

# **LENDING CASE STUDY SUBMISSION**

Sharukh Shaikh  
Jawaad Shaikh



# Analysis Overview

- **Contents**

- Analysis Anecdote.
- About the Data
- Cleaning Chronology
- Univariate Analysis
- Bivariate Analysis
- Driving Factors on Defaulting on a Loan
- Conclusion Pointers

- **Goals**

- A consumer finance company intends to identify risks pertaining to loan defaulters
- Analyzing applicants who are likely to repay the loan
- Analyzing applicants who are likely to default on loans
- Identify Driver Variables that cause an applicant to default
- Quantify the magnitude of probability for an applicant to default

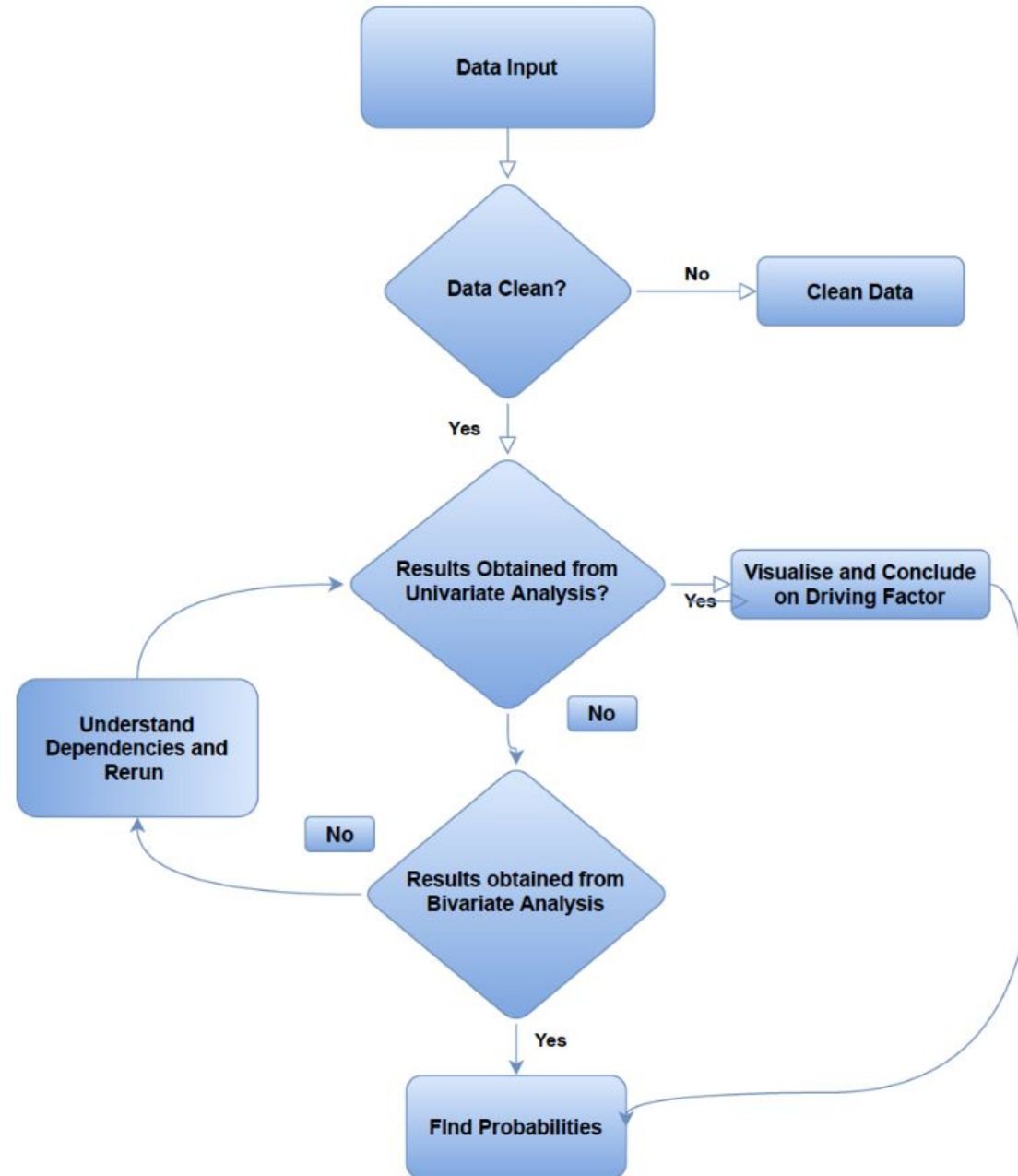
- **Assumptions**

- Historical dataset shall be used to infer on a singular application
- The dataset contains accepted applications and the respective status of the loan



# About The Data

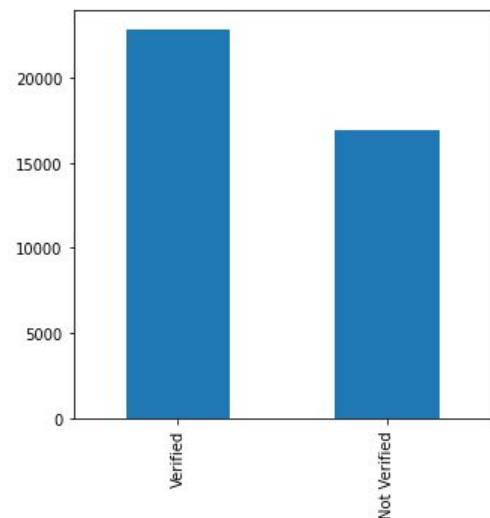
- **loans**: Table including various lending details
- **Data-dictionary**: *Metadata* about the column names
- **Loans** dataset contains applicants' loan status, under three categories
  - **Fully Paid**: Applicant has fully paid the loan
  - **Current**: Applicant is in the process of paying the instalments
  - **Charged Off**: Applicant has not paid the instalments in due time for a long period of time
- Based on the loan status, there are different associated variables such as:
  - Annual Income of Applicants
  - Reason for Loan and Others
- Referring to the loan status and the associated variables, the analysis is conducted to understand the “driver variables” to understand different factors that cause an individual to default on loans.



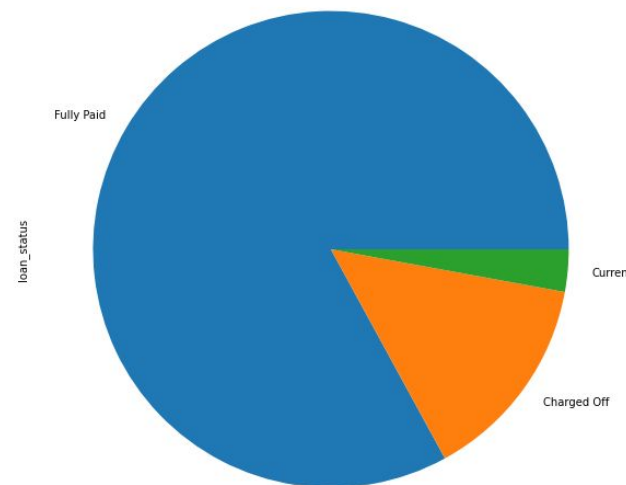
- Removed columns / dimensions which have **real values less than 2**;
  - ***63 columns removed, 47 columns remain***
- Checking columns for **junk values greater than 90%**;
  - ***2 columns removed, 45 columns remain***
- Removing Columns having **% missing data greater than 30%**;
  - ***2 columns removed, 43 columns remain***
- Corrected Value errors such as:
  - Object / String values to integer and float (***converting to proper dtypes***)
  - ***Date – Time Corrections***
  - Eliminated String Errors such as; ***Interest Rate is 5%; removed %***



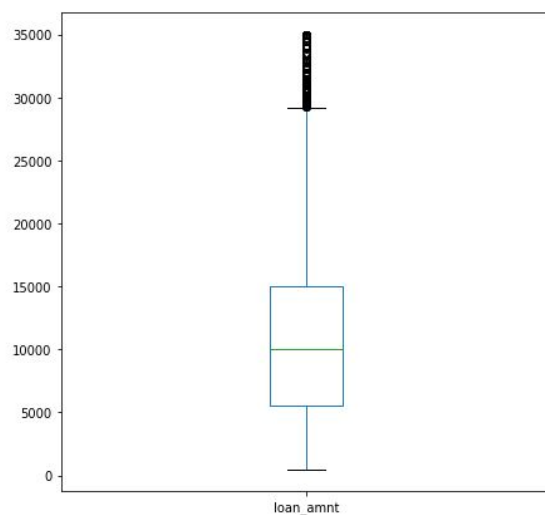
# Univariate Analysis - 1



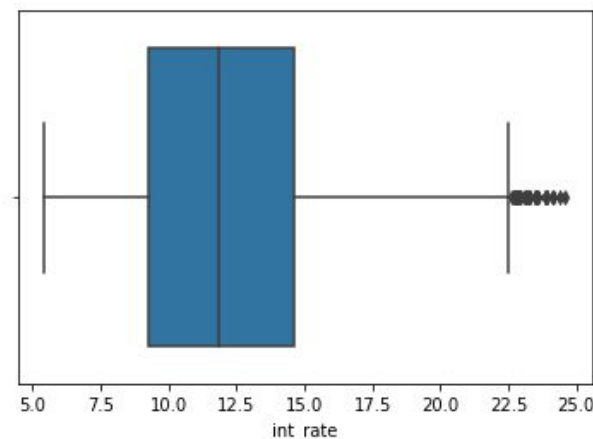
**Most Accepted Applicants have their source verified for a specific loan**



**Most of the loan applicants have fully paid their loans**



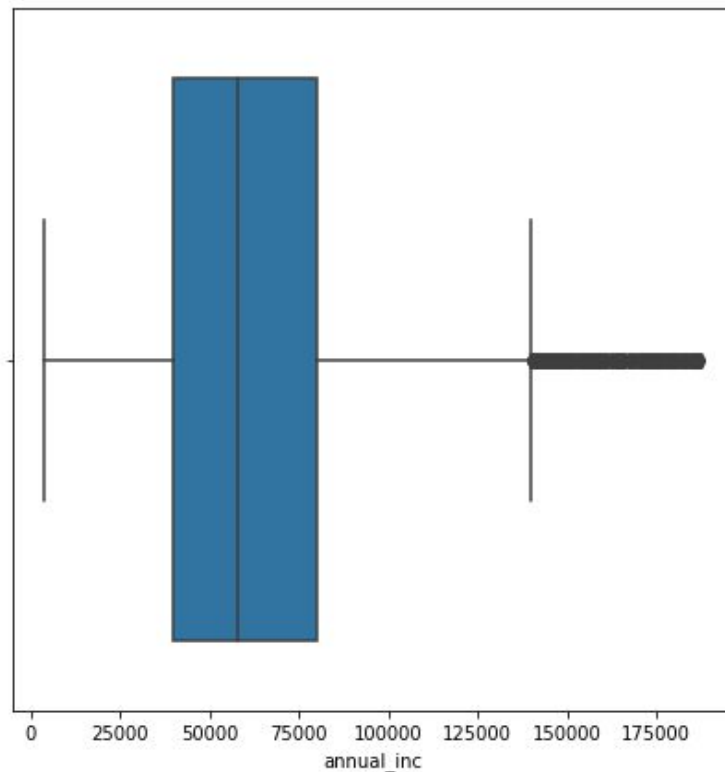
**The loan amount is heavily varying between 5k to 15k USD; with some outliers**



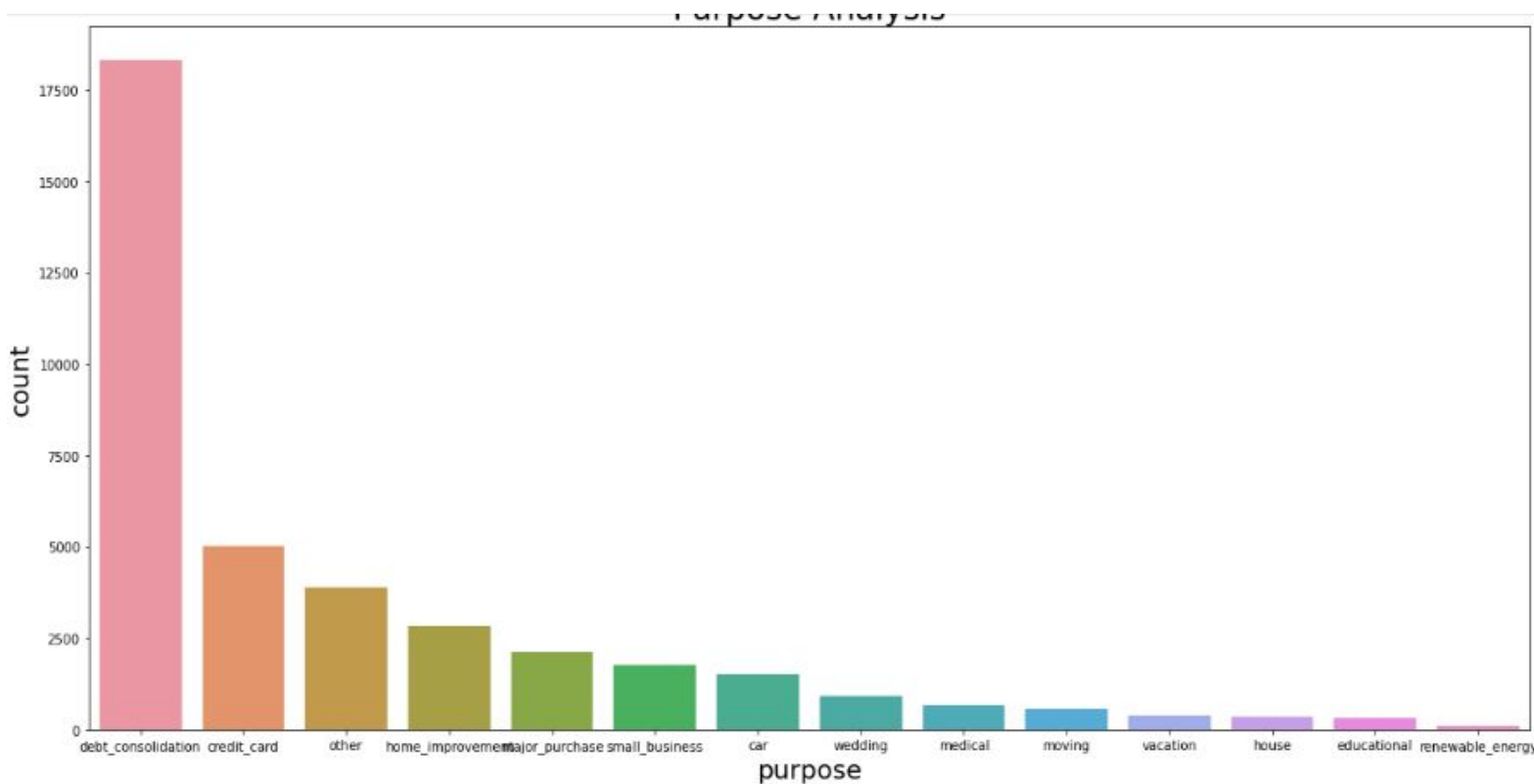
**The interest rate varies from 9% to 15%; however, there are cases where the interest rate is shot up to 25%.**



## Univariate Analysis - 2

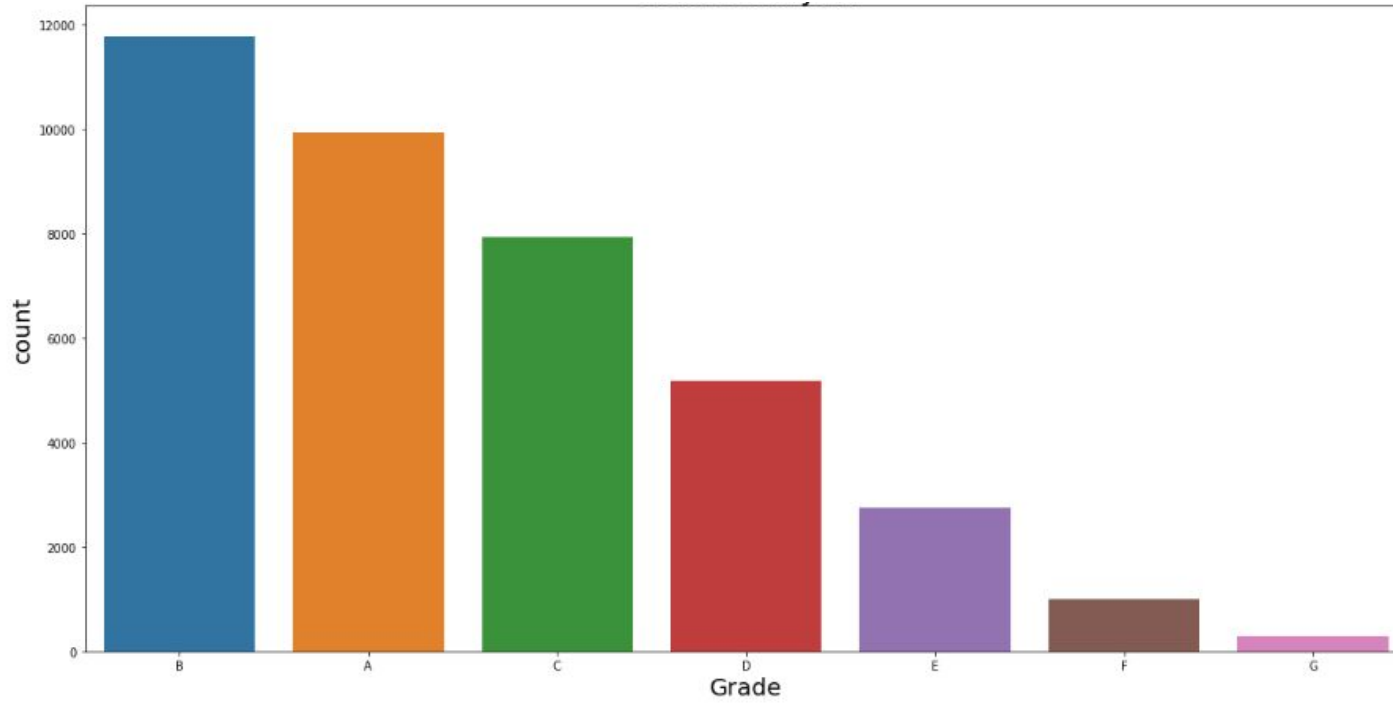


A most alarming driver is noted to be annual income onto the loan applicants. The **nominal income ranges between 30k to 80k USD**.

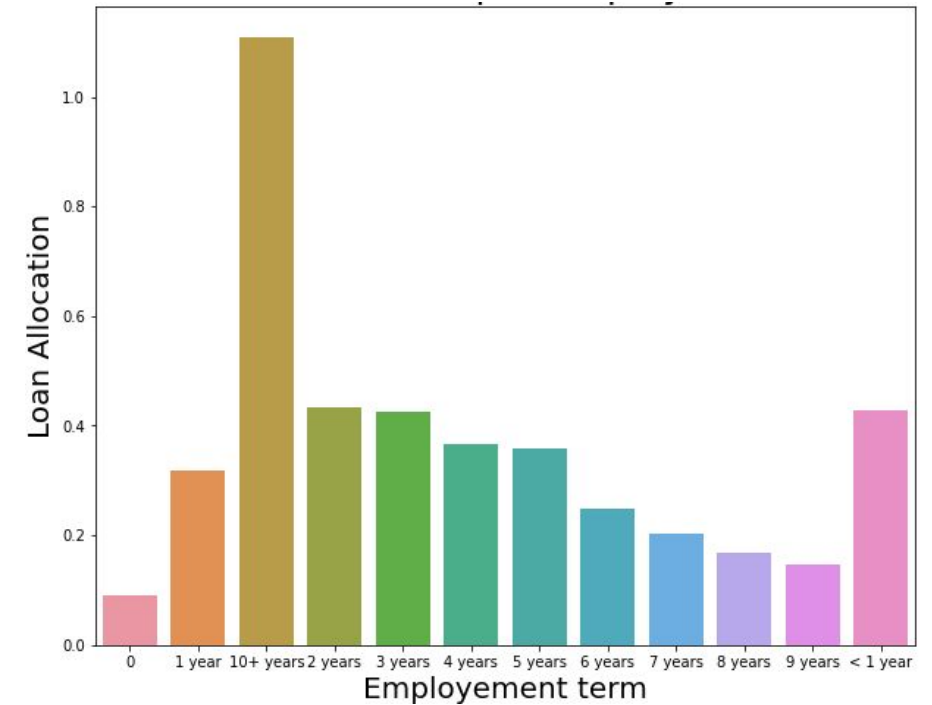


Another dominating factor is the reason for applying for loan. Majority of the **applicants are debt – ridden**. This factor should be carefully accounted in clearing an individual's application.





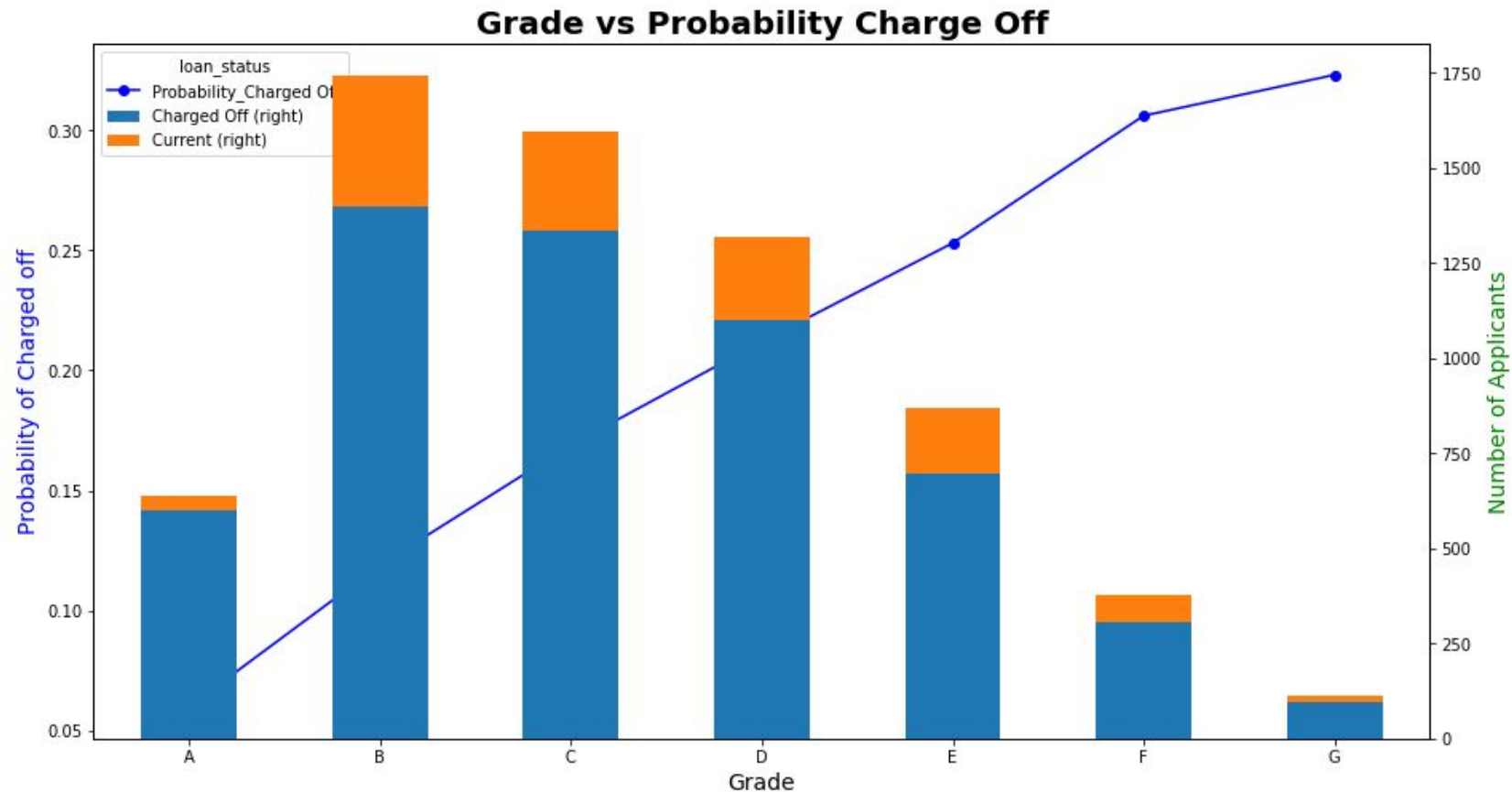
Based on the assigned grades on an individual's application. It is noted that ***most applications are grade as "B"***, which is closely followed by A and C



The highest number of applications are noted from those individuals whose working ***experience is greater than 10 years***

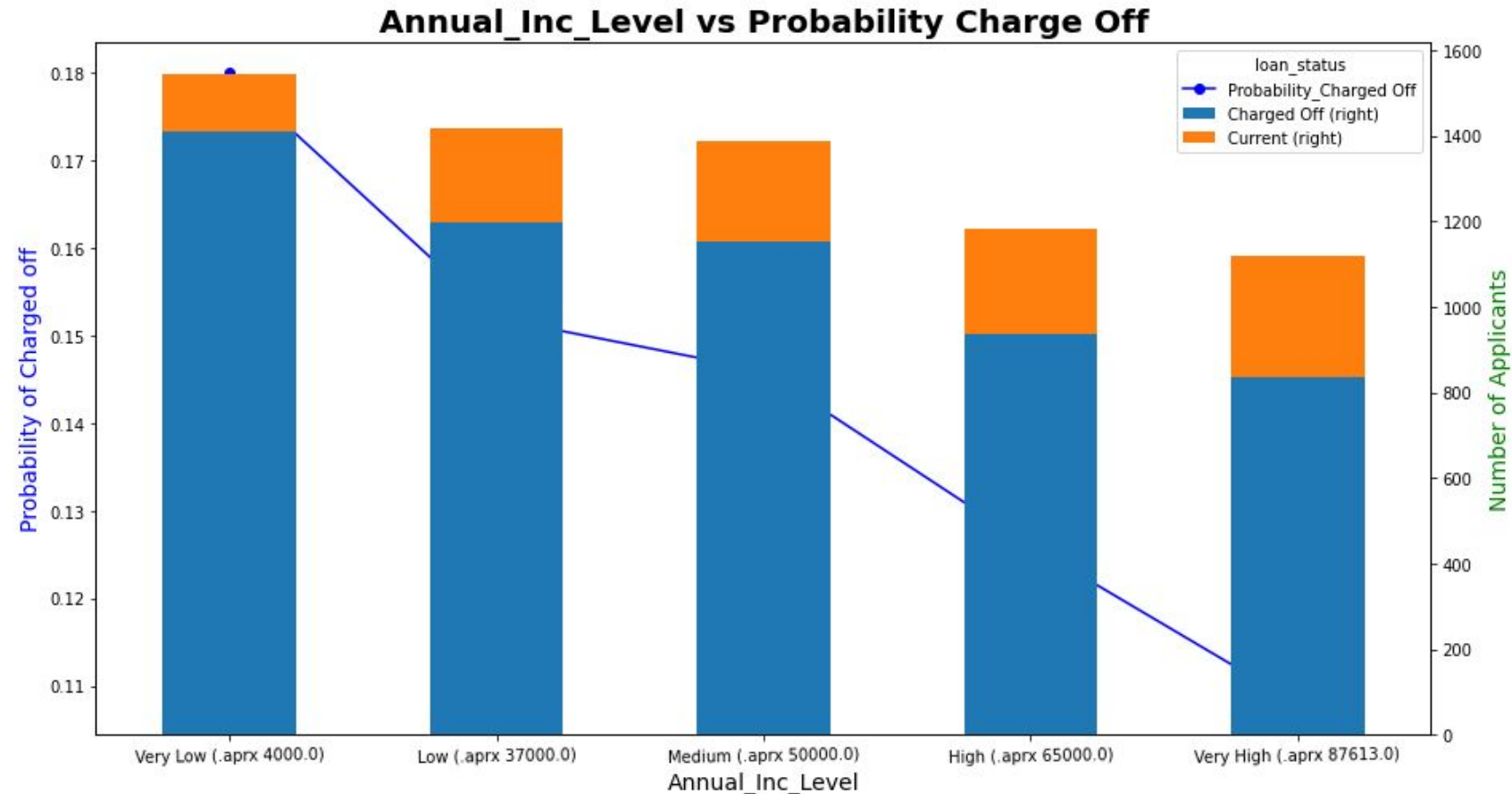


- The interest rate and the amount is highly correlated, indicating, ***more the amount greater the interest.***
- The installments are greater if the loan amount is greater.



**The probability of an individual to default on loans increases for grades after D**

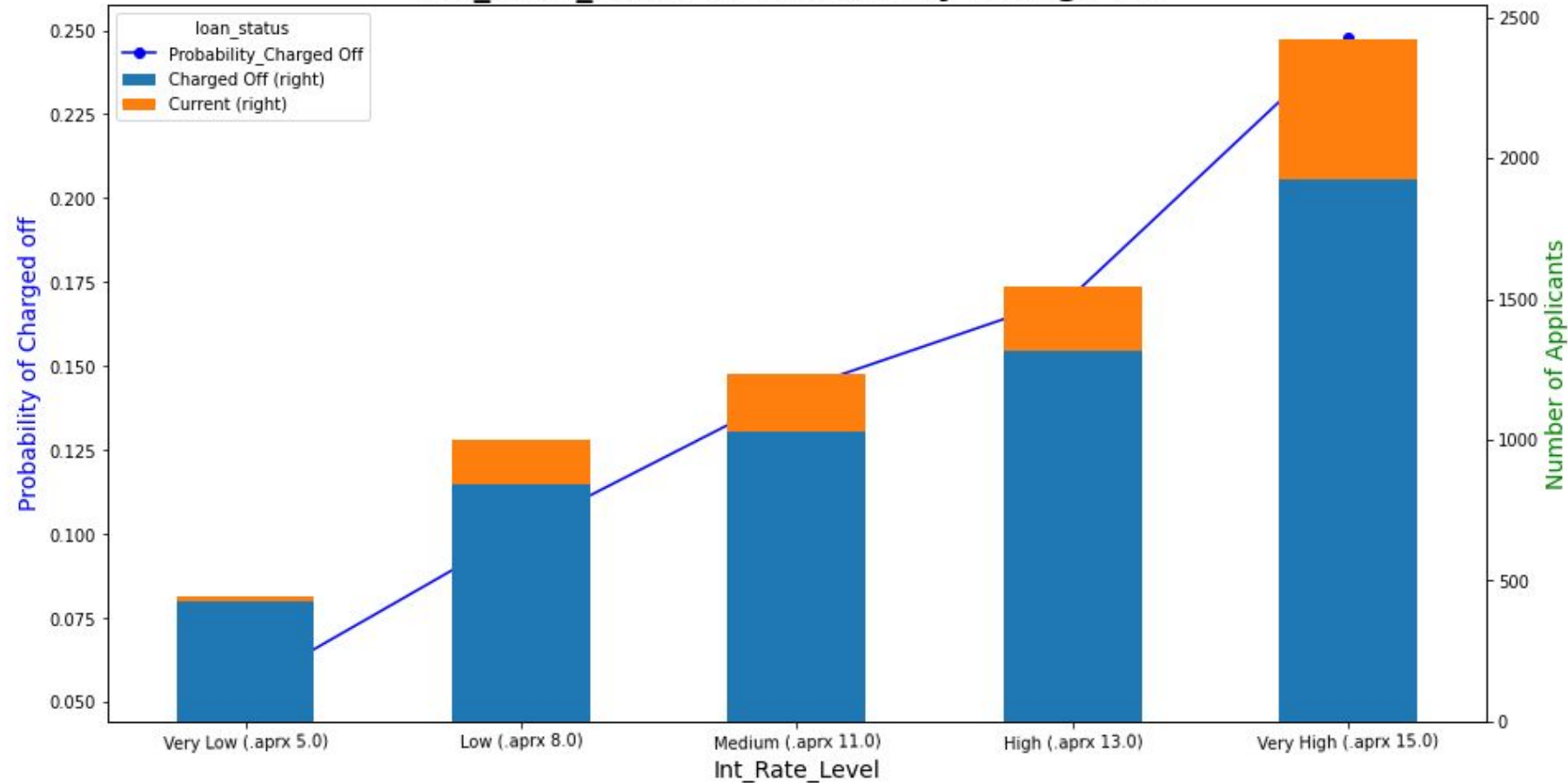
# Probabilities – Annual Income v/s Loan Defaults



**The probability of an individual to default on loans increases for individuals with low annual incomes;  
i.e below 4k USD**

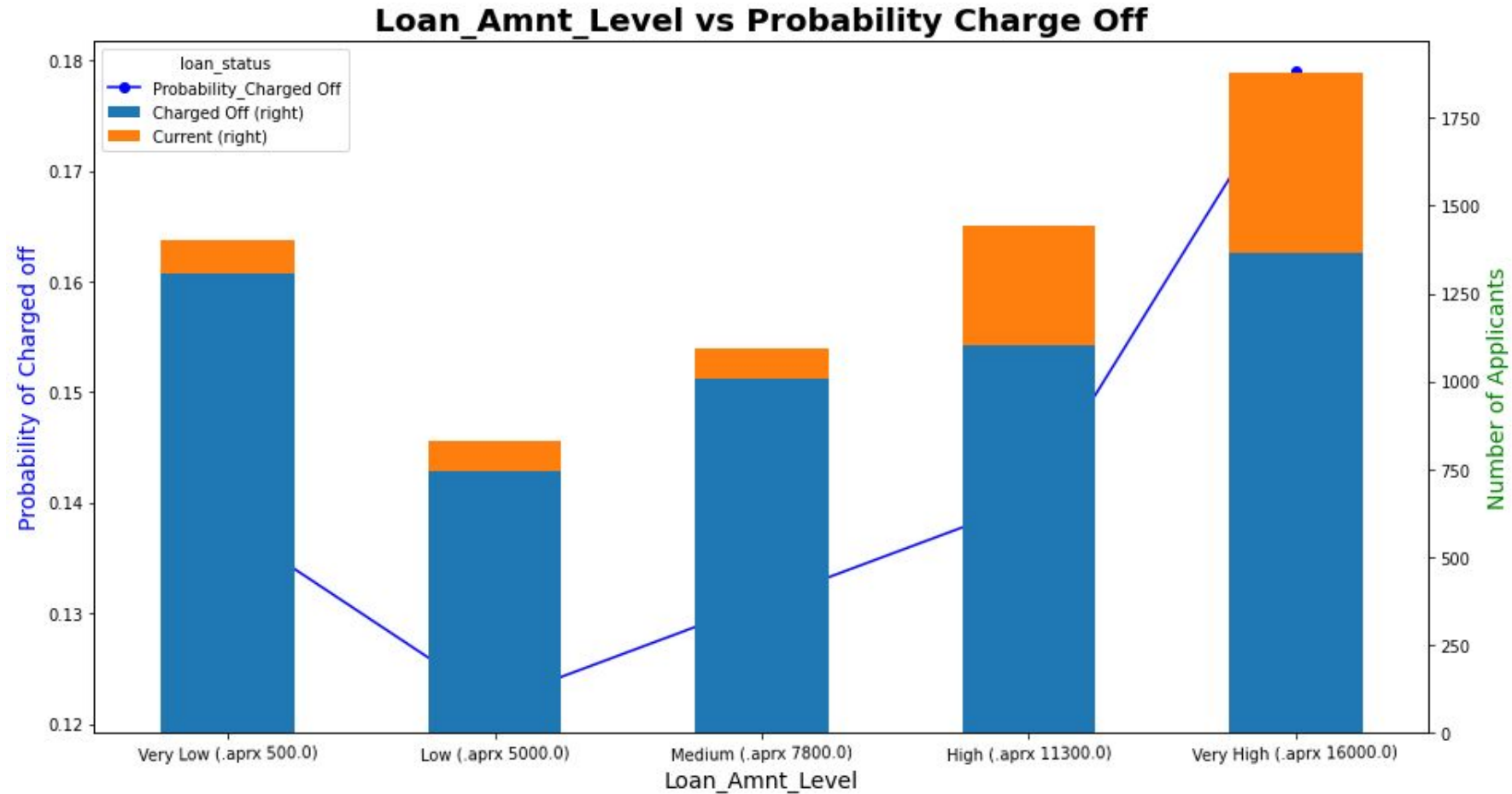
# Probabilities – Interest Rates v/s Loan Defaults

Int\_Rate\_Level vs Probability Charge Off



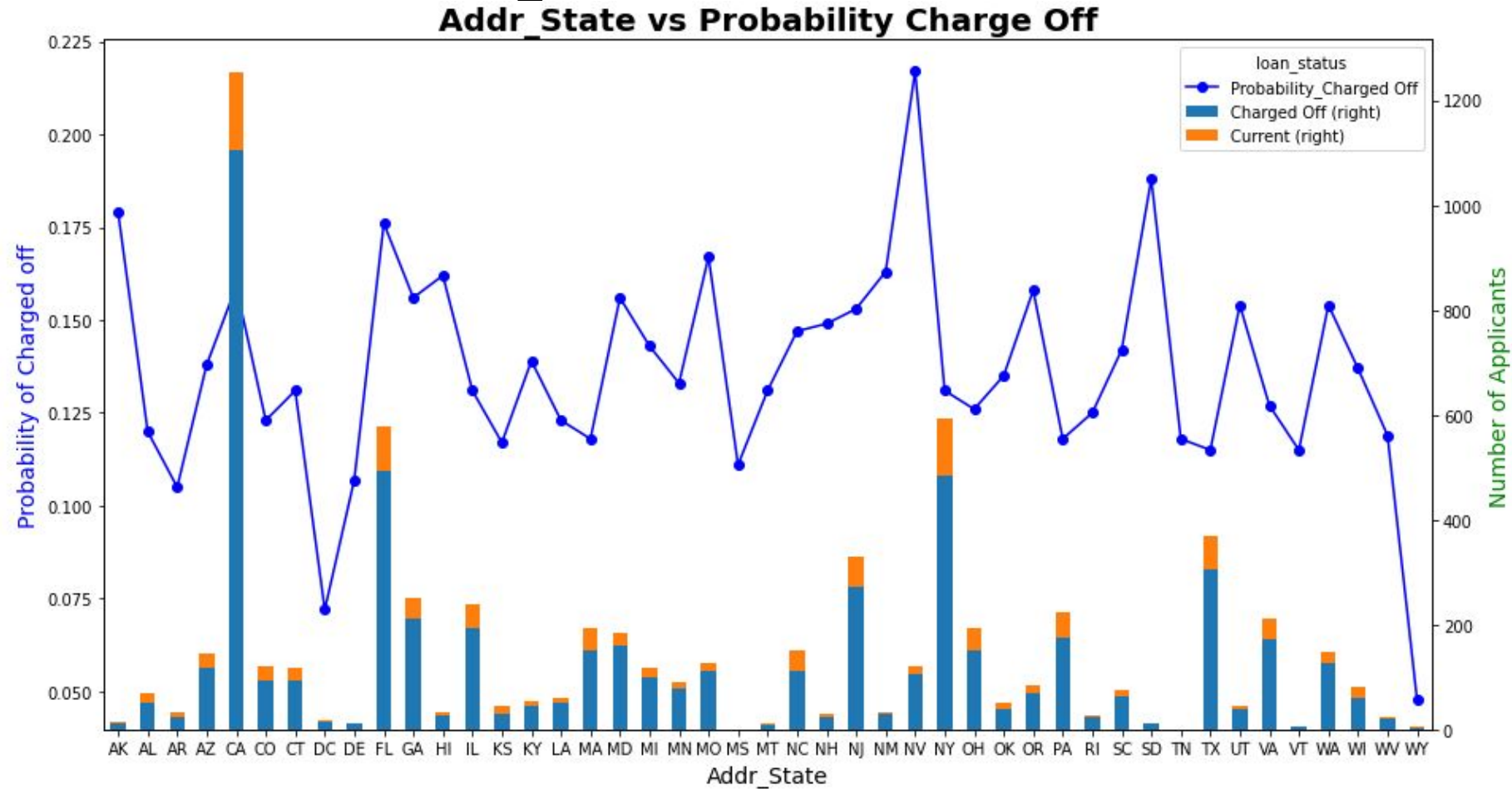
**The probability of an individual to default on loans increases for individuals with high interest rates; i.e interest rates greater than 13%**

# Probabilities – Loan Amount v/s Loan Defaults



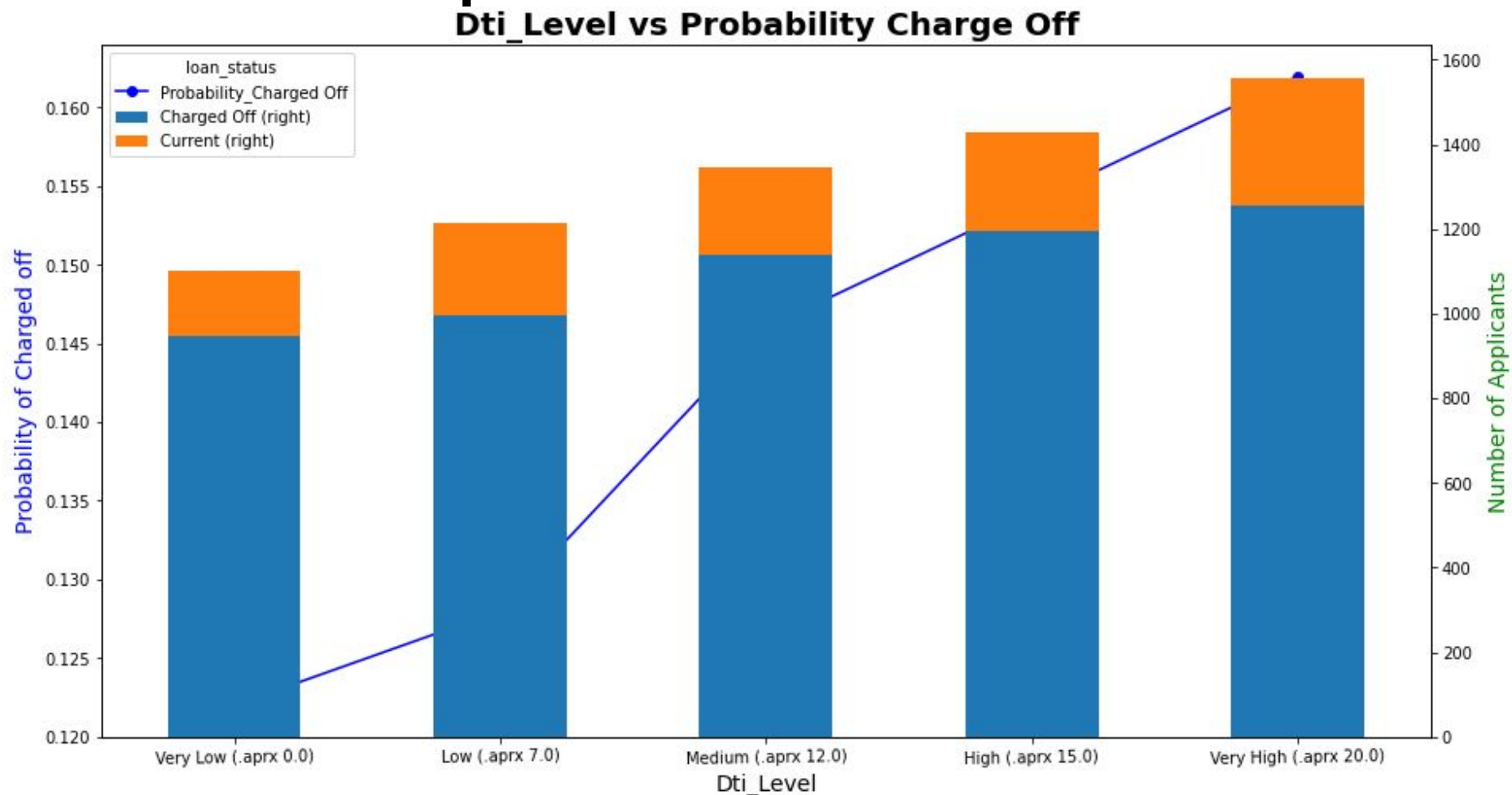
**The probability of default is increased with higher loan amounts; however, it is observed even for loan amounts lesser than 500USD, there are significant defaulters**

# Probabilities – Region v/s Loan Defaults



**Nevada (NV) has the highest defaulters, followed by South Carolina (SC) and Florida (FL)**

# Probabilities – Impact of DTI



**As the DTI (Debt to Income Ratio) increases, the probability of defaulting increases**



We have performed different types of analysis to find few of the important factors/variables that needs to be considered to determine is an applicant will default or not. Those variables are as follows:-

1. Grades (grade)

From A to G including the subgrades the trend is the probability of being a loan default increases.

2. Annual Income (annual\_inc)

It has a negative co-relation, the more the income, the lesser chance to default the

3. Interest Rate (int\_rate)

It is positively co-related, the more the interest rate the more chance of an applicant to be a loan defaulter

4. Loan Amount (loan\_amnt)

It is positively co-related, the more the loan amount, the more there is a chance of an applicant to be a loan defaulter

5. Debt To Income ratio (dti)

It is positively co-related, the more the DTI, the more there is a chance of an applicant to be a loan defaulter