

7 Assuming a set of documents that need to be classified, use one naive Bayesian classification model to perform this task. Build a JAVA class/API can be used to create the program, calculate the accuracy, precision and recall for your datasets

```

→ import pandas as pd
msg = pd.read_csv('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 used
      below")
x5, y5 = x[0:5], msg['labelnum'][0:5]
for x, y in zip(x5, y5):
      print(x, ' ', y)

```



```
from sklearn.model_selection import  
train_test_split
```

```
x_train, x_test, y_train, y_test = train_test_split  
X, y)
```

```
print("Dataset is split into Training  
and Testing samples")
```

```
print("Total training instances :",  
x_train.shape[0])
```

```
print("Total testing instances :",  
x_test.shape[0])
```

```
from sklearn.feature_extraction import  
CountVecorizer
```

```
count_vec = CountVecorizer()
```

```
x_train_dtm = count_vec.fit_transform  
x_train)
```

```
x_test_dtm = count_vec.transform  
x_test)
```

```
print("Total features extracted  
using CountVecorizer :")
```

```
x_train_dtm.shape[1])
```

```
print("Features for first 5 training  
instances are listed below")
```

```
df = pd.DataFrame(x_train_dtm.  
toarray(), columns = count_vec.  
get_feature_names())
```

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```
print(df.co:57)
```

```
from sklearn.metrics import  
multinomialNB
```

```
df = multinomialNB().fit(train_data,  
                           train_labels)
```

```
predicted = df.predict(test_data)
```

```
print("\n Classification results of  
testing samples are given  
below")
```

```
for doc, p in zip(test_data, predicted):
```

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

```
    print("%s → %s" % (doc, pred))
```

```
from sklearn.metrics
```

```
print("\n Accuracy metrics")
```

```
print("\n Accuracy of the classifier  
is", metrics.accuracy_score(test_data,  
                              predicted))
```

```
print("\n Recall:", metrics.recall_score  
      (test_data, predicted))
```

```
print("\n Precision:", metrics.precision  
      _score(test_data, predicted))
```

```
print("\n Confusion matrix")
```

```
print(metrics.confusion_matrix  
      (test_data, predicted))
```

Output:

Total instances in the dataset: 18

The message and its label of first 5 instances are used 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 video, pos

Dataset is split into training & testing samples:

Total training instances: 13

Total testing instances: 5

Total features extracted using countvectorizer: 46

Features for first 5 training instances are used below:

	about	am	an	awesome	beers	best	boss	can	deal	documented
0	0	0	0	0	0	0	0	0	0	1
1	0	0	0	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0	1	1	0
3	0	0	1	1	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0

	tomorrow	very	new	we	went	what	will	with	new
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 → neg

Accuracy metrics

Accuracy of the classifier is 1.0

Recall : 1.0

Precision : 1.0

confusion matrix:

$\begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$