

Problem Statement

We work for a consumer finance company which specializes in lending various types of loans to urban customers. When the company receives a loan application, the company has to make a decision for loan approval based on the applicant's profile. 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

The aim of this case study is to identify patterns which indicate if a person is likely to default, which may be used for taking actions such as denying the loan, reducing the amount of loan, lending (to risky applicants) at a higher interest rate, etc.

Analysis Approach

In this case study, we have used EDA to understand how **consumer attributes** and **loan attributes** influence the tendency of default.

Following steps are followed in the analysis:

Conclusion

Create plots to check for univariate and Bivariate analysis

Create derived columns if required.

Drop rows for Current as we are not interested in this: we are interested in Fully paid and Charged off.

Describe dataset.

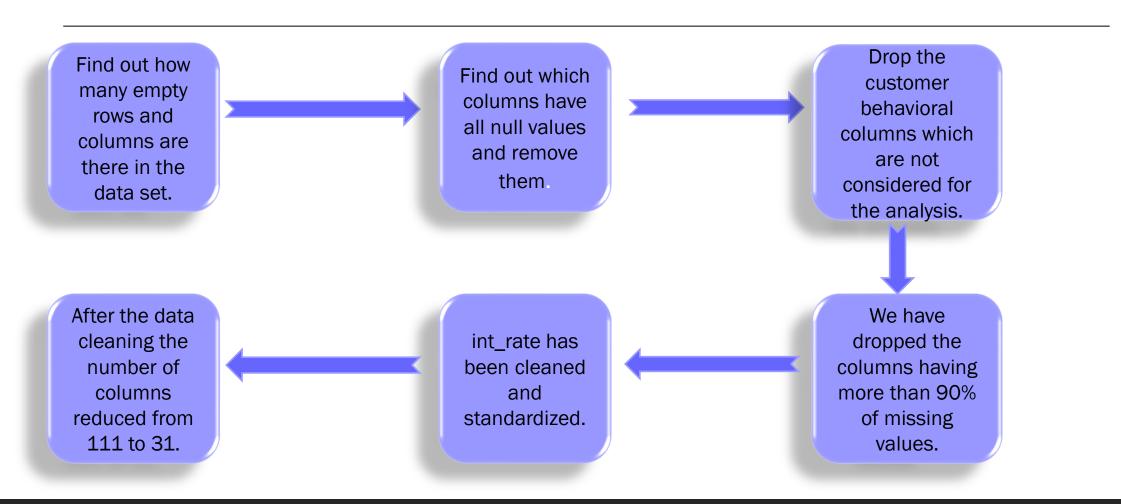
Drop columns like URL or cutomer behavior variables which may not have any impact.

Drop all coulmns with all value Nan or if column values are same like 0.

First load data and get shape.

Data Cleaning

Steps followed in the data cleaning



Derived Metrics

- has_defaulted column has been derived from loan_status.
- Created income_category from annual_inc column.

Categorical division of the following columns is done

Loan Amount. Invested Interest **Annual** Rate Income Emp Length

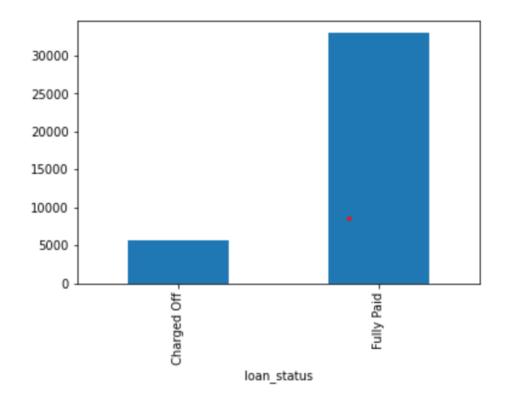
Univariate Analysis

Loan Amount Invested Annual Income Term of Ioan Emp Length

Observations LOAN Amount

- ❖ Loan amount is in the range 5000 to 15000.
- ❖ Around 14.16% loans were charged off out of total loan issued

Fully Paid	83 %
Charged off	14%
Current	3 %



Bivariate Analysis

has_defaulted column has been used for bivariate analysis for following variables:

Loan Term

Interest rate

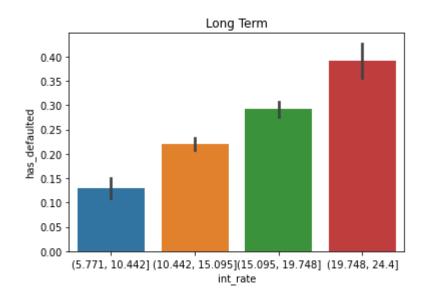
Annual Income

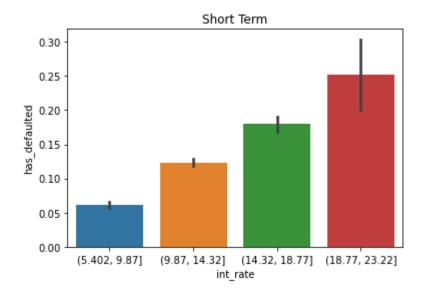
Grade

Purpose

Previous bankrupt cies

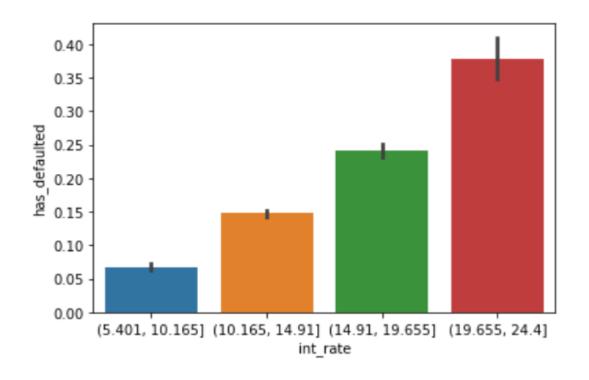
Loan Term observation





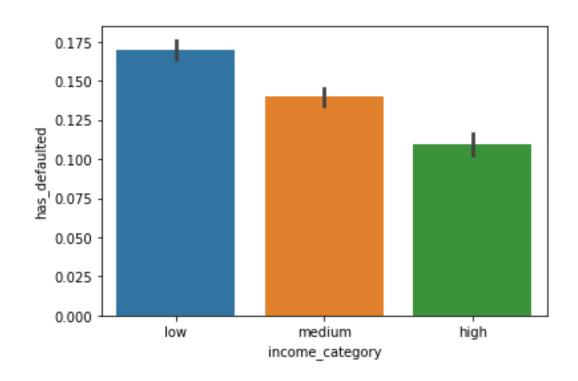
Long term loans has high defaults.

Interest rate observation



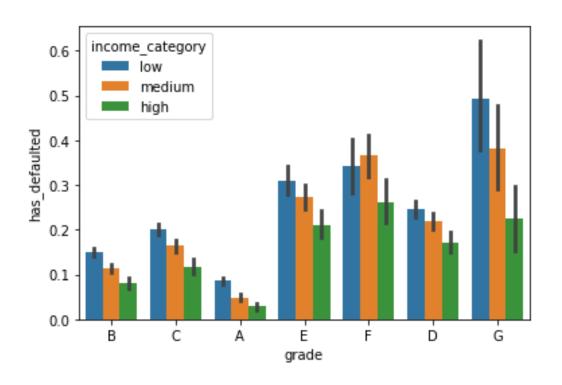
Higher interest rate is a driving factor for high defaults

Annual Income observation



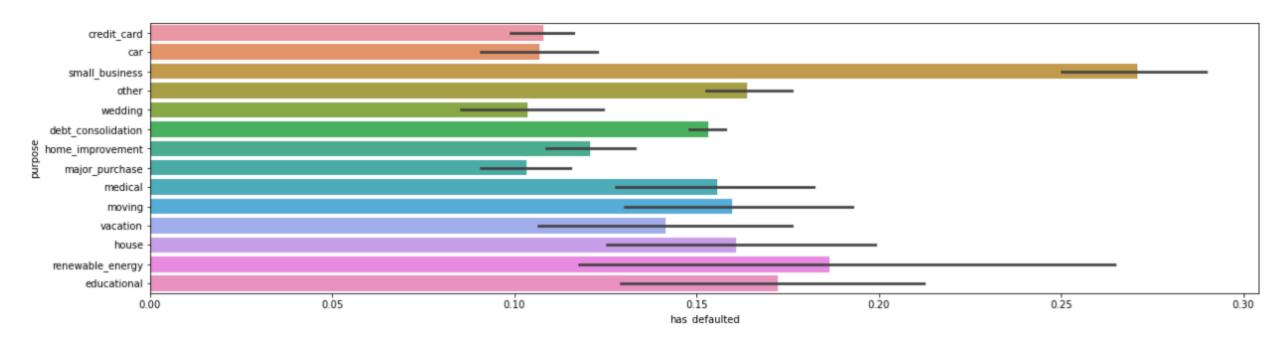
The defaulters are present more in low income group.

Grade observation



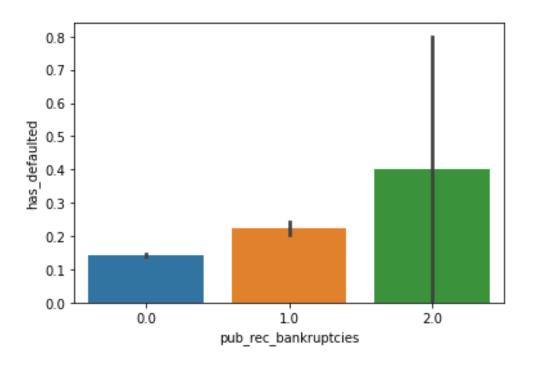
In loans with different grades, loan with G grade have defaulted the most.

Purpose observation



Borrowers with purpose 'small_business' have highest rate for defaults.

Previous bankruptcies observation



Borrowers, who went bankrupt earlier, will most probably default.

Consolidated observations

Annual income:

- ❖loan amount increases when borrower has high income level.
- ❖low income group borrower are less likely to pay the loan.
- ❖Interest rate: Higher interest rate has led to high defaults.
- ❖Term of loan: Long term loans has high defaults.
- ❖Grade Analysis: Providing loan to a borrower with credit grade G could be risky as loan with G grade have defaulted the most.
- *Purpose of loan: Borrowers with purpose 'small_business' have highest rate for defaults
- *Earlier Bankrupt Borrowers: Borrowers, who went bankrupt earlier, will most probably default.

Recommendation

- * Reduce giving high interest rate loans with low income group.
- Stop high risk loan like earlier bankrupt borrowers for small business purposes.
- Long term loans should be minimized and focus can be on short term loans.
- Focus on increasing safe loans like
 - ❖Borrowers having record of bankruptcy
 - Loan with low interest rate
 - ❖Short term loans



Thank you

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