

QUIZ1 ADEI_20-21Q1: Template for solutions to questions

Júlia Gasull

November 10th, 2020

List of Questions

The data set for this exercise contains 396 observations for the mean daily values of the variables included in the table referred to a wastewater treatment plant. The plant has measures on the quality of the wastewater at the entrance of the plant, they are the variables xxx.e, from here they go to a first decantation process (Primary Treatment) where it is intended that they settle the solids in suspensions. Then it goes into the Biological Reactor. This is the most critical part. Here is a biological mud that literally “lives by eating organic matter.” It is activated by temperature and aeration. This process is carried out by many species of microorganisms. These microorganisms work at different temperatures. If they are not balanced with the composition of the water, they eat each other. The control variables are those that graduate aeration, temperature, recirculation and purge of the bioreactor. This is the most difficult part: if things go well, dirty water and biological mud enter the bioreactor and end up with clean water and colonies of microorganisms that have consumed the organic matter. Then the water goes through a second decanting process where the microorganisms settle because if the water is not aerated they fall to the ground. And then clean can be poured into the river. The purified water, at the end of the process, before being poured into the river must have neither DBO, nor DQO, nor SS nor SSV, absolute zeros are impossible and therefore the current legislation has permissible limits that are not dangerous for the life in rivers. Available variables:

- date id from 1 to number of observations
- dateformatted dd-mm-yy
- datenorm dd/mm/yyyy
- q.e Input Flow
- qb.b Flow after biological reactor
- qr.g Recirculation Flow
- qp.g Purge Flow
- qa.g Air inflow
- fe.e Iron pretreatment
- ph.e Hydrogen potential
- ss.e Input Solid in Suspension
- ssv.e Input Suspended Volatile Solids
- dco.e Input Fraction of degradable organic matter
- dco.e Input BIOdegradable organic matter fraction
- nkt.e Input Hydrogen potential
- nh4.e Input Ammonium concentration
- p.e Input Phosphor concentration
- ph.d Decantation Hydrogen potential at the settler
- ss.d Decantation Solid in Suspension at the settler
- ssv.d Decantation Suspended Volatile Solids at the settler
- dco.d Decantation Fraction of degradable organic matter at the settler
- dco.d Decantation BIOdegradable organic matter fraction at the settler
- nkt.d Decantation Hydrogen potential at the settler
- nh4.d Decantation Ammonium concentration at the settler

- p.d Decantation Phosphor concentration at the settler
- ph.s Output Hydrogen potential
- ss.s Output Solid in Suspension
- ssv.s Output Suspended Volatile Solids
- dgo.s Output Fraction of degradable organic matter
- dbo.s Output BIOdegradable organic matter fraction
- nk.s Unknown
- nh4.s Output Ammonium concentration
- p.s Output Phosphor concentration
- v30.b Biological Volumetric Analysis
- mlss.b Biological Mixed Liquor Suspended Solids
- mlvss.b Biological Volatile solids in suspension liquor mixture
- im.b Unknown
- cm1.b Unknown
- cm2.b Unknown
- mcert.b Biological Cell Age
- trh.c Unknown (non important)
- dbo.dqoe Input Quocient DBO.E into DQO.E
- dbo.dqod Quocient DBO.D into DQO.D at the settler
- dbo.dqos Output Quocient DBO.S into DQO.S
- weekday Day of the week
- season Year season

The data technically correspond to daily measurements and there is a temporal correlation that cannot be dealt with in this subject. You only have to work in this exercise with the data in randomized order. The response variables are considered the fraction of biodegradable organic matter DBO.S, degradable organic matter DQO.S or solids in suspension, either volatile (SSV.S) or not (SS.S) in the OUTPUT of the plant. The response variable DQO.S is initially considered.

Firstly, load dataset and check available variables.

```
# Clear plots
if(!is.null(dev.list())) dev.off()

## null device
##          1

# Clean workspace
rm(list=ls())
```

Load Required Packages for this deliverable

We load the necessary packages and set working directory

```
# Load Required Packages
options(contrasts=c("contr.treatment","contr.treatment"))

requiredPackages <- c("missMDA","chemometrics","mvoutlier","effects","FactoMineR","car", "factoextra","lme4")
missingPackages <- requiredPackages[!(requiredPackages %in% installed.packages()[,"Package"])]

if(length(missingPackages)) install.packages(missingPackages)
lapply(requiredPackages, require, character.only = TRUE)
```

Some useful functions

```
calcQ <- function(x) { # Function to calculate the different quartiles
  s.x <- summary(x)
  iqr<-s.x[5]-s.x[2]
  list(souti=s.x[2]-3*iqr, mouti=s.x[2]-1.5*iqr, min=s.x[1], q1=s.x[2], q2=s.x[3],
       q3=s.x[5], max=s.x[6], mouts=s.x[5]+1.5*iqr, souts=s.x[5]+3*iqr )
}

countNA <- function(x) { # Function to count the NA values
  mis_x <- NULL
  for (j in 1:ncol(x)) {mis_x[j] <- sum(is.na(x[,j])) }
  mis_x <- as.data.frame(mis_x)
  rownames(mis_x) <- names(x)
  mis_i <- rep(0,nrow(x))
  for (j in 1:ncol(x)) {mis_i <- mis_i + as.numeric(is.na(x[,j])) }
  list(mis_col=mis_x,mis_ind=mis_i)
}

countX <- function(x,X) { # Function to count a specific number of appearances
  n_x <- NULL
  for (j in 1:ncol(x)) {n_x[j] <- sum(x[,j]==X) }
  n_x <- as.data.frame(n_x)
  rownames(n_x) <- names(x)
  nx_i <- rep(0,nrow(x))
  for (j in 1:ncol(x)) {nx_i <- nx_i + as.numeric(x[,j]==X) }
  list(nx_col=n_x,nx_ind=nx_i)
}

setwd("~/Documents/uni/FIB-ADEI/Data and Questions for Quiz 1-20201110")
load("~/Documents/uni/FIB-ADEI/Data and Questions for Quiz 1-20201110/WasteWater.RData")
summary(df)
```

```
##      date      dateformatted      datenorm      q.e
## Min.   : 1.00   Length:396      Length:396   Min.   :20500
## 1st Qu.: 99.75   Class :character  Class :character  1st Qu.:38832
## Median :198.50   Mode  :character  Mode  :character  Median :42880
## Mean   :198.50                                     Mean   :41809
## 3rd Qu.:297.25                                     3rd Qu.:45124
## Max.   :396.00                                     Max.   :54089
##
##      qb.b      qr.g      qp.g      qa.g
## Min.   :19883   Min.   :17933   Min.   : 0.0   Min.   : 96451
## 1st Qu.:38498   1st Qu.:40592   1st Qu.: 527.8 1st Qu.:193010
## Median :39000   Median :41920   Median : 653.8 Median :227740
## Mean   :38903   Mean   :40965   Mean   : 621.6 Mean   :231732
## 3rd Qu.:39000   3rd Qu.:42850   3rd Qu.: 689.0 3rd Qu.:273438
## Max.   :52245   Max.   :49527   Max.   :1080.0 Max.   :367840
##
##      fe.e      ph.e      ss.e      ssv.e
## Min.   : 0.00   Min.   :7.200   Min.   : 62.0   Min.   : 19.0
## 1st Qu.:41.67   1st Qu.:7.550   1st Qu.:154.0   1st Qu.:118.9
## Median :48.25   Median :7.600   Median :187.5   Median :146.0
## Mean   :45.39   Mean   :7.619   Mean   :209.7   Mean   :157.5
```

```

## 3rd Qu.:54.35 3rd Qu.:7.700 3rd Qu.:245.2 3rd Qu.:183.2
## Max. :89.80 Max. :8.000 Max. :655.0 Max. :593.0
##
##      dgo.e      dbo.e      nkt.e      nh4.e
## Min. : 27.0 Min. : 69.0 Min. :17.90 Min. : 8.10
## 1st Qu.: 335.8 1st Qu.:155.8 1st Qu.:35.18 1st Qu.: 24.98
## Median : 422.5 Median :197.0 Median :41.39 Median : 27.56
## Mean : 442.4 Mean :213.6 Mean :41.95 Mean : 41.14
## 3rd Qu.: 517.0 3rd Qu.:250.0 3rd Qu.:46.86 3rd Qu.: 31.80
## Max. :1579.0 Max. :987.0 Max. :82.00 Max. :347.00
##
##      p.e      ph.d      ss.d      ssv.d
## Min. : 2.100 Min. :7.100 Min. : 40.00 Min. : 13.00
## 1st Qu.: 5.825 1st Qu.:7.500 1st Qu.: 76.00 1st Qu.: 56.00
## Median : 6.800 Median :7.600 Median : 88.00 Median : 64.00
## Mean : 9.902 Mean :7.563 Mean : 88.72 Mean : 64.91
## 3rd Qu.:16.000 3rd Qu.:7.700 3rd Qu.: 98.12 3rd Qu.: 73.00
## Max. :24.000 Max. :7.900 Max. :192.00 Max. :134.00
##
##      dgo.d      dbo.d      nkt.d      nh4.d
## Min. : 27.0 Min. : 36.0 Min. :15.10 Min. : 8.90
## 1st Qu.:217.0 1st Qu.: 95.0 1st Qu.:30.80 1st Qu.:23.80
## Median :252.5 Median :119.5 Median :35.15 Median :27.03
## Mean :249.6 Mean :120.0 Mean :36.42 Mean :25.94
## 3rd Qu.:286.2 3rd Qu.:142.0 3rd Qu.:40.02 3rd Qu.:28.63
## Max. :538.0 Max. :274.0 Max. :74.00 Max. :37.10
##
##      p.d      ph.s      ss.s      ssv.s
## Min. : 1.500 Min. :7.000 Min. : 2.800 Min. : 1.60
## 1st Qu.: 3.450 1st Qu.:7.400 1st Qu.: 9.275 1st Qu.: 6.50
## Median : 4.369 Median :7.500 Median :12.800 Median : 9.20
## Mean : 6.071 Mean :7.533 Mean :16.511 Mean :12.36
## 3rd Qu.: 9.961 3rd Qu.:7.700 3rd Qu.:18.000 3rd Qu.:14.00
## Max. :14.700 Max. :8.000 Max. :174.800 Max. :134.80
##
##      dgo.s      dbo.s      nk.s      nh4.s
## Min. : 9.00 Min. : 2.00 Min. : 2.00 Min. : 0.500
## 1st Qu.: 34.00 1st Qu.:12.00 1st Qu.: 9.00 1st Qu.: 5.293
## Median : 45.50 Median :17.00 Median :19.52 Median :11.164
## Mean : 51.26 Mean :18.81 Mean :19.23 Mean :12.170
## 3rd Qu.: 63.00 3rd Qu.:22.62 3rd Qu.:26.81 3rd Qu.:18.472
## Max. :163.00 Max. :84.00 Max. :67.00 Max. :31.500
##
##      p.s      v30.b      mlss.b      mlvss.b      im.b
## Min. :0.600 Min. : 77.0 Min. : 754 Min. : 185 Min. : 58.50
## 1st Qu.:1.350 1st Qu.:170.0 1st Qu.:1538 1st Qu.:1180 1st Qu.: 94.65
## Median :1.900 Median :210.0 Median :1760 Median :1343 Median :118.00
## Mean :2.746 Mean :262.7 Mean :1767 Mean :1344 Mean :155.90
## 3rd Qu.:4.757 3rd Qu.:320.0 3rd Qu.:1944 3rd Qu.:1494 3rd Qu.:199.93
## Max. :7.000 Max. :770.0 Max. :3294 Max. :2100 Max. :577.00
##
##      cm1.b      cm2.b      mcrt.b      trh.c
## Min. :0.0200 Min. :0.0500 Min. : 1.780 Min. :2.160
## 1st Qu.:0.5000 1st Qu.:0.2300 1st Qu.: 8.658 1st Qu.:2.417

```

```
## Median :0.6200 Median :0.3200 Median : 10.195 Median :4.845
## Mean :0.6198 Mean :0.3599 Mean : 14.290 Mean :4.196
## 3rd Qu.:0.7400 3rd Qu.:0.4412 3rd Qu.: 12.582 3rd Qu.:4.910
## Max. :1.4300 Max. :3.8600 Max. :341.990 Max. :9.630
##
##      dbo.dqoe      dbo.dqod      dbo.dqos      weekday      season
## Min. :0.1500 Min. :0.1500 Min. :0.0700 Sunday :57 Autumn:101
## 1st Qu.:0.4100 1st Qu.:0.4100 1st Qu.:0.2800 Thursday :56 Spring: 92
## Median :0.4600 Median :0.4717 Median :0.3817 Monday :57 Summer:112
## Mean :0.4788 Mean :0.4832 Mean :0.4068 Tuesday :56 Winter: 91
## 3rd Qu.:0.5400 3rd Qu.:0.5500 3rd Qu.:0.5112 Wednesday:56
## Max. :1.0400 Max. :1.0000 Max. :1.0000 Saturday :57
## Friday :57
```

1. Produeix un conjunt de dades aleatori per destruir la correlació en sèrie.

```
set.seed(10121)
sam<-as.vector(sample(1:nrow(df)))
```

Verificació i emmagatzematge de la mostra

```
head(df)

##      date dateformatted datenorm      q.e      qb.b      qr.g      qp.g      qa.g      fe.e      ph.e
## 227 227      14-IV-96 14/04/1996 37144.6 36643.4 42864.9 699.7 209741 35.7 7.8
## 21 21      21-IX-95 21/09/1995 48650.0 39000.0 41766.0 385.9 263090 0.0 7.5
## 291 291      17-VI-96 17/06/1996 44319.4 43731.4 41484.2 840.6 291791 56.9 7.9
## 354 354      19-VIII-96 19/08/1996 36000.0 35416.1 43516.8 676.3 199481 55.2 7.6
## 241 241      28-IV-96 28/04/1996 41040.0 39000.0 34946.2 642.4 267831 15.7 7.6
## 228 228      15-IV-96 15/04/1996 43200.0 42682.0 42477.0 663.0 237011 34.8 7.7
##      ss.e ssv.e dqo.e dbo.e nkt.e      nh4.e      p.e      ph.d      ss.d      ssv.d      dqo.d      dbo.d
## 227 151 119 588 140 44.64 38.0000 6.560 7.6 63 53 280 70
## 21 154 115 335 257 29.70 18.3000 15.800 7.5 90 64 270 160
## 291 296 228 800 290 56.18 36.6800 7.140 7.6 89 70 329 110
## 354 204 159 498 247 43.10 30.7000 7.800 7.5 87 65 267 118
## 241 410 319 618 294 29.75 17.1375 4.425 7.5 82 62 240 100
## 228 157 126 314 154 42.80 39.2000 6.500 7.6 80 62 305 151
##      nkt.d      nh4.d      p.d      ph.s      ss.s      ssv.s      dqo.s      dbo.s      nk.s      nh4.s      p.s
## 227 43.4600 33.840 4.4600 7.6 6.8 4.0 30 13 28.460 20.0000 1.5000
## 21 30.8000 18.800 14.7000 7.5 33.2 24.4 74 35 8.400 7.5000 7.0000
## 291 50.2600 35.100 4.1000 7.6 17.4 14.0 65 10 28.660 2.9000 2.0800
## 354 38.0000 29.700 4.5000 7.8 15.6 12.1 46 29 19.600 17.6000 2.1000
## 241 28.5625 18.775 2.4375 7.3 16.4 12.4 65 23 6.275 4.1625 0.9625
## 228 43.1000 35.000 4.4000 7.7 8.0 6.6 43 14 26.800 17.4000 1.5000
##      v30.b      mlss.b      mlvss.b      im.b      cm1.b      cm2.b      mcrt.b      trh.c      dbo.dqoe      dbo.dqod
## 227 620 1706 1326 269.0 0.67 0.15 9.65 5.23 0.24 0.25
## 21 130 1877 1310 69.3 0.70 0.60 9.48 2.37 0.77 0.59
## 291 160 2296 1757 69.0 0.63 0.24 7.86 4.38 0.36 0.33
## 354 120 1556 1178 77.0 0.63 0.26 9.68 5.41 0.50 0.44
## 241 160 1574 1240 95.0 0.55 0.24 8.01 4.91 0.48 0.42
## 228 610 1872 1438 322.0 0.75 0.39 10.32 4.49 0.49 0.50
##      dbo.dqos      weekday      season
## 227 0.43 Sunday Spring
## 21 0.47 Thursday Autumn
## 291 0.15 Monday Spring
## 354 0.63 Monday Summer
```

```
## 241      0.35  Sunday Spring
## 228      0.33  Monday Spring
```

```
df<-df[sam,]
```

2. S'han tractat les dades que falten, però algunes NA codificades com a valors 0 encara romanen a fe.e i qp.g i s'han de tractar aplicant eines d'imputació explicades a la classe.

fe.e

```
summary(df$fe.e)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.00  41.67   48.25   45.39   54.35   89.80
```

```
sel_fe.e <- which(df$fe.e == 0)
```

```
df[sel_fe.e, "fe.e"] <- NA
```

```
summary(df$fe.e)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##      6.50  42.90   48.70   47.43   54.80   89.80     17
```

qp.g

```
summary(df$qp.g)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.0   527.8   653.8   621.6   689.0  1080.0
```

```
sel_qp.g <- which(df$qp.g == 0)
```

```
df[sel_qp.g, "qp.g"] <- NA
```

```
summary(df$qp.g)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##     188.0   530.9   654.4   631.2   689.6  1080.0      6
```

imputació

```
library(missMDA)
```

```
vars_to_impute <- names(df)[c(7,9)]
```

```
res.imputation<-imputePCA(df[,vars_to_impute],ncp=1)
```

```
summary(res.imputation$completeObs)
```

```
##      qp.g      fe.e
##  Min.   : 188.0  Min.   : 6.50
## 1st Qu.: 531.5  1st Qu.:43.17
##  Median : 653.8  Median :48.25
##   Mean   : 630.7  Mean   :47.37
## 3rd Qu.: 689.0  3rd Qu.:54.35
##   Max.   :1080.0  Max.   :89.80
```

```
df[,vars_to_impute] <- res.imputation$completeObs
```

```
summary(df$fe.e)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      6.50  43.17   48.25   47.37   54.35   89.80
```

```
summary(df$qp.g)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##     188.0   531.5   653.8   630.7   689.0  1080.0
```

3. Els outliers univariants per a la variable de sortida DQO.S també són presents i han de ser tractats. Fes-ho.

```
summary(df$dqo.s)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      9.00  34.00  45.50   51.26  63.00  163.00
```

```
mean<-mean(df$dqo.s)
```

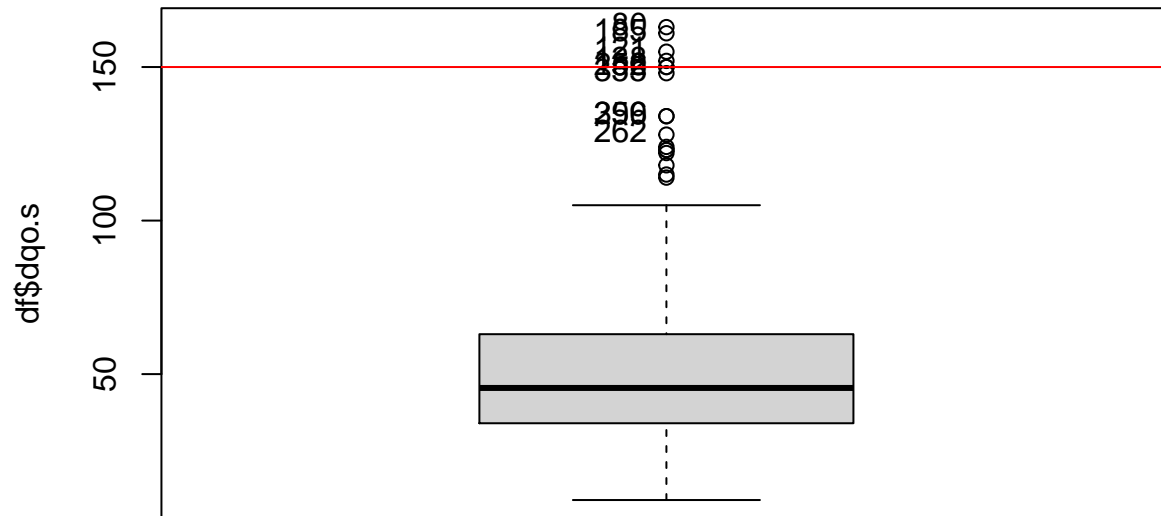
```
Boxplot(df$dqo.s)
```

```
## [1] 80 185 121 138 156 252 355 259 396 262
```

```
var_out<-calcQ(df$dqo.s)
```

```
abline(h=var_out$souts,col="red")
```

```
abline(h=var_out$souti,col="red")
```



```
llout<-which((df$dqo.s>150))
```

```
df[llout,"dqo.s"]<-mean
```

4. Hi ha outliers multivariants? Troba'ls. Intenta explicar la seva singularitat. Els valors atípics multivariants no es tractaran en aquest exercici: mantingueu-los tal com són.

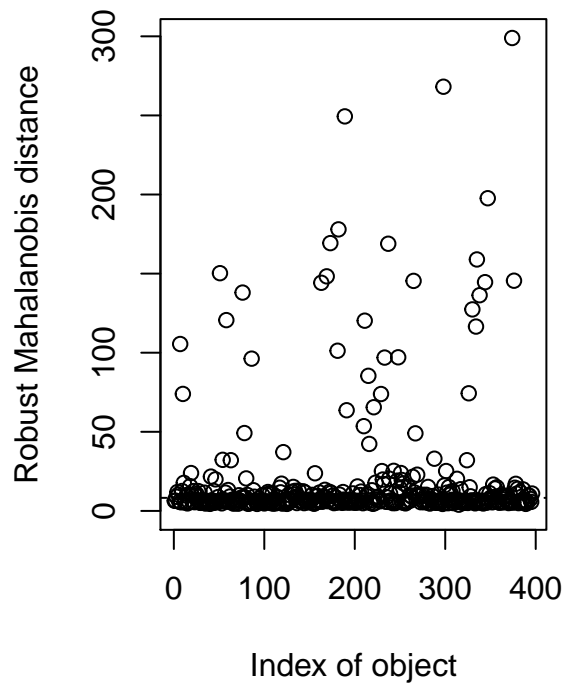
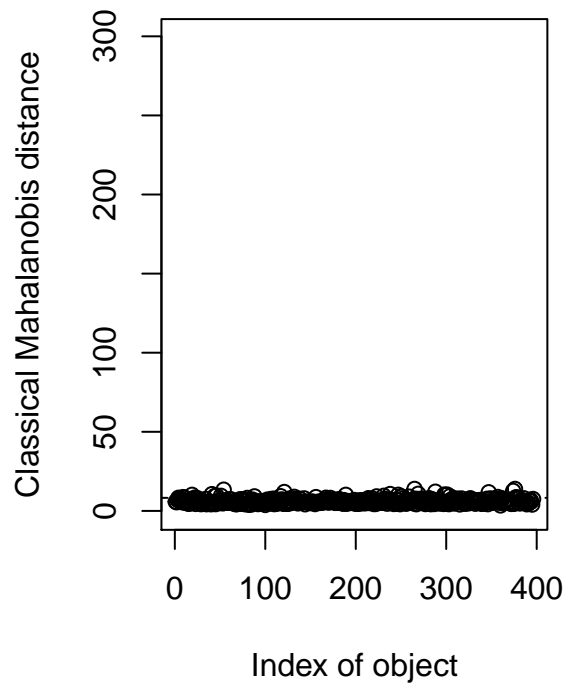
```
library(mvoutlier)
```

```
library(chemometrics)
```

```
names(df)
```

```
## [1] "date"           "dateformatted" "datenorm"      "q.e"           "qb.b"
## [6] "qr.g"           "qp.g"          "qa.g"          "fe.e"          "ph.e"
## [11] "ss.e"           "ssv.e"         "dqo.e"         "dbo.e"         "nkt.e"
## [16] "nh4.e"          "p.e"           "ph.d"          "ss.d"          "ssv.d"
## [21] "dqo.d"          "dbo.d"         "nkt.d"         "nh4.d"         "p.d"
## [26] "ph.s"           "ss.s"          "ssv.s"         "dqo.s"         "dbo.s"
## [31] "nk.s"           "nh4.s"         "p.s"           "v30.b"         "mlss.b"
## [36] "mlvss.b"        "im.b"          "cm1.b"         "cm2.b"         "mcrt.b"
## [41] "trh.c"          "dbo.dqoe"      "dbo.dqod"      "dbo.dqos"      "weekday"
## [46] "season"
```

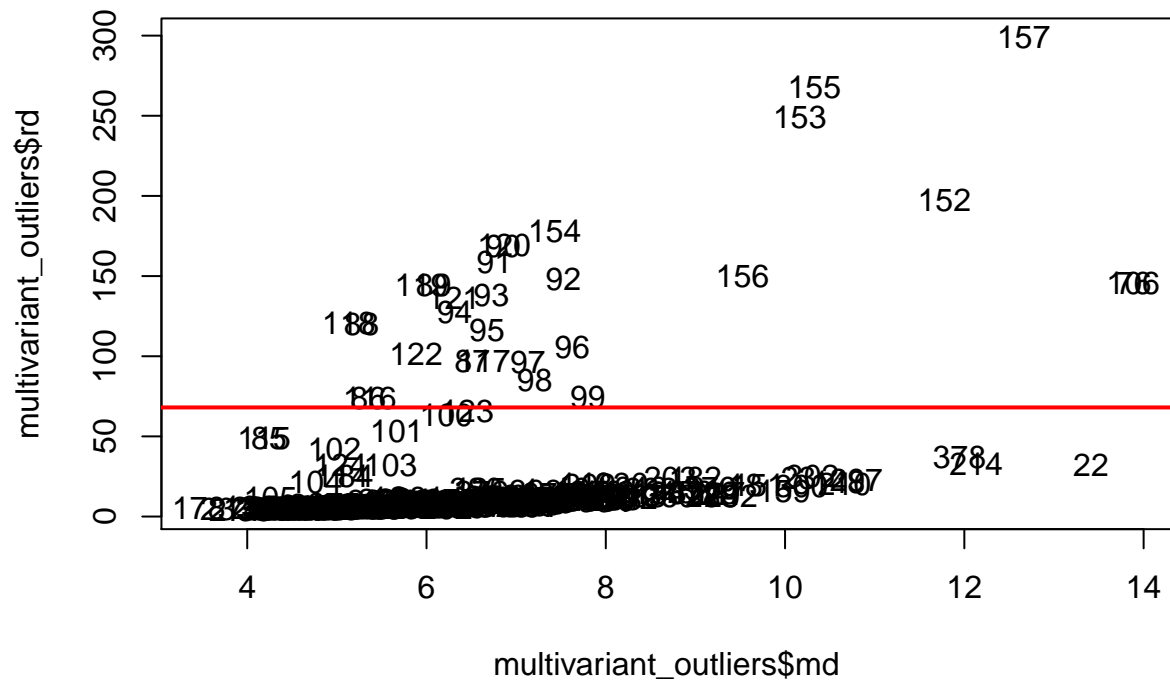
```
multivariant_outliers <- Moutlier(df[, c(4:44)], quantile = 0.995)
```



```
multivariant_outliers$cutoff
```

```
## [1] 8.249408
```

```
par(mfrow=c(1,1))
plot(multivariant_outliers$md, multivariant_outliers$rd, type="n")
text(multivariant_outliers$md, multivariant_outliers$rd, labels=rownames(df[, c(4:44)]))
abline(col="red",lwd=2, h=qchisq(0.995, ncol(df[, c(4:44)])))
```



Singularitat:


```
df[which(row.names(df)=="157"), 1:46]
```

```
##      date dateformatted   datenorm      q.e qb.b   qr.g   qp.g   qa.g   fe.e
## 157 157      4-II-96 04/02/1996 23662.9 22891 18150.3 630.7379 143151 47.3702
##      ph.e ss.e ssv.e dco.e dco.e nkt.e nh4.e p.e ph.d ss.d ssv.d dco.d dco.d
## 157 7.6 77    51 180    73 17.9 8.1 2.1 7.7 40 30 152 54
##      nkt.d nh4.d p.d ph.s ss.s ssv.s dco.s dco.s nk.s nh4.s p.s v30.b mlss.b
## 157 15.1 8.9 1.9 7.6 5.2    4 66    5 2 0.5 1.1 310 2936
##      mlvss.b im.b cm1.b cm2.b mcrt.b trh.c dco.dco dco.dco dco.dcos weekday
## 157 1878 105.6 0.08 0.05 341.99 8.37    0.41    0.36    0.08 Sunday
##      season
## 157 Winter
```

5. Indiqueu-ho mitjançant eines d'anàlisi de dades exploratòries que aparentment són les variables més associades a la variable de resposta (utilitzeu només les variables indicades). Utilitzeu també eines de perfilat FactoMineR.

```
vars_res<-names(df)[c(29)]
vars_quantitatives<-names(df)[c(4:28, 30:44)]
vars_categorical<-names(df)[c(45:46)]

res.condes <- condes(df[, c(vars_res,vars_quantitatives, vars_categorical)],1)
res.condes$quanti
```

```
##      correlation      p.value
## dco.s      0.4132442 9.150194e-18
## ssv.s      0.3789540 5.697798e-15
## ss.s       0.3665570 4.875618e-14
## dco.d      0.2744302 2.844414e-08
## nk.s       0.2177364 1.232463e-05
## im.b       0.2049816 3.958146e-05
## ph.s       0.1960249 8.612797e-05
## ssv.e      0.1772202 3.947384e-04
## nh4.s      0.1737394 5.148898e-04
## nkt.d      0.1713036 6.182922e-04
## dco.e      0.1683242 7.708725e-04
## qr.g       0.1663809 8.884228e-04
## ss.e       0.1489591 2.963754e-03
## qp.g       0.1212841 1.574322e-02
## dco.e      0.1196286 1.723704e-02
## dco.d      0.1196110 1.725356e-02
## nkt.e      0.1179772 1.884911e-02
## ssv.d      0.1155087 2.150337e-02
## ph.d       0.1148294 2.228833e-02
## v30.b      0.1129367 2.460796e-02
## mcrt.b     -0.1431737 4.305904e-03
## mlvss.b    -0.1628865 1.142315e-03
## mlss.b     -0.1885617 1.604054e-04
## dco.dcos   -0.3187916 8.357317e-11
```

```
res.condes$quali
```

```
##      R2      p.value
## season 0.03122189 0.005990612
```

```
res.condes$category
```

```
##           Estimate      p.value
## season=Spring  4.77248 0.040570839
## season=Winter -7.22961 0.001142338
```

6. Definiu els factors polítòmics f.dbo.s, f.dqo.s, f.sst.s (de SSV.S més SS.S) per a les covariables segons el límit legal (DBO 25 mg/l O₂, DQO 125 mg/l O₂ i sòlids totals en suspensió 35 mg/l). Perfil factor f.dqo.s.

Factoritzar:

```
# f.dbo.s DBO 25 mg/l O2
table(df$dbo.s)
```

```
##
##           2           4           5 5.33333333333333
##           2           3           8           1
## 5.66666666666667      6 6.33333333333333      6.5
##           1           7           1           1
## 6.66666666666667      7           7.5           8
##           1           8           1           5
## 8.33333333333333      9           9.7          10
##           1           4           1          23
##          10.5          11          11.5         12
##           2          14           2          18
## 12.3333333333333      12.5          13 13.666666666667
##           2           3           20           1
##          13.7          14          14.5 14.666666666667
##           3          15           5           2
##          14.75          15 15.3333333333333      15.5
##           1          13           1           2
##           16          16.5 16.6666666666667      17
##           22           1           1          13
##          17.3          17.5 17.6666666666667      18
##           2           2           1          17
##          18.5          19          19.5         20
##           4          14           1          13
## 20.3333333333333      21 21.3333333333333      22
##           1          16           1          13
##          22.5          23          23.5         24
##           3          11           2           6
##          24.25          25 25.3333333333333      25.5
##           1           7           1           2
##          26 26.6666666666667      27          28
##           2           1           3           8
##          29          30          32          33
##           7           5           5           1
##          34 34.6666666666667      35          36
##           5           1           8           1
##          37          38          39          40
##           1           3           1           4
##          41          42          43          44
##           1           1           1           1
## 44.6666666666667      45          50          51
```

```
##          1          2          1          1
## 60.3333333333333 76      79      84
##          1          1          1          1
```

```
df$dbo.s[df$dbo.s <= 25] = 0
df$dbo.s[df$dbo.s > 25] = 1
df$f.dbo.s <- factor(df$dbo.s, labels =c("Legal","Illegal"))
```

```
# f.dqo.s DQO 125 mg/l O2
table(df$dqo.s)
```

```
##
##          9          10          17          18
##          4          4          1          10
##         19          20          24          26
##          3          14          1          10
##         27         27.5          28          29
##         24          1          8          1
##        29.3          30          31          32
##          1          10          3          2
##         34          35         35.3          36
##          3          14          2          20
##         37          38          39          40
##         11          5          2          3
##        40.5          41          42         42.5
##          1          2          3          1
##         43          44          45          46
##          3          11          20          11
##         47          48         49.5          50
##          7          6          1          1
##        51 51.2573232323232          52          53
##          1          4          2          9
##         54         54.7          55          56
##         15          1          8          6
##         57          58          59          60
##          5          4          3          2
##         61          62          63          64
##          1          10          10          6
##         65         65.5          66          67
##          5          2          3          2
##         68          69          70          71
##          1          2          4          1
##         72          74          75          76
##          3          3          3          5
##        77.5          78          79          80
##          2          4          4          2
##         81         81.3          82          84
##          2          1          2          2
##         85          87          89          91
##          2          2          2          1
##         93          94          95          96
##          2          1          2          1
##         98         100         103         105
##          2          2          1          1
##        114         115         118         122
```

```
##          1          1          1          1
##        123        124        128        134
##          2          1          1          2
##        148        150
##          1          2
```

```
df$dqo.s[df$dqo.s <= 125] = 0
df$dqo.s[df$dqo.s > 125] = 1
df$f.dqo.s <- factor(df$dqo.s, labels =c("Legal","Illegal"))
```

```
# f.sst.s 35 mg/l
df$sst.s=rowSums(cbind(df$ssv.s,df$ss.s),na.rm=TRUE)
table(df$sst.s)
```

```
##
##          4.8          5          5.2          5.4
##          1          1          1          1
##          6          6.2          6.4          6.6
##          1          1          1          1
##          7.6          8          8.4          8.6
##          1          1          2          1
##          8.8 8.866666666666666 9.2          9.5
##          4          1          5          1
##          9.6          9.8          10         10.8
##          4          2          1          4
##          11.2 11.533333333333333 11.6          12
##          7          1          2          3
##          12.4          12.8          13         13.2
##          2          1          1          5
##          13.4          13.6          13.8 13.866666666666667
##          1          2          1          1
##          14          14.2          14.3          14.4
##          6          1          1          5
##          14.6 14.733333333333333 14.8          15
##          1          1          9          1
##          15.2          15.4          15.6         15.65
##          4          1          4          1
##          15.7          16          16.2          16.4
##          1          5          2          2
##          16.5          16.8          16.9          17
##          1          6          1          4
##          17.2          17.4          17.6          17.7
##          1          2          6          1
##          17.8 17.866666666666666 17.9          18
##          2          1          1          5
##          18.4          18.6          18.8 18.933333333333334
##          3          1          10          1
##          19          19.2          19.3          19.4
##          1          10          2          1
##          19.6          19.8          20         20.4
##          3          3          7          6
##          20.6          21          21.2          21.5
##          3          2          6          1
##          21.55          21.6          22         22.2
##          1          4          3          1
```

##	22.4	22.8	22.9	23.2
##	3	1	1	1
##	23.4	23.46666666666667	23.6	23.7
##	1	1	5	2
##	23.8	24	24.4	24.6
##	2	3	3	2
##	24.8	24.9	25	25.2
##	5	1	1	4
##	25.6	26	26.4	26.8
##	2	4	3	1
##	27.2	27.6	27.7	27.8
##	3	5	1	1
##	28	28.13333333333333	28.4	28.6
##	2	1	5	1
##	28.8	29	29.2	29.5
##	2	2	2	1
##	29.6	30.4	30.6	30.66666666666666
##	1	2	1	1
##	30.8	31	31.1	31.2
##	3	1	1	3
##	31.4	31.6	31.7	32
##	1	3	1	2
##	32.33333333333334	32.4	32.6	32.8
##	1	2	1	2
##	33	33.2	33.6	34
##	1	1	1	7
##	34.2	34.4	34.6	34.8
##	1	5	1	1
##	35.4	35.6	36	36.2
##	1	3	5	1
##	37	37.2	37.6	38
##	1	4	3	3
##	38.4	39.6	39.8	40.7
##	1	1	1	1
##	41.2	42	42.4	44
##	1	2	2	1
##	44.8	46	46.4	47.2
##	2	3	1	2
##	48	48.4	49.2	50
##	2	1	1	3
##	56	57.6	62	67
##	2	1	2	1
##	72.4	74.4	74.8	77.2
##	1	2	1	1
##	79	82.8	101	105.5
##	1	1	1	1
##	107	107.2	118	126.4
##	1	1	1	1
##	129	132	142	154.4
##	1	1	1	1
##	162.8	196.4	233.2	309.6
##	1	1	1	1

```
df$sst.s[df$sst.s <= 35] = 0
df$sst.s[df$sst.s > 35] = 1
df$f.sst.s <- factor(df$sst.s, labels =c("Legal","Illegal"))
```

Profiling de f.dqo.s:

```
vars_res<-names(df)[c(48)]
vars_quantitatives<-names(df)[c(4:44)]
vars_categorical<-names(df)[c(45:47, 50)]

res.catdes <- catdes(df[, c(vars_res,vars_quantitatives, vars_categorical)],1)
res.catdes$test.chi2
```

```
##           p.value df
## f.sst.s 3.610966e-05 1
## f.dbo.s 1.917010e-03 1
```

```
res.catdes$quanti.var
```

```
##           Eta2      P-value
## dqo.s  1.00000000 0.000000e+00
## ssv.s  0.16606401 2.828654e-17
## ss.s   0.16279494 6.173060e-17
## im.b   0.02463311 1.731371e-03
## dbo.s  0.02431149 1.857973e-03
## mlss.b 0.01089606 3.786359e-02
```

```
res.catdes$category
```

```
## $Legal
##           Cla/Mod Mod/Cla  Global    p.value    v.test
## f.sst.s=Legal  99.69040 82.5641 81.56566 0.001006991 3.288566
## f.dbo.s=Legal  99.38272 82.5641 81.81818 0.012259831 2.504576
## f.dbo.s=Illegal 94.44444 17.4359 18.18182 0.012259831 -2.504576
## f.sst.s=Illegal 93.15068 17.4359 18.43434 0.001006991 -3.288566
##
## $Illegal
##           Cla/Mod Mod/Cla  Global    p.value    v.test
## f.sst.s=Illegal 6.8493151 83.33333 18.43434 0.001006991 3.288566
## f.dbo.s=Illegal 5.5555556 66.66667 18.18182 0.012259831 2.504576
## f.dbo.s=Legal  0.6172840 33.33333 81.81818 0.012259831 -2.504576
## f.sst.s=Legal  0.3095975 16.66667 81.56566 0.001006991 -3.288566
```

7. Una anàlisi normalitzada de components principals s'aborda utilitzant com a variables addicionals disponibles el factor disponible i les variables de sortida de xxxx.s.

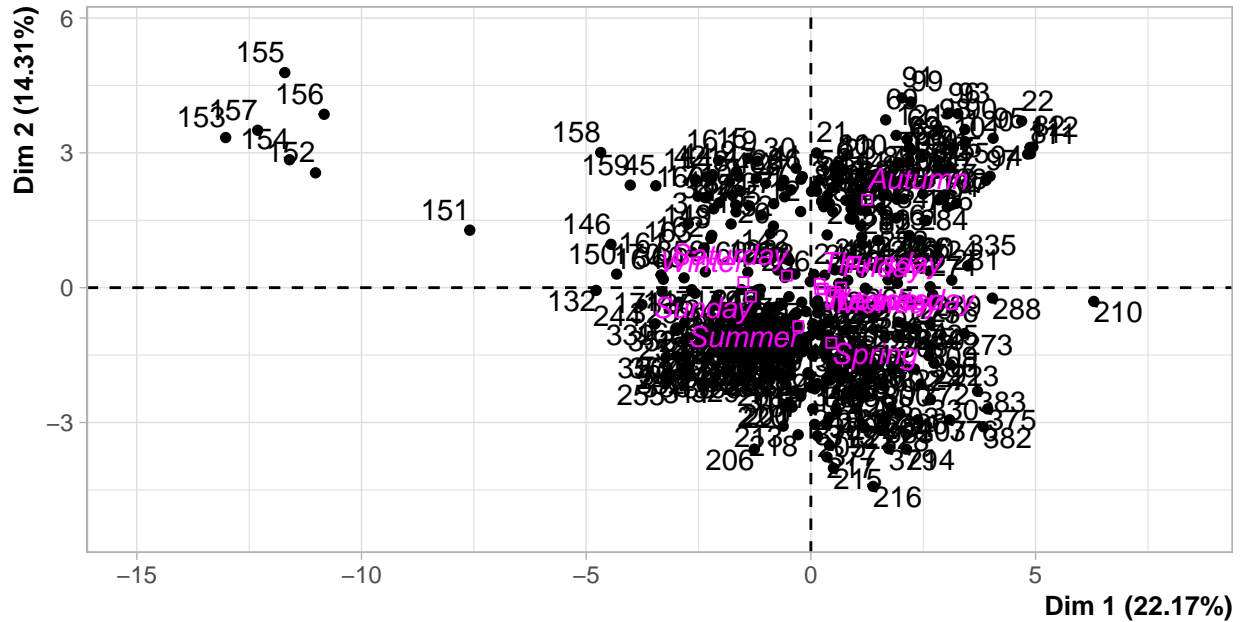
```
names(df)
```

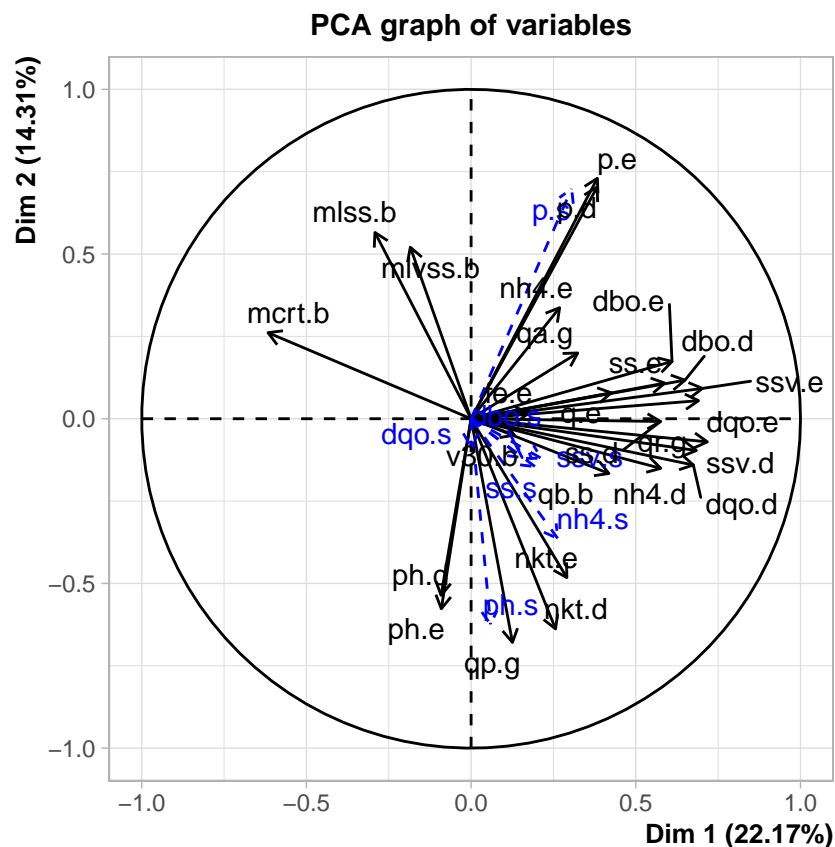
```
## [1] "date"           "dateformatted" "datenorm"      "q.e"           "qb.b"
## [6] "qr.g"           "qp.g"          "qa.g"          "fe.e"          "ph.e"
## [11] "ss.e"           "ssv.e"         "dqo.e"         "dbo.e"         "nkt.e"
## [16] "nh4.e"          "p.e"           "ph.d"          "ss.d"          "ssv.d"
## [21] "dqo.d"          "dbo.d"         "nkt.d"         "nh4.d"         "p.d"
## [26] "ph.s"           "ss.s"          "ssv.s"         "dqo.s"         "dbo.s"
## [31] "nk.s"           "nh4.s"         "p.s"           "v30.b"         "mlss.b"
## [36] "mlvss.b"        "im.b"          "cm1.b"         "cm2.b"         "mcrt.b"
## [41] "trh.c"          "dbo.dqoe"      "dbo.dqod"      "dbo.dqos"      "weekday"
```

```
## [46] "season"      "f.dbo.s"      "f.dqo.s"      "sst.s"      "f.sst.s"
names(df[, c(4:30,32:36,40,45:46)])

## [1] "q.e"      "qb.b"      "qr.g"      "qp.g"      "qa.g"      "fe.e"      "ph.e"
## [8] "ss.e"      "ssv.e"     "dgo.e"     "dbo.e"     "nkt.e"     "nh4.e"     "p.e"
## [15] "ph.d"      "ss.d"      "ssv.d"     "dgo.d"     "dbo.d"     "nkt.d"     "nh4.d"
## [22] "p.d"      "ph.s"      "ss.s"      "ssv.s"     "dgo.s"     "dbo.s"     "nh4.s"
## [29] "p.s"      "v30.b"     "mlss.b"    "mlvss.b"   "mcrt.b"    "weekday"    "season"
res.pca <- PCA(df[, c(4:30,32:36,40,45:46)], quanti.sup=c(23:29), quali.sup=c(34:35))
```

PCA graph of individuals





```
summary(res.pca,nbind=0,nbelements = 25)
```

```
##
## Call:
## PCA(X = df[, c(4:30, 32:36, 40, 45:46)], quanti.sup = c(23:29),
##     quali.sup = c(34:35))
##
##
## Eigenvalues
##
```

	Dim.1	Dim.2	Dim.3	Dim.4	Dim.5	Dim.6	Dim.7
## Variance	5.765	3.721	2.377	2.215	1.663	1.542	1.431
## % of var.	22.175	14.312	9.142	8.520	6.395	5.932	5.503
## Cumulative % of var.	22.175	36.487	45.629	54.149	60.545	66.476	71.979

```
##
```

	Dim.8	Dim.9	Dim.10	Dim.11	Dim.12	Dim.13	Dim.14
## Variance	1.215	1.046	0.722	0.654	0.597	0.499	0.453
## % of var.	4.673	4.024	2.776	2.515	2.295	1.919	1.743
## Cumulative % of var.	76.652	80.675	83.451	85.967	88.262	90.181	91.924

```
##
```

	Dim.15	Dim.16	Dim.17	Dim.18	Dim.19	Dim.20	Dim.21
## Variance	0.414	0.348	0.324	0.214	0.196	0.175	0.131
## % of var.	1.592	1.338	1.244	0.825	0.754	0.674	0.504
## Cumulative % of var.	93.516	94.854	96.099	96.924	97.678	98.352	98.856

```
##
```

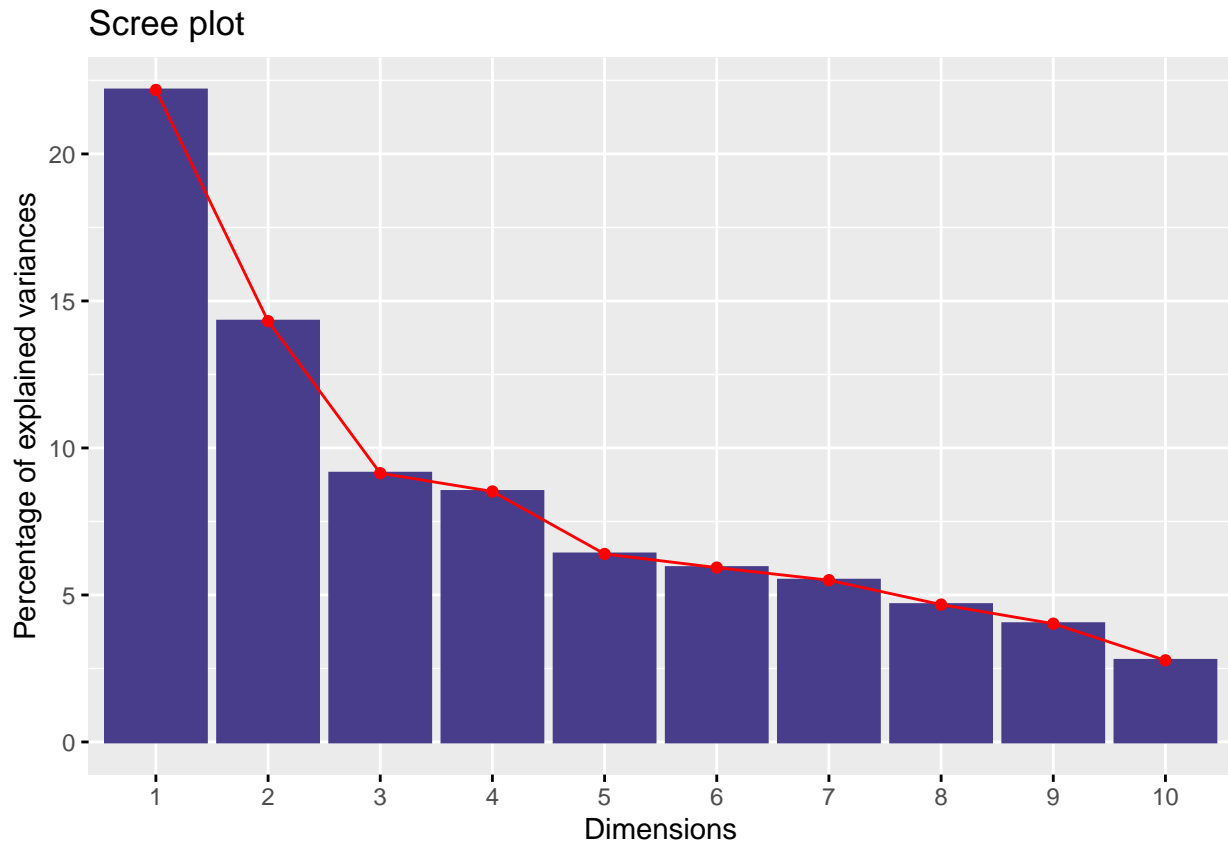
	Dim.22	Dim.23	Dim.24	Dim.25	Dim.26
## Variance	0.105	0.078	0.048	0.041	0.025
## % of var.	0.402	0.302	0.185	0.158	0.097
## Cumulative % of var.	99.258	99.560	99.745	99.903	100.000

```
##
## Variables (the 25 first)
```


[illegible]

```
##          Dim.1  cos2    Dim.2  cos2    Dim.3  cos2
## ph.s      |  0.059 0.004 | -0.624 0.390 |  0.041 0.002 |
## ss.s      |  0.194 0.038 | -0.145 0.021 | -0.093 0.009 |
## ssv.s     |  0.213 0.045 | -0.120 0.014 | -0.098 0.010 |
## dqs.s     | -0.001 0.000 | -0.084 0.007 | -0.027 0.001 |
## dso.s     |  0.155 0.024 | -0.118 0.014 |  0.013 0.000 |
## nh4.s     |  0.263 0.069 | -0.363 0.132 | -0.307 0.094 |
## p.s       |  0.305 0.093 |  0.698 0.488 | -0.356 0.127 |
##
## Supplementary categories
##          Dist    Dim.1  cos2 v.test    Dim.2  cos2 v.test    Dim.3
## Sunday      |  1.974 | -1.333 0.456 -4.524 | -0.199 0.010 -0.840 | -0.928
## Thursday    |  0.457 |  0.196 0.185  0.660 |  0.087 0.036  0.362 |  0.082
## Monday      |  0.674 |  0.540 0.643  1.833 | -0.095 0.020 -0.400 |  0.219
## Tuesday     |  0.452 |  0.266 0.347  0.894 | -0.022 0.002 -0.094 |  0.161
## Wednesday   |  0.577 |  0.209 0.131  0.702 | -0.038 0.004 -0.159 |  0.260
## Saturday    |  0.851 | -0.535 0.396 -1.817 |  0.266 0.098  1.125 | -0.081
## Friday      |  1.012 |  0.668 0.436  2.269 |  0.002 0.000  0.007 |  0.296
## Autumn      |  2.471 |  1.249 0.255  6.049 |  1.955 0.626 11.787 | -0.338
## Spring      |  1.919 |  0.455 0.056  2.073 | -1.227 0.409 -6.956 |  0.794
## Summer      |  1.694 | -0.279 0.027 -1.453 | -0.856 0.255 -5.537 | -0.592
## Winter      |  1.898 | -1.503 0.626 -6.793 |  0.124 0.004  0.698 |  0.301
##
##          cos2 v.test
## Sunday    0.221 -4.903 |
## Thursday  0.032  0.428 |
## Monday    0.105  1.155 |
## Tuesday   0.127  0.840 |
## Wednesday 0.203  1.360 |
## Saturday  0.009 -0.426 |
## Friday    0.086  1.565 |
## Autumn    0.019 -2.546 |
## Spring    0.171  5.628 |
## Summer    0.122 -4.795 |
## Winter    0.025  2.122 |
```

```
fviz_screplot(
  res.pca,
  barfill = "darkslateblue",
  barcolor = "darkslateblue",
  linecolor = "red",
  ggtheme = theme_gray())
```



Quants eixos heu de conservar segons els criteris de Kaiser? 9 dimensions. I segons la regla de Elbow? 5 dimensions. Quina és la inèrcia que expliquen els components principals conservats basats en Kaiser? 80.675%

7. Intenta explicar el significat dels eixos en el primer pla factorial. Quines són les 3 variables amb més correlació amb el primer pla factorial?

```
summary(res.pca,nb.dec=2,nbind=0,nbelements = 25,ncp=2)
```

```
##
## Call:
## PCA(X = df[, c(4:30, 32:36, 40, 45:46)], quanti.sup = c(23:29),
##     quali.sup = c(34:35))
##
##
## Eigenvalues
##
##          Dim.1  Dim.2  Dim.3  Dim.4  Dim.5  Dim.6  Dim.7  Dim.8
## Variance      5.77   3.72   2.38   2.22   1.66   1.54   1.43   1.21
## % of var.     22.17  14.31   9.14   8.52   6.40   5.93   5.50   4.67
## Cumulative % of var. 22.17  36.49  45.63  54.15  60.54  66.48  71.98  76.65
##
##          Dim.9 Dim.10 Dim.11 Dim.12 Dim.13 Dim.14 Dim.15 Dim.16
## Variance      1.05   0.72   0.65   0.60   0.50   0.45   0.41   0.35
## % of var.      4.02   2.78   2.52   2.29   1.92   1.74   1.59   1.34
## Cumulative % of var. 80.68  83.45  85.97  88.26  90.18  91.92  93.52  94.85
##
##          Dim.17 Dim.18 Dim.19 Dim.20 Dim.21 Dim.22 Dim.23 Dim.24
## Variance      0.32   0.21   0.20   0.18   0.13   0.10   0.08   0.05
## % of var.      1.24   0.82   0.75   0.67   0.50   0.40   0.30   0.19
## Cumulative % of var. 96.10  96.92  97.68  98.35  98.86  99.26  99.56  99.75
##
##          Dim.25 Dim.26
```

```

## Variance          0.04  0.03
## % of var.         0.16  0.10
## Cumulative % of var. 99.90 100.00
##
## Variables (the 25 first)
##      Dim.1  ctr  cos2  Dim.2  ctr  cos2
## q.e      |  0.43 3.14 0.18 |  0.08 0.17 0.01 |
## qb.b      |  0.42 3.04 0.18 | -0.17 0.74 0.03 |
## qr.g      |  0.68 8.11 0.47 | -0.10 0.25 0.01 |
## qp.g      |  0.13 0.27 0.02 | -0.68 12.39 0.46 |
## qa.g      |  0.32 1.82 0.10 |  0.20 1.07 0.04 |
## fe.e      |  0.18 0.53 0.03 |  0.02 0.01 0.00 |
## ph.e      | -0.09 0.14 0.01 | -0.58 8.94 0.33 |
## ss.e      |  0.59 6.00 0.35 |  0.11 0.31 0.01 |
## ssv.e     |  0.70 8.55 0.49 |  0.09 0.23 0.01 |
## dqo.e     |  0.69 8.27 0.48 |  0.05 0.08 0.00 |
## dbo.e     |  0.61 6.45 0.37 |  0.17 0.81 0.03 |
## nkt.e     |  0.29 1.47 0.08 | -0.48 6.25 0.23 |
## nh4.e     |  0.27 1.25 0.07 |  0.34 3.05 0.11 |
## p.e       |  0.38 2.55 0.15 |  0.73 14.31 0.53 |
## ph.d      | -0.09 0.14 0.01 | -0.54 7.72 0.29 |
## ss.d      |  0.58 5.74 0.33 | -0.01 0.00 0.00 |
## ssv.d     |  0.72 8.92 0.51 | -0.07 0.13 0.00 |
## dqo.d     |  0.67 7.90 0.46 | -0.14 0.53 0.02 |
## dbo.d     |  0.65 7.31 0.42 |  0.12 0.37 0.01 |
## nkt.d     |  0.26 1.14 0.07 | -0.64 10.94 0.41 |
## nh4.d     |  0.58 5.75 0.33 | -0.15 0.61 0.02 |
## p.d       |  0.38 2.54 0.15 |  0.70 13.30 0.49 |
## v30.b     |  0.13 0.30 0.02 | -0.05 0.07 0.00 |
## mlss.b    | -0.29 1.48 0.09 |  0.57 8.59 0.32 |
## mlvss.b   | -0.18 0.59 0.03 |  0.52 7.28 0.27 |
##
## Supplementary continuous variables
##      Dim.1  cos2  Dim.2  cos2
## ph.s      |  0.06 0.00 | -0.62 0.39 |
## ss.s      |  0.19 0.04 | -0.15 0.02 |
## ssv.s     |  0.21 0.05 | -0.12 0.01 |
## dqo.s     |  0.00 0.00 | -0.08 0.01 |
## dbo.s     |  0.16 0.02 | -0.12 0.01 |
## nh4.s     |  0.26 0.07 | -0.36 0.13 |
## p.s       |  0.31 0.09 |  0.70 0.49 |
##
## Supplementary categories
##      Dist  Dim.1  cos2  v.test  Dim.2  cos2  v.test
## Sunday   |  1.97 | -1.33 0.46 -4.52 | -0.20 0.01 -0.84 |
## Thursday |  0.46 |  0.20 0.18  0.66 |  0.09 0.04  0.36 |
## Monday   |  0.67 |  0.54 0.64  1.83 | -0.09 0.02 -0.40 |
## Tuesday  |  0.45 |  0.27 0.35  0.89 | -0.02 0.00 -0.09 |
## Wednesday | 0.58 |  0.21 0.13  0.70 | -0.04 0.00 -0.16 |
## Saturday | 0.85 | -0.54 0.40 -1.82 |  0.27 0.10  1.12 |
## Friday   |  1.01 |  0.67 0.44  2.27 |  0.00 0.00  0.01 |
## Autumn   |  2.47 |  1.25 0.26  6.05 |  1.96 0.63 11.79 |
## Spring   |  1.92 |  0.46 0.06  2.07 | -1.23 0.41 -6.96 |
## Summer   |  1.69 | -0.28 0.03 -1.45 | -0.86 0.26 -5.54 |

```

```
## Winter      |  1.90 | -1.50  0.63 -6.79 |  0.12  0.00   0.70 |
```

```
ddd<-dimdesc(res.pca,axes=1:2)
```

```
ddd$Dim.1
```

```
## $quanti
```

	correlation	p.value
## ssv.d	0.7172345	8.702867e-64
## ssv.e	0.7022531	4.130053e-60
## dgo.e	0.6903675	2.358761e-57
## qr.g	0.6839592	6.373242e-56
## dgo.d	0.6747353	6.332099e-54
## dbo.d	0.6492347	9.182255e-49
## dbo.e	0.6099533	1.020883e-41
## ss.e	0.5879781	3.452770e-38
## nh4.d	0.5759342	2.294667e-36
## ss.d	0.5750476	3.103914e-36
## q.e	0.4257444	7.242552e-19
## qb.b	0.4183785	3.269946e-18
## p.e	0.3832906	2.630581e-15
## p.d	0.3828651	2.839262e-15
## qa.g	0.3236407	4.155347e-11
## p.s	0.3051301	5.599140e-10
## nkt.e	0.2906850	3.769062e-09
## nh4.e	0.2687437	5.595631e-08
## nh4.s	0.2628777	1.106464e-07
## nkt.d	0.2563545	2.316727e-07
## ssv.s	0.2127715	1.957327e-05
## ss.s	0.1943145	9.952683e-05
## fe.e	0.1750958	4.645130e-04
## dbo.s	0.1550750	1.968511e-03
## v30.b	0.1316599	8.712224e-03
## qp.g	0.1255977	1.237126e-02
## mlvss.b	-0.1845031	2.227597e-04
## mlss.b	-0.2918414	3.248056e-09
## mcrt.b	-0.6165374	7.888993e-43

```
##
```

```
## $quali
```

	R2	p.value
## season	0.17117557	6.918856e-16
## weekday	0.07369764	4.050303e-05

```
##
```

```
## $category
```

	Estimate	p.value
## season=Autumn	1.2684654	6.204938e-10
## weekday=Friday	0.6667511	2.309695e-02
## season=Spring	0.4746323	3.804156e-02
## weekday=Sunday	-1.3345613	4.751263e-06
## season=Winter	-1.4830732	2.711014e-12

```
##
```

```
## attr(,"class")
```

```
## [1] "condes" "list "
```

```
ddd$Dim.2
```

```
## $quanti
```

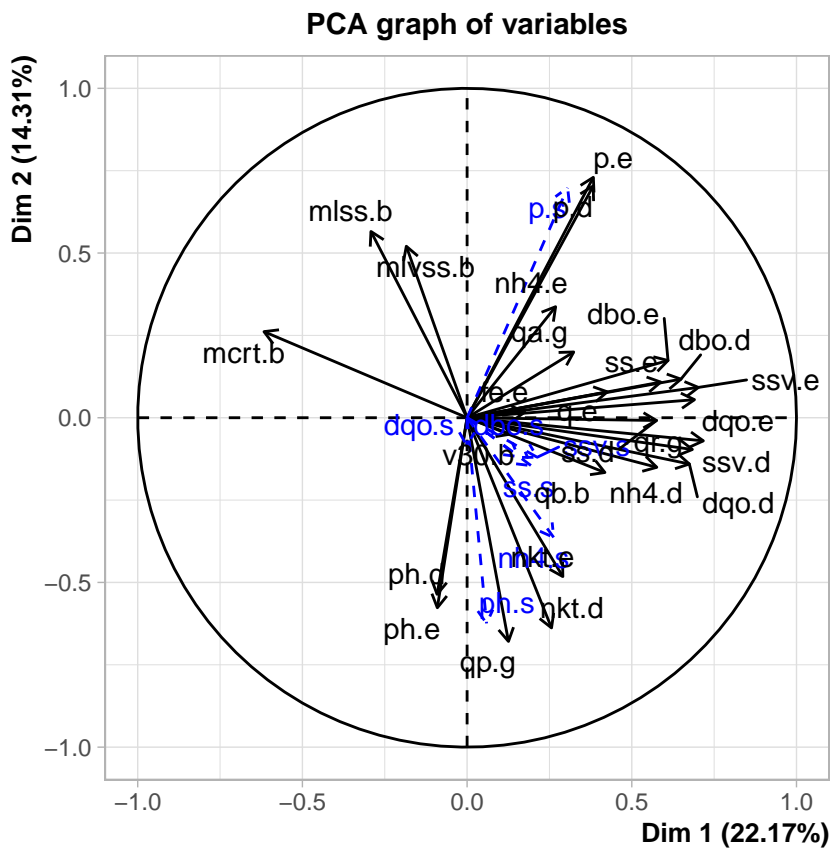
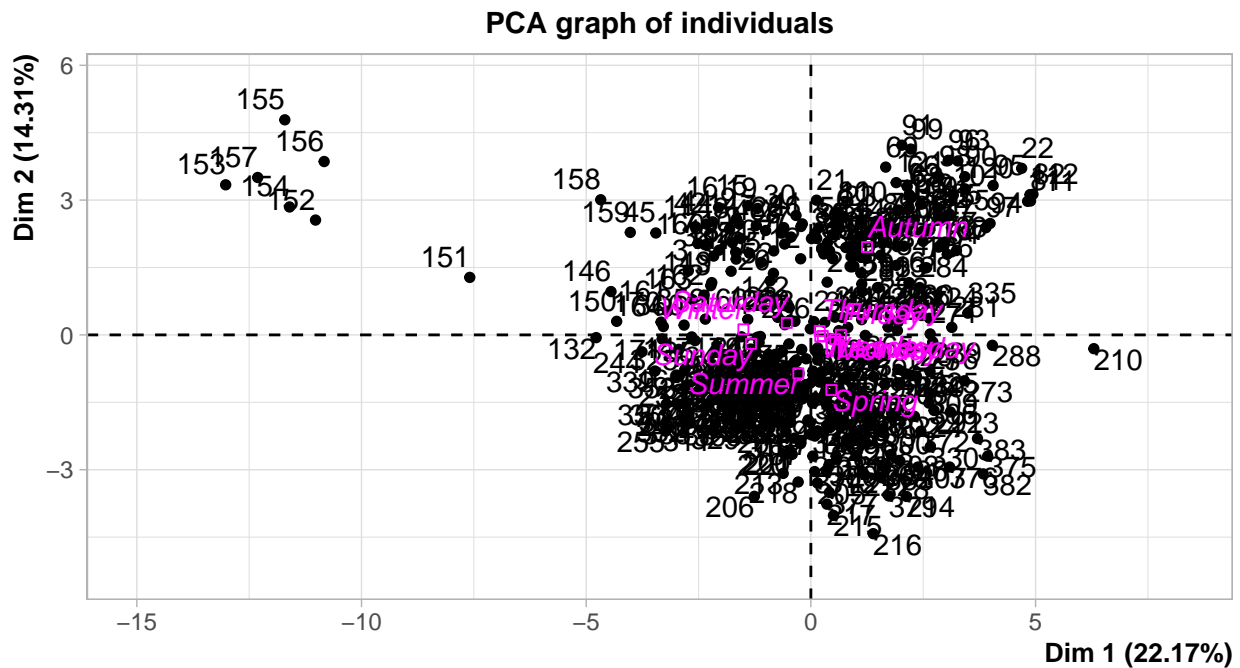
```
##          correlation      p.value
## p.e      0.7298391 4.533983e-67
## p.d      0.7035411 2.036918e-60
## p.s      0.6983003 3.530509e-59
## mlss.b    0.5653911 7.859456e-35
## mlvss.b   0.5205493 7.001826e-29
## nh4.e     0.3370715 5.615938e-12
## mcrt.b    0.2613343 1.320249e-07
## qa.g      0.1995670 6.358950e-05
## dbo.e     0.1738470 5.107139e-04
## dbo.d     0.1170809 1.977781e-02
## ss.e      0.1081635 3.140211e-02
## dbo.s     -0.1179087 1.891867e-02
## ssv.s     -0.1199365 1.695015e-02
## dgo.d     -0.1404357 5.115187e-03
## ss.s      -0.1454264 3.728858e-03
## nh4.d     -0.1504904 2.678851e-03
## qb.b      -0.1662918 8.941924e-04
## nh4.s     -0.3632852 8.461935e-14
## nkt.e     -0.4821699 1.900945e-24
## ph.d      -0.5361076 7.607530e-31
## ph.e      -0.5768834 1.658854e-36
## ph.s      -0.6243612 3.475558e-44
## nkt.d     -0.6380109 1.204814e-46
## qp.g      -0.6790493 7.526410e-55
##
## $quali
##          R2      p.value
## season 0.4126682 5.12791e-45
##
## $category
##          Estimate      p.value
## season=Autumn  1.9561567 5.582386e-39
## season=Summer -0.8548954 1.710921e-08
## season=Spring -1.2263131 7.458427e-13
##
## attr(,"class")
## [1] "condes" "list "
```

Eix 1: * És difícil resumir-lo, però la correlació positiva amb l'eix 1 és + ssv.d + ssv.e * i negativa correlacionada amb * mcrt.b * mlss.b

Eix 2: * Per a l'eix 2, apareix una correlació positiva per a + p.e + p.d * s'associa inversament a + qp.g + nkt.d

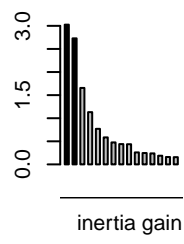
8. A Hierarchical Clustering is addressed. How many clusters are needed to represent 60% of the total inertia.

```
# 6 dimensions have to be selected according to Kaiser's criteria
res.pca <- PCA(df[, c(4:30,32:36,40,45:46)], quanti.sup=c(23:29), quali.sup=c(34:35))
```

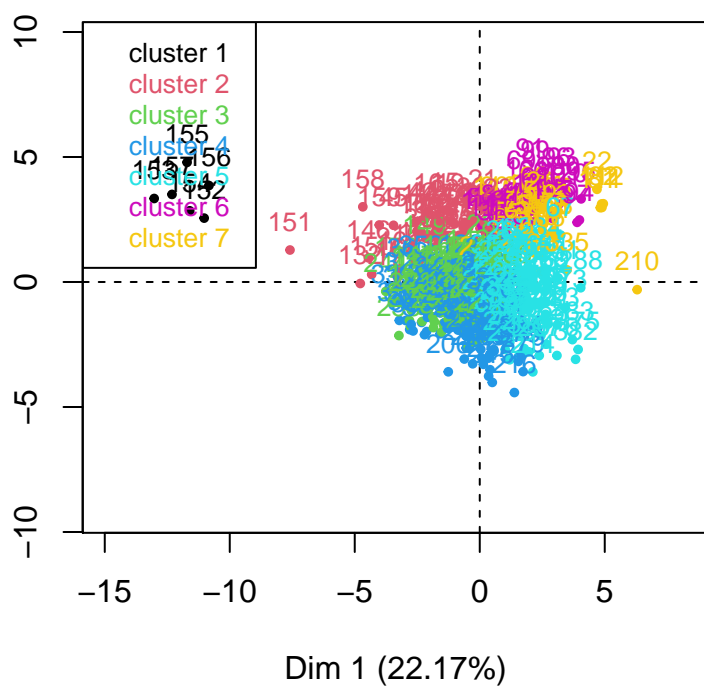
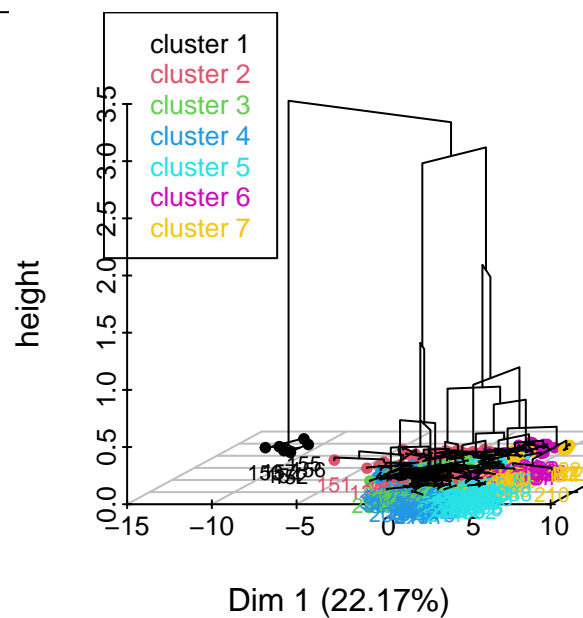
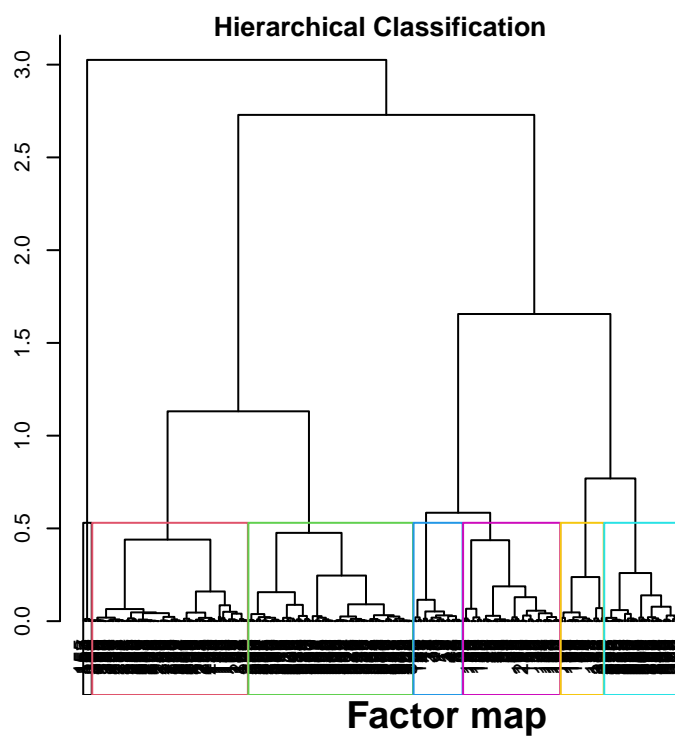


```
res.hcpc <- HCPC(res.pca,nb.clust=7, graph=T)
```

Hierarchical Clustering



Hierarchical clustering on the



```
res.hcpc$desc.var
```

```
##
```



```

## Link between the cluster variable and the categorical variables (chi-square test)
## =====
##           p.value df
## season 1.348225e-67 18
##
## Description of each cluster by the categories
## =====
## $`1`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Winter 6.593407      100 22.9798 0.0001292797 3.827827
##
## $`2`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Autumn 26.73267 45.76271 25.50505 2.550525e-04 3.657133
## season=Spring 0.00000 0.00000 23.23232 3.834159e-08 -5.498325
##
## $`3`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Winter 45.054945 47.674419 22.97980 7.055466e-09 5.789599
## season=Spring 42.391304 45.348837 23.23232 2.146518e-07 5.186180
## season=Summer 4.464286 5.813953 28.28283 9.833661e-09 -5.733573
## season=Autumn 0.990099 1.162791 25.50505 9.561075e-12 -6.812960
##
## $`4`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Summer 60.714286 70.833333 28.28283 1.679165e-24 10.216127
## weekday=Sunday 40.350877 23.958333 14.39394 3.613743e-03 2.910047
## season=Spring 15.217391 14.583333 23.23232 1.871663e-02 -2.351127
## season=Winter 8.791209 8.333333 22.97980 3.097796e-05 -4.166155
## season=Autumn 5.940594 6.250000 25.50505 6.432604e-08 -5.406339
##
## $`5`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Spring 40.217391 51.388889 23.23232 5.736828e-09 5.824254
## weekday=Sunday 3.508772 2.777778 14.39394 5.760943e-04 -3.442627
## season=Autumn 4.950495 6.944444 25.50505 1.480660e-05 -4.331546
##
## $`6`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Autumn 30.6930693 77.5 25.50505 3.501513e-13 7.273554
## season=Summer 0.8928571 2.5 28.28283 1.462911e-05 -4.334200
## season=Spring 0.0000000 0.0 23.23232 1.338267e-05 -4.353753
##
## $`7`
##           Cla/Mod Mod/Cla Global      p.value    v.test
## season=Autumn 30.693069 83.783784 25.50505 5.460010e-15 7.815834
## season=Winter 2.197802 5.405405 22.97980 3.792140e-03 -2.894954
## season=Spring 2.173913 5.405405 23.23232 3.417305e-03 -2.927471
## season=Summer 1.785714 5.405405 28.28283 3.767414e-04 -3.555862
##
## Link between the cluster variable and the quantitative variables
## =====
##           Eta2      P-value

```

```

## mcrt.b 0.93898517 9.914575e-233
## p.e 0.71293982 3.690727e-102
## p.d 0.67450573 1.359173e-91
## p.s 0.67337209 2.663855e-91
## qr.g 0.60623525 1.334174e-75
## qp.g 0.54775435 5.427995e-64
## ssv.e 0.51169712 1.431562e-57
## nh4.e 0.49402239 1.346049e-54
## q.e 0.49086892 4.450176e-54
## qb.b 0.46430530 7.882009e-50
## dgo.e 0.44245360 1.707933e-46
## ss.e 0.42937147 1.464882e-44
## ssv.d 0.42473874 6.910469e-44
## nh4.d 0.37902450 1.588827e-37
## dbo.e 0.37873285 1.738118e-37
## mlss.b 0.37787682 2.261778e-37
## ph.s 0.35355001 3.447846e-34
## dbo.d 0.35326769 3.747544e-34
## dgo.d 0.34878244 1.401556e-33
## nkt.d 0.33062051 2.657144e-31
## ss.d 0.32462719 1.451161e-30
## nkt.e 0.31395131 2.870430e-29
## nh4.s 0.29191383 1.165500e-26
## mlvss.b 0.29071902 1.604821e-26
## ph.e 0.28824569 3.105839e-26
## ph.d 0.22100966 7.779382e-19
## qa.g 0.19084698 9.454592e-16
## v30.b 0.17144284 7.707267e-14
## fe.e 0.12533723 1.579860e-09
## ssv.s 0.03910414 1.609294e-02
## ss.s 0.03420714 3.432497e-02

```

```

##
## Description of each cluster by quantitative variables
## =====

```

```

## $`1`

```

##	v.test	Mean in category	Overall mean	sd in category	Overall sd
## mcrt.b	19.241642	2.582033e+02	1.429049e+01	5.794879e+01	3.124888e+01
## mlss.b	7.934118	2.929833e+03	1.766515e+03	2.365928e+02	3.614446e+02
## mlvss.b	5.512139	1.937167e+03	1.344251e+03	9.663232e+01	2.651639e+02
## ph.s	-2.202465	7.358333e+00	7.533333e+00	1.426437e-01	1.958714e-01
## ph.e	-2.229630	7.491667e+00	7.618813e+00	1.017213e-01	1.405767e-01
## ss.e	-2.253882	1.325000e+02	2.096944e+02	8.287692e+01	8.442999e+01
## p.s	-2.459619	9.750000e-01	2.746086e+00	8.539126e-02	1.775063e+00
## p.d	-2.920306	1.800000e+00	6.071465e+00	6.831301e-02	3.605706e+00
## fe.e	-3.017039	3.483013e+01	4.737019e+01	1.783332e+01	1.024615e+01
## p.e	-3.149663	2.675000e+00	9.902146e+00	3.927998e-01	5.656459e+00
## nh4.s	-3.279402	1.800000e+00	1.216982e+01	8.880691e-01	7.795045e+00
## dbo.e	-3.333698	8.341667e+01	2.136061e+02	7.115938e+00	9.627007e+01
## ss.d	-3.730008	5.891667e+01	8.872096e+01	1.934429e+01	1.969749e+01
## dbo.d	-4.501487	5.066667e+01	1.199558e+02	2.277100e+00	3.794470e+01
## ssv.e	-4.701110	3.716667e+01	1.574583e+02	9.770989e+00	6.307787e+01
## dgo.e	-4.951190	1.101667e+02	4.423813e+02	6.359092e+01	1.654059e+02
## nkt.d	-5.170245	1.973462e+01	3.641604e+01	3.166045e+00	7.953600e+00
## nkt.e	-5.292623	2.210000e+01	4.195316e+01	2.869146e+00	9.246992e+00

```

## qa.g      -5.361405      1.228293e+05 2.317322e+05      1.400397e+04 5.007290e+04
## ssv.d     -6.371728      2.583333e+01 6.490783e+01      1.096079e+01 1.511743e+01
## dqo.d     -6.629825      8.833333e+01 2.496414e+02      5.250608e+01 5.997854e+01
## nh4.d     -8.168626      1.065000e+01 2.594192e+01      1.195478e+00 4.614821e+00
## q.e       -9.284755      2.256380e+04 4.180854e+04      1.066996e+03 5.109559e+03
## qb.b      -10.336716     2.179937e+04 3.890254e+04      1.013377e+03 4.078833e+03
## qr.g      -14.106104     1.813152e+04 4.096489e+04      1.267343e+02 3.990294e+03
##           p.value
## mcrt.b    1.658705e-82
## mlss.b    2.119973e-15
## mlvss.b   3.544987e-08
## ph.s      2.763251e-02
## ph.e      2.577200e-02
## ss.e      2.420360e-02
## p.s       1.390846e-02
## p.d       3.496882e-03
## fe.e      2.552569e-03
## p.e       1.634590e-03
## nh4.s     1.040274e-03
## dbo.e     8.569976e-04
## ss.d      1.914739e-04
## dbo.d     6.747975e-06
## ssv.e     2.587515e-06
## dqo.e     7.376090e-07
## nkt.d     2.337869e-07
## nkt.e     1.205745e-07
## qa.g      8.257733e-08
## ssv.d     1.869103e-10
## dqo.d     3.360847e-11
## nh4.d     3.119218e-16
## q.e       1.620793e-20
## qb.b      4.807277e-25
## qr.g      3.482973e-45
##
## $`2`
##           v.test Mean in category Overall mean sd in category Overall sd
## p.s       8.265379           4.510363      2.746086      1.9182178      1.7750626
## p.d       6.214796           8.766147      6.071465      3.9459771      3.6057065
## p.e       5.855474          13.885026      9.902146      5.3318790      5.6564592
## mlss.b    4.716562          1971.516949 1766.515152    299.4233861    361.4446128
## mlvss.b   4.021792          1472.491525 1344.251263    192.7648407    265.1639251
## q.e       2.752603          43499.827119 41808.539773  4897.0590615  5109.5594150
## ph.d      -3.778421           7.499153      7.562879      0.1703664      0.1402548
## dbo.e     -4.394829          162.728814    213.606061    66.4209306    96.2700653
## dbo.d     -4.410633          99.830508    119.955808    28.5098274    37.9446951
## ph.e      -4.521873           7.542373      7.618813      0.1653998      0.1405767
## nkt.e     -4.560575          36.881959     41.953157     7.0592318     9.2469924
## nkt.d     -4.647924          31.970612     36.416035     4.4723368     7.9535998
## v30.b     -4.714438          187.169492    262.714646    51.3918653    133.2558559
## ss.d      -4.733698          77.508475     88.720960    14.8622179    19.6974930
## nh4.d     -4.790340          23.283573     25.941919     4.6200681     4.6148210
## ss.e      -4.860417          160.347458    209.694444    46.6544505    84.4299924
## dqo.e     -5.321985          336.525424    442.381313    104.5651781    165.4059089
## dqo.d     -5.875723          207.262712    249.641414    52.5820066    59.9785437

```

```

## ssv.e      -5.895824      112.737288      157.458333      36.7601428      63.0778719
## ssv.d      -5.995596      54.008475      64.907828      10.6335301      15.1174319
## ph.s       -6.632244      7.377119      7.533333      0.1690758      0.1958714
## nh4.s      -7.866669      4.795878      12.169823      3.1739990      7.7950446
## qp.g       -11.260392     430.684746     630.737857     82.4954006     147.7411209
##           p.value
## p.s        1.392512e-16
## p.d        5.139134e-10
## p.e        4.756510e-09
## mlss.b     2.398637e-06
## mlvss.b    5.775715e-05
## q.e        5.912358e-03
## ph.d       1.578261e-04
## dbo.e      1.108602e-05
## dbo.d      1.030690e-05
## ph.e       6.129478e-06
## nkt.e      5.101363e-06
## nkt.d      3.352931e-06
## v30.b      2.423789e-06
## ss.d       2.204660e-06
## nh4.d      1.664991e-06
## ss.e       1.171390e-06
## dqo.e      1.026413e-07
## dqo.d      4.210009e-09
## ssv.e      3.728160e-09
## ssv.d      2.027411e-09
## ph.s       3.306222e-11
## nh4.s      3.642084e-15
## qp.g       2.058401e-29
##
## $`3`
##           v.test Mean in category Overall mean sd in category Overall sd
## qb.b      7.596793      4.186259e+04 3.890254e+04 3.736642e+03 4.078833e+03
## ph.e      6.998062      7.712791e+00 7.618813e+00 1.151809e-01 1.405767e-01
## q.e       6.213987      4.484164e+04 4.180854e+04 3.097591e+03 5.109559e+03
## ph.d      3.725233      7.612791e+00 7.562879e+00 1.154609e-01 1.402548e-01
## qa.g      3.441173      2.481927e+05 2.317322e+05 4.119189e+04 5.007290e+04
## ssv.d     -2.267158      6.163372e+01 6.490783e+01 9.795975e+00 1.511743e+01
## ss.s      -2.565305      1.252674e+01 1.651111e+01 6.968606e+00 1.625876e+01
## ssv.s     -2.792663      8.992442e+00 1.235758e+01 5.524150e+00 1.261394e+01
## dqo.e     -2.890822      3.967035e+02 4.423813e+02 8.052204e+01 1.654059e+02
## nh4.s     -2.992351      9.941572e+00 1.216982e+01 6.502827e+00 7.795045e+00
## nh4.e     -3.196035      2.482396e+01 4.114419e+01 4.905343e+00 5.345419e+01
## qr.g      -3.712500      3.954973e+04 4.096489e+04 2.921590e+03 3.990294e+03
## dbo.e     -4.043637      1.764186e+02 2.136061e+02 5.374166e+01 9.627007e+01
## ssv.e     -4.196953      1.321686e+02 1.574583e+02 2.550039e+01 6.307787e+01
## fe.e      -4.241511      4.321860e+01 4.737019e+01 1.216351e+01 1.024615e+01
## dbo.d     -4.318439      1.043023e+02 1.199558e+02 2.595910e+01 3.794470e+01
## ss.e      -4.468116      1.736570e+02 2.096944e+02 3.473013e+01 8.442999e+01
## nh4.d     -4.814362      2.381952e+01 2.594192e+01 4.253499e+00 4.614821e+00
## p.e       -7.462399      5.869815e+00 9.902146e+00 1.131790e+00 5.656459e+00
## p.s       -7.636893      1.451105e+00 2.746086e+00 8.383211e-01 1.775063e+00
## p.d       -8.365265      3.190067e+00 6.071465e+00 8.400500e-01 3.605706e+00
##           p.value

```

```

## qb.b 3.035599e-14
## ph.e 2.595269e-12
## q.e 5.165683e-10
## ph.d 1.951345e-04
## qa.g 5.791976e-04
## ssv.d 2.338057e-02
## ss.s 1.030853e-02
## ssv.s 5.227609e-03
## dgo.e 3.842359e-03
## nh4.s 2.768377e-03
## nh4.e 1.393300e-03
## qr.g 2.052224e-04
## dbo.e 5.262845e-05
## ssv.e 2.705305e-05
## fe.e 2.220198e-05
## dbo.d 1.571363e-05
## ss.e 7.891150e-06
## nh4.d 1.476709e-06
## p.e 8.496134e-14
## p.s 2.225257e-14
## p.d 5.997998e-17
##
## $`4`
##      v.test Mean in category Overall mean sd in category Overall sd
## qp.g      8.369860      7.407260e+02 6.307379e+02 1.019273e+02 1.477411e+02
## ph.s      7.882365      7.670660e+00 7.533333e+00 1.511864e-01 1.958714e-01
## nh4.s      7.260424      1.720375e+01 1.216982e+01 6.991434e+00 7.795045e+00
## nkt.d      6.091541      4.072544e+01 3.641604e+01 9.888748e+00 7.953600e+00
## nkt.e      5.665565      4.661299e+01 4.195316e+01 9.802301e+00 9.246992e+00
## fe.e      5.361318      5.225625e+01 4.737019e+01 7.045681e+00 1.024615e+01
## nh4.d      3.641203      2.743652e+01 2.594192e+01 2.427227e+00 4.614821e+00
## qr.g      3.372423      4.216183e+04 4.096489e+04 1.259576e+03 3.990294e+03
## ph.d      3.226123      7.603125e+00 7.562879e+00 9.236829e-02 1.402548e-01
## ph.e      2.577513      7.651042e+00 7.618813e+00 8.798847e-02 1.405767e-01
## nh4.e     -2.093786      3.118921e+01 4.114419e+01 1.358853e+01 5.345419e+01
## dbo.e     -2.830604      1.893681e+02 2.136061e+02 5.007340e+01 9.627007e+01
## ssv.e     -2.939045      1.409688e+02 1.574583e+02 3.744897e+01 6.307787e+01
## ssv.d     -3.081832      6.076389e+01 6.490783e+01 1.041396e+01 1.511743e+01
## dgo.e     -3.511547      3.907188e+02 4.423813e+02 1.128559e+02 1.654059e+02
## dbo.d     -3.532657      1.080330e+02 1.199558e+02 3.089277e+01 3.794470e+01
## ss.e      -3.808495      1.810938e+02 2.096944e+02 5.155376e+01 8.442999e+01
## ss.d      -4.637792      8.059549e+01 8.872096e+01 1.477450e+01 1.969749e+01
## p.d       -4.783936      4.537193e+00 6.071465e+00 1.504983e+00 3.605706e+00
## p.e       -5.541238      7.114242e+00 9.902146e+00 2.566078e+00 5.656459e+00
## p.s       -5.587146      1.863961e+00 2.746086e+00 4.651973e-01 1.775063e+00
## qa.g      -6.122437      2.044642e+05 2.317322e+05 4.097892e+04 5.007290e+04
## qb.b      -6.244545      3.663704e+04 3.890254e+04 3.339490e+03 4.078833e+03
## mlvss.b   -8.274550      1.149094e+03 1.344251e+03 2.057226e+02 2.651639e+02
## mlss.b    -8.750404      1.485198e+03 1.766515e+03 2.811038e+02 3.614446e+02
## q.e       -9.262379      3.759902e+04 4.180854e+04 3.839056e+03 5.109559e+03
##
##      p.value
## qp.g      5.768686e-17
## ph.s      3.212422e-15
## nh4.s     3.858797e-13

```

```

## nkt.d 1.118292e-09
## nkt.e 1.465409e-08
## fe.e 8.261702e-08
## nh4.d 2.713665e-04
## qr.g 7.450997e-04
## ph.d 1.254794e-03
## ph.e 9.951428e-03
## nh4.e 3.627905e-02
## dbo.e 4.646018e-03
## ssv.e 3.292256e-03
## ssv.d 2.057312e-03
## dqo.e 4.455069e-04
## dbo.d 4.114051e-04
## ss.e 1.398153e-04
## ss.d 3.521510e-06
## p.d 1.718956e-06
## p.e 3.003411e-08
## p.s 2.308316e-08
## qa.g 9.215473e-10
## qb.b 4.250360e-10
## mlvss.b 1.289419e-16
## mlss.b 2.125897e-18
## q.e 1.999270e-20
##
## $`5`
##      v.test Mean in category Overall mean sd in category Overall sd
## ssv.d 9.548430      8.031481e+01 6.490783e+01 1.530824e+01 15.1174319
## ss.d 9.089570      1.078310e+02 8.872096e+01 2.239636e+01 19.6974930
## dqo.d 8.397822      3.034028e+02 2.496414e+02 4.792838e+01 59.9785437
## dbo.d 6.937578      1.480532e+02 1.199558e+02 3.612720e+01 37.9446951
## ss.e 6.023836      2.639792e+02 2.096944e+02 9.162323e+01 84.4299924
## dqo.e 5.909328      5.467083e+02 4.423813e+02 1.245167e+02 165.4059089
## nkt.d 5.803392      4.134270e+01 3.641604e+01 6.830697e+00 7.9535998
## qp.g 5.415340      7.161333e+02 6.307379e+02 1.215539e+02 147.7411209
## ssv.e 5.237758      1.927222e+02 1.574583e+02 5.045831e+01 63.0778719
## nh4.d 4.761835      2.828743e+01 2.594192e+01 4.223998e+00 4.6148210
## ph.s 4.362483      7.624537e+00 7.533333e+00 1.669205e-01 0.1958714
## nkt.e 4.315820      4.621278e+01 4.195316e+01 8.274843e+00 9.2469924
## ph.d 3.500238      7.615278e+00 7.562879e+00 9.952627e-02 0.1402548
## dbo.e 3.427859      2.488287e+02 2.136061e+02 6.646388e+01 96.2700653
## ssv.s 2.674456      1.595833e+01 1.235758e+01 1.877528e+01 12.6139385
## ss.s 2.653115      2.111528e+01 1.651111e+01 2.430286e+01 16.2587559
## dbo.s 2.330972      2.777778e-01 1.818182e-01 4.479032e-01 0.3856946
## ph.e 2.124786      7.650694e+00 7.618813e+00 9.446231e-02 0.1405767
## qb.b 2.041321      3.979124e+04 3.890254e+04 3.070174e+03 4078.8329019
## v30.b 1.975021      2.908056e+02 2.627146e+02 1.457519e+02 133.2558559
## p.d -3.956847      4.548649e+00 6.071465e+00 1.547965e+00 3.6057065
## p.e -4.380875      7.257222e+00 9.902146e+00 2.222683e+00 5.6564592
## p.s -4.653603      1.864406e+00 2.746086e+00 7.213007e-01 1.7750626
##      p.value
## ssv.d 1.316765e-21
## ss.d 9.943106e-20
## dqo.d 4.548371e-17
## dbo.d 3.988798e-12

```

```

## ss.e 1.703307e-09
## dgo.e 3.435068e-09
## nkt.d 6.498681e-09
## qp.g 6.117220e-08
## ssv.e 1.625386e-07
## nh4.d 1.918403e-06
## ph.s 1.285945e-05
## nkt.e 1.590116e-05
## ph.d 4.648427e-04
## dbo.e 6.083610e-04
## ssv.s 7.485059e-03
## ss.s 7.975278e-03
## dbo.s 1.975481e-02
## ph.e 3.360449e-02
## qb.b 4.121893e-02
## v30.b 4.826573e-02
## p.d 7.594566e-05
## p.e 1.182036e-05
## p.s 3.261844e-06
##
## $`6`
##      v.test Mean in category Overall mean sd in category Overall sd
## nh4.e 13.908195      152.740286      41.144192      109.4581539      53.4541918
## p.e 11.428445      19.605643      9.902146      2.5264603      5.6564592
## p.d 10.136979      11.557964      6.071465      0.9721012      3.6057065
## p.s 8.983253      5.139643      2.746086      0.4133082      1.7750626
## v30.b 6.640961      395.550000      262.714646      140.5862636      133.2558559
## dbo.d 5.457040      151.037500      119.955808      27.8837738      37.9446951
## nh4.d 4.438122      29.016250      25.941919      2.3294040      4.6148210
## qr.g 3.914289      43309.410000      40964.888636      1542.8587806      3990.2935927
## mlvss.b 3.412438      1480.075000      1344.251263      205.1073362      265.1639251
## ss.d 2.956518      97.462500      88.720960      13.3099528      19.6974930
## ssv.d 2.679203      70.987500      64.907828      11.0405432      15.1174319
## mlss.b 2.589814      1907.025000      1766.515152      227.4090464      361.4446128
## nkt.e 2.299765      45.145286      41.953157      8.8360203      9.2469924
## dbo.e 2.105886      244.037500      213.606061      62.5530163      96.2700653
## nkt.d -3.726221      31.967380      36.416035      3.4349947      7.9535998
## ph.e -4.564316      7.522500      7.618813      0.1254741      0.1405767
## qp.g -4.785547      524.610000      630.737857      55.0190322      147.7411209
## ph.s -5.427753      7.373750      7.533333      0.1710217      0.1958714
## ph.d -5.836639      7.440000      7.562879      0.1475635      0.1402548
##      p.value
## nh4.e 5.648705e-44
## p.e 3.014595e-30
## p.d 3.786292e-24
## p.s 2.628762e-19
## v30.b 3.116449e-11
## dbo.d 4.841376e-08
## nh4.d 9.074698e-06
## qr.g 9.067098e-05
## mlvss.b 6.438455e-04
## ss.d 3.111339e-03
## ssv.d 7.379760e-03
## mlss.b 9.602784e-03

```

```

## nkt.e 2.146153e-02
## dbo.e 3.521429e-02
## nkt.d 1.943720e-04
## ph.e 5.011250e-06
## qp.g 1.705226e-06
## ph.s 5.706794e-08
## ph.d 5.326427e-09
##
## $`7`
##          v.test Mean in category Overall mean sd in category Overall sd
## ssv.e 10.786488 2.640946e+02 1.574583e+02 7.926426e+01 6.307787e+01
## dbo.e 9.949702 3.637297e+02 2.136061e+02 1.689280e+02 9.627007e+01
## ss.e 9.812605 3.395405e+02 2.096944e+02 9.487997e+01 8.442999e+01
## dqo.e 9.501222 6.886892e+02 4.423813e+02 2.377629e+02 1.654059e+02
## p.d 7.265442 1.017729e+01 6.071465e+00 2.694517e+00 3.605706e+00
## p.e 6.862087 1.598558e+01 9.902146e+00 4.045566e+00 5.656459e+00
## p.s 6.831997 4.646766e+00 2.746086e+00 1.324410e+00 1.775063e+00
## dbo.d 3.761303 1.423243e+02 1.199558e+02 3.530164e+01 3.794470e+01
## qr.g 2.719729 4.266579e+04 4.096489e+04 2.030601e+03 3.990294e+03
## qa.g 2.553264 2.517699e+05 2.317322e+05 5.099712e+04 5.007290e+04
## ssv.d 2.331714 7.043243e+01 6.490783e+01 1.072643e+01 1.511743e+01
## mlvss.b -2.481215 1.241135e+03 1.344251e+03 2.933543e+02 2.651639e+02
## qp.g -3.020391 5.608000e+02 6.307379e+02 1.123057e+02 1.477411e+02
## ph.s -3.022703 7.440541e+00 7.533333e+00 1.240302e-01 1.958714e-01
## ph.d -4.151467 7.471622e+00 7.562879e+00 1.353377e-01 1.402548e-01
## nkt.d -4.408933 3.092006e+01 3.641604e+01 3.258221e+00 7.953600e+00
## ph.e -5.331328 7.501351e+00 7.618813e+00 1.411758e-01 1.405767e-01
## nkt.e -5.890872 3.341571e+01 4.195316e+01 3.093561e+00 9.246992e+00
##          p.value
## ssv.e 3.987375e-27
## dbo.e 2.529386e-23
## ss.e 9.936922e-23
## dqo.e 2.074415e-21
## p.d 3.718204e-13
## p.e 6.786160e-12
## p.s 8.374078e-12
## dbo.d 1.690307e-04
## qr.g 6.533552e-03
## qa.g 1.067184e-02
## ssv.d 1.971573e-02
## mlvss.b 1.309355e-02
## qp.g 2.524488e-03
## ph.s 2.505282e-03
## ph.d 3.303515e-05
## nkt.d 1.038810e-05
## ph.e 9.749690e-08
## nkt.e 3.841633e-09

```

```

(res.hcpc$call$t$within[1]-res.hcpc$call$t$within[7])/res.hcpc$call$t$within[1] # representació de la i

```

```

## [1] 0.6286321

```

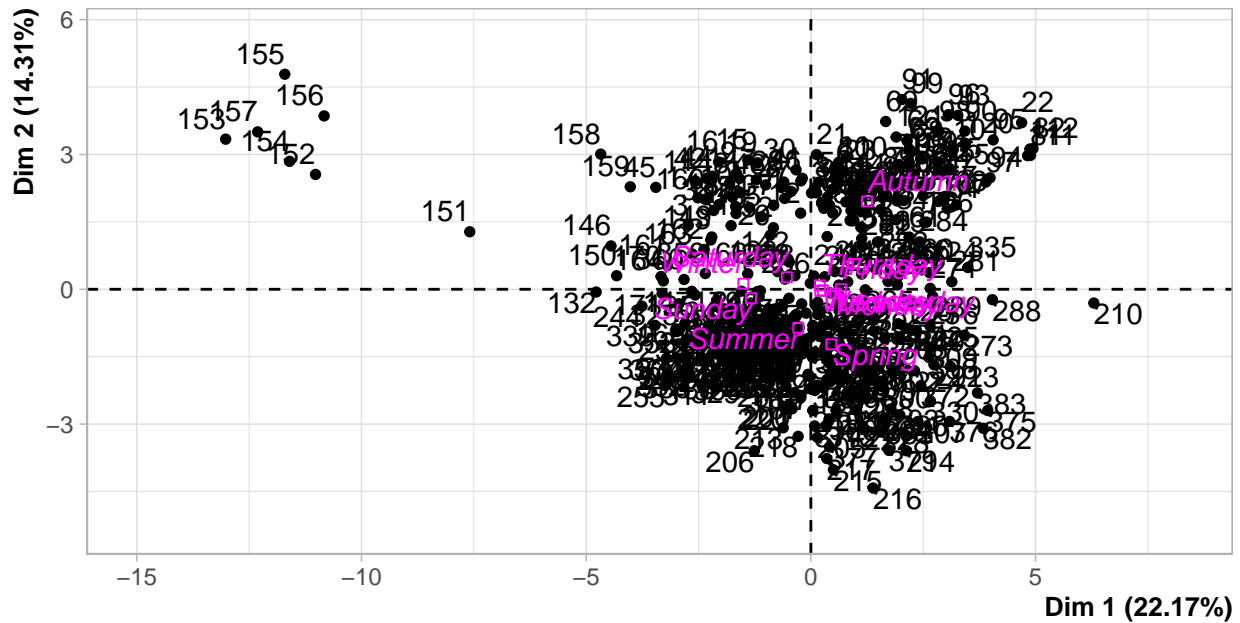
Clusters: 7. Inèrcia: 62.86%

9. A nondefault criteria for selecting the number of clusters to 3 has to be set. Explain the characteristics of cluster number 3.

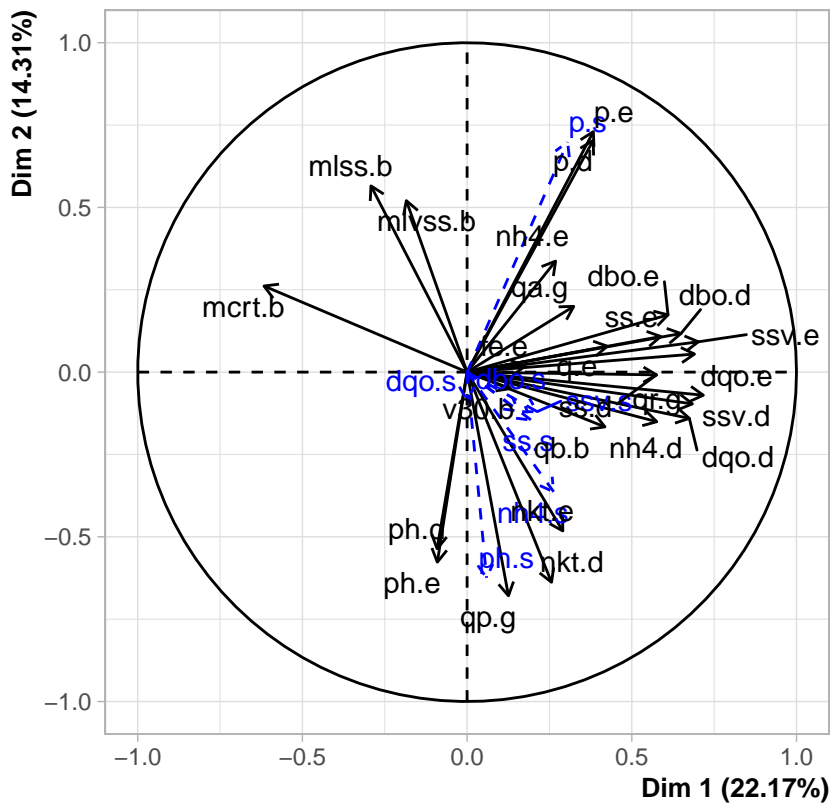
6 dimensions have to be selected according to Kaiser's criteria

```
res.pca <- PCA(df[, c(4:30,32:36,40,45:46)], quanti.sup=c(23:29), quali.sup=c(34:35))
```

PCA graph of individuals

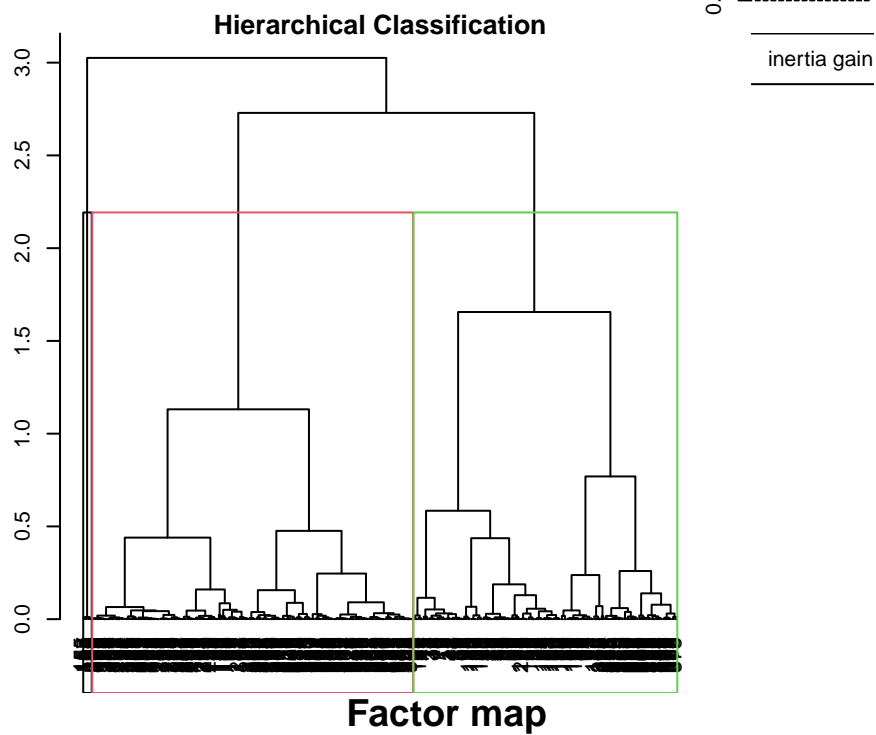


PCA graph of variables

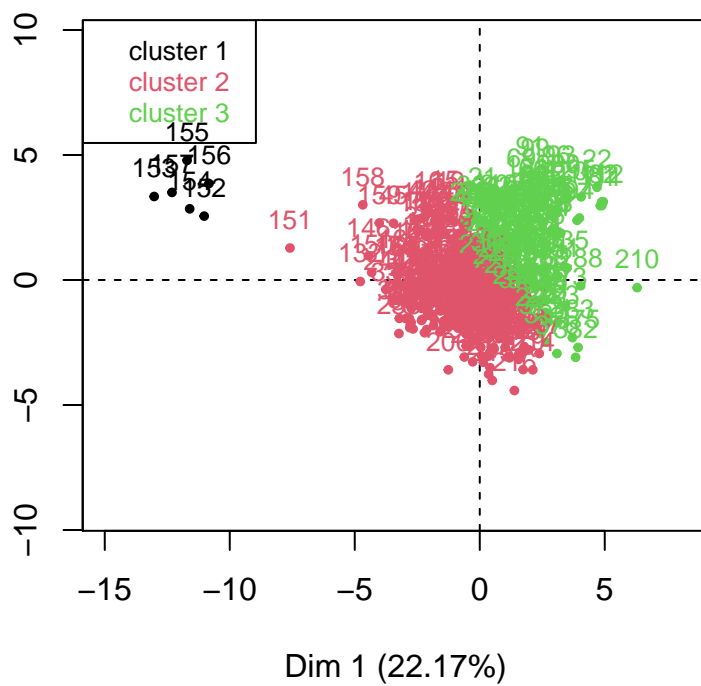
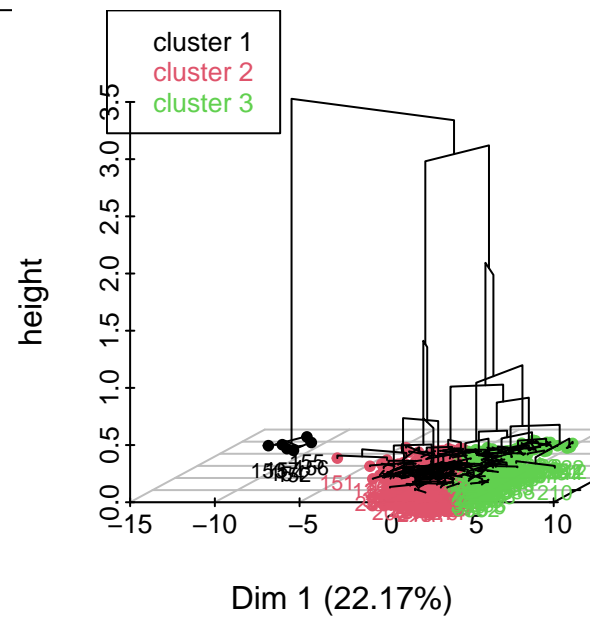


```
res.hcpc <- HCPC(res.pca,nb.clust=3, graph=T)
```

Hierarchical Clustering



Hierarchical clustering on the



```
(res.hcpc$call$t$within[1]-res.hcpc$call$t$within[3])/res.hcpc$call$t$within[1] # representació de la i
```

```
## [1] 0.3655823
```

10. Use a partition method to group available data into the selected number of clusters found in Question 7. Determine the quality of the partition and plot the resulting partition in the first factorial plane.

```
# Ja no m'ha donat temps.
```

Do not forget to Knit to .pdf before posting your answers in ATENEA.