32.19.Amazon_food_review_RF

July 1, 2018

1 Amazon food review dataset apply RF to predict polarity review

Data set from https://www.kaggle.com/snap/amazon-fine-food-reviews

2 Objective

Here try BOW,TFIDF,avg W2V, avg TFIDFW2V and different hyperparameter RF(no of base learner) # Import data and libraries

```
In [25]: import sqlite3
         import pandas as pd
         import numpy as np
         import nltk
         import string
         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 roc_curve, auc
         from nltk.stem.porter import PorterStemmer
         con = sqlite3.connect('database.sqlite')
         #get only +ve and -ve review
         raw_data = pd.read_sql_query("""SELECT * FROM Reviews WHERE Score != 3""", con)
```

3 Data preprocessing

```
In [26]: filtered_data=raw_data
    # Score>3 a positive rating, and score<3 a negative rating.
    def partition(x):
        if x < 3:
            return 'negative'
        return 'positive'</pre>
```

```
#changing reviews with score less than 3 to be positive and vice-versa
         actualScore = filtered_data['Score']
         positiveNegative = actualScore.map(partition)
         filtered_data['Score'] = positiveNegative
         #filtered_data.sample(5)
         filtered_data['Score'].value_counts()
         #Sorting data according to ProductId in ascending order
         sorted_data=filtered_data.sort_values('ProductId', axis=0, ascending=True, inplace=Fa
         #Deduplication of entries for same profilename, userid, time, text and take first elem
         sorted_data=sorted_data.drop_duplicates(subset={"UserId", "ProfileName", "Time", "Text"}
In [27]: #take only 2500 + 2500 data
         from sklearn.cross_validation import train_test_split,KFold
         _ , clean_data = train_test_split(sorted_data, test_size = 40000, random_state=0,stra
         clean_data['Score'].value_counts()
Out[27]: positive
                     33727
         negative
                      6273
         Name: Score, dtype: int64
In [28]: # Clean html tag and punctuation
         import re
         import string
         from nltk.corpus import stopwords
         from nltk.stem import PorterStemmer
         from nltk.stem.wordnet import WordNetLemmatizer
         stop = set(stopwords.words('english')) #set of stopwords
         sno = nltk.stem.SnowballStemmer('english') #initialising the snowball stemmer
         #substitute html tag and punctuation
         def cleanhtml(sentence): #function to clean the word of any html-tags
             cleanr = re.compile('<.*?>')
             cleantext = re.sub(cleanr, ' ', sentence)
             return cleantext
         def cleanpunc(sentence): #function to clean the word of any punctuation or special ch
             cleaned = re.sub(r'[?|!|\'|"|#]',r'',sentence)
             cleaned = re.sub(r'[.|,|)|(||/|,r'|,cleaned)
             return cleaned
         #print(sno.stem('tasty'))
         i = 0
         str1=' '
         final_string=[]
         all_positive_words=[] # store words from +ve reviews here
         all_negative_words=[] # store words from -ve reviews here.
         #Create new catagory as Cleanedtext after removing htmltag and punctuation and upperc
         for sent in clean_data['Text'].values:
```

```
filtered_sentence=[]
             #print(sent);
             sent=cleanhtml(sent) # remove HTMl tags
             for w in sent.split():
                 for cleaned_words in cleanpunc(w).split():
                     if((cleaned_words.isalpha()) & (len(cleaned_words)>2)):
                         if(cleaned_words.lower() not in stop):
                             s=(sno.stem(cleaned_words.lower())).encode('utf8')
                             filtered_sentence.append(s)
                             if (clean_data['Score'].values)[i] == 'positive':
                                 all_positive_words.append(s) #list of all words used to descr
                             if(clean_data['Score'].values)[i] == 'negative':
                                 all_negative_words.append(s) #list of all words used to descr
                         else:
                             continue
                     else:
                         continue
             str1 = b" ".join(filtered_sentence) #final string of cleaned words
             final_string.append(str1)
             i+=1
         clean_data['CleanedText']=final_string
         clean_data.shape
         #Sort data on timestamp
         clean_data=clean_data.sort_values(by=['Time'],ascending=False)
         clean_data.sample(2)
Out [28]:
                     Ιd
                          ProductId
                                             UserId
                                                                   ProfileName \
         242151 262623 B003UGZOCO A1I4M2EYH36A5V
                                                                 Shannon Joyce
         320453 346891 B003CZGOSG A1BEHDHARZDIMR Tammie Sokoloff "Nashira"
                 HelpfulnessNumerator HelpfulnessDenominator
                                                                               Time \
                                                                  Score
         242151
                                    3
                                                            3 positive 1322352000
         320453
                                    2
                                                            2 positive
                                                                         1295222400
                                  Summary \
         242151 High Quality from Hill's
         320453
                              THE BEST!!!
                                                              Text \
         242151 We live in Santiago, Chile and the pickings we...
         320453 This stuff is great, tastes just like sugar an...
                                                       CleanedText
         242151 b'live santiago chile pick slim far qualiti co...
         320453 b'stuff great tast like sugar measur like suga...
```

4 Split train and test data

```
In [29]: x=clean_data['CleanedText'].values
         y = clean_data['Score']
         n=x.shape[0]
         n1=int(n*.3)
         X_{test_raw} = x[0:n1]
         X_train_raw= x[n1:n+1]
         y_{test=y[0:n1]}
         y_train=y[n1:n+1]
         print('size of X_train, X_test, y_train , y_test ',X_train_raw.shape, X_test_raw.shape
         print("positive and negative review in train and test\n",y_train.value_counts(),"\n",;
size of X_train, X_test, y_train, y_test (28000,) (12000,) (28000,) (12000,)
positive and negative review in train and test
             23774
positive
             4226
negative
Name: Score, dtype: int64
             9953
positive
negative
            2047
Name: Score, dtype: int64
```

5 Build Random Forest

6 BOW

7 plot accuracy with k value

```
from sklearn.grid_search import GridSearchCV
                   #GradientBoostingClassifier(learning_rate=0.1, n_estimators=100, max_depth=3, random_
                  n_{estimators}=[10,50,100]
                  tuned_parameters=dict(n_estimators=n_estimators)
                   #Using GridSearchCV
                  model = GridSearchCV(RandomForestClassifier(), tuned_parameters, cv=5)
                  model.fit(X_train, y_train)
                  print('Best parameters \n', model.best_estimator_)
                   #print('Model test score', model.score(X_test, y_test))
                   optimumn_estimators=model.best_estimator_.n_estimators
                   #build model with optimum parameter
                  model = RandomForestClassifier(n_estimators=optimumn_estimators)
                  model.fit(X_train, y_train)
                   #Store scores
                  pred=model.predict(X_test)
                  mat=pd.crosstab(y_test, pred, rownames=['Actual'], colnames=['Predicted'], margins=Tr
                   tp=mat.iloc[1,1] ;tn=mat.iloc[0,0] ;fp=mat.iloc[0,1];fn=mat.iloc[1,0];precision=tp/(t
                  recall=tp/(tp+fn);fscoretest=2*precision*recall/(precision+recall)
                  pred=model.predict(X_train)
                   mat=pd.crosstab(y_train, pred, rownames=['Actual'], colnames=['Predicted'], margins=Tates | Tates | Ta
                  print(mat);tp=mat.iloc[1,1];tn=mat.iloc[0,0];fp=mat.iloc[0,1];fn=mat.iloc[1,0];pre
                  recall=tp/(tp+fn);fscoretrain=2*precision*recall/(precision+recall)
                   aa=pd.DataFrame({'type':['BOW RF'],'depth':['na'],'estimator':[optimumn_estimators], '
                                                         'learningrate':['na'],'accuracy_train':[model.score(X_train,y_train
                                                       'fscore_train':[fscoretrain], 'accuracy_test':[model.score(X_test,y_test)
(28000, 20761) (12000, 20761)
Best parameters
 RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                        max_depth=None, max_features='auto', max_leaf_nodes=None,
                        min_impurity_decrease=0.0, min_impurity_split=None,
                        min_samples_leaf=1, min_samples_split=2,
                        min_weight_fraction_leaf=0.0, n_estimators=50, n_jobs=1,
                         oob_score=False, random_state=None, verbose=0,
                         warm_start=False)
Predicted negative positive
                                                                     All
```

```
Actual negative 4225 1 4226 positive 0 23774 23774 All 4225 23775 28000
```

8 TFIDF

```
In [20]: tf_idf_vect = TfidfVectorizer()
         final_counts = tf_idf_vect.fit_transform(X_train_raw)
         #use the same vectors to convert test data
         X_test=count_vect.transform(X_test_raw)
         print(X_train.get_shape(), X_test.get_shape())
         #Use scale of train and apply to test
         from sklearn.preprocessing import StandardScaler
         scaler = StandardScaler(with mean=False).fit(X train)
         \#X\_train = scaler.transform(X\_train)
         \#X\_test = scaler.transform(X\_test)
         print('size of X_train, X_test, y_train , y_test ',X_train.shape, X_test.shape,y_train
         # Use RF
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import GradientBoostingClassifier,RandomForestClassifier
         #RandomForestClassifier(n_estimators=10)
         from sklearn.grid_search import GridSearchCV
         #GradientBoostingClassifier(learning_rate=0.1, n_estimators=100, max_depth=3, random_
         n_{estimators}=[10,50,100]
         tuned_parameters=dict(n_estimators=n_estimators)
         #Using GridSearchCV
         model = GridSearchCV(RandomForestClassifier(), tuned_parameters, cv=5)
         model.fit(X_train, y_train)
         print('Best parameters \n', model.best_estimator_)
         #print('Model test score', model.score(X_test, y_test))
         optimumn_estimators=model.best_estimator_.n_estimators
         #build model with optimum parameter
         model = RandomForestClassifier(n_estimators=optimumn_estimators)
         model.fit(X_train, y_train)
```

```
#Store scores
                     pred=model.predict(X_test)
                     mat=pd.crosstab(y_test, pred, rownames=['Actual'], colnames=['Predicted'], margins=Tr
                      tp=mat.iloc[1,1] ; tn=mat.iloc[0,0] ; fp=mat.iloc[0,1]; fn=mat.iloc[1,0]; precision=tp/(tp+mat.iloc[1,0]); fn=mat.iloc[1,0]; fn=mat.iloc
                      recall=tp/(tp+fn);fscoretest=2*precision*recall/(precision+recall)
                     pred=model.predict(X_train)
                     mat=pd.crosstab(y_train, pred, rownames=['Actual'], colnames=['Predicted'], margins=T:
                     print(mat);tp=mat.iloc[1,1] ;tn=mat.iloc[0,0];fp=mat.iloc[0,1] ;fn=mat.iloc[1,0] ;pre-
                     recall=tp/(tp+fn);fscoretrain=2*precision*recall/(precision+recall)
                     bb=pd.DataFrame({'type':['TFIDF RF'],'depth':['na'],'estimator':[optimumn_estimators]
                                                                  'learningrate':['na'], 'accuracy_train':[model.score(X_train,y_train
                                                                'fscore_train':[fscoretrain],'accuracy_test':[model.score(X_test,y_test)
                      aa=aa.append(bb)
(28000, 20761) (12000, 20761)
size of X_train, X_test, y_train , y_test (28000, 20761) (12000, 20761) (28000,) (12000,)
Best parameters
  RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                             max_depth=None, max_features='auto', max_leaf_nodes=None,
                            min_impurity_decrease=0.0, min_impurity_split=None,
                            min_samples_leaf=1, min_samples_split=2,
                            min_weight_fraction_leaf=0.0, n_estimators=50, n_jobs=1,
                             oob_score=False, random_state=None, verbose=0,
                             warm_start=False)
Predicted negative positive
                                                                                All
Actual
                                                                              4226
                                    4222
negative
positive
                                                           23774 23774
                                           0
                                    4222
All
                                                          23778 28000
```

9 AVG W2V

```
In [36]: #ignore warning
    import warnings
    warnings.filterwarnings('ignore')
    from gensim.models import Word2Vec
    from gensim.models import KeyedVectors
    import pickle
    model = KeyedVectors.load_word2vec_format('GoogleNews-vectors-negative300.bin.gz', bis import gensim
    i=0
    #create a list of list to be used in W2V
```

```
list_of_sent_train=[]
for sent in X_train_raw: #clean_data['CleanedText'].values:
    filtered_sentence=[]
    #sent=cleanhtml(sent)
    for w in sent.split():
        #for cleaned_words in cleanpunc(w).split():
         for cleaned_words in w.split():
            if(cleaned_words.isalpha()):
                filtered_sentence.append(cleaned_words.lower().decode('utf8'))
            else:
                continue
    list_of_sent_train.append(filtered_sentence)
#convert each sentence's words to a vector of 50 dimension. Dont construct vec if wor
#and 4 core processor
w2v_model=gensim.models.Word2Vec(list_of_sent_train,min_count=5,size=50, workers=4)
# average Word2Vec
# for each sentence make average of vectors by (vectors of each words)/(total no of w
# compute average word2vec for each review.
sent_vectors_train = []; # the aug-w2v for each sentence/review is stored in this lis
for sent in list_of_sent_train: # for each review/sentence
    sent_vec = np.zeros(50) # as word vectors are of zero length
    cnt_words =0; # num of words with a valid vector in the sentence/review
    for word in sent: # for each word in a review/sentence
        try:
            vec = w2v_model.wv[word]
            sent_vec += vec
            cnt_words += 1
        except:
            pass
    sent_vec /= cnt_words
    sent_vectors_train.append(sent_vec)
#convert W2V test data
i=0
#create a list of list to be used in W2V
list_of_sent_test=[]
for sent in X_test_raw: #clean_data['CleanedText'].values:
    filtered sentence=[]
    #sent=cleanhtml(sent)
    for w in sent.split():
        #for cleaned_words in cleanpunc(w).split():
         for cleaned_words in w.split():
            if(cleaned_words.isalpha()):
                filtered_sentence.append(cleaned_words.lower().decode('utf8'))
            else:
                continue
```

```
list_of_sent_test.append(filtered_sentence)
#convert each sentence's words to a vector of 50 dimension. Dont construct vec if wor
#and 4 core processor
w2v_model=gensim.models.Word2Vec(list_of_sent_test,min_count=5,size=50, workers=4)
# average Word2Vec
# for each sentence make average of vectors by (vectors of each words)/(total no of w
# compute average word2vec for each review.
sent_vectors_test = []; # the avg-w2v for each sentence/review is stored in this list
for sent in list_of_sent_test: # for each review/sentence
    sent_vec = np.zeros(50) # as word vectors are of zero length
    cnt_words =0; # num of words with a valid vector in the sentence/review
    for word in sent: # for each word in a review/sentence
        try:
            vec = w2v_model.wv[word]
            sent_vec += vec
            cnt_words += 1
        except:
            pass
    sent_vec /= cnt_words
    sent_vectors_test.append(sent_vec)
# try
X_train = pd.DataFrame(sent_vectors_train)
X_test = pd.DataFrame(sent_vectors_test)
#X_train = sent_vectors_train
#X_test = sent_vectors_test
\#X\_train\_w2v=X\_train
\#X\_test\_w2v=\!X\_test
print('size of X_train, X_test, y_train , y_test ',X_train.shape, X_test.shape,y_train
#to avoid "ValueError: Input contains NaN, infinity or a value too large for dtype('f
X_test.fillna(X_test.mean())
X_train.fillna(X_train.mean())
X_train=np.nan_to_num(X_train) # replace nan with zero and inf with finite numbers
X_test=np.nan_to_num(X_test) # replace nan with zero and inf with finite numbers
#np.nan_to_num(X_train)
#np.nan_to_num(X_test)
print("Checkinf for NaN and Inf")
print("np.inf=", np.where(np.isnan(X_train)))
print("is.inf=", np.where(np.isinf(X_train)))
n_estimators=[10,50,100]
```

```
tuned_parameters=dict(n_estimators=n_estimators)
         model = GridSearchCV(RandomForestClassifier(), tuned_parameters, cv=5)
         model.fit(X_train, y_train)
         print('Best parameters \n', model.best_estimator_)
         optimumn_estimators=model.best_estimator_.n_estimators
         #build model with optimum parameter
         model = RandomForestClassifier(n_estimators=optimumn_estimators)
         model.fit(X_train, y_train)
         #Store scores
         pred=model.predict(X_test)
         mat=pd.crosstab(y_test, pred, rownames=['Actual'], colnames=['Predicted'], margins=Tr
         tp=mat.iloc[1,1] ;tn=mat.iloc[0,0] ;fp=mat.iloc[0,1];fn=mat.iloc[1,0];precision=tp/(t)
         recall=tp/(tp+fn);fscoretest=2*precision*recall/(precision+recall)
         pred=model.predict(X_train)
         mat=pd.crosstab(y_train, pred, rownames=['Actual'], colnames=['Predicted'], margins=T:
         print(mat);tp=mat.iloc[1,1] ;tn=mat.iloc[0,0];fp=mat.iloc[0,1] ;fn=mat.iloc[1,0] ;pre-
         recall=tp/(tp+fn);fscoretrain=2*precision*recall/(precision+recall)
         bb=pd.DataFrame({'type':['AVG W2V RF'],'depth':['na'],'estimator':[optimumn_estimator']
                           'learningrate':['na'],'accuracy_train':[model.score(X_train,y_train
                          'fscore_train':[fscoretrain], 'accuracy_test':[model.score(X_test,y_test)
         aa=aa.append(bb)
size of X_train, X_test, y_train, y_test (28000, 50) (12000, 50) (28000,) (12000,)
Checkinf for NaN and Inf
np.inf= (array([], dtype=int64), array([], dtype=int64))
is.inf= (array([], dtype=int64), array([], dtype=int64))
Best parameters
RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
           max_depth=None, max_features='auto', max_leaf_nodes=None,
           min_impurity_decrease=0.0, min_impurity_split=None,
           min_samples_leaf=1, min_samples_split=2,
           min_weight_fraction_leaf=0.0, n_estimators=100, n_jobs=1,
            oob_score=False, random_state=None, verbose=0,
            warm_start=False)
Predicted negative positive
                                 All
Actual
               4226
                                4226
negative
                            0
positive
                  0
                        23774 23774
All
               4226
                        23774 28000
```

10 AVG TFIDF

```
In [37]: #ignore warning
    import warnings
```

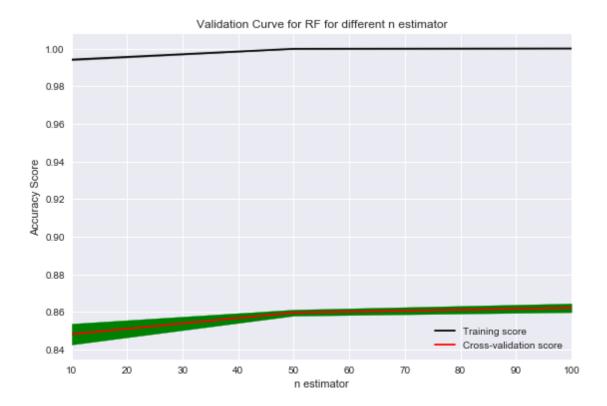
```
warnings.filterwarnings('ignore')
from gensim.models import Word2Vec
from gensim.models import KeyedVectors
import pickle
model = KeyedVectors.load_word2vec_format('GoogleNews-vectors-negative300.bin.gz', bi
import gensim
tf_idf_vect = TfidfVectorizer()
final_tf_idf=tf_idf_vect.fit_transform(X_train_raw)
tfidf_feat = tf_idf_vect.get_feature_names() # tfidf_words/col-names
# final_tf_idf is the sparse matrix with row= sentence, col=word and cell_val = tfidf
tfidf_sent_vectors_train = []; # the tfidf-w2v for each sentence/review is stored in
#calculate avg tfidf score for each sentences
for sent in list_of_sent_train: # for each review/sentence
    sent_vec = np.zeros(50) # as word vectors are of zero length
    weight_sum =0; # num of words with a valid vector in the sentence/review
    for word in sent: # for each word in a review/sentence
            vec = w2v model.wv[word] #calculate w2v for each word
            # obtain the tf_idfidf of a word in a sentence/review
            tf_idf = final_tf_idf[row, tfidf_feat.index(word)] #get tfidf score of eac
            sent_vec += (vec * tf_idf) # multiply vec with tfidf of each word and cum
            weight_sum += tf_idf # also add tfidf sums in each sentence
        except:
            pass
    sent_vec /= weight_sum
    tfidf_sent_vectors_train.append(sent_vec)
    row += 1
\#tfidf\_sent\_vectors.
# do for test
final_tf_idf=tf_idf_vect.transform(X_test_raw)
tfidf_sent_vectors_test = []; # the tfidf-w2v for each sentence/review is stored in t
#calculate avg tfidf score for each sentences
for sent in list_of_sent_test: # for each review/sentence
    sent_vec = np.zeros(50) # as word vectors are of zero length
    weight_sum =0; # num of words with a valid vector in the sentence/review
    for word in sent: # for each word in a review/sentence
        try:
            vec = w2v_model.wv[word] #calculate w2v for each word
            # obtain the tf_idfidf of a word in a sentence/review
            tf_idf = final_tf_idf[row, tfidf_feat.index(word)] #get tfidf score of eac
            sent_vec += (vec * tf_idf) # multiply vec with tfidf of each word and cum
            weight_sum += tf_idf # also add tfidf sums in each sentence
```

```
except:
            pass
    sent_vec /= weight_sum
    tfidf_sent_vectors_test.append(sent_vec)
    row += 1
\#X\_train = pd.DataFrame(tfidf\_sent\_vectors\_train)
\#X\_test = pd.DataFrame(tfidf\_sent\_vectors\_test)
X_train = tfidf_sent_vectors_train
X_test = tfidf_sent_vectors_test
\#print('size\ of\ X\_train,\ X\_test,\ y\_train\ ,\ y\_test\ ',X\_train.shape,\ X\_test.shape,y\_train
#to avoid "ValueError: Input contains NaN, infinity or a value too large for dtype('f
\#X_{test.fillna}(X_{test.mean}())
\#X_train.fillna(X_train.mean())
X_train=np.nan_to_num(X_train) # replace nan with zero and inf with finite numbers
X_test=np.nan_to_num(X_test) # replace nan with zero and inf with finite numbers
#X_train=np.nan_to_num(X_train) # replace nan with zero and inf with finite numbers
n_{estimators}=[10,50,100]
tuned_parameters=dict(n_estimators=n_estimators)
model = GridSearchCV(RandomForestClassifier(), tuned_parameters, cv=5)
model.fit(X_train, y_train)
print('Best parameters \n', model.best_estimator_)
optimumn_estimators=model.best_estimator_.n_estimators
\#build\ model\ with\ optimum\ parameter
model = RandomForestClassifier(n_estimators=optimumn_estimators)
model.fit(X_train, y_train)
#Store scores
pred=model.predict(X_test)
mat=pd.crosstab(y_test, pred, rownames=['Actual'], colnames=['Predicted'], margins=Tr
tp=mat.iloc[1,1];tn=mat.iloc[0,0];fp=mat.iloc[0,1];fn=mat.iloc[1,0];precision=tp/(tj
recall=tp/(tp+fn);fscoretest=2*precision*recall/(precision+recall)
pred=model.predict(X_train)
mat=pd.crosstab(y_train, pred, rownames=['Actual'], colnames=['Predicted'], margins=T
print(mat);tp=mat.iloc[1,1] ;tn=mat.iloc[0,0];fp=mat.iloc[0,1] ;fn=mat.iloc[1,0] ;pre-
recall=tp/(tp+fn);fscoretrain=2*precision*recall/(precision+recall)
bb=pd.DataFrame({'type':['AVG W2V RF'],'depth':['na'],'estimator':[optimumn_estimator
                   'learningrate':['na'], 'accuracy_train':[model.score(X_train,y_train
                  'fscore_train':[fscoretrain],'accuracy_test':[model.score(X_test,y_test)
aa=aa.append(bb)
```

```
Best parameters
RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
           max_depth=None, max_features='auto', max_leaf_nodes=None,
           min_impurity_decrease=0.0, min_impurity_split=None,
           min_samples_leaf=1, min_samples_split=2,
           min_weight_fraction_leaf=0.0, n_estimators=100, n_jobs=1,
            oob_score=False, random_state=None, verbose=0,
            warm_start=False)
Predicted negative positive
                                 All
Actual
               4226
                                4226
                            0
negative
                  0
                        23774 23774
positive
               4226
All
                        23774 28000
```

11 Plot accuracy with n_extimator

```
In [38]: import warnings
         warnings.filterwarnings('ignore')
         from sklearn.model_selection import validation_curve
         #create plot for training and test validation
         # We cannot put 0 in parameter it will give error while validation curve
         n_estimators=[10,50,100]
         param_range=[10,50,100]
         train_scores, test_scores = validation_curve(RandomForestClassifier(), X_train, y_tra
                                                      param_range=n_estimators,cv=5)
         #print(train_scores, test_scores)
         train_scores_mean = np.mean(train_scores, axis=1)
         train scores std = np.std(train scores, axis=1)
         test_scores_mean = np.mean(test_scores, axis=1)
         test_scores_std = np.std(test_scores, axis=1)
         plt.plot(param_range, train_scores_mean, label="Training score", color="black")
         plt.plot(param_range, test_scores_mean, label="Cross-validation score", color="red")
         plt.fill_between(param_range, train_scores_mean - train_scores_std, train_scores_mean
         plt.fill between(param range, test_scores mean - test_scores std, test_scores_mean + range)
         plt.title("Validation Curve for RF for different n estimator")
         plt.xlabel("n estimator")
         plt.ylabel("Accuracy Score")
         plt.xlim(10,100)
         plt.tight_layout()
         plt.legend(loc="best")
         plt.show()
```



12 Steps followed

Only !=3 reviews are taken Mark >3 as positive and <3 as negative. Sort data as per product id in ascending order Deduplication of entries for same profilename, userid, time, text and take first element Get stratified sampling of 50k data Clean html and punctuation Convert to uppercase and word<3 are rejected data sorted on time Split the data in train and test to 70:30

BOW BOW vec created using train data test data is converted using above on train and same applied to test do grid search for different value of n_estimator best model is established with best hyperparameter. model metric is stored in dataframe and crosstable is printed.

TFIDF TFIDF form tfidf vec using train same is used in test to convert rest are same

AVG W2V gensim is used to convert train and test text to W2V AVG seperately rest are same AVG TFIDF form thidf vec using train same is used in test to convert gensim and above thidf is used to convert train and test text to W2V AVG seperately

plot cv error with n_estimator

Below is metric

In [40]: aa

Out[40]:	accuracy_test	accuracy_train	depth	estimator	fscore_test	fscore_train	\
0	0.858583	0.999964	na	50	0.921191	0.999979	
0	0.859167	0.999857	na	50	0.921534	0.999916	
0	0.828667	1.000000	na	100	0.905558	1.000000	

0	0.84950	0 1.00000	0 na	100	0.915536	1.000000
lear	rningrate	type				
0	na	BOW RF				
0	na	TFIDF RF				
0	na	AVG W2V RF				
0	na	AVG W2V RF				