

# Preprocessing

FERRAN ARMENGOL Y SALOMÓN SABAL

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Preprocessing

```
#install.packages("dplyr")
library(dplyr)

train <- read.csv("train.csv", header = TRUE, sep = ",")
test <- read.csv("test.csv", header = TRUE, sep = ",")
datos <- bind_rows(train, test)

dim(datos)

## [1] 10000    24

head(datos)

##   X Tenure Gender EducationLevel LoanStatus NetPromoterScore TransactionFrequency Surname Age Ge
## 1 1      NA Female University       <NA>          10                 34      <NA> 29
## 2 2      1   <NA> University     No loan           9                  31 Chiagoziem 41
## 3 3      8   Male University     No loan           8                  26      <NA> 43
## 4 4      NA Female High School Active loan        NA                 32 Mackay 48
## 5 5      3   Male High School Active loan           6                 41      <NA> 34
## 6 6      NA   Male University       <NA>          9                  33 O'Loughlin NA
##   ComplaintsCount HasCrCard EstimatedSalary IsActiveMember AvgTransactionAmount CustomerSegment Mari
## 1              0       NA             NA            0                   NA      Mass Market
## 2              0       NA             NA            0                   NA      Mass Market
## 3              0       1            135650.72         0                228.83782 Mass Market
## 4              0       0            102640.52        NA               133.89406 Affluent
## 5             NA       1            83773.02         1                91.57003      <NA>
## 6             NA      NA            116706.00         0                87.96360      <NA>
##   DigitalEngagementScore ID CreditScore SavingsAccountFlag Balance NumOfProducts Exited
## 1                      60 6222          832            1       NA            2            0
## 2                     NA 3217          NA            1       NA            NA            0
## 3                     NA  NA          577            1    79757.21            1            1
## 4                     NA  NA          482            0       NA            NA            1
## 5                     67 5511          635            0       NA            NA            0
## 6                     43 3575          656            NA      0.00            2            0

summary(datos)
```

```

##      X          Tenure        Gender       EducationLevel     LoanStatus   NetPromote
## Min. : 1    Min. : 0.000  Length:10000  Length:10000  Length:10000  Min. : 0
## 1st Qu.:1751 1st Qu.: 3.000  Class :character  Class :character  Class :character  1st Qu.: 4
## Median :3500 Median : 5.000  Mode  :character  Mode  :character  Mode  :character  Median : 8
## Mean   :3500  Mean   : 5.005
## 3rd Qu.:5250 3rd Qu.: 7.000
## Max.   :7000  Max.   :10.000
## NA's   :3000  NA's   :3000
## TransactionFrequency Surname        Age        Geography   ComplaintsCount HasCrCard
## Min. :13.00      Length:10000  Min. :18.00  Length:10000  Min. :0.0000  Min.
## 1st Qu.:26.00      Class :character 1st Qu.:32.00  Class :character 1st Qu.:0.0000  1st Qu.
## Median :30.00      Mode  :character Median :37.00  Mode  :character Median :0.0000  Median
## Mean   :30.06
## 3rd Qu.:34.00
## Max.   :58.00
## NA's   :3000
## EstimatedSalary IsActiveMember AvgTransactionAmount CustomerSegment MaritalStatus  Digit
## Min. : 11.58  Min. :0.0000  Min. : 19.60  Length:10000  Length:10000  Min.
## 1st Qu.: 50755.81 1st Qu.:0.0000  1st Qu.: 70.12  Class :character  Class :character  1st
## Median : 99796.85  Median :1.0000  Median : 98.61  Mode  :character  Mode  :character  Med
## Mean   : 99912.26  Mean   :0.5167  Mean   :111.76
## 3rd Qu.:148823.10 3rd Qu.:1.0000  3rd Qu.:138.02
## Max.   :199992.48  Max.   :1.0000  Max.   :611.35
## NA's   :3000
## ID      CreditScore SavingsAccountFlag Balance  NumOfProducts Exited
## Min. : 1    Min. :350.0  Min. :0.0000  Min. : 0  Min. :1.000  Min. :0.0000
## 1st Qu.:1167 1st Qu.:584.0  1st Qu.:0.0000  1st Qu.: 0  1st Qu.:1.000  1st Qu.:0.0000
## Median :2334  Median :652.0  Median :1.0000  Median : 96951  Median :1.000  Median :0.0000
## Mean   :2750  Mean   :650.7  Mean   :0.6573  Mean   : 76267  Mean   :1.536  Mean   :0.2071
## 3rd Qu.:4210 3rd Qu.:718.0  3rd Qu.:1.0000  3rd Qu.:127686 3rd Qu.:2.000  3rd Qu.:0.0000
## Max.   :6999  Max.   :850.0  Max.   :1.0000  Max.   :250898  Max.   :4.000  Max.   :1.0000
## NA's   :2100  NA's   :3000  NA's   :3000  NA's   :3000  NA's   :3000  NA's   :3000

clases <- sapply(datos, class)
datos$HasCrCard <- as.factor(datos$HasCrCard)
datos$IsActiveMember <- as.factor(datos$IsActiveMember)
datos$SavingsAccountFlag <- as.factor(datos$SavingsAccountFlag)

varNum <- names(datos)[sapply(datos, is.numeric)]
varCat <- names(datos)[!sapply(datos, is.numeric)]

#install.packages("dlookr")
library(dlookr)
dlookr::diagnose(datos)

## # A tibble: 24 x 6
##   variables      types missing_count missing_percent unique_count unique_rate
##   <chr>        <chr>        <int>            <dbl>        <int>        <dbl>
## 1 X             integer       3000             0.700      7001        0.700
## 2 Tenure        integer       3000             0.700       12        0.0012
## 3 Gender        character      3000             0.700        3        0.0003
## 4 EducationLevel character      3000             0.700        5        0.0005

```

```

## 5 LoanStatus      character      3000      30      4    0.0004
## 6 NetPromoterScore integer      3000      30     12    0.0012
## 7 TransactionFrequency integer      3000      30     42    0.0042
## 8 Surname        character      3000      30   2427    0.243
## 9 Age            integer      3000      30     70    0.007
## 10 Geography     character      3000      30      4    0.0004
## # i 14 more rows

```

```
head(overview(datos), n = 9)
```

```

##      division          metrics  value
## 1      size    observations 10000
## 2      size       variables    24
## 3      size        values 240000
## 4      size   memory size 1509120
## 5 duplicated duplicate observation    0
## 6 missing   complete observation    0
## 7 missing   missing observation 10000
## 8 missing   missing variables    24
## 9 missing   missing values 71100

```

```
diagnose_numeric(datos)
```

	variables	min	Q1	mean	median	Q3	max	zero
## 1	X	1.00000	1750.75000	3.500500e+03	3500.50000	5250.2500	7000.0000	0
## 2	Tenure	0.00000	3.00000	5.005286e+00	5.00000	7.0000	10.0000	293
## 3	NetPromoterScore	0.00000	4.00000	6.440286e+00	8.00000	9.0000	10.0000	423
## 4	TransactionFrequency	13.00000	26.00000	3.005900e+01	30.00000	34.0000	58.0000	0
## 5	Age	18.00000	32.00000	3.900557e+01	37.00000	44.0000	92.0000	0
## 6	ComplaintsCount	0.00000	0.00000	3.695714e-01	0.00000	0.0000	5.0000	5617
## 7	EstimatedSalary	11.58000	50755.80750	9.991226e+04	99796.84500	148823.0950	199992.4800	0
## 8	AvgTransactionAmount	19.60371	70.11772	1.117624e+02	98.61161	138.0194	611.3528	0
## 9	DigitalEngagementScore	5.00000	50.00000	5.955029e+01	60.00000	70.0000	100.0000	0
## 10	ID	1.00000	1167.00000	2.749697e+03	2334.00000	4210.2500	6999.0000	0
## 11	CreditScore	350.00000	584.00000	6.506946e+02	652.00000	718.0000	850.0000	0
## 12	Balance	0.00000	0.00000	7.626726e+04	96950.71000	127685.6825	250898.0900	254
## 13	NumOfProducts	1.00000	1.00000	1.535714e+00	1.00000	2.0000	4.0000	0
## 14	Exited	0.00000	0.00000	2.071429e-01	0.00000	0.0000	1.0000	555

```
diagnose_category(datos)
```

	variables	levels	N	freq	ratio	rank
## 1	Gender	Male	10000	3813	38.13	1
## 2	Gender	Female	10000	3187	31.87	2
## 3	Gender	<NA>	10000	3000	30.00	3
## 4	EducationLevel	University	10000	3206	32.06	1
## 5	EducationLevel	<NA>	10000	3000	30.00	2
## 6	EducationLevel	High School	10000	2400	24.00	3
## 7	EducationLevel	Postgraduate	10000	1064	10.64	4
## 8	EducationLevel	Other	10000	330	3.30	5
## 9	LoanStatus	No loan	10000	4216	42.16	1
## 10	LoanStatus	<NA>	10000	3000	30.00	2

```

## 11      LoanStatus    Active loan 10000 2065 20.65    3
## 12      LoanStatus    Default risk 10000 719  7.19    4
## 13      Surname      <NA> 10000 3000 30.00    1
## 14      Surname      Walker 10000 25   0.25    2
## 15      Surname      Smith 10000 24   0.24    3
## 16      Surname      Martin 10000 22   0.22    4
## 17      Surname      Scott 10000 21   0.21    5
## 18      Surname      Yeh   10000 21   0.21    5
## 19      Surname      Maclean 10000 20   0.20    7
## 20      Surname      Brown 10000 19   0.19    8
## 21      Surname      Shih   10000 18   0.18    9
## 22      Surname      Sun    10000 18   0.18    9
## 23      Geography    France 10000 3522 35.22    1
## 24      Geography    <NA> 10000 3000 30.00    2
## 25      Geography    Germany 10000 1755 17.55    3
## 26      Geography    Spain  10000 1723 17.23    4
## 27      HasCrCard    1 10000 4918 49.18    1
## 28      HasCrCard    <NA> 10000 3000 30.00    2
## 29      HasCrCard    0 10000 2082 20.82    3
## 30      IsActiveMember 1 10000 3617 36.17    1
## 31      IsActiveMember 0 10000 3383 33.83    2
## 32      IsActiveMember <NA> 10000 3000 30.00    3
## 33      CustomerSegment Mass Market 10000 3566 35.66    1
## 34      CustomerSegment <NA> 10000 3000 30.00    2
## 35      CustomerSegment Affluent 10000 2069 20.69    3
## 36      CustomerSegment High Net Worth 10000 1365 13.65    4
## 37      MaritalStatus Married 10000 3549 35.49    1
## 38      MaritalStatus <NA> 10000 3000 30.00    2
## 39      MaritalStatus Single 10000 2074 20.74    3
## 40      MaritalStatus Divorced 10000 1029 10.29    4
## 41      MaritalStatus Widowed 10000 348  3.48    5
## 42      SavingsAccountFlag 1 10000 4601 46.01    1
## 43      SavingsAccountFlag <NA> 10000 3000 30.00    2
## 44      SavingsAccountFlag 0 10000 2399 23.99    3

```

```

mapply(function(x, name) {
  if (is.numeric(x)) {
    cat("var. ", name, ":\n\t min: ", min(x, na.rm = TRUE), "\n\t max: ", max(x, na.rm = TRUE), "\n")
  }
  invisible(NULL)
}, datos[, varNum], colnames(datos[, varNum]))


## var. X :
##   min: 1
##   max: 7000
## var. Tenure :
##   min: 0
##   max: 10
## var. NetPromoterScore :
##   min: 0
##   max: 10
## var. TransactionFrequency :
##   min: 13
##   max: 58

```

```

## var. Age :
##   min: 18
##   max: 92
## var. ComplaintsCount :
##   min: 0
##   max: 5
## var. EstimatedSalary :
##   min: 11.58
##   max: 199992.5
## var. AvgTransactionAmount :
##   min: 19.60371
##   max: 611.3528
## var. DigitalEngagementScore :
##   min: 5
##   max: 100
## var. ID :
##   min: 1
##   max: 6999
## var. CreditScore :
##   min: 350
##   max: 850
## var. Balance :
##   min: 0
##   max: 250898.1
## var. NumOfProducts :
##   min: 1
##   max: 4
## var.Exited :
##   min: 0
##   max: 1

## $X
## NULL
##
## $Tenure
## NULL
##
## $NetPromoterScore
## NULL
##
## $TransactionFrequency
## NULL
##
## $Age
## NULL
##
## $ComplaintsCount
## NULL
##
## $EstimatedSalary
## NULL
##
## $AvgTransactionAmount
## NULL

```

```

## 
## $DigitalEngagementScore
## NULL
##
## $ID
## NULL
##
## $CreditScore
## NULL
##
## $Balance
## NULL
##
## $NumOfProducts
## NULL
##
## $Exited
## NULL

library(EnvStats)

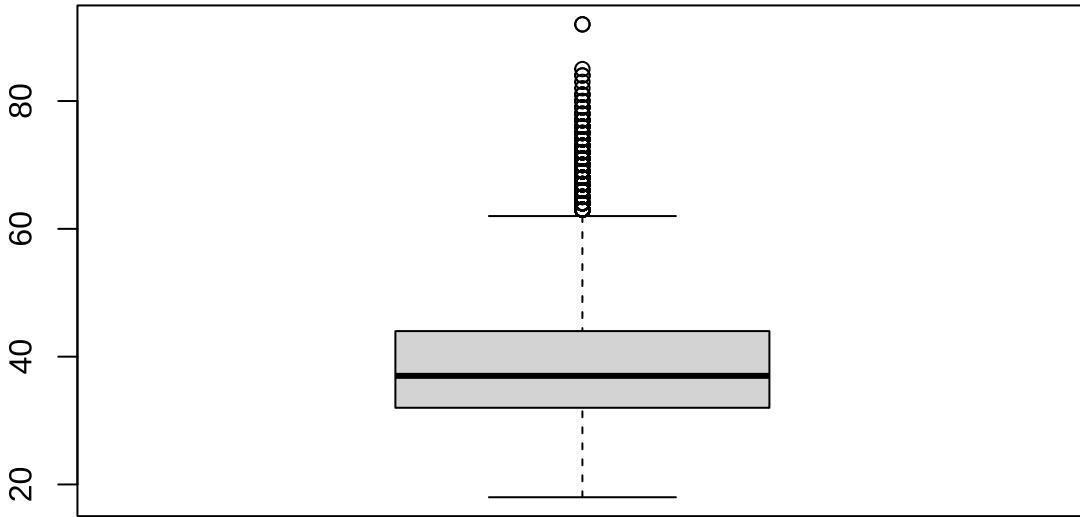
IQROutlier <- function(variable, rmnas = TRUE) {
  variable <- variable[!is.na(variable)] # eliminar NAs
  IQ <- iqr(variable, na.rm = rmnas)
  intInf <- quantile(variable, probs = 0.25, na.rm = TRUE) - 1.5 * IQ
  intSup <- quantile(variable, probs = 0.75, na.rm = TRUE) + 1.5 * IQ
  posicions <- which(variable >= intSup | variable <= intInf)
  if (length(posicions) > 0) {
    cat("Existeixen outliers en les posicions:", paste0(posicions, collapse = ", "))
  } else {
    cat("No existeixen outliers")
  }
  return(posicions)
}

outliers_todos <- lapply(varNum, function(v) {
  cat("\n\nVariable:", v, "\n")
  IQROutlier(datos[[v]])
})

## 
## Variable: X
## No existeixen outliers
##
## Variable: Tenure
## No existeixen outliers
##
## Variable: NetPromoterScore
## No existeixen outliers
##
## Variable: TransactionFrequency
## Existeixen outliers en les posicions: 110, 183, 359, 643, 1099, 1464, 1468, 1511, 1584, 1827, 2060, ...
##

```





```

boxplot.stats(datos[, variable])$out

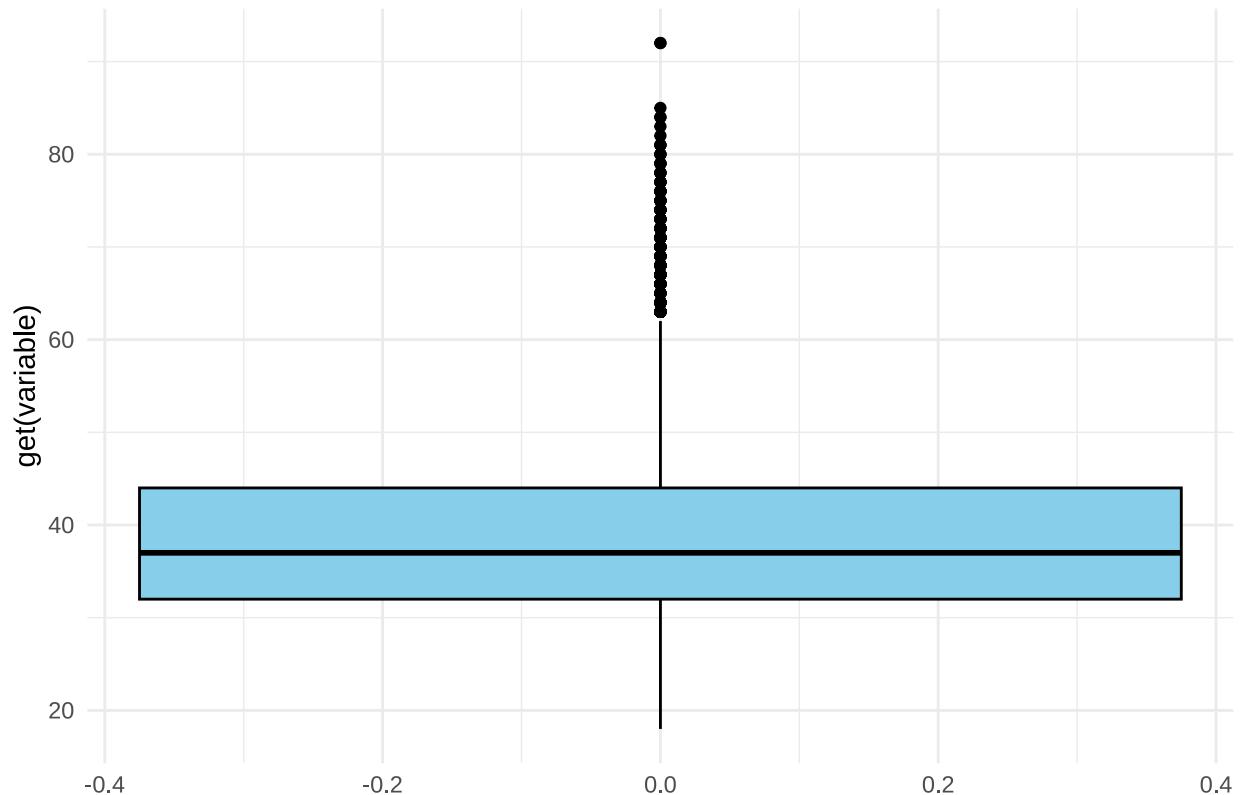
## [1] 74 67 68 64 67 63 77 72 66 75 66 64 70 67 63 68 66 75 69 78 77 67 76 72 71 76 69 67 63 63 85 6
## [39] 63 66 63 64 66 65 71 76 73 92 67 67 75 71 64 79 67 75 64 80 66 66 70 63 68 82 63 67 73 66 72 6
## [77] 68 63 72 68 66 70 63 64 63 76 69 79 84 66 63 78 69 66 72 63 76 63 75 66 67 67 68 63 74 64 68 7
## [115] 66 73 64 66 68 66 76 69 63 76 73 67 63 66 80 73 63 72 67 67 66 63 64 74 67 66 70 70 67 66 70 6
## [153] 77 77 74 63 69 73 73 63 71 67 64 71 63 92 74 70 69 67 81 63 67 66 69 68 70 74 75 70 65 74 63 6
## [191] 69 70 74 63 72 84 67 65 71 66 68 67 71 64 80 64 77 63 69 73 66 63 64 71 65 83 64 66 67 63 64 7
## [229] 69 66 73 71 67 77 71 65 68 73 63 63 70 64 70 68 64 71 78 71 68 69 67

# Crear un boxplot
ggplot(datos, aes(y = get(variable))) +
  geom_boxplot(fill = "skyblue", color = "black") +
  labs(title = paste0("Boxplot de ", variable)) +
  theme_minimal()

## Warning: Removed 3000 rows containing non-finite outside the scale range ('stat_boxplot()').

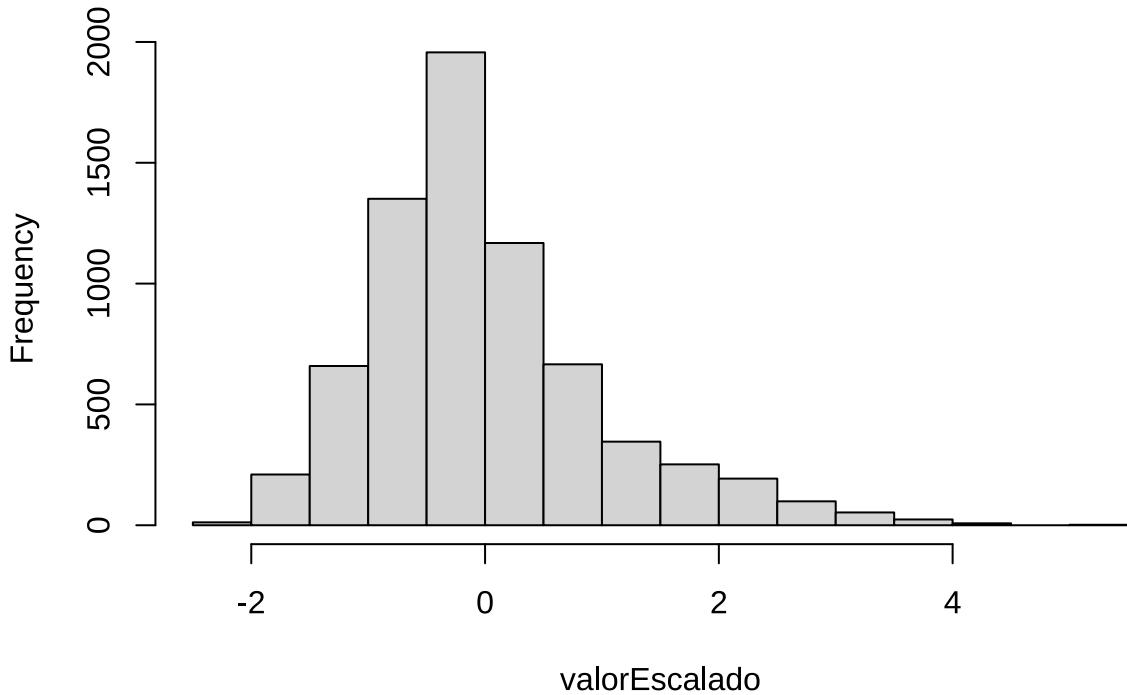
```

## Boxplot de Age



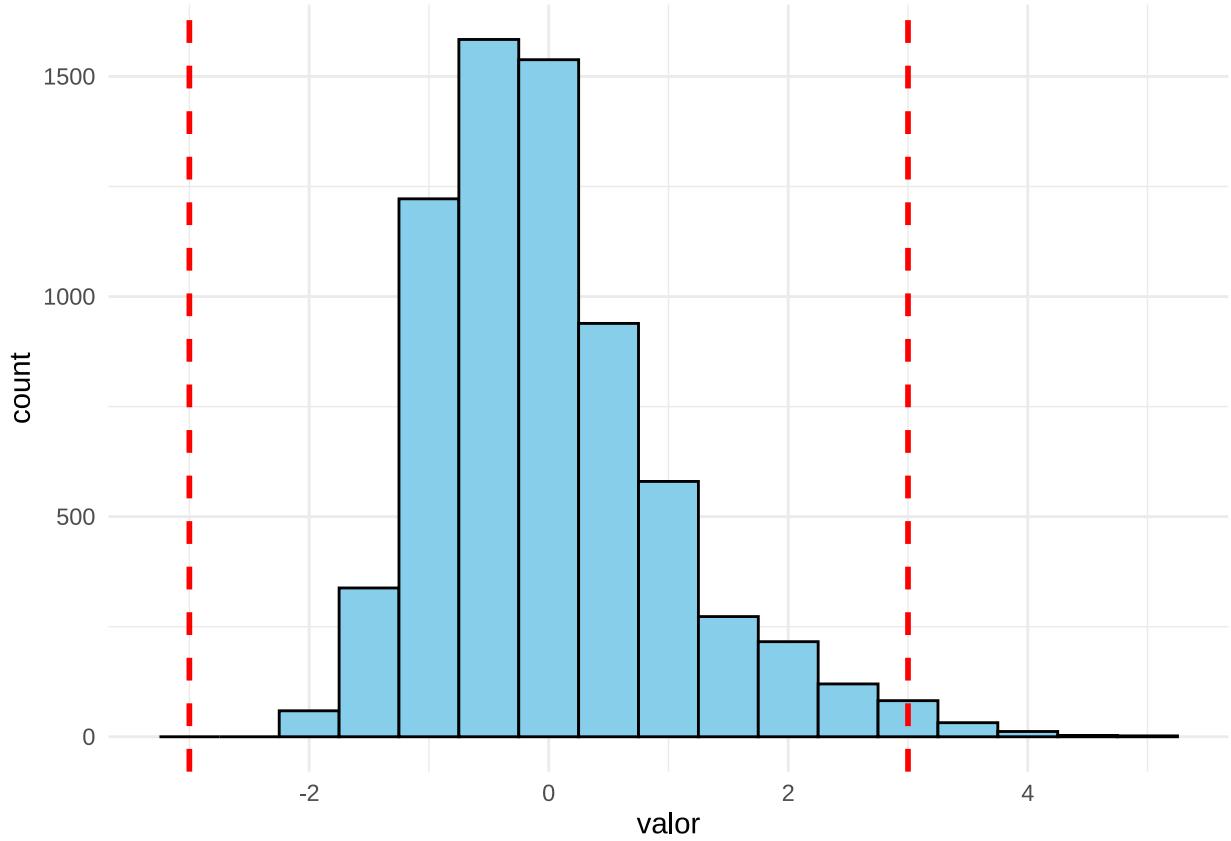
```
variable <- "Age"  
valorEscalado <- scale(datos[, variable])  
hist(valorEscalado)
```

## Histogram of valorEscalado



```
ggplot(data.frame(valor = valorEscalado), aes(x = valor)) +  
  geom_histogram(binwidth = 0.5, fill = "skyblue", color = "black") + # Histograma  
  geom_vline(xintercept = c(3, -3), linetype = "dashed", color = "red", size = 1) + # Líneas horizontales  
  theme_minimal()
```

```
## Warning: Removed 3000 rows containing non-finite outside the scale range ('stat_bin()').
```



```
#library(dplyr)

train <- read.csv("train.csv", header = TRUE, sep = ",")
test <- read.csv("test.csv", header = TRUE, sep = ",")
datos <- bind_rows(train, test)

varNum <- names(datos)[sapply(datos, is.numeric)]
datos_num <- datos[, varNum]
summary(datos_num)
```

##	X	Tenure	NetPromoterScore	TransactionFrequency	Age	ComplaintsCount
## Min. :	1	Min. : 0.000	Min. : 0.00	Min. :13.00	Min. :18.00	Min. :0.0000
## 1st Qu.:	1751	1st Qu.: 3.000	1st Qu.: 4.00	1st Qu.:26.00	1st Qu.:32.00	1st Qu.:0.0000
## Median :	3500	Median : 5.000	Median : 8.00	Median :30.00	Median :37.00	Median :0.0000
## Mean :	3500	Mean : 5.005	Mean : 6.44	Mean :30.06	Mean :39.01	Mean :0.3690
## 3rd Qu.:	5250	3rd Qu.: 7.000	3rd Qu.: 9.00	3rd Qu.:34.00	3rd Qu.:44.00	3rd Qu.:0.0000
## Max. :	7000	Max. :10.000	Max. :10.00	Max. :58.00	Max. :92.00	Max. :5.0000
## NA's :	3000	NA's :3000	NA's :3000	NA's :3000	NA's :3000	NA's :3000
## EstimatedSalary		IsActiveMember	AvgTransactionAmount	DigitalEngagementScore	ID	Cr
## Min. :	11.58	Min. :0.0000	Min. : 19.60	Min. : 5.00	Min. : 1	Min
## 1st Qu.:	50755.81	1st Qu.:0.0000	1st Qu.: 70.12	1st Qu.: 50.00	1st Qu.:1167	1st
## Median :	99796.85	Median :1.0000	Median : 98.61	Median : 60.00	Median :2334	Med
## Mean :	99912.26	Mean :0.5167	Mean :111.76	Mean : 59.55	Mean :2750	Mean
## 3rd Qu.:	148823.10	3rd Qu.:1.0000	3rd Qu.:138.02	3rd Qu.: 70.00	3rd Qu.:4210	3rd
## Max. :	199992.48	Max. :1.0000	Max. :611.35	Max. :100.00	Max. :6999	Max

```

##  NA's :3000      NA's :3000      NA's :3000      NA's :3000      NA's :2100  NA's :
## SavingsAccountFlag   Balance    NumOfProducts   Exited
##  Min. :0.0000     Min. :    0     Min. :1.000     Min. :0.0000
##  1st Qu.:0.0000    1st Qu.:    0     1st Qu.:1.000    1st Qu.:0.0000
##  Median :1.0000    Median : 96951    Median :1.000    Median :0.0000
##  Mean   :0.6573    Mean   : 76267    Mean   :1.536    Mean   :0.2071
##  3rd Qu.:1.0000    3rd Qu.:127686   3rd Qu.:2.000    3rd Qu.:0.0000
##  Max.  :1.0000    Max.  :250898    Max.  :4.000    Max.  :1.0000
##  NA's  :3000      NA's  :3000      NA's  :3000      NA's  :3000

center <- colMeans(datos_num, na.rm = TRUE)
cov_matrix <- cov(datos_num, use = "complete.obs")

dist_mahal <- mahalanobis(datos_num, center, cov_matrix)

umbral <- qchisq(0.999, df = ncol(datos_num))

outliers_mahal <- which(dist_mahal > umbral)
cat("Número de outliers multivariantes detectados:", length(outliers_mahal), "\n")

## Número de outliers multivariantes detectados: 0

summary(dist_mahal)

##      Min. 1st Qu.  Median  Mean 3rd Qu.  Max.  NA's
##  10.82   15.31   17.52  17.61  20.08  26.31  9970

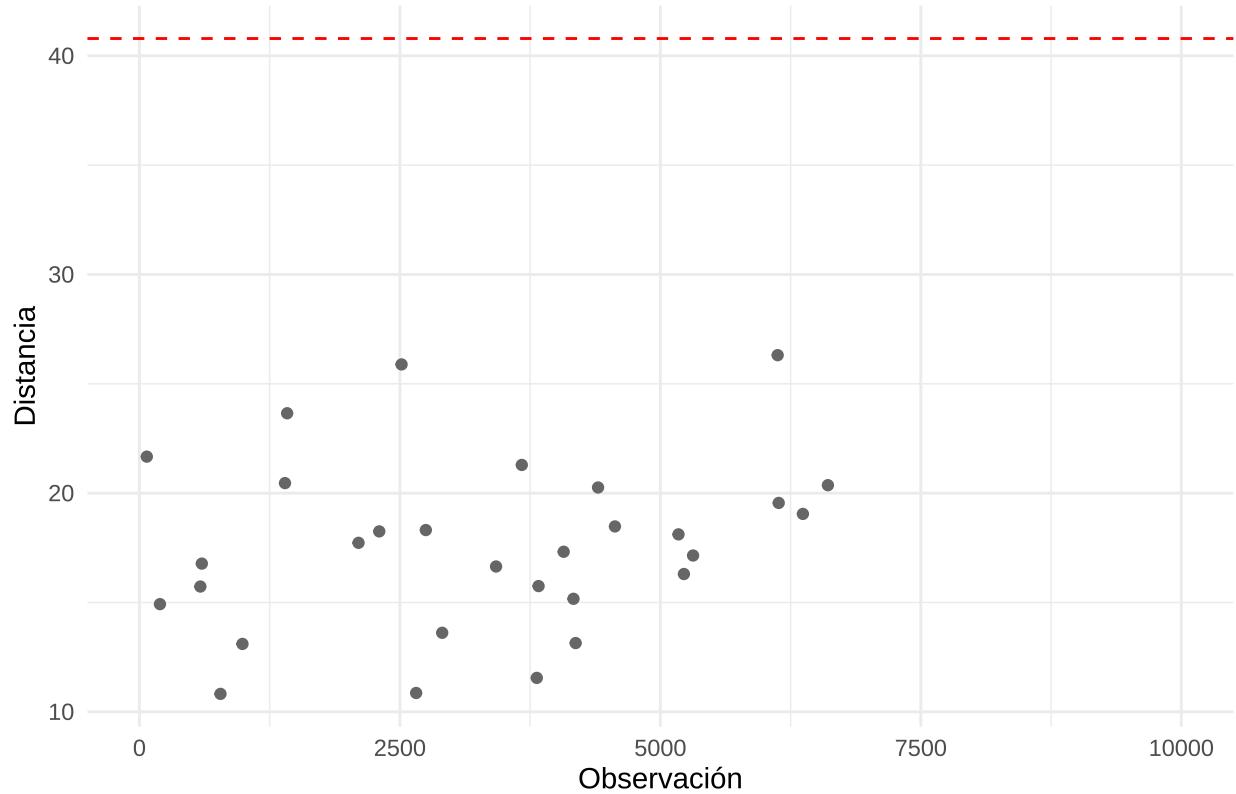
#library(ggplot2)

ggplot(data.frame(dist_mahal), aes(x = 1:length(dist_mahal), y = dist_mahal)) +
  geom_point(color = "grey40") +
  geom_hline(yintercept = umbral, color = "red", linetype = "dashed") +
  labs(title = "Distancia de Mahalanobis", x = "Observación", y = "Distancia") +
  theme_minimal()

## Warning: Removed 9970 rows containing missing values or values outside the scale range ('geom_point()')

```

## Distancia de Mahalanobis



```

library(FactoMineR)
library(factoextra)
library(ggplot2)

# PCA
pca <- PCA(datos_num, scale.unit = TRUE, graph = FALSE)

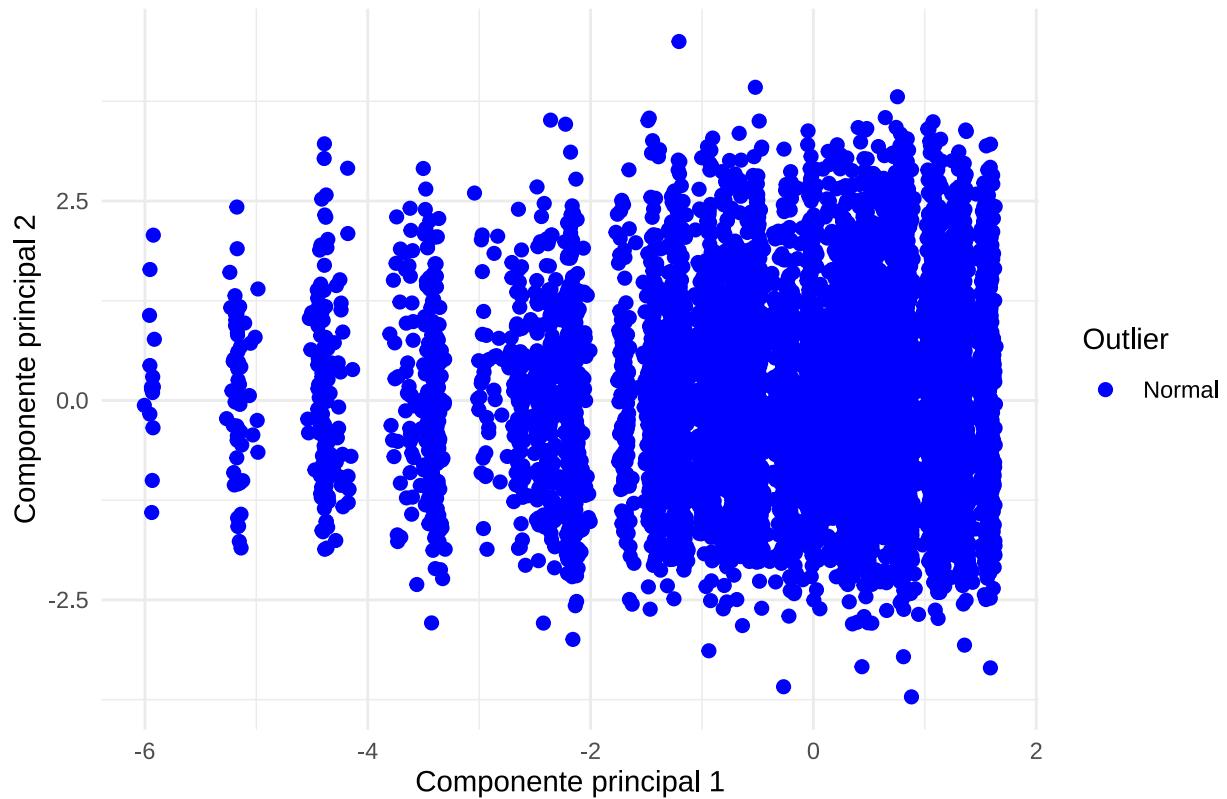
## Warning in PCA(datos_num, scale.unit = TRUE, graph = FALSE): Missing values are imputed by the mean
## should use the imputePCA function of the missMDA package

# Crear dataframe con coordenadas
coords <- as.data.frame(pca$ind$coord)
coords$Outlier <- factor(ifelse(1:nrow(datos_num) %in% outliers_mahal, "Outlier", "Normal"))

# Gráfico manual con ggplot2 (sin errores de color)
ggplot(coords, aes(x = Dim.1, y = Dim.2, color = Outlier)) +
  geom_point(size = 2) +
  scale_color_manual(values = c("blue", "red")) +
  labs(title = "PCA - Outliers multivariantes (rojo = outliers)",
       x = "Componente principal 1",
       y = "Componente principal 2") +
  theme_minimal()

```

PCA - Outliers multivariantes (rojo = outliers)



```
datos$Outlier_Mahalanobis <- FALSE  
datos$Outlier_Mahalanobis[outliers_mahal] <- TRUE  
  
write.csv(datos, "datos_outliers_multivariantes.csv", row.names = FALSE)
```