Name - Parth Dipak Kulkarni PRN - 202201040007 Program - TY B.Tech. Computer Engineering Course - MDM Deep Learning Practical Assignment 4

NLP Preprocessing And Text Classification

- 1. Understand and apply NLP preprocessing techniques such as tokenization, stopword removal, stemming, and lemmatization.
- 2. Implement text vectorization techniques such as TF-IDF and CountVectorizer.
- 3. Develop a text classification model using a machine learning algorithm.
- 4. Evaluate the performance of the model using suitable metrics.

Colab File:- Parth Kulkarni NLP Lab_Assignmnent_4.ipynb

Dataset: Emotion Dataset for Emotion Recognition

Train Dataset:-

https://drive.google.com/file/d/1GctUwtQdJrWIqkqpSVnQtheSezzL3mNf/view?usp=sharing

Test Dataset:- https://drive.google.com/file/d/1MQm8t97E3nzDuwhIqF5vYHg_b-swlL8/view?usp=sharing

Validation Dataset:-

https://drive.google.com/file/d/1wdsCfVtx5PrKWGTzy6xcRcKeHiIPgeXd/view?usp=sharing

☐ Google Colab Lab Assignment -NLP

Course Name: Deep Learning (MDM)_Labs DSG-SSB Lab Title: NLP Techniques for Text Classification

Student Name: Parth Kulkarni Student ID:202201040007 Date of Submission: 01/04/2025 Group Members: Rohit Jagtap

Rohan Wagh Rohan Agarwal

Objective The objective of this assignment is to implement NLP preprocessing techniques and build a text classification model using machine learning techniques.

Learning Outcomes:

- 1. Understand and apply NLP preprocessing techniques such as tokenization, stopword removal, stemming, and lemmatization.
- 2. Implement text vectorization techniques such as TF-IDF and CountVectorizer.
- 3. Develop a text classification model using a machine learning algorithm.
- 4. Evaluate the performance of the model using suitable metrics.

☐ Assignment Instructions:

Part 1: NLP Preprocessing

Dataset Selection:

Choose any text dataset from **Best Datasets for Text** https://en.innovatiana.com/post/best-datasets-for-text-classification Classification, such as SMS Spam Collection, IMDb Reviews, or any other relevant dataset.

Download the dataset and upload it to Google Colab.

Load the dataset into a Pandas DataFrame and explore its structure (e.g., check missing values, data types, and label distribution).

Text Preprocessing:

Convert text to lowercase.

Perform tokenization using NLTK or spaCy.

Remove stopwords using NLTK or spaCy.

Apply stemming using PorterStemmer or SnowballStemmer.

Apply lemmatization using WordNetLemmatizer.

Vectorization Techniques:

Convert text data into numerical format using TF-IDF and CountVectorizer.

```
1 from google.colab import drive
2 drive.mount('/content/drive')

The Mounted at /content/drive

1 import pandas as pd
2
3 df_train = pd.read_csv('/content/drive/MyDrive/training.csv')
4 df_test = pd.read_csv('/content/drive/MyDrive/test.csv')
5 df_valid = pd.read_csv('/content/drive/MyDrive/validation.csv')
6
7 print("Training Set:")
8 print(df_train.info(), "\n")
9 print(df_train.head(), "\n")
10
11 print("Testing Set:")
12 print(df_test.info(), "\n")
13 print(df_test.head(), "\n")
```

```
15 print("Validation Set:")
16 print(df_valid.info(), "\n")
17 print(df_valid.head(), "\n")
19 print("Training Set Labels:\n", df_train['label'].value_counts(), "\n")
20 print("Testing Set Labels:\n", df_test['label'].value_counts(), "\n")
21 print("Validation Set Labels:\n", df_valid['label'].value_counts(), "\n")
→ Training Set:
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 16000 entries, 0 to 15999
    Data columns (total 2 columns):
     # Column Non-Null Count Dtype
        ----
                 -----
                 16000 non-null object
     0
        text
     1 label 16000 non-null int64
    dtypes: int64(1), object(1)
    memory usage: 250.1+ KB
    None
                                                   text label
                                i didnt feel humiliated
                                                             0
    1 i can go from feeling so hopeless to so damned...
       im grabbing a minute to post i feel greedy wrong
                                                             3
    3 \, i am ever feeling nostalgic about the fireplac...
                                   i am feeling grouchy
    Testing Set:
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 2000 entries, 0 to 1999
    Data columns (total 2 columns):
     # Column Non-Null Count Dtype
                 2000 non-null
     0 text
        label
                 2000 non-null
                                int64
    dtypes: int64(1), object(1)
    memory usage: 31.4+ KB
    None
                                                   text label
    0 im feeling rather rotten so im not very ambiti...
                                                             0
              im updating my blog because i feel shitty
                                                             a
    2 i never make her separate from me because i do...
                                                             0
    3
      i left with my bouquet of red and yellow tulip...
                                                             1
       i was feeling a little vain when i did this one
    Validation Set:
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 2000 entries, 0 to 1999
    Data columns (total 2 columns):
     # Column Non-Null Count Dtype
        -----
     0
        text
                 2000 non-null
                                object
     1 label 2000 non-null
    dtypes: int64(1), object(1)
    memory usage: 31.4+ KB
    None
                                                   text label
    0 im feeling quite sad and sorry for myself but ...
    1 i feel like i am still looking at a blank canv...
                                                             0
                         i feel like a faithful servant
                                                             2
                      i am just feeling cranky and blue
                                                             3
    4 i can have for a treat or if i am feeling festive
                                                             1
    Training Set Labels:
```

Text Processing

```
1 import shutil
2 import os
3 import nltk
4 nltk_data_path = os.path.expanduser('~/nltk_data')
5 shutil.rmtree(nltk_data_path, ignore_errors=True)

1 import nltk
2 nltk.download('punkt')
3 nltk.download('stopwords')
4 nltk.download('wordnet')
5 nltk.download('omw-1.4')
6 nltk.download('averaged_perceptron_tagger') # For better tokenization
7

[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!
    [nltk_data] Downloading package omw-1.4 to /root/nltk_data...
    [nltk_data] Downloading package averaged_perceptron_tagger to
    [nltk_data]
                  /root/nltk_data...
   [nltk_data] Unzipping taggers/averaged_perceptron_tagger.zip.
   True
1 nltk.download('stopwords')
  [nltk_data] Downloading package stopwords to /root/nltk_data...
    [nltk_data]
                Package stopwords is already up-to-date!
   True
1 nltk.download('punkt')
  [nltk data] Downloading package punkt to /root/nltk data...
    [nltk_data] Package punkt is already up-to-date!
2 nltk.download('wordnet')
   [nltk_data] Downloading package wordnet to /root/nltk_data...
    [nltk_data] Package wordnet is already up-to-date!
    True
1 import nltk
 2 from nltk.tokenize import word_tokenize
 3 from nltk.corpus import stopwords
4 from nltk.stem import PorterStemmer, WordNetLemmatizer
5 import spacy
7 nltk.download('punkt')
8 nltk.download('stopwords')
9 nltk.download('wordnet')
10
11 stop_words = set(stopwords.words('english'))
12 ps = PorterStemmer()
13 lemmatizer = WordNetLemmatizer()
14 nlp = spacy.load('en_core_web_sm')
15
16 # Function for text preprocessing
17 def preprocess(text):
18
      text = text.lower() # Lowercase
19
      tokens = word_tokenize(text) # Tokenization
      tokens = [word for word in tokens if word.isalnum()] # Remove punctuation
20
21
      tokens = [word for word in tokens if word not in stop_words] # Stopword removal
      tokens = [ps.stem(word) for word in tokens] # Stemming
      tokens = [lemmatizer.lemmatize(word) for word in tokens] # Lemmatization
23
      return " ".join(tokens) # Convert back to text
24
25
26 # Apply preprocessing to all datasets
27 df_train['processed_text'] = df_train['text'].apply(preprocess)
28 df_test['processed_text'] = df_test['text'].apply(preprocess)
29 df_valid['processed_text'] = df_valid['text'].apply(preprocess)
```

```
[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!
    OSError
                                             Traceback (most recent call last)
    <ipython-input-39-90ac0d6a0383> in <cell line: 0>()
         9 nltk.download('wordnet')
         10
    ---> 11 stop_words = set(stopwords.words('english'))
        12 ps = PorterStemmer()
         13 lemmatizer = WordNetLemmatizer()
                                     2 4 frames
    /usr/local/lib/python3.11/dist-packages/nltk/data.py in __init__(self, _path)
                  _path = os.path.abspath(_path)
if not os.path.exists(_path):
        309
        310
                      raise OSError("No such file or directory: %r" % _path)
    --> 311
       312
                   self._path = _path
        313
    OSError: No such file or directory: '/root/nltk_data/corpora/stopwords/english'
```

Vectorization

```
1 from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer
2
3 # Use TF-IDF
4 tfidf = TfidfVectorizer()
5 X_train_tfidf = tfidf.fit_transform(df_train['processed_text'])
6 X_test_tfidf = tfidf.transform(df_test['processed_text'])
7 X_valid_tfidf = tfidf.transform(df_valid['processed_text'])
8
9 # Use CountVectorizer
10 cv = CountVectorizer
10 tx_train_cv = cv.fit_transform(df_train['processed_text'])
12 X_test_cv = cv.transform(df_test['processed_text'])
13 X_valid_cv = cv.transform(df_valid['processed_text'])
14
15 # Labels (assuming the label column is named 'label')
16 y_train = df_train['label']
17 y_test = df_test['label']
18 y_valid = df_valid['label']
19
```

Splitting the Data:

Divide the dataset into training and testing sets (e.g., 80% training, 20% testing).

Building the Classification Model:

Train a text classification model using Logistic Regression, Naïve Bayes, or any other suitable algorithm.

Implement the model using scikit-learn.

Model Evaluation:

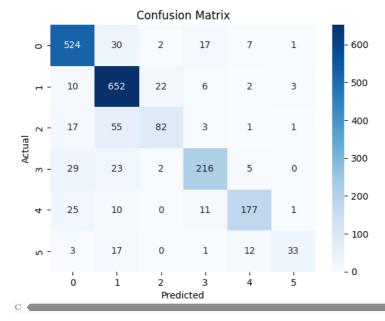
Evaluate the model using accuracy, precision, recall, and F1-score.

Use a confusion matrix to visualize the results.

```
1 from sklearn.linear_model import LogisticRegression
2
3 # Train classifier
4 clf = LogisticRegression()
5 clf.fit(X_train_tfidf, y_train)
6
7 # Predictions
8 y_pred = clf.predict(X_test_tfidf)
9

1 from sklearn.metrics import accuracy_score, classification_report, confusion_matrix
2 import seaborn as sns
3 import matplotlib.pyplot as plt
4
5 # Accuracy
6 print("Accuracy:", accuracy_score(y_test, y_pred))
7
```

```
8 # Classification report
9 print("Classification Report:\n", classification_report(y_test, y_pred))
10
11 # Confusion matrix
12 cm = confusion_matrix(y_test, y_pred)
13 sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
14 plt.xlabel("Predicted")
15 plt.ylabel("Actual")
16 plt.title("Confusion Matrix")
17 plt.show()
   Accuracy: 0.842
    Classification Report:
                                 recall f1-score
                    precision
                                                      support
                0
                        0.86
                                   0.90
                                             0.88
                                                         581
                1
                        0.83
                                   0.94
                                             0.88
                                                         695
                2
                        0.76
                                   0.52
                                                         159
                                             0.61
                3
                        0.85
                                   0.79
                                             0.82
                                                         275
                                   0.79
                4
                        0.87
                                             0.83
                                                         224
                5
                        0.85
                                   0.50
                                             0.63
                                                          66
        accuracy
                                             0.84
                                                        2000
       macro avg
                        0.84
                                   9.74
                                             0.77
                                                        2000
    weighted avg
                        0.84
                                   0.84
                                             0.84
                                                        2000
```



Submission Guidelines:

Google Colab Notebook Submission:

Save your notebook as NLP_Text_Classification_YourName.ipynb.

Ensure all code cells are executed, and the output is visible.

Include proper documentation and comments explaining each step.

Report Submission (Optional):

Prepare a short report (2-3 pages) summarizing your approach, findings, and model performance.

Upload the report along with the Colab Notebook.

Grading Criteria:

Correct implementation of NLP preprocessing (30%)

Effective use of vectorization techniques (20%)

Model accuracy and performance evaluation (30%)

Code clarity, documentation, and presentation (20%)

1 Start coding or ge_nerate with AI.

Declaration

I, Parth Kulkarni, confirm that the work submitted in this assignment is my own and has been completed following academic integrity guidelines. The code is uploaded on my GitHub repository account, and the repository link is provided below:

GitHub Repository Link: https://github.com/parthdk16/Deep_Learning

Signature: Parth Kulkarni

Submission Checklist

- ✓ Ultralitycs Platform Documentsation Like hel file for Given Task
- √ Code file (Python Notebook or Script)
- ✓ Dataset or link to the dataset
- √ Visualizations (if applicable)
- √ Screenshots of model performance metrics
- √ Readme File