University of Dundee

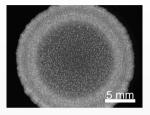
Competition and coexistence in bacterial range expansion: the role of founder cells and spatial heterogeneities Microbial communities: current approaches and open challenges

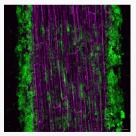
11/10/2022

Lukas Eigentler

Biofilms

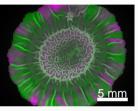
- Bacterial biofilms are surface-adhering multicellular collectives embedded in a self-produced extracellular matrix.
- Biofilms can have both beneficial and detrimental effects on the surrounding environment.
- Example: the soil-dwelling bacterium *Bacillus subtilis* forms biofilms on the roots of plants, where some strains promote the growth of plants.
- To fully realise their potential as biocontrol agents, strains need to be capable of coexisting with (or outcompeting) other biofilm-forming strains in the rhizosphere.

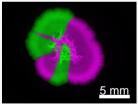




Competition within biofilms

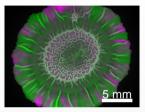
- Competition in biofilms is underpinned by kin discrimination.
- Many mechanisms of kin discrimination require spatial co-location of strains.
- Take a step back: need to understand the role of spatial structure first.

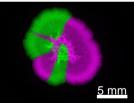




Competition within biofilms

- Spatial structure is best studied using isogenic strains: all other competitive mechanisms (e.g. kin discrimination) are excluded from the model system by design.
- Isogenic strains: Low founder densities promote spatial segregation and formation of spatial sectors.^{1,2}
- Questions: How does spatial structure arise and how does it affect competitive interactions?

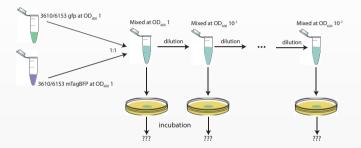




¹van Gestel, J. et al.: *ISME J.* 8.10 (2014) ²Martinez-Garcia, R. et al.: *PLoS Comput. Biol.* 14.4 (2018)

Methods

Experimental assay:



Tested for

- *B. subtilis* NCIB3610 ("standard" lab strain).
- B. subtilis NRS6153 (isolated from garden soil in Tayport, Fife, UK).

Methods

Mathematical model for isogenic strain pair:

$$\begin{aligned} \frac{\partial B_1}{\partial t} &= \nabla \cdot \left(\left(1 - (B_1 + B_2) \right) \nabla B_1 \right) + B_1 \left(1 - (B_1 + B_2) \right), \\ \frac{\partial B_2}{\partial t} &= \nabla \cdot \left(\left(1 - (B_1 + B_2) \right) \nabla B_2 \right) + B_2 \left(1 - (B_1 + B_2) \right). \end{aligned}$$

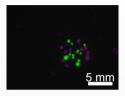
- Circular domain $\Omega = \{ x \in \mathbb{R}^2 : ||x|| \le R \}.$
- Logistic growth
- Diffusion with density-dependent diffusion coefficient, motivated by experimental observation that initially separated colonies abut rather than merge¹.

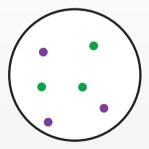


¹Matoz-Fernandez, D. et al.: *Soft Matter* 16.13 (2020)

Initial conditions

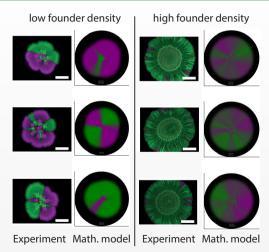
- What are appropriate initial conditions?
- In experiments, cells settle at random locations within the initial spot and grow to small micro-colonies.
- In the model, we position initial "cell patches" at random locations in the domain centre
 Ω₀ = {x ∈ ℝ² : ||x|| ≤ R₀} ⊂ Ω.
- Each model patch represents 1 microcolony ⇒ tool to modulate founder density.





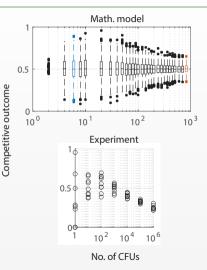
Variability in competitive outcome

- High founder density: no spatial structure and initial strain ratio consistently determines competitive outcome.
- Low founder density: spatial segregation occurs. Large variability in competitive outcome for fixed initial strain ratio.
- Founder density significantly affects phenotype and variability in competitive outcome.



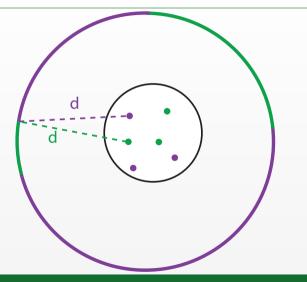
Variability in competitive outcome

- Founder density significantly affects phenotype and variability in competitive outcome.
- Variability increases with decreasing founder density.
- Note the computational power of the mathematical model: 1000 model simulations each vs 12 technical replicates each of experimental assay.



Disentangling variability

- Hypothesis: only initial patches that can drive the biofilm's radial expansion contribute to outcome density.
- We define a quantity that, based on the initial cell locations, measures a strain's "access to free space"



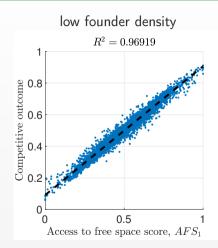
Disentangling variability

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t = 24ht = 48ht = 72h

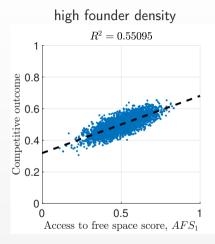
Access to free space predicts outcome

 Access to free space determines competitive outcome in the absence of any other competitive dynamics (isogenic strains).



Access to free space predicts outcome

- Access to free space determines competitive outcome in the absence of any other competitive dynamics (isogenic strains).
- Slope of relation between access to free space and competitive outcome depends on founder density.



• Do these results also hold if strains are non-isogenic and interact through local antagonisms?

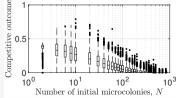
$$\frac{\partial B_1}{\partial t} = \nabla \cdot \left(\left(1 - \frac{B_1 + B_2}{k} \right) \nabla B_1 \right) + B_1 \left(1 - \frac{B_1 + B_2}{k} \right) - B_1 B_2,$$

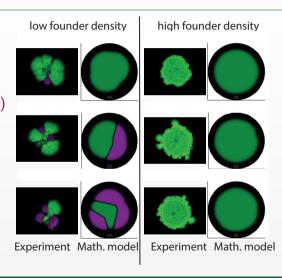
$$\frac{\partial B_2}{\partial t} = \nabla \cdot \left(d \left(1 - \frac{B_1 + B_2}{k} \right) \nabla B_2 \right) + r B_2 \left(1 - \frac{B_1 + B_2}{k} \right) - c B_2 B_1.$$

Nondimensional model, i.e. d, r, c are ratios of corresponding dimensional parameters.

Non-isogenic strains

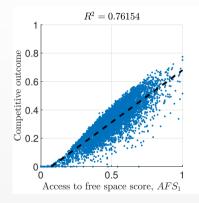
- High founder density: competitive exclusion.
- Low founder density: spatial segregation enables coexistence.
- Decreases in founder density cause (i) increased variability in competitive outcome, (ii) higher (on average) densities of weaker strain.





Access to free space predicts outcome

- Access to free space remains a reliable predictor of competitive outcome for low founder densities.
- Competition for space is the dominant mode of interaction for low founder densities.



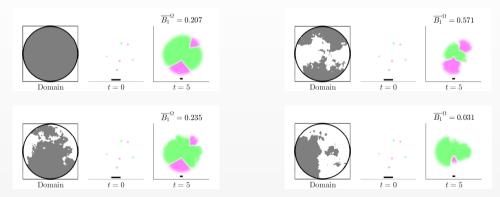
Q: What if environmental conditions are spatially heterogeneous?

$$\frac{\partial B_1}{\partial t} = \nabla \cdot \left(\mathbf{d_1}(\mathbf{x}) \left(1 - \frac{B_1 + B_2}{k} \right) \nabla B_1 \right) + \mathbf{r_1}(\mathbf{x}) B_1 \left(1 - \frac{B_1 + B_2}{k} \right) - \mathbf{c_{12}} B_1 B_2,$$

$$\frac{\partial B_2}{\partial t} = \nabla \cdot \left(\mathbf{d_2}(\mathbf{x}) \left(1 - \frac{B_1 + B_2}{k} \right) \nabla B_2 \right) + \mathbf{r_2}(\mathbf{x}) B_2 \left(1 - \frac{B_1 + B_2}{k} \right) - \mathbf{c_{21}} B_1 B_2.$$

Impact of spatial heterogeneity

Q: What if environmental conditions are spatially heterogeneous?



A: Spatial heterogeneity adds more variability \Rightarrow Access to free space score cannot make accurate predictions

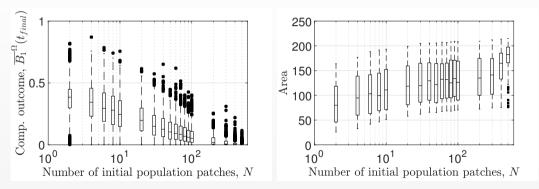
Lukas Eigentler (Dundee)

Impact of spatial heterogeneity

Q: Which source of variability (low founder density or spatial heterogeneous environmental conditions) dominates?

Impact of spatial heterogeneity

Q: Which source of variability (low founder density or spatial heterogeneous environmental conditions) dominates?



A: Low founder density determines variability in competitive outcome but spatial heterogeneities determine variability in footprint.

Lukas Eigentler (Dundee)

Q: Can we still make predictions of competitive outcome?

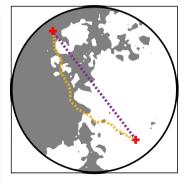
Q: Can we still make predictions of competitive outcome?A: Yes, using Voronoi tessellations based on an appropriate metric.







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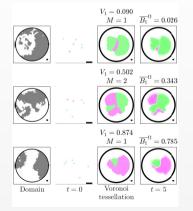


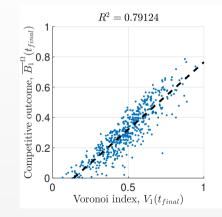






Q: Can we still make predictions for competitive outcome?A: Yes, using Voronoi tessellations based on an appropriate metric.





Conclusions

- Large variability in competitive outcome occurs for biofilms inoculated at low founder density, induced by the random positions of founder cells within the inoculum.
- Large variability in biofilm footprint occurs in spatially heterogeneous environments.
- Competitive outcome can be predicted based on founder cell locations and information on the spatial environment.
- Predictions hold true even if kin discrimination occurs ⇒ "Race for space" is more important than antagonistic actions at low founder densities.
- Impact on applications (e.g. use of *B. subtilis* as biofertilizer): Competitive success across all founder densities can only be guaranteed if a strain spreads fast and kills efficiently.

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- Natalie Bamford
- Alistair Bonsall
- Joana Moreira Carneiro
- Thibault Rosazza
- David Stevenson
- Tetyana Sukhodub

References

Slides are available on my website. http://lukaseigentler.github.io

- Eigentler, L., Davidson, F. A. and Stanley-Wall, N. R.: 'Mechanisms driving spatial distribution of residents in colony biofilms: an interdisciplinary perspective'. (Submitted).
- Eigentler, L., Kalamara, M., Ball, G., MacPhee, C. E., Stanley-Wall, N. R. and Davidson, F. A.: 'Founder cell configuration drives competitive outcome within colony biofilms'. *ISME J* 16.6 (2022), pp. 1512–1522.
- Eigentler, L., Stanley-Wall, N. R. and Davidson, F. A.: 'A theoretical framework for multi-species range expansion in spatially heterogeneous landscapes'. *Oikos* 2022.8 (2022), e09077.

Additional slides

Q: Can we still make predictions for competitive outcome?

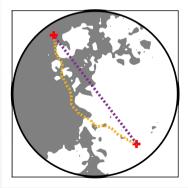
Q: Can we still make predictions for competitive outcome? A: Yes, using Voronoi tessellations based on an appropriate metric. Let $\mathcal{P}(x, y) := \{P = p([0, 1]) \subset \Omega, \text{ where } p : [0, 1] \rightarrow \Omega : p \in \mathcal{C}^1 \text{a.e.}, p(0) = x, p(1) = y\}$ be the set of all paths from x to y. For a given path $P \in \mathcal{P}(x, y)$, the time taken to move along the path is given by

$$I(P) := \int_P rac{1}{c(\mathbf{x})} \mathrm{d} s = \int_0^1 rac{1}{c(p(\tau))} \| p'(\tau) \| \mathrm{d} au,$$

where $0 \le c(x) < \infty$ represents the propagation speed along the path *P*. We define the *front propagation metric* from x to y as

$$t_{\mathsf{FP}}(\mathsf{x},\mathsf{y}) := \inf_{P \in \mathcal{P}(\mathsf{x},\mathsf{y})} I(P).$$

Q: Can we still make predictions of competitive outcome?A: Yes, using Voronoi tessellations based on an appropriate metric.



Q: Can we still make predictions for competitive outcome?A: Yes, using Voronoi tessellations based on an appropriate metric.Voronoi tessellation of whole domain:

$$\Delta_{\mathcal{B}_i}^{\Omega} := \left\{ \mathsf{x} \in \Omega : \min_{\mathsf{x}_i \in \mathcal{B}_i} t_{\mathsf{FP}}(\mathsf{x},\mathsf{x}_i) \leq \min_{\mathsf{x}_j \in \mathcal{B}_j} t_{\mathsf{FP}}(\mathsf{x},\mathsf{x}_j), i \neq j \right\},\$$

where $\mathcal{B}_i := \{x \in \{x_1, \dots x_N\} : B_i(x, 0) > 0\}, i = 1, 2.$ Restrict to area expected to be occupied by time *t*:

$$\Delta_{B_i}(t) := \left\{ \mathsf{x} \in \Delta^\Omega_{B_i} : \min_{\mathsf{y} \in \mathcal{B}_i} t_{\mathsf{FP}}(\mathsf{x},\mathsf{y}) \leq t
ight\}, \quad i=1,2$$

Voronoi index:

$$V_i(t) := rac{\operatorname{Area}\left(\Delta_{B_1}(t)
ight)}{\operatorname{Area}\left(\Delta_{B_1}(t)
ight) + \operatorname{Area}\left(\Delta_{B_2}(t)
ight)}, \quad i=1,2,$$





