

**Problem 1**

- (a) Let  $M_{m,j}$  denote the  $j$ th largest order statistic in a random sample of size  $m$  (so  $M_{m,1}$  is the maximum), and suppose that the rescaled quantities  $((M_{m,1}-b_m)/a_m, \dots, (M_{m,r}-b_m)/a_m) \xrightarrow{D} (Y_1, \dots, Y_r)$  as  $m \rightarrow \infty$  for fixed  $r$ , where  $Y_1$  has distribution function  $\exp\{-\Lambda(y)\}$  and  $Y_1 > \dots > Y_r$ . Explain why the cumulative distribution function of  $Y_2$ , conditional on  $Y_1 = y_1$ , must be  $\exp\{\Lambda(y_1) - \Lambda(y_2)\}$  for  $y_2 < y_1$ , and deduce that the joint density of  $Y_1$  and  $Y_2$  must be

$$\{-\dot{\Lambda}(y_1)\}\{-\dot{\Lambda}(y_2)\} \exp\{-\Lambda(y_2)\}, \quad y_2 < y_1.$$

Hence show (e.g., by induction) that the joint density of  $Y_1, \dots, Y_r$  is  $\exp\{-\Lambda(y_r)\} \prod_{j=1}^r \{-\dot{\Lambda}(y_j)\}$ , with  $y_r < \dots < y_1$ .

- (b) Deduce that

$$f_{Y_r}(y_r) = \frac{\{-\dot{\Lambda}(y_r)\} \Lambda(y_r)^{r-1}}{(r-1)!} \exp\{-\Lambda(y_r)\}, \quad y_r \in \mathbb{R}.$$

- (c) Hence show that the joint conditional density of  $Y_1, \dots, Y_r$  given that  $Y_{r+1} = u$  is

$$r! \prod_{j=1}^r \frac{-\dot{\Lambda}(y_j)}{\Lambda(u)}, \quad y_1 > \dots > y_r > u.$$

How is this related to the joint density of  $r$  independent variables, each having distribution function  $H(y) = 1 - \Lambda(y)/\Lambda(u)$ , for  $y > u$ ? What is  $H$ ?

**Problem 2** Verify the formulae for  $x_p$  given on slide 102 of the notes.

**Problem 3** Suppose that the GEV distribution with parameters  $\eta$ ,  $\tau$  and  $\xi$  has been fitted to data  $y_1, \dots, y_n$ , and a profile log likelihood is required for the  $1/p$ -year return level  $x_p$ .

- (a) Show that  $x_p = \eta + \tau a_p(\xi)$ , where for  $p \in (0, 1)$  we set  $a_p(\xi) = [ \{-\log(1-p)\}^{-\xi} - 1 ] / \xi$ .
- (b) The required profile log likelihood  $\ell_p(x_p)$  is defined as  $\max_{\tau, \xi} \ell^*(x_p, \tau, \xi)$ , where  $\ell^*$  is the log likelihood function parametrised in terms of  $x_p$ ,  $\tau$  and  $\xi$ . Show that

$$\ell_p(x_p) = \max_{\tau, \xi} \ell\{x_p - \tau a_p(\xi), \tau, \xi\},$$

where  $\ell$  is the log likelihood parametrized in terms of  $\eta$ ,  $\tau$  and  $\xi$ , and hence explain how you would compute  $\ell_p(x_p)$ .