

# Project 1 Decision

```
#Load all of the libraries  
library(tidyr)  
library(rpart)
```

Warning: package 'rpart' was built under R version 4.3.1

```
library(rpart.plot)  
library(forecast)
```

Registered S3 method overwritten by 'quantmod':  
method from  
as.zoo.data.frame zoo

```
library(caret)
```

Loading required package: ggplot2

Loading required package: lattice

```
library(ROSE)
```

Loaded ROSE 0.0-4

kNN Model

```

# Load data
data <- read.csv("credit_fa2023_23.csv", header = TRUE)

# Add new fields into data frame to improve model accuracy

for (i in 1:nrow(data)) {
  data$Income_Credit_Ratio[i] <- data$AMT_INCOME_TOTAL[i] / data$AMT_CREDIT[i]
}

for (i in 1:nrow(data)) {
  data$Annuity_Income_Ratio[i] <- data$AMT_ANNUITY[i] / data$AMT_INCOME_TOTAL[i]
}

for (i in 1:nrow(data)) {
  data$Credit_As_Percentage[i] <- data$AMT_CREDIT[i] / data$AMT_INCOME_TOTAL[i]
}

for (i in 1:nrow(data)) {
  data$Percent_Days_Employed[i] <- data$DAYS_EMPLOYED[i] / data$DAYS_BIRTH[i]
}

for (i in 1:nrow(data)) {
  data$Income_Per_Person[i] <- data$AMT_INCOME_TOTAL[i] / data$CNT_FAM_MEMBERS[i]
}

# Remove XNA from CODE_GENDER variable and convert to factor
data <- data[data$CODE_GENDER != "XNA", ]

data$CODE_GENDER <- factor(data$CODE_GENDER)

# Explore data
names(data)

```

[1] "X"	"SK_ID_CURR"
[3] "TARGET"	"NAME_CONTRACT_TYPE"
[5] "CODE_GENDER"	"FLAG_OWN_CAR"
[7] "FLAG_OWN_REALTY"	"CNT_CHILDREN"
[9] "AMT_INCOME_TOTAL"	"AMT_CREDIT"
[11] "AMT_ANNUITY"	"AMT_GOODS_PRICE"
[13] "NAME_TYPE_SUITE"	"NAME_INCOME_TYPE"
[15] "NAME_EDUCATION_TYPE"	"NAME_FAMILY_STATUS"
[17] "NAME_HOUSING_TYPE"	"DAYS_BIRTH"

[19]	"DAYS_EMPLOYED"	"DAYS_REGISTRATION"
[21]	"DAYS_ID_PUBLISH"	"OWN_CAR_AGE"
[23]	"FLAG_MOBIL"	"FLAG_EMP_PHONE"
[25]	"FLAG_WORK_PHONE"	"FLAG_CONT_MOBILE"
[27]	"FLAG_PHONE"	"FLAG_EMAIL"
[29]	"OCCUPATION_TYPE"	"CNT_FAM_MEMBERS"
[31]	"REGION_RATING_CLIENT"	"REGION_RATING_CLIENT_W_CITY"
[33]	"WEEKDAY_APPR_PROCESS_START"	"HOUR_APPR_PROCESS_START"
[35]	"REG_REGION_NOT_LIVE_REGION"	"REG_REGION_NOT_WORK_REGION"
[37]	"LIVE_REGION_NOT_WORK_REGION"	"REG_CITY_NOT_LIVE_CITY"
[39]	"REG_CITY_NOT_WORK_CITY"	"LIVE_CITY_NOT_WORK_CITY"
[41]	"ORGANIZATION_TYPE"	"DAYS_LAST_PHONE_CHANGE"
[43]	"FLAG_DOCUMENT_2"	"FLAG_DOCUMENT_3"
[45]	"FLAG_DOCUMENT_4"	"FLAG_DOCUMENT_5"
[47]	"FLAG_DOCUMENT_6"	"FLAG_DOCUMENT_7"
[49]	"FLAG_DOCUMENT_8"	"FLAG_DOCUMENT_9"
[51]	"FLAG_DOCUMENT_10"	"FLAG_DOCUMENT_11"
[53]	"FLAG_DOCUMENT_12"	"FLAG_DOCUMENT_13"
[55]	"FLAG_DOCUMENT_14"	"FLAG_DOCUMENT_15"
[57]	"FLAG_DOCUMENT_16"	"FLAG_DOCUMENT_17"
[59]	"FLAG_DOCUMENT_18"	"FLAG_DOCUMENT_19"
[61]	"FLAG_DOCUMENT_20"	"FLAG_DOCUMENT_21"
[63]	"AMT_REQ_CREDIT_BUREAU_HOUR"	"AMT_REQ_CREDIT_BUREAU_DAY"
[65]	"AMT_REQ_CREDIT_BUREAU_WEEK"	"AMT_REQ_CREDIT_BUREAU_MON"
[67]	"AMT_REQ_CREDIT_BUREAU_QRT"	"AMT_REQ_CREDIT_BUREAU_YEAR"
[69]	"Income_Credit_Ratio"	"Annuity_Income_Ratio"
[71]	"Credit_As_Percentage"	"Percent_Days_Employed"
[73]	"Income_Per_Person"	

```
str(data)
```

```
'data.frame': 29999 obs. of 73 variables:
 $ X : int 300440 217645 70440 300551 86881 146804 263212 70439 11...
 $ SK_ID_CURR : int 448070 352176 181716 448195 200835 270206 404772 181715...
 $ TARGET : int 0 0 1 0 1 0 0 1 1 0 ...
 $ NAME_CONTRACT_TYPE : chr "Cash loans" "Cash loans" "Cash loans" "Cash loans" ...
 $ CODE_GENDER : Factor w/ 2 levels "F","M": 1 1 2 1 2 1 2 1 2 1 ...
 $ FLAG_OWN_CAR : chr "N" "N" "N" "N" ...
 $ FLAG_OWN_REALTY : chr "N" "Y" "Y" "Y" ...
 $ CNT_CHILDREN : int 0 0 0 0 0 1 0 0 0 1 ...
 $ AMT_INCOME_TOTAL : num 157500 90000 180000 171000 135000 ...
```

\$ AMT_CREDIT	: num	640080 573628 292500 757598 381528 ...
\$ AMT_ANNUITY	: num	29970 22878 34844 40491 25628 ...
\$ AMT_GOODS_PRICE	: num	450000 463500 292500 702000 315000 ...
\$ NAME_TYPE_SUITE	: chr	"Unaccompanied" "Children" "Unaccompanied" "Unaccompanied"
\$ NAME_INCOME_TYPE	: chr	"Commercial associate" "Working" "Working" "State servan"
\$ NAME_EDUCATION_TYPE	: chr	"Secondary / secondary special" "Secondary / secondary"
\$ NAME_FAMILY_STATUS	: chr	"Separated" "Widow" "Civil marriage" "Married" ...
\$ NAME_HOUSING_TYPE	: chr	"With parents" "House / apartment" "House / apartment"
\$ DAYS_BIRTH	: int	-10953 -20075 -13898 -21445 -10240 -13857 -21167 -17146
\$ DAYS_EMPLOYED	: int	-3005 -1715 -539 -4657 -921 -3113 -3320 -1509 -295 3652
\$ DAYS_REGISTRATION	: int	-5485 -1409 -2070 -3980 -1113 -7952 -2434 -8268 -205 -4
\$ DAYS_ID_PUBLISH	: int	-1284 -3573 -258 -4154 -123 -4604 -3253 -695 -1315 -446
\$ OWN_CAR_AGE	: int	NA NA NA NA 2 NA NA NA NA NA ...
\$ FLAG_MOBIL	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ FLAG_EMP_PHONE	: int	1 1 1 1 1 1 1 1 1 0 ...
\$ FLAG_WORK_PHONE	: int	0 0 0 0 0 0 0 1 0 0 ...
\$ FLAG_CONT_MOBILE	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ FLAG_PHONE	: int	0 1 0 0 0 0 1 1 0 0 ...
\$ FLAG_EMAIL	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ OCCUPATION_TYPE	: chr	" " "Cooking staff" "Laborers" "Core staff" ...
\$ CNT_FAM_MEMBERS	: int	1 1 2 2 1 3 2 1 2 3 ...
\$ REGION_RATING_CLIENT	: int	2 2 2 2 3 1 2 3 3 2 ...
\$ REGION_RATING_CLIENT_W_CITY	: int	2 2 2 2 3 1 2 3 3 2 ...
\$ WEEKDAY_APPR_PROCESS_START	: chr	"SATURDAY" "FRIDAY" "TUESDAY" "SATURDAY" ...
\$ HOUR_APPR_PROCESS_START	: int	13 11 8 11 17 18 12 11 10 11 ...
\$ REG_REGION_NOT_LIVE_REGION	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ REG_REGION_NOT_WORK_REGION	: int	0 0 1 0 0 0 0 0 0 0 ...
\$ LIVE_REGION_NOT_WORK_REGION	: int	0 0 1 0 0 0 0 0 0 0 ...
\$ REG_CITY_NOT_LIVE_CITY	: int	1 0 0 0 1 0 0 0 0 0 ...
\$ REG_CITY_NOT_WORK_CITY	: int	1 0 0 0 1 0 0 0 1 0 ...
\$ LIVE_CITY_NOT_WORK_CITY	: int	0 0 0 0 0 0 0 0 1 0 ...
\$ ORGANIZATION_TYPE	: chr	"Self-employed" "Hotel" "Business Entity Type 1" "School"
\$ DAYS_LAST_PHONE_CHANGE	: int	-2411 -1513 0 -2778 -20 -1827 -1471 -657 0 -796 ...
\$ FLAG_DOCUMENT_2	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_3	: int	1 1 1 1 1 1 1 1 1 0 ...
\$ FLAG_DOCUMENT_4	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_5	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_6	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_7	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_8	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_9	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_10	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_11	: int	0 0 0 0 0 0 0 0 0 0 ...

```

$ FLAG_DOCUMENT_12      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_13      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_14      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_15      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_16      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_17      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_18      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_19      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_20      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_21      : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_HOUR : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_DAY  : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_WEEK : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_MON  : int  0 0 1 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_QRT  : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_YEAR : int  0 4 2 3 1 1 5 2 3 6 ...
$ Income_Credit_Ratio      : num  0.246 0.157 0.615 0.226 0.354 ...
$ Annuity_Income_Ratio     : num  0.19 0.254 0.194 0.237 0.19 ...
$ Credit_As_Percentage     : num  4.06 6.37 1.62 4.43 2.83 ...
$ Percent_Days_Employed    : num  0.2744 0.0854 0.0388 0.2172 0.0899 ...
$ Income_Per_Person        : num  157500 90000 90000 85500 135000 ...

```

```
summary(data$TARGET)
```

```

      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
0.0000   0.0000   0.0000  0.1951  0.0000   1.0000

```

```

# Convert education type to factor with levels across education
data$NAME_EDUCATION_TYPE <- factor(data$NAME_EDUCATION_TYPE, levels = c(
  "Secondary / secondary special",
  "Higher education",
  "Lower secondary",
  "Incomplete higher",
  "Academic degree"))

```

```

# Set Target variable as factor
data$TARGET <- as.factor(data$TARGET)

```

```

# Variable list
# Percent_Days_Employed, NAME_EDUCATION_TYPE, REGION_RATING_CLIENT_W_CITY, AMT_GOODS_PRICE

```

```
# Remove unused variables
data <- data[ , -c(1:2, 4, 6:9, 13:14, 16:17, 22:31, 33:69, 71, 73)]
names(data)
```

```
[1] "TARGET"                                "CODE_GENDER"
[3] "AMT_CREDIT"                            "AMT_ANNUITY"
[5] "AMT_GOODS_PRICE"                      "NAME_EDUCATION_TYPE"
[7] "DAYS_BIRTH"                           "DAYS_EMPLOYED"
[9] "DAYS_REGISTRATION"                    "DAYS_ID_PUBLISH"
[11] "REGION_RATING_CLIENT_W_CITY" "Annuity_Income_Ratio"
[13] "Percent_Days_Employed"
```

```
# Training - Validation split
set.seed(666)
train_index <- sample(1:nrow(data), 0.7 * nrow(data))
valid_index <- setdiff(1:nrow(data), train_index)
train_df <- data[train_index, ]
valid_df <- data[valid_index, ]

# Double check
nrow(train_df)
```

```
[1] 20999
```

```
nrow(valid_df)
```

```
[1] 9000
```

```
head(train_df)
```

	TARGET	CODE_GENDER	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE
17983	1	F	467257.5	17743.5	328500
12926	0	F	260640.0	31059.0	225000
13195	0	F	855882.0	36391.5	765000
23676	0	M	266832.0	24601.5	238500
15901	1	F	360000.0	10102.5	360000
873	0	F	157500.0	14575.5	157500

	NAME_EDUCATION_TYPE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION
17983	Higher education	-10631	-200	-4795
12926	Secondary / secondary special	-17424	-10763	-9653
13195	Secondary / secondary special	-13370	-3493	-6795
23676	Secondary / secondary special	-10606	-383	-5120
15901	Higher education	-14135	-1400	-7982
873	Higher education	-11931	-1154	-10268

	DAYS_ID_PUBLISH	REGION_RATING_CLIENT_W_CITY	Annuity_Income_Ratio
17983	-3303	2	0.14082143
12926	-969	2	0.16051163
13195	-4908	2	0.10782667
23676	-3208	2	0.14775676
15901	-3265	2	0.10204545
873	-788	2	0.09254286

	Percent_Days_Employed
17983	0.01881291
12926	0.61771120
13195	0.26125654
23676	0.03611163
15901	0.09904492
873	0.09672282

```
head(valid_df)
```

	TARGET	CODE_GENDER	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE
2	0	F	573628.5	22878.0	463500
3	1	M	292500.0	34843.5	292500
6	0	F	1198548.0	50913.0	1102500
9	1	M	545040.0	26640.0	450000
10	0	F	337500.0	16875.0	337500
12	1	F	312768.0	17095.5	270000

	NAME_EDUCATION_TYPE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION
2	Secondary / secondary special	-20075	-1715	-1409
3	Secondary / secondary special	-13898	-539	-2070
6	Higher education	-13857	-3113	-7952
9	Secondary / secondary special	-18234	-295	-205
10	Secondary / secondary special	-16897	365243	-4399
12	Secondary / secondary special	-12185	-2602	-4412

	DAYS_ID_PUBLISH	REGION_RATING_CLIENT_W_CITY	Annuity_Income_Ratio
2	-3573	2	0.2542000
3	-258	2	0.1935750

6	-4604	1	0.2057091
9	-1315	3	0.1315556
10	-446	2	0.1500000
12	-3509	2	0.2532667
Percent_Days_Employed			
2	0.08542964		
3	0.03878256		
6	0.22465180		
9	0.01617857		
10	-21.61584897		
12	0.21354124		

```
str(train_df)
```

```
'data.frame': 20999 obs. of 13 variables:
 $ TARGET          : Factor w/ 2 levels "0","1": 2 1 1 1 2 1 1 2 1 1 ...
 $ CODE_GENDER     : Factor w/ 2 levels "F","M": 1 1 1 2 1 1 2 2 2 1 ...
 $ AMT_CREDIT      : num  467258 260640 855882 266832 360000 ...
 $ AMT_ANNUITY     : num  17744 31059 36392 24602 10102 ...
 $ AMT_GOODS_PRICE : num  328500 225000 765000 238500 360000 ...
 $ NAME_EDUCATION_TYPE : Factor w/ 5 levels "Secondary / secondary special",...: 2 1 1 ...
 $ DAYS_BIRTH      : int   -10631 -17424 -13370 -10606 -14135 -11931 -18336 -12296 ...
 $ DAYS_EMPLOYED   : int    -200 -10763 -3493 -383 -1400 -1154 -6561 -1212 -2285 -6 ...
 $ DAYS_REGISTRATION : int   -4795 -9653 -6795 -5120 -7982 -10268 -1038 -2417 -3801 - ...
 $ DAYS_ID_PUBLISH : int   -3303 -969 -4908 -3208 -3265 -788 -1898 -2422 -4156 -47 ...
 $ REGION_RATING_CLIENT_W_CITY: int    2 2 2 2 2 2 2 2 2 2 ...
 $ Annuity_Income_Ratio : num   0.141 0.161 0.108 0.148 0.102 ...
 $ Percent_Days_Employed : num   0.0188 0.6177 0.2613 0.0361 0.099 ...
```

```
str(valid_df)
```

```
'data.frame': 9000 obs. of 13 variables:
 $ TARGET          : Factor w/ 2 levels "0","1": 1 2 1 2 1 2 1 2 1 1 ...
 $ CODE_GENDER     : Factor w/ 2 levels "F","M": 1 2 1 2 1 1 2 1 1 1 ...
 $ AMT_CREDIT      : num   573628 292500 1198548 545040 337500 ...
 $ AMT_ANNUITY     : num    22878 34844 50913 26640 16875 ...
 $ AMT_GOODS_PRICE : num   463500 292500 1102500 450000 337500 ...
 $ NAME_EDUCATION_TYPE : Factor w/ 5 levels "Secondary / secondary special",...: 1 1 2 ...
 $ DAYS_BIRTH      : int   -20075 -13898 -13857 -18234 -16897 -12185 -10579 -11938 ...
 $ DAYS_EMPLOYED   : int   -1715 -539 -3113 -295 365243 -2602 -246 -680 365243 -15 ...
```



```

$ DAYS_REGISTRATION      : int   -1409 -2070 -7952 -205 -4399 -4412 -4665 -5949 -4764 -4764 -4764 ...
$ DAYS_ID_PUBLISH        : int   -3573 -258 -4604 -1315 -446 -3509 -3226 -4153 -82 -2312 ...
$ REGION_RATING_CLIENT_W_CITY: int    2  2  1  3  2  2  2  2  2  2  2 ...
$ Annuity_Income_Ratio    : num    0.254 0.194 0.206 0.132 0.15 ...
$ Percent_Days_Employed   : num    0.0854 0.0388 0.2247 0.0162 -21.6158 ...

```

```

# Use ROSE to to balance model
train_df_rose <- ROSE(TARGET ~ Percent_Days_Employed + NAME_EDUCATION_TYPE + REGION_RATING_CLIENT_W_CITY,
                      data = train_df, seed = 666)$data

table(train_df_rose$TARGET)

```

```

      0      1
10337 10640

```

```

# Check variables
names(data)

```

```

[1] "TARGET"                "CODE_GENDER"
[3] "AMT_CREDIT"             "AMT_ANNUITY"
[5] "AMT_GOODS_PRICE"        "NAME_EDUCATION_TYPE"
[7] "DAYS_BIRTH"             "DAYS_EMPLOYED"
[9] "DAYS_REGISTRATION"      "DAYS_ID_PUBLISH"
[11] "REGION_RATING_CLIENT_W_CITY" "Annuity_Income_Ratio"
[13] "Percent_Days_Employed"

```

```

# Normalization algorithm
train_norm <- train_df_rose
valid_norm <- valid_df

```

```

names(train_df)

```

```

[1] "TARGET"                "CODE_GENDER"
[3] "AMT_CREDIT"             "AMT_ANNUITY"
[5] "AMT_GOODS_PRICE"        "NAME_EDUCATION_TYPE"
[7] "DAYS_BIRTH"             "DAYS_EMPLOYED"
[9] "DAYS_REGISTRATION"      "DAYS_ID_PUBLISH"
[11] "REGION_RATING_CLIENT_W_CITY" "Annuity_Income_Ratio"
[13] "Percent_Days_Employed"

```

```

norm_values <- preProcess(train_df_rose[, -c(1)],
                           method = c("center",
                                       "scale"))
train_norm[, -c(1)] <- predict(norm_values,
                               train_df_rose[, -c(1)])

head(train_norm)

```

	TARGET	CODE_GENDER	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE
1	0	F	0.99268089	1.5528871	1.0235252
2	0	F	-0.55697501	-1.7304333	-0.8443363
3	0	F	1.72159500	0.8693993	1.2456352
4	0	M	-1.62704136	-1.0453429	-0.6409312
5	0	F	0.02538308	-0.9634157	-0.7529268
6	0	M	0.79678441	0.2707055	0.3107206

	NAME_EDUCATION_TYPE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION
1	Secondary / secondary special	0.07110103	-0.6251980	1.0871392
2	Incomplete higher	1.93744114	-1.4197362	-0.5301268
3	Lower secondary	-1.64281333	1.8690287	0.7773543
4	Secondary / secondary special	1.00341927	0.6162391	-0.4667237
5	Secondary / secondary special	0.51491802	-0.2473539	0.8483135
6	Secondary / secondary special	1.67805126	-0.6616488	0.5031771

	DAYS_ID_PUBLISH	REGION_RATING_CLIENT_W_CITY	Annuity_Income_Ratio
1	-0.63250269	-1.919067483	-0.09484436
2	0.77819545	-0.330154347	-0.81335026
3	-1.80369065	0.045849235	0.97555523
4	-0.04891247	-0.171642515	-1.10781415
5	-0.35311456	-0.009725689	-0.19377584
6	0.80158561	-1.541164503	-0.29677068

	Percent_Days_Employed
1	0.95414420
2	-0.13951258
3	-2.57417168
4	0.40890915
5	0.13264204
6	0.06014959

```

# Apply to validation set
valid_norm[, -c(1)] <- predict(norm_values,
                               valid_df[, -c(1)])

```

```
head(valid_norm)
```

	TARGET	CODE_GENDER	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE
2	0	F	-0.02155441	-0.27565214	-0.1405352
3	1	M	-0.69261860	0.52747947	-0.5902807
6	0	F	1.47015169	1.60607405	1.5400928
9	1	M	-0.08979622	-0.02314442	-0.1760414
10	0	F	-0.58520193	-0.67857714	-0.4719267
12	1	F	-0.64423813	-0.66377704	-0.6494578

	NAME_EDUCATION_TYPE	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION
2	Secondary / secondary special	-0.9462974	-0.3838620	0.87385304
3	Secondary / secondary special	0.3392137	-0.3759077	0.69759733
6	Higher education	0.3477464	-0.3933178	-0.87083853
9	Secondary / secondary special	-0.5631622	-0.3742573	1.19489975
10	Secondary / secondary special	-0.2849158	2.0981828	0.07656926
12	Secondary / secondary special	0.6957105	-0.3898615	0.07310280

	DAYS_ID_PUBLISH	REGION_RATING_CLIENT_W_CITY	Annuity_Income_Ratio
2	-0.4122812	-0.1450661	0.64784698
3	1.5503652	-0.1450661	0.07502373
6	-1.0226850	-1.9364656	0.18967428
9	0.9245682	1.6463335	-0.51097511
10	1.4390598	-0.1450661	-0.33670036
12	-0.3743900	-0.1450661	0.63902826

	Percent_Days_Employed
2	0.3728429
3	0.3661824
6	0.3927217
9	0.3629550
10	-2.7257592
12	0.3911352

```
# drop missing values
valid_norm <- drop_na(valid_norm)
```

```
# Train kNN model using k = 5
knn_model <- caret::knn3(TARGET ~ ., data = train_norm, k = 5)
```

```
# Prediction on training set
knn_pred_train <- predict(knn_model, newdata = train_norm[, -c(1)],
                          type = "class")
```

```
head(knn_pred_train)
```

```
[1] 0 0 0 1 0 0  
Levels: 0 1
```

```
# Prediction on validation set  
knn_pred_valid <- predict(knn_model, newdata = valid_norm[, -c(1)],  
                           type = "class")  
head(knn_pred_valid)
```

```
[1] 1 1 0 1 1 1  
Levels: 0 1
```

```
# Confusion matrix on training set  
confusionMatrix(knn_pred_train, as.factor(train_norm[, 1]),  
                 positive = "1")
```

#### Confusion Matrix and Statistics

	Reference	
Prediction	0	1
0	7120	2257
1	3217	8383

Accuracy : 0.739  
95% CI : (0.733, 0.745)  
No Information Rate : 0.5072  
P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.4773

McNemar's Test P-Value : < 2.2e-16

Sensitivity : 0.7879  
Specificity : 0.6888  
Pos Pred Value : 0.7227  
Neg Pred Value : 0.7593  
Prevalence : 0.5072  
Detection Rate : 0.3996

Detection Prevalence : 0.5530  
Balanced Accuracy : 0.7383

'Positive' Class : 1

```
# Confusion matrix on validation set
confusionMatrix(knn_pred_valid, as.factor(valid_norm[, 1]),
                positive = "1")
```

#### Confusion Matrix and Statistics

	Reference	
Prediction	0	1
0	3601	631
1	3667	1094

Accuracy : 0.5221  
95% CI : (0.5117, 0.5324)  
No Information Rate : 0.8082  
P-Value [Acc > NIR] : 1

Kappa : 0.0776

McNemar's Test P-Value : <2e-16

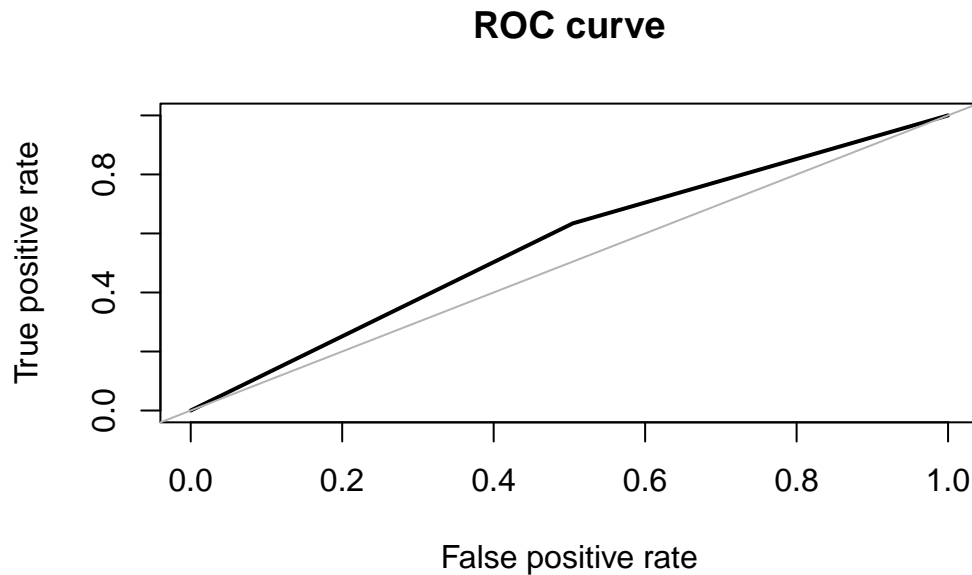
Sensitivity : 0.6342  
Specificity : 0.4955  
Pos Pred Value : 0.2298  
Neg Pred Value : 0.8509  
Prevalence : 0.1918  
Detection Rate : 0.1217  
Detection Prevalence : 0.5294  
Balanced Accuracy : 0.5648

'Positive' Class : 1

#### Model Evaluation

```
library(ROSE)
```

```
ROSE::roc.curve(valid_norm$TARGET, knn_pred_valid)
```



Area under the curve (AUC): 0.565

#### Decision Tree Model

```
# Load data for decision tree model and add new fields again
data1 <- read.csv("credit_fa2023_23.csv", header = TRUE)
data1$Income_Credit_Ratio <- NA

for (i in 1:nrow(data1)) {
  data1$Income_Credit_Ratio[i] <- data1$AMT_INCOME_TOTAL[i] / data1$AMT_CREDIT[i]
}

for (i in 1:nrow(data1)) {
  data1$Annuity_Income_Ratio[i] <- data1$AMT_ANNUITY[i] / data1$AMT_INCOME_TOTAL[i]
}

for (i in 1:nrow(data1)) {
```

```

    data1$Credit_As_Percentage[i] <- data1$AMT_CREDIT[i] / data1$AMT_INCOME_TOTAL[i]
  }

  for (i in 1:nrow(data1)) {
    data1$Percent_Days_Employed[i] <- data1$DAYS_EMPLOYED[i] / data1$DAYS_BIRTH[i]
  }

  for (i in 1:nrow(data1)) {
    data1$Income_Per_Person[i] <- data1$AMT_INCOME_TOTAL[i] / data1$CNT_FAM_MEMBERS[i]
  }

  #Remove XNA from CODE_GENDER variable
  data1 <- data1[data1$CODE_GENDER != "XNA", ]

  #Save modified data into new data variable. Investigate data
  data <- data1
  names(data)

```

[1] "X"	"SK_ID_CURR"
[3] "TARGET"	"NAME_CONTRACT_TYPE"
[5] "CODE_GENDER"	"FLAG_OWN_CAR"
[7] "FLAG_OWN_REALTY"	"CNT_CHILDREN"
[9] "AMT_INCOME_TOTAL"	"AMT_CREDIT"
[11] "AMT_ANNUITY"	"AMT_GOODS_PRICE"
[13] "NAME_TYPE_SUITE"	"NAME_INCOME_TYPE"
[15] "NAME_EDUCATION_TYPE"	"NAME_FAMILY_STATUS"
[17] "NAME_HOUSING_TYPE"	"DAYS_BIRTH"
[19] "DAYS_EMPLOYED"	"DAYS_REGISTRATION"
[21] "DAYS_ID_PUBLISH"	"OWN_CAR_AGE"
[23] "FLAG_MOBIL"	"FLAG_EMP_PHONE"
[25] "FLAG_WORK_PHONE"	"FLAG_CONT_MOBILE"
[27] "FLAG_PHONE"	"FLAG_EMAIL"
[29] "OCCUPATION_TYPE"	"CNT_FAM_MEMBERS"
[31] "REGION_RATING_CLIENT"	"REGION_RATING_CLIENT_W_CITY"
[33] "WEEKDAY_APPR_PROCESS_START"	"HOUR_APPR_PROCESS_START"
[35] "REG_REGION_NOT_LIVE_REGION"	"REG_REGION_NOT_WORK_REGION"
[37] "LIVE_REGION_NOT_WORK_REGION"	"REG_CITY_NOT_LIVE_CITY"
[39] "REG_CITY_NOT_WORK_CITY"	"LIVE_CITY_NOT_WORK_CITY"
[41] "ORGANIZATION_TYPE"	"DAYS_LAST_PHONE_CHANGE"
[43] "FLAG_DOCUMENT_2"	"FLAG_DOCUMENT_3"
[45] "FLAG_DOCUMENT_4"	"FLAG_DOCUMENT_5"

[47] "FLAG_DOCUMENT_6"	"FLAG_DOCUMENT_7"
[49] "FLAG_DOCUMENT_8"	"FLAG_DOCUMENT_9"
[51] "FLAG_DOCUMENT_10"	"FLAG_DOCUMENT_11"
[53] "FLAG_DOCUMENT_12"	"FLAG_DOCUMENT_13"
[55] "FLAG_DOCUMENT_14"	"FLAG_DOCUMENT_15"
[57] "FLAG_DOCUMENT_16"	"FLAG_DOCUMENT_17"
[59] "FLAG_DOCUMENT_18"	"FLAG_DOCUMENT_19"
[61] "FLAG_DOCUMENT_20"	"FLAG_DOCUMENT_21"
[63] "AMT_REQ_CREDIT_BUREAU_HOUR"	"AMT_REQ_CREDIT_BUREAU_DAY"
[65] "AMT_REQ_CREDIT_BUREAU_WEEK"	"AMT_REQ_CREDIT_BUREAU_MON"
[67] "AMT_REQ_CREDIT_BUREAU_QRT"	"AMT_REQ_CREDIT_BUREAU_YEAR"
[69] "Income_Credit_Ratio"	"Annuity_Income_Ratio"
[71] "Credit_As_Percentage"	"Percent_Days_Employed"
[73] "Income_Per_Person"	

```
str(data)
```

```
'data.frame': 29999 obs. of 73 variables:
 $ X                : int  300440 217645 70440 300551 86881 146804 263212 70439 11...
 $ SK_ID_CURR       : int  448070 352176 181716 448195 200835 270206 404772 181715...
 $ TARGET           : int  0 0 1 0 1 0 0 1 1 0 ...
 $ NAME_CONTRACT_TYPE : chr  "Cash loans" "Cash loans" "Cash loans" "Cash loans" ...
 $ CODE_GENDER      : chr  "F" "F" "M" "F" ...
 $ FLAG_OWN_CAR     : chr  "N" "N" "N" "N" ...
 $ FLAG_OWN_REALTY  : chr  "N" "Y" "Y" "Y" ...
 $ CNT_CHILDREN     : int  0 0 0 0 0 1 0 0 0 1 ...
 $ AMT_INCOME_TOTAL : num  157500 90000 180000 171000 135000 ...
 $ AMT_CREDIT       : num  640080 573628 292500 757598 381528 ...
 $ AMT_ANNUITY      : num  29970 22878 34844 40491 25628 ...
 $ AMT_GOODS_PRICE  : num  450000 463500 292500 702000 315000 ...
 $ NAME_TYPE_SUITE  : chr  "Unaccompanied" "Children" "Unaccompanied" "Unaccompanied" ...
 $ NAME_INCOME_TYPE : chr  "Commercial associate" "Working" "Working" "State servant" ...
 $ NAME_EDUCATION_TYPE : chr  "Secondary / secondary special" "Secondary / secondary" ...
 $ NAME_FAMILY_STATUS : chr  "Separated" "Widow" "Civil marriage" "Married" ...
 $ NAME_HOUSING_TYPE : chr  "With parents" "House / apartment" "House / apartment" ...
 $ DAYS_BIRTH       : int  -10953 -20075 -13898 -21445 -10240 -13857 -21167 -17146 ...
 $ DAYS_EMPLOYED    : int  -3005 -1715 -539 -4657 -921 -3113 -3320 -1509 -295 3652 ...
 $ DAYS_REGISTRATION : int  -5485 -1409 -2070 -3980 -1113 -7952 -2434 -8268 -205 -4...
 $ DAYS_ID_PUBLISH  : int  -1284 -3573 -258 -4154 -123 -4604 -3253 -695 -1315 -446 ...
 $ OWN_CAR_AGE      : int  NA NA NA NA 2 NA NA NA NA NA ...
 $ FLAG_MOBIL       : int  1 1 1 1 1 1 1 1 1 1 ...
```



```

$ FLAG_EMP_PHONE           : int  1 1 1 1 1 1 1 1 1 0 ...
$ FLAG_WORK_PHONE         : int  0 0 0 0 0 0 0 1 0 0 ...
$ FLAG_CONT_MOBILE        : int  1 1 1 1 1 1 1 1 1 1 ...
$ FLAG_PHONE              : int  0 1 0 0 0 0 1 1 0 0 ...
$ FLAG_EMAIL              : int  0 0 0 0 0 0 0 0 0 0 ...
$ OCCUPATION_TYPE         : chr   "" "Cooking staff" "Laborers" "Core staff" ...
$ CNT_FAM_MEMBERS         : int  1 1 2 2 1 3 2 1 2 3 ...
$ REGION_RATING_CLIENT    : int  2 2 2 2 3 1 2 3 3 2 ...
$ REGION_RATING_CLIENT_W_CITY: int  2 2 2 2 3 1 2 3 3 2 ...
$ WEEKDAY_APPR_PROCESS_START : chr  "SATURDAY" "FRIDAY" "TUESDAY" "SATURDAY" ...
$ HOUR_APPR_PROCESS_START  : int  13 11 8 11 17 18 12 11 10 11 ...
$ REG_REGION_NOT_LIVE_REGION : int  0 0 0 0 0 0 0 0 0 0 ...
$ REG_REGION_NOT_WORK_REGION : int  0 0 1 0 0 0 0 0 0 0 ...
$ LIVE_REGION_NOT_WORK_REGION: int  0 0 1 0 0 0 0 0 0 0 ...
$ REG_CITY_NOT_LIVE_CITY   : int  1 0 0 0 1 0 0 0 0 0 ...
$ REG_CITY_NOT_WORK_CITY   : int  1 0 0 0 1 0 0 0 1 0 ...
$ LIVE_CITY_NOT_WORK_CITY  : int  0 0 0 0 0 0 0 0 1 0 ...
$ ORGANIZATION_TYPE       : chr   "Self-employed" "Hotel" "Business Entity Type 1" "School" ...
$ DAYS_LAST_PHONE_CHANGE   : int  -2411 -1513 0 -2778 -20 -1827 -1471 -657 0 -796 ...
$ FLAG_DOCUMENT_2         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_3         : int  1 1 1 1 1 1 1 1 1 0 ...
$ FLAG_DOCUMENT_4         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_5         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_6         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_7         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_8         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_9         : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_10        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_11        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_12        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_13        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_14        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_15        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_16        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_17        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_18        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_19        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_20        : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_21        : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_HOUR : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_DAY  : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_WEEK : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_MON  : int  0 0 1 0 0 0 0 0 0 0 ...

```

```

$ AMT_REQ_CREDIT_BUREAU_QRT : int 0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_YEAR : int 0 4 2 3 1 1 5 2 3 6 ...
$ Income_Credit_Ratio       : num 0.246 0.157 0.615 0.226 0.354 ...
$ Annuity_Income_Ratio      : num 0.19 0.254 0.194 0.237 0.19 ...
$ Credit_As_Percentage      : num 4.06 6.37 1.62 4.43 2.83 ...
$ Percent_Days_Employed     : num 0.2744 0.0854 0.0388 0.2172 0.0899 ...
$ Income_Per_Person         : num 157500 90000 90000 85500 135000 ...

```

```
head(data)
```

```

      X SK_ID_CURR TARGET NAME_CONTRACT_TYPE CODE_GENDER FLAG_OWN_CAR
1 300440    448070      0      Cash loans           F           N
2 217645    352176      0      Cash loans           F           N
3  70440    181716      1      Cash loans           M           N
4 300551    448195      0      Cash loans           F           N
5  86881    200835      1      Cash loans           M           Y
6 146804    270206      0      Cash loans           F           N
  FLAG_OWN_REALTY CNT_CHILDREN AMT_INCOME_TOTAL AMT_CREDIT AMT_ANNUITY
1              N              0          157500    640080.0    29970.0
2              Y              0           90000    573628.5    22878.0
3              Y              0          180000    292500.0    34843.5
4              Y              0          171000    757597.5    40491.0
5              Y              0          135000    381528.0    25627.5
6              Y              1          247500   1198548.0    50913.0
  AMT_GOODS_PRICE NAME_TYPE_SUITE NAME_INCOME_TYPE
1         450000  Unaccompanied Commercial associate
2         463500      Children           Working
3         292500  Unaccompanied           Working
4         702000  Unaccompanied State servant
5         315000  Unaccompanied           Working
6        1102500  Unaccompanied Commercial associate
  NAME_EDUCATION_TYPE NAME_FAMILY_STATUS NAME_HOUSING_TYPE
1 Secondary / secondary special      Separated      With parents
2 Secondary / secondary special      Widow House / apartment
3 Secondary / secondary special Civil marriage House / apartment
4 Secondary / secondary special      Married House / apartment
5 Secondary / secondary special Single / not married House / apartment
6           Higher education Civil marriage House / apartment
  DAYS_BIRTH DAYS_EMPLOYED DAYS_REGISTRATION DAYS_ID_PUBLISH OWN_CAR_AGE
1     -10953       -3005       -5485       -1284          NA
2     -20075       -1715       -1409       -3573          NA

```

3	-13898	-539	-2070	-258	NA
4	-21445	-4657	-3980	-4154	NA
5	-10240	-921	-1113	-123	2
6	-13857	-3113	-7952	-4604	NA
	FLAG_MOBIL	FLAG_EMP_PHONE	FLAG_WORK_PHONE	FLAG_CONT_MOBILE	FLAG_PHONE
1	1	1	0	1	0
2	1	1	0	1	1
3	1	1	0	1	0
4	1	1	0	1	0
5	1	1	0	1	0
6	1	1	0	1	0
	FLAG_EMAIL	OCCUPATION_TYPE	CNT_FAM_MEMBERS	REGION_RATING_CLIENT	
1	0		1	2	
2	0	Cooking staff	1	2	
3	0	Laborers	2	2	
4	0	Core staff	2	2	
5	0		1	3	
6	0		3	1	
	REGION_RATING_CLIENT_W_CITY	WEEKDAY_APPR_PROCESS_START			
1		2	SATURDAY		
2		2	FRIDAY		
3		2	TUESDAY		
4		2	SATURDAY		
5		3	WEDNESDAY		
6		1	WEDNESDAY		
	HOOR_APPR_PROCESS_START	REG_REGION_NOT_LIVE_REGION	REG_REGION_NOT_WORK_REGION		
1		13	0		0
2		11	0		0
3		8	0		1
4		11	0		0
5		17	0		0
6		18	0		0
	LIVE_REGION_NOT_WORK_REGION	REG_CITY_NOT_LIVE_CITY	REG_CITY_NOT_WORK_CITY		
1		0	1		1
2		0	0		0
3		1	0		0
4		0	0		0
5		0	1		1
6		0	0		0
	LIVE_CITY_NOT_WORK_CITY	ORGANIZATION_TYPE	DAYS_LAST_PHONE_CHANGE		
1		0 Self-employed	-2411		
2		0 Hotel	-1513		
3		0 Business Entity Type 1	0		

4	0	School	-2778
5	0	Industry: type 4	-20
6	0	Business Entity Type 3	-1827

	FLAG_DOCUMENT_2	FLAG_DOCUMENT_3	FLAG_DOCUMENT_4	FLAG_DOCUMENT_5
1	0	1	0	0
2	0	1	0	0
3	0	1	0	0
4	0	1	0	0
5	0	1	0	0
6	0	1	0	0

	FLAG_DOCUMENT_6	FLAG_DOCUMENT_7	FLAG_DOCUMENT_8	FLAG_DOCUMENT_9
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0

	FLAG_DOCUMENT_10	FLAG_DOCUMENT_11	FLAG_DOCUMENT_12	FLAG_DOCUMENT_13
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0

	FLAG_DOCUMENT_14	FLAG_DOCUMENT_15	FLAG_DOCUMENT_16	FLAG_DOCUMENT_17
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0

	FLAG_DOCUMENT_18	FLAG_DOCUMENT_19	FLAG_DOCUMENT_20	FLAG_DOCUMENT_21
1	0	0	0	0
2	0	0	0	0
3	0	0	0	0
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0

	AMT_REQ_CREDIT_BUREAU_HOUR	AMT_REQ_CREDIT_BUREAU_DAY
1	0	0
2	0	0
3	0	0
4	0	0

5	0	0
6	0	0
AMT_REQ_CREDIT_BUREAU_WEEK AMT_REQ_CREDIT_BUREAU_MON		
1	0	0
2	0	0
3	0	1
4	0	0
5	0	0
6	0	0
AMT_REQ_CREDIT_BUREAU_QRT AMT_REQ_CREDIT_BUREAU_YEAR Income_Credit_Ratio		
1	0	0 0.2460630
2	0	4 0.1568960
3	0	2 0.6153846
4	0	3 0.2257135
5	0	1 0.3538403
6	0	1 0.2064999
Annuity_Income_Ratio Credit_As_Percentage Percent_Days_Employed		
1	0.1902857	4.064000 0.27435406
2	0.2542000	6.373650 0.08542964
3	0.1935750	1.625000 0.03878256
4	0.2367895	4.430395 0.21716018
5	0.1898333	2.826133 0.08994141
6	0.2057091	4.842618 0.22465180
Income_Per_Person		
1	157500	
2	90000	
3	90000	
4	85500	
5	135000	
6	82500	

```

set.seed(666)
train_index <- sample(1:nrow(data), 0.7 * nrow(data))
valid_index <- setdiff(1:nrow(data), train_index)
train_df <- data[train_index, ]
valid_df <- data[valid_index, ]
nrow(train_df)

```

```
[1] 20999
```

```
nrow(valid_df)
```

[1] 9000

```
head(train_df)
```

	X	SK_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODE_GENDER	FLAG_OWN_CAR
17983	235506	372786	1	Cash loans	F	N
12926	92468	207374	0	Cash loans	F	Y
13195	190897	321346	0	Cash loans	F	N
23676	103209	219794	0	Cash loans	M	N
15901	166080	292539	1	Cash loans	F	N
873	222231	357432	0	Cash loans	F	N
	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	
17983	Y	0	126000	467257.5	17743.5	
12926	Y	0	193500	260640.0	31059.0	
13195	N	2	337500	855882.0	36391.5	
23676	Y	0	166500	266832.0	24601.5	
15901	N	1	99000	360000.0	10102.5	
873	Y	1	157500	157500.0	14575.5	
	AMT_GOODS_PRICE	NAME_TYPE_SUITE	NAME_INCOME_TYPE			
17983	328500	Unaccompanied	Commercial associate			
12926	225000	Family	Working			
13195	765000	Unaccompanied	Commercial associate			
23676	238500	Family	Working			
15901	360000	Unaccompanied	Working			
873	157500	Unaccompanied	Working			
	NAME_EDUCATION_TYPE	NAME_FAMILY_STATUS	NAME_HOUSING_TYPE			
17983	Higher education	Single / not married	House / apartment			
12926	Secondary / secondary special	Single / not married	House / apartment			
13195	Secondary / secondary special	Separated	House / apartment			
23676	Secondary / secondary special	Civil marriage	House / apartment			
15901	Higher education	Single / not married	House / apartment			
873	Higher education	Married	House / apartment			
	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION	DAYS_ID_PUBLISH	OWN_CAR_AGE	
17983	-10631	-200	-4795	-3303	NA	
12926	-17424	-10763	-9653	-969	13	
13195	-13370	-3493	-6795	-4908	NA	
23676	-10606	-383	-5120	-3208	NA	
15901	-14135	-1400	-7982	-3265	NA	
873	-11931	-1154	-10268	-788	NA	
	FLAG_MOBIL	FLAG_EMP_PHONE	FLAG_WORK_PHONE	FLAG_CONT_MOBILE	FLAG_PHONE	
17983	1	1	0	1	0	

12926	1	1	0	1	0
13195	1	1	0	1	0
23676	1	1	0	1	0
15901	1	1	0	1	1
873	1	1	0	1	1
FLAG_EMAIL OCCUPATION_TYPE CNT_FAM_MEMBERS REGION_RATING_CLIENT					
17983	0	Sales staff	1		2
12926	0	Laborers	1		2
13195	0	Cleaning staff	3		2
23676	1	Laborers	2		2
15901	0	High skill tech staff	2		2
873	0	Sales staff	3		2
REGION_RATING_CLIENT_W_CITY WEEKDAY_APPR_PROCESS_START					
17983		2		WEDNESDAY	
12926		2		TUESDAY	
13195		2		WEDNESDAY	
23676		2		SUNDAY	
15901		2		WEDNESDAY	
873		2		TUESDAY	
HOUR_APPR_PROCESS_START REG_REGION_NOT_LIVE_REGION					
17983		17		0	
12926		16		0	
13195		13		0	
23676		12		0	
15901		19		0	
873		12		0	
REG_REGION_NOT_WORK_REGION LIVE_REGION_NOT_WORK_REGION					
17983		0		0	
12926		0		0	
13195		0		0	
23676		1		1	
15901		0		0	
873		0		0	
REG_CITY_NOT_LIVE_CITY REG_CITY_NOT_WORK_CITY LIVE_CITY_NOT_WORK_CITY					
17983		1		1	0
12926		0		1	1
13195		0		0	0
23676		0		0	0
15901		0		0	0
873		1		1	0
ORGANIZATION_TYPE DAYS_LAST_PHONE_CHANGE FLAG_DOCUMENT_2					
17983	Self-employed		0		0
12926	Business Entity Type 2		-1918		0

13195	Business Entity Type 3	-1370	0
23676	Business Entity Type 3	-282	0
15901	Industry: type 5	-1498	0
873	Self-employed	-1457	0
	FLAG_DOCUMENT_3	FLAG_DOCUMENT_4	FLAG_DOCUMENT_5
17983	1	0	0
12926	1	0	0
13195	1	0	0
23676	1	0	0
15901	1	0	0
873	1	0	0
	FLAG_DOCUMENT_7	FLAG_DOCUMENT_8	FLAG_DOCUMENT_9
17983	0	0	0
12926	0	0	0
13195	0	0	0
23676	0	0	0
15901	0	0	0
873	0	0	0
	FLAG_DOCUMENT_11	FLAG_DOCUMENT_12	FLAG_DOCUMENT_13
17983	0	0	0
12926	0	0	0
13195	0	0	0
23676	0	0	0
15901	0	0	0
873	0	0	0
	FLAG_DOCUMENT_15	FLAG_DOCUMENT_16	FLAG_DOCUMENT_17
17983	0	0	0
12926	0	0	0
13195	0	0	0
23676	0	0	0
15901	0	0	0
873	0	0	0
	FLAG_DOCUMENT_19	FLAG_DOCUMENT_20	FLAG_DOCUMENT_21
17983	0	0	0
12926	0	0	0
13195	0	0	0
23676	0	0	0
15901	0	0	0
873	0	0	0
	AMT_REQ_CREDIT_BUREAU_HOUR	AMT_REQ_CREDIT_BUREAU_DAY	
17983	0	0	
12926	0	0	
13195	0	0	



23676	0	0
15901	0	0
873	0	0
AMT_REQ_CREDIT_BUREAU_WEEK AMT_REQ_CREDIT_BUREAU_MON		
17983	0	0
12926	0	0
13195	0	0
23676	0	0
15901	0	0
873	0	1
AMT_REQ_CREDIT_BUREAU_QRT AMT_REQ_CREDIT_BUREAU_YEAR Income_Credit_Ratio		
17983	0	2 0.2696586
12926	1	4 0.7424033
13195	0	2 0.3943301
23676	0	6 0.6239881
15901	2	4 0.2750000
873	0	1 1.0000000
Annuity_Income_Ratio Credit_As_Percentage Percent_Days_Employed		
17983	0.14082143	3.708393 0.01881291
12926	0.16051163	1.346977 0.61771120
13195	0.10782667	2.535947 0.26125654
23676	0.14775676	1.602595 0.03611163
15901	0.10204545	3.636364 0.09904492
873	0.09254286	1.000000 0.09672282
Income_Per_Person		
17983	126000	
12926	193500	
13195	112500	
23676	83250	
15901	49500	
873	52500	

```
head(valid_df)
```

	X	SK_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODE_GENDER	FLAG_OWN_CAR
2	217645	352176	0	Cash loans	F	N
3	70440	181716	1	Cash loans	M	N
6	146804	270206	0	Cash loans	F	N
9	114242	232484	1	Cash loans	M	N
10	251026	390460	0	Revolving loans	F	N
12	229721	366072	1	Cash loans	F	N

	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY
2	Y	0	90000	573628.5	22878.0
3	Y	0	180000	292500.0	34843.5
6	Y	1	247500	1198548.0	50913.0
9	N	0	202500	545040.0	26640.0
10	Y	1	112500	337500.0	16875.0
12	Y	3	67500	312768.0	17095.5
	AMT_GOODS_PRICE	NAME_TYPE_SUITE	NAME_INCOME_TYPE		
2	463500	Children	Working		
3	292500	Unaccompanied	Working		
6	1102500	Unaccompanied	Commercial associate		
9	450000	Unaccompanied	Working		
10	337500	Unaccompanied	Pensioner		
12	270000	Unaccompanied	State servant		
	NAME_EDUCATION_TYPE	NAME_FAMILY_STATUS	NAME_HOUSING_TYPE		
2	Secondary / secondary special	Widow	House / apartment		
3	Secondary / secondary special	Civil marriage	House / apartment		
6	Higher education	Civil marriage	House / apartment		
9	Secondary / secondary special	Civil marriage	House / apartment		
10	Secondary / secondary special	Civil marriage	House / apartment		
12	Secondary / secondary special	Civil marriage	With parents		
	DAYS_BIRTH	DAYS_EMPLOYED	DAYS_REGISTRATION	DAYS_ID_PUBLISH	OWN_CAR_AGE
2	-20075	-1715	-1409	-3573	NA
3	-13898	-539	-2070	-258	NA
6	-13857	-3113	-7952	-4604	NA
9	-18234	-295	-205	-1315	NA
10	-16897	365243	-4399	-446	NA
12	-12185	-2602	-4412	-3509	NA
	FLAG_MOBIL	FLAG_EMP_PHONE	FLAG_WORK_PHONE	FLAG_CONT_MOBILE	FLAG_PHONE
2	1	1	0	1	1
3	1	1	0	1	0
6	1	1	0	1	0
9	1	1	0	1	0
10	1	0	0	1	0
12	1	1	1	1	0
	FLAG_EMAIL	OCCUPATION_TYPE	CNT_FAM_MEMBERS	REGION_RATING_CLIENT	
2	0	Cooking staff	1	2	
3	0	Laborers	2	2	
6	0		3	1	
9	0	Laborers	2	3	
10	0		3	2	
12	0	Core staff	5	2	
	REGION_RATING_CLIENT_W_CITY	WEEKDAY_APPR_PROCESS_START			

2	2	FRIDAY
3	2	TUESDAY
6	1	WEDNESDAY
9	3	THURSDAY
10	2	THURSDAY
12	2	SATURDAY

	HOUR_APPR_PROCESS_START	REG_REGION_NOT_LIVE_REGION
2	11	0
3	8	0
6	18	0
9	10	0
10	11	0
12	9	0

	REG_REGION_NOT_WORK_REGION	LIVE_REGION_NOT_WORK_REGION
2	0	0
3	1	1
6	0	0
9	0	0
10	0	0
12	0	0

	REG_CITY_NOT_LIVE_CITY	REG_CITY_NOT_WORK_CITY	LIVE_CITY_NOT_WORK_CITY
2	0	0	0
3	0	0	0
6	0	0	0
9	0	1	1
10	0	0	0
12	0	0	0

	ORGANIZATION_TYPE	DAYS_LAST_PHONE_CHANGE	FLAG_DOCUMENT_2
2	Hotel	-1513	0
3	Business Entity Type 1	0	0
6	Business Entity Type 3	-1827	0
9	Business Entity Type 3	0	0
10	XNA	-796	0
12	Kindergarten	-1079	0

	FLAG_DOCUMENT_3	FLAG_DOCUMENT_4	FLAG_DOCUMENT_5	FLAG_DOCUMENT_6
2	1	0	0	0
3	1	0	0	0
6	1	0	0	0
9	1	0	0	0
10	0	0	0	0
12	1	0	0	0

	FLAG_DOCUMENT_7	FLAG_DOCUMENT_8	FLAG_DOCUMENT_9	FLAG_DOCUMENT_10
2	0	0	0	0

3	0	0	0	0
6	0	0	0	0
9	0	0	0	0
10	0	0	0	0
12	0	0	0	0
	FLAG_DOCUMENT_11	FLAG_DOCUMENT_12	FLAG_DOCUMENT_13	FLAG_DOCUMENT_14
2	0	0	0	0
3	0	0	0	0
6	0	0	0	0
9	0	0	0	0
10	0	0	0	0
12	0	0	0	0
	FLAG_DOCUMENT_15	FLAG_DOCUMENT_16	FLAG_DOCUMENT_17	FLAG_DOCUMENT_18
2	0	0	0	0
3	0	0	0	0
6	0	0	0	0
9	0	0	0	0
10	0	0	0	0
12	0	0	0	0
	FLAG_DOCUMENT_19	FLAG_DOCUMENT_20	FLAG_DOCUMENT_21	
2	0	0	0	
3	0	0	0	
6	0	0	0	
9	0	0	0	
10	0	0	0	
12	0	0	0	
	AMT_REQ_CREDIT_BUREAU_HOUR	AMT_REQ_CREDIT_BUREAU_DAY		
2		0	0	
3		0	0	
6		0	0	
9		0	0	
10		0	0	
12		0	0	
	AMT_REQ_CREDIT_BUREAU_WEEK	AMT_REQ_CREDIT_BUREAU_MON		
2		0	0	
3		0	1	
6		0	0	
9		0	0	
10		0	0	
12		0	0	
	AMT_REQ_CREDIT_BUREAU_QRT	AMT_REQ_CREDIT_BUREAU_YEAR	Income_Credit_Ratio	
2		0	4	0.1568960
3		0	2	0.6153846

6	0	1	0.2064999
9	0	3	0.3715324
10	0	6	0.3333333
12	0	1	0.2158149

	Annuity_Income_Ratio	Credit_As_Percentage	Percent_Days_Employed
2	0.2542000	6.373650	0.08542964
3	0.1935750	1.625000	0.03878256
6	0.2057091	4.842618	0.22465180
9	0.1315556	2.691556	0.01617857
10	0.1500000	3.000000	-21.61584897
12	0.2532667	4.633600	0.21354124

	Income_Per_Person
2	90000
3	90000
6	82500
9	101250
10	37500
12	13500

```
str(train_df)
```

```
'data.frame': 20999 obs. of 73 variables:
 $ X                : int  235506 92468 190897 103209 166080 222231 30005 194123 1...
 $ SK_ID_CURR       : int  372786 207374 321346 219794 292539 357432 134832 325093
 $ TARGET           : int  1 0 0 0 1 0 0 1 0 0 ...
 $ NAME_CONTRACT_TYPE : chr  "Cash loans" "Cash loans" "Cash loans" "Cash loans" ...
 $ CODE_GENDER      : chr  "F" "F" "F" "M" ...
 $ FLAG_OWN_CAR     : chr  "N" "Y" "N" "N" ...
 $ FLAG_OWN_REALTY  : chr  "Y" "Y" "N" "Y" ...
 $ CNT_CHILDREN     : int  0 0 2 0 1 1 0 0 0 0 ...
 $ AMT_INCOME_TOTAL : num  126000 193500 337500 166500 99000 ...
 $ AMT_CREDIT       : num  467258 260640 855882 266832 360000 ...
 $ AMT_ANNUITY      : num  17744 31059 36392 24602 10102 ...
 $ AMT_GOODS_PRICE  : num  328500 225000 765000 238500 360000 ...
 $ NAME_TYPE_SUITE   : chr  "Unaccompanied" "Family" "Unaccompanied" "Family" ...
 $ NAME_INCOME_TYPE  : chr  "Commercial associate" "Working" "Commercial associate"
 $ NAME_EDUCATION_TYPE : chr  "Higher education" "Secondary / secondary special" "Seco
 $ NAME_FAMILY_STATUS : chr  "Single / not married" "Single / not married" "Separated
 $ NAME_HOUSING_TYPE  : chr  "House / apartment" "House / apartment" "House / apartm
 $ DAYS_BIRTH        : int  -10631 -17424 -13370 -10606 -14135 -11931 -18336 -12296
 $ DAYS_EMPLOYED     : int  -200 -10763 -3493 -383 -1400 -1154 -6561 -1212 -2285 -6
```

\$ DAYS_REGISTRATION	: int	-4795 -9653 -6795 -5120 -7982 -10268 -1038 -2417 -3801 -
\$ DAYS_ID_PUBLISH	: int	-3303 -969 -4908 -3208 -3265 -788 -1898 -2422 -4156 -47
\$ OWN_CAR_AGE	: int	NA 13 NA NA NA NA NA NA 15 NA ...
\$ FLAG_MOBIL	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ FLAG_EMP_PHONE	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ FLAG_WORK_PHONE	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_CONT_MOBILE	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ FLAG_PHONE	: int	0 0 0 0 1 1 0 0 0 0 ...
\$ FLAG_EMAIL	: int	0 0 0 1 0 0 0 0 0 0 ...
\$ OCCUPATION_TYPE	: chr	"Sales staff" "Laborers" "Cleaning staff" "Laborers" ..
\$ CNT_FAM_MEMBERS	: int	1 1 3 2 2 3 2 1 1 2 ...
\$ REGION_RATING_CLIENT	: int	2 2 2 2 2 2 2 2 2 2 ...
\$ REGION_RATING_CLIENT_W_CITY	: int	2 2 2 2 2 2 2 2 2 2 ...
\$ WEEKDAY_APPR_PROCESS_START	: chr	"WEDNESDAY" "TUESDAY" "WEDNESDAY" "SUNDAY" ...
\$ HOUR_APPR_PROCESS_START	: int	17 16 13 12 19 12 14 19 10 12 ...
\$ REG_REGION_NOT_LIVE_REGION	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ REG_REGION_NOT_WORK_REGION	: int	0 0 0 1 0 0 0 0 0 0 ...
\$ LIVE_REGION_NOT_WORK_REGION	: int	0 0 0 1 0 0 0 0 0 0 ...
\$ REG_CITY_NOT_LIVE_CITY	: int	1 0 0 0 0 1 0 0 0 0 ...
\$ REG_CITY_NOT_WORK_CITY	: int	1 1 0 0 0 1 0 1 0 1 ...
\$ LIVE_CITY_NOT_WORK_CITY	: int	0 1 0 0 0 0 0 1 0 1 ...
\$ ORGANIZATION_TYPE	: chr	"Self-employed" "Business Entity Type 2" "Business Enti
\$ DAYS_LAST_PHONE_CHANGE	: int	0 -1918 -1370 -282 -1498 -1457 -108 -394 -740 -1555 ...
\$ FLAG_DOCUMENT_2	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_3	: int	1 1 1 1 1 1 1 1 1 1 ...
\$ FLAG_DOCUMENT_4	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_5	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_6	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_7	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_8	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_9	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_10	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_11	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_12	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_13	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_14	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_15	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_16	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_17	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_18	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_19	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_20	: int	0 0 0 0 0 0 0 0 0 0 ...
\$ FLAG_DOCUMENT_21	: int	0 0 0 0 0 0 0 0 0 0 ...

```

$ AMT_REQ_CREDIT_BUREAU_HOUR : int 0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_DAY  : int 0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_WEEK : int 0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_MON  : int 0 0 0 0 0 1 0 0 0 1 ...
$ AMT_REQ_CREDIT_BUREAU_QRT  : int 0 1 0 0 2 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_YEAR : int 2 4 2 6 4 1 3 7 0 5 ...
$ Income_Credit_Ratio        : num 0.27 0.742 0.394 0.624 0.275 ...
$ Annuity_Income_Ratio       : num 0.141 0.161 0.108 0.148 0.102 ...
$ Credit_As_Percentage       : num 3.71 1.35 2.54 1.6 3.64 ...
$ Percent_Days_Employed      : num 0.0188 0.6177 0.2613 0.0361 0.099 ...
$ Income_Per_Person          : num 126000 193500 112500 83250 49500 ...

```

```
str(valid_df)
```

```

'data.frame':  9000 obs. of  73 variables:
 $ X                : int  217645 70440 146804 114242 251026 229721 85195 283368 1...
 $ SK_ID_CURR       : int  352176 181716 270206 232484 390460 366072 198841 428168 ...
 $ TARGET           : int  0 1 0 1 0 1 0 1 0 0 ...
 $ NAME_CONTRACT_TYPE : chr  "Cash loans" "Cash loans" "Cash loans" "Cash loans" ...
 $ CODE_GENDER      : chr  "F" "M" "F" "M" ...
 $ FLAG_OWN_CAR     : chr  "N" "N" "N" "N" ...
 $ FLAG_OWN_REALTY  : chr  "Y" "Y" "Y" "N" ...
 $ CNT_CHILDREN     : int  0 0 1 0 1 3 0 0 0 1 ...
 $ AMT_INCOME_TOTAL : num  90000 180000 247500 202500 112500 ...
 $ AMT_CREDIT       : num  573628 292500 1198548 545040 337500 ...
 $ AMT_ANNUITY      : num  22878 34844 50913 26640 16875 ...
 $ AMT_GOODS_PRICE  : num  463500 292500 1102500 450000 337500 ...
 $ NAME_TYPE_SUITE  : chr  "Children" "Unaccompanied" "Unaccompanied" "Unaccompanied" ...
 $ NAME_INCOME_TYPE : chr  "Working" "Working" "Commercial associate" "Working" ...
 $ NAME_EDUCATION_TYPE : chr  "Secondary / secondary special" "Secondary / secondary" ...
 $ NAME_FAMILY_STATUS : chr  "Widow" "Civil marriage" "Civil marriage" "Civil marriage" ...
 $ NAME_HOUSING_TYPE : chr  "House / apartment" "House / apartment" "House / apartment" ...
 $ DAYS_BIRTH       : int  -20075 -13898 -13857 -18234 -16897 -12185 -10579 -11938 ...
 $ DAYS_EMPLOYED    : int  -1715 -539 -3113 -295 365243 -2602 -246 -680 365243 -15...
 $ DAYS_REGISTRATION : int  -1409 -2070 -7952 -205 -4399 -4412 -4665 -5949 -4764 -4...
 $ DAYS_ID_PUBLISH  : int  -3573 -258 -4604 -1315 -446 -3509 -3226 -4153 -82 -2312 ...
 $ OWN_CAR_AGE      : int  NA NA NA NA NA NA NA NA NA 8 ...
 $ FLAG_MOBIL       : int  1 1 1 1 1 1 1 1 1 1 ...
 $ FLAG_EMP_PHONE    : int  1 1 1 1 0 1 1 1 0 1 ...
 $ FLAG_WORK_PHONE   : int  0 0 0 0 0 1 0 0 0 1 ...
 $ FLAG_CONT_MOBILE  : int  1 1 1 1 1 1 1 1 1 1 ...

```

```

$ FLAG_PHONE           : int  1 0 0 0 0 0 1 0 0 1 ...
$ FLAG_EMAIL           : int  0 0 0 0 0 0 0 0 0 0 ...
$ OCCUPATION_TYPE      : chr  "Cooking staff" "Laborers" "" "Laborers" ...
$ CNT_FAM_MEMBERS      : int  1 2 3 2 3 5 1 1 2 3 ...
$ REGION_RATING_CLIENT : int  2 2 1 3 2 2 2 2 2 2 ...
$ REGION_RATING_CLIENT_W_CITY: int  2 2 1 3 2 2 2 2 2 2 ...
$ WEEKDAY_APPR_PROCESS_START : chr  "FRIDAY" "TUESDAY" "WEDNESDAY" "THURSDAY" ...
$ HOUR_APPR_PROCESS_START : int  11 8 18 10 11 9 14 8 9 11 ...
$ REG_REGION_NOT_LIVE_REGION : int  0 0 0 0 0 0 0 0 0 0 ...
$ REG_REGION_NOT_WORK_REGION : int  0 1 0 0 0 0 0 0 0 0 ...
$ LIVE_REGION_NOT_WORK_REGION: int  0 1 0 0 0 0 0 0 0 0 ...
$ REG_CITY_NOT_LIVE_CITY : int  0 0 0 0 0 0 0 0 0 0 ...
$ REG_CITY_NOT_WORK_CITY : int  0 0 0 1 0 0 1 0 0 1 ...
$ LIVE_CITY_NOT_WORK_CITY : int  0 0 0 1 0 0 1 0 0 1 ...
$ ORGANIZATION_TYPE    : chr  "Hotel" "Business Entity Type 1" "Business Entity Type 3" ...
$ DAYS_LAST_PHONE_CHANGE : int  -1513 0 -1827 0 -796 -1079 -558 -2209 0 -1105 ...
$ FLAG_DOCUMENT_2      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_3      : int  1 1 1 1 0 1 1 1 1 1 ...
$ FLAG_DOCUMENT_4      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_5      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_6      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_7      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_8      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_9      : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_10     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_11     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_12     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_13     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_14     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_15     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_16     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_17     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_18     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_19     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_20     : int  0 0 0 0 0 0 0 0 0 0 ...
$ FLAG_DOCUMENT_21     : int  0 0 0 0 0 0 0 0 0 0 ...
$ AMT_REQ_CREDIT_BUREAU_HOUR : int  0 0 0 0 0 0 0 0 0 NA ...
$ AMT_REQ_CREDIT_BUREAU_DAY : int  0 0 0 0 0 0 0 0 0 NA ...
$ AMT_REQ_CREDIT_BUREAU_WEEK : int  0 0 0 0 0 0 0 0 1 NA ...
$ AMT_REQ_CREDIT_BUREAU_MON : int  0 1 0 0 0 0 0 0 0 NA ...
$ AMT_REQ_CREDIT_BUREAU_QRT : int  0 0 0 0 0 0 0 0 0 NA ...
$ AMT_REQ_CREDIT_BUREAU_YEAR : int  4 2 1 3 6 1 0 4 0 NA ...
$ Income_Credit_Ratio    : num  0.157 0.615 0.206 0.372 0.333 ...

```



```
$ Annuity_Income_Ratio      : num  0.254 0.194 0.206 0.132 0.15 ...
$ Credit_As_Percentage      : num  6.37 1.62 4.84 2.69 3 ...
$ Percent_Days_Employed     : num  0.0854 0.0388 0.2247 0.0162 -21.6158 ...
$ Income_Per_Person         : num  90000 90000 82500 101250 37500 ...
```

```
table(train_df$CODE_GENDER)
```

```
      F      M
13627  7372
```

```
#Variable list
#Income_Credit_Ratio + Annuity_Income_Ratio + Credit_As_Percentage + Percent_Days_Employed
```

```
#Convert all categorical variables into factors
train_df$OCCUPATION_TYPE <- as.factor(train_df$OCCUPATION_TYPE)
train_df$ORGANIZATION_TYPE <- as.factor(train_df$ORGANIZATION_TYPE)
train_df$NAME_EDUCATION_TYPE <- as.factor(train_df$NAME_EDUCATION_TYPE)
train_df <- train_df[train_df$CODE_GENDER != "XNA", ]
```

```
#Use ROSE to oversample target variable in order to balance model
train_df_rose <- ROSE(TARGET ~ Percent_Days_Employed + NAME_EDUCATION_TYPE + REGION_RATING_
                      data = train_df, seed = 666)$data
```

```
table(train_df_rose$TARGET)
```

```
      0      1
10337 10640
```

```
#Create classification decision tree with relevant fields
class_tr_cl <- rpart(TARGET ~ Percent_Days_Employed + NAME_EDUCATION_TYPE + REGION_RATING_
                    data = train_df_rose, method = "class", control = rpart.control(cp = 0

prp(class_tr_cl, cex = 0.8, tweak = 1)
```



P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.2088

McNemar's Test P-Value : < 2.2e-16

Sensitivity : 0.6818  
Specificity : 0.5266  
Pos Pred Value : 0.5971  
Neg Pred Value : 0.6165  
Prevalence : 0.5072  
Detection Rate : 0.3458  
Detection Prevalence : 0.5791  
Balanced Accuracy : 0.6042

'Positive' Class : 1

```
confusionMatrix(class_tr_valid, valid_df$TARGET, positive = "1")
```

#### Confusion Matrix and Statistics

	Reference	
Prediction	0	1
0	3853	626
1	3420	1101

Accuracy : 0.5504  
95% CI : (0.5401, 0.5608)  
No Information Rate : 0.8081  
P-Value [Acc > NIR] : 1

Kappa : 0.1035

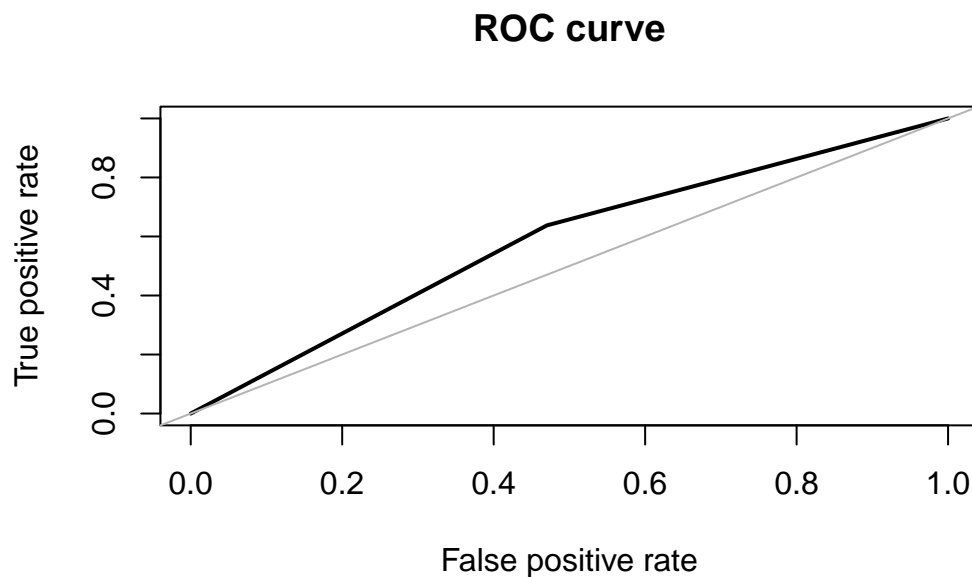
McNemar's Test P-Value : <2e-16

Sensitivity : 0.6375  
Specificity : 0.5298  
Pos Pred Value : 0.2435  
Neg Pred Value : 0.8602  
Prevalence : 0.1919

Detection Rate : 0.1223  
Detection Prevalence : 0.5023  
Balanced Accuracy : 0.5836

'Positive' Class : 1

```
library(ROSE)
ROSE::roc.curve(valid_df$TARGET, class_tr_valid)
```



Area under the curve (AUC): 0.584

```
# Load new customer data
test <- read.csv("credit_test_fa2023_23.csv", header = TRUE)
#Add percent_days_employed field
for (i in 1:nrow(test)) {
  test$Percent_Days_Employed[i] <- test$DAYS_EMPLOYED[i] / test$DAYS_BIRTH[i]
}

# Predict risk of new customers
```

```
credit_prediction <- predict(class_tr_cl, newdata = test,
                             type = "class")
credit_prediction
```

```
1 2 3 4 5
0 0 1 1 0
Levels: 0 1
```

Write-up:

Problem Description:

Stark Enterprises decided it wanted to branch out into the financial industry. They want to create a model to assist them in this endeavor.

Objective: Create a model to predict which customers for a loan are likely to be high risk.

Data description: The data includes the characteristics and financial situation of our customers. This includes fields which are strictly personal, like their gender and education, and fields which are finance-related, like their income and loan annuity

Data Modifications: percent\_days\_employed was created by dividing Days employed by customer's age. The target variable for the training set was also over-sampled to raise sensitivity.

Both models were tailored to sensitivity. As a company which is branching into a new industry, it may be more important for them to catch all high risk customers to prevent losses. We chose to select the decision tree model because it was better at distinguishing risk (58.4% AUC to 56.6% AUC). Our chosen model correctly identifies high risk candidates 63.75% of the time. While this metric is high, the model pays for it with a lower accuracy of 55.04% and a low pos pred value of 24.35%. What this means is that this model has a tendency to incorrectly classify low risk candidates as high risk candidates. While this is regrettable, improving this metric would necessitate a reduction in sensitivity, which would be dangerous as more high risk customers would slip under the radar. Looking at it another way, 86.02% of low risk predictions are accurate, meaning that ~14% of low risk candidates may be unfairly rejected for a loan due to this model. The question of the efficacy of this model therefore lies in what the revenue loss will be from 14% of low risk candidates rejected vs. the potential revenue loss from giving loans to more high-risk candidates. Ultimately, as Stark Enterprises becomes more established, we believe the model could be tweaked to have less false positives, at the expense of false negatives, in order to give out more loans.

```
Accuracy : 0.5504
          95% CI : (0.5401, 0.5608)
No Information Rate : 0.8081
```

P-Value [Acc > NIR] : 1

Kappa : 0.1035

McNemar's Test P-Value : <2e-16

Sensitivity : 0.6375

Specificity : 0.5298

Pos Pred Value : 0.2435

Neg Pred Value : 0.8602

Prevalence : 0.1919

Detection Rate : 0.1223

Detection Prevalence : 0.5023

Balanced Accuracy : 0.5836