

# CREDIT EDA ASSIGNMENT

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# Problem Statement

- Use data minimize the risk of losing money while lending to customers
- Application data contains information about the customers who have taken loans
- 1 indicates that they had difficulty paying back whereas 0 indicates that they paid all the dues on time
- Other data set contains details about previous loan applications, indicating whether the loan was approved, refused, rejected or unused
- Using these data sets we need to help make a smart decision that whether an application should be rejected or accepted

# Assumptions

- 'XNA' in both the datasets means Not Available
- 'XAP' in both the datasets means Not Applicable

# Overall Approach

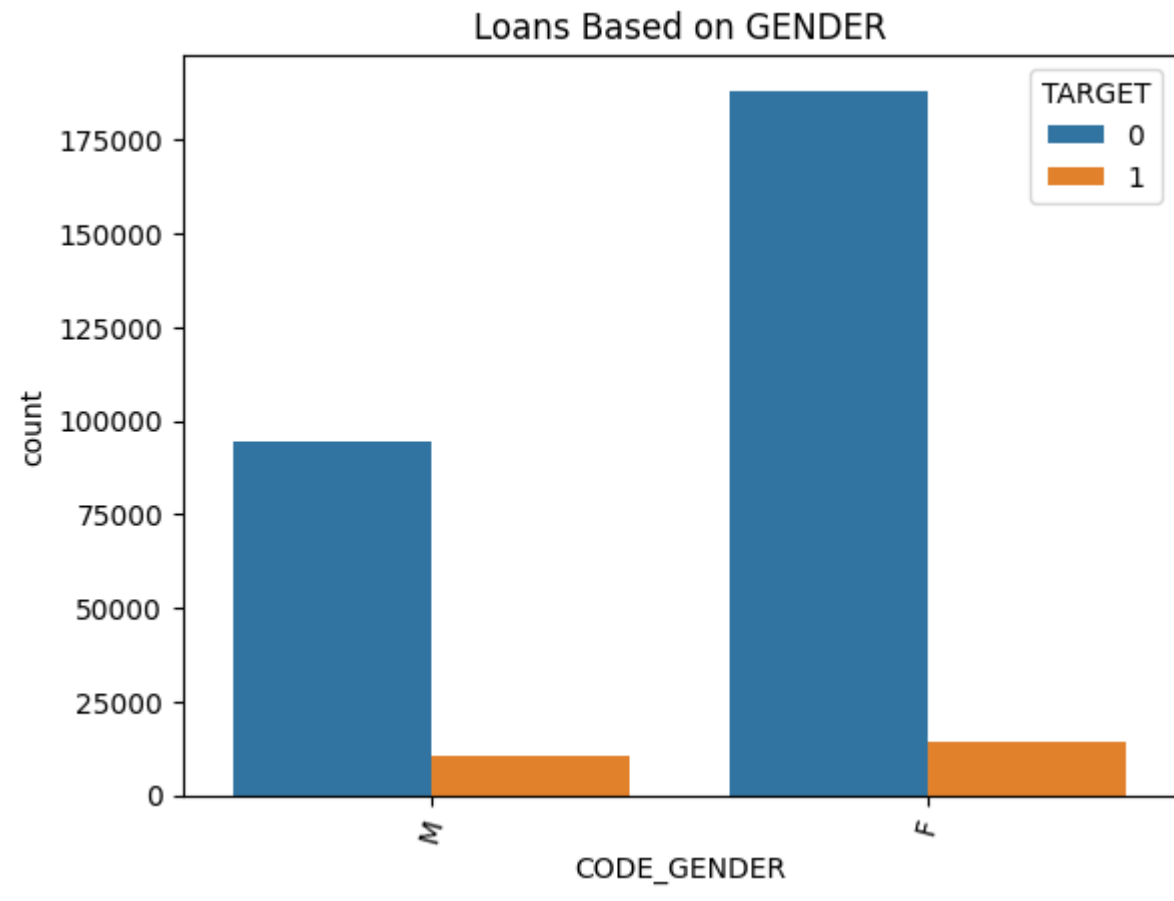
- We import the data present in csv files into data frames
- We analyse data and impute all the blank values
- We clean all the incorrect data present in the files
- We checked for outliers and how to impute them
- Once the data is ready, we perform univariant and bivariant analysis on it
- Once we have performed analysis on both the data sets, we merge the data in a single data frame
- We perform analysis on the merged data to get more insights

# Univariate Analysis

# Number of Credits taken based on gender

From the graph we can observe-

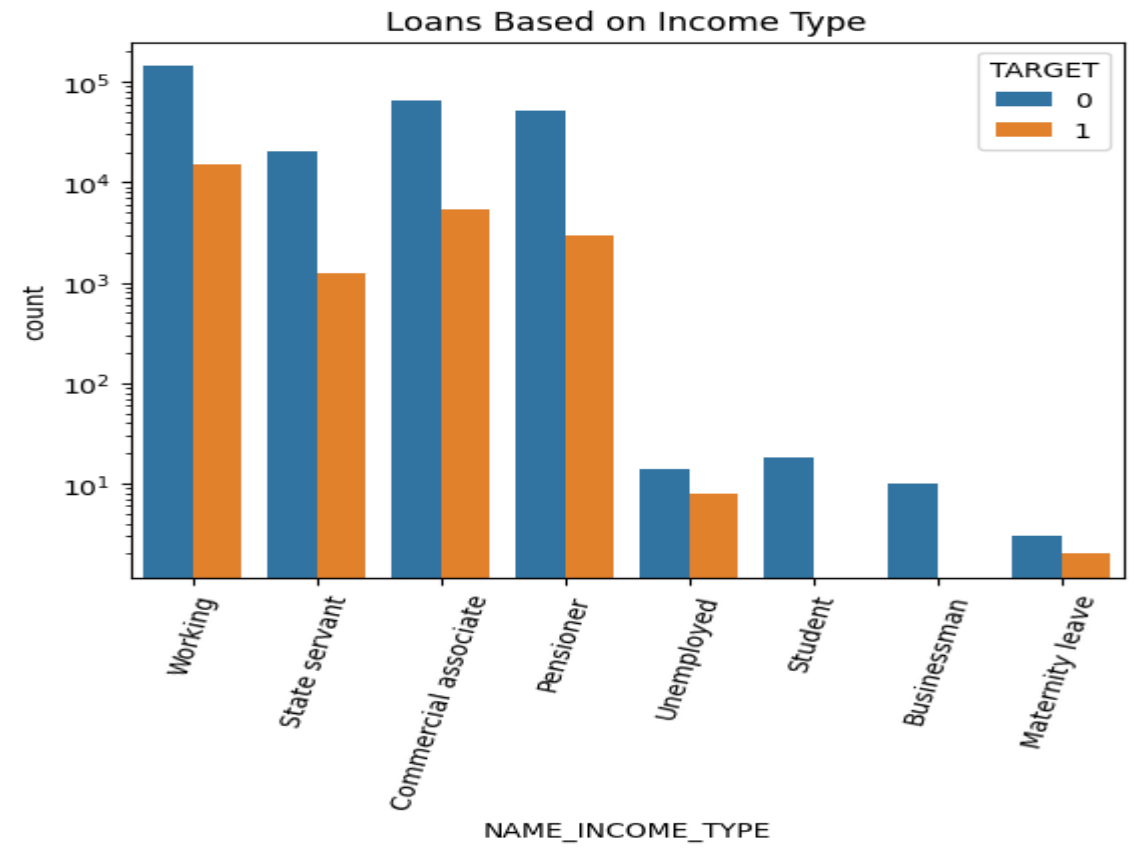
- ▶ Females take more number of credits than males
- ▶ Females have more successful loan payments than males
- ▶ Females also made slightly more unsuccessful payments than males



# Credits based on Income Type

From this we can observe the following-

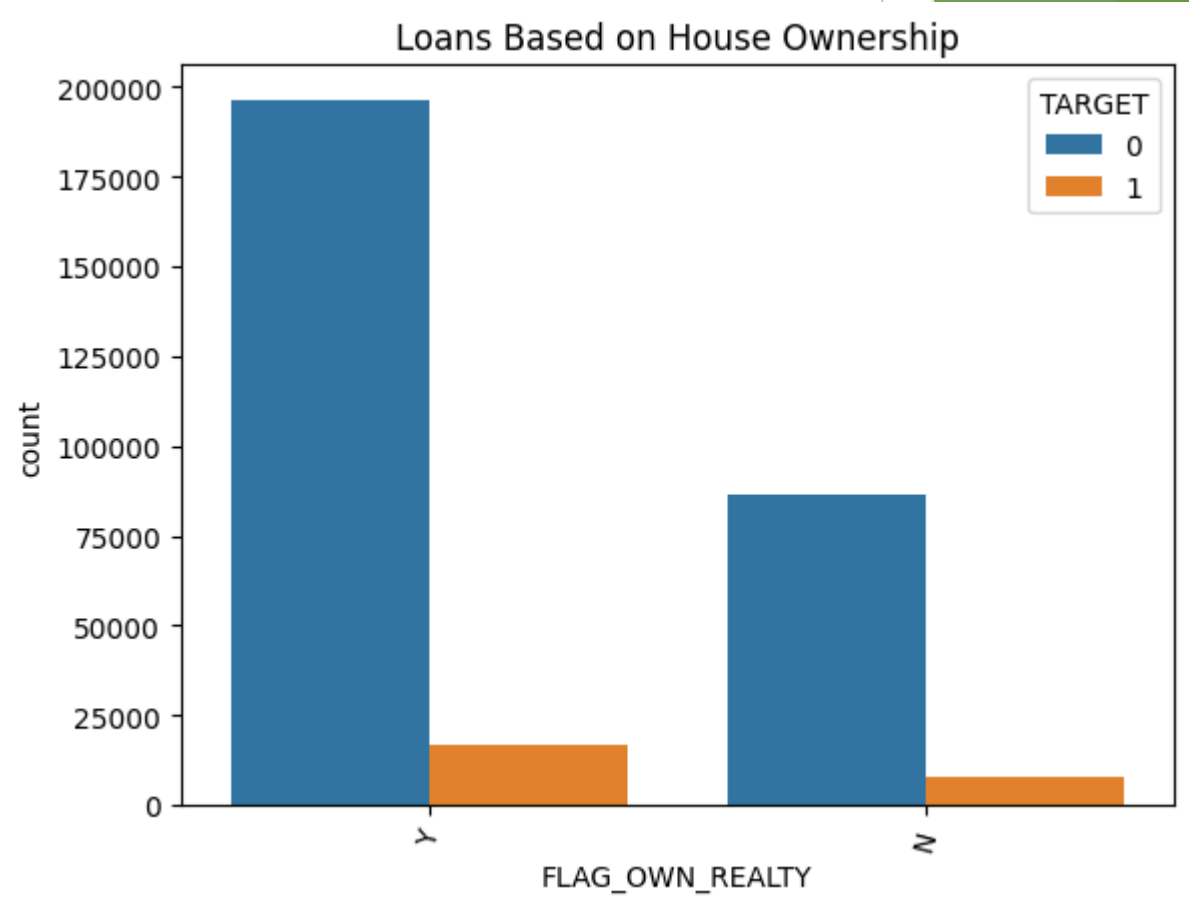
- ▶ Customers whose income type is working take the most number of credits
- ▶ Customers with income type as working have the maximum number of unsuccessful payments
- ▶ Customers who are students or businessman have the least number of unsuccessful payments



# Credits based on ownership of real estate

From this we can observe the following-

- ▶ Customers who own a real estate take more number of loans
- ▶ Customers who own real estate have more successful payments

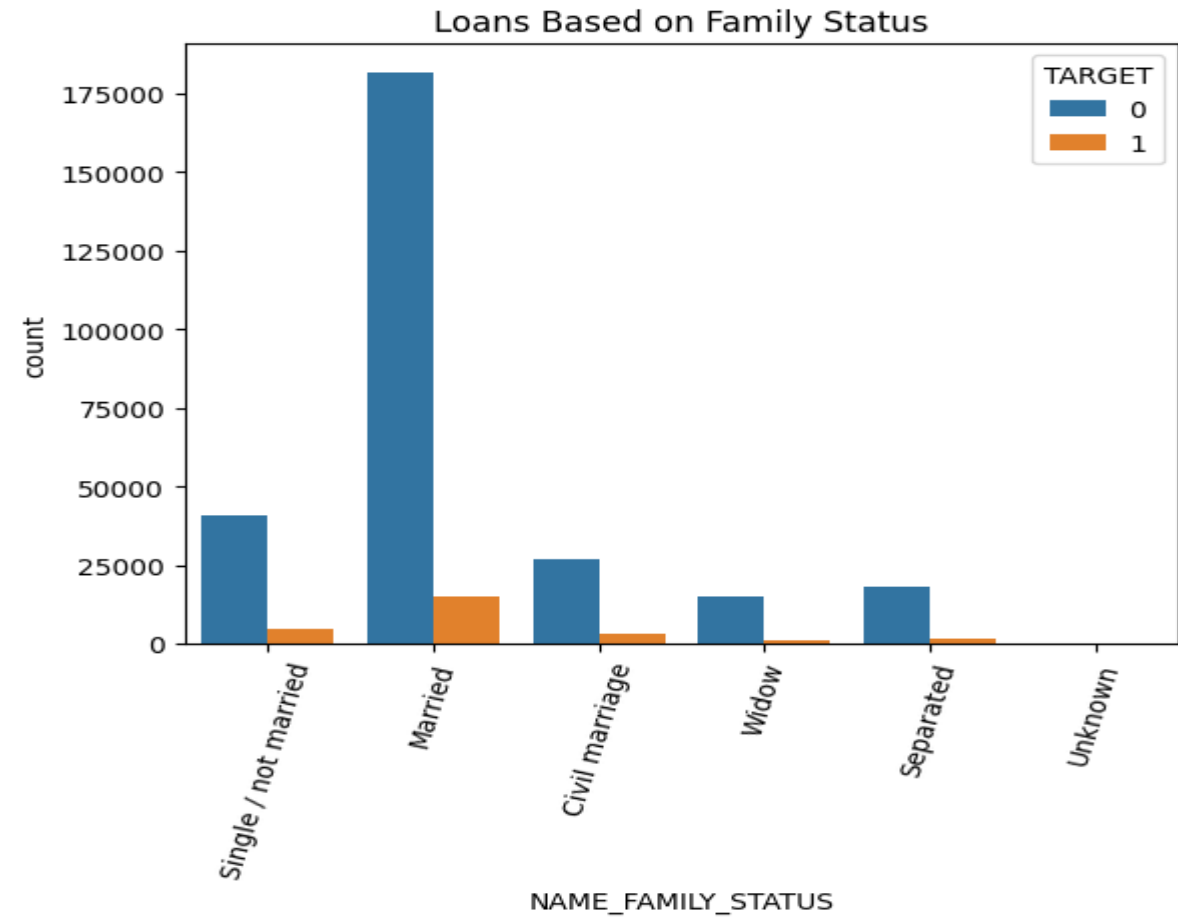




# Credits based on family status

From this we can observe the following-

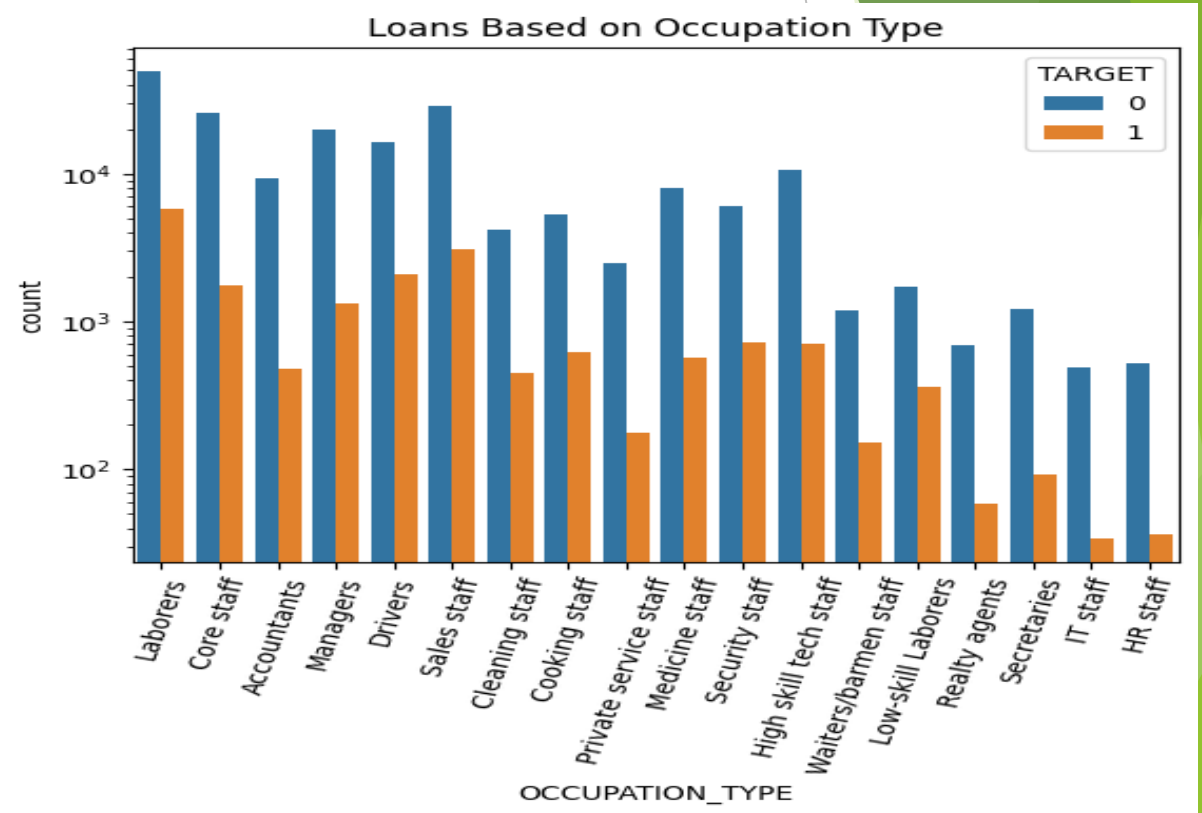
- ▶ Married people tend to take more number of loans
- ▶ Widows and customers who have separated take the least number of loans
- ▶ Married people also have the highest number of unsuccessful payments



# Credits based on Occupation

From this we can observe the following-

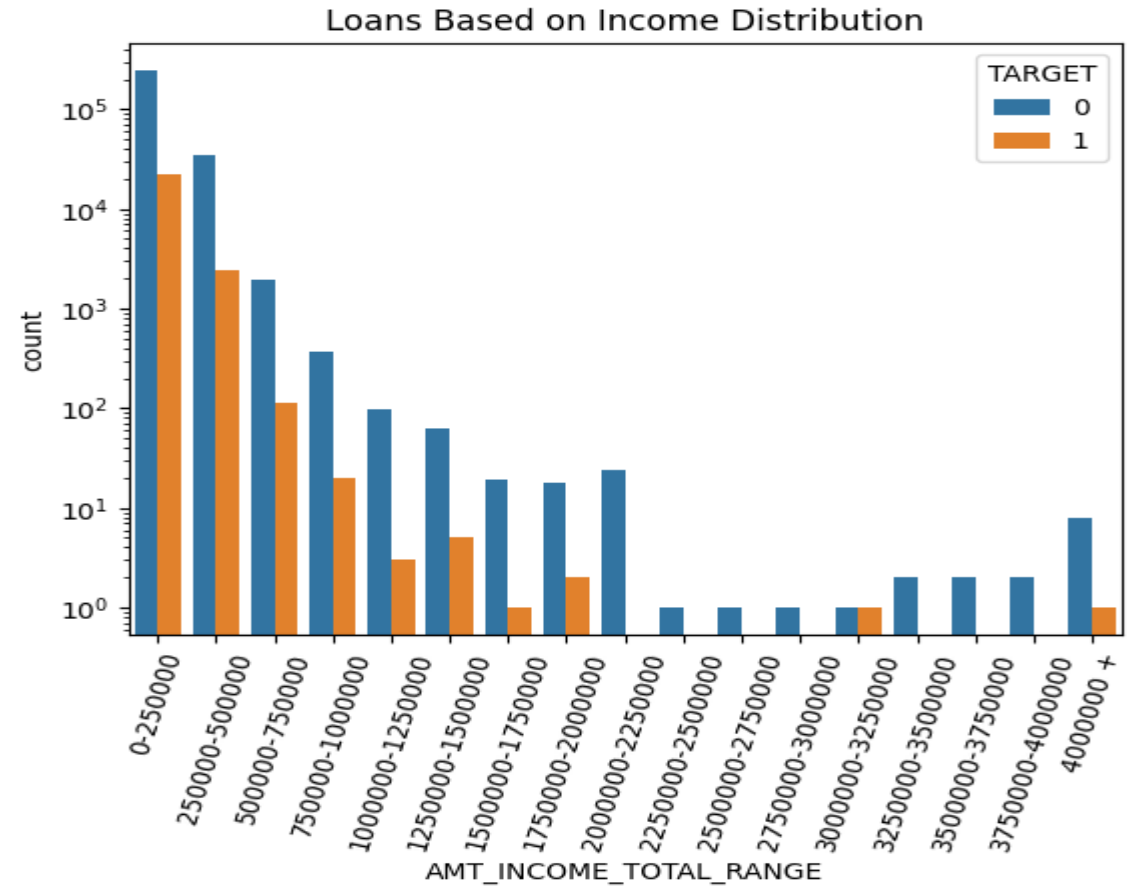
- ▶ Laborers and sales staff take the most number of loans
- ▶ IT staff and HR staff take the least number of loans
- ▶ IT staff and HR staff customers have the least number of unsuccessful payments
- ▶ Laborers and sales staff have the most number of unsuccessful payments



# Credit based on Income Range

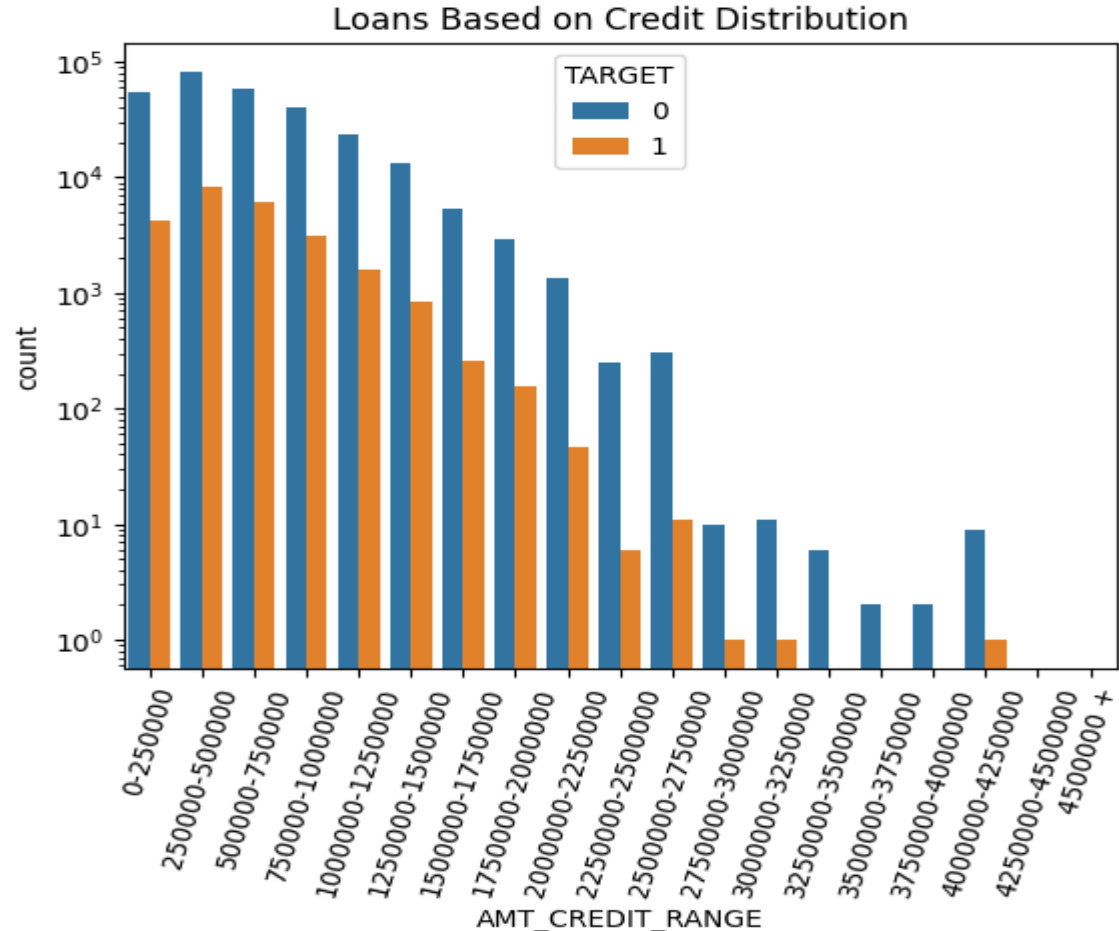
From this we can observe -

- ▶ Customers in income range of 0-500000 take the most number of loans
- ▶ Customers with income range of 2000000 - 2750000 and 3000000 - 4000000 usually repay their loans successfully
- ▶ Customers with income range of income range of 0-250000 have the most unsuccessful loan payments



# Credits based on the credit amount

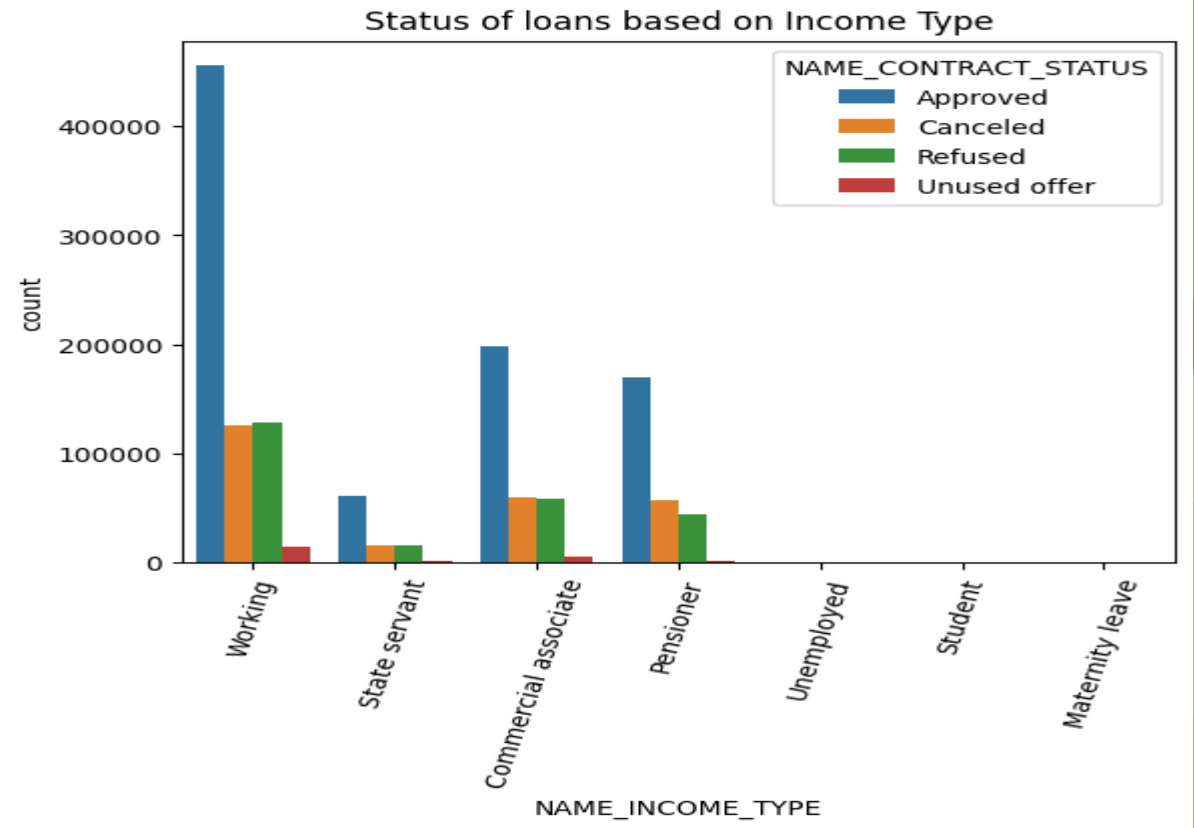
- ▶ Customers whose credit range 3000000 - 3750000 usually don't have difficulties in paying back their loans
- ▶ Customers whose credit is in the range 250000 - 500000 take the maximum number of loans and have the highest unsuccessful loan payments



# Application Status based on Income Type

From this we can observe that -

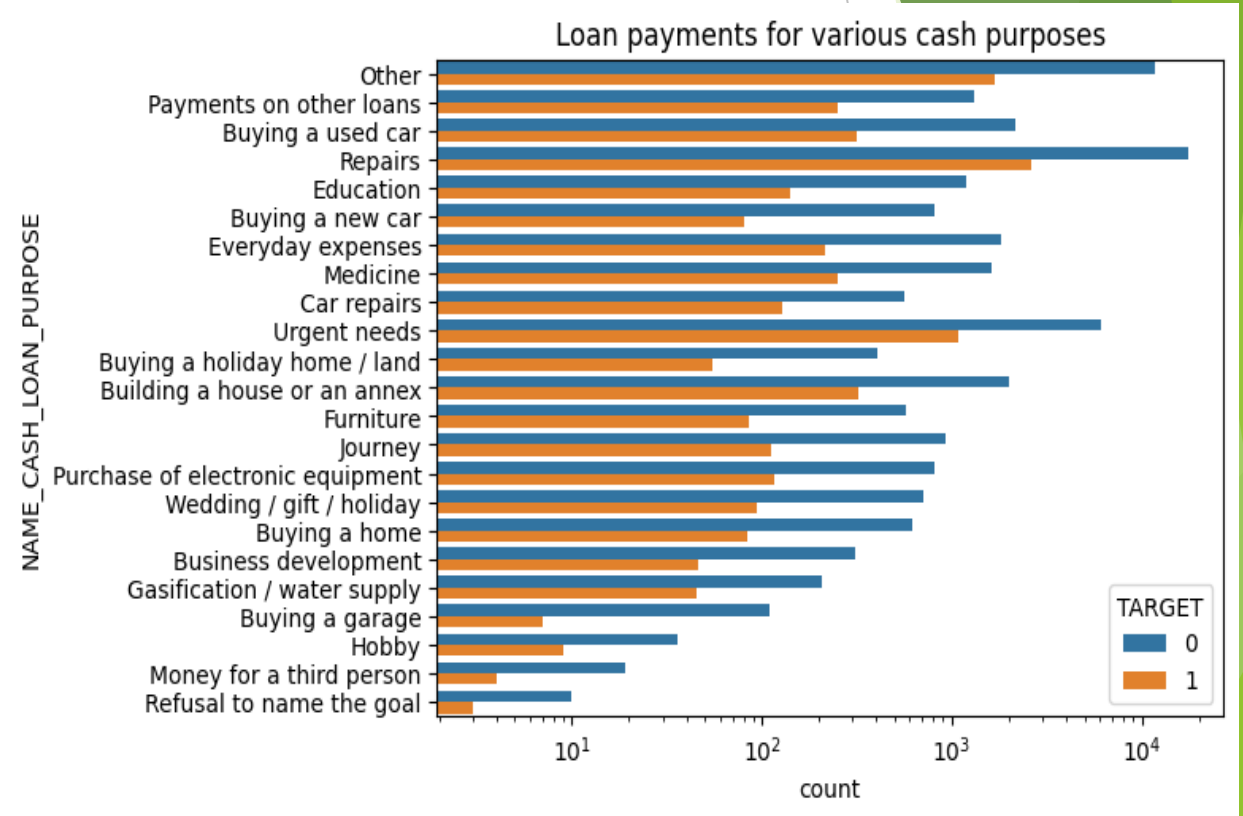
- ▶ Working professionals have the highest number of approved loans followed by commercial associates and pensioners
- ▶ Customers who are either unemployed, students or on maternity leave have very less number of loans



# Credits for various purposes

From above we can observe that-

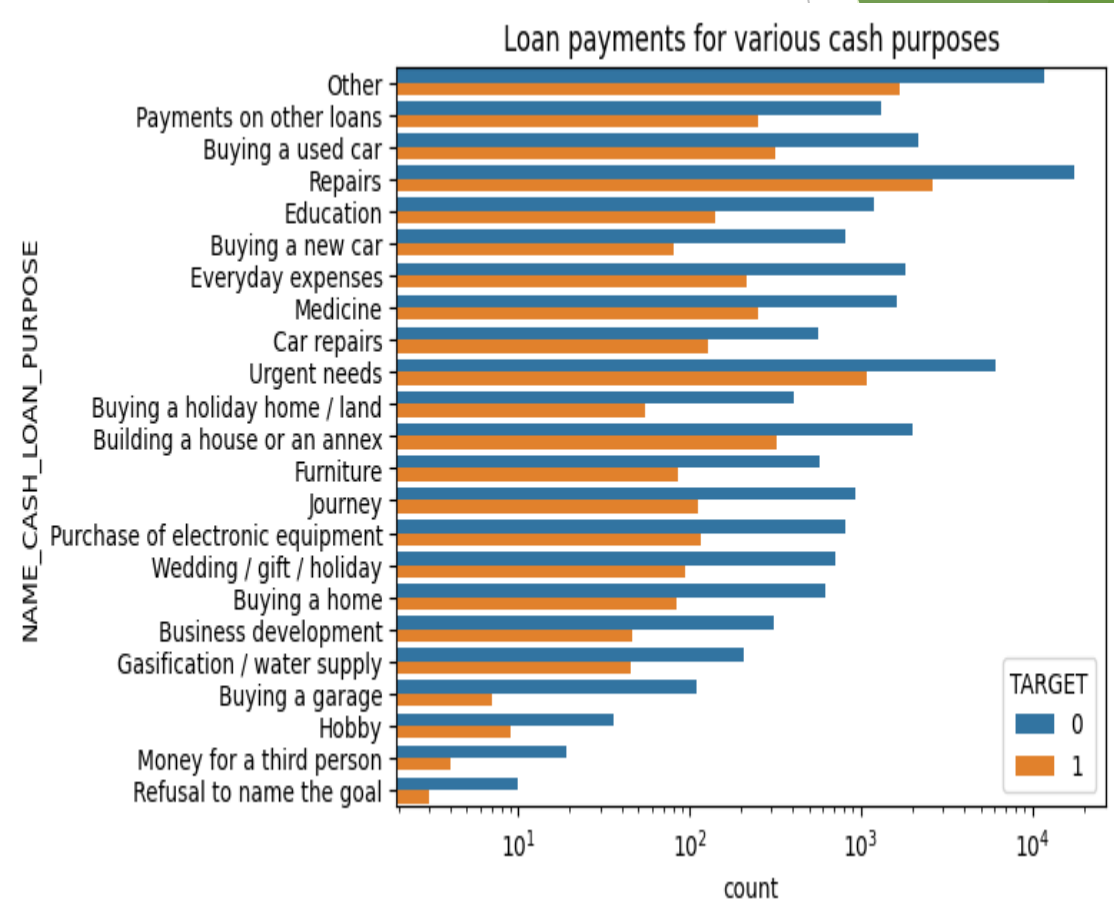
- ▶ Maximum number of loans are being taken for the purpose of repairs
- ▶ Repairs also have the highest number of defaulters as compared to others
- ▶ Customers who have taken a loan for the purpose of buying a garage usually don't face any difficulty in paying back the loan



# Credit for various cash purposes

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- ▶ Maximum number of loans are being taken for the purpose of repairs
- ▶ Repairs also have the highest number of defaulters as compared to others
- ▶ Customers who have taken a loan for the purpose of buying a garage usually don't face any difficulty in paying back the loan



# Bivariant Analysis

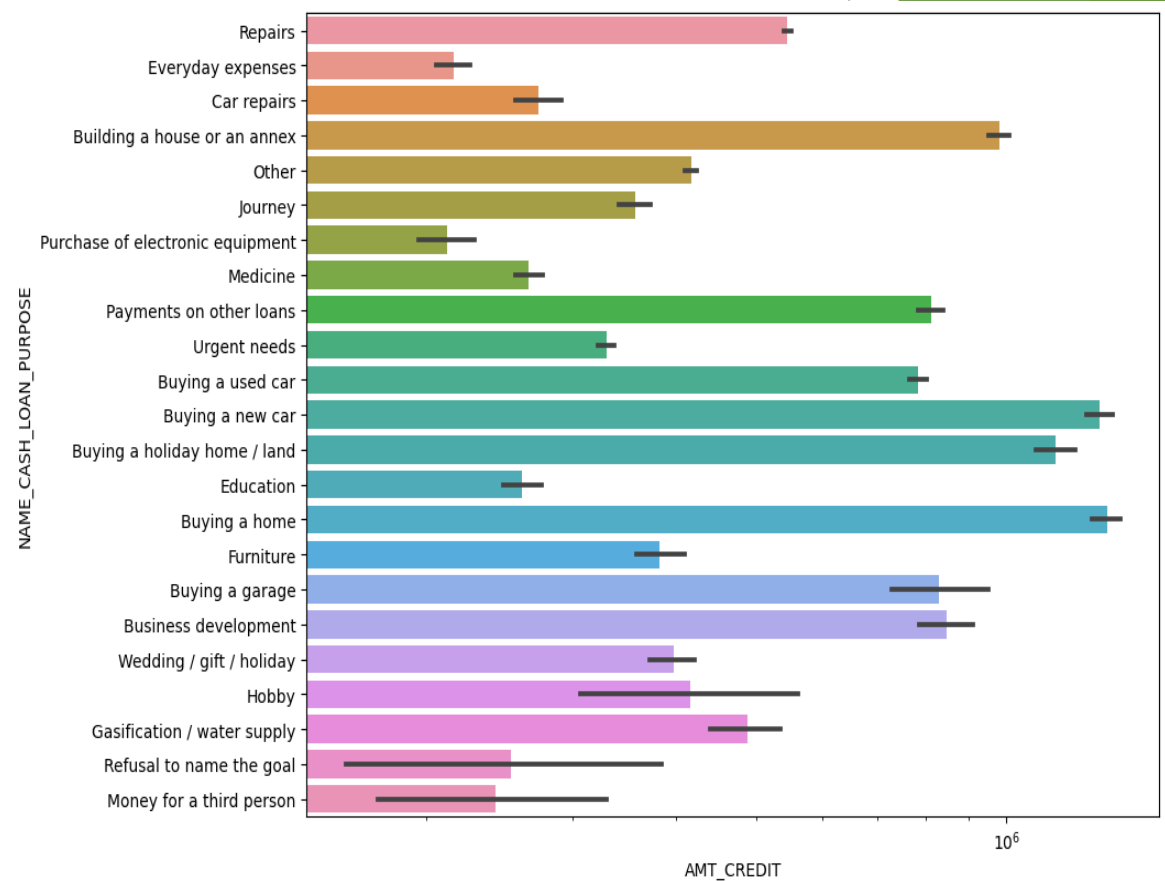
The background of the slide features abstract, overlapping geometric shapes in various shades of green, ranging from light lime to dark forest green. These shapes are primarily located on the right side and bottom of the frame, creating a modern, layered effect. The main area of the slide is a plain, light gray.



# Purpose of loan VS credit amount

From above we can observe the following-

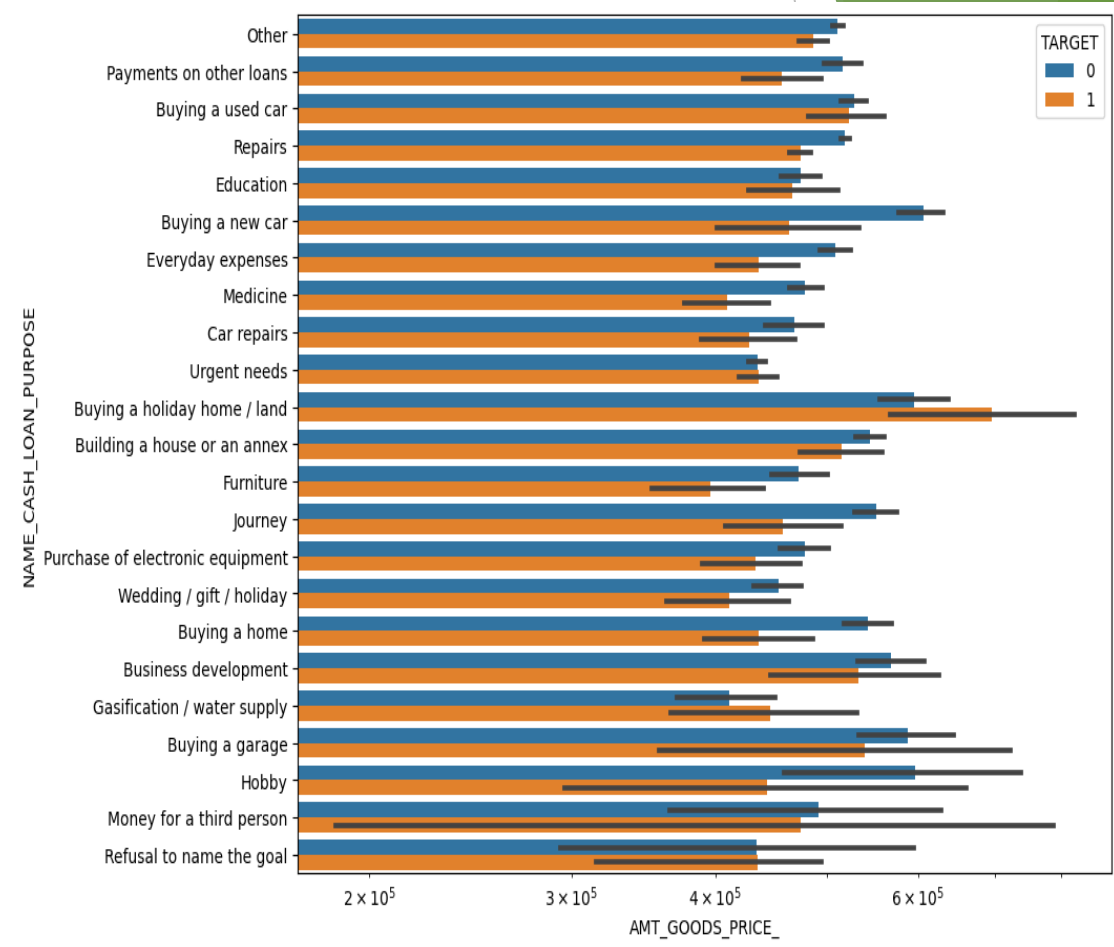
- ▶ Least credit was taken for everyday expenses and purchase of electronic equipment
- ▶ The highest amount of credit was taken for buying a new home and car



# Loan purpose VS Goods Price

From this we can observe the following-

- ▶ Customers who took a loan for buying a new car had the most number of successful loan payments
- ▶ Customers who took loan for buying a holiday home/land had the most number of unsuccessful loan payments



# Conclusion-

- Banks should avoid giving loans for the purpose of buying a holiday home/land as they have high unsuccessful rate
- Banks should give more loans to students and businessman as they have the least unsuccessful payments
- Banks should give more loans for buying house and cars as the credit amount taken for them is high and also the successful payment percentage is decent