naive-bayes-and-text-mining

November 25, 2024

```
[1]: import pandas as pd
     df = pd.read_csv('/content/blogs.csv')
[2]: print(df.head())
     print(df.info())
     print(df.describe())
                                                     Data
                                                                Labels
    O Path: cantaloupe.srv.cs.cmu.edu!magnesium.club... alt.atheism
    1 Newsgroups: alt.atheism\nPath: cantaloupe.srv... alt.atheism
    2 Path: cantaloupe.srv.cs.cmu.edu!das-news.harva... alt.atheism
    3 Path: cantaloupe.srv.cs.cmu.edu!magnesium.club... alt.atheism
    4 Xref: cantaloupe.srv.cs.cmu.edu alt.atheism:53... alt.atheism
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 2000 entries, 0 to 1999
    Data columns (total 2 columns):
         Column Non-Null Count Dtype
                 2000 non-null object
     0
         Data
         Labels 2000 non-null
                                 object
    dtypes: object(2)
    memory usage: 31.4+ KB
    None
                                                          Data
                                                                     Labels
    count
                                                          2000
                                                                       2000
                                                          2000
    unique
            Path: cantaloupe.srv.cs.cmu.edu!magnesium.club... alt.atheism
    top
                                                             1
    freq
                                                                        100
[4]: import string
     from nltk.corpus import stopwords
     from nltk.tokenize import word_tokenize
     from sklearn.feature_extraction.text import TfidfVectorizer
     from nltk.stem import WordNetLemmatizer
     import nltk
     nltk.download('punkt')
     nltk.download('stopwords')
```

```
nltk.download('wordnet')
     # Download the 'punkt_tab' resource
     nltk.download('punkt_tab') # This line was added to download the missing_
      ⇔resource
     stop words = set(stopwords.words('english'))
     lemmatizer = WordNetLemmatizer()
     def preprocess_text(text):
         # Convert to lowercase
         text = text.lower()
         # Remove punctuation
         text = text.translate(str.maketrans('', '', string.punctuation))
         # Tokenize text
         tokens = word_tokenize(text)
         # Remove stopwords and lemmatize
         tokens = [lemmatizer.lemmatize(word) for word in tokens if word not in,
      ⇔stop_words]
         return ' '.join(tokens)
     df['Data'] = df['Data'].apply(preprocess_text)
    [nltk_data] Downloading package punkt to /root/nltk_data...
                  Package punkt is already up-to-date!
    [nltk_data]
    [nltk_data] Downloading package stopwords to /root/nltk_data...
                 Package stopwords is already up-to-date!
    [nltk_data]
    [nltk_data] Downloading package wordnet to /root/nltk_data...
    [nltk_data]
                 Package wordnet is already up-to-date!
    [nltk_data] Downloading package punkt_tab to /root/nltk_data...
    [nltk_data]
                 Unzipping tokenizers/punkt_tab.zip.
[5]: vectorizer = TfidfVectorizer()
     X = vectorizer.fit transform(df['Data'])
     y = df['Labels']
[6]: from sklearn.model_selection import train_test_split
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
      ⇔random_state=42)
[7]: from sklearn.naive_bayes import MultinomialNB
     from sklearn.metrics import accuracy_score, precision_score, recall_score, u
      ⇒f1_score, classification_report
     model = MultinomialNB()
     model.fit(X_train, y_train)
     y_pred = model.predict(X_test)
```

```
precision
                                              recall f1-score
                                                                   support
                   alt.atheism
                                      0.53
                                                 0.94
                                                           0.68
                                                                        18
                                      0.71
                                                 0.83
                                                           0.77
                                                                        18
                 comp.graphics
                                                                        22
      comp.os.ms-windows.misc
                                      0.86
                                                 0.82
                                                           0.84
     comp.sys.ibm.pc.hardware
                                      0.79
                                                 0.76
                                                           0.78
                                                                        25
         comp.sys.mac.hardware
                                      0.86
                                                 0.86
                                                           0.86
                                                                        21
                                                 0.80
                                                           0.89
                                                                        25
                comp.windows.x
                                      1.00
                  misc.forsale
                                      0.83
                                                 0.56
                                                           0.67
                                                                        18
                                                 0.89
                     rec.autos
                                      0.89
                                                           0.89
                                                                        18
               rec.motorcycles
                                      0.83
                                                 0.94
                                                           0.88
                                                                        16
           rec.sport.baseball
                                      0.71
                                                 0.94
                                                           0.81
                                                                        18
              rec.sport.hockey
                                                 1.00
                                                           0.94
                                                                        15
                                      0.88
                     sci.crypt
                                      0.86
                                                 0.95
                                                           0.90
                                                                        19
               sci.electronics
                                      0.67
                                                 0.75
                                                           0.71
                                                                        16
                       sci.med
                                      0.88
                                                 0.88
                                                           0.88
                                                                        17
                                                 0.81
                                                           0.89
                     sci.space
                                      1.00
                                                                        21
       soc.religion.christian
                                      0.85
                                                 0.96
                                                           0.90
                                                                        23
            talk.politics.guns
                                      0.95
                                                 0.68
                                                           0.79
                                                                        28
        talk.politics.mideast
                                                 0.95
                                                           0.95
                                                                        20
                                      0.95
                                                 0.89
                                                           0.71
            talk.politics.misc
                                      0.59
                                                                        18
            talk.religion.misc
                                      0.83
                                                 0.21
                                                           0.33
                                                                        24
                                                           0.81
                                                                       400
                      accuracy
                     macro avg
                                      0.82
                                                 0.82
                                                           0.80
                                                                       400
                  weighted avg
                                      0.83
                                                 0.81
                                                           0.80
                                                                       400
 [9]: from textblob import TextBlob
[10]: def get_sentiment(text):
          analysis = TextBlob(text)
          if analysis.sentiment.polarity > 0:
              return 'positive'
          elif analysis.sentiment.polarity == 0:
              return 'neutral'
          else:
              return 'negative'
      df['Sentiment'] = df['Data'].apply(get_sentiment)
[11]: sentiment_distribution = df.groupby(['Labels', 'Sentiment']).size().unstack().
       →fillna(0)
      print(sentiment_distribution)
```

[8]: print(classification_report(y_test, y_pred))

Sentiment	${\tt negative}$	neutral	positive
Labels			
alt.atheism	36.0	0.0	64.0
comp.graphics	28.0	0.0	72.0
comp.os.ms-windows.misc	23.0	0.0	77.0
<pre>comp.sys.ibm.pc.hardware</pre>	18.0	0.0	82.0
comp.sys.mac.hardware	26.0	0.0	74.0
comp.windows.x	20.0	2.0	78.0
misc.forsale	20.0	0.0	80.0
rec.autos	24.0	0.0	76.0
rec.motorcycles	29.0	0.0	71.0
rec.sport.baseball	38.0	0.0	62.0
rec.sport.hockey	44.0	0.0	56.0
sci.crypt	23.0	0.0	77.0
sci.electronics	24.0	0.0	76.0
sci.med	34.0	0.0	66.0
sci.space	29.0	0.0	71.0
soc.religion.christian	25.0	0.0	75.0
talk.politics.guns	41.0	1.0	58.0
talk.politics.mideast	27.0	0.0	73.0
talk.politics.misc	25.0	0.0	75.0
talk.religion.misc	21.0	0.0	79.0

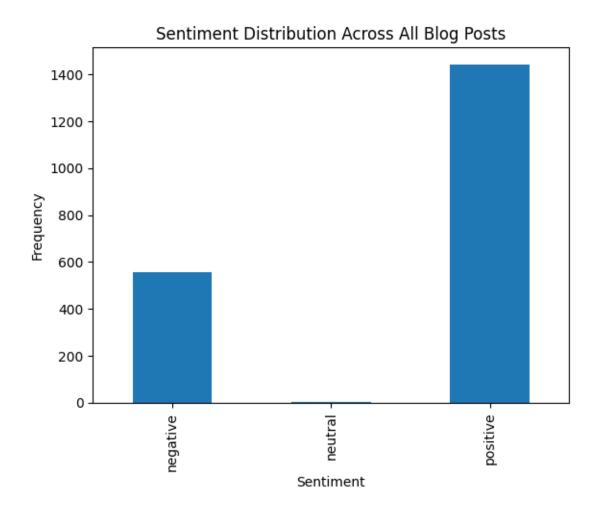
```
[12]: print(f"Accuracy: {accuracy_score(y_test, y_pred)}")
    print(f"Precision: {precision_score(y_test, y_pred, average='weighted')}")
    print(f"Recall: {recall_score(y_test, y_pred, average='weighted')}")
    print(f"F1-Score: {f1_score(y_test, y_pred, average='weighted')}")
```

Accuracy: 0.8075

Precision: 0.8326691647812972

Recall: 0.8075

F1-Score: 0.7991740766856441



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