

LENDING CLUB CASE STUDY

(Using Exploratory Data Analysis)

Group:

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Agenda

1. Introduction
2. Problem Statement
3. Data Understanding
4. Data Cleaning and Manipulating
5. Data analysis
6. Presentation and Recommendations

Introduction

- ❑ 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).
- ❑ Main Intention of this assignment is to identify the risky loan applicants, then such loans can be reduced thereby cutting down the amount of credit loss
- ❑ We will be using the Exploratory Data analysis to identify the driving factors behind loan default and other factors which are strong indicators of default.

Problem Statement

The data given for assignment contains the information about past loan applicants and whether they ‘defaulted’ or not. Borrowers who default cause the largest amount of loss to the lenders. Main Intention of this assignment is to identify the risky loan applicants and identify the driving factors behind loan default and other factors which are strong indicators of default.

Data Understanding

Dataset has 39717 records with 111 different variable

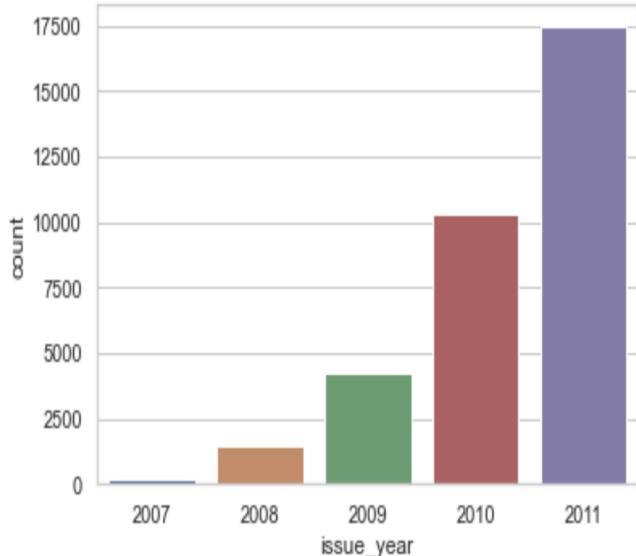
Data has been collected between 2007 to 2011

Dataset has other variables like, loan_ammount, intrest_rate, DTI, Purpose, Employment details and so on.

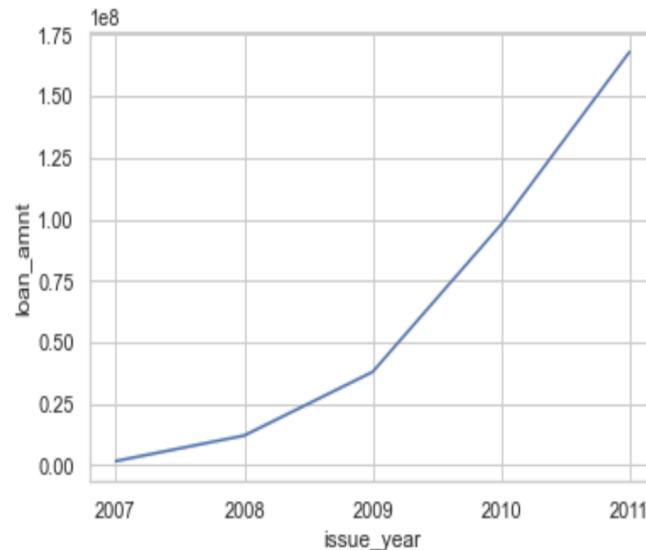
loan_status has been categorised into 3 different category ("Fully Paid", "Charged Off", "Current")

Data also has alot to unwanted Variables and 57 columns with >50% of Missing data

Growth of Number of Loans sanctioned



Growth of Loan amount Issued

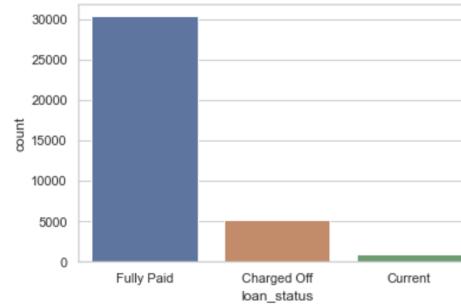
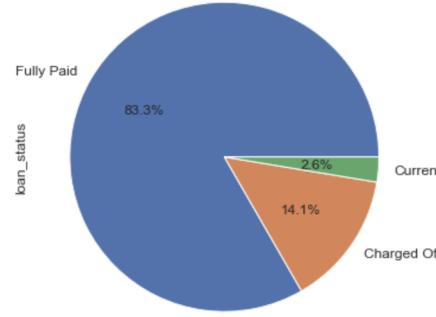
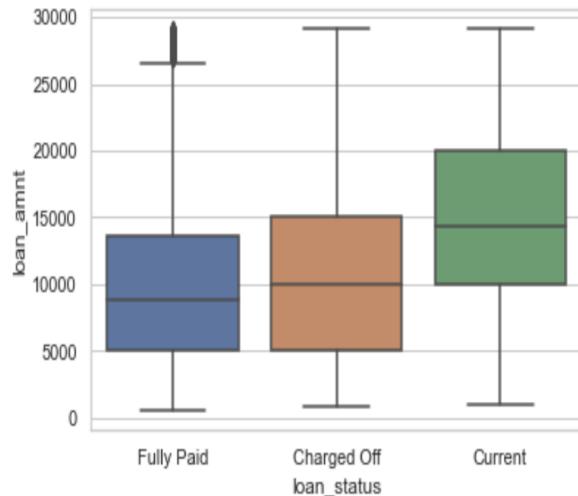


Observation:

Number of issued Loan has increased over past years

Loan amount also has significantly increased over the years

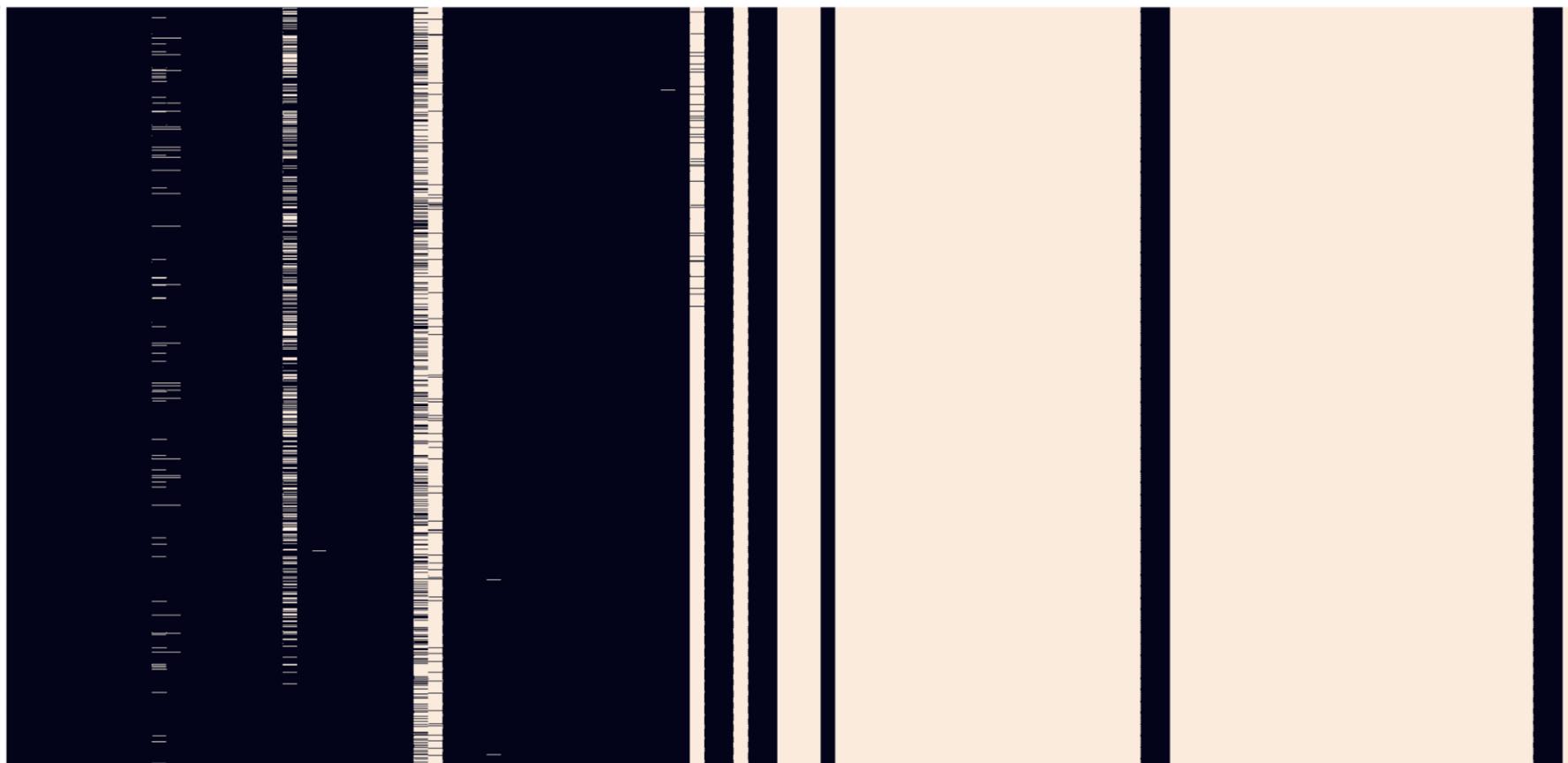
Loan Status by volume and percentage



Data Cleaning and Manipulating

- ❖ As part of data cleaning removed 57 columns with >50% of Missing data
- ❖ Removed unwanted columns by intuition
- ❖ Removed unwanted prefixes and suffixes in the data and converted to appropriate datatype
- ❖ Manipulated the columns which has object colums with string value to appropriate type (int, float)
- ❖ Filled missing values with median values
- ❖ identified and removed outliers

Heatmap to Represent the Missing Value



Data manipulate

- ❑ Removing strings 'years' and '<' and '+' from emp_length
- ❑ Remove '%' symbol from int_rate and revol_util
- ❑ Fill missing values with median value for revol_util
- ❑ Remove string ' months' from term

```
term int_rate grade sub_grade emp_length home_ownership \
0 36 months 10.65% B B2 10+ years RENT
1 60 months 15.27% C C4 < 1 year RENT
2 36 months 15.96% C C5 10+ years RENT
3 36 months 13.49% C C1 10+ years RENT
4 60 months 12.69% B B5 1 year RENT

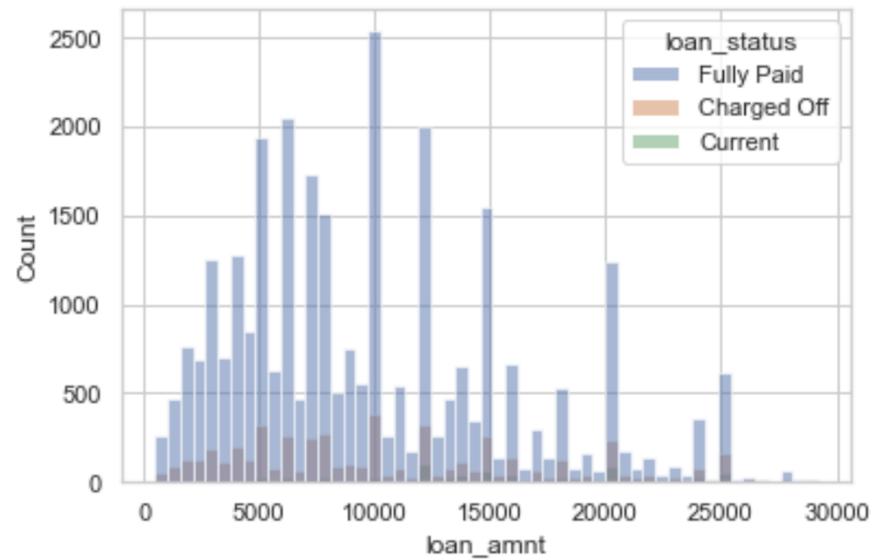
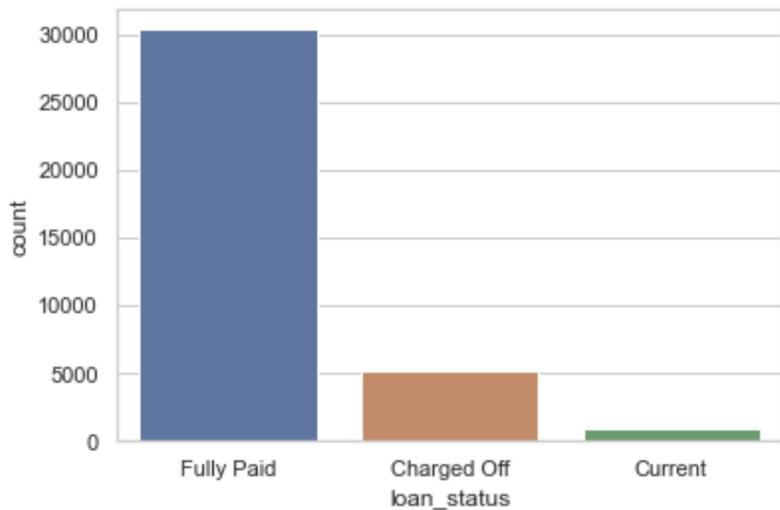
verification_status issue_d loan_status purpose addr_state \
0 Verified Dec-11 Fully Paid credit_card AZ
1 Source Verified Dec-11 Charged Off car GA
2 Not Verified Dec-11 Fully Paid small_business IL
3 Source Verified Dec-11 Fully Paid other CA
4 Source Verified Dec-11 Current other OR

revol_util
0 83.70%
1 9.40%
2 98.50%
3 21%
4 53.90%
```

After Cleaning Dataset

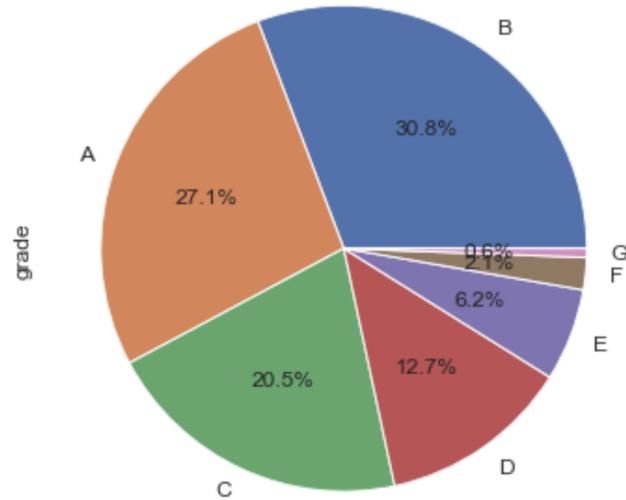
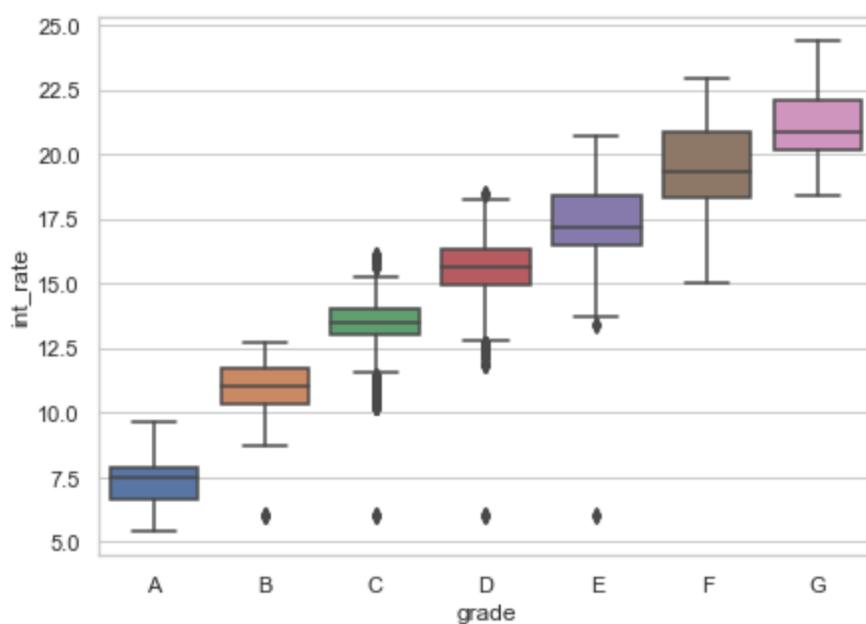
loan_amnt	0
funded_amnt_inv	1622
term	2433
int_rate	3244
installment	4065
grade	4866
sub_grade	5677
emp_length	6488
home_ownership	7299
annual_inc	8110
verification_status	8921
loan_status	9732
purpose	10543
addr_state	11354
dti	12165
revol_bal	12976
total_pymnt	13787
total_rec_prncp	14598
issue_month	15409
issue_year	16220
non_compliant	17031
inq_last_6mths	17842
revol_util	18653
total_pymnt_inv	19464
total_rec_prncp	20275
total_rec_ppd	21086
total_rev_hi_lim	21897
open_acc	22708
open_il	23519
open_il_12m	24330
open_il_6m	25141
open_il_3m	25952
open_il_6m	26763
open_il_3m	27574
open_il_12m	28385
open_il_36m	29196
open_il_6m	30007
open_il_3m	30818
open_il_12m	31629
open_il_36m	32440
open_il_6m	33251
open_il_3m	34062
open_il_12m	34873
open_il_36m	35684
open_il_6m	36495
open_il_3m	37306
open_il_12m	38117
open_il_36m	38928

Univariate Data Analysis of Loan_status and Loan_amount



Observation:
Majority of the loan are Fully paid.

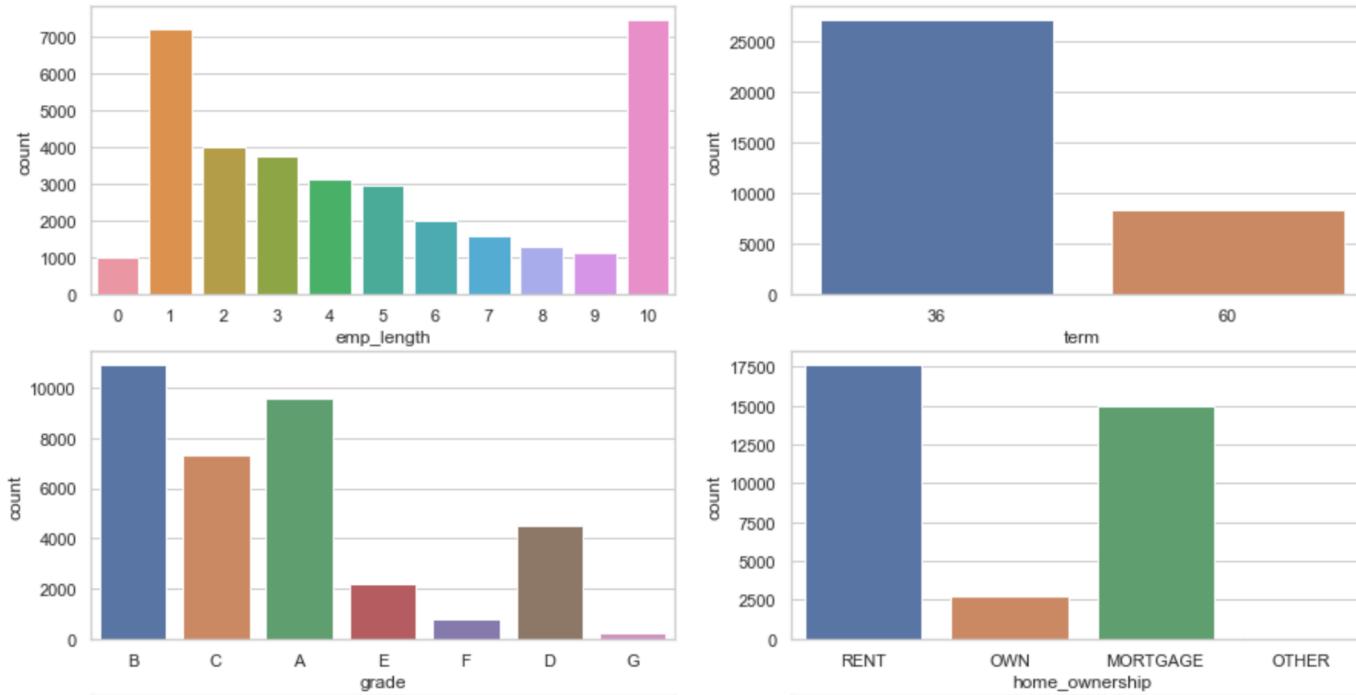
Grade categorised based on Interest rate and majority



Observation:

Grade is categorised into 7 grade from A-to-G based on the int_rate.. grade 'A' being lowest int_rate
Most of the loans are sanctioned with low grade A,B,C means lower interest rate.

UDA of Emp_length, Term, Grade, House_ownership



Observation:

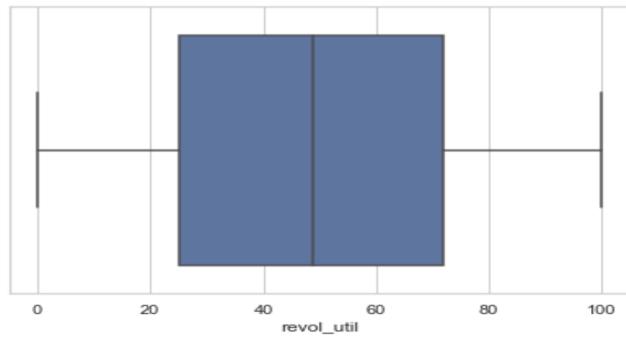
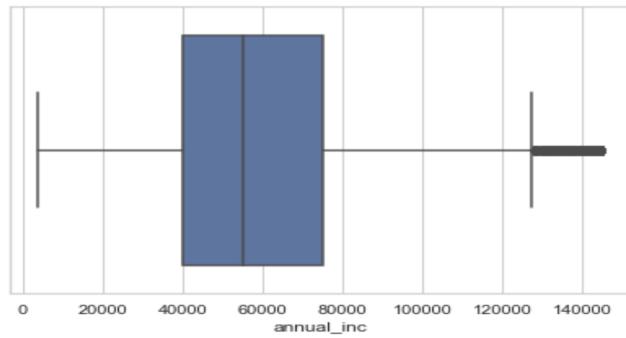
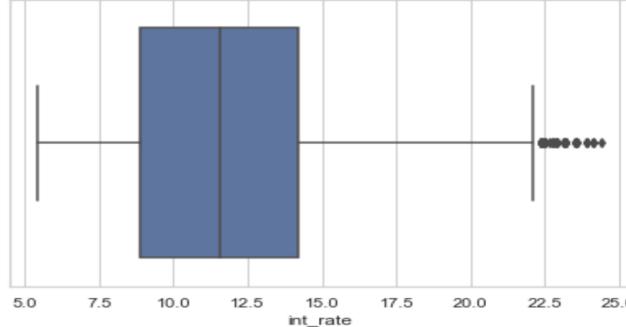
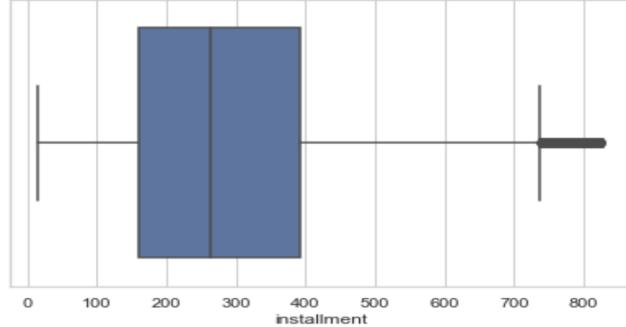
Majority of borrowers has more than 10+ years of experience or < 1 year of experience

Majority of borrowers has taken 36 months of term

Most of the loan are sanctioned for lower loan_grade (A being the lowest grade)

Most of borrowers either has Rented or Mortgaged

UDA of installment, int rate, annual income, revol util



Observation:

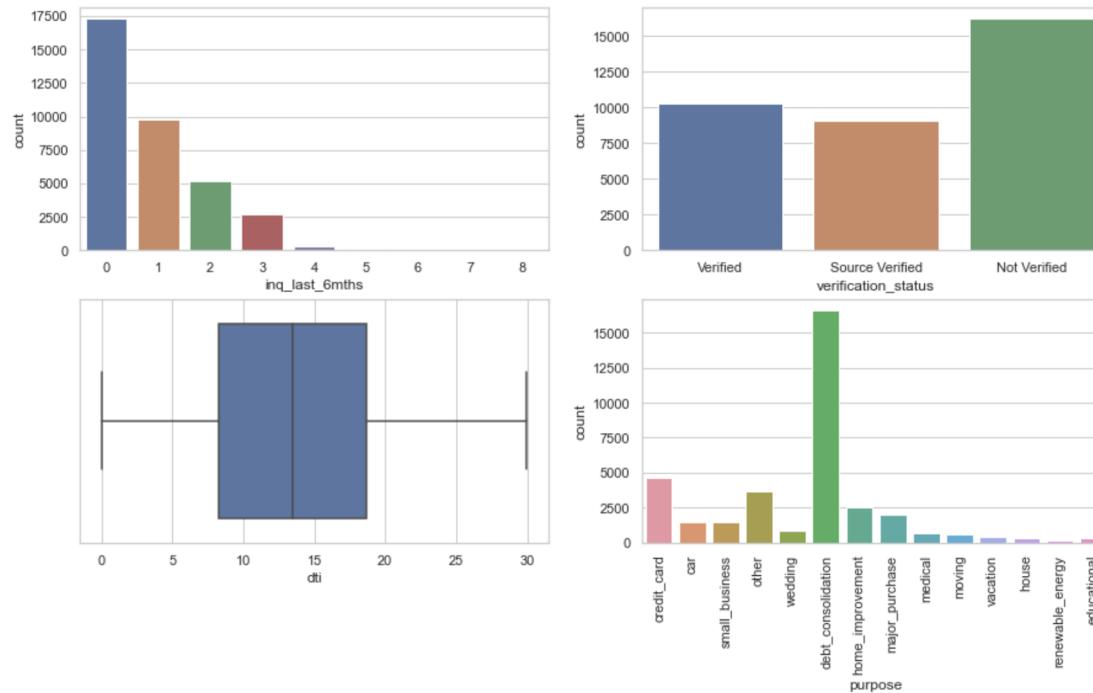
Installment has mean of 292.19

int_rate has mean of 11.75

annul_inc has mean of 60027

revol_util has mean of 48

UDA of inq_last_6months, Verification_status, dti, purpose



Observation:

Very few borrowers inquired in last 6 months

Majority of borrowers are not verified

dti has mean of 13.4

Most people borrowed the loan for debt_consolidation, credit_card, home_improvement

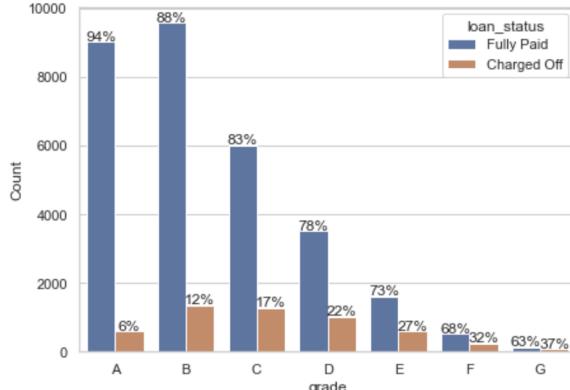
Bivariate Analysis

For understanding the relation between two variables we are calculating the percentage as well per category wise

Example: we have two category of loan_status "Fully_Paid" and "Charged_off" (we have dropped "Current" loan_status as they do not have significance)

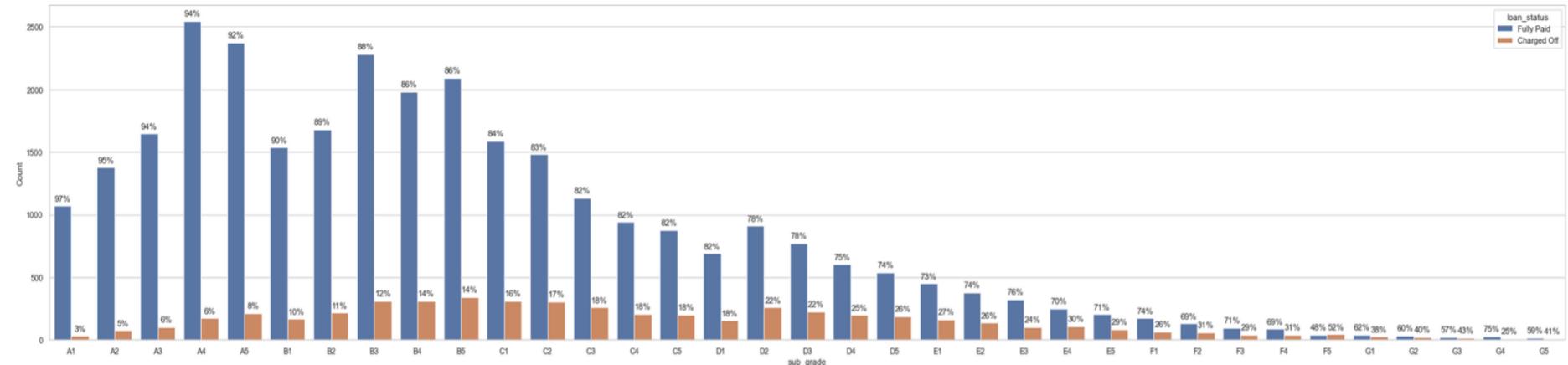
We also have 7 difference category of grade from A-Z, To understand the relation of each Grade with loan_status, we are computing the percentage of Paid and non-Paid for each Grade. This would help in understadning the behavior of people of every grade.

Grade, Subgrade relation to Loan_status

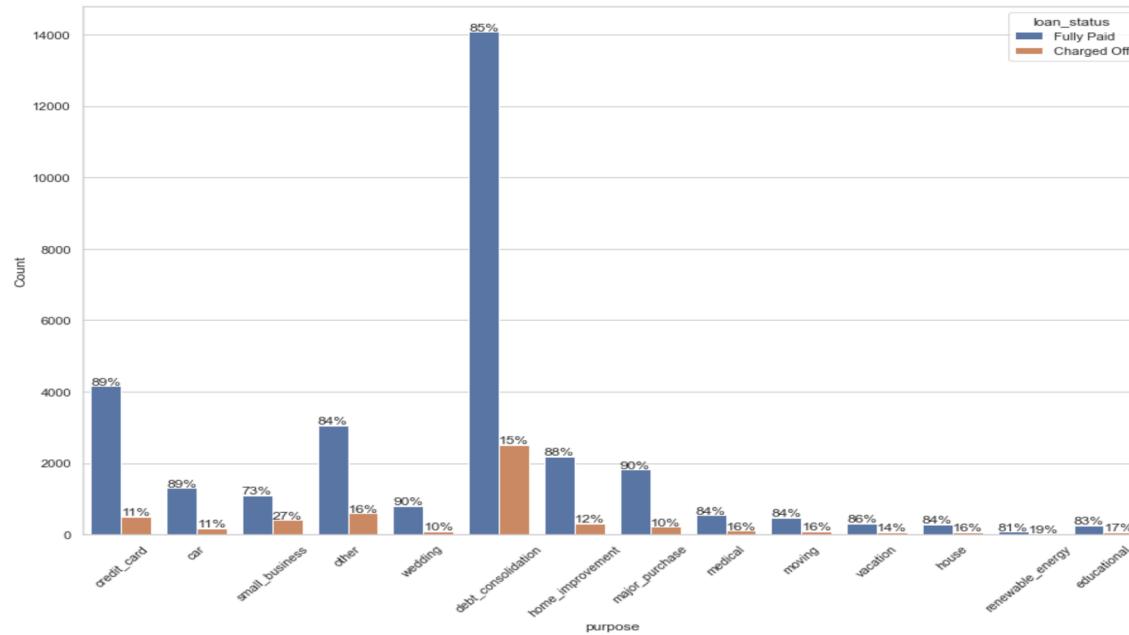


Observation:

Grade-wise percentage is calculated, as we can see in the countplot, Ratio or percentage of defaulter is increased with the higher grade.
(same observation is seen in subgrade as well)



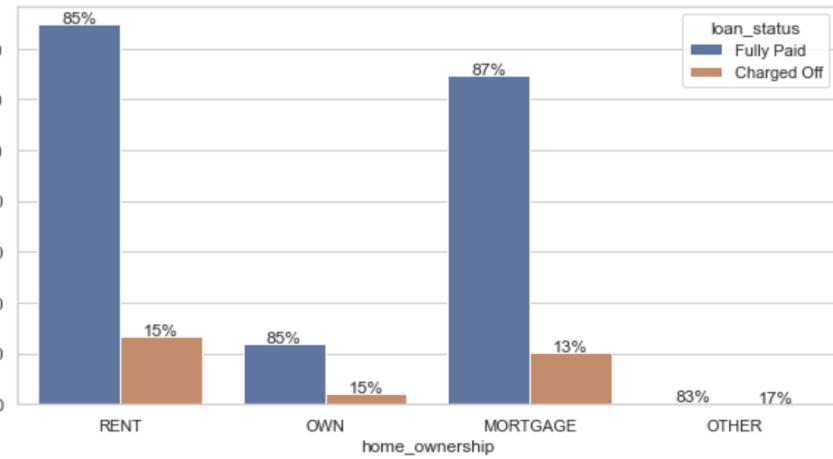
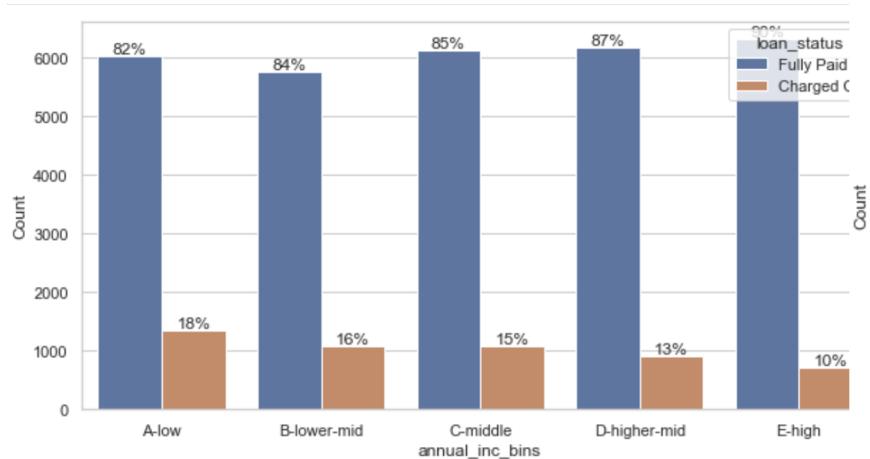
Relation of Purpose to Loan_Status



Observation:

Higher percentage of defaulter seen in "small_business (27%)", renewable_energy(19%), Educational(%17%)

Relation of Annual_income and House_ownership to Loan_status

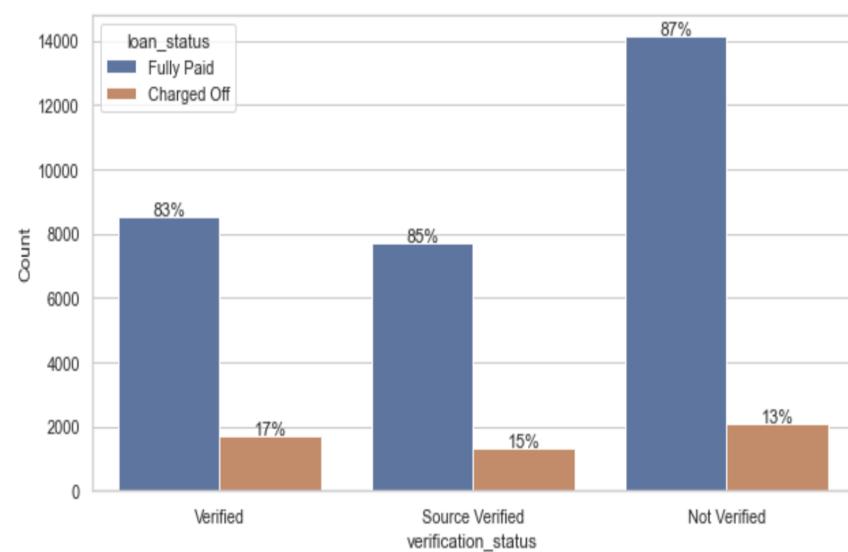
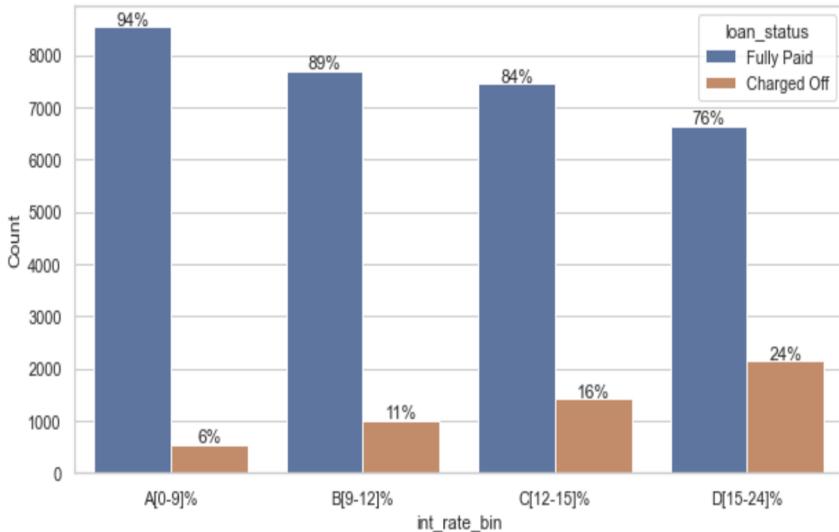


Observation:

Higher percentage of defaulter are more likely in the lower income. as the chart indicate the percentage of defaulter increases with lower income range

Higher percentage of re-paid customers are those who have mortgaged ownership

Relation of int_rate and Verification_status to Loan_Status

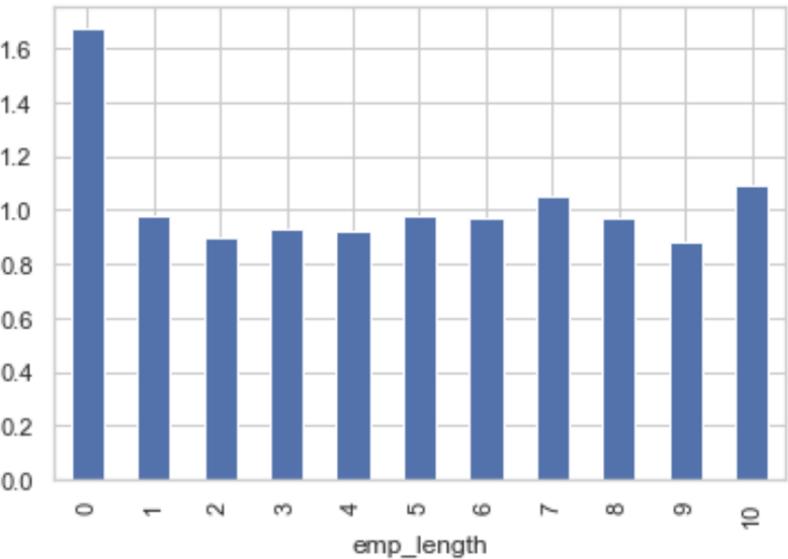
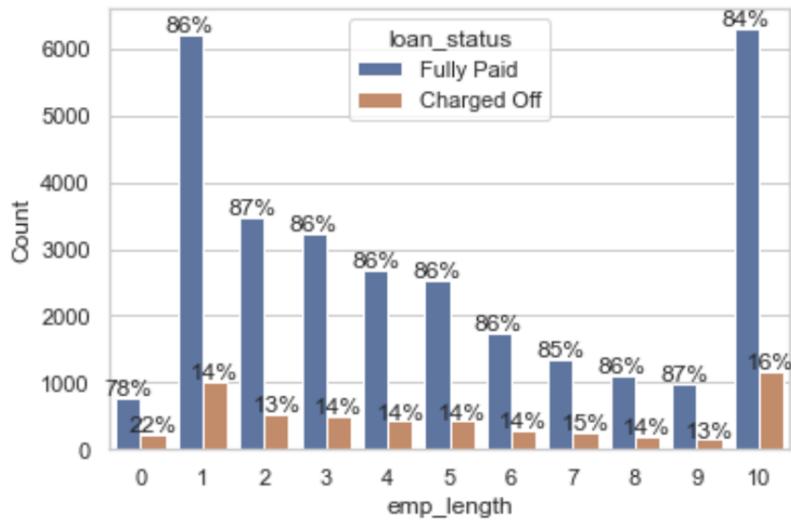


Observation:

The defaulter percentage is higher with high rate of interest

The defaulter percentage Verified borrowers

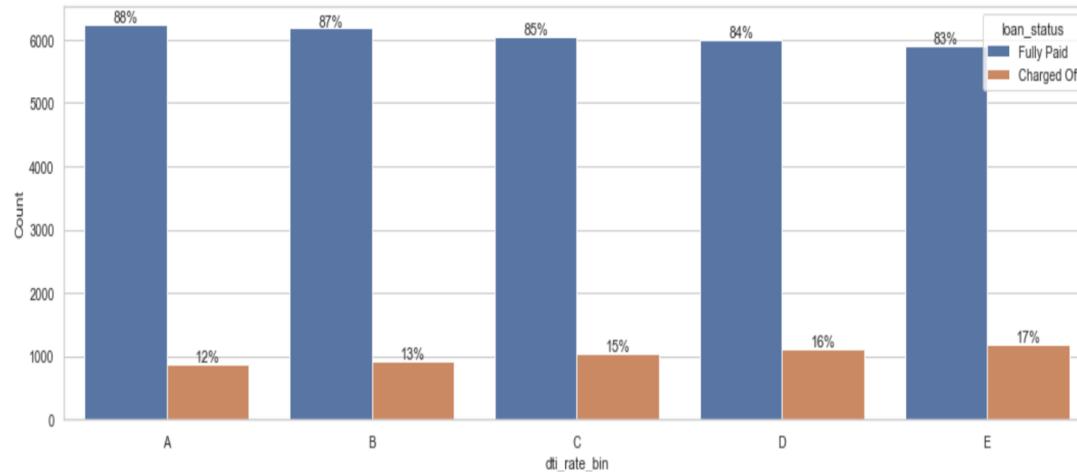
Relation of Emp_length to Loan_status



Observation:

Defaulter percentage is almost same across all Emp_length and does not indicate any specific trend
The plot on the right indicate the ratio to "Fully Paid" to "Charged Off", which indicate almost all has same ratio, There is not significant difference.

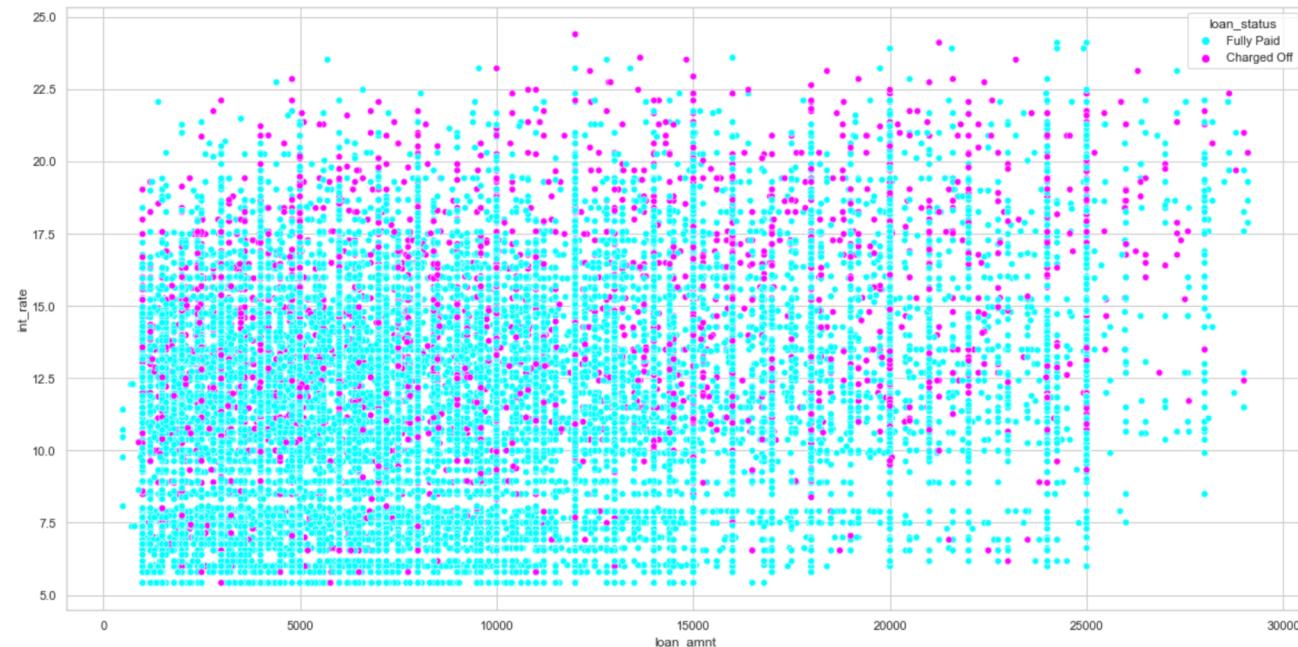
Relation of DTI to Loan_status



Observation:

Higher percentage of defaulter are more likely to have hight dti percentage, as the chart indicate the percetage of defautler increases with Higher dti percentage.

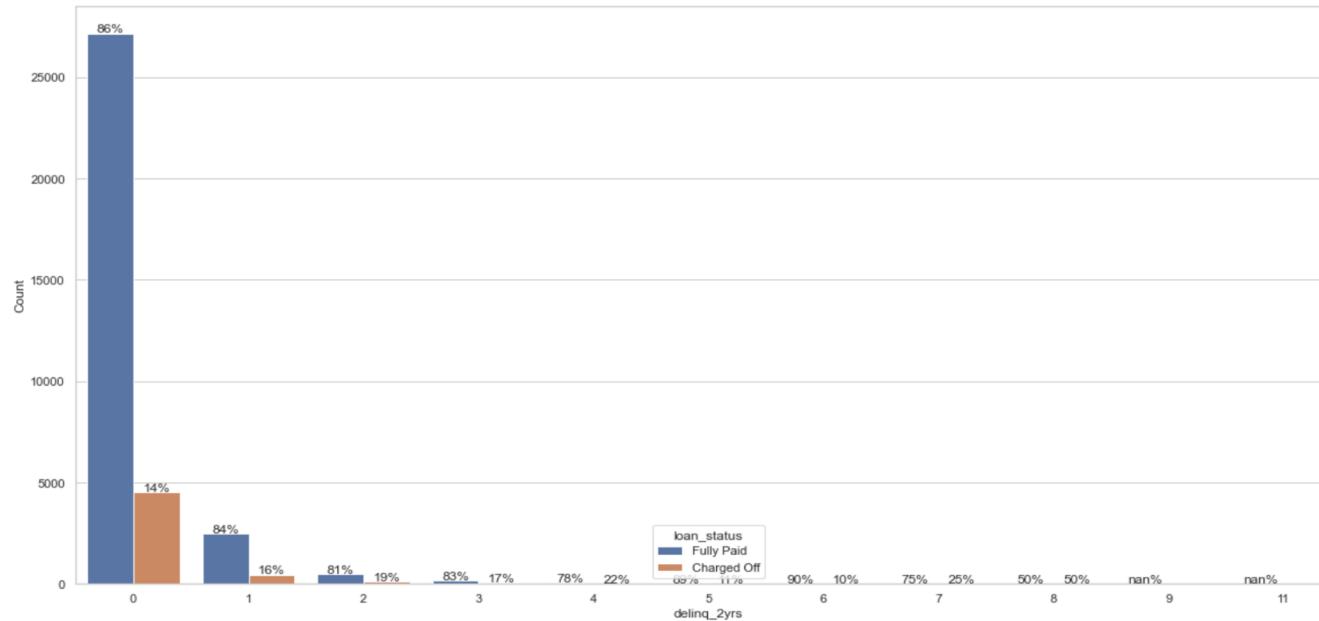
Distibution of of Loan_ammount and int_rate w.r.t Loan_status



Observation:

The Majority of defaulter is seen with high rate of interest

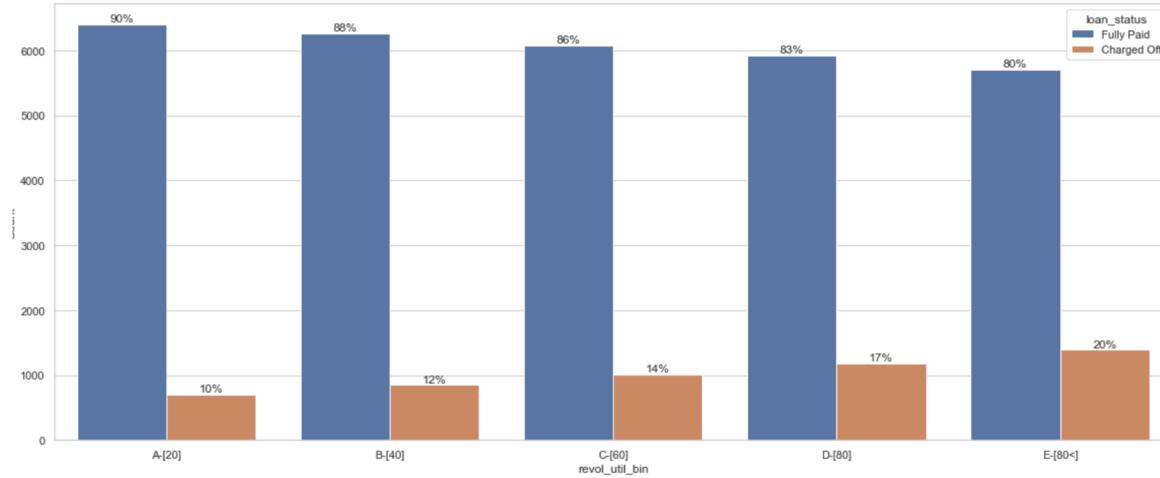
Relation of Delinq_2yrs to Loan_Status



Observation:

Percentage of defaulter increasing with value of delinq_2yrs increase

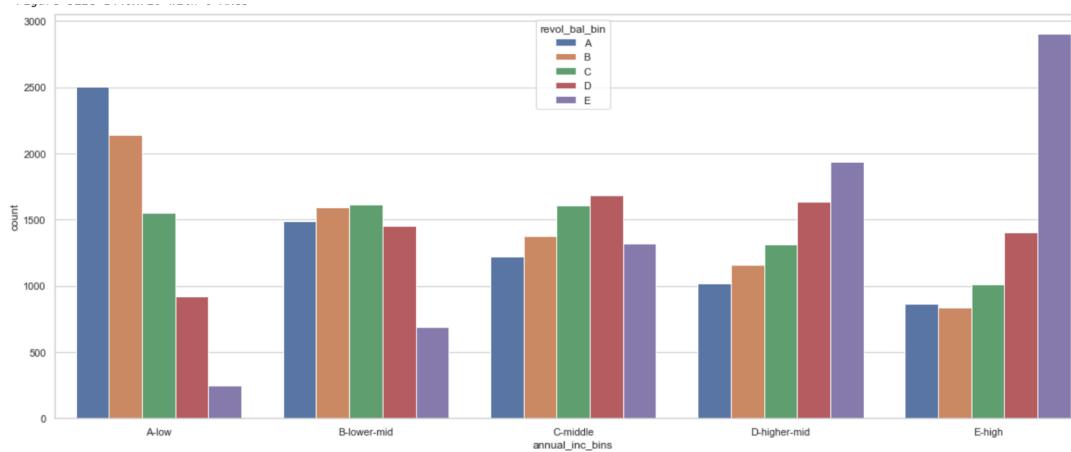
Relation of Revol_util to Loan_status



Observation:

Percentage of defaulter increasing with higher revol_util value

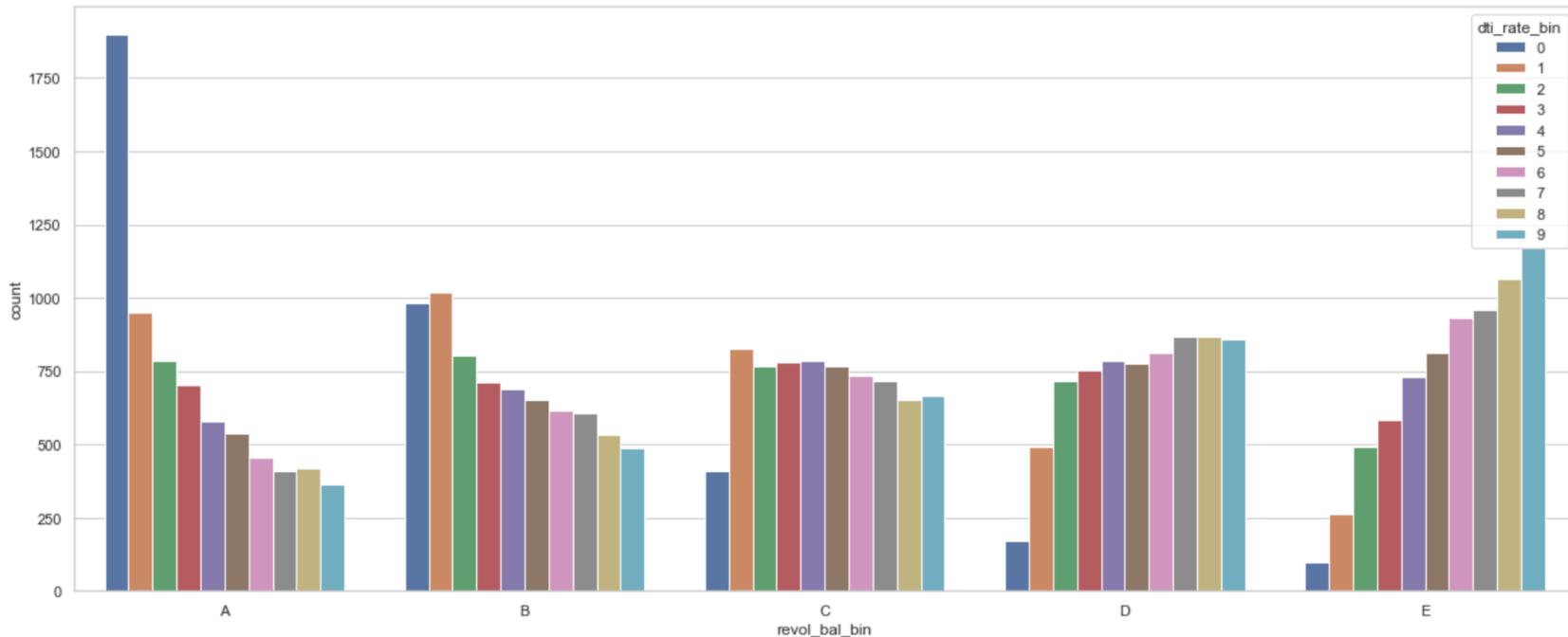
Multivariate Analysis: Relation between Revol_bal, Annual_income and loan_status



Observation:

the revol_bal is increasing as the annual_income is increased

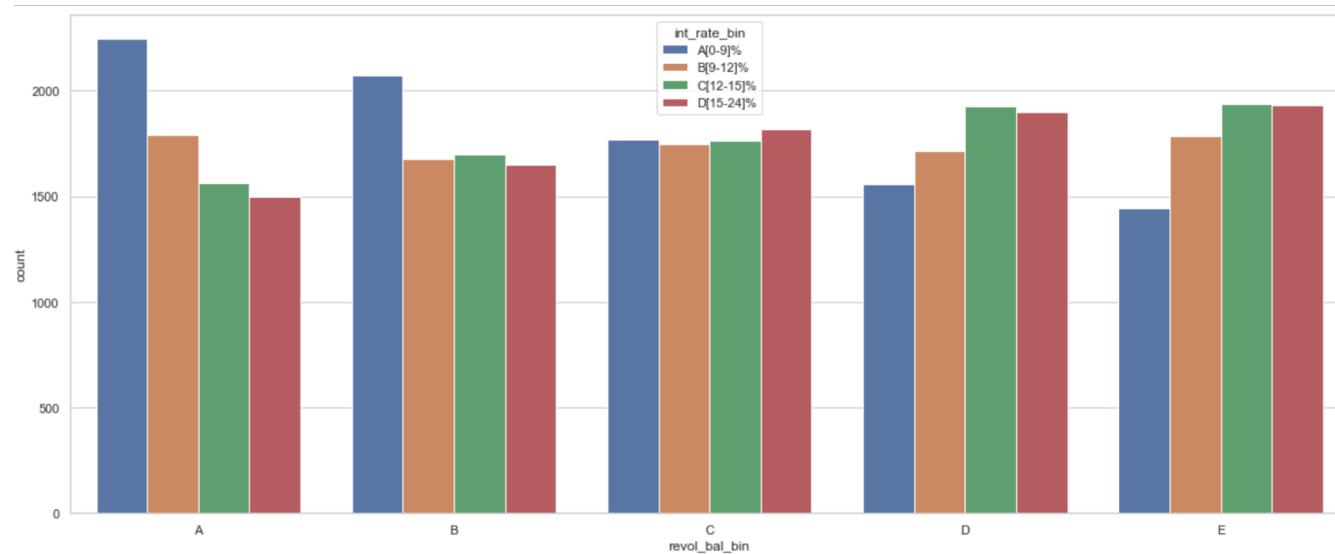
Multivariate Analysis: Relation between dti, revol_bal and loan_status



Observation:

the `revol_bal` is increases as the `dti` value is increased

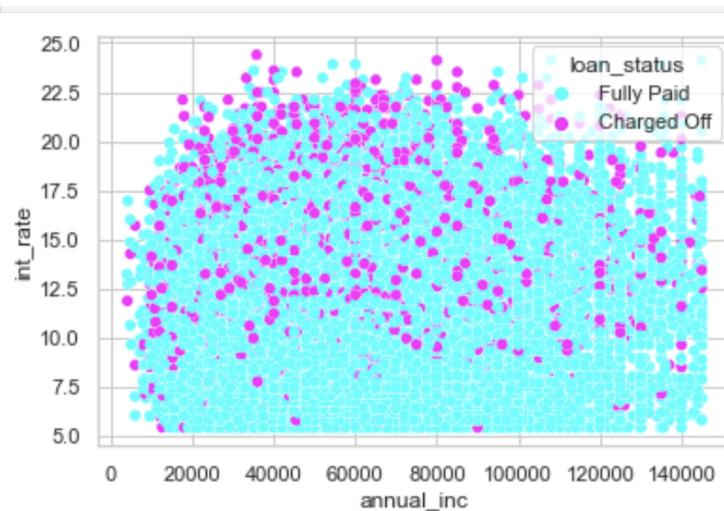
Multivariate Analysis: Relation between dti, revol_bal and loan_status



Observation:

the revol_bal is increases as the Int_rate is increased

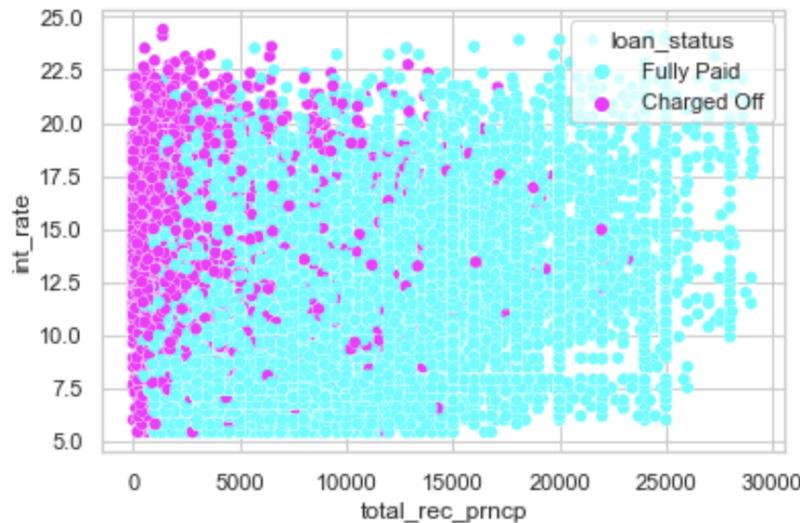
Multivariate Analysis: Relation between dti, revol_bal and loan_status



Observation:

Most defaulter as seen on Higher interest_rate

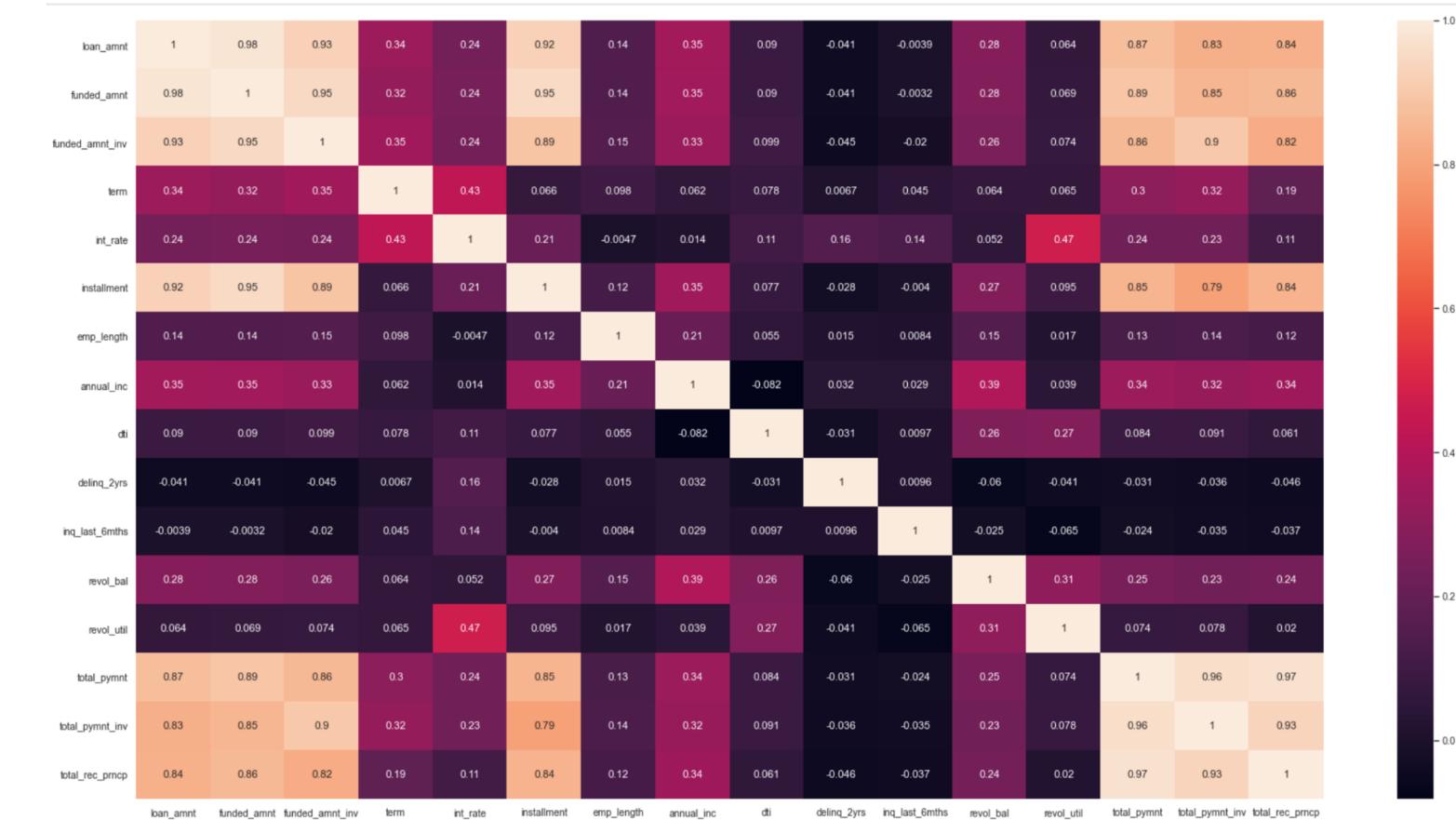
Multivariate Analysis: Relation between dti, revol_bal and loan_status



Observation:

Most defaulter as seen on Higher intrest_rate and Most defaulter are defaulting in the initial repayment itself (higher number of people who have defaulted paid 5000 or less)

Correlation matrix for Multivariate Analysis



Outcome of Correlation matrix analysis

High correlation is observed with the following variables:

Annual_inc (very good indicator for loan eligibility):

positive correlation with:

- loan_amount, total_paymt, total_rec_prncp, installment variable

negative correlation with:

- int_rate, revol_bal, revol_util, dti

int_rate has positive correlation with term, revol_bal, revol_util

Recommendation

- Most people borrowed loan amount between range of 5000-15000
- People with high DTI and revol_bal are most likely to default. Not recommended to approve the loan for those who has DTI high
- People with low annual income are also more likely to default. also Majority of loan borrower are from lower income category, not approving the loan will also be loss for the company, recommended to consider other parameter as well revol_util and DTI for lower income borrower.
- People borrow loan for the following reason,
- > debt_consolidation, small_business , renewable_energy Educational
- recommended to approve loan If annual_income is more than average or (higher-mid, high)
- Observed Defaulter are high with Higher rate of interest across all income range. Recommended to approve the loan with high interest rate for people who belongs to =annual_income category higher-mid/high