# Course Practical Assignment - 1st Delivery (17 de març del 2019)

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## Bank client data

# Description of input variables:

- 1. age (numeric)
- 2. job: type of job (categorical: 'admin', 'blue-collar', 'entrepreneur', 'housemaid', 'management', 'retired', 'self-employed', 'services', 'student', 'technician', 'unemployed', 'unknown')
- 3. marital : marital status (categorical: 'divorced', 'married', 'single', 'unknown'; note: 'divorced' means divorced or widowed)
- 4. education (categorical: 'basic.4y', 'basic.6y', 'basic.9y', 'high.school', 'illiterate', 'professional.course', 'university.degree', 'unknown')
- 5. default: has credit in default? (categorical: 'no', 'yes', 'unknown')
- 6. housing: has housing loan? (categorical: 'no', 'yes', 'unknown')
- 7. loan: has personal loan? (categorical: 'no', 'yes', 'unknown')# related with the last contact of the current campaign:
- 8. contact: contact communication type (categorical:'cellular', 'telephone')
- 9. month: last contact month of year (categorical: 'jan', 'feb', 'mar',..., 'nov', 'dec')
- 10. day of week: last contact day of the week (categorical:'mon','tue','wed','thu','fri')
- 11. duration: last contact duration, in seconds (numeric). Important note: this attribute highly affects the output target (e.g., if duration=0 then y='no'). Yet, the duration is not known before a call is performed. Also, after the end of the call y is obviously known. Thus, this input should only be included for benchmark purposes and should be discarded if the intention is to have a realistic predictive model.
- 12. campaign: number of contacts performed during this campaign and for this client (numeric, includes last contact)
- 13. pdays: number of days that passed by after the client was last contacted from a previous campaign (numeric; 999 means client was not previously contacted)
- 14. previous: number of contacts performed before this campaign and for this client (numeric)
- 15. poutcome: outcome of the previous marketing campaign (categorical: 'failure', 'nonexistent', 'success') # social and economic context attributes
- 16. emp.var.rate: employment variation rate quarterly indicator (numeric)
- 17. cons.price.idx: consumer price index monthly indicator (numeric)
- 18. cons.conf.idx: consumer confidence index monthly indicator (numeric)
- 19. euribor3m: euribor 3 month rate daily indicator (numeric)
- 20. nr.employed: number of employees quarterly indicator (numeric)
- 21. y has the client subscribed a term deposit? (binary: 'yes', 'no')

# Loading packages:

# Loading data:

```
dirwd<-"D:/Users/Usuari/Documents/ADEIpractica"
#dirwd<-"D:/Documents/GitHub/ADEI"
setwd(dirwd)

df<-read.table( paste0(dirwd, "/bank-additional/bank-additional-full.csv"), header=TRUE, sep=";")
# General description of the bank data</pre>
```

# Our chosen sample:

```
#load( pasteO(dirwd, "/bank-additional/Bank5000_raw.RData") )
summary(df)
```

```
##
         age
                             job
                                            marital
   Min.
          :18.00
                    admin.
                               :1234
                                        divorced: 556
##
   1st Qu.:32.00
                    blue-collar:1154
                                       married:3053
  Median :38.00
                    technician: 794
                                       single :1381
          :40.07
                    services : 500
                                       unknown: 10
##
  Mean
##
   3rd Qu.:47.00
                    management: 413
##
   Max. :87.00
                    retired
                              : 205
##
                    (Other)
                               : 700
##
                  education
                                  default
                                                                   loan
                                                  housing
                                                                      :4091
##
  university.degree :1472
                                       :3966
                                                      :2219
                               no
                                               no
                                                              no
## high.school
                       :1171
                               unknown:1034
                                               unknown: 137
                                                              unknown: 137
## basic.9y
                       : 716
                               yes
                                               yes
                                                      :2644
                                                                     : 772
                                                              yes
##
   professional.course: 602
##
   basic.4y
                       : 513
##
   basic.6y
                       : 291
                       : 235
##
   (Other)
##
         contact
                         month
                                    day of week
                                                    duration
##
   cellular :3130
                            :1743
                                    fri: 924
                                                 Min.
                                                            1.0
                     may
   telephone:1870
                            : 831
                                    mon:1018
                                                 1st Qu.: 101.0
                     jul
##
                            : 699
                                    thu:1039
                                                 Median : 178.0
                     aug
##
                            : 653
                                    tue:1045
                                                 Mean
                                                        : 254.8
                     jun
##
                     nov
                            : 509
                                    wed: 974
                                                 3rd Qu.: 317.0
##
                                                        :3785.0
                     apr
                            : 310
                                                 Max.
##
                     (Other): 255
                         pdays
                                        previous
##
       campaign
                                                              poutcome
##
                           : 0.0
                                             :0.0000
          : 1.000
                     Min.
                                      Min.
                                                       failure
                                                                  : 478
   1st Qu.: 1.000
                     1st Qu.:999.0
                                      1st Qu.:0.0000
                                                       nonexistent:4363
  Median : 2.000
                     Median :999.0
                                      Median :0.0000
                                                       success
                                                                  : 159
```

```
: 2.583
                            :963.2
                                            :0.1606
##
   Mean
                     Mean
                                     Mean
                    3rd Qu.:999.0
##
   3rd Qu.: 3.000
                                     3rd Qu.:0.0000
##
   Max.
          :33.000
                    Max.
                           :999.0
                                    Max.
                                            :4.0000
##
##
    emp.var.rate
                       cons.price.idx cons.conf.idx
                                                          euribor3m
##
           :-3.40000
                     Min.
                              :92.20
                                       Min.
                                              :-50.80
                                                               :0.635
  Min.
                                                       \mathtt{Min}.
   1st Qu.:-1.80000
                      1st Qu.:93.08
                                       1st Qu.:-42.70
                                                        1st Qu.:1.334
  Median : 1.10000
##
                      Median :93.77
                                       Median :-41.80
                                                        Median :4.857
##
   Mean : 0.06326
                      Mean
                              :93.57
                                      Mean
                                              :-40.43
                                                        Mean
                                                               :3.613
##
   3rd Qu.: 1.40000
                       3rd Qu.:93.99
                                       3rd Qu.:-36.40
                                                        3rd Qu.:4.961
##
   Max.
          : 1.40000
                      Max.
                              :94.77
                                       Max.
                                              :-26.90
                                                        Max.
                                                               :5.000
##
    nr.employed
##
                    У
##
  Min.
           :4964
                  no:4435
##
   1st Qu.:5099
                  yes: 565
##
   Median:5191
##
  Mean
           :5166
##
   3rd Qu.:5228
   Max.
           :5228
##
##
```

# Inicialitzacio del control d'errors, missings i outliers:

```
columnes <- names(df) #list of column names

# creem 3 dataframes inicialitzats a 0 d'una fila amb les columnes de la nostra mostra;
# en ells hi posarem el nombre d'errors, missings i outliers per a cada variable
errors <- data.frame(matrix(0, ncol = length(columnes), nrow = 1))
colnames(errors)<-columnes

missings <- data.frame(matrix(0, ncol = length(columnes), nrow = 1))
colnames(missings)<-columnes

outliers <- data.frame(matrix(0, ncol = length(columnes), nrow = 1))
colnames(outliers)<-columnes

# columnes que portaran el control per individu:
df$num_missings <- 0
df$num_outliers <- 0
df$num_outliers <- 0
df$num_errors <- 0</pre>
```

# UNIVARIATE DESCRIPTIVE ANALYSIS (to be included for each variable):

Aqui estudiem cada variable buscant missing values, outliers i possibles errors. En el cas que en trobem, els transformem en NAs i procedim a una imputacio manual o els eliminem, o una imputacio automatica (en un chunck posterior d'Imputation).

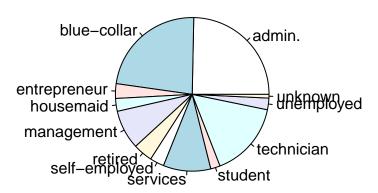
### **QUALITATIVE VARIABLES:**

També factoritzem aqui les categories (levels) de les variables qualitatives (discretes).

### Job

Jobs "unknown" són considerats com a categoria.

```
# Jobs "unknown" will be considered a category, not a missing value.
table(df$job, useNA="always")
##
##
                    blue-collar entrepreneur
          admin.
                                                     housemaid
                                                                   management
             1234
                            1154
##
                                                           135
                                                                          413
##
         retired self-employed
                                       services
                                                       student
                                                                   technician
                                                                          794
##
              205
                             149
                                            500
                                                           100
                                           <NA>
##
      unemployed
                         unknown
##
              122
                              39
                                              0
# Missings:
miss<-which(is.na(df$job));</pre>
missings$job<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "job-":
df$job<-factor(df$job)</pre>
pie(summary(df$job))
```



```
levels(df$job)<-paste0("job-",levels(df$job))</pre>
```

### Marital

Els "unknowns" seran imputats mes endavant automaticament.

```
# Marital "unknown" will be a missing value (set to NA):
sel<-which(df$marital=="unknown");length(sel)

## [1] 10

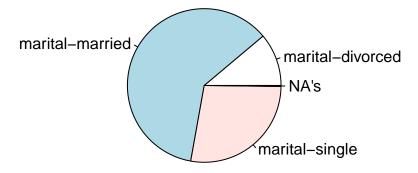
df$marital[sel]<-NA

# Missings:
miss<-which(is.na(df$marital));
missings$marital<-length(miss); length(miss)

## [1] 10

df[miss, "num_missings"]<- df[miss, "num_missings"]+1

# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "marital-":
df$marital<-factor(df$marital)
levels(df$marital)<-paste0("marital-",levels(df$marital))
pie(summary(df$marital))</pre>
```

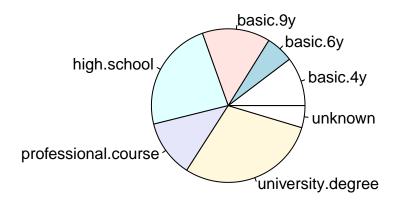


```
## marital-divorced marital-married marital-single NA's
## 556 3053 1381 10
```

### Education

Education "unknown" es considerada com a categoria. La categoria "illiterate" és inclosa manualment a "basic.4y".

```
# Education "unknown" will be considered a category, not a missing value.
table(df$education, useNA="always")
##
##
               basic.4y
                                    basic.6y
                                                          basic.9y
##
                    513
                                          291
                                                                716
##
           high.school
                                  illiterate professional.course
##
                   1171
                                                                602
                                                               <NA>
##
     university.degree
                                      unknown
##
                   1472
                                          232
                                                                  0
# Illiterates are consired as basic.4y.educated:
sel<-which(df$education=="illiterate");length(sel)</pre>
## [1] 3
df[sel, "education"] <- "basic.4y"</pre>
# Missings:
miss<-which(is.na(df$education));</pre>
missings$education<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "education-":
df$education<-factor(df$education)</pre>
pie(summary(df$education))
```



```
levels(df$education)<-paste0("education-",levels(df$education))</pre>
```

### Default (has credit in default?)

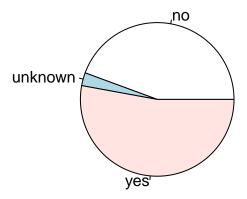
Default "unknown" sera considerada com a una categoria, no com a missing value.

```
table(df$default, useNA="always")
##
##
        no unknown
                                 <NA>
                         yes
##
      3966
               1034
                           0
                                    0
# Missings:
miss<-which(is.na(df$default));</pre>
missings$default<-length(miss); length(miss)</pre>
## [1] O
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "default-":
df$default<-factor(df$default)</pre>
summary(df$default)
##
        no unknown
##
      3966
               1034
levels(df$default)<-paste0("default-",levels(df$default))</pre>
```

### Housing

Housing "unknown" sera considerada com a una categoria, no com a missing value.

```
table(df$housing, useNA="always")
##
                                <NA>
##
        no unknown
                        yes
##
      2219
                137
                        2644
                                    0
# Missings:
miss<-which(is.na(df$housing));</pre>
missings$housing<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "housing-":
df$housing<-factor(df$housing)</pre>
pie(summary(df$housing))
```



levels(df\$housing)<-paste0("housing-",levels(df\$housing))</pre>

```
Loan (has personal loan?)

Loan "unknown" sera considerat com a missing value (NA) sera imputat mes endavant automaticament.

sel<-which(df$loan=="unknown");length(sel)

## [1] 137

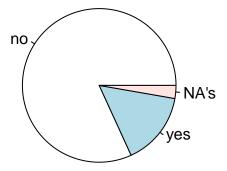
df$loan[sel]<-NA

# Missings:
miss<-which(is.na(df$loan));
missings$loan<-length(miss); length(miss)

## [1] 137

df[miss, "num_missings"]<- df[miss, "num_missings"]+1

# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "loan-":
df$loan<-factor(df$loan)
pie(summary(df$loan))
```



```
levels(df$loan)<-paste0("loan-",levels(df$loan))</pre>
```

### Contact

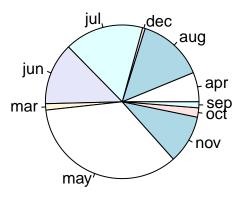
```
summary(df$contact)
##
    cellular telephone
##
        3130
                   1870
# Missings:
miss<-which(is.na(df$contact));</pre>
missings$contact<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"]<- df[miss, "num_missings"]+1</pre>
# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "contact-":
df$contact<-factor(df$contact)</pre>
summary(df$contact)
    cellular telephone
##
                   1870
        3130
levels(df$contact) <-paste0("contact-",levels(df$contact))</pre>
```

### Month

```
miss<-which(is.na(df$month));
missings$month<-length(miss); length(miss)</pre>
```

```
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1

# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "month-":
df$month <- factor (df$month)
pie (summary (df$month))</pre>
```



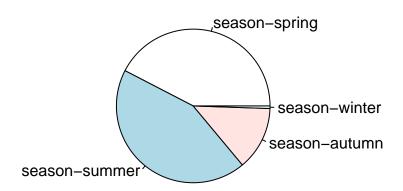
```
levels(df$month) <-paste0("month-",levels(df$month))</pre>
```

### Month -> definim noves factor categories per Season.

(New factors grouping original levels will be considered very positively.)

```
summary(df$f.season);pie(summary(df$f.season))

## season-spring season-summer season-autumn season-winter
## 2120 2183 670 27
```



### Day\_of\_week

```
miss<-which(is.na(df$day_of_week));
missings$day_of_week<-length(miss); length(miss)

## [1] 0

df[miss, "num_missings"]<- df[miss, "num_missings"]+1

# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "day_of_week-":
df$day_of_week<-factor(df$day_of_week)

summary(df$day_of_week)

## fri mon thu tue wed

## 924 1018 1039 1045 974

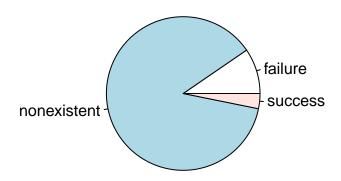
levels(df$day_of_week)<-paste0("day_of_week-",levels(df$day_of_week))</pre>
```

# Poutcome (outcome of previous marketing campaign)

```
# Poutcome "nonexistent" will be considered a category, not a missing value.
table(df$poutcome, useNA="always")

##
## failure nonexistent success <NA>
```

```
## 478 4363 159 0
# All missing data indicated as NA:
miss<-which(is.na(df$poutcome));
missings$poutcome<-length(miss); length(miss)
## [1] 0
df[miss, "num_missings"]<- df[miss, "num_missings"]+1
# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "poutcome-":
df$poutcome<-factor(df$poutcome)
pie(summary(df$poutcome))</pre>
```



```
levels(df$poutcome) <-paste0("poutcome-",levels(df$poutcome))

y (has the client subscribed a term deposit?)

miss<-which(is.na(df$y));
missings$y<-length(miss); length(miss)

## [1] 0

df[miss, "num_missings"] <- df[miss, "num_missings"] +1

# Factoritzem les categories (levels) de la columna i afegim l'etiqueta "y-":
df$y<-factor(df$y)
summary(df$y)

## no yes
## 4435 565</pre>
```

```
levels(df$y)<-paste0("y-",levels(df$y))</pre>
```

# **QUANTITATIVES VARIABLES:**

Funcio de gran utilitat per a la deteccio d'outliers:

```
calcQ <- function(x){
   s.x <- summary(x)

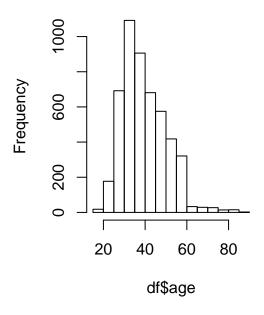
iqr <- s.x[5]-s.x[2] # IQR = Q3([5]) - Q1([2])

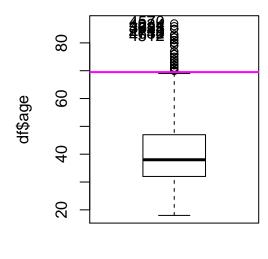
list(souti=s.x[2]-3*iqr, mouti=s.x[2]-1.5*iqr, min=s.x[1], q1=s.x[2],
   q2=s.x[3], q3=s.x[5], max=s.x[6], mouts=s.x[5]+1.5*iqr, souts=s.x[5]+3*iqr)
}</pre>
```

### Age

```
summary(df$age)
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                 Max.
     18.00
            32.00
                      38.00
                               40.07
                                       47.00
                                                87.00
# No tenim cap missing NA!
miss<-which(is.na(df$age))
missings$age<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
par(mfrow=c(1,2))
hist(df$age, breaks=10, main="age - histogram")
Boxplot(df$age)
## [1] 4570 4634 3623 3628 3631 4755 4612 4734 4740 4512
# Errors are under aged people:
err<-which(df$age < 18)</pre>
errors$age<-length(err); length(err)</pre>
## [1] 0
if(length(err)>0) df<-df[-err,]</pre>
# Outliers:
out.var <- calcQ(df$age)</pre>
abline(h=out.var[["mouts"]], col="magenta", lwd=2); out.var[["mouts"]]
## 3rd Qu.
##
      69.5
# But our outliers will be the ones above 100 years (there is none):
abline(h=100, col="red", lwd=2)
```

# age - histogram





```
out<-which(df$age > 100)
outliers$age<-length(out); length(out)
## [1] 0
if(length(out)>0) df<-df[-out,]</pre>
```

### Duration

Els outliers en la variable duracio han estat eliminats. Corresponen a duracions per sota els 5 segons (trucada massa curta a un client que potser no podia parlar en aquell moment o penja per error) i per sobre dels 1600 segons (26 minuts).

```
summary(df$duration)
##
      Min. 1st Qu.
                     Median
                                Mean 3rd Qu.
                                                 Max.
##
              101.0
                      178.0
                               254.8
                                        317.0
                                               3785.0
# No tenim cap missing NA!
miss<-which(is.na(df$duration));</pre>
missings$duration<-length(miss); length(miss)</pre>
df[miss, "num_missings"]<- df[miss, "num_missings"]+1</pre>
par(mfrow=c(1,2))
hist(df$duration, breaks=20, main="duration - histogram")
Boxplot(df$duration)
```

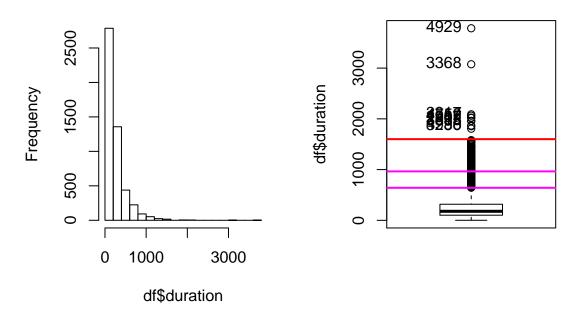
[1] 4929 3368 2817 4759 1285 2907 2033 3815 4998 3280

```
# Outliers:
out.var <- calcQ(df$duration)
abline(h=out.var[["mouts"]], col="magenta", lwd=2); out.var[["mouts"]]

## 3rd Qu.
## 641
abline(h=out.var[["souts"]], col="magenta", lwd=2); out.var[["souts"]]

## 3rd Qu.
## 965
# But our outliers will be the ones above 1600 and below 5 seconds:
abline(h=1600, col="red", lwd=2)</pre>
```

# duration - histogram



```
out<-which( (df$duration < 5) | (df$duration > 1600) )
outliers$duration=length(out); length(out)

## [1] 14

df[out, "num_outliers"]<- df[out, "num_outliers"]+1

df[out, "duration"]<-NA

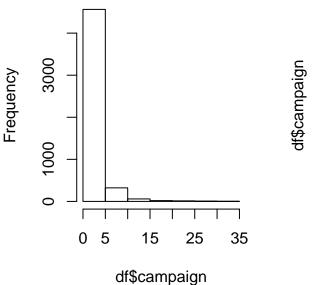
# Eliminem els outliers:
if(length(out)>0) df<-df[-out,]

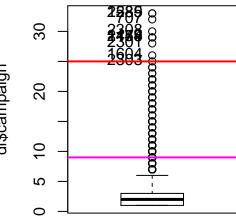
# Final summary of duration variable:
# par(mfrow=c(1,1))
# summary(df$duration)
# Boxplot(df$duration)</pre>
```

### Duration -> creem una columna de duracio en minuts:

```
df$minutes<-df$duration/60
summary(df$minutes)
       Min. 1st Qu. Median
                                  Mean 3rd Qu.
                                                     Max.
## 0.08333 1.68333 2.95000 4.17703 5.26667 26.33333
Campaign
# summary(df$campaiqn)
# No tenim cap missing NA!
miss<-which(is.na(df$campaign));</pre>
missings$campaign<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
par(mfrow=c(1,2))
hist(df$campaign, breaks=10, main="campaign - histogram")
Boxplot(df$campaign)
## [1] 1589 2285 707 2308 1158 1474 2149 2301 1604 2303
# Outliers:
out.var <- calcQ(df$campaign)</pre>
abline(h=out.var[["souts"]], col="magenta", lwd=2); out.var[["souts"]]
## 3rd Qu.
##
# But our outliers will be the ones contacted more than 25 times:
abline(h=25, col="red", lwd=2)
```

# campaign - histogram





```
out<-which(df$campaign > 25)
df[out, "num_outliers"] <- df[out, "num_outliers"] +1</pre>
outliers$campaign=length(out); length(out)
## [1] 9
df[out, "campaign"] <-NA</pre>
# Final summary of campaign variable:
# par(mfrow=c(1,1))
# summary(df$campaign)
# Boxplot(df$campaign)
```

### **Pdays**

```
# No tenim cap missing NA!
miss<-which(is.na(df$pdays));</pre>
missings$pdays<-length(miss); length(miss)</pre>
## [1] 0
df[miss, "num_missings"] <- df[miss, "num_missings"] +1</pre>
# Values that are 999 mean never contacted before:
never<-which(df$pdays==999)</pre>
# They correspond to this percentage of rows:
length(never)/5000*100
```

```
## [1] 96.18
```

```
# No outliers!
# Final summary of pdays variable:
summary(df$pdays)
      Min. 1st Qu.
                    Median
                               Mean 3rd Qu.
                                               Max.
##
       0.0
             999.0
                     999.0
                              963.7
                                      999.0
                                              999.0
par(mfrow=c(1,2))
hist(df$pdays, breaks=10, main="pdays - histogram")
Boxplot(df$pdays)
```

# pdays – histogram O 1000 3000 2000 1000 0 1

1000

## [1] 3148 4902 3576 4135 4366 3627 3642 3644 3646 4352

600

df\$pdays

### Previous

##

0 200

Min. 1st Qu. Median

```
# No tenim cap missing NA!
miss<-which(is.na(df$previous));
missings$previous<-length(miss); length(miss)

## [1] 0

df[miss, "num_missings"]<- df[miss, "num_missings"]+1

par(mfrow=c(1,2))
hist(df$previous, main="previous - histogram")

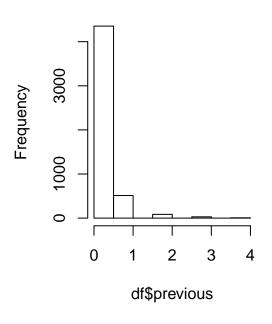
# Final summary of previous variable:
summary(df$previous)</pre>
```

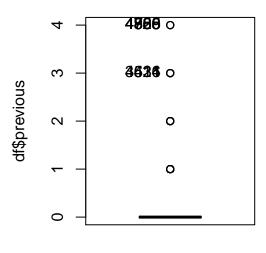
Max.

Mean 3rd Qu.

Boxplot(df\$previous)

# previous - histogram





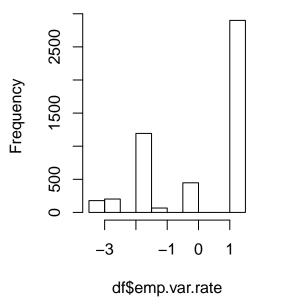
## [1] 4769 4786 4805 4826 4850 4888 4925 3431 4516 4624

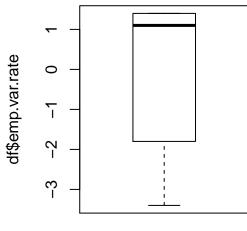
# emp.var.rate

```
# Neither missing, outliers nor error values.
par(mfrow=c(1,2))
hist(df$emp.var.rate, main="emp.var.rate - histogram")
summary(df$emp.var.rate)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -3.40000 -1.80000 1.10000 0.06446 1.40000 1.40000
Boxplot(df$emp.var.rate)
```

# emp.var.rate - histogram

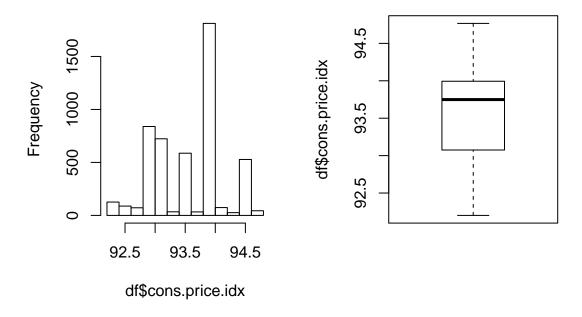




# cons.price.idx

```
# Neither missing, outliers nor error values.
par(mfrow=c(1,2))
hist(df$cons.price.idx, main="cons.price.idx - histogram")
summary(df$cons.price.idx)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
##
     92.20
             93.08
                     93.75
                             93.57
                                     93.99
                                             94.77
Boxplot(df$cons.price.idx)
```

# cons.price.idx - histogram



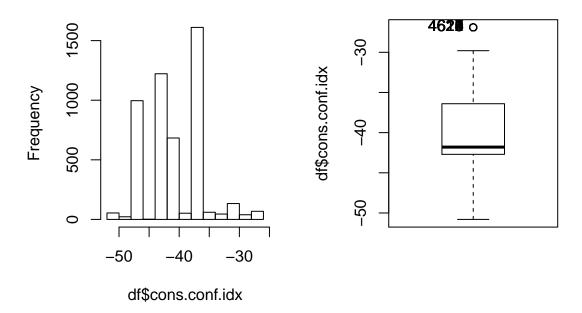
### cons.conf.idx

```
# Neither missing, outliers nor error values.
par(mfrow=c(1,2))
hist(df$cons.conf.idx, main="cons.conf.idx - histogram")
summary(df$cons.conf.idx)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## -50.80 -42.70 -41.80 -40.43 -36.40 -26.90

Boxplot(df$cons.conf.idx)
```

# cons.conf.idx - histogram



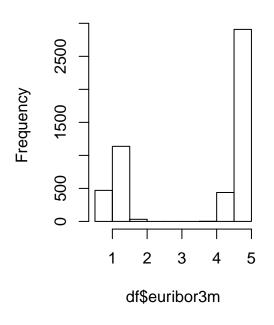
## [1] 4617 4618 4619 4620 4621 4622 4623 4624 4625 4626

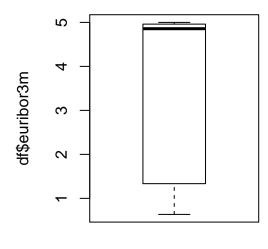
### euribor3m

```
# Neither missing, outliers nor error values.
par(mfrow=c(1,2))
hist(df$euribor3m, main="euribor3m - histogram")
summary(df$euribor3m)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.635 1.334 4.857 3.614 4.961 5.000
Boxplot(df$euribor3m)
```

# euribor3m - histogram

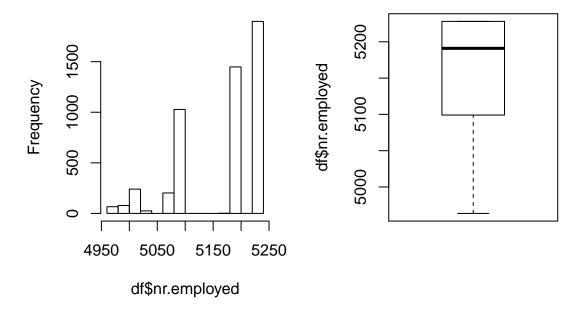




# nr.employed

```
# Neither missing, outliers nor error values.
par(mfrow=c(1,2))
hist(df$nr.employed, main="nr.employed - histogram")
summary(df$nr.employed)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
##
      4964
              5099
                      5191
                              5166
                                      5228
                                              5228
Boxplot(df$nr.employed)
```

# nr.employed - histogram



# DISCRETITZACIO DE VARIABLES NUMERIQUES:

Les variables numeriques originals que corresponen a conceptes quantitatius reals es mantenen com a numeriques, pero tambe s'han de crear factors addicionals com a discretitzacio de cada variable numerica. Les etiquetes adicionals als factors s'afegeixen posterior als grafics per una questio estetica, es redueix la mida de les etiquetes i es poden veure amb m?s claredad cada una de les variables.

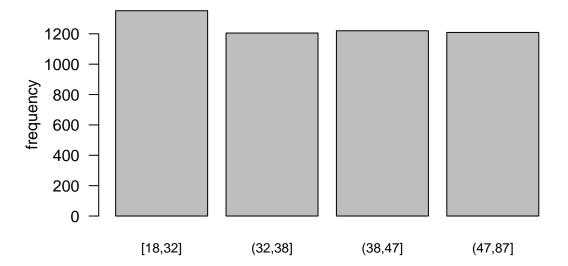
```
par(mfrow=c(1,1))

# AGE
qulist<-quantile(df$age, seq(0,1,0.25), na.rm=TRUE)

df$f.age<-factor( cut(df$age, breaks=qulist, include.lowest=T) )

# Es mostra una distribucio d'edats equitativa amb aquesta factoritzacio:
barplot(table(df$f.age), main="f.age - additional factors", ylab="frequency", las=1, cex.names=0.8)</pre>
```

# f.age - additional factors



```
summary(df$f.age)

## [18,32] (32,38] (38,47] (47,87]

## 1352 1205 1220 1209

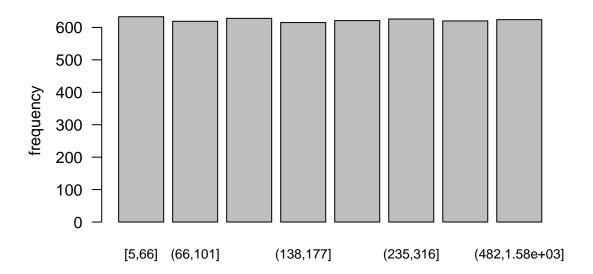
levels(df$f.age)<-paste0("f.age-", levels(df$f.age))

# DURATION
qulist<-quantile(df$duration, seq(0,1,0.125), na.rm=TRUE)

df$f.duration<-factor( cut(df$duration, breaks=qulist, include.lowest=T) )

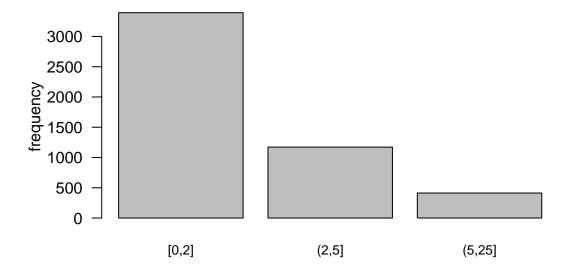
# Es mostra una distribucio de duracions de la trucada equitativa amb aquesta factoritzacio:
barplot(table(df$f.duration), main="f.duration - additional factors", ylab="frequency", las=1, cex.name")</pre>
```

# f.duration - additional factors

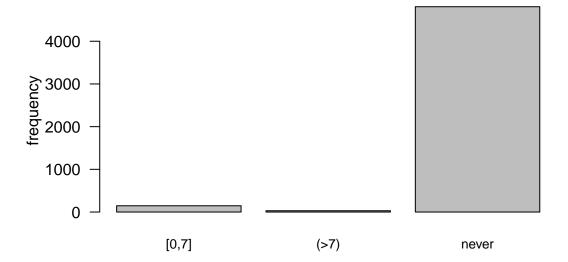


```
levels(df$f.duration)<-paste0("f.duration-", levels(df$f.duration) )</pre>
summary(df$f.duration)
##
           f.duration-[5,66]
                                    f.duration-(66,101]
##
##
        f.duration-(101,138]
                                   f.duration-(138,177]
##
##
        f.duration-(177,235]
                                   f.duration-(235,316]
##
        f.duration-(316,482] f.duration-(482,1.58e+03]
##
##
                                                     624
# CAMPAIGN
qulist<-quantile(df$campaign, seq(0,1,0.5), na.rm=TRUE)
df$f.campaign<-factor( cut(df$campaign, breaks=c(0,2,5,25), include.lowest=T) )</pre>
# Resultat de la factoritzacio de cops que s'ha contactat al client en la campanya actual:
barplot(table(df$f.campaign), main="f.campaign - additional factors", ylab="frequency", las=1, cex.name
```

# f.campaign - additional factors



# f.pdays - additional factors



```
levels(df$f.pdays)<-paste0("f.pdays-", levels(df$f.pdays) )
summary(df$f.pdays)

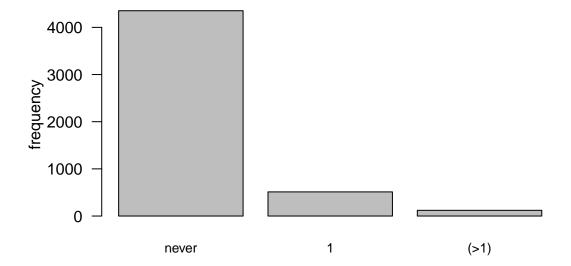
## f.pdays-[0,7] f.pdays-(>7) f.pdays-never
## 147 30 4809

# PREVIOUS
df$f.previous<-factor( cut(df$previous, breaks=c(-Inf, 0, 1, +Inf), include.lowest=T) )

levels(df$f.previous)<-c("never", "1", "(>1)")

# Resultat de la factoritzacio de number of contacts performed
# before this campaign and for this client:
barplot(table(df$f.previous), main="f.previous - additional factors", ylab="frequency", las=1, cex.name
```

# f.previous - additional factors



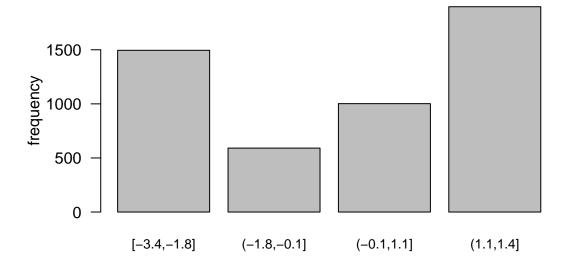
```
levels(df$f.previous)<-paste0("f.previous-", levels(df$f.previous))
summary(df$f.previous)

## f.previous-never f.previous-1 f.previous-(>1)
## 4353 512 121

# EMP.VAR.RATE
qulist<-quantile(df$emp.var.rate, seq(0,1,0.125), na.rm=TRUE)

df$f.emp.var.rate <-factor( cut(df$emp.var.rate , breaks=unique(qulist), include.lowest=T) )
barplot(table(df$f.emp.var.rate), main="f.emp.var.rate - additional factors", ylab="frequency", las=1,</pre>
```

# f.emp.var.rate - additional factors



```
levels(df$f.emp.var.rate) <-paste0("f.emp.var.rate-", levels(df$f.emp.var.rate))
summary(df$f.emp.var.rate)

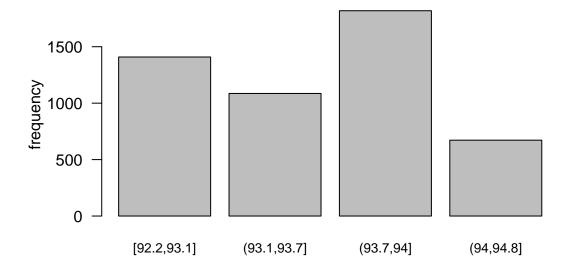
## f.emp.var.rate-[-3.4,-1.8] f.emp.var.rate-(-1.8,-0.1]
## 1495 591

## f.emp.var.rate-(-0.1,1.1] f.emp.var.rate-(1.1,1.4]
## 1002 1898

# CONS.PRICE.IDX
qulist<-quantile(df$cons.price.idx, seq(0,1,0.25), na.rm=TRUE)

df$f.cons.price.idx <-factor( cut(df$cons.price.idx , breaks=unique(qulist), include.lowest=T) )
barplot(table(df$f.cons.price.idx), main="f.cons.price.idx - additional factors", ylab="frequency", las")</pre>
```

# f.cons.price.idx - additional factors



```
levels(df$f.cons.price.idx) <-pasteO("f.cons.price.idx-", levels(df$f.cons.price.idx))
summary(df$f.cons.price.idx)

## f.cons.price.idx-[92.2,93.1] f.cons.price.idx-(93.1,93.7]
## 1409 1086

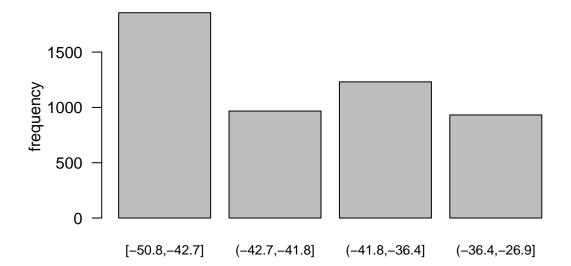
## f.cons.price.idx-(93.7,94] f.cons.price.idx-(94,94.8]
## 672

## CONS.CONF.IDX
qulist<-quantile(df$cons.conf.idx, seq(0,1,0.25), na.rm=TRUE)

df$f.cons.conf.idx <-factor( cut(df$cons.conf.idx , breaks=unique(qulist), include.lowest=T) )

barplot(table(df$f.cons.conf.idx), main="f.cons.conf.idx - additional factors", ylab="frequency", las=1")</pre>
```

# f.cons.conf.idx - additional factors



```
levels(df$f.cons.conf.idx) <-paste0("f.cons.conf.idx-", levels(df$f.cons.conf.idx) )
summary(df$f.cons.conf.idx)

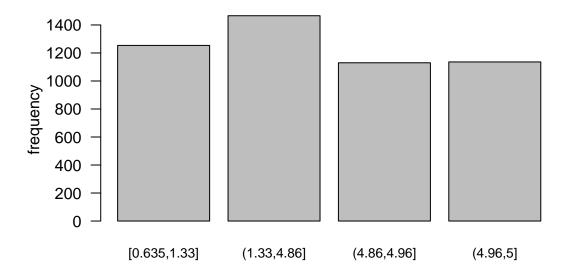
## f.cons.conf.idx-[-50.8,-42.7] f.cons.conf.idx-(-42.7,-41.8]
## 1856 967

## f.cons.conf.idx-(-41.8,-36.4] f.cons.conf.idx-(-36.4,-26.9]
## 1231 932

# EURIBOR3M
qulist<-quantile(df$euribor3m, seq(0,1,0.25), na.rm=TRUE)

df$f.euribor3m <-factor( cut(df$euribor3m , breaks=unique(qulist), include.lowest=T) )
barplot(table(df$f.euribor3m), main="f.euribor3m - additional factors", ylab="frequency", las=1, cex.na</pre>
```

# f.euribor3m - additional factors



```
levels(df$f.euribor3m) <-paste0("f.euribor3m-", levels(df$f.euribor3m))
summary(df$f.euribor3m)

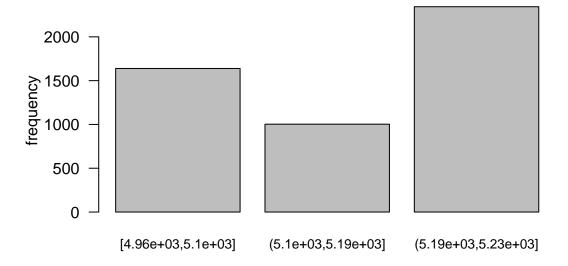
## f.euribor3m-[0.635,1.33] f.euribor3m-(1.33,4.86] f.euribor3m-(4.86,4.96]
## 1254 1466 1130

## f.euribor3m-(4.96,5]
## 1136

# NR.EMPLOYED
qulist<-quantile(df$nr.employed, seq(0,1,0.25), na.rm=TRUE)

df$f.nr.employed <-factor( cut(df$nr.employed , breaks=unique(qulist), include.lowest=T) )
barplot(table(df$f.nr.employed), main="f.nr.employed - additional factors", ylab="frequency", las=1, cettings.")</pre>
```

# f.nr.employed - additional factors



```
levels(df$f.nr.employed) <-paste0("f.nr.employed-", levels(df$f.nr.employed))
summary(df$f.nr.employed)

## f.nr.employed-[4.96e+03,5.1e+03] f.nr.employed-(5.1e+03,5.19e+03]
## 1639 1003

## f.nr.employed-(5.19e+03,5.23e+03]
## 2344</pre>
```

### Llistat de variables continues i discretes:

```
vars<-names(df); vars</pre>
    [1] "age"
                             "job"
                                                 "marital"
##
    [4] "education"
                             "default"
                                                 "housing"
   [7] "loan"
                             "contact"
                                                 "month"
                             "duration"
                                                 "campaign"
## [10] "day_of_week"
        "pdays"
                             "previous"
                                                 "poutcome"
## [13]
## [16]
        "emp.var.rate"
                             "cons.price.idx"
                                                 "cons.conf.idx"
                                                 "у"
## [19] "euribor3m"
                             "nr.employed"
## [22] "num_missings"
                             "num_outliers"
                                                 "num_errors"
## [25] "f.season"
                             "minutes"
                                                 "f.age"
                                                 "f.pdays"
## [28] "f.duration"
                             "f.campaign"
## [31] "f.previous"
                                                 "f.cons.price.idx"
                             "f.emp.var.rate"
## [34] "f.cons.conf.idx"
                            "f.euribor3m"
                                                 "f.nr.employed"
# Variables continues
vars_con<-names(df)[c(1, 11:14, 16:20)]; vars_con</pre>
    [1] "age"
##
                           "duration"
                                             "campaign"
                                                               "pdays"
    [5] "previous"
                           "emp.var.rate"
                                             "cons.price.idx" "cons.conf.idx"
```

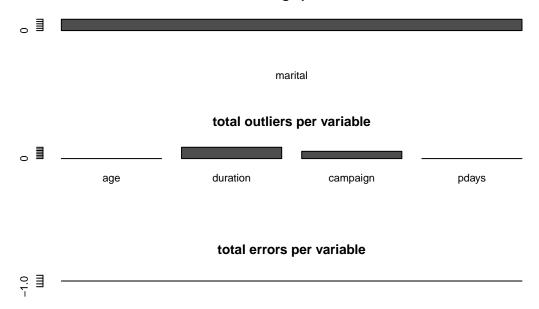
```
[9] "euribor3m"
                           "nr.employed"
# Variables discretes
vars_dis<-names(df)[c(2:10, 15, 21, 25, 27:36)]; vars_dis</pre>
    [1] "job"
                             "marital"
                                                 "education"
##
    [4] "default"
                             "housing"
                                                 "loan"
##
##
    [7] "contact"
                             "month"
                                                 "day_of_week"
       "poutcome"
##
   [10]
                                                 "f.season"
  [13] "f.age"
                             "f.duration"
                                                 "f.campaign"
  [16] "f.pdays"
                             "f.previous"
                                                 "f.emp.var.rate"
  [19] "f.cons.price.idx" "f.cons.conf.idx"
                                                 "f.euribor3m"
  [22] "f.nr.employed"
```

# DATA QUALITY REPORT:

### Per variable:

```
par(mfrow=c(3,1))
barplot( t(c(missings[, 3])), main="total missings per variable", xlab="marital")
barplot( t(c(outliers[, c(1, 11, 12, 13)])), main="total outliers per variable")
barplot( t(c(errors[, 13])), main="total errors per variable")
```

### total missings per variable



### Per individu:

Cap individu en te mes d'un. Es mostra en format taula el numero d'individus que tenen 0 i/o 1 (o mes) missings, errors i outliers. Per ultim, es mostren alguns dels individus que han tingut algun outlier i que aquest ha estat imputat.

```
par(mfrow=c(1,1))
table(df$num_missings)
##
##
      0
           1
        147
## 4839
table(df$num_errors)
##
##
      0
## 4986
table(df$num_outliers)
##
##
      0
           1
## 4977
           9
head(df[which(df\u00e4num_outliers>0), ], 2) #individus amb algun outlier
##
        age
                         job
                                     marital
                                                                education
## 5565
        39
                 job-admin. marital-married education-university.degree
        30 job-blue-collar marital-married
                                                       education-basic.9y
##
           default
                       housing
                                   loan
                                                   contact
                                                               month
## 5565 default-no housing-yes loan-no contact-telephone month-may
  9014 default-no housing-no loan-no contact-telephone month-jun
##
            day_of_week duration campaign pdays previous
                                                                       poutcome
## 5565 day_of_week-mon
                               14
                                             999
                                                         0 poutcome-nonexistent
   9014 day_of_week-thu
                                             999
                                                         0 poutcome-nonexistent
##
                               53
                                        NA
        emp.var.rate cons.price.idx cons.conf.idx euribor3m nr.employed
##
## 5565
                              93.994
                                             -36.4
                 1.1
                                                        4.857
                                                                   5191.0 y-no
                                                        4.866
## 9014
                 1.4
                              94.465
                                             -41.8
                                                                   5228.1 y-no
##
        num_missings num_outliers num_errors
                                                   f.season
                                                               minutes
                                            0 season-spring 0.2333333
## 5565
                   0
                                 1
## 9014
                   0
                                            0 season-summer 0.8833333
                                 1
                              f.duration f.campaign
##
                f.age
                                                           f.pdays
## 5565 f.age-(38,47] f.duration-[5,66]
                                               <NA> f.pdays-never
  9014 f.age-[18,32] f.duration-[5,66]
                                               <NA> f.pdays-never
##
              f.previous
                                     f.emp.var.rate
                                                               f.cons.price.idx
## 5565 f.previous-never f.emp.var.rate-(-0.1,1.1] f.cons.price.idx-(93.7,94]
## 9014 f.previous-never f.emp.var.rate-(1.1,1.4] f.cons.price.idx-(94,94.8]
                      f.cons.conf.idx
                                                   f.euribor3m
## 5565 f.cons.conf.idx-(-41.8,-36.4] f.euribor3m-(1.33,4.86]
## 9014 f.cons.conf.idx-(-42.7,-41.8] f.euribor3m-(4.86,4.96]
##
                             f.nr.employed
## 5565 f.nr.employed-(5.1e+03,5.19e+03]
## 9014 f.nr.employed-(5.19e+03,5.23e+03]
```

### **Outliers Multivariants:**

No hem aconseguit trobar una configuracio del aq.plot que ens doni una bona grafica per a veure les distancies de Mahalanobis i detecter outliers multivariants.

```
# Consider subset of numeric variables:
# summary(df[,vars_con])
vars_con_sub<-vars_con[c(1,2,3,6:10)]</pre>
```

```
x<-df[,vars_con_sub]
# aq.plot(x, delta=qchisq(0.995, df=ncol(x)) )</pre>
```

# **IMPUTATION:**

#### **Factors:**

De totes les variables discretes que hem analitzat, hem vist que el "marital" status es podria imputar facilment amb imputeMCA(), ja que els unknown (passats previament a NA) corresponen nomes una petita part de la mostra. El mateix fem amb la variable "loan". Com hem vist previament, els unknowns han estat considerats categoria propia en altres variables.

```
res.impf<-imputeMCA(df[,vars_dis], ncp=10)
# Original:
summary(df$marital)
## marital-divorced
                      marital-married
                                         marital-single
                                                                      NA's
##
                 554
                                  3046
                                                    1376
                                                                        10
summary(df$loan)
##
    loan-no loan-yes
                          NA's
       4080
                  769
                           137
# Amb dades imputades:
summary(res.impf$completeObs$marital)
## marital-divorced
                     marital-married
                                         marital-single
##
                 554
                                  3055
                                                    1377
summary(res.impf$completeObs$loan)
##
    loan-no loan-yes
##
       4217
                  769
# Acceptem la imputació:
df$loan<-res.impf$completeObs[,"marital"]</pre>
df$loan<-res.impf$completeObs[,"loan"]</pre>
#summary(df[,vars_dis])
```

#### **Numeric Variables:**

La variable numerica campaign te certs individus que han estat considerats outliers previament. Aqui els imputem mitjaçant la imputacio automatica imputePCA().

```
res.imp<-imputePCA(df[,vars_con], ncp=8)
# Original:
summary(df$campaign)
##
      Min. 1st Qu. Median
                               Mean 3rd Qu.
                                                Max.
                                                        NA's
     1.000
             1.000
                     2.000
                              2.535
                                      3.000
                                             25.000
                                                           9
# Amb dades imputades:
# Acceptem la imputació:
```

```
df$campaign<-res.imp$completeObs[,"campaign"]
#summary(df[,vars_con])</pre>
```

### **PROFILING:**

## CONTINOUS DESCRIPTION - Numeric Target (Duration):

La funcio d'R "condes" ens descriu la variable continua "duration" a partir d'altres variables quantitatives o de les variables categoriques. Aixo ho fa mitjançant els tres outputs diferenciats mes avall; etiquetats con a "quanti", "quali" i "\$category".

El primer dels quals (\$quanti) ens mostra la correlacio de la variable estudiada "duration" amb altres variables numeriques, mostrant nomes les correlacions que tenen un p-value per sota del llindar o nivell de significacio del 5% (en aquest cas). Com mes petit es el p-valor, menys evidencia hi ha de que la hipotesi nul.la sigui certa i mes segurs estem del rebuig de la hipotesi nul.la. Aquesta hipotesi nul.la H0 afirma que la correlacio o resultat obtingut es fruit d'una aleatorietat de les dades i no pot ser atribuible a una causa especifica. Per tant, a partir d'ara, direm que quan el p-valor esta per sota del nivell de significacio establert, els resultats son significatius.

Comentar que ens apareix el valor NA però no tenim cap valor en la nostra mostra (ho vam estar mirant a classe), tot i aixi no afecta al resultat obtingut, simplement l'obviem. De la mateixa manera obviem la correlació d'1 entre la duració de la trucada en segons i en minuts, ja que és una correlació perfecta deguda a una conversió d'unitats. Dit això, observem lleugeres correlacions negatives significatives (ordenades de més correlació positiva a no correlació i després a més correlació negativa) entre la duració de la trucada i la variable pdays, euribor3m, nr.employed i campaign. Es pot veure com la duració de la trucada augmenta com menys cops s'ha contactat al client en aquesta campanya (campaign), el quals és lògic perquè un client molt contactat estara cansat ja de rebre trucades. També es pot veure com la duració de la trucada augmenta com menys dies fa que s'ha contactat a un client en relació a una campanya anterior (pdays), el que pot estar relacionat amb l'interès del client per les diferents campanyes actuals que se li han exposat. Finalment tenim dos indicados socioecònomics que tenen una lleugera correlació negativa amb la duració de la trucada.

El segon output (\$quali) ens mostra els factors (variables categoriques) que estan mes relacionades amb la variable target "duration". Ens mostra els resultats significatius ordenats per factors de mes a menys relacionats la duracio. Obviant la discretitzacio de la duracio (f.duration) que obviament esta molt relacionada, observem com la la decisio final (y) del client a contractar un servei esta forca relacionada amb la duracio d'una trucada. Molt menys relacionades (pero llegurament) ho estan les variables "f.campaign", "month", aixi com altres indicadors socioeconomics.

El tercer output (\$category) ens indica una estimacio de les unitats que la durada de la trucada esta per sobre (+) o per sota (-) de la mitja global quan el registre pertany a la categoria en questio; ordenades per p-valor. Deixant de banda les categories de f.duration que son fruit de la discretitzacio, pot veure com quan el producte es contractat (y-yes), la duracio de la trucada esta 148 segons per sobre, com era d'esperar en una contractacio per telefon. Altres resultats obtinguts interessants son que la duracio de la trucada esta 72 segons per sobre quan s'ha contactat amb el client en aquesta campanya 1 o 2 cops (f-campaign-[0,2]) i que tambe augmenta en 38 segons quan el resultat de la campanya anterior va ser positiu pel mateix client (poutcome-success). Tambe podem destacar el mes d'abril (month-apr), en el qual les duracions de les trucades estan 28 segons per sobre de la mitja, o la primavera (season-spring) amb 18 segons per sobre de la mitja. D'altra banda podem veure com en el mes d'agost (month-aug) la duracio de les trucades esta 28 segons per sota la mitja, en el novembre (month-nov) 20 segons per sota, i que els clients que mai han estat contactats abans (f.pdays-never) estan 28 segons menys al telefon que la mitja.

El oneway.test d'R ens compara si dues o mes mostres de variables amb distribucio normal tenen o no la mateixa mitjana (no cal assumir igualtat de variancies pels grups implicats que es comparen). En aquest cas ens permet concluore que la mitjana de la durada de la trucada en els casos que s'ha contractat el servei es significativament diferent a la dels casos en els quals no s'ha contractat el servei. L'estadadistic de contrast

```
segueix una distribucio F de Fisher i pren el valor 447.7, que es molt significatiu (p-value < 1e-16).
```

pos\_duration<-which(names(df)=="duration"); pos\_duration</pre>

#### ## [1] 11

```
condes(df, num.var=pos_duration, proba = 0.05)
```

```
## $quanti
##
                correlation
                                 p.value
## <NA>
                         NΑ
                                      NΑ
                 1.00000000 0.000000e+00
## minutes
                -0.03478274 1.404179e-02
## pdays
## euribor3m
                -0.03512962 1.311237e-02
## num outliers -0.04065979 4.085021e-03
## nr.employed -0.04831097 6.438109e-04
##
  campaign
                -0.07479201 1.241577e-07
##
## $quali
##
                             R.2
                                      p.value
## f.duration
                    0.855794028 0.000000e+00
## y
                    0.164777620 3.759496e-197
## f.campaign
                    0.006187857
                                8.807648e-07
## f.cons.conf.idx
                   0.004067507
                                 1.465565e-04
## f.nr.employed
                    0.002912867
                                 6.975062e-04
## f.cons.price.idx 0.003246051
                                1.031905e-03
## month
                    0.005064462
                                 2.674014e-03
## f.euribor3m
                    0.002462249 6.473152e-03
## f.season
                                7.627865e-03
                    0.002391458
## poutcome
                    0.001851161 9.887924e-03
## day_of_week
                    0.002352912 1.942616e-02
## f.pdays
                    0.001214169
                                4.846375e-02
## f.emp.var.rate
                    0.001574759 4.916221e-02
##
## $category
##
                                        Estimate
                                                       p.value
## f.duration-(482,1.58e+03]
                                      493.613665 0.000000e+00
## y-yes
                                      148.441504 3.759496e-197
                                      134.394010 8.476109e-56
## f.duration-(316,482]
## f.campaign-(5,25]
                                       14.794426
                                                  2.638343e-06
## season-spring
                                       17.952283 5.877554e-04
## poutcome-success
                                       38.359032 5.480212e-03
## f.campaign-[0,2]
                                                  7.136472e-03
                                       71.765001
## f.nr.employed-[4.96e+03,5.1e+03]
                                        9.017147
                                                  8.355482e-03
## f.duration-(235,316]
                                       22.169724 9.317648e-03
## f.cons.conf.idx-[-50.8,-42.7]
                                       14.076002 1.238528e-02
## NA
                                      132.886872 1.491425e-02
## month-may
                                        9.867780 1.599295e-02
## f.cons.price.idx-(93.7,94]
                                       11.621760 2.081111e-02
## f.pdays-[0,7]
                                       16.460640 2.262020e-02
## f.cons.conf.idx-(-41.8,-36.4]
                                       16.349262 2.392080e-02
## month-apr
                                       27.731238 2.403940e-02
## education-high.school
                                        9.358222 4.228302e-02
## day_of_week-wed
                                       13.376659 4.495212e-02
## month-nov
                                      -20.376410 4.421467e-02
```

```
## education-university.degree
                                      -14.109465 2.294239e-02
## f.emp.var.rate-(1.1,1.4]
                                      -10.129703 2.036833e-02
## day of week-mon
                                      -15.133836 1.838350e-02
## season-summer
                                       -3.899443 1.752241e-02
## f.pdays-never
                                      -27.755294
                                                  1.396985e-02
## f.cons.conf.idx-(-36.4, -26.9]
                                      -14.862166 7.024095e-03
## f.cons.conf.idx-(-42.7,-41.8]
                                      -15.563098 4.192506e-03
## NA
                                     -154.540521 4.085021e-03
## f.euribor3m-(4.96,5]
                                      -19.423787
                                                  1.079935e-03
## month-aug
                                      -28.383026 6.707022e-04
## f.nr.employed-(5.19e+03,5.23e+03]
                                      -16.466612 1.395228e-04
## f.cons.price.idx-(93.1,93.7]
                                      -22.699701
                                                  8.027710e-05
## f.duration-(177,235]
                                      -47.149040 5.572506e-08
## f.duration-(138,177]
                                      -94.204089 1.668437e-27
## f.duration-(101,138]
                                     -131.656740 5.328783e-54
## f.duration-(66,101]
                                     -167.038569 1.102835e-85
## f.duration-[5,66]
                                     -210.128961 1.924209e-141
## y-no
                                     -148.441504 3.759496e-197
# mitjana de la duració per categoria de la duració
# tapply(df$duration, df$f.duration, mean)
# duració global
summary(df$duration)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                              Max.
##
       5.0
             101.0
                     177.0
                             250.6
                                     316.0
                                            1580.0
# mitjana de la duració per categoria de la y
tapply(df$duration, df$y, mean)
##
       y-no
               y-yes
## 217.4563 514.3393
oneway.test(df$duration~df$y)
##
##
   One-way analysis of means (not assuming equal variances)
##
## data: df$duration and df$y
## F = 447.7, num df = 1.00, denom df = 605.83, p-value < 2.2e-16
```

## CATEGORICAL DESCRIPTION - Factor (Y, Final Decision):

La funcio d'R "catdes" ens descriu la variable categorica "y" a partir d'altres variables categoriques o de les variables quantitatives. Aixo ho fa mitjancant outputs diferenciats mes avall. Notem que el nostre llindar de signifiacio en aquest cas es del 0.025 per tal de limitar una mica la gran quantitat de resultats mostrats.

L'apartat "Link between the cluster variable and the categorical variables (chi-square test)" ens mostra les variables categoriques que han caracteritzat al factor "y" ordenades de mes a menys caracteritzacio del factor (de menys a mes p-value). La columna "df" mostra els Degrees of Freedom, que corresponen amb el nombre de categories del factor menys 1. Les variables categoriques que han influenciat mes en la decisio final (y) son la f.duration (pero es una dada que s'obte a posteriori de la trucada, no ens serveix per a generar un perfil de client), f.pdays (nombre de dies des de l'ultim contacte), poutcome (si la ultima campanya va ser acceptada per aquest client o no), el mes (month), previous (si havia estat contactat o no abans d'aquesta campanya), diferents indicadors socioeconomics, contact (via de contacte), el job (feina), etc.

L'apartat "Description of each cluster by the categories" ens mostra per a cada categoria de la "y" (y-yes, y-no), una descripcio de les variables categoriques per tal de poder estudiar-ne el seu enllac. La primera columna, Cla/Mod, en mostra el tant percent d'individus corresponents a la categoria de la fila que pertanyen tambe al cluster. Per altra banda, la segona columna (Mod/Cla) mostra la operacio inversa, es a dir, dels individus que pertanyen al cluster quin tant percent pertany tambe a la categoria de la fila en questio. La columna Global ens mostra quin tant percent de la poblacio total posseix la caracteristica de la fila. Per acabar, v.test ens indica si la categoria de la fila es troba sobre representada (v.test>0) o menys representada (v.test<0) entre els individus que conformen el cluster. Per al cluster "y-no" podem deduir que el fet de la haver realitzat un contacte previ i que es realitzin menys de 2 contactes en l'actual campanya te un gran pes en la decisio final. A mes, per a valors positius del euribor (desfavorables per al client) hi a una tendencia a l'alca a refusar el producte. El mateix succeix amb valor elevats del IPC i quan es redueix la taxa d'ocupacio. En relacio al cluster "y-yes" les presones que van ser contactades fa menys de 7 dies per altres campanyes i aquelles que el resultat en la campanya anterior va ser exitos tendeixen a donar un si com a resposta. Tambe quan els valors socioeconomics es mostren favorables al client aquest tendeix a adquirir el producte amb una major probabilitat.

L'apartat "Link between the cluster variable and the quantitative variables" ens mostra les variables quantitatives que han caracteritzat al factor "y" ordenades de mes a menys caracteritzacio del factor (de menys a mes correlacio). Les variables quantitatives que han influenciat mes en la decisio final (y) son la duration i minutes (pero son dades que s'obtenen a posteriori de la trucada, no ens serveixen per a generar un perfil de client), pdays (nombre de dies des de l'ultim contacte), previous (si havia estat contactat o no abans d'aquesta campanya), diferents indicadors socioeconomics, etc.

L'apartat "Description of each cluster by quantitative variables". D'aquesta part de l'analisi no en podem extreure informacio dels individus que conformen el cluster "y-no", donat que els valors que es presenten de les categories dins el cluster i de manera general no presenten una diversificacio notable. Per altra banda del cluster "y-yes" si que en poden extreure informacio, podem veure que la mitjana de la duracio de les trucades dels individus del cluster duplica la mitjana global (tot i que la duracio no ens pot servir com a indicador, donat que aquelles persones no interesades en el producte refusen la trucada mes rapidament o el fet de contractar el proucte pot repercutir en un increment de la durada). Altres factors con l'euribor o la taxa de variacio de la ocupacio tambe tenen un impacte en la decisio final.

```
pos_y<-which(names(df)=="y"); pos_y</pre>
## [1] 21
catdes(df, num.var=pos y, proba = 0.025)
##
## Link between the cluster variable and the categorical variables (chi-square test)
##
  _____
##
                          p.value df
## f.duration
                    2.794524e-159
                                  7
## f.pdays
                    9.362887e-100
                                   2
## poutcome
                     3.053387e-95
                                   2
## f.nr.employed
                     1.703080e-89
                                   2
## f.euribor3m
                     5.470503e-79
## month
                     1.690776e-65
                     7.969229e-62
## f.emp.var.rate
                                   3
## f.previous
                     5.590487e-45
## f.cons.price.idx
                    5.572278e-38
## f.cons.conf.idx
                     4.786677e-23
## contact
                     2.110136e-21
                                  1
## job
                     8.420857e-16 11
## default
                     9.768051e-13
## f.season
                     1.176664e-10
```

```
## f.age
                    7.936723e-09 3
## education
                     6.361426e-06 6
## marital
                     1.452705e-04 3
## f.campaign
                     1.037416e-03 3
## Description of each cluster by the categories
## $`y-no`
##
                                                   Cla/Mod
                                                              Mod/Cla
                                                   90.64255 98.4195078
## f.pdays=f.pdays-never
## f.nr.employed=f.nr.employed-(5.19e+03,5.23e+03] 94.70990 50.1241815
## f.previous=f.previous-never
                                                   91.01769 89.4558591
## poutcome=poutcome-nonexistent
                                                   91.01769 89.4558591
## f.duration=f.duration-[5,66]
                                                   99.52607 14.2244299
## f.emp.var.rate=f.emp.var.rate-(1.1,1.4]
                                                   94.52055 40.5057575
## contact=contact-telephone
                                                   94.31330 39.6929329
## f.duration=f.duration-(66,101]
                                                   98.38449 13.7502822
## f.cons.price.idx=f.cons.price.idx-(93.7,94]
                                                   94.11765 38.6543238
## f.nr.employed=f.nr.employed-(5.1e+03,5.19e+03]
                                                  96.11167 21.7656356
## f.emp.var.rate=f.emp.var.rate-(-0.1,1.1]
                                                   96.10778 21.7430571
## f.cons.conf.idx=f.cons.conf.idx-(-42.7,-41.8]
                                                   96.07032 20.9753895
## default=default-unknown
                                                   95.05814 22.1494694
## month=month-may
                                                   93.33716 36.6899977
## f.euribor3m=f.euribor3m-(4.86,4.96]
                                                   94.51327 24.1137954
## f.euribor3m=f.euribor3m-(4.96,5]
                                                  94.36620 24.2041093
## f.duration=f.duration-(101,138]
                                                   96.01911 13.6148115
## job=job-blue-collar
                                                   93.74457 24.3621585
## f.euribor3m=f.euribor3m-(1.33,4.86]
                                                   92.70123 30.6841273
## f.duration=f.duration-(138,177]
                                                   94.79675 13.1632423
## f.cons.price.idx=f.cons.price.idx-(93.1,93.7]
                                                  92.90976 22.7816663
## f.age=f.age-(38,47]
                                                   92.54098 25.4910815
## f.campaign=f.campaign-(5,25]
                                                  94.18886 8.7830210
## education=education-basic.9y
                                                  92.72727 14.9695191
## marital=marital-married
                                                  89.92121 61.8424023
## month=month-jul
                                                   91.31484 17.0918943
## education=education-basic.6y
                                                  93.07958 6.0736058
## f.season=season-spring
                                                  90.08030 43.0571235
## f.age=f.age-(32,38]
                                                  90.62241 24.6556785
## poutcome=poutcome-failure
                                                   85.53459 9.2120117
## education=education-unknown
                                                   82.68398 4.3124859
## f.cons.price.idx=f.cons.price.idx-(94,94.8]
                                                   85.41667 12.9600361
## f.campaign=f.campaign-[0,2]
                                                   87.94222 67.3515466
## f.season=season-winter
                                                   65.38462 0.3838338
## month=month-dec
                                                   65.38462 0.3838338
## education=education-university.degree
                                                   86.51226 28.6746444
                                                   84.09475 11.2214947
## f.emp.var.rate=f.emp.var.rate-(-1.8,-0.1]
## f.duration=f.duration-(316,482]
                                                   83.87097 11.7407993
## job=job-retired
                                                   78.92157 3.6351321
## marital=marital-single
                                                   85.68314 26.6200045
## f.age=f.age-[18,32]
                                                   85.35503 26.0555430
## f.pdays=f.pdays-(>7)
                                                   53.33333 0.3612554
## job=job-student
                                                  70.00000 1.5804922
## month=month-apr
                                                  78.70968 5.5091443
## f.season=season-autumn
                                                  82.25564 12.3504177
```

```
## month=month-sep
                                                   57.37705 0.7902461
## month=month-mar
                                                   57.57576 0.8579815
## f.cons.conf.idx=f.cons.conf.idx-(-36.4, -26.9]
                                                   81.22318 17.0918943
## default=default-no
                                                   87.20283 77.8505306
## f.previous=f.previous-1
                                                   77.53906 8.9636487
## month=month-oct
                                                   54.63918 1.1966584
## f.previous=f.previous-(>1)
                                                   57.85124 1.5804922
## contact=contact-cellular
                                                   85.55413 60.3070671
## f.cons.price.idx=f.cons.price.idx-[92.2,93.1]
                                                   80.48261 25.6039738
## f.emp.var.rate=f.emp.var.rate-[-3.4,-1.8]
                                                   78.59532 26.5296907
## f.pdays=f.pdays-[0,7]
                                                   36.73469
                                                            1.2192368
## poutcome=poutcome-success
                                                   37.82051
                                                             1.3321291
## f.euribor3m=f.euribor3m-[0.635,1.33]
                                                   74.16268 20.9979679
## f.nr.employed=f.nr.employed-[4.96e+03,5.1e+03]
                                                   75.96095 28.1101829
## f.duration=f.duration-(482,1.58e+03]
                                                   59.13462 8.3314518
##
                                                       Global
                                                                    p.value
## f.pdays=f.pdays-never
                                                   96.4500602 2.410684e-59
## f.nr.employed=f.nr.employed-(5.19e+03,5.23e+03] 47.0116326 2.158488e-37
## f.previous=f.previous-never
                                                   87.3044525 1.438650e-30
## poutcome=poutcome-nonexistent
                                                   87.3044525 1.438650e-30
## f.duration=f.duration-[5,66]
                                                   12.6955475 1.487124e-30
## f.emp.var.rate=f.emp.var.rate-(1.1,1.4]
                                                   38.0665864 1.340920e-25
                                                   37.3846771 3.447929e-23
## contact=contact-telephone
## f.duration=f.duration-(66,101]
                                                   12.4147613 7.696941e-22
## f.cons.price.idx=f.cons.price.idx-(93.7,94]
                                                   36.4821500 7.057265e-21
## f.nr.employed=f.nr.employed-(5.1e+03,5.19e+03]
                                                   20.1163257 1.424235e-19
## f.emp.var.rate=f.emp.var.rate-(-0.1,1.1]
                                                   20.0962696 1.574618e-19
## f.cons.conf.idx=f.cons.conf.idx-(-42.7,-41.8]
                                                   19.3943041 1.401017e-18
## default=default-unknown
                                                   20.6979543 1.230324e-14
## month=month-may
                                                   34.9177698 1.726364e-14
## f.euribor3m=f.euribor3m-(4.86,4.96]
                                                   22.6634577 1.693548e-13
## f.euribor3m=f.euribor3m-(4.96,5]
                                                   22.7837946 6.639818e-13
## f.duration=f.duration-(101,138]
                                                   12.5952667 1.010774e-11
                                                   23.0846370 1.884818e-10
## job=job-blue-collar
## f.euribor3m=f.euribor3m-(1.33,4.86]
                                                   29.4023265 6.796806e-09
## f.duration=f.duration-(138,177]
                                                   12.3345367 5.342775e-08
## f.cons.price.idx=f.cons.price.idx-(93.1,93.7]
                                                   21.7809868 4.701642e-07
## f.age=f.age-(38,47]
                                                   24.4685118 9.135370e-07
## f.campaign=f.campaign-(5,25]
                                                    8.2831929 1.084374e-04
## education=education-basic.9y
                                                   14.3401524 1.876745e-04
## marital=marital-married
                                                   61.0910550 2.314946e-03
## month=month-jul
                                                   16.6265544 1.093857e-02
## education=education-basic.6y
                                                    5.7962294 1.335614e-02
## f.season=season-spring
                                                   42.4588849 1.562952e-02
## f.age=f.age-(32,38]
                                                   24.1676695 2.153346e-02
## poutcome=poutcome-failure
                                                    9.5667870 1.986516e-02
## education=education-unknown
                                                    4.6329723 4.270710e-03
## f.cons.price.idx=f.cons.price.idx-(94,94.8]
                                                   13.4777377 3.445794e-03
## f.campaign=f.campaign-[0,2]
                                                   68.0304854 3.359672e-03
## f.season=season-winter
                                                    0.5214601 1.657365e-03
## month=month-dec
                                                    0.5214601 1.657365e-03
## education=education-university.degree
                                                   29.4424388 9.565525e-04
## f.emp.var.rate=f.emp.var.rate-(-1.8,-0.1]
                                                   11.8531889 1.984797e-04
## f.duration=f.duration-(316,482]
                                                   12.4348175 6.392065e-05
```

```
## job=job-retired
                                                    4.0914561 2.982842e-05
## marital=marital-single
                                                   27.5972724 2.055013e-05
## f.age=f.age-[18,32]
                                                   27.1159246 3.567657e-06
## f.pdays=f.pdays-(>7)
                                                    0.6016847 1.202754e-06
                                                    2.0056157 2.508620e-07
## job=job-student
## month=month-apr
                                                    6.2174087 1.047741e-07
## f.season=season-autumn
                                                   13.3373446 5.062563e-08
                                                    1.2234256 3.276634e-10
## month=month-sep
## month=month-mar
                                                    1.3237064 7.597160e-11
## f.cons.conf.idx=f.cons.conf.idx-(-36.4,-26.9]
                                                   18.6923385 1.352020e-14
## default=default-no
                                                   79.3020457 1.230324e-14
                                                   10.2687525 7.464256e-15
## f.previous=f.previous-1
                                                    1.9454473 8.959508e-18
## month=month-oct
## f.previous=f.previous-(>1)
                                                    2.4267950 1.002106e-18
## contact=contact-cellular
                                                   62.6153229 3.447929e-23
                                                   28.2591256 3.335427e-29
## f.cons.price.idx=f.cons.price.idx-[92.2,93.1]
## f.emp.var.rate=f.emp.var.rate-[-3.4,-1.8]
                                                   29.9839551 1.289177e-46
## f.pdays=f.pdays-[0,7]
                                                    2.9482551 6.682675e-54
                                                    3.1287605 2.946325e-55
## poutcome=poutcome-success
## f.euribor3m=f.euribor3m-[0.635,1.33]
                                                   25.1504212 3.042037e-70
## f.nr.employed=f.nr.employed-[4.96e+03,5.1e+03]
                                                   32.8720417 1.759629e-84
## f.duration=f.duration-(482,1.58e+03]
                                                   12.5150421 4.894928e-100
##
                                                       v.test
## f.pdays=f.pdays-never
                                                    16.245323
## f.nr.employed=f.nr.employed-(5.19e+03,5.23e+03]
                                                    12.778626
## f.previous=f.previous-never
                                                    11.492513
## poutcome=poutcome-nonexistent
                                                     11.492513
## f.duration=f.duration-[5,66]
                                                     11.489650
## f.emp.var.rate=f.emp.var.rate-(1.1,1.4]
                                                    10.458406
## contact=contact-telephone
                                                     9.918824
## f.duration=f.duration-(66,101]
                                                     9.603908
## f.cons.price.idx=f.cons.price.idx-(93.7,94]
                                                     9.372891
## f.nr.employed=f.nr.employed-(5.1e+03,5.19e+03]
                                                     9.050417
## f.emp.var.rate=f.emp.var.rate-(-0.1,1.1]
                                                     9.039450
## f.cons.conf.idx=f.cons.conf.idx-(-42.7, -41.8]
                                                     8.797336
## default=default-unknown
                                                     7.712857
## month=month-may
                                                     7.669524
## f.euribor3m=f.euribor3m-(4.86,4.96]
                                                     7.370998
## f.euribor3m=f.euribor3m-(4.96,5]
                                                     7.186654
## f.duration=f.duration-(101,138]
                                                     6.804960
## job=job-blue-collar
                                                     6.370444
## f.euribor3m=f.euribor3m-(1.33,4.86]
                                                     5.795870
## f.duration=f.duration-(138,177]
                                                     5.439509
## f.cons.price.idx=f.cons.price.idx-(93.1,93.7]
                                                     5.038105
## f.age=f.age-(38,47]
                                                     4.909404
## f.campaign=f.campaign-(5,25]
                                                     3.870893
## education=education-basic.9v
                                                     3.735055
## marital=marital-married
                                                     3.046536
## month=month-jul
                                                     2.544655
## education=education-basic.6y
                                                     2.474129
## f.season=season-spring
                                                     2,417454
## f.age=f.age-(32,38]
                                                     2.298498
## poutcome=poutcome-failure
                                                    -2.328885
## education=education-unknown
                                                    -2.857442
```

```
## f.cons.price.idx=f.cons.price.idx-(94,94.8]
                                                     -2.924889
## f.campaign=f.campaign-[0,2]
                                                     -2.932757
                                                     -3.145618
## f.season=season-winter
## month=month-dec
                                                     -3.145618
## education=education-university.degree
                                                     -3.303003
## f.emp.var.rate=f.emp.var.rate-(-1.8,-0.1]
                                                     -3.720944
## f.duration=f.duration-(316,482]
                                                     -3.997849
## job=job-retired
                                                     -4.174772
## marital=marital-single
                                                     -4.258828
## f.age=f.age-[18,32]
                                                     -4.635100
## f.pdays=f.pdays-(>7)
                                                     -4.855183
## job=job-student
                                                     -5.157057
## month=month-apr
                                                     -5.318243
## f.season=season-autumn
                                                     -5.449099
## month=month-sep
                                                     -6.285090
## month=month-mar
                                                     -6.508368
## f.cons.conf.idx=f.cons.conf.idx-(-36.4,-26.9]
                                                     -7.700814
## default=default-no
                                                     -7.712857
## f.previous=f.previous-1
                                                     -7.776358
## month=month-oct
                                                     -8.586582
## f.previous=f.previous-(>1)
                                                     -8.834875
## contact=contact-cellular
                                                     -9.918824
## f.cons.price.idx=f.cons.price.idx-[92.2,93.1]
                                                    -11.217779
## f.emp.var.rate=f.emp.var.rate-[-3.4,-1.8]
                                                    -14.336770
## f.pdays=f.pdays-[0,7]
                                                    -15.457815
## poutcome=poutcome-success
                                                    -15.657639
## f.euribor3m=f.euribor3m-[0.635,1.33]
                                                    -17.718064
## f.nr.employed=f.nr.employed-[4.96e+03,5.1e+03]
                                                    -19.475855
## f.duration=f.duration-(482,1.58e+03]
                                                    -21.231431
##
## $`y-yes`
##
                                                       Cla/Mod
                                                                  Mod/Cla
## f.duration=f.duration-(482,1.58e+03]
                                                    40.8653846 45.7809695
## f.nr.employed=f.nr.employed-[4.96e+03,5.1e+03]
                                                    24.0390482 70.7360862
## f.euribor3m=f.euribor3m-[0.635,1.33]
                                                    25.8373206 58.1687612
## poutcome=poutcome-success
                                                    62.1794872 17.4147217
## f.pdays=f.pdays-[0,7]
                                                    63.2653061 16.6965889
## f.emp.var.rate=f.emp.var.rate-[-3.4,-1.8]
                                                    21.4046823 57.4506284
## f.cons.price.idx=f.cons.price.idx-[92.2,93.1]
                                                    19.5173882 49.3716338
## contact=contact-cellular
                                                    14.4458680 80.9694794
## f.previous=f.previous-(>1)
                                                    42.1487603 9.1561939
## month=month-oct
                                                    45.3608247 7.8994614
## f.previous=f.previous-1
                                                    22.4609375 20.6463196
## default=default-no
                                                    12.7971674 90.8438061
## f.cons.conf.idx=f.cons.conf.idx-(-36.4,-26.9]
                                                    18.7768240 31.4183124
## month=month-mar
                                                    42.42424 5.0269300
## month=month-sep
                                                    42.6229508 4.6678636
## f.season=season-autumn
                                                    17.7443609 21.1849192
## month=month-apr
                                                    21.2903226 11.8491921
## job=job-student
                                                    30.0000000 5.3859964
## f.pdays=f.pdays-(>7)
                                                    46.6666667 2.5134650
## f.age=f.age-[18,32]
                                                    14.6449704 35.5475763
## marital=marital-single
                                                    14.3168605 35.3680431
## job=job-retired
                                                    21.0784314 7.7199282
```

```
## f.duration=f.duration-(316,482]
                                                   16.1290323 17.9533214
## f.emp.var.rate=f.emp.var.rate-(-1.8,-0.1]
                                                   15.9052453 16.8761221
## education=education-university.degree
                                                   13.4877384 35.5475763
## f.season=season-winter
                                                   34.6153846 1.6157989
## month=month-dec
                                                   34.6153846 1.6157989
## f.campaign=f.campaign-[0,2]
                                                   12.0577830 73.4290844
## f.cons.price.idx=f.cons.price.idx-(94,94.8]
                                                   14.5833333 17.5942549
## education=education-unknown
                                                   17.3160173 7.1813285
## poutcome=poutcome-failure
                                                   14.4654088 12.3877917
## f.age=f.age-(32,38]
                                                    9.3775934 20.2872531
## f.season=season-spring
                                                    9.9196977 37.7019749
## education=education-basic.6y
                                                    6.9204152 3.5906643
## month=month-jul
                                                    8.6851628 12.9263914
                                                   10.0787919 55.1166966
## marital=marital-married
## education=education-basic.9y
                                                    7.2727273 9.3357271
## f.campaign=f.campaign-(5,25]
                                                    5.8111380 4.3087971
## f.age=f.age-(38,47]
                                                    7.4590164 16.3375224
## f.cons.price.idx=f.cons.price.idx-(93.1,93.7]
                                                    7.0902394 13.8240575
## f.duration=f.duration-(138,177]
                                                    5.2032520 5.7450628
## f.euribor3m=f.euribor3m-(1.33,4.86]
                                                    7.2987722 19.2100539
## job=job-blue-collar
                                                    6.2554301 12.9263914
## f.duration=f.duration-(101,138]
                                                    3.9808917 4.4883303
## f.euribor3m=f.euribor3m-(4.96,5]
                                                    5.6338028 11.4901257
## f.euribor3m=f.euribor3m-(4.86,4.96]
                                                    5.4867257 11.1310592
                                                    6.6628374 20.8258528
## month=month-may
## default=default-unknown
                                                    4.9418605 9.1561939
## f.cons.conf.idx=f.cons.conf.idx-(-42.7,-41.8]
                                                    3.9296794 6.8222621
## f.emp.var.rate=f.emp.var.rate-(-0.1,1.1]
                                                    3.8922156 7.0017953
## f.nr.employed=f.nr.employed-(5.1e+03,5.19e+03]
                                                    3.8883350 7.0017953
## f.cons.price.idx=f.cons.price.idx-(93.7,94]
                                                    5.8823529 19.2100539
## f.duration=f.duration-(66,101]
                                                    1.6155089 1.7953321
## contact=contact-telephone
                                                    5.6866953 19.0305206
## f.emp.var.rate=f.emp.var.rate-(1.1,1.4]
                                                    5.4794521 18.6714542
## f.duration=f.duration-[5,66]
                                                    0.4739336 0.5385996
## f.previous=f.previous-never
                                                    8.9823110 70.1974865
## poutcome=poutcome-nonexistent
                                                    8.9823110 70.1974865
## f.nr.employed=f.nr.employed-(5.19e+03,5.23e+03]
                                                   5.2901024 22.2621185
## f.pdays=f.pdays-never
                                                    9.3574548 80.7899461
##
                                                       Global
                                                                    p.value
## f.duration=f.duration-(482,1.58e+03]
                                                   12.5150421 4.894928e-100
## f.nr.employed=f.nr.employed-[4.96e+03,5.1e+03]
                                                   32.8720417 1.759629e-84
                                                   25.1504212 3.042037e-70
## f.euribor3m=f.euribor3m-[0.635,1.33]
## poutcome=poutcome-success
                                                    3.1287605 2.946325e-55
## f.pdays=f.pdays-[0,7]
                                                    2.9482551 6.682675e-54
## f.emp.var.rate=f.emp.var.rate-[-3.4,-1.8]
                                                   29.9839551 1.289177e-46
## f.cons.price.idx=f.cons.price.idx-[92.2,93.1]
                                                   28.2591256 3.335427e-29
## contact=contact-cellular
                                                   62.6153229 3.447929e-23
## f.previous=f.previous-(>1)
                                                    2.4267950 1.002106e-18
## month=month-oct
                                                    1.9454473 8.959508e-18
## f.previous=f.previous-1
                                                   10.2687525 7.464256e-15
## default=default-no
                                                   79.3020457 1.230324e-14
## f.cons.conf.idx=f.cons.conf.idx-(-36.4,-26.9]
                                                   18.6923385 1.352020e-14
## month=month-mar
                                                    1.3237064 7.597160e-11
                                                    1.2234256 3.276634e-10
## month=month-sep
```

```
## f.season=season-autumn
                                                   13.3373446 5.062563e-08
## month=month-apr
                                                    6.2174087 1.047741e-07
                                                    2.0056157 2.508620e-07
## job=job-student
## f.pdays=f.pdays-(>7)
                                                    0.6016847 1.202754e-06
## f.age=f.age-[18,32]
                                                   27.1159246 3.567657e-06
## marital=marital-single
                                                   27.5972724 2.055013e-05
## job=job-retired
                                                    4.0914561 2.982842e-05
## f.duration=f.duration-(316,482]
                                                   12.4348175 6.392065e-05
## f.emp.var.rate=f.emp.var.rate-(-1.8,-0.1]
                                                   11.8531889 1.984797e-04
## education=education-university.degree
                                                   29.4424388 9.565525e-04
## f.season=season-winter
                                                    0.5214601 1.657365e-03
## month=month-dec
                                                    0.5214601 1.657365e-03
## f.campaign=f.campaign-[0,2]
                                                   68.0304854 3.359672e-03
## f.cons.price.idx=f.cons.price.idx-(94,94.8]
                                                   13.4777377 3.445794e-03
## education=education-unknown
                                                    4.6329723 4.270710e-03
## poutcome=poutcome-failure
                                                    9.5667870 1.986516e-02
## f.age=f.age-(32,38]
                                                   24.1676695 2.153346e-02
## f.season=season-spring
                                                   42.4588849 1.562952e-02
## education=education-basic.6y
                                                    5.7962294 1.335614e-02
## month=month-jul
                                                   16.6265544 1.093857e-02
## marital=marital-married
                                                   61.0910550 2.314946e-03
## education=education-basic.9y
                                                   14.3401524 1.876745e-04
## f.campaign=f.campaign-(5,25]
                                                    8.2831929 1.084374e-04
## f.age=f.age-(38,47]
                                                   24.4685118 9.135370e-07
## f.cons.price.idx=f.cons.price.idx-(93.1,93.7]
                                                   21.7809868 4.701642e-07
## f.duration=f.duration-(138,177]
                                                   12.3345367 5.342775e-08
## f.euribor3m=f.euribor3m-(1.33,4.86]
                                                   29.4023265 6.796806e-09
## job=job-blue-collar
                                                   23.0846370 1.884818e-10
## f.duration=f.duration-(101,138]
                                                   12.5952667 1.010774e-11
## f.euribor3m=f.euribor3m-(4.96,5]
                                                   22.7837946 6.639818e-13
## f.euribor3m=f.euribor3m-(4.86,4.96]
                                                   22.6634577 1.693548e-13
## month=month-may
                                                   34.9177698 1.726364e-14
## default-default-unknown
                                                   20.6979543 1.230324e-14
## f.cons.conf.idx=f.cons.conf.idx-(-42.7,-41.8]
                                                   19.3943041 1.401017e-18
## f.emp.var.rate=f.emp.var.rate-(-0.1,1.1]
                                                   20.0962696 1.574618e-19
## f.nr.employed=f.nr.employed-(5.1e+03,5.19e+03]
                                                   20.1163257 1.424235e-19
## f.cons.price.idx=f.cons.price.idx-(93.7,94]
                                                   36.4821500 7.057265e-21
## f.duration=f.duration-(66,101]
                                                   12.4147613 7.696941e-22
## contact=contact-telephone
                                                   37.3846771 3.447929e-23
## f.emp.var.rate=f.emp.var.rate-(1.1,1.4]
                                                   38.0665864 1.340920e-25
## f.duration=f.duration-[5,66]
                                                   12.6955475 1.487124e-30
## f.previous=f.previous-never
                                                   87.3044525 1.438650e-30
## poutcome=poutcome-nonexistent
                                                   87.3044525 1.438650e-30
## f.nr.employed=f.nr.employed-(5.19e+03,5.23e+03] 47.0116326 2.158488e-37
## f.pdays=f.pdays-never
                                                   96.4500602 2.410684e-59
##
                                                       v.test
## f.duration=f.duration-(482,1.58e+03]
                                                    21.231431
## f.nr.employed=f.nr.employed-[4.96e+03,5.1e+03]
                                                    19.475855
## f.euribor3m=f.euribor3m-[0.635,1.33]
                                                    17.718064
## poutcome=poutcome-success
                                                    15.657639
## f.pdays=f.pdays-[0,7]
                                                    15.457815
## f.emp.var.rate=f.emp.var.rate-[-3.4,-1.8]
                                                    14.336770
## f.cons.price.idx=f.cons.price.idx-[92.2,93.1]
                                                    11.217779
## contact=contact-cellular
                                                     9.918824
```

```
## f.previous=f.previous-(>1)
                                                      8.834875
                                                      8.586582
## month=month-oct
                                                      7.776358
## f.previous=f.previous-1
## default=default-no
                                                      7.712857
## f.cons.conf.idx=f.cons.conf.idx-(-36.4,-26.9]
                                                      7.700814
## month=month-mar
                                                      6.508368
## month-month-sep
                                                      6.285090
## f.season=season-autumn
                                                      5.449099
## month=month-apr
                                                      5.318243
## job=job-student
                                                      5.157057
## f.pdays=f.pdays-(>7)
                                                      4.855183
## f.age=f.age-[18,32]
                                                      4.635100
## marital=marital-single
                                                      4.258828
## job=job-retired
                                                      4.174772
## f.duration=f.duration-(316,482]
                                                      3.997849
## f.emp.var.rate=f.emp.var.rate-(-1.8,-0.1]
                                                      3.720944
## education=education-university.degree
                                                      3.303003
## f.season=season-winter
                                                      3.145618
## month=month-dec
                                                      3.145618
## f.campaign=f.campaign-[0,2]
                                                      2.932757
## f.cons.price.idx=f.cons.price.idx-(94,94.8]
                                                      2.924889
## education=education-unknown
                                                      2.857442
## poutcome=poutcome-failure
                                                      2.328885
                                                     -2.298498
## f.age=f.age-(32,38]
## f.season=season-spring
                                                     -2.417454
## education=education-basic.6y
                                                     -2.474129
## month=month-jul
                                                     -2.544655
## marital=marital-married
                                                     -3.046536
## education=education-basic.9y
                                                     -3.735055
## f.campaign=f.campaign-(5,25]
                                                     -3.870893
## f.age=f.age-(38,47]
                                                     -4.909404
## f.cons.price.idx=f.cons.price.idx-(93.1,93.7]
                                                     -5.038105
## f.duration=f.duration-(138,177]
                                                     -5.439509
## f.euribor3m=f.euribor3m-(1.33,4.86]
                                                     -5.795870
## job=job-blue-collar
                                                     -6.370444
## f.duration=f.duration-(101,138]
                                                     -6.804960
## f.euribor3m=f.euribor3m-(4.96,5]
                                                     -7.186654
## f.euribor3m=f.euribor3m-(4.86,4.96]
                                                     -7.370998
## month=month-may
                                                     -7.669524
## default=default-unknown
                                                     -7.712857
## f.cons.conf.idx=f.cons.conf.idx-(-42.7,-41.8]
                                                     -8.797336
## f.emp.var.rate=f.emp.var.rate-(-0.1,1.1]
                                                     -9.039450
## f.nr.employed=f.nr.employed-(5.1e+03,5.19e+03]
                                                     -9.050417
## f.cons.price.idx=f.cons.price.idx-(93.7,94]
                                                     -9.372891
## f.duration=f.duration-(66,101]
                                                     -9.603908
## contact=contact-telephone
                                                     -9.918824
## f.emp.var.rate=f.emp.var.rate-(1.1,1.4]
                                                    -10.458406
## f.duration=f.duration-[5,66]
                                                    -11.489650
## f.previous=f.previous-never
                                                    -11.492513
## poutcome=poutcome-nonexistent
                                                    -11.492513
## f.nr.employed=f.nr.employed-(5.19e+03,5.23e+03] -12.778626
## f.pdays=f.pdays-never
                                                    -16.245323
##
##
```

```
## Link between the cluster variable and the quantitative variables
##
                                  P-value
                       Eta2
                 0.164777620 3.759496e-197
## duration
## minutes
                 0.164777620 3.759496e-197
## nr.employed
                 0.121012601 8.238443e-142
## pdays
                 0.090100788 2.433135e-104
## euribor3m
                 0.090010720 3.115343e-104
## emp.var.rate
                 0.085417483 8.992557e-99
## previous
                 0.042523921 5.101307e-49
## cons.price.idx 0.018386453 6.794885e-22
## cons.conf.idx 0.004669195 1.369222e-06
## campaign
                 0.004489049 2.189052e-06
## <NA>
                         NΑ
##
## Description of each cluster by quantitative variables
## $`v-no`
##
                    v.test Mean in category Overall mean sd in category
## nr.employed
                  24.561104
                               5175.3298261 5166.47621340
                                                             64.3842715
## pdays
                  21.193217
                                983.3030029 963.73706378
                                                            123.8692868
## euribor3m
                  21.182621
                                  3.7992890
                                               3.61448034
                                                              1.6425449
## emp.var.rate
                  20.635071
                                  0.2287424
                                                              1.4946001
                                               0.06446049
## cons.price.idx
                                 93.6004884
                   9.573739
                                              93.57245006
                                                              0.5619158
## campaign
                   4.730529
                                  2.5940750
                                              2.53512998
                                                              2.5654605
## cons.conf.idx
                  -4.824514
                                -40.5398961 -40.42591256
                                                              4.4454152
## previous
                 -14.559593
                                  0.1255362
                                              0.15984757
                                                              0.4004406
## duration
                 -28.660364
                                217.4563107
                                             250.62194144
                                                            191.6321071
## minutes
                 -28.660364
                                  3.6242718
                                               4.17703236
                                                              3.1938685
                                  p.value
##
                  Overall sd
## nr.employed
                  71.7679377 3.291367e-133
## pdays
                 183.8068310 1.102990e-99
## euribor3m
                   1.7370025 1.381286e-99
                   1.5850448 1.329502e-94
## emp.var.rate
## cons.price.idx
                   0.5830800 1.031083e-21
                   2.4808187 2.239350e-06
## campaign
## cons.conf.idx
                   4.7037753 1.403451e-06
## previous
                   0.4691873 5.075919e-48
## duration
                 230.3904064 1.190744e-180
                   3.8398401 1.190744e-180
## minutes
##
## $`y-yes`
                    v.test Mean in category
                                             Overall mean sd in category
## minutes
                  28.660364
                                   8.572322
                                               4.17703236
                                                              5.3967235
                                                            323.8034093
## duration
                  28.660364
                                 514.339318
                                             250.62194144
## previous
                  14.559593
                                   0.432675
                                               0.15984757
                                                              0.7821222
## cons.conf.idx
                  4.824514
                                 -39.519569
                                             -40.42591256
                                                              6.3242738
## campaign
                  -4.730529
                                   2.066427
                                               2.53512998
                                                              1.5845655
## cons.price.idx -9.573739
                                  93.349503
                                             93.57245006
                                                              0.6904449
## emp.var.rate
                 -20.635071
                                  -1.241831
                                               0.06446049
                                                              1.6751620
## euribor3m
                 -21.182621
                                   2.144969
                                               3.61448034
                                                              1.7676126
## pdays
                 -21.193217
                                 808.157989 963.73706378
                                                            391.3731388
                 -24.561104
## nr.employed
                                5096.076481 5166.47621340
                                                             86.9764988
##
                  Overall sd
                                  p.value
```

```
## minutes
                 3.8398401 1.190744e-180
## duration
               230.3904064 1.190744e-180
## previous
                0.4691873 5.075919e-48
## cons.conf.idx 4.7037753 1.403451e-06
## campaign
                  2.4808187 2.239350e-06
## cons.price.idx 0.5830800 1.031083e-21
## emp.var.rate 1.5850448 1.329502e-94
## euribor3m
                 1.7370025 1.381286e-99
## pdays
                183.8068310 1.102990e-99
## nr.employed
                71.7679377 3.291367e-133
# $`y-yes`
                                                Mod/Cla Global
                                                                     p.value
                                        Cla/Mod
                                                                               v.test
# f.duration=f.duration-(483,1.58e+03] 40.8064516 44.7787611 12.40 2.180784e-97 20.942837
# poutcome=poutcome-success 62.2641509 17.5221239 3.18 5.331532e-56 15.766007
# f.pdays=f.pdays-[0,6]
                                    62.222222 14.8672566 2.70 2.653287e-47 14.446089
# contact=contact-cellular
                                    14.5686901 80.7079646 62.60 6.688527e-23 9.852462
# Dins el cluster que s'ha acceptat el producte financer, la "durada (483 a 1580]" és el 44,778% dels va
```