

The background features abstract, overlapping green geometric shapes, primarily triangles and polygons, in various shades of green, creating a modern and dynamic visual effect.

EDA CASE STUDY

Credit Risk Analysis

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Problem Statement

- ▶ There are some risk associated with loan request. These are as follows:
 - 1) If the applicant is not likely to repay the loan that means he/she is likely to default, then approving the loan may lead to a financial loss for the bank.
 - 2) If the applicant is likely to repay the loan, then not approving the loan application results in loss of revenue to the company.

To find out solution to overcome this, we have analysed the data in Jupyter Notebook.

Methodology

- ▶ After loading the data sets, first we cleaned it for better analysis.
- ▶ Found out Out co-relation and imbalance between data.
- ▶ Did univariate and bivariate analysis on each dataset independently.
- ▶ After that merged the data set.
- ▶ Repeated univariate and bivariate analysis on merged dataset.

Dropping the columns which having more missing values

```
In [10]: ##get the columns having missing value percentage more then 22 to a list  
missing_percentage_pre_app_col=missing_percentage_pre_app[missing_percentage_pre_app>=22]  
missing_percentage_pre_app_col
```

```
Out[10]: AMT_ANNUITY          22.286665  
AMT_DOWN_PAYMENT          53.636480  
AMT_GOODS_PRICE           23.081773  
RATE_DOWN_PAYMENT        53.636480  
RATE_INTEREST_PRIMARY     99.643698  
RATE_INTEREST_PRIVILEGED  99.643698  
NAME_TYPE_SUITE           49.119754  
CNT_PAYMENT              22.286366  
DAYS_FIRST_DRAWING        40.298129  
DAYS_FIRST_DUE            40.298129  
DAYS_LAST_DUE_1ST_VERSION 40.298129  
DAYS_LAST_DUE            40.298129  
DAYS_TERMINATION          40.298129  
NFLAG_INSURED_ON_APPROVAL 40.298129  
dtype: float64
```

```
In [11]: ##drop columns with percentage of missing value >22  
pre_app=pre_app.drop(columns=missing_percentage_pre_app_col.index)  
  
##validating the shape after dropping the columns  
pre_app.shape
```

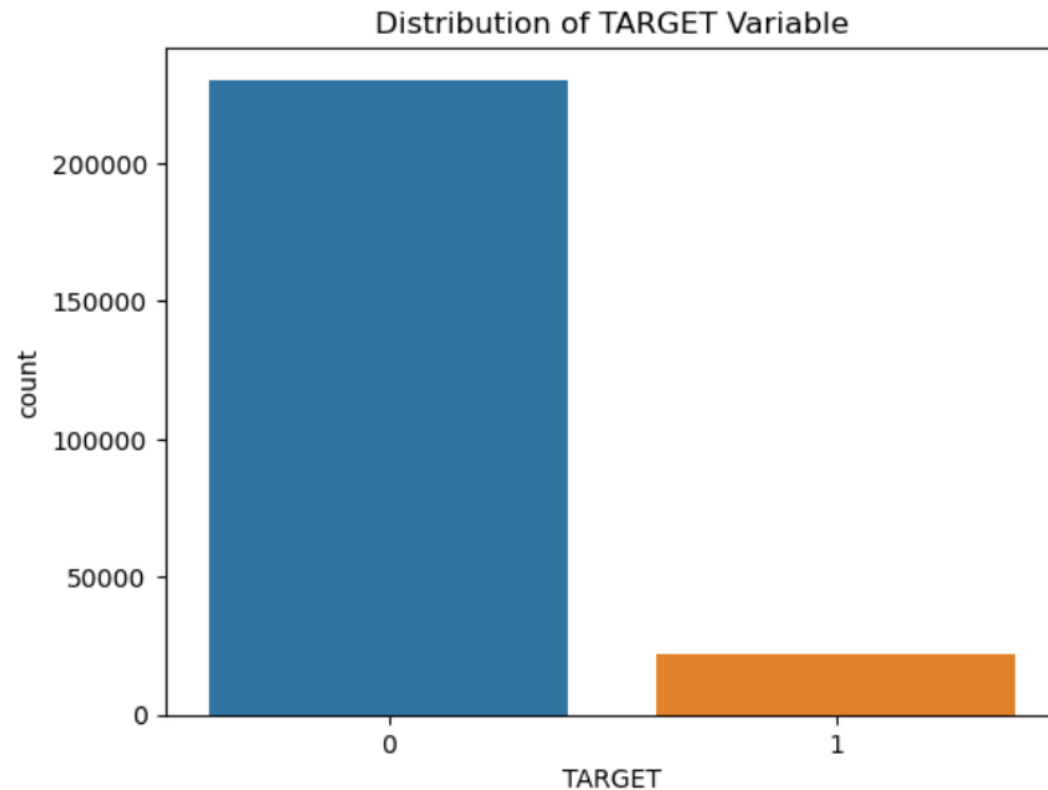
```
Out[11]: (1670214, 23)
```

```
In [81]: ##drop columns with percentage of missing value >40  
app_data_1=app_data.drop(columns=app_data_missing_percentage_col.index)
```

```
In [82]: ##validating the shape after dropping the columns  
app_data_1.shape
```

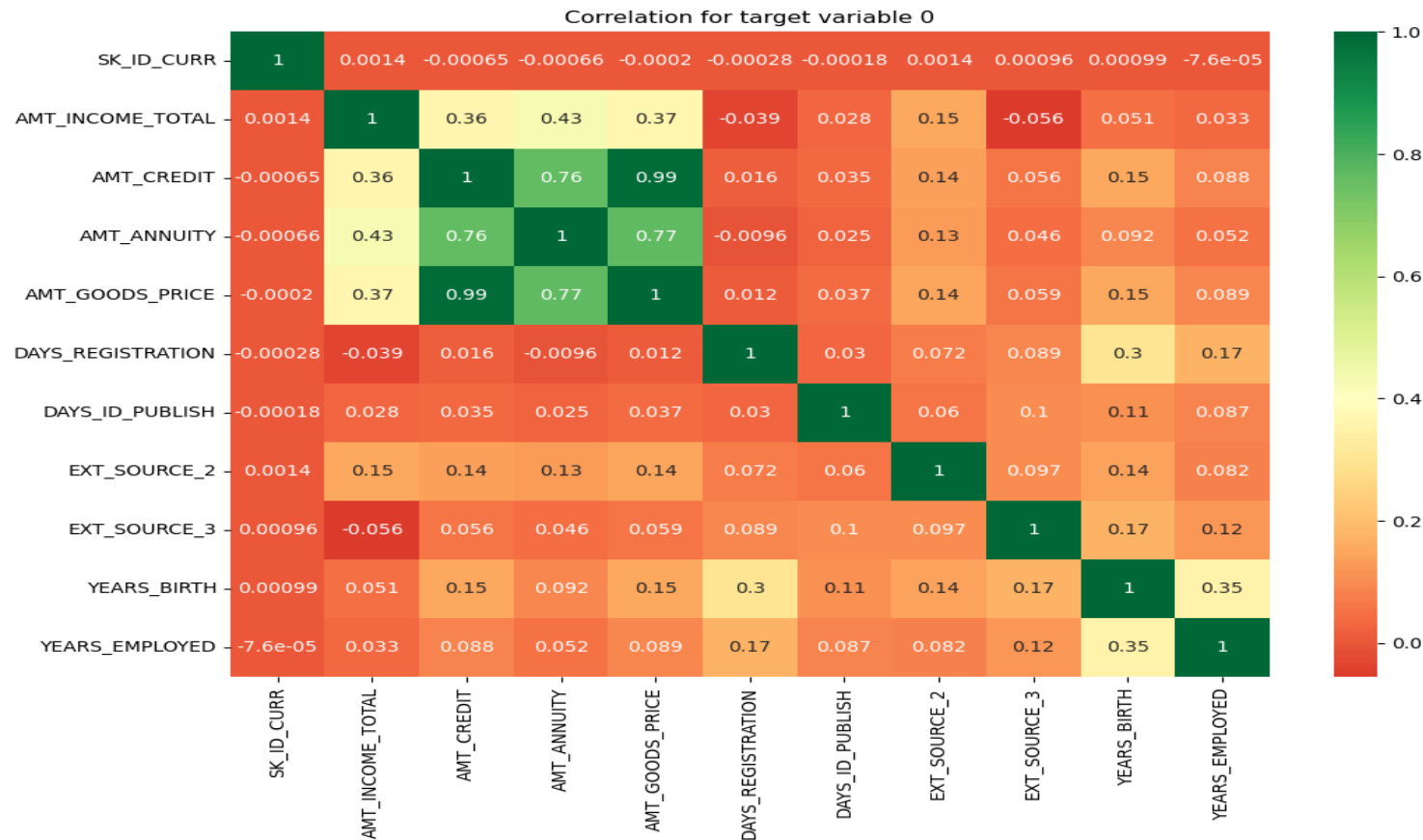
```
Out[82]: (307511, 73)
```

Finding Imbalance



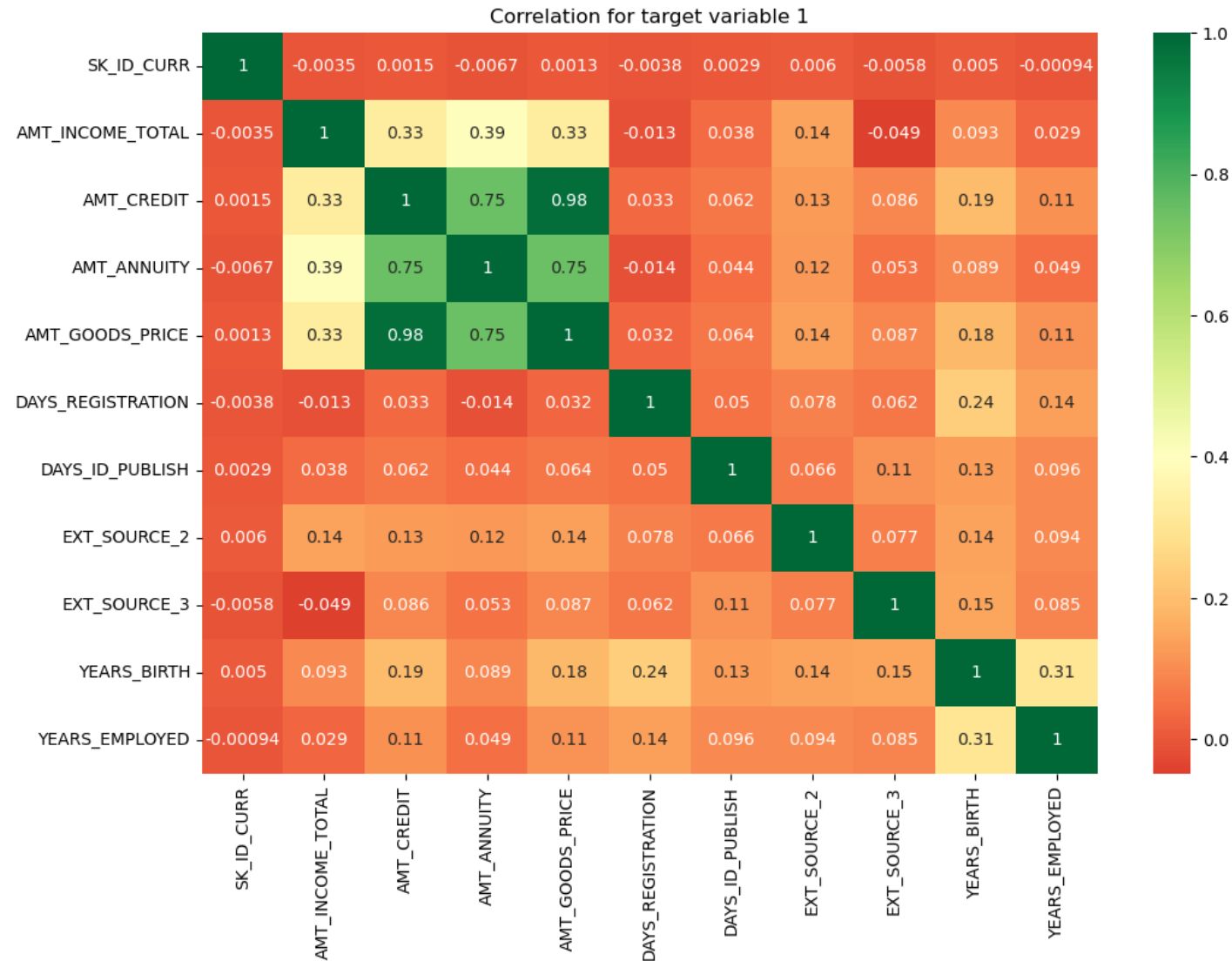
- 1) Percentage of customer being defaulter is imbalanced as compare to not being defaulter
- 2) The gap between the perecentage of taget values are significant

Correlation For Target Variable 0



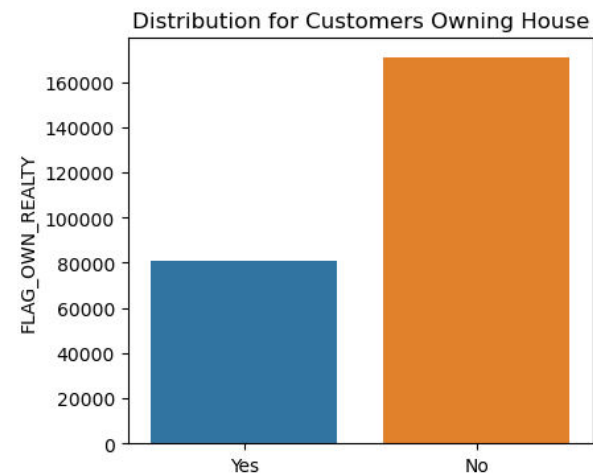
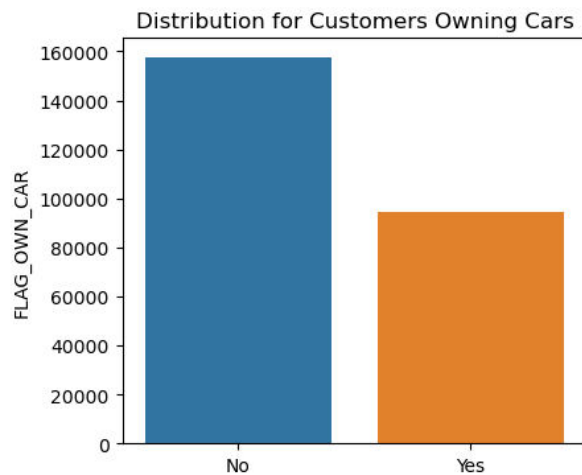
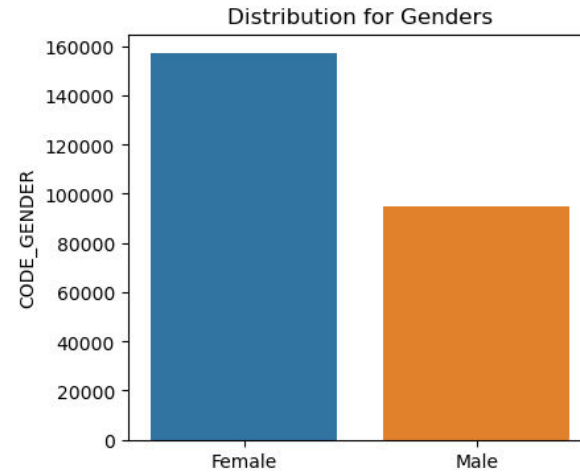
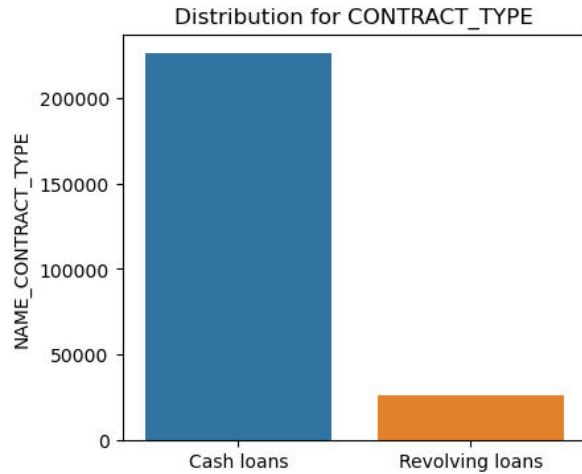
- 1) AMT_GOODS_PRICE and AMT_CREDIT have highest correlation around 0.99
- 2) AMT_GOODS_PRICE and AMT_ANNUITY is also showing better correlation
- 3) Days Registration is having negative correlation with AMT_INCOME_TOTAL

Correlation For Target Variable 1



AMT_CREDIT and
AMT_GOODS_PRICE
have heighest
correlation around
0.98

Univariate Analysis



1) Number of Cash loan is more than Revolving loans it means more clients are opting for cash loans

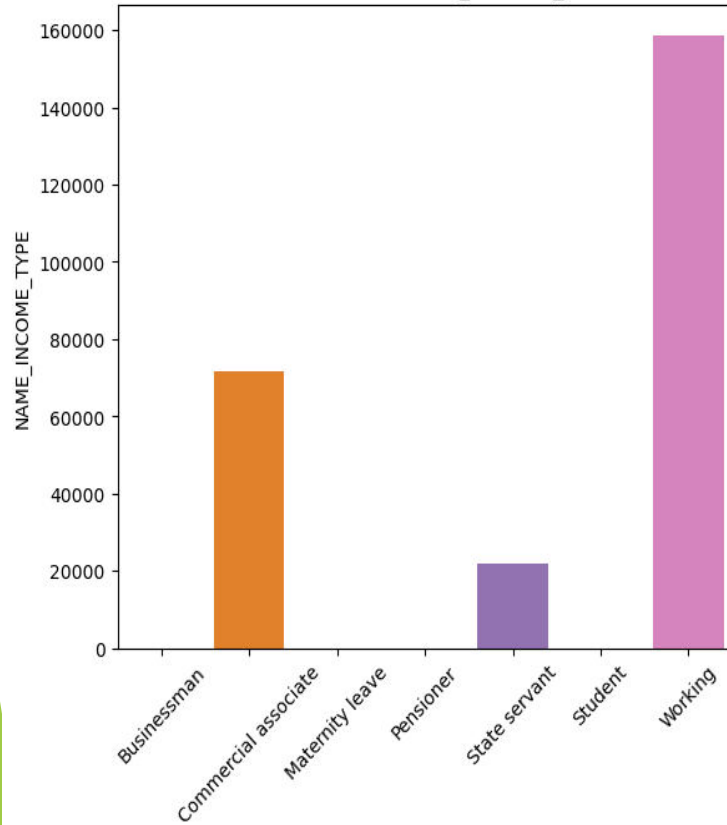
2) Number of female clients is more than male clients

3) Clients who owns a car is less than clients who does not owns a car

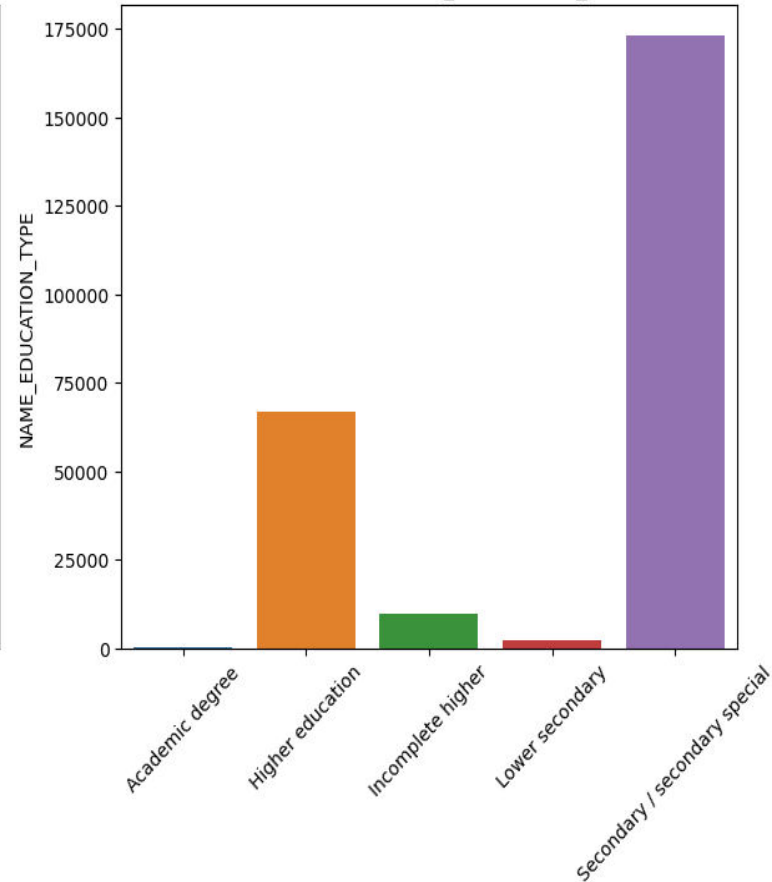
4) Clients who owns reality is around 50 % less than clients who does not owns a reality

Univariate Analysis

Distribution for NAME_INCOME_TYPE



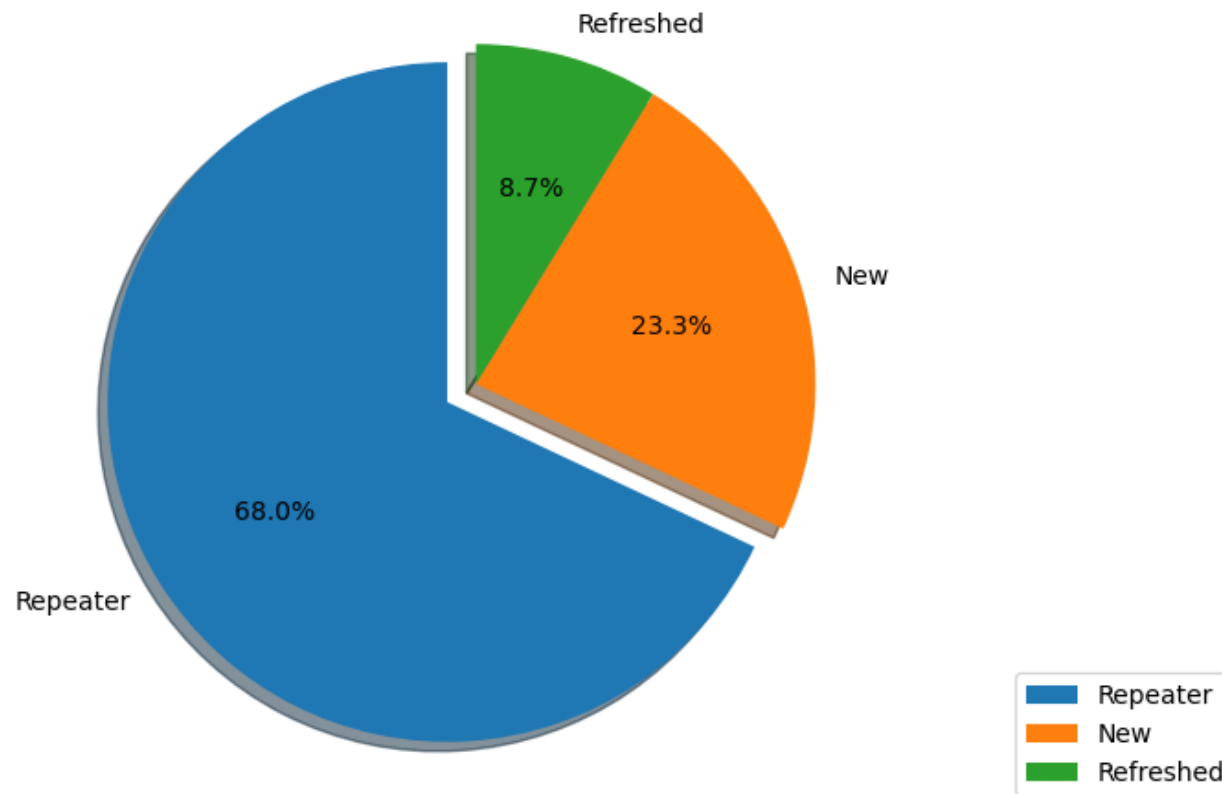
Distribution for NAME_EDUCATION_TYPE



People with
secondary/ special,
higher education
likely to make less
default

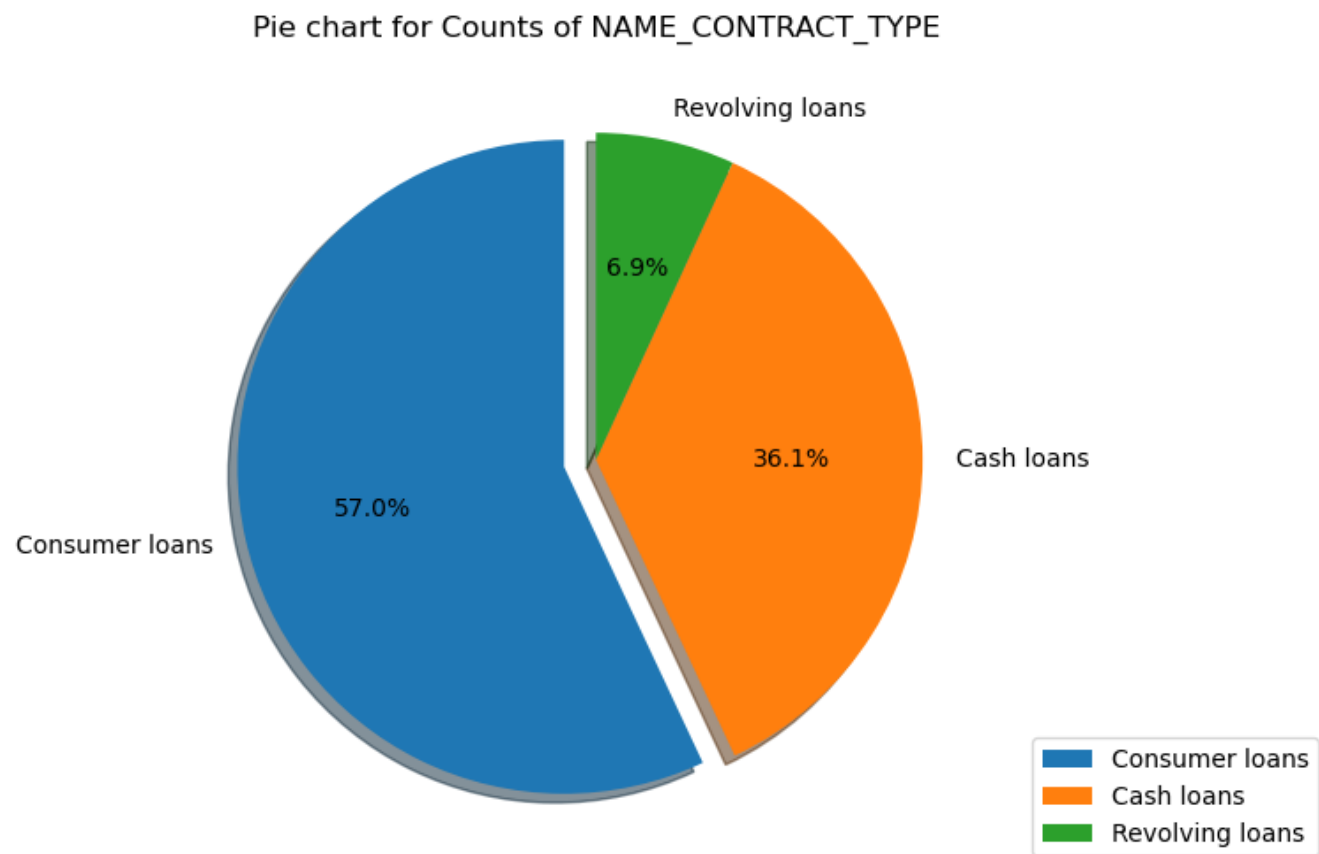
Univariate Analysis

Pie chart for Counts of NAME_CLIENT_TYPE

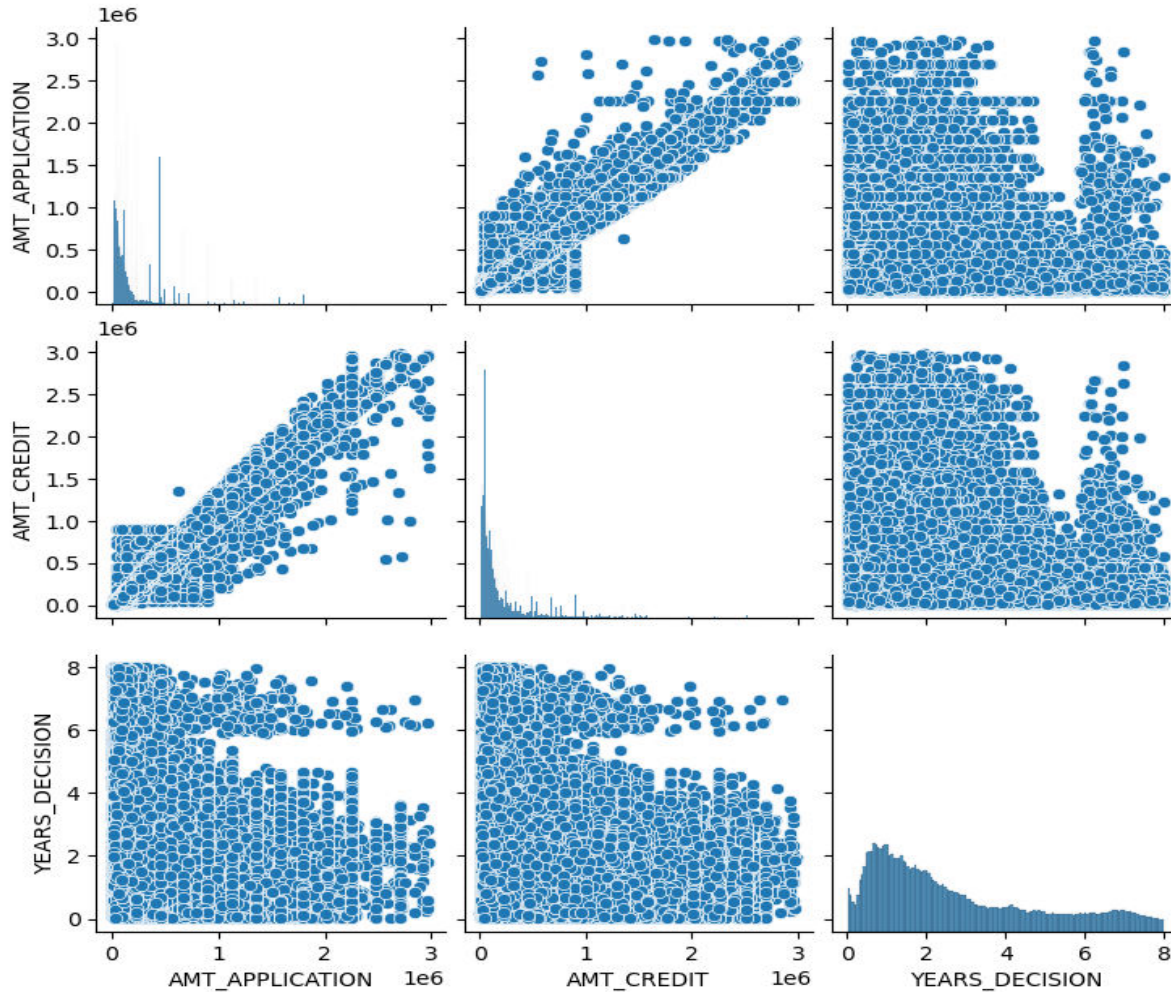


Most of
the
applicants
are
repeater

Univariate Analysis

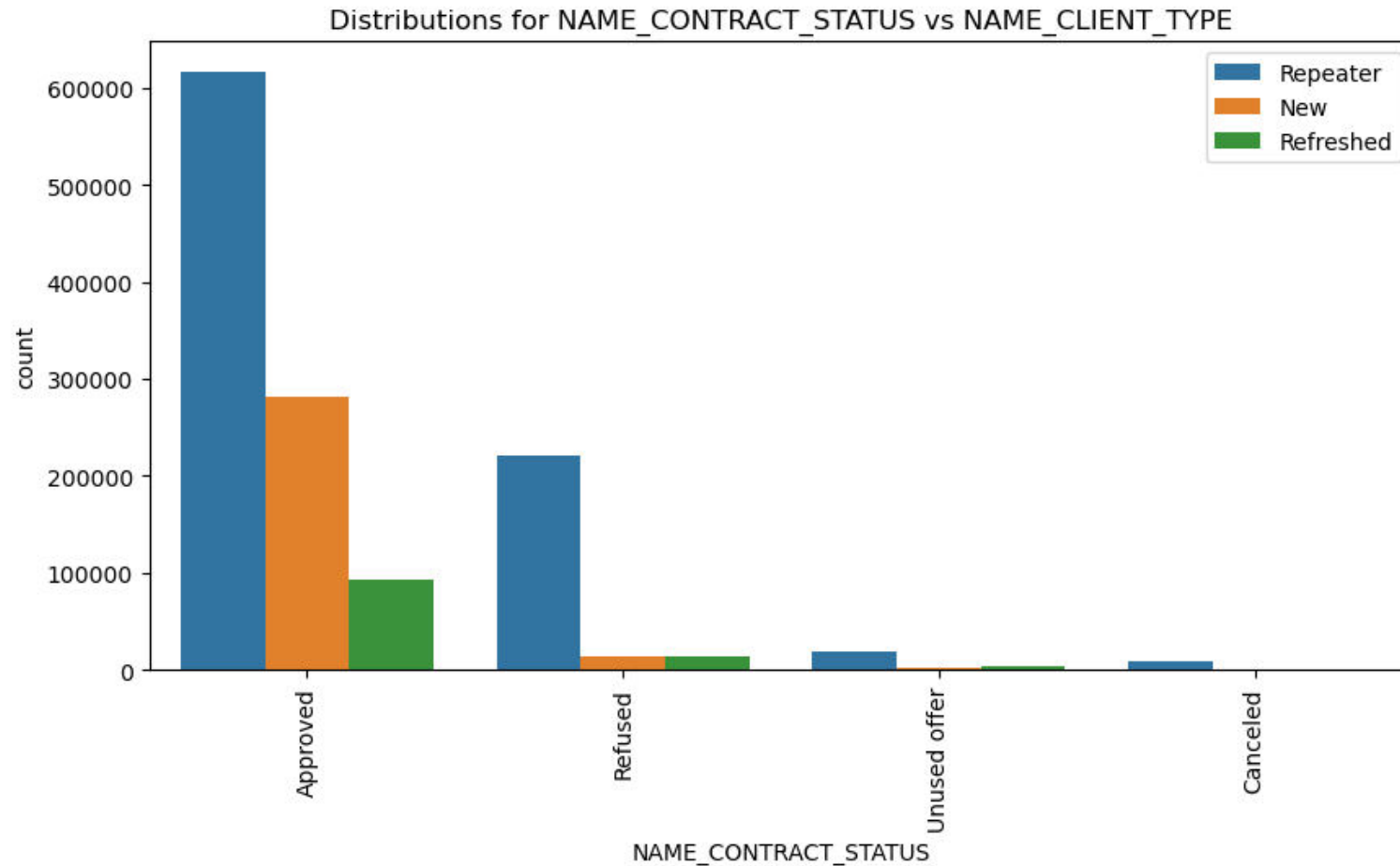


Bivariate Analysis



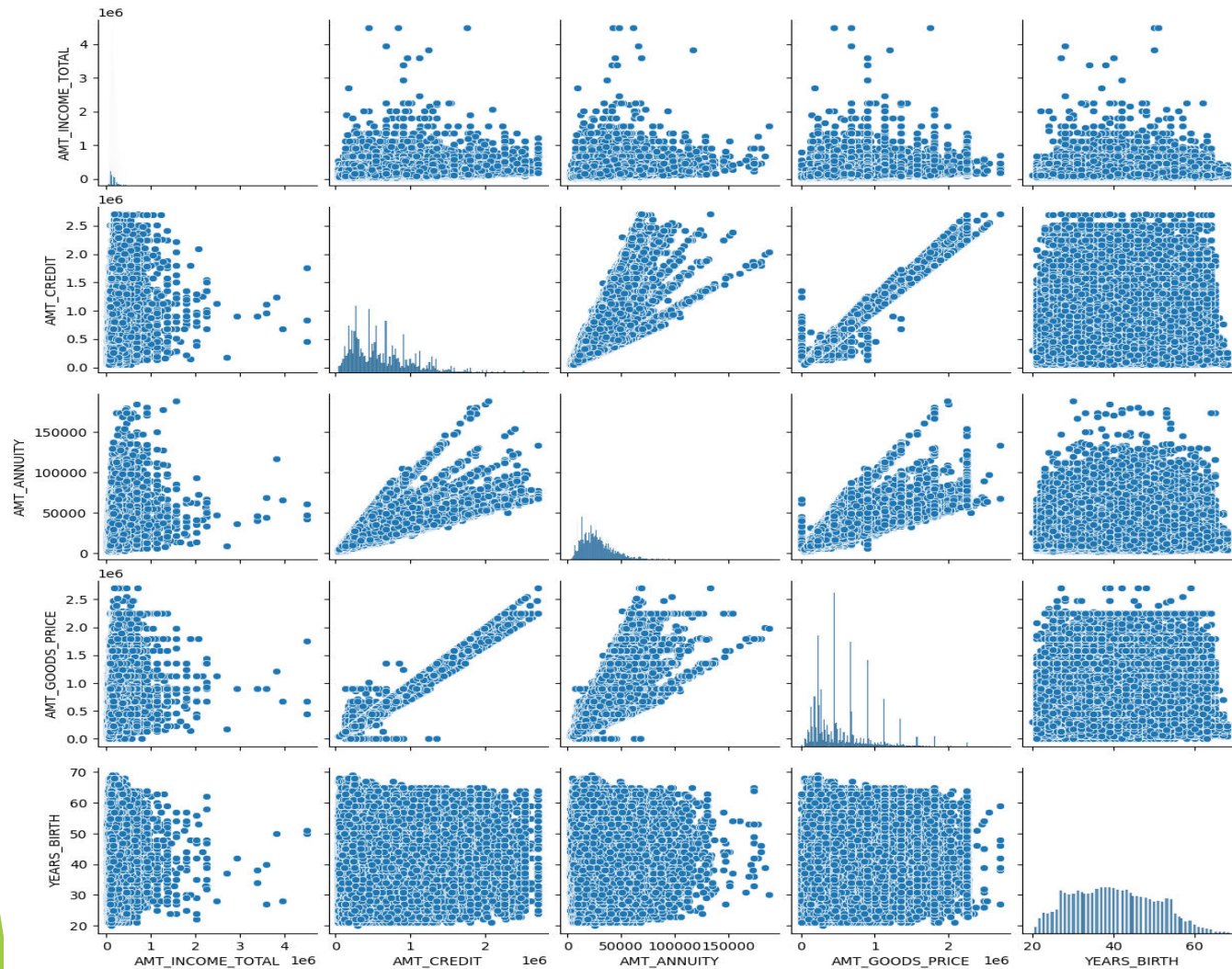
There is linear relationship between AMT_APPLICATION & AMT_CREDIT

Bivariate Analysis



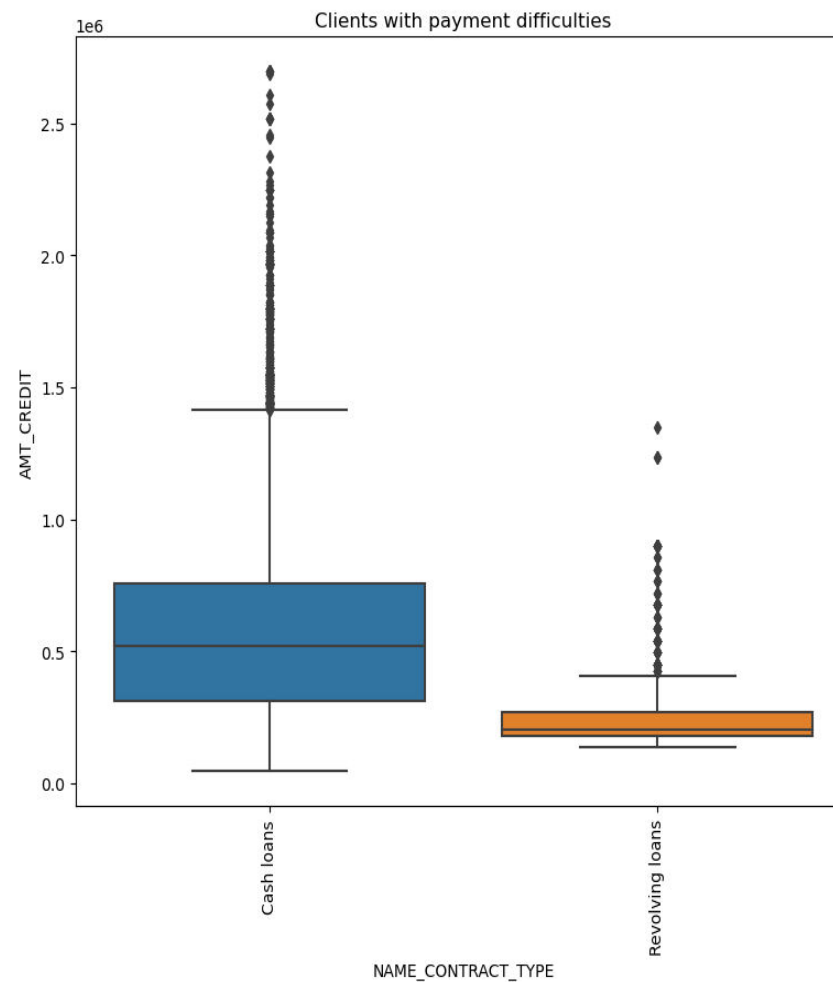
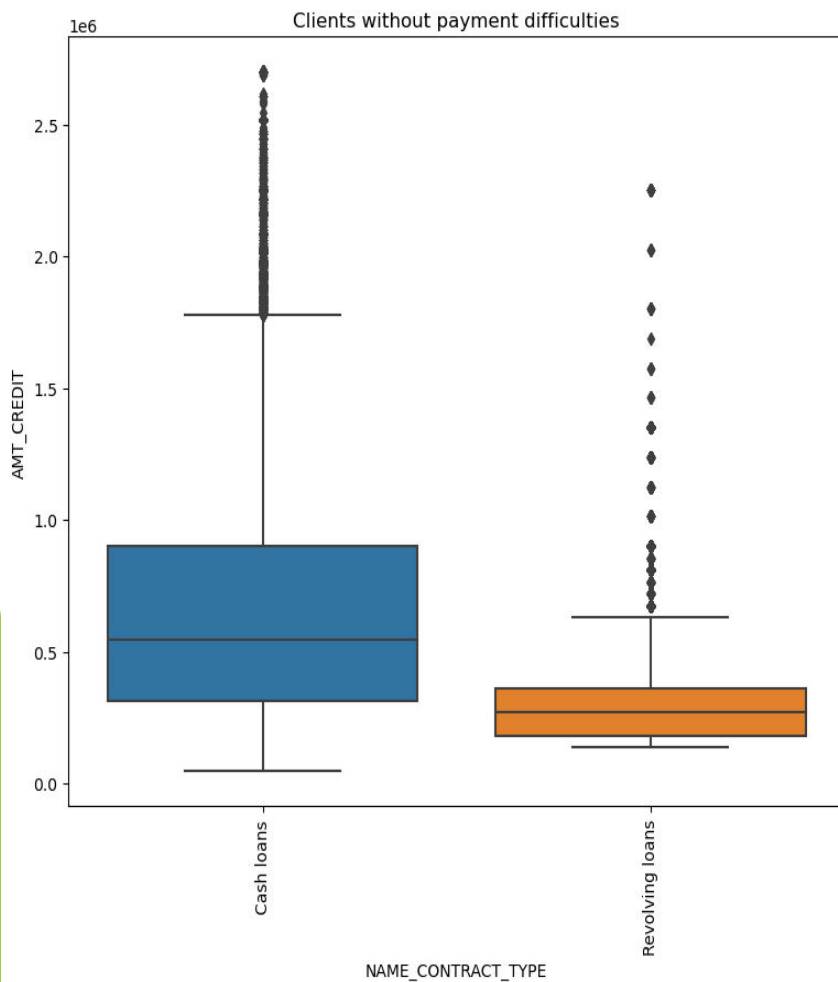
Approval rate for repeater applicants is higher than any other category

Bivariate Analysis



Linear correlation
present between
AMT_GOODS_PRICE
vs AMT_CREDIT ,
AMT_GOODS_PRICE
vs AMT_ANNUITY

Bivariate Analysis



Applicants more approached towards cash loans than revolving loans.

Conclusion

- ▶ Occupation type such as IT staff , HR staff , Managers etc have lower difficulty in repayment when observed in comparison to drivers, Cleaning staff ,laborers
- ▶ To increase the chances of repaying the loan provided by the companies should less focus on the clients having income type as Working
- ▶ In general more number of Cash loans are provided by bank in comparison to Revolving loan post chances to get defaulted increases
- ▶ People with higher education and older people default less
- ▶ Higher amount of income, less default
- ▶ Loans which previously refused or cancelled, high chance of default
- ▶ Giving loan to married people rather than single or widow is safer