



UNIVERSITAT DE
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Cluster analysis

Píndoles d'estadística avançada
STeL (Març 2021)
Sessió 3

PROF. S. CIVIT

Partitioning Around Medoids (PAM) (1/4)

Function `pam()`, simplified format:
`pam(x, k)`

x: possible values includes:

- **Numeric data matrix or numeric data frame**: each row corresponds to an observation, and each column corresponds to a variable.
- **Dissimilarity matrix**: in this case **x** is typically the output of `daisy()` or `dist()`

k: The number of clusters

Partitioning Around Medoids (PAM) (2/4)

Example: analysis of the file `paises.txt`

```
countries<-read.table("Paises.txt",sep="\t",header=TRUE)

# Compute Mahalanobis or any other distance (using dist(),
# vegdist() function)
library(vegan)
countries.MAH<-vegdist(countries[,-1], method="mahalanobis")

# Compute PAM with k = 2
library(cluster)
pam.res.2 <- pam(countries.MAH, 2, diss = TRUE)

# medoids: Objects that represent clusters
pam.res.2$medoids
```

Partitioning Around Medoids (PAM) (3/4)

Extract clustering vectors

pam.res.2\$cluster

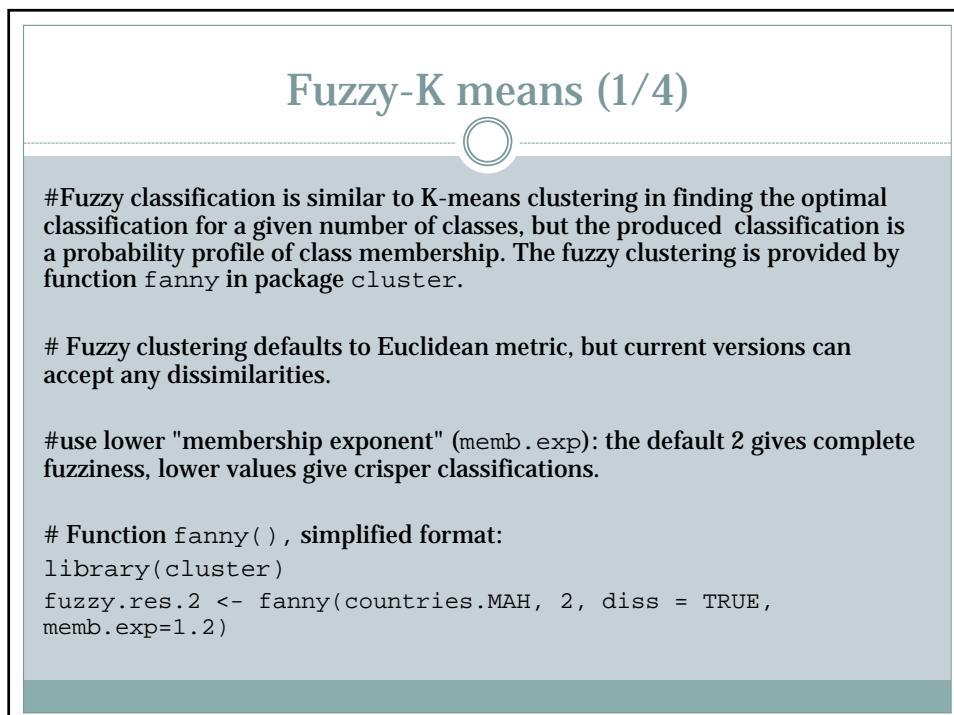
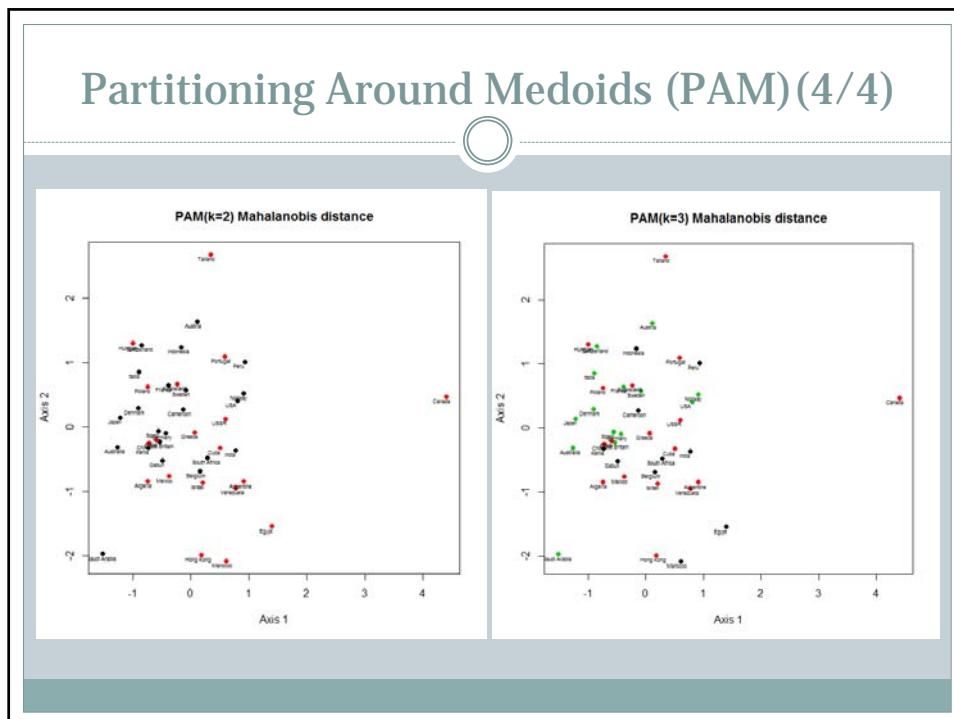
Extract Silhouette average

pam.res.2\$silinfo\$avg.width

The result can be plotted using the function cmdscale() in cluster package.

mds.pam.res.2 <- cmdscale(as.dist(countries.MAH), eig=TRUE)

```
plot(mds.pam.res.2$points[,1], mds.pam.res.2$points[,2],
main="PAM(k=2) Mahalanobis distance", xlab="Axis 1", ylab="Axis 2",
col=pam.res.2$cluster, pch=19)
text(mds.pam.res.2$points[,1], mds.pam.res.2$points[,2], labels=
countries[,1], pos=1, cex=0.5, offset=0.15)
```



Fuzzy-K means (2/4)

```
#The function returns an object with the following items:
names(fuzzy.res.2)
"membership" "coeff" "memb.exp" "clustering" "k.crisp" "objective"
"convergence" "diss" "call" "silinfo"

# membership is the probability profile of belonging to a certain class, #clustering
is the most likely crisp classification.
fuzzy.res.2$membership
[,1]      [,2]
[1,] 0.5177944 0.4822056
[2,] 0.6067280 0.3932720
[3,] 0.4546626 0.5453374
[4,] 0.3385004 0.6614996
[5,] 0.4812201 0.5187799
...
fuzzy.res.2$clustering
fuzzy.res.2$clustering
[1] 1 1 2 2 2 1 2 2 2 1 1 1 1 2 1 2 2 1 1 2 1 2 2 1 1 1 2 2 1 1 1 2 2 1 1
[39] 2 1 1
```

Fuzzy-K means (3/4)

#To summarise, the *interpretation* of 2D representation we put the correlation between **cmdscale axes** and **original variables**

```
Axis1<-round(cor(mds.fuzzy.res.2$points[,1], countries[,2:11]),3)
Pob    PIB     Urb   Analf Estud Vida     Nutric ContInd ContVeh SecPrim
0.543 -0.017 -0.291  0.279  0.083 -0.221 -0.241   0.047 -0.024   0.121

Axis2<-round(cor(mds.fuzzy.res.2$points[,2], countries[,2:11]),3)
Pob    PIB     Urb   Analf Estud Vida     Nutric ContInd ContVeh SecPrim
0.016 -0.781 -0.238  0.267 -0.172 -0.344 -0.042 -0.526 -0.548   0.469
```

Fuzzy-K means (4/4)

#It is difficult to show the fuzzy results graphically, but here is one idea:
rgb function (**easy for 3 groups**) to show colors (mixed colors)
corresponding to the given intensities (between 0 and max) of the red,
green and blue primaries.

```
mds.fuzzy.res.3 <- cmdscale(as.dist(countries.MAH), eig=TRUE)

plot(mds.fuzzy.res.3$points[,1], mds.fuzzy.res.3$points[,2],
main="Fuzzy(k=3) Mahalanobis distance", xlab="Axis 1", ylab="Axis
2", col=rgb(fuzzy.res.3$membership[,1], fuzzy.res.3$membership[,2],
fuzzy.res.3$membership[,3]), pch=19)
text(mds.fuzzy.res.3$points[,1], mds.fuzzy.res.3$points[,2], labels=
countries[,1], pos=1, cex=0.5, offset=0.15)
```