

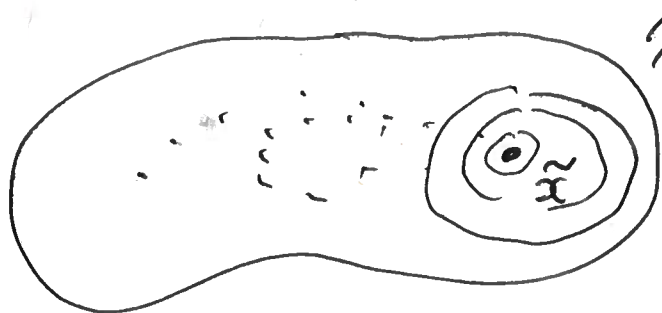
1) Almost sure growth of maxima

m.p.t $f: X \rightarrow X$, prob space (X, μ, \mathcal{B}) .

Process $X_n(x) = \phi(f^{n-1}x)$, $x \in X$,

observable $\phi: X \rightarrow \mathbb{R}$.

Maxima process ~~M_n~~ $M_n(x) = \max_{k \leq n} X_k(x)$.



Take $\Psi: [0, \infty) \rightarrow \mathbb{R}$.
 $\phi(x) = \Psi(\text{dist}(x, \tilde{x}))$

e.g. $\Psi(z) = -\log z, z^{-\alpha}, \sup \phi = \infty$.

$$\partial B(\tilde{x}, r) = \{x : \phi(x) = \Psi(r)\}.$$

- If ergodic, $M_n \rightarrow \infty$, μ -a.e. $x \in X$.
- Determine u_n, v_n such that for μ -a.e. $x \in X$, $v_n \leq M_n(x) \leq u_n, \forall n \geq N_x$.
- Recall the strong-BC property.

Let $S_n(x) = \sum_{k=1}^n \mathbb{1}_{B_k}(f^{k-1}x)$, targets $(B_k)_{k \geq 1}$. Let $E_n = \sum_{k=1}^n \mu(B_k) \rightarrow \infty$.
 Then for μ -a.e. $x \in X$, $\frac{S_n(x)}{E_n} \rightarrow 1$ ($n \rightarrow \infty$).

Easy case: Suppose $u_n \nearrow \infty$ monotone.

Then $\mu(M_n \geq u_n, i.o.) = \mu(X_n \geq u_n, i.o.)$.

"i.o." = infinitely often.

If $u_n = \psi(\Gamma_n)$, $\sum_{n=1}^{\infty} \mu(B(\tilde{x}, \Gamma_n)) < \infty$, then

$$BCI \Rightarrow \mu(X_n \geq u_n, i.o.) = 0.$$

$$\Rightarrow \mu(M_n \geq u_n, i.o.) = 0$$

$$\Rightarrow \mu(M_n < u_n, \text{eventually}) = 1.$$

e.g. $\Gamma_n = \frac{1}{n(\log n)^{1+\varepsilon}}$.

Lower bound.

$$\text{let } L = \sup \left\{ k \leq n : f^k(x) \in B(\tilde{x}, \Gamma_k) \right\}.$$

$$\sum_{k=1}^{\infty} \mu(B(\tilde{x}, \Gamma_k)) = \infty.$$

$$L \equiv L(n) \rightarrow \infty, n \rightarrow \infty.$$

Using SBC: $S_n(x) = S_L(x).$

$$|S_n(x) - E_n| = o(E_n).$$

$$|S_L(x) - E_L| = o(E_L) \quad , E_L \leq E_n.$$

$$E_n = \sum_{k=1}^n \nu(B(\tilde{x}, r_k)).$$

$$\Rightarrow E_L = (1 + o(1))E_n.$$

We get $M_n(x) \geq M_L(x) = \Psi(\Gamma_{L(n)}).$

E.g. $\nu(B(\tilde{x}, r_n)) = \frac{1}{n}.$

Then $E_n \sim \log n$, $E_L \sim \log L.$

$$\Rightarrow \log L = (1 + o(1)) \log n.$$

$$\Rightarrow \forall \varepsilon > 0, L \geq n^{1-\varepsilon}.$$

$$\Rightarrow M_n(x) \geq \Psi(\Gamma_{n^{1-\varepsilon}}),$$

$$\nu(B(\tilde{x}, r_{n^{1-\varepsilon}})) = \frac{1}{n^{1-\varepsilon}}.$$

• Balls of measure $\frac{1}{n(\log n)}.$

However $\log \log L = (1 + o(1)) \log \log n.$

Using this SBC approach:

$M_n(x) \geq V_n$ eventually, μ -a.e. $x \in \mathbb{R}$,
where $V_n = \mathcal{N}(\Gamma_n)$, $\mu(B(\bar{x}, r_n)) \geq \frac{(\log n)^B}{n}$.
 $B > 2$.

- Kim 2007
- Gupta-Nicol-Ott 2010.
- H, Nicol, Török 2016.
- Kirsebom, Kunde, Persson 2020.

I.I.D. class 1984, 1985, Galambos.

$(X_n)_{n \geq 1}$, distribution function

$$F_{X_1}(x) = \mu(X_1 \leq x), \quad \bar{F}_{X_1}(x) = \mu(X_1 \geq x).$$

- If $\bar{F}_X(u_n) \geq \frac{c \log \log n}{n}$ for $c > 1$,
then $\mu(M_n \geq u_n, \text{ eventually}) = 1$.
- If $\bar{F}_X(u_n) \leq \frac{c \log \log n}{n}$ for ~~some~~ $c < 1$
 $\mu(M_n \leq u_n, i.o.) = \mu(M_n > u_n, i.o.) = 1$.
 $\sum_{n=1}^{\infty} \bar{F}_X(u_n) = \infty$

- Continued fractions: Philipp 1976.

$$x = [a_0(x) \dots a_n(x) \dots]$$

$$\text{Then } \liminf_{n \rightarrow \infty} \frac{\log \log n}{n} \max_{k \leq n} a_k(x) = \frac{1}{\log 2}.$$

$$\mu\text{-a.e. } x \in [0, 1].$$

- Caution: $(X_n)_{n \geq 1}$ i.i.d.

$$Y_n = \max(X_n, X_{nt_1}).$$

Corresponding c as stated in the class conditions for $F_Y(u_n)$ is 2.

$$\max(Y_1, \dots, Y_n) = \max(X_1, \dots, X_{nt_1}).$$

$$F_Y(u) = F_X(u)^2$$

$$\bar{F}_Y(u) = 2\bar{F}_X(u) + \text{Error}.$$

Distributional convergence of maxima.

Set up (X, μ, \mathcal{B}) , $f: X \rightarrow X$.

$$M_n(x) = \max_{k \leq n} \phi(f^{k-1}x), \quad x \in X. \quad \phi: X \rightarrow \mathbb{R} \text{ observable.}$$

$\mu(M_n \leq u_n)$. Determine sequences a_n, b_n and a limit function G such that

$$\lim_{n \rightarrow \infty} \mu(a_n(M_n - b_n) \leq u) = G(u).$$

Theorem: Consider (X_n) i.i.d., Given $\tau > 0$ let u_n be such that $n\mu(X_1 \geq u_n) \rightarrow \tau$. ($n \rightarrow \infty$). (i.e. $u_n \equiv u_n(\tau)$). Then:

$$\mu(M_n \leq u_n) \rightarrow e^{-\tau} \quad (n \rightarrow \infty),$$

and conversely.

proof:

$$\begin{aligned} \mu(M_n \leq u_n) &= \mu(X_1 \leq u_n, \dots, X_n \leq u_n) \\ &= \mu(X_1 \leq u_n)^n \\ &= (1 - \mu(X_1 \geq u_n))^n \\ &= \left(1 - \frac{\tau}{n} (1 + o(1))\right)^n \rightarrow e^{-\tau}. \end{aligned}$$

E.g. $Y_n = \text{Max}(X_n, X_{n+1})$, $(X_n)_{n \geq 1}$ i.i.d.

Choose u_n : $n \bar{F}_Y(u_n) \rightarrow \tau$.

$$\text{Then } F_Y(u_n) = F_X(u_n)^2$$

$$\bar{F}_Y(u_n) = \mu(Y > u_n)$$

$$= 1 - (1 - \bar{F}_X(u_n))^2$$

$$\sim \cancel{2} 2 \bar{F}_X(u_n)$$

$$\Rightarrow \mu(M_n^Y \leq u_n) \xrightarrow{n \rightarrow \infty} e^{-\tau/2}. \quad \theta = \frac{1}{2}$$

Def: Suppose $n \mu(X_1 \geq u_n) \rightarrow \tau$. We say

$(X_n)_{n \geq 1}$ has extremal index θ if

$$\mu(M_n \leq u_n) \rightarrow e^{-\theta \tau} \quad \forall \tau > 0.$$

Remark: $\theta \in [0, 1]$.

• Dependent processes. (Blocking method).

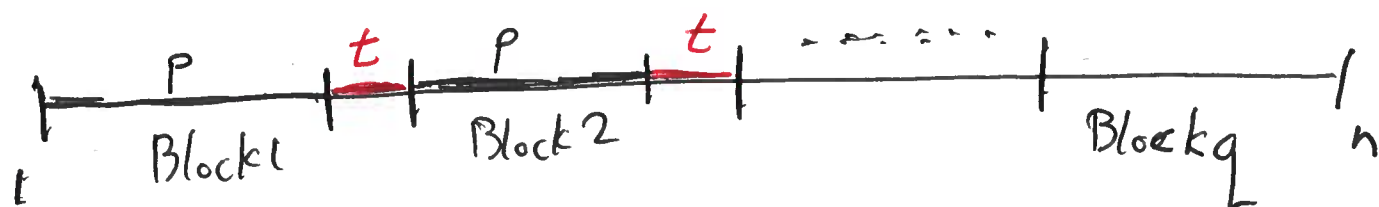
Leadbetter, et.al 1983.

Collet 2001.

Lucarini, et.al 2016.

- $X_n = \phi \circ f^{n-1}, \quad f: X \rightarrow X.$

$n, (p, q, t \rightarrow \infty, n \rightarrow \infty). \quad p, q \sim \sqrt{n}$
 $n = q(p+t). \quad t \sim (\log n)^2.$



In block 1: $\mu(M_p \leq u_n) = 1 - p\mu(X, \geq u_n) + \text{Error}.$

$$\left| \mu(M_n \leq u_n) - \left(1 - p\mu(X, \geq u_n)\right)^q \right| \leq \underline{\underline{\mathcal{E}(p, q, t)}}$$

- $$(1 - p\mu(X, \geq u_n))^q = \exp\{q \log(1 - p\mu(X, \geq u_n))\}$$

$$= \exp\{-pq\mu(X, \geq u_n)\} (1 + o(1)).$$

$$\sim \exp\{-n\mu(X, \geq u_n)\} \rightarrow e^{-\tau}.$$

- Correlation decay. Error depends on $C(p, q) \alpha^t, \quad \alpha \in (0, 1).$

• Error contribution:

$$n \sum_{j=2}^P \mu(X_1 \geq u_n, X_j \geq u_n). \quad (*)$$

Problem if: $\mu(X_1 \geq u_n, X_j \geq u_n) \approx \frac{1}{n}$.

want $(*) \rightarrow 0$ as $n \rightarrow \infty$. $D'(u_n)$.

What about $a_n, b_n, G(u)$?

$f(x) = 2x \bmod 1$, $\mu = \text{Leb}$.

$\phi(x) = -\log \text{dist}(x, \tilde{x})$. $u_n = \frac{u}{a_n} + b_n$

$$n \mu(X_1 \geq \frac{u}{a_n} + b_n) = n \text{Leb} \left\{ d(x, \tilde{x}) \leq e^{-\frac{u}{a_n} - b_n} \right\}$$

$$\text{Choose } a_n = 1, b_n = \log n + \log 2$$

$$\downarrow$$

$$= e^{-u}$$

Then (Collet 2001), $\mu(M_n \leq u + \log 2 + \log n)$
 for μ -a.e. $\tilde{x} \in \mathcal{X}$ $\rightarrow \exp\{-e^{-u}\}$.

Gumbel distribution. $u \in (-\infty, \infty)$.