

# Analysis of my 5 favorite stocks

```
In [1]: # import Libraries

import pandas as pd
import yfinance as yf
from datetime import datetime

In [2]: # create start and end date

start_date = datetime.now() - pd.DateOffset(months = 6)
end_date = datetime.now()

In [5]: # MY FAVORITES :)

tickers = ['SPY', 'QQQ', 'AAPL', 'AMD', 'SHOP']

In [6]: # download data for the last 3 months from yfinance

df_list = [] # start with an empty List

for ticker in tickers: # iterates through each ticker symbol
    data = yf.download(ticker, start=start_date, end=end_date) # downloads historical stock data
    data['DailyRange'] = data['High'] - data['Low']
    df_list.append(data) # append List and data

df = pd.concat(df_list, keys=tickers, names=['Ticker', 'Date']) # concat List into df

#print
print(df.head())

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[*****100%*****] 1 of 1 completed
[*****100%*****] 1 of 1 completed
Open      High      Low      Close  Adj Close  \
Ticker Date
SPY  2023-06-27  432.350006  436.809998  431.880005  436.170013  432.881836
    2023-06-28  435.049988  437.440002  434.410004  436.390015  433.100189
    2023-06-29  435.959991  438.279999  435.540009  438.109985  434.807159
    2023-06-30  441.440002  444.299988  441.109985  443.279999  439.938202
    2023-07-03  442.920013  444.079987  442.630005  443.790009  440.444366

Volume  DailyRange
Ticker Date
SPY  2023-06-27  72813700  4.929993
    2023-06-28  75636000  3.029999
    2023-06-29  67882300  2.739990
    2023-06-30  104921500  3.190002
    2023-07-03  32793400  1.449982

In [7]: # reset index from Date col

df = df.reset_index()
print(df.head())

Ticker  Date      Open      High      Low      Close  \
0  SPY  2023-06-27  432.350006  436.809998  431.880005  436.170013
1  SPY  2023-06-28  435.049988  437.440002  434.410004  436.390015
2  SPY  2023-06-29  435.959991  438.279999  435.540009  438.109985
3  SPY  2023-06-30  441.440002  444.299988  441.109985  443.279999
4  SPY  2023-07-03  442.920013  444.079987  442.630005  443.790009

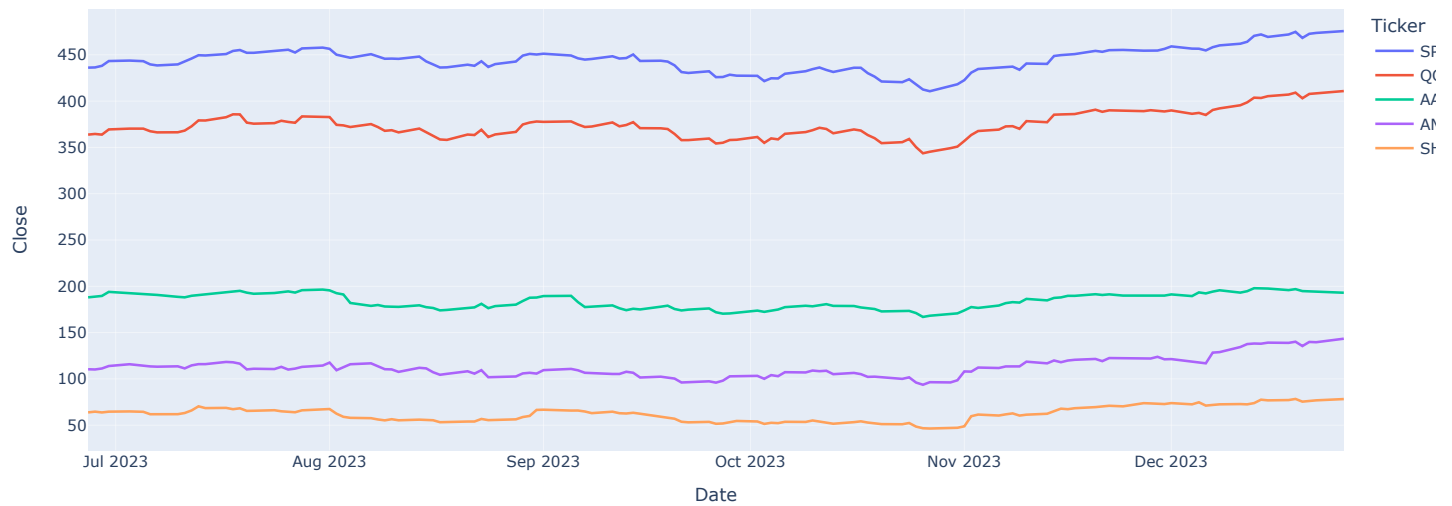
Adj Close  Volume  DailyRange
0  432.881836  72813700  4.929993
1  433.100189  75636000  3.029999
2  434.807159  67882300  2.739990
3  439.938202  104921500  3.190002
4  440.444366  32793400  1.449982

In [8]: # visualize stock performance

import plotly.express as px
fig = px.line(df, x='Date', y='Close', color='Ticker', title='Stock Market Performance (6M)')

fig.show()
```

## Stock Market Performance (6M)

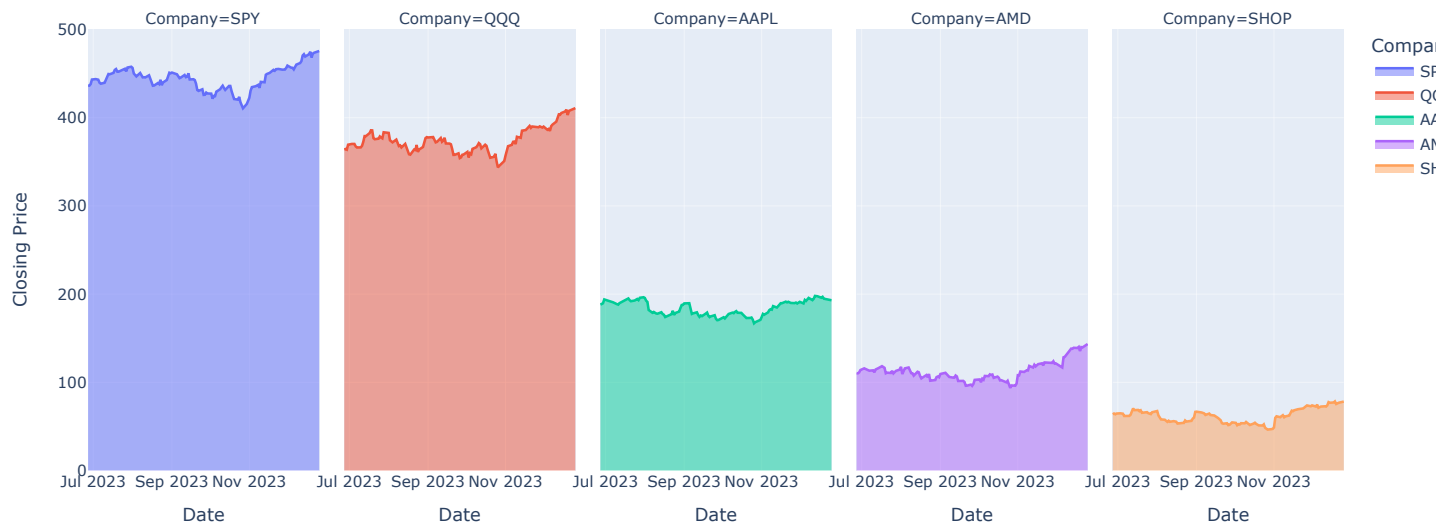


```
In [9]: # facet area chart

fig = px.area(df, x='Date', y='Close', color='Ticker',
              facet_col = 'Ticker',
              labels = {'Date': 'Date', 'Close': 'Closing Price', 'Ticker': 'Company'},
              title = 'Stock Prices (6M)')

fig.show()
```

## Stock Prices (6M)



## Moving Averages

```
In [10]: # create moving averages (8 & 21)

df['MA8'] = df.groupby('Ticker')['Close'].rolling(window=8).mean().reset_index(0, drop=True)
df['MA21'] = df.groupby('Ticker')['Close'].rolling(window=21).mean().reset_index(0, drop=True)

# analyze moving averages

for ticker, group in df.groupby('Ticker'):
    print(f'Moving Averages for {ticker}')
    print(group[['MA8', 'MA21']])
```

Moving Averages for AAPL

	MA8	MA21
254	NaN	NaN
255	NaN	NaN
256	NaN	NaN
257	NaN	NaN
258	NaN	NaN
..	...	...
376	196.258753	193.030000
377	196.148752	193.190953
378	196.336252	193.383333
379	196.197502	193.492382
380	195.583752	193.639048

[127 rows x 2 columns]  
Moving Averages for AMD

	MA8	MA21
381	NaN	NaN
382	NaN	NaN
383	NaN	NaN
384	NaN	NaN
385	NaN	NaN
..	...	...
503	136.916248	127.335237
504	137.734999	127.999047
505	138.422499	128.987142
506	138.671249	129.800952
507	139.323750	130.805714

[127 rows x 2 columns]  
Moving Averages for QQQ

	MA8	MA21
127	NaN	NaN
128	NaN	NaN
129	NaN	NaN
130	NaN	NaN
131	NaN	NaN
..	...	...
249	401.883751	393.851429
250	403.247498	394.439048
251	404.778748	395.358095
252	405.992496	396.230476
253	406.884998	397.248095

[127 rows x 2 columns]  
Moving Averages for SHOP

	MA8	MA21
508	NaN	NaN
509	NaN	NaN
510	NaN	NaN
511	NaN	NaN
512	NaN	NaN
..	...	...
630	75.191249	73.315714
631	75.553749	73.597143
632	75.969999	73.902380
633	76.511250	74.181904
634	77.054999	74.556190

[127 rows x 2 columns]  
Moving Averages for SPY

	MA8	MA21
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
..	...	...
122	468.117500	460.409047
123	469.125000	461.075714
124	470.463753	462.000953
125	471.657501	462.888096
126	472.301250	463.857144

[127 rows x 2 columns]

```
In [11]: # visualize moving averages

for ticker, group in df.groupby('Ticker'):
    fig = px.line(group, x='Date', y=['Close', 'MA8', 'MA21'],
                  title= f'{ticker} Moving Averages')
    fig.show()
```

AAPL Moving Averages



AMD Moving Averages



QQQ Moving Averages



## SHOP Moving Averages



## SPY Moving Averages

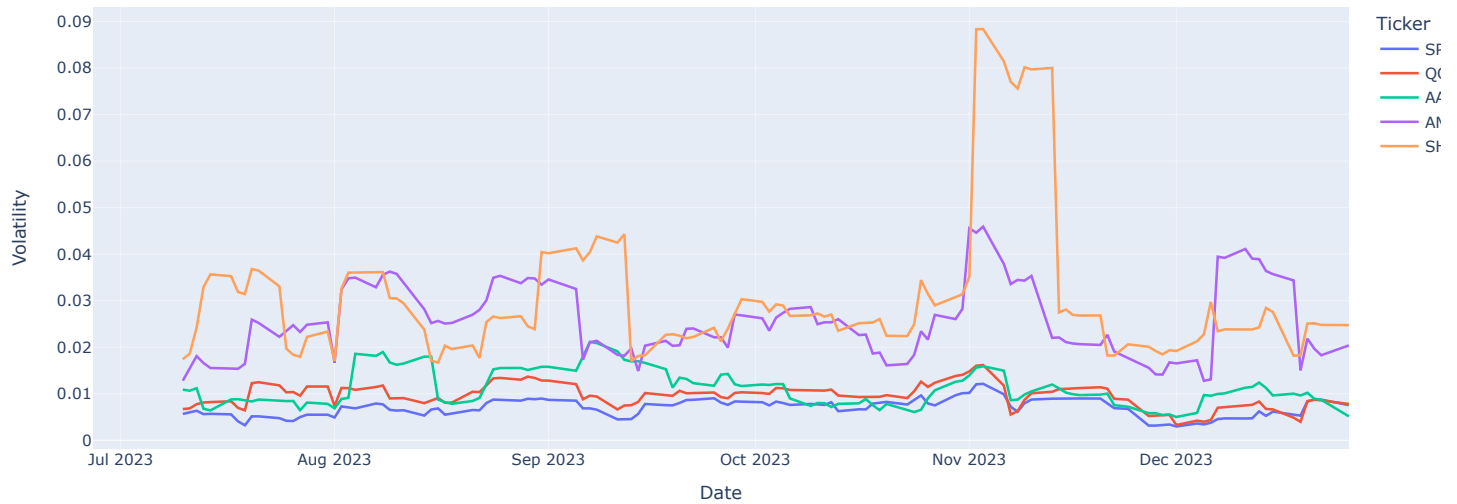


## Volatility

```
In [13]: # create volatility df
df['Volatility'] = df.groupby('Ticker')['Close'].pct_change().rolling(window=8).std().reset_index(0, drop=True)

#visualize
fig = px.line(df, x='Date', y='Volatility', color='Ticker', title='Volatility of My Top 5')
fig.show()
```

## Volatility of My Top 5



## Correlation

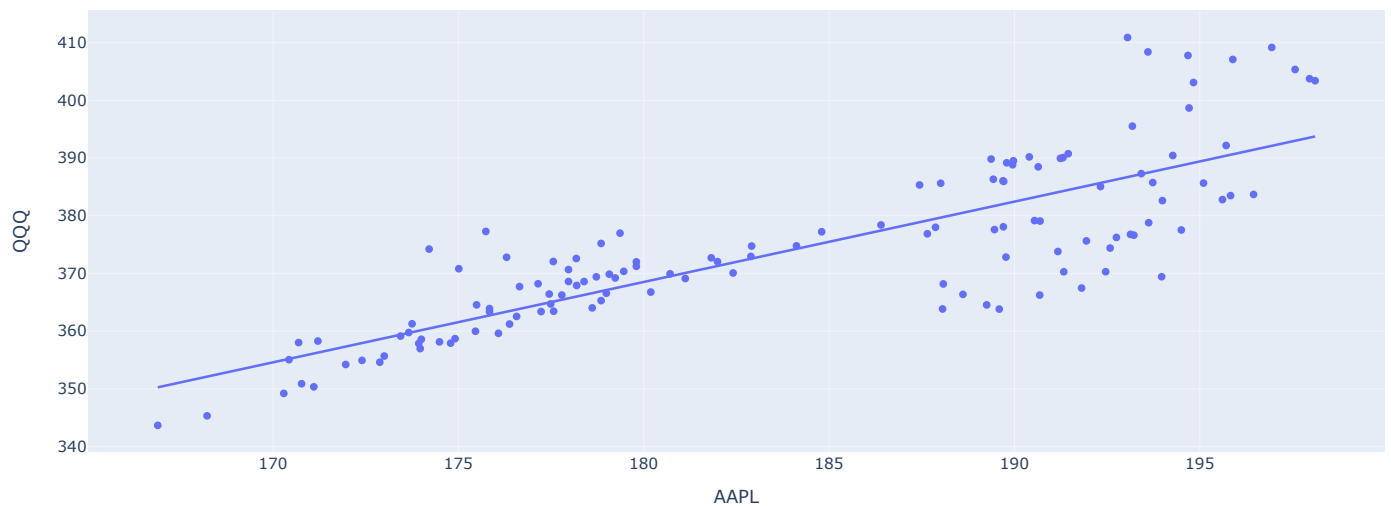
```
In [14]: import statsmodels.api as sm
```

```
In [15]: # create dataframe with AAPL & QQQ
apple = df.loc[df['Ticker'] == 'AAPL', ['Date', 'Close']].rename(columns={'Close': 'AAPL'})
qqq = df.loc[df['Ticker'] == 'QQQ', ['Date', 'Close']].rename(columns={'Close': 'QQQ'})
df_corr = pd.merge(apple, qqq, on='Date')

#create scatterplot to visualize

fig = px.scatter(df_corr, x='AAPL', y='QQQ', trendline='ols', title='Correlation between AAPL & QQQ')
fig.show()
```

### Correlation between AAPL & QQQ



```
In [16]: # create dataframe with AMD & QQQ
amd = df.loc[df['Ticker'] == 'AMD', ['Date', 'Close']].rename(columns={'Close': 'AMD'})
qqq = df.loc[df['Ticker'] == 'QQQ', ['Date', 'Close']].rename(columns={'Close': 'QQQ'})
df_corr = pd.merge(amd, qqq, on='Date')

#create scatterplot to visualize

fig = px.scatter(df_corr, x='AMD', y='QQQ', trendline='ols', title='Correlation between AMD & QQQ')
fig.show()
```

Correlation between AMD & QQQ

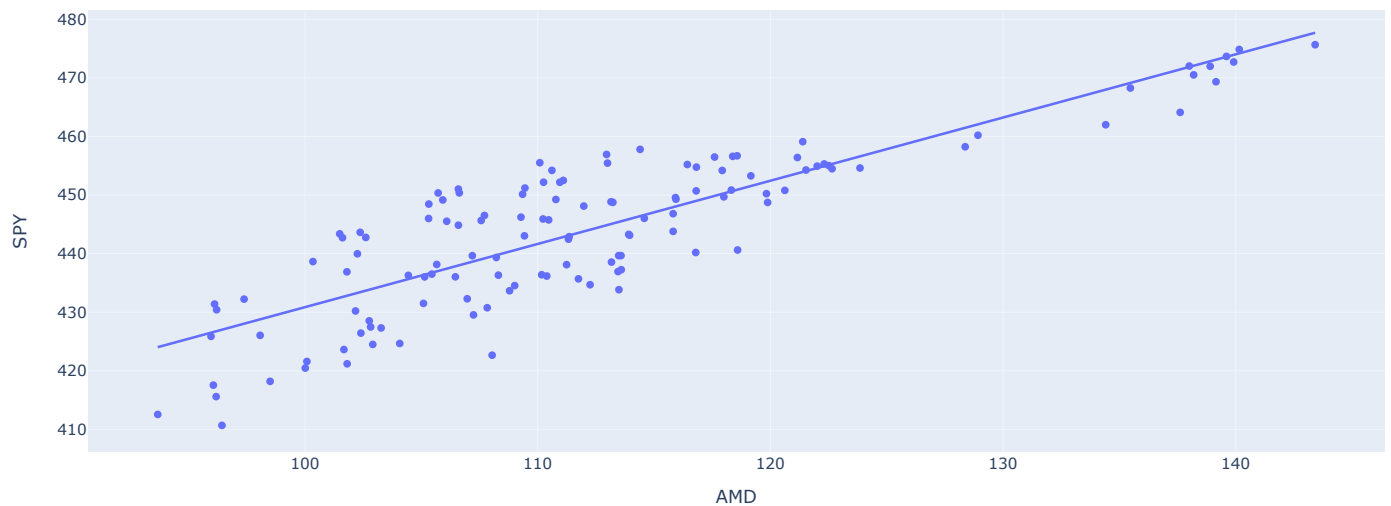


```
In [17]: # create dataframe with AMD & SPY
amd = df.loc[df['Ticker'] == 'AMD', ['Date', 'Close']].rename(columns={'Close': 'AMD'})
spy = df.loc[df['Ticker'] == 'SPY', ['Date', 'Close']].rename(columns={'Close': 'SPY'})
df_corr = pd.merge(amd, spy, on= 'Date')

#create scatterplot to visualize

fig = px.scatter(df_corr, x='AMD', y='SPY', trendline='ols', title='Correlation between AMD & SPY')
fig.show()
```

Correlation between AMD & SPY

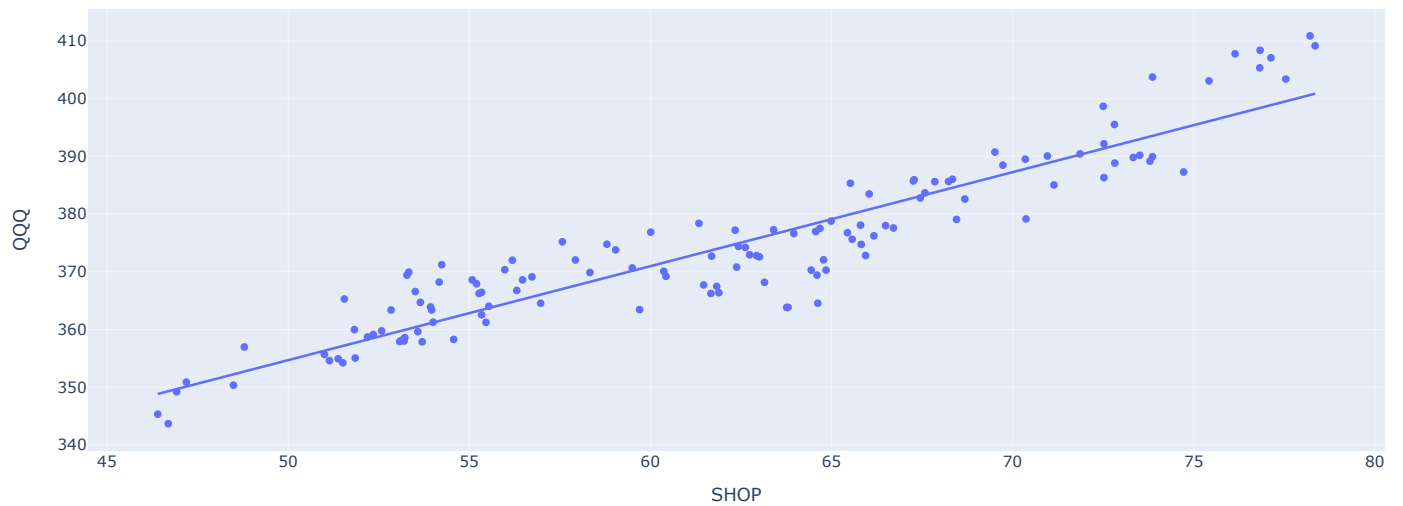


```
In [18]: # create dataframe with SHOP & QQQ
shop = df.loc[df['Ticker'] == 'SHOP', ['Date', 'Close']].rename(columns={'Close': 'SHOP'})
qqq = df.loc[df['Ticker'] == 'QQQ', ['Date', 'Close']].rename(columns={'Close': 'QQQ'})
df_corr = pd.merge(shop, qqq, on= 'Date')

#create scatterplot to visualize

fig = px.scatter(df_corr, x='SHOP', y='QQQ', trendline='ols', title='Correlation between SHOP & QQQ')
fig.show()
```

Correlation between SHOP & QQQ

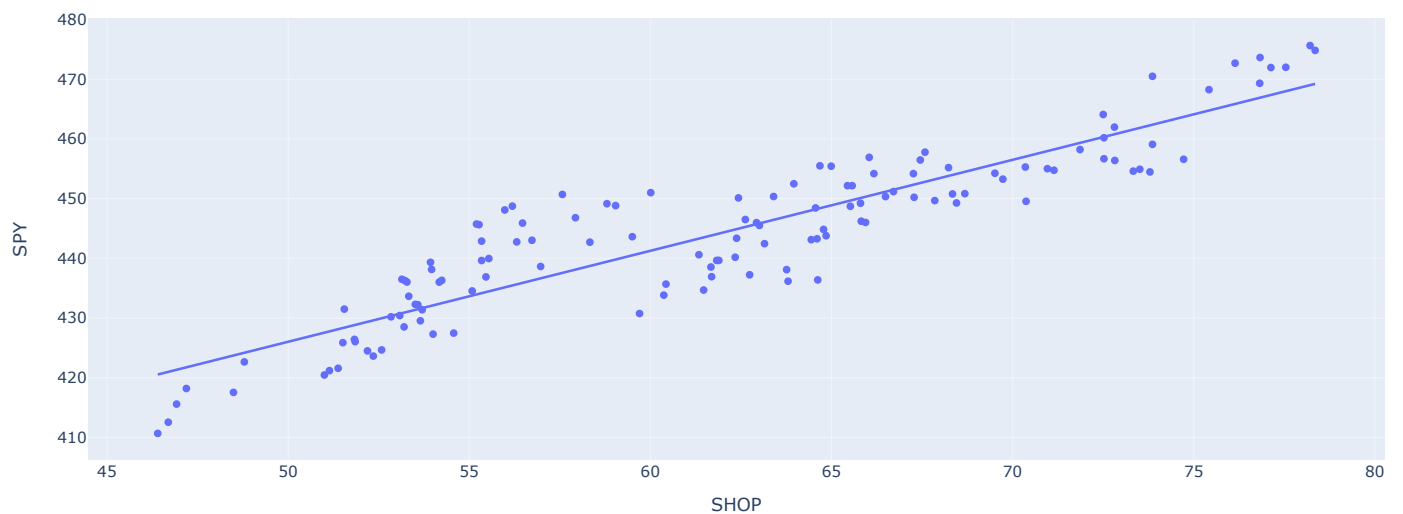


```
In [19]: # create dataframe with SHOP & SPY
shop = df.loc[df['Ticker'] == 'SHOP', ['Date', 'Close']].rename(columns={'Close': 'SHOP'})
spy = df.loc[df['Ticker'] == 'SPY', ['Date', 'Close']].rename(columns={'Close': 'SPY'})
df_corr = pd.merge(shop, spy, on= 'Date')

#create scatterplot to visualize

fig = px.scatter(df_corr, x='SHOP', y='SPY', trendline='ols', title='Correlation between SHOP & SPY')
fig.show()
```

Correlation between SHOP & SPY



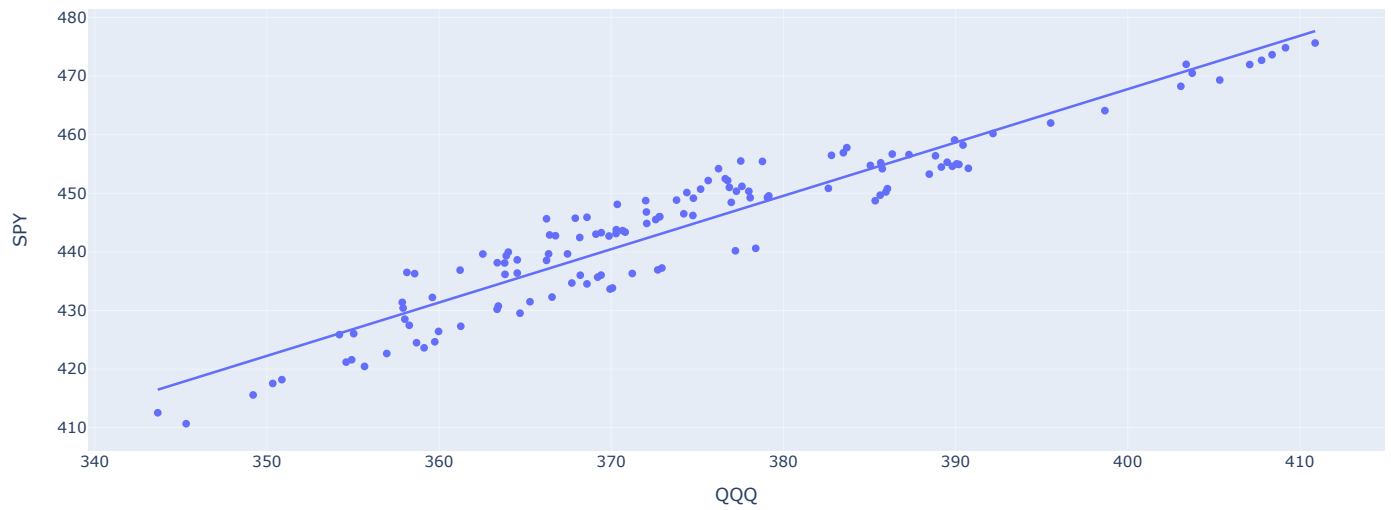
```
In [20]: # create dataframe with QQQ & SPY
qqq = df.loc[df['Ticker'] == 'QQQ', ['Date', 'Close']].rename(columns={'Close': 'QQQ'})
spy = df.loc[df['Ticker'] == 'SPY', ['Date', 'Close']].rename(columns={'Close': 'SPY'})
df_corr = pd.merge(qqq, spy, on= 'Date')

#create scatterplot to visualize

fig = px.scatter(df_corr, x='QQQ', y='SPY', trendline='ols', title='Correlation between QQQ & SPY')
fig.show()
```



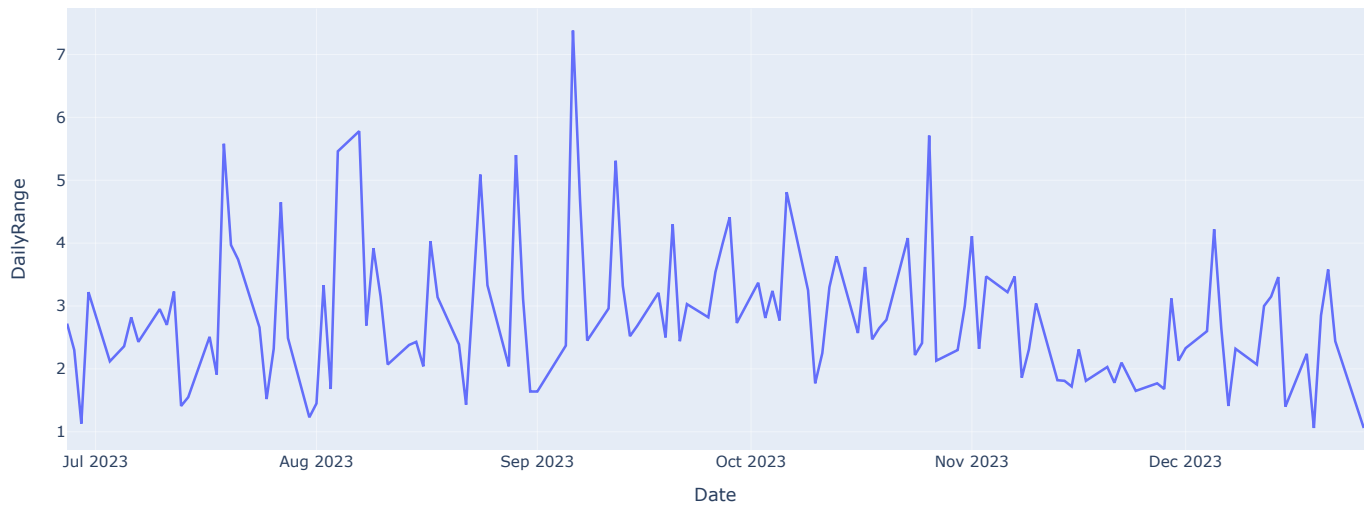
Correlation between QQQ & SPY



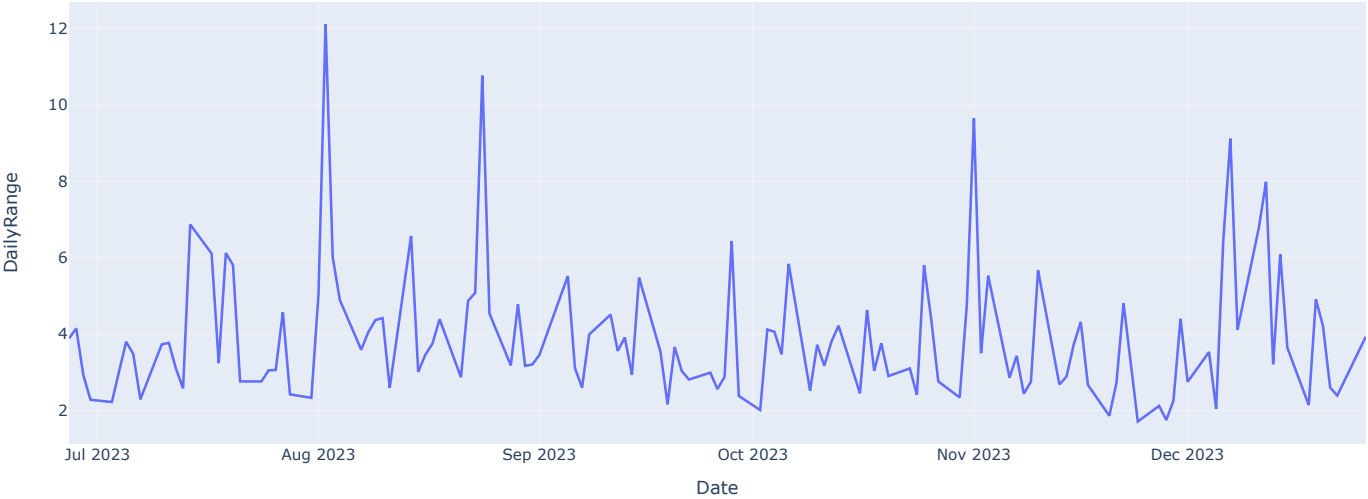
## Daily Range

```
In [21]: for ticker, group in df.groupby('Ticker'):
# reset index to include data column
group.reset_index(inplace=True)
#plot daily range for each stock
fig = px.line(group, x='Date', y='DailyRange',
              title= f'{ticker} Daily Range')
fig.show()
```

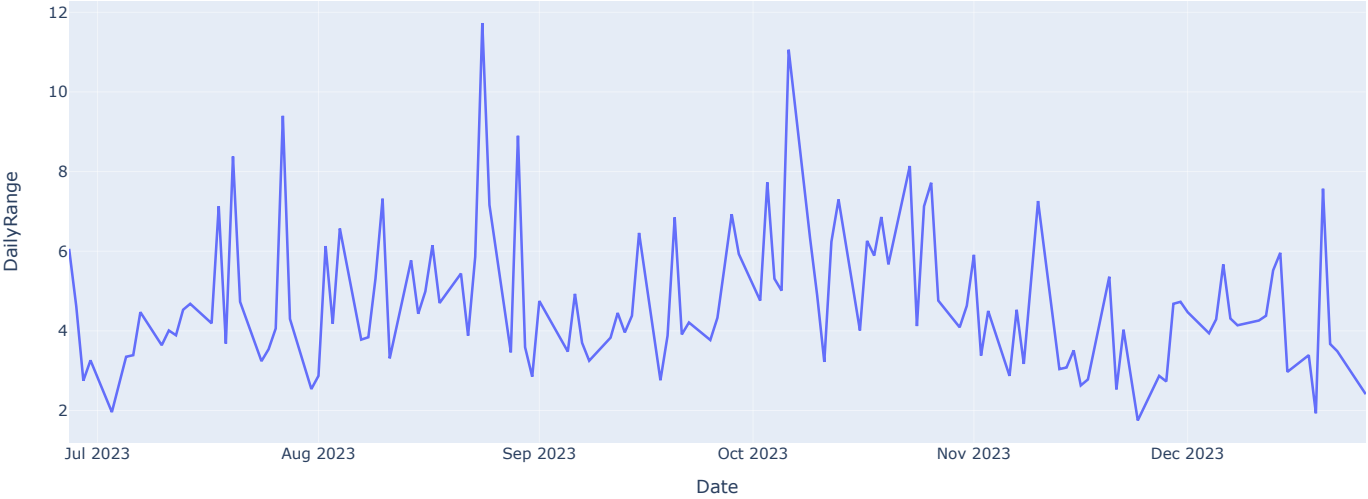
AAPL Daily Range



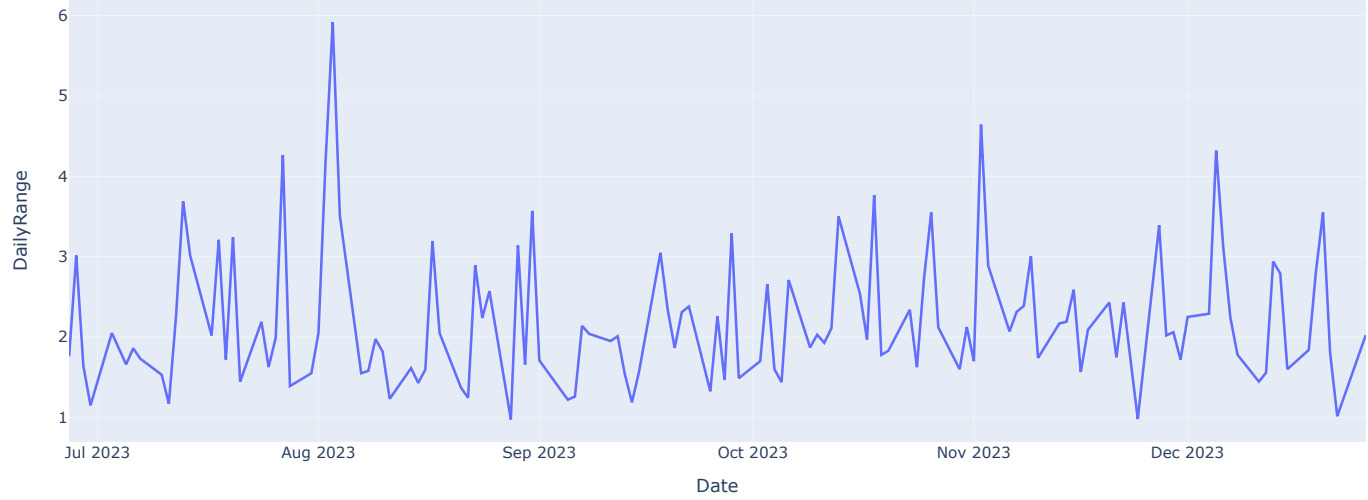
AMD Daily Range



QQQ Daily Range



SHOP Daily Range



SPY Daily Range

