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### PROBLEM STATEMENT

Lending Club is the largest online loan marketplace, facilitating personal loans, business loans, and financing of medical procedures. Borrowers can easily access lower interest rate loans through a fast online interface.

Lending loans to 'risky' applicants is the largest source of financial loss (called credit loss).

Credit loss is the amount of money lost by the lender when the borrower refuses to pay or runs away with the money owed.

In other words, borrowers who default cause the largest amount of loss to the lenders. In this case, the customers labelled as 'charged-off' are the 'defaulters'.

Aim of this case study is Identify the risky loan applicants using Exploratory Data Analysis, then such loans can be reduced thereby cutting down the amount of credit loss.

Lending Club wants to understand the driving factors behind loan default and identify the variables which are strong indicators of default.



**Data Cleansing** 

Find Columns to apply to Exploratory data analysis methods

Data Analysis

- Univariate Analysis
- ByVariate Analysis
- Plot and charts

Observations

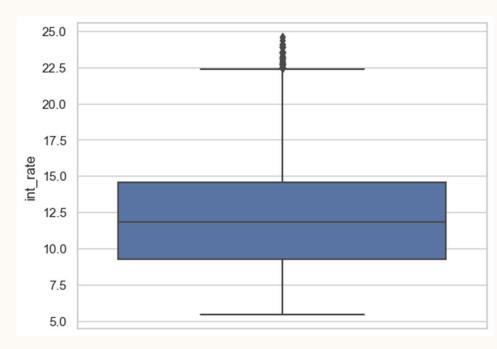


Pandas, numpy, seaborn and matplotlib.pyplot libraries used for the analysis Almost 50+ columns identified with null values are dropped from the data Derived variables created from Date and year columns of loan issued year # % to be stripped from intrest rate column and convert emp\_length column to have only digit values

converted 'loan\_amnt', 'funded\_amnt', 'int\_rate', 'funded\_amnt\_inv', 'installment', 'annual\_inc', 'dti', 'emp\_length', 'total\_pymnt' columns into numeric data for considering to calculations

### UNIVARAITE ANALYSIS

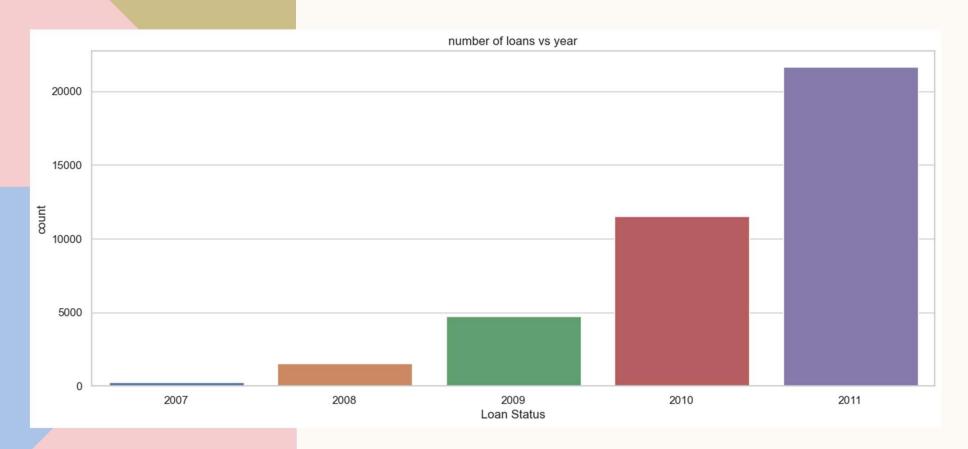




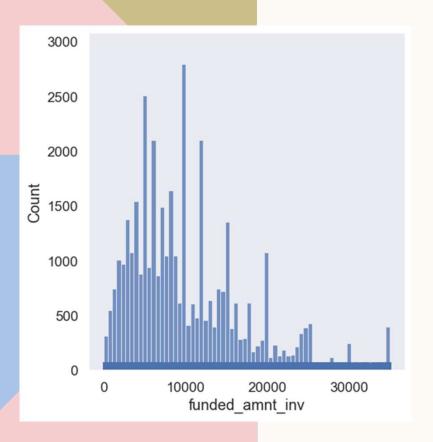
Most loan amounts within the range of 5000 to 15000

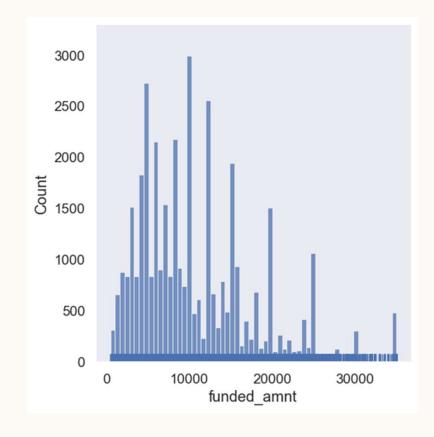
Most loan interest are within the range of 8% to 15%

### # Looking at loan\_status vs int\_rate vs issued year by plotting

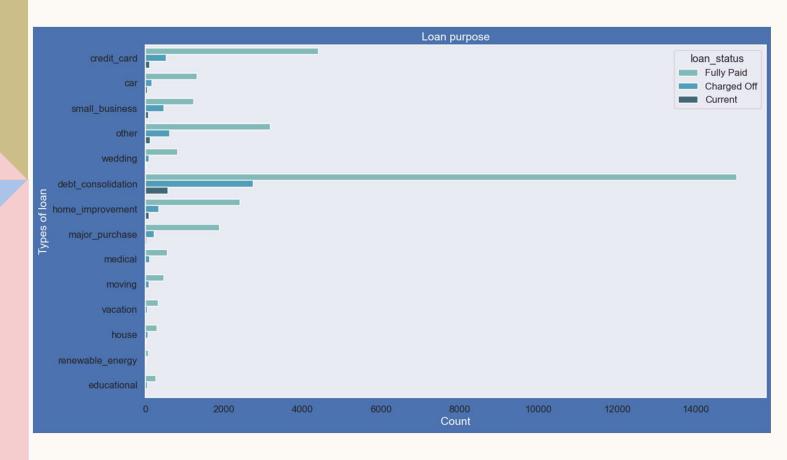


### Most loan amounts within the range of 5000 to 15000

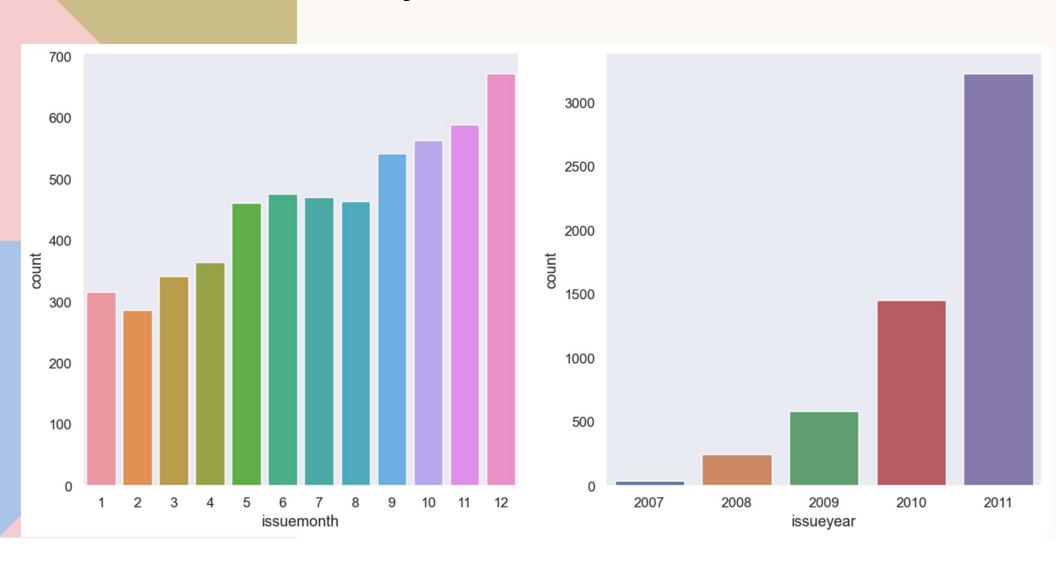




### Loan purpose with status

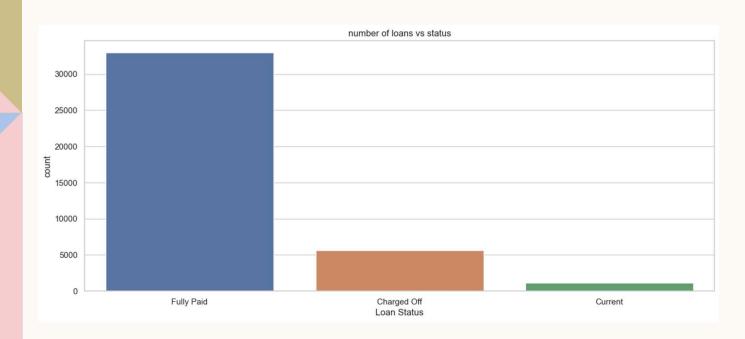


### Charged off Loans vs Issued Month and Year



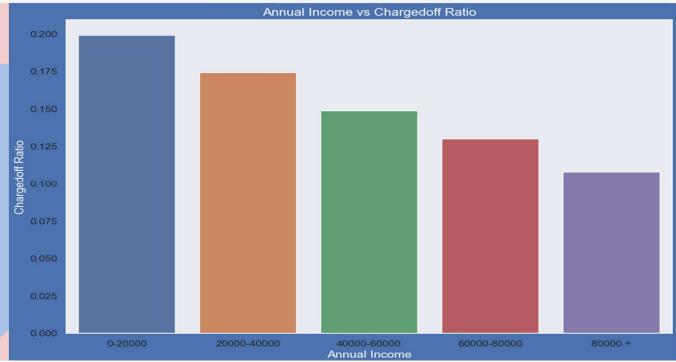
### **BIVARAITE ANALYSIS**

# Looking at loan\_status vs int\_rate vs issued year by plotting



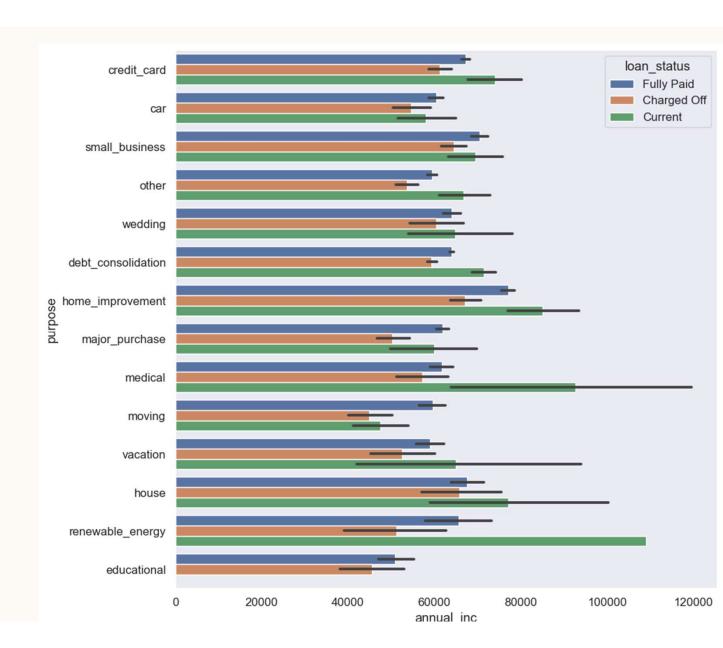
### # Bivariate Analysis of annual income versus default rate -

loan_status	annual_inc_grp	Charged Off	Current	Fully Paid	Total	Charged_off_ratio
0	0-20000	237	9	943	1189	0.199327
1	20000-40000	1514	170	7004	8688	0.174263
2	40000-60000	1729	345	9534	11608	0.148949
3	60000-80000	1024	240	6597	7861	0.130263
4	80000 +	1045	352	8181	9578	0.108292

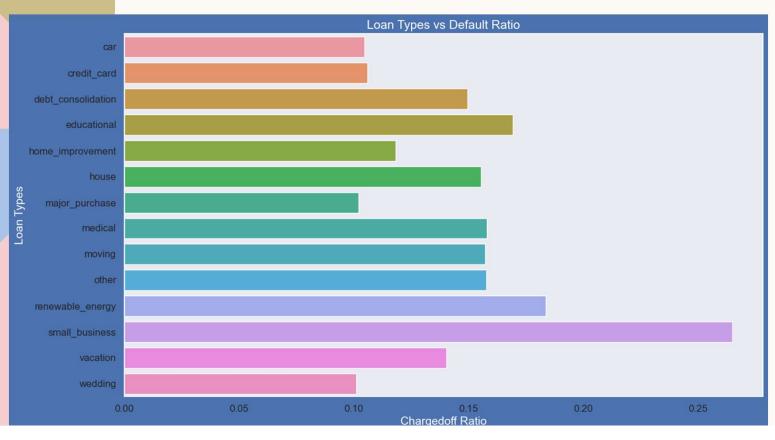


- # Observations:
- # small Business applicants have high charged off.
- # renewable\_energy where changed off pas compare to other categories.

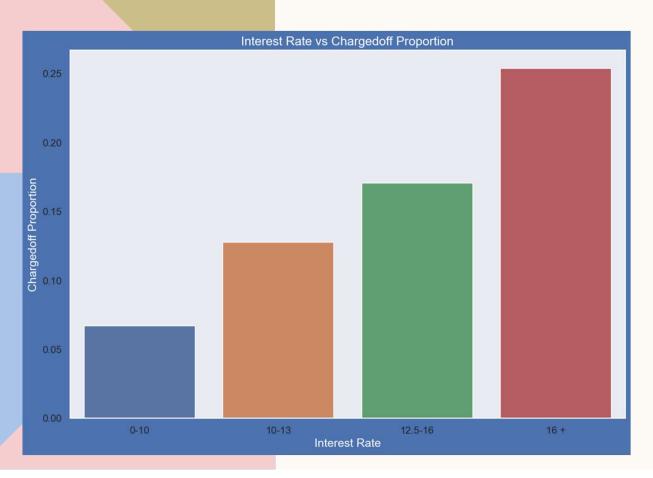
# ANNUAL INCOME VS LOAN PURPOSE



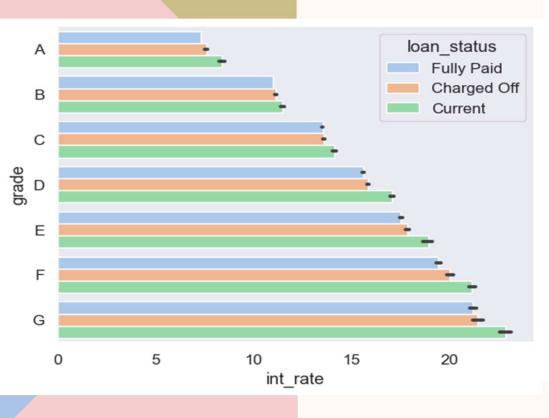
# LOAN PURPOSE VS DEFAULT PROPORTION



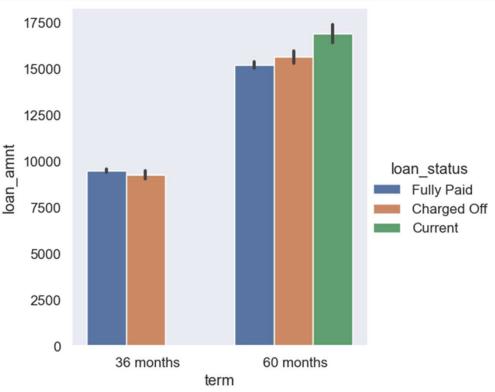
# ANNUAL INCOME VS CHARGED OFF PROPORTION



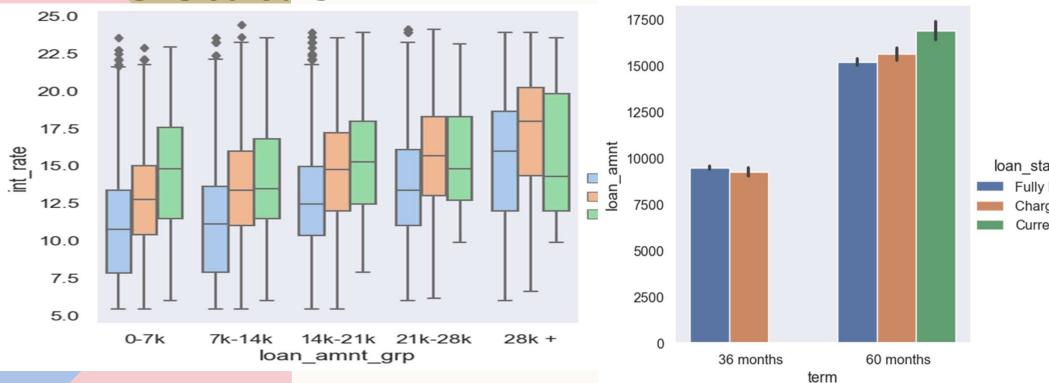
### GRADE VS LOAN STATUS



### **Duration vs Loan Status**



## Interest rate vs annual income vs status



### **OBSERVATIONS**

The number of loans defaulted are the highest in the loan pupose "debt\_consolation", But the annual income is not the highest.

The Loan is higher for all the income groups, people who defaulted.

The interest rate for charged off loans is higher, compared to that of fully paid loans in all the groups. This is a driving factor for loan defaulting.

The loan was sanctioned/issued in December are maximum defaulted and Loans issued in the year 2011 are also higher comparing other years

Employees with highest tenure in an organization got higher Loan amount.

Interest rate below10% are less charged off and these are starting from minimin 5 %.

Interest rate above 15% are highly charged off compared other interest rates.

More or higher intrest rates and higher the chances of default

### RECOMMENDATIONS

Modest interest for all types of loan would help in minimizing load default rate

Higher income group and lower income group, moderate loan amount would help in reducing defaulters

The number of loans defaulted are the highest in the loan pupose "debt\_consolation", number of loans, loan amount can be reduced

December month loans seems seasonal with high default rate, thus can pushed to a month before or later, gives time to employees think and apply for loans

Employees with highest tenure likely to repay the loan successfully



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