## nlp-assignment

September 2, 2024

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def preprocess\_text(text):

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[3]: import pandas as pd
     import numpy as np
     import nltk
     from nltk.corpus import stopwords
     from nltk.tokenize import word_tokenize
     from nltk.stem import WordNetLemmatizer
     import re
     from sklearn.model_selection import train_test_split
     from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
     from sklearn.linear_model import LogisticRegression
     from sklearn.svm import SVC
     from sklearn.ensemble import RandomForestClassifier
     from sklearn.metrics import classification_report, confusion_matrix
     import seaborn as sns
     import matplotlib.pyplot as plt
[4]: df = pd.read_csv('IMDB Dataset.csv')
     df.head()
[4]:
                                                   review sentiment
     One of the other reviewers has mentioned that ... positive
     1 A wonderful little production. <br /><br />The... positive
     2 I thought this was a wonderful way to spend ti... positive
     3 Basically there's a family where a little boy ... negative
     4 Petter Mattei's "Love in the Time of Money" is... positive
[8]: # Preprocessing
     nltk.download('punkt')
     nltk.download('wordnet')
     nltk.download('stopwords')
     stop_words = set(stopwords.words('english'))
     lemmatizer = WordNetLemmatizer()
```

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# Remove URLs
         text = re.sub(r"http\S+|www\S+|https\S+", '', text, flags=re.MULTILINE)
         # Remove punctuation and numbers
         text = re.sub(r'\W', '', text)
         # Tokenization
         tokens = word_tokenize(text)
         # Remove stopwords and lemmatize
         tokens = [lemmatizer.lemmatize(word.lower()) for word in tokens if word.
      →lower() not in stop_words]
         return ' '.join(tokens)
     # Apply preprocessing to the 'review' column
     df['review'] = df['review'].apply(preprocess text)
     df.head()
    [nltk_data] Downloading package punkt to
                    C:\Users\samee\AppData\Roaming\nltk_data...
    [nltk data]
    [nltk_data]
                  Package punkt is already up-to-date!
    [nltk_data] Downloading package wordnet to
    [nltk_data]
                    C:\Users\samee\AppData\Roaming\nltk_data...
    [nltk data]
                  Package wordnet is already up-to-date!
    [nltk_data] Downloading package stopwords to
    [nltk data]
                    C:\Users\samee\AppData\Roaming\nltk data...
                  Package stopwords is already up-to-date!
    [nltk_data]
[8]:
                                                   review sentiment
     O one reviewer mentioned watching 1 oz episode h... positive
     1 wonderful little production br br filming tech... positive
     2 thought wonderful way spend time hot summer we... positive
     3 basically family little boy jake think zombie ... negative
     4 petter mattei love time money visually stunnin... positive
[9]: # Vectorization
     # Split the data into training and testing sets
     X_train, X_test, y_train, y_test = train_test_split(df['review'],_
     ⇔df['sentiment'], test_size=0.2, random_state=42)
     # Vectorization
     count_vect = CountVectorizer()
     tfidf_vect = TfidfVectorizer()
     X_train_count = count_vect.fit_transform(X_train)
     X_test_count = count_vect.transform(X_test)
```

```
X_train_tfidf = tfidf_vect.fit_transform(X_train)
X_test_tfidf = tfidf_vect.transform(X_test)
```

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[10]: # Step 6: Model Training and Evaluation
      def evaluate_model(model, X_train, X_test, y_train, y_test):
          model.fit(X_train, y_train)
          y_pred = model.predict(X_test)
          # Print classification report
          print(classification_report(y_test, y_pred))
          # Plot confusion matrix
          cm = confusion_matrix(y_test, y_pred, labels=model.classes_)
          sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=model.
       ⇔classes_, yticklabels=model.classes_)
          plt.ylabel('Actual')
          plt.xlabel('Predicted')
          plt.show()
      # Logistic Regression
      print("Logistic Regression with CountVectorizer")
      evaluate_model(LogisticRegression(), X_train_count, X_test_count, y_train,_u

y_test)

      print("Logistic Regression with TfidfVectorizer")
      evaluate_model(LogisticRegression(), X_train_tfidf, X_test_tfidf, y_train,_u

y_test)

      # SVC
      print("SVC with CountVectorizer")
      evaluate_model(SVC(), X_train_count, X_test_count, y_train, y_test)
      print("SVC with TfidfVectorizer")
      evaluate_model(SVC(), X_train_tfidf, X_test_tfidf, y_train, y_test)
      # Random Forest
      print("Random Forest with CountVectorizer")
      evaluate_model(RandomForestClassifier(), X_train_count, X_test_count, y_train,_

y_test)

      print("Random Forest with TfidfVectorizer")
      evaluate_model(RandomForestClassifier(), X_train_tfidf, X_test_tfidf, y_train,_

y_test)
```

Logistic Regression with CountVectorizer

 $\label{local-Packages-PythonSoftwareFoundation.Python.3.12_qbz5n} C: \Users \same \App Data \Local \Packages \PythonSoftware Foundation. Python. 3.12_qbz5n$ 

2kfra8p0\LocalCache\local-packages\Python312\site-packages\sklearn\linear\_model\\_logistic.py:469: ConvergenceWarning: lbfgs failed to converge (status=1):

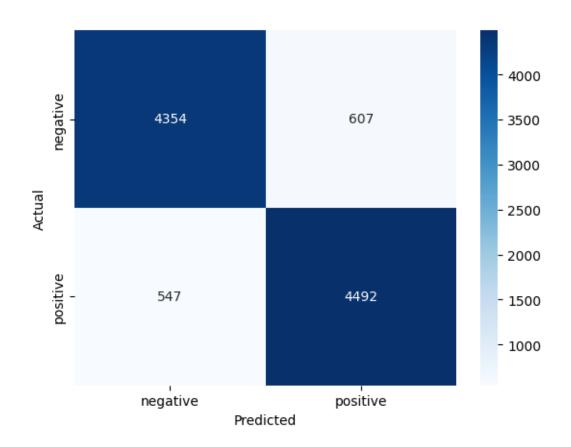
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max\_iter) or scale the data as shown in:
 https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:

 $\verb|https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression| \\$ 

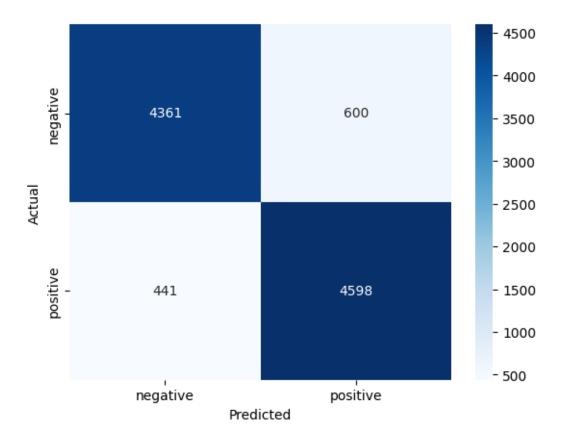
n\_iter\_i = \_check\_optimize\_result(

pr	ecision	recall	f1-score	support
negative	0.89	0.88	0.88	4961
positive	0.88		0.89	5039
accuracy			0.88	10000
macro avg	0.88	0.88	0.88	10000
weighted avg	0.88	0.88	0.88	10000



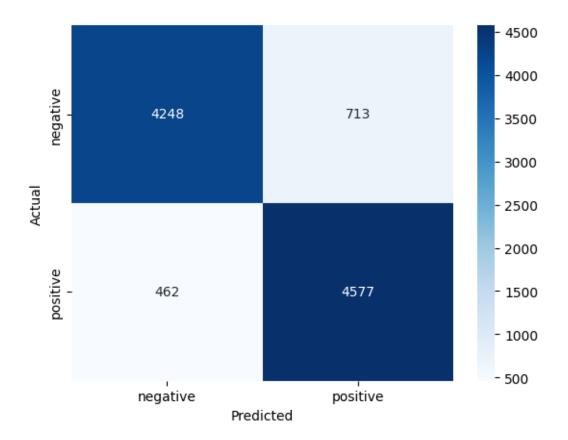
Logistic Regression with TfidfVectorizer

	precision	recall	f1-score	support
negative	0.91	0.88	0.89	4961
positive	0.88	0.91	0.90	5039
accuracy			0.90	10000
macro avg	0.90	0.90	0.90	10000
weighted avg	0.90	0.90	0.90	10000

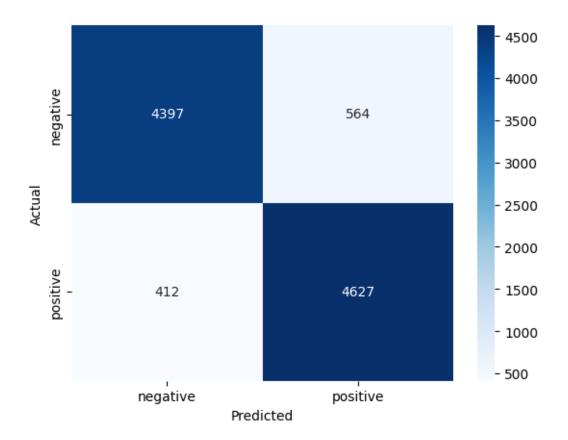


SVC with CountVectorizer

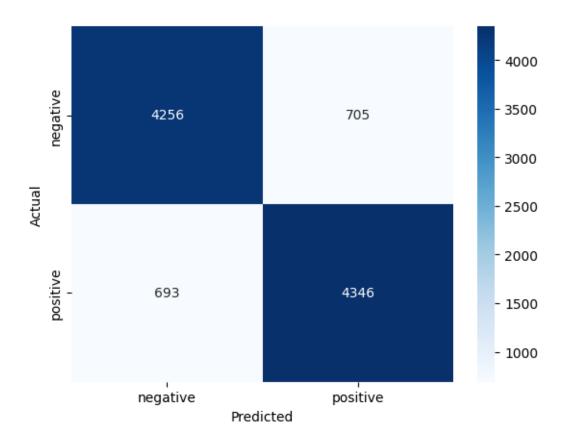
	precision	recall	f1-score	support
negative positive	0.90 0.87	0.86 0.91	0.88	4961 5039
accuracy macro avg weighted avg	0.88 0.88	0.88 0.88	0.88 0.88 0.88	10000 10000 10000



SVC with Tfid	fVectorizer			
	precision	recall	f1-score	support
negative	0.91	0.89	0.90	4961
positive	0.89	0.92	0.90	5039
accuracy			0.90	10000
macro avg	0.90	0.90	0.90	10000
weighted avg	0.90	0.90	0.90	10000



Random Forest with CountVectorizer				
	precision	recall	f1-score	support
negative	0.86	0.86	0.86	4961
positive	0.86	0.86	0.86	5039
accuracy			0.86	10000
macro avg	0.86	0.86	0.86	10000
weighted avg	0.86	0.86	0.86	10000



Random Forest	with TfidfV	ectorizer		
	precision	recall	f1-score	support
negative	0.85	0.87	0.86	4961
positive	0.86	0.85	0.86	5039
accuracy			0.86	10000
macro avg	0.86	0.86	0.86	10000
weighted avg	0.86	0.86	0.86	10000

