

LENDING CLUB CASE STUDY

Exploratory Data Analysis Assignment

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BUSINESS OBJECTIVE

A leading consumer finance company, which facilitates disbursal of personal loans, business loans, and loans for other purposes, via an online marketplace wants to identify "risky" loan applicants so that it can reduce disbursal of bad loans. At the same time legit applications shall not be rejected. Thus the key objectives are to:

- Study the data available for past loans
- Apply Exploratory Data Analysis techniques and identify patterns which indicate if a person is likely to default.
- Use EDA to understand how consumer attributes and loan attributes influence the tendency of 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

APPROACH

The entire study was broken down into simple steps, which were further divided into smaller sub steps.

The broader steps are shown below (in the order of their occurrence)...

Data Assessment

Data Cleaning

Data Preparation

Data Analysis

(univariate & bivariate analysis)

Conclusion

DATA ASSESSMENT

Load data file and examine rows/columns using read_csv method.

Understand columns using the Data Dictionary

Obtain info on dataframe using **info** method such as the number of columns, column labels, column data types, memory usage, range index, and the number of cells in each column (non-null values)

Using the **describe()** method calculated some statistical data like percentile, mean and standard deviation of the numerical value.

DATA CLEANING

Drop columns with high null values

Drop unique columns

Drop columns not relevant for analysis

Drop rows where loan status is "Current"

DATA PREPARATION

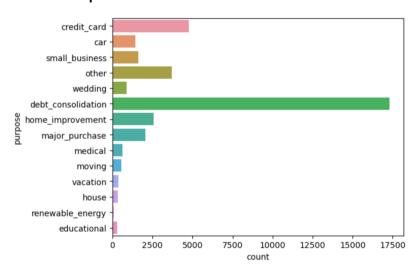
Converting text values to numeric values, e.g. int_rate, loan_status, term, etc.

Setting correct data type on numeric columns, e.g. emp_length, loan_amnt, annual_inc, etc.

Outlier treatment

Creating data groups on continuous data columsn, e.g. annual_inc, loan_amnt, dti

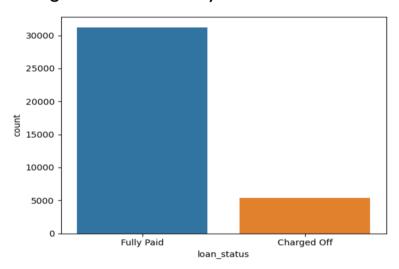
Loan Purpose



Observation:-

Nearly 50% of the loans were taken for debt consolidation. The next big head was that of credit card.

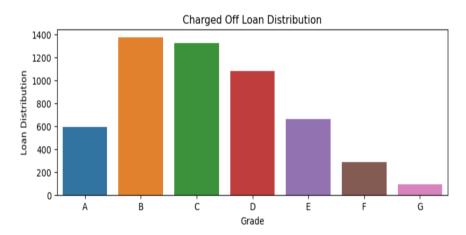
Charged off Loan vs Fully Paid Loans



Observation:-

Nearly 14% of loans turn out to be bad loans.

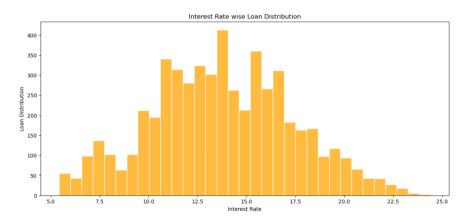
Charged off loan distribution / Grade



Observation:-

Loan default rate is high for Grades B, C & D.

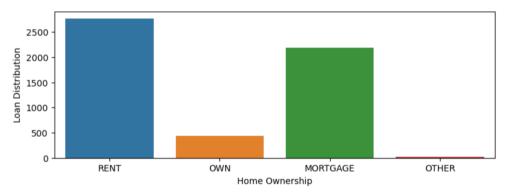
Interest rate wise loan distribution



Observation:-

The graph indicates that loans taken at higher interest rates are more likely to default.

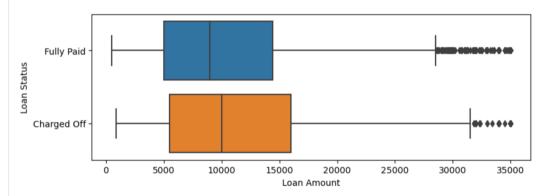
Charged off loan distribution / Home Ownership



Observation:-

People who live in Rented accommodation or have their homes mortgaged are more likely to default.

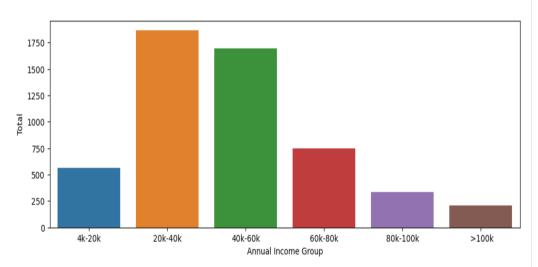
Loan Status / Loan Amount



Observation:-

It can be inferred from the above graph that if the loan amount is higher, chances of default are more.

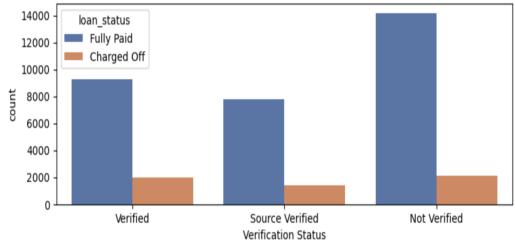
Monthly income group vs Charged off loans



Observation:-

Maximum loan charge off happened in the annual income group ranging from 20k to 60k. Higher the annual income, lesser the default.

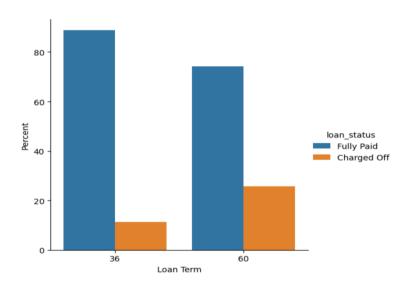
Verification status and its impact on Charged off loans



Observation:-

Not verified loans default the max. However, there are large number of verified loans also which have defaulted. Verification process is not robust.

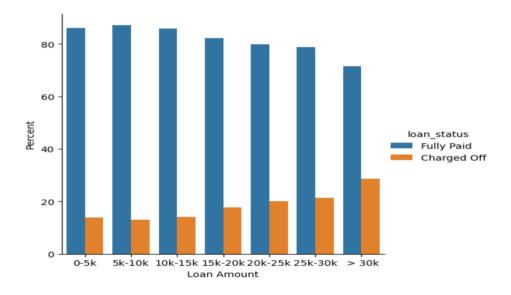
Loan Term and its effect on Loan Status



Observation:-

Loans given for longer terms are more likely to get charged off.

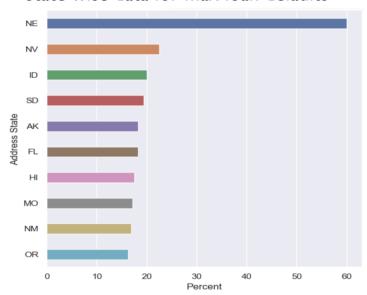
Loan Amount and its effect on Loan Status



Observation:-

Likelihood of loan default increases with loan amount.

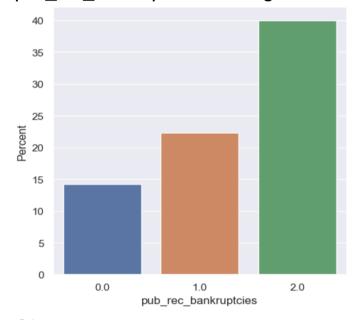
State wise data for max loan defaults



Observation:-

Maximum loan default comes from states: NE, NV, ID, SD, AK, FL.

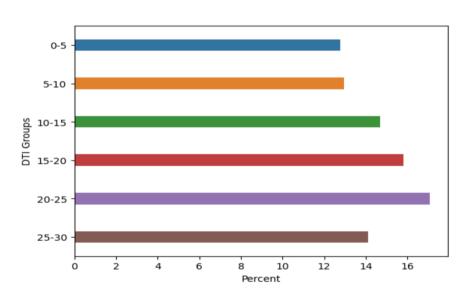
pub_rec_bankruptices vs charged off loans



Observation:-

Higher the public records for bankruptcies, greater is the risk of loan default.

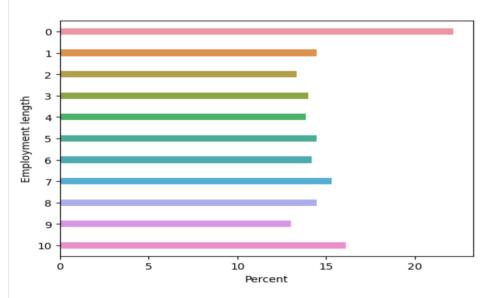
DTI Groups vs Charged off Loans



Observation:-

Increase in Debt to Income ratio may lead to higher default rate.

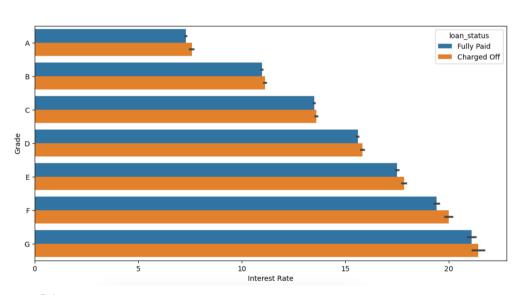
Employment length vs Charged off Loans



Observation:-

Unknown Employment Duration has max charged off loans.

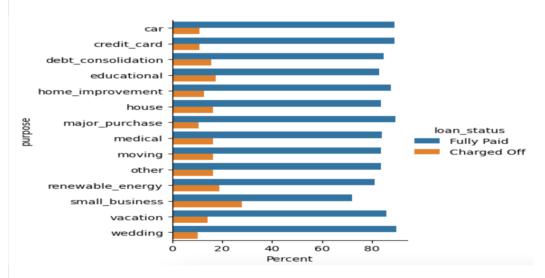
Grade & Interest Rate effect on loan status



Observation:-

Higher the Grade & Interest Rate, higher the Charged Off loans rate.

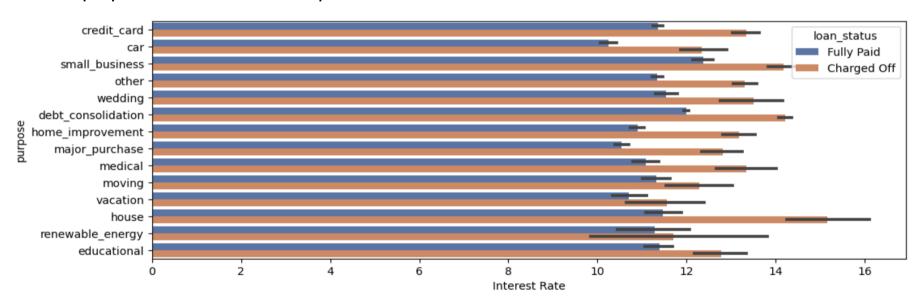
Loan purpose vs Loan status



Observation:-

Small Business loans are more likely to get charged off.

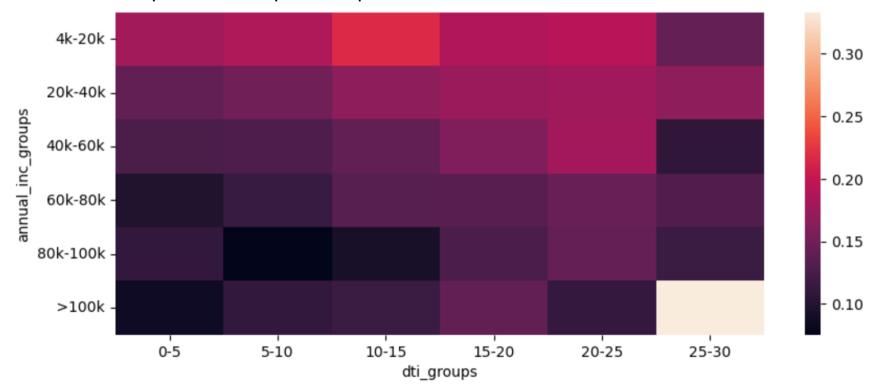
Loan purpose & Interest Rate's impact on Loan Status



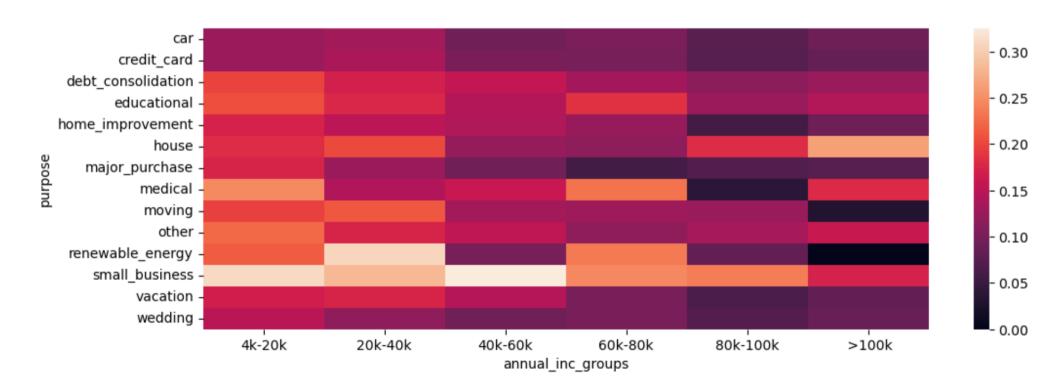
Observation:-

Small business and House purpose loans were given at very high interest rates, which led to high percentage of default. Interest rate has huge impact on loan default.

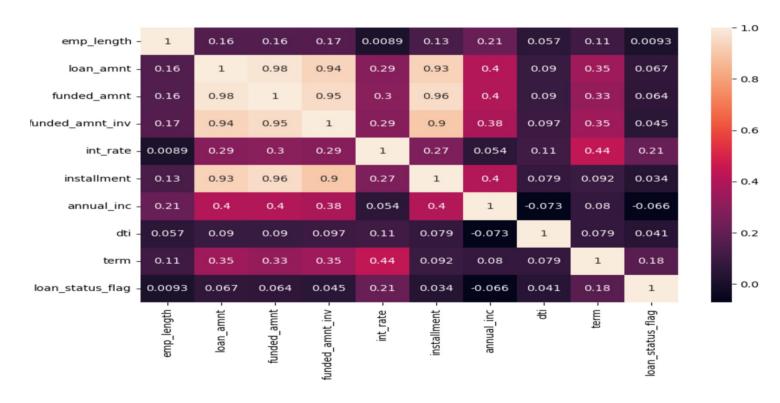
Annual Income Groups – DTI Groups heatmap



Annual Income Groups – Loan purpose heatmap



Heatmap of correlation among numeric variables



CONCLUSION / RECOMMENDATION

- Loan default rate is high for Grades B, C & D.
- Higher the Grade & Interest Rate, higher the Charged Off loan rate.
- People who live in Rented accommodation or have their homes mortgaged are more likely to default.
- Caution has to be exercised while giving loans in the lower annual income group specially at higher interest rate.
- Not verified loans default the max. However, there are large number of verified loans also which have defaulted. Verification process needs to be improved to reduce loan default.
- Loans given for longer terms are more likely to get charged off.
- Likelihood of loan default increases with loan amount.
- Maximum loan default comes from states: NE, NV, ID, SD, AK, FL.
- Higher the public records for bankruptcies, greater is the risk of loan default.
- Increase in Debt to Income ratio leads to higher default rate.
- Unknown Employment Duration has max charged off loans.
- Small business and House purpose loans were given at very high interest rates, which led to high percentage of default.
 Interest rate has huge impact on loan default.