

Lending case study

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Problem Statement

Two types of risks are associated with the bank's decision:

- If the applicant is **likely to repay the loan**, then not approving the loan results in a **loss of business** to the company
- If the applicant is **not likely to repay the loan**, i.e. he/she is likely to default, then approving the loan may lead to a **financial loss** for the company
- Now this would require a preliminary understanding of the features that define whether applicant's loan should be accepted or rejected.

Analysis approach

- Understanding data

Perform Data cleaning

- Remove irrelevant columns
- Remove null columns and rows
- Check for null values
- Imputation of null values
- Finding outliers with box plot

Univariate analysis

- Univariate analysis on Continuous variables
- Univariate analysis on Categorical variables

Segmented and Business driven analysis

Segmentation based on defaulter across different variables

Bivariate analysis

- Bivariate analysis on Continuous vs Continuous variables
- Bivariate analysis on Categorical vs Continuous variables

Derived matric analysis

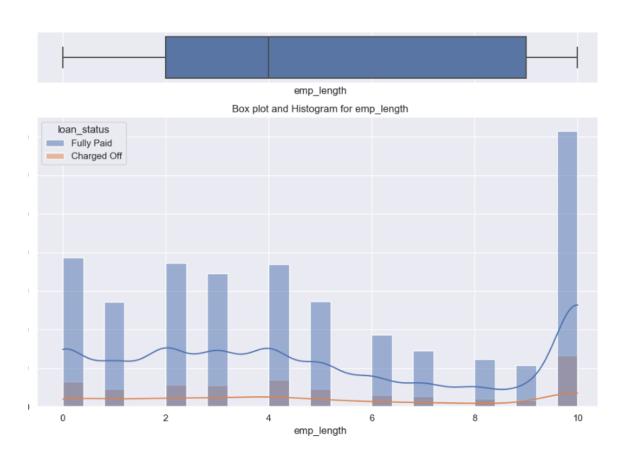
- Type driven analysis
- Data driven analysis

Conclusion

Data Understanding and Data cleaning

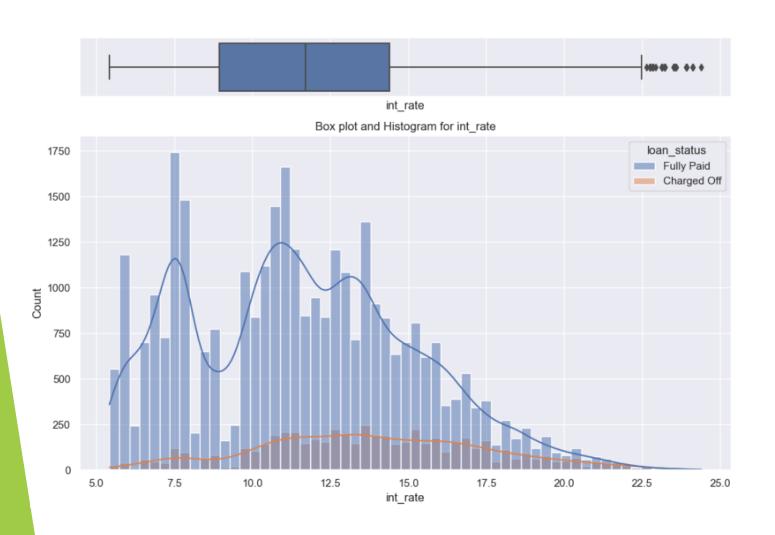
- Analysis performed on the data set with 39717 applicants with 111 attributes
- 111 attributes includes:
 - applicant relevant information
 - Loan characteristics
 - Customer behavior variables
- Through the analysis, most of the columns are dropped to identify the real driving factors(or driver variables) behind loan default
- Used appropriate imputation technique to fill out the missing values.

Univariate on Tenure or Work experience



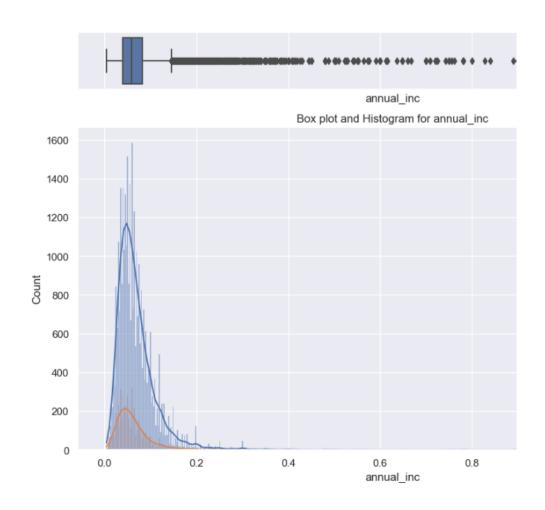
 Most of the loan applicants are having 10 or more years of work experience. So mostly they are in 30s

Univariate on "Interest rate"



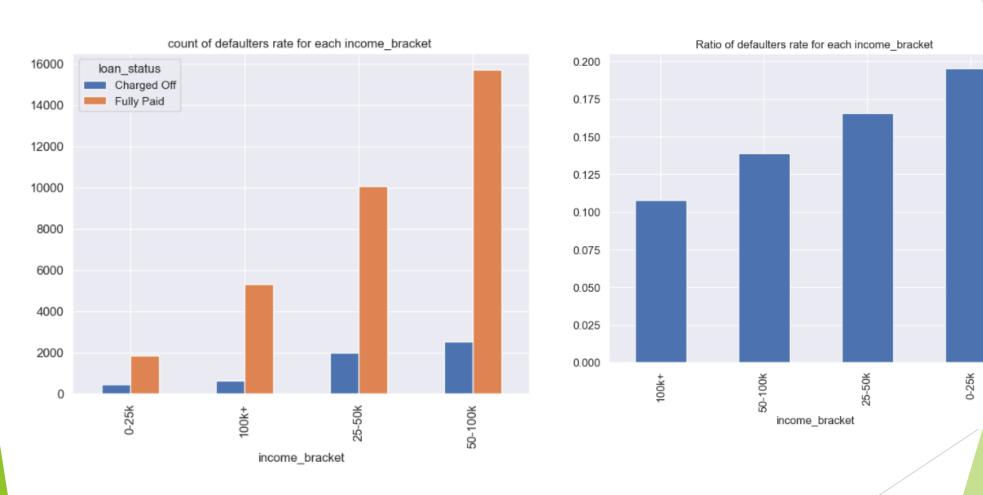
Interest rate is mostly falls between 10-14.5%

Univariate on annual income



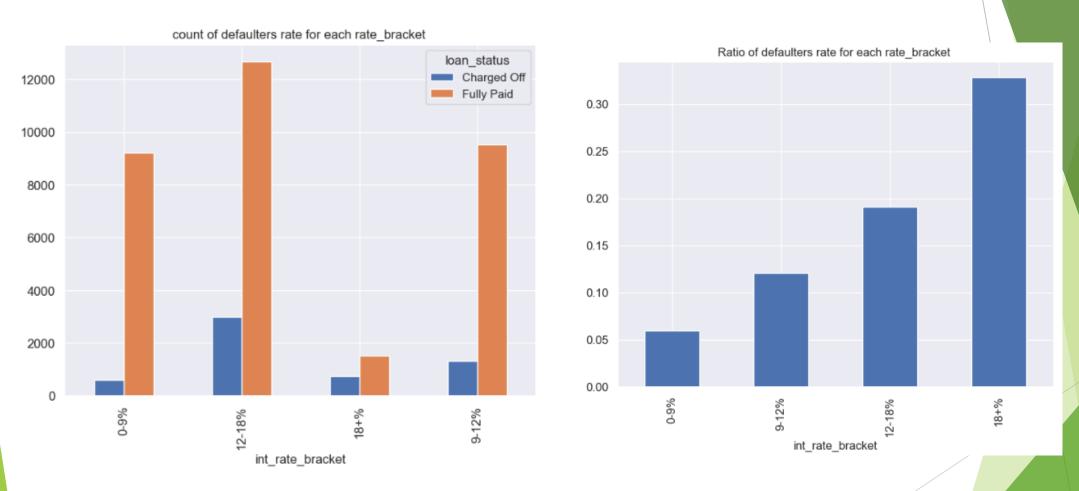
HIGH PORTION OF THE APPLICANTS ARE HAVING ANNUAL INCOME BETWEEN 50K-100K.

Applicants defaulted based on Annual income



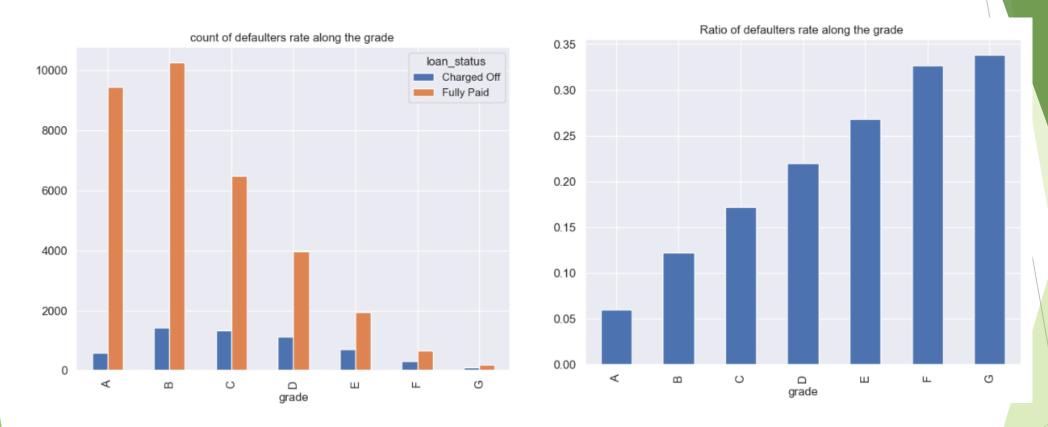
Most of the defaulters are in 0-25k bracket

Applicants defaulted based on Interest rate



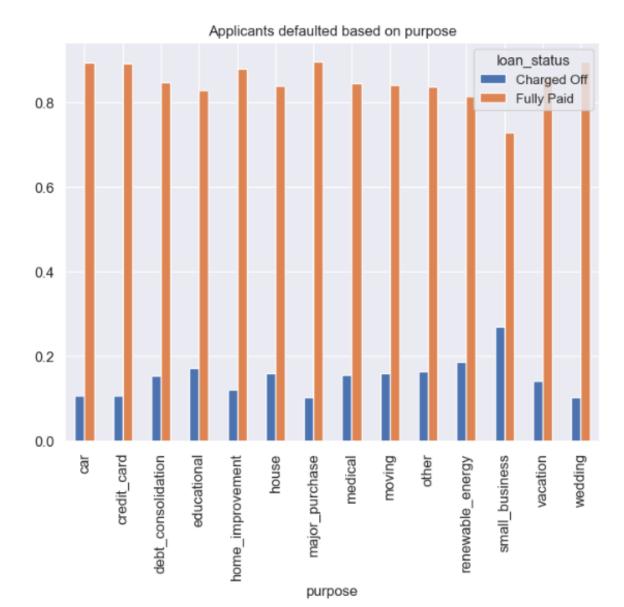
Clearly there are many defaulters in 18+% interest rate bracket

Applicants defaulted based on Grade assigned



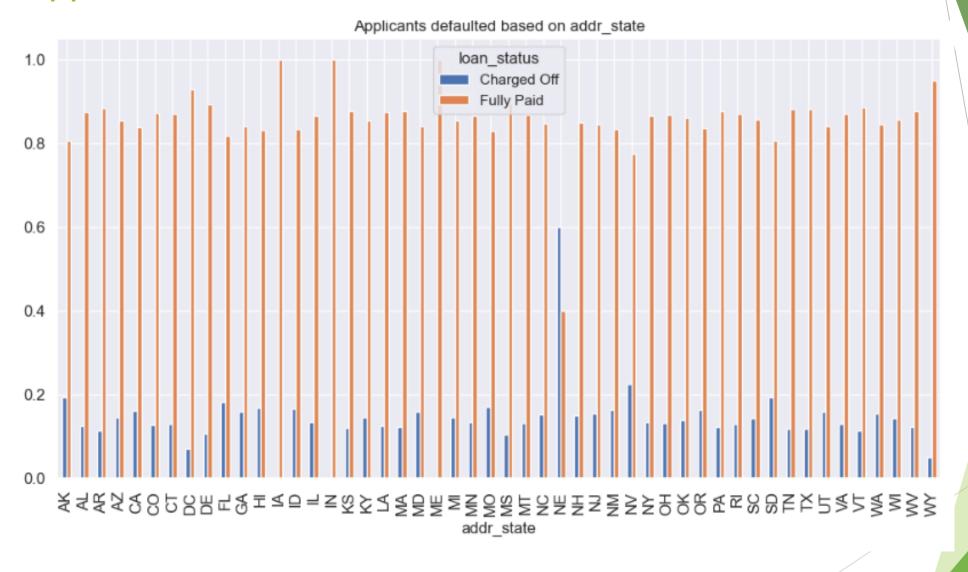
Defaulters increase with the grade but the count of F and G grade records are very less.

Applicants defaulted based on purpose stated



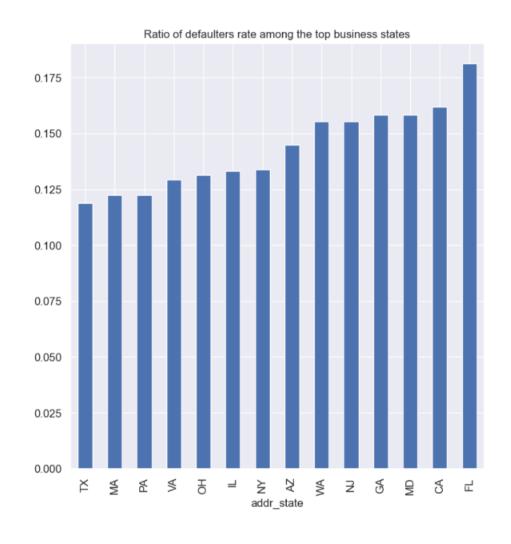
Ratio of defaulters are more in "small business".

Applicants defaulted based on Address State



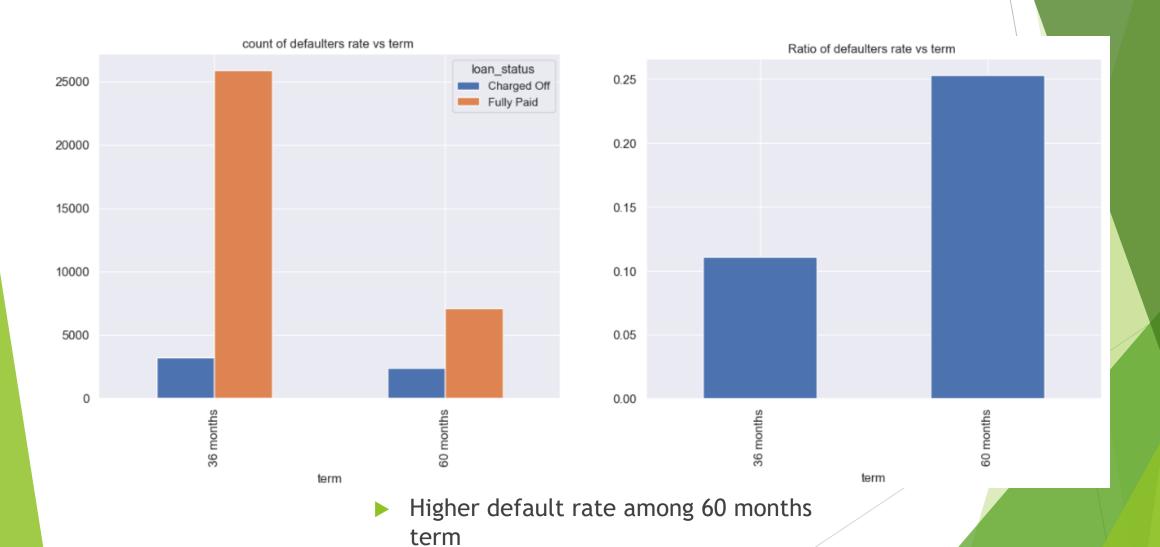
Initial observation: High defaulters in NE and IA/ME/IN is good business area with zero defaulters. Conti...

Conti: Applicants defaulted based on Address State

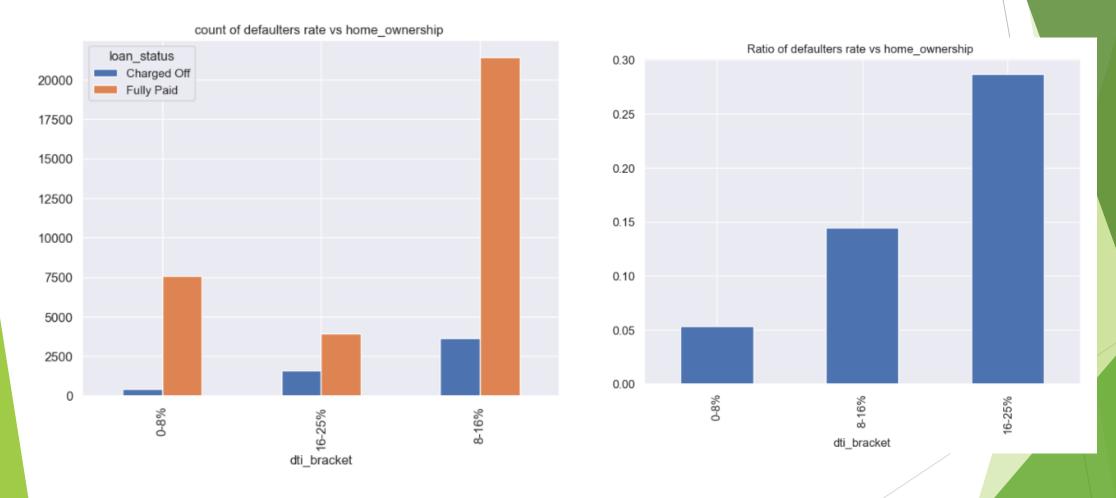


- It is observed that applicants from NE/IA/ME/IN region are in single digital.
- On continuing further analysis on top State with maximum applicant, it is found that FL is the worst performing state.

Applicants defaulted based on Term selected



Applicants defaulted based on dti



Defaulters increases with dti score

Bivariate analysis - between continuous columns

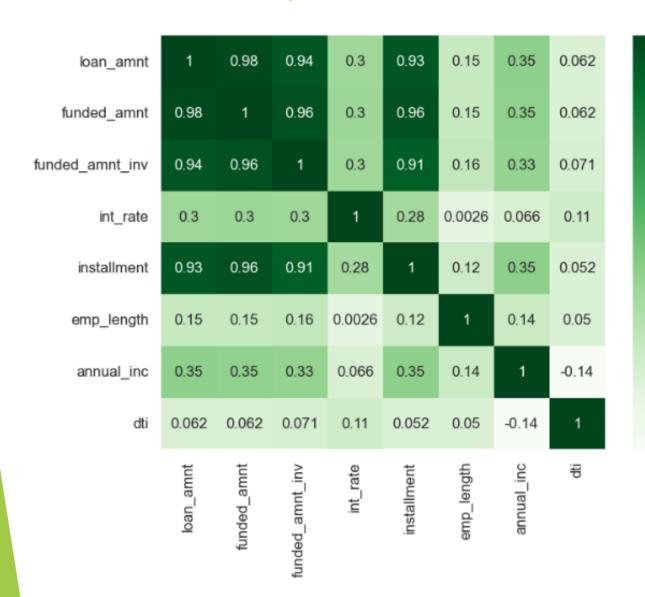
- 0.8

- 0.6

- 0.4

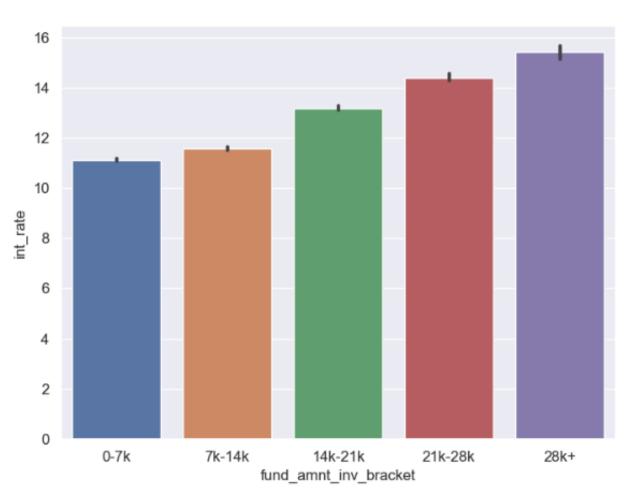
- 0.2

-0.0



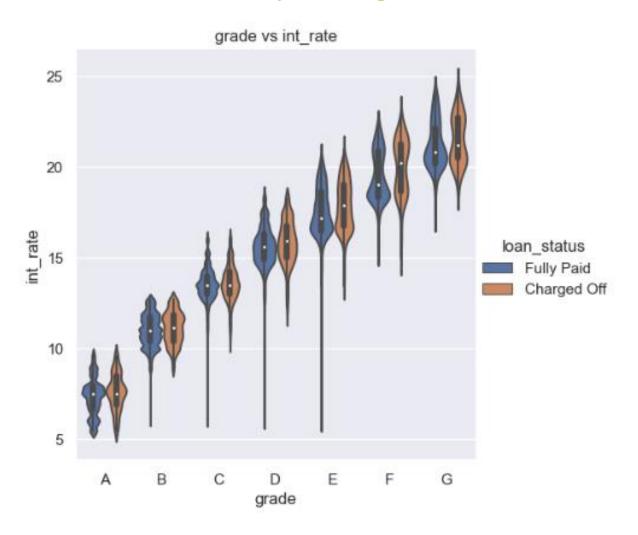
- Installment has string correlation with the amount columns which is obvious.
- Negative co-relation between annual_inc and dti
- Relatively positive co-relation between annual_inc to loan_amnt.
- As observed before, strong correlation between loan amount/funded amount /funded amount invested
- Also 0.3 positive cor-relation of int_rate with loan ampt.

Bivariate analysis - Loan amount vs Interest rate



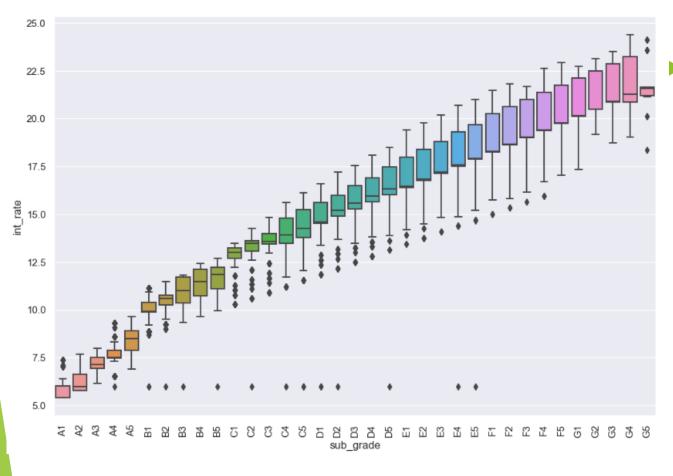
 As already observed in heatmap, the interest rate increase with the loan amount related columns

Bivariate analysis - grade vs Interest rate



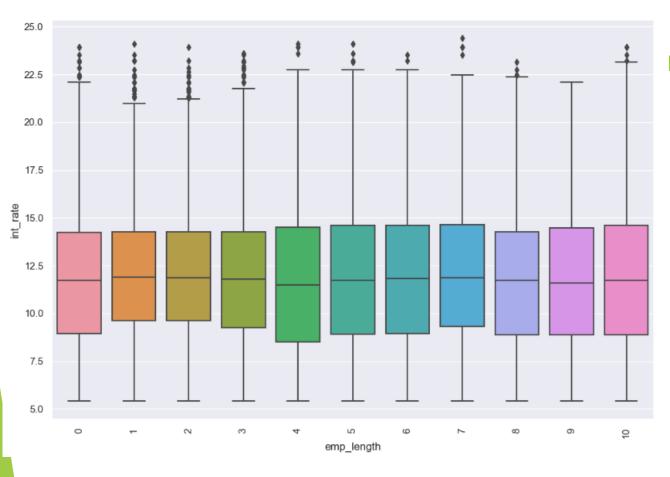
Interest rate increase with grade A->G

Bivariate analysis - subgrade vs Interest rate



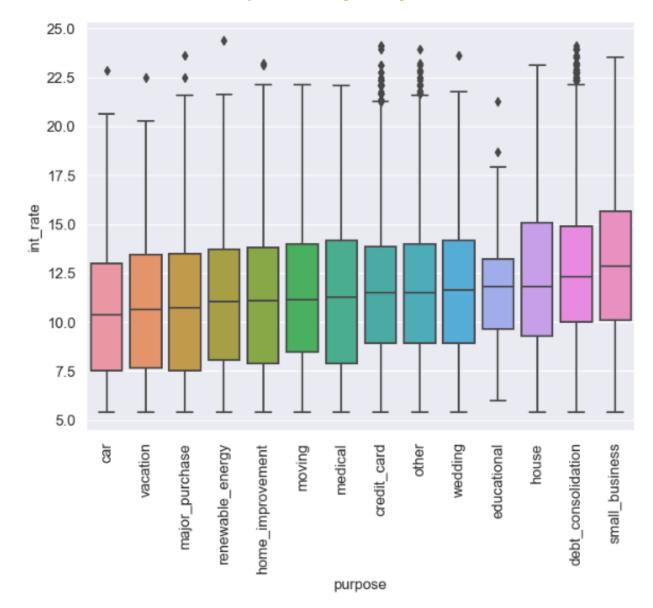
Same as grade, Interest rate increase with sub grade Ax-->Gx

Bivariate analysis - tenure vs Interest rate



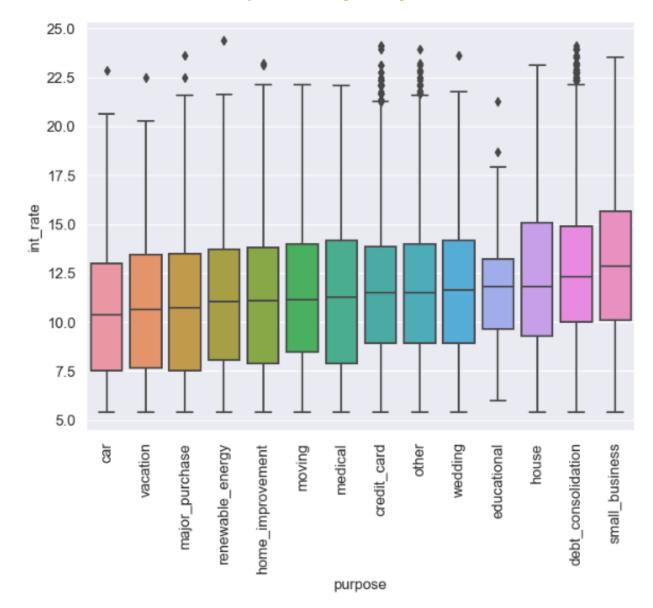
Interest rate is mostly same across all the applicant tenure.

Bivariate analysis - purpose vs Interest rate



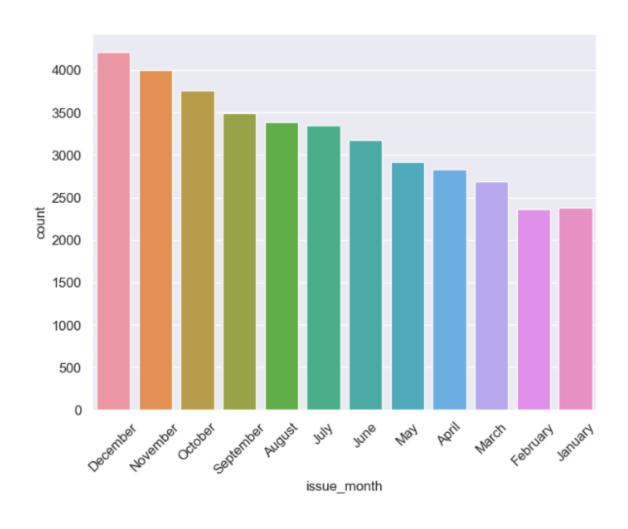
 Clearyly the 'small business' have higher interest rates followed by debt_consolidation

Bivariate analysis - purpose vs Interest rate



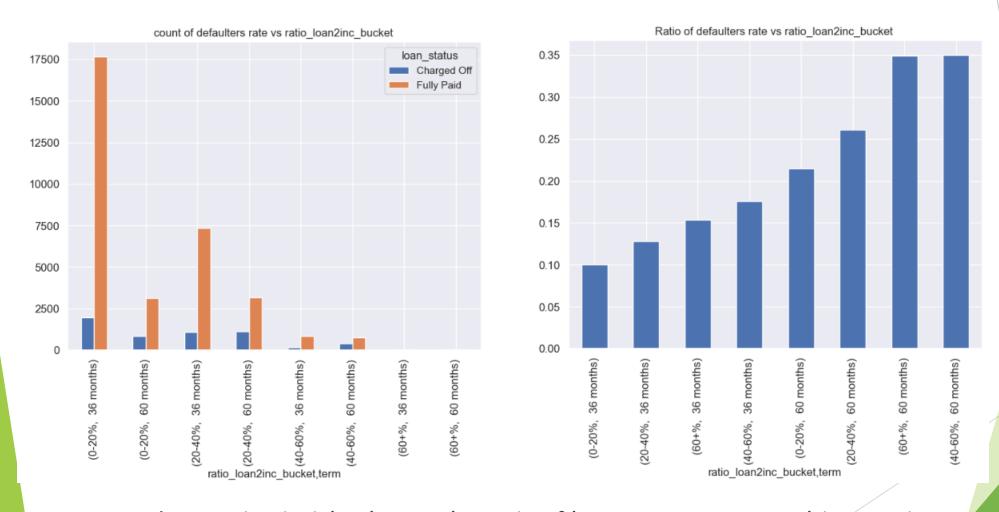
 Clearyly the 'small business' have higher interest rates followed by debt_consolidation

Derived metrics - Month from issue date



Application for loan increases towards year end. Could be because of vacation period.

Derived metrics - Defaulters rate to "Portion of loan amount to annual income"



Interesting insight that as the ratio of loan amount to annual income increase, the defaulters count increases.

Conclusions:

- Annual income Applicant with lesser income tend to default especially in the range of 0-25K. Income with more than 100K are good and 50-100K are moderate.
- Term Rate of defaulters increase with the term. As the length of period is more, there could be many factors coming with time which could give more chance for defaulting.
- ▶ Interest rate Applicant with interest rate greater than 15% are more likely to default.
- Home ownership Home ownership is relatively not affecting the defaulters count.
- ▶ DTI: Applicant with lower DTI score are preferrable. Customer falling under 16-25% highly risky applicants.
- ► Grade It is clear from analysis that with the loan amount, grade increases, which inturn increase the interest rate, so LC should be careful while granting loans with higher grade($A \rightarrow G$)
- Sub-grade: Same as grade
- Address State: Among the top business states(based on maximum applicants), highest defaulters are from FL. Order: FL>CA>MD>GA>NJ>WA>AZ>NY>IL>OH>VA>PA>MA>TX
- Purpose: LC should be careful while granding loan to "small business" have higher defaulters.
- Portion of loan amount to annual income: As per analysis, LC also should make sure the portion of the loan amount doesnot exceed 20-40% of the annual salary irrespective of term.