Part 2 — Workshop 5

TECH2: Introduction to Programming, Data, and Information Technology

Richard Foltyn Norwegian School of Economics (NHH)

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See GitHub repository for notebooks and data:

https://github.com/richardfoltyn/TECH2-H24

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1 Exercise: Daily returns of US stock market indices

In this exercise, we examine how the three major US stock market indices performed this year. Use the data in indices.csv from the folder ../data/stockmarket to solve the following tasks:

1. Load the CSV data and restrict the sample to the calendar year 2024.

Hint: You have to first figure out which character is used to separate individual columns.

Hint: You should use pd.read_csv(..., parse_dates=['Date']) to automatically parse strings stored in the Date column as dates.

2. The data comes in a "long" format where each date-ticker combination is stored in a separate row. For the following analysis, the data needs to be reshaped such that each ticker symbol is in its own column ("wide" format).

Use the pivot() method to reshape the DataFrame into the desired form. Consult the user guide to see a graphical illustration.

Your pivoted DataFrame should look as follows:

```
Ticker DJIA NASDAQ SP500
Date
2024-01-02 37715.0 14765.9 4742.8
2024-01-03 37430.2 14592.2 4704.8
```

- 3. Create a line plot which shows the time series for each of the three ticker symbols in a single graph, for example using DataFrame.plot().
- 4. The graph you just created is not very satisfactory as the three indices are recorded on vastly different scales. Express each index relative to its initial value in 2024 (so that all three start with the value 1.0) and recreate the previous graph with this normalized data.

- 5. Another way to check whether the three indices are co-moving is to compute and plot their daily returns. For each index, compute the daily returns, i.e., the relative change vs. the previous closing price in percent (e.g., using pct_change()), and plot the three time series of daily returns in a single graph.
- 6. The previous plots suggests that the three indices co-move a lot.
 - 1. In order to quantify the strength of this co-movement, compute the pairwise correlations of daily returns using corr().
 - 2. Create a figure with 3-by-3 subplots where each off-diagonal plot shows the scatter plot for two of the indices.
 - Hint: You can use the function scatter_matrix() to accomplish this task. Alternatively, you can create a figure with 3-by-3 subplots using Matplotlib's subplots(), iterate over all rows and columns and add a scatter() plot to each axes object.
 - 3. Add the correlation coefficient for each stockmarket index pair as text to the corresponding subplot in the figure you created (this can be done using text()).

2 Exercise: Long-run returns of the S&P 500

For this question, we restrict our attention to the S&P 500 and examine its long-run annual returns. Use the data from SP500.csv from the folder .../data/stockmarket to solve the following tasks:

- 1. Load the CSV data and set the Date column as the index.
 - Hint: You should use pd.read_csv(..., parse_dates=['Date']) to automatically parse strings stored in the Date column as dates.
- 2. You are interested in computing annual returns, but the data you just loaded contains daily observations. For each calendar year, select the first observation to get an annualized time series.
 - *Hint*: Use the resample() method to aggregate by year.
- 3. Compute the annual returns of the S&P 500, defined as the relative change from the previous-year's price on the first trading day (e.g., using pct_change()). Plot the resulting annual return time series.
- 4. What are the mean and standard deviation of annual returns? Indicate the average annual returns over this period using a horizontal line in the plot you previously created.

3 Exercise: Daily returns of Apple, Nvidia and Tesla

In this exercise, you are asked to compute various statistics of daily returns for selected stocks and plot these as bar charts. You'll be using data from files located in the folder .../data/stockmarket.

1. Load the price data for Apple Inc (AAPL.csv), Nvidia Inc (NVDA.csv) and Tesla Inc (TSLA.csv). Set the Date column as the index and merge these data sets into a single DataFrame, keeping only those dates for which you have observations on all three stocks. Restrict your data to the calendar years 2015-2023.

Hint: All three time series have a Price column which you should rename to the respective ticker symbol (e.g., AAPL) before merging the data to avoid overlapping column names.

Hint: You should use pd.read_csv(..., parse_dates=['Date']) to automatically parse strings stored in the Date column as dates.

- 2. Compute the average daily returns for each of the three stocks, e.g., using pct_change(). For simplicity, we assume that daily returns are defined as the percentage change vs. the previous observation even though these may be multiple days apart.
 - Compute the average daily returns, the standard deviation of daily returns, and the Sharpe ratio of daily returns by calendar year (using resample()). For this exercise, we define Sharpe ratio as the ratio of average returns divided by their standard deviation.
- 3. Create three figures visualizing these statistics. Each figure should contain a bar chart that shows one of the statistics (mean, std. deviation, Sharpe ratio) for all three companies by calendar year. Properly label your figures (using labels for x-axis, y-axis and titles), and add a horizontal line at zero for the figures depicting average returns and the Sharpe ratio.