## Are there limits to inference in the simplest ecological models?

Some thoughts and speculations

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## After many discussions with probabilists and statisticians

- Vincent Bansaye (CMAP, École Polytechnique)
- Éliza Vergu (INRA Jouy-en-Josas)
- Jean-René Chazottes (CPHT, École Polytechnique)
- Maud Delattre (AgroParisTech)
- Chi Tran (Université Gustave Eiffel)
- Nicolas Champagnat (Institut E. Cartan, INRIA)
- Peter Czuppon (University of Muenster)
- Sylvie Méléard (CMAP, École Polytechnique)
- Pierre Collet (CPHT, École Polytechnique)
- ...


## PATTERNS

## DATA

## PROCESSES

Information about the world from data?

## Examples: Genetic diversity and demography reconstruction

African genetic diversity and adaptation inform a precision medicine agenda
Luisa Pereira ${ }^{1,2}$, Leon Mutesa $)^{3}$. Paulina Tindana ${ }^{4}$ and Michèle Ramsay $9^{5 \times 4}$


## Examples: Species diversity and processes

## Barro Colorado Island

## A test of the unified neutral theory of biodiversity

Brian J. McGill


## Limits to inference

- Models much simpler than natural populations
- Models very complex (many parameters)
- Several processes can give similar patterns
- No clues from the past (most often) (no idea of initial conditions)
- No independent data to validate inference (except from computer simulations but ... circularity issues)
- Big data but little information (autocorrelation, non-independence)


## What can we infer from the simplest population system?

- bacteria (clonal, relatively simple)
- individual count (exact dynamics data)
- homogeneous environment
- few individuals (no competition)
- small timescale (no mutation, no trait evolution)
- controlled experiment (repeatability)
- Direct observations are possible
$\rightarrow$ A simple birth-death model.


# Antibiotic-induced population fluctuations and stochastic clearance of bacteria 

Jessica Coates ${ }^{1 \dagger}$, Bo Ryoung Park ${ }^{2 \dagger}$, Dai Le ${ }^{2}$, Emrah Şimşek ${ }^{2}$, Waqas Chaudhry ${ }^{2}$, Minsu Kim ${ }^{1,2,3 *}$


## Coates et al. 2018's Experiment

## Protocol

- Inoculation of colonies on plates with a single cell
- Picture every $\sim 10$ mins
- Rich medium
- Automatized cell count
- Three types of medium:

1. Antibiotics-free (control)
2. Bacteriostatic (slows down or stops bacteria growth and division)
3. Bactericidal (kills bacteria)

## What we want to know:

Can we recover that antibiotics differently affect

- the individual birth rate
- the individual death rate


## Movies



## Data: population dynamics



## Inference from a pure birth model (Yule's process)

The model

- $X_{t}$ : the number of cells at time $t$
- $b$ : cell birth rate
- $\Delta t$ : the time frame between observations

Distribution: Negative binomial

$$
\begin{aligned}
& \mathbb{P}\left(X_{t+\Delta t}-X_{t}=k \mid X_{t}=n\right)=\binom{n+k-1}{n-1} p^{n}(1-p)^{k} \\
& \text { with } p=e^{-b \Delta t}
\end{aligned}
$$

Method: Exact Likelihood

$$
\mathcal{L}(b)=\sum_{i} \log \mathbb{P}\left(X_{t_{i}}-X_{t_{i-1}}=k_{i} \mid X_{t_{i-1}}=k_{i-1}, b\right)
$$

with

- $k_{i}$ the number of cells in observation $i$
- $t_{i}$ the time of observation $i$


## Inference from a pure birth model (Yule's process)

## Results

- Control (no antibiotics): $\widehat{b}=0.023$ (AIC: 801.8, BIC: 804.7)
- Chloramphenicol (bacteriostatic): $\widehat{b}=0.014$ (AIC: 1082.8, BIC: 1085.7)
- Cefsulodin (bactericidal, supercritical): $\widehat{b}=$ ?
(Likelihood not defined)
- Cefsulodin (bactericidal, subcritical): $\widehat{b}=$ ? (Likelihood not defined)

Direct observations
Cell division rate: $0.025 \mathrm{~min} .^{-1}$ (Minsu Kim, pers. comm.)

## Inference from a linear birth-death model

## The model

- $X_{t}$ : the number of cells at time $t$
- $b$ : cell birth rate
- d: cell death rate
- $\gamma=b+d$ : an allometry parameter scaling demographic rates
- $\Delta t$ : the time frame between observations
- Assumption: $b$ and $d$, and the population size, scale with $K \rightarrow \infty$

SDE Approximation (e.g. Bansaye \& Méléard 2015
$d X_{t}=(b-d) X_{t} d t+\sqrt{2 \gamma X_{t}} d B_{t}$

## Inference from a linear birth-death model

Discretization with an Euler's scheme
$X_{t+\Delta t}-X_{t}=(b-d) X_{t} \Delta t+\sqrt{2(b+d) X_{t}}\left(B_{t+\Delta t}-B_{t}\right)$
Method: Likelihood aproximation in an Euler's Scheme
$\mathcal{L}(b, d)=-\frac{1}{2}\left(\sum_{i} \frac{\left(X_{t_{i}}-X_{t_{i-1}}-(b-d) X_{t_{i-1}} \Delta t_{i}\right)^{2}}{2(b+d) X_{t_{i-1}} \Delta t_{i}}\right)+\sum_{i} \log \left(4 \pi(b+d) X_{t_{i}} \Delta t_{i}\right)$
Algorithm

- R package MsdeParEst from Delattre et al 2016
- Mixed effect in the drift part
- Fixed effect in the diffusion part


## Results

- Control (no antibiotics): $\widehat{b}=0.039$ and $\widehat{d}=0.016$ (AIC: 864.9, BIC: 866.1)
- Chloramphenicol (bacteriostatic): $\widehat{b}=0.016$ and $\widehat{d}=0.002$ (AIC: 902.7, BIC: 902.6)
- Cefsulodin (bactericidal, supercritical): $\widehat{b}=0.023$ and $\widehat{d}=0.017$ (AIC: 297.3, BIC: 299.9)
- Cefsulodin (bactericidal, subcritical): $\widehat{b}=0.007$ and $\widehat{d}=0.008$ (AIC: 170.9, BIC: 173.5)


## Issues from the models

| Model | Medium | Rates estimate | Growth rate |
| :---: | :---: | :---: | :---: |
| Pure Birth | Control | $\widehat{b}=0.023$ | 0.023 |
| (AIC: 801) | Bacteriostatic | $\widehat{b}=0.014$ | 0.014 |
| (AIC: 1083) | Bactericidal (supercritical) | - | - |
| - | Bactericidal (subcritical) | - | - |
| Birth-Death | Control | $\widehat{b}=0.039, \widehat{d}=0.016$ | 0.023 |
| (AIC: 865) | Bacteriostatic | $\widehat{b}=0.016, \widehat{d}=0.002$ | 0.014 |
| (AIC: 903) | Bactericidal (supercritical) | $\widehat{b}=0.023, \widehat{d}=0.017$ | 0.006 |
| (AIC: 297) | $\widehat{b}=0.007, \widehat{d}=0.008$ | -0.001 |  |
| (AIC: 171) | Bactericidal (subcritical) |  |  |

## Issues from the data




- No extinction before first observation
$\rightarrow$ Observations conditional on
survival
$\rightarrow$ Observation bias


## Issues from the interpretation

- Population clearly spatialized
- Competition between individuals
- Medium certainly not homogeneous, not constant
- Cells growth before division: multiscale models?
$\rightarrow$ interpretation of $\widehat{b}$ and $\widehat{d}$ ?


## Probability of extinction: unexplained

- Pure birth model: $P\left(\right.$ extinction $\left.\mid X_{0}=1\right)=0$
- $($ supercritical $)$ Birth-Death model : $P\left(\right.$ extinction $\left.\mid X_{0}=1\right)=\frac{d}{b}$



## Perspectives

- Even simpler biological systems?
- Include conditional on survival?
- Include observations protocol and apparels into models?
- More informative statistics?
- (a priori) Choice of the scale of observation? (why not focusing on a single cell?)
- (a priori) Choice of the relevant processes? (why not bacteria movement?)
- Time is exponential in models, can we test this?
- Inherent and unsurpassable limits?


## The remaining question

Most relevant model to estimate $b$ and $d$ on such data?

## Acknowledgements

Minsu Kim (for sharing data)
Maud Delattre (for her help with parameter estimations from the SDE)
The Organizers


