





# MECHANISTIC MODELLING OF MICROBIAL COMMUNITIES WITH INSIGHTS FROM AN IN VITRO PIG GUT MODEL

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### BACKGROUND

The gut microbiota play a key role in the health of animals and humans. However, the dynamic properties and stability of the microbiota are poorly understood. We propose a regression technique for parameter inference of a mechanistic model to describe the temporal dynamics of these microbial communities. The model could be used for measuring community resilience against external perturbing factors, such as antibiotic therapy.

### METHODS

For model validation, we generated our own *in silico* time-series data of microbial abundances, based on an agentbased method that recreates the essential characteristics of experimental data [1]. To model these data, we applied a set of stochastic differential equations to realistically mimic microbial growth, mortality, and species synergy. An external signal was included to represent the impact of perturbations such as periodic circadian rhythms or administration of antibiotics. The model parameters were inferred using Bayesian linear regression.

## The generative model (Lotka-Volterra):

$$\frac{\mathrm{d}}{\mathrm{d}t}x_i(t) = x_i(t)\left(\alpha_i + \sum_j \beta_{ij}x_j(t) + \gamma_i u(t) + \sigma \dot{W}(t)\right)$$

This dynamical equation specifies the change in the abundance,  $x_i$ , of the *i*-th simulated taxon at each time point, *t*.

Specifically, we modelled five microbial taxa: each of which grows with rate  $\alpha_i$  and interacts with other taxa according to an interaction matrix  $\boldsymbol{\beta}$ . The interaction components  $\beta_{ij}$  refer to the impact of taxa *j* on *i*, which can be negative (parasitism), zero, or positive (symbiosis). The self-interaction ( $\beta_{ii} < 0$ ) represents the mortality of taxon *i*. The susceptibility term  $\gamma_i$  refers to the impact of a perturbation signal, *u*, on the taxon, and *W* is a stochastic process with noise amplitude  $\sigma$ .



#### We successfully produced time-series data from preliminary simulations that are qualitatively similar to those measured from experiments.

- We implemented an external perturbation signal, modelled as a periodic step function (Fig. 1).
- Fig 2. demonstrates the strong potential of our regression technique for parameter estimation on real data.
- The entire set of 'self-interaction' parameters are inferred as negative: indicating a reliable fit to these data.



## **CONCLUSION AND PERSPECTIVES**

Mathematically modelling microbial dynamics is valuable for predicting the long-term stability of the gut microbiota in response to a wide variety of perturbing factors. Future work will explore the application of this probabilistic regression technique to data that will be obtained from an *in vitro* pig gut model that is subject to perturbation.

### **Reference:**

[1] Descheemaeker, L., & De Buyl, S. (2020). Stochastic logistic models reproduce experimental time series of microbial communities. Elife, 9, e55650. doi: https://doi.org/10.7554/eLife.55650

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