```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt

In [2]: import seaborn as sns

In [3]: import glob

In [4]: #import all the packages

In [5]: glob.glob(r'/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/>
```

```
Out[5]:
```

```
['/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/XRX_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/GS data.cs
۷',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SPGI data.
```

- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MTB data.c
- sv', '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/V_data.cs
- '/Users/rivalachuriva/Desktop/Python Project/individual stocks 5yr/CTAS data.
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ESRX data.
- csv',
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/APH data.c sv'
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/BBT data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/RHI_data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/GGP data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HCA data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HD data.cs
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PEP data.c sv',
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/INTC_data. csv',
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MAC data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CLX_data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/UNP data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WAT_data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EMN data.c
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- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ALK data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/VZ_data.cs ۷',
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- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AON data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/APC data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/WRK data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/XYL_data.c

```
Stock_Price_Analysis
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sv'
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 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CHD_data.c
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 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/SNA_data.c
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 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/LNC_data.c
```

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sv',

```
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 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WMT_data.c
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Stock_Price_Analysis '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CAT_data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/KMX data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CELG data. '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/INCY data. '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CBOE_data. csv', '/Users/rivalachuriva/Desktop/Python Project/individual stocks 5yr/ALL data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AMT data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MDT data.c sv' '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PNC data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/APD data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HBAN data. '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EBAY data. '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/BBY data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ICE data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WDC_data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EXC data.c '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/TROW_data. '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ZTS data.c sv' '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CMG_data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/TRV data.c '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/MKC_data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ABC data.c '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/FRT_data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/TXN data (1).csv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DVN data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/VIAB_data. csv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ARNC data. csv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/LLL data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MNST data.

'/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/LEN_data.c

localhost:8888/nbconvert/html/Stock_Price_Analysis.ipynb?download=false

sv',

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```

```
Stock_Price_Analysis
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 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AMD data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PNR data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/RHT data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AME data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/APTV data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CTSH data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MRK data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/WU data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/IBM_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MSFT data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/STT_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FMC data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/UHS_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PPG data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/KMI_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/RE data.cs
۷',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HON data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/PX_data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AAPL data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/RMD data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MAT data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/PEG_data.c
```

```
Stock_Price_Analysis
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/INTU_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SHW data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/KEY data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ADM data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/MMM_data.c
sv'
 '/Users/rivalachuriva/Desktop/Python Project/individual stocks 5yr/AOS data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CVX data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PLD data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CL data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DISH data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/A data.cs
٧',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EQIX data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/GE data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/GD data.cs
۷',
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/ALLE_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PHM data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/AIG_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/LKQ data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/HSIC_data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ADS data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/VRTX_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MDLZ data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/MON_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MPC data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/APA data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/BRK.B_dat
a.csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AMP data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FCX data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PCAR data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/NLSN_data.
csv',
```

```
Stock_Price_Analysis
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/TJX_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PYPL data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FBHS data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AXP data.c
sv'.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/BA_data.cs
 '/Users/rivalachuriva/Desktop/Pvthon Project/individual stocks 5vr/LNT data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/ALB_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PG data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/XEL data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/ATVI_data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FFIV data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EFX data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CMI data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EL data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/ESS_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PCG data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/ALGN_data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/KORS data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/HSY_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HLT data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/T_data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CHRW data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/AYI_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/LYB data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HUM data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/FB_data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DRI data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SYK data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MAA data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/SCHW_data.
csv',
```

```
Stock_Price_Analysis
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/USB_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/TXT data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PM data.cs
ν¹,
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/NTRS data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/BAX_data.c
sv'
 '/Users/rivalachuriva/Desktop/Pvthon Project/individual stocks 5vr/ITW data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/BK data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DVA data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/GPN data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/IT data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CCL data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SWKS data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HAS data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DHR data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WMB_data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HCN data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CBS_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EA data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CME_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SEE data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/VTR_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HST data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/SLG_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/JWN data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/LB data.cs
٧',
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/S0_data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/UDR data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ALXN data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/NUE data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/GPC_data.c
sv',
```

```
Stock_Price_Analysis
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/FDX_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SYF data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/LUV data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/TGT data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/VAR_data.c
sv',
 '/Users/rivalachuriva/Desktop/Pvthon Project/individual stocks 5vr/DISCA dat
a.csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FIS data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EXPE data.
csv'.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/EXPD data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DRE data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/LH data.cs
٧',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DXC data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/REG data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PWR data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/WLTW data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/XOM data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/IDXX_data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AVY data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/BF.B_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PXD data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WM_data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AMAT data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/JEC_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MYL data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ROST data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/G00GL_dat
a.csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/BSX data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CXO data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HPQ data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WYNN_data.
csv',
```

```
Stock_Price_Analysis
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/PDCO_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HRL data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ROK data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FE data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WHR_data.c
sv',
 '/Users/rivalachuriva/Desktop/Python Project/individual stocks 5yr/ETFC data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FTI data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AWK data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/NDAQ data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DPS data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DISCK dat
a.csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ETR data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/VNO data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/XLNX data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/HII_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/KO data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/LEG_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/IR data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/MLM_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/TSCO data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/NWS_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/RTN data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/TSS_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HRS data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/BLK data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/GILD_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PAYX data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DG data.cs
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MCD data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/FTV_data.c
```

```
Stock_Price_Analysis
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/IPG_data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/KIM data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/STX data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AES data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/PFE_data.c
sv'
 '/Users/rivalachuriva/Desktop/Python Project/individual stocks 5vr/YUM data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CAH data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AGN data.c
sv'
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MGM data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CHK data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HBI data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/DTE data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/NWL data.c
sv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/M data.cs
٧¹
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/L_data.cs
۷',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ABT data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WBA_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/BMY data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/H0LX_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/PBCT data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/URI_data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/COL data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/AMZN_data.
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/JCI data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CBG data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/NCLH_data.
csv',
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/RCL data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/NBL data.c
 '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SYMC data.
 '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/CA_data.cs
٧',
```

- Stock_Price_Analysis '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/JBHT_data. csv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FOX data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/FAST data. '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ZION data. '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/ABBV_data. csv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MCO data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/SYY data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MAS data.c sv' '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/OXY data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MAR data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AAP data.c '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/HPE_data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/MO data.cs '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ARE data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/ORLY data. csv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/HCP data.c '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/T_data(1). '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/NKE data.c '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/HAL_data.c sv', '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/CERN data. '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/SJM_data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/WY data.cs '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/WYN_data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/WFC data.c '/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/QCOM data.
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/TWX_data (1).csv',
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/COF_data.c
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/COG_data.csv',
- '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/F_data.cs v']

```
#import data in the python # only csv file # it is raw string
 In [6]:
 In [7]:
         len(glob.glob(r'/Users/riyalachuriya/Desktop/Python Project/individual_stocks
         509
 Out[7]:
 In [8]:
         #glob is use for specific file pattern or extension
 In [9]:
         company list = [
              r'/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/AAPL_dates
              r'/Users/riyalachuriya/Desktop/Python Project/individual stocks 5yr/AMZN d
              r'/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/G00G_d
              r'/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/MSFT_d
In [10]:
         #import specific data file from above data set
          #will take stocks of this four company # amazon,apple,google,microsoft
In [11]: all data = pd.DataFrame() # blank dataframe
         for file in company_list:
              current df = pd.read csv(file)
              all_data = pd.concat([all_data, current_df], ignore_index=True)
         # Now 'all data' should contain the concatenated data from all CSV files.
         all_data.shape
In [12]:
         (4752, 7)
Out[12]:
In [13]:
         all_data.head(6)
Out[13]:
                  date
                                  high
                                          low
                                                          volume Name
                         open
                                                 close
         0 2013-02-08
                        67.7142 68.4014 66.8928
                                               67.8542
                                                       158168416
                                                                  AAPL
          1 2013-02-11 68.0714 69.2771
                                       67.6071
                                               68.5614 129029425
                                                                  AAPL
          2 2013-02-12 68.5014
                               68.9114 66.8205 66.8428
                                                      151829363
                                                                 AAPL
         3 2013-02-13 66.7442 67.6628
                                       66.1742
                                               66.7156
                                                       118721995
                                                                  AAPL
         4 2013-02-14 66.3599
                                                                 AAPL
                               67.3771 66.2885 66.6556
                                                        88809154
         5 2013-02-15 66.9785 67.1656 65.7028
                                               65.7371
                                                        97924631
                                                                 AAPL
         all data['Name'].unique()
In [14]:
         array(['AAPL', 'AMZN', 'GOOG', 'MSFT'], dtype=object)
Out[14]:
In [15]:
         all data.isnull()
```

Out[15]:		date	open	high	low	close	volume	Name
	0	False	False	False	False	False	False	False
	1	False	False	False	False	False	False	False
	2	False	False	False	False	False	False	False
	3	False	False	False	False	False	False	False
	4	False	False	False	False	False	False	False
	•••		•••	•••	•••		•••	•••
	4747	False	False	False	False	False	False	False
	4748	False	False	False	False	False	False	False
	4749	False	False	False	False	False	False	False
	4750	False	False	False	False	False	False	False
	4751	False	False	False	False	False	False	False

4752 rows × 7 columns

```
In [16]: #False means in that row there is no missing value
In [17]: all_data.isnull().sum()
                   0
         date
Out[17]:
                   0
         open
         high
                    0
         low
                    0
         close
                   0
         volume
                   0
         Name
         dtype: int64
In [18]: all_data.dtypes
                    object
         date
Out[18]:
                   float64
         open
                    float64
         high
         low
                   float64
         close
                   float64
         volume
                     int64
         Name
                    object
         dtype: object
In [19]: #date can't be string it should be date time object
         #date time have it's own type which is datetime64[ns] ns is nenosec
         ##or it can be uts as well which is universal co-ordinated
         ## 64 means 64 bits
In [20]: all_data['date'] = pd.to_datetime(all_data['date'])
In [21]: all_data['date']
```

```
2013-02-08
Out[21]:
          1
                  2013-02-11
          2
                  2013-02-12
          3
                  2013-02-13
                  2013-02-14
          4747
                  2018-02-01
          4748
                  2018-02-02
          4749
                  2018-02-05
          4750
                  2018-02-06
          4751
                  2018-02-07
          Name: date, Length: 4752, dtype: datetime64[ns]
          tech_list = all_data['Name'].unique()
In [22]:
In [23]:
          plt.figure(figsize=(20,12))
           for index , company in enumerate(tech list , 1):
               plt.subplot(2 , 2 , index) ## creating subplot for each stock
               filter1 = all_data['Name'] == company
               df = all data[filter1]
               plt.plot(df['date'] , df['close']) ## plotting "date" vs "close"
               plt.title(company)
                                                                               AMZN
           160
           140
                                                          1000
                                                          800
           100
                                                          600
                                                          400
                                                          200
                                                                   2014
                                                                                                2018
             2013
                           2015
                                  2016
                                         2017
                                                 2018
                                                                          2015
                                                                                  2016
                                                                                         2017
                               GOOG
                                                                               MSFT
          1200
          1100
          1000
                                                           60
           800
           700
           600
                   2015-01 2015-07 2016-01 2016-07
               2014-07
                                      2017-01
                                           2017-07
                                                                   2014
                                                                           2015
                                                                                  2016
                                                                                         2017
                                                                                                2018
          #First graph (AAPL)
In [24]:
          #The stock price has steadily increased over the entire period, more than doub
          #There have been a few periods of dips and fluctuations, but the overall trend
          #Second graph (AMZN)
In [25]:
          #the stock price of AMZN has also increased significantly over the period, from
          #The growth of AMZN's stock price has been even more impressive than that of AV
          #There have also been some dips and fluctuations for AMZN, but the overall tree
```

#GOOG's stock price has not shown a clear upward trend over the entire period.

localhost:8888/nbconvert/html/Stock_Price_Analysis.ipynb?download=false

In [26]: #Third graph (G00G)

#There have been periods of both growth and decline, with the stock price reach #G00G's stock price has been more volatile than AAPL and AMZN.

In [27]: #Fourth graph (MSFT)

#MSFT's stock price has not had a clear upward or downward trend over the entile #There have been periods of both growth and decline, with the stock price reach #MSFT's stock price has also been more volatile than AAPL and AMZN.

In [28]: #Overall

#The stocks of AAPL and AMZN have significantly outperformed the stocks of GOOC #AAPL and AMZN have both shown steady upward trends in their stock prices, whi

In [29]: all_data.head(15)

Out[29]:		date	open	high	low	close	volume	Name
	0	2013-02-08	67.7142	68.4014	66.8928	67.8542	158168416	AAPL
	1	2013-02-11	68.0714	69.2771	67.6071	68.5614	129029425	AAPL
	2	2013-02-12	68.5014	68.9114	66.8205	66.8428	151829363	AAPL
	3	2013-02-13	66.7442	67.6628	66.1742	66.7156	118721995	AAPL
	4	2013-02-14	66.3599	67.3771	66.2885	66.6556	88809154	AAPL
	5	2013-02-15	66.9785	67.1656	65.7028	65.7371	97924631	AAPL
	6	2013-02-19	65.8714	66.1042	64.8356	65.7128	108854046	AAPL
	7	2013-02-20	65.3842	65.3842	64.1142	64.1214	118891367	AAPL
	8	2013-02-21	63.7142	64.1671	63.2599	63.7228	111596821	AAPL
	9	2013-02-22	64.1785	64.5142	63.7999	64.4014	82583823	AAPL
	10	2013-02-25	64.8356	65.0171	63.2242	63.2571	92899597	AAPL
	11	2013-02-26	63.4028	64.5056	62.5228	64.1385	125096657	AAPL
	12	2013-02-27	64.0614	64.6342	62.9499	63.5099	146674682	AAPL
	13	2013-02-28	63.4357	63.9814	63.0571	63.0571	80532382	AAPL
	14	2013-03-01	62.5714	62.5971	61.4257	61.4957	137899041	AAPL

In [30]: all_data['close'].rolling(window = 10).mean().head(15)

```
NaN
Out[30]:
          1
                     NaN
          2
                     NaN
          3
                     NaN
          4
                     NaN
          5
                     NaN
          6
                     NaN
          7
                     NaN
          8
                     NaN
          9
                66.03251
                65.57280
          10
                65.13051
          11
          12
                64.79722
          13
                64.43137
          14
                63.91538
         Name: close, dtype: float64
In [31]: new_data = all_data.copy()
          #### now lets consider different windows of rolling ,ie 10 days ,20 days ,30 days
In [32]:
          ma_day = [10, 20, 50]
          for ma in ma_day:
              new_data['close_'+str(ma)] = new_data['close'].rolling(ma).mean()
          new_data
In [33]:
```

Out[33]:		date	open	high	low	close	volume	Name	close_10	close_20	close
	0	2013- 02- 08	67.7142	68.4014	66.8928	67.8542	158168416	AAPL	NaN	NaN	
	1	2013- 02-11	68.0714	69.2771	67.6071	68.5614	129029425	AAPL	NaN	NaN	
	2	2013- 02-12	68.5014	68.9114	66.8205	66.8428	151829363	AAPL	NaN	NaN	
	3	2013- 02-13	66.7442	67.6628	66.1742	66.7156	118721995	AAPL	NaN	NaN	
	4	2013- 02- 14	66.3599	67.3771	66.2885	66.6556	88809154	AAPL	NaN	NaN	
	•••										
	4747	2018- 02-01	94.7900	96.0700	93.5813	94.2600	47227882	MSFT	92.765	90.6770	86.9
	4748	2018- 02- 02	93.6400	93.9700	91.5000	91.7800	47867753	MSFT	92.943	90.9105	87.
	4749	2018- 02- 05	90.5600	93.2400	88.0000	88.0000	51031465	MSFT	92.582	90.9010	87.2
	4750	2018- 02- 06	86.8900	91.4750	85.2500	91.3300	67998564	MSFT	92.525	91.0535	87.4
	4751	2018- 02- 07	90.4900	91.7700	89.2000	89.6100	41107592	MSFT	92.304	91.1230	87.!

4752 rows × 10 columns

In [34]: new_data.tail(17)

Out[34]:		date	open	high	low	close	volume	Name	close_10	close_20	close_50
	4735	2018- 01-16	90.100	90.790	88.0104	88.35	36599736	MSFT	87.795	86.7955	84.9602
	4736	2018- 01-17	89.080	90.280	88.7500	90.14	25621164	MSFT	88.214	86.9600	85.0820
	4737	2018- 01-18	89.800	90.670	89.6600	90.10	24159683	MSFT	88.589	87.1460	85.2012
	4738	2018- 01-19	90.140	90.610	89.6600	90.00	36875013	MSFT	88.878	87.3545	85.3118
	4739	2018- 01-22	90.000	91.620	89.7400	91.61	23601555	MSFT	89.220	87.6590	85.4586
	4740	2018- 01-23	91.900	92.300	91.5400	91.90	23412841	MSFT	89.582	87.9790	85.6054
	4741	2018- 01-24	92.550	93.430	91.5800	91.82	33277483	MSFT	89.942	88.2945	85.7600
	4742	2018- 01-25	92.465	93.240	91.9300	92.33	26383238	MSFT	90.393	88.6410	85.9292
	4743	2018- 01-26	93.120	94.060	92.5800	94.06	29172167	MSFT	90.991	89.0585	86.1318
	4744	2018- 01-29	95.140	95.450	93.7200	93.92	31569940	MSFT	91.423	89.4685	86.3292
	4745	2018- 01- 30	93.300	93.660	92.1000	92.74	38635053	MSFT	91.862	89.8285	86.5244
	4746	2018- 01-31	93.750	95.400	93.5100	95.01	48756338	MSFT	92.349	90.2815	86.7606
	4747	2018- 02-01	94.790	96.070	93.5813	94.26	47227882	MSFT	92.765	90.6770	86.9978
	4748	2018- 02- 02	93.640	93.970	91.5000	91.78	47867753	MSFT	92.943	90.9105	87.1828
	4749	2018- 02- 05	90.560	93.240	88.0000	88.00	51031465	MSFT	92.582	90.9010	87.2684
	4750	2018- 02- 06	86.890	91.475	85.2500	91.33	67998564	MSFT	92.525	91.0535	87.4328
	4751	2018- 02- 07	90.490	91.770	89.2000	89.61	41107592	MSFT	92.304	91.1230	87.5598
In [35]:	new_d	lata.se	et_inde	x('date	', inpla	ce = 1	True)				
In [36]:	new_d	lata.co	lumns								
Out[36]:	Index		en', 'hi se_20',			lose',	'volume'	, 'Nam	e', 'clo	se_10',	

dtype='object')

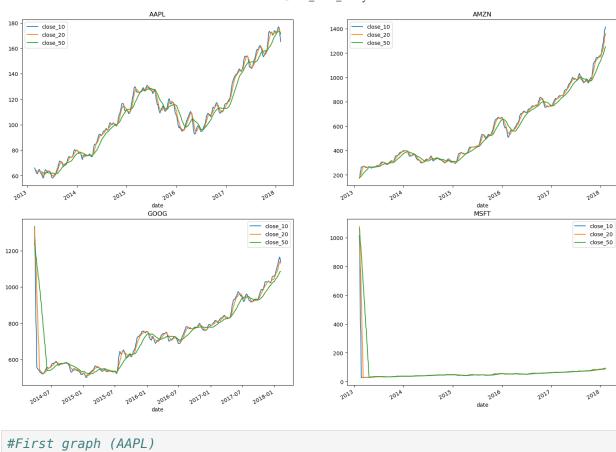
In [37]:	new_data										
Out[37]:		open	high	low	close	volume	Name	close_10	close_20	close_50	
	date										
	2013- 02- 08	67.7142	68.4014	66.8928	67.8542	158168416	AAPL	NaN	NaN	NaN	
	2013- 02-11	68.0714	69.2771	67.6071	68.5614	129029425	AAPL	NaN	NaN	NaN	
	2013- 02-12	68.5014	68.9114	66.8205	66.8428	151829363	AAPL	NaN	NaN	NaN	
	2013- 02-13	66.7442	67.6628	66.1742	66.7156	118721995	AAPL	NaN	NaN	NaN	
	2013- 02-14	66.3599	67.3771	66.2885	66.6556	88809154	AAPL	NaN	NaN	NaN	
	•••					•••					
	2018- 02-01	94.7900	96.0700	93.5813	94.2600	47227882	MSFT	92.765	90.6770	86.9978	
	2018- 02- 02	93.6400	93.9700	91.5000	91.7800	47867753	MSFT	92.943	90.9105	87.1828	
	2018- 02- 05	90.5600	93.2400	88.0000	88.0000	51031465	MSFT	92.582	90.9010	87.2684	
	2018- 02- 06	86.8900	91.4750	85.2500	91.3300	67998564	MSFT	92.525	91.0535	87.4328	
	2018- 02- 07	90.4900	91.7700	89.2000	89.6100	41107592	MSFT	92.304	91.1230	87.5598	

4752 rows × 9 columns

```
In []:

In [38]: plt.figure(figsize=(20,14))

for index , company in enumerate(tech_list , 1):
    plt.subplot(2 , 2 , index)
    filter1 = new_data['Name']==company
    df = new_data[filter1]
    df[['close_10','close_20', 'close_50']].plot(ax=plt.gca())
    plt.title(company)
```



- In [39]: #First graph (AAPL)

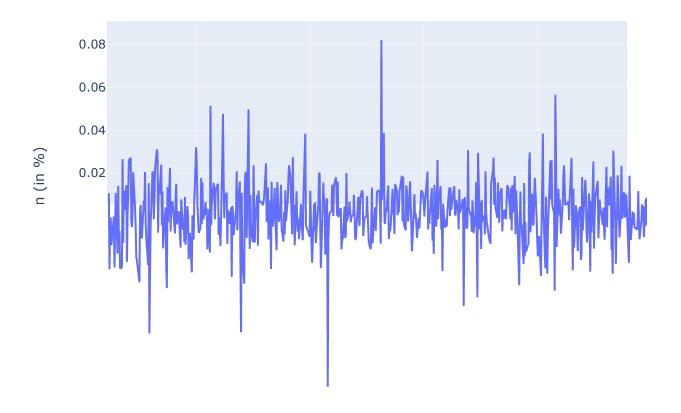
 #Both moving averages show an upward trend over the entire period, indicating and the second secon
- In [40]: #Second graph (AMZN)

 #Both moving averages show a very strong upward trend over the entire period,
 #Like AAPL, the 10-day moving average for AMZN is more volatile than the 50-day
- In [41]: #Third graph (GOOG)

 #Unlike AAPL and AMZN, the moving averages for GOOG do not show a clear upward

 #The 10-day moving average for GOOG is more volatile than the 50-day moving ave
- In [42]: #Fourth graph (MSFT)
 #Similar to GOOG, the moving averages for MSFT do not show a clear upward or do
 #The 10-day moving average for MSFT is more volatile than the 50-day moving ave
- In [43]: #Overall
 #The moving averages confirm that AAPL and AMZN outperformed GOOG and MSFT over
 #AAPL and AMZN showed strong upward trends in their moving averages, while GOOG
 #This suggests that AAPL and AMZN were more stable and predictable in their sta
- In []:
 In []:
 In [44]: apple = pd.read_csv(r'/Users/riyalachuriya/Desktop/Python Project/individual_s
- In [45]: apple.head(5)

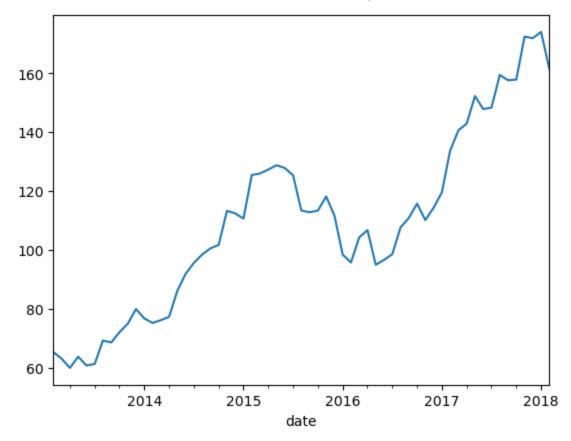
```
Out[45]:
                   date
                                    high
                                              low
                                                     close
                                                              volume Name
                           open
          0 2013-02-08
                          67.7142 68.4014 66.8928
                                                   67.8542
                                                           158168416
                                                                       AAPL
              2013-02-11
                         68.0714
                                  69.2771
                                           67.6071
                                                   68.5614
                                                           129029425
                                                                       AAPL
             2013-02-12 68.5014
                                  68.9114
                                          66.8205
                                                  66.8428
                                                           151829363
                                                                       AAPL
              2013-02-13 66.7442
                                 67.6628
                                          66.1742
                                                   66.7156
                                                            118721995
                                                                       AAPL
             2013-02-14 66.3599
                                  67.3771
                                          66.2885
                                                  66.6556
                                                            88809154
                                                                       AAPL
In [46]: apple['close']
                    67.8542
Out[46]:
          1
                    68.5614
          2
                    66.8428
          3
                    66.7156
                    66.6556
                      . . .
          1254
                   167.7800
          1255
                   160.5000
          1256
                   156.4900
          1257
                   163.0300
                   159.5400
          1258
          Name: close, Length: 1259, dtype: float64
In [47]:
          apple.head(4)
Out [47]:
                   date
                           open
                                    high
                                             low
                                                     close
                                                              volume Name
          0 2013-02-08 67.7142 68.4014
                                         66.8928
                                                  67.8542
                                                           158168416
                                                                       AAPL
           1 2013-02-11 68.0714
                                 69.2771
                                          67.6071
                                                  68.5614
                                                          129029425
                                                                       AAPL
                                                                       AAPL
            2013-02-12 68.5014
                                 68.9114
                                          66.8205
                                                  66.8428
                                                          151829363
          3 2013-02-13 66.7442 67.6628
                                          66.1742
                                                   66.7156
                                                           118721995
                                                                       AAPL
In [48]:
          apple['Daily return (in %)']= apple['close'].pct_change()
In [49]:
          apple.head(4)
Out[49]:
                   date
                           open
                                    high
                                              low
                                                     close
                                                              volume
                                                                      Name Daily return (in %)
          0 2013-02-08
                         67.7142
                                 68.4014
                                         66.8928
                                                   67.8542
                                                           158168416
                                                                       AAPL
                                                                                         NaN
           1 2013-02-11 68.0714
                                 69.2771
                                          67.6071
                                                   68.5614 129029425
                                                                       AAPL
                                                                                     0.010422
          2 2013-02-12 68.5014
                                 68.9114
                                          66.8205
                                                  66.8428
                                                                                    -0.025067
                                                           151829363
                                                                       AAPL
             2013-02-13 66.7442 67.6628
                                          66.1742
                                                   66.7156
                                                           118721995
                                                                       AAPL
                                                                                    -0.001903
In [50]:
          import plotly.express as px
          px.line(apple, x = 'date', y = 'Daily return (in %)')
In [51]:
```



In [52]:	#The graph you sent me	appears to show the daily percentage change in Apple's							
In [53]:	#Apple's stock price has been volatile over the past five years, with daily cha #There have been several periods of significant volatility, such as in the sum #Overall, the stock price has trended upwards over the period shown, with some								
In [54]:	apple.dtypes								
Out[54]:	date open high low close volume Name Daily return (in %) dtype: object	object float64 float64 float64 float64 int64 object float64							
In [55]:	apple['date'] = pd.to_	<pre>datetime(apple['date'])</pre>							
In [56]:	apple.dtypes								

2/8/24, 10:50 AM

```
datetime64[ns]
          date
Out[56]:
                                          float64
          open
                                          float64
          high
          low
                                          float64
                                          float64
          close
          volume
                                            int64
          Name
                                           object
          Daily return (in %)
                                          float64
          dtype: object
          apple.set index('date', inplace = True)
In [57]:
          apple.head(4)
In [58]:
Out[58]:
                        open
                                 high
                                          low
                                                close
                                                         volume Name Daily return (in %)
                 date
          2013-02-08 67.7142 68.4014 66.8928
                                              67.8542
                                                       158168416
                                                                  AAPL
                                                                                   NaN
           2013-02-11 68.0714 69.2771
                                      67.6071 68.5614 129029425
                                                                  AAPL
                                                                               0.010422
          2013-02-12 68.5014
                              68.9114
                                      66.8205 66.8428
                                                      151829363
                                                                  AAPL
                                                                               -0.025067
          2013-02-13 66.7442 67.6628
                                      66.1742
                                              66.7156
                                                       118721995
                                                                  AAPL
                                                                               -0.001903
          apple['close'].resample('M').mean()
In [59]:
          date
Out[59]:
          2013-02-28
                          65.306264
                          63.120110
          2013-03-31
          2013-04-30
                          59.966432
          2013-05-31
                          63.778927
          2013-06-30
                          60.791120
                            . . .
          2017-10-31
                         157.817273
          2017-11-30
                         172.406190
          2017-12-31
                         171.891500
          2018-01-31
                         174,005238
          2018-02-28
                         161.468000
          Freq: M, Name: close, Length: 61, dtype: float64
In [60]: apple['close'].resample('M').mean().plot()
          <Axes: xlabel='date'>
Out[60]:
```

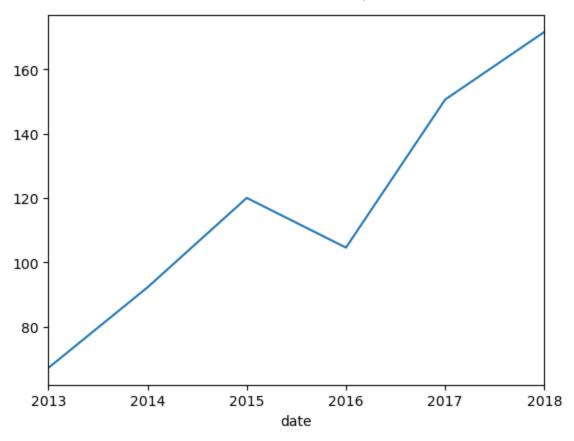


In [61]: #The price has increased overall from around \$80 in early 2014 to around \$160 in #There is a lot of volatility in the data, with some months seeing price increases #However, the overall trend is clearly upwards.

In [62]: #The largest price increase occurred in January 2018, when the price jumped by #The largest price decrease occurred in August 2015, when the price fell by ove #There appears to be a seasonal pattern to the data, with prices tending to be

```
apple['close'].resample('Y').mean()
In [63]:
         date
Out[63]:
                         67.237839
         2013-12-31
         2014-12-31
                         92.264531
         2015-12-31
                        120.039861
         2016-12-31
                        104.604008
         2017-12-31
                        150.585080
         2018-12-31
                        171.594231
         Freq: A-DEC, Name: close, dtype: float64
In [64]:
         apple['close'].resample('Y').mean().plot()
```

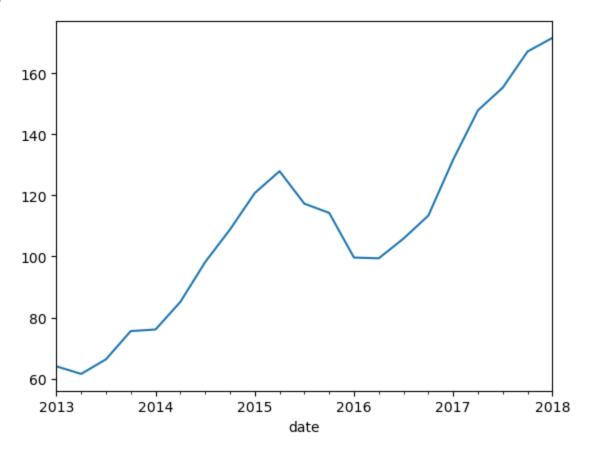
Out[64]: <Axes: xlabel='date'>



```
#Overall, the stock price has increased significantly over the past 11 years.
In [65]:
          #There has been significant volatility year-to-year. While the overall trend is
In [66]:
In [67]:
          #The stock price has reached new all-time highs in recent years. In 2023, the
          apple['close'].resample('Q').mean()
In [68]:
          date
Out[68]:
          2013-03-31
                         64.020291
                         61.534692
          2013-06-30
          2013-09-30
                         66.320670
                         75.567478
          2013-12-31
                         76.086293
          2014-03-31
          2014-06-30
                         85.117475
          2014-09-30
                         98.163311
          2014-12-31
                        108.821016
          2015-03-31
                        120.776721
          2015-06-30
                        127.937937
          2015-09-30
                        117.303438
          2015-12-31
                        114.299297
                         99.655082
          2016-03-31
          2016-06-30
                         99.401250
                        105.866094
          2016-09-30
          2016-12-31
                        113.399048
          2017-03-31
                        131.712500
          2017-06-30
                        147.875397
          2017-09-30
                        155.304603
          2017-12-31
                        167.148254
          2018-03-31
                        171.594231
          Freq: Q-DEC, Name: close, dtype: float64
```

```
In [69]: apple['close'].resample('Q').mean().plot()
```

Out[69]: <Axes: xlabel='date'>



```
#Apple's stock price has generally increased over the timeframe shown. There is
In [70]:
In [71]:
         #There have been several periods of significant price increases. These include
         #There have also been several periods of significant price declines. These inc
In [72]:
         #Overall, the quarterly closing price of Apple stock has shown a generally upwa
In [73]:
         company list[0]
In [74]:
         '/Users/riyalachuriya/Desktop/Python Project/individual_stocks_5yr/AAPL_data.c
Out[74]:
         app = pd.read_csv(company_list[0])
In [75]:
         amzn = pd.read_csv(company_list[1])
         google = pd.read_csv(company_list[2])
         msft = pd.read csv(company list[3])
In [76]:
         closing_price = pd.DataFrame()
In [77]: closing_price['apple_close'] = app['close']
         closing_price['amzn_close'] = amzn['close']
         closing_price['google_close'] = google['close']
         closing_price['msft_close'] = msft['close']
```

In [78]: closing_price

\cap	1.11	+	Γ	7	0]	
U	u	L	L	/	0]	

	apple_close	amzn_close	google_close	msft_close
0	67.8542	261.95	558.46	27.55
1	68.5614	257.21	559.99	27.86
2	66.8428	258.70	556.97	27.88
3	66.7156	269.47	567.16	28.03
4	66.6556	269.24	567.00	28.04
•••				
1254	167.7800	1390.00	NaN	94.26
1255	160.5000	1429.95	NaN	91.78
1256	156.4900	1390.00	NaN	88.00
1257	163.0300	1442.84	NaN	91.33
1258	159.5400	1416.78	NaN	89.61

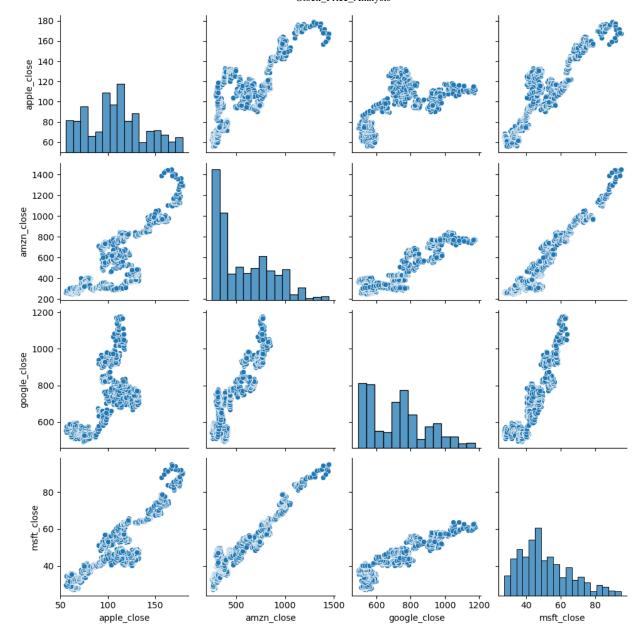
1259 rows × 4 columns

In [79]: sns.pairplot(closing_price)

/Users/riyalachuriya/anaconda3/lib/python3.11/site-packages/seaborn/axisgrid.p y:118: UserWarning:

The figure layout has changed to tight

Out[79]: <seaborn.axisgrid.PairGrid at 0x1597de110>



In [80]: #apple nd amazon have a straight line graph which mean they have higher co rela

In [81]: closing_price.corr()

 Out [81]:
 apple_close amzn_close google_close msft_close

 apple close
 1.000000
 0.819078
 0.640522
 0.899689

apple_close	1.000000	0.819078	0.640522	0.899689
amzn_close	0.819078	1.000000	0.888456	0.955977
google_close	0.640522	0.888456	1.000000	0.907011
msft_close	0.899689	0.955977	0.907011	1.000000

In [82]: sns.heatmap(closing_price.corr(), annot= True)

Out[82]: <Axes: >



```
In [83]:
          closing_price['apple_close']
                   67.8542
Out[83]:
          1
                   68.5614
          2
                   66.8428
          3
                   66.7156
          4
                   66.6556
          1254
                  167.7800
                  160.5000
          1255
          1256
                  156,4900
          1257
                  163.0300
          1258
                  159.5400
          Name: apple_close, Length: 1259, dtype: float64
In [84]:
         closing_price['apple_close'].shift(1)
                       NaN
Out[84]:
          1
                   67.8542
          2
                   68.5614
          3
                   66.8428
          4
                   66.7156
          1254
                  167.4300
          1255
                  167.7800
          1256
                  160.5000
          1257
                  156.4900
          1258
                  163,0300
          Name: apple_close, Length: 1259, dtype: float64
In [85]: closing_price['apple_close'] - closing_price['apple_close'].shift(1)
```

```
NaN
Out[85]:
          1
                  0.7072
          2
                 -1.7186
          3
                 -0.1272
                 -0.0600
                   . . .
          1254
                  0.3500
          1255
                 -7.2800
          1256
                 -4.0100
          1257
                  6.5400
          1258
                 -3.4900
         Name: apple_close, Length: 1259, dtype: float64
In [86]:
          (closing_price['apple_close'] - closing_price['apple_close'].shift(1))/closing
                       NaN
Out[86]:
          1
                  0.010422
          2
                 -0.025067
          3
                 -0.001903
          4
                 -0.000899
          1254
                  0.002090
          1255
                 -0.043390
          1256
                 -0.024984
          1257
                  0.041792
          1258
                 -0.021407
         Name: apple_close, Length: 1259, dtype: float64
In [87]: (closing_price['apple_close'] - closing_price['apple_close'].shift(1))/closing
                       NaN
Out[87]:
          1
                  1.042235
          2
                 -2.506658
          3
                 -0.190297
                 -0.089934
          1254
                  0.209043
          1255
                 -4.339015
                 -2.498442
          1256
          1257
                  4.179181
          1258
                 -2.140710
         Name: apple_close, Length: 1259, dtype: float64
          for col in closing_price.columns:
In [88]:
              closing_price[col + '_pct_change'] = (closing_price[col] - closing_price[col]
In [89]:
          closing_price
```

Out[89]:		apple_close	amzn_close	google_close	msft_close	apple_close_pct_change	amzn_close
	0	67.8542	261.95	558.46	27.55	NaN	
	1	68.5614	257.21	559.99	27.86	1.042235	
	2	66.8428	258.70	556.97	27.88	-2.506658	
	3	66.7156	269.47	567.16	28.03	-0.190297	
	4	66.6556	269.24	567.00	28.04	-0.089934	
	•••						
	1254	167.7800	1390.00	NaN	94.26	0.209043	
	1255	160.5000	1429.95	NaN	91.78	-4.339015	
	1256	156.4900	1390.00	NaN	88.00	-2.498442	
	1257	163.0300	1442.84	NaN	91.33	4.179181	
	1258	159.5400	1416.78	NaN	89.61	-2.140710	

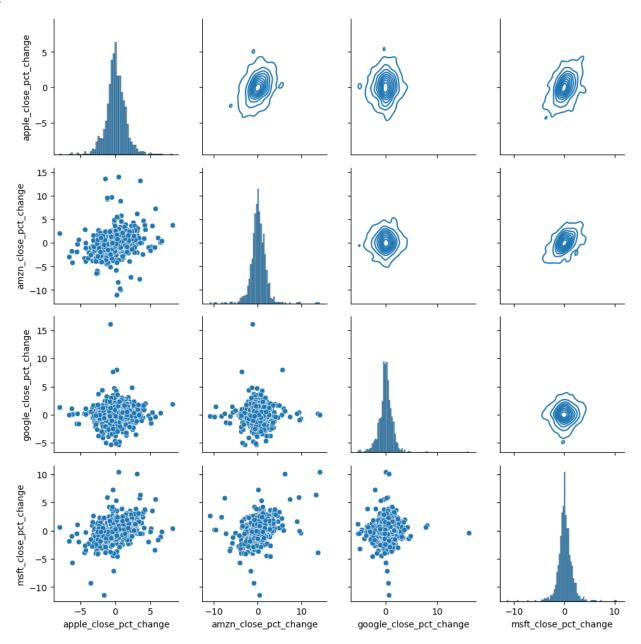
1259 rows × 8 columns

```
In [90]:
          closing_price.columns
          Index(['apple_close', 'amzn_close', 'google_close', 'msft_close',
Out[90]:
                 'apple_close_pct_change', 'amzn_close_pct_change',
'google_close_pct_change', 'msft_close_pct_change'],
                dtype='object')
          In [91]:
In [92]:
          clsing p
Out [92]:
                apple_close_pct_change amzn_close_pct_change google_close_pct_change msft_close_
             0
                                 NaN
                                                       NaN
                                                                              NaN
             1
                             1.042235
                                                  -1.809506
                                                                          0.273968
             2
                            -2.506658
                                                   0.579293
                                                                         -0.539295
             3
                            -0.190297
                                                   4.163123
                                                                          1.829542
             4
                            -0.089934
                                                  -0.085353
                                                                          -0.028211
          1254
                             0.209043
                                                  -4.196734
                                                                              NaN
          1255
                            -4.339015
                                                    2.874101
                                                                              NaN
          1256
                            -2.498442
                                                  -2.793804
                                                                              NaN
          1257
                                                   3.801439
                                                                              NaN
                             4.179181
          1258
                            -2.140710
                                                   -1.806160
                                                                              NaN
```

1259 rows × 4 columns

```
In [93]: g = sns.PairGrid(data = clsing_p)
    g.map_diag(sns.histplot)
    g.map_lower(sns.scatterplot)
    g.map_upper(sns.kdeplot)
```

Out[93]: <seaborn.axisgrid.PairGrid at 0x15a91ce90>

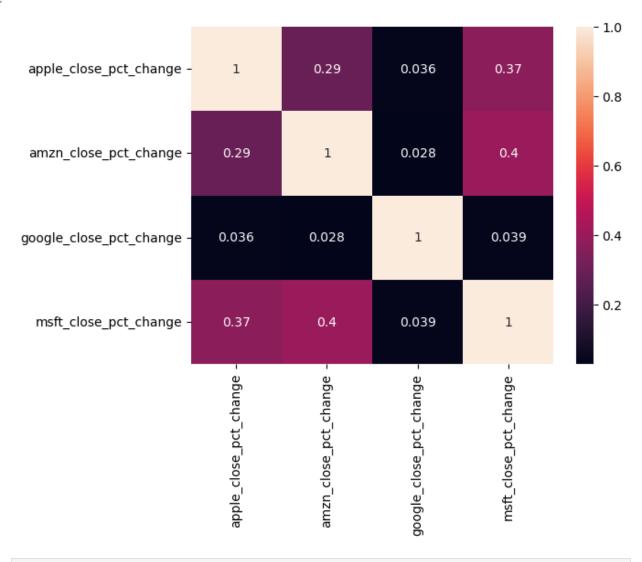


In [94]: clsing_p.corr()

Out[94]:		apple_close_pct_change	amzn_close_pct_change	google_close_pct_c
	apple_close_pct_change	1.000000	0.287659	0.0
	amzn_close_pct_change	0.287659	1.000000	0.0
	google_close_pct_change	0.036202	0.027698	1.0
	msft_close_pct_change	0.366598	0.402678	0.0

In [95]: sns.heatmap(clsing_p.corr(), annot= True)

Out[95]: <Axes: >



In []: