

BTC vs All, last 180d.

4 Nov 2023

```
import yfinance as yf
import pandas as pd
import datetime

# Definisci i simboli di Bitcoin e S&P 500 su Yahoo Finance
symbols = ['BTC-USD', '^GSPC', 'GC=F', 'CL=F']

# Calcola la data di 90 giorni fa
end_date = datetime.date.today()
start_date = end_date - datetime.timedelta(days=180)

# Crea un DataFrame vuoto
df = pd.DataFrame()

# Scarica i dati e aggiungi al DataFrame
for symbol in symbols:
    data = yf.download(symbol, start=start_date, end=end_date)
    if df.empty:
        df = data[['Close']].rename(columns={'Close': symbol})
    else:
        df = df.join(data[['Close']].rename(columns={'Close': symbol}), how='inner')

# Rinomina le colonne
df.columns = ['BTC', 'S&P 500', 'Gold', 'Crude Oil']

# Stampa il DataFrame
print(df)
```

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          BTC      S&P 500      Gold Crude Oil
Date
2023-05-08  27694.273438  4138.120117  2026.300049  73.160004
2023-05-09  27658.775391  4119.169922  2036.199951  73.709999
2023-05-10  27621.755859  4137.640137  2030.500000  72.559998
2023-05-11  27000.789062  4130.620117  2014.699951  70.870003
2023-05-12  26804.990234  4124.080078  2014.500000  70.040001
...
2023-10-30  34502.363281  4166.819824  1996.199951  82.309998
2023-10-31  34667.781250  4193.799805  1985.199951  81.019997
2023-11-01  35437.253906  4237.859863  1978.800049  80.440002
2023-11-02  34938.242188  4317.779785  1985.599976  82.459999
2023-11-03  34732.324219  4358.339844  1991.500000  80.510002

[126 rows x 4 columns]
```

```
# Ordina df2 in ordine cronologico
df2 = df.sort_index(ascending=True)

# Normalizza i valori rispetto al primo valore di ogni colonna
df2 = df2 / df2.iloc[0]

# Visualizza il DataFrame normalizzato
print(df2)
```

```
          BTC      S&P 500      Gold Crude Oil
Date
2023-05-08  1.000000  1.000000  1.000000  1.000000
2023-05-09  0.998718  0.995421  1.004886  1.007518
2023-05-10  0.997381  0.999884  1.002073  0.991799
2023-05-11  0.974959  0.998188  0.994275  0.968699
2023-05-12  0.967889  0.996607  0.994177  0.957354
...
...
...
...
...
```

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2023-10-30 1.245830 1.006935 0.985145 1.125068
2023-10-31 1.251803 1.013455 0.979717 1.107436
2023-11-01 1.279588 1.024103 0.976558 1.099508
2023-11-02 1.261569 1.043416 0.979914 1.127119
2023-11-03 1.254134 1.053217 0.982826 1.100465

```

[126 rows x 4 columns]

```

import matplotlib.pyplot as plt

# Creare un grafico
plt.figure(figsize=(10,6))

# Plottare la colonna 'btc' in blu
plt.plot(df2['BTC'], color='blue', label='BTC')

# Plottare la colonna 'sp500' in verde
plt.plot(df2['S&P 500'], color='green', label='S&P 500')

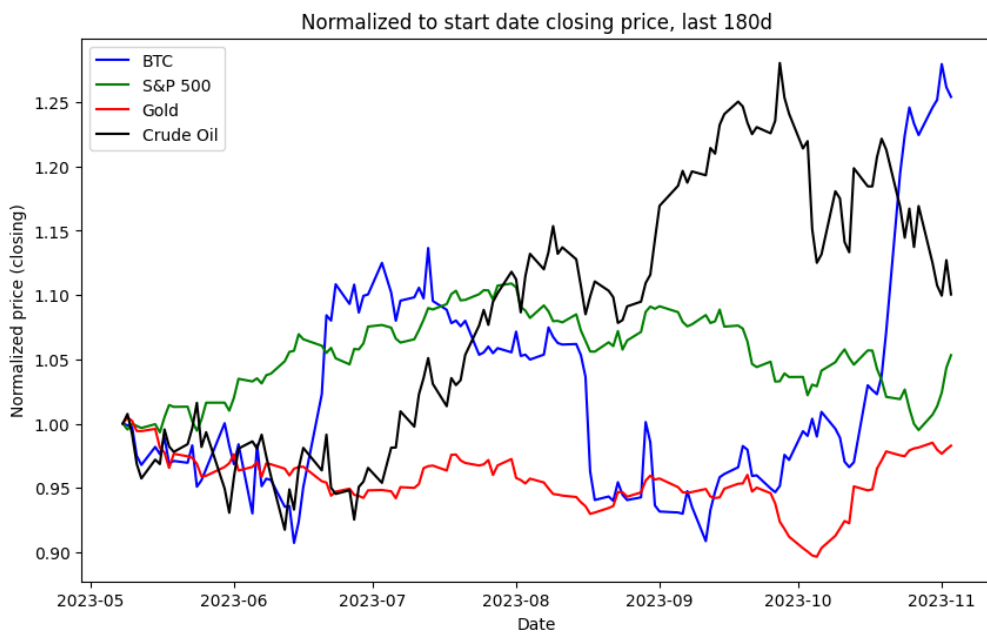
# Plottare la colonna 'gold' in rosso
plt.plot(df2['Gold'], color='red', label='Gold')

# Plottare la colonna 'oil' in nero
plt.plot(df2['Crude Oil'], color='black', label='Crude Oil')

# Aggiungere titoli e etichette
plt.title('Normalized to start date closing price, last 180d')
plt.xlabel('Date')
plt.ylabel('Normalized price (closing)')
plt.legend()

# Mostrare il grafico
plt.show()

```



```

ret = df2.pct_change().dropna()
print(ret)

```

```

          BTC  S&P 500  Gold  Crude Oil
Date
2023-05-09 -0.001282 -0.004579  0.004886  0.007518
2023-05-10 -0.001338  0.004484 -0.002799 -0.015602
2023-05-11 -0.022481 -0.001697 -0.007781 -0.023291
2023-05-12 -0.007252 -0.001583 -0.000099 -0.011712
2023-05-15  0.014464  0.002958  0.001737  0.015277
...
...
2023-10-30  0.017475  0.012010  0.003822 -0.037760
2023-10-31  0.004794  0.006475 -0.005510 -0.015672
2023-11-01  0.022196  0.010506 -0.003224 -0.007159
2023-11-02 -0.014082  0.018859  0.003436  0.025112
2023-11-03 -0.005894  0.009394  0.002971 -0.023648

```

[125 rows x 4 columns]

```
# Calcola la matrice di correlazione
correlation_matrix = ret.corr()

# Stampa la matrice di correlazione
print(correlation_matrix)
```

```
          BTC   S&P 500   Gold  Crude Oil
BTC      1.000000  0.150308  0.038880 -0.120946
S&P 500  0.150308  1.000000  0.099293 -0.051337
Gold     0.038880  0.099293  1.000000  0.213201
Crude Oil -0.120946 -0.051337  0.213201  1.000000
```

```
import seaborn as sns

# Crea un heatmap della matrice di correlazione
plt.figure(figsize=(10, 7))
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', linewidths=0.5)
plt.title("Daily return correlation heatmap")
plt.show()
```

