



# **Lending Club Case Study**

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Prepared by:  
Ejaz Sayyed  
Manikandan

# Agenda

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- Problem Statement
- Technologies Used
- Data Understanding
- Analysis
- Conclusion(s)



# Problem Statement

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- Understand the **driving factors (or driver variables)** behind loan default, i.e., the variables which are strong indicators of default

# Technologies Used

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We have used below technologies for analysis

- **Python** – Open-source language best suited for handling data. Python provides rich libraries like matplotlib & seaborn for visualization and many other libraries for statistical analysis and data manipulation
- **Jupyter Notebooks** – An open-source web application to create and share documents that contain live code (e.g. Python), equations, visualizations, and narrative text.

# Data Understanding – As-IS

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After analyzing the raw data, below are the initial observations:

- 39717 records with 111 columns
- Enough columns to derive the useful metrics and perform loan default analysis
- But also, more than 50 columns which had no data
- Many attributes which are not required for this analysis
- Some missing values in the data

# Data Cleansing Activities



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We used below techniques to cleanse the data and make it suitable for analysis -

- Fill in some columns with missing values i.e. imputation (using mean, median, mode techniques)
- Converted columns from String to Day/Month/Year
- Removing columns where 'all' values are null
- Dropping columns which are not relevant for this analysis
- Converting few column values from String to Integer/Float for analysis



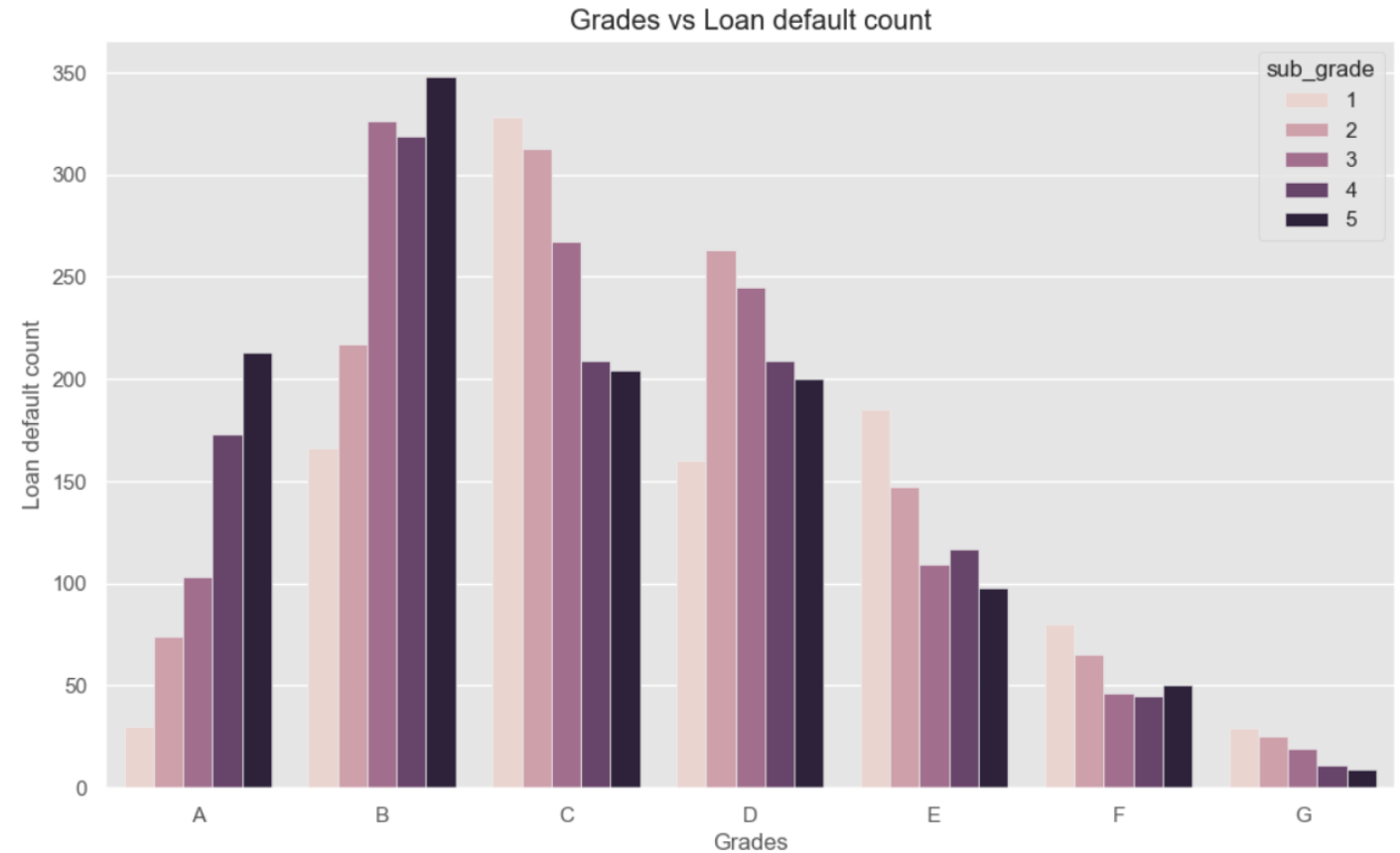
# Analysis on Data - I

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- Univariate Analysis (i.e. using one variable/field/column/attribute at a time)

# Observation 1

Many defaulters are from B and C grades and specifically, borrowers from 5th subgrade of B are higher in %.

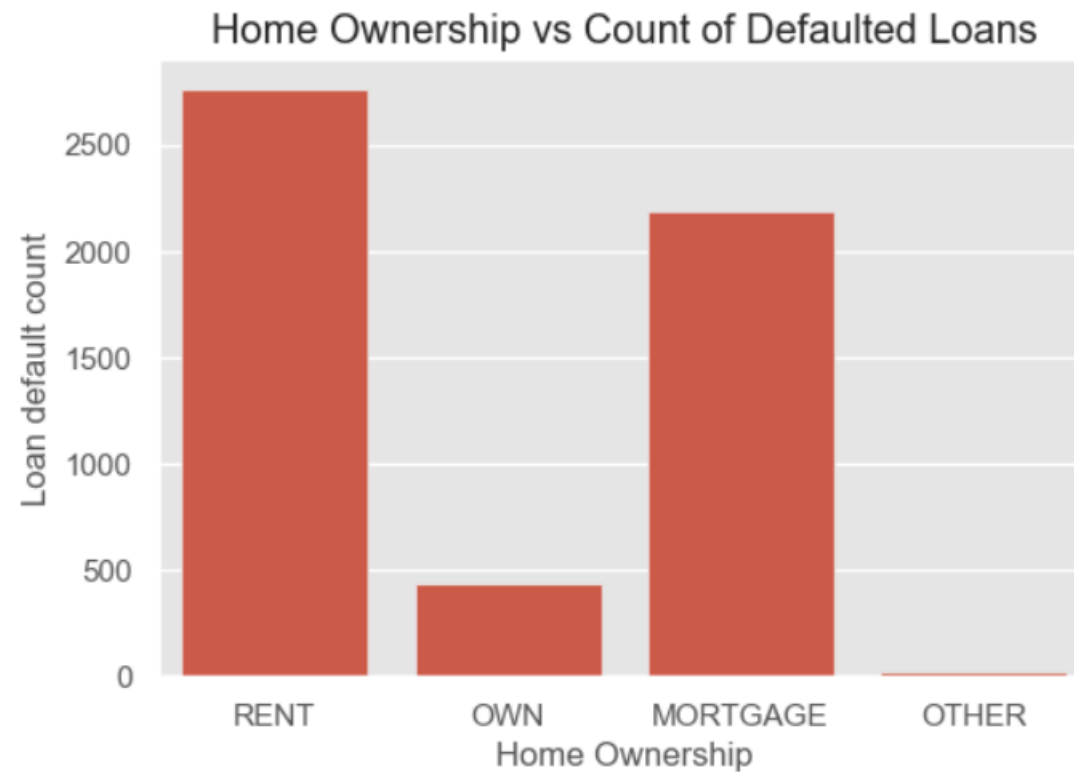




# Observation 2

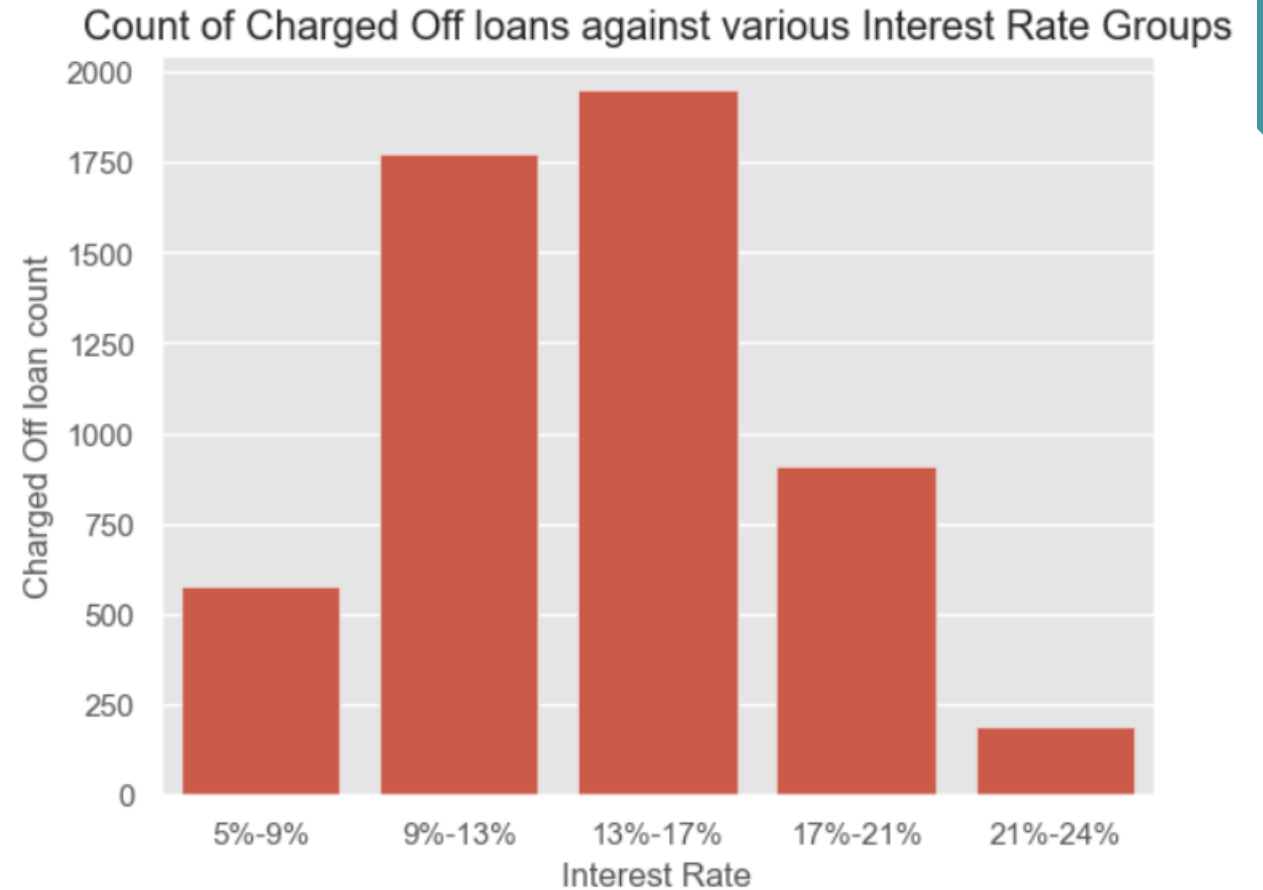
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Loan Defaulter have 'Home Ownership' mostly as 'Rent' or 'Mortgage'



# Observation 3

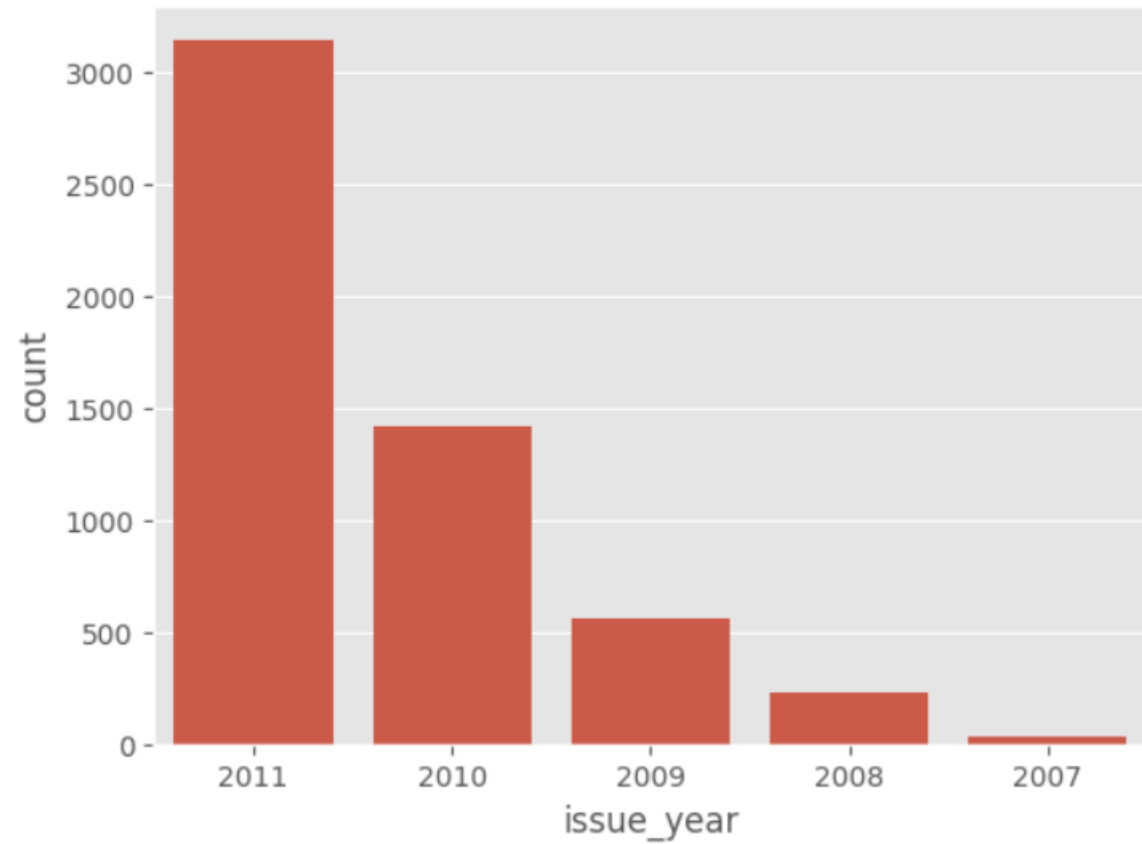
Loans are 'Charged Off' or 'Defaulted' when 'Interest Rate' is high.



# Observation 4

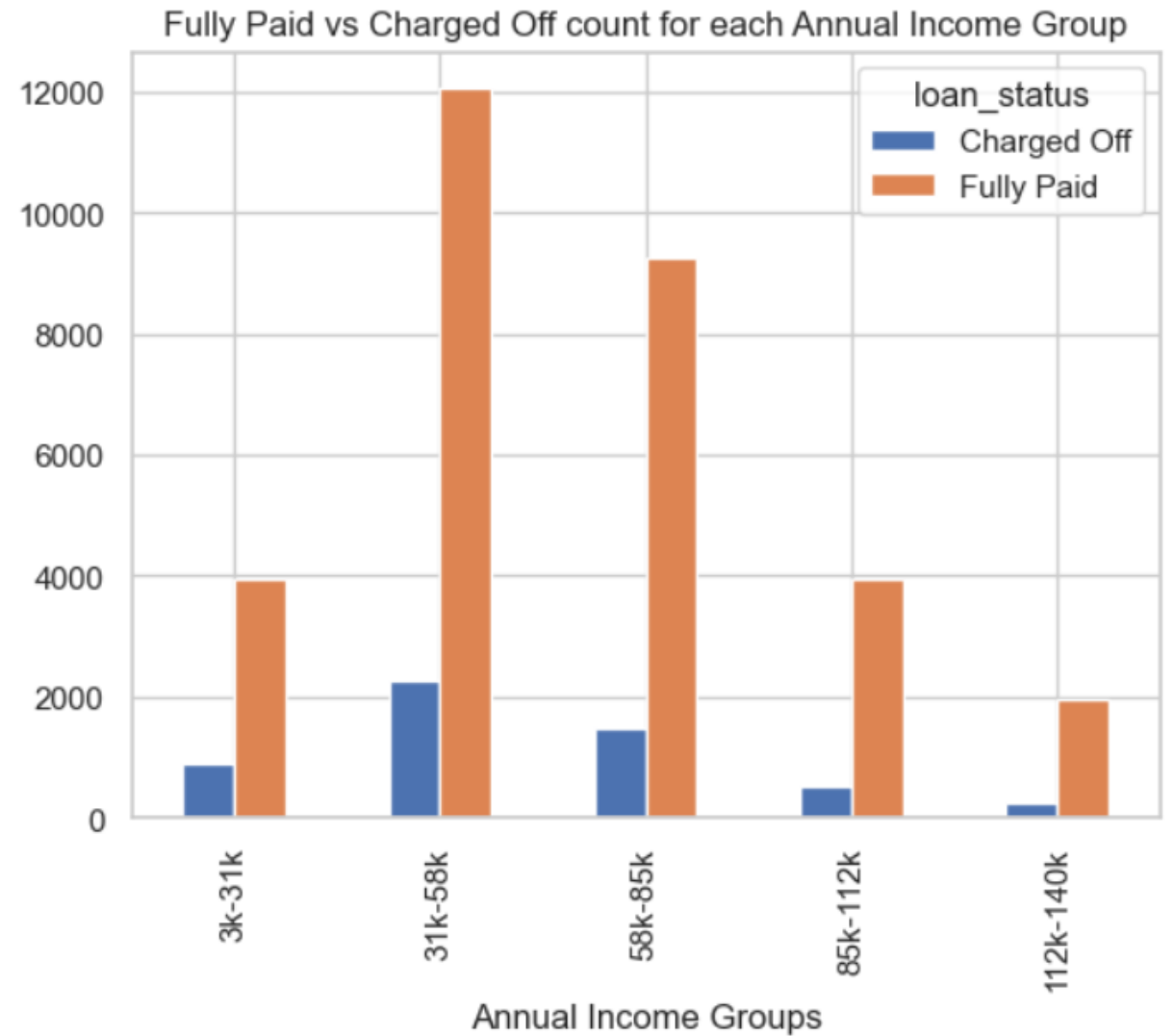
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Most Loan defaults have occurred in year 2011.



# Observation 5

Charged Off (i.e. Defaulters) count is high in the annual income group of 31k-58k.





# Analysis on Data - II

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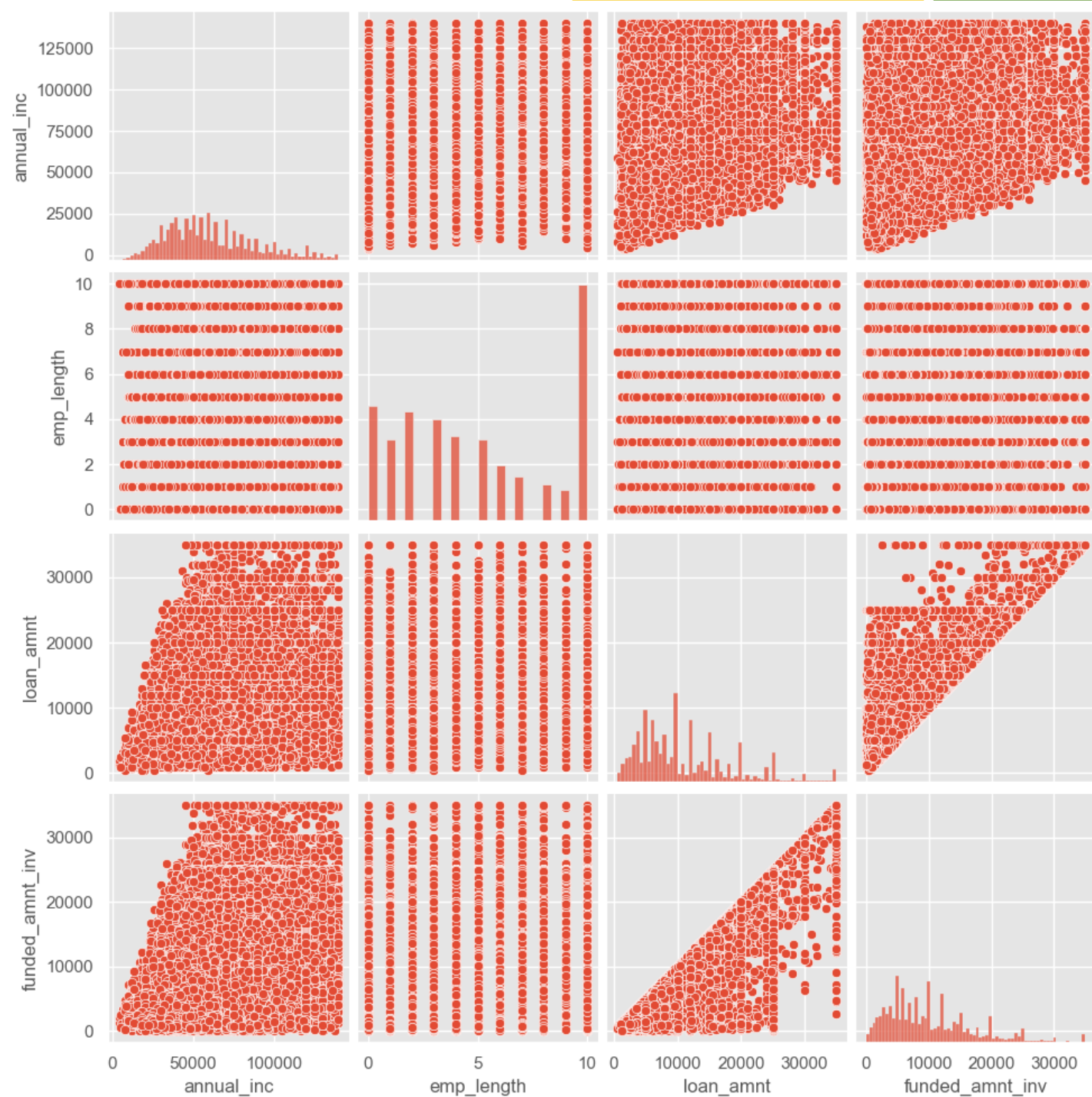
- Bivariate or Multivariate Analysis (i.e. using **two or more** variables/fields/columns/attributes at a time)

# Observation 6

This is called as Pairplot and is used to show relationship between multiple variables/attributes in the dataset.

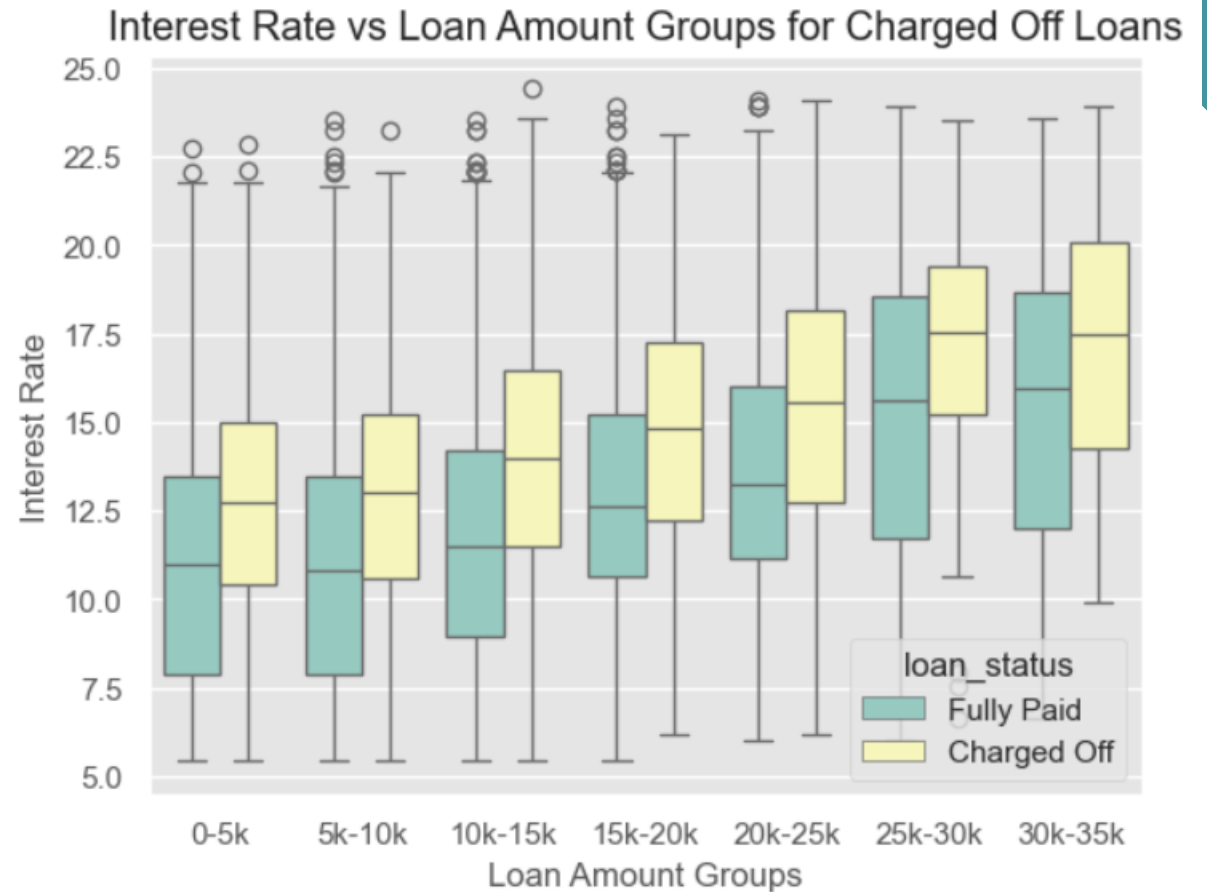
From the graph, it can be observed that –

- Loan Amount and Funded Amount are related
- No relationship between Employment Length/Duration and Loan Amount!



# Observation 7

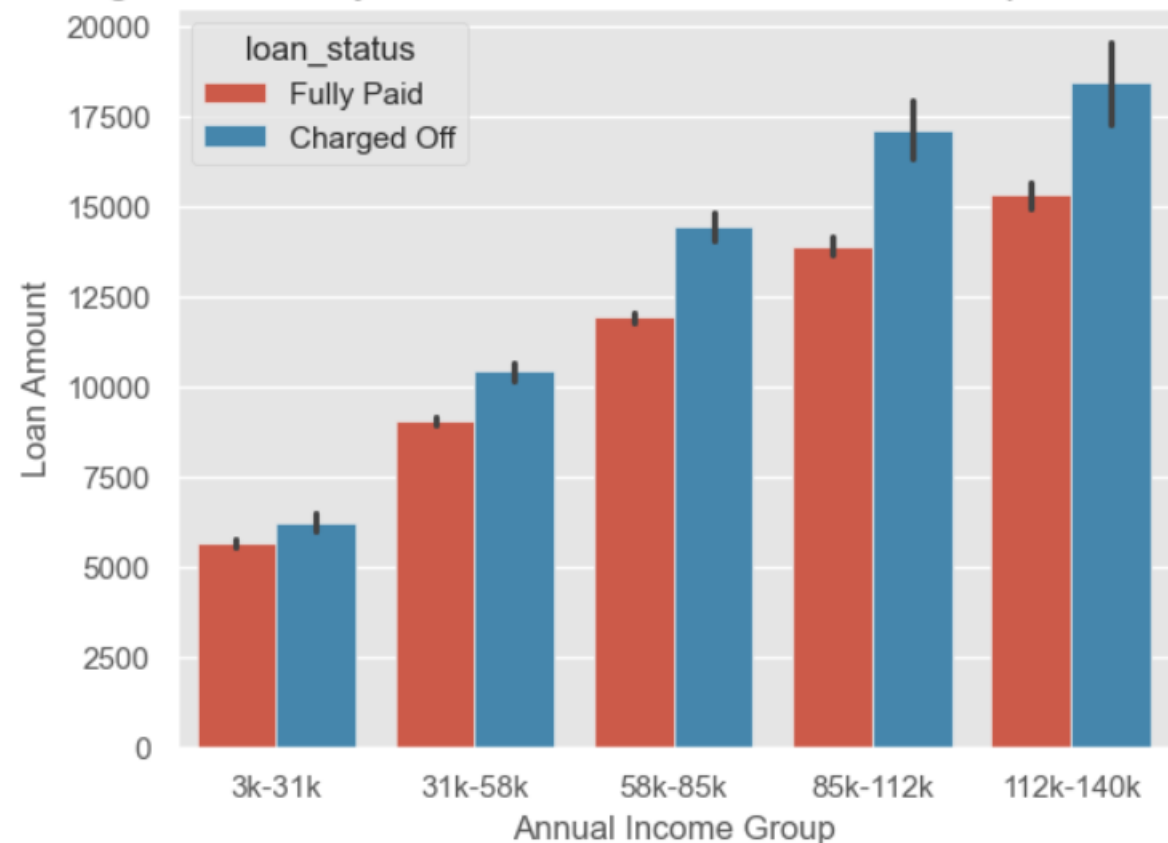
For high loan amounts, more interest is being charged for the loans which are defaulted.



# Observation 8

As the Loan Amount gets higher, % of loan defaults is increased. These high amount loans are taken by people with high annual income.

Charged Off & Fully Paid loans for Annual Income Group vs Loan Amount

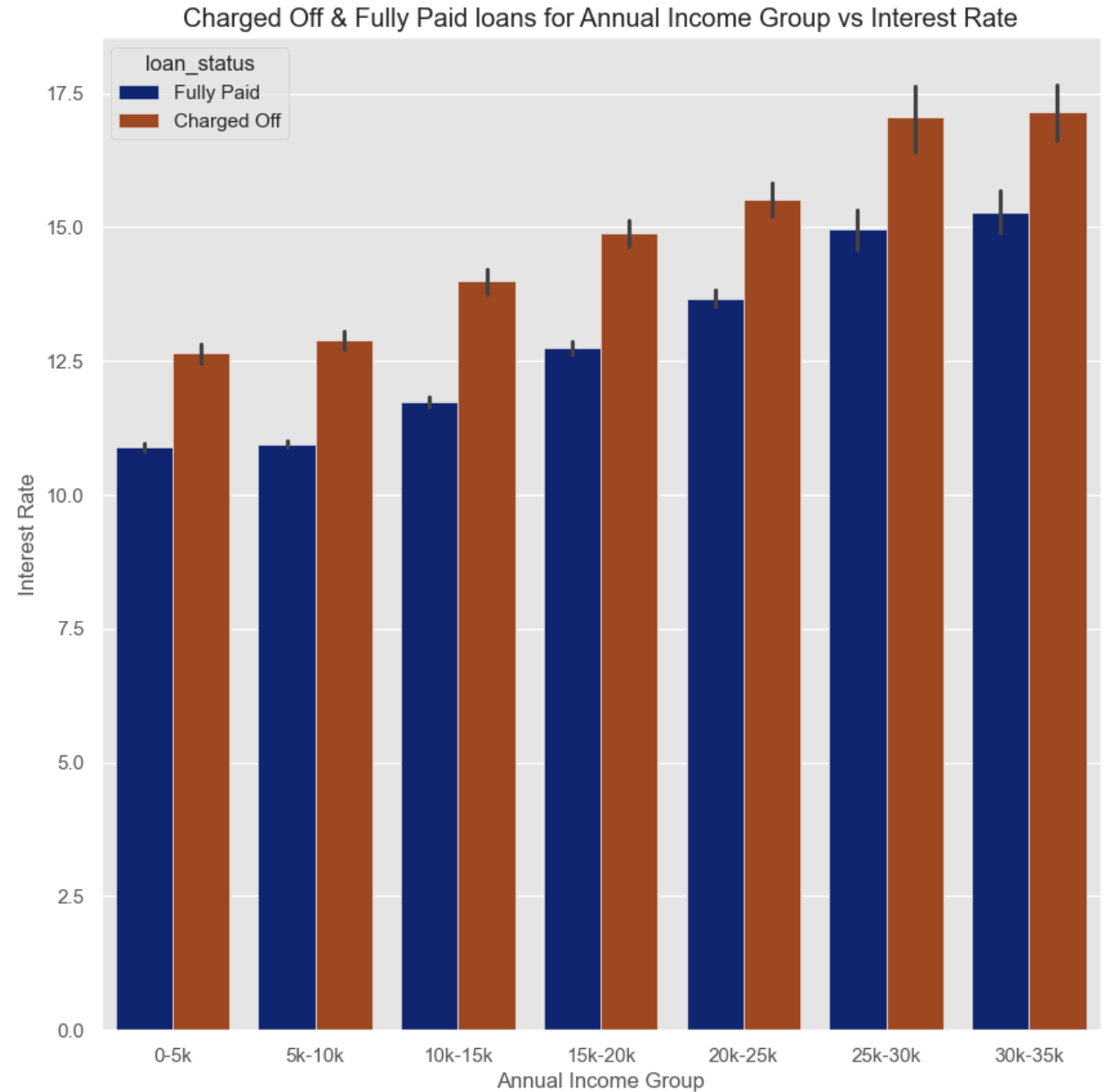




# Observation 9

High Interest Rate Loans are taken by people with high Annual Income.

‘Fully Paid’ Loans are always taken at Low Interest Rates than ‘Charged Off’ loans.



# Observation 10

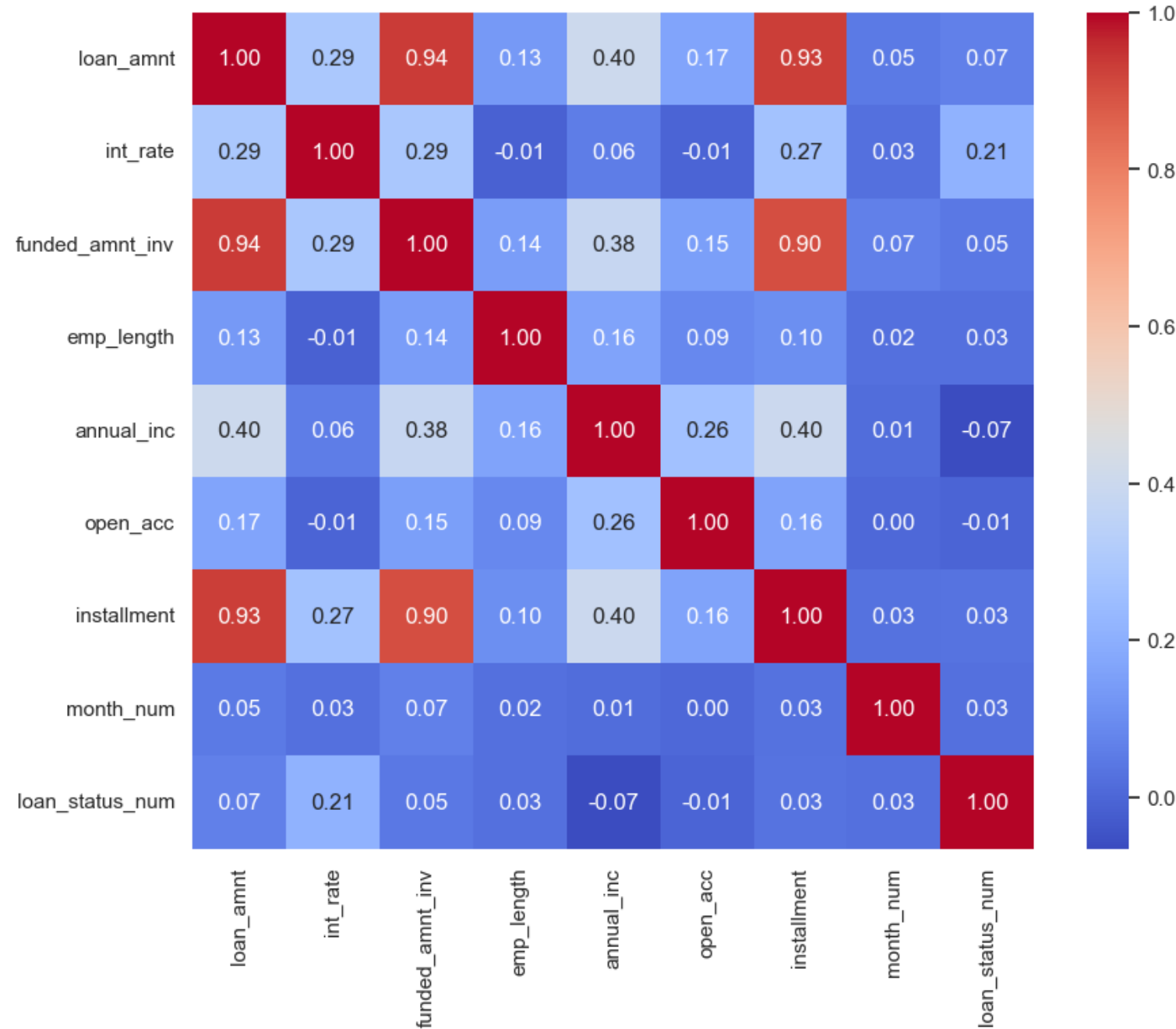
Heatmap is shown. It shows relationship of various attributes with every other attribute.

There is high correlation between –

- Funded Amount Investment & Installment
- Loan Amount & Installment
- Loan Status & Interest Rate

There is low correlation between –

- Loan Status & Annual Income of a person



# Conclusions

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Below are the main factors having influence on loan defaults –

- Interest Rate (and thus Installment)
- Grade
- Year (2011) – Can be due to market conditions in the year!
- Home Ownership

**Thank you**

