# Donor Application Screening

### Introduction

- Donorchoose.org is an organisation which funds school level projects.
- Teachers from different schools request for the material they require for the students.
- Each year they receive thousands of applications for which they require lot of manpower and time to screen this applications.
- So, to make it more easier they want to automate this application screening process.

### **Objective**

- The goal here is to predict whether the proposal would be approved or not based on the description provided in application.
- Use efficient algorithms for training the model for predicting accurately.

### **Dataset**

- Dataset we worked on is taken from kaggle.
  (<a href="https://www.kaggle.com/c/donorschoose-application-screening/data">https://www.kaggle.com/c/donorschoose-application-screening/data</a>)
- The size of zip file of this dataset is 203.9MB. It contains following files:
- Train.csv: It contains the data of 182080 proposals and has 16 features.
- Test.csv: test data for 78035 proposals and has 15 features.
- Resource.csv: data of resources requested, their quantity and price. It has 1541272 entries and have 4 features.

### **Technologies Used**

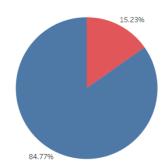
- Anaconda Navigator for jupyter notebook
- Python as Programming Language
- Tableau for better visualization
- Libraries used: scipy, numpy, pandas, nltk, sklearn, matplotlib, plotly, seaborn

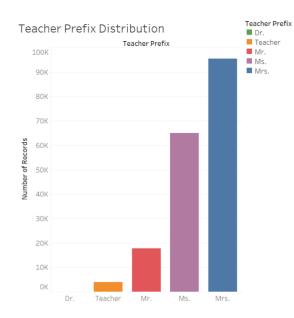


### **Visualization**



#### Class Distribution





#### Project Category Distribution



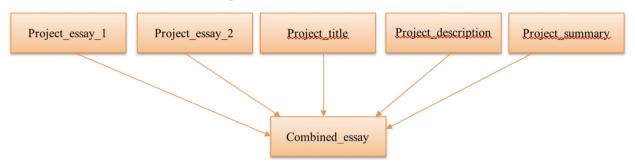
### **Data Preprocessing**

- We merged train.csv and resource.csv based on id.
- Based on visualisation and data exploration, we did feature extraction. The features which we selected to use were text based.
- Replacing null values: We checked for the null values in the data set and replaced those null values with some string.
- Applied Tfidf vectorization on the text features of data set for removing special characters, stop words, stemming and for generating tokens.

### **Data Preprocessing**

#### Feature extraction:

Created Text feature from using features



Created total price and total quantity feature from resource.csv

### **Data Preprocessing**

#### Conversion of Categorical Features into Numeric Vectors

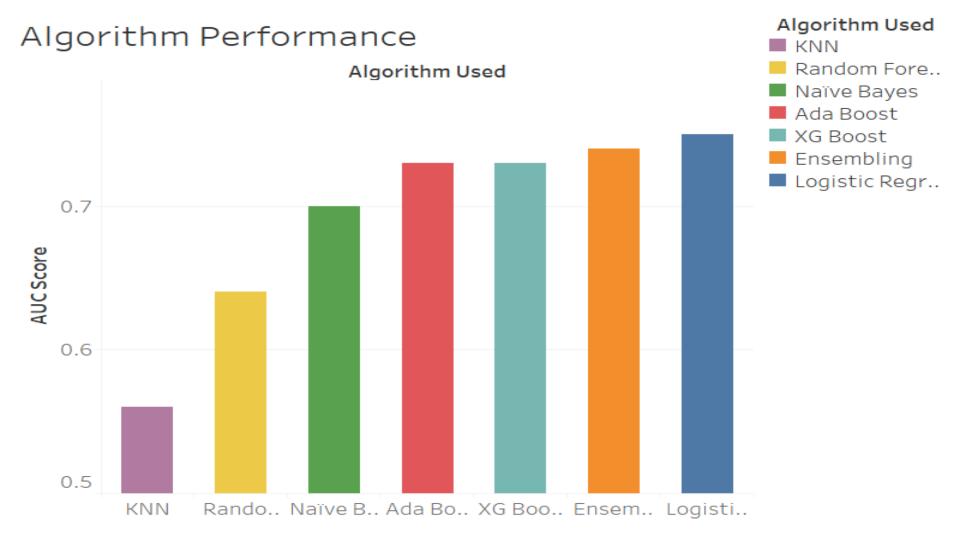
- school\_state
- project\_subject\_categories
- project\_subject\_subcategories
- teacher\_prefix

#### Feature Removal

- id
- project\_title
- project\_essay\_3
- project\_essay\_4
- teacher\_number\_of\_previously\_posted\_projects
- Categorical Features

### **Algorithms**

- Logistic regression
- SVM
- K-nearest neighbors
- Naive bayes
- Xgboost
- Lightgbm
- Random forest
- Decision tree/Gradient Boosting based
  - Adaboost
  - Xgboost
  - Lightgbm



### Algorithms- worked for us

Based on the accuracy, selected the following algorithms for final model.

- Logistic regression
- Naive bayes
- Xgboost
- Random forest

### **Evaluation and Methodology**

- We followed n-fold cross validation for value of n=5. Considered average as final validation score.
- We performed cross validation on all experimented algorithm and then selected the models which gave best accuracy.
- Ensemble learning: We used the combination of the algorithms which gave us best results by ensembling it.

## Things that worked /Things that didn't worked

- > Things that worked:
- Referred kaggle's beginner tutorial.
- All worked on preprocessing to make it faster and easier.
- Experimenting different models to get more accuracy for getting optimized results.
- > Things that didn't work well:
- Some algorithms were taking very long time to implement. Like SVM took a lot of time to implement.

### Conclusion

- Spending quality of time in preprocessing of data is necessary to get optimized results.
- Feature Engineering can make algorithms to work better.
- Simple algorithms works better than complex on some datasets.

### References

- <a href="https://towardsdatascience.com/machine-learning-nlp-text-classification-using-scikit-learn-python-and-nltk-c52b92a7c73a">https://towardsdatascience.com/machine-learning-nlp-text-classification-using-scikit-learn-python-and-nltk-c52b92a7c73a</a>
- https://www.youtube.com/watch?v=ZiKMluYidY0&t=4931s
- https://www.kaggle.com/c/donorschoose-application-screening
- http://scikitlearn.org/stable/modules/generated/sklearn.feature\_extraction.text.TfidfVectorizer.html