

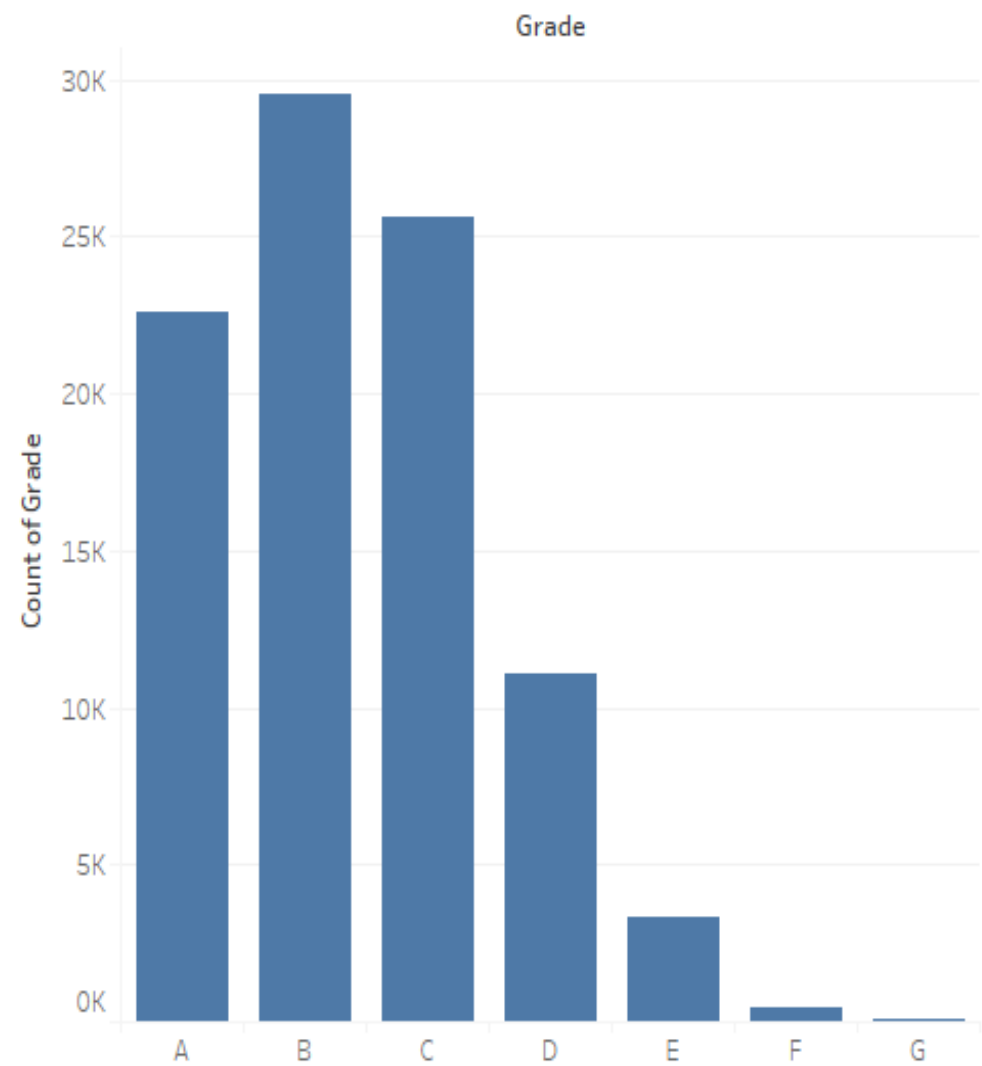
# About Lending Club



- LendingClub is the largest personal loans provider in the **U.S.** and a marketplace
- It is essentially a peer to peer lending platform.
- **Peer-to-peer** (P2P) lending, is the practice of lending money to individuals or businesses through online services that match lenders with borrowers.



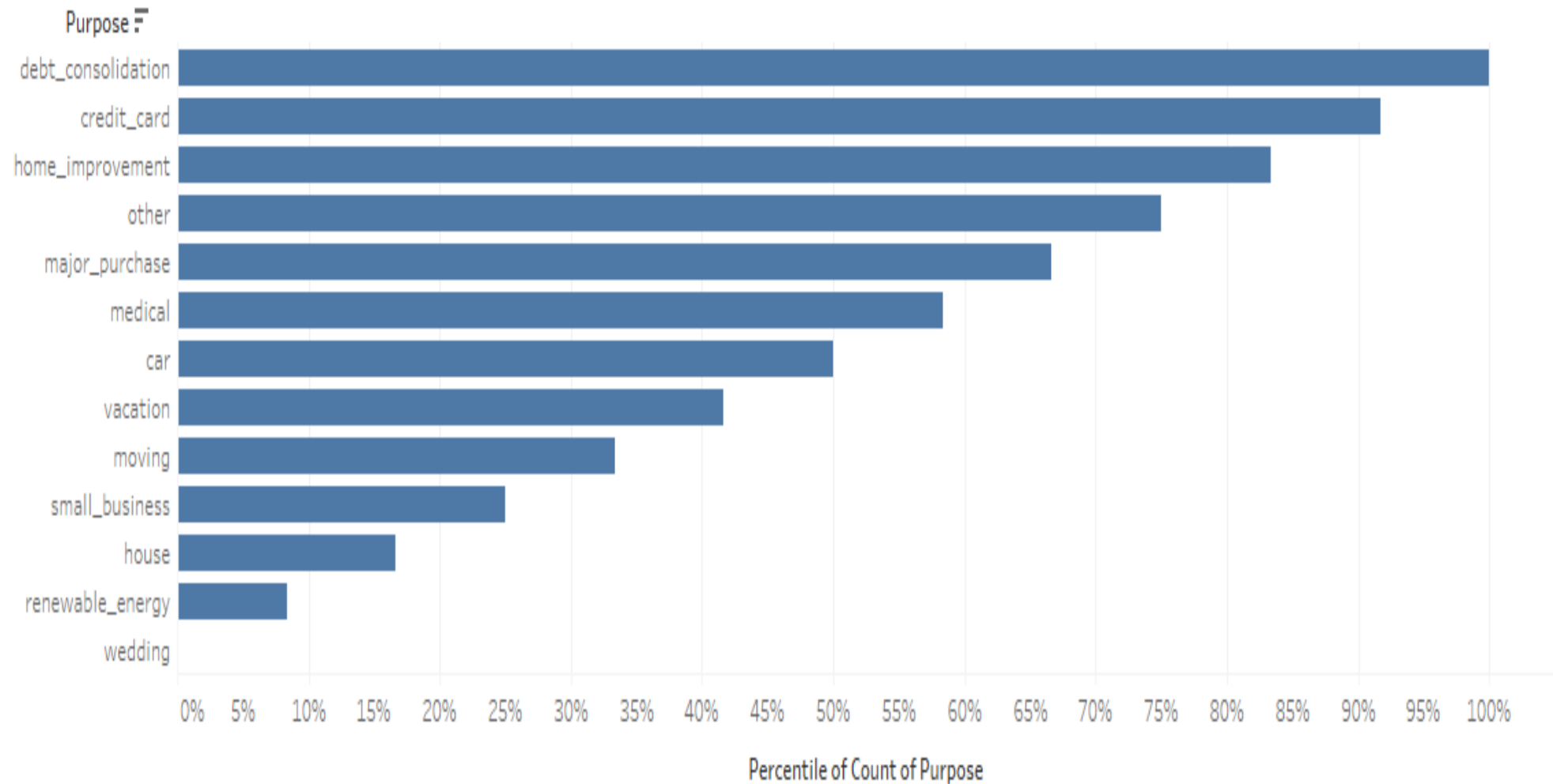
No.of Loan by Grade



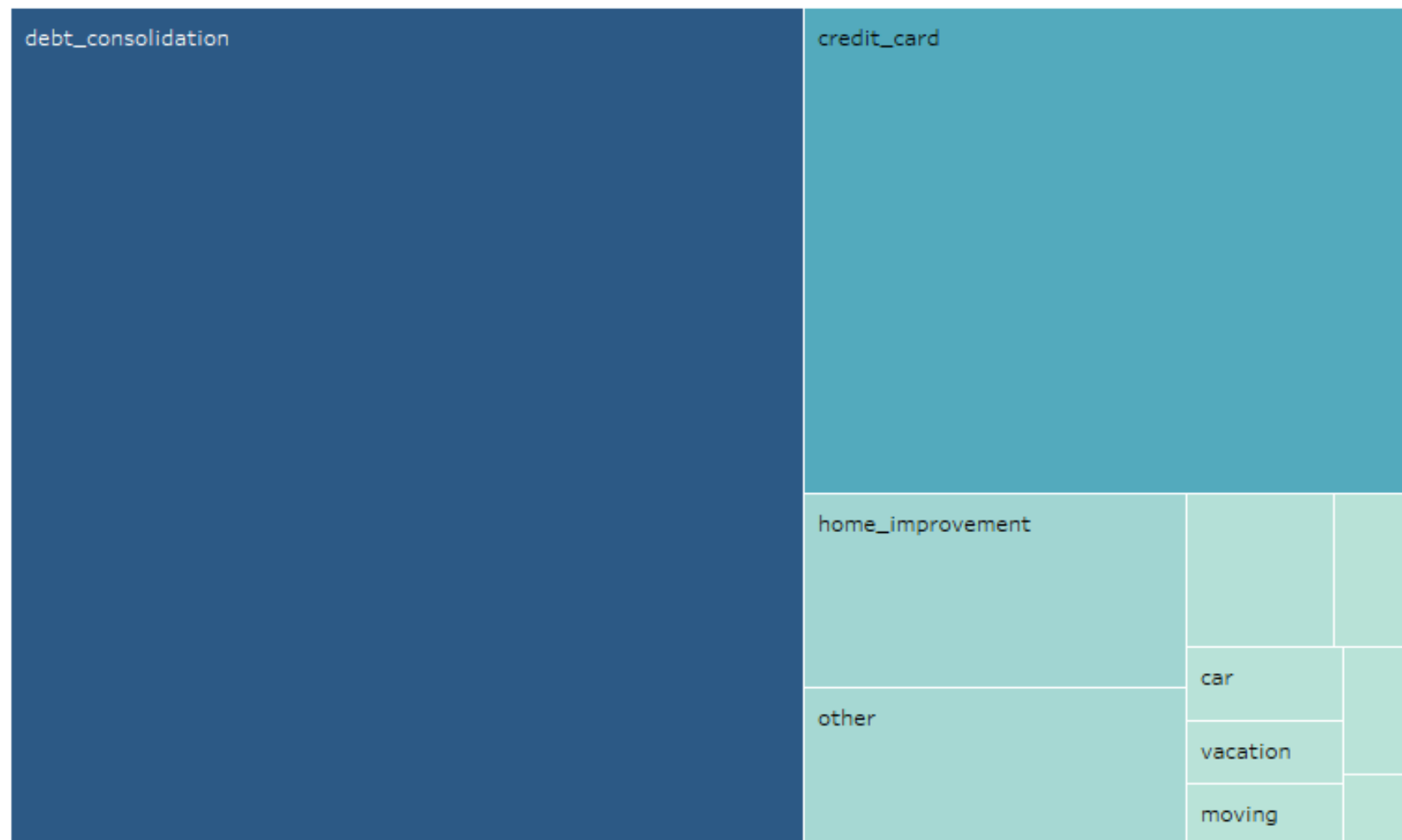
Maximum loans is observed to be falling under the category of B and C.

# Major categories of loan purpose

Loan purpose in percentile

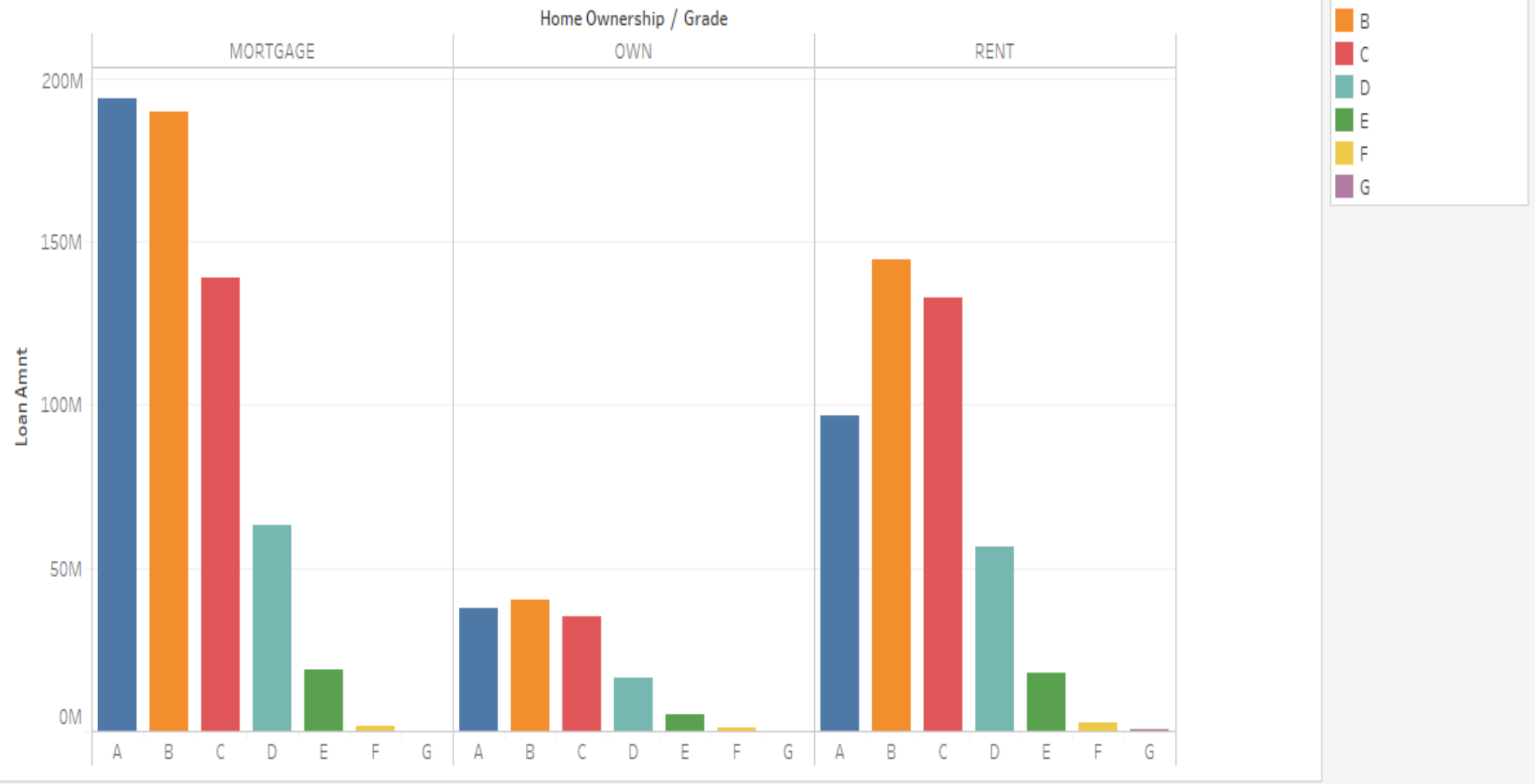


Tree Map showing Loan Purpose

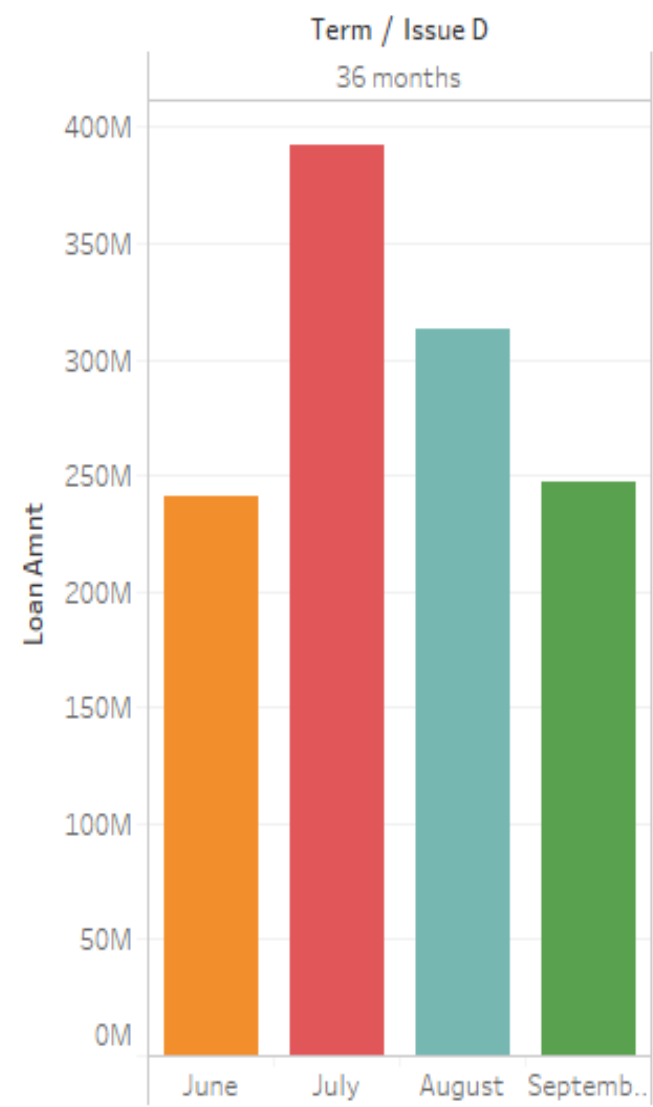


# Distribution of loans based on Homeownership

Sheet 7



Loan volume by Term

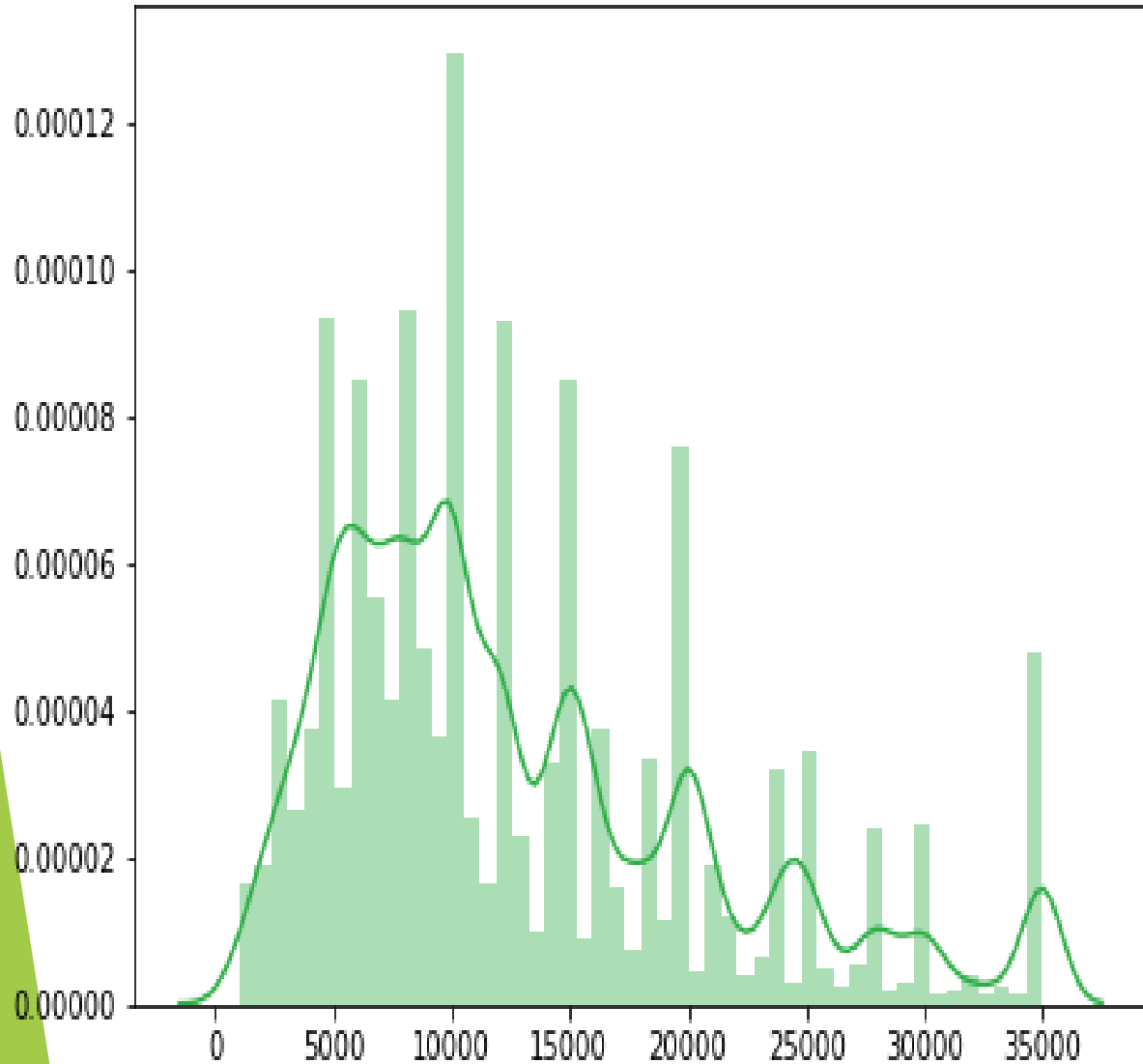


Overall loan monthly wise

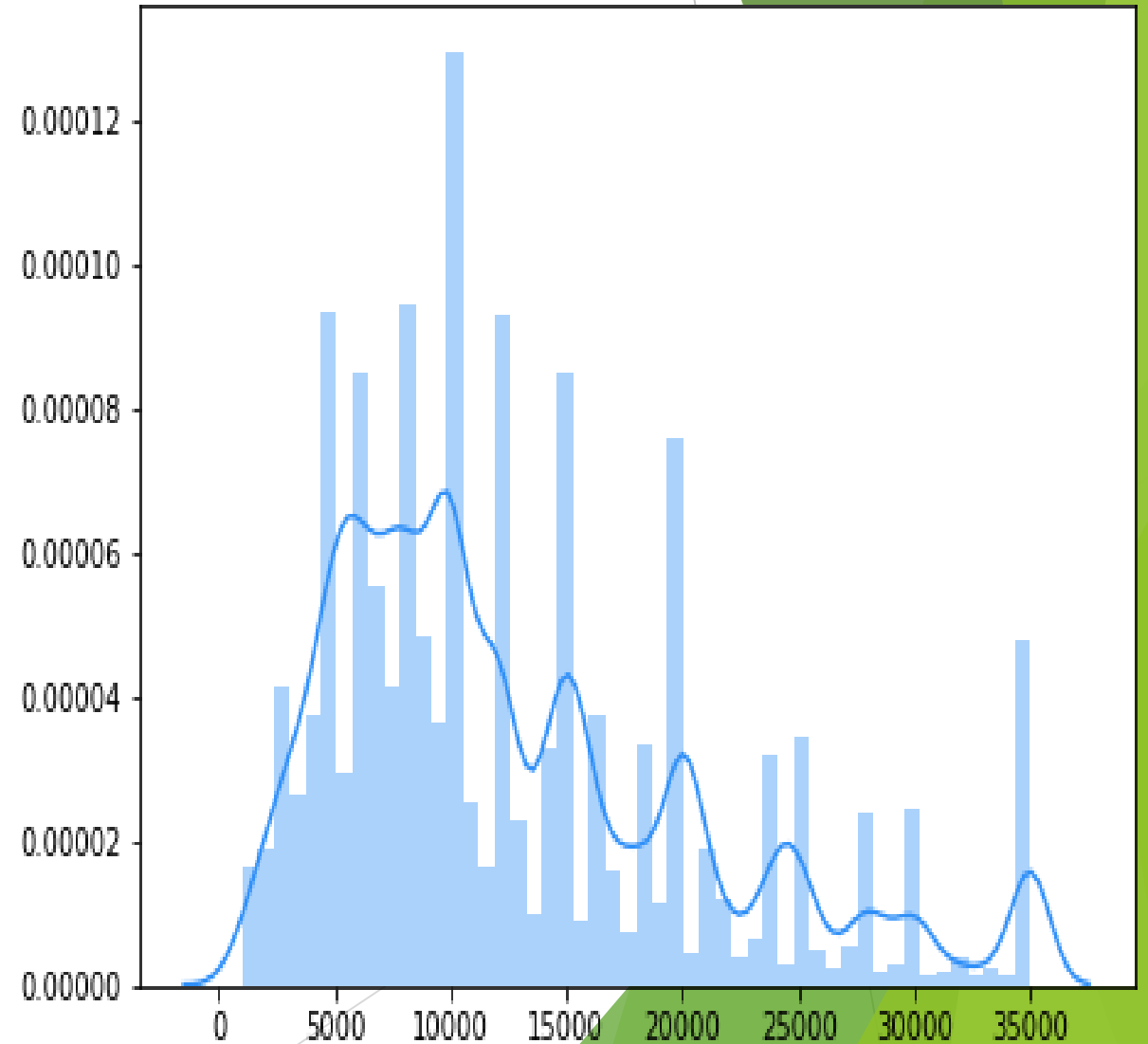


# Proportion of Supply and Demand

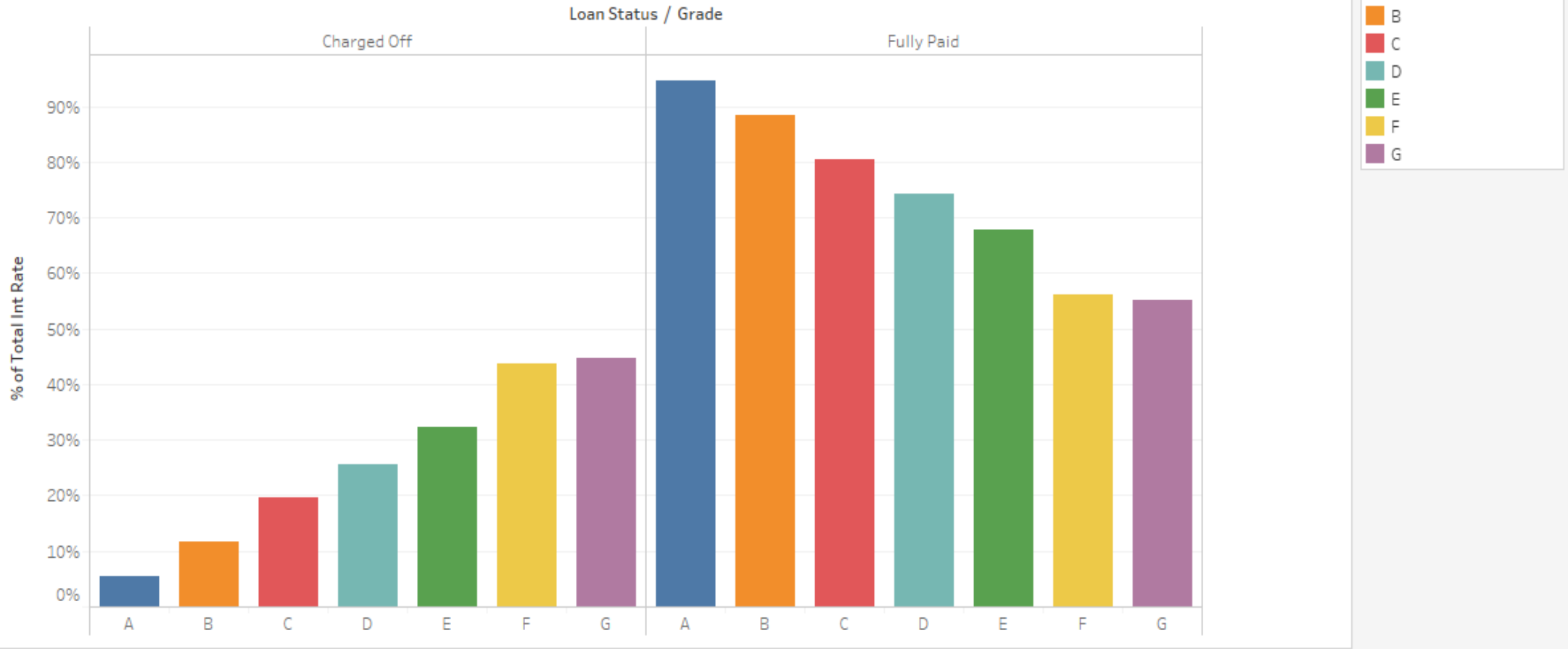
Loan Applied by the Borrower



Amount Funded by the Lender



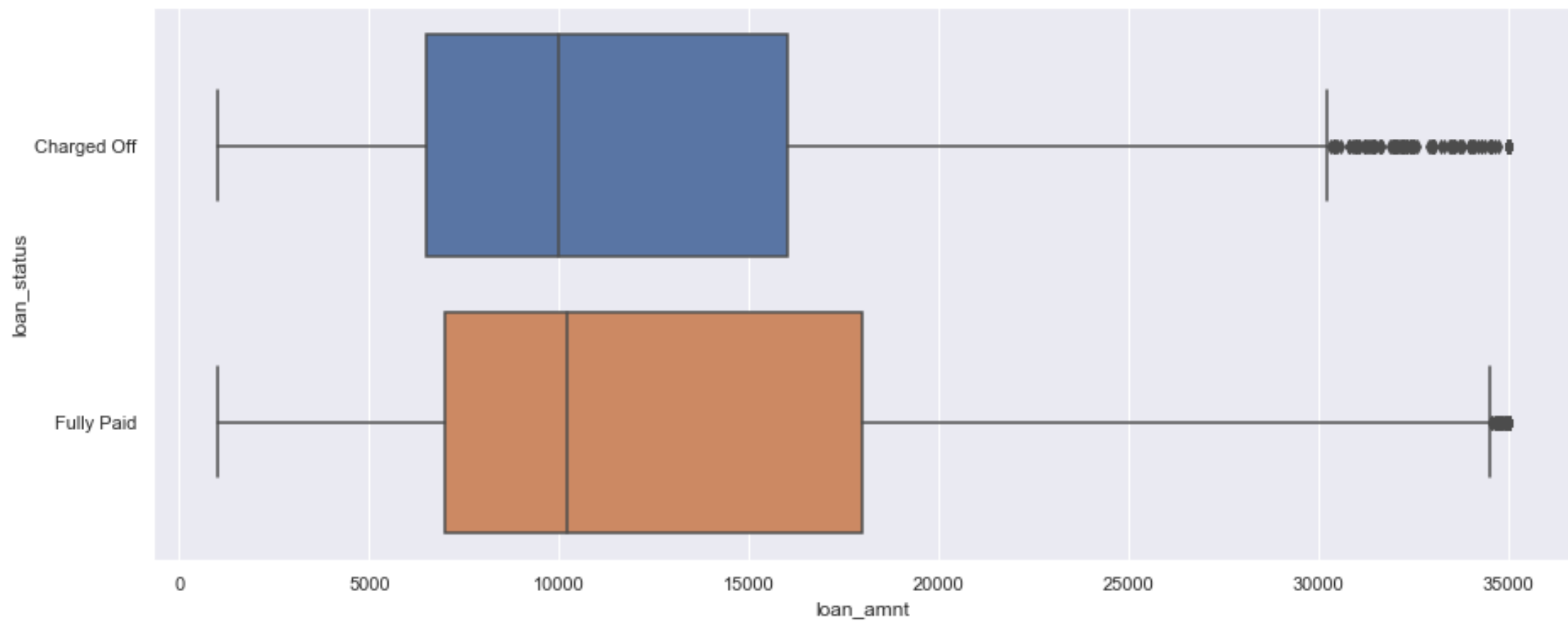
## Loan Status



Loan Status categorised as Fully Paid and Charged off loan

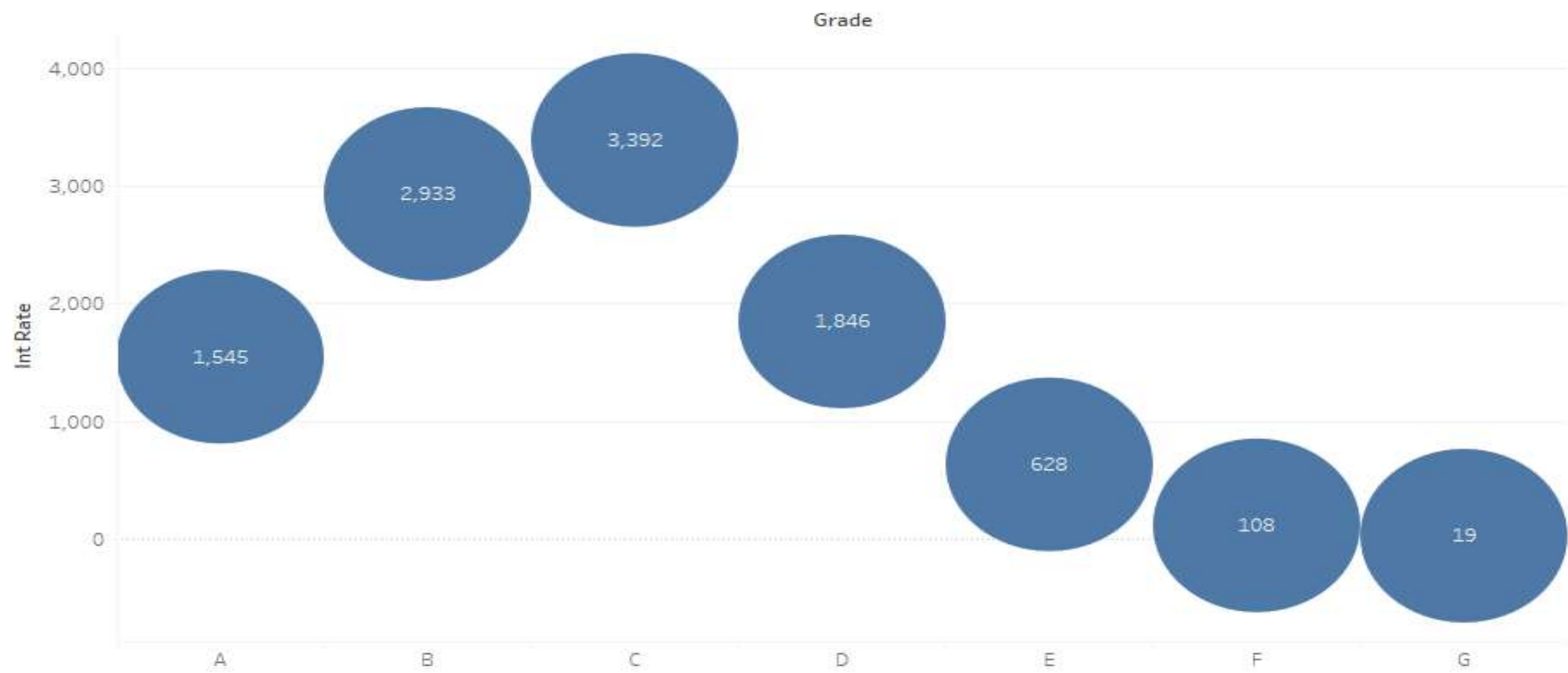
1. Based on Grade
2. Based on interest Rate





# Interest rate distribution over Grade

Interest Rate - Grade



# Problem Statement

The task is the development of the appropriate ML model which can be used to predict ‘**Fully Paid**’ vs “ **Charged Off**” loans in the Lending Club platform. In this task, you need to develop models to predict which loans are likely to be paid off and which will default. You also have to judge the interrelationship among the variables and determine an effective investment decisions based on your predictive models – we need examine this in details here. You need to focus on parameter tuning, and reliable performance estimates through resampling and cross-validation.

# Data Cleaning

- ❑ Deleting columns with no values
- ❑ Excluding Univariate columns
- ❑ Removal of attributes such as url ,zipcode etc based on description

Data consists of 92624 rows and 79 attributes.

Tools Used : Tableau

Programming language : Python

Models are based on Machine Learning Algorithms.



## Numerical variable Feature Selection

13 attributes were extracted after decision tree modelling

Model	Random Forest	Decision Tree
Accuracy Score	96.53	96.56

## Categorical variable

Pi-square test has been applied on all variables and accordingly , attributes of importance is sorted out

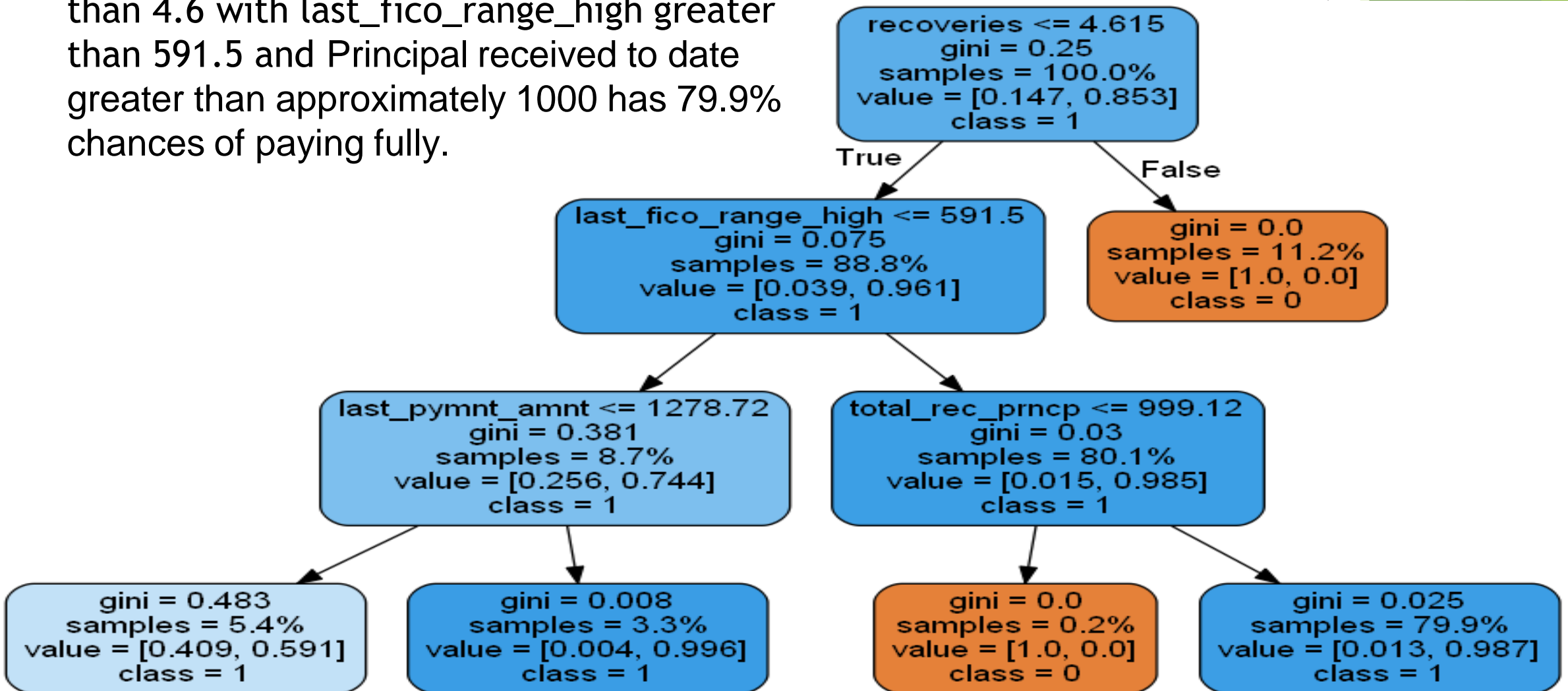
# Overall Performance of Models on Loan Status

Models were applied on data obtained after cleaning

MODEL	DECISION TREE	BAGGING	RANDOM FOREST	ADA BOOSTING	GRADIENT BOOSTING
ACCURACY SCORE	<b>95.99%</b>	<b>99.62%</b>	<b>99.48%</b>	<b>99.28%</b>	<b>99.26%</b>

# Loan Status interpretation

Given post charge off gross recovery less than 4.6 with last\_fico\_range\_high greater than 591.5 and Principal received to date greater than approximately 1000 has 79.9% chances of paying fully.





# Factors affecting loan\_status

- ❑ **Loan amount** : The listed amount of the loan applied for by the borrower. If at some point in time, the credit department reduces the loan amount, then it will be reflected in this value.
- ❑ **Funded\_amnt\_inv** : The total amount committed by investors for that loan at that point in time.
- ❑ **Purpose\_credit\_card** : A category provided by the borrower for the loan request.
- ❑ **Home\_ownership** : The home ownership status provided by the borrower during registration. Our values are: **RENT**, **OWN**, **MORTGAGE**, **OTHER**.
- ❑ **Total\_rec\_prncp** : Principal received to date
- ❑ **Int\_rate** : Interest Rate on the loan
- ❑ **Total\_pymnt** : Payments received to date for total amount funded
- ❑ **Title\_Debt consolidation** : The loan title provided by the borrower
- ❑ **Emp\_length\_10+ years** : Employment length in years.