

# Segmentation

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October 1, 2018

## Objectifs

Classification des données en utilisant les algorithmes de K-means et CAH.

Application de K-means sur les données :

```
setwd("C:/")
redwine=read.table("winequality-red.csv", header=T, sep=";", dec=".", na.string="")
summary(redwine)
```

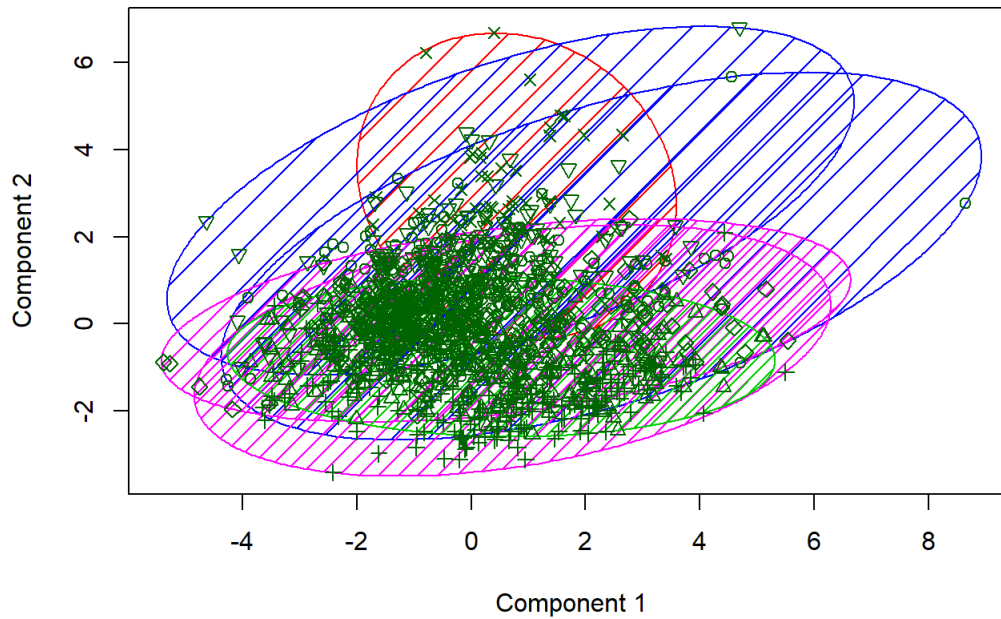
```
## fixed.acidity    volatile.acidity    citric.acid    residual.sugar
## Min.      : 4.60    Min.      :0.1200    Min.      :0.000    Min.      : 0.900
## 1st Qu.: 7.10    1st Qu.:0.3900    1st Qu.:0.090    1st Qu.: 1.900
## Median : 7.90    Median :0.5200    Median :0.260    Median : 2.200
## Mean      : 8.32    Mean      :0.5278    Mean      :0.271    Mean      : 2.539
## 3rd Qu.: 9.20    3rd Qu.:0.6400    3rd Qu.:0.420    3rd Qu.: 2.600
## Max.      :15.90    Max.      :1.5800    Max.      :1.000    Max.      :15.500
## chlorides        free.sulfur.dioxide    total.sulfur.dioxide
## Min.      :0.01200    Min.      : 1.00      Min.      : 6.00
## 1st Qu.:0.07000    1st Qu.: 7.00      1st Qu.: 22.00
## Median :0.07900    Median :14.00      Median : 38.00
## Mean      :0.08747    Mean      :15.87      Mean      : 46.47
## 3rd Qu.:0.09000    3rd Qu.:21.00      3rd Qu.: 62.00
## Max.      :0.61100    Max.      :72.00      Max.      :289.00
## density          pH          sulphates          alcohol
## Min.      :0.9901    Min.      :2.740    Min.      :0.3300    Min.      : 8.40
## 1st Qu.:0.9956    1st Qu.:3.210    1st Qu.:0.5500    1st Qu.: 9.50
## Median :0.9968    Median :3.310    Median :0.6200    Median :10.20
## Mean      :0.9967    Mean      :3.311    Mean      :0.6581    Mean      :10.42
## 3rd Qu.:0.9978    3rd Qu.:3.400    3rd Qu.:0.7300    3rd Qu.:11.10
## Max.      :1.0037    Max.      :4.010    Max.      :2.0000    Max.      :14.90
## quality
## Min.      :3.000
## 1st Qu.:5.000
## Median :6.000
## Mean      :5.636
## 3rd Qu.:6.000
## Max.      :8.000
```

```
rw1=redwine[,1:11]
kml=kmeans(rw1,6)
table(redwine$quality,kml$cluster)
```

```
##
##      1      2      3      4      5      6
## 3      0      1      6      0      3      0
## 4      7     12     21      1      8      4
## 5    114    138    133     60    120    116
## 6    105    179    154      6    148     46
## 7      15      52      77      2     42     11
## 8       8       1       3       9      0       3       2
```

```
library(cluster)
clusplot(rw1,kml$cluster,lines=0,color=T,shade=T,main=paste('Visualisation des clusters k-means'))
```

## Visualisation des clusters k-means

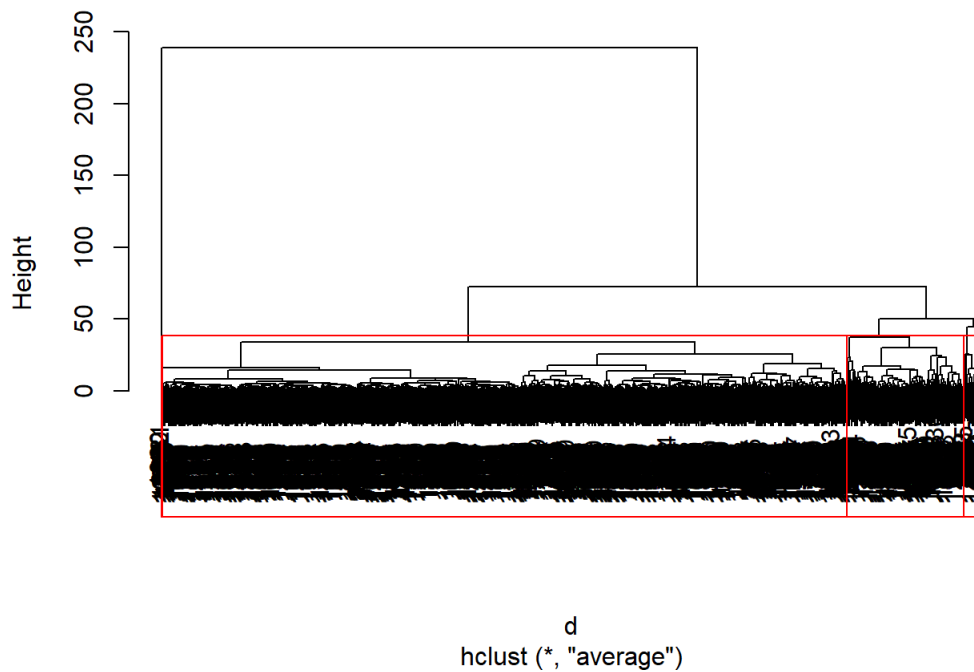


These two components explain 45.68 % of the point variability.

Application de la classification ascendante hiérarchique sur les données :

```
d=dist(rwl,"euclidean")
hc=hclust(d, method="average")
plot(hc)
rect.hclust(hc, k=6)
```

## Cluster Dendrogram

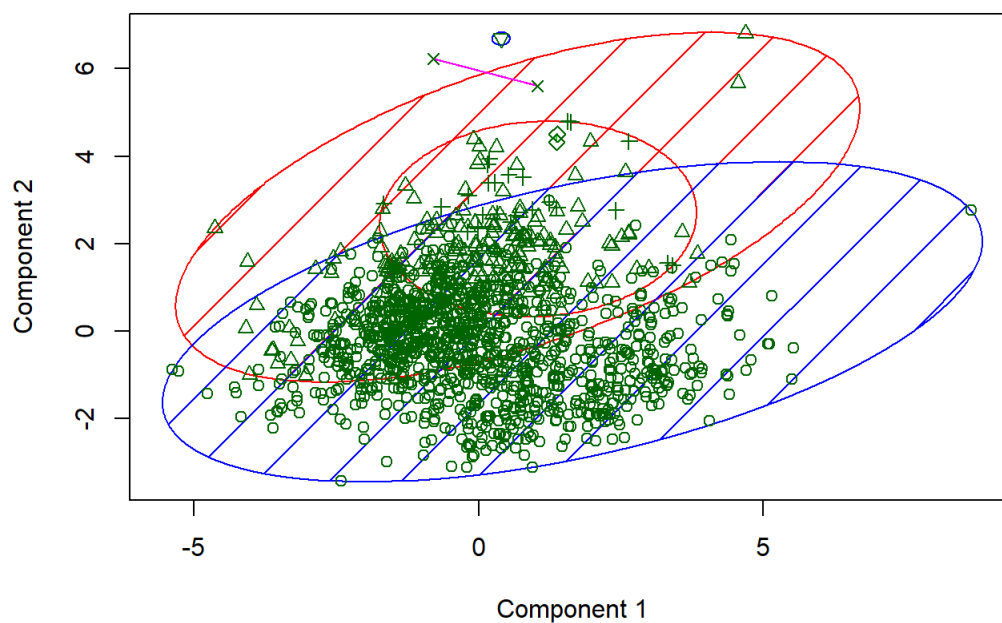


```
groupes=cutree(hc,6)
table(redwine$quality,groupes)
```

```
##      groupes
##      1  2  3  4  5  6
##  3  10  0  0  0  0  0
##  4  48  5  0  0  0  0
##  5 497 155 26  3  0  0
##  6 580 52  5  0  0  1
##  7 184 13  0  0  2  0
##  8  16  2  0  0  0  0
```

```
library(cluster)
clusplot(rw1, groupes, lines = 0, color=T, shade=T, main = paste('Visualisation des clusters CAH'))
```

### Visualisation des clusters CAH



These two components explain 45.68 % of the point variability.

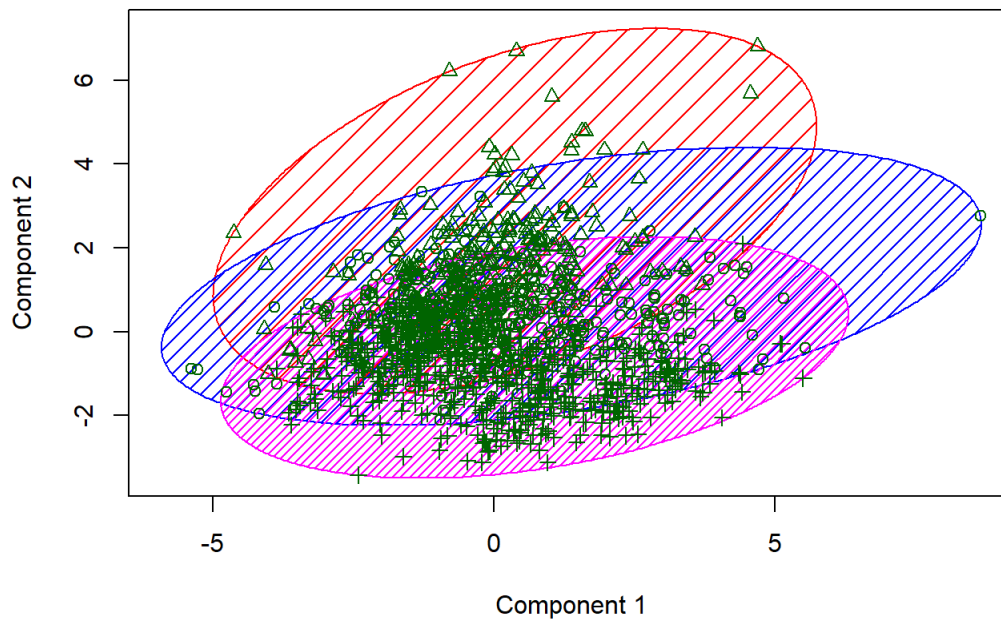
Application de K-means sur les données avec 3 groupes :

```
rw2=redwine
rw2[which(rw2$quality>6), "review"]='fine'
rw2[which(rw2$quality %in% c(5,6)), "review"]='average'
rw2[which(rw2$quality<5), "review"]='bad'
km2=kmeans(rw1,3)
table(rw2$review, km2$cluster)
```

```
##
##      1  2  3
## average 470 220 629
## bad      18  5  40
## fine     56 16 145
```

```
library(cluster)
clusplot(rw1, km2$cluster, lines=0, color=T, shade=T, main=paste('Visualisation des clusters k-means'))
```

## Visualisation des clusters k-means



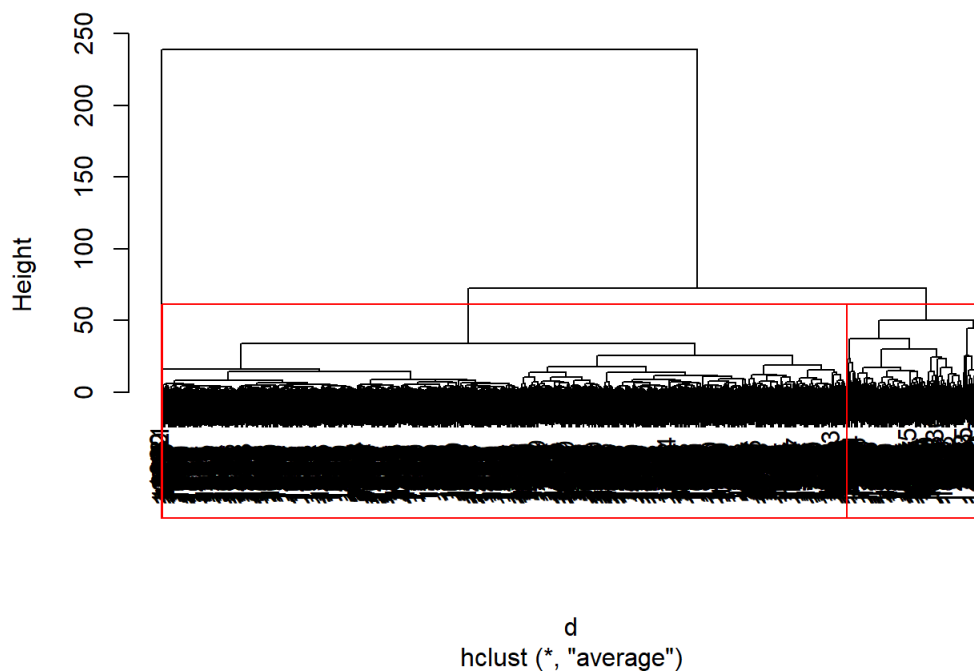
Application de la classification ascendante hiérarchique sur les données avec 3 groupes :

```
d=dist(rw1, "euclidean")
hc=hclust(d, method="average")
hc
```

```
##
## Call:
## hclust(d = d, method = "average")
##
## Cluster method   : average
## Distance         : euclidean
## Number of objects: 1599
```

```
plot(hc)
rect.hclust(hc, k=3)
```

## Cluster Dendrogram

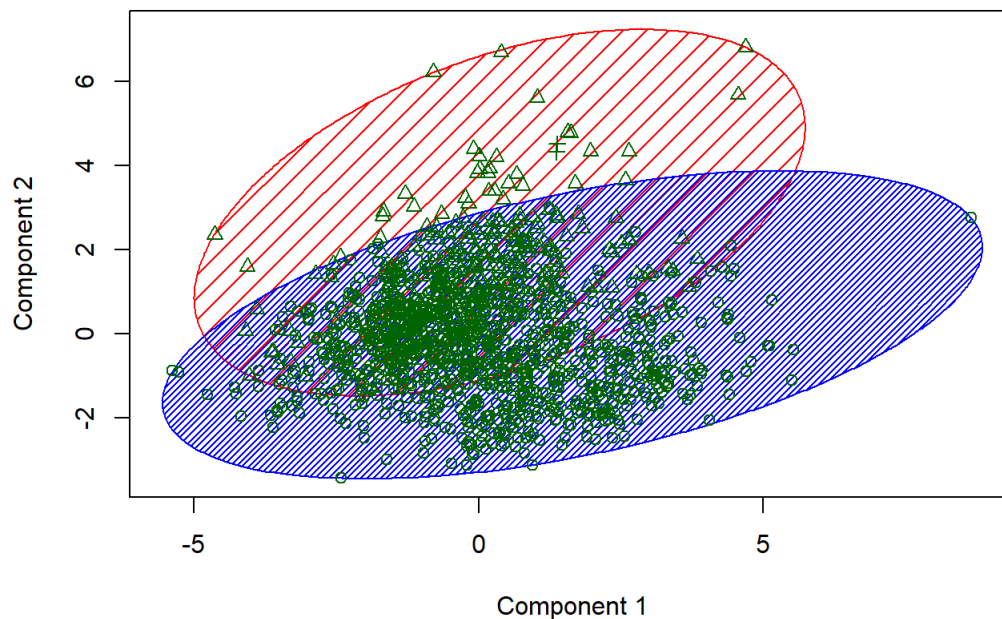


```
groupes=cutree(hc,3)
table(rw2$review,groupes)
```

```
##          groupes
##           1    2    3
## average 1077  242    0
## bad      58    5    0
## fine    200   15    2
```

```
library(cluster)
clusplot(rw1, groupes, lines = 0, color= T, shade= T, main = paste('Visualisation des clusters CAH'))
```

### Visualisation des clusters CAH



These two components explain 45.68 % of the point variability.