# Importing libraries

```
In [1]: import nltk
         import glob
         import os
         import numpy as np
         import string
         import pickle
In [2]: from gensim.models import Doc2Vec
         from gensim.models.doc2vec import LabeledSentence
         from tqdm import tqdm
         from sklearn import utils
         from sklearn.svm import LinearSVC
         from sklearn.neural network import MLPClassifier
         from sklearn.metrics import roc curve
         from sklearn.metrics import roc auc score
         from matplotlib import pyplot
         C:\ProgramData\Anaconda3\lib\site-packages\gensim\utils.py:1197: UserWarning: de
         tected Windows; aliasing chunkize to chunkize serial
           warnings.warn("detected Windows; aliasing chunkize to chunkize serial")
In [3]: from nltk import sent tokenize
         from nltk.tokenize import word tokenize
         from nltk.corpus import stopwords
         from nltk.stem.porter import PorterStemmer
         from sklearn.feature_extraction.text import CountVectorizer
         from sklearn.feature_extraction.text import TfidfVectorizer
         from sklearn.naive bayes import MultinomialNB
         from sklearn.naive_bayes import BernoulliNB
         from sklearn.naive_bayes import GaussianNB
         from sklearn.linear model import LogisticRegression
         from sklearn.ensemble import RandomForestClassifier
         from sklearn.metrics import accuracy score
         from collections import Counter
         from collections import defaultdict
In [64]: | X_train_text = []
         Y train = []
         X_test_text =[]
         Y_test = []
         Vocab = {} {}
         VocabFile = "aclImdb/imdb.vocab"
```

# **Create Vocabulary Function**

### **Cleaning Data**

```
In [93]: def clean_review(text):
    tokens = word_tokenize(text)
    tokens = [w.lower() for w in tokens]
    table = str.maketrans('', '', string.punctuation)
    stripped = [w.translate(table) for w in tokens]
    words = [word for word in stripped if word.isalpha()]
    stop_words = set(stopwords.words('english'))
    words = [w for w in words if not w in stop_words]
    return words
```

# **Generating Word Matrices**

```
In [67]: def BoWMatrix(docs):
    vectorizer = CountVectorizer(binary=True, vocabulary = Vocab)
    Doc_Term_matrix = vectorizer.fit_transform(docs)
    return Doc_Term_matrix

def TfidfMatrix(docs):
    vectorizer = TfidfVectorizer(vocabulary = Vocab, norm = '11')
    Doc_Term_matrix = vectorizer.fit_transform(docs)
    return Doc_Term_matrix
```

### **ROC Curve Function**

```
In [69]: def ROC2(X, pred, pred1, pred2):
    fpr, tpr, thresholds = roc_curve(X, pred)
    pyplot.plot([0, 1], [0, 1], linestyle='--')
    pyplot.plot(fpr, tpr, marker='.')
    fpr1, tpr1, thresholds1 = roc_curve(X, pred1)
    pyplot.plot([0, 1], [0, 1], linestyle='--')
    pyplot.plot(fpr1, tpr1, marker='.')
    fpr2, tpr2, thresholds2 = roc_curve(X, pred2)
    pyplot.plot([0, 1], [0, 1], linestyle='--')
    pyplot.plot(fpr2, tpr2, marker='.')
    pyplot.show()
```

### **Naive Bayes Function**

```
In [70]: | def NB(X,Y train, Xtest,Y test, mtype):
             if mtype == "Bow":
                 model = BernoulliNB()
             elif mtype == "Tfidf":
                 model = MultinomialNB()
             else:
                 model = GaussianNB()
             model.fit(X,Y train)
             pred1 = model.predict(X)
             pred2 = model.predict(Xtest)
             acc1 = accuracy score(Y train, pred1)
             acc2 = accuracy_score(Y_test,pred2)
             print("NaiveBayes + " + mtype + " Train Accuracy: " + str(acc1*100) + "%")
             print("NaiveBayes + " + mtype + " Test Accuracy: " + str(acc2*100) + "%")
             prob1 = model.predict proba(X)
             prob1 = prob1[:, 1]
             prob2 = model.predict_proba(Xtest)
             prob2 = prob2[:, 1]
             #ROC(Y train, pred1, Y test, pred2)
             ROC(Y train, prob1, Y test, prob2)
```

### **Logistic Regression Function**

```
In [71]: def LR(X,Y_train,Xtest,Y_test,mtype):
    model = LogisticRegression()
    model.fit(X,Y_train)
    pred1 = model.predict(X)
    pred2 = model.predict(Xtest)
    acc1 = accuracy_score(Y_train,pred1)
    acc2 = accuracy_score(Y_test,pred2)
    print("LogisticRegression + " + mtype + " Train Accuracy: " + str(acc1*100) +
    "%")
    print("LogisticRegression + " + mtype + " Test Accuracy: " + str(acc2*100) +
    "%")
    prob1 = model.predict_proba(X)
    prob1 = prob1[:, 1]
    prob2 = model.predict_proba(Xtest)
    prob2 = prob2[:, 1]
    #ROC(Y_train, pred1, Y_test, pred2)
    ROC(Y_train, prob1, Y_test, prob2)
```

#### **Random Forest Function**

```
In [72]: def RF(X,Y train, Xtest, Y test, mtype):
             if mtype == "Bow":
                 n = 400
                 md = 100
             elif mtype == "Tfidf":
                 n = 400
                 md = 100
             else:
                 n = 100
                 md = 10
             model = RandomForestClassifier(n_estimators=n, bootstrap=True,
                                          max depth=md, max features='auto',
                                          min samples leaf=4, min samples split=10)
             model.fit(X,Y train)
             pred1 = model.predict(X)
             pred2 = model.predict(Xtest)
             acc1 = accuracy score(Y train, pred1)
             acc2 = accuracy_score(Y_test,pred2)
             print("RandomForest + " + mtype + " Train Accuracy: " + str(acc1*100) + "%")
             print("RandomForest + " + mtype + " Test Accuracy: " + str(acc2*100) + "%")
             prob1 = model.predict_proba(X)
             prob1 = prob1[:, 1]
             prob2 = model.predict_proba(Xtest)
             prob2 = prob2[:, 1]
             #ROC(Y_train, pred1, Y_test, pred2)
             ROC(Y_train, prob1, Y_test, prob2)
```

# **Support Vector Machine Function**

```
In [73]: def SVM(X,Y_train,Xtest,Y_test,mtype):
    model = LinearSVC()
    model.fit(X,Y_train)
    pred1 = model.predict(X)
    pred2 = model.predict(Xtest)
    acc1 = accuracy_score(Y_train,pred1)
    acc2 = accuracy_score(Y_test,pred2)
    print("SVM + " + mtype + " Train Accuracy: " + str(acc1*100) + "%")
    print("SVM + " + mtype + " Test Accuracy: " + str(acc2*100) + "%")
    ROC(Y_train, pred1, Y_test, pred2)
```

### **Forward Feed Neural Network Function**

```
In [74]: def NN(X,Y_train,Xtest,Y_test,mtype):
             model = MLPClassifier(hidden_layer_sizes=(10,10),activation='relu',max_iter=20
         0)
             model.fit(X,Y_train)
             pred1 = model.predict(X)
             pred2 = model.predict(Xtest)
             acc1 = accuracy_score(Y_train,pred1)
             acc2 = accuracy_score(Y_test,pred2)
             print("FFN + " + mtype + " Train Accuracy: " + str(acc1*100) + "%")
             print("FFN + " + mtype + " Test Accuracy: " + str(acc2*100) + "%")
             prob1 = model.predict proba(X)
             prob1 = prob1[:, 1]
             prob2 = model.predict proba(Xtest)
             prob2 = prob2[:, 1]
             #ROC(Y train, pred1, Y test, pred2)
             ROC(Y_train, prob1, Y_test, prob2)
```

# **Loading Data**

```
In [4]: path1 = 'aclImdb/train/pos/*.txt'
    path2 = 'aclImdb/train/neg/*.txt'
    path3 = 'aclImdb/test/pos/*.txt'
    path4 = 'aclImdb/test/neg/*.txt'

files1 = glob.glob(path1)
    files2 = glob.glob(path2)
    files3 = glob.glob(path3)
    files4 = glob.glob(path4)
```

```
In [77]: #Positive labels
         for i, filename in enumerate(files1):
             f = open(filename, "r+", encoding='latin-1')
             text = f.read()
             f.close()
             X train text.append(text)
             Y train.append(1)
          #Neg labels
         for j, filename in enumerate(files2):
             f = open(filename, "r+", encoding='latin-1')
             text = f.read()
             f.close()
             X train text.append(text)
             Y_train.append(0)
          #Test labels +
         for k, filename in enumerate(files3):
             f = open(filename, "r+", encoding='latin-1')
             text = f.read()
             f.close()
             X test text.append(text)
             Y_test.append(1)
          #Test labels +
         for 1, filename in enumerate(files4):
             f = open(filename, "r+", encoding='latin-1')
             text = f.read()
             f.close()
             X_test_text.append(text)
             Y test.append(0)
         CreateVocab();
```

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# **Generating Word Matrix for Test & Train Data**

```
In [75]: def Getbowvec(X_train_text,Y_train,X_test_text,Y_test):
    X = BoWMatrix(X_train_text)
    Xtest = BoWMatrix(X_test_text)
    return X,Xtest

def Gettfidfvec(X_train_text,Y_train,X_test_text,Y_test):
    X = TfidfMatrix(X_train_text)
    Xtest = TfidfMatrix(X_test_text)
    return X,Xtest
```

# **Doc2Vec Representation**

```
In [79]: | '''
        def LabelRev(reviews, label string):
           result = []
           prefix = label_string
            for i, t in enumerate (reviews):
               # print(t)
               result.append(LabeledSentence(t, [prefix + ' %s' % i]))
            return result
        LabelledXtrain = LabelRev(X train text, "review")
        LabelledXtest = LabelRev(X test text, "test")
        LabelledData = LabelledXtrain + LabelledXtest
        modeld2v = Doc2Vec(dm=1, min count=2, alpha=0.065, min alpha=0.065)
        modeld2v.build vocab([x for x in tqdm(LabelledData)])
        print("Training the Doc2Vec Model....")
        for epoch in range (50):
            print("epoch : ",epoch)
            modeld2v.train(utils.shuffle([x for x in tqdm(LabelledData)]), total examples=1
        en(LabelledData), epochs=1)
            modeld2v.alpha -= 0.002
            modeld2v.min_alpha = modeld2v.alpha
        print("Saving Doc2Vec1 Model....")
        modeld2v.save('doc2vec1.model')
        #print("Saving Doc2Vec Model....")
        #modeld2v.save('doc2vec.model')
Out[79]: '\ndef LabelRev(reviews,label string):\n result = []\n prefix = label stri
              for i, t in enumerate(reviews):\n
                                                     # print(t)\n result.appe
        t,"test")\n\nLabelledData = LabelledXtrain + LabelledXtest\n\nmodeld2v = Doc2Vec
        in tqdm(LabelledData)]) \n\nprint("Training the Doc2Vec Model....") \nfor epoch i
        n range(50):\n print("epoch : ",epoch)\n modeld2v.train(utils.shuffle([x f
        or x in tqdm(LabelledData)]), total examples=len(LabelledData), epochs=1)\n
        odeld2v.alpha -= 0.002\n modeld2v.min alpha = modeld2v.alpha\n \nprint("Sa
        ving Doc2Vec1 Model....")\nmodeld2v.save(\'doc2vec1.model\')\n#print("Saving Doc
        2Vec Model....") \n#modeld2v.save(\'doc2vec.model\') \n'
In [85]: def Doc2vec(X_train_text,Y_train,X_test_text,Y_test):
            model = Doc2Vec.load('doc2vec.model')
            #model = Doc2Vec.load('doc2vec1.model')
            X = []
            Xtest = []
            for i,l in enumerate(X train text):
               temp = "review" + " " + str(i)
               X.append(model.docvecs[temp])
            for i,l in enumerate(X test text):
               temp = "test" + "_" + str(i)
```

Xtest.append(model.docvecs[temp])

return X,Xtest

```
In [86]: print("Bag of Words is being built...")
         X,Xtest = Getbowvec(X_train_text,Y_train,X_test_text,Y_test)
         print("Tf-idf is being built...")
         X1,Xtest1 = Gettfidfvec(X_train_text,Y_train,X_test_text,Y_test)
         print("Doc2Vec is being built...")
         X2,Xtest2 = Doc2vec(X train text,Y train,X test text,Y test)
         Bag of Words is being built...
         Tf-idf is being built...
         Doc2Vec is being built...
In [112]: len(X[0])
          [ 3.8730502e-01 -2.4226126e-01 1.8447070e-01 2.7821556e-01
           5.6896007e-01 2.4535358e-01 -2.5278392e-01 1.5472387e-01
          -1.1508914e-01 4.5354521e-01 3.8365748e-02 -2.5258625e-01
          -5.8337015e-01 1.0790733e+00 -1.6636238e-01 4.3921784e-02
           6.0108167e-01 -7.4332672e-01 -3.5194703e-03 5.2995592e-01
          -4.6524012e-01 -2.2355888e-02 8.3163187e-02 -3.0640143e-01
          -7.0134312e-01 6.8939185e-01 2.0314547e-01 -4.4697830e-01
          -1.9118084e-02 -1.1067226e-01 -3.0236110e-01 3.7518761e-01
          -5.2174801e-01 -6.5070170e-01 4.4410905e-01 -3.6672121e-01
          -6.6573453e-01 -7.9081794e-03 -9.3441790e-01 1.5675272e-01
           2.4286245e-01 -5.1078118e-02 2.6444617e-01 1.8430072e-01
          -3.6135960e-01 1.0735059e+00 -1.7230248e-01 1.9120649e-01
           2.9940462e-01 1.2391033e+00 -3.0393973e-01 2.6779616e-01
           4.8723632e-01 2.8240904e-03 -2.5652254e-01 1.0942936e+00
           6.4555228e-01 -6.1286968e-01 8.6758316e-01 -1.3204034e-01
           3.7257919e-01 -2.8581244e-01 7.1880788e-01 2.4354255e-01
           8.9242440e-01 -6.6156977e-01 2.5719035e-01 2.2389887e-01
           1.4597312e-01 2.1936417e-01 2.6718158e-02 -1.9517705e-01
           3.0567044e-01 4.3458107e-01 3.6869454e-01 1.0570779e+00
          -4.2633000e-01 -1.2129361e-01 -3.2494134e-01 9.7133152e-02
          -1.2926124e-01 3.7891394e-01 -8.6885536e-01 1.1091402e-02
          -8.0138183e-01 5.7643318e-01 -3.5828719e-01 6.8130858e-02
          -6.5640682e-01 1.1339425e+00 7.7646202e-04 1.5335666e-01
          -3.6238903e-01 -1.3119204e-01 -2.6151622e-02 9.3961424e-01
          -8.3501214e-01 3.9072767e-01 2.4230620e-01 6.3034648e-01]
```

# **Applying Classification Algorithms**

```
In [87]: print("Naive Bayes:")
          NB(X,Y_train,Xtest,Y_test,"Bow")
          NB(X1,Y_train,Xtest1,Y_test,"Tfidf")
          NB(X2,Y_train,Xtest2,Y_test,"Doc2Vec")
          Naive Bayes:
          NaiveBayes + Bow Train Accuracy: 90.3800000000001%
          NaiveBayes + Bow Test Accuracy: 81.872%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                      Train
                                                    Test
           0.0
               0.0
                       0.2
                               0.4
                                        0.6
                                                        1.0
          NaiveBayes + Tfidf Train Accuracy: 89.6000000000001%
          NaiveBayes + Tfidf Test Accuracy: 85.048%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                     - Train
                                                      Test
           0.0
                               0.4
                       0.2
                                        0.6
                                                0.8
                                                        1.0
          NaiveBayes + Doc2Vec Train Accuracy: 86.168%
          NaiveBayes + Doc2Vec Test Accuracy: 86.444%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                      Train
                                                      Test
           0.0
```

0.2

0.4

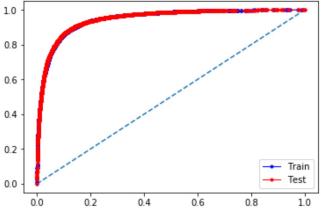
0.6

9 of 13 26-10-2019, 18:44

0.8

1.0

```
In [88]: print("Logistic Regression:")
          LR(X,Y_train,Xtest,Y_test,"Bow")
          LR(X1,Y_train,Xtest1,Y_test,"Tfidf")
          LR(X2,Y_train,Xtest2,Y_test,"Doc2Vec")
          Logistic Regression:
          LogisticRegression + Bow Train Accuracy: 99.752%
          LogisticRegression + Bow Test Accuracy: 86.8639999999999998
          1.0
           0.8
           0.6
           0.4
           0.2
                                                    Train
                                                   Test
           0.0
                      0.2
                              0.4
                                      0.6
                                                      1.0
          LogisticRegression + Tfidf Train Accuracy: 83.82%
          LogisticRegression + Tfidf Test Accuracy: 82.84%
          1.0
           0.8
           0.6
           0.4
           0.2
                                                   Train
                                                    Test
           0.0
                      0.2
                              0.4
                                      0.6
                                                      1.0
                                              0.8
          LogisticRegression + Doc2Vec Train Accuracy: 87.968%
          LogisticRegression + Doc2Vec Test Accuracy: 88.1480000000001%
```



```
In [89]: print("Random Forest:")
          RF(X,Y_train,Xtest,Y_test,"Bow")
          RF(X1,Y_train,Xtest1,Y_test,"Tfidf")
          RF(X2,Y_train,Xtest2,Y_test,"Doc2Vec")
          Random Forest:
          RandomForest + Bow Train Accuracy: 93.5880000000001%
          RandomForest + Bow Test Accuracy: 86.572%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                       Train
                                                     Test
           0.0
               0.0
                       0.2
                                0.4
                                        0.6
                                                        1.0
          RandomForest + Tfidf Train Accuracy: 95.916%
          RandomForest + Tfidf Test Accuracy: 86.051999999999998
           1.0
           0.8
           0.6
           0.4
           0.2
                                                     - Train
                                                       Test
           0.0
                       0.2
                                0.4
                                        0.6
                                                        1.0
                                                0.8
          RandomForest + Doc2Vec Train Accuracy: 95.136%
          RandomForest + Doc2Vec Test Accuracy: 81.632%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                       Train
                                                       Test
           0.0
                       0.2
                                0.4
                                        0.6
                                                0.8
                                                        1.0
```

```
In [90]: print("SVM:")
          SVM(X,Y_train,Xtest,Y_test,"Bow")
          SVM(X1,Y_train,Xtest1,Y_test,"Tfidf")
          SVM(X2,Y_train,Xtest2,Y_test,"Doc2Vec")
          SVM:
          SVM + Bow Train Accuracy: 100.0%
          SVM + Bow Test Accuracy: 84.892%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                        Train
                                                      Test
           0.0
                0.0
                        0.2
                                0.4
                                         0.6
                                                         1.0
          SVM + Tfidf Train Accuracy: 89.27199999999998
          SVM + Tfidf Test Accuracy: 86.848%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                       - Train
                                                        Test
           0.0
                                0.4
                        0.2
                                         0.6
                                                         1.0
          SVM + Doc2Vec Train Accuracy: 87.952%
          SVM + Doc2Vec Test Accuracy: 88.152%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                        Train
                                                        Test
           0.0
                        0.2
                                0.4
                                         0.6
                0.0
                                                 0.8
                                                          1.0
```

```
In [91]: print("Neural Networks:")
          NN(X,Y_train,Xtest,Y_test,"Bow")
          NN(X1,Y_train,Xtest1,Y_test,"Tfidf")
          NN(X2,Y_train,Xtest2,Y_test,"Doc2Vec")
          Neural Networks:
          FFN + Bow Train Accuracy: 100.0%
          FFN + Bow Test Accuracy: 85.6%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                       Train
                                                      Test
           0.0
               0.0
                        0.2
                                0.4
                                        0.6
                                                         1.0
          FFN + Tfidf Train Accuracy: 100.0%
          FFN + Tfidf Test Accuracy: 85.84%
           1.0
           0.8
           0.6
           0.4
           0.2
                                                      - Train
                                                       Test
           0.0
                                0.4
                        0.2
                                        0.6
                                                         1.0
                                                 0.8
          FFN + Doc2Vec Train Accuracy: 89.776%
          FFN + Doc2Vec Test Accuracy: 87.30799999999998
           1.0
           0.8
           0.6
           0.4
           0.2
                                                       Train
                                                        Test
           0.0
                        0.2
                                0.4
                                         0.6
                                                 0.8
                                                         1.0
 In [ ]:
```