LENDING CLUB CASE STUDY

Submitted By:

Ravi Shankar

Abhishek Ranjan

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PROBLEM STATEMENT

- Lending Club, a consumer finance marketplace that provides various loans to urban customers, faces a significant challenge in managing its loan approval process. When assessing loan applications, the company needs to make informed decisions to reduce financial losses, particularly from loans granted to applicants deemed "risky."
- The main objective of this exercise is to help Lending Club reduce credit losses. This challenge stems from two possible scenarios:
 - o Identifying applicants who are likely to repay their loans is essential, as they can generate profits for the company through interest payments. Rejecting these applicants would mean missing out on potential business opportunities.
 - o Conversely, approving loans for applicants who are unlikely to repay and pose a risk of default can lead to significant financial losses for the company.
- The company aims to identify the key factors (or driver variables) that contribute to loan defaults, specifically the variables that strongly indicate a likelihood of default. This understanding can be leveraged for better portfolio management and risk assessment.

DATA OVERVIEW

- Historical data provided by Lending Club.
- Contains borrower credit history and loan details.
- Size: Over 39,717 records and 111 columns.
- Data dictionary with descriptions

DATA UNDERSTANDING

- Target attribute: Loan Status
- Target attribute can have three distinct values
 - o Fully Paid: This category includes applicants who have successfully repaid both the principal and interest of the loan.
 - o Current: Applicants in this group are actively making loan installments, meaning the loan term is still ongoing. These individuals are not considered *defaulted*
 - o Charged-off: This classification applies to applicants who have not made timely payments for an extended period, leading to a default on the loan.

Cont.

DATA UNDERSTANDING

- Customer attributes
 - Annual Income (annual_inc)
 - Home ownership (home_ownership)
 - Debt to Income (dti)
 - State (addr_state)
 - Employemnt year (emp_length)
- Loan attributes
 - Loan amount (loan_amnt)
 - Loan Term (term)
 - Interest rate (int_rate)
 - Loan date (issue_d)
 - Grade (grade)
 - Loan purpose (purpose)
 - Public record bankrupticies (pub_rec_bankruptcies)

DATA CLEANING

- Removal of columns having all null values
- Removal of columns which have single value
- Removal of columns which have post loan approval features
- Removal of columns which has string data like *emp_title*, *desc*, *title*, *url*
- Removal of entries which has loan status of *current* (as they don't help in analysis)
- Handling of missing values
 - o Removal of entries which has missing values (~ 4.48% rows deleted)

Cont.

DATA CLEANING

- Modifying values
 - Removing string *months* from column *term*
 - o Removal of char % from the interest rate colum *int_rate*
 - Modifying *emp_length* column by removing string *years* and converting 10+ to 10 and converting < 1 to 0
- Deriving columns issue_month & issue_year from loan issue column issue_d
- Removing outliers
 - o Plotting boxplot for columns : loan_amnt, int_rate, annual_inc, dti
 - Use standard IQR technique to remove outlier
 - \square Only data falling between [Q1 1.5*IQR, Q3 + 1.5 * IQR] are kept for analysis

Cont.

DATA CLEANING

- Categorizing columns: Since some of the loan attribute columns like *loan_amnt*, *int_rate* are continuous numerical values, hence we can categorize this values into bucket (in new columns) for better analysis
 - O Loan amount (loan_amnt): [0-5k, 5k 10k, 10k 15k, 15k above], column: loan_amnt_bucket
 - o Loan interest rate (int_rate): [Very low, Low, Moderate, High, Very high], column: int_rate_bucket
 - o Annual Income (*annual_inc*) : [*0-40k* , *40k -50k*, *50k 60k*, *60k 70k*, *70k 80k*, *80k above*] , column: annual_inc_bucket
 - o Debt to income (dti): [Very low, Low, Moderate, High, Very high], column: dti_bucket

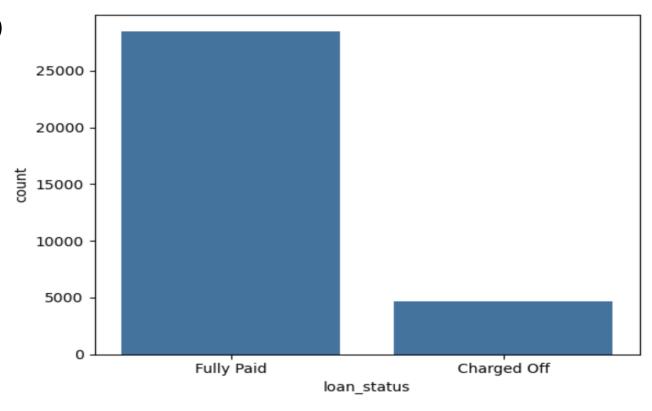
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DATA ANALYSIS

- # of interested columns = 25
- Date-time : [issue_d , issue_month, issue_year]
- Numerical: [loan_amnt, funded_amnt, funded_amnt_inv, term, int_rate, installment, annual_inc, dti]
- Categorical: [grade, sub_grade, emp_length, home_ownership, verification_status, loan_status, purpose, addr_state, pub_rec, pub_rec_bankruptcies, loan_amnt_bucket, int_rate_bucket, annual_inc_bucket, dti_bucket]
- Target variable : [*loan_status*]

Analysis of Loan status (loan_status)

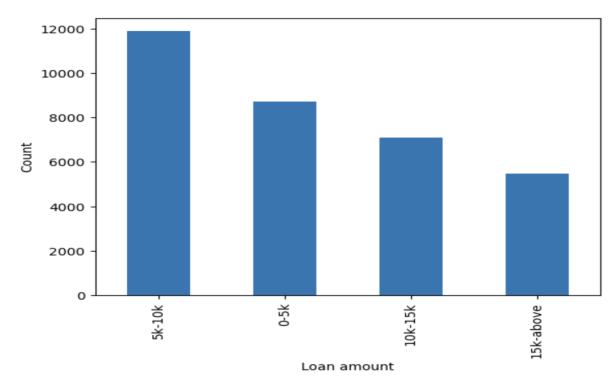
Loan Status	
Fully Paid	28494
Charged Off	4699



Observation: 14.16% of loan in the data set has defaulted

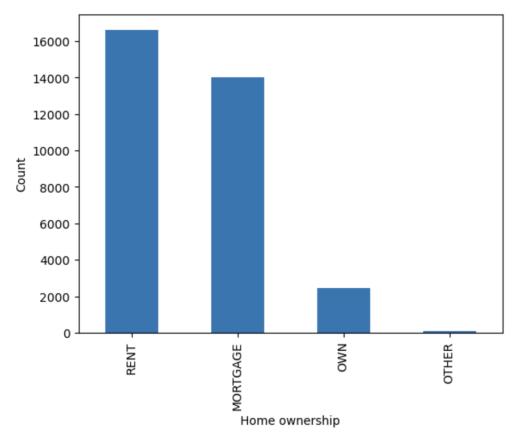
Cont.

• Analysis of Loan amount (loan_amnt_bucket)



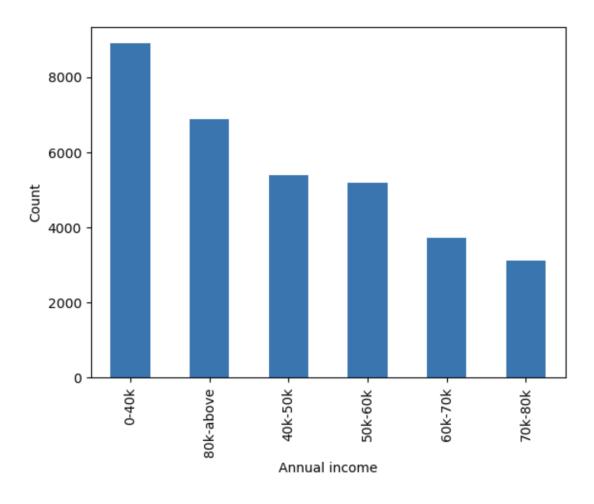
Observation: Most of the loan application is for amount between 5000 -10000

• Analysis of home ownership (*home_ownership*)



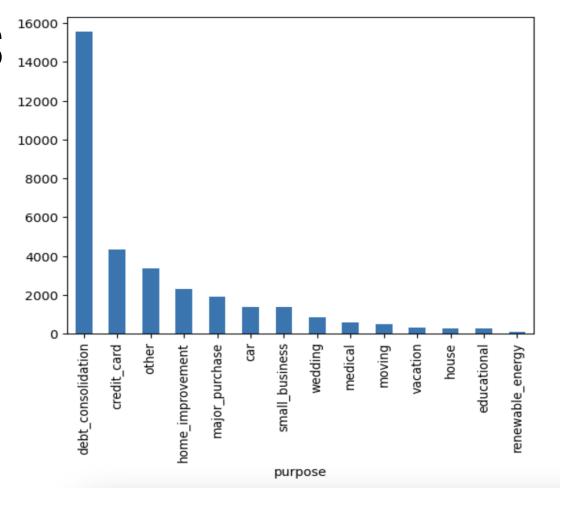
Observation: Majority of the loan applicants are ether living on rent or have house on mortgage

• Analysis of applicant annual income



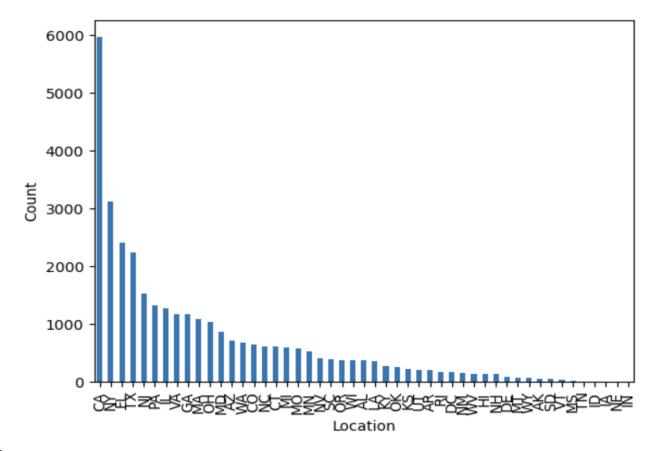
Observation: Majority of the applicants has income between 0-40k

• Analysis of loan purpose (purpose)



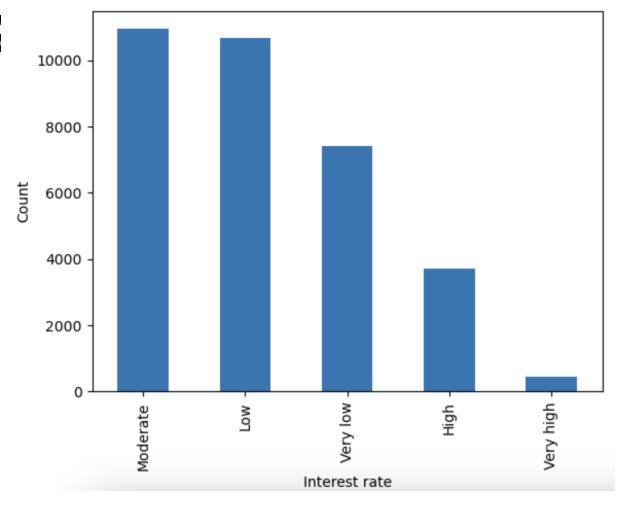
Observation: Most of the applicants apply loan for the debt consolidation followed by settlement of credit card

• Analysis of location (addr_state)



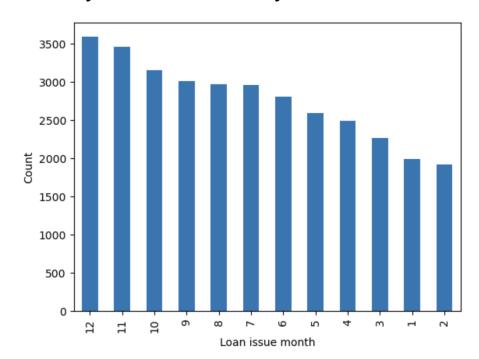
<u>Observation</u>: State CA has most of the applicants

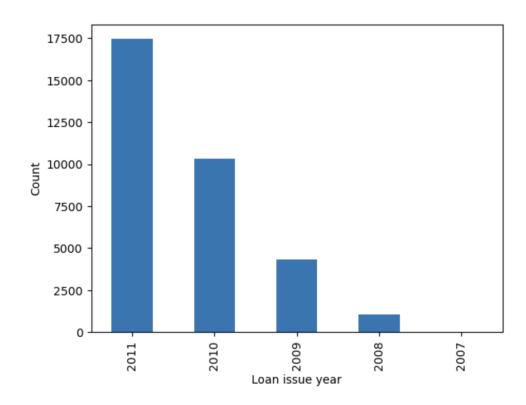
• Analysis of interest rate (*int_rate*)



Observation: Most of the applicants have either low (<8%) or moderate(12% to 16%) interest rate.

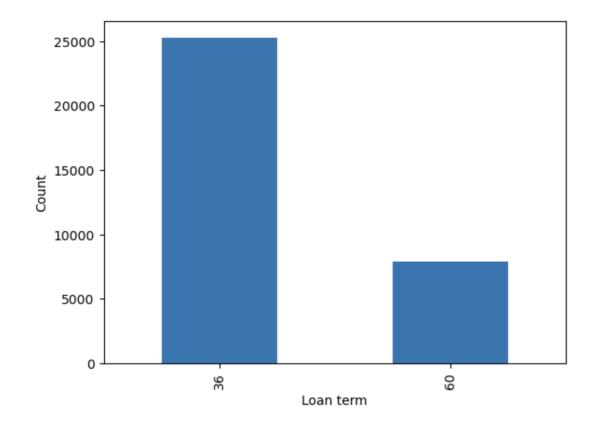
• Analysis of month & year





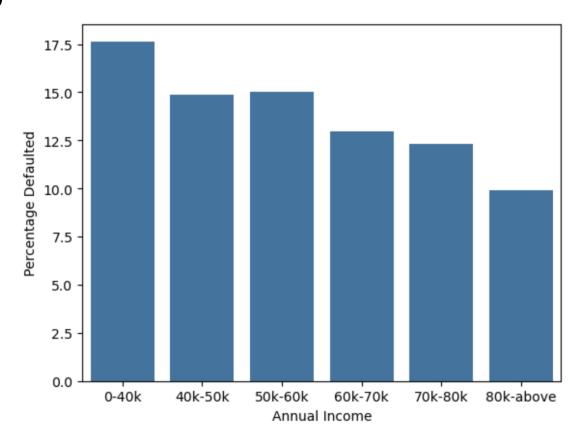
Observation: In year 2011 and month of dec the loan application is highest

• Analysis of loan term



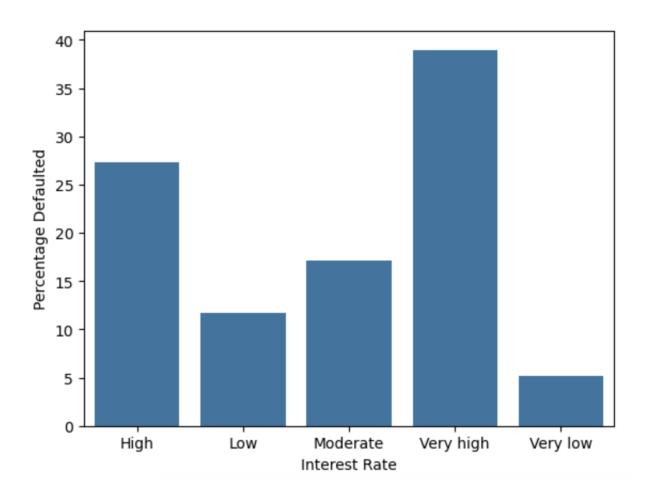
<u>Observation</u>: Most of the applicants took loan for 36 months term

• Analysis of annual_inc vs loan status



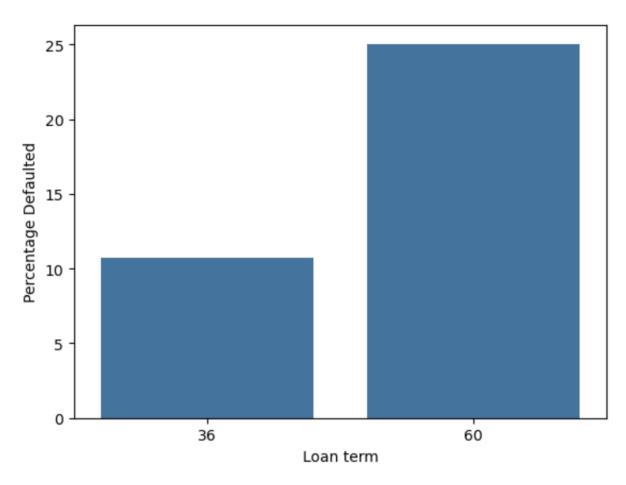
Observation: Applicant with annual income in range 0-40k are most defaulted one 1

• Analysis of int_rate vs loan status



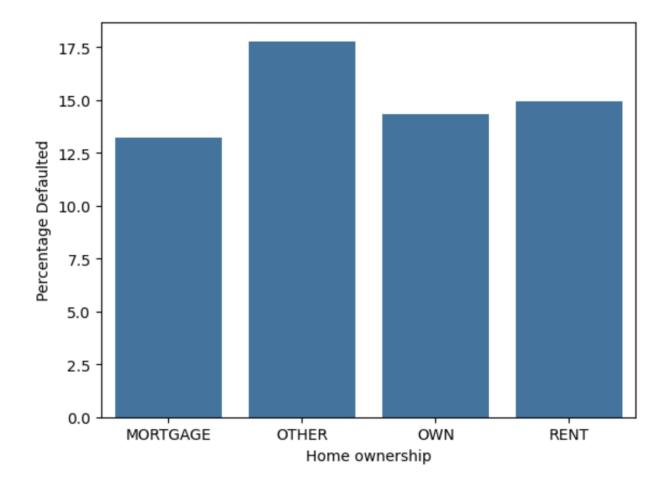
Observation: Applicants whose loan interest rate is more than 20%v are likely to get default

• Analysis of loan term vs loan status



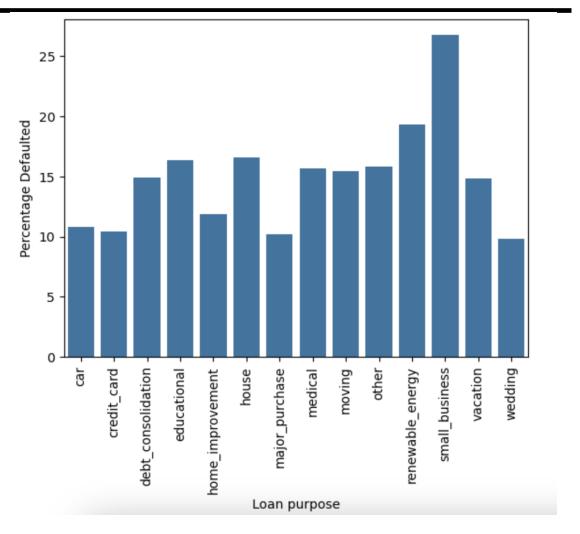
Observation: Applicants whose loan term is 60 months are likely to get default

• Analysis of home ownership vs *loan status*



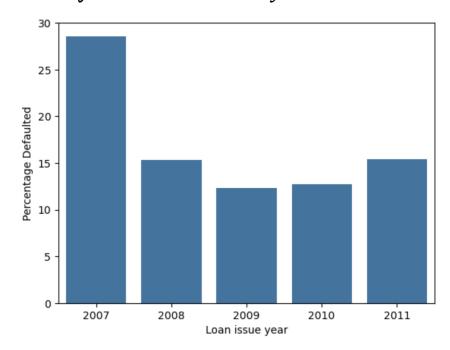
<u>Observation</u>: We cannot get clear indication if home ownership data can tell us about the loan status as almost all categorical value are equally defaulted

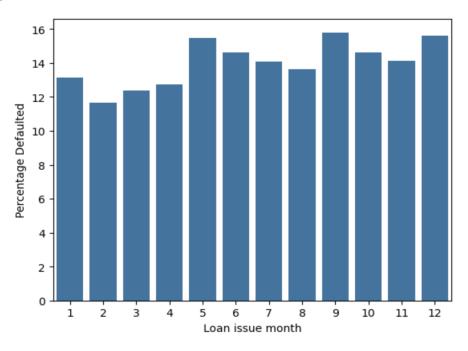
• Analysis of Loan purpose vs *loan status*



Observation: Applicants who availed loan for small businesses are most defaulted

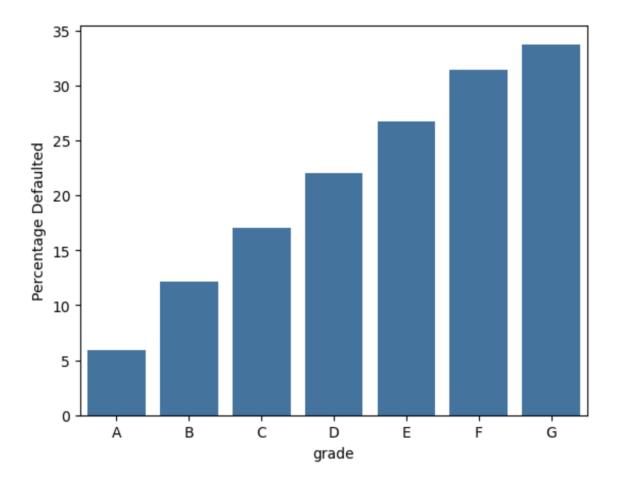
• Analysis of Loan issue year/month vs Loan status





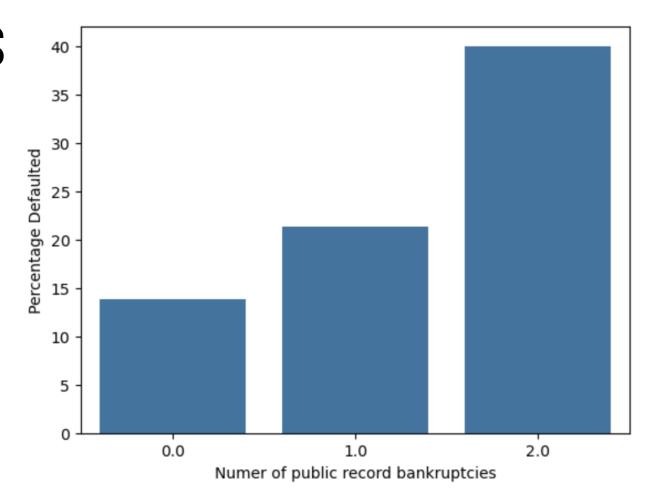
Observation: Loan issued in year 2007 and in months of sept is more likely to get default

• Analysis of Grade vs Loan status



Observation: Grade G get default more followed by F

• Analysis of Num of *Public record* bankruptcies vs Loan status



Observation: Public record bankruptcies of count 2 are more default

OBSERVATIONS

The analysis of charged-off loans for each variable indicates the following. The likelihood of default increases when:

- Applicant has annual income in range 0-40k
- Applicant's interest rate is very high: > 20%
- Loan term is of 60 months
- Loan amount is more than 15k
- Applicants use the loan for small businesses
- Loan is issued in year 2007 or in month of sept to dec
- Applicant is from TX state
- Applicant has more that 1 bankruptcies in past
- DTI is more than 20%

RECOMMENDATIONS

- Consider strict check when applicants annual income is in range 0-40k
- Refrain from giving high interest rate of > 20%
- When lending loan for 60 months period, extra measures need to be taken.
- Implement rigorous audit and scrutiny measures when loan is being availed for small businesses.
- Take extra care when applicant is from TX state.
- Refrain from lending loan to applicants who have went bankruptcies more than 1 in past.
- Don't lend loan when DTI is more than 20%

THANK YOU