

**MATH4/68181: Extreme values and financial risk**  
**Semester 1**  
**Solutions to problem sheet for Week 4**

1. The cdf of  $X$  is

$$\begin{aligned}
 F_X(x) &= \Pr(X \leq x) \\
 &= \Pr(\max(X_1, \dots, X_\alpha) \leq x) \\
 &= \Pr(X_1 \leq x, \dots, X_\alpha \leq x) \\
 &= \Pr(X_1 \leq x) \cdots \Pr(X_\alpha \leq x) \\
 &= [1 - \exp(-\lambda x)] \cdots [1 - \exp(-\lambda x)] \\
 &= [1 - \exp(-\lambda x)]^\alpha.
 \end{aligned}$$

2. Differentiating

$$F_X(x) = [1 - \exp(-\lambda x)]^\alpha$$

with respect to  $x$  gives the pdf of  $X$  as

$$f_X(x) = \alpha \lambda \exp(-\lambda x) [1 - \exp(-\lambda x)]^{\alpha-1}.$$

3. The  $n$ th moment of  $X$  can be calculated as

$$\begin{aligned}
 E(X^n) &= \alpha \lambda \int_0^\infty x^n \exp(-\lambda x) [1 - \exp(-\lambda x)]^{\alpha-1} dx \\
 &= (-1)^n \alpha \lambda^{-n} \int_0^1 (\log y)^n [1 - y]^{\alpha-1} dy \\
 &= (-1)^n \alpha \lambda^{-n} \frac{\partial^n}{\partial \beta^n} \int_0^1 y^\beta [1 - y]^{\alpha-1} dy \Big|_{\beta=0} \\
 &= (-1)^n \alpha \lambda^{-n} \frac{\partial^n}{\partial \beta^n} B(\beta + 1, \alpha) \Big|_{\beta=0}.
 \end{aligned}$$

4. The mean of  $X$  can be calculated as

$$\begin{aligned}
 E(X) &= -\alpha \lambda^{-1} \frac{\partial}{\partial \beta} B(\beta + 1, \alpha) \Big|_{\beta=0} \\
 &= -\alpha \lambda^{-1} \left\{ \frac{\Gamma(\alpha) \Gamma'(\beta + 1)}{\Gamma(\alpha + \beta + 1)} - \frac{\Gamma(\alpha) \Gamma(\beta + 1) \Gamma'(\alpha + \beta + 1)}{\Gamma^2(\alpha + \beta + 1)} \right\} \Big|_{\beta=0} \\
 &= -\alpha \lambda^{-1} \left\{ \frac{\Gamma'(1)}{\alpha} - \frac{\Gamma(\alpha) \Gamma'(\alpha + 1)}{\Gamma^2(\alpha + 1)} \right\}.
 \end{aligned}$$

5. Since

$$E(X^2) = -\alpha \lambda^{-2} \frac{\partial^2}{\partial \beta^2} B(\beta + 1, \alpha) \Big|_{\beta=0}$$

$$\begin{aligned}
&= -\alpha\lambda^{-2} \left\{ \frac{\Gamma(\alpha)\Gamma''(\beta+1)}{\Gamma(\alpha+\beta+1)} - \frac{2\Gamma(\alpha)\Gamma'(\beta+1)\Gamma'(\alpha+\beta+1)}{\Gamma^2(\alpha+\beta+1)} \right. \\
&\quad \left. - \frac{\Gamma(\alpha)\Gamma(\beta+1)\Gamma''(\alpha+\beta+1)}{\Gamma^2(\alpha+\beta+1)} + \frac{2\Gamma(\alpha)\Gamma(\beta+1)\left(\Gamma'(\alpha+\beta+1)\right)^2}{\Gamma^3(\alpha+\beta+1)} \right\}_{\beta=0} \\
&= -\alpha\lambda^{-2} \left\{ \frac{\Gamma''(1)}{\alpha} - \frac{2\Gamma(\alpha)\Gamma'(1)\Gamma'(\alpha+1)}{\Gamma^2(\alpha+1)} \right. \\
&\quad \left. - \frac{\Gamma(\alpha)\Gamma''(\alpha+1)}{\Gamma^2(\alpha+1)} + \frac{2\Gamma(\alpha)\left(\Gamma'(\alpha+1)\right)^2}{\Gamma^3(\alpha+1)} \right\},
\end{aligned}$$

the variance of  $X$  is

$$\begin{aligned}
Var(X) &= -\alpha\lambda^{-2} \left\{ \frac{\Gamma''(1)}{\alpha} - \frac{2\Gamma(\alpha)\Gamma'(1)\Gamma'(\alpha+1)}{\Gamma^2(\alpha+1)} \right. \\
&\quad \left. - \frac{\Gamma(\alpha)\Gamma''(\alpha+1)}{\Gamma^2(\alpha+1)} + \frac{2\Gamma(\alpha)\left(\Gamma'(\alpha+1)\right)^2}{\Gamma^3(\alpha+1)} \right\} - E^2(X).
\end{aligned}$$

6. Setting

$$[1 - \exp(-\lambda x)]^\alpha = p$$

gives

$$\text{VaR}_p(X) = -\frac{1}{\lambda} \log(1 - p^{1/\alpha}).$$

7. The expected shortfall is

$$\begin{aligned}
\text{ES}_p(X) &= -\frac{1}{\lambda p} \int_0^p \log(1 - v^{1/\alpha}) dv \\
&= \frac{1}{\lambda p} \int_0^p \sum_{i=0}^{\infty} \frac{v^{i/\alpha}}{i} dv \\
&= \frac{1}{\lambda p} \sum_{i=0}^{\infty} \int_0^p \frac{v^{i/\alpha}}{i} dv \\
&= \frac{1}{\lambda p} \sum_{i=0}^{\infty} \frac{p^{i/\alpha+1}}{i(i/\alpha+1)}.
\end{aligned}$$

8. The likelihood function is

$$L(\alpha, \lambda) = \alpha^n \lambda^n \exp \left( -\lambda \sum_{i=1}^n x_i \right) \left\{ \prod_{i=1}^n [1 - \exp(-\lambda x_i)] \right\}^{\alpha-1}.$$

The log-likelihood function is

$$\log L = n \log(\alpha\lambda) - \lambda \sum_{i=1}^n x_i + (\alpha-1) \sum_{i=1}^n \log [1 - \exp(-\lambda x_i)].$$

The partial derivatives with respect to  $\alpha$  and  $\lambda$  are

$$\frac{\partial \log L}{\partial \alpha} = \frac{n}{\alpha} + \sum_{i=1}^n \log [1 - \exp(-\lambda x_i)]$$

and

$$\frac{\partial \log L}{\partial \lambda} = \frac{n}{\lambda} - \sum_{i=1}^n x_i + (\alpha - 1) \sum_{i=1}^n \frac{x_i \exp(-\lambda x_i)}{1 - \exp(-\lambda x_i)}.$$

Setting these to zero, we obtain the mle of  $\alpha$  as

$$\hat{\alpha} = -\frac{n}{\sum_{i=1}^n \log [1 - \exp(-\lambda x_i)]}.$$

The mle of  $\lambda$  is the root of the equation

$$\frac{n}{\lambda} - \sum_{i=1}^n x_i - \left( \frac{n}{\sum_{i=1}^n \log [1 - \exp(-\lambda x_i)]} + 1 \right) \sum_{i=1}^n \frac{x_i \exp(-\lambda x_i)}{1 - \exp(-\lambda x_i)}.$$