

LECTURE

2 OCTOBER

9:00-10:00AM

MATH3/4/68181

For a practitioner, dealing with 3 distributions is not convenient.

Is there a distribution that unifies the 3 distributions into a single form?

Yes.

GEV (Generalized Extreme Value)

Distribution has the CDF

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

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

$$-\infty < \gamma < \infty \quad \text{"shape"}$$

$$-\infty < \mu < \infty \quad \text{"location"}$$

$$\sigma > 0 \quad \text{"scale"}$$

Particular Cases

i) $\gamma = 0$

$$\lim_{\gamma \rightarrow 0} e^{-\left(1 + \gamma \frac{x-\mu}{\sigma}\right)^{-\frac{1}{\gamma}}}$$

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

Set $m = \frac{1}{\gamma}$

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

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

$\left(1 + \frac{y}{m}\right)^m \rightarrow e^y$
as $m \rightarrow \infty$

$$= e^{-\left[e^{\frac{x-\mu}{\sigma}}\right]^{-1}}$$

$$= e^{-e^{-\frac{x-\mu}{\sigma}}}$$

the same type
as $\Lambda(x) = e^{-e^{-x}}$

$\Rightarrow \Lambda(x)$ is a particular case of
GEV when $\gamma = 0$

$$\text{ii) } \boxed{\gamma > 0}$$

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

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

$\boxed{\text{Set } \alpha = \frac{1}{\gamma}}$

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

$\boxed{a = \frac{1}{\sigma\alpha}, b = 1 - \frac{\mu}{\alpha\sigma}}$

$$\begin{aligned} a &> 0 \\ b &\in \mathbb{R} \end{aligned}$$

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

= the same type as $\Phi_\alpha(x)$

$\Rightarrow \Phi_\alpha(x)$ is a particular case of
GEV when $\gamma > 0$.

iii) $\boxed{\xi < 0}$

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

Set $\alpha = -\frac{1}{\xi}$

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

$a = \frac{1}{\alpha \sigma}, b = -1 - \frac{\mu}{\alpha \sigma}$

$$\begin{aligned} a &> 0 \\ b &\in \mathbb{R} \end{aligned}$$

$$= e^{-(-b + ax)^\alpha}$$

= same type as $\Psi_\alpha(x)$

$\Rightarrow \Psi_\alpha(x)$ is a particular case of GEV when $\xi < 0$.

G E V Distribution

CDF

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

PDF

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

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

$$\cdot e^{-\left(1 + \frac{x-\mu}{\sigma}\right)^{-\frac{1}{\gamma}}}$$

Quantile function

$$e^{-\left(1 + \frac{x-\mu}{\sigma}\right)^{-\frac{1}{\gamma}}} = p, \quad 0 < p < 1$$

$$\Rightarrow -\left(1 + \frac{x-\mu}{\sigma}\right)^{-\frac{1}{\gamma}} = \log p$$

$$\Rightarrow \left(1 + \frac{x-\mu}{\sigma}\right)^{-\frac{1}{\gamma}} = -\log p$$

$$\Rightarrow 1 + \frac{x-\mu}{\sigma} = (-\log p)^{-\frac{1}{\gamma}}$$

$$\Rightarrow x = \mu + \frac{\sigma}{\gamma} \left[(-\log p)^{-\frac{1}{\gamma}} - 1 \right]$$

pth quantile

For example, if $p = \frac{1}{2}$

$$\text{Median} = \mu + \frac{\sigma}{\sqrt{3}} \left[\left(\log 2 \right)^{-\frac{2}{3}} - 1 \right]$$

Return level

is the quantile when

$$p = 1 - \frac{1}{T} \quad \begin{matrix} \text{No. of} \\ \text{years} \end{matrix}$$

For example if $T = 100$ years

then the quantile will be the extreme value that occurs on average once in 100 years.

Return level for T years

$$= \mu + \frac{\sigma}{\sqrt{3}} \left[\left(-\log \left(1 - \frac{1}{T} \right) \right)^{-\frac{2}{3}} - 1 \right].$$

ML estimation

(Maximum Likelihood Estimation)

Suppose x_1, x_2, \dots, x_n is a random sample from the GEV distribution.

$$L(\mu, \sigma, \xi) = \prod_{i=1}^n g(x_i)$$

$$= \prod_{i=1}^n \left[\frac{1}{\sigma} \left(1 + \xi \frac{x_i - \mu}{\sigma} \right)^{-\frac{1}{\xi}} - 1 \right]$$

$$e^{-\left(1 + \xi \frac{x_i - \mu}{\sigma} \right)^{-\frac{1}{\xi}}} \right]$$

$$= \frac{1}{\sigma^n} \left[\prod_{i=1}^n \left(1 + \xi \frac{x_i - \mu}{\sigma} \right) \right]^{-\frac{1}{\xi}} - 1$$

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

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

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

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

MLE equations for the GEV distribution

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

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

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

and

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

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

fgev in the R software will compute the MLEs $\hat{\mu}$, $\hat{\sigma}$ and $\hat{\xi}$.

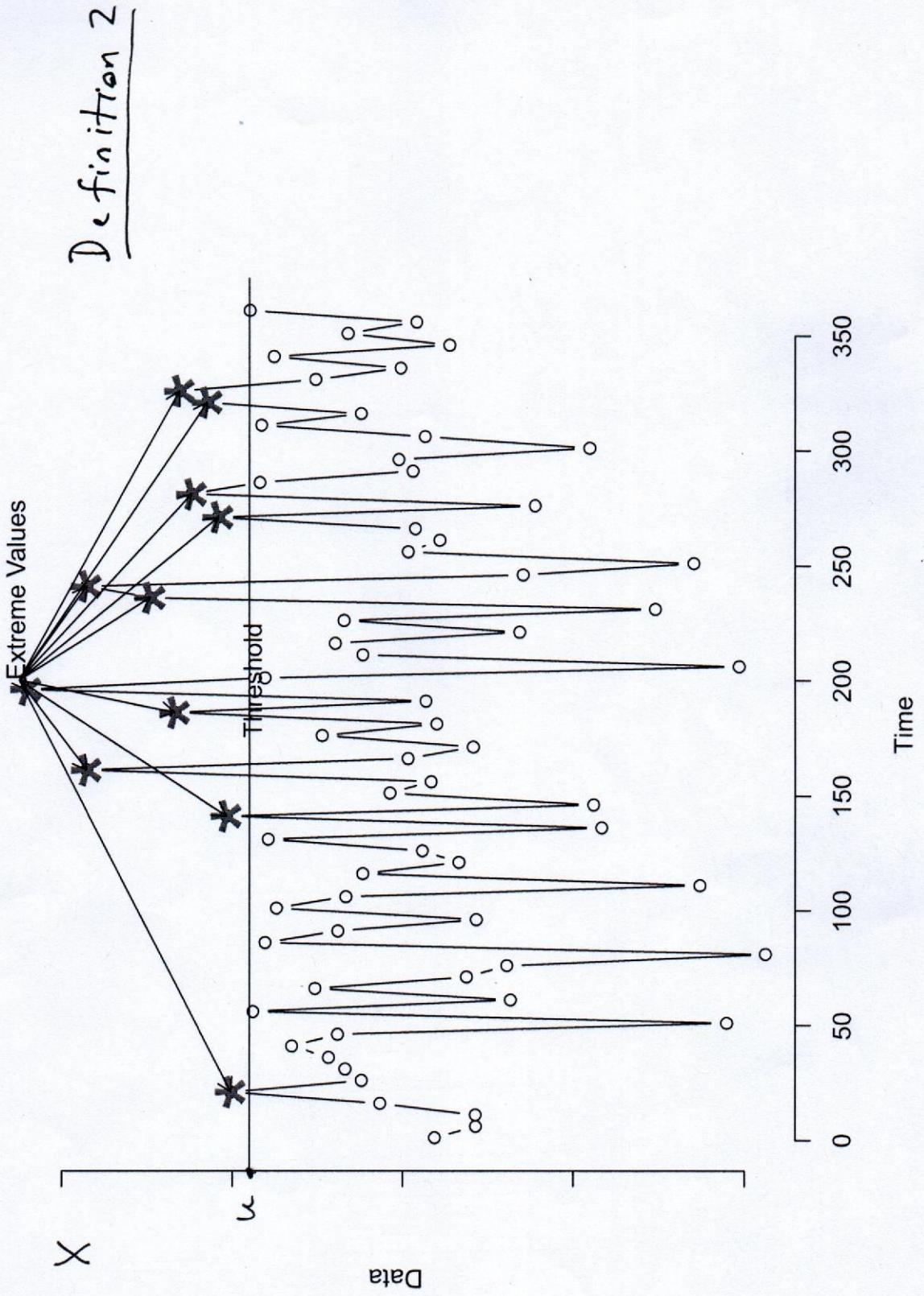
Definition 2

X = variable of interest
(eg stock returns)

u = threshold (considered high enough)

X is an extreme value
if $X > u$

$$X - u = \frac{\text{Excess}}{\text{amount}}$$



What is the distribution of
 $X - u$ (the excess) if $X > u$
(that is X is an extreme value)?

$$P(X - u < \infty | X > u) = ?$$