SVM on Amazon Food Reviews

May 28, 2018

0.0.1 SVM on Amazon food reviews

```
In [6]: #importing required Modules
        %matplotlib inline
        import sqlite3
        import pandas as pd
        import numpy as np
        import nltk
        import string
        import pickle
        import matplotlib.pyplot as plt
        import seaborn as sns
        from sklearn.feature_extraction.text import TfidfTransformer
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.feature_extraction.text import CountVectorizer
        from sklearn.metrics import confusion_matrix
        from sklearn import metrics
        from sklearn.metrics import accuracy_score
        from sklearn.metrics import roc_curve, auc
        from nltk.stem.porter import PorterStemmer
        from sklearn.preprocessing import StandardScaler
        from sklearn.model_selection import TimeSeriesSplit
        from sklearn.metrics import precision_score
        from sklearn.metrics import recall_score
        from sklearn.metrics import confusion_matrix
        from sklearn.model_selection import GridSearchCV
        from sklearn.model_selection import RandomizedSearchCV
        from sklearn.model_selection import ParameterGrid
        from sklearn.svm import SVC
        from sklearn.linear_model import SGDClassifier
In [7]: import warnings
        warnings.filterwarnings('ignore')
In [8]: def cleanpunc(sentence):
            function to clean the word of any punctuation or special characters
```

```
cleaned = re.sub(r'[?|!||'|#]',r'',sentence)
            cleaned = re.sub(r'[.|,|)|(||/|,r'',cleaned)
            return cleaned
        def cleanhtml(sentence):
            function to clean the word of any html-tags
            cleanr = re.compile('<.*?>')
            cleantext = re.sub(cleanr, ' ', sentence)
            return cleantext
        def reduce_lengthening(text):
            pattern = re.compile(r''(.)\1{2,}")
            return pattern.sub(r"\1\1", text)
In [9]: #getting stop words
        from nltk.corpus import stopwords
        stop = set(stopwords.words('english'))
        stop.remove('not')
        stop.remove('very')
        #from autocorrect import spell
In [10]: conn = sqlite3.connect('final_clean_LR.sqlite')
         final_review = pd.read_sql_query("""
         SELECT *
         FROM Reviews_final
         """, conn)
In [11]: s = final_review.sample(n=30000,random_state=0)
In [12]: #SORT by time for TBS
         s = s.sort_values(by='Time')
In [13]: s.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 30000 entries, 431 to 339792
Data columns (total 15 columns):
level_0
                          30000 non-null int64
index
                          30000 non-null int64
Ιd
                          30000 non-null int64
ProductId
                          30000 non-null object
UserId
                          30000 non-null object
ProfileName
                          30000 non-null object
HelpfulnessNumerator
                          30000 non-null int64
HelpfulnessDenominator
                          30000 non-null int64
Score
                          30000 non-null object
Time
                          30000 non-null int64
                          30000 non-null object
Summary
Text
                          30000 non-null object
```

```
CleanedTextBow
                          30000 non-null object
final_text
                          30000 non-null object
final_stem_text
                          30000 non-null object
dtypes: int64(6), object(9)
memory usage: 3.7+ MB
In [14]: #changing lables to 1 or 0
         s.Score = final_review.Score.apply(lambda x:
                              1 if x == 'positive' else 0)
In [15]: #Converting to int8
         s.HelpfulnessNumerator = s.HelpfulnessNumerator.astype(np.int8)
         s.HelpfulnessDenominator = s.HelpfulnessDenominator.astype(np.int8)
In [16]: #Splitting Dataframe for train and test
         train_df = s.iloc[:round(s.shape[0]*0.70),:]
         test df = s.iloc[round(s.shape[0]*0.70):,:]
In [17]: train df.to csv('train df svm.csv',index=False)
         test_df.to_csv('test_df_svm.csv',index=False)
In [18]: print(train_df.shape)
        print(test_df.shape)
(21000, 15)
(9000, 15)
0.0.2 Bag of Words:
In [25]: #BoW with cleaned data and without stopwords
         #simple cv for train data
         scores_train = []
         from nltk.corpus import stopwords
         stop = set(stopwords.words('english'))
         stop.remove('not')
         stop.remove('very')
         #CountVectorizer for BoW
         count_vect = CountVectorizer(stop_words=list(stop),dtype=np.int8)
         X_train = train_df.iloc[:round(train_df.shape[0]*0.70),:]
         X_test_cv = train_df.iloc[round(train_df.shape[0]*0.70):,:]
         final_counts_train = count_vect.fit_transform(
                 X_train['final_text'].values)
         #test
         X_test = count_vect.transform(X_test_cv['final_text'].values)
         scale =StandardScaler(with_mean=False)
         X_train_scale = scale.fit_transform(final_counts_train)
         X_test = scale.transform(X_test)
```

```
In [16]: for i in ParameterGrid({'C':[0.001, 0.01, 0.1, 1, 10],
                             'gamma':[0.01, 0.001, 0.1, 1]}):
             model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
             model.fit(X_train_scale,X_train.Score)
             train score = model.score(X train scale, X train.Score)
             test_score = model.score(X_test, X_test_cv.Score)
             print('C',i['C'],'Gamma',i['gamma'],'Train Score',train_score,
                              'Test Score', test score)
C 0.001 Gamma 0.01 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.001 Gamma 0.001 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.001 Gamma 0.1 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.001 Gamma 1 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.01 Gamma 0.01 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.01 Gamma 0.001 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.01 Gamma 0.1 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.01 Gamma 1 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.1 Gamma 0.01 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.1 Gamma 0.001 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.1 Gamma 0.1 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 0.1 Gamma 1 Train Score 0.8605306122448979 Test Score 0.8275238095238096
C 1 Gamma 0.01 Train Score 0.9997959183673469 Test Score 0.8253968253968254
C 1 Gamma 0.001 Train Score 0.9951700680272109 Test Score 0.8274603174603175
C 1 Gamma 0.1 Train Score 1.0 Test Score 0.8253968253968254
C 1 Gamma 1 Train Score 1.0 Test Score 0.8253968253968254
C 10 Gamma 0.01 Train Score 1.0 Test Score 0.8253968253968254
C 10 Gamma 0.001 Train Score 1.0 Test Score 0.8312698412698413
C 10 Gamma 0.1 Train Score 1.0 Test Score 0.8253968253968254
C 10 Gamma 1 Train Score 1.0 Test Score 0.8253968253968254
In [17]: for i in ParameterGrid({'C':[0.2,0.5,0.8],
                             'gamma':[0.01, 0.001, 0.0001, 0.005]}):
             model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
             model.fit(X_train_scale,X_train.Score)
             train_score = model.score(X_train_scale, X_train.Score)
             test_score = model.score(X_test, X_test_cv.Score)
             print('C',i['C'],'Gamma',i['gamma'],'Train Score',train_score,
                              'Test Score', test_score)
C 0.2 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.2 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.2 Gamma 0.0001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.2 Gamma 0.005 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.5 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.5 Gamma 0.001 Train Score 0.8568707482993198 Test Score 0.8253968253968254
C 0.5 Gamma 0.0001 Train Score 0.8601360544217687 Test Score 0.8271428571428572
C 0.5 Gamma 0.005 Train Score 0.8563945578231292 Test Score 0.8253968253968254
```

```
C 0.8 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.8 Gamma 0.001 Train Score 0.8745578231292517 Test Score 0.8265079365079365
C 0.8 Gamma 0.0001 Train Score 0.9240136054421769 Test Score 0.8353968253968254
C 0.8 Gamma 0.005 Train Score 0.85666666666666667 Test Score 0.8253968253968254
```

Observed that for high C- training data is overfitting so much. for low c. and low values of gamma is giving somewat better scores than high.

```
In [79]: c = [0.005, 0.01, 0.4, 0.8, 1.2]
        gamma = [0.000009, 0.0008, 0.001, 0.04, 0.2, 5, 10, 15]
        model_grid_bow = GridSearchCV(make_pipeline(CountVectorizer(stop_words=list(stop)),
                                                     StandardScaler(with_mean=False),SVC()),
                                      param_grid={'svc_C': c,'svc_gamma':gamma},
                                     cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_grid_bow.fit(train_df.final_text,train_df.Score)
In [56]: dict_scores = []
         idx = 0
        for i in model_grid_bow.grid_scores_:
             dict_score = []
            dict_score.append(i[0]['svc__gamma'])
            dict_score.append(i[0]['svc__C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_grid_bow.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
             idx = idx + 1
         scores_df = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                        'Test_std', 'Train_score'])
In [70]: #top scores wit grid search
         scores_df.sort_values('Test_score', ascending=False).head(15)
Out [70]:
                           C Test_score Test_std Train_score
                 gamma
         33
             0.000800 1.200
                                0.845207 0.019683
                                                        0.995821
         34
             0.001000 1.200
                                0.844159 0.020326
                                                        0.996686
                                0.843478 0.020834
         32
             0.000009 1.200
                                                        0.883865
         25
             0.000800 0.800
                                0.843269 0.021011
                                                        0.883656
                                0.842902 0.021275
         26
             0.001000 0.800
                                                        0.880644
         17
             0.000800 0.400
                                0.842378 0.021622
                                                        0.869664
                                0.842326 0.021650
         18
             0.001000 0.400
                                                       0.869558
         0
                                0.842273 0.021690
             0.000009 0.005
                                                        0.869488
         28
             0.200000 0.800
                                0.842273 0.021690
                                                        0.869498
         23 15.000000 0.400
                                0.842273 0.021690
                                                        0.869488
         24
            0.000009 0.800
                                0.842273 0.021703
                                                        0.871792
         27
             0.040000 0.800
                                0.842273 0.021690
                                                        0.869498
         30 10.000000 0.800
                                0.842273 0.021690
                                                        0.869498
         29
             5.000000 0.800
                                0.842273 0.021690
                                                       0.869498
         21
             5.000000 0.400
                                0.842273 0.021690
                                                       0.869488
```

```
In [63]: #RandomSearch
        model_random_bow = RandomizedSearchCV(
                            make_pipeline(CountVectorizer(stop_words=list(stop)),
                            StandardScaler(with_mean=False),SVC()),
                            param distributions={'svc C': uniform(loc=0,scale=0.7),
                                'svc_gamma':uniform(loc=0,scale=0.01)},n_iter=15,
                                    cv=TimeSeriesSplit(n splits=10),n jobs=-1)
        model_random_bow.fit(train_df.final_text,train_df.Score)
In [66]: dict_scores = []
        idx = 0
        for i in model_random_bow.grid_scores_:
            dict score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc_C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_random_bow.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df1 = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                       'Test_std', 'Train_score'])
In [72]: scores_df1.sort_values('Test_score',ascending=False).head(10)
Out [72]:
                                Test_score Test_std Train_score
               gamma
            0.000415 0.473474
                                  0.842850 0.021271
                                                         0.870966
        14 0.001121 0.545602
                                  0.842378 0.021622
                                                         0.870091
            0.006791 0.237241
                                  0.842273 0.021690
                                                         0.869488
            0.004997 0.257014
                                  0.842273 0.021690
                                                        0.869488
        2
            0.005213 0.182690
                                  0.842273 0.021690
                                                        0.869488
            0.007353 0.527309 0.842273 0.021690
        3
                                                        0.869498
        5
            0.003778 0.325648
                                  0.842273 0.021690
                                                        0.869488
            0.001744 0.498052
                                  0.842273 0.021690
        6
                                                        0.869552
        7
            0.003124 0.151355
                                  0.842273 0.021690
                                                         0.869488
            0.006712 0.141517
                                  0.842273 0.021690
                                                         0.869488
```

It seems like for high C values, it is giving somewhat better cv score but it is overfitting so much. There is a difference of >15% in train and test scores. so found that gamma = 0.000800 C = 0.400 are the better params with cv score of 0.842378

```
count_vect = CountVectorizer(stop_words=list(stop),binary=True,dtype=np.int8)
        X_train = train_df.iloc[:round(train_df.shape[0]*0.70),:]
        X_test_cv = train_df.iloc[round(train_df.shape[0]*0.70):,:]
        final_counts_train = count_vect.fit_transform(
                X_train['final_text'].values)
        #test
        X_test = count_vect.transform(X_test_cv['final_text'].values)
In [53]: for i in ParameterGrid({'C':[0.4, 0.8, 0.1, 1, 10],
                            'gamma': [0.0008,0.005, 0.1, 1]}):
            model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
            model.fit(final_counts_train,X_train.Score)
            train_score = model.score(final_counts_train,X_train.Score)
            test_score = model.score(X_test, X_test_cv.Score)
            print('C',i['C'],'Gamma',i['gamma'],'Train Score',train_score,
                             'Test Score',test_score)
C 0.4 Gamma 0.0008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.4 Gamma 0.005 Train Score 0.8725850340136054 Test Score 0.840952380952381
C 0.4 Gamma 0.1 Train Score 0.8656462585034014 Test Score 0.8284126984126984
C 0.4 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.8 Gamma 0.0008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.8 Gamma 0.005 Train Score 0.9155102040816326 Test Score 0.8819047619047619
C 0.8 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.0008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.005 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 1 Gamma 0.0008 Train Score 0.8570748299319728 Test Score 0.8257142857142857
C 1 Gamma 0.005 Train Score 0.9269387755102041 Test Score 0.8920634920634921
C 1 Gamma 0.1 Train Score 0.9973469387755102 Test Score 0.8485714285714285
C 1 Gamma 1 Train Score 1.0 Test Score 0.8253968253968254
C 10 Gamma 0.0008 Train Score 0.9442857142857143 Test Score 0.9082539682539682
C 10 Gamma 0.005 Train Score 0.9850340136054422 Test Score 0.92
C 10 Gamma 0.1 Train Score 1.0 Test Score 0.8571428571428571
C 10 Gamma 1 Train Score 1.0 Test Score 0.8253968253968254
```

Tried with binary Count vectorizer and found some interesting results with high accuracy for gamma in range of 0.001-0.01 and and c > 0.8, i am getting some high test scores for the data.

model_grid_bow_binary.fit(train_df.final_text,train_df.Score)

```
In [85]: dict_scores = []
        idx = 0
        for i in model_grid_bow_binary.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc__C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_grid_bow_binary.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df_bin = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                       'Test_std', 'Train_score'])
In [87]: #top scores
        scores_df_bin.sort_values('Test_score', ascending=False).head(10)
Out [87]:
            gamma
                      C Test_score Test_std Train_score
        53 0.010
                    7.0
                           0.921215 0.005992
                                                  0.993688
        61 0.010 10.0
                           0.920639 0.005565
                                                  0.996447
        45 0.010
                         0.920377 0.006399
                   5.0
                                                  0.989241
        60 0.008 10.0
                         0.920063 0.005074
                                                  0.994219
        52 0.008 7.0
                        0.920010 0.005494
                                                  0.990022
        59 0.005 10.0 0.919329 0.005046
                                                  0.986794
        44 0.008 5.0 0.919068 0.006339
                                                  0.984562
        51 0.005 7.0
                        0.918649 0.007152
                                                  0.980237
        58 0.003 10.0 0.917444 0.007297
                                                  0.975592
        37 0.010
                   3.0
                           0.916763 0.009404
                                                  0.979108
In [99]: #random search
        model_random_bow_binary = RandomizedSearchCV(
                            make_pipeline(CountVectorizer(stop_words=list(stop),binary=True),
                                    SVC()),
                         param_distributions={'svc__C': uniform(loc=0,scale=10),
                         'svc_gamma':uniform(loc=0.003,scale=0.017)},n_iter=20,
                                    cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_random_bow_binary.fit(train_df.final_text,train_df.Score)
In [100]: dict_scores = []
         idx = 0
         for i in model_random_bow_binary.grid_scores_:
             dict_score = []
             dict_score.append(i[0]['svc_gamma'])
             dict_score.append(i[0]['svc__C'])
             dict_score.append(i[1])
             dict_score.append(i[2].std())
             dict_score.append(model_random_bow_binary.cv_results_['mean_train_score'][idx])
             dict_scores.append(dict_score)
             idx = idx + 1
```

```
scores_df1_bin = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                        'Test_std', 'Train_score'])
In [103]: #top scores
         scores_df1_bin.sort_values('Test_score',ascending=False).head(10)
Out [103]:
                              C Test_score Test_std Train_score
                gamma
             0.012632 5.387201
                                  0.921320 0.006144
         9
                                                         0.994085
         7
            0.010988 8.434631
                                  0.920744 0.005843
                                                         0.996239
         18 0.013324 5.211789 0.920691 0.006230
                                                         0.994289
         13 0.015818 4.657298 0.920482 0.007147
                                                         0.995270
             0.008742 6.182782
                                  0.920377 0.005578
                                                         0.989817
         6
             0.015201 7.367153 0.920272 0.007147
                                                         0.997701
         2
             0.017782 4.186368 0.920168 0.007176 0.995562
         14 0.006675 7.241270 0.919906 0.006305
                                                         0.987144
             0.018738 5.846448 0.919749 0.007964
                                                         0.997884
         1
         15 0.015909 8.272383
                                   0.919644 0.008027
                                                         0.998387
  Comapared to Non-binary Bag of word binary bag of words score was high and best score
found at gamma = 0.010, C = 7.0 and cv mean score is 0.921215
In [21]: #test scores
        scores_train = []
        from nltk.corpus import stopwords
        stop = set(stopwords.words('english'))
        stop.remove('not')
        stop.remove('very')
        #CountVectorizer for BoW
        count_vect = CountVectorizer(stop_words=list(stop),binary=True,dtype=np.int8)
        final_counts_train = count_vect.fit_transform(
                train_df['final_text'].values)
        #t.est
        X_test = count_vect.transform(test_df['final_text'].values)
        model = SVC(C=7,kernel='rbf',gamma=0.010)
        model.fit(final_counts_train,train_df.Score)
        #Predicting training data
        train_list = model.predict(final_counts_train)
        #Accuracy score
        score_train = accuracy_score(train_df.Score,train_list)
        #predict test cv
        test_list = model.predict(X_test)
        #Accuracy score
        score_test = accuracy_score(test_df.Score,test_list)
        #precision
        #precision
        test_precision = precision_score(test_df.Score,test_list)
        #recall
```

test_recall = recall_score(test_df.Score,test_list)

```
#confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print('C' ,7,'gamma',0.010)
         print('Train Score', score_train)
         print('Test Score',score test)
         print('Test Precision',test_precision)
         print('Test Recall',test_recall)
         print('Test ConfusionMatrix',confusion_matrix_test)
C 7 gamma 0.01
Train Score 0.9915238095238095
Test Score 0.924666666666666
Test Precision 0.9447222953408791
Test Recall 0.9653039268423884
Test ConfusionMatrix [[1144 420]
 [ 258 7178]]
In [22]: #no of support vectrors for each class
         model.n_support_
Out[22]: array([2245, 3591], dtype=int32)
  SGD With BoW
In [88]: #random search
         model_random_bow_binary = RandomizedSearchCV(make_pipeline(
                                       CountVectorizer(stop_words=list(stop),binary=True),
                                        SGDClassifier(n_jobs=-1)),
                              param_distributions={'sgdclassifier_penalty':['11','12'],
                                 'sgdclassifier__alpha':uniform(loc=0.00001,scale=0.069),
                                 'sgdclassifier__l1_ratio':uniform(loc=0,scale=1)},
                                                          n iter=100,
                                     cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
         model_random_bow_binary.fit(train_df.final_text,train_df.Score)
In [89]: dict scores = []
         idx = 0
         for i in model_random_bow_binary.grid_scores_:
             dict_score = []
             dict_score.append(i[0]['sgdclassifier__alpha'])
             dict_score.append(i[0]['sgdclassifier__l1_ratio'])
             dict_score.append(i[0]['sgdclassifier__penalty'])
             dict_score.append(i[1])
             dict_score.append(i[2].std())
             dict_score.append(model_random_bow_binary.cv_results_['mean_train_score'][idx])
             dict_scores.append(dict_score)
             idx = idx + 1
         scores_df1_bin = pd.DataFrame(dict_scores,columns=['alpha','l1_rato','penality','Test
                                                         'Test_std', 'Train_score'])
```

```
In [90]: scores_df1_bin.sort_values('Test_score',ascending=False).head(10)
Out [90]:
                      11_rato penality Test_score Test_std Train_score
               alpha
           0.003570 0.912537
                                         0.915977 0.007592
                                                               0.961918
        13
                                   12
        95
           0.000654 0.544642
                                   12
                                         0.915872 0.006177
                                                               0.980992
        38 0.000256 0.797285
                                   12
                                         0.910529 0.006600
                                                               0.983756
        46 0.007463 0.314634
                                   12
                                         0.906129 0.010484
                                                               0.946527
        80 0.010533 0.760320
                                   12
                                         0.901414 0.010897
                                                               0.937890
        7
           0.000199 0.744540
                                   11
                                         0.900681 0.006887
                                                               0.962679
        63 0.011535 0.273312
                                       0.900210 0.011199
                                   12
                                                               0.935924
        56 0.012332 0.480412
                                   12
                                         0.898638 0.010939
                                                               0.934297
                                   12
        22 0.013868 0.611535
                                         0.893557 0.009607
                                                               0.928721
        77 0.015297 0.045345
                                   12
                                         0.892981 0.011083
                                                               0.926446
```

Got best mean cv at alpha = 0.003570. l1_ratio = 0.912537 and penality l2 and corresponding mean cv test score is 0.915977

```
In [93]: #test scores
         scores_train = []
         from nltk.corpus import stopwords
         stop = set(stopwords.words('english'))
         stop.remove('not')
         stop.remove('very')
         #CountVectorizer for BoW
         count_vect = CountVectorizer(stop_words=list(stop),binary=True,dtype=np.int8)
         final_counts_train = count_vect.fit_transform(
                 train_df['final_text'].values)
         X_test = count_vect.transform(test_df['final_text'].values)
         model = SGDClassifier(penalty='12',alpha=0.003570,l1_ratio=0.912537,n_jobs=-1) #0.003
         model.fit(final_counts_train,train_df.Score)
         #Predicting training data
         train_list = model.predict(final_counts_train)
         #Accuracy score
         score_train = accuracy_score(train_df.Score,train_list)
         #predict test cv
         test_list = model.predict(X_test)
         #Accuracy score
         score_test = accuracy_score(test_df.Score,test_list)
         #precision
         #precision
         test_precision = precision_score(test_df.Score,test_list)
         test_recall = recall_score(test_df.Score,test_list)
         #confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print("With SGD Classifier penalty='12',alpha=0.003570,l1_ratio=0.912537 ")
```

```
print('Train Score', score_train)
         print('Test Score',score_test)
         print('Test Precision',test_precision)
         print('Test Recall',test_recall)
         print('Test ConfusionMatrix',confusion matrix test)
With SGD Classifier penalty='12',alpha=0.003570,l1_ratio=0.912537
Train Score 0.9445238095238095
Test Score 0.91977777777778
Test Precision 0.930053804765565
Test Recall 0.9763313609467456
Test ConfusionMatrix [[1018 546]
 [ 176 7260]]
0.0.3 Tf-Idf
In [13]: #TFIDF with (1,2) gram with cleaned data
         #simple cv for train data
         #tfidf vec
         tf_idf_vect = TfidfVectorizer(ngram_range=(1,2))
         X_train = train_df.iloc[:round(train_df.shape[0]*0.70),:]
         X_test_cv = train_df.iloc[round(train_df.shape[0]*0.70):,:]
         final_counts_train = tf_idf_vect.fit_transform(
                 X_train['final_text'].values)
         #test
         X_test = tf_idf_vect.transform(X_test_cv['final_text'].values)
In [24]: for i in ParameterGrid(('C':[0.001,0.01,0.1,1,5,10],
                             'gamma': [0.001,0.008,0.01,0.1,0.5,1,10]}):
             model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
             model.fit(final_counts_train,X_train.Score)
             train_score = model.score(final_counts_train,X_train.Score)
             test_score = model.score(X_test, X_test_cv.Score)
             print('C',i['C'],'Gamma',i['gamma'],'Train Score',train_score,
                              'Test Score',test_score)
C 0.001 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
```

```
C 0.01 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 1 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 1 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 1 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 1 Gamma 0.1 Train Score 0.8623129251700681 Test Score 0.8311111111111111
C 1 Gamma 0.5 Train Score 0.978843537414966 Test Score 0.90111111111111111
C 1 Gamma 1 Train Score 0.9975510204081632 Test Score 0.8942857142857142
C 1 Gamma 10 Train Score 1.0 Test Score 0.8253968253968254
C 5 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 5 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 5 Gamma 0.01 Train Score 0.8564625850340136 Test Score 0.8253968253968254
C 5 Gamma 0.5 Train Score 1.0 Test Score 0.93
C 5 Gamma 1 Train Score 1.0 Test Score 0.9146031746031746
C 5 Gamma 10 Train Score 1.0 Test Score 0.8253968253968254
C 10 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 10 Gamma 0.008 Train Score 0.8604081632653061 Test Score 0.8293650793650794
C 10 Gamma 0.01 Train Score 0.8687074829931973 Test Score 0.8380952380952381
C 10 Gamma 0.1 Train Score 0.9999319727891156 Test Score 0.9382539682539682
C 10 Gamma 0.5 Train Score 1.0 Test Score 0.93
C 10 Gamma 1 Train Score 1.0 Test Score 0.9146031746031746
C 10 Gamma 10 Train Score 1.0 Test Score 0.8253968253968254
In [66]: c = [0.1, 0.5, 0.8, 1, 5, 7, 10, 20]
        gamma = [0.008, 0.007, 0.1, 0.3, 0.5, 1, 3, 10]
        model_grid_tfidf = GridSearchCV(make_pipeline(TfidfVectorizer(ngram_range=(1,2)),
                                                    SVC()).
                                     param_grid={'svc__C': c,'svc__gamma':gamma},
                                    cv=TimeSeriesSplit(n splits=10),n jobs=-1)
        model_grid_tfidf.fit(train_df.final_text,train_df.Score)
In [67]: dict_scores = []
        idx = 0
        for i in model_grid_tfidf.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc_C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
```

```
dict_score.append(model_grid_tfidf.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                      'Test std', 'Train score'])
In [71]: scores_df.sort_values('Test_score',ascending=False).head(10)
Out [71]:
                      C Test_score Test_std Train_score
            gamma
              0.1 10.0
        50
                          0.925982 0.013537
                                                 0.999961
        58
              0.1 20.0
                          0.925930 0.013911
                                                 1.000000
        42
              0.1 7.0
                        0.922263 0.014279
                                                 0.999523
        35
              0.3
                  5.0
                        0.922263 0.014587
                                                 1.000000
              0.3 10.0
                        0.922211 0.014668
        51
                                                 1.000000
        43
              0.3 7.0 0.922211 0.014668
                                                 1.000000
        59
              0.3 20.0 0.922211 0.014668
                                                 1.000000
        36
              0.5 5.0 0.918806 0.014783
                                                 1.000000
                          0.918806 0.014783
        52
              0.5 10.0
                                                 1.000000
        44
              0.5
                  7.0
                          0.918806 0.014783
                                                 1.000000
```

for high values of c model is overfitting to train data and for each c with reasonable gamma is giving good score than low or high gamma.

```
In [12]: model_random_tfidf = RandomizedSearchCV(
                               make_pipeline(TfidfVectorizer(ngram_range=(1,2)),SVC()),
                               param_distributions={'svc__C':uniform(loc=0,scale=12),
                                                'svc gamma':uniform(loc=0,scale=0.7)},
                                     n_iter=20,cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_random_tfidf.fit(train_df.final_text,train_df.Score)
In [13]: dict_scores = []
         idx = 0
         for i in model_random_tfidf.grid_scores_:
             dict_score = []
             dict_score.append(i[0]['svc_gamma'])
             dict_score.append(i[0]['svc_C'])
             dict_score.append(i[1])
             dict_score.append(i[2].std())
             dict_score.append(model_random_tfidf.cv_results_['mean_train_score'][idx])
             dict_scores.append(dict_score)
             idx = idx + 1
         scores_df1 = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                        'Test_std', 'Train_score'])
In [17]: #top scores with random search
         scores_df1.sort_values('Test_score',ascending=False).head(10)
Out[17]:
                                 Test_score Test_std Train_score
                gamma
         13 0.175194
                       7.107109
                                    0.924987 0.014313
                                                           1.000000
```

```
4
            0.205345
                       7.555669
                                  0.924620 0.014065
                                                         1.000000
            0.192857 4.893355 0.923258 0.014181
        0
                                                         0.999950
            0.266751 8.332227
                                 0.923206 0.014551
                                                         1.000000
        8
        11 0.324423 5.865958
                                  0.922001 0.014653
                                                         1.000000
        15 0.311839 10.320605 0.921896 0.014627
                                                         1.000000
        18 0.273510 3.244258
                                  0.921320 0.014689
                                                         0.999898
        19 0.347285 9.050680 0.921163 0.014763
                                                         1.000000
            0.430498 11.934278 0.920430 0.014574
        6
                                                         1.000000
        14 0.068950 8.277344
                                  0.919958 0.015449
                                                         0.998265
  From 10 fold cv got high mean cv at gamma = 0.175194, C = 7.107109 and mean cv is
0.924987
In [19]: #test scores
        #TFIDF with (1,2) gram with cleaned data
        #tfidf vec
        tf idf vect = TfidfVectorizer(ngram range=(1,2))
        final_counts_train = tf_idf_vect.fit_transform(
                train_df['final_text'].values)
        #test
        X_test = tf_idf_vect.transform(test_df['final_text'].values)
        model = SVC(C=7.107109,kernel='rbf',gamma=0.175194)
        model.fit(final_counts_train,train_df.Score)
        #Predicting training data
        train_list = model.predict(final_counts_train)
        #Accuracy score
        score_train = accuracy_score(train_df.Score,train_list)
        #predict test cv
        test_list = model.predict(X_test)
        #Accuracy score
        score test = accuracy score(test df.Score,test list)
        #precision
        #precision
        test_precision = precision_score(test_df.Score,test_list)
        #recall
        test_recall = recall_score(test_df.Score,test_list)
        #confusion matrix
        confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
        print('C' ,7,'gamma',0.175194)
        print('Train Score', score_train)
        print('Test Score',score_test)
        print('Test Precision',test_precision)
        print('Test Recall',test_recall)
        print('Test ConfusionMatrix',confusion_matrix_test)
        print('No of support vectors for each class',model.n_support_)
C 7 gamma 0.175194
```

Train Score 1.0

```
Test Score 0.941555555555556
Test Precision 0.9508089770354906
Test Recall 0.9799623453469607
Test ConfusionMatrix [[1187 377]
 [ 149 7287]]
No of support vectors for each class [2810 6720]
  SGD Classifier
In [14]: model_random_tfidf = RandomizedSearchCV(make_pipeline(TfidfVectorizer(ngram_range=(1,
                                                    SGDClassifier(n_jobs=-1)),
                                     param_distributions={'sgdclassifier__penalty':['11','12']
                                         'sgdclassifier__alpha':uniform(loc=0.00001,scale=0.06
                                           'sgdclassifier__l1_ratio':uniform(loc=0,scale=1)},n
                                                 ,cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_random_tfidf.fit(train_df.final_text,train_df.Score)
In [18]: dict_scores = []
        idx = 0
        for i in model_random_tfidf.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['sgdclassifier__alpha'])
            dict_score.append(i[0]['sgdclassifier__l1_ratio'])
            dict_score.append(i[0]['sgdclassifier__penalty'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_random_tfidf.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
             idx = idx + 1
         scores_df1 = pd.DataFrame(dict_scores,columns=['alpha','l1_ratio','penality','Test_sc
                                                        'Test_std', 'Train_score'])
In [19]: scores_df1.sort_values('Test_score',ascending=False).head(10)
Out [19]:
                alpha l1_ratio penality Test_score Test_std Train_score
        206 0.000116 0.738118
                                      12
                                            0.912729 0.007599
                                                                   0.983543
         263 0.000233 0.550213
                                      12
                                            0.878104 0.017691
                                                                   0.935960
         153 0.000264 0.305034
                                      12
                                            0.870875 0.019746
                                                                   0.925270
         184 0.000207 0.538170
                                            0.865008 0.015358
                                      11
                                                                   0.885320
         29
             0.000437 0.431934
                                      12
                                            0.847564 0.023918
                                                                   0.886908
         268 0.000545 0.509713
                                      12
                                            0.843426 0.022213
                                                                   0.879669
         179 0.000690 0.546427
                                      12
                                            0.842326 0.021736
                                                                   0.873685
                                            0.842273 0.021690
         202 0.043333 0.990595
                                      11
                                                                   0.869488
         201 0.025632 0.245308
                                      11
                                            0.842273 0.021690
                                                                   0.869488
        200 0.016020 0.509449
                                      11
                                            0.842273 0.021690
                                                                   0.869488
In [24]: for i in ParameterGrid({'alpha':[0.00005,0.00008,0.0001,0.00012,0.00018,0.00023],
                             'l1_ratio': [0,0.03,0.05,0.08,0.1,0.15,0.25,0.35,0.45,0.55,0.65,0.
```

```
train_score = model.score(final_counts_train,X_train.Score)
           test score = model.score(X test, X test cv.Score)
           print('Alpha',i['alpha'],'11_ratio',i['11_ratio'],'Penality',i['penality'],
                 'Train Score', train score, 'Test Score', test score)
Alpha 5e-05 11_ratio 0 Penality 11 Train Score 0.9470748299319728 Test Score 0.930317460317460
Alpha 5e-05 11 ratio 0 Penality 12 Train Score 0.998639455782313 Test Score 0.93444444444444444
Alpha 5e-05 11_ratio 0.03 Penality 11 Train Score 0.9455102040816327 Test Score 0.927936507936
Alpha 5e-05 11 ratio 0.03 Penality 12 Train Score 0.9984353741496599 Test Score 0.935079365079
Alpha 5e-05 11_ratio 0.05 Penality 11 Train Score 0.9473469387755102 Test Score 0.928095238095
Alpha 5e-05 11 ratio 0.05 Penality 12 Train Score 0.9985034013605442 Test Score 0.935079365079
Alpha 5e-05 11 ratio 0.08 Penality 12 Train Score 0.998639455782313 Test Score 0.9326984126984
Alpha 5e-05 11 ratio 0.1 Penality 11 Train Score 0.9444897959183673 Test Score 0.9287301587301
Alpha 5e-05 11_ratio 0.1 Penality 12 Train Score 0.998639455782313 Test Score 0.93571428571428
Alpha 5e-05 11_ratio 0.15 Penality 11 Train Score 0.9457142857142857 Test Score 0.927142857142
Alpha 5e-05 11_ratio 0.25 Penality 11 Train Score 0.9439455782312925 Test Score 0.924603174603
Alpha 5e-05 11 ratio 0.35 Penality 11 Train Score 0.9451020408163265 Test Score 0.925714285714
Alpha 5e-05 11_ratio 0.35 Penality 12 Train Score 0.9983673469387755 Test Score 0.932857142857
Alpha 5e-05 11 ratio 0.45 Penality 11 Train Score 0.9470748299319728 Test Score 0.928412698412
Alpha 5e-05 11_ratio 0.45 Penality 12 Train Score 0.998639455782313 Test Score 0.9368253968253
Alpha 5e-05 11_ratio 0.55 Penality 11 Train Score 0.9471428571428572 Test Score 0.929523809523
Alpha 5e-05 11 ratio 0.55 Penality 12 Train Score 0.998639455782313 Test Score 0.93507936507936
Alpha 5e-05 11 ratio 0.65 Penality 11 Train Score 0.9452380952380952 Test Score 0.9266666666666
Alpha 5e-05 11_ratio 0.65 Penality 12 Train Score 0.9983673469387755 Test Score 0.932539682539
Alpha 5e-05 11 ratio 0.75 Penality 11 Train Score 0.9455102040816327 Test Score 0.9265079365079
Alpha 5e-05 11 ratio 0.75 Penality 12 Train Score 0.9984353741496599 Test Score 0.9358730158730
Alpha 5e-05 11_ratio 0.85 Penality 11 Train Score 0.947687074829932 Test Score 0.9298412698412
Alpha 5e-05 11 ratio 0.85 Penality 12 Train Score 0.9985034013605442 Test Score 0.933650793650
Alpha 5e-05 11_ratio 0.95 Penality 11 Train Score 0.9448979591836735 Test Score 0.925238095238
Alpha 5e-05 11_ratio 0.95 Penality 12 Train Score 0.9985714285714286 Test Score 0.936984126984
Alpha 8e-05 11_ratio 0 Penality 11 Train Score 0.9295918367346939 Test Score 0.914761904761904
Alpha 8e-05 11_ratio 0 Penality 12 Train Score 0.9891836734693877 Test Score 0.926825396825396
Alpha 8e-05 11_ratio 0.03 Penality 11 Train Score 0.9272789115646258 Test Score 0.9141269841269
Alpha 8e-05 11 ratio 0.03 Penality 12 Train Score 0.9894557823129252 Test Score 0.926984126984
Alpha 8e-05 11_ratio 0.05 Penality 11 Train Score 0.9251700680272109 Test Score 0.912539682539
Alpha 8e-05 11 ratio 0.05 Penality 12 Train Score 0.9907482993197279 Test Score 0.928095238095
Alpha 8e-05 11_ratio 0.08 Penality 11 Train Score 0.9274829931972789 Test Score 0.9130158730158
Alpha 8e-05 11 ratio 0.08 Penality 12 Train Score 0.9907482993197279 Test Score 0.927936507936
Alpha 8e-05 11 ratio 0.1 Penality 11 Train Score 0.9248979591836735 Test Score 0.9139682539682
Alpha 8e-05 11_ratio 0.1 Penality 12 Train Score 0.988843537414966 Test Score 0.924761904761904
Alpha 8e-05 11_ratio 0.15 Penality 11 Train Score 0.9279591836734694 Test Score 0.912857142857
```

'penality':['11','12']}):

model.fit(final_counts_train,X_train.Score)

model = SGDClassifier(penalty=i['penality'],alpha=i['alpha'],l1_ratio=i['l1_ratio

```
Alpha 8e-05 11_ratio 0.25 Penality 11 Train Score 0.9262585034013605 Test Score 0.909682539682
Alpha 8e-05 11_ratio 0.25 Penality 12 Train Score 0.9885034013605443 Test Score 0.9266666666666
Alpha 8e-05 11 ratio 0.35 Penality 11 Train Score 0.9283673469387755 Test Score 0.915396825396
Alpha 8e-05 11_ratio 0.35 Penality 12 Train Score 0.9893877551020408 Test Score 0.926031746031
Alpha 8e-05 11 ratio 0.45 Penality 11 Train Score 0.9261224489795918 Test Score 0.911428571428
Alpha 8e-05 11_ratio 0.45 Penality 12 Train Score 0.9889795918367347 Test Score 0.9247619047619
Alpha 8e-05 11 ratio 0.55 Penality 11 Train Score 0.9268027210884354 Test Score 0.9123809523809
Alpha 8e-05 11_ratio 0.55 Penality 12 Train Score 0.9903401360544217 Test Score 0.9285714285714
Alpha 8e-05 11 ratio 0.65 Penality 11 Train Score 0.9248979591836735 Test Score 0.908412698412
Alpha 8e-05 11_ratio 0.65 Penality 12 Train Score 0.9908163265306122 Test Score 0.929365079365
Alpha 8e-05 11 ratio 0.75 Penality 11 Train Score 0.9285034013605442 Test Score 0.913650793650
Alpha 8e-05 11 ratio 0.75 Penality 12 Train Score 0.9907482993197279 Test Score 0.928730158730
Alpha 8e-05 11_ratio 0.85 Penality 11 Train Score 0.926734693877551 Test Score 0.9125396825396
Alpha 8e-05 11 ratio 0.85 Penality 12 Train Score 0.9890476190476191 Test Score 0.924920634920
Alpha 8e-05 11_ratio 0.95 Penality 11 Train Score 0.9276190476190476 Test Score 0.911904761904
Alpha 8e-05 11 ratio 0.95 Penality 12 Train Score 0.991156462585034 Test Score 0.93
Alpha 0.0001 11_ratio 0 Penality 11 Train Score 0.9206122448979592 Test Score 0.90825396825396
Alpha 0.0001 11 ratio 0 Penality 12 Train Score 0.976326530612245 Test Score 0.9174603174603174
Alpha 0.0001 11_ratio 0.03 Penality 11 Train Score 0.9180272108843538 Test Score 0.90380952380
Alpha 0.0001 11 ratio 0.03 Penality 12 Train Score 0.9778231292517007 Test Score 0.91793650793
Alpha 0.0001 11 ratio 0.05 Penality 11 Train Score 0.9162585034013605 Test Score 0.90285714285
Alpha 0.0001 11 ratio 0.05 Penality 12 Train Score 0.9795918367346939 Test Score 0.91936507936
Alpha 0.0001 11_ratio 0.08 Penality 11 Train Score 0.9176190476190477 Test Score 0.90428571428
Alpha 0.0001 11_ratio 0.1 Penality 11 Train Score 0.9201360544217687 Test Score 0.9066666666666
Alpha 0.0001 11_ratio 0.1 Penality 12 Train Score 0.9780952380952381 Test Score 0.919365079365
Alpha 0.0001 11 ratio 0.15 Penality 11 Train Score 0.9177551020408163 Test Score 0.90523809523
Alpha 0.0001 11_ratio 0.15 Penality 12 Train Score 0.9774149659863945 Test Score 0.91841269841
Alpha 0.0001 11 ratio 0.25 Penality 12 Train Score 0.9759183673469388 Test Score 0.91619047619
Alpha 0.0001 11 ratio 0.35 Penality 11 Train Score 0.9171428571428571 Test Score 0.905079365079
Alpha 0.0001 11_ratio 0.35 Penality 12 Train Score 0.9795238095238096 Test Score 0.92158730158
Alpha 0.0001 11 ratio 0.45 Penality 11 Train Score 0.917687074829932 Test Score 0.9047619047619
Alpha 0.0001 11_ratio 0.45 Penality 12 Train Score 0.9797959183673469 Test Score 0.92079365079
Alpha 0.0001 11 ratio 0.55 Penality 11 Train Score 0.9180952380952381 Test Score 0.90476190476
Alpha 0.0001 l1_ratio 0.55 Penality 12 Train Score 0.98 Test Score 0.9203174603174603
Alpha 0.0001 11 ratio 0.65 Penality 11 Train Score 0.9184353741496598 Test Score 0.90555555555
Alpha 0.0001 11_ratio 0.65 Penality 12 Train Score 0.9772789115646259 Test Score 0.91698412698
Alpha 0.0001 11_ratio 0.75 Penality 11 Train Score 0.9205442176870748 Test Score 0.90714285714
Alpha 0.0001 11_ratio 0.75 Penality 12 Train Score 0.9800680272108844 Test Score 0.92206349206
Alpha 0.0001 11_ratio 0.85 Penality 11 Train Score 0.9163945578231293 Test Score 0.90126984126
Alpha 0.0001 11 ratio 0.85 Penality 12 Train Score 0.9791836734693877 Test Score 0.92142857142
Alpha 0.0001 11_ratio 0.95 Penality 11 Train Score 0.9194557823129251 Test Score 0.90619047619
Alpha 0.0001 11 ratio 0.95 Penality 12 Train Score 0.9761224489795919 Test Score 0.91809523809
Alpha 0.00012 11_ratio 0 Penality 11 Train Score 0.9089795918367347 Test Score 0.8968253968253
Alpha 0.00012 11_ratio 0 Penality 12 Train Score 0.9679591836734693 Test Score 0.91174603174603
Alpha 0.00012 11_ratio 0.03 Penality 11 Train Score 0.9079591836734694 Test Score 0.8942857142
Alpha 0.00012 11_ratio 0.03 Penality 12 Train Score 0.9718367346938775 Test Score 0.9139682539
```

```
Alpha 0.00012 11_ratio 0.05 Penality 11 Train Score 0.9091156462585034 Test Score 0.8969841269
Alpha 0.00012 11_ratio 0.05 Penality 12 Train Score 0.9671428571428572 Test Score 0.9101587301
Alpha 0.00012 11_ratio 0.08 Penality 11 Train Score 0.9091836734693878 Test Score 0.8949206349
Alpha 0.00012 11_ratio 0.08 Penality 12 Train Score 0.9682312925170068 Test Score 0.9120634920
Alpha 0.00012 11 ratio 0.1 Penality 11 Train Score 0.9068707482993197 Test Score 0.89476190476
Alpha 0.00012 11_ratio 0.1 Penality 12 Train Score 0.9696598639455782 Test Score 0.91238095238
Alpha 0.00012 11 ratio 0.15 Penality 11 Train Score 0.9097278911564626 Test Score 0.8973015873
Alpha 0.00012 11_ratio 0.15 Penality 12 Train Score 0.9698639455782313 Test Score 0.9144444444
Alpha 0.00012 11 ratio 0.25 Penality 11 Train Score 0.9082993197278911 Test Score 0.8950793650
Alpha 0.00012 11_ratio 0.25 Penality 12 Train Score 0.9674829931972789 Test Score 0.9119047619
Alpha 0.00012 11_ratio 0.35 Penality 11 Train Score 0.9074829931972789 Test Score 0.8949206349
Alpha 0.00012 11_ratio 0.35 Penality 12 Train Score 0.969047619047619 Test Score 0.91222222222
Alpha 0.00012 11_ratio 0.45 Penality 11 Train Score 0.9074829931972789 Test Score 0.8947619047
Alpha 0.00012 11_ratio 0.45 Penality 12 Train Score 0.9695238095238096 Test Score 0.9131746031
Alpha 0.00012 11_ratio 0.55 Penality 11 Train Score 0.9112925170068027 Test Score 0.8987301587
Alpha 0.00012 11_ratio 0.55 Penality 12 Train Score 0.9710884353741497 Test Score 0.9125396825
Alpha 0.00012 11_ratio 0.65 Penality 11 Train Score 0.9098639455782312 Test Score 0.8979365079
Alpha 0.00012 11_ratio 0.65 Penality 12 Train Score 0.968843537414966 Test Score 0.91174603174
Alpha 0.00012 11_ratio 0.75 Penality 11 Train Score 0.9104081632653062 Test Score 0.8977777777
Alpha 0.00012 11 ratio 0.75 Penality 12 Train Score 0.9705442176870749 Test Score 0.9144444444
Alpha 0.00012 11_ratio 0.85 Penality 11 Train Score 0.9074149659863946 Test Score 0.8950793650
Alpha 0.00012 11 ratio 0.85 Penality 12 Train Score 0.9674149659863945 Test Score 0.9111111111
Alpha 0.00012 11_ratio 0.95 Penality 11 Train Score 0.9091836734693878 Test Score 0.8965079365
Alpha 0.00012 11_ratio 0.95 Penality 12 Train Score 0.9676190476190476 Test Score 0.9114285714
Alpha 0.00018 11_ratio 0 Penality 11 Train Score 0.8865306122448979 Test Score 0.87158730158730
Alpha 0.00018 11_ratio 0 Penality 12 Train Score 0.9310884353741496 Test Score 0.8846031746031
Alpha 0.00018 11_ratio 0.03 Penality 11 Train Score 0.8848979591836734 Test Score 0.8677777777
Alpha 0.00018 11_ratio 0.03 Penality 12 Train Score 0.937891156462585 Test Score 0.89111111111
Alpha 0.00018 11_ratio 0.05 Penality 11 Train Score 0.8859183673469387 Test Score 0.8695238095
Alpha 0.00018 11_ratio 0.05 Penality 12 Train Score 0.9338095238095238 Test Score 0.8876190476
Alpha 0.00018 11 ratio 0.08 Penality 11 Train Score 0.8852380952380953 Test Score 0.8690476190
Alpha 0.00018 11_ratio 0.1 Penality 11 Train Score 0.8846938775510204 Test Score 0.86841269841
Alpha 0.00018 11_ratio 0.1 Penality 12 Train Score 0.9347619047619048 Test Score 0.88809523809
Alpha 0.00018 11 ratio 0.15 Penality 11 Train Score 0.8858503401360545 Test Score 0.87079365079
Alpha 0.00018 11_ratio 0.15 Penality 12 Train Score 0.9362585034013605 Test Score 0.8911111111
Alpha 0.00018 11 ratio 0.25 Penality 11 Train Score 0.8900680272108844 Test Score 0.8741269841
Alpha 0.00018 11_ratio 0.25 Penality 12 Train Score 0.932312925170068 Test Score 0.885714285714
Alpha 0.00018 11_ratio 0.35 Penality 11 Train Score 0.8859863945578231 Test Score 0.8703174603
Alpha 0.00018 11_ratio 0.35 Penality 12 Train Score 0.9297278911564626 Test Score 0.8838095238
Alpha 0.00018 11_ratio 0.45 Penality 11 Train Score 0.8850340136054422 Test Score 0.8679365079
Alpha 0.00018 11_ratio 0.45 Penality 12 Train Score 0.9336734693877551 Test Score 0.8880952380
Alpha 0.00018 11_ratio 0.55 Penality 11 Train Score 0.886734693877551 Test Score 0.871904761904
Alpha 0.00018 11_ratio 0.55 Penality 12 Train Score 0.936734693877551 Test Score 0.890476190476
Alpha 0.00018 11_ratio 0.65 Penality 11 Train Score 0.886666666666667 Test Score 0.8711111111
Alpha 0.00018 11_ratio 0.65 Penality 12 Train Score 0.9308843537414966 Test Score 0.8849206349
Alpha 0.00018 11_ratio 0.75 Penality 11 Train Score 0.8862585034013606 Test Score 0.8698412698
Alpha 0.00018 11_ratio 0.75 Penality 12 Train Score 0.934625850340136 Test Score 0.88841269841
```

```
Alpha 0.00018 11_ratio 0.85 Penality 11 Train Score 0.8873469387755102 Test Score 0.87079365079
Alpha 0.00018 l1_ratio 0.85 Penality 12 Train Score 0.9358503401360544 Test Score 0.89
Alpha 0.00018 11_ratio 0.95 Penality 11 Train Score 0.886734693877551 Test Score 0.87158730158
Alpha 0.00018 11_ratio 0.95 Penality 12 Train Score 0.9342176870748299 Test Score 0.8879365079
Alpha 0.00023 11 ratio 0 Penality 11 Train Score 0.872312925170068 Test Score 0.85190476190476
Alpha 0.00023 11_ratio 0 Penality 12 Train Score 0.9082993197278911 Test Score 0.8690476190476
Alpha 0.00023 11 ratio 0.03 Penality 11 Train Score 0.8717006802721089 Test Score 0.8496825396
Alpha 0.00023 11_ratio 0.03 Penality 12 Train Score 0.9046258503401361 Test Score 0.8663492063
Alpha 0.00023 11_ratio 0.05 Penality 11 Train Score 0.8714965986394558 Test Score 0.8496825396
Alpha 0.00023 11_ratio 0.05 Penality 12 Train Score 0.9078231292517007 Test Score 0.8684126984
Alpha 0.00023 11_ratio 0.08 Penality 11 Train Score 0.8709523809523809 Test Score 0.8509523809
Alpha 0.00023 11_ratio 0.08 Penality 12 Train Score 0.9056462585034014 Test Score 0.8665079365
Alpha 0.00023 11_ratio 0.1 Penality 11 Train Score 0.867482993197279 Test Score 0.845238095238
Alpha 0.00023 11_ratio 0.1 Penality 12 Train Score 0.9051020408163265 Test Score 0.86682539682
Alpha 0.00023 11_ratio 0.15 Penality 11 Train Score 0.8693197278911564 Test Score 0.84714285714
Alpha 0.00023 11_ratio 0.15 Penality 12 Train Score 0.9060544217687074 Test Score 0.86730158730
Alpha 0.00023 11_ratio 0.25 Penality 11 Train Score 0.8701360544217687 Test Score 0.84714285714
Alpha 0.00023 11_ratio 0.25 Penality 12 Train Score 0.9053061224489796 Test Score 0.8668253968
Alpha 0.00023 11_ratio 0.35 Penality 11 Train Score 0.8735374149659864 Test Score 0.8526984126
Alpha 0.00023 11 ratio 0.35 Penality 12 Train Score 0.906734693877551 Test Score 0.86809523809
Alpha 0.00023 11_ratio 0.45 Penality 11 Train Score 0.8731972789115646 Test Score 0.85222222222
Alpha 0.00023 11 ratio 0.45 Penality 12 Train Score 0.9056462585034014 Test Score 0.8666666666
Alpha 0.00023 11_ratio 0.55 Penality 11 Train Score 0.8721088435374149 Test Score 0.8514285714:
Alpha 0.00023 11_ratio 0.55 Penality 12 Train Score 0.9038095238095238 Test Score 0.8652380952
Alpha 0.00023 11_ratio 0.65 Penality 11 Train Score 0.8682312925170068 Test Score 0.8476190476
Alpha 0.00023 11_ratio 0.65 Penality 12 Train Score 0.904421768707483 Test Score 0.86619047619
Alpha 0.00023 11_ratio 0.75 Penality 12 Train Score 0.9059183673469388 Test Score 0.8665079365
Alpha 0.00023 11_ratio 0.85 Penality 11 Train Score 0.871156462585034 Test Score 0.849206349206
Alpha 0.00023 11_ratio 0.85 Penality 12 Train Score 0.9070068027210885 Test Score 0.8673015873
Alpha 0.00023 11_ratio 0.95 Penality 11 Train Score 0.8695918367346939 Test Score 0.84904761904
Alpha 0.00023 11_ratio 0.95 Penality 12 Train Score 0.9082312925170068 Test Score 0.8695238095
```

In Random search i didnt got some results wit high 11 ratio and 12 peanality so i dicided to try some low 11 ratios wit same learning rate range and 11 penality. i sisnt get this case in random search. so did some initial investigation above and got some good scores without varince problem also.

```
In [70]: dict_scores = []
         idx = 0
         for i in model_grid_tfidf.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['sgdclassifier__alpha'])
            dict_score.append(i[0]['sgdclassifier__l1_ratio'])
            dict_score.append(i[0]['sgdclassifier__penalty'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_grid_tfidf.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
             idx = idx + 1
         scores_df = pd.DataFrame(dict_scores,columns=['alpha','l1_ratio','penality','Test_scores
                                                        'Test_std', 'Train_score'])
In [72]: scores_df.sort_values('Test_score',ascending=False).head(10)
Out [72]:
              alpha l1_ratio penality Test_score Test_std Train_score
         10 0.00003
                         0.15
                                          0.929282 0.008460
                                                                 0.976729
            0.00003
                         0.00
                                    11
                                          0.929178 0.010054
         0
                                                                 0.976633
                         0.03
         2
            0.00003
                                    11
                                         0.929125 0.010393
                                                                0.976605
         4
            0.00003
                         0.05
                                    11
                                          0.929073 0.008819
                                                                0.976753
         12 0.00003
                         0.25
                                    11
                                          0.928968 0.008937
                                                                0.976537
            0.00003
                        0.10
                                    11
                                          0.928968 0.007552
        8
                                                                0.976478
            0.00003
                        0.08
         6
                                    11
                                          0.928916 0.008576
                                                                0.976547
         16 0.00003
                         0.45
                                    11
                                          0.928759 0.008902
                                                                 0.976561
         18 0.00003
                         0.55
                                    11
                                          0.928654 0.008387
                                                                 0.976760
         24 0.00003
                         0.85
                                    11
                                          0.928444 0.007099
                                                                 0.977262
  Got best scores at alpha = 0.00003, 11_ratio = 0.15, penalty = 11 and mean cv score is 0.929282
In [88]: #test scores
         #TFIDF with (1,2) gram with cleaned data
         #tfidf vec
        tf_idf_vect = TfidfVectorizer(ngram_range=(1,2))
        final_counts_train = tf_idf_vect.fit_transform(
                train_df['final_text'].values)
         #test
        X_test = tf_idf_vect.transform(test_df['final_text'].values)
        model = SGDClassifier(penalty='11',alpha=0.00003,11_ratio=0.15)
        model.fit(final_counts_train,train_df.Score)
         #Predicting training data
         train_list = model.predict(final_counts_train)
         #Accuracy score
         score_train = accuracy_score(train_df.Score,train_list)
         #predict test cv
         test_list = model.predict(X_test)
         #Accuracy score
```

```
score_test = accuracy_score(test_df.Score,test_list)
         #precision
         #precision
         test_precision = precision_score(test_df.Score,test_list)
         #recall
         test_recall = recall_score(test_df.Score,test_list)
         #confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print("penalty='11',alpha=0.00003,11_ratio=0.15")
         print('Train Score', score_train)
         print('Test Score',score_test)
         print('Test Precision',test_precision)
         print('Test Recall',test_recall)
         print('Test ConfusionMatrix',confusion_matrix_test)
penalty='11',alpha=0.00003,l1_ratio=0.15
Train Score 0.9569047619047619
Test Score 0.93555555555556
Test Precision 0.9517659462308908
Test Recall 0.9712210866057019
Test ConfusionMatrix [[1198 366]
[ 214 7222]]
0.0.4 Word2Vec
In [19]: #importing
         from gensim.models import Word2Vec
         from gensim.models import KeyedVectors
         import pickle
         import gensim
In [20]: import gensim
        list_of_sent=[]
         for sent in final review.final text.values:
             list_of_sent.append(sent.split())
In [24]: #word2vec model with 50 dim vector
         w2v_model_50=gensim.models.Word2Vec(list_of_sent,min_count=5,size=50, workers=8)
         #word2vec model with 100 dim vector
         w2v_model_100=gensim.models.Word2Vec(list_of_sent,min_count=5,size=100, workers=8)
         #word2vec model with 300 dim vector
         w2v model 300=gensim.models.Word2Vec(list_of_sent,min_count=5,size=300, workers=8)
In [27]: #saving to disk
         pickle.dump(w2v_model_50,open('w2v_model_svm_50.p','wb'))
         pickle.dump(w2v_model_100,open('w2v_model_svm_100.p','wb'))
         pickle.dump(w2v_model_300,open('w2v_model_svm_300.p','wb'))
```

```
In [21]: #loading from disk
         w2v_model_100 = pickle.load(open('w2v_model_svm_100.p','rb'))
         w2v_model_50 = pickle.load(open('w2v_model_svm_50.p','rb'))
         w2v_model_300 = pickle.load(open('w2v_model_svm_300.p','rb'))
Avg Word2Vec
In [16]: # the avg-w2v for each sentence/review is stored in this list
         def avg_w2v(list_of_sent,model,d):
             111
             Returns average of word vectors for
             each sentance with dimension of model given
             sent_vectors = []
             for sent in list_of_sent: # for each review/sentence
                 doc = [word for word in sent if word in model.wv.vocab]
                     sent_vec = np.mean(model.wv[doc],axis=0)
                 else:
                     sent_vec = np.zeros(d)
                 sent_vectors.append(sent_vec)
             return sent_vectors
In [17]: list_of_sent_train=[]
         for sent in train_df.final_text.values:
             list_of_sent_train.append(sent.split())
In [18]: #avg word2vec for
         sent_vector_avgw2v_300 = avg_w2v(list_of_sent_train,w2v_model_300,300)
         #stacking columns
         train_avgw2v_300 = np.hstack((sent_vector_avgw2v_300,
                     train_df[['HelpfulnessNumerator','HelpfulnessDenominator','Score']]))
         column = list(range(0,300))
         column.extend(['HelpfulnessNumerator','HelpfulnessDenominator','Score'])
         train_df_avgw2v_300 = pd.DataFrame(train_avgw2v_300,columns=column)
In [19]: #CountVectorizer for BoW
         X_train = train_df_avgw2v_300.iloc[:round(train_df.shape[0]*0.70),:]
         X_test_cv = train_df_avgw2v_300.iloc[round(train_df.shape[0]*0.70):,:]
         scale = StandardScaler()
         X_train_sc = scale.fit_transform(X_train.drop('Score',axis=1))
         X_test_cv_sc = scale.transform(X_test_cv.drop('Score',axis=1))
In [19]: for i in ParameterGrid({'C':[0.001,0.01,0.1,1,5],
                             'gamma': [0.001,0.008,0.01,0.1,0.5,1,10]}):
             model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
             model.fit(X_train_sc,X_train.Score)
             train_score = model.score(X_train_sc,X_train.Score)
             test_score = model.score(X_test_cv_sc,X_test_cv.Score)
```

```
'Test Score',test_score)
C 0.001 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.001 Train Score 0.9078231292517007 Test Score 0.8917460317460317
C 0.1 Gamma 0.008 Train Score 0.8776190476190476 Test Score 0.8468253968253968
C 0.1 Gamma 0.01 Train Score 0.8648299319727891 Test Score 0.8352380952380952
C 0.1 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 1 Gamma 0.001 Train Score 0.9448299319727891 Test Score 0.9274603174603174
C 1 Gamma 0.008 Train Score 0.98578231292517 Test Score 0.9068253968253969
C 1 Gamma 0.01 Train Score 0.9902721088435374 Test Score 0.8960317460317461
C 1 Gamma 0.1 Train Score 1.0 Test Score 0.8253968253968254
C 1 Gamma 0.5 Train Score 1.0 Test Score 0.8253968253968254
C 1 Gamma 1 Train Score 1.0 Test Score 0.8253968253968254
C 1 Gamma 10 Train Score 1.0 Test Score 0.8253968253968254
C 5 Gamma 0.001 Train Score 0.9637414965986395 Test Score 0.936031746031746
C 5 Gamma 0.008 Train Score 1.0 Test Score 0.9119047619047619
C 5 Gamma 0.01 Train Score 1.0 Test Score 0.9014285714285715
C 5 Gamma 0.1 Train Score 1.0 Test Score 0.8253968253968254
C 5 Gamma 0.5 Train Score 1.0 Test Score 0.8253968253968254
C 5 Gamma 1 Train Score 1.0 Test Score 0.8253968253968254
C 5 Gamma 10 Train Score 1.0 Test Score 0.8253968253968254
```

print('C',i['C'],'Gamma',i['gamma'],'Train Score',train_score,

We can observe that for low values of c we ave ig bias in model and for high values of c we are overfitting to the train data. and for some values of gamma(0.001) and C(1-10) the cv score are better than the others. we do have some generalization error if we are regularizing max also so my be with moderate High C wit low gamma be better for this data.

```
model_grid_avgw2v = GridSearchCV(make_pipeline(StandardScaler(),
                                                   SVC()),
                                   param_grid={'svc_C': c,'svc_gamma':gamma},
                                cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_grid_avgw2v.fit(train_df_avgw2v_300.drop('Score',axis=1),train_df_avgw2v_300.Sc
In [39]: dict_scores = []
        idx = 0
        for i in model_grid_avgw2v.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc_C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_grid_avgw2v.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                      'Test std', 'Train score'])
In [41]: scores_df.sort_values('Test_score', ascending=False).head(10)
Out [41]:
                       C Test_score Test_std Train_score
              gamma
        52 0.00100 10.0
                             0.933630 0.006818
                                                   0.978743
        51 0.00095 10.0
                          0.933578 0.006756
                                                   0.977413
        50 0.00080 10.0 0.932897 0.006800
                                                   0.973149
        45 0.00100 7.0 0.932216 0.006664
                                                   0.973193
        43 0.00080 7.0 0.932006 0.007466
                                                   0.968160
        44 0.00095 7.0 0.931797 0.007015
                                                  0.971904
        38 0.00100 5.0 0.931797 0.007716
                                                   0.968227
        37 0.00095 5.0 0.931744 0.007464
                                                   0.967187
        49 0.00050 10.0 0.931640 0.006226
                                                   0.962913
        39 0.00300
                     5.0 0.931535 0.004997
                                                   0.993881
In [50]: c = [10,20,30,40,50,60,70,80,90,100]
        gamma = [0.00095, 0.001]
        model_grid_avgw2v2 = GridSearchCV(make_pipeline(StandardScaler(),
                                                   SVC()),
                                    param_grid={'svc__C': c,'svc__gamma':gamma},
                                    cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_grid_avgw2v2.fit(train_df_avgw2v_300.drop('Score',axis=1),train_df_avgw2v_300.S
In [51]: dict_scores = []
        idx = 0
        for i in model_grid_avgw2v2.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc__C'])
            dict_score.append(i[1])
```

```
dict_score.append(i[2].std())
            dict_score.append(model_grid_avgw2v2.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
             idx = idx + 1
        scores_df2 = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                       'Test_std', 'Train_score'])
In [53]: scores_df2.sort_values('Test_score',ascending=False).head(10)
Out [53]:
                       Test_score Test_std Train_score
             gamma
        1 0.00100 10
                          0.933630 0.006818
                                                 0.978743
        0 0.00095 10
                          0.933578 0.006756
                                                 0.977413
        3 0.00100 20
                          0.932111 0.006485
                                                0.987762
        2 0.00095 20
                          0.931640 0.007013
                                                0.986663
        5 0.00100 30
                          0.931221 0.006858
                                                0.992037
        4 0.00095 30
                          0.931116 0.006455
                                                0.991255
        6 0.00095 40 0.930906 0.006735
                                                0.993642
        7 0.00100 40
                          0.930592 0.007110
                                                0.994460
        8 0.00095 50
                          0.930016 0.007068
                                                 0.995593
        9 0.00100 50
                          0.929387 0.006629
                                                 0.996349
In [76]: from scipy.stats import uniform
        model_random_avgw2v = RandomizedSearchCV(make_pipeline(StandardScaler(),SVC()),
                             param_distributions={'svc__C': uniform(loc=0,scale=13),
                             'svc__gamma':uniform(loc=0.0008,scale=0.004)},n_iter=25,
                                    cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_random_avgw2v.fit(train_df_avgw2v_300.drop('Score',axis=1),train_df_avgw2v_300.drop('Score')
In [77]: dict_scores = []
        idx = 0
        for i in model_random_avgw2v.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc__C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_random_avgw2v.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df1 = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                       'Test_std', 'Train_score'])
In [78]: scores_df1.sort_values('Test_score',ascending=False).head(10)
Out [78]:
               gamma
                              C Test_score Test_std Train_score
            0.001337
                       7.135116
                                   0.932792 0.006643
                                                          0.980713
        21 0.001729
                       6.847431
                                   0.932740 0.006190
                                                          0.986392
            0.001961
                       4.655458 0.932635 0.006548
                                                          0.983589
        5
        19 0.001799
                       6.445765 0.932583 0.006304
                                                         0.986443
```

```
22 0.001156 12.931930
                        0.932478 0.006379
                                             0.985614
13 0.001583 10.511907
                        0.932373 0.006640
                                             0.990104
1
   0.002504 3.154084
                        0.932216 0.006163
                                             0.983583
12 0.002473 5.143757
                        0.932163 0.006208
                                             0.990524
18 0.001417 12.460012
                        0.932111 0.006425
                                             0.989660
16 0.002093 2.559956
                        0.931954 0.006908
                                             0.975369
```

Best cv score for 10 fold cv got at gamma = 0.00100, C = 10 and mean cv score is 0.933630

```
In [82]: #testscore
         list_of_sent_train=[]
         for sent in train_df.final_text.values:
             list_of_sent_train.append(sent.split())
         #avg word2vec for
         sent_vector_avgw2v_300 = avg_w2v(list_of_sent_train,w2v_model 300,300)
         #stacking columns
         train_avgw2v_300 = np.hstack((sent_vector_avgw2v_300,
                     train_df[['HelpfulnessNumerator','HelpfulnessDenominator','Score']]))
         column = list(range(0,300))
         column.extend(['HelpfulnessNumerator','HelpfulnessDenominator','Score'])
         train_df_avgw2v_300 = pd.DataFrame(train_avgw2v_300,columns=column)
         list_of_sent_test=[]
         for sent in test_df.final_text.values:
             list_of_sent_test.append(sent.split())
         #avg word2vec for
         sent_vector_avgw2v_300_test = avg_w2v(list_of_sent_test,w2v_model_300,300)
         #stacking columns
         test_avgw2v_300 = np.hstack((sent_vector_avgw2v_300 test,
                     test_df[['HelpfulnessNumerator','HelpfulnessDenominator','Score']]))
         column = list(range(0,300))
         column.extend(['HelpfulnessNumerator','HelpfulnessDenominator','Score'])
         test_df_avgw2v_300 = pd.DataFrame(test_avgw2v_300,columns=column)
         scale = StandardScaler()
         X_train_sc = scale.fit_transform(train_df_avgw2v_300.drop('Score',axis=1))
         X test cv sc = scale.transform(test df avgw2v 300.drop('Score',axis=1))
         model = SVC(C=10,kernel='rbf',gamma=0.00100)
         model.fit(X train sc,train df.Score)
         #Predicting training data
         train list = model.predict(X train sc)
         #Accuracy score
         score_train = accuracy_score(train_df.Score,train_list)
         #predict test cv
```

```
#Accuracy score
         score_test = accuracy_score(test_df.Score,test_list)
         #precision
         #precision
         test_precision = precision_score(test_df.Score,test_list)
         test_recall = recall_score(test_df.Score,test_list)
         #confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print('C' ,10,'gamma',0.00100)
         print('Train Score', score_train)
         print('Test Score',score_test)
         print('Test Precision',test_precision)
         print('Test Recall',test_recall)
         print('Test ConfusionMatrix',confusion_matrix_test)
         print('No of support vectors for each class',model.n_support_)
C 10 gamma 0.001
Train Score 0.9725714285714285
Test Score 0.93322222222222
Test Precision 0.9459741615555266
Test Recall 0.9748520710059172
Test ConfusionMatrix [[1150 414]
 [ 187 7249]]
No of support vectors for each class [1693 2078]
  SGD Classifier
In [27]: for i in ParameterGrid({'alpha':[0.00005,0.00008,0.0001,0.00012],
                             'll_ratio': [0,0.03,0.05,0.08,0.1,0.15,0.25,0.35,
                                            0.45, 0.55, 0.65, 0.75, 0.85, 0.95,
                             'penality':['11','12','elasticnet']}):
             model = SGDClassifier(penalty=i['penality'],alpha=i['alpha'],11_ratio=i['11_ratio
             model.fit(X_train_sc,X_train.Score)
             train_score = model.score(X_train_sc,X_train.Score)
             test_score = model.score(X_test_cv_sc,X_test_cv.Score)
             print('Alpha',i['alpha'],'l1_ratio',i['l1_ratio'],'Penality',i['penality'],
                   'Train Score', train_score, 'Test Score', test_score)
Alpha 5e-05 11_ratio 0 Penality 11 Train Score 0.9415646258503402 Test Score 0.929365079365079
```

test_list = model.predict(X_test_cv_sc)

Alpha 5e-05 l1_ratio 0 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936508 Alpha 5e-05 l1_ratio 0 Penality elasticnet Train Score 0.9393197278911565 Test Score 0.92793650793650 Alpha 5e-05 l1_ratio 0.03 Penality l1 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 l1_ratio 0.03 Penality l2 Train Score 0.9393197278911565 Test Score 0.92793650793650 Alpha 5e-05 l1_ratio 0.03 Penality elasticnet Train Score 0.9377551020408164 Test Score 0.92500 Alpha 5e-05 l1_ratio 0.05 Penality l1 Train Score 0.9415646258503402 Test Score 0.9293650793650

Alpha 5e-05 11_ratio 0.05 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1_ratio 0.05 Penality elasticnet Train Score 0.9386394557823129 Test Score 0.9258 Alpha 5e-05 11 ratio 0.08 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11_ratio 0.08 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1 ratio 0.08 Penality elasticnet Train Score 0.9385034013605442 Test Score 0.9255 Alpha 5e-05 11_ratio 0.1 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11 ratio 0.1 Penality 12 Train Score 0.9393197278911565 Test Score 0.92793650793650 Alpha 5e-05 11_ratio 0.1 Penality elasticnet Train Score 0.9387074829931973 Test Score 0.92777 Alpha 5e-05 11 ratio 0.15 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11_ratio 0.15 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1 ratio 0.15 Penality elasticnet Train Score 0.9374829931972789 Test Score 0.9230 Alpha 5e-05 11 ratio 0.25 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11_ratio 0.25 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1 ratio 0.25 Penality elasticnet Train Score 0.939047619047619 Test Score 0.92619 Alpha 5e-05 11_ratio 0.35 Penality 11 Train Score 0.9415646258503402 Test Score 0.929365079365 Alpha 5e-05 11 ratio 0.35 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1_ratio 0.35 Penality elasticnet Train Score 0.9387074829931973 Test Score 0.9265 Alpha 5e-05 11 ratio 0.45 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11_ratio 0.45 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1 ratio 0.45 Penality elasticnet Train Score 0.9392517006802721 Test Score 0.9273 Alpha 5e-05 11_ratio 0.55 Penality 11 Train Score 0.9415646258503402 Test Score 0.929365079365 Alpha 5e-05 11 ratio 0.55 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 11_ratio 0.55 Penality elasticnet Train Score 0.941156462585034 Test Score 0.92984 Alpha 5e-05 l1_ratio 0.65 Penality 11 Train Score 0.9415646258503402 Test Score 0.929365079365 Alpha 5e-05 11_ratio 0.65 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 11 ratio 0.65 Penality elasticnet Train Score 0.9406802721088435 Test Score 0.9263 Alpha 5e-05 11 ratio 0.75 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11_ratio 0.75 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1 ratio 0.75 Penality elasticnet Train Score 0.9422448979591836 Test Score 0.9292 Alpha 5e-05 11_ratio 0.85 Penality 11 Train Score 0.9415646258503402 Test Score 0.929365079365 Alpha 5e-05 11 ratio 0.85 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 11_ratio 0.85 Penality elasticnet Train Score 0.9417006802721088 Test Score 0.9290 Alpha 5e-05 11 ratio 0.95 Penality 11 Train Score 0.9415646258503402 Test Score 0.9293650793650 Alpha 5e-05 11_ratio 0.95 Penality 12 Train Score 0.9393197278911565 Test Score 0.927936507936 Alpha 5e-05 l1 ratio 0.95 Penality elasticnet Train Score 0.943469387755102 Test Score 0.92746 Alpha 8e-05 11_ratio 0 Penality 11 Train Score 0.9438095238095238 Test Score 0.928571428571428 Alpha 8e-05 11 ratio 0 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603174 Alpha 8e-05 11_ratio 0 Penality elasticnet Train Score 0.9384353741496598 Test Score 0.9246031 Alpha 8e-05 11_ratio 0.03 Penality 11 Train Score 0.9438095238095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.03 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 l1_ratio 0.03 Penality elasticnet Train Score 0.937687074829932 Test Score 0.92365 Alpha 8e-05 11 ratio 0.05 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.05 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11 ratio 0.05 Penality elasticnet Train Score 0.9385034013605442 Test Score 0.9246 Alpha 8e-05 11_ratio 0.08 Penality 11 Train Score 0.9438095238095238 Test Score 0.9285714285714 Alpha 8e-05 11 ratio 0.08 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11_ratio 0.08 Penality elasticnet Train Score 0.9373469387755102 Test Score 0.9233 Alpha 8e-05 11 ratio 0.1 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11 ratio 0.1 Penality 12 Train Score 0.9384353741496598 Test Score 0.9246031746031 Alpha 8e-05 11_ratio 0.1 Penality elasticnet Train Score 0.9389115646258503 Test Score 0.92396 Alpha 8e-05 11 ratio 0.15 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.15 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 l1 ratio 0.15 Penality elasticnet Train Score 0.9386394557823129 Test Score 0.9239 Alpha 8e-05 11_ratio 0.25 Penality 11 Train Score 0.9438095238095238 Test Score 0.9285714285714 Alpha 8e-05 11 ratio 0.25 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11_ratio 0.25 Penality elasticnet Train Score 0.939047619047619 Test Score 0.92571 Alpha 8e-05 11 ratio 0.35 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.35 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11 ratio 0.35 Penality elasticnet Train Score 0.9385714285714286 Test Score 0.9255 Alpha 8e-05 11 ratio 0.45 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.45 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11 ratio 0.45 Penality elasticnet Train Score 0.9403401360544218 Test Score 0.9265 Alpha 8e-05 11_ratio 0.55 Penality 11 Train Score 0.9438095238095238 Test Score 0.9285714285714 Alpha 8e-05 11 ratio 0.55 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11_ratio 0.55 Penality elasticnet Train Score 0.9396598639455782 Test Score 0.9269 Alpha 8e-05 11 ratio 0.65 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.65 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 l1 ratio 0.65 Penality elasticnet Train Score 0.9417006802721088 Test Score 0.9280 Alpha 8e-05 11_ratio 0.75 Penality 11 Train Score 0.9438095238095238 Test Score 0.9285714285714 Alpha 8e-05 11 ratio 0.75 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11_ratio 0.75 Penality elasticnet Train Score 0.9415646258503402 Test Score 0.92904 Alpha 8e-05 11 ratio 0.85 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.85 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 11 ratio 0.85 Penality elasticnet Train Score 0.9398639455782313 Test Score 0.9280 Alpha 8e-05 11 ratio 0.95 Penality 11 Train Score 0.9438095238 Test Score 0.9285714285714 Alpha 8e-05 11_ratio 0.95 Penality 12 Train Score 0.9384353741496598 Test Score 0.924603174603 Alpha 8e-05 l1 ratio 0.95 Penality elasticnet Train Score 0.9425850340136055 Test Score 0.9273 Alpha 0.0001 11_ratio 0 Penality 11 Train Score 0.9437414965986395 Test Score 0.927936507936507 Alpha 0.0001 11 ratio 0.03 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11_ratio 0.03 Penality 12 Train Score 0.9366666666666666 Test Score 0.92285714285 Alpha 0.0001 11 ratio 0.03 Penality elasticnet Train Score 0.9374829931972789 Test Score 0.922 Alpha 0.0001 11_ratio 0.05 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11 ratio 0.05 Penality 12 Train Score 0.936666666666666 Test Score 0.92285714285 Alpha 0.0001 l1_ratio 0.05 Penality elasticnet Train Score 0.9378231292517006 Test Score 0.924 Alpha 0.0001 11_ratio 0.08 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11_ratio 0.08 Penality 12 Train Score 0.936666666666666 Test Score 0.92285714285 Alpha 0.0001 l1_ratio 0.08 Penality elasticnet Train Score 0.9373469387755102 Test Score 0.923 Alpha 0.0001 11 ratio 0.1 Penality 11 Train Score 0.9437414965986395 Test Score 0.927936507936 Alpha 0.0001 11_ratio 0.1 Penality 12 Train Score 0.936666666666666 Test Score 0.922857142857 Alpha 0.0001 11 ratio 0.1 Penality elasticnet Train Score 0.9380272108843537 Test Score 0.9236 Alpha 0.0001 11_ratio 0.15 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11 ratio 0.15 Penality 12 Train Score 0.93666666666666666666 Test Score 0.92285714285 Alpha 0.0001 11_ratio 0.15 Penality elasticnet Train Score 0.9380272108843537 Test Score 0.923 Alpha 0.0001 11 ratio 0.25 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11_ratio 0.25 Penality 12 Train Score 0.936666666666666 Test Score 0.92285714285 Alpha 0.0001 11_ratio 0.25 Penality elasticnet Train Score 0.9382993197278912 Test Score 0.924 Alpha 0.0001 11 ratio 0.35 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11 ratio 0.35 Penality elasticnet Train Score 0.9391836734693878 Test Score 0.923 Alpha 0.0001 11_ratio 0.45 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11 ratio 0.45 Penality 12 Train Score 0.936666666666666 Test Score 0.92285714285 Alpha 0.0001 11_ratio 0.45 Penality elasticnet Train Score 0.940204081632653 Test Score 0.9255 Alpha 0.0001 11 ratio 0.55 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11_ratio 0.55 Penality 12 Train Score 0.9366666666666666 Test Score 0.92285714285 Alpha 0.0001 11 ratio 0.55 Penality elasticnet Train Score 0.9401360544217687 Test Score 0.927 Alpha 0.0001 11 ratio 0.65 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11_ratio 0.65 Penality 12 Train Score 0.936666666666666 Test Score 0.92285714285 Alpha 0.0001 11 ratio 0.65 Penality elasticnet Train Score 0.9406802721088435 Test Score 0.926 Alpha 0.0001 11_ratio 0.75 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11 ratio 0.75 Penality 12 Train Score 0.9366666666666666666 Test Score 0.92285714285 Alpha 0.0001 11_ratio 0.75 Penality elasticnet Train Score 0.9400680272108843 Test Score 0.9279 Alpha 0.0001 11 ratio 0.85 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11_ratio 0.85 Penality 12 Train Score 0.9366666666666666 Test Score 0.92285714285 Alpha 0.0001 11 ratio 0.85 Penality elasticnet Train Score 0.9415646258503402 Test Score 0.927 Alpha 0.0001 11 ratio 0.95 Penality 11 Train Score 0.9437414965986395 Test Score 0.92793650793 Alpha 0.0001 11 ratio 0.95 Penality 12 Train Score 0.936666666666666 Test Score 0.92285714285 Alpha 0.0001 11_ratio 0.95 Penality elasticnet Train Score 0.9410884353741497 Test Score 0.926 Alpha 0.00012 11_ratio 0 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873015 Alpha 0.00012 11_ratio 0 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539682 Alpha 0.00012 11_ratio 0 Penality elasticnet Train Score 0.9385714285714286 Test Score 0.92396 Alpha 0.00012 11_ratio 0.03 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873 Alpha 0.00012 11_ratio 0.03 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539 Alpha 0.00012 11 ratio 0.03 Penality elasticnet Train Score 0.9372789115646258 Test Score 0.92 Alpha 0.00012 11_ratio 0.05 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873 Alpha 0.00012 11 ratio 0.05 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539 Alpha 0.00012 11_ratio 0.05 Penality elasticnet Train Score 0.9374149659863945 Test Score 0.92 Alpha 0.00012 11_ratio 0.08 Penality 11 Train Score 0.9426530612244898 Test Score 0.92730158730 Alpha 0.00012 11_ratio 0.08 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539 Alpha 0.00012 11 ratio 0.08 Penality elasticnet Train Score 0.9379591836734694 Test Score 0.92 Alpha 0.00012 11_ratio 0.1 Penality 11 Train Score 0.9426530612244898 Test Score 0.92730158730 Alpha 0.00012 11 ratio 0.1 Penality 12 Train Score 0.9385714285714286 Test Score 0.92396825396 Alpha 0.00012 11_ratio 0.1 Penality elasticnet Train Score 0.9379591836734694 Test Score 0.923 Alpha 0.00012 11_ratio 0.15 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873 Alpha 0.00012 11_ratio 0.15 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539 Alpha 0.00012 l1_ratio 0.15 Penality elasticnet Train Score 0.9386394557823129 Test Score 0.92 Alpha 0.00012 11_ratio 0.25 Penality 11 Train Score 0.9426530612244898 Test Score 0.92730158730 Alpha 0.00012 11_ratio 0.25 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539 Alpha 0.00012 11 ratio 0.25 Penality elasticnet Train Score 0.9379591836734694 Test Score 0.92 Alpha 0.00012 11_ratio 0.35 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873 Alpha 0.00012 11_ratio 0.35 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539 Alpha 0.00012 11_ratio 0.35 Penality elasticnet Train Score 0.9382993197278912 Test Score 0.926 Alpha 0.00012 11_ratio 0.45 Penality 11 Train Score 0.9426530612244898 Test Score 0.92730158730

```
Alpha 0.00012 11_ratio 0.45 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539
Alpha 0.00012 11_ratio 0.45 Penality elasticnet Train Score 0.9382993197278912 Test Score 0.92
Alpha 0.00012 11_ratio 0.55 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873
Alpha 0.00012 11_ratio 0.55 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539
Alpha 0.00012 l1_ratio 0.55 Penality elasticnet Train Score 0.9393197278911565 Test Score 0.92
Alpha 0.00012 11_ratio 0.65 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873
Alpha 0.00012 11_ratio 0.65 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539
Alpha 0.00012 11_ratio 0.65 Penality elasticnet Train Score 0.9402721088435374 Test Score 0.92
Alpha 0.00012 11_ratio 0.75 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873
Alpha 0.00012 11_ratio 0.75 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539
Alpha 0.00012 11 ratio 0.75 Penality elasticnet Train Score 0.9406122448979591 Test Score 0.926
Alpha 0.00012 11_ratio 0.85 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873
Alpha 0.00012 11_ratio 0.85 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539
Alpha 0.00012 11_ratio 0.85 Penality elasticnet Train Score 0.9415646258503402 Test Score 0.92
Alpha 0.00012 11_ratio 0.95 Penality 11 Train Score 0.9426530612244898 Test Score 0.9273015873
Alpha 0.00012 11_ratio 0.95 Penality 12 Train Score 0.9385714285714286 Test Score 0.9239682539
Alpha 0.00012 11_ratio 0.95 Penality elasticnet Train Score 0.9427210884353742 Test Score 0.92
In [44]: param_grid={'sgdclassifier__penalty':['11','12','elasticnet'],
         'sgdclassifier__alpha': [0.00003,0.00005,0.00007,0.00008,0.0001,
                                             0.00012,0.00018,0.00023],
         'sgdclassifier__l1_ratio':[0,0.03,0.05,0.08,0.1,0.15,0.25,0.35,
                                        0.45,0.55,0.65,0.75,0.85,0.95]}
         model_grid_avgw2v = GridSearchCV(make_pipeline(StandardScaler(),
                                                 SGDClassifier(n_jobs=-1)),
                                                    param_grid=param_grid,
                                          cv=TimeSeriesSplit(n_splits=10),
                                                               n_jobs=-1
         model_grid_avgw2v.fit(train_df_avgw2v_300.drop('Score',axis=1),train_df_avgw2v_300.Score')
In [48]: dict_scores = []
         idx = 0
         for i in model_grid_avgw2v.grid_scores_:
             dict_score = []
             dict_score.append(i[0]['sgdclassifier__alpha'])
             dict_score.append(i[0]['sgdclassifier__l1_ratio'])
             dict_score.append(i[0]['sgdclassifier__penalty'])
             dict_score.append(i[1])
             dict_score.append(i[2].std())
             dict_score.append(model_grid_avgw2v.cv_results_['mean_train_score'][idx])
             dict_scores.append(dict_score)
             idx = idx + 1
         scores_df = pd.DataFrame(dict_scores,columns=['alpha','11_ratio','penalty','Test_score
                                                        'Test_std', 'Train_score'])
In [49]: scores_df.sort_values('Test_score',ascending=False).head(10)
Out [49]:
                alpha l1_ratio
                                    penalty Test_score Test_std Train_score
         228 0.00012
                           0.25
                                               0.925406 0.005465
                                                                      0.949701
                                         11
```

```
231 0.00012
                 0.35
                                     0.925354 0.008048
                                                           0.949089
                               11
251 0.00012
                                     0.925039 0.007527
                 0.95
                       elasticnet
                                                           0.948532
213 0.00012
                 0.03
                                     0.924987 0.006550
                                                           0.948186
                               11
66
    0.00005
                 0.45
                               11
                                     0.924935 0.005871
                                                           0.949894
234 0.00012
                 0.45
                               11
                                     0.924830 0.006851
                                                           0.948310
186 0.00010
                 0.25
                                     0.924725 0.006626
                               11
                                                           0.948830
138 0.00008
                 0.10
                               11
                                     0.924673 0.004765
                                                           0.950189
177 0.00010
                 0.08
                               11
                                     0.924515 0.005506
                                                           0.948358
    0.00005
                 0.85
                               12
                                     0.924463 0.005463
                                                           0.943389
79
```

Got best cv mean score at alpha = 0.00012,l1_ratio= 0.25 penlty = l1 and mean cv score is 0.925406

```
In [68]: print('With SGD')
         #testscore
         list_of_sent_train=[]
         for sent in train_df.final_text.values:
             list_of_sent_train.append(sent.split())
         #avg word2vec for
         sent_vector_avgw2v_300 = avg_w2v(list_of_sent_train,w2v_model_300,300)
         #stacking columns
         train_avgw2v_300 = np.hstack((sent_vector_avgw2v_300,
                     train df[['HelpfulnessNumerator','HelpfulnessDenominator','Score']]))
         column = list(range(0,300))
         column.extend(['HelpfulnessNumerator','HelpfulnessDenominator','Score'])
         train df avgw2v 300 = pd.DataFrame(train avgw2v 300,columns=column)
         list_of_sent_test=[]
         for sent in test df.final text.values:
             list_of_sent_test.append(sent.split())
         #avg word2vec for
         sent_vector_avgw2v_300_test = avg_w2v(list_of_sent_test,w2v_model_300,300)
         #stacking columns
         test_avgw2v_300 = np.hstack((sent_vector_avgw2v_300_test,
                     test_df[['HelpfulnessNumerator','HelpfulnessDenominator','Score']]))
         column = list(range(0,300))
         column.extend(['HelpfulnessNumerator','HelpfulnessDenominator','Score'])
         test_df_avgw2v_300 = pd.DataFrame(test_avgw2v_300,columns=column)
         scale = StandardScaler()
         X train sc = scale.fit transform(train df avgw2v 300.drop('Score',axis=1))
         X_test_cv_sc = scale.transform(test_df_avgw2v_300.drop('Score',axis=1))
         model = SGDClassifier(penalty='l1',alpha=0.00012,l1_ratio=0.25,random_state=25)
         model.fit(X_train_sc,train_df.Score)
```

```
#Predicting training data
         train_list = model.predict(X_train_sc)
         #Accuracy score
         score_train = accuracy_score(train_df.Score,train_list)
         #predict test cv
         test_list = model.predict(X_test_cv_sc)
         #Accuracy score
         score_test = accuracy_score(test_df.Score,test_list)
         #precision
         #precision
         test_precision = precision_score(test_df.Score,test_list)
         #recall
         test_recall = recall_score(test_df.Score,test_list)
         #confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print("penalty='l1',alpha=0.00012,l1_ratio=0.25")
         print('Train Score', score_train)
         print('Test Score',score_test)
         print('Test Precision',test_precision)
         print('Test Recall',test recall)
         print('Test ConfusionMatrix',confusion_matrix_test)
With SGD
penalty='11',alpha=0.00012,l1_ratio=0.25
Train Score 0.9392380952380952
Test Score 0.92877777777778
Test Precision 0.9474516001580403
Test Recall 0.9674556213017751
Test ConfusionMatrix [[1165 399]
 [ 242 7194]]
0.0.5 Tf-Idf Word2Vec
In [22]: from sklearn.base import BaseEstimator, TransformerMixin
         class TfidfWeightedWord2Vec(BaseEstimator, TransformerMixin):
             111
             Class for Tfidf Weighted Word2Vec Calculations
             def __init__(self, word2vec):
                 self.word2vec = word2vec
                 self.word2weight = None
                 self.dim = word2vec.vector_size
                 self.tfidf = None
             def fit(self, X, y=None):
                 tfidf = TfidfVectorizer()
```

```
tfidf.fit(X[:,0])
    self.tfidf = tfidf
    #print(self.word2vec.wv.vocab.keys())
    return self
def tf_idf_W2V(self,feature_names,tf_idf_trans_arr,list_of_sent):
    tfidf weighted word2vec calculation
    import operator
    dict_tfidf = {k: v for v, k in enumerate(feature_names)}
    sent_vectors = []
    i = 0
    for sent in list_of_sent: # for each review/sentence
        doc = [word for word in sent if word in self.word2vec.wv.vocab.keys()]
        if doc:
            #itemgetter
            f = operator.itemgetter(*doc)
            try:
                #itemgetter from dict
                final = f(dict tfidf)
                final = tf idf trans arr[i,final]
                #converting to dense
                final = final.toarray()
                #converting to diagnol matrix for multiplication
                final= np.diag(final[0])
                sent_vec = np.dot(final,np.array(self.word2vec.wv[doc]))
                #tfidf weighted word to vec
                sent_vec = np.sum(sent_vec,axis=0) / np.sum(final)
            except:
                sent_vec = np.zeros(self.dim)
        else:
            sent_vec = np.zeros(self.dim)
        sent_vectors.append(sent_vec)
        i = i+1
    return sent_vectors
def transform(self, X):
    #transform data
    tf_idf_trans_arr = self.tfidf.transform(X[:,0])
    feature_names = self.tfidf.get_feature_names()
    list_of_sent = []
    for sent in X[:,0]:
        list_of_sent.append(sent.split())
    temp_vec = self.tf_idf_W2V(feature_names,tf_idf_trans_arr,list_of_sent)
    temp_vec= np.hstack((temp_vec,X[:,[1,2]]))
    return temp_vec
```

```
In [23]: # For simple cv
        #Train data
        X_train = train_df.iloc[:round(train_df.shape[0]*0.70),:]
        X_test_cv = train_df.iloc[round(train_df.shape[0]*0.70):,:]
        #transforming to tfidf weighted word2vec
        tfidfvect w2v = TfidfWeightedWord2Vec(w2v model 300)
        tfidfvect w2v.fit(X train[['final text', 'HelpfulnessNumerator',
                                   'HelpfulnessDenominator']].values)
        X_train_tfw2v = tfidfvect_w2v.transform(X_train[['final_text',
                        'HelpfulnessNumerator', 'HelpfulnessDenominator']].values)
        X_cv_tfw2v = tfidfvect_w2v.transform(X_test_cv[['final_text',
                         'HelpfulnessNumerator', 'HelpfulnessDenominator']].values)
In [24]: #scaling the data
        scale = StandardScaler()
        X_train_sc = scale.fit_transform(X_train_tfw2v)
        X_test_cv_sc = scale.transform(X_cv_tfw2v)
In [18]: for i in ParameterGrid(\{'C':[1,5,7],
                            'gamma': [0.001,0.008,0.01,0.1,0.5,1,10]}):
            model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
            model.fit(X_train_sc,X_train.Score)
            train_score = model.score(X_train_sc,X_train.Score)
            test_score = model.score(X_test_cv_sc,X_test_cv.Score)
            print('C',i['C'],'Gamma',i['gamma'],'Train Score',train_score,
                             'Test Score', test_score)
C 1 Gamma 0.008 Train Score 0.9641496598639456 Test Score 0.8671428571428571
C 1 Gamma 0.01 Train Score 0.9679591836734693 Test Score 0.86111111111111112
C 1 Gamma 0.1 Train Score 0.9785714285714285 Test Score 0.8268253968253968
C 1 Gamma 0.5 Train Score 0.9789795918367347 Test Score 0.8287301587301588
C 1 Gamma 1 Train Score 0.9792517006802721 Test Score 0.829047619047619
C 1 Gamma 10 Train Score 0.9814285714285714 Test Score 0.8323809523809523
C 5 Gamma 0.001 Train Score 0.9479591836734694 Test Score 0.8898412698412699
C 5 Gamma 0.008 Train Score 0.9782993197278912 Test Score 0.87
C 5 Gamma 0.01 Train Score 0.9786394557823129 Test Score 0.8653968253968254
C 5 Gamma 0.1 Train Score 0.978843537414966 Test Score 0.8274603174603175
C 5 Gamma 0.5 Train Score 0.9793877551020408 Test Score 0.8282539682539682
C 5 Gamma 1 Train Score 0.9798639455782313 Test Score 0.8304761904761905
C 5 Gamma 10 Train Score 0.9821768707482993 Test Score 0.8320634920634921
C 7 Gamma 0.001 Train Score 0.9527210884353742 Test Score 0.8895238095238095
C 7 Gamma 0.008 Train Score 0.9785714285714285 Test Score 0.8701587301587301
C 7 Gamma 0.01 Train Score 0.9785034013605443 Test Score 0.8653968253968254
C 7 Gamma 0.1 Train Score 0.9787755102040816 Test Score 0.8274603174603175
C 7 Gamma 0.5 Train Score 0.9795238095238096 Test Score 0.8284126984126984
C 7 Gamma 1 Train Score 0.980204081632653 Test Score 0.8314285714285714
C 7 Gamma 10 Train Score 0.9823129251700681 Test Score 0.8319047619047619
```

```
In [17]: for i in ParameterGrid({'C':[0.001,0.01,0.1],
                             'gamma': [0.001,0.008,0.01,0.1,0.5,1,10]}):
             model = SVC(C=i['C'],kernel='rbf',gamma=i['gamma'])
             model.fit(X_train_sc,X_train.Score)
             train score = model.score(X train sc,X train.Score)
             test_score = model.score(X_test_cv_sc,X_test_cv.Score)
             print('C',i['C'],'Gamma',i['gamma'],'Train Score',train score.
                              'Test Score', test score)
C 0.001 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.001 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.001 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.008 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.01 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 0.5 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.01 Gamma 10 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.001 Train Score 0.8851020408163265 Test Score 0.8536507936507937
C 0.1 Gamma 0.008 Train Score 0.8639455782312925 Test Score 0.8319047619047619
C 0.1 Gamma 0.01 Train Score 0.8584353741496599 Test Score 0.8273015873015873
C 0.1 Gamma 0.1 Train Score 0.8563945578231292 Test Score 0.8253968253968254
C 0.1 Gamma 0.5 Train Score 0.8571428571428571 Test Score 0.8258730158730159
C 0.1 Gamma 1 Train Score 0.8572789115646259 Test Score 0.8257142857142857
C 0.1 Gamma 10 Train Score 0.8565986394557823 Test Score 0.8257142857142857
In [20]: c = [10.5, 0.85, 1, 2.5, 5, 10, 12, 20]
         gamma = [0.00095, 0.001, 0.0013, 0.0015, 0.0024, 0.007, 0.01, 1, 10]
         model_grid_tfidfw2v = GridSearchCV(
                                     make_pipeline(TfidfWeightedWord2Vec(w2v_model_300),
                                     StandardScaler(),SVC()),
                                      param_grid={'svc_C': c,'svc_gamma':gamma},
                                     cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
         model_grid_tfidfw2v.fit(train_df[['final_text','HelpfulnessNumerator',
                                     'HelpfulnessDenominator']].values,train_df.Score)
In [22]: dict_scores = []
         idx = 0
         for i in model_grid_tfidfw2v.grid_scores_:
             dict_score = []
             dict_score.append(i[0]['svc_gamma'])
```

```
dict_score.append(i[0]['svc__C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_grid_tfidfw2v.cv_results_['mean_train_score'][idx])
            dict scores.append(dict score)
            idx = idx + 1
        scores df = pd.DataFrame(dict scores,columns=['gamma','C','Test score',
                                                       'Test_std', 'Train_score'])
In [25]: scores_df.sort_values('Test_score',ascending=False).head(10)
Out [25]:
                        С
                          Test_score Test_std Train_score
              gamma
        64 0.00100 20.0
                             0.892457 0.004307
                                                    0.972534
        63 0.00095 20.0
                             0.892404 0.004359
                                                    0.971365
        54 0.00095 12.0
                             0.892142 0.005162
                                                    0.964579
        47 0.00130 10.0
                             0.892090 0.005534
                                                    0.969587
        55 0.00100 12.0
                             0.892090 0.005008
                                                    0.965825
        2
            0.00130 10.5
                             0.892038 0.005365
                                                    0.970219
        56 0.00130 12.0 0.892038 0.005081
                                                    0.971907
        1
            0.00100 10.5
                             0.891933 0.006247
                                                    0.963971
            0.00095 10.5
                             0.891776 0.005768
        0
                                                    0.962865
        45 0.00095 10.0
                             0.891776 0.006028
                                                    0.961995
In [36]: model_random_tfidfw2v = RandomizedSearchCV(
                                make_pipeline(TfidfWeightedWord2Vec(w2v_model_300),
                                    StandardScaler(),SVC()),
                                param_distributions={'svc_C': uniform(loc=0,scale=3.5),
                                'svc_gamma':uniform(loc=0.0008,scale=0.004)},n_iter=15,
                                    cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_random_tfidfw2v.fit(train_df[['final_text','HelpfulnessNumerator',
                                        'HelpfulnessDenominator']].values,train_df.Score)
In [38]: dict_scores = []
        idx = 0
        for i in model_random_tfidfw2v.grid_scores_:
            dict score = []
            dict_score.append(i[0]['svc_gamma'])
            dict_score.append(i[0]['svc_C'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_random_tfidfw2v.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df = pd.DataFrame(dict_scores,columns=['gamma','C','Test_score',
                                                       'Test_std', 'Train_score'])
In [39]: scores_df.sort_values('Test_score', ascending=False).head(10)
Out [39]:
                             C Test_score Test_std Train_score
               gamma
            0.002625 2.978150
                                0.889314 0.006145
                                                         0.968132
```

```
11 0.002274 2.929955
                       0.889104 0.005996
                                            0.964255
10 0.003434 3.078954
                       0.887742 0.006286
                                            0.974421
4
   0.003403 2.501247
                       0.887690 0.006506
                                            0.971187
12 0.003382 3.446907
                       0.887533 0.006390
                                            0.975409
   0.002573 1.670520
                       0.887218 0.006713
6
                                            0.955952
9
   0.003608 2.015362
                       0.886695 0.006492
                                            0.968607
7
   0.003702 2.062496
                       0.886433 0.006320
                                            0.969583
14 0.003818 2.920594
                       0.886276 0.006151
                                            0.975324
   0.004238 2.525873
                       0.885961 0.006509
                                            0.975278
```

best cv score for tfidf word2vec got at gamma = 0.00100 C = 20.0 and mean cv score is 0.892457

```
In [41]: #testscore
         # For simple cv
         #transforming to tfidf weighted word2vec
         tfidfvect_w2v = TfidfWeightedWord2Vec(w2v_model_300)
         tfidfvect_w2v.fit(train_df[['final_text', 'HelpfulnessNumerator',
                                    'HelpfulnessDenominator']].values)
         X_train_tfw2v = tfidfvect_w2v.transform(train_df[['final_text',
                         'HelpfulnessNumerator', 'HelpfulnessDenominator']].values)
         X_cv_tfw2v = tfidfvect_w2v.transform(test_df[['final_text',
                          'HelpfulnessNumerator', 'HelpfulnessDenominator']].values)
         #scaling the data
         scale = StandardScaler()
         X_train_sc = scale.fit_transform(X_train_tfw2v)
         X_test_cv_sc = scale.transform(X_cv_tfw2v)
         model = SVC(C=20,kernel='rbf',gamma=0.00100)
         model.fit(X_train_sc,train_df.Score)
         #Predicting training data
         train_list = model.predict(X_train_sc)
         #Accuracy score
         score_train = accuracy_score(train_df.Score,train_list)
         #predict test cv
         test_list = model.predict(X_test_cv_sc)
         #Accuracy score
         score_test = accuracy_score(test_df.Score,test_list)
         #precision
         #precision
         test_precision = precision_score(test_df.Score,test_list)
         #recall
         test_recall = recall_score(test_df.Score,test_list)
         #confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print('C' ,20,'gamma',0.00100)
         print('Train Score', score_train)
         print('Test Score',score_test)
```

```
print('No of support vectors for each class',model.n_support_)
C 20 gamma 0.001
Train Score 0.9643809523809523
Test Score 0.890444444444445
Test Precision 0.8969719350073855
Test Recall 0.9799623453469607
Test ConfusionMatrix [[ 727 837]
 [ 149 7287]]
No of support vectors for each class [2031 2750]
  SGD Classifier
In [58]: for i in ParameterGrid({'alpha':[0.00005,0.00008,0.0001,0.00012],
                             'l1_ratio': [0,0.03,0.05,0.08,0.1,0.15,0.25,0.35,
                                            0.45, 0.55, 0.65, 0.75, 0.85, 0.95,
                             'penality':['11','12','elasticnet']}):
             model = SGDClassifier(penalty=i['penality'],alpha=i['alpha'],
                                   11_ratio=i['l1_ratio'],random_state=25)
             model.fit(X_train_sc,X_train.Score)
             train_score = model.score(X_train_sc,X_train.Score)
             test_score = model.score(X_test_cv_sc,X_test_cv.Score)
             print('Alpha',i['alpha'],'l1_ratio',i['l1_ratio'],'Penality',i['penality'],
                   'Train Score', train_score, 'Test Score', test_score)
Alpha 5e-05 11 ratio 0 Penality 11 Train Score 0.9236734693877551 Test Score 0.8922222222222222
Alpha 5e-05 11_ratio 0 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047619
Alpha 5e-05 l1_ratio 0 Penality elasticnet Train Score 0.9210884353741496 Test Score 0.8890476
Alpha 5e-05 11 ratio 0.03 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222
Alpha 5e-05 11_ratio 0.03 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047
Alpha 5e-05 l1 ratio 0.03 Penality elasticnet Train Score 0.9169387755102041 Test Score 0.8917
Alpha 5e-05 11_ratio 0.05 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222
Alpha 5e-05 11 ratio 0.05 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047
Alpha 5e-05 11_ratio 0.05 Penality elasticnet Train Score 0.9196598639455782 Test Score 0.8933
Alpha 5e-05 11_ratio 0.08 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222
Alpha 5e-05 11_ratio 0.08 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047
Alpha 5e-05 l1 ratio 0.08 Penality elasticnet Train Score 0.9186394557823129 Test Score 0.8884
Alpha 5e-05 11_ratio 0.1 Penality 11 Train Score 0.9236734693877551 Test Score 0.89222222222222
Alpha 5e-05 11_ratio 0.1 Penality 12 Train Score 0.9210884353741496 Test Score 0.8890476190476
Alpha 5e-05 11_ratio 0.1 Penality elasticnet Train Score 0.9102040816326531 Test Score 0.88650
Alpha 5e-05 11_ratio 0.15 Penality 11 Train Score 0.9236734693877551 Test Score 0.8922222222222
Alpha 5e-05 11 ratio 0.15 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047
Alpha 5e-05 l1_ratio 0.15 Penality elasticnet Train Score 0.9124489795918367 Test Score 0.8831
Alpha 5e-05 11_ratio 0.25 Penality 11 Train Score 0.9236734693877551 Test Score 0.8922222222222
```

print('Test Precision',test_precision)

print('Test ConfusionMatrix',confusion_matrix_test)

print('Test Recall',test_recall)

Alpha 5e-05 11_ratio 0.25 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047 Alpha 5e-05 l1_ratio 0.25 Penality elasticnet Train Score 0.9208163265306123 Test Score 0.8923 Alpha 5e-05 11 ratio 0.35 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222 Alpha 5e-05 11_ratio 0.35 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047 Alpha 5e-05 l1 ratio 0.35 Penality elasticnet Train Score 0.9231972789115647 Test Score 0.8953 Alpha 5e-05 11_ratio 0.45 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222 Alpha 5e-05 11 ratio 0.45 Penality 12 Train Score 0.9210884353741496 Test Score 0.8890476190476 Alpha 5e-05 11_ratio 0.45 Penality elasticnet Train Score 0.922312925170068 Test Score 0.89333 Alpha 5e-05 11 ratio 0.55 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222 Alpha 5e-05 11_ratio 0.55 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047 Alpha 5e-05 l1 ratio 0.55 Penality elasticnet Train Score 0.9128571428571428 Test Score 0.8839 Alpha 5e-05 11 ratio 0.65 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222 Alpha 5e-05 11_ratio 0.65 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047 Alpha 5e-05 l1 ratio 0.65 Penality elasticnet Train Score 0.9136734693877551 Test Score 0.8855 Alpha 5e-05 11_ratio 0.75 Penality 11 Train Score 0.9236734693877551 Test Score 0.8922222222222 Alpha 5e-05 11 ratio 0.75 Penality 12 Train Score 0.9210884353741496 Test Score 0.8890476190476 Alpha 5e-05 l1_ratio 0.75 Penality elasticnet Train Score 0.9219727891156463 Test Score 0.8926 Alpha 5e-05 11 ratio 0.85 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222 Alpha 5e-05 11_ratio 0.85 Penality 12 Train Score 0.9210884353741496 Test Score 0.889047619047 Alpha 5e-05 l1 ratio 0.85 Penality elasticnet Train Score 0.9263945578231293 Test Score 0.8974 Alpha 5e-05 11_ratio 0.95 Penality 11 Train Score 0.9236734693877551 Test Score 0.892222222222 Alpha 5e-05 11 ratio 0.95 Penality 12 Train Score 0.9210884353741496 Test Score 0.8890476190476 Alpha 5e-05 11_ratio 0.95 Penality elasticnet Train Score 0.9259863945578232 Test Score 0.8965 Alpha 8e-05 11_ratio 0 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920634 Alpha 8e-05 11_ratio 0 Penality 12 Train Score 0.9206122448979592 Test Score 0.892063492063492 Alpha 8e-05 11 ratio 0 Penality elasticnet Train Score 0.9206122448979592 Test Score 0.8920634 Alpha 8e-05 11 ratio 0.03 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920 Alpha 8e-05 11 ratio 0.03 Penality elasticnet Train Score 0.9195238095238095 Test Score 0.8931 Alpha 8e-05 11_ratio 0.05 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920 Alpha 8e-05 11 ratio 0.05 Penality 12 Train Score 0.9206122448979592 Test Score 0.892063492063 Alpha 8e-05 11_ratio 0.05 Penality elasticnet Train Score 0.9224489795918367 Test Score 0.8944 Alpha 8e-05 11 ratio 0.08 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920 Alpha 8e-05 11_ratio 0.08 Penality 12 Train Score 0.9206122448979592 Test Score 0.8920634920634 Alpha 8e-05 l1 ratio 0.08 Penality elasticnet Train Score 0.9149659863945578 Test Score 0.8860 Alpha 8e-05 11_ratio 0.1 Penality 11 Train Score 0.9240136054421769 Test Score 0.8949206349206 Alpha 8e-05 11 ratio 0.1 Penality 12 Train Score 0.9206122448979592 Test Score 0.89206349206349 Alpha 8e-05 11_ratio 0.1 Penality elasticnet Train Score 0.9220408163265306 Test Score 0.89301 Alpha 8e-05 11_ratio 0.15 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920 Alpha 8e-05 11_ratio 0.15 Penality 12 Train Score 0.9206122448979592 Test Score 0.8920634920634 Alpha 8e-05 l1_ratio 0.15 Penality elasticnet Train Score 0.9148979591836734 Test Score 0.8853 Alpha 8e-05 11 ratio 0.25 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920 Alpha 8e-05 11 ratio 0.25 Penality elasticnet Train Score 0.9210884353741496 Test Score 0.8936 Alpha 8e-05 11_ratio 0.35 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920 Alpha 8e-05 11 ratio 0.35 Penality 12 Train Score 0.9206122448979592 Test Score 0.892063492063 Alpha 8e-05 11_ratio 0.35 Penality elasticnet Train Score 0.9227210884353741 Test Score 0.8946 Alpha 8e-05 11 ratio 0.45 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920

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Alpha 8e-05 11_ratio 0.45 Penality 12 Train Score 0.9206122448979592 Test Score 0.8920634920634
Alpha 8e-05 11_ratio 0.45 Penality elasticnet Train Score 0.9160544217687074 Test Score 0.8896
Alpha 8e-05 11 ratio 0.55 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920
Alpha 8e-05 11_ratio 0.55 Penality 12 Train Score 0.9206122448979592 Test Score 0.8920634920634
Alpha 8e-05 l1 ratio 0.55 Penality elasticnet Train Score 0.9230612244897959 Test Score 0.8961
Alpha 8e-05 11_ratio 0.65 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920
Alpha 8e-05 11 ratio 0.65 Penality 12 Train Score 0.9206122448979592 Test Score 0.892063492063-
Alpha 8e-05 11_ratio 0.65 Penality elasticnet Train Score 0.9244897959183673 Test Score 0.8957
Alpha 8e-05 11 ratio 0.75 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920
Alpha 8e-05 11_ratio 0.75 Penality 12 Train Score 0.9206122448979592 Test Score 0.8920634920634
Alpha 8e-05 l1 ratio 0.75 Penality elasticnet Train Score 0.925578231292517 Test Score 0.89428
Alpha 8e-05 11 ratio 0.85 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920
Alpha 8e-05 11 ratio 0.85 Penality elasticnet Train Score 0.923265306122449 Test Score 0.89682
Alpha 8e-05 11_ratio 0.95 Penality 11 Train Score 0.9240136054421769 Test Score 0.894920634920
Alpha 8e-05 11 ratio 0.95 Penality 12 Train Score 0.9206122448979592 Test Score 0.892063492063
Alpha 8e-05 l1_ratio 0.95 Penality elasticnet Train Score 0.9178231292517007 Test Score 0.8874
Alpha 0.0001 11 ratio 0 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396825
Alpha 0.0001 11 ratio 0 Penality elasticnet Train Score 0.9231292517006803 Test Score 0.893968
Alpha 0.0001 11_ratio 0.03 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.03 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11_ratio 0.03 Penality elasticnet Train Score 0.921156462585034 Test Score 0.8931
Alpha 0.0001 11_ratio 0.05 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.05 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11 ratio 0.05 Penality elasticnet Train Score 0.921156462585034 Test Score 0.8928
Alpha 0.0001 11 ratio 0.08 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.08 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11 ratio 0.08 Penality elasticnet Train Score 0.9155102040816326 Test Score 0.886
Alpha 0.0001 11_ratio 0.1 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11 ratio 0.1 Penality 12 Train Score 0.9231292517006803 Test Score 0.893968253968
Alpha 0.0001 11_ratio 0.1 Penality elasticnet Train Score 0.9214965986394558 Test Score 0.8934
Alpha 0.0001 11 ratio 0.15 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.15 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11 ratio 0.15 Penality elasticnet Train Score 0.9210884353741496 Test Score 0.893
Alpha 0.0001 11_ratio 0.25 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11 ratio 0.25 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11_ratio 0.25 Penality elasticnet Train Score 0.9206122448979592 Test Score 0.895
Alpha 0.0001 11_ratio 0.35 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.35 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 l1_ratio 0.35 Penality elasticnet Train Score 0.9197278911564626 Test Score 0.888
Alpha 0.0001 11 ratio 0.45 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.45 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11 ratio 0.45 Penality elasticnet Train Score 0.9230612244897959 Test Score 0.893
Alpha 0.0001 11_ratio 0.55 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11 ratio 0.55 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11_ratio 0.55 Penality elasticnet Train Score 0.9242857142857143 Test Score 0.895
Alpha 0.0001 11 ratio 0.65 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
```

```
Alpha 0.0001 11_ratio 0.65 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11_ratio 0.65 Penality elasticnet Train Score 0.92421768707483 Test Score 0.89492
Alpha 0.0001 11 ratio 0.75 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.75 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11 ratio 0.75 Penality elasticnet Train Score 0.9238095238095239 Test Score 0.895
Alpha 0.0001 11_ratio 0.85 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11 ratio 0.85 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 l1_ratio 0.85 Penality elasticnet Train Score 0.9238775510204081 Test Score 0.895
Alpha 0.0001 11 ratio 0.95 Penality 11 Train Score 0.9182312925170067 Test Score 0.89
Alpha 0.0001 11_ratio 0.95 Penality 12 Train Score 0.9231292517006803 Test Score 0.89396825396
Alpha 0.0001 11 ratio 0.95 Penality elasticnet Train Score 0.9229251700680272 Test Score 0.895
Alpha 0.00012 11_ratio 0 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746031
Alpha 0.00012 11_ratio 0 Penality 12 Train Score 0.916666666666666 Test Score 0.88841269841269
Alpha 0.00012 11_ratio 0 Penality elasticnet Train Score 0.916666666666666666 Test Score 0.88841
Alpha 0.00012 11_ratio 0.03 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.03 Penality 12 Train Score 0.9166666666666666 Test Score 0.8884126984
Alpha 0.00012 11_ratio 0.03 Penality elasticnet Train Score 0.9204761904761904 Test Score 0.894
Alpha 0.00012 11_ratio 0.05 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.05 Penality 12 Train Score 0.91666666666666666 Test Score 0.8884126984
Alpha 0.00012 l1 ratio 0.05 Penality elasticnet Train Score 0.9219727891156463 Test Score 0.89
Alpha 0.00012 11_ratio 0.08 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11 ratio 0.08 Penality 12 Train Score 0.9166666666666666 Test Score 0.8884126984
Alpha 0.00012 11_ratio 0.08 Penality elasticnet Train Score 0.9227210884353741 Test Score 0.89
Alpha 0.00012 11 ratio 0.1 Penality 11 Train Score 0.9231972789115647 Test Score 0.89460317460
Alpha 0.00012 11_ratio 0.1 Penality 12 Train Score 0.9166666666666666666 Test Score 0.88841269841
Alpha 0.00012 l1_ratio 0.1 Penality elasticnet Train Score 0.9228571428571428 Test Score 0.894
Alpha 0.00012 11_ratio 0.15 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.15 Penality elasticnet Train Score 0.9222448979591836 Test Score 0.89
Alpha 0.00012 11_ratio 0.25 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11 ratio 0.25 Penality 12 Train Score 0.91666666666666666666 Test Score 0.8884126984
Alpha 0.00012 11_ratio 0.25 Penality elasticnet Train Score 0.921156462585034 Test Score 0.893
Alpha 0.00012 11_ratio 0.35 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.35 Penality 12 Train Score 0.91666666666666666 Test Score 0.8884126984
Alpha 0.00012 11 ratio 0.35 Penality elasticnet Train Score 0.924625850340136 Test Score 0.893
Alpha 0.00012 11_ratio 0.45 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11 ratio 0.45 Penality 12 Train Score 0.9166666666666666 Test Score 0.8884126984
Alpha 0.00012 11_ratio 0.45 Penality elasticnet Train Score 0.9231292517006803 Test Score 0.89
Alpha 0.00012 11_ratio 0.55 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.55 Penality 12 Train Score 0.91666666666666666 Test Score 0.8884126984
Alpha 0.00012 l1_ratio 0.55 Penality elasticnet Train Score 0.9229931972789116 Test Score 0.894
Alpha 0.00012 11_ratio 0.65 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.65 Penality elasticnet Train Score 0.9238775510204081 Test Score 0.89
Alpha 0.00012 11_ratio 0.75 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 11_ratio 0.75 Penality 12 Train Score 0.9166666666666666 Test Score 0.8884126984
Alpha 0.00012 11_ratio 0.75 Penality elasticnet Train Score 0.9242857142857143 Test Score 0.894
Alpha 0.00012 11_ratio 0.85 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
```

```
Alpha 0.00012 11_ratio 0.85 Penality 12 Train Score 0.9166666666666666 Test Score 0.8884126984
Alpha 0.00012 11_ratio 0.85 Penality elasticnet Train Score 0.9204081632653062 Test Score 0.89
Alpha 0.00012 11_ratio 0.95 Penality 11 Train Score 0.9231972789115647 Test Score 0.8946031746
Alpha 0.00012 l1_ratio 0.95 Penality elasticnet Train Score 0.9236054421768708 Test Score 0.89
In [25]: model_random_tfidfw2v = RandomizedSearchCV(
                           make_pipeline(TfidfWeightedWord2Vec(w2v_model_300),
                           StandardScaler(),SGDClassifier(n_jobs=-1)),
                           param_distributions={'sgdclassifier__penalty':['11','12'],
                            'sgdclassifier_alpha':uniform(loc=0.00001,scale=0.0049),
                            'sgdclassifier__l1_ratio':uniform(loc=0,scale=1)},n_iter=40,
                             cv=TimeSeriesSplit(n_splits=10),n_jobs=-1)
        model_random_tfidfw2v.fit(train_df[['final_text','HelpfulnessNumerator',
                                       'HelpfulnessDenominator']].values,train_df.Score)
In [27]: dict_scores = []
        idx = 0
        for i in model_random_tfidfw2v.grid_scores_:
            dict_score = []
            dict_score.append(i[0]['sgdclassifier__alpha'])
            dict_score.append(i[0]['sgdclassifier__l1_ratio'])
            dict_score.append(i[0]['sgdclassifier__penalty'])
            dict_score.append(i[1])
            dict_score.append(i[2].std())
            dict_score.append(model_random_tfidfw2v.cv_results_['mean_train_score'][idx])
            dict_scores.append(dict_score)
            idx = idx + 1
        scores_df = pd.DataFrame(dict_scores,columns=['alpha','11_ratio','penalty','Test_score
                                                     'Test_std', 'Train_score'])
In [29]: scores_df.sort_values('Test_score',ascending=False).head(10)
Out [29]:
               alpha l1_ratio penalty Test_score
                                                  Test_std Train_score
        25 0.000823 0.464483
                                         0.892981 0.008003
                                                               0.924661
                                   11
        16 0.000456 0.864083
                                   11
                                         0.892823 0.007788
                                                               0.926553
        29 0.000856 0.372678
                                   11
                                         0.891881 0.008156
                                                               0.924244
        36 0.000393 0.995435
                                   11
                                         0.891828 0.007841
                                                               0.928939
        31 0.000906 0.480030
                                   11
                                         0.891671 0.005621
                                                               0.924880
        37 0.000734 0.360432
                                   11
                                         0.891566 0.007266
                                                               0.926045
        15 0.000525 0.521241
                                   11
                                         0.891409 0.008473
                                                               0.927267
        19 0.000608 0.692168
                                   11
                                         0.891252 0.007907
                                                               0.926134
        4
            0.000154 0.842079
                                   12
                                         0.891200 0.007756
                                                               0.925755
        14 0.000231 0.884751
                                   12
                                         0.890676 0.006560
                                                               0.924619
```

Got best cv scores at alpha = 0.000823 11_ratio = 0.464483 penalty = 11 and mean cv score is 0.892981.

```
In [34]: #testscore
         #transforming to tfidf weighted word2vec
         tfidfvect_w2v = TfidfWeightedWord2Vec(w2v_model_300)
         tfidfvect_w2v.fit(train_df[['final_text', 'HelpfulnessNumerator',
                                    'HelpfulnessDenominator']].values)
         X_train_tfw2v = tfidfvect_w2v.transform(train_df[['final_text',
                         'HelpfulnessNumerator', 'HelpfulnessDenominator']].values)
         X_cv_tfw2v = tfidfvect_w2v.transform(test_df[['final_text',
                          'HelpfulnessNumerator', 'HelpfulnessDenominator']].values)
         #scaling the data
         scale = StandardScaler()
         X_train_sc = scale.fit_transform(X_train_tfw2v)
         X_test_cv_sc = scale.transform(X_cv_tfw2v)
         model = SGDClassifier(penalty='11',alpha=0.000823,11_ratio=0.464483,random_state=25)
         model.fit(X_train_sc,train_df.Score)
         #Predicting training data
         train_list = model.predict(X_train_sc)
         #Accuracy score
         score_train = accuracy_score(train_df.Score,train_list)
         #predict test cv
         test_list = model.predict(X_test_cv_sc)
         #Accuracy score
         score_test = accuracy_score(test_df.Score,test_list)
         #precision
         #precision
         test_precision = precision_score(test_df.Score,test_list)
         test_recall = recall_score(test_df.Score,test_list)
         #confusion matrix
         confusion_matrix_test = confusion_matrix(test_df.Score,test_list)
         print("penalty='11',alpha=0.000823,11_ratio=0.464483")
         print('Train Score', score_train)
         print('Test Score',score test)
         print('Test Precision',test_precision)
         print('Test Recall',test_recall)
         print('Test ConfusionMatrix',confusion_matrix_test)
penalty='11',alpha=0.000823,l1_ratio=0.464483
Train Score 0.9137142857142857
Test Score 0.889
Test Precision 0.9022622172228472
Test Recall 0.9708176438945669
Test ConfusionMatrix [[ 782 782]
 [ 217 7219]]
```

Observations:

- 1. For Binary Bag of Words got high mean cv at gamma = 0.010 ,C = 7.0 and cv mean score is 0.921215
 - Train Score 0.9915238095238095
 - Test Score 0.92466666666666
 - Test Precision 0.9447222953408791
 - Test Recall 0.9653039268423884
 - No of support vectors for each class [2245, 3591]
 - Test Confusion Matrix

$$\begin{bmatrix}
 1144 & 420 \\
 258 & 7287
 \end{bmatrix}
 \tag{1}$$

- 2. SGD: For Binary bagof words wit SGD got best cv score at alpha = 0.003570, 11_ratio = 0.912537, penalty 12 and corresponding mean cv test score is 0.915977
 - Train Score 0.9445238095238095
 - Test Score 0.9197777777778
 - Test Precision 0.930053804765565
 - Test Recall 0.9763313609467456
 - Test Confusion Matrix

$$\begin{bmatrix}
1018 & 546 \\
176 & 7260
\end{bmatrix}$$
(2)

- 3. For Tf-Idf got high mean cv at gamma = 0.175194, C = 7.107109 and mean cv is 0.924987
 - Train Score 1.0
 - Test Score 0.941555555555556
 - Test Precision 0.9508089770354906
 - Test Recall 0.9799623453469607
 - No of support vectors for each class [2810 6720]
 - Test Confusion Matrix

$$\begin{bmatrix}
1187 & 377 \\
149 & 7287
\end{bmatrix}$$
(3)

- 4. SGD: For Tf-Idf with sgd got best scores at alpha = 0.00003, l1_ratio = 0.15, penalty = 11 and mean cv score is 0.929282
 - Train Score 0.9569047619047619
 - Test Score 0.93555555555556
 - Test Precision 0.9517659462308908
 - Test Recall 0.9712210866057019
 - Test Confusion Matrix

$$\begin{bmatrix}
 1198 & 366 \\
 214 & 7222
 \end{bmatrix}
 \tag{4}$$

5. For Avg Word2Vec got high mean cv at gamma = 0.00100, C = 10 and mean cv score is 0.933630

- Train Score 0.9725714285714285
- Test Score 0.93322222222222
- Test Precision 0.9459741615555266
- Test Recall 0.9748520710059172
- No of support vectors for each class [1693 2078]

$$\begin{bmatrix}
1150 & 414 \\
187 & 7249
\end{bmatrix}$$
(5)

- 6. SGD: For Avg Word2Vec got best cv mean score at alpha = 0.00012,l1_ratio= 0.25 penlty = 11 and mean cv score is 0.925406
 - Train Score 0.9392380952380952
 - Test Score 0.9287777777778
 - Test Precision 0.9474516001580403
 - Test Recall 0.9674556213017751
 - Test ConfusionMatrix

$$\begin{bmatrix}
1165 & 399 \\
242 & 7194
\end{bmatrix}$$
(6)

- 7. For Tf-Idf Word2Vec got high mean cv at gamma = 0.00100 C = 20.0 and mean cv score is 0.892457
 - Train Score 0.9643809523809523
 - Test Score 0.89044444444445
 - Test Precision 0.8969719350073855
 - Test Recall 0.9799623453469607
 - No of support vectors for each class [2031 2750]

$$\begin{bmatrix}
727 & 837 \\
149 & 7287
\end{bmatrix}$$
(7)

- 8. SGD: Got best cv scores at alpha = 0.000823 l1_ratio = 0.464483 penalty = 11 and mean cv score is 0.892981.
 - Train Score 0.9137142857142857
 - Test Score 0.889
 - Test Precision 0.9022622172228472
 - Test Recall 0.9708176438945669
 - Test ConfusionMatrix

$$\begin{bmatrix}
 782 & 782 \\
 217 & 7219
 \end{bmatrix}
 \tag{8}$$