

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

1) The likelihood function is

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

The log-likelihood function is

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

The derivative with respect to  $\sigma$  is

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

So, the mle of  $\sigma$  is the root of the equation

$$n\sigma = \sum_{i=1}^n x_i - \sum_{i=1}^n x_i \exp\left(-\frac{x_i}{\sigma}\right).$$

2) The likelihood function is

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

The log-likelihood function is

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

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

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

and

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

Setting these to zero and rearranging, we obtain the mle of  $\sigma$  as

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

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

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

3) The likelihood function is

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

The log-likelihood function is

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

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

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

and

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

Setting these to zero and rearranging, we obtain the mle of  $\sigma$  as

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

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

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

4) The likelihood function is

$$L(\lambda) = \left( \prod_{i=1}^n (1 - \lambda x_i) \right)^{1/\lambda-1}.$$

The log-likelihood function is

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

The derivative with respect to  $\lambda$  is

$$\frac{d \log L(\lambda)}{d\lambda} = -\lambda^{-2} \sum_{i=1}^n \log(1 - \lambda x_i) - (1/\lambda - 1) \sum_{i=1}^n \frac{x_i}{1 - \lambda x_i}.$$

So, the mle of  $\lambda$  is the root of the equation

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