HW Problems for Assignment 2 - Part 2

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

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

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:

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

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

(b) (25 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, ..., 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

$$\mu_{T_2} = \lambda \mu_{T_1} + (1 - \lambda) X_{T_1};$$

$$\Sigma_{T_2} = \lambda \Sigma_{T_1} + (1 - \theta) (X_{T_1} - \mu_{T_1}) (X_{T_1} - \mu_{T_1})^{\mathrm{T}}.$$

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.