Advanced Risk Management

Computer Exercise Week 1

In this exercise, we compare the RiskMetrics VaR to the (Weighted) Historical Simulation approach. We compare the time patterns of the three VaR measures applied to S&P 500 index returns in the period January 2001 through December 2010. Next, we evaluate the three approaches using backtests.

Efficient programming of the Weighted Historical Simulation method requires a function that can deliver weighted percentiles from a sample. The code for this has been found on the web, and imported in the Jupyter notebook ComputerExercise1.ipynb, which also loads the necessary data (using the pandas DataReader connected to Yahoo! Finance; this does not always work properly, and alternatively the data can be imported from SP500.csv). The notebook will need to be completed for this exercise.

- 1. Calculate the 1%, one-day VaR for S&P 500 index returns for each day in the evaluation period (January 2001 through December 2010), using each of the three methods, and keep the result in three vectors / arrays. The scripts already contain the definition of a weight vector w, based on a historical period m=250, and a parameter $\eta=\lambda=0.94$ for the WHS and RM methods. Note that WHS and the σ_{t+1} sequence in RM need a start-up sample period, for which we use data from the year 2000 (252 observations). Note: an easy way to implement the RM method to construct σ_t^2 is to use the series .ewm()
 - Note: an easy way to implement the RM method to construct σ_t^2 is to use the series.ewm() function from pandas. Note that in that function, $\alpha = 1 \lambda$, and also that you need to apply this to the *lagged* squared return (so using the series.shift() function).
- 2. Make a figure where each of the three VaR measures are plotted against time. Discuss the similarities and differences.
- 3. Investigate the effect of changing the η parameter in the WHS method: what happens if we give η a value very close to 1, e.g. 0.999?
- 4. Construct the hit sequences $I_{t+1} = \mathbb{I}(R_{t+1} < -VaR_{t+1}^{0.01})$, for each of the three VaR methods. Next, test unconditional coverage and independence using the methods described in the slides (you may also use Christoffersen's LR tests for comparison). What do you conclude?

 $^{{}^{1}\}mathbb{I}(\cdot)$ is the indicator function, i.e., $\mathbb{I}(A)=1$ if A is true and 0 otherwise.