Identifying Machine Learning Models to Predict Serious Credit Delinquency

Noah Keogh

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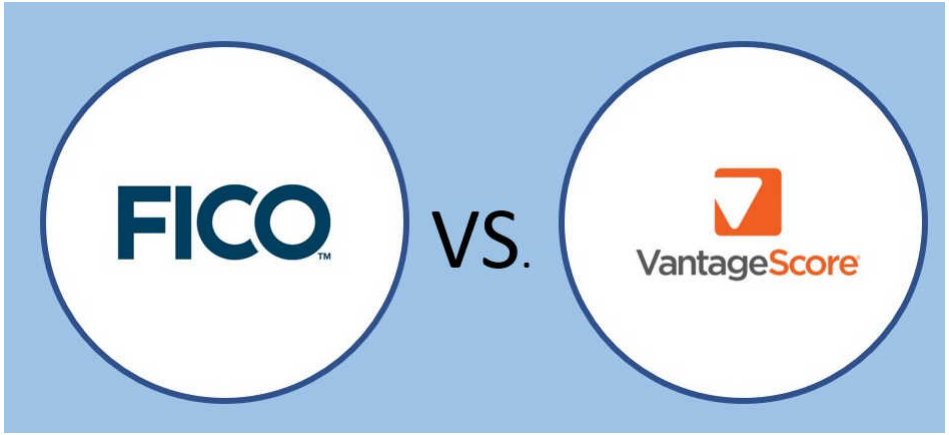
What is a Credit Score?

- Created in 1989 with Introduction of FICO score [1].
- Gives quantitative measure to a borrower's potential riskiness [2].
- Scores range from 300 to 850 [3].

How do Models Calculate Credit Score?

Common Credit Score Factors:

1) Payment history
2) Credit utilization ratio
3) Total debt
4) Credit mix
5) Account age/depth of credit
6) Hard inquiries

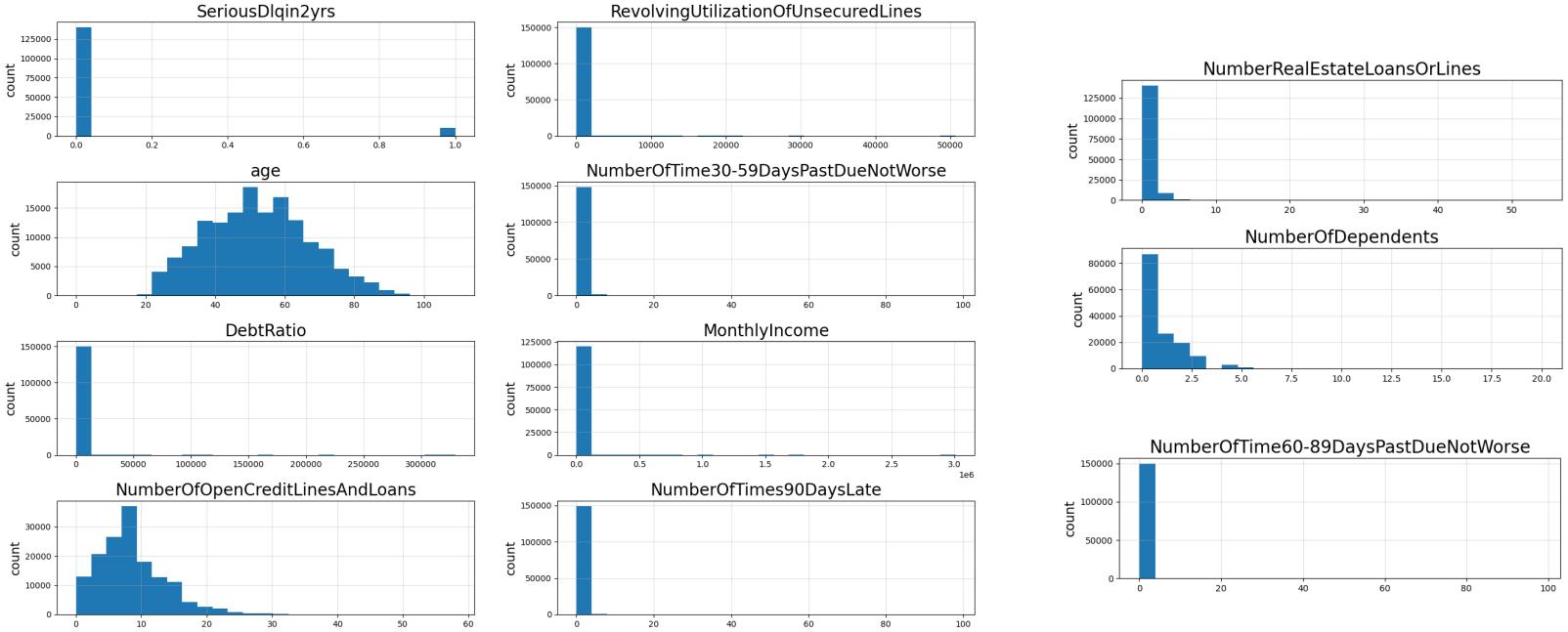


Business Objective

Supervised Model: To classify a given person as being likely or not likely to experience serious delinquency.

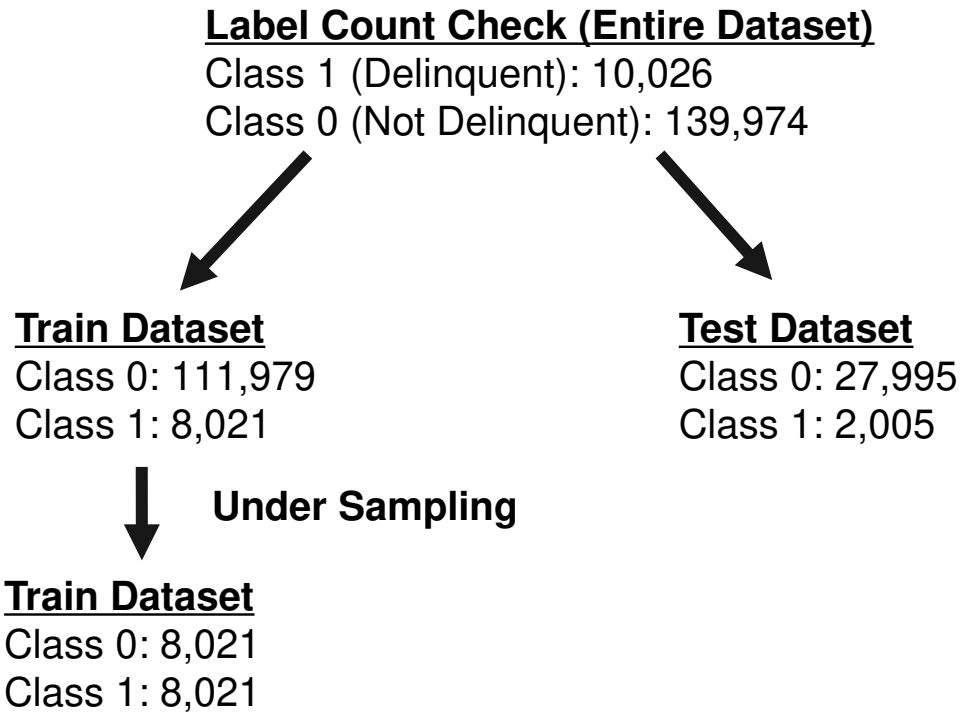
Unsupervised Model: Any financial tendencies or behaviors of people in the dataset. Do these tendencies have any correlation to risk of serious delinquency?

Dataset



- Publicly Available on Kaggle
- 11 Features (one considered label)

Data Wrangling



Data Wrangling

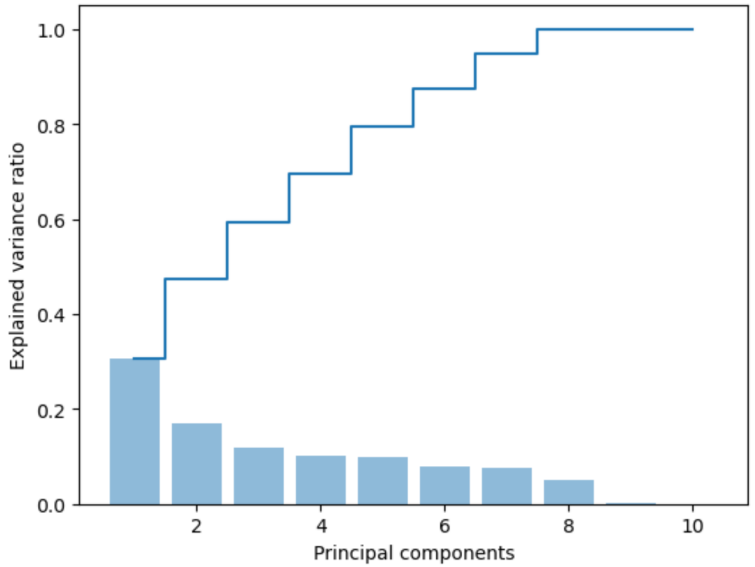
1) Missing Values (Entire Dataset):
Monthly Income: 29,731
Number of Dependents: 3,924

2) Label Balancing (under sampling)

3) Feature Standardization

4) PCA Dimensionality Reduction
10 dimensions 7 dimensions
(94.8% explained variance)

```
Explained Variance Ratio:
[0.30587879 0.16867738 0.11931453 0.10115368 0.09935303 0.07923475
 0.07456405 0.05026824 0.00102548 0.00053007]
Cumulative Sum of EVR:
 [0.30587879 0.47455617 0.5938707  0.69502438 0.79437741 0.87361216
 0.94817621 0.99844445 0.99946993 1.        ]
```



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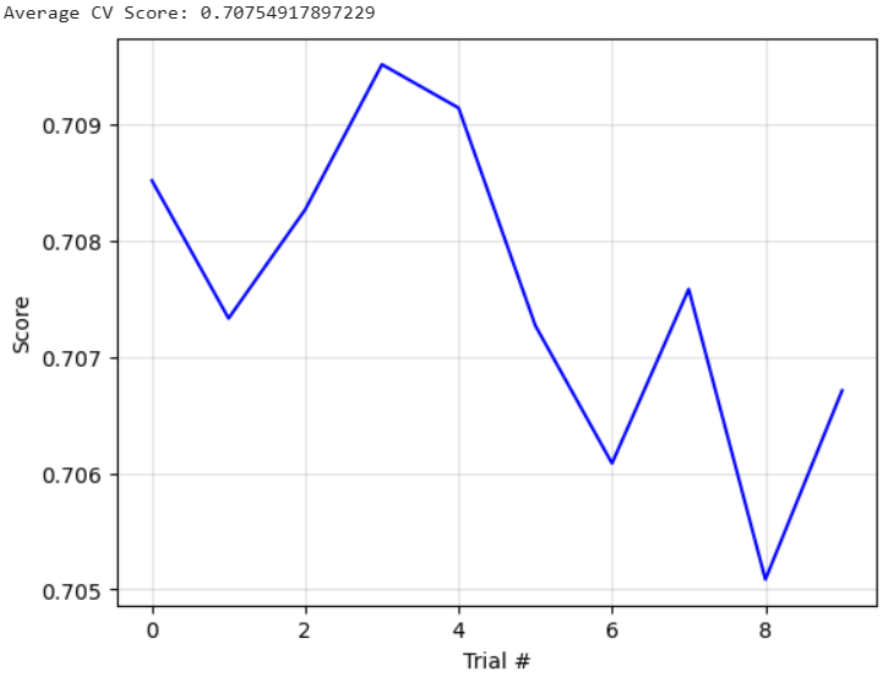
Model Testing

Data Modeling: Logistic Regression



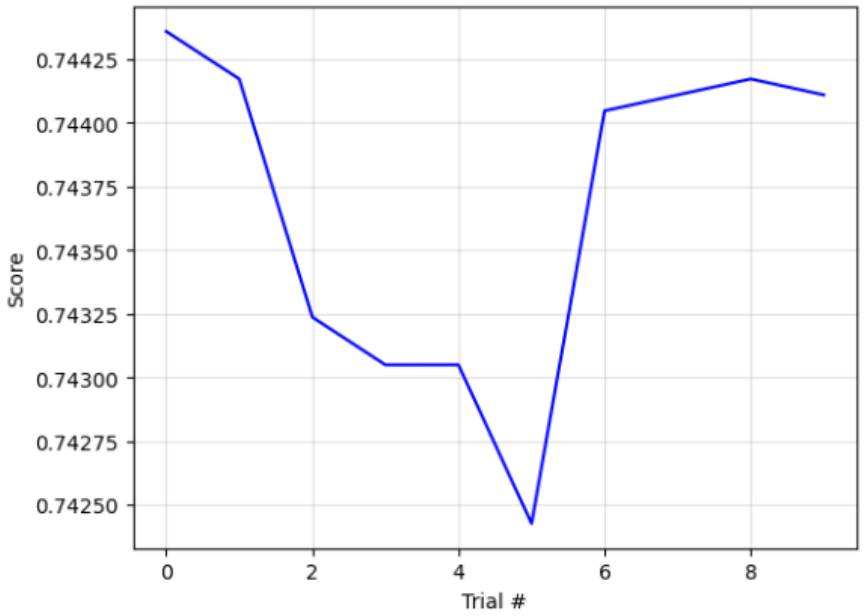
Average Accuracy: 70.75%

Nested Cross Validation:
10 trials conducted

Parameters Tested
1) C: 1, 10, 100
2) Solver: lbfgs, liblinear

Data Modeling: Support Vector Machine (SVM)

Average CV Score: 0.7436730257378245



Average Accuracy: 74.37%

Nested Cross Validation:
10 trials conducted

Parameters Tested
1) C: 1, 10
2) Gamma: 0.01, 0.1
3) Solver: rbf

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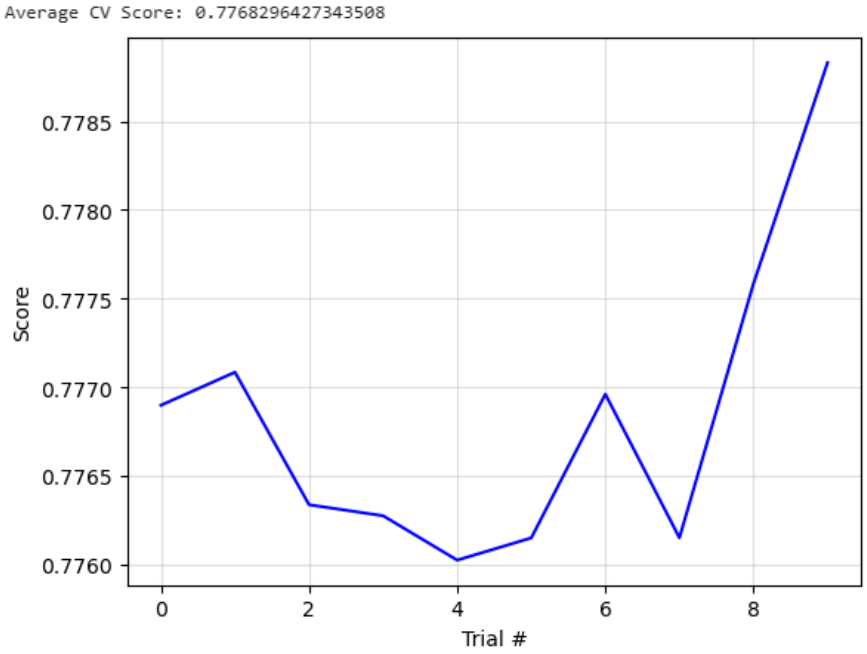
Data Modeling: Random Forest Classifier



Average Accuracy: 77.68%

Nested Cross Validation:
10 trials conducted

Parameters Tested
1) Max Depth: 4, 6, 8
2) Min Samples Leaf: 30, 100

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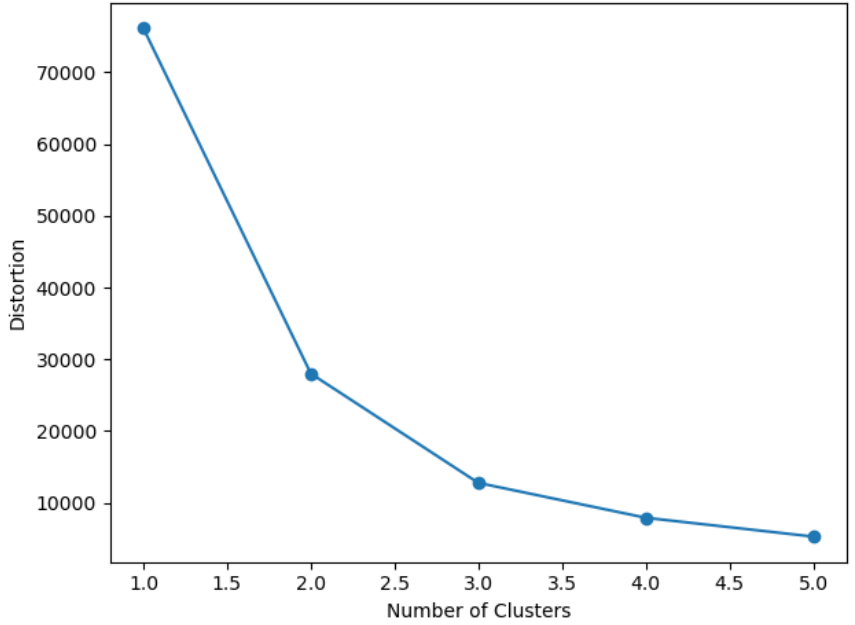
Data Modeling: K-Means-Clustering

1) Dimensionality Reduction
10 dimensions 2 dimensions
(47.46% variance)

2) Clustering
Use K=2 to perform clustering



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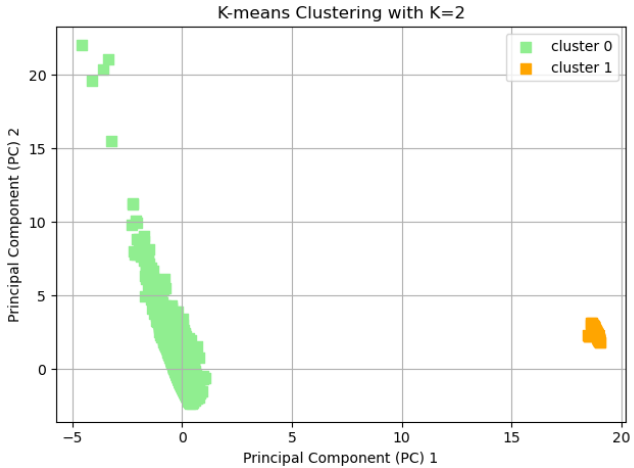
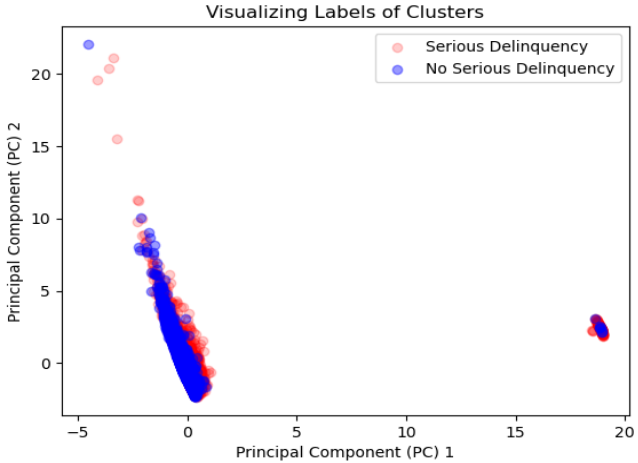
Data Modeling: K-Means-Clustering




```
Most Important Features for PC 1:
NumberOfTime30-59DaysPastDueNotWorse    0.560500
NumberOfTimes90DaysLate                 0.563166
NumberOfTime60-89DaysPastDueNotWorse    0.562465
Name: PC_1, dtype: float64

Most Important Features for PC 2:
MonthlyIncome                           0.443187
NumberOfOpenCreditLinesAndLoans         0.563457
NumberRealEstateLoansOrLines            0.592006
Name: PC_2, dtype: float64
```



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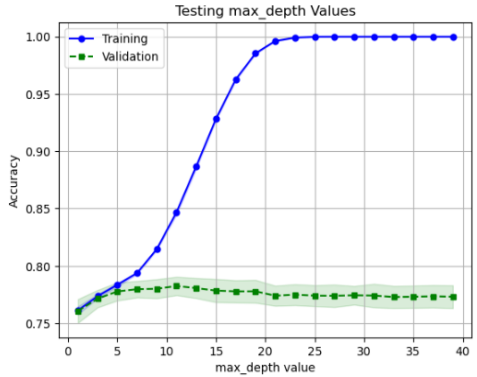
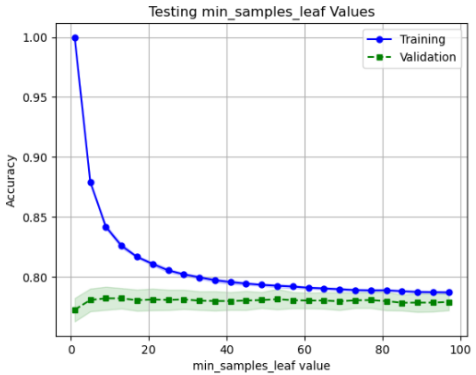
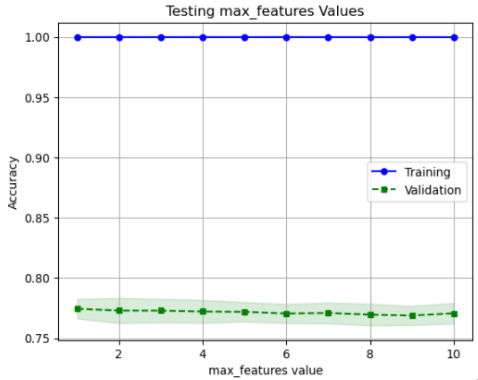
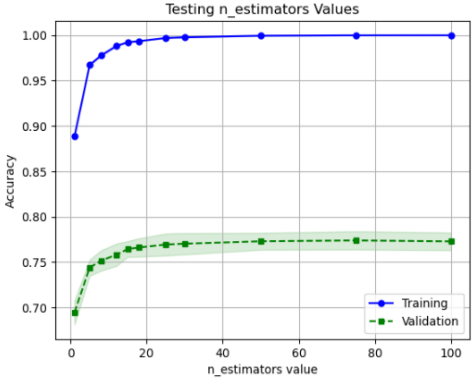
Random Forest Classifier Optimization

Hyperparameter Testing: Validation Curves









Optimal Values for Each Hyperparameter to Test in Randomized Search

max_depth: 3 to 8
min_samples_leaf: 45 to 100
max_features: 1 to 10
n_estimators: 15 to 25

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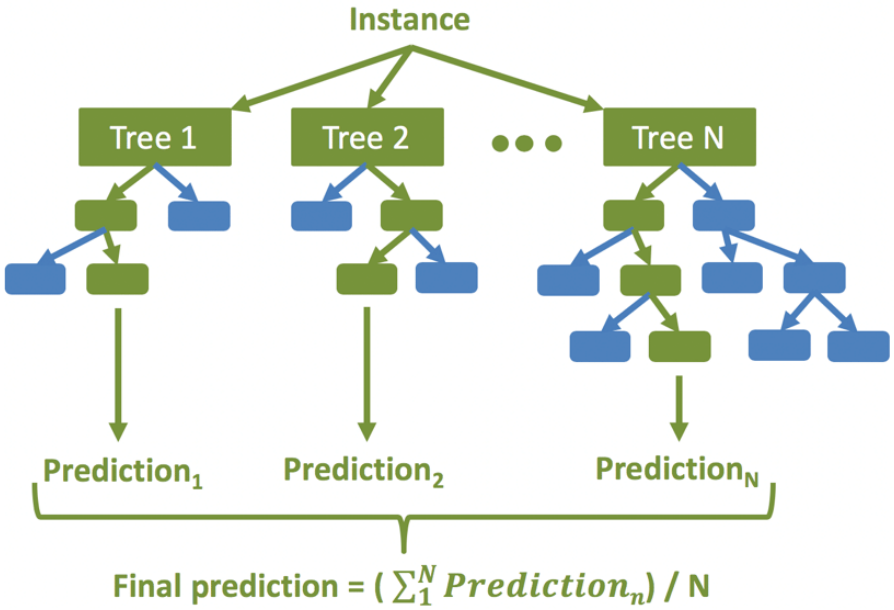
Randomized CV Search for Final Model



Accuracy in CV: 78.12%

Model Hyperparameters

1) max_depth: 7
2) max_features: 5
3) min_samples_leaf: 48
4) n_estimators: 19

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Assessing Accuracy

Test Dataset
Class 0: 27,995
Class 1: 2,005

```
Number of Samples: 30000
Number Positive: 2005
Number Negative: 27995
------------------
True Positive (TN): 1579
True Negative (TN): 21815
False Positive (TN): 6180
False Negative (TN): 426
```

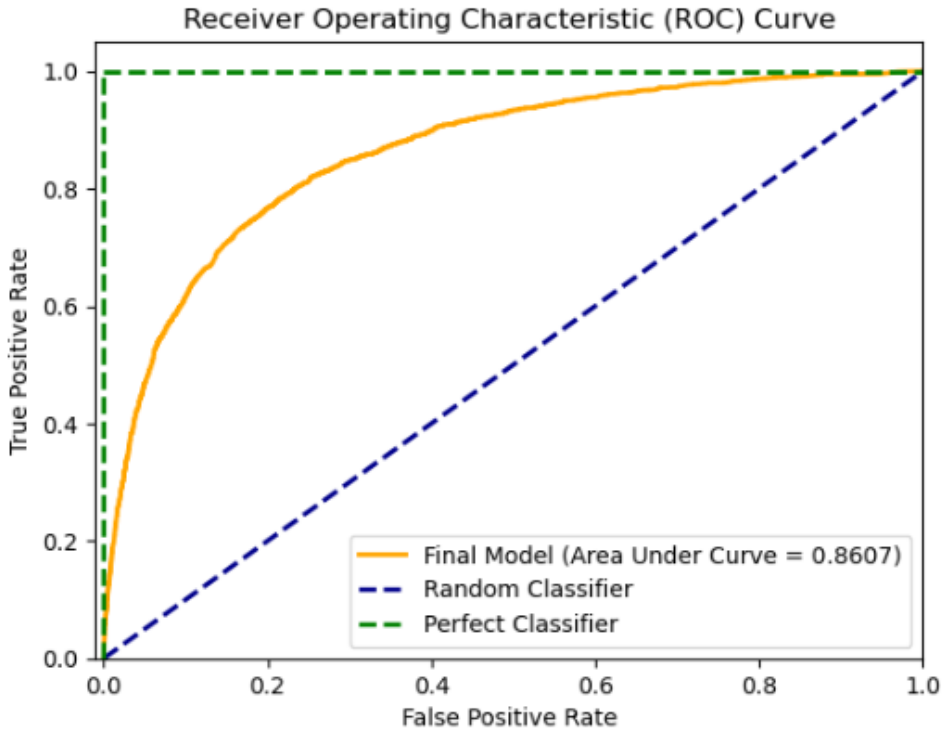
```
Model Accuracy: 0.7798
Precision: 0.2035
Recall: 0.7875
F1 Score: 0.3234
```

Assessing Accuracy

Actual \ Predicted	0	1
0	TN	FP
1	FN	TP

False Positive Rate (FPR) = $\frac{FP}{(FP + TN)}$

True Positive Rate(TPR) = $\frac{TP}{(TP + FN)}$



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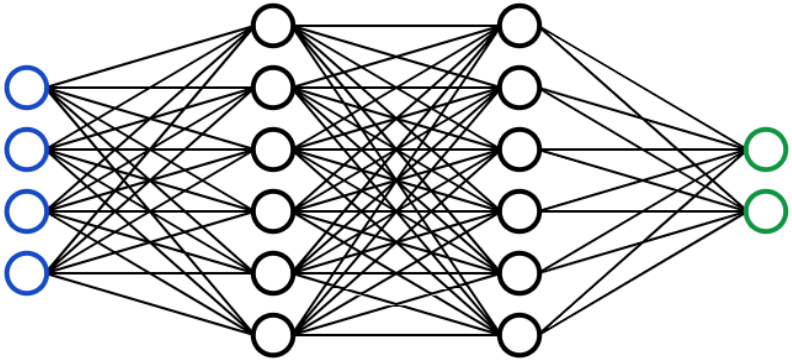
Future Explorations



- Exploring different models such as neural networks.
- Further optimizing tested models such as logistic regression and SVMs.
- Collecting more data and training examples to improve performance.

Questions?

Sources

[1] https://www.creditrepair.com/blog/education/when-were-credit-scores-invented/
[2] https://www.onemainfinancial.com/resources/credit/credit-scoring-models#:~:text=A%20credit%20scoring%20model%20is%20an%20algorithm%20used,helps%20lenders%20make%20informed%20decisions%20when%20approving%20loans.
[3] https://www.bankrate.com/personal-finance/credit/no-credit-score-zero-credit/