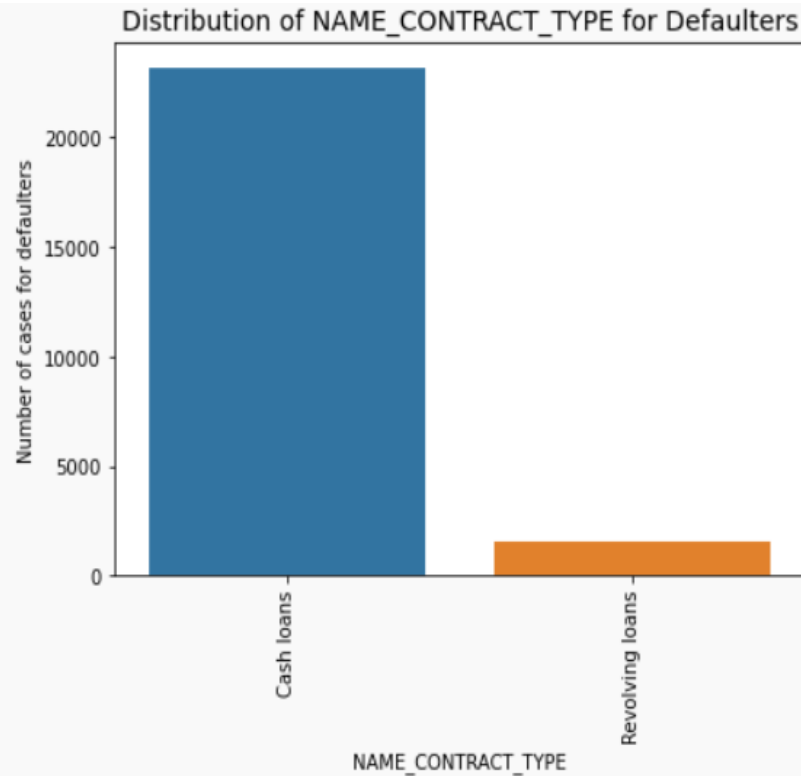
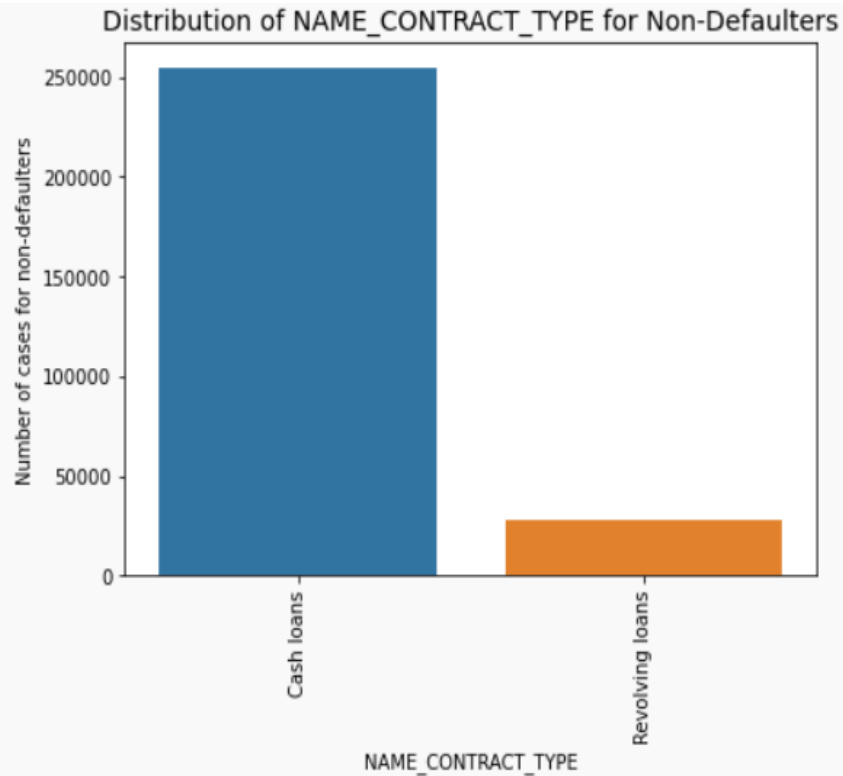




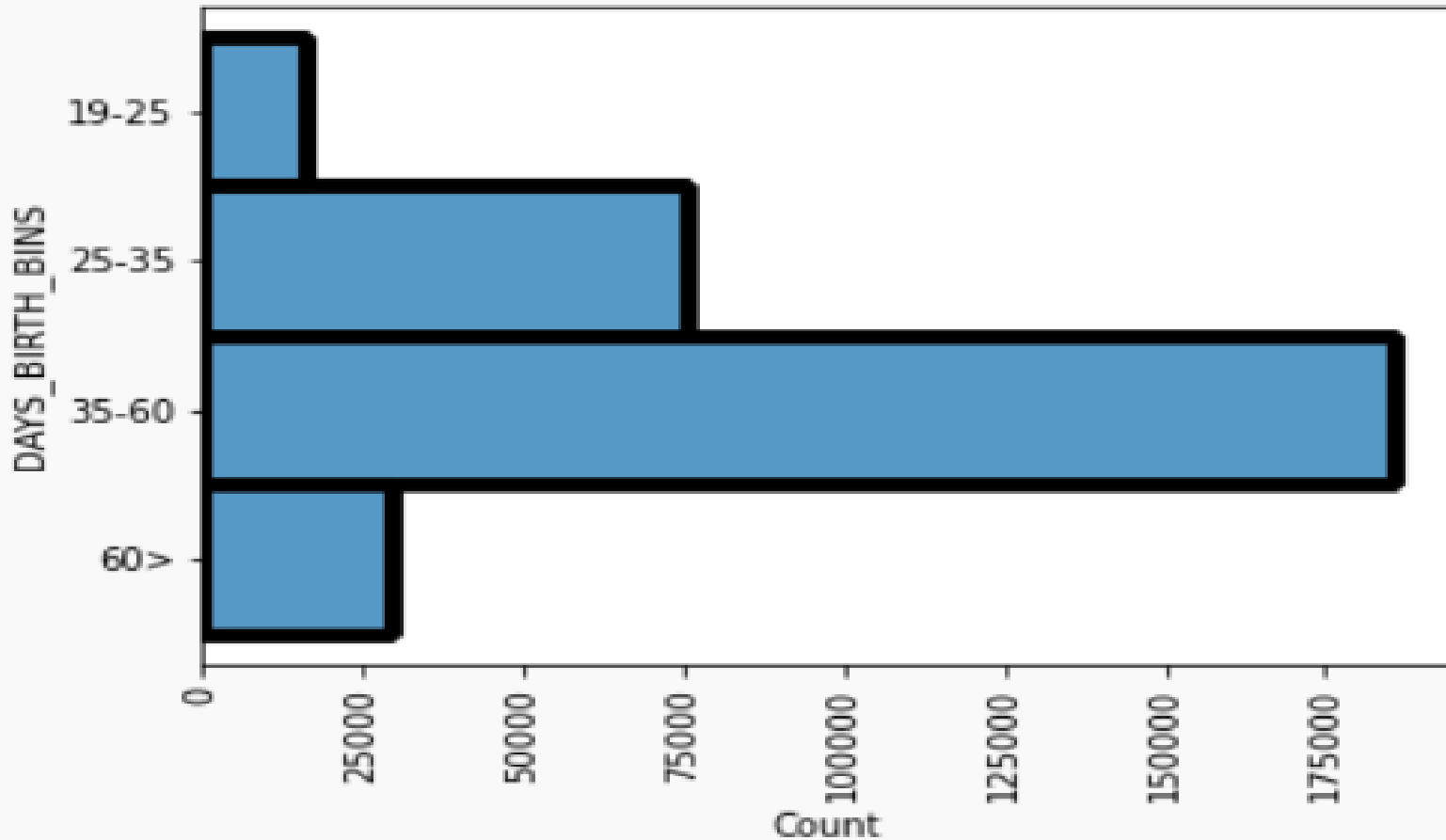
# CREDIT EDA CASE STUDY

By- Syed Sha Khalid



# Cash Loans vs Revolving Loans

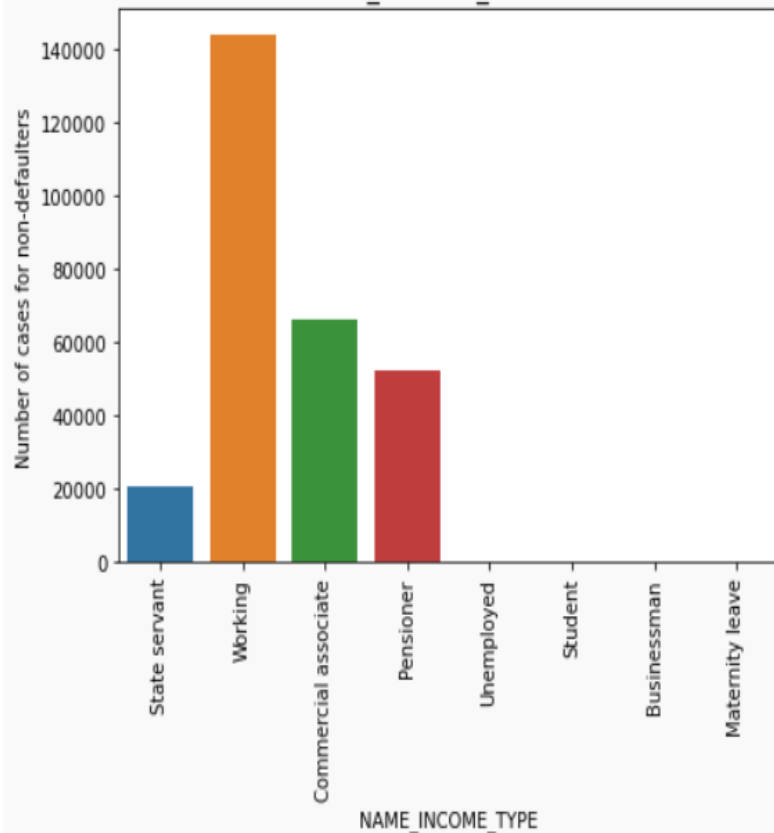
- We can notice that revolving loans are lesser in the defaulted population.
- Hence we can infer that revolving loans are comparatively safer.
- This may be attributed to the Nature of revolving loan as it is considered a flexible financing tool due to its repayment and re-borrowing flexibility



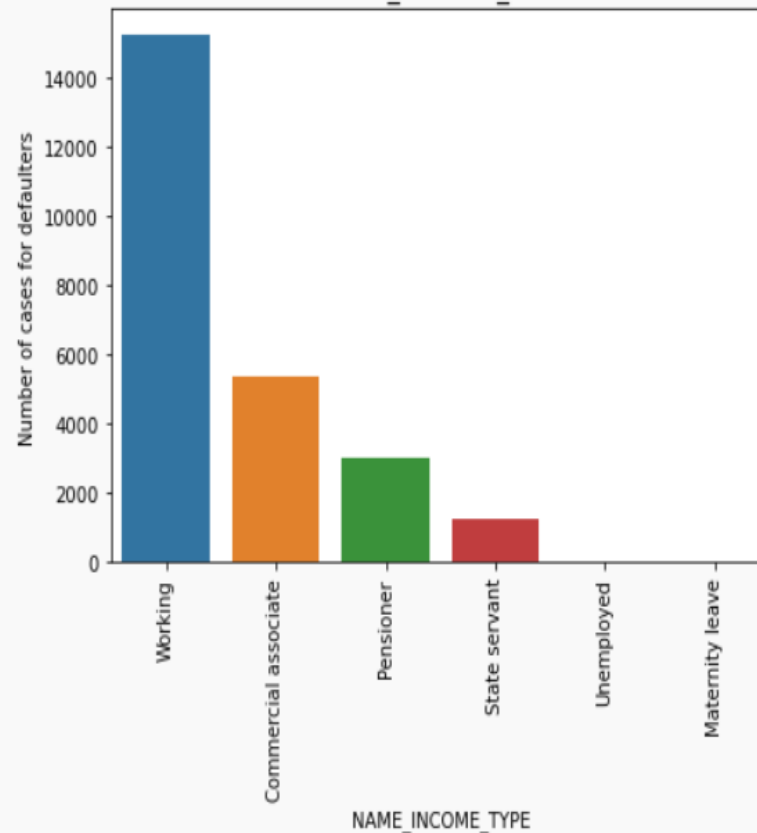
## Age analysis after binning

- We can observe that most of our population lies between the age of 35-60.
- The least amount population lies in between the age group 19-25.

Distribution of NAME\_INCOME\_TYPE for Non-Defaulters

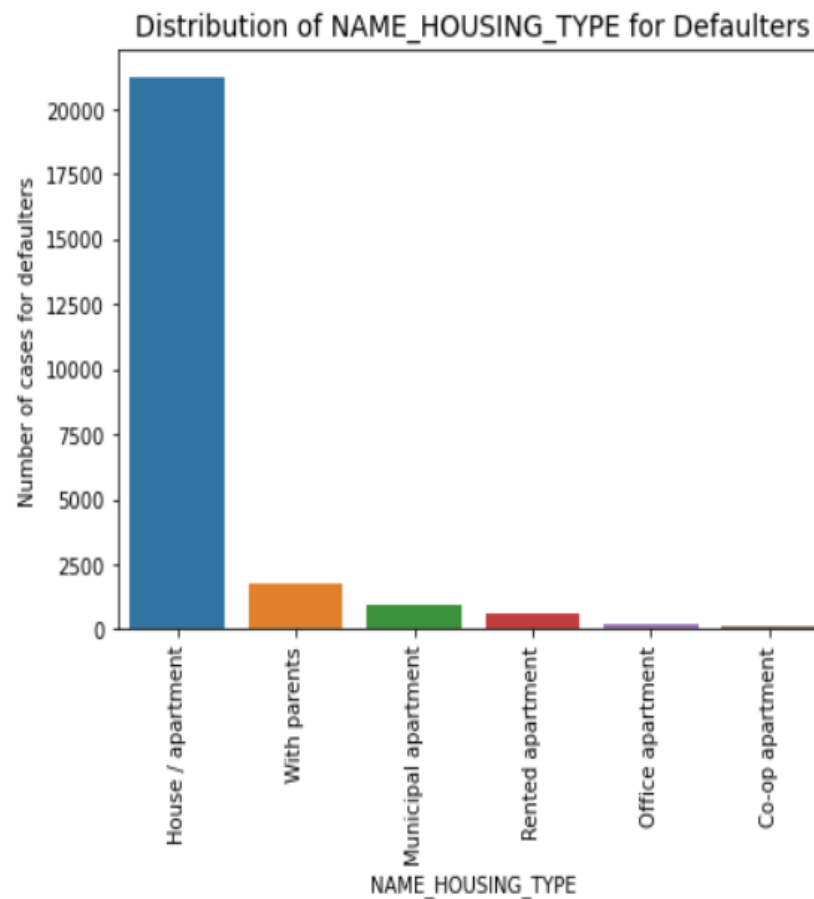
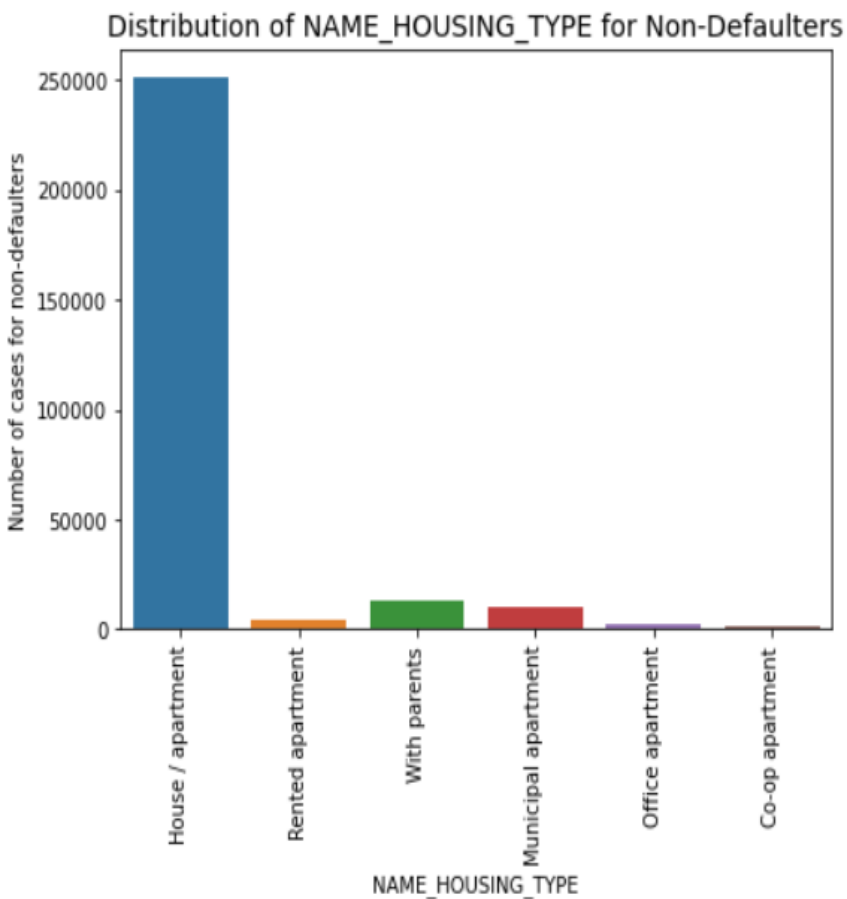


Distribution of NAME\_INCOME\_TYPE for Defaulters



# Income Types Distribution

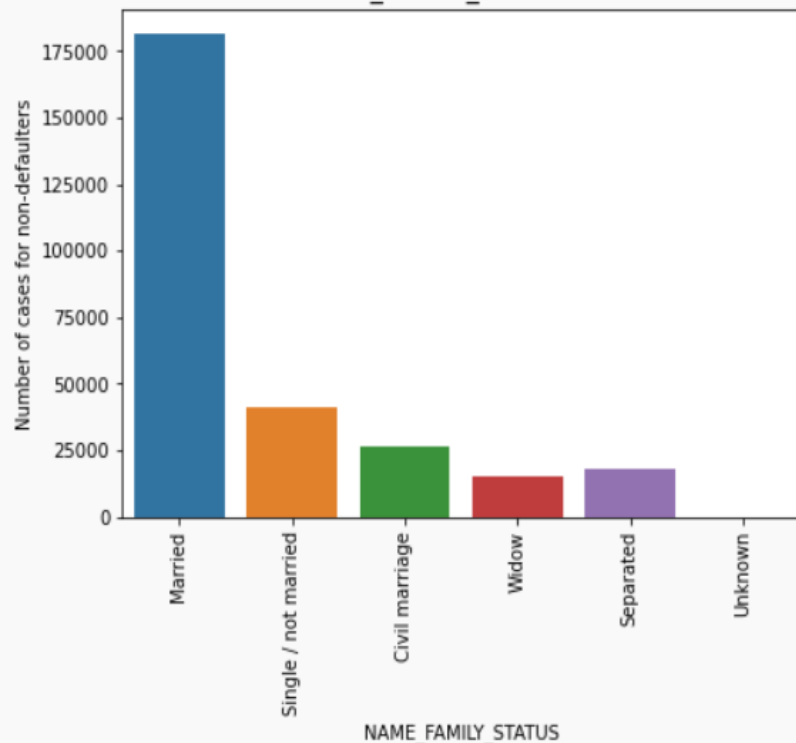
- Working population has the highest amount of defaulters
- State Servant have very low default population as they have secure jobs with steady income.



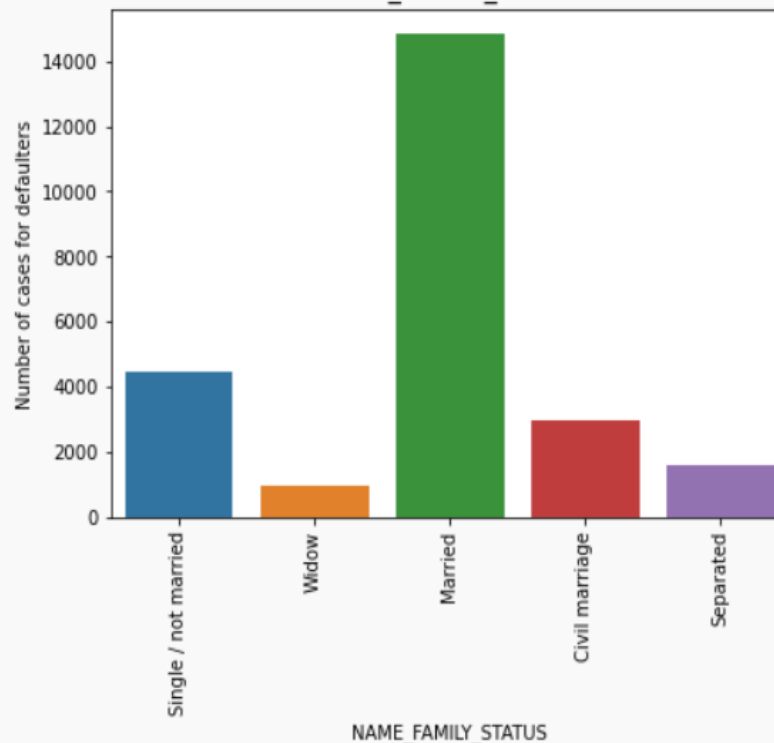
# Housing type distribution

- Population living in Rented apartments and those living with parents have higher default rate.
- Living in rental apartment means a cash outflow towards rent and thus less cash left for repayment of loan.
- Living with parents may suggest that the income is not too high and thus difficulty in repayment of loan.

Distribution of NAME\_FAMILY\_STATUS for Non-Defaulters



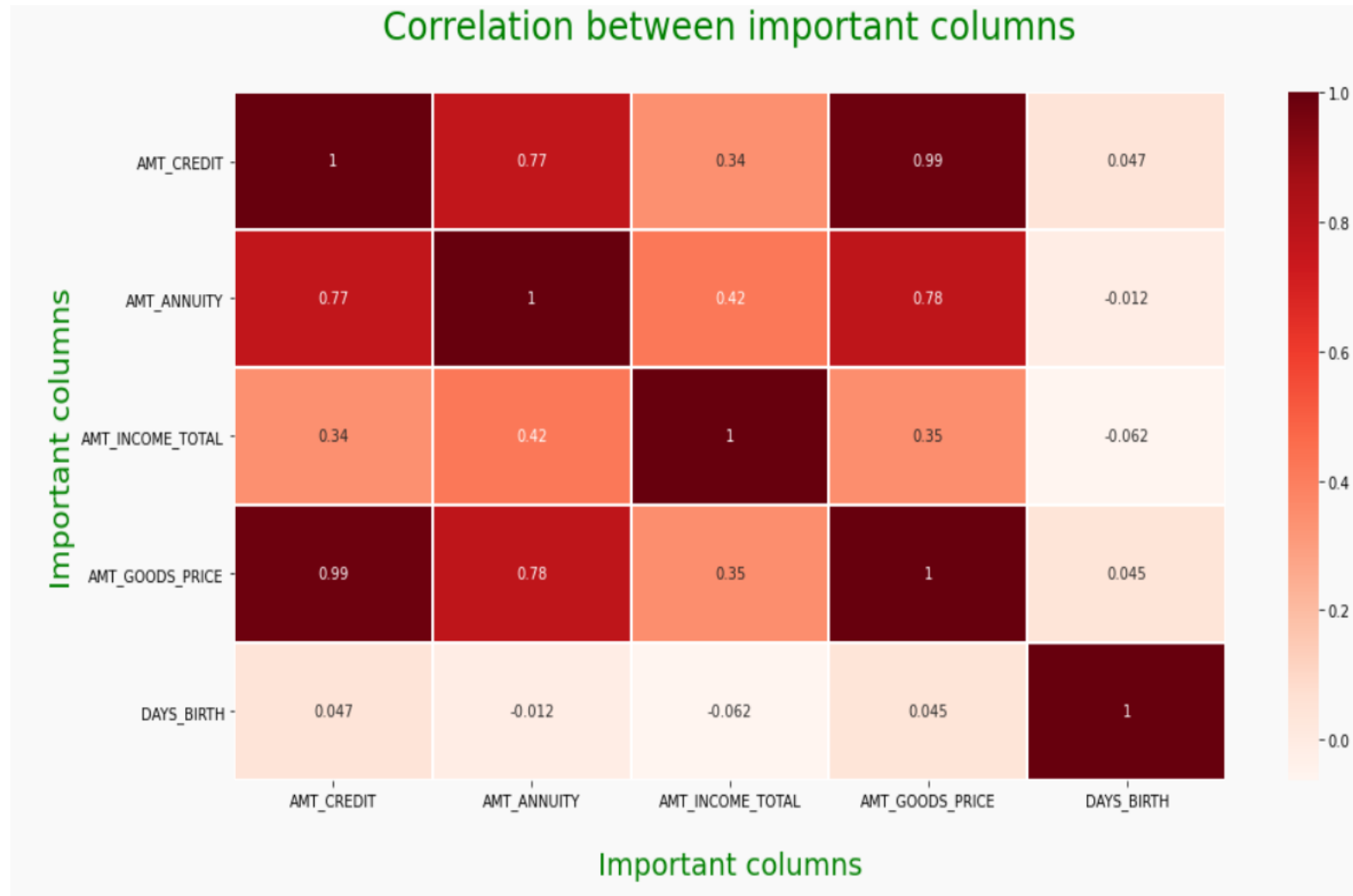
Distribution of NAME\_FAMILY\_STATUS for Defaulters



# Family Status Distributions

- The highest amount of non-defaulters are married people.
- On the other side we can also observe the highest amount from defaulters population are married.
- The least amount of defaulters are widow and separated.

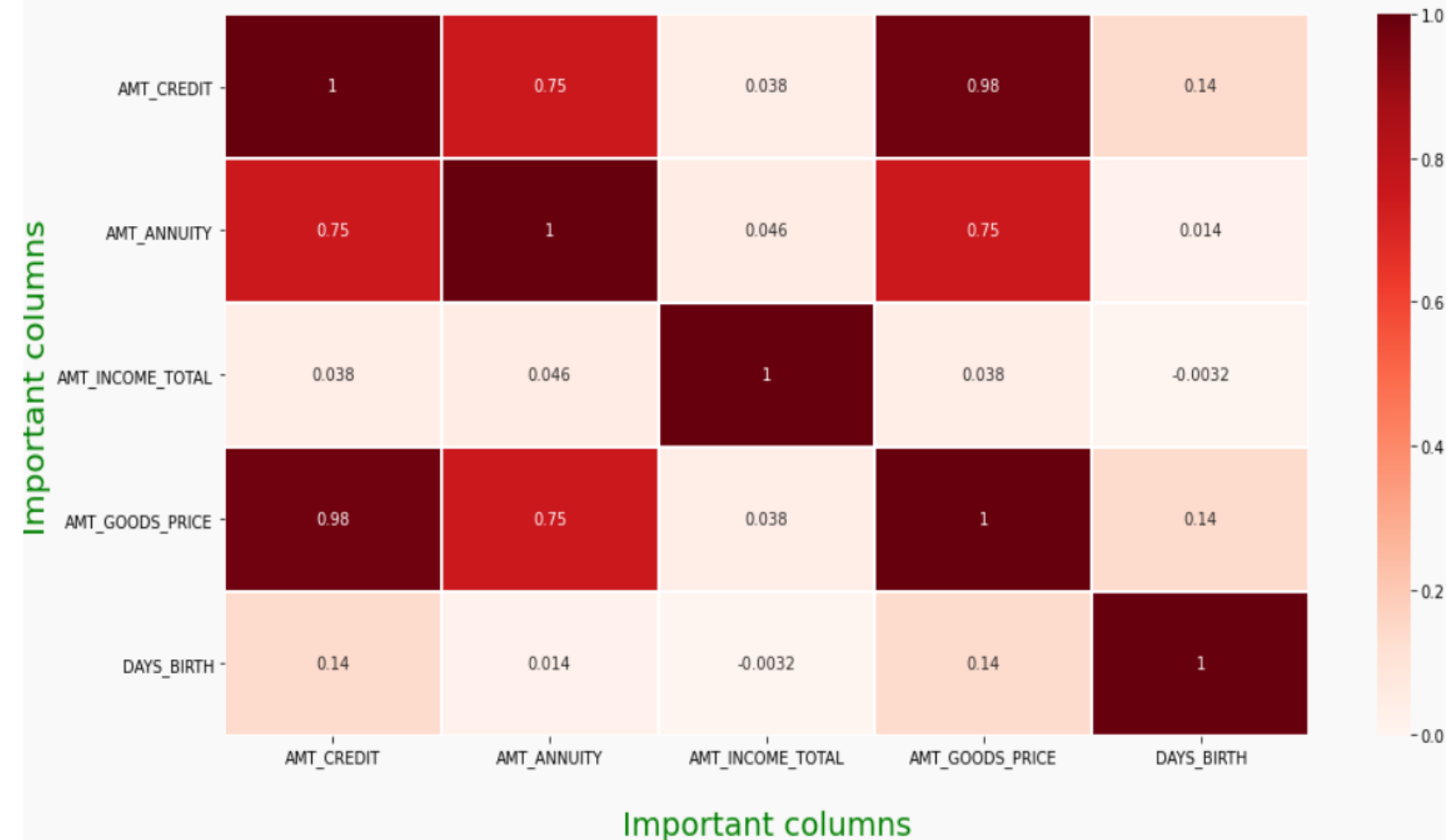
# Correlation between Important columns (Target 0)



- Credit amount is related to Goods price as the higher the price the higher the amount of credit taken by customer and vice versa.
- Annuity amount and Credit amount is also related as higher credit amount has a higher annuity.
- Therefore annuity amount and Goods price amount is also correlated as higher price of goods results in higher credit amount which eventually results in higher annuity amount.

# Correlation between Important columns (Target 1)

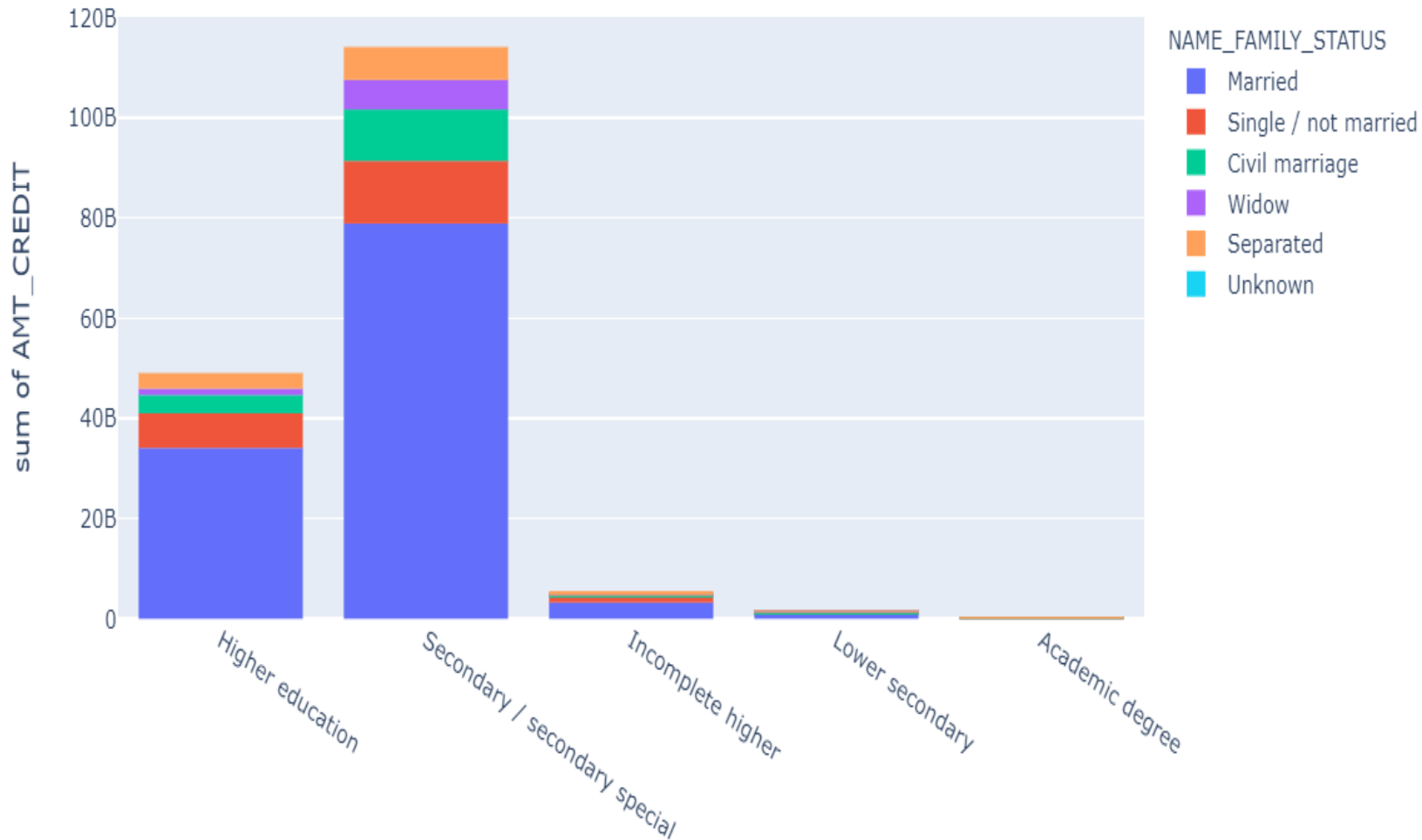
Correlation between important columns



- The Target 1 population are people with payment difficulties.
- Target 1 heat map looks similar but there are a few changes like there is very little relation between Amount of total income and amount of annuity compared to Target 0 population



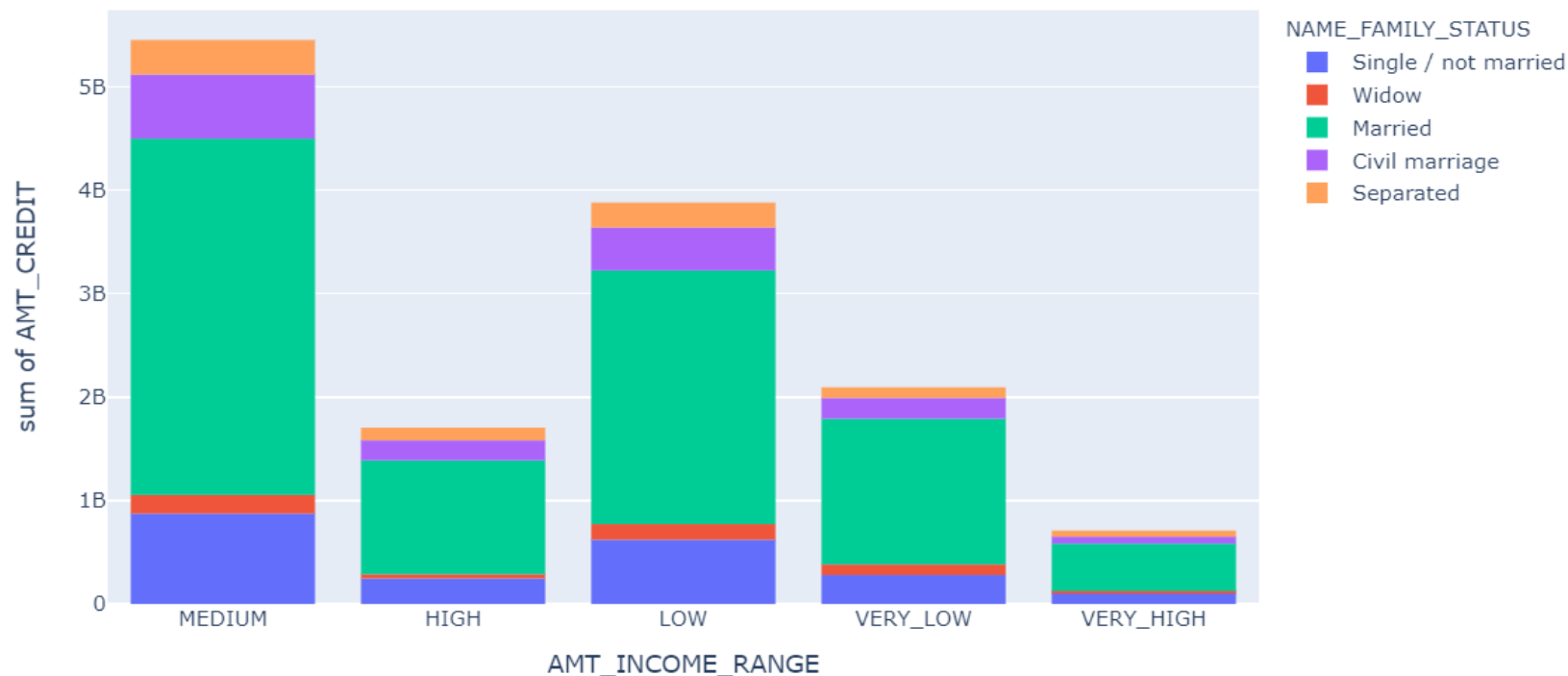
Credit amount vs Education of Loan-Payment Difficulties



## Credit amount vs Education of loan-Payment difficulties

- People with secondary and higher education have higher credit amount compared to other categories.
- The majority of higher and secondary credit amount is occupied by married people

Income range vs Credit amount of Loan Payment Difficulties

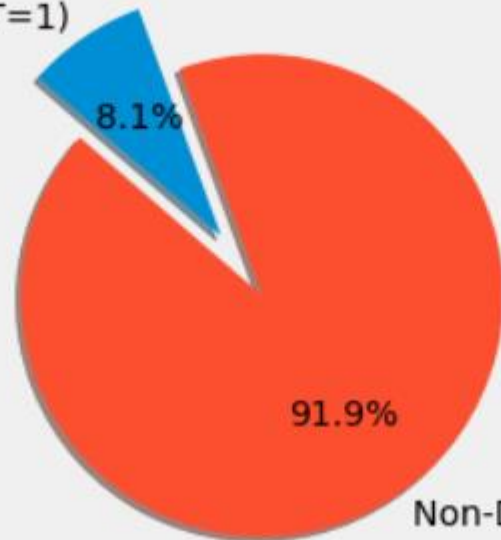


## Income range vs Credit amount

- The highest amount of credit is for people with medium income, followed by people with low income.
- The lowest amount of credit is for people with very high income, as people with very high income rarely need credit or loans.

## Data imbalance

Defaulted Population(TARGET=1)

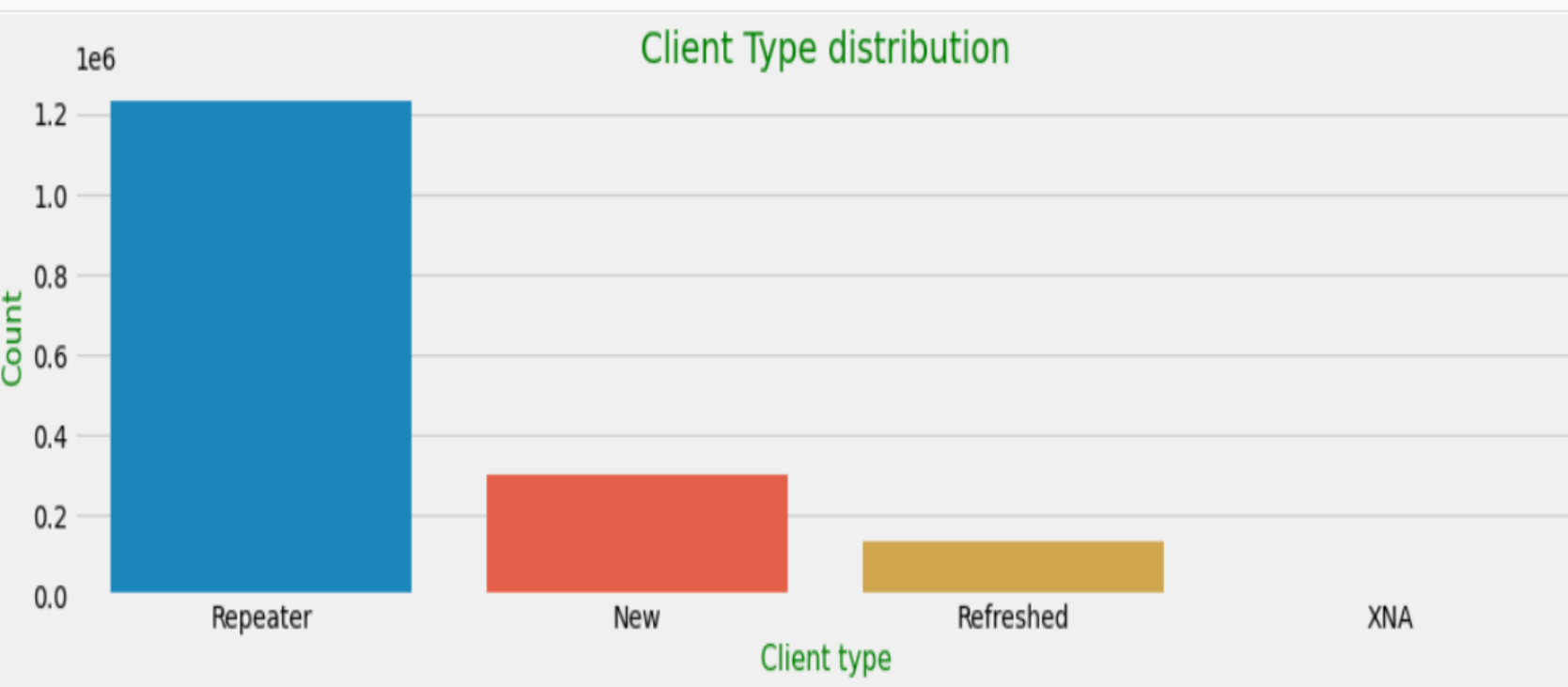


Non-Defaulted Population(TARGET=0)

## Data Imbalance

- **More than 90% of the population consists of people who are Target=0, i.e. people with no payment difficulties/defaulted on loans.**

# Previous Data Application Analysis.



## Client type while applying for the previous application

- We can observe that the client type is highest for repeater while applying for the previous application.
- The second highest population is for the new client type followed by refreshed.

# Bivariate Analysis After Merging Previous Data

% of Loan Payment Difficulties for NAME\_CONTRACT\_STATUS and NAME\_CLIENT\_TYPE

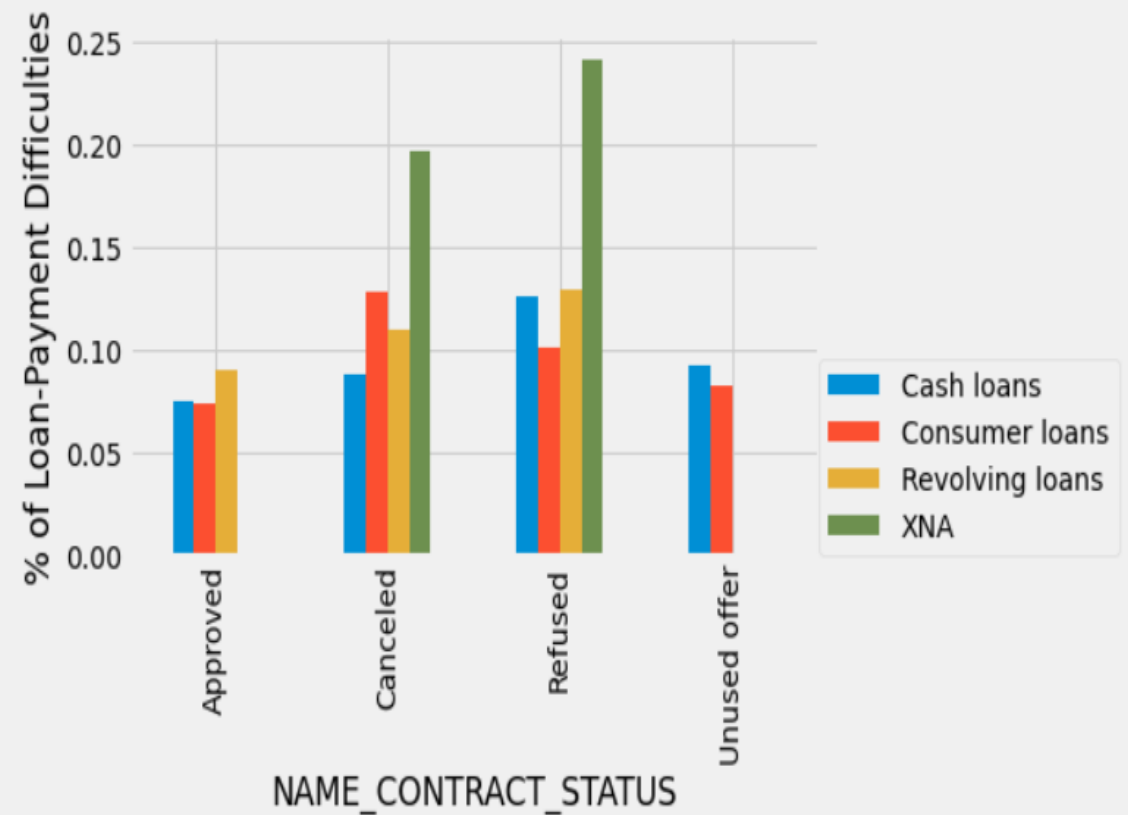


## Contract status vs Client type

- The highest amount of approved loans are of new client type.
- The highest amount of cancelled loans are also of new client type.
- The highest amount of refused loans are also of repeater client type.
- The highest amount of unused offer are of new client type

# Bivariate Analysis After Merging Previous Data

% of Loan Payment Difficulties for NAME\_CONTRACT\_STATUS and NAME\_CONTRACT\_TYPE



## Contract status vs Contract type

- The highest amount approved loans are revolving loans.
- The highest amount of cancelled loans are consumer loans
- The highest amount of refused loans are revolving loans.
- The most unused offer are cash loans.

# Conclusion

- Banks should focus less on income type 'Working' as they are having most number of unsuccessful payments.
- Banks should focus more on contract type 'Student' , ' pensioner' , 'state servant' and 'Businessman' for successful payments.
- Banks should accept more revolving loans as they are less defaulted compared to cash loans.
- Banks should also focus on people with Family status as 'widow' and 'separated' as they have the least amount of defaulters compared to other Family Status.