Modeling the Interplay between HDV and HBV in Chronic HDV/HBV Patients

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Abstract: Hepatitis D virus is an infectious subviral agent that can only propagate in people infected with hepatitis B virus. In this study, we modified and further developed a recent model for early hepatitis D virus and hepatitis B virus kinetics to better reproduce hepatitis D virus and hepatitis B virus kinetics measured in infected patients during anti-hepatitis D virus treatment. The analytical solutions were provided to highlight the new features of the modified model. The improved model offered significantly better prospects for modeling hepatitis D virus and hepatitis B virus interactions.

Keywords: hepatitis delta virus; hepatitis B virus; anti-HDV treatment; analytical solution

MSC: 92-10

1. Introduction

Chronic hepatitis D virus (HDV) is a serious clinical concern with an estimated prevalence of ~10–20 million persons worldwide [1,2]. It was first described by Rizzetto and colleagues in the mid-1970s in patients with chronic hepatitis B virus (HBV) infection [3]. It was found that HDV requires the presence of HBV infection to assemble infectious HDV progeny virions [4]. In addition, HDV and HBV need the same proteins to enter hepatocytes [5]. There is no therapy approved for HDV by the US Food and Drug Administration [6]. Pegylated interferon-α treatment [7], which is endorsed by expert guidelines [8,9] with suboptimal outcomes, affects both HDV and HBV, which precludes modeling the interplay between the HBV and HDV [10].

Because HBV and HDV replicate in the same cells and influence each other and the host, we recently developed the first mathematical model to reproduce HDV and HBV kinetics in two patients receiving the prenylation inhibitor Lonafarnib (termed here anti-HDV) treatment in the absence of an anti-HBV therapy [11]. In the current study, we showed that the published model in [11] failed to explain the additional patients’ kinetic data reported in the LOWR HDV-1 (Lonafarnib with and without Ritonavir in HDV-1) study by Yurdaydin et al. [12] and suggested modifications to the model along with its analytical solutions. We provided the analytical solution of the improved model for both HDV and HBV using hypergeometric functions for HBV [13].
2. Background

We recently developed a model [11], that accounts for both HDV and HBV dynamics after the onset of anti-HDV treatment using the following differential equations:

\[
\frac{dD}{dt} = (1 - \varepsilon)p_1 I_0 e^{-\gamma t} - cD(t) \quad (1)
\]

\[
\frac{dB}{dt} = p_2 I_0 n^{-4} \left( \frac{D_0}{D(t)} \right) - cB(t) \quad (2)
\]

where \(D\) and \(B\) represent the HDV viral load and HBV viral load in blood, respectively. Parameters \(p_1\) and \(p_2\) represent the production rate constant of HDV and HBV from infected cell number, \(I_0\). We assumed that the clearance rate constant, \(c\), is the same for HDV and HBV and that it is within the range of the HDV clearance rate estimated in [14,15]. Parameter \(\varepsilon\) represents the efficacy of the treatment for HDV, with \(0 < \varepsilon < 1\), and \(g\) is the assumed additional treatment inhibitory effect in blocking HDV production as previously conducted under antiviral treatment for HBV [16,17]. Parameter \(n\) governs the HBV production rate increase under anti-HDV treatment.

Before treatment, HBV and HDV levels are in equilibrium with the production rate of HDV given by

\[ p_1 = \frac{(cD_0)}{I_0} \quad (3) \]

and that of HBV represented by

\[ p_2 = \frac{(cB_0)}{I_0} \quad (4) \]

where \(D_0\) and \(B_0\) are the HDV and HBV levels at the onset of treatment, respectively. The number of the infected cells, \(I_0\), was kept constant under treatment and prior to treatment onset, where \(\varepsilon = g = 0\), and \(n = 1\).

The original model Equations (1) and (2) simulated well the digitized serum HBV DNA and HDV kinetic data in two patients that were provided in Figure 5 in [12]. The two digitized patients displayed a moderate decrease in the second phase of HDV decline (i.e., low parameter \(g\) values) during anti-HDV treatment along with a sharp increase in HBV levels that could be reproduced by the model, as recently shown in Figures 3 and 4 in [11]. However, in the current study, we reported that the original model Equations (1) and (2) failed to explain additional patients who had a faster second phase decline in HDV (i.e., higher parameter \(g\) values) concomitantly with a moderate (or extremely slow) increase in HBV levels (Figure 1).

![Figure 1](image-url)

**Figure 1.** The model Equations (1) and (2) failed to reproduce HBV kinetics in additional patients reported in Yurdaydin et al. [5]. Model parameter values were \(D_0 = 1.2 \times 10^6\), \(B_0 = 282\), \(c = 0.51\), \(g = 0.089\), \(\varepsilon = 0.97\).
3. Modified Model

To address the inability of the model Equations (1) and (2) to reproduce HDV and HBV kinetics in some patients (Figure 1), we modified Equation (2) by replacing the term \( n^{-4}(D_0/D(t))^n \) with \( (1 + (k/D(t))^n) \) as follows and in Figure 2.

\[
\frac{dB}{dt} = p_2 I_0 \left( 1 + \left( \frac{k}{D(t)} \right)^n \right) - cB
\]  

Figure 2. Schematic diagram for the modified model, Equations (1) and (5).

Obtaining

where \( \kappa \) represents the threshold HDV level in blood that triggers an increase in HBV production, and \( \kappa \) is unique to each patient. As an example, in a patient with pre-treatment \( \kappa/D_0 = 0 \), as \( D(t) \) decreased under treatment, because \( D_0 > \kappa \), the production of HBV increased, e.g., at time \( t_1 \) after the initiation of treatment. When \( D(t_1) = \kappa \), the pre-treatment HBV production \( p_2 \) doubled (Figure 3). The time \( t_1 \) at which \( p_2 \) double depends on the pre-treatment ratio \( \kappa/D_0 \) as shown in Figure 3.

Figure 3. One-way sensitivity analysis on the impact of parameter \( \kappa \) on predicted HBV kinetics, B, (red curves) under anti-HDV treatment. Model parameters were set to \( D_0 = 1,202,260 \), \( B_0 = 282 \), \( c = 0.47 \), \( n = 1.25 \), \( g = 0.094 \), \( \tau = 0.1 \), and \( \varepsilon = 0.97 \). HDV predicted kinetics, D (blue curve) was not affected by different \( \kappa \) values.
Before treatment, the HBV and HDV are in equilibrium, with the production rate of HDV given by \( p_1 \) in Equation (3), and \( p_2 \) for HBV in Equation (4) is replaced by

\[
p_2 = \frac{cB_0}{I_0(1 + \left( \frac{\kappa}{I_0} \right)^n)}
\]

(6)

The number of the infected cells, \( I_0 \), was kept constant under treatment and prior to treatment onset, where \( \varepsilon = g = 0 \) as assumed in the original model. Furthermore, we assumed that HDV decay begins at time \( \tau \) (days), corresponding to the delay observed in the data and possibly reflecting Lonafarnib pharmacokinetics.

A one-way sensitivity analysis of our proposed model was carried out to investigate its robustness. The sensitivity analysis was carried out on the fittings for patient 10 as shown in Figures A1–A5, which are provided in Appendix A. We introduced some noise into our model by varying five parameters (\( g, c, n, \varepsilon, \) and \( \kappa \)) with a variation of \(+/− 10\%\) around our baseline values. It is worth stating that the variation range of \( \varepsilon \) was a bit lower than \(+/− 10\%\), since it was already closer to its upper bound, i.e., 1. We can clearly see that varying \( c, n, \) and \( \kappa \) had no impact on the HDV dynamics, while varying \( g \) and \( \varepsilon \) had a small impact. Additionally, it can be noted that varying of all the five parameters had no major effect on the HBV dynamics.

4. Results
4.1. The Modified Model Simulates the Measured Data

Using the Berkely Madonna software (berkeley-madonna.myshopify.com (accessed on 1 August 2022)) with the standard Runge–Kutta scheme of the fourth order, we reproduced the data well, as shown for four representative patients (Figure 4).
Figure 4. Model agreement with measured HDV RNA (circles) and HBV DNA (triangles) kinetics in four representative patients. Model parameters were (a) \( B_0 = 30,200, D_0 = 1.14815 \times 10^6, \varepsilon = 0.53, \) 
\( g = 0.077, n = 1.8, \) and \( \kappa = 9799, \) (b) \( B_0 = 151, D_0 = 6.30957 \times 10^5, \varepsilon = 0.97, \) 
\( g = 0.054, n = 2, \) and \( \kappa = 25,503, \) (c) \( B_0 = 5.88844 \times 10^5, D_0 = 5.01187 \times 10^7, \varepsilon = 0.97, \) 
\( g = 0.020, n = 1.1, \) and \( \kappa = 1.23198 \times 10^6, \) and (d) \( B_0 = 282, D_0 = 1.20226 \times 10^6, \varepsilon = 0.97, \) 
\( g = 0.094, n = 1.25, \) and \( \kappa = 26,137. \) The other model parameters \( c \) and \( \tau \) were fixed to 0.47 and 0.1, respectively, except for Patient 2 with \( \tau = 0.6, \) as depicted in Figure 4a.

4.2. Analytic Solutions
This subsection sought to determine the analytical solutions for \( D(t) \) and \( B(t). \) The analytical solution of \( D(t) \) was already solved in [11] and is provided herein again for convenience.

4.3. Analytic Solutions for HDV
We can derive the analytical solution of \( D(t) \) as follows:

From Equation (1), we have that

\[
\frac{dD}{dt} + cD = (1 - \varepsilon)p_1e^{-\gamma t}I_0.
\]

We then determined the integrating factor as, \( e^{ct} \), and by multiplying both sides of Equation (7) by the integrating factor, we have

\[
\frac{d}{dt}(De^{ct}) = (1 - \varepsilon)p_1e^{-\gamma t}I_0 e^{ct}.
\]

By integrating both sides of the Equation (8), it follows that

\[
(De^{ct}) = Me^{-\gamma t} e^{ct} + K,
\]

with \( M = \frac{(1-\varepsilon)p_1e^{-\gamma t}I_0}{c-g} \), which then simplifies to
D = Me^{-\beta t} e^{\alpha t} + Ke^{-ct}. \quad (10)

At time t = 0, we have D = D_0, and evaluating Equation (10) we now have

K = D_0 - M.

Thus, the exact solution for D is given by

\[ D(t) = Me^{-\beta t} + (D_0 - M)e^{-ct} \] \quad (11)

4.4. Analytical Solution for HBV and Plots

Using Wolfram Mathematica version 13.0.1.0, we managed to determine the analytical solution for HBV (Equation (5)) is given by

\[ b(t) \xrightarrow{\text{e}^{-ct}B_0} + \int_0^t B_0 \cdot e^{cx} \left( 1 + \left( -\frac{e^{c\cdot x}(c-g) \cdot e^{ct} - K}{D_0 - c - e^{ct} + e^{ct} - c + c + e^{ct} + e^{ct} - c + c} \right) \right) dx \]

Using the Berkeley Madonna simulation for the newly simulated Patient 10, we found its best-fit parameters:

\[ n = 1.25, \; \alpha = 0.97, \; \beta = 0.094, \; k = 26137 \; \text{and} \; c = 0.47 \]

With the above, the HBV analytical solution becomes

\[ b(t) \xrightarrow{279.66 + 1.28e^{-0.47t}} + 35.70 \left( \frac{1}{115715.75e^{-0.47t} + 45084.75e^{-0.0947}} \right)^{1.25} (115715.75 + 45084.75e^{0.3767t}) \]

\[ + \text{Hypergeometric2F1}\left[ 1, 2.5625, 3.8125, -0.0394e^{0.3767t} \right] \]

where Hypergeometric2F1 is the Gaussian hypergeometric function \( _2F_1(\alpha, \beta; \gamma; z) \), where \( \alpha, \beta, \) and \( \gamma \) are the function parameters and \( z \) is the variable of the Gaussian hypergeometric function.

In Figure 5, we show the simulations of the HDV and HBV plots for Patient 10.

![Patient 10 HDV](a)
for HDV using D. Madonna fit for Patient 10, which was a result of fitting the modified model Equations (1) and (5) (Figure 5. (b) follows a similar approach as conducted in 5a but for HDV using D.

Using the Berkeley Madonna simulation for Patient 2, we found its best-fit parameters:

\[ n = 1.80, \varepsilon = 0.53, \varsigma = 0.070, k = 9799 \text{ and the constant } c = 0.47 \]

With the above, the HBV analytical solution becomes

\[
b[t] \\
\rightarrow 30194.30 + 2.98e^{-0.47t} \\
+320523.47e^{-1.66 \times 10^{-16} t} \left( \frac{1}{514084.16e^{-0.47t} + 634065.8375e^{-0.07t}} \right)^{1.8} (514084.16) \\
+634065.84^{0.47})^{1}\text{Hypergeometric2F1} \left[ 1, 2.49, 4.29, -1.23^{0.47} \right]
\]

In Figure 6, we show the simulations of HDV and HBV plots for Patient 2.
Figure 6. (a) shows the HBV Mathematica analytical solution (i.e., Equation (14) herein), HBV Berkeley Madonna fit for Patient 2, which was a result of fitting the modified model with measured data using Berkeley Madonna as reported in [4] and the HBV raw data. (b) follows a similar approach as conducted in 6a but for HDV using D.

Using the Madonna simulation for Patient 4, we found its best-fit parameters:

\[ n = 2.00, \varepsilon = 0.97, g = 0.054, k = 25503 \text{ and the constant } c = 0.47 \]

With the above, the HBV analytical solution becomes

\[
\begin{align*}
    b(t) &\rightarrow 150.75 + 0.16e^{-0.47t} \\
    &+ 53617.29 \left( \frac{1}{609571.20e^{-0.47t} + 21385.80e^{-0.054t}} \right)^2 \cdot 609571.20 \\
    &+ 21385.80e^{0.416t}^{1} \text{ Hypergeometric2F1} \left[ 1, 2.39, 4.39, -0.035e^{0.416t} \right]
\end{align*}
\]

(15)

In Figure 7, we show the simulations of HDV and HBV plots for Patient 4.

Figure 7. Cont.
Using the Madonna simulation for the newly simulated Patient 5, we found its best-fit parameters:

\[ n = 0.81, \epsilon = 0.97, g = 0.021, k = 1716544 \text{ and the constant } c = 0.47 \]

With the above, the HBV analytical solution becomes

\[
\begin{align*}
    b(t) & \rightarrow 552892.94 + 15911.46e^{-0.47t} \\
    & + 706.14\left( \frac{1}{4.85 \times 10^7 e^{-0.47t} + 1573883.45e^{-0.021t}} \right)^{0.81} (4.85 \times 10^7) \\
    & + 1573883.45e^{0.449t} \ Hypergeometric2F1 \left[ 1, 2.08, 2.89, -0.032e^{0.449t} \right] 
\end{align*}
\]

In Figure 8, we show the simulations of HDV and HBV plots for Patient 5.
The term 
\[ (1+\left(\frac{\kappa}{D}\right)^n) \] 
allows starting with a pre-treatment HBV production rate, which could increase as a result of HDV decline by the several suggested mechanisms that were reviewed in [18]. The value of \( \kappa \), which was unique for each patient (Figure 4), could represent, in part, the interplay between HDV and HBV.

The modified model could explain HDV decline with different HBV kinetic changes: (i) no change (Figure 4a), (ii) a delay followed by an increase (Figure 4b,d), and (iii) a fast increase followed by a slower increase (Figure 4c). In addition, we provided the analytical solution of the HBV modified model using hypergeometric functions [6].

In future work, our modified model could be examined in a much larger number of patients and anti-HDV combination therapies. The modified model could also be extended to account for additional biological features, e.g., immune response, molecular interactions, and/or susceptible and infected cell dynamics. In addition, previously reported HDV/HBV mathematical models [19–21] could be extended to account for the unique HDV and HBV dynamics proposed in this study. As such, the modified model is a better starting point for developing more comprehensive HDV/HBV models for understanding the interactions between HBV and HDV that, in turn, may help to design therapeutic strategies to fight HDV.


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Conflicts of Interest: Yurdaydin and Etzion are on the speakers’ bureau and received grants from Eiger. Glenn is the founder and a director of, holds intellectual property rights with, and owns stock and royalty rights from Eiger. The other authors declare no conflict of interest.

Appendix A

![Figure A1](image1.png)

**Figure A1.** One-way sensitivity analysis on the impact of parameter c on predicted HBV/HDV kinetics under anti-HDV treatment. Model parameters were set to $D_0 = 1,202,260$, $B_0 = 282$, $\kappa = 26,137$, $n = 1.25$, $g = 0.094$, $\tau = 0.1$, and $\varepsilon = 0.97$, with $c$ being set at $c = (0.423, 0.4465, 0.47, 0.4935, 0.517)$. The arrows depict the direction of increase of the parameter $c$.

![Figure A2](image2.png)

**Figure A2.** One-way sensitivity analysis on the impact of parameter n on predicted HBV/HDV kinetics under anti-HDV treatment. Model parameters were set to $D_0 = 1,202,260$, $B_0 = 282$, $\kappa = 26,137$, $c = 0.47$, $g = 0.094$, $\tau = 0.1$, and $\varepsilon = 0.97$, with $n$ being set at $n = (1.125, 1.1875, 1.25, 1.3125, 1.375)$. HDV predicted kinetics were not affected by different n values.
**Figure A3.** One-way sensitivity analysis on the impact of parameter $\kappa$ on predicted HBV/HDV kinetics under anti-HDV treatment. Model parameters were set to $D_0 = 1,202,260$, $B_0 = 282$, $n = 1.25$, $c = 0.47$, $g = 0.094$, $\tau = 0.1$, and $\epsilon = 0.97$, with $\kappa$ being set at $\kappa = (23,523, 24,830, 26,137, 27,444, 28,751)$. The arrows depict the direction of increase of the parameter $\kappa$. HDV predicted kinetics were not affected by different $n$ values.

**Figure A4.** One-way sensitivity analysis on the impact of parameter $g$ on predicted HBV/HDV kinetics under anti-HDV treatment. Model parameters were set to $D_0 = 1,202,260$, $B_0 = 282$, $\kappa = 26,137$, $c = 0.47$, $n = 1.25$, $\tau = 0.1$, and $\epsilon = 0.97$, with $g$ being set at $g = (0.0846, 0.0893, 0.094, 0.0987, 0.1034)$.

**Figure A5.** One-way sensitivity analysis on the impact of parameter $\epsilon$ on predicted HBV/HDV kinetics under anti-HDV treatment. Model parameters were set to $D_0 = 1,202,260$, $B_0 = 282$, $\kappa = 26,137$, $c = 0.47$, $n = 1.25$, $\tau = 0.1$, and $g = 0.094$, with $\epsilon$ being set at $\epsilon = (0.95, 0.96, 0.97, 0.98, 0.99)$. 