

Lending Club Case Study

(Assignment submission)

Venkata Prakash Reddy A
Ganga Gowthami

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Problem statement

The data given contains information about past loan applicants and whether they 'defaulted' or not. The aim 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.

Step 1: Data Sourcing

The data provided in this case study is private data. all the data is provided in one file loan.csv

We need to perform the

Manual inspection and columns cleaning –

- Below mentioned columns will be removed as they donot have valid data.
- Some of these columns are customer behaviour columns for which we will not be having data at the time of loan application hence deleting them.
- 26 columns are left after cleaning

Rows Cleaning –

- Removing the rows with loan status current as they will not be usefull for loan default analysis

Treating missing values –

- Checking the percentage of missing values for each column.
- Removing the columns having high percentage of missing values
- Imputing missing values for columns with less percentage of missing values

Prepared clean data set without missing values

- After performing the above mentioned steps we got a cleaned data set without missing values and 22 columns.

Step3- Derived Columns

- Derived 2 new columns issue_month and issue_year from column issue_d

Step4- Univariate Analysis

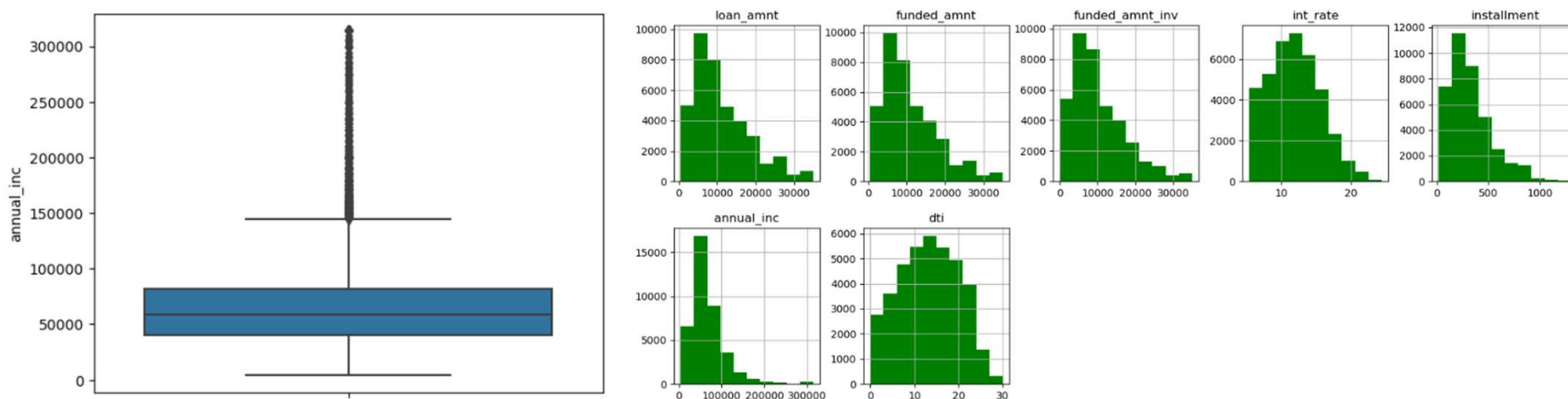
- Splitting the columns into categorical and quantitative columns
- This splitting is done based on manual inspection and the number of unique values present in each column

Quantitative variables univariate analysis

- Plotted Histograms for understanding distribution of Quantitative variables
- All the distributions look fine except annual_inc.

Outliers detection and correction

- As the histogram of Annual_inc is not distributed properly there might be chances of outliers.
- Used box plot to understand the outliers of Annual_inc.
- Created function to check outliers for all the columns.
- while treating outliers considered 5th percentile as Q1 and 95th percentile as Q2 as this is financial data
- verified if there are any outliers in other columns using the function.



Categorical variables univariate analysis

- Plotted bar graph for all the categorical variables to understand distribution of data with respect to categorical variables.

Inferences from categorical variables univariate analysis

- Initial_list_status feature can be ignored as it has only one value.
- The number of loans gradually increases year by year.
- More loans are issued at the year end in the month of december. It might be because of festival season.
- Most of the borrowers houses are under mortgage or they are renting the house.

Step5- Bivariate Analysis

- Bivariate analysis is used to find the relation between 2 features of the dataset.

Categorical variables Bivariate analysis

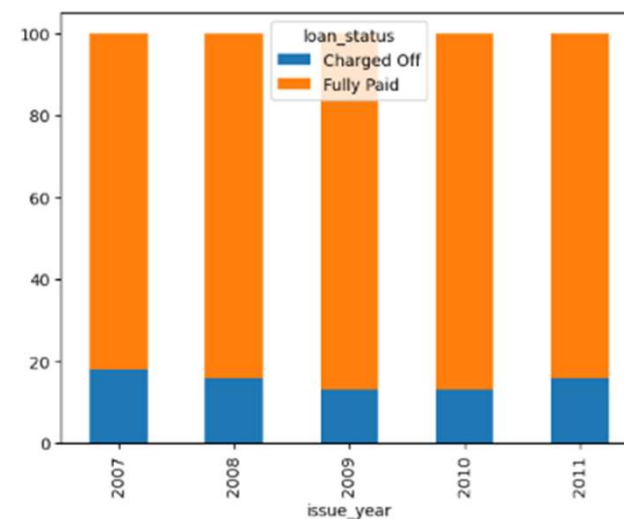
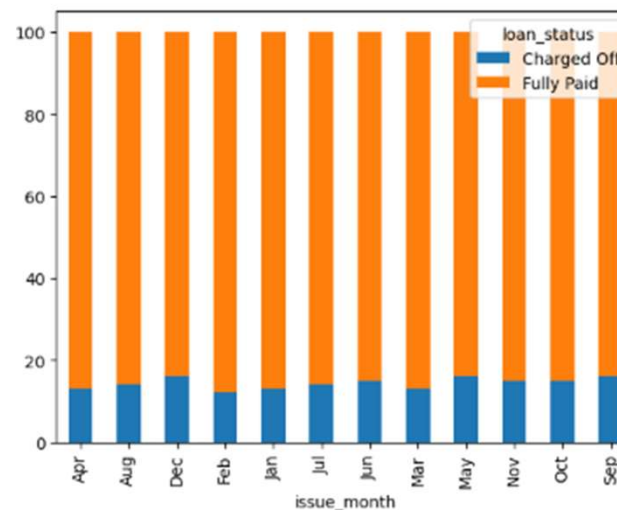
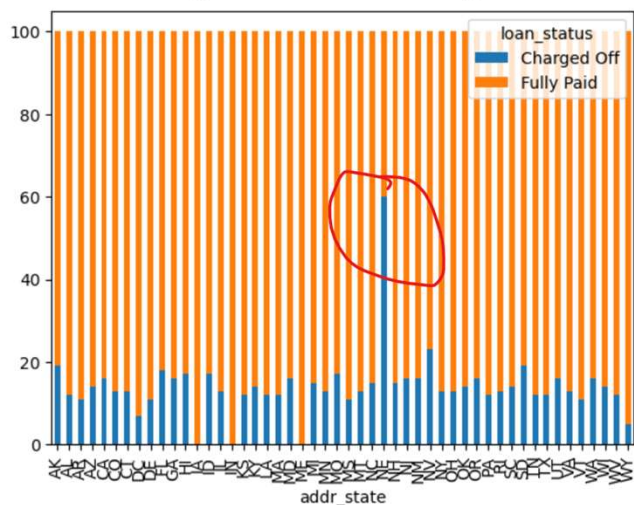
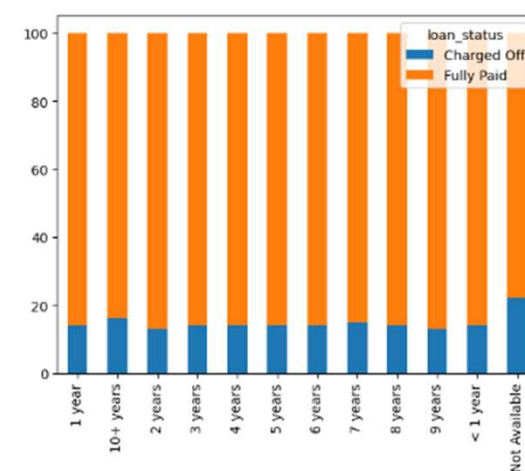
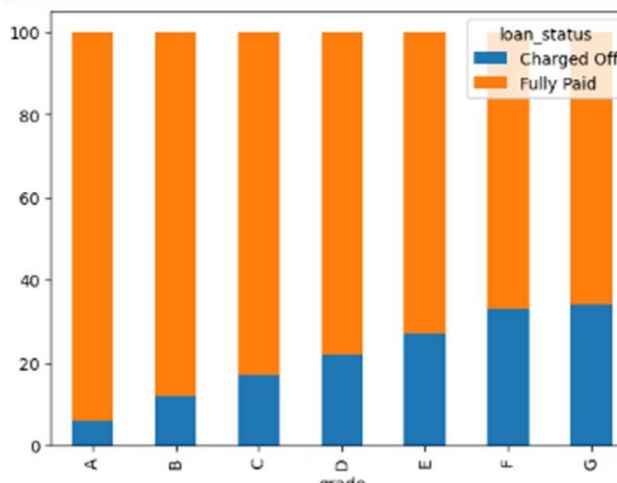
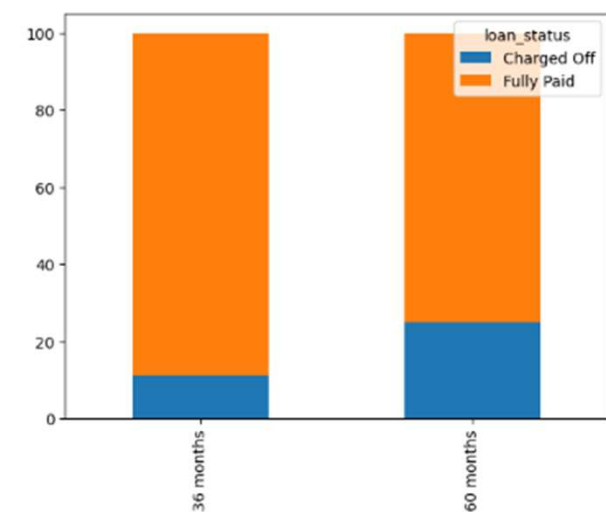
- Bivariate analysis of categorical variables is performed against the target loan status with stacked bar charts.

Inferences from categorical variables bivariate analysis

- Term - 60 Month term loans have higher percentage of defaulters
- Grade - The percentage of defaulters increases gradually as the grade changes from A to G
- Sub_grade - sub grade F5 has highest number of Defaulters between 45 to 50 %
- Emp_length - There is no impact of emp length on default percentages.
- Home_ownership - there is no specific trend of defaulters based on home ownership.
- Verification_status - surprisingly verified customer have slightly more percentage of defaulters than non verified
- Purpose - Loans taken for small business have high default rate.
- State - The state NE(Nebraska) has very high number of defaulters, also in this state defaulters are more than fully paid borrowers
- Pub_rec_bankrupcies - higher the number of bankruptcies higher the chances of defaulting the loan.
- Issue_month - no specific trend of defaulters percentage across the months.
- Issue_year - no specific trend of defaulters percentage across the years

Step5- Bivariate Analysis

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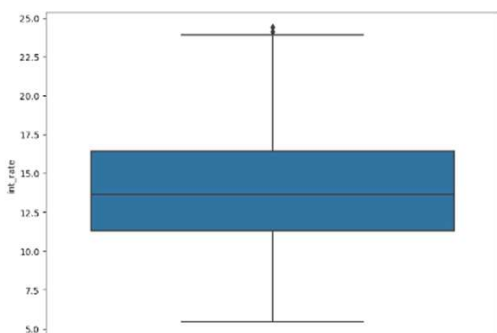
Step5- Bivariate Analysis

Quantitative variables Bivariate analysis

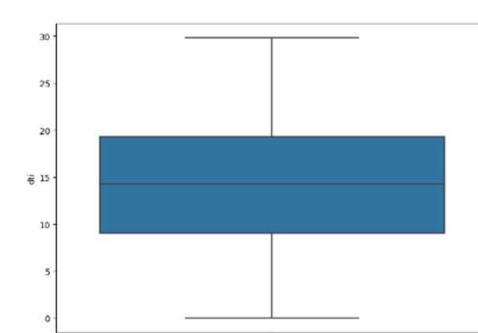
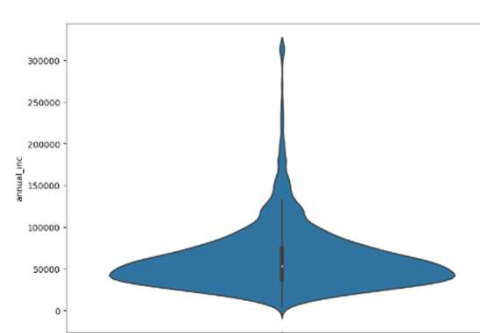
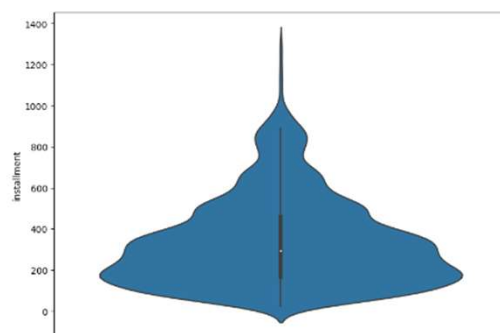
- For this analysis we will consider the data with loan_status Charged Off only.
- With this data we can check how the defaulted loans are spread with respect to quantitative variables.

Inferences from quantitative variables bivariate analysis

- Int_rate - most of the loans with interest rate between 11 to 16 % are defaulted
- Installment - most of the defaulted loans have installements less than 1000
- Annual_inc - most of the defaulted loans have annual income less than 16000
- Dti - most of the defaulted loans have dti between 9 to 20.
- Loan_amnt - the loan amount of most of the defaulted loans is between 5000 to 16000
- Funded_amnt - same behaviour as loan_amt
- Funded_amnt_inv - same behaviour as loan_amt



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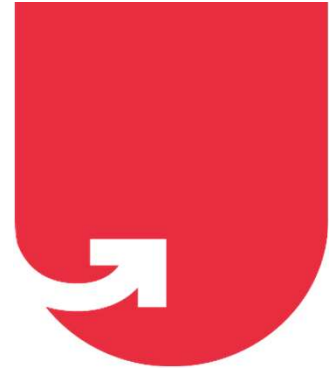


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Below are the high level conclusions. for more details check inferences sections of different analysis.

- Loans take in the state of NE(Nebraska) has very high very high risk of getting defaulted.
- Loans with smaller amounts have high risk of getting defaulted.
- Loans given to the customer with history of bancruptcy have high chances of defaulting.

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Thank You!