

# EDA Case Study

Bank Loan Application Data

# Analysis of Application data file

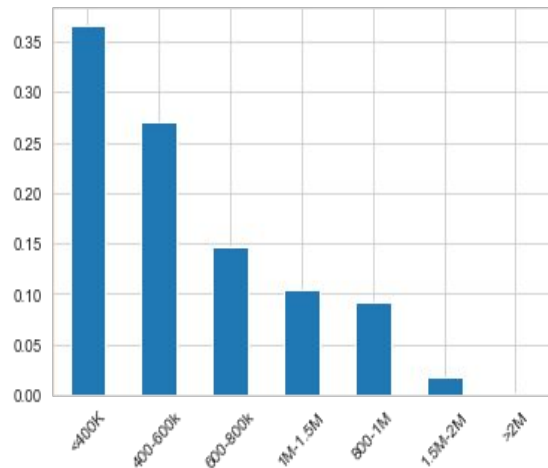
- Load Application data file and check the shape of the file
- Check the column length
- Print the data of the file and understand the used and unused data.

# Data Cleaning

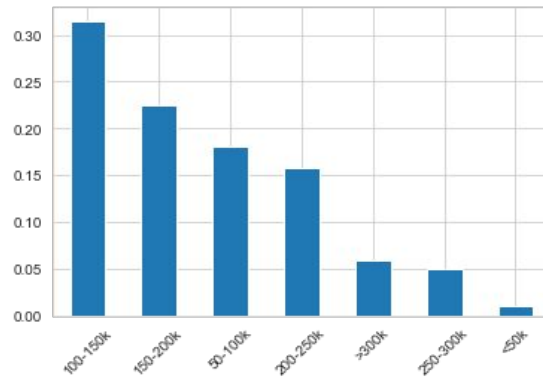
- Remove all the column which are null more than 50%
- Also remove all the unused column which are not required in analysis.
- Treat the null values with column 'mode' value
- Calculate the null % of each columns
- Change the Data type of the column which are wrong datatype.
- Rename the column name and correct the data from negative to positive.
- Create categorical data of continuous columns to analysis in better ways.
- Calculate the defaulters and non-defaulter %.
- Create two data frame based on 'TARGET' column.

# Default(Target=1) analysis and Visualize Univariate and bivariate

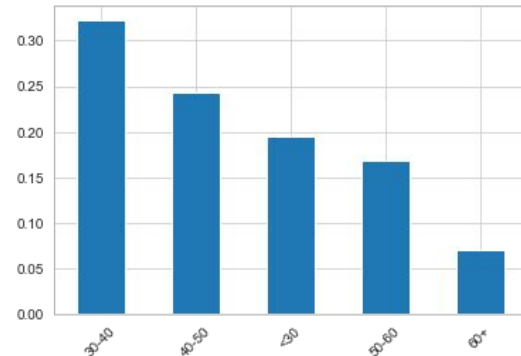
Credit amount Group



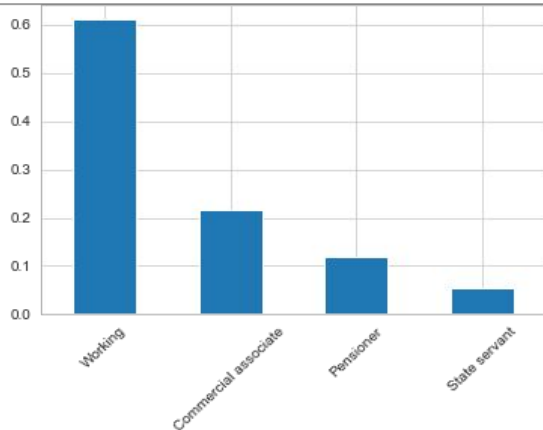
Total Income Group



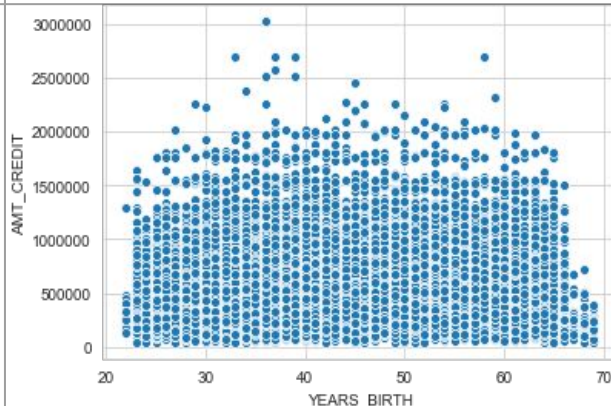
Age Group



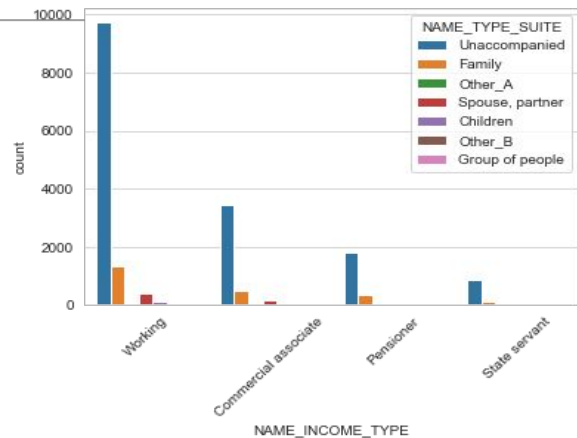
### Income Type



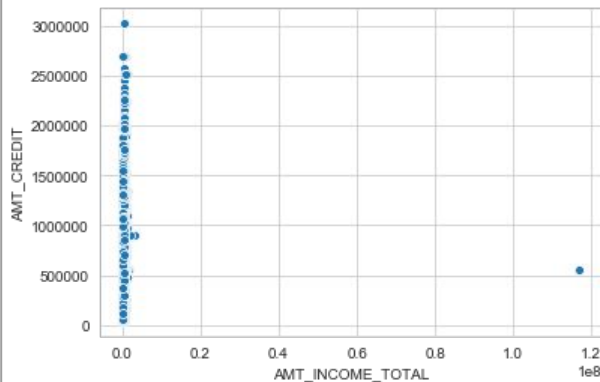
### Age and Credit amount



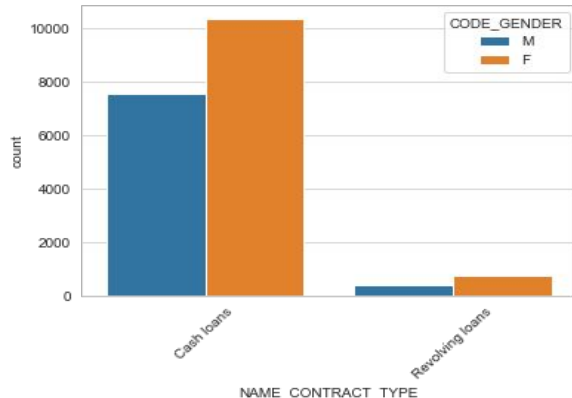
### Income Type & Suite



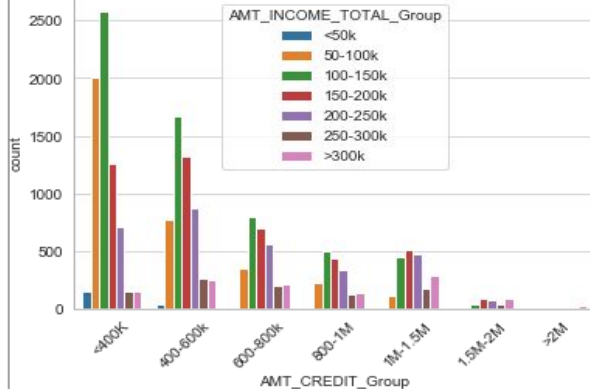
### Total Income & Credit



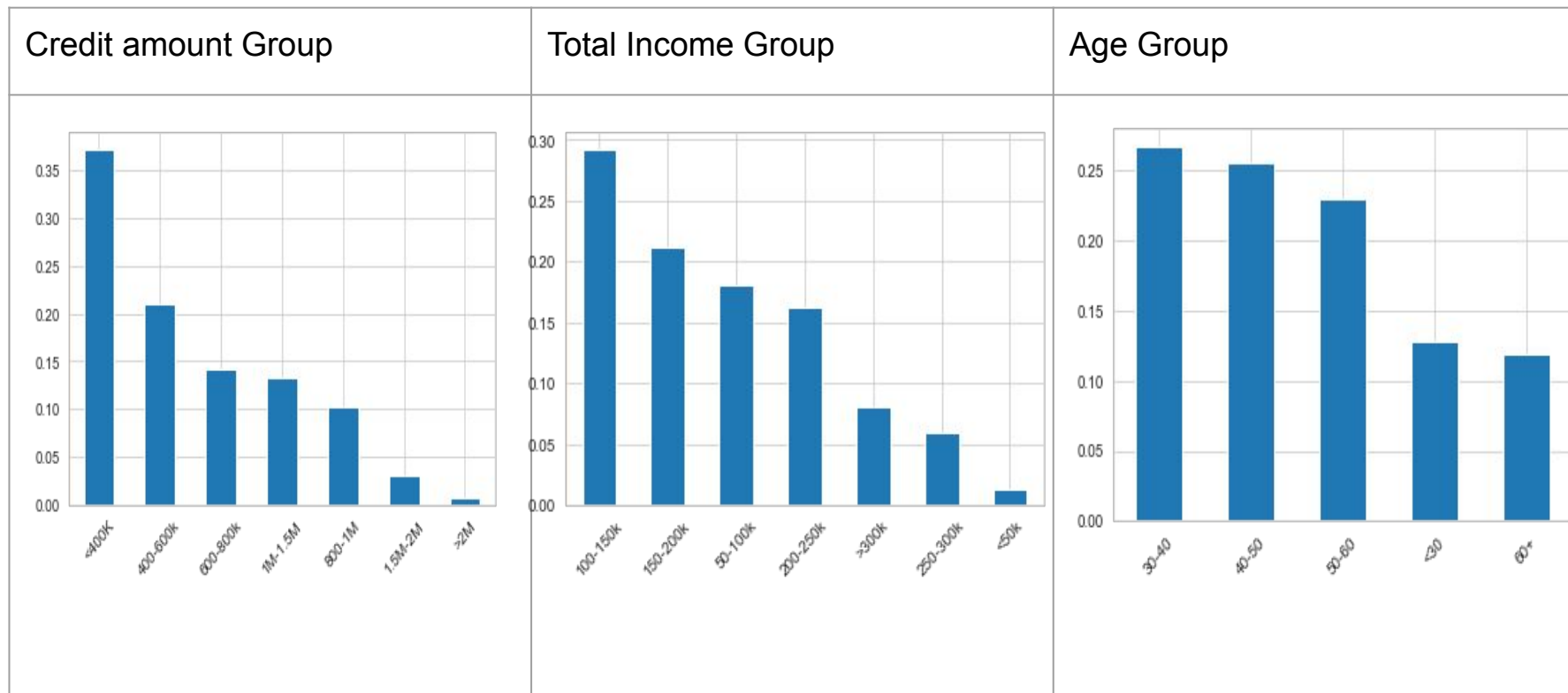
### Gender & Loan More Defaulter



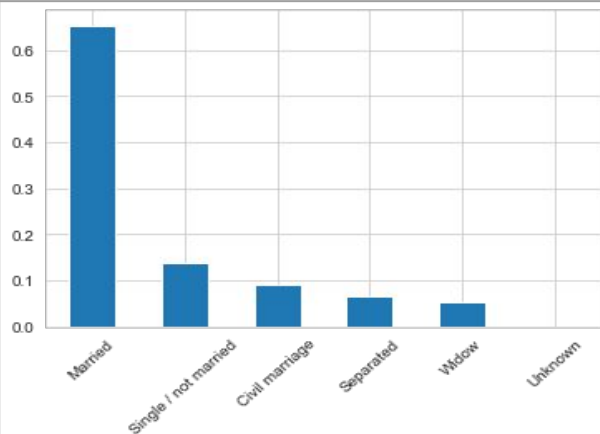
### Credit & Income Groupwise



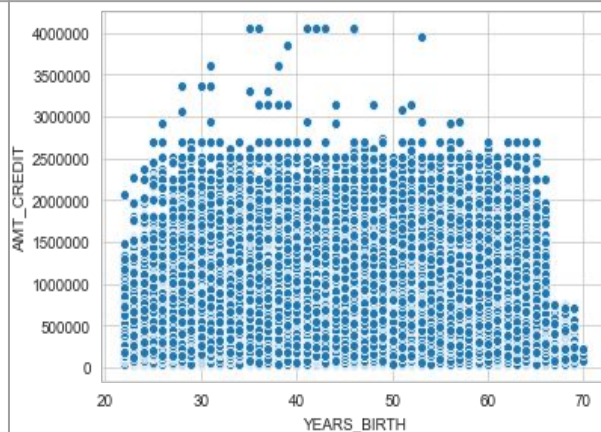
# Non Default(Target=0) analysis & Visualize Univariate and bivariate



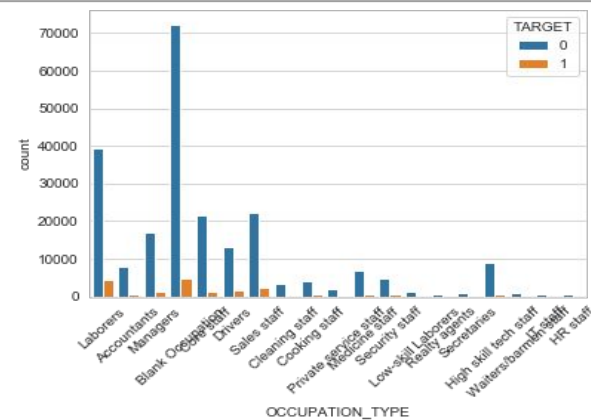
### Non default Based Family Type



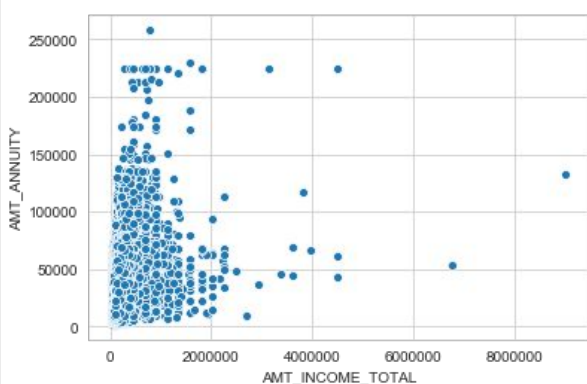
### Age and Credit amount



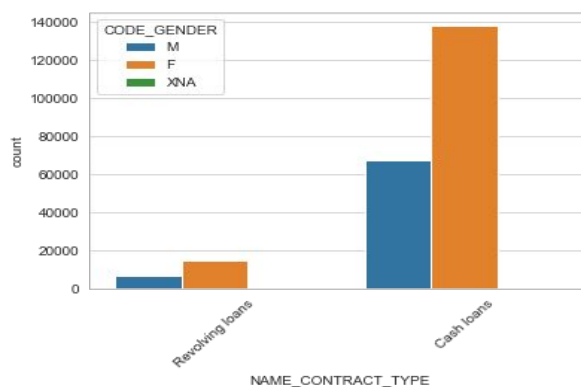
### Occupation and Target Column



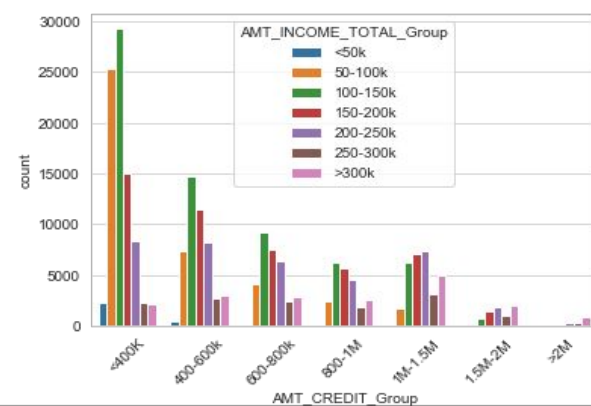
### Total Income & Annuity



### Gender & Loan More Defaulter



### Credit & Income Groupwise



# Merge Application and Previous Application

- Merge Both file , create df3 dataframe from both
- Drop Column who have more than 40 % missing value
- Calculate mean of missing Value
- Droup more column that look like not useful for this analysis

## **Data Cleaning:**

- CODE\_GENDER fillNa value with 'F' because its having percentage
- FLAG\_OWN\_REALTY fillNa value with 'Y' because its having percentage
- AMT\_ANNUITY\_x and AMT\_ANNUITY\_y updated with mean value (Both mean and median have almost same value(minor diff))
- CNT\_FAM\_MEMBERS 'Na' updated with median (Because mean and median almost same.

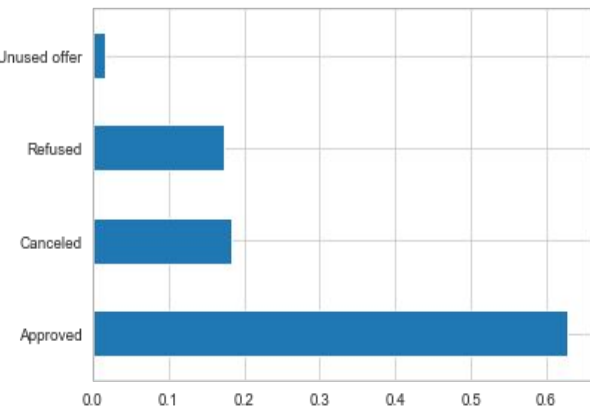


- NAME\_CONTRACT\_STATUS filter NaN value because very less row having NaN

## **Univariant**

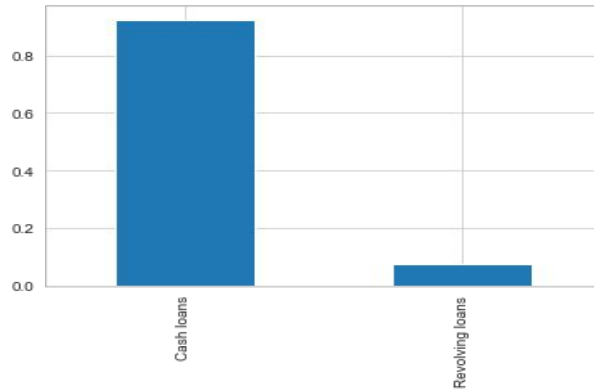
- Converted DAYS\_BIRTH into Years
- Perform Univaraint on NAME\_CONTRACT\_STATUS
- Perform on NAME\_CONTRACT\_TYPE\_x
- Perform on Code Gender
- Perform on CODE\_GENDER
- Perform on NAME\_INCOME\_TYPE
- Perform on NAME\_EDUCATION\_TYPE
- Created Age group for DAYS\_BIRTH
-

NAME\_CONTRACT\_STATUS

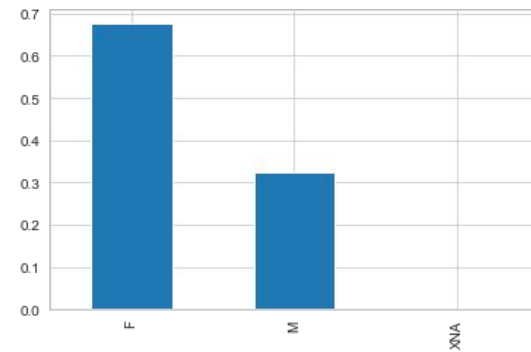


NAME\_CONTRACT\_TYPE\_

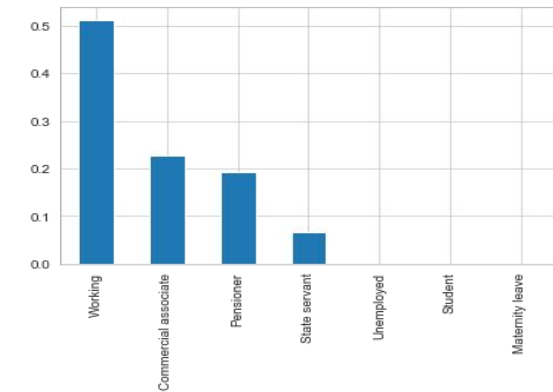
X



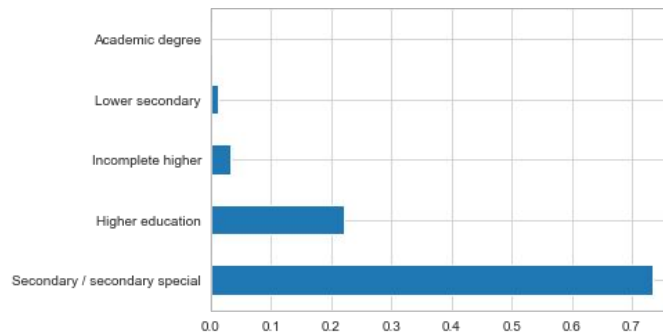
CODE\_GENDER



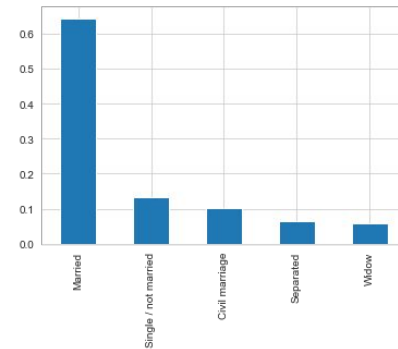
NAME\_INCOME\_TYPE



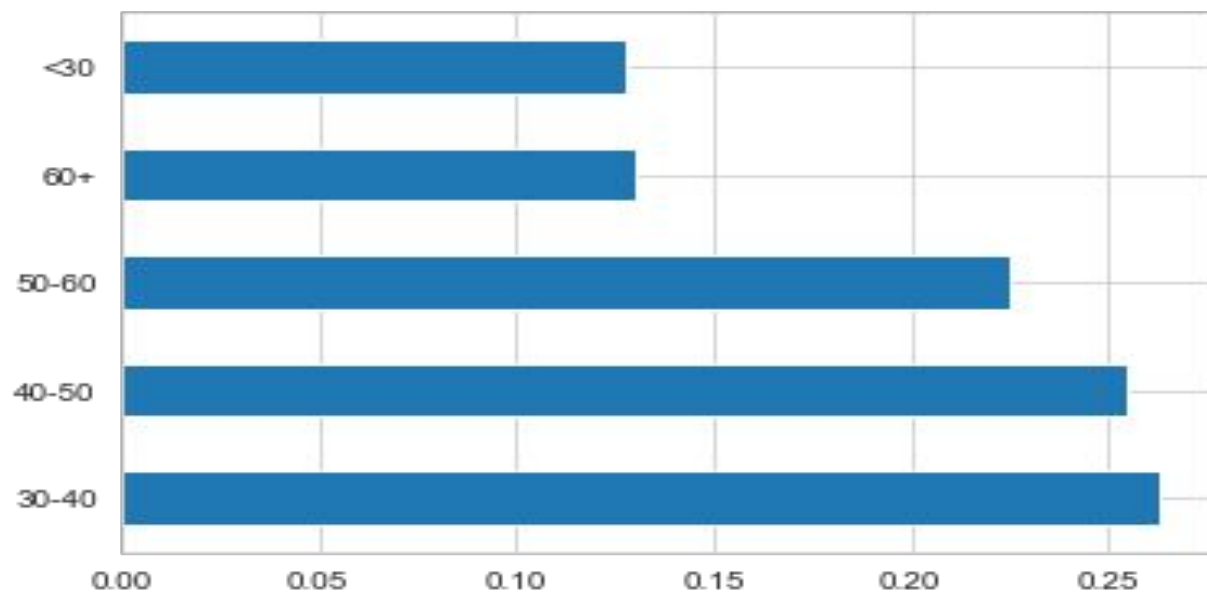
NAME\_EDUCATION\_TYPE



NAME\_FAMILY\_STATUS



YEARS\_BIRTH

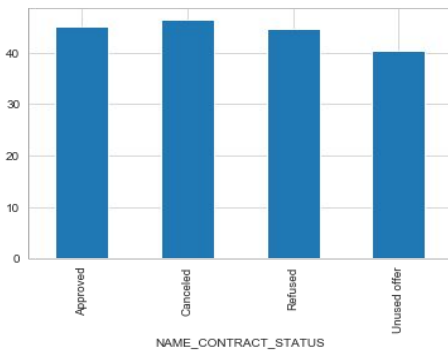


# Bivariate(With NAME\_CONTRACT\_STATUS)

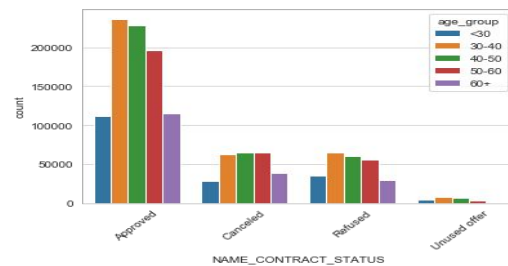
Perform Bivariate analysis below Attributes

- YEARS\_BIRTH
- Age\_group
- NAME\_CONTRACT\_TYPE\_x
- CODE\_GENDER
- AMT\_INCOME\_TOTAL
- AMT\_ANNUITY\_y
- NAME\_INCOME\_TYPE
- NAME\_FAMILY\_STATUS
- CNT\_PAYMENT
- AMT\_APPLICATION
-

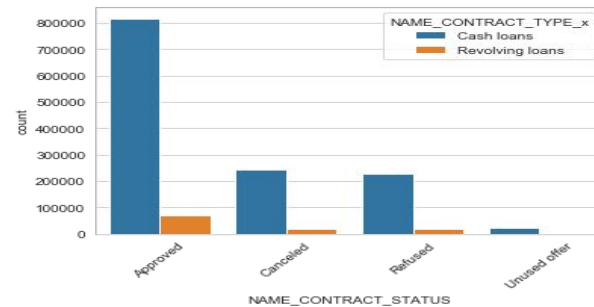
YEARS\_BIRTH



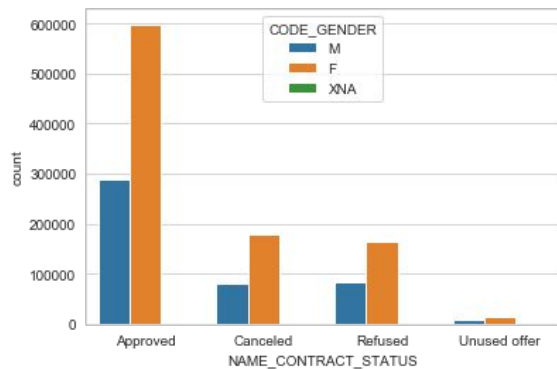
age\_group



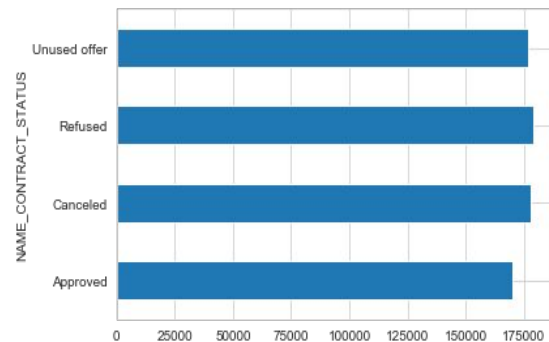
NAME\_CONTRACT\_TYPE\_x



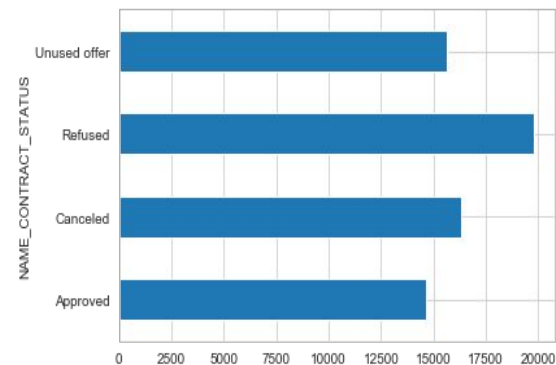
CODE\_GENDER



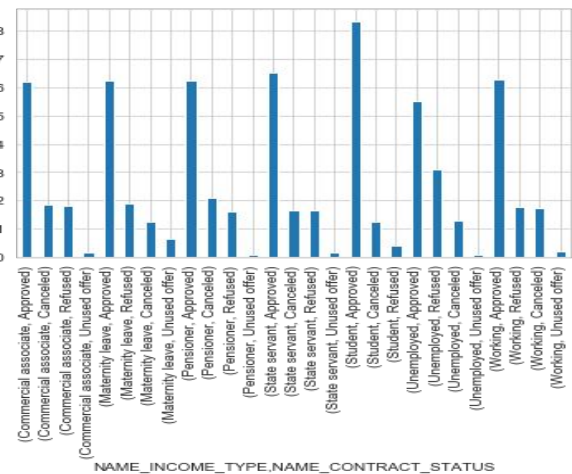
AMT\_INCOME\_TOTAL



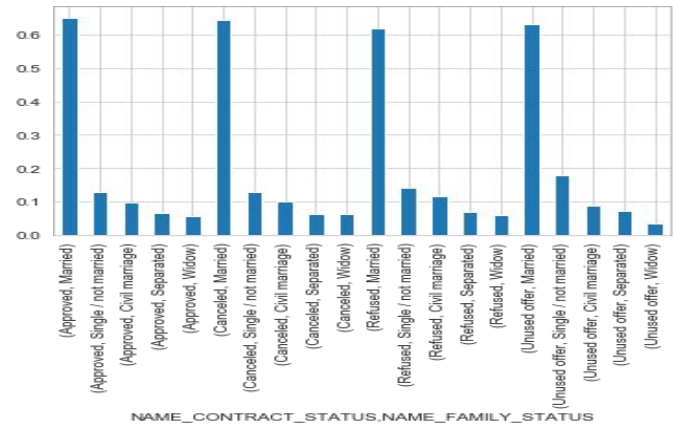
AMT\_ANNUITY\_y



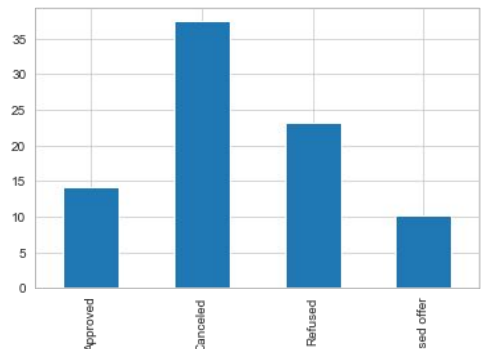
NAME\_INCOME\_TYPE



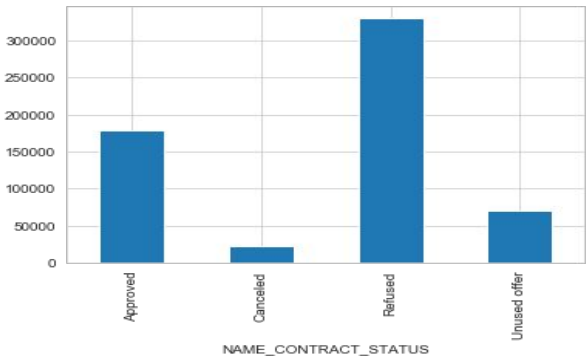
NAME\_FAMILY\_STATUS



CNT\_PAYMENT



AMT\_APPLICATION



# Conclusion Of Analysis

- We try to show many reason of rejection or approval of loan
- Try to find Outliers form Univariant and bivariate
- Perform Comparison for Target with many Column in application to see the reason for any action(Rejection or Approval)
- Merge the both application and perform Univariate to see who applied the loan (Category)
- Perform bivariate on merge dataframe from 'NAME\_CONTRACT\_STATUS' with many column to see the reason of approval ,Canceled, Refused