

Approach

DATA IMPORTING	DATA CLEANING	DATA ANALYSIS	DATA VISUALIZATION
Imported Data using pandas(pd.read_csv)	 NA's analysis Duplicate Changing the class of observation. 	 Exploratory data analysis –Question the data to perform: Univariate* analysis of both categorical and continuous variables. 	 Using Matplotlib & Seaborn to create graphs that aid in Defining the issues Analysis (univariate, bivariate) Segmentation analysis
Tools Used Pandas Numpy Matplotlib.pyplot Seaborn	Formatting and standardizing date time –(issue_d), percentage etc.	Bivariate analysis* – •Categorical vs categorical •Categorical vs continuous •Continues vs continuous	2. Communicate inferences, understanding with supporting analysis and graphs to decision making audience and any larger audience.
	Creating derived metrics -	Correlation matrix.	

Creating derived metrics

- •Charge off amount
- Charge off as a percentage of funded amount

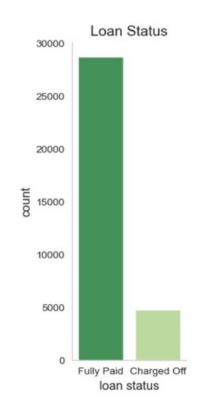
Correlation matrix.

**Univariate analysis, Bivariate analysis includes Derived metrics as well

We are going to work with the loan data

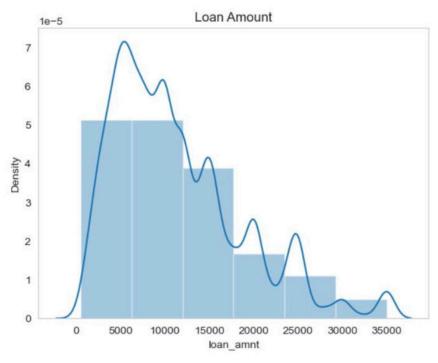
	Louistativew	Description
2	acc_now_delinq	The number of accounts on which the borrower is now delinquent.
3	acc_open_past_24mths	Number of trades opened in past 24 months.
4	addr_state	The state provided by the borrower in the loan application
5	all_util	Balance to credit limit on all trades
6	annual_inc	The self-reported annual income provided by the borrower during registration.
7	annual_inc_joint	The combined self-reported annual income provided by the co-borrowers during registration
8	application_type	Indicates whether the loan is an individual application or a joint application with two co-borrowers
9	avg_cur_bal	Average current balance of all accounts
10	bc_open_to_buy	Total open to buy on revolving bankcards.
11	bc_util	Ratio of total current balance to high credit/credit limit for all bankcard accounts.
12	chargeoff_within_12_mths	Number of charge-offs within 12 months
13	collection_recovery_fee	post charge off collection fee
14	collections_12_mths_ex_med	Number of collections in 12 months excluding medical collections
15	delinq_2yrs	The number of 30+ days past-due incidences of delinquency in the borrower's credit file for the past 2 years
16	delinq_amnt	The past-due amount owed for the accounts on which the borrower is now delinquent.
17	desc	Loan description provided by the borrower
18	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.
19	dti_joint	A ratio calculated using the co-borrowers' total monthly payments on the total debt obligations, excluding mortgages and the requested LC loan, divided by the co-borrowers' combined self-reported monthly income
20	earliest_cr_line	The month the borrower's earliest reported credit line was opened
21	emp_length	Employment length in years. Possible values are between 0 and 10 where 0 means less than one year and 10 means ten or more years.
22	emp_title	The job title supplied by the Borrower when applying for the loan.*
23	fico_range_high	The upper boundary range the borrower's FICO at loan origination belongs to.
24	fico_range_low	The lower boundary range the borrower's FICO at loan origination belongs to.
25	funded_amnt	The total amount committed to that loan at that point in time.
26	funded_amnt_inv	The total amount committed by investors for that loan at that point in time.
27	grade	LC assigned loan grade
28	home_ownership	The home ownership status provided by the borrower during registration. Our values are: RENT, OWN, MORTGAGE, OTHER.
29	id	A unique LC assigned ID for the loan listing.
30	il_util	Ratio of total current balance to high credit/credit limit on all install acct
31	initial_list_status	The initial listing status of the loan. Possible values are – W, F
32	inq_fi	Number of personal finance inquiries
33	inq_last_12m	Number of credit inquiries in past 12 months

Univariate Analysis

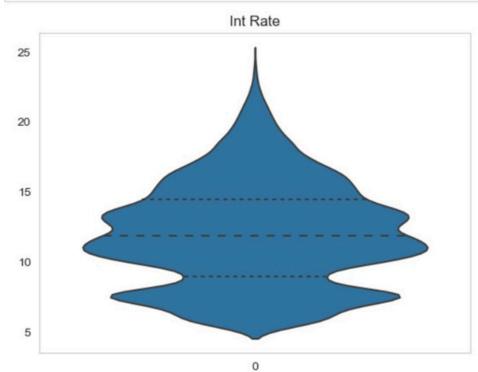


The graph shows that around 5000 people have defaulted/charged off.

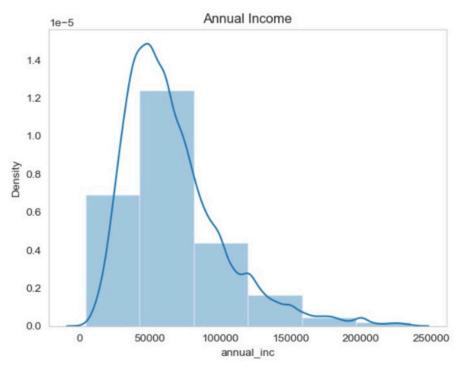
Distribution of loan amount



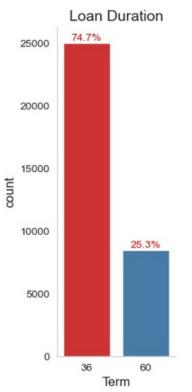
The graph above shows that the loan amount is majorly spread around say 6000 to 18000 approximately



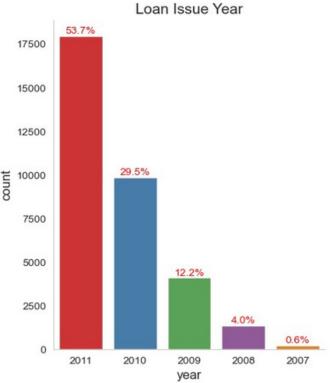
The above graph shows that the interest rate is spread majorly between 8% and 14% approximately



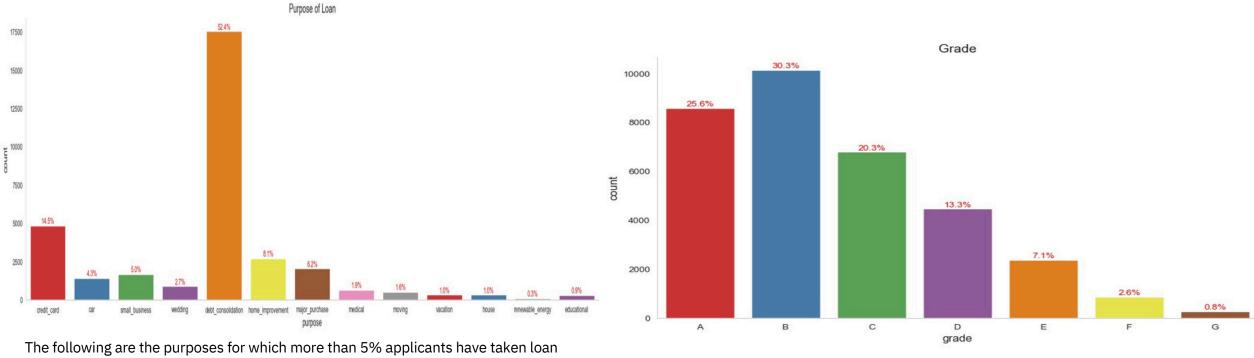
As per the above graph, majority of the applicants have an annual income ranging approximately between 40000 USD to 90000 USD.



As per the above graph, majority of the applicants have taken the loan duration as 36 months (74.7%)



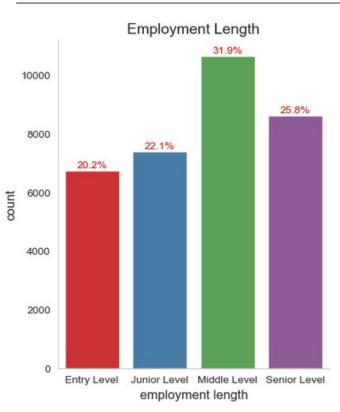
Per the above graph, applicants for loan increased as the years increased. In 2011 the number of applicants for loan was 53.7%. Since the variable issue year does not provide us any direction in the analysis, we would not be using this variable for any further analysis.

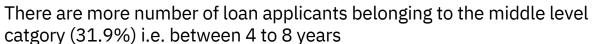


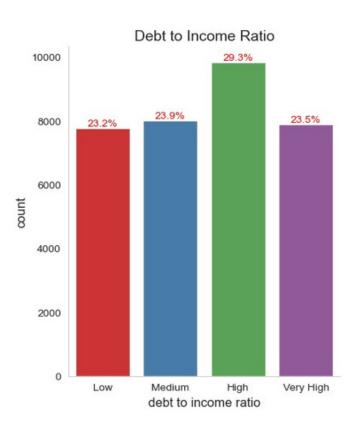
- debt_consolidation 52.4%
- credit_card 14.5%
- home_improvement 8.1%
- major_purchase 6.2%
- small business 5.0%

The above graph shows that most of the applicants fall under the grade B(30.3%), followed by A(25.6%) and C(20.3%)

Univariate Analysis





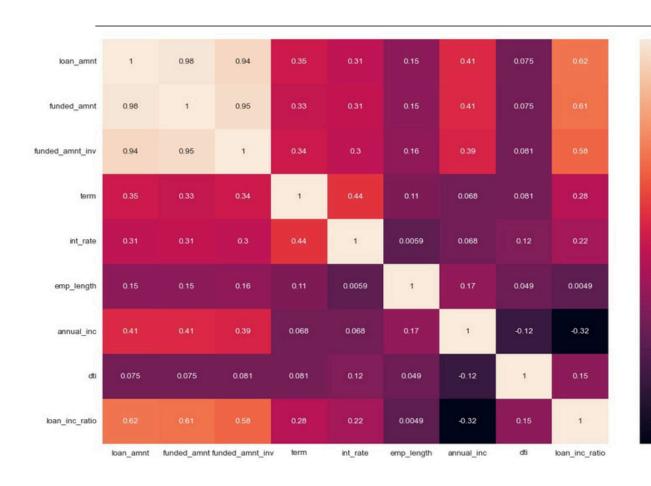


Analysis – Key Graph 6 Bivariate Analysis

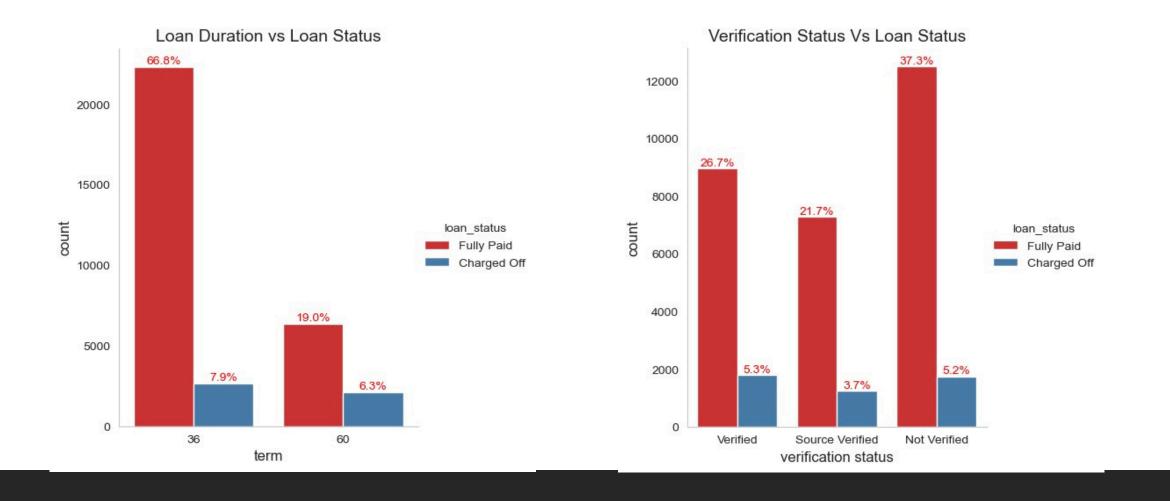
- 0.8

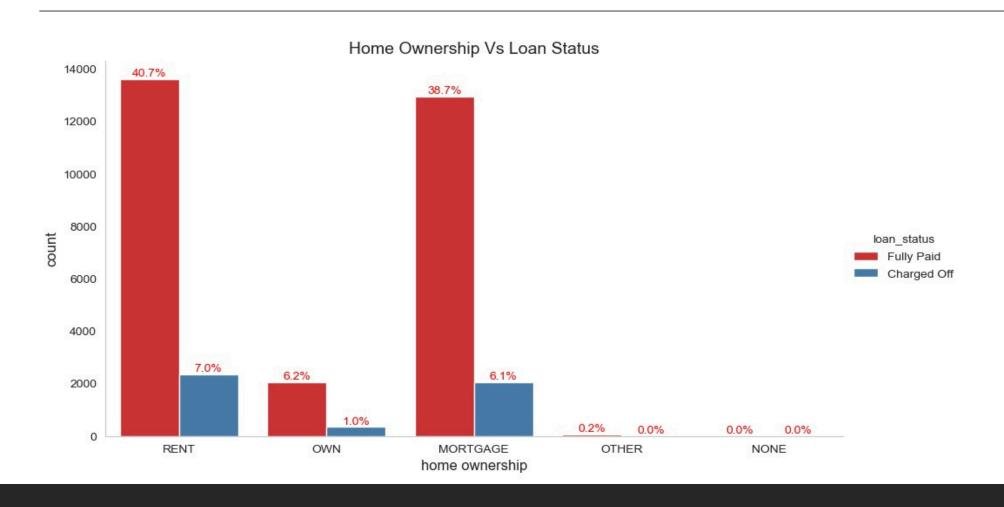
- 0.6

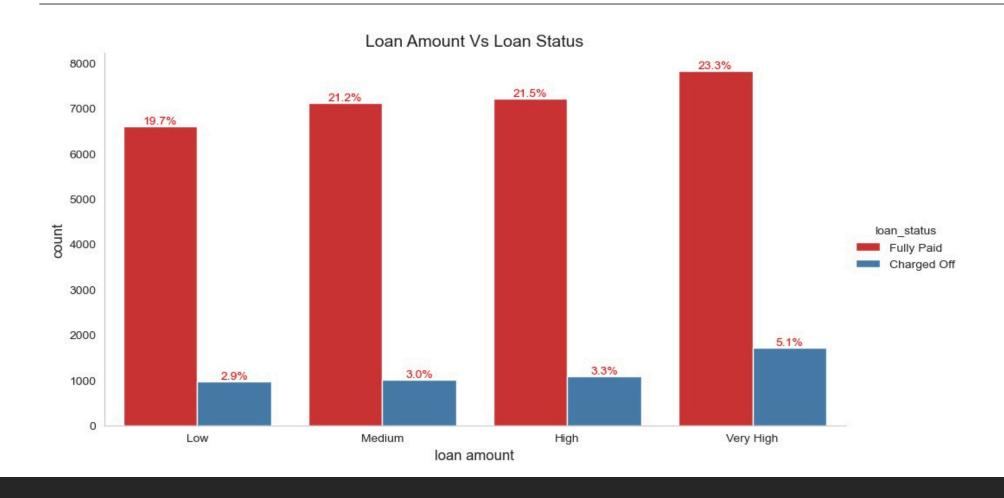
- 0.4

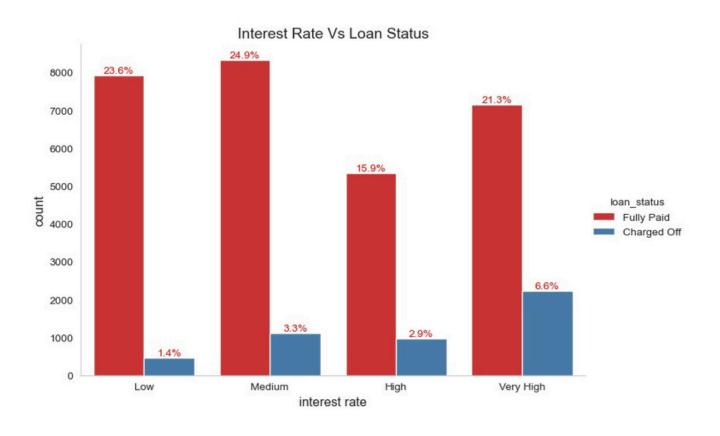


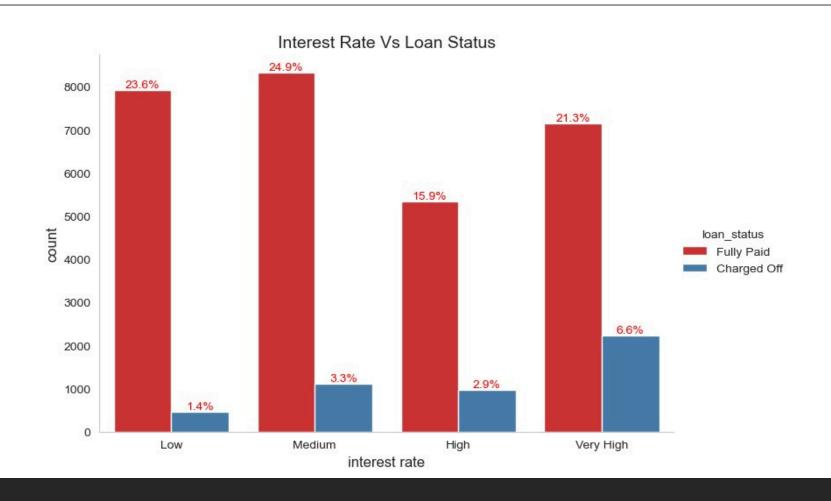
The heat map shows that the loan_amount, funded_amount and funded_amount_inv are very closely correlated. Hence we can safely take any one of the fields from the above 3 fields for our analysis.

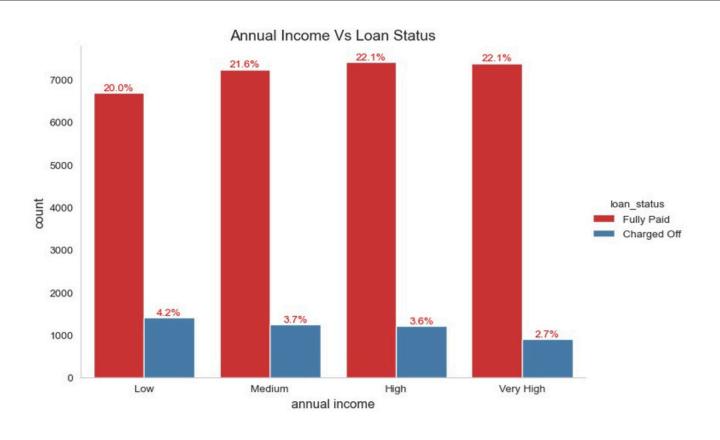


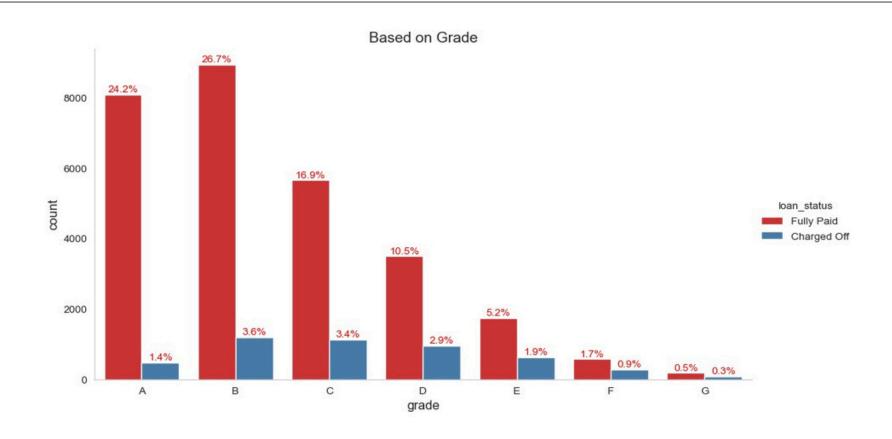


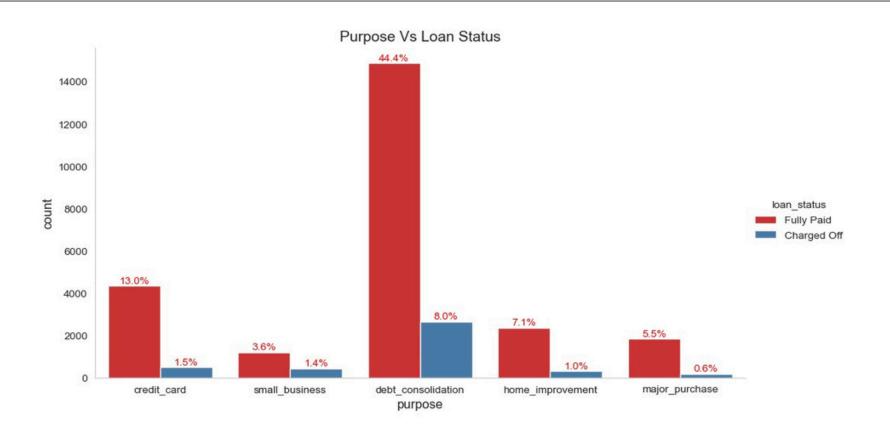


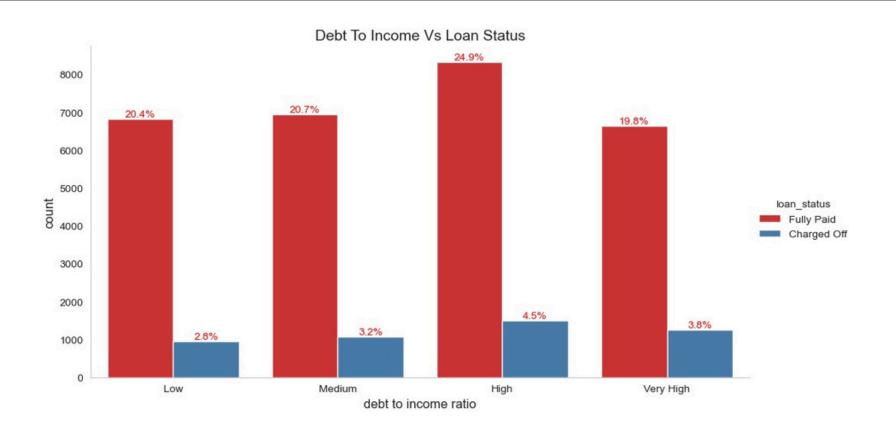












THANK YOU