# CREDIT EDA CASE STUDY

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#### AGENDA

- Give an idea of applying EDA in a real business scenario.
- Develop a basic understanding of risk analytics in banking and financial services
- Understand how data is used to minimise the risk of losing money while lending to customers

#### **Business Understanding**

When the company receives a loan application, the company has to decide for loan approval based on the applicant's profile. Two types of risks are associated with the bank's 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.

#### TARGET VARIABLE

The dataset 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.

#### TYPES OF DECISION

When a client applies for a loan, there are four types of decisions that could be taken by the client/company

- **Approved:** The Company has approved loan Application
- **Cancelled:** The client cancelled the application sometime during approval. Either the client changed her/his mind about the loan or in some cases due to a higher risk of the client he received worse pricing which he did not want.
- **Refused:** The company had rejected the loan (because the client does not meet their requirements etc.).
- **Unused offer:** Loan has been cancelled by the client but on different stages of the process.

#### PURPOSE OF CASE STUDY

 Using EDA techniques analyse the dataset to understand how consumer attributes and loan attributes influence the tendency of default

#### DATA UNDERSTANDING

- 'application\_data.csv' contains all the information of the client at the time of application.
  - The data is about whether a client has payment difficulties.
- 'previous\_application.csv' contains information about the client's previous loan data. It contains the data whether the previous application had been Approved, Cancelled, Refused or Unused offer.

#### PROBLEM STATEMENT

- Identify patterns which indicate if a client has difficulty paying their installments
- Ensuring consumers capable of repaying the loan are not rejected
- Find out client variables and loan variables that are high indicators of defaulting

### Exploratory Data Analysis Approach

The dataset is being analysed using following steps:

- Data Sourcing
- Data Cleansing
- Derived Metrics
- Univariate Analysis
- Segmented Univariate Analysis
- Bivariate Analysis
- Correlation Analysis
- Arriving at Insights

### Data Sourcing

- 'application\_data.csv' and 'previous\_application.csv' datasets are meged on current application ID
- The combined dataset is used for analysis
- The columns are divided into numerical and catrgorical for ease of analysis

### Data Cleansing

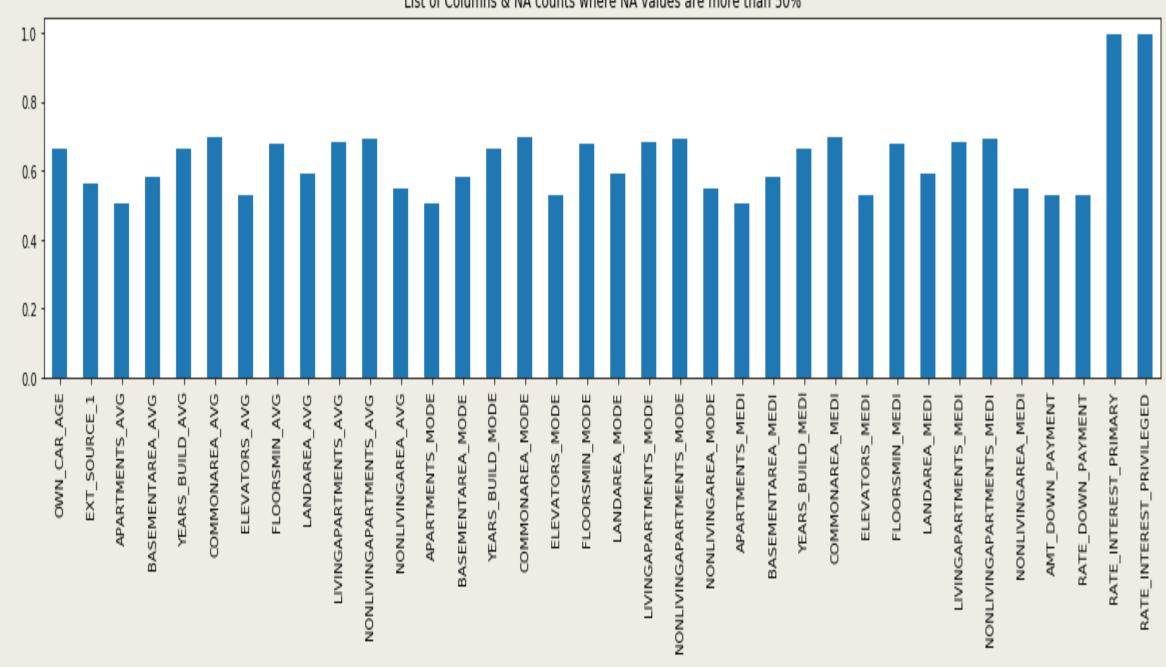
- **■** Fixing Rows and Columns
- Dealing with missing values
- Detecting Outliers
- Finding Data Imbalance

#### Fixing Rows and Columns

- Duplicate Rows and Columns are removed
- Certain columns are renamed
- Datatype of certain columns changed to category
- Columns segregated based on datatype as numerical, object and categorical

#### Dealing with missing values

- Null values in categorical columns replaced with appropriate values
- Numerical Columns with more than 50% missing values identified
- Insignificant columns are removed
- RATE\_INTEREST\_PRIMARY and RATE\_INTEREST\_PRIVILEGED removed



#### **Detecting Outliers**

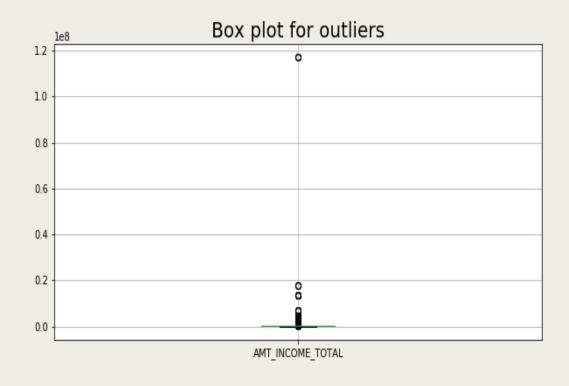
- Numerical columns are analysed for identifying outliers using Box plots
- Outlier Datapoints are detected using IQR method
- Datapoints that are beyond 1.5 times Inter Quartile Regions are considered as outliers
- Rows with Significant Outliers for particular variables are dropped

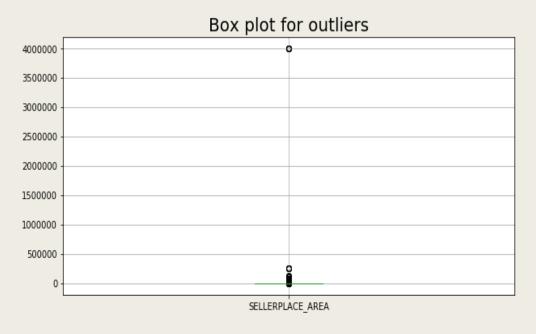
#### Detected and Removed Outliers

SELLERPLACE\_AREA:[4000000]

AMT\_INCOME\_TOTAL: [117000000.0, 18000090.0, 13500000.0]

#### Box Plots for outliers





#### Pseudo Outliers

Highly Significant outliers were found as follows:

DAYS\_EMPLOYED:[365243]

DAYS\_FIRST\_DUE:[365243.0]

DAYS\_LAST\_DUE\_1ST\_VERSION:[365243.0]

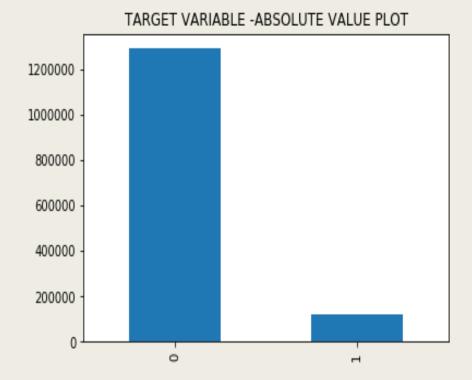
DAYS\_LAST\_DUE:[365243.0]

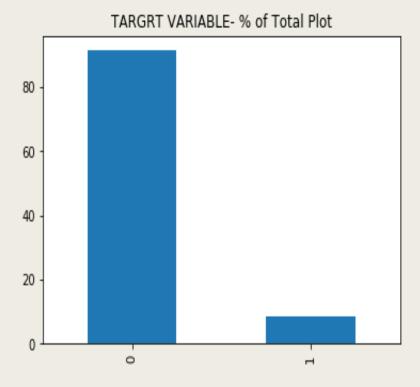
DAYS\_TERMINATION:[365243.0]

Since the value is spread over many rows, it is assumed as indication of unemployment or other category

#### Data Imbalance

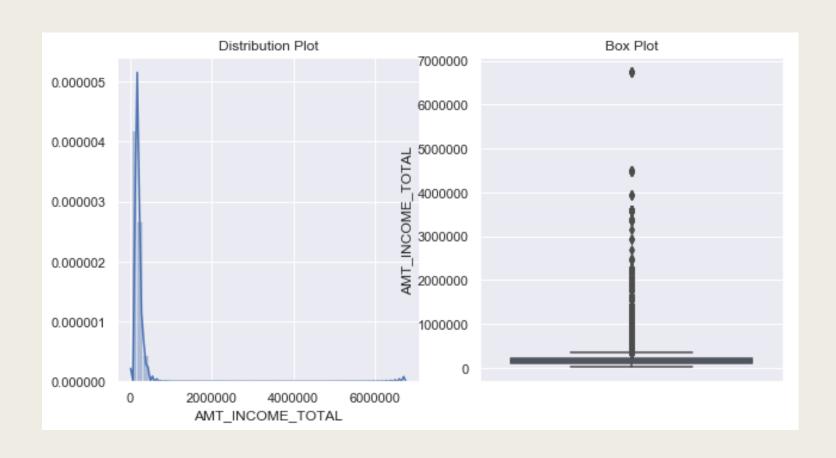
- Dataset is segmented into Repayers with Target=0 and Defaulters with Target=1
- Countplot of Repayers and Defaulters indicates DATA IMBALANCE
- Count of Repayers = 1291326
- Count of Defaulters =122357
- Ratio of Data Imbalance is 1291326:122357 ie. 10.554:1



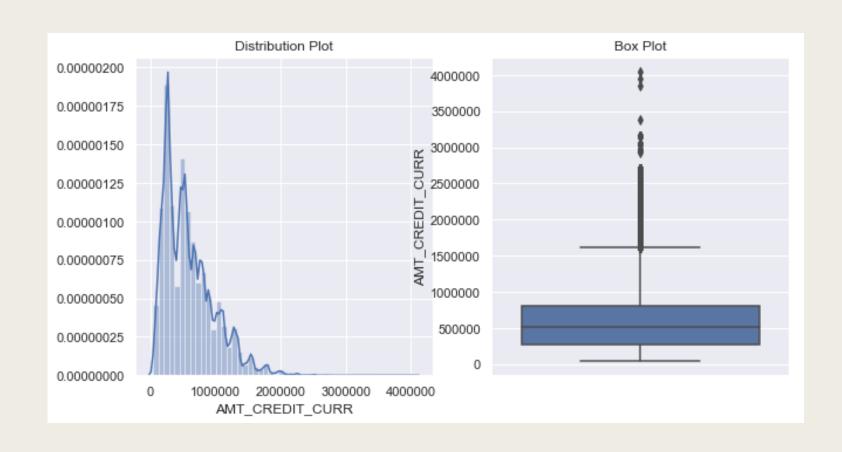


## Univariate Analysis

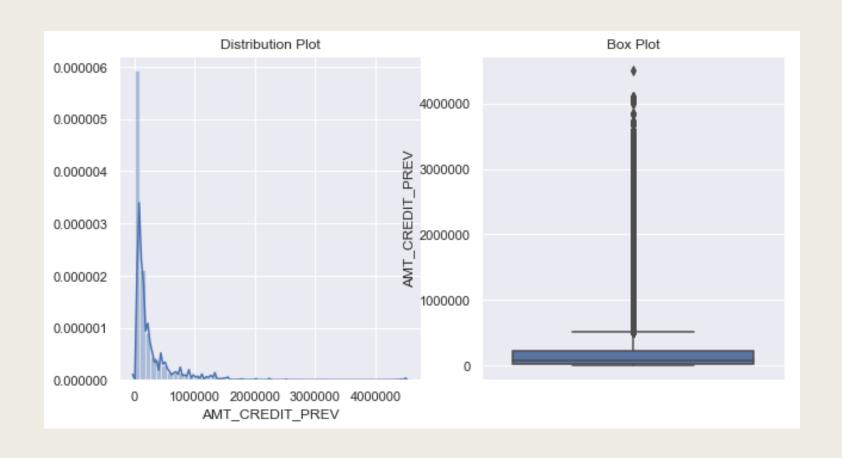
#### TOTAL INCOME



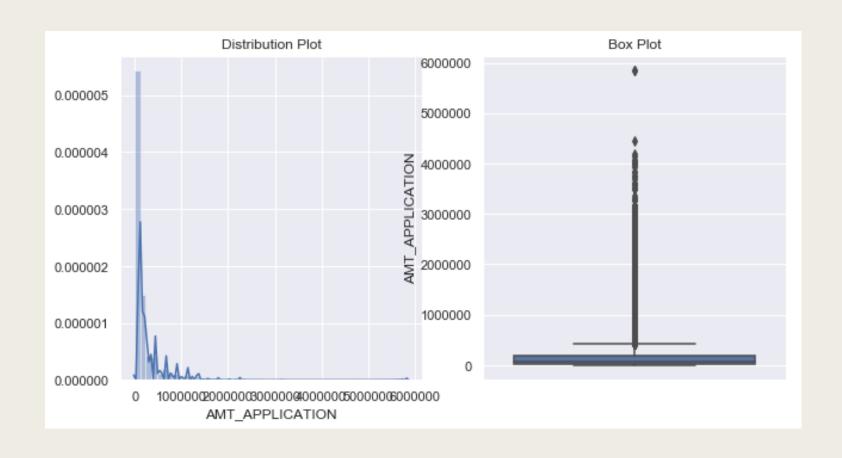
#### **CURRENT CREDIT AMOUNT**



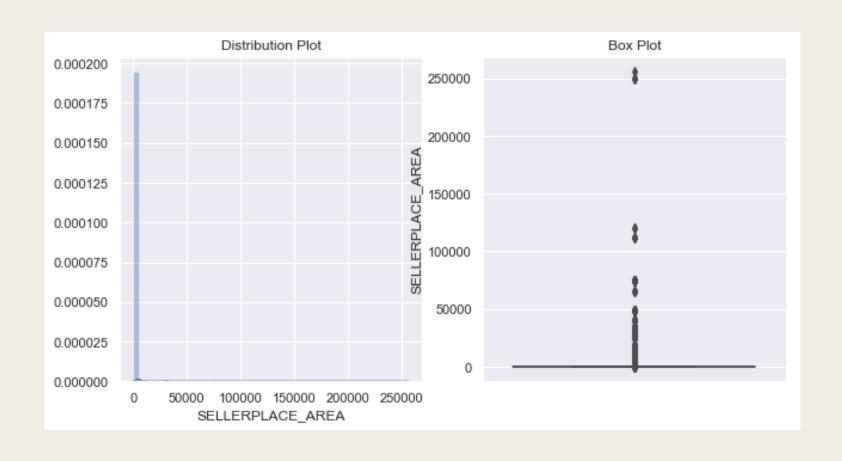
#### PREVIOUS CREDIT AMOUNT



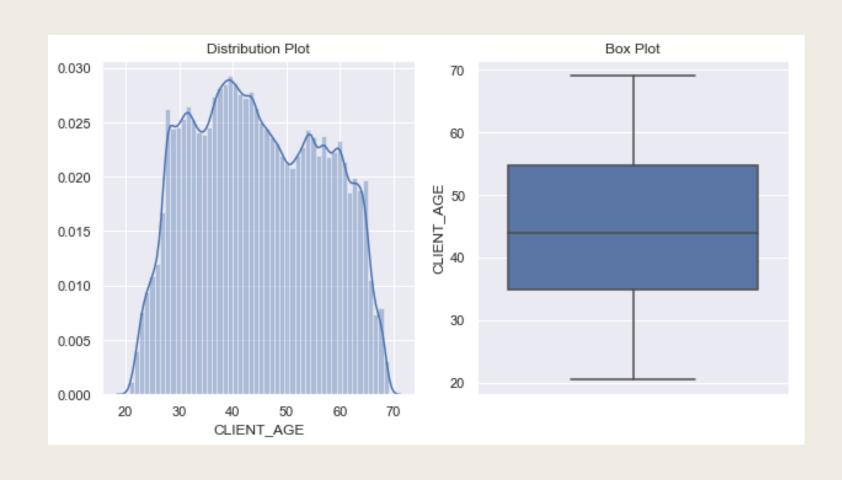
#### APPLICATION AMOUNT



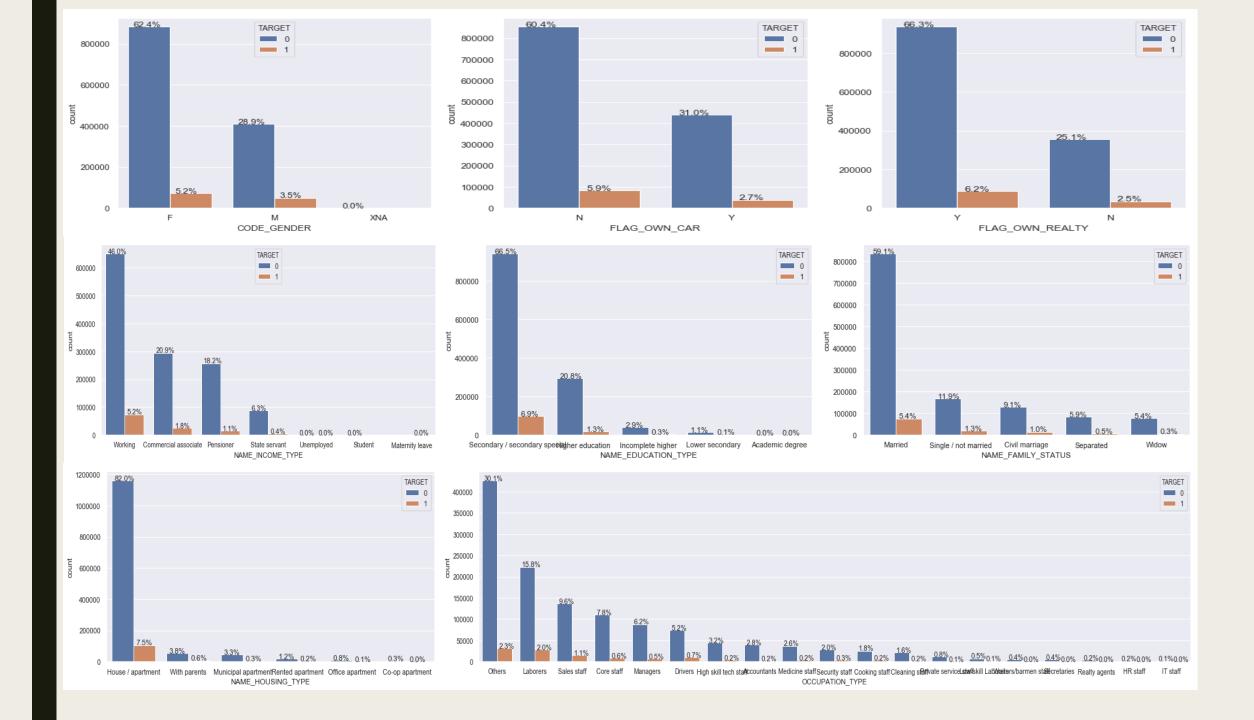
#### SELLER PLACE AREA

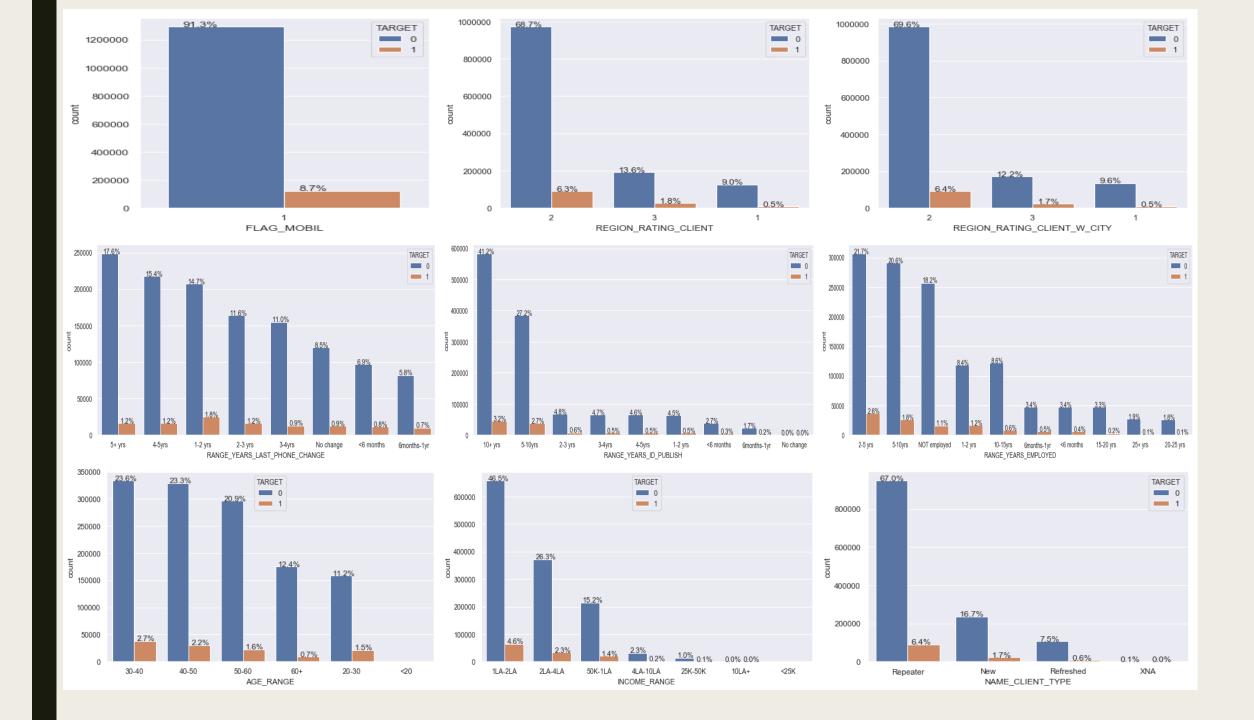


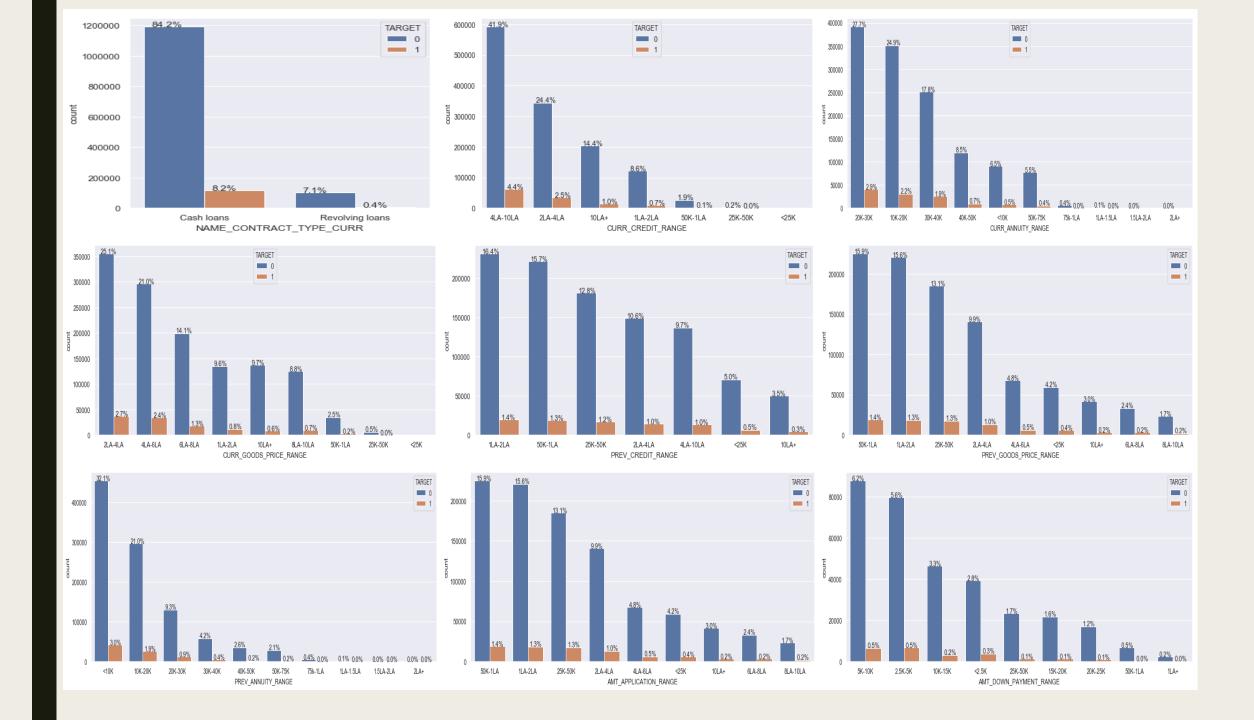
#### CLIENT AGE

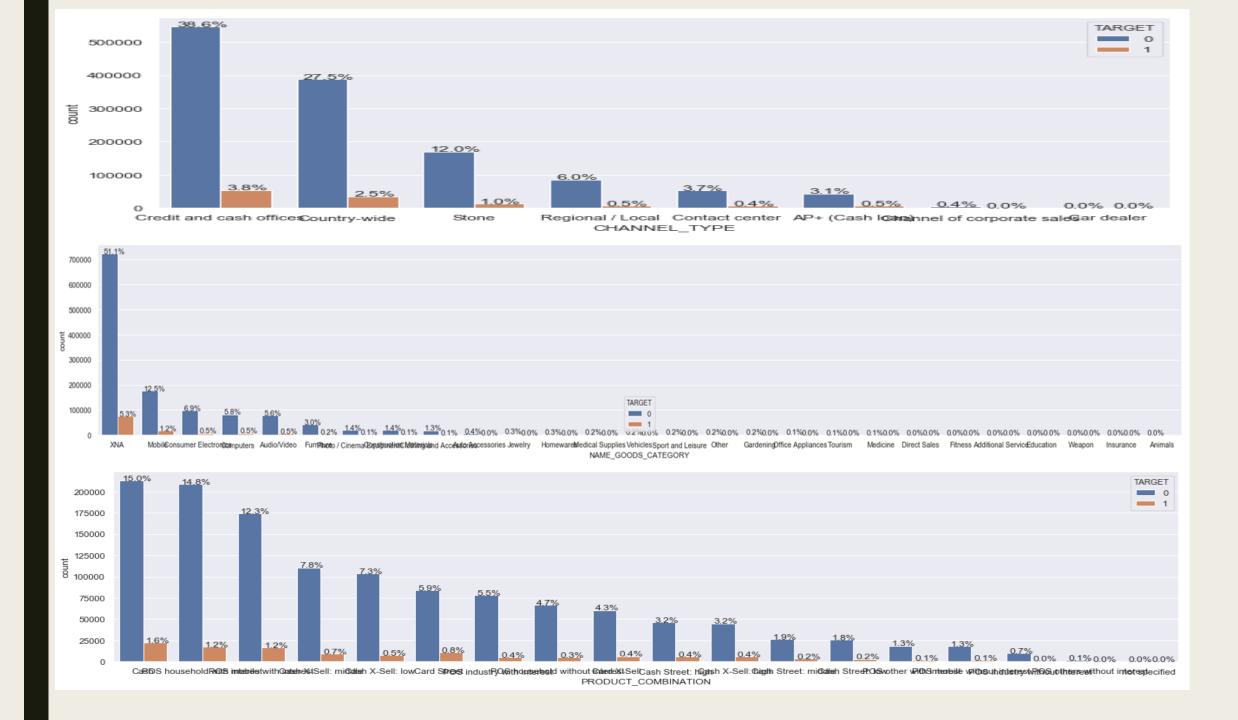


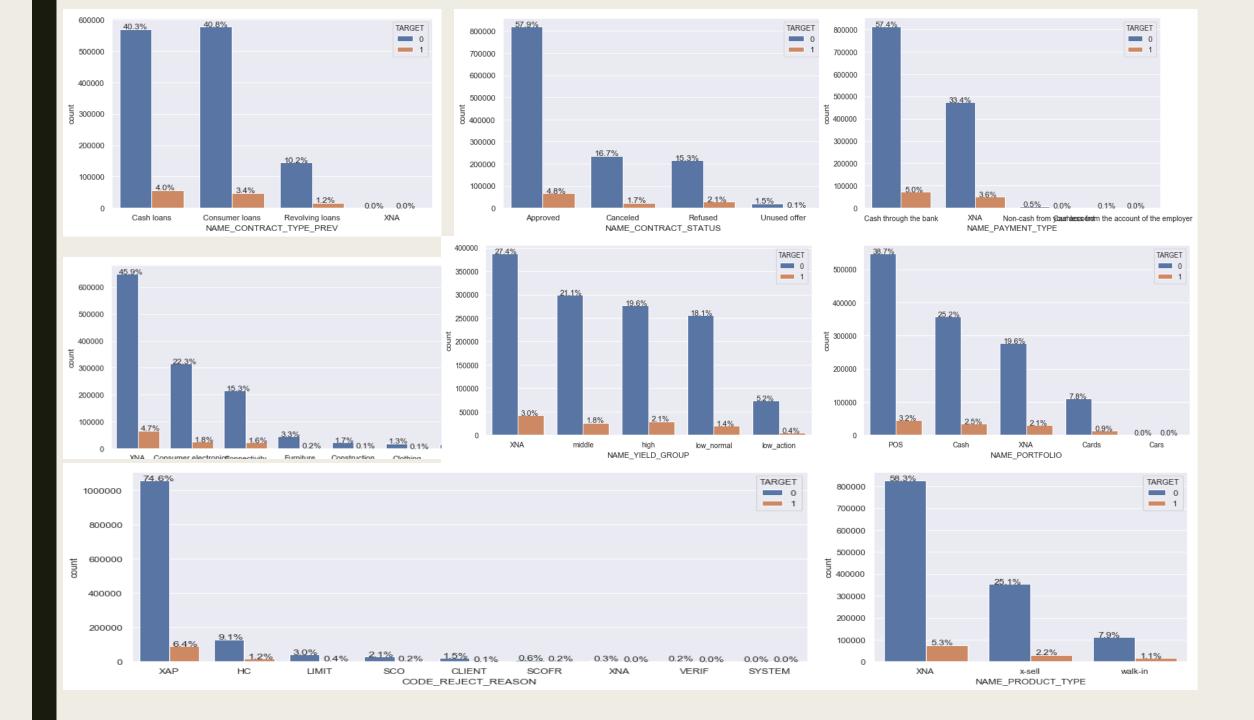
## Segmented Univariate Analysis





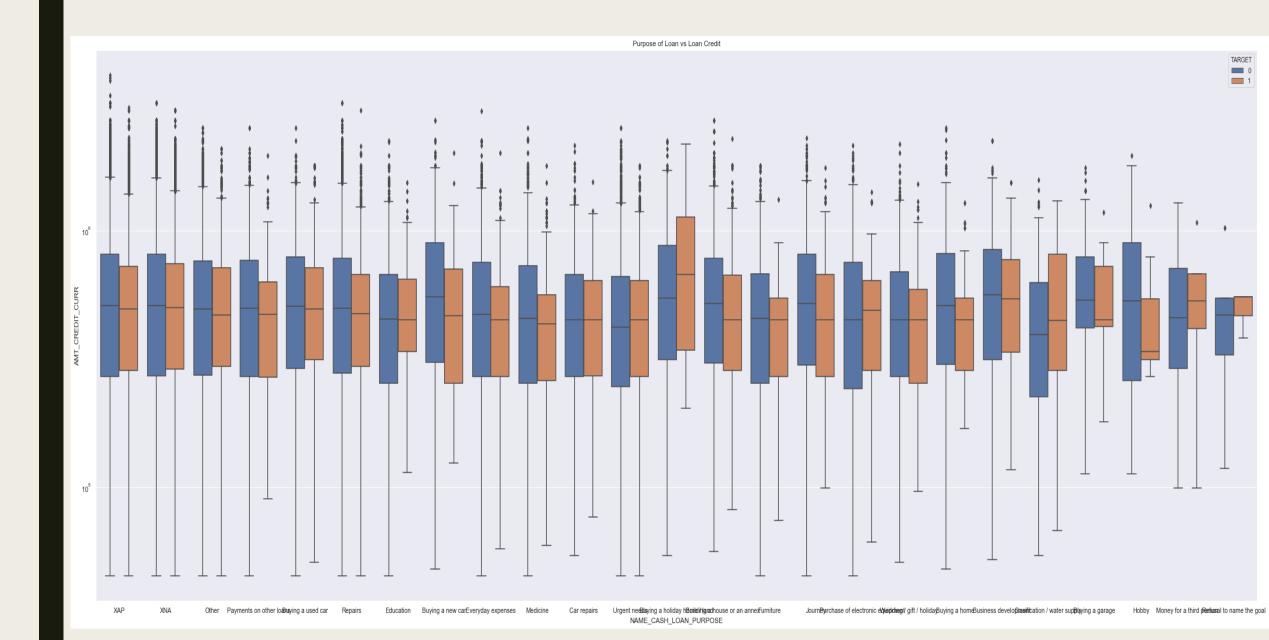




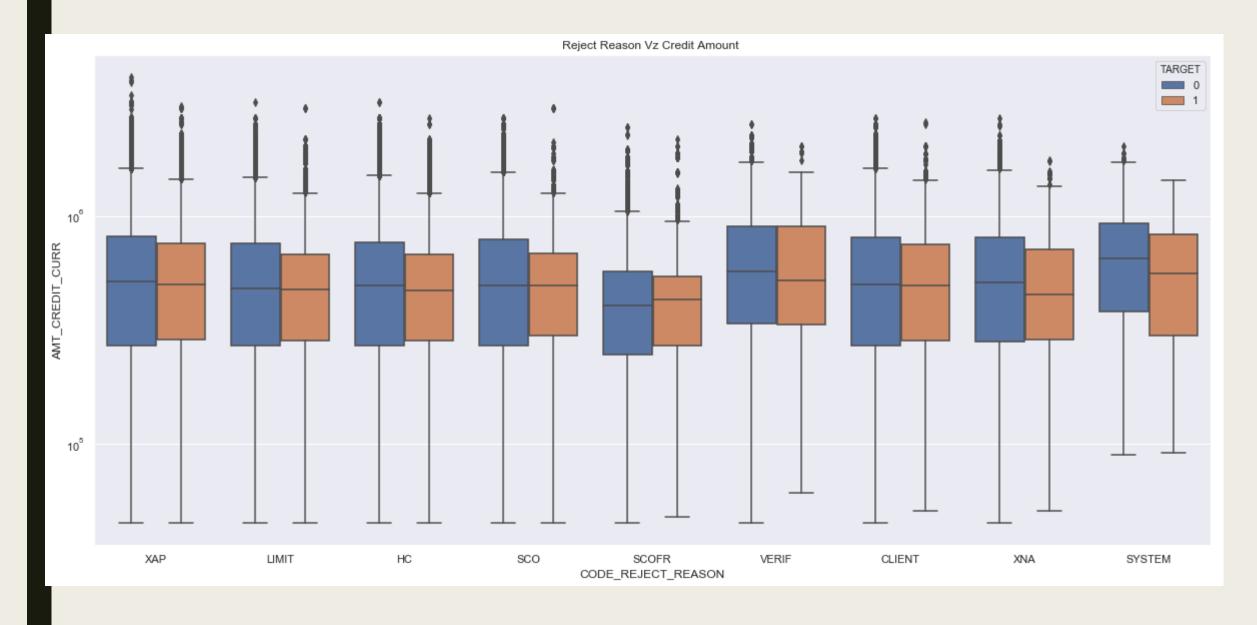


### Bivariate Analysis

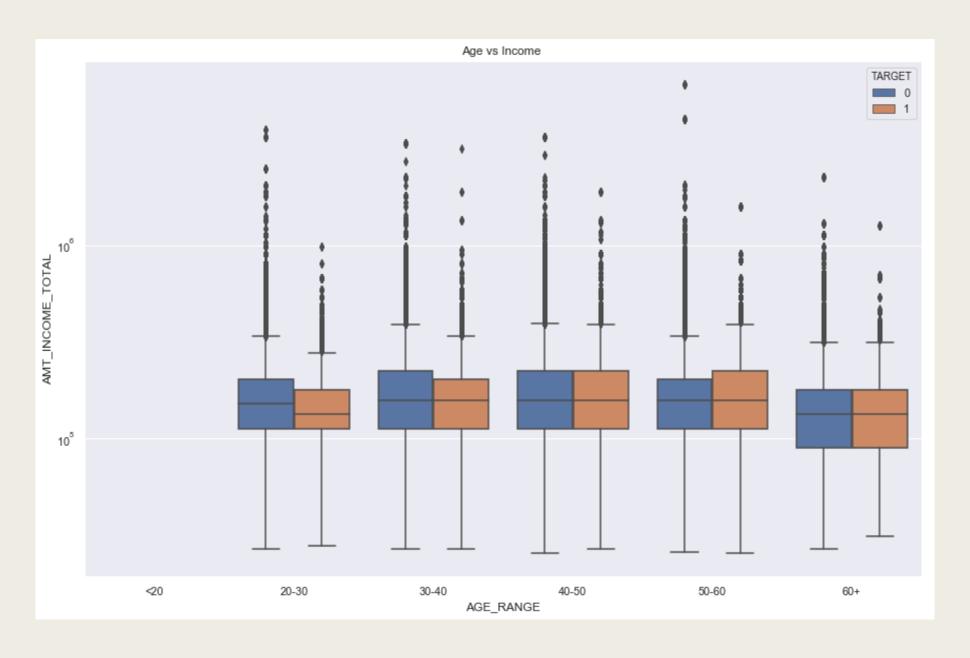
#### **Purpose of Loan vs Loan Credit**



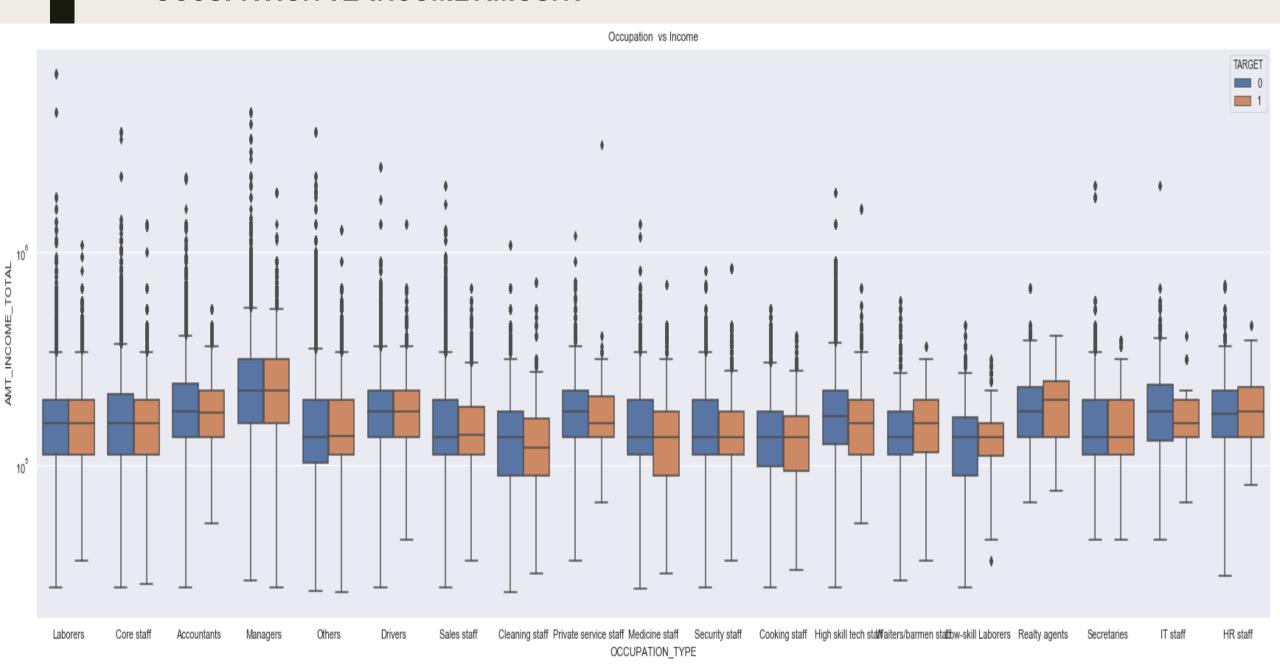
#### **REJECT REASON VZ CREDIT AMOUNT**



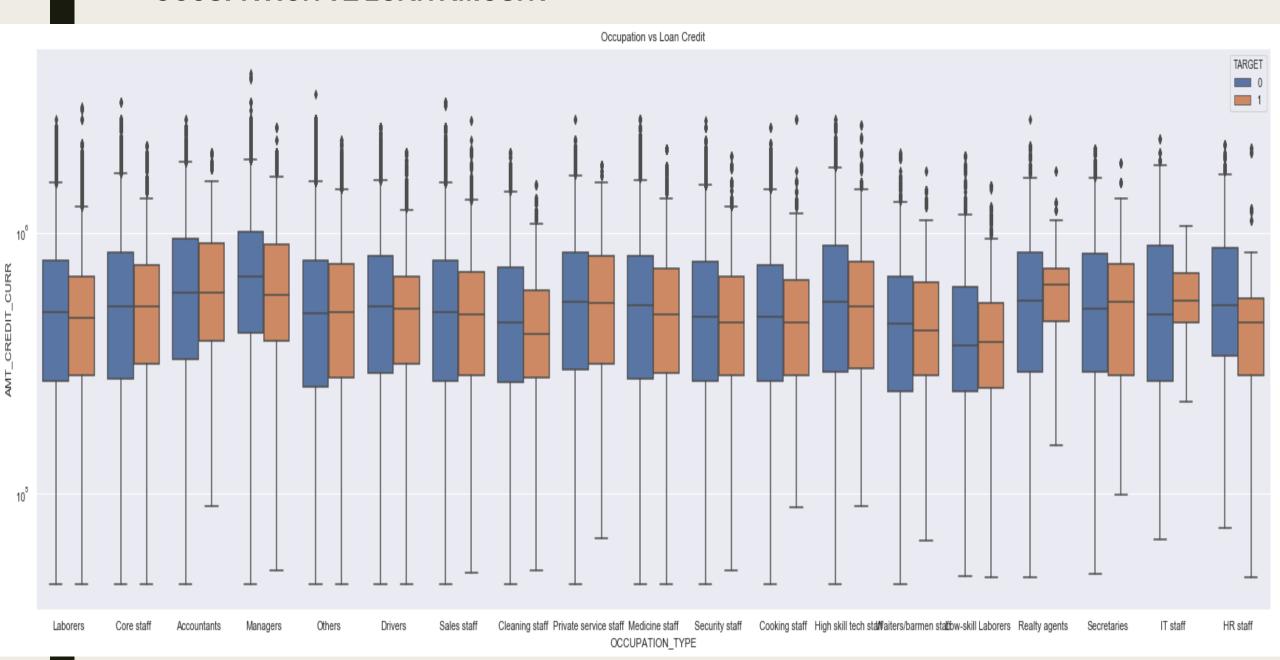
## **AGE VZ INCOME AMOUNT**



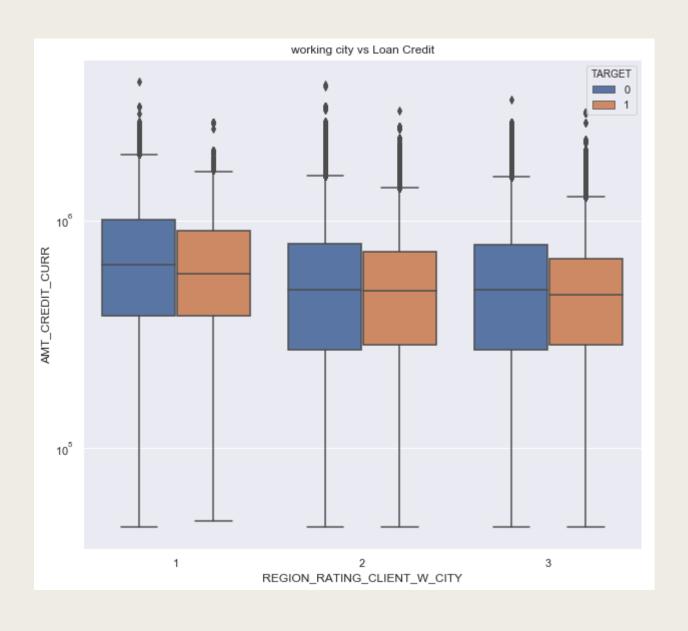
#### **OCCUPATION VZ INCOME AMOUNT**



#### **OCCUPATION VZ LOAN AMOUNT**



### **WORKING CITY VZ LOAN AMOUNT**



# REPAYERS CORRELATION MATRIX-HEATMAP

## TOP 10 correlation for REPAYERS

1	AMT_CREDIT_CURR	AMT_GOODS_PRICE_CURR	0.9998
2	AMT_APPLICATION	AMT_GOODS_PRICE_PREV	0.9926
3	YEARS_TERMINATION	YEARS_LAST_DUE	0.9902
4	AMT_CREDIT_PREV	AMT_GOODS_PRICE_PREV	0.9902
5	AMT_APPLICATION	AMT_CREDIT_PREV	0.989
6	AMT_ANNUITY_PREV	AMT_GOODS_PRICE_PREV	0.964
7	AMT_ANNUITY_PREV	AMT_CREDIT_PREV	0.9584
8	AMT_ANNUITY_PREV	AMT_APPLICATION	0.9583
9	CREDIT_TO_INCOME	ANNUITY_T0_INCOME	0.9449
10	AMT_ANNUITY_CURR	AMT_GOODS_PRICE_CURR	0.944

# DEFAULTERS CORRELATION MATRIX-HEATMAP

## TOP 10 correlation for DEFAULTERS

1	AMT_CREDIT_CURR	AMT_GOODS_PRICE_CURR	0.9998
2	YEARS_TERMINATION	YEARS_LAST_DUE	0.9942
3	AMT_APPLICATION	AMT_GOODS_PRICE_PREV	0.993
4	AMT_APPLICATION	AMT_CREDIT_PREV	0.989
5	AMT_CREDIT_PREV	AMT_GOODS_PRICE_PREV	0.989
6	AMT_ANNUITY_PREV	AMT_GOODS_PRICE_PREV	0.969
7	YEARS_FIRST_DRAWING	YEARS_LAST_DUE_1ST_VERSION	0.9688
8	AMT_ANNUITY_PREV	AMT_CREDIT_PREV	0.9665
9	AMT_ANNUITY_PREV	AMT_APPLICATION	0.9646
10	CREDIT_TO_INCOME	ANNUITY_T0_INCOME	0.9508

# **Business Insights**

- Majority Loanees are more in Region\_Rating\_Client=2
- Majority Loanees are more in Region\_Rating\_Client\_W\_City=2
- No one loanee without mobile number
- Majority Loanees both repayers and defaulters are married and working with secondary education
- Majority of the loanee are employed within 2-5 years
- Majority of loanees are in the age range 30-40
- Defaulters are very less in the age range of 60+
- Majority of loanees in the range of income 1LA-2LA
- Majority of loanees are of organization type -Busniess Entity
- Majority of loanees are getting credit range 4LA-10LA and Annuity range 20k-30K
- Majority of loanees have current good price in the range 2la-6la
- Majority of loanees application amount is between 50k-4LA
- Majority of loanees Down payment in the range 2,5k-10K

- Majority of the repayers last phone change is 5+ years before application
- Majority of the defaulters last phone change is 1-2 years before application
- Defaulters are more in Region\_Rating\_Client\_W\_city=2
- Defaulters occupation is majorly Others and Laborers
- Defaulters are very less in the age range of 60+
- Deafulting high in Cash Loans and Approved (Contract Status), and cash through bank payment gtype
- Defaulting tendency more in Client type: Repeaters
- Defaulers are very less in IT SECTOR
- Defaulters are more on high yield group than middle and lower
- Defaulters are more in Loan Credited for the purpose: Buying a Holiday Home
- Loan with SCOFR reject reason defaulted less
- Main code reject reason is XAP then, HC

# **THANK YOU**