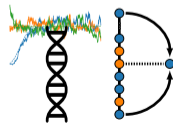


Within-host dynamics of antibiotic resistance

Pete Czippon

(in collaboration with Troy Day, Florence Débarre & François Blanquart)

BEVAS, Lausanne, April 2023



Antibiotic resistance

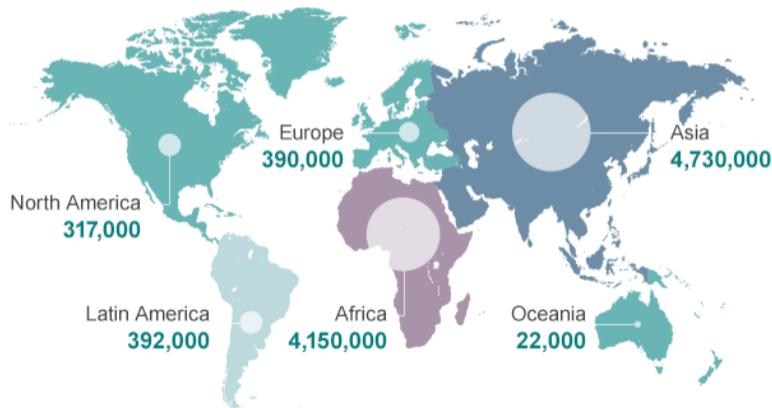


Antibiotic resistance – a public health problem

Currently: ~ 1 million deaths per year

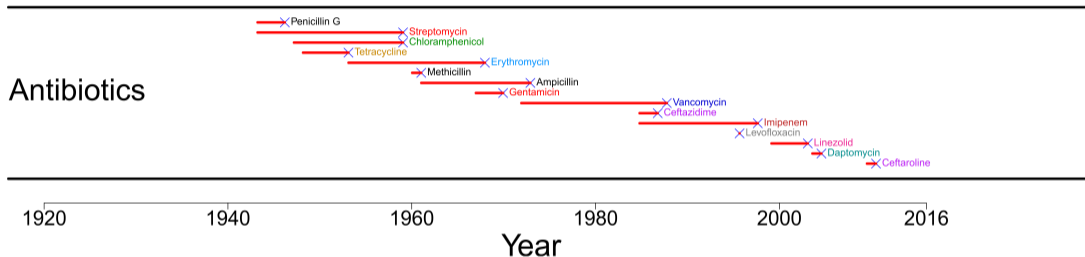
(Murray et al. (2022), Global burden of bacterial antimicrobial resistance in 2019: a systematic analysis, *The Lancet*)

Deaths attributable to antimicrobial resistance every year by 2050



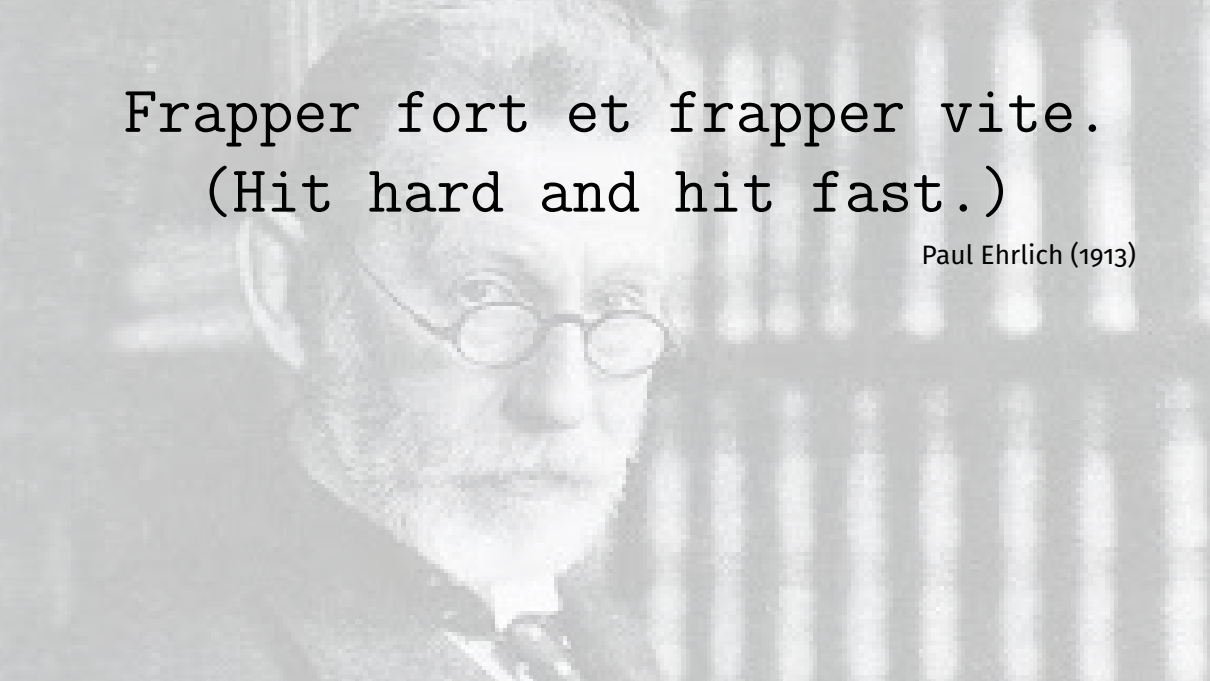
Source: Review on Antimicrobial Resistance 2014

Resistance always evolves



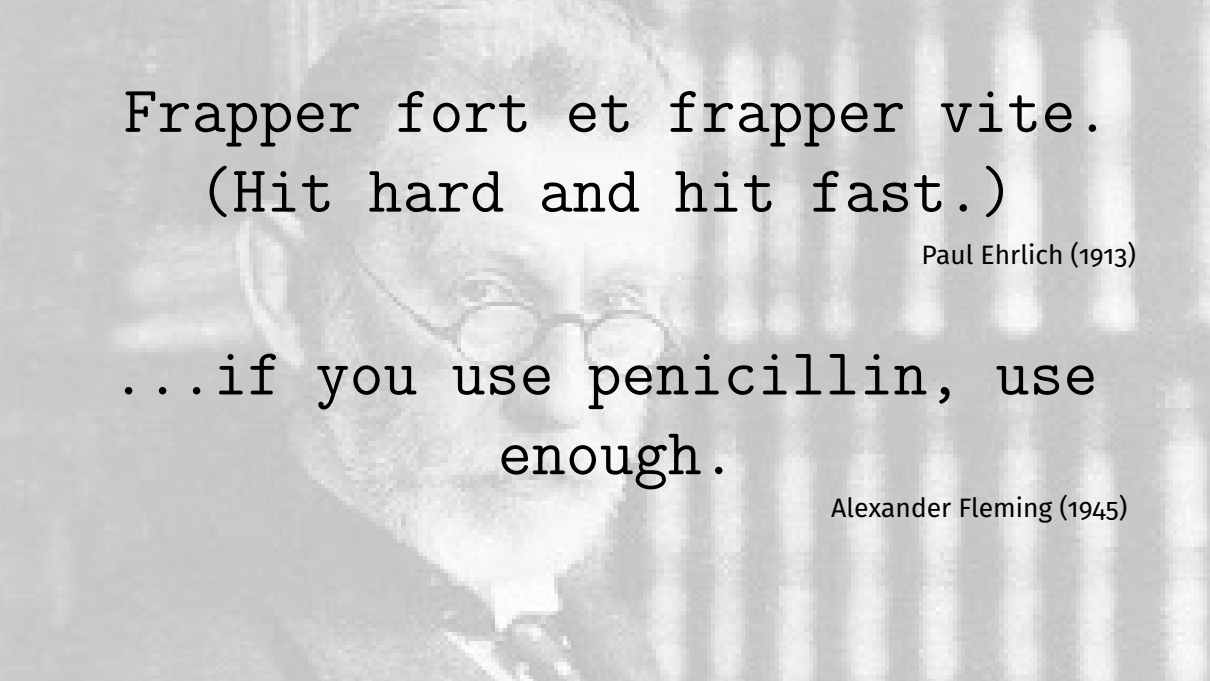
Kennedy et al. (2018), *PNAS*

Crosses indicate the appearance of a resistant strain



Frapper fort et frapper vite.
(Hit hard and hit fast.)

Paul Ehrlich (1913)

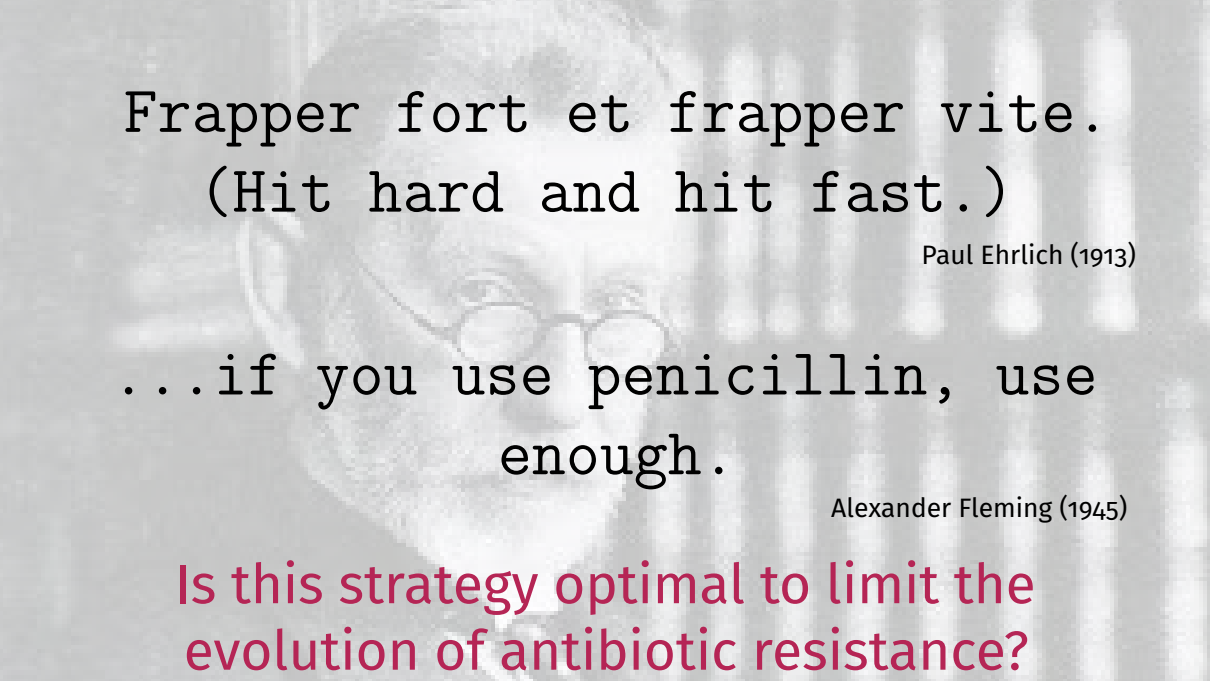


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...if you use penicillin, use
enough.

Alexander Fleming (1945)



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Is this strategy optimal to limit the
evolution of antibiotic resistance?

Arguments exist in favor and against this strategy in the context of resistance evolution

Ankomah & Levin (2014), PNAS

‘‘The results of this computer-assisted theoretical study support this century-old recommendation.’’

Arguments exist in favor and against this strategy in the context of resistance evolution

Ankomah & Levin (2014), PNAS

‘‘The results of this computer-assisted theoretical study support this century-old recommendation.’’

Day & Read (2016), PLoS CB

‘‘Theory does not support using the highest tolerable dose as a rule of thumb.’’

**Predicting the establishment
probability of a resistant subcolony
during treatment**

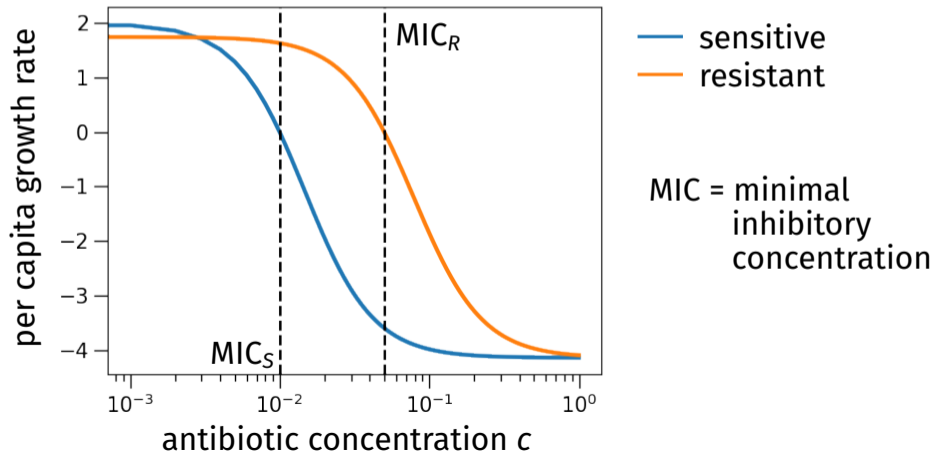
Modeling antibiotic treatment

deterministic growth rate = $\beta - \delta - \alpha(c)$

antibiotic response $\alpha(c)$ is modeled by a sigmoid (e.g. Regoes et al., 2004)

Modeling antibiotic treatment

$$\text{deterministic growth rate} = \beta - \delta - \alpha(c)$$



Modeling bacterial population dynamics

- ▶ per capita birth (β) and death rate (δ)
- ▶ density regulation of bacteria (through competition, γ)
 - ▶ reducing birth rate (resource competition)
 - ▶ increasing death rate (toxin production)

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Example: birth competition and biocidal treatment

($x(t)$ = bacterial concentration at time t)

birth rate:

$$\lambda = \max(0, \beta - \gamma x(t))$$

death rate:

$$\mu = \delta + \alpha(c)$$

Deterministic population dynamics

birth competition + biocidal treatment

$$x_i(t) = \lambda(t) - \mu(t) = x_i(t) (\max(0, \beta_i - \gamma(x_S(t) + x_R(t))) - \delta_i - \alpha_i(c))$$

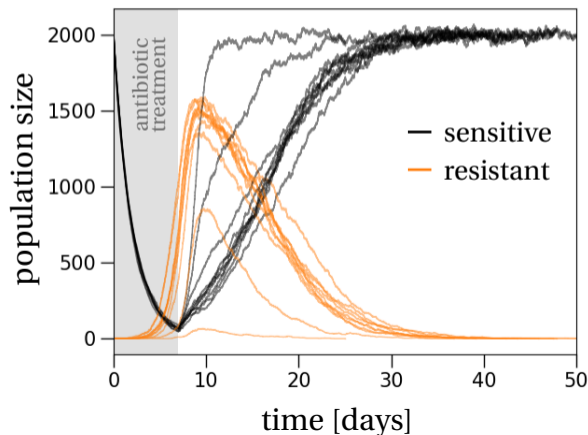
Assumptions

- ▶ **resistance:** $\alpha_R(c) \leq \alpha_S(c)$ for all concentrations $c \geq 0$
- ▶ **cost of resistance:** $\beta_S - \delta_S > \beta_R - \delta_R$

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Survival probability until the end of treatment

...continuous-time branching process theory (Kendall, 1948; Uecker & Hermisson, 2011)

Survival probability $\varphi(\tau)$ for a treatment of duration τ when started with a single resistant cell is given by

$$\varphi(\tau) = \frac{1}{1 + \int_0^\tau \mu_R(t) \exp \left(\int_0^t (\lambda_R(t') - \mu_R(t')) dt' \right) dt}$$

Survival probability until the end of treatment

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Example: birth regulation and biocidal treatment

Abbreviations

- ▶ par capita maximal growth rate of strain k : $\rho_k = \beta_k - \delta_k - \alpha_k(c)$
- ▶ selection coefficient: $s = \rho_R - \rho_S > 0$!

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Survival probability ($\tau = \infty$)

$$\varphi = \frac{1}{1 + \frac{\delta_R + \alpha_R}{\rho_R} \left(\frac{x_S(0)\gamma}{s} + 1 \right)}$$

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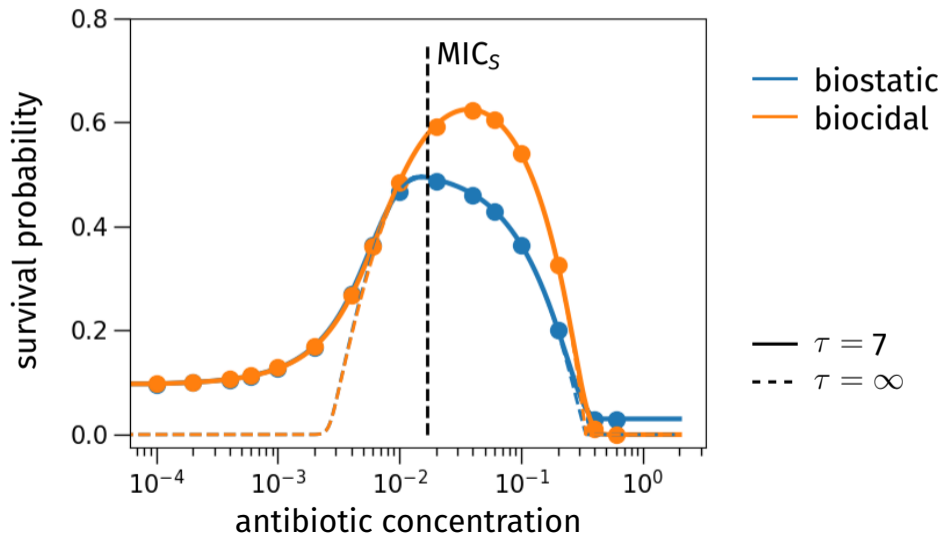
Analogous expressions can be obtained for death competition + biostatic/biocidal treatment

birth competition + biostatic treatment is more complicated

Survival probability

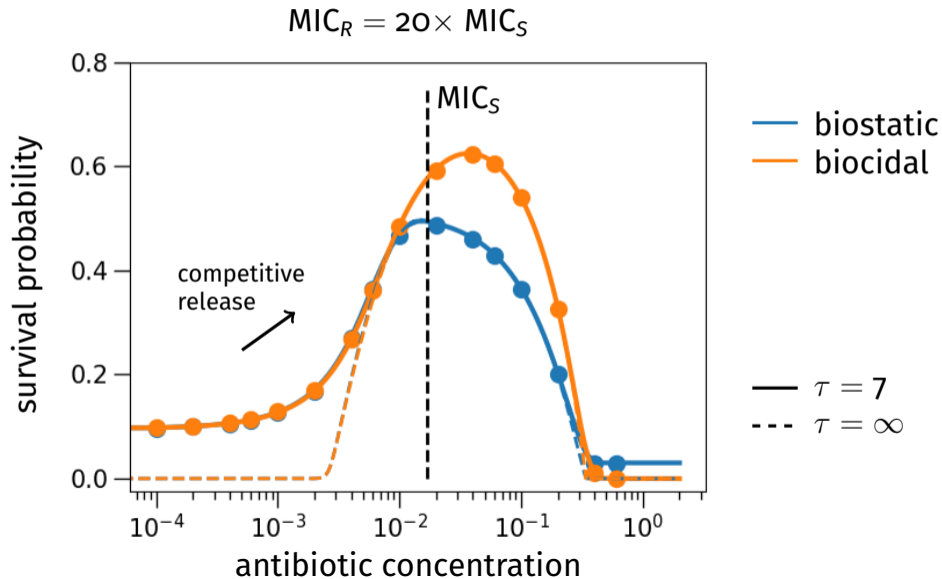
birth competition

$$\text{MIC}_R = 20 \times \text{MIC}_S$$



Survival probability

birth competition



**Predicting the location of the maximal
risk of resistance establishment**

Maximizing the survival probability

$$\varphi(c) = \frac{1}{1+f(c)} \quad \Rightarrow \quad \varphi'(c) = -\frac{f'(c)}{(1+f(c))^2}$$

And hence

$$\varphi'(c) = 0 \quad \Leftrightarrow \quad f'(c) = 0$$

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Example: birth competition + biocidal treatment

$$\frac{\alpha'_S(c)}{\alpha'_R(c)} = 1 + \frac{\beta_R s(c)}{\rho_R(c)(\delta_R + \alpha_R(c))} + \underbrace{\frac{\beta_R s(c)^2}{x_S(0)\gamma\rho_R(c)(\delta_R + \alpha_R(c))}}_{\approx 0 \text{ for weak selection}}$$

Maximizing the survival probability

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...not possible to resolve

But: we can derive a condition for a critical concentration \tilde{c} below which the maximizing concentration has to be!

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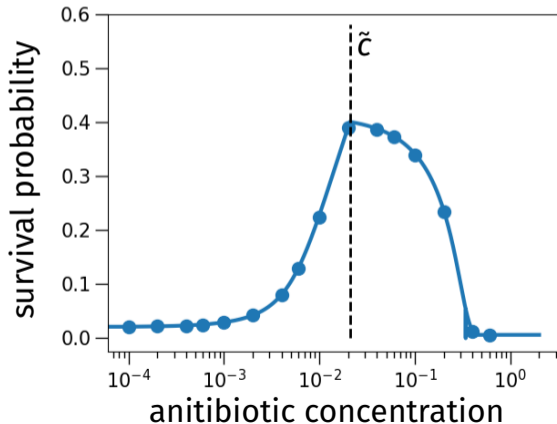
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Maximizing the survival probability

death competition + biostatic treatment

$$\tilde{c} = \inf\{c : \beta_S - \alpha_S(c) = 0\}$$

This threshold is independent of the resistant type!



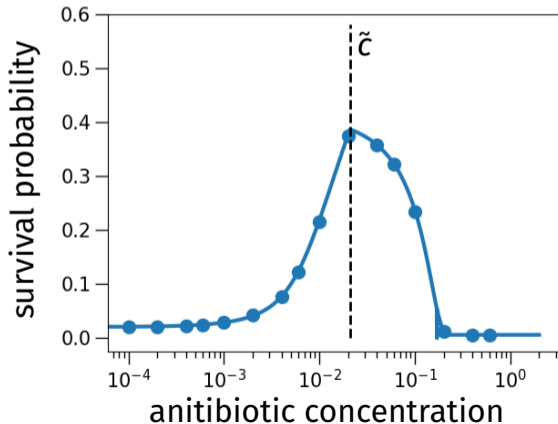
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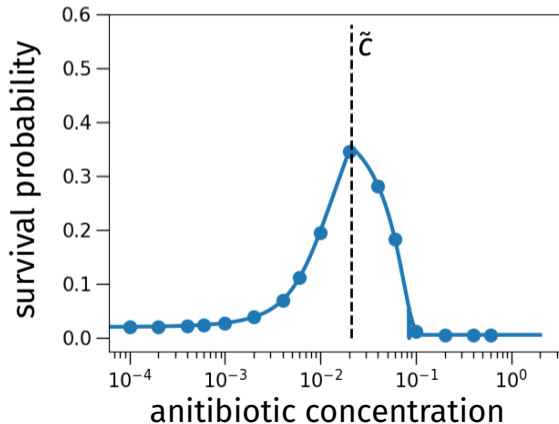
$$MIC_R = 10 \times MIC_S$$

Maximizing the survival probability

death competition + biostatic treatment

$$\tilde{c} = \inf\{c : \beta_S - \alpha_S(c) = 0\}$$

This threshold is independent of the resistant type!

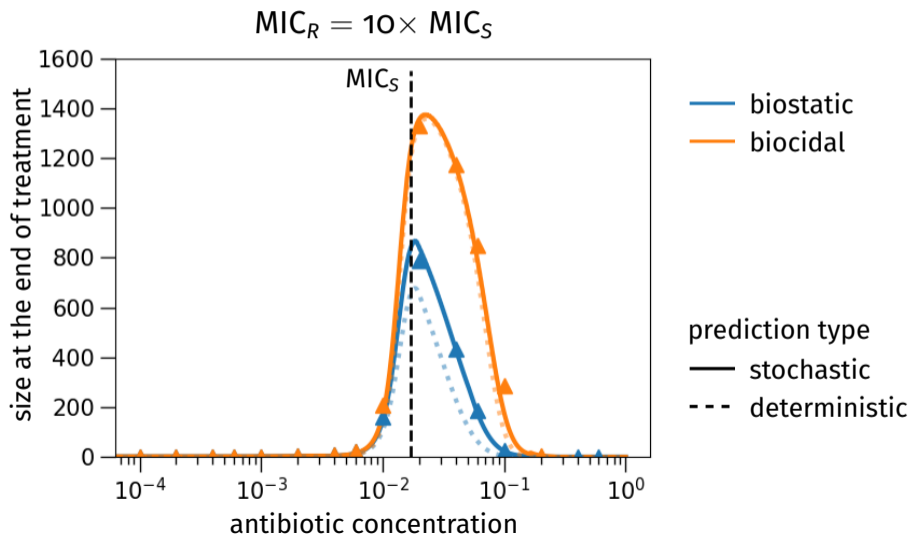


$$\text{MIC}_R = 5 \times \text{MIC}_S$$

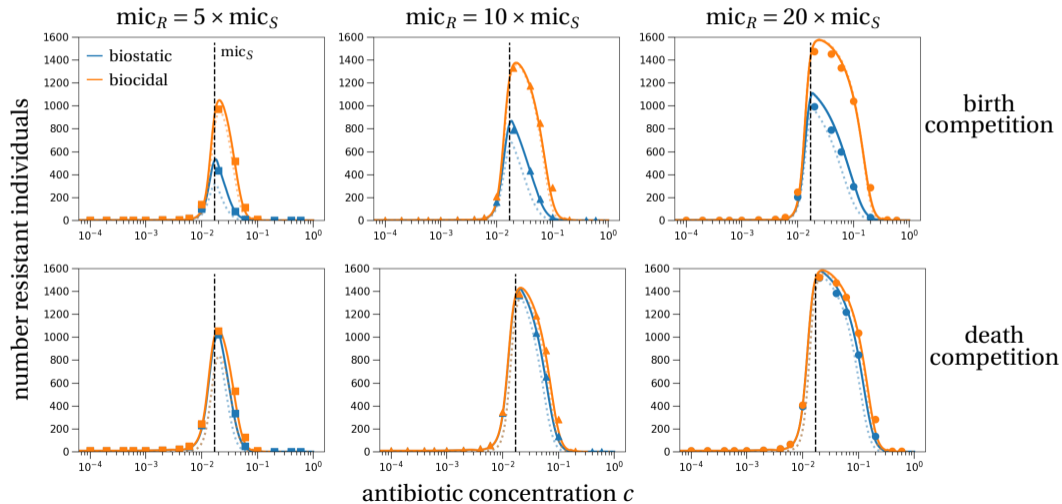
survival \neq establishment
Does survival matter?

The resistant subpopulation size at the end of treatment

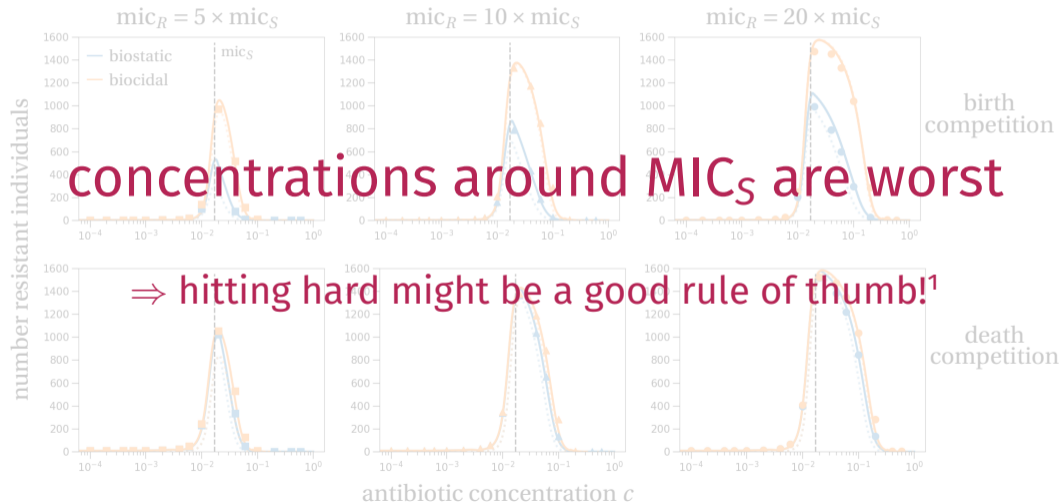
birth competition



The resistant subpopulation size at the end of treatment



The resistant subpopulation size at the end of treatment



¹my personal conclusion, which is not shared by all of my coauthors

Summary & Conclusion

- ▶ **continuous-time branching processes** allow us to compute the survival probability explicitly for several scenarios of (self-limited) bacterial population dynamics

Summary & Conclusion

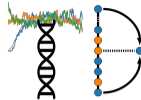
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- ▶ ...and further helps to derive conditions for the **resistance-survival-maximizing antibiotic concentration**

Summary & Conclusion

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- ▶ survival is not necessarily clinically relevant
→ size of the resistant subpopulation

Summary & Conclusion

- ▶ **continuous-time branching processes** allow us to compute the survival probability explicitly for several scenarios of (self-limited) bacterial population dynamics
- ▶ ...and further helps to derive conditions for the **resistance-survival-maximizing antibiotic concentration**
- ▶ survival is not necessarily clinically relevant
→ size of the resistant subpopulation
- ▶ the **resistant subpopulation size is maximized typically at the MIC_S** (in our model)
we ran a lot of additional scenarios (immune response, different antibiotic response curves, different population dynamical models) and the result is robust (in all of our considered parameter ranges ...but we don't have analytical solutions in those cases)



Thank you for your attention!

Collaborators



François
Blanquart
(Collège de France)



Troy
Day
(Queen's Uni)



Florence
Débarre
(Sorbonne Uni)

Preprint



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