Apple stock prices from yahoo finance

Meron and MD

PySpark advantages over Hadoop

- High-level programming interface
- Built on Apache Spark for performance and scalability
- Supports a wide range of data sources and formats

Advantages of PySpark

- High-level, Pythonic API for data processing:
- DataFrame API for powerful and easy-to-use interface
- Built-in functions and transformations for data processing
- Designed for distributed computing and can scale to handle large datasets:
- Includes MLlib for machine learning, GraphX for graph processing, and Spark Streaming for real-time data processing

Dataset Overview

- The dataset used in this project is the daily historical stock price data for Apple Inc. (AAPL) from Yahoo Finance.
- The dataset contains daily stock prices for AAPL from 1980 to the present.
- The dataset is provided in CSV format and has a file size of approximately 250 MB.
- The dataset is used in this project to demonstrate how to fetch and process data using PySpark and Python, and to perform basic data analysis and visualization on the results.

Code Overview

```
In [1]: import yfinance as yf
```

PySpark and Initializing Spark Session

```
In [2]: import pyspark
In [3]: from pyspark.sql import SparkSession
In [4]: spark=SparkSession.builder.appName("Gold").getOrCreate()
        Setting default log level to "WARN".
        To adjust logging level use sc.setLogLevel(newLevel). For SparkR, use setLogLevel(newLevel).
        23/03/06 12:07:30 WARN NativeCodeLoader: Unable to load native-hadoop library for your platfo
        rm... using builtin-java classes where applicable
In [5]: spark
Out[5]: SparkSession - in-memory
        SparkContext
        Spark UI
        Version
        v3.3.2
        Master
        local[*]
        AppName
        Gold
```

Fetching dataset

```
In [7]: # Fetch data from yfinance
       data = yf.download("AAPL", start="1980-01-01", end="2022-03-01", interval="1d")
       data['Close'].plot(figsize = (20,8))
        [********* 100%********** 1 of 1 completed
Out[7]: <AxesSubplot:xlabel='Date'>
        175
        150
        125
        100
         75 -
         50
         25
```

Converting dataset to PySpark dataframe

```
In [23]: # Convert the pandas DataFrame to a PySpark DataFrame
df = spark.createDataFrame(data.reset_index()).select(['Date','Open','Close'])
df.count()
Out[23]: 10391
```

Smart anaysis

```
In [21]: # Calculate the rolling mean and standard deviation of the closing price
         window = Window.orderBy(col("Date")).rowsBetween(-200, 0)
         df = df.withColumn("rolling mean", mean(col("Close")).over(window)) \
               .withColumn("rolling stddev", stddev(col("Close")).over(window)).na.drop()
In [22]: df.show(4)
                                         Open
                                                    Close
                                                                         rolling mean
                      Date
                                                                                            rollin
      g stddev
      1980-12-15 00:00:00|0.12221000343561172|0.12165199965238571| 0.1249999962747097| 0.00473478
      22300699
      1980-12-16 00:00:00 0.1132809966802597 0.11272300034761429 0.12090766429901123 0.00783904
      48640804
      1980-12-17 00:00:00|0.11551299691200256|0.11551299691200256|0.11955899745225906| 0.006945695
      93164169
      1980-12-18 00:00:00|0.11886200308799744|0.11886200308799744|0.11941959857940673|0.0060232200
      only showing top 4 rows
```

Simple Visualization

Converting to pandas for visualization

```
In [24]: import matplotlib.pyplot as plt
In [25]: pdf = df.toPandas()
In [26]: pdf.head()
Out[26]:
                  Date
                          Open
                                  Close
           0 1980-12-12 0.128348 0.128348
           1 1980-12-15 0.122210 0.121652
           2 1980-12-16 0.113281 0.112723
           3 1980-12-17 0.115513 0.115513
           4 1980-12-18 0.118862 0.118862
```

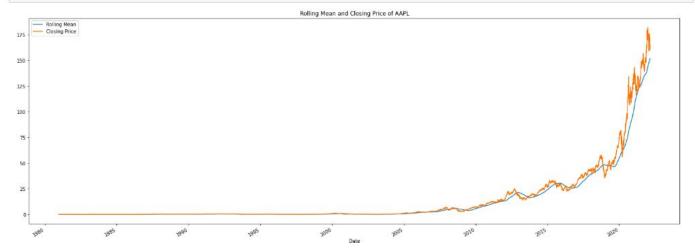
A graph of rolling mean

```
In [14]: # Set the date column as the index for the Pandas DataFrame
    pdf.set_index('Date', inplace=True)

# Plot the rolling mean and closing price against the index
    pdf[['rolling_mean', 'Close']].plot(figsize=(25,9))

# Set the title and legend for the plot
    plt.title("Rolling Mean and Closing Price of AAPL")
    plt.legend(['Rolling Mean', 'Closing Price'])

# Show the plot
    plt.show()
```



A graph for standard deviation

```
In [15]: # Plot the rolling standard deviation and closing price against the index
pdf[['rolling_stddev', 'Close']].plot(figsize=(25,9))

# Set the title and legend for the plot
plt.title("Rolling Standard Deviation and Closing Price of AAPL")
plt.legend(['Rolling Standard Deviation', 'Closing Price'])

# Show the plot
plt.show()
```

