# **Decision Trees on Donors Choose dataset**

## **Importing Libraries**

```
In [1]:
              %matplotlib inline
           2 import warnings
           3 warnings.filterwarnings("ignore")
           5 import pandas as pd
           6 import numpy as np
           7 import math as m
           8 import matplotlib.pyplot as plt
           9 import seaborn as sns
          10 import nltk
          11 import re
          12 | from scipy import sparse
          13 from sklearn.feature_extraction.text import TfidfVectorizer
          14 | from sklearn.feature_extraction.text import CountVectorizer
          15 | from sklearn.preprocessing import Normalizer
          16 | from sklearn.metrics import confusion_matrix
          17 | from sklearn.model_selection import train_test_split
          18 from sklearn import metrics
          19 from sklearn.metrics import roc_curve, auc
          20 | from tqdm import tqdm
          21
          22 import nltk
          23  nltk.download('vader_lexicon')
          24 | from nltk.sentiment.vader import SentimentIntensityAnalyzer
          25 | sid = SentimentIntensityAnalyzer()
         [nltk_data] Downloading package vader_lexicon to /root/nltk_data...
In [2]:
           1 import tensorflow as tf
           2 tf.test.gpu_device_name()
Out[2]: '/device:GPU:0'
         Importing Dataset
In [9]:
              !wget --header="Host: doc-14-3k-docs.googleusercontent.com" --header="User-Agent: Mozilla/5.0 (Windows NT 10.0; Win64
         --2020-04-21 13:29:17-- https://doc-14-3k-docs.googleusercontent.com/docs/securesc/8h8uvuh5ifib89027b7ihknt22nu5vgc/t7
         va9bgvmpo50k1actnt4qtpenckrn6g/1587475725000/00484516897554883881/05866892802988797180/1GU3LIJJ3zS1xLXXe-sdItSJHtI5txjV
         O?e=download&authuser=0&nonce=n1kem9jsfroba&user=05866892802988797180&hash=52na0kdlokv88v5r8oe0k3796nblteim (https://do
         c-14-3k-docs.googleusercontent.com/docs/securesc/8h8uvuh5ifib89027b7ihknt22nu5vgc/t7va9bgvmpo50k1actnt4qtpenckrn6g/1587
         475725000/00484516897554883881/05866892802988797180/1GU3LIJJ3zS1xLXXe-sdItSJHtI5txjVO?e=download&authuser=0&nonce=n1kem
         9jsfroba&user=05866892802988797180&hash=52na0kdlokv88v5r8oe0k3796nblteim)
         Resolving doc-14-3k-docs.googleusercontent.com (doc-14-3k-docs.googleusercontent.com)... 64.233.167.132, 2a00:1450:400
         c:c0a::84
         Connecting to doc-14-3k-docs.googleusercontent.com (doc-14-3k-docs.googleusercontent.com) 64.233.167.132 :443... connec
         ted.
         HTTP request sent, awaiting response... 200 OK
         Length: unspecified [text/csv]
         Saving to: 'preprocessed_data.csv'
         preprocessed_data.c
                                                       ] 118.69M 52.3MB/s
                                                                               in 2.3s
         2020-04-21 13:29:20 (52.3 MB/s) - 'preprocessed_data.csv' saved [124454659]
           1 | data = pd.read_csv('preprocessed_data.csv')
           1 data.head(2)
In [11]:
Out[11]:
             school_state teacher_prefix project_grade_category teacher_number_of_previously_posted_projects project_is_approved clean_categories clean_s
          0
                                                                                           53
                                                                                                                   math_science
                                            grades_prek_2
                                                                                                              1
                     ca
                                mrs
                                                                                                                                  hea
          1
                                                                                                                    specialneeds
                     ut
                                 ms
                                              grades_3_5
```

```
In [0]:
           1 negative = []
              positive = []
              neutral = []
              compound = []
           5
              def update_sentiments(values):
           6
           7
                  negative.append(values["neg"])
           8
                  positive.append(values["pos"])
                  neutral.append(values["neu"])
                  compound.append(values["compound"])
          10
In [13]:
           1 | from tqdm import tqdm
              for essay in tqdm(data["essay"]):
                  update_sentiments(sid.polarity_scores(essay))
           3
          100%
                       | 109248/109248 [03:27<00:00, 526.52it/s]
              data["neg"] = negative
In [0]:
           2 data["pos"] = positive
           3 | data["neu"] = neutral
                                             # adding new features to dataset based on Sentiment Intensity Analyzer
              data["compound"] = compound
In [15]:
           1 data.head(1)
Out[15]:
             school_state teacher_prefix project_grade_category teacher_number_of_previously_posted_projects project_is_approved clean_categories clean_s
          0
                                                                                              53
                                                                                                                      math_science
                                             grades_prek_2
                                                                                                                 1
                     ca
                                 mrs
```

## Splitting Data Into Train And Cross Validation(or test): Stratified Sampling

```
In [0]: 1  y = data['project_is_approved'].values
2  X = data.drop(['project_is_approved'], axis=1)

In [0]: 1  X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, stratify=y,random_state=42)

In [18]: 1  print("Total data points in Train Dataset =",len(y_train))
2  print("Total data points in Test Dataset =",len(y_test))

Total data points in Train Dataset = 73196
```

## Make Data Model Ready: Encoding Eassay(text feature)

Total data points in Test Dataset = 36052

## TFIDF Vectorizer

## TFIDF W2V

```
In [20]:
              !wget --header="Host: doc-10-3k-docs.googleusercontent.com" --header="User-Agent: Mozilla/5.0 (Windows NT 10.0; Win64
         --2020-04-21 13:34:26-- https://doc-10-3k-docs.googleusercontent.com/docs/securesc/8h8uvuh5ifib89027b7ihknt22nu5vgc/t3
         0frjril2s6r6rt84jvdfcc5im19uck/1587476025000/00484516897554883881/05866892802988797180/1zJbDcbwvM2ueudqJHPp0b3Z9V2QrGd2
         r?e=download&authuser=0&nonce=uaih9emd25nk6&user=05866892802988797180&hash=939k2qd0inucqnck97ei14sqtt2c7if3 (https://do
         c-10-3k-docs.googleusercontent.com/docs/securesc/8h8uvuh5ifib89027b7ihknt22nu5vgc/t30frjril2s6r6rt84jvdfcc5im19uck/1587
         476025000/00484516897554883881/05866892802988797180/1zJbDcbwvM2ueudqJHPpOb3Z9V2QrGd2r?e=download&authuser=0&nonce=uaih9
         emd25nk6&user=05866892802988797180&hash=939k2qd0inucqnck97ei14sqtt2c7if3)
         Resolving doc-10-3k-docs.googleusercontent.com (doc-10-3k-docs.googleusercontent.com)... 64.233.167.132, 2a00:1450:400
         c:c0a::84
         Connecting to doc-10-3k-docs.googleusercontent.com (doc-10-3k-docs.googleusercontent.com) 64.233.167.132:443... connec
         HTTP request sent, awaiting response... 200 OK
         Length: unspecified [application/octet-stream]
         Saving to: 'glove_vectors'
         glove_vectors
                                                      ] 121.60M 46.6MB/s
                                                                             in 2.6s
                                        <=>
         2020-04-21 13:34:29 (46.6 MB/s) - 'glove_vectors' saved [127506004]
 In [0]:
             import pickle
           1
             with open('glove_vectors', 'rb') as f:
                  model = pickle.load(f)
           4
                  glove_words = set(model.keys())
 In [0]:
           1 tfidf_model = TfidfVectorizer()
           2 tfidf_model.fit(X_train['essay'].values)
           3 | dictionary = dict(zip(tfidf_model.get_feature_names(), list(tfidf_model.idf_)))
           4 tfidf_words = set(tfidf_model.get_feature_names())
In [23]:
           1 #TFIDF W2V for train dataset
             train_tfidf_w2v_essays = [] # the tfidf-w2v for each essay is stored in this list
              for sentence in tqdm(X_train['essay']):
                  vector = np.zeros(300)
                  tf_idf_weight =0;
           5
                  for word in sentence.split():
           6
           7
                      if (word in glove_words) and (word in tfidf_words):
           8
                          vec = model[word]
           9
                          tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))
          10
                          vector += (vec * tf_idf)
          11
                          tf_idf_weight += tf_idf
          12
                  if tf_idf_weight != 0:
          13
                      vector /= tf_idf_weight
          14
                  train_tfidf_w2v_essays.append(vector)
          15 | X_train_essay_tfidf_w2v= sparse.csr_matrix(train_tfidf_w2v_essays)
             print("After vectorizations")
             print(X_train_essay_tfidf_w2v.shape, y_train.shape)
          17
             print("="*40)
         100% | 73196/73196 [02:31<00:00, 483.50it/s]
         After vectorizations
         (73196, 300) (73196,)
In [24]:
           1 | #TFIDF W2V for test dataset
             test_tfidf_w2v_essays = [] # the tfidf_w2v for each essay is stored in this list
           3
             for sentence in tqdm(X_test['essay']):
           4
                 vector = np.zeros(300)
           5
                  tf_idf_weight =0;
                  for word in sentence.split():
           6
           7
                      if (word in glove_words) and (word in tfidf_words):
           8
                          vec = model[word]
           9
                          tf_idf = dictionary[word]*(sentence.count(word)/len(sentence.split()))
                          vector += (vec * tf_idf)
          10
          11
                          tf_idf_weight += tf_idf
                 if tf_idf_weight != 0:
          12
          13
                     vector /= tf_idf_weight
          14
                 test_tfidf_w2v_essays.append(vector)
          15 | X_test_essay_tfidf_w2v= sparse.csr_matrix(test_tfidf_w2v_essays)
          16 | print("After vectorizations")
              print(X_test_essay_tfidf_w2v.shape, y_test.shape)
              print("="*40)
                          36052/36052 [01:15<00:00, 479.44it/s]
         After vectorizations
         (36052, 300) (36052,)
         _____
```

#### **Response Encoding**

```
In [0]:
             def fit(feature):
          1
                 X_train['class_label']=y_train # adding 'project_is_approved' column to x_train
                 count = X_train[ feature ].value_counts() # getting value counts(denominator) of each category
          3
          4
                 feature_dictionary = dict()
          5
                 for i, denominator in count.items():
          6
                     vector = []
          7
                     for j in range(2):
                         compare =X_train.loc[ ( X_train['class_label'] == j ) & (X_train[feature] == i ) ]
          8
          9
                         vector.append( len( compare) / denominator )
                     feature_dictionary[i] = vector # adding probability of each class label for a pariticular category of feat
         10
         11
                 return feature_dictionary
         12
             def transform(feature, df ):
         13
                 feature_dictionary = fit(feature)
         14
                 count = X_train[feature].value_counts()
         15
                 f=[]
         16
                 for cat in df[feature]:
                     if cat in dict( count ).keys():# transform test data with trainning probabilities
         17
                         f.append( feature_dictionary[cat] )
         18
         19
         20
                         f.append([0.5, 0.05])
         21
                 return f
```

#### **Encoding Categorical Features: School State**

```
In [0]: 1  X_train_state_rc =np.array(transform('school_state',X_train))
2  X_test_state_rc =np.array(transform('school_state',X_test))
3  print("After vectorizations")
4  print(X_train_state_rc.shape, y_train.shape)
5  print(X_test_state_rc.shape, y_test.shape)
6  print("="*40)

After vectorizations
(73196, 2) (73196,)
(36052, 2) (36052,)
```

#### **Encoding Categorical Features: teacher\_prefix**

```
In [0]: 1 X_train_teacher_rc =np.array(transform('teacher_prefix',X_train))
2 X_test_teacher_rc = np.array(transform('teacher_prefix',X_test))
3 print("After vectorizations")
4 print(X_train_teacher_rc.shape, y_train.shape)
5 print(X_test_teacher_rc.shape, y_test.shape)
6 print("="*40)

After vectorizations
(73196, 2) (73196,)
(36052, 2) (36052,)
```

## Encoding Categorical Features: project\_grade\_category

\_\_\_\_\_\_

### **Encoding Categorical Features: clean\_categories**

```
(73196, 2) (73196,)
(36052, 2) (36052,)
```

#### Encoding Categorical Features: clean\_subcategories

#### **Encoding Numerical Features**

#### **Encoding Numerical Feature : price**

#### Encoding Numerical Features:teacher\_number\_of\_previously\_posted\_projects

#### **Encoding Numerical Features:neg**

\_\_\_\_\_

```
In [0]:
          1 normalizer = Normalizer()
            normalizer.fit(X_train['neg'].values.reshape(1,-1)) #fitting
          4 X_train_neg_norm = normalizer.transform(X_train['neg'].values.reshape(1,-1)) #transform
          5 | X_test_neg_norm = normalizer.transform(X_test['neg'].values.reshape(1,-1))
          7
            X_train_neg_norm =X_train_neg_norm .reshape(-1,1)
            X_test_neg_norm=X_test_neg_norm.reshape(-1,1)
          8
         10 | print("After vectorizations")
         11 | print(X_train_neg_norm.shape, y_train.shape)
         12 print(X_test_neg_norm.shape, y_test.shape)
         13 print("="*40)
        After vectorizations
        (73196, 1) (73196,)
        (36052, 1) (36052,)
```

## **Encoding Numerical Features:pos**

```
In [0]:
         1 normalizer = Normalizer()
            normalizer.fit(X_train['pos'].values.reshape(1,-1)) #fitting
            X_train_pos_norm = normalizer.transform(X_train['pos'].values.reshape(1,-1)) #transform
         5
           X_test_pos_norm = normalizer.transform(X_test['pos'].values.reshape(1,-1))
         7 X_train_pos_norm =X_train_pos_norm .reshape(-1,1)
         8 X_test_pos_norm=X_test_neg_norm.reshape(-1,1)
        10 | print("After vectorizations")
            print(X_train_pos_norm.shape, y_train.shape)
            print(X_test_pos_norm.shape, y_test.shape)
        13 | print("="*40)
        After vectorizations
        (73196, 1) (73196,)
        (36052, 1)(36052,)
        _____
```

#### **Encoding Numerical Features:neu**

## **Encoding Numerical Features:compound**

```
In [0]:
         1 normalizer = Normalizer()
            normalizer.fit(X_train['compound'].values.reshape(1,-1)) #fitting
            X_train_compound_norm = normalizer.transform(X_train['compound'].values.reshape(1,-1)) #transform
           X_test_compound_norm = normalizer.transform(X_test['compound'].values.reshape(1,-1))
            X_train_compound_norm =X_train_compound_norm .reshape(-1,1)
           X_test_compound_norm=X_test_neu_norm.reshape(-1,1)
        10 | print("After vectorizations")
            print(X_train_compound_norm.shape, y_train.shape)
            print(X_test_compound_norm.shape, y_test.shape)
        13
            print("="*40)
        14
        After vectorizations
        (73196, 1) (73196,)
        (36052, 1) (36052,)
        _____
```

#### **Concatinating All The Features**

#### SET-1

• Set 1: categorical, numerical features + preprocessed\_eassay (TFIDF)

**GBDT** 4/21/2020

#### SET-2

Set 2: categorical, numerical features + preprocessed\_eassay (TFIDF W2V)

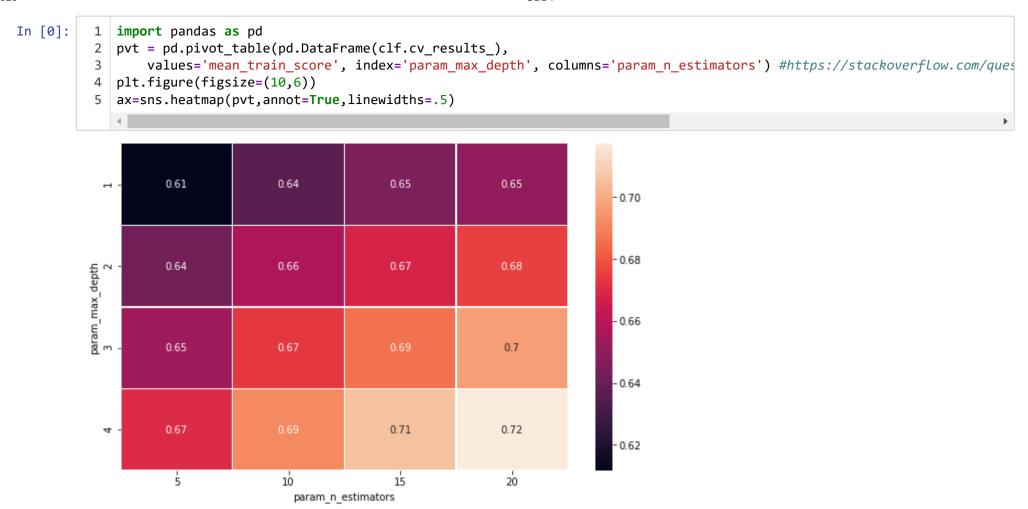
```
In [0]:
          1 from scipy.sparse import hstack
          2 X_tr_set_two = hstack((X_train_essay_tfidf_w2v, X_train_state_rc, X_train_teacher_rc, X_train_grade_rc, X_train_price
          3 X_te_set_two = hstack((X_test_essay_tfidf_w2v, X_test_state_rc, X_test_teacher_rc, X_test_grade_rc, X_test_price_norm
In [0]:
          1 print("SHAPE OF TRAIN AND TEST AFTER STACKING")
          2 print(X_tr_set_two.shape)
          3 print(X_te_set_two.shape)
         SHAPE OF TRAIN AND TEST AFTER STACKING
         (73196, 312)
         (36052, 312)
         GRADIENT BOOSTING CLASSIFIER USING GRID SEARCH CROSS VALIDATION (SET - 1)
 In [0]:
          1 from sklearn.ensemble import GradientBoostingClassifier
             from sklearn.model_selection import GridSearchCV
          3 parameters = {"max_depth":[1,2,3,4],"n_estimators":[5,10,15,20] }
          4 clf = GridSearchCV(GradientBoostingClassifier(), parameters, cv=5, scoring='roc_auc',return_train_score=True,n_jobs=
          5 clf.fit(X_tr_set_one,y_train)
Out[76]: GridSearchCV(cv=5, error_score=nan,
                      estimator=GradientBoostingClassifier(ccp_alpha=0.0,
                                                           criterion='friedman_mse',
                                                           init=None, learning_rate=0.1,
                                                           loss='deviance', max_depth=3,
                                                           max_features=None,
                                                           max_leaf_nodes=None,
                                                           min_impurity_decrease=0.0,
                                                          min_impurity_split=None,
                                                           min_samples_leaf=1,
                                                           min_samples_split=2,
                                                           min_weight_fraction_leaf=0.0,
                                                           n_estimators=100,
                                                           n_iter_no_change=None,
                                                           presort='deprecated',
                                                           random_state=None,
                                                           subsample=1.0, tol=0.0001,
                                                           validation_fraction=0.1,
                                                           verbose=0, warm_start=False),
                      iid='deprecated', n_jobs=-1,
                      param_grid={'max_depth': [1, 2, 3, 4],
                                  'n_estimators': [5, 10, 15, 20]},
                      pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
                      scoring='roc_auc', verbose=0)
In [0]:
          1 | train_auc= clf.cv_results_['mean_train_score']
            train_auc_std= clf.cv_results_['std_train_score']
          3 | cv_auc = clf.cv_results_['mean_test_score']
          4 cv_auc_std= clf.cv_results_['std_test_score']
In [0]:
          1 print('Best score: ',clf.best_score_)
```

```
2 print('Best Hyper parameters: ',clf.best_params_)
```

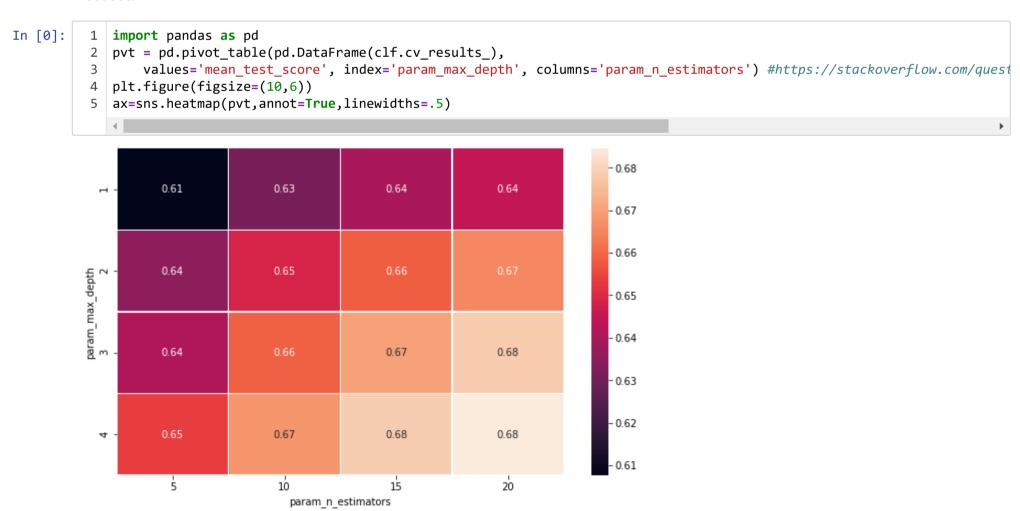
```
Best score: 0.684567328741456
Best Hyper parameters: {'max_depth': 4, 'n_estimators': 20}
```

**Roc Plot Of Train And Test Data** 

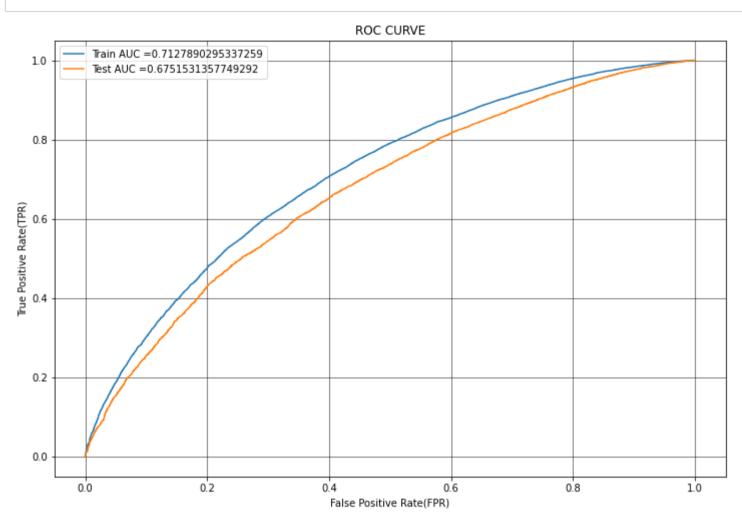
Train data



#### Test data



Roc Plot Of Train And Test Data



## **Confusion Matrix**

```
In [0]:
          1
             def find_best_threshold(threshould, fpr, tpr):
                 t = threshould[np.argmax(tpr*(1-fpr))]
          3
                 # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
          4
                 print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
          5
                 return t
          6
          7
             def predict_with_best_t(proba, threshould):
                 predictions = []
          8
                 for i in proba:
          9
                     if i>=threshould:
         10
                         predictions.append(1)
         11
         12
                     else:
         13
                         predictions.append(0)
         14
                 return predictions
```

#### **Train Data**

```
In [0]:

best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
cm=metrics.confusion_matrix(y_train,predict_with_best_t(y_train_probs, best_t)) # https://stackoverflow.com/question

print("CONFUSION MATRIX OF TRAIN DATA")
print("\n")
print(cm)
sns.heatmap(cm, annot=True, fmt='d',cmap='GnBu',annot_kws = {"size":16})
plt.ylabel('True Label',size=12)
plt.xlabel('Predicted Label',size=12)
plt.title('Train Data Confusion Matrix',size=12)
```

the maximum value of tpr\*(1-fpr) 0.4277456121087118 for threshold 0.846 CONFUSION MATRIX OF TRAIN DATA

```
[[ 7046 4037]
[20322 41791]]
```

Out[112]: Text(0.5, 1.0, 'Train Data Confusion Matrix')



#### **Test Data**

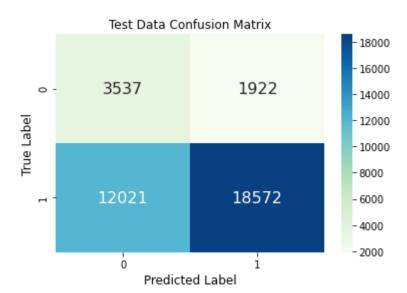
```
In [0]: 1 best_t = find_best_threshold(te_thresholds, test_fpr, test_tpr)
cm=metrics.confusion_matrix(y_test,predict_with_best_t(y_test_probs, best_t))

4 print("CONFUSION MATRIX OF TEST DATA")
5 print('\n')
6 print(cm)
7 sns.heatmap(cm, annot=True, fmt='d',cmap='GnBu',annot_kws = {"size":16})
8 plt.ylabel('True Label',size=12)
9 plt.xlabel('Predicted Label',size=12)
10 plt.title('Test Data Confusion Matrix',size=12)
```

the maximum value of tpr\*(1-fpr) 0.39333136004500213 for threshold 0.85 CONFUSION MATRIX OF TEST DATA

```
[[ 3537 1922]
[12021 18572]]
```

Out[113]: Text(0.5, 1.0, 'Test Data Confusion Matrix')



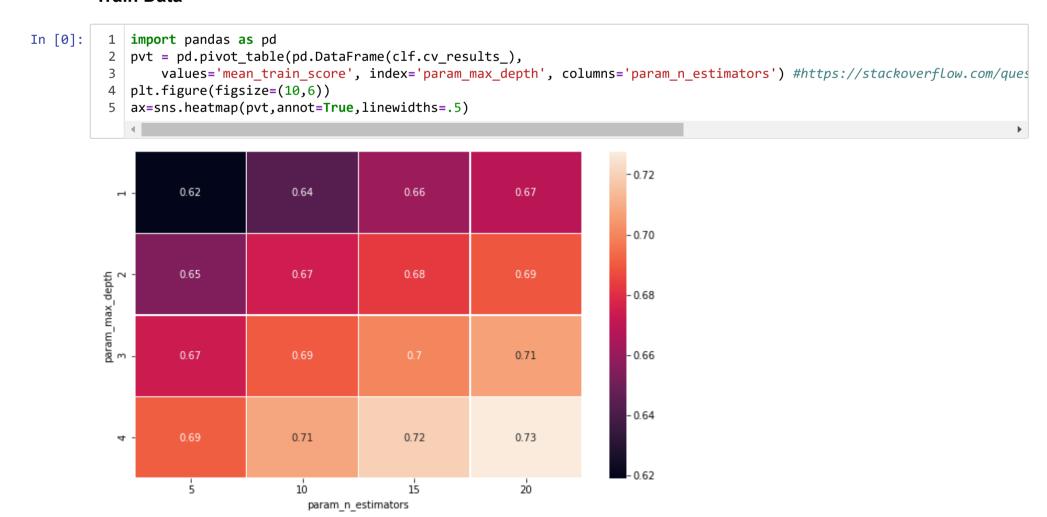
## **GRADIENT BOOSTING CLASSIFIER USING GRID SEARCH CROSS VALIDATION (SET - 2)**

```
1 from sklearn.ensemble import GradientBoostingClassifier
 In [0]:
            2 from sklearn.model_selection import GridSearchCV
            3 | parameters = {"max_depth":[1,2,3,4],"n_estimators":[5,10,15,20] }
            4 clf = GridSearchCV(GradientBoostingClassifier(), parameters, cv=5, scoring='roc_auc',return_train_score=True,n_jobs=
            5 clf.fit(X_tr_set_two,y_train)
Out[114]: GridSearchCV(cv=5, error_score=nan,
                       estimator=GradientBoostingClassifier(ccp_alpha=0.0,
                                                             criterion='friedman_mse',
                                                             init=None, learning_rate=0.1,
                                                             loss='deviance', max_depth=3,
                                                             max_features=None,
                                                             max_leaf_nodes=None,
                                                             min_impurity_decrease=0.0,
                                                             min_impurity_split=None,
                                                             min_samples_leaf=1,
                                                             min_samples_split=2,
                                                             min_weight_fraction_leaf=0.0,
                                                             n_estimators=100,
                                                             n_iter_no_change=None,
                                                             presort='deprecated',
                                                             random_state=None,
                                                             subsample=1.0, tol=0.0001,
                                                             validation_fraction=0.1,
                                                             verbose=0, warm_start=False),
                       iid='deprecated', n_jobs=-1,
                       param_grid={'max_depth': [1, 2, 3, 4],
                                    'n_estimators': [5, 10, 15, 20]},
                       pre_dispatch='2*n_jobs', refit=True, return_train_score=True,
                       scoring='roc_auc', verbose=0)
 In [0]:
            1 train_auc= clf.cv_results_['mean_train_score']
            2 train_auc_std= clf.cv_results_['std_train_score']
            3 cv_auc = clf.cv_results_['mean_test_score']
              cv_auc_std= clf.cv_results_['std_test_score']
 In [0]:
            1 print('Best score: ',clf.best_score_)
              print('Best Hyper parameters: ',clf.best params )
          Best score: 0.6902930550494676
```

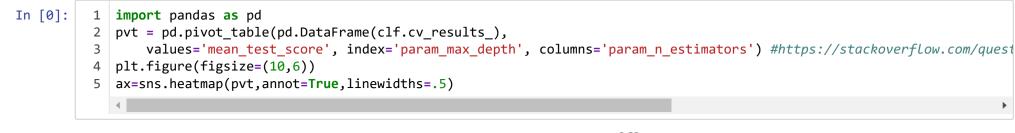
# Plotting Hyperparameter v/s Auc

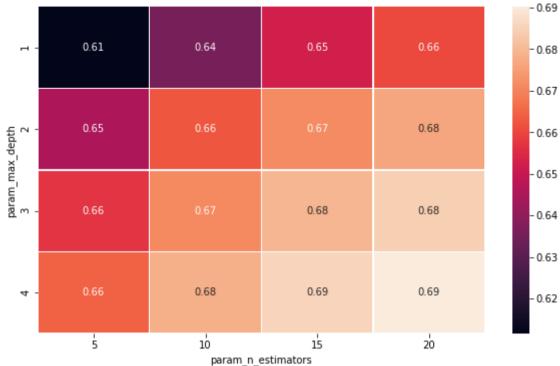
Best Hyper parameters: {'max\_depth': 4, 'n\_estimators': 20}

## **Train Data**



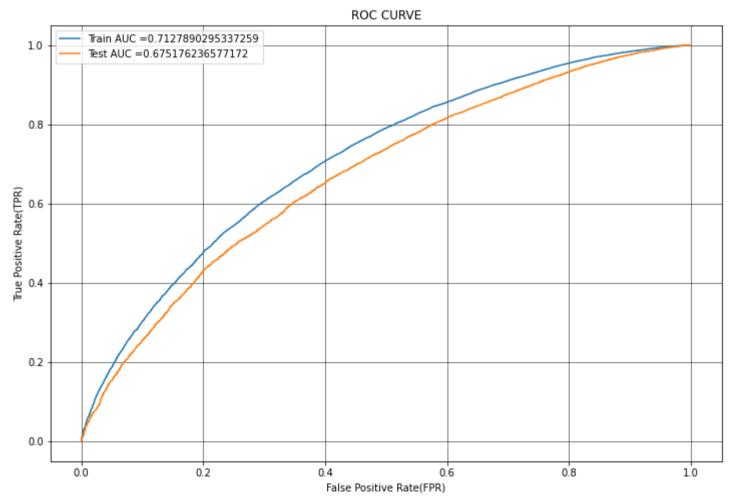
#### **Cross Validation data**





#### Roc Plot Of Train And Test Data

```
In [0]:
             model_set2=GradientBoostingClassifier(max_depth = clf.best_params_["max_depth"], n_estimators= clf.best_params_["n_estimators= clf.best_params_"]
             model_set2.fit(X_tr_set_one,y_train)
          3 | y_train_probs = model_set2.predict_proba(X_tr_set_one)[:,1] # converting train and test output into probability
          4 y_test_probs= model_set2.predict_proba(X_te_set_one )[:,1]
          6 train_fpr, train_tpr, tr_thresholds = roc_curve(y_train, y_train_probs) # storing values of fpr and tpr
             test_fpr, test_tpr, te_thresholds = roc_curve(y_test, y_test_probs)
          8
          9
             plt.figure(figsize=(12,8))
         10 plt.plot(train_fpr, train_tpr, label="Train AUC ="+str(auc(train_fpr, train_tpr)))
         11 plt.plot(test_fpr, test_tpr, label="Test AUC ="+str(auc(test_fpr, test_tpr)))
         12 plt.legend()
         plt.xlabel("False Positive Rate(FPR)")
             plt.ylabel("True Positive Rate(TPR)")
             plt.title("ROC CURVE")
            plt.grid(color='black',lw=0.5)
```



## **Confusion Matrix**

```
In [0]:
             def find_best_threshold(threshould, fpr, tpr):
                 t = threshould[np.argmax(tpr*(1-fpr))]
                 # (tpr*(1-fpr)) will be maximum if your fpr is very low and tpr is very high
          3
                 print("the maximum value of tpr*(1-fpr)", max(tpr*(1-fpr)), "for threshold", np.round(t,3))
          4
          5
                 return t
          6
             def predict_with_best_t(proba, threshould):
          7
          8
                 predictions = []
          9
                 for i in proba:
         10
                     if i>=threshould:
         11
                         predictions.append(1)
         12
                     else:
         13
                         predictions.append(0)
         14
                 return predictions
```

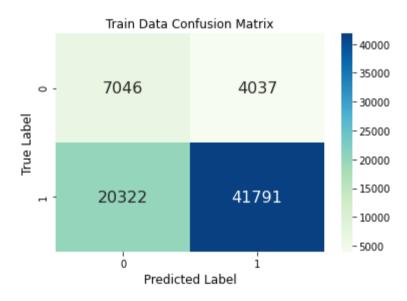
#### **Train Data**

```
In [0]: 1 best_t = find_best_threshold(tr_thresholds, train_fpr, train_tpr)
2 cm=metrics.confusion_matrix(y_train,predict_with_best_t(y_train_probs, best_t)) # https://stackoverflow.com/question
3
4 print("CONFUSION MATRIX OF TRAIN DATA")
5 print("\n")
6 print(cm)
7 sns.heatmap(cm, annot=True, fmt='d',cmap='GnBu',annot_kws = {"size":16})
8 plt.ylabel('True Label',size=12)
9 plt.xlabel('Predicted Label',size=12)
10 plt.title('Train Data Confusion Matrix',size=12)
```

the maximum value of tpr\*(1-fpr) 0.4277456121087118 for threshold 0.846 CONFUSION MATRIX OF TRAIN DATA

```
[[ 7046 4037]
[20322 41791]]
```

Out[121]: Text(0.5, 1.0, 'Train Data Confusion Matrix')

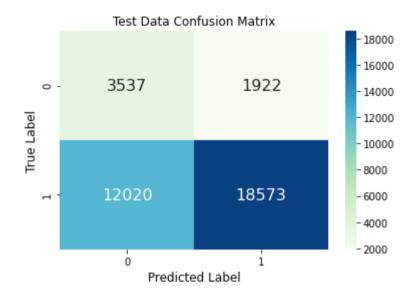


**Test Data** 

the maximum value of tpr\*(1-fpr) 0.3933525387742744 for threshold 0.85 CONFUSION MATRIX OF TEST DATA

```
[[ 3537 1922]
[12020 18573]]
```

Out[122]: Text(0.5, 1.0, 'Test Data Confusion Matrix')



# **Summary**

S1.N0	Vectorizer	·		++   Test-AUC   ++
1	TFIDF	GRADIENT BOOSTING CLASSIFIER	max_depth =4 , n_estimators=20	0.67515
		·	max_depth =4 , n_estimators=20	

```
In [0]: 1
```