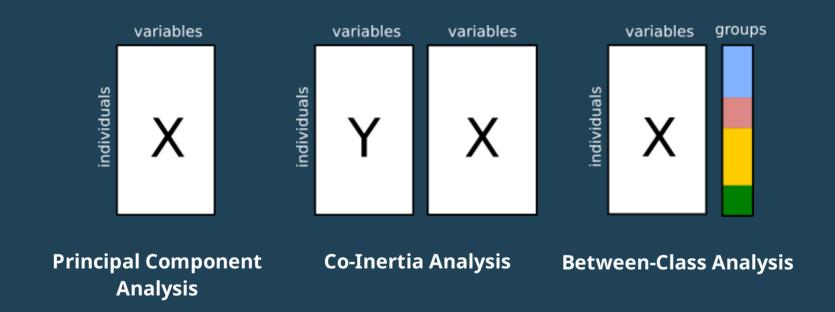
# Raw-data and distance based methods

in practice

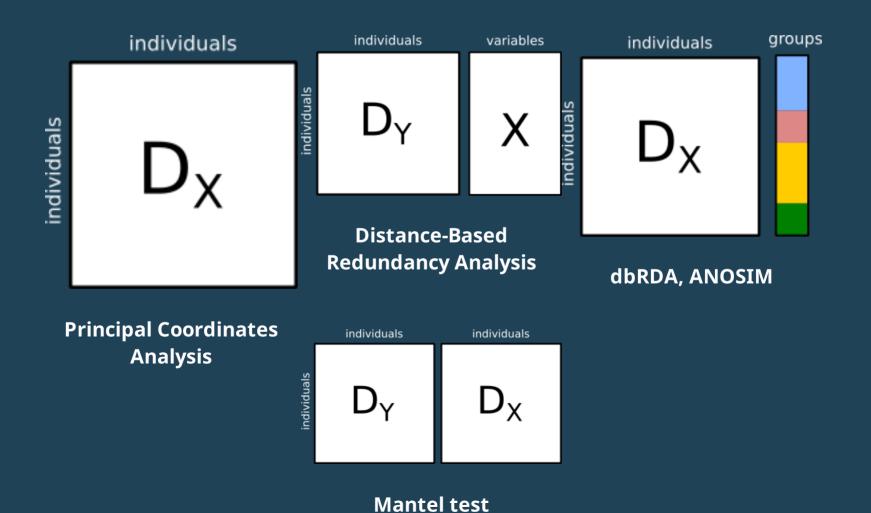
Stéphane Dray

2024-01-08

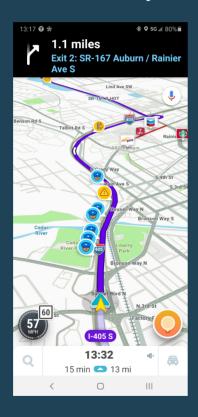
## Raw data

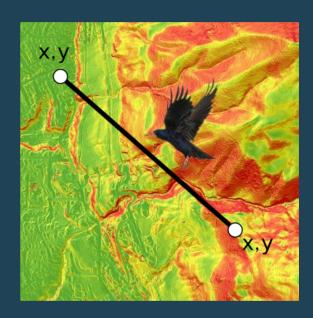


#### **Distances**



• Distances can be directly measured or inherited from raw-data



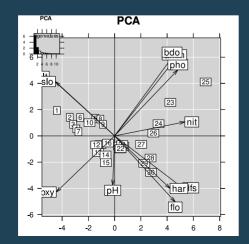


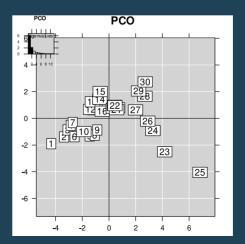
• Raw-data methods produce information on both individuals and variables when distance-based methods focus only on individuals

```
library(ade4)
library(adegraphics)
data(doubs)
pca <- dudi.pca(doubs$env, scannf = FALSE)
pco <- dudi.pco(dist(scale(doubs$env)), scannf = FALSE)</pre>
```

```
scatter(pca, main = "PCA")
```

scatter(pco, main = "PCO")





• Distance-based methods allow for more flexibility and can be more suitable in some contexts

```
Χ
         species 1 species 2 species 3 species 4
## site 1
## site 2
## site 3
## site 4
## Euclidean
                                   ## Jaccard
round(dist(x), 2)
                                   round(dist.binary(x, method = 1)
  site 1 site 2 site 3
                                  ## site 1 site 2 site 3
##
## site 2 1.41
                                  ## site 2 0.82
## site 3 1.00 1.73
                                  ## site 3 0.50 0.87
## site 4 1.73 1.00 1.41
                                  ## site 4 0.87 0.71 0.71
```

Sites 3-4 are closer than sites 1-2 when considering only presences as a measure of similarity

• In some cases, both approaches can be equivalent (Euclidean distances)

For the univariate case, Euclidean distance is  $d_{ij} = \sqrt{(x_i - x_j)^2}$  and we have:

$$var(\mathbf{x}) = rac{1}{n} \sum_{i=1}^n (x_i - ar{x})^2 = rac{1}{2n^2} \sum_{i=1}^n \sum_{j=1}^n d_{ij}^2.$$

```
x <- rnorm(10)
var(x) * 9/10
```

## [1] 0.9402963

```
sum(as.matrix(dist(x))^2)/(2 * 10^2)
```

## [1] 0.9402963

When using Euclidean distance, several raw- and distance-based analysis would produce the same results.

#### Your turn

- 1. Create an Rmd file
- 2. Load the meaudret data set from ade4
- 3. Perform the principal component analysis (dudi.pca) and principal coordinates analysis (dudi.pco) using Euclidean distances (dist) on faunistic data. Compare the outputs and conclude.
- 4. Transform the data into presence-absence (ifelse(meaudret\$spe>0, 1, 0)). Peform principal coordinates analysis (dudi.pco) using Euclidean (dist) and Jaccard distances (dist.binary). Compare the results.
- 5. Perform the between-class analysis (bca) and distance-based RDA (vegan::dbrda) using Euclidean distances using the factor meaudret\$design\$season as an exploratory variable. Compare the results (be aware that vegan use  $\frac{1}{n-1}$  to compute variances while ade4 uses  $\frac{1}{n}$ ).
- 6. Look at the percentage of variation explained by the between-class analysis (stored in the object). Perform permutational multivariate analysis of variance with the function vegan::adon's using the Euclidean distances. Compare

# **Summary**

When using Euclidean distances, we have:

- Permutational Distance-based Multivariate Analysis of Variance 
   ⇔ Permutation test of Between-Class Analysis

```
## Answer 2
library(ade4)
library(adegraphics)
data(meaudret)
## Answer 3
pca.meau <- dudi.pca(meaudret$spe, scale = FALSE, scannf = FALSE)</pre>
pco.meau <- dudi.pco(dist(meaudret$spe), scannf = FALSE)</pre>
pca.meau$eig - pco.meau$eig ## Equivalence between the two analyses
## Answer 4
meau01 <- ifelse(meaudret$spe > 0, 1, 0)
pco.meau1 <- dudi.pco(dist(meau01), scannf = FALSE)</pre>
pco.meau2 <- dudi.pco(dist.binary(meau01, method = 1),</pre>
    scannf = FALSE)
cbindADEg(s.class(pco.meau1$li, meaudret$design$season,
    col = TRUE), s.class(pco.meau2$li, meaudret$design$season,
    col = TRUE))
## Answer 5
bca1 <- bca(pco.meau, meaudret$design$season, scannf = FALSE)</pre>
dbrda1 <- vegan::dbrda(dist(meaudret$spe) ~ meaudret$design$season)</pre>
dbrda1$CCA$eig * (nrow(meaudret$spe) - 1)/nrow(meaudret$spe) -
    bca1$eig ## Equivalence between the two analyses
## Answer 6
vegan::adonis(dist(meaudret$spe) ~ meaudret$design$season)
randtest(bca1)
```