### Week 22 Challenge - NLP

Vader Sentiment Analyzer from the Natural Language Toolkit (NLTK) library to perform sentiment analysis on the Twitter data.

The Vader Sentiment Analyzer is a lexicon and rule-based sentiment analysis tool that is specifically designed for analyzing sentiment in social media text, such as tweets. It assigns a sentiment score to each piece of text, indicating the overall sentiment polarity (positive, negative, or neutral) and intensity.

Before we begin with the mongoDB task, we need to make sure we have installed mongoDB - PyMongo

```
In [4]: import pymongo

Connect the jupyter with the MongoDB server by using MongoDB details

In [5]: client = pymongo.MongoClient("mongodb://localhost:27017/")
    db = client("Trial")
    checking the list of database

In [6]: print(client.list_database_names())
    ['Trial', 'admin', 'config', 'local']

In [7]: import pandas as pd

# Load the Twitter dataset from the .json files
    tweets_df = pd.read_csv(r"C:\Users\junai\OneDrive - Middlesex University\Applied Data Analytics\week 22\twitter_cleaned\tweets.cs
    users_df = pd.read_csv(r"C:\Users\junai\OneDrive - Middlesex University\Applied Data Analytics\week 22\twitter_cleaned\tweets.csv"

# Display the first few rows of the datasets
    tweets_df.head(), users_df.head()
```

```
id
                             timestamp \
Out[7]:
             8653 2009-04-06T21:21:56Z
         1 12020 2009-04-06T21:22:47Z
         2 23858 2009-04-06T21:25:53Z
         3 51844 2009-04-06T21:33:11Z
         4 52341 2009-04-06T21:33:20Z
                                                       text
         0 falling asleep, just heard about that tracy gi...
         1 i have a sad feeling that dallas is not going ...
         2 @statravelau just got ur newsletter, those far...
         3 @djalizay i really don't think people choose t...
         4 my mind and body are severely protesting this ... ,
               id
                           user
                                   age
            8653
                    hpfangirl94
                                   old
         1 12020
                   HybridMink young
         2 23858 driveaway2008
                                  old
         3 51844
                     lennytoups
                                   old
         4 52341
                        Jemimus young)
```

this code segment fetches data from the MongoDB collection named 'Trial', retrieves specific fields specified in the fields\_to\_retrieve dictionary, and stores the retrieved data in a list of dictionaries named data\_list, making it easier to work with the data in Python.

```
In [8]: collection_name = 'Trial'

# Access the collection
collection = db[collection_name]

# Define fields to retrieve
fields_to_retrieve = {
    'user_id': 1,
    'tweet_content': 1,
    'timestamp': 1,
    'user_age': 1,
    # Add other relevant fields here
}

# Retrieve data from the collection
data = collection.find({}, fields_to_retrieve)

# Convert retrieved data to a list of dictionaries
data_list = list(data)
```

```
In [10]: print(data_list)
         []
In [11]: import pymongo
         # Connection details
         host = 'localhost:27017'
         port = 27017
         database_name = 'ADA_trial'
         # Establish connection to MongoDB without authentication
         client = pymongo.MongoClient(host, port)
         # Access the database
         db = client[database name]
In [12]: print(client.list_database_names())
         ['ADA_trial', 'Trial', 'admin', 'config', 'local']
In [43]: import pandas as pd
         # Load the Twitter dataset from the .json files
         tweets df = pd.read csv(r"C:\Users\junai\OneDrive - Middlesex University\Applied Data Analytics\week 22\twitter cleaned\tweets.cs
         users df = pd.read csv(r"C:\Users\junai\OneDrive - Middlesex University\Applied Data Analytics\week 22\twitter cleaned\users.csv"
         # Display the first few rows of the datasets
         tweets df.head(), users df.head()
```

```
id
                               timestamp \
Out[43]:
              8653 2009-04-06T21:21:56Z
          1 12020 2009-04-06T21:22:47Z
          2 23858 2009-04-06T21:25:53Z
          3 51844 2009-04-06T21:33:11Z
          4 52341 2009-04-06T21:33:20Z
                                                         text
          0 falling asleep, just heard about that tracy gi...
          1 i have a sad feeling that dallas is not going ...
          2 @statravelau just got ur newsletter, those far...
          3 @djalizay i really don't think people choose t...
          4 my mind and body are severely protesting this ... ,
                id
                             user
                                     age
             8653
                      hpfangirl94
                                     old
          1 12020
                    HybridMink young
          2 23858 driveaway2008
                                    old
          3 51844
                       lennytoups
                                    old
          4 52341
                          Jemimus young)
In [44]: # Convert DataFrame to dictionary (each row becomes a dictionary)
         tweets_data_dict = tweets_df.to_dict(orient='records')
         # Access the MongoDB collection for tweets
         tweets collection = db['tweets collection']
         # Insert tweets data into MongoDB collection
         tweets collection.insert many(tweets data dict)
         print("Tweets data inserted into MongoDB successfully!")
         Tweets data inserted into MongoDB successfully!
In [45]: # Convert DataFrame to dictionary (each row becomes a dictionary)
         users data dict = users df.to dict(orient='records')
         # Access the MongoDB collection for users
         users collection = db['users collection']
         # Insert users data into MongoDB collection
         users collection.insert many(users data dict)
         print("Users data inserted into MongoDB successfully!")
```

```
In [47]: # Assuming you have two DataFrames: tweets_df and users_df
         # Merge the datasets based on the 'user id' column
         merged df = pd.merge(tweets df, users df, on='id', how='left')
In [48]: # Access the MongoDB collection for tweets
         tweets collection = db['tweets collection']
         # Define fields to retrieve
         tweet fields = {
             'id': 1,
             'text': 1,
             'timestamp': 1,
             # Add more fields as needed
         # Retrieve tweets data from MongoDB collection
         tweets data = tweets collection.find({}, tweet fields)
         # Convert retrieved tweets data to a list of dictionaries
         tweets_data_list = list(tweets_data)
In [49]: # Access the MongoDB collection for users
         users collection = db['users collection']
         # Define fields to retrieve
         user fields = {
             'id': 1,
             'user': 1,
             'age': 1,
             # Add more fields as needed
         # Retrieve user data from MongoDB collection
         users_data = users_collection.find({}, user_fields)
         # Convert retrieved user data to a list of dictionaries
         users_data_list = list(users_data)
In [50]: import pandas as pd
```

```
# Convert retrieved tweets data to a pandas DataFrame
tweets_df = pd.DataFrame(tweets_data_list)

# Convert retrieved users data to a pandas DataFrame
users_df = pd.DataFrame(users_data_list)
```

NLTK stands for Natural Language Toolkit. It is a leading platform for building Python programs to work with human language data. NLTK provides easy-to-use interfaces and libraries for tasks such as tokenization, stemming, tagging, parsing, and semantic reasoning.

Stopwords are common words that are often filtered out during text preprocessing because they typically do not carry significant meaning or information for analysis.

Stemming is the process of reducing words to their base or root form. The Porter stemming algorithm is a widely used stemming algorithm that applies a set of rules to reduce words to their stems. Stemming helps in reducing the dimensionality of the text data and can improve the performance of text analysis tasks such as information retrieval or classification.

```
In [51]: import nltk
         from nltk.tokenize import word tokenize
          from nltk.corpus import stopwords
         from nltk.stem import PorterStemmer
         import string
          # Download NLTK resources (run this only once)
          nltk.download('punkt')
         nltk.download('stopwords')
         # Initialize NLTK components
         stop words = set(stopwords.words('english'))
         ps = PorterStemmer()
          # Function to preprocess text
         def preprocess text(text):
             # Tokenization
             tokens = word tokenize(text)
             # Lowercasing
             tokens = [token.lower() for token in tokens]
             # Removing punctuation
             tokens = [token for token in tokens if token not in string.punctuation]
```

```
# Removing stopwords
             tokens = [token for token in tokens if token not in stop words]
             # Stemmina (optional)
             # tokens = [ps.stem(token) for token in tokens]
             # Join tokens back into text
             preprocessed text = ' '.join(tokens)
             return preprocessed text
         # Apply preprocessing to tweet content
         tweets df['clean tweet'] = tweets df['text'].apply(preprocess text)
         [nltk data] Downloading package punkt to
         [nltk data]
                         C:\Users\junai\AppData\Roaming\nltk data...
                      Package punkt is already up-to-date!
         [nltk data]
         [nltk data] Downloading package stopwords to
                       C:\Users\junai\AppData\Roaming\nltk_data...
         [nltk_data]
         [nltk data]
                      Package stopwords is already up-to-date!
In [52]: print(tweets_df.head())
                                        id
                                                       timestamp \
                                 id
         0 65f8dbecaad7e1701be77c0e 8653 2009-04-06T21:21:56Z
         1 65f8dbecaad7e1701be77c0f 12020 2009-04-06T21:22:47Z
         2 65f8dbecaad7e1701be77c10 23858 2009-04-06T21:25:53Z
         3 65f8dbecaad7e1701be77c11 51844 2009-04-06T21:33:11Z
         4 65f8dbecaad7e1701be77c12 52341 2009-04-06T21:33:20Z
                                                        text \
         0 falling asleep, just heard about that tracy gi...
         1 i have a sad feeling that dallas is not going ...
         2 @statravelau just got ur newsletter, those far...
         3 @djalizay i really don't think people choose t...
         4 my mind and body are severely protesting this ...
                                                  clean tweet
         0 falling asleep heard tracy girl 's body found ...
         1 sad feeling dallas going show got ta say thoug...
         2 statravelau got ur newsletter fares really unb...
         3 djalizay really n't think people choose way th...
         4 mind body severely protesting quot getting quo...
```

```
In [53]: # Select a sample tweet and print original and preprocessed text
         sample tweet index = 0 # Change this to the index of the tweet you want to inspect
         original tweet = tweets df.loc[sample tweet index, 'text']
         preprocessed tweet = tweets df.loc[sample tweet index, 'clean tweet']
          print("Original Tweet:")
          print(original tweet)
         print("\nPreprocessed Tweet:")
         print(preprocessed tweet)
         Original Tweet:
         falling asleep. just heard about that tracy girl's body being found. how sad my heart breaks for that family.
         Preprocessed Tweet:
         falling asleep heard tracy girl 's body found sad heart breaks family
In [54]: # Check length of preprocessed text for all tweets
         tweets df['clean tweet length'] = tweets df['clean tweet'].apply(len)
         print(tweets df['clean tweet length'].describe())
                  62434.000000
         count
                     64,468046
         mean
                     25.388203
         std
                     5.000000
         min
         25%
                     45.000000
         50%
                     65.000000
         75%
                     84.000000
                    154.000000
         max
         Name: clean_tweet_length, dtype: float64
```

this code snippet demonstrates how to use NLTK's Vader Sentiment Analyzer to analyze the sentiment of tweets and add sentiment scores to a DataFrame for further analysis or visualization.

This analyzer is pre-trained on sentiment analysis tasks and is specifically tailored for social media text like tweets. This code defines a function get\_sentiment\_score(text) that takes a piece of text as input and returns the compound sentiment score. The compound score is a normalized score ranging from -1 (extremely negative) to 1 (extremely positive), with values closer to 0 indicating neutral sentiment.

```
In [55]: from nltk.sentiment.vader import SentimentIntensityAnalyzer

# Initialize the Vader Sentiment Analyzer
analyzer = SentimentIntensityAnalyzer()
```

```
# Function to get sentiment score for a text
         def get_sentiment_score(text):
             return analyzer.polarity scores(text)['compound']
         # Apply sentiment analysis to the tweet content
          tweets df['sentiment score'] = tweets df['text'].apply(get sentiment score)
In [56]: # Function to categorize sentiment
          def categorize sentiment(score):
             if score > 0.05:
                 return 'positive'
             elif score < -0.05:</pre>
                 return 'negative'
             else:
                  return 'neutral'
         # Apply sentiment categorization
          tweets_df['sentiment_category'] = tweets_df['sentiment_score'].apply(categorize_sentiment)
In [57]: import matplotlib.pyplot as plt
          # Plot sentiment distribution
          sentiment_distribution = tweets_df['sentiment_category'].value_counts()
          plt.bar(sentiment_distribution.index, sentiment_distribution.values)
         plt.xlabel('Sentiment Category')
         plt.ylabel('Count')
         plt.title('Sentiment Distribution')
         plt.show()
```

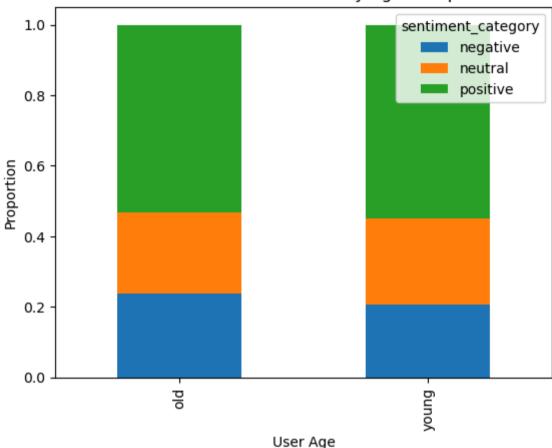
# Sentiment Distribution 35000 -30000 -25000 -20000 15000 -10000 -5000 positive neutral negative Sentiment Category

```
In [58]: # Assuming 'users_df' contains user demographics including age

# Merge 'tweets_df' with 'users_df' based on 'user_id' column
merged_df = pd.merge(tweets_df, users_df, on='id', how='left')

# Analyze sentiment distribution across different age groups
sentiment_by_age = merged_df.groupby('age')['sentiment_category'].value_counts(normalize=True).unstack()
sentiment_by_age.plot(kind='bar', stacked=True)
plt.xlabel('User Age')
plt.ylabel('Proportion')
plt.title('Sentiment Distribution by Age Group')
plt.show()
```

#### Sentiment Distribution by Age Group



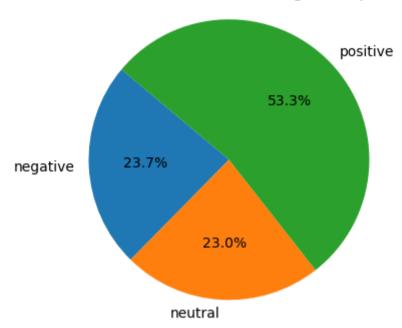
```
import matplotlib.pyplot as plt

# Analyze sentiment distribution across different age groups
sentiment_by_age = merged_df.groupby('age')['sentiment_category'].value_counts(normalize=True).unstack()

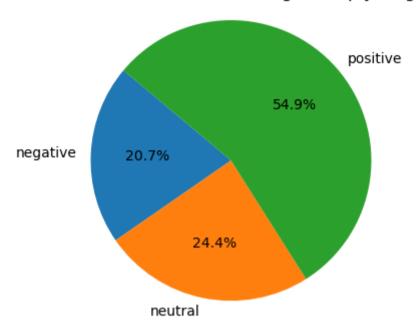
# Plot pie charts for each age group
for age_group in sentiment_by_age.index:
    sentiment_distribution = sentiment_by_age.loc[age_group]
    plt.figure(figsize=(6, 4)) # Adjust figure size as needed
    plt.pie(sentiment_distribution, labels=sentiment_distribution.index, autopct='%1.1f%%', startangle=140)
    plt.title(f'Sentiment Distribution for Age Group {age_group}')
```

plt.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle
plt.show()

## Sentiment Distribution for Age Group old



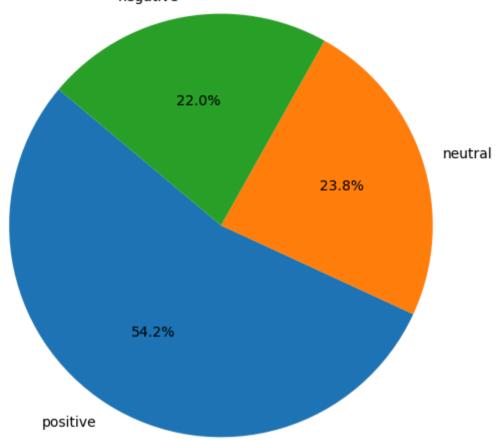
### Sentiment Distribution for Age Group young



```
import matplotlib.pyplot as plt

# Plotting sentiment distribution using a pie chart
sentiment_counts = tweets_df['sentiment_category'].value_counts()
plt.figure(figsize=(6, 6))
plt.pie(sentiment_counts, labels=sentiment_counts.index, autopct='%1.1f%', startangle=140)
plt.title('Sentiment Distribution across All Tweets')
plt.axis('equal')
plt.show()
```

## Sentiment Distribution across All Tweets negative



```
import matplotlib.pyplot as plt
import seaborn as sns

# Create subplots
fig, axs = plt.subplots(2, 2, figsize=(15, 10))

# Plot Violin Plot
sns.violinplot(x='age', y='sentiment_score', data=merged_df, ax=axs[0, 0])
axs[0, 0].set_title('Violin Plot - Sentiment Distribution by Age Group')

# Plot Bar Plot
sns.barplot(x='age', y='sentiment_score', data=merged_df, ci=None, ax=axs[0, 1])
```

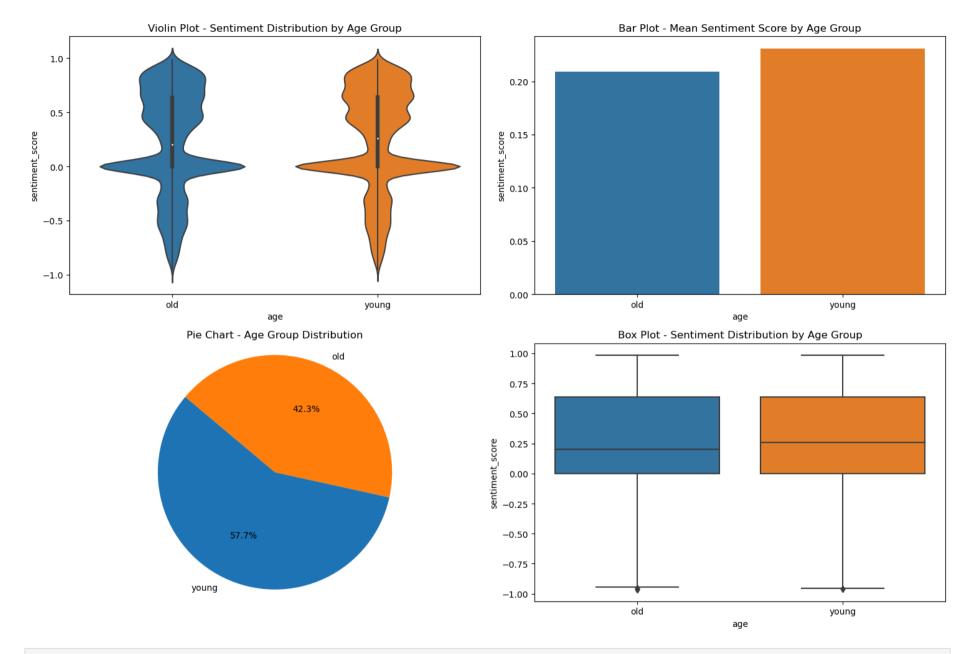
```
axs[0, 1].set_title('Bar Plot - Mean Sentiment Score by Age Group')

# Plot Pie Chart
age_counts = merged_df['age'].value_counts()
axs[1, 0].pie(age_counts, labels=age_counts.index, autopct='%1.1f%', startangle=140)
axs[1, 0].set_title('Pie Chart - Age Group Distribution')
axs[1, 0].axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.

# Plot Box Plot
sns.boxplot(x='age', y='sentiment_score', data=merged_df, ax=axs[1, 1])
axs[1, 1].set_title('Box Plot - Sentiment Distribution by Age Group')

plt.tight_layout()
plt.show()

C:\Users\junai\AppData\Local\Temp\ipykernel_11148\3937623405.py:12: FutureWarning:
The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.
sns.barplot(x='age', y='sentiment_score', data=merged_df, ci=None, ax=axs[0, 1])
```



In [68]: # Assuming 'merged\_df' is your DataFrame containing tweet data and sentiment scores
print(merged\_df['sentiment\_score'])

```
-0.5719
         0
         1
                  -0.5719
         2
                  -0.3818
         3
                  -0.3818
         4
                  -0.3150
                    . . .
                   0.6395
         124863
         124864
                   0.8625
         124865
                   0.8625
         124866
                  -0.1531
                  -0.1531
         124867
         Name: sentiment_score, Length: 124868, dtype: float64
In [69]: # View the first few rows of the DataFrame along with sentiment scores
         print(merged df[['text', 'sentiment score']].head())
                                                         text sentiment score
         0 falling asleep. just heard about that tracy gi...
                                                                       -0.5719
         1 falling asleep. just heard about that tracy gi...
                                                                       -0.5719
         2 i have a sad feeling that dallas is not going ...
                                                                       -0.3818
         3 i have a sad feeling that dallas is not going ...
                                                                       -0.3818
         4 @statravelau just got ur newsletter, those far...
                                                                       -0.3150
In [70]: import matplotlib.pyplot as plt
         # Plot histogram of sentiment scores
         plt.figure(figsize=(8, 6))
         plt.hist(merged df['sentiment score'], bins=30, color='skyblue', edgecolor='black')
         plt.xlabel('Sentiment Score')
         plt.ylabel('Frequency')
         plt.title('Distribution of Sentiment Scores')
         plt.grid(True)
         plt.show()
```

### Distribution of Sentiment Scores

