

# Exercise Session - May 22, 2024

Python Lab – A.Y. 2023/2024

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## Exercise 1

Consider the tickers:

GOOG, AMZN, and TSLA.

1. Download from Yahoo! finance a joint DataFrame of closing prices for the last 1 year.
2. Build a portfolio with weights  $w_1 = 0.2$ ,  $w_2 = 0.3$ ,  $w_3 = 0.5$  and initial investment equal to \$100000.
3. Compute the return rates as

$$u_i^j = \frac{S_i^j - S_{i-1}^j}{S_{i-1}^j} = \frac{S_i^j}{S_{i-1}^j} - 1 \quad \text{for } j = 1, 2, 3.$$

4. Determine the mean vector  $\underline{\mu}$  and the variance-covariance matrix  $\Sigma$  of the return rates.
5. Compute the mean and standard deviation of the portfolio.
6. Compute the (absolute) VaR of the portfolio for a 2 days horizon and a confidence level of 95%.
7. Extract the return rates of GOOG and AMZN, and save them in a csv file named 'GARR.csv', *not saving the index as a new column*.

## Exercise 2

Load the file 'GARR.csv' in a DataFrame named `return_rates`.

1. Draw the plot and histogram of return rates of GOOG and AMZN using the DataFrame built-in methods.
2. Compute the parameters  $a, b$  of the regression line  $y = ax + b$  of GOOG return rates (seen as  $y$  variable) over AMZN return rates (seen as  $x$  variable) and plot the regression line.
3. Define a function named `prediction(x, a, b)` that returns the prediction of the regression line and call it on  $x = 0$ , with  $a$  and  $b$  estimated in previous point.

## Exercise 3

Consider the one-period three-state market whose payoffs at time  $T = 1$  are

$$\begin{array}{ccccc} & B_1 & S_1^1 & S_1^2 & \\ \text{Payoffs: } M_1 = & \begin{bmatrix} 10 & 20 & 5 \\ 10 & 30 & 5 \\ 10 & 5 & 10 \end{bmatrix} \end{array}$$

1. Build the market.
2. Verify that the market is complete and there are no redundant assets.
3. Define the payoff  $C_1 = \max\{S_1^1 - K, 0\}$  of a call option on  $S_1^1$  with strike price  $K = 20$ .
4. Compute the replicating portfolio  $\theta$  of the payoff of the call.
5. Compute the "best hedge" portfolio  $\theta_{\text{best}}$  through ordinary least-square regression, and the Mean Absolute Error (MAE) between  $\theta$  and  $\theta_{\text{best}}$ .