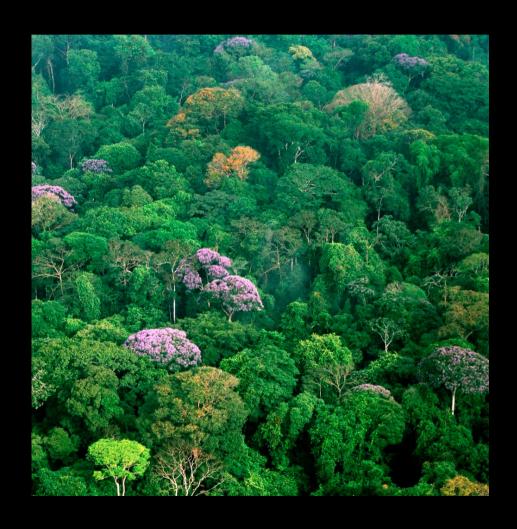
# Analyse spatiale multi-échelles de données écologiques avec adespatial

Stéphane Dray

Rencontres R - 2017

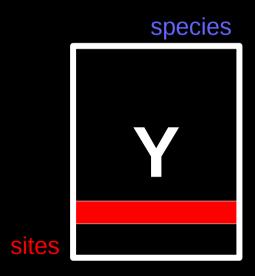
## Ecological communities



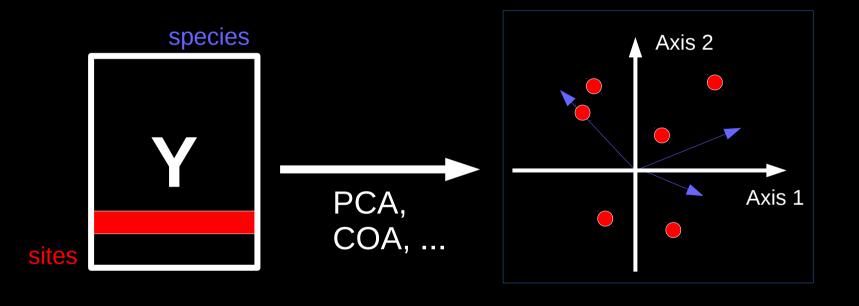
## Community ecology

- Describe communities (species assemblages)
- Study how their composition vary (in space or time)
- Identify factors/processes behind these patterns

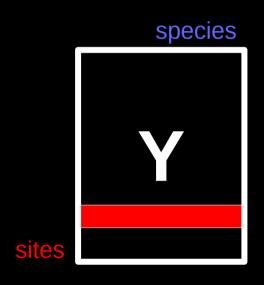
## One-table ordination

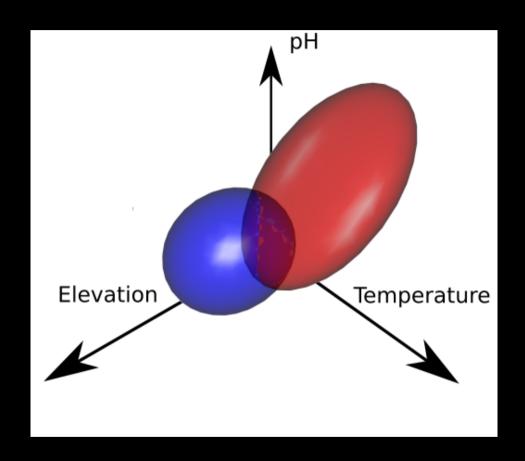


## One-table ordination

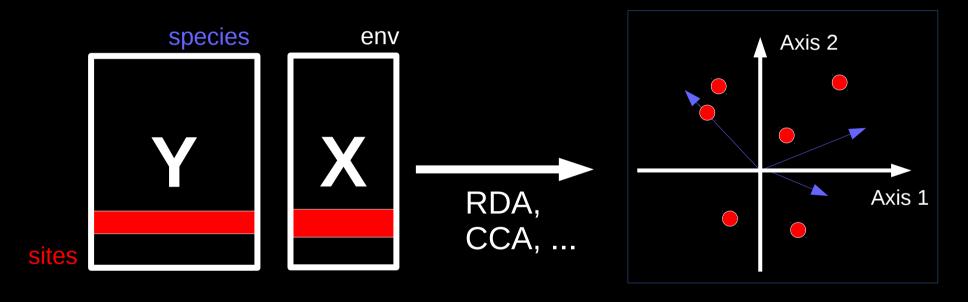


## The niche theory





### Two-tables ordination



Environmental constraint Axis =  $f(X)=a_1x_1+a_2x_2+...+a_px_p$ 

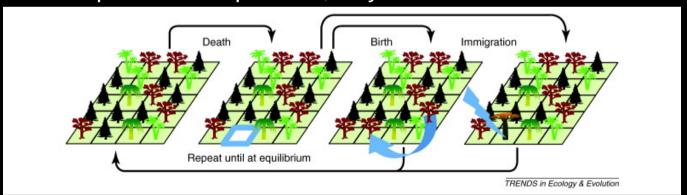
## Spatial patterns as proxies

- Some missing environmental variables
- Other ecological processes could act

## Spatial patterns as proxies

- Some missing environmental variables
- Other ecological processes could act

#### Species are equivalent, only stochastic events



## Spatial patterns as proxies

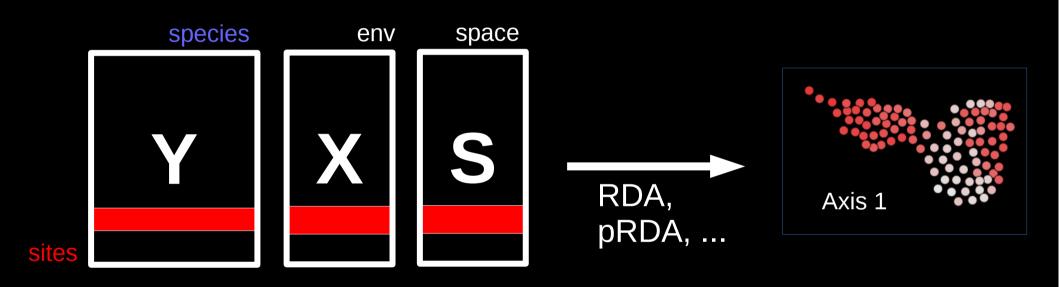
- Some missing environmental variables
- Other ecological processes could act
- Spatial structures as proxies of unmeasured variables / unknown processes

Ecology, 90(1), 2009, pp. 46-56 © 2009 by the Ecological Society of America

Beyond description: the active and effective way to infer processes from spatial patterns

ELIOT J. B. McIntire<sup>1,4</sup> and Alex Fajardo<sup>2,3</sup>

## Spatial patterns



- Spatial patterns linked to environment  $(Y \sim X \leftrightarrow S)$
- Spatial patterns due to other factors (Y $|X\leftrightarrow S$ )
- Which scales?

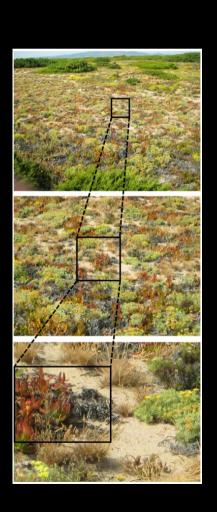
## How integrate spatial information?

- Mapping (Goodall, 1954)
- Spatial weights (Moran, 1954)
- Spatial predictors:
  - Polynomials (Gittins, 1968)
  - PCNM (Borcard & Legendre, 2002)
  - MEM (Dray et al, 2006)
  - AEM (Blanchet et al, 2008)

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- Table: sites x variables
- Variables = scales
- Orthonormal predictors





## REVIEWS

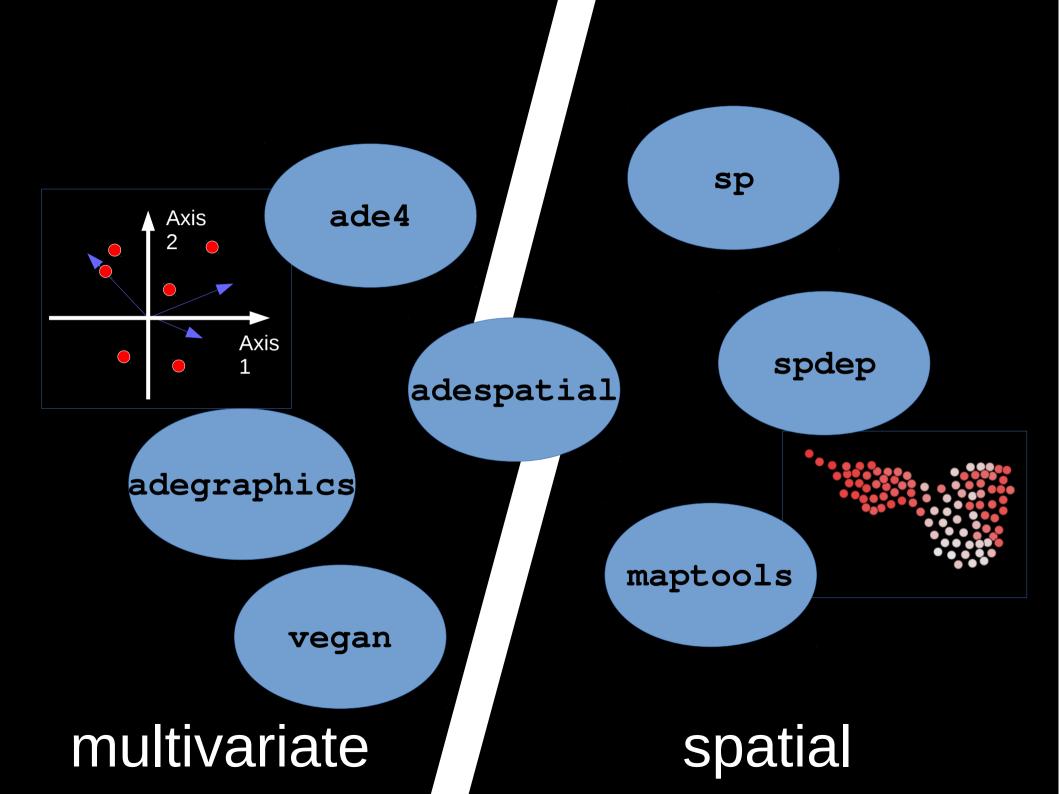
Monographs, 82(3), 2012, pp. 257–275 the Ecological Society of America

Community ecology in the age of multivariate multiscale spatial analysis

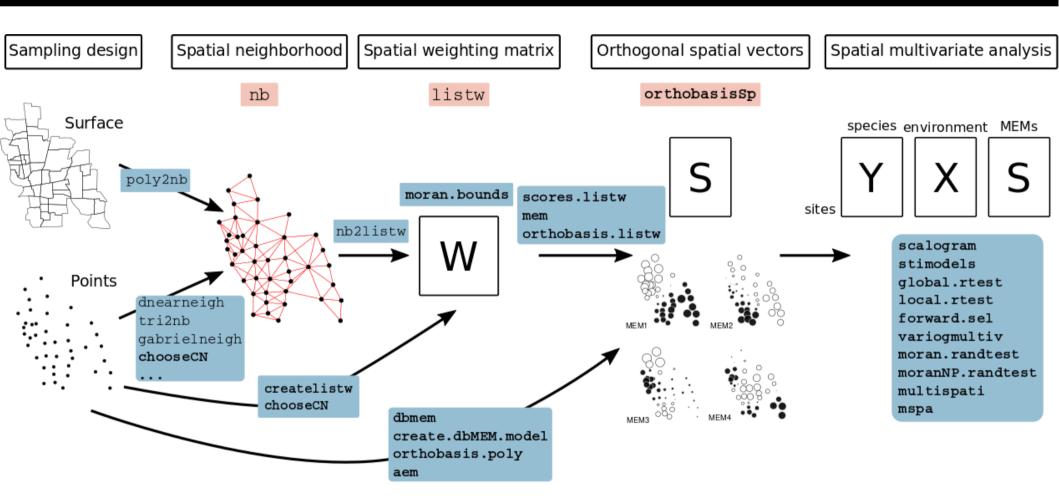
S. Dray,<sup>1,16</sup> R. Pélissier,<sup>2,3</sup> P. Couteron,<sup>2</sup> M.-J. Fortin,<sup>4</sup> P. Legendre,<sup>5</sup> P. R. Peres-Neto,<sup>6</sup> E. Bellier,<sup>7,8</sup> R. Bivand,<sup>9</sup> F. G. Blanchet,<sup>10</sup> M. De Cáceres,<sup>11</sup> A.-B. Dufour,<sup>1</sup> E. Heegaard,<sup>12</sup> T. Jombart,<sup>1,13</sup> F. Munoz,<sup>2</sup> J. Oksanen,<sup>14</sup> J. Thioulouse,<sup>1</sup> and H. H. Wagner<sup>15</sup>

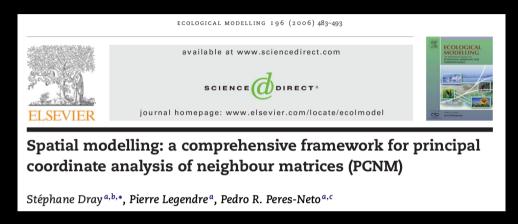


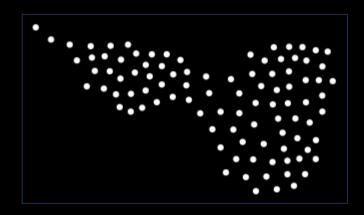
adespatial 0.0-1 on CRAN

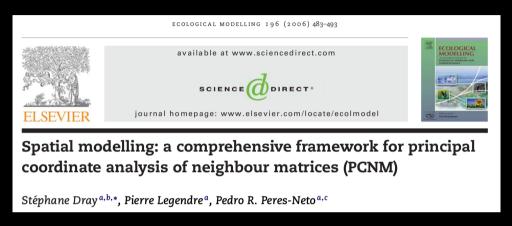


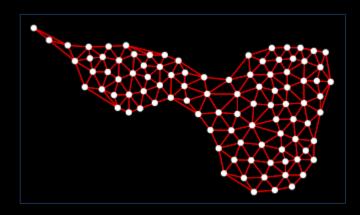
## adespatial: an overview

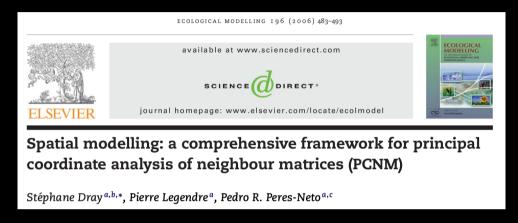




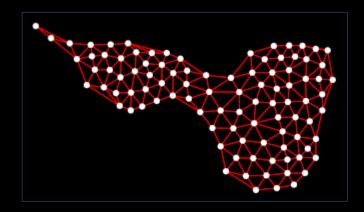








$$\mathbf{W} = \begin{bmatrix} 0 & 1 & \cdots & 0 \\ 1 & 0 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 1 & \cdots & 0 \end{bmatrix}$$





Spatial modelling: a comprehensive framework for principal coordinate analysis of neighbour matrices (PCNM)

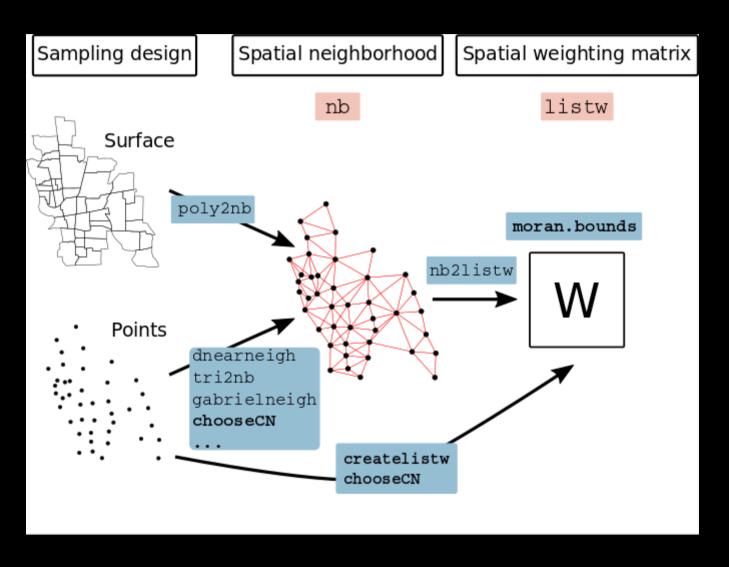
Stéphane Dray a,b,\*, Pierre Legendre a, Pedro R. Peres-Neto a,c

$$\mathbf{W} = \begin{bmatrix} 0 & 1 & \cdots & 0 \\ 1 & 0 & \cdots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 1 & \cdots & 0 \end{bmatrix}$$



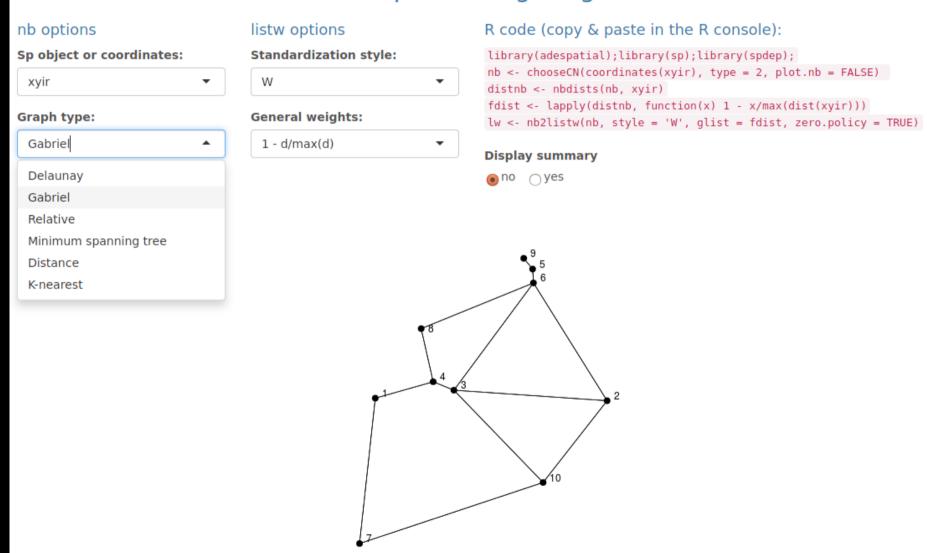
$$MC(\mathbf{x}) = rac{n}{\mathbf{1}_n^{\top} \mathbf{W} \mathbf{1}_n} rac{\mathbf{x}^{\top} \mathbf{H} \mathbf{W} \mathbf{H} \mathbf{z}}{\mathbf{z}^{\top} \mathbf{H} \mathbf{H} \mathbf{z}} = rac{n \sum_{i,j} w_{ij} (x_i - \bar{x}) (x_j - \bar{x})}{\sum_{i,j} w_{ij} \sum_i (x_i - \bar{x})^2}$$

## From spatial sampling to weights

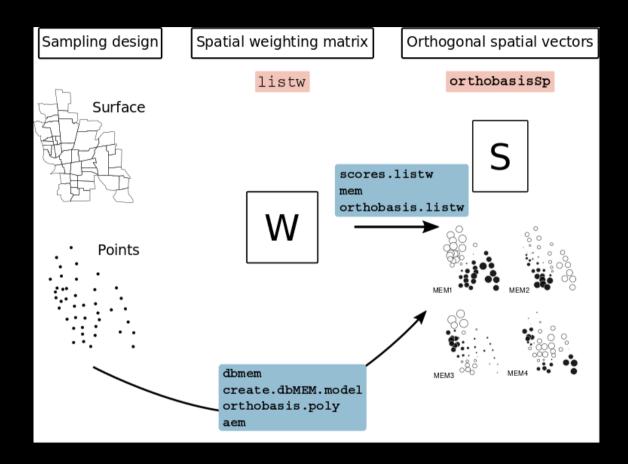


### createlistw

#### Generate R code to create a spatial weighting matrix



## From weights to spatial predictors



#### orthobasisSp

- · plot, summary, print methods
- inherits from data.frame

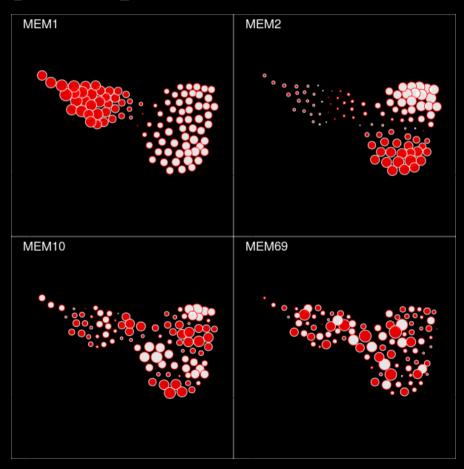
## Moran's Eigenvector Maps

$$\mathbf{HWHV} = \mathbf{V}\boldsymbol{\Lambda}$$

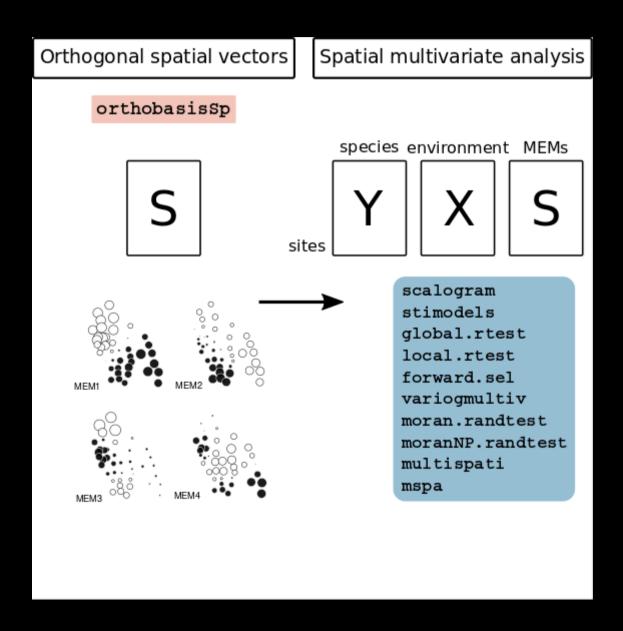
Eigenvectors are orthogonal and maximize Moran's coefficient :

$$MC(\mathbf{v}_i) = \lambda_i$$

mymem <- mem(mylistw)
plot(mymem[,c(1,2,10,69], xy</pre>



## Spatial multiscale/multivariate methods



#### Univariate multiscale methods

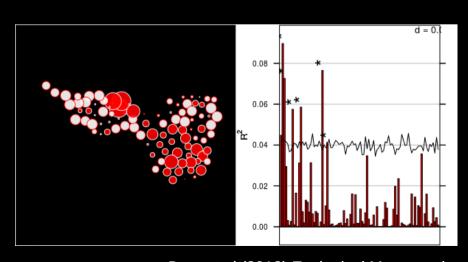
- Moran's Index: moran.randtest(x, mylistw)
- Positive/Negative decomposition: moranNP.randtest(x, mylistw)

$$MC(\mathbf{x}) = \sum_{i=1}^{n-1} \lambda_i cor^2(\mathbf{x}, \mathbf{v}_i)$$

Dray (2011) Geographical Analysis

Multiscale decomposition:

$$\sum_{i=1}^{n-1} cor^2(\mathbf{x}, \mathbf{v}_i) = 1$$



Dray et al (2012) Ecological Monographs Jombart, Dufour & Dray (2009) Ecography

## Spatial multivariate methods

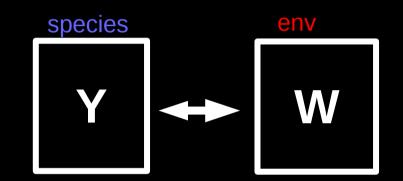
Multivariate autocorrelation: global.rtest(Y, mylistw)

Jombart et al (2008) Heredity

Spatial multivariate analysis:

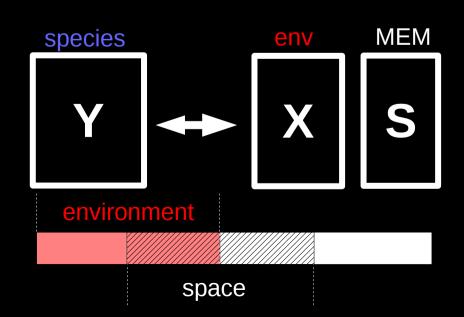
```
multispati(mypca, mylistw)
```

Dray et al (2008) Journal of Vegetation Science



Constrained ordination/Variation partitioning:

Dray et al (2012) Ecological Monographs



## The two faces of spatial autocorrelation

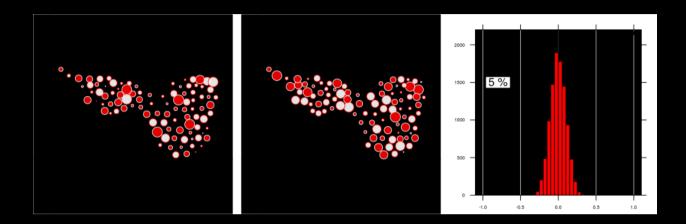


Proxies for unmeasured processes

Non-independent observations

## Inference and spatial autocorrelation

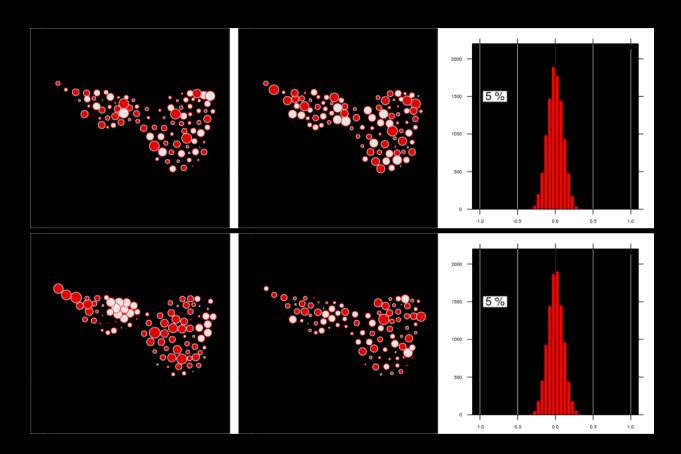
No spatial structure



## Inference and spatial autocorrelation

No spatial structure

One spatial structure

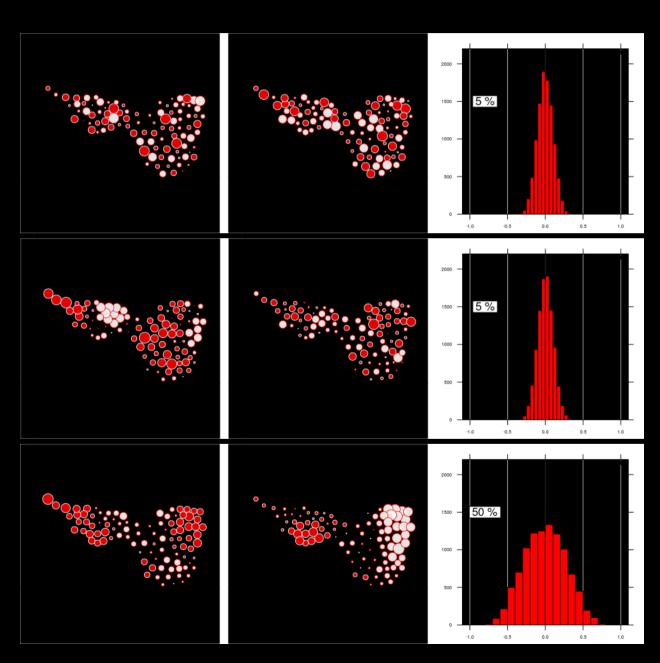


## Inference and spatial autocorrelation

No spatial structure

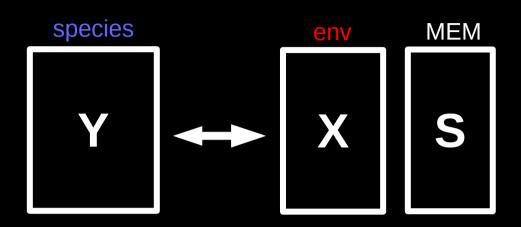
One spatial structure

 Two independent spatial structures



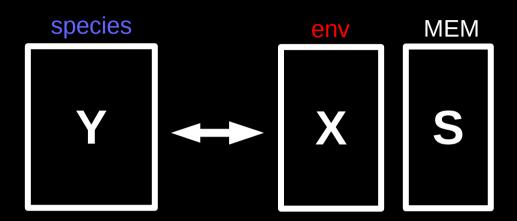
## Consequences in ordination

X and Y independents
 X and Y spatially structured



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X and Y independents
 X and Y spatially structured

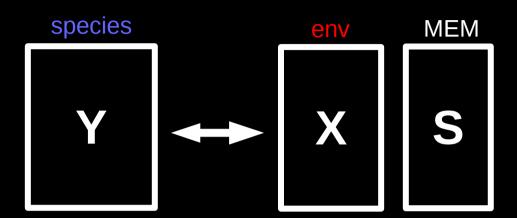


In theory



## Consequences in ordination

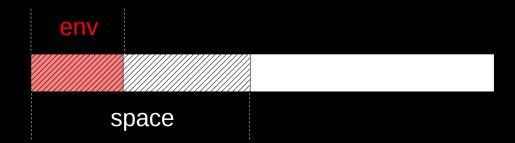
X and Y independents
 X and Y spatially structured



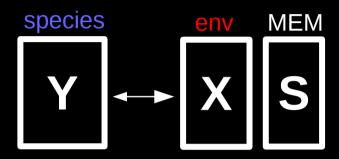
In theory

space

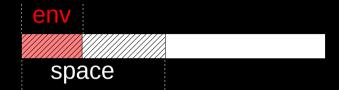
In practice



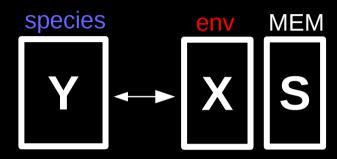
Observed data



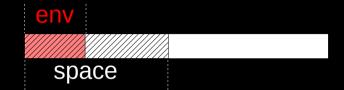
Obs:



Observed data

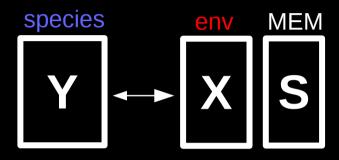


Obs:



• Simulated data

Observed data

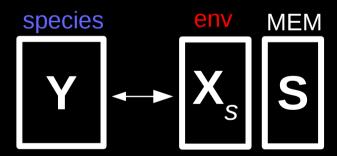


Obs: space

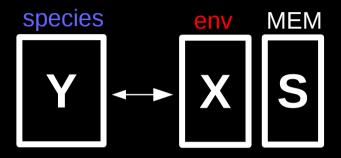
Sim 1:



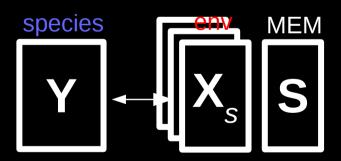
• Simulated data



Observed data

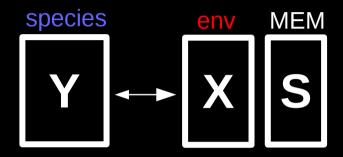


• Simulated data

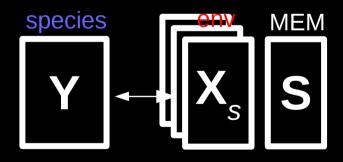


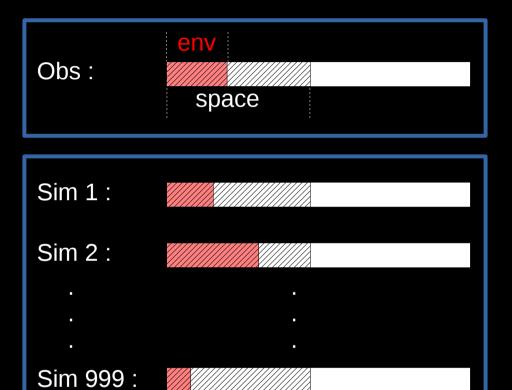


Observed data



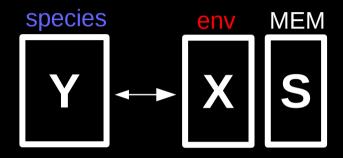
• Simulated data



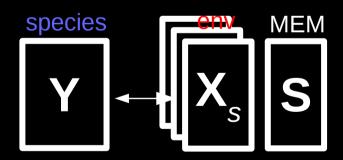


Comparison

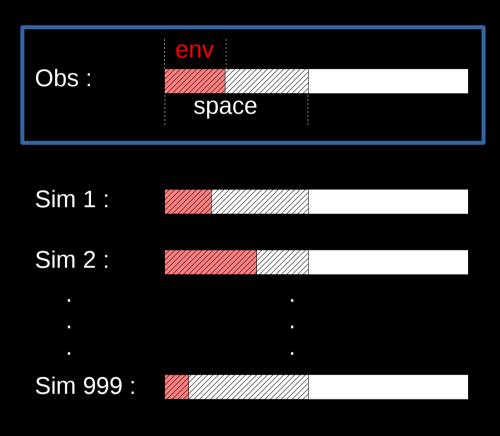
Observed data

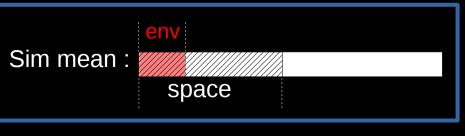


• Simulated data



Comparison





## Moran spectral randomization

Moran's coefficient decomposition:

$$MC(\mathbf{x}) = \sum_{i=1}^{n-1} \lambda_i cor^2(\mathbf{x}, \mathbf{v}_i)$$

Variable decomposition:

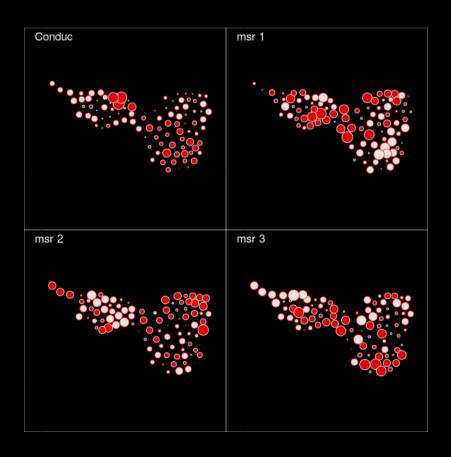
$$\mathbf{x} = \bar{x} + s(\mathbf{x}) \sum_{i=1}^{n-1} cor(\mathbf{x}, \mathbf{v}_i) \mathbf{v}_i$$

Random replicates:

$$\mathbf{x}_{sim} = \bar{x} + s(\mathbf{x}) \sum_{i=1}^{n-1} a_i \mathbf{v}_i$$

with 
$$a_i = \pm cor(\mathbf{x}, \mathbf{v}_i)$$

msr(X, mymem)





Pierre Legendre



Guillaume Larocque



Thibaut Jombart



Helene Wagner



**Daniel Borcard** 



Guillaume Blanchet

Naima Madi

Guillaume Guenard

## GitHub https://github.com/sdray/adespatial

## Vignette