

MATH48181/68181: EXTREME VALUES AND FINANCIAL RISK
SEMESTER 1
SOLUTIONS TO QUIZ PROBLEM 8

Let X represent loss and suppose x_1, x_2, \dots, x_n is a random sample on X . Suppose that the probability density function of X is given by

$$f(x) = \exp(a - x)$$

where $x > a > 0$.

First, we need to determine the cumulative distribution function of X as follows

$$\begin{aligned} F(x) &= \int_a^x f(y) dy \\ &= \int_a^x \exp(a - y) dy \\ &= \exp(a) \int_a^x \exp(-y) dy \\ &= \exp(a) [-\exp(-y)]_a^x \\ &= \exp(a) [\exp(-a) - \exp(-x)] \\ &= 1 - \exp(a - x). \end{aligned}$$

Hence, the cumulative distribution function of X is

$$F(x) = 1 - \exp(a - x).$$

Next, we invert $F(x) = p$ to find the value at risk at p . This gives

$$\text{VaR}_p(X) = a - \log(1 - p).$$

Next, we find the expected shortfall at p as follows

$$\begin{aligned} \text{ES}_p(X) &= \frac{1}{p} \int_0^p [a - \log(1 - t)] dt \\ &= \frac{1}{p} \left[\int_0^p a dt - \int_0^p \log(1 - t) dt \right] \\ &= \frac{1}{p} \left[ap - \int_0^p \log(1 - t) dt \right] \\ &= a - \frac{1}{p} \int_0^p \log(1 - t) dt \\ &= a - \frac{1}{p} \left\{ [t \log(1 - t)]_0^p + \int_0^p \frac{t}{1 - t} dt \right\} \end{aligned}$$

$$\begin{aligned}
&= a - \frac{1}{p} \left\{ [p \log(1-p) - 0] + \int_0^p \frac{t-1+1}{1-t} dt \right\} \\
&= a - \frac{1}{p} \left\{ p \log(1-p) + \int_0^p \left[-1 + \frac{1}{1-t} \right] dt \right\} \\
&= a - \frac{1}{p} \{ p \log(1-p) + [-t - \log(1-t)]_0^p \} \\
&= a - \frac{1}{p} \{ p \log(1-p) + [-p - \log(1-p) - 0] \} \\
&= a - \frac{1}{p} \{ p \log(1-p) - p - \log(1-p) \}.
\end{aligned}$$

Finally, to find the maximum likelihood estimator of $ES_p(X)$, we need to find the maximum likelihood estimator of a . The likelihood function of a is

$$\begin{aligned}
L(a) &= \prod_{i=1}^n [\exp(a - x_i) I\{x_i > a\}] \\
&= \exp\left(na - \sum_{i=1}^n x_i\right) \prod_{i=1}^n [I\{x_i > a\}] \\
&= \exp\left(na - \sum_{i=1}^n x_i\right) I\{\min(x_1, x_2, \dots, x_n) > a\}.
\end{aligned}$$

We see that $L(a)$ is an increasing function of a over $(-\infty, \min(x_1, x_2, \dots, x_n))$ and is zero elsewhere. Hence, the maximum likelihood estimator of a is $\min(x_1, x_2, \dots, x_n)$.

Hence, the maximum likelihood estimator of $ES_p(X)$ is

$$\min(x_1, x_2, \dots, x_n) - \frac{1}{p} \{ p \log(1-p) - p - \log(1-p) \}.$$