Global Dynamics of a Delayed Fractional-Order Viral Infection Model With Latently Infected Cells

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In this paper, we propose a fractional-order viral infection model, which includes latent infection, a Holling type II response function, and a time-delay representing viral production. Based on the characteristic equations for the model, certain sufficient conditions guarantee local asymptotic stability of infection-free and interior steady states. Whenever the time-delay crosses its critical value (threshold parameter), a Hopf bifurcation occurs. Furthermore, we use LaSalle’s invariance principle and Lyapunov functions to examine global stability for infection-free and interior steady states. Our results are illustrated by numerical simulations.

Keywords: bifurcation, fractional order, viral infection model, stability, time-delay

1 INTRODUCTION

Various mathematical models have been developed to describe, within-host, the dynamics of various viral infections, with a focus on virus-to-cell transmission, such as human immunodeficiency virus (HIV) [1], COVID-19 [2,3], hepatitis C virus (HCV) [4], hepatitis B virus (HBV) [5], and human T-cell lymphotropic virus 1 (HTLV-1) [6]. Classical viral infection models are composed of the interactions between susceptible cells, infected target cells, and free viruses. Furthermore, some authors studied latent infection to describe the mechanism of latency. Wang et al. [7] investigated the HIV model with latent infection incorporating both modes of time-delays and transmissions between viral entry and viral production or integration and also discussed the basic reproductive number, showing existence of asymptotic stability of endemic equilibrium points. Wen et al. [8] studied the virus-to-cell and cell-to-cell HIV virus transmission dynamics with latently infected cells. Pan et al. [9] discussed the HCV infection model, which includes the routes of infection and spread, such as virus-to-cell and cell-to-cell transmission dynamics, and explained numerically the four different HCV models.

The infection persists when the virus weakens and suppresses the immune response. Immune system response refers to the process that when the virus enters the human body, the immune system receives the signal of virus attack and spreads it to the immune organs, which secrete lymphocytes to purge the virus. Moreover, the adaptive immune response plays a crucial role in the control of infection process. When the virus spreads in the human body, the human body produces double modes of immune responses: one is the B cell that makes humoral immune response and the other is the cytotoxic T lymphocyte (CTL) that makes cellular immune response. Previous studies have shown that the humoral immune response is more active than the cellular immune responses. Elaiw et al. [10] discussed the dynamical behaviors of viral infection models with latently infected cells, humoral immune response, and general non-linear incidence rate function. The authors in [11] investigated the global asymptotic stability of reaction–diffusion viral infection model with
homogeneous environments and non-linear incidence in heterogeneous and humoral immunity. Wang et al. [12] discussed the global stability results of HIV viral infection model with latently infected cells, B-cell immune response, Beddington–DeAngelis functional response, and various time-delays. The authors in [13] reported the stability and bifurcation results of generalized viral infection system with humoral immunity and distributed delays in virus production and cell infection, and time lags described the time needed to activate the immune response.

The fractional-order modeling provides more reliable, accurate, and precise information about the dynamics of biological models. Description of memory and hereditary properties make it superior to the classical-order model. Additionally, non-integer–order models can easily demonstrate and explore the dynamics between two non-local points and are more comprehensive to discuss real-world problems for the global dynamics. In recent years, various ideas and theories about the fractional-order derivatives have been developed and introduced (see [14–16]). Rakkhiyappan et al. [17] discussed the fractional-order Zika viral infection model with different time-delays, such as the latency infection in the infected host and in a vector, and sufficient conditions for stability and bifurcation results with respect to time lags. The authors in [18] analyzed the local stability results of fractional-order Ebola viral infection model with time-delayed immune response (cytotoxic T-lymphocyte term) in heterogeneous complex networks. Rihan et al. [19] studied the fractional-order dynamics of hepatitis C virus (HCV) incorporating the intracellular delay between the initial infection of a cell by HCV and the release of new virions and the interferon-α (IFN-α) treatment. Tamilagaran et al. [20] reported the characteristic dynamics for the non-integer–order HIV infection model with CTL immune responses and antibody and also compared it with that of the integer-order model. The authors reported in [21] derived the sufficient conditions of stability and optimal control results for the fractional-order HIV model with transmission dynamics. Researchers investigated fractional-order viral infection models to understand how the immune system works and examined how immune cells eliminate virus [22,23].

With motivation of the above-mentioned studies, in this paper, we extend the integer-order differential model into the fractional-order model for the virus-immune system with latently infected cell compartments. We incorporate fractional order and time-delay to represent long-run and short-run memory. As part of the Holling type II functional response, viral pathogens spread from virus to cell and from cell to cell. To avoid dimensional mismatches in the fractional equations, a parameter modification is used. We derive the positivity and boundedness of the solution for the considered model. We examine the local stability of existing equilibrium points and the bifurcation results of the considered model. We also discuss some necessary conditions for global stability using a novel suitable Lyapunov function combined with LaSalle’s invariance principle. The rest of this paper is organized as follows: In Section 2, we formulate the viral infection model and also study the positivity and boundedness of the solution. Local and global stability results of such a model are derived in Section 3 and Section 4, respectively. The numerical simulations are provided in Section 5, to validate the obtained theoretical results. Section 6 contains conclusion.

## 2 MODEL FORMULATION

In the process of viral infection, the immune system plays a critical role. Viral dynamics can be modeled properly to provide insights into understanding the disease and the clinical treatments used to treat it. In adaptive immune responses, lymphocytes are responsible for specificity and memory. The two main types of lymphocytes are B cells and T cells. The function of T cells is to recognize and kill infected cells, while the function of B cells is to produce antibodies to neutralize the viruses. Researchers have studied the effects of immune responses such as CTL responses and antibody responses [24–27]. Some other researchers have also taken into account the effect of CTL responses and intracellular delays [16,19,28]. The mathematical model that describes the effect of humoral immune response on virus dynamics is presented as [29]

\[
\begin{align*}
\dot{x}(t) &= \lambda - \alpha x(t) - \beta x(t)v(t) - \beta^* x(t)y(t), \\
\dot{y}(t) &= \beta x(t)v(t) + \beta^* x(t)y(t) - a y(t), \\
\dot{v}(t) &= ky(t) - \mu v(t) - \xi v(t)w(t), \\
\dot{w}(t) &= g v(t)w(t) - hw(t),
\end{align*}
\]

where \(x(t), y(t), v(t),\) and \(w(t)\) represent the concentration of uninfected target cells, actively infected cells, free viruses, and antibodies/B cells at time \(t\), respectively. The uninfected cells \(x(t)\) are assumed to produce/grow at a constant rate \(\lambda\) and death rate \(d\). \(\beta\) is the infection rate by free virus, and \(\beta^*\) is the rate at which uninfected cells are converted to productively infected ones per both cells. \(\alpha, \beta, h\) are the death rate of infected cells, free virus particles, and antibodies/B cells, respectively. Free virus particles are produced from productively infected cells at the rate \(\hat{k}\), and \(\hat{\xi}\) is the rate of neutralization by antibodies. \(\hat{g}\) is the rate of antibodies activated against the virus.

Herein, we upgrade the model (1) to include the latent infection component. We assume that the uninfected cell \(x(t)\) becomes infected by free virus or by direct contact with an infected cell at the rate \(\frac{\beta x(t)v(t)}{1 + v(t)} + \frac{\beta^* x(t)y(t)}{1 + y(t)}\) with Holling type II functional response. We also assume that \((1 - \varphi)\) and \(\varphi \in (0, 1)\) are the proportions of infection that lead to latency and productivity, respectively. A fractional order \((0 < \alpha \leq 1)\) is also considered in the model to represent the long-run memory. The model takes the form

\[
\begin{align*}
D^\alpha x(t) &= \lambda - \alpha x(t) - \beta x(t)v(t) + \beta^* x(t)y(t) - \frac{\beta x(t)v(t)}{1 + v(t)} - \frac{\beta^* x(t)y(t)}{1 + y(t)}, \\
D^\alpha y(t) &= (1 - \varphi)\left(\beta x(t)v(t) + \beta^* x(t)y(t)\right) - (m + \gamma)l(t), \\
D^\alpha v(t) &= \varphi\left(\beta x(t)v(t) + \beta^* x(t)y(t)\right) + \gamma l(t) - a y(t), \\
D^\alpha w(t) &= k e^{-\alpha t} y(t - \tau) - \mu v(t) - \xi v(t)w(t), \\
D^\alpha \omega(t) &= g v(t)w(t) - hw(t),
\end{align*}
\]

where \(D^\alpha\) is the Riemann–Liouville fractional derivative of order \(\alpha\).
with initial values $x(0) > 0$, $l(0) > 0$, $y(s) = \chi(s) > 0$, $v(0) > 0$, $w(0) > 0$, $s \in [-\tau, 0]$ and $\chi(s)$ being the smooth function. Here, $D^\alpha$ is the Caputo fractional derivative of order $\alpha$. $l(t)$ denotes the concentrations of infected cells in the latent stage at $t$. We avoid dimensional mismatches for the fractional order model, define the parameter values $\lambda = \lambda^1$, $\beta_3 = \beta_3^1$, $\beta_2 = \beta_2^1$, $a = a^1$, $k = k^1$, $m = m^1$, $\mu = \mu^1$, $\xi = \xi^1$, $g = g^1$, $h = h^1$, and $m = m^a$ be the death rate of $l(t)$ and latent infection become productively infected cells at the rate $\gamma = \gamma^1$. $\tau$ is delay which defines the time it takes for the newly produced virus to be mature and then infectious. $e^{-m^1\tau}$ is the survival probability of immature virions.

Definition 1 [15]. The Caputo derivative of fractional order $\alpha$ for a function $f(t)$ is described as

$$D^\alpha f(t) = \frac{1}{\Gamma(n-\alpha)} \int_0^t (t-\tau)^{n-\alpha-1} f^n(\tau)d\tau,$$

where $n-1 < \alpha < n \in \mathbb{Z}^+$ and $\Gamma(\cdot)$ is the gamma function.

The Laplace transform of Caputo derivative is described as

$$\mathcal{L}[D^\alpha f(t); s] = s^n \mathcal{F}(s) - \sum_{i=1}^{n-1} s^{n-i-1} f^{(i)}(0),$$

where $\mathcal{F}(s) = \mathcal{L}[f(t)]$. In particular, when $f^{(0)}(0) = 0$, $i = 1, 2, \ldots$, $n-1$, then $\mathcal{L}[D^\alpha f(t); s] = s^n \mathcal{F}(s)$.

Remark 1. The fractional derivative $\alpha \in (0, 1)$ is defined by Caputo sense [15], so introducing a convolution integral with a power-law memory kernel is a good way to describe memory effects in dynamical systems. The decaying rate of the memory kernel is determined by $\alpha$. A lower value of $\alpha$ corresponds to slower-decaying time-correlation functions which result in long-run memory.

Theorem 1. For any initial values $x(0) > 0$, $l(0) > 0$, $y(s) = \chi(s) > 0$, $v(0) > 0$, $w(0) > 0$, $s \in [-\tau, 0]$, the model (2) has a unique solution in $[0, +\infty)$. Moreover, this solution remains positive and bounded.

Proof 1. Assume that $\mathbb{R}_+^5 = \{(x, l, y, v, w) \in \mathbb{R}_+^5 : x \geq 0, l \geq 0, y \geq 0, v \geq 0, w \geq 0\}$ is positively invariant. The model (2) becomes

$$D^\alpha Z(t) = \mathcal{K}(Z(t)).$$

Here, $Z(t) = (x(t), l(t), y(t), v(t), w(t))^T$, and

$$\mathcal{K}(Z(t)) = \begin{bmatrix}
\lambda - dx(t) - \beta_1 x(t)v(t) + \beta_2 x(t)y(t) \\
(1 - \varphi)\left(\frac{\beta_1 x(t)v(t)}{1 + v(t)} + \frac{\beta_2 x(t)y(t)}{1 + y(t)}\right) - (m + \mu)l(t) \\
\varphi\left(\frac{\beta_1 x(t)v(t)}{1 + v(t)} + \frac{\beta_2 x(t)y(t)}{1 + y(t)}\right) + \gamma l(t) - ay(t) \\
ke^{-mt}y(t - \tau) - \mu v(t) - \xi v(t)w(t) \\
gv(t)w(t) - hw(t)
\end{bmatrix}.$$

$Z_0 = (x(0), l(0), y(s), v(0), w(0))^T \in \mathbb{R}_+^5$. For that, we analyze the direction of $\mathcal{K}(Z(t))$ on each coordinate space and see whether the vector field points to the interior of $\mathbb{R}_+^5$. From Eq. 2, we get

$$D^\alpha x(t)|_{v=0} = \lambda \geq 0,$$

$$D^\alpha l(t)|_{v=0} = (1 - \varphi)\left(\frac{\beta_1 x(t)v(t)}{1 + v(t)} + \frac{\beta_2 x(t)y(t)}{1 + y(t)}\right) \geq 0,$$

$$D^\alpha y(t)|_{v=0} = \varphi\beta_2 x(t)y(t) \geq 0,$$

$$D^\alpha v(t)|_{v=0} = ke^{-mt}y(t - \tau) \geq 0,$$

$$D^\alpha w(t)|_{v=0} = 0.$$
\[ \mathcal{L} = \begin{bmatrix} m + \gamma & 0 & 0 & 0 \\ -\gamma & a & 0 & 0 \\ 0 & -ke^{-\mu \tau} & \mu & 0 \\ 0 & 0 & 0 & h \end{bmatrix} \]

Based on the method of next-generation matrix \cite{32-34}, the threshold quantity \( R_0 \) is the spectral radius of

\[
\begin{bmatrix}
\psi, \beta(1 - \psi) (m + \gamma) \alpha & \beta(1 - \psi) (m + \gamma) \eta & \beta(1 - \psi) (m + \gamma) \\
\beta(1 - \psi) (m + \gamma) \alpha & \beta(1 - \psi) (m + \gamma) \eta & \beta(1 - \psi) (m + \gamma) \\
0 & 0 & 0 \\
0 & 0 & 0
\end{bmatrix}^{-1}
\]

Therefore,

\[
R_0 = \rho(\mathcal{W}^{-1}) = \left(\frac{\beta(1 - \psi)(m + \gamma)}{\alpha} + \frac{\beta(1 - \psi)(m + \gamma)}{\alpha} \right) \left( \frac{1 - \psi}{(m + \gamma)} \right).
\]

Note that \( R_0 \) is of dimension \([time]^{-1}\). Hence, it is not a dimensionless quantity as generally defined for the basic reproduction number in the integer-order setting. The above threshold quantity \( R_0 \) is memory-dependent (see \cite{32,33}).

### 3 Local Stability

In this section, we discussed the local stability and bifurcation results of equilibrium point by using the linearization technique and Jacobian matrix. The linearized system of Eq. 2 at any equilibrium point \( \mathcal{E}(x^*, l^*, y^*, \nu^*, w^*) \) is described as

\[ D^*x(t) = \left(-d - \frac{\beta_1 v^*}{1 + v^*} - \frac{\beta_2 y^*}{1 + y^*} \right) x(t) - \frac{\beta_3 x^*}{(1 + y^*)} y(t) - \frac{\beta_4 x^*}{(1 + v^*)} v(t), \]

\[ D^*l(t) = (1 - \psi) \left( \frac{\beta_5 v^*}{1 + v^*} + \frac{\beta_2 y^*}{1 + y^*} \right) x(t) + (1 - \psi) \left( \frac{\beta_7 x^*}{(1 + y^*)} \right) y(t) \]

\[ + (1 - \psi) \left( \frac{\beta_8 x^*}{(1 + v^*)} \right) v(t), \]

\[ D^*y(t) = \left( \frac{\beta_2 y^*}{1 + y^*} \right) x(t) + \frac{\beta_9 y^*}{1 + y^*} y(t) - \mu y(t), \]

\[ + \psi \left( \frac{\beta_1 y^*}{1 + v^*} \right) y(t) + \alpha y(t), \]

\[ D^*v(t) = ke^{-\mu \tau} y(t - \tau) - \mu v(t) - \xi \omega v(t) - \xi v^* w(t), \]

\[ D^*w(t) = gw^*v(t) + (g v^* - h) w(t). \]

Apply the Laplace transform of Eq. 4, so that \( X(s) = \mathcal{L}(x(t)), L(s) = \mathcal{L}(l(t)), Y(s) = \mathcal{L}(y(t)), V(s) = \mathcal{L}(v(t)), \) and \( W(s) = \mathcal{L}(w(t)) \). Eq. 4 is described as

\[
\begin{bmatrix}
s^2 + a_1 & s^2 + a_5 & a_3 + a_5 & 0 \\
0 & s^2 + a_5 & a_3 + a_5 & 0 \\
0 & 0 & s^2 + a_5 & a_3 + a_5 \\
0 & 0 & 0 & s^2 + a_5 + a_7 \end{bmatrix} \begin{bmatrix}
X(s) \\
L(s) \\
Y(s) \\
V(s) \end{bmatrix} = \begin{bmatrix}
d_1(s) \\
d_2(s) \\
d_3(s) \\
d_4(s) \end{bmatrix},
\]

where

\[
a_1 = d + \frac{\beta_1 v^*}{1 + v^*} + \frac{\beta_2 y^*}{1 + y^*} \quad a_2 = \frac{\beta_3 x^*}{(1 + y^*)} \quad a_3 = \frac{\beta_4 x^*}{(1 + v^*)},
\]

\[
a_4 = -(1 - \psi) \left( \frac{\beta_1 v^*}{1 + v^*} + \frac{\beta_2 y^*}{1 + y^*} \right),
\]

\[
a_5 = m + \gamma, a_6 = -(1 - \psi) \left( \frac{\beta_1 x^*}{(1 + y^*)} \right),
\]

\[
a_7 = -(1 - \psi) \left( \frac{\beta_1 v^*}{1 + v^*} + \frac{\beta_2 y^*}{1 + y^*} \right),
\]

\[
a_9 = -\gamma, a_{10} = a,
\]

\[
a_{11} = -\psi \left( \frac{\beta_1 v^*}{1 + v^*} \right), a_{12} = -\psi \left( \frac{\beta_1 v^*}{1 + v^*} \right), a_{13} = -ke^{-\mu \tau},
\]

\[
a_{14} = \mu + \xi \omega^*, a_{15} = \xi v^*, a_{16} = -gw^*, a_{17} = -(gw^* - h)
\]

and

\[
d_1(s) = s^{a_1} x(0), d_2(s) = s^{a_2} l(0), d_3(s) = s^{a_3} y(0), \]

\[
d_4(s) = s^{a_4} v(0) + a_{13} e^{-\mu \tau} \int_{-\tau}^{0} e^{-\mu \tau} (t) dt, \quad d_5(s) = s^{a_5} w(0). \]

The characteristic equation becomes

\[
D_1(s) + D_2(s)e^{-\mu \tau} = 0,
\]

where \( D_1(s) = (s^6) + \rho_1 (s^4) + \rho_2 (s^3)^2 + \rho_3 (s^2)^3 + \rho_4 s^3 + \rho_5 D_2(s) = \rho_6 (s^9) + \rho_7 (s^8)^2 + \rho_8 s^7 + \rho_9 \)

and

\[
d_6 = a_{14} a_{17} - a_{13} a_{16}, \rho_1 = a_1 + a_5 + a_9 + a_{14} + a_{17} + a_{15},
\]

\[
+ a_{12} (a_1 + a_5 + a_9 + a_{14} + a_{17}),
\]

\[
\rho_2 = a_1 a_2 a_3 + (a_1 + a_5 + a_9 + a_{14} + a_{17} + a_{13} + a_{15}),
\]

\[
+ a_{12} (a_1 + a_5 + a_9 + a_{14} + a_{17}) + a_{13} + a_{15},
\]

\[
+ a_3 a_5 + a_3 a_{15},
\]

\[
\rho_3 = a_1 a_2 a_3 + (a_1 + a_5 + a_{14} + a_{17}),
\]

\[
+ a_{12} (a_1 + a_5 + a_{14} + a_{17}),
\]

\[
\rho_4 = (a_{14} + a_{17}) a_1 a_2 a_{10} + d_6 (a_1 a_5 + a_1 a_{12}),
\]

\[
+ a_2 a_3 a_4 + a_2 a_4 + a_{12} (a_1 + a_4 + a_{17})),
\]

\[
+ a_{12} a_3 (a_1 + a_4 + a_{17} + a_{13} + a_{15} + a_{17}),
\]

\[
+ a_{12} (a_1 + a_5 + a_{14} + a_{17}),
\]

\[
\rho_5 = a_1 a_2 a_3 + a_2 a_3 + a_2 a_{15} + a_3 a_5 + a_3 a_{15} + a_3 a_{17} + a_3 a_{18},
\]

\[
\rho_6 = a_{12} a_{13},
\]

\[
\rho_7 = a_{12} (a_1 a_5 + a_5 + a_{12} (a_1 + a_5)) + a_2 a_5 a_{12}.
\]
Suppose the time-delay $\tau = 0$, then the characteristic equation (5) is

$$(s^n) + p_1(s^n) + (p_2 + p_3) (s^n) + (p_3 + p_4) (s^n) + (p_4 + p_5) s^n\quad \begin{cases} 
\rho_3 = a_3 a_4 a_5 (a_4 + a_5) - a_1 a_2 (a_1 a_2 + a_3), \\
\rho_4 = a_3 a_5 (a_4 - a_2) + a_4 a_5 (a_4 - a_1) \end{cases}
$$

Theorem 2. If $\mathcal{R}_0 < 1$, then the infection-free steady state $\mathcal{E}_0$ of (2) is locally asymptotically stable, and if $\mathcal{R}_0 > 1$, then $\mathcal{E}_0$ is unstable.

Proof 2. The characteristic equation of Eq. 2 at $\mathcal{E}_0$ is

$$(s^3 + m + \rho) (s^3 + \mu) = (s^3 + m + \rho) (s^3 + \mu) (s^3 + a) + k_3 x^* \left(1 - \phi\right) + \gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r = 0.$$ (6)

Since some coefficients of the above equations are delay-dependent, we utilize the geometric method, discussed in [35], to explore the possible stability switch in a general delay differential system with delay-stability dependent parameters. For the underlying model, we prove the stability of equilibrium points by direct comparison of the modulus of the underlying model, we prove the stability of equilibrium points to (2) by direct comparison of the modulus of (6), where $\mathcal{R}_0 > 1$.

Since we substitute $s = 1 + 1$, a solution is $E_0$, and then the final real part $x_1 < 0$. But taking contradiction, suppose that $x_1 \geq 0$.

Dividing by $(s^3 + m + \rho) (s^3 + \mu)$ on both sides of Eq. 6, we get

$$\begin{align*}
1 &= \frac{\beta_2 x^*}{(s^3 + m + \rho)} + \frac{\beta_2 x^*}{(s^3 + \mu)} + \frac{\gamma k_3 x^* (1 - \phi)}{(s^3 + m + \rho) (s^3 + \mu)} + \frac{\gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r}{(m + \gamma) a} \\
&\leq \frac{\beta_2 x^*}{a} + \frac{\beta_2 x^*}{m + \gamma} + \frac{\gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r}{(m + \gamma) a} + \frac{\gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r}{(m + \gamma) a} \\
&\leq \frac{\beta_2 x^*}{a d} + \frac{\beta_2 x^*}{m + \gamma} + \frac{\gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r}{(m + \gamma) a} + \frac{\gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r}{(m + \gamma) a} \\
&= \mathcal{R}_0.
\end{align*}$$

This leads to contradiction because $\mathcal{R}_0 < 1$. So, all the eigenvalues $s^3$ have negative real parts, and then $\mathcal{E}_0$ is locally stable if $\mathcal{R}_0 < 1$. If $\mathcal{R}_0 > 1$, then Eq. 6 is written as

$$H(s^3) = (s^n)^3 + h_1 s^n + h_2 s^n + h_3 = 0,$$

where

$$h_1 = m + \gamma + \mu + a - \beta_2 x^*,
$$

$$h_2 = \beta_2 x^* \left(1 - \phi\right) + (m + \gamma) a + \beta_2 x^*,
$$

$$h_3 = (m + \gamma) \mu - \beta_2 x^* \left(1 - \phi\right) + \gamma k_3 x^* (1 - \phi) e^{-(m+1)^3} r$$

Clearly, $H(0) = h_3 = \mu (m + \gamma) (1 - \mathcal{R}_0) < 0$ and $\lim s \to \cos H(s^3) = + \infty$. Thus, there exists at least one non-negative real root such that $H(s^n) = 0$, which implies that $\mathcal{E}_0$ is unstable if $\mathcal{R}_0 > 1$.

Next, we derive the local stability at the equilibrium state $\mathcal{E}_0$ of Eq. 2.
Then,
\[
\text{Re} \left( \frac{ds}{d\tau} \right) \Big|_{s = 0} = \frac{N_{11}N_{21} + N_{12}N_{22}}{N_{21} + N_{22}} \neq 0,
\]
where
\[
N_{11} = \gamma_0U_2 \sin \gamma_0t^* - \gamma_0V_2 \cos \gamma_0t^*, \quad N_{12} = \gamma_0V_2 \sin \gamma_0t^* + \gamma_0U_2 \cos \gamma_0t^*,
\]
\[
N_{21} = (U_j^* - \tau^*U_j^*) \cos \gamma_0t^* + U_j^* + (V_j^* - \tau^*V_j^*) \sin \gamma_0t^*,
\]
\[
N_{22} = (V_j^* - \tau^*V_j^*) \cos \gamma_0t^* + V_j^* - (U_j^* - \tau^*U_j^*) \sin \gamma_0t^*.
\]

We arrive at the following result.

Theorem 3. For the model (2), the following results hold:

i) If \( \tau = 0 \), then \( \mathcal{E}_2 \) is asymptotically stable if \( p_i > 0, \ i = 1, 2, \ldots, 5 \), and \( p_1p_2p_3 > \beta_3^2 + \beta_1^2 \beta_2^2 \gamma + \beta_1p_4p_5 - \beta_3^2 - p_1^2p_4 > p_5(p_1p_2 - p_3)^2 + p_1p_5^2 \).

ii) If \( \tau > 0 \), then \( \mathcal{E}_2 \) is asymptotically stable for \( 0 < \tau < \tau^* \) and the model (2) can undergo a Hopf bifurcation at \( \tau = \tau^* \).

The local asymptotic stability at the immune response–free equilibrium point \( \mathcal{E}_1 \) is similar to that at the endemic equilibrium point \( \mathcal{E}_2 \); hence, it is omitted.

### 4 Global Stability

Lemma 1 [36]. Let \( x(t) \in \mathbb{R}^+ \) be a continuous and derivative function. Then, for any time \( t \geq t_0 \),
\[
\epsilon_t D^\alpha_t \left[ x(t) - x^* - x^* \ln \frac{x(t)}{x^*} \right] \leq \left( 1 - \frac{x^*}{x(t)} \right) D^\alpha_t x(t),
\]
\( \forall a \in (0, 1), x^* \in \mathbb{R}^+ \).

Theorem 4. If \( x^* < \left( \frac{\epsilon_{\tau^*}}{1 - \tau^*} \right) \) and \( g < \frac{\epsilon_{\tau^*}}{k} \), then the model (2) is globally stable at \( \mathcal{E}_0 \).

Proof 3. We define the non-negative definite function at \( \mathcal{E}_0 \) of Eq. 2 as
\[
V_1(t) = \left( x(t) - x^* - x^* \ln \frac{x(t)}{x^*} \right) + l(t) + y(t) + \frac{\epsilon_{\tau^*}}{k} \nu(t) + w(t) + D_\tau^\alpha y(t - \sigma).
\]
Applying the fractional-order derivatives and using Lemma 1, we get
\[
D^\beta V_1(t) \leq \left( 1 - \frac{x^*}{x(t)} \right) \left( \lambda - dx(t) - \beta_1 x(t)v(t) + \beta_2 x(t)y(t) \right) + \left( 1 - \varphi \right) \left( \frac{\beta_2 x(t)v(t)}{1 + v(t)} + \frac{\beta_1 x(t)y(t)}{1 + y(t)} \right) - (m + y)l + \varphi \left( \frac{\beta_2 x(t)v(t)}{1 + v(t)} + \frac{\beta_1 x(t)y(t)}{1 + y(t)} \right) + yl - ay
\]
\[
+ \left( \frac{\epsilon_{\tau^*}}{k} \right) \left[ ke^{-\mu t} y(t - \tau) - \mu v - \xi uw \right] + gwv - hw
\]
\[
- D^\alpha_x \left( D^\beta_w y(t - \sigma) \right)
\]
\[
\leq - \frac{d}{x} (x - x^*)^2 + \beta_1 \nu x^* + \beta_2 y x^* - ay + y(t - \tau) - \frac{\mu v}{k}
\]
\[
- \frac{\xi uw}{k} + gwv - hw - y(t - \tau) + y(t)
\]
\[
\leq - \frac{d}{x} (x - x^*)^2 + \left( \beta_1 \nu x^* + \frac{\mu v}{k} \right) v + \left( \beta_1 x^* + 1 - a \right) y
\]
\[
+ \left( g - \frac{\epsilon_{\tau^*}}{k} \right) v.
\]

Based on the assumptions \( \lambda = dx^* x^* \left( \frac{\epsilon_{\tau^*}}{1 - \tau^*} \right) \) and \( g < \frac{\epsilon_{\tau^*}}{k} \), we can get \( D^\alpha V_1(t) < 0 \), and then the infection-free steady state \( \mathcal{E}_0 \) is globally asymptotically stable.

Theorem 5. Assume that \( g < \frac{\epsilon_{\tau^*}}{k} \), then the model (2) is globally stable at \( \mathcal{E}_2 \).

Proof 4. We define the non-negative definite function at \( \mathcal{E}_2 \) of Eq. 2 as
\[
V_2(t) = \left( x(t) - x^* - x^* \ln \frac{x(t)}{x^*} \right) + \left( l(t) - l^* - l^* \ln \frac{l(t)}{l^*} \right)
\]
\[
+ \left( y(t) - y^* - y^* \ln \frac{y(t)}{y^*} \right) + \frac{1}{2(m + d)x^*}
\]
\[
\times \left( (x - x^*)^2 + (l - l^*) + (y - y^*)^2 \right)
\]
\[
+ \left( \beta_1 x^* v^* + \beta_2 x^* y^* \right) e^{\epsilon_{\tau^*}} \left( \frac{v(t) - v^* - v^* \ln \frac{v(t)}{v^*}}{k} \right)
\]
\[
+ w(t) + (\beta_1 x^* v^* + \beta_2 x^* y^*)
\]
\[
\times \left( D^\alpha_x \left( \frac{y(t - \sigma)}{y^*} - 1 - \ln \frac{y(t - \sigma)}{y^*} \right) \right).
\]

Applying the fractional derivative on both sides, using Lemma 1, and assuming \( \lambda = dx^* x^* + \beta_1 x^* v^* + \beta_2 x^* y^*, m + y = \beta_1 x^* v^* + \beta_2 x^* y^*, a = \beta_1 \frac{\mu v}{k} + \beta_2 x^* v^*, \gamma = k y^* \frac{\epsilon_{\tau^*}}{k} \), we get
\[
D^\beta V(t) \leq \left( 1 - \frac{x^*}{x(t)} \right) \left( \lambda - dx(t) - \beta_1 x(t)v(t) + \beta_2 x(t)y(t) \right)
\]
\[
+ \left( 1 - \frac{l^*}{l(t)} \right) \left( 1 - \varphi \right) \left( \frac{\beta_2 x(t)v(t)}{1 + v(t)} + \frac{\beta_1 x(t)y(t)}{1 + y(t)} \right)
\]
\[
- (m + y)l + \varphi \left( \frac{\beta_2 x(t)v(t)}{1 + v(t)} + \frac{\beta_1 x(t)y(t)}{1 + y(t)} \right) + yl - ay
\]
\[
+ \frac{y}{2(m + d)x^*} (x - x^*)^2
\]
\[
+ (l - l^*) + (y - y^*) (\lambda - dx(t) - ay(t) - ml(t))
\]
\[
+ \left( \beta_1 x^* v^* + \beta_2 x^* y^* \right) e^{\epsilon_{\tau^*}} \left( 1 - \frac{v(t)}{v^*} \right)
\]
FIGURE 1 | Time trajectories for uninfected cells $x(t)$, latently infected cells $l(t)$, infected cells $y(t)$, viruses $v(t)$, and antibodies $w(t)$ of Eq. 2 with $\alpha = 1$ (left banner) and $\alpha = 0.9$ (right banner) and $R_0 = 0.93 < 1$.

FIGURE 2 | Time trajectories for uninfected cells $x(t)$, latently infected cells $l(t)$, infected cells $y(t)$, and viruses $v(t)$ of Eq. 2 with $\tau = 14 > \tau^* = 13.5$, $\alpha = 0.9$ (left banner) and $\alpha = 0.8$ (right banner).

FIGURE 3 | Time and space plane trajectories for uninfected cells $x(t)$, latently infected cells $l(t)$, infected cells $y(t)$, and viruses $v(t)$ of Eq. 2 with $\tau = 13 < \tau^* = 13.5$ and $\alpha = 0.9$. 
\( \alpha \leq \beta x^s v^s + \beta x^s y^s \)
\( (3 - \frac{x}{x^s} - y(t - t)^{v^s} - \frac{xy^s}{y} + \frac{y(t - t)^w}{y^s}) + (\beta x^s v^s + \beta x^s y^s) \)
\( (2 - \frac{1}{l^s} - \frac{y(t - t)^v}{y} - \frac{ly^s}{y^s} - \frac{y(t - t)^w}{y^s}) \)
\( -(\beta x^s v^s + \beta x^s y^s) \)
\( \frac{y(t - t)^v}{y^s} - \frac{y}{y^s} - \ln \frac{y(t - t)^v}{y^s} - \ln \frac{xy^s}{y^x y^y} \)
\( \ln \frac{1}{x} + \ln \frac{x^s}{x} + \ln \frac{l^w}{l^s} \)

\( \leq -\left( \frac{d}{x} + \frac{ym}{(m + d)x^s} \right)(x - x^s)^2 - \frac{ya}{(m + d)x^s}(l - l^w)^2 \)
\( -\frac{ym}{(m + d)x^s}(y - y^s)^2 \)

Clearly, by using Lyapunov–LaSalle’s invariant principle and [37], \( D^\alpha V(t) \leq 0 \), the equilibrium point \( E_2 \) is globally stable.
5 NUMERICAL SIMULATION

The numerical results of fractional-order delay differential system (2) are discussed, using implicit Euler’s scheme in [38]. The parameter values are as follows: $\lambda = 8, d = 1, \beta_1 = 5, \beta_2 = 5, \varphi = 0.5, m = 2, \gamma = 100, a = 10, k = 10, \mu = 10, \xi = 1, g = 1, \beta = 1, \varepsilon = 1, \beta = 1,$ and $h = 1$

**Figure 1** shows the stable behaviors of the model (2) which converge to the infection-free steady state $E_0$ and $R_0 = 0.93 < 1$, with different fractional orders $\alpha = 1, 0.9$ and initial conditions $x(t) = 2.3, l(t) = 2.3, y(t) = 2.3, v(t) = 2.3$, and $w(t) = 2.3$. The immune response–free steady state $E_1$ is unstable when $\tau = 14 > \tau^* = 13.5$ and fractional orders $\alpha = 0.9, 0.8$, which is displayed in **Figure 2**, while the immune response–free steady state $E_1$ is stable when $\tau = 13 < \tau^* = 13.5$, which is shown in **Figure 3**. The fractional derivatives are a useful tool for reducing the infection.

**Figure 4** displays the time trajectories of the considered model (2), which converge to the endemic equilibrium point $E_2$, and stable behavior, when $\tau = 13 < \tau^* = 13.5$. By increasing the fractional-order values, the curves get more flat and have significant effects on the cell dynamics.

Remark 2. These are the results from the fractional-order viral infection model with latent infection and delay, which are different, complement, and expand upon those in [7,9,12]. The fractional-order stability regions in **Figures 1–4** are larger than the corresponding integer-order stability regions in [7,12]. In a fractional-order model, the unstable equilibrium of an integer-order model may be a stable model. Furthermore, time-delays enhance the dynamics of the model and increase its complexity.

6 CONCLUDING REMARKS

Although the classical integer-order differential models can be useful for the study of disease dynamics, the fractional-order models are more useful for exploring disease dynamics. Several factors contribute to this, including data fitting, memory effects, and crossover effects. The fractional calculus is a powerful tool to describe physical and biological systems that have long-term memory and long-range spatial interactions. This paper examines the global dynamics of a fractional-order viral infection model with latent infection. We studied the positivity and boundedness of the solutions. Based on this formula, we derived the basic reproductive number $R_0$, which acts as a threshold parameter in the viral infection disease status. For such a model, we examined the existence of equilibrium points and their corresponding asymptotic stability. Hopf bifurcation occurs at the critical values of the time-delay $\tau^*$. The Lyapunov–LaSalle method is used to determine the global stability of the model. A numerical solution for the proposed fractional-order viral infection model is obtained by employing Euler’s implicit scheme. Fractional order and time-delays increase the complexity and enrich the dynamics of the model.

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/Supplementary Materials, and further inquiries can be directed to the corresponding author.

AUTHOR CONTRIBUTIONS

FR was involved in project conceptualization and supervision, writing the original draft, and reviewing and editing the paper. CR was responsible for data visualization and validation, running the software, and performing the methodology.

FUNDING

This research was funded by UAE University, fund # 12S005-UPAR 2020.

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