Credit Card Default

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August 16, 2019
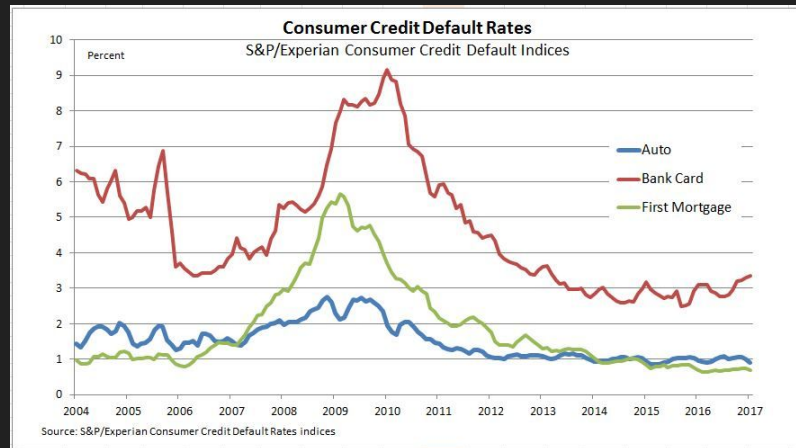
Introduction

Goal:

- Predicting who is about to default on his/her credit card

Data:

- UCI Machine Learning Repository
- Number of Instances: 30,000
- Number of Attributes: 24

Consumer Credit Default Rates
S&P/Experian Consumer Credit Default Indices

Source: S&P/Experian Consumer Credit Default Rates Indices

- Predictors:
  - credit limit
  - basic personal info
  - past payments and bill statements info
- Target Variable:
  - default payment next month

Processed Data

Convert all categorical variables into dummy variables

Transform all numerical features using Yeo-Johnson transformation
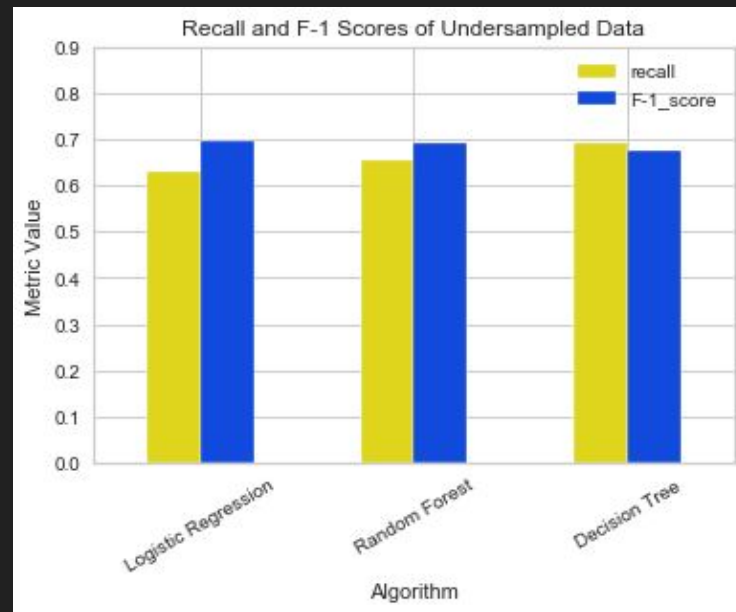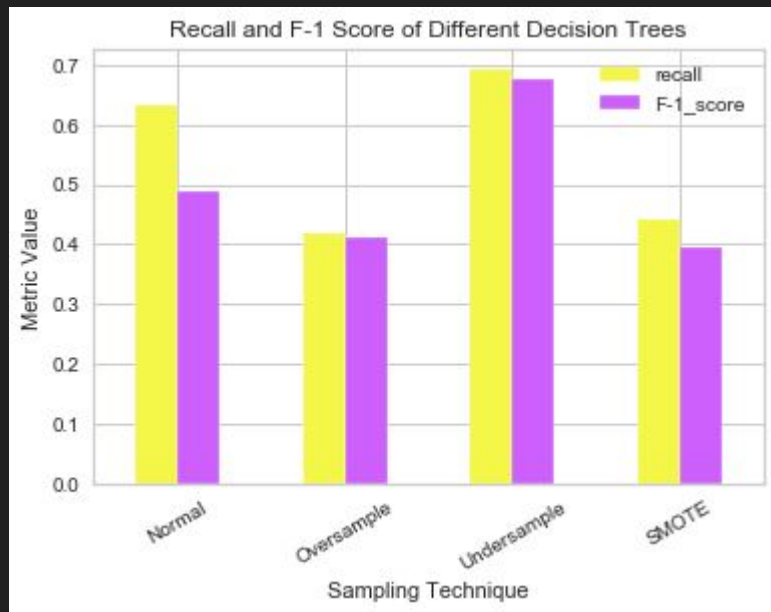
Scale all numerical features using Min-Max scaler

Final datasets:

- As-is
- Over-sampled
- Under-sampled
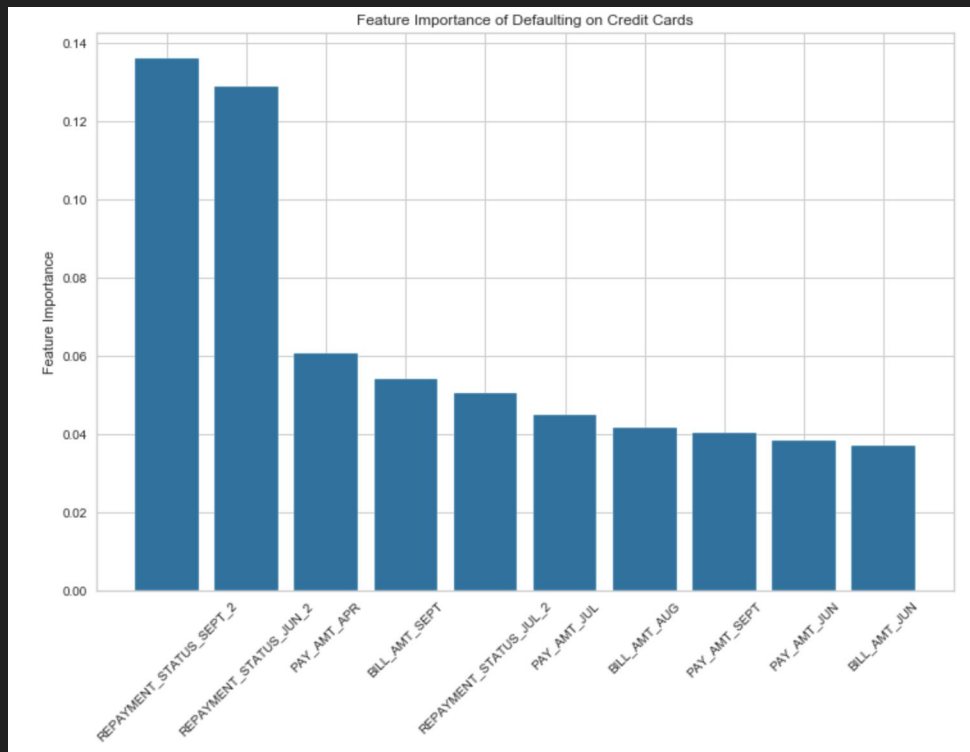- SMOTE

Models Comparison

Best Model

Model:

- Decision Tree

Data:

- Under-sampled

Conclusions

- Decision Tree on under-sampled data yielded the best model, according to recall
- Being 2 months late on payments is a strong signal that a customer will default
- Bill amount also associated with default prediction

Next Steps for Future Improvements

- Collect more (and relevant) features (IS_EMPLOYED, SALARY, etc.)
- Collect more data
- Try more Machine Learning algorithms

Questions?

THANK YOU!

Sources

- https://archive.ics.uci.edu/ml/datasets/default+of+credit+card+clients

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- https://scikit-learn.org/stable/modules/generated/sklearn.ensemble.RandomForestClassifier.html