

1/14/2025 p.6

Alternative Models for Extremes

- (a) Exceedances over thresholds
- (b) largest order statistics approach
- (c) point process approach

Consider (a) first. and 1/14/25

Begin 1/16/25 here Motivation for an exceedances-based approach:
given that the variable of interest is above some threshold, how high can it go?

Practical interpretation: "Let the tail speak for itself"

Consider distribution conditional on exceeding a threshold u ,

$$F_u(y) = P\{Y \leq u+y \mid Y > u\} = \frac{F(u+y) - F(u)}{1 - F(u)}, y \geq 0.$$

Scaling result in this case:

Define $w_F = \sup\{\gamma : F(\gamma) < 1\}$, consider

limits of form $\lim_{u \uparrow w_F} F_u\left(\frac{y}{\sigma_u}\right) = G(y) \quad (**)$

for some $\sigma_u > 0$, nondegenerate G .

1/14/2025 p. 7

Balkema-de Haan-Richards theorem (1974, 1975)

If limit (**) exists, then $G(y)$ must be of form

$$G(y; \sigma, \xi) = \begin{cases} 1 - \left(1 + \xi \frac{y}{\sigma}\right)_+^{-1/\xi} & \xi \neq 0 \\ 1 - e^{-y/\sigma} & \xi = 0 \end{cases}$$

and the limit (**) exists if and only if $\lim_{y \rightarrow \infty} F(y)$ exists for the same F .

Informal motivation: Poisson-GPD model

Consider a fixed time unit (say, 1 year)

N : # of exceedances in a year, assume Poisson(λ)

Conditional on N , exceedances Y_1, \dots, Y_N are GPD(σ, ξ)

$$\begin{aligned} P\left\{\max_{1 \leq i \leq N} Y_i \leq x\right\} &= P(N=0) + \sum_{n=1}^{\infty} P(N=n, Y_1 \leq x, \dots, Y_n \leq x) \\ &= e^{-\lambda} + \sum_{n=1}^{\infty} \frac{\lambda^n e^{-\lambda}}{n!} \left\{1 - \left(1 + \xi \frac{x - \mu}{\sigma}\right)_+^{-1/\xi}\right\} \\ &= \exp\left\{-\lambda \left(1 + \xi \frac{x - \mu}{\sigma}\right)_+^{-1/\xi}\right\} \end{aligned}$$

which is of GEV form.

STOR 834 p. 8

(b) r-largest order statistics approach

Formulation: X_1, X_2, X_3, \dots IID (F)

Sample size n , order $X_{n,1} \geq X_{n,2} \geq \dots \geq X_{n,r}$

Assume $\frac{X_{n,i} - b_n}{a_n} \xrightarrow{d} Y_i$ where $Y_i \sim H$ ($H = GEV$)

Also assume or

Also assume convergence of densities (a little bit stronger) [do we need this?]

Then for fixed r as $n \rightarrow \infty$,

$$\left(\frac{X_{n,1} - b_n}{a_n}, \frac{X_{n,2} - b_n}{a_n}, \dots, \frac{X_{n,r} - b_n}{a_n} \right) \xrightarrow{d} (Y_1, \dots, Y_r)$$

where the joint density of (Y_1, \dots, Y_r) at (y_1, \dots, y_r) is

$$h(y_1, \dots, y_r) = \psi^{-r} \prod_{i=1}^r \left(1 + \zeta \frac{y_i - \mu}{4} \right)^{-1/3-1} \cdot \exp \left\{ - \left(1 + \zeta \frac{y_r - \mu}{4} \right)^{-1/3} \right\}$$

valid so long as $\left(1 + \zeta \frac{y_i - \mu}{4} \right) > 0$ for each $i = 1, \dots, r$

Origins: papers by Dwass, Lamperti, Weissman, etc.
on probability derivations

Later: statistical applications (hydrology extremes,
track records etc.). Caution about independence
of raw observations.

Point Process Approach

Some basic notation and terminology

Sample space \mathcal{S} is a complete separable metric space

For us: \mathcal{S} is a measurable subset of \mathbb{R}^d , some $d \geq 1$

A point process is a random subset of points in \mathcal{S}
on \mathcal{S}

Defined by $N(A)$, all $A \in \mathcal{A}$ where \mathcal{A} is some family of subsets of \mathcal{S} , e.g. all Borel sets

"Simple" means no multiple points

i.e. $N(\{x\}) = 1$ or 0 for any singleton

set $A = \{x\}$.

N is a non-homogeneous Poisson process with intensity Λ if it is simple and

(i) $N(A_1), \dots, N(A_k)$ are independent RVs for any finite sequence of disjoint sets A_1, \dots, A_k
($A_j \cap A_k = \emptyset$ whenever $j \neq k$)

(ii) $P\{N(A) = n\} = \frac{\Lambda(A)^n e^{-\Lambda(A)}}{n!} \quad n=0, 1, 2, \dots$

Here Λ is any measure on \mathcal{A} .

STOR 834 p.10

Now assume Λ has a density, i.e. there exists a function

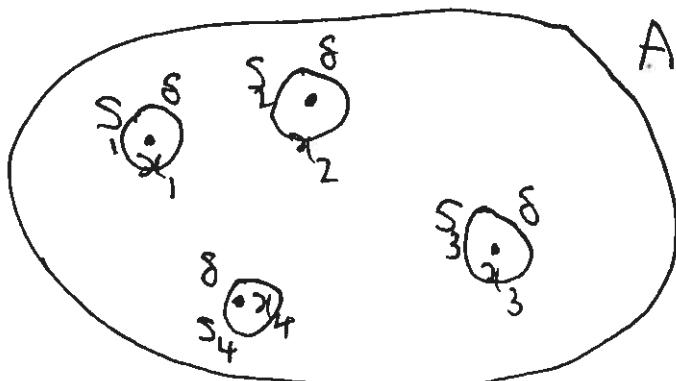
$$\lambda(x), x \in S, \text{ such that } \Lambda(A) = \int_A \lambda(x) dx, \text{ all } A \in \mathcal{A}$$

Notation in book: vector x because in general $S \subseteq \mathbb{R}^d$ some $d \geq 1$, x is a vector of length d (but theory does apply to $d=1$)

Question: Fix a set $A \in \mathcal{A}$, points at $x_1, x_2, \dots, x_{N(A)}$, what is joint density of $X_1 = x_1, \dots, X_{N(A)} = x_{N(A)}$?

Assume $N(A) = n$, assume a small sphere of volume δ around each of x_1, \dots, x_n

What is prob. There is exactly one point in each of S_1, \dots, S_n and no others in A ?



Answer is

$$\Lambda(S_1)e^{-\Lambda(S_1)} \cdot \Lambda(S_2)e^{-\Lambda(S_2)} \cdots \Lambda(S_n)e^{-\Lambda(S_n)} \cdot e^{-\Lambda(A - S_1 - S_2 - \cdots - S_n)}$$

$$= \prod_{i=1}^n \Lambda(S_i) e^{-\Lambda(A)} \approx \delta^n \lambda(x_1) \cdots \lambda(x_n) e^{-\Lambda(A)}$$

Divide by δ , let $\delta \rightarrow 0$

illustration with $n=4$

End 1/16/25