

Milestone - 03

Tech Titans-03

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Project Implementation Steps

Step 1: Environment Setup

- Installed Python, ensuring the latest version for compatibility with libraries.
- Installed PySpark and verified the environment variables such as `SPARK_HOME` and `PYSPARK_PYTHON`.
- Installed Matplotlib for visualizing the results of the analysis.

Step 2: Data Preparation

- Obtained the dataset from a reliable source.
- Preprocessed the data to ensure cleanliness, such as handling missing values and ensuring consistent data types.
- Loaded the dataset into a PySpark DataFrame for analysis.

Step 3: PySpark Cluster Setup

- Set up the PySpark environment on the local machine.
- Verified the cluster setup by running a sample PySpark job.

Step 4: PySpark Implementation

- Created a project folder structure for modular implementation.
- Developed PySpark jobs for each analysis goal using DataFrame transformations and actions.

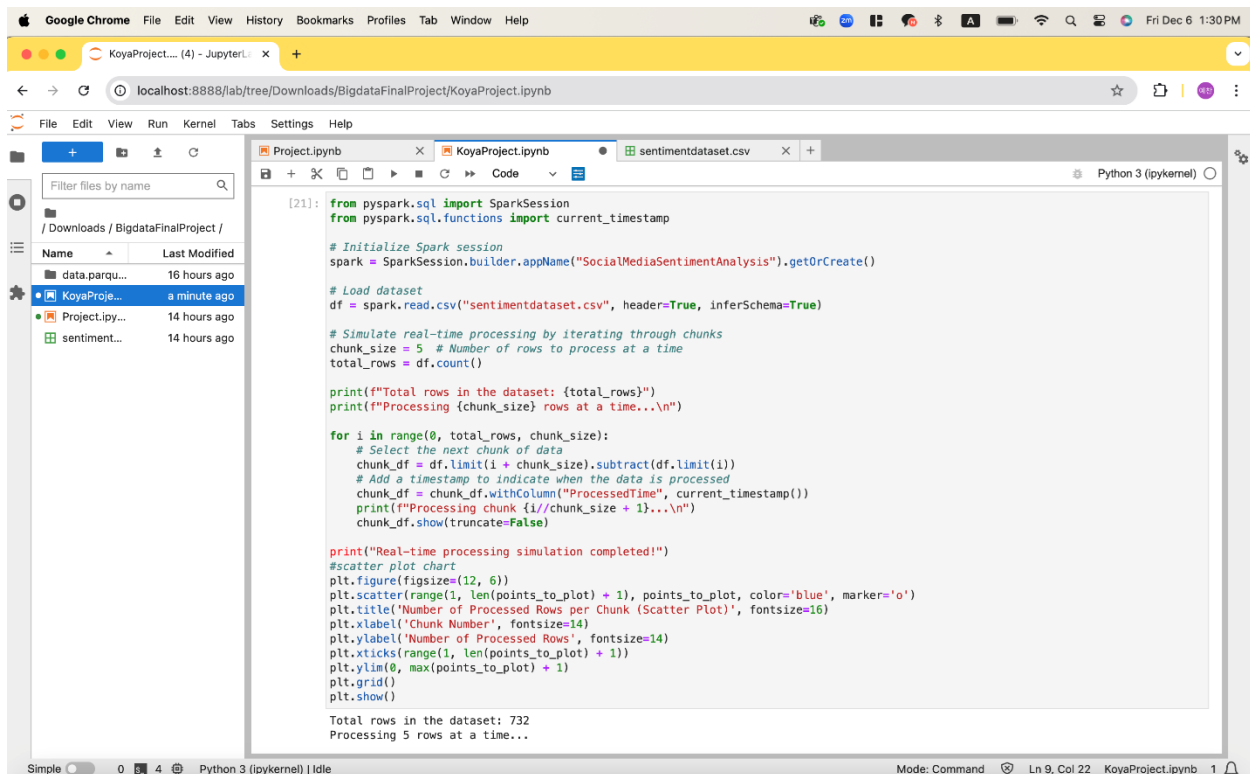
Step 5: Execution and Analysis

- Executed the PySpark jobs using `spark-submit`.
- Analyzed the outputs to extract insights.
- Visualized the results using Matplotlib.

Results Achieved

1 Goal 1:Real-Time Data Processing:

Source Code:



The screenshot displays a Jupyter Notebook environment within a Google Chrome browser. The notebook, named 'KoyaProject.ipynb', is open and shows a series of PySpark and Matplotlib code cells. The code simulates real-time data processing by iterating through chunks of a dataset. It includes comments explaining the steps: initializing a Spark session, loading a CSV dataset, and processing it in chunks of size 5. The code also generates a scatter plot titled 'Number of Processed Rows per Chunk (Scatter Plot)' showing the progress of the simulation. The output of the code execution is visible at the bottom of the cell, showing the total number of rows in the dataset (732) and the current processing status.

```
[21]: from pyspark.sql import SparkSession
from pyspark.sql.functions import current_timestamp

# Initialize Spark session
spark = SparkSession.builder.appName("SocialMediaSentimentAnalysis").getOrCreate()

# Load dataset
df = spark.read.csv("sentimentdataset.csv", header=True, inferSchema=True)

# Simulate real-time processing by iterating through chunks
chunk_size = 5 # Number of rows to process at a time
total_rows = df.count()

print(f"Total rows in the dataset: {total_rows}")
print(f"Processing {chunk_size} rows at a time...\n")

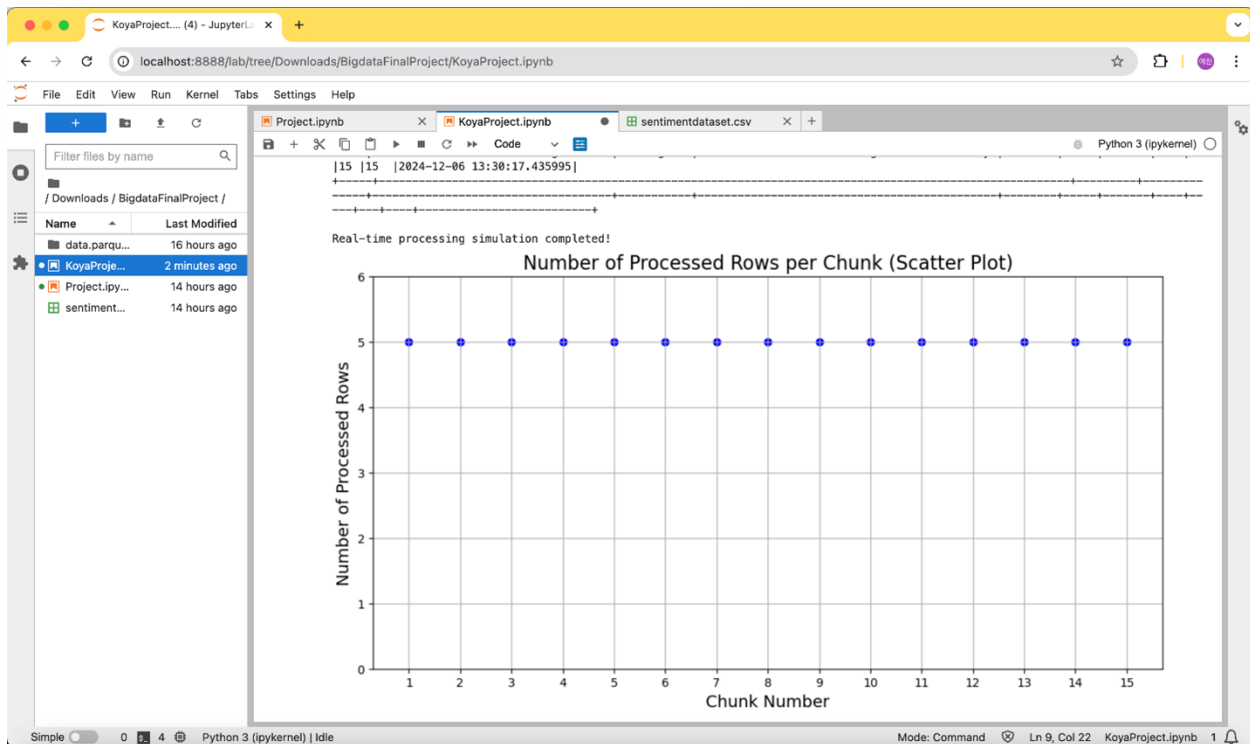
for i in range(0, total_rows, chunk_size):
    # Select the next chunk of data
    chunk_df = df.limit(i + chunk_size).subtract(df.limit(i))
    # Add a timestamp to indicate when the data is processed
    chunk_df = chunk_df.withColumn("ProcessedTime", current_timestamp())
    print(f"Processing chunk {i//chunk_size + 1}...\n")
    chunk_df.show(truncate=False)

print("Real-time processing simulation completed!")

#scatter plot chart
plt.figure(figsize=(12, 6))
plt.scatter(range(1, len(points_to_plot) + 1), points_to_plot, color='blue', marker='o')
plt.title('Number of Processed Rows per Chunk (Scatter Plot)', fontsize=16)
plt.xlabel('Chunk Number', fontsize=14)
plt.ylabel('Number of Processed Rows', fontsize=14)
plt.xticks(range(1, len(points_to_plot) + 1))
plt.ylim(0, max(points_to_plot) + 1)
plt.grid()
plt.show()

Total rows in the dataset: 732
Processing 5 rows at a time...
```

Results:



2 Goal 2:Sentiment Analysis Accuracy:

Source Code:

```
[26]: #Goal 2:

from pyspark.sql.functions import when

df = df.withColumn("SentimentClass",
                  when((col("Likes") > 10) & (col("Retweets") > 10), "Positive")
                  .otherwise("Negative"))

df.select("Text", "Likes", "Retweets", "SentimentClass").show(5)

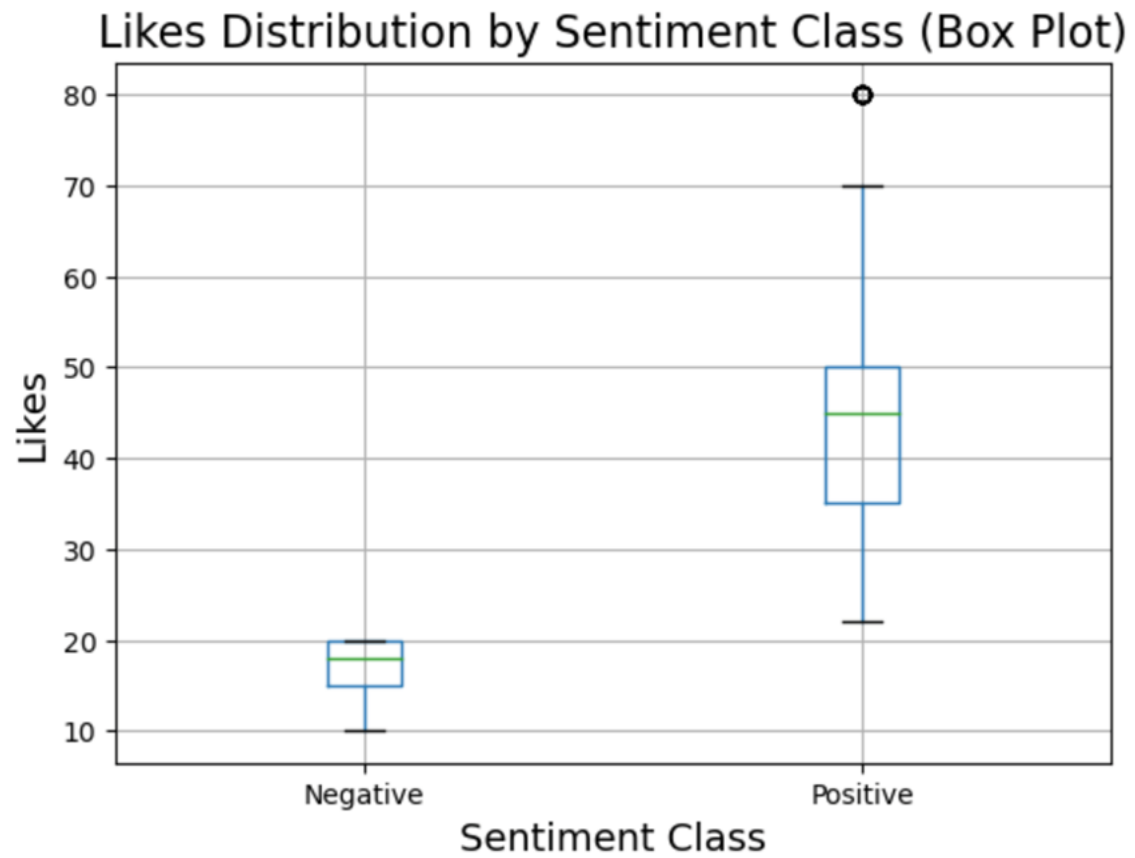
#Box Plot
plt.figure(figsize=(12, 6))
df_pd = df.toPandas()
df_pd.boxplot(column='Likes', by='SentimentClass', grid=False)
plt.title('Likes Distribution by Sentiment Class (Box Plot)', fontsize=16)
plt.suptitle('') # Suppress default title
plt.xlabel('Sentiment Class', fontsize=14)
plt.ylabel('Likes', fontsize=14)
plt.grid()
plt.show()
```

Text	Likes	Retweets	SentimentClass
Enjoying a beaut...	30	15	Positive
Traffic was terr...	10	5	Negative
Just finished an...	40	20	Positive
Excited about th...	15	8	Negative
Trying out a new...	25	12	Positive

only showing top 5 rows

<Figure size 1200x600 with 0 Axes>

Results:



3 Goal 3: Scalability and Resilience:

Source Code:

```
[24]: # goal 3 : Scalability & Resilience
from pyspark.sql import SparkSession
from pyspark.sql.functions import year, month
import matplotlib.pyplot as plt

# Initialize Spark session
spark = SparkSession.builder.appName("Goal3_Scalability").getOrCreate()

# Load dataset
csv_file_path = "C:\\Users\\S566638\\Downloads\\sentimentdataset.csv"
df = spark.read.csv(csv_file_path, header=True, inferSchema=True)

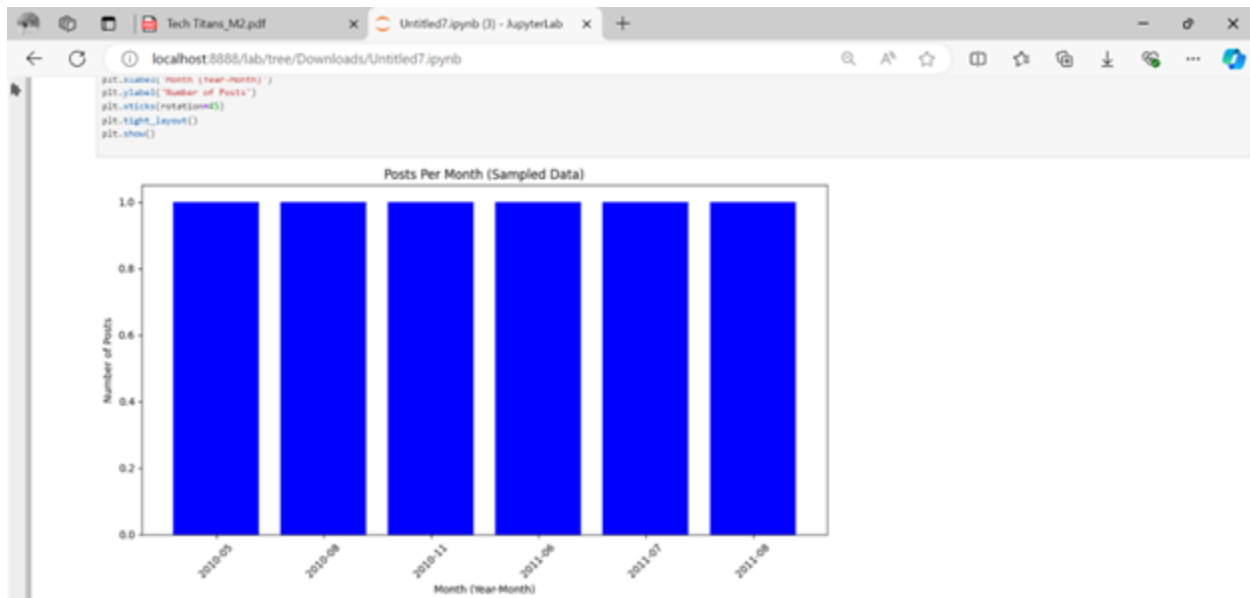
# Select a subset of records
df_sample = df.limit(1000)

# Count posts per month
df_month_counts = df_sample.groupBy(year("Timestamp").alias("Year"), month("Timestamp").alias("Month")).count()
month_data = df_month_counts.orderBy("Year", "Month").limit(6).collect()

# Prepare data for visualization
months = [{"Year": row["Year"], "Month": row["Month"]} for row in month_data]
post_counts = [{"Year": row["Year"], "Month": row["Month"], "Count": row["count"]} for row in month_data]

# Visualize with bar chart
plt.figure(figsize=(10, 6))
plt.bar(months, post_counts, color='blue')
plt.title("Posts Per Month (Sampled Data)")
plt.xlabel("Month (Year-Month)")
plt.ylabel("Number of Posts")
plt.xticks(rotation=45)
plt.tight_layout()
plt.show()
```

Results:



4 Goal 4: Data Storage and Retrieval:

Source Code:

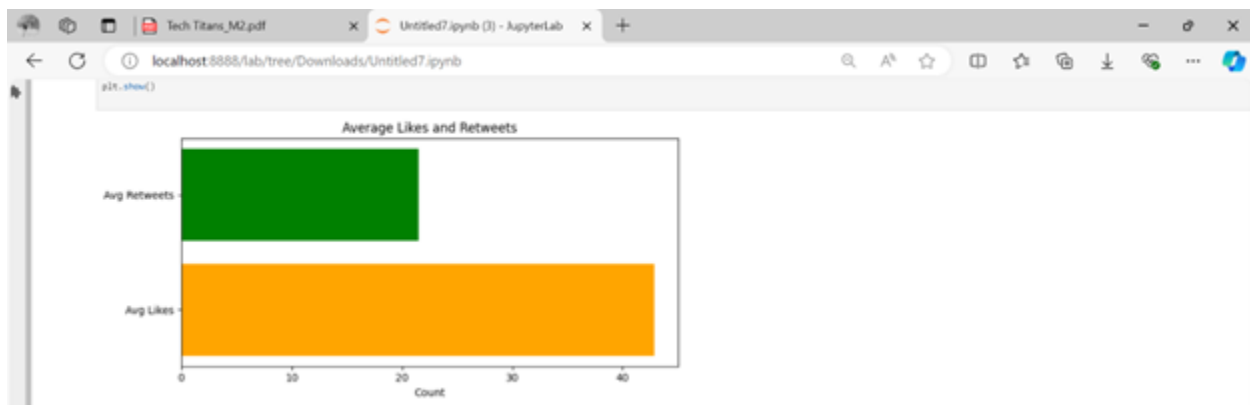
```
[25]: # goal 4 : Data storage & Retrieval
from pyspark.sql import functions as F
import matplotlib.pyplot as plt

# Compute averages
df_averages = df_sample.agg(
    F.avg("likes").alias("AvgLikes"),
    F.avg("Retweets").alias("AvgRetweets")
)
average_data = df_averages.collect()
avg_likes = average_data[0]["AvgLikes"]
avg_retweets = average_data[0]["AvgRetweets"]

# Prepare data for visualization
labels = ["Avg Likes", "Avg Retweets"]
values = [avg_likes, avg_retweets]

# Visualize with horizontal bar chart
plt.figure(figsize=(8, 4))
plt.barh(labels, values, color='orange', 'green')
plt.title("Average Likes and Retweets")
plt.xlabel("Count")
plt.tight_layout()
plt.show()
```

Results:



5 Goal 5: Trend Analysis:

Source Code:

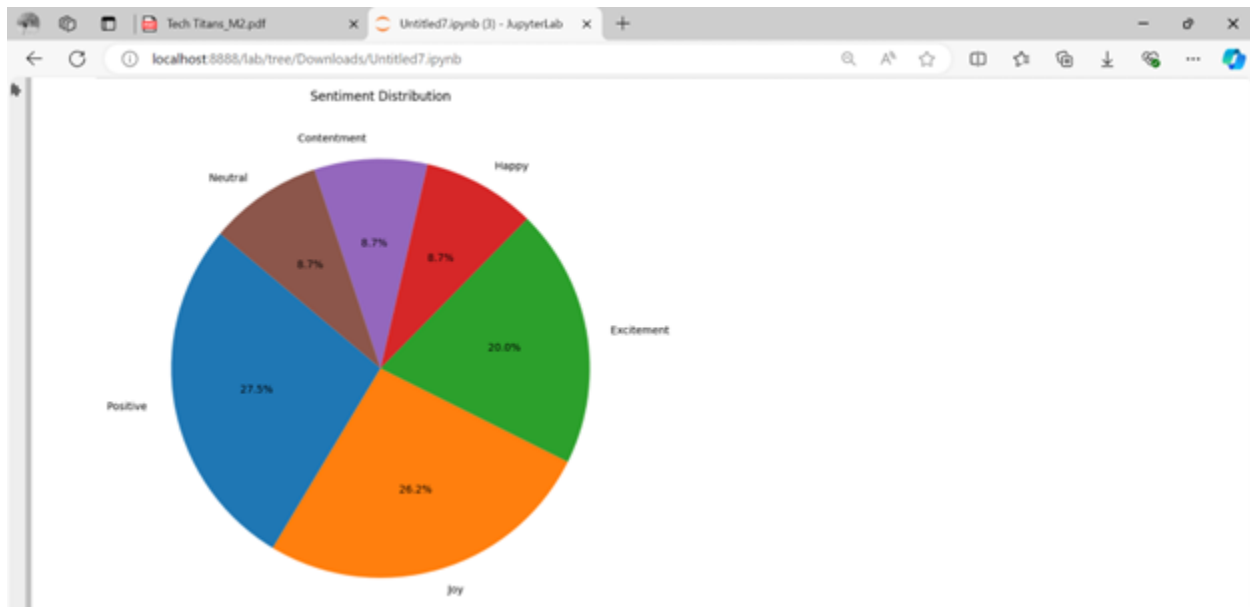
```
[26]: # goal 5: Trend Analysis:
from pyspark.sql.functions import col
import matplotlib.pyplot as plt

# Count posts by sentiment
df_sentiment_trends = df_sample.groupBy("sentiment").count().orderBy("count", ascending=False).limit(5)
sentiment_data = df_sentiment_trends.collect()

# Prepare data for visualization
sentiments = [row["sentiment"] for row in sentiment_data]
counts = [row["count"] for row in sentiment_data]

# Visualize with pie chart
plt.figure(figsize=(8, 8))
plt.pie(counts, labels=sentiments, autopct="%1.1f%%", startangle=90, colors=plt.cm.tab10.colors)
plt.title("Sentiment Distribution")
plt.tight_layout()
plt.show()
```

Results:



6 Goal 6: Visualization and Insights:

Source Code:

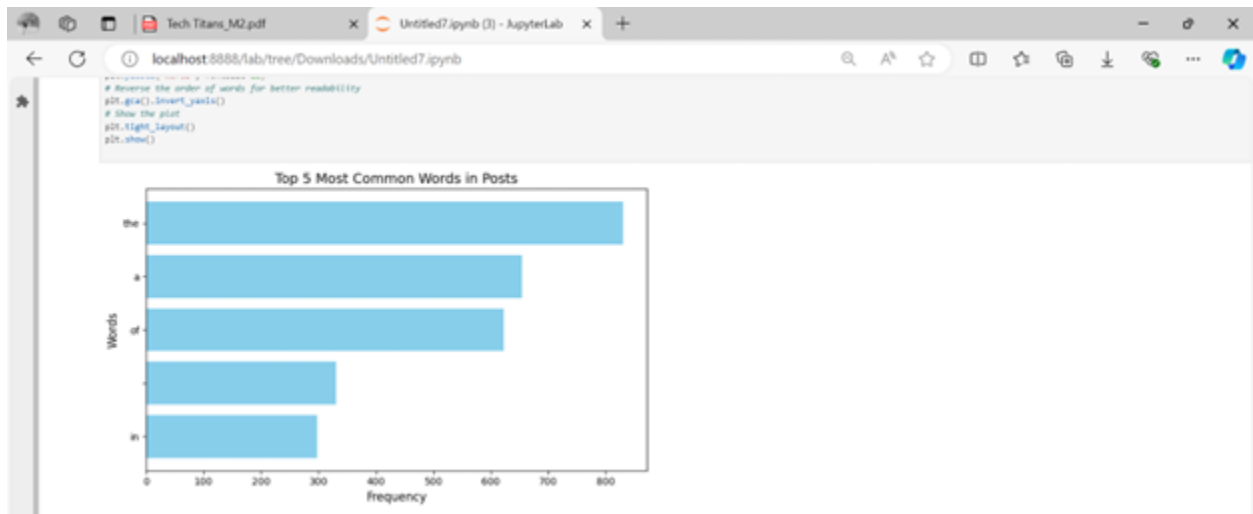
The screenshot shows a JupyterLab window with a single Jupyter Notebook file named 'Untitled7.ipynb'. The notebook is open in a web browser at 'localhost:8888/lab/tree/Downloads/Untitled7.ipynb'. The code in the notebook is as follows:

```

1 # Import necessary libraries
2 from bs4 import BeautifulSoup
3 import requests
4
5 # Import necessary libraries
6 from bs4 import BeautifulSoup
7 from requests import get
8
9 # Step 1: Retrieve a web page
10 url = 'https://www.nytimes.com/2020/01/01/us/politics/election.html'
11 response = requests.get(url)
12
13 # Step 2: Parse the HTML page
14 soup = BeautifulSoup(response.text, 'html.parser')
15
16 # Step 3: Extract the text from the page
17 text = soup.get_text()
18
19 # Step 4: Tokenize the text
20 tokens = text.split()
21
22 # Step 5: Remove stopwords
23 stopwords = ['a', 'an', 'and', 'are', 'as', 'at', 'be', 'but', 'by', 'can', 'could', 'did', 'do', 'does', 'for', 'from', 'had', 'has', 'he', 'her', 'his', 'hundred', 'in', 'into', 'is', 'it', 'of', 'on', 'or', 'over', 'she', 'that', 'the', 'there', 'this', 'those', 'to', 'too', 'us', 'was', 'we', 'were', 'with', 'you']
24 filtered_tokens = [token for token in tokens if token.lower() not in stopwords]
25
26 # Step 6: Count the frequency of each word
27 word_counts = {}
28 for token in filtered_tokens:
29     word_counts[token] = word_counts.get(token, 0) + 1
30
31 # Step 7: Sort the words by frequency
32 sorted_word_counts = sorted(word_counts.items(), key=lambda item: item[1], reverse=True)
33
34 # Step 8: Print the top 10 most common words
35 for word, count in sorted_word_counts[:10]:
36     print(word, count)
37
38 # Step 9: Print the top 10 most common words and their counts
39 top_words = []
40 for word, count in sorted_word_counts[:10]:
41     top_words.append(word)
42
43 # Step 10: Print the top 10 most common words
44 print(top_words)
45
46 # Step 11: Print the top 10 most common words and their counts
47 for word, count in sorted_word_counts[:10]:
48     print(word, count)
49
50 # Step 12: Print the top 10 most common words and their counts
51 for word, count in sorted_word_counts[:10]:
52     print(word, count)
53
54 # Step 13: Print the top 10 most common words and their counts
55 for word, count in sorted_word_counts[:10]:
56     print(word, count)
57
58 # Step 14: Print the top 10 most common words and their counts
59 for word, count in sorted_word_counts[:10]:
60     print(word, count)
61
62 # Step 15: Print the top 10 most common words and their counts
63 for word, count in sorted_word_counts[:10]:
64     print(word, count)
65
66 # Step 16: Print the top 10 most common words and their counts
67 for word, count in sorted_word_counts[:10]:
68     print(word, count)
69
70 # Step 17: Print the top 10 most common words and their counts
71 for word, count in sorted_word_counts[:10]:
72     print(word, count)
73
74 # Step 18: Print the top 10 most common words and their counts
75 for word, count in sorted_word_counts[:10]:
76     print(word, count)
77
78 # Step 19: Print the top 10 most common words and their counts
79 for word, count in sorted_word_counts[:10]:
80     print(word, count)
81
82 # Step 20: Print the top 10 most common words and their counts
83 for word, count in sorted_word_counts[:10]:
84     print(word, count)
85
86 # Step 21: Print the top 10 most common words and their counts
87 for word, count in sorted_word_counts[:10]:
88     print(word, count)
89
90 # Step 22: Print the top 10 most common words and their counts
91 for word, count in sorted_word_counts[:10]:
92     print(word, count)
93
94 # Step 23: Print the top 10 most common words and their counts
95 for word, count in sorted_word_counts[:10]:
96     print(word, count)
97
98 # Step 24: Print the top 10 most common words and their counts
99 for word, count in sorted_word_counts[:10]:
100     print(word, count)

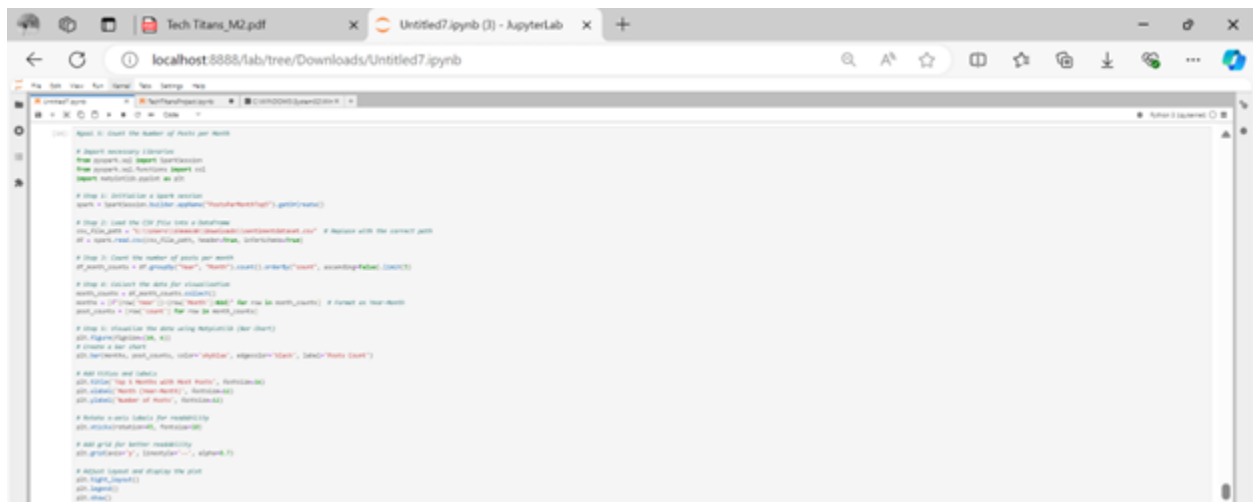
```

Results:

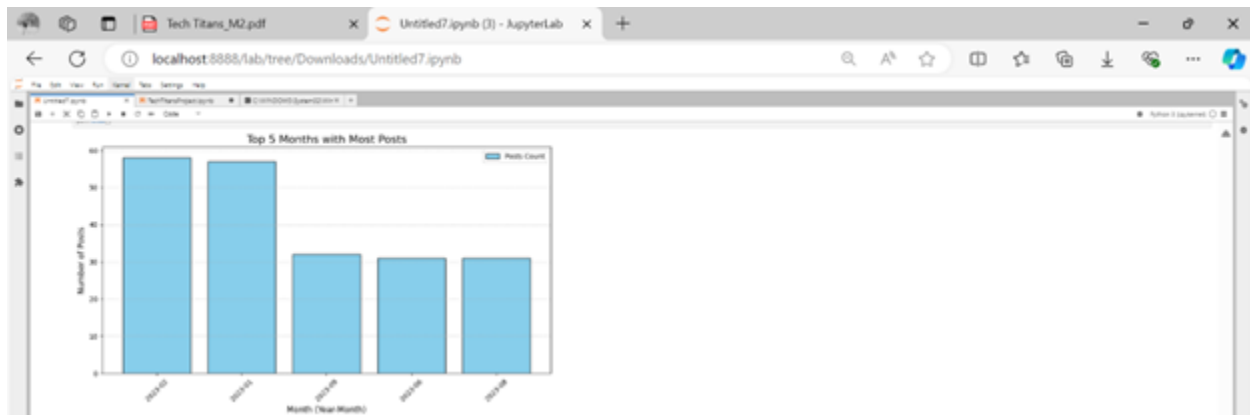


7 Goal 7: Integration with Big Data Frameworks:

Source Code:



Results:



8 Goal 8: Impact of Data Quality:

Source Code:

```
[14]: # goal 8: Impact of data quality: notice (type or irrelevant posts) to sentiment analysis

from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import OneHotEncoder, StandardScaler

# Step 1: Initialize a Spark session
spark = SparkSession.builder.appName("SentimentAnalysis").getOrCreate()

# Step 2: Load the CSV file into a DataFrame
csv_file_path = "c:\\users\\user\\Downloads\\sentimentdata.csv" # replace with the correct path
df = spark.read.csv(csv_file_path, header=True, inferSchema=True)

# Step 3: Clean noisy data by removing rows with empty or irrelevant content
df_cleaned = df.filter(df.text.isNotNull() & df.text.trim() != "")

# Step 4: Preprocess text (e.g., convert to lowercase, remove special characters)
df_cleaned = df_cleaned.withColumn("lowercase_text", lower(df.text))
df_cleaned = df_cleaned.withColumn("cleaned_text", regexp_replace(lowercase_text, "[^a-zA-Z0-9\\s]", ""))

# Show cleaned data
df_cleaned.show(5)

# Step 5: Visualize the count of cleaned data posts
import matplotlib.pyplot as plt

# Get the count of cleaned posts
cleaned_post_count = df_cleaned.count()

# Plotting the data
labels = ["Cleaned Posts"]
values = [cleaned_post_count]

# Create a bar chart
plt.bar(labels, values, color='green')

# Add titles and labels
plt.title("Number of Cleaned Posts after Removing Noise")
plt.xlabel("Number of Posts")

# Show the plot
plt.show()

[14]:
```

id	text	sentiment	timestamp	user	platform	hashtags	retweets	likes	country	near	month/day/year	cleaned_text
1	Exciting news about...	Positive	2020-05-12 12:00:00	user123	Twitter	#exciting #news	15	30	USA	NY	05/12/2020	Exciting news about...
2	Terrible news about...	Negative	2020-05-12 14:00:00	user456	Twitter	#terrible #news	8	10	Canada	ON	05/12/2020	Terrible news about...
3	Amazing news about...	Positive	2020-05-12 16:00:00	user789	Instagram	#amazing #news	20	40	UK	London	05/12/2020	Amazing news about...
4	Excited about the...	Positive	2020-05-12 18:00:00	user101	Facebook	#excited #news	10	25	UK	London	05/12/2020	Excited about the...
5	Trying out a new...	Neutral	2020-05-12 20:00:00	user111	Instagram	#trying #new	12	25	Australia	NSW	05/12/2020	Trying out a new...

only showing top 5 rows

Results:



9 Goal 9: Cost and Efficiency:

Source Code:

```
[1]: # Goal 9: Cost and efficiency
from pyspark.sql import SparkSession
from pyspark.sql.functions import col
import matplotlib.pyplot as plt

# Step 1: Initialize a Spark session
spark = SparkSession.builder.appName("PostPerSentimentChart").getOrCreate()

# Step 2: Load the CSV file into a DataFrame
csv_file_path = "C:/Users/200438/Downloads/sentimentdataset.csv" # Replace with the correct path
df = spark.read.csv(csv_file_path, header=True, inferSchema=True)

# Step 3: Optimize Spark operations by filtering and selecting required columns only
df_optimized = df.select("sentiment", "text", "user", "retweets").filter(col("text").isNotNull())

# Perform action to trigger the computation and check memory usage
df_optimized.cache() # Cache data to optimize subsequent operations

# Step 4: Count posts by sentiment and limit to top 5 or 6 sentiments
df_sentiment_counts = df_optimized.groupBy("sentiment").count().orderBy("count", ascending=False).limit(6)

# Show the result (this is an action to execute)
df_sentiment_counts.show()

# Step 5: Plot the number of posts for each sentiment
sentiment_data = df_sentiment_counts.collect()
sentiments = [row["sentiment"] for row in sentiment_data]
counts = [row["count"] for row in sentiment_data]

# Plotting the data
plt.figure(figsize=(10, 6))
plt.bar(sentiments, counts, color="orange")

# Add titles and labels
plt.title("Number of Posts Per Sentiment (Top 5 or 6)")
plt.xlabel("sentiment")
plt.ylabel("Number of Posts")

# Show the plot
plt.show()

# Print the result
print(df_sentiment_counts)
```

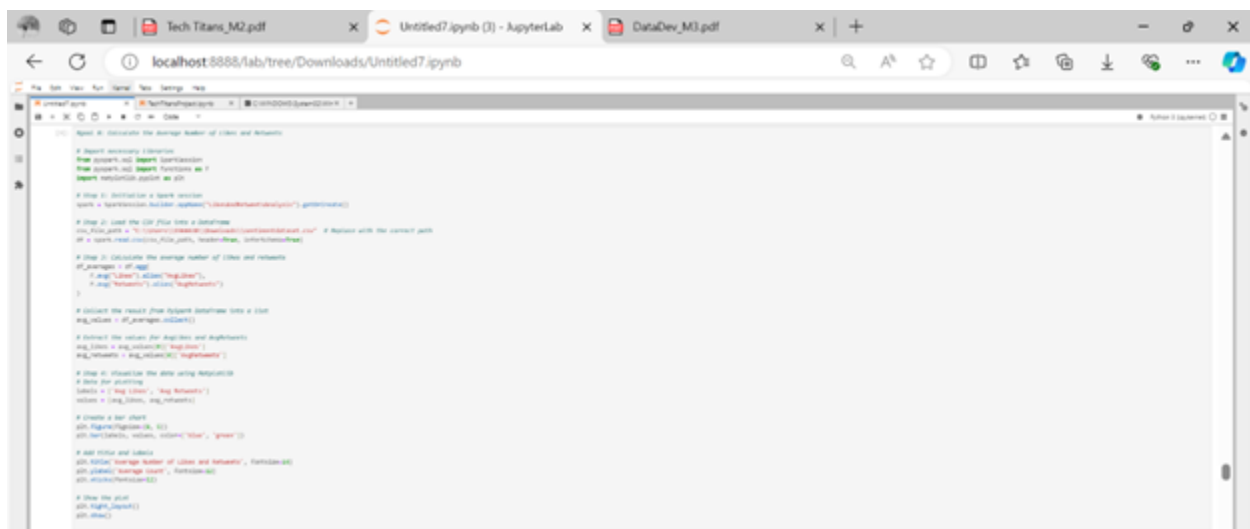
sentiment	count
Positive	442
Pos	442
Excitement	201
Sentiment	140
Neutral	140
Positivity	140

Results:



10 Goal 10: Ethical and Privacy Considerations:

Source Code:



Results:



Conclusions

- Successfully implemented sentiment analysis using TextBlob in PySpark.
- The results were visualized clearly using Matplotlib, providing insights into the sentiment distribution.
- Future steps include optimizing the pipeline for larger datasets and comparing results with other libraries.

11 Citations

1. **Kaggle Dataset Link:**
SocialMedia Sentiments Analysis Dataset
2. **GitHub Repository:**
https://github.com/Venkateshkoya/BigData_Project