Numerical Modeling of Hepatitis B Dynamics with Vertical Transmission and Treatment

Anmole Razzaq¹, A Raza¹ and M Rafiq²¹Department of Mathematics, National College of Business, Administration and Economics, Gujarat, Pakistan²Faculty of Engineering, University of Central Punjab, Lahore, Pakistan***Corresponding Author:** Anmole Razzaq, Department of Mathematics, National College of Business, Administration and Economics, Gujarat, Pakistan.**Received:** June 25, 2019; **Published:** July 12, 2019**DOI:** 10.31080/ASMS.2019.03.0353

Abstract

Numerical modeling of communicable disease is a device to understand the instrument of how disease blowouts and how it can be measured. In this paper, we have studied the dynamics of hepatitis B with vertical transmission and treatment numerically. We formulate an unconditionally stable non-standard finite difference (NSFD) scheme for a mathematical model of Hepatitis B disease. The developed numerical scheme is bounded, dynamically consistent and preserves the positivity of the solutions. NSFD scheme shows convergence to the true equilibrium points of the model for any time step sizes. But Euler and RK-4 fail for large time step sizes.

Keywords: Hepatitis B Disease; Dynamical System; Numerical Modeling; Convergence

Introduction

Hepatitis B effects the liver [5]. We can say that Hepatitis B is a liver disease that emanates from the infection with Hepatitis B virus (HBV) [1-4]. Its acute infection and chronic infection will be caused [6]. HBV is exposure to infectious blood or body fluids by transmitted.

- HBV from contact with other people's blood during childhood.
- HBV arrived the time of birth.
- HBV interchange from person to person through blood.
- HBV transfer child from mother during delivery.

Approximately 2 billion people infected this virus Hepatitis B, 360 million people effected the chronic HBV. 600,000 peoples die in each year with infection of HBV [5-13].

Mathematical model

Variables and parameters

$S_c(t)$	Denoted by susceptible to chronic infection any time of t.
$U_c(t)$	Denoted by chronic infection any time of t.
$I_c(t)$	Denoted by chronic infection any time of t.
$S_a(t)$	Denoted by susceptible to acute infection any time of t.
$U_a(t)$	Denoted by acute infection any time of t.
$I_a(t)$	Denoted by acute infection any time of t.
$T_a(t)$	Denoted by treatment who infected by acute virus any time of t.
$i=a$	Denoted by adult females and juveniles.

N	Total population size.
λ	Denoted by HBV infection rate.
β	Denoted by contact the rate with infection individuals.
$\wedge$	Denoted by susceptible adult female's rate.
π	Denoted by susceptible juvenile's female's rate.
b	Denoted by increase susceptible juveniles by birth.
p	Denoted by birth from HBV acute females is assumed to susceptible.
$1-p$	Denoted by remaining birth are infected infants who are in acute virus.
γ_a	Denoted by acute status progress in chronic virus.
γ_c	Denoted by adult females and juveniles.
ϵ	Denoted by adult females treated rate.
η	Denoted by adult female's recovery rate and become susceptible.
α	Denoted by treated is not perfect and female may progress to chronic stage rate.
ρ	Denoted by birth from treated adult females is assumed to be susceptible.
$1-\rho$	Denoted by birth from treated adult females is assumed to be susceptible and remaining proportion is infected and join acute virus.
μ	Denoted by adult female experience natural death.
μ_c	Denoted by juvenile's female experience natural death.
δ_a	Denoted by adult female death rate.
δ_c	Denoted by juvenile's female death rate.

The Scheme of Nonlinear Differential Equations(DE) on behalf of the Typical remains specified by:

$$\begin{aligned}\frac{dS_c}{dt} &= bS_a + pbU_a - (\pi + \mu_c)S_c \\ \frac{dU_c}{dt} &= (1-p)bU_a - (\mu_c + \gamma_c)U_c \\ \frac{dI_c}{dt} &= \gamma_c U_c - (\delta_c + \mu_c)I_c \\ \frac{dS_a}{dt} &= \Lambda + \pi S_c - (\lambda + \mu)S_a \quad (1) \\ \frac{dU_a}{dt} &= \lambda S_a - (\mu + \gamma_a + b)U_a \\ \frac{dI_a}{dt} &= \gamma_a U_a - (\delta_a + \mu)I_a\end{aligned}$$

Analysis of the model figure 1

We describe equilibrium points of system i.e Disease free equilibrium(DFE).

$$\mathcal{E}_1 = \left(\frac{\Lambda(\pi + \mu_c)}{\mu(\pi + \mu_c) - \pi b}, \frac{b\Lambda}{\mu(\pi + \mu_c) - \pi b}, 0, 0, 0, 0 \right) \text{ are stability facts of scheme (1),}$$

$$\text{Where } R_0 = \frac{\beta_c \mu(\pi + \mu_c)(\mu_c + \gamma_c)(1 + \theta \gamma_a)}{(\mu(\pi + \mu_c) - \pi b)(\mu + \gamma_a + b)(\delta_a + \mu)}$$

R_0 is a beginning influence who describe the disease of the exit or persist? If $R_0 < 1$ then we say that the scheme will observed disease Free Equilibrium (DFE) and iff $R_0 > 1$ the scheme to involve-ment Endemic Equilibrium (EE).

Numerical modeling

Now we have conferred two standard finite difference structures to unravel the endless dynamical scheme (1) i.e. Euler's Method and Runge-Kutta Method of Order 4.

Euler method

The Forward Euler's Structure for the unceasing model (1) certain through:

$$\begin{aligned}S_c^{n+1} - S_c^n &= h\{bS_a^n + pbU_a^n - (\pi + \mu_c)S_c^n\} \\ U_c^{n+1} - U_c^n &= h\{(1-p)bU_a^n - (\mu_c + \gamma_c)U_c^n\} \\ I_c^{n+1} - I_c^n &= h\{\gamma_c U_c^n - (\delta_c + \mu_c)I_c^n\} \\ S_a^{n+1} - S_a^n &= h\{\Lambda + \pi S_c^n - (\lambda + \mu)S_a^n\} \\ U_a^{n+1} - U_a^n &= h\{\lambda S_a^n - (\mu + \gamma_a + b)U_a^n\} \\ I_a^{n+1} - I_a^n &= h\{\gamma_a U_a^n - (\delta_a + \mu)I_a^n\}\end{aligned}$$

Numerical experiments

Now solve numerical tryouts by expending the values of given parameters Table 1 [6].

Figure 2: Euler Method (DFE), h=0.001.

Figure 3: Euler Method (DFE), h=0.01.

Figure 4: Euler Method (DFE), h=0.1.

Figure 4: Euler Method (DFE), h=0.1.

Parameters	Values
	0.5
	0.4
	10.5
	0.8
	0.7
	0.054
	0.88797
	0.6
	0
	0.03
	0.4
	0.47
	0.04
	0.4
	0
	0
	0

Table 1

Fourth Order Runge-Kutta Scheme

For Stage-1

$$K_1 = h\{bs_a^n + pbU_a^n - (\pi + \mu_c)s_c^n\}$$
$$L_1 = h\{(1 - p)bU_a^n - (\mu_c + \gamma_c)U_c^n\}$$
$$M_1 = h\{\gamma_cU_c^n - (\delta_c + \mu_c)I_c^n\}$$
$$N_1 = h\{\wedge + \pi s_c^n - (\lambda + \mu)s_a^n\}$$
$$O_1 = h\{\lambda s_a^n - (\mu + \gamma_a + b)U_a^n\}$$
$$P_1 = h\{\gamma_aU_a^n - (\delta_a + \mu)I_a^n\}$$

For Stage-2

$$K_2 = h\left[b\left(s_a^n + \frac{N_1}{2}\right) + pb\left(U_a^n + \frac{O_1}{2}\right) - (\pi + \mu_c)\left(s_c^n + \frac{K_1}{2}\right)\right]$$
$$L_2 = h\left[(1 - p)b\left(U_a^n + \frac{O_1}{2}\right) - (\mu_c + \gamma_c)\left(U_c^n + \frac{L_1}{2}\right)\right]$$
$$M_2 = h\left[\gamma_c\left(U_c^n + \frac{L_1}{2}\right) - (\delta_c + \mu_c)\left(I_c^n + \frac{M_1}{2}\right)\right]$$
$$N_2 = h\left[\wedge + \pi\left(s_c^n + \frac{K_1}{2}\right) - (\lambda + \mu)\left(s_a^n + \frac{N_1}{2}\right)\right]$$
$$O_2 = h\left[\lambda\left(s_a^n + \frac{N_1}{2}\right) - (\mu + \gamma_a + b)\left(U_a^n + \frac{O_1}{2}\right)\right]$$
$$P_2 = h\left[\gamma_a\left(U_a^n + \frac{O_1}{2}\right) - (\delta_a + \mu)\left(I_a^n + \frac{P_1}{2}\right)\right]$$

For Stage-3

$$K_3 = h\left[b\left(s_a^n + \frac{N_2}{2}\right) + pb\left(U_a^n + \frac{O_2}{2}\right) - (\pi + \mu_c)\left(s_c^n + \frac{K_2}{2}\right)\right]$$
$$L_3 = h\left[(1 - p)b\left(U_a^n + \frac{O_2}{2}\right) - (\mu_c + \gamma_c)\left(U_c^n + \frac{L_2}{2}\right)\right]$$
$$M_3 = h\left[\gamma_c\left(U_c^n + \frac{L_2}{2}\right) - (\delta_c + \mu_c)\left(I_c^n + \frac{M_2}{2}\right)\right]$$
$$N_3 = h\left[\wedge + \pi\left(s_c^n + \frac{K_2}{2}\right) - (\lambda + \mu)\left(s_a^n + \frac{N_2}{2}\right)\right]$$
$$O_3 = h\left[\lambda\left(s_a^n + \frac{N_2}{2}\right) - (\mu + \gamma_a + b)\left(U_a^n + \frac{O_2}{2}\right)\right]$$
$$P_3 = h\left[\gamma_a\left(U_a^n + \frac{O_2}{2}\right) - (\delta_a + \mu)\left(I_a^n + \frac{P_2}{2}\right)\right]$$

For Stage-4

$$K_4 = h\left[b\left(s_a^n + N_3\right) + pb\left(U_a^n + O_3\right) - (\pi + \mu_c)\left(s_c^n + K_3\right)\right]$$
$$L_4 = h\left[(1 - p)b\left(U_a^n + O_3\right) - (\mu_c + \gamma_c)\left(U_c^n + L_3\right)\right]$$
$$M_4 = h\left[\gamma_c\left(U_c^n + L_3\right) - (\delta_c + \mu_c)\left(I_c^n + M_3\right)\right]$$
$$N_4 = h\left[\wedge + \pi\left(s_c^n + K_3\right) - (\lambda + \mu)\left(s_a^n + N_3\right)\right]$$
$$O_4 = h\left[\lambda\left(s_a^n + N_3\right) - (\mu + \gamma_a + b)\left(U_a^n + O_3\right)\right]$$
$$P_4 = h\left[\gamma_a\left(U_a^n + O_3\right) - (\delta_a + \mu)\left(I_a^n + P_3\right)\right]$$

Finally

$$s_c^{n+1} = s_c^n + \frac{1}{6}[K_1 + 2K_2 + 2K_3 + K_4]$$
$$U_c^{n+1} = U_c^n + \frac{1}{6}[L_1 + 2L_2 + 2L_3 + L_4]$$
$$I_c^{n+1} = I_c^n + \frac{1}{6}[M_1 + 2M_2 + 2M_3 + M_4]$$
$$S_a^{n+1} = S_a^n + \frac{1}{6}[N_1 + 2N_2 + 2N_3 + N_4]$$
$$U_a^{n+1} = U_a^n + \frac{1}{6}[O_1 + 2O_2 + 2O_3 + O_4]$$
$$I_a^{n+1} = I_a^n + \frac{1}{6}[P_1 + 2P_2 + 2P_3 + P_4]$$

Figure 6: RK-4 Method (DFE), h=0.01.

Figure 9: RK-4 Method (DFE), h=10.

Figure 7: RK-4 Method (DFE), h=0.1.

Figure 8: RK-4 Method (DFE), h=1.

Non-standard finite difference model

Now we show an unreservedly convergent non-standard finite difference(NSFD) numerical model which be there describe on non-standard finite difference modeling concept introduced by Micken’s [3]. Now show the covergenence scrutiny of the suggested structure. The NSFD model for the incessant dynamical system is given by:

$$\begin{aligned} S_c^{n+1} &= \frac{s_c^n + h(bs_a^n + pbU_a^n)}{(1+h(\pi + \mu_c))} \\ U_c^{n+1} &= \frac{U_c^n + h(1-p)bU_a^n}{(1+h(\mu_c + \gamma_c))} \\ I_c^{n+1} &= \frac{I_c^n + h\gamma_c U_c^n}{(1+h(\delta_c + \mu_c))} \\ S_a^{n+1} &= \frac{s_a^n + h(\wedge + \pi s_c^n)}{(1+h(\lambda + \mu))} \\ U_a^{n+1} &= \frac{U_a^n + h\lambda s_a^n}{(1+h(\mu + \gamma_a + b))} \\ I_a^{n+1} &= \frac{I_a^n + h\gamma_a U_a^n}{(1+h(\delta_a + \mu))} \end{aligned}$$

Convergence analysis of NSFD scheme

Let us define

$$\begin{aligned} F &= \frac{s_c^n + h(bs_a^n + pbU_a^n)}{(1+h(\pi + \mu_c))} \\ G &= \frac{U_c^n + h(1-p)bU_a^n}{(1+h(\mu_c + \gamma_c))} \\ H &= \frac{I_c^n + h\gamma_c U_c^n}{(1+h(\delta_c + \mu_c))} \\ I &= \frac{s_a^n + h(\wedge + \pi s_c^n)}{(1+h(\lambda + \mu))} \\ J &= \frac{U_a^n + h\lambda s_a^n}{(1+h(\mu + \gamma_a + b))} \\ K &= \frac{I_a^n + h\gamma_a U_a^n}{(1+h(\delta_a + \mu))} \end{aligned}$$

Now the Jacobian Matrix is given by

$$J = \begin{bmatrix} \frac{\partial F}{\partial S_c} & \frac{\partial F}{\partial U_c} & \frac{\partial F}{\partial I_c} & \frac{\partial F}{\partial S_a} & \frac{\partial F}{\partial U_a} & \frac{\partial F}{\partial I_a} \\ \frac{\partial G}{\partial S_c} & \frac{\partial G}{\partial U_c} & \frac{\partial G}{\partial I_c} & \frac{\partial G}{\partial S_a} & \frac{\partial G}{\partial U_a} & \frac{\partial G}{\partial I_a} \\ \frac{\partial S_c}{\partial H} & \frac{\partial S_c}{\partial U_c} & \frac{\partial S_c}{\partial I_c} & \frac{\partial S_c}{\partial S_a} & \frac{\partial S_c}{\partial U_a} & \frac{\partial S_c}{\partial I_a} \\ \frac{\partial H}{\partial S_c} & \frac{\partial H}{\partial U_c} & \frac{\partial H}{\partial I_c} & \frac{\partial H}{\partial S_a} & \frac{\partial H}{\partial U_a} & \frac{\partial H}{\partial I_a} \\ \frac{\partial S_c}{\partial I} & \frac{\partial S_c}{\partial U_c} & \frac{\partial S_c}{\partial I_c} & \frac{\partial S_c}{\partial S_a} & \frac{\partial S_c}{\partial U_a} & \frac{\partial S_c}{\partial I_a} \\ \frac{\partial I}{\partial S_c} & \frac{\partial I}{\partial U_c} & \frac{\partial I}{\partial I_c} & \frac{\partial I}{\partial S_a} & \frac{\partial I}{\partial U_a} & \frac{\partial I}{\partial I_a} \\ \frac{\partial S_c}{\partial J} & \frac{\partial S_c}{\partial U_c} & \frac{\partial S_c}{\partial I_c} & \frac{\partial S_c}{\partial S_a} & \frac{\partial S_c}{\partial U_a} & \frac{\partial S_c}{\partial I_a} \\ \frac{\partial J}{\partial S_c} & \frac{\partial J}{\partial U_c} & \frac{\partial J}{\partial I_c} & \frac{\partial J}{\partial S_a} & \frac{\partial J}{\partial U_a} & \frac{\partial J}{\partial I_a} \\ \frac{\partial S_c}{\partial K} & \frac{\partial S_c}{\partial U_c} & \frac{\partial S_c}{\partial I_c} & \frac{\partial S_c}{\partial S_a} & \frac{\partial S_c}{\partial U_a} & \frac{\partial S_c}{\partial I_a} \\ \frac{\partial K}{\partial S_c} & \frac{\partial K}{\partial U_c} & \frac{\partial K}{\partial I_c} & \frac{\partial K}{\partial S_a} & \frac{\partial K}{\partial U_a} & \frac{\partial K}{\partial I_a} \end{bmatrix}$$

At DiseaseFree Equilibrium

$$\mathcal{E}_1 = (\frac{\wedge(\pi+\mu_c)}{\mu(\pi+\mu_c)-b\pi}, \frac{b\wedge}{\mu(\pi+\mu_c)-b\pi}, 0,0,0,0)$$

There are the following eigen values of above jacobian matrix is:

$$\Rightarrow \lambda_1 = \frac{1}{(1+h(\pi+\mu_c))} < 1$$
$$\Rightarrow \lambda_2 = \frac{1}{(1+h(\mu_c+\gamma_c))} < 1$$
$$\Rightarrow \lambda_3 = \frac{1}{(1+h(\delta_c+\mu_c))} < 1$$
$$\Rightarrow \lambda_4 = \frac{1}{(1+h(\lambda+\mu))} < 1$$
$$\Rightarrow \lambda_5 = \frac{1}{(1+h(\mu+\gamma_a+b))} < 1$$
$$\Rightarrow \lambda_6 = \frac{1}{(1+h(\delta_a+\mu))} < 1$$

So,
The scheme is LAS.

Numerical experiments

Figure 10: NSFD Method (DFE), h=0.01.

Figure 11: NSFD Method (DFE), h=1.

Figure 12: NSFD Method (DFE), h=100.

Figure 13: NSFD Method (DFE), h=10000.

Comparison analysis

In this section, we see the comparison among of two standard difference schemes and non-standard difference scheme in epidemiology

Figure 14: Comparison (DFE), h=0.1.

Figure 15: Comparison (DFE), h=1.

Figure 16: Comparison (DFE), h=5.

Figure 17: Comparison (DFE), h=100.

Figure 18: Comparison (DFE), h=1000.

Figure 19: Comparison (EE), h=100000.

Results and Discussion

The model of transmission dynamics of Hepatitis B virus disease consumes introduced expending PSIT Model. (i.e Threatened, Susceptible, Infected and Treated). The constancy of solid positions i.e the Disease free equilibrium(DFE) deliberated numerically. We describe an unqualifiedly constant Non-Standard Finite Difference (NSFD) structure aimed at the incessant dynamical system. The suggested structure exists dynamical consistant, numerically steady and holds all the athentic assets of the incessant model. The outcomes equaled well known standard finite difference schemes i.e Euler’s and Runge-Kutta method of order 4 (RK-4). The Euler and RK-4 are provisionally convergent and diverge of the assured ethics of step size ‘h’ while the constructed NSFD scheme for every assessment used to residues convergent.

Conclusion

The non - standard finite difference scheme created for the communication dynamics of HBV is unconditionally convergent. Applied schemes in this paper like Euler and RK-4 are unsuccessful they depend on step size. So, Euler and RK-4 are conditionally convergent. When we intensify the step size, the graph of Euler and RK-4 are divergent and from time to time give variation in solution. The new advanced numerical scheme like non-standard finite difference scheme is independent on step size. Uncertainty we intensify the step size in hundreds and thousands then NSFD motionless convergent. The NSFD scheme is informal implement that gives mathematically stable, positivity, bounded-ness and shows an equal behaviour of the continuous model and discrete model.

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