Bahceci

```
# Importing the necessary packages
#!pip install demoji
#!pip install transformers
import ast
import re
import demoji
import matplotlib.pyplot as plt
import nltk
import pandas as pd
import spacy
import textblob
from nltk import word_tokenize
from nltk.corpus import stopwords
from \ nltk.sentiment.vader \ import \ SentimentIntensityAnalyzer
from nltk.stem import PorterStemmer, WordNetLemmatizer
from sklearn import metrics
from sklearn.feature extraction.text import CountVectorizer
from transformers import pipeline
nltk.download('vader_lexicon')
nltk.download('wordnet')
```

1. Text Preprocessing

```
Code for data cleaning.
nlp = spacy.load("en_core_web_sm")
nltk.download('stopwords')
stop_words = set(stopwords.words('english'))
nltk.download('punkt')
PATH = 'twitter_training.csv'
twitter_df = pd.read_csv(PATH)
# Run summary/descriptive statistics tests on the data
# Rename the dataframe to pre_df
pre_df = twitter_df.copy()
print("Head of the dataframe:")
print(pre_df.head())
print("\nDescriptive statistics of the dataframe:")
print(pre_df.describe())
pre_df.columns = ["ID", "Game", "Sentiment", "Text"]
# Determine if there are missing values
missing_values = pre_df.isnull().sum()
print("\nMissing values in the dataframe:")
print(missing_values)
# Drop rows with missing 'Text' values
pre_df.fillna("", inplace=True)
pre_df["ID"] = pd.to_numeric(pre_df["ID"], errors="coerce")
pre_df["ID"] = pre_df["ID"].fillna(0).astype(int)
print("Dropped na rows.")
# Lowercase the entire text
pre_df['Text'] = pre_df['Text'].str.lower()
# Remove /n from the text
pre_df['Text'] = pre_df['Text'].str.replace(r'\n.*', '', regex=True)
def remove_emojis(text):
```

```
Remove emojis, of course the emojis may effect the analysis
    as there are positive and negative emojis.
    return demoji.replace(text, '')
pre_df['Text'] = pre_df['Text'].apply(remove_emojis)
print("Removed emojis")
#Correcting the spelling errors can help us as we will
#have more data that can be processed and
#it can give us more meaningful results.
# Remove all non-ASCII characters and fill with whitespace
def remove_non_ascii(text):
    Function to remove non-ascii characters
    cleaned_text = ""
    for char in text:
        if ord(char) < 128:
            cleaned_text += char
            cleaned_text += " "
    return cleaned_text
pre_df["Text"] = pre_df["Text"].apply(remove_non_ascii)
pre_df["Text"] = pre_df["Text"].str.strip()
print(pre_df)
# Tokenize the text and add it as a new column
pre_df['tokenized_text'] = pre_df['Text'].apply(nltk.word_tokenize)
print(pre_df)
print("tokenization done")
# Remove stopwords in the tokenized column
def remove_stopwords_nltk(tokens):
    Remove stopwords nltk.
    filtered = [word for word in tokens if word.lower() not in stop_words]
    return filtered
def remove_stopwords_spacy(tokens):
    Remove stopwords spacy.
    stopwords_set = set(nlp.Defaults.stop_words)
    filtered = [word for word in tokens if word.lower() not in stopwords_set]
    return filtered
pre_df['tokenized_text_nltk'] = pre_df['tokenized_text'].apply(
    remove_stopwords_nltk)
pre_df['tokenized_text_spacy'] = pre_df['tokenized_text'].apply(
    remove_stopwords_spacy)
print(pre_df)
print("Removed stopwords.")
# Adding 4 columns for stem and lemmas. 2 for nltk stopword
# removed column and 2 for plain tokenized column.
lemmatizer = WordNetLemmatizer()
porter = PorterStemmer()
def lemmatize(tokens):
    """Lemmatizer"""
    return [lemmatizer.lemmatize(token) for token in tokens]
```

```
3/29/24, 9:16 AM
                                                        application_final_rbb.ipynb - Colaboratory
   get stem(tokens):
       """Stemmer"""
       return [porter.stem(token) for token in tokens]
   pre_df['lemmatized_text'] = pre_df['tokenized_text'].apply(lemmatize)
   pre_df['stemmed_text'] = pre_df['tokenized_text'].apply(stem)
   pre_df['lemmatized_text_nltk'] = pre_df['tokenized_text_nltk'].apply(lemmatize)
   pre_df['stemmed_text_nltk'] = pre_df['tokenized_text_nltk'].apply(stem)
   print(pre_df)
   print("Lemma and Stem columns added.")
   def remove_numbers(tokens):
       Removing numbers.
       return [re.sub(r'\d+', '', word) for word in tokens]
   pre_df['lemmatized_text_nltk'] = pre_df['lemmatized_text_nltk'].apply(
        remove_numbers)
   print(pre_df)
   print("removed numbers")
   #Find the non-alphabetic words.
   pattern = re.compile(r'[^a-zA-Z]+')
   def find_non_alphabetic_words(lst):
       """Function to find non alphabetic words."""
       non_alphabetic_words = []
       for word in lst:
           if pattern.match(word):
               non_alphabetic_words.append(word)
        return non_alphabetic_words
   pre_df['non_alphabetic_words'] = pre_df['lemmatized_text_nltk'].apply(
        find_non_alphabetic_words)
   pre_df['lemmatized_text_nltk'] = pre_df['lemmatized_text_nltk'].apply(" ".join)
   print(pre_df)
   print("Found the non-alphabetic words.")
   #We see that we only have punctuation at the end.
   #After removing them we can only have meaningful part of the text.
   #pre_df.to_csv("twitter_output.csv", index=False)
   print("the end")
   class DataCleaner:
       def init (self, file path):
           if isinstance(file_path, str):
               self.file_path = file_path
            else:
                raise ValueError("Input needs to be a string.")
       def read_data(self):
```

self.data = pd.read_csv(self.file_path)

return self.data

cleaned_text = "" for char in text:

return cleaned_text

"""Cleaning the data"""

def clean_data(self):

def remove_non_ascii(self, text): """Removing non-ascii chars."""

if ord(char) < 128:

cleaned_text += char

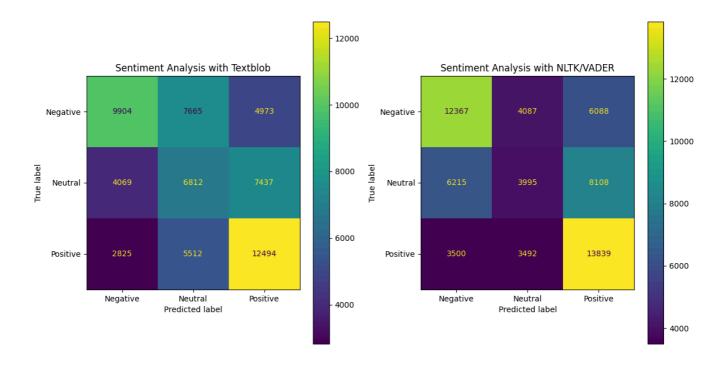
cleaned_text += " "

```
# Rename the dataframe to pre_df
    pre_df = self.data.copy()
    pre_df.columns = ["ID", "Game", "Sentiment", "Text"]
    # Determine if there are missing values
    missing_values = pre_df.isnull().sum()
    # Drop rows with missing 'Text' values
    pre_df.fillna("", inplace=True)
    pre_df["ID"] = pd.to_numeric(pre_df["ID"], errors="coerce")
    pre_df["ID"] = pre_df["ID"].fillna(0).astype(int)
    # Lowercase the entire text
    pre_df["Text"] = pre_df["Text"].str.lower()
    pre_df["Text"] = pre_df["Text"].str.strip()
    pre_df["Words"] = pre_df["Text"].apply(lambda x: word_tokenize(x))
    pre_df["Text"] = pre_df["Text"].apply(self.remove_non_ascii)
    pre_df["Text"] = pre_df["Text"].str.strip()
    self.data = pre_df
    return self.data
def features(self):
    """Getting the features"""
    vectorizer = CountVectorizer()
    X = vectorizer.fit_transform(self.data['Text'])
    self.data['Text_length'] = self.data['Text'].apply(len)
    print(self.data)
def plot_text_length(self):
    """Plotting text length."""
    plt.hist(self.data['Text_length'], bins=20)
    plt.xlabel('Text Length')
    plt.ylabel('Frequency')
    plt.title('Distribution of Text Lengths')
    plt.show()
def tokenizer(self):
    """Tokenization"""
    self.data['tokenized_text'] = self.data['Text'].apply(
       nltk.word_tokenize)
    self.data['tokenized_text'] = self.data['tokenized_text'].apply(
        ' '.join)
    print(self.data)
def lemmatize(self):
    """Lemmatizer"""
    self.data['lemmatized_text'] = self.data['tokenized_text'].apply(
        lemmatizer.lemmatize)
    self.data['lemmatized_text'] = self.data['lemmatized_text'].apply(
        ' '.join)
def stem(self):
    """Stemmer"""
    self.data['stemmed_text'] = self.data['lemmatized_text'].apply(
        porter.stem)
    self.data['stemmed_text'] = self.data['stemmed_text'].apply(' '.join)
```

2. Use NLTK/VADER, Textblob and Huggingface for Sentiment Analysis

```
def score_to_sentiment(score):
    Convert scores to sentiments.
    if score < -0.05:
        return "Negative"
    elif score > 0.05:
       return "Positive"
    else:
        return "Neutral"
sia = SentimentIntensityAnalyzer()
def sentiment_nltk(text):
    Apply sentiment analysis using nltk.
    sentiment = sia.polarity_scores(text)
    return sentiment["compound"]
pre df["sentiment nltk"] = pre df["lemmatized text nltk"].apply(sentiment nltk)
pre_df["sentiment_nltk_word"] = pre_df["sentiment_nltk"].apply(
    score to sentiment)
def sentiment_textblob(text):
    Apply sentiment analysis using TextBlob.
    blob = textblob.TextBlob(text) # Create a TextBlob object
    sentiment_score = blob.sentiment.polarity # Get the polarity score
    return sentiment_score
pre_df["sentiment_textblob"] = pre_df["lemmatized_text_nltk"].apply(
    sentiment_textblob)
pre_df["sentiment_textblob_word"] = pre_df["sentiment_textblob"].apply(
    score_to_sentiment)
# This takes more than 30 min to run, I ran it on my laptop.
huggingface_analyzer = pipeline("sentiment-analysis")
def sentiment_huggingface(text):
    Apply sentiment analysis using Huggingface.
    sentiment_score = huggingface_analyzer(text) # Get the polarity score
    return sentiment_score[0]["label"]
pre_df["sentiment_huggingf"] = pre_df["lemmatized_text_nltk"].swifter.apply(
    sentiment_huggingface)
pre_df["sentiment_huggingf_word"] = pre_df["sentiment_huggingf"].str.title()
    No model was supplied, defaulted to distilbert/distilbert-base-uncased-finetuned-sst-2-english and revision af0f9
    Using a pipeline without specifying a model name and revision in production is not recommended.
print(pre_df)
```

```
#pre_df.to_csv((PATH / "combined_version.csv"), index=False)
# Drop "Irrelevant" labels for comparison
new_df = pre_df[pre_df["Sentiment"] != "Irrelevant"]
# Visualize the two prediction models with a confusion matrix
confusion_matrix1 = metrics.confusion_matrix(new_df["Sentiment"],
                                             new_df["sentiment_textblob_word"])
confusion_matrix2 = metrics.confusion_matrix(new_df["Sentiment"],
                                             new_df["sentiment_nltk_word"])
labels = ["Negative", "Neutral", "Positive"]
cm_display1 = metrics.ConfusionMatrixDisplay(confusion_matrix=confusion_matrix1,
                                             display_labels=labels)
cm display2 = metrics.ConfusionMatrixDisplay(confusion matrix=confusion matrix2,
                                             display_labels=labels)
fig, axes = plt.subplots(1, 2, figsize=(12, 6))
cm_display1.plot(ax=axes[0])
axes[0].set_title("Sentiment Analysis with Textblob")
cm_display2.plot(ax=axes[1])
axes[1].set_title("Sentiment Analysis with NLTK/VADER")
plt.tight_layout()
plt.show()
```



```
*********** Module application_final_rbb
application_final_rbb.py:403:0: C0304: Final newline missing (missing-final-newline)
application_final_rbb.py:41:0: W0105: String statement has no effect (pointless-string-statement)
application_final_rbb.py:216:0: C0115: Missing class docstring (missing-class-docstring)
application_final_rbb.py:223:4: C0116: Missing function or method docstring (missing-function-docstring)
application_final_rbb.py:242:8: W0621: Redefining name 'pre_df' from outer scope (line 56) (redefined-outer-name)
application_final_rbb.py:247:8: W0621: Redefining name 'missing_values' from outer scope (line 65) (redefined-out
application_final_rbb.py:258:47: W0108: Lambda may not be necessary (unnecessary-lambda)
application_final_rbb.py:247:8: W0612: Unused variable 'missing_values' (unused-variable)
application_final_rbb.py:271:8: C0103: Variable name "X" doesn't conform to snake_case naming style (invalid-name
application_final_rbb.py:271:8: W0612: Unused variable 'X' (unused-variable)
application_final_rbb.py:224:8: W0201: Attribute 'data' defined outside __init__ (attribute-defined-outside-init)
application_final_rbb.py:305:0: W0105: String statement has no effect (pointless-string-statement)
application_final_rbb.py:311:4: R1705: Unnecessary "elif" after "return", remove the leading "el" from "elif" (no
application_final_rbb.py:396:0: E1123: Unexpected keyword argument 'title' in method call (unexpected-keyword-arg
```

```
import pandas as pd
In []:
        import re
        import nltk
        import matplotlib.pyplot as plt
        import spacy
        import seaborn as sns
        import numpy as np
        from nltk.corpus import stopwords
        from spacy.lang.en import STOP_WORDS as stopwords_spacy
        from sklearn.feature_extraction.text import ENGLISH_STOP_WORDS as stopwords
        from nltk.tokenize import word_tokenize
        from nltk.stem import PorterStemmer, WordNetLemmatizer
        from nltk.sentiment.vader import SentimentIntensityAnalyzer
        from textblob import TextBlob
        from sklearn.metrics import confusion matrix
        from transformers import pipeline
        from textblob import TextBlob
```

1. Preprocess Twitter Dataset

- · Preliminary Data Cleaning
- Advanced Data Cleaning

Read in the twitter_training.csv

```
In []: # Read in the twitter_training.csv
TWITTER_FILE_PATH = "twitter_training.csv"
twitter_data = pd.read_csv(TWITTER_FILE_PATH)
```

Run summary/descriptive statistics tests on the data (e.g. head, describe, etc)

```
In [ ]: twitter_data_length = len(twitter_data)

print(f"Twitter Data Length: {twitter_data_length}\n")
print(twitter_data.head())
print("")
print(twitter_data.describe())
```

```
Twitter Data Length: 74681
   2401 Borderlands Positive \
  2401 Borderlands Positive
1 2401 Borderlands Positive
2 2401 Borderlands Positive
  2401 Borderlands Positive
4 2401 Borderlands Positive
  im getting on borderlands and i will murder you all ,
0 I am coming to the borders and I will kill you...
1 im getting on borderlands and i will kill you ...
2 im coming on borderlands and i will murder you...
3 im getting on borderlands 2 and i will murder ...
4 im getting into borderlands and i can murder y...
               2401
count 74681.000000
        6432,640149
mean
        3740,423819
std
min
           1.000000
25%
        3195.000000
        6422.000000
50%
75%
        9601,000000
max
       13200.000000
```

Rename this dataframe to pre_df

```
In []: pre_df = twitter_data.copy()

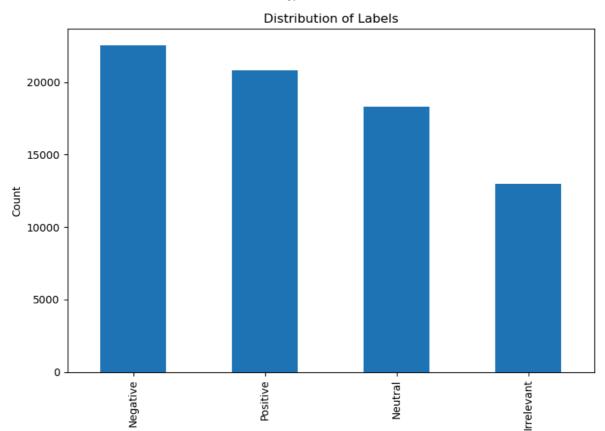
In []: # Add a row on the top of the dataframe to represent column names of each column_names = ["ID", "Game", "Label", "Text"]
    pre_df.columns = column_names

In []: pre_df["Label"].unique()
Out[]: array(['Positive', 'Neutral', 'Negative', 'Irrelevant'], dtype=object)
```

Data Visualization

```
In []: label_counts = pre_df["Label"].value_counts()

plt.figure(figsize=(8, 6))
label_counts.plot(kind="bar")
plt.title("Distribution of Labels")
plt.xlabel("Labels")
plt.ylabel("Count")
# plt.xticks(rotation=45)
plt.tight_layout()
```



Labels

1.1 Preliminary Data Cleaning

Complete several text transoformations:

- Determine if there are missing values
- Determine what to fill these missing values with

```
In []: # Check for missing values in the "Text" column
missing_text_rows = pre_df[pre_df["Text"] == " "]

print(f"Missing Text Rows:\n {missing_text_rows}")

# Check for NaN in the "Text" column
nan_rows = pre_df[pre_df["Text"].isna()]

print(f"NaN Rows:\n {nan_rows}")
```

```
Missing Text Rows:
                  ID
                                          Game
                                                     Label Text
        2291
               1602 CallOfDutyBlackopsColdWar
                                               Irrelevant
        2993
              1719 CallOfDutyBlackopsColdWar
                                                 Positive
               1763 CallOfDutyBlackopsColdWar
        3239
                                                  Neutral
        3935
               1880 CallOfDutyBlackopsColdWar
                                                 Negative
        4229
               1929 CallOfDutyBlackopsColdWar
                                                 Negative
        . . .
               . . .
                                                      . . .
        73229
              8945
                                       Nvidia
                                                 Positive
        73517 8993
                                       Nvidia
                                                  Neutral
        73757
              9036
                                       Nvidia
                                                 Negative
        73967
              9073
                                       Nvidia
                                                 Positive
        74417 9154
                                       Nvidia
                                                 Positive
        [172 rows x 4 columns]
        NaN Rows:
                  ID
                            Game
                                     Label Text
        60
               2411 Borderlands
                                  Neutral NaN
                                  Neutral
        552
               2496
                    Borderlands
                                           NaN
        588
               2503 Borderlands
                                 Neutral
                                           NaN
        744
               2532 Borderlands Positive NaN
        1104
               2595 Borderlands Positive
                                           NaN
        73971 9073
                         Nvidia Positive
                                           NaN
        73972
              9073
                         Nvidia Positive
                                           NaN
        74420 9154
                         Nvidia Positive
                                           NaN
                         Nvidia Positive
        74421 9154
                                           NaN
        74422 9154
                         Nvidia Positive NaN
        [686 rows x 4 columns]
In []: # Fill missing values with "Unknown" in the "Text" column
        pre_df.loc[missing_text_rows.index, "Text"] = "Unknown"
        pre_df.loc[nan_rows.index, "Text"] = "Unknown"
```

Change/verify relevant column data types

```
In []:
        # Check the data types of the columns
        column_datatypes = pre_df.dtypes
        print(column_datatypes)
        ID
                   int64
        Game
                  object
        Label
                  object
        Text
                  object
        dtype: object
```

Lowercase

```
# Lower case
pre_df["Lower_Text"] = pre_df["Text"].str.lower()
```

Remove non-ASCII characters and fill with whitespace

```
# Remove all non-ASCII characters and fill with whitespace
pre_df["Remove_non_Ascii"] = pre_df["Lower_Text"].apply(lambda x: re.sub(r'
```

Remove additional whitespace (stripping)

```
In []: # Remove additional whitespace (stripping)
    pre_df["Remove_Whitespace"] = pre_df["Remove_non_Ascii"].str.strip()
```

Split the string info into a list of strings that are each one word (tokenization)

```
In [ ]: pre_df["Tokenized_Text"] = pre_df["Remove_Whitespace"].apply(word_tokenize)
```

1.2 Advanced Data Cleaning

Remove "\n" and other symbols followed by letters

```
In []: # Remove "\n" and other symbols followed by letters
pre_df["Remove_n"] = pre_df["Remove_Whitespace"].apply(lambda x: re.sub(r"\r")
```

Remove emojis and emoticons

Handling emojis and emoticons:

During sentiment analysis, the model only considers textual content. Elements like emojis and emoticons aren't important for the task, thus can be omitted.

However, if we aim to retain the impact of emojis and emoticons, there's a possibility to convert them into numerical values.

```
In [ ]: pre_df["Remove_Emoji"] = pre_df["Remove_n"].apply(lambda x: x.encode("ascii")
```

Fix Spelling Error

Handling spelling errors:

Spelling mistakes can introduce noise in sentiment analysis tasks. One approach is to simply ignore the errors. Alternatively, we can employ libraries such as PySpellChecker or TextBlob to solve spelling issues.

```
In []:
    def correct_spelling(text):
        """Correct the spelling of the given text using TextBlob."""
        blob = TextBlob(text)
        return str(blob.correct())

    pre_df["Fixed_Spelling"] = pre_df["Remove_Emoji"].apply(correct_spelling)
    pre_df
```

The execution took too long

Remove stopwords

```
In []: nltk.download("stopwords")
        def remove_stopwords_nltk(text):
            """Remove stopwords from the given text using NLTK."""
            stopwords_nltk = set(stopwords.words("english"))
            tokens = word_tokenize(text)
             return " ".join([token for token in tokens if token not in stopwords_nlt
        def remove stopwords spacy(text):
            """Remove stopwords from the given text using SpaCy."""
            tokens = word tokenize(text)
            return " ".join([token for token in tokens if token not in stopwords_spa
        def remove_stopwords_textblob(text):
            """Remove stopwords from the given text using TextBlob."""
            tokens = word tokenize(text)
             stopwords_textblob = set(stopwords.words("english"))
             return " ".join([token for token in tokens if token not in stopwords_tex
        def remove_stopwords_sklearn(text):
            """Remove stopwords from the given text using Scikit-learn."""
            tokens = word tokenize(text)
             return " ".join([token for token in tokens if token not in stopwords_sk]
        [nltk_data] Downloading package stopwords to /Users/yu/nltk_data...
                    Package stopwords is already up-to-date!
        [nltk data]
        pre_df["Remove_Stopwords_NLTK"] = pre_df["Remove_Emoji"].apply(remove_stopwords_NLTK"]
        pre_df["Remove_Stopwords_SpaCy"] = pre_df["Remove_Emoji"].apply(remove_stop)
        pre_df["Remove_Stopwords_Textblob"] = pre_df["Remove_Emoji"].apply(remove_st
        pre df["Remove Stopwords Sklearn"] = pre df["Remove Emoji"].apply(remove sto
```

Stemming and Lemmatization (on NLTK)

In my opinion, stemming is preferable to lemmatization. Instead of reducing words to their base form, stemming only removes suffixes from the root, which can preserve more information and also may be faster in computation.

```
In []:
        stemmer = PorterStemmer()
         lemmatizer = WordNetLemmatizer()
        # Stemming and lemmatization on ntlk
        def stemming_nltk(text):
             """Perform stemming using NLTK."""
             tokens = word tokenize(text)
             stemmed_tokens = [stemmer.stem(token) for token in tokens]
             return " ".join(stemmed_tokens)
        def lemmatization_nltk(text):
             """Perform lemmatization using NLTK."""
             tokens = word tokenize(text)
             lemmatized_tokens = [lemmatizer.lemmatize(token) for token in tokens]
             return " ".join(lemmatized_tokens)
        pre_df["Stemmed_Text_NLTK"] = pre_df["Remove_Stopwords_NLTK"].apply(stemming)
        pre_df["Lemmatized_Text_NLTK"] = pre_df["Remove_Stopwords_NLTK"].apply(lemmatized_Text_NLTK"]
```

Remove numbers

Whether or not to remove numbers depends on their role in the dataset. In Twitter datasets, numbers often represent user IDs, timestamps, email addresses, and more. However, such numerical data isn't important to sentiment analysis. Therefore, removing numbers is a viable option.

```
In [ ]: pre_df["Remove_Numbers"] = pre_df["Stemmed_Text_NLTK"].apply(lambda x: re.st
```

Remove non alphabetic words

```
In []: def remove_non_alphabetic(text):
    """Remove non_alphabetic characters from the given text."""
    return " ".join(re.findall(r"\b[a-zA-Z]+\b", text))

pre_df["Remove_non_Alphabetic"] = pre_df["Remove_Numbers"].apply(remove_non_
```

Create an output of this dataset as a csv file

```
In [ ]: OUTPUT_FILE_PATH = "/Users/yu/Desktop/uds_ws2324/tools_for_nlp/assignment/f:
    pre_df.to_csv(OUTPUT_FILE_PATH, index=False)
```

2. Create a functioning class that packages all the items

```
class TwitterDataPreprocessor:
In [ ]:
            """A class for preprocessing Twitter data"""
                  _init__(self, file_path):
                """Initialize the TwitterDataPreprocessor object."""
                self.file_path = file_path
                self.data = None
            def load_data(self):
                """Load data."""
                self.data = pd.read_csv(self.file_path)
            def preprocess(self):
                """Preprocess the loaded data."""
                pre_df = self.data.copy()
                column_names = ["ID", "Game", "Label", "Text"]
                pre_df.columns = column_names
                # Check for missing values in the "Text" column
                pre_df["Text"].fillna("Unknown", inplace=True)
                # Convert text to lowercase
                pre_df["Lower_Text"] = pre_df["Text"].str.lower()
                # Remove non-ASCII characters
                pre_df["Remove_non_Ascii"] = pre_df["Lower_Text"].apply(lambda x: re
```

```
# Remove whitespace
        pre_df["Remove_Whitespace"] = pre_df["Remove_non_Ascii"].str.strip(
        # Remove newlines
        pre_df["Remove_n"] = pre_df["Remove_Whitespace"].apply(lambda x: re
        # Remove emoiis
        pre df["Remove Emoji"] = pre df["Remove n"].apply(lambda x: x.encode
        def remove stopwords nltk(text):
            """Remove stopwords using NLTK."""
            stopwords_nltk = set(stopwords.words("english"))
            tokens = word tokenize(text)
            return " ".join([token for token in tokens if token not in stop)
        pre df["Remove Stopwords NLTK"] = pre df["Remove Emoji"].apply(remove
        # Stemming and lemmatization on nltk
        stemmer = PorterStemmer()
        lemmatizer = WordNetLemmatizer()
        def stemming nltk(text):
            """Perform stemming using NLTK."""
            tokens = word tokenize(text)
            stemmed_tokens = [stemmer.stem(token) for token in tokens]
            return " ".join(stemmed_tokens)
        def lemmatization_nltk(text):
            """Perform lemmatization using NLTK."""
            tokens = word_tokenize(text)
            lemmatized_tokens = [lemmatizer.lemmatize(token) for token in token
            return " ".join(lemmatized tokens)
        def remove non alphabetic(text):
            """Remove non-alphabetic characters."""
            return " ".join(re.findall(r"\b[a-zA-Z]+\b", text))
        pre_df["Stemmed_Text_NLTK"] = pre_df["Remove_Stopwords_NLTK"].apply
        pre_df["Lemmatized_Text_NLTK"] = pre_df["Remove_Stopwords_NLTK"].apr
        # Remove numbers
        pre_df["Remove_Numbers"] = pre_df["Stemmed_Text_NLTK"].apply(lambda
        # Remove non-alphabetic characters
        pre_df["Remove_non_Alphabetic"] = pre_df["Remove_Numbers"].apply(rer
        return pre_df
    def save_preprocessed_data(self, output_file_path):
        """Save preprocessed data to a CSV file."""
        preprocessed_data = self.preprocess()
        preprocessed_data.to_csv(output_file_path, index=False)
        print("Preprocessed data saved.")
TWITTER_FILE_PATH = "twitter_training.csv"
OUTPUT_FILE_PATH = "twitter_data_preprocessed.csv"
preprocessor = TwitterDataPreprocessor(TWITTER_FILE PATH)
preprocessor.load_data()
preprocessor.save_preprocessed_data(OUTPUT_FILE_PATH)
Preprocessed data saved.
```

4. Use NLTK/VADER, SpaCy/Textblob, and HuggingFace Sentiment Analysis

- Make a function for NLTK/VADER to pass through your preprocessed dataset
- Make a function to do the same with SpaCy/Textblob

NLTK/VADER sentiment analysis

```
In []: # Initialize VADER
    nltk.download("vader_lexicon")
    sid = SentimentIntensityAnalyzer()

def nltk_vader_analysis(text):
        """
        Get sentiment scores using VADER.
        """
        scores = sid.polarity_scores(text)
        return scores

pre_df["Sentiment_Scores_NLTK_VADER"] = pre_df["Remove_non_Alphabetic"].app

[nltk_data] Downloading package vader_lexicon to
    [nltk_data] /Users/yu/nltk_data...
    [nltk_data] Package vader_lexicon is already up-to-date!
```

SpaCy/Textblob sentiment analysis

```
In [ ]: def spacy_textblob_analysis(data):
            Sentiment Analysis: Using SpaCy for tokenization and TextBlob for senting
            nlp = spacy.load("en_core_web_sm")
            def tokenize(text):
                Tokenize text using SpaCy.
                doc = nlp(text)
                tokens = [token.text for token in doc]
                return tokens
            def get_sentiment_scores(text):
                Get sentiment scores using TextBlob.
                blob = TextBlob(text)
                polarity = blob.sentiment.polarity
                subjectivity = blob.sentiment.subjectivity
                return {"polarity": polarity, "subjectivity": subjectivity}
            data["Sentiment_Scores_SpaCy_TextBlob"] = data["Remove_non_Alphabetic"]
            return data
        pre_df = spacy_textblob_analysis(pre_df)
```

HuggingFace sentiment analysis

```
In []: sentiment_analysis_pipeline = pipeline("sentiment-analysis")

def hugging_face_sentiment_analysis(text):
    """
    Get sentiment scores using HugginFace
    """
    results = sentiment_analysis_pipeline(text)
    # print(results[0])

# Extract sentiment label and score
    sentiment_label = results[0]["label"]
    sentiment_score = results[0]["score"]

    return {"label": sentiment_label, "score": sentiment_score}

pre_df["Sentiment_Scores_Hugging_Face"] = pre_df["Remove_non_Alphabetic"].ap

No model was supplied, defaulted to distilbert-base-uncased-finetuned-sst-2
    -english and revision af0f99b (https://huggingface.co/distilbert-base-uncased-finetuned-sst-2-english).
Using a pipeline without specifying a model name and revision in production is not recommended.
```

5. Run and show the results via visualizations (e.g. Confusion matrix / bar plots) comparing the results

Confusion Matrix

```
nltk_scores = pre_df["Sentiment_Scores_NLTK_VADER"]
In [ ]:
        spacy_scores = pre_df["Sentiment_Scores_SpaCy_TextBlob"]
        actual_sentiments = []
        nltk_sentiments = []
        spacy_sentiments = []
        hugging_face_sentiments = []
In [ ]: for nltk_score, spacy_score in zip(nltk_scores, spacy_scores):
            nltk_sentiment = "Negative" if -1 <= nltk_score["compound"] < -0.05 else</pre>
            spacy_sentiment = "Negative" if spacy_score["polarity"] < 0 else "Neutra</pre>
            pre_df["Sentiment_Scores_NLTK_VADER"] = nltk_sentiment
            pre_df["Sentiment_Scores_SpaCy_TextBlob"] = spacy_sentiment
            nltk_sentiments.append(nltk_sentiment)
             spacy_sentiments.append(spacy_sentiment)
In [ ]: for index, row in pre_df.iterrows():
            hugging_face_sentiment = row["Sentiment_Scores_Hugging_Face"]["label"]
            hugging_face_sentiments.append((hugging_face_sentiment.capitalize()))
        for index, row in pre_df.iterrows():
In []:
            actual_sentiment = row["Label"]
            actual_sentiments.append(actual_sentiment)
```

```
labels = ["Negative", "Neutral", "Positive"]
In [ ]:
        nltk_confusion_matrix = confusion_matrix(actual_sentiments, nltk_sentiments)
        spacy_confusion_matrix = confusion_matrix(actual_sentiments, spacy_sentiment
        hugging_face_confusion_matrix = confusion_matrix(actual_sentiments, hugging_
        print(f"NLTK Confusion Matrix: {nltk_confusion_matrix}\n")
        print(f"SpaCy Confusion Matrix: {spacy confusion matrix}\n")
        print(f"Hugging Face Confusion Matrix: {hugging_face_confusion_matrix}\n")
        NLTK Confusion Matrix: [[11939 5071 5532]
         [ 5849 5062 7407]
         [ 3858 5062 11911]]
        SpaCy Confusion Matrix: [[ 9738 7875 4929]
         [ 4516 7175 6627]
         [ 3854 6846 10131]]
        Hugging Face Confusion Matrix: [[19809
                                                 0 2733]
         [13739
                    0 45791
         [10532
                    0 1029911
```

Visualization

```
In []: plt.figure(figsize=(20, 9))
    plt.subplot(1, 3, 1)
    sns.heatmap(nltk_confusion_matrix, annot=True, cmap="Greens", fmt="d", xticl
    plt.title("NLTK Confusion Matrix")

    plt.subplot(1, 3, 2)
    sns.heatmap(spacy_confusion_matrix, annot=True, cmap="Greens", fmt="d", xticl
    plt.title("SpaCy Confusion Matrix")

    plt.subplot(1, 3, 3)
    sns.heatmap(hugging_face_confusion_matrix, annot=True, cmap="Greens", fmt="c
    plt.title("Hugging Face Confusion Matrix")

    plt.tight_layout()
    plt.show()

    nltk_accuracy = np.trace(nltk_confusion_matrix) / np.sum(nltk_confusion_matrix) spacy_accuracy = np.trace(spacy_confusion_matrix) / np.sum(spacy_confusion_r
    hugging_face_accuracy = np.trace(hugging_face_confusion_matrix) / np.sum(hug
```

