z = 0 \) and \( W = W_2. \) Therefore, the largest compact invariant set in \( \{(T, I, V, W, Z) \mid \frac{dL_2}{dt} = 0\} \) is the singleton \( \{E_2\}, \) and the proof of (ii) is completed.
For (iii), consider the following Lyapunov functional:

\[
L_3(t) = T_3 \Phi \left( \frac{T(t)}{T_3} \right) + \frac{1}{\eta_1} I_3 \Phi \left( \frac{I(t)}{I_3} \right) + \frac{\beta_1 T_3 V_2}{\kappa \eta_2 I_3} V_3 \Phi \left( \frac{V(t)}{V_3} \right) \\
+ \frac{q \beta_1 T_3}{q \mu} W(t) + \frac{p}{c \eta_1} Z_2 \Phi \left( \frac{Z(t)}{Z_2} \right) \\
+ \frac{\beta_1 T_3 V_3}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \int_{t-\tau}^t F \left( \frac{T(s)V(s)}{T_3 V_3} \right) ds d\tau \\
+ \frac{\beta_2 T_3 I_3}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \int_{t-\tau}^t F \left( \frac{T(s)I(s)}{T_3 I_3} \right) ds d\tau \\
+ \frac{\beta_2 T_3 V_3}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_2 \tau} \int_{t-\tau}^t F \left( \frac{I(s)}{I_3} \right) ds d\tau.
\]

From \( \lambda = dT_3 + \beta_1 T_3 V_3 + \beta_2 T_3 I_3 = dT_3 + \frac{a}{c} I_2 + \frac{p}{c} I_3 Z_3, I_3 = \frac{b}{c} \), and \( \kappa \eta_2 I_3 = \mu V_3 \), we easily have

\[
\frac{dL_3}{dt} \big|_{(3)} = - \frac{d}{T} (T - T_3)^2 + \frac{hq \beta_1 T_3}{q \mu} (R_3^W - 1) W \\
- \frac{\beta_1 T_3 V_3}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ \Phi \left( \frac{T_3}{T} \right) + \Phi \left( \frac{T_3 V_3 I_3}{T_3 V_3 I_3} \right) \right] d\tau \\
- \frac{\beta_2 T_3 I_3}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ \Phi \left( \frac{T_3}{T} \right) + \Phi \left( \frac{T_3 I_3}{T_3 I_3} \right) \right] d\tau \\
- \frac{\beta_2 T_3 V_3}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_2 \tau} \Phi \left( \frac{V_3 I_3}{V_3 I_3} \right) d\tau.
\]

Consequently, \( \frac{dL_3}{dt} \big|_{(3)} \leq 0 \) with equality if and only if \( T = T_3, I = I_3, \) and \( V = V_3 \). It follows from \( l = 0 \) and \( \dot{V} = 0 \) that \( Z = Z_3 \) and \( W = 0 \). By LaSalle’s invariance principle, we deduce that \( E_3 \) is globally asymptotically stable.

Finally, we show (iv) by considering the following Lyapunov functional:

\[
L_4(t) = T_4 \Phi \left( \frac{T(t)}{T_4} \right) + \frac{1}{\eta_1} I_4 \Phi \left( \frac{I(t)}{I_4} \right) + \frac{\beta_1 T_4 V_4}{\kappa \eta_2 I_4} V_4 \Phi \left( \frac{V(t)}{V_4} \right) \\
+ \frac{q \beta_1 T_4 V_4}{q \mu} W_4 \Phi \left( \frac{W(t)}{W_4} \right) + \frac{p}{c \eta_1} Z_4 \Phi \left( \frac{Z(t)}{Z_4} \right) \\
+ \frac{\beta_1 T_4 V_4}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \int_{t-\tau}^t F \left( \frac{T(s)V(s)}{T_4 V_4} \right) ds d\tau \\
+ \frac{\beta_2 T_4 I_4}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \int_{t-\tau}^t F \left( \frac{T(s)I(s)}{T_4 I_4} \right) ds d\tau \\
+ \frac{\beta_2 T_4 V_4}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_2 \tau} \int_{t-\tau}^t F \left( \frac{I(s)}{I_4} \right) ds d\tau.
\]

By \( \lambda = dT_4 + \beta_1 T_4 V_4 + \beta_2 T_4 I_4 = dT_4 + \frac{a}{c} I_4 + \frac{p}{c} I_4 Z_4, I_4 = \frac{b}{c} \), and \( \kappa \eta_2 I_4 = (\mu + q W_4) V_4 \), we obtain

\[
\frac{dL_4}{dt} \big|_{(3)} = - \frac{d}{T} (T - T_4)^2 + \frac{\beta_1 T_4 V_4}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ 3 - \frac{T_4}{T} - \frac{T_4 V_4 I_4}{T_4 V_4 I_4} \ln \left( \frac{T_4 V_4}{T V} \right) \right] d\tau \\
+ \frac{\beta_2 T_4 I_4}{\eta_1} \int_0^\infty f_1(\tau) e^{-\alpha_1 \tau} \left[ 2 - \frac{T_4}{T} - \frac{T_4 I_4}{T_4 I_4} \ln \left( \frac{T_4 I_4}{T I} \right) \right] d\tau \\
- \frac{\beta_1 T_4 V_4}{\eta_2} \int_0^\infty f_2(\tau) e^{-\alpha_2 \tau} \left[ \frac{V_4 I_4}{V I_4} - \frac{I_4}{I} \right] d\tau.
\]
Hence,
\[
\frac{dL_4}{dt} \big|_{(3)} = \frac{d}{dt} (T - T_4)^2 - \frac{\beta_1 T_4 V_4}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1\tau} \left[ \Phi\left(\frac{T_4}{T}\right) + \Phi\left(\frac{T_4 V_4}{T_4 V_4 I_4}\right) \right] d\tau
\]
\[
- \frac{\beta_2 T_4 I_4}{\eta_1} \int_0^\infty f_1(\tau)e^{-\alpha_1\tau} \left[ \Phi\left(\frac{T_4}{T}\right) + \Phi\left(\frac{T_4 I_4}{T_4 V_4}\right) \right] d\tau
\]
\[
- \frac{\beta_1 T_4 V_4}{\eta_2} \int_0^\infty f_2(\tau)e^{-\alpha_2\tau} \Phi\left(\frac{V_4 I_4}{V I_4}\right) d\tau.
\]

Thus, \(\frac{dL_4}{dt} \big|_{(3)} \leq 0\) with equality holds if and only if \(T = T_4, I = I_4,\) and \(V = V_4.\) Let
\[
\Gamma = \{(T, I, V, W, Z)|\frac{dL_4}{dt} = 0\}.
\]

From the second and third equations of the model given by Equation (3), we have
\[
\begin{align*}
\dot{I} &= \eta_1 (\beta_1 T_4 V_4 + \beta_2 T_4 I_4) - a I_4 - p I_4 Z = 0, \\
\dot{V} &= k \eta_2 I_4 - \mu V_4 - q V_4 W = 0,
\end{align*}
\]
which implies that \(Z = Z_4\) and \(W = W_4.\) Then the largest compact invariant set in \(\Gamma\) is the singleton \(\{E_4\}.\) Therefore, \(E_4\) is globally asymptotically stable. □

The conditions of the global stability of \(E_2\) and those of \(E_3\) given in (ii) and (iii) of Theorem 4 do not hold simultaneously. In fact, supposing the contrary, then \(R_1^W > 1 \geq R_2^W\) and \(R_1^Z > 1 \geq R_2^W.\) Because \(R_3^W \leq 1\) and \(R_2^Z > \frac{1}{R_3^Z},\) we have \(R_2^Z > 1.\) This is a contradiction with \(R_2^Z \leq 1.\)

According to Equation (12) and Theorem 4, we have the following important result.

**Remark 2.** Assume \(R_0 > 1.\)

1. If \(\max(R_1^W, R_1^Z) \leq 1,\) then Equation (3) converges to \(E_1\) without immunity.
2. If \(\max(R_1^W, R_1^Z) > 1,\) two cases arise:
   
   (i) When \(\max(R_1^W, R_1^Z) = R_1^W,\) the humoral immunity is dominant, and Equation (3) converges to \(E_2\) if \(R_2^Z \leq 1\) or to \(E_4\) if \(R_2^Z > 1.
   
   (ii) When \(\max(R_1^W, R_1^Z) = R_1^Z,\) the cellular immunity is dominant, and Equation (3) converges to \(E_3\) without humoral immunity.

From this important remark, we can define the over-domination of humoral immunity when \(R_2^Z > 1\) and \(R_3^W > 1\) and the over-domination of cellular immunity when \(R_2^Z > 1\) and \(R_3^W < 1\).

4. **Numerical Simulations**

In this section, we present some numerical simulations in order to validate our theoretical results. For simplicity, we chose \(f_1(\tau) = \delta(\tau - \tau_1)\) and \(f_2(\tau) = \delta(\tau - \tau_2)\) with \(\tau_1\) and \(\tau_2\) to be the delays in cell infection and virus production, respectively, and \(\delta(\cdot)\) to be the Dirac function; then our model becomes

\[
\begin{align*}
T(t) &= \lambda - d T(t) - \beta_1 T(t) V(t) - \beta_2 T(t) I(t), \\
I(t) &= \beta_1 T(t - \tau_1) V(t - \tau_1) + \beta_2 T(t - \tau_1) I(t - \tau_1) e^{-\alpha_1 \tau_1} - a I(t) - p I(t) Z(t), \\
V(t) &= k I(t - \tau_2) e^{-\alpha_2 \tau_2} - \mu V(t) - q V(t) W(t), \\
W(t) &= g V(t) W(t) - h W(t), \\
Z(t) &= c I(t) Z(t) - b Z(t).
\end{align*}
\]

(13)
This system improves the model presented in 2017 by Lin et al. [7], which considered only the humoral immunity and discrete delay in cell infection and ignored the time delay in virus production; that is, \( \tau_2 = 0 \). In addition, Equation (13) includes many special cases existing in the literature. For example, when \( \beta_2 = 0 \) and \( \tau_1 = \tau_2 = 0 \), we obtain the model presented by Wodarz in [15] and analyzed by Hattaf et al. in [16]. When \( \beta_2 = 0 \) and \( \tau_2 = 0 \), we obtain the model of Yan and Wang [17].

The algorithm for the numerical treatment of the delay differential system given by Equation (13) can be derived for the numerical method presented in [18,19]. Recently, this numerical method has been used for delayed partial differential equations [20]; it is called the “mixed” Euler method, as it is a mixture of both forward and backward Euler methods. In addition, it is shown that this mixed Euler method preserves the qualitative properties of the corresponding continuous system, such as positivity, boundedness, and global behaviors of solutions. Hence, we discretize the continuous system given by Equation (13) by this numerical method. Thus, we let \( \Delta t \) be a time step size and assume that there exist two integers \( (m_1, m_2) \in \mathbb{N}^2 \) with \( \tau_1 = m_1 \Delta t \) and \( \tau_2 = m_2 \Delta t \). The grids points are \( t_n = n \Delta t \) for \( n \in \mathbb{N} \). By applying the mixed Euler method and using the approximations \( T(t_n) \approx T_n, I(t_n) \approx I_n, \quad V(t_n) \approx V_n, W(t_n) \approx W_n, \) and \( Z(t_n) \approx Z_n \), we obtain the following discrete model:

\[
\begin{align*}
T_{n+1} &= T_n + \left( \lambda - d T_{n+1} - \beta_1 T_{n+1} V_n - \beta_2 T_{n+1} I_n \right) \Delta t, \\
I_{n+1} &= I_n + \left[ \beta_1 T_{n-m_1+1} V_{n-m_1} + \beta_2 T_{n-m_1+1} I_{n-m_1} \right] e^{-\alpha_1 \tau_1} - a I_{n+1} - p I_{n+1} Z_n \Delta t, \\
V_{n+1} &= V_n + \left( k e^{-\alpha_2 \tau_2} I_{n-m_2+1} - \mu V_{n+1} - q V_{n+1} W_n \right) \Delta t, \\
W_{n+1} &= W_n + \left( g V_{n+1} W_{n+1} - h W_{n+1} \right) \Delta t, \\
Z_{n+1} &= Z_n + \left( c I_{n+1} Z_{n+1} - b Z_{n+1} \right) \Delta t,
\end{align*}
\]

where the discrete initial conditions are

\[
T_s = \phi_1(t_s), \quad I_s = \phi_2(t_s), \quad V_s = \phi_3(t_s), \quad W_s = \phi_4(t_s), \quad Z_s = \phi_5(t_s),
\]

for \( s \in \{-m, -m+1, \cdots, 0\} \) and \( m = \max(m_1, m_2) \).

The five threshold parameters \( R_0, R_1^V, R_2^V, R_3^V, \) and \( R_4^V \) for Equation (13) are given by Equations (7)–(11) with \( \eta_1 = e^{-\alpha_1 \tau_1} \) and \( \eta_2 = e^{-\alpha_2 \tau_2} \). In order to study the impact of cell-to-cell transmission and both arms of adaptive immunity on the HIV dynamics, we chose \( \beta_2, g, \) and \( c \) as free parameters. The units of state variables \( T, I, \) and \( Z \) were given by cells \( \mu \text{L}^{-1} \). Further, the units of \( V \) and \( W \) were given by virions \( \mu \text{L}^{-1} \) and molecules \( \mu \text{L}^{-1} \), respectively. The other parameter values for the simulation are listed in Table 1.
Table 1. List of parameters.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Unit</th>
<th>Value</th>
<th>Range</th>
<th>Source</th>
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<tr>
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<td>day$^{-1}$</td>
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<td>—</td>
<td>[22]</td>
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<td>$\beta_1$</td>
<td>$\mu$L virion$^{-1}$ day$^{-1}$</td>
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<td>$2.4 \times 10^{-5}$–$4.8 \times 10^{-3}$</td>
<td>[3,23]</td>
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<td>0.2666–0.7073</td>
<td>[3,24,25]</td>
</tr>
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<td>$a_1$</td>
<td>day$^{-1}$</td>
<td>0.01</td>
<td>—</td>
<td>[26]</td>
</tr>
<tr>
<td>$\mu$</td>
<td>day$^{-1}$</td>
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<td>2.06–3.81</td>
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<tr>
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<td>virion cell$^{-1}$ day$^{-1}$</td>
<td>50</td>
<td>27–7073</td>
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<td>day$^{-1}$</td>
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<td>—</td>
<td>[26]</td>
</tr>
<tr>
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<td>molecule$^{-1}$ $\mu$L day$^{-1}$</td>
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<td>—</td>
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<td>$h$</td>
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<td>—</td>
<td>Assumed</td>
</tr>
<tr>
<td>$b$</td>
<td>day$^{-1}$</td>
<td>0.1</td>
<td>0.05–0.15</td>
<td>[28,29]</td>
</tr>
<tr>
<td>$\tau_1$</td>
<td>days</td>
<td>1.5</td>
<td>0–2</td>
<td>[31]</td>
</tr>
<tr>
<td>$\tau_2$</td>
<td>days</td>
<td>0.5</td>
<td>—</td>
<td>[32]</td>
</tr>
</tbody>
</table>

For the case in which $\beta_2 = 10^{-6}$, $g = 10^{-5}$, and $c = 0.002$, we obtained $R_0 = 0.9751$. It follows from Theorem 2 that Equation (13) has one infection-free equilibrium $E_0(719.4245, 0, 0, 0, 0)$. From Figure 1, we see that the concentration of uninfected CD4$^+$ T cells increased and tended to the value $T_0 = 719.4245$, while the concentrations of infected cells, free HIV particles, antibodies, and CTL cells decreased and tended to zero. This means that $E_0$ is globally asymptotically stable and that the virus will be cleared. This confirms the result in Theorem 3.

For the case in which $\beta_2 = 1.5 \times 10^{-4}$, $g = 10^{-5}$, and $c = 0.002$, we obtained $R_0 = 1.3392$, $R_{W1} = 0.0143$, and $R_{Z1} = 0.1721$. It follows that case (i) of Theorem 4 occurs, and the infection equilibrium without immunity $E_1(537.9692, 8.5643, 142.0261, 0, 0)$ is globally asymptotically stable (see Figure 2).

For the case in which $\beta_2 = 3.4 \times 10^{-4}$, $g = 10^{-3}$, and $c = 0.002$, we obtained $R_0 = 1.8036$, $R_{W1} = 2.5099$, and $R_{Z2} = 0.8292$. By (ii) of Theorem 4, the infection equilibrium with only humoral immunity $E_2(507.5951, 10.0014, 99.9785, 3.9525, 0)$ is globally asymptotically stable (see Figure 3).

For the case in which $\beta_2 = 5.4 \times 10^{-4}$, $g = 10^{-3}$, and $c = 0.02$, we obtained $R_0 = 2.2923$, $R_{W1} = 3.8301$, and $R_{Z2} = 1.8783$. By (iii) of Theorem 4, the infection equilibrium with only cellular immunity $E_3(537.3806, 5.0015, 83.1539, 0, 211.6047)$ is globally asymptotically stable (see Figure 4).

For the case in which $\beta_2 = 5.4 \times 10^{-4}$, $g = 10^{-2}$, and $c = 0.02$, we obtained $R_0 = 2.2923$, $R_{Z2} = 1.8783$, and $R_{W3} = 8.2918$. Using (iv) of Theorem 4, we deduce that the infection equilibrium with both arms of immunity $E_4(594.0271, 4.9769, 10.0246, 43.3843, 53.3821)$ is globally asymptotically stable (see Figure 5).
Figure 1. Demonstration of the global stability of the infection-free equilibrium $E_0$ for $R_0 = 0.9751 \leq 1$.

Figure 2. Demonstration of the global stability of the infection equilibrium without immunity $E_1$ for $R_0 = 1.3392 > 1$, $R_1^W = 0.0143 \leq 1$, and $R_1^Z = 0.1721 \leq 1$. 
Figure 3. Demonstration of the global stability of the infection equilibrium with only humoral immunity $E_2$ for $R_0 = 1.8036 > 1$, $R^W_1 = 2.5099 > 1$, and $R^Z_2 = 0.2000 \leq 1$.

Figure 4. Demonstration of the global stability of the infection equilibrium with only cellular immunity $E_3$ for $R_0 = 2.2923 > 1$, $R^Z_1 = 3.8301 > 1$, and $R^W_3 = 0.8292 \leq 1$. 
Figure 5. Demonstration of the global stability of the infection equilibrium with both arms of immunity $E_4$ for $R_0 = 2.2923 > 1$, $R_Z^2 = 1.8783 > 1$, and $R_W^3 = 8.2918 > 1$.

5. Conclusions

In this paper, we have modeled the role of the adaptive immunity in HIV infection by proposing a new mathematical model that takes into account the classical virus-to-cell infection, the direct cell-to-cell transmission, and the two kinds of delays during infection processes and virus production. By a rigorous mathematical analysis, we have proved that the global dynamics of the proposed model is fully determined by five threshold parameters, which are the basic reproduction number $R_0$, and the reproduction numbers for humoral immunity $R_W^1$, for cellular immunity $R_Z^1$, for cellular immunity in competition $R_Z^2$, and for humoral immunity in competition $R_W^3$. More precisely, the infection-free equilibrium is globally asymptotically stable if $R_0 \leq 1$, which biologically means that the HIV is cleared and the infection dies out. When $R_0 > 1$, our model has four infection equilibria, which are the following: (i) the infection equilibrium without immunity is globally asymptotically stable if $R_W^1 \leq 1$ and $R_Z^2 \leq 1$; (ii) the infection equilibrium with only humoral immunity is globally asymptotically stable if $R_W^1 > 1$ and $R_Z^2 \leq 1$; (iii) the infection equilibrium with only cellular immunity is globally asymptotically stable if $R_Z^1 > 1$ and $R_W^3 \leq 1$; (iv) the infection equilibrium with both cellular and humoral immune responses is globally asymptotically stable if $R_Z^2 > 1$ and $R_W^3 > 1$. Biologically, this implies that the HIV persists and the infection becomes chronic when the basic reproduction number is greater than 1. Additionally, the activation of one or both arms of immunity is unable to eliminate the virus in the human body.

From Remark 2, we can deduce that the over-domination of cellular immunity leads to the absence of the humoral immunity, and the over-domination of the humoral immunity leads to the persistence of HIV infection with a weak response of both arms of immunity. This biological result may be an explanation for the dysfunction of the adaptive immunity in HIV-infected patients. Our model can be adapted for HBV infection, and the above result can also explain the dysfunction of the adaptive immune response in patients infected with HBV, which is still largely incomplete [33].

Furthermore, the model and the results presented in this study improve and generalize the models and the corresponding results in more recent papers with only cellular immunity [5], with only humoral immunity [6–9], and with both arms of immunity [15,17].
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References


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