Assignment 5.

NB: The code for the assignment has to be developed in Python, using the IDE PyCharm by JetBrains.

0. Exercise. Variance-covariance method for VaR & ES in linear portfolio: a simple example of data mining

At the end of the 20th of March 2019 for an equally weighted equity portfolio with *Adidas*, *Allianz*, *Munich Re* and *L'Oréal*. Compute daily VaR and ES with a 3y estimation using the dataset provided via Gaussian parametric approach. The notional of the portfolio is \in 10 MIO and the significance level is α =0.95%.

[Hint: Pay attention that the trading days of the different stocks are not the same. Add previous day value in case of missing share price.]

1. Case study: Historical (HS & WHS) Simulation, Bootstrap and PCA for VaR & ES in a linear portfolio

At the end of the 20^{th} of March 2019 an asset manager due to different rules on three company units (portfolios) has to compute risk measures according to the following rules (α =0.99%):

- A. Portfolio 1 with: *Total* (25K shares), *Danone* (20K Shares), *Sanofi* (20K Shares), *Volkswagen* (10K Shares). Compute daily VaR and ES with a 3y estimation using the dataset provided via a Historical Simulation approach and a Bootstrap method with 200 simulations.
- B. Portfolio 2 with equally weighted equity: *Adidas*, *Airbus*, *BBVA*, *BMW* and *Schneider*. Compute daily VaR and ES with a 3y estimation using the dataset provided via a Weighted Historical Simulation approach with lambda = 0.97.
- C. Portfolio 3. An equally weighted equity portfolio with shares of the first 20 companies in the provided csv file "_indexes.csv" (leave out *Adyen* due to missing data and take *Eni*, the 21st, instead). Compute 10 days VaR and ES with a 3y estimation using the dataset provided via a Gaussian parametric PCA approach using the first n principal components, with the parameter n =1,...6. Comment the results.

For the three portfolios he checks results' order of magnitude via a Plausibility check.

2. Exercise: Full Monte-Carlo and Delta normal VaR

At the end of the 31st of January 2023 consider a portfolio formed by stocks of *Vonovia* for 25,870,000 Euro and the same <u>number</u> of put options with expiry on the 5th of April 2023, with strike 25 Euro and volatility equal to 15.4% (dividend yield of 3.1%). Compute a 10dd/99% VaR via a Full Monte-Carlo and a Delta normal approaches (only delta term). Use a 2y Historical Simulation for the underlying.

Can you improve the Delta normal VaR? How?

Why the Full Monte-Carlo can be numerical intensive for an <u>exotic</u> derivative that cannot be priced via a closed formula?

3. Case Study: Pricing in presence of counterparty risk

On the 31st of January 2023 at 10:45 C.E.T. bank XX buys form ISP a 4y Cliquet option for a 50 MIO € notional. Option yearly payoff (annual bond) at payment date is

$$[S(t_i) - S(t_{i-1})]^+$$
.

The option is on an equity stock (with no dividends) and constant volatility 25%.

What should be the correct price? At what price ISP would try to sell it?

[Hint: consider the dynamics of the underlying not of the corresponding forward]

Function signatures

[ES, Var] = AnalyticalNormalMeasures(alpha, weights, portfolioValue, riskMeasureTimeIntervalInDay, returns)

[ES, VaR] = HSMeasurements(returns, alpha, weights, portfolioValue, riskMeasureTimeIntervalInDay) samples = bootstrapStatistical(numberOfSamplesToBootstrap, returns)

[ES, VaR] = WHSMeasurements(returns, alpha, lambda, weights, portfolioValue, riskMeasureTimeIntervalInDay)

[ES, VaR] = PrincCompAnalysis(yearlyCovariance, yearlyMeanReturns, weights, H, alpha, numberOfPrincipalComponents, portfolioValue)

VaR = plausibilityCheck(returns, portfolioWeights, alpha, portfolioValue, riskMeasureTimeIntervalInDay)

VaR = FullMonteCarloVaR(logReturns, numberOfShares, numberOfPuts, stockPrice, strike, rate, dividend, volatility, timeToMaturityInYears, riskMeasureTimeIntervalInYears, alpha,NumberOfDaysPerYears)

VaR = DeltaNormalVaR(logReturns, numberOfShares, numberOfPuts, stockPrice, strike, rate, dividend, volatility, timeToMaturityInYears, riskMeasureTimeIntervalInYears, alpha, NumberOfDaysPerYears)

Delivery date: Friday 31st of March at 14:00.

Delivery date for correctors: Tuesday 4th of April at 23:00.

Corrector groups: 1,6,7,14.

Useful Python packages:

numpy for vector and matrix management pandas for working with dataframes

scipy (and scipy.stats) for PDFs, CDFs and quantile functions

datetime for dates management