How Quorum Sensing Interactions Affect Population Structure 02-712 Final Project

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Background

Quorum-Sensing Systems

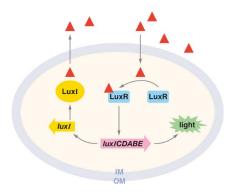


Figure 1: Waters and Bassler (2005)

 Signal-Receptor molecule pairs that modulate gene expression

Quorum-Sensing Systems

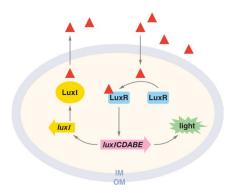


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- Once threshold density is reached, enough signal is received to upregulated target genes

Quorum-Sensing Systems

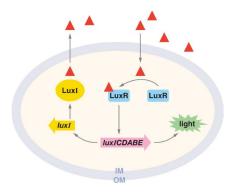


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How Quorum Sensing Interactions Affect Population Structure

- Signal-Receptor molecule pairs that modulate gene expression
- Once threshold density is reached, enough signal is received to upregulated target genes
- Can lead to biofilms, antibiotic production etc.

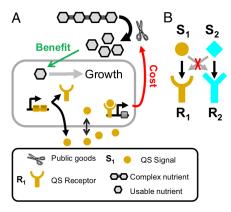


Figure 2: Eldar (2011)

When quorum is reached, bacteria produce a "public good"

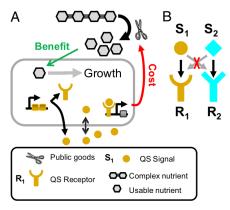


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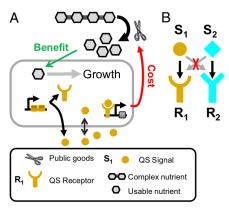


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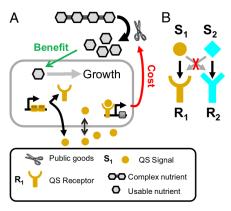


Figure 2: Eldar (2011)

- When quorum is reached, bacteria produce a "public good"
- Everyone benefits from this even if they don't contribute
- Must produce the receptor, signal molecule and good to contribute
- Cheaters DO prosper (if you are a bacterium)

Who Cares?

▶ check the discussion from Eldar (2011) for references

Maintaining Freeloaders as a Diversity Reservoir

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Kin Recognition for Strains

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Maintaining Freeloaders as a Diversity Reservoir

Kin Recognition for Strains

Designing Cheaters to Disrupt Pathogen Growth

Methods

Basic ODE Model

Social conflict drives the evolutionary divergence of quorum sensing

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Model Equations

$$\begin{split} \frac{dn_i}{dt} &= n_i (\frac{P_d}{P_d + 1} (1 - rf(R_i^{active})) - n_{tot} - \gamma_n) \\ \frac{dS_i}{dt} &= \beta_S (n_i - S_i) \\ \frac{dE}{dt} &= -\beta_E E + \sum_i f(R_i^{active}) n_i \\ \frac{dP_d}{dt} &= J_{P_d} + V_{max} E - \beta_{P_d} (\frac{P_d}{P_d + 1}) n_{tot} \end{split}$$

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- ▶ K_{ac} represents all receptors-signal pairs (R_iS_i) produced in each strain
- $ightharpoonup K_{ac}$ is of dimension $|R| \times |S| = |n| \times |n|$

Facultative Cheaters

Matrix for 2 strains R_1S_1 and R_2S_2

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

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Obligate Cheater

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Obligate Cheater

Matrix for 2 strains R_1S_1 and R_0S_0

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Custom Matrix

Matrix for 2 strains $R_1R_2S_1$ and R_2S_2

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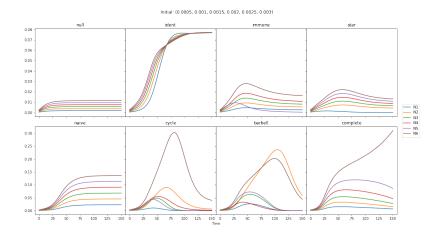
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- Simulate using gut microbiome data as initial state
- All code/results easily available to use on Github

Results

Comparing Different K_{ac} Matrices



How K_{ac} Sparsity Affects Population Structure

Simulating With Human Gut Microbiome Data

Discussion

Cheating works...

Cheating works...

(for bacteria)

Cheating works...

(for bacteria)

but cooperating is better!

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