#### Information about data:

- ->We have the amazon reviews dataset from kaggle
- ->Reviews are given for the product
- ->The features of the data were:

Ιd

ProductId- unique identifier for the product

UserId- unqiue identifier for the user

ProfileName

 $\label{eq:helpfullnessNumerator-number of users who foliand the review helpful} HelpfullnessNumerator- number of users who foliated the second seco$ 

 ${\tt HelpfulnessDenominator-\ number\ of\ users\ who\ ir}$  dicated whether they found the review

helpful or not

Score-rating between 1 and 5

Time-timestamp for the review

Summary- brief summary of the review

Text- text of the review

 $\,$  -> Based on the score of the review we review we classify them into positive and negative

Number of reviews: 568,454

# Objective:

- -> Cleaning the dataset by classifying them into positive and negati ve reviews based on the rating provided and removing the duplicates
- -> Converting the text data to vectors by using Bag of words, Tfidf, word2vec, Average word2vec
- -> Implementing Support Vector Machines with kernel as RBF

- -> Implementing Linear Kernel SVM
- -> Implementing nu-SVM
- -> Using Grid Search and Random Search th determine the best hyper p arameters
- -> To check the performance with different "C" and "GAMMA" values

# Importing the required libraries to process the data

#### In [1]:

```
from sklearn.feature_extraction.text import
TfidfTransformer,TfidfVectorizer,CountVectorizer
import sqlite3
import pandas as pd
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import RandomizedSearchCV
from sklearn.svm import SVC
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
```

- -> Establishing the connection and loading the data
- -> Shape of the data
- -> Dimensionality of the data
- -> Attributes of the data
- -> Sample of data

#### In [2]:

```
connection = sqlite3.connect("database.sqlite")
data = pd.read_sql_query("SELECT * FROM Reviews WHERE Score != 3",connectio
print(data.shape)
print (data.ndim)
print(data.columns)
print(data.head(5))
(525814, 10)
Index(['Id', 'ProductId', 'UserId', 'ProfileName', 'HelpfulnessNumerator',
       'HelpfulnessDenominator', 'Score', 'Time', 'Summary', 'Text'],
      dtype='object')
  Id ProductId
                          UserId
                                                      ProfileName
0
   1 B001E4KFG0 A3SGXH7AUHU8GW
                                                       delmartian
1
   2 B00813GRG4 A1D87F6ZCVE5NK
                                                           dll pa
                  ABXLMWJIXXAIN Natalia Corres "Natalia Corres"
2
   3 B000LQOCH0
3
  4 B000UA0QIQ A395BORC6FGVXV
                                                             Karl
  5 B006K2ZZ7K A1UQRSCLF8GW1T Michael D. Bigham "M. Wassir"
```

```
HelpfulnessNumerator HelpfulnessDenominator Score
                                                              Time
0
                                                       1303862400
                                                     5
                                              1
1
                      0
                                              0
                                                     1
                                                       1346976000
2
                      1
                                              1
                                                       1219017600
3
                      3
                                              3
                                                     2
                                                       1307923200
4
                      0
                                              0
                                                     5
                                                       1350777600
                 Summary
                                                                       Text
  Good Quality Dog Food I have bought several of the Vitality canned d...
1
      Not as Advertised Product arrived labeled as Jumbo Salted Peanut...
2
   "Delight" says it all This is a confection that has been around a fe...
3
          Cough Medicine If you are looking for the secret ingredient i...
4
            Great taffy Great taffy at a great price. There was a wid...
```

# Data Pre-processing:

- -> Cleaning the data
- -> Removing duplicates

# In [3]:

```
def score(n):
    if n>3:
        return 'positive'
    return 'negative'
rating = data['Score']
rating = rating.map(score)
data['Score'] = rating
data.head(6)
```

# Out[3]:

	ld	ProductId	Userld	ProfileName	HelpfulnessNumerator	Helpfulness
0	1	B001E4KFG0	A3SGXH7AUHU8GW	delmartian	1	1
1	2	B00813GRG4	A1D87F6ZCVE5NK	dll pa	0	0

2	βd	воо <b>втофсно</b>	ABXLMWJIXX <b>Aljserid</b>	Natalia Corres ProfileName "Natalia Corres"	elpfulnessNumerator	⊭elpfulness		
				Natalia Golfee				
3	4	B000UA0QIQ	A395BORC6FGVXV	Karl	3	3		
4	5	B006K2ZZ7K	A1UQRSCLF8GW1T	Michael D. Bigham "M. Wassir"	0	0		
5	6	B006K2ZZ7K	ADT0SRK1MGOEU	Twoapennything	0	0		
4								

# In [4]:

```
sorting_data = data.sort_values('ProductId',axis=0,ascending=True,inplace=F
alse,kind='quicksort',na_position='last')
```

# In [5]:

```
cleaned_data = sorting_data.drop_duplicates(subset={"UserId","ProfileName",
"Time","Text"}, keep='first',inplace=False)
```

# Information about the cleaned data:

- -> Shape of the data
- -> Dimensionality of the data
- -> Attributes of the data
- -> Sample of cleaned data

# In [6]:

```
print(cleaned_data.shape)
print(cleaned_data.ndim)
print(cleaned_data.columns)
print(cleaned_data.head(5))
```

(364173, 10)

```
2
Index(['Id', 'ProductId', 'UserId', 'ProfileName', 'HelpfulnessNumerator',
      'HelpfulnessDenominator', 'Score', 'Time', 'Summary', 'Text'],
     dtype='object')
           Id
               ProductId
                                                          ProfileName
                                  UserId
138706 150524 0006641040 ACITT7DI6IDDL
                                                      shari zychinski
138688 150506 0006641040 A2IW4PEEKO2ROU
                                                                Tracy
138689 150507 0006641040 A1S4A3IQ2MU7V4 sally sue "sally sue"
                            AZGXZ2UUK6X Catherine Hallberg "(Kate)"
138690 150508 0006641040
138691 150509 0006641040 A3CMRKGE0P909G
       HelpfulnessNumerator HelpfulnessDenominator
                                                                    Time
                                                       Score
138706
                          0
                                                 0 positive 939340800
                          1
138688
                                                 1 positive 1194739200
138689
                          1
                                                 1 positive 1191456000
                                                 1 positive 1076025600
138690
                          1
138691
                          3
                                                 4 positive 1018396800
                                         Summary \
138706
                        EVERY book is educational
138688 Love the book, miss the hard cover version
                    chicken soup with rice months
138689
138690
           a good swingy rhythm for reading aloud
                  A great way to learn the months
138691
                                                   Text
138706 this witty little book makes my son laugh at l...
138688 I grew up reading these Sendak books, and watc...
138689 This is a fun way for children to learn their ...
138690 This is a great little book to read aloud- it ...
138691 This is a book of poetry about the months of t...
```

Keeping in mind about the performance capability of the box and time efficiency here i am taking the subset of the data

```
In [7]:
```

```
sample_data = cleaned_data.sample(n=100000)
```

#### Information about the sampled data:

- -> Shape of the data
- -> Dimensionality of the data
- -> Attributes if the data
- -> Sample of modified data

# In [8]:

```
print(sample_data.shape)
print(sample_data.ndim)
print(sample_data.columns)
print(sample_data.head(5))
```

```
(100000, 10)
2
Index(['Id', 'ProductId', 'UserId', 'ProfileName', 'HelpfulnessNumerator',
       'HelpfulnessDenominator', 'Score', 'Time', 'Summary', 'Text'],
      dtype='object')
           Ιd
               ProductId
                                   UserId
                                                          ProfileName
255735 277245 B000VK8AVK A1HLYNGUBXXW9S
                                            Maureen in WA "maureeng"
173730 188447 B0051ZCRGY A3L601NV34AJ2J Richard Odato "Slingboxer"
282621 306193 B000JSM344 A3E5D99WX49BK7
                                                        Michelle Cohn
       17583 B0000GH6UQ A18ZDBQ8LUNVO6
16096
                                                   Jessica McCormick
477841 516720 B007WTQAKQ A2YAAF9BCQ4AJC
                                                      Leigh Somebody
       HelpfulnessNumerator HelpfulnessDenominator
                                                        Score
                                                                    Time
255735
                         16
                                                 17 positive 1215561600
173730
                          0
                                                    positive 1345334400
                          3
282621
                                                  5 positive 1199750400
16096
                          0
                                                  0 positive 1318377600
                          0
477841
                                                  0 positive 1337904000
                           Summary \
                  Excellent Chips!
255735
173730 Our Cocker LOVES Pegetables
                Thank you so much!
282621
              So long, Swiss Miss!
16096
477841
               Surprisingly Works!
                                                    Text
255735 These chips are really, really good. I don't ...
173730 Our Cocker LOVES Pegetables! She seems to rea...
282621 I use to be able to buy this in town and when ...
16096
      A couple years ago, my husband pointed this co...
477841 Just on a whim, I thought I'd try this and it ...
```

#### SVM WITH RBF KERNEL:

#### **BAG OF WORDS:**

TIME BASED SPLITTING OF DATA:

- -> Implementing Support Vector Classifier with "RBF" as kernel
- -> Using both Grid Search and Random Search to determine the be st hyper parameters
- $\rightarrow$  To check performance measure with different hyper parameter values

# In [9]:

```
cv_data = sample_data.sort_values("Time",axis=0,ascending=True,kind='quicks
ort',na_position='last',inplace=False)
```

#### In [10]:

```
count vectorizer = Countvectorizer()
cv data vect = count vectorizer.fit transform(cv data['Text'].values)
In [11]:
score = sample data['Score']
In [12]:
cv data vect.shape
Out[12]:
(100000, 61485)
In [13]:
xtrain = cv data vect[0:70000]
xtest = cv data vect[70000:]
ytrain = score[0:70000]
ytest = score[70000:]
In [14]:
print(xtrain.shape)
print(xtest.shape)
print(ytrain.shape)
print(ytest.shape)
(70000, 61485)
(30000, 61485)
(70000,)
(30000,)
BAG OF WORDS: GRID SEARCH TO FIND HYPER PARAMETERS
In [15]:
parameters = {'kernel':['rbf'], 'C':[0.01,0.1,1],'gamma':[0.01,0.1,1]}
In [16]:
classifier = SVC()
In [17]:
model = GridSearchCV(classifier,param grid=parameters,scoring='accuracy')
model.fit(xtrain,ytrain)
Out[17]:
GridSearchCV(cv=None, error score='raise',
       estimator=SVC(C=1.0, cache size=200, class weight=None, coef0=0.0,
 decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
 max iter=-1, probability=False, random state=None, shrinking=True,
  tol=0.001, verbose=False),
       fit params=None, iid=True, n jobs=1,
       param grid={'kernel': ['rbf'], 'C': [0.01, 0.1, 1], 'gamma': [0.01,
0.1, 1]},
       pre dispatch='2*n jobs', refit=True, return train score='warn',
       scoring='accuracy', verbose=0)
```

```
In [26]:
classif = SVC(C=1,gamma=1,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy score(ytest,pred)
print(acc)
0.8419
In [27]:
classif = SVC(C=0.1,gamma=1,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print(acc)
0.8419
In [28]:
classif = SVC(C=0.1,gamma=0.1,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print(acc)
0.8419
In [29]:
classif = SVC(C=0.1,gamma=0.01,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print(acc)
0.8419
In [30]:
classif = SVC(C=0.01, gamma=1, kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print (acc)
0.8419
BAG OF WORDS: RANDOM SEARCH TO FIND HYPER PARAMETERS
In [15]:
import scipy
In [16]:
parameters = {'C': scipy.stats.randint.rvs(1,5,size=5), 'gamma':
scipy.stats.randint.rvs(1,5,size=5),'kernel': ['rbf']}
```

```
In [17]:
from sklearn.linear model import SGDClassifier
In [20]:
classifier = SVC()
In [24]:
model = RandomizedSearchCV(classifier,param distributions=parameters,scorin
g='accuracy',cv=5,n iter=5,n jobs=-1)
model.fit(xtrain,ytrain)
print(model.best estimator )
SVC(C=3, cache size=200, class weight=None, coef0=0.0,
 decision function shape='ovr', degree=3, gamma=2, kernel='rbf',
 max iter=-1, probability=False, random state=None, shrinking=True,
  tol=0.001, verbose=False)
In [25]:
model
Out [25]:
RandomizedSearchCV(cv=5, error score='raise',
          estimator=SVC(C=1.0, cache size=200, class weight=None,
coef0=0.0,
 decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
 max iter=-1, probability=False, random state=None, shrinking=True,
 tol=0.001, verbose=False),
          fit params=None, iid=True, n iter=5, n jobs=-1,
          param distributions={'C': array([2, 2, 3, 3, 4]), 'gamma':
array([3, 4, 3, 2, 4]), 'kernel': ['rbf']},
          pre_dispatch='2*n_jobs', random_state=None, refit=True,
          return_train_score='warn', scoring='accuracy', verbose=0)
In [26]:
model.best estimator
Out [26]:
SVC(C=3, cache size=200, class weight=None, coef0=0.0,
 decision function shape='ovr', degree=3, gamma=2, kernel='rbf',
 max iter=-1, probability=False, random state=None, shrinking=True,
 tol=0.001, verbose=False)
In [27]:
classif = SVC(C=3,gamma=2,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy score(ytest,pred)
print(acc)
0.8424666666666667
In [28]:
```

classif = SVC(C=2.gamma=2.kernel='rbf')

```
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print(acc)
```

#### 0.8424666666666667

#### In [29]:

```
classif = SVC(C=1,gamma=2,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print(acc)
```

#### 0.8424666666666667

# In [30]:

```
classif = SVC(C=3,gamma=1,kernel='rbf')
classif.fit(xtrain,ytrain)
pred = classif.predict(xtest)
acc = accuracy_score(ytest,pred)
print(acc)
```

#### 0.8424666666666667

#### Observation:

- -> BAG OF WORDS:
- $\rightarrow$  By using Grid Search with "RBF" as kernel the best values of hyper parameters were C = 1 and gamma = 3
- $\rightarrow$  By using Random search with "RBF" as kernel the best values of hyper parameters were C = 3 and gamma = 2
  - -> The accurancy by grid search is 84.19
  - -> The accurancy of random search is 84.24

# SVM WITH LINEAR KERNEL:

#### TFIDF:

TIME BASED SPLITTING OF DATA:

- -> Implementing Support Vector Classifier with Linear kernel
- $\,$  -> Using both Grid Search and Random Search to determine the be st hyper parameters

```
In [31]:
tfid = TfidfVectorizer(ngram range=(1,2))
In [32]:
tfid data = sample data.sort values("Time", axis=0, ascending=True, kind='quic
ksort', na position='last', inplace=False)
In [33]:
tfid vect data = tfid.fit transform(tfid data['Text'].values)
In [34]:
tfid vect data.shape
Out[34]:
(100000, 1283627)
In [36]:
sample = tfid_vect_data[0:10000]
score = tfid data['Score'][0:10000]
In [37]:
xtrain = sample[0:7000]
xtest = sample[7000:]
ytrain = score[0:7000]
ytest = score[7000:]
In [38]:
print(xtrain.shape)
print(ytrain.shape)
print(xtest.shape)
print(ytest.shape)
(7000, 1283627)
(7000,)
(3000, 1283627)
(3000,)
TFIDF: GRID SEARCH TO FIND HYPER PARAMETERS
In [54]:
classifier = SGDClassifier(loss="hinge")
In [55]:
parameters = { 'alpha': [0.01, 0.1, 1, 10, 100] }
In [57]:
model = GridSearchCV(classifier,param grid=parameters,scoring='accuracy',cv
model.fit(xtrain, vtrain)
```

```
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print(model)
print (score)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
{\bf x} iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max_iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max_iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear_model.stochastic_gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
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, max iter defaults to max iter=1000. From 0.21, default max iter will be 1\,
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
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are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1\,
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
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are left unset, they default to max_iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
```

```
"and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
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are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1\,
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
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are left unset, they default to max_iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
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  "and default tol will be 1e-3." % type(self), FutureWarning)
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packages/sklearn/linear_model/stochastic_gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max_iter=5 and tol=None. If tol is not None
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  "and default tol will be 1e-3." % type(self), FutureWarning)
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are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1\,
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
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'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max_iter defaults to max_iter=1000. From 0.21, default max_iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
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'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
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'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max_iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
```

```
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
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are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
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are left unset, they default to max iter=5 and tol=None. If tol is not None
, max_iter defaults to max_iter=1000. From 0.21, default max_iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
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are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x_iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear_model/stochastic_gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max_iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tel will be 10.2 " & time/celf). EntureManning)
```

```
"and delault tol will be re-3." % type(sell), futurewarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear_model.stochastic_gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max_iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x_{iter} and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max_iter defaults to max_iter=1000. From 0.21, default max_iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
GridSearchCV(cv=5, error score='raise',
       estimator=SGDClassifier(alpha=0.0001, average=False,
class weight=None, epsilon=0.1,
       eta0=0.0, fit_intercept=True, l1_ratio=0.15,
       learning rate='optimal', loss='hinge', max iter=None, n iter=None,
       n jobs=1, penalty='12', power t=0.5, random state=None,
       shuffle=True, tol=None, verbose=0, warm_start=False),
       fit params=None, iid=True, n jobs=1,
       param grid={'alpha': [0.01, 0.1, 1, 10, 100]},
       pre dispatch='2*n jobs', refit=True, return train score='warn',
       scoring='accuracy', verbose=0)
0.8936666666666667
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
/Users/vthumati/anaconda3/lib/python3.6/site-
packages/sklearn/linear model/stochastic gradient.py:128: FutureWarning: ma
x iter and tol parameters have been added in <class
'sklearn.linear model.stochastic gradient.SGDClassifier'> in 0.19. If both
are left unset, they default to max iter=5 and tol=None. If tol is not None
, max iter defaults to max iter=1000. From 0.21, default max iter will be 1\,
000, and default tol will be 1e-3.
  "and default tol will be 1e-3." % type(self), FutureWarning)
In [69]:
```

#### from sklearn.svm import LinearSVC

# In [71]:

```
model = LinearSVC(C = 0.0001)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy_score(ytest,pred)
print(model)
print(score)
```

LinearSVC(C=0.0001. class weight=None. dual=True. fit intercept=True.

```
induition to the transmission name, and it is induited in its
     intercept scaling=1, loss='squared hinge', max iter=1000,
     multi class='ovr', penalty='12', random state=None, tol=0.0001,
    verbose=0)
0.8936666666666667
In [72]:
model = SVC(C = 0.01)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print(model)
print(score)
SVC(C=0.01, cache size=200, class weight=None, coef0=0.0,
 decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
 max iter=-1, probability=False, random state=None, shrinking=True,
 tol=0.001, verbose=False)
0.8936666666666667
TFIDF: RANDOM SEARCH TO FIND HYPER PARAMETERS
In [74]:
parameters = {'C': scipy.stats.randint.rvs(1,10,size=5)}
parameters
Out [74]:
{'C': array([8, 1, 9, 4, 5])}
In [75]:
classifier = LinearSVC(loss='hinge')
model = RandomizedSearchCV(classifier,param distributions=parameters,scorin
g='accuracy',cv=5,n iter=5,n jobs=-1)
model.fit(xtrain,ytrain)
print(model.best estimator )
LinearSVC(C=8, class weight=None, dual=True, fit intercept=True,
     intercept_scaling=1, loss='hinge', max_iter=1000, multi_class='ovr',
     penalty='12', random state=None, tol=0.0001, verbose=0)
In [78]:
model = SVC(C = 8)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print (model)
SVC(C=8, cache size=200, class weight=None, coef0=0.0,
 decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
 max iter=-1, probability=False, random state=None, shrinking=True,
  tol=0.001, verbose=False)
In [81]:
print(score)
Out[81]:
```

```
87.14
```

```
In [79]:
```

```
model = SVC(C = 0.0001, gamma= 0.01)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy_score(ytest,pred)
print(model)
```

SVC(C=0.0001, cache\_size=200, class\_weight=None, coef0=0.0,
 decision\_function\_shape='ovr', degree=3, gamma=0.01, kernel='rbf',
 max\_iter=-1, probability=False, random\_state=None, shrinking=True,
 tol=0.001, verbose=False)

# In [82]:

```
print(score)
```

# Out[82]:

87.14

#### Observation:

#### TFIDF:

- $\rightarrow$  By using Grid Search with Linear SVM the best values of hyper par ameters were C = 0.001
- $\rightarrow$  By using Random search with Linear SVM the best values of hyper p arameters were C = 8
- -> The accurancy by grid search is 89.36
- -> The accurancy of random search is 87.14

#### WORD2VEC:

CONSTRUCTING VECTOR REPRESENTATION OF EACH IN THE DATA BY USING WORD2VEC

## In [83]:

```
import gensim
from gensim.models import word2vec
```

- -> Importing the required libraries
- -> Functions to clean the sentences

# In [84]:

```
import re
def cleanhtml(sentence):
```

```
clean = re.compile("<.*?>")
    cleantext = re.sub(clean, " ", sentence)
    return cleantext
def cleanpunct(sentence):
    cleanr = re.sub(r"[?|!|\|'|#|.|,|)|(|/]",r'',sentence)
    return cleanr
In [85]:
sorted w2vec = sample data.sort values("Time", axis=0, ascending=True, kind='q
uicksort', na position='last', inplace=False)
Information about the sorted data:
   -> Shape of the data
   -> Dimensionality of the data
   -> Attributes if the data
   -> Sample of modified data
In [86]:
print(sorted w2vec.shape)
print(sorted w2vec.ndim)
print(sorted w2vec.columns)
print(sorted w2vec.head(5))
(100000, 10)
2
Index(['Id', 'ProductId', 'UserId', 'ProfileName', 'HelpfulnessNumerator',
       'HelpfulnessDenominator', 'Score', 'Time', 'Summary', 'Text'],
```

```
dtype='object')
           Id ProductId
                                  UserId
                                                      ProfileName \
138683 150501 0006641040 AJ46FKXOVC7NR
                                               Nicholas A Mesiano
417838 451855 B00004CXX9 AJH6LUC1UT1ON The Phantom of the Opera
417883 451903 B00004CXX9 A2DEE7F9XKP3ZR
                                                           jerome
        1245 B00002Z754 A29Z5PI9BW2PU3
1146
                                                           Robbie
346115 374421 B00004CI84 A1FJOY14X3MUHE
                                                    Justin Howard
       HelpfulnessNumerator HelpfulnessDenominator Score
                                                              Time
                                                 2 positive 940809600
138683
                          2
417838
                          0
                                                 0 positive 946857600
417883
                          0
                                                 1 positive 959990400
                          7
1146
                                                   positive 961718400
                          2
346115
                                                 2 positive 966297600
                                                Summary
138683 This whole series is great way to spend time w...
417838
                                             FANTASTIC!
417883
                                               Research
                                          Great Product
1146
346115 A fresh, original film from master storyteller...
138683 I can remember seeing the show when it aired o...
417838 Beetlejuice is an excellent and funny movie. K...
```

```
417883 I'm getting crazy. Is it really impossible t...
        This was a really good idea and the final prod...
1146
346115 This is such a great film, I don't even know h...
                                                                          •
4
In [87]:
i=0
sentences list=[]
for sent in sorted w2vec['Text'].values:
    filtered sentences = []
    sent = cleanhtml(sent)
    for w in sent.split():
        for cleanedwords in cleanpunct(w).split():
            if (cleanedwords.isalpha()):
                filtered sentences.append(cleanedwords.lower())
    sentences list.append(filtered sentences)
In [88]:
print(len(sentences list))
print(type(sentences list))
100000
<class 'list'>
In [89]:
w2vmodel =
gensim.models.Word2Vec(sentences list,min count=4, size=200, workers=4)
-> Most similar word
-> Similarity between the words
-> Dimensionality representation of a word
In [91]:
print(w2vmodel.most similar("the"))
print(w2vmodel.similarity("this", 'these'))
print(w2vmodel.wv['hello'])
[('foil', 0.3661514222621918), ('printed', 0.3577941656112671), ('marked',
0.3473374545574188), ('metal', 0.34421518445014954), ('contents', 0.3353686
3327026367), ('loosely', 0.33394986391067505), ('divided',
0.3335522413253784), ('each', 0.3309260606765747), ('torn',
0.3271692097187042), ('this', 0.326387882232666)]
0.15672317384225404
[ 7.46376738e-02 -3.91671211e-02 -9.74655803e-03 1.17327280e-01
  4.30680700e-02 1.42935008e-01 1.69160645e-02 -9.99073833e-02
 -8.94245226e-03 7.31556639e-02 2.71437708e-02 8.49297866e-02
 -3.79108302e-02 -4.03551161e-02 1.79700971e-01 -8.83629546e-02
  6.21749647e-02 1.70396626e-01 -9.88507271e-02 -3.23529937e-03
 -1.54427374e-02 -3.88860442e-02 2.32633371e-02 -8.95061940e-02
 -5.09793237e-02 -8.53051022e-02 -2.16724798e-01 -8.22060779e-02
 1.60596505e-01 2.38274857e-01 6.77032471e-02 2.92483121e-02
 -1.52913973e-01 -8.54681283e-02 3.81469503e-02 1.63954377e-01
 -1.04502574e-01 -2.37784889e-02 2.73989085e-02 1.55599207e-01
 -2.77195079e-03 1.89030655e-02 4.68748361e-02 2.11649723e-02
```

```
3.29707451e-02 -9.35959592e-02 1.53364509e-01 8.30018222e-02
 1.56344399e-01 2.09414542e-01 -1.09019009e-02 -8.27383846e-02
-1.52124390e-01 2.04615861e-01 1.84260547e-01 1.78548589e-01
 1.75115541e-02 1.01964595e-02 2.85195210e-03 8.62797499e-02
1.57599911e-01 -6.92393258e-02  2.03184187e-01  3.93796858e-04
 2.88663972e-02 2.57268101e-01 -9.62323770e-02 -1.19149141e-01
1.88405007e-01 1.53981045e-01 9.49712843e-02 -7.32565746e-02
-4.71446812e-02 6.39423653e-02 -2.86645442e-02 -7.28903562e-02
 2.74329204e-02 1.42536268e-01 3.79656479e-02 -2.90902182e-02
-7.72973225e-02 -3.79364908e-01 1.44835740e-01 2.89077967e-01
 1.83294430e-01 -8.90121460e-02 -1.18728027e-01 -3.65421809e-02
1.50683001e-01 5.75584024e-02 6.35507181e-02 -6.93864524e-02
-7.18736127e-02 4.85871844e-02 -7.92968273e-02 -7.60947764e-02
1.75751209e-01 1.56454127e-02 4.42015640e-02 1.00735486e-01
 2.36615837e-02 1.41660385e-02 1.73439924e-03 3.85778993e-02
 1.47505954e-01 -5.57372011e-02 1.99366614e-01 1.77332927e-02
1.39017720e-02 5.20809088e-03 -3.51771899e-02 -1.91033900e-01
 1.33832600e-02 -9.94292200e-02 3.76986302e-02 9.38725099e-03
 2.85761002e-02 -4.29738127e-02 7.43370620e-04 -2.47689307e-01
 8.02268460e-02 2.27401834e-02 2.83477485e-01 1.24741815e-01
 5.13344586e-01 -1.04342841e-01 9.24930423e-02 1.10017419e-01
-3.39711964e-01 -3.63127738e-02 -6.21374734e-02 -9.39397067e-02
-4.27645631e-02 4.19991016e-02 1.27547383e-01 -1.16916738e-01
1.53501838e-01 -1.62372038e-01 -7.45334476e-02 -1.64226457e-01
 1.24249734e-01 2.45201990e-01 -2.82593012e-01 -2.52310280e-02
 5.01009524e-02 1.28962547e-01 3.16310897e-02 8.67929310e-02
 3.72633100e-01 2.05868393e-01 -1.23229228e-01 -2.31928229e-01
 1.47817180e-01 -1.47775665e-01 7.49128535e-02 9.20652598e-02
1.92804068e-01 -2.17391551e-01 -6.01991778e-03 8.68222117e-02
 4.75134291e-02 -1.32257745e-01 -2.10440829e-02 -1.56973869e-01
-6.19931072e-02 1.25609227e-02 -1.23218983e-01 2.53714114e-01
1.86279014e-01 2.62722641e-01 1.78902242e-02 -1.56033605e-01
 9.97112147e-05 8.21634233e-02 -1.82665229e-01 5.08229099e-02
-4.72846329e-02 -1.46754622e-01 5.48403300e-02 1.43918321e-01
 1.39083862e-01 2.09148712e-02 -8.22087154e-02 -6.02755025e-02
 6.76847994e-02 -1.21093611e-03 -9.16820243e-02 8.49751532e-02
-1.14586517e-01 1.89506263e-01 -1.04682297e-02 -5.79984039e-02
 2.53672209e-02 1.11608185e-01 -9.44144577e-02 -3.44396085e-02
 6.49087224e-03 1.61137000e-01 -1.84098765e-01 -6.27396954e-03]
```

/Users/vthumati/anaconda3/lib/python3.6/site-packages/ipykernel\_launcher.py:1: DeprecationWarning: Call to deprecated `most\_similar` (Method will be removed in 4.0.0, use self.wv.most\_similar() i

"""Entry point for launching an IPython kernel.

/Users/vthumati/anaconda3/lib/python3.6/site-

packages/ipykernel\_launcher.py:2: DeprecationWarning: Call to deprecated `s
imilarity` (Method will be removed in 4.0.0, use self.wv.similarity() inste
ad).

## Observation:

nstead).

- -> We have constructed the vector representation of each word
- -> Using this model to construct vector representation of each sente nce in average word2vec
- -> Each word in the model is of 200-dimensions

# Nu - Support Vector Classification

#### **AVERAGE WORD2VEC:**

TIME BASED SPLITTING OF DATA:

- -> Implementing Nu Support Vector Classifier
- -> Using both Grid Search and Random Search to determine the be st hyper parameters
- $\ \ ->$  To check performance measure with different hyper parameter values

# In [97]:

```
sent_vectors = []
for sent in sentences_list:
    sent_vec = np.zeros(200)
    cnt=0
    for word in sent:
        try:
        vec = w2vmodel.wv[word]
        sent_vec += vec
        cnt += 1
        except:
        pass
    sent_vec /= cnt
        sent_vectors.append(sent_vec)
print(len(sent_vectors))
print(len(sent_vectors[88888]))
```

100000

# In [121]:

```
sent_vectors = np.nan_to_num(sent_vectors)
```

# In [122]:

```
sample_vect = sent_vectors[0:10000]
sample_score = sorted_w2vec['Score'][0:10000]
```

#### In [123]:

```
xtrain = sample_vect[0:7000]
xtest = sample_vect[7000:]
ytrain = sample_score[0:7000]
ytest = sample_score[7000:]
```

#### In [141]:

```
print((xtrain.shape))
print((xtest.shape))
print(vtrain.shape)
```

```
print (ytrain.snape)
print(ytest.shape)
(7000, 200)
(3000, 200)
(7000,)
(3000,)
AVERAGE WORD2VEC: GRID SEARCH TO FIND HYPER PARAMETERS
In [128]:
from sklearn.svm import NuSVC
In [175]:
parameters = { 'nu':[0.1,0.2,0.15,0.05]}
In [146]:
model = GridSearchCV(classifier,param grid=parameters,scoring='accuracy',cv
=5)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print(model)
print(score)
GridSearchCV(cv=5, error score='raise',
       estimator=NuSVC(cache size=200, class weight=None, coef0=0.0,
   decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max iter=-1, nu=0.5, probability=False, random state=None,
   shrinking=True, tol=0.001, verbose=False),
       fit params=None, iid=True, n jobs=1,
       param grid={'nu': [0.1, 0.2, 0.15, 0.05]}, pre dispatch='2*n jobs',
       refit=True, return train score='warn', scoring='accuracy',
       verbose=0)
0.916
In [149]:
model = NuSVC (nu=0.1)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print(model)
print(score)
NuSVC (cache size=200, class weight=None, coef0=0.0,
   decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max_iter=-1, nu=0.1, probability=False, random state=None,
   shrinking=True, tol=0.001, verbose=False)
0.9156666666666666
In [150]:
model = NuSVC (nu=0.2)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print (model)
```

```
print(score)
NuSVC(cache size=200, class weight=None, coef0=0.0,
   decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max iter=-1, nu=0.2, probability=False, random state=None,
   shrinking=True, tol=0.001, verbose=False)
0.916
In [151]:
model = NuSVC (nu=0.01)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print(model)
print(score)
NuSVC(cache size=200, class weight=None, coef0=0.0,
   decision function shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max_iter=-1, nu=0.01, probability=False, random_state=None,
   shrinking=True, tol=0.001, verbose=False)
0.85333333333333334
AVERAGE WORD2VEC: RANDOM SEARCH TO FIND HYPER PARAMETERS
In [180]:
parameters = {'nu': np.linspace(0.0,1,5)}
parameters
Out[180]:
{'nu': array([0. , 0.25, 0.5 , 0.75, 1. ])}
In [173]:
classifier = NuSVC()
In [178]:
model = RandomizedSearchCV(classifier,param distributions=parameters,cv=5,n
iter=4)
model.fit(xtrain,ytrain)
Out[178]:
RandomizedSearchCV(cv=5, error score='raise',
          estimator=NuSVC(cache size=200, class weight=None, coef0=0.0,
   decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max iter=-1, nu=0.5, probability=False, random state=None,
   shrinking=True, tol=0.001, verbose=False),
          fit params=None, iid=True, n iter=4, n jobs=1,
          param distributions={'nu': [0.1, 0.2, 0.15, 0.05]},
          pre_dispatch='2*n_jobs', random_state=None, refit=True,
          return train score='warn', scoring=None, verbose=0)
In [181]:
pred = model.predict(xtest)
score = accuracy score(ytest,pred)
print(score)
```

# In [184]:

```
model = NuSVC(nu=0.01)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy_score(ytest,pred)
print(model)
print(score)
```

NusVC(cache\_size=200, class\_weight=None, coef0=0.0,
 decision\_function\_shape='ovr', degree=3, gamma='auto', kernel='rbf',
 max\_iter=-1, nu=0.01, probability=False, random\_state=None,
 shrinking=True, tol=0.001, verbose=False)
0.85333333333333334

# In [185]:

```
model = NuSVC(nu=0.1)
model.fit(xtrain,ytrain)
pred = model.predict(xtest)
score = accuracy_score(ytest,pred)
print(model)
print(score)
```

# Observation:

AVERAGE word2VEC:

- $\rightarrow$  By using Grid Search with NuSVC the best values of hyper parameters were nu = 0.5
- -> The accurancy by grid search is 91.6
- -> The accurancy of random search is 91.6

# **CONCLUSION:**

RBF SVM:

BAG OF WORDS:

- -> Implemented RBF SVM with Grid Search and random Search
- -> By using Grid Search with "RBF" as kernel the best values of hyper parameters were C = 1 and Gamma = 3

- $\rightarrow$  By using Random search with "RBF" as kernel the best values of hyper parameters were C = 3 and gamma= 2
  - -> The accurancy by grid search is 84.19
  - -> The accurancy of random search is 84.24
- $\rightarrow$  Measured accuracy with differenct combinations of hyper p arameters

# LINEAR SVM:

#### TFIDF:

- -> Implemented Linear SVM with Grid Search and random Search
- $\rightarrow$  By using Grid Search with Linear SVM the best values of h yper parameters were C = 0.001
- $\,$  -> By using Random search with Linear SVM the best values of hyper parameters were C = 8
  - -> The accurancy by grid search is 89.36
  - -> The accurancy of random search is 87.14

#### NuSVM:

#### AVERAGE WORD2VEC:

- -> Implemented Nu SVM with Grid Search and random Search
- $\ \ ->$  By using Grid Search with NuSVC the best values of hyper parameters were nu = 0.5
- $\,$  -> By using Random search with NuSVC the best values of hyper parameters were nu = 0.5
  - -> The accurancy by grid search is 91.6
  - -> The accurancy of random search is 91.6