# Training in ade4 in R - Module I: Basic methods

Correspondence analysis

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#### Data

We will analyze the doubs data set (see ?doubs)

```
library(ade4)
library(adegraphics)
data(doubs)
names(doubs)

## [1] "env"  "fish"  "xy"  "species"

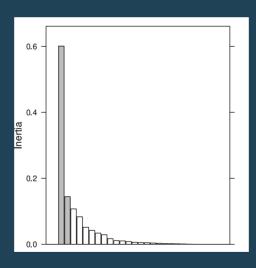
names(doubs$fish)

## [1] "Cogo" "Satr" "Phph" "Neba" "Thth" "Teso" "Chna" "Chto" "Lele" "Lece"
## [12] "Spbi" "Gogo" "Eslu" "Pefl" "Rham" "Legi" "Scer" "Cyca" "Titi" "Abbr"
## [23] "Acce" "Ruru" "Blbj" "Alal" "Anan"
```

## Correspondence Analysis

- Perform CA
- Display the barplot of eigenvalues

```
coal <- dudi.coa(doubs$fish, scannf = FALSE)
screeplot(coa1, main = " ", xlab = " ")</pre>
```



### **Inertia statistics**

Compute the percentage of variation explained by the first COA axes

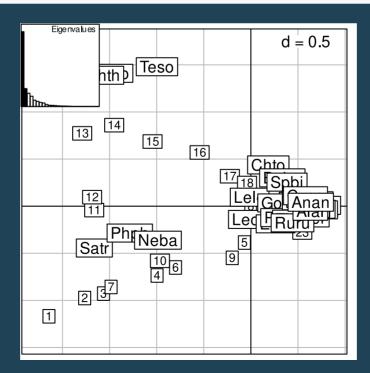
#### summary(coa1)

```
## Class: coa dudi
  Call: dudi.coa(df = doubs$fish, scannf = FALSE)
##
## Total inertia: 1.167
##
  Eigenvalues:
##
      Ax1
              Ax2
                     Ax3
                             Ax4
                                     Ax5
  0.60099 0.14437 0.10729 0.08337 0.05158
##
  Projected inertia (%):
      Ax1
              Ax2
                     Ax3
                                     Ax5
##
                             Ax4
                                  4.420
   51.503 12.372 9.195 7.145
##
  Cumulative projected inertia (%):
##
      Ax1
            Ax1:2 Ax1:3 Ax1:4
                                   Ax1:5
##
    51.50 63.87 73.07 80.21 84.63
##
  (Only 5 dimensions (out of 26) are shown)
```

# Graphical representation of CA results

Plot the results using the biplot function

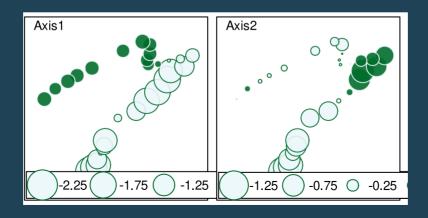
biplot(coa1)



# CA scores on the geographical map

- Draw maps of CA scores on the first two axes
- Interpret the maps to describe how the fish communities vary along the river

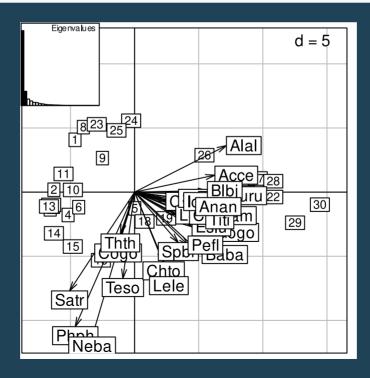
```
mypal <- colorRampPalette(c("#EDF8FB", "#006D2C"))
s.value(doubs$xy, coa1$li[, 1:2], pgrid.draw = FALSE,
    porigin.draw = FALSE, method = "size", symbol = "circle",
    col = mypal(2), ppoints.cex = 1)</pre>
```



## Principal Component Analysis

PCA can also be applied on the abundance table. Perform PCA on doubs\$fish table. Should we scale or not?

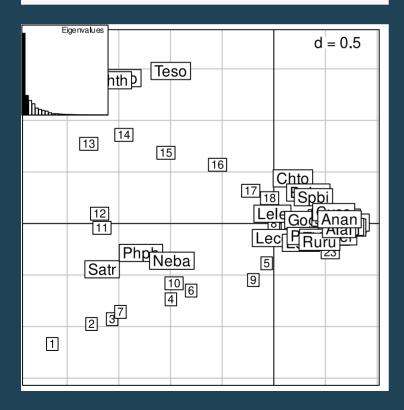
```
pca1 <- dudi.pca(doubs$fish, scale = FALSE, scannf = FALSE)
biplot(pca1)</pre>
```



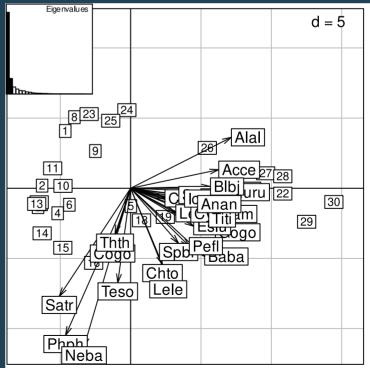
### PCA vs CA

#### Compare the biplots of CA and PCA

biplot(coa1)



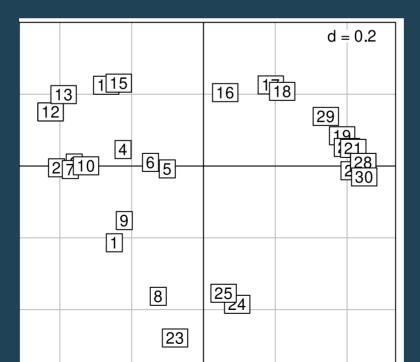
biplot(pca1)



## Principal Coordinates Analysis

- Compute Jaccard distances between sites
- Perform principal coordinates analysis and display the ordination of sites

```
dis <- dist.binary(doubs$fish, method = 1)
pco1 <- dudi.pco(dis, scannf = FALSE)
s.label(pco1$li)</pre>
```



### PCA and PCoA

 Compare the results of PCA and those of PCoA applied on Euclidean distance (function dist)

```
dis2 <- dist(doubs$fish)
pco2 <- dudi.pco(dis2, scannf = FALSE)</pre>
```

```
head(pca1$eig, 2)
## [1] 42.746273 8.158339
```

head(pca1\$li)

```
## Axis1 Axis2

## 1 -4.64572246 4.0779692

## 2 -6.29163098 0.1850865

## 3 -6.58040507 -1.3415646

## 4 -5.18637301 -1.7826102

## 5 0.01763648 -1.2682324

## 6 -4.35691748 -1.1761457
```

```
head(pco2$eig, 2)

## [1] 42.746273 8.158339

head(pco2$li)
```

```
## A1 A2
## 1 -4.64572246 4.0779692
## 2 -6.29163098 0.1850865
## 3 -6.58040507 -1.3415646
## 4 -5.18637301 -1.7826102
## 5 0.01763648 -1.2682324
## 6 -4.35691748 -1.1761457
```