6. Assuming a set of documents that need to classified use the naive Bayesian Classifier model to perform this task. Built-in Java Classes/API can be used to write the program. Calculate the accuracy, precision and recall for your dataset.

```
import pandas as pd
msg = pd.read_csv('C:/Users/User/Desktop/ML Lab/
                    lab6.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("In The 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)

from sklearn.model_selection import train_test_split:
  xtrain, xtest, ytrain, ytest = train_test_split(x,y)
  print("Dataset is split into Training and Testing
                  Samples")
  print(" Total training instances:", xtrain.shape[0])
  print("Total testing instances:", xtest.shape[0])
```

Teacher's Signature :

```
from sklearn.feature_extraction.text import
                                    CountVectorizer
count_vect = CountVectorizer()
xtrain_dtm = count_vect.fit_transform(xtrain)

xtest_dtm = count_vect.fit transform(xtest)
print('In Total features extracted using CountVectorizer:'
                                    xtrain_dtm.shape[1])
print("In features for first 5 training instances are
                                    listed below")
df = pd.DataFrame(xtrain_dtm.toarray(), column =
                                    count_vect.get_feature_names())
print (df[0:5])

from sklearn.naive_bayes import MultinomialNB
df = MultinomialNB().fit(xtrain_dtm, ytrain)
predicted = df.predict(xtest_dtm)
print("In Classification 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))

from sklearn import metrics
print('In Accuracy metrics')
print("In Accuracy of the classifier is", matrices.
                    accuracy_score(ytest, predicted))
```

Teacher's Signature : _____

```
print ("Recall :", metrices.recall(ytest, predicted"))
print ("Precision:", metrices.precision_treescore
                                (ytest, predicted))
print (" Confusion matrix")
print(" metrices.confusion-matrix (ytest, predicted))
```

Teacher's Signature : _____

output:-

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
I do not like restaurant	neg
Iam tired of this stuff	neg
I can't deal with this	neg
He is my sworn enemy	neg
My boss is horrible	neg
This is an awesome place	pos
Idont like taste of this juice	neg
I love to dance	pos
Iam sick and tired of this place	neg
what a great holiday	pos
This is a bad locality to stay	neg
We will have good fun tomarrow	pos
I went to my enemy's house today	neg

Dimension of Dataset : (18, 2)

Total number of training data : (13,)
Total number of testing data : (5,)

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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

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

Total features extracted using countVectorizer : 46

Features for first 5 training instances are listed below

	about	am	an	awesome	beers	best	boss	can	deal	do ... today
0	0	0	0	0	0	0	0	0	0	1 ... 0
1	0	0	0	0	0	0	0	0	0	0 ... 1
2	0	0	0	0	0	0	0	1	1	0 ... 0
3	0	0	1	1	0	0	0	0	0	0 ... 0
4	0	0	0	0	0	0	0	0	0	0 ... 0

	tomarrow	very	view	we	went	what	will	with	work
0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	1	0	0	0	0
2	0	0	0	0	0	0	0	1	0
3	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0

[5 rows x 46 columns]

Classification results of testing samples are given below:

I love to dance → pos

I am sick and tired of this place → neg

This is an amazing place → pos

what a great holiday → pos

This is a bad locality to stay → ~~pos~~ neg

Accuracy metrices

Accuracy of the decision is 1.0

Recall : 1.0

Precision : 1.0

Confusion Matrix :

[[2 0]
 [0 3]]