P.Janaki-2211CS020388-AIML-THETA

!pip install nltk scikit-learn seaborn

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Requirement already satisfied: nltk in /usr/local/lib/python3.11/dist-packages (3.9.1
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.11/dist-package
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Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.11/dist
Requirement already satisfied: pandas>=1.2 in /usr/local/lib/python3.11/dist-packages
Requirement already satisfied: matplotlib!=3.6.1,>=3.4 in /usr/local/lib/python3.11/d
Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.11/dist-pac
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Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11/dist-pack
Requirement already satisfied: pillow>=8 in /usr/local/lib/python3.11/dist-packages (
Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.11/dist-pac
Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.11/dist
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-package
Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packa
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (f
```

```
# Import necessary libraries
import pandas as pd
import numpy as np
import re
import nltk
from nltk.corpus import stopwords
from nltk.tokenize import word tokenize
from nltk.stem import WordNetLemmatizer
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.linear model import LogisticRegression
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, conf
import seaborn as sns
import matplotlib.pyplot as plt
# Download necessary NLTK resources
nltk.download('punkt')
nltk.download('stopwords')
nltk.download('wordnet')
```

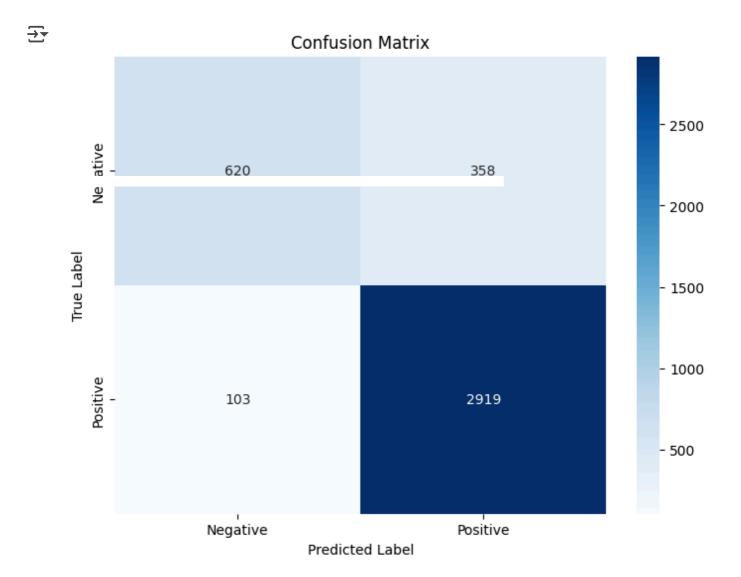
[nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data] Package punkt is already up-to-date!

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk_data]
                    Package stopwords is already up-to-date!
     [nltk data] Downloading package wordnet to /root/nltk data...
                    Package wordnet is already up-to-date!
     [nltk data]
     True
from google.colab import files
uploaded = files.upload()
# Load the dataset
df = pd.read csv('amazon.csv') # Ensure 'amazon.csv' is the correct file name
      Choose Files No file chosen
                                         Upload widget is only available when the cell has been
     executed in the current browser session. Please rerun this cell to enable.
     Saving amazon.csv to amazon.csv
# Check the first few rows of the dataset
df.head()
→
                                             Text label
                                                       1
         This is the best apps acording to a bunch of ...
      1 This is a pretty good version of the game for ...
                                                       1
      2
           this is a really . there are a bunch of levels...
                                                       1
      3
         This is a silly game and can be frustrating, b...
      4 This is a terrific game on any pad. Hrs of fun...
                                                       1
# Download necessary NLTK resources
import nltk
nltk.download('punkt')
nltk.download('stopwords')
nltk.download('wordnet')
→ [nltk data] Downloading package punkt to /root/nltk data...
     [nltk data] Package punkt is already up-to-date!
     [nltk_data] Downloading package stopwords to /root/nltk_data...
     [nltk data] Package stopwords is already up-to-date!
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [nltk data] Package wordnet is already up-to-date!
     True
import nltk
nltk.download('punkt') # Ensure the correct 'punkt' resource is downloaded
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk data]
                    Package punkt is already up-to-date!
     True
```

```
import nltk
nltk.data.path.append('/usr/share/nltk_data') # Add the resource path
nltk.download('punkt', download_dir='/usr/share/nltk_data') # Force download to this dir
→ [nltk_data] Downloading package punkt to /usr/share/nltk_data...
     [nltk_data] Unzipping tokenizers/punkt.zip.
     True
import nltk
# Add the resource path to NLTK
nltk.data.path.append('/usr/share/nltk_data') # Ensure nltk looks in the correct directo
# Download the punkt resource and punkt_tab resource
nltk.download('punkt', download_dir='/usr/share/nltk_data') # Force download to the spec
nltk.download('punkt_tab', download_dir='/usr/share/nltk_data') # Ensure punkt_tab is al
→ [nltk_data] Downloading package punkt to /usr/share/nltk_data...
     [nltk_data] Package punkt is already up-to-date!
     [nltk_data] Downloading package punkt_tab to /usr/share/nltk_data...
     [nltk_data] Unzipping tokenizers/punkt_tab.zip.
     True
Start coding or generate with AI.
Start coding or generate with AI.
# Data Preprocessing Functions
# Clean the text data
def clean_text(text):
   text = re.sub(r'\W', ' ', text) # Remove non-alphanumeric characters
   text = re.sub(r'\s+', ' ', text) # Remove extra spaces
   text = text.lower() # Convert to lowercase
    text = re.sub(r'\d+', '', text) # Remove numbers
    stop words = set(stopwords.words('english')) # Stopwords
    text = ' '.join([word for word in text.split() if word not in stop words])
    return text
# Lemmatize words
lemmatizer = WordNetLemmatizer()
def lemmatize words(text):
    tokens = word_tokenize(text)
    return ' '.join([lemmatizer.lemmatize(word) for word in tokens])
# Apply the cleaning and lemmatization functions
# Apply the cleaning and lemmatization functions
df['cleaned_reviews'] = df['Text'].apply(clean_text) # Use 'Text' as the column name
```

```
df['lemmatized_reviews'] = df['cleaned_reviews'].apply(lemmatize_words)
# Vectorization using TF-IDF
vectorizer = TfidfVectorizer(max_features=5000)
X = vectorizer.fit_transform(df['lemmatized_reviews']).toarray()
# Split the data into train and test sets
X_train, X_test, y_train, y_test = train_test_split(X, df['label'], test_size=0.2, random
# Initialize the Logistic Regression model
model = LogisticRegression()
# Train the model
model.fit(X_train, y_train)
      ▼ LogisticRegression ① ?
     LogisticRegression()
# Predict the sentiment of the test set
y_pred = model.predict(X_test)
# Evaluate model performance
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred, average='binary') # Adjust for multiclass if
recall = recall_score(y_test, y_pred, average='binary')
f1 = f1_score(y_test, y_pred, average='binary')
print(f'Accuracy: {accuracy}')
print(f'Precision: {precision}')
print(f'Recall: {recall}')
print(f'F1 Score: {f1}')
→ Accuracy: 0.88475
     Precision: 0.8907537381751602
     Recall: 0.9659166115155526
     F1 Score: 0.9268137799650739
# Generate and visualize the confusion matrix
cm = confusion matrix(y test, y pred)
plt.figure(figsize=(8, 6))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['Negative', 'Positive'],
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
```

plt.show()



```
import joblib

# Save the model
joblib.dump(model, 'sentiment_model.pkl')

# Save the vectorizer
joblib.dump(vectorizer, 'vectorizer.pkl')

# ['vectorizer.pkl']

def predict_sentiment(text):
    cleaned_text = clean_text(text)
    lemmatized_text = lemmatize_words(cleaned_text)
    vectorized_text = vectorizer.transform([lemmatized_text]).toarray(
    return model.predict(vectorized_text)

# Test with a new review
new_review = "This product is amazing! Highly recommend it."
sentiment = predict_sentiment(new_review)
```

```
NLP_2211CS020384_PROJECT (3) (1).ipynb - Colab
print(f"The sentiment of the review is: {sentiment[0]}")
The sentiment of the review is: 1
import joblib
# Save the trained model
joblib.dump(model, 'sentiment_model.pkl')
# Save the TF-IDF vectorizer
joblib.dump(vectorizer, 'vectorizer.pkl')
→ ['vectorizer.pkl']
!zip sentiment_analysis_files384.zip sentiment_model.pkl vectorizer.pkl
\rightarrow
       adding: sentiment_model.pkl (deflated 5%)
       adding: vectorizer.pkl (deflated 72%)
!1s
     amazon.csv sentiment_analysis_files384.zip sentiment_model.pkl
     sample_data sentiment_analysis_files.zip vectorizer.pkl
from google.colab import files
files.download('sentiment_analysis_files384.zip')
\rightarrow
```