financial-market-analysis

June 27, 2024

[1]: !pip install yfinance

```
Requirement already satisfied: yfinance in c:\users\zeine\anaconda3\lib\site-
packages (0.2.40)
Requirement already satisfied: pandas>=1.3.0 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (2.1.4)
Requirement already satisfied: numpy>=1.16.5 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (1.26.4)
Requirement already satisfied: requests>=2.31 in
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c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (0.0.11)
Requirement already satisfied: lxml>=4.9.1 in c:\users\zeine\anaconda3\lib\site-
packages (from yfinance) (4.9.3)
Requirement already satisfied: platformdirs>=2.0.0 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (3.10.0)
Requirement already satisfied: pytz>=2022.5 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (2023.3.post1)
Requirement already satisfied: frozendict>=2.3.4 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (2.4.2)
Requirement already satisfied: peewee>=3.16.2 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (3.17.5)
Requirement already satisfied: beautifulsoup4>=4.11.1 in
c:\users\zeine\anaconda3\lib\site-packages (from yfinance) (4.12.2)
Requirement already satisfied: html5lib>=1.1 in
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beautifulsoup4>=4.11.1->yfinance) (2.5)
Requirement already satisfied: six>=1.9 in c:\users\zeine\anaconda3\lib\site-
packages (from html5lib>=1.1->yfinance) (1.16.0)
Requirement already satisfied: webencodings in
c:\users\zeine\anaconda3\lib\site-packages (from html5lib>=1.1->yfinance)
(0.5.1)
Requirement already satisfied: python-dateutil>=2.8.2 in
c:\users\zeine\anaconda3\lib\site-packages (from pandas>=1.3.0->yfinance)
(2.8.2)
Requirement already satisfied: tzdata>=2022.1 in
```

```
c:\users\zeine\anaconda3\lib\site-packages (from pandas>=1.3.0->yfinance)
(2023.3)
Requirement already satisfied: charset-normalizer<4,>=2 in
c:\users\zeine\anaconda3\lib\site-packages (from requests>=2.31->yfinance)
(2.0.4)
Requirement already satisfied: idna<4,>=2.5 in
c:\users\zeine\anaconda3\lib\site-packages (from requests>=2.31->yfinance) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in
c:\users\zeine\anaconda3\lib\site-packages (from requests>=2.31->yfinance)
(2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in
c:\users\zeine\anaconda3\lib\site-packages (from requests>=2.31->yfinance)
(2024.6.2)
```

[3]: !pip install quantstats

```
Requirement already satisfied: quantstats in c:\users\zeine\anaconda3\lib\site-
packages (0.0.62)
Requirement already satisfied: pandas>=0.24.0 in
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Requirement already satisfied: numpy>=1.16.5 in
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Requirement already satisfied: seaborn>=0.9.0 in
c:\users\zeine\anaconda3\lib\site-packages (from quantstats) (0.12.2)
Requirement already satisfied: matplotlib>=3.0.0 in
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Requirement already satisfied: scipy>=1.2.0 in
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Requirement already satisfied: tabulate>=0.8.0 in
c:\users\zeine\anaconda3\lib\site-packages (from quantstats) (0.9.0)
Requirement already satisfied: yfinance>=0.1.70 in
c:\users\zeine\anaconda3\lib\site-packages (from quantstats) (0.2.40)
Requirement already satisfied: python-dateutil>=2.0 in
c:\users\zeine\anaconda3\lib\site-packages (from quantstats) (2.8.2)
Requirement already satisfied: contourpy>=1.0.1 in
c:\users\zeine\anaconda3\lib\site-packages (from matplotlib>=3.0.0->quantstats)
Requirement already satisfied: cycler>=0.10 in
c:\users\zeine\anaconda3\lib\site-packages (from matplotlib>=3.0.0->quantstats)
Requirement already satisfied: fonttools>=4.22.0 in
c:\users\zeine\anaconda3\lib\site-packages (from matplotlib>=3.0.0->quantstats)
Requirement already satisfied: kiwisolver>=1.0.1 in
c:\users\zeine\anaconda3\lib\site-packages (from matplotlib>=3.0.0->quantstats)
Requirement already satisfied: packaging>=20.0 in
c:\users\zeine\anaconda3\lib\site-packages (from matplotlib>=3.0.0->quantstats)
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(23.1)
Requirement already satisfied: pillow>=6.2.0 in
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Requirement already satisfied: pyparsing>=2.3.1 in
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Requirement already satisfied: pytz>=2020.1 in
c:\users\zeine\anaconda3\lib\site-packages (from pandas>=0.24.0->quantstats)
(2023.3.post1)
Requirement already satisfied: tzdata>=2022.1 in
c:\users\zeine\anaconda3\lib\site-packages (from pandas>=0.24.0->quantstats)
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Requirement already satisfied: frozendict>=2.3.4 in
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Requirement already satisfied: beautifulsoup4>=4.11.1 in
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Requirement already satisfied: html5lib>=1.1 in
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Requirement already satisfied: soupsieve>1.2 in
c:\users\zeine\anaconda3\lib\site-packages (from
beautifulsoup4>=4.11.1->yfinance>=0.1.70->quantstats) (2.5)
Requirement already satisfied: webencodings in
c:\users\zeine\anaconda3\lib\site-packages (from
html5lib >= 1.1 - yfinance >= 0.1.70 - yquantstats) (0.5.1)
Requirement already satisfied: charset-normalizer<4,>=2 in
c:\users\zeine\anaconda3\lib\site-packages (from
requests>=2.31->yfinance>=0.1.70->quantstats) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in
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c:\users\zeine\anaconda3\lib\site-packages (from
     requests>=2.31->yfinance>=0.1.70->quantstats) (3.4)
     Requirement already satisfied: urllib3<3,>=1.21.1 in
     c:\users\zeine\anaconda3\lib\site-packages (from
     requests>=2.31->yfinance>=0.1.70->quantstats) (2.0.7)
     Requirement already satisfied: certifi>=2017.4.17 in
     c:\users\zeine\anaconda3\lib\site-packages (from
     requests>=2.31->yfinance>=0.1.70->quantstats) (2024.6.2)
[80]: !pip install ta
     Collecting ta
       Downloading ta-0.11.0.tar.gz (25 kB)
       Preparing metadata (setup.py): started
       Preparing metadata (setup.py): finished with status 'done'
     Requirement already satisfied: numpy in c:\users\zeine\anaconda3\lib\site-
     packages (from ta) (1.26.4)
     Requirement already satisfied: pandas in c:\users\zeine\anaconda3\lib\site-
     packages (from ta) (2.1.4)
     Requirement already satisfied: python-dateutil>=2.8.2 in
     c:\users\zeine\anaconda3\lib\site-packages (from pandas->ta) (2.8.2)
     Requirement already satisfied: pytz>=2020.1 in
     c:\users\zeine\anaconda3\lib\site-packages (from pandas->ta) (2023.3.post1)
     Requirement already satisfied: tzdata>=2022.1 in
     c:\users\zeine\anaconda3\lib\site-packages (from pandas->ta) (2023.3)
     Requirement already satisfied: six>=1.5 in c:\users\zeine\anaconda3\lib\site-
     packages (from python-dateutil>=2.8.2->pandas->ta) (1.16.0)
     Building wheels for collected packages: ta
       Building wheel for ta (setup.py): started
       Building wheel for ta (setup.py): finished with status 'done'
       Created wheel for ta: filename=ta-0.11.0-py3-none-any.whl size=29420
     \verb|sha| 256 = \verb|fe| 1737b 15e 7c51553b e 508d 12c 2cb 236a6c 7fc 3de 91a6b 587e 8cc 469699cd 2c9 \\
       Stored in directory: c:\users\zeine\appdata\local\pip\cache\wheels\a1\d7\29\77
     81cc5eb9a3659d032d7d15bdd0f49d07d2b24fec29f44bc4
     Successfully built ta
     Installing collected packages: ta
     Successfully installed ta-0.11.0
[86]: # Importing Libraries
      # Data handling and statistical analysis
      import pandas as pd
      from pandas_datareader import data
      import numpy as np
      from scipy import stats
      # Data visualization
```

```
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
import plotly.graph_objects as go
from plotly.subplots import make_subplots
# Financial data
import quantstats as qs
import ta
import yfinance as yf
# Linear Regression Model
from sklearn.linear_model import LinearRegression
# Enabling Plotly offline
from plotly.offline import init_notebook_mode
init_notebook_mode(connected=True)
# Datetime and hiding warnings
import datetime as dt
import warnings
warnings.filterwarnings("ignore")
```

1 Daily Returns for Stocks

Definition: Daily returns measure the percentage change in the value of an investment from one day to the next. For example, if a stock's price increases from 100 to 102 in one day, the daily return is 2%.

```
[98]: azn = qs.utils.download_returns('AZN')
azn = azn.loc['2010-07-01':'2024-06-20']

bp = qs.utils.download_returns('BP')
bp = bp.loc['2010-07-01':'2024-06-20']

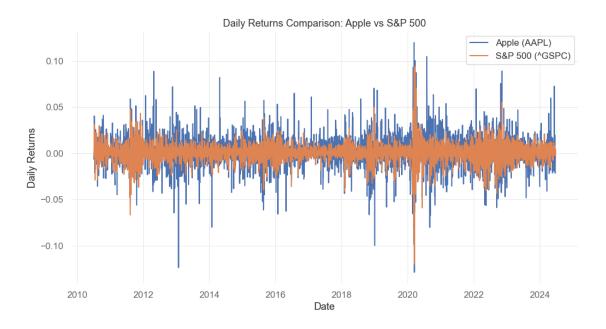
aapl = qs.utils.download_returns('AAPL')
aapl = aapl.loc['2010-07-01':'2024-06-20']

msft = qs.utils.download_returns('MSFT')
msft = msft.loc['2010-07-01':'2024-06-20']
```

```
[100]: try:
           azn.index = azn.index.tz_localize('UTC').tz_convert(None)
           bp.index = bp.index.tz_localize('UTC').tz_convert(None)
           aapl.index = aapl.index.tz_localize('UTC').tz_convert(None)
           msft.index = msft.index.tz_localize('UTC').tz_convert(None)
       except Exception as e:
           print(f"An error occurred during timezone localization and conversion: {e}")
[96]: # Function to download daily returns
       def download_daily_returns(ticker):
           data = yf.download(ticker, start='2010-07-01', end='2024-06-26')
           returns = data['Adj Close'].pct_change().dropna()
           returns.index = returns.index.tz_localize('UTC')
           return returns
       # Downloading Apple (AAPL) and S&P 500 (~GSPC) daily returns
       aapl = download_daily_returns('AAPL')
       sp500 = download_daily_returns('^GSPC')
       # Plotting daily returns For Apple
       print('\n')
       print('\nApple Daily Returns Plot:\n')
       plt.figure(figsize=(12, 6))
       plt.plot(aapl.index, aapl, label='Apple (AAPL)')
       plt.plot(sp500.index, sp500, label='S&P 500 (^GSPC)')
       plt.title('Daily Returns Comparison: Apple vs S&P 500')
       plt.xlabel('Date')
       plt.ylabel('Daily Returns')
       plt.legend()
       plt.grid(True)
       plt.show()
```

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

Apple Daily Returns Plot:

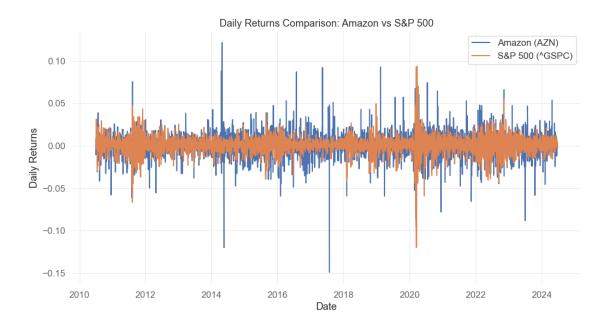


1.1 Apple analysis:

- Huge volatility in 2013 and 2020 (due to COVID)
- In general stocks seem stable and a predictable investment option

```
[110]: # Plotting daily returns for Amazon
    print('\n')
    print('\nAmazon Daily Returns Plot:\n')
    plt.figure(figsize=(12,6))
    plt.plot(azn.index, azn, label='Amazon (AZN)')
    plt.plot(sp500.index, sp500, label='S&P 500 (^GSPC)')
    plt.title('Daily Returns Comparison: Amazon vs S&P 500')
    plt.xlabel('Date')
    plt.ylabel('Daily Returns')
    plt.legend()
    plt.grid(True)
    plt.show()
```

Amazon Daily Returns Plot:

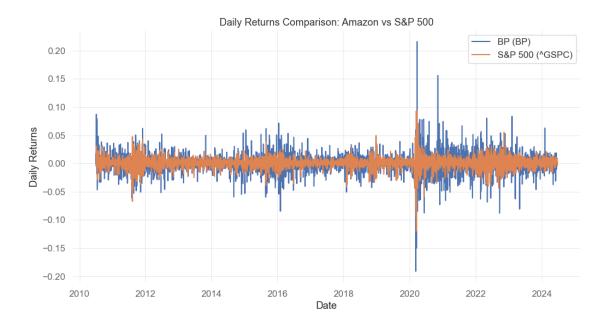


1.2 Amazon analysis:

- Unusual variation seen in late 2014/early 2015 and late 2016/early2017
- This indivates higher volatility suggesting it could be a riskier investment

```
[120]: # Plotting daily returns for Amazon
    print('\n')
    print('\nBP Daily Returns Plot:\n')
    plt.figure(figsize=(12,6))
    plt.plot(bp.index, bp, label= 'BP (BP)')
    plt.plot(sp500.index, sp500, label='S&P 500 (^GSPC)')
    plt.title('Daily Returns Comparison: Amazon vs S&P 500')
    plt.xlabel('Date')
    plt.ylabel('Daily Returns')
    plt.legend()
    plt.grid(True)
    plt.show()
```

BP Daily Returns Plot:

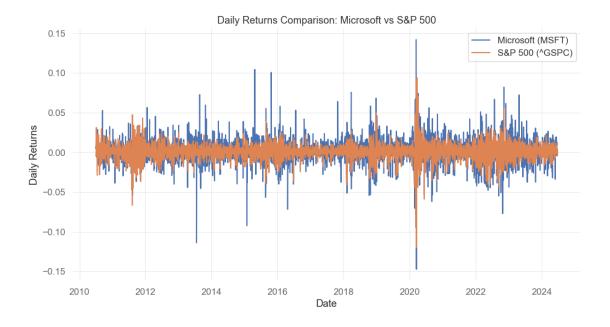


1.3 BP analysis:

- As expected, volatility seen in 2020
- In geenral the stocks seem stable and a predictable investment option

```
[126]: # Plotting daily returns for Microsoft
print('\n')
print('\nMicrosoft Daily Returns Plot:\n')
plt.figure(figsize=(12,6))
plt.plot(msft.index, msft, label = 'Microsoft (MSFT)')
plt.plot(sp500.index, sp500, label = 'S&P 500 (^GSPC)')
plt.title('Daily Returns Comparison: Microsoft vs S&P 500')
plt.xlabel('Date')
plt.ylabel('Daily Returns')
plt.grid(True)
plt.legend()
plt.show()
```

Microsoft Daily Returns Plot:



1.4 Microsoft analysis:

- As expected, volatility seen in 2020- but bigger than usual
- IUnusal volatility seen around 2013, and 2015-2016 year.
- This might be a riskier investment

2 Cumulative Returns

Definition: Cumulative return is the total change in the value of an investment over a specified period, expressed as a percentage of the initial value. For example, if you invest 1,000 in a stock and its value grows to 1,500 over three years, the cumulative return is 50%.

```
[160]: # print('\n')
    print('\nApple Cumulative Returns Plot\n')
    qs.plots.returns(aapl)
    print('\n')
    print('\n')
    print('\nAmazon Cumulative Returns Plot\n')
    qs.plots.returns(azn)
    print('\n')
    print('\n')
    print('\n')
    print('\nBP Cumulative Returns Plot\n')
    qs.plots.returns(bp)
    print('\n')
    qs.plots.returns(msft)
```

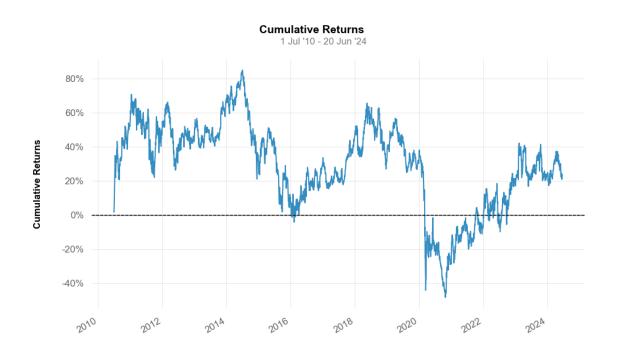
Apple Cumulative Returns Plot



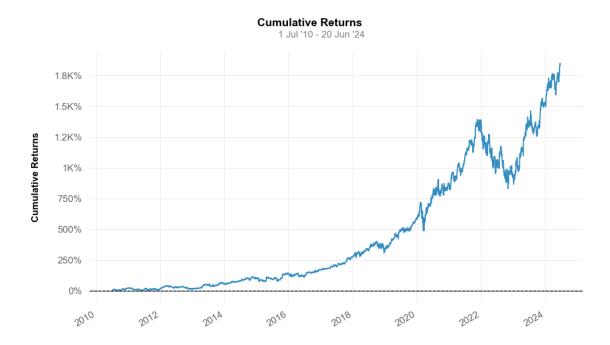
Amazon Cumulative Returns Plot



BP Cumulative Returns Plot



Microsoft Cumulative Returns Plot



2.1 Cumulative Returns Analysis

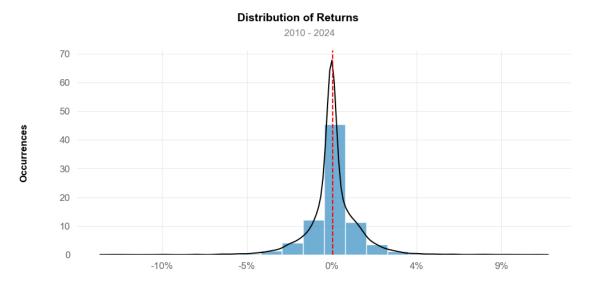
- Apple has the highest percentage of Cumulative return peak- surpassing 2k in 2024 as opposed to the lowest peak of above 80% in 2015
- Stark difference in negative cumulative return between BP and all the other stocks- BP cumulative returns was at an alaraming low percentage, exceeding -40% in 2020-2021
- However, in 2024 it has now increased back to just above 40%
- BP seems to be yield extreme volatile cumulative returns
- Microsoft also yielded good cumulative returns, exceeding 1.8k in 2024

3 Histogram analysis

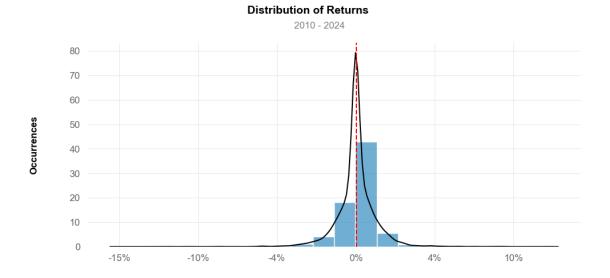
```
[164]: # Plotting histograms for daily returns
print('\n')
print('\nApple Daily Returns Histogram')
```

```
qs.plots.histogram(aapl, resample = 'D')
print('\n')
print('\nAmazon Daily Returns Histogram')
qs.plots.histogram(azn, resample = 'D')
print('\n')
print('\nBP Daily Returns Histogram')
qs.plots.histogram(bp, resample = 'D')
print('\n')
print('\n')
print('\nMicrosoft Daily Returns Histogram')
qs.plots.histogram(msft, resample = 'D')
```

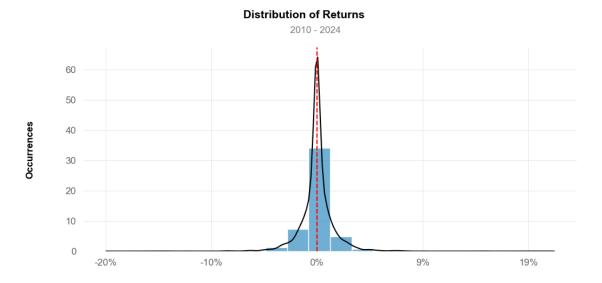
Apple Daily Returns Histogram



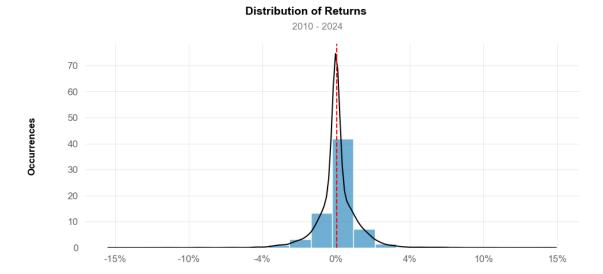
Amazon Daily Returns Histogram



BP Daily Returns Histogram



Microsoft Daily Returns Histogram



4 Histogram Analysis

- $\bullet\,$ Most Daily returns are close to 0
- Extreme values seem to be consistent across stocks- never exceeding -20% and 20%
- All stocks seem to have balaced retrns- values on extreme sides being closer to eachother

4.1 Kurtosis

```
[172]: print('\n')
    print("Apple's kurtosis: ", qs.stats.kurtosis(aapl).round(2))
    print('\n')
    print("Amazon's kurtosis: ", qs.stats.kurtosis(azn).round(2))
    print('\n')
    print("BP's kurtosis: ", qs.stats.kurtosis(bp).round(3))
    print('\n')
    print("Microsoft's kurtosis: ", qs.stats.kurtosis(msft).round(3))
```

Apple's kurtosis: 5.35

Amazon's kurtosis: 9.71

BP's kurtosis: 14.151

Microsoft's kurtosis: 7.807

4.2 Analysis:

- All four stocks have positive and high kurtosis- indicating that there is a high concentration of daily return observations in the tails of their distribution
- This suggests all four stocks can be subject to high levels of volatility and therefore risk-there can be a good amount of price fluctuations which deviate significantly from their average returns
- BP has thew highest kurtosis level- meaning it is subjected to high levels of volatility and risk (as seen from the other above ghraphs)

4.3 Skewness

```
[176]: print('\n')
    print("Apple's skewness: ", qs.stats.skew(aapl).round(2))
    print('\n')
    print("Amazon's skewness: ", qs.stats.skew(azn).round(2))
    print('\n')
    print("BP's skewness: ", qs.stats.skew(bp).round(3))
    print('\n')
    print("Microsoft's skewness: ", qs.stats.skew(msft).round(3))
```

```
Amazon's skewness: -0.23
BP's skewness: 0.053
```

Apple's skewness:

Microsoft's skewness: 0.054

• Generally slight level of skewness across all four stocks - confirming insights from Histogram

4.4 Correlation Matrix and Pairplots

-0.04

```
[193]: df = pd.concat([aapl, azn, bp, msft], join = 'outer', axis = 1)
df.columns = ['Apple', 'Amazon', 'BP', 'Microsoft']
df
```

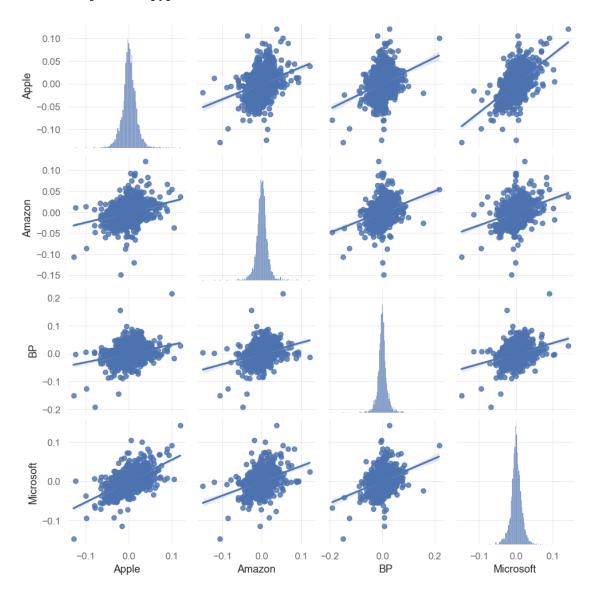
```
[193]:
                                    Apple
                                                               Microsoft
                                             Amazon
      Date
      2010-07-02 00:00:00+00:00 -0.006198 -0.010515 -0.001361
                                                                0.004750
      2010-07-06 00:00:00+00:00 0.006844 0.015303 0.087223
                                                                0.023635
      2010-07-07 00:00:00+00:00 0.040381
                                           0.011723
                                                     0.040113
                                                                0.020151
      2010-07-08 00:00:00+00:00 -0.002242 0.009104 0.016571
                                                                0.004527
      2010-07-09 00:00:00+00:00 0.005928 0.004716 0.009188
                                                              -0.005735
```

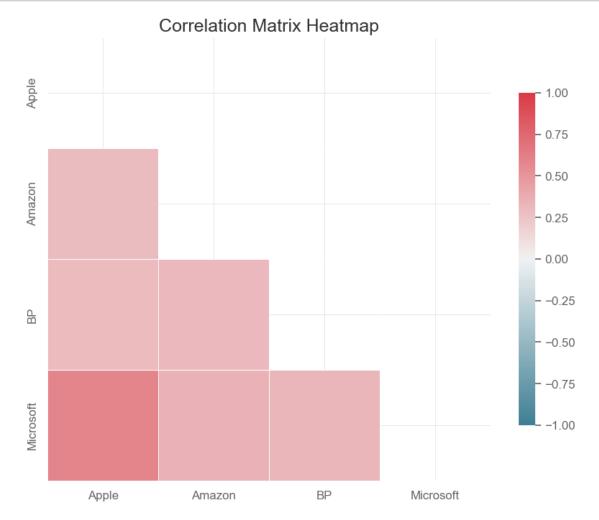
```
2024-06-18 00:00:00+00:00 -0.010984 -0.011222
                                                         -0.004528
                                               0.007975
2024-06-20 00:00:00+00:00 -0.021513
                                                         -0.001434
                                     0.001785
                                              0.009042
2024-06-21 00:00:00+00:00 -0.010444
                                     0.004073 -0.005601
                                                          0.009154
2024-06-24 00:00:00+00:00 0.003133
                                    0.006212 0.020839
                                                         -0.004691
2024-06-25 00:00:00+00:00 0.004468
                                    0.004788 -0.004690
                                                          0.007327
```

[3518 rows x 4 columns]

```
[195]: sns.pairplot(df, kind='reg') plt.show
```

[195]: <function matplotlib.pyplot.show(close=None, block=None)>





4.5 Analysis

- Strong Positive correlation between Apple and Microsoft
- Relative positive correlation between all other stocks
- Should not use Long-Short strategies

5 Sharpe Ratio

```
[226]: print('\n')
    print("Sharpe Ratio for AAPL: ", qs.stats.sharpe(aapl).round(2))
    print('\n')
    print("Sharpe Ratio for AZN: ", qs.stats.sharpe(azn).round(2))
    print('\n')
    print("Sharpe Ratio for BP: ", qs.stats.sharpe(bp).round(2))
    print('\n')
    print("Sharpe Ratio for MSFT: ", qs.stats.sharpe(msft).round(2))
```

```
Sharpe Ratio for AZN: 0.65

Sharpe Ratio for BP: 0.38

Sharpe Ratio for MSFT: 1.03
```

Sharpe Ratio for AAPL: 0.99

5.1 Analysis

• Apple and Microsoft offer a better risk-return relationship

[]: