

# Problem Statement



Consumer finance company wants to understand the **driving factors** (or **driver variables**) behind loan default



The company can utilize this knowledge for its portfolio and risk assessment



Analysis will help company to make a decision for loan approval based on the applicant's profile Analysis approach

Data understanding and sourcing

Check data quality issues and fix missing values

Perform univariate data analysis – categorical and numerical variables

Perform bivariate data analysis - categorical and numerical variables

Identify correlation between continues variables

# Data Understaning



- ■There are 111 columns having various data types like object, int, float and 305711 rows.
- ■There are many columns with 0 values or NULL values.
- ■There are columns with special characters like interest rate, employment length, standardizing is required

# Data quality check

There are 58 columns having missing value more than 30%. Hence we exclude these columns from analysis

```
## list of columns where missing values are above 30%
nullcol_gt30 = (round((data.isnull().sum()*100/len(data)).sort_values(ascending = False),2))[round((data.isnull().sum()*100/len(data)).sort_values(ascending = False),2)>30]

Python

print("Num of columns having missing values more than 30% :",len(nullcol_gt30))

v 0.3s

Num of columns having missing values more than 30% : 58
```

# Data standardizing

 Convert emp\_length column to int by removing special characters and alphabets

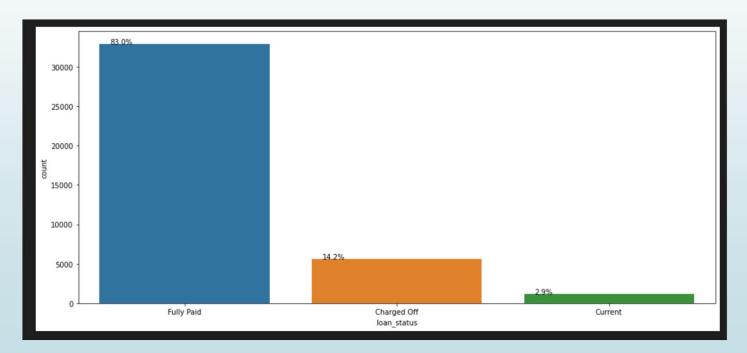
# Data standardizing

 Clean up int\_rate column to int by removing '%' character and convert to numeric



### Categorical column 'loan\_status':

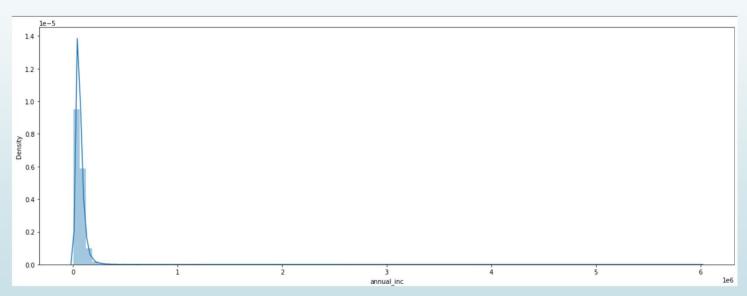
Out of 39717 applicants, 14.2% that is 5609 applicants defaulted.

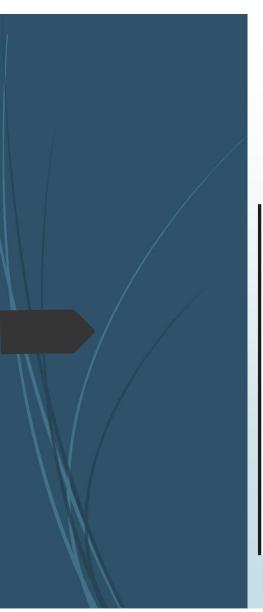




### **Annual Income:**

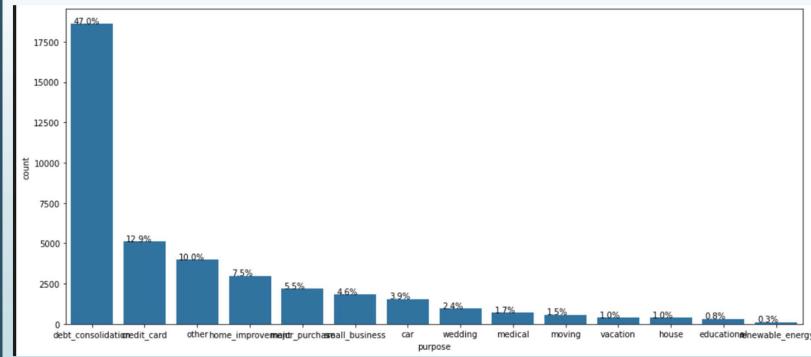
Most of applicants have income less than 50000





### Loan Purpose:

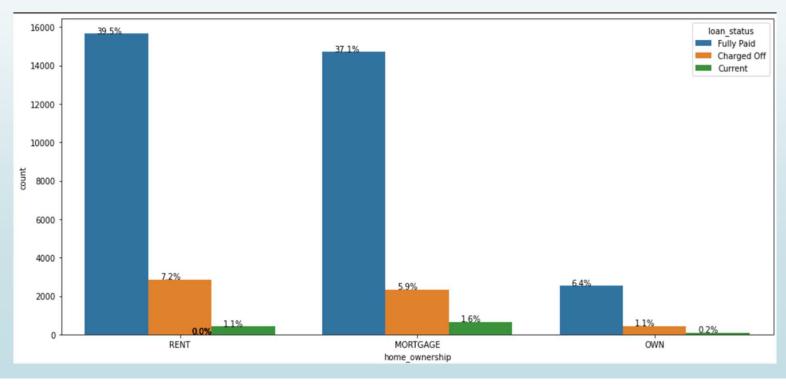
Majority of loan - 47% falls under debt consolidation category





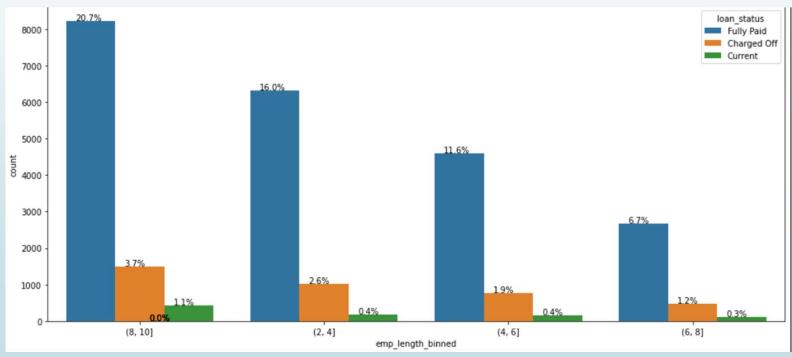
### Home Ownership vs loan status of applicants:

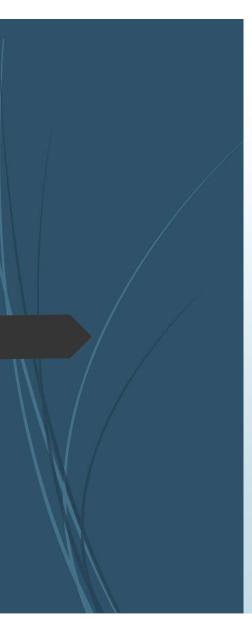
Approx 48% of applicants live in rented home and 45% of applicants have mortgage on their house. There are very less loan applications from house owners.





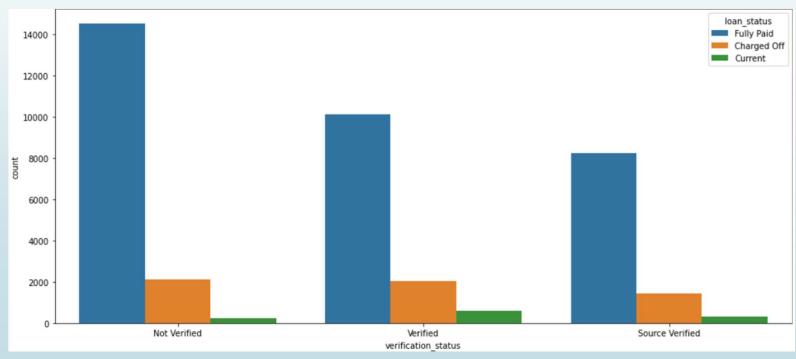
Employment length vs Loan status of applicants:
There is not much difference in defaulters with respect employment length





**Application verification vs loan status:** From not verified, around 86% applicants fully repaid loan, apund 12.64% charged off

Charged off percentage is higher 16% in Source verified category, compared to other two.

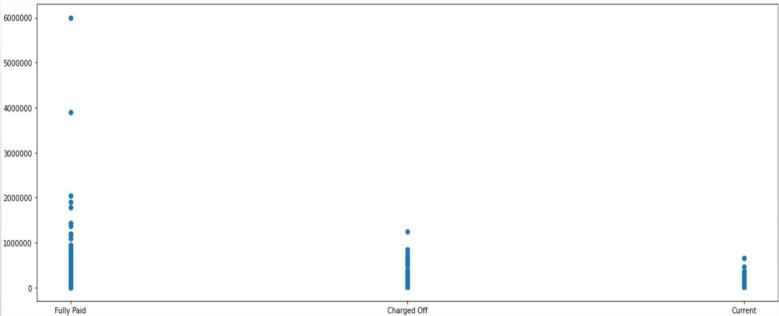




# **Bivariate Analysis**

### **Annual Income:**

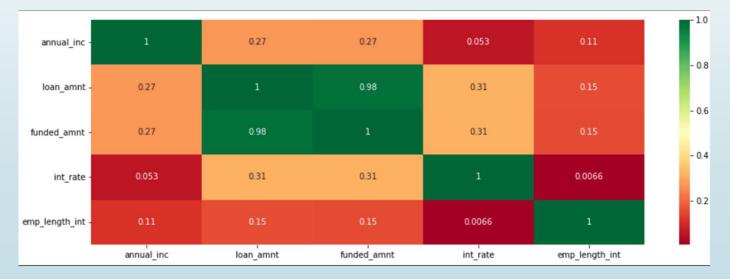
All the defaulters have annual income below 150000





# Correlation

- There is strong correlation between funded amount and loan amout
- Loan amount and funded amout shows positive correlation with Annual income
- Loan amount and funded amout shows positive correlation with interest rate
- Correlation is less between employment length and annual income, loan amount and funded amount.



# Conclusion

### Decisive Factor whether an applicant will be Repayer:

- 1. Annual\_income: Annual income below 150000 have more defaults.
- 2. Verification status: Source verification has less defaults than not verified and verified.
- 3. Housing status with Rent is having more defaults.

# Thank you