Machine Learning Lab – 20ISL68A

Program 2 – For or a given set of training data examples stored in a .CSV file, implement and demonstrate the Document classifier using Naiive Bayes.

Step 1: Importing Libraries

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn import metrics
```

Step 2: Understanding Datasets

Dataset:

```
Message Label
                This is an amazing place
                                           pos
      I feel very good about these beers
                                           pos
                    This is my best work
2
                                           pos
3
                    What an awesome view
                                           pos
4
            I do not like this restaurant
                I am tired of this stuff
5
                                          neg
                   I can't deal with this
6
                                           neg
7
                    He is my sworn enemy
                                           neg
8
                     My boss is horrible
                                          neg
                This is an awesome place
9
                                           pos
10 I do not like the taste of this juice
                                           neg
                         I love to dance
11
                                           pos
       I am sick and tired of this place
12
                                           neg
13
                    What a great holiday
                                           pos
14
          That is a bad locality to stay
                                           neg
15
          We will have good fun tomorrow
                                           pos
16
        I went to my enemy's house today
                                           neg
```

Step 3: Printing Dimensions of data set

```
print('The dimensions of the dataset',data.shape)
```

The dimensions of the dataset (17, 2)

Step 4: Converting the labels to numerical data

```
data['Labelnum']=data.Label.map({'pos':1, 'neg':0})
x=data.Message
y=data.Labelnum
print(x)
print(y)
                  This is an amazing place
1
        I feel very good about these beers
2
                      This is my best work
                     What an awesome view
             I do not like this restaurant
4
5
                 I am tired of this stuff
6
                    I can't deal with this
7
                      He is my sworn enemy
8
                      My boss is horrible
9
                  This is an awesome place
10
   I do not like the taste of this juice
11
                          I love to dance
12
        I am sick and tired of this place
13
                     What a great holiday
            That is a bad locality to stay
15
            We will have good fun tomorrow
16
          I went to my enemy's house today
Name: Message, dtype: object
0
     1
1
     1
2
     1
3
     1
4
     0
5
     0
6
     0
7
8
9
     1
10
     0
11
     1
12
13
14
     0
15
     1
Name: Labelnum, dtype: int64
```

Step 5: Converting the message to numerical data

```
vectorizer=TfidfVectorizer()
data=vectorizer.fit_transform(x)
```

Step 6: Displaying TFIDF features

```
print("\n The TFIDF features of Dataset:\n")
df=pd.DataFrame(data.toarray(), columns=vectorizer.get_feature_names())
df.head()
```

The TFIDF features of Dataset:

	about	am	amazing	an	and	awesome	bad	beers	best	boss	 today	tomorrow	very	view	we	went	what	will	with
0	0.000000	0.0	0.589549	0.461737	0.0	0.000000	0.0	0.000000	0.000000	0.0	 0.0	0.0	0.000000	0.000000	0.0	0.0	0.000000	0.0	0.0
1	0.416578	0.0	0.000000	0.000000	0.0	0.000000	0.0	0.416578	0.000000	0.0	 0.0	0.0	0.416578	0.000000	0.0	0.0	0.000000	0.0	0.0
2	0.000000	0.0	0.000000	0.000000	0.0	0.000000	0.0	0.000000	0.562609	0.0	 0.0	0.0	0.000000	0.000000	0.0	0.0	0.000000	0.0	0.0
3	0.000000	0.0	0.000000	0.442107	0.0	0.492899	0.0	0.000000	0.000000	0.0	 0.0	0.0	0.000000	0.564485	0.0	0.0	0.492899	0.0	0.0
4	0.000000	0.0	0.000000	0.000000	0.0	0.000000	0.0	0.000000	0.000000	0.0	 0.0	0.0	0.000000	0.000000	0.0	0.0	0.000000	0.0	0.0

5 rows × 55 columns

Step 7: Dividing dataset into training and testing data

```
print("\n Train Test Split:\n")
xtrain,xtest,ytrain,ytest=train_test_split(data,y,test_size=0.3,random_state=2)
print('\n The total number of Training Data:',ytrain.shape)
print('\n The total number of Test Data:',ytest.shape)

Train Test Split:

The total number of Training Data: (11,)
The total number of Test Data: (6,)
```

Step 8: Training Naïve Bayes classifier on training data

```
clf= MultinomialNB().fit(xtrain, ytrain)
predicted = clf.predict(xtest)

#printing accuracy, Confusion matrix, Precision and Recall
print("\n Accuracy of the classifier is:", metrics.accuracy_score(ytest,predicted))
print("\nConfision Matrix is:", metrics.confusion_matrix(ytest,predicted))
print("\nClassification Report:", metrics.classification_report(ytest,predicted))
print("\nThe value of Precision :", metrics.precision_score(ytest,predicted))
print("\nThe value of Recall:", metrics.recall_score(ytest,predicted))
```

Accuracy of the classifier is: 0.6666666666666666

Confision Matrix is: [[3 0] [2 1]]

Classification Report: precision recall f1-score support 1.00 0.75 0 0.60 1.00 0.33 0.50 0.67 6 accuracy 0.80 0.67 0.62 6 macro avg 0.80 0.67 0.62 weighted avg

The value of Precision : 1.0

The value of Recall: 0.33333333333333333

Description

- ➤ Naïve Bayes methods are a set of supervised learning algorithms
- ➤ Applications → Real time prediction, Multiclass prediction
- ➤ 3 types of Naïve Bayes → Gaussian, Multinomial, Bernoulli
- ➤ Tfidf Vectorizer → Term Frequency Inverse Document Frequency