

1 **Mathematical modeling of typhoid fever disease incorporating unprotected humans**
2 **in the spread dynamics.**

3

4 **Abstract**

5 In this study we have develop a deterministic mathematical model for spread dynamics of
6 typhoid fever disease incorporating unprotected humans. The model result into a system of
7 ordinary differential equations which are used to study the spread dynamics of typhoid fever.
8 The model incorporating Susceptible, unprotected, Infectious and Recovered humans which are
9 analyzed mathematically. The existence of steady states of the mathematical model is
10 determined. More so we show the existence and positivity of a solution and finally computed the
11 basic reproductive number using next generation matrix.

12 **Key words:** Basic reproduction number, invariant region, positivity of solution, Mathematical
13 model, Disease Free Equilibrium, Endemic equilibrium point.

14 **Introduction**

15 Typhoid fever is an endemic disease that is classified as an enteritis disease. The disease is
16 caused by a bacterium called *Salmonella Typhi*. It is a common infectious disease in human beings
17 and is transmitted through food and water contaminated with faeces and urine of an infected
18 person [3].The disease is endemic in developing countries where it continuously causes illness
19 and death. This is contributed by unsafe water supply, poor food hygiene and wanting
20 environmental sanitation. According to World Health Organization an estimated 17 million
21 illness cases of typhoid fever were reported per year worldwide resulting to 0.6 million deaths
22 annually[4,5].

23 **1. Description and model formulation**

24 We formulated a deterministic model for spread dynamics of typhoid fever that considers
25 human population at time t . The model is divided into four compartments as follows.
26 Susceptible(S), Unprotected (E), Infective (I) and Recovered(R).The model has the following
27 flow. $S \rightarrow E \rightarrow I \rightarrow R \rightarrow S$.We use the following parameters in our model.(i) μ is the natural
28 death rate (ii) α is the disease induced death rate.(iii) Λ human recruitment rate (birth). (iv) β
29 disease interaction rate .(v) Ω unprotected symptoms showing rate(vi) γ Infective recovery rate
30 and finally(vii) δ this is the rate at which recovered humans loses temporary immunity obtained
31 through treatment and get the disease back again. All the compartments are positive in the
32 feasible region ϕ where $\{S, E, I, R\} \in \phi \subset R_+^4$. All the solutions are also bounded in ϕ such that

33 $0 \leq N \leq \frac{\Lambda}{\mu}$. Thus the model is epidemiologically well posed in the region ϕ .

34 The following flow chart shows various compartments in the model.

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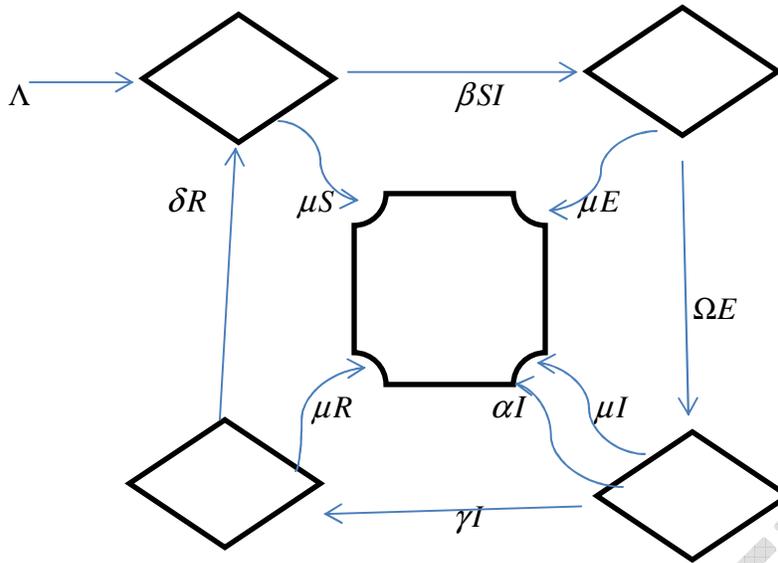
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The model dynamics results to four differential equations as shown equation 1.

$$\begin{aligned}
 \frac{dS}{dt} &= \Lambda + \delta R - \beta SI - \mu S \\
 \frac{dE}{dt} &= \beta SI - \Omega E - \mu E \\
 \frac{dI}{dt} &= \Omega E - \gamma I - \alpha I - \mu I \\
 \frac{dR}{dt} &= \gamma I - \delta R - \mu R
 \end{aligned}
 \tag{1}$$

2. Disease free equilibrium point and endemic equilibrium point

The disease free equilibrium of the model is obtained by setting

$$\frac{dS}{dt} = \frac{dE}{dt} = \frac{dI}{dt} = \frac{dR}{dt} = 0$$

In absence of disease

$$E = 0, I = 0, R = 0$$

Setting the right hand side of equations of system 1 to zero we have

$$\begin{aligned}
& \Lambda + \delta R - \beta SI - \mu S = 0. \\
& \beta SI - \Omega E - \mu E = 0 \\
& \Omega E - \gamma I - \alpha I - \mu I = 0 \\
& \gamma I - \delta R - \mu R = 0
\end{aligned}
\tag{2}$$

Hence model has a disease free equilibrium given by

$$(S^* E^* I^* R^*) = \left(\frac{\Lambda}{\mu}, 0, 0, 0 \right)
\tag{3}$$

The basic reproductive number (R_0) which is average number of secondary infections caused by one infectious individual introduced in a completely susceptible population is obtained using

next generation matrix as $R_0 = \frac{\beta S \Omega}{(\Omega + \mu)(\alpha + \mu + \gamma)}$ where at disease free equilibrium

$$R_0 = \frac{\beta \Lambda \Omega}{\mu(\Omega + \mu)(\alpha + \mu + \gamma)}.$$

Theorem 1

If $\Omega < \frac{\mu^2(\alpha + \mu + \gamma)}{\beta \Lambda - u(\alpha + \mu + \gamma)}$, there disease free equilibrium will be stable and typhoid disease will not have a hand in the population.

Proof

When $R < 1$; this means that $\frac{\beta \Lambda \Omega}{\mu(\Omega + \mu)(\alpha + \mu + \gamma)} < 1$.

Making Ω the subject, $\Omega < \frac{\mu^2(\alpha + \mu + \gamma)}{\beta \Lambda - u(\alpha + \mu + \gamma)}$

Disease free equilibrium point therefore is locally asymptotically stable if the basic reproduction number (R_0) less than one ($R_0 < 1$) and unstable if the basic reproduction number is greater than ($R_0 > 1$).

3. Endemic equilibrium point

Endemic equilibrium E_2^* ; disease exists. Evaluating the state variables of equations of the system 2, the endemic equilibrium points are in this form

$$E_2^* = \{S^{**}, E^{**}, I^{**}, R^{**}\}$$

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Where

$$S^{**} = \frac{(\Omega + \mu)(\gamma + \alpha + \mu)}{\Omega\beta}$$

$$E^{**} = \frac{(\gamma + \alpha + \mu)(\delta + \mu)\{\Lambda\Omega\beta - \mu\{(\Omega + \mu)(\gamma + \alpha + \mu)\}\}}{\beta\Omega\{(\delta + \mu)(\Omega + \mu)(\gamma + \alpha + \mu)\} - \gamma\Omega\delta}$$

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$$I^{**} = \frac{(\delta + \mu)}{\beta} \cdot \frac{\Lambda\Omega\beta - \mu\{(\Omega + \mu)(\gamma + \alpha + \mu)\}}{\{(\delta + \mu)(\Omega + \mu)(\gamma + \alpha + \mu)\} - \gamma\Omega\delta} \quad (4)$$

$$R^{**} = \frac{1}{\beta} \left\{ \frac{\gamma\Lambda\Omega\beta - \gamma\mu\{(\Omega + \mu)(\gamma + \alpha + \mu)\}}{\{(\delta + \mu)(\Omega + \mu)(\gamma + \alpha + \mu)\} - \gamma\Omega\delta} \right\}$$

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4. Stability of endemic equilibrium

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Proof: By the use of lyapunov function defined by LaSalle [1976]

$$L(S^{**}, E^{**}, I^{**}, R^{**}) = (S - S^{**} - S^{**} \ln \frac{S^{**}}{S}) + \left(E - E^{**} - E^{**} \ln \frac{E^{**}}{E} \right) + I - I^{**} - I^{**} \ln \frac{I^{**}}{I} + R - R^{**} - R^{**} \ln \frac{R^{**}}{R}$$

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computing the derivative of L along the solutions of the system is directly:

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$$\frac{dL}{dt} = \left(\frac{s - s^{**}}{s} \right) \frac{dS}{dt} + \left(\frac{E - E^{**}}{E} \right) \frac{dE}{dt} + \left(\frac{I - I^{**}}{I} \right) \frac{dI}{dt} + \left(\frac{R - R^{**}}{R} \right) \frac{dR}{dt} \quad (5)$$

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Substituting the equations of system 1 in equation 5, the equation becomes

$$\begin{aligned} \frac{dL}{dt} = & \left[\left(\frac{s - s^{**}}{s} \right) \Lambda + \delta R - (\beta I + \mu) S \right] + \left[\left(\frac{E - E^{**}}{E} \right) \beta SI - (\Omega + \mu) E \right] + \\ & \left[\left(\frac{I - I^{**}}{I} \right) \Omega E - (\gamma + \alpha + \mu) I \right] + \left[\left(\frac{R - R^{**}}{R} \right) \gamma I - (\delta + \mu) R \right] \end{aligned} \quad (6)$$

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Expanding equation 6, it produces

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$$\begin{aligned} \frac{dL}{dt} = & \Lambda + \delta R - (\beta I + \mu) S - \Lambda \frac{S^{**}}{S} - \delta R \frac{S^{**}}{S} + (\beta I + \mu) S^{**} + \beta SI - (\Omega + \mu) E - \beta SI \frac{E^{**}}{E} + \\ & (\Omega + \mu) E^{**} + \Omega E - (\gamma + \alpha + \mu) I - \Omega E \frac{I^{**}}{I} + (\gamma + \alpha + \mu) I^{**} + \gamma I - (\delta + \mu) R - \gamma I \frac{R^{**}}{R} + (\delta + \mu) R^{**} \end{aligned}$$

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Further simplification result to

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$$\begin{aligned} \frac{dL}{dt} = & \left[\Lambda + \delta R + (\beta I + \mu) S^{**} + \beta SI + (\Omega + \mu) E^{**} + \Omega E + (\gamma + \alpha + \mu) I^{**} + \gamma I + (\delta + \mu) R^{**} \right] \\ & + \left[-(\beta I + \mu) S - \Lambda \frac{S^{**}}{S} - \delta R \frac{S^{**}}{S} - (\Omega + \mu) E - \beta SI \frac{E^{**}}{E} - (\gamma + \alpha + \mu) I - \Omega E \frac{I^{**}}{I} - (\delta + \mu) R - \gamma I \frac{R^{**}}{R} \right] \end{aligned}$$

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Or

$$\frac{dL}{dt} = \left[\Lambda + \delta R + (\beta I + \mu) S^{**} + \beta SI + (\Omega + \mu) E^{**} + \Omega E + (\gamma + \alpha + \mu) I^{**} + \gamma I + (\delta + \mu) R^{**} \right]$$

$$- \left[(\beta I + \mu) S + \Lambda \frac{S^{**}}{S} + \delta R \frac{S^{**}}{S} + (\Omega + \mu) E + \beta SI \frac{E^{**}}{E} + (\gamma + \alpha + \mu) I + \Omega E \frac{I^{**}}{I} + (\delta + \mu) R + \gamma I \frac{R^{**}}{R} \right]$$

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89 From equation it's clear that ; $\frac{dL}{dt} = A - B$. Where A are the positive terms and B are the
90 negative ones, such that;

$$A = \Lambda + \delta R + (\beta I + \mu) S^{**} + \beta SI + (\Omega + \mu) E^{**} + \Omega E + (\gamma + \alpha + \mu) I^{**} + \gamma I + (\delta + \mu) R^{**}$$

$$B = (\beta I + \mu) S + \Lambda \frac{S^{**}}{S} + \delta R \frac{S^{**}}{S} + (\Omega + \mu) E + \beta SI \frac{E^{**}}{E} + (\gamma + \alpha + \mu) I + \Omega E \frac{I^{**}}{I} + (\delta + \mu) R + \gamma I \frac{R^{**}}{R}$$

91
92 If $A < B$ then $\frac{dL}{dt} \leq 0$

93 $\frac{dL}{dt} = 0$ Only if $S = S^{**}, E = E^{**}, I = I^{**}, R = R^{**}$

94 The largest invariant set in $\{(S, E, I, R) \in \varphi : \frac{dL}{dt} = 0\}$ is a singleton E_2^* . Where E_2^* is the
95 endemic equilibrium. Therefore, the endemic equilibrium is globally asymptotically stable
96 in the invariant region φ if $A < B$ [1,2].

97 Conclusion.

98 From our finding if $\Omega < \frac{\mu^2 (\alpha + \mu + \gamma)}{\beta \Lambda - \mu (\alpha + \mu + \gamma)}$, there disease equilibrium will be stable and typhoid

99 disease will not have a hand in the population. However if $\Omega > \frac{\mu^2 (\alpha + \mu + \gamma)}{\beta \Lambda - \mu (\alpha + \mu + \gamma)}$, then disease

100 will be dependent on prevailing circumstances. We also performed numerical simulations to
101 determine the changes in various compartments with time using MATLAB ode solve software.
102 There is direct variation relationship between the unprotected and infectious compartments,
103 therefore the unprotected humans contribute significantly to the spread dynamics of typhoid
104 fever disease.

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