

EDA - Practice Project

Joseph Pushnam

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1 Introduction: Statement of the Business and Analytic Problems

The primary objective of this project is to predict loan default risk using data from loan applicants, helping the business to make informed decisions and mitigate financial risk associated with loan approvals. The dataset includes various demographic, financial, and loan-specific features, and the target variable (`TARGET`) indicates whether an applicant has defaulted on their loan (`1` for default, `0` for no default).

To understand the data and its potential for predictive modeling, the following key questions are addressed in the Exploratory Data Analysis (EDA):

1.1 Questions Addressed in the EDA:

Is the target variable (loan default) imbalanced, and what would be the accuracy of a simple majority class classifier?

- This question explores whether the target variable is imbalanced, which is important for deciding whether any balancing techniques (e.g., oversampling or undersampling) will be needed. We also calculate the accuracy of a simple model that always predicts the majority class, serving as a baseline for future modeling efforts.

What is the relationship between the target and the predictors? Are there potentially strong predictors that could be included in a model?

- We investigate correlations and other relationships between the target and key predictors, looking for features that may strongly influence loan default. Identifying strong predictors at this stage helps in feature selection and engineering for future modeling.

How can the `skimr` and `janitor` packages in R assist with data exploration and cleaning?

- The `skimr` package provides useful data summaries, including information on data types, missing values, and basic statistics. The `janitor` package helps streamline data cleaning by removing empty rows/columns and renaming columns. These tools are used throughout the EDA to simplify and enhance data exploration and preparation.

What is the scope of missing data, and what are the possible solutions? Should we remove rows, remove columns, or impute missing values?

- Missing data can impact model accuracy if not handled appropriately. We explore the extent of missing data across the dataset and consider different strategies to address it, such as removing rows or columns, or using imputation methods like mean/median imputation for numeric variables or mode imputation for categorical variables.

2 Brief Exploratory Data Analysis

2.1 Checking for Class Imbalance:

```
# Set CRAN mirror
options(repos = c(CRAN = "https://cran.rstudio.com/"))
```

```
# Load libraries
library(skimr)
library(janitor)
```

```
##
## Attaching package: 'janitor'
```

```
## The following objects are masked from 'package:stats':
##
##   chisq.test, fisher.test
```

```
library(ggplot2)

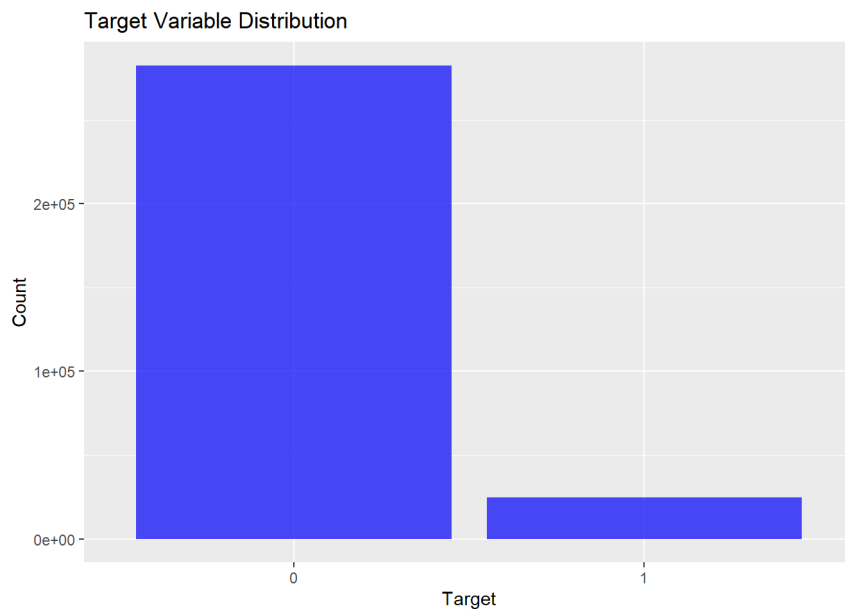
# Set working directory
setwd("C:\\Users\\Joseph\\OneDrive\\Desktop\\UoU\\DB_R\\3. Fall 2024\\Practice Project")

# Read the dataset
train_data <- read.csv("application_train.csv")

# Count of target classes
table(train_data$TARGET)
```

```
##  
##      0      1  
## 282686 24825
```

```
# Plotting the distribution  
ggplot(train_data, aes(x = as.factor(TARGET))) +  
  geom_bar(fill = "blue", alpha = 0.7) +  
  labs(title = "Target Variable Distribution", x = "Target", y = "Count")
```



```
# Summary of the data  
summary(train_data)
```

```

## SK_ID_CURR TARGET NAME_CONTRACT_TYPE CODE_GENDER
## Min. :100002 Min. :0.00000 Length:307511 Length:307511
## 1st Qu.:189146 1st Qu.:0.00000 Class :character Class :character
## Median :278202 Median :0.00000 Mode :character Mode :character
## Mean :278181 Mean :0.00073
## 3rd Qu.:367143 3rd Qu.:0.00000
## Max. :456255 Max. :1.00000
##
## FLAG_OWN_CAR FLAG_OWN_REALTY CNT_CHILDREN AMT_INCOME_TOTAL
## Length:307511 Length:307511 Min. : 0.0000 Min. : 25650
## Class :character Class :character 1st Qu.: 0.0000 1st Qu.: 112500
## Mode :character Mode :character Median : 0.0000 Median : 147150
## Mean : 0.4171 Mean : 168798
## 3rd Qu.: 1.0000 3rd Qu.: 202500
## Max. :19.0000 Max. :117000000
##
## AMT_CREDIT AMT_ANNUITY AMT_GOODS_PRICE NAME_TYPE_SUITE
## Min. : 45000 Min. : 1616 Min. : 40500 Length:307511
## 1st Qu.: 270000 1st Qu.: 16524 1st Qu.: 238500 Class :character
## Median : 513531 Median : 24903 Median : 450000 Mode :character
## Mean : 599026 Mean : 27109 Mean : 538396
## 3rd Qu.: 808650 3rd Qu.: 34596 3rd Qu.: 679500
## Max. :4050000 Max. :258026 Max. :4050000
## NA's :12 NA's :278
## NAME_INCOME_TYPE NAME_EDUCATION_TYPE NAME_FAMILY_STATUS NAME_HOUSING_TYPE
## Length:307511 Length:307511 Length:307511 Length:307511
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
##
##
##
## REGION_POPULATION_RELATIVE DAYS_BIRTH DAYS_EMPLOYED DAYS_REGISTRATION
## Min. :0.00029 Min. : -25229 Min. : -17912 Min. : -24672
## 1st Qu.:0.01001 1st Qu.: -19682 1st Qu.: -2760 1st Qu.: -7480
## Median :0.01885 Median : -15750 Median : -1213 Median : -4504
## Mean :0.02087 Mean : -16037 Mean : 63815 Mean : -4986
## 3rd Qu.:0.02866 3rd Qu.: -12413 3rd Qu.: -289 3rd Qu.: -2010
## Max. :0.07251 Max. : -7489 Max. :365243 Max. : 0
##
## DAYS_ID_PUBLISH OWN_CAR_AGE FLAG_MOBIL FLAG_EMP_PHONE
## Min. : -7197 Min. : 0.00 Min. :0 Min. :0.0000
## 1st Qu.: -4299 1st Qu.: 5.00 1st Qu.:1 1st Qu.:1.0000
## Median : -3254 Median : 9.00 Median :1 Median :1.0000
## Mean : -2994 Mean :12.06 Mean :1 Mean :0.8199
## 3rd Qu.: -1720 3rd Qu.:15.00 3rd Qu.:1 3rd Qu.:1.0000
## Max. : 0 Max. :91.00 Max. :1 Max. :1.0000
## NA's :202929
## FLAG_WORK_PHONE FLAG_CONT_MOBILE FLAG_PHONE FLAG_EMAIL
## Min. :0.0000 Min. :0.0000 Min. :0.0000 Min. :0.00000
## 1st Qu.:0.0000 1st Qu.:1.0000 1st Qu.:0.0000 1st Qu.:0.00000
## Median :0.0000 Median :1.0000 Median :0.0000 Median :0.00000
## Mean :0.1994 Mean :0.9981 Mean :0.2811 Mean :0.05672
## 3rd Qu.:0.0000 3rd Qu.:1.0000 3rd Qu.:1.0000 3rd Qu.:0.00000
## Max. :1.0000 Max. :1.0000 Max. :1.0000 Max. :1.00000
##
## OCCUPATION_TYPE CNT_FAM_MEMBERS REGION_RATING_CLIENT
## Length:307511 Min. : 1.000 Min. :1.000
## Class :character 1st Qu.: 2.000 1st Qu.:2.000
## Mode :character Median : 2.000 Median :2.000
## Mean : 2.153 Mean :2.052
## 3rd Qu.: 3.000 3rd Qu.:2.000
## Max. :20.000 Max. :3.000
## NA's :2
## REGION_RATING_CLIENT_W_CITY WEEKDAY_APPR_PROCESS_START HOUR_APPR_PROCESS_START
## Min. :1.000 Length:307511 Min. : 0.00
## 1st Qu.:2.000 Class :character 1st Qu.:10.00
## Median :2.000 Mode :character Median :12.00
## Mean :2.032 Mean :12.06
## 3rd Qu.:2.000 3rd Qu.:14.00
## Max. :3.000 Max. :23.00
##
## REG_REGION_NOT_LIVE_REGION REG_REGION_NOT_WORK_REGION
## Min. :0.00000 Min. :0.00000
## 1st Qu.:0.00000 1st Qu.:0.00000
## Median :0.00000 Median :0.00000
## Mean :0.01514 Mean :0.05077
## 3rd Qu.:0.00000 3rd Qu.:0.00000
## Max. :1.00000 Max. :1.00000
##
## LIVE_REGION_NOT_WORK_REGION REG_CITY_NOT_LIVE_CITY REG_CITY_NOT_WORK_CITY

```

```

## Min. :0.00000 Min. :0.00000 Min. :0.0000
## 1st Qu.:0.00000 1st Qu.:0.00000 1st Qu.:0.0000
## Median :0.00000 Median :0.00000 Median :0.0000
## Mean :0.04066 Mean :0.07817 Mean :0.2305
## 3rd Qu.:0.00000 3rd Qu.:0.00000 3rd Qu.:0.0000
## Max. :1.00000 Max. :1.00000 Max. :1.0000
##
## LIVE_CITY_NOT_WORK_CITY ORGANIZATION_TYPE EXT_SOURCE_1 EXT_SOURCE_2
## Min. :0.0000 Length:307511 Min. :0.01 Min. :0.0000
## 1st Qu.:0.0000 Class :character 1st Qu.:0.33 1st Qu.:0.3925
## Median :0.0000 Mode :character Median :0.51 Median :0.5660
## Mean :0.1796 Mean :0.50 Mean :0.5144
## 3rd Qu.:0.0000 3rd Qu.:0.68 3rd Qu.:0.6636
## Max. :1.0000 Max. :0.96 Max. :0.8550
##
## NA's :173378 NA's :660
## EXT_SOURCE_3 APARTMENTS_AVG BASEMENTAREA_AVG YEARS_BEGINEXPLUATATION_AVG
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.37 1st Qu.:0.06 1st Qu.:0.04 1st Qu.:0.98
## Median :0.54 Median :0.09 Median :0.08 Median :0.98
## Mean :0.51 Mean :0.12 Mean :0.09 Mean :0.98
## 3rd Qu.:0.67 3rd Qu.:0.15 3rd Qu.:0.11 3rd Qu.:0.99
## Max. :0.90 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :60965 NA's :156061 NA's :179943 NA's :150007
## YEARS_BUILD_AVG COMMONAREA_AVG ELEVATORS_AVG ENTRANCES_AVG
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.69 1st Qu.:0.01 1st Qu.:0.00 1st Qu.:0.07
## Median :0.76 Median :0.02 Median :0.00 Median :0.14
## Mean :0.75 Mean :0.04 Mean :0.08 Mean :0.15
## 3rd Qu.:0.82 3rd Qu.:0.05 3rd Qu.:0.12 3rd Qu.:0.21
## Max. :1.00 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :204488 NA's :214865 NA's :163891 NA's :154828
## FLOORSMAX_AVG FLOORSMIN_AVG LANDAREA_AVG LIVINGAPARTMENTS_AVG
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.17 1st Qu.:0.08 1st Qu.:0.02 1st Qu.:0.05
## Median :0.17 Median :0.21 Median :0.05 Median :0.08
## Mean :0.23 Mean :0.23 Mean :0.07 Mean :0.10
## 3rd Qu.:0.33 3rd Qu.:0.38 3rd Qu.:0.09 3rd Qu.:0.12
## Max. :1.00 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :153020 NA's :208642 NA's :182590 NA's :210199
## LIVINGAREA_AVG NONLIVINGAPARTMENTS_AVG NONLIVINGAREA_AVG APARTMENTS_MODE
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.05 1st Qu.:0.00 1st Qu.:0.00 1st Qu.:0.05
## Median :0.07 Median :0.00 Median :0.00 Median :0.08
## Mean :0.11 Mean :0.01 Mean :0.03 Mean :0.11
## 3rd Qu.:0.13 3rd Qu.:0.00 3rd Qu.:0.03 3rd Qu.:0.14
## Max. :1.00 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :154350 NA's :213514 NA's :169682 NA's :156061
## BASEMENTAREA_MODE YEARS_BEGINEXPLUATATION_MODE YEARS_BUILD_MODE
## Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.04 1st Qu.:0.98 1st Qu.:0.70
## Median :0.07 Median :0.98 Median :0.76
## Mean :0.09 Mean :0.98 Mean :0.76
## 3rd Qu.:0.11 3rd Qu.:0.99 3rd Qu.:0.82
## Max. :1.00 Max. :1.00 Max. :1.00
## NA's :179943 NA's :150007 NA's :204488
## COMMONAREA_MODE ELEVATORS_MODE ENTRANCES_MODE FLOORSMAX_MODE
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.01 1st Qu.:0.00 1st Qu.:0.07 1st Qu.:0.17
## Median :0.02 Median :0.00 Median :0.14 Median :0.17
## Mean :0.04 Mean :0.07 Mean :0.15 Mean :0.22
## 3rd Qu.:0.05 3rd Qu.:0.12 3rd Qu.:0.21 3rd Qu.:0.33
## Max. :1.00 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :214865 NA's :163891 NA's :154828 NA's :153020
## FLOORSMIN_MODE LANDAREA_MODE LIVINGAPARTMENTS_MODE LIVINGAREA_MODE
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.08 1st Qu.:0.02 1st Qu.:0.05 1st Qu.:0.04
## Median :0.21 Median :0.05 Median :0.08 Median :0.07
## Mean :0.23 Mean :0.06 Mean :0.11 Mean :0.11
## 3rd Qu.:0.38 3rd Qu.:0.08 3rd Qu.:0.13 3rd Qu.:0.13
## Max. :1.00 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :208642 NA's :182590 NA's :210199 NA's :154350
## NONLIVINGAPARTMENTS_MODE NONLIVINGAREA_MODE APARTMENTS_MEDI BASEMENTAREA_MEDI
## Min. :0.00 Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.00 1st Qu.:0.00 1st Qu.:0.06 1st Qu.:0.04
## Median :0.00 Median :0.00 Median :0.09 Median :0.08
## Mean :0.01 Mean :0.03 Mean :0.12 Mean :0.09
## 3rd Qu.:0.00 3rd Qu.:0.02 3rd Qu.:0.15 3rd Qu.:0.11
## Max. :1.00 Max. :1.00 Max. :1.00 Max. :1.00
## NA's :213514 NA's :169682 NA's :156061 NA's :179943
## YEARS_BEGINEXPLUATATION_MEDI YEARS_BUILD_MEDI COMMONAREA_MEDI
## Min. :0.00 Min. :0.00 Min. :0.00
## 1st Qu.:0.98 1st Qu.:0.69 1st Qu.:0.01

```

```

## Median :0.98          Median :0.76      Median :0.02
## Mean   :0.98          Mean   :0.76      Mean   :0.04
## 3rd Qu.:0.99          3rd Qu.:0.83      3rd Qu.:0.05
## Max.   :1.00          Max.   :1.00      Max.   :1.00
## NA's   :150007        NA's   :204488    NA's   :214865
## ELEVATORS_MEDI ENTRANCES_MEDI FLOORSMAX_MEDI FLOORSMIN_MEDI
## Min.    :0.00      Min.    :0.00      Min.    :0.00      Min.    :0.00
## 1st Qu.:0.00      1st Qu.:0.07      1st Qu.:0.17      1st Qu.:0.08
## Median :0.00      Median :0.14      Median :0.17      Median :0.21
## Mean   :0.08      Mean   :0.15      Mean   :0.23      Mean   :0.23
## 3rd Qu.:0.12      3rd Qu.:0.21      3rd Qu.:0.33      3rd Qu.:0.38
## Max.   :1.00      Max.   :1.00      Max.   :1.00      Max.   :1.00
## NA's   :163891    NA's   :154828    NA's   :153020    NA's   :208642
## LANDAREA_MEDI LIVINGAPARTMENTS_MEDI LIVINGAREA_MEDI
## Min.    :0.00      Min.    :0.00      Min.    :0.00
## 1st Qu.:0.02      1st Qu.:0.05      1st Qu.:0.05
## Median :0.05      Median :0.08      Median :0.07
## Mean   :0.07      Mean   :0.10      Mean   :0.11
## 3rd Qu.:0.09      3rd Qu.:0.12      3rd Qu.:0.13
## Max.   :1.00      Max.   :1.00      Max.   :1.00
## NA's   :182590    NA's   :210199    NA's   :154350
## NONLIVINGAPARTMENTS_MEDI NONLIVINGAREA_MEDI FONDKAPREMONT_MODE
## Min.    :0.00      Min.    :0.00      Length:307511
## 1st Qu.:0.00      1st Qu.:0.00      Class :character
## Median :0.00      Median :0.00      Mode  :character
## Mean   :0.01      Mean   :0.03
## 3rd Qu.:0.00      3rd Qu.:0.03
## Max.   :1.00      Max.   :1.00
## NA's   :213514      NA's   :169682
## HOUSETYPE_MODE TOTALAREA_MODE WALLSMATERIAL_MODE EMERGENCYSTATE_MODE
## Length:307511 Min.    :0.00      Length:307511      Length:307511
## Class :character 1st Qu.:0.04      Class :character    Class :character
## Mode :character Median :0.07      Mode :character     Mode :character
## Mean :0.10
## 3rd Qu.:0.13
## Max. :1.00
## NA's :148431
## OBS_30_CNT_SOCIAL_CIRCLE DEF_30_CNT_SOCIAL_CIRCLE OBS_60_CNT_SOCIAL_CIRCLE
## Min. : 0.000      Min. : 0.0000      Min. : 0.000
## 1st Qu.: 0.000      1st Qu.: 0.0000      1st Qu.: 0.000
## Median : 0.000      Median : 0.0000      Median : 0.000
## Mean : 1.422      Mean : 0.1434      Mean : 1.405
## 3rd Qu.: 2.000      3rd Qu.: 0.0000      3rd Qu.: 2.000
## Max. : 348.000      Max. : 34.0000      Max. : 344.000
## NA's :1021      NA's :1021      NA's :1021
## DEF_60_CNT_SOCIAL_CIRCLE DAYS_LAST_PHONE_CHANGE FLAG_DOCUMENT_2
## Min. : 0.0      Min. : -4292.0      Min. : 0.00e+00
## 1st Qu.: 0.0      1st Qu.: -1570.0      1st Qu.: 0.00e+00
## Median : 0.0      Median : -757.0      Median : 0.00e+00
## Mean : 0.1      Mean : -962.9      Mean : 4.23e-05
## 3rd Qu.: 0.0      3rd Qu.: -274.0      3rd Qu.: 0.00e+00
## Max. : 24.0      Max. : 0.0      Max. : 1.00e+00
## NA's :1021      NA's :1
## FLAG_DOCUMENT_3 FLAG_DOCUMENT_4 FLAG_DOCUMENT_5 FLAG_DOCUMENT_6
## Min. :0.00      Min. :0.00e+00      Min. :0.00000      Min. :0.00000
## 1st Qu.:0.00      1st Qu.:0.00e+00      1st Qu.:0.00000      1st Qu.:0.00000
## Median :1.00      Median :0.00e+00      Median :0.00000      Median :0.00000
## Mean :0.71      Mean :8.13e-05      Mean :0.01511      Mean :0.08806
## 3rd Qu.:1.00      3rd Qu.:0.00e+00      3rd Qu.:0.00000      3rd Qu.:0.00000
## Max. :1.00      Max. :1.00e+00      Max. :1.00000      Max. :1.00000
##
## FLAG_DOCUMENT_7 FLAG_DOCUMENT_8 FLAG_DOCUMENT_9 FLAG_DOCUMENT_10
## Min. :0.0000000      Min. :0.00000      Min. :0.000000      Min. :0.00e+00
## 1st Qu.:0.0000000      1st Qu.:0.00000      1st Qu.:0.000000      1st Qu.:0.00e+00
## Median :0.0000000      Median :0.00000      Median :0.000000      Median :0.00e+00
## Mean :0.0001919      Mean :0.08138      Mean :0.003896      Mean :2.28e-05
## 3rd Qu.:0.0000000      3rd Qu.:0.00000      3rd Qu.:0.000000      3rd Qu.:0.00e+00
## Max. :1.0000000      Max. :1.00000      Max. :1.000000      Max. :1.00e+00
##
## FLAG_DOCUMENT_11 FLAG_DOCUMENT_12 FLAG_DOCUMENT_13 FLAG_DOCUMENT_14
## Min. :0.000000      Min. :0.0e+00      Min. :0.000000      Min. :0.000000
## 1st Qu.:0.000000      1st Qu.:0.0e+00      1st Qu.:0.000000      1st Qu.:0.000000
## Median :0.000000      Median :0.0e+00      Median :0.000000      Median :0.000000
## Mean :0.003912      Mean :6.5e-06      Mean :0.003525      Mean :0.002936
## 3rd Qu.:0.000000      3rd Qu.:0.0e+00      3rd Qu.:0.000000      3rd Qu.:0.000000
## Max. :1.000000      Max. :1.0e+00      Max. :1.000000      Max. :1.000000
##
## FLAG_DOCUMENT_15 FLAG_DOCUMENT_16 FLAG_DOCUMENT_17 FLAG_DOCUMENT_18
## Min. :0.00000      Min. :0.000000      Min. :0.0000000      Min. :0.00000
## 1st Qu.:0.00000      1st Qu.:0.000000      1st Qu.:0.0000000      1st Qu.:0.00000
## Median :0.00000      Median :0.000000      Median :0.0000000      Median :0.00000
## Mean :0.00121      Mean :0.009928      Mean :0.0002667      Mean :0.00813

```

```
## 3rd Qu.:0.00000 3rd Qu.:0.00000 3rd Qu.:0.0000000 3rd Qu.:0.00000
## Max. :1.00000 Max. :1.00000 Max. :1.0000000 Max. :1.00000
##
## FLAG_DOCUMENT_19 FLAG_DOCUMENT_20 FLAG_DOCUMENT_21
## Min. :0.0000000 Min. :0.0000000 Min. :0.0000000
## 1st Qu.:0.0000000 1st Qu.:0.0000000 1st Qu.:0.0000000
## Median :0.0000000 Median :0.0000000 Median :0.0000000
## Mean :0.0005951 Mean :0.0005073 Mean :0.0003349
## 3rd Qu.:0.0000000 3rd Qu.:0.0000000 3rd Qu.:0.0000000
## Max. :1.0000000 Max. :1.0000000 Max. :1.0000000
##
## AMT_REQ_CREDIT_BUREAU_HOUR AMT_REQ_CREDIT_BUREAU_DAY
## Min. :0.00 Min. :0.00
## 1st Qu.:0.00 1st Qu.:0.00
## Median :0.00 Median :0.00
## Mean :0.01 Mean :0.01
## 3rd Qu.:0.00 3rd Qu.:0.00
## Max. :4.00 Max. :9.00
## NA's :41519 NA's :41519
## AMT_REQ_CREDIT_BUREAU_WEEK AMT_REQ_CREDIT_BUREAU_MON AMT_REQ_CREDIT_BUREAU_QRT
## Min. :0.00 Min. : 0.00 Min. : 0.00
## 1st Qu.:0.00 1st Qu.: 0.00 1st Qu.: 0.00
## Median :0.00 Median : 0.00 Median : 0.00
## Mean :0.03 Mean : 0.27 Mean : 0.27
## 3rd Qu.:0.00 3rd Qu.: 0.00 3rd Qu.: 0.00
## Max. :8.00 Max. :27.00 Max. :261.00
## NA's :41519 NA's :41519 NA's :41519
## AMT_REQ_CREDIT_BUREAU_YEAR
## Min. : 0.0
## 1st Qu.: 0.0
## Median : 1.0
## Mean : 1.9
## 3rd Qu.: 3.0
## Max. :25.0
## NA's :41519
```

2.2 Majority Class Classifier Accuracy:

2.2.1 Correlation with Numeric Variables: You can use `cor()` for a quick look at correlations between the numeric features and the target.

```
majority_class <- max(table(train_data$TARGET)) / nrow(train_data)
majority_class # This will give the accuracy of a simple majority class model
```

```
## [1] 0.9192712
```

2.3 Explore the Relationship Between Target and Predictors

2.3.1 Correlation with Numeric Variables: You can use `cor()` for a quick look at correlations between the numeric features and the target.

```
numeric_columns <- sapply(train_data, is.numeric)
cor_matrix <- cor(train_data[, numeric_columns], use = "complete.obs")
```

```
## Warning in cor(train_data[, numeric_columns], use = "complete.obs"): the
## standard deviation is zero
```

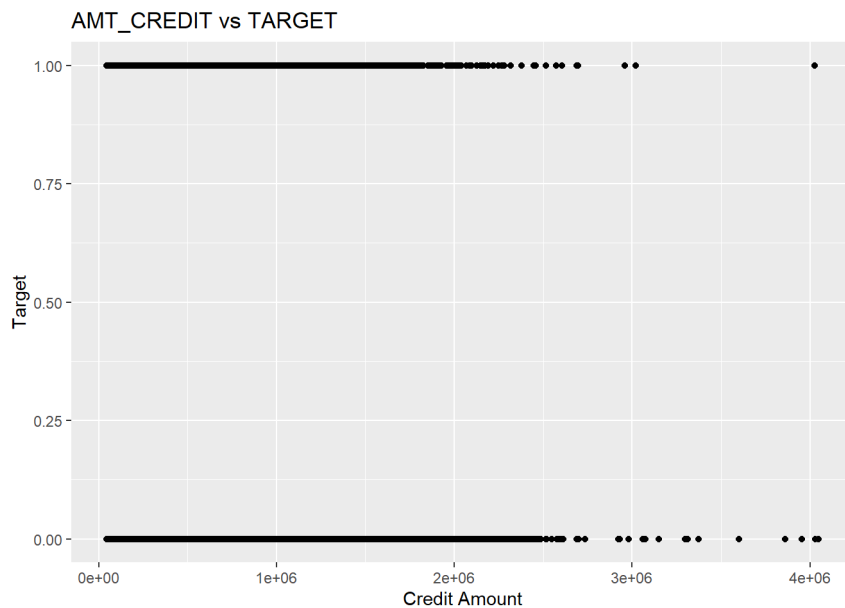
```
cor_target <- cor_matrix[, "TARGET"]
print(cor_target[order(-abs(cor_target))]) # Sorted by strength of correlation
```

##	TARGET	EXT_SOURCE_3
##	1.00000e+00	-1.586054e-01
##	EXT_SOURCE_2	EXT_SOURCE_1
##	-1.371169e-01	-1.351815e-01
##	REGION_RATING_CLIENT_W_CITY	REGION_RATING_CLIENT
##	6.426384e-02	5.623161e-02
##	FLAG_DOCUMENT_3	FLOORSMAX_AVG
##	5.239061e-02	-4.934060e-02
##	FLOORSMAX_MODE	FLOORSMAX_MEDI
##	-4.927142e-02	-4.880812e-02
##	DAYS_BIRTH	AMT_INCOME_TOTAL
##	4.618023e-02	-4.599431e-02
##	OWN_CAR_AGE	TOTALAREA_MODE
##	3.837052e-02	-3.311888e-02
##	AMT_REQ_CREDIT_BUREAU_YEAR	FLOORSMIN_MEDI
##	3.279824e-02	-3.254174e-02
##	FLOORSMIN_AVG	ELEVATORS_AVG
##	-3.252648e-02	-3.203765e-02
##	APARTMENTS_AVG	LIVINGAPARTMENTS_AVG
##	-3.185354e-02	-3.156523e-02
##	FLOORSMIN_MODE	LIVINGAREA_AVG
##	-3.123269e-02	-3.108370e-02
##	ELEVATORS_MEDI	APARTMENTS_MEDI
##	-3.057312e-02	-3.052623e-02
##	REGION_POPULATION_RELATIVE	LIVINGAPARTMENTS_MEDI
##	-3.027419e-02	-3.015265e-02
##	LIVINGAREA_MEDI	AMT_GOODS_PRICE
##	-3.013481e-02	-3.011508e-02
##	DAYS_ID_PUBLISH	LIVINGAREA_MODE
##	2.987001e-02	-2.915797e-02
##	FLAG_DOCUMENT_13	LIVINGAPARTMENTS_MODE
##	-2.890926e-02	-2.792125e-02
##	OBS_30_CNT_SOCIAL_CIRCLE	ELEVATORS_MODE
##	2.788881e-02	-2.751939e-02
##	OBS_60_CNT_SOCIAL_CIRCLE	APARTMENTS_MODE
##	2.750033e-02	-2.665493e-02
##	NONLIVINGAREA_MEDI	NONLIVINGAREA_AVG
##	-2.654524e-02	-2.613666e-02
##	YEARS_BUILD_AVG	YEARS_BUILD_MEDI
##	-2.561819e-02	-2.512312e-02
##	DEF_60_CNT_SOCIAL_CIRCLE	NONLIVINGAREA_MODE
##	2.412805e-02	-2.364367e-02
##	YEARS_BUILD_MODE	AMT_CREDIT
##	-2.284402e-02	-2.213439e-02
##	BASEMENTAREA_AVG	FLAG_DOCUMENT_16
##	-2.087791e-02	-1.965151e-02
##	DEF_30_CNT_SOCIAL_CIRCLE	CNT_FAM_MEMBERS
##	1.943179e-02	-1.929619e-02
##	BASEMENTAREA_MEDI	DAYS_LAST_PHONE_CHANGE
##	-1.921669e-02	1.790208e-02
##	COMMONAREA_AVG	COMMONAREA_MEDI
##	-1.776508e-02	-1.727416e-02
##	CNT_CHILDREN	HOURL_APPR_PROCESS_START
##	-1.633035e-02	-1.625102e-02
##	AMT_REQ_CREDIT_BUREAU_WEEK	AMT_REQ_CREDIT_BUREAU_DAY
##	1.616949e-02	1.515155e-02
##	FLAG_DOCUMENT_7	FLAG_WORK_PHONE
##	1.495330e-02	1.464951e-02
##	BASEMENTAREA_MODE	COMMONAREA_MODE
##	-1.460262e-02	-1.452439e-02
##	SK_ID_CURR	YEARS_BEGINEXPLUATATION_MEDI
##	1.451369e-02	-1.439711e-02
##	FLAG_DOCUMENT_8	FLAG_EMP_PHONE
##	-1.439409e-02	1.428056e-02
##	AMT_REQ_CREDIT_BUREAU_QRT	FLAG_PHONE
##	1.326654e-02	-1.310217e-02
##	FLAG_DOCUMENT_5	DAYS_EMPLOYED
##	-1.282136e-02	-1.259943e-02
##	REG_CITY_NOT_WORK_CITY	REG_REGION_NOT_LIVE_REGION
##	1.211448e-02	-1.100620e-02
##	YEARS_BEGINEXPLUATATION_AVG	ENTRANCES_AVG
##	-1.063228e-02	-1.058016e-02
##	ENTRANCES_MEDI	FLAG_DOCUMENT_18
##	-1.039053e-02	-9.594059e-03
##	LANDAREA_AVG	LANDAREA_MEDI
##	-9.462830e-03	-9.342955e-03
##	ENTRANCES_MODE	FLAG_DOCUMENT_19
##	-9.135605e-03	8.607177e-03
##	LANDAREA_MODE	FLAG_CONT_MOBILE
##	-8.097740e-03	8.001784e-03
##	REG_REGION_NOT_WORK_REGION	FLAG_DOCUMENT_6

```
##          -7.993280e-03          -7.917716e-03
##          AMT_ANNUITY    AMT_REQ_CREDIT_BUREAU_HOUR
##          -6.698003e-03          6.329620e-03
##          FLAG_DOCUMENT_11    NONLIVINGAPARTMENTS_MODE
##          -6.267669e-03          5.963353e-03
##    AMT_REQ_CREDIT_BUREAU_MON          FLAG_EMAIL
##          -5.736402e-03          5.313036e-03
##          FLAG_DOCUMENT_14    LIVE_CITY_NOT_WORK_CITY
##          4.620775e-03          4.403652e-03
##          DAYS_REGISTRATION          FLAG_DOCUMENT_21
##          4.309767e-03          -4.120458e-03
##    NONLIVINGAPARTMENTS_MEDI          FLAG_DOCUMENT_17
##          3.795607e-03          -3.364191e-03
##    NONLIVINGAPARTMENTS_AVG          FLAG_DOCUMENT_15
##          2.617824e-03          -2.595100e-03
##          FLAG_DOCUMENT_10          FLAG_DOCUMENT_4
##          -2.378738e-03          -2.378738e-03
##          FLAG_DOCUMENT_12          FLAG_DOCUMENT_20
##          -2.378738e-03          2.357059e-03
##    YEARS_BEGINEXPLUATATION_MODE    REG_CITY_NOT_LIVE_CITY
##          -1.659073e-03          8.729674e-04
##    LIVE_REGION_NOT_WORK_REGION          FLAG_DOCUMENT_9
##          -8.081022e-04          -9.695645e-05
##          FLAG_MOBIL          FLAG_DOCUMENT_2
##          NA          NA
```

2.3.2 Exploring Strong Predictors: You can also visualize potential strong predictors with scatter plots:

```
ggplot(train_data, aes(x = AMT_CREDIT, y = TARGET)) +
  geom_point() +
  labs(title = "AMT_CREDIT vs TARGET", x = "Credit Amount", y = "Target")
```



2.4 Missing Data Exploration

2.4.1 Setup to discover missing values

```
# Skim for a general overview
skim(train_data)
```

Data summary

Name	train_data
Number of rows	307511
Number of columns	122

Column type frequency:

character	16
numeric	106

Group variables

None

Variable type: character

skim_variable	n_missing	complete_rate	min	max	empty	n_unique	whitespace
NAME_CONTRACT_TYPE	0	1	10	15	0	2	0
CODE_GENDER	0	1	1	3	0	3	0
FLAG_OWN_CAR	0	1	1	1	0	2	0
FLAG_OWN_REALTY	0	1	1	1	0	2	0
NAME_TYPE_SUITE	0	1	0	15	1292	8	0
NAME_INCOME_TYPE	0	1	7	20	0	8	0
NAME_EDUCATION_TYPE	0	1	15	29	0	5	0
NAME_FAMILY_STATUS	0	1	5	20	0	6	0
NAME_HOUSING_TYPE	0	1	12	19	0	6	0
OCCUPATION_TYPE	0	1	0	21	96391	19	0
WEEKDAY_APPR_PROCESS_START	0	1	6	9	0	7	0
ORGANIZATION_TYPE	0	1	3	22	0	58	0
FONDKAPREMONT_MODE	0	1	0	21	210295	5	0
HOUSETYPE_MODE	0	1	0	16	154297	4	0
WALLSMATERIAL_MODE	0	1	0	12	156341	8	0
EMERGENCYSTATE_MODE	0	1	0	3	145755	3	0

Variable type: numeric

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
SK_ID_CURR	0	1.00	278180.52	102790.18	100002.00	189145.50	278202.00	367142.50	456255.00	█
TARGET	0	1.00	0.08	0.27	0.00	0.00	0.00	0.00	1.00	█
CNT_CHILDREN	0	1.00	0.42	0.72	0.00	0.00	0.00	1.00	19.00	█
AMT_INCOME_TOTAL	0	1.00	168797.92	237123.15	25650.00	112500.00	147150.00	202500.00	117000000.00	█
AMT_CREDIT	0	1.00	599026.00	402490.78	45000.00	270000.00	513531.00	808650.00	4050000.00	█
AMT_ANNUITY	12	1.00	27108.57	14493.74	1615.50	16524.00	24903.00	34596.00	258025.50	█
AMT_GOODS_PRICE	278	1.00	538396.21	369446.46	40500.00	238500.00	450000.00	679500.00	4050000.00	█
REGION_POPULATION_RELATIVE	0	1.00	0.02	0.01	0.00	0.01	0.02	0.03	0.07	█
DAYS_BIRTH	0	1.00	-16037.00	4363.99	-25229.00	-19682.00	-15750.00	-12413.00	-7489.00	█
DAYS_EMPLOYED	0	1.00	63815.05	141275.77	-17912.00	-2760.00	-1213.00	-289.00	365243.00	█
DAYS_REGISTRATION	0	1.00	-4986.12	3522.89	-24672.00	-7479.50	-4504.00	-2010.00	0.00	█
DAYS_ID_PUBLISH	0	1.00	-2994.20	1509.45	-7197.00	-4299.00	-3254.00	-1720.00	0.00	█
OWN_CAR_AGE	202929	0.34	12.06	11.94	0.00	5.00	9.00	15.00	91.00	█
FLAG_MOBIL	0	1.00	1.00	0.00	0.00	1.00	1.00	1.00	1.00	█
FLAG_EMP_PHONE	0	1.00	0.82	0.38	0.00	1.00	1.00	1.00	1.00	█
FLAG_WORK_PHONE	0	1.00	0.20	0.40	0.00	0.00	0.00	0.00	1.00	█
FLAG_CONT_MOBILE	0	1.00	1.00	0.04	0.00	1.00	1.00	1.00	1.00	█
FLAG_PHONE	0	1.00	0.28	0.45	0.00	0.00	0.00	1.00	1.00	█
FLAG_EMAIL	0	1.00	0.06	0.23	0.00	0.00	0.00	0.00	1.00	█
CNT_FAM_MEMBERS	2	1.00	2.15	0.91	1.00	2.00	2.00	3.00	20.00	█
REGION_RATING_CLIENT	0	1.00	2.05	0.51	1.00	2.00	2.00	2.00	3.00	█
REGION_RATING_CLIENT_W_CITY	0	1.00	2.03	0.50	1.00	2.00	2.00	2.00	3.00	█
HOUR_APPR_PROCESS_START	0	1.00	12.06	3.27	0.00	10.00	12.00	14.00	23.00	█
REG_REGION_NOT_LIVE_REGION	0	1.00	0.02	0.12	0.00	0.00	0.00	0.00	1.00	█
REG_REGION_NOT_WORK_REGION	0	1.00	0.05	0.22	0.00	0.00	0.00	0.00	1.00	█
LIVE_REGION_NOT_WORK_REGION	0	1.00	0.04	0.20	0.00	0.00	0.00	0.00	1.00	█
REG_CITY_NOT_LIVE_CITY	0	1.00	0.08	0.27	0.00	0.00	0.00	0.00	1.00	█
REG_CITY_NOT_WORK_CITY	0	1.00	0.23	0.42	0.00	0.00	0.00	0.00	1.00	█

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
LIVE_CITY_NOT_WORK_CITY	0	1.00	0.18	0.38	0.00	0.00	0.00	0.00	1.00	
EXT_SOURCE_1	173378	0.44	0.50	0.21	0.01	0.33	0.51	0.68	0.96	
EXT_SOURCE_2	660	1.00	0.51	0.19	0.00	0.39	0.57	0.66	0.85	
EXT_SOURCE_3	60965	0.80	0.51	0.19	0.00	0.37	0.54	0.67	0.90	
APARTMENTS_AVG	156061	0.49	0.12	0.11	0.00	0.06	0.09	0.15	1.00	
BASEMENTAREA_AVG	179943	0.41	0.09	0.08	0.00	0.04	0.08	0.11	1.00	
YEARS_BEGINEXPLUATATION_AVG	150007	0.51	0.98	0.06	0.00	0.98	0.98	0.99	1.00	
YEARS_BUILD_AVG	204488	0.34	0.75	0.11	0.00	0.69	0.76	0.82	1.00	
COMMONAREA_AVG	214865	0.30	0.04	0.08	0.00	0.01	0.02	0.05	1.00	
ELEVATORS_AVG	163891	0.47	0.08	0.13	0.00	0.00	0.00	0.12	1.00	
ENTRANCES_AVG	154828	0.50	0.15	0.10	0.00	0.07	0.14	0.21	1.00	
FLOORSMAX_AVG	153020	0.50	0.23	0.14	0.00	0.17	0.17	0.33	1.00	
FLOORSMIN_AVG	208642	0.32	0.23	0.16	0.00	0.08	0.21	0.38	1.00	
LANDAREA_AVG	182590	0.41	0.07	0.08	0.00	0.02	0.05	0.09	1.00	
LIVINGAPARTMENTS_AVG	210199	0.32	0.10	0.09	0.00	0.05	0.08	0.12	1.00	
LIVINGAREA_AVG	154350	0.50	0.11	0.11	0.00	0.05	0.07	0.13	1.00	
NONLIVINGAPARTMENTS_AVG	213514	0.31	0.01	0.05	0.00	0.00	0.00	0.00	1.00	
NONLIVINGAREA_AVG	169682	0.45	0.03	0.07	0.00	0.00	0.00	0.03	1.00	
APARTMENTS_MODE	156061	0.49	0.11	0.11	0.00	0.05	0.08	0.14	1.00	
BASEMENTAREA_MODE	179943	0.41	0.09	0.08	0.00	0.04	0.07	0.11	1.00	
YEARS_BEGINEXPLUATATION_MODE	150007	0.51	0.98	0.06	0.00	0.98	0.98	0.99	1.00	
YEARS_BUILD_MODE	204488	0.34	0.76	0.11	0.00	0.70	0.76	0.82	1.00	
COMMONAREA_MODE	214865	0.30	0.04	0.07	0.00	0.01	0.02	0.05	1.00	
ELEVATORS_MODE	163891	0.47	0.07	0.13	0.00	0.00	0.00	0.12	1.00	
ENTRANCES_MODE	154828	0.50	0.15	0.10	0.00	0.07	0.14	0.21	1.00	
FLOORSMAX_MODE	153020	0.50	0.22	0.14	0.00	0.17	0.17	0.33	1.00	
FLOORSMIN_MODE	208642	0.32	0.23	0.16	0.00	0.08	0.21	0.38	1.00	
LANDAREA_MODE	182590	0.41	0.06	0.08	0.00	0.02	0.05	0.08	1.00	
LIVINGAPARTMENTS_MODE	210199	0.32	0.11	0.10	0.00	0.05	0.08	0.13	1.00	
LIVINGAREA_MODE	154350	0.50	0.11	0.11	0.00	0.04	0.07	0.13	1.00	
NONLIVINGAPARTMENTS_MODE	213514	0.31	0.01	0.05	0.00	0.00	0.00	0.00	1.00	
NONLIVINGAREA_MODE	169682	0.45	0.03	0.07	0.00	0.00	0.00	0.02	1.00	
APARTMENTS_MEDI	156061	0.49	0.12	0.11	0.00	0.06	0.09	0.15	1.00	
BASEMENTAREA_MEDI	179943	0.41	0.09	0.08	0.00	0.04	0.08	0.11	1.00	
YEARS_BEGINEXPLUATATION_MEDI	150007	0.51	0.98	0.06	0.00	0.98	0.98	0.99	1.00	
YEARS_BUILD_MEDI	204488	0.34	0.76	0.11	0.00	0.69	0.76	0.83	1.00	
COMMONAREA_MEDI	214865	0.30	0.04	0.08	0.00	0.01	0.02	0.05	1.00	
ELEVATORS_MEDI	163891	0.47	0.08	0.13	0.00	0.00	0.00	0.12	1.00	
ENTRANCES_MEDI	154828	0.50	0.15	0.10	0.00	0.07	0.14	0.21	1.00	
FLOORSMAX_MEDI	153020	0.50	0.23	0.15	0.00	0.17	0.17	0.33	1.00	
FLOORSMIN_MEDI	208642	0.32	0.23	0.16	0.00	0.08	0.21	0.38	1.00	
LANDAREA_MEDI	182590	0.41	0.07	0.08	0.00	0.02	0.05	0.09	1.00	
LIVINGAPARTMENTS_MEDI	210199	0.32	0.10	0.09	0.00	0.05	0.08	0.12	1.00	
LIVINGAREA_MEDI	154350	0.50	0.11	0.11	0.00	0.05	0.07	0.13	1.00	
NONLIVINGAPARTMENTS_MEDI	213514	0.31	0.01	0.05	0.00	0.00	0.00	0.00	1.00	
NONLIVINGAREA_MEDI	169682	0.45	0.03	0.07	0.00	0.00	0.00	0.03	1.00	
TOTALAREA_MODE	148431	0.52	0.10	0.11	0.00	0.04	0.07	0.13	1.00	
OBS_30_CNT_SOCIAL_CIRCLE	1021	1.00	1.42	2.40	0.00	0.00	0.00	2.00	348.00	

skim_variable	n_missing	complete_rate	mean	sd	p0	p25	p50	p75	p100	hist
DEF_30_CNT_SOCIAL_CIRCLE	1021	1.00	0.14	0.45	0.00	0.00	0.00	0.00	34.00	█
OBS_60_CNT_SOCIAL_CIRCLE	1021	1.00	1.41	2.38	0.00	0.00	0.00	2.00	344.00	█
DEF_60_CNT_SOCIAL_CIRCLE	1021	1.00	0.10	0.36	0.00	0.00	0.00	0.00	24.00	█
DAYS_LAST_PHONE_CHANGE	1	1.00	-962.86	826.81	-4292.00	-1570.00	-757.00	-274.00	0.00	█
FLAG_DOCUMENT_2	0	1.00	0.00	0.01	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_3	0	1.00	0.71	0.45	0.00	0.00	1.00	1.00	1.00	█
FLAG_DOCUMENT_4	0	1.00	0.00	0.01	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_5	0	1.00	0.02	0.12	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_6	0	1.00	0.09	0.28	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_7	0	1.00	0.00	0.01	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_8	0	1.00	0.08	0.27	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_9	0	1.00	0.00	0.06	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_10	0	1.00	0.00	0.00	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_11	0	1.00	0.00	0.06	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_12	0	1.00	0.00	0.00	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_13	0	1.00	0.00	0.06	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_14	0	1.00	0.00	0.05	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_15	0	1.00	0.00	0.03	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_16	0	1.00	0.01	0.10	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_17	0	1.00	0.00	0.02	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_18	0	1.00	0.01	0.09	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_19	0	1.00	0.00	0.02	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_20	0	1.00	0.00	0.02	0.00	0.00	0.00	0.00	1.00	█
FLAG_DOCUMENT_21	0	1.00	0.00	0.02	0.00	0.00	0.00	0.00	1.00	█
AMT_REQ_CREDIT_BUREAU_HOUR	41519	0.86	0.01	0.08	0.00	0.00	0.00	0.00	4.00	█
AMT_REQ_CREDIT_BUREAU_DAY	41519	0.86	0.01	0.11	0.00	0.00	0.00	0.00	9.00	█
AMT_REQ_CREDIT_BUREAU_WEEK	41519	0.86	0.03	0.20	0.00	0.00	0.00	0.00	8.00	█
AMT_REQ_CREDIT_BUREAU_MON	41519	0.86	0.27	0.92	0.00	0.00	0.00	0.00	27.00	█
AMT_REQ_CREDIT_BUREAU_QRT	41519	0.86	0.27	0.79	0.00	0.00	0.00	0.00	261.00	█
AMT_REQ_CREDIT_BUREAU_YEAR	41519	0.86	1.90	1.87	0.00	0.00	1.00	3.00	25.00	█

```
# Checking missing values
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

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

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

```
train_data %>%
  summarise(across(everything(), ~ sum(is.na(.))))
```

```

## SK_ID_CURR TARGET NAME_CONTRACT_TYPE CODE_GENDER FLAG_OWN_CAR FLAG_OWN_REALTY
## 1 0 0 0 0 0 0
## CNT_CHILDREN AMT_INCOME_TOTAL AMT_CREDIT AMT_ANNUITY AMT_GOODS_PRICE
## 1 0 0 0 12 278
## NAME_TYPE_SUITE NAME_INCOME_TYPE NAME_EDUCATION_TYPE NAME_FAMILY_STATUS
## 1 0 0 0 0
## NAME_HOUSING_TYPE REGION_POPULATION_RELATIVE DAYS_BIRTH DAYS_EMPLOYED
## 1 0 0 0 0
## DAYS_REGISTRATION DAYS_ID_PUBLISH OWN_CAR_AGE FLAG_MOBIL FLAG_EMP_PHONE
## 1 0 0 202929 0 0
## FLAG_WORK_PHONE FLAG_CONT_MOBILE FLAG_PHONE FLAG_EMAIL OCCUPATION_TYPE
## 1 0 0 0 0 0
## CNT_FAM_MEMBERS REGION_RATING_CLIENT REGION_RATING_CLIENT_W_CITY
## 1 2 0 0
## WEEKDAY_APPR_PROCESS_START HOUR_APPR_PROCESS_START REG_REGION_NOT_LIVE_REGION
## 1 0 0 0
## REG_REGION_NOT_WORK_REGION LIVE_REGION_NOT_WORK_REGION REG_CITY_NOT_LIVE_CITY
## 1 0 0 0
## REG_CITY_NOT_WORK_CITY LIVE_CITY_NOT_WORK_CITY ORGANIZATION_TYPE EXT_SOURCE_1
## 1 0 0 0 173378
## EXT_SOURCE_2 EXT_SOURCE_3 APARTMENTS_AVG BASEMENTAREA_AVG
## 1 660 60965 150061 179943
## YEARS_BEGINEXPLUATATION_AVG YEARS_BUILD_AVG COMMONAREA_AVG ELEVATORS_AVG
## 1 150007 204488 214865 163891
## ENTRANCES_AVG FLOORSMAX_AVG FLOORSMIN_AVG LANDAREA_AVG LIVINGAPARTMENTS_AVG
## 1 154828 153020 208642 182590 210199
## LIVINGAREA_AVG NONLIVINGAPARTMENTS_AVG NONLIVINGAREA_AVG APARTMENTS_MODE
## 1 154350 213514 169682 156061
## BASEMENTAREA_MODE YEARS_BEGINEXPLUATATION_MODE YEARS_BUILD_MODE
## 1 179943 150007 204488
## COMMONAREA_MODE ELEVATORS_MODE ENTRANCES_MODE FLOORSMAX_MODE FLOORSMIN_MODE
## 1 214865 163891 154828 153020 208642
## LANDAREA_MODE LIVINGAPARTMENTS_MODE LIVINGAREA_MODE NONLIVINGAPARTMENTS_MODE
## 1 182590 210199 154350 213514
## NONLIVINGAREA_MODE APARTMENTS_MEDI BASEMENTAREA_MEDI
## 1 169682 156061 179943
## YEARS_BEGINEXPLUATATION_MEDI YEARS_BUILD_MEDI COMMONAREA_MEDI ELEVATORS_MEDI
## 1 150007 204488 214865 163891
## ENTRANCES_MEDI FLOORSMAX_MEDI FLOORSMIN_MEDI LANDAREA_MEDI
## 1 154828 153020 208642 182590
## LIVINGAPARTMENTS_MEDI LIVINGAREA_MEDI NONLIVINGAPARTMENTS_MEDI
## 1 210199 154350 213514
## NONLIVINGAREA_MEDI FONDKAPREMONT_MODE HOUSETYPE_MODE TOTALAREA_MODE
## 1 169682 0 0 148431
## WALLSMATERIAL_MODE EMERGENCYSTATE_MODE OBS_30_CNT_SOCIAL_CIRCLE
## 1 0 0 1021
## DEF_30_CNT_SOCIAL_CIRCLE OBS_60_CNT_SOCIAL_CIRCLE DEF_60_CNT_SOCIAL_CIRCLE
## 1 1021 1021 1021
## DAYS_LAST_PHONE_CHANGE FLAG_DOCUMENT_2 FLAG_DOCUMENT_3 FLAG_DOCUMENT_4
## 1 1 0 0 0
## FLAG_DOCUMENT_5 FLAG_DOCUMENT_6 FLAG_DOCUMENT_7 FLAG_DOCUMENT_8
## 1 0 0 0 0
## FLAG_DOCUMENT_9 FLAG_DOCUMENT_10 FLAG_DOCUMENT_11 FLAG_DOCUMENT_12
## 1 0 0 0 0
## FLAG_DOCUMENT_13 FLAG_DOCUMENT_14 FLAG_DOCUMENT_15 FLAG_DOCUMENT_16
## 1 0 0 0 0
## FLAG_DOCUMENT_17 FLAG_DOCUMENT_18 FLAG_DOCUMENT_19 FLAG_DOCUMENT_20
## 1 0 0 0 0
## FLAG_DOCUMENT_21 AMT_REQ_CREDIT_BUREAU_HOUR AMT_REQ_CREDIT_BUREAU_DAY
## 1 0 41519 41519
## AMT_REQ_CREDIT_BUREAU_WEEK AMT_REQ_CREDIT_BUREAU_MON
## 1 41519 41519
## AMT_REQ_CREDIT_BUREAU_QRT AMT_REQ_CREDIT_BUREAU_YEAR
## 1 41519 41519

```

```

# Using janitor for a cleaner view of missing data
train_data_clean <- train_data %>% remove_empty("cols")

```

```

# Removing columns with more than 50% missing values
train_data <- train_data[, colMeans(is.na(train_data)) < 0.5]

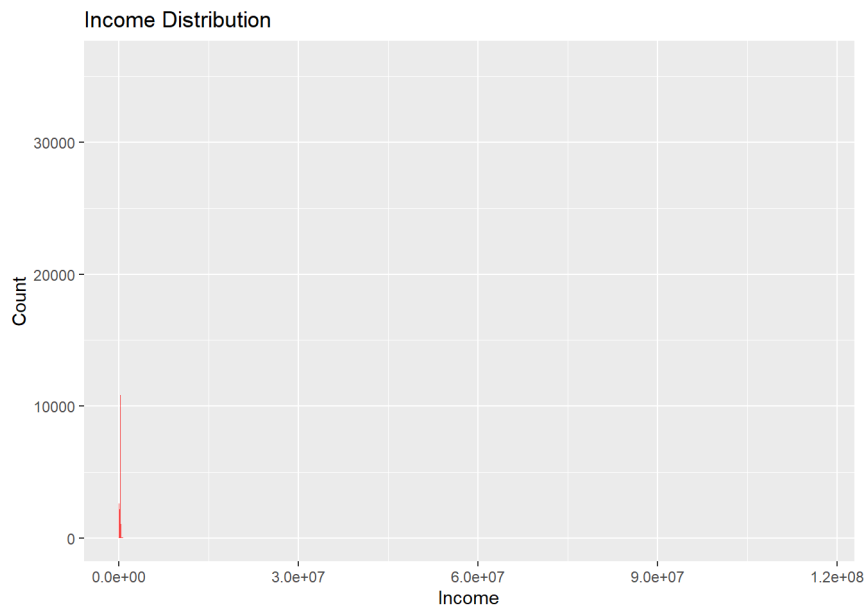
# Imputing with median
train_data$AMT_INCOME_TOTAL[is.na(train_data$AMT_INCOME_TOTAL)] <- median(train_data$AMT_INCOME_TOTAL, na.rm = TRUE)

```

```

ggplot(train_data, aes(x = AMT_INCOME_TOTAL)) +
  geom_histogram(binwidth = 5000, fill = "red", alpha = 0.7) +
  labs(title = "Income Distribution", x = "Income", y = "Count")

```



```
# Converting categorical variables into factors
train_data$CODE_GENDER <- as.factor(train_data$CODE_GENDER)

# For models requiring dummy variables
train_data <- model.matrix(~ CODE_GENDER + FLAG_OWN_CAR, data = train_data)
```

3 Additional Questions to Guide Exploration

Some questions you can explore during EDA include:

- Is the data balanced or imbalanced with respect to the target variable (loan default)?
- What are the relationships between key features (e.g., loan amount, income) and the target variable?
- Which features are the strongest predictors of loan default?
- Are there any significant data quality issues, such as missing or outlier values?
- Does the data need preprocessing, such as normalization or encoding of categorical variables?
- How are demographic variables (e.g., age, family status) related to loan default?
- What trends can be seen with credit amount, loan annuity, and income total?

4 Data Description

Your dataset includes information about loan applicants, such as demographic information (e.g., age, gender), financial information (e.g., income, credit amount), and loan application details. The file `application_train.csv` is the primary dataset, and additional data files (such as `bureau.csv`) can be used to enrich the analysis by joining based on `SK_ID_CURR`.

Summary of the Data:

- The dataset contains both numeric and categorical variables.
- Key variables include `AMT_CREDIT`, `AMT_INCOME_TOTAL`, `AMT_ANNUITY`, and `DAYS_BIRTH`.
- The target variable is `TARGET`, which indicates loan repayment status.

5 Missing Data

During the data exploration, you might encounter missing values in several columns. For instance, columns related to property or financial information may have missing values for some applicants. Addressing missing data is crucial for accurate model development.

Proposed Solutions:

- For columns with a significant portion of missing data (e.g., more than 50%), consider dropping these columns.
- For columns with fewer missing values, consider imputing with median or mean values (for numerical columns) or the most frequent value (for categorical columns).
- You can also consider more sophisticated imputation methods like k-nearest neighbors or predictive modeling.

6 Findings and Discussion

Based on the initial analysis:

- The target variable (`TARGET`) is imbalanced, with a majority of applicants repaying their loans on time and a smaller proportion defaulting.
- Certain variables, such as `AMT_CREDIT` and `AMT_INCOME_TOTAL`, show potential for being strong predictors of loan default, based on their distribution and correlation with the target variable.

- There are some data quality issues, including missing values in multiple columns, especially those related to property ownership and financial records.
- The exploratory visualizations suggest that applicants with lower income or higher loan amounts might be at greater risk of defaulting on their loans.