# Population Dynamics of Bactria inside Humans

Battle of Bacteria, Antibiotics and Immune system

Pooria Assarehha Mani Moradi Mohammad Hossein Naderi

Department of Mathematics, Statistics and Computer Science



Summer 2025

- Introduction
- 2 Dynamic Model
- Equiliberia
- 4 Stability Analysis
- Numerical Computation
- 6 Code
- References



- Bacteria lives inside us
- Some of them unwanted
- Antibiotics to get rid of them
- interactions lead to different equiliberia

- Introduction
- 2 Dynamic Model
- 3 Equiliberia
- 4 Stability Analysis
- Numerical Computation
- 6 Code
- References



#### Variable Definitions

A: Concentration of Antibiotics

5 : Biomass of Non-Antibiotic-Resistant Bacteria

R: Biomass of Antibiotic-Resistant Bacteria

P: Immune Cell Population

N: S + R total bacteria

 $\Psi$  and  $\Phi$ : functions in  $\mathcal{C}^1(\mathbb{R}_+)$ 

# Model Equations

#### Modified Logistic Model

$$\dot{A}(t) = \Lambda - \mu A, 
\dot{S}(t) = \eta_s \left( 1 - \frac{S+R}{K} \right) S - \bar{\alpha} A S - \beta \frac{SR}{N} - \Gamma S P, 
\dot{R}(t) = \eta_r \left( 1 - \frac{S+R}{K} \right) R + \beta \frac{SR}{N} - \Gamma R P, 
\dot{P}(t) = \Phi(N) P \left( 1 - \frac{P}{P_{max}} \right) - \Psi(N) P,$$

#### Parameter and Term Definitions

Λ : administration rate of Antibiotics

 $\mu$ : absorbtion rate of Antibiotic

 $\eta_S$  and  $\eta_R$ : reproduction rate of S and R

K: carrying capacity (Limiting the reproduction)

 $\Gamma$ : transfer rate of resistant gene

 $P_{\text{max}}$ : limit of prolifiration of immune cells

# **Key Assumptions**

- $\eta_S > \eta_R$  cost of resistance
- Resistance genes transfer  $\beta \frac{SR}{N}$
- Immune response  $\Gamma SP \Gamma RP$

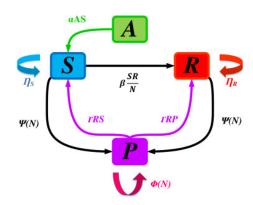


Fig. 1: Schematic Diagram [1]

#### **Transformation**

$$a = \frac{A}{A/\mu} \qquad s = \frac{S}{K} \qquad r = \frac{R}{K} \qquad p = \frac{P}{P_{\text{max}}}$$

$$\alpha = \frac{\bar{\alpha}A}{\mu} \quad \gamma = \Gamma P_{\text{max}} \quad n = s + r \quad \phi(n) = \Phi(Kn) \quad \psi(n) = \Psi(Kn)$$

$$\dot{a}(t) = \mu(1 - a)$$

$$\dot{s}(t) = \eta_s(1 - n)s - \alpha as - \beta \frac{sr}{n} - \gamma sp$$

$$\dot{r}(t) = \eta_r(1 - n)r + \beta \frac{sr}{n} - \gamma rp$$

$$\dot{p}(t) = \phi(n)p(1 - p) - \psi(n)p$$

#### **Boundedness**

- $\mathbb{R}^4_+ = \{(a, s, r, p) \in \mathbb{R}^4 \mid a \ge 0 \ s \ge 0 \ , \ r \ge 0 \ , \ p \ge 0\}$
- ullet Right hand side of our system  $\in \mathcal{C}^1\left(\mathsf{Int}(\mathbb{R}^4_+),\mathbb{R}^4_+\right)$
- Unique Solution exists  $\in [0, T_{\text{max}}]$
- $\mathcal{A} = \left\{ (a, s, r, p) \in \operatorname{Int}\left(\mathbb{R}^4_+\right), a \leq 1, s + r \leq 1, p \leq 1 \right\}$
- ullet lemma:  ${\cal A}$  is positively invariant with recpect to 1
- hence mathematically and biologically well posed



- Introduction
- 2 Dynamic Model
- 3 Equiliberia
- 4 Stability Analysis
- 5 Numerical Computation
- 6 Code
- References



Analysing reveals 4 Equilibria - biologically meaningfull

- clearance of infection
- infection under S
- infection under R
- infection under both

### deriving Equilibria

- Equiliberia  $\equiv$  Zero Change  $\equiv$   $\dot{\mathbf{X}} = \vec{0}$
- let  $f(n) = 1 \frac{\psi(n)}{\phi(n)}$  with f(0) > 0 and simplify
- we want  $f \uparrow$  for small n and  $f \downarrow$  for large n (How Immunity works)
- f either remains + or after a threshold drops < 0
- we focus on 2nd scenario (remmeber n is bounded )

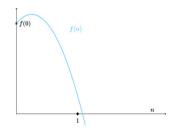


Fig. 2: Schematic Rep. of what f can be [1]

$$\mu(1-a) = 0 \qquad \qquad a = 1$$

$$\eta_s(1-n)s - \alpha as - \beta \frac{sr}{n} - \gamma sp = 0 \qquad \Rightarrow \qquad s = 0 \quad \text{or} \quad \eta_s(1-n) - \alpha - \beta \frac{r}{n} - \gamma p = 0$$

$$\eta_r(1-n)r + \beta \frac{sr}{n} - \gamma rp = 0 \qquad \qquad r = 0 \quad \text{or} \quad \eta_r(1-n) + \beta \frac{s}{n} - \gamma p = 0$$

$$\phi(n)p(f(n)-p) = 0 \qquad \qquad p = 0 \quad \text{or} \quad p = f(n)$$



# Equiliberia

#### 7 Cases

- Case 1:  $E_0(1,0,0,0)$  r=s=p=0
- Case 2:  $E_1(1,0,0,f(0))$  r=s=0 ,  $p \neq 0$
- Case 3:  $E_2(1,0,1,0)$  s=p=0 ,  $r\neq 0$

#### Case 4

Case 4: 
$$E_+$$
  $s=0$  ,  $r\neq 0$  ,  $p\neq 0$ 

$$r = \lambda_{+}$$
  $p = f(\lambda_{+})$   $\Rightarrow f(r) = \eta_{r} \frac{1-r}{\gamma}$ 

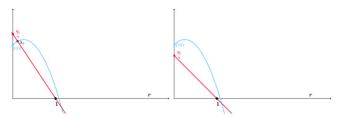


Fig. 3: f and  $\lambda_+$  [1]

if  $f(0) < rac{\eta_r}{\gamma}$  exists Unique  $0 < \lambda_+ < 1$ 



#### Case 5 an 6

- Case 5  $E_3(1,1-rac{lpha}{\eta_s},0,0)$  s 
  eq 0 , r=0 , p=0
- Case 6  $E_ s \neq 0$  , r = 0 ,  $p \neq 0$

$$s = \lambda_{-}$$
 ,  $p = f(\lambda_{-}) \Rightarrow f(s) = \frac{\eta_{s} - \alpha}{\gamma} - \frac{\eta_{s}}{\gamma} s$ 

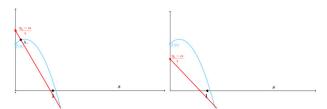


Fig. 4: f and  $\lambda_{-}$  [1]

Like Case 4 if  $f(0) < \frac{\eta_s - \alpha}{\gamma}$  exists Unique  $0 < \lambda_- < 1$ 



# Case 7: *E*\*

$$s \neq 0$$
,  $r \neq 0$ ,  $p \neq 0$   
 $n_* = s + r = 1 - \frac{\alpha + \beta}{\eta_s - \eta_r}$   
 $n_* > 0$  if  $\alpha + \beta + \eta_r < \eta_s$ 

$$p_* = f(n_*)$$

$$s_* = \frac{n_*}{\beta} \left( \gamma f(n_*) - \eta_r \frac{\alpha + \beta}{\eta_s - \eta_r} \right)$$

$$r_* = \frac{n_*}{\beta} \left( \eta_s \frac{\alpha + \beta}{\eta_s - \eta_r} - \alpha - \gamma f(n_*) \right)$$

$$p_*$$
,  $s_*$ ,  $r_* > 0$  if  $\eta_r \frac{\alpha + \beta}{\eta_s - \eta_r} < \gamma f(n_*) < \frac{\alpha \eta_r + \beta \eta_s}{\eta_s - \eta_r}$ 

- Introduction
- 2 Dynamic Model
- Equiliberia
- Stability Analysis
- 5 Numerical Computation
- 6 Code
- References



#### Linearinzation

- System is non linear,
- Linearinzation around fixed points ⇒ Jacobian Matrix
- Signs of the eigenvalues of the Jacobian matrix shall be determined

$$\begin{pmatrix} -\mu & 0 & 0 & 0 \\ -\alpha s & \eta_s(1-n) - \eta_s s - \alpha a - \beta \frac{r^2}{n^2} - \gamma p & -\eta_s s - \beta \frac{s^2}{n^2} & -\gamma s \\ 0 & -\eta_r r + \beta \frac{r^2}{n^2} & \eta_r(1-n) - \eta_r r + \beta \frac{s^2}{n^2} - \gamma p & -\gamma r \\ 0 & \phi(n)(f(n)-p) + \dot{f}(n)p\phi(n) & p\dot{\phi}(n)(f(n)-p) + \dot{f}(n)p\phi(n) & \phi(n)(f(n)-2p) \end{pmatrix}$$

• let  $C_+ = \eta_s (1 - \lambda_+) - \alpha - \beta$ ,  $C_- = \eta_r (1 - \lambda_-) + \beta$  and

$$C_* = -\frac{(\eta_{s}s_* + \eta_{r}r_*) \beta s_* r_* (\eta_{s} - \eta_{r})}{f(n_*)\phi(n_*)(n_*)^2 (\eta_{s}s_* + \eta_{r}r_* + f(n_*)\phi(n_*))} - \frac{\eta_{s}s_* - \eta_{r}r_*}{\eta_*}.$$

will show up for finding stability criteria



# Stability Table

Table 1: Conditions for the stability of equilibria.

Equilibrium	Biological existence	Stability
$E_0$ (1,0,0,0)	Always exists	Always unstable
$E_1$ (1,0,0, $f$ (0))	Always exists	$lpha > \eta_{s} \;  ext{and} \; \gamma f(0) > \eta_{r}$
$E_2$ (1,0,1,0)	Always exists	Always unstable
$E_3\left(1,1-rac{lpha}{\eta_s},0,0 ight)$	$\eta_s > \alpha$	Always unstable
$E_+$ $(1,0,\lambda_+,f(\lambda_+))$	$\eta_r > \gamma f(0)$	$\gamma f(\lambda_+) > C_+ \text{ and } \gamma f(\lambda_+) > \eta_r$ $\gamma f(\lambda) > C \text{ and } \gamma f(\lambda) > \eta_s$
$E$ $(1, \lambda, 0, f(\lambda))$	$\eta_s - \alpha > \gamma f(0)$	$\gamma f(\lambda) > \mathcal{C} \text{ and } \gamma f(\lambda) > \eta_s$
$E_* (1, s_*, r_*, f(n_*))$	$\begin{split} &\eta_s>\alpha\\ &\eta_r>\gamma f(0)\\ &\eta_s-\alpha>\gamma f(0)\\ &\begin{cases} \eta_r\frac{\alpha+\beta}{\eta_s-\eta_r}<\gamma f(n_*)<\frac{\eta_r\alpha+\eta_s\beta}{\eta_s-\eta_r}\\ \text{and} \end{cases} \end{split}$	$\gamma f(n_*) > C_*$
	$\left  \begin{array}{c} \eta_{s} > \eta_{r} + \alpha + \beta \end{array} \right $	

- Introduction
- 2 Dynamic Model
- 3 Equiliberia
- Stability Analysis
- **5** Numerical Computation
- 6 Code
- References



# **Numerical Computation**

• we have chosen the function  $f(n) = -n^2 + n + \frac{3}{4}$  the aligns with our hypothesis

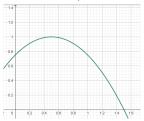


Fig. 5:  $f(n) = -n^2 + n + \frac{3}{4}$ 

 $\bullet$  Since  $E_1$  is the equilibria which the system will reach an infection free state we will display it

# **Numerical Computation**

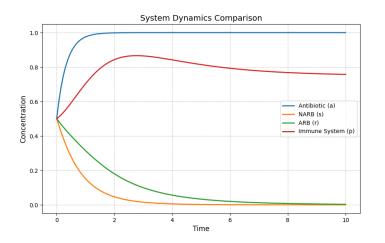


Fig. 6: Stability of  $E_1$ 

# **Numerical Computation**

Specifically, for  $E_1$  to be stable, two crucial conditions must be met:

- the antibiotic must be administered at a sufficiently high rate compared to the reproduction rate of non-resistant bacteria
- the immune response must be strong enough to outcompete and eliminate resistant bacteria

- Introduction
- 2 Dynamic Model
- 3 Equiliberia
- Stability Analysis
- Numerical Computation
- 6 Code
- References



### Code I

- Introduction
- 2 Dynamic Model
- 3 Equiliberia
- Stability Analysis
- Numerical Computation
- 6 Code
- References



#### References I

[1] Imene Meriem Mostefaoui and Abdellatif Seghiour. "Modeling the dynamics of interactions between antibiotic-resistant bacteria and immune response". In: Communications in Nonlinear Science and Numerical Simulation 140 (2025), p. 108412. ISSN: 1007-5704. DOI: https://doi.org/10.1016/j.cnsns.2024.108412. URL: https://www.sciencedirect.com/science/article/pii/ S1007570424005975.

#### For Your Attention

Thank You!