```
In [ ]:
         The following code uses movie reviews data from nltk.corpus. The da
         ta contains files with movie reviews and
         their corresponding labels(positive or negative)
         It does text preprocessing such as removing stop words and tokeniza
         tion.
         After preprocessing different classifiers like Naive Bayesian class
         ifier, SVC classifier, BernoulliNB classifier,
         SGDClassifier classifier, LogisticRegression classifier are implemen
         ted.
         Finally a aggregator classifer which outputs the most voted label a
         nd its confidence is implemented.
         __author__='Soniya Rode'
          _citation__="pythonprogramming"
In [78]: import nltk
         import random
         from nltk.corpus import movie reviews
         from nltk.tokenize import RegexpTokenizer
         from nltk.corpus import stopwords
         import pickle
         from nltk.classify.scikitlearn import SklearnClassifier
         from sklearn.naive bayes import MultinomialNB, BernoulliNB
         from sklearn.linear model import LogisticRegression,SGDClassifier
         from sklearn.svm import SVC
         from nltk.classify import ClassifierI
         from statistics import mode
In [31]: #Initalise the stop words set and tokenizer to remove punctuations.
         stopwords = set(stopwords.words('english'))
         tokenizer = RegexpTokenizer(r'\w+')
In [32]: | #Get all words in the movie reviews
         words = []
         for w in movie reviews.words():
             #If the word is not a punctuation mark and not in stopwords add
         it to the list
             if tokenizer.tokenize(w) and w not in stopwords:
                 words.append(w.lower())
                 #print(words)
         #Get frequencies of all words
         words = nltk.FreqDist(words)
```

## Get top most 10 common words

```
In [33]: words.most common(10)
Out[33]: [('film', 9517),
          ('one', 5852),
          ('movie', 5771),
          ('like', 3690),
          ('even', 2565),
          ('good', 2411),
          ('time', 2411),
          ('story', 2169),
          ('would', 2109),
          ('much', 2049)]
In [34]: | #Get the top 4000 common words
         commonWords=list(words.keys())[:4000]
In [35]: | #There are two categories pos,neg
         #for every category, get the files associated with it.
         # Add the review words, the category as a tuple to the reviews
         reviews = [(list(movie reviews.words(fileid)), category)
                       for category in movie reviews.categories()
                       for fileid in movie reviews.fileids(category)]
         #Shuffle the data since arranged categorywise
         #reviews variable contains all reviews with their labels
         random.shuffle(reviews)
In [36]: #Function to return boolean value for words from the review which a
         re amongst the most common words.
         #Return True if the word in the review is among the top 4000 words.
         def check commonWords(words):
             words = set(words)
             listOfCommonWords = {}
             for w in commonWords:
                 listOfCommonWords[w] = (w in words)
             return listOfCommonWords
In [28]: #Now
         featuresets = [(check commonWords(rev), category) for (rev, categor
         y) in reviews]
In [37]: print(len(featuresets))
```

2000

```
In [38]: #split the data into training and testing
         training set = featuresets[:1800]
         # set that we'll test against.
         testing set = featuresets[1800:]
In [42]: | #Use NLTK's Naive Bayes classifier
         classifier = nltk.NaiveBayesClassifier.train(training set)
         print("accuracy:",(nltk.classify.accuracy(classifier, testing set))
In [43]:
         *100)
         accuracy: 80.5
In [44]: classifier.show_most_informative_features(15)
         Most Informative Features
                         bothered = True
                                                       neg: pos
                                                                           8
         .9:1.0
                          miscast = True
                                                       neg : pos
                                                                           8
         .1:1.0
                          frances = True
                                                       pos : neg
                                                                           7
         .7:1.0
                         sundance = True
                                                       pos : neg
         .7 : 1.0
                       schumacher = True
                                                       neg: pos
                                                                           7
         .4:1.0
                    unimaginative = True
                                                       neg : pos
                                                                           7
         .0:1.0
                           shoddy = True
                                                       neg: pos
                                                                           7
         .0 : 1.0
                        atrocious = True
                                                                           7
                                                       neg : pos
         .0 : 1.0
                         jeopardy = True
                                                       neg : pos
                                                                           6
         .3:1.0
                            suvari = True
                                                       neg : pos
         .3:1.0
                             mena = True
                                                       neg: pos
         .3 : 1.0
                           wasted = True
                                                       neg : pos
                                                                           5
         .7 : 1.0
                          singers = True
                                                                           5
                                                       pos : neg
         .7:1.0
                          bronson = True
                                                       neg: pos
                                                                           5
         .6:1.0
                        underwood = True
                                                       neg: pos
                                                                           5
         .6:1.0
```

```
In [48]: #Store the classifier
Pickle_classifier = open("naivebayes.pickle","wb")
pickle.dump(classifier, Pickle_classifier)
Pickle_classifier.close()
```

## Using sklearn classifiers

```
In [73]: #Logistic Regression Classifier
          LogisticRegression classifier = SklearnClassifier(LogisticRegressio
          n(solver='lbfgs'))
          LogisticRegression classifier.train(training set)
 Out[73]: <SklearnClassifier(LogisticRegression(C=1.0, class weight=None, du
          al=False, fit_intercept=True,
                    intercept_scaling=1, max_iter=100, multi_class='warn',
                    n jobs=None, penalty='12', random state=None, solver='lb
          fgs',
                    tol=0.0001, verbose=0, warm_start=False))>
In [108]: #stochastic gradient descent (SGD) classifier
          SGDClassifier classifier = SklearnClassifier(SGDClassifier( max ite
          r=1000)
          SGDClassifier classifier.train(training set)
Out[108]: <SklearnClassifier(SGDClassifier(alpha=0.0001, average=False, clas
          s weight=None,
                 early stopping=False, epsilon=0.1, eta0=0.0, fit intercept=
          True,
                 11 ratio=0.15, learning rate='optimal', loss='hinge', max i
          ter=1000,
                 n_iter=None, n_iter_no_change=5, n_jobs=None, penalty='12',
                 power t=0.5, random state=None, shuffle=True, tol=None,
                 validation fraction=0.1, verbose=0, warm start=False))>
 In [83]: #Support Vector machine Classifier
          SVC classifier = SklearnClassifier(SVC(gamma='auto'))
          SVC classifier.train(training set)
 Out[83]: <SklearnClassifier(SVC(C=1.0, cache size=200, class weight=None, c</pre>
          oef0=0.0,
            decision function shape='ovr', degree=3, gamma='auto', kernel='r
            max iter=-1, probability=False, random state=None, shrinking=Tru
            tol=0.001, verbose=False))>
```

BernoulliNB classifier accuracy percent: 81.0

```
In [84]: #Classifier Accuracies:
    print("Classifier accuracies :")
    print("Naive Bayes Classifier accuracy:",(nltk.classify.accuracy(classifier, testing_set))*100)
    print("LogisticRegression_classifier accuracy:", (nltk.classify.accuracy(LogisticRegression_classifier, testing_set))*100)

    print("SGDClassifier_classifier accuracy:", (nltk.classify.accuracy(SGDClassifier_classifier, testing_set))*100)
    print("SVC_classifier accuracy:", (nltk.classify.accuracy(SVC_classifier, testing_set))*100)
```

Classifier accuracies:
Naive Bayes Classifier accuracy: 80.5
LogisticRegression\_classifier accuracy: 78.5
SGDClassifier\_classifier accuracy: 80.0
SVC\_classifier accuracy: 80.0

```
In [99]:
         Aggregated Classifier takes different classifiers as input. To clas
         sify it takes vote from each classifier, and
         returns the class label with most number of votes (mode)
         The get confidence method returns the confidence on the vote(class
         label)
         The get confidence uses mode method from statistics which raisies s
         tatistics.StatisticsError
         statistics. Statistics Error: no unique mode; found 2 equally common
         Since number of classifiers used in uneven, this error is avoided.
         class AggregatedClassifier(ClassifierI):
             def init (self, *classifiers):
                 self._classifiers = classifiers
             def classify(self, features):
                     votes = []
                     for classifier in self. classifiers:
                         votes.append(classifier.classify(features))
                     return mode(votes)
             def get confidence(self, features):
                 votes = []
                 for classifier in self. classifiers:
                     votes.append(classifier.classify(features))
                 confidence = votes.count(mode(votes)) / len(votes)
                 return confidence
```

## In [104]:

agg\_classifier = AggregatedClassifier(classifier,SVC\_classifier,Ber noulliNB\_classifier,SGDClassifier\_classifier,LogisticRegression\_cla ssifier)

#print("voted\_classifier accuracy percent:", (nltk.classify.accurac
y(voted\_classifier, testing\_set))\*100)
for i in range(20):
 print("Predicted Class label:", agg\_classifier.classify(testing\_set[i][0]), "Confidence %:",agg\_classifier.get\_confidence(testing\_set[i][0])\*100)

```
Predicted Class label: pos Confidence %: 100.0
Predicted Class label: neg Confidence %: 80.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: pos Confidence %: 80.0
Predicted Class label: neg Confidence %: 80.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: neg Confidence %: 60.0
Predicted Class label: neg Confidence %: 80.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: pos Confidence %: 100.0
Predicted Class label: pos Confidence %: 100.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: pos Confidence %: 100.0
Predicted Class label: neg Confidence %: 60.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: neg Confidence %: 100.0
Predicted Class label: neg Confidence %: 60.0
Predicted Class label: neg Confidence %: 80.0
```