

## Lab Program-6

Assuming a set of documents that need to be 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.

Algorithm to train and derive inference from the Naive Bayes model

TRAINMULTINOMIALNB( $C, D$ )

- 1  $V \leftarrow \text{ExtractVocabulary}(D)$
- 2  $N \leftarrow \text{CountDocs}(D)$
- 3 for each  $c \in C$
- 4 do  $N_c \leftarrow \text{Countdocsinclass}(D, c)$
- 5  $\text{prior}[c] \leftarrow N_c / N$
- 6  $\text{text}_c \leftarrow \text{ConcatenateTextofalldocsinclass}(D, c)$
- 7 for each  $t \in V$
- 8 do  $T_{ct} \leftarrow \text{Counttokensofterms}(\text{text}_c, t)$
- 9 for each  $d \in V$
- 10 do  $\text{condprob}[t][c] \leftarrow \frac{T_{ct} + 1}{\sum_i (T_{ci} + 1)}$
- 11 return  $V, \text{prior}, \text{condprob}$



Apply Multinomial NB( $C, V, \text{prior}, \text{cond prob}, d$ )

- 1  $w \leftarrow \text{Extract Tokens from doc } (V, d)$
- 2 for each  $c \in C$
- 3 do  $\text{score}[c] \leftarrow \log \text{prior}[c]$
- 4 for each  $t \in w$
- 5 do  $\text{score}[c] += \log \text{cond prob}[t][c]$
- 6 return arg max $_c \text{score}[c]$

$$P(t/c) = \frac{T_{ct} + 1}{\sum_{t' \in V} (T_{ct'} + 1)} = \frac{T_{ct} + 1}{(\sum_{t' \in V} T_{ct'}) + B}$$

Program-

```
from sklearn.datasets import fetch_20newsgroups
from sklearn.metrics import confusion_matrix
from sklearn.metrics import classification_report
import numpy as np
categories = ['alt.atheism', 'soc.religion.christian',
              'comp.graphics', 'sci.med']
```

```
twenty_train = fetch_20newsgroups(subset='train',
                                   categories=categories, shuffle=True)
```

```
twenty_test = fetch_20newsgroups(subset='test',
                                   categories=categories, shuffle=True)
```



```
print(len(twenty_train.data))
print(len(twenty_test.data))
print(twenty_train.target_names)
print("\n".join(twenty_train.data[0].split("\n")))
print("Target", twenty_train.target)
from sklearn.feature_extraction.text import CountVectorizer
count_vect = CountVectorizer()
X_train_tf = count_vect.fit_transform(twenty_train.data)
from sklearn.feature_extraction.text import TfidfTransformer
tfidf_transformer = TfidfTransformer()
X_train_tfidf = tfidf_transformer.fit_transform(X_train_tf)

from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score
from sklearn import metrics
mod = MultinomialNB()
mod.fit(X_train_tfidf, twenty_train.target)
X_test_tf = count_vect.transform(twenty_test.data)
X_test_tfidf = tfidf_transformer.transform(X_test_tf)
predicted = mod.predict(X_test_tfidf)
print("Predicted", predicted)
print("Accuracy", accuracy_score(twenty_test.target,
                                  predicted))

print("Classification report (twenty_test.target, predicted,
      target_names = twenty_test.target_names)")
print("Confusion matrix is", metrics.confusion_matrix(twenty_test.
      target, predicted)).
```



Total instances in the dataset: 8

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 a great holiday , pos

Dataset is split into Training and Testing samples

Total training instances : 6

Total testing instances : 2

Total features extracted using CountVectorizer: 26

Features for first 5 training instances are listed below

	about	amazing	an	beers	best	enemy	feel	fun	good	have	...	these \
0	0	1	1	0	0	0	0	0	0	...	0	
1	0	0	0	0	1	0	0	0	0	...	0	
2	1	0	0	1	0	0	1	0	1	...	1	
3	0	0	0	0	0	1	0	0	0	...	0	
4	0	0	0	0	0	0	0	0	0	...	0	

	this	to	today	tomorrow	very	we	went	will	work
0	1	0	0	0	0	0	0	0	0
1	1	0	0	0	0	0	0	0	1
2	0	0	0	0	1	0	0	0	0
3	0	1	1	0	0	0	1	0	0
4	1	0	0	0	0	0	0	0	0

[5 rows x 26 columns]

Classification results of testing samples are given below

Accuracy metrics

Accuracy of the classifier is 0.5

Recall : 1.0

Precision : 0.5

Confusion matrix

[[0 1]

[0 1]]

1