

Fish trait analyses

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Framework

This document presents the analyses done on the fish trait matrix.

Data

Abundances

DATRAS CGFS

Fish “traits”

Sources

Sources: - our own work on fishbase and IUCN - Beuhkof et al Pangea database and - benthos : xxx

Raw data

Missing values imputation

```
##                               nbid nbNA
## Species                      84     0
## habitat                      6     0
## feeding.mode                 4     0
## tl                          71     0
## age.maturity                 37     2
## growth.coefficient          77     0
## length.max                  65     0
## age.max                     37     0
## IUCN.status                  6     0
```

```
##
## iter imp variable
## 1 1 age.maturity
## 2 1 age.maturity
## 3 1 age.maturity
## 4 1 age.maturity
## 5 1 age.maturity
```

```
##                               nbid nbNA
## Species                      84     0
## habitat                      6     0
```

```
## feeding.mode      4    0
## tl                71    0
## age.maturity      36    0
## growth.coefficient 77    0
## length.max        65    0
## age.max           37    0
## IUCN.status        6    0
```

Quantitative to qualitative data

```
## [1] 0 1 2 3 4 5
```

```
## [1] 0 2 3 4 5 14
```

```
## [1] 0.0 0.1 0.2 0.3 0.4 0.5 2.0
```

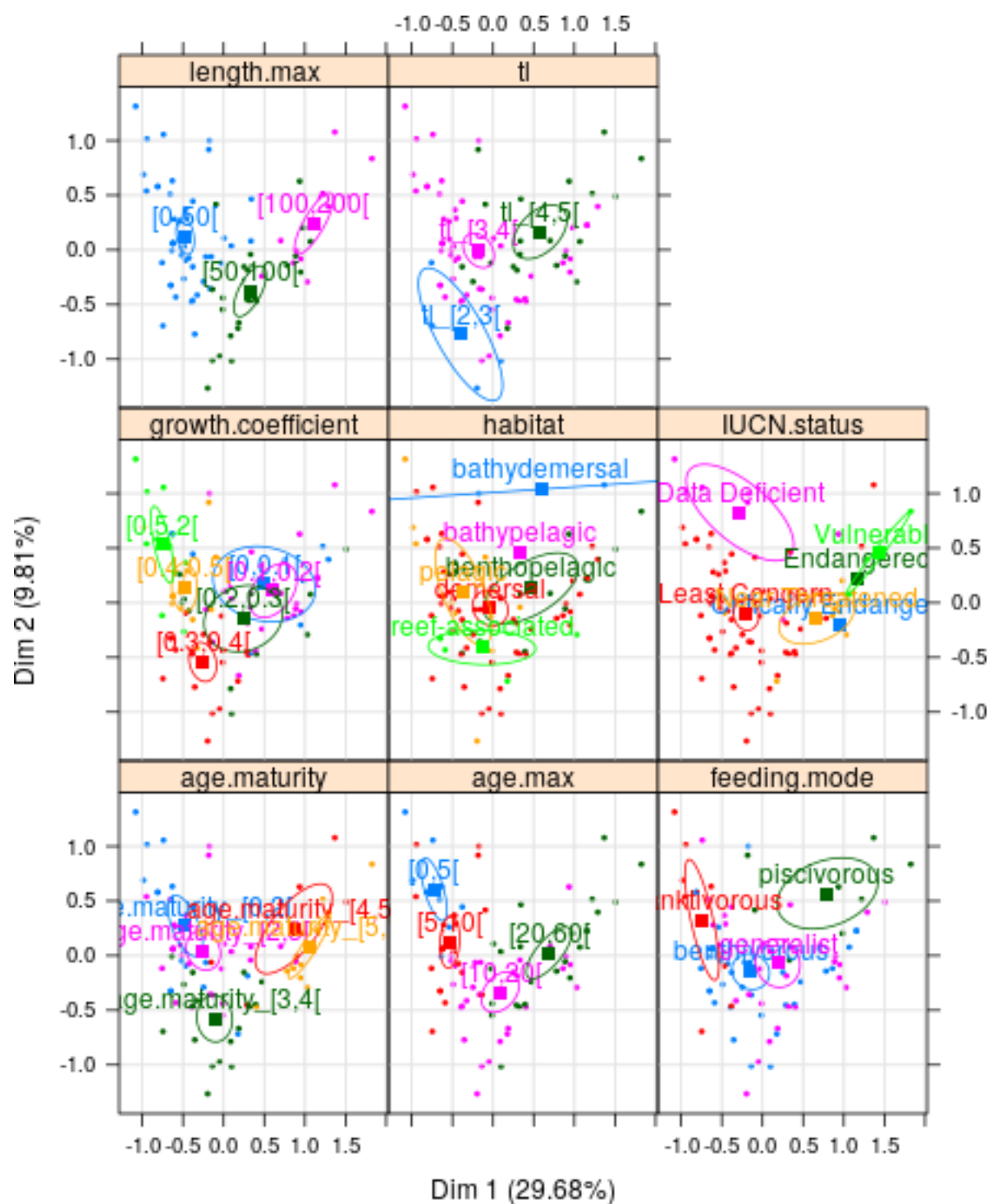
```
## [1] 0 50 100 200
```

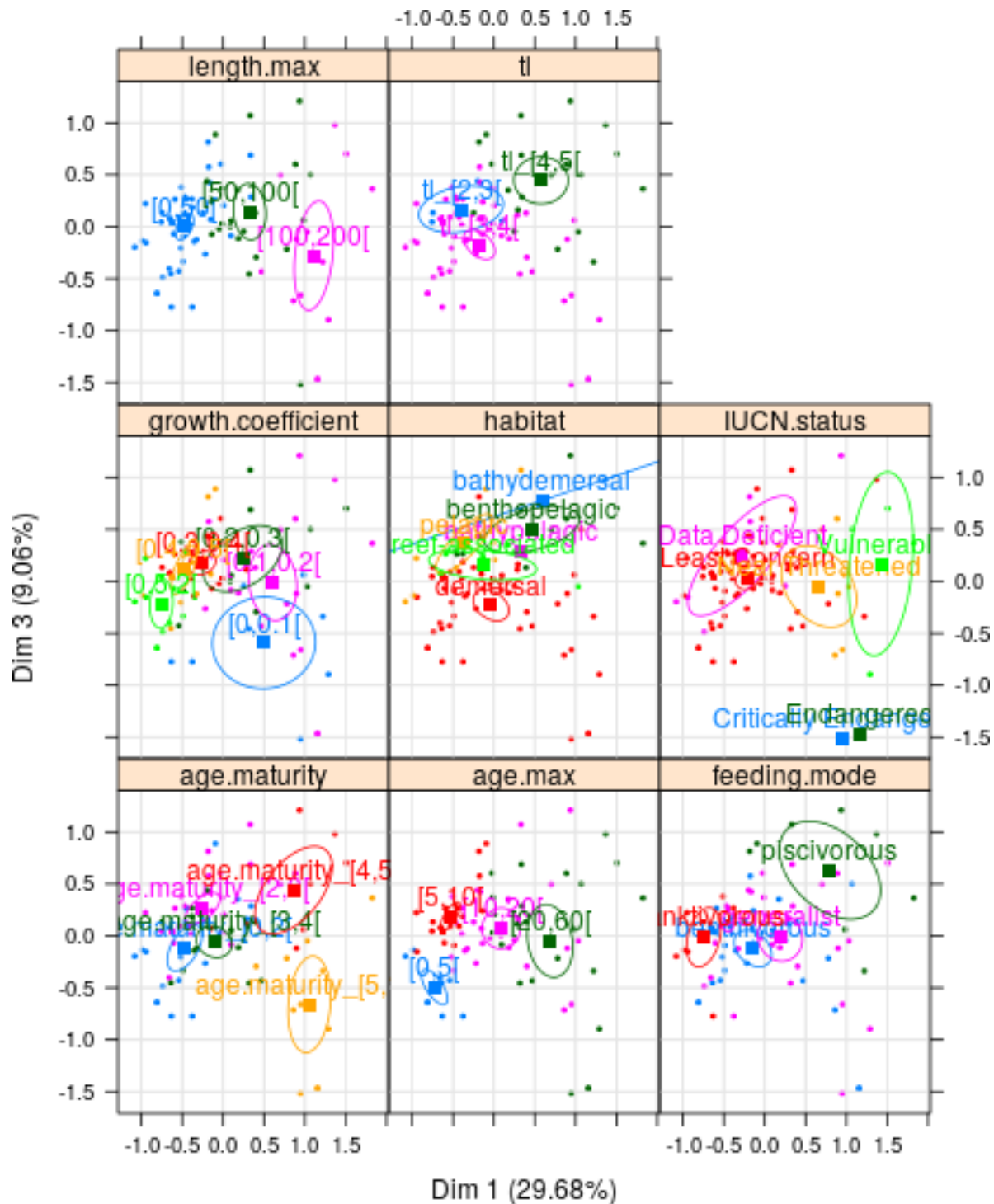
```
## [1] 0 5 10 20 60
```

```
##          nbid nbNA
## Species      84    0
## habitat       6    0
## feeding.mode   4    0
## tl            3    0
## age.maturity   5    0
## growth.coefficient 6    0
## length.max     3    0
## age.max        4    0
## IUCN.status    6    0
```

Multiple correspondance analysis

Multiple correspondance analyses on the trait matrix using Burt tables.





With three axes 48.55% of total variance mapped. Quick analyses: axe one for big fishes with intermediate growth parameters (size, trophic level, age, age maturity) versus small one with low life expectancy, low trophic level and co. Axe two and three not so clear, but help to distinguish the axe one properties.

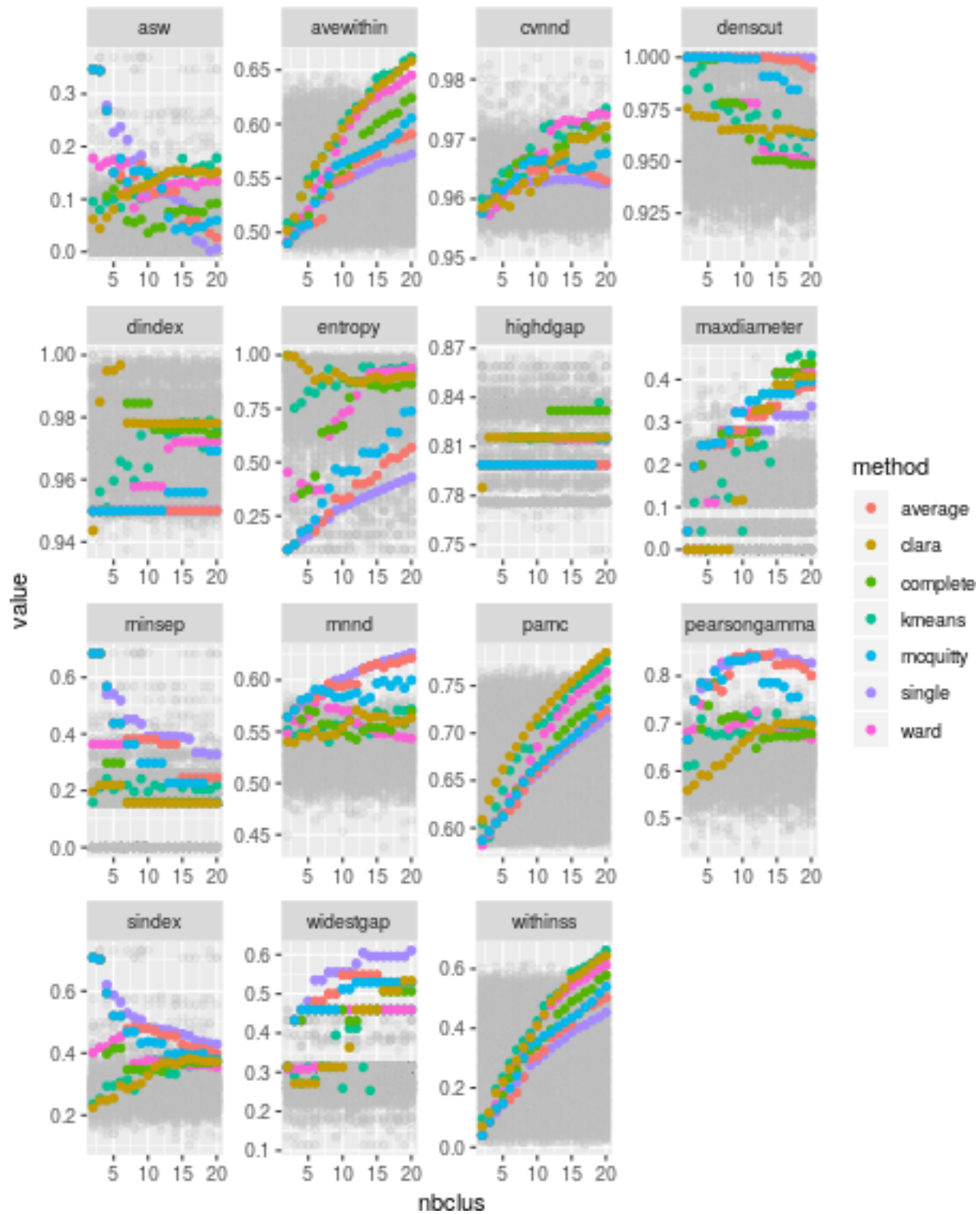
Groups identification in the multiple correspondance analysis spaces

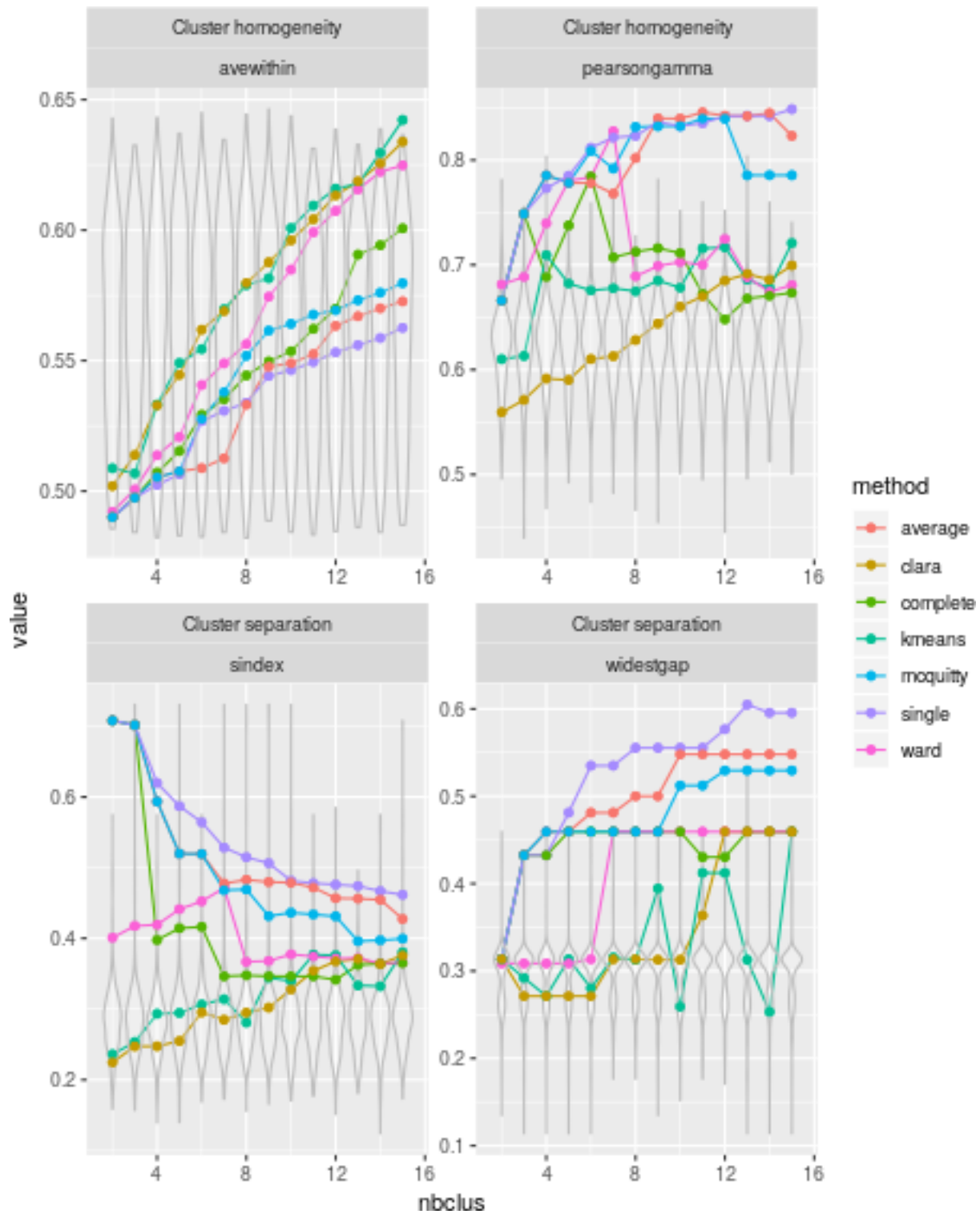
The MCA results are used to identify group of similar traits characteristics in the MCA space. First methods test using Hennig approach and the fpc packages, then group with the best method. Methods tested with 2 to 20 groups:

- 4 random classification (cf “stupid” classifiers in Hennig),
- hierarchical clustering with ward, single, complete, average, mcquitty criterions,
- kmeans,
- clara methods (large dataset computation methods).

This comes from the quick reading of the Hennig paper found on ArXiv (fpc paper, the metric agregation and the cluster strategy and selection to bee data - yes bee). To select the methods and the number of cluster, differents metrics are plotted and interpreted. Two type of metrics: metrics to assess cluster homogeneity and metrics to assess cluster separation. The choice of the metrics has to be in line with the cluster objectives (here to reduce the dimension of the species number in DATRAS, according to some traits similarity). Choice of the measures link to the aim of the clustering. Some random notes taken during the reading of the Hennig’s papers:

- cluster homogeneity:
 - average within-cluster dissimilarities or average distance with cluster (avewithin): individuals within cluster should be functionally more similar than individuals in different cluster (smaller values indicates better clustering quality),
 - correlation between distances and a 0-1 vector where 0 means same cluster, 1 means different clusters (pearsongamma): group of species should represent the functional distance in the MCA space (youhouu, headache incoming), it is a representation of dissimilarity structure by clustering (a guess: higher is better?). Measure of the quality of the cluster representatiion “in some way”.
- cluster seperation
 - separation index, based on the distance for every point to closest point not in the same cluster (sindex): individuals should be functionaly separated from other species (smaller values are better)
 - widest within-cluster gap (widestgap): assess connectivity (no gap in cluster : smaller value are better)



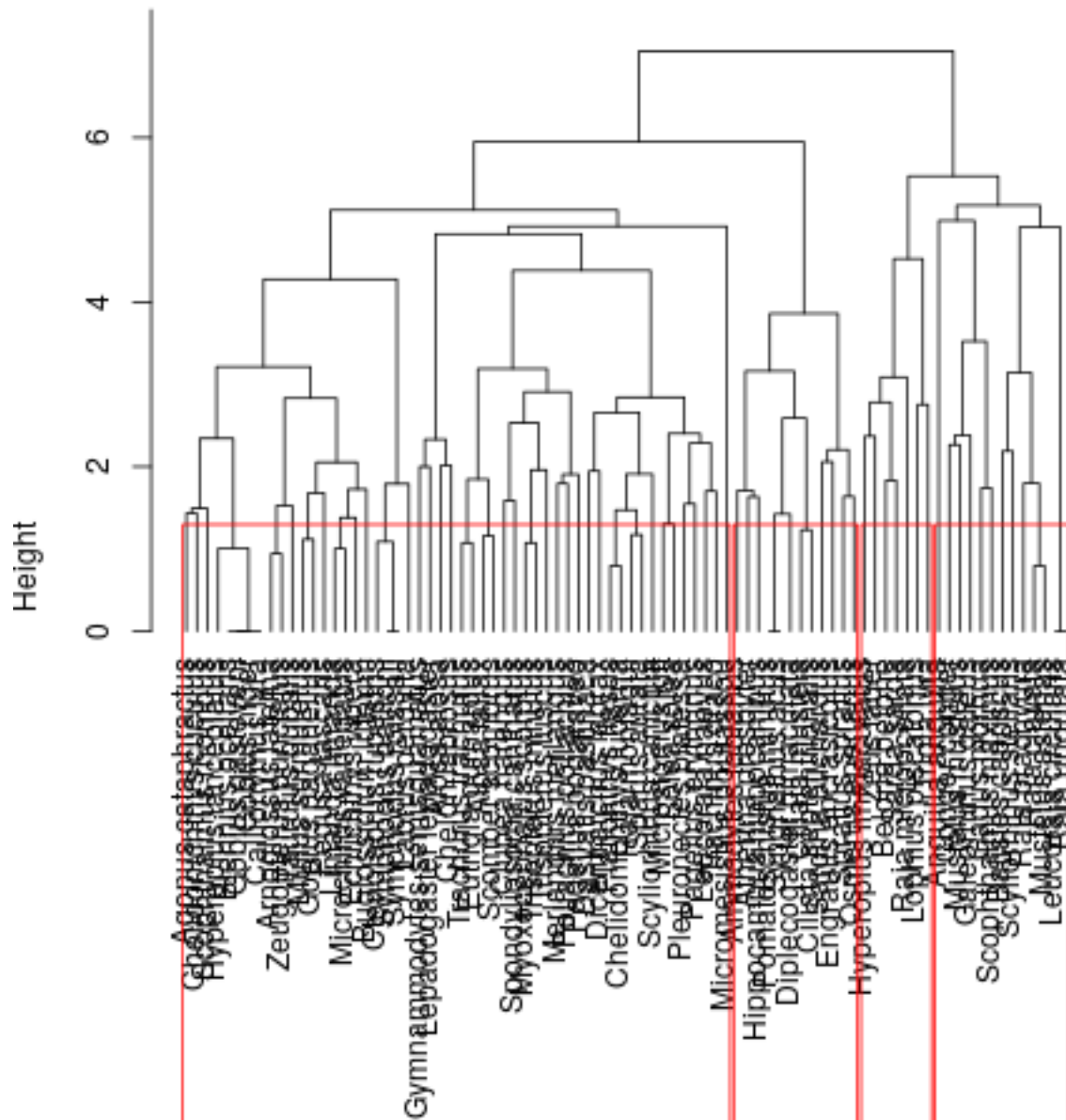


Results :

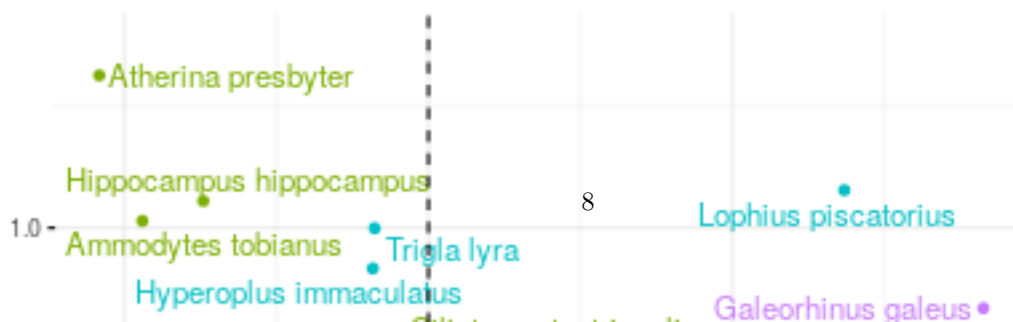
- 7 clusters! 4 to 5 should be better

Final temporary results and Representation for interpretation

Dendrogram of cluster::agnes(x = rez\$ind\$coord, method = "ward", par.method = 1)



rez\$ind\$coord
Agglomerative Coefficient = 0.79
Individuals - MCA



Individuals - MCA



Nice plot but impossible to include in the markdown for now...



Rasteriiiiiize

