

LECTURE

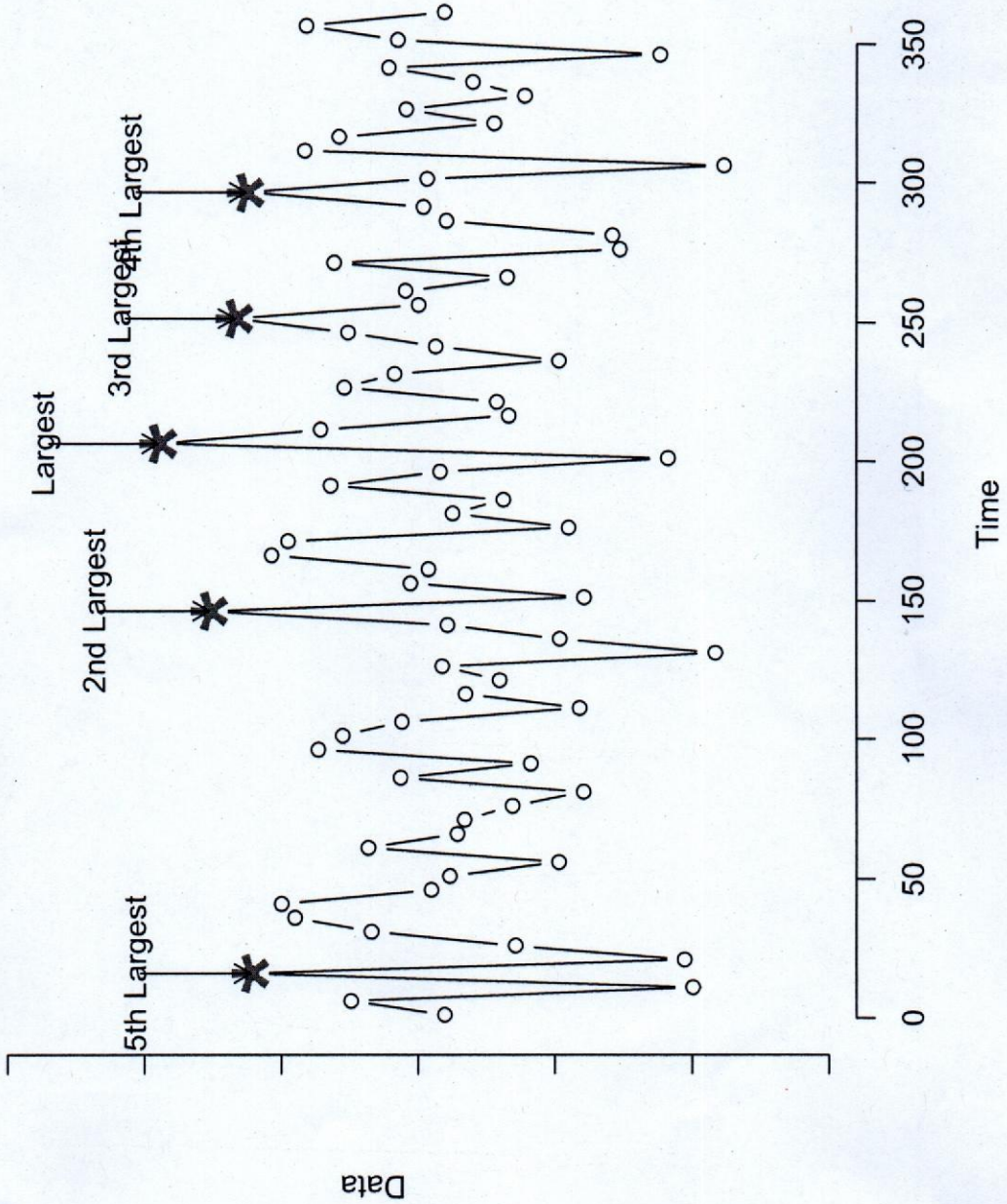
5 OCTOBER

10:00-11:00AM

MATH4/68181

Definition 3

"r largest" method



ML Estimation

$$x_{1,1}, x_{1,2}, \dots, x_{1,n}$$

$$x_{1,1}, x_{2,1}, \dots, x_{n,1}$$

(largest obsns for n years)

$$x_{1,2}, x_{2,2}, \dots, x_{n,2}$$

(2nd largest obsns for n years)

⋮

$$x_{1,r}, x_{2,r}, \dots, x_{n,r}$$

(r th largest obsn for n years)

Definition 3

Suppose X_1, X_2, \dots, X_n are IID observations with CDF $F(\cdot)$

Let $M_n^{(1)} =$ largest of X_1, \dots, X_n

$M_n^{(2)} =$ 2nd " " "

⋮

$M_n^{(r)} =$ rth " " "

It can be shown under certain conditions that

$$\Pr \left(\frac{M_n^{(1)} - b_n}{a_n} < x_1, \dots, \frac{M_n^{(r)} - b_n}{a_n} < x_r \right)$$

$$\xrightarrow{\text{as } n \rightarrow \infty} \sum_{s_1=0}^1 \sum_{s_2=0}^{2-s_1} \dots \sum_{s_{r-1}=0}^{r-1-s_1-\dots-s_{r-2}}$$

$$\frac{(\gamma_2 - \gamma_1)^{s_1}}{s_1!} \dots \frac{(\gamma_r - \gamma_{r-1})^{s_{r-1}}}{s_{r-1}!} e^{-\gamma_r} \quad (*)$$

$$\text{where } \gamma_i = \left(1 + \sum \frac{x_i - 0}{1} \right)^{-\frac{1}{\alpha}}$$

$$= \left(1 + \sum x_i \right)^{-\frac{1}{\alpha}}, \quad i = 1, 2, \dots, r$$

It can be deduced from (*) that the joint PDF of $(M_n^{(1)}, M_n^{(2)}, \dots, M_n^{(r)})$ is

$$f(x_1, x_2, \dots, x_r) = \sigma^{-r} e^{-\left(1 + \sum \frac{x_i - \mu}{\sigma}\right)^{-\frac{1}{\xi}}} \times e^{-\left(\frac{1}{\xi} + 1\right) \sum_{i=1}^r \log\left(1 + \frac{x_i - \mu}{\sigma}\right)}$$

where $x_1 \geq x_2 \geq \dots \geq x_r$ and

$$1 + \sum \frac{(x_i - \mu)}{\sigma} > 0, \quad i = 1, 2, \dots, r$$

$$L(\mu, \sigma, \xi) = \prod_{i=1}^n f(x_{i,1}, x_{i,2}, \dots, x_{i,r})$$

$$= \prod_{i=1}^n \left[\sigma^{-r} e^{-\left(1 + \frac{1}{\xi}\right) \frac{x_{i,r} - \mu}{\sigma}} \right. \\ \left. e^{-\left(1 + \frac{1}{\xi}\right) \sum_{j=1}^r \log\left(1 + \frac{x_{i,j} - \mu}{\sigma}\right)} \right]$$

$$= \sigma^{-nr} e^{-\sum_{i=1}^n \left(1 + \frac{1}{\xi}\right) \frac{x_{i,r} - \mu}{\sigma}}$$

$$e^{-\left(1 + \frac{1}{\xi}\right) \sum_{i=1}^n \sum_{j=1}^r \log\left(1 + \frac{x_{i,j} - \mu}{\sigma}\right)}$$

$$\Rightarrow \log L(\mu, \sigma, \xi) = -nr \log \sigma$$

$$- \sum_{i=1}^n \left(1 + \frac{1}{\xi}\right) \frac{x_{i,r} - \mu}{\sigma}$$

$$- \left(1 + \frac{1}{\xi}\right) \sum_{i=1}^n \sum_{j=1}^r \log\left(1 + \frac{x_{i,j} - \mu}{\sigma}\right).$$

MLE equations for the r largest distribution

The MLEs of μ , σ and ξ are the simultaneous solutions of

$$\begin{aligned} \frac{\partial \log L}{\partial \mu} &= -\frac{1}{\sigma} \sum_{i=1}^n \left(1 + \xi \frac{x_{i,r} - \mu}{\sigma}\right)^{-\frac{1}{\xi}-1} \\ &\quad - \frac{1 + \xi}{\sigma} \sum_{i=1}^n \sum_{j=1}^r \left(1 + \xi \frac{x_{i,j} - \mu}{\sigma}\right)^{-1} \\ &= 0, \end{aligned} \quad (1)$$

$$\begin{aligned} \frac{\partial \log L}{\partial \sigma} &= -\frac{nr}{\sigma} - \frac{1}{\sigma^2} \sum_{i=1}^n (x_{i,r} - \mu) \left(1 + \xi \frac{x_{i,r} - \mu}{\sigma}\right)^{-\frac{1}{\xi}-1} \\ &\quad + \frac{1 + \xi}{\sigma^2} \sum_{i=1}^n \sum_{j=1}^r (x_{i,j} - \mu) \left(1 + \xi \frac{x_{i,j} - \mu}{\sigma}\right)^{-1} \\ &= 0, \end{aligned} \quad (2)$$

and

$$\begin{aligned} \frac{\partial \log L}{\partial \xi} &= -\frac{1}{\xi^2} \sum_{i=1}^n \log \left(1 + \xi \frac{x_{i,r} - \mu}{\sigma}\right) \left(1 + \xi \frac{x_{i,r} - \mu}{\sigma}\right)^{-\frac{1}{\xi}} \\ &\quad + \frac{1}{\xi \sigma} \sum_{i=1}^n (x_{i,r} - \mu) \left(1 + \xi \frac{x_{i,r} - \mu}{\sigma}\right)^{-\frac{1}{\xi}-1} \\ &\quad + \frac{1}{\xi^2} \sum_{i=1}^n \sum_{j=1}^r \log \left(1 + \xi \frac{x_{i,j} - \mu}{\sigma}\right) \\ &\quad - \frac{1 + \xi}{\xi \sigma} \sum_{i=1}^n \sum_{j=1}^r (x_{i,j} - \mu) \left(1 + \xi \frac{x_{i,j} - \mu}{\sigma}\right)^{-1} \\ &= 0. \end{aligned} \quad (3)$$

The MLEs of μ , σ & ξ are the simultaneous solutions of (1), (2) and (3). The R package

Evd can compute the MLEs μ , σ & ξ .

$$e^y \approx 1 + y \quad \text{small } y$$

$$e^y = 1 + y + \frac{y^2}{2!} + \frac{y^3}{3!} + \dots$$

$$\underline{\underline{\gamma(t)}} > 0 \quad \forall t$$