EX.NO.: 15

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NAIVE BAYES ON NLTK MOVIE REVIEW DATASET

To implement a Naïve Bayes classifier using the NLTK library for text classification on two datasets:

- 1. Movie Reviews Dataset from NLTK.
- 2. **20 Newsgroups Dataset** from scikit-learn.

PROCEDURE:

Part 1: Movie Reviews Classification

- 1. Import necessary libraries (nltk, random).
- 2. Download and load the movie reviews dataset from NLTK.
- 3. Store reviews as (word_list, category) tuples and shuffle them.
- 4. Define a document features() function to extract word features.
- 5. Convert data into feature sets using document features().
- 6. Split the dataset into training (80%) and testing (20%) sets.
- 7. Train a Naïve Bayes Classifier using the training set.
- 8. Evaluate model performance using **accuracy** on the test set.
- 9. Display the **most informative features** used by the classifier.

Part 2: 20 Newsgroups Classification

- 1. Import required libraries (sklearn.datasets, nltk).
- 2. Load the **20 Newsgroups** dataset (subset of 4 categories).
- 3. Remove headers, footers, and quotes to clean the text.
- 4. Convert text data into numerical features using CountVectorizer (bag-of-words).
- 5. Apply TfidfTransformer to normalize feature importance.
- 6. Convert feature matrices into **NLTK-compatible dictionaries**.
- 7. Split the dataset into **training and testing** sets.
- 8. Train a Naïve Bayes Classifier using the processed feature sets.
- 9. Evaluate the model using **accuracy** on the test set.
- 10. Display the **most informative words** influencing classification.

CODE AND OUTPUT

```
lef document features(words):
    return {word: True for word in words}
feature sets = [(document features(doc), category) for (doc, category) in documents]
train set, test set = feature sets[:1600], feature sets[1600:]
classifier = NaiveBayesClassifier.train(train set)
print("Accuracy:", accuracy(classifier, test set))
classifier.show most informative features(10)
 [nltk data] Downloading package movie reviews to
            C:\Users\Hema\AppData\Roaming\nltk_data...
 [nltk data]
 [nltk_data] Package movie_reviews is already up-to-date!
Accuracy: 0.735
Most Informative Features
            insulting = True
                                                15.8 : 1.0
                                  neg:pos =
         breathtaking = True
                                  pos : neg = 14.1 : 1.0
              avoids = True
                                                12.9 : 1.0
                                  pos:neg =
                                                 12.9 : 1.0
           vulnerable = True
                                  pos:neg =
            ludicrous = True
                                  neg:pos =
                                                 12.7 : 1.0
            stupidity = True
                                  neg: pos
                                                 12.3 : 1.0
            chilling = True
                                   pos:neg =
                                                 12.2 : 1.0
              seagal = True
                                                 11.1 : 1.0
                                  neg : pos
           astounding = True
                                                 10.9 : 1.0
                                  pos:neg =
                                  neg : pos = 10.4 : 1.0
           schumacher = True
  com sklearn.datasets
from sklearn.feature extraction.text import CountVectorizer, TfidfTransformer
import nltk
categories = ['alt.atheism', 'comp.graphics', 'sci.med', 'soc.religion.christian']
newsgroups train = fetch 20newsgroups(subset='train', categories=categories,
remove=('headers', 'footers', 'quotes'))
newsgroups test = fetch 20newsgroups(subset='test', categories=categories,
remove=('headers', 'footers', 'quotes'))
vectorizer = CountVectorizer(stop words='english', max features=3000)
X train counts = vectorizer.fit transform(newsgroups train.data)
X test counts = vectorizer.transform(newsgroups test.data)
tfidf transformer = TfidfTransformer()
X train tfidf = tfidf transformer.fit transform(X train counts)
 test tfidf = tfidf transformer.transform(X test counts)
```

```
train features = [(dict(zip(vectorizer.get feature names out(), row.toarray()[0])),
category)
                  for row, category in zip(X train tfidf, newsgroups train.target)]
test features = [(dict(zip(vectorizer.get feature names out(), row.toarray()[0])),
category)
                 for row, category in zip(X test tfidf, newsgroups test.target)]
classifier = NaiveBayesClassifier.train(train features)
print("Accuracy:", accuracy(classifier, test features))
classifier.show most informative features(10)
 Accuracy: 0.2669773635153129
 Most Informative Features
                   god = np.float64(0.0)
                                          2:3
                                                         1.9:1.0
                people = np.float64(0.0)
                                          1:3
                                                          1.5 : 1.0
                 jesus = np.float64(0.0)
                                                          1.4:1.0
                                          2:3
                   say = np.float64(0.0)
                                                          1.3 : 1.0
                   don = np.float64(0.0)
                                         1:0
                                                         1.3 : 1.0
               believe = np.float64(0.0)
                                         1:3
                                                         1.3:1.0
                christ = np.float64(0.0)
                                          2:3
                                                          1.3:1.0
               graphics = np.float64(0.0)
                                         3:1
                                                         1.3:1.0
              christian = np.float64(0.0)
                                         2:3
                                                          1.3:1.0
                 bible = np.float64(0.0)
                                                          1.3:1.0
```