



## Group Facilitator

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## Batch

ML C50



# Lending Club Case Study

upGrad



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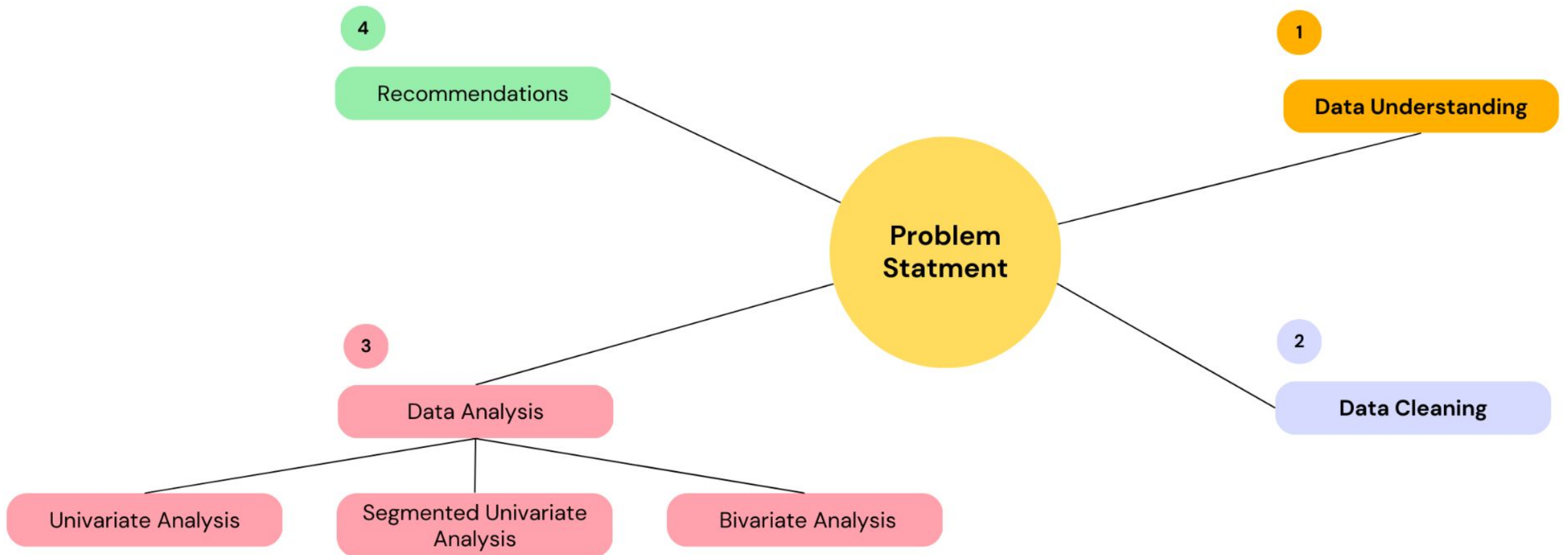
1. Problem Statement
2. Approach
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# Problem Statement

Lending Club is the leading online loan marketplace, offering a wide range of financial products including personal loans, business loans, and financing for medical procedures. One of the biggest challenges for lenders is managing credit risk, which is the financial loss that occurs when borrowers default on their loans.

To address this challenge, Lending Club is focused on understanding the key drivers behind loan defaults, including the factors that strongly indicate when a borrower is likely to default. By identifying these key variables, Lending Club can better manage its portfolio and assess risks to ensure the best possible outcomes for both borrowers and investors.

# Problem solving approach



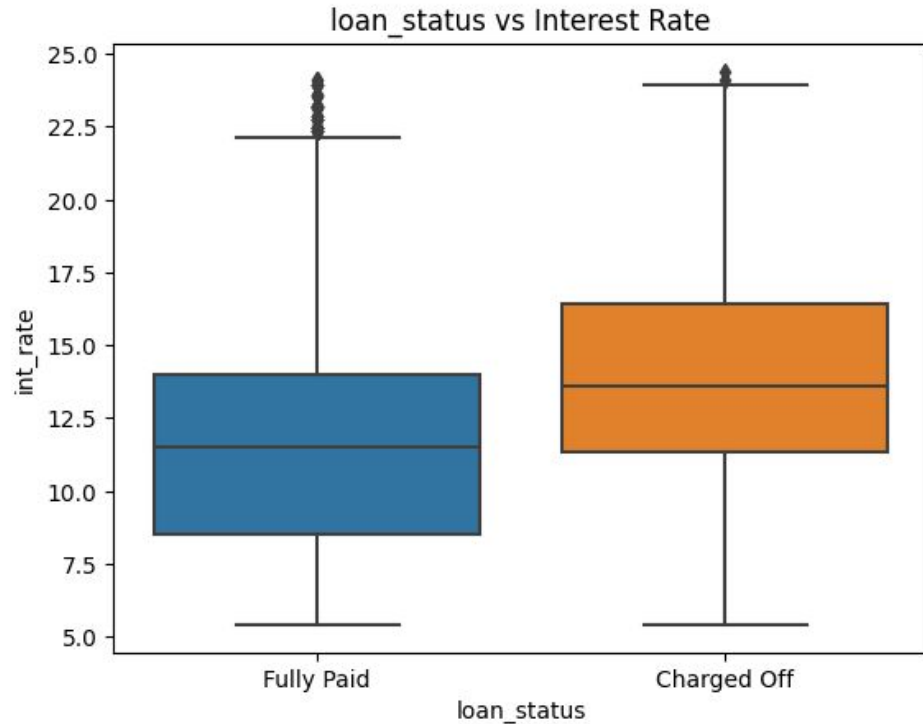
# Data Cleaning

- ❖ We removed columns with Null % greater than 30 %
- ❖ We removed trailing spaces from the categorical data
- ❖ We dropped columns which had 1 unique values
- ❖ We dropped columns which had no significant impact on loan default prediction
- ❖ We dropped records for which loan payment is in progress
- ❖ loan\_amnt , funded\_amnt , int\_rate & revol\_util casted to float as they are amount & rate of interest respectively
- ❖ issue\_d converted to date format and month & year derived as additional columns
- ❖ loan\_status\_num created to identify loan status numerically



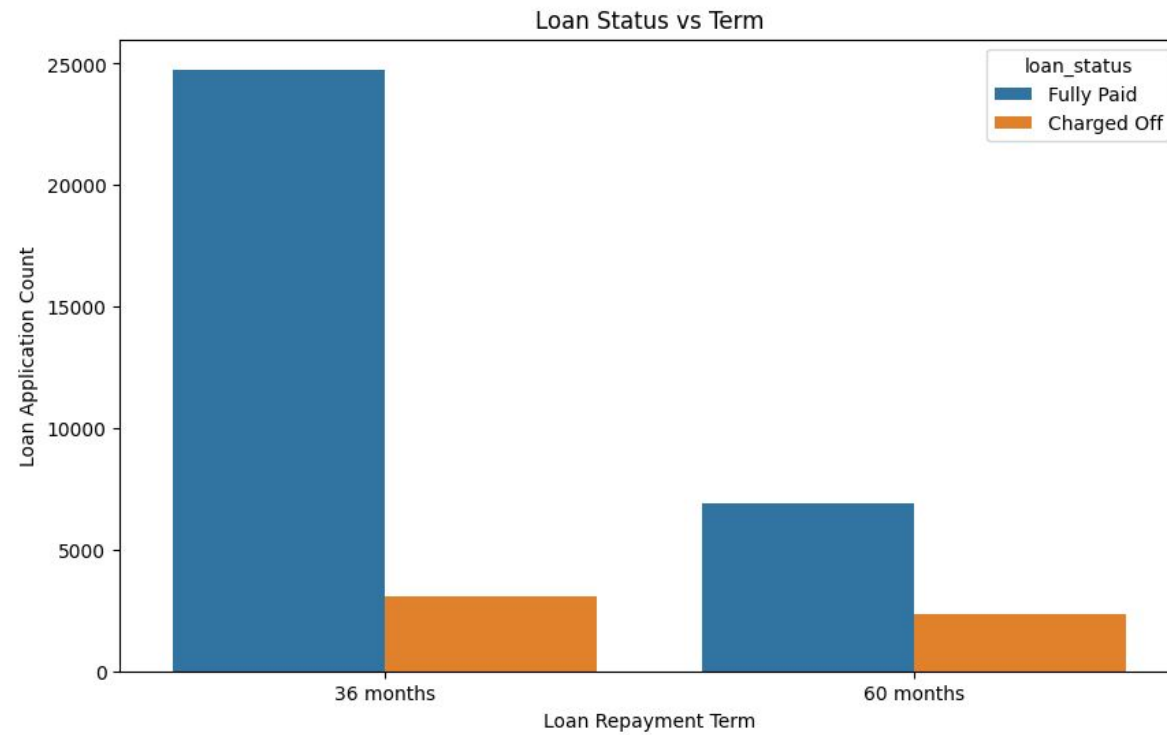
# **Data Analysis & Observation**

# Interest Rates



1. Interest rates less than 10% have very low chances of resulting in charge-offs. The minimum interest rate for charge-offs is 5.4%.
2. According to the quartile distribution, interest rates above 14% have a higher probability of resulting in default.
3. The proportion of charge-offs increases with higher interest rates.

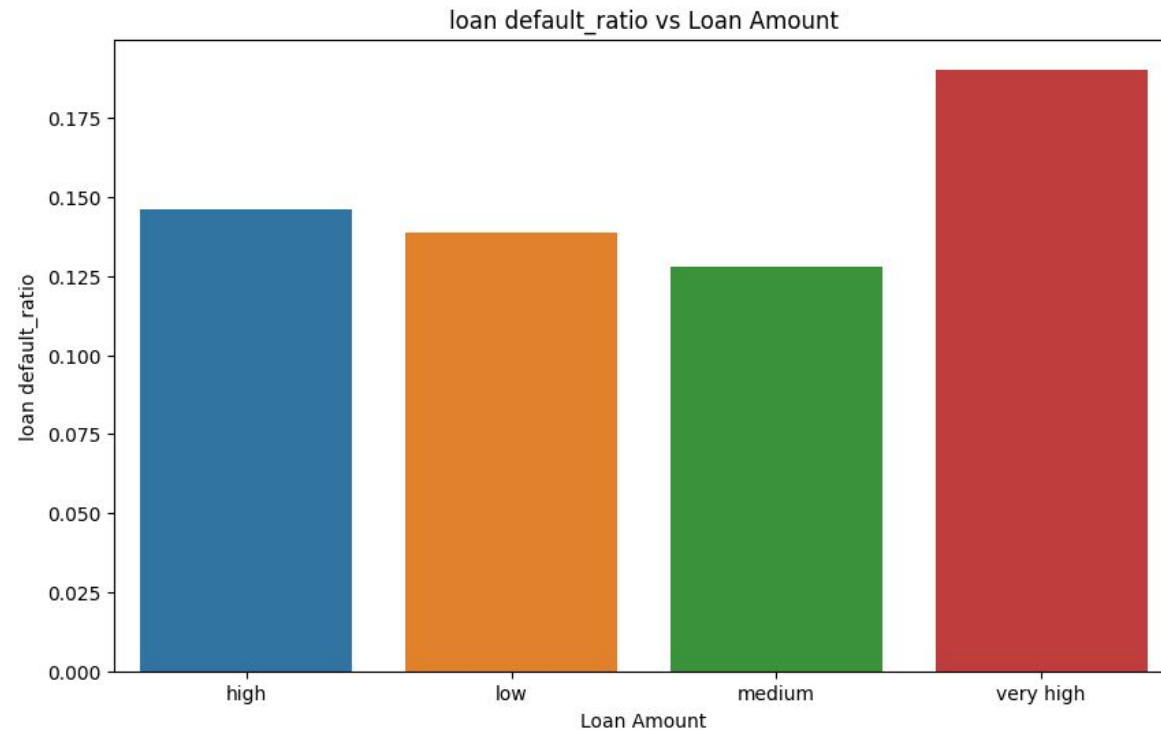
# Term



People who have taken loan for a tenure of 60 months have higher chances of defaulting

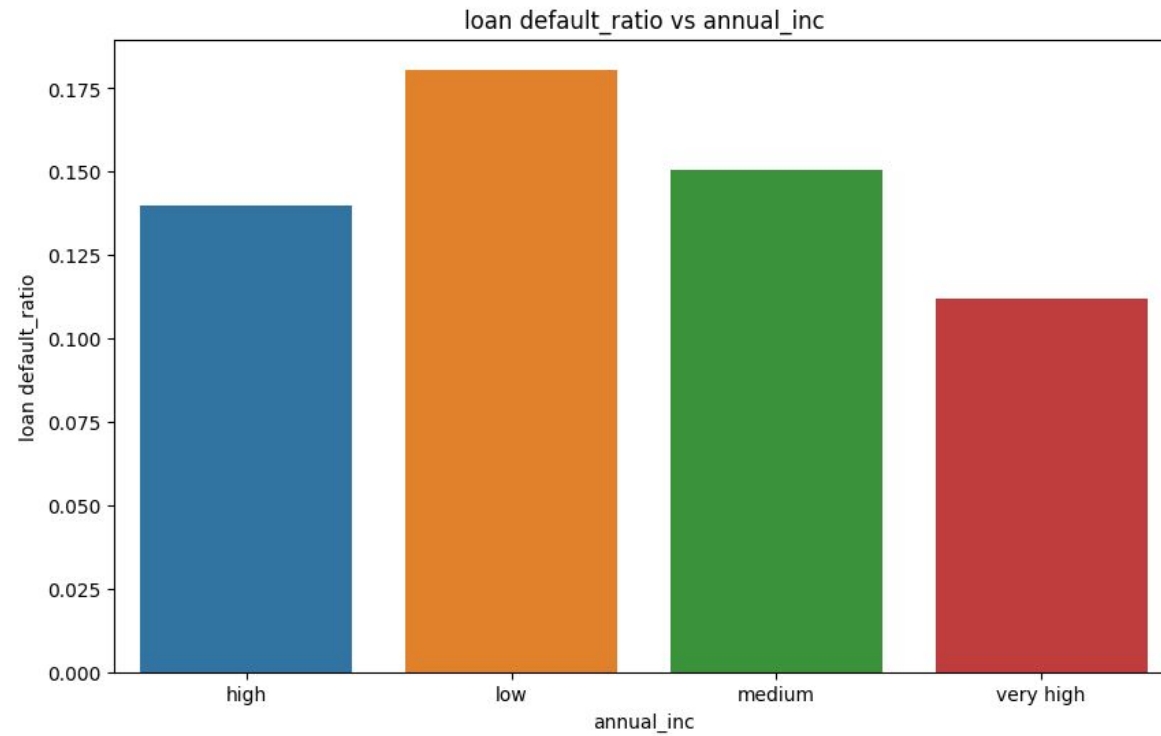


# Loan Amount



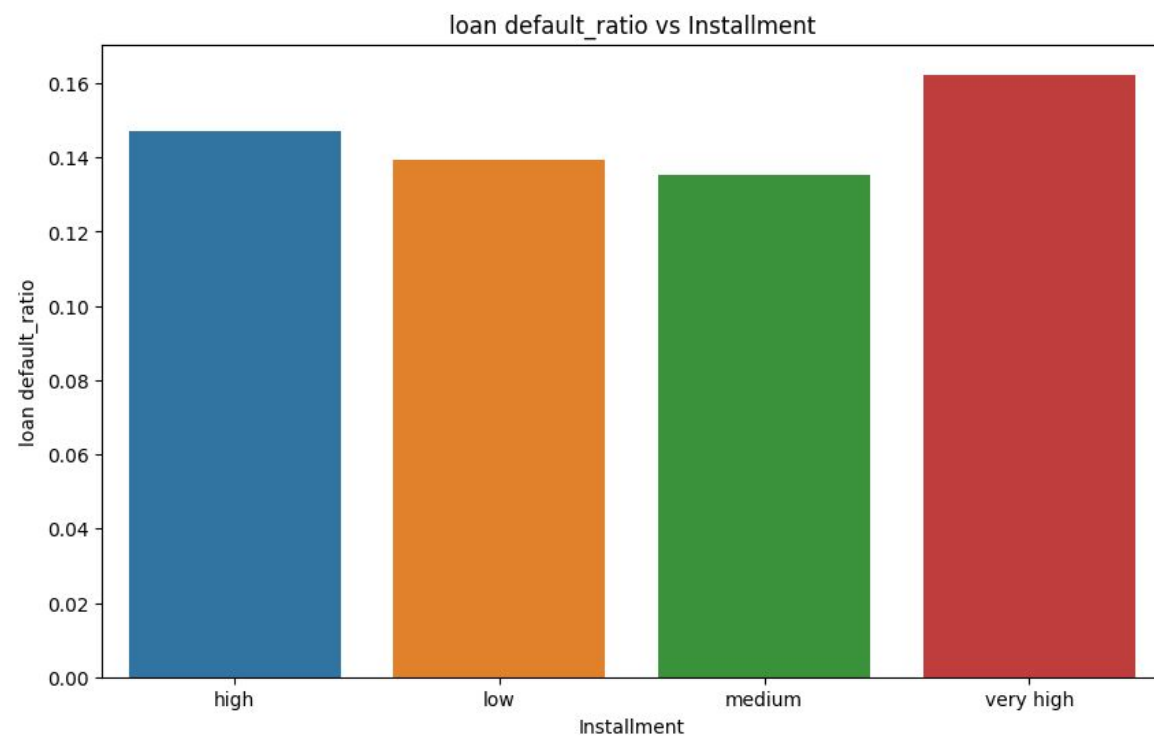
Higher the loan amount, greater the chance of the loan getting default.

# Annual Income



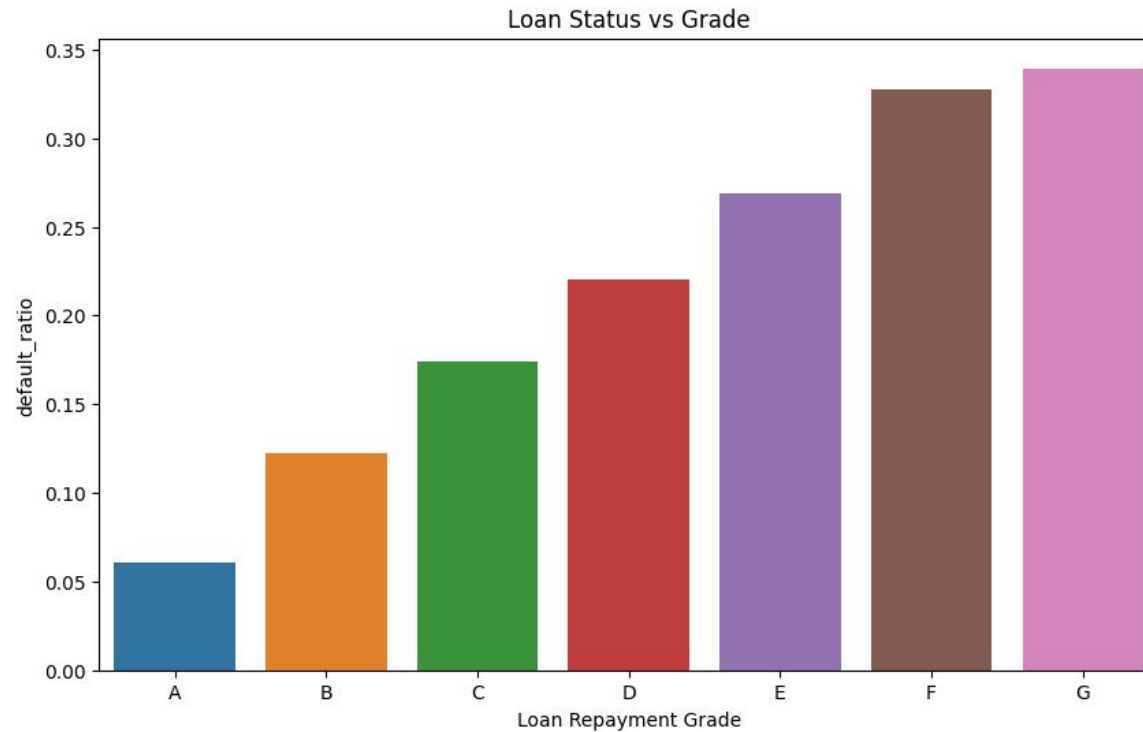
Low Annual Income have slight chances to default

# Installment



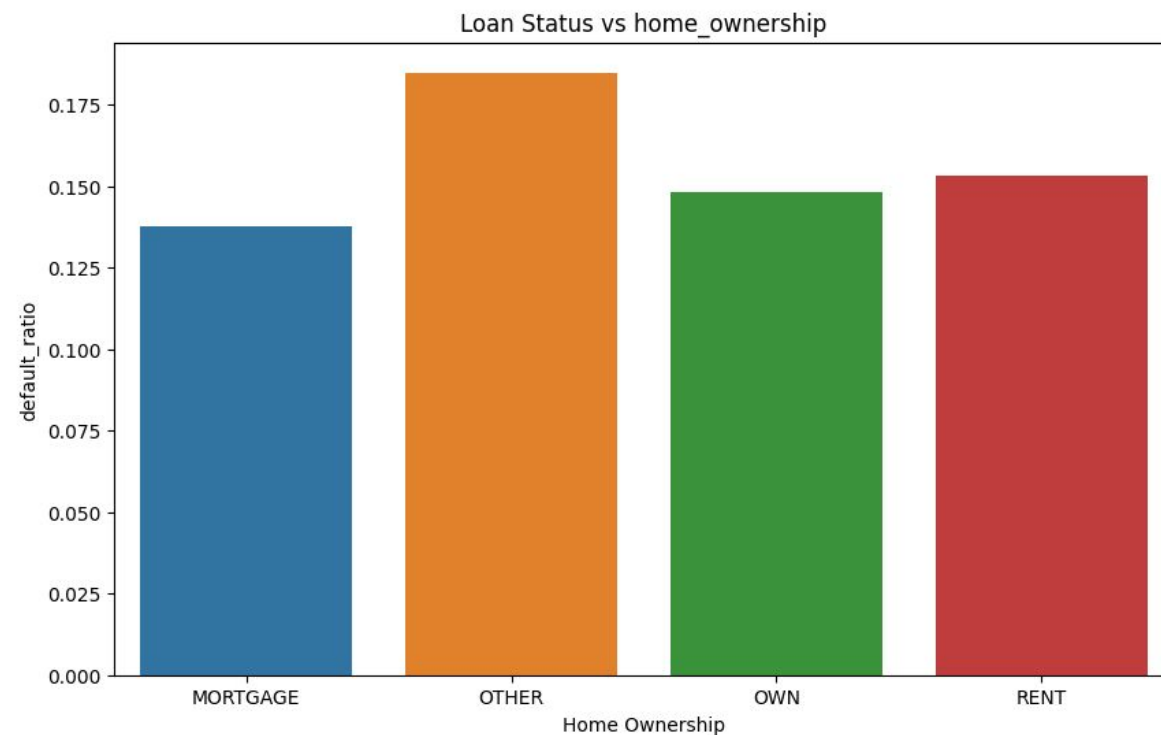
Higher installment have around 16 % chances of default

# Grade



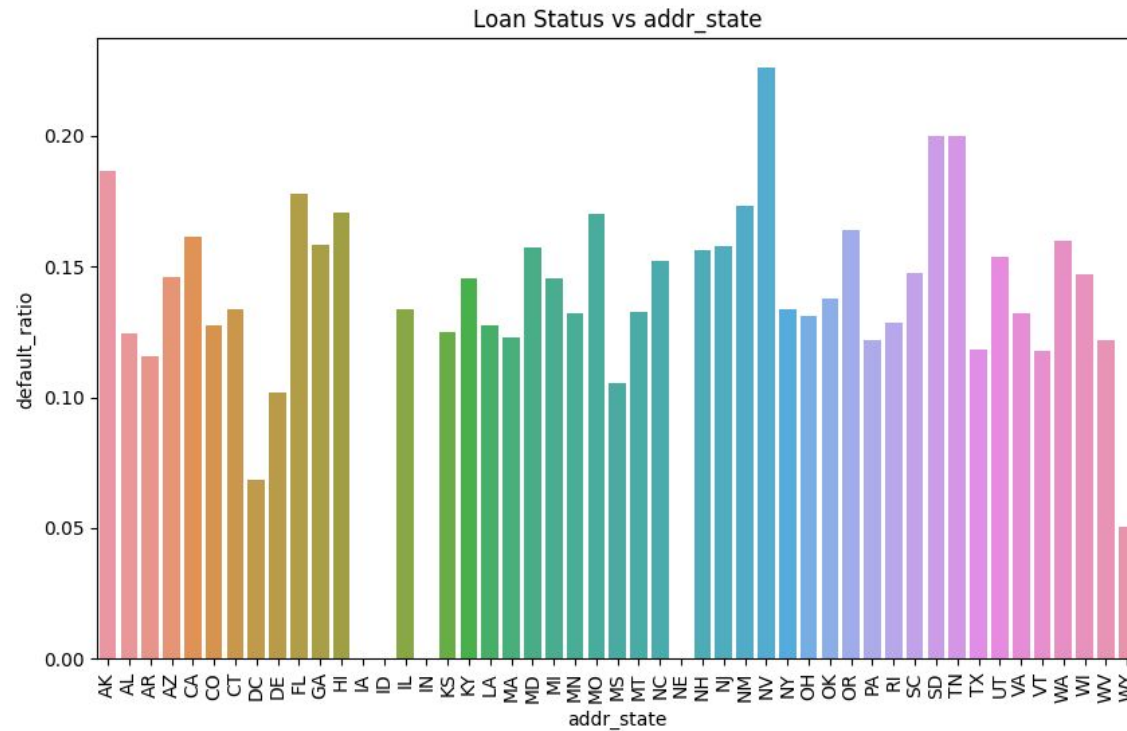
1. Grade F & G have highest chances of default
2. Grade A has the least chance of default
3. As we saw earlier Grade and sub grade have same trend hence this behaviour would be same in subgrade also , where in subgrades of F & G would be high risk to default

# Home Ownership



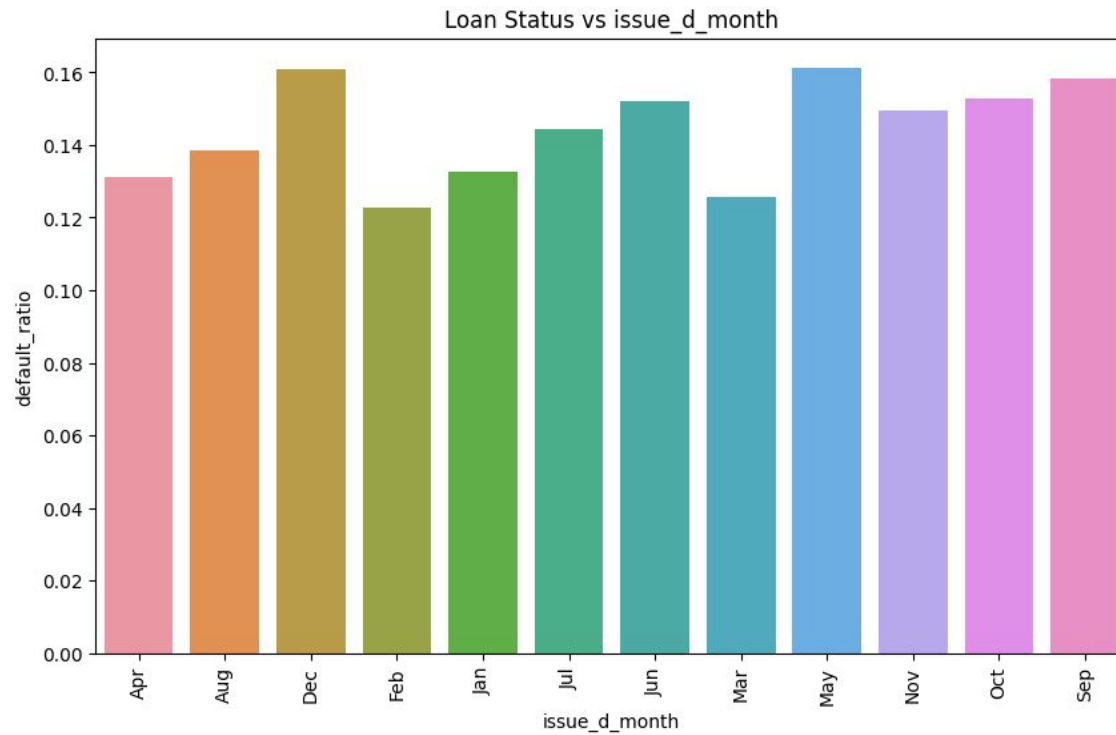
There is slight impact of other category in home\_ownership to default

# Address State



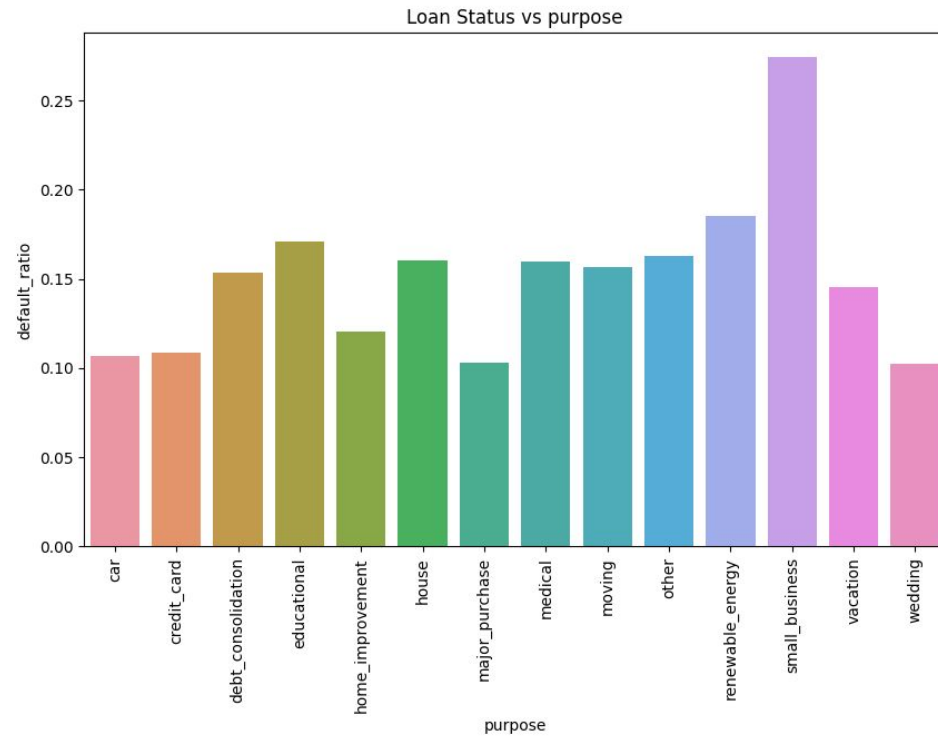
High chances of default are in NV,SD,TN,NM,MO & the least are in IA,ID,IN,NE

# Issue Month



The issuance of loans in May, September, October, November, and December has minimal impact on loan defaults.

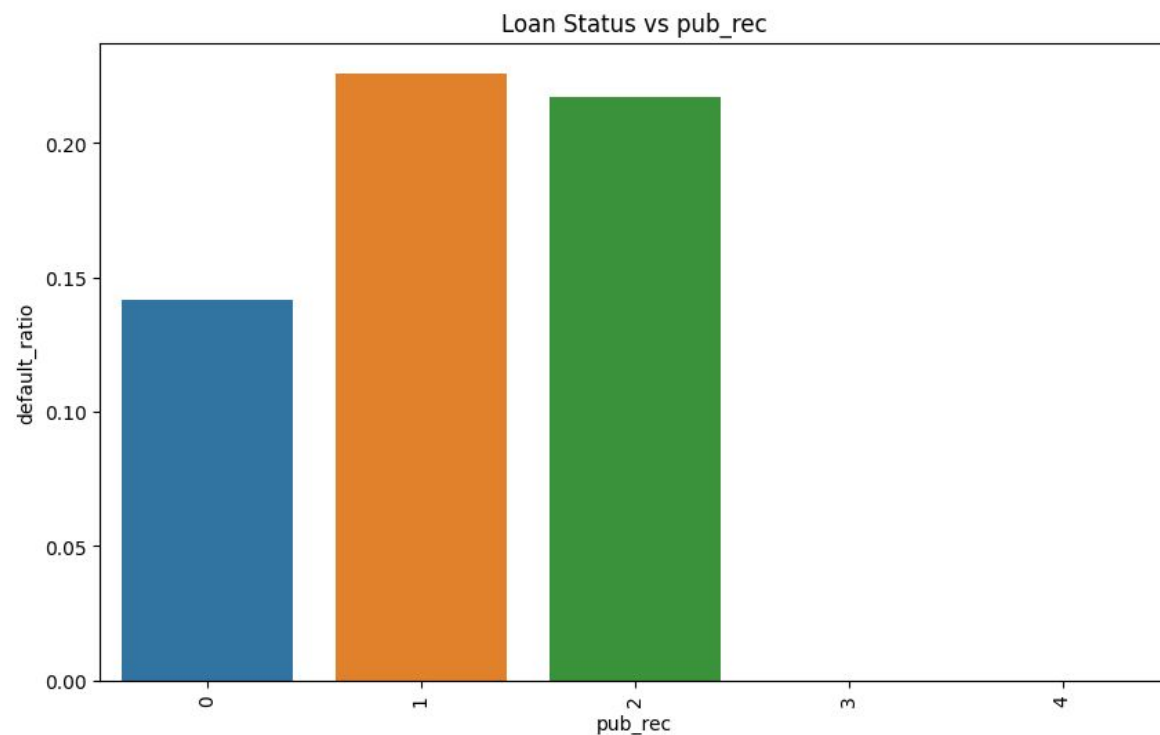
# Purpose



Small businesses, renewable energy, and educational purposes have a higher probability of defaulting on their loans.

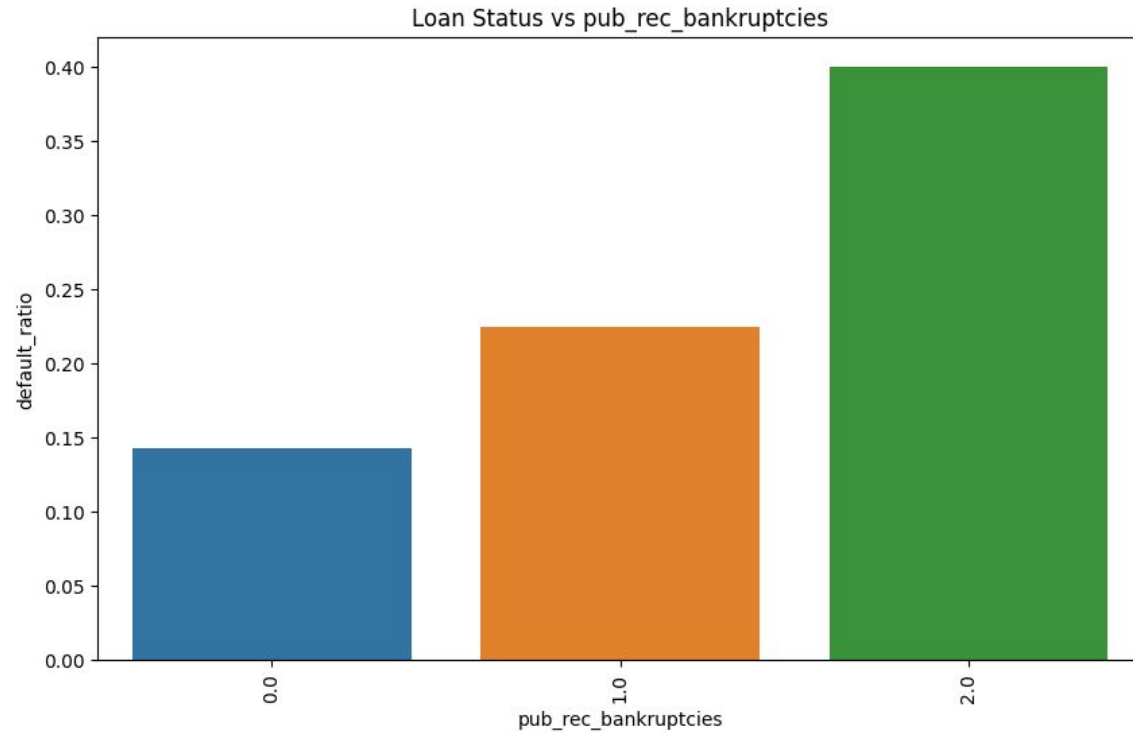


# Number of derogatory public records



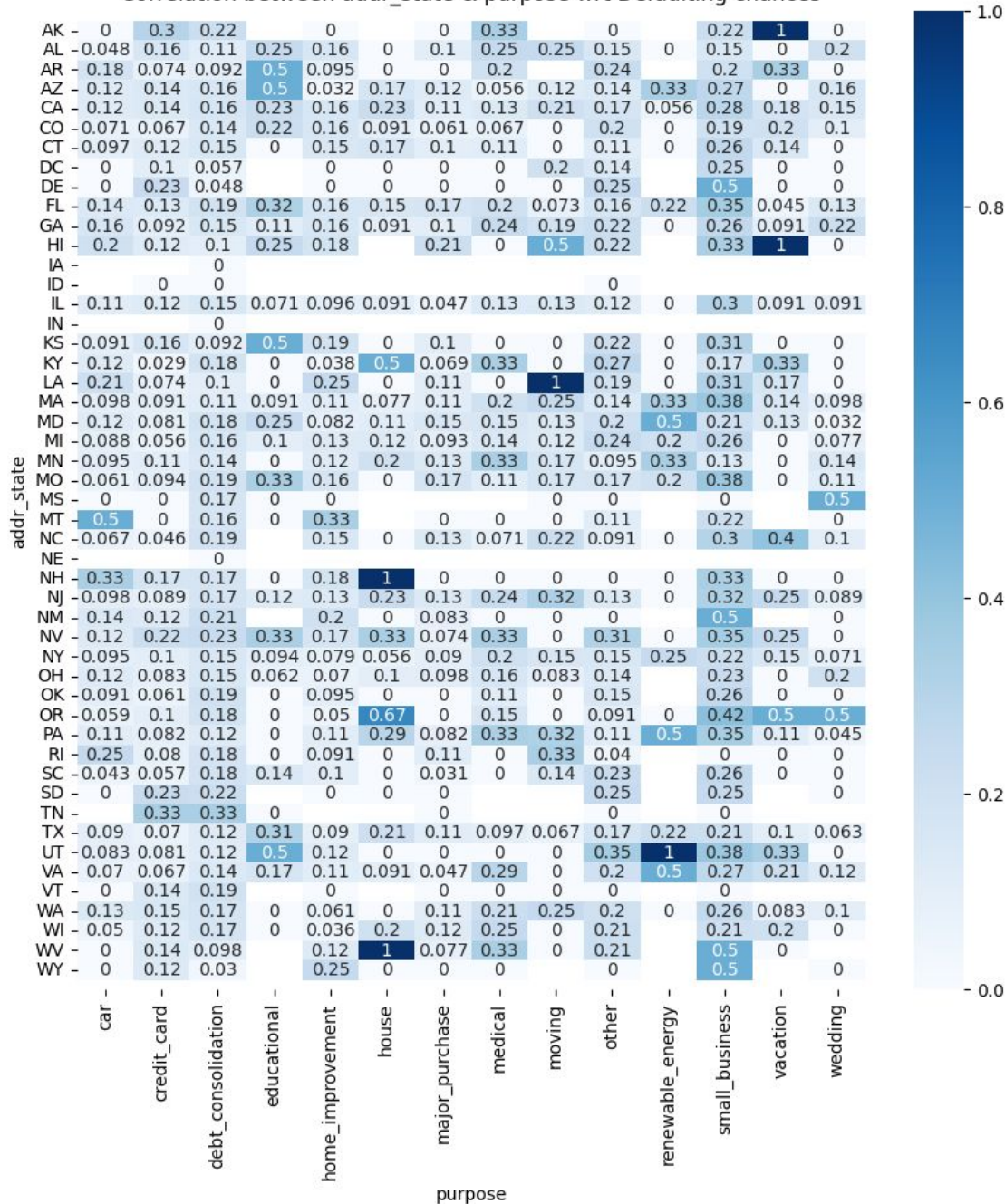
Borrowers with derogatory public records, specifically those with public records marked as 1 or 2, have a higher likelihood of defaulting on their loans.

# Number of public record bankruptcies



Borrowers with public bankruptcy records (1 & 2) are more likely to default on their loans.

Correlation between addr\_state &amp; purpose wrt Defaulting chances

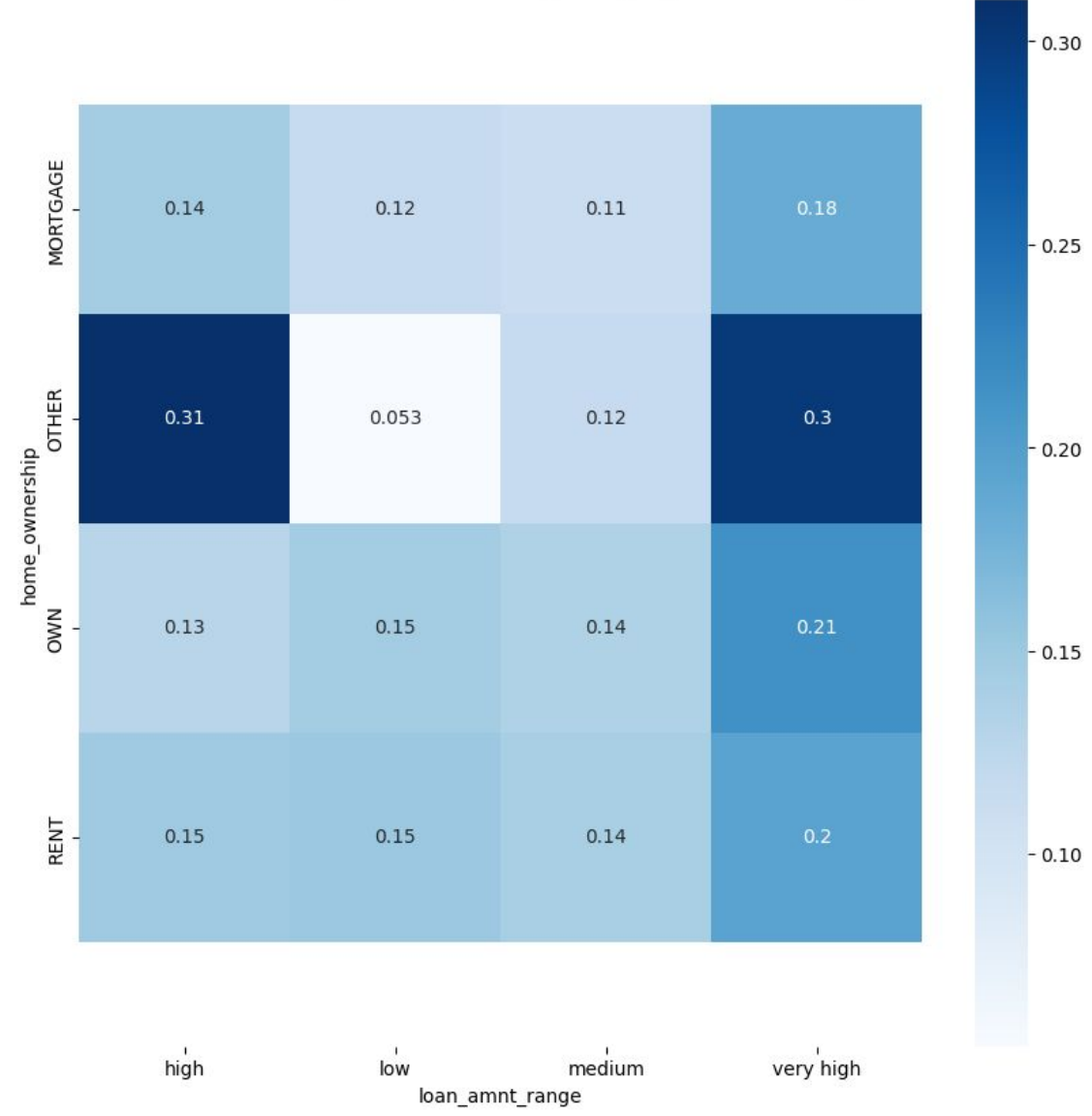


## addr\_state & purpose wrt Defaulting chances

Based on the above plot, the following observations can be made regarding risky loans:

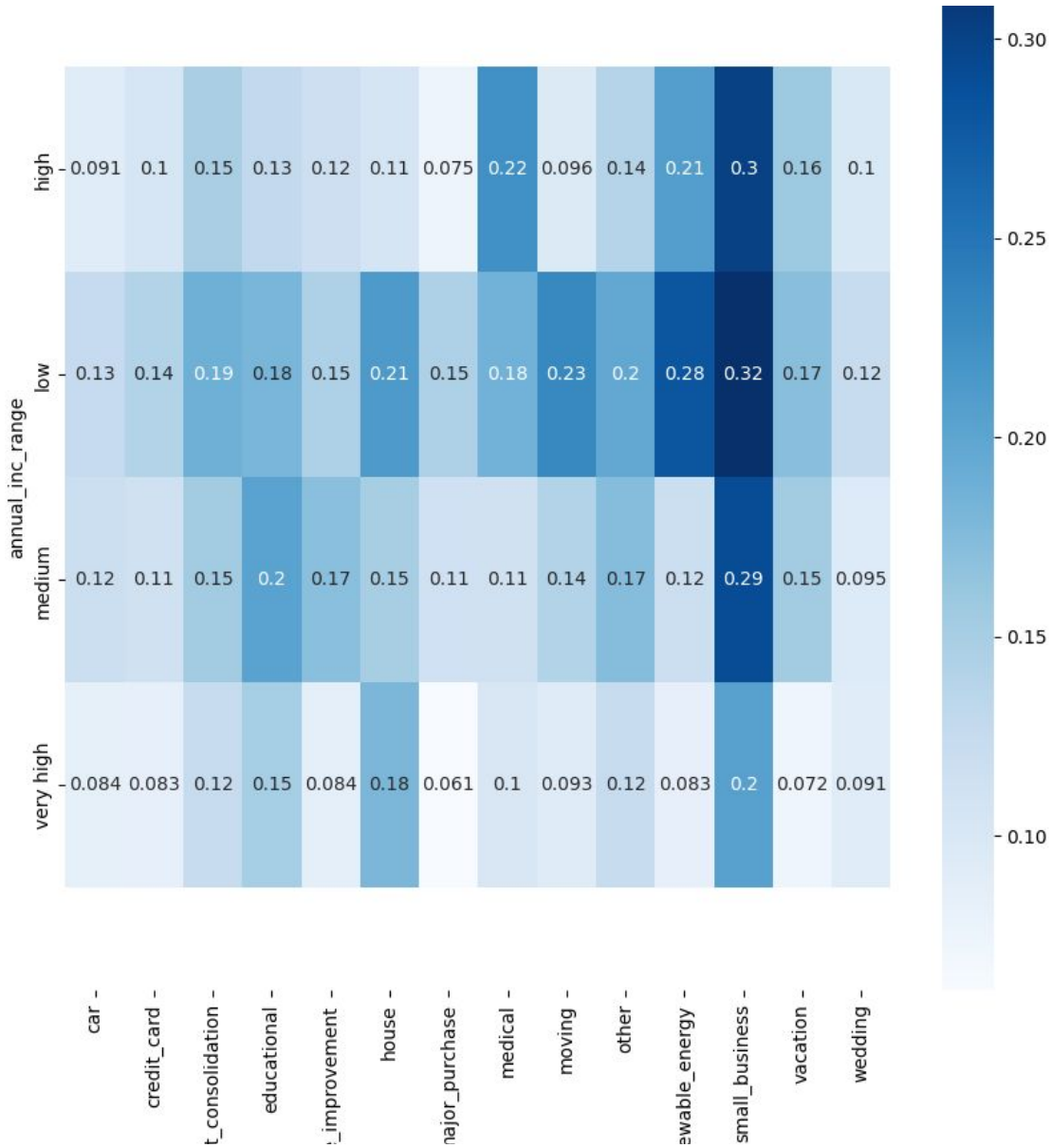
1. Education loans in AR, KS, and UT
2. Small business loans in DE, NM, WV, and WY
3. Vacation loans in AK and HI
4. Housing loans in NH and WV
5. Car loans in MT
6. Moving loans in LA and HI

Correlation between home\_ownership & loan\_amnt\_range wrt Defaulting chances



# Home\_ownership & loan\_amnt\_range wrt Defaulting chances

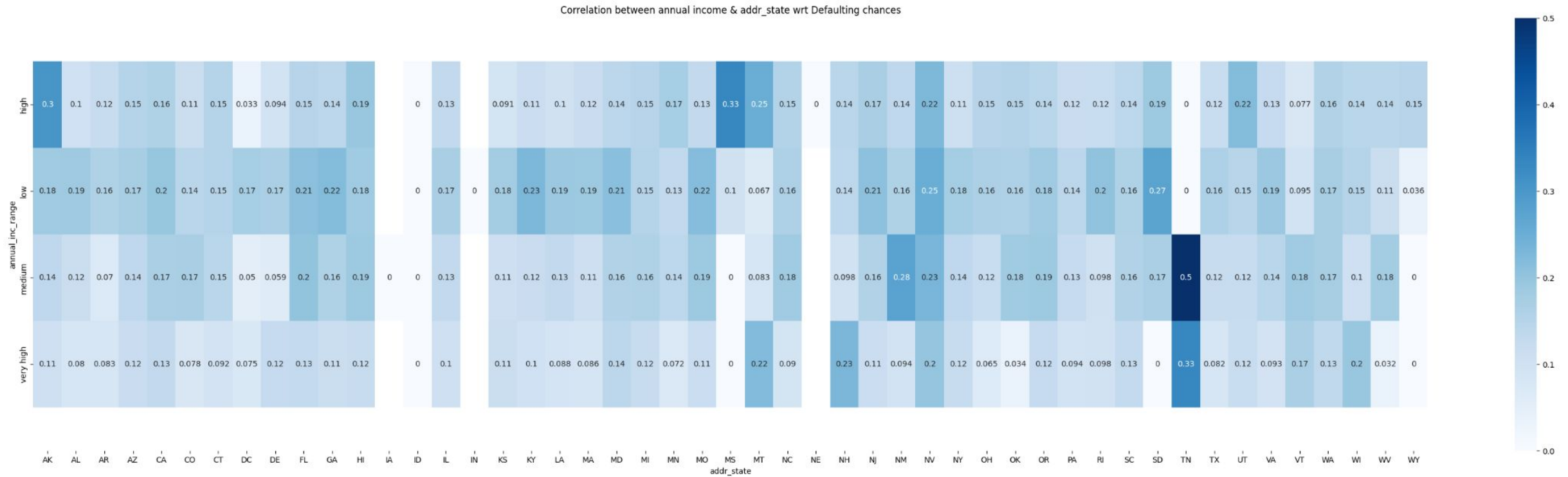
People with High and Very High Income with Homeownership(other) tend to default on Loan



## Annual income & purpose wrt Defaulting chances

People with both high and low income who are involved in small business tend to have a higher tendency to default on loans.

# Annual income & addr\_state wrt Defaulting chances



Based on the analysis, it was observed that people with high and low incomes in small business tend to default on loans. Additionally, individuals with high incomes in AK, MS, and MT have a medium chance of default, while those with medium incomes in TN and NM have a high chance of default. Lastly, very high-income earners in TN, NH, and MT have a medium chance of default."

# Recommendations

1. Interest rate more than 14% have high probability to default
2. People who have taken loan for a tenure of 60 months have higher chances of defaulting
3. Higher the loan amount, greater the chance of the loan getting default.
4. Low Annual Income have slight chances to default
5. Higher installment have around 16 % chances of default(300 +)
6. Grade F & G along with its subgrade have highest chances of default
7. There is slight impact of other category in home\_ownership to default
8. High chances of default are in NV,SD,TN,NM,MO
9. Small Business , renewable energy ,educational purpose have higher chances to default
10. Derogatory public records (1 & 2) have high default chances

# Recommendations

11. Public bankruptcy records (1 & 2) have higher chances to default
12. Vacation loans in AK, HI
13. Housing loans in WV,NH
14. Small Business in DE, NM, WV, WY
15. Car Loan in MT
16. Moving Loan in LA , HI
17. People with High and Very High Income with Homeownership(other) tend to default on Loan
18. High Income People in AK,MS,MT have medium chances of default
19. Medium Income people in TN,NM have high chances of default
20. Very High Income people in TN,NH,MT have medium chances of default



# Thankyou

