Data Retrieval

April 10, 2023

1 Data Retrieval

This code file completes the data collection and storage process. Headlines and financial data are gathered from respective sources, processed into final forms and stored in the PostgreSQL database.

*The database used in this work is locally hosted, usage of this file outside of the environment in which it was created will be unsuccessful

```
[]: !pip install yfinance
     !pip install pyspark
     !pip install findspark
     !pip install dateparser
     !pip install vaderSentiment
     import pandas as pd
     import csv
     import datetime
     import yfinance as yf
     import numpy as np
     from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
     from bs4 import BeautifulSoup as bs
     import requests
     from pyspark.sql.functions import sum, max, min, mean, count
     import datetime as dt
     import pyspark
     from pyspark.sql import SparkSession
     import pyspark.pandas as ps
     import findspark
     import yaml
     from yaml.loader import SafeLoader
     from os.path import abspath
     warehouse_location = abspath('spark-warehouse')
     with open('cfg.yml') as f:
         config = yaml.load(f, Loader = SafeLoader)
     findspark.init()
     spark = SparkSession.builder \
         .master(config['spark']['spark_master'])\
```

```
.appName('gather')\
.enableHiveSupport()\
.config('spark.sql.warehouse.dir', warehouse_location)\
.config(config['spark']['spark_jars'], config['spark']['spark_jars_path'])\
.config('spark.cores.max', '2')\
.config('spark.executor.cores', '2')\
.getOrCreate()
spark.sparkContext.setLogLevel("WARN")
spark
```

```
[2]: url = config['postgres']['url']
props = {
        'user': config['postgres']['user'],
        'password' : config['postgres']['user'],
        'url': url,
        'driver': config['postgres']['driver']
```

2 Retrieve Headlines for Sentiment Analysis

```
[4]: #retrieve headlines from financial post
     headers = {'User-Agent':
             'Mozilla/5.0 (X11; Ubuntu; Linux x86_64; rv:108.0) Gecko/20100101

→Firefox/108.0'}

     def gather headlines(company name, ticker):
        headlines = []
        dates = []
        for i in range(10, 30000, 10): # Running for-loop
             info_url = "https://financialpost.com/search/?
      search_text="+company_name +"&date_range=-3650d&sort=asc&from="+str(i)
             page = requests.get(info url, headers = headers)
            parser = bs(page.content, "html.parser" )
            date = parser.body.find_all('div', attrs={'class':__
      ⇔'article-card__meta-bottom'})
            for span in date:
                 dates.append(span.text.split(" ")[1])
            headline = parser.body.find_all('h3', class_ = 'article-card_headline_
      →text-size--extra-large--sm-up')
             for x in headline:
                headlines.append(x.text)
        dates = dates[:len(headlines)]
        file = {'date' : dates, "headline" : headlines}
        file = pd.DataFrame(file)
        print(file.head())
        file['ticker'] = ticker
```

```
#calculate sentiment scores for each headlines and append to dataset

def analyze_sent(df):
    analyze_obj = SentimentIntensityAnalyzer()
    df['sentiment']=df['headline'].apply(lambda headline: analyze_obj.
    polarity_scores(str(headline))['compound'])
    df.fillna(0, inplace = True)
    return df

def final_sentiment(df):
    return df.withColumn("sent_score", df.mean_sentiment*(df.
    headline_count**2)).drop('headline', 'headline_count', 'mean_sentiment')
```

3 Process Sentiment Scores and Write to Database

```
[5]: import dateparser
     ticker_list = ['MSFT','GOOG','NFLX','TSLA', 'AMZN']
     company_list = ['microsoft', 'google', 'netflix', 'tesla', 'amazon']
     def process_headlines(ticker_list, company_list):
         dfs = []
         for tick, company in zip(ticker_list, company_list):
             data = gather headlines(company, tick)
             dfs.append(data)
         full_df = pd.concat(dfs)
         dates = []
         for index, row in full_df.iterrows():
             date = dateparser.parse(row['date'], date_formats = ["%d-%m-%y"])
             dates.append(date.date())
         full_df['date'] = dates
         full_df = ps.from_pandas(full_df)
         print(full df.head())
         full_df = analyze_sent(full_df)
         full_df = full_df.to_spark()
         full_df.show()
         aggregated = full_df.groupBy('date', 'ticker').agg(count('headline').
      alias('headline_count'), mean('sentiment').alias("mean_sentiment"))
         final_news = final_sentiment(aggregated)
         final_news.write.format("jdbc")\
             .option("url", "jdbc:postgresql://localhost:5432/financials") \
             .option("driver", "org.postgresql.Driver").option("dbtable", __

¬"sentiment") \

             .option("user", "adam").option("password", "green").mode('append').
      ⇒save()
```

process_headlines(ticker_list, company_list)

O April 10, 2013 Personal computer shipments shrink 14% in wor					
1 Appel 11 0012 [Thet man mail to be a form mail to					
1 April 11, 2013 What you need to know before markets open					
2 April 11, 2013 Microsoft's Windows 8 gets blame for worst PC					
3 April 11, 2013 Microsoft falls after Goldman warns of PC los					
4 April 11, 2013 Electronic Arts lays off employees at Montrea					
date headline					
0 April 5, 2013 4.05.13: 'Flurry of disappointment' for Canad					
1 April 5, 2013 Facebook Home Q&A with mobile engineering dir					
2 April 6, 2013 Discontinued products that we still miss dearly					
3 April 8, 2013 How DIRTT has built a successful green manufa					
4 April 8, 2013 Review: The HTC One is the most beautiful And					
date headline					
O April 22, 2013 Google chairman Schmidt explores future of In					
1 April 22, 2013 Netflix shares surge as profit, U.S. streamin					
2 April 22, 2013 Closing Bell: TSX closes modestly higher amid					
3 April 23, 2013 4.23.13: Mark Carney, travelling light but ta					
4 April 23, 2013 Netflix becomes S&P's top performer as it stu					
date headline					
O July 9, 2013 What you need to know before markets open					
1 July 9, 2013 Tesla rises to record high as stock heads for					
2 July 13, 2013 13 expensive side projects keeping the bigges					
3 July 16, 2013 Tesla CEO Elon Musk morphs from Tony Stark to					
4 July 18, 2013 GM is so afraid of Tesla it has assembled a s					
date headline					
0 May 1, 2013 Transforce ready to ride out energy slump					
1 May 6, 2013 How Al Gore amassed a \$200-million fortune af					
2 May 7, 2013 The 40 most undervalued stocks in the market					
3 May 7, 2013 Desire2Learn launches software that predicts					
4 May 9, 2013 Canadian TV producers DHX Media, Nelvana, OUT					
23/03/30 14:01:26 WARN TaskSetManager: Stage 0 contains a task of very large					
size (1907 KiB). The maximum recommended task size is 1000 KiB.					

date	headline
ticker	
0 2013-04-10	Personal computer shipments shrink 14% in worst-ever decline
MSFT	
1 2013-04-11	What you need to know before markets open
MSFT	
2 2013-04-11	Microsoft's Windows 8 gets blame for worst PC decline on record
MSFT	
3 2013-04-11	Microsoft falls after Goldman warns of PC losses
MSFT	

4 2013-04-11 Electronic Arts lays off employees at Montreal studio MSFT

23/03/30 14:01:29 WARN TaskSetManager: Stage 1 contains a task of very large size (1907 KiB). The maximum recommended task size is 1000 KiB.

/home/cis6180/anaconda3/lib/python3.9/site-packages/pyspark/pandas/utils.py:975: PandasAPIOnSparkAdviceWarning: If `index_col` is not specified for `to_spark`, the existing index is lost when converting to Spark DataFrame.

warnings.warn(message, PandasAPIOnSparkAdviceWarning)

23/03/30 14:01:29 WARN TaskSetManager: Stage 2 contains a task of very large size (1907 KiB). The maximum recommended task size is 1000 KiB.

+			
date		ticker	sentiment
2013-04-10	Personal compute	MSFT	0.01
2013-04-11	What you need to	MSFT	0.0
2013-04-11	Microsoft's Wind	MSFT	-0.7579
2013-04-11	Microsoft falls	MSFT	-0.4767
2013-04-11	Electronic Arts	MSFT	0.0
2013-04-12	4.12.13: BlackBe	MSFT	0.0
2013-04-12	Motocross Madnes	MSFT	-0.4939
2013-04-12	Who says account	MSFT	-0.3182
2013-04-15	4.15.13: Gold an	MSFT	0.0
2013-04-15	Microsoft smartw	MSFT	0.0
2013-04-16	Facebook, Apple	MSFT	0.0
2013-04-16	Facebook Home se	MSFT	0.0
2013-04-17	Buying defensive	MSFT	-0.1531
2013-04-17	4.17.13: Stickin	MSFT	0.0
2013-04-17	TSX tumbles as g	MSFT	-0.2023
2013-04-17	The bull case fo	MSFT	0.5216
2013-04-17	Learning to use	MSFT	0.0
2013-04-18	Private equity t	MSFT	0.0
2013-04-18	Microsoft CFO Kl	MSFT	0.4019
2013-04-19	What you need to	MSFT	0.0
++		+	·+
	+ 00		

only showing top 20 rows

23/03/30 14:01:34 WARN TaskSetManager: Stage 3 contains a task of very large size (1907 KiB). The maximum recommended task size is 1000 KiB.

4 Retrieve, Process and Store Financial Data

```
[3]: def get_financials(ticker, start):
         time_delt = dt.timedelta(days = 150)
         start_day = start - time_delt
         data = yf.download(str(ticker), start_day)
         data['ticker'] = ticker
         data = data.reset index()
         data = data.rename(columns = {'Date':'date', 'Open':'open', 'High':'high', __
      -'Low':'low', 'Close':'close', 'Adj Close': 'adj_close', 'Volume':'volume'})
         print('success!')
         return data
     def EWMA(data, ndays):
         EMA = pd.Series(data['close'].ewm(span = ndays, min_periods = ndays - 1).
      ⊶mean(),
                      name = 'EWMA_' + str(ndays))
         data = data.join(EMA)
         return data
     def rsi(close, periods = 14):
         close_delta = close.diff()
         # Make two series: one for lower closes and one for higher closes
         up = close_delta.clip(lower=0)
         down = -1 * close_delta.clip(upper=0)
         ma_up = up.ewm(com = periods - 1, adjust=True, min_periods = periods).mean()
         ma_down = down.ewm(com = periods - 1, adjust=True, min_periods = periods).
      →mean()
         rsi = ma_up / ma_down
         rsi = 100 - (100/(1 + rsi))
         return rsi
     def BBANDS(data, window):
         MA = data.close.rolling(window).mean()
         SD = data.close.rolling(window).std()
         data['MiddleBand'] = MA
         data['UpperBand'] = MA + (2 * SD)
         data['LowerBand'] = MA - (2 * SD)
         return data
     def prep_financials(df):
        df = pd.DataFrame(df)
```

```
#df.set_index('date')
df['target'] = (df['close'].rolling(10).mean()
df['tenmda'] = df['close'].rolling(20).mean()
df['fiftymda'] = df['close'].rolling(50).mean()
df['hundredmda'] = df['close'].rolling(100).mean()
df = EWMA(df, 20)
df = EWMA(df, 50)
df = EWMA(df, 50)
df = EWMA(df, 100)
df['rsi'] = rsi(df['close'])
df = BBANDS(df, 40)
df.dropna(inplace = True)
df.reset_index()
print(df.head())
return df
```

```
[ ]: def process_finance(ticker_list):
        finance_dfs = []
        for tick in ticker_list:
            data = get_financials(tick, dt.date(2015,1, 1))
            data = prep_financials(data)
            finance_dfs.append(data)
        final_finance = pd.concat(finance_dfs)
        final_finance = spark.createDataFrame(final_finance)
        final_finance.write.format("jdbc")\
             .option("url", "jdbc:postgresql://localhost:5432/financials") \
             .option("driver", "org.postgresql.Driver").option("dbtable", __
      .option("user", "adam").option("password", "green").mode('append').
      ⇒save()
    ticker_list = ['MSFT','GOOG','NFLX','AMZN', 'TSLA']
    process_finance(ticker_list)
```

```
[]: spark.stop()
```