

NIH Public Access

Author Manuscript

Hepatology. Author manuscript; available in PMC 2013 July 16.

Published in final edited form as:

Hepatology. 2009 January ; 49(1): 32–38. doi:10.1002/hep.22586.

Modeling Complex Decay Profiles of Hepatitis B Virus during Antiviral Therapy

Harel Dahari1, **Emi Shudo**2, **Ruy M. Ribeiro**2, and **Alan S. Perelson**2,*

¹Department of Medicine, University of Illinois at Chicago, Chicago, IL 60612

²Theoretical Biology and Biophysics, MS-K710, Los Alamos National Laboratory, New Mexico 87545

Abstract

Typically, hepatitis B virus (HBV) decays in patients under therapy in a biphasic manner. However, more complex decay profiles of HBV DNA (e.g. flat partial response, triphasic, stepwise) have also been observed in some treated patients with no clear understanding of their origin. We recently introduced the notion of a critical drug efficacy, ε_c , such that if overall drug efficacy, ε_{tot} , is higher than the critical drug efficacy, i.e., $\varepsilon_{\text{tot}} > \varepsilon_c$, then viral levels will continually decline on therapy, while if $\varepsilon_{\text{tot}} < \varepsilon_{\text{c}}$ then viral loads will initially decline but ultimately stabilize at a new set point, as seen in flat partial responders. Using the idea of critical efficacy and including hepatocyte proliferation in a viral kinetic model we can account for these complex HBV DNA decays. The model predicts that complex profiles such as those exhibiting a plateau or shoulder phase as well as a class of stepwise declines occur only in patients in whom the majority of hepatocytes are infected before therapy. In conclusion, we show via kinetic modeling how a variety of HBV DNA decay profiles can arise in treated patients.

Keywords

mathematical modeling; hepatitis B virus; multiphase viral decay; critical drug efficacy

INTRODUCTION

Infection with hepatitis B virus (HBV) affects hundreds of millions of people worldwide and is the major cause of hepatocellular carcinoma (1). Moreover, 10% of HIV-infected people are co-infected with HBV \sim 4 million people worldwide). This important health burden has led to the development of several approved therapies for chronic HBV infection (1), including interferon-α (IFN) and nucleot(s)ide analogues (NA) that interfere with the reverse transcription step of the HBV lifecycle.

While in most patients HBV DNA levels fall during therapy, the pattern of fall is varied. Often HBV DNA falls in a biphasic manner, with a fast first phase decrease of several logs over a few weeks, followed by a slower but continued decrease over many weeks (Fig. 1A). In others, HBV DNA decay under therapy seems to proceed in triphasic-like (Fig. 1C) or staircase-like (Fig. 1D) fashions (2), with periods of viral decrease interspersed with more or less long periods of stable viral load. However, in many patients primary treatment failure is observed, with less than 1 log decay over the first 6 months of treatment (3); in others, secondary treatment failure occurs, when in spite of an initial viral response, HBV DNA

^{*}Corresponding author: Address: Theoretical Biology and Biophysics, MS-K710, T-10, Los Alamos National Laboratory, New Mexico 87545. Tel: (505)- 667-6829; Fax: (505) – 665-3493; asp@lanl.gov.

remains flat (Fig. 1B) or increases (Fig. 1E) more than 1 log over nadir in two successive treatment visits over more than a month apart (3).

A better understanding of the origins of these varied profiles of HBV DNA decay under therapy could lead to improved management of HBV treatment. While host factors, such as genetic background, drug metabolism, and strength of the immune response, and viral factors, such as genotype, quasi-species diversity, and drug resistance could affect the HBV DNA decay profile, here we show how viral dynamic factors alone can explain many of the observed patterns. To provide this explanation we introduce the concept of a critical drug effectiveness, which has previously been used to explain HIV $(4-6)$ and hepatitis C (7) infection dynamics under therapy. We show how the effectiveness of therapy in a given patient, relative to the patient's critical drug effectiveness, and the fraction of hepatocytes infected at treatment baseline dynamically influence the pattern of HBV DNA decay under therapy.

METHODS

Model Description

Recently, we extended (7, 8) a model of HCV infection (9) to account for the proliferation of hepatocytes. Along analogous lines, here we generalize a previous model of HBV infection and treatment (2) by including proliferation of uninfected hepatocytes, T, and infected hepatocytes, I. Our previous model for HCV infection (7, 8) does not include the possibility of noncytolytic cure of infected cells, which we include in the HBV model. The models also differ in that HCV treatment involves ribavirin whose effect was modeled by the production of noninfectious virions (7), an effect not needed when modeling HBV treatment (2, 10–12). The equations describing our new model of HBV infection and treatment are

$$
\frac{dT}{dt} = s + rT T(1 - \frac{T+I}{T_{\text{max}}}) - dT T - (1-\eta)\beta VT + \rho I \frac{dI}{dt} = (1-\eta)\beta VT + rI I(1 - \frac{T+I}{T_{\text{max}}}) - (\delta + \rho)I \frac{dV}{dt} = (1-\varepsilon)pI - cV \tag{1}
$$

where *V* represents virus, HBV. The model assumes that uninfected hepatocytes susceptible to HBV infection (target cells), T, are produced at a constant rate s, die at rate, d_T , per cell and are infected by virus, V , at constant rate β . Uninfected and infected hepatocytes can proliferate with maximum proliferation rates r_T and r_L , respectively. However, as the total hepatocyte number, $T+I$, approaches its natural level, T_{max} , we assume proliferation slows down according to a homeostasis process in which only the total number of hepatocytes matter and not whether they are infected or not (13). In order to have a physiologically realistic model (*i.e.*, T before infection less than T_{max}) the hepatocyte influx rate s, needs to be less than $d_T T_{max}$. Here the influx represents both uninfected hepatocytes that become susceptible to infection and *de novo* generation of hepatocytes from stem cells (14). Due to the burden of supporting HBV replication, we allow infected cells to proliferate slower than uninfected cells, i.e., r_I r_T . Infected hepatocytes die at a constant rate δ per cell and are possibly "cured" by noncytolytic processes at a constant rate ρ per cell. As hepatocytes die the total number of hepatocytes will be reduced from T_{max} , and proliferation will resume. In the absence of treatment, each productively infected cell releases new infectious virions at rate p . Virus particles are assumed to be cleared at rate c per particle. IFN and NA therapies are assumed to lower p by a factor $(1-e)$ – where e is the effectiveness of drug therapy in blocking virion production – and may also lower β by a fraction $(1-\eta)$ – where η is the effectiveness of therapy in blocking infection. For simplicity, we define in our model an overall drug effectiveness, ε_{tot} , where $1 - \varepsilon_{tot} = (1 - \eta)(1 - \varepsilon)$. Before treatment, a patient with chronic HBV infection is assumed to be in an infected steady state with $\varepsilon_{tot}=0$.

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We can find a critical drug efficacy, ε_c , such that when ε_{tot} is higher than ε_c , viral levels are predicted to continually decline on therapy $(4-8)$. One can calculate ε_c from our model in a manner analogous to the methods we previously used for analyzing HCV models (7, 8). As shown in the online suppplemental material, we find

$$
\varepsilon_c = 1 - \frac{c((\delta + \rho)T_{\text{max}} + r_I(\bar{T}_0 - T_{\text{max}}))}{\rho \beta T_{\text{max}} \bar{T}_0}, \quad (2)
$$

where \bar{T}_0 is the steady state number of hepatocytes in an uninfected individual (see online supplemental material).

RESULTS

We introduced a model of treatment of HBV infection, Eq. (1), that includes proliferation of uninfected and infected hepatocytes. In the context of this model, we define the concept of a critical drug efficacy, ε_c (see Methods), such that treatments with efficacy higher than the critical efficacy are predicted to be successful in reducing HBV viral load. Moreover, in the case of *successful* drug therapy, i.e., $\varepsilon_{\text{tot}} > \varepsilon_o$ HBV DNA is predicted to decline in either the typical biphasic manner (Fig. 2A; thick solid line) or in a triphasic manner (Fig. 2B; thick solid line). When the efficacy of treatment is lower than the critical drug efficacy, i.e., $\varepsilon_{\text{tot}} < \varepsilon$, therapy is predicted to be unsuccessful with HBV DNA initially falling, but then achieving a new steady state, i.e. plateau level. In this case, the predicted patterns of HBV DNA decay include a flat second phase after an initial fast first phase decline (Fig. 2A, dashed line); a slow approach to the new steady-state (Fig. 2A, thin line); a stepwise decline pattern, with an intermediate period of constant viral load (Fig. 2B).

Which pattern is actually observed within a patient depends on the relationship between the critical drug efficacy and the overall treatment efficacy for that individual (Table 1). When the total drug effectiveness is higher than ε_c , the viral load should decline to below the limit of detection. The model predicts that this decline may be biphasic (Fig. 2A) or triphasic (Fig. 2B), depending on whether or not a shoulder phase exists. During the shoulder phase, not only the virus but also the number of infected cells must remain constant. Under potent therapy, few new infections occur (15), and thus proliferation of infected cells is the major source of new infected cells. Thus, proliferation helps compensate for the loss of infected cells, a feature that was not possible in earlier models (2, 11, 12). This proliferation can keep the number of infected cells constant for some period of time, but because uninfected cells are assumed to proliferate faster than infected ones their relative number increases, and due to competition for space and other resources this ultimately slows down the proliferation of infected cells to the point that infected cell numbers decline and the shoulder phase ends. We recently showed (7) that the shoulder phase ends approximately when the number of uninfected hepatocytes reaches the same order of magnitude as the infected ones. This demarks the transition from the shoulder phase to the third phase of viral decline. When the ratio of uninfected cells to infected cells at the start of therapy is close to or greater than one, the shoulder phase does not exist. Thus, our model predicts that triphasic responses will only be seen in patients in which a majority of hepatocytes are infected before therapy is begun.

Interestingly, the model predicts that the length of the shoulder phase is dependent on how fast the number of uninfected hepatocytes, T, reaches the same order of magnitude as the infected ones, I, assuming that $T < I$ at baseline. An illustration of this notion is shown in Fig. 2 in which the time needed for T to reach I is governed by the ratio of the target cell to the infected cell replication rate, r_T/r_I (Fig. 2B). It may be possible that some patients exhibiting biphasic decline also have $T \leq I$ at baseline but a shoulder phase will not be observed if fast

replication of uninfected hepatocytes causes T to increase to approximately I during the first phase of viral decline.

Clearance of HBV infected hepatocytes has been shown to occur through non-cytolytic mechanisms in HBV transgenic mice (16) and strongly suggested to occur during acute HBV infection in chimpanzees (17) and possibly humans. Analysis of our model shows that "cure" of infected cells, ρ, can shorten the shoulder phase of the triphasic decline or even abolish it completely, depending on its magnitude (Fig. 2C).

For reasons that are not fully understood, therapy is not equally effective in all patients. For patients in which the overall efficacy of drug therapy is lower than e_c , the viral load will decline to a viral plateau lower than baseline, and if the patient is compliant with therapy the virus will remain at the plateau until the end of therapy, until therapy is changed (18), or until drug resistance develops (3). Again, the specific profile of decay depends on the ratio of uninfected cells to infected cells at baseline as well as viral kinetic parameters. According to the model, a flat second phase can be obtained with large death rates of infected cells, δ, a feature that was not possible in earlier models in which target cells were held constant (11, 12).

Lewin et al. (2) observed that some HBV treated patients exhibited a staircase or stepwise decline in HBV DNA under treatment. Our model can explain a class of stepwise decays in which there is a flat second phase followed by a second drop in HBV DNA and a final flat phase (Fig. 2B). Note that in our model all parameters are held constant and thus the stepwise decreases in viral load are not caused by parameter changes. The "two-step declines" that our model generates are a special case of triphasic decays, that are characterized by having a shoulder phase, but in which $e_{\text{tot}} < e_c$ so that therapy is not potent enough to continually eliminate virus and hence leads to a final state in which virus is still present. However, simulating the model with over 275,000 choices of sets of parameters spanning the biologically feasible ranges of all the parameters showed that the model could only generate modest sized second drops, i.e., less than 1 log (Fig. 2B). For patients in whom this second drop is very large (e.g., Patients 6 and 11 in Fig. 5 from (2)), other effects such as changes in the immune response may be playing a role.

In some treated patients a viral rebound is observed during therapy (15). Several explanations for this viral rebound have been suggested, such as the development of antiviral drug resistance or/and poor adherence to therapy (3). Here we offer another explanation. Our model shows that viral rebound can occur during suboptimal therapy, i.e., when $\varepsilon_{\text{tot}} < \varepsilon_c$, in the absence of resistance if the total number of hepatocytes susceptible to HBV infection increases during therapy (Fig. 2D; Table 1). Moreover, depending of the total drug efficacy, HBV DNA levels may rebound to baseline levels or rebound and stabilize at a lower therapy-induced set point (Fig. 2D).

We compared our model with experimental HBV DNA data. By varying patient specific parameters the kinetics of HBV decline predicted by our model can be shown to quantitatively agree with: (i) a biphasic decline (Fig. 3A) observed in a HBeAg-positive patient (Patient 16 from (2)) treated with lamivudine and famciclovir), (ii) a triphasic decline (Fig. 2B) observed in an HBeAg-positive patient (Patient F in Fig. 5 from (11)) treated with adefovir dipivoxil (ADV), (iii) a flat second phase decline (Fig. 3C) observed in HBeAgnegative patients (e.g., Patient 8472 from (19)) treated with pegylated intereferon-α2a, and (iv) a slow viral rebound (Fig. 3D) observed in an HBeAg-positive patient (Patient 10 from (2)) treated with lamivudine.

Discussion

The field of viral dynamic modeling is predicated on the assumption that the kinetic pattern of viral decay under therapy can be informative. In the case of HIV, viral dynamic analyses revealed that very rapid viral replication and clearance were occurring despite the relatively constant level of viremia observed in untreated patients. Similar analyses then were performed for HCV and HBV (see review (20)). Here we have extended viral kinetic analyses of HBV infection in order to shed light on the origin of some complex HBV DNA decay patterns. In many patients on treatment, the now classic biphasic decay of HBV DNA is observed (2, 10, 11, 19, 21). However, Lewin et al. (2) pointed out that in some patients more complex decay profiles occur. Here we have shown, as in the case of HCV infection (7), that patients can exhibit a triphasic decay or a fast first phase followed by a gradual approach to a steady state HBV level – patterns that are not a classic second phase. We also showed that a class of stepwise declines in which there is $<$ 1 log drop in HBV DNA after a flat second phase (Fig. 2B) as well as viral rebounds (Fig. 2D and Fig. 3D) could be also be explained by this new model.

Including hepatocyte proliferation (of both infected and uninfected cells) in an HBV kinetic model and introducing the notion of critical drug efficacy allows us to explain not only the typical biphasic HBV decay but also more complex viral kinetic patterns that have been observed under therapy. Four main features, according to the model, define the HBV DNA decay patterns under therapy: (i) the ratio of HBV-infected cells to uninfected cells at the start of therapy, (ii) the ratio at the start of therapy of the total number of cells $(T+I)$ to the total number of hepatocytes in a healthy liver, T_{max} , (iii) the overall drug effectiveness, ε_{tot} , and (iv) the patient's critical drug efficacy, ε_c . The relationship between these model features and outcome of HBV DNA kinetic patterns are summarized in Table 1. Unfortunately, the model presented here has additional parameters that characterize uninfected and infected cell kinetics, which cannot be estimated from HBV DNA kinetics. Thus, a future challenge will be to collect additional data from liver biopsies about hepatocyte population kinetics both before and during treatment and correlate it with HBV DNA decay profiles under treatment.

Viral rebound can occur, according to the model, when $\varepsilon_{tot} < \varepsilon_c$ and $T + K < T_{max}$ at the start of therapy. If T_{max} represents total hepatocyte number in an uninfected healthy liver then T $+K < T_{\text{max}}$ may not be realistic as it indicates a drastic decrease in hepatocyte number. However, if one assumes that our model is only concerned with a subset of hepatocytes that are susceptible to infection due to their state of differentiation or other host factors, then T_{max} would represent this number of susceptible hepatocytes in a healthy liver and T and I the subsets of these cells that are uninfected and infected during chronic infection. In this case, having $I + T \ll T_{max}$ (pretreatment) may be reasonable as this does not necessarily imply a huge decrease in total hepatocyte number. Experimental results during acute HBV infection (22, 23) or woodchuck hepatitis (24), show that the majority of hepatocytes are HBV infected before the specific immune response against HBV is induced. However, once a cell-mediated immune response is present, HBV DNA is cleared from serum or reaches a viral set point. Recently, using a mathematical model of acute HBV infection in humans, we predicted the appearance of a population of cells refractory to infection or in which viral production is inhibited (25). If, in patients that go on to chronicity, this "antiviral state" is maintained, T_{max} may represent only a small fraction of susceptible cells to HBV infection. This possibility is suggested by studies involving chronically infected patients (26–28) and animals (29) showing that only a small fraction of hepatocytes appear to be infected.

Drug resistance develops over long periods (months to years) of treatment with nucleos(t)ide analogues (NAs) against HBV (1). Virological breakthrough, during this period of time, is

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defined as a $1 \log_{10}$ cp/ml increase in HBV DNA level from nadir in two consecutive samples 1 month apart. However, as shown here, in cases of viral rebound within days or weeks post initiation of therapy and especially under IFN monotherapy or in combination with NAs, in which drug resistance is less likely to develop, an increase of susceptible cells may lead to an increase in HBV replication. In these cases, careful analysis for the nature of viral breakthrough (i.e., drug resistance vs. an increase of susceptible cells) needs to be preformed in order to distinguish between the two processes.

While our new model can successfully explain a large number of observed HBV decay patterns, there are patterns that the model does not explain such as the step-wise viral declines seen with lamivudine alone or in combination with famciclovir (patients 6 and 11 from Lewin et al. (2), respectively) in which there is >1 log decline in viral load from the shoulder phase to the *final* steady-state level that lasts until end of therapy. The model also suggests that from a viral dynamic viewpoint when $\varepsilon_{\text{tot}} > \varepsilon_c$, HBV DNA should ultimately be cleared. Thus, the model does not properly account for the low rate of treatment success (<8%); i.e., loss of HBsAg, seroconversion to anti-HBs and undetectable level of HBV DNA six month after stopping therapy (1). However, the model does not consider covalently closed circular HBV DNA, HbsAg, nor a dynamic immune response to HBV. Future models will need to address these issues.

In summary, we show that by including proliferation of both uninfected and infected cells, a viral kinetic model can account for complex HBV DNA decays observed in treated patients. In addition, we introduced the notion of critical drug efficacy that implies that each patient has a particular viral-host state in which a certain drug effectiveness needs to be imposed in order to reduce HBV viral load below detection. In addition, we predict that viral breakthrough early on treatment can be explained by an increase of susceptible cells without drug resistance development and/or poor treatment compliance. Further studies of viral and host dynamics, based on these results, may be able to better identify the factors giving rise to successful therapy.

Supplementary Material

Refer to Web version on PubMed Central for supplementary material.

Acknowledgments

Financial support: This research was performed under the auspices of the U.S. Department of Energy under contract DE-AC52-06NA25396 and supported by National Institutes of Health (NIH) grant RR06555 (to A.S.P). R.M.R and H.D were supported by grant P20-RR18754 from the National Center for Research Resources (NCRR), a component of the NIH. H.D is supported by the University of Illinois Gastrointestinal and Liver Disease (UIC GILD) Association. This study is solely the responsibility of the authors and does not necessarily represent the official views of NCRR or NIH.

Abbreviations

HBV Hepatitis B virus

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Figure 1.

Complex HBV DNA decline profiles observed in patients during drug therapy. (A) Biphasic decay (patient 16 from Lewin et al. (2), treated with lamivudine and famciclovir), (B) flat 2nd phase (patient 8472 digitized from Colombatto et al. (19); patient was treated with pegylated intereferon-α2a), (C) triphasic decay (Figure 2B digitized from Tsiang et al. (11); patient was treated with adefovir dipivoxil), (D) staircase decay (patient 6 from Lewin et al. (2), treated with lamivudine and famciclovir), and (E) slow rebound (patient 10 from Lewin et al. (2), treated with lamivudine). Experimental data are represented by open circles. Solid lines were used to emphasize plausible phases of viral decay.

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Figure 2.

Model simulations predict complex HBV DNA decline profiles during drug therapy. We solved our model (Eq. 1) numerically using Berkeley-Madonna (version 7.0.2; [http://](http://www.berkeleymadonna.com) [www.berkeleymadonna.com\)](http://www.berkeleymadonna.com). (A) A flat second phase occurs when $\varepsilon_{tot} < \varepsilon_c$ (dashed line; here ε_{tot} =0.88; ε_c =0.934). When ε_{tot} is close to ε_c (but still ε_{tot} < ε_c), then after the rapid viral decay phase an intermediate slower phase is predicted that approaches a flat final phase (thin solid line; ε_{tot} =0.92). When $\varepsilon_{\text{tot}} > \varepsilon_c$ we obtain a biphasic viral decline (thick line; ε_{tot} =0.94). (B) If we decrease the ratio of the target cell to the infected cell replication rate from $r_T/r_T = 8.5$ (used in (A)) to $r_T/r_T \geq 2$, a shoulder phase emerges. Staircase and triphasic viral decay profiles are represented by thin and thick solid lines, respectively. For the staircase profile total drug efficacy, ε_{tot} =0.93, is lower than the critical drug efficacy, ε_c =0.934 and therefore after the third phase of viral decay the viral load stabilizes in a steady state until end of treatment. For the triphasic profile $\varepsilon_{\text{tot}} > \varepsilon_c$ ($\varepsilon_{\text{tot}} = 0.97$) and thus viral will continue to decline under therapy. (C) If we increase the rate constant for cure of infected cells, ρ , from 0 (i.e., no cure) to 0.01 day⁻¹, the shoulder phase of the triphasic decline disappears and we obtain a biphasic viral decline. Here ε_{tot} =0.94 and r_T/r_T = 1.25. (D) Viral rebound during treatment occurs when the hepatocyte replication rate is low,

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r_T=0.51 day⁻¹, *r*₁/*r*_I = 2 and ε_{tot} < ε_{c} (η=0). Except as noted, model parameters used were: $T_{\text{max}} = 1.9 \times 10^7$ cells ml⁻¹, $d_T = 0.004 \text{ day}^{-1}$, $p = 92.3$ virions cell⁻¹ day⁻¹, β= 7.1×10⁻⁹ ml day⁻¹ virions⁻¹, $c = 3.5$ day⁻¹, $\delta = 0.25$ day⁻¹, $\rho = 0$ day⁻¹, $r_T = 3.0$ day⁻¹, $r_T/r_T = 8.5$, $\eta = 0.5$ and $s=1$ cell ml⁻¹ day⁻¹.

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Figure 3.

The model is consistent with experimental data. To show that the model (Eq. 1) agrees with experimental data (as shown in Fig. 1), exhibiting biphasic (A), triphasic (B), flat 2nd phase (C) viral decays, and slow viral rebound (D), the solution for $V(t)$ from Eq. (1) was compared (solid lines) to the HBV DNA (●) using Berkeley-Madonna (version 7.0.2; [http://](http://www.berkeleymadonna.com/) [www.berkeleymadonna.com/\)](http://www.berkeleymadonna.com/). For the theoretical curves, we fixed $T_{max} = 1.9 \times 10^7$ cells ml⁻¹; ρ =0 day⁻¹; s = 1 cell day⁻¹ ml⁻¹. The other parameter values used to generate the theoretical curves in panels (A) to (D), respectively, are: $d_T = 0.2 \times 10^{-3}$, 2.7×10^{-3} , 5.3×10^{-3} , and 9.0×10^{-4} day⁻¹; δ= 0.07, 0.22, 0.25, and 0.06 day⁻¹; β= 1×10^{-10} , 2.5×10^{-7} , 1.9×10^{-6} , and 6.6×10^{-8} ml virions⁻¹ day⁻¹; r_T = 3.0, 0.9, 0.8, and 0.5 day⁻¹; r_T : r_T = 5.0, 3.3, 2.9, and 7.2; $c = 0.6, 0.7, 0.18,$ and 1.0 day^{-1} ; $p = 87, 5, 1.4$, and 164 virions cell⁻¹ day⁻¹; ε = 0.999934, 0.991, 0.9988, and 0.997; η =0.5, 0.5, 0.2, and 0.5; ε_c = 0.75, 0.994, 0.99908, and 0.9987. Previous model predictions (dashed lines) for (A) and (D) were calculated from (2) and for (B) and (C) were digitized from (11) and (19), respectively.

Hepatology. Author manuscript; available in PMC 2013 July 16.

Table 1

HBV DNA kinetic patterns under therapy

