Training in ade4 in R - Module II: Advanced methods

Introduction

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Material

The content of the course is available at

https://github.com/sdray/LausanneAdvanced/

In R, you can download the course by

usethis::use_course("sdray/LausanneAdvanced", destdir = "~/")

Online version at https://sdray.github.io/LausanneAdvanced

Required packages

- ade4 to run the analyses
- adegraphics to represent results
- adespatial and spdep for spatial analysis
- rgl to understand multivariate methods in interactive 3D

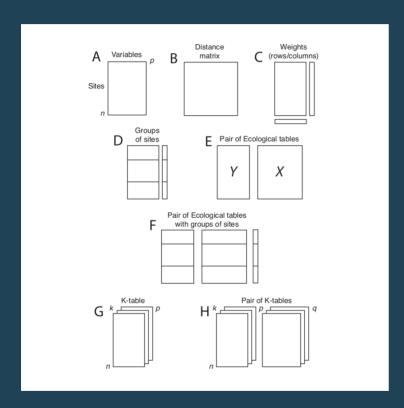
ade4

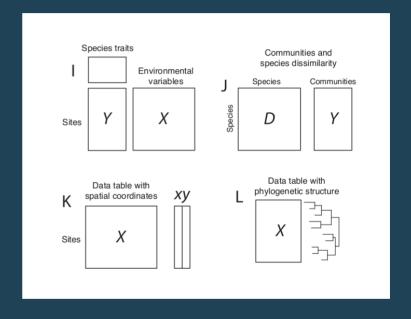
- R package since 2002
- Exploratory analysis of ecological data
 - Multivariate methods
 - Graphical functions

3,778 Environmental Sciences Ecology	2,225 Zoology	1,578 Biochemistry Molecular Biology	1,496 1 Life Sciences Biomedicine A Other Topics	,259 griculture
2,493 Genetics Heredity	1,687 Plant Sciences	1,041 Marine Freshwater Biology	904 Meteorology Atmospheric Sciences	688 Microbiology
2,337 Biodiversity Conservation	1,647 Evolutionary Biology	962 Science Technology Other Topics	797 Forestry	655 Anatomy Morphology

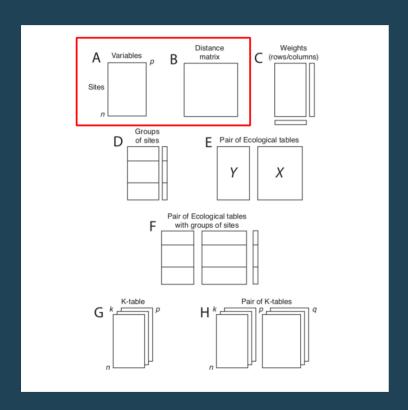
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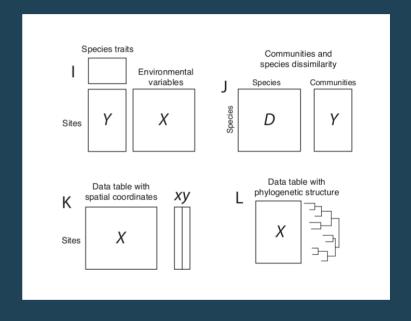
Data structure



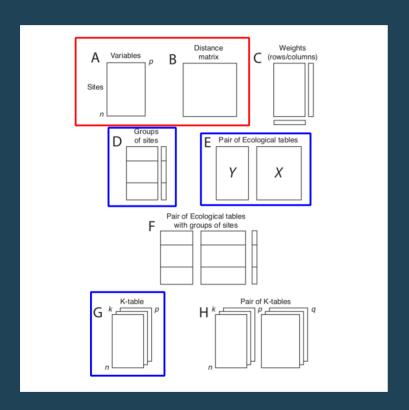


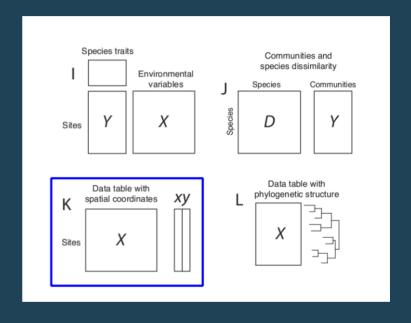
Data structure





Data structure





Module 1

variables



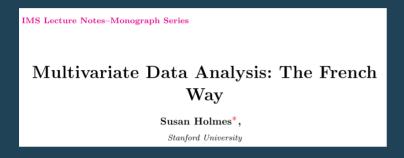
- Environmental variables
 - Quantitative variables → Principal Component
 Analysis (dudi.pca)
 - Categorical variables → Multiple Correspondence Analysis (dudi.acm)
 - Mix of both → Hill-Smith Analysis (dudi.hillsmith)
- Species table
 - Contingency table → Correspondence Analysis (dudi.coa)
 - Distance matrix → Principal Coordinates Analysis (dudi.pco)

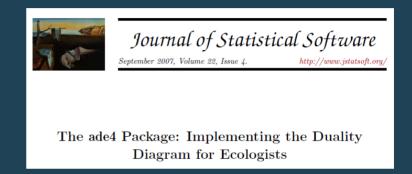
Module 2: course outline

We will explore the geometric properties, outputs and interpretation of multivariate analysis focusing on one-table methods. Last afternoon for case studies.

- One table + one categorical variable
- Two tables
- K tables
- One table + spatial information

ade4: the French way

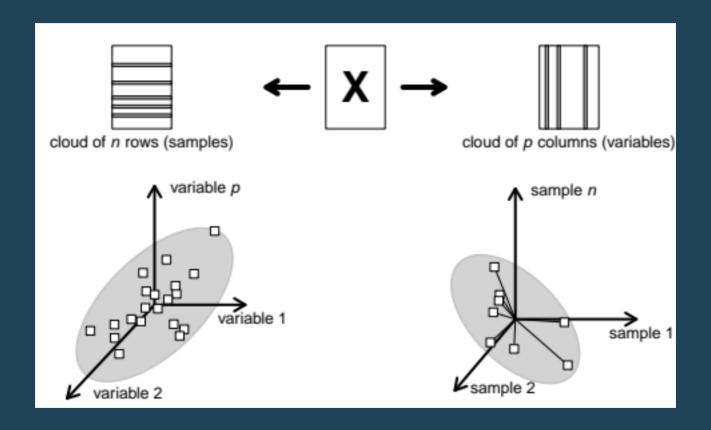




Implementation of functions in ade4 follows the duality diagram theory

More details are provided in the paper published in Journal of Statistical Software available here

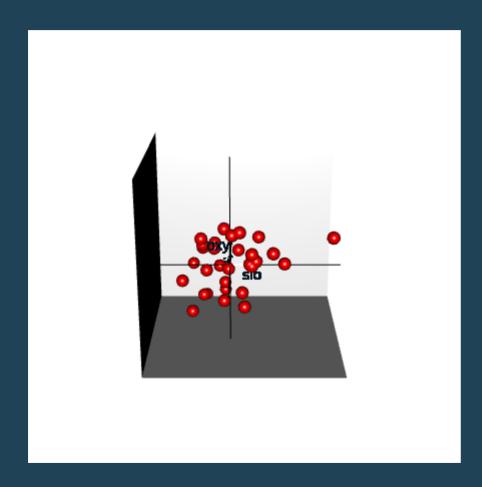
Two geometric views



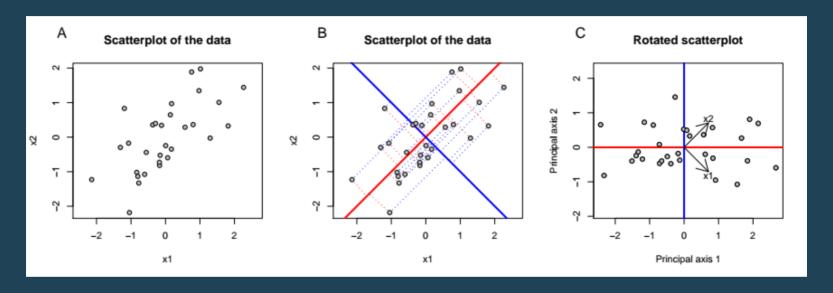
what are the main similarities and differences between the individuals?

what are the main relationships between the variables?

Explore the space of individuals



Geometric view for individuals



- Multivariate methods only perform geometric operations (rotations) to obtain the best viewpoint on the data
- When many variables are considered, dimension reduction is also applied to simplify the interpretation

Statistical triplet

Multivariate methods aim to answer these two questions and seek for small dimension hyperspaces (few axes) where the representations of individuals and variables are as close as possible to the original ones.

To answer the two previous questions, we define

- ${f Q}$, a p imes p positive symmetric matrix, used as an inner product in ${\Bbb R}^p$ and thus allows to measure distances between the n individuals
- \mathbf{D} , a $n \times n$ positive symmetric matrix, used as an inner product in \mathbb{R}^n and thus allows to measure relationships between the p variables.

 $(\mathbf{X}, \mathbf{Q}, \mathbf{D})$

Duality diagram theory

$$\mathbf{X}\mathbf{Q}\mathbf{X}^{\mathsf{T}}\mathbf{D}\mathbf{B} = \mathbf{B}\mathbf{\Lambda}$$

 $\mathbf{X}^{\mathsf{T}}\mathbf{D}\mathbf{X}\mathbf{Q}\mathbf{A} = \mathbf{A}\mathbf{\Lambda}$

- ullet ${f B}$ contains the principal components (${f B}^{ op}{f D}{f B}={f I}_r$).
- ullet ${f A}$ contains the principal axis (${f A}^{ op}{f Q}{f A}={f I}_r$).
- $\mathbf{L} = \mathbf{XQA}$ contains the row scores (projection of the rows of \mathbf{X} onto the principal axes)
- ${f C}={f X}^{ op}{f D}{f B}$ contains the column scores (projection of the columns of ${f X}$ onto the principal components)

Maximization of:

$$Q(\mathbf{a}) = \mathbf{a}^{ op} \mathbf{Q}^{ op} \mathbf{X}^{ op} \mathbf{D} \mathbf{X} \mathbf{Q} \mathbf{a} = \lambda ext{ and } S(\mathbf{b}) = \mathbf{b}^{ op} \mathbf{D}^{ op} \mathbf{X} \mathbf{Q} \mathbf{X}^{ op} \mathbf{D} \mathbf{b} = \lambda$$
 $\langle \mathbf{X} \mathbf{Q} \mathbf{a} | \mathbf{k}
angle_{\mathbf{D}} = \langle \mathbf{X}^t \mathbf{D} \mathbf{b} | \mathbf{a}
angle_{\mathbf{Q}} = \sqrt{\lambda}$

Inertia

• The total amount of information (variation) contained in the data is called the inertia

$$I_{(\mathbf{X}, \mathbf{Q}, \mathbf{D})} = Trace(\mathbf{X}^ op \mathbf{D} \mathbf{X} \mathbf{Q})$$

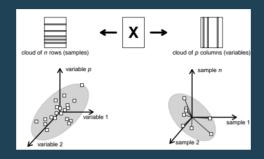
- Multivariate analysis aims to find new axes maximizing the projected inertia (i.e., the inertia of the projections).
- Total inertia is equal to the sum of eigenvalues
- In the case of PCA, total inertia is a sum of variances and an eigenvalue is equal to the variance of the projections on the associated axis

Implementation in ade4

Computations are performed by the function as.dudi. This functions takes 3 arguments defining the statistical triplet and returns an object of class dudi that contains:

ade4	theory	Definition
tab	\mathbf{X}	(transformed) data table
CW	${f Q}$	inner product for rows
lw	D	inner product for columns
eig	Λ	eigenvalues
l1	В	principal components
c1	\mathbf{A}	principal axes
li	${f L}$	row scores
СО	C	column scores

From the theory



• The principal axes

$$\mathbf{X}^{\mathsf{T}}\mathbf{D}\mathbf{X}\mathbf{Q}\mathbf{A} = \mathbf{A}\mathbf{\Lambda}$$

• The row scores

$$L = XQA$$

Maximization of

$$egin{aligned} Q(\mathbf{a}) &= \mathbf{a}^ op \mathbf{Q}^ op \mathbf{X}^ op \mathbf{D} \mathbf{X} \mathbf{Q} \mathbf{a} = \lambda \ Q(\mathbf{a}) &= \| \mathbf{X} \mathbf{Q} \mathbf{a} \|_{\mathbf{D}}^2 = \lambda \end{aligned}$$

• The principal components

$$\mathbf{X}\mathbf{Q}\mathbf{X}^{\mathsf{T}}\mathbf{D}\mathbf{K} = \mathbf{B}\mathbf{\Lambda}$$

• The column scores

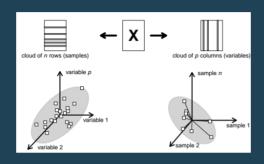
$$\mathbf{C} = \mathbf{X}^{\mathsf{T}} \mathbf{D} \mathbf{B}$$

Maximization of

$$S(\mathbf{b}) = \mathbf{b}^{ op} \mathbf{D}^{ op} \mathbf{X} \mathbf{Q} \mathbf{X}^{ op} \mathbf{D} \mathbf{b} = \lambda$$

$$S(\mathbf{b}) = \|\mathbf{X}^{ op}\mathbf{D}\mathbf{b}\|_{\mathbf{Q}}^2 = \lambda$$

To the practice in ade4



• The principal axes

\$c1

• The row scores

\$li

Maximization of

\$eig

• The principal components

\$11

• The column scores

\$co

Maximization of

\$eig

Available methods

Different definitions of a statistical triplet correspond to different methods

Function name	Analysis name	
dudi.pca	Principal component analysis	
dudi.pco	Principal coordinate analysis	
dudi.coa	Correspondence analysis	
dudi.acm	Multiple correspondence analysis	
dudi.dec	Decentered correspondence analysis	
dudi.fca	Fuzzy correspondence analysis	
dudi.fpca	Fuzzy PCA	
dudi.mix	Mixed nalysis	
dudi.hillsmith	Hill-Smith analysis	
dudi.nsc	Non-symmetric correspondence analysis	

Graphical functions

- Outputs of multivariate methods are usually provided as plots
- ade4 contains several graphical functions
- they have been re-implemented in a much more flexible way in the package adegraphics

A comprehensive overview of the package is available in its vignette available online or in R by:

vignette("adegraphics")

See also the paper published in the R Journal here

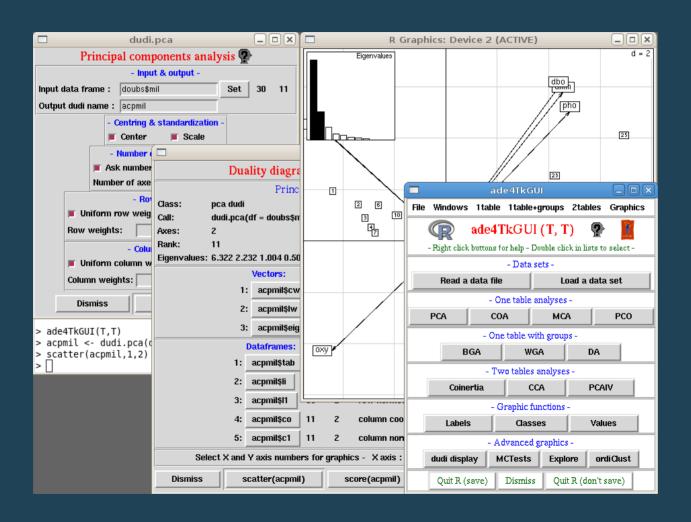
The ade packages

- adegraphics: S4-lattice based multivariate graphics
- adespatial: spatial multiscale multivariate analysis
- adiv: analysis of diversity
- adehabitat: analysis of habitat selection by animals
- adegenet: classes and methods for the multivariate analysis of genetic markers
- adephylo: exploratory analyses for the phylogenetic comparative method
- ade4TkGUI: graphical interface

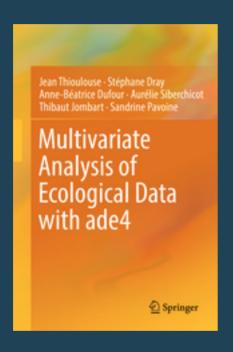
Other packages for multivariate analysis

- vegan: various tools devoted to ecology(diversity, dissimilarity), formulas, conditioning, variation partitioning
- FactoMineR: structured data, missing data

ade4TkGUI



Resources



• Mailing list:

http://listes.univlyon1.fr/wws/info/adelist

• Development:

https://github.com/sdray/ade4

• Courses (in French):

http://pbil.univlyon1.fr/R/enseignement.html

https://www.springer.com/fr/book/9781493988488