

Simulation and Theory of Bacterial Transformation

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Outline

Introduction

Motivation

Biological Background

Simulation

Results

Conclusions

Motivation

- Ubiquitous threat of antibiotic resistance
- Transmission of resistance via plasmids
- Transformation of susceptible bacteria to resistant
- Identify what most significantly affects resistant cell dominance.

Question

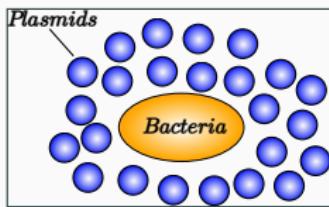
What conditions lead to emergence of antibiotic resistance?

Plasmids

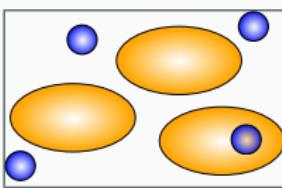
- Small, independently replicating genetic material
- Often include DNA segments encoding antibiotic resistance
- Imposes a fitness cost on host cell

Transformation and Conjugation

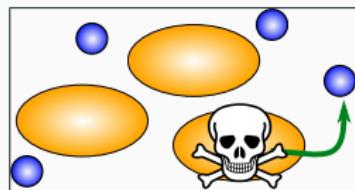
- **Conjugation:** Plasmid transferred between cells
- **Transformation:** Cell incorporates plasmid from environment
 - Three main regimes



Constant α



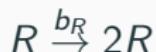
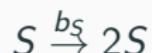
Linear α



Recycled α

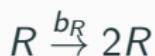
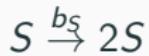
Population Dynamics Model - Constant

Reactions



Population Dynamics Model - Constant

Reactions



Equations

$$\frac{dS}{dt} = b_S \left(1 - \frac{S+R}{K}\right) S - \alpha S$$

$$\frac{dR}{dt} = b_R \left(1 - \frac{S+R}{K}\right) R + \alpha S - \delta R$$

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Simulation

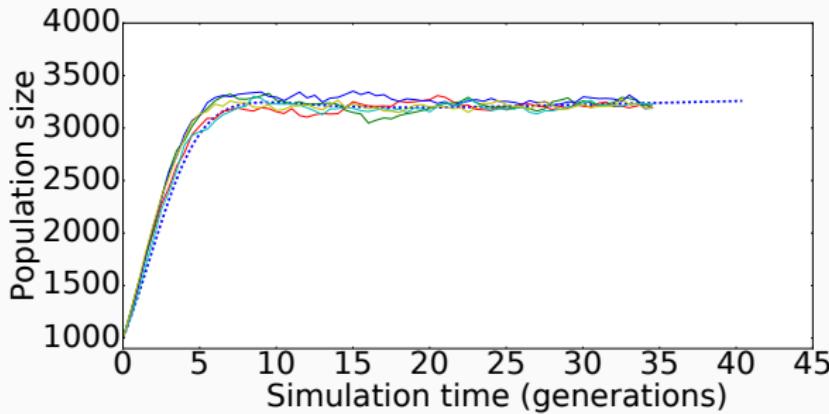
Population Dynamics Model

Results

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Simulation vs Modeling

- Stochastic vs Deterministic
- Information about average behavior vs specific trajectories
- Individual realizations noisily follow model



$$\frac{dS}{dt} = b_S \left(1 - \frac{S + R}{K}\right) S$$

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Constant α

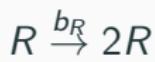
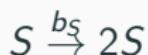
Linear α

Recycled α

Conclusions

Population Dynamics Model - Constant

Reactions

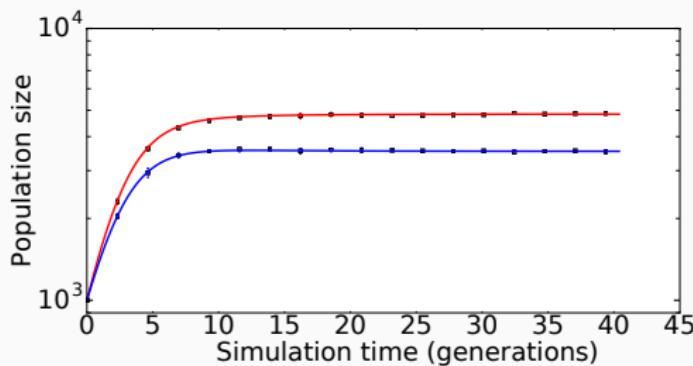


Equations

$$\frac{dS}{dt} = b_S \left(1 - \frac{S+R}{K}\right) S - \alpha S$$

$$\frac{dR}{dt} = b_R \left(1 - \frac{S+R}{K}\right) R + \alpha S - \delta R$$

Well-Mixed - Constant α



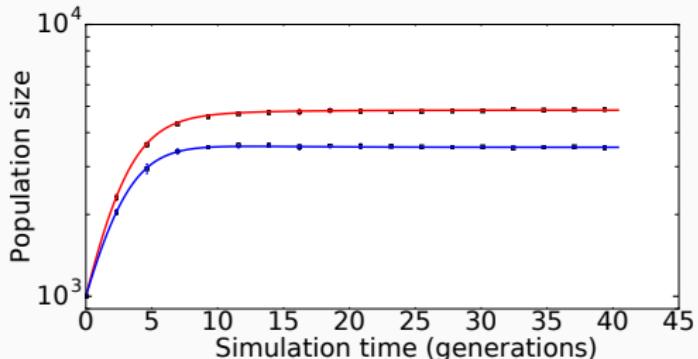
Parameters

α	.13	$\frac{b_S}{b_R}$	1.07
δ	.3	S_0	10^3
R_0	10^3	K	10^4

● Susceptible

● Resistant

Well-Mixed - Constant α

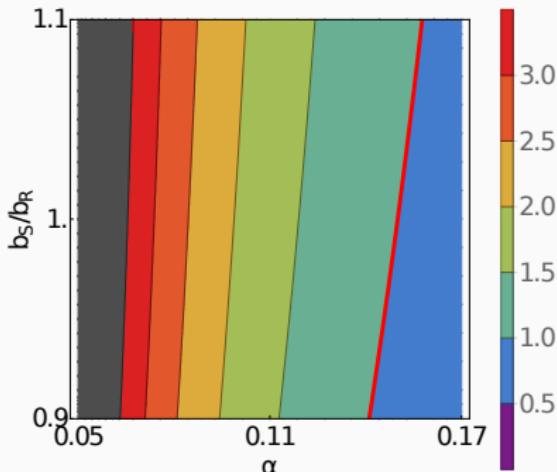


Parameters

α	.13	$\frac{b_S}{b_R}$	1.07
δ	.3	S_0	10^3
R_0	10^3	K	10^4

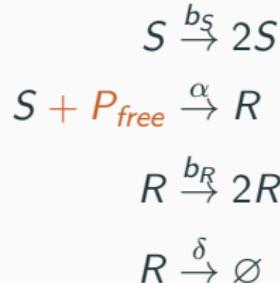
● Susceptible

● Resistant



Population Dynamics Model - Linear

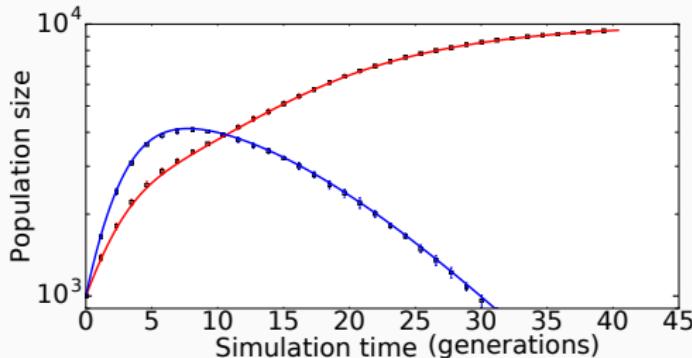
Reactions



Equations

$$\begin{aligned} \frac{dS}{dt} &= b_S \left(1 - \frac{S+R}{K}\right) S - \alpha \left(\frac{P_f}{P}\right) S \\ \frac{dR}{dt} &= b_R \left(1 - \frac{S+R}{K}\right) R + \alpha \left(\frac{P_f}{P}\right) S - \delta R \\ \frac{dP_f}{dt} &= -\alpha \left(\frac{P_f}{P}\right) S \\ \frac{dP}{dt} &= b_R \left(1 - \frac{S+R}{K}\right) R \end{aligned}$$

Well-Mixed - Linear α



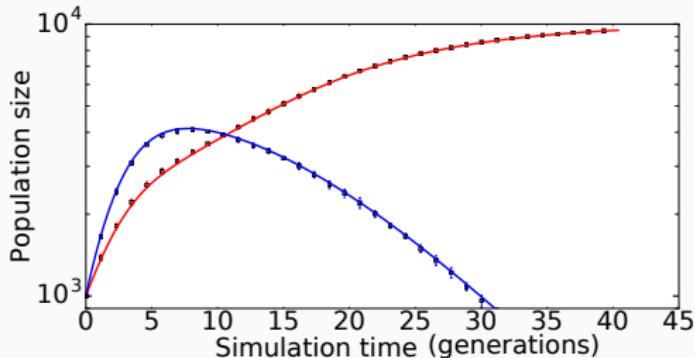
Parameters

α	.13	$\frac{b_S}{b_R}$	1.07
δ	.3	S_0	10^3
R_0	10^3	K	10^4
P_0	10^4		

● Susceptible

● Resistant

Well-Mixed - Linear α

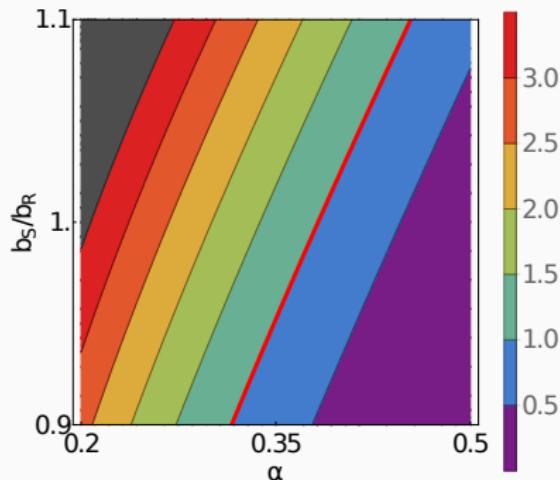


Parameters

α	.13	$\frac{b_S}{b_R}$	1.07
δ	.3	S_0	10^3
R_0	10^3	K	10^4
P_0	10^4		

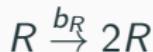
● Susceptible

● Resistant



Population Dynamics Model - Recycled

Reactions



Equations

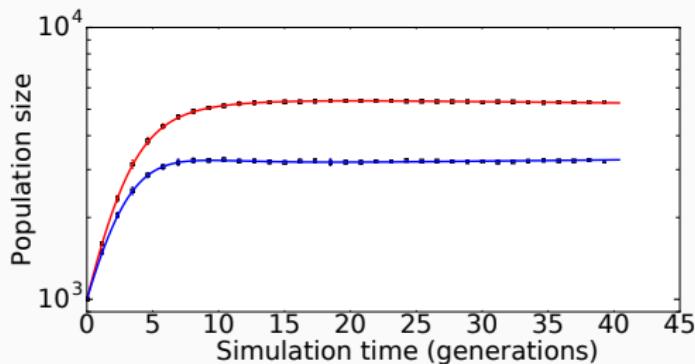
$$\frac{dS}{dt} = b_S \left(1 - \frac{S+R}{K}\right) S - \alpha \left(\frac{P_f}{P}\right) S$$

$$\frac{dR}{dt} = b_R \left(1 - \frac{S+R}{K}\right) R + \alpha \left(\frac{P_f}{P}\right) S - \delta R$$

$$\frac{dP_f}{dt} = -\alpha \left(\frac{P_f}{P}\right) S + \delta R$$

$$\frac{dP}{dt} = b_R \left(1 - \frac{S+R}{K}\right) R$$

Well-Mixed - Recycled α

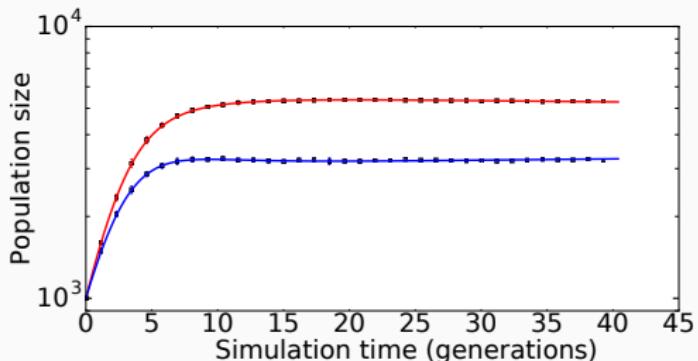


Parameters

α	.13	$\frac{b_S}{b_R}$	1.07
δ	.3	R_0	10^3
S_0	10^3	K	10^4
P_0	10^4		

● Resistant ● Susceptible

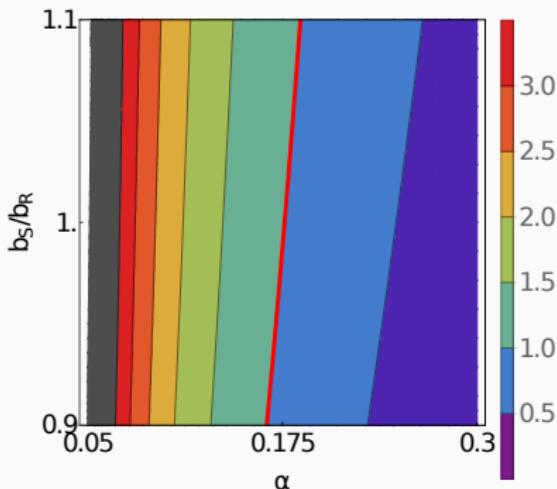
Well-Mixed - Recycled α



Parameters

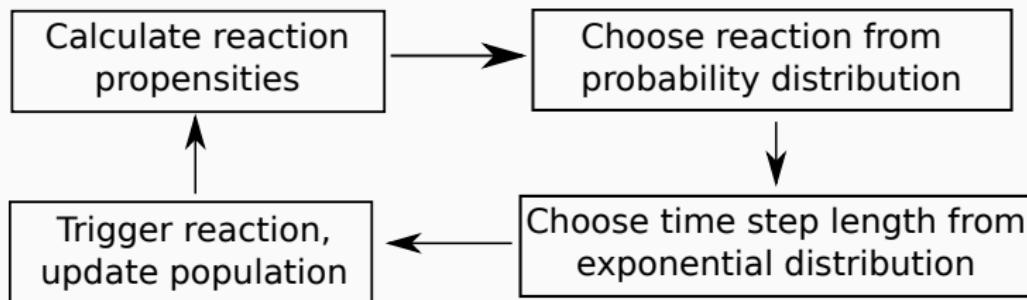
α	.13	$\frac{b_S}{b_R}$	1.07
δ	.3	R_0	10^3
S_0	10^3	K	10^4
P_0	10^4		

● Resistant ● Susceptible



Kinetic Monte Carlo Method

- Initially used to simulate chemical reactions
- Useful for simulating any reaction that occurs with a rate
- Captures information about dynamics of a growing system
- Self-adjusting timescales



Simulation Details

- Slower plasmid carrier growth rate
- Carrying capacity
- Periodic boundary conditions
- Fixed death rate
- Mapping simulation time to real time

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What conditions lead to emergence of antibiotic resistance?

- Transition point where population dominance switches
- Transition point heavily dependent on α mechanism, not growth rate ratio
- Little growth rate dependence, except linear case
- Linear case tends to population extinction, with interesting dynamics in between

Ongoing Work

- Continue gathering lattice data
- Simulate larger lattices

Future Work

- Incorporate diffusion
- Add conjugation reaction
- Simulate antibiotic dosing

Questions?

Acknowledgments

- Department of Physics and Astronomy, Bucknell University
- NSF-DMR #1248387
- Sandy!

Lattice Results

Optimizations

- Occupancy lists
- Sets vs. lists
- imshow vs. plcolor
- Parallelization

Gillespie Algorithm¹

1. Initialize simulation
2. Calculate propensity a for each reaction
3. Choose reaction μ according to the distribution

$$P(\text{reaction } \mu) = a_\mu / \sum_i a_i$$

4. Choose time step length τ according to the distribution

$$P(\tau) = \left(\sum_i a_i \right) \cdot \exp \left(-\tau \sum_i a_i \right)$$

5. Update populations with results of reaction
6. Go to Step 2

¹Heiko Rieger. *Kinetic Monte Carlo*. Powerpoint Presentation. 2012. URL:
https://www.uni-oldenburg.de/fileadmin/user_upload/physik/ag_compphys/download/Alexander/dpg_school/talk-rieger.pdf.

Mapping Simulation Time to Realtime

$$\frac{dS}{dt} = b_S S$$

$$\begin{aligned} S(t) &= S_0 e^{b_S t} \\ &= S_0 e^{\ln 2 t / \tau} \end{aligned}$$

$$b_S = \ln 2 / \tau$$

$$\tau = \ln 2 / b_S$$

Lattice Simulation Results

Sharp Transition

Lattice Simulation Results

Long Run

Lattice Simulation Results

Large Lattice