Big Tech Simple Analysis

December 30, 2024

0.0.1 Simple Analysis

This program will show compute portfolio simple returns, get daily returns and volatility.

[16]: pip install yfinance Collecting yfinance Downloading yfinance-0.2.51-py2.py3-none-any.whl.metadata (5.5 kB) Requirement already satisfied: pandas>=1.3.0 in /opt/conda/lib/python3.11/sitepackages (from yfinance) (2.1.3) Requirement already satisfied: numpy>=1.16.5 in /opt/conda/lib/python3.11/sitepackages (from yfinance) (1.24.4) Requirement already satisfied: requests>=2.31 in /opt/conda/lib/python3.11/sitepackages (from yfinance) (2.31.0) Collecting multitasking>=0.0.7 (from yfinance) Downloading multitasking-0.0.11-py3-none-any.whl.metadata (5.5 kB) Requirement already satisfied: lxml>=4.9.1 in /opt/conda/lib/python3.11/sitepackages (from yfinance) (5.2.2) Requirement already satisfied: platformdirs>=2.0.0 in /opt/conda/lib/python3.11/site-packages (from yfinance) (4.1.0) Requirement already satisfied: pytz>=2022.5 in /opt/conda/lib/python3.11/sitepackages (from yfinance) (2023.3.post1) Collecting frozendict>=2.3.4 (from yfinance) Downloading frozendict-2.4.6-py311-none-any.whl.metadata (23 kB) Collecting peewee>=3.16.2 (from yfinance) Downloading peewee-3.17.8.tar.gz (948 kB) 948.2/948.2 kB 6.7 MB/s eta 0:00:0000:01 Installing build dependencies ... done Getting requirements to build wheel ... done Preparing metadata (pyproject.toml) ... done Requirement already satisfied: beautifulsoup4>=4.11.1 in /opt/conda/lib/python3.11/site-packages (from yfinance) (4.12.2) Requirement already satisfied: html5lib>=1.1 in /opt/conda/lib/python3.11/sitepackages (from yfinance) (1.1) Requirement already satisfied: soupsieve>1.2 in /opt/conda/lib/python3.11/sitepackages (from beautifulsoup4>=4.11.1->yfinance) (2.5) Requirement already satisfied: six>=1.9 in /opt/conda/lib/python3.11/site-

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packages (from html5lib>=1.1->yfinance) (1.16.0)
     Requirement already satisfied: webencodings in /opt/conda/lib/python3.11/site-
     packages (from html5lib>=1.1->yfinance) (0.5.1)
     Requirement already satisfied: python-dateutil>=2.8.2 in
     /opt/conda/lib/python3.11/site-packages (from pandas>=1.3.0->yfinance) (2.8.2)
     Requirement already satisfied: tzdata>=2022.1 in /opt/conda/lib/python3.11/site-
     packages (from pandas>=1.3.0->yfinance) (2023.3)
     Requirement already satisfied: charset-normalizer<4,>=2 in
     /opt/conda/lib/python3.11/site-packages (from requests>=2.31->yfinance) (3.3.2)
     Requirement already satisfied: idna<4,>=2.5 in /opt/conda/lib/python3.11/site-
     packages (from requests>=2.31->yfinance) (3.6)
     Requirement already satisfied: urllib3<3,>=1.21.1 in
     /opt/conda/lib/python3.11/site-packages (from requests>=2.31->yfinance) (2.1.0)
     Requirement already satisfied: certifi>=2017.4.17 in
     /opt/conda/lib/python3.11/site-packages (from requests>=2.31->yfinance)
     (2023.11.17)
     Downloading yfinance-0.2.51-py2.py3-none-any.whl (104 kB)
                             104.7/104.7 kB
     841.3 kB/s eta 0:00:0000:01
     Downloading frozendict-2.4.6-py311-none-any.whl (16 kB)
     Downloading multitasking-0.0.11-py3-none-any.whl (8.5 kB)
     Building wheels for collected packages: peewee
       Building wheel for peewee (pyproject.toml) ... done
       Created wheel for peewee:
     filename=peewee-3.17.8-cp311-cp311-linux_x86_64.whl size=300748
     Stored in directory: /home/jovyan/.cache/pip/wheels/ff/6c/15/506e25bc390de450a
     7fa53c155cd9b0fbd13ad3e84a9abc183
     Successfully built peewee
     Installing collected packages: peewee, multitasking, frozendict, yfinance
     Successfully installed frozendict-2.4.6 multitasking-0.0.11 peewee-3.17.8
     yfinance-0.2.51
     Note: you may need to restart the kernel to use updated packages.
[17]: #import
     import numpy as np
     import pandas as pd
     import yfinance as yf
     import matplotlib.pyplot as plt
     from datetime import datetime
     plt. style.use('fivethirtyeight')
 [6]: #Stock symbols
     stockSymbols = ["GOOGL", "AAPL", "AMZN", "META", "MSFT", "NVDA", "TSLA"]
```

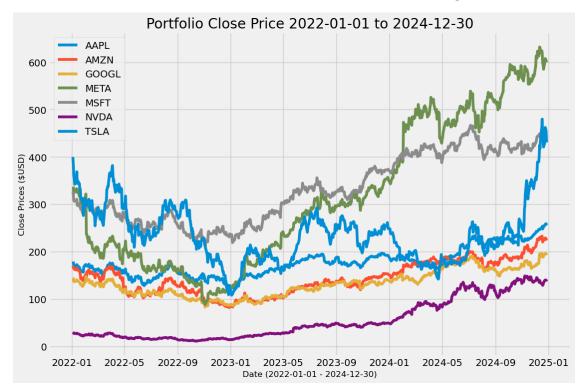
```
[8]: #start date
     stockStartDate = '2022-01-01'
 [9]: #todays date YYYY-mm-dd
     today = datetime.today().strftime('%Y-%m-%d')
     print(today)
     2024-12-30
[12]: #Get the number of assets
     numAssets = len(stockSymbols)
      #print(f'You have {numAssets} assets in your portfolio.')
[73]: #Get stock prices
     def getStockPrices(stocks = stockSymbols, start = stockStartDate, end = today,__

col = 'Close'):
         data = yf.download(stocks, start=start, end=end)[col]
         return data
     stocks_prices = getStockPrices(stockSymbols)
     #stocks_prices
     [********* 7 of 7 completed
[32]: def graph(stocks = stockSymbols, start = stockStartDate, end = today, col = __
      #t.i.t.l.e.
         title = f'Portfolio {col} Price {start} to {end}'
         #get the stocks
         stocks_prices = getStockPrices(stocks= stocks, start = start, end = end,__
       \hookrightarrowcol = col)
         #figure size
         plt.figure(figsize = (12, 8))
         #Plot iteration
         for i in stocks_prices.columns.values:
             plt.plot(stocks_prices[i], label=i)
         plt.title(title)
         plt.xlabel(f'Date ({start} - {end})', fontsize = 12)
         plt.ylabel(f'{col} Prices ($USD)', fontsize = 12)
```

```
plt.legend(stocks_prices.columns.values, loc = 'upper left')
plt.show()
```

[33]: graph(stockSymbols)

[********* 7 of 7 completed



[74]: # 2 years returns for Big Tech daily_returns = stocks_prices.pct_change(1) #fractional changes between current__ and prior element #daily_returns

[36]: daily_returns.corr()

[36]:	Ticker Ticker	AAPL	AMZN	GOOGL	META	MSFT	NVDA	TSLA
	AAPL	1.000000	0.563115	0.622199	0.502307	0.684440	0.554810	0.498880
	AMZN	0.563115	1.000000	0.649585	0.598642	0.689201	0.556612	0.448672
	GOOGL	0.622199	0.649585	1.000000	0.597194	0.694386	0.548561	0.402920
	META	0.502307	0.598642	0.597194	1.000000	0.593425	0.502581	0.316783
	MSFT	0 684440	0 689201	0 694386	0 593425	1 000000	0 638074	0 418782

```
NVDA 0.554810 0.556612 0.548561 0.502581 0.638074 1.000000 0.458779 TSLA 0.498880 0.448672 0.402920 0.316783 0.418782 0.458779 1.000000
```

```
[40]: daily_returns.cov() #daily_returns.var()
```

[40]:	Ticker	AAPL	AMZN	GOOGL	META	MSFT	NVDA	TSLA
	Ticker							
	AAPL	0.000292	0.000233	0.000219	0.000264	0.000203	0.000330	0.000329
	AMZN	0.000233	0.000584	0.000324	0.000445	0.000290	0.000469	0.000419
	GOOGL	0.000219	0.000324	0.000426	0.000379	0.000249	0.000395	0.000321
	META	0.000264	0.000445	0.000379	0.000944	0.000317	0.000538	0.000376
	MSFT	0.000203	0.000290	0.000249	0.000317	0.000302	0.000387	0.000281
	NVDA	0.000330	0.000469	0.000395	0.000538	0.000387	0.001215	0.000618
	TSLA	0.000329	0.000419	0.000321	0.000376	0.000281	0.000618	0.001492

0.0.2 Results from covariance matrix

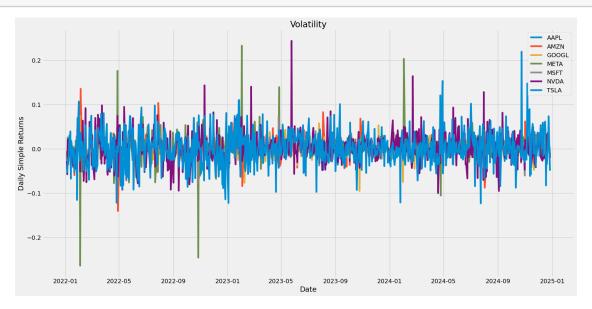
Diagonals shows the variance of each stocks symbols and higher the variance the higher risk, higher returns and higher volatility. In constrast, lower variance, lower risk and returns.

Non-diagonals shows covariance between two different stock symbols. If it's positive covariance, it shows that the two assets tend to move in the same direction. Since we are investigating the Big Tech companies, they obviously have some covariance between them.

0.0.3 Volatility

```
[49]: daily_returns.std()
[49]: Ticker
      AAPL
               0.017080
               0.024173
      AMZN
      GOOGL
               0.020648
     META
               0.030725
     MSFT
               0.017386
     NVDA
               0.034851
      TSLA
               0.038630
      dtype: float64
[53]: plt.figure(figsize = (20,10))
      for i in daily_returns.columns.values:
          plt.plot(daily_returns[i], label = i)
      plt.legend(loc='upper right')
      plt.title('Volatility')
      plt.xlabel('Date')
      plt.ylabel('Daily Simple Returns')
```

plt.show()



```
[77]: dailyMean = daily_returns.mean() #dailyMean
```

```
[61]: #expected daily return

#5% AAPL, 5% AMZN, 5% GOOGL, 20% META, 15% MSFT, 25% NVDA, 25% TSLA
randomWeights = np.array([0.05, 0.05, 0.05, 0.2, 0.15, 0.25, 0.25])

portfolioReturn = np.sum(dailyMean * randomWeights)

print(f"The daily expected portfolio return: {portfolioReturn}")
```

The daily expected portfolio return: 0.001287295662819741

```
[64]: #yearly return

print(f"Expected annualised portfolio return: {portfolioReturn * 251}")
```

Expected annualised portfolio return: 0.323111211367755

About 32.3% return with the weights of

5% AAPL, 5% AMZN, 5% GOOGL, 20% META, 15% MSFT, 25% NVDA, 25% TSLA

```
[75]: #Growth of the investment
growthReturns = (daily_returns+1).cumprod() #Return cumulative product
```

#growthReturns

```
[72]: #Visualize daily cumulative returns

plt.figure(figsize = (15,7))

for i in growthReturns.columns.values:
    plt.plot(growthReturns.index, growthReturns[i], label = i)

plt.legend(loc = 'upper left')
plt.xlabel(f'Date {stockStartDate} - {today}')
plt.ylabel(f'Growth of $1 investment')
plt.title("Daily Cumulative Simple Returns")
plt.show()
```

