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Stephen Ekwueme Ekwueme , Obiora Cornelius Collins , Ifeanyi Sunday Onah *

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Analysis and Optimal Control Measures of a Typhoid Fever Mathematical Model for a Multiple Socio-Economic Population

Stephen Ekwueme [1,†,‡], Obiora Collins [2,‡] and Ifeanyi Onah [3,‡*]

[1] Department of Mathematics, University of Nigeria, Nsukka; stephen.aniaku@unn.edu.ng
[2] Institute of Systems Science, Durban University of Technology, Durban 4000, South Africa.; obiora.c.collins@gmail.com
[3] School of Mathematics and Statistics, Mathematics and Statistics Building, University of Glasgow, United Kingdom, G12 8QW
* Correspondence: i.onah.1@research.gla.ac.uk; +44-7513-337-217 (Onah I.S.)

Abstract: Typhoid fever is an infectious disease that affects humanity worldwide particularly in the lower socio-economic communities where many individuals are exposed to a dirty environment and unclean food. A mathematical model is formulated to analyze the impact of control measures such as vaccination of susceptible humans, treatment of infected humans and sanitation in the different socio-economic communities. The model assumed that the population comprises of two socio-economic classes. The essential dynamical system analysis of our model was appropriately carried out. The impact of the control measures was analyzed, and the optimal control theory was applied on the control model to explore the impact of the different control measures. Numerical simulation of the models and the optimal controls were carried out and results obtained indicates that the overall combination of the control measures eradicates Typhoid fever in the population but the controls are more optimal in the higher socio-economic communities.

Keywords: typhoid fever; reproduction number; stability analysis; optimal control; numerical analysis

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1. Introduction

Typhoid fever is a life-threatening infection that originated from the bacterium *Salmonella Typhi* and is a major cause of illness and mortality in regions of the world with limited access to treated water and sanitation [1,2]. Recent statistics show that an average of 15 million cases and 145 000 typhoid-related deaths occur annually worldwide and its more endemic in many developing countries, and despite recent interventions in sanitation coverage, the disease remains a significant public health problem [1,3]. People who are infected with Salmonella Typhi, often referred to as "typhoid carriers," shed the bacteria in their feces (stool) and, to a lesser extent, in their urine. These individuals may have symptoms of typhoid fever or be asymptomatic carriers. The transmission of the disease is primarily a result of poor sanitation and lack of clean drinking water, but can also be transmitted via person-to-person on unclean surfaces [1]. Symptoms of typhoid fever include headache, weakness, loss of appetite, prolonged fever, nausea and constipation, or sometimes diarrhea [1].

In developing countries, the public health goal of preventing and controlling typhoid disease through sanitation and adequate medical care is challenging. Besides that, delays in diagnosis and treatment occur due to barriers to medical care, such as difficulty accessing medical facilities caused by delayed referral, the distance to hospital, and the cost of getting healthcare. The socio-economic class (SEC) of individuals has been shown to influence the dynamics of some infectious diseases [4–6,17]. Since typhoid fever is linked with poor sanitation and unclean water, individuals in the SEC are expected to be more exposed to typhoid fever compared to the individuals in a higher SEC [17]. In this work, we analyze the influence of control measures on the dynamics of typhoid fever disease

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for multiple socio-community. Specifically, we look at a case when the community consists of two socio-economic classes (i.e, lower SEC and higher SEC).

Several control measures have been implemented in fighting typhoid fever. Some of the effective ones include sanitation, vaccination, and treatment [1]. Medically, each of these three control measures (sanitation, vaccination, and treatment) are independent and hence can be applied simultaneously. Others include accurate diagnosis at the right time and treatment of typhoid disease in the community to prevent hospitalization complications. Prevention of the disease through improving sanitation and accessibility to safe water and food, health education to increase public awareness and induce healthy behavioral changes [3,7–9].

A mathematical model is a powerful approach to learning how disease spreads among both human and animal populations. Mathematical models help us to understand the transmission dynamics quantitatively and allow us to check hypotheses to understand their importance [10]. Interpreting a mathematical model for an infectious disease requires some assumptions about the spreading infection mechanism [11]. The basic compartmental mathematical models to describe the spread of disease are introduced [12,13]. This approach may test and compare different disease interventions as a strategy to prevent and control the disease. The origin of a mathematical model for typhoid disease could be related to Branko Cvjetanovi´c, an assistant professor at the Zagreb School of Medicine, who implemented the medical trial of the first typhoid vaccine that was funded by the US Public Health Service and WHO in the 1950s [16]. He was involved in research about the vaccines for diphtheria, pertussis, tetanus, cholera, and typhoid. In the 1973s, Cvjetanovi´c noted that no controlled trials had been run to demonstrate the extent to which typhoid transmission can control the spread through various sanitation strategies. Moreover, some researchers developed a non-autonomous mathematical model [14,15] to study typhoid transmission by considering the effect of seasonal conditions and some time-dependent parameters.

The epidemiology of some infectious disease has been studied extensively using mathematical models [17–26,32,35]. In this study, we utilize a mathematical model to ascertain the influence of control measures in reducing typhoid fever in a diverse socio-economic community.

2. Model formulation

A multiple socio-economic community with total human inhabitant N is considered. We assume that the community is made up of two socio-economic classes whose sub-population is N_i. Suppose there is a typhoid fever outbreak within the two socio-economic classes of the community. Assumed that each of these socio-economic community (N_i) engages three control measures (vaccination, treatment, and sanitation) in fighting the disease. Based on these assumptions, the formulation of mathematical model requires that the total population (N_i) for each socio-economic class is partitioned into susceptible population $S_i(t)$, vaccinated population $V_i(t)$, infected population $I_i(t)$, treated population $T_i(t)$ and recovered population $R_i(t)$. The variable $P_i(t)$ represented the pathogen in the environment for each SEC i. Epidemiologically, the transmission of typhoid fever disease is either through contact with infected humans or through exposure to the bacteria causing the illness. Recruitment of individuals into each of the susceptible class $S_i(t)$ occur at a rate $\mu_i N_i(t)$. Individuals in each $S_i(t)$ moves to $V_i(t)$ as they get vaccinated at a rate ϕ_i. Direct transmission from $I_i(t)$ to $S_i(t)$ and $V_i(t)$ occur at a rate β_i while the indirect transmission from $P_i(t)$ to $S_i(t)$ and $V_i(t)$ occur at a rate α_i. Note that the vaccinated individuals have less chances of being infected because they are vaccinated. This is captured in the model by assuming that the efficacy of the vaccine is ε_i for SEC i. Each infected class $I_i(t)$ get treated at a rate σ_i. The treated class $T_i(t)$ recovers at a rate ρ_i. The $I_i(t)$ who did not get treatment can recover naturally at a rate γ_i. Note that we discourage not getting treatment because typhoid fever can be very fatal and there are available treatment for the disease. Natural death occur at each of the SEC $N_i(t)$ at a rate μ_i. Each of the recovered class $R_i(t)$ can loss immunity and become susceptible again at a rate φ_i. Susceptible individuals move from $S_i(t)$ to $S_j(t)$ at a rate k_{ij} whereas infected individuals move from $I_i(t)$ to $I_j(t)$ at a rate b_{ij}. Infected individuals $I_i(t)$ shed pathogens

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into the environment $P_i(t)$ at a rate δ_i and the pathogen $P_i(t)$ decay at a rate ξ. Sanitation enhances pathogens $P_i(t)$ decay at a rate θ_i. Based on these explanations, we obtained the typhoid fever control model given by

$$\begin{cases} \frac{dS_1(t)}{dt} = N_1(t)\mu_1 - (\beta_1 I_1(t) + \alpha_1 P_1(t))S_1(t) - (\mu_1 + \phi_1)S_1(t) \\ \qquad + \varphi_1 R_1(t) - k_{12}S_1(t) + k_{21}S_2(t), \\ \frac{dV_1(t)}{dt} = \phi_1 S_1(t) - (1-\varepsilon_1)(\beta_1 I_1(t) + \alpha_1 P_1(t))V_1(t) - \mu_1 V_1(t), \\ \frac{dI_1(t)}{dt} = (\beta_1 I_1(t) + \alpha_1 P_1(t))S_1(t) + (1-\varepsilon_1)(\beta_1 I_1(t) + \alpha_1 P_1(t))V_1(t) \\ \qquad - (\mu_1 + \gamma_1 + \sigma_1)I_1(t) - b_{12}I_1(t) + b_{21}I_2(t), \\ \frac{dT_1(t)}{dt} = \sigma_1 I_1(t) - (\mu_1 + \rho_1)T_1(t), \\ \frac{dR_1(t)}{dt} = \gamma_1 I_1(t) + \rho_1 T_1(t) - (\mu_1 + \varphi_1)R_1(t), \\ \frac{dP_1(t)}{dt} = \delta_1 I_1(t) - (\xi + \theta_1)P_1(t), \\ \frac{dS_2(t)}{dt} = N_2(t)\mu_2 - (\beta_2 I_2(t) + \alpha_2 P_2(t))S_2(t) - (\mu_2 + \phi_2)S_2(t) \\ \qquad + \varphi_2 R_2(t) + k_{12}S_1(t) - k_{21}S_2(t), \\ \frac{dV_2(t)}{dt} = \phi_2 S_2(t) - (1-\varepsilon_2)(\beta_2 I_2(t) + \alpha_2 P_2(t))V_2(t) - \mu_2 V_2(t), \\ \frac{dI_2(t)}{dt} = (\beta_2 I_2(t) + \alpha_2 P_2(t))S_2(t) + (1-\varepsilon_2)(\beta_2 I_2(t) + \alpha_2 P_2(t))V_2(t) \\ \qquad - (\mu_2 + \gamma_2 + \sigma_2)I_2(t) + b_{12}I_1(t) - b_{21}I_2(t), \\ \frac{dT_2(t)}{dt} = \sigma_2 I_2(t) - (\mu_2 + \rho_2)T_2(t), \\ \frac{dR_2(t)}{dt} = \gamma_2 I_2(t) + \rho_2 T_2(t) - (\mu_2 + \varphi_2)R_2(t), \\ \frac{dP_2(t)}{dt} = \delta_2 I_2(t) - (\xi + \theta_2)P_2(t). \end{cases} \tag{1}$$

The variables and parameters meanings can be found in Table 1 and 2 respectively.

Table 1. Meaning of variables in model 1

Variable	Meaning
$N_i(t)$	Total population of individuals in SEC i
$S_i(t)$	Susceptible population in SEC i
$V_i(t)$	Vaccinated population in SEC i
$I_i(t)$	Infected population in SEC i
$T_i(t)$	Treated population in SEC i
$R_i(t)$	Recovered population in SEC i
$P_i(t)$	Pathogens in the environment in SEC i

Table 2. Meaning of parameters used in model 1

Parameter	Meaning
β_i	Contact rate of susceptible with infected population in SEC i
α_i	Contact rate of susceptible population with pathogens in SEC i
μ_i	Natural mortality rate of humans in the SEC i
ϕ_i	Vaccination rate of individuals in the SEC i
ε_i	Efficacy of vaccination in the SEC i
γ_i	Natural recovery rate of infected population in SEC i
γ_i	Recovery rate of infected population due to treatment in SEC i
σ_i	Treatment rate of infected population in SEC i
φ_i	Rate at which recovered population becomes susceptible in SEC i
δ_i	Shedding rate of $P_i(t)$ by the infected population in SEC i
ξ	Natural death rate of pathogens in the environment
θ_i	Decay rate of $P_i(t)$ due to sanitation
k_{ij}	Movement rate of susceptible population from $S_i(t)$ to $S_j(t)$
b_{ij}	Movement rate of infected population from $I_i(t)$ to $I_j(t)$

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Let the initial conditions of the multiple control model be assumed as:

$$S_i(0) > 0,\ V_i(0) > 0,\ I_i(0) \geq 0,\ T_i(0) \geq 0,\ R_i(0) \geq 0,\ P_i(0) > 0,\ \ i = 1, 2. \tag{2}$$

3. Model analysis

In this section, we present the dynamical system analysis of the multiple control model 1. The analysis will improve our understanding of typhoid fever disease dynamics. Mathematically, there exists a unique disease-free equilibrium (DFE) for the multiple control model 1

$$(S_1^0, V_1^0, I_1^0, T_1^0, R_1^0, P_1^0, S_2^0, V_2^0, I_2^0, T_2^0, R_2^0,\ P_2^0) = \left(S_1^0,\ V_1^0,\ 0,\ 0,\ 0,\ 0,\ S_2^0,\ V_2^0,\ 0,\ 0,\ 0,\ 0\right), \tag{3}$$

where $S_1^0 = \frac{k_{21}N}{\varphi_1 k_{12}+\varphi_2 k_{21}}$, $S_2^0 = \frac{k_{12}N}{\varphi_1 k_{12}+\varphi_2 k_{21}}$, $V_1^0 = \frac{\phi_1 S_1^0}{\mu_1}$, $V_2^0 = \frac{\phi_2 S_2^0}{\mu_2}$, $\varphi_1 = \frac{\mu_1+\phi_1}{\mu_1}$ and $\varphi_2 = \frac{\mu_2+\phi_2}{\mu_2}$.

The basic reproduction number for the multiple control model 1 can be referred to as the expected number of new infections of typhoid fever produced when an infected individual is brought into contact with the population susceptible to typhoid fever in the presence of vaccination and sanitation. Mathematically, the basic reproduction number of model 1, using the next generation matrix approach [20] is

$$\mathcal{R}_0 = \frac{\mathcal{R}_{11} + \mathcal{R}_{44} + \sqrt{(\mathcal{R}_{11} + \mathcal{R}_{44})^2 + 4(\mathcal{R}_{14}\mathcal{R}_{41} - \mathcal{R}_{11}\mathcal{R}_{44})}}{2}, \tag{4}$$

where $\mathcal{R}_{11} = \frac{(\beta_1 S_1^0+(1-\varepsilon_1)\beta_1 V_1^0)\psi_2}{\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12}} + \frac{(\alpha_1 S_1^0+(1-\varepsilon_1)\alpha_1 V_1^0)\delta_1\psi_2}{(\xi+\theta_1)(\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12})}$, $\mathcal{R}_{14} = \frac{(\beta_1 S_1^0+(1-\varepsilon_1)\beta_1 V_1^0)b_{21}}{\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12}} + \frac{(\alpha_1 S_1^0+(1-\varepsilon_1)\alpha_1 V_1^0)\delta_1 b_{21}}{(\xi+\theta_1)(\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12})}$, $\mathcal{R}_{41} = \frac{(\beta_2 S_2^0+(1-\varepsilon_2)\beta_2 V_2^0)b_{12}}{\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12}} + \frac{(\alpha_2 S_2^0+(1-\varepsilon_2)\alpha_2 V_2^0)\delta_2 b_{12}}{(\xi+\theta_2)(\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12})}$, $\mathcal{R}_{44} = \frac{(\beta_2 S_2^0+(1-\varepsilon_2)\beta_2 V_2^0)\psi_1}{\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12}} + \frac{(\alpha_2 S_2^0+(1-\varepsilon_2)\alpha_2 V_2^0)\delta_2\psi_2}{(\xi+\theta_2)(\varrho_1\varrho_2+\varrho_1 b_{21}+\varrho_2 b_{12})}$, $\varrho_1 = \mu_1 + \gamma_1 + \sigma_1$, $\varrho_2 = \mu_2 + \gamma_2 + \sigma_2$, $\psi_1 = \varrho_1 + b_{12}$ and $\psi_2 = \varrho_2 + b_{21}$.

Epidemiologically, when $\mathcal{R}_0 < 1$, the disease can be eradicated from the two socio-economic classes. This can be shown by proving that the disease free equilibrium is stable when $\mathcal{R}_0 < 1$ [20–22]. This implies that the control measures ensure that the basic reproduction number is less than unity so that the disease will not be established in any of the socio-economic class in the community. On the contrary, if the control measures are not effectual in decreasing $\mathcal{R}_0$ below unity, a typhoid fever outbreak is likely to occur. The outbreak may persist or remain endemic in either or both socio-economic classes of the population [18,20–22]. Further investigation on the influence of the control measures on typhoid fever disease dynamics is considered via numerical illustrations in the subsequent section.

4. Numerical illustrations

Numerical illustrations are presented here to analyze the influence of the control measures on typhoid fever disease dynamics for the diverse socio-economic community. The parameter values used in the numerical illustrations are given in Table 3.

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Table 3. Parameter values used for the numerical illustrations

Symbol of the parameters	Parameter values	Source
μ_i	0.0200	[18,23]
β	0.00002	Estimated
β_1	1.6 β	Estimated
β_2	0.4 β	Estimated
α	0.00001	Estimated
α_1	1.6 α	Estimated
α_2	0.4 α	Estimated
φ	0.001	Estimated
φ_1	0.4 φ	Estimated
φ_2	1.6 φ	Estimated
γ	0.0445	[21]
γ_1	0.4 γ	Estimated
γ_2	1.6 γ	Estimated
ξ	0.0333	[18,21]
k_{12}	0.20	[17]
k_{21}	0.20	[17]
b_{12}	0.20	[17]
b_{21}	0.20	[17]
ε	0.78	[34]
ε_1	0.4 ε	Estimated
ε_2	1.6 ε	Estimated
ϕ_1	0.20	Estimated
ϕ_2	0.80	Estimated
θ_1	0.04	Estimated
θ_2	0.16	Estimated
σ_1	0.18	Estimated
σ_2	0.72	Estimated
δ	ξ	[21]
δ_1	1.6 ξ	Estimated
δ_2	0.4 ξ	Estimated

Vaccination is one of the effective control measures for minimizing typhoid fever [1]. Figure 1 illustrates the influence of vaccination rate in the population. We observe from the figure that increasing vaccination rates lead to a decrease in infected humans in both socio-economic classes. We observe that the infected populace is greater in the lower SEC 1 in the presence of vaccination. Hence, to achieve disease eradication, this lower SEC 1 should be the main target of vaccination.

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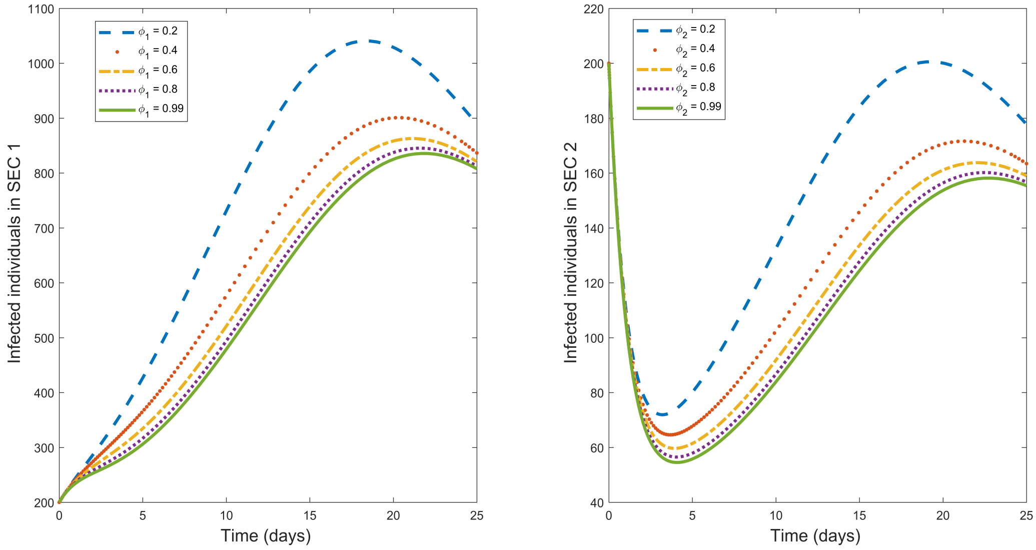

Figure 1. Plot illustrating the effects of vaccination rate ϕ_i on the dynamics of typhoid fever infections in SEC 1 and SEC 2.

Vaccine efficacy is a major factor in vaccination that determines the percentage reduction of the disease in a vaccinated group. Figure 2 illustrates the impact of vaccine efficacy ε_i on the dynamics of typhoid fever. The figure shows that an increase in vaccine efficacy decreases typhoid fever infected humans in the entire community. Hence, considering a vaccination with a very high efficacy (say 99% as we have in Figure 2) will results in faster disease eradication in the two socio-economic classes if the vaccine is applied uniformly in the entire populace.

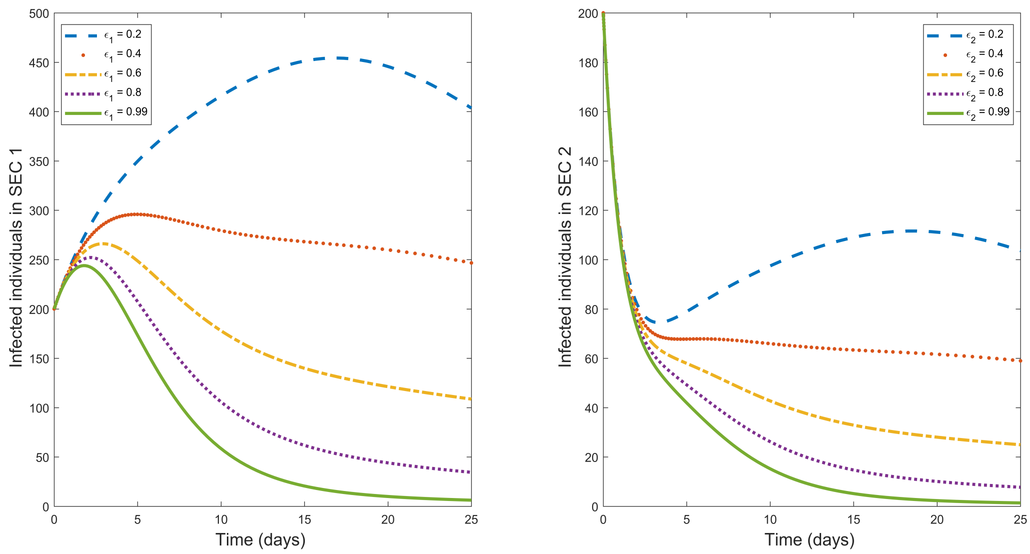

Figure 2. Plot illustrating the impact of vaccine efficacy ε_i on the dynamics of typhoid fever infections in SEC 1 and SEC 2.

Typhoid fever can be treated with appropriate antibiotic medicine [1]. Treatment of infected individuals is one of the effective control measures for reducing typhoid fever infections. Figure 3 is a graphical illustration of the effect of treatment rate in decreasing typhoid fever. The figure illustrates

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that an increase in treatment rate results in to decrease in typhoid fever in both socio-economic classes. Based on this, effective treatment of infected humans is recommended in the entire population.

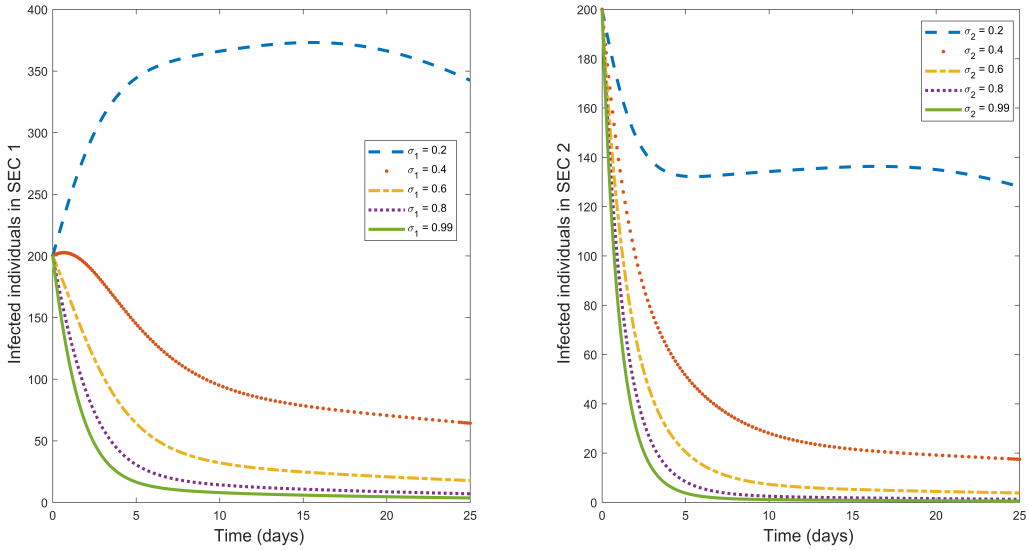

Figure 3. Plot illustrating the influence of treatment σ_i on the dynamics of typhoid fever infections in SEC 1 and SEC 2.

Contaminated food and environment are one of the major routes of contracting typhoid fever [1]. So, to reduce typhoid fever infections, sanitation should be maintained in society. Figure 4 is a graphical representation of the impact of sanitation θ_i on the dynamics of typhoid fever. From the figure, we observe that an increase in sanitation results in a decrease in typhoid fever infected humans. The effects of sanitation are less in the higher SEC 2. A possible explanation for this could be because the higher SEC 2 have a certain level of sanitation in their environment, so introducing what is already in existence in their environment will not lead to major results unlike in the lower SEC 1 that have limited access to sanitation. Based on these, the lower SEC 1 should be the target of sanitation for maximum results.

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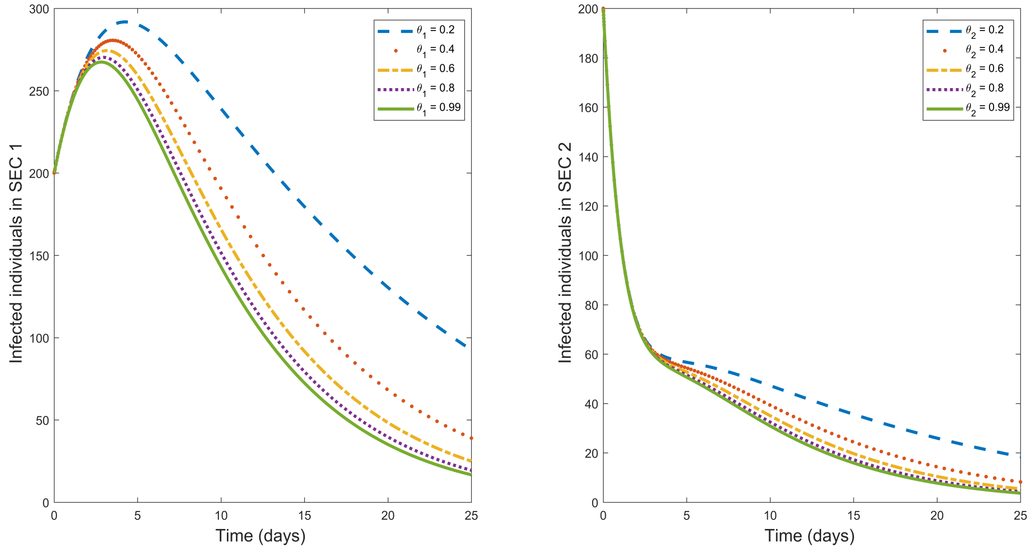

Figure 4. Plot illustrating the influence of sanitation θ_i on the dynamics of typhoid fever infections in SEC 1 and SEC 2.

Multiple control measures in this study are the situation when different possible control measures are introduced simultaneously in fighting a particular disease. In this study, we have discussed three possible control measures that can be used in fighting typhoid fever outbreaks. Figure 5 describes the effects of introducing these three control measures in fighting typhoid fever. The figure shows that using multiple control measures has maximum influence in decreasing the infected population (in both socio-economic classes) when compared with any of the single control measures. Therefore, whenever a typhoid fever outbreak occurs, multiple control measures should be considered for the faster eradication of the disease.

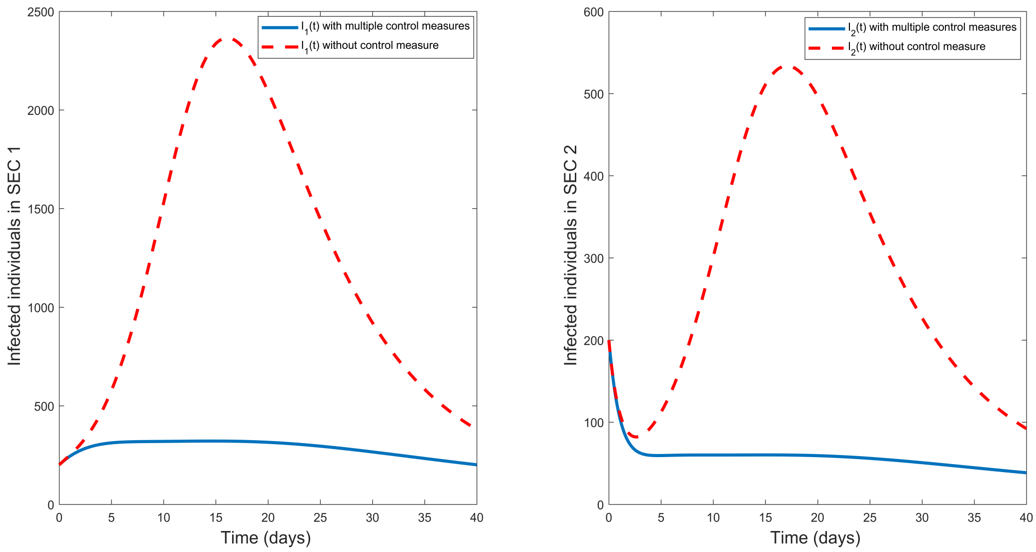

Figure 5. Plot illustrating the impact of multiple control measures on the dynamics of typhoid fever infections in SEC 1 and SEC 2.

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5. Optimal control analysis

Qualitative and numerical analysis of our model showed that implementing multiple control measures as aforementioned plays major roles in reducing the influence of Typhoid fever among humanity. Here, we intend to carryout optimal control analysis to determine the most effective control strategy for minimizing the number of humans affected by Typhoid fever along different socioeconomic classes. To minimize the cost of implementing the controls. we assume that the control parameters ϕ_i, σ_i and θ_i denoting vaccination, treatment and sanitation, respectively are measurable functions of time and then we formulate an appropriate optimal control function that minimizes the cost of implementing the controls subject to the model (1). For simplicity, we write the control strategies as control functions given as $\phi_i = u_i(t)$, $\sigma_i = v_i(t)$ and $\theta_i = w_i$ which are bounded, lebesgue integrable functions. Given the above, we now write the optimal control model as

$$\begin{cases} \frac{dS_1(t)}{dt} = N_1(t)\mu_1 - (\beta_1 I_1(t) + \alpha_1 P_1(t))S_1(t) - (\mu_1 + u_1)S_1(t) \\ \qquad + \varphi_1 R_1(t) - k_{12}S_1(t) + k_{21}S_2(t), \\ \frac{dV_1(t)}{dt} = u_1 S_1(t) - (1-\varepsilon_1)(\beta_1 I_1(t) + \alpha_1 P_1(t))V_1(t) - \mu_1 V_1(t), \\ \frac{dI_1(t)}{dt} = (\beta_1 I_1(t) + \alpha_1 P_1(t))S_1(t) + (1-\varepsilon_1)(\beta_1 I_1(t) + \alpha_1 P_1(t))V_1(t) \\ \qquad - (\mu_1 + \gamma_1 + v_1)I_1(t) - b_{12}I_1(t) + b_{21}I_2(t), \\ \frac{dT_1(t)}{dt} = v_1 I_1(t) - (\mu_1 + \rho_1)T_1(t), \\ \frac{dR_1(t)}{dt} = \gamma_1 I_1(t) + \rho_1 T_1(t) - (\mu_1 + \varphi_1)R_1(t), \\ \frac{dP_1(t)}{dt} = \delta_1 I_1(t) - (\xi + w_1)P_1(t), \\ \frac{dS_2(t)}{dt} = N_2(t)\mu_2 - (\beta_2 I_2(t) + \alpha_2 P_2(t))S_2(t) - (\mu_2 + u_2)S_2(t) \\ \qquad + \varphi_2 R_2(t) + k_{12}S_1(t) - k_{21}S_2(t), \\ \frac{dV_2(t)}{dt} = u_2 S_2(t) - (1-\varepsilon_2)(\beta_2 I_2(t) + \alpha_2 P_2(t))V_2(t) - \mu_2 V_2(t), \\ \frac{dI_2(t)}{dt} = (\beta_2 I_2(t) + \alpha_2 P_2(t))S_2(t) + (1-\varepsilon_2)(\beta_2 I_2(t) + \alpha_2 P_2(t))V_2(t) \\ \qquad - (\mu_2 + \gamma_2 + v_2)I_2(t) + b_{12}I_1(t) - b_{21}I_2(t), \\ \frac{dT_2(t)}{dt} = v_2 I_2(t) - (\mu_2 + \rho_2)T_2(t), \\ \frac{dR_2(t)}{dt} = \gamma_2 I_2(t) + \rho_2 T_2(t) - (\mu_2 + \varphi_2)R_2(t), \\ \frac{dP_2(t)}{dt} = \delta_2 I_2(t) - (\xi + w_2)P_2(t), \end{cases} \tag{5}$$

subject to the initial conditions $S_1(0) = S_1^0$, $V_1(0) = V_1^0$, $I_1 = I_1^0$, $T_1(0) = T_1^0$, $R_1(0) = R_1^0$, $S_2(0) = S_2^0$, $V_2(0) = V_2^0$, $I_2(0) = I_2^0$, $T_2(0) = T_2^0$, $R_2(0) = R_2^0$.
This implies that the optimal control model is said to be optimal if it minimizes the objective functional

$$\mathcal{J}(u_i, v_i, w_i) = \int_0^T \sum_{i=1}^{2} \left\{ A_i S_1 + B_i I_i + C_i P_i + d_i u_i^2 + e_i v_i^2 + f_i w_i^2 \right\} dt, \tag{6}$$

subject to the model (5), where the coefficients A_i, B_i, C_i, d_i, e_i and f_i are cost balancing coefficients that transform the integral into money expended over time T. Here, A_i, is the direct cost associated with reducing the number of susceptibility to disease in each SEC, B_i is the direct cost associated with reducing the number of infected humans in each SEC and C_i is the direct cost associated with reducing the number of bacteria in the environment, while d_i, e_i and f_i are relative costs for enforcing the control strategies u_i, v_i, w_i. The goal is to minimise the number of humans susceptible to Typhoid fever among different SECs, minimize the infectious humans in all SECs and minimize the bacteria that causes Typhoid. In doing this, we anticipate nonlinear costs arising from these controls and so we consider quadratic functions for measuring the control costs [19,27–31].

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The goal is to determine an optimal control u_i, v_i and w_i such that

$$\mathcal{J}(u_1, v_i, w_i) = \min_{\Omega} \mathcal{J}(u_1, v_i, w_i), \tag{7}$$

where $\Omega = \{u_i(t), w_i(t), w_i(t) | 0 \leq u_i(t), v_i(t), w_i(t) \leq 1$ are measurable$\}$.

The Pontryagins Maximum Principle [36] introduces adjoint functions that enable us to combine the state system to the objective functional. With the Pontryagins principle we can convert the problem of minimizing the objective functional to the state system into a problem that involves minimizing a Hamiltonian H, with respect to $u_i(t)$, $v_i(t)$ and $w_i(t)$. From the idea above we now have the Hamiltonian for the objective functional and the state system given as

$$\begin{aligned}
H &= A_1S_1(t) + B_1I_1(t) + C_1P_1(t) + d_1u_1^2(t) + e_1v_1^2(t) + f_1w_1^2(t) \\
&+ \lambda_{S_1}\Big(\mu_1N_1(t) - (\beta_1I_1(t) + \alpha_1P_1(t))S_1(t) - (\mu_1 + u_1)S_1(t) + \varphi_1R_1(t) \\
&- k_{12}S_1(t) + k_{21}S_2(t)\Big) + \lambda_{V_1}\Big(u_1S_1(t) - (1-\epsilon_1)(\beta_1I_1(t) + \alpha_1P_1(t))V_1(t) - \mu_1V_1(t)\Big) \\
&+ \lambda_{I_1}\Big((\beta_1I_1(t) + \alpha_1P_1(t))S_1(t) + (1-\epsilon)(\beta_1I_1(t) + \alpha_1P_1(t))V_1(t) \\
&- (\mu_1 + \gamma_1 + \nu_1)I_1(t) - b_{12}I(t) + b_{21}I_2(t)\Big) + \lambda_{T_1}\Big(\nu_1I_1(t) - (\mu_1 + \rho_1)T_1(t)\Big) \\
&+ \lambda_{R_1}\Big(\gamma_1I_1(t) + \rho_1T_1(t) - (\mu_1 + \varphi_1)R_1(t)\Big) + \lambda_{P_1}\Big(\delta_1I_1(t) - (\xi_1 + w_1)P_1(t)\Big) \\
&+ A_2S_2(t) + B_2I_2(t) + C_2P_2(t) + d_2u_2^2(t) + e_2v_2^2(t) + f_2w_2^2(t) \\
&+ \lambda_{S_2}\Big(\mu_2N_2(t) - (\beta_2I_2(t) + \alpha_2P_2(t))S_2(t) - (\mu_2 + u_2)S_2(t) + \varphi_2R_2(t) + k_{12}S_1(t) \\
&- k_{21}S_2(t)\Big) + \lambda_{V_2}\Big(u_2S_2(t) - (1-\epsilon_2)(\beta_2I_2(t) + \alpha_2P_2(t))V_2(t) - \mu_2V_2(t)\Big) \\
&+ \lambda_{I_2}\Big((\beta_2I_2(t) + \alpha_2P_2(t))S_2(t) + (1-\epsilon)(\beta_2I_1(t) + \alpha_2P_2(t))V_2(t) - (\mu_2 + \gamma_2 + \nu_2)I_2(t) \\
&+ b_{12}I(t) - b_{21}I_2(t)\Big) + \lambda_{T_2}\Big(\nu_2I_2(t) - (\mu_2 + \rho_2)T_2(t)\Big) \\
&+ \lambda_{R_2}\Big(\gamma_2I_2(t) + \rho_2T_2(t) - (\mu_2 + \varphi_2)R_2(t)\Big) + \lambda_{P_2}\Big(\delta_2I_2(t) - (\xi_2 + w_2)P_2(t)\Big),
\end{aligned}$$

where λ_{S1}, λ_{V1}, λ_{I1}, λ_{T1}, λ_{R1}, λ_{P1}, λ_{S2}, λ_{V2}, λ_{I2}, λ_{T2}, λ_{R2}, and λ_{P2} are associated adjoint for the states S_i, V_i, I_i, T_i, R_i and P_i, respectively. Given an optimal control triple $(u_i^*(t), v_i^*(t), w_i^*(t))$ together with corresponding states $(S_i^*, V_i^*, I_i^*, T_i^*, R_i^*, P_i^*)$ that minimizes $\mathcal{J}(u_i, v_i, w_i)$ over Ω, there exits adjoint variables λ_{Si}, λ_{Vi}, λ_{Ii}, λ_{Ti}, λ_{Ri}, and λ_{Pi} $(i = 1, 2)$ that satisfies

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$$
\begin{aligned}
\frac{d\lambda_{S_1}}{dt} &= -A_1 + \lambda_{S_1}\Big(\beta_1 I_1(t) + \alpha_1 P_1(t) + (\mu_1 + u_1) - k_{12}\Big) - \lambda_{V_1} u_1 - \lambda_{I_1}(\beta_1 I_1 + \alpha_1 P_1(t)),\\
\frac{d\lambda_{V_1}}{dt} &= \lambda_{V_1}\Big((1-\epsilon_1)(\beta_1 I_1(t) + \alpha_1 P_1(t)) + \mu_1\Big) - \lambda_{I_1}\Big((1-\epsilon_1)(\beta_1 I_1(t) + \alpha_1 P_1(t))\Big),\\
\frac{d\lambda_{I_1}}{dt} &= -B_1 + \lambda_{S_1}\beta_1 S_1(t) + \lambda_{V_1}(1-\epsilon_1)\beta_1 V_1(t) - \lambda_{I_1}\Big(\beta_1 S_1(t) + (1-\epsilon_1)\beta_1 V_1(t)\\
&- (\mu_1 + \gamma_1 + \nu_1) - b_{12}\Big) - \lambda_{T_1} v_1 - \lambda_{R_1}\gamma_1 - \lambda_{P_1}\delta_1,\\
\frac{d\lambda_{T_1}}{dt} &= \lambda_{T_1}(\mu_1 + \rho_1) - \lambda_{R_1}\rho_1,\\
\frac{d\lambda_{R_1}}{dt} &= -\lambda_{S_1}\varphi_1 + \lambda_{R_1}(\mu_1 + \varphi_1),\\
\frac{d\lambda_{P_1}}{dt} &= -C_1 + \lambda_{S_1}\alpha_1 S_1(t) + \lambda_{V_1}(1-\epsilon_1)\alpha_1 V_1(t) - \lambda_{I_1}\alpha_1\Big(S_1(t) + (1-\epsilon_1)V_1(t)\Big)\\
&+ \lambda_{P_1}(\xi_1 + w_1),\\
\frac{d\lambda_{S_2}}{dt} &= -A_2 + \lambda_{S_2}\Big(\beta_2 I_2(t) + \alpha_2 P_2(t) + (\mu_2 + u_2) + k_{21}\Big) - \lambda_{V_2} u_2 - \lambda_{I_2}(\beta_2 I_2 + \alpha_2 P_2(t)),\\
\frac{d\lambda_{V_2}}{dt} &= \lambda_{V_2}\Big((1-\epsilon_2)(\beta_2 I_2(t) + \alpha_2 P_2(t)) + \mu_2\Big) - \lambda_{I_2}(1-\epsilon_2)(\beta_2 I_2(t) + \alpha_2 P_2(t)),\\
\frac{d\lambda_{I_2}}{dt} &= -B_2 + \lambda_{S_2}\beta_2 S_2(t) + \lambda_{V_2}(1-\epsilon_2)\beta_2 V_2(t) - \lambda_{I_2}\Big(\beta_2 S_2(t) + (1-\epsilon_2)\beta_2 V_2(t)\\
&- (\mu_2 + \gamma_2 + \nu_2) + b_{21}\Big) - \lambda_{T_2} v_2 - \lambda_{R_2}\gamma_2 - \lambda_{P_2}\delta_2,\\
\frac{d\lambda_{T_2}}{dt} &= \lambda_{T_2}(\mu_2 + \rho_2) - \lambda_{R_2}\rho_2,\\
\frac{d\lambda_{R_2}}{dt} &= -\lambda_{S_2}\varphi_2 + \lambda_{R_2}(\mu_2 + \varphi_2),\\
\frac{d\lambda_{P_2}}{dt} &= -C_2 + \lambda_{S_2}\alpha_2 S_2(t) + \lambda_{V_2}(1-\epsilon_2)\alpha_2 V_2(t) - \lambda_{I_2}\alpha_2\Big(S_2(t) + (1-\epsilon_2 V_2(t)\Big) \qquad (8)\\
&+ \lambda_{P_2}(\xi_2 + w_2),
\end{aligned}
$$

together with the transversality conditions $\lambda_k(t_f) = 0$, for $k = S_i, V_i, I_i, T_i, R_i$ and P_i.

Note that we get the differential equation (8) which governs the adjoint variables by differentiating the appropriate Hamiltonian function (8) with respect to the corresponding state as follows:

$$\frac{d\lambda_k}{dt} = \frac{dH}{dk}. \qquad (9)$$

Now, consider the optimality conditions

$$\frac{\partial H}{\partial u_i} = 0, \quad \frac{\partial H}{\partial v_i} = 0, \quad \frac{\partial H}{\partial w_i} = 0. \qquad (10)$$

So for the control triplet u_i^*, v_i^* and w_i^* to satisfy the optimality condition we have;

For u_i we have

$$2d_i u_i^* - \lambda_{S_i} S_i(t) + \lambda_{V_i} S_i(t) = 0. \qquad (11)$$

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The solving for u_i using the optimality condition (10), we have

$$u_i^* = \frac{S_i(t)(\lambda_{S_i} - \lambda_{V_i})}{2d_i}, \tag{12}$$

and subsequently taking bounds into consideration, we have

$$u_i^* = \min\left(1, \max\left(0, \frac{S_i(t)(\lambda_{S_i} - \lambda_{V_i})}{2d_i}\right)\right). \tag{13}$$

Solving for v_i using the optimality condition, we have

$$v_i^* = \frac{I_i(t)(\lambda_{I_i} - \lambda_{T_i})}{2e_i}, \tag{14}$$

and subsequently taking bounds into consideration, we have

$$v_i^* = \min\left(1, \max\left(0, \frac{I_i(t)(\lambda_{I_i} - \lambda_{T_i})}{2e_i}\right)\right). \tag{15}$$

Similarly for w_i, we have

$$w_i^* = \min\left(1, \max\left(0, \frac{\lambda_{P_i} P_i(t)}{2f_i}\right)\right), \quad (i = 1, 2). \tag{16}$$

The results obtained above, shows that the optimal triple (u_i^*, v_i^*, w_i^*) has the tendency of reducing the impact of typhoid fever in any human population, if there is application of the disease control measures at a minimum cost. The optimal control triple is parameter dependent, hence further analysis is to be carried out to determine the exact effect of these optimal control parameters in eradicating these diseases and its magnitude of reduction taking into consideration cost of application. This is done using numerical simulation to give pictorial view of this impact using published data in similar peer reviewed works.

5.1. Existence of the optimal control

Let $x = (u_i, v_i, w_i) \in [L^2(0, T)]^3$ and $\mathcal{I} = (S_i, I_i, P_i)$. Hence, a reduced function corresponding to (6) is given by

$$\mathcal{J}(x, I^x) = \int_0^T \sum_{i=1}^{2} \left\{ A_i S_i + B_i I_i + C_i P_i + d_i u_i^2 + e_i v_i^2 + f_i w_i^2 \right\} dt, \quad x \in \Omega. \tag{17}$$

Lemma 1. *Set Ω is convex and closed.*

Proof. To prove that Ω is a closed set, assume that $x_{m\in\mathbb{N}} \longrightarrow x^*$ in $L^2(0, T)$ for $x_m \in \Omega$ but $x^* \notin \Omega$, i.e., $x^* < 0$ or $x^* > 1$ on a set of positive measure. Then taking $x^* < 0$, from lebesgue measure methods there exists $\epsilon > 0$ and a positive measure set $(0, t) \subset (0, T)$ such that $x^* \leq 0 - \epsilon$ on $(0, t)$ [37]. This implies that

$$\int_0^T (x_m - x^*)^2 dt \geq \int_0^t (x_m - x^*)^2 dt \geq \int_0^t (0 - x^*)^2 dt \geq \int_0^t \epsilon^2 dt > 0,$$

a contradiction. Thus, set Ω is closed.

To prove convexity of set Ω, it suffices to show that if Ω is a convex set and $u_i, v_i, w_i \in \Omega$, then any convex combination of any of u_i, v_i, w_i (say u_i) $\sum_{i=1}^2 \phi_i u_i$ for $\sum_{i=1}^2 \phi_i = 1$, $\phi_1, \phi_2 \geq 0$ is also contained in Ω.

The proof is by induction. For $i = 1$, since $u_1 \in \Omega$ then $\phi_1 u_1 \in \Omega$. For $i = 2$, since $(u_1, u_2) \in \Omega$,

$$0 \leq u_1 \leq 1, \tag{18}$$

and

$$0 \leq u_2 \leq 1. \tag{19}$$

Multiplying (18) by ϕ_1 and (19) by ϕ_2 gives

$$0 \leq \phi_1 u_1 \leq \phi_1, \tag{20}$$

and

$$0 \leq \phi_2 u_2 \leq \phi_2. \tag{21}$$

Adding up equations (20) and (21), we have

$$0 \leq \phi_1 u_1 + \phi_2 u_2 \leq 1.$$

Thus, $\phi_1 u_1 + \phi_2 u_2 \in \Omega$. This also relates to v_1, v_2 and w_1, w_2. This can also be extended to the nth socioeconomic class.

For $i = n - 1$, suppose that $\sum_{i=1}^{n-1} \phi_i u_i \in \Omega$. By the inductive hypothesis,

$$y = \frac{\phi_1 u_1}{\sum_{i=1}^{n-1} \phi_i} + \frac{\phi_2 u_2}{\sum_{i=1}^{n-1} \phi_1} + ... + \frac{\phi_{n-1} u_{n-1}}{\sum_{i=1}^{n-1} \phi_i} \in \Omega. \tag{22}$$

For $i = n$,

$$\sum_{i=1}^{n} \phi_i u_i = \sum_{i=1}^{n-1} \phi_i u_i + \phi_n u_n. \tag{23}$$

From (22), $\sum_{i=1}^{n-1} \phi_i u_i = y \sum_{i=1}^{n-1} \phi_i$. Then (23) gives

$$\sum_{i=1}^{n} \phi_i u_i = y \sum_{i=1}^{n-1} \phi_i + \phi_n u_n. \tag{24}$$

Since $u_n \in \Omega$ it follows that the RHS of (24) is a convex combination of two points of Ω. Thus, $y \sum_{i=1}^{n-1} \phi_i + \phi_n u_n \in \Omega$, so, is $y \sum_{i=1}^{n-1} \phi_i + \phi_n v_n \in \Omega$ and $y \sum_{i=1}^{n-1} \phi_i + \phi_n w_n \in \Omega$ hence, Ω is convex. □

Theorem 1. *There exists an optimal control pair $(x^*, \mathcal{I}^{x^*})$ to the optimization problem (7).*

Proof. Set

$$b = \sup_{x \in \Omega} \mathcal{J}(x, I^x).$$

This implies, for any $m \in \mathbb{N}$, there exists $x_m \in \Omega$ so that

$$b - \frac{1}{m} < \mathcal{J}(x_m, \mathcal{I}^{x_m}) \leq b. \tag{25}$$

As set Ω is a bounded subset of $L^2(0, T)$, it follows from Bolzano-Weierstrass theorem, that there exists a subsequence $\{x_{m_r}\}_{r \in \mathbb{N}}$ such that

$$x_{m_r} \longrightarrow x^*, \tag{26}$$

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weakly in $L^2(0,T)$. From (2) we have that all non-negative initial conditions are bounded. Thus, there exists a subsequence $\{\mathcal{I}^{x_{mr}}\}_{r\in\mathbb{N}}$ such that

$$\mathcal{I}^{x_{mr}} \longrightarrow \mathcal{I}^{x^*} \quad \text{in} \quad C([0,T]). \tag{27}$$

From (25),

$$b - \frac{1}{m} < \int_0^T \sum_{i=1}^{2} \left\{ A_i S_i^{u_{imr}}(t) + B_i I_i^{v_{imr}}(t) + C_i P_i^{w_{imr}}(t) + d_i u_{i_{mr}}^2 + e_i v_{i_{mr}}^2 + f_i w_{i_{mr}}^2 \right\} dt \le b. \tag{28}$$

By (25) and (27), passing to the limit in (30),

$$b = \int_0^T \sum_{i=1}^{2} \left\{ A_i S_i^{u_i^*}(t) + B_i I_i^{v_i^*}(t) + C_i P_i^{w_i^*}(t) + d_i (u_i^*)^2 + e_i (v_i^*)^2 + f_i (w_i^*)^2 \right\} dt, \tag{29}$$

that is, $((u_i^*, v_i^*, w_i^*), (S_i^{u^*}, I_i^{v^*}, P_i^{w^*}))$, $i = 1, 2$ is an optimal pair where u_i^*, v_i^* and w_i^* are optimal controls for (7). □

5.2. *Uniqueness of the optimal control system*

The optimality system of our optimal control problem is the combination of model (5) and the adjoint variables (8). So we have

$$\begin{aligned}
\frac{dS_i(t)}{dt} &= N_i(t)\mu_i - (\beta_i I_i(t) + \alpha_i P_i(t))S_i(t) - (\mu_i + u_i)S_i(t) + \varphi_i R_i(t) - k_{i,j} S_i(t) + k_{j,i} S_j(t), \\
\frac{dV_i(t)}{dt} &= u_i S_i(t) - (1 - \varepsilon_i)(\beta_i I_i(t) + \alpha_i P_i(t))V_i(t) - \mu_i V_i(t), \\
\frac{dI_i(t)}{dt} &= (\beta_i I_i(t) + \alpha_i P_i(t))S_i(t) + (1 - \varepsilon_i)(\beta_i I_i(t) + \alpha_i P_i(t))V_i(t) - (\mu_i + \gamma_i + v_i)I_i(t) \\
&- b_{i,j} I_i(t) + b_{j,i} I_j(t), \\
\frac{dT_i(t)}{dt} &= v_i I_i(t) - (\mu_i + \rho_i)T_i(t), \\
\frac{dR_i(t)}{dt} &= \gamma_i I_i(t) + \rho_i T_i(t) - (\mu_i + \varphi_i)R_i(t), \\
\frac{dP_i(t)}{dt} &= \delta_i I_i(t) - (\xi_i + w_i)P_i(t), \\
\frac{d\lambda_{S_i}}{dt} &= -A_i + \lambda_{S_i}\Big(\alpha_i P_i(t) + (\mu_i + u_i) - k_{ij}\Big) - \lambda_{V_i} u_i + \lambda_{I_i} \alpha_i P_i(t), \\
\frac{d\lambda_{V_i}}{dt} &= \lambda_{V_i}\Big((1 - \epsilon_i)(\beta_i I_i(t) + \alpha_i P_i(t)) + \mu_i\Big) - \lambda_{I_i}(1 - \epsilon_i)(\beta_i I_i(t) + \alpha_i P_i(t)), \\
\frac{d\lambda_{I_i}}{dt} &= -B_i - \lambda_{I_i}\Big(\beta_i S_i(t) + (1 - \epsilon_i)\beta_i V_i(t) - (\mu_i + \gamma_i + \nu_i) - b_{ij}\Big) - \lambda_{I_i} V_i(t) \\
&- \lambda_{R_i}\gamma_i - \lambda_{P_i}\delta_i, \\
\frac{d\lambda_{T_i}}{dt} &= \lambda_{T_i}(\mu_i + \rho_i) - \lambda_{R_i}\rho_i, \\
\frac{d\lambda_{R_i}}{dt} &= -\varphi_i + \lambda_{R_i}(\mu_i + \varphi_i), \\
\frac{d\lambda_{P_i}}{dt} &= -C_i + \lambda_{S_i}\alpha_i S_i(t) + (1 - \epsilon_i)\alpha_i V_i(t) + \lambda_{I_i}\alpha_i S_i(t) - (1 - \epsilon_i)\alpha_i V_i(t) + \lambda_{P_i}(\xi_i + w_i),
\end{aligned} \tag{30}$$

where $S_i(t_0)$, $I_i(t_0)$, $P_i(t_0) \ge 0$, and $\lambda_{S_i} = 0$, $\lambda_{V_i} = 0$, $\lambda_{I_i} = 0$, $\lambda_{T_i} = 0$, $\lambda_{R_i} = 0$, $\lambda_{P_i} = 0$.

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Theorem 2. *For sufficiently small t_f, the solution to the optimality system (30) of the optimal control problem is unique.*

Proof. Suppose $(S_i, V_i, I_i, T_i, R_i, P_i, \lambda_{S_i}, \lambda_{V_i}, \lambda_{I_i}, \lambda_{T_i}, \lambda_{R_i}, \lambda_{R_i})$, and $(\tilde{S}_i, \tilde{V}_i, \tilde{I}_i, \tilde{T}_i, \tilde{R}_i, \tilde{P}_i, \tilde{\lambda}_{S_i}, \tilde{\lambda}_{V_i}, \tilde{\lambda}_{I_i}, \tilde{\lambda}_{T_i}, \tilde{\lambda}_{R_i}, \tilde{\lambda}_{R_i})$ are two solutions of the optimality system (30). Let $S_i = e^{\lambda i t}x_1$, $V_i = e^{\lambda i t}x_2$, $I_i = e^{\lambda i t}x_3$, $T_i = e^{\lambda i t}x_4$, $R_i = e^{\lambda i t}x_5$, $P_i = e^{\lambda i t}x_6$, $\lambda_{S_i} = e^{-\lambda i t}y_1$, $\lambda_{V_i} = e^{-\lambda i t}y_2$, $\lambda_{I_i} = e^{-\lambda i t}y_3$, $\lambda_{T_i} = e^{-\lambda i t}y_4$, $\lambda_{R_i} = e^{-\lambda i t}y_5$, $\lambda_{P_i} = e^{-\lambda i t}y_6$, $\tilde{S}_i = e^{\lambda i t}\tilde{x}_1$, $\tilde{V}_i = e^{\lambda i t}\tilde{x}_2$, $\tilde{I}_i = e^{\lambda i t}\tilde{x}_3$, $\tilde{T}_i = e^{\lambda i t}\tilde{x}_4$, $\tilde{R}_i = e^{\lambda i t}\tilde{x}_5$, $\tilde{P}_i = e^{\lambda i t}\tilde{x}_6$, $\tilde{\lambda}_{S_i} = e^{-\lambda i t}\tilde{y}_1$, $\tilde{\lambda}_{V_i} = e^{-\lambda i t}\tilde{y}_2$, $\tilde{\lambda}_{I_i} = e^{-\lambda i t}\tilde{y}_3$, $\tilde{\lambda}_{T_i} = e^{-\lambda i t}\tilde{y}_4$, $\tilde{\lambda}_{R_i} = e^{-\lambda i t}\tilde{y}_5$, $\tilde{\lambda}_{P_i} = e^{-\lambda i t}\tilde{y}_6$, where λ is chosen arbitrarily. We now let

$$
\begin{aligned}
u_i^* &= \min\left(1, \max\left(0, \frac{x_1(y_1 - y_2)}{2d_i}\right)\right), \\
v_i^* &= \min\left(1, \max\left(0, \frac{x_3(y_3 - y_4)}{2e_i}\right)\right), \\
w_i^* &= \min\left(1, \max\left(0, \frac{y_6 x_6}{2f_i}\right)\right), \\
\tilde{u}_i^* &= \min\left(1, \max\left(0, \frac{\tilde{x}_1(\tilde{y}_1 - \tilde{y}_2)}{2d_i}\right)\right), \\
\tilde{v}_i^* &= \min\left(1, \max\left(0, \frac{\tilde{x}_3(\tilde{y}_3 - \tilde{y}_4)}{2e_i}\right)\right), \\
\tilde{w}_i^* &= \min\left(1, \max\left(0, \frac{\tilde{y}_6 \tilde{x}_6}{2f_i}\right)\right), \quad (i = 1, 2)
\end{aligned}
$$

Now lets consider the first equation of (30), we have

$$
\begin{aligned}
\dot{x}_1 + \lambda_i x_1 &= \mu_i N_i e^{-\lambda t} - \beta_i x_3 x_1 - \alpha_i x_6 x_1 - \mu_i x_1 + \varphi_i x_5 - k_{ij} x_1 + k_{ji} x_2, \\
\dot{\tilde{x}}_1 + \lambda_i \tilde{x}_1 &= \mu_i N_i e^{-\lambda t} - \beta_i \tilde{x}_3 \tilde{x}_1 - \alpha_i \tilde{x}_6 \tilde{x}_1 - \mu_i \tilde{x}_1 + \varphi_i \tilde{x}_5 - k_{ij} \tilde{x}_1 + k_{ji} \tilde{x}_2.
\end{aligned}
$$

For simplicity, we assume that there is no movement between S_1 and S_2 in this prove.

By subtracting and integrating from t_0 to t_f for the above two equations, we have

$$
\begin{aligned}
\frac{1}{2}(x_1(t_f) - \tilde{x}_1(t_f))^2 + (\lambda_i + \mu_i)\int_{t_0}^{t_f}(x_1 - \tilde{x}_1)^2 dt &= -\beta_i \int_{t_0}^{t_f}(x_1 x_3 - \tilde{x}_1 \tilde{x}_3)(x_1 - \tilde{x}_1)dt \\
&- \alpha_i \int_{t_0}^{t_f}(x_1 x_6 - \tilde{x}_1 \tilde{x}_6)(x_1 - \tilde{x}_1)dt \\
&+ \varphi \int_{t_0}^{t_f}(x_5 - \tilde{x}_5)(x_1 - \tilde{x}_1)dt. \qquad (31)
\end{aligned}
$$

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Note that

$$\begin{aligned}
\int_{t_0}^{t_f}(u_i-\tilde{u}_i)^2dt &\leq \left(\frac{1}{2d_i}\right)^2\int_{t_0}^{t_f}[x_1(y_1-y_2)-\tilde{x}_1(\tilde{y}_1-\tilde{y}_2)]^2dt,\\
&\leq \left(\frac{1}{2d_i}\right)L_1\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2]dt,\\
\int_{t_0}^{t_f}(v_i-\tilde{v}_i)^2dt &\leq \left(\frac{1}{2e_i}\right)^2\int_{t_0}^{t_f}[x_3(y_3-y_4)-\tilde{x}_3(\tilde{y}_3-\tilde{y}_4)]^2dt,\\
&\leq \left(\frac{1}{2e_i}\right)L_2\int_{t_0}^{t_f}[(x_3-\tilde{x}_3)^2+(y_3-\tilde{y}_3)^2+(y_4-\tilde{y}_4)^2]dt,\\
\int_{t_0}^{t_f}(w_i-\tilde{w}_i)^2dt &\leq \left(\frac{1}{2f_i}\right)^2\int_{t_0}^{t_f}[x_6y_6-\tilde{x}_6\tilde{y}_6]^2dt,\\
&\leq \left(\frac{1}{2f_i}\right)L_3\int_{t_0}^{t_f}[(x_6-\tilde{x}_6)^2+(y_6-\tilde{y}_6)^2]dt,\\
\int_{t_0}^{t_f}(x_1x_3-\tilde{x}_1\tilde{x}_3)(x_1-\tilde{x}_1)dt &\leq \int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2x_3+\tilde{x}_1(x_3-\tilde{x}_3)(x_1-\tilde{x}_1)]dt,\\
&\leq C_1\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_3-\tilde{x}_3)^2]dt,\\
\int_{t_0}^{t_f}(x_1x_6-\tilde{x}_1\tilde{x}_6)(x_1-\tilde{x}_1)dt &= \int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2x_6+\tilde{x}_1(x_6-\tilde{x}_6)(x_1-\tilde{x}_1)]dt,\\
&\leq C_2\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_6-\tilde{x}_6)^2]dt,\\
\int_{t_0}^{t_f}(x_5-\tilde{x}_5)(x_1-\tilde{x}_1)dt &\leq C_3\int_{t_0}^{t_f}(x_5-\tilde{x}_5)dt,
\end{aligned}$$

where C_1 depends on the bounds of $\tilde{x}_1, x_3$, C_2 depends on the bounds of $\tilde{x}_1, x_6$, C_3 depends on the bounds of $x_5, \tilde{x}_5$. So, by (31), we have

$$\begin{aligned}
\frac{1}{2}(x_1(t_f)-\tilde{x}_1(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(x_1-\tilde{x}_1)^2dt &\leq M_1\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2(x_3-\tilde{x}_3)^2+(x_5-\tilde{x}_5)^2+(x_6-\tilde{x}_6)^2]dt\\
&+ N_1\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(y_1-\tilde{y}_1)^2\\
&+ (y_2-\tilde{y}_2)^2]dt, \qquad (32)
\end{aligned}$$

where M_1 is an appropriate upper-bound. Similarly, we can get the following inequalities for $(x_k(t_f), \tilde{x}_k(t_f))$ and $(y_l(t_f), \tilde{y}_l(t_0))$ $(k = 1,2,3,5,6,\ l = 1,2,3,5,6)$:

$$\begin{aligned}
\frac{1}{2}(x_2(t_f)-\tilde{x}_2(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(x_2-\tilde{x}_2)^2dt &\leq M_2\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_2-\tilde{x}_2)^2+(x_3-\tilde{x}_3)^2\\
&+ (x_5-\tilde{x}_5)^2+(x_6-\tilde{x}_6)^2]dt\\
&+ N_2\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2\\
&+ (y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2]dt, \qquad (33)
\end{aligned}$$

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$$\begin{aligned}\frac{1}{2}(x_3(t_f)-\tilde{x}_3(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(x_3-\tilde{x}_3)^2dt &\leq M_3\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_3-\tilde{x}_3)^2]dt\\ &+ N_3\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(y_1-\tilde{y}_1)^2\\ &+ (y_2-\tilde{y}_2)^2]dt\\ &+ K_1\int_{t_0}^{t_f}[(x_2-\tilde{x}_2)^2+(x_3-\tilde{x}_3)^2]dt, \quad (34)\end{aligned}$$

$$\begin{aligned}\frac{1}{2}(x_5(t_f)-\tilde{x}_5(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(x_5-\tilde{x}_5)^2dt &\leq M_4\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_3-\tilde{x}_3)^2+(x_5-\tilde{x}_5)^2\\ &+ (x_6-\tilde{x}_6)^2]dt+N_4\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2\\ &+ (y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2++(y_3-\tilde{y}_3)^2]dt\\ &+ K_2\int_{t_0}^{t_f}[(x_2-\tilde{x}_2)^2+(x_5-\tilde{x}_5)^2]dt, \quad (35)\end{aligned}$$

$$\begin{aligned}\frac{1}{2}(x_6(t_f)-\tilde{x}_6(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(x_6-\tilde{x}_6)^2dt &\leq M_5\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_6-\tilde{x}_6)^2]dt\\ &+ N_5\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(y_1-\tilde{y}_1)^2\\ &+ (y_2-\tilde{y}_2)^2+(y_5-\tilde{y}_5)^2]dt\\ &+ K_3\int_{t_0}^{t_f}[(x_5-\tilde{x}_5)^2+(x_6-\tilde{x}_6)^2]dt, \quad (36)\end{aligned}$$

$$\begin{aligned}\frac{1}{2}(y_1(t_f)-\tilde{y}_1(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(y_1-\tilde{y}_1)^2dt &\leq M_6\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_2-\tilde{x}_2)^2+(x_3-\tilde{x}_3)^2+(x_5-\tilde{x}_5)^2\\ &+ (x_6-\tilde{x}_6)^2+(y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2\\ &+ (y_3-\tilde{y}_3)^2+(y_5-\tilde{y}_5)^2+(y_6-\tilde{y}_6)^2]dt\\ &+ D_1\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_5-\tilde{x}_5)^2+(y_1-\tilde{y}_1)^2\\ &+ (y_5-\tilde{y}_5)^2]dt+N_6\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2\\ &+ (y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2]dt, \quad (37)\end{aligned}$$

$$\begin{aligned}\frac{1}{2}(y_2(t_f)-\tilde{y}_2(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(y_2-\tilde{y}_2)^2dt &\leq M_7\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(y_1-\tilde{y}_1)^2\\ &+ (y_2-\tilde{y}_2)^2]dt\\ &+ K_4\int_{t_0}^{t_f}[(y_2-\tilde{y}_2)^2+(y_5-\tilde{y}_5)^2]dt, \quad (38)\end{aligned}$$

$$\begin{aligned}\frac{1}{2}(y_3(t_f)-\tilde{y}_3(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(y_3-\tilde{y}_3)^2dt &\leq M_8\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_3-\tilde{x}_3)^2+(y_1-\tilde{y}_1)^2\\ &+ (y_2-\tilde{y}_2)^2+(y_3-\tilde{y}_3)^2]dt\\ &+ N_7\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_3-\tilde{x}_3)^2\\ &+ (y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2]dt, \quad (39)\end{aligned}$$

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$$
\begin{aligned}
\frac{1}{2}(y_5(t_f)-\tilde{y}_5(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(y_5-\tilde{y}_5)^2dt \quad \leq \quad & M_9\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_5-\tilde{x}_5)^2+(y_1-\tilde{y}_1)^2\\
+ \quad & (y_5-\tilde{y}_5)^2+D_2\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_5-\tilde{x}_5)^2\\
+ \quad & (y_1-\tilde{y}_1)^2+(y_5-\tilde{y}_5)^2]dt\\
+ \quad & K_5\int_{t_0}^{t_f}[(x_5-\tilde{x}_5)^2\\
+ \quad & (y_5-\tilde{y}_5)^2+(y_6-\tilde{y}_6)^2]dt, \qquad (40)
\end{aligned}
$$

$$
\begin{aligned}
\frac{1}{2}(y_6(t_f)-\tilde{y}_6(t_f))^2+(\lambda_i+\mu_i)\int_{t_0}^{t_f}(y_6-\tilde{y}_6)^2dt \quad \leq \quad & M_{10}\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2+(x_6-\tilde{x}_6)^2\\
+ \quad & (y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2+(y_5-\tilde{y}_5)^2\\
+ \quad & (y_6-\tilde{y}_6)^2]dt+N_8\int_{t_0}^{t_f}[(x_1-\tilde{x}_1)^2\\
+ \quad & (x_6-\tilde{x}_6)^2+(y_1-\tilde{y}_1)^2+(y_2-\tilde{y}_2)^2\\
+ \quad & (y_5-\tilde{y}_5)^2]dt, \qquad (41)
\end{aligned}
$$

where $M_k(k=1,2,...,10)$, $N_l(l=1,2,...,8)$, $D_k(k=1,2)$ and $K_l(l=1,2,...,5)$ depend on the coefficients and the bounds of the state variables and co-state variables. Adding up equations (32) to (41), we have

$$
\begin{aligned}
& \left[(\lambda_i+\mu_i)-\left(\sum_{k=1}^{10}M_k\right)-D_2-\left(\sum_{l=1}^{8}N_l\right)\right]\int_{t_0}^{t_f}(x_1-\tilde{x}_1)^2dt\\
& +[(\lambda_i+K_1+K_2+\mu_i)-(M_1+M_2+M_5)]\int_{t_0}^{t_f}(x_2-\tilde{x}_2)^2dt\\
& +[(\lambda_i+\mu_i-K_1-K_3-K_5)-(M_1+M_4+M_6)-(D_1+D_2)]\int_{t_0}^{t_f}(x_5-\tilde{x}_5)^2dt\\
& +\left[(\lambda_i+\mu_i+K_1+K_3)-\left(\sum_{k=1}^{6}M_k+M_{10}\right)-\left(\sum_{l=1}^{6}N_l\right)\right]\int_{t_0}^{t_f}[(x_3-\tilde{x}_3)^2+(x_6-\tilde{x}_6)^2]dt\\
& +[(\lambda_i+\mu_i)-\left(\sum_{k=1}^{10}M_k\right)-D_1-\left(\sum_{l=1}^{8}N_l\right)]\int_{t_0}^{t_f}(y_1-\tilde{y}_1)^2dt\\
& +\left[(\lambda_i+K_1+K_3-K_4+\mu_i)-(M_6+M_7+M_8+M_{10})\left(\sum_{l=1}^{8}N_l\right)\right]\int_{t_0}^{t_f}(y_2-\tilde{y}_2)^2dt\\
& +\left[(\lambda_i+\mu_i+K_4-K_5)-\left(M_6+\sum_{k=1}^{10}M_k\right)-D_2-\left(\sum_{l=1}^{8}N_l\right)\right]\int_{t_0}^{t_f}(y_5-\tilde{y}_5)^2dt\\
& +[(\lambda_i+\mu_i-K_5)-(M_6+M_8+M_{10})+(N_3-N_5)]\int_{t_0}^{t_f}(y_3-\tilde{y}_3)^2+(y_6-\tilde{y}_6)^2dt\\
& \leq 0. \qquad (42)
\end{aligned}
$$

From equation (42), we can see clearly that the coefficients of the integrals and non-negative anytime we choose a large λ_i is chosen and in turn choosing a small value of t_f. For instance, if we take $\lambda_i > -\mu_i+\sum_{k=1}^{10}M_k+D_2+\sum_{l=1}^{8}N_k$ and also $t_f < \frac{1}{3\lambda_i}\ln\frac{\lambda_i+\mu_i}{A_i}$, $A_i := \sum_{k=1}^{10}M_k+D_2+\sum_{l=1}^{8}N_l$, then we see that the coefficient $(\lambda_i+\mu_i)-(\sum_{k=1}^{10}M_k)-D_2-(\sum_{l=1}^{8}N_l)\geq 0$ in relation to the integral $\int_{t_0}^{t_f}(x_1-\tilde{y_2})^2dt$. This is also application to the various $\lambda_i s$ relative to different x and y's, which shows that each integral of (42) is non negative.

To this effect, we can see that $x_1=\tilde{x_1}$, $x_2=\tilde{x_2}$, $x_3=\tilde{x_3}$, $x_5=\tilde{x_5}$, $x_6=\tilde{x_6}$, $y_1=\tilde{y_1}$, $y_2=\tilde{y_2}$, $y_3=\tilde{y_3}$,

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$y_5 = \tilde{y_5}$, $y_6 = \tilde{y_6}$, and $S_i = \tilde{S_i}$, $V = \tilde{V_i}$, $I_i = \tilde{I_i}$, $T_i = \tilde{T_i}$, $R_i = \tilde{R_i}$, $P_i = \tilde{P_i}$. We can conclude that the solution of (42) is unique for small time t. □

The unique optimal control triple (u_i^*, v_i^*, w_i^*) is characterised in terms of the unique solution of the optimal system. Therefore, the optimal triple gives us the optimal control strategy that is effective in preventing the incidence of Typhoid fever in any human population.

6. Numerical illustration of optimal control

Here we present the numerical solution of the optimal control problem. We first consider the effect of the controls on different socio-economic status independently as seen in Figure 6 and then systematically show the effects of the optimal controls over the controls on the different classes as illustrated in Figures 7-11. To illustrate this we use the parameter values as given in Table 3 with the following assigned cost factors: $A_1 = 0.8$, $A_2 = 0.8$, $B_1 = 0.7$, $B_2 = 0.7$, $C_1 = 0.9$, $C_2 = 0.9$, $d_1 0.4$, $d_2 = 0.4$, $e_1 = 0.2$, $e_2 = 0.2$, $f_1 = 0.3$, $f_2 = 0.3$. We carried out iterative technique by employing the forward-backward algorithm postulated by Lenhart and Workman [33] to obtain the optimal control functions (u_1, v_1, w_1) and (u_2, v_2, w_2) as shown in Figure 6. The figure shows that it is most appropriate or optimal to commence treatment early and to make it readily available to affected victims of Typhoid fever and also to ensure that vaccination is adequately provided across all socio-economic communities and lastly adherence to good sanitation. This result is realistic, since it agrees with disease epidemiology in human, that supports good treatment and introduction of vaccination at early stage before the onset of an epidemic while ensuring adequate treatment of individuals throughout the endemic period.

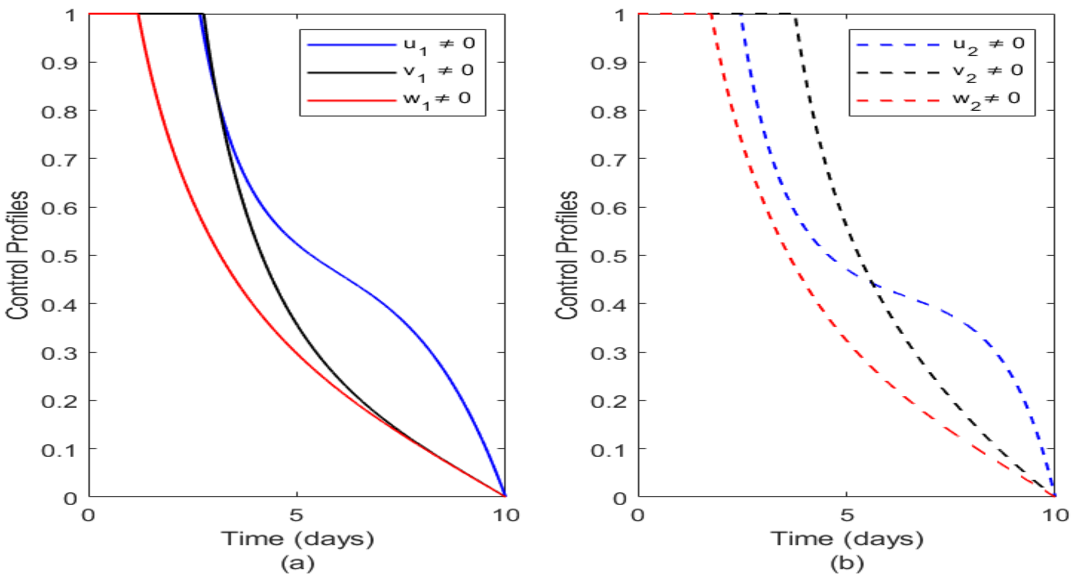

Figure 6. Plot illustrating the control profiles of the two socioeconomic classes. Figure 6(a) illustrates the control profiles for SEC 1 while Figure 6(b) illustrates control profiles for SEC 2.

Figure 6 illustrates the optimal control profiles of the two socioeconomic profiles. Here vaccination is denoted by red thick line on SEC 1 and red dashed line on SEC 2. From Figure 6(a), it is seen that the vaccination is very effective at the onset but weans out with time on SEC 1. Also we have that for SEC 2 vaccination also thrives at the onset but waens out over a period of time but when compared with vaccination on SEC 1, we have that vaccination in SEC 1 weans faster than that of SEC 2 and this could be attributed to the healthy or unhealthy activities carried out in either of the classes. Treatment on the other hand denoted by blue thick line on SEC 1 and blue dashed line on SEC 2 weans out almost the same time on both socio-economics classes and this agrees with real life intuition as treatment may

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be helpful but if the epidemic is not controlled, does requiring treatment may outnumber available medical equipment's and practitioners. This also shows that treatment when properly administered has same effect on both socioeconomic classes. Finally, we have sanitation denoted by black thick line in SEC 1 and black dashed line in SEC 2. It is seen that sanitation plays a major role in reducing the impact of Typhoid fever but it is more active on SEC 2 than SEC 1 and this readily agrees with real life scenario as high socio-economics class individuals tends to reside more in an environment that is hygienic and they obey sanitation regulations.

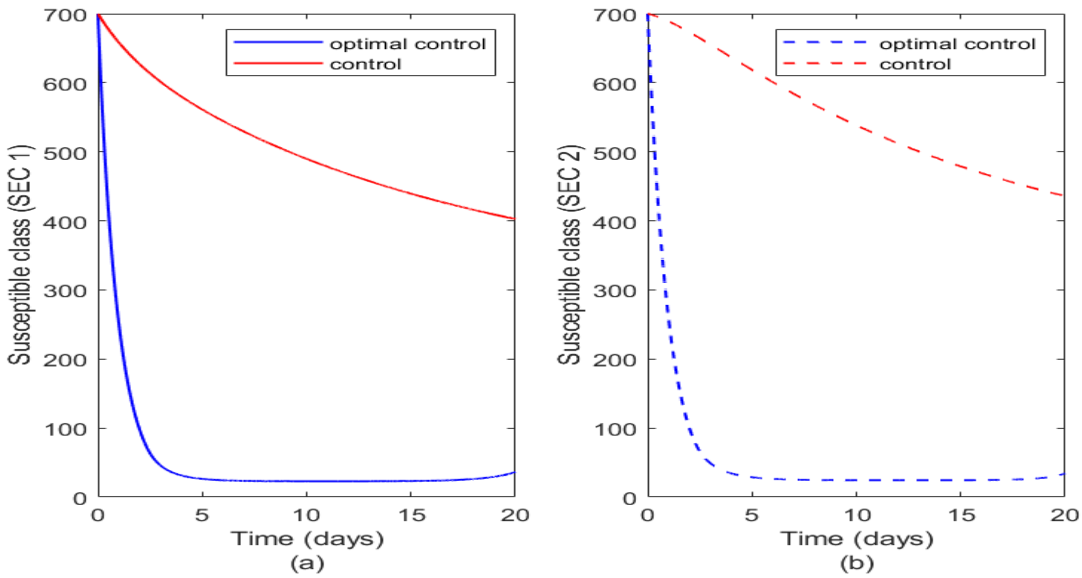

Figure 7. Plot illustrating effect of optimal control and controls on the two socioeconomic classes. Figure 7(a) illustrates the effect of optimal control and controls on susceptible class of SEC 1 while Figure 7(b) illustrates the effect of optimal control and controls on susceptible class of SEC 2.

Figure 7 illustrates the impact of control and optimal control on susceptible humans in both socioeconomic classes. The optimal control of vaccination reduces completely susceptibility to typhoid fever in both socioeconomic class population. Alternatively just combining the controls have more impact on SEC 1 than on SEC 2 and this may be consequent on the fact that vaccination maybe more targeted on SEC 1 population than SEC 2.

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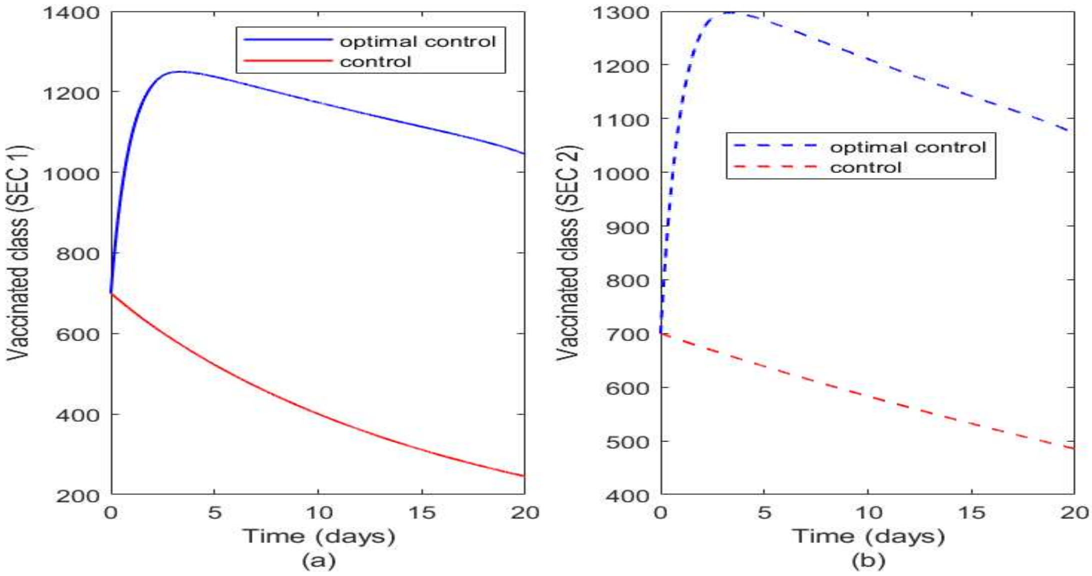

Figure 8. Plot illustrating effect of optimal control and controls on the two socioeconomic classes. Figure 8(a) illustrates the effect of optimal control and controls on vaccinated class of SEC 1 while Figure 8(b) illustrates the effect of optimal control and controls on vaccinated class of SEC 2.

Figure 8 illustrates the effect of optimal control and controls on the vaccinated class. The vaccinated class in SEC 1 increases when vaccination is applied optimally but the impact of vaccination quickly reduces rather than in SEC 2 where there is more impact of optimal vaccination as shown in Figure 8(b). Also, even without optimal control we have that the control by vaccination weans out more in SEC 1 when compared to SEC 2 even though we have more reduction in susceptibility in SEC 1 (Figure 7a). This agrees with our earlier prediction that vaccination weans out faster in SEC 1 population due to unhealthy practices and due to poor living standard has less immunity to diseases than those in SEC 2.

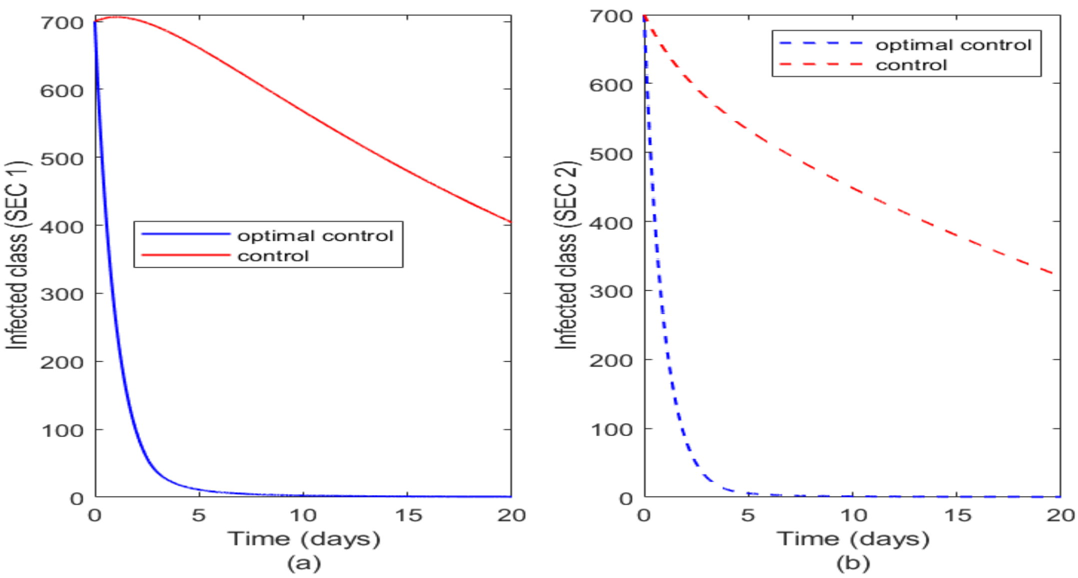

Figure 9. Plot illustrating effect of optimal control and controls on the two socioeconomic classes. Figure 9(a) illustrates the effect of optimal control and controls on infected class of SEC 1 while Figure 9(b) illustrates the effect of optimal control and controls on infected class of SEC 2.

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Figure 9 also illustrates the impact of optimal control and controls on the infected class of the two socioeconomic classes. Optimal control has same effect on the two classes but the controls are more effective on SEC 2 than SEC 1. There is more reduction in the population of infected class in SEC 2 (see Figure 9b) than in SEC 1 (Figure 9a) when the controls are not optimal. This agrees with real life intuition as those in higher socio-economic status tend to benefit more from disease control strategies than those in lower socio-economic status. But when either or all the controls are optimally used in both communities it yields the same result.

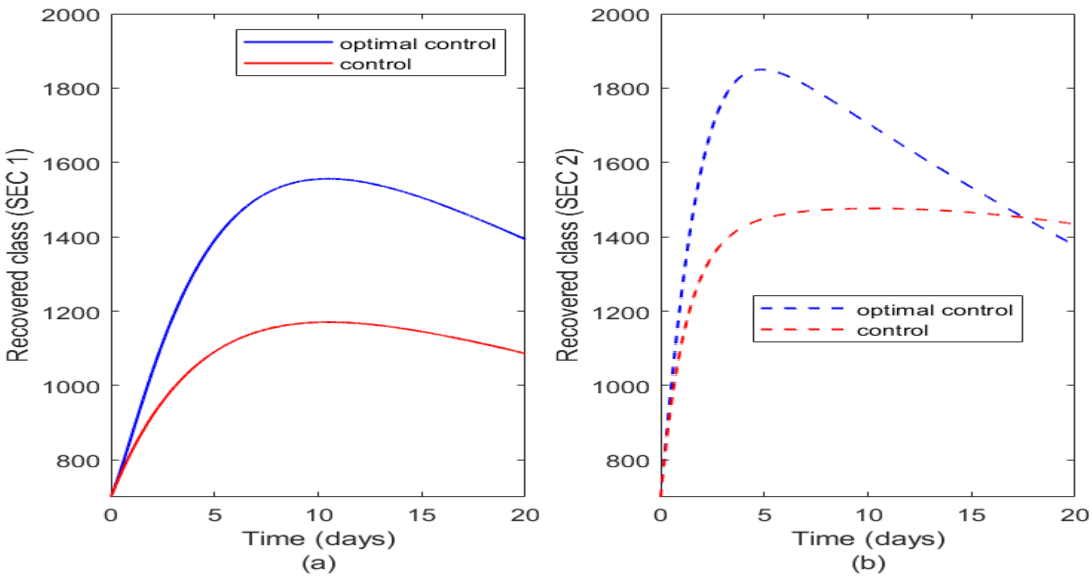

Figure 10. Plot illustrating effect of optimal control and controls on the two socioeconomic classes. Figure 10(a) illustrates the effect of optimal control and controls on recovered class of SEC 1 while Figure 10(b) illustrates the effect of optimal control and controls on recovered class of SEC 2.

From Figure 10 it is shown that with either controls or optimal control there is higher number of recovered humans in SEC 2 (Figure 10b) than in SEC 1 (Figure 10a) and this is attributed to good medical practices, good sanitation and early exposure to vaccination and generally good standard of living which boosts their immune system to enable them recover faster even when infected.

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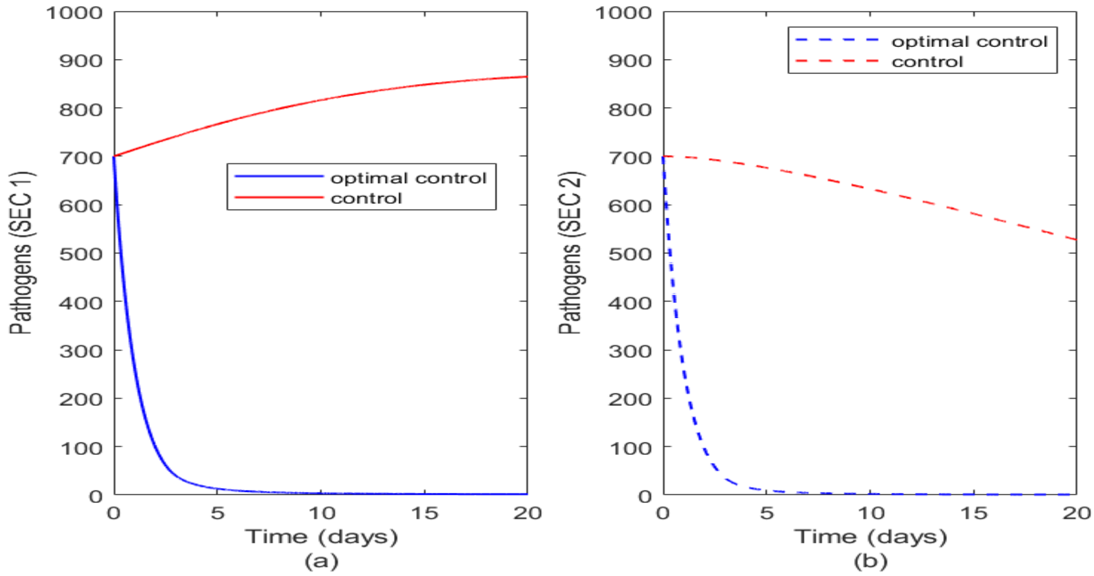

Figure 11. Plot illustrating effect of optimal control and controls on the two socioeconomic classes. Figure 11(a) illustrates the effect of optimal control and controls on pathogens in the environment of SEC 1 while Figure 11(b) illustrates the effect of optimal control and controls on pathogens in the environment of SEC 2.

Finally, Figure 11 illustrates the impact of optimal control of sanitation on the pathogens in the two socioeconomic classes. Analogous to other cases discussed, optimal sanitation reduces pathogens in both classes but just applying sanitation in relation to what is obtainable in the different socioeconomic classes. It is seen that it is more effective in SEC 2 than in SEC 1.

7. Discussion

Typhoid fever is a fatal illness affecting humans especially those in the lower socio-economic community with limited access to clean food and a neat environment. The disease can be prevented or controlled by adopting effective control intervention measures. Some of the effective control measures for decreasing typhoid fever infections in some affected communities include vaccination, treatment, and sanitation. Many countries/communities are comprised of individuals in different socio-economic classes. For more accurate results on the dynamics of typhoid fever, the socio-economic classes of individuals in the community must be taken into consideration. This study used a mathematical epidemiological model to analyze the influence of control measures (vaccination, treatment, and sanitation) in decreasing typhoid fever in multiple socio-economic communities. By developing and analyzing a mathematical epidemiological model for typhoid fever for multiple socio-economic community, the dynamics of the disease were explored. The results of our analysis showed that the disease can be eradicated from the two socio-economic classes using the control measures provided that the basic reproduction number remained below unity. In contrast, when no control measure is introduced, the disease remains endemic in the community, especially in the lower socio-economic community. Further analysis revealed that under uniform movement rates, the lower SEC have a greater infected population, so control measures should be the focus on this class for faster disease eradication.

Next, the influence of each of the control measures was investigated numerically. Each of the control measures was found to have some influence in reducing typhoid fever. The combined effects of the multiple control measures yield better results when compared with the no-control measure and single control measure. Based on these findings, multiple control measures are highly recommended for controlling typhoid fever. However, if it is not available, any of the single control measures can be used because each of them is shown to have some positive influence in reducing infections in the two socio-economic communities.

Finally, we carried out optimal control analysis on our control model and it was observed that optimal control has effect on both socio-economic classes but optimizing treatment has more effect on SEC 2 than SEC 1 followed by vaccination and then sanitation. We also compared optimal control and controls on each of the classes in the two socioeconomic classes. Our analysis showed that optimal control has good effect on both classes even though it may wean out with time rather than using the controls collectively or independently. In all our analysis showed that more attention should be paid on communities of low socio-economic status in the event of typhoid fever epidemic.

Author Contributions: Obiora Collins conceived the project and provided the conceptual framework and early mathematical model and analysis. Ifeanyi Onah further developed and refined the model to add optimal control analysis and Stephen Aniaku contributed in the analysis and wrote the manuscript with assistance of Obiora Collins and Ifeanyi Onah.

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Informed Consent Statement: This research is a deterministic mathematical model which provides postulations on Typhoid disease epidemic in any human population but no real data was used in the study.

Data Availability Statement: Data sharing not applicable in this research article.

Conflicts of Interest: The authors declare no conflict of interest.

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