

CREDIT EDA CASE STUDY

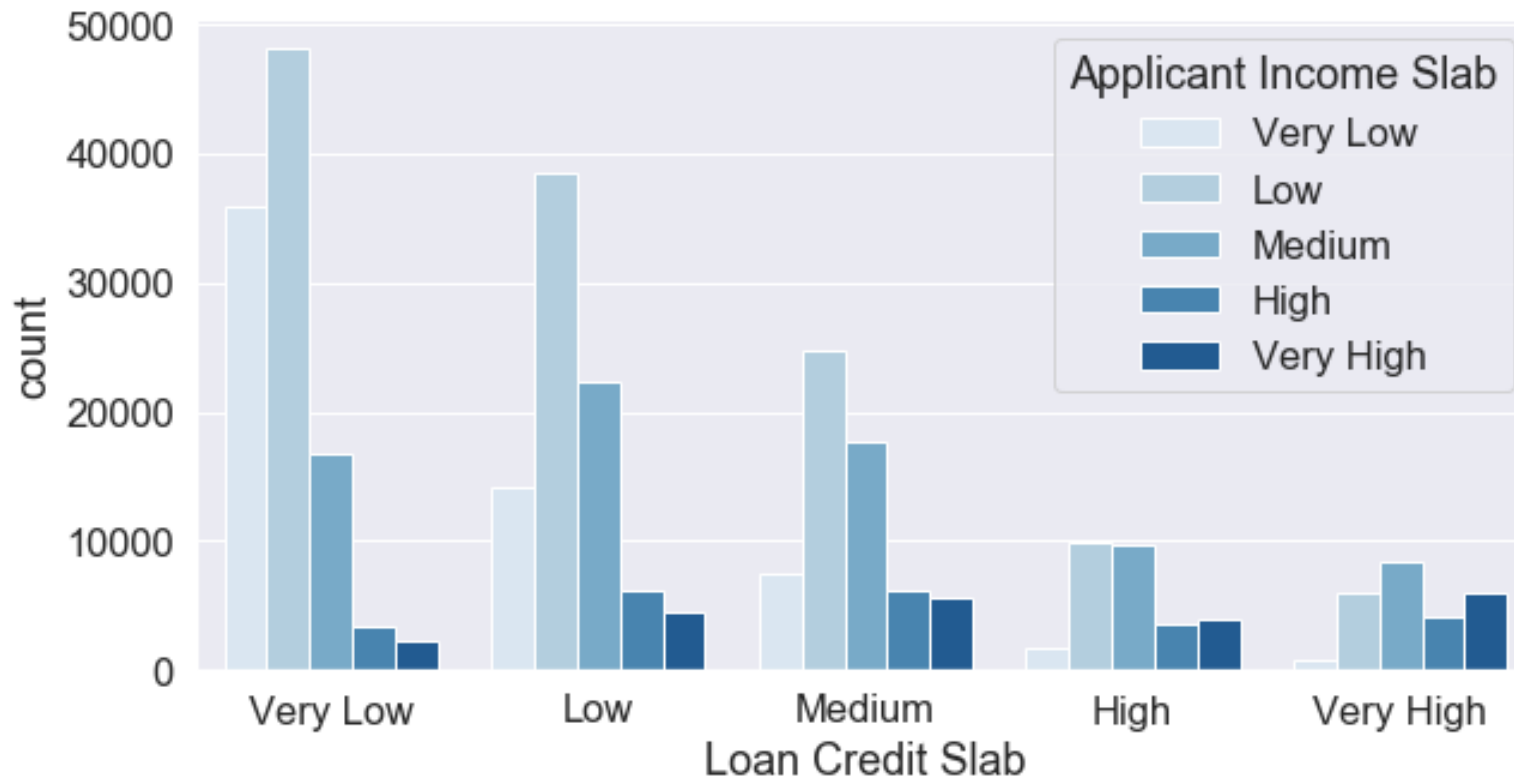
Group Name:

1. Karan Joseph
2. Meghana C

Business Objectives

- 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.

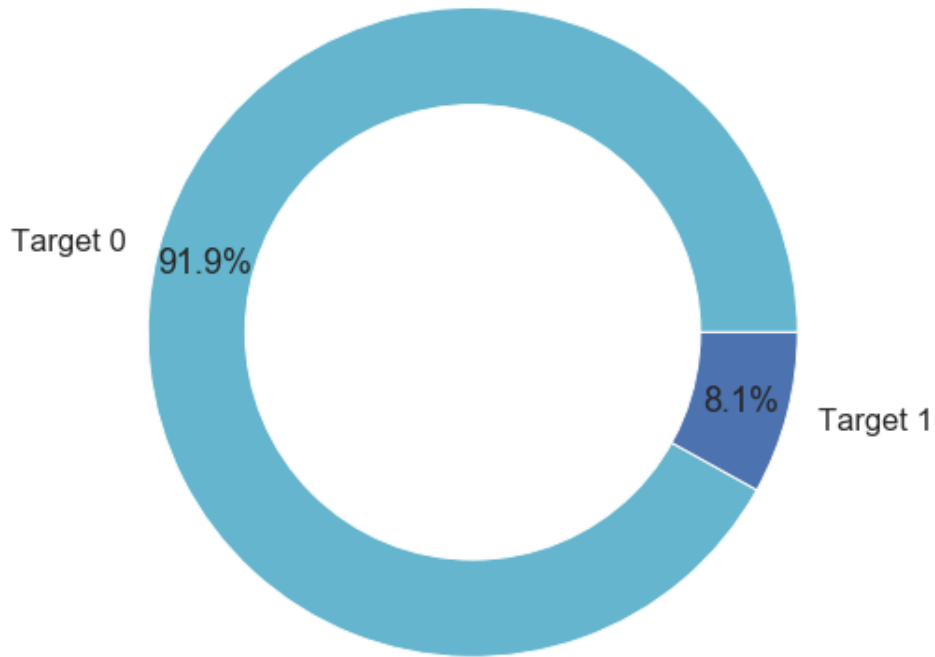
Applicant Income vs Loan Credit



- ❖ Most of loans applied are in the Very Low and Low categories, and the highest applicants in both such cases have low income slabs.
- ❖ Most applicants have either low salary or medium salary, followed by people with very low salary.
- ❖ Highest applicants in Very High loan category are from the Medium salary slab

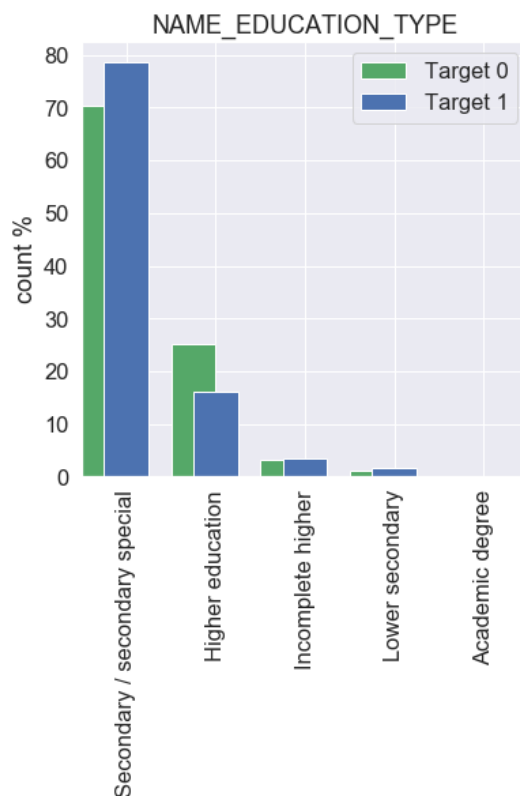
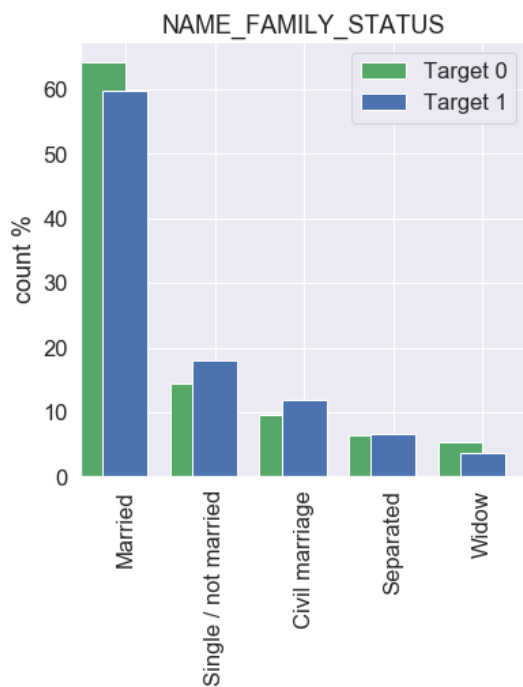
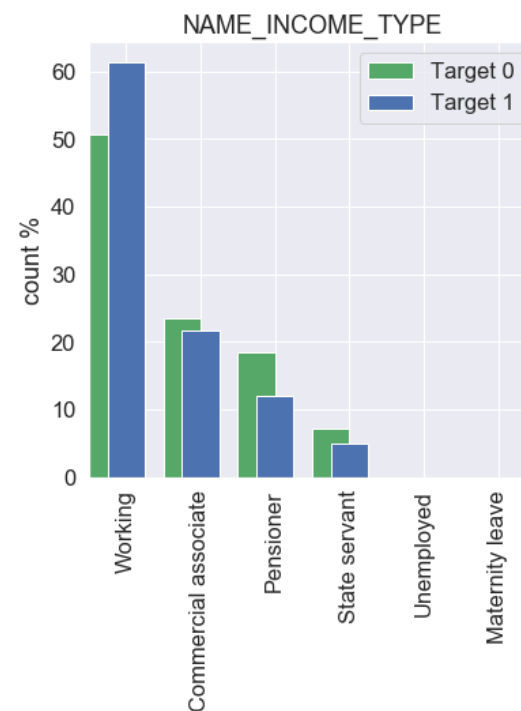
Data Analysis

– Data Imbalance Check:



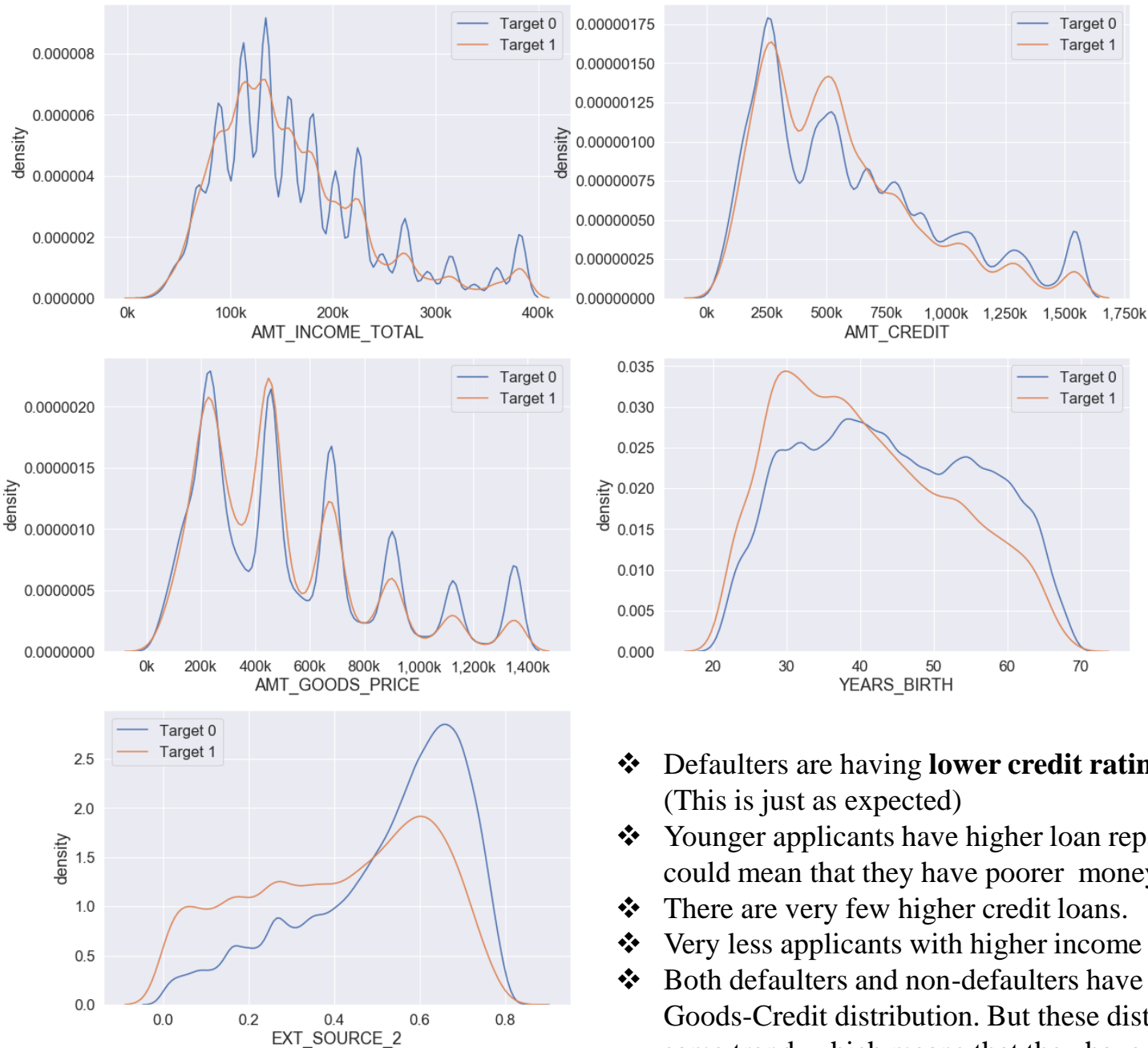
- ❖ 91.9% of the applicants do have any history late payments.
- ❖ These applicants are potentially capable of repaying the loans.
- ❖ This is **not an indication** of their future repayment capabilities; which spurs the need for further studies on the dataset.

Univariate analysis for Categorical Variables



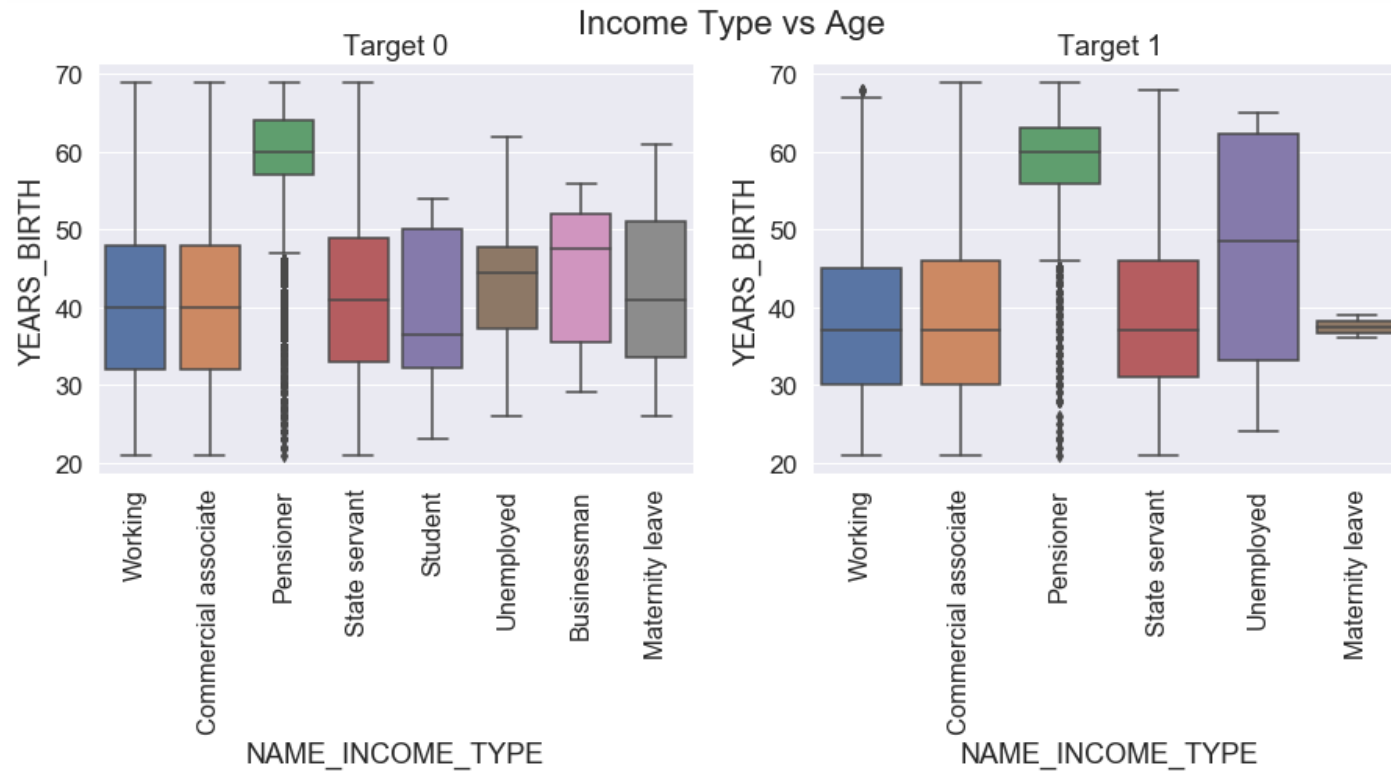
- ❖ Majority of the loan applicants are female.
- ❖ There are more males having **payment troubles**, even though there are fewer male loan applicants.
- ❖ Pensioner defaulters are fewer than non-defaulters.
- ❖ Majority of loan applicants are Working class.
- ❖ Single applicants have higher percentage of defaulters.
- ❖ Married applicants have more non-defaulters than defaulters.
- ❖ People Owning Cars have lesser default ratio.
- ❖ Majority of the applicants are Married.
- ❖ Majority applicants have lesser academic qualifications.
- ❖ **Default rate is higher for applicants with just Secondary education.**

Univariate analysis for Numerical Variables

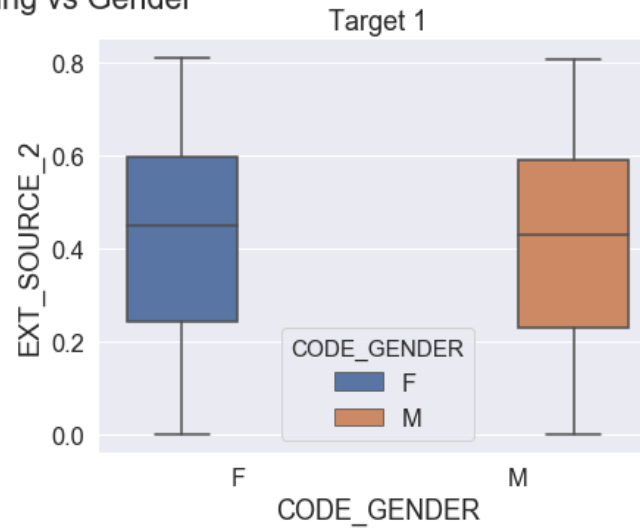
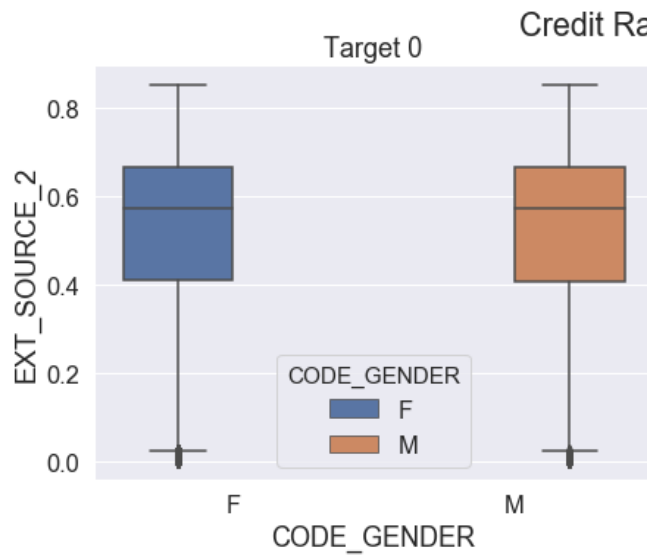


- ❖ Defaulters are having **lower credit ratings** than non-defaulters. (This is just as expected)
- ❖ Younger applicants have higher loan repayment troubles. This could mean that they have poorer money management skills.
- ❖ There are very few higher credit loans.
- ❖ Very less applicants with higher income slabs.
- ❖ Both defaulters and non-defaulters have almost similar Income-Goods-Credit distribution. But these distributions follow the same trend, which means that they have a strong correlation between them.

Bivariate Analysis

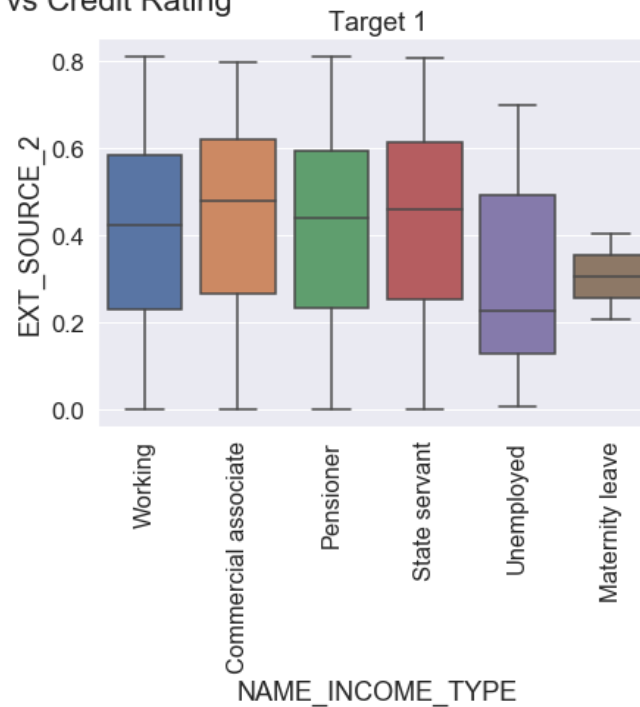
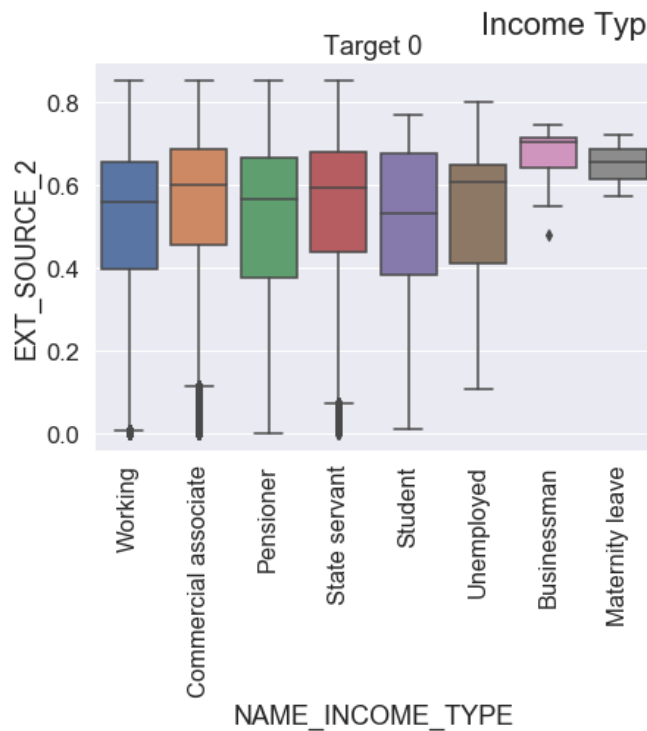


- From the above plots which shows the bivariate analysis for both Target 0 and Target 1, we can state the following:
 - ❖ **Businessmen** have highest age cap among all the salaried applicants.
 - ❖ Giving loans to **Unemployed** applicants with more than 50 years of age is the riskiest(more chance to default).
 - ❖ Most of the **working** applicants are of the age group 30-50 years.



❖ Credit Ratings vs Gender

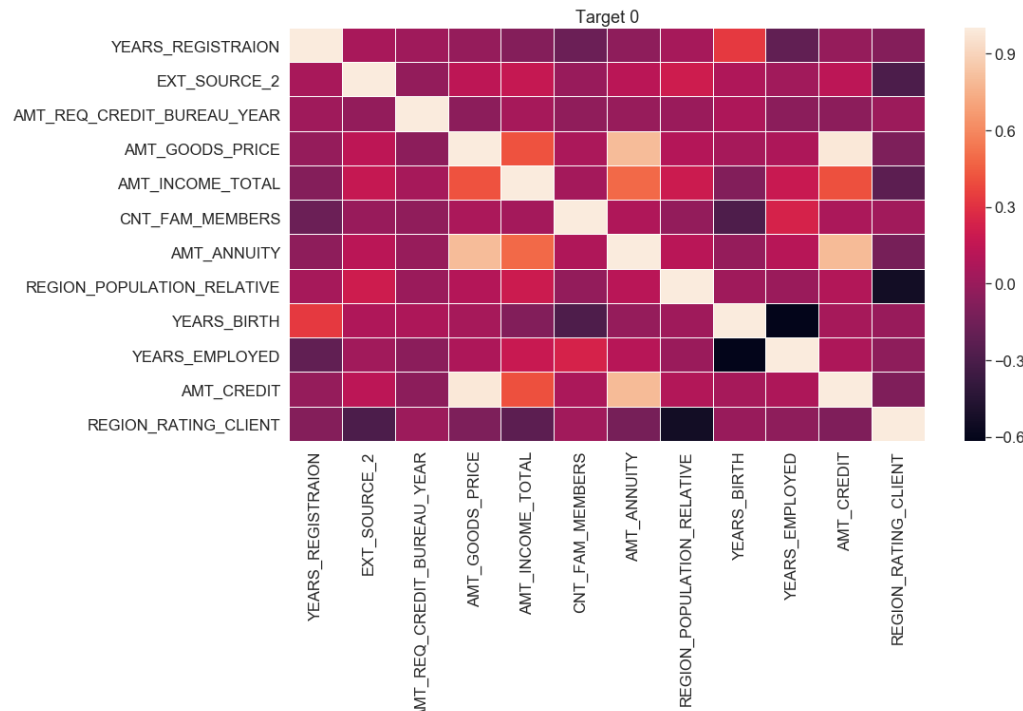
Credit rating does not depend on the gender of the applicant, but on their loan repayment capabilities; as evident from the box plots.



❖ Job vs Credit Ratings

- **Businessmen have the highest credit ratings.**
- Unemployed applicants with less than 0.4 credit rating are very risky candidates.
- Women on Maternity leave, but with less credit ratings are risky applicants.

Correlation Matrix

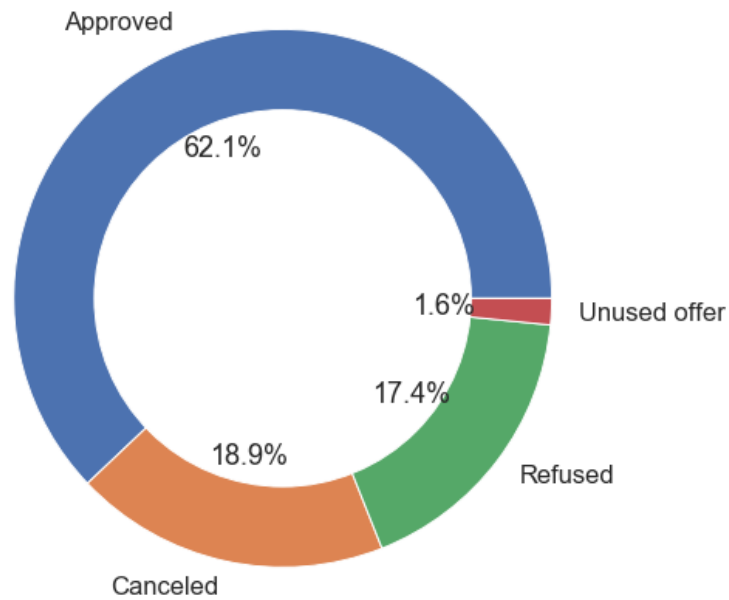


➤ **Correlation matrix is not the absolute method to distinguish between which variable has the highest impact in identifying who will default or not.**

But it can help highlight some interesting interactions:

- Loan Credit amount is highly correlated with the value of the collateral goods. The amount of loan you get is directly dependent on the value of the goods one submits.
- Annuity has good correlation with Loan Amount. Hence as loan amount increase, annuity amount will also increase.
- Age of the employee and years employed has a moderate correlation.
- High populous regions have better region rating.
- Annuity amount has a decent correlation with the income of the applicant. Applicants with more income might repay the loan at higher annuity amounts and lesser periods.

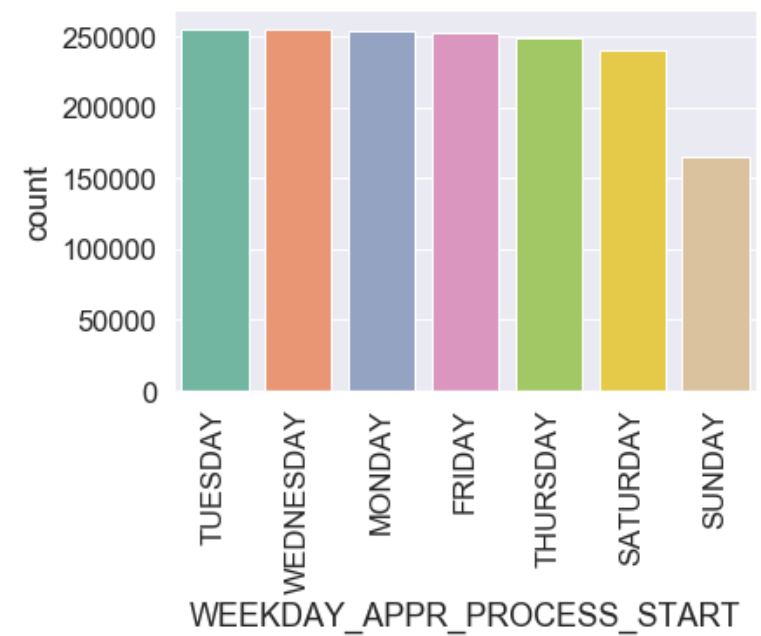
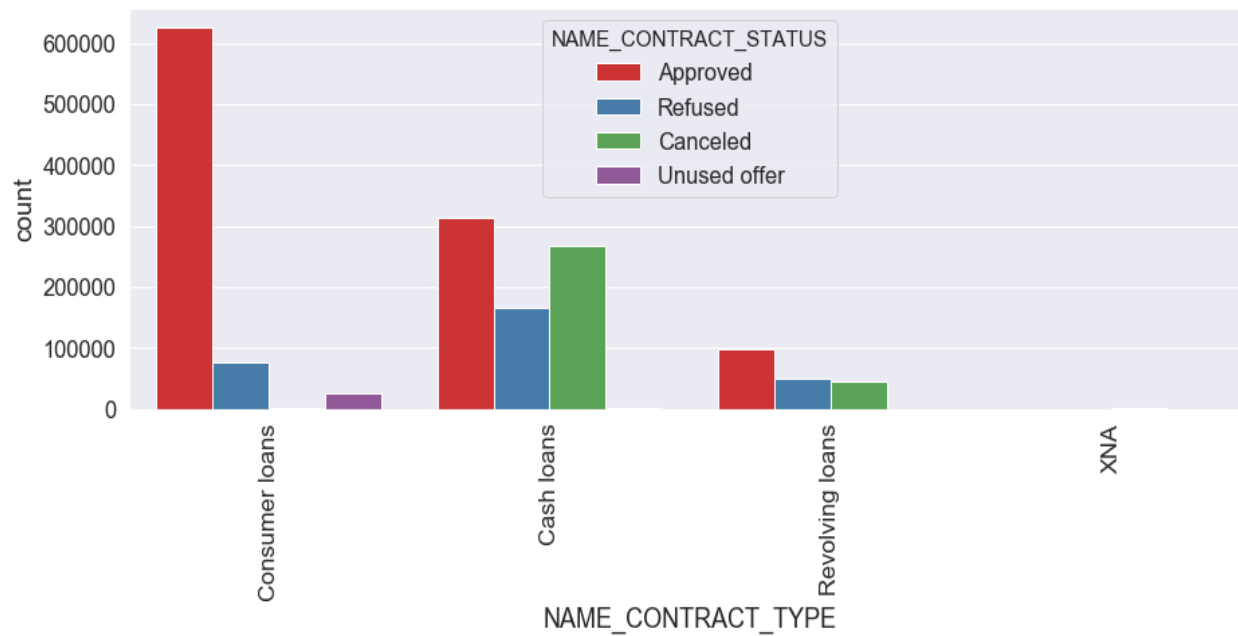
VAR1	VAR2	CORR
AMT_CREDIT	AMT_GOODS_PRICE	0.986680
AMT_ANNUITY	AMT_GOODS_PRICE	0.799405
AMT_CREDIT	AMT_ANNUITY	0.796758
YEARS_EMPLOYED	YEARS_BIRTH	0.617763
REGION_RATING_CLIENT	REGION_POPULATION_RELATIVE	0.539005
AMT_ANNUITY	AMT_INCOME_TOTAL	0.489586
AMT_INCOME_TOTAL	AMT_GOODS_PRICE	0.416020
AMT_CREDIT	AMT_INCOME_TOTAL	0.411285
YEARS_BIRTH	YEARS_REGISTRAION	0.332980
REGION_RATING_CLIENT	EXT_SOURCE_2	0.291624



62% of the previous applications were approved and only 17% of the loan applicants were refused.

- **Repeater clients and Refresher clients apply for the biggest loans.**
- New clients generally apply for smaller amounts.





- ❖ Most loan applications occur along Tue, Wed, Mon, Fri and Thurs. Hence, weekdays are much busier than weekends.
- ❖ **Cash loans have the highest rejection and cancellation chances.** Hence, most of the applied loans are consumer loans. They have the least chance of cancellation; and only a small percentage of consumer loans are refused.
- ❖ **New clients have the lowest loan refusal rate.** However, majority of the loan applicants are Repeaters; and they have the highest chance of loan cancellation. Repeater clients have a nearly 40% chance that their application will get rejected.

Conclusion

- Some of the inferences that can be made based on our analysis are:
 - Giving loans to Unemployed applicants with more 50 Years of age is the riskiest.
 - There are more youngsters having higher payment issues.
 - Default rate is higher for applicants with just Secondary education
 - Single/ not married defaulters are more than non-defaulters.
 - Repeater and Refresher clients apply for the biggest loans.
 - New clients generally apply for smaller amounts.
 - Repeater clients has nearly 40% chance of application getting rejected.
 - Consumer loans have the least chance of getting cancelled.
- ➔ From these inferences, the loan providing companies can minimize the risk of losing money while lending to customers.