

In [9]:

```
import os
import numpy as np
import random
import math
from collections import Counter
```

In [10]:

```
def make_Dictionary(train_dir):
    emails = [os.path.join(train_dir,f) for f in os.listdir(train_dir)]
    all_words = []
    for mail in emails:
        with open(mail) as m:
            for i,line in enumerate(m):
                if i == 2: #Body of email is only 3rd line of text file
                    words = line.split()
                    all_words += words
    dictionary = Counter(all_words)
    dictionary_temp = Counter(all_words)

    # Paste code for non-word removal here (code snippet is given below)
    list_to_remove = dictionary_temp.keys()
    for item in list_to_remove:
        if item.isalpha() == False: #Determine whether it is punctuation
            del dictionary[item]
        elif len(item) == 1: #
            del dictionary[item]
    dictionary = dictionary.most_common(3000)
    return dictionary
```

In [11]:

```
def extract_features(root_dir,dictionary):
    emails = [os.path.join(root_dir,f) for f in os.listdir(root_dir)]
    all_words = []
    features_matrix = np.zeros((len(emails),3000))
    docID = 0
    for mail in emails:
        with open(mail) as m:
            for i,line in enumerate(m):
                if i == 2:
                    words = line.split()
                    for word in words:
                        wordID = 0
                        for i,d in enumerate(dictionary):
                            if d[0] == word:
                                wordID = i
                                features_matrix[docID,wordID] = words.count(word)

            docID = docID + 1
    return features_matrix
```

In [12]:

```
Path = "train-mails"
dir = make_Dictionary(Path)
# print('finish dir:\n', dir)
train_matrix = extract_features(Path, dir)
# print(train_matrix)

Path_test = "test-mails"
test_matrix = extract_features(Path_test, dir)
# print(test_matrix)
```

In [46]:

```

class NaiveBayes:
    train_data = []
    Mean = {}
    Std = {}
    Prob = {}
    def fit(self, x, y):
        #divide x by x
        DivideX = {}
        for i in range(len(y)):
            if (y[i] not in DivideX):
                DivideX[y[i]] = []
                DivideX[y[i]].append(x[i])
            else:
                DivideX[y[i]] = np.vstack((DivideX[y[i]], x[i]))
        self.train_data = DivideX
        for key in DivideX:
            self.Mean[key] = []
            self.Std[key] = []
            self.Prob[key] = []
            for i in range(DivideX[key].shape[1]):
                self.Mean[key].append(np.mean(DivideX[key][:, i]))
                #Divide
                c = 0
                for j in DivideX[key][:, i]:
                    if j >= self.Mean[key][i]:
                        c += 1
                num = len(DivideX[key][:, i])
                self.Prob[key].append(c + 1/num + 1)
                #
                if np.std(DivideX[key][:, i]) == 0:
                    self.Std[key].append(0.00001)
                else:
                    self.Std[key].append(np.std(DivideX[key][:, i]))
            # print(self.Mean)
        def probabiltiy(self, data, key):
            prob = 1
            # print(data)
            for i in range(len(data)):
                if self.Std[key][i] == 0:
                    if data[i] >= self.Mean[key][i]:
                        prob *= self.Prob[key][i]
                    else:
                        prob *= (1 - self.Prob[key][i])
                else:
                    exp = math.exp(-(math.pow(data[i] - self.Mean[key][i], 2)/(2*math.pow(self.Std[key][i], 2))))
                    prob *= (1/(math.sqrt(2*math.pi)*self.Std[key][i])) * exp
            return prob
        def predict(self, data):
            result = []
            for i in range(data.shape[0]):
                MaxProb = 0
                MaxProb_key = 0
                for key in self.Mean:
                    Prob_temp = self.probabiltiy(data[i], key)
                    if (Prob_temp > MaxProb):
                        MaxProb = Prob_temp
                        MaxProb_key = key
                result.append(MaxProb_key)

```

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    return result
def accuracy(self, data, prediction):
    correct = 0
    for x in range(len(data)):
        if data[x] == prediction[x]:
            correct += 1
    return (correct/float(len(data)))
def ConfusionMatrix(self, data, prediction):
    TP = 0
    TN = 0
    FP = 0
    FN = 0
    for i in range(len(data)):
        if data[i] == 1 and prediction[i] == 1:
            TP += 1
        if data[i] == 1 and prediction[i] == 0:
            FP += 1
        if data[i] == 0 and prediction[i] == 1:
            FN += 1
        if data[i] == 0 and prediction[i] == 0:
            TN += 1
    return TP, TN, FP, FN
def Precision(self, data, prediction):
    TP, TN, FP, FN = self.ConfusionMatrix(data, prediction)
    return TP/(TP + FP)
def Recall(self, data, prediction):
    TP, TN, FP, FN = self.ConfusionMatrix(data, prediction)
    return TP/(TP + FN)
def F1_score(self, data, prediction, Beta = 1):
    TP, TN, FP, FN = self.ConfusionMatrix(data, prediction)
    return (1 + Beta ** 2)*self.Precision(data, prediction) * self.Recall(data, prediction) /
    (Beta ** 2 * self.Precision(data, prediction) + self.Recall(data, prediction))

```

In [47]:

```

NB = NaiveBayes()
train_labels = np.zeros(train_matrix.shape[0])
train_labels[351:701] = 1
NB.fit(train_matrix, train_labels)
pred = NB.predict(train_matrix)
# print(NB.accuracy(train_labels, pred))
print('train data:')
train_accuracy = NB.accuracy(train_labels, pred)
train_TP, train_TN, train_FP, train_FN = NB.ConfusionMatrix(train_labels, pred)
train_Precision = NB.Precision(train_labels, pred)
train_Recall = NB.Recall(train_labels, pred)
train_F1score = NB.F1_score(train_labels, pred)
print('train_accuracy: {}'.format(train_accuracy))
print('train_Precision: {}'.format(train_Precision))
print('train_Recall: {}'.format(train_Recall))
print('train_F1_score: {}'.format(train_F1score))

```

```

train data:
train_accuracy:0.8988603988603988
train_Precision:0.7971428571428572
train_Recall:1.0
train_F1_score:0.8871224165341812

```

In [48]:

```
pred_test = NB.predict(test_matrix)
test_labels = np.zeros(test_matrix.shape[0])
test_labels[130:] = 1
# print(pred_test)
# print(NB.accuracy(test_label, pred_test))

Test_TP, Test_TN, Test_FP, Test_FN = NB.ConfusionMatrix(test_labels, pred_test)
print('test data:')
test_accuracy = NB.accuracy(test_labels, pred_test)
test_TP, test_TN, test_FP, test_FN = NB.ConfusionMatrix(test_labels, pred)
test_Precision = NB.Precision(test_labels, pred_test)
test_Recall = NB.Recall(test_labels, pred_test)
test_F1score = NB.F1_score(test_labels, pred_test)
print('test_accuracy: {}'.format(test_accuracy))
print('test_Precision: {}'.format(test_Precision))
print('test_Recall: {}'.format(test_Recall))
print('test_F1_score: {}'.format(test_F1score))
```

```
test data:
test_accuracy:0.8038461538461539
test_Precision:0.6076923076923076
test_Recall:1.0
test_F1_score:0.7559808612440191
```