## RESEARCH ARTICLE

# A Mathematical Model of Tuberculosis with Respect to Drug Resistance to the First and Second Line of the Treatment 

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#### Abstract

In this research paper work, we developed a mathematical model for tuberculosis (TB). Disease was formulated and rigorously analyzed. The model sub-divided into six compartments. The model has equilibria; the diseases-free equilibrium. The equilibrium states were obtained and analyzed for their stability relatively to the effective reproduction number. The result shows that the disease-free equilibrium state was stable state is established. We able to show that the TB disease free equilibrium is locally and globally asymptotically stable $R_{0}<1$. Using the number of treatment, individual increases as the rate at which recovery rate of which makes them recovery back from disease. The analytical solution was obtained using homotopy perturbation method and effective reproduction number was computed to measure the relative impact for individual or combined intervention for effective disease control. Numerical simulations of the model show that lose their immunity at the rate decreases at immunity wanes of which makes them susceptible back to disease is the most effective way to combat the epidemiology of TB.


Key words: Mathematical model, tuberculosis, resistance, second and fist line, treatment

## INTRODUCTION

Tuberculosis (TB) is a worldwide pandemic disease. According to the World Health Organization (WHO) [1], one-third of the world's population is currently infected by the TB bacillus bacteria. Being a disease of poverty, the vast majority of TB deaths are in the developing world with more than half occurring in Asia (Bleed et al., 2001) [2]. The estimated global incidence rate is falling very slowly from the peak of 141 cases per 100,000 population in 2002 to 128 cases per 100,000 population in 2010. The TB death rate has also fallen by $40 \%$ since 1990 and the number of deaths is also declining.
Globally, the percentage of people successfully treated reached its highest level at $87 \%$ in 2009. TB caused by infection with the bacillus Mycobacterium tuberculosis is a very common and an infectious airborne disease. It typically affects the lungs (pulmonary TB) but can affect other sites as well (extrapulmonary TB). It is estimated that one-third of the world's population has been infected with the M. tuberculosis (the WHO, 2010) [3]. Moreover, an estimated 8.6 million people developed TB and 1.3 million died from the disease (including 320 thousand deaths among HIV-positive people) in 2012 (Kochi, 2001)[3]. Although the rate of new TB cases and the TB incidence rates are falling worldwide and the TB mortality rate has been reduced, the absolute number of incident cases of TB is increasing due to population growth. Therefore, TB remains a major global health problem (the WHO, 2013) [4].
In 2011, the treatment success rate continued to be high at $87 \%$ among all new TB cases. However, there were about 3 million people who developed TB and were missed by national notification systems (the WHO, 2013)[4]. On the other hand, the treatment interruptions are frequent in active TB cases during the intensive phase and the continuation phase because of a wide range of reasons (Jakubowiak

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et al., 2009)[5]. It may be recognized that the treatment interruptions and the missed TB cases are the key factors to cause the more drug-resistant TB cases and the high TB mortality (Tsai et al., 2010)[6]. The factor of treatment interruptions may result in more susceptible people infected as well. In 2012, there was estimation that 450 thousand individuals developed multidrug-resistant TB (MDR-TB) and an estimation of 170 thousand deaths from MDR-TB [2], which is currently a main threat to tuberculosis control programs and community health (van Helden et al., 2006)[7].
According to the WHO, the number of people falling ill with TB each year is declining (the WHO, 2010)[1]. However, this downward trend is threaten by the increasing number of TB cases in immigrants especially in countries that have substantial levels of immigration from areas with a high prevalence of the disease (Jia et al., 2008)[8]. The immigrants here are generally the people who are travelling from less to more economically developed geographical areas in search of jobs and better living conditions. As an air-borne infectious disease, it is impossible for any country to isolate itself. In long-term, the best defense against TB is to bring the disease under control worldwide.
The model subdivides the population into six mutually-exclusive classes, namely, susceptible $S(t)$, early latent $E_{1}(t)$, late latent $E_{2}(t)$, non-isolated infectious $I(t)$, isolated infectious $J(t)$, and treated or recovered $T(t)$.
Some of the limitations are;
i. Unavailability of records of TB disease case
ii. Poor data documentation habit of public servants
iii. Scanty scholarly article on TB disease
iv. Is not age structured?

## MODEL FORMATION

Following Andrawus et al., 2020[9], the model considers human population N. The population at time $t$ is divided into six (6) subpopulations. Susceptible $S(t)$ this class includes those individuals who are at risk for developing an infection from TB. Latent Class $E(t)$ this class refers to susceptible individuals who become infected. Infected $I(t)$ : This class includes all individuals who are showing the symptom of the disease. First Line Treatment $T_{N}(t)$ : This class includes all individual that failed to take drugs (treatment). Second Line Treatment $T_{D}(t)$ : This class includes all individual that failed to take drugs (treatment) for the second treatment term. Recovered $R(t)$ : This class includes all individuals that have recovered from the disease and got temporary immunity. The susceptible class is increased by birth or emigration at the rate of $\pi$ : The susceptible class will get TB when they mingled with infectious individuals at the, $\beta$ is the effective contact rate. The latent class is generated at a rate $\beta S I$, it decreases at rate $\alpha$ which is the progression rate. The infectious class is generated at a rate $\alpha$, it also reduces at rate of the treatment $r_{1}$ and it further reduces at a rate $\gamma_{1}$ which is default of the treatment. The first line treatment is generated at rate $\gamma_{1}$, it reduces at a rate of the treatment $r_{1}$ and it further reduces at a rate $\gamma_{2}$ which is second default of the treatment. The second line treatment class is generated at rate $\gamma_{2}$, it reduces at a rate of the treatment $r_{1}$ and recovered class is generated at rate $r_{1}, r_{2}$ and $r_{3}$ The infectious, first line treatment, and second line treatment class are reduced at the rates $\delta_{1}, \delta_{2}$ and $\delta_{3}$ while the whole classes reduce at the rate $\propto$ those who lose their partial immunity at the rate $\sigma$ which recovered humans become susceptible. Which is the natural mortality rate, Hence, Figure 1, shows the schematic diagram:

## Mathematical Model

Our assumptions are as follow;

- That the population size in a compartment is differentiable with respect to time and that the epidemic process is deterministic
- That the population is heterogeneous. That is, the individuals that make up the population can be grouped into different compartments or groups according to their epidemiological state
- That a proportion of the population of newborn in immunized against TB infection through treatment
- That the immunity conferred on individuals by the treatment expires after some time at given rate
- That people in each compartment have equal natural death rate of $\propto$
- That there are no immigrants and emigrants. The only way of entry into the population is through new-born babies and the only way of exit is through death from natural causes of death from TBrelated causes
- That the infection does not confer immunity to the curved and recovered individuals and so they go back to the susceptible class at a given rates
- That all newborns are previously uninfected by TB and therefore joint either the immunized compartment or the susceptible compartment depending on whether they are treated or not.
The model equations are;
$\frac{d S}{d t}=\pi+\sigma R-\beta S I-\mu S$
$\frac{d E}{d t}=\beta S I-(\mu+\alpha) E$
$\frac{d I}{d t}=\alpha E-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I$
$\frac{d T_{N}}{d t}=\gamma_{1} I-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N}$
$\frac{d T_{D}}{d t}=\gamma_{2} T_{N}-\left(r_{3}+\mu+\delta_{3}\right) T_{D}$
$\frac{d R}{d t}=r_{1} I+r_{2} T_{N}+r_{3} T_{D}-(\mu+\sigma) R$


## Positive Invariant Region

The entire population size $N$ can be determined from equations (1) to (6).
The total population size is $N=S+E+I+T_{N}+T_{D}+R$
Adding equation (1) to equation (6)
$\frac{d N}{d t}=\frac{d S}{d t}+\frac{d E}{d t}+\frac{d I}{d t}+\frac{d T_{N}}{d t}+\frac{d T_{D}}{d t}+\frac{d R}{d t}$
$\frac{d N}{d t}=\pi-\mu S-\mu E-\mu I-\delta_{1} I-\mu T_{N}$

$$
\begin{equation*}
-\delta_{2} T_{N}-\mu T_{D}-\delta_{3} T_{D}-\mu R \tag{9}
\end{equation*}
$$

In the absence of the disease ( $\delta_{1}=\delta_{2}=\delta_{3}=0$ ) then (9) given
The positive invariant region can be obtained using the following theorem.

## Theorem 2.1

The solutions of the system of equations (1) to (6) are feasible for $t>0$ if they enter that the invariant region $D$ is given as equation (10).

## Proof

Let $D=\left(S, E, I, T_{N}, T_{D}, R\right) \in R^{6}$

Be any solution of the system of equations (1) to (6) with non-zero initial conditions.
Assuming there is no disease-induced deaths equation (9)
$\frac{d N}{d t} \leq \pi-\mu N$
$\frac{d N}{d t}+\mu N \leq \pi$
The integrating fact for equation (13) is multiplying both sides of equation (12) by give
$\frac{d N}{d t}+\mu N e^{-\mu t} \leq \pi e^{\mu t}$
$\frac{d N}{d t} \leq \mu$
Integrating both sides we have
$N(t)$
$N(t)=\frac{\pi}{\mu}+c$
Applying the initial condition $t=0, N(0)=N_{0}$
$N_{0} \leq \frac{\pi}{\mu}+c \Rightarrow N_{0}-\frac{\pi}{\mu} \leq c$
$\rightarrow N \leq \frac{\pi}{\mu}+\left(N_{0}-\frac{\pi}{\mu}\right) e^{-\mu t}$
Therefore, as $t=\infty$ in (18) to humans $N$ approaches $K=\frac{\pi}{\mu}$ (that is $N \rightarrow K=\frac{\pi}{\mu}$ ) the parameter $K=\frac{\pi}{\mu}$ is called the carrying capacity. Hence, all feasible solution set of the human of the model equation (1) to (6) enter the region.

$$
\begin{equation*}
\left.D=\left\{\left(S, E, I, T_{N}, T_{D}, R\right) \in R^{6}: S>0, E>0, I>0, T_{N}>0, \mathrm{~T}_{D}>0, \mathrm{R}>0, \mathrm{~N} \leq \frac{\pi}{\mu}\right)\right\} \tag{19}
\end{equation*}
$$

With is positively invariant (i.e., solution remain positive for all time $t$ ) and the model is epidemiologically meaningful and mathematically well pose.

## POSITIVE OF SOLUTIONS

Since equation (1) to (6) represent the population in each compartment and all model parameters are all positive, then if lies in region $D$ defined by

## Theorem 2

Let the initial data for the model equation be given as this
$S(0) \geq 0, E(0) \geq 0, I(0) \geq 0, T_{N}(0) \geq 0, T_{D}(0) \geq 0, R(0) \geq 0$
Then, the solutions $\left(S(t), E(t), I(t), T_{N}(t), T_{D}(t), R\right)$ of the model equation with non-negative initial data with remain non-negative for at time $t>0$

## Proof

For equation (1)
$\frac{d S}{d t}=\pi+\sigma R-\beta I S-\mu S \geq-\mu S(t)$
$\frac{d S}{d t} \geq-\mu S \mu S(t)$
Separating the variables and integrating both sides, we have
$\frac{d S}{S} \geq-\mu d t$
In $S(t) \geq \propto t+c$
$S(t)=e^{-\alpha t+c}$
$S(t)=e^{-\alpha t}$

Where $k=e^{c}$
Using the initial condition $t=0 \Rightarrow S(0) \geq k$
Therefore, $S(t) \geq S(0) e^{-\alpha \tau} \geq 0$
From equation (2)
$\frac{d E}{d t}=\beta S I-(\mu+\alpha) E \geq-(\mu+\alpha) E$
$\frac{d E}{d t} \geq-(\mu+\alpha) E$
Separating the variables and integrating both sides, we have
$\frac{d E}{E} \geq-(\mu+\alpha) d t$
$\operatorname{InE}(t) \geq-(\mu+\alpha)+c$
$\left.E(t)=e^{-(\alpha+} \alpha\right)^{l t+c}$
$E(t)=e^{-(\alpha+} \alpha^{t}$

Where $k=e^{c}$
Using the initial condition $t=0 \Rightarrow E(0) \geq k$
From equation (3)
$\frac{d I}{d t}=\alpha E-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I \geq-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I$
$\frac{d I}{d t} \geq-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I$
Separating the variables and integrating both sides, we have
$\frac{d I}{I} \geq-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) d t$
$\operatorname{InI}(t) \geq-\left(\gamma_{1}+r_{1}+\infty+\delta_{1}\right)$
$I(t)=e^{--} \boldsymbol{\gamma}^{1+r 1+\alpha+} \boldsymbol{\delta}^{1)+c}$
$I(t)=e^{-( } \gamma^{1+r l^{\prime}+\alpha+} \delta^{1) t}$

Where $k=e^{c}$
Using the initial condition $t=0 \Rightarrow I(0) \geq k$
Therefore, $I(t) \geq I(0) e^{-( } \gamma^{1+r 1+\alpha+} \delta^{1) t} \geq 0$
From equation (4)
$\frac{d T_{N}}{d t}=\gamma_{1} I-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N} \geq-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N}$
$\frac{d T_{N}}{d t} \geq-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N}$
Separating the variables and integrating both sides, we have
$\frac{d T_{N}}{T_{N}} \geq-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) d t$
$\operatorname{InT} T_{N}(t) \geq-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)+c$
$T_{N}(t)=e^{-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) t+c}$
$T_{N}(t)=e^{-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) t}$

Where $k=e^{c}$
Using the initial condition $t=0 \Rightarrow T_{D}(0) \geq k$
Therefore, $T_{N}(t) \geq T_{N}(0) e^{-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) t} \geq 0$
From equation (5)
$\frac{d T_{D}}{d t}=\gamma_{2} T_{N}-\left(r_{3}+\mu+\delta_{3}\right) T_{D} \geq-\left(r_{3}+\mu+\delta_{3}\right) T_{D}$
$\frac{d T_{D}}{d t} \geq-\left(r_{3}+\mu+\delta_{3}\right) T_{D}$
Separating the variables and integrating both sides, we have
$\frac{d T_{D}}{T_{D}} \geq-\left(r_{3}+\mu+\delta_{3}\right) d t$
$\operatorname{In} T_{D}(t) \geq-\left(r_{3}+\mu+\delta_{3}\right)+c$
$T_{D}(t)=e^{-\left(r_{3}+\mu+\delta_{3}\right) t+c}$
$T_{D}(t)=e^{-\left(r_{3}+\mu+\delta_{3}\right) t}$
Where $k=e^{c}$
Using the initial condition $t=0 \Rightarrow T_{D}(0) \geq k$
Therefore, $T_{D}(t) \geq T_{D}(0) e^{-\left(r_{3}+\mu+\delta_{3}\right) t} \geq 0$
From equation (6)
$\frac{d R}{d t}=r_{1} I+r_{2} T_{N}+r_{3} T_{D}-(\sigma+\mu) R \geq-(\sigma+\mu) R$
$\frac{d R}{d t} \geq-(\sigma+\mu) R$

Separating the variables and integrating both sides, we have
$\frac{d R}{R} \geq-(\sigma+\mu) d t$
$\operatorname{InR}(t) \geq-(\sigma+\mu) t+c$
$R(t)=e^{-(\sigma+\mu) t+c}$
$R(t)=e^{-(\sigma+\mu) t}$

Where $k=e^{c}$
Using the initial condition $t=0 \Rightarrow R(0) \geq k$
Therefore, $R(t) \geq R(0) e^{-(\sigma+\mu) t} \geq 0$

## THE EXISTENCE AND UNIQUENESS OF SOLUTION

The validity and implementation of any mathematical model depend on whether the given system of equations has a solution, and if it has, there is need to check if the solution is unique (Ayoade et al., 2019) [10].

## Theorem 4.1

Let D denotes the region $\pi \in \mathfrak{R}^{+}$. Then, the model system (1) to (6) has a unique solution if it is established that $\frac{\partial m_{i}}{\partial m_{i}}, \mathrm{i}=1,2,3,4,5,6$ are continuous and bounded in D .

## Proof

Let equations (1) to (2) be represented by $m_{1}, m_{2}, m_{3}, m_{4}, m_{5}$ and $m_{6}$ respectively
From equation (1), the following partial derivatives are obtain

$$
\left.\begin{array}{l}
\left|\frac{\partial m_{1}}{\partial S}\right|=|-\beta I-\mu|<\infty ;\left|\frac{\partial m_{1}}{\partial E}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial I}\right|=|-\beta S|<\infty ; \\
\left|\frac{\partial m_{1}}{\partial T_{N}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial T_{D}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial R}\right|=|\sigma|<\infty \tag{68}
\end{array}\right\}
$$

The above partial derivatives exist, continuous and are bounded. From equation (2), we obtained the following partial derivatives

$$
\left.\begin{array}{l}
\left|\frac{\partial m_{1}}{\partial S}\right|=|\beta I|<\infty ;\left|\frac{\partial m_{1}}{\partial E}\right|=|-(\alpha+\mu)|<\infty ;\left|\frac{\partial m_{1}}{\partial I}\right|=|\beta S|<\infty ; \\
\left|\frac{\partial m_{1}}{\partial T_{N}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial T_{D}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial R}\right|=0<\infty \tag{69}
\end{array}\right\}
$$

The above partial derivatives exist, continuous and are bounded. From equation (3), we obtained the following partial derivatives

$$
\left.\begin{array}{c}
\left|\frac{\partial m_{1}}{\partial S}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial E}\right|=|\alpha|<\infty ;\left|\frac{\partial m_{1}}{\partial I}\right|=\left|-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)\right|<\infty ; \\
\left|\frac{\partial m_{1}}{\partial T_{N}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial T_{D}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial R}\right|=0<\infty \tag{70}
\end{array}\right\}
$$

The above partial derivatives exist, continuous, and are bounded. From equation (4), we obtained the following partial derivatives

$$
\left.\begin{array}{l}
\left|\frac{\partial m_{1}}{\partial S}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial E}\right|=|\alpha|<\infty ;\left|\frac{\partial m_{1}}{\partial I}\right|=\left|\gamma_{1}\right|<\infty ; \\
\left|\frac{\partial m_{1}}{\partial T_{N}}\right|=\left|-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)\right|<\infty ;\left|\frac{\partial m_{1}}{\partial T_{D}}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial R}\right|=0<\infty
\end{array}\right\}
$$

The above partial derivatives exist, continuous, and are bounded. From equation (5), we obtained the following partial derivatives

$$
\left.\begin{array}{c}
\left|\frac{\partial m_{1}}{\partial S}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial E}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial I}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial T_{N}}\right|=\left|\gamma_{2}\right|<\infty ; \\
\left|\frac{\partial m_{1}}{\partial T_{D}}\right|=\left|-\left(+r_{3}+\mu+\delta_{3}\right)\right|<\infty ;\left|\frac{\partial m_{1}}{\partial R}\right|=0<\infty \tag{72}
\end{array}\right\}
$$

The above partial derivatives exist, continuous, and are bounded. From equation (6), we obtained the following partial derivatives

$$
\left.\begin{array}{l}
\left|\frac{\partial m_{1}}{\partial S}\right|=0<\infty ;\left|\frac{\partial m_{1}}{\partial E}\right|=|\alpha|<\infty ;\left|\frac{\partial m_{1}}{\partial I}\right|=\left|r_{1}\right|<\infty ; \\
\left|\frac{\partial m_{1}}{\partial T_{N}}\right|=\left|r_{2}\right|<\infty ;\left|\frac{\partial m_{1}}{\partial T_{D}}\right|=\left|r_{3}\right|<\infty ;\left|\frac{\partial m_{1}}{\partial R}\right|=|-(\sigma+\mu)|<\infty \tag{73}
\end{array}\right\}
$$

Since all the partial derivatives exist, bounded, and defined, then system of equations (1) - (6) exists and has solution in $\mathfrak{K}^{6}$.

## Equilibrium Points of the Model

At equilibrium
$\frac{d S}{d t}=\frac{d E}{d t}=\frac{d I}{d t}=\frac{d T_{N}}{d t}=\frac{d T_{D}}{d t}=\frac{d R}{d t}$
$\pi+\sigma R-\beta S I-\propto S=0$
$\beta S I-(\propto+\alpha) E=0$
$\alpha E-\left(\gamma_{1}+r_{1}+\infty+\delta_{1}\right) I=0$
$\gamma_{1} I-\left(\gamma_{2}+r_{2}+\infty+\delta_{2}\right) T_{N}=0$
$\gamma_{2} T_{N}-\left(r_{3}+\infty+\delta_{3}\right) T_{D}=0$
$r_{1} I+r_{2} T_{N}+r_{3} T_{D}-(\propto+\sigma) R=0$
From equation (77) we have
$\alpha E=\left(\gamma_{1}+r_{1}+\propto+\delta_{1}\right) I$
$I=\frac{\alpha E}{\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)}$
Substitute equation (82) into equation (76)
$\frac{\beta S \alpha E}{\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)}-(\mu+\alpha) E=0$
Cross multiplication
$\beta S \alpha E=\left(\gamma_{1}+r_{1}+\propto+\delta_{1}\right)(\propto+\alpha) E=0$
$\left[\beta S \alpha-\left(\gamma_{1}+r_{1}+\infty+\delta_{1}\right)(\propto+\alpha)\right] E=0$
$E=0$ or $\beta S \alpha-\left(\gamma_{1}+r_{1}+\propto+\delta_{1}\right)(\propto+\alpha)=0$
Suppose $E=0$

Substitute equation (87) into (77) we have
$\alpha(0)-\left(\gamma_{1}+r_{1}+\infty+\delta_{1}\right) I=0$
Therefore
$I=0$
Substitute equation (89) into equation (77), we have
$\gamma_{2}(0)=\left(\gamma_{2}+r_{2}+\infty+\delta_{2}\right) T_{N}$
Therefore
$T_{N}=0$
Substitute equation (91) into equation (79)
$\gamma_{2}(0)=\left(+r_{3+} \propto+\delta_{3}\right) T_{D}$
Therefore
$T_{D}=0$
Substitute equation (79), (91) and (93) into equation (80), we have
$r_{1}(0)+r_{2}(0)+r_{3}(0)=(\propto+\sigma) R$
Therefore,
$R=0$
Substitute equation (89) and (95) into equation (75), we have
$\pi+\sigma(0)-\beta S(0)=\propto S$
$\pi=\propto S$
Therefore,
$S=\frac{\pi}{\mu}$

## Disease-Free Equilibrium

The equilibrium state in the absence of infection is known as zero equilibrium state or disease-free equilibrium. Therefore, from equations (75) to (80), disease-free equilibrium is given as
$E^{0}=\left(S^{*}, E^{*}, I^{*}, T_{M}^{*}, T_{D}^{*}, R^{*}\right)=\left(\frac{\pi}{\mu}, 0,0,0,0,0\right)$

## EFFECTIVE REPRODUCTION NUMBER

In epidemiology, the basic reproduction number (sometimes called basic reproductive rate, basic reproductive ratio and is denoted by $R_{0}$ ) of an infection can be thought of as the number of cases which generate on the average over the course of its infection period, in an otherwise uninfected population.

## When

$R_{0}<1$ The infection will die out in the long run, but if
$R_{0}>1$ The infection will be able to spread in a population
In this model, the spectral radius of the equation is given as largest Eigen value given as Equation (90).
$R_{0}=\rho k \rightarrow f v^{-1}$
From equation (1) to equation (6), we have
$F_{i}=\left(\begin{array}{c}\beta S I \\ \alpha E \\ \gamma_{1} I \\ \gamma_{2} T_{N}\end{array}\right)$
$V_{i}=\left(\begin{array}{c}-(\mu+\alpha) E \\ -\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I \\ -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{M} \\ -\left(r_{3}+\mu+\delta_{3}\right) T_{D}\end{array}\right)$
$F=\left(\begin{array}{cccc}0 & \frac{\beta \pi}{\mu} & 0 & 0 \\ \alpha & 0 & 0 & 0 \\ 0 & \gamma_{1} & 0 & 0 \\ 0 & 0 & \gamma_{2} & 0\end{array}\right)$
$V=\left(\begin{array}{cccc}-(\mu+\alpha) & 0 & 0 & 0 \\ 0 & -\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) & 0 & 0 \\ 0 & 0 & -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) & 0 \\ 0 & 0 & 0 & -\left(r_{3}+\mu+\delta_{3}\right)\end{array}\right)$
Therefore, the inverse
$V^{-1}=\frac{\text { adjo int }}{\operatorname{det} e r \min a n t}$
$V^{-1}=\left(\begin{array}{cccc}-\frac{1}{(\mu+\alpha)} & 0 & 0 & 0 \\ 0 & -\frac{1}{\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)} & 0 & 0 \\ 0 & 0 & -\frac{1}{\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)} & 0 \\ 0 & 0 & 0 & -\frac{1}{\left(r_{3}+\mu+\delta_{3}\right)}\end{array}\right)$

$$
\begin{align*}
& F V^{-1}=\left(\begin{array}{cccc}
0 & \frac{\beta \pi}{\mu} & 0 & 0 \\
\alpha & 0 & 0 & 0 \\
0 & \gamma_{1} & 0 & 0 \\
0 & 0 & \gamma_{2} & 0
\end{array}\right) \\
& \left(\begin{array}{cccc}
-\frac{1}{(\mu+\alpha)} & 0 & 0 & 0 \\
0 & -\frac{1}{\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)} & 0 & 0 \\
0 & 0 & -\frac{1}{\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)} & 0 \\
0 & 0 & 0 & -\frac{1}{\left(r_{3}+\mu+\delta_{3}\right)}
\end{array}\right)  \tag{107}\\
& F V^{-1}=\left(\begin{array}{cccc}
0 & -\frac{\beta \pi}{\mu\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)} & 0 & 0 \\
-\frac{\alpha}{(\mu+\alpha)} & 0 & 0 & 0 \\
0 & -\frac{\gamma_{1}}{\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)} & 0 & 0 \\
0 & 0 & -\frac{\gamma_{2}}{\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)} & 0
\end{array}\right) \tag{108}
\end{align*}
$$

To compute, the Eigen values
$F V^{-1}=\left(\begin{array}{cccc}\lambda & -\frac{\beta \pi}{\mu\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)} & 0 & 0 \\ -\frac{\alpha}{(\mu+\alpha)} & \lambda & 0 & 0 \\ 0 & -\frac{\gamma_{1}}{\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)} & \lambda & 0 \\ 0 & 0 & -\frac{\gamma_{2}}{\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)} & \lambda\end{array}\right)=0$
The characteristics equation given as equation (110) to equation (111)
$\lambda=0$
$\lambda=\frac{\sqrt{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right) \pi \alpha \beta}}{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}$
and

$$
\begin{equation*}
\lambda=-\frac{\sqrt{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right) \pi \alpha \beta}}{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)} \tag{111}
\end{equation*}
$$

Then,
$R_{0}=\frac{\sqrt{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right) \pi \alpha \beta}}{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}$
Therefore
$\frac{\sqrt{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right) \pi \alpha \beta}}{\mu\left(\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}<1$

## LOCAL STABILITY OF DISEASE-FREE EQUILIBRIUM POINT

To analyze local stability at the disease-free equilibrium point, the Jacobian matrix of the system of equations (1) to (6) at DFE is evaluated. Then, the stability is determined based on the sign of the Eigen values of the Jacobian matrix.

## Theorem 7.1

The disease-free equilibrium point is said to be locally asymptotically stable, if all the Eigen values of the Jacobian matrix at DFE are negative or unstable otherwise.

## Proof:

The Jacobian matrix of the system of equations is;
$J=\left(\begin{array}{cccccc}-\mu & 0 & -\frac{\beta \pi}{\mu} & 0 & 0 & \sigma \\ 0 & -(\mu+\alpha) & \frac{\beta \pi}{\mu} & 0 & 0 & 0 \\ 0 & \alpha & -\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) & 0 & 0 & 0 \\ 0 & 0 & \gamma_{1} & -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) & 0 & 0 \\ 0 & 0 & 0 & \gamma_{2} & -\left(r_{3}+\mu+\delta_{3}\right) & 0 \\ 0 & 0 & r_{1} & r_{2} & r_{3} & -(\mu+\sigma)\end{array}\right)$
Reducing equation (116) to upper triangular matrix,
$J=\left(\begin{array}{cccccc}-\mu & 0 & -\frac{\beta \pi}{\mu} & 0 & 0 & \sigma \\ 0 & -(\mu+\alpha) & \frac{\beta \pi}{\mu} & 0 & 0 & 0 \\ 0 & 0 & -\frac{\mu\left(\pi \beta \alpha+\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}{\mu(\mu+\alpha)} & 0 & 0 & 0 \\ 0 & 0 & 0 & -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) & 0 & 0 \\ 0 & 0 & 0 & 0 & -\left(r_{3}+\mu+\delta_{3}\right) & 0 \\ 0 & 0 & 0 & 0 & 0 & -(\mu+\sigma)\end{array}\right)$

The characteristics equation is;
$\left|J\left(E^{0}\right)-\lambda I\right|=$
$\left(\begin{array}{cccccc}-\mu-\lambda & 0 & -\frac{\beta \pi}{\mu} & 0 & 0 & \sigma \\ 0 & -(\mu+\alpha)-\lambda & \frac{\beta \pi}{\mu} & 0 & 0 & 0 \\ 0 & 0 & \left.-\frac{\mu\left(\pi \beta \alpha+\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}\right)}{+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}}\right) \\ 0 & \mu(\mu+\alpha) & 0 & 0 & 0 \\ 0 & 0 & 0 & & & \\ 0 & 0 & 0 & -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)-\lambda & 0 & 0 \\ 0 & 0 & 0 & 0 & -\left(r_{3}+\mu+\delta_{3}\right)-\lambda & 0 \\ & & & 0 & 0 & -(\mu+\sigma)-\lambda\end{array}\right)=0$

The determinates give;

$$
\left[\begin{array}{l}
\left(-\mu-\lambda_{1}\right)\left(-(\mu+\alpha)-\lambda_{2}\right)\left(-\frac{\mu\left(\pi \beta \alpha+\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}{\mu(\mu+\alpha)}-\lambda_{3}\right) \\
\left(-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)-\lambda_{4}\right)\left(-\left(r_{3}+\mu+\delta_{3}\right)-\lambda_{5}\right)\left(-(\mu+\sigma)-\lambda_{6}\right)=0
\end{array}\right]
$$

Either
$-\propto-\lambda=0 \quad$ or $\quad-(\propto+\alpha)-\lambda_{2}=0 \quad$ or $-\frac{\mu\left(\pi \beta \alpha+\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}{\mu(\mu+\alpha)}-\lambda_{3}=0$ or -
$\left(\gamma_{2}+r_{2}+\propto+\delta_{2}\right)-\lambda_{4}=0$ or $-\left(r_{3}+\propto+\delta_{3}\right)-\lambda_{5}=0$ or $-(\propto+\sigma)-\lambda_{6}=0$
Therefore,
$\lambda_{1}=-\mu$ or $\lambda_{2}=-(\mu+\alpha) \quad$ or $\lambda_{3}=-\frac{\mu\left(\pi \beta \alpha+\mu^{2}+\mu \alpha+\mu r_{1}+\mu \delta_{1}+\mu \gamma_{1}+\alpha r_{1}+\alpha \delta_{1}+\alpha \gamma_{1}\right)}{\mu(\mu+\alpha)}$ or
$\lambda_{4}=-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) \quad$ or $\lambda_{5}=-\left(r_{3}+\mu+\delta_{3}\right) \quad$ or $\lambda_{6}=-(\mu+\sigma)$
From equation (119)
$\lambda_{1}, \lambda_{2}, \lambda_{3}, \lambda_{4}, \lambda_{5}, \lambda_{6}<0$
Hence, the disease-free equilibrium point is locally asymptotically stable.

## GLOBAL STABILITY OF DISEASE-FREE EQUILIBRIUM

## Theorem 3.9.1

The disease-free equilibrium of equations (45) to (50) is globally asymptotically stable provided $R_{0}<1$ and unstable if $R_{0}>1$
Proof: Referring to Castillo-Chaves et al. (2004)[11], the system of equations (45) to (50) can be written
as,
$\frac{d x(t)}{d t}=F(x, y)$,
$\frac{d y(t)}{d t}=G(x, y)$
Where $x=\left(S, E, I, T_{N}, T_{D}, R\right) \in \mathfrak{R}_{+}^{6}$ denote the different compartments of uninfected humans, $y=\left(E, I, T_{N}, T_{D}\right) \in \mathfrak{R}_{+}^{4}$ denote the different compartments of infected humans.

The disease free equilibrium $(\mathrm{DFE})=\left(x_{0}, 0\right)$, where $x_{0}=\left(\frac{\pi}{\mu}, 0\right)$
We are required to proof that,
$\frac{d x(t)}{d t}=F(x, 0), x_{0}$ is globally asymptotically
stable, and $G(x, y)=C y-\hat{G}(x, y)$,

$$
\begin{equation*}
\hat{G}(x, y) \geq 0 \text { for }(x, y) \in \Omega \tag{57}
\end{equation*}
$$

Case 1: Consider the uninfected subsystem,

$$
\begin{equation*}
\frac{d x(t)}{d t}=F(x, y)=\binom{\pi+\sigma R-\beta S I-\mu S}{r_{1} I+r_{2} T_{N}+r_{3} T_{D}-(\mu+\sigma) R} \tag{58}
\end{equation*}
$$

When $y=0$ that is $E=I=T_{N}=T_{D}=0$
Then, equation (58) becomes,
$F(x, 0)=\binom{\pi-\mu S}{0}$
Solving equation (60), gives
$\frac{d S(t)}{d t}=\pi-\mu S$
$\frac{d S(t)}{d t}+\mu S=\pi$
Multiply equation (62) by its integrating factor $\left(\ell^{\mu t}\right)$, gives

$$
\begin{equation*}
\frac{d S(t)}{d t} \ell^{\mu t}+\mu S \ell^{\mu t}=\Lambda \ell^{\mu t} \tag{63}
\end{equation*}
$$

$\frac{d}{d t}\left[S \ell^{-\mu t}\right]=\pi \ell^{\mu t}$

Integrate equation (65), gives
$\int \frac{d}{d t}\left[S \ell^{\mu t}\right]=\int \pi \ell^{\mu t}$
$S \ell^{(1+\mu) t}=\left(\frac{\Lambda}{(1+\mu)}\right) \ell^{(1+\mu) t}+c$
Divide equation (68) by $\ell^{\alpha+}$
$S(t)=\left(\frac{\Lambda_{d}}{(1+\mu)}\right)+c \ell^{-(1+\mu) t}$
When $t=0$, equation (69) becomes,
$c=S(0)-\left(\frac{\pi}{\mu}\right)$
Substitute equation (70) into equation (69) gives,
$S(t)=\left(\frac{\pi}{\mu}\right)-\left(\frac{\pi}{\mu}\right) \ell^{-\mu t}+S(0) \ell^{-\mu t}$
As $t \rightarrow \infty, S \rightarrow \frac{\pi}{\mu}, R \rightarrow 0$ regardless of the value of $S(0), R(0)$
Therefore,
$x_{0}=\left(\frac{\pi}{\mu}, 0\right)$ is globally asymptotically stable.
Case 2: Consider an infected subsystem
$y^{\prime}=G(x, y)=\left(\begin{array}{l}\beta S I-(\mu+\alpha) E \\ \alpha E-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I \\ \gamma_{1} I-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N} \\ \gamma_{2} T_{N}-\left(r_{3}+\mu+\delta_{3}\right) T_{D}\end{array}\right)$
Given that,
$G(x, y)=C y-\hat{G}(x, y)$
Then,
$\hat{G}(x, y)=C y-G(x, y)$
Where $C=\frac{\partial G(x, 0)}{\partial t}$
$C=\left[\begin{array}{cccc}-(\mu+\alpha) & \beta S & 0 & 0 \\ \alpha & -\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) & 0 & 0 \\ 0 & \gamma_{1} & -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) & 0 \\ 0 & 0 & \gamma_{2} & -\left(r_{3}+\mu+\delta_{3}\right)\end{array}\right]$
$C y=\left[\begin{array}{cccc}-(\mu+\alpha) & \beta S & 0 & 0 \\ \alpha & -\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) & 0 & 0 \\ 0 & \gamma_{1} & -\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) & 0 \\ 0 & 0 & \gamma_{2} & -\left(r_{3}+\mu+\delta_{3}\right)\end{array}\right]\left[\begin{array}{c}E \\ I \\ T_{N} \\ T_{D}\end{array}\right]$
Evaluating equation (76), we have
$C y=\left(\begin{array}{l}\beta S I-(\mu+\alpha) E \\ \alpha E-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I \\ \gamma_{1} I-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N} \\ \gamma_{2} T_{N}-\left(r_{3}+\mu+\delta_{3}\right) T_{D}\end{array}\right)$
Substituting equation (77) and equation (72) into equation (74), gives
$\hat{G}(x, y)=\left(\begin{array}{l}\beta S I-(\mu+\alpha) E \\ \alpha E-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I \\ \gamma_{1} I-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N} \\ \gamma_{2} T_{N}-\left(r_{3}+\mu+\delta_{3}\right) T_{D}\end{array}\right)-\left(\begin{array}{l}\beta S I-(\mu+\alpha) E \\ \alpha E-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I \\ \gamma_{1} I-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N} \\ \gamma_{2} T_{N}-\left(r_{3}+\mu+\delta_{3}\right) T_{D}\end{array}\right)$
$\hat{G}(x, y)=\left(\begin{array}{c}\hat{G}_{1}(x, y) \\ \hat{G}_{2}(x, y) \\ \hat{G}_{3}(x, y) \\ \hat{G}_{4}(x, y)\end{array}\right)=\left(\begin{array}{l}0 \\ 0 \\ 0 \\ 0\end{array}\right)$
Then, $\hat{G}(x, y)=0$.
Therefore, the disease-free equilibrium point is globally asymptotically stable when $R_{0}<1$

## ANALYTICAL SOLUTION OF EQUATION (4.4.1) USING HOMOTOPY PERTURBATION

## Method

The system of equation

$$
\left\{\begin{array}{l}
\frac{d S}{d t}+\mu S+\beta S I-\sigma R-\pi=0 \\
\frac{d E}{d t}+(\mu+\alpha) E-\beta S I=0 \\
\frac{d I}{d t}+\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I-\alpha E=0 \\
\frac{d T_{N}}{d t}+\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N}-\gamma_{1} I=0 \\
\frac{d T_{D}}{d t}+\left(r_{3}+\mu+\delta_{3}\right) T_{D}-\gamma_{2} T_{N}=0 \\
\frac{d R}{d t}+(\mu+\sigma) R-r_{3} T_{D}-r_{2} T_{N}-r_{1} I=0
\end{array}\right.
$$

Conditions for the differential equation above
We construct a homotopy:

$$
\left\{\begin{array}{l}
(1-h) \frac{d S}{d t}+h\left[\frac{d S}{d t}+\mu S+\beta S I-\sigma R-\pi\right]=0 \\
(1-h) \frac{d E}{d t}+h\left[\frac{d E}{d t}+(\mu+\alpha) E-\beta S I\right]=0 \\
(1-h) \frac{d I}{d t}+h\left[\frac{d I}{d t}+\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I-\alpha E\right]=0 \\
(1-h) \frac{d T_{N}}{d t}+h\left[\frac{d T_{N}}{d t}+\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N}-\gamma_{1} I\right]=0 \\
(1-h) \frac{d T_{D}}{d t}+h\left[\frac{d T_{D}}{d t}+\left(r_{3}+\mu+\delta_{3}\right) T_{D}-\gamma_{2} T_{N}\right]=0 \\
(1-h) \frac{d R}{d t}+h\left[\frac{d R}{d t}+(\mu+\sigma) R-r_{3} T_{D}-r_{2} T_{N}-r_{1} I\right]=0
\end{array}\right.
$$

The basic assumption is that the solution of equation (1) can be written as a series in powers of p

$$
\left\{\begin{array}{l}
S(t)=a_{0}+h a_{1}+h^{2} a_{2}+\ldots \\
E(t)=b_{0}+h b_{1}+h^{2} b_{2}+\ldots \\
I(t)=c_{0}+h c_{1}+h^{2} c_{2}+\ldots \\
T_{N}(t)=d_{0}+h d_{1}+h^{2} d_{2}+\ldots \\
T_{D}(t)=e_{0}+h e_{1}+h^{2} e_{2}+\ldots \\
R(t)=f_{0}+h f_{1}+h^{2} f_{2}+\ldots
\end{array}\right.
$$

If $p \rightarrow 1$, we recall the solution of the series
Substituting equation (4) into equation (3), we will have;

$$
\begin{aligned}
& {\left[\left(a_{0}{ }_{0}+h a^{\prime}{ }_{1}+h^{2} a^{\prime}{ }_{2}+\ldots\right)+h\left[\begin{array}{l}
\beta\left(a_{0}+h a_{1}+h^{2} a_{2}+\ldots\right)\left(c_{0}+h c_{1}+h^{2} c_{2}+\ldots\right)+ \\
\mu\left(a_{0}+h a_{1}+h^{2} a_{2}+\ldots\right)-\sigma\left(f_{0}+h f_{1}+h^{2} f_{2}+\ldots\right)-\pi
\end{array}\right]=0\right.} \\
& \left(b_{0}^{\prime}+h b_{1}^{\prime}+h^{2} b_{2}^{\prime}+\ldots\right)+h\left[\begin{array}{l}
(\mu+\alpha)\left(b_{0}+h b_{1}+h^{2} b_{2}+\ldots\right)-\beta\left(a_{0}+h a_{1}+h^{2} a_{2}+\ldots\right) \\
\left(c_{0}+h c_{1}+h^{2} c_{2}+\ldots\right)
\end{array}\right]=0 \\
& \left(c^{\prime}{ }_{0}+h c^{\prime}{ }_{1}+h^{2} c^{\prime}{ }_{2}+\ldots\right)+h\left[\begin{array}{l}
\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)\left(c_{0}+h c_{1}+h^{2} b_{2}+\ldots\right)- \\
\alpha\left(b_{0}+h b_{1}+h^{2} b_{2}+\ldots\right)
\end{array}\right]=0 \\
& \left(d^{\prime}{ }_{0}+h d^{\prime}{ }_{1}+h^{2} d^{\prime}{ }_{2}+\ldots\right)+h\left[\begin{array}{l}
\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)\left(d_{0}+h d_{1}+h^{2} d_{2}+\ldots\right)- \\
\gamma_{1}\left(c_{0}+h c_{1}+h^{2} c_{2}+\ldots\right)
\end{array}\right]=0 \\
& \left(e_{0}^{\prime}+h e_{1}^{\prime}+h^{2} e^{\prime}+\ldots\right)+h\left[\begin{array}{l}
\left(r_{3}+\mu+\delta_{3}\right)\left(e_{0}+h e_{1}+h^{2} e_{2}+\ldots\right)- \\
\gamma_{2}\left(d_{0}+h d_{1}+h^{2} d_{2}+\ldots\right)
\end{array}\right]=0 \\
& \left(f^{\prime}{ }_{0}+h f^{\prime}{ }_{1}+h^{2} f^{\prime}{ }_{2}+\ldots\right)+h\left[\begin{array}{l}
(\mu+\sigma)\left(f_{0}+h f_{1}+h^{2} f_{2}+\ldots\right)-r_{3}\left(e_{0}+h e_{1}+h^{2} e_{2}+\ldots\right) \\
r_{2}\left(d_{0}+h d_{1}+h^{2} d_{2}+\ldots\right)-r_{1}\left(c_{0}+h c_{1}+h^{2} c_{2}+\ldots\right)
\end{array}\right]=0
\end{aligned}
$$

We now compare the identical powers of $p$ as follow:

$$
\left\{\begin{array}{l}
S(t)=S_{0}+\left(\pi+\sigma R_{0}-\beta S_{0} I_{0}-\mu S_{0}\right) t+ \\
\left(\begin{array}{l}
\sigma\left(r_{1} I_{0}+r_{2} T_{N 0}+r_{3} T_{D 0}-(\mu+\sigma) R_{0}\right)-\beta\binom{\left(\pi+\sigma R_{0}-\beta S_{0} I_{0}-\mu S_{0}\right) I_{0}+}{S_{0}\left(\alpha E_{0}-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I_{0}\right)}
\end{array}\right) \frac{t^{2}}{2} \\
-\mu\left(\pi+\sigma R_{0}-\beta S_{0} I_{0}-\mu S_{0}\right) \\
E(t)=E_{0}+\left(\beta S_{0} I_{0}-(\mu+\alpha) E_{0}\right) t+ \\
\binom{\left.\beta\left(\left(\pi+\sigma R_{0}-\beta S_{0} I_{0}-\mu S_{0}\right) I_{0}+S_{0}\left(\alpha E_{0}-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I_{0}\right)\right)\right) \frac{t^{2}}{2}}{-(\mu+\alpha)\left(\beta S_{0} I_{0}-(\mu+\alpha) E_{0}\right)} \frac{I_{0}}{2} \\
I(t)=I_{0}+\left(\alpha E_{0}-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I_{0}\right) t+\binom{\alpha\left(\beta S_{0} I_{0}-(\mu+\alpha) E_{0}\right)-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right)}{\left(\alpha E_{0}-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I_{0}\right)} \frac{t^{2}}{2} \\
T_{N}(t)=T_{N 0}+\left(\gamma_{1} I_{0}-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N 0}\right) t+\binom{\gamma_{1}\left(\alpha E_{0}-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I_{0}\right)-}{\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right)\left(\gamma_{1} I_{0}-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N 0}\right)} \frac{t^{2}}{2} \\
T_{D}(t)=T_{D 0}+\left(\gamma_{2} T_{N 0}-\left(r_{3}+\mu+\delta_{3}\right) T_{D 0}\right) t+\binom{\gamma_{2}\left(\gamma_{1} I_{0}-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N 0}\right)-}{\left(r_{3}+\mu+\delta_{3}\right)\left(\gamma_{2} T_{N 0}-\left(r_{3}+\mu+\delta_{3}\right) T_{D 0}\right)} \frac{t^{2}}{2} \\
R(t)=R_{0}+\left(r_{1} I_{0}+r_{2} T_{N 0}+r_{3} T_{D 0}-(\mu+\sigma) R_{0}\right) t+ \\
\binom{r_{1}\left(\alpha E_{0}-\left(\gamma_{1}+r_{1}+\mu+\delta_{1}\right) I_{0}\right)-r_{2}\left(\gamma_{1} I_{0}-\left(\gamma_{2}+r_{2}+\mu+\delta_{2}\right) T_{N 0}\right)-}{r_{3}\left(\gamma_{2} T_{N 0}-\left(r_{3}+\mu+\delta_{3}\right) T_{D 0}\right)+(\mu+\sigma)\left(r_{1} I_{0}+r_{2} T_{N 0}+r_{3} T_{D 0}-(\mu+\sigma) R_{0}\right)} \frac{t^{2}}{2}
\end{array}\right.
$$

## NUMERICAL SIMULATION AND RESULTS

In this chapter, the model will be analyzed using the parameter values as well as estimated initial values of the susceptible, exposed, infected, recovered, and vaccinated individuals. The results obtained will be discussed. MATLAB was used to get the numerical solution of the model using ode45.
Table 1 shows the Notation and definition of variables and parameter used in the paper.
Table 2 parameter and estimated values for initial conditions for the $S E I T_{N} T_{D} R$ models

## DISCUSSION OF RESULTS

Figure 2: Is the graph of susceptible human against time for different of rate at which the susceptible
Table 1: Notation and definition of variables and parameter

| Symbol | Description |
| :--- | :--- |
| $S(t)$ | Susceptible at time t |
| $E_{( }(t)$ | Exposed at time t |
| $I(t)$ | Infected at time t |
| $T_{N}(t)$ | First Line Treatment t |
| $T_{D}(t)$ | Second Line Treatment t |
| $R(t)$ | Recovered t |
| $\pi$ | Recruitment rate |
| $\sigma$ | Lose their immunity at the rate |
| $\beta$ | Rate at which the susceptible become exposed to Mtb |
| $\alpha$ | Infection rate |
| $\gamma_{1}$ | First line treatment class |
| $\gamma_{2}$ | Second line treatment class |
| $r_{1}, r_{2} r_{3}$ | Reduces at rate of treatment class |
| $\propto$ | Rate of natural death |
| $\delta_{1}$ | Disease induced death rate |
| $\delta_{2}$ | Disease induced death rate |
| $\delta_{3}$ | Disease induced death rate |

Table 2: Variables and model parameters values

| Parameters and State Variables | Value | Source |
| :--- | :---: | :---: |
| $\beta$ | 0.00003 | Assumed |
| $\sigma$ | 0.470104 | Assumed |
| $\mu$ | 0.019896 | Assumed |
| $\alpha$ | 0.01 | Assumed |
| $S$ | 85,000 | Assumed |
| $E$ | 32,000 | Assumed |
| $I$ | 1800 | Assumed |
| $T_{N}$ | 2300 | Assumed |
| $T_{D}$ | 1500 | Assumed |
| $R$ | 1700 | Assumed |
| $r_{1}$ | 0.3 | Assumed |
| $r_{2}$ | 0.3 | Assumed |
| $r_{3}$ | 0.2 | Assumed |
| $\gamma_{1}$ | 0.5 | Assumed |
| $\gamma_{2}$ | 0.2 | Assumed |
| $\delta_{1}$ | 0.01 | Assumed |
| $\delta_{2}$ | 0.01 | Assumed |
| $\delta_{3}$ | 0.01 | Assumed |
| $\pi$ | 0.03 | Assumed |

become exposed to Mtb. It is observed that the population susceptible human decreases as the rate at which the susceptible become exposed to Mtb increases.
Figure 3: Is the graph of exposed individual against time for different of contact rate for human. It is observed that the population exposed individual increases as the rate at which the susceptible become exposed to Mtb increases.
Figure 4: Is the graph of first line treatment individuals against time for different of first line treatment class. It is observed that the population first line treatment decreases as the first line treatment class increases.
Figure 5: Is the graph of second line treatment individuals against time for different of second line treatment class. It is observed that the population second line treatment decreases as the second line treatment class increases.


Figure 1: Schematic diagram of the model


Figure 2: Graph of Susceptible individuals against time for different of Rate at which the susceptible become exposed to Mtb


Figure 3: Graph of Exposed individuals against time for different of Rate at which the susceptible become exposed to Mtb

Figure 6: Is the graph of infected individuals against time for different of first line treatment class. It is observed that the population infected decreases as the first line treatment class increases.
Figure 7: Is the graph of recovered individuals against time for different of reduces at rate of treatment class. It is observed that the population of recovered increases as reduces at rate of treatment class increases.

## SUMMARY

TB is the most dreadful diseases that could be transmitted from human. To curb its spread, a mathematical model was proposed in this work so as to understand the transmission dynamics of the disease and to


Figure 4: Graph of Frist Line Treatment individuals against time for different of first line treatment class


Figure 5: Graph of Second Line Treatment individuals against time for different of second line treatment class


Figure 6: Graph of Infected individuals against time for different of first line treatment class


Figure 7: Graph of Recovered individuals against time for different of reduces at rate of treatment class
proffer solution by introducing vaccination to facilitate the prevention of the spread of the disease.
The population under consideration was a non-constant population. The SEIR and SEIV models were considered and the disease free points were obtained. Their basic reproduction number was obtained using the next generation matrix approach. The stability of the disease free points was analyzed. With estimated values for parameters and initial numbers of the population classes, MATLAB was used to get the numerical solution of the model using ode 45.

## CONCLUSION

Mathematical model for TB disease was developed in this study. The disease-free state (DFE) was analyzed for stability, and it revealed that it is stable. The reproduction number was analyzed, and the result shows the stability of the disease, which implies that the disease would be reduced after the treatment used as a control parameter.
From the numerical simulation, we observe the different dynamics of the population from the graphical profiles of Figure 6 is it observed infected individuals decreases at first line treatment increases and from Figure 7, at rate of treatment class increases, its observed that the disease will be eradicated completely from the population.
The following findings are established:

1. Existence of the disease free state
2. Existence of the endemic state
3. Effective reproduction number shows the stability of the disease.

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