

Fraudulent or not?

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Introduction

The goal in this project is to learn how to predict a fraudulent financial transaction. The data used here is called Synthetic Financial Datasets for Fraud Detection generated by the PaySim mobile money simulator (<https://www.kaggle.com/ntnu-testimon/paysim1>). As described on the web page, the dataset is a synthetic one, generated using the simulator called PaySim. It uses aggregated data from a private dataset to generate a synthetic dataset that resembles the normal operation of transactions and injects malicious behaviour.

PaySim simulates mobile money transactions based on a sample of real transactions extracted from one month of financial logs from a mobile money service implemented in an African country. The synthetic dataset is scaled down 1/4 of the original dataset.

I have downloaded the dataset from the net (the link above) and I have unzipped it to the same folder where my R script and the rmd file are. Here, I am reading the data from my folder.

```
<<<<<<< HEAD
```

The dataset, here referred with a variable name `fraud_or_not`, has the following dimensions

```
## Parsed with column specification:
## cols(
##   step = col_double(),
##   type = col_character(),
##   amount = col_double(),
##   nameOrig = col_character(),
##   oldbalanceOrg = col_double(),
##   newbalanceOrg = col_double(),
##   nameDest = col_character(),
##   oldbalanceDest = col_double(),
##   newbalanceDest = col_double(),
##   isFraud = col_double(),
##   isFlaggedFraud = col_double()
## )
## [1] 6362620      11
```

Next I will analyse the data and split it to training and test sets. I will use different machine learning algorithms to try to predict which transaction is fraudulent and which not. In this kind of a case the speciality is that the amount of fraudulent transactions is very minor compared to the amount of non-fraudulent transactions, as we will see.

Analysis

Let's look the data first as is. Like can be seen from the summary below, there are e.g. no NA values which would need to be cleaned.

```
summary(fraud_or_not)
```

```
##      step      type      amount      nameOrig
## Min.   : 1.0 Length:6362620 Min.    : 0 Length:6362620
## 1st Qu.:156.0 Class :character 1st Qu.: 13390 Class :character
## Median :239.0 Mode  :character Median : 74872 Mode  :character
```

```
## Mean :243.4 Mean : 179862
## 3rd Qu.:335.0 3rd Qu.: 208721
## Max. :743.0 Max. :92445517
## oldbalanceOrig newbalanceOrig nameDest
## Min. : 0 Min. : 0 Length:6362620
## 1st Qu.: 0 1st Qu.: 0 Class :character
## Median : 14208 Median : 0 Mode :character
## Mean : 833883 Mean : 855114
## 3rd Qu.: 107315 3rd Qu.: 144258
## Max. :59585040 Max. :49585040
## oldbalanceDest newbalanceDest isFraud
## Min. : 0 Min. : 0 Min. :0.000000
## 1st Qu.: 0 1st Qu.: 0 1st Qu.:0.000000
## Median : 132706 Median : 214661 Median :0.000000
## Mean : 1100702 Mean : 1224996 Mean :0.001291
## 3rd Qu.: 943037 3rd Qu.: 1111909 3rd Qu.:0.000000
## Max. :356015889 Max. :356179279 Max. :1.000000
## isFlaggedFraud
## Min. :0.0e+00
## 1st Qu.:0.0e+00
## Median :0.0e+00
## Mean :2.5e-06
## 3rd Qu.:0.0e+00
## Max. :1.0e+00
```

```
str(fraud_or_not)
```

```
## Classes 'spec_tbl_df', 'tbl_df', 'tbl' and 'data.frame': 6362620 obs. of 11 variables:
## $ step : num 1 1 1 1 1 1 1 1 1 1 ...
## $ type : chr "PAYMENT" "PAYMENT" "TRANSFER" "CASH_OUT" ...
## $ amount : num 9840 1864 181 181 11668 ...
## $ nameOrig : chr "C1231006815" "C1666544295" "C1305486145" "C840083671" ...
## $ oldbalanceOrig : num 170136 21249 181 181 41554 ...
## $ newbalanceOrig: num 160296 19385 0 0 29886 ...
## $ nameDest : chr "M1979787155" "M2044282225" "C553264065" "C38997010" ...
## $ oldbalanceDest: num 0 0 0 21182 0 ...
## $ newbalanceDest: num 0 0 0 0 0 ...
## $ isFraud : num 0 0 1 1 0 0 0 0 0 0 ...
## $ isFlaggedFraud: num 0 0 0 0 0 0 0 0 0 0 ...
## - attr(*, "spec")=
## .. cols(
## .. step = col_double(),
## .. type = col_character(),
## .. amount = col_double(),
## .. nameOrig = col_character(),
## .. oldbalanceOrig = col_double(),
## .. newbalanceOrig = col_double(),
## .. nameDest = col_character(),
## .. oldbalanceDest = col_double(),
## .. newbalanceDest = col_double(),
## .. isFraud = col_double(),
## .. isFlaggedFraud = col_double()
## .. )
```

```
fraud_or_not %>% head()
```

```
## # A tibble: 6 x 11
##   step type amount nameOrig oldbalanceOrg newbalanceOrig nameDest
##   <dbl> <chr> <dbl> <chr>          <dbl>          <dbl> <chr>
## 1     1 PAYM~  9840. C123100~        170136        160296. M197978~
## 2     1 PAYM~  1864. C166654~        21249         19385. M204428~
## 3     1 TRAN~   181 C130548~          181           0 C553264~
## 4     1 CASH~   181 C840083~          181           0 C389970~
## 5     1 PAYM~ 11668. C204853~        41554        29886. M123070~
## 6     1 PAYM~  7818. C900456~        53860        46042. M573487~
## # ... with 4 more variables: oldbalanceDest <dbl>, newbalanceDest <dbl>,
## #   isFraud <dbl>, isFlaggedFraud <dbl>
```

The data has 11 columns which are:

Table 1: Explanations of the features

feature	expl
step	maps a unit of time in the real world. 1 step is 1 hour of time. Total steps 744 (30 days simulation).
type	CASH-IN, CASH-OUT, DEBIT, PAYMENT and TRANSFER.
amount	amount of the transaction in local currency.
nameOrig	customer who started the transaction
oldbalanceOrg	initial balance before the transaction
newbalanceOrig	new balance after the transaction
nameDest	customer who is the recipient of the transaction
oldbalanceDest	initial balance recipient before the transaction
newbalanceDest	new balance recipient after the transaction
isFraud	Transactions made by the fraudulent agents inside the simulation
isFlaggedFraud	An illegal attempt in this dataset is an attempt to transfer more than 200.000 in a single transaction.

Results

Conclusion