

**MATH4/68181: Extreme values and financial risk**  
**Semester 1**  
**Problem sheet for Week 1**

1. If  $x_1, x_2, \dots, x_n$  is a random sample from

$$f(x) = \sigma^{-1} \exp\left(-\frac{1}{\sigma}x\right) \exp\left\{-\exp\left(-\frac{x}{\sigma}\right)\right\}$$

find the mle of  $\sigma$ .

2. If  $x_1, x_2, \dots, x_n$  is a random sample from

$$f(x) = \lambda \sigma^\lambda x^{-\lambda-1} \exp\left(-\sigma^\lambda x^{-\lambda}\right)$$

find the mles of  $\lambda$  and  $\sigma$ .

3. If  $x_1, x_2, \dots, x_n$  is a random sample from

$$f(x) = \lambda \sigma^{-\lambda} x^{\lambda-1} \exp\left(-\sigma^{-\lambda} x^\lambda\right)$$

find the mles of  $\lambda$  and  $\sigma$ .

4. If  $x_1, x_2, \dots, x_n$  is a random sample from

$$f(x) = (1 - \lambda x)^{1/\lambda-1}$$

find the mle of  $\lambda$ .

1) Suppose  $x_1, \dots, x_n$  IID from

$$f(x) = \frac{1}{\sigma} e^{-\frac{x}{\sigma}} e^{-e^{-\frac{x}{\sigma}}}.$$

The likelihood function is

$$\begin{aligned} L(\sigma) &= \prod_{i=1}^n \left[ \frac{1}{\sigma} e^{-\frac{x_i}{\sigma}} e^{-e^{-\frac{x_i}{\sigma}}} \right] \\ &= \frac{1}{\sigma^n} e^{-\frac{1}{\sigma} \sum_{i=1}^n x_i} e^{-\sum_{i=1}^n e^{-\frac{x_i}{\sigma}}}. \end{aligned}$$

The log-likelihood function is

$$\log L(\sigma) = -n \log \sigma - \frac{1}{\sigma} \sum_{i=1}^n x_i - \sum_{i=1}^n e^{-\frac{x_i}{\sigma}}$$

The derivative wrt  $\sigma$  is

$$\frac{d \log L(\sigma)}{d \sigma} = -\frac{n}{\sigma} + \frac{1}{\sigma^2} \sum_{i=1}^n x_i - \frac{1}{\sigma^2} \sum_{i=1}^n x_i e^{-\frac{x_i}{\sigma}}$$

The MLE of  $\sigma$  is the root of

$$\frac{d \log L(\sigma)}{d \sigma} = 0$$

$$\Leftrightarrow n \sigma = \sum_{i=1}^n x_i - \sum_{i=1}^n x_i e^{-\frac{x_i}{\sigma}}$$

This must be solved numerically for  $\sigma$ .

2) Suppose  $x_1, \dots, x_n$  IID with PDF

$$f(x) = \lambda \sigma^\lambda x^{-\lambda-1} e^{-\sigma^\lambda x^{-\lambda}}.$$

The likelihood function

$$\begin{aligned} L(\lambda, \sigma) &= \prod_{i=1}^n \left[ \lambda \sigma^\lambda x_i^{-\lambda-1} e^{-\sigma^\lambda x_i^{-\lambda}} \right] \\ &= \lambda^n \sigma^{n\lambda} \left( \prod_{i=1}^n x_i \right)^{-\lambda-1} e^{-\sigma^\lambda \sum_{i=1}^n x_i^{-\lambda}}. \end{aligned}$$

The log-likelihood function is

$$\begin{aligned} \log L(\lambda, \sigma) &= n \log \lambda + n\lambda \log \sigma \\ &\quad - (\lambda+1) \sum_{i=1}^n \log x_i \\ &\quad - \sigma^\lambda \sum_{i=1}^n x_i^{-\lambda}. \end{aligned}$$

The partial derivative wrt  $\lambda$  and  $\sigma$  are

$$\begin{aligned} \frac{\partial \log L}{\partial \lambda} &= \frac{n}{\lambda} + n \log \sigma - \sum_{i=1}^n \log x_i - \sum_{i=1}^n \left( \frac{\sigma}{x_i} \right)^\lambda \log \left( \frac{\sigma}{x_i} \right) \\ &= 0 \quad \text{--- (1)} \end{aligned}$$

$$\frac{\partial \log L}{\partial \sigma} = \frac{n\lambda}{\sigma} - \lambda \sigma^{\lambda-1} \sum_{i=1}^n x_i^{-\lambda} = 0 \quad \text{--- (2)}$$

$$(2) \Rightarrow \frac{n}{\sigma^\lambda} = \sum_{i=1}^n x_i^{-\lambda}$$

$$\Rightarrow \sigma = \left[ \frac{n}{\sum_{i=1}^n x_i^{-\lambda}} \right]^{\frac{1}{\lambda}} \quad (3)$$

Substituting (3) into (1) gives

$$\frac{n}{\lambda} + n \log \sigma - \sum_{i=1}^n \log x_i$$

$$- \left[ \frac{n}{\sum_{i=1}^n x_i^{-\lambda}} \right] \sum_{i=1}^n x_i^{-\lambda} [\log \sigma - \log x_i] = 0$$

$$\Leftrightarrow \frac{n}{\lambda} + n \log \sigma - \sum_{i=1}^n \log x_i$$

$$- \left[ \frac{n}{\sum_{i=1}^n x_i^{-\lambda}} \right] \left[ \sum_{i=1}^n x_i^{-\lambda} \right] \log \sigma$$

$$+ \frac{n}{\sum_{i=1}^n x_i^{-\lambda}} \sum_{i=1}^n x_i^{-\lambda} \log x_i = 0$$

$$\Leftrightarrow \frac{n}{\lambda} + n \log \sigma - \sum_{i=1}^n \log x_i - n \log \sigma$$

$$+ \frac{n}{\sum_{i=1}^n x_i^{-\lambda}} \sum_{i=1}^n x_i^{-\lambda} \log x_i = 0$$

$$\Leftrightarrow \frac{n}{\lambda} = \sum_{i=1}^n \log x_i - n \frac{\sum_{i=1}^n x_i^{-\lambda} \log x_i}{\sum_{i=1}^n x_i^{-\lambda}} \quad (4)$$

Solve (4) to get  $\hat{\lambda}$ . Substitute  $\hat{\lambda}$  into (3) to

3) Suppose  $x_1, \dots, x_n$  are IID with PDF

$$f(x) = \lambda \sigma^{-\lambda} x^{\lambda-1} e^{-\sigma^{-\lambda} x^\lambda}$$

The likelihood function is

$$\begin{aligned} L(\lambda, \sigma) &= \prod_{i=1}^n \left[ \lambda \sigma^{-\lambda} x_i^{\lambda-1} e^{-\sigma^{-\lambda} x_i^\lambda} \right] \\ &= \lambda^n \sigma^{-n\lambda} \left( \prod_{i=1}^n x_i \right)^{\lambda-1} e^{-\sigma^{-\lambda} \sum_{i=1}^n x_i^\lambda} \end{aligned}$$

The log-likelihood is

$$\begin{aligned} \log L(\lambda, \sigma) &= n \log \lambda - n\lambda \log \sigma \\ &\quad + (\lambda-1) \sum_{i=1}^n \log x_i \\ &\quad - \sigma^{-\lambda} \sum_{i=1}^n x_i^\lambda \end{aligned}$$

The partial derivatives wrt  $\lambda$  and  $\sigma$  are

$$\begin{aligned} \frac{\partial \log L}{\partial \lambda} &= \frac{n}{\lambda} - n \log \sigma + \sum_{i=1}^n \log x_i - \sum_{i=1}^n \left( \frac{x_i}{\sigma} \right)^\lambda \log \left( \frac{x_i}{\sigma} \right) \\ &= 0 \quad (1) \end{aligned}$$

$$\frac{\partial \log L}{\partial \sigma} = -\frac{n\lambda}{\sigma} + \lambda \sigma^{-\lambda-1} \sum_{i=1}^n x_i^\lambda = 0 \quad (2)$$



$$(2) \Rightarrow n \sigma^\lambda = \sum_{i=1}^n x_i^\lambda$$

$$\Rightarrow \sigma = \left[ \frac{\sum_{i=1}^n x_i^\lambda}{n} \right]^{\frac{1}{\lambda}} \quad \text{--- (3)}$$

Substituting (3) into (1) gives

$$\frac{n}{\lambda} - n \log \sigma + \sum_{i=1}^n \log x_i - \left[ \frac{\sum_{i=1}^n x_i^\lambda}{n} \right]^{\frac{n}{\lambda}} \sum_{i=1}^n x_i^\lambda [\log x_i - \log \sigma] = 0$$

$$\Leftrightarrow \frac{n}{\lambda} - n \log \sigma + \sum_{i=1}^n \log x_i - \left[ \frac{\sum_{i=1}^n x_i^\lambda}{n} \right]^{\frac{n}{\lambda}} \sum_{i=1}^n x_i^\lambda \log x_i$$

$$+ \left[ \frac{\sum_{i=1}^n x_i^\lambda}{n} \right]^{\frac{n}{\lambda}} \left[ \sum_{i=1}^n x_i^\lambda \right] \log \sigma = 0$$

$$\Leftrightarrow \frac{n}{\lambda} - n \log \sigma + \sum_{i=1}^n \log x_i - n \frac{\sum_{i=1}^n x_i^\lambda \log x_i}{\sum_{i=1}^n x_i^\lambda} + n \log \sigma = 0$$

$$\Leftrightarrow \frac{n}{\lambda} + \sum_{i=1}^n \log x_i = n \frac{\sum_{i=1}^n x_i^\lambda \log x_i}{\sum_{i=1}^n x_i^\lambda} = 0 \quad \text{--- (4)}$$

Solve (4) to get  $\lambda$ . Substitute  $\lambda$  into (3) to get  $\sigma$ .

4) Suppose  $x_1, \dots, x_n$  IID with PDF

$$f(x) = (1 - \lambda x)^{\frac{1}{\lambda} - 1}.$$

The likelihood function is

$$\begin{aligned} L(\lambda) &= \prod_{i=1}^n (1 - \lambda x_i)^{\frac{1}{\lambda} - 1} \\ &= \left[ \prod_{i=1}^n (1 - \lambda x_i) \right]^{\frac{1}{\lambda} - 1}. \end{aligned}$$

The log-likelihood is

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

The derivative wrt  $\lambda$  is

$$\begin{aligned} \frac{d \log L(\lambda)}{d\lambda} &= -\lambda^{-2} \sum_{i=1}^n \log(1 - \lambda x_i) \\ &\quad - \left( \frac{1}{\lambda} - 1 \right) \sum_{i=1}^n \frac{x_i}{1 - \lambda x_i}. \end{aligned}$$

The MLE of  $\lambda$  is to root of

$$\frac{d \log L(\lambda)}{d\lambda} = 0$$

$$\Leftrightarrow \sum_{i=1}^n \log(1 - \lambda x_i) = \lambda(\lambda - 1) \sum_{i=1}^n \frac{x_i}{1 - \lambda x_i}.$$

This must be solved for  $\lambda$  numerically