

Space-Time Extremes & Applications to Environmental Data

Ana Ferreira
Instituto Superior Técnico
Maths Department

*MMAC day
14 March, 2024*

Outline

- ▶ Extremes and Maximum Domain of Attraction (MDA)
- ▶ Extreme value index - $\gamma \in \mathbb{R}$ -
- ▶ Extensions of MDA to more dimensions
- ▶ Applications

Probability and Statistics

Extreme Values commonly relate to **maxima** and **minima**,

$$X_{n,n} = \max(X_1, X_2, \dots, X_n) \quad X_{1,n} = \min(X_1, X_2, \dots, X_n)$$

with X_1, \dots, X_n a sample of independent and identically distributed (common d.f. F) random variables.

Let's look at the d.f. of the **maxima**:

$$\max(X_1, X_2, \dots, X_n) \leq x \quad \text{iif} \quad X_1 \leq x, X_2 \leq x, \dots, X_n \leq x$$

hence,

$$P(X_{n,n} \leq x) = \prod_{i=1}^n P(X_i \leq x) = F^n(x), \quad x \in \mathbb{R}.$$

Maximum Domain of Attraction condition and GEV_γ d.f.s

If $\exists a_n > 0$ and $b_n \in \mathbb{R}$ s.t. convergence in distribution holds,

$$\lim_{n \rightarrow \infty} P\left(\frac{X_{n,n} - b_n}{a_n} \leq x\right) = \lim_{n \rightarrow \infty} F^n(a_n x + b_n) = G(x) \text{ exists,}$$

then

$$G_\gamma(x) = e^{-(1+\gamma x)^{-1/\gamma}}, \quad 1 + \gamma x > 0, \quad \gamma \in \mathbb{R}.$$

Central Limit Theorem and Gaussian distribution:

with $E[X_i] = \mu$, $\text{var}(X_i) = \sigma^2 < \infty$, convergence in distribution holds,

$$\frac{\sum_{i=1}^n X_i - n\mu}{\sqrt{n\sigma^2}} = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \sim^a N(0, 1)$$

$$\lim_{n \rightarrow \infty} P\left(\frac{\sum_{i=1}^n X_i - n\mu}{\sigma\sqrt{n}} \leq z\right) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx.$$

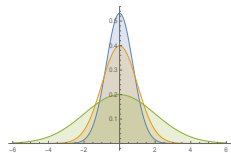


Figure: $N(\mu, \sigma^2)$

MDA and the extreme value index - $\gamma \in \mathbb{R}$ -

Examples:

1. - $\gamma = 1$ - Pareto $F_X(x) = 1 - 1/x, x > 1$:

$$\begin{aligned} F^n(a_n x + b_n) &= \left(1 - \frac{1}{a_n x + b_n}\right) \Big|_{a_n=n, b_n=n}^n \\ &= \left(1 - \frac{(x+1)^{-1}}{n}\right)^n \rightarrow_{n \rightarrow \infty} e^{-(x+1)^{-1}}, x+1 > 0. \end{aligned}$$

2. - $\gamma = 0$ - Exponential $F_X(x) = 1 - e^{-x}, x > 0$:

$$\begin{aligned} F^n(a_n x + b_n) &= \left(1 - e^{-a_n x - b_n}\right) \Big|_{a_n=1, b_n=\log n}^n \\ &= \left(1 - \frac{e^{-x}}{n}\right)^n \rightarrow_{n \rightarrow \infty} e^{-e^{-x}}, x > 0. \end{aligned}$$

Extreme value index - $\gamma \in \mathbb{R}$ -

- ▶ $\gamma > 0$: $1 - G_\gamma(x) \sim (\gamma x)^{-1/\gamma}$, $x \rightarrow \infty$
- ▶ $\gamma = 0$: $1 - G_0(x) \sim e^{-x}$, $x \rightarrow \infty$
- ▶ $\gamma < 0$: $1 - G_\gamma(-\gamma^{-1} - x) \sim (-\gamma x)^{-1/\gamma}$, $x \downarrow 0$.

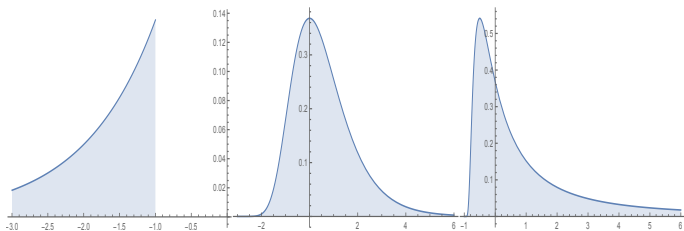


Figure: GEV densities: $\gamma = -1, 0, 1$.

Extreme value index estimation

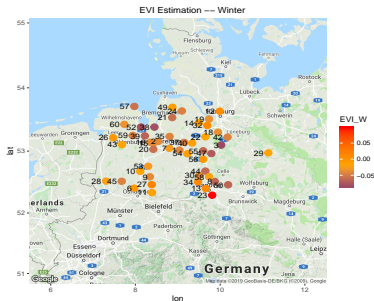


Figure: Estimation of the extreme value index for precipitation data from observational stations

- ▶ extreme value index, j -th moment of log-spacings

$$\sum_{i=0}^{k-1} (\log X_{n-i,n} - \log X_{n-k,n})^j$$

- ▶ exceedance probabilities
- ▶ high quantiles
- ▶ value-at-risk
- ▶ return value
- ▶ return period

Spatial-temporal data

Observed space-time process $X = \{X_{i,j}\}_{i,j}^{n,m}$

- ▶ (time - days $i = 1, \dots, n$)
- ▶ (space - stations $j = 1, \dots, m$)

$(X_{i,1}, X_{i,2}, \dots, X_{i,m})$ joint dependence $F_{i;1,\dots,m}$

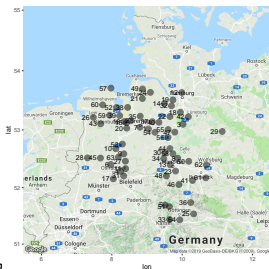


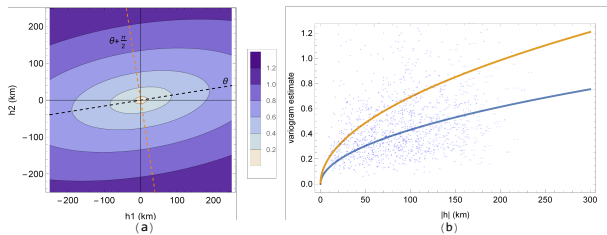
Figure: Weather stations in NW Germany

Bivariate Maximum Domain of Attraction condition

GEV $_{\gamma_1, \gamma_2}$ d.f.s: $\exists a_n, c_n > 0, b_n, d_n \in \mathbb{R}$ s.t. conv. in distrib. holds,

$$\begin{aligned} \lim_{n \rightarrow \infty} P \left(\frac{X_{n,n} - b_n}{a_n} \leq x, \frac{Y_{n,n} - d_n}{c_n} \leq y \right) \\ = \lim_{n \rightarrow \infty} F^n(a_n x + b_n, c_n y + d_n) = G_{\gamma_1, \gamma_2}(x, y), \quad \gamma_1, \gamma_2 \in \mathbb{R}. \end{aligned}$$

Winter variogram and non-stationary stations



Figures (a)&(b).
Winter variogram

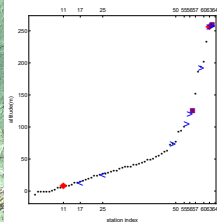
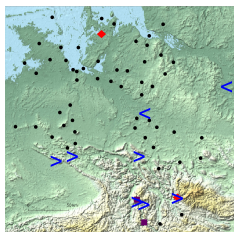


Figure 2.

Non-stationary
stations over time
and over space

THANK YOU !!