

# SIMPLIFYING COMPLEXITY

---

## LARGE COMMUNITIES AND RANDOM INTERACTIONS

---

Matthieu Barbier

(Plant Health Institute Montpellier, CIRAD  
& Institut Natura e Teoria en Pirenèus)

20/03/2025

Before everything – background for why I am giving you this lecture, in 2 questions and 1 graph!

**The questions:** when I analyze data, do I believe

**The questions:** when I analyze data, do I believe

- I can neatly separate what I measure into causes and consequences (independent & dependent variables), or not?

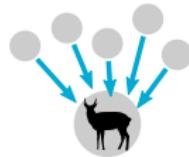
**The questions:** when I analyze data, do I believe

- I can neatly separate what I measure into causes and consequences (independent & dependent variables), or not?
- I can only focus on a few things that matter, or not?

**The questions:** when I analyze data, do I believe

- I can neatly separate what I measure into causes and consequences (independent & dependent variables), or not?
- I can only focus on a few things that matter, or not?

Many things  
matter



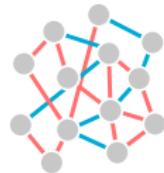
Few things  
matter




One-way  
causality



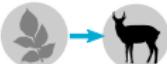
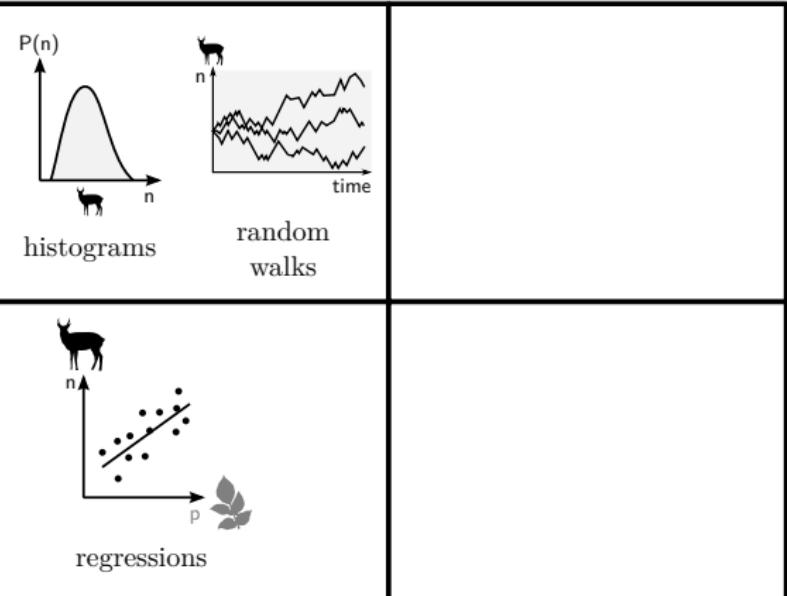
Feedbacks



forget who does what,  
count what happens

## Many things matter

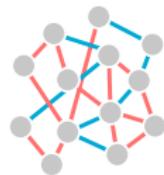
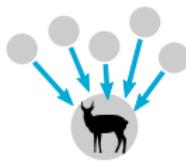
PCA,  
network  
metrics,  
...



## One-way causality



## Feedbacks



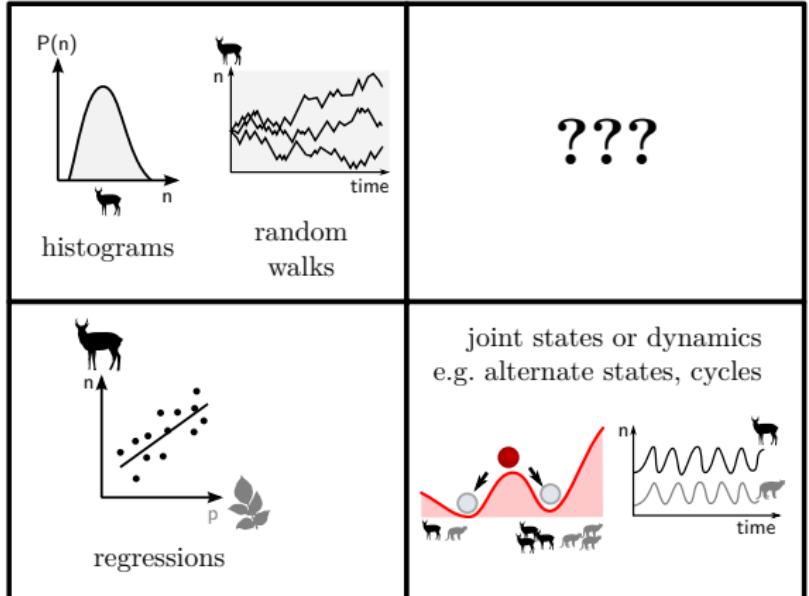
forget who does what,  
count what happens

## Many things matter

PCA,  
network  
metrics,  
...

## Few things matter

identify who does what



## One-way causality

more stats  $\longleftrightarrow$  more dyn



## Feedbacks

## Points of this lecture:

- We always think that simple system = small system  
(e.g. only study one species in isolation, or a pair)

## Points of this lecture:

- We always think that simple system = small system  
(e.g. only study one species in isolation, or a pair)
- In fact, it is possible to be large (many-species...) and still simple

## Points of this lecture:

- We always think that simple system = small system  
(e.g. only study one species in isolation, or a pair)
- In fact, it is possible to be large (many-species...) and still simple
- Complexity with simple consequences is (or can be modelled by)  
“randomness”

## Points of this lecture:

- We always think that simple system = small system  
(e.g. only study one species in isolation, or a pair)
- In fact, it is possible to be large (many-species...) and still simple
- Complexity with simple consequences is (or can be modelled by)  
“randomness”
- Small and large simplicity are both wrong, but both are valid starting points, and they can be combined to model reality

## Points of this lecture:

- We always think that simple system = small system  
(e.g. only study one species in isolation, or a pair)
- In fact, it is possible to be large (many-species...) and still simple
- Complexity with simple consequences is (or can be modelled by)  
“randomness”
- Small and large simplicity are both wrong, but both are valid starting points, and they can be combined to model reality

**Note:** Hereafter I will be talking about community ecology & species interactions, but many of the ideas apply for complex systems of genes, behaviors...

## PREAMBLE: AN ILLUSTRATION IN SIMULATED DATA

# MODEL

Assume we can use a model roughly like Lotka-Volterra with  $S$  species

$$\frac{dN_i}{dt} = r_i N_i \left(1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j\right) \quad (1)$$

# MODEL

Assume we can use a model roughly like Lotka-Volterra with  $S$  species

$$\frac{dN_i}{dt} = r_i N_i \left(1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j\right) \quad (1)$$

we need many parameters:

# MODEL

Assume we can use a model roughly like Lotka-Volterra with  $S$  species

$$\frac{dN_i}{dt} = r_i N_i \left(1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j\right) \quad (1)$$

we need many parameters:

- growth rates  $r_i$  ( $S$  numbers, 1 per species)

$$r = (? , ? , ? \dots) \quad (2)$$

# MODEL

Assume we can use a model roughly like Lotka-Volterra with  $S$  species

$$\frac{dN_i}{dt} = r_i N_i \left(1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j\right) \quad (1)$$

we need many parameters:

- growth rates  $r_i$  ( $S$  numbers, 1 per species)

$$r = (? , ? , ? \dots) \quad (2)$$

- even harder, interactions  $a_{ij}$  ( $S^2$  numbers,  $S$  per species)

$$a = \begin{pmatrix} ? & ? & \dots \\ ? & & \\ \dots & & \end{pmatrix} \quad (3)$$

# MODEL

Practical modeller's strategy: don't know  $r_i$  and  $a_{ij}$  in your system? take them at random!

$$a = \begin{pmatrix} 0.29 & 0.54 & 0.53 & 0.02 & \dots \\ 0.57 & 0.86 & 0.90 & 0.81 & \dots \\ \dots & & & & \end{pmatrix} \quad (4)$$

# MODEL

Practical modeller's strategy: don't know  $r_i$  and  $a_{ij}$  in your system? take them at random!

$$a = \begin{pmatrix} 0.29 & 0.54 & 0.53 & 0.02 & \dots \\ 0.57 & 0.86 & 0.90 & 0.81 & \dots \\ \dots & & & & \end{pmatrix} \quad (4)$$

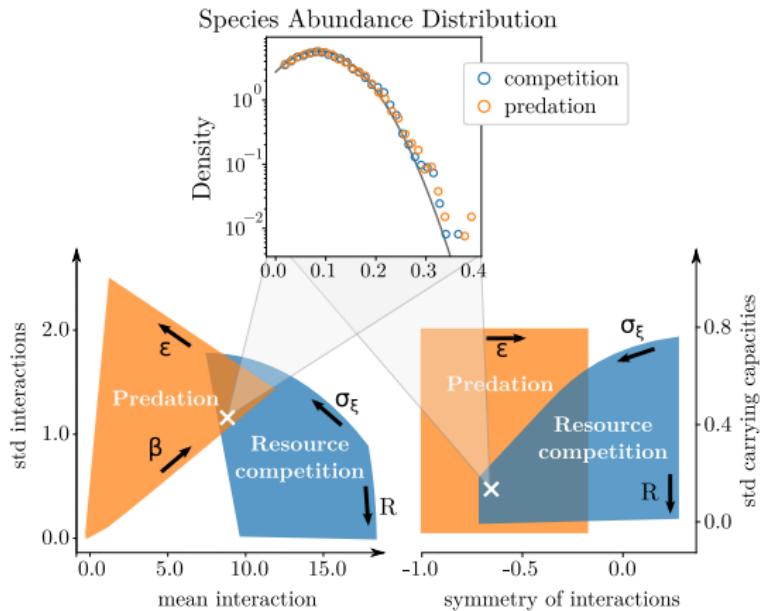
**Problem:** results could depend on many details of the matrix, e.g. each value, or how we drew the random numbers (normal, uniform, etc.)

# PREDICTIONS

In fact, if you simulate again and again, many results only depend on a few global statistics of the community:

- mean of interactions  $\langle a_{ij} \rangle$  ( $= E[a_{ij}]$  if you prefer)
- standard deviation  $\text{std}(a_{ij})$
- and symmetry  $\text{corr}(a_{ij}, a_{ji})$

# PREDICTIONS



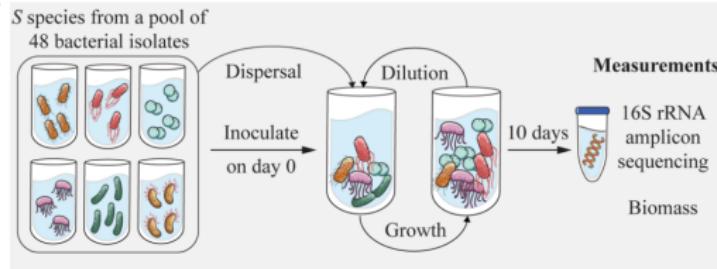
In particular, nature of interactions (competitive, trophic, parasitism...) is *irrelevant!*

e.g. two models, one with predation, one with competition, give same results (abundance distribution, etc.) if they correspond to the same mean, variance and symmetry of  $a_{ij}$

# EMPIRICAL TEST

## Experimental setup: soil bacteria competition

(Hu et al, Science 2022)



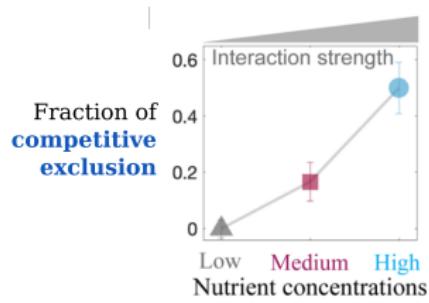
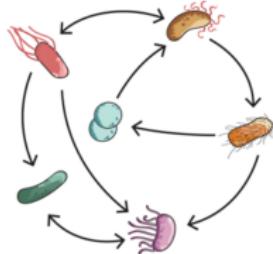
Jiliang Hu



Jeff Gore

Very well controlled environment & replicable results

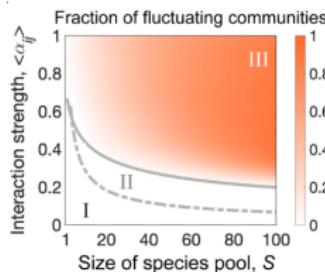
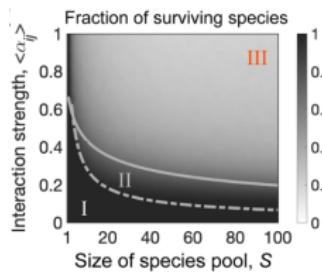
Unique feature: ability to **control how strongly species interact**



# EMPIRICAL TEST

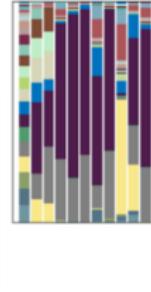
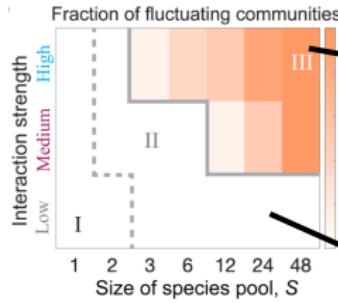
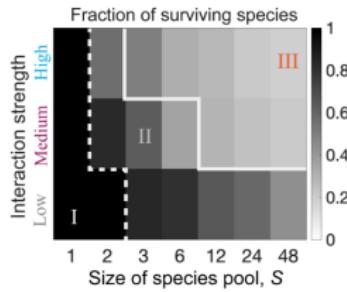
white = unique equilibrium, orange = fluctuations (in theory, chaos)

## Random Lotka-Volterra Theory



— Survival boundary  
— Stability boundary  
Phase I: stable full coexistence  
Phase II: stable partial coexistence  
Phase III: persistent fluctuation

## Microbial experiments



Now that I gave this illustration, two questions:

- how is that possible?
- how does this help us understand ecology?

## I. INTRODUCTION: COMPLEXITY AND SIMPLICITY

# WHERE TO PUT COMPLEXITY

Basic question of modelling: which details are important to include?

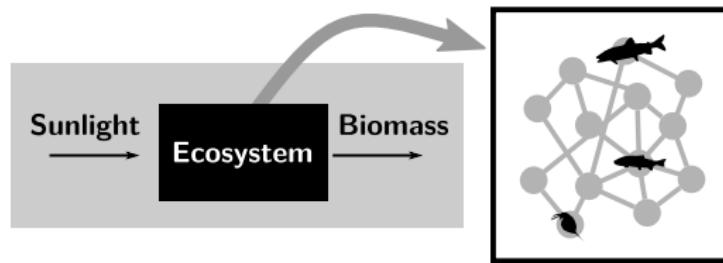


# WHERE TO PUT COMPLEXITY

Basic question of modelling: which details are important to include?



- whenever we write a simple model in biology, we are hiding complexity

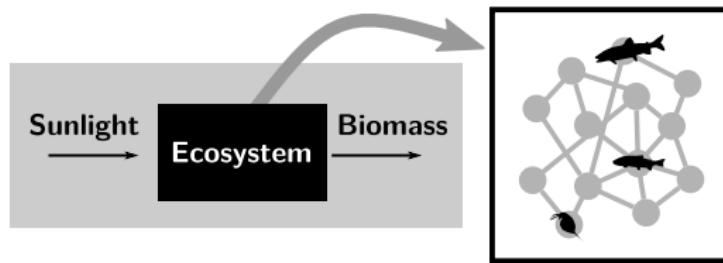


# WHERE TO PUT COMPLEXITY

Basic question of modelling: which details are important to include?



- whenever we write a simple model in biology, we are hiding complexity



- is there a *principled* way of understanding when this is a valid choice?

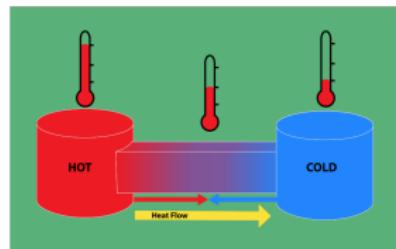
# IDEA ORIGINATING FROM PHYSICS

When a system has *many* variables, a much simpler description is often possible

$10^{23}$  variables:  
position of every molecule



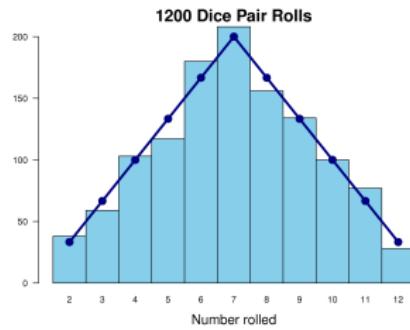
2 variables: temperature & pressure



uncountable factors, chaotic motion



1 probability distribution

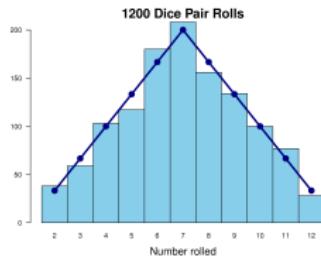


# MEANING OF RANDOMNESS

uncountable factors, chaotic motion



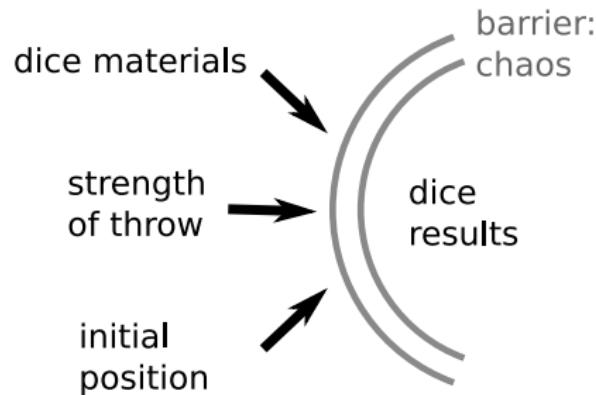
1 probability distribution



dice are simple *because* they are extremely sensitive to many details, making their movement chaotic and impossible to control

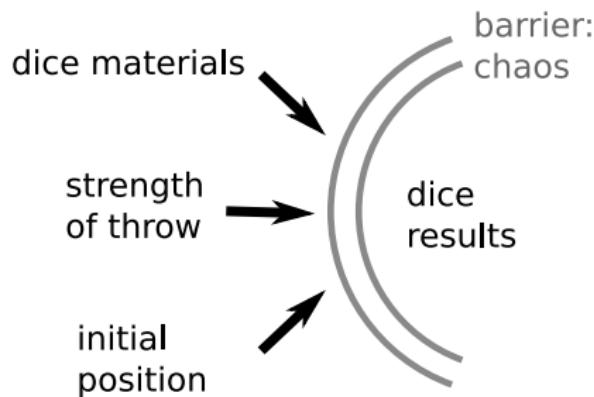
(vs e.g. basketball throws, where you need to know who is throwing, how far...)

# MEANING OF RANDOMNESS



- “barrier” against details = chaos, motion unpredictable even if you know almost all details

# MEANING OF RANDOMNESS



- “barrier” against details = chaos, motion unpredictable even if you know almost all details
- result = randomness, unpredictability becomes simplicity

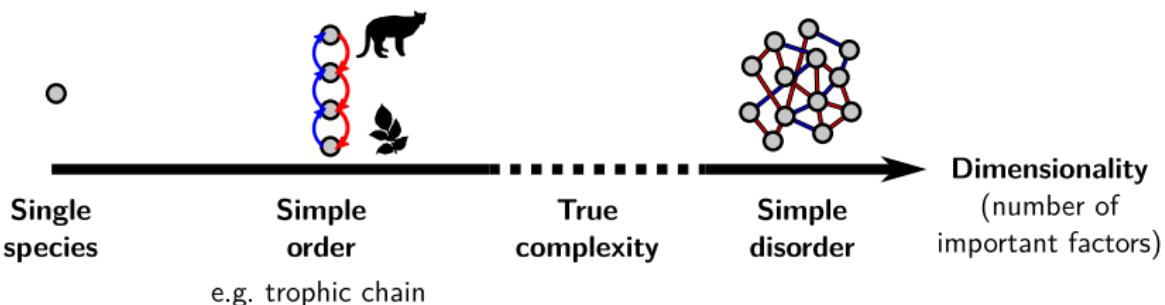
“Random” means “too many factors”, so complex mechanistically that it becomes simple statistically

## SMALL AND LARGE SYSTEMS

All that to ask: is there simplicity from apparent complexity in ecology?

# SMALL AND LARGE SYSTEMS

All that to ask: is there simplicity from apparent complexity in ecology?



Modelling an ecological community can start

- from “small simplicity” (e.g. a 3-species trophic chain)
- or from “large simplicity” = many-species networks...  
but when & how are they simple?

## II. MANY-SPECIES COMMUNITIES

---

PART 1: WHAT OBSERVATIONS ARE WE TRYING TO EXPLAIN

Forget about randomness for now, just study communities with many populations



Hereafter “species”, but could be strains, phenotypes, etc.

# OBSERVABLES IN LARGE COMMUNITIES

What is interesting in large communities:

- we lose focus on individual species – they are often unpredictable, maybe impacted by dozens or hundreds of others

# OBSERVABLES IN LARGE COMMUNITIES

What is interesting in large communities:

- we lose focus on individual species – they are often unpredictable, maybe impacted by dozens or hundreds of others
- we gain aggregate properties
  - static properties
  - dynamical properties

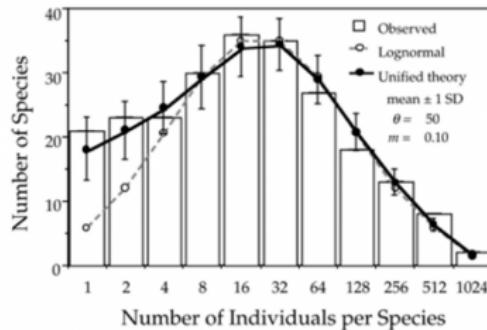
## STATIC PROPERTIES

Measurable from a single/few snapshots:

# STATIC PROPERTIES

Measurable from a single/few snapshots:

- Distributions (= histograms, frequencies)

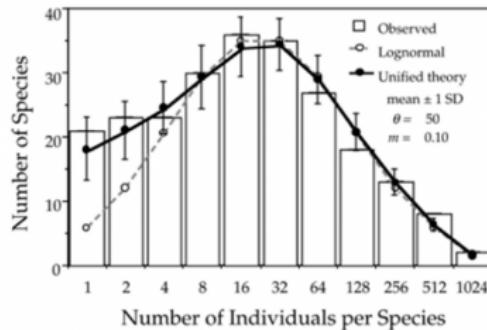


abundances, number of offsprings/production, variation in space or time...

# STATIC PROPERTIES

Measurable from a single/few snapshots:

- Distributions (= histograms, frequencies)



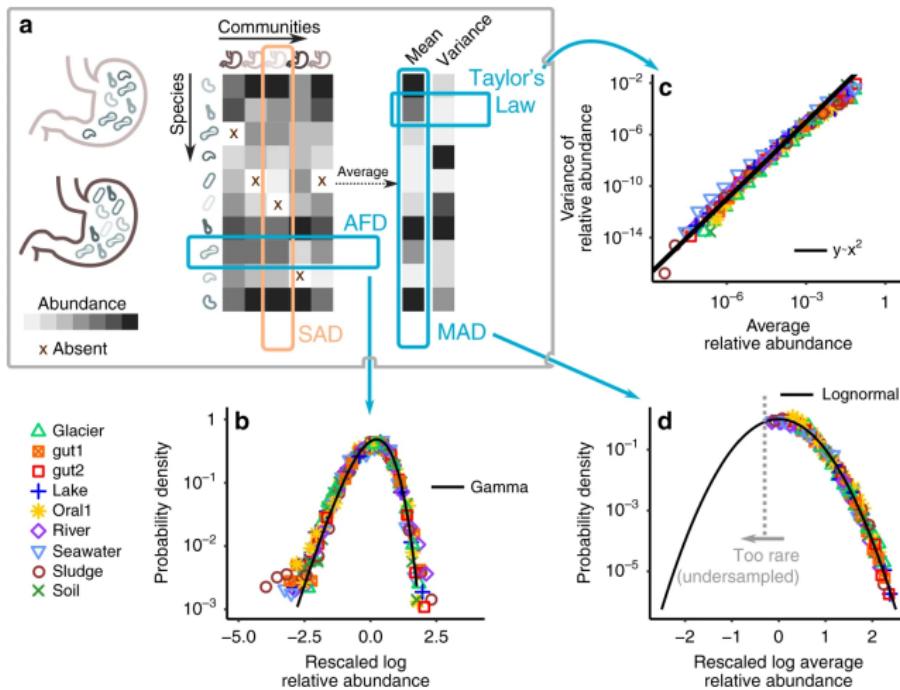
abundances, number of offsprings/production, variation in space or time...

- Statistics on these distributions:

- diversity (number of coexisting species, Shannon)
- total abundance  $\sum_i N_i$ , total production  $\sum_i r_i N_i$

# STATIC PROPERTIES

Many common patterns are different ways of aggregating same basic data



## DYNAMICAL PROPERTIES

Properties that can only be observed by tracking species over time, e.g.

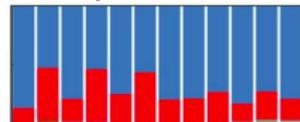
- Is an ecosystem in a stable equilibrium or other dynamical regime?
- How does it respond when you disturb it?

# DYNAMICAL PROPERTIES

What is the usual regime of a given ecosystem?

- equilibrium (with small fluctuations)

*example: near-constant populations of microbial functional groups*

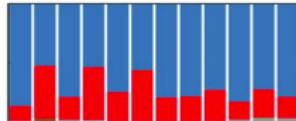


# DYNAMICAL PROPERTIES

What is the usual regime of a given ecosystem?

- equilibrium (with small fluctuations)

*example: near-constant populations of microbial functional groups*

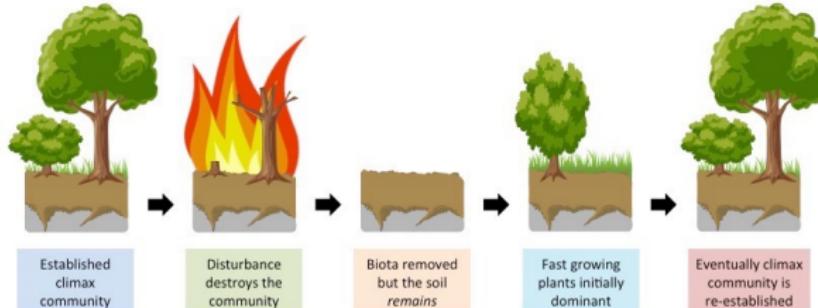


- directional trajectory

*example: microbial succession during decomposition*

- stationary nonequilibrium

*example: cycles, chaos, constant flux of species invading and dying*



# DYNAMICAL PROPERTIES

How does an ecosystem respond when you disturb it?

- “elastic”, goes back to its state or trajectory

*example: gut microbiome disturbed by sickness then re-colonized*

# DYNAMICAL PROPERTIES

How does an ecosystem respond when you disturb it?

- “elastic”, goes back to its state or trajectory

*example: gut microbiome disturbed by sickness then re-colonized*

- “malleable”, remains modified, does not go back

*example: humans plant trees outside their original range, they remain in the new biome*

# DYNAMICAL PROPERTIES

How does an ecosystem respond when you disturb it?

- “elastic”, goes back to its state or trajectory

*example: gut microbiome disturbed by sickness then re-colonized*

- “malleable”, remains modified, does not go back

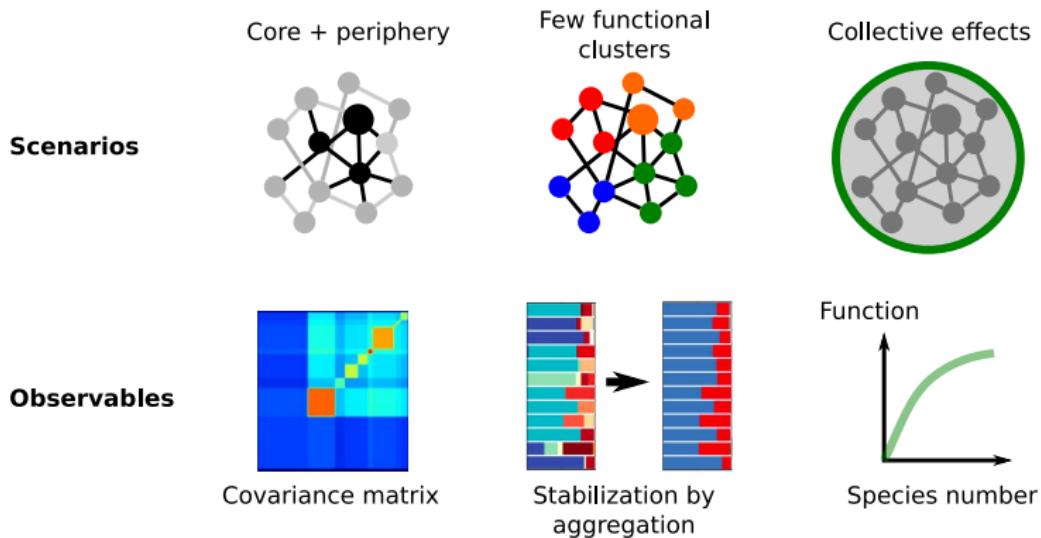
*example: humans plant trees outside their original range, they remain in the new biome*

- “chaotic”, becomes more and more different

*example: a single invasive species causes a cascade of extinctions, then invasions, then extinctions...*

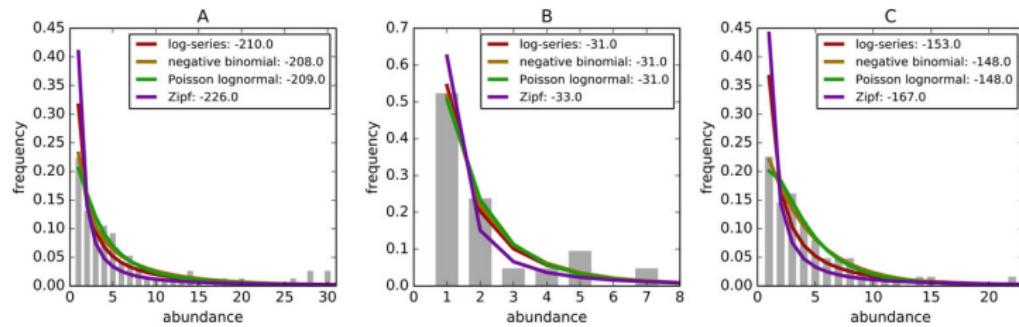
# FINGERPRINTS OF ECOLOGICAL SCENARIOS

**Why do we care?** Often, these patterns are used as “fingerprints” to reveal some ecological scenarios...



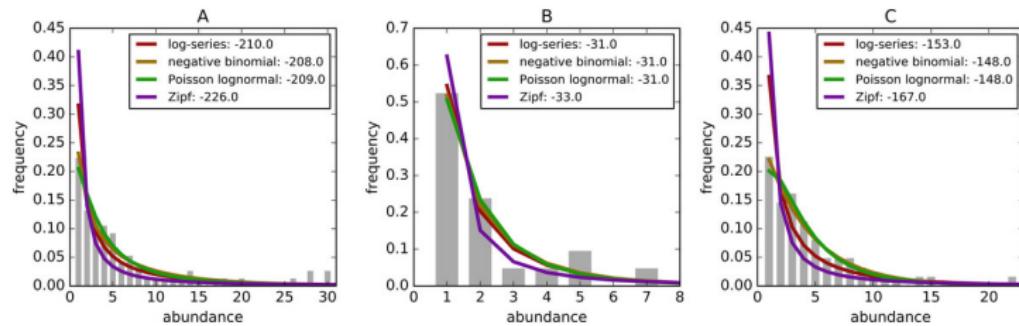
# FINGERPRINTS OF ECOLOGICAL SCENARIOS

... But usually no single pattern is a “smoking gun” (e.g. many different explanations for same abundance histograms)



# FINGERPRINTS OF ECOLOGICAL SCENARIOS

... But usually no single pattern is a “smoking gun” (e.g. many different explanations for same abundance histograms)



⇒ Goal: How do we construct a simple model that explains as many community patterns as possible? (to allow a real test)

## II. MANY-SPECIES COMMUNITIES

---

### PART 2: HOW DO WE EXPLAIN OBSERVATIONS

# MODEL

Back to our Lotka-Volterra

$$\frac{dN_i}{dt} = r_i N_i (1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j) \quad (5)$$

# MODEL

Back to our Lotka-Volterra

$$\frac{dN_i}{dt} = r_i N_i (1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j) \quad (5)$$

- maybe the snapshot properties (e.g. abundance distribution) are given by some *equilibrium* of the equation?

$$\frac{dN_i}{dt} = 0 \Rightarrow \begin{cases} N_i = 0 & \text{extinct species} \\ N_i = (1 - \sum_{j \neq i}^S a_{ij} N_j) / a_{ii} & \text{surviving species} \end{cases} \quad (6)$$

# MODEL

Back to our Lotka-Volterra

$$\frac{dN_i}{dt} = r_i N_i (1 - a_{ii} N_i - \sum_{j \neq i}^S a_{ij} N_j) \quad (5)$$

- maybe the snapshot properties (e.g. abundance distribution) are given by some *equilibrium* of the equation?

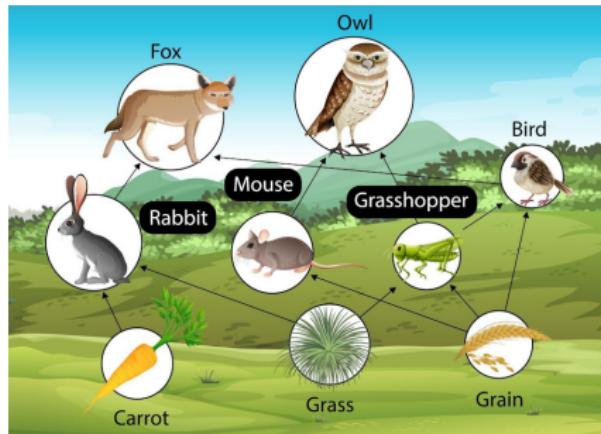
$$\frac{dN_i}{dt} = 0 \Rightarrow \begin{cases} N_i = 0 & \text{extinct species} \\ N_i = (1 - \sum_{j \neq i}^S a_{ij} N_j) / a_{ii} & \text{surviving species} \end{cases} \quad (6)$$

- maybe dynamical properties can be understood, e.g. how does  $N_i(t)$  respond to perturbation, does it go to equilibrium or cycle...

# SPECIES INTERACTION NETWORKS

How do we obtain the matrix of interactions  $a_{ij}$ ?

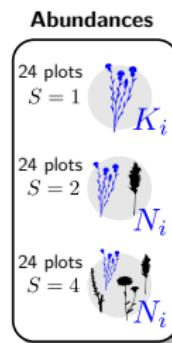
- *Good news:* qualitative structure ( $a_{ij} = 0$  or  $\neq 0$ ) can be known for some interaction types, e.g. who eats who



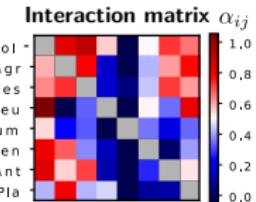
# SPECIES INTERACTION NETWORKS

How do we obtain the matrix of interactions  $a_{ij}$ ?

- *Bad news:* quantitative strength ( $a_{ij}$  values) is very rarely measured directly for every pair of species  $i, j$  (few experiments doing all that)

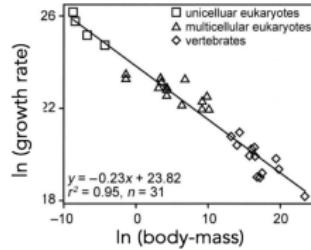


Multilinear fit



# SPECIES INTERACTION NETWORKS

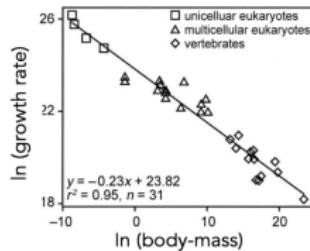
- Most of the time, theoretical assumptions are needed to put numbers into the model:
  - Metabolic scaling,  $r_i$  and  $a_{ij}$  given by body sizes of species  $i$  and  $j$



- Ecopath model (see with Claire this afternoon)
- ...

# SPECIES INTERACTION NETWORKS

- Most of the time, theoretical assumptions are needed to put numbers into the model:
  - Metabolic scaling,  $r_i$  and  $a_{ij}$  given by body sizes of species  $i$  and  $j$



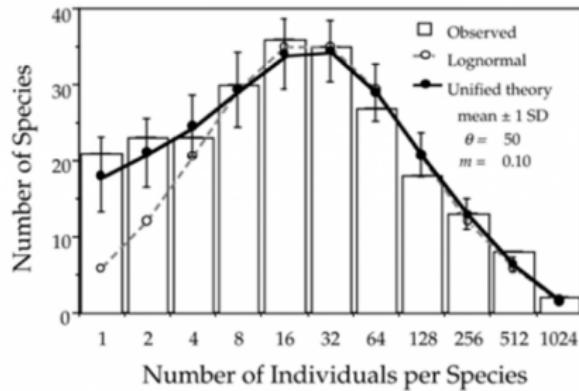
- Ecopath model (see with Claire this afternoon)
- ...
- What do we do if we cannot or do not want to assume anything?

# NEUTRALITY

- Extreme simplification: neutrality, all species identical, e.g.  
 $r_i = a_{ij} = 1$
- Different outcomes for different species only due to chance: random events of birth, death and migration

# NEUTRALITY

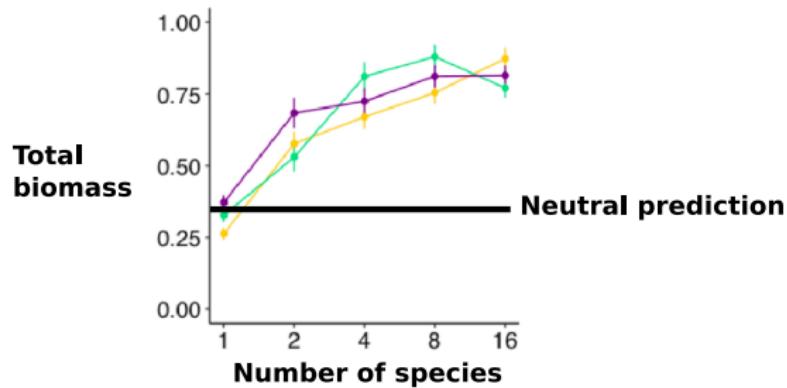
- Extreme simplification: neutrality, all species identical, e.g.  $r_i = a_{ij} = 1$
- Different outcomes for different species only due to chance: random events of birth, death and migration
- Why use it? Because it can suffice to predict some patterns, e.g. abundance distributions



# NEUTRALITY

Why go beyond neutral? It fails for other patterns, e.g.

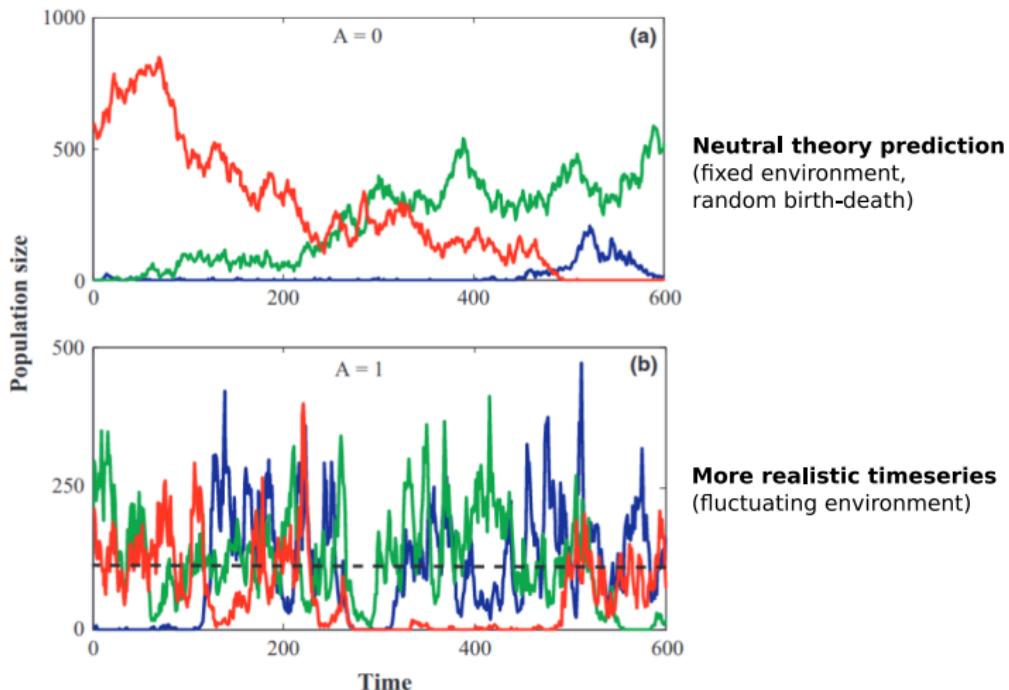
- More biomass when more species (neutral theory = zero-sum game, total biomass is fixed)



# NEUTRALITY

Why go beyond neutral? It fails for other patterns, e.g.

- Temporal fluctuations from original neutral theory are too slow



# RANDOM INTERACTIONS

Next simplest thing:

- neutrality = identical interactions

$$a = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \quad (7)$$

# RANDOM INTERACTIONS

Next simplest thing:

- neutrality = identical interactions

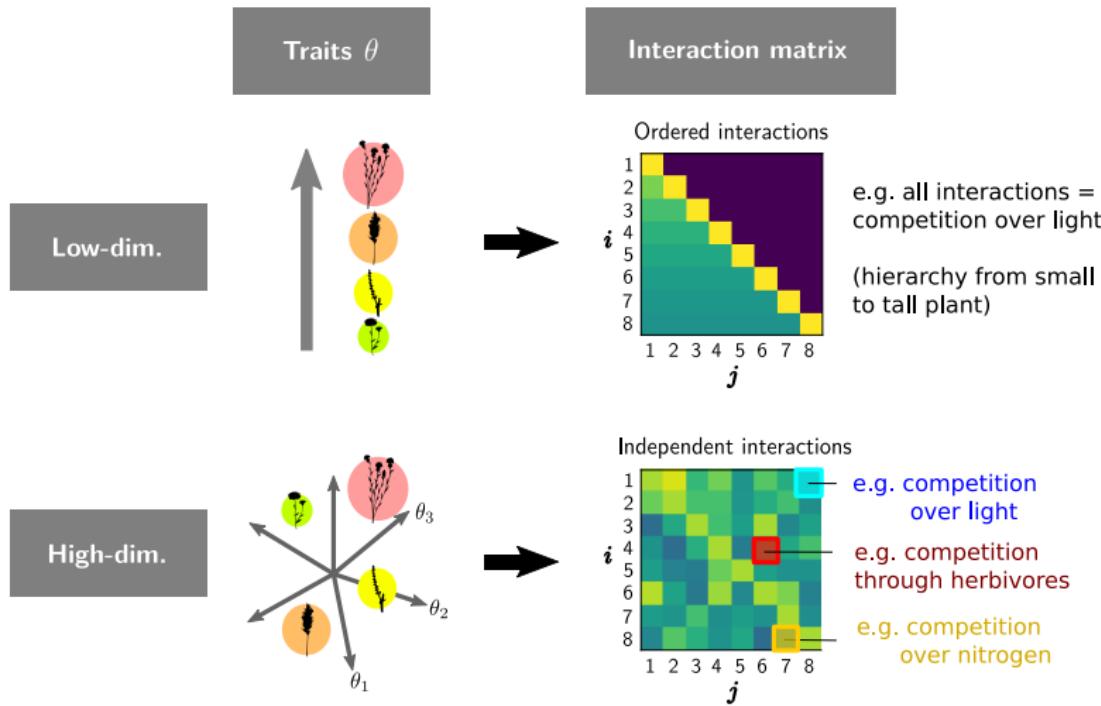
$$a = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 \end{pmatrix} \quad (7)$$

- instead, take interactions  $a_{ij}$  that are different, but drawn at random

$$a = \begin{pmatrix} 0.29 & 0.54 & 0.53 & 0.02 & 0.40 \\ 0.57 & 0.86 & 0.90 & 0.81 & 0.76 \\ 0.53 & 0.11 & 0.42 & 0.44 & 0.09 \\ 0.15 & 0.72 & 0.84 & 0.27 & 0.94 \\ 0.87 & 0.85 & 0.61 & 0.36 & 0.63 \end{pmatrix} \quad (8)$$

# RANDOM INTERACTIONS

Justification: interactions not “really” uncertain, but caused by many independent ecological traits, mechanisms, etc.



# PREDICTIONS

As I said at the start, under broad conditions, results only depend on 3 parameters

- mean of interactions  $\langle a_{ij} \rangle$  ( $= E[a_{ij}]$  if you prefer)
- standard deviation  $std(a_{ij})$
- and symmetry  $corr(a_{ij}, a_{ji})$

# How?

From a system of  $S$  equilibrium equations on all the abundances  
(here  $a_{ii} = 1$  for simplicity)

$$N_1 = 1 - \sum_{j \neq 1}^S a_{1j} N_j \quad (9)$$

$$N_2 = 1 - \sum_{j \neq 2}^S a_{2j} N_j \quad (10)$$

$$\dots \quad (11)$$

# How?

From a system of  $S$  equilibrium equations on all the abundances  
(here  $a_{ii} = 1$  for simplicity)

$$N_1 = 1 - \sum_{j \neq 1}^S a_{1j} N_j \quad (9)$$

$$N_2 = 1 - \sum_{j \neq 2}^S a_{2j} N_j \quad (10)$$

$$\dots \quad (11)$$

We can go to a few equations on statistics of abundances, almost like this (a bit more complicated)

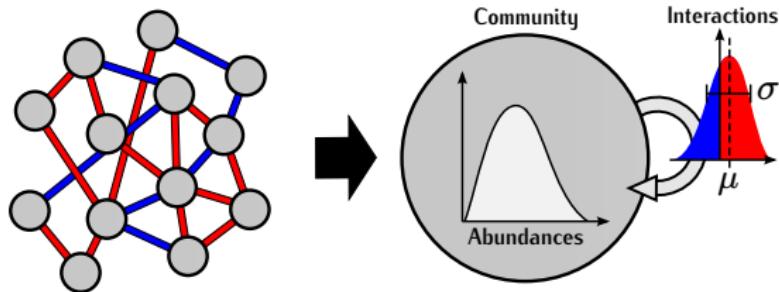
$$\langle N \rangle = 1 - S \langle a \rangle \langle N \rangle \quad (12)$$

$$\text{var}(N) = S \text{ var}(a) \langle N^2 \rangle \quad (13)$$

where the **only parameters** are number of species  $S$ , and statistics of interactions  $\langle a \rangle$ ,  $\text{var}(a)$  (or  $\text{std}(a)$ )

# WHY?

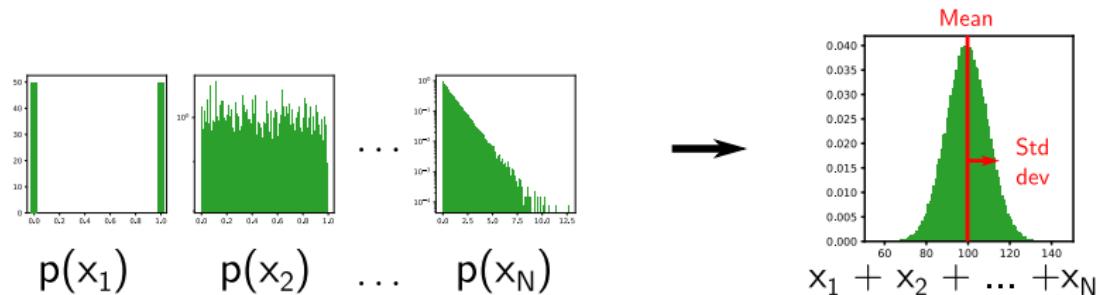
Why does it work to just write equations on statistics (e.g. mean and variance) of abundance?



In a disordered network, all species are different but statistically equivalent  
(no special role/position, any one at random is a fair sample from whole distribution)

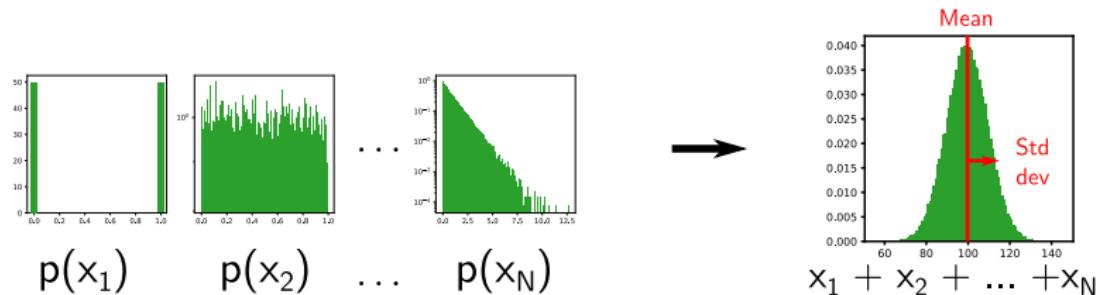
# WHY?

- Like Central Limit Theorem: many independent variables together create a Gaussian, with only 2 parameters: mean and variance



# WHY?

- Like Central Limit Theorem: many independent variables together create a Gaussian, with only 2 parameters: mean and variance



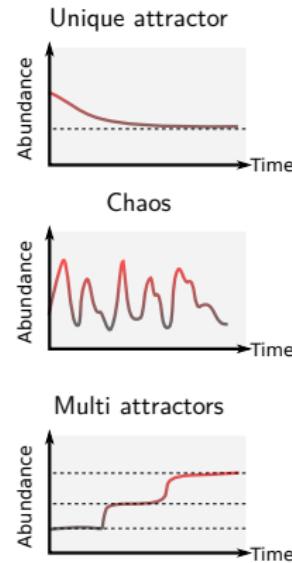
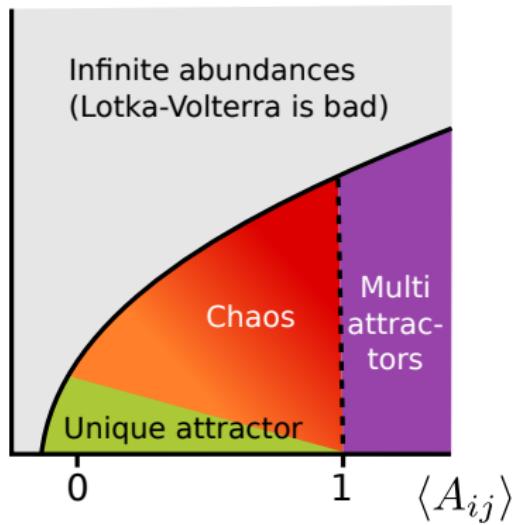
- Same is true with networks: many independent interactions together create a simple statistical result with only 3 parameters

How do we prove that result? Mathematical methods from physics

# PREDICTIONS: DYNAMICS

- Same reasoning works for dynamics, " $d\langle N \rangle / dt$ ,  $d\text{var}(N) / dt \dots$ "
- Hence, with few parameters, we can answer the question of how a many-species network behaves dynamically "by default" (when random)

$$\text{std}(A_{ij})$$

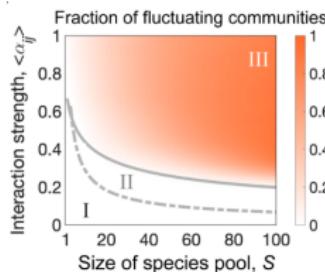
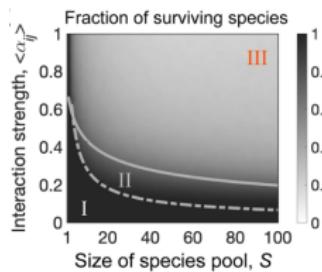


NB: Chaotic phase shows "realistic" fluctuations

# EMPIRICAL TEST

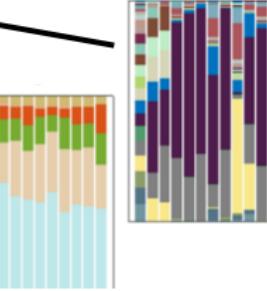
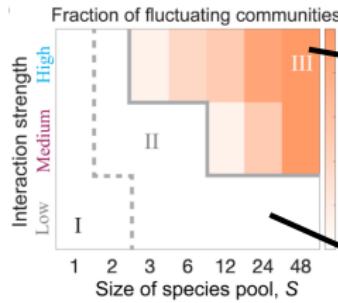
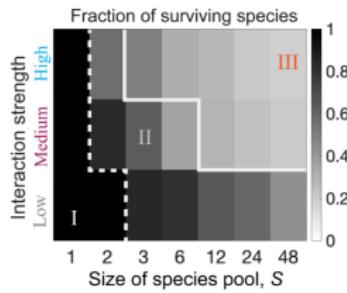
white = unique equilibrium, orange = fluctuations (in theory, chaos)

## Random Lotka-Volterra Theory



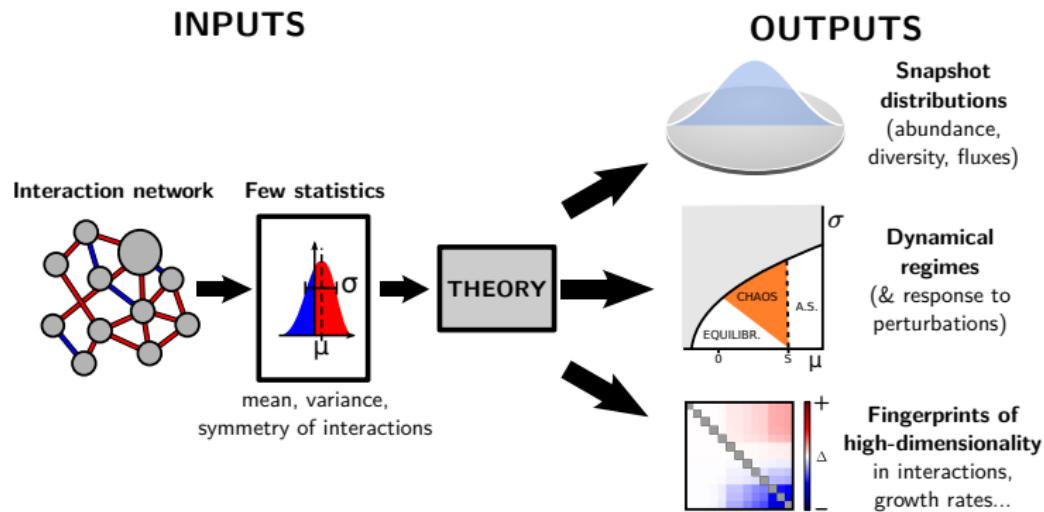
- Survival boundary
- Stability boundary
- Phase I: stable full coexistence
- Phase II: stable partial coexistence
- Phase III: persistent fluctuation

## Microbial experiments



# RANDOM COMMUNITIES: A SUMMARY

Random interactions = a few input parameters, many testable outputs



# RANDOM INTERACTIONS

40 years of empirical successes for random interaction models:

- In physics: glass, large atoms
- In chemistry: large reaction networks
- In neuroscience: neural networks
- In computer science: complex satisfaction problems (NP-hardness)

# RANDOM INTERACTIONS

40 years of empirical successes for random interaction models:

- In physics: glass, large atoms
- In chemistry: large reaction networks
- In neuroscience: neural networks
- In computer science: complex satisfaction problems (NP-hardness)

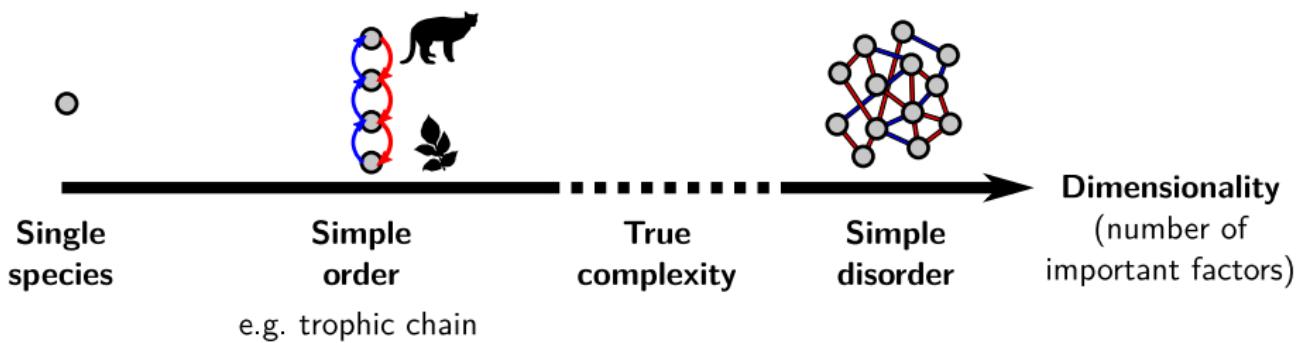
... but do we really believe that ecological systems can be modelled as completely random?

### III. ORDER AND DISORDER

# COMBINING ORDER AND DISORDER

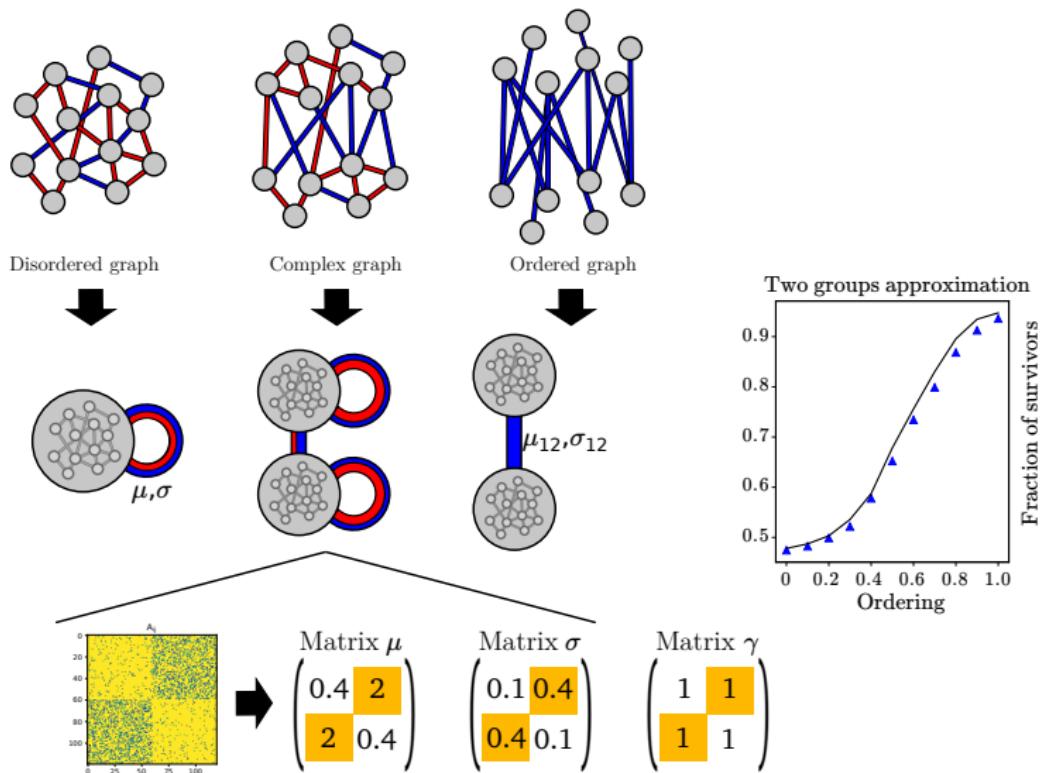
*"There is a fundamental dichotomy between structure and randomness, which in turn leads to a decomposition of any object into a structured (low-complexity) component and a random (discorrelated) component."*

– Terence Tao

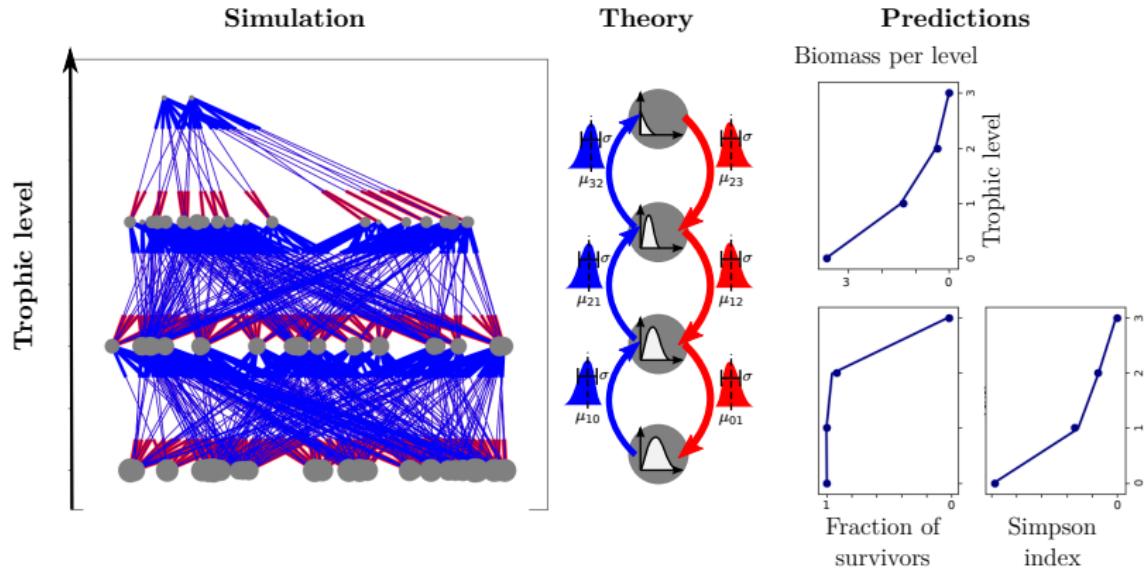


**Claim:** Often, apparently complex systems behave like interpolation between simple order & disorder

# EXAMPLE 1: COMPETITORS AND MUTUALISTS



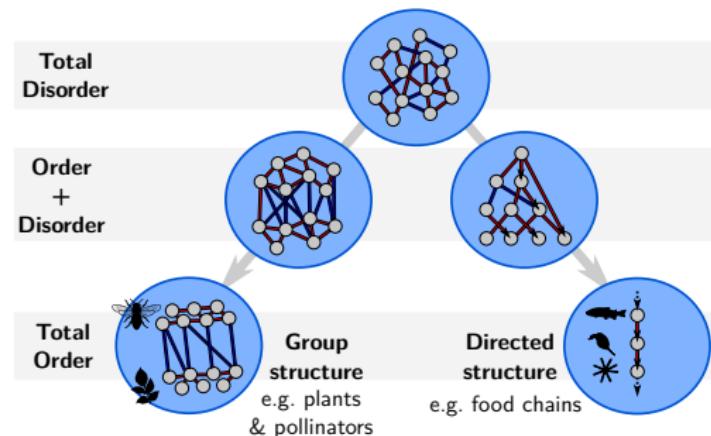
## EXAMPLE 2: FOOD WEBS



(but also size hierarchy, nestedness, trade-offs...)

# TWO SIMPLICITIES

In brief:

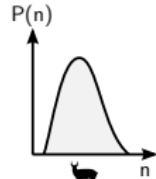
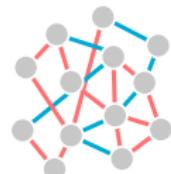
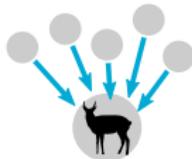


- Disorder = plausible null model for (single-functional group) communities with many factors causing interactions
- Order+disorder decomposition can reduce more complex systems to only few more parameters, but there are different types of simple order (most classically: blocks, nestedness, directedness)

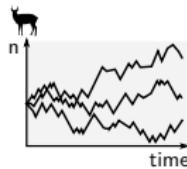
forget who does what,  
count what happens

## Many things matter

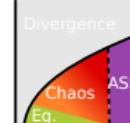
keystone spp/traits,  
PCA,  
network metrics,  
...



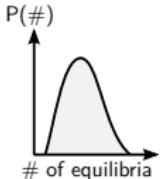
histograms



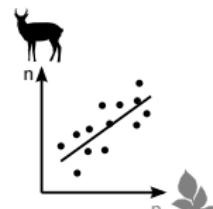
random walks



collective  
dynamical  
regimes

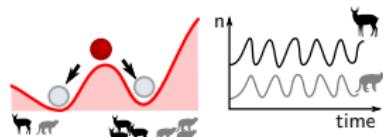


counting  
states



regressions

joint states or dynamics  
e.g. alternate states, cycles



## Few things matter

identify who does what



## One-way causality

more stats  $\longleftrightarrow$  more dyn



## Feedbacks

## TAKE AWAYS

- Observables in large systems:
  - snapshot patterns: distribution and statistics
  - dynamics: number & nature of attractors, sensitivity to perturbations
- Can we really identify ecological mechanisms from observables, or can we explain observables without knowing much about mechanisms?
  - For many observables, not all details of mechanisms matter; the art of modelling involves understanding when and which details are lost
- Randomness = particular case where we can prove that all details are lost except a few basic statistics
  - **drawing things at random does not mean you explore all possibilities!**
  - useful as null model; to know if network structure is important for a result, compare to result of random networks with similar statistics
  - can be mixed with simple structure (e.g. functional groups, nestedness...) to model “complex” networks  
⇒ what seems complex may be largely random