

From species cooccurrences...

...to species interactions



# Part 1.

# A historical perspective

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A historical perspective

Part 2.

A "modern" permutation-based null model

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A historical perspective

Part 2.

A "modern" permutation-based null model

Part 3.

Geographical scale and cooccurrence patterns



jareddiamond.org



New guinea



A Fruit Dove (*Ptilinopus* sp.)

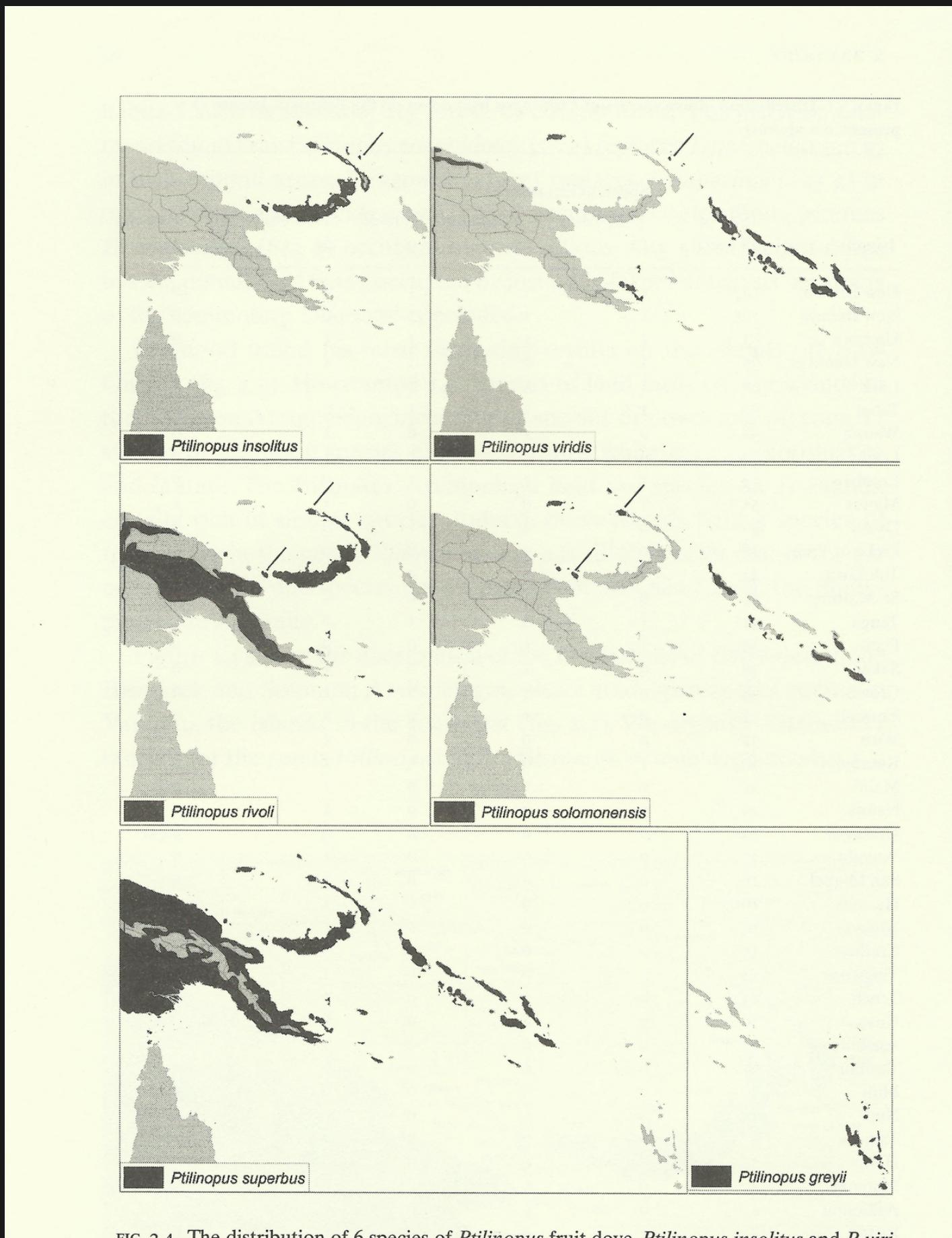


FIG. 2.4. The distribution of 6 species of *Ptilinopus* fruit dove. *Ptilinopus insolitus* and *P. viridis*

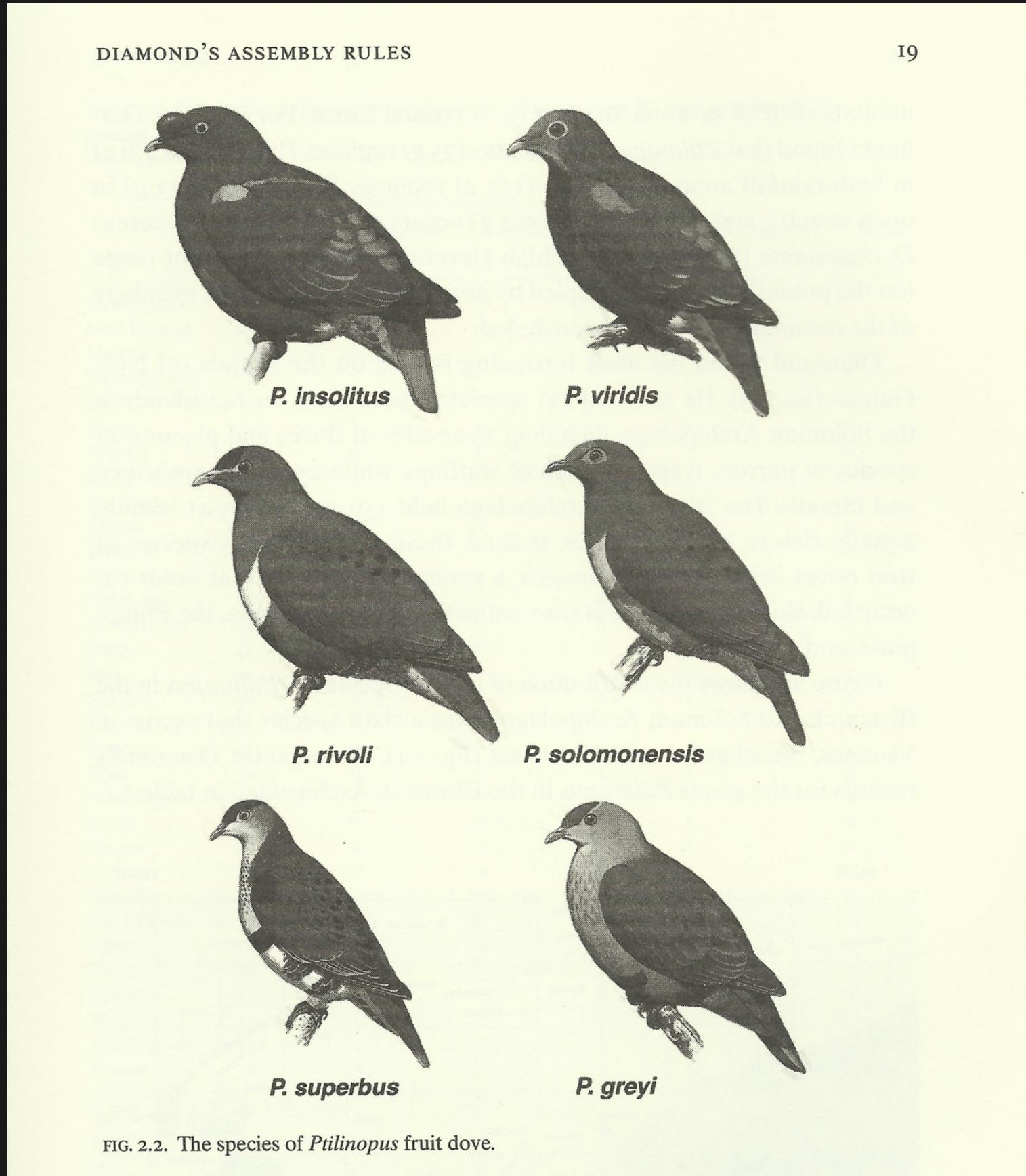


FIG. 2.2. The species of *Ptilinopus* fruit dove.

# Checkerboards



# Checkerboards



Evidence of competition shaping  
communities at large-scale ?

« In a few instances, competition expresses itself in ‘simple’ checkerboards distributions, by which species replace each other one-for-one »

Diamond, 1975 (In Cody and Diamond, 1975)

## Assembly rules

In a widely cited new approach to the interpretation of biogeographic distributions, Diamond (1975) asserts that the assembly of bird communities manifests the following patterns:

- a. "If one considers all the combinations that can be formed from a group of related species, only certain ones of these combinations exist in nature."
- b. "Permissible combinations resist invaders that would transform them into forbidden combinations."
- c. "A combination that is stable on a large or species-rich island may be unstable on a small or species-poor island."
- d. "On a small or species-poor island, a combination may resist invaders that would be incorporated on a larger or more species-rich island."
- e. "Some pairs of species never coexist, either by themselves or as a part of a larger combination."
- f. "Some pairs of species that form an unstable combination by themselves may form part of a stable larger combination."
- g. "Conversely, some combinations that are composed entirely of stable subcombinations are themselves unstable."

« We show that every assembly rule is either tautological, trivial, or a pattern expected were species distributed at random »

Connor and Simberloff, 1979

New Hebrides (Vanuatu) birds :  
58 species x 28 islands

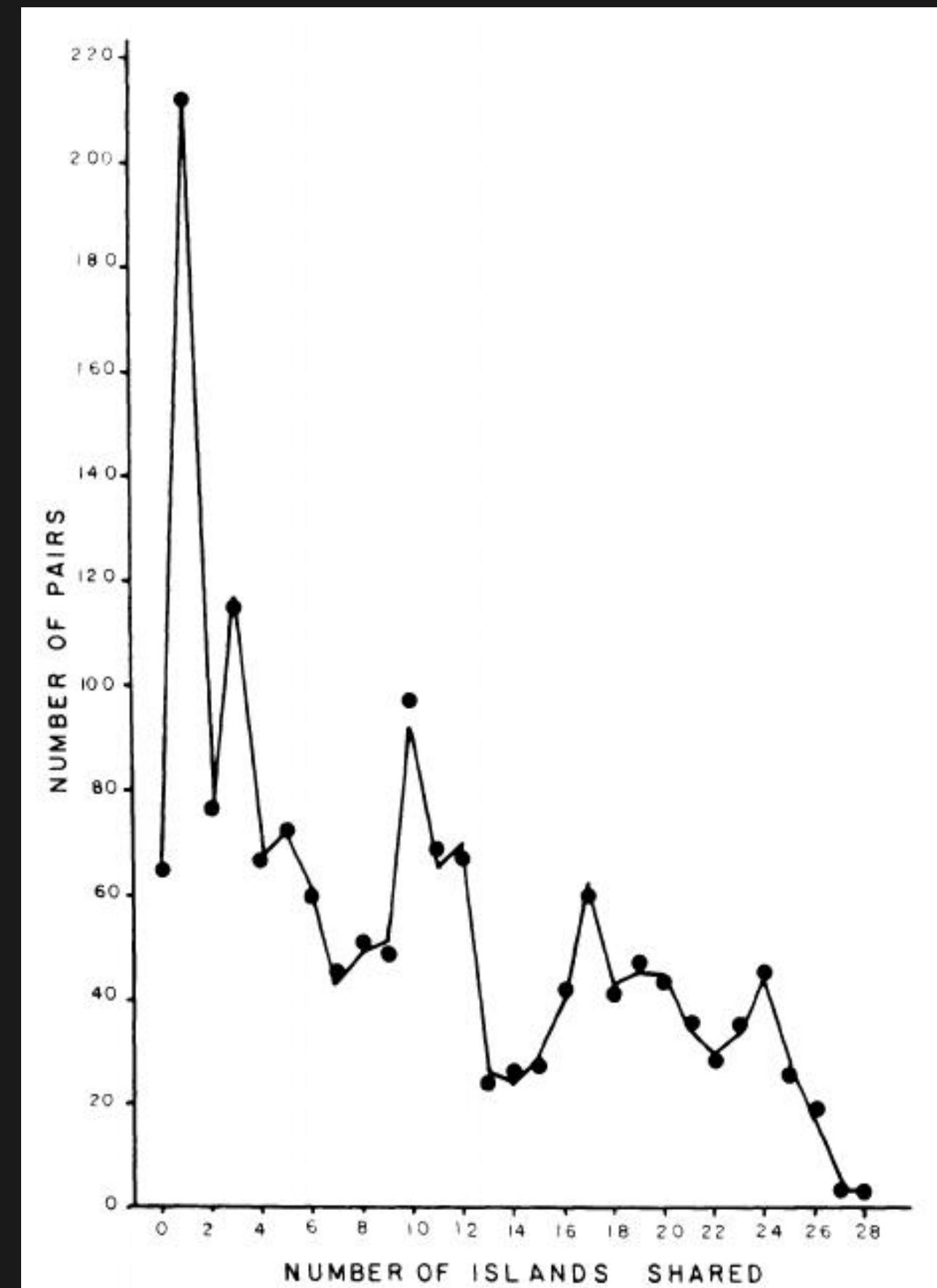


FIG. 1. Distribution of number of species pairs vs. number of islands shared for New Hebrides birds. Solid line is the expected distribution given the three constraints discussed in text. Dots are the observed values.

« We show that every assembly rule is either tautological, trivial, or a pattern expected were species distributed at random »  
Connor and Simberloff, 1979

« Diamond (1975) assumed competition to be the primary determinant and then sought post facto to rationalize the observed data in the light of this assumption »

Connor and Simberloff, 1979

New Hebrides (Vanuatu) birds :  
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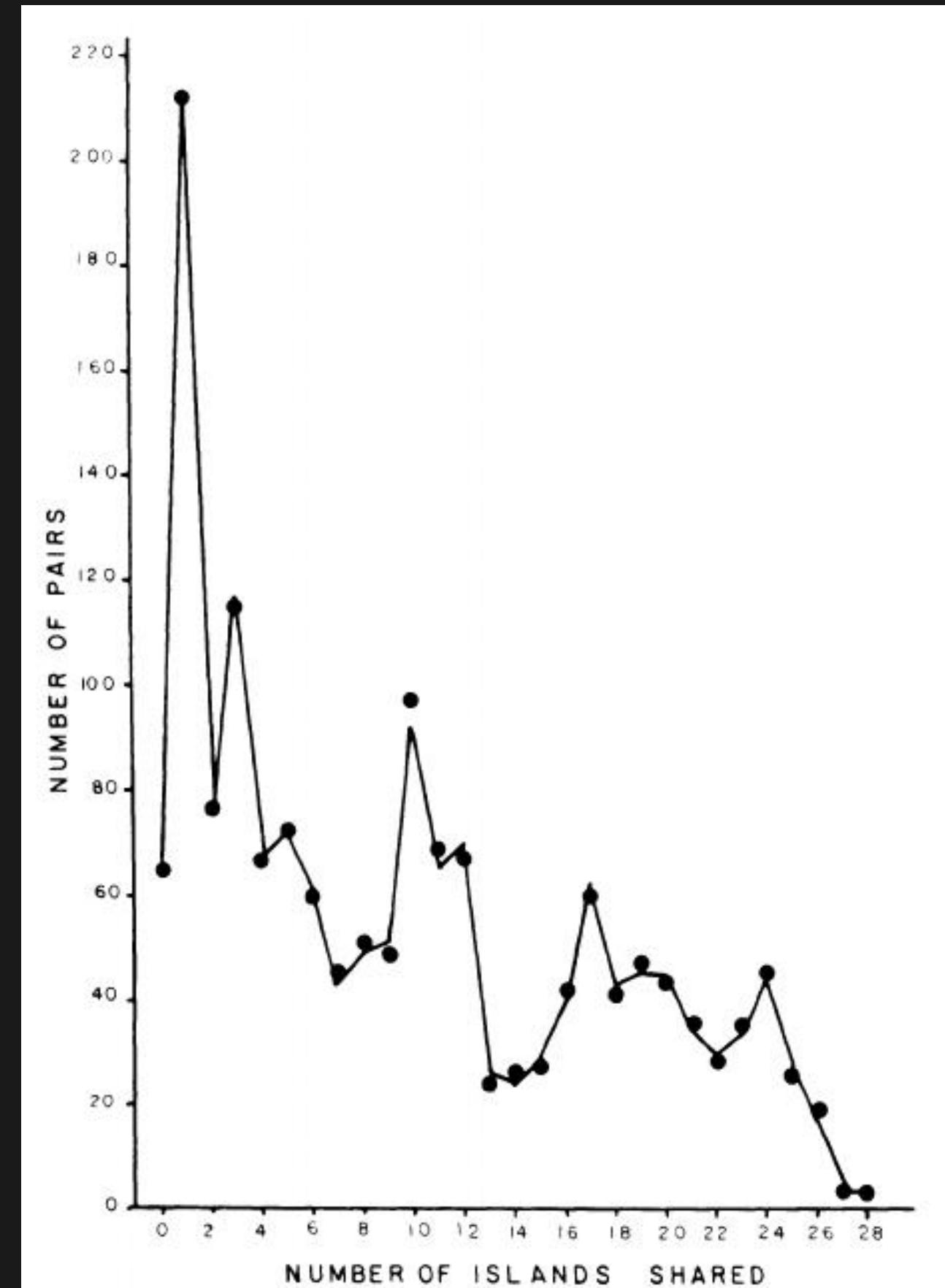


FIG. 1. Distribution of number of species pairs vs. number of islands shared for New Hebrides birds. Solid line is the expected distribution given the three constraints discussed in text. Dots are the observed values.

## Connor and Simberloff's null model

A incidence matrix :

Species	Sites							
	1	2	3	4	5	6	7	8
1	0	0	0	0	0	1	0	1
2	0	0	1	0	1	1	0	3
3	1	0	1	1	1	0	0	4
4	0	1	1	0	1	0	0	3
5	1	1	3	1	3	2	0	

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5	1	1	3	1	3	2	0	

Islands can inherently be "bigger" (hold more species)  
> the **column sum** is constrained

Species have an inherent dispersal ability  
> the **row sum** is constrained

Species occur in a given "community size"  
> if a species occur in a community with S species,  
then it can only occur with S species in the null model

## Connor and Simberloff's null model

	Sites							
Species	0	0	0	0	0	1	0	1
	0	0	1	0	1	1	0	3
	1	0	1	1	1	0	0	4
	0	1	1	0	1	0	0	3
	1	1	3	1	3	2	0	

$$\begin{matrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 1 & 1 \end{matrix} \longrightarrow \begin{matrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 \end{matrix}$$

a Swap algorithm

# Connor and Simberloff's null model

"several swaps", nobody knows (?)

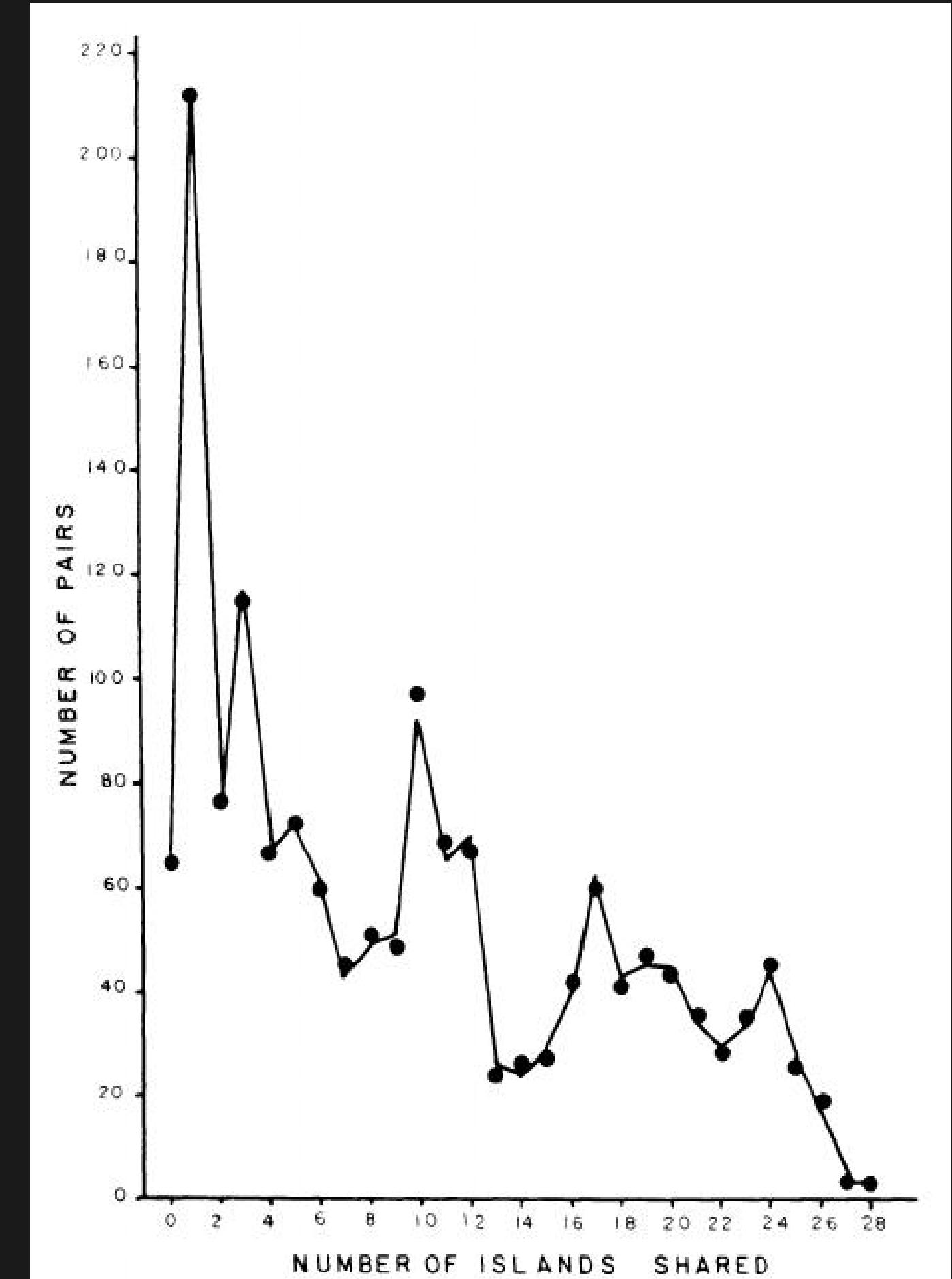


FIG. 1. Distribution of number of species pairs vs. number of islands shared for New Hebrides birds. Solid line is the expected distribution given the three constraints discussed in text. Dots are the observed values.

A "good" permutation-based null model

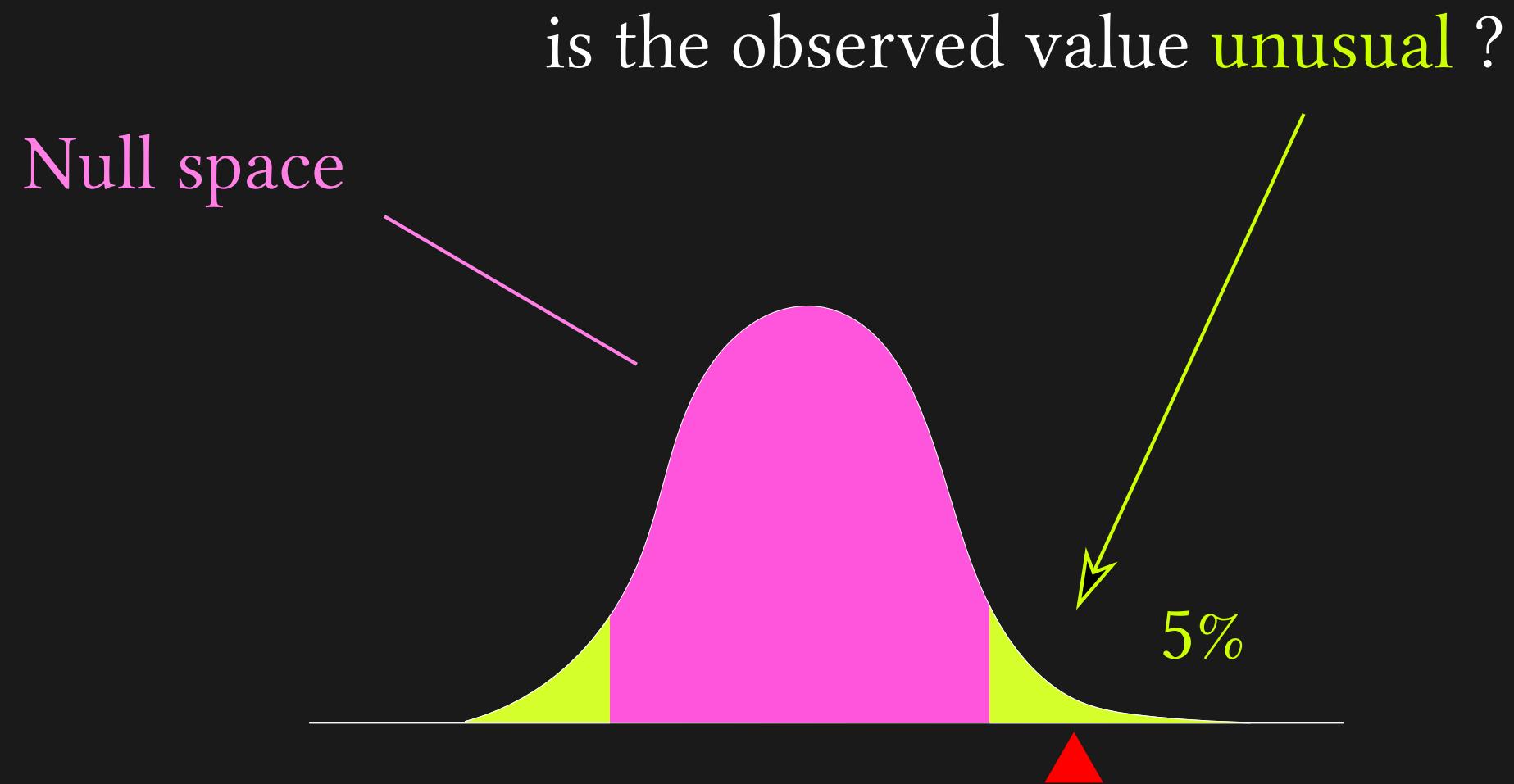
Apparent mutual exclusion, measured by the number of co-ocurrences

A pattern, measured by a metric

A "good" permutation-based null model

Apparent mutual exclusion, measured by the number of co-occurrences

A pattern, measured by a metric



## Metrics

- \* The number of co-occurrences

$$S_{ij}$$

$$\overline{S}_{ij}$$

Species-pair  
level

Matrix level

## Metrics

- \* The number of co-occurrences

$$S_{ij} \quad \overline{S}_{ij}$$

- \* The checkerboard score

Species-pair  
level      Matrix level

$$C_{ij} = (r_i - S_{ij})(r_j - S_{ij})$$

$$C_{score} = \frac{2}{n * (n - 1)} \sum_{i=1}^{n-1} \sum_{j=i+1}^n C_{ij}$$

- \* Togetherness, etc...

# Generating the null space

Constraints

Row sums

Column sums

(Incidence) ————— The Narcissus effect  
Colwell & Winkler 1984



Many, many, many matrices

Species	Sites
0	0 0 0 0 0 1 0 1
0	0 0 1 0 1 1 0 3
1	0 1 1 1 1 0 0 4
0	1 1 1 0 1 0 0 3
1	1 1 3 1 3 2 0

# Generating the null space

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Row sums

Column sums

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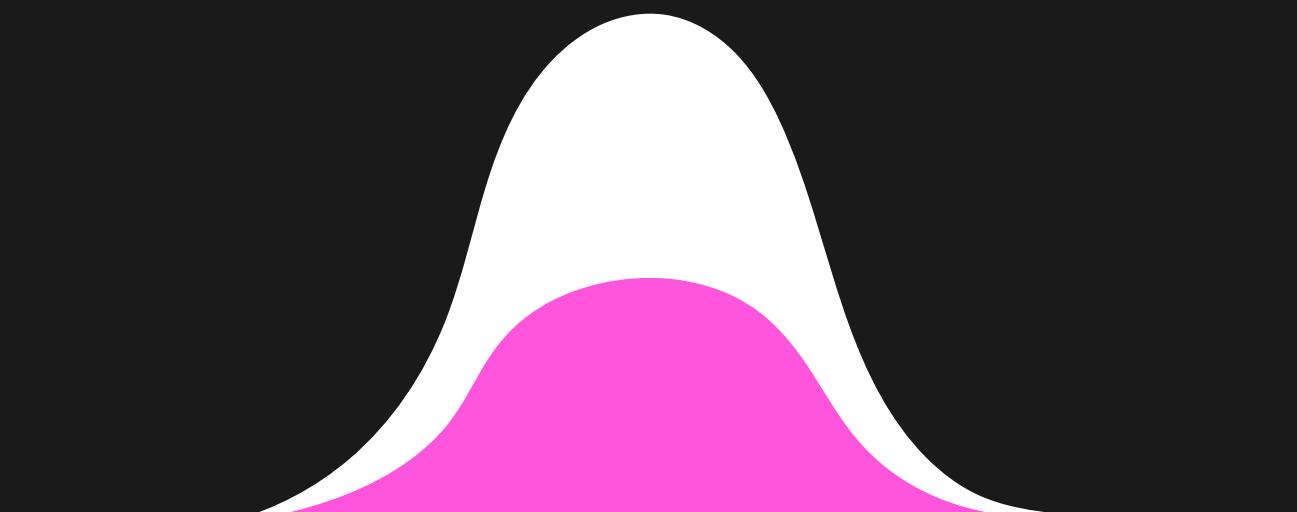
Many, many, many matrices

Very often, the full null space is  
impossible to compute !

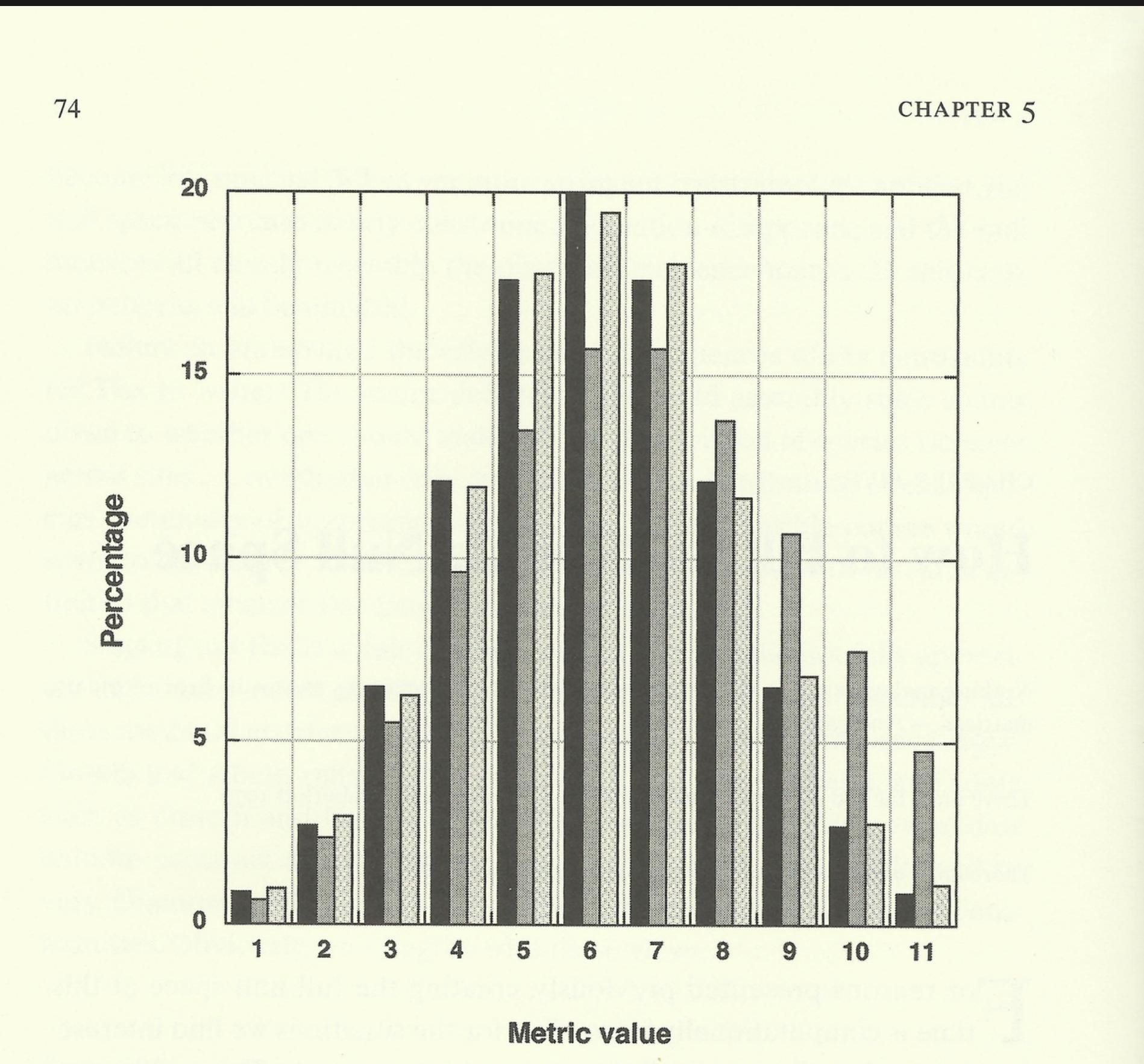


We compute a sample null space

Sites	0	0	0	0	0	1	0	1
Species	0	0	1	0	1	1	0	3
1	0	1	1	1	1	0	0	4
0	1	1	0	1	0	0	0	3
1	1	3	1	3	2	0		



## Generating the null space



We compute a sample null space

# Generating the null space

Enumeration algorithms

*expensive*  
build all the possible matrices

Swap algorithms

*cheap*  
start from a matrix, then swap submatrices

the trial-swap method (Miklos & Podani, 2004)

Construction algorithms

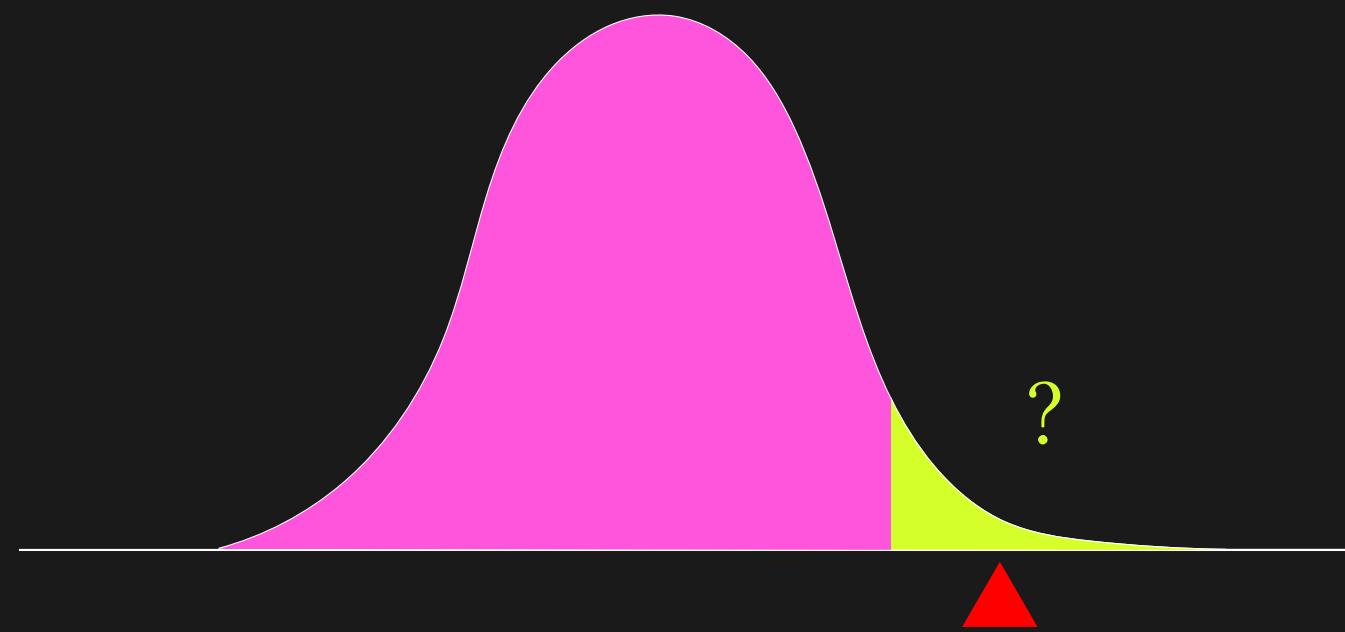
*expensive*  
start from zeroes/random, then build a  
matrix satisfying the constraints

We compute a sample null space

# Testing

Does our metric value fall within the tails ?

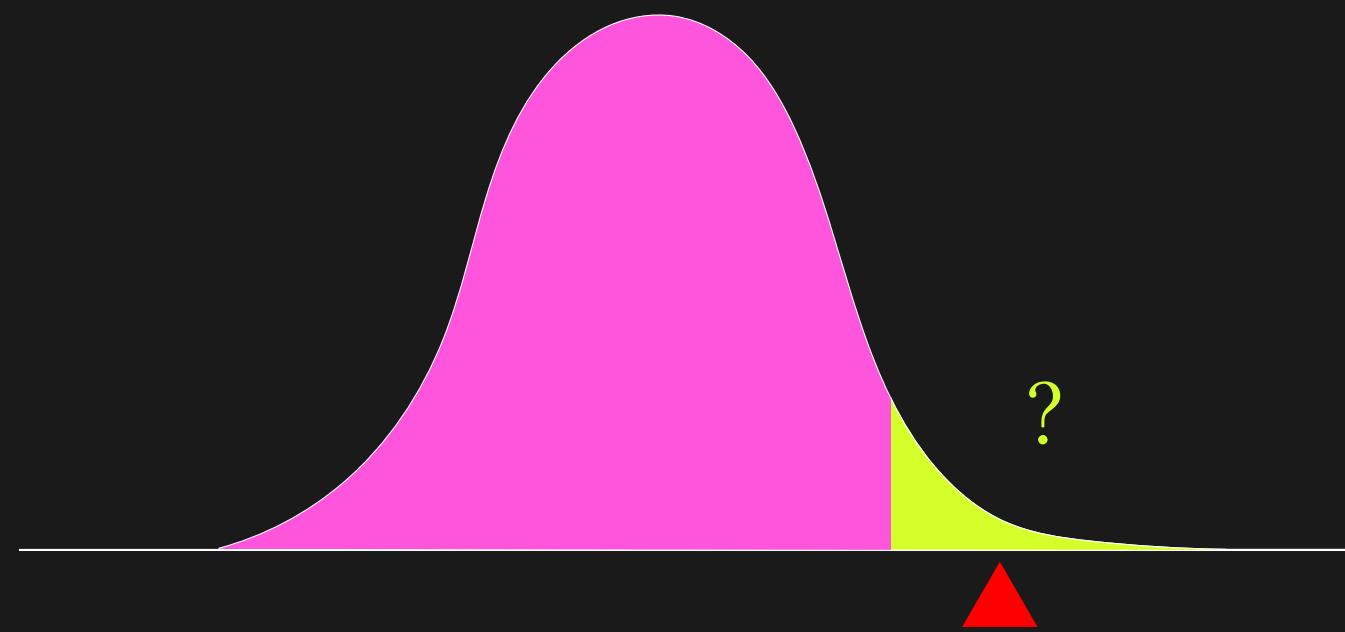
(assume matrix-level metric -> one test)



# Testing

Does our metric value fall within the tails ?

(assume matrix-level metric -> one test)



-> Your turn !

[http://alex.lecairn.org/files/01\\_nerdchat.R](http://alex.lecairn.org/files/01_nerdchat.R)

[http://alex.lecairn.org/files/02\\_nerdchat.R](http://alex.lecairn.org/files/02_nerdchat.R)

# Problems

Problems: the pringle effect



we lump everything into one single measure

> use a different, pair-level measure

> do it

## Problems

Problems: multiple tests corrections

Conservative: Bonferroni correction

at least one false discovery

$$\text{m tests at } 5\% \longrightarrow \frac{5\%}{m}$$

Reasonable: Benjamini-Hochberg procedure (1995)

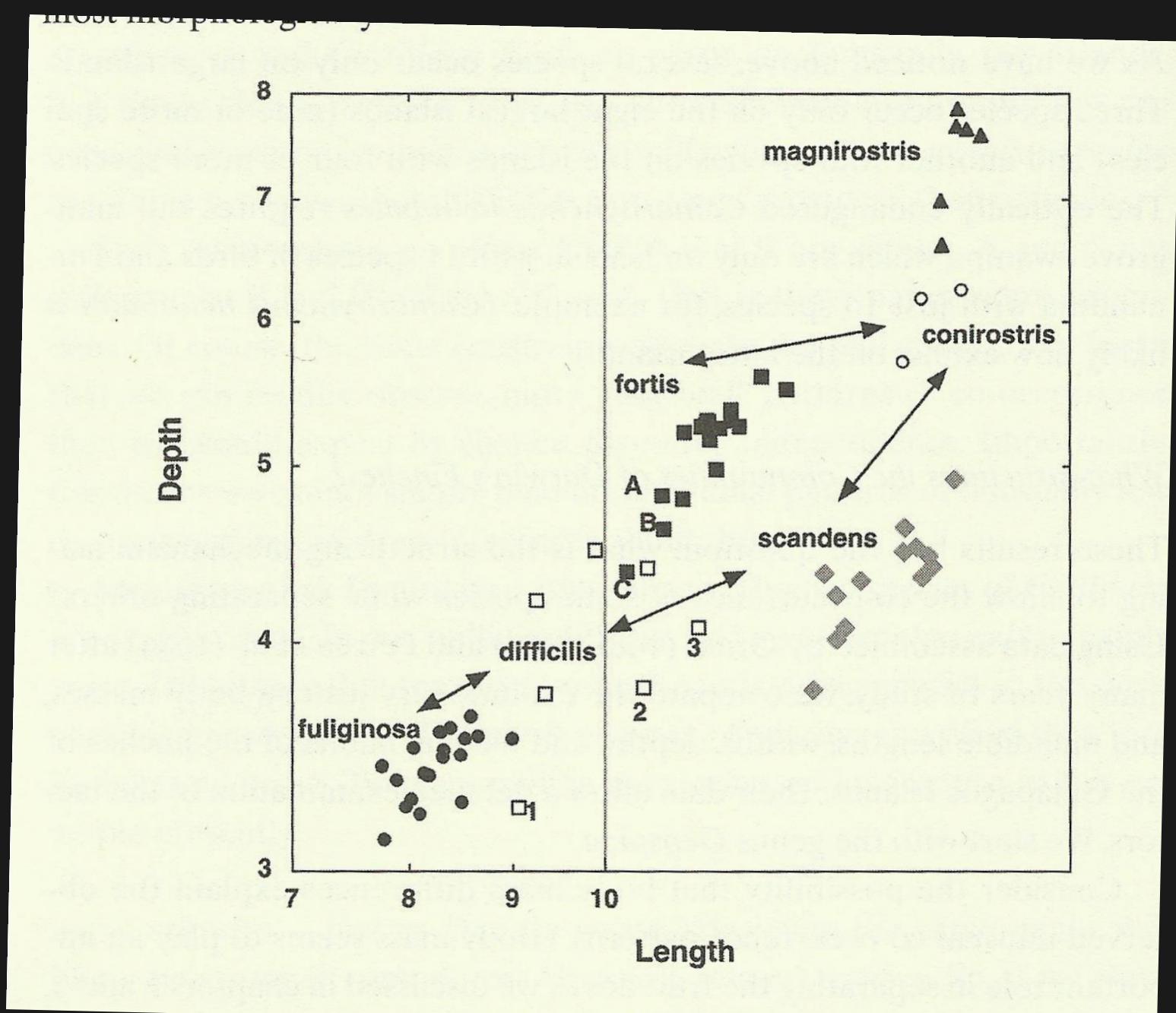
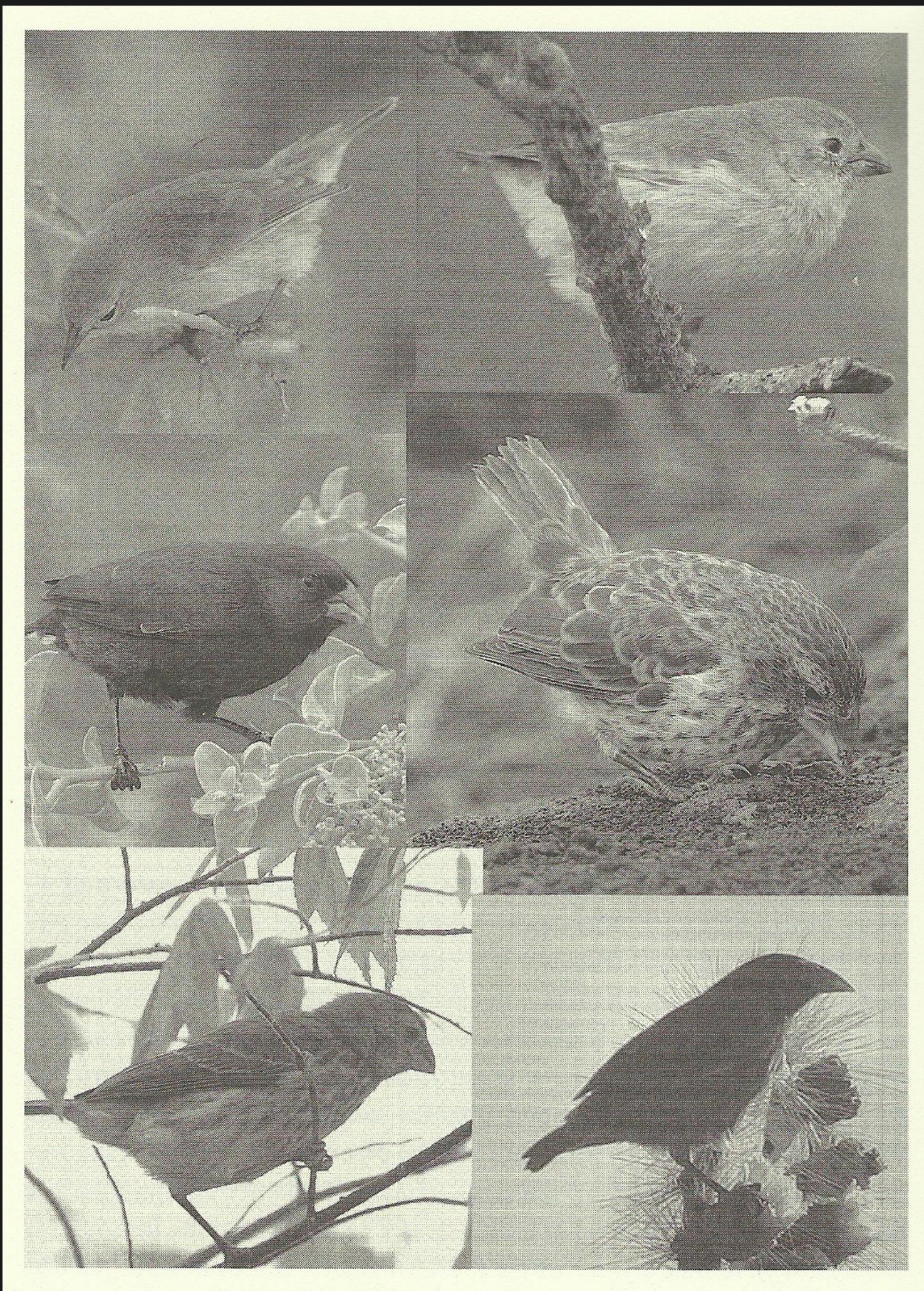
control rate of false discoveries

m tests

$$P_1, P_2, \dots, P_m$$

$$P_{(k)} \leq \frac{k}{m} \alpha$$

# Case study - the darwin finches: is it a true interaction ?



Positive associations ?

« do we think that species that coocurr frequently are mutualists ? They could be, but there is a simpler explanation. [...] Some islands [...] have habitats not found on smaller ones »

# The habitat problem

species that **share** habitat preferences will cooccur  
more frequently

species that **differ** in habitat preferences will cooccur  
less frequently

# Case study: Morueta-Holme, 2016

associations. For instance, *Carya aquatica* and *Taxodium distichum* are both species of Coastal Plain, strongly associated with large river backswamps that are periodically flooded. A similar habitat is preferred by *Nyssa aquatica* and *Planera aquatica*, although this species pair is found more up-stream than the first ones, in areas where flooding is less prolonged. *Quercus palustris*, *Q. bicolor* and *Carya laciniosa* also prefer swamp habitats, though more from interior flatlands on more calcareous soils. Habitat pref-

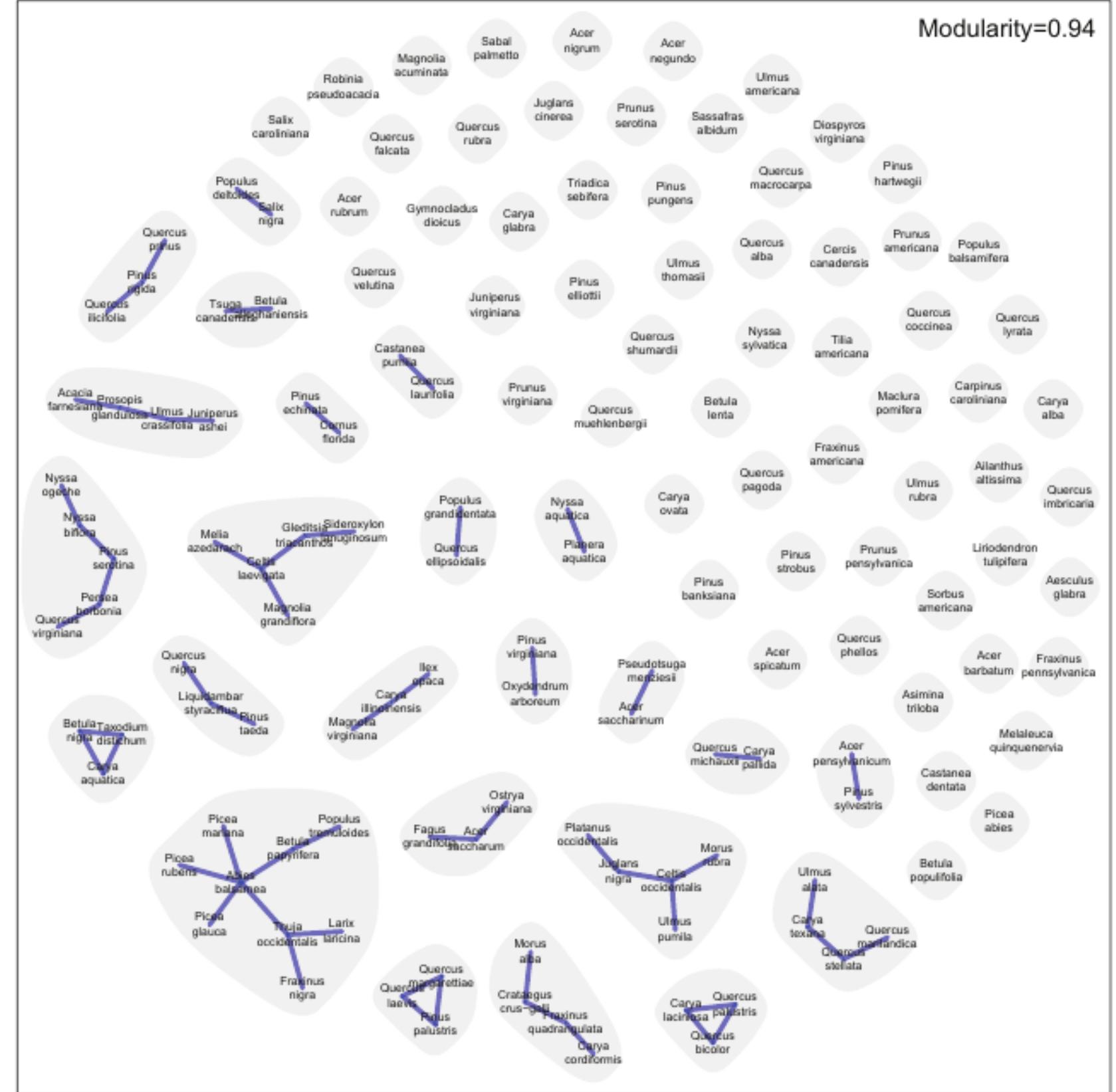


Figure 1. Empirical association network for North American trees. The network was constructed using the MaxEnt regional null model and a shrinkage inverse covariance estimator using 1000 null replicates and a false discovery rate of  $\alpha = 0.05$ . Gray envelopes indicate distinct modules. Positive associations are shown as blue lines; negative would be shown as red lines but none were found.

## A network approach for inferring species associations from co-occurrence data

# Case study: Sfenthourakis 2005

**Table 1** Plausible assumptions concerning causality of significant co-occurrence patterns, based on the simultaneous analysis of geographical and ecological data matrices at the species pair level. The geographical matrix contains species presence-absence data, and the ecological matrix contains binary data on species per ecological variable. Positive interaction means higher co-occurrence than expected by chance, negative means less co-occurrence than expected by chance, and none signifies the absence of deviating co-occurrence pattern. Here, allopatry is not necessarily meant in the strict sense, but denotes a tendency

Ecological matrix				
	<i>Interaction</i>	<b>Positive</b>	<b>Negative</b>	<b>None</b>
Geographic matrix	<b>Positive</b>	common ecology and history — neutral/symbiotic	common history	common history
	<b>Negative</b>	competition — ecological 'allopatry'	irrelevance — historical 'allopatry'	historical 'allopatry'
	<b>None</b>	common ecology and history — neutral	different ecology	random occurrence

*Global Ecology and Biogeography*, (*Global Ecol. Biogeogr.*) (2005) **15**, 39–49

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## Species co-occurrence: the case of congeneric species and a causal approach to patterns of species association

Spyros Sfenthourakis,\* Evangelos Tzanatos and Sinos Giokas

# Case study: Sfenthourakis 2005

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Ecological matrix			
Interaction	Positive	Negative	None
Geographic matrix	Positive	common ecology and history — neutral/symbiotic	common history
	Negative	competition — ecological 'allopatry'	irrelevance — historical 'allopatry'
	None	common ecology and history — neutral	historical 'allopatry'

**Table 6** Terrestrial isopod species pairs whose co-occurrence frequency deviates from that expected by chance in both matrices (species per sampling station and species per ecological variable), and the presumed causal explanation of the observed combination of interactions. The two pairs that are suspect for direct competition are given in boldface

Species pairs that deviate from random in both matrices	Type of interaction in each matrix		
	geographic	ecological	Causal explanation
<i>Ligidium ghigii</i> × <i>Paraschizidium album</i>	positive	negative	common history
<i>Cretoniscellus strinatii</i> × <i>Trichodillidium malickyi</i>	positive	positive	common ecology/history
<i>Bathytropa granulata</i> × <i>Rodoniscus anophthalmus</i>	positive	positive	common ecology/history
<i>Bathytropa granulata</i> × <i>Leptotrichus kossugi</i>	positive	positive	common ecology/history
<i>Bathytropa granulata</i> × <i>Porcellio laevis</i>	positive	positive	common ecology/history
<b><i>Bathytropa granulata</i> × <i>Proporcellio quadriseriatus</i></b>	negative	positive	<b>competition</b>
<i>Bathytropa granulata</i> × <i>Protracheoniscus babori</i>	positive	positive	common ecology/history
<i>Bathytropa granulata</i> × <i>Schizidium hybridum</i>	positive	positive	common ecology/history
<i>Chaetophiloscia elongata</i> × <i>Armadillidium marmoratum</i>	positive	positive	common ecology/history
<i>Chaetophiloscia elongata</i> × <i>Armadillidium vulgare</i>	positive	positive	common ecology/history
<b><i>Platyarthrus lindbergi</i> × <i>Armadillidium vulgare</i></b>	negative	positive	<b>competition</b>
<i>Porcellio wernerii</i> × <i>Armadillidium ameglioii</i>	positive	positive	common ecology/history
<i>Trachelipus aegaeus</i> × <i>Armadillidium vulgare</i>	positive	positive	common ecology/history
<i>Paraschizidium aegaeum</i> × <i>Schizidium tinum</i>	positive	positive	common ecology/history

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## Species co-occurrence: the case of congeneric species and a causal approach to patterns of species association

Spyros Sfenthourakis,\* Evangelos Tzanatos and Sinos Giokas

# The habitat problem

...how well do you describe your habitat ?

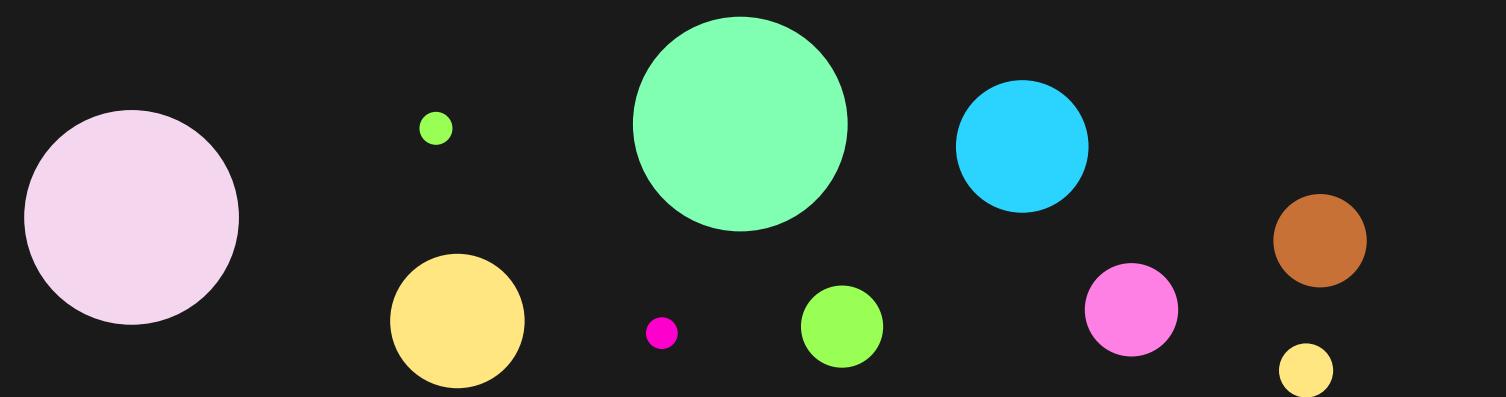
is there a hidden variable ?

The dispersal problem

The habitat problem

Species pool

Geographical scale of processes



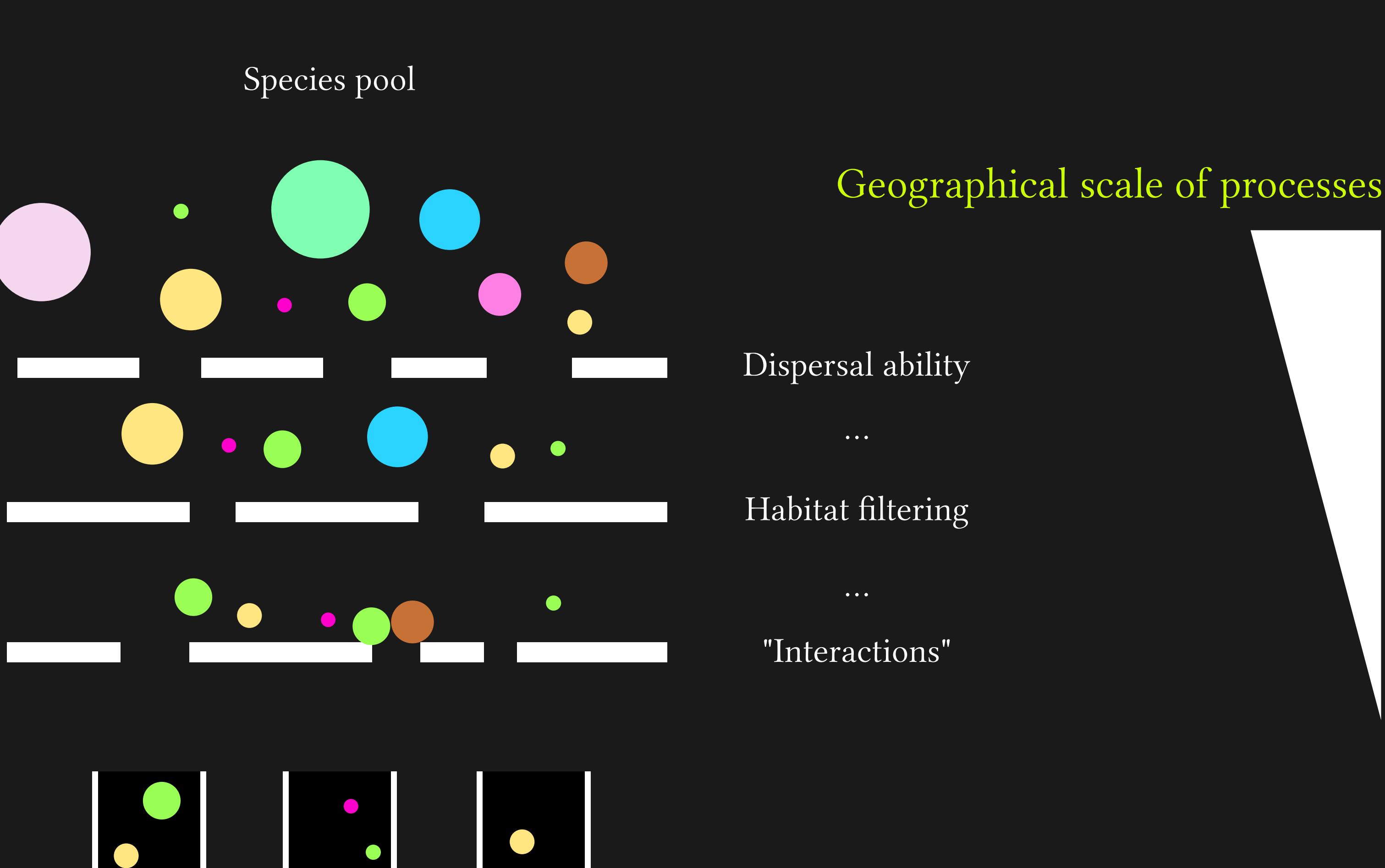
Dispersal ability

...

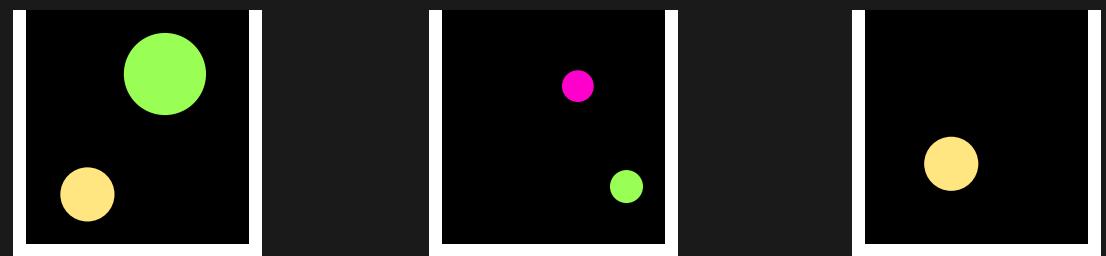
Habitat filtering

...

"Interactions"

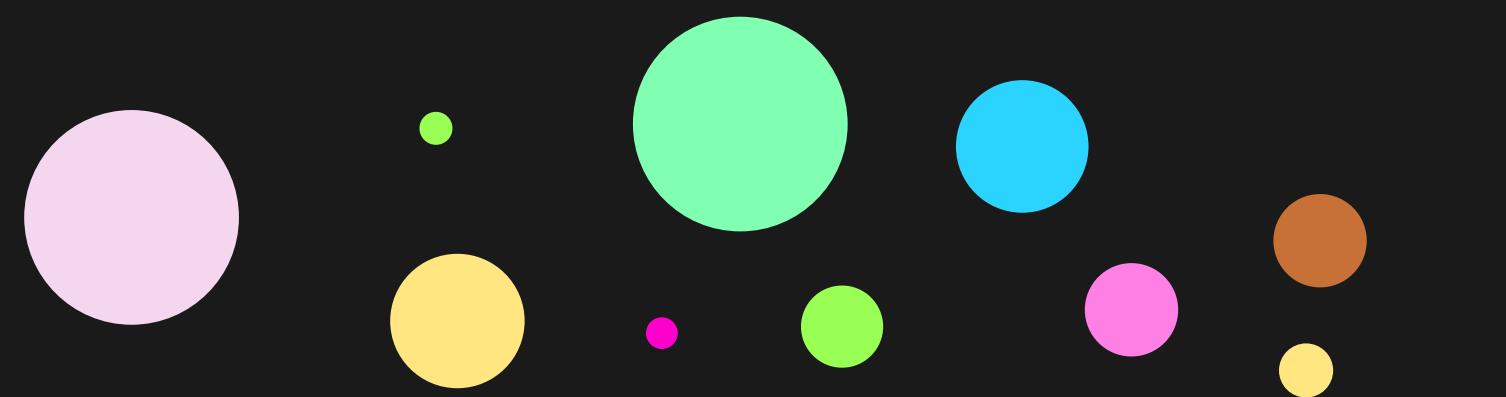


Communities



Species pool

Geographical scale of processes



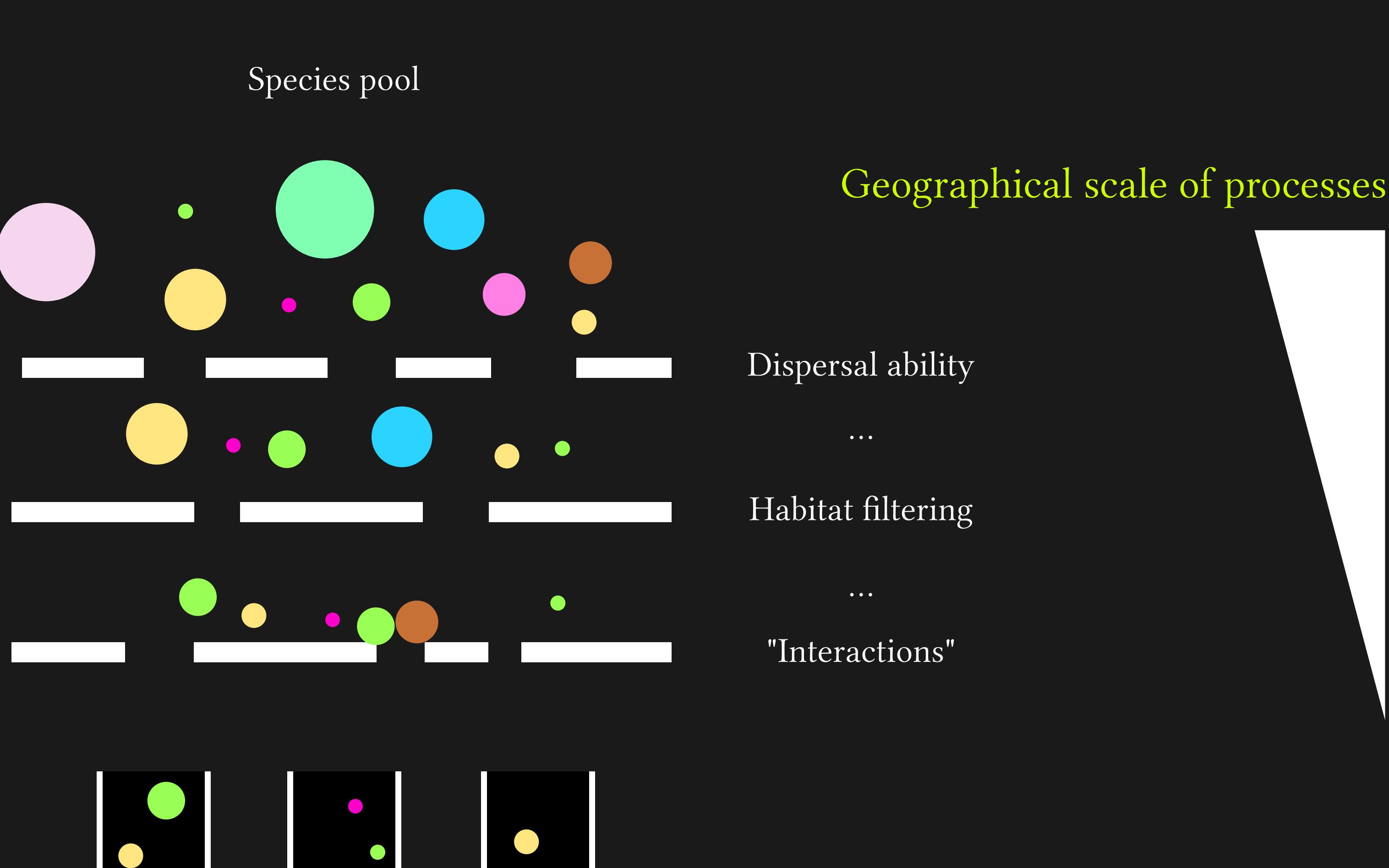
Dispersal ability

...

Habitat filtering

...

"Interactions"



Communities

JSDMs?

# JSDMs ?

### Understanding co-occurrence by modelling species simultaneously with a Joint Species Distribution Model (JSDM)

Laura J. Pollock<sup>1†</sup>, Reid Tingley<sup>1†</sup>, William K. Morris<sup>1</sup>, Nick Golding<sup>2</sup>, Robert B. O'Hara<sup>3</sup>, Kirsten M. Parris<sup>1</sup>, Peter A. Vesk<sup>1</sup> and Michael A. McCarthy<sup>1\*</sup>

## Frogs

shared responses to environmental variables. Facilitative interactions between frogs seem unlikely, but positive residual correlations between frog species could have been due to a shared response to an abiotic variable that was not considered in our model, such as the presence of fish (Hamer & Parris (2013).

# JSDMs ?

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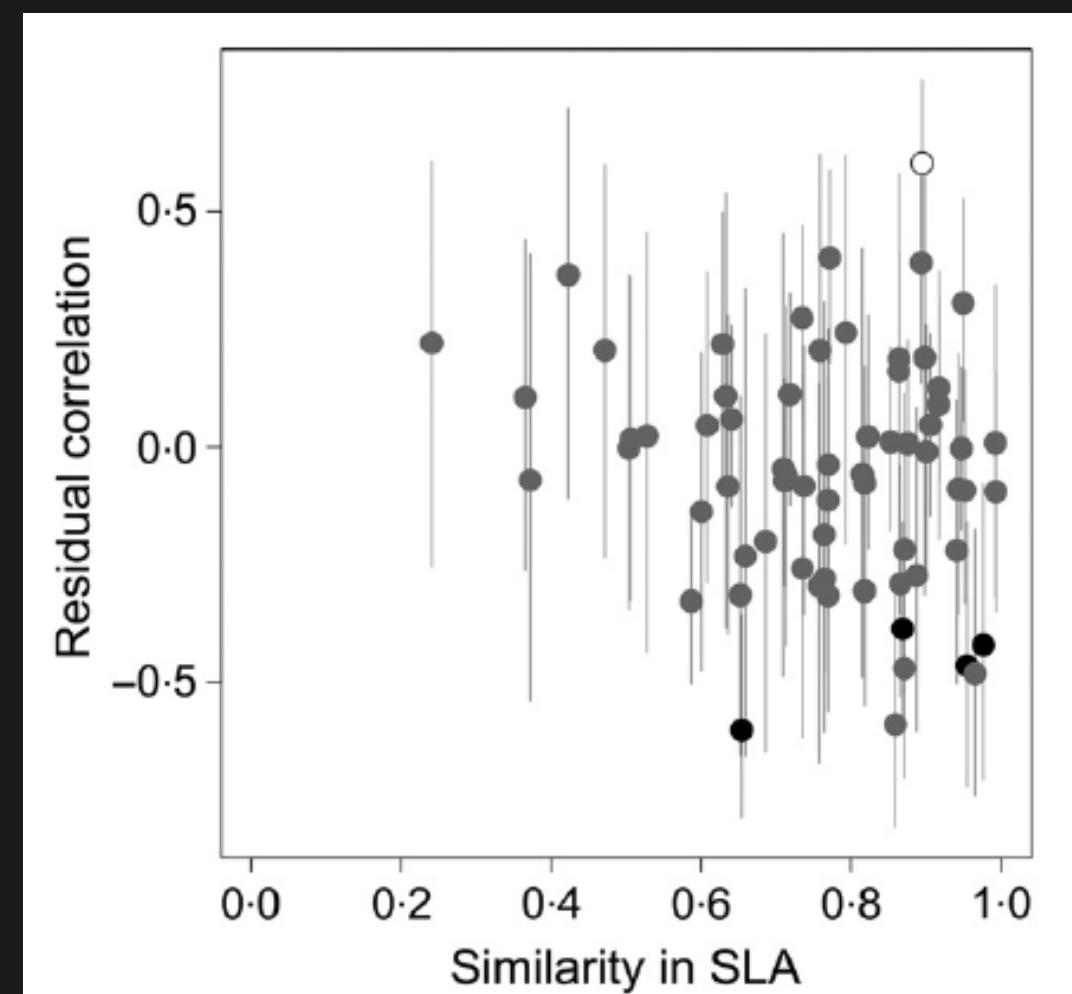
Laura J. Pollock<sup>1†</sup>, Reid Tingley<sup>1†</sup>, William K. Morris<sup>1</sup>, Nick Golding<sup>2</sup>, Robert B. O'Hara<sup>3</sup>, Kirsten M. Parris<sup>1</sup>, Peter A. Vesk<sup>1</sup> and Michael A. McCarthy<sup>1\*</sup>

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as Pryor's rule (Pryor 1953). A possible explanation for this pattern is that species from different subgenera are able to differentiate resource use, thereby alleviating competition (Austin, Cunningham & Wood 1983). Another potential explanation is that species from different subgenera are not able to interbreed (Ellis, Sedgley & Gardner 1991). If species

## Eucalyptus



# Fine-scale cooccurrence patterns

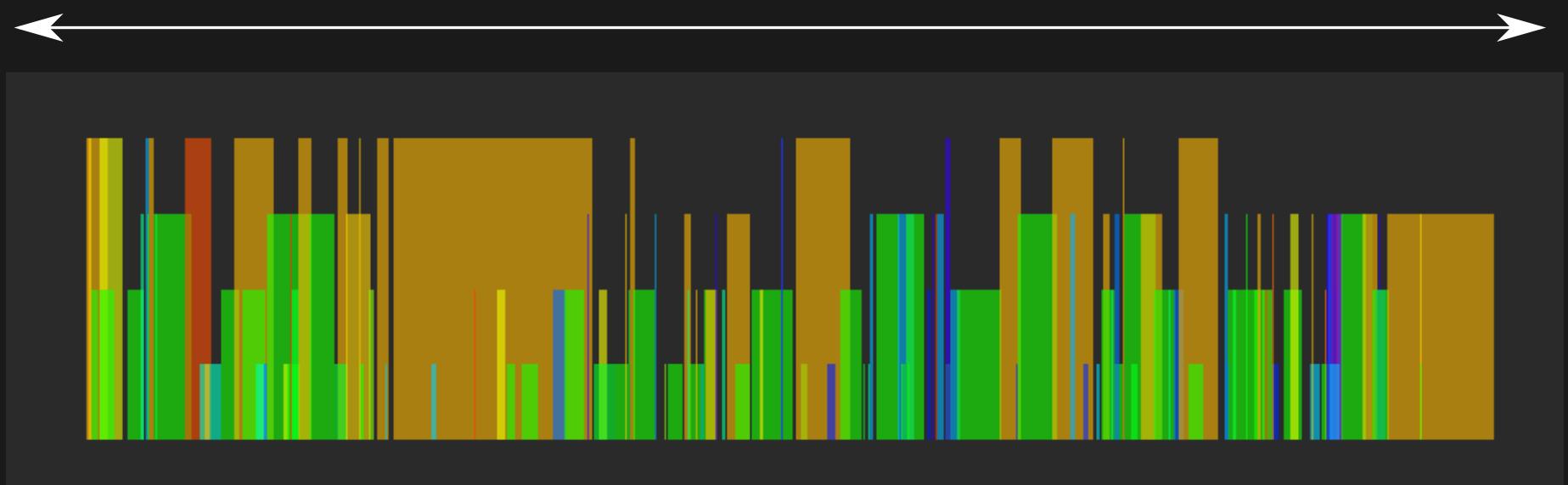
because you believe they compete/facilitate each other

Fine-scale cooccurrence patterns

because they really occur in the same place

because they really share the same habitat

5 meters



500 meters

OPEN ACCESS Freely available online

PLoS one

## Changes in Semi-Arid Plant Species Associations along a Livestock Grazing Gradient

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Grupo de Conservación de la biodiversidad, Departamento de Biodiversidad y Restauración, Instituto Pirenaico de Ecología, Consejo Superior de Investigaciones Científicas, Zaragoza, Spain

# Concluding remarks

Testing cooccurrence through permutations is...

easy to get wrong !

more so at large scales !

metric  
algorithm  
tests

...but also very studied  
(30+ years of research)

confounding factors in  
the assembly process

# Concluding remarks

Testing cooccurrence through permutations is...

easy to get wrong !

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the unit size of sampling (one "community")  
should not exceed much the length of the  
interaction (?)

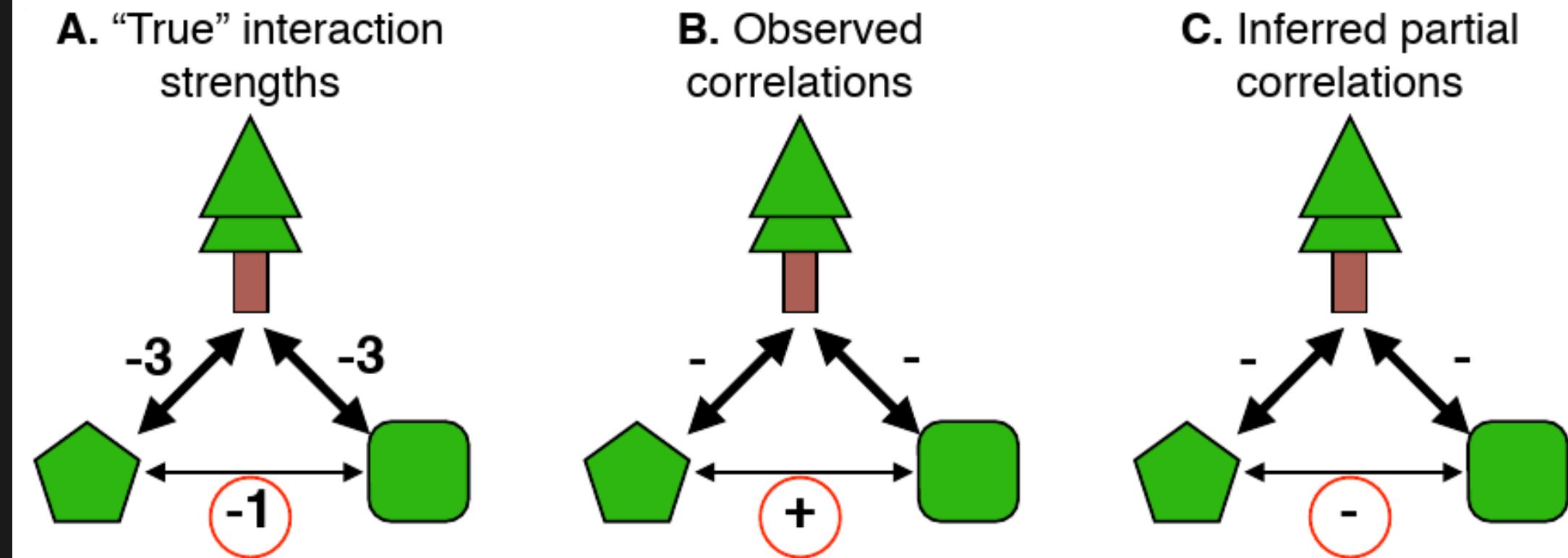


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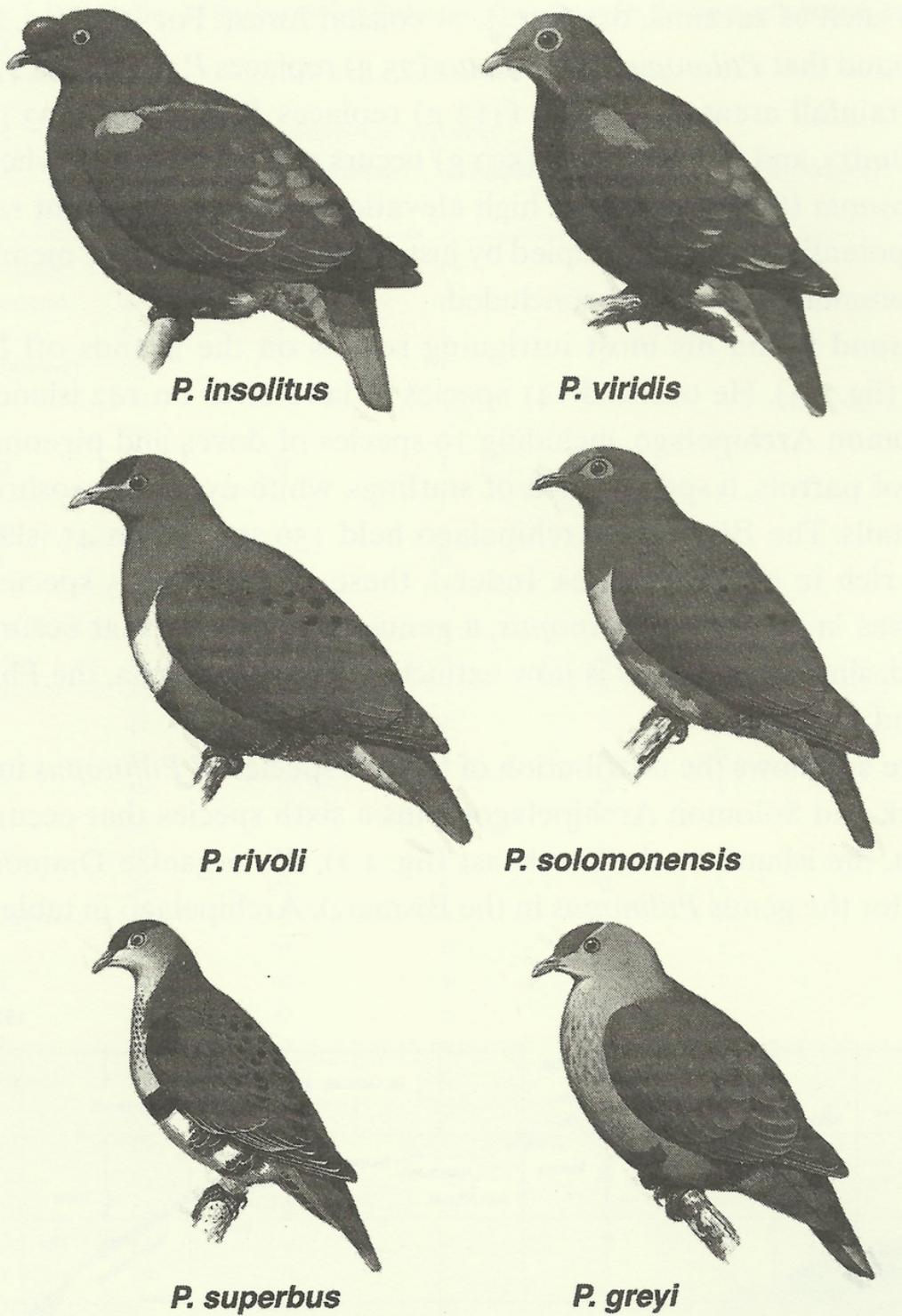
Title: Inferring species interactions from co-occurrence data with Markov networks

Author: David J. Harris: Department of Wildlife Ecology and Conservation, University of Florida, 110 Newins-Ziegler Hall PO Box 110430, Gainesville, FL 32611

## Markov networks



...towards more efficient methods ?

FIG. 2.2. The species of *Ptilinopus* fruit dove.

Does competition shape communities at large scale ?  
^  
or mutualism

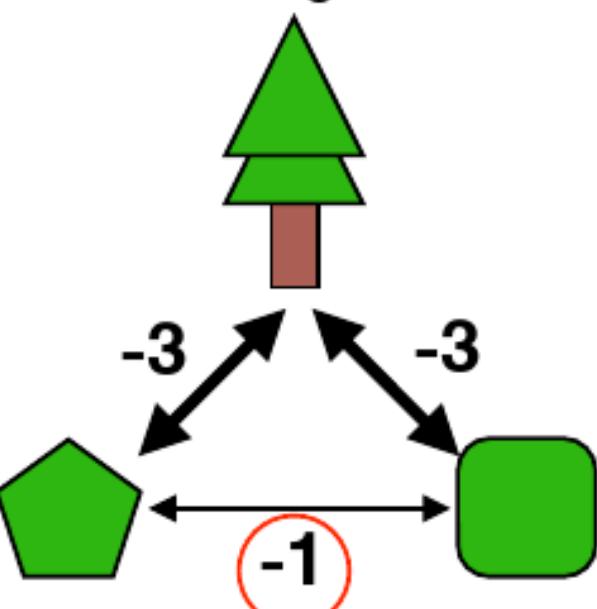
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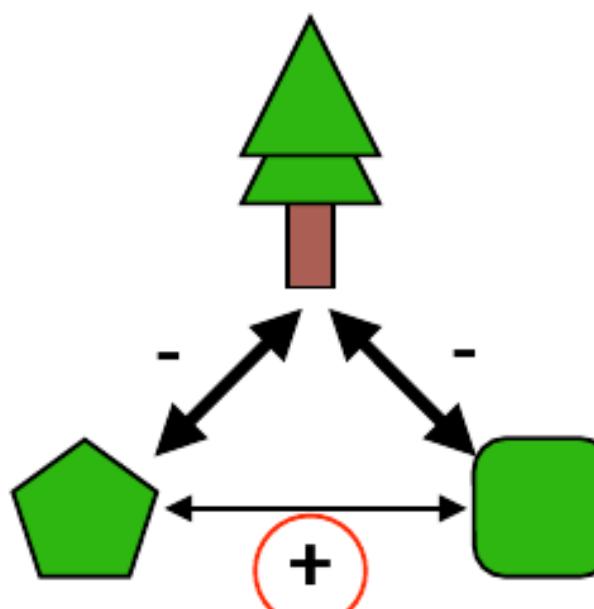
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## Markov networks

A. “True” interaction strengths



B. Observed correlations



C. Inferred partial correlations

