# CREDIT EDA ASSIGNMENT

Divya N L August 28<sup>th</sup>, 2022

## **CONTENTS**

- **01.** Problem statement
- **02.** Steps involved in Analysis
- 03. Conclusions & Recommendations

### **Problem Statement**

- This case study aims to identify patterns if an loan applicant has difficulty in repaying his/her loan in a Finance Based Company.
- \* Also to ensure that applicants capable of repaying are not denied the application request and applicants who are incapable of repaying are approved/accepted the application request.
- Bank wants to understand the driving factors (or driver variables) behind loan default, i.e. the variables which are strong indicators of default.
- The bank can utilize this knowledge for its portfolio and risk assessment.

# Steps involved in analysis

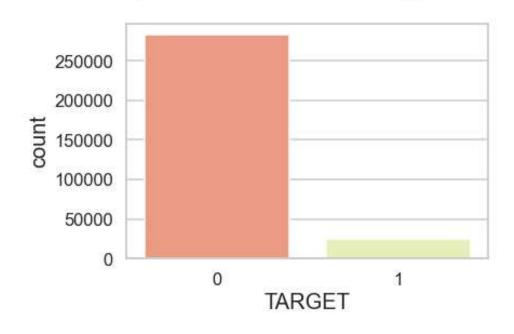
- Import Warnings and Libraries
- Read the Datasets
- Analyze, Clean the Dataset
  - Eliminating null values and Identifying Outliers
- Check the Imbalance Ratio
- Univariate, Segmented and Bivariate Analysis
- Summary and Recommendations

# **IMBALANCE RATIO**

#### **OBSERVATIONS**:

- The data is highly imbalanced.
- Data Imbalance Ratio 8:92
- Number of defaulters(Target = 1): 24825
- Number of non-defaulters (Target = 0): 282686

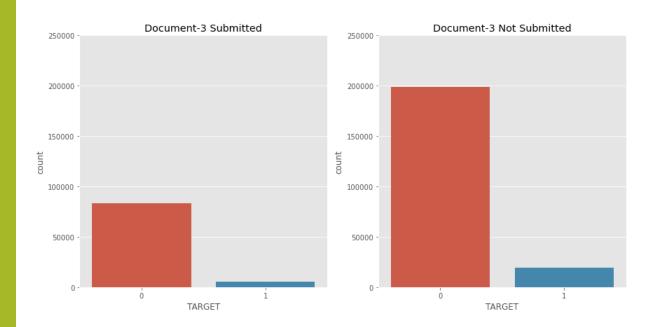
#### Payment Status of Current Application



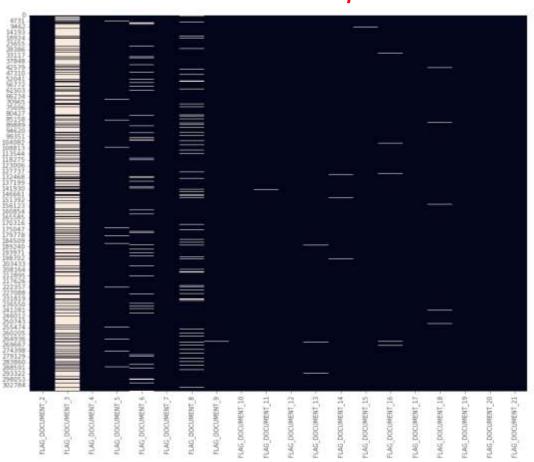
# Segment wise analysis

#### **Documents submitted by applicants:**

- except Document\_3, other documents were not provided by applicants in majority of the cases.
- Document\_3 shows similar trends for both defaulters and non-defaulters.



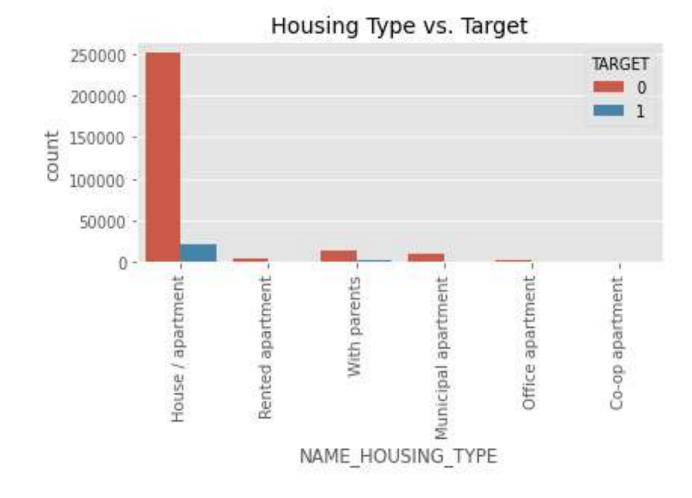
#### **Heat Map**



# **Housing Type**

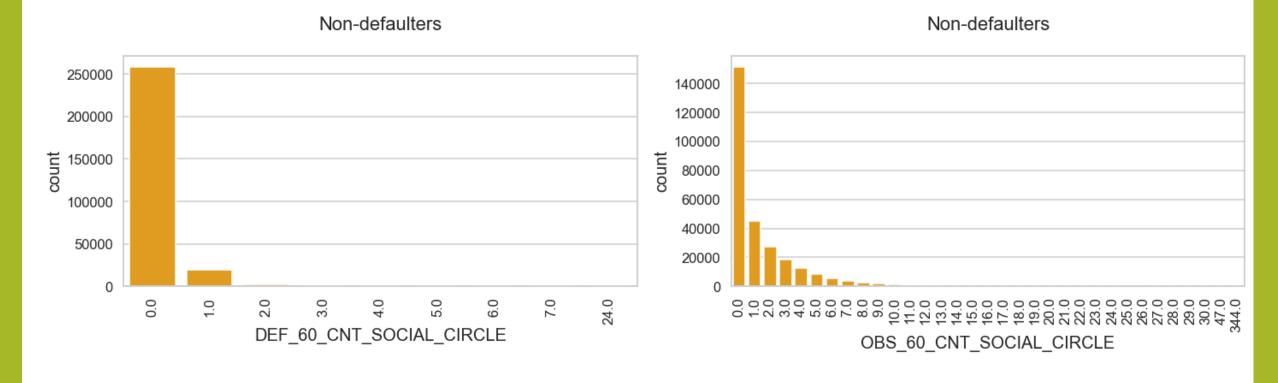
#### • OBSERVATIONS:

- Most of the loan applicants live in House/Apartment
- The most defaulters live in rented apartment or live with their parents

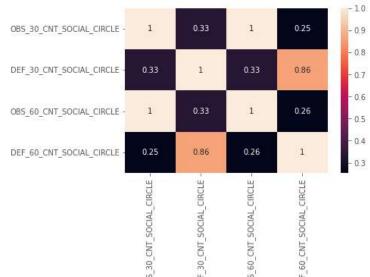


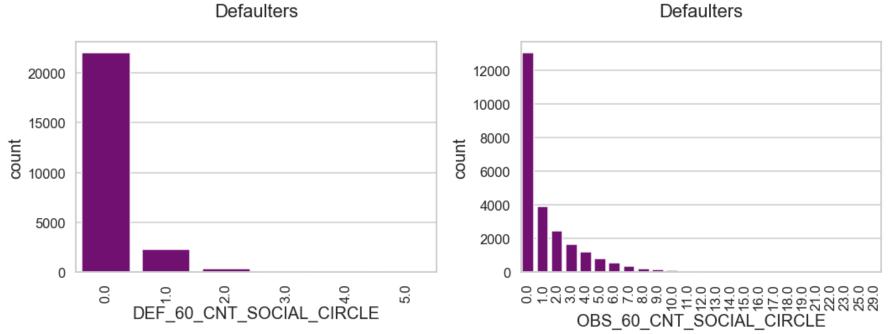
## SOCIAL CIRCLE

• For defaulters and non-defaulters 'DEF\_60\_CNT\_SOCIAL\_CIRCLE', 'OBS\_60\_CNT\_SOCIAL\_CIRCLE' features show similar trend.



- DEF\_30\_CNT\_SOCIAL\_CIRCLE and DEF\_60\_CNT\_SOCIAL\_CIRCLE are highly correlated
- OBS\_30\_CNT\_SOCIAL\_CIRCLE and OBS\_60\_CNT\_SOCIAL\_CIRCLE are identical column



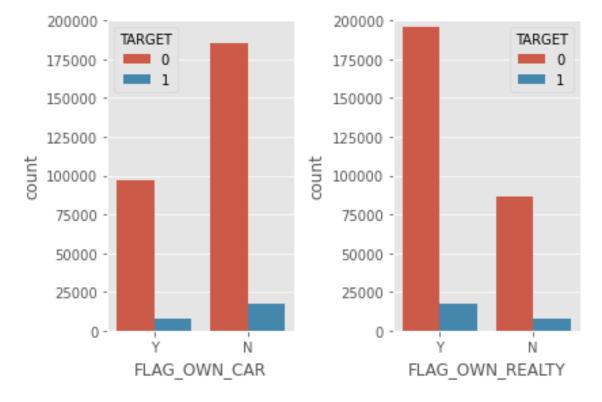


## **ASSET DETAILS**

#### **OBSERVATIONS:**

• People not owning realty and car and have a slightly higher default rate than the people who own reality and

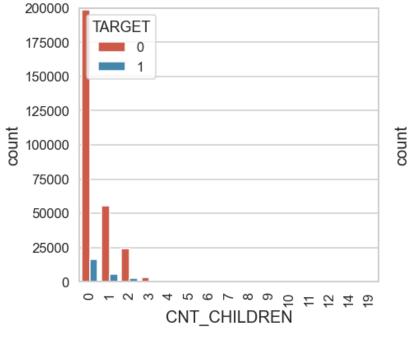
car.

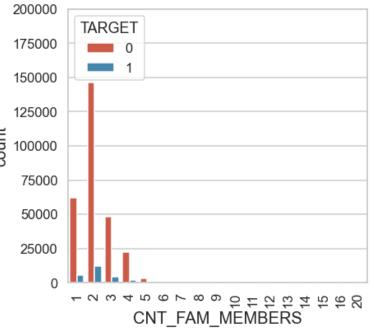


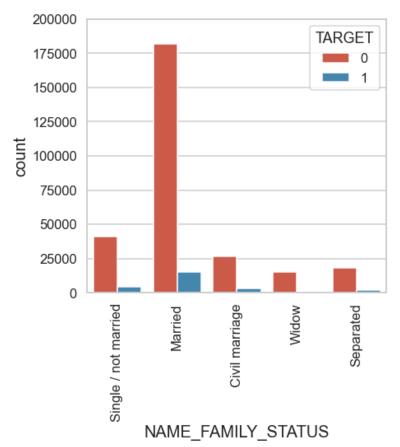
## **FAMILY RELATED INFO**

#### **OBSERVATIONS:**

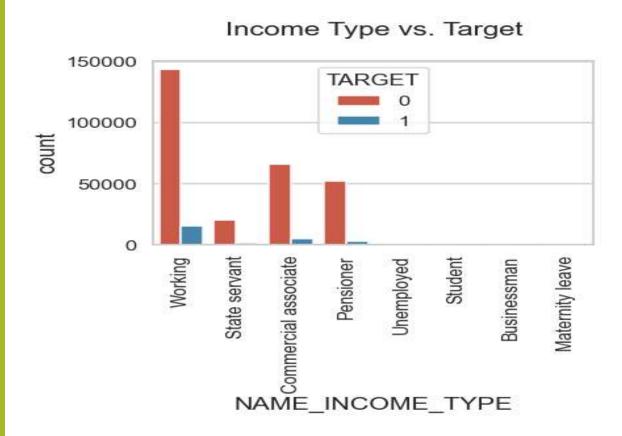
- Default rate is highest for Civil Marriage and Single applicants
- Most of the applicants are married (and/or) no children (and/or) 2 family members.
- Applicants with relatively more number of children (and/or) family members have higher default percentage.
- For some of the cases where count children/family members is high, and the default rate is very high or very low. These cases cannot be taken as a conclusion as number of applicants having a large family is very low.

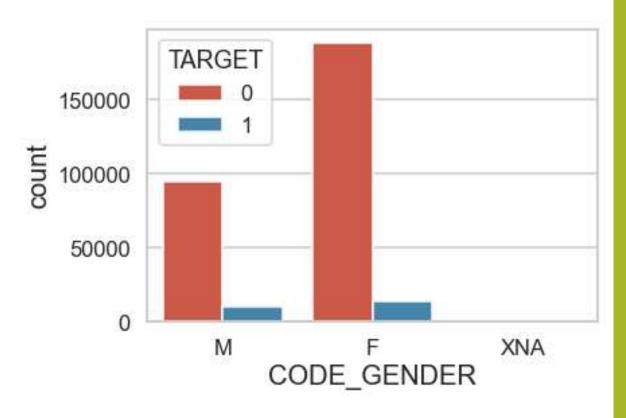






## Income Type and gender

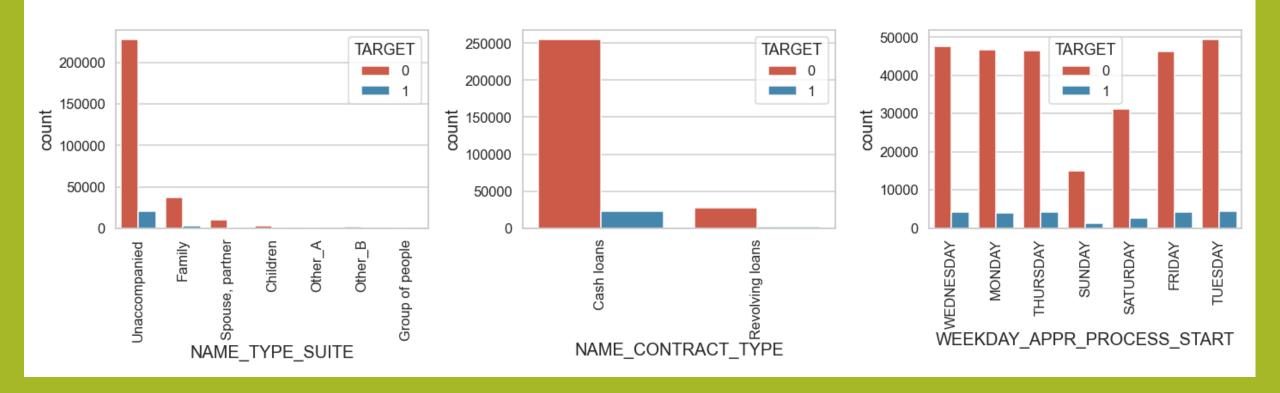




#### **OBSERVATIONS**

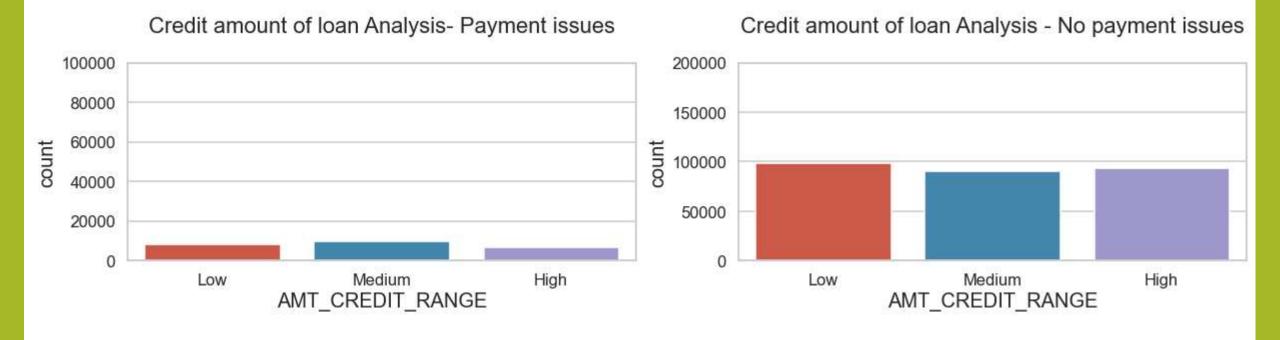
- Female applicants are more than male applicants
- Defaulter percentage is higher for male applicants
- Applicants having "Lower secondary" education have highest percentage of Defaulter. Applicants having 'Higher education' and 'academic degree' have lowest percentage of defaulters
- Applicants on Maternity Leave and Unemployed has highest percentage of Defaulter
- Businessman have lowest (0) percentage of Defaulter However, if we see the count, the applicants of income type('Unemployed', 'Student', 'Businessman', 'Maternity leave') are very few in the dataset to contribute in the analysis. Most of the applicants are working.

- Most of the applicants are unaccompanied while applying for loan
- Number Cash loans is quite higher than Revolving Loans
- All weekdays have similar number of applicants than weekend(Saturday and Sunday)

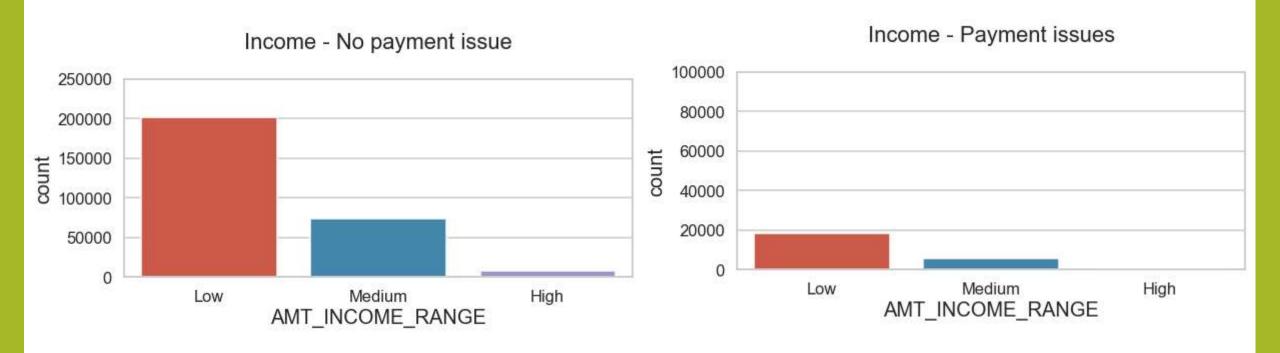


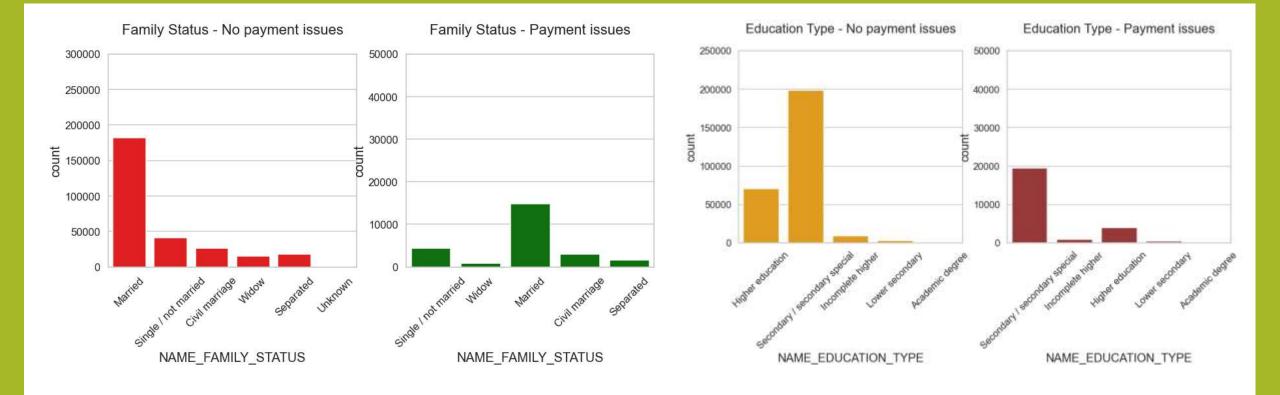
#### **Credit Amount Analysis:**

Customers who have low credit amount are more likely to pay back the loan



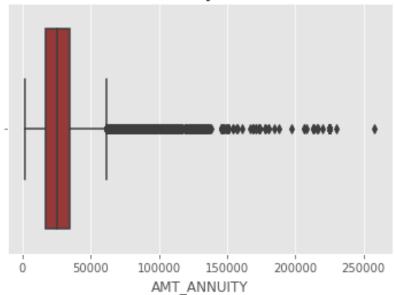
• As compared to the other categories, clients having low income are more likely to repay the loan.



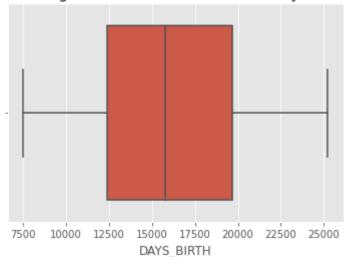


- People most likely to repay the loan have secondary/Secondary special education status whereas clients having an academic degree are the most defaulters.
- Widows are the least likely category to repay whereas as seen here, Married category is most likely to repay the loan amount.

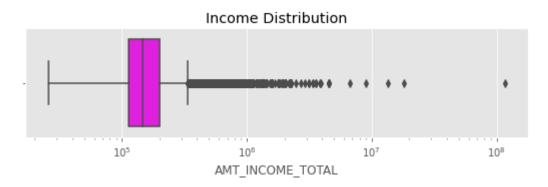
#### Amount Annuity Distribution

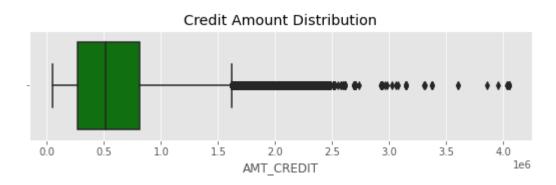


#### Age Distribution in the form of days

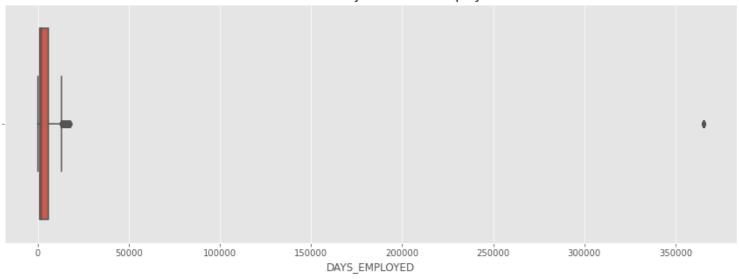


# **Identifying Outliers**





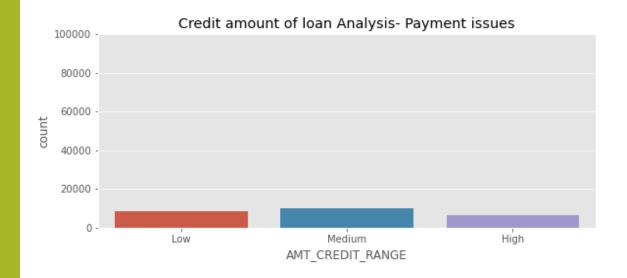
#### Distribution of Days the client employed

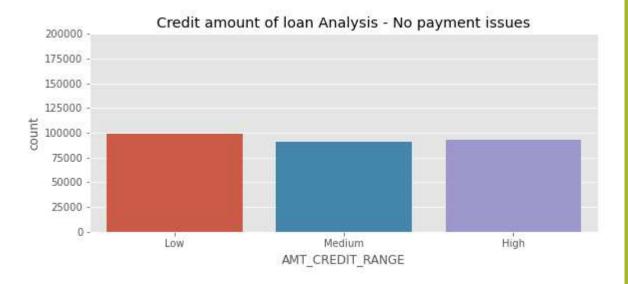


#### **OBSERVATIONS:**

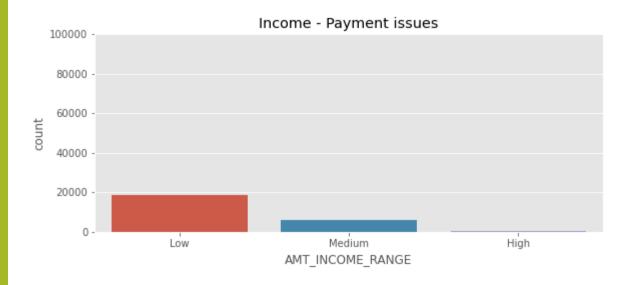
- Amount\_Annuity: outlier is present at 258025
- Here in 'AMT\_INCOME\_TOTAL' variable outlier values stands at 1.17x10^8. As the 95th and 99th quantile values differ largely, we can conclude the presence of outliers in the data set.
- Amt\_credit: The outliers here are present after the 99th quantile.
- No outliers are present in Days\_birth.
- we have an outlier at 365243 in Days\_employed.

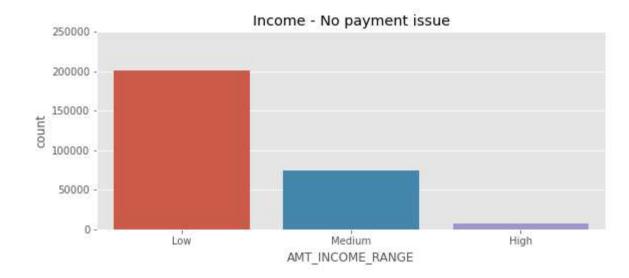
## **Univariate Analysis - Numeric variables**





Defaulters Non Defaulters

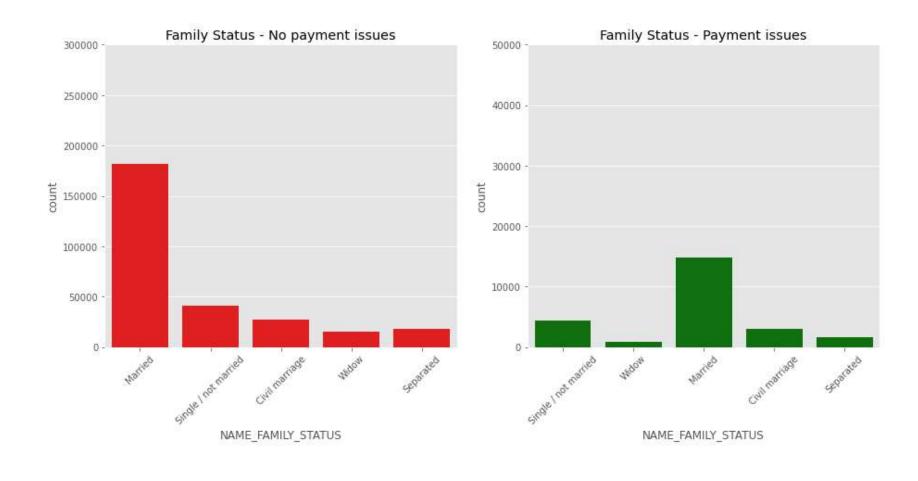


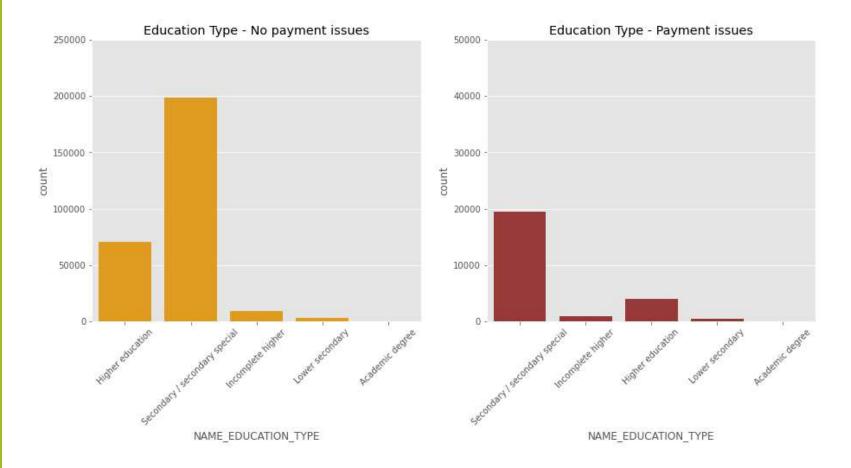


#### Defaulters

Non Defaulters

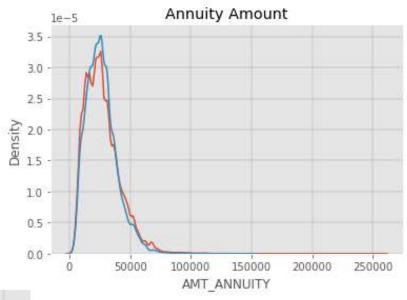
#### **Univariate analysis - Categorical variables**

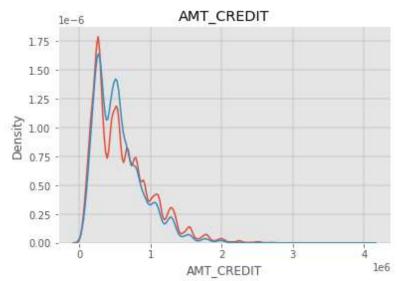


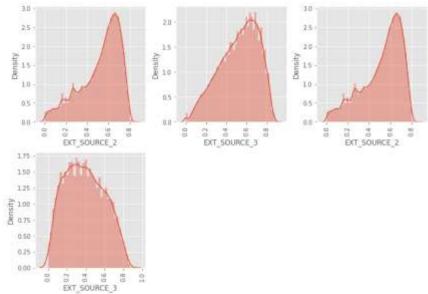


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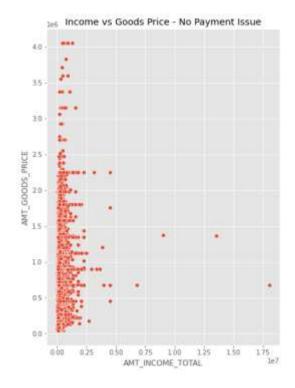
# **Univariate Analysis - Continuous variables**

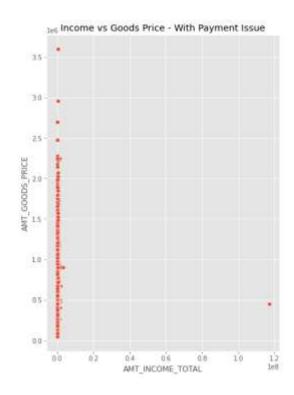


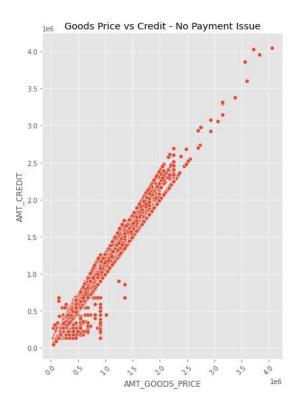


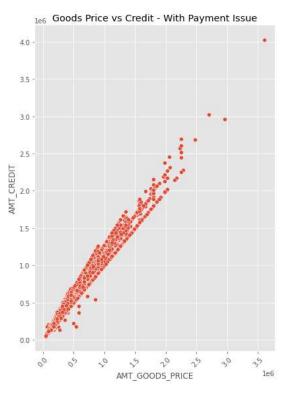


# **Bivariate Analysis**

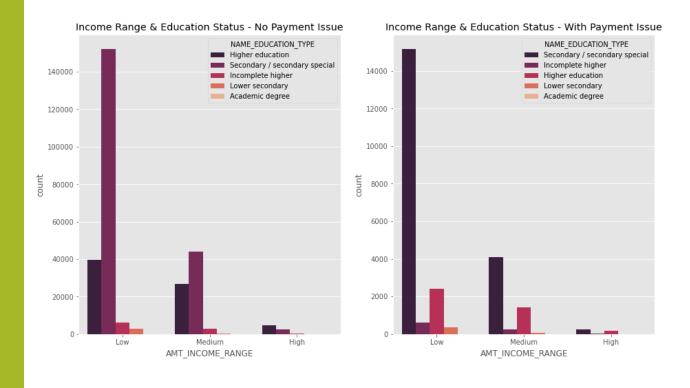


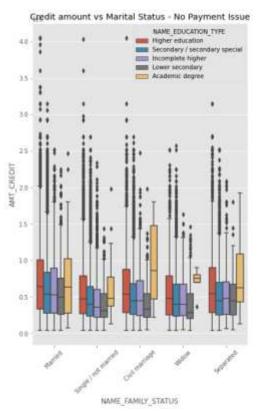


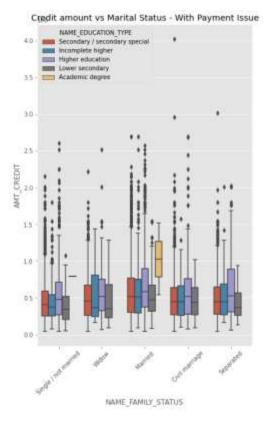




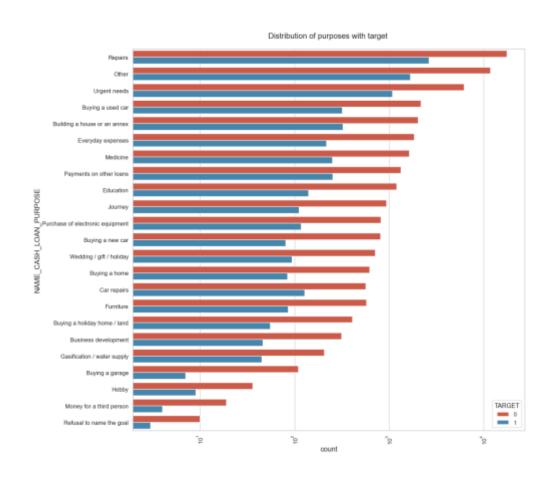
# **Numeric - Categorical**





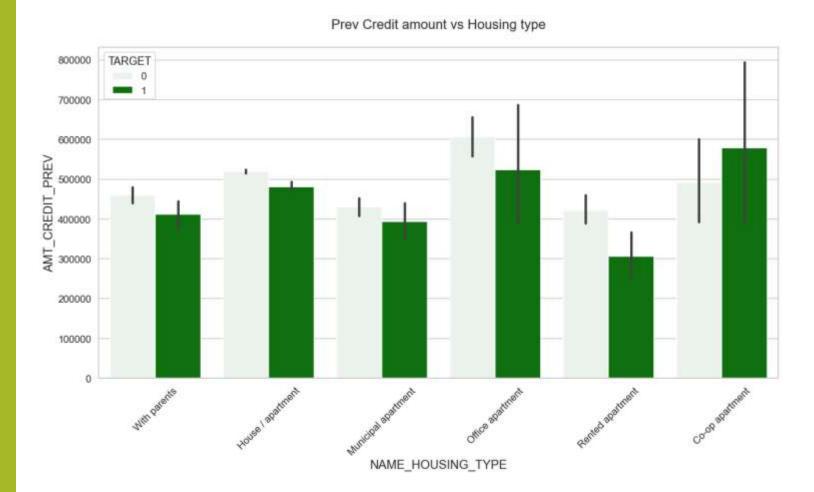


# Univariate analysis on merged data



• Categories like 'Buying a garage', 'Money for a third person', etc are the categories who have a high chance of successful loan repayment and should be prioritized.

## Bivariate analysis on merged data



• When it comes to the ease of payment, "office apartment" category has the higher credit as compared to others. Also, bank should be careful while approving loans for 'co-op apartment' categories.

# **Top 10 correlation for Defaulters**

• SK_ID_CURR	SK_ID_CURR	1.000000
• OBS_60_CNT_SOCIAL_CIRCLE	OBS_30_CNT_SOCIAL_CIRCLE	0.998269
<ul> <li>AMT_GOODS_PRICE</li> </ul>	AMT_CREDIT	0.983103
• REGION_RATING_CLIENT_W_CITY	REGION_RATING_CLIENT	0.956637
• CNT_CHILDREN	CNT_FAM_MEMBERS	0.885484
• DEF_60_CNT_SOCIAL_CIRCLE	DEF_30_CNT_SOCIAL_CIRCLE	0.868994
• REG_REGION_NOT_WORK_REGION	LIVE_REGION_NOT_WORK_REGION	0.847885
<ul><li>LIVE_CITY_NOT_WORK_CITY</li></ul>	REG_CITY_NOT_WORK_CITY	0.778540
<ul> <li>AMT_ANNUITY</li> </ul>	AMT_GOODS_PRICE	0.752699
• AMT_CREDIT	AMT_ANNUITY	0.752195
<ul> <li>DAYS_EMPLOYED</li> </ul>	FLAG_DOCUMENT_6	0.617307

#### CORELATION MATRIX FOR DEFAULTERS

SK_ID_CURR	1	0.053	0.048	-0.05	0.039	0.015	0.053	-0.04	0.049	- 1.6
OBS_60_CNT_SOCIAL_CIRCLE	0.053	1	0.051	0.51	0.029	0.031	1	-0.03	0.57	-14
AMT_ANNUITY	0.048	0.051	1	-0.13	0.95	0:088	0.051	0.95	0.12	-12
DEF_60_CNT_SOCIAL_CIRCLE	-0.05	0.51	-0.13	1	-0.12	0.05	0.5	0.11	0.99	-10
AMT_CREDIT	0.039	0.029	0.95	-0.12	1	0.051	-0.03	1	-0.11	
DAYS_EMPLOYED	0.015	0.031	0.088	0.05	0.051	1	0.031	0.056	0.046	- 0.8
OBS_30_CNT_SOCIAL_CIRCLE	0.053	1	0.053	0.5	-0.03	0.031	1	-0.03	0.57	- 0.6
AMT_GOODS_PRICE	-0.04	-0.03	0.95	-0.11	1	0.056	-0.03	1	-0.11	- 0.4
DEF_30_CNT_SOCIAL_CIRCLE	0.049	0.57	-0.12	0.99	-0.11	0.046	0.57	0.11	1	- 0.2
	SK_ID_CURR-	OBS_60_CNT_SOCIAL_CIRCLE -	AMT ANNUITY	DEF_60_CNT_SOCIAL_CIRCLE -	AMT_CREDIT -	DAYS_EMPLOYED -	OBS_30_CNT_SOCIAL_CIRCLE -	AMT_GOODS_PRICE -	DEF_30_CNT_SOCIAL_CIRCLE -	V.E.

# **Top 10 correlation for Non Defaulters**

• SK_ID_CURR	SK_ID_CURR	1.000000
• OBS_60_CNT_SOCIAL_CIRCLE	OBS_30_CNT_SOCIAL_CIRCLE	0.998508
• AMT_GOODS_PRICE	AMT_CREDIT	0.987250
<ul> <li>REGION_RATING_CLIENT_W_CITY</li> </ul>	REGION_RATING_CLIENT	0.950149
• CNT_CHILDREN	CNT_FAM_MEMBERS	0.878571
<ul><li>LIVE_REGION_NOT_WORK_REGION</li></ul>	REG_REGION_NOT_WORK_REGION	0.861861
• DEF_30_CNT_SOCIAL_CIRCLE	DEF_60_CNT_SOCIAL_CIRCLE	0.859332
<ul><li>REG_CITY_NOT_WORK_CITY</li></ul>	LIVE_CITY_NOT_WORK_CITY	0.830381
• AMT_ANNUITY	AMT_GOODS_PRICE	0.776686
• FLAG_EMP_PHONE	DAYS_BIRTH	0.622073

#### CORELATION MATRIX FOR NON DEFAULTERS



# **Summary and Recommendations**

- People most likely to repay the loan have secondary/Secondary special education status and a lower income range.
- Be careful while approving loan to low skilled laborers, male applicants and widows, and those who belong to 'co-apartment' categories. Suggestion is to give *low credit amount* to these category people, because as per our analysis, 'Customers who have low credit amount are more likely to pay back the loan.'
- While approving, we can ignore the social circle, family having children or not, though married people repay the loan at a higher rate.

- Cash loans is quite higher than Revolving Loans
- Applicants who are working are more likely to apply for loan.
- Working category people apply for loans when compared to unemployed, maternity leave and businessmen. Hence, bank can create schemes to encourage other category people to apply for loans.