

Stability Analysis of a Schistosomiasis Transmission Model with Control Strategies

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Abstract—We have established and rigorously analyzed a new mathematical model that describes the dynamics schistosomiasis infection. This model incorporates several realistic features including density-dependent births rate of snails and reduced fecundity in snail hosts. Our qualitative analysis of the deterministic model is made with respect to the stability of the disease free equilibrium and the unique endemic equilibrium. Some biological consequences and control strategies are discussed. We have derived the basic reproduction number above which the infection will be controlled under certain levels. We have shown that the disease free equilibrium is globally asymptotically stable when the basic reproduction number \mathcal{R}_0 is less than one. We have proved the existence and global asymptotic stability of an endemic steady state when $\mathcal{R}_0 > 1$. This mathematical analysis of the model gives insight about the effects of the reduced fecundity and intermediate host density-dependent birth rate.

Finally, numerical simulations are performed to illustrate the main results.

Keywords—Epidemic models; Nonlinear dynamical systems; Monotone systems; Global stability; Reproduction number; Schistosomiasis.

I. INTRODUCTION

Mathematical modeling of the spread of infectious diseases is an important instrument in the comprehension of the dynamics of diseases and in decision making processes regarding intervention programs for controlling these diseases in many countries.

The high prevalence of infectious diseases as schistosomiasis has prompted the mathematical modelers to deploy multiple infectious disease models in recent years, and different mechanisms have been suggested to explain their occurrence. Schistosomiasis is transmitted by human contact with contaminated fresh water (lakes and ponds, rivers, dams) inhabited by snails carrying the parasite. Many people die from schistosomiasis disease every year. Schistosomiasis is the most deadly NTD (Neglected Tropical Diseases), killing an estimated 280,000 people each year [20]. It can result in liver damage and anemia, especially among children [22].

It is found mostly in rural areas in tropical and subtropical countries and infects humans and other vertebrates, using snails in most cases as intermediate hosts. Infection takes place when

larvae of the parasite released by the intermediate hosts, penetrate the skin of an individual when in contact with infested water. In the body, the larvae develop and cross over into adult schistosomes. These parasites live in the blood vessels, where the females lay their eggs. Some eggs out of the body via the feces or urine and the parasitic life cycle continues. Others are trapped in the tissues of the body, causing an immune reaction in active lesions and organs.

A possible method of struggle against schistosomiasis is to treat water bodies with a molluscicide to minimize the number of snail intermediate hosts, thus breaking the cycle of disease transmission [23], [24]. Beside this one, there are other such that biological control, chemical molluscicides, chemotherapy, and more permanent methods such as the provision of safe water and sanitary facilities [1], [6]. Control strategies focused on intermediate hosts which are the sites of intense proliferation of this parasite are seen as a priority of the reduction of schistosomiasis transmission [28].

Many researchers studied the dynamics of schistosomiasis using systems similar to the one we are interested [2], [7], [13], [14], [15], [16], [21]. That systems depend on the epidemiological formulation, but also on the demographic process incorporated into the model under different control strategies. Almost all of the work to date on schistosomiasis transmission has assumed constant immigration in the snail population [25], [36] or linear birth rates [26], [1], [27] not allowing the possibility of the reduced fecundity of the infected population snails. A drawback of the models with birth and death rates proportional to the size of the population are that the population size decreases or increases exponentially, except in the special case where births exactly balance deaths. Moreover, researchers in laboratory reported the influence of a trematode on life history traits of adult *Lymnaea elodes* snails. It has been displayed reduced fecundity relative to uninfected snails [29]. To our knowledge, due to various considerations of the factors related to the transmission of the

disease, impaired fertility and density-dependent birth rate are not taken account simultaneously with chemotherapy and chemical molluscicides as control strategies.

We shall introduce and analyze a compartmental model, which considers a host (human), intermediate host (snail) as well as free-living cercariae and miracidia and their interaction. The model focuses essentially on the aquatic life stages of the parasite. A full mathematical analysis of the model is derived. The aim focus of this paper is to study the effect of chemotherapy combined with chemical molluscicide as control strategies on the dynamics of the model with density-dependent birth rate and impaired fertility in snail host.

The paper is organized as follows. In Section II we start by defining the mathematical framework we use and focus on the different processes of transmission that might be appropriate to understand schistosomiasis. The main results are developed in Section III, including the determination of \mathcal{R}_0 the basic reproductive number of the model and the analysis of its stability properties. We show that the disease-free equilibrium is global asymptotic stability when the basic reproductive number is less than one. When the basic reproductive number is larger than one, we prove the global asymptotic stability of a unique endemic equilibrium by using some properties of K-monotone systems (see [17]). Section IV is devoted to the numerical analysis and control strategies.

Finally, in Section V, concluding remarks close the paper.

II. MODEL DERIVATION

We build an evolutionary outcomes model of interactions between a complex life-cycle parasite *schistosoma* and its hosts (humans and snails). The parasite populations at the free-living stages are modeled explicitly through miracidia and cercariae. The model consists of a system of ordinary differential equations.

Furthermore, it is admitted that the infected snails did not recover from schistosomiasis that their lifetimes are short compared to that of humans. The model sub-divides the total human

population at time t , denoted by $N_h(t)$, into the following sub-populations of susceptible individuals ($S_h(t)$) and individuals with schistosomiasis symptoms ($I_h(t)$). So that

$$N_h(t) = S_h(t) + I_h(t).$$

The total snails population at time t , denoted by $N_s(t)$, is sub-divided into susceptible snails ($S_s(t)$) and infectious snails ($I_s(t)$). Thus,

$$N_s(t) = S_s(t) + I_s(t).$$

Let W_m and W_p be the population of miracidia and cercariae, respectively. This model assumes a constant per capita rate of exposure between hosts and sensitive parasites, where exposure and susceptibility to infection are regrouped in the transmission coefficient. We considered the mass action transmission model. We follow of the available model for schistosomiasis [30] by incorporating the human interaction.

Susceptible host snails increase through density-dependent births with maximum rate b_s and competitive intensity c . Assume that individuals are born uninfected. Infected hosts suffer from reduced fecundity ($0 \leq \rho < 1$). Susceptible snails die at background death rate d_s , due to parasite natural death rate d_{se} and the elimination rates θ_s of snails and become infected at the per capita infection rate β_s , through contact with infectious miracidia. Infected snails die at an elevated death rate, d_s , due to parasite natural death rate d_{se} and the elimination rates θ_s of snails at which is added the rate α due to parasite virulence. The susceptible human population is increased by the recruitment of individuals in the population (assumed susceptible), at a rate Λ . The population of individuals is further decreased by natural death (at a rate d_h). It is also assumed that infected individuals have additional host mortality during the given short time considered at the rate μ . Susceptible humans become infected only through contact with free-living pathogen cercariae in infested water at the per capita infection rate β_h and recover at the per capita rate η . Free-living miracidia are introduced into the aquatic environment at a per capita rate k , but they are depleted

during the infection process at the per capita rate δ and die naturally if they do not find snails to infect at the rate d_m . Here, depletion of parasites through transmission depends on total host density. Infected snails will then free up second form of larvae called cercariae at a rate γ to be able to infect humans. Some cercariae also die at an elevated death rate d_c due to the natural death rate d_{ce} and cercariae elimination θ_c by the chemical molluscicide.

The time evolution of the different populations is governed by the following system of equations:

$$\left\{ \begin{aligned} \frac{dS_s}{dt} &= b_s (S_s + \rho I_s)(1 - c(S_s + I_s)) \\ &\quad - \overbrace{(d_{se} + \theta_s)}^{d_s} S_s - \beta_s S_s W_m, \\ \frac{dI_s}{dt} &= \beta_s S_s W_m - (d_{se} + \theta_s + \alpha) I_s, \\ \frac{dW_m}{dt} &= k I_h - \delta (S_s + I_s) W_m - d_m W_m, \\ \frac{dW_c}{dt} &= \gamma I_s - \overbrace{(d_{ce} + \theta_c)}^{d_c} W_c, \\ \frac{dS_h}{dt} &= \Lambda - \beta_h W_c S_h - d_h S_h + \eta I_h, \\ \frac{dI_h}{dt} &= \beta_h W_c S_h - (d_h + \mu + \eta) I_h. \end{aligned} \right. \tag{1}$$

For convenience of simplicity, we denote

$$d_s = d_{se} + \theta_s, \quad d_c = d_{ce} + \theta_s.$$

The dynamics of the total snails population ($N_s = S_s + I_s$) is governed by

$$\begin{aligned} \frac{dN_s}{dt} &= b_s (S_s + \rho I_s) (1 - c N_s) - d_s N_s - \alpha I_s \\ &\leq b_s N_s - b_s c N_s^2 - d_s N_s \\ &\leq \left(\frac{b_s - d_s}{b_s c} - N_s \right) b_s c N_s. \end{aligned}$$

The dynamics of the total humans population ($N_h = S_h + I_h$) is governed by

$$\frac{dN_h}{dt} = \Lambda - d_h N_h - \mu I_h \leq \frac{\Lambda}{d_h}.$$

Thus the region

$$\mathcal{D} = \{(S_s, I_s, W_m, W_c, S_h, I_h) \in \mathbb{R}_+^6 : N_s \leq \frac{b_s - d_s}{c b_s}, W_m \geq 0, W_c \geq 0, N_h \leq \frac{\Lambda}{d_h}\}$$

is a positively invariant set of system (1).

Furthermore, the model (1) is well-posed epidemiologically and we will consider dynamic behavior of model (1) on \mathcal{D} .

III. MAIN RESULTS

In this section, we firstly derive the basic reproduction number for system (1) by the method of next generation matrix formulated in [9], [5]. The basic reproduction number \mathcal{R}_0 is used to assess the stability of the disease free equilibrium and the endemic equilibrium. Then we derive the asymptotic behavior of the system (1) by its limiting system. Discussions of the relation of the dynamics between (1) and its limiting system can be found in [32], [33], [34]. In particular, there holds the following significant result :

Lemma III.1. Consider the following two systems

$$\frac{dx}{dt} = f(t, x), \quad \frac{dy}{dt} = g(y),$$

where $x, y \in \mathbb{R}^n$, f and g are continuous, satisfy a local Lipschitz condition in any compact set $\Omega \in \mathbb{R}^n$, and $f(t, x) \rightarrow g(x)$ as $t \rightarrow \infty$, so that the second system is the limit system for the first system. Let $\Phi(t, t_0, x_0)$ and $\tilde{\Phi}(t, x)$ be solutions of these systems, respectively. Suppose that $p \in \Omega$ is a locally asymptotically stable equilibrium of the limit system and its attractive region is

$$W(p) = \{y \in \Omega | \tilde{\Phi}(t, t_0, y) \rightarrow p, t \rightarrow +\infty\}.$$

Let W_Φ be the omega limit set of $\Phi(t, t_0, x_0)$. If $W_\Phi \cap W(p) \neq \emptyset$, then $\lim_{t \rightarrow +\infty} \Phi(t, t_0, x_0) = p$.

This limiting system is the key to understanding the global dynamics of system (1). By a method due to Castillo-Chavez et al [31], the stability of the disease free equilibrium of the proposed limiting system is discussed. Moreover, the stability of positive equilibrium of a proposed limiting system of model (1) is analyzed theoretically thanks to monotone dynamical system theory [11].

A. The reproduction number

The reproduction number is the expected number of secondary cases produced in a completely susceptible population by a typical infective individual. It is easy to see that system (1) admits always a disease-free equilibrium, $E_0 = (\frac{b_s - d_s}{b_s c}, 0, 0, 0, \frac{\Lambda}{\mu})$.

Let $x = (I_s, I_h, W_m, W_p, S_s, S_h)^T$. Then system (1) can be written as

$$x' = \mathcal{F}(x) - \mathcal{V}(x),$$

$$\text{where } \mathcal{F} = \begin{pmatrix} \beta_s S_s W_m \\ \beta_h S_h W_c \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}, \quad \text{and}$$

$$\mathcal{V} = \begin{pmatrix} (d_s + \alpha) I_s \\ (\mu + d_h + \eta) I_h \\ -k I_h + \delta (S_s + I_s) W_m + d_m W_m \\ -\gamma I_s + d_c W_c \\ -b_s (S_s - \rho I_s)(1 - c (S_s + I_s)) + d_s S_s + \beta_s S_s W_m \\ -\Lambda + \beta_h W_c S_h + \mu I_h + \eta I_h \end{pmatrix}.$$

We can get

$$F = \begin{pmatrix} 0 & 0 & \frac{\beta_s (b_s - d_s)}{b_s c} & 0 \\ 0 & 0 & 0 & \frac{\beta_h \Lambda}{\mu} \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix} \quad \text{and}$$

$$V = \begin{pmatrix} \alpha + d_s & 0 & 0 & 0 \\ 0 & \eta + \mu & 0 & 0 \\ 0 & -k & \frac{\delta (b_s - d_s)}{b_s c} + d_m & 0 \\ -\gamma & 0 & 0 & d_c \end{pmatrix}$$

FV^{-1} is the next generation matrix for model (1). It then follows that the spectral radius

of matrix FV^{-1} is $\rho(FV^{-1})$.

According to Theorem 2 in [18], the basic reproduction number of model (1) is

$$\mathcal{R}_0 = \rho(FV^{-1}) = \sqrt{\frac{\beta \gamma k \Lambda (b_s - d_s) \beta_h}{d_h d_c (d_s + \alpha) (\eta + \mu + d_h) (b_s (c d_m + \delta) - \delta d_s)}}$$

B. The local stability of the DFE

Before proceeding with the mathematical analysis of the model (1), the following simplification would be made. To make the mathematical analysis more tractable, we will consider the case where the the disease-induced death and the parasite virulence parameters in the model will be set to zero. That is, from now on, it is assumed that $\alpha = \mu = 0$ (it should be mentioned that such assumption may not be appropriate in modeling the transmission dynamics in hardest hit areas). However, this assumption will be relaxed in the numerical simulations section.

Without loss of generality (see [33], [32]), we assume that our population of humans has reached its limiting value, i.e, $N_h^* \equiv \frac{\Lambda}{d_h}$ (with $\alpha = \mu = 0$).

By eliminating the equation for $\frac{dS_h}{dt}$, we get from (1) the equivalent limiting system :

$$\begin{cases} \frac{dS_s}{dt} = b_s (S_s + \rho I_s)(1 - c(S_s + I_s)) - d_s S_s - \beta_s S_s W_m, \\ \frac{dI_s}{dt} = \beta_s S_s W_m - d_s I_s, \\ \frac{dW_m}{dt} = k I_h - \delta (S_s + I_s) W_m - d_m W_m, \\ \frac{dW_c}{dt} = \gamma I_s - d_c W_c, \\ \frac{dI_h}{dt} = \beta_h W_c (N_h^* - I_h) - (d_h + \eta) I_h. \end{cases} \tag{2}$$

defined on

$$\mathcal{D}_0 = \{(S_s, I_s, W_m, W_c, I_h) \in \mathbb{R}_+^5 : N_s \leq \frac{b_s - d_s}{c b_s}, W_m \geq 0, W_c \geq 0, I_h \leq \frac{\Lambda}{d_h}\}$$

We now study the local behavior of the disease free equilibrium for system (2), which also gives the analogous behavior for system (1).

Proposition III.1. *The DFE for limiting system (2) E_0 is LAS (locally asymptotically stable) if $\mathcal{R}_0 < 1$ with $\rho \leq 1 - d_s/b_s$ and is unstable if $\mathcal{R}_0 > 1$.*

Proof: The Jacobian matrix of (2) at E_0 is

$$J_0 = \begin{pmatrix} J_{11} & J_{12} \\ J_{21} & J_{22} \end{pmatrix}$$

where,

$$J_{11} = \begin{pmatrix} d_s - b_s (\rho + 1) & -b_s \\ 0 & -d_s \end{pmatrix},$$

$$J_{12} = \begin{pmatrix} \frac{\beta_s (d_s - b_s)}{b_s c} & 0 & 0 \\ \frac{\beta_s (b_s - d_s)}{b_s c} & 0 & 0 \end{pmatrix}, J_{21} = \begin{pmatrix} 0 & 0 \\ 0 & \gamma \\ 0 & 0 \end{pmatrix},$$

$$J_{22} = \begin{pmatrix} \frac{\delta d_s - b_s (\delta + c d_m)}{b_s c} & 0 & k \\ 0 & -d_c & 0 \\ 0 & N_h \beta_h & -\eta - d_h \end{pmatrix}.$$

Let us do the following transformation:

$$J_{s0} = T.J_0.T$$

where,

$$T = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 & 0 \\ 0 & 0 & -1 & 0 & 0 \\ 0 & 0 & 0 & -1 & 0 \\ 0 & 0 & 0 & 0 & -1 \end{pmatrix}.$$

Then, J_{s0} and J_0 are similar. J_{s0} is a Metzler matrix and we can write $J_{s0} = F + V$ with

$$F = \begin{pmatrix} 0 & b_s - d_s & \frac{\beta_s (b_s - d_s)}{b_s c} & 0 & 0 \\ 0 & 0 & \frac{\beta_s (b_s - d_s)}{b_s c} & 0 & 0 \\ 0 & 0 & 0 & 0 & k \\ 0 & \gamma & 0 & 0 & 0 \\ 0 & 0 & 0 & N_h \beta_h & 0 \end{pmatrix},$$

$$V = \begin{pmatrix} d_s - b_s - \rho d_s & 0 & 0 & 0 \\ 0 & -d_s & 0 & 0 \\ 0 & 0 & \frac{\delta d_s - b_s(\delta + c d_m)}{b_s c} & 0 \\ 0 & 0 & 0 & -d_c \\ 0 & 0 & 0 & 0 \end{pmatrix} - \eta I_5.$$

We have $F > 0$ and V is Metzler stable, see [3], [12], [10], [4]. Thanks to Varga's Theorem in [19]: $s(J_0) \leq 0$ iff $\rho(-F V^{-1}) \leq 1$. Since $\rho(-F V^{-1}) = \mathcal{R}_0^4$, we then deduce that E_0 is LAS if $\mathcal{R}_0 < 1$ and is unstable if $\mathcal{R}_0 > 1$. ■

C. Global stability of the DFE

The object of this subsection is to prove the GAS result using the second generation approach given in Castillo-Chavez et al [31]. Herein stated to be self-contained.

Theorem III.1. *If the system (2) can be written in the form*

$$\begin{aligned} \frac{dX}{dt} &= F(\mathbf{x}, Z), \\ \frac{dZ}{dt} &= G(X, Z), G(\mathbf{x}, Z) = 0, \end{aligned}$$

where the components of $X \in \mathbb{R}^m$ denotes

the number of uninfected individuals and the components of $Z \in \mathbb{R}^n$ the number of infected individuals. let $E_0 = (\mathbf{x}^*, 0)$ be the disease free equilibrium of the system. Assume that

- For $\frac{dX}{dt} = F(X, 0)$, X^* is globally asymptotically stable (GAS),
- $G(X, Z) = A Z - \hat{G}(X, Z)$, $\hat{G}(X, Z) \geq 0$ for $(X, Z) \in \mathcal{D}$, where $A = D_Z G(X^*, 0)$ is an M-Matrix (the off diagonal of A are nonnegative) and \mathcal{D} the region where the model makes biological sense.

Then the fixed $E_0 = (\mathbf{x}^*, 0)$ is a globally asymptotically stable equilibrium of model system (2) provided that $\mathcal{R}_0 < 1$.

Applying Theorem III.1 to system (2) gives

$$\hat{G}(X, Z) = \begin{bmatrix} W_m \beta_s (S^* - S_s) \\ W_m \delta (I_s + S_s - S_s^*) \\ 0 \\ W_c I_h \beta_h \end{bmatrix}.$$

Since, $S \leq S^* \equiv N_h^*$ thus $\hat{G}(X, Z) \geq 0$, and by Theorem III.1, E_0 is GAS.

We summarize the result in the following theorem:

Theorem III.2. *If $\mathcal{R}_0 < 1$ then the DFE is GAS.*

D. Local Stability of the Endemic Equilibrium

We proceed to investigate the local stability of the model. We use the following limiting system (3) to obtain the information for the whole system.

Since the population size, $N_s(t)$ (with $\alpha = \mu = 0$) also satisfies

$$N_s \rightarrow N_s^* := \frac{b_s - d_s}{c b_s} \text{ as } t \rightarrow \infty.$$

Using results from Lemma III.1, we can get analytical results by considering the limiting system of (3) in which the total population of snails and humans both have reached the limiting states. Then, we obtain the reduced limiting dynamical system:

$$\begin{cases} \frac{dI_s}{dt} = \beta_s (N_s - I_s) W_m - d_s I_s, \\ \frac{dW_m}{dt} = k I_h - \delta N_s^* W_m - d_m W_m, \\ \frac{dW_c}{dt} = \gamma I_s - d_c W_c, \\ \frac{dI_h}{dt} = \beta_h W_c (N_h^* - I_h) - (d_h + \eta) I_h. \end{cases} \quad (3)$$

The dynamics of system (1) can be focused on this restricted region

$$\mathcal{D}_1 = \{(I_s, W_m, W_c, I_h) \in \mathbb{R}_+^4 : I_s \leq \frac{b_s - d_s}{c b_s}, W_m \geq 0, W_c \geq 0, I_h \leq \frac{\Lambda}{d_h}\}$$

Clearly, the system (1) is asymptotic to the limiting system (3). Thus, the dynamics of system (1) are qualitatively equivalent to the dynamics of its limiting system. The variation matrix of system (3) at E^* an equilibrium point is

$$J(E^*) = \begin{pmatrix} J_{11}(E^*) & J_{12}(E^*) \\ J_{21}(E^*) & J_{22}(E^*) \end{pmatrix},$$

where,

$$J_{11}(E^*) = \begin{pmatrix} -W_m^* \beta_s - d_s & (N_s^* - I_s^*) \beta_s \\ 0 & -N_s^* \delta - d_m \end{pmatrix},$$

$$J_{12}(E^*) = \begin{pmatrix} 0 & 0 \\ 0 & k \end{pmatrix},$$

$$J_{21}(E^*) = \begin{pmatrix} \gamma & 0 \\ 0 & 0 \end{pmatrix},$$

$$J_{22}(E^*) = \begin{pmatrix} -d_c & 0 \\ (N_h^* - I_h^*)\beta_h & -\eta - W_c^* \beta_h - d_h \end{pmatrix}$$

Theorem III.3. *If $\mathcal{R}_0 > 1$, then the positive endemic equilibrium E^* of the limiting system (3) is locally asymptotically stable.*

Before proving Theorem III.3, we state the following useful lemma :

Lemma III.2. (Kamkang [35], Proposition 3.1) *Let M be a square Metzler matrix written in block form $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$, with A and D square matrices. M is Metzler stable if only if matrices A and $D - C A^{-1} B$ are metzler stable.*

Proof of Theorem III.3: $J(E^*)$ is a Metzler matrix and we can write : $J(E^*) = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$, where

$$A = \begin{pmatrix} -W_m^* \beta_s - d_s & (N_s^* - I_s^*)\beta_s \\ 0 & -N_s^* \delta - d_m \end{pmatrix},$$

$$D = \begin{pmatrix} -d_c & 0 \\ (N_h^* - I_h^*)\beta_h & -\eta - W_c^* \beta_h - d_h \end{pmatrix}.$$

$$B = \begin{pmatrix} 0 & 0 \\ 0 & k \end{pmatrix}, C = \begin{pmatrix} \gamma & 0 \\ 0 & 0 \end{pmatrix}.$$

Clearly, A is a stable Metzler matrix. Then, after some computations, we get

$$D - C A^{-1} B = \begin{pmatrix} M_{11} & M_{12} \\ M_{21} & M_{22} \end{pmatrix}.$$

where,

$$M_{11} = -d_c, \\ M_{12} = \frac{k(N_s^* - I_s^*)\beta_s \gamma}{W_m^* N_s^* \beta \delta + N_s^* d_s \delta + W_m^* \beta d_m + d_s d_m}$$

$$M_{21} = (N_h^* - I_h^*)\beta_h, M_{22} = -\eta - W_c^* \beta_h - d_h.$$

$D - C A^{-1} B$ is a stable Metzler matrix iff

$$\chi := \beta \gamma k \beta_h (I_h^* - N_s^*) (N_h^* - I_h^*) + d_c (d_m + \delta N_s^*) (d_s + \beta W_m^*) \cdot (W_c^* \beta_h + \eta + d_h) > 0.$$

The endemic equilibrium satisfies

$$\begin{cases} \beta_s (I_s - N_s) W_m & = -d_s I_s, \\ W_c & = \frac{\gamma}{d_c} I_s. \\ k I_h & = (\delta N_s + d_m) W_m. \end{cases} \quad (4)$$

Hence,

$$\begin{aligned} \chi &= -\gamma k \beta_h \frac{d_s}{W_m} I_s (N_h^* - I_h^*) + d_c (d_m + \delta N_s^*) \cdot (d_s + \beta_s W_m^*) (W_c^* \beta_h + \eta + d_h) \\ &> -\gamma k \beta_h \frac{d_s}{W_m} I_s (N_h^* - I_h^*) + d_c (d_m + \delta N_s^*) d_s \beta_h \frac{\gamma I_s}{d_c} \\ &> d_s \beta_h \gamma I_s \left[-\frac{k}{W_m} (N_h^* - I_h^*) + (d_m + \delta N_s^*) \right] \\ &> d_s \beta_h \gamma I_s \left[-\frac{k}{W_m} I_h^* + (d_m + \delta N_s^*) \right] \\ &> d_s \beta_h \gamma I_s \left[-\frac{W_m (\delta N_s + d_m)}{W_m} I_h^* + (d_m + \delta N_s^*) \right] > 0, \end{aligned}$$

which implies $J(E^*)$ is a Metzler stable matrix. Thus the unique endemic equilibrium is locally asymptotically stable. ■

E. Global Stability of the Endemic Equilibrium

In this section we will establish the global stability of the unique endemic equilibrium point when $\mathcal{R}_0 > 1$. We shall use the properties of K-monotone systems for the analysis of our system (see [17]).

Theorem III.4. *If $\mathcal{R}_0 > 1$, then the positive endemic equilibrium state E^* of the limiting system (3) is globally asymptotically stable in the interior of the set \mathcal{D}_1 .*

Proof: It is useful to be able to test monotonicity directly in terms of vector fields. We recall here a wider criterion to define monotonicity. A system $X = f(X)$ is said to be monotone in the set \mathcal{K} if there exists a diagonal matrix $H = \text{diag}(\epsilon_1, \epsilon_2, \dots, \epsilon_n)$ where each ϵ_i is 1 or -1 such that the matrix product $H Df(X) H$ has only non positive values outside the diagonal for any X (see Smith [17], Lemma 2.1). In the above statement, $Df(X)$ is the Jacobian matrix of f and \mathcal{K} is a convex set. The Jacobian of the system (3), $J(E^*)$, is a Metzler matrix and is irreducible, which implies the strong monotonicity of the system under a usual order defined by the orthant

$$\mathcal{K} = \{(I_s, W_m, W_c, I_h) \in \mathbb{R}_+^4 : I_s, W_m, W_c, I_h > 0\}$$

The significant result of convergence proved by Hirsh in the late 1980s can be described in our framework as follows : given an autonomous system that is strongly monotone with respect to some proper cone and assuming that there is a unique equilibrium in an open set of points with compact orbit closures, every initial condition with bounded solution converges to the unique equilibrium; (see Hirsh [11], theorem 10.3).

Thanks Hirsch’s theorem and the fact that we have only one endemic equilibrium E^* in \mathcal{D}_1 which is locally asymptotically stable when $\mathcal{R}_0 > 1$ we can conclude that E^* is globally asymptotically stable in \mathcal{D}_1 when $\mathcal{R}_0 > 1$. ■

IV. NUMERICAL ANALYSIS AND CONTROL STRATEGIES

In this section, we present some numerical simulation results to confirm our analytical predictions on the global dynamics of the schistosomiasis models. The two selected control strategies identified in this study against the spread of schistosomiasis are snails reduction strategies and

chemotherapy treatments. Snail reduction strategies include the elimination of snails as well as the elimination of free-living cercariae using appropriate biological agents. The strategy of applying chemotherapy treatments requires an effective drug delivery as praziquantel to infected humans. In order to examine the effects of these two classes of anti-schistosomiasis strategies, the model (1) is simulated using the set of parameter values given in Table I.

The effect of chemical molluscicide is incorporated in our model by increasing the death rate parameter of the snails and free-living cercariae at the rate θ_s and θ_c , respectively. In other words, a higher values in θ_s and θ_c serve to reduce the number of snails and free-living cercariae, and thus can be used to investigate the effects of chemical molluscicide on the epidemic. Similarly, the effects of the drug administration method are modelled by increasing a recover rate η to the schistosomiasis.

The basic reproduction number can be applied to measuring the control efforts needed to reduce or eliminate an infection. Furthermore, since \mathcal{R}_0 is a decreasing function of θ_s , θ_c and η , the use of any preventive strategy that can increase θ_s , θ_c or η results in a reduction of schistosomiasis infections.

We, consequently, seek to compare the effects of chemical molluscicide and drug administration on the control of the spread of schistosomiasis in humans. We first consider the original no-control model by setting $\eta = \theta_s = \theta_c = 0$ in system (1). We set the population size of initial conditions as $S_s(0) = 10^4$, $I_s(0) = 5 \times 10^4$; $W_m(0) = 5 \times 10^3$; $W_c(0) = 9 \times 10^3$; $S_h(0) = 2 \times 10^3$; $I_h(0) = 550$, and take the values of the parameters from table I. It should be mentioned that some of these parameters (e.g. θ_s , θ_c , η) are estimated in the convenience use. We pick the following conclusions for the parameter values used.

The results in Fig. 1 show a marked increase in the number of asymptotically infected snails and infected humans at steady state with no control strategies ($\theta_s \approx \theta_c \approx \eta \approx 0$). With the same configurations, we now add control measures and simulate the control model (1). The simulation

TABLE I
PARAMETER VALUES ESTIMED

Parameter	Definition	Default value	Literature source
b_s	Maximum birth rate of hosts	0.06 day ⁻¹	[30]
ρ	Relative fecundity of infected hosts	0.75 day ⁻¹	--
c	Strength of density dependence on host birth rate	0.025 L day ⁻¹	--
d_h	Natural mortality rate of humans	0.0000384 day ⁻¹	--
d_s	Natural mortality rate of snails	0.003 day ⁻¹	--
μ	Parasite dependent mortality of humans	0.0039 day ⁻¹	--
d_m	Loss rate of free-living miracidia	2.5 day ⁻¹	--
d_c	Loss rate of free-living cercariae	2 day ⁻¹	--
β_s	Rate at which the miracidia successfully infects a snail	0.615 L mir ⁻¹ day ⁻¹	--
β_h	Rate at which susceptible humans get infected	0.406 L mir ⁻¹ day ⁻¹	[36]
α	Parasite virulence on survival	0.007 day ⁻¹	[30]
γ	Per capita production rate of cercariae by infected snails	100 cerc host ⁻¹ day ⁻¹	[36]
k	Per capita production rate of miracidia by infected humans	0.00232 day ⁻¹	[36]
η	Treatment rate of infected human	0.03 day ⁻¹	Estimated
δ	Per capita depletion rate of miracidia	0.0039 day ⁻¹	[30]
θ_s, θ_c	elimination rates of snails and cercariae, respectively	0.05 day ⁻¹	Estimated
Λ	Recruitment rate for humans	8000 day ⁻¹	[36]

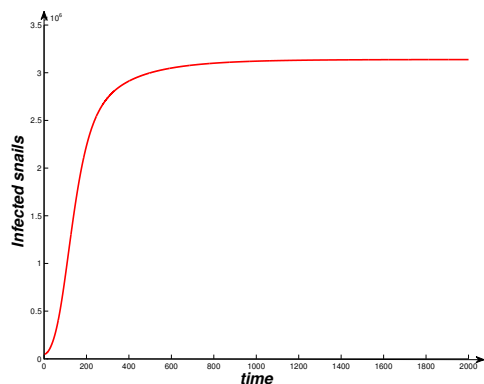
shown in Figure 2 demonstrate the effectiveness of the control strategies adopted. We observe that an epidemic outbreak occurs for some period of time. For instance, if human recover at the rate $\eta = 0.9$ and snails be eliminated at the rate $\eta = 0.05$, then the number of asymptotically infected humans tend to zeros after a period of time $t \approx 1050$ days. Mention should be made here of the fact that, in the last scenario, we have values for each of the following parameters η, θ_s, θ_c , so that $\mathcal{R}_0 < 1$. Fig. 3 and 4 also show that the efficacy of snail elimination can be important for such strategies to make a meaningful impact in combatting schistosomiasis in humans.

V. SUMMARY AND CONCLUSIONS

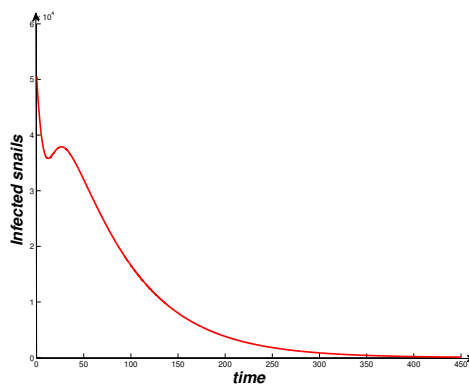
In this paper, we have presented a stability analysis of a schistosomiasis infection model that explicitly includes density dependent births rate and impaired fecundity in the snail host population. The model also captures the effect of two control strategies: chemotherapy and snail

elimination by molluscicides. Six sub population sizes were considered: human host susceptible and infected, snail intermediate host susceptible and infected, free-living miracidia and cercariae. It is shown that the combination of chemotherapy and snail elimination can be effective control strategy that is there will not be an epidemic.

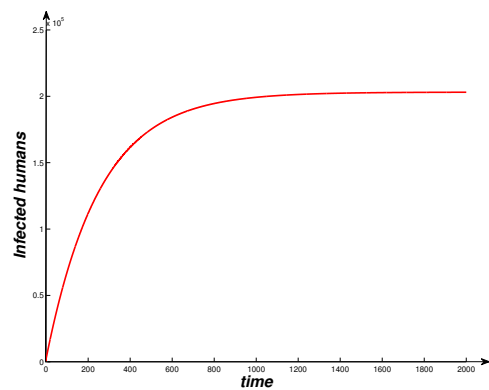
Mathematical properties of the model are analyzed and used to reduce the dimension of the system under consideration. The reproductive number \mathcal{R}_0 is then analytically and explicitly computed. We proved that the disease-free steady state E_0 is globally asymptotically stable if $\mathcal{R}_0 \leq 1$. We have also established the global asymptotic stability of the endemic equilibrium E^* when it exists i.e., when $\mathcal{R}_0 > 1$ using some properties of monotone systems.



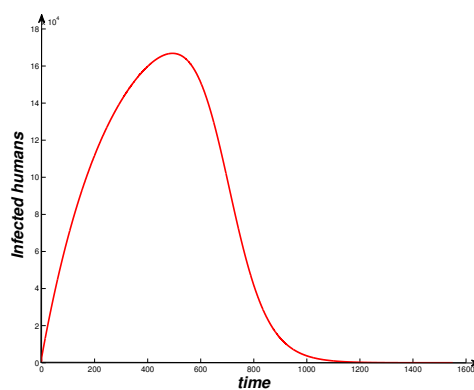
(a)



(a)



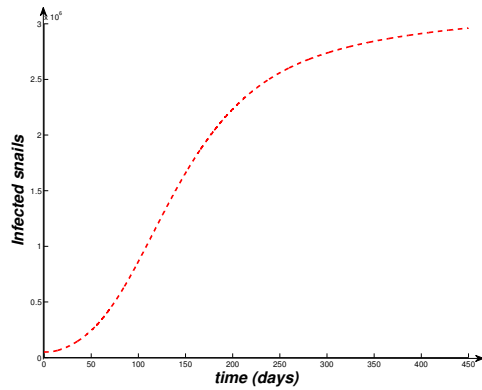
(b)



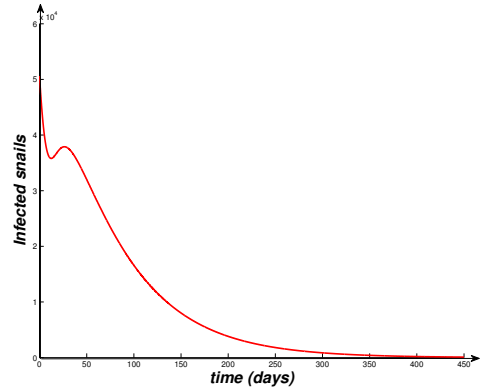
(b)

Fig. 1. Time evolution of the number of infected snails population (a) and humans population (b) without control strategies with parameter values defined in Table I: $\theta_c = 0$, $\theta_s = 0$, $\eta = 0$. These parameters correspond to $\mathcal{R}_0 > 1$. The initial condition is $S_s = 10000$, $I_s = 50000$, $W_m = 5000$, $W_c = 9000$, $S_h = 2000$, $I_h = 550$.

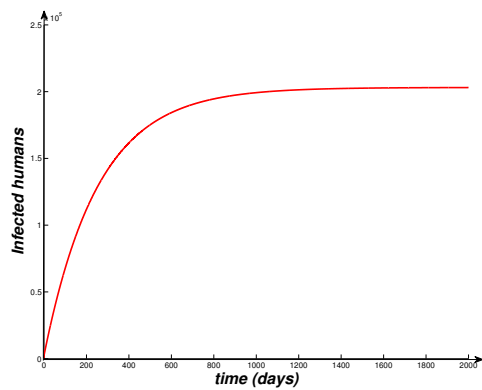
Fig. 2. Effect of combined human treatment and snails elimination on snails population (a) and humans population behavior (b) with parameter values defined in Table I: $\theta_c = 10$, $\theta_s = 0.05$, $\eta = 0.90$. These parameters correspond to $\mathcal{R}_0 < 1$. The initial condition is $S_s = 10000$, $I_s = 50000$, $W_m = 5000$, $W_c = 9000$, $S_h = 2000$, $I_h = 550$.



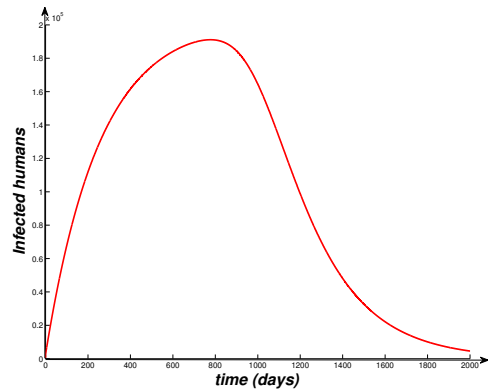
(a)



(a)



(b)



(b)

Fig. 3. Effect of human treatment only, by drug administration, on snails population (a) and humans population behavior (b) with parameter values defined in Table I: $\theta_c = 0$, $\theta_s = 0$, $\eta = 0.90$. The initial condition is $S_s = 10000$, $I_s = 50000$, $W_m = 5000$, $W_c = 9000$, $S_h = 2000$, $I_h = 550$.

Fig. 4. Effect of snails elimination only, by molluscicide, on snails population (a) and humans population behavior (b) with parameter values defined in Table I: $\theta_c = 10$, $\theta_s = 0.05$, $\eta = 0$. The initial condition is $S_s = 10000$, $I_s = 50000$, $W_m = 5000$, $W_c = 9000$, $S_h = 2000$, $I_h = 550$.

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