

# TFG Codigo Modelos PDF

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## Librerias

```
library(MASS)
library(glmnet)
library(rpart)
library(rpart.plot)
library(randomForest)
library(tidyverse)
library(readr)
library(psych)
library(ggplot2)
library(dplyr)
library(corrplot)
library(RColorBrewer)
library(gridExtra)
library(caret)
library(pROC)
library(car)
library(klaR)
```

Warning: package 'klaR' was built under R version 4.4.3

```
# library(MXM)
# library(parallel)
# library(doParallel)
setwd("C:\\Users\\diego\\OneDrive\\Escritorio\\UCM\\Cuarto\\Segundo Cuatri\\TFG")
```

## Modelos

Ya tenemos la base de datos depurada, por tanto vamos a llamarla, y seguimos trabajando aquí

### Base de datos depurada

```
load("DatosDepurados.Rda")
datos$WALLSMATERIAL_MODE <- as.factor(datos$WALLSMATERIAL_MODE)
datos$FONDKAPREMONT_MODE <- as.factor(datos$FONDKAPREMONT_MODE)
datos$TARGET <- as.factor(datos$TARGET)
datos$TARGET <- factor(datos$TARGET, levels = c(0,1), labels = c("PAYER", "NOT_PAYER"))
table(datos$TARGET)/nrow(datos)
```

```
      PAYER  NOT_PAYER
0.91334135 0.08665865
```

### Train / test

```
set.seed(12345)
trainIndex <- createDataPartition(datos$TARGET, p=0.8, list=FALSE)
data_train <- datos[trainIndex,]
data_test <- datos[-trainIndex,]
data.frame(sort(colSums(is.na(data_train))))
```

```
sort.colSums.is.na.data_train...
EXT_SOURCE_3      0
EXT_SOURCE_2      0
DAYS_BIRTH        0
AMT_GOODS_PRICE   0
FLAG_OWN_CAR      0
EXT_SOURCE_1      0
CODE_GENDER       0
NAME_EDUCATION_TYPE 0
DAYS_EMPLOYED     0
REGION_RATING_CLIENT 0
AMT_CREDIT        0
NAME_INCOME_TYPE  0
```

NAME_CONTRACT_TYPE	0
AMT_CREDIT_RANGE	0
REGION_POPULATION_RELATIVE	0
NAME_HOUSING_TYPE	0
FLAG_WORK_PHONE	0
DEF_30_CNT_SOCIAL_CIRCLE	0
REG_CITY_NOT_LIVE_CITY	0
DAYS_REGISTRATION	0
REGION_RATING_CLIENT_W_CITY	0
FLAG_DOCUMENT_3	0
AGE_GROUP	0
EMPLOYMENT_YEAR	0
FLAG_PHONE	0
OWN_CAR_AGE	0
CNT_CHILDREN	0
DAYS_LAST_PHONE_CHANGE	0
FLAG_DOCUMENT_18	0
NAME_TYPE_SUITE	0
FLAG_DOCUMENT_16	0
WEEKDAY_APPR_PROCESS_START	0
REG_CITY_NOT_WORK_CITY	0
AMT_ANNUITY	0
WALLSMATERIAL_MODE	0
AMT_INCOME_TOTAL	0
HOURLY_APPR_PROCESS_START	0
AMT_REQ_CREDIT_BUREAU_QRT	0
APARTMENTS_AVG	0
FLOORSMAX_AVG	0
FLAG_DOCUMENT_5	0
FLAG_DOCUMENT_2	0
FONDKAPREMONT_MODE	0
OBS_30_CNT_SOCIAL_CIRCLE	0
YEARS_EMPLOYED	0
TARGET	0

## Forward

## AIC

```
# null<-glm(TARGET~1,data=data_train,family=binomial)
# full<-glm(TARGET~.,data=data_train,family=binomial)
# AIC
```

```
#modeloForwAIC<-stats::step(null, scope=list(lower=null, upper=full), direction="forward", tr
# una vez sabemos el modelo podemos escribirlo de esta forma
modeloForwAIC <- glm(TARGET~EXT_SOURCE_2+EXT_SOURCE_3+EXT_SOURCE_1+
  AMT_CREDIT_RANGE+NAME_HOUSING_TYPE, data = data_train, family = "binomial")
```

## BIC

```
#modeloForwBIC<-stats::step(null, scope=list(lower=null, upper=full), direction="forward", k
# de la misma forma hacemos lo mismo, esto se realiza porque cada vez que vamos a ejecutar l
modeloForwBIC <- glm(TARGET~EXT_SOURCE_2+EXT_SOURCE_3+EXT_SOURCE_1+NAME_EDUCATION_TYPE+
  REGION_RATING_CLIENT_W_CITY+
  DAYS_LAST_PHONE_CHANGE+
  NAME_CONTRACT_TYPE+AMT_I
  FLAG_PHONE+FLOORSMAX_AVG, data = data_train,family = "binomial")
```

## Discriminante

```
moddis <- lda(TARGET~., data=data_train)
```

Warning in lda.default(x, grouping, ...): variables are collinear

Tenemos este problema

```
# Assume your data is in `df` # Step 1: Expand factor variables like LDA would
X_expanded <- model.matrix(~ . - 1, data = data_train)
# Step 2: Identify constant or near-zero variance columns
nzv <- nearZeroVar(X_expanded, saveMetrics = TRUE)
# View constant or near-constant columns
constant_or_nzv <- nzv[nzv$zeroVar | nzv$nzv, ]
print(constant_or_nzv)
```

	freqRatio	percentUnique	zeroVar	nzv
EXT_SOURCE_3	47.74501	0.3979240471	FALSE	TRUE
NAME_EDUCATION_TYPEIncomplete higher	24.94902	0.0009923293	FALSE	TRUE
NAME_EDUCATION_TYPELower secondary	108.29826	0.0009923293	FALSE	TRUE
NAME_INCOME_TYPEMaternity leave	40308.20000	0.0009923293	FALSE	TRUE
NAME_INCOME_TYPEPensioner	25192.25000	0.0009923293	FALSE	TRUE
NAME_INCOME_TYPEStudent	13435.40000	0.0009923293	FALSE	TRUE
NAME_INCOME_TYPEUnemployed	11854.64706	0.0009923293	FALSE	TRUE

AMT_CREDIT_RANGE900K-1M	32.35750	0.0009923293	FALSE	TRUE
NAME_HOUSING_TYPEMunicipal apartment	26.86479	0.0009923293	FALSE	TRUE
NAME_HOUSING_TYPEOffice apartment	110.22848	0.0009923293	FALSE	TRUE
NAME_HOUSING_TYPERented apartment	52.90372	0.0009923293	FALSE	TRUE
EMPLOYMENT_YEAR20-30	29.07700	0.0009923293	FALSE	TRUE
EMPLOYMENT_YEAR30-40	104.79843	0.0009923293	FALSE	TRUE
EMPLOYMENT_YEAR40-50	1428.40426	0.0009923293	FALSE	TRUE
OWN_CAR_AGE	23.99482	0.0292737142	FALSE	TRUE
FLAG_DOCUMENT_181	101.88208	0.0009923293	FALSE	TRUE
NAME_TYPE_SUITEChildren	111.65847	0.0009923293	FALSE	TRUE
NAME_TYPE_SUITEGroup of people	1170.77907	0.0009923293	FALSE	TRUE
NAME_TYPE_SUITEOther_A	329.94581	0.0009923293	FALSE	TRUE
NAME_TYPE_SUITEOther_B	169.36855	0.0009923293	FALSE	TRUE
NAME_TYPE_SUITESpouse, partner	25.16461	0.0009923293	FALSE	TRUE
FLAG_DOCUMENT_161	83.04754	0.0009923293	FALSE	TRUE
WALLSMATERIAL_MODEBlock	32.58540	0.0009923293	FALSE	TRUE
WALLSMATERIAL_MODEMixed	129.11362	0.0009923293	FALSE	TRUE
WALLSMATERIAL_MODEMonolithic	173.34775	0.0009923293	FALSE	TRUE
WALLSMATERIAL_MODEOthers	183.73511	0.0009923293	FALSE	TRUE
WALLSMATERIAL_MODEWooden	56.20863	0.0009923293	FALSE	TRUE
APARTMENTS_AVG	23.29487	1.0880890715	FALSE	TRUE
FLAG_DOCUMENT_51	60.39080	0.0009923293	FALSE	TRUE
FLAG_DOCUMENT_21	16794.50000	0.0009923293	FALSE	TRUE
FONDKAPREMONT_MODEnot specified	51.39043	0.0009923293	FALSE	TRUE
FONDKAPREMONT_MODEorg spec account	53.53084	0.0009923293	FALSE	TRUE
FONDKAPREMONT_MODEreg oper spec account	24.18695	0.0009923293	FALSE	TRUE

**otros**

```
(modelo_rf <- randomForest(TARGET ~ ., data = data_train, ntree = 100))
```

Call:

```
randomForest(formula = TARGET ~ ., data = data_train, ntree = 100)
```

```
  Type of random forest: classification
```

```
    Number of trees: 100
```

```
No. of variables tried at each split: 6
```

```
  OOB estimate of  error rate: 8.67%
```

Confusion matrix:

	PAYER	NOT_PAYER	class.error
PAYER	183898	182	0.0009887006
NOT_PAYER	17290	176	0.9899232795

```
# Call:
# randomForest(formula = TARGET ~ ., data = data_train, ntree = 100)
#           Type of random forest: classification
#           Number of trees: 100
# No. of variables tried at each split: 6
#
# OOB estimate of error rate: 8.66%
#Confusion matrix: #           PAYER NOT_PAYER class.error
# PAYER      183884      186 0.001010485
# NOT_PAYER  17266      194 0.988888889
```

```
modelos<-list(modeloForwBIC, modeloForwAIC)
sapply(modelos,function(x) formula(x))
```

[[1]]

```
TARGET ~ EXT_SOURCE_2 + EXT_SOURCE_3 + EXT_SOURCE_1 + NAME_EDUCATION_TYPE +
CODE_GENDER + FLAG_DOCUMENT_3 + FLAG_OWN_CAR + REG_CITY_NOT_LIVE_CITY +
DEF_30_CNT_SOCIAL_CIRCLE + DAYS_EMPLOYED + REGION_RATING_CLIENT_W_CITY +
DAYS_LAST_PHONE_CHANGE + NAME_CONTRACT_TYPE + AMT_REQ_CREDIT_BUREAU_QRT +
FLAG_DOCUMENT_18 + FLAG_DOCUMENT_16 + OWN_CAR_AGE + AMT_ANNUITY +
AMT_GOODS_PRICE + AMT_CREDIT + FLAG_WORK_PHONE + DAYS_REGISTRATION +
NAME_INCOME_TYPE + FLAG_PHONE + FLOORSMAX_AVG
```

[[2]]

```
TARGET ~ EXT_SOURCE_2 + EXT_SOURCE_3 + EXT_SOURCE_1 + NAME_EDUCATION_TYPE +
CODE_GENDER + FLAG_DOCUMENT_3 + FLAG_OWN_CAR + REG_CITY_NOT_LIVE_CITY +
DEF_30_CNT_SOCIAL_CIRCLE + DAYS_EMPLOYED + REGION_RATING_CLIENT_W_CITY +
DAYS_LAST_PHONE_CHANGE + NAME_CONTRACT_TYPE + WALLSMATERIAL_MODE +
AMT_REQ_CREDIT_BUREAU_QRT + FLAG_DOCUMENT_18 + FLAG_DOCUMENT_16 +
OWN_CAR_AGE + AMT_ANNUITY + AMT_GOODS_PRICE + AMT_CREDIT +
FLAG_WORK_PHONE + DAYS_REGISTRATION + NAME_INCOME_TYPE +
FLAG_PHONE + FLOORSMAX_AVG + FONDKAPREMONT_MODE + DAYS_BIRTH +
NAME_TYPE_SUITE + FLAG_DOCUMENT_2 + FLAG_DOCUMENT_5 + YEARS_EMPLOYED +
REGION_RATING_CLIENT + REGION_POPULATION_RELATIVE + HOUR_APPR_PROCESS_START +
WEEKDAY_APPR_PROCESS_START + AMT_CREDIT_RANGE + NAME_HOUSING_TYPE
```

## Analisis de modelos

### Validacion cruzada

```
data_train$TARGET <- relevel(data_train$TARGET, ref = "NOT_PAYER")
# evaluamos los modelos glm
modelos<-list(moddis, modeloForwBIC, modeloForwAIC)
metodos<-list("lda", "glm", "glm")
titulos<-list("LDA", "Forward BIC", "Forward AIC")
vcrTodosModelos<-list()
predClassCounts <- list()
for (i in 1:length(modelos)){
  set.seed(12345)
  vcr<-train(formula(modelos[[i]]),
             data = data_train,
             method = metodos[[i]],
             family="binomial",
             trControl = trainControl(method="repeatedcv",
                                     number=5,
                                     repeats=5,
                                     summaryFunction=twoClassSummary,
                                     classProbs=TRUE,
                                     savePredictions = TRUE)
             )
  vcrTodosModelos[[ titulos[[i]] ]] <- vcr
  # Predict class probabilities on test data
  probs <- predict(vcr, newdata = data_test, type = "prob")
  # Apply 0.5 threshold to NOT_PAYER probability
  pred_classes <- ifelse(probs$NOT_PAYER >= 0.5, "NOT_PAYER", "PAYER")
  # Count predicted classes
  class_counts <- table(Predicted = pred_classes)
  predClassCounts[[ titulos[[i]] ]] <- class_counts
}
```

Warning in train.default(x, y, weights = w, ...): The metric "Accuracy" was not in the result set. ROC will be used instead.

Warning in lda.default(x, grouping, ...): variables are collinear

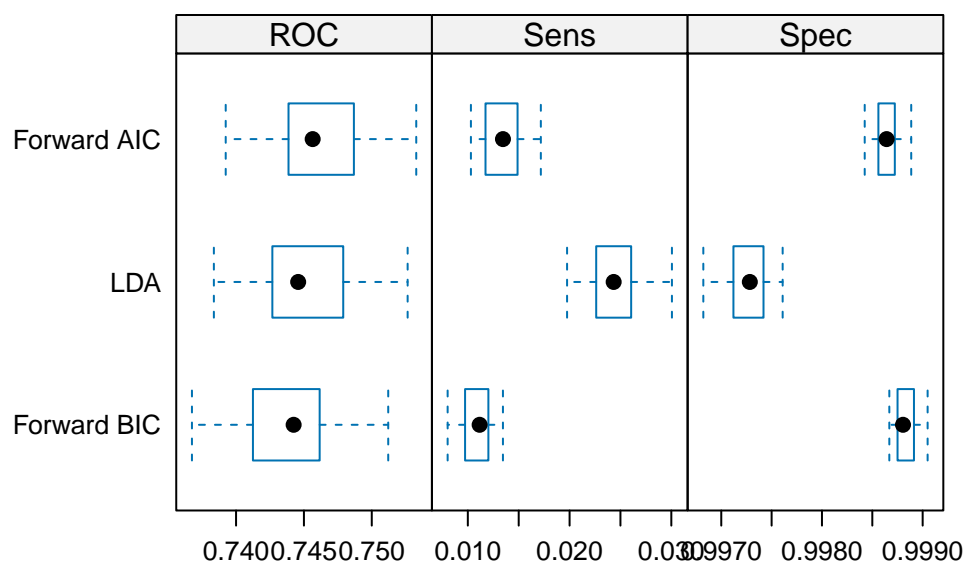
Warning in lda.default(x, grouping, ...): variables are collinear

Warning in lda.default(x, grouping, ...): variables are collinear

Warning in lda.default(x, grouping, ...): variables are collinear







```
summary(resamples(vcrTodosModelos), metric=c("ROC", "Sens", "Spec"))
```

Call:

```
summary.resamples(object = resamples(vcrTodosModelos), metric =  
  c("ROC", "Sens", "Spec"))
```

Models: LDA, Forward BIC, Forward AIC

Number of resamples: 25

ROC

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
LDA	0.7383652	0.7426743	0.7445765	0.7451961	0.7478972	0.7526424	0
Forward BIC	0.7367485	0.7412434	0.7442415	0.7441938	0.7461598	0.7512106	0
Forward AIC	0.7392366	0.7438641	0.7456450	0.7461731	0.7486817	0.7532757	0

Sens

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
LDA	0.019753793	0.022616662	0.02433438	0.02447037	0.02605210	0.03006012	
Forward BIC	0.008016032	0.009730967	0.01116199	0.01090116	0.01202061	0.01345163	
Forward AIC	0.010303377	0.011737761	0.01345548	0.01338600	0.01488692	0.01717721	

```
LDA          0
Forward BIC  0
Forward AIC  0
```

Spec

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
LDA	0.9968220	0.9971208	0.9972838	0.9972784	0.9974196	0.9976097	0
Forward BIC	0.9986691	0.9987505	0.9988049	0.9988266	0.9989135	0.9990493	0
Forward AIC	0.9984246	0.9985604	0.9986419	0.9986430	0.9987234	0.9988864	0

## Matrices de confusion

```
# Get predicted probabilities
probs <- predict(vcrTodosModelos[["Forward BIC"]], data_train, type = "prob")
# Use ROC analysis to find best threshold
# roc_obj <- roc(data_train$TARGET, probs)
# best_thresh <- coords(roc_obj, "best", ret = "threshold", best.method = "youden")
#
# print(best_thresh)
```

## Train

```
probs <- predict(vcrTodosModelos[["Forward BIC"]], data_train, type = "prob")

cm2<-confusionMatrix(data=as.factor(ifelse(probs$NOT_PAYER>=0.08,"NOT_PAYER","PAYER")),
                      reference=data_train$TARGET, positive="NOT_PAYER")

cm2$table
```

	Reference	
Prediction	NOT_PAYER	PAYER
NOT_PAYER	12437	64953
PAYER	5029	119127

```
cm2$overall[1:2]
```

	Accuracy	Kappa
	0.6527741	0.1407215

```
cm2$byClass[1:2]
```

```
Sensitivity Specificity  
0.7120692 0.6471480
```

## Test

```
probs <- predict(vcrTodosModelos[["Forward BIC"]], data_test, type = "prob")  
cm_test<-confusionMatrix(data=as.factor(ifelse(probs$NOT_PAYER>=0.08,"NOT_PAYER","PAYER")),  
                          reference=data_test$TARGET, positive="NOT_PAYER")
```

Warning in confusionMatrix.default(data = as.factor(ifelse(probs\$NOT\_PAYER >= :  
Levels are not in the same order for reference and data. Refactoring data to  
match.

```
cm_test$table
```

	Reference	
Prediction	PAYER	NOT_PAYER
PAYER	29806	1233
NOT_PAYER	16213	3133

```
cm_test$overall[1:2]
```

```
Accuracy Kappa  
0.6537462 0.1430914
```

```
cm_test$byClass[1:2]
```

```
Sensitivity Specificity  
0.7175905 0.6476890
```

## Curva ROC

### Train

```
probs <- predict(modeloForwBIC, data_train, type = "response")
curvaROC<-roc(data_train$TARGET, probs)
```

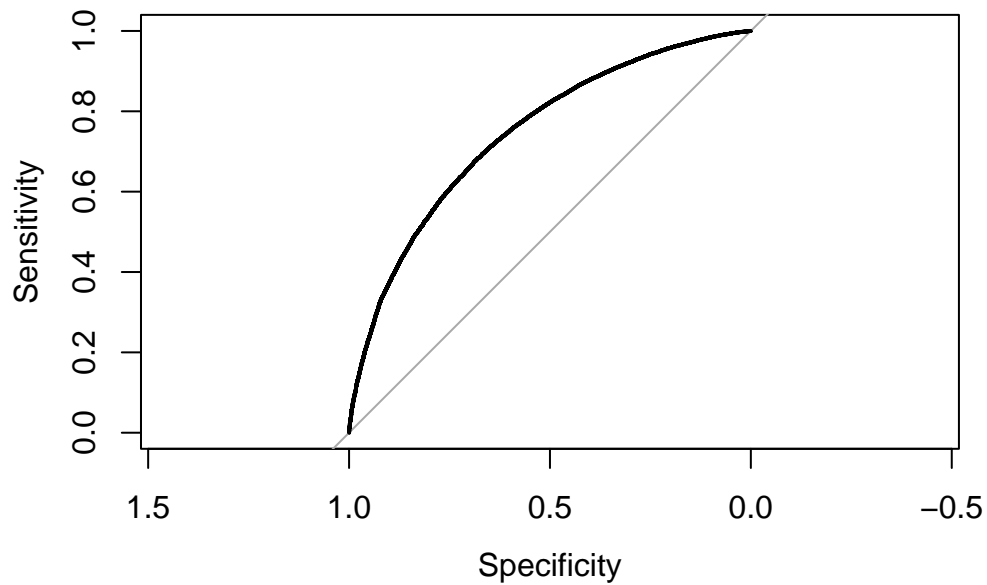
Setting levels: control = NOT\_PAYER, case = PAYER

Setting direction: controls > cases

```
curvaROC$auc
```

Area under the curve: 0.7448

```
plot(curvaROC)
```



## Test

```
probs_test <- predict(modeloForwBIC, data_test, type = "response")
curvaROC_test<-roc(data_test$TARGET, probs_test)
```

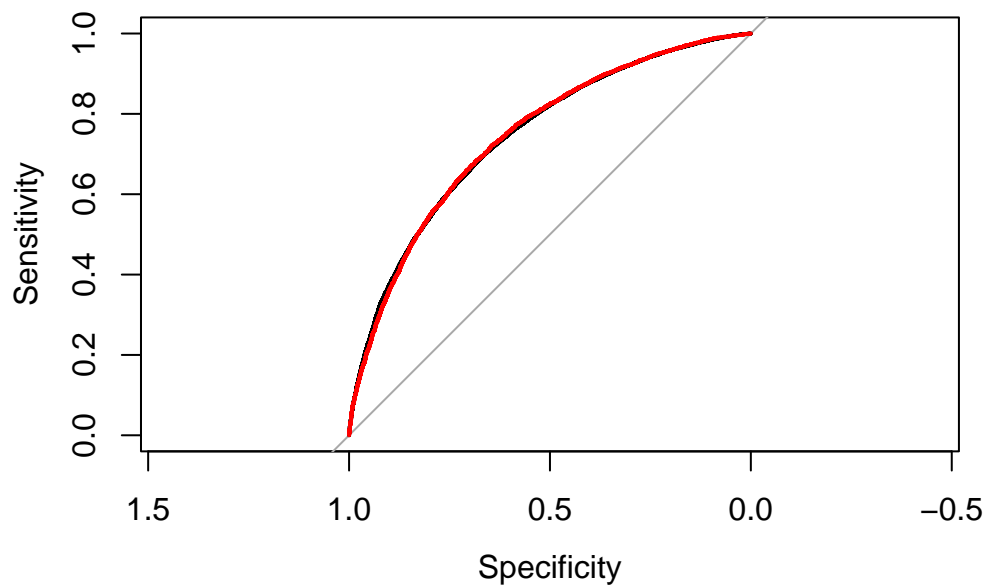
Setting levels: control = PAYER, case = NOT\_PAYER

Setting direction: controls < cases

```
curvaROC_test$auc
```

Area under the curve: 0.7448

```
plot(curvaROC)  
plot(curvaROC_test, add=T, col="red")
```



```
anova(modeloForwBIC)
```

Analysis of Deviance Table

Model: binomial, link: logit

Response: TARGET

Terms added sequentially (first to last)

	Df	Deviance	Resid. Df	Resid. Dev	Pr(>Chi)
NULL			201545	118808	

EXT_SOURCE_2	1	5398.8	201544	113409	< 2.2e-16	***
EXT_SOURCE_3	1	3685.8	201543	109723	< 2.2e-16	***
EXT_SOURCE_1	1	789.7	201542	108934	< 2.2e-16	***
NAME_EDUCATION_TYPE	4	608.3	201538	108325	< 2.2e-16	***
CODE_GENDER	1	222.9	201537	108102	< 2.2e-16	***
FLAG_DOCUMENT_3	1	289.8	201536	107813	< 2.2e-16	***
FLAG_OWN_CAR	1	231.1	201535	107582	< 2.2e-16	***
REG_CITY_NOT_LIVE_CITY	1	74.8	201534	107507	< 2.2e-16	***
DEF_30_CNT_SOCIAL_CIRCLE	1	104.6	201533	107402	< 2.2e-16	***
DAYS_EMPLOYED	1	150.8	201532	107251	< 2.2e-16	***
REGION_RATING_CLIENT_W_CITY	2	100.4	201530	107151	< 2.2e-16	***
DAYS_LAST_PHONE_CHANGE	1	61.0	201529	107090	5.852e-15	***
NAME_CONTRACT_TYPE	1	71.1	201528	107019	< 2.2e-16	***
AMT_REQ_CREDIT_BUREAU_QRT	1	35.8	201527	106983	2.231e-09	***
FLAG_DOCUMENT_18	1	43.5	201526	106940	4.329e-11	***
FLAG_DOCUMENT_16	1	37.7	201525	106902	8.286e-10	***
OWN_CAR_AGE	1	31.9	201524	106870	1.594e-08	***
AMT_ANNUITY	1	23.5	201523	106847	1.224e-06	***
AMT_GOODS_PRICE	1	131.4	201522	106715	< 2.2e-16	***
AMT_CREDIT	1	302.0	201521	106413	< 2.2e-16	***
FLAG_WORK_PHONE	1	54.8	201520	106358	1.364e-13	***
DAYS_REGISTRATION	1	17.6	201519	106341	2.775e-05	***
NAME_INCOME_TYPE	7	328.5	201512	106012	< 2.2e-16	***
FLAG_PHONE	1	12.8	201511	106000	0.0003396	***
FLOORSMAX_AVG	1	50.4	201510	105949	1.268e-12	***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

## Otros modelos AUC, Matrices

Train

```
probs <- predict(vcrTodosModelos[["LDA"]], data_train, type = "prob")

cm2<-confusionMatrix(data=as.factor(ifelse(probs$NOT_PAYER>=0.08,"NOT_PAYER","PAYER")),
                      reference=data_train$TARGET, positive="NOT_PAYER")

cm2$table
```

	Reference	
Prediction	NOT_PAYER	PAYER

NOT_PAYER	12301	62154
PAYER	5165	121926

```
cm2$overall[1:2]
```

Accuracy	Kappa
0.6659869	0.1480381

```
cm2$byClass[1:2]
```

Sensitivity	Specificity
0.7042826	0.6623533

Test

```
probs <- predict(vcrTodosModelos[["LDA"]], data_test, type = "prob")
cm_test<-confusionMatrix(data=as.factor(ifelse(probs$NOT_PAYER>=0.08,"NOT_PAYER","PAYER")),
                          reference=data_test$TARGET, positive="NOT_PAYER")
```

Warning in confusionMatrix.default(data = as.factor(ifelse(probs\$NOT\_PAYER >= :  
Levels are not in the same order for reference and data. Refactoring data to  
match.

```
cm_test$table
```

	Reference	
Prediction	PAYER	NOT_PAYER
PAYER	30529	1318
NOT_PAYER	15490	3048

```
cm_test$overall[1:2]
```

Accuracy	Kappa
0.6664087	0.1464234

```
cm_test$byClass[1:2]
```

```
Sensitivity Specificity  
0.6981219 0.6633999
```

Curva ROC

Train

```
probs <- predict(vcrTodosModelos[["LDA"]], data_train, type = "prob")  
str(probs)
```

```
'data.frame': 201546 obs. of 2 variables:  
 $ NOT_PAYER: num 0.6281 0.031 0.0268 0.0427 0.1045 ...  
 $ PAYER : num 0.372 0.969 0.973 0.957 0.895 ...
```

```
probs_pos <- probs[, "NOT_PAYER"]  
curvaROC <- roc(data_train$TARGET, probs_pos, levels = c("PAYER", "NOT_PAYER"))
```

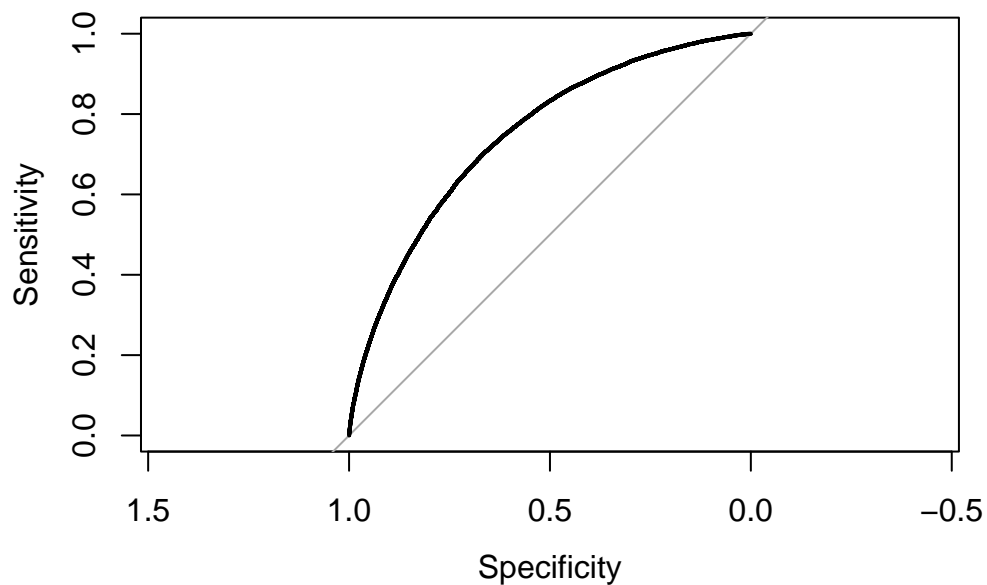
Setting direction: controls < cases

```
curvaROC$auc
```

Area under the curve: 0.7466

```
plot(curvaROC)
```





Test

```
probs <- predict(vcrTodosModelos[["LDA"]], data_test, type = "prob")
str(probs)
```

```
'data.frame':  50385 obs. of  2 variables:
 $ NOT_PAYER: num  0.1033 0.0292 0.1087 0.0326 0.0216 ...
 $ PAYER    : num  0.897 0.971 0.891 0.967 0.978 ...
```

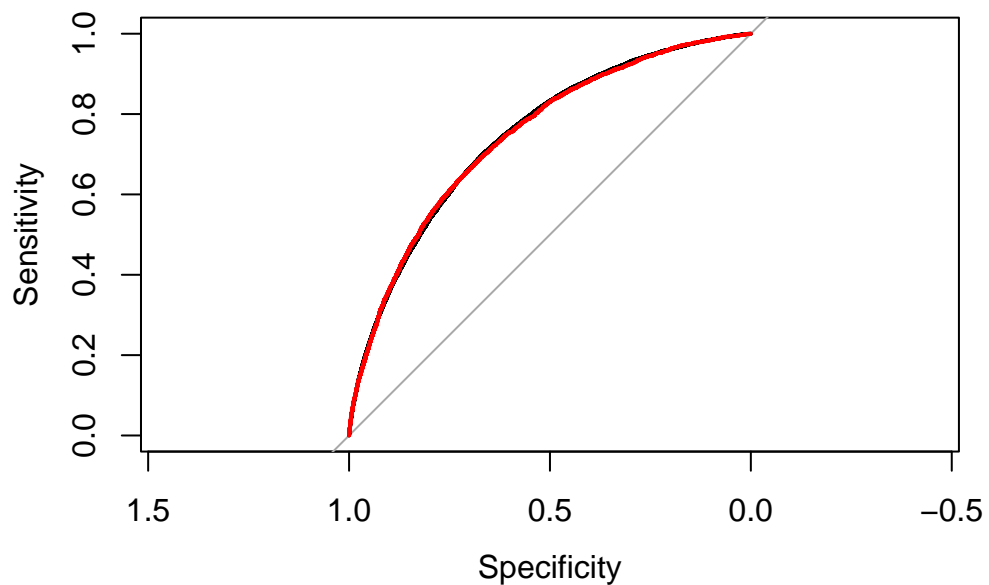
```
probs_pos <- probs[, "NOT_PAYER"]
curvaROC_test <- roc(data_test$TARGET, probs_pos, levels = c("PAYER", "NOT_PAYER"))
```

Setting direction: controls < cases

```
curvaROC$auc
```

Area under the curve: 0.7466

```
plot(curvaROC)
plot(curvaROC_test, add=T, col="red")
```



```
coeffs <- moddis$scaling
importance <- abs(coeffs[,1])
sorted_importance <- sort(importance, decreasing = TRUE)
print(sorted_importance)
```

```
NAME_INCOME_TYPEUnemployed
3.020513e+01
NAME_INCOME_TYEMaternity leave
5.447024e+00
FLAG_DOCUMENT_21
4.345755e+00
EXT_SOURCE_3
2.912404e+00
EXT_SOURCE_2
2.688135e+00
REGION_POPULATION_RELATIVE
2.085941e+00
AGE_GROUP30-40
1.475423e+00
AGE_GROUP40-50
1.395338e+00
AGE_GROUP20-30
```

1.386304e+00  
 NAME\_EDUCATION\_TYPELower secondary  
 1.336682e+00  
 AGE\_GROUP50 above  
 1.264941e+00  
 NAME\_INCOME\_TYPEStudent  
 1.182644e+00  
 EXT\_SOURCE\_1  
 1.149213e+00  
 NAME\_EDUCATION\_TYPESecondary / secondary special  
 1.004006e+00  
 NAME\_INCOME\_TYPEPensioner  
 7.826655e-01  
 NAME\_EDUCATION\_TYPEIncomplete higher  
 7.267212e-01  
 NAME\_EDUCATION\_TYPEHigher education  
 7.067526e-01  
 REGION\_RATING\_CLIENT\_W\_CITY3  
 6.857041e-01  
 AMT\_CREDIT\_RANGE400K-500K  
 5.614867e-01  
 AMT\_CREDIT\_RANGE500K-600K  
 5.559021e-01  
 FLAG\_DOCUMENT\_181  
 5.218544e-01  
 NAME\_TYPE\_SUITEGroup of people  
 5.021683e-01  
 AMT\_CREDIT\_RANGE300K-400K  
 4.808357e-01  
 FLAG\_DOCUMENT\_161  
 4.489856e-01  
 AMT\_CREDIT\_RANGE600K-700K  
 4.410201e-01  
 REGION\_RATING\_CLIENT3  
 4.395788e-01  
 EMPLOYMENT\_YEAR30-40  
 4.043022e-01  
 REGION\_RATING\_CLIENT\_W\_CITY2  
 3.954369e-01  
 AMT\_CREDIT\_RANGE200K-300K  
 3.828897e-01  
 CODE\_GENDERM  
 3.817189e-01

REGION\_RATING\_CLIENT2  
 3.375708e-01  
 FLAG\_OWN\_CARY  
 3.127267e-01  
 EMPLOYMENT\_YEAR40-50  
 3.001720e-01  
 FLOORSMAX\_AVG  
 2.771353e-01  
 NAME\_HOUSING\_TYPEOffice apartment  
 2.618962e-01  
 AMT\_CREDIT  
 2.600076e-01  
 EMPLOYMENT\_YEAR20-30  
 2.583529e-01  
 NAME\_TYPE\_SUITEOther\_B  
 2.474244e-01  
 AMT\_CREDIT\_RANGE100K-200K  
 2.382407e-01  
 AMT\_CREDIT\_RANGE700K-800K  
 2.252928e-01  
 FLAG\_DOCUMENT\_31  
 2.231567e-01  
 FLAG\_DOCUMENT\_51  
 2.210624e-01  
 DEF\_30\_CNT\_SOCIAL\_CIRCLE  
 2.194826e-01  
 NAME\_INCOME\_TYPEWorking  
 2.097342e-01  
 REG\_CITY\_NOT\_LIVE\_CITY1  
 2.030575e-01  
 NAME\_HOUSING\_TYPERented apartment  
 1.859280e-01  
 AMT\_CREDIT\_RANGE800K-900K  
 1.796256e-01  
 FLAG\_WORK\_PHONE1  
 1.743235e-01  
 NAME\_TYPE\_SUITEUnaccompanied  
 1.544677e-01  
 NAME\_HOUSING\_TYPEWith parents  
 1.490864e-01  
 WALLSMATERIAL\_MODEMonolithic  
 1.453681e-01  
 AMT\_CREDIT\_RANGE1M Above

	1.411061e-01
WALLSMATERIAL_MODE	Others
	1.337589e-01
NAME_HOUSING_TYPE	Municipal apartment
	1.179861e-01
AMT_CREDIT_RANGE	900K-1M
	1.151345e-01
FONDKAPREMONT_MODE	reg oper spec account
	1.122304e-01
NAME_INCOME_TYPE	Commercial associate
	1.036708e-01
NAME_TYPE_SUITE	Family
	1.021477e-01
WALLSMATERIAL_MODE	Panel
	9.489312e-02
FONDKAPREMONT_MODE	org spec account
	8.719861e-02
WALLSMATERIAL_MODE	Block
	7.668154e-02
WEEKDAY_APPR_PROCESS_START	MONDAY
	7.262149e-02
EMPLOYMENT_YEAR	5-10
	7.132580e-02
YEARS_EMPLOYED	
	6.839381e-02
WALLSMATERIAL_MODE	Wooden
	6.622466e-02
WEEKDAY_APPR_PROCESS_START	SUNDAY
	6.540212e-02
NAME_TYPE_SUITE	Children
	6.478319e-02
FLAG_PHONE	1
	6.299659e-02
NAME_HOUSING_TYPE	House / apartment
	5.600838e-02
FONDKAPREMONT_MODE	not specified
	5.117367e-02
WEEKDAY_APPR_PROCESS_START	TUESDAY
	3.937959e-02
NAME_TYPE_SUITE	Spouse, partner
	3.732481e-02
WEEKDAY_APPR_PROCESS_START	SATURDAY
	3.692898e-02

NAME_CONTRACT_TYPE	Revolving loans
	3.249683e-02
AMT_REQ_CREDIT_BUREAU_QRT	
	3.243485e-02
REG_CITY_NOT_WORK_CITY	1
	3.100387e-02
APARTMENTS_AVG	
	3.010494e-02
CNT_CHILDREN	
	2.998628e-02
WALLSMATERIAL_MODE	Stone, brick
	2.653276e-02
EMPLOYMENT_YEAR	10-20
	2.310469e-02
FONDKAPREMONT_MODE	reg oper account
	2.120956e-02
WEEKDAY_APPR_PROCESS_START	WEDNESDAY
	1.706198e-02
WEEKDAY_APPR_PROCESS_START	THURSDAY
	1.406181e-02
AMT_INCOME_TOTAL	
	1.286044e-02
NAME_TYPE_SUITE	Other_A
	1.168900e-02
WALLSMATERIAL_MODE	Mixed
	1.080305e-02
NAME_INCOME_TYPE	State servant
	7.643848e-03
OWN_CAR_AGE	
	4.730482e-03
OBS_30_CNT_SOCIAL_CIRCLE	
	4.414556e-03
HOUR_APPR_PROCESS_START	
	3.657671e-03
DAYS_EMPLOYED	
	2.611506e-04
EMPLOYMENT_YEAR	0-5
	1.358216e-04
DAYS_LAST_PHONE_CHANGE	
	7.116356e-05
DAYS_BIRTH	
	1.639156e-05
AMT_ANNUITY	

9.853604e-06  
DAYS\_REGISTRATION  
9.701877e-06  
AMT\_GOODS\_PRICE  
2.866606e-06