Research Article

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Asymptotic analysis of hepatitis B epidemic model using Caputo Fabrizio fractional operator

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Abstract: A mathematical model representing the temporal dynamics of hepatitis B virus (HBV) is discussed in this research work. This is based on the asymptomatic carriers and symptomatic individuals keeping in view the characteristics of the disease. We also incorporate the vaccination parameter to vaccinate susceptible individuals. Moreover, we use fractional calculus to extend the model to its associated fractional-order. For this, we particularly use the fractional operator of the Caputo-Fabrizio type to fractionalize the proposed model. First, the model formulation has been derived in classic order and then extended to its associated fractional-order version for generalization. The model equilibria was calculated, and the basic reproductive number was found. Then we will discuss the existence with properties of the uniqueness of the proposed fractional version of the model that is under consideration. The positivity with boundedness is shown to investigate that the considered model is feasible biologically as well as mathematically. Finally, we use the Mittag-Leffler approach to visualize the model of fractional-order and to support the results carried out in the theocratical part. We also demonstrate the solution curves for different values of the fractional parameter to differentiate between integer-order and fractional-order on the disease transmission.

Keywords: epidemic model, hepatitis B virus, CF-fractional derivative, asymptomatic and symptomatic, numerical simulation

1 Research background

Hepatitis B is a contagious disease among other infectious diseases. This causes inflammation of the liver. Around 350 million individuals globally have chronic HBV, with 25–40 percent of these cases leading to chronic liver disease, hepatocellular carcinoma, and cirrhosis [1,2]. As a result, HBV infection has become a huge public health issue all over the world. A patient with a chronic stage has a serum burden [3]. If we assume the average mass of the liver to be 1.5 kg, a human liver should have the same number of cells. Due to the large numbers, we must use the conventional incidence rate rather than the mass action incidence rate.

Epidemiological models yield some intriguing mathematical insights, particularly in the prevention and management of infectious diseases, see, for instance, [4-9]. Varied models of infectious diseases with various properties and characteristics have been produced and studied in the literature. Different elements (such as temperature, precipitation, absolute humidity, and others) have a major impact on the force of infection in human viral infections. Various deterministic and stochastic models have been proposed by different authors and investigated the dynamics of infectious diseases with some real applications (see, for instance, [10-16]). However, models with fractional-order are proved to be useful for describing the complicated and cross behavior because the range of derivative order is optional in this case [17-22]. The information about the spectrum between the integer orders has been provided via fractional-order calculus [23]. Fractional calculus is more realistic from integer order while producing more accurate outputs [24-29]. Therefore, problems with real-world situations can be studied with derivative of fractional-order, with a greater degree of freedom. Instead of classical cases nowadays, researchers are very much interested in fractional-order

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derivatives while applying them to real-world problems. With this, many analyses has been made, *i.e.*, the existence theory, uniqueness theory, approximating solutions, and stabilities of the fractional equation. Physical problems in various forms have been formulated using differential, functional, and integral equations with fractional-order, such as a logistic model, model of microorganism, dingy problem, and Lotka-Volterra equation. These equations are foundation models in the field of formulating models in case of biological situations [31–35]. Further various concepts are examined for FODEs using analytical, semi-analytical, and approximation analysis. Many more methods like Euler, Taylor, Adams-Bashforth, predictor-corrector, etc., are the transforms with many more other well-known techniques [36,37]. Various fractional derivatives have been applied in different shapes of biological models to show the temporal dynamics of distinct infectious diseases (see, for more detail, [38-42]).

In the present article, we analyze and formulate a novel model to represent the temporal dynamics of HBV using the CF fractional operator based on the asymptomatic, symptomatic, and vaccination assumptions from the findings of the previous studies. All of the models in the literature have distinct perspectives for studying HBV dynamics, but we use asymptomatic and symptomatic carriers with vaccination of susceptible individuals to examine the dynamics of the disease more thoroughly. HBV-infected people can live for 30-180 days without showing symptoms and can infect others, which can lead to mortality in humans [43]. As a result, the goal of the study is to forecast and develop a novel model with asymptomatic, symptomatic carries, and vaccination parameter for immunization. The model is established first with integer order and then extending to the generalized form of fractional derivatives keeping in view the CF derivative operator. We calculate the basic reproductive number and examine the equilibria of the classical model. Once the reported model fractionalizes, we will then study the existence analysis with unique properties. Finally, we perform some numerical simulations and show a graphical representation of the analytical results.

We organized the reported studies as the formulation of the novel HBV model, and primary definitions of the CF derivative are presented in Section 2. In Section 3, we discuss the analysis of existence with unique properties of a positive solution. In Section 4, some numerical visualization has been provided and also compared the results to show for different fractional values. We end our work with the conclusion in Section 5.

2 HBV transmission models

We consider a model to describe the time dynamics of hepatitis B virus transmission by following the work presented in ref. [44]. For this purpose, we will introduce some notations in the form of state variables and parameters. Particularly, we assume that W(t), X(t), Y(t), and Z(t), respectively, denote the vulnerable population, the fraction of population with HBV positive, the portion of the population with positive HBV chronic, and recovered individuals. Further, the vulnerable are those individuals who at any time may catch the disease. Likewise, we divide the total population as symbolized by N(t) in various compartments of W, X, Y, and Z. Obviously, the entire compartments are dependent on time, which can be written as N(t) = Z(t) + Y(t) + X(t) + W(t), and so the model looks like

$$\frac{d\mathbf{W}(\mathbf{t})}{dt} = \Lambda - \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_0 + \nu)\mathbf{W}(\mathbf{t}) - \delta \Lambda \mathbf{Y}(\mathbf{t}),$$

$$\frac{d\mathbf{X}(\mathbf{t})}{dt} = \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_1 + r_1 + d_0)\mathbf{X}(\mathbf{t}),$$

$$\frac{d\mathbf{Y}(\mathbf{t})}{dt} = r_1\mathbf{X}(\mathbf{t}) - (d + \kappa + d_0 - \delta \Lambda)\mathbf{Y}(\mathbf{t}),$$

$$\frac{d\mathbf{Z}(\mathbf{t})}{dt} = d_1\mathbf{X}(\mathbf{t}) - d_0\mathbf{Z}(\mathbf{t}) + \nu\mathbf{W}(\mathbf{t}) + \kappa\mathbf{Y}(\mathbf{t}),$$
(1)

where the description of the parameters is as Λ is the rate of birth and ν is assumed to be the rate of vaccination. The rate of offspring prenatally infected portion is denoted by δ , while d_2 , d_0 , and d_1 are reduced rate of disease transmission, natural, and disease death rates respectively. Further, β and d_1 are the transmission co-efficient and recovery rate, but κ and r_1 symbolize the transfer rate of acute to chronic population.

To shorten our calculation, we assume that $y_1 = v + d_0$, $y_1 = d_1 + r_1 + d_0$, and $y_1 = \kappa + d_1 + d - \Lambda \eta$, then using the theory of the dynamical system, we always admits a disease-free point for our proposed model as symbolized by $\mathbf{E_0}$, which looks like

$$\mathbf{E_0} = \left(\frac{\Lambda}{\gamma_1}, 0, 0, \frac{\nu\Lambda}{d_0\gamma_1}\right). \tag{2}$$

We use the disease-free state and applying the next-generation approach to calculate the reproductive number. Let us assume that this quantity is denoted by $\mathbf{R_0^D}$, then

$$\mathbf{R_0^D} = \frac{\beta \Lambda}{\gamma_1 \gamma_2} + \frac{\mathbf{d}_2 \gamma_2 \Lambda \beta}{\gamma_1 \gamma_2 \gamma_3}.$$
 (3)

Further, the dynamical properties of the reported model (1) has been described as follows:

- The equilibrium point E_0 is table if $R_0^D < 1$, but if $\mathbf{R_0^D} > 1$, then the model is unstable.
- If $\mathbf{R_0^D} > 1$, the unique endemic state $\mathbf{E}_* = (\mathbf{W}_*, \mathbf{X}_*, \mathbf{Y}_*, \mathbf{Z}_*)$ is stable both locally and globally, implies that the infection of HBV persists.

We will now study the fractional-order version of the afore-reported model, so first we present some preliminary knowing of CF fractional derivatives to derive acquainted with customary notations [7,23].

Definition 1. Let $\phi \in H^1(0, T)$, if $n - 1 < \theta < n, \theta > 0$ and $n \in \mathbb{N}$. The derivative in sense of Caputo as well as in Caputo-Fabrizio operators are defined by

$${}^{\mathrm{CF}}D_{0,t}^{\theta}\{\varphi(t)\} = \frac{K(\theta)}{(1-\theta)} \int_{0}^{t} \varphi'(z) \exp\left(\frac{(z-t)\theta}{1-\theta}\right) \mathrm{d}z \quad (4)$$

and

$${}^{C}D_{0,t}^{\vartheta}\{\varphi(t)\} = \frac{1}{\Gamma(-\vartheta+n)} \int_{0}^{t} (t-z)^{-1+n-\vartheta} \varphi^{n}(z) \mathrm{d}z, \quad (5)$$

where CF and C symbolizes the Caputo-Fabrizio and Caputo, respectively.

Definition 2. Let $0 < \theta < 1$, then the integral

$${}^{\mathrm{RL}}J_{0,t}^{\theta}\{\varphi(t)\} = \frac{1}{\Gamma(\theta)} \int_{0}^{t} (t-z)^{\theta-1} \varphi(z) \mathrm{d}z \tag{6}$$

is the Riemann-Liouville and

$$^{\text{CF}}J_{0,t}^{\theta}\{\varphi(t)\} = \frac{2}{(2-\theta)K(\theta)} \times \left\{ (1-\theta)\varphi(t) + \theta \int_{0}^{t} \varphi(z) dz \right\}$$
(7)

is the Caputo-Fabrizio-Caputo (CF) integral operators.

We now incorporate the aforementioned fractional operator into model (1) to study the associated fraction version, which looks like

$$^{\mathrm{CF}}D_{0,t}^{9}\mathbf{W}(\mathbf{t}) = \Lambda - \frac{\beta\mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_{2}\mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_{0} + \nu)\mathbf{W}(\mathbf{t}) - \delta\Lambda\mathbf{Y}(\mathbf{t}),$$

$$^{\mathrm{CF}}D_{0,t}^{9}\mathbf{X}(\mathbf{t}) = \frac{\beta\mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_{2}\mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_{1} + r_{1} + d_{0})\mathbf{X}(\mathbf{t}),$$

$$^{\mathrm{CF}}D_{0,t}^{9}\mathbf{Y}(\mathbf{t}) = r_{1}\mathbf{X}(\mathbf{t}) - (d + \kappa + d_{0} - \delta\Lambda)\mathbf{Y}(\mathbf{t}),$$

$$^{\mathrm{CF}}D_{0,t}^{9}\mathbf{Z}(\mathbf{t}) = d_{1}\mathbf{X}(\mathbf{t}) - d_{0}\mathbf{Z}(\mathbf{t}) + \nu\mathbf{W}(\mathbf{t}) + \kappa\mathbf{Y}(\mathbf{t}).$$
(8)

Existence and uniqueness

The proposed fractional-order system will be transformed to its associated integral equation system to show the existence analysis. We will apply the fixed-point theorem and show existence uniqueness. Thus taking the transformation of the aforementioned system leads to the integral system as given by

$$\mathbf{W}(\mathbf{t}) = {^{\mathrm{CF}}J_{0,t}^{\theta}} \left\{ \Lambda - \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_0 + \nu)\mathbf{W}(\mathbf{t}) - \delta \Lambda \mathbf{Y}(\mathbf{t}) \right\} + \mathbf{W}(\mathbf{0}),$$

$$\mathbf{X}(\mathbf{t}) = \mathbf{X}(\mathbf{0}) + {^{\mathrm{CFC}}J_{0,t}^{\theta}} \left\{ \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_1 + r_1 + d_0)\mathbf{X}(\mathbf{t}) \right\},$$

$$\mathbf{Y}(\mathbf{t}) = {^{\mathrm{CFC}}J_{0,t}^{\theta}} \left\{ r_1 \mathbf{X}(\mathbf{t}) - (d + \kappa + d_0 - \delta \Lambda)\mathbf{Y}(\mathbf{t}) \right\} + \mathbf{Y}(\mathbf{0}),$$

$$\mathbf{Z}(\mathbf{t}) = \mathbf{Z}(\mathbf{0}) + {^{\mathrm{CFC}}J_{0,t}^{\theta}} \left\{ d_1 \mathbf{X}(\mathbf{t}) - d_0 \mathbf{Z}(\mathbf{t}) + \nu \mathbf{W}(\mathbf{t}) + \kappa \mathbf{Y}(\mathbf{t}) \right\}.$$

To simplify our calculations, we assume $Y = K(\theta)(2 - \theta)$ for simplicity, and by applying integration in CF sense, we obtain the following equivalent system:

$$\begin{aligned} \mathbf{W}(\mathbf{t}) &= \mathbf{W}(\mathbf{0}) + \frac{2(1-\vartheta)}{\Upsilon} \left\{ \Lambda - \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} \right. \\ &- (d_0 + \nu)\mathbf{W}(\mathbf{t}) - \delta \Lambda \mathbf{Y}(\mathbf{t}) \right\} \\ &+ 2\vartheta \int_0^t \left\{ \Lambda - \frac{\beta \mathbf{W}(\mathbf{x})\mathbf{X}(\mathbf{x})}{\mathbf{N}} - \frac{\beta d_2 \mathbf{W}(\mathbf{x})\mathbf{Y}(\mathbf{x})}{\mathbf{N}} \right. \\ &- (d_0 + \nu)\mathbf{W}(\mathbf{x}) - \delta \Lambda \mathbf{Y}(\mathbf{x}) \right\} dx \left(\frac{1}{(2-\vartheta)K(\vartheta)} \right), \end{aligned}$$

(7)
$$\mathbf{X}(\mathbf{t}) = \frac{2(1-\vartheta)}{K(\vartheta)} \left\{ \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_1 + r_1 + d_0)\mathbf{X}(\mathbf{t}) \right\} \frac{1}{(2-\vartheta)} + \mathbf{X}(\mathbf{0})$$

$$+ \frac{1}{K(\vartheta)} \int_0^t \left\{ \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} \right\}$$

$$- (d_1 + r_1 + d_0)\mathbf{X}(\mathbf{t}) \right\} dx \left\{ \frac{2\vartheta}{(2-\vartheta)} \right\},$$

$$\mathbf{Y}(\mathbf{t}) = \frac{(1-\vartheta)}{K(\vartheta)} \left\{ r_1 \mathbf{X}(\mathbf{t}) - (d + \kappa + d_0 - \delta \Lambda)\mathbf{Y}(\mathbf{t}) \right\} \frac{2}{(2-\vartheta)} + \mathbf{Y}(\mathbf{0})$$

$$+ \frac{2\vartheta}{(2-\vartheta)K(\vartheta)} \int_0^t \left\{ r_1 \mathbf{X}(\mathbf{x}) - (d + \kappa + d_0 - \delta \Lambda)\mathbf{Y}(\mathbf{x}) \right\} dx,$$

$$\mathbf{Z}(\mathbf{t}) = \mathbf{Z}(\mathbf{0}) + \frac{2(1-\vartheta)}{\Upsilon} \left\{ d_1 \mathbf{X}(\mathbf{t}) - d_0 \mathbf{Z}(\mathbf{t}) + \nu \mathbf{W}(\mathbf{t}) + \kappa \mathbf{Y}(\mathbf{t}) \right\}$$

$$+ \frac{1}{(2-\vartheta)} \int_0^t \left\{ d_1 \mathbf{X}(\mathbf{x}) - d_0 \mathbf{Z}(\mathbf{x}) + \nu \mathbf{W}(\mathbf{x}) + \kappa \mathbf{Y}(\mathbf{x}) \right\} dx \left(\frac{2\vartheta}{K(\vartheta)} \right).$$

Let l_1 , l_2 , l_3 , and l_4 represents the kernels, then

$$l_{1}(\mathbf{W}(\mathbf{t}), t) = \left\{ \Lambda - \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_{2}\mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_{0} + \nu)\mathbf{W}(\mathbf{t}) - \delta \Lambda \mathbf{Y}(\mathbf{t}) \right\},$$

$$l_{2}(\mathbf{X}(\mathbf{t}), t) = \left\{ \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_{2}\mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_{1} + r_{1} + d_{0})\mathbf{X}(\mathbf{t}) \right\},$$

$$l_{3}(\mathbf{Y}(\mathbf{t}), t) = \left\{ r_{1}\mathbf{X}(\mathbf{t}) - (d + \kappa + d_{0} - \delta \Lambda)\mathbf{Y}(\mathbf{t}) \right\},$$

$$l_{4}(\mathbf{Z}(\mathbf{t}), t) = \left\{ d_{1}\mathbf{X}(\mathbf{t}) - d_{0}\mathbf{Z}(\mathbf{t}) + \nu \mathbf{W}(\mathbf{t}) + \kappa \mathbf{Y}(\mathbf{t}) \right\}.$$
(9)

Theorem 1. Lipschitz conditions hold for the aforementioned kernels l_i , i - 1, 2, 3, 4.

Proof. We assume **W** and **W**₁, **X**, and **X**₁, **Y** and **Y**₁, **Z**, and **Z**₁ are the functions of l_i ; thus, we have

$$\begin{split} l_{l}(\textbf{W}(\textbf{t}),\,t) &= l_{l}(\textbf{W}_{l}(\textbf{t}),\,t) + \left\{ \Lambda - \frac{\beta(\textbf{W}(\textbf{t}) - \textbf{W}_{l}(\textbf{t}))\textbf{X}(\textbf{t})}{\textbf{N}} \right. \\ &- \frac{\beta d_{2}(\textbf{W}(\textbf{t}) - \textbf{W}_{l}(\textbf{t}))\textbf{Y}(\textbf{t})}{\textbf{N}} \\ &- (d_{0} + \nu)(\textbf{W}(\textbf{t}) - \textbf{W}_{l}(\textbf{t})) - \delta \Lambda \textbf{Y}(\textbf{t}) \right\}, \\ l_{l}(\textbf{X}(\textbf{t}),\,t) &= l_{l}(\textbf{X}_{l}(\textbf{t}),\,t) + \left\{ \frac{\beta \textbf{W}(\textbf{t})(\textbf{X}(\textbf{t}) - \textbf{X}_{l}(\textbf{t}))}{\textbf{N}} + \frac{\beta d_{2}\textbf{W}(\textbf{t})\textbf{Y}(\textbf{t})}{\textbf{N}} \right. \\ &- (d_{1} + r_{1} + d_{0})(\textbf{X}(\textbf{t}) - \textbf{X}_{l}(\textbf{t})) \right\}, \\ l_{l}(\textbf{Y}(\textbf{t}),\,t) &= l_{l}(\textbf{Y}_{l}(\textbf{t}),\,t) + \{r_{l}\textbf{X}(\textbf{t}) - (d + \kappa + d_{0} - \delta \Lambda)(\textbf{Y}(\textbf{t}) - \textbf{Y}_{l}(\textbf{t}))\}, \\ l_{l}(\textbf{Z}(\textbf{t}),\,t) &= l_{l}(\textbf{Z}_{l}(\textbf{t}),\,t) + \{d_{l}\textbf{X}(\textbf{t}) - d_{0}(\textbf{Z}(\textbf{t}) - \textbf{Z}_{l}(\textbf{t})) + \nu \textbf{W}(\textbf{t}) + \kappa \textbf{Y}(\textbf{t})\}. \end{split}$$

We apply the Cauchy's inequality implies that

$$\|l_1(\mathbf{W}(\mathbf{t}), t) - l_1(\mathbf{W}_1(\mathbf{t}), t)\|$$

$$\leq \left\| \left\{ \Lambda - \frac{\beta(\mathbf{W}(\mathbf{t}) - \mathbf{W}_{\mathbf{I}}(\mathbf{t}))\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_2(\mathbf{W}(\mathbf{t}) - \mathbf{W}_{\mathbf{I}}(\mathbf{t}))\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_0 + \nu)(\mathbf{W}(\mathbf{t}) - \mathbf{W}_{\mathbf{I}}(\mathbf{t})) - \delta \Lambda \mathbf{Y}(\mathbf{t}) \right\} \right\|,$$

$$\|l_1(\mathbf{X}(\mathbf{t}), t) - l_1(\mathbf{X}_{\mathbf{I}}(\mathbf{t}), t)\|$$

$$\leq \left\| \left\{ \frac{\beta \mathbf{W}(\mathbf{t})(\mathbf{X}(\mathbf{t}) - \mathbf{X}_{\mathbf{I}}(\mathbf{t}))}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_1 + r_1 + d_0)(\mathbf{X}(\mathbf{t}) - \mathbf{X}_{\mathbf{I}}(\mathbf{t})) \right\} \right\|,$$

$$||l_{1}(\mathbf{Y}(\mathbf{t}), t) - l_{1}(\mathbf{Y}_{1}(\mathbf{t}), t)||$$

$$\leq ||\{r_{1}\mathbf{X}(\mathbf{t}) - (d + \kappa + d_{0} - \delta\Lambda)(\mathbf{Y}(\mathbf{t}) - \mathbf{Y}_{1}(\mathbf{t}))\}||,$$

$$||l_{1}(\mathbf{Z}(\mathbf{t}), t) - l_{1}(\mathbf{Z}_{1}(\mathbf{t}), t)||$$

$$\leq \|\{d_1\mathbf{X}(t)-d_0(\mathbf{Z}(t)-\mathbf{Z_1}(t))+\nu\mathbf{W}(t)+\kappa\mathbf{Y}(t)\}\|.$$

We then obtain the relations given by

$$\mathbf{W}(\mathbf{t}) = \frac{2(1 - \theta)l_{1}(\mathbf{W}_{n-1}(t), t)}{Y} + \frac{2\theta}{Y} \int_{0}^{t} l_{1}(\mathbf{W}_{n-1}(x), x) dx,$$

$$\mathbf{X}(\mathbf{t}) = \frac{2(1 - \theta)l_{2}(\mathbf{X}_{n-1}(t), t)}{Y} + \frac{2\theta}{Y} \int_{0}^{t} l_{2}(\mathbf{X}_{n-1}(x), x) dx,$$

$$\mathbf{Y}(\mathbf{t}) = \frac{2(1 - \theta)l_{3}(\mathbf{Y}_{n-1}(t), t)}{(2 - \theta)K(\theta)} + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{3}(\mathbf{Y}_{n-1}(x), x) dx,$$

$$\mathbf{Z}(\mathbf{t}) = \frac{2(1 - \theta)l_{4}(\mathbf{Z}_{n-1}(t), t)}{(2 - \theta)K(\theta)} + \frac{2\theta}{K(\theta)(2 - \theta)} \int_{0}^{t} l_{4}(\mathbf{Z}_{n-1}(x), x) dx.$$
(10)

Taking the difference between successive terms and applying norm as well as majorizing, we then lead to the assertions

$$\|\mathbf{A}_{n}(t)\| = \|\mathbf{W}_{n}(t) - \mathbf{W}_{1,n-1}(t)\|$$

$$\leq \frac{2(1-\vartheta)}{Y} \|l_{1}(\mathbf{W}_{n-1}(t), t) - l_{1}(\mathbf{W}_{1,n-2}(t), t)\|$$

$$+ \frac{2\vartheta}{K(\vartheta)(2-\vartheta)} \left\| \int_{0}^{t} \left[l_{1}(\mathbf{W}_{n-1}(x), x) - l_{1}(\mathbf{W}_{1,n-2}(x), x) \right] dx \right\|,$$

$$\|\mathbf{B}_{n}(t)\| = \|\mathbf{X}_{n}(t) - \mathbf{X}_{1,n-1}(t)\|$$

$$\leq \|l_{2}(\mathbf{X}_{n-1}(t), t) - l_{2}(\mathbf{X}_{1,n-2}(t), t)\| \frac{2(1-\vartheta)}{K(\vartheta)(2-\vartheta)}$$

$$+ \frac{2\vartheta}{K(\vartheta)(2-\vartheta)} \left\| \int_{0}^{t} \left[l_{2}(\mathbf{X}_{n-1}(x), x) - l_{2}(\mathbf{X}_{1,n-2}(x), x) \right] dx \right\|,$$

$$\|\mathbf{C}_{n}(t)\| = \|\mathbf{Y}_{n}(t) - \mathbf{Y}_{1,n-1}(t)\|$$

$$\leq \frac{2(1-\vartheta)}{K(\vartheta)(3-\vartheta)} \|l_{3}(\mathbf{Y}_{n-1}(t), t) - l_{1}(\mathbf{Y}_{1,n-2}(t), t)\|$$

$$+ \frac{2\vartheta}{K(\vartheta)(2-\vartheta)} \left\| \int_{0}^{t} \left[l_{3}(\mathbf{Y}_{n-1}(x), x) - l_{3}(\mathbf{Y}_{1,n-2}(x), x) \right] dx \right\|,$$

$$\|\mathbf{D}_{n}(t)\| = \|\mathbf{Z}_{n}(t) - \mathbf{Z}_{1,n-1}(t)\|$$

$$\leq \|l_{4}(\mathbf{Z}_{n-1}(t), t) - l_{4}(\mathbf{Z}_{1,n-2}(t), t)\| \frac{2(1-\vartheta)}{K(\vartheta)(2-\vartheta)}$$

$$+ \frac{2\vartheta}{Y} \left\| \int_{0}^{t} \left[l_{4}(\mathbf{Z}_{n-1}(x), x) - l_{4}(\mathbf{Z}_{1,n-2}(x), x) \right] dx \right\|.$$

In the aforementioned system

$$\sum_{i=0}^{\infty} \mathbf{A}_{i}(t) = \mathbf{W}_{n}(t), \quad \sum_{i=0}^{\infty} \mathbf{B}_{i}(t) = \mathbf{X}_{n}(t),$$

$$\sum_{i=0}^{\infty} \mathbf{C}_{i}(t) = \mathbf{Y}_{n}(t), \quad \sum_{i=0}^{\infty} \mathbf{D}_{i}(t) = \mathbf{Z}_{n}(t).$$
(12)

Since, the kernels l_i satisfy the conditions of Lipschitz, so

$$\|\mathbf{A}_{n}(t)\| = \|\mathbf{W}_{n}(t) - \mathbf{W}_{1,n-1}(t)\|$$

$$\leq \frac{2(1-\vartheta)}{\Upsilon} \tau_{1} \|\mathbf{W}_{n-1}(t) - \mathbf{W}_{1,n-2}(t)\|$$

$$+ \frac{2\vartheta}{(2-\vartheta)K(\vartheta)} \tau_{2} \int_{0}^{t} \|\mathbf{W}_{n-1}(x) - \mathbf{W}_{1,n-2}(x)\| dx,$$

$$\|\mathbf{B}_{n}(t)\| = \|\mathbf{X}_{n}(t) - \mathbf{X}_{1,n-1}(t)\|$$

$$\leq \tau_{3} \|\mathbf{X}_{n-1}(t) - \mathbf{X}_{1,n-2}(t)\| \frac{2(1-\vartheta)}{\Upsilon}$$

$$+ \frac{2\vartheta}{\Upsilon} \tau_{4} \int_{0}^{t} \|\mathbf{X}_{n-1}(x) - \mathbf{X}_{1,n-2}(x)\| dx,$$

$$\|\mathbf{C}_{n}(t)\| = \|\mathbf{Y}_{n}(t) - \mathbf{Y}_{1,n-1}(t)\|$$

$$\leq \tau_{5} \|\mathbf{Y}_{n-1}(t) - \mathbf{Y}_{1,n-2}(t)\| \frac{2(1-\vartheta)}{\Upsilon}$$

$$+ \frac{2\vartheta}{(2-\vartheta)K(\vartheta)} \tau_{6} \int_{0}^{t} \|\mathbf{Y}_{n-1}(x) - \mathbf{Y}_{1,n-2}(x)\| dx,$$

$$\|\mathbf{D}_{n}(t)\| = \|\mathbf{Z}_{n}(t) - \mathbf{Z}_{1,n-1}(t)\|$$

$$\leq \tau_{7} \|\mathbf{Z}_{n-1}(t) - \mathbf{Z}_{1,n-2}(t)\| \frac{(1-\vartheta)(2)}{(2-\vartheta)K(\vartheta)}$$

$$+ \tau_{8} \int_{0}^{t} \|\mathbf{Z}_{n-1}(x) - \mathbf{Z}_{1,n-2}(x)\| dx \left(\frac{2\vartheta}{(2-\vartheta)K(\vartheta)}\right). \square$$

Theorem 2. Model (8) poses a solution.

Proof. Using Eq. (12) with the recursive scheme leads to the assertion as follows:

$$\|\mathbf{A}_{n}(t)\| \leq \|\mathbf{W}(0)\| + \left\{ \left(\frac{2(1-\theta)\tau_{1}}{Y} \right)^{n} \right\} + \left\{ \left(\frac{2t\tau_{2}\theta}{Y} \right)^{n} \right\},$$

$$\|\mathbf{B}_{n}(t)\| \leq \|\mathbf{X}(0)\| + \left\{ \left(\frac{2\tau_{3}(1-\theta)}{Y} \right)^{n} \right\}$$

$$+ \left\{ \left(\frac{2\tau_{4}t\theta}{(2-\theta)K(\theta)} \right)^{n} \right\},$$

$$\|\mathbf{C}_{n}(t)\| \leq \|\mathbf{Y}(0)\| + \left\{ \left(\frac{2\tau_{5}(1-\theta)}{\Upsilon} \right)^{n} \right\}$$

$$+ \left\{ \left(\frac{2\tau_{6}\theta t}{K(\theta)(2-\theta)} \right)^{n} \right\},$$

$$\|\mathbf{D}_{n}(t)\| \leq \|\mathbf{Z}(0)\| + \left\{ \left(\frac{2\tau_{7}(1-\theta)}{(2-\theta)K(\theta)} \right)^{n} \right\}$$

$$+ \left\{ \left(\frac{2\tau_{8}t\theta}{K(\theta)(2-\theta)} \right)^{n} \right\}.$$
(14)

Furthermore, let us assume that $Y_{1,i}(t)$, i = 1,..., 5 represent the remainder terms, and then to prove that Eq. (14) satisfies the model (8), we make the substitution as follows:

$$\mathbf{W}(t) = \mathbf{W}_{n}(t) + Y_{1,n}(t),$$

$$\mathbf{X}(t) = \mathbf{X}_{n}(t) + Y_{2,n}(t), \quad \mathbf{Y}(t) + Y_{3,n}(t) = \mathbf{Y}_{n}(t), \quad (15)$$

$$\mathbf{Z}(t) = +Y_{4,n}(t) + \mathbf{Z}_{n}(t).$$

Thus,

$$\mathbf{W}(t) - \mathbf{W}_{n-1}(t) = \frac{2(1 - \theta)l_{1}(\mathbf{W}(t) - Y_{1,n}(t))}{K(\theta)(2 - \theta)} + \frac{2\theta}{K(\theta)(2 - \theta)} \int_{0}^{t} l_{1}(\mathbf{W}(x)) + \frac{2\theta}{K(\theta)(2 - \theta)} \int_{0}^{t} l_{1}(\mathbf{W}(x)) + \frac{2\theta}{K(\theta)(2 - \theta)} + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{2}(\mathbf{X}(x)) + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{2}(\mathbf{X}(x)) + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{3}(\mathbf{Y}(x)) + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{3}(\mathbf{Y}(x)) + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{3}(\mathbf{Y}(x)) + \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{4}(\mathbf{Z}(x)) + \frac{2\theta}{(2 -$$

The application of norm with Lipschitz conditions may lead to

$$\left\| \mathbf{W}(t) - \frac{2l_{1}(\mathbf{W}(t), t)(1 - \theta)}{K(\theta)} \frac{1}{(2 - \theta)} - \mathbf{W}(0) \right\|$$

$$- \frac{2\theta}{Y} \int_{0}^{t} l_{1}(\mathbf{W}(x), x) dx \left\|$$

$$\leq \|Y_{1,n}(t)\| \left\{ 1 + \left(\frac{2(1 - \theta)\tau_{1}}{Y} + \frac{2\theta\tau_{2}t}{Y} \right) \right\},$$

$$\left\| \mathbf{X}(t) - \frac{2l_{2}(\mathbf{X}(t), t)(1 - \theta)}{Y} - \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{2}(\mathbf{X}(x), x) dx \right\| - \mathbf{X}(0)$$

$$\leq \|Y_{2,n}(t)\| \left\{ 1 + \left(\frac{2(1 - \theta)\tau_{3}}{Y} + \frac{2\theta\tau_{4}t}{Y} \right) \right\},$$

$$\left\| \mathbf{Y}(t) - \frac{2l_{3}(\mathbf{Y}(t), t)(1 - \theta)}{Y} - \mathbf{Y}(0) - \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{3}(\mathbf{Y}(x), x) dx \right\|$$

$$\leq \|Y_{3,n}(t)\| \left\{ 1 + \left(\frac{2\tau_{5}(1 - \theta)}{Y} + \frac{2\theta\tau_{6}t}{Y} \right) \right\},$$

$$\left\| \mathbf{Z}(t) - \frac{2l_{4}(\mathbf{Z}(t), t)(1 - \theta)}{Y} - \frac{2\theta}{(2 - \theta)K(\theta)} \int_{0}^{t} l_{4}(\mathbf{Z}(x), x) dx \right\| - \mathbf{Z}(0)$$

$$\leq \|Y_{4,n}(t)\| \left\{ 1 + \left(\frac{2(1 - \theta)\tau_{7}}{Y} + \frac{2\theta\tau_{8}t}{Y} \right) \right\}.$$

Taking \lim as t approaches ∞ , we reach to

$$\mathbf{W}(t) = \frac{2l_1(\mathbf{W}(t), t)(1 - \theta)}{\Upsilon} + \int_0^t l_1(\mathbf{W}(x), x) dx \left(\frac{2\theta}{\Upsilon}\right) + \mathbf{W}(0),$$

$$\mathbf{X}(t) = \frac{2l_2(\mathbf{X}(t), t)(1 - \theta)}{\Upsilon} + \int_0^t l_2(\mathbf{X}(x), x) dx \left(\frac{2\theta}{\Upsilon}\right) + \mathbf{X}(0),$$

$$\mathbf{Y}(t) = \frac{2l_3(\mathbf{Y}(t), t)(1 - \theta)}{\Upsilon} + \frac{2\theta}{\Upsilon} \int_0^t l_3(\mathbf{Y}(x), x) dx + \mathbf{Y}(0),$$

$$\mathbf{Z}(t) = \frac{2(1 - \theta)l_4(\mathbf{Z}(t), t)}{\Upsilon} + \frac{2\theta}{\Upsilon} \int_0^t l_4(\mathbf{Z}(x), x) dx + \mathbf{Z}(0).$$
(18)

This completes the conclusion.

Theorem 3. The proposed model as reported by system (8) exists a unique solution.

Proof. Let $(\mathbf{W}^+(t), \mathbf{X}^+(t), \mathbf{Y}^+(t), \mathbf{Z}^+(t))$ is also the solution, then

$$\mathbf{W}(t) = \mathbf{W}^{+}(t) + \frac{\{l_{1}(\mathbf{W}(t), t) - l_{1}(\mathbf{W}^{+}(t), t)\}2(1 - \theta)}{\Upsilon}$$

$$+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \{l_{1}(\mathbf{W}(y), y) - l_{1}(\mathbf{W}^{+}(y), y)\}dx,$$

$$\mathbf{X}(t) = \mathbf{X}^{+}(t) + \frac{\{l_{2}(\mathbf{X}(t), t) - l_{2}(\mathbf{X}^{+}(t), t)\}2(1 - \theta)}{\Upsilon}$$

$$+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \{l_{2}(\mathbf{X}(y), y) - l_{2}(\mathbf{X}^{+}(x), x)\}dx,$$

$$\mathbf{Y}(t) = \mathbf{Y}^{+}(t) + \frac{\{l_{3}(\mathbf{Y}(t), t) - l_{3}(\mathbf{Y}^{+}(t), t)\}2(1 - \theta)}{\Upsilon}$$

$$+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \{l_{3}(\mathbf{Y}(x), x) - l_{3}(\mathbf{Y}^{+}(x), x)\}dx,$$

$$\mathbf{Z}(t) = \mathbf{Z}^{+}(t) + \frac{\{l_{4}(\mathbf{Z}(t), t) - l_{4}(\mathbf{Z}^{+}(t), t)\}2(1 - \theta)}{\Upsilon}$$

$$+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \{l_{4}(\mathbf{Z}(x), x) - l_{4}(\mathbf{Z}^{+}(x), x)\}dx.$$

The application of majorizing gives

$$\begin{aligned} \|\mathbf{W}(t) - \mathbf{W}^{+}(t)\| &= \frac{\|l_{1}(\mathbf{W}(t), t) - l_{1}(\mathbf{W}^{+}(t), t)\|2(1 - \theta)}{\Upsilon} \\ &+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \|l_{1}(\mathbf{W}(x), x) - l_{1}(\mathbf{W}^{+}(x), x)\| dx, \\ \|\mathbf{X}(t) - \mathbf{X}^{+}(t)\| &= \frac{\|l_{2}(\mathbf{X}(t), t) - l_{2}(\mathbf{X}^{+}(t), t)\|2(1 - \theta)}{\Upsilon} \\ &+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \|l_{2}(\mathbf{X}(x), x) - l_{2}(\mathbf{X}^{+}(x), x)\| dx, \\ \|\mathbf{Y}(t) - \mathbf{Y}^{+}(t)\| &= \frac{\|l_{3}(\mathbf{Y}(t), t) - l_{3}(\mathbf{Y}^{+}(t), t)\|2(1 - \theta)}{\Upsilon} \\ &+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \|l_{3}(\mathbf{Y}(x), x) - l_{3}(\mathbf{Y}^{+}(x), x)\| dx, \end{aligned}$$

$$\|\mathbf{Z}(t) - \mathbf{Z}^{+}(t)\| &= \frac{\|l_{4}(\mathbf{Z}(t), t) - l_{4}(\mathbf{Z}^{+}(t), t)\|2(1 - \theta)}{\Upsilon} \\ &+ \frac{2\theta}{\Upsilon} \int_{0}^{t} \|l_{4}(\mathbf{Z}(x), x) - l_{4}(\mathbf{Z}^{+}(x), x)\| dx. \end{aligned}$$

By using Theorems 1 and 2, we may lead to the assertion as described by

$$\|\mathbf{W}(t) - \mathbf{W}^{+}(t)\| \leq \frac{2\tau_{1}\psi_{1}(1-\theta)}{\Upsilon} + \left(\frac{2\theta\tau_{2}t\phi_{2}}{\Upsilon}\right)^{n},$$

$$\|\mathbf{X}(t) - \mathbf{X}^{+}(t)\| \leq \frac{2\tau_{3}(1-\theta)\psi_{3}}{\Upsilon} + \left(\frac{2\tau_{4}\theta\phi_{4}t}{\Upsilon}\right)^{n},$$

$$\|\mathbf{Y}(t) - \mathbf{Y}^{+}(t)\| \leq \frac{2\tau_{5}\psi_{5}(1-\theta)}{\Upsilon} + \left(\frac{2\theta\tau_{6}\phi_{6}t}{\Upsilon}\right)^{n},$$

$$\|\mathbf{Z}(t) - \mathbf{Z}^{+}(t)\| \leq \frac{2\tau_{7}\psi_{7}(1-\theta)}{\Upsilon} + \left(\frac{2\theta\tau_{8}\phi_{8}t}{\Upsilon}\right)^{n}.$$
(21)

This implies that the aforementioned formulas hold for every n, so we obtain

$$\mathbf{W}(t) = \mathbf{W}^{+}(t), \ \mathbf{X}(t) = \mathbf{X}^{+}(t), \ \mathbf{Y}(t) = \mathbf{Y}^{+}(t),$$
$$\mathbf{Z}(t) = \mathbf{Z}^{+}(t).$$

Further, we will discuss that the model under consideration is feasible by Lemma 1.

Lemma 1. Since $(\mathbf{W}(\mathbf{t}), \mathbf{X}(\mathbf{t}), \mathbf{Y}(\mathbf{t}), \mathbf{Z}(\mathbf{t}))$ is the solution of system (8). Let us assume that the initial population sizes are nonnegative, then $(\mathbf{W}(\mathbf{t}), \mathbf{X}(\mathbf{t}), \mathbf{Y}(\mathbf{t}), \mathbf{Z}(\mathbf{t}))$ is also nonnegative for every $t \geq 0$.

Proof. Since ϑ is the fractional parameter, then

$${}^{G}D_{0,t}^{g}(\mathbf{W}(\mathbf{t})) = \Lambda - \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_{2}\mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}}$$

$$- (d_{0} + \nu)\mathbf{W}(\mathbf{t}) - \delta\Lambda\mathbf{Y}(\mathbf{t}),$$

$${}^{G}D_{0,t}^{g}(\mathbf{X}(\mathbf{t})) = \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_{2}\mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}}$$

$$- (d_{1} + r_{1} + d_{0})\mathbf{X}(\mathbf{t}),$$

$${}^{G}D_{0,t}^{g}(\mathbf{Y}(\mathbf{t})) = r_{1}\mathbf{X}(\mathbf{t}) - (d + \kappa + d_{0} - \delta\Lambda)\mathbf{Y}(\mathbf{t}),$$

$${}^{G}D_{0,t}^{g}(\mathbf{Z}(\mathbf{t})) = d_{1}\mathbf{X}(\mathbf{t}) - d_{0}\mathbf{Z}(\mathbf{t}) + \nu\mathbf{W}(\mathbf{t}) + \kappa\mathbf{Y}(\mathbf{t}),$$

$$(23)$$

where G is a general fractional operator. Thus, we have

$$\begin{cases}
GD_{0,t}^{\theta}(\mathbf{W}(\mathbf{t}))|_{\kappa(\mathbf{W}(\mathbf{t}))} = \Lambda > 0, \\
GD_{0,t}^{\theta}(\mathbf{X}(\mathbf{t}))|_{\kappa(\mathbf{X}(\mathbf{t}))} = \frac{\beta \mathbf{W}(\mathbf{t})\mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t})\mathbf{Y}(\mathbf{t})}{\mathbf{N}} \ge 0 \\
GD_{0,t}^{\theta}(\mathbf{Y}(\mathbf{t}))|_{\kappa(\mathbf{Y}(\mathbf{t}))} = r_1\mathbf{X}(\mathbf{t}) \ge 0, \\
GD_{0,t}^{\theta}(\mathbf{Z}(\mathbf{t}))|_{\kappa(\mathbf{Z}(\mathbf{t}))} = d_1\mathbf{X}(\mathbf{t}) + \nu \mathbf{W}(\mathbf{t}) + \kappa \mathbf{Y}(\mathbf{t}) \ge 0,
\end{cases}$$
(24)

where $\kappa(\xi) = \{\xi = 0 \text{ and } \mathbf{W}(\mathbf{t}), \mathbf{X}(\mathbf{t}), \mathbf{Y}(\mathbf{t}), \mathbf{Z}(\mathbf{t}) \text{ is in } C(R_+ \times R_+)\}$ and $\xi \in \{\mathbf{W}(\mathbf{t}), \mathbf{X}(\mathbf{t}), \mathbf{Y}(\mathbf{t}), \mathbf{Z}(\mathbf{t})\}$. We then reach to our conclusion by using the same methodology as adopted in [30], that the solutions of the considered model are nonnegative for all $t \ge 0$.

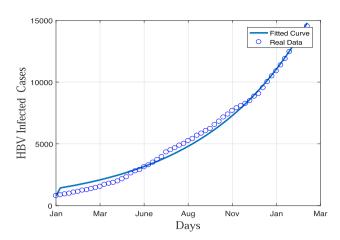


Figure 1: The graph visualizes the fitting of model parameters to real data for the reported HBV model.

Table 1: The estimated value of epidemic parameters

Parameter	Description	Value	Source
٨	Birth rate	260,479	Assumed
β	Transmission coefficient	14.24	Fitted
d_2	Reduced transmission coefficient	7.02529	Fitted
V	Vaccination rate	5.8480	Fitted
r_1	Portion of acute leads to chronic	5.5650	Fitted
d_1	Recovery rate of acute	0.97	Fitted
d_0	Natural death	0.0005	Fitted
d_1	Death from disease	0.001253133	Fitted
δ	Offspring prenatally infected portion	0.0004	Fitted
κ	Rate of recovery of chronic	0.00001	Fitted

Table 2: Parameters values used in simulating model 8

Parameter	Value	Source
۸	1.5	Assumed
β	0.5	Assumed
d_2	0.03	[44]
v	0.07	Assumed
r_1	0.01	Assumed
d_1	0.9	Assumed
d_0	0.005	Assumed
d_1	0.001253133	Fitted
δ	0.02	[44]
К	0.00001	Assumed
W	110	[44]
X	80	[44]
Υ	30	[44]
Z	10	[44]

4 Parameters of estimation and numerical simulation

In this section, we will present some numerical simulations to show the significance of our model. We will first estimate the model parameters and then use the estimated value to present the long-run numerical simulation.

4.1 Estimation of parameters

We use real data to estimate our model parameters that are important to show the validation of the reported model. We used a number of infected cases as reported in district Swat Khyber Pakhtunkhwa Pakistan from January 1, 2020, to December 28, 2020. To estimate the value of epidemic parameters, the method of parameter estimation technique under lsqcurvefit routine via MATLAB software has been used, while to measure the goodness of fit, the associated

average relative error of formula
$$\frac{1}{12}\sum_{k=1}^{12}\left|\frac{x_k^{\rm real}-x_k^{\rm approximate}}{x_k^{\rm real}}\right|\approx$$

 1.5685×10^{-1} is used. This verifies the suitability of small relative error's value (1.5685×10^{-1}). Once to perform the estimation analysis as presented in Figure 1, we estimate the epidemic parameters presented in Table 1.

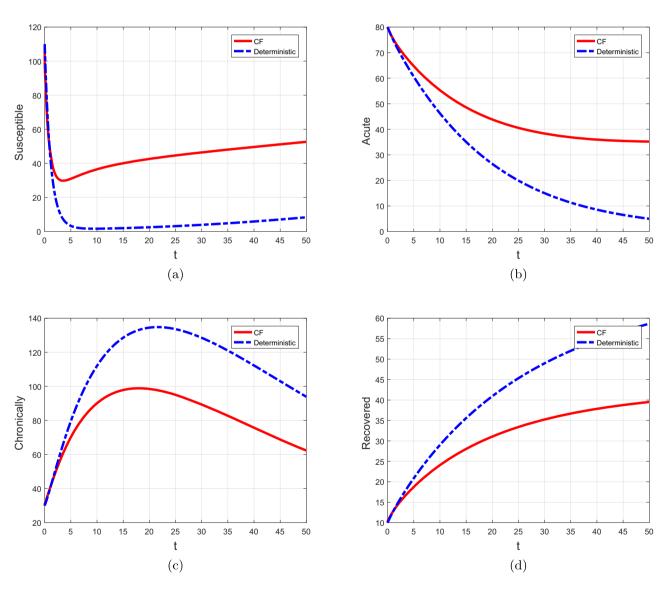


Figure 2: The visualization of the dynamics of system (8), when $\theta = 0.90$. (a) Susceptible individuals, (b) acute individuals, (c) chronically individuals, and (d) recovered individuals.

4.2 Numerical simulation

We will present some graphic descriptions of the analytical results to present the justification as well as the authenticity of the analytical work. We present the simulation of the reported fractional HBV epidemic model (8). For this, we assume that the units of time is 50, while the value of parameters is taken from Table 2. Moreover, we also use various values of fractional-order parameters to show the impact of fractional-order on disease transmission. We further assume that time step integration is $h = 10^{-3}$ and [0, t] is the interval of time. Moreover, let $n = \frac{T}{h}, n \in \mathbb{N}$, and u = 0, 1, 2, ..., n, then the discretization of the proposed model looks like:

$$\begin{split} ^{\text{CF}}\mathbf{W}_{u+1} &= \mathbf{W}(0) + (1-\vartheta) \bigg\{ \Lambda - \frac{\beta \mathbf{W}(\mathbf{t}) \mathbf{X}(\mathbf{t})}{\mathbf{N}} \\ &- \frac{\beta d_2 \mathbf{W}(\mathbf{t}) \mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_0 + \nu) \mathbf{W}(\mathbf{t}) - \delta \Lambda \mathbf{Y}(\mathbf{t}) \bigg\} \\ &+ \vartheta h \sum_{k=0}^{u} \bigg\{ \Lambda - \frac{\beta \mathbf{W}(\mathbf{t}) \mathbf{X}(\mathbf{t})}{\mathbf{N}} - \frac{\beta d_2 \mathbf{W}(\mathbf{t}) \mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_0 + \nu) \mathbf{W}(\mathbf{t}) - \delta \Lambda \mathbf{Y}(\mathbf{t}) \bigg\}, \\ ^{\text{CF}}\mathbf{X}_{u+1} &= \mathbf{X}(0) + (1-\vartheta) \bigg\{ \frac{\beta \mathbf{W}(\mathbf{t}) \mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t}) \mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_1 + r_1 + d_0) \mathbf{X}(\mathbf{t}) \bigg\} \\ &+ \vartheta h \sum_{k=0}^{u} \bigg\{ \frac{\beta \mathbf{W}(\mathbf{t}) \mathbf{X}(\mathbf{t})}{\mathbf{N}} + \frac{\beta d_2 \mathbf{W}(\mathbf{t}) \mathbf{Y}(\mathbf{t})}{\mathbf{N}} - (d_1 + r_1 + d_0) \mathbf{X}(\mathbf{t}) \bigg\}, \\ ^{\text{CF}}\mathbf{Y}_{u+1} &= \mathbf{Y}(0) + (1-\vartheta) \{r_1 \mathbf{X}(\mathbf{t}) - (d + \kappa + d_0 - \delta \Lambda) \mathbf{Y}(\mathbf{t})\} \\ &+ \vartheta h \sum_{k=0}^{u} (1-p) \{r_1 \mathbf{X}(\mathbf{t}) - (d + \kappa + d_0 - \delta \Lambda) \mathbf{Y}(\mathbf{t})\}, \\ ^{\text{CF}}\mathbf{Z}_{u+1} &= (1-\vartheta) \{d_1 \mathbf{X}(\mathbf{t}) - d_0 \mathbf{Z}(\mathbf{t}) + \nu \mathbf{W}(\mathbf{t}) + \kappa \mathbf{Y}(\mathbf{t})\} \\ &+ \vartheta h \sum_{k=0}^{u} \{d_1 \mathbf{X}(\mathbf{t}) - d_0 \mathbf{Z}(\mathbf{t}) + \nu \mathbf{W}(\mathbf{t}) + \kappa \mathbf{Y}(\mathbf{t})\} + \mathbf{Z}(0). \end{split}$$

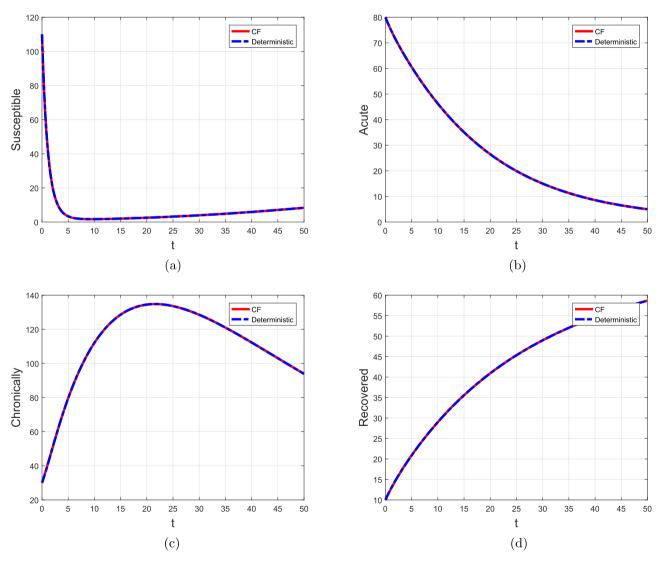


Figure 3: The visualization of the dynamics of system (8), when $\theta = 1.0$. (a) Susceptible individuals, (b) acute individuals, (c) chronically individuals, and (d) recovered individuals.

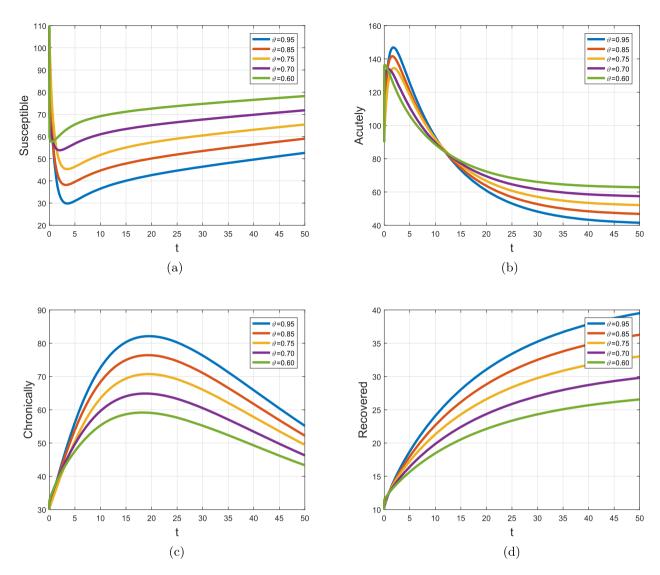


Figure 4: The visualization of the dynamics of system (8), when $\theta = 0.95$, 0.85, 0.75, 0.70, 0.60. (a) Susceptible individuals, (b) acute individuals, (c) chronically individuals, and (d) recovered individuals.

The collective execration parameters values taken from Table 2 with the aforementioned algorithm leads to the graphical representation in Figure 2. To show the difference between the classical and fractional-order, we present the simulation for $\theta=1$ and $\theta=0.95$ (see Figure 2), which clearly shows the significance of the fractional-order. Moreover, we represent the dynamics of the different compartmental of the proposed model for $\theta=1.01$ to show the effect of fractional on disease dynamics as described in Figure 3. Likewise to perform the numerical simulation for the various compartments of reported model for various values fractional-order, we present the simulation carried out in Figure 4. Particularly, Figure 4(a) demonstrates the time variation of

vulnerable individuals, which describe that decrease, or increases in the value of ϑ cause increase or decrease in the size of vulnerable population. This verify that there is an inverse relation between the time dynamics of vulnerable population and fractional parameter. Similarly, the time dynamics of acute and chronic is given in Figure 4(b) and (c), while the dynamics of recovered is described in Figure 4(d), which, respectively, indicates the relation between compartmental sizes and fractional parameter. We also demonstrated the phase portraits of system (8) to show the relative impact of compartmental population on each other as shown in Figure 5. There is inverse relation of ϑ with acute and chronic sizes while there is direct relation with recovered compartment. So the numerical

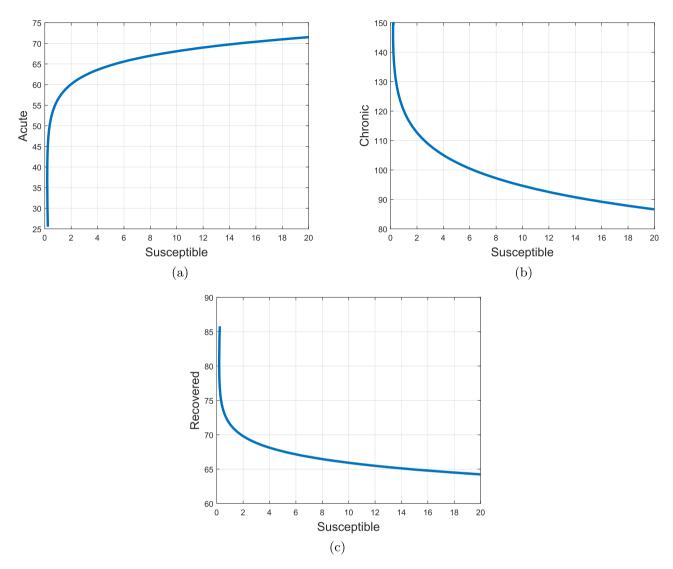


Figure 5: The graphs represent the phase portrait diagram for susceptible vs other compartments of the proposed system. (a) Phase portrait of the acute vs susceptible population, (b) phase portrait of the susceptible vs chronic carries population, and (c) phase portrait of the recovered vs susceptible population.

visualization analyzes that there is a strong dependency of fractional parameter on the compartmental population and disease transmission, which are clearly visible.

5 Conclusion

We formulated a novel model for the dynamics of HBV transmission with the asymptomatic carriers and symptomatic individuals as well as vaccination of susceptible. By formulating the model, then we studied the equilibria and calculated the basic reproductive number. Further, we used the fractional calculus and extended the model

to its associated fractional-order. For this, we used the operator in the sense of CF derivative. Moreover, the fixed-point theorem has been utilized and showed the existence analysis with unique properties. Also, the biological feasibility has been discussed. We also used real data of hepatitis B to parameterize the proposed model and estimate the value of epidemic parameters. Based on estimated parameters values, all the theocratical results and the impact of fractional-order on disease transmission have been presented with the help of some graphical visualization. The numerical results are produced for various values of the fractional parameters. It has been also observed that vaccination has a positive and effective role in eradicating the infection of hepatitis B.

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References

- Zhuo X. Global analysis of a general HBV infection model. 2011 IEEE International Conference on Systems Biology (ISB). IEEE; 2011.
- [2] Nowak M, May RM. Virus dynamics: mathematical principles of immunology and virology: mathematical principles of immunology and virology. UK: Oxford University Press; 2000.
- Wodarz D, May RM, Nowak MA. The role of antigen-independent persistence of memory cytotoxic T lymphocytes. Int Immunol. 2000;12(4):467-77.
- Nowak A, Bonhoeffer S, Hill AM, Boehme R, Thomas HC, McDade H. Viral dynamics in hepatitis B virus infection. Proc National Acad Sci. 1996;939:4398-402. 2022, 30(1), Article ID 2240026, 12 pages. doi: 10.1142/ S0218348X22400266.
- [5] Din A, Li Y, Yusuf A. Delayed hepatitis B epidemic model with stochastic analysis. Chaos Soliton Fractal. 2021;146:110839.
- Ullah S, Khan MA, Farooq M. A new fractional model for the dynamics of the hepatitis B virus using the Caputo-Fabrizio derivative. Europ Phys J Plus. 2018;133(6):1-14.
- Rashid S, Sultana S, Karaca Y, Khalid A, Chu YM. Some further extensions considering discrete proportional fractional operators. Fractals. 2022;30:2240026.
- [8] Hajiseyedazizi SN, Samei ME, Alzabut J, Chu Y-M. On multistep methods for singular fractional q-integro-differential equations. Open Math. 2021;19:1378-405. doi: 10.1515/math-
- [9] Khan T, Ullah Z, Ali N, Zaman G. Modeling and control of the hepatitis B virus spreading using an epidemic model. Chaos Solitons Fractals. 2019;124:1-9.

- [10] Khan T, Zaman G, Alshomrani AS. Spreading dynamic of acute and carrier hepatitis B with nonlinear incidence. PloS One. 2018;13(4):e0191914.
- [11] He Z-Y, Abbes A, Jahanshahi H, Alotaibi ND, Wang Y. Fractional-order discrete-time SIR epidemic model with vaccination: Chaos and complexity. Mathematics. 2022;10(2):165, 18 pages, doi: 10.3390/math10020165.
- [12] Khan T, Zaman G, Chohan MI. The transmission dynamic of different hepatitis B-infected individuals with the effect of hospitalization. J Biol Dynam. 2018;12(1):611-31.
- [13] Khan T, Jung IH, Khan A, Zaman G. Classification and sensitivity analysis of the transmission dynamic of hepatitis B. Theoret Biol Med Modell. 2017;14(1):1-17.
- [14] Moonsamy S, Suchard M, Pillay P, Prabdial-Sing N. Prevalence and incidence rates of laboratory-confirmed hepatitis B infection in South Africa, 2015 to 2019. BMC Public Health 2022;22(1):1-13.
- [15] Jajarmi A, Baleanu D, ZarghamiVahid K, Mobayen S. A general fractional formulation and tracking control for immunogenic tumor dynamics. Math Meth Appl Sci. 2022;45(2):667-80.
- [16] Baleanu D, Zibaei S, Namjoo M, Jajarmi A. A nonstandard finite difference scheme for the modeling and nonidentical synchronization of a novel fractional chaotic system. Adv Differ Equ. 2021;2021(1):1-19.
- [17] Arif MS, Raza A, Rafiq M, Bibi M. A reliable numerical analysis for stochastic hepatitis B virus epidemic model with the migration effect. Iranian J Sci Technol Trans A Sci. 2019;43(5):2477-92.
- [18] Lu Q. Stability of SIRS system with random perturbations. Phys A Statist Mechanic Appl. 2009;388(18):3677-86.
- [19] Din A, Li Y. Stationary distribution extinction and optimal control for the stochastic hepatitis B epidemic model with partial immunity. Phys Scr. 2021;96(7):074005.
- [20] Zhang X-B, Wang X-D, Huo H-F. Extinction and stationary distribution of a stochastic SIRS epidemic model with standard incidence rate and partial immunity. Physica A Statist Mechanic Appl. 2019;531:121548.
- [21] Jin F, Qian Z-S, Chu Y-M, ur Rahman M. On nonlinear evolution model for drinking behavior under Caputo-Fabrizio derivative. 2022;12(2):790-806. doi: 10.11948/20210357.
- [22] Baleanu D, Alsaedi A. Analysis of time-fractional Hunter-Saxton equation: a model of neumatic liquid crystal. Open Phys. 2016;14(1):145-9.
- [23] Baleanu D, Machado JA, Luo AC, eds. Fractional dynamics and control. New York Dordrecht Heidelberg London: Springer Science and Business Media; 2011.
- [24] Baleanu D, Hassan Abadi M, Jajarmi A, Zarghami Vahid K, Nieto JJ. A new comparative study on the general fractional model of COVID-19 with isolation and quarantine effects. Alexandria Eng J. 2022;61(6):4779-91.
- [25] Jajarmi A, Baleanu D, Zarghami Vahid K, Mohammadi Pirouz H, Asad JH. A new and general fractional Lagrangian approach: A capacitor microphone case study. Results Phys. 2021;31:104950.
- [26] Erturk VS, Godwe E, Baleanu D, Kumar P, Asad J, Jajarmi A. Novel Fractional-Order Lagrangian to Describe Motion of Beam on Nanowire. Acta Physica Polonica A. 2021;140(3):265-72.
- [27] Jajarmi A, Baleanu D, Zarghami Vahid K, Mobayen S. A general fractional formulation and tracking control for immunogenic tumor dynamics. Math Meth Appl Sci. 2022;45(2):667-80.

- [28] Kumar P, Baleanu D, SuatErturk V, Inc M, Govindaraj V. A delayed plant disease model with Caputo fractional derivatives. Adv Contin Discrete Models 2022;2022(1):1-22.
- [29] Khan FM, Ullah Khan Z, Lv Y-P, Yusuf A, Din A. Investigating of fractional order dengue epidemic model with ABC operator. Results Phys. 2021;24:104075.
- [30] Khan T, Qian Z-S, Ullah R, AlAlwan B, Zaman G, Al-Mdallal QM, et al. The transmission dynamics of hepatitis B virus via the fractional-order epidemiological model. Complexity 2021;2021:1-18.
- [31] Chen SB, Soradi-Zeid S, Jahanshahi H, Alcaraz R, Gómez-Aguilar JF, Bekiros S, et al. Optimal control of time-delay fractional equations via a joint application of radial basis functions and collocation method. Entropy. 2020;22(11):1213.
- [32] Rashid S, Sultana S, Karaca Y, Khalid A, Chu YM. Some further extensions considering discrete proportional fractional operators. Fractals. 2021;30:2240026.
- [33] Rashid S, Abouelmagd EI, Khalid A, Farooq FB, Chu Y-M. Some recent developments on dynamical discrete fractional type inequalities in the frame of nonsingular and nonlocal kernels. Fractals. 2022;30(2):2240110, 15 pages. doi: 10.1142/ S0218348X22401107.
- [34] Wang F-Z, Khan MN, Ahmad I, Ahmad H, Abu-Zinadah H, Chu Y-M. Numerical solution of traveling waves in chemical kinetics: time- fractional fishers equations. Fractals. 2022;30(2):22400051, 11.
- [35] Baleanu D, Diethelm K, Scalas E Trujillo JJ. Fractional calculus: models and numerical methods. Vol. 3. London: World Scientific; 2012.

- [36] Baleanu D, Atangana A. New fractional derivatives with nonlocal and non-singular kernel: theory and application to heat transfer model. 2016. arXiv: http://arXiv.org/abs/arXiv:1602. 03408.
- [37] Atangana A, Koca I. Chaos in a simple nonlinear system with Atangana Baleanu derivatives with fractional order. Chaos Solitons Fractals. 2016;89:447-54.
- [38] Zarin R, Ahmed I, Kumam P, Zeb A, Din A. Fractional modeling and optimal control analysis of rabies virus under the convex incidence rate. Results Phys. 2021;28:104665.
- Khan MA, Atangana A. Modeling the dynamics of novel coronavirus (2019-nCov) with fractional derivative. Alexandria Eng. J. 2020;59(4):2379-89.
- [40] Baleanu D, Orkun YC, Kurt A. New analytical solutions for conformable fractional PDEs arising in mathematical physics by exp-function method. Open Phys. 2017;15(1):647-51.
- [41] Din A, Li Y, Muhammad Khan F, UllahKhan Z, Liu P. On analysis of fractional order mathematical model of Hepatitis B using Atangana-Baleanu Caputo (ABC) derivative. Fractals. 2022;30(1):2240017.
- [42] Baleanu D, Mohammadi H, Rezapour S. Analysis of the model of HIV-1 infection of CD4+ CD4+ T-cell with a new approach of fractional derivative. Adv Differ Equ. 2020;2020(1):1-17.
- Shepard CW, Simard EP, Finelli L, Fiore AE, Bell BP. Hepatitis B virus infection: epidemiology and vaccination. Epidemiologic Rev. 2006;28(1):112-25.
- [44] Din A, Li Y. Stochastic optimal analysis for the hepatitis B epidemic model with Markovian switching. Math Meth Appl Sci. 2022;2022:1-26. doi: 10.1002/mma.8218.