003 - Ploting data with Cufflinks

November 11, 2023

1 #003 Ploting data with Cufflinks

for future: - develope some about techinical analysis with that data

In this code will generate some graphical data as Candlestick plots, MACD, Bollinger Bands and other to improve analysis about one asset.

Implemented the same code from study #001, to obtain data from Yfinance, except for utilizing Cufflings to connect Pandas and Plotly for generate more useful info.

[1]:
```
#   !pip install pandas
#   !pip install pandas-datareader
#   !pip install yfinance
#   !pip install datetime
#   !pip install plotly_express
#   !pip install cufflinks
```

[2]:
```
#import libraries

import pandas as pd
from pandas_datareader import data as pdr
import numpy as np
import yfinance as yf
import datetime as dt
import plotly.express as px
import plotly.graph_objects as go
import cufflinks as cf

yf.pdr_override()
cf.go_offline() # Enabling offline mode for interactive data visualization␣
 ↪locally
```

1.1 3.1 Import, ajust and analyse DataFrame

[3]:
```
# Define the start and end dates, last 2 years
end = dt.datetime.now()
start = end - dt.timedelta(days = 365*2)
```

[4]:
```
# define Tickers
tk = input('Enter the ticker code: ')
```

```
Enter the ticker code: AAPL
```

[5]:
```
#obtain data from Yahoo Finance
df = pdr.get_data_yahoo(tk, start = start, end = end)
df
```

```
[**********100%%**********]  1 of 1 completed
```

[5]:

Date	Open	High	Low	Close	Adj Close	Volume
2021-11-11	148.960007	149.429993	147.679993	147.869995	146.199554	41000000
2021-11-12	148.429993	150.399994	147.479996	149.990005	148.295624	63804000
2021-11-15	150.369995	151.880005	149.429993	150.000000	148.305496	59222800
2021-11-16	149.940002	151.490005	149.339996	151.000000	149.294189	59256200
2021-11-17	151.000000	155.000000	150.990005	153.490005	151.756104	88807000
…	…	…	…	…	…	…
2023-11-06	176.380005	179.429993	176.210007	179.229996	178.994186	63841300
2023-11-07	179.179993	182.440002	178.970001	181.820007	181.580780	70530000
2023-11-08	182.350006	183.449997	181.589996	182.889999	182.649368	49340300
2023-11-09	182.960007	184.119995	181.809998	182.410004	182.169998	53763500
2023-11-10	183.970001	186.570007	183.529999	186.399994	186.399994	66133400

[503 rows x 6 columns]

[6]:
```
#test for Null values on DataFrame
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
DatetimeIndex: 503 entries, 2021-11-11 to 2023-11-10
Data columns (total 6 columns):
 #   Column     Non-Null Count  Dtype
---  ------     --------------  -----
```

```
 0   Open       503 non-null    float64
 1   High       503 non-null    float64
 2   Low        503 non-null    float64
 3   Close      503 non-null    float64
 4   Adj Close  503 non-null    float64
 5   Volume     503 non-null    int64
dtypes: float64(5), int64(1)
memory usage: 27.5 KB
```

[7]:
```
#calc a daily percentage change for adjusted close price
df['Daily Return'] = df['Adj Close'].pct_change(1) * 100
df['Daily Return'].replace(np.nan, 0, inplace = True) #replace the first row,␣
 ↪changes null for 0

df
```

[7]:

Date	Open	High	Low	Close	Adj Close	Volume	Daily Return
2021-11-11	148.960007	149.429993	147.679993	147.869995	146.199554	41000000	0.000000
2021-11-12	148.429993	150.399994	147.479996	149.990005	148.295624	63804000	1.433704
2021-11-15	150.369995	151.880005	149.429993	150.000000	148.305496	59222800	0.006657
2021-11-16	149.940002	151.490005	149.339996	151.000000	149.294189	59256200	0.666660
2021-11-17	151.000000	155.000000	150.990005	153.490005	151.756104	88807000	1.649035
…	…	…	…	…	…	…	…
2023-11-06	176.380005	179.429993	176.210007	179.229996	178.994186	63841300	1.460520
2023-11-07	179.179993	182.440002	178.970001	181.820007	181.580780	70530000	1.445071
2023-11-08	182.350006	183.449997	181.589996	182.889999	182.649368	49340300	0.588492
2023-11-09	182.960007	184.119995	181.809998	182.410004	182.169998	53763500	-0.262454
2023-11-10	183.970001	186.570007	183.529999	186.399994	186.399994	66133400	2.322005

[503 rows x 7 columns]

[8]:
```
df.describe().round(2)
```

[8]:

	Open	High	Low	Close	Adj Close	Volume	Daily Return
count	503.00	503.00	503.00	503.00	503.00	5.030000e+02	503.00
mean	161.83	163.79	160.12	162.04	161.04	7.725599e+07	0.07
std	16.22	16.00	16.39	16.19	16.29	2.694275e+07	1.88
min	126.01	127.77	124.17	125.02	124.33	3.145820e+07	-5.87
25%	148.86	150.76	147.26	149.30	148.28	5.695555e+07	-0.98
50%	163.06	165.39	161.00	163.64	162.25	7.159840e+07	0.08
75%	174.02	175.87	172.58	174.61	173.37	9.007030e+07	1.22
max	196.24	198.23	195.28	196.45	195.93	1.954327e+08	8.90

1.2 3.2 Ploting results with Cufflinks

[21]:
```
cf.set_config_file(theme='pearl', sharing='public', offline=True)
```

[22]:
```
# Plot Candlestick figure using Cufflinks QuantFig module,
figure = cf.QuantFig(df, title = tk + ' - Candlestick, RSI Chart', name =␣
 ↪tk,legend='top', rangeslider=False)
figure.add_sma(periods =[14, 21], column = 'Close', color = ['magenta',␣
 ↪'green'])   # plot 14 and 21, Simple Moving Average
figure.add_volume()
figure.add_rsi(periods=20, color='java') #plot RSI, with close price and 20␣
 ↪periods
figure.iplot(up_color = 'green', down_color = 'red')
```

[23]:
```
figure = cf.QuantFig(df, title = tk + ' - Bollinger Bands, MACD Chart ', name =␣
 ↪tk, legend='top', rangeslider=False)
figure.add_bollinger_bands(periods=20, boll_std=2, colors=['magenta','grey'],␣
 ↪fill=True)

figure.add_volume()
figure.add_macd()

figure.iplot(up_color = 'green', down_color = 'red')
```

[31]:
```
figure = cf.QuantFig(df, title= tk + ' - Another Quant Figures', legend='top',␣
 ↪name = tk)

figure.add_adx(color = 'purple')    #Plot Average Directional Index    (ADX)
#figure.add_cci(color = 'java')     #Plot Commodity Channel Indicator  (CCI)
figure.add_dmi()                    #Plot Directional Movement Index   (DMI)
figure.add_ema(color = 'magenta')   #Plot Exponencial Moving Average   (EMA)
figure.add_sma(color = 'java')      #Plot Simple Moving Average        (SMA)

figure.iplot(up_color = 'green', down_color = 'red')
```

[ ]:
```

```