# Lending Club Case Study

**Exploratory Data Analysis** 

### **Business Overview**

- The company specializes in lending loans to urban customers.
- The company has to make a decision for loan approval based on the applicant's profile.
- Two types of risks are associated with the bank's decision:
  - If the applicant is likely to repay the loan, then not approving the loan results in a loss of business to the company.
  - If the applicant is not likely to repay the loan, then approving the loan may lead to a financial loss for the company.
- The data provided contains information about past loan applicants and whether they defaulted or not.
- There are a total of 111 columns and 39717 records in the loan.csv file.

The aim is to identify patterns that indicate if a person is likely to default, which may be used for taking actions such as denying the loan, reducing the amount of loan, lending (to risky applicants) at a higher interest rate, etc.

# **Data Cleaning**

- There are a total of 111 columns and 39717 rows.
- Majority of the columns has only Null values -
  - Hence dropping all the columns whose null values are more than 20000
- We are left with 16 columns and 39717 rows.
- There are no rows in the data with all null records.
- Dropping all the rows whose loan\_status = Current as they have no effect on our analysis

Output -The final cleaned dataset has a total of 16 columns and 36800 rows.

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 39717 entries, 0 to 39716
Data columns (total 16 columns):
    Column
                          Non-Null Count Dtype
     loan_amnt
                          39717 non-null int64
     term
                          39717 non-null object
                          39717 non-null object
     int rate
     installment
                           39717 non-null float64
     grade
                          39717 non-null object
     emp length
                          38642 non-null object
    home ownership
                          39717 non-null object
     annual inc
                          39717 non-null float64
    verification status
                          39717 non-null object
     issue d
                          39717 non-null object
    loan status
                          39717 non-null object
    purpose
                          39717 non-null object
 12 dti
                          39717 non-null float64
    pub rec
                          39717 non-null int64
14 revol_util
                          39667 non-null object
   pub rec bankruptcies 39020 non-null float64
dtypes: float64(4), int64(2), object(10)
memory usage: 4.8+ MB
```

# **Data Formatting**

- The following columns with inconsistent data formats are corrected for appropriate analysis
  - Employment length
    - Special chars & Strings are removed
    - Converted into float
  - I oan issue date
    - Object to date
  - Term of the loan
    - String object to years
  - Interest rate & Revolving credit utilization rate
    - Removed % and converted to float

Output - Cleaned data without data inconsistencies and ready for further EDA steps

```
final loan data.info()
<class 'pandas.core.frame.DataFrame'>
Index: 36800 entries, 0 to 39680
Data columns (total 16 columns):
    Column
                          Non-Null Count Dtvpe
    loan amnt
                          36800 non-null int64
    term
                          36800 non-null float64
    int rate
                          36800 non-null float64
    installment
                          36800 non-null float64
    grade
                          36800 non-null object
    emp_length
                          36800 non-null object
    home_ownership
                          36800 non-null object
    annual inc
                          36800 non-null float64
    verification status
                          36800 non-null object
    issue d
                          36800 non-null datetime64[ns]
    loan_status
                          36800 non-null object
    purpose
                          36800 non-null object
   dti
                          36800 non-null float64
    pub rec
                          36800 non-null int64
 14 revol util
                          36800 non-null float64
   pub_rec_bankruptcies 36800 non-null float64
dtypes: datetime64[ns](1), float64(7), int64(2), object(6)
memory usage: 4.8+ MB
```

# **Univariate Analysis**

### 1. Purpose of the loan

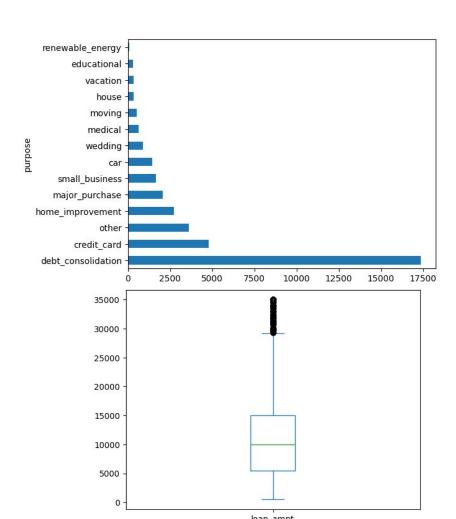
Debt Consolidation, Credit card and Home improvement are some of the popular reasons for taking loan

#### 2. Loan Amount

- Median loan amount: 10000.0

- Largest loan amount: 35000

- Smallest loan amount: 500



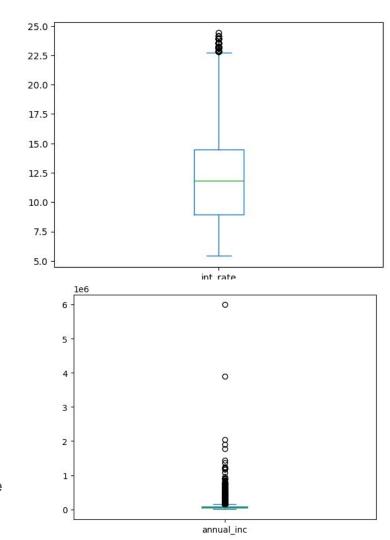
### 3. Interest Rate

- Median interest rate: 11.83
- Max interest rate: 24.4
- Lowest interest rate: 5.42

### 4. Annual Income

- Median Annual Income: 10000.0
- Largest Annual Income: 6000000.0\*
- Lowest Annual Income: 500

Post mean and median comparison the outliers in the columns are dropped to ensure unbiased analysis.



# Segmented Univariate Analysis

Create bins for quantitative columns like :1.Debt to income ratio (DTI) | 2.Interest rate | 3.Revolving line utilization rate.

#### 1. Debt to income ratio:

- DTI <=8 : Very Low

- 8 - 12 : Low

- 12 - 14 : Moderate

- 16 - 20 : High

DTI > 20 : Very High

#### 2. Interest Rate:

- Interest Rate < = 9 : Very Low

- 9 - 11 : Low

- 11 - 13 : Moderate

- 13 - 15 : High

Interest Rate > 15 : Very High

# 3. Revolving line utilization rate

- 0 - 20 : Very Low

- 20 - 40 : Low

- 40 - 50 : Moderate

- 50 - 60 : High

- 60 - 100 : Very High

# Segmented Univariate Analysis

Name: count, dtype: int64

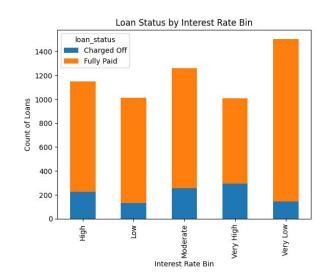
loan_status	dti_b		loan_status	int_rate_b		loan_status	revol_util_b	
Charged Off	Very Low	246	Charged Off	Very High	293	Charged Off	Very High	897
Fully Paid	Very High	223	Fully Paid	Moderate	253		Very Low	116
	Moderate	205		High	226		Moderate	14
	High	204		Very Low	144		High	12
	Low	171		Low	133		Low	10
	Very Low	1245		Very Low	1362	Fully Paid	Very High	3971
	Very High	1071		Moderate	1009		Very Low	599
	Moderate	919		High	922		High	112
	High	860		Low	879		Low	104
	Low	793		Very High	716		Moderate	102

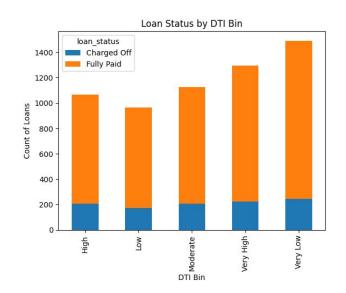
Name: count, dtype: int64
Name: count, dtype: int64

### Bi Variate Analysis -

### <u>Segmented DTI ratios Vs Loan</u> <u>Status</u>

Conclusion: Loans given to higher DTI
 ratio > 16 for applicants are more likely to
 be defaulted



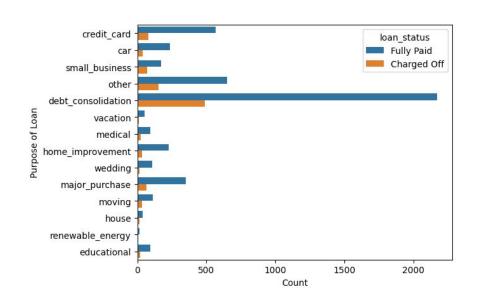


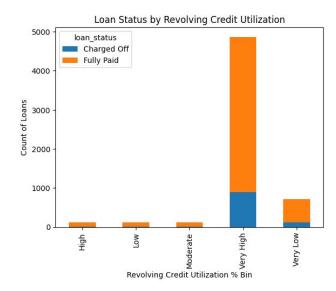
### <u>Segmented interest rates Vs Loan</u> <u>Status</u>

- Conclusion : Higher interest rates loans >
13% are more likely to get defaulted.

### <u>Segmented Credit line utilization Rate Vs Loan</u> <u>Status</u>

- Conclusion : Customers who have maxed out their credit utilization (high %) are more likely to default



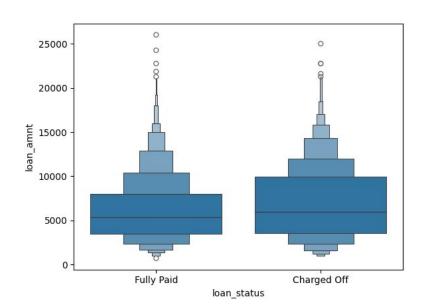


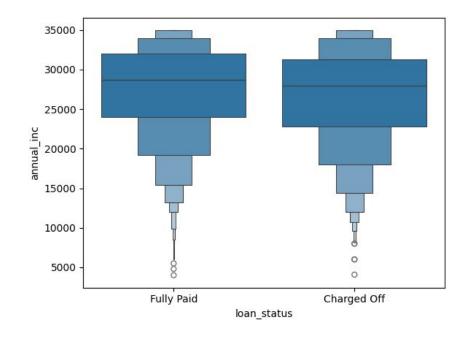
### Purpose of the Loan Vs Loan Status

- Conclusion: Most of the loan defaults are occurring under debt consolidation category.

### <u>Annual income Vs Loan Status</u>

- **Conclusion**: Majority of loan defaults are happening in the case of applicants whose annual income < 20000



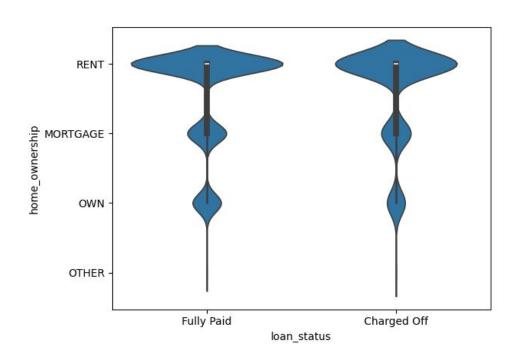


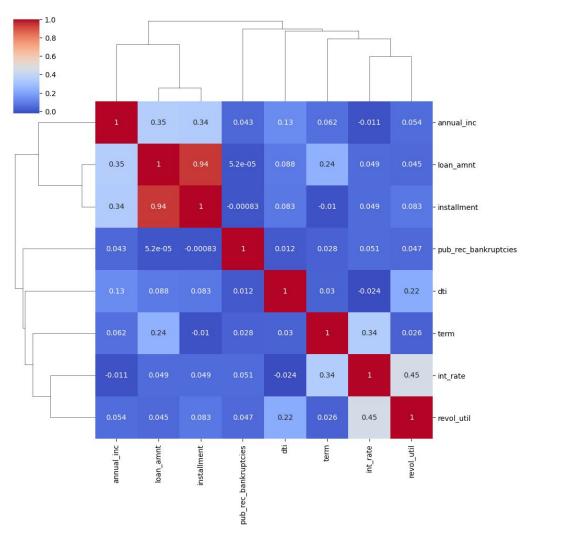
### Loan amount taken Vs Loan Status

- **Conclusion**: We can infer that lower value loans has a higher probability of being charged off.
Especially true for loan amounts between 2500 to 7500 with the average loan amount being 5000

### Home Ownership Vs Loan Status

- **Conclusion**: Most of the loan defaults are by applicant who are living in rented homes





#### **Correlation Analysis**

- Find correlation between the most impacted factors from the lessons drawn from previous analysis

#### All the factors given below has strong correlation with Loan status

- Loan Amount
- Term of the loan
- Interest Rate
- Installment amount
- DTI
- Annual Income

#### - Conclusions

- 1. Annual Income, Loan Amount and installment shows a high degree of correlation that confirms our earlier analysis observations.
- 2. Interest rate granted to the loan is highly r=dependant on the credit utilization percentage of the customer.
- 3. The loan term/duration is highly influenced by the interest rate.
- 4. Interest rates are negatively correlated chances of default DTI and Annual Income of the applicant which implies that borrowers with strong indicators of financial health get low interest rates hence less chances of default.

# Ideal Customer persona

- Customer who are working in a firm for 10+ years
- Whose DTI ratio is low
- Who live in their own home without any mortgage on it
- Loan amount is between 2500 to 7500
- Annual income is between 20000 to 32000
- Interest on the loan < 13%