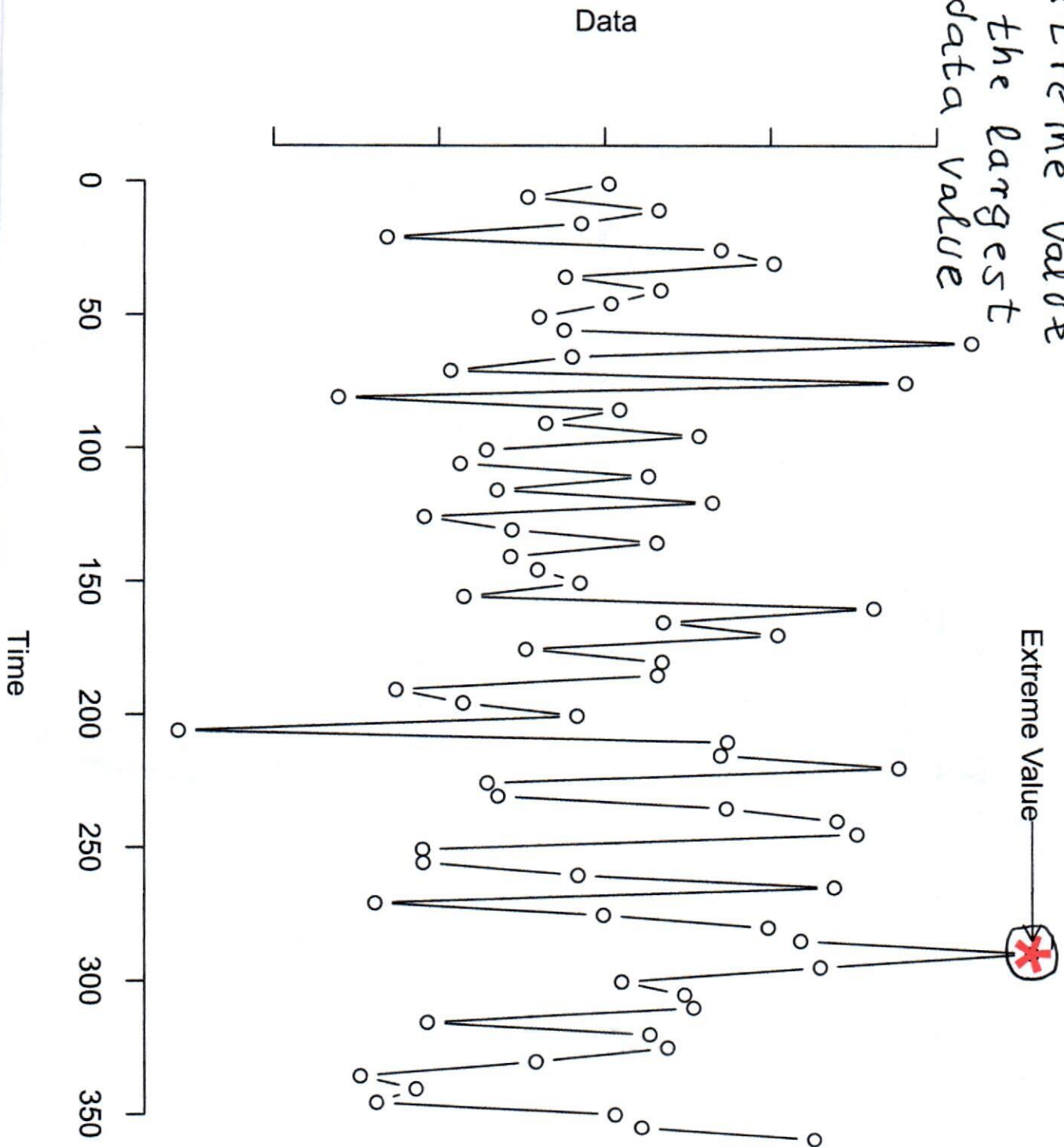


Definition 1

Extreme value
= the largest
data_value



Can the 3 limits be combined into 1.

The ETT says that there can be 3 possible limits for the sample maximum. A practitioner may not know how to check the conditions I-III giving rise to the 3 limits. So it will be convenient if the 3 limits can be combined into one mathematical form.

The combined form is known as the Generalised Extreme Value (GEV) distribution.

GEV distribution

A random variable X follows the GEV distribution if its CDF is

$$G(x) = \exp\left[-\left(1 + \xi \frac{x-\mu}{\sigma}\right)^{-\frac{1}{\xi}}\right]$$

where $1 + \xi \frac{x-\mu}{\sigma} > 0$

$-\infty < \xi < \infty$ "shape parameter"

$\sigma > 0$ "scale parameter"

$-\infty < \mu < \infty$ "location parameter"

Notation: $X \sim \text{GEV}(\mu, \sigma, \xi)$

Domain of X

$$\xi < 0$$

$$1 + \xi \frac{x - \mu}{\sigma} > 0$$

$$\Leftrightarrow \xi \frac{x - \mu}{\sigma} > -1$$

$$\Leftrightarrow \frac{x - \mu}{\sigma} < -\frac{1}{\xi}$$

$$\Leftrightarrow x < \mu - \frac{\sigma}{\xi}$$

domain is $(-\infty, \mu - \frac{\sigma}{\xi})$

$$\xi = 0$$

$$1 + \xi \frac{x - \mu}{\sigma} = 1 > 0$$

\Leftrightarrow domain is $(-\infty, \infty)$

$$\xi > 0$$

$$1 + \xi \frac{x - \mu}{\sigma} > 0$$

$$\Leftrightarrow \xi \frac{x - \mu}{\sigma} > -1$$

$$\Leftrightarrow \frac{x - \mu}{\sigma} > -\frac{1}{\xi}$$

$$\Leftrightarrow x > \mu - \frac{\sigma}{\xi}$$

\Leftrightarrow domain is $(\mu - \frac{\sigma}{\xi}, \infty)$

$$\text{Domain of } X = \begin{cases} (\mu - \frac{\sigma}{\xi}, \infty) & \text{if } \xi > 0 \\ (-\infty, \infty) & \text{if } \xi = 0 \\ (-\infty, \mu - \frac{\sigma}{\xi}) & \text{if } \xi < 0 \end{cases}$$

$$\xi > 0$$

$$G(x) = \exp \left\{ - \left(1 + \xi \frac{x - \mu}{\sigma} \right)^{-\frac{1}{\xi}} \right\}$$

$$= \exp \left\{ - \left(\underbrace{1 - \frac{\xi \mu}{\sigma}}_b + \underbrace{\frac{\xi \sigma}{\sigma}}_a x \right)^{-\frac{1}{\xi}} \right\}$$

$$\begin{array}{l} a > 0 \\ \alpha > 0 \\ b \in \mathbb{R} \end{array}$$

$$= \exp \left\{ - (ax + b)^{-\alpha} \right\}$$

$$= \text{same type as } \exp(-x^{-\alpha})$$

\Rightarrow Fréchet limit is a particular case of the GEV for $\xi > 0$.

$$\xi = 0$$

$$G(x) = \lim_{\xi \rightarrow 0} \exp \left\{ - \left(1 + \xi \frac{x - \mu}{\sigma} \right)^{-\frac{1}{\xi}} \right\}$$

$$= \lim_{\xi \rightarrow 0} \exp \left\{ - \left(1 + \frac{\frac{x - \mu}{\sigma}}{\frac{1}{\xi}} \right)^{-\frac{1}{\xi}} \right\}$$

$$\text{Set } m = \frac{1}{\xi}$$

$$= \lim_{m \rightarrow \infty} \exp \left\{ - \left(1 + \frac{x - \mu}{\sigma m} \right)^{-m} \right\}$$

$$= \lim_{m \rightarrow \infty} \exp \left\{ - \left[\left(1 + \frac{x - \mu}{\sigma m} \right)^m \right]^{-1} \right\}$$

$$\lim_{m \rightarrow \infty} \left(1 + \frac{z}{m} \right)^m = e^z$$

$$= \lim_{m \rightarrow \infty} \exp \left\{ - \left[\exp \left(\frac{x - \mu}{\sigma} \right) \right]^{-1} \right\}$$

$$= \exp \left\{ - \exp \left(- \frac{x - \mu}{\sigma} \right) \right\}$$

$$= \exp \left\{ - \exp(-ax + b) \right\}$$

$$a = \frac{1}{\sigma} > 0, \quad b = \frac{\mu}{\sigma} \in \mathbb{R}$$

$$= \text{same type as } \exp \{ - \exp(-x) \}$$

\Rightarrow Gumbel limit is the particular case of the GEV for $\xi = 0$.

- Gumbel limit is particular case for $\xi = 0$
- Weibull " " " " " $\xi < 0$
- Fréchet " " " " " $\xi > 0$

PDF of GEV distribution

$$g(x) = \frac{d}{dx} G(x)$$

$$= \frac{d}{dx} \exp \left[- \left(1 + \xi \frac{x-\mu}{\sigma} \right)^{-\frac{1}{\xi}} \right]$$

$$= \frac{1}{\sigma} \left(1 + \xi \frac{x-\mu}{\sigma} \right)^{-\frac{1}{\xi} - 1}$$

$$\bullet \exp \left[- \left(1 + \xi \frac{x-\mu}{\sigma} \right)^{-\frac{1}{\xi}} \right]$$

Quantile of GEV distribution

A quantile x_p of a RV X is defined by

$$P(X \leq x_p) = p \dots (†)$$

For the GEV distribution,

$$(†) \Leftrightarrow G(x_p) = p$$

$$\Leftrightarrow \exp \left[- \left(1 + \xi \frac{x_p - \mu}{\sigma} \right)^{-\frac{1}{\xi}} \right] = p$$

$$\Leftrightarrow - \left(1 + \xi \frac{x_p - \mu}{\sigma} \right)^{-\frac{1}{\xi}} = \log p$$

$$\Leftrightarrow 1 + \xi \frac{x_p - \mu}{\sigma} = (-\log p)^{-\xi}$$

$$\Leftrightarrow \xi \frac{x_p - \mu}{\sigma} = (-\log p)^{-\xi} - 1$$

$$\Leftrightarrow x_p = \mu + \frac{\sigma}{\xi} \left[(-\log p)^{-\xi} - 1 \right]$$

↑ p th quantile of the GEV distribution

In particular

$$\begin{aligned} x_{\frac{1}{2}} &= \text{Median of GEV} \\ &= \mu + \frac{\sigma}{\xi} \left[(\log 2)^{-\xi} - 1 \right] \end{aligned}$$

Return level of GEV distribution

A return level with period T (in years) is the level expected to be exceeded on average once in every T years. Let x_T denote the return level with period T . Then

$$P(X > x_T) = \frac{1}{T}$$

$$\Leftrightarrow P(X \leq x_T) = 1 - \frac{1}{T}$$

$$\Leftrightarrow G(x_T) = 1 - \frac{1}{T}$$

$$\Leftrightarrow \exp \left\{ - \left(1 + \xi \frac{x_T - \mu}{\sigma} \right)^{-\frac{1}{\xi}} \right\} = 1 - \frac{1}{T}$$

$$\Leftrightarrow x_T = \mu + \frac{\sigma}{\xi} \left[\left(-\log \left(1 - \frac{1}{T} \right) \right)^{-\xi} - 1 \right]$$

Estimation of GEV distribution

Suppose x_1, x_2, \dots, x_n are independent and identical data on $X \sim \text{GEV}(\mu, \sigma, \xi)$.

The most popular method for estimation is the method of maximum likelihood.

The likelihood function of (μ, σ, ξ) is

$$\begin{aligned} L(\mu, \sigma, \xi) &= \prod_{i=1}^n \left\{ \frac{1}{\sigma} \left(1 + \xi \frac{x_i - \mu}{\sigma} \right)^{-\frac{1}{\xi} - 1} \right. \\ &\quad \left. \cdot \exp \left[- \left(1 + \xi \frac{x_i - \mu}{\sigma} \right)^{-\frac{1}{\xi}} \right] \right\} \\ &= \frac{1}{\sigma^n} \left[\prod_{i=1}^n \left(1 + \xi \frac{x_i - \mu}{\sigma} \right) \right]^{-\frac{1}{\xi} - 1} \\ &\quad \cdot \exp \left[- \sum_{i=1}^n \left(1 + \xi \frac{x_i - \mu}{\sigma} \right)^{-\frac{1}{\xi}} \right] \end{aligned}$$

The log-likelihood function is

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

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

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

$$\frac{\partial \log L}{\partial \mu} = 0 ,$$

$$\frac{\partial \log L}{\partial \sigma} = 0 ,$$

$$\frac{\partial \log L}{\partial \xi} = 0 .$$

MLE equations for the GEV distribution

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

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

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

and

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

MLEs of μ , σ and ξ are the solutions of (1), (2) and (3).

The fgev(.) command in R can compute the MLEs of μ , σ and ξ .