

Research Article

Various Gradients of the Second Phase of the Viral Kinetics of Hepatitis C Virus Infection Drawn using the Rapid Equilibrium Model

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Abstract

Mathematical models of Hepatitis C Virus (HCV) infection during antiviral treatment have revealed the major mechanism of the function of Interferon (IFN) and the estimated clearance rates of HCV virions and HCV-infected cells. However, differences in the estimated clearance rates of infected cells have been observed between subjects who are and are not treated with Protease Inhibitor. In this study, the author demonstrated that various gradients in the second phase were difficult to obtain by regulating the efficacy of treatment in the predominately used basic model but were easily obtained by regulating the clearance rate of infected cells. The author considers that the estimated clearance rate of infected cells differs depending on the type of treatment because fitting the second phase in the basic model to viral load by regulating efficacy is difficult; however, fitting the second phase by regulating the clearance rate of infected cells is not difficult. Consequently, different infected-cell clearance rates were obtained for different types of treatments. Previously, the author proposed the Rapid Equilibrium (RE) model, which is capable of displaying viral kinetics without rebounds or oscillations under some conditions. In this study the author demonstrated the applicability of the RE model by displaying various gradients of the second phase with various efficacies of treatment.

Keywords: HCV; Mathematical Model; Viral Hepatitis; Rapid Equilibrium Model

Abbreviations

HCV: Hepatitis C Virus; HCC: Hepatocellular Carcinoma; IFN: Interferon; RBV: Ribavirin; PEG-IFN: Pegylated Interferon; PI: Protease Inhibitor; SVR: Sustained Virological Response; HIV: Human Immunodeficiency Virus; HBV: Hepatitis B Virus; RE: Rapid Equilibrium

Introduction

Hepatitis C Virus (HCV) is known to cause chronic hepatitis, liver cirrhosis and Hepatocellular Carcinoma (HCC) and infects approximately 130 to 150 million people worldwide [1]. Interferon (IFN) mono-therapy, IFN in combination with Ribavirin (RBV), Pegylated Interferon (PEG-IFN) mono-therapy and PEG-IFN in combination with RBV have been prescribed to chronic HCV hepatitis subjects. Recently, Protease Inhibitor (PI) has been administered to HCV genotype 1-infected subjects [2-4]. Despite drug-resistant mutations and serious side effects on the skin, the administration of PI improves the rate of Sustained Virological Response (SVR), which is defined as the absence of detectable serum HCV RNA 24 weeks after the cessation of treatment. Similar to Human Immunodeficiency Virus (HIV) infection [9,10] and Hepatitis B Virus (HBV) hepatitis [8,11-13], HCV hepatitis has been analyzed using various mathematical models [5-8]. Moreover, the clearance rates of HCV virions and HCV-infected cells have been estimated

in mathematical studies [4,5,14-16]; however, the estimated HCV-infected cell clearance rates of subjects treated with and without PI are different [4]. The infected-cell clearance rate of subjects treated with PI is estimated to be approximately ten times higher than that of subjects treated without PI [4]. Adiwijaya et al, assumed different infected-cell clearance rates for subjects treated with and without PI [17], and Adiwijaya et al. assumed that clearance rates depended on the efficacy of treatment [18]. However, PI, which inhibits the protease of HCV, is not expected to directly increase the infected-cell clearance rate.

The author published a study dealing the Rapid Equilibrium (RE) model for chronic viral hepatitis during anti-viral treatment [8]. In the previous study, the author demonstrated that the RE model could exhibit viral kinetics without rebounds or oscillations under some conditions. In the present paper, the author compared the well-known basic model [5] with the RE model [8], particularly in terms of the second viral decline. The author discusses the discrepancy in the estimated infected-cell clearance rates between subjects who are treated with and without PI.

Materials and Methods**Modeling****Basic model**

The basic model [5] of viral infection is described by equation (1)

below;

$$\begin{cases} \frac{dT}{dt} = s - \delta_T T - (1-\eta)\beta_1 VT \\ \frac{dI}{dt} = (1-\eta)\beta_1 VT - \delta_I I \\ \frac{dV}{dt} = (1-\varepsilon)pI - cV \end{cases} \quad (1)$$

Where T is the number of target cells, I is the number of infected cells, and V is the viral load. Target cells are produced at rate s and die at rate δ_T . Target cells become infected at rate $\beta_1 V$, and infected cells die at rate δ_I . HCV virions are produced by an infected cell at rate p and are cleared at rate c . Antiviral treatment reduces the infection rate by the fraction $(1 - \eta)$ and/or the viral production rate by the fraction $(1 - \varepsilon)$.

Rapid Equilibrium model

The RE model [8] is described by equation (2) below;

$$\begin{cases} \frac{dT}{dt} = s - \delta_T T - \frac{(1-\eta)\beta_2 VT}{K+V} \\ \frac{dI}{dt} = \frac{(1-\eta)\beta_2 VT}{K+V} - \delta_I I \\ \frac{dV}{dt} = (1-\varepsilon)pI - cV \end{cases} \quad (2)$$

Where the parameter K is analogous to the Michaelis constant K_m , i.e., K is the viral load for the half-maximal production of infected hepatocytes from uninfected hepatocytes and virus. Target cells become infected at rate $\beta_2 V / (K + V)$.

Values of the Parameters

To investigate the differences between the basic model and the RE model, the following identical values based on previous studies [5,16] were used: initial $V(V_0) = 1 \times 10^6$ virions ml^{-1} ; $\delta_T = 0.003 \text{ day}^{-1}$; $\delta_I = 0.51 \text{ day}^{-1}$; $p = 100$ virions $\text{cell}^{-1} \text{ day}^{-1}$; and $c = 12.2 \text{ day}^{-1}$.

In the previous study [8], the author assumed $\beta_1 = 2.25 \times 10^{-9}$ virions $^{-1} \text{ ml day}^{-1}$ to exhibit various biphasic viral declines with various values of parameter ε . In this study, $\beta_1 = 3 \times 10^{-7}$ virions $^{-1} \text{ ml day}^{-1}$ was assumed, because the value of parameter β_1 was assumed to be numbers in minus seven figures in the previous studies [5-7].

Because the production of infected hepatocytes is given by $\beta_1 VT$ in the basic model and by $\beta_2 VT / (K + V)$ in the RE model, $\beta_2 = \beta_1 K$ was assumed. Parameter $K = 7.5 \times 10^4$ virions ml^{-1} was determined to make the gradient of the second phase change as the value of parameter ε changed; consequently $\beta_2 = 2.25 \times 10^{-2} \text{ day}^{-1}$ was obtained.

Assuming a steady state ($dT/dt = dI/dt = dV/dt = 0$) before treatment ($\eta = \varepsilon = 0$), equations (3) and (4) were obtained for the basic model, and equations (5) and (6) were obtained for the RE model. The values of parameter s were obtained by equations (4) and (6) for the basic model and the RE model, respectively.

$$\begin{aligned} V_0 &= \frac{ps}{c\delta_I} - \frac{\delta_T}{\beta_1} \quad (3) \\ \therefore s &= \frac{c\delta_I}{p} \left(V_0 + \frac{\delta_T}{\beta_1} \right) \quad (4) \end{aligned}$$

$$V_0 = \frac{ps\beta_2 - c\delta_I\delta_T K}{c\delta_I(\beta_2 + \delta_T)} \quad (5)$$

$$\therefore s = \frac{c\delta_I\delta_T}{p\beta_2} (K + V_0) + \frac{c\delta_I V_0}{p} \quad (6)$$

Results

Figure 1 A and B show the simulations of the basic model and the RE model, respectively, from day 0 to 28 with various values of parameter ε . In Figure 1A, as the value of ε increases between 0.8 and 0.999, the first viral decline increases. In contrast, the gradient of the second phase changes only minimally while $0.95 \leq \varepsilon \leq 0.999$. However, when $\varepsilon = 0.9$, a downwardly convex curve was obtained in the second phase of viral decline, and when $\varepsilon = 0.8$, a rebounding curve was obtained in the second phase.

As shown in Figure 1B, as the value of ε increases between 0.8 and 0.999, the first viral decline increases as in Figure 1A. Furthermore, the gradient of the second phase decreases as the value of ε decreases between 0.97 and 0.999. When $\varepsilon = 0.95$, a downwardly convex curve was obtained in the second phase of viral decline, and when $\varepsilon = 0.8$ and 0.9, flat phases were obtained after the first phase of viral decline.

To confirm the ease of obtaining various gradients of the second phase using the basic model with various values of δ_I , Figure 2 shows the simulations using the basic model from day 0 to 28 with various values of parameter δ_I . To maintain $V_0 = 1 \times 10^6$ virions ml^{-1} , the values of parameters were obtained by equation (4). Figure 2 shows the viral kinetics with $\delta_I = 0.05, 0.1, 0.2, 0.3, 0.4$ and 0.51 day^{-1} , and $s = 6.161 \times 10^3, 1.2322 \times 10^4, 2.4644 \times 10^4, 3.6966 \times 10^4, 4.9288 \times 10^4$ and $6.28422 \times 10^4 \text{ cell ml}^{-1} \text{ day}^{-1}$, respectively. A comparison of Figure 1A and Figure 2 reveals that, as expected, the regulation of δ_I was more effective in changing the gradient of the second phase than the regulation of ε .

Discussion

In the present study, the author demonstrated that the gradient of the second phase of viral decline changed according to changes in efficacy using the RE model shown in Figure 1B. As the clinical data [3,19] show, the first and second phases depend on efficacy as in

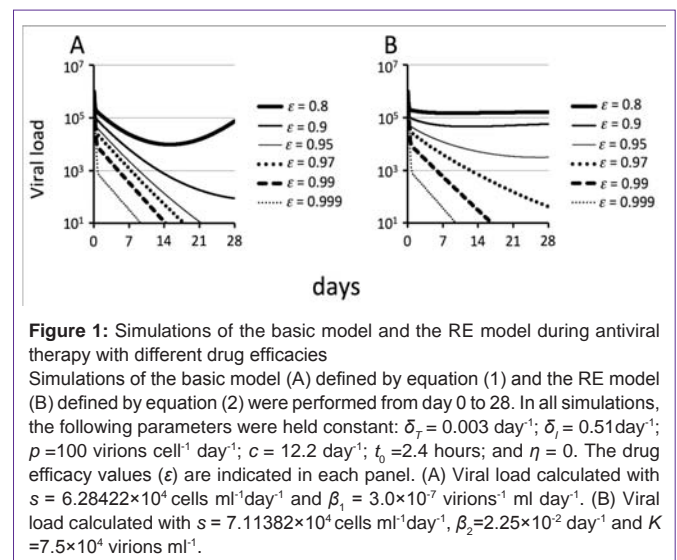


Figure 1: Simulations of the basic model and the RE model during antiviral therapy with different drug efficacies. Simulations of the basic model (A) defined by equation (1) and the RE model (B) defined by equation (2) were performed from day 0 to 28. In all simulations, the following parameters were held constant: $\delta_T = 0.003 \text{ day}^{-1}$; $\delta_I = 0.51 \text{ day}^{-1}$; $p = 100$ virions $\text{cell}^{-1} \text{ day}^{-1}$; $c = 12.2 \text{ day}^{-1}$; $t_0 = 2.4$ hours; and $\eta = 0$. The drug efficacy values (ε) are indicated in each panel. (A) Viral load calculated with $s = 6.28422 \times 10^4 \text{ cells ml}^{-1} \text{ day}^{-1}$ and $\beta_1 = 3.0 \times 10^{-7} \text{ virions}^{-1} \text{ ml day}^{-1}$. (B) Viral load calculated with $s = 7.11382 \times 10^4 \text{ cells ml}^{-1} \text{ day}^{-1}$, $\beta_2 = 2.25 \times 10^{-2} \text{ day}^{-1}$ and $K = 7.5 \times 10^4 \text{ virions ml}^{-1}$.

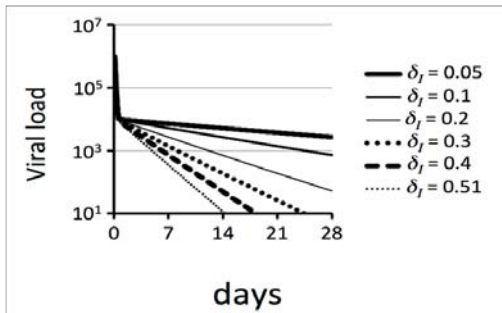


Figure 2: Simulations of the basic model during antiviral therapy with various values of parameters s and δ_i . Simulations of the basic model defined by equation (1) were performed from day 0 to 28. In all simulations, the following parameters were held constant: $\delta_e = 0.003 \text{ day}^{-1}$; $\beta_1 = 3.0 \times 10^{-7} \text{ virions}^{-1} \text{ ml day}^{-1}$; $p = 100 \text{ virions cell}^{-1} \text{ day}^{-1}$; $c = 12.2 \text{ day}^{-1}$; $t_0 = 2.4 \text{ hours}$; $\eta = 0$; and $\epsilon = 0.99$. To maintain $V_0 = 1 \times 10^6 \text{ virions ml}^{-1}$, the value of parameter s was calculated for each value of δ_i by equation (4). For the sake of clarity of the figure, only the values of δ_i are indicated in the graph, and $s = 6.161 \times 10^3$, 1.2322×10^4 , 2.4644×10^4 , 3.6966×10^4 , 4.9288×10^4 and $6.28422 \times 10^4 \text{ cell ml}^{-1} \text{ day}^{-1}$ when $\delta_i = 0.05, 0.1, 0.2, 0.3, 0.4$ and 0.51 day^{-1} , respectively.

Figure 1B. However, the author was unable to obtain the viral kinetics similar to those reported in the clinical data using the basic model with $\beta_1 = 3 \times 10^{-7} \text{ virions}^{-1} \text{ ml day}^{-1}$ as shown in Figure 1A. The second phase changed only minimally while $0.95 \leq \epsilon \leq 0.999$; furthermore, the second phase became a rebounding curve when $\epsilon = 0.8$.

Because the administration of PI improves the rate of SVR, the value of parameter ϵ should be larger with PI treatment than without PI treatment. Although PI is not expected to directly increase the infected-cell clearance rate, the infected-cell clearance rate was assumed to be larger in the subjects who are treated with PI than in those who are not treated with PI [4,17,18]. The author doubted the veracity of the discrepancy in the estimated infected-cell clearance rates, which led the author consider the following thought process. When fitting the basic model to HCV RNA viral load, one has to estimate the value of parameter ϵ as large for the second phase to be straight. Then, one must fit the second phase to the HCV RNA viral load by regulating the value of the infected-cell clearance rate. Consequently, the infected-cell clearance rate of the subjects who are treated with PI is estimated to be larger than of the subjects who are treated without PI. Thus, the administration of treatments that are more effective than the current available therapies would necessitate estimate even larger estimations of the infected-cell clearance rate.

In the previous study [8], the author suggested that the basic model is capable of explaining early viral kinetics during therapies but is incapable of explaining long-term viral kinetics. Under the conditions utilized in this paper, the basic model might not be able to explain the second phase. The author considers that the RE model might be more appropriate for analyzing the second phase than the basic model because the RE model is capable of exhibiting the various gradients of the second phase according to various values of parameter ϵ as shown in Figure 1B.

The discrepancy in the estimated infected-cell clearance rates was investigated in this study. The author considers that we have to assure whether the values of the other parameters are also reasonable medically. Because a new problem of drug resistant mutations has

occurred, modeling HCV infection during treatment has to be more complicated [17,18]. Nevertheless, in the future, modeling HCV infection will help us with making predictions of treatment outcomes clinically.

Conclusion

In the present study, the author compared the basic model [5] to the RE model [8] in terms of the second phase. Under the conditions of this study, regulating the efficacy of treatment in the basic model was incapable of producing a trend in the second phase that mimicked the clinical data [3,19]; however, regulating the infected-cell clearance rate was capable of mimicking this phase. The author considers that the estimated infected-cell clearance rate differs depending on the type of treatment because fitting the second phase in the basic model to viral load by regulating efficacy is difficult, but fitting it by regulating the infected-cell clearance rate is not difficult. In contrast, regulating the efficacy of treatment in the RE model was able to produce a trend in the second phase that mimicked the clinical data [3,19]. The author considers that the RE model is thus more appropriate for analyzing the second phase.

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