

# Pruebas duplicados

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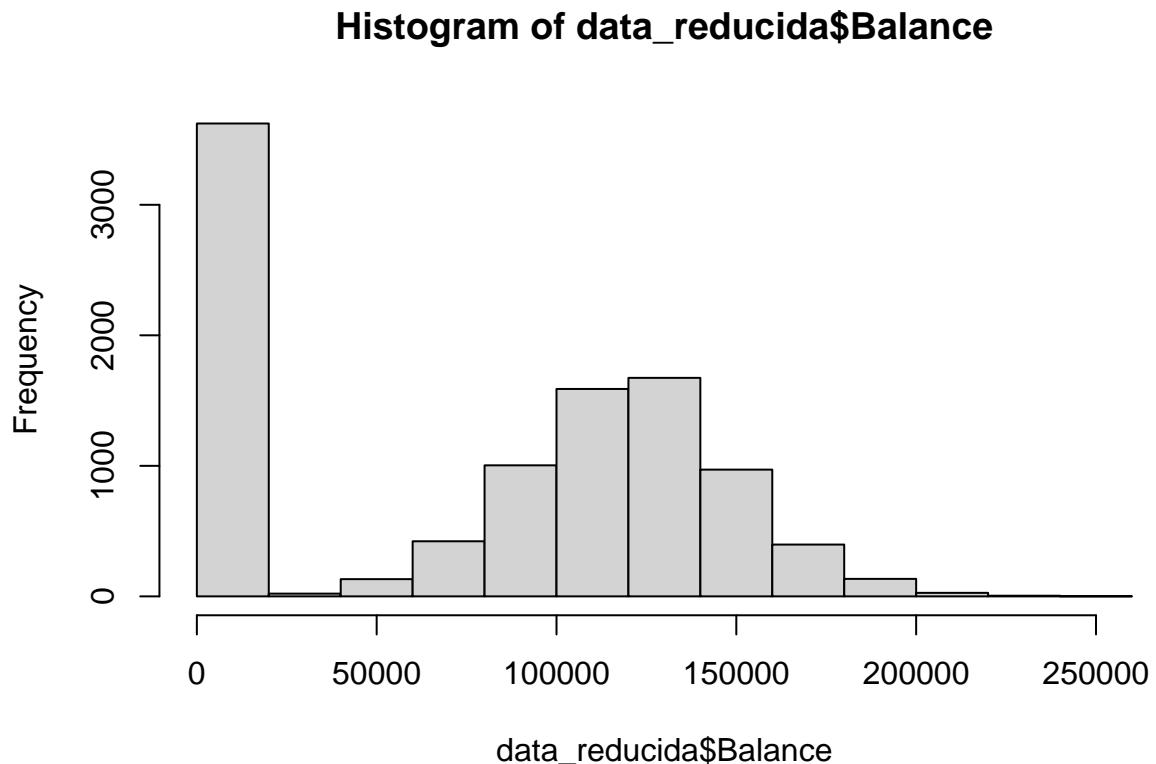
```
load("~/GitHub/Mineria/DATA/dataaaaaaaaaaaaa.RData")
library(dplyr)

## 
## Adjuntando el paquete: 'dplyr'

## The following objects are masked from 'package:stats':
## 
##     filter, lag

## The following objects are masked from 'package:base':
## 
##     intersect, setdiff, setequal, union

hist(data_reducida$Balance)
```



A continuación se hará la prueba de buscar valores duplicados discretizando la variable Balance

```

cortes <- c(-Inf, 0, 30000, 60000, 90000, 120000, 150000, 180000, 210000, 240000, Inf)
etiquetas <- c("Cero", "0-30k", "30k-60k", "60k-90k", "90k-120k", "120k-150k",
             "150k-180k", "180k-210k", "210k-240k", ">240k")
data_reducida$Balance <- cut(
  data_reducida$Balance,
  breaks = cortes,
  labels = etiquetas,
  right = TRUE,
  include.lowest = TRUE
)

train_df <- data_reducida[data_reducida$group == "train", ]
test_df <- data_reducida[data_reducida$group == "test", ]

```

Conteo duplicados

```

feature_cols <- setdiff(names(train_df), c("Exited", "group"))

dup_completos_train <- train_df[
  duplicated(train_df) |
  duplicated(train_df, fromLast = TRUE),
]

nrow(dup_completos_train)

## [1] 4479

```

Número idéntico

```

table(dup_completos_train$Exited)

##
##      0      1
## 3987  492

prop.table(table(dup_completos_train$Exited))

##
##          0           1
## 0.8901541 0.1098459

```

Duplicados más frecuentes en train:

```

freq_dup_train <- train_df %>%
  group_by(across(all_of(feature_cols))) %>%
  summarise(
    n      = n(),
    n_0   = sum(Exited == 0),
    n_1   = sum(Exited == 1),
    prop_0 = mean(Exited == 0),

```

```

    .groups = "drop"
) %>%
  arrange(desc(n), desc(n_0))
freq_dup_train

## # A tibble: 3,283 x 10
##   Geography Gender IsActiveMember NumOfProducts_grupo   Age Balance     n   n_0
##   <fct>     <fct>   <fct>           <dbl> <fct>   <int> <int>
## 1 France     Male     0             2                 37 Cero      22   18
## 2 France     Male     0             2                 38 Cero      21   19
## 3 France     Male     1             2                 32 Cero      18   18
## 4 France     Male     1             2                 33 Cero      18   18
## 5 France     Female   0             2                 35 Cero      17   15
## 6 France     Male     0             2                 30 Cero      16   16
## 7 France     Male     1             2                 37 Cero      16   16
## 8 France     Male     1             2                 38 Cero      16   16
## 9 France     Male     1             2                 40 Cero      16   16
## 10 Spain    Male     0             2                 36 Cero      14   14
## # i 3,273 more rows
## # i 2 more variables: n_1 <int>, prop_0 <dbl>

```

## Imputación de coincidencias perfectas

```

library(dplyr)

MIN_N <- 7
# Identificar combinaciones perfectamente predictivas de Exited = 0 con soporte mínimo
perfect_combinations <- freq_dup_train %>%
  filter(prop_0 == 1, n >= MIN_N) %>%
  select(all_of(feature_cols))
# Convertir columnas a carácter para la comparación
perfect_combinations_clean <- perfect_combinations %>%
  mutate(across(all_of(feature_cols), as.character))

# poner todas como carácter para que coincidan con las combinaciones de antes, exceptoExited
test_df_cleaned <- test_df %>%
  mutate(across(all_of(feature_cols), as.character),
        Exited = as.numeric(as.character(Exited)))

# Unir el data frame de prueba con las combinaciones perfectas e imputar Exited = 0 en caso de coincidencia
test_df_imputed <- test_df_cleaned %>%
  left_join(
    perfect_combinations_clean %>% mutate(match_found = TRUE),
    by = feature_cols
  ) %>%
  mutate(
    Exited = if_else(
      match_found == TRUE,
      0,
      Exited
    )
  )

```

```

        )
) %>%
select(-match_found)
# Crear el data frame final añadiendo la columna Exited imputada al data frame original
test_df_final <- test_df %>%
  mutate(Exited = test_df_imputed$Exited)

summary(test_df_final)

##      Geography          Gender         Exited      IsActiveMember NumOfProducts_grupo
##  France :1527   Female:1368   Min.    :0       0:1434           1     :1491
##  Germany: 738   Male  :1632   1st Qu.:0      1:1566           2     :1409
##  Spain  : 735                    Median :0      3 o más: 100
##                               Mean   :0
##                               3rd Qu.:0
##                               Max.   :0
##                               NA's    :2772
##      Age                  Balance      group
##  Min.   :18.00    Cero    :1108    test  :3000
##  1st Qu.:32.00  90k-120k : 653    train:   0
##  Median :37.00  120k-150k: 645
##  Mean   :39.02  60k-90k  : 253
##  3rd Qu.:44.00  150k-180k: 237
##  Max.   :84.00  180k-210k:  50
##                (Other)   : 54

```