

Lending Club Case Study

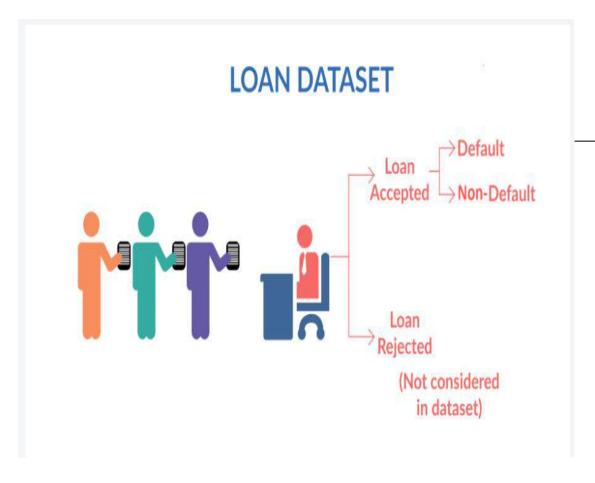
GROUP MEMBERS

- AAKANKSHA
- GUNEET SINGH GANDHI

PROBLEM STATEMENT

As a data scientist working for Lending Club analyze the dataset containing information about past loan applicants using EDA to understand how consumer attributes and loan attributes influence the tendency of default

DATA SET UNDERSTANDING



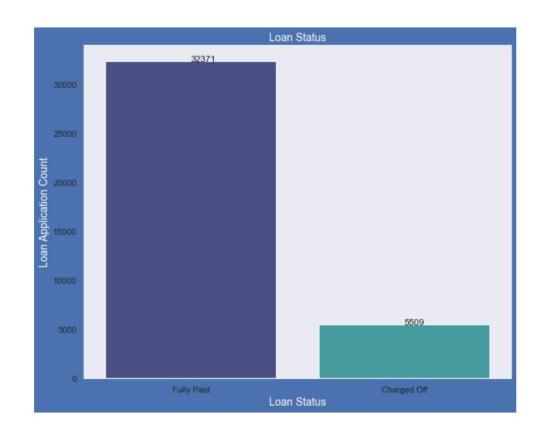
When a person applies for a loan, there are **two types of decisions** that could be taken by the company:

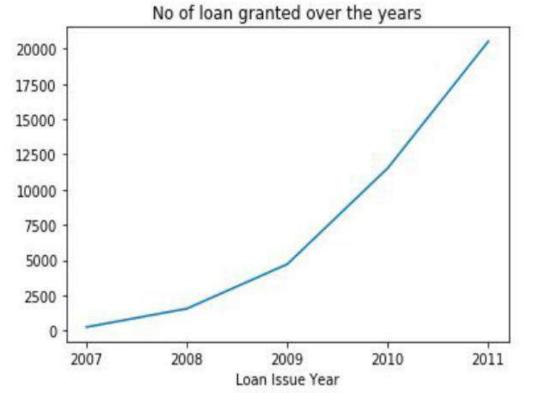
- **1.Loan accepted:** If the company approves the loan, there are 3 possible scenarios described below:
 - **1. Fully paid**: Applicant has fully paid the loan (the principal and the interest rate)
 - 2. Current: Applicant is in the process of paying the instalments, i.e. the tenure of the loan is not 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.). Since the loan was rejected, there is no transactional history of those applicants with the company and so this data is not available with the company (and thus in this dataset)

PROBLEM SOLVING APPROACH

Univariate Segmented Bivariate **Data Preparation &** Recommendations **Data Cleaning Understanding** /summary analysis analysis analysis Removing the null Working with the Data Analysing each Analysing the Do correlation Analysing all plots Dictionary and getting continuous data valued columns, column, plotting analysis and and single or unique knowledge of all the the distributions columns with Check how two Detecting/deciding Which variables or columns and their of each relevant variables affect valued columns, respect to the domain specific uses. column. categorical each other or a columns contribute unnecessary Check third variable. Variables. column. to loan defaulters. Check for outliers in Check for missing distributions and Analyse variables Based on that decide against segments values and missing data. frequencies of value percentage. various of other variables and conclude the Eliminate further numerical analysis and come up Convert values to unnecessary and and categorical with proper int, irrelevant information variables. recommendations to float, date facilitate business representations. from the data set that Create derived variables. was cleaned. growth. Box plots, histograms used.

LOAN STATUS ANALYSIS

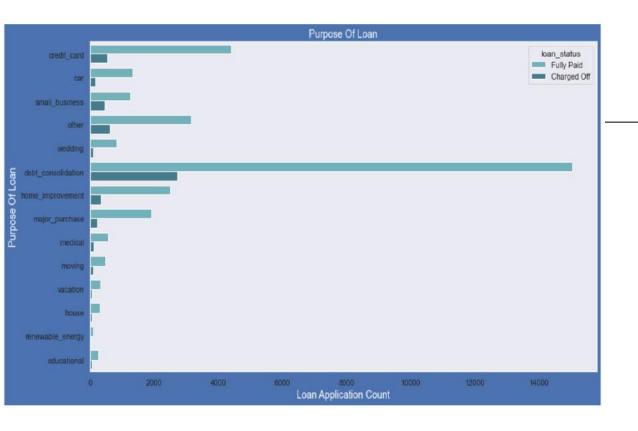


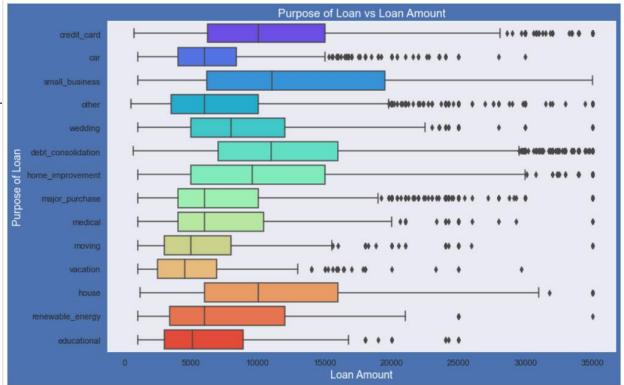


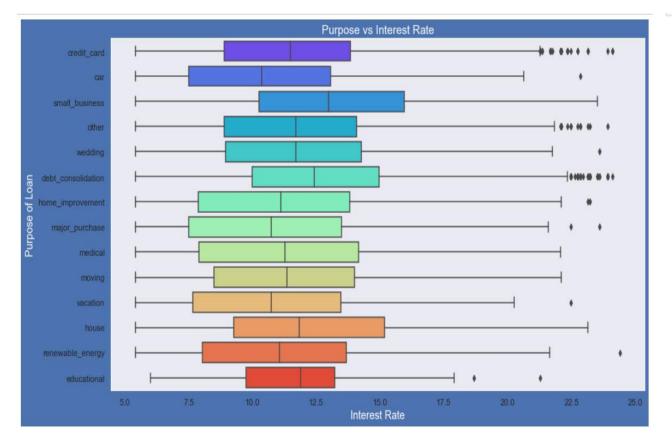
Observations –

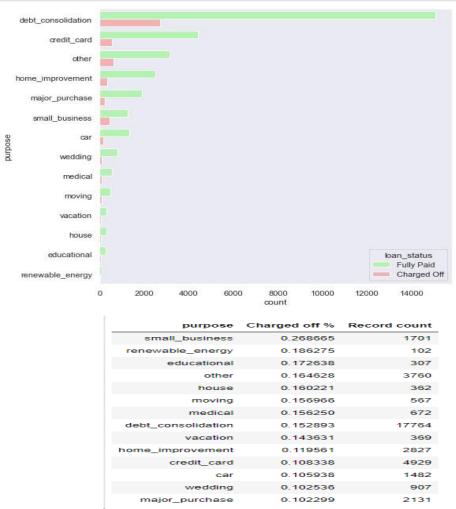
- 1. About 14% of loan are having status as defaulters.
- 2. The loan has been increasing exponentially over the years.

LOAN PURPOSE





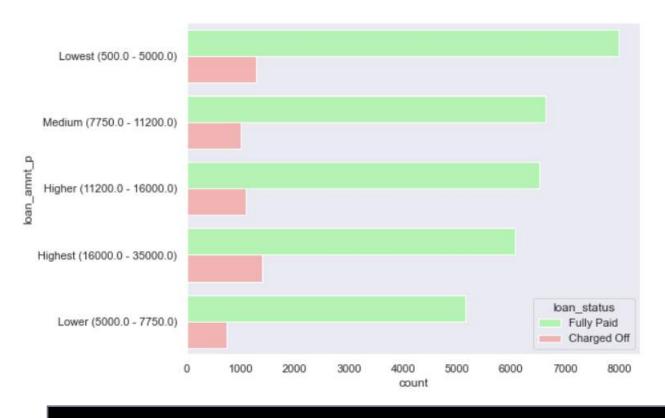




Observations:

- 1. Most of the loans were taken for the purpose of debt consolidation & paying credit card bill.
- 2. Number of charged off count also high too for these loans.
- 3.Loan amount is highest for loan taken for small business purpose among all purposes. Debt consolidation is second and Credit card comes 3rd.
- 4.Loans taken for small business are the risky ones followed by renewable energy.

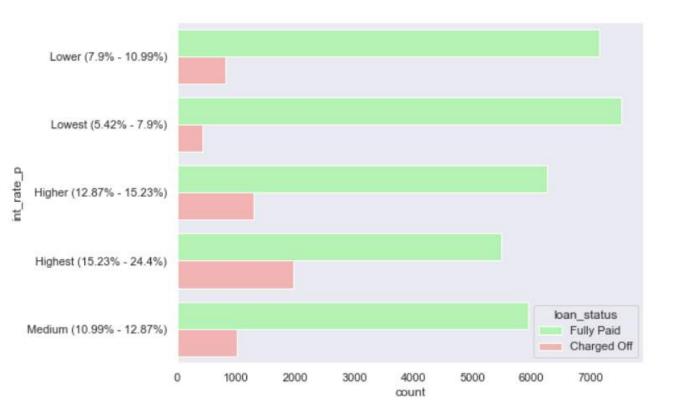
UNDERSTANDING LOAN AMOUNT



loan_amnt_p	Charged off %	Record count
Highest (16000.0 - 35000.0)	0.186009	7462
Higher (11200.0 - 16000.0)	0.144656	7625
Lowest (500.0 - 5000.0)	0.138898	9273
Medium (7750.0 - 11200.0)	0.130589	7627
Lower (5000.0 - 7750.0)	0.124555	5893

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

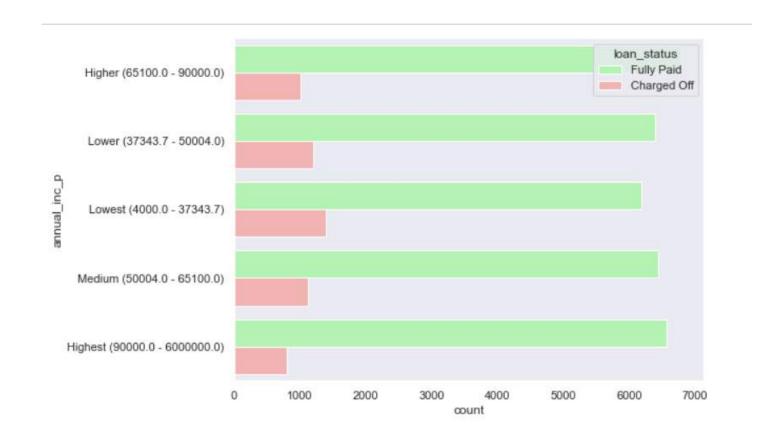
UNDERSTANDING INTEREST RATE



int_rate_p	Charged off %	Record count
Highest (15.23% - 24.4%)	0.263574	7459
Higher (12.87% - 15.23%)	0.172086	7566
Medium (10.99% - 12.87%)	0.144420	6945
Lower (7.9% - 10.99%)	0.102110	7962
Lowest (5.42% - 7.9%)	0.053473	7948

Higher the interest rate leads to higher charged off%

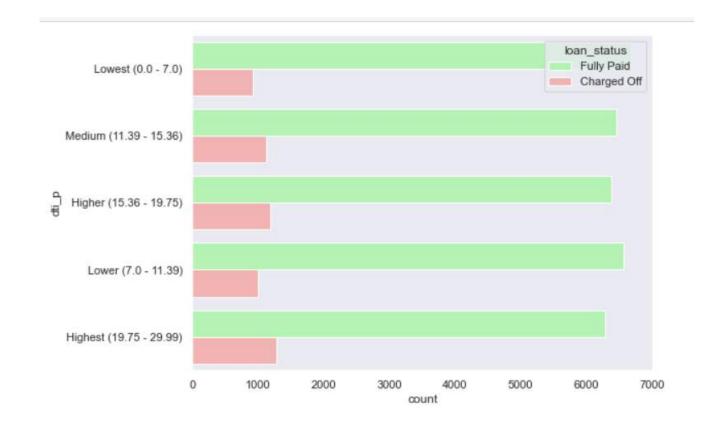
ANNUAL INCOME



annual_inc_p	Charged off %	Record count
Lowest (4000.0 - 37343.7)	0.183474	7576
Lower (37343.7 - 50004.0)	0.157597	7589
Medium (50004.0 - 65100.0)	0.148995	7564
Higher (65100.0 - 90000.0)	0.128767	7797
Highest (90000.0 - 6000000.0)	0.107696	7354

Higher the annual income higher the repayment percentage.

DTI (Debt to Income Ratio)



dti_p	Charged off %	Record count
Highest (19.75 - 29.99)	0.168585	7557
Higher (15.36 - 19.75)	0.156852	7574
Medium (11.39 - 15.36)	0.148524	7588
Lower (7.0 - 11.39)	0.131711	7562
Lowest (0.0 - 7.0)	0.121595	7599

DTI- A ratio calculated using the borrower's total monthly debt payments on the total debt obligations, excluding mortgage and the requested LC loan, divided by the borrower's self-reported monthly income.

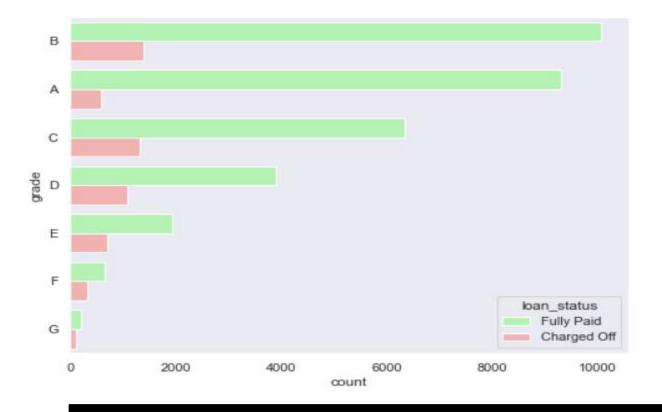
Higher DTI will lead to higher charged off.

LOAN TERM



For loans with 60 months repayment term, the default percent is 25%. For Loans with 36 months loan repayment term, the default is only for 11% of the cases.

LOAN GRADES

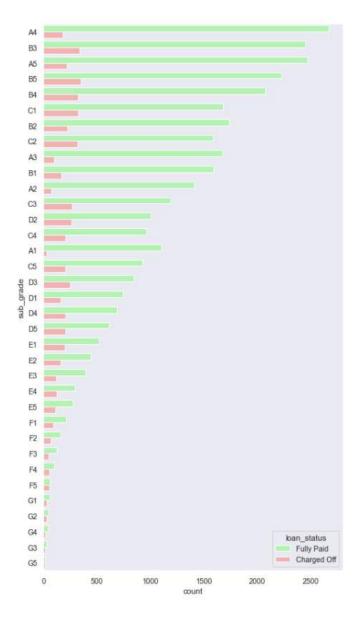


grade	Charged off %	Record count
G	0.340067	297
F	0.325702	961
Е	0.268265	2628
D	0.218134	4974
С	0.171612	7651
В	0.121751	11466
Α	0.060184	9903

Loan grades having highest default percentages - G, F, E and D form grades where default rate is much higher than others. Lower grades have higher incidence of defaults on loans.

The grading system is working!

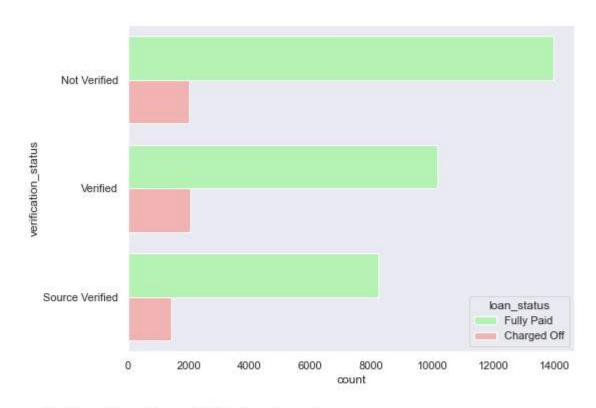




Record count	Charged off %	aub_grade
112	0.473214	F5
45	0.422222	G3
77	0.363636	G2
148	0.358108	F4
26	0.357143	G5
93	0.333333	G1
228	0.302632	F2
415	0.298795	E4
302	0.294702	E1
171	0.286550	F3
382	0.282723	E5
713	0.273492	E1
605	0.264463	E2
820	0.251220	D5
54	0.240741	G4
895	0.231265	D4
513	0.230019	E3
1095	0.226311	D3
1261	0.206979	D2
1453	0.182381	C3
1125	0,179556	C5
903	0.178296	D1
1171	0.175918	C4
1898	0.165964	C2
2004	0.162176	C1
2573	0.135639	B5
2395	0.134864	B4
2789	0.121190	B3
1955	0.112532	82
1754	0.094641	B1
2683	0.080507	A5
2843	0.062258	A4
1771	0.056465	A3
1478	0.049391	A2
1128	0.026596	A1

Loan subgrades G3 and F5 sub categories that have above 40% default rate.

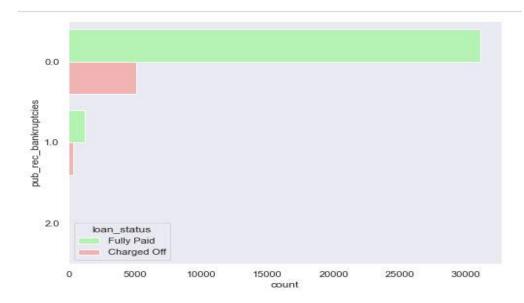
VERIFICATION STATUS



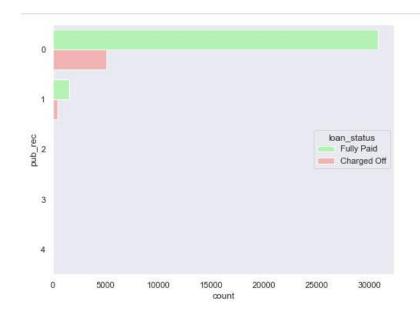
verification_status	Charged off %	Record count
Verified	0.168046	12205
Source Verified	0.148186	9677
Not Verified	0.126516	15998

Verified applicants have a higher charged off percentage . The verification process needs to be validated/ checked or changed.

PUBLIC RECORDS



pub_rec_bankruptcies	Charged off %	Record count
2.0	0.400000	5
1.0	0.223580	1637
0.0	0.141868	36238

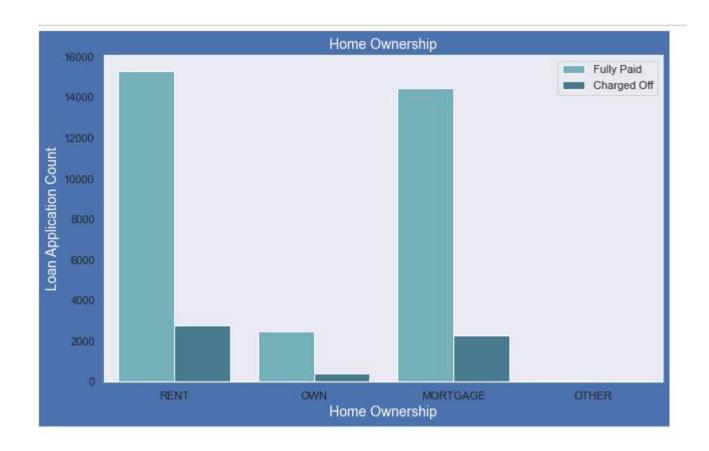


Record count	Charged off %	pub_rec
1994	0.226179	1
47	0.212766	2
35830	0.140888	0

Higher the number of public bankruptcy records, bigger the chance of defaulting the loan. Having even 1 derogatory record increases the chances of Charge Off

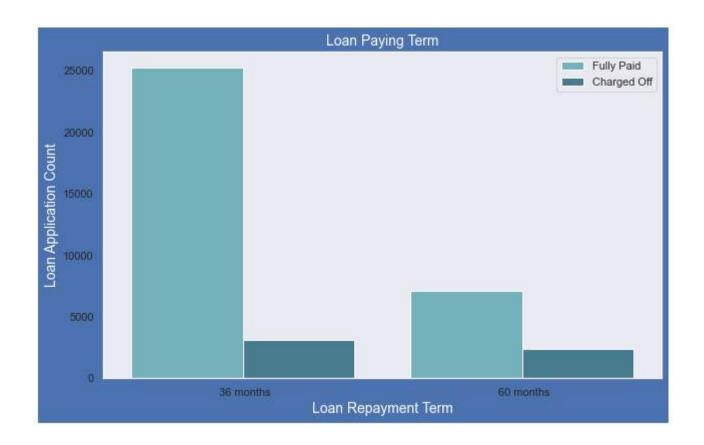
significantly. These two factors are corelated and we can use any one to conclude.

HOME OWNERSHIP



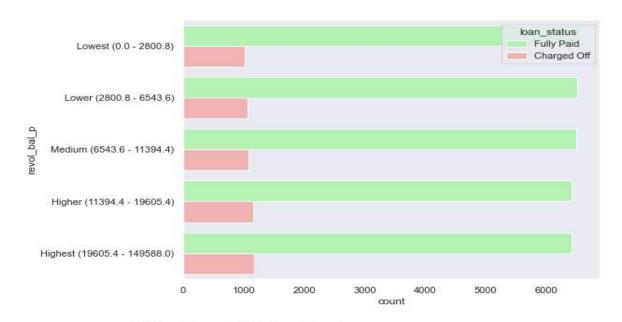
People who live in a rented home or a mortgaged one are most likely to default.

LOAN PAYING TERM



Those who had taken loan to repay in 60 months had more % of number of applicants getting charged off as compared to applicants who had taken loan for 36 months.

REVOLVING BALANCE

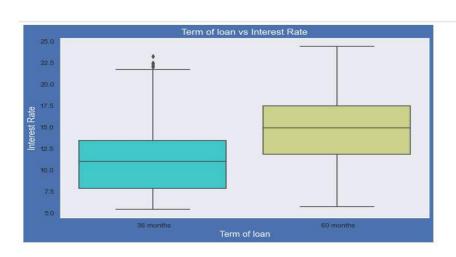


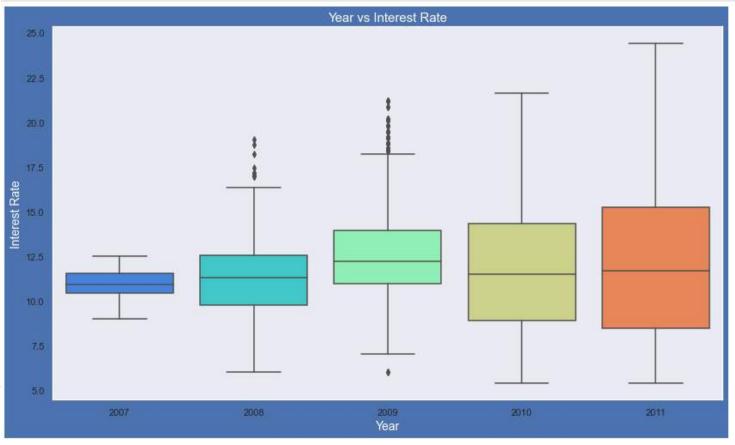
Charged off %	Record count
0.154303	7576
0.153775	7576
0.143347	7576
0.140576	7576
0.135164	7576
	0.154303 0.153775 0.143347 0.140576

Total credit revolving balance - Higher the revolving balance, bigger the chance of the loan getting defaulted. People with high utilization of Revolving Line of Credit at the time of taking loan default more.

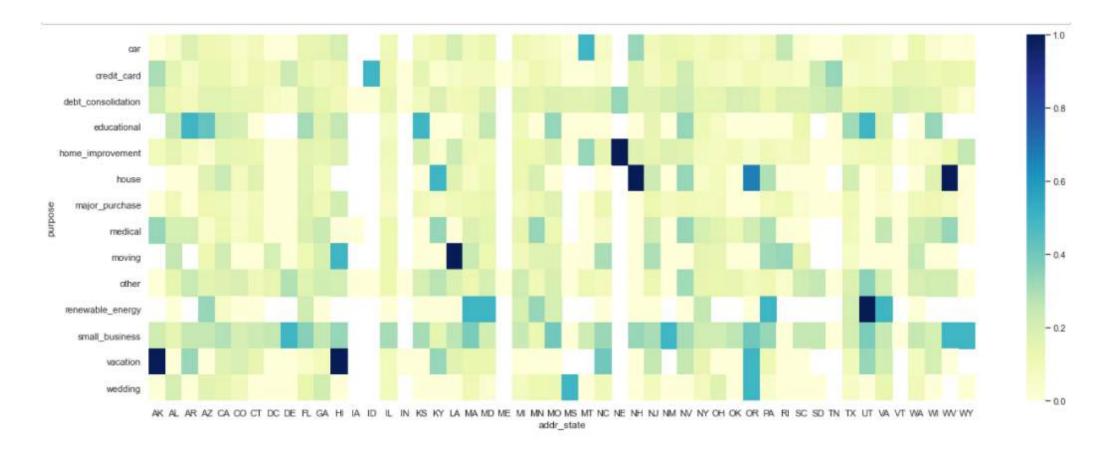
Loans with utilization > 75% are risky.

Plot shows interest rate is increasing slowly with increase in year. Average interest rate is higher for 60 months loan term. Most of the loans issued for longer term had higher interest rates for repayment.





ADDRESS STATE



As per the above plot, the darker the intersection of addr_state has with the purpose of the loan, the risker the loan application is. Some of the examples are below: •vacation loans in AK, HI, OR •education loans in AR, KS, UT •small business loans in DE, NM, WV, wY •wedding loans in MS, OR

RECOMMENDATIONS

- 1.Lending club should reduce the high interest loans for 60 months tenure, they are prone to loan default.
- 2.Grades are good metric for detecting defaulters. Low grade loans have high tendency to default. Grading system is working as expected. Lending club should examine more information
- from borrowers before issuing loans to Low grade (G to A).
- 3.Small business loans are defaulted more. Lending club should stop/reduce issuing the loans to them. Reduce the number of approvals where purpose is small business. A deep scrutiny can also be done when the purpose is Debt consolidation.
- 4. Borrowers with mortgage home ownership are taking higher loans and defaulting the approved loans.
- 5.People with more number of public derogatory records are having more chance of filing a bankruptcy. Lending club should make sure there are no public derogatory records for borrower. Stop approving loans to people with prior bad record.
- 6.Loans having higher interest rate have more defaulters. Check the background of applicant thoroughly if interest rate is high.
- 7. When approving loans for a high amount, check the background of the applicant thoroughly as the default rate is directly proportional to amount.
- 8. Prefer approving loans for people with higher incomes for profit making.
- 9. Start charging higher interest rates for loans with DTI greater than 20
- 10. Verified applicants have a higher charged off percentage. The verification process needs to be validated/checked or changed.

SUMMARY

VARIABLES HAVING A MINOR EFFECT

Higher loan amount

- Higher installment amount
- Lower annual income
- Higher debt to income ratio
- •Applicant's address state (NV, SD, AK, FL, etc.)
- Loan issue month

VARIABLES HAVING A MAJOR EFFECT

- Higher interest rate
- •Higher revolving line utilization rate.
- •Repayment term (5 years)
- Loan grade & sub-grade (D to G)
- Loan purpose (small business, renewable energy, educational)
- Derogatory public records (1 or 2)
- •Public bankruptcy records (1 or 2)

COMBINED VARIABLES

- •High loan amount & interest rate for lower income group
- •High installment and longer repayment term
- •Home ownership (other) and loan purpose (car, moving or small business)
- Income group and loan purpose