

Predicting Loan Defaults

By Stephanie Duncan 11th March 2021





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Introduction to Data

- Provided by an American online loan provider
- Over 5 year period: 2007 2011
- Client is concerned about default rate on their loans
 - Trend by Employment Length
 - Home Ownership
 - Fico Score
- Who should be lended to in the future?

Data Cleaning

- Project was carried out in Python
- Bias zip code was removed to avoid bias by location
- Missing Values
 - Many rows/columns fully missing removed
 - Employment length
- Over 100 variables removed those with single values

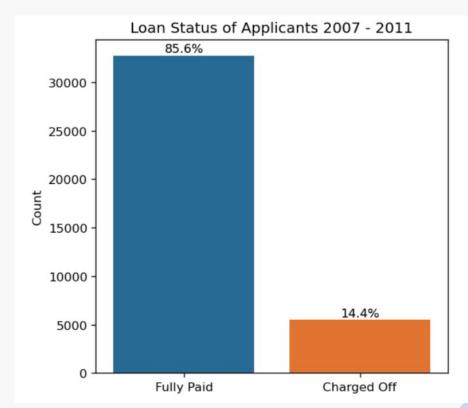




DATA ANALYSIS

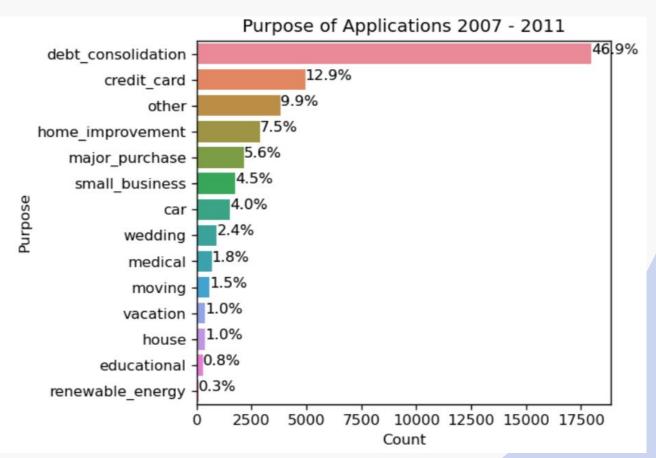


Loan Status

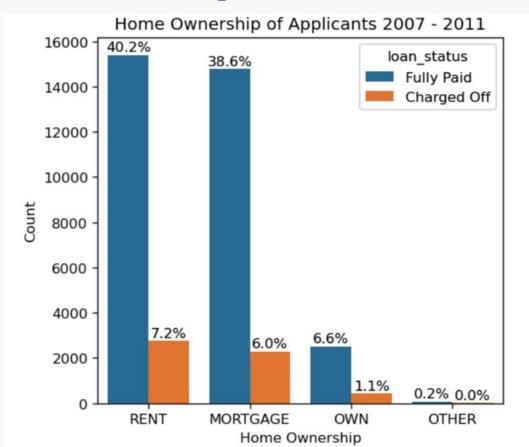




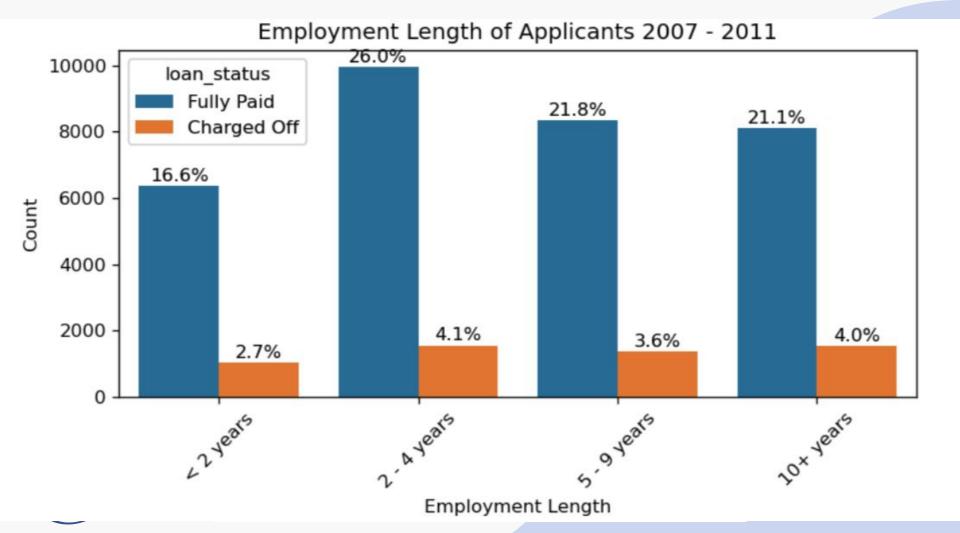
Purpose of Application



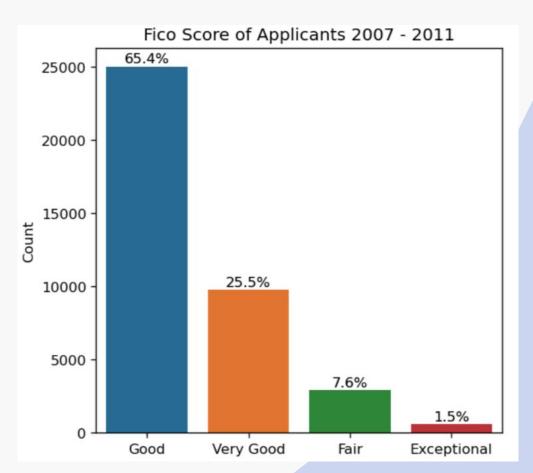
Home Ownership





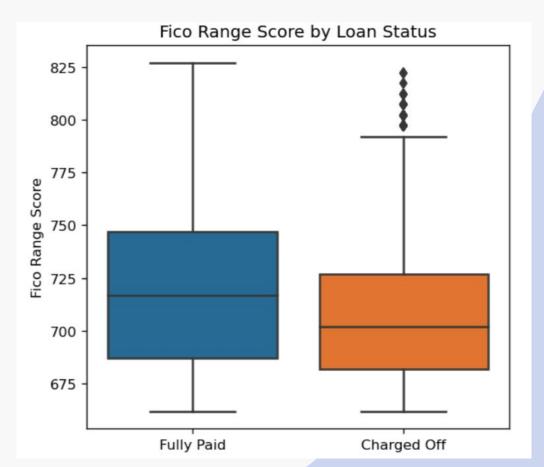


Fico Score



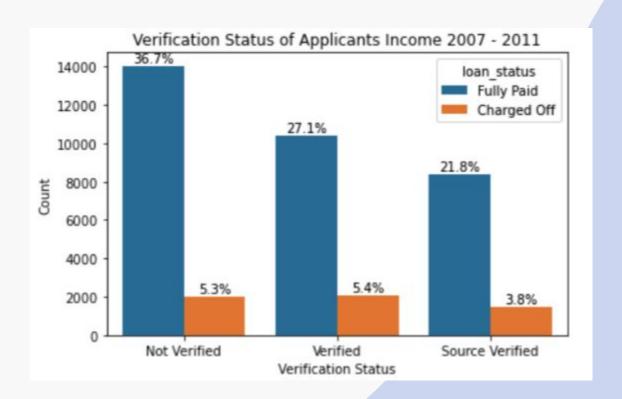


Fico Score



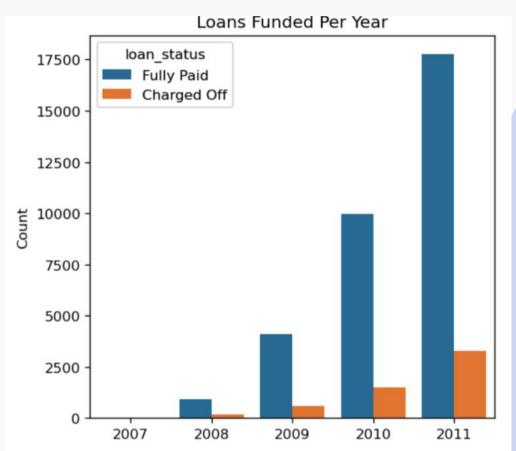


Verification Status



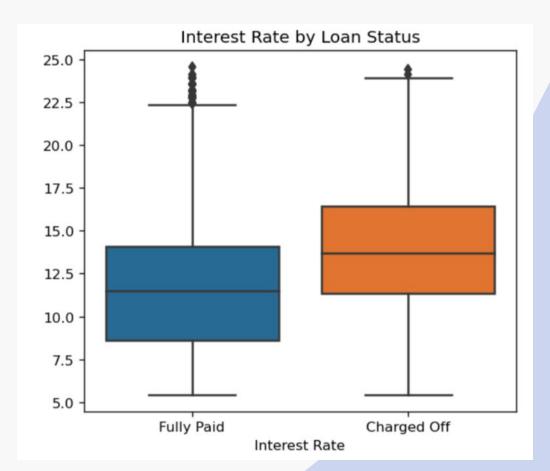


Applications over Time



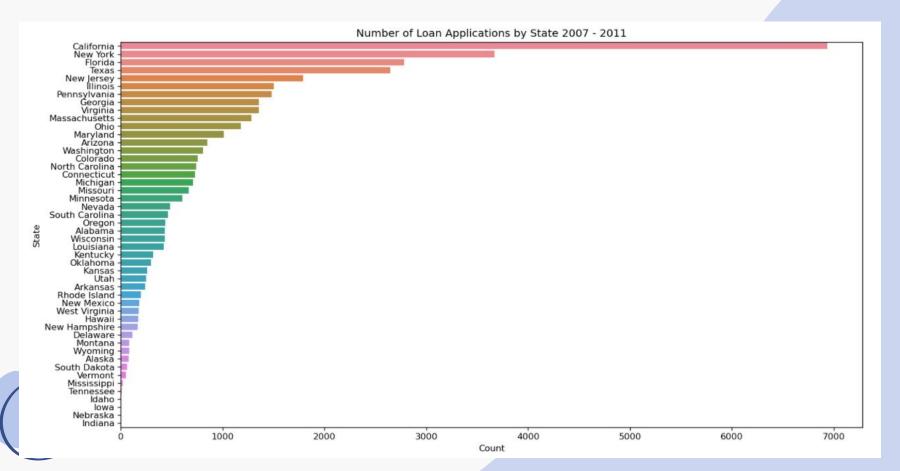


Interest Rate

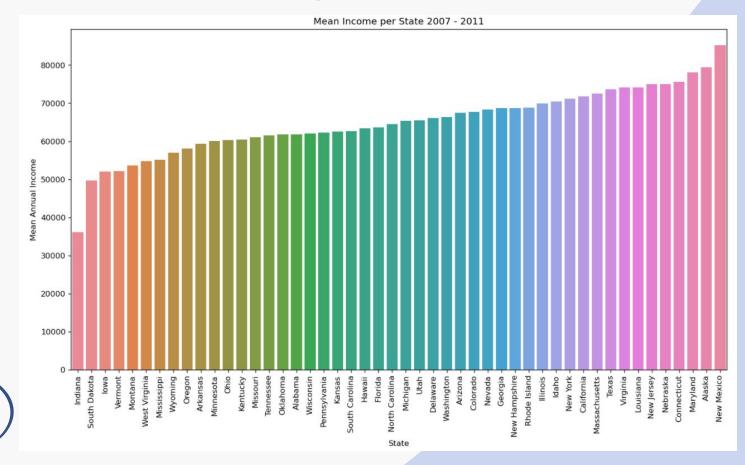




Location



Mean Income by Location





PREDICTIVE MODELS



Target Variable

• Loan Status - Fully Paid or Charged Off where encoded as Fully Paid == 1 and Charged Off == 0 for the purpose of modelling.

Predictors

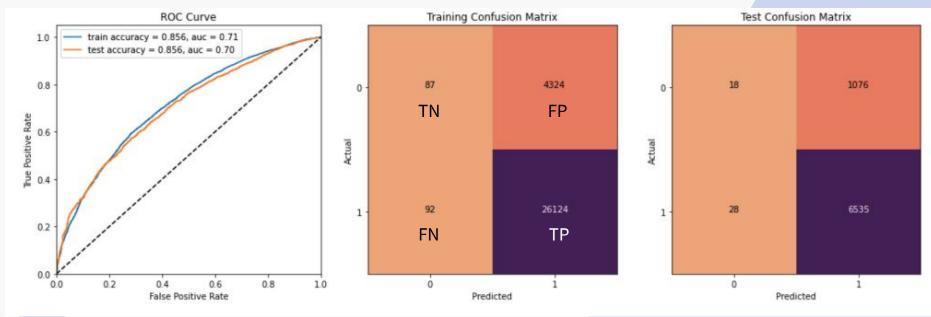
- Categorical variables were converted to dummy variables.
- Panda's Profiling showed the annual income variable is highly skewed so the log-transform of the values was taken.

Train/Test Split

• 80/20 conventional split

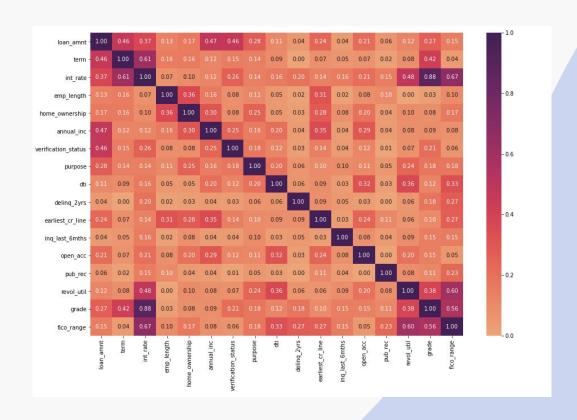


Logistic Regression Model Results





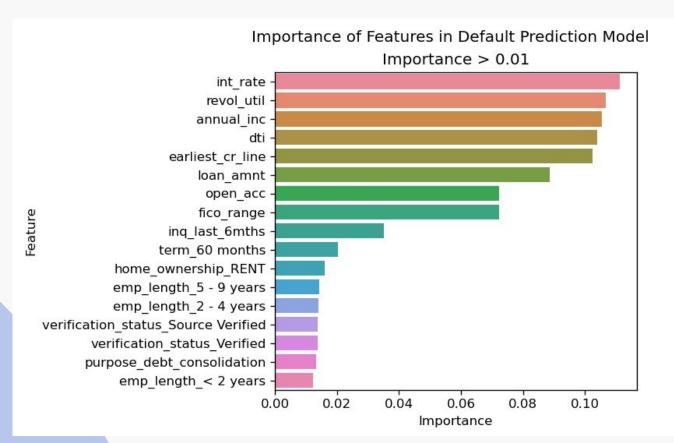
Correlations between Predictors





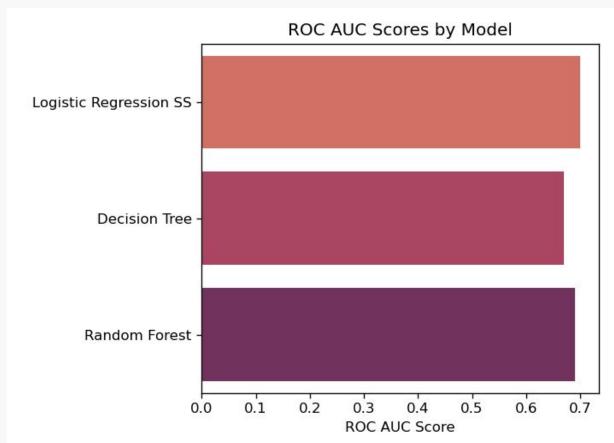
Feature Importance







Model Comparison





Summary

- Interest Rate loans with higher rates more likely to be charged off
- Other features of high importance in default prediction:
 - Revolution Credit Utilization Rate
 - Annual Income
 - Debt to Income Ratio





Thank you for Listening Question Time

