

# Quantitative Risk Management

## Assignment 9 Solutions

December 3, 2019

**Question 1:** Part 1) Selecting the sequences  $c_n = \frac{\log(n)}{\beta}$  and  $d_n = \frac{1}{\beta}$  and setting  $M_n = \max\{X_1, \dots, X_n\}$ , we compute:

$$\begin{aligned}
 \mathbb{P}\left(\frac{M_n - c_n}{d_n} \leq x\right) &= \mathbb{P}\left(M_n \leq d_n x + c_n\right) \\
 &= F^n(d_n x + c_n) \\
 &= \left(1 - e^{-\beta(d_n x + c_n)}\right)^n \\
 &= \left(1 - e^{-\beta\left(\frac{x}{\beta} + \frac{\log(n)}{\beta}\right)}\right)^n \\
 &= \left(1 - \frac{1}{n}e^{-x}\right)^n
 \end{aligned}$$

Taking the limit as  $n \rightarrow \infty$  gives us the limiting distribution:

$$H(x) = \exp\{-e^{-x}\}$$

which is a GEV distribution with parameter  $\xi = 0$ .

Part 2) We now select the sequences  $c_n = \kappa n^{1/\alpha} - \kappa$  and  $d_n = \frac{\kappa n^{1/\alpha}}{\alpha}$  and again compute:

$$\begin{aligned}
 \mathbb{P}\left(\frac{M_n - c_n}{d_n} \leq x\right) &= \mathbb{P}\left(M_n \leq d_n x + c_n\right) \\
 &= F^n(d_n x + c_n) \\
 &= \left(1 - \left(\frac{\kappa}{\kappa + d_n x + c_n}\right)^\alpha\right)^n \\
 &= \left(1 - \left(\frac{\kappa}{\kappa + \frac{\kappa n^{1/\alpha}}{\alpha} x + \kappa n^{1/\alpha} - \kappa}\right)^\alpha\right)^n \\
 &= \left(1 - \frac{1}{n} \left(1 + \frac{x}{\alpha}\right)^{-\alpha}\right)^n
 \end{aligned}$$

Taking the limit  $n \rightarrow \infty$  gives us the limiting distribution:

$$H(x) = \exp\left\{-\left(1 + \frac{x}{\alpha}\right)^{-\alpha}\right\}$$

which is a GEV distribution with  $\xi = \frac{1}{\alpha}$ .

**Question 2:** If  $X$  has excess distribution over the threshold  $u$  of  $F_u = G_{\xi,\beta}$ , this means:

$$\mathbb{P}(X - u \leq x | X > u) = F_u(x) = G_{\xi,\beta}(x)$$

We now compute:

$$\begin{aligned} \mathbb{P}(X - v \leq x | X > v) &= 1 - \mathbb{P}(X - v > x | X > v) \\ &= 1 - \frac{\mathbb{P}(X > x + v)}{\mathbb{P}(X > v)} \\ &= 1 - \frac{\mathbb{P}(X > x + v)}{\mathbb{P}(X > u)} \frac{\mathbb{P}(X > u)}{\mathbb{P}(X > v)} \\ &= 1 - \frac{\mathbb{P}(X - u > x + v - u)}{\mathbb{P}(X > u)} \frac{\mathbb{P}(X > u)}{\mathbb{P}(X - u > v - u)} \\ &= 1 - \mathbb{P}(X - u > x + v - u | X > u) \frac{1}{\mathbb{P}(X - u > v - u | X > u)} \\ &= 1 - \frac{1 - \mathbb{P}(X - u \leq x + v - u | X > u)}{1 - \mathbb{P}(X - u \leq v - u | X > u)} \\ &= 1 - \frac{1 - G_{\xi,\beta}(x + v - u)}{1 - G_{\xi,\beta}(v - u)} \\ &= 1 - \left( \frac{1 + \frac{\xi(x+v-u)}{\beta}}{1 + \frac{\xi(v-u)}{\beta}} \right)^{-\frac{1}{\xi}} \\ &= 1 - \left( \frac{\beta + \xi(x + v - u)}{\beta + \xi(v - u)} \right)^{-\frac{1}{\xi}} \\ &= 1 - \left( 1 + \frac{\xi x}{\beta + \xi(v - u)} \right)^{-\frac{1}{\xi}} \\ &= G_{\xi,\beta+(v-u)\xi}(x) \end{aligned}$$

**Question 3:** Part 1) The negative log returns are computed, and the sample mean excess function defined by:

$$e_n(v) = \frac{\sum_{i=1}^n (X_i - v) \mathbb{1}_{X_i > v}}{\sum_{i=1}^n \mathbb{1}_{X_i > v}}$$

is plotted at the points  $X_1, \dots, X_n$  to give the result in Figure 1.

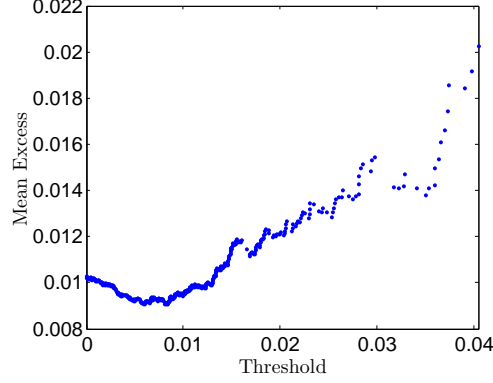


Figure 1: Sample mean excess plot for MSFT data.

Part 2) In performing a GPD fit to the data, a threshold of  $u = 0.01$  is selected. The data is transformed into  $\hat{X}_i = X_i - u$ . This leaves  $N_u$  points that are positive, which are the only ones we consider, and we relabel the data as  $\hat{X}_1, \dots, \hat{X}_{N_u}$ . We assume that  $\hat{X}_1, \dots, \hat{X}_{N_u}$  come from a GPD distribution and estimate the parameters  $\xi$  and  $\beta$  by maximizing the log likelihood function:

$$\log(L(\xi, \beta; \hat{X}_1, \dots, \hat{X}_{N_u})) = \sum_{i=1}^{N_u} \log(g_{\xi, \beta}(\hat{X}_i))$$

The resulting estimates are:

$$\hat{\xi} = 0.2171$$

$$\hat{\beta} = 0.0075$$