AmazonReviews_RF_GBDT

September 2, 2018

1 Amazon Reviews Classification Using Random Forest and Gradient-Boosting

```
In [10]: %matplotlib inline
         import pandas as pd #for data frames
         import numpy as np #numpy array operations
         import string
         import matplotlib.pyplot as plt
         from sklearn.feature_extraction.text import TfidfTransformer
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.feature_extraction.text import CountVectorizer
         import pickle
         from sklearn.cross_validation import train_test_split
         import seaborn as sn
         from sklearn.metrics import accuracy_score
         from sklearn.cross_validation import cross_val_score
         from sklearn import cross_validation
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import average_precision_score,f1_score,precision_score,recall_s
         from sklearn.grid_search import GridSearchCV
         from sklearn.model_selection import RandomizedSearchCV
         from sklearn.metrics import roc_curve, auc
         from scipy.stats import expon
         import random
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.ensemble import GradientBoostingClassifier
         from xgboost import XGBClassifier
In [2]: pickle_in=open("cleanedData.pickle","rb")
        final = pickle.load(pickle_in)
        import pickle
        pickle_in = open("BOW_tfidf_avgW2V_Train_test_data.pickle","rb")
```

```
count_vect = pickle.load(pickle_in) #BOW
        final_counts_train = pickle.load(pickle_in) #BOW
        final_counts_test = pickle.load(pickle_in) #BOW
        tf_idf_vect = pickle.load(pickle_in) #tfidf
        final_tf_idf_train = pickle.load(pickle_in) #tfidf
        final_tf_idf_test = pickle.load(pickle_in) #tfidf
        features = pickle.load(pickle_in)
        sent_vectors_train = pickle.load(pickle_in) #avgW2v Vectors
        sent_vectors_test = pickle.load(pickle_in) #avgW2v Vectors
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\base.py:315: UserWarning: Trying to unpickle
  UserWarning)
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\base.py:315: UserWarning: Trying to unpickle
  UserWarning)
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\base.py:315: UserWarning: Trying to unpickle
  UserWarning)
In [3]: train_data = final.head(int(0.80*final.shape[0]))
        test_data = final.head(int(0.20*final.shape[0])+1)
        scores = final['Score'].get_values()
        len(scores)
Out[3]: 364171
In [28]: def convScores(scores):
             li = lambda x: 1 if x=='positive' else 0
             final_scores = []
             for i in range(0,len(scores)):
                 final_scores.append(li(scores[i]))
             return final_scores
         def convToNpArray(arr):
             if(type(arr) == list):
                 arr = np.array(arr)
                 return arr
             else:
                 return arr;
         def confusionMatrix(y_test,pred):
             df_cm = pd.DataFrame(confusion_matrix(y_test, pred), index = ['False','True'],
                               columns = ['False','True'])
             sn.heatmap(df_cm, annot=True)
             plt.title('Confusion Matrix')
             plt.ylabel('Actual')
             plt.xlabel('Predicted')
             plt.show()
         def auc_roc(y_test,pred):
             fpr, tpr, thresholds = roc_curve(y_test,pred)
```

```
acc = auc(fpr, tpr)
   print("Area Under The Curve is : ",acc)
   plt.figure()
   plt.plot(fpr, tpr, color='darkorange',
              label='ROC curve (area/auc = %0.2f)' % acc)
   plt.xlabel('False Positive Rate')
   plt.ylabel('True Positive Rate')
   plt.title('Receiver operating characteristic Curve')
   plt.legend(loc="lower right")
   plt.show()
def show_most_informative_features(vectorizer, clf, n=20):
   feature_names = vectorizer.get_feature_names()
    coefs_with_fns = sorted(zip(clf.feature_importances_, feature_names))
   top = zip(coefs_with_fns[:n], coefs_with_fns[:-(n + 1):-1])
   print("These are the top {} important Features Which are most widely used in Posi
   print("")
   print("\tPositive: \t\tNegative:")
   for (coef_1, fn_1), (coef_2, fn_2) in top:
       print("")
       print("\t%-15s\t\t\t\-15s" % (fn_2,fn_1))
```

2 RF

2.1 Avg W2V RF

```
In [29]: x_1 = sent_vectors_train[0:50000]

# this is only Score/rating of data

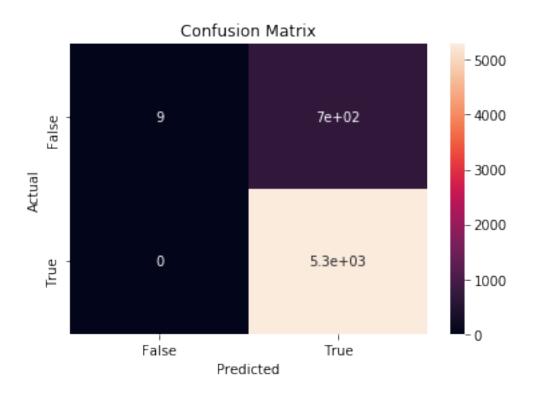
y_1 = convScores(train_data['Score'].get_values())[0:50000]

x_test = sent_vectors_test[0:6000]
 y_test = convScores(test_data['Score'].get_values())[0:6000]

x_1 = convToNpArray(x_1)
 x_test = convToNpArray(x_test)
 y_1 = convToNpArray(y_1)
 y_test = convToNpArray(y_test)
```

Considering n_estimators/k and max_depth as hyperparameters

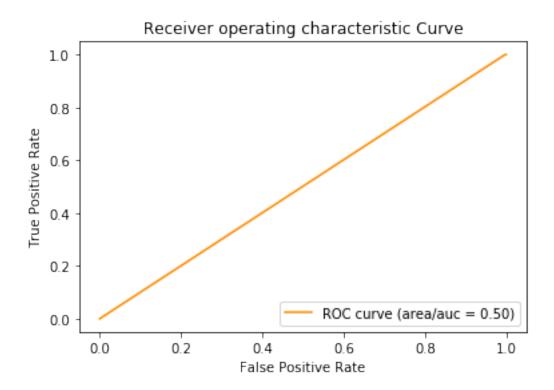
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: 'ChangedBehaviorWarning)



AUC

In [16]: auc_roc(y_test,pred)

Area Under The Curve is: 0.50131996277



2.2 BOW RF

```
In [33]: # Total data frame

x_1 = final_counts_train[0:50000]

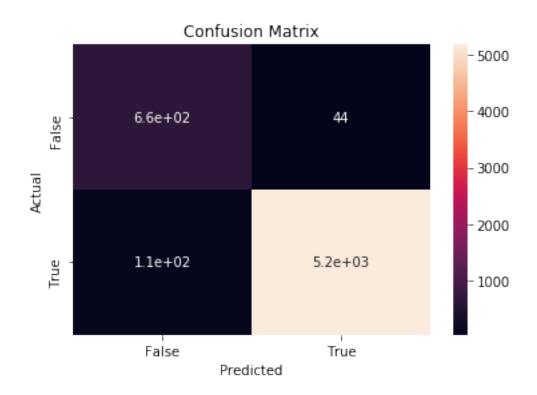
# this is only Score/rating of data

y_1 = convScores(train_data['Score'].get_values())[0:50000]

x_test = final_counts_test[0:6000]
 y_test = convScores(test_data['Score'].get_values())[0:6000]

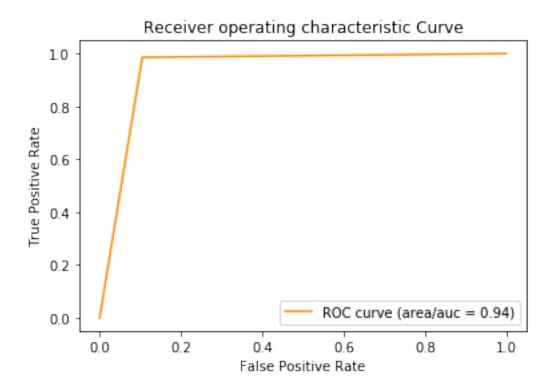
#x_1, x_test, y_1, y_test = train_test_split(x,y, test_size=0.3, random_state=0)
```

```
x_1 = convToNpArray(x_1)
         x_test = convToNpArray(x_test)
         y_1 = convToNpArray(y_1)
         y_test = convToNpArray(y_test)
In [34]: tuned_parameters = {'n_estimators':[10,20,40,50],
                            'max_depth': [20,50,100]}
         rf_model = RandomForestClassifier(n_jobs=-1,oob_score=True,class_weight='balanced')
         model = GridSearchCV(rf_model,tuned_parameters,
                              scoring='accuracy',cv=3,n_jobs=-1)
         model.fit(x_1,y_1)
         print(model.best_estimator_)
         print("Score: ",model.score(x_test,y_test))
RandomForestClassifier(bootstrap=True, class_weight='balanced',
            criterion='gini', max_depth=50, max_features='auto',
            max_leaf_nodes=None, min_impurity_split=1e-07,
           min_samples_leaf=1, min_samples_split=2,
           min_weight_fraction_leaf=0.0, n_estimators=50, n_jobs=-1,
            oob_score=True, random_state=None, verbose=0, warm_start=False)
Score: 0.974
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: '
  ChangedBehaviorWarning)
In [35]: best_rf_model = model.best_estimator_
         best_rf_model.fit(x_1,y_1)
         pred = best_rf_model.predict(x_test)
         confusionMatrix(y_test,pred)
```



AUC

In [21]: auc_roc(y_test,pred)



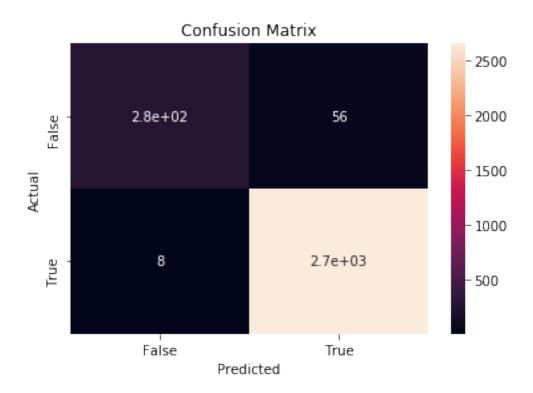
In [37]: show_most_informative_features(count_vect,best_rf_model,10)

These are the top 10 important Features Which are most widely used in Positive and Negative Re

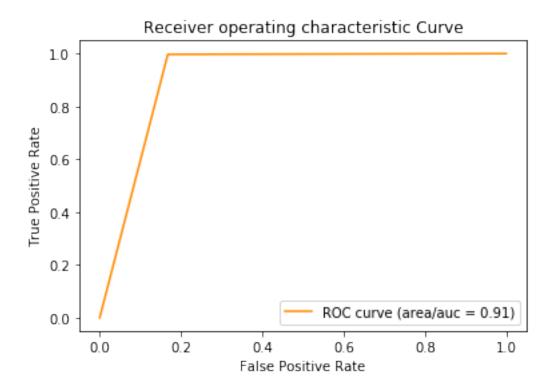
Positive:	Negative:
great	aa
love	aaa
disappoint	aaaa
best	aaaaa
delici	aaaaaaaaaaaaaaa
would	aaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaa
perfect	aaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaa
wast	aaaaaaaagghh
favorit	aaaaaahhhhhh

2.3 TFIdf

```
In [22]: x_1 = final_tf_idf_train[0:10000]
         y_1 = convScores(train_data['Score'].get_values())[0:10000]
         x_test = final_tf_idf_test[0:3000]
         y_test = convScores(test_data['Score'].get_values())[0:3000]
         x_1 = convToNpArray(x_1)
         x_test = convToNpArray(x_test)
         y_1 = convToNpArray(y_1)
         y_test = convToNpArray(y_test)
In [23]: tuned_parameters = {'n_estimators':[10,20,40,50],
                            'max_depth': [20,50,100]}
         rf_model = RandomForestClassifier(n_jobs=-1,oob_score=True,class_weight='balanced')
         model = GridSearchCV(rf_model,tuned_parameters,
                              scoring='accuracy',cv=3,n_jobs=-1)
         model.fit(x_1,y_1)
         print(model.best_estimator_)
         print("Score: ",model.score(x_test,y_test))
RandomForestClassifier(bootstrap=True, class_weight='balanced',
            criterion='gini', max_depth=50, max_features='auto',
            max_leaf_nodes=None, min_impurity_split=1e-07,
            min_samples_leaf=1, min_samples_split=2,
            min_weight_fraction_leaf=0.0, n_estimators=50, n_jobs=-1,
            oob_score=True, random_state=None, verbose=0, warm_start=False)
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: '
  ChangedBehaviorWarning)
Score: 0.982
In [24]: best_rf_model = model.best_estimator_
         best_rf_model.fit(x_1,y_1)
         pred = best_rf_model.predict(x_test)
         confusionMatrix(y_test,pred)
```



In [25]: auc_roc(y_test,pred)



2.4 Important Features

In [27]: show_most_informative_features(tf_idf_vect,best_rf_model,10)

These are the top 10important Features Which are most widely used in Positive and Negative Rev.

Positive:	Negative:
money	aa
order	aa pleas
return	aa state
great	aaa
horribl	aaa aaa
away	aaa condit
snack	aaa dont
love	aaa hockey

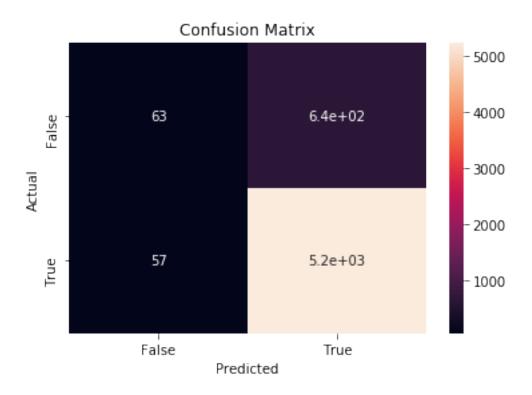
high recommend aaa job
bland aaa magazin

3 GBDT

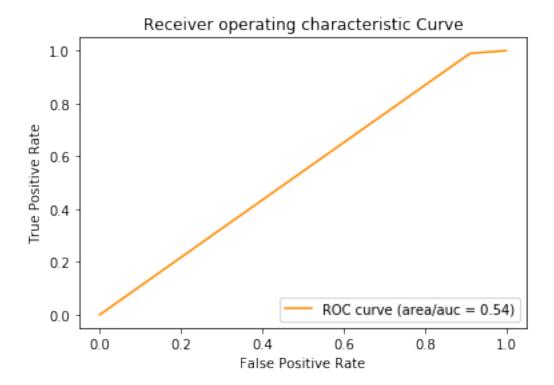
3.1 **Avg W2V**

```
In [48]: x_1 = sent_vectors_train[0:30000]
         # this is only Score/rating of data
         y_1 = convScores(train_data['Score'].get_values())[0:30000]
         x_test = sent_vectors_test[0:6000]
         y_test = convScores(test_data['Score'].get_values())[0:6000]
         x_1 = convToNpArray(x_1)
         x_test = convToNpArray(x_test)
         y_1 = convToNpArray(y_1)
         y_test = convToNpArray(y_test)
In [49]: tuned_parameters = {'n_estimators':[50,75,100,125],
                            'learning_rate':[0.01,0.1,0.5],
                            'max_depth': [3,5,7]}
         #using 70% of data to train the base models (Row Sampling)
         gb_model = GradientBoostingClassifier(subsample=0.7)
         model = GridSearchCV(gb_model,tuned_parameters,
                              scoring='accuracy',cv=3,n_jobs=-1)
         model.fit(x_1,y_1)
         print(model.best_estimator_)
         print("Score: ",model.score(x_test,y_test))
GradientBoostingClassifier(criterion='friedman_mse', init=None,
              learning_rate=0.1, loss='deviance', max_depth=7,
              max_features=None, max_leaf_nodes=None,
              min_impurity_split=1e-07, min_samples_leaf=1,
              min_samples_split=2, min_weight_fraction_leaf=0.0,
              n_estimators=100, presort='auto', random_state=None,
              subsample=0.7, verbose=0, warm_start=False)
Score: 0.887666666667
```

C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: 'ChangedBehaviorWarning)



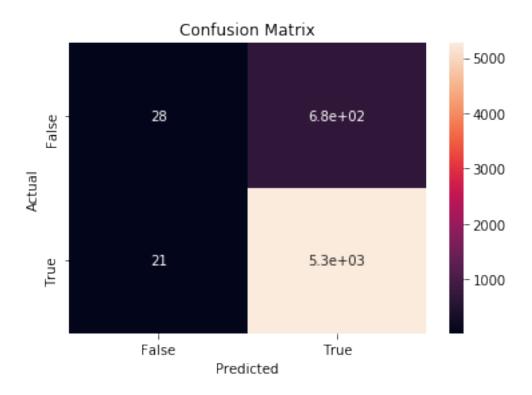
In [51]: auc_roc(y_test,pred)



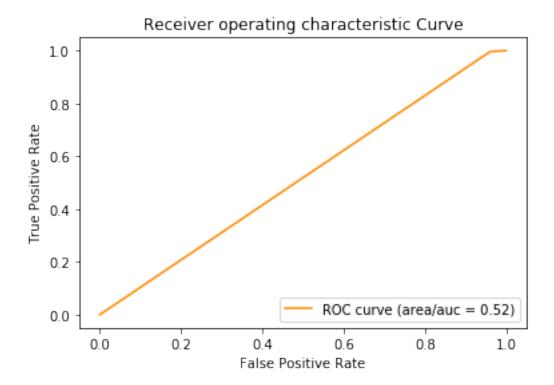
3.1.1 Using XGBClassifier

```
In [54]: tuned parameters = {'n estimators':[75,100,120],
                            'learning_rate':[0.01,0.1,0.5]}
         #using 70% of data to train the base models (Row Sampling)
         gb_model = XGBClassifier(subsample=0.7,n_jobs=4)
         model = GridSearchCV(gb_model,tuned_parameters,
                              scoring='accuracy',cv=3,n_jobs=-1)
         model.fit(x_1,y_1)
         print(model.best_estimator_)
         print("Score: ",model.score(x_test,y_test))
XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
       colsample_bytree=1, gamma=0, learning_rate=0.1, max_delta_step=0,
       max_depth=3, min_child_weight=1, missing=None, n_estimators=120,
       n_jobs=4, nthread=None, objective='binary:logistic', random_state=0,
       reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
       silent=True, subsample=0.7)
Score: 0.883333333333
```

C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: 'ChangedBehaviorWarning)

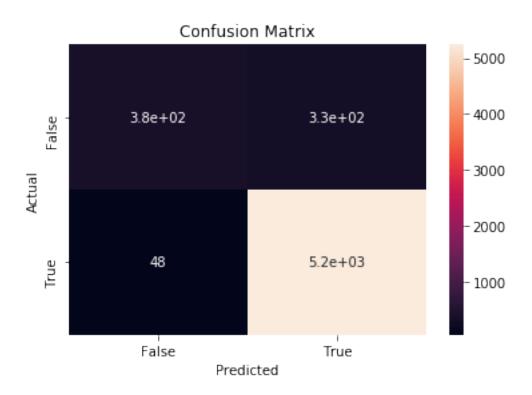


In [56]: auc_roc(y_test,pred)

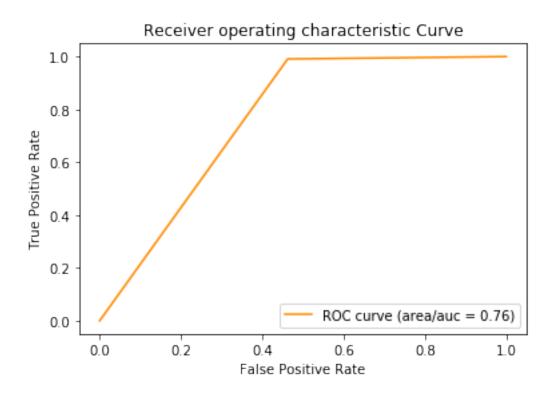


4 BOW

```
#using 70% of data to train the base models (Row Sampling)
         gb_model = XGBClassifier(subsample=0.7,n_jobs=4)
         model = GridSearchCV(gb_model,tuned_parameters,
                              scoring='accuracy',cv=3,n_jobs=-1)
         model.fit(x_1,y_1)
         print(model.best_estimator_)
         print("Score: ",model.score(x_test,y_test))
XGBClassifier(base_score=0.5, booster='gbtree', colsample_bylevel=1,
       colsample_bytree=1, gamma=0, learning_rate=0.5, max_delta_step=0,
      max_depth=3, min_child_weight=1, missing=None, n_estimators=120,
       n_jobs=4, nthread=None, objective='binary:logistic', random_state=0,
       reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
       silent=True, subsample=0.7)
Score: 0.9375
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: '
  ChangedBehaviorWarning)
In [40]: best_gb_model = model.best_estimator_
         best_gb_model.fit(x_1,y_1)
         pred = best_gb_model.predict(x_test)
         confusionMatrix(y_test,pred)
```



In [41]: auc_roc(y_test,pred)



4.1 Important Features

In [42]: show_most_informative_features(count_vect,best_gb_model,10)

These are the top 10 important Features Which are most widely used in Positive and Negative Re

Positive:	Negati	ive:
like	8	aa
great	č	aaa
tast	8	aaaa
product	8	aaaaa
love	8	aaaaaaaaaaaaa
best	8	aaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaa
tri	8	aaaaaaaaaaaaaaaaangh
littl	ā	aaaaaaaagghh

```
good aaaaaaahhhhhh
```

y_1 = convScores(train_data['Score'].get_values())[0:10000]

flavor

In [43]: x_1 = final_tf_idf_train[0:10000]

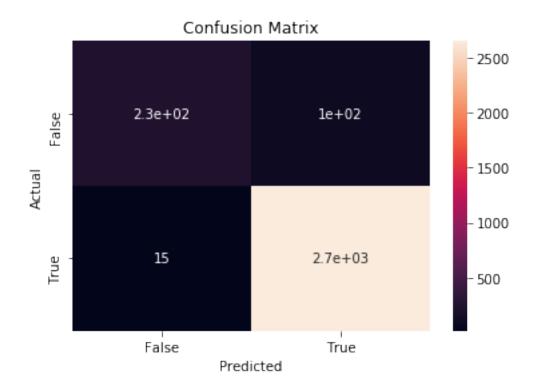
aaaaaarrrrrggghhh

4.2 TFIdf

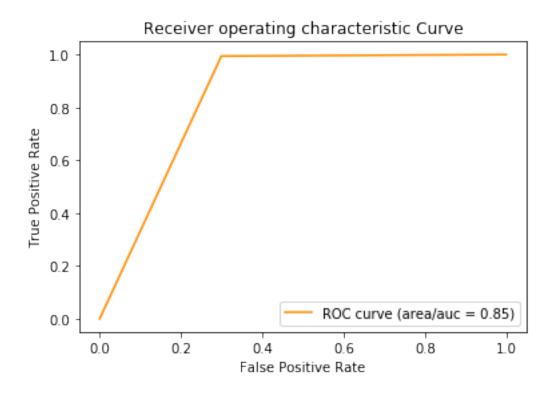
```
x_test = final_tf_idf_test[0:3000]
         y_test = convScores(test_data['Score'].get_values())[0:3000]
        x_1 = convToNpArray(x_1)
         x_test = convToNpArray(x_test)
         y_1 = convToNpArray(y_1)
         y_test = convToNpArray(y_test)
In [44]: tuned_parameters = {'n_estimators':[75,100,120],
                            'learning_rate':[0.01,0.1,0.5]}
         #using 70% of data to train the base models (Row Sampling)
         gb_model = XGBClassifier(subsample=0.7,n_jobs=4)
         model = GridSearchCV(gb_model,tuned_parameters,
                              scoring='accuracy',cv=3,n_jobs=-1)
         model.fit(x_1,y_1)
         print(model.best_estimator_)
         print("Score: ",model.score(x_test,y_test))
XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1,
       colsample_bytree=1, gamma=0, learning_rate=0.5, max_delta_step=0,
       max_depth=3, min_child_weight=1, missing=None, n_estimators=120,
      n_jobs=4, nthread=None, objective='binary:logistic', random_state=0,
       reg_alpha=0, reg_lambda=1, scale_pos_weight=1, seed=None,
       silent=True, subsample=0.7)
C:\Users\Dell\Anaconda3\lib\site-packages\sklearn\grid_search.py:438: ChangedBehaviorWarning: '
  ChangedBehaviorWarning)
Score: 0.961666666667
In [45]: best_gb_model = model.best_estimator_
```

best_gb_model.fit(x_1,y_1)

pred = best_gb_model.predict(x_test)
confusionMatrix(y_test,pred)



In [46]: auc_roc(y_test,pred)



4.3 Important Features

In [47]: show_most_informative_features(tf_idf_vect,best_gb_model,10)

These are the top 10 important Features Which are most widely used in Positive and Negative Re

Positive:	Negative:
good	aa
like	aa pleas
one	aa state
tast	aaa
tri	aaa aaa
product	aaa condit
get	aaa dont
use	aaa hockey

love aaa job

aaa magazin

Summary:

great

```
Classifier
  Vectorizer
  Hyper Parameters
  Confusion Matrix
  Accuracy
  AUC
Random Forest
  Avg W2v
  n_estimators: 40<br>
    max_depth: 20
  TPR :0.999622
               TNR : 0.004243<br>
               FNR: 0.000378
    FPR: 0.995757
  88%<br>
  0.5
Random Forest
  BOW
  n_estimators: 40<br>
    max_depth: 50
  TPR :0.983374
               TNR : 0.903819<br>
    FPR : 0.096181
               FNR: 0.016626
  97.5%
  0.94
Random Forest
  TfIdf
  n_estimators: 50<br>
```

```
max_depth: 50
   TPR :0.999625
                  TNR : 0.847305<br>
     FPR: 0.152695 FNR: 0.000375
  97.9%
   0.91
Gradient Boost / XGBClassifier
   Avg W2v
   n_estimators: 100<br>
      learning_rate: 0.1<br>
     max_depth: 7<br>
      <b>Using XGBClassifier</b><br>
     n_estimators: 120<br>
      learning_rate: 0.1
  TPR :0.983374
                  TNR : 0.128713<br>
     FPR: 0.871287 FNR: 0.016626
  88.33%
   0.54
Gradient Boost / XGBClassifier
   BOW
   n_estimators: 120<br>
      learning_rate: 0.5
  TPR :0.990931
                  TNR : 0.537482<br>
     FPR: 0.462518 FNR: 0.009069
  93.75%
  0.76
Gradient Boost / XGBClassifier
   TfIdf
   n_estimators: 120<br>
      learning_rate: 0.5<br>
   TPR :0.994374
                  TNR : 0.700599<br>
     FPR: 0.299401 FNR: 0.005626
```

-> With Both the classifiers BOW and Tfidf have given Good Results in terms of TPR TNR and Accuracy. -> Even Interms of AUC also BOW and Tfidf Have performed well. -> Interms of AUC Random Forest has Performed well when compared to Gradient Boost.

Note: -> Didnt use Sklearn's GradientBoost Classifier for Tfidf and BOW because fit() method is expecting dense array, as there is large amount of data with more no.of dimensions it requires more Memory, So used XGBClassifier().

-> XGBClassifier took very Less time to execute when compared to GradientBoostingClassifier() because XGBClassifier supports Parallel Execution.