



# **OBJECTIVE**

To perform Exploratory data analysis, on the provided bank customer data, to come up with the insights on the scenarios when a customer would default the loan.

# STEPS FOLLOWED FOR EDA

# DATA UNDERSTANDING

Going through the given info from data dictionary

#### **DATA CLEANING**

Performing steps to create a better dataset

# UNIVARIATE ANALYSIS

Analysis based on a single variable

#### BIVARIATE ANALYSIS

Analysis based on comparison with double variable

#### MULTIVARIATE ANALYSIS

Exploring relationships between the variables

**INSIGHTS** 

### **STEP I: DATA SOURCING**

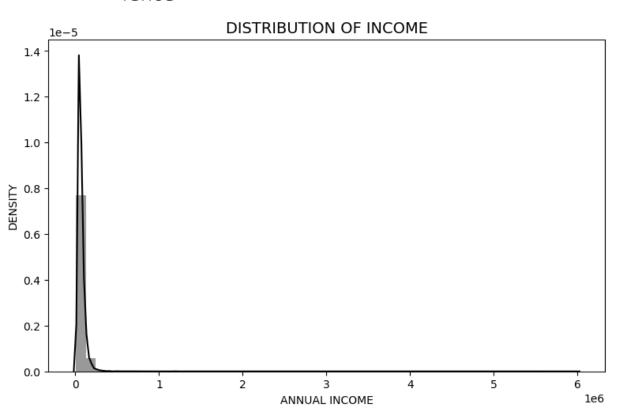
- We are provided with the dataset by UPGRAD, loan.csv dataset and a data dictionary to go through and understand the column details from it.
- Totally there are 39717 rows, and 111 columns
- We can briefly see and understand that there are many columns with null values only, and few columns that require to be filled with missing values.

#### **STEP 2: DATA CLEANING**

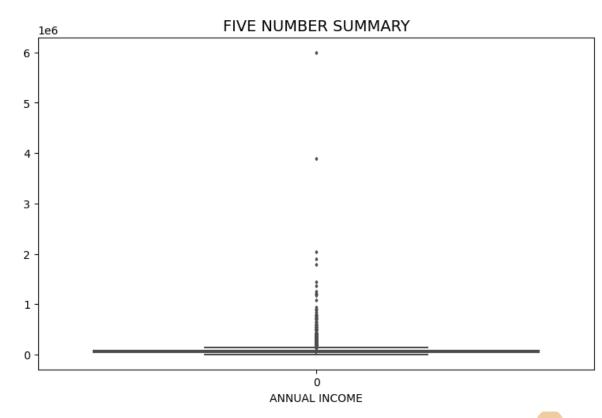
- 1. We have removed all the columns containing the null values.
- 2. We have then found that there are columns containing only single values, so we removed it as it would not help in the analysis.
- 3. Then we calculated the percentage of null values present in each column of the dataset and removed the columns having null value percentage above 60%.
- 4. We dropped the columns the irrelevant columns that wouldn't provide us much help with our objective.
- 5. Dropped the rows containing loan status as "current", as our goal is to predict if in future the customer would default or not.
- 6. We have handled the missing values using median for numerical columns and mode for categorical columns
- 7. We have standardized the columns like int\_rate, revol\_util by removing the % symbol and converting it to a numerical value.
- 8. Finally we have removed the outliers by visualizing it using box plots and removing it using the percentile method.

We can see a high range of customers in the following scenarios:

 the feature "annual\_inc" has outliers in the upper fence

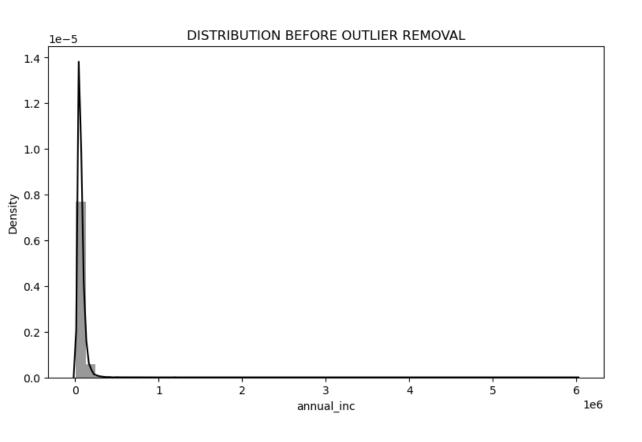


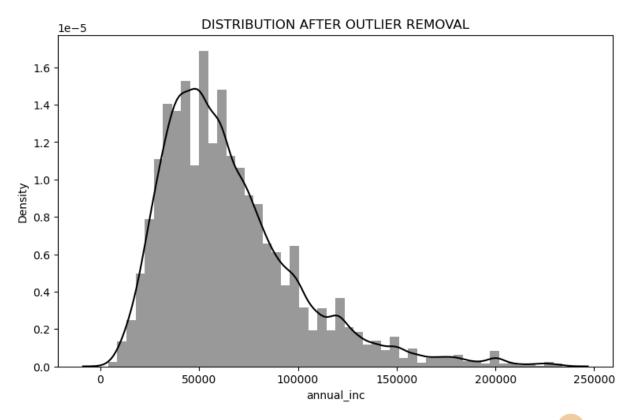
 There is a huge jump between 99 and 100 % of data in "annual\_inc" and that can be trimmed.



We can see a high range of customers in the following scenarios:

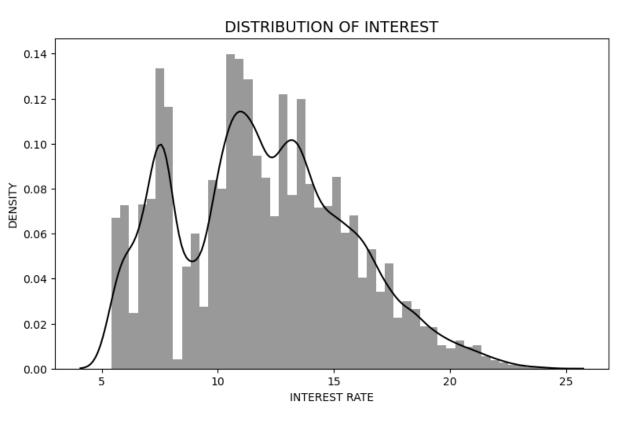
• After outlier removal, we can see that the majority of borrower's income lies between 40k-80k range

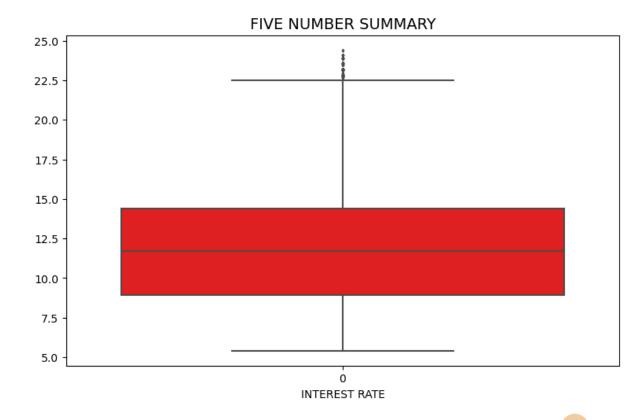




#### We can see a high range of customers in the following scenarios:

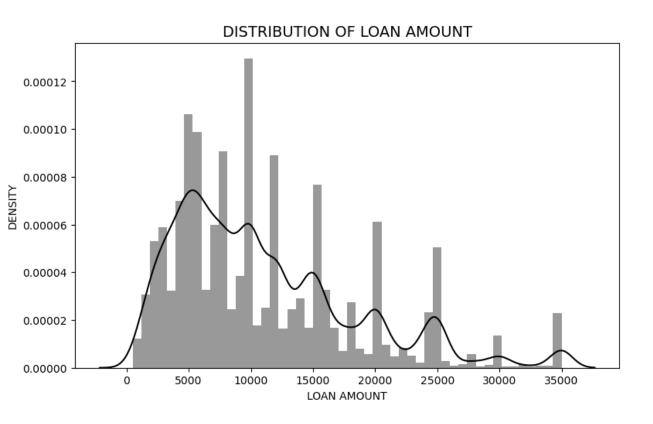
 The average interest rate is 12% but there is huge jump after 75 percentile.

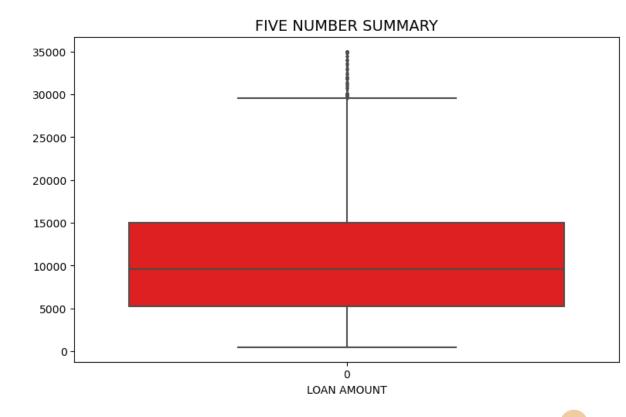




We can see a high range of customers in the following scenarios:

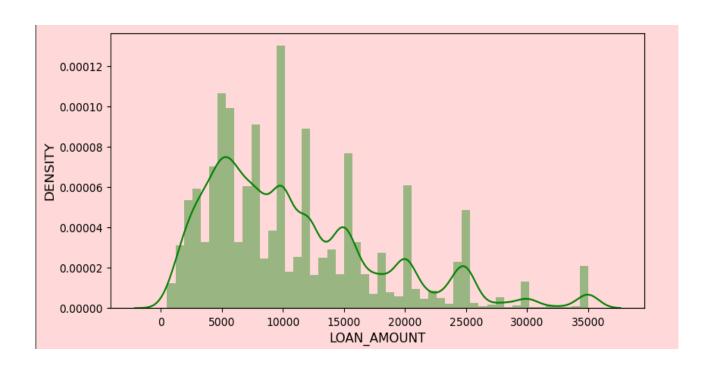
 Most of the borrowers prefer to get loan amount between 5k to 15k.





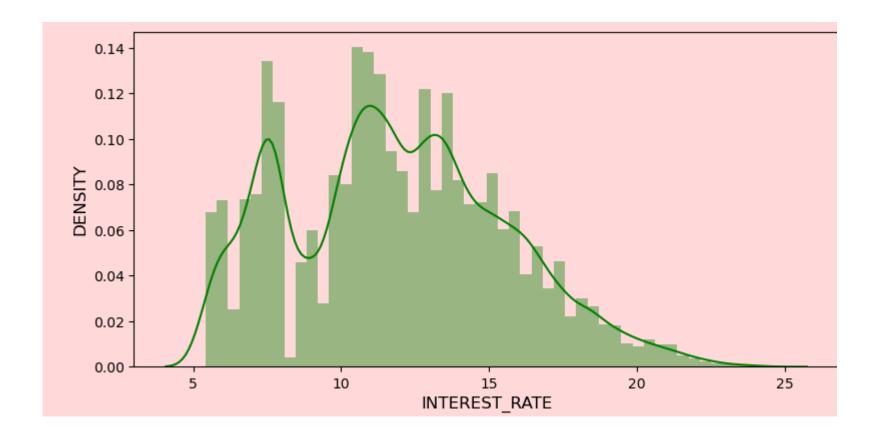
We can see a high range of customers in the following scenarios:

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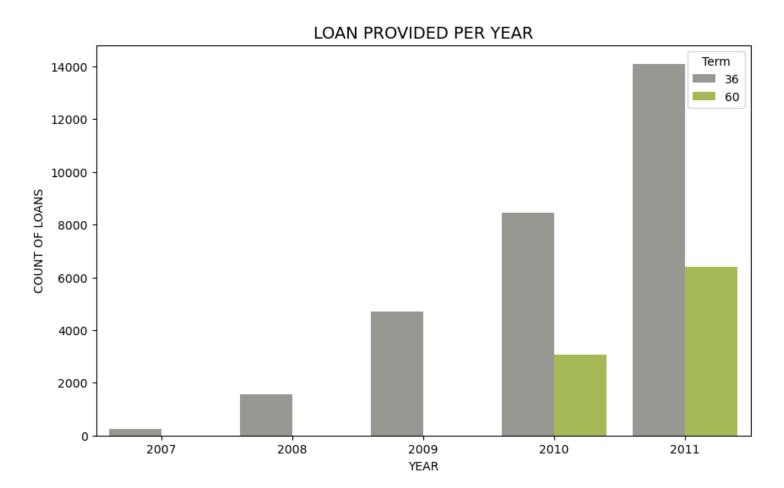
We can see a high range of customers in the following scenarios:

• Almost of the borrowers gets interest rate between 9% to 14%.



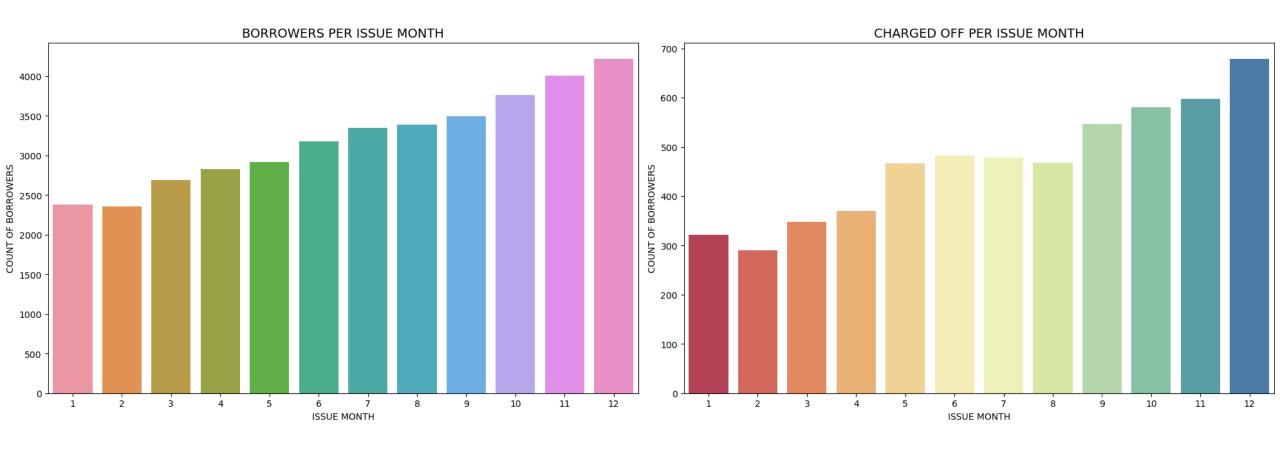
We can see a high range of customers in the following scenarios:

• For each year number of borrowers keeps on increasing and they opt for 36 months tenure.



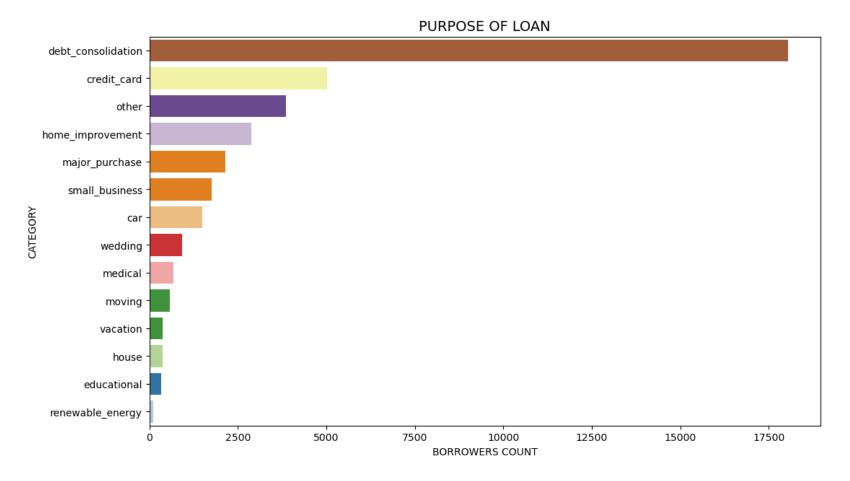
We can see a high range of customers in the following scenarios:

Loans approved on Q4 are more likely to default.



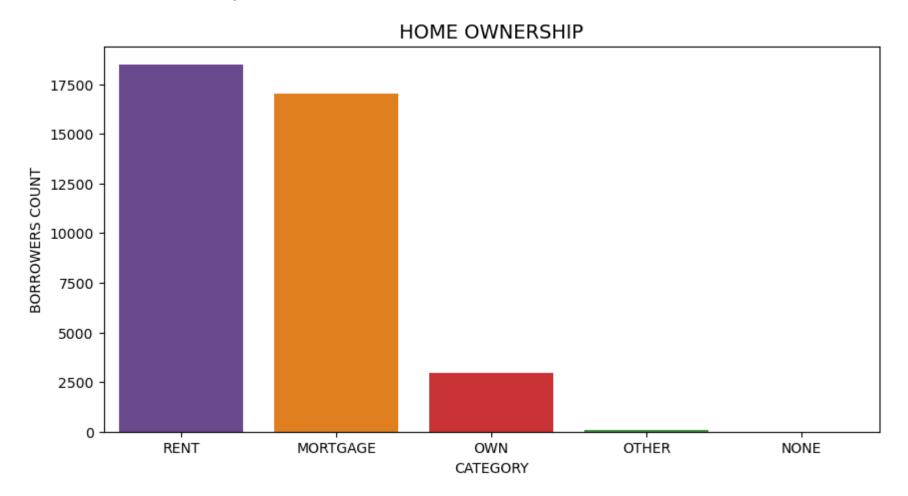
We can see a high range of customers in the following scenarios:

Most of the borrowings happens for debt consolidation and credit card bill payment.



We can see a high range of customers in the following scenarios:

• Very least of the borrowers has a own house.



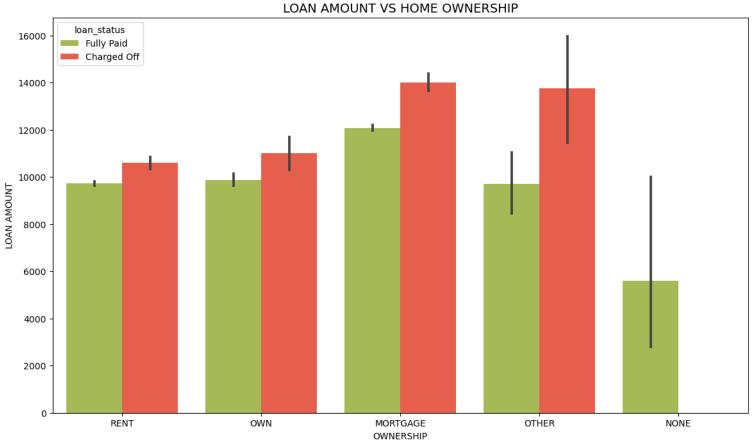
#### EMPLOYMENT LENGTH VS LOAN STATUS

Borrowers with 10+ years of employment length are more likely to charge off.

#### **EMPLOYMENT LENGTH VS LOAN STATUS** 8000 loan\_status **Fully Paid** Charged Off 7000 6000 5000 COUNT 4000 3000 2000 1000 < 1 year 1 year 5 years 2 years 3 years 4 years 6 years 7 years 8 years 9 years 10+ years EMPLOYMENT LENGTH

#### HOME OWNERSHIP VS LOAN AMOUNT

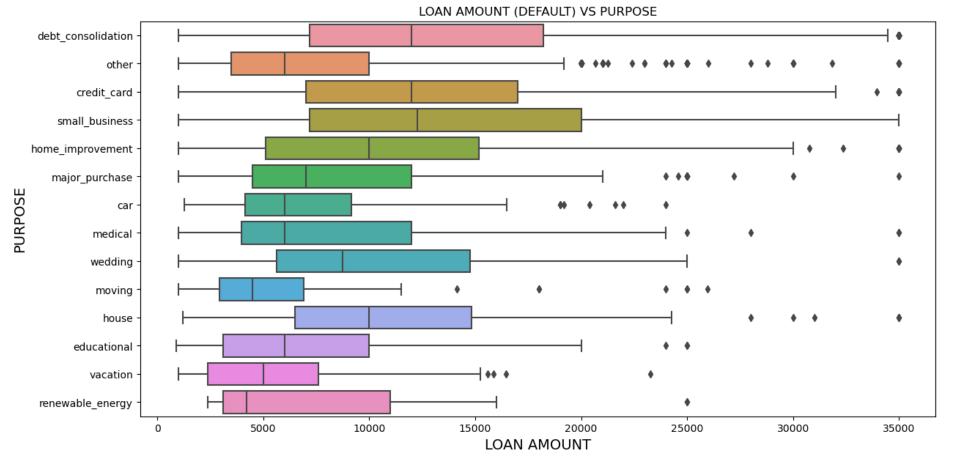
• Borrowers living on mortgage with loan amount greater than 12k are likely to charge off.





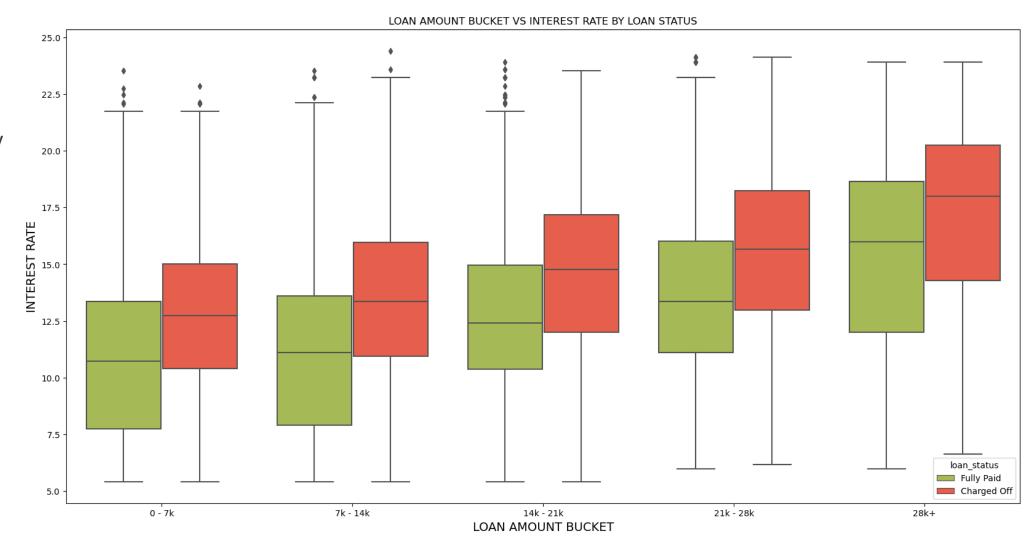
#### LOAN AMOUNT VS PURPOSE

 Loan for debt consolidation, credit card, small business with average loan amount greater than 12000 are more likely to default.



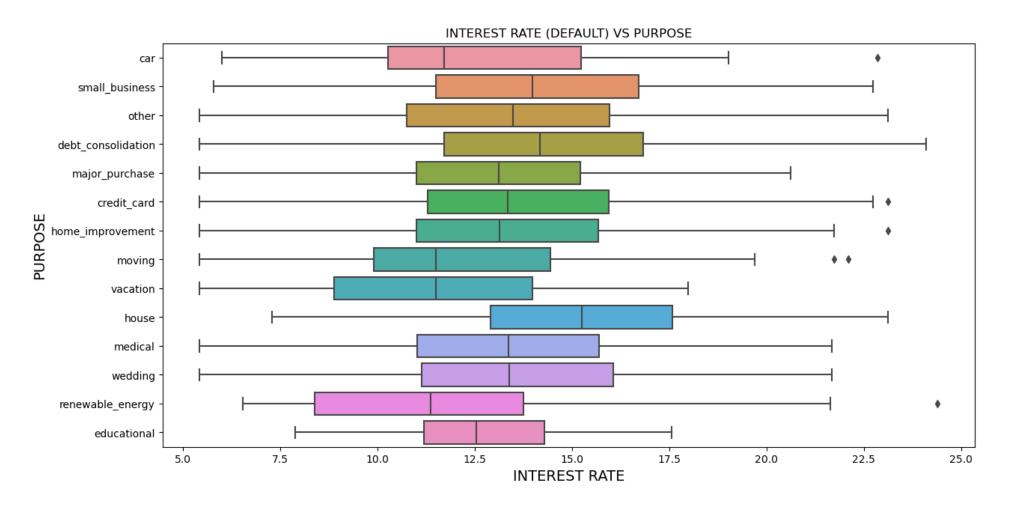
#### loan amount bucket vs interest rate

 Loan with highest interest rate in all buckets are more likely to charge off.



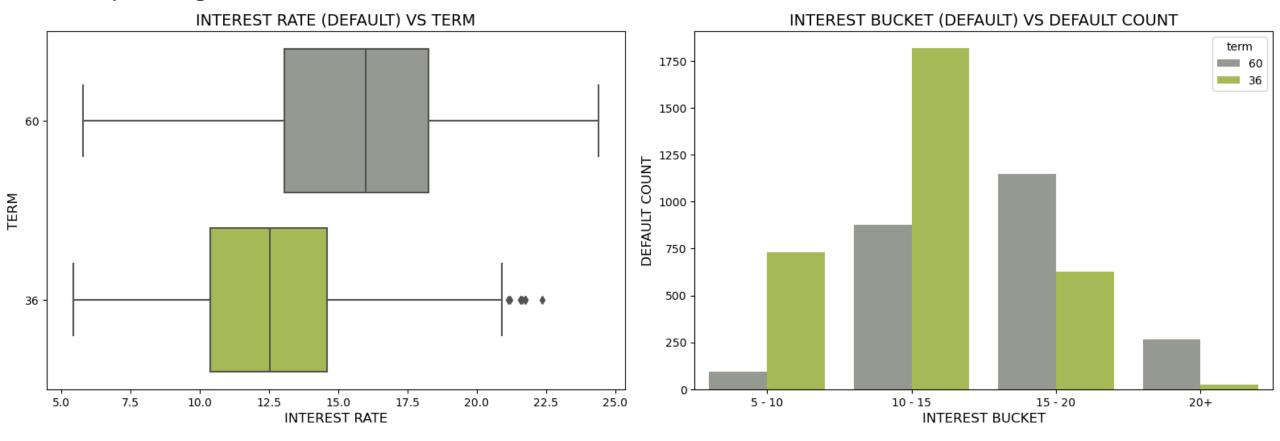
#### interest rate vs purpose:

Loan provided for house, debt consolidation, small business with highest interest rate are more likely to charge off.



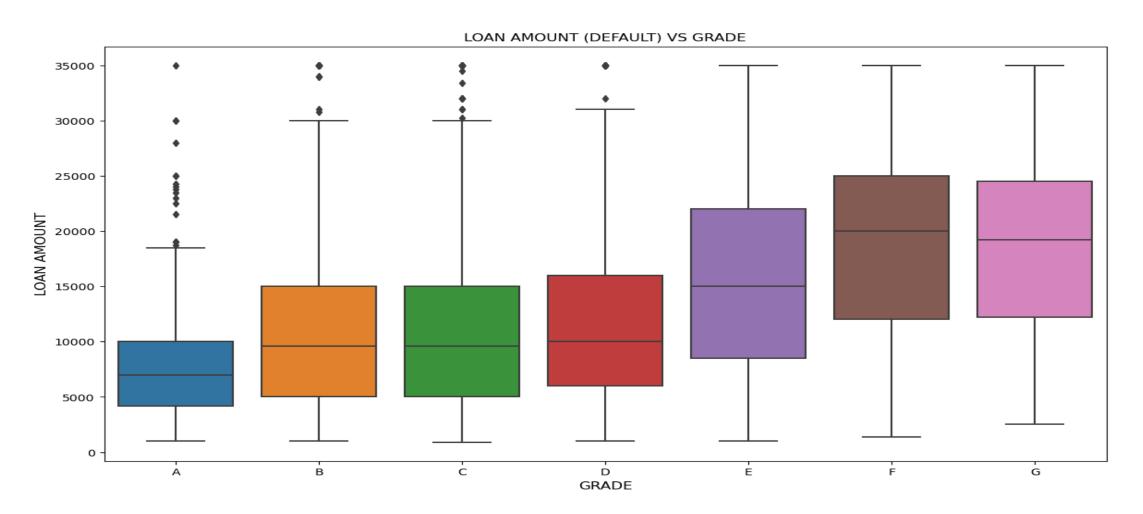
#### **INTEREST RATE VS TERM:**

• Longer tenure has the higher interest rate and people in this term with interest rate between 15-20% is more likely to charge off. Shorter tenure has the lower interest rate and people in this term with interest rate between 10-15% is more likely to charge off.



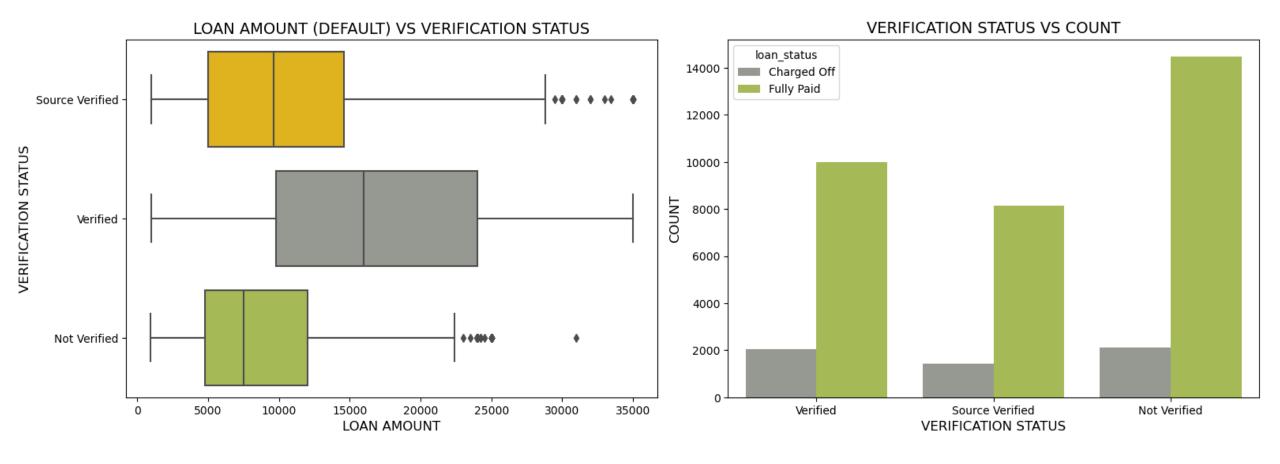
#### LOAN AMOUNT VS GRADE

• Borrower within grade E,F,G and with average loan amount greater than 15000 are more likely to charge off.



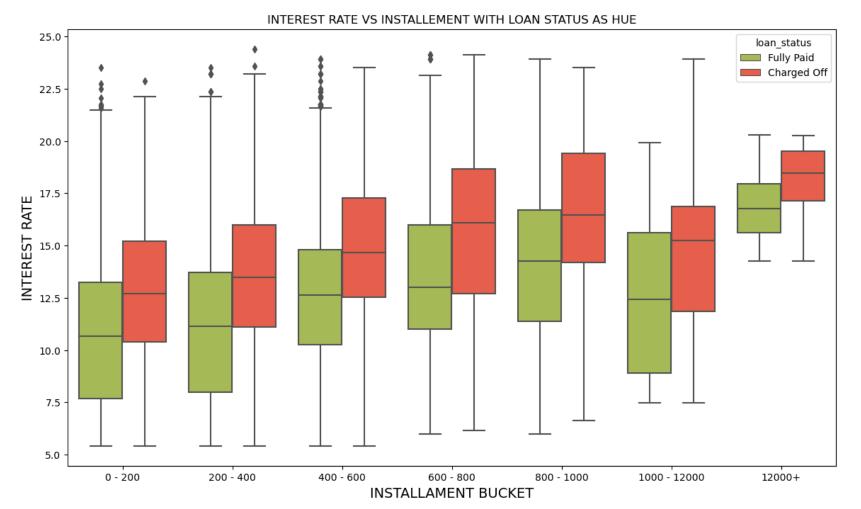
#### LOAN AMOUNT VS VERIFICATION STATUS

- Borrowers with verification status "not-verified" has loan amount with 5k to 12.5k are more likely to charge off.
- Borrowers with verification status "not-verified" and "verified" has a equal proportion in charge off



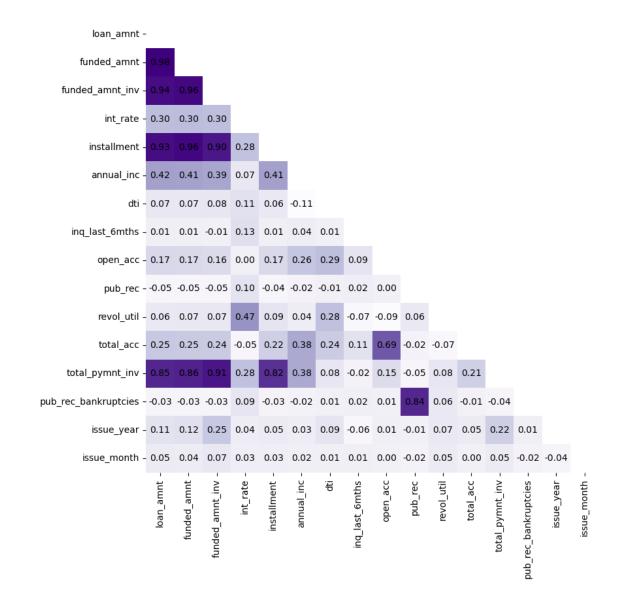
#### Loan amount vs employment period:

As the number of installment increase the borrowers with high interest rate are more likely to charge off.



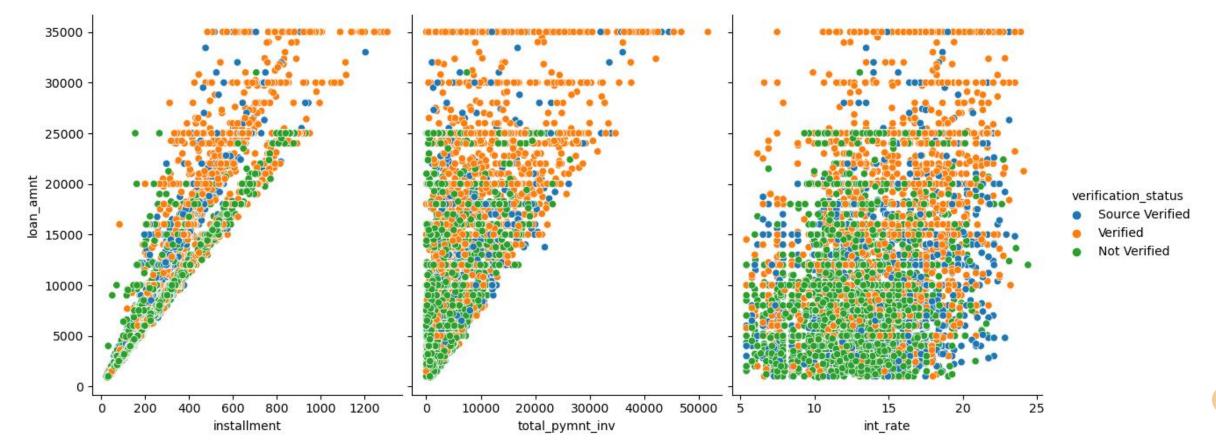
#### Correlation heatmap

- loan\_amnt has strong positive correlation with installment, total\_payment\_inv.
- loan\_amnt has intermediate positive correlation with annual\_inc, int\_rate, total\_acc.
- open\_acc has strong positive correlation with total\_acc.
- pub\_rec has strong positive correlation with pub\_rec\_bankruptcies.



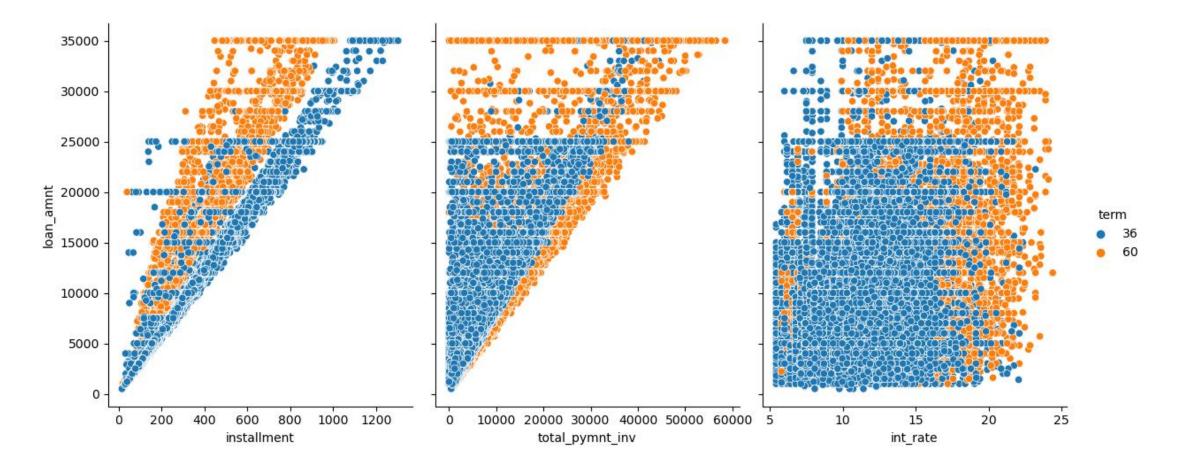
Plotting a pair plot for 'installment', 'total\_pymnt\_inv', 'int\_rate' vs loan amount by verification status as hue

- When the loan amount increases, number of installment, total\_pymnt\_inv, int\_rate also increases.
- Borrowers with verification status "Not-Verified" are most likely to get loan less than 25k (Charge off).
- Borrowers with verification status "Verified" are most likely to get loan greater than 25k (Charge off).



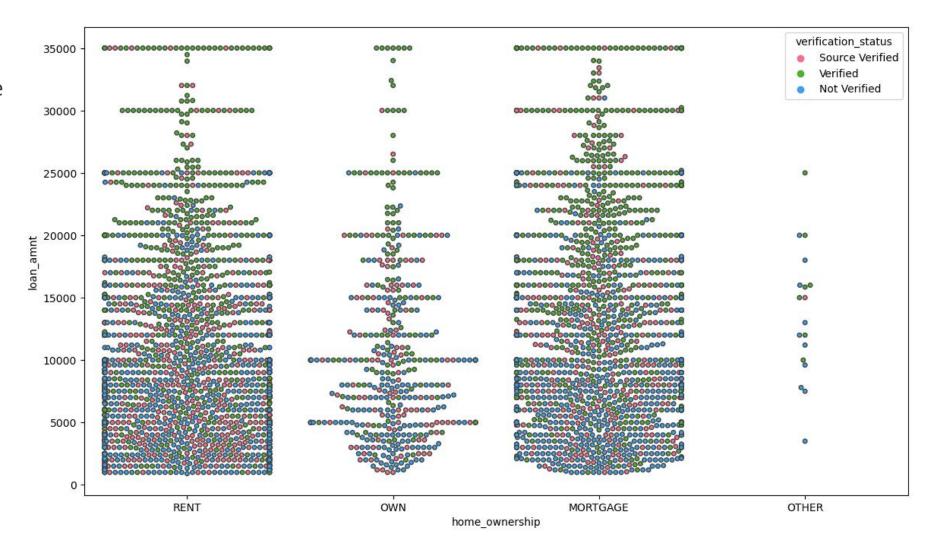
Plotting a pair plot for 'installment', 'total\_pymnt\_inv', 'int\_rate' vs loan amount by term as hue

- Borrowers with lower term has higher installments comparatively.
- Borrowers with 60 months term has highest share to total\_pymnt\_inv.



Plotting a swarm plot for home ownership vs loan amount by term as hue

- Across all categories of home ownership, borrowers with loan amount less than 25k are most like to be source verified.
- Across all categories of home ownership, borrowers with loan amount greater than 25k gets verified.



#### CONCLUSION



#### It is risky for the bank to give loan when

- Borrowers with high interest and 10+ years of employment length.
- Borrowers living on mortgage with loan amount greater than 12000.
- Loan for debt consolidation, credit card, small business with loan amount greater than 12000.
- For any loan amount provided with an average interest rate higher than 13%.
- Loan provided for house purpose with average interest grater than 15%.
- Loan provided with an average of 12.5% interest rate for 36months of tenure.
- Loan provided with an average of 16.5% interest rate for 60months of tenure.
- Average loan amount greater than 15000 with grade E, F, G.
- Installments between 800 12000 with average interest greater than 15%.
- Average interest rate of 17% with installment greater than 12000.

