

# Predictive Modeling Results Report Date: April 19, 2020

## **A. Data Input Summary**

1. Total Data Sample: 3,800

2. 80% of data used for training the model: 3,040

3. 20% of data used for testing the model: 760

- 4. Donation Columns:
  - a. funding1
  - b. fund2
  - c. donations3
  - d. gifts4
  - e. donat5
  - f. transac6
  - g. grants7
  - h. money8
  - i. gift9
  - j. gift10
  - k. funding11
  - I. donations12
  - m. donations13
- 5. Categorical Columns:
  - a. Cand\_Office

## B. Running the Predictive Model: A Step by Step Guide

- 1. Read the input data file provided
- 2. Cleaning up of data: remove null rows and columns and impute missing values
- 3. Identifying columns containing categorical and textual data and convert it to numerical values
- 4. Assigning Target Value: Target values are the dependent variable
- 5. Splitting the dataset for training and testing to train a total of 10 different classifiers (for example Logistic Regression, Naive Bayes and Random Forest etc)
- 6. Calculating Feature Importance for each classifier. Feature importance gives a score for each feature of your data

- 7. Plot Confusion Matrix and Classification report. A confusion matrix is a table that is used to describe the performance of a model
- 8. Identifying and selecting the best fit classifier using the F1-Score. The F1-score is a measure of a test's (model's) accuracy
- 9. Receiver Operating Characteristic (ROC) Curve. ROC is a probability curve. It tells how much a model is capable of distinguishing between classes
- 10. Identifying the optimal threshold (accuracy of the model) and predict

## C. Important Terms Used in Predictive Modeling

- 1. F1-score: It is a harmonic mean of precision and recall
- 2. Precision: It is a fraction of correctly classified instances among all predicted instances
- 3. Recall: It is fraction of correctly classified instances among all actual/valid instances
- 4. Support: Number of samples used for the experiment
- 5. Confusion Matrix Plot: It is a plot of the true count (x-axis) versus predicted count (y-axis) for both the classes. The top left box represents the count of true negatives, the top right box represents the count of false negatives, bottom left box represents the count of false positive and bottom rightbox represents the count of true positives
- 6. Feature Importance Plot: Y-axis: variable present in input file and X-axis: relative % of feature importance

## D. Best Fit Model Used in Predictive Modeling

Best fit classifier is selected (out of 10 classifiers) based on F1-score and used for prediction. We identified optimal threshold to separate donor and non-donor classes. Following are f1-score, threshold and count of donor samples

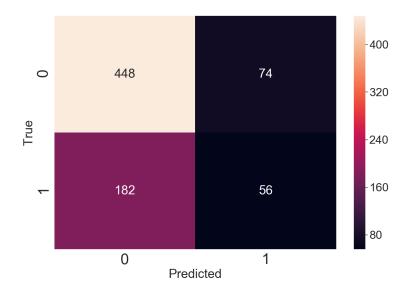
#### Model: BernoulliNB

a. F1-score: 0.63

b. Threshold used: 0.8

c. Donor predicted: 60.92% (2,315 out of 3,800)

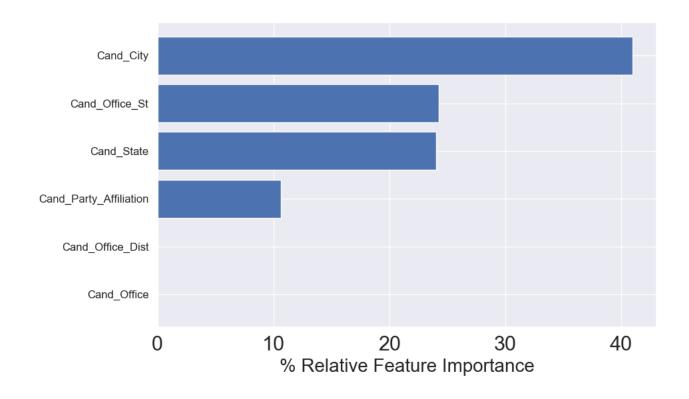
#### # Confusion Matrix Plot



# # Classification Report Table

Index	F1-score	Precision	Recall	Support
Non-donor class	0.78	0.71	0.86	522
Donor class	0.3	0.43	0.24	238
Macro avg	0.54	0.57	0.55	760
Weighted avg	0.63	0.62	0.66	760

# # Feature Importance Plot



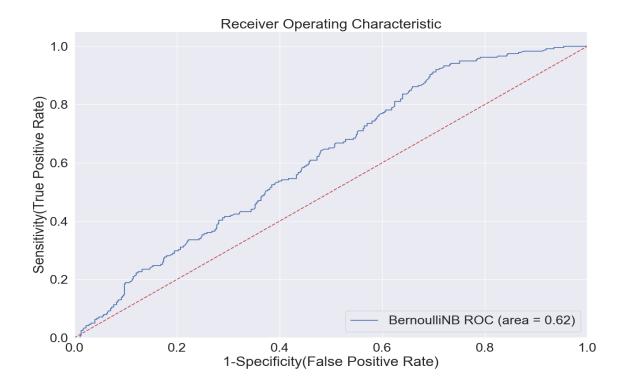
# # Receiver Operating Characteristic (ROC) Curve

It is a plot of the false positive rate (x-axis) versus the true positive rate (y-axis). True positive rate or

sensitivity describes how good the model is at predicting the positive class when the actual outcome is positive. False positive rate explains how often a positive class is predicted when the actual result is negative.

A model with high accuracy is represented by a line that travels from the bottom left of the plot to the top left and then across the top to the top right and has Area Under Curve (AUC) as 1. A model with less accuracy is represented by a diagonal line from the bottom left of the plot to the top right and has an AUC of 0.5.

We can compare multiple models using AUC value; the best model will have AUC close to 1.



#### # Correlation Plot

Correlation explains how one or more variables are related to each other.

