# CREDIT EDA

ASSIGNMENT

### PROBLEM STATEMENT

The loan providing companies find it hard to give loans to the people due to their insufficient or non-existent credit history. Because of that, some consumers use it to their advantage by becoming a defaulter. Suppose you work for a consumer finance company which specialises in lending various types of loans to urban customers. You have to use EDA to analyse the patterns present in the data. This will ensure that the applicants capable of repaying the loan are not rejected.

When the company receives a loan application, the company has to decide for loan approval based on the applicants profile. Two types of risks are associated with the banks decision:

- ▶ If the applicant is likely to repay the loan, then not approving the loan results in a loss of business to the company
- ▶ If the applicant is not likely to repay the loan, i.e. he/she is likely to default, then approving the loan may lead to a financial loss for the company.

### PROBLEM STATEMENT

The data given below contains the information about the loan application at the time of applying for the loan. It contains two types of scenarios:

- ► The client with payment difficulties: he/she had late payment more than X days on at least one of the first Y instalments of the loan in our sample,
- ▶ All other cases: All other cases when the payment is paid on time.

# **Business Objectives**

This case study aims to identify patterns which indicate if a client has difficulty paying their instalments which may be used for taking actions such as denying the loan, reducing the amount of loan, lending (to risky applicants) at a higher interest rate, etc. This will ensure that the consumers capable of repaying the loan are not rejected. Identification of such applicants using EDA is the aim of this case study.

In other words, the company wants to understand the driving factors (or driver variables) behind loan default, i.e. the variables which are strong indicators of default. The company can utilise this knowledge for its portfolio and risk assessment.

To develop your understanding of the domain, you are advised to independently research a little about risk analytics - understanding the types of variables and their significance should be enough.

# Assumptions

- Dropping columns have null values greater than or equal to 30%.
- Dropping below mentioned columns

```
FLAG DOCUMENT 13
EXT SOURCE 2
                                FLAG DOCUMENT 14
EXT SOURCE 3FLAG DOCUMENT 2
                                FLAG DOCUMENT 15
FLAG DOCUMENT 3
                                FLAG DOCUMENT 16
FLAG DOCUMENT 4
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FLAG DOCUMENT 5
                                FLAG_DOCUMENT_18
FLAG DOCUMENT 6
                                FLAG DOCUMENT 19
                                FLAG DOCUMENT 20
FLAG DOCUMENT 7
                                FLAG DOCUMENT 21
FLAG DOCUMENT 8
                               AMT_REQ_CREDIT_BUREAU_HOUR
FLAG DOCUMENT 9
                               AMT REQ CREDIT BUREAU DAY
FLAG DOCUMENT 10
                               AMT_REQ_CREDIT_BUREAU_WEEK
FLAG DOCUMENT 11
                               AMT REQ CREDIT BUREAU MON
FLAG DOCUMENT 12
                               AMT REQ CREDIT BUREAU QRT
FLAG DOCUMENT 13
                               AMT_REQ_CREDIT_BUREAU_YEAR
```

- ▶ Filled negative employment values in Employment (in Yrs) column with null
- Filled 'XNA' in CODE\_GENDER column with null

# Approach

Importing the csv file of application data and previous application data Understanding the column description file

Checking the structure of the data

Missing value check

Handling missing values

- Dropping columns have null values greater than or equal to 30%.
- Dropping unnecessary columns.
- Dropping rows containing missing values.

# Approach

#### **Checking and Handling Outlier**

Removing row having salary greater than 100000.

#### Performing Univariate and Bivariate Analysis

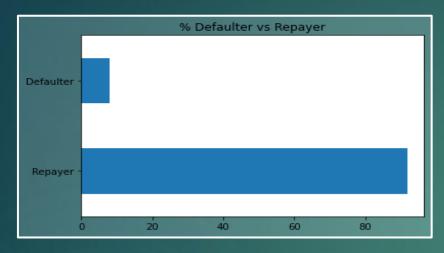
Analyzing the data using bar chart, pie chart and line chart.

#### **Performing Multivariate Analysis**

Analyzing using bar chart, pie chart, scatter plot, pair plot and heatmap.

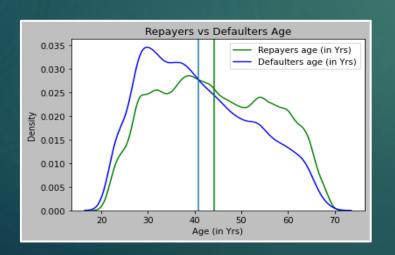
#### **Analyzing Previous application data**

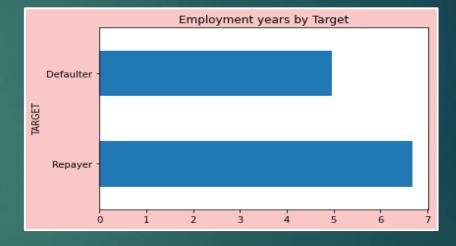
### **RESULTS**



More than 90% of the Clients are repayers

Average of employment years is less in case of defaulters as compared to repayers

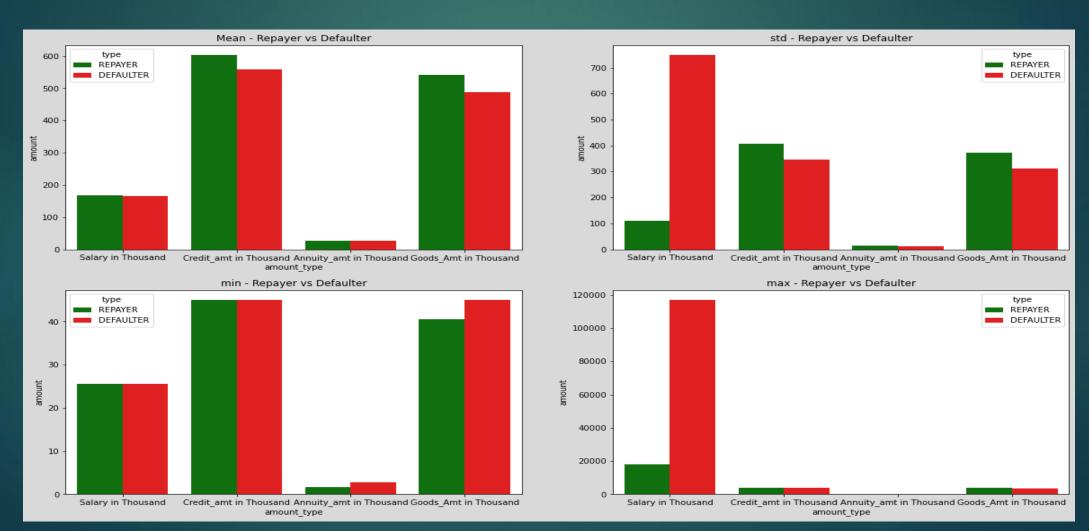




- Maximum defaulters have the age between 28 to 33
- Average age of defaulters is less as compared to repayers
- Defaulters tend to decrease as the age increases

### **RESULTS**

- Average income of defaulter and non-defaulter client are same
- Standard deviation of income of defaulter is very high as compared to non-defaulter
- Max income of defaulter is very high as compared to non-defaulter

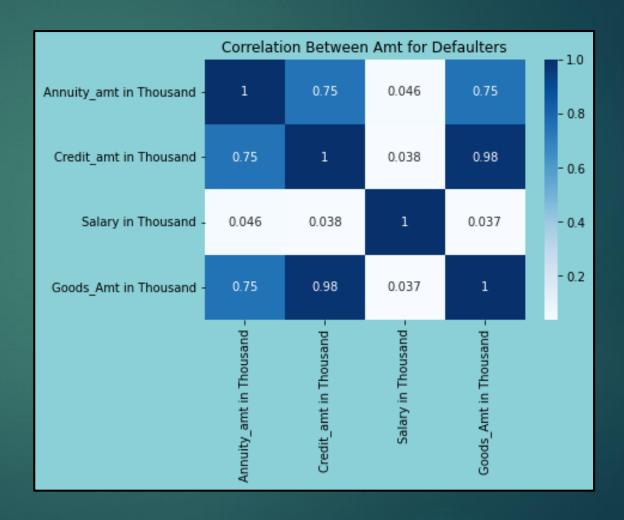


## Results

There is a strong positive linear relationship between credit amount and annuity amount for defaulters

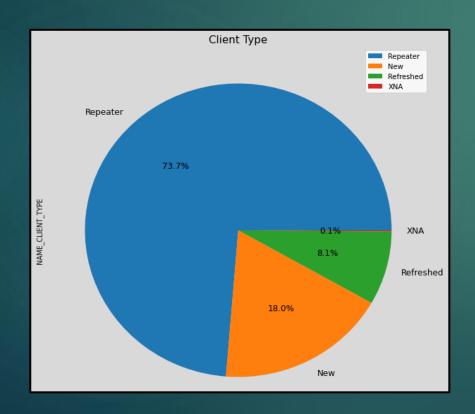
There is a strong positive linear relationship between credit amount and goods amount for defaulters

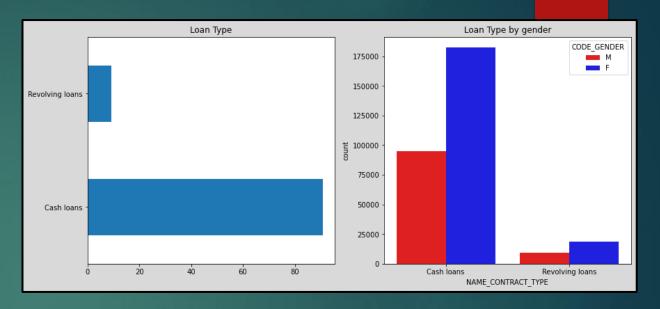
There is a strong positive linear relationship between goods amount and annuity amount for defaulters



## Results

- Majority of the clients prefer cash loans
- Both the gender prefer cash loans over revolving loans
- Female prefers cash loan more as compared to male

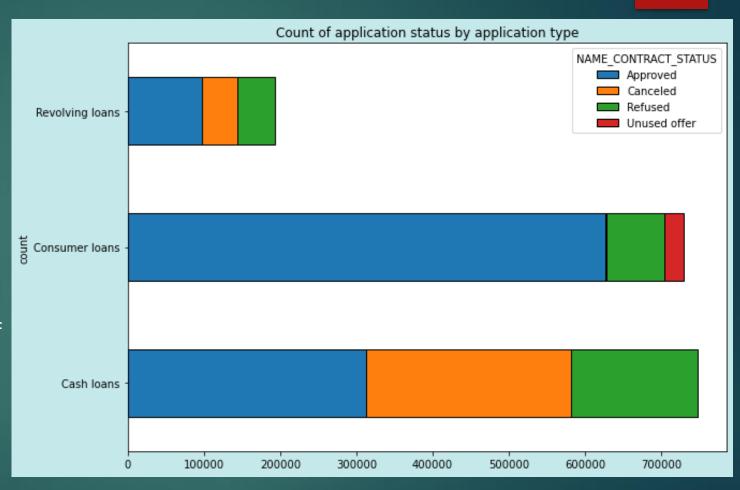




73.7% Clients are old clients whereas 18% Clients are new

## RESULTS

- Most Approved loans are Consumer loans
- Most Canceled and Refused loans are Cash loans
- Unused offer is more in case of Consumer loans



### CONCLUSION

- ▶ There is a high chance of refusal for the customers applying the loan for the purpose of paying the other loan.
- Average salary for defaulters is less as compared to nondefaulters.
- Self-employed customers have a high chance of defaulting due to non-stable income