

Trabajo3

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```
#Librerías utilizadas.  
library("caret")
```

```
## Loading required package: lattice
```

```
## Loading required package: ggplot2
```

```
library("leaps")  
library("e1071")
```

1. default of credit card clients Data Set (Clasificaci?n)

La base de datos se centra en el caso de los pagos por defecto de los clientes en Taiwán y compara la precisión predictiva de la probabilidad de incumplimiento entre seis métodos de minería de datos. Desde la perspectiva de la gestión de riesgos, el resultado de la precisión predictiva de la probabilidad estimada de incumplimiento ser más valioso que el resultado binario de la clasificación - clientes creíbles o no creíbles. Con la probabilidad real de default como variable de respuesta (Y), y la probabilidad predictiva de default como variable independiente (X), el resultado de regresión lineal simple ($Y = A + BX$) muestra que el modelo de predicción producido por la red neuronal artificial tiene el mayor coeficiente de determinación. Su intercepción de regresión (A) es cercana a cero, y el coeficiente de regresión (B) a uno. Por lo tanto, entre las seis técnicas de minería de datos, la red neuronal artificial es la única que puede estimar con precisión la probabilidad real de incumplimiento. Dichas características son:

SBP: Presión arterial Tobacco: El tabaco acululado en kg. LDL: Lipoproteina de baja densidad. Adiposity: Acumulacion excesiva de grasa en el organismo. Famhist: Historia familiar sobre casos de enfermedades cardiacas. Typea: Personalidad tipo A. Obesity: Obesidad. Alcohol Consumo de alcohol actual. Age: Edad. Chd: Respuesta a la enfermedad coronaria (la variable que queremos aprender).

```
credit_card = read.csv("default_of_credict_card_clients.csv"  
                      , sep="," , header = TRUE, row.names =1)
```

```
summary(credit_card)
```

##	LIMIT_BAL	SEX	EDUCATION	MARRIAGE
##	Min. : 10000	Min. :1.000	Min. :0.000	Min. :0.000
##	1st Qu.: 50000	1st Qu.:1.000	1st Qu.:1.000	1st Qu.:1.000
##	Median : 140000	Median :2.000	Median :2.000	Median :2.000
##	Mean : 167484	Mean :1.604	Mean :1.853	Mean :1.552
##	3rd Qu.: 240000	3rd Qu.:2.000	3rd Qu.:2.000	3rd Qu.:2.000
##	Max. :1000000	Max. :2.000	Max. :6.000	Max. :3.000
##	AGE	PAY_0	PAY_2	PAY_3
##	Min. :21.00	Min. : -2.0000	Min. : -2.0000	Min. : -2.0000
##	1st Qu.:28.00	1st Qu.: -1.0000	1st Qu.: -1.0000	1st Qu.: -1.0000
##	Median :34.00	Median : 0.0000	Median : 0.0000	Median : 0.0000
##	Mean :35.49	Mean : -0.0167	Mean : -0.1338	Mean : -0.1662
##	3rd Qu.:41.00	3rd Qu.: 0.0000	3rd Qu.: 0.0000	3rd Qu.: 0.0000
##	Max. :79.00	Max. : 8.0000	Max. : 8.0000	Max. : 8.0000
##	PAY_4	PAY_5	PAY_6	BILL_AMT1

```
## Min.      :-2.0000    Min.      :-2.0000    Min.      :-2.0000    Min.      :-165580
## 1st Qu.: -1.0000    1st Qu.: -1.0000    1st Qu.: -1.0000    1st Qu.:   3559
## Median :  0.0000    Median :  0.0000    Median :  0.0000    Median :  22382
## Mean    :-0.2207    Mean     :-0.2662    Mean     :-0.2911    Mean     :  51223
## 3rd Qu.:  0.0000    3rd Qu.:  0.0000    3rd Qu.:  0.0000    3rd Qu.:  67091
## Max.     :  8.0000    Max.      :  8.0000    Max.      :  8.0000    Max.      : 964511
## BILL_AMT2    BILL_AMT3    BILL_AMT4    BILL_AMT5
## Min.      :-69777    Min.      :-157264    Min.      :-170000    Min.      :-81334
## 1st Qu.:   2985    1st Qu.:   2666    1st Qu.:   2327    1st Qu.:   1763
## Median :   21200    Median :   20089    Median :   19052    Median :   18105
## Mean     :   49179    Mean      :   47013    Mean      :   43263    Mean      :   40311
## 3rd Qu.:  64006    3rd Qu.:  60165    3rd Qu.:  54506    3rd Qu.:  50191
## Max.      : 983931    Max.      :1664089    Max.      : 891586    Max.      :927171
## BILL_AMT6    PAY_AMT1    PAY_AMT2    PAY_AMT3
## Min.      :-339603    Min.       :      0    Min.       :      0    Min.       :      0
## 1st Qu.:   1256    1st Qu.:   1000    1st Qu.:    833    1st Qu.:    390
## Median :   17071    Median :    2100    Median :    2009    Median :    1800
## Mean     :   38872    Mean      :   5664    Mean      :   5921    Mean      :   5226
## 3rd Qu.:   49198    3rd Qu.:   5006    3rd Qu.:   5000    3rd Qu.:   4505
## Max.      : 961664    Max.      :873552    Max.      :1684259    Max.      :896040
## PAY_AMT4    PAY_AMT5    PAY_AMT6
## Min.       :      0    Min.       :    0.0    Min.       :    0.0
## 1st Qu.:    296    1st Qu.:   252.5    1st Qu.:   117.8
## Median :   1500    Median :   1500.0    Median :   1500.0
## Mean     :   4826    Mean      :  4799.4    Mean      :  5215.5
## 3rd Qu.:   4013    3rd Qu.:  4031.5    3rd Qu.:  4000.0
## Max.      :621000    Max.      :426529.0    Max.      :528666.0
## default.payment.next.month
## Min.      :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean     :0.2212
## 3rd Qu.:0.0000
## Max.      :1.0000
```

```
set.seed(1)
train = sample (nrow(credit_card), round(nrow(credit_card)*0.7))
credit_card.train = credit_card[train,]
credit_card.test = credit_card[-train,]
```

2. Preprocesado de los datos.

Lo primero que queremos hacer es comprobar si hay datos perdidos, y si es así; reemplazaremos el valor perdido.

```
anyNA(credit_card.train)
```

```
## [1] FALSE
```

Como no tenemos ningún dato perdido, no tendremos que reemplazar los valores. Si hubiéramos tenido valores tenidos, podríamos haber utilizado la función *knnImputation* para reemplazar los valores perdidos por los k vecinos más cercanos (normalmente k=3). También podríamos utilizar la media como sustituto del valor perdido.

Lo siguiente que vamos a hacer es modificar aquellas columnas que separan los datos en variables “clases”

como por ejemplo la columna *EDUCATION*, que indica que tipo de estudios tiene cada persona. Por cada tipo en los que los separe, crearemos una nueva columna que indique con 0s y 1s la pertenencia a ese tipo. También tenemos que realizar este proceso con la columna *MARRIAGE*

```
# Modificamos la columna 2, llamada sex, para dividir los datos en 0=mujer, 1=hombre.
credit_card.train$SEX = ifelse(credit_card.train$SEX == 2, 0, 1)
summary(credit_card.train)
```

```
##      LIMIT_BAL      SEX      EDUCATION      MARRIAGE
## Min.   : 10000   Min.   :0.0000   Min.    :0.000   Min.    :0.000
## 1st Qu.: 50000   1st Qu.:0.0000   1st Qu.:1.000   1st Qu.:1.000
## Median :140000   Median :0.0000   Median :2.000   Median :2.000
## Mean   :167948   Mean    :0.3974   Mean    :1.853   Mean    :1.549
## 3rd Qu.:240000   3rd Qu.:1.0000   3rd Qu.:2.000   3rd Qu.:2.000
## Max.   :1000000   Max.    :1.0000   Max.    :6.000   Max.    :3.000
##      AGE      PAY_0      PAY_2      PAY_3
## Min.   :21.0   Min.   :-2.00000   Min.   :-2.0000   Min.   :-2.0000
## 1st Qu.:28.0   1st Qu.:-1.00000   1st Qu.:-1.0000   1st Qu.:-1.0000
## Median :34.0   Median : 0.00000   Median : 0.0000   Median : 0.0000
## Mean   :35.5   Mean    :-0.01524   Mean    :-0.1321   Mean    :-0.1697
## 3rd Qu.:41.0   3rd Qu.: 0.00000   3rd Qu.: 0.0000   3rd Qu.: 0.0000
## Max.   :79.0   Max.    : 8.00000   Max.    : 8.0000   Max.    : 8.0000
##      PAY_4      PAY_5      PAY_6      BILL_AMT1
## Min.   :-2.000   Min.   :-2.0000   Min.   :-2.000   Min.   :-154973
## 1st Qu.:-1.000   1st Qu.:-1.0000   1st Qu.:-1.000   1st Qu.: 3508
## Median : 0.000   Median : 0.0000   Median : 0.000   Median : 22143
## Mean   :-0.221   Mean    :-0.2678   Mean    :-0.289   Mean    : 51238
## 3rd Qu.: 0.000   3rd Qu.: 0.0000   3rd Qu.: 0.000   3rd Qu.: 66175
## Max.    : 8.000   Max.    : 8.0000   Max.    : 8.000   Max.    : 964511
##      BILL_AMT2      BILL_AMT3      BILL_AMT4      BILL_AMT5
## Min.   :-69777   Min.    : -46127   Min.   :-170000   Min.   :-81334
## 1st Qu.: 2956    1st Qu.: 2552    1st Qu.: 2276    1st Qu.: 1730
## Median : 20895   Median : 20012   Median : 19044   Median : 18093
## Mean    : 49092   Mean     : 47104   Mean     : 43452   Mean     : 40395
## 3rd Qu.: 62795   3rd Qu.: 59674   3rd Qu.: 53536   3rd Qu.: 49719
## Max.    :983931   Max.    :1664089   Max.    : 891586   Max.    :927171
##      BILL_AMT6      PAY_AMT1      PAY_AMT2      PAY_AMT3
## Min.   :-339603   Min.    : 0       Min.    : 0.0     Min.    : 0
## 1st Qu.: 1261     1st Qu.: 1000    1st Qu.: 809.8    1st Qu.: 390
## Median : 17036    Median : 2100    Median : 2000.0    Median : 1794
## Mean    : 38948    Mean     : 5608    Mean     : 5988.9    Mean     : 5218
## 3rd Qu.: 48740    3rd Qu.: 5006    3rd Qu.: 5000.0    3rd Qu.: 4525
## Max.    :961664    Max.    :493358    Max.    :1684259.0   Max.    :896040
##      PAY_AMT4      PAY_AMT5      PAY_AMT6
## Min.    : 0       Min.    : 0.0     Min.    : 0
## 1st Qu.: 300      1st Qu.: 269.8    1st Qu.: 100
## Median : 1500     Median : 1500.0    Median : 1500
## Mean    : 4793     Mean     : 4838.0    Mean     : 5252
## 3rd Qu.: 4010     3rd Qu.: 4037.8    3rd Qu.: 4000
## Max.    :528897    Max.    :426529.0   Max.    :528666
## default.payment.next.month
## Min.    :0.0000
## 1st Qu.:0.0000
## Median :0.0000
## Mean    :0.2209
```

```
## 3rd Qu.:0.0000
## Max. :1.0000
```

```
# También tenemos que modificar la columna EDUCATION, la dividiremos en cuatro columnas diferentes:
# ed.other, ed.university, ed.high_school, ed.school
ed.other = ifelse(credit_card.train$EDUCATION == 4, 1, 0)
ed.university = ifelse(credit_card.train$EDUCATION == 2, 1, 0)
ed.high_school = ifelse(credit_card.train$EDUCATION == 3 | credit_card.train$EDUCATION == 2, 1, 0)
ed.school = ifelse(credit_card.train$EDUCATION == 1 | credit_card.train$EDUCATION == 2 | credit_card.train$EDUCATION == 3, 1, 0)

credit_card.train = cbind(credit_card.train, ed.other, ed.high_school, ed.school, ed.university)

# Borramos la columna EDUCATION.
credit_card.train = credit_card.train[, -which(colnames(credit_card.train) == "EDUCATION")]
summary(credit_card.train)
```

```
##      LIMIT_BAL      SEX      MARRIAGE      AGE
## Min.   : 10000   Min.   :0.0000   Min.   :0.000   Min.   :21.0
## 1st Qu.: 50000   1st Qu.:0.0000   1st Qu.:1.000   1st Qu.:28.0
## Median :140000   Median :0.0000   Median :2.000   Median :34.0
## Mean   :167948   Mean   :0.3974   Mean   :1.549   Mean   :35.5
## 3rd Qu.:240000   3rd Qu.:1.0000   3rd Qu.:2.000   3rd Qu.:41.0
## Max.   :1000000   Max.   :1.0000   Max.   :3.000   Max.   :79.0
##      PAY_0      PAY_2      PAY_3      PAY_4
## Min.   : -2.00000   Min.   : -2.0000   Min.   : -2.0000   Min.   : -2.000
## 1st Qu.: -1.00000   1st Qu.: -1.0000   1st Qu.: -1.0000   1st Qu.: -1.000
## Median : 0.00000   Median : 0.0000   Median : 0.0000   Median : 0.000
## Mean   : -0.01524   Mean   : -0.1321   Mean   : -0.1697   Mean   : -0.221
## 3rd Qu.: 0.00000   3rd Qu.: 0.0000   3rd Qu.: 0.0000   3rd Qu.: 0.000
## Max.   : 8.00000   Max.   : 8.0000   Max.   : 8.0000   Max.   : 8.000
##      PAY_5      PAY_6      BILL_AMT1      BILL_AMT2
## Min.   : -2.0000   Min.   : -2.000   Min.   : -154973   Min.   : -69777
## 1st Qu.: -1.0000   1st Qu.: -1.000   1st Qu.: 3508      1st Qu.: 2956
## Median : 0.0000   Median : 0.000    Median : 22143     Median : 20895
## Mean   : -0.2678   Mean   : -0.289    Mean   : 51238     Mean   : 49092
## 3rd Qu.: 0.0000   3rd Qu.: 0.000    3rd Qu.: 66175     3rd Qu.: 62795
## Max.   : 8.0000   Max.   : 8.000    Max.   : 964511     Max.   : 983931
##      BILL_AMT3      BILL_AMT4      BILL_AMT5      BILL_AMT6
## Min.   : -46127   Min.   : -170000   Min.   : -81334   Min.   : -339603
## 1st Qu.: 2552     1st Qu.: 2276     1st Qu.: 1730     1st Qu.: 1261
## Median : 20012    Median : 19044    Median : 18093    Median : 17036
## Mean   : 47104    Mean   : 43452    Mean   : 40395    Mean   : 38948
## 3rd Qu.: 59674    3rd Qu.: 53536    3rd Qu.: 49719    3rd Qu.: 48740
## Max.   :1664089    Max.   : 891586    Max.   : 927171    Max.   : 961664
##      PAY_AMT1      PAY_AMT2      PAY_AMT3      PAY_AMT4
## Min.   : 0         Min.   : 0.0      Min.   : 0         Min.   : 0
## 1st Qu.: 1000      1st Qu.: 809.8    1st Qu.: 390      1st Qu.: 300
## Median : 2100      Median : 2000.0    Median : 1794     Median : 1500
## Mean   : 5608      Mean   : 5988.9    Mean   : 5218     Mean   : 4793
## 3rd Qu.: 5006      3rd Qu.: 5000.0    3rd Qu.: 4525     3rd Qu.: 4010
## Max.   :493358     Max.   :1684259.0   Max.   :896040     Max.   :528897
##      PAY_AMT5      PAY_AMT6      default.payment.next.month
## Min.   : 0.0       Min.   : 0        Min.   :0.0000
## 1st Qu.: 269.8     1st Qu.: 100      1st Qu.:0.0000
## Median : 1500.0    Median : 1500     Median :0.0000
```

```
## Mean : 4838.0 Mean : 5252 Mean :0.2209
## 3rd Qu.: 4037.8 3rd Qu.: 4000 3rd Qu.:0.0000
## Max. :426529.0 Max. :528666 Max. :1.0000
## ed.other ed.high_school ed.school ed.university
## Min. :0.000000 Min. :0.0000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.000000 1st Qu.:0.0000 1st Qu.:1.0000 1st Qu.:0.0000
## Median :0.000000 Median :1.0000 Median :1.0000 Median :0.0000
## Mean :0.003857 Mean :0.6332 Mean :0.9844 Mean :0.4708
## 3rd Qu.:0.000000 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000
## Max. :1.000000 Max. :1.0000 Max. :1.0000 Max. :1.0000
```

También tenemos que modificar la columna marriage. Introduciremos tres nueva columnas: marriage.married, marriage.single, marriage.others

```
marriage.married = ifelse(credit_card.train$MARRIAGE == 1, 1,0)
marriage.single = ifelse(credit_card.train$MARRIAGE == 2, 1,0)
marriage.others = ifelse(credit_card.train$MARRIAGE == 3, 1,0)
```

Introducimos los datos.

```
credit_card.train = cbind(credit_card.train, marriage.married, marriage.single, marriage.others)
```

Borramos la variable MARRIAGE.

```
credit_card.train = credit_card.train[, -which(colnames(credit_card.train) == "MARRIAGE")]
summary(credit_card.train)
```

```
## LIMIT_BAL SEX AGE PAY_0
## Min. : 10000 Min. :0.0000 Min. :21.0 Min. : -2.00000
## 1st Qu.: 50000 1st Qu.:0.0000 1st Qu.:28.0 1st Qu.: -1.00000
## Median : 140000 Median :0.0000 Median :34.0 Median : 0.00000
## Mean : 167948 Mean :0.3974 Mean :35.5 Mean : -0.01524
## 3rd Qu.: 240000 3rd Qu.:1.0000 3rd Qu.:41.0 3rd Qu.: 0.00000
## Max. :1000000 Max. :1.0000 Max. :79.0 Max. : 8.00000
## PAY_2 PAY_3 PAY_4 PAY_5
## Min. : -2.0000 Min. : -2.0000 Min. : -2.0000 Min. : -2.0000
## 1st Qu.: -1.0000 1st Qu.: -1.0000 1st Qu.: -1.0000 1st Qu.: -1.0000
## Median : 0.0000 Median : 0.0000 Median : 0.0000 Median : 0.0000
## Mean : -0.1321 Mean : -0.1697 Mean : -0.221 Mean : -0.2678
## 3rd Qu.: 0.0000 3rd Qu.: 0.0000 3rd Qu.: 0.0000 3rd Qu.: 0.0000
## Max. : 8.0000 Max. : 8.0000 Max. : 8.0000 Max. : 8.0000
## PAY_6 BILL_AMT1 BILL_AMT2 BILL_AMT3
## Min. : -2.000 Min. : -154973 Min. : -69777 Min. : -46127
## 1st Qu.: -1.000 1st Qu.: 3508 1st Qu.: 2956 1st Qu.: 2552
## Median : 0.000 Median : 22143 Median : 20895 Median : 20012
## Mean : -0.289 Mean : 51238 Mean : 49092 Mean : 47104
## 3rd Qu.: 0.000 3rd Qu.: 66175 3rd Qu.: 62795 3rd Qu.: 59674
## Max. : 8.000 Max. : 964511 Max. : 983931 Max. :1664089
## BILL_AMT4 BILL_AMT5 BILL_AMT6 PAY_AMT1
## Min. : -170000 Min. : -81334 Min. : -339603 Min. : 0
## 1st Qu.: 2276 1st Qu.: 1730 1st Qu.: 1261 1st Qu.: 1000
## Median : 19044 Median : 18093 Median : 17036 Median : 2100
## Mean : 43452 Mean : 40395 Mean : 38948 Mean : 5608
## 3rd Qu.: 53536 3rd Qu.: 49719 3rd Qu.: 48740 3rd Qu.: 5006
## Max. : 891586 Max. : 927171 Max. : 961664 Max. :493358
## PAY_AMT2 PAY_AMT3 PAY_AMT4 PAY_AMT5
## Min. : 0.0 Min. : 0 Min. : 0 Min. : 0.0
## 1st Qu.: 809.8 1st Qu.: 390 1st Qu.: 300 1st Qu.: 269.8
## Median : 2000.0 Median : 1794 Median : 1500 Median : 1500.0
```

```
## Mean : 5988.9 Mean : 5218 Mean : 4793 Mean : 4838.0
## 3rd Qu.: 5000.0 3rd Qu.: 4525 3rd Qu.: 4010 3rd Qu.: 4037.8
## Max. :1684259.0 Max. :896040 Max. :528897 Max. :426529.0
## PAY_AMT6 default.payment.next.month ed.other
## Min. : 0 Min. :0.0000 Min. :0.000000
## 1st Qu.: 100 1st Qu.:0.0000 1st Qu.:0.000000
## Median : 1500 Median :0.0000 Median :0.000000
## Mean : 5252 Mean :0.2209 Mean :0.003857
## 3rd Qu.: 4000 3rd Qu.:0.0000 3rd Qu.:0.000000
## Max. :528666 Max. :1.0000 Max. :1.000000
## ed.high_school ed.school ed.university marriage.married
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:1.0000 1st Qu.:0.0000 1st Qu.:0.000
## Median :1.0000 Median :1.0000 Median :0.0000 Median :0.000
## Mean :0.6332 Mean :0.9844 Mean :0.4708 Mean :0.458
## 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:1.000
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.000
## marriage.single marriage.others
## Min. :0.0000 Min. :0.000000
## 1st Qu.:0.0000 1st Qu.:0.000000
## Median :1.0000 Median :0.000000
## Mean :0.5296 Mean :0.01076
## 3rd Qu.:1.0000 3rd Qu.:0.000000
## Max. :1.0000 Max. :1.000000
```

Tambi n vamos a cambiar el nombre de la columna *PAY0* por *PAY1*.

```
colnames(credit_card.train)[which(colnames(credit_card.train)=="PAY_0")]="PAY_1"
summary(credit_card.train)
```

```
## LIMIT_BAL SEX AGE PAY_1
## Min. : 10000 Min. :0.0000 Min. :21.0 Min. : -2.00000
## 1st Qu.: 50000 1st Qu.:0.0000 1st Qu.:28.0 1st Qu.: -1.00000
## Median : 140000 Median :0.0000 Median :34.0 Median : 0.00000
## Mean : 167948 Mean :0.3974 Mean :35.5 Mean : -0.01524
## 3rd Qu.: 240000 3rd Qu.:1.0000 3rd Qu.:41.0 3rd Qu.: 0.00000
## Max. :1000000 Max. :1.0000 Max. :79.0 Max. : 8.00000
## PAY_2 PAY_3 PAY_4 PAY_5
## Min. : -2.0000 Min. : -2.0000 Min. : -2.000 Min. : -2.0000
## 1st Qu.: -1.0000 1st Qu.: -1.0000 1st Qu.: -1.000 1st Qu.: -1.0000
## Median : 0.0000 Median : 0.0000 Median : 0.000 Median : 0.0000
## Mean : -0.1321 Mean : -0.1697 Mean : -0.221 Mean : -0.2678
## 3rd Qu.: 0.0000 3rd Qu.: 0.0000 3rd Qu.: 0.000 3rd Qu.: 0.0000
## Max. : 8.0000 Max. : 8.0000 Max. : 8.000 Max. : 8.0000
## PAY_6 BILL_AMT1 BILL_AMT2 BILL_AMT3
## Min. : -2.000 Min. : -154973 Min. : -69777 Min. : -46127
## 1st Qu.: -1.000 1st Qu.: 3508 1st Qu.: 2956 1st Qu.: 2552
## Median : 0.000 Median : 22143 Median : 20895 Median : 20012
## Mean : -0.289 Mean : 51238 Mean : 49092 Mean : 47104
## 3rd Qu.: 0.000 3rd Qu.: 66175 3rd Qu.: 62795 3rd Qu.: 59674
## Max. : 8.000 Max. : 964511 Max. : 983931 Max. : 1664089
## BILL_AMT4 BILL_AMT5 BILL_AMT6 PAY_AMT1
## Min. : -170000 Min. : -81334 Min. : -339603 Min. : 0
## 1st Qu.: 2276 1st Qu.: 1730 1st Qu.: 1261 1st Qu.: 1000
## Median : 19044 Median : 18093 Median : 17036 Median : 2100
```

```
## Mean : 43452 Mean : 40395 Mean : 38948 Mean : 5608
## 3rd Qu.: 53536 3rd Qu.: 49719 3rd Qu.: 48740 3rd Qu.: 5006
## Max. : 891586 Max. : 927171 Max. : 961664 Max. : 493358
## PAY_AMT2 PAY_AMT3 PAY_AMT4 PAY_AMT5
## Min. : 0.0 Min. : 0 Min. : 0 Min. : 0.0
## 1st Qu.: 809.8 1st Qu.: 390 1st Qu.: 300 1st Qu.: 269.8
## Median : 2000.0 Median : 1794 Median : 1500 Median : 1500.0
## Mean : 5988.9 Mean : 5218 Mean : 4793 Mean : 4838.0
## 3rd Qu.: 5000.0 3rd Qu.: 4525 3rd Qu.: 4010 3rd Qu.: 4037.8
## Max. : 1684259.0 Max. : 896040 Max. : 528897 Max. : 426529.0
## PAY_AMT6 default.payment.next.month ed.other
## Min. : 0 Min. : 0.0000 Min. : 0.000000
## 1st Qu.: 100 1st Qu.: 0.0000 1st Qu.: 0.000000
## Median : 1500 Median : 0.0000 Median : 0.000000
## Mean : 5252 Mean : 0.2209 Mean : 0.003857
## 3rd Qu.: 4000 3rd Qu.: 0.0000 3rd Qu.: 0.000000
## Max. : 528666 Max. : 1.0000 Max. : 1.000000
## ed.high_school ed.school ed.university marriage.married
## Min. : 0.0000 Min. : 0.0000 Min. : 0.0000 Min. : 0.000
## 1st Qu.: 0.0000 1st Qu.: 1.0000 1st Qu.: 0.0000 1st Qu.: 0.000
## Median : 1.0000 Median : 1.0000 Median : 0.0000 Median : 0.000
## Mean : 0.6332 Mean : 0.9844 Mean : 0.4708 Mean : 0.458
## 3rd Qu.: 1.0000 3rd Qu.: 1.0000 3rd Qu.: 1.0000 3rd Qu.: 1.000
## Max. : 1.0000 Max. : 1.0000 Max. : 1.0000 Max. : 1.000
## marriage.single marriage.others
## Min. : 0.0000 Min. : 0.000000
## 1st Qu.: 0.0000 1st Qu.: 0.000000
## Median : 1.0000 Median : 0.000000
## Mean : 0.5296 Mean : 0.01076
## 3rd Qu.: 1.0000 3rd Qu.: 0.000000
## Max. : 1.0000 Max. : 1.000000
```

Por Ãltimo.

```
# Por Ãltimo, utilizamos la funciÃ³n preprocess.
trans = preProcess(credit_card.train, c("BoxCox") )
trainTransformado = predict(trans, credit_card.train)
summary(trainTransformado)
```

```
## LIMIT_BAL SEX AGE PAY_1
## Min. : 49.50 Min. : 0.0000 Min. : 1.564 Min. : -2.00000
## 1st Qu.: 82.29 1st Qu.: 0.0000 1st Qu.: 1.622 1st Qu.: -1.00000
## Median : 113.27 Median : 0.0000 Median : 1.657 Median : 0.00000
## Mean : 111.23 Mean : 0.3974 Mean : 1.656 Mean : -0.01524
## 3rd Qu.: 133.74 3rd Qu.: 1.0000 3rd Qu.: 1.688 3rd Qu.: 0.00000
## Max. : 206.99 Max. : 1.0000 Max. : 1.775 Max. : 8.00000
## PAY_2 PAY_3 PAY_4 PAY_5
## Min. : -2.0000 Min. : -2.0000 Min. : -2.000 Min. : -2.0000
## 1st Qu.: -1.0000 1st Qu.: -1.0000 1st Qu.: -1.000 1st Qu.: -1.0000
## Median : 0.0000 Median : 0.0000 Median : 0.000 Median : 0.0000
## Mean : -0.1321 Mean : -0.1697 Mean : -0.221 Mean : -0.2678
## 3rd Qu.: 0.0000 3rd Qu.: 0.0000 3rd Qu.: 0.000 3rd Qu.: 0.0000
## Max. : 8.0000 Max. : 8.0000 Max. : 8.000 Max. : 8.0000
## PAY_6 BILL_AMT1 BILL_AMT2 BILL_AMT3
## Min. : -2.000 Min. : -154973 Min. : -69777 Min. : -46127
```

```
## 1st Qu.: -1.000 1st Qu.: 3508 1st Qu.: 2956 1st Qu.: 2552
## Median : 0.000 Median : 22143 Median : 20895 Median : 20012
## Mean : -0.289 Mean : 51238 Mean : 49092 Mean : 47104
## 3rd Qu.: 0.000 3rd Qu.: 66175 3rd Qu.: 62795 3rd Qu.: 59674
## Max. : 8.000 Max. : 964511 Max. : 983931 Max. : 1664089
## BILL_AMT4 BILL_AMT5 BILL_AMT6 PAY_AMT1
## Min. : -170000 Min. : -81334 Min. : -339603 Min. : 0
## 1st Qu.: 2276 1st Qu.: 1730 1st Qu.: 1261 1st Qu.: 1000
## Median : 19044 Median : 18093 Median : 17036 Median : 2100
## Mean : 43452 Mean : 40395 Mean : 38948 Mean : 5608
## 3rd Qu.: 53536 3rd Qu.: 49719 3rd Qu.: 48740 3rd Qu.: 5006
## Max. : 891586 Max. : 927171 Max. : 961664 Max. : 493358
## PAY_AMT2 PAY_AMT3 PAY_AMT4 PAY_AMT5
## Min. : 0.0 Min. : 0 Min. : 0 Min. : 0.0
## 1st Qu.: 809.8 1st Qu.: 390 1st Qu.: 300 1st Qu.: 269.8
## Median : 2000.0 Median : 1794 Median : 1500 Median : 1500.0
## Mean : 5988.9 Mean : 5218 Mean : 4793 Mean : 4838.0
## 3rd Qu.: 5000.0 3rd Qu.: 4525 3rd Qu.: 4010 3rd Qu.: 4037.8
## Max. : 1684259.0 Max. : 896040 Max. : 528897 Max. : 426529.0
## PAY_AMT6 default.payment.next.month ed.other
## Min. : 0 Min. : 0.0000 Min. : 0.000000
## 1st Qu.: 100 1st Qu.: 0.0000 1st Qu.: 0.000000
## Median : 1500 Median : 0.0000 Median : 0.000000
## Mean : 5252 Mean : 0.2209 Mean : 0.003857
## 3rd Qu.: 4000 3rd Qu.: 0.0000 3rd Qu.: 0.000000
## Max. : 528666 Max. : 1.0000 Max. : 1.000000
## ed.high_school ed.school ed.university marriage.married
## Min. : 0.0000 Min. : 0.0000 Min. : 0.0000 Min. : 0.000
## 1st Qu.: 0.0000 1st Qu.: 1.0000 1st Qu.: 0.0000 1st Qu.: 0.000
## Median : 1.0000 Median : 1.0000 Median : 0.0000 Median : 0.000
## Mean : 0.6332 Mean : 0.9844 Mean : 0.4708 Mean : 0.458
## 3rd Qu.: 1.0000 3rd Qu.: 1.0000 3rd Qu.: 1.0000 3rd Qu.: 1.000
## Max. : 1.0000 Max. : 1.0000 Max. : 1.0000 Max. : 1.000
## marriage.single marriage.others
## Min. : 0.0000 Min. : 0.000000
## 1st Qu.: 0.0000 1st Qu.: 0.000000
## Median : 1.0000 Median : 0.000000
## Mean : 0.5296 Mean : 0.01076
## 3rd Qu.: 1.0000 3rd Qu.: 0.000000
## Max. : 1.0000 Max. : 1.000000
```

Para hacer más sencillo hacer las transformaciones al conjunto de test, se crearán funciones para reemplazar los valores faltantes.

Función para comprobar si hay datos perdidos y reemplazarlos.

```
reemplazarCol = function(x){
  # Columna SEX
  x$SEX = ifelse(x$SEX == 2, 0, 1)

  # Columna EDUCATION.
  x.other = ifelse(x$EDUCATION == 4, 1, 0)
  x.university = ifelse(x$EDUCATION == 2, 1, 0)
  x.high_school = ifelse(x$EDUCATION == 3 | x$EDUCATION == 2, 1, 0)
  x.school = ifelse(x$EDUCATION == 1 | x$EDUCATION == 2 | x$EDUCATION == 3, 1, 0)
  x = cbind(x, x.other, x.high_school, x.school, x.university)
```



```

# Borramos la columna EDUCATION.
x = x[,-which(colnames(x) == "EDUCATION")]

# Columna MARRIAGE.
x.married = ifelse(x$MARRIAGE == 1, 1,0)
x.single = ifelse(x$MARRIAGE == 2, 1,0)
x.others = ifelse(x$MARRIAGE == 3, 1,0)

# Introducimos los datos.
x = cbind(x.train, x.married, x.single, x.others)

# Borramos la variable MARRIAGE.
x = x[, -which(colnames(x) == "MARRIAGE")]

# Cambiamos el nombre de la variables PAY_0.
colnames(x)[which(colnames(x)=="PAY_0")]="PAY_1"
x
}
preprocesar = function(x,pred=trans){
  transTest = predict(pred, x)
  transTest
}

```