



LENDING CLUB CASE STUDY SUBMISSION

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Business Understanding

As Lending Club specialises in lending various loans to urban customers.

As an financial organization, Lending club want to profile application to identify below factor:

- What is the risk of Applicant not paying loan.
- Which application should be accepted or rejected.
- What are the factor drives the applicant to be a defaulter.
- How organization can minimize the risk if an applicant turns out to be a defaulter.
- How organization can curb the Financial loss and Business loss





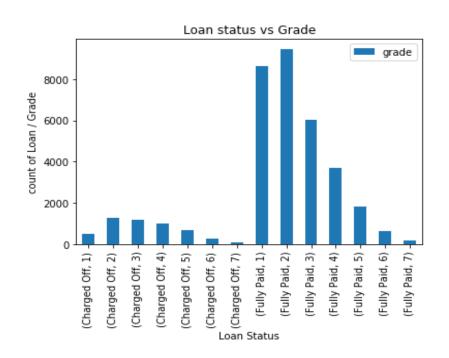
Analysis Strategy

Data **Bivariate Observations** Collect the Analysis of data on Cleaning **Analysis** • Defining a correlation Data and the basis categories between the variables • Depict the • Analysis of data on the understand and variables • Analyzing two or and their observation on basis categories and its Metadata more variables interdependencies Segmented the basis of the variables against a **Data Collection and** Derived Univariate analysis categorical Data Metrics **Understanding Analysis**



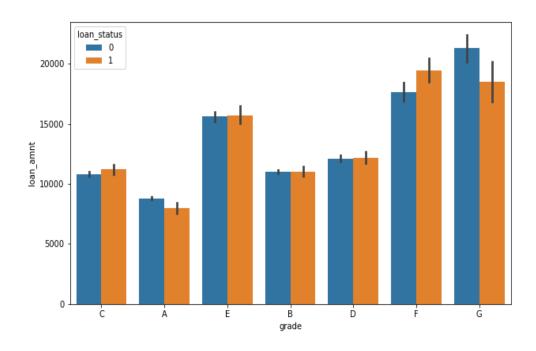


Data Graphs Loan Status Vs Grade





- We get number of the fully paid loans are from Grade rank 2 (Grade B)
- We get the number of charged off loans with respect to grade of applicants.
- Based on the grade we can identify whether an applicant will be a defaulter or not.

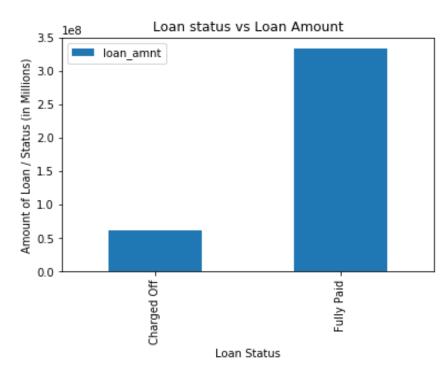


- The above graph shows the trend which grade has larger share of loan amount.
- The above graph also provide comparison of amount for defaulter and fully paid applicants.
- It can be clearly seen that Applicant from Grade F & G are higher in defaulter category.



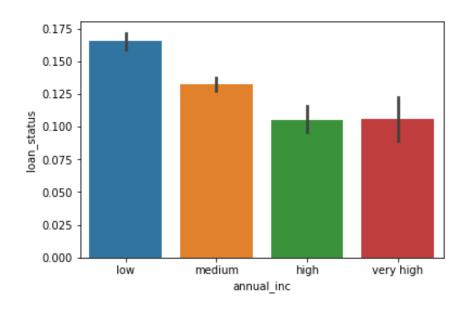


Data Graphs Loan Status





- We get the amount (in dollars) that are charged off by company
- The charged off loan amount is ~0.5Milion which is 14% of the total amount of loan provide by company.



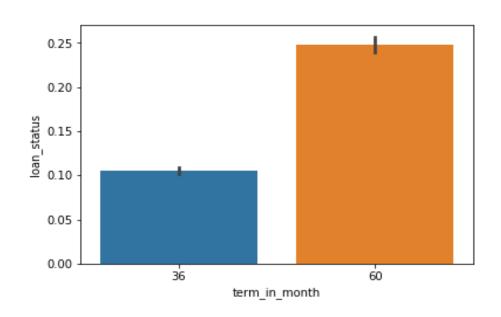
loan status 0-Fully Paid 1-Charged Off

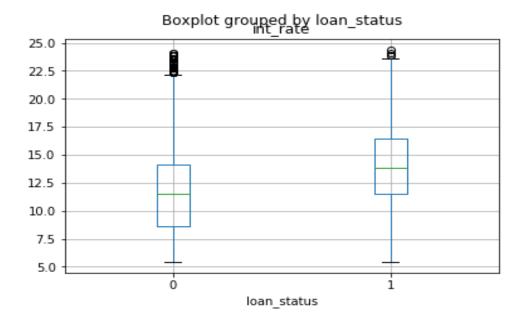
- Above graph shows comparison between the Annual income and Loan status.
- We can clearly see that lower the annual income higher the chance for an applicant to be a defaulter.





Data Graphs Tenure and Interest Rate





Inferences

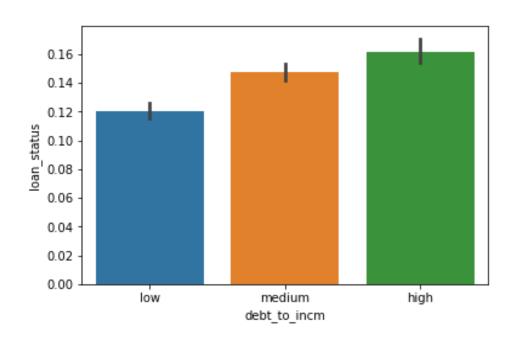
- Above graphs shows the comparison between the tenure of loan and status of loan.
- We can clearly infer that higher the tenure for loan, higher the chances of applicant to be defaulter.

- Above graph shows comparison between rate of interest and Loan status.
- It can seen here that company have already started a good practice of providing loan at higher interest rate to defaulter, in order to reduce the financial loss.



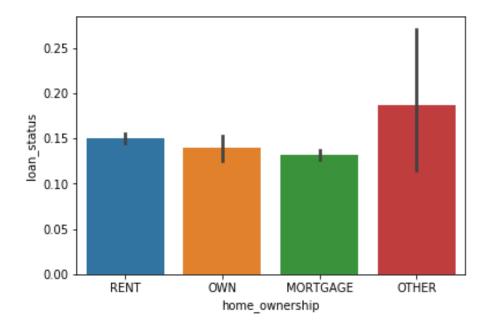


Data Graphs DTI and Home Ownership





- Above graphs shows the comparison between debt to income and status of loan.
- We can clearly infer that chances of loan applicant with higher debt to income ration turning defaulter are high.

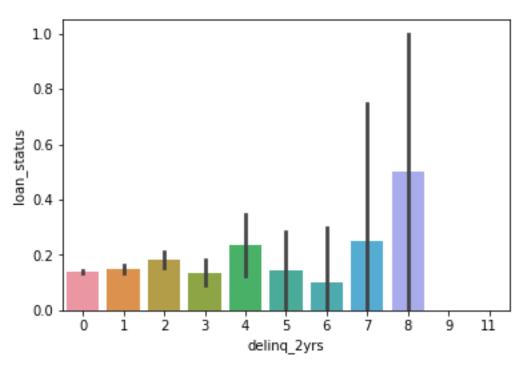


- Above graph shows comparison between home ownership and Loan status.
- It can observe Applicant having rented or other type of ownership having more chances of turning to be as defaulter.



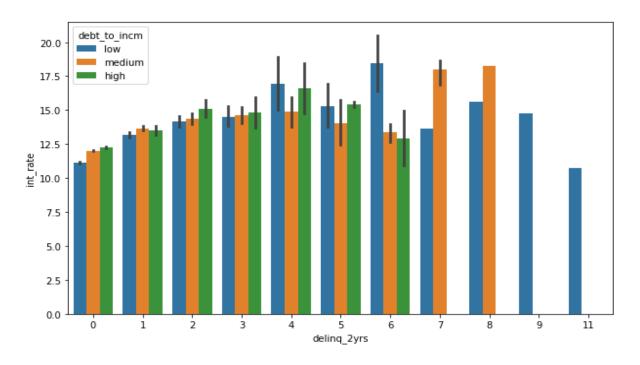


Data Graphs Deling 2 Years





- Above graphs shows the comparison between deling in 2 year and status of loan.
- Delinq in 2 years provides the information of how many installment missed by an Applicant and above graphs shows the trend with higher value for delinq 2 years value are more prone to be defaulters.

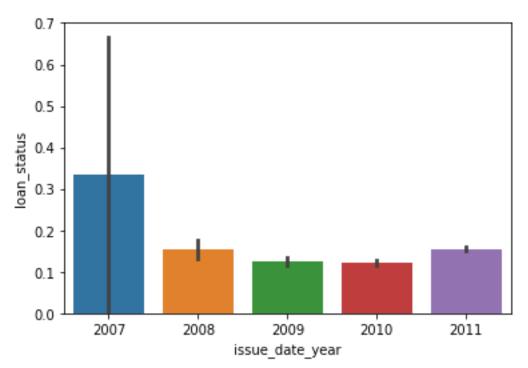


- Above graphs shows the comparison between deling in 2 year and status of loan with respect to DTI.
- Above graphs shows the trend with higher value for deling 2 years value are more prone to be defaulters also the DTI starts to diminish as deling value rises.
- Making applicant more prone to be defaulter.



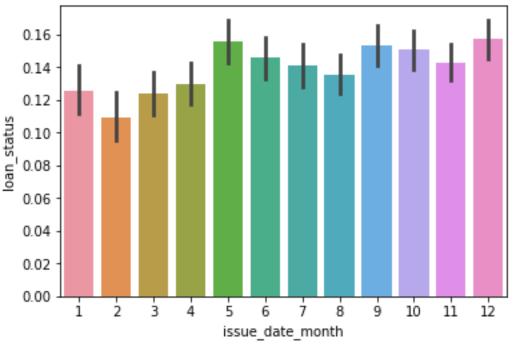


Data Graphs Issue Date



Inferences

- Above graphs shows the comparison between status of loan and Issued Date years.
- Above graphs shows the trend that loan that were approved in year of 2007 having more number of defaulter applicant.
- Post 2007 the trend decrease for 2008-10 then in 2011 the number of defaulters increase.
- Assumption: Defaulter for the year of 2007 may have lost their job in 2008-10 economics crises causing them to turn as defaulter.

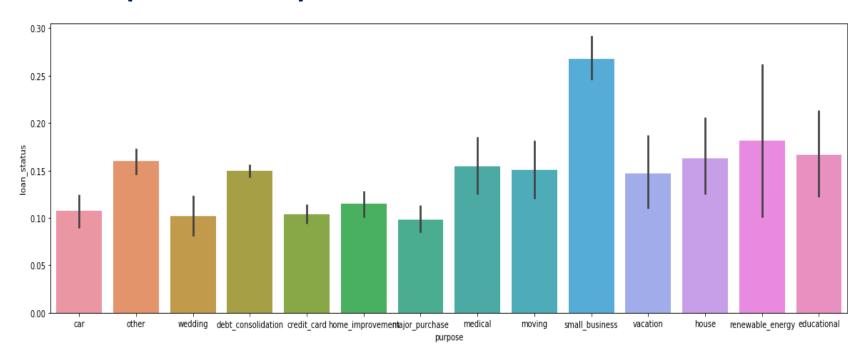


- Above graphs shows the comparison between status of loan and Issued Date months.
- Above graphs shows the trend that loan that were approved for the month of May and December are turning defaulters.
- Assumption: The reason for these Defaulter may be they spend on Summer and Christmas Vacations but failed to repay the loan lately.





Data Graphs Loan Purpose

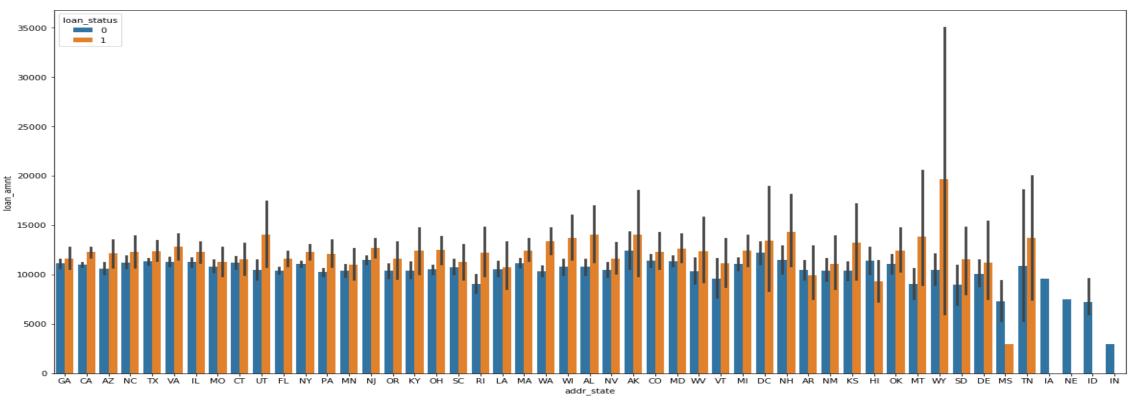


- Above graphs shows the comparison between the purpose of loan and status of loan.
- We can clearly infer applicants who are applying loan for small business are tend to be defaulters.
- Hence the company can keep interest rate higher for such type of loans.





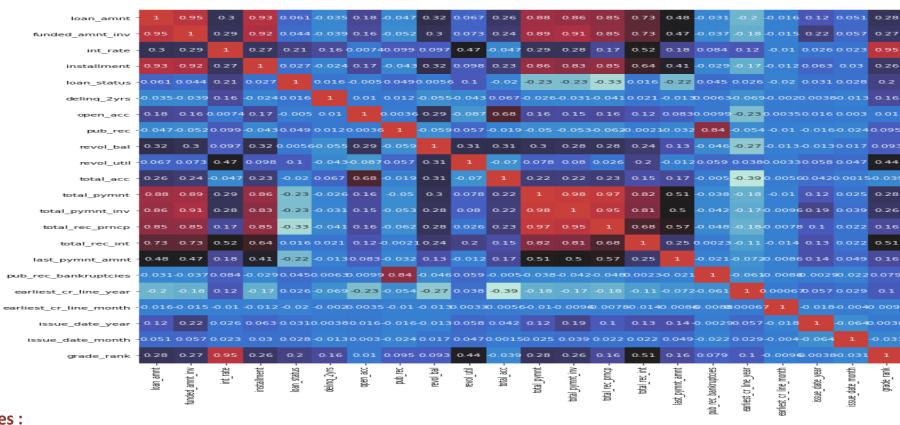
Data Graphs Demographics



- Above graphs shows the comparison between the State of applicant and loan amount with respect to loan Status.
- We can clearly infer applicants who are applying loan from State WY (Wyoming) are on higher risk for company they can turn to be as Defaulter.
- Hence the company can keep interest rate higher while doing business from WY state.









- 1. Loan amount is correlated to last payment amount with r factor.48, as expected
- 2. Interest rate is highly correlated to revolve util with r factor of .47, which shows that company is charging the defaulter with higher rate of interest.
- 3. Interest rate is highly correlated with grade rank, as we have also seen in graphs above.
- 4. Loan amount revolve balance are correlated with r factor .32 This is not good as it suggests that higher loan amount is being approved to riskier borrowers.
- 5. Delinq 2yrs is totally un-correlated with public record of bankruptcy. Therefore they represent distinct features with individual predictive value.
- 6. Revolve util is inversely or un-correlated with earliest credit line year. We depict that these two features are independent.





Defaulter Driving Factors

- Lower Income group, lower the income higher the chances of Applicant turning into defaulter.
- DTI, as the parameter increases the risk of Applicant turning into defaulter increases.
- Months like May and December shows more defaulters. This could be due to holiday season, applicant tend to get the loan but failed to repay.
- Purpose of loan, we have identifies most of the defaulter are from Small Business type of loan category.
- State of applicants, we have identify that Wyoming state has more defaulters, applicants having rented or other type of home Ownership are more likely to be Defaulters.

Recommendation

- To provide the loan at higher ROI to lower income group Applicants. This may lead to business loss as lower income group do not want to lend money at higher rate.
- To overcome the listed factor company should provide loan for less amount and for less tenure, which will reduce the risk of financial loss.
- For applicants having high DTI Company should provide loan at higher ROI.
- Our recommendation is to verify the purpose of loan before approval. On the basis of this company can decide the rate of interest and tenure of the loan.
- Company should consider Applicants parameter like state, home ownership, employment length before approving the application and deciding the amount, tenure and interest rate for the loan.