

HW Problems for Assignment 2 - Part 2

Due 6:30 PM Tuesday, October 5th, 2021

- 1. (10 Points, 5 Points Each) Practice with ES.** Explicitly compute $\text{VaR}_\alpha(L)$ and $\text{ES}_\alpha(L)$, assuming L has the following distributions.
- $L \sim \text{Exp}(1/\theta)$ is exponentially distributed with mean θ .
 - $L \sim \text{Pareto}(1, 1/\xi)$ is Pareto distributed with unit scale and $1/\xi$ shape parameter for $\xi > 0$. Here, L has CDF

$$F(\tau) = \begin{cases} 0 & \tau \leq 1 \\ 1 - \tau^{-\frac{1}{\xi}} & \tau > 1 \end{cases}.$$

How does your answer depend on ξ ?

- 2. (30 Points) Time aggregated risk measures for a constant weight portfolio of equities.** In this exercise you will estimate various risk measures at a $K = 10$ day horizon for a portfolio of equities, assuming the portfolio weights are held constant over the K day horizon.

The portfolio consists of Boeing, McDonald's, Nike and Walmart stock. At the start (time $t = 9/1/2020$) the portfolio value is $V_t = \$1M$. The weights are determined using the time t market capitalizations from (in billions of dollars) of

Boeing : 97.39; McDonald's : 158.20; Nike : 179.01; Walmart : 417.97.

Historical prices from 8/31/2015 – 8/31/2020 are in the file “Prices.csv”. The first column is the date (in Excel numeric format) while columns 2-5 give the stock price data. Data is sorted oldest to newest.

Our goal is to write a simulation to estimate the distribution of the K day losses $L_{t+K\Delta}$, and then use the distribution to estimate the risk measures. We work in the normal log-returns framework, using **full losses** and **EWMA** to estimate the conditional mean and covariance. However, as our time horizon is longer than 1 day, there are some subtleties when writing the simulation. Thus, complete the following steps:

- (5 Points)** (pen and paper problem) As an abuse of notation, for a vector $x = (x^{(1)}, \dots, x^{(d)})$ write e^x for the vector $(e^{x^{(1)}}, \dots, e^{x^{(d)}})$. Show that for constant weights, the full loss over the K day horizon is

$$L_{t+K\Delta} = -V_t \left(\prod_{k=1}^K w^T e^{X_{t+k\Delta}} - 1 \right).$$

- (15 Points)** Write the simulation to obtain the time-aggregated risk measures, building into your simulation that (1) weights are held constant and (2) we use EWMA *each day* to obtain a new conditional mean and covariance estimate.

As for (2), use the historical data to obtain an estimate $\mu_{t+\Delta}, \Sigma_{t+\Delta}$ as of the initial time t . Since we have 5 years of data (about 1200 points) it is OK to start the EWMA procedure off with $\mu_{t_0} = 0 = \Sigma_{t_0}$.

Write $T_k = t+k\Delta$. From the historical data we have obtained μ_{T_1}, Σ_{T_1} and can sample $X_{T_1} \sim N(\mu_{T_1}, \Sigma_{T_1})$. Next, for $k = 2, \dots, K$ write your simulation to include updating the mean and covariance estimates! Indeed, once we have sampled X_{T_1} we obtain our next estimate by setting

$$\begin{aligned}\mu_{T_2} &= \lambda\mu_{T_1} + (1 - \lambda)X_{T_1}; \\ \Sigma_{T_2} &= \lambda\Sigma_{T_1} + (1 - \theta)(X_{T_1} - \mu_{T_1})(X_{T_1} - \mu_{T_1})^T.\end{aligned}$$

We then sample $X_{T_2} \sim N(\mu_{T_2}, \Sigma_{T_2})$ and obtain μ_{T_3}, Σ_{T_3} by EWMA accordingly. We repeat this over the K day horizon.

With this methodology, write a simulation to estimate the K day VaR and ES. To compare with square root of time, estimate the risk measures for both a $K = 1$ day and $K = 10$ day horizon. Output the ten day estimates as well as the square root of time approximation found by multiplying the one day estimates by \sqrt{K} . For parameter values take $\alpha = .99$, $\gamma = 30$, $\lambda = .94$, $\theta = .97$. Have your simulations perform $N = 50,000$ trials.

For error checking purposes, I am obtaining around \$93,000 and \$111,000 for the VaR and ES risk measure respectively.