In []:

- 1 Program:6
- 2 Assuming a set of documents that need to be classified, use the naïve Bayesian Classifi
- 3 model to perform this task. Built-in Java classes/API can be used to write the program.
- 4 Calculate the accuracy, precision, and recall for your data set.

In [13]:

```
1 import pandas as pd
```

In [14]:

```
msg=pd.read_csv('document.csv',names=['message','label'])
print('Total instances in the dataset:',msg.shape[0])
msg['labelnum']=msg.label.map({'pos':1,'neg':0})

X=msg.message
Y=msg.labelnum
print('\nThe message and its label of first 5 instances are listed below')

X5, Y5 = X[0:5], msg.label[0:5]
for x, y in zip(X5,Y5):
    print(x,',',y)
```

Total instances in the dataset: 18

The message and its label of first 5 instances are listed below I love this sandwich , pos
This is an amazing place , pos
I feel very good about these beers , pos
This is my best work , pos
What an awesome view , pos

In [15]:

```
# Splitting the dataset into train and test data
from sklearn.model_selection import train_test_split
xtrain,xtest,ytrain,ytest=train_test_split(X,Y)
print('\nDataset is split into Training and Testing samples')
print('Total training instances :', xtrain.shape[0])
print('Total testing instances :', xtest.shape[0])
```

Dataset is split into Training and Testing samples Total training instances : 13 Total testing instances : 5

In [16]:

```
# Output of count vectoriser is a sparse matrix
# CountVectorizer - stands for 'feature extraction'
from sklearn.feature_extraction.text import CountVectorizer
count_vect = CountVectorizer()

xtrain_dtm = count_vect.fit_transform(xtrain) #Sparse matrix

xtest_dtm = count_vect.transform(xtest)
print('\nTotal features extracted using CountVectorizer:',xtrain_dtm.shape[1])
print('\nFeatures for first 5 training instances are listed below')
df=pd.DataFrame(xtrain_dtm.toarray(),columns=count_vect.get_feature_names())
print(df[0:5])#tabular representation
```

Total features extracted using CountVectorizer: 42

```
Features for first 5 training instances are listed below
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```

[5 rows x 42 columns]

In [17]:

```
#print(xtrain_dtm) #Same as above but sparse matrix representation
# Training Naive Bayes (NB) classifier on training data.

from sklearn.naive_bayes import MultinomialNB

clf = MultinomialNB().fit(xtrain_dtm,ytrain)

predicted = clf.predict(xtest_dtm)

print('\nClassstification results of testing samples are given below')

for doc, p in zip(xtest, predicted):

pred = 'pos' if p==1 else 'neg'

print('%s -> %s ' % (doc, pred))
```

Classstification results of testing samples are given below I am tired of this stuff -> neg

```
In [18]:
```

```
#printing accuracy metrics
from sklearn import metrics
print('\nAccuracy metrics')
print('Accuracy of the classifer is',metrics.accuracy_score(ytest,predicted))
print('Recall :',metrics.recall_score(ytest,predicted),
    '\nPrecison :',metrics.precision_score(ytest,predicted))
print('Confusion matrix')
print(metrics.confusion_matrix(ytest,predicted))
```

In []:

1