

Lending Club Case Study

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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.
- The company wants to understand the **driving factors** (or **driver variables**) behind loan default, i.e. the variables which are strong indicators of default. The company can utilise this knowledge for its portfolio and risk assessment.





- 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:
 - 1. If the applicant is likely to **repay the loan**, then not approving the loan results in a **loss of business** to the company
 - 2. 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**

Decision Factor



- When a person applies for a loan, there are two types of decisions that could be taken by the company:
 - **1. Loan accepted**: If the company approves the loan, there are 3 possible scenarios described below:
 - **1. Fully paid**: Applicant has fully paid the loan (the principal and the interest rate)
 - **2. Curren**t: Applicant is in the process of paying the instalments, i.e. the tenure of the loan is not yet completed. These candidates are not labelled as 'defaulted'.
 - **3. Charged-off**: Applicant has not paid the instalments in due time for a long period of time, i.e. he/she has defaulted on the loan
 - 2. Loan rejected: The company had rejected the loan (because the candidate does not meet their requirements etc.).





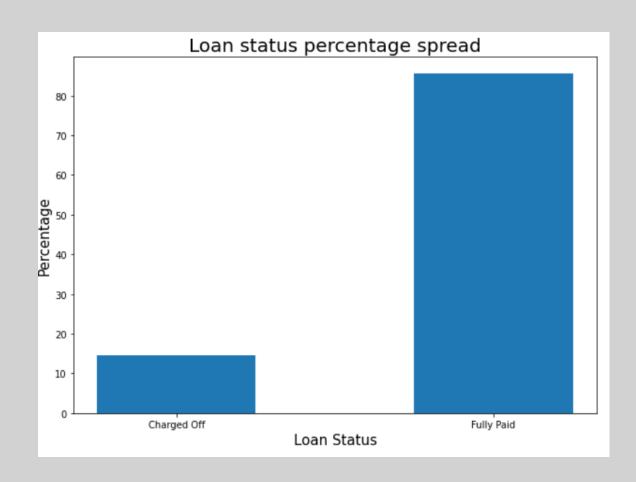
• Remove null values • Fix rows and columns **Data Cleaning** • Filter data • Understanding the impact using single variable Univariate Analysis • Understanding the impact using the double variables **Bivariate Analysis** • Final conclusion based on the previous analysis Conclusion/Insights



Defaulter vs Non-defaulter Percentage

Out of the total Loan application

- Charged Off(Defaulter) Loan percentage ~ 15%
- Fully Paid Loan percentage ~ 85%

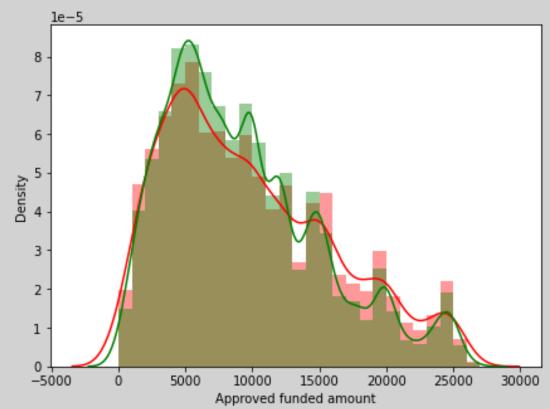




Analysis based on Approved Loan Amount

- People with approved loan amount ranging from 4k
 to 12k has high chance of repaying the loan.
- People with approved Loan amount beyond 12k, has high chance of defaulting

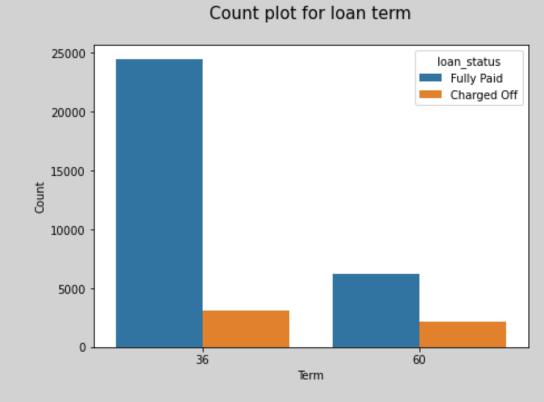
Approved funded amount data spread





Loan Term relationship with Defaulter

- More people opt for 36 months term than 60 months term
- People opting loan for 36 months has high chance of re-paying up the loan as against the one opting loan for 60 months

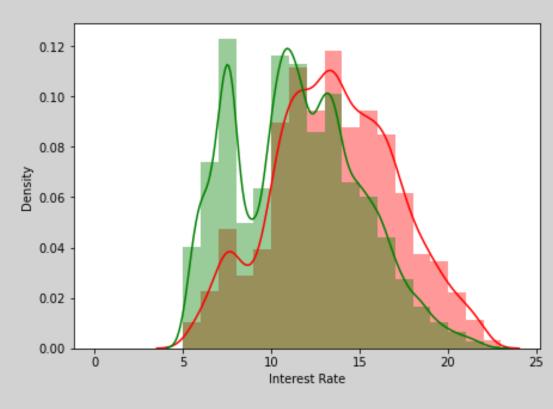


Distribution of Interest rate



- People with interest rate less than 12 has high chance of re-paying the loan
- However, people going for high interest rate (More than
 12) has high chance of defaulting

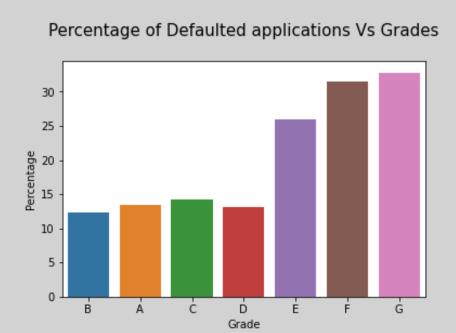
Interest Rate data spread

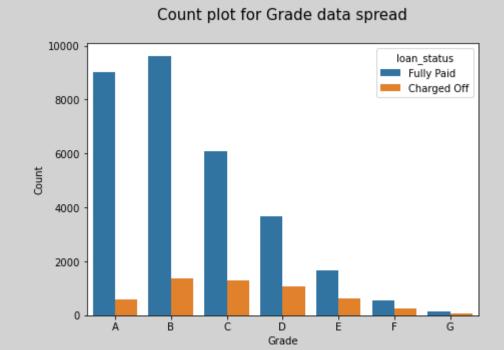






- Grade A, B, C has high number of applications
- At a high level Ic Grade B, C, D has high number of defaulters.
- However, the percentage of getting defaulting is high when the application moving from D to G



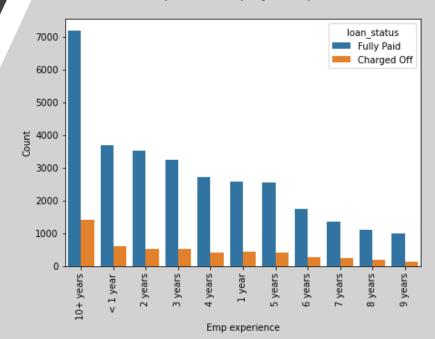




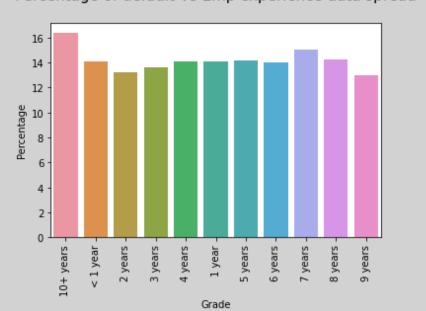
Employment Experience Analysis

- We are getting more applications with 10+ years of experience and they also have high percentage of loan default
- With experience of 1, 4, 5, 6 years have almost similar percentage of defaulting

Count plot for employee experience



Percentage of default Vs Emp experience data spread

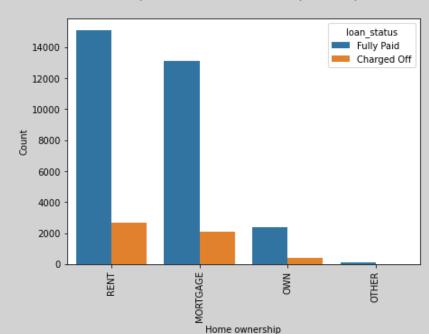




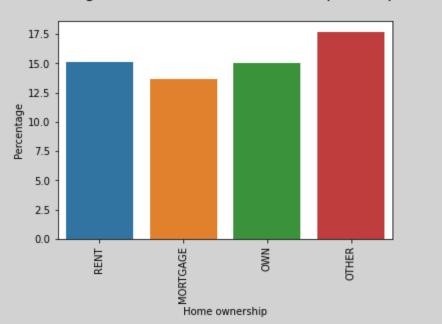


- People staying in rented house or mortgage has higher percentage of loan application
- However, people with OTHER category has the highest defaulter percentage, followed by RENT, OWN and MORTGAGE

Count plot for Home ownership data spread



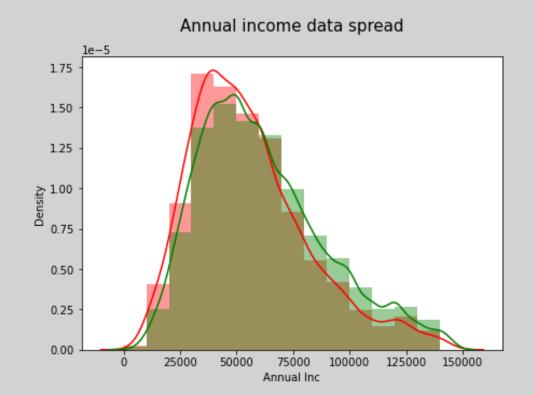
Percentage of default Vs Home ownership data spread



Annual Income Analysis



• People with annual income between 30k to 50k has high chance of getting default than the people with income more than 50k

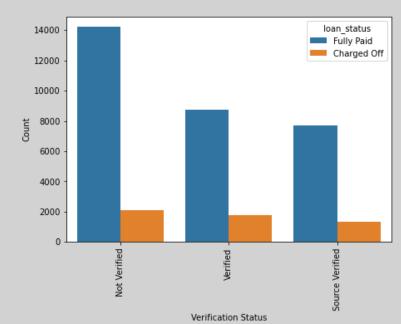




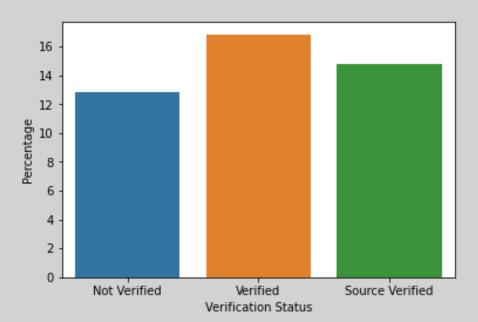


- People with Not verified income status has higher number of loan applications
- On the other hand, people with Verified income status has more chance of being defaulter

Count plot for Verification status data spread



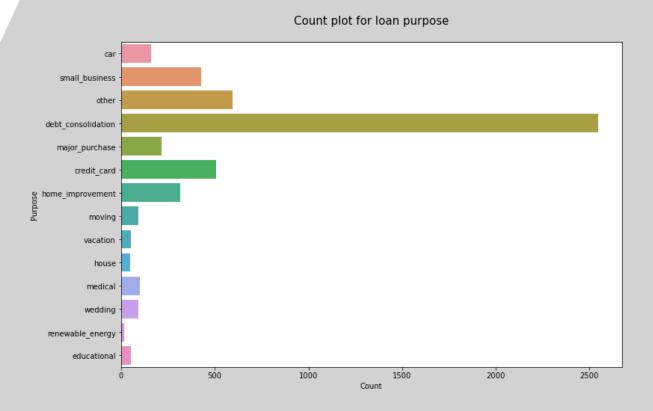
Percentage of Defaulted applications Vs Verification Status

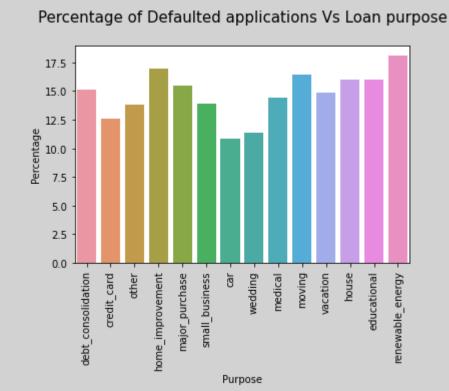


Loan Purpose Analysis



- Highest number of borrower took loan for the purpose of "debt_consolidation" and they have the highest number of defaulters
- However, the defaulter chances are more for the loan taken for "renewable_energy" and "home_improvement"

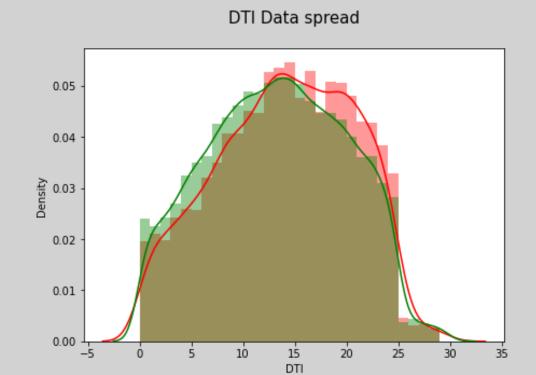








- Borrower with dti less than 12 has high chance of re-paying the loan
- Borrower with more dti (>12) has high change of defaulting

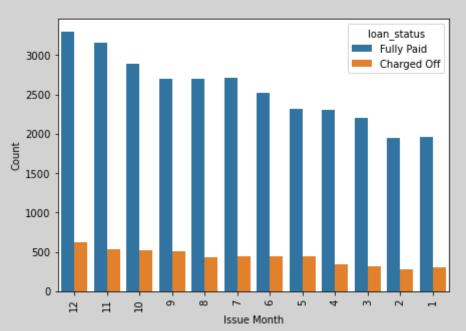






- •There is a slight upward trend in the number of applications from Jan to Dec.
- •December has highest number of applications and high number of defaulters.

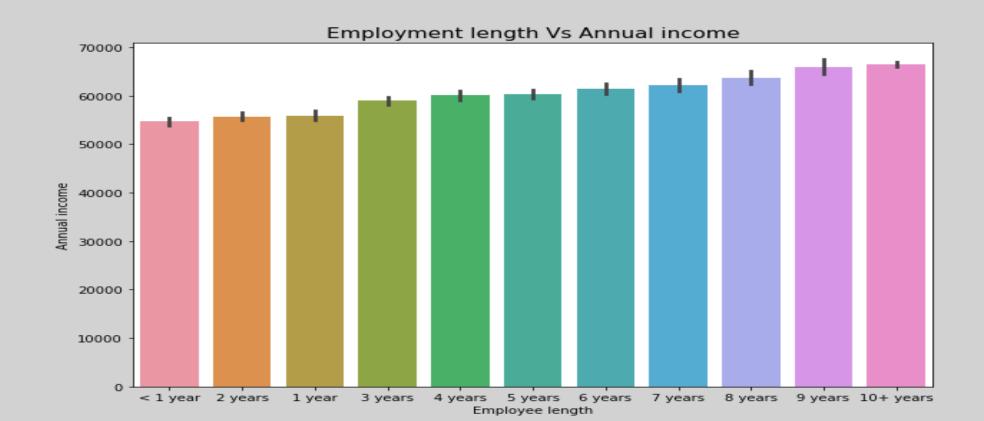
Count plot for Issue month





Emp length Vs Annual income (Bi-variate)

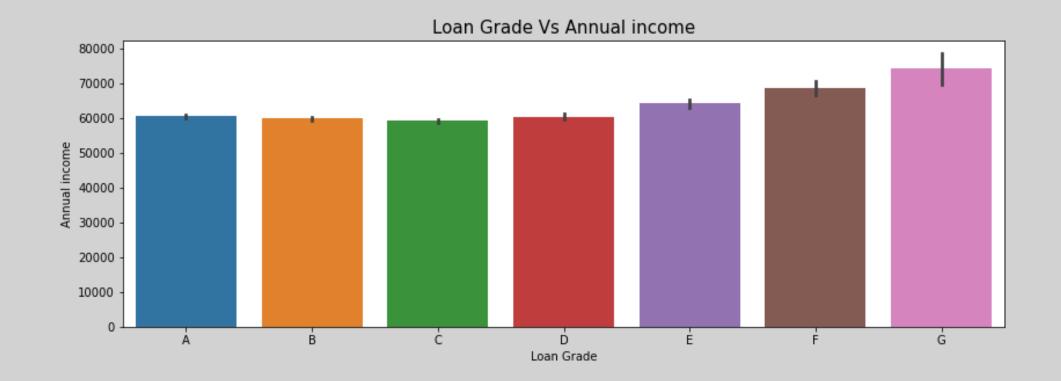
• As the experience is increasing the average annual income is also increasing, but at a slow rate



Grade Vs Annual income (Bi-variate)



•As the grade moves from A to G, annual income(average) is increasing.

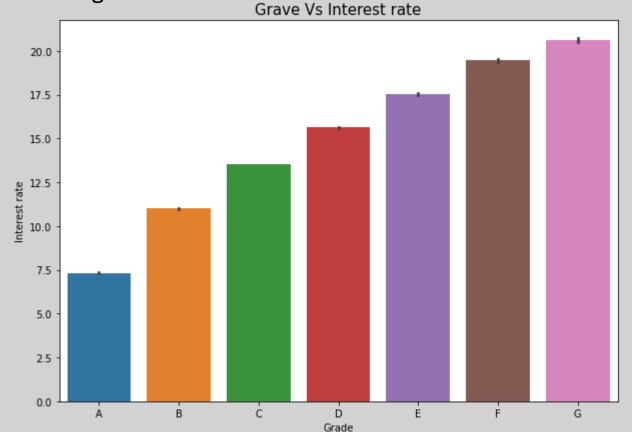




upGrad

Grade Vs interest rate (Bi-variate)

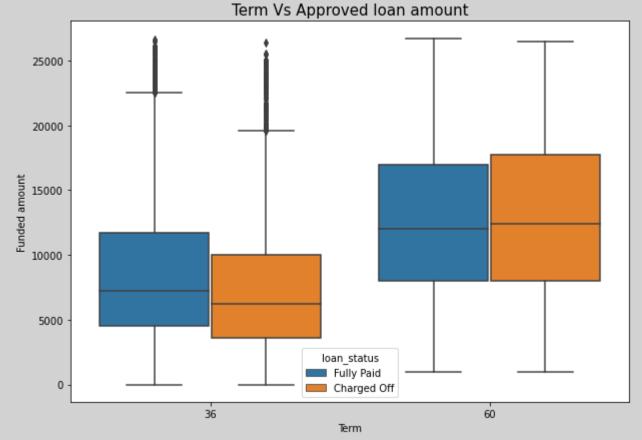
- LC Grade F & G are attracting high interest rates
- Grade A has low interest rates among all the loan categories







- Borrower with higher loan approved amount will go for higher loan term.
- More Borrower with funded approved amount are opting for 36 months as term and they are re-paying the loan as well

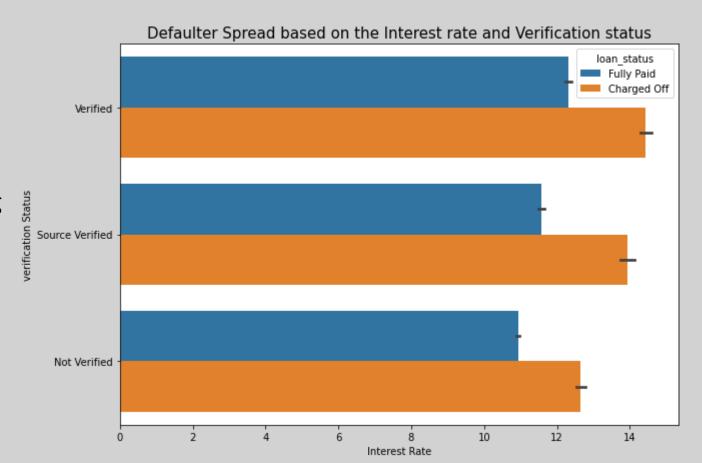




upGrad

Interest rate Vs Verification status (Bi-variate

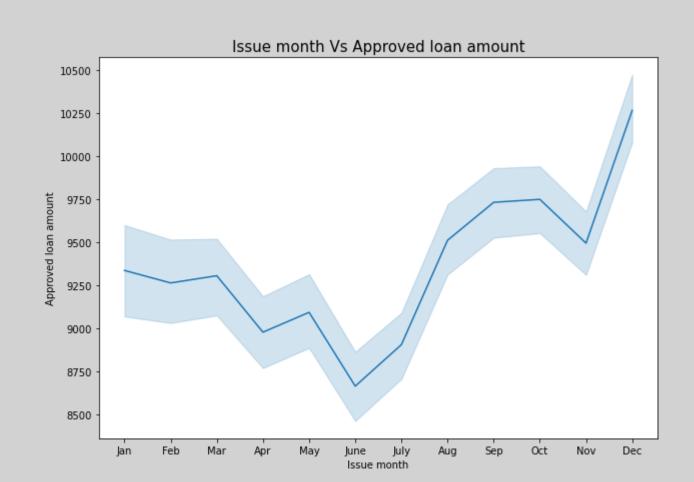
- If borrower with income verified is going for high interest rate (Average), chances of getting defaulted is high
- In all categories, if the borrower is going with low interest rates, chances of re-paying the loan is high





Issue month Vs funded amount (Bi-variate)

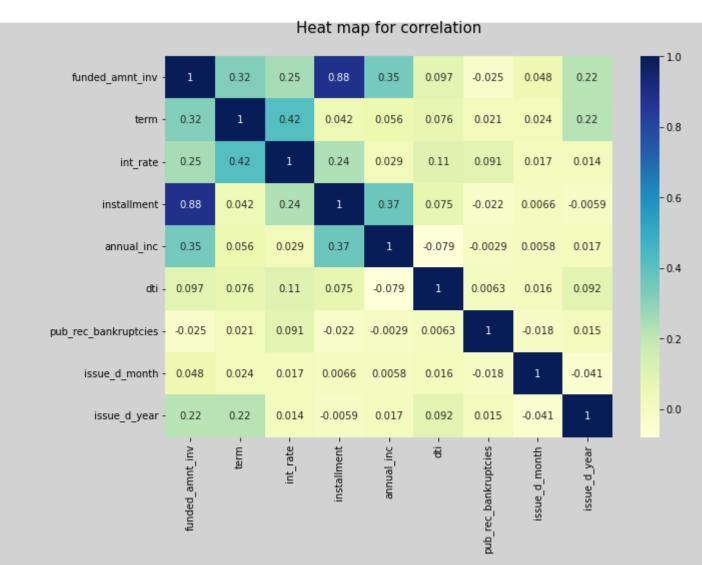
- •The average approved loan amount is very less in middle of the year (June month).
- •From June, again we can see an upward trend with another dip in Nov.
- •December has the highest approved loan amount (Average).



Correlation Matrix



- Approved funded amount has a positive correlation with installments (0.88). As the funded amount increases, installment increases.
- Interest rate to term has positive correlation(0.42). As the loan term increases, interest rate increases



Conclusions



Driving factor (Variables) in determining whether the borrower can be defaulter are as below:

- Term(**term**) 60 months (Categorical Variable)
- Loan amount(funded_amnt_inv) > 12k (Continuous Variable)
- Interest rate(int_rate) > 12% (Continuous Variable)
- LC Grade(grade) D,E,G (Categorical Variable)
- Employment Length(emp_length) >= 10 years (Continuous Variable)
- Home Ownership (home_ownership) is OTHER (Categorical Variable)
- Annual Income (annual_inc) between 30K to 50K (Continuous Variable)
- Verification status(verification_status) is Verified (Categorical Variable)
- Purpose of Loan(purpose) is for "renewable energy" of "home improvement" (Categorical Variable)
- DTI(dti) > 12 (Continuous Variable)
- Loan disbursed/approved in December (Derived Variable)
- Borrower with verified income and high interest rate



Thank You

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