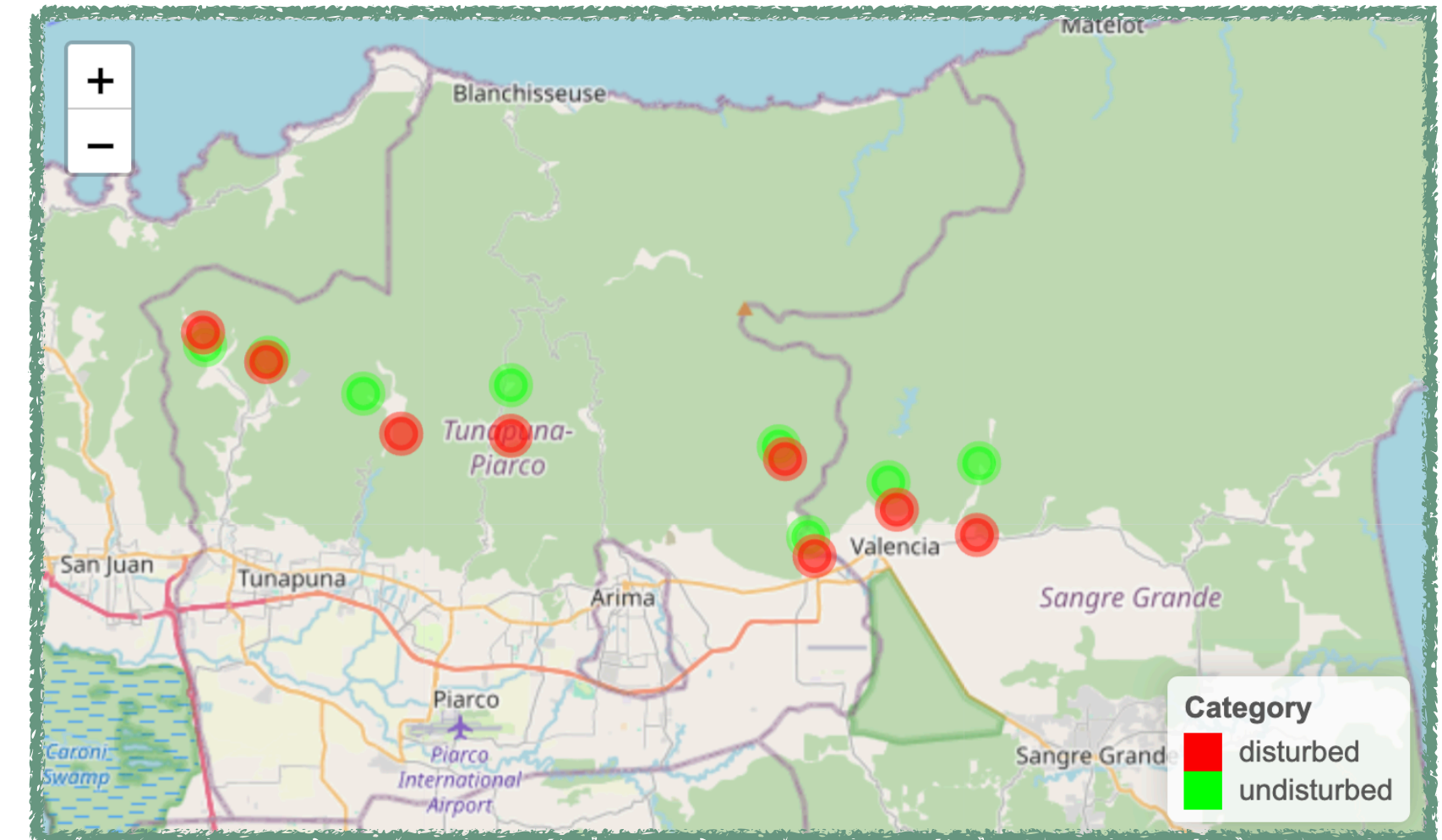
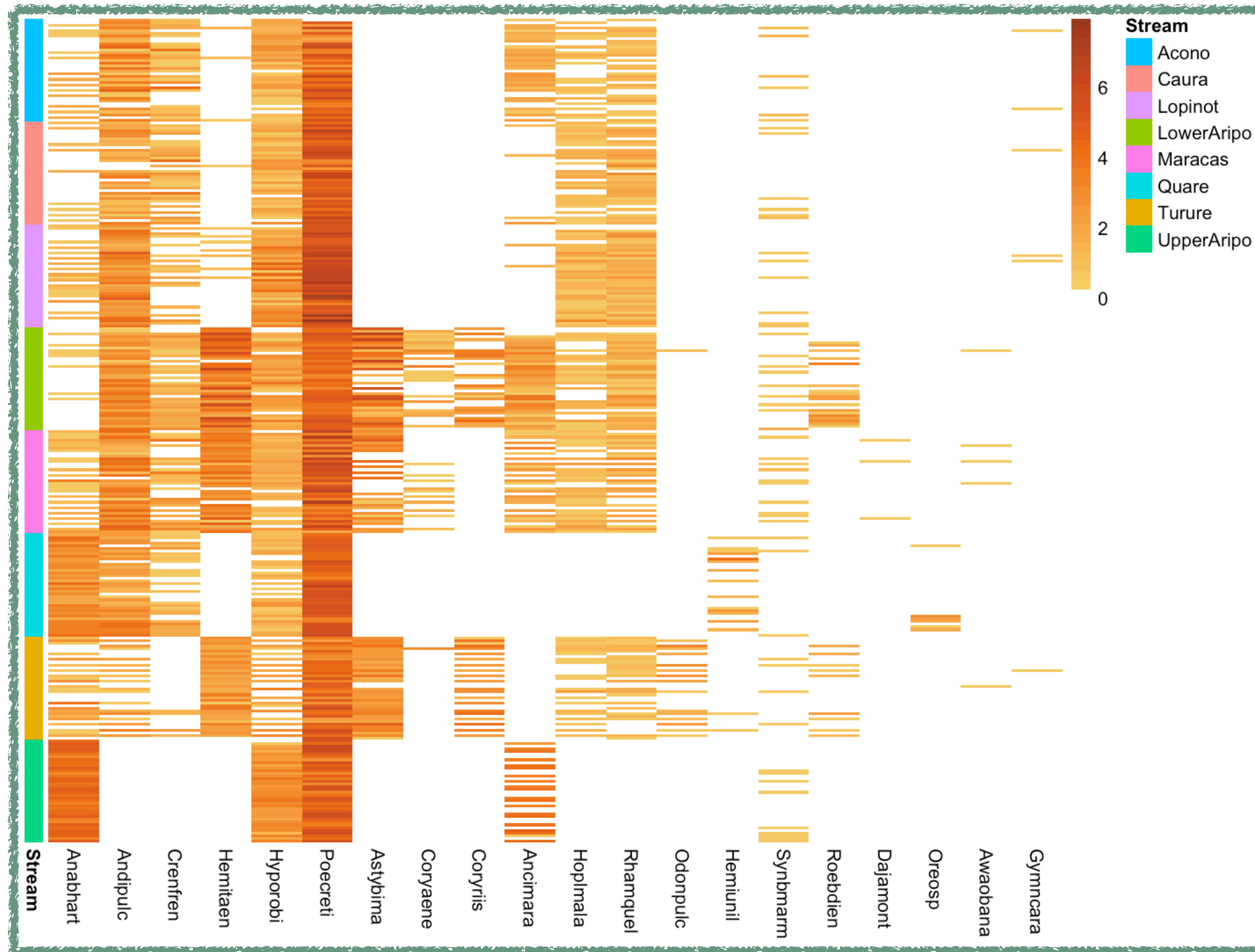


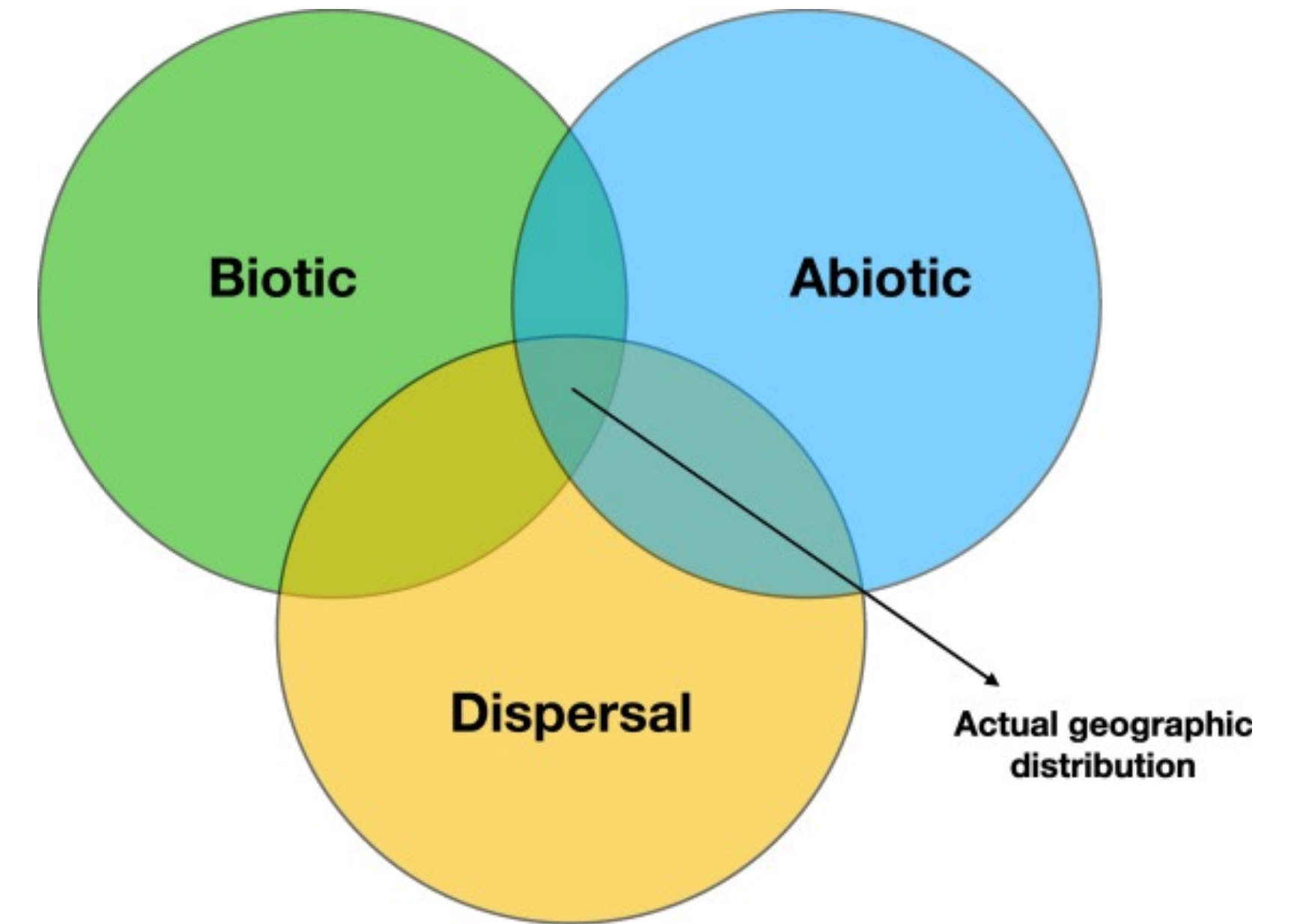
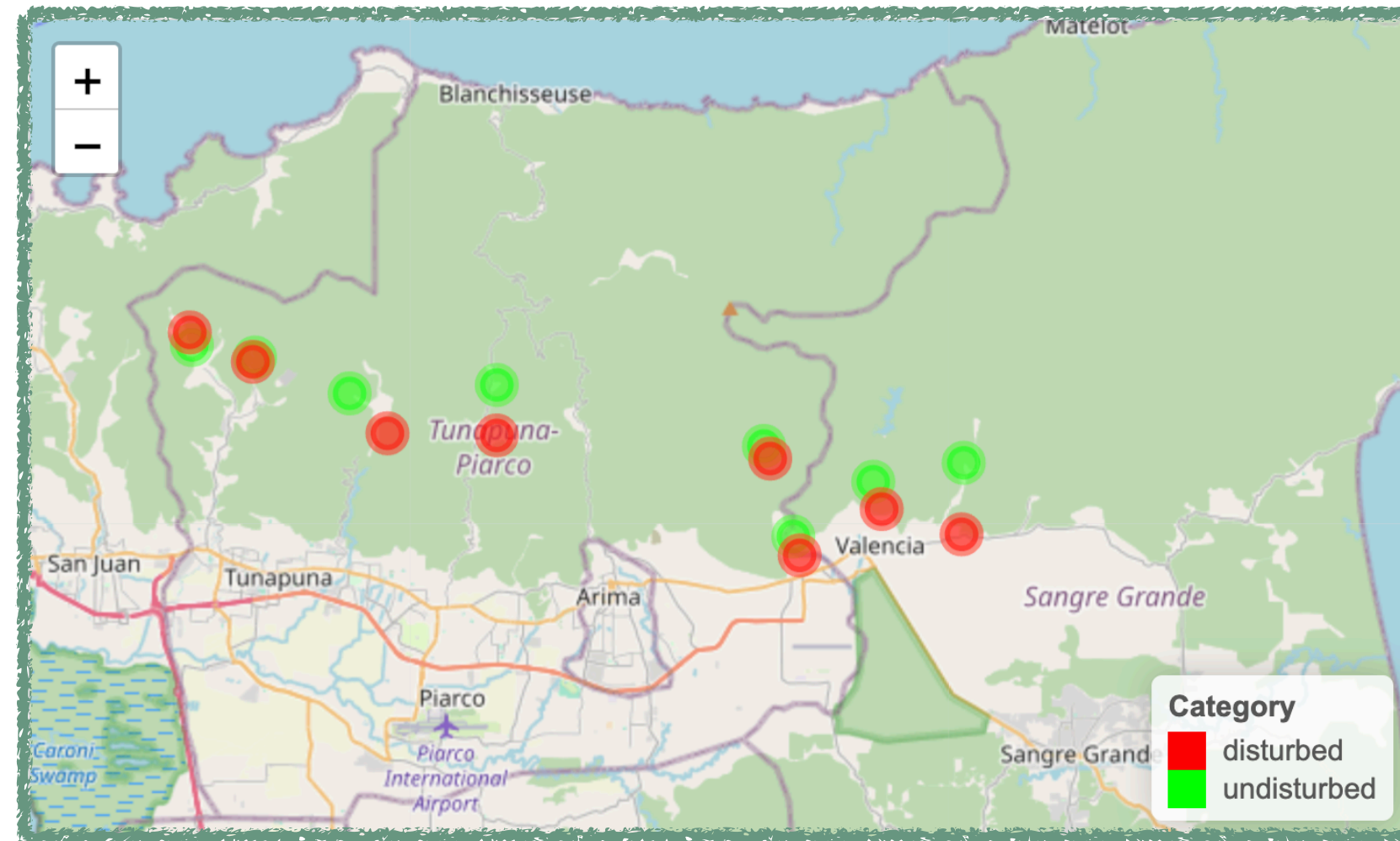
**A simulation study to better understand the correspondence between biotic interactions and JSDM-inferred associations in species distribution data.**

*Questioning the relationships between ecological processes and statistical patterns in species distribution data.*

# 1. Species distribution data

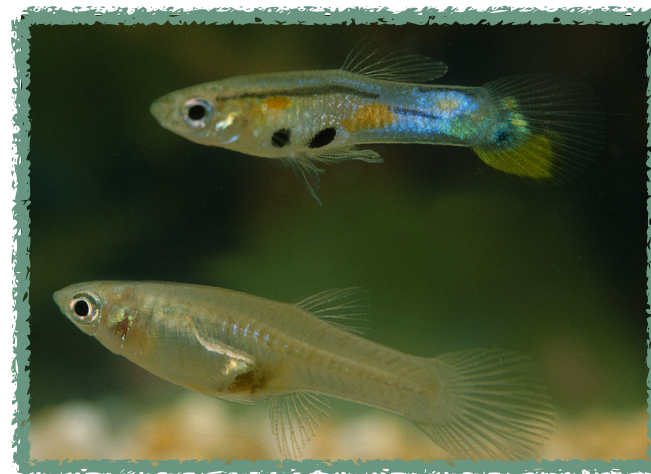


# 1. Species distribution data



Trends in Ecology & Evolution

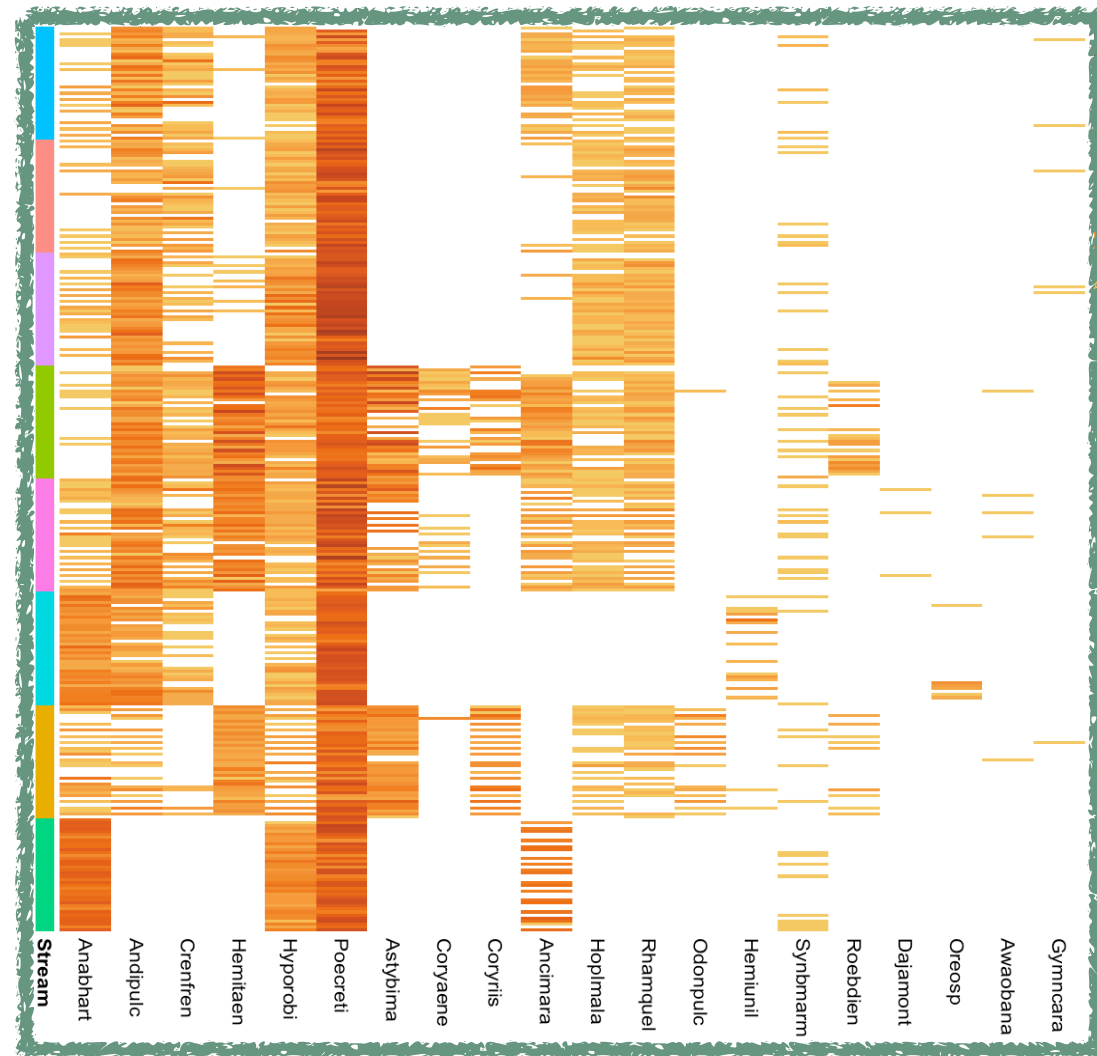
Poggiato et al. [2021]



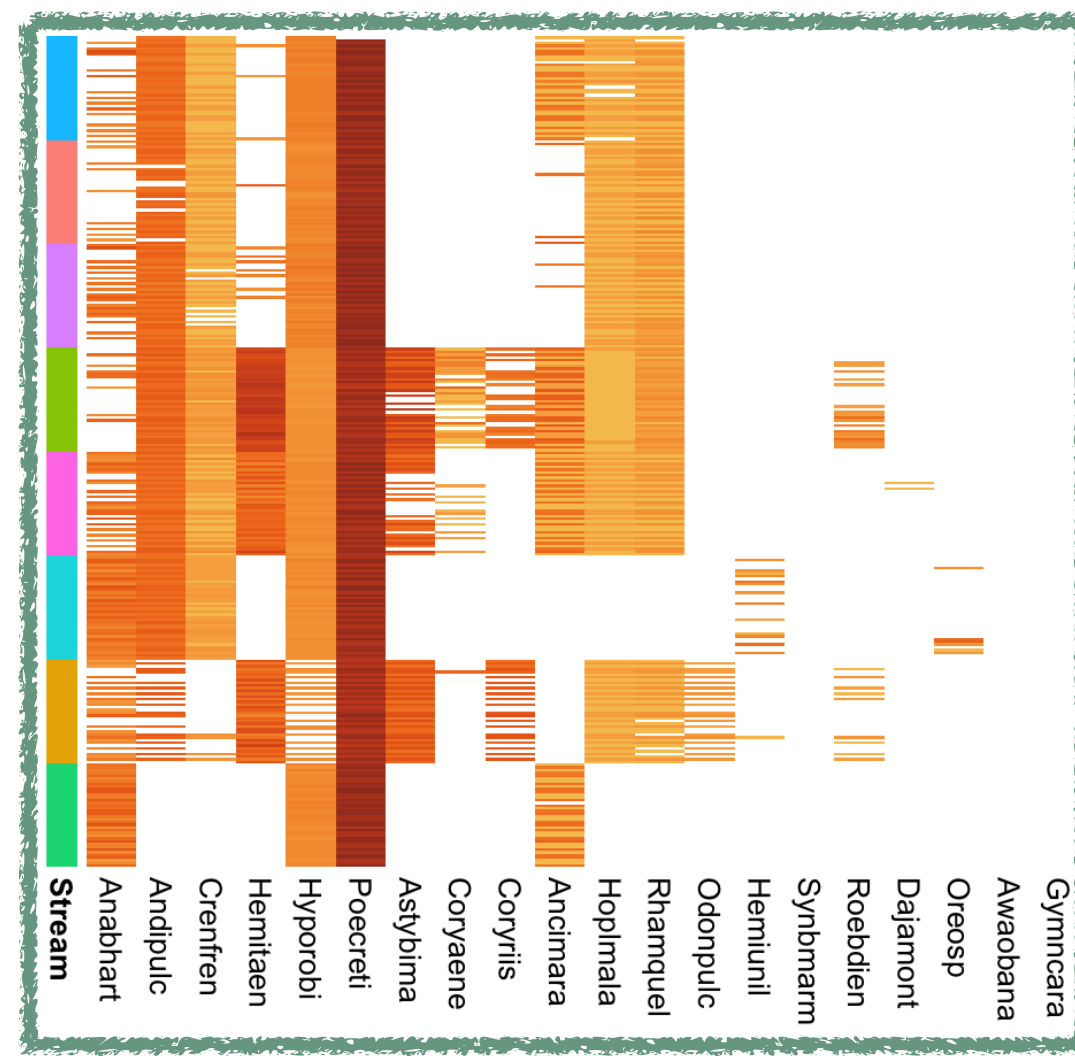
## 2. Patterns, processes and mechanisms

- Anand (1994) : *"patterns are what we perceive, processes describe how these patterns come about, and the mechanisms provide explanations as to why these patterns occur"*
- We often don't have a direct access to ecological processes / mechanisms.
- Old idea in ecology: identify patterns to understand processes.
  - Null models, SDMs, JSDBMs...

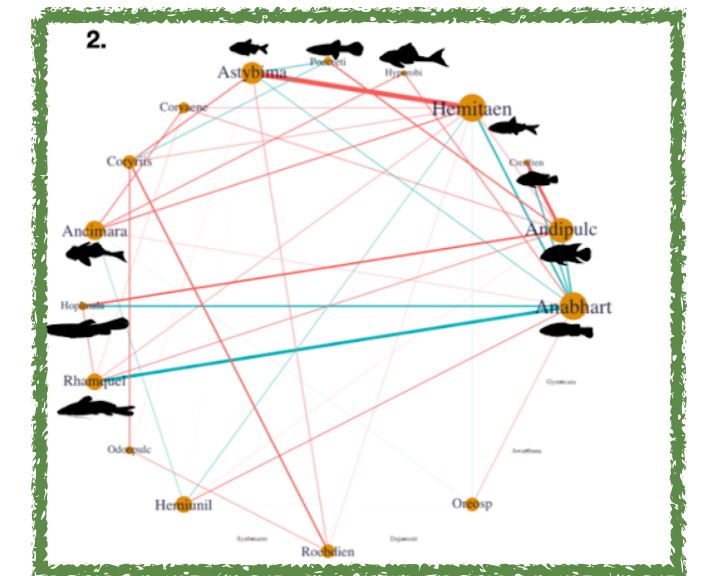
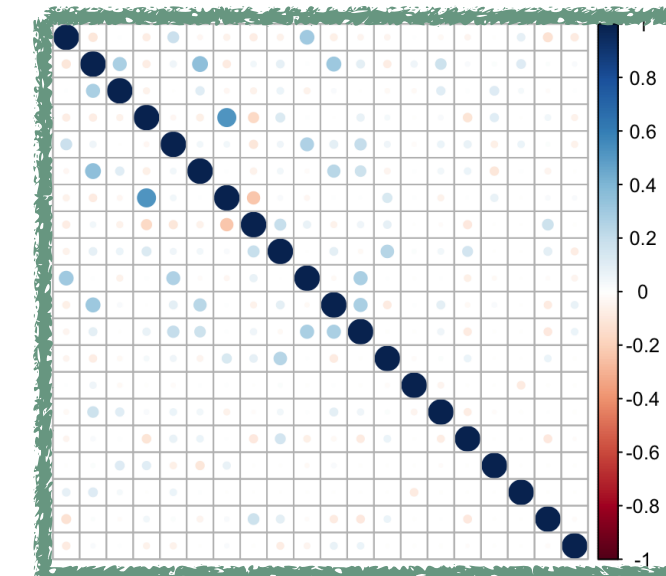
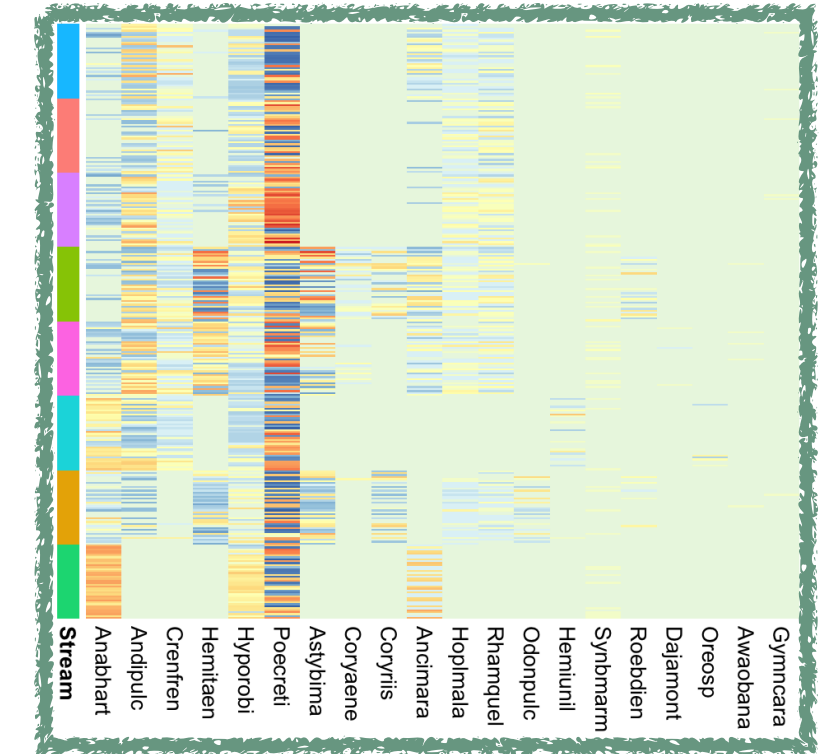
# 3. Joint Species Distribution Models (JSDMs)



Explain what you can  
with abiotic  
covariates



Look at the residuals



# 3. Joint Species Distribution Models (JSDMs)

Graphical models = possible JSDMs

Ex: PLN-network [Chiquet, Robin and Mariadassou, 2019]

$$Z_i \sim \mathcal{N}(0, \Omega^{-1})$$

$$Y_{ij} | Z_{ij} \sim \mathcal{P}(\exp(X_i B_j + Z_{ij}))$$

# 3. Joint Species Distribution Models (JSDMs)

Graphical models = possible JSDMs

Ex: PLN-network [Chiquet, Robin and Mariadassou, 2019]

$$Z_i \sim \mathcal{N}(0, \Omega^{-1})$$

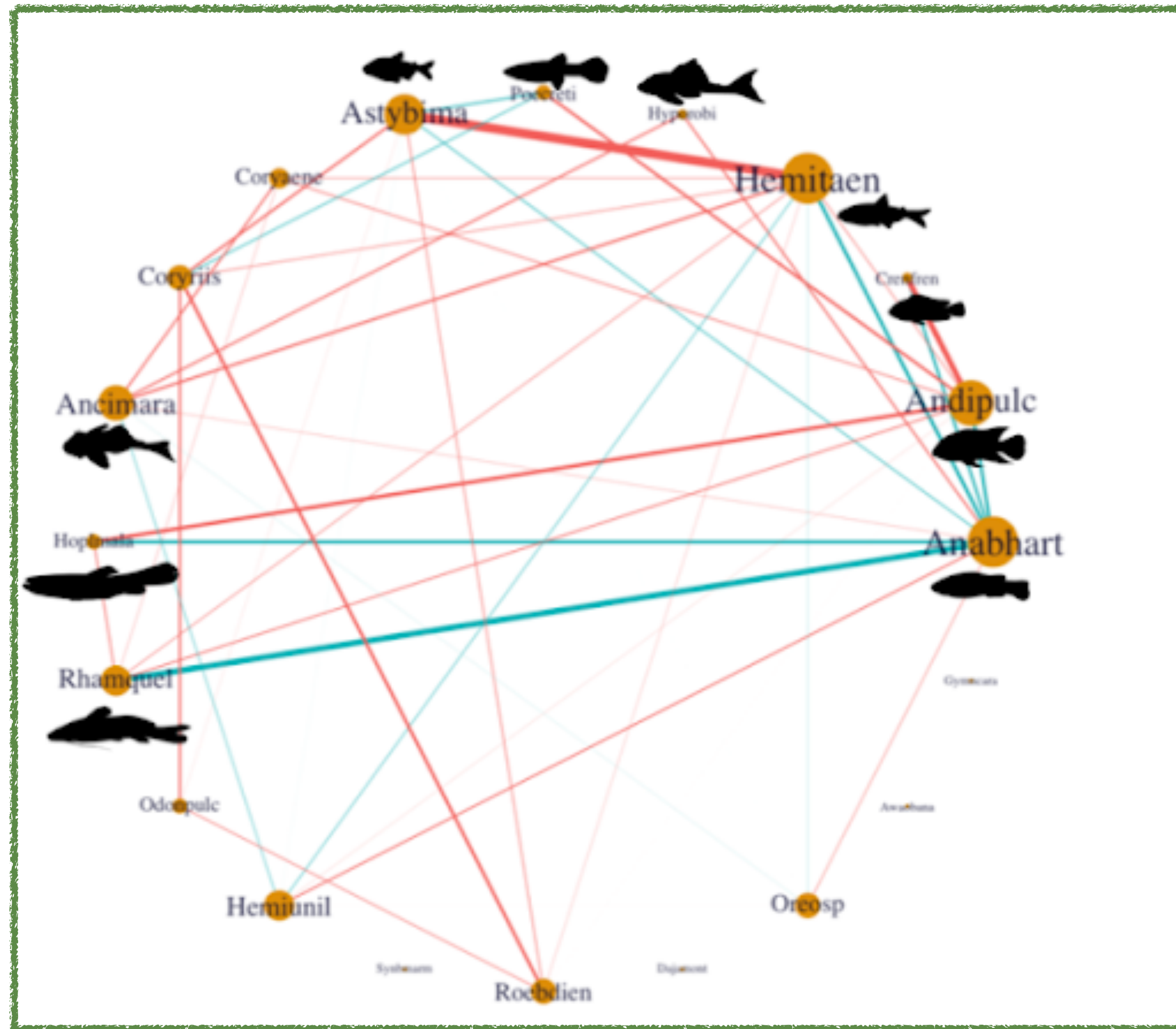
Residual partial correlations

Count data  $Y_{ij}$

$$Z_{ij} \sim \mathcal{P}(\exp(X_i B_j + Z_{ij}))$$

Measured abiotic effects

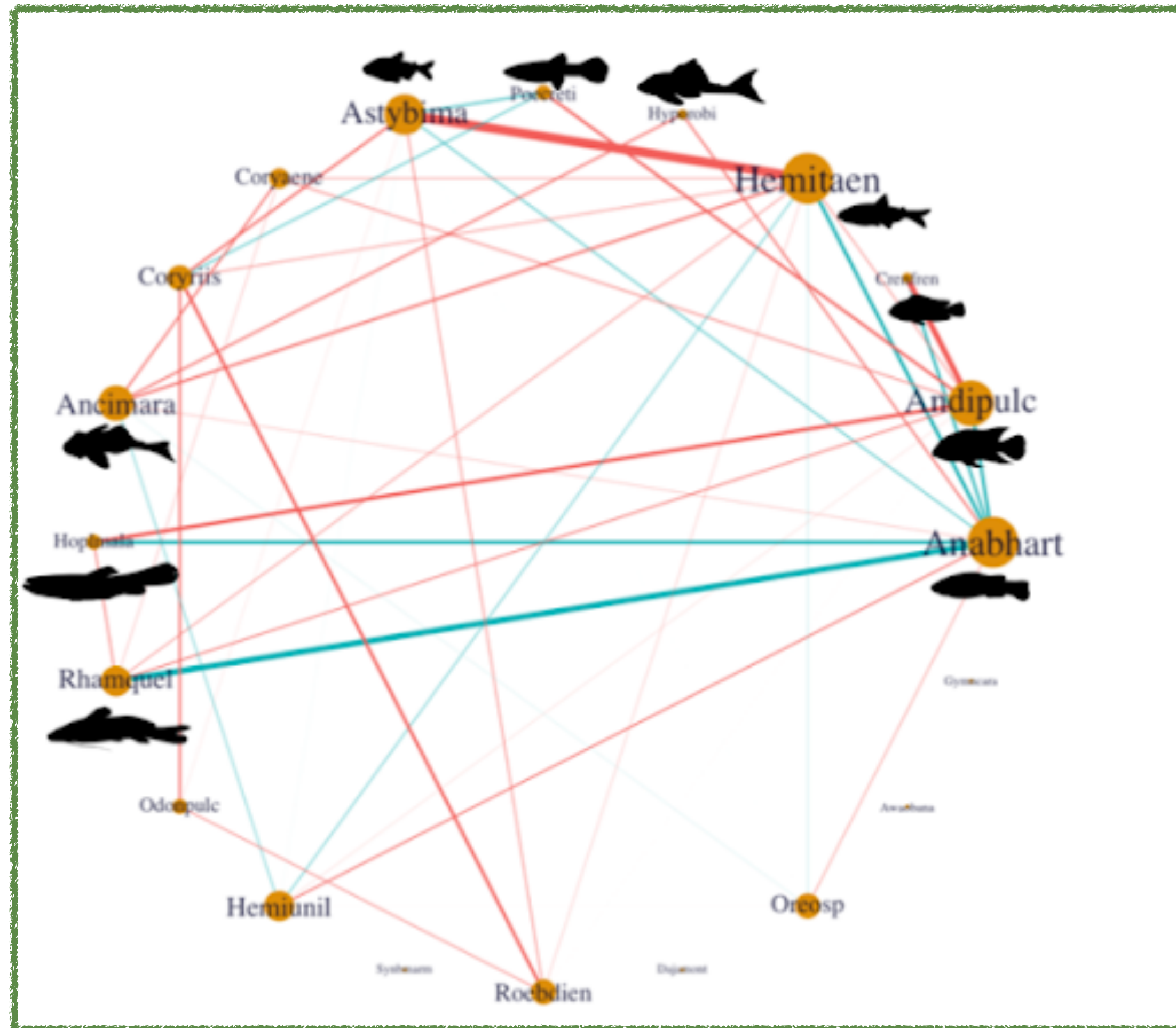
# 3. Joint Species Distribution Models (JSDMs)



Possible interpretations :

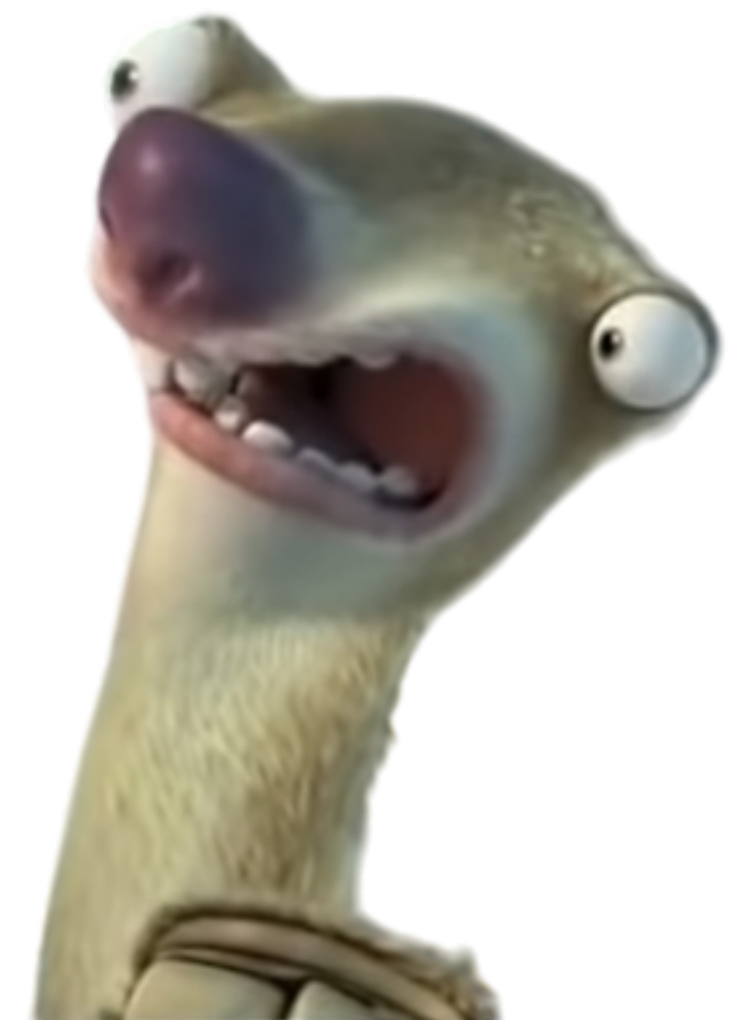
- Model misspecification.
- Missing abiotic covariates.
- Missing species.
- Biotic interactions.

# 3. Joint Species Distribution Models (JSDMs)



Possible interpretations :

- Model misspecification.
- Missing abiotic covariates.
- Missing species.
- ***Biotic interactions ?***



***What if we knew exactly what interactions  
were involved in the distribution ?***

# 4. A “virtual ecologist” approach

## General method

Virtual Ecologist idea (Zurell et al., 2010) :

An “approach where simulated data and observer models are used to mimic real species and how they are ‘virtually’ observed. ”

# 4. A “virtual ecologist” approach

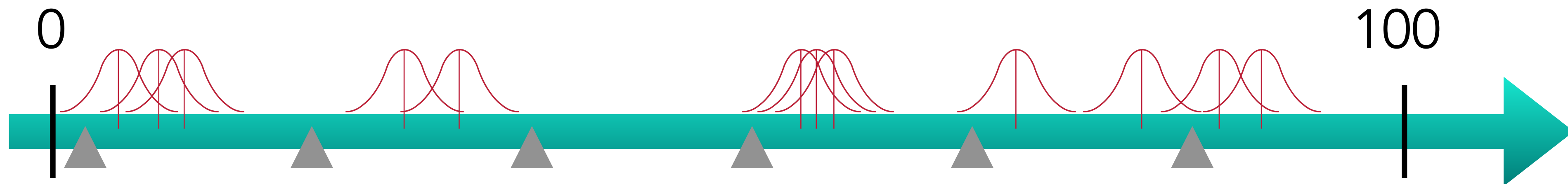
## General method

1. Simulate species distribution models with a biotic interactions based model.
2. Run JSDMs over the simulated data.
3. What level of correspondence between biotic interactions (*processes*) and statistical correlation (*patterns*).
- 3'. Bonus: how is this influenced by: the interactions type / intensities / densities / symmetry...

## 4. A “virtual ecologist” approach

### The VirtualCom model (Münkemüller and Gallien, 2015)

- Triple filtering: environmental, biotic interactions and reproduction.
- 1D environmental gradient. Sites and species niches over this gradient.



# 4. A “virtual ecologist” approach

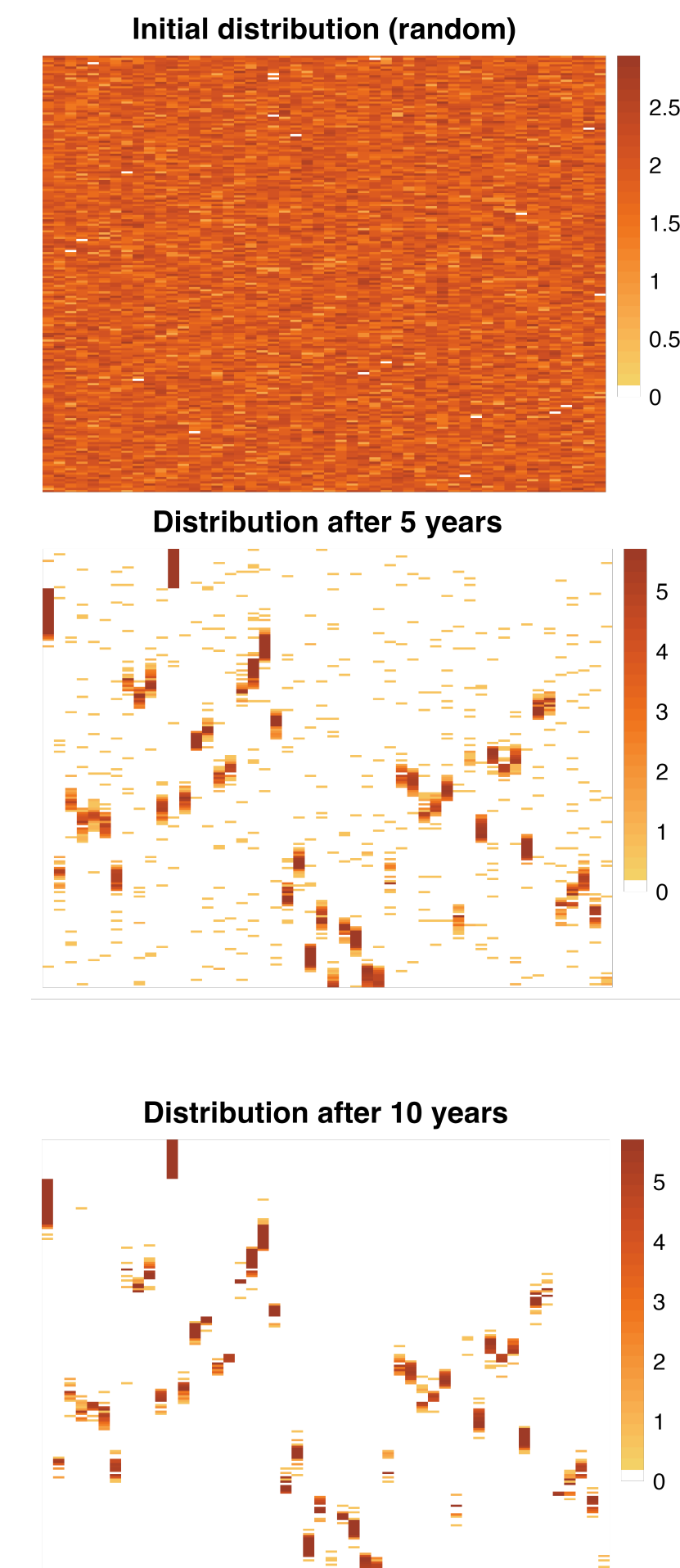
## The VirtualCom model (Münkemüller and Gallien, 2015)

For each site  $i$ :

1. Random initialization:  $K$  individuals are drawn from the species pool
2.  $K$  individuals are successively drawn and replaced with an individual from species  $j$  with probability

$$P_{ij}$$

3. Step 2 is repeated over  $n$  years.



# 4. A “virtual ecologist” approach

## The VirtualCom model (Münkemüller and Gallien, 2015)

Filtering allowed by  $p$ , resulting from the normalization of:

$$W_{ij} = \exp(\beta_{\text{env}} \log(p_{\text{env},i,j}) + \beta_{\text{abun}} \log(p_{\text{abun},i,j}) + \beta_{\text{comp}} \log(p_{\text{comp},i,j}))$$

**Environmental filter**  $p_{\text{env}} = f(E_i; \text{opt}_j, \text{sd}_j) / f(E_i; E_i, \text{sd}_j)$ , with  $f$  the density of a Gaussian distribution

**Reproduction filter**  $p_{\text{abun},i,j} = \frac{N_{ij}}{K}$

# 4. A “virtual ecologist” approach

## The VirtualCom model (Münkemüller and Gallien, 2015)

Filtering allowed by  $p$ , resulting from the normalization of:

Competition / Facilitation filter 
$$P_{comp} = 1 - \frac{1}{K} \sum_k a_{jk} N_{ik}$$

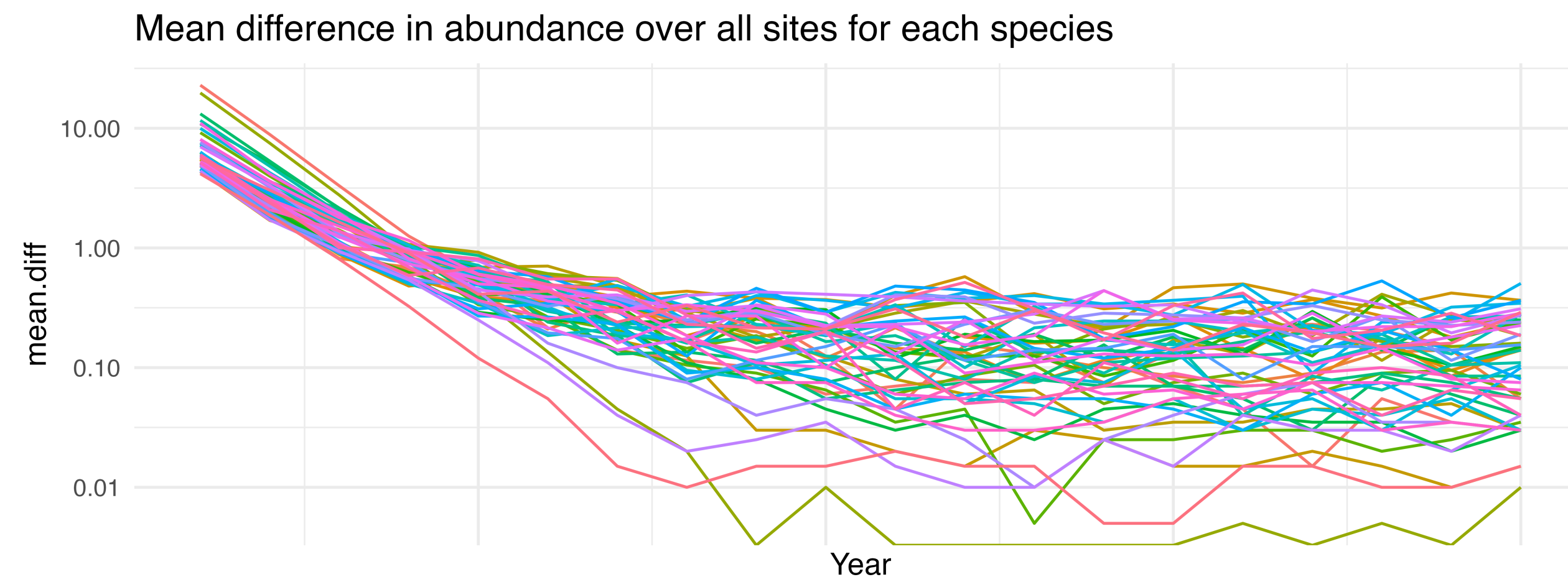
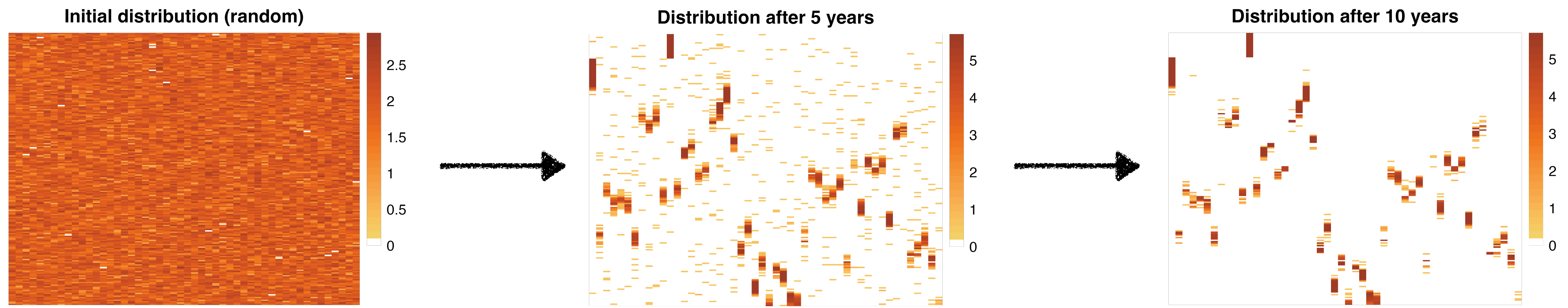
$a_{jk} > 0$  if  $j$  and  $k$  in competition,  $a_{jk} < 0$  if  $j$  and  $k$  facilitate each other's presence.

To generate the model we don't use the original approach but instead fix:

- A niche distance threshold beyond which two species can't interact.
- Competition and facilitation densities among the species whose niches are closer than this threshold.

# 4. A “virtual ecologist” approach

## The VirtualCom model (Münkemüller and Gallien, 2015)



# 4. A “virtual ecologist” approach

## Simulation protocol

- Simulate VirtualCom data with 50 species, 200 sites, niche distance threshold of 10.
- Changing interaction intensities & densities.
- We also test asymmetric interactions and sampling effect.

# 4. A “virtual ecologist” approach

## Simulation protocol

- We test the adequation between the intuitive effect of interactions and what they really look like with a *relative abundance index*:

$$\text{RAI}_{jk} = \text{avg} \left( \{ \delta_{i,j} \mid 1 \leq i \leq n, y_{ij} > 0, y_{ik} > 0 \} \right)$$

$$\text{with } \delta_{i,j} = y_{ij} - \bar{y}_j,$$

- We run several penalized methods evaluated in terms of AUC
  - True positive = competition matching negative residual correlations or facilitation matching positive residual correlations.

# 4. A “virtual ecologist” approach

## Simulation protocol

We test several JSDBMs including a Poisson Log-normal model with:

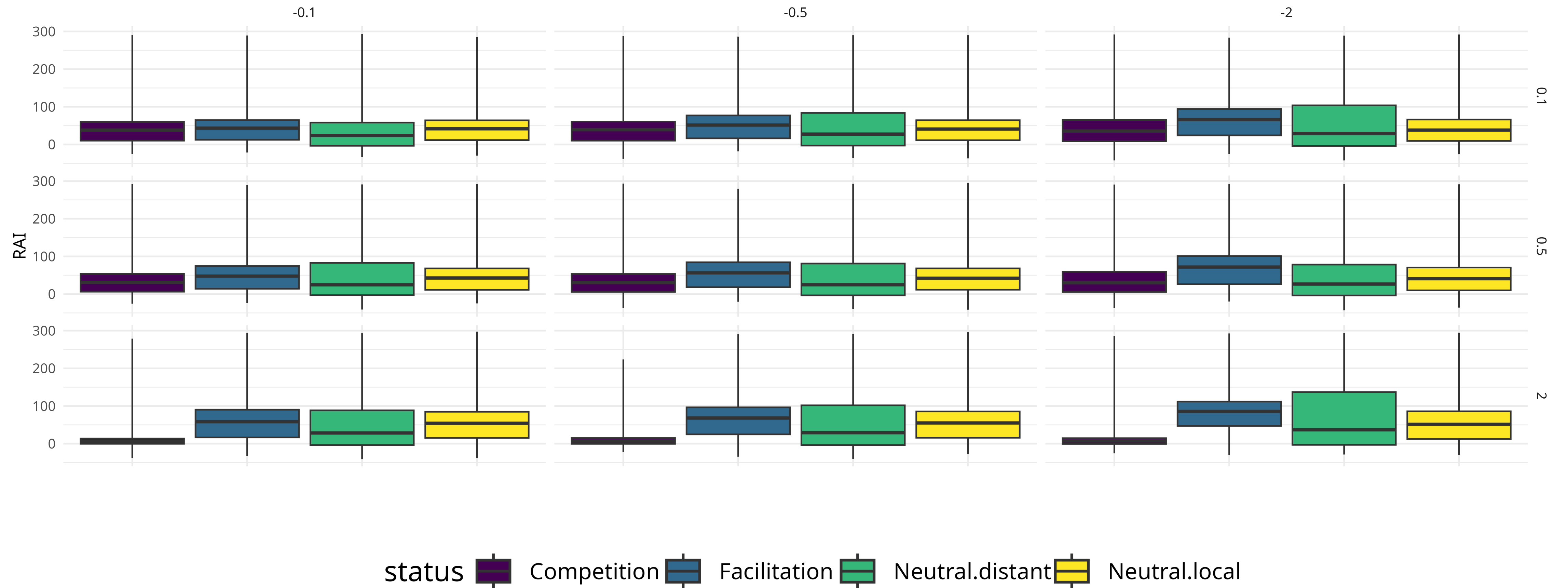
- No covariate.
- The environmental gradient as a continuous covariate.
- A discretized version of the environmental gradient.
  - Goal = account for the niche modelling in the covariate effect.

# 4. A "virtual ecologist" approach

## Results: competition

RAI (competition.intensity ~ facilitation.intensity)

Row = competition intensity | Col = facilitation intensity

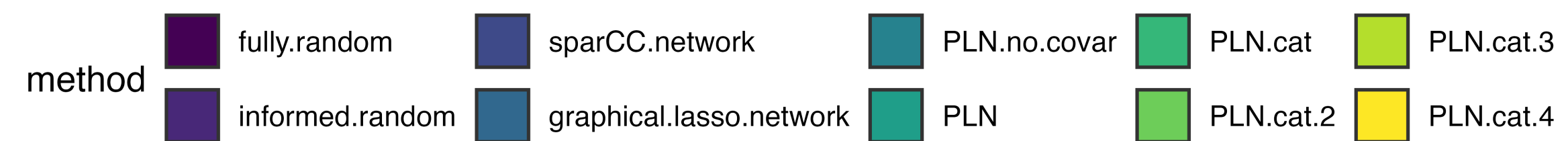
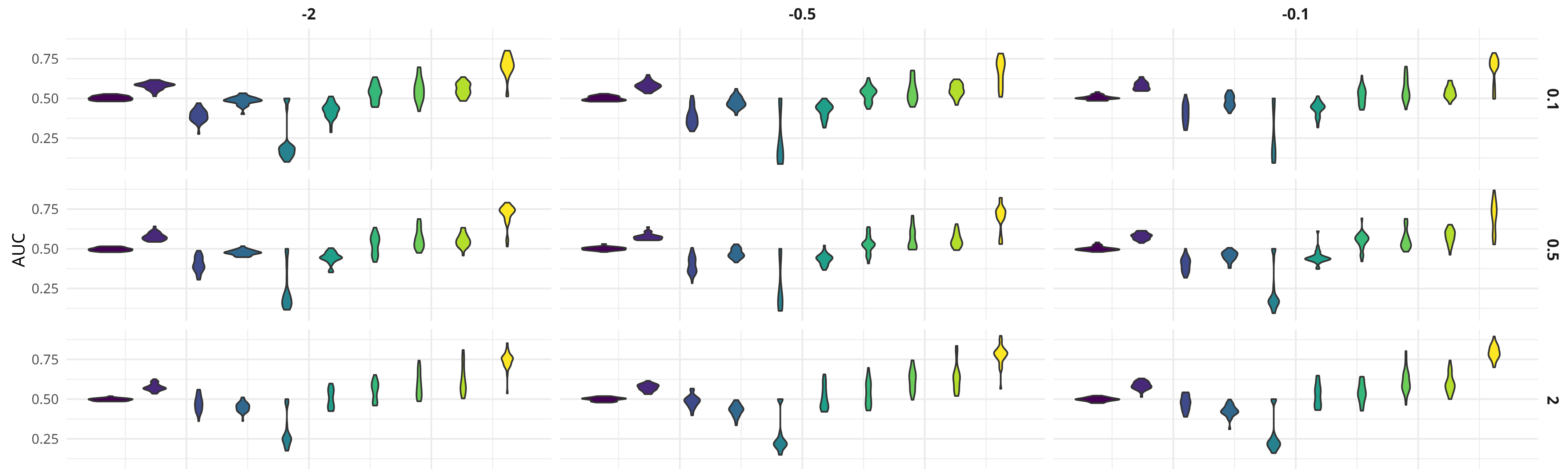


# 4. A "virtual ecologist" approach

## Results: competition

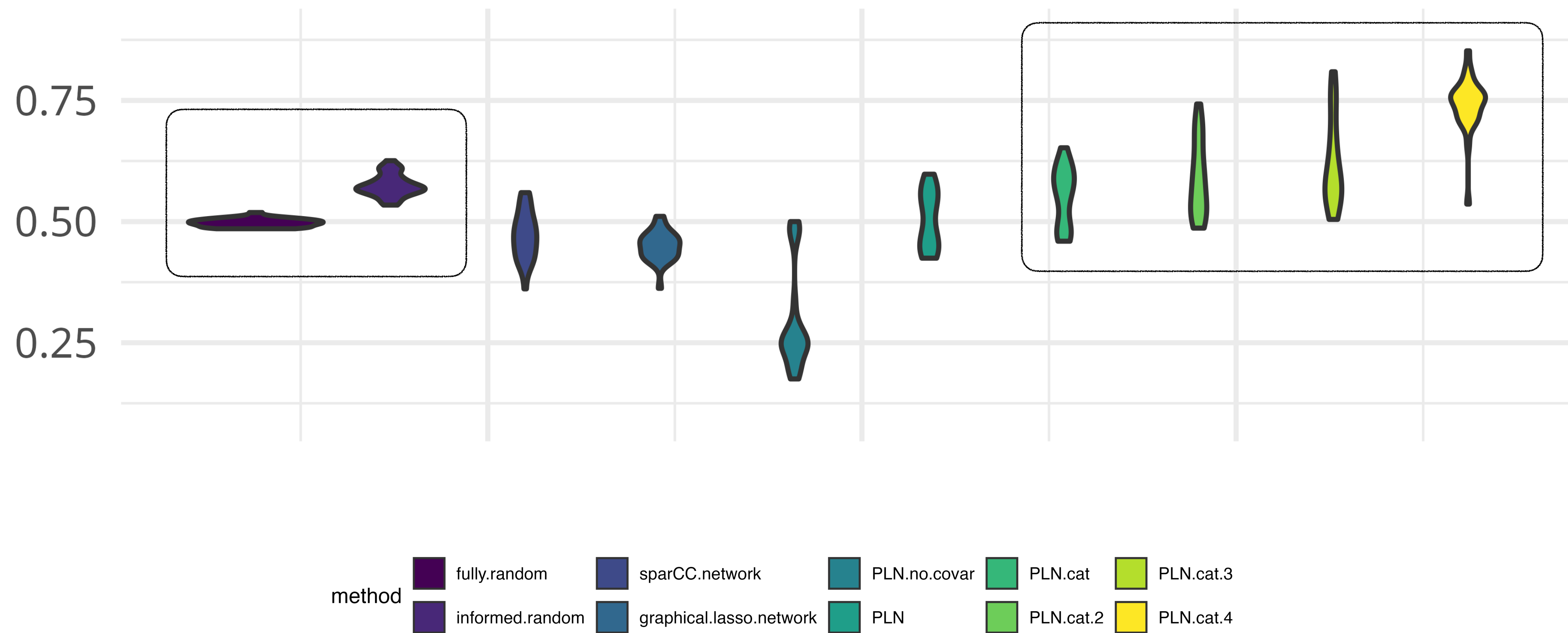
Competition AUC (competition.intensity ~ facilitation.intensity)

Row = competition intensity | Col = facilitation.intensity



# 4. A "virtual ecologist" approach

## Results: competition

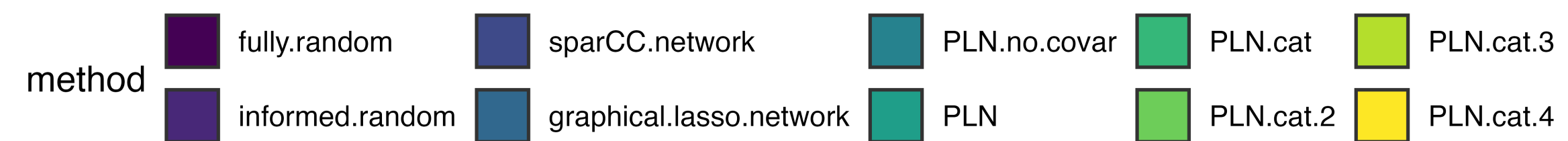
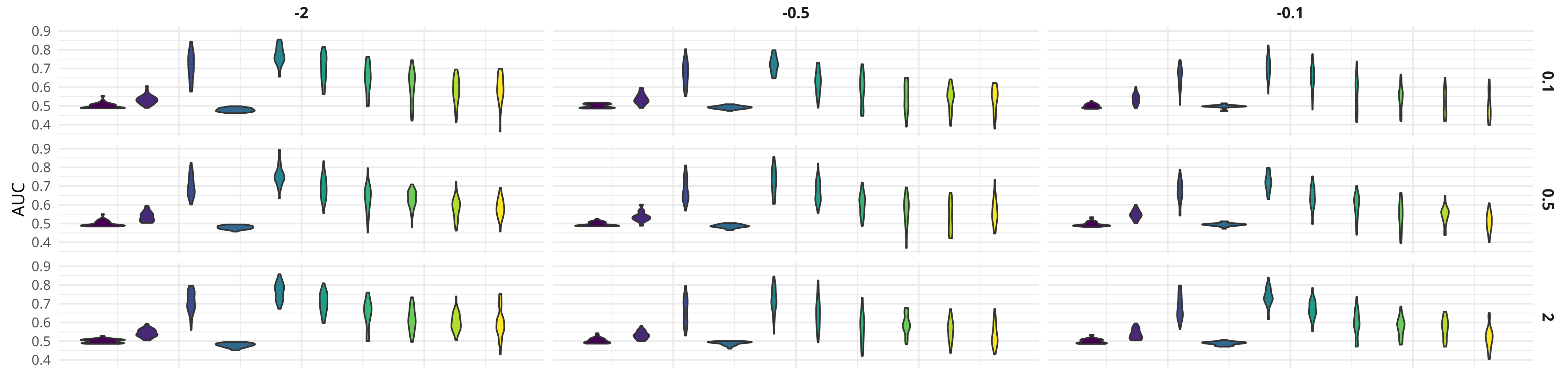


# 4. A "virtual ecologist" approach

## Results: facilitation

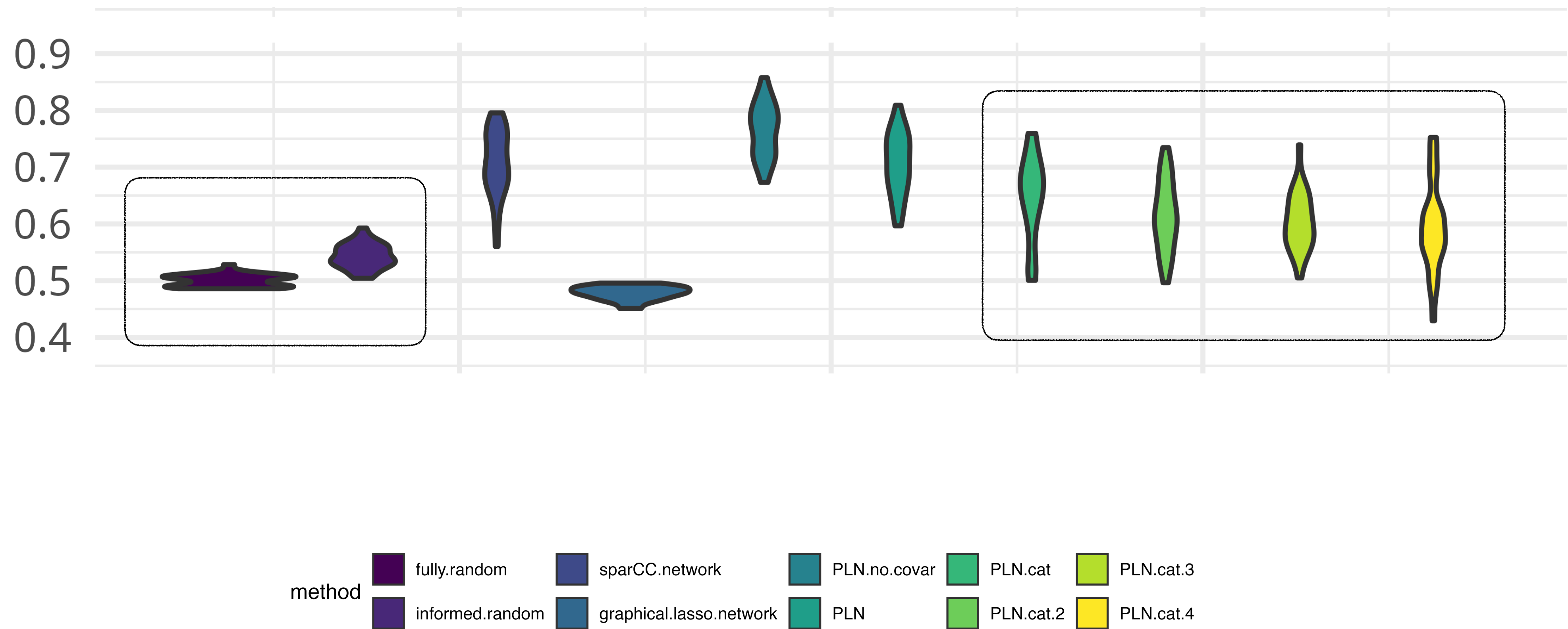
Facilitation AUC (competition.intensity ~ facilitation.intensity)

Row = competition intensity | Col = facilitation.intensity



# 4. A "virtual ecologist" approach

## Results: facilitation



# 4. A “virtual ecologist” approach

## Interpretation

- Reversed results for competition and facilitation: antagonistic vs synergistic effects ?
- Importance of adjusting the modelling of abiotic effects.
- Results robust to changes in intensities and densities.
- Slightly more surprising: they're also robust to asymmetry (competition / neutral)

# 5. Conclusion

- An original virtual framework to test a type of processes - patterns correspondence.
- Good correspondence for competition.
- Things are more complex for facilitation, that's probably owing to the synergetic vs antagonistic situation in this framework.
- Underlines the importance of a careful abiotic modelling.

# 6. Limits & perspectives

- There are more rigorous ways to model abiotic effects, eg using splines.
- Other interactions?
- Proximity between VirtualCom simulations and real data?
- To what extent is this a circular reasoning? 🤔



**Thank you !**

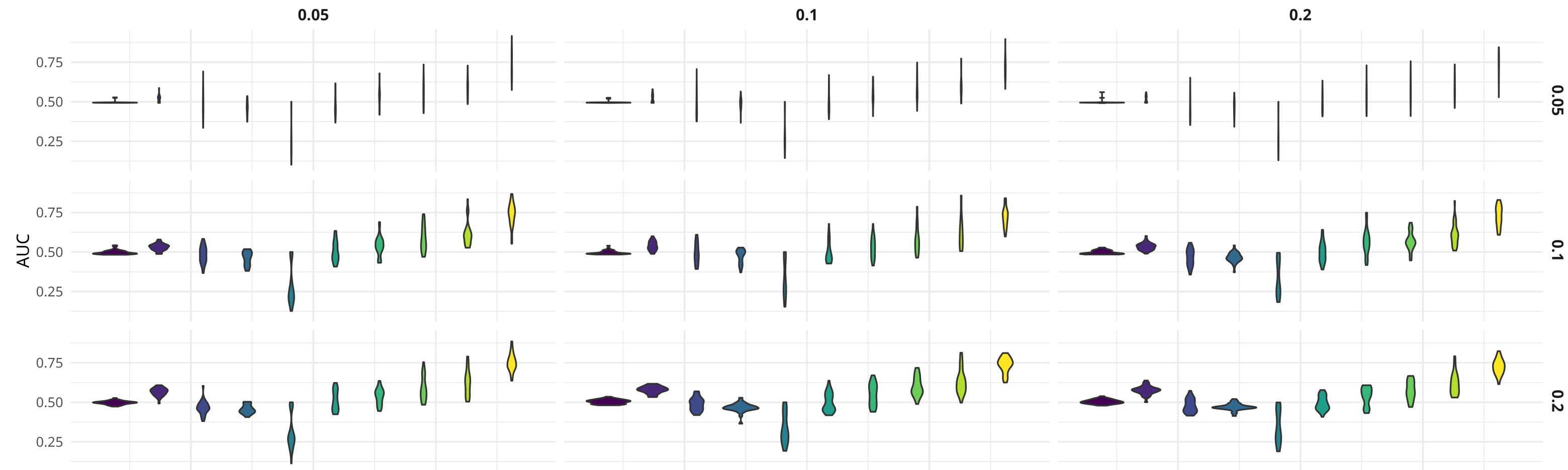
**Questions ? Suggestions ?**

**Remarks ?**

# Appendix 1: results for changing densities

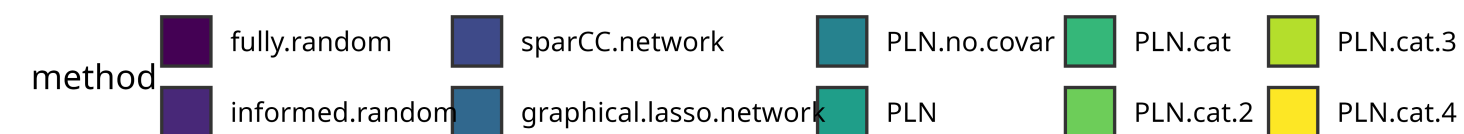
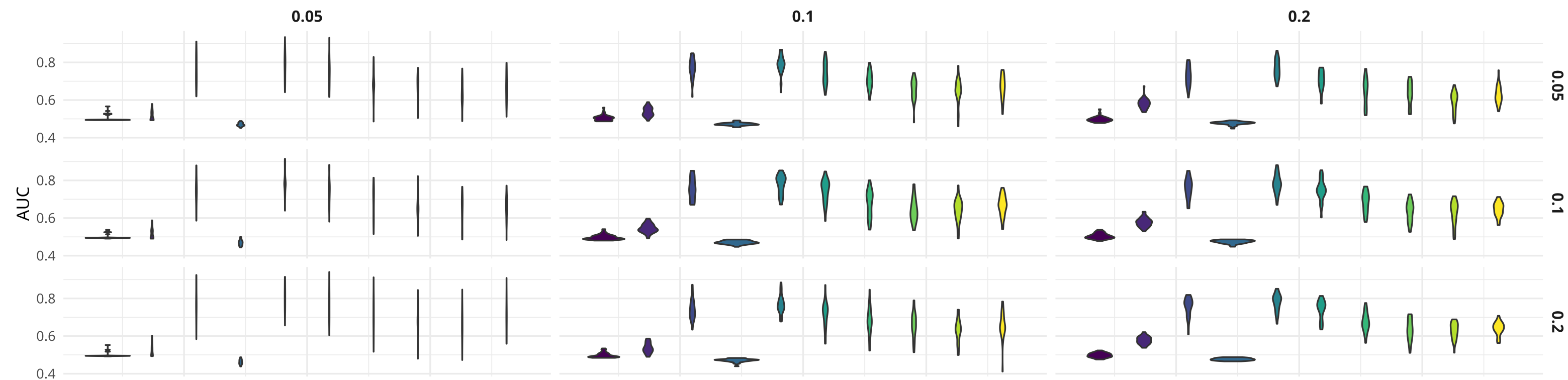
Competition AUC (competition.density ~ facilitation.density)

Row = competition density | Col = facilitation.density

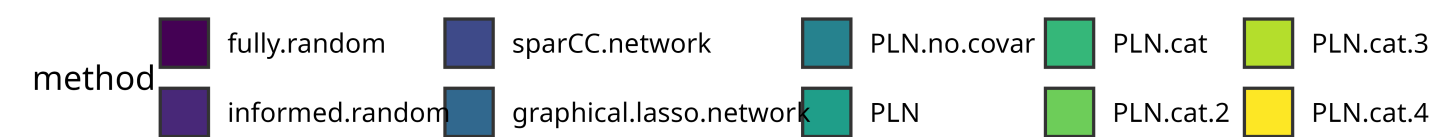
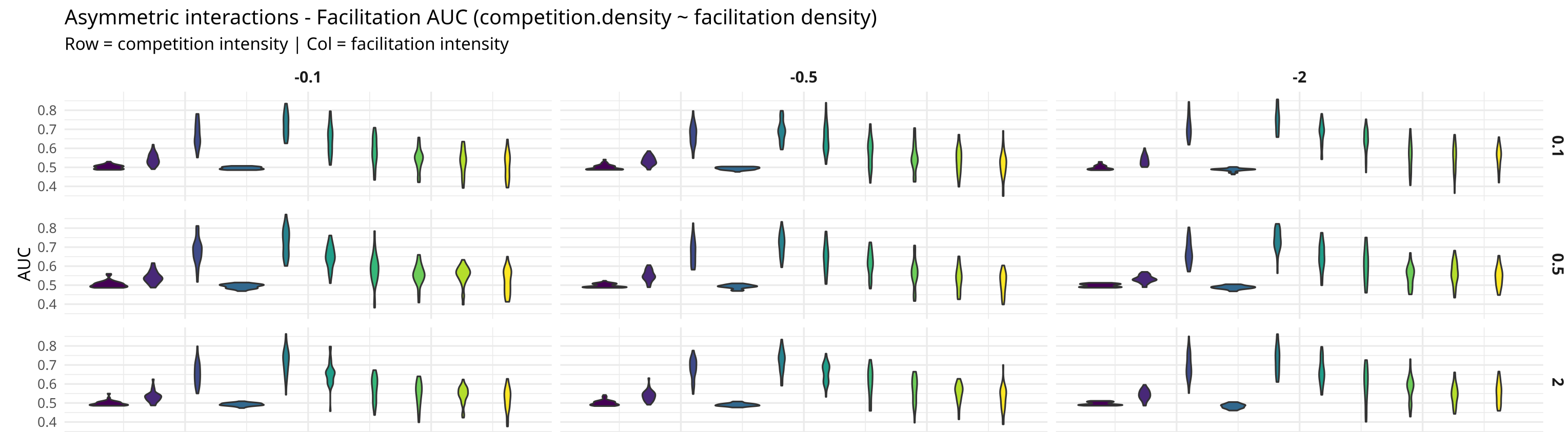
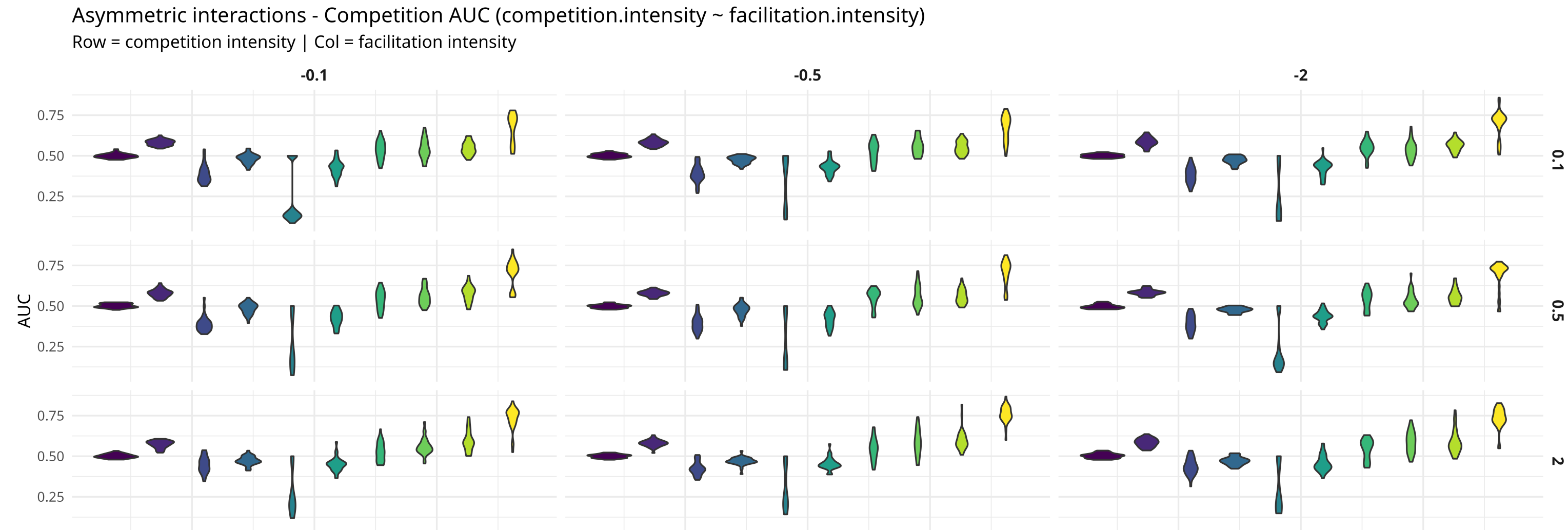


Facilitation AUC (competition.density ~ facilitation.density)

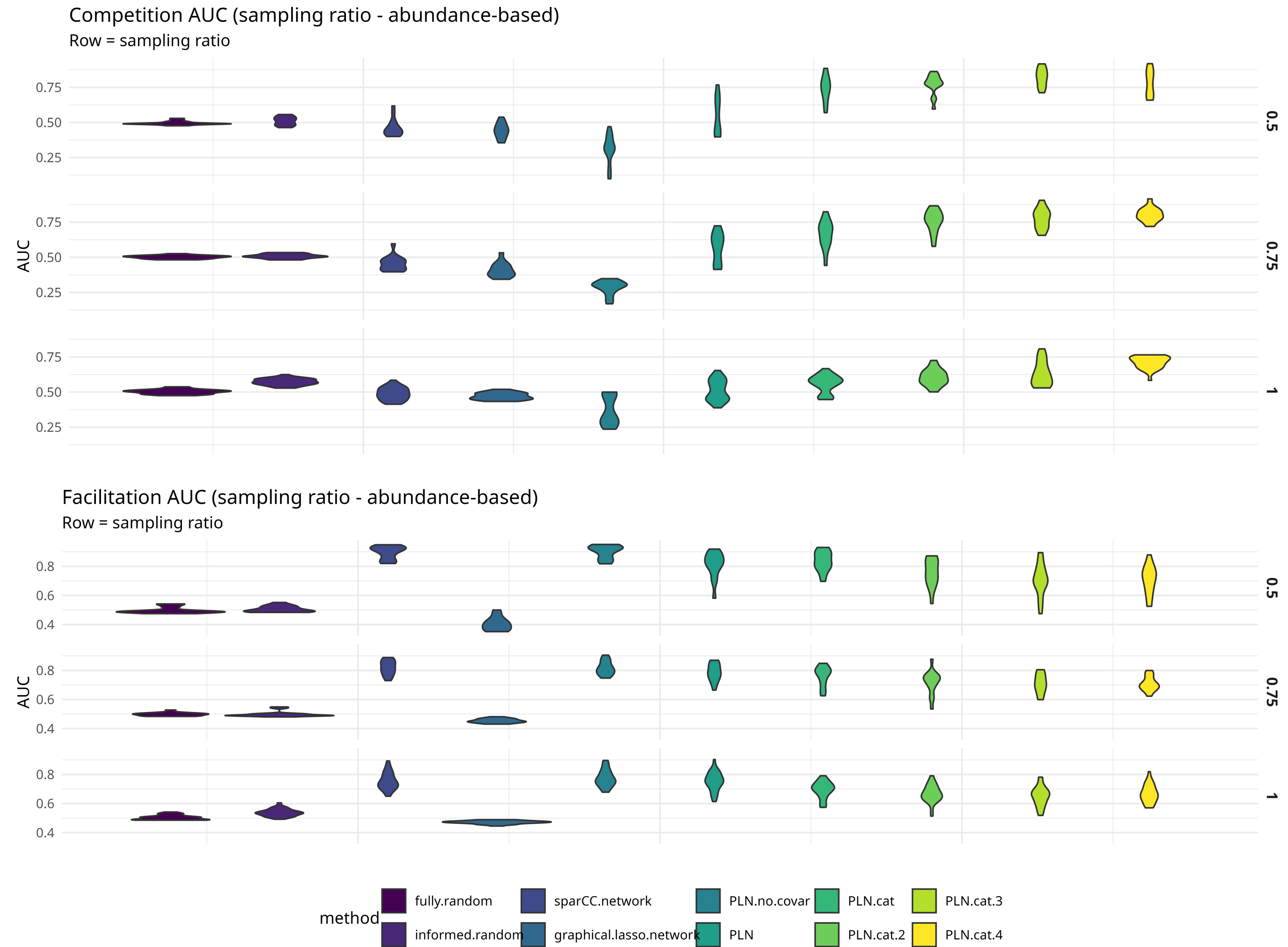
Row = competition density | Col = facilitation.density



# Appendix 2: results for asymmetric interactions



# Appendix 3: abundance-based sampling effect



# Appendix 4: random sampling effect

