Random Forest and XGBoost on Amazon Food Reviews

Data Source: https://www.kaggle.com/snap/amazon-fine-food-reviews (https://www.kaggle.com/snap/amazon-fine-food-reviews)

The Amazon Fine Food Reviews dataset consists of reviews of fine foods from Amazon.

Number of reviews: 568,454 Number of users: 256,059 Number of products: 74,258 Timespan: Oct 1999 - Oct 2012 Number of Attributes/Columns in data: 10

Attribute Information:

- 1. index
- 2. ld
- 3. Productld unique identifier for the product
- 4. Userld ungiue identifier for the user
- 5. ProfileName
- 6. HelpfulnessNumerator number of users who found the review helpful
- 7. HelpfulnessDenominator number of users who indicated whether they found the review helpful or not
- 8. Score rating between 1 and 5
- 9. Time timestamp for the review
- 10. Summary brief summary of the review
- 11. Text text of the review
- 12. ProcessedText Cleaned & Preprocessed Text of the review

Objective: Given Amazon Food reviews, convert all the reviews into a vector using two techniques:

- 1. Average W2V.
- 2. Average TFIDF-W2V.

Then perform following tasks under each technique:

- Task 1. Split train and test data in a ratio of 80:20.
- Task 2. Perform GridSearch Cross Validation to find optimal value of number of base models in Random Forest.
- Task 3. Apply Random Forest and report accuracy. Also check for train error.

Task 4. Perform GridSearch Cross Validation to find optimal number of base models and optimal value of depth of decision tree base models in XGBoost.

Task 5. Apply XGBoost and report accuracy. Also check for train error.

[Q] How to determine if a review is positive or negative?

[Ans] We could use the Score/Rating. A rating of 4 or 5 could be cosnidered a positive review. A review of 1 or 2 could be considered negative. A review of 3 is nuetral and ignored. This is an approximate and proxy way of determining the polarity (positivity/negativity) of a review.

Loading the data

SQLite Database

In order to load the data, We have used the SQLITE dataset as it easier to guery the data and visualise the data efficiently. Here as we only want to get the global sentiment of the recommendations (positive or negative), we will purposefully ignore all Scores equal to 3. If the score id above 3, then the recommendation wil be set to "positive". Otherwise, it will be set to "negative".

```
In [3]:
        import sqlite3
        import pandas as pd
        import numpy as np
        import seaborn as sns
        import matplotlib.pyplot as plt
        from gensim.models import Word2Vec
        import gensim
        from sklearn.metrics import accuracy score, confusion matrix
        from sklearn.feature extraction.text import TfidfVectorizer
        from sklearn.ensemble import RandomForestClassifier
        import xgboost as xgb
        from sklearn.cross validation import train test split
        from sklearn.grid search import GridSearchCV
```

```
connection = sqlite3.connect('FinalAmazonFoodReviewsDataset.sqlite')
In [4]:
        data = pd.read sql query("SELECT * FROM Reviews", connection)
In [5]:
```

In [6]: data.head()

Out[6]:

t[6]: _	ind	ex	ld	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score	Time	Summary	
	0	0	1	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	1	Positive	1303862400	Good Quality Dog Food	Sŧ
	1	1	2	B00813GRG4	A1D87F6ZCVE5NK	dll pa	0	0	Negative	1346976000	Not as Advertised	lal F
	2	2	3	B000LQOCH0	ABXLMWJIXXAIN	Natalia Corres "Natalia Corres"	1	1	Positive	1219017600	"Delight" says it all	CC E
	3	4	5	B006K2ZZ7K	A1UQRSCLF8GW1T	Michael D. Bigham "M. Wassir"	0	0	Positive	1350777600	Great taffy	1
	4	5	6	B006K2ZZ7K	ADT0SRK1MGOEU	Twoapennything	0	0	Positive	1342051200	Nice Taffy	
4												•

In [7]: data.shape

Out[7]: (364171, 12)

```
In [8]: data["Score"].value counts()
 Out[8]: Positive
                     307061
         Negative
                      57110
         Name: Score, dtype: int64
In [9]:
         def changingScores(score):
             if score == "Positive":
                 return 1
             else:
                 return 0
In [10]: # changing score
         # Positive = 1
         # Negative = 0
         actualScore = list(data["Score"])
         positiveNegative = list(map(changingScores, actualScore)) #map(function, list of numbers)
         data['Score'] = positiveNegative
In [11]: #taking 10000 random samples
         data = data.sample(n = 10000)
```

In [12]: data.head()

Out[12]:

	index	ld	ProductId	UserId	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score	Time	Summa
1771	2319	2525	B000NY4SZQ	A1IDF30FSGME1L	Shari	2	2	1	1283040000	Thor Musta Mani
307570	442577	478546	B000E3XCG2	A3N8XPE3QXZ62Q	Connie L. Roberts "Daeshii"	1	1	1	1185321600	C Yumm
309959	445341	481519	B004OY7FD8	A18F3YI8GZCIIY	Karmali	0	0	1	1340496000	Delicio cracke
183392	259085	280861	B000FFLHSY	A3JETXRRUHIOUH	rain21004	3	4	0	1271980800	Dra clear
303846	436877	472419	B001F1U5WG	A1GZVRXW89KZST	kawtamer	1	1	1	1301011200	Gre
1										>

In [13]: data.shape

Out[13]: (10000, 12)

```
data["Score"].value_counts()
In [14]:
Out[14]: 1
              8456
              1544
         Name: Score, dtype: int64
In [15]:
         data = data.sort_values('Time', axis=0, ascending=True)
In [16]:
         Data = data
         Data_Labels = data["Score"]
In [17]:
In [18]:
         print(Data.shape)
         print(Data_Labels.shape)
         (10000, 12)
         (10000,)
```

In [19]: Data.head()

Out[19]:

	index	ld	ProductId	Userld	ProfileName	HelpfulnessNumerator	HelpfulnessDenominator	Score	Time	Summa
97415	138001	149770	B00004S1C5	A1KXONFPU2XQ5K	Stephanie Manley	8	8	1	965779200	Very ea to u
238632	346030	374332	B00004CI84	AEPJYN0NAX9N4	Jody L. Schoth	0	0	1	1014163200	Excellen Hilario
308591	443674	479736	B00005U2FA	A1UUIV2251UKHJ	S. McMillian	2	2	1	1052179200	Vacu \
267514	388413	419994	B0000A0BS5	A238V1XTSK9NFE	Andrew Lynn	46	59	0	1064361600	actually to use espres machin
129891	184396	200021	B0000D9N5Q	A176XYKMRF1U8Q	Lance Arthur	0	0	1	1067990400	A sm taste heav
4										•

1. Avg W2V

```
In [20]: i = 0
         listOfSentences = []
         for sentence in Data["ProcessedText"].values:
             subSentence = []
             for word in sentence.split():
                 subSentence.append(word)
             listOfSentences.append(subSentence)
         print(Data['ProcessedText'].values[0])
In [21]:
         print("\n")
         print(listOfSentences[0:2])
         print("\n")
         print(type(listOfSentences))
         this are much easier use than the wilson past color color are vibrant and not taint the frost like some color can these
         are simpl use and not make mess onli complaint that did not find these year ago this must have you decor often
         [['this', 'are', 'much', 'easier', 'use', 'than', 'the', 'wilson', 'past', 'color', 'color', 'are', 'vibrant', 'and',
         'not', 'taint', 'the', 'frost', 'like', 'some', 'color', 'can', 'these', 'are', 'simpl', 'use', 'and', 'not', 'make',
         'mess', 'onli', 'complaint', 'that', 'did', 'not', 'find', 'these', 'year', 'ago', 'this', 'must', 'have', 'you', 'deco
         r', 'often'], ['just', 'hilari', 'favorit', 'part', 'the', 'dine', 'room', 'scene', 'with', 'all', 'the', 'sing', 'an
         d', 'danc', 'classic', 'moment', 'not', 'miss', 'winona', 'rider', 'and', 'geena', 'davi', 'are', 'spectacular', 'thi
         s', 'movi', 'and', 'michael', 'keaton', 'unbeliev', 'funni', 'couldnt', 'stop', 'laugh', 'great', 'famili', 'movi', 'wi
         11', 'keep', 'you', 'stitch']]
         <class 'list'>
In [22]: w2vModel = gensim.models.Word2Vec(listOfSentences, size=300, min count=5, workers=4)
```

```
In [23]:
         # compute average word2vec for each review.
         sentenceAsW2V = []
         for sentence in listOfSentences:
             sentenceVector = np.zeros(300)
             TotalWordsPerSentence = 0
             for word in sentence:
                 try:
                      vect = w2vModel.wv[word]
                      sentenceVector += vect
                      TotalWordsPerSentence += 1
                  except:
                      pass
             if TotalWordsPerSentence!= 0:
                  sentenceVector /= TotalWordsPerSentence
                 sentenceAsW2V.append(sentenceVector)
         print(type(sentenceAsW2V))
         print(len(sentenceAsW2V))
         print(len(sentenceAsW2V[0]))
         <class 'list'>
         10000
         300
```

Task 1. Split train and test data in a ratio of 80:20.

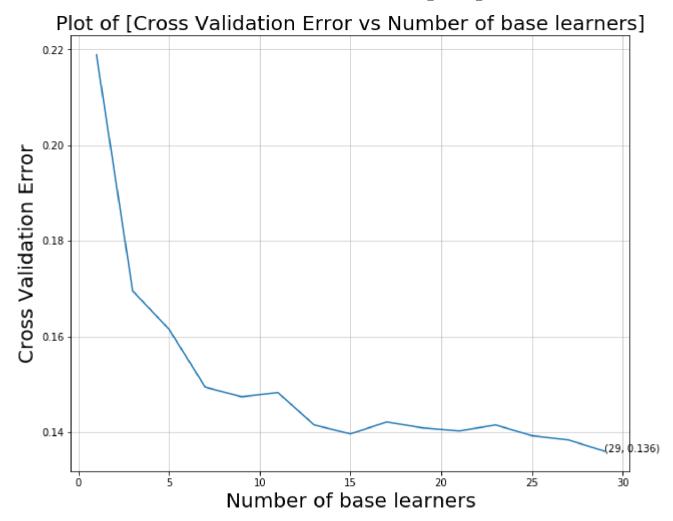
```
In [24]: X train AvgW2V, X test AvgW2V, Y train labels AvgW2V, Y test labels AvgW2V = train test split(sentenceAsW2V, Data Labels,
In [25]: train AvgW2V = np.array(X train AvgW2V)
         test AvgW2V = np.array(X test AvgW2V)
         train labels AvgW2V = np.array(Y train labels AvgW2V)
         test labels AvgW2V = np.array(Y test labels AvgW2V)
In [26]: train AvgW2V.shape, test AvgW2V.shape, train labels AvgW2V.shape, test labels AvgW2V.shape
Out[26]: ((8000, 300), (2000, 300), (8000,), (2000,))
```

Task 2. Perform GridSearch Cross Validation to find optimal value of number of base models in Random Forest.

```
In [125]:
          values = []
          for i in range(1, 31, 2):
              values.append(i)
          clf = RandomForestClassifier(n jobs = -1)
          hyper parameters = {'n estimators': values}
          bestScore = GridSearchCV(clf, hyper parameters, scoring = "accuracy", cv = 3)
          bestScore.fit(train AvgW2V, train labels AvgW2V)
          print(bestScore.best estimator )
          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=29, n jobs=-1,
                      oob score=False, random state=None, verbose=0,
                      warm start=False)
          best parameter = bestScore.best params
In [126]:
          bp = best parameter["n estimators"]
          bp
Out[126]: 29
```

```
In [127]:
          scoreData = bestScore.grid scores
           scoreData
Out[127]: [mean: 0.78112, std: 0.00458, params: {'n estimators': 1},
           mean: 0.83050, std: 0.00185, params: {'n estimators': 3},
           mean: 0.83850, std: 0.00621, params: {'n estimators': 5},
           mean: 0.85062, std: 0.00498, params: {'n estimators': 7},
           mean: 0.85262, std: 0.00159, params: {'n estimators': 9},
           mean: 0.85175, std: 0.00097, params: {'n estimators': 11},
           mean: 0.85850, std: 0.00356, params: {'n estimators': 13},
           mean: 0.86038, std: 0.00272, params: {'n estimators': 15},
           mean: 0.85788, std: 0.00082, params: {'n estimators': 17},
           mean: 0.85913, std: 0.00034, params: {'n estimators': 19},
           mean: 0.85975, std: 0.00293, params: {'n estimators': 21},
           mean: 0.85850, std: 0.00204, params: {'n estimators': 23},
           mean: 0.86075, std: 0.00428, params: {'n estimators': 25},
           mean: 0.86162, std: 0.00359, params: {'n estimators': 27},
           mean: 0.86400, std: 0.00015, params: {'n estimators': 29}]
```

```
In [128]: error = []
          parameter = []
          for i in range(len(scoreData)):
              error.append(1 - scoreData[i][1])
              parameter.append(scoreData[i][0]["n estimators"])
          plt.figure(figsize=(10,8))
          plt.plot(parameter, error)
          plt.title("Plot of [Cross Validation Error vs Number of base learners]", fontsize=20)
          plt.xlabel("Number of base learners", fontsize=20)
          plt.ylabel("Cross Validation Error", fontsize=20)
          plt.grid(linestyle='-', linewidth=0.5)
          errorMin = min(error)
           errorMin
          for xy in zip(parameter, error):
              if xy == (bp, errorMin):
                  plt.annotate(xy, xy)
```



Task 3. Apply Random Forest and report accuracy. Also check for train error.

```
In [129]:
          #First checking for train error
          model rf tr = RandomForestClassifier(n estimators = bp, n jobs = -1)
          model rf tr.fit(train AvgW2V, train labels AvgW2V)
          prediction AvgW2V rf tr = model rf tr.predict(train AvgW2V)
          AccuracyScore AvgW2V rf tr = accuracy score(train labels AvgW2V, prediction AvgW2V rf tr)
          print("Train error of Random Forest = "+str(np.round(((1 - AccuracyScore AvgW2V rf tr) * 100), 4))+"%")
          Train error of Random Forest = 0.1%
          model rf = RandomForestClassifier(n estimators = bp, n jobs = -1)
In [130]:
          model rf.fit(train AvgW2V, train labels AvgW2V)
          prediction AvgW2V rf = model rf.predict(test AvgW2V)
          AccuracyScore AvgW2V rf = accuracy score(test labels AvgW2V, prediction AvgW2V rf) * 100
          print("Accuracy score of Random Forest = "+str(AccuracyScore AvgW2V rf)+"%")
          Accuracy score of Random Forest = 85.25%
          Confusion Matrix = confusion matrix(test labels AvgW2V, prediction AvgW2V rf)
In [131]:
          print("Confusion Matrix on L2 regularization \n"+str(Confusion Matrix))
          Confusion Matrix on L2 regularization
          [[ 82 262]
           [ 33 1623]]
In [132]: tn, fp, fn, tp = confusion matrix(test labels AvgW2V, prediction AvgW2V rf).ravel()
          tn, fp, fn, tp
Out[132]: (82, 262, 33, 1623)
```

Task 4. Perform GridSearch Cross Validation to find optimal number of base models and optimal value of depth of decision tree base models in XGBoost.

```
xgb model = xgb.XGBClassifier()
In [31]:
         hyperParameters = { 'max depth':[1, 2, 3, 4], 'n estimators':[30, 50, 70, 100]}
         clf = GridSearchCV(estimator = xgb model, param grid = hyperParameters, scoring='accuracy', cv=3)
         clf.fit(train AvgW2V, train labels AvgW2V)
         print(clf.best estimator )
```

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if diff:

C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check that an array is not empty.

if diff:

XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1, colsample bytree=1, gamma=0, learning rate=0.1, max delta step=0, max depth=4, min child weight=1, missing=None, n estimators=100, n jobs=1, nthread=None, objective='binary:logistic', random state=0, reg alpha=0, reg lambda=1, scale pos weight=1, seed=None, silent=True, subsample=1)

```
best parameter = clf.best params
In [32]:
         best parameter
Out[32]: {'max depth': 4, 'n estimators': 100}
In [33]: | scoreData = clf.grid scores
         scoreData
Out[33]: [mean: 0.84350, std: 0.00012, params: {'max depth': 1, 'n estimators': 30},
          mean: 0.84613, std: 0.00072, params: {'max depth': 1, 'n estimators': 50},
          mean: 0.85037, std: 0.00078, params: {'max depth': 1, 'n estimators': 70},
          mean: 0.85525, std: 0.00086, params: {'max depth': 1, 'n estimators': 100},
          mean: 0.85200, std: 0.00186, params: {'max depth': 2, 'n estimators': 30},
          mean: 0.85650, std: 0.00204, params: {'max depth': 2, 'n estimators': 50},
          mean: 0.85800, std: 0.00450, params: {'max depth': 2, 'n estimators': 70},
          mean: 0.85825, std: 0.00455, params: {'max depth': 2, 'n estimators': 100},
          mean: 0.85588, std: 0.00390, params: {'max depth': 3, 'n estimators': 30},
          mean: 0.86087, std: 0.00615, params: {'max depth': 3, 'n estimators': 50},
          mean: 0.86038, std: 0.00522, params: {'max depth': 3, 'n estimators': 70},
          mean: 0.86075, std: 0.00586, params: {'max depth': 3, 'n estimators': 100},
          mean: 0.85975, std: 0.00527, params: {'max_depth': 4, 'n estimators': 30},
          mean: 0.86087, std: 0.00606, params: {'max depth': 4, 'n estimators': 50},
          mean: 0.86287, std: 0.00540, params: {'max depth': 4, 'n estimators': 70},
          mean: 0.86513, std: 0.00589, params: {'max depth': 4, 'n estimators': 100}]
In [34]: error = []
         eachError = []
         for i in range(len(scoreData)):
             eachError.append(np.round((1 - scoreData[i][1]), 4))
             if i == 3 or i == 7 or i == 11 or i == 15:
                 error.append(eachError)
                 eachError = []
In [35]:
         error
Out[35]: [[0.1565, 0.1539, 0.1496, 0.1448],
          [0.148, 0.1435, 0.142, 0.1418],
          [0.1441, 0.1391, 0.1396, 0.1392],
          [0.1402, 0.1391, 0.1371, 0.1349]]
```

```
In [36]:
          columnNames = [30, 50, 70, 100]
          errorFrame = pd.DataFrame(error, columns = columnNames)
In [37]:
In [38]:
          errorFrame
Out[38]:
                30
                       50
                              70
                                    100
          0 0.1565 0.1539 0.1496 0.1448
           1 0.1480 0.1435 0.1420 0.1418
           2 0.1441 0.1391 0.1396 0.1392
           3 0.1402 0.1391 0.1371 0.1349
In [39]:
          indexNames = [1, 2, 3, 4]
          errorFrame["Max depth"] = indexNames
In [40]:
          errorFrame
Out[40]:
                                    100 Max_depth
                30
                       50
                              70
          0 0.1565 0.1539 0.1496 0.1448
                                                1
           1 0.1480 0.1435 0.1420 0.1418
                                                2
           2 0.1441 0.1391 0.1396 0.1392
           3 0.1402 0.1391 0.1371 0.1349
                                                4
          errorFrame.set_index("Max_depth", append = False, drop = True, inplace = True)
In [41]:
```

In [42]: errorFrame

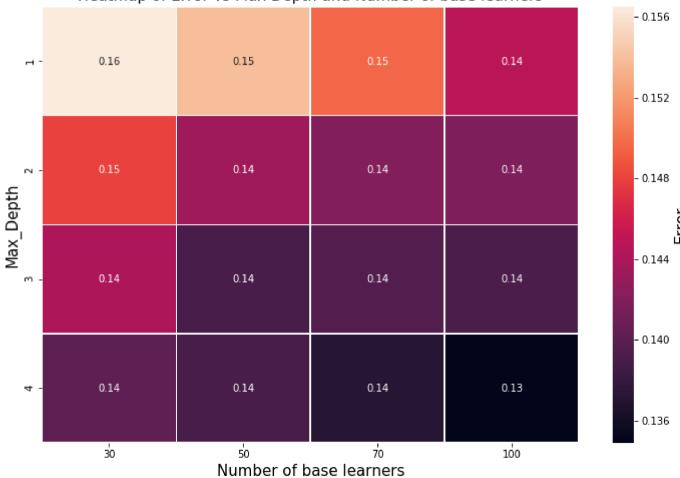
Out[42]:

	30	50	70	100
Max_depth				
1	0.1565	0.1539	0.1496	0.1448
2	0.1480	0.1435	0.1420	0.1418
3	0.1441	0.1391	0.1396	0.1392
4	0.1402	0.1391	0.1371	0.1349

```
In [43]:
         plt.figure(figsize=(12, 8))
         plt.title("Heatmap of Error vs Max Depth and Number of base learners", size = 15)
         ax = sns.heatmap(errorFrame, annot = True, linewidths=.5)
         ax.figure.axes[0].set xlabel('Number of base learners', size = 15)
         ax.figure.axes[0].set ylabel('Max Depth', size = 15)
         ax.figure.axes[-1].set_ylabel('Error', size = 15)
```

Out[43]: Text(738.529,0.5,'Error')

Heatmap of Error vs Max Depth and Number of base learners



Here above in heatmap you can see that the error is minimum when Number of base learners are 100 and Max Depth value is 4. Therefore,

from grid search we are considering our Number of base learners value as 100 and Max Depth Value as 4.

Task 5. Apply XGBoost and report accuracy. Also check for train error.

```
In [45]: #checking for train error
         model xg tr = xgb.XGBClassifier(max depth = 4, n estimators = 100)
         model xg tr.fit(train AvgW2V, train labels AvgW2V)
         prediction AvgW2V xg tr = model xg tr.predict(train AvgW2V)
         AccuracyScore AvgW2V xg tr = accuracy score(train labels AvgW2V, prediction AvgW2V xg tr)
         print("Train Error of XGBoost = "+str((1 - AccuracyScore AvgW2V xg tr) * 100)+"%")
         Train Error of XGBoost = 8.1125%
         C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of
         an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check
         that an array is not empty.
           if diff:
         model xg = xgb.XGBClassifier(max depth = 4, n estimators = 100)
In [49]:
         model xg.fit(train AvgW2V, train labels AvgW2V)
         prediction xg = model_xg.predict(test_AvgW2V)
         AccuracyScore xg = accuracy score(test labels AvgW2V, prediction xg) * 100
         print("Accuracy score of XGBoost = "+str(AccuracyScore xg)+"%")
         Accuracy score of XGBoost = 86.55000000000001%
```

C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check that an array is not empty.

2. Average TFIDF-W2V.

```
In [152]:
          # TF-IDF weighted Word2Vec
          tfidf features = tfidf vect.get feature names()
          tfidf w2v = []
          reviews = 0
          for sentence in listOfSentences:
               sentenceVector = np.zeros(300)
              weightTfidfSum = 0
              for word in sentence:
                   try:
                       W2V Vector = w2v Model.wv[word]
                       tfidfVector = tfidf[reviews, tfidf features.index(word)]
                       sentenceVector += (W2V Vector * tfidfVector)
                       weightTfidfSum += tfidfVector
                   except:
                       pass
              sentenceVector /= weightTfidfSum
              tfidf w2v.append(sentenceVector)
               reviews += 1
```

Task 1. Split train and test data in a ratio of 80:20.

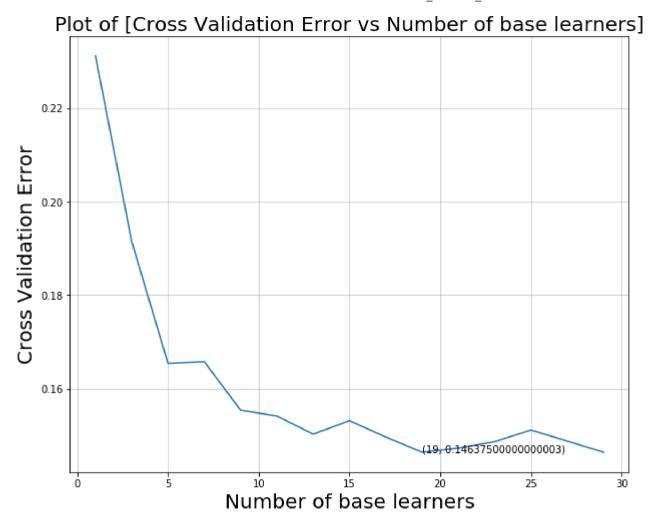
```
In [153]: X train TFIDF W2V, X test TFIDF W2V, Y train labels TFIDF W2V, Y test labels TFIDF W2V = train test split(tfidf w2v, Data
In [154]:
          train TFIDF W2V = np.array(X train TFIDF W2V)
          test TFIDF W2V = np.array(X test TFIDF W2V)
          train labels TFIDF W2V = np.array(Y train labels TFIDF W2V)
          test labels TFIDF W2V = np.array(Y test labels TFIDF W2V)
In [155]: train TFIDF W2V.shape, test TFIDF W2V.shape, train labels TFIDF W2V.shape, test labels TFIDF W2V.shape
Out[155]: ((8000, 300), (2000, 300), (8000,), (2000,))
```

Task 2. Perform GridSearch Cross Validation to find optimal value of number of base models in Random Forest.

```
In [156]: values = []
          for i in range(1, 31, 2):
              values.append(i)
          clf = RandomForestClassifier(n jobs = -1)
          hyper parameters = {'n estimators': values}
          bestScore = GridSearchCV(clf, hyper parameters, scoring = "accuracy", cv = 3)
          bestScore.fit(train TFIDF W2V, train labels TFIDF W2V)
          print(bestScore.best estimator )
          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=19, n jobs=-1,
                      oob score=False, random state=None, verbose=0,
                      warm start=False)
In [157]: best parameter = bestScore.best params
          bp = best parameter["n estimators"]
          bp
Out[157]: 19
```

```
In [158]:
          scoreData = bestScore.grid scores
           scoreData
Out[158]: [mean: 0.76875, std: 0.01000, params: {'n estimators': 1},
           mean: 0.80850, std: 0.00427, params: {'n estimators': 3},
           mean: 0.83462, std: 0.00339, params: {'n estimators': 5},
           mean: 0.83425, std: 0.00236, params: {'n estimators': 7},
           mean: 0.84462, std: 0.00230, params: {'n estimators': 9},
           mean: 0.84588, std: 0.00193, params: {'n estimators': 11},
           mean: 0.84975, std: 0.00361, params: {'n estimators': 13},
           mean: 0.84688, std: 0.00181, params: {'n estimators': 15},
           mean: 0.85037, std: 0.00317, params: {'n estimators': 17},
           mean: 0.85362, std: 0.00073, params: {'n estimators': 19},
           mean: 0.85275, std: 0.00297, params: {'n estimators': 21},
           mean: 0.85137, std: 0.00131, params: {'n estimators': 23},
           mean: 0.84888, std: 0.00280, params: {'n estimators': 25},
           mean: 0.85125, std: 0.00277, params: {'n estimators': 27},
           mean: 0.85362, std: 0.00154, params: {'n estimators': 29}]
```

```
In [159]: error = []
          parameter = []
          for i in range(len(scoreData)):
              error.append(1 - scoreData[i][1])
              parameter.append(scoreData[i][0]["n estimators"])
          plt.figure(figsize=(10,8))
          plt.plot(parameter, error)
          plt.title("Plot of [Cross Validation Error vs Number of base learners]", fontsize=20)
          plt.xlabel("Number of base learners", fontsize=20)
          plt.ylabel("Cross Validation Error", fontsize=20)
          plt.grid(linestyle='-', linewidth=0.5)
          errorMin = min(error)
           errorMin
          for xy in zip(parameter, error):
              if xy == (bp, errorMin):
                  plt.annotate(xy, xy)
```



Task 3. Apply Random Forest and report accuracy. Also check for train error.

```
#First checking for train error
In [160]:
          model rf tr = RandomForestClassifier(n estimators = bp, n jobs = -1)
          model rf tr.fit(train TFIDF W2V, train labels TFIDF W2V)
          prediction TFIDF W2V rf tr = model rf tr.predict(train TFIDF W2V)
          AccuracyScore TFIDF W2V rf tr = accuracy score(train labels TFIDF W2V, prediction TFIDF W2V rf tr)
          print("Train error of Random Forest = "+str(np.round(((1 - AccuracyScore TFIDF W2V rf tr) * 100), 4))+"%")
          Train error of Random Forest = 0.225%
          model rf = RandomForestClassifier(n estimators = bp, n jobs = -1)
In [161]:
          model rf.fit(train TFIDF W2V, train labels TFIDF W2V)
          prediction TFIDF W2V rf = model rf.predict(test TFIDF W2V)
          AccuracyScore_TFIDF_W2V_rf = accuracy_score(test_labels_TFIDF W2V, prediction TFIDF W2V rf) * 100
          print("Accuracy score of Random Forest = "+str(AccuracyScore TFIDF W2V rf)+"%")
          Accuracy score of Random Forest = 84.1%
          Confusion Matrix = confusion matrix(test labels TFIDF W2V, prediction TFIDF W2V rf)
In [162]:
          print("Confusion Matrix on L2 regularization \n"+str(Confusion Matrix))
          Confusion Matrix on L2 regularization
          [[ 53 272]
           [ 46 1629]]
In [163]: tn, fp, fn, tp = confusion matrix(test labels TFIDF W2V, prediction TFIDF W2V rf).ravel()
          tn, fp, fn, tp
Out[163]: (53, 272, 46, 1629)
```

Task 4. Perform GridSearch Cross Validation to find optimal number of base models and optimal value of depth of decision tree base models in XGBoost.

```
xgb model = xgb.XGBClassifier()
In [164]:
          hyperParameters = { 'max depth':[1, 2, 3, 4], 'n estimators':[30, 50, 70, 100]}
          clf = GridSearchCV(estimator = xgb model, param grid = hyperParameters, scoring='accuracy', cv=3)
          clf.fit(train TFIDF W2V, train labels TFIDF W2V)
          print(clf.best estimator )
```

C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check that an array is not empty.

if diff:

C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check that an array is not empty.

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if diff:

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if diff:

XGBClassifier(base score=0.5, booster='gbtree', colsample bylevel=1, colsample bytree=1, gamma=0, learning rate=0.1, max delta step=0, max depth=4, min child weight=1, missing=None, n estimators=70, n jobs=1, nthread=None, objective='binary:logistic', random state=0, reg alpha=0, reg lambda=1, scale pos weight=1, seed=None, silent=True, subsample=1)

```
best parameter = clf.best params
In [165]:
           best parameter
Out[165]: {'max depth': 4, 'n estimators': 70}
In [166]: | scoreData = clf.grid scores
           scoreData
Out[166]: [mean: 0.84500, std: 0.00012, params: {'max depth': 1, 'n estimators': 30},
           mean: 0.84500, std: 0.00012, params: {'max depth': 1, 'n estimators': 50},
           mean: 0.84950, std: 0.00154, params: {'max depth': 1, 'n estimators': 70},
           mean: 0.85175, std: 0.00108, params: {'max depth': 1, 'n estimators': 100},
           mean: 0.84975, std: 0.00204, params: {'max depth': 2, 'n estimators': 30},
           mean: 0.85488, std: 0.00157, params: {'max depth': 2, 'n estimators': 50},
           mean: 0.85675, std: 0.00294, params: {'max depth': 2, 'n estimators': 70},
           mean: 0.85888, std: 0.00327, params: {'max depth': 2, 'n estimators': 100},
           mean: 0.85600, std: 0.00245, params: {'max depth': 3, 'n estimators': 30},
           mean: 0.85813, std: 0.00413, params: {'max depth': 3, 'n estimators': 50},
           mean: 0.85950, std: 0.00468, params: {'max depth': 3, 'n estimators': 70},
           mean: 0.86150, std: 0.00348, params: {'max depth': 3, 'n estimators': 100},
           mean: 0.85788, std: 0.00277, params: {'max_depth': 4, 'n estimators': 30},
           mean: 0.86087, std: 0.00370, params: {'max depth': 4, 'n estimators': 50},
           mean: 0.86413, std: 0.00645, params: {'max depth': 4, 'n estimators': 70},
           mean: 0.86200, std: 0.00447, params: {'max depth': 4, 'n estimators': 100}]
          error = []
In [167]:
           eachError = []
           for i in range(len(scoreData)):
              eachError.append(np.round((1 - scoreData[i][1]), 4))
              if i == 3 or i == 7 or i == 11 or i == 15:
                   error.append(eachError)
                   eachError = []
In [168]:
          error
Out[168]: [[0.155, 0.155, 0.1505, 0.1482],
           [0.1502, 0.1451, 0.1432, 0.1411],
           [0.144, 0.1419, 0.1405, 0.1385],
           [0.1421, 0.1391, 0.1359, 0.138]]
```

```
In [169]:
           columnNames = [30, 50, 70, 100]
In [170]:
           errorFrame = pd.DataFrame(error, columns = columnNames)
           errorFrame
In [171]:
Out[171]:
                 30
                        50
                               70
                                     100
           0 0.1550 0.1550 0.1505 0.1482
            1 0.1502 0.1451 0.1432 0.1411
            2 0.1440 0.1419 0.1405 0.1385
            3 0.1421 0.1391 0.1359 0.1380
In [172]:
           indexNames = [1, 2, 3, 4]
           errorFrame["Max depth"] = indexNames
In [173]:
           errorFrame
Out[173]:
                                     100 Max_depth
                 30
                        50
                               70
           0 0.1550 0.1550 0.1505 0.1482
                                                 1
            1 0.1502 0.1451 0.1432 0.1411
                                                 2
            2 0.1440 0.1419 0.1405 0.1385
            3 0.1421 0.1391 0.1359 0.1380
                                                 4
           errorFrame.set_index("Max_depth", append = False, drop = True, inplace = True)
In [174]:
```

In [175]: errorFrame

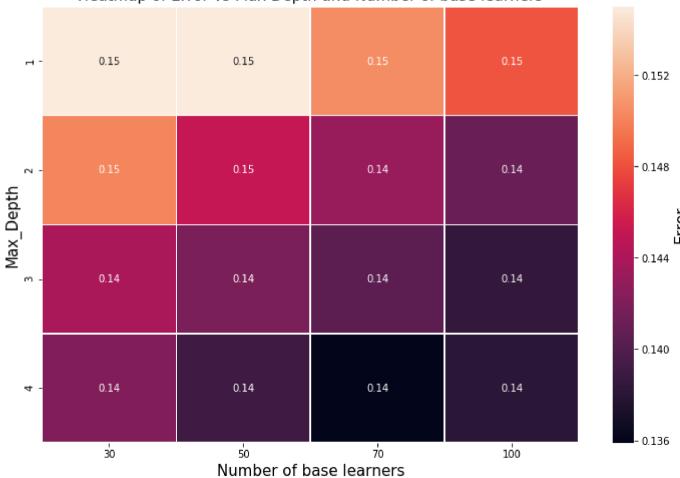
Out[175]:

	30	50	70	100
Max_depth				
1	0.1550	0.1550	0.1505	0.1482
2	0.1502	0.1451	0.1432	0.1411
3	0.1440	0.1419	0.1405	0.1385
4	0.1421	0.1391	0.1359	0.1380

```
In [176]: plt.figure(figsize=(12, 8))
   plt.title("Heatmap of Error vs Max Depth and Number of base learners", size = 15)
   ax = sns.heatmap(errorFrame, annot = True, linewidths=.5)
   ax.figure.axes[0].set_xlabel('Number of base learners', size = 15)
   ax.figure.axes[0].set_ylabel('Max_Depth', size = 15)
  ax.figure.axes[-1].set_ylabel('Error', size = 15)
```

Out[176]: Text(738.529,0.5,'Error')

Heatmap of Error vs Max Depth and Number of base learners



Here above in heatmap you can see that the error is minimum when Number of base learners are 70 and Max_Depth value is 4. Therefore,

from grid search we are considering our Number of base learners value as 70 and Max Depth Value as 4.

Task 5. Apply XGBoost and report accuracy. Also check for train error.

```
#checking for train error
In [177]:
          model xg tr = xgb.XGBClassifier(max depth = 4, n estimators = 70)
          model xg tr.fit(train TFIDF W2V, train labels TFIDF W2V)
          prediction TFIDF W2V xg tr = model xg tr.predict(train TFIDF W2V)
          AccuracyScore TFIDF W2V xg tr = accuracy score(train labels TFIDF W2V, prediction TFIDF W2V xg tr)
          print("Train Error of XGBoost = "+str((1 - AccuracyScore TFIDF W2V xg tr) * 100)+"%")
          Train Error of XGBoost = 10.09999999999998%
          C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of
          an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check
          that an array is not empty.
            if diff:
          model xg = xgb.XGBClassifier(max depth = 4, n estimators = 70)
In [179]:
          model xg.fit(train TFIDF W2V, train labels TFIDF W2V)
          prediction xg = model xg.predict(test TFIDF W2V)
          AccuracyScore xg = accuracy score(test labels TFIDF W2V, prediction xg) * 100
          print("Accuracy score of XGBoost = "+str(AccuracyScore xg)+"%")
          Accuracy score of XGBoost = 84.65%
```

C:\Users\GauravP\Anaconda3\lib\site-packages\sklearn\preprocessing\label.py:151: DeprecationWarning: The truth value of an empty array is ambiguous. Returning False, but in future this will result in an error. Use `array.size > 0` to check that an array is not empty.

Summary

Avg W2V

- 1. Optimal Value of number of base learners in Random Forest from Grid Search = 29
- 2. Train Error in Random Forest = 0.1%
- 3. Accuracy of Random Forest = 85.25%
- 4. Optimal Value of number of base learners and maximum depth of base learners in XGBoost from Grid Search = 100 & 4
- 5. Train Error in XGBoost 8.11%
- 6. Accuracy of XGBoost = 86.55%

TFIDF-W2V

- 1. Optimal Value of number of base learners in Random Forest from Grid Search = 19
- 2. Train Error in Random Forest = 0.225%
- 3. Accuracy of Random Forest = 84.1%
- 4. Optimal Value of number of base learners and maximum depth of base learners in XGBoost from Grid Search = 70 & 4
- 5. Train Error in XGBoost 10.09%
- 6. Accuracy of XGBoost = 84.65%