

# Maxima under Dependence

Johanna G. Nešlehová, McGill University  
(joint work with K. Herrmann and M. Hofert)

Wirtschaftsuniversität Wien

May 3, 2023

# Motivation

- Maxima of random variables or vectors are of prime interest in risk management.
- When the variables are *i.i.d.*, asymptotic behavior of maxima is *well understood* (classical EVT). There is also an extensive literature on the case when the variables form a *time series*.
- In this talk, the aim is to investigate what happens when the variables are *identically distributed but dependent*.
- You can think of a large, homogeneous portfolio of claims, in which the *claims are dependent*, e.g., through some common factor(s).

# The central question cast in mathematical terms

Suppose  $X_1, X_2, \dots$  is a sequence of **identically distributed** univariate random variables (e.g., claims) that are generally **not independent**.

Define

$$M_n = \max(X_1, \dots, X_n).$$

## Questions to be addressed today:

Under which conditions do there exist sequences of reals  $(a_n)$ ,  $a_n > 0$ , and  $(b_n)$  so that

$$\frac{M_n - b_n}{a_n} \rightsquigarrow H$$

for some **non-degenerate** df  $H$ , and what does  $H$  look like?

## The i.i.d. case

When  $X_1, X_2, \dots$  are i.i.d., these questions have been long answered by the **Fisher–Tippett–Gnedenko Theorem**:

If there exist sequences of constants  $(a_n)$ ,  $a_n > 0$ , and  $(b_n)$  so that

$$\frac{M_n - b_n}{a_n} \rightsquigarrow H$$

for some **non-degenerate df**  $H$ , then  $H$  must be a generalized **extreme-value distribution**, given by

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

for all  $x$  such that  $1 + \xi(x - \mu)/\sigma > 0$ .

## The time series case: Leadbetter et al. (1983)

Consider a stationary sequence  $X_1, X_2, \dots$  with a limited long-range dependence, i.e., so that the so-called  $\mathcal{D}(u_n)$  condition holds for a series of suitable thresholds.

If there exist sequences of constants  $(a_n)$ ,  $a_n > 0$ , and  $(b_n)$  so that

$$\frac{M_n - b_n}{a_n} \rightsquigarrow H$$

for some non-degenerate df  $H$ ,  $H$  must be generalized extreme-value.

Let  $X_1^*, X_2^*, \dots$  be an i.i.d. sequence with the same marginal distribution and set  $M_n^* = \max(X_1^*, \dots, X_n^*)$ . Then under regularity conditions

$$\frac{M_n^* - b_n}{a_n} \rightsquigarrow H^* \quad \text{and} \quad \frac{M_n - b_n}{a_n} \rightsquigarrow H$$

where  $H = (H^*)^\theta$  for some extremal index  $\theta \in (0, 1]$ .

## Example

Suppose that  $X_1, X_2, \dots$  is an **i.i.d.** sequence of **standard normal** variables.

Set the norming constants to be

$$b_n = \Phi^{-1}\left(1 - \frac{1}{n}\right), \quad a_n = \frac{\bar{\Phi}(b_n)}{\varphi(b_n)}$$

where  $\Phi$  and  $\varphi$  denote the standard normal cdf and density.

Then for all  $x \in \mathbb{R}$ ,

$$\Pr\left(\frac{M_n - b_n}{a_n} \leq x\right) \rightarrow \Lambda(x) = \exp\{-\exp(-x)\}.$$

Here,  $\Lambda$  denotes the Gumbel extreme-value distribution.

## Example cont'd

Let  $X_1, X_2, \dots$  be an **stationary** sequence of **standard normal** variables.

Suppose that **Berman's** condition holds: With  $\gamma(n) = \text{cov}(X_1, X_n)$ ,

$$\lim_{n \rightarrow \infty} \gamma(n) \ln(n) = 0.$$

Then as in the i.i.d. case, for all  $x \in \mathbb{R}$ ,

$$\Pr\left(\frac{M_n - b_n}{a_n} \leq x\right) \rightarrow \Lambda(x) = \exp\{-\exp(-x)\}.$$

## Example cont'd: Berman (1962a)

Now consider the sequence  $X_1, X_2, \dots$  where for each  $n \in \mathbb{N}$ ,

$$X_n = \sqrt{\varrho}Z_0 + \sqrt{1 - \varrho}Z_n;$$

here,  $\varrho \in (0, 1)$  and  $Z_0, Z_1, \dots$  are i.i.d. standard normal variables.

Obviously,  $X_1, X_2, \dots$  is a stationary sequence of standard normal variables. However,  $\gamma(n) = \varrho$  for all  $n \geq 2$ , and hence

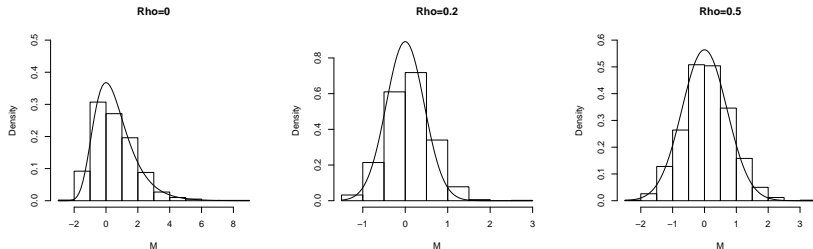
$$\lim_{n \rightarrow \infty} \gamma(n) \ln(n) = \infty.$$

Interestingly, as  $n \rightarrow \infty$ ,

$$M_n - \sqrt{1 - \varrho}b_n \rightsquigarrow \mathcal{N}(0, \varrho).$$



# Illustration



Histograms of the normalized maxima of  $n = 10^5$  variables in the i.i.d. case (left), and dependent case with  $\rho = 0.2$  (middle) and  $\rho = 0.5$  (right). Overlaid are the asymptotic densities of the Gumbel (left), and Normal with variance  $\rho$  (middle and right).

## Extension to normal variance mixtures

Consider the stationary sequence  $X_1, X_2, \dots$  with

$$X_n = \sqrt{\rho}\sigma Z_0 + \sqrt{1 - \rho}\sigma Z_n$$

and define, for  $n \in \mathbb{N}$ ,

$$Y_n = \mu + \sqrt{W}X_n,$$

where  $\mu \in \mathbb{R}$  and  $W$  is a positive random variable independent of  $(X_i)$ .

$Y_1, Y_2, \dots$  is then a stationary sequence whose finite-dimensional distributions are **elliptical**;  $W \sim \text{Ig}(\nu/2, \nu/2)$  leads to the  $t$  distribution.

A direct calculation shows that if  $M_n = \max(Y_1, \dots, Y_n)$ , as  $n \rightarrow \infty$ ,

$$\frac{M_n - \mu}{b_n} \rightsquigarrow \sigma \sqrt{1 - \rho} \sqrt{W}.$$

## Can we do anything at all?

Consider a sequence  $X_1, X_2, \dots$  of **identically distributed** random variables with common distribution  $F$  which is assumed to be **continuous**.

For each  $n \in \mathbb{N}$  and  $x_1, \dots, x_n \in \mathbb{R}$ , let also

$$F_n(x_1, \dots, x_n) = \Pr(X_1 \leq x_1, \dots, X_n \leq x_n).$$

From **Sklar's Theorem**, there exists a unique **copula**  $C_n$  so that

$$F_n(x_1, \dots, x_n) = C_n\{F(x_1), \dots, F(x_n)\}.$$

Using the **copula diagonal** (Jaworski, 2009)  $\delta_n(u) = C_n(u, \dots, u)$ , one has

$$\Pr\left(\frac{M_n - b_n}{a_n} \leq x\right) = \delta_n\{F(a_n x + b_n)\}.$$

## Basic insight

Consider some **suitable rate**  $r(n)$  (typically  $r(n) \rightarrow \infty$ ) and write

$$\begin{aligned} \Pr\left(\frac{M_n - b_n}{a_n} \leq x\right) &= \delta_n\{F(a_n x + b_n)\} \\ &= \delta_n[\{F^{r(n)}(a_n x + b_n)\}^{1/r(n)}] \end{aligned}$$

Whether the maximum  $M_n$  converges weakly will depend on:

- $F$ , or, in other words, the behavior of the maximum  $M_n^*$  of the **i.i.d. sequence**  $X_1^*, X_2^*, \dots$  with the same common distribution  $F$ .
- The behavior of the **copula diagonal**, notably the limit

$$\lim_{n \rightarrow \infty} \delta_n(u^{1/r(n)}), \quad u \in (0, 1).$$

## First result

Let  $X_1, X_2, \dots$  be a sequence of identically distributed rvs with continuous marginal  $F$ , and suppose that the following conditions hold:

- (a) There exist sequences  $(a_n)$ ,  $a_n > 0$  and  $(b_n)$  such that for all  $x \in \mathbb{R}$ ,

$$F^n(a_n x + b_n) \rightarrow H_{\xi, \mu, \sigma}(x).$$

- (b) There exists a rate function  $r : \mathbb{N} \rightarrow (0, \infty)$  with  $r(n) \rightarrow \infty$  as  $n \rightarrow \infty$  and a **continuous function  $D$**  such that for all  $u \in [0, 1]$ ,

$$\delta_n \{u^{1/r(n)}\} \rightarrow D(u).$$

Then  $D$  is in fact a distribution function and for all  $x \in \mathbb{R}$ ,

$$\Pr\left(\frac{M_n - b_{\lceil r(n) \rceil}}{a_{\lceil r(n) \rceil}} \leq x\right) \rightarrow D\{H_{\xi, \mu, \sigma}(x)\}.$$

## Moving maxima example

Consider the **moving maximum process**: Take a sequence  $(Z_i)_{i=-k+1}^{\infty}$  of i.i.d. unit Fréchet variables ( $Z_i \sim \Phi_1$ ) and set

$$X_i = \frac{1}{k+1} \max_{0 \leq j \leq k} Z_{i-j}$$

This process has limited long-range dependence and its **extremal index** is  $\theta = 1/(k+1)$  (Beirlant et al., 2004). One can also show that

$$C_n(u_1, \dots, u_n) = \prod_{j=0}^{k-1} \min_{1 \leq \ell \leq k-j} u_{\ell}^{1/(k+1)} \min_{n-j \leq \ell \leq n} u_{\ell}^{1/(k+1)} \times \prod_{j=1}^{n-k} \min_{j \leq \ell \leq j+k-1} u_{\ell}^{1/(k+1)}$$

so that  $\delta_n(u) = u^{\eta_n}$ ,  $\eta_n = (k^2 + n)/(k+1)$ . Obviously,

$$\lim_{n \rightarrow \infty} \frac{\eta_n}{n} = \frac{1}{(k+1)} = \theta.$$

## Moving maxima (cont'd)

This means that we can set  $r(n) = n$  for each  $n \in \mathbb{N}$  and have

$$\delta_n(u^{1/n}) \rightarrow u^\theta$$

as  $n \rightarrow \infty$  for each  $u \in [0, 1]$ . Consequently,

$$(M_n - a_n)/b_n \rightsquigarrow (\Phi_1)^\theta,$$

where  $(a_n)$  and  $(b_n)$  are the normalizing sequences from the iid case.

For a continuous  $F$  in the maximum domain of attraction of a  $H$  and

$$Y_i = F^{-1}\{\exp(-1/X_i)\}, \quad i \geq 1,$$

we further have that the maximum  $N_n = \max(Y_1, \dots, Y_n)$  satisfies

$$(N_n - a_n)/b_n \rightsquigarrow H^\theta,$$

where  $(a_n)$  and  $(b_n)$  are the normalizing sequences from the iid case.

## First nasty example

Let  $X_1, X_2, \dots$  be a sequence of identically distributed random variables such that

$$X_1 = X_2 = \dots \quad \text{almost surely}$$

and  $X_i \sim F$  for some continuous  $F$ .

Then for each  $n \in \mathbb{N}$ ,

$$C_n(u_1, \dots, u_n) = \min(u_1, \dots, u_n)$$

is the Fréchet–Hoeffding upper bound and

$$\delta_n(u) = u.$$

This means that for all  $n \in \mathbb{N}$ ,  $r(n) = 1$  and in fact

$$M_n \sim F.$$



## Second nasty example

From Example 5 in Mai (2018), there exists a sequence of identically distributed random variables such that for all  $n \geq 2$ ,

$$C_n(u_1, \dots, u_n) = \prod_{i=1}^n (1 - \theta)^{n-i} u_{(i)}$$

where  $u_{(1)} \leq \dots \leq u_{(n)}$  and  $\theta \in (0, 1)$ .

$C_n$  is the so-called **Cuadras–Augé copula** and we have

$$\delta_n(u) = u^{(1 - (1 - \theta)^n)/\theta} = u^{r(n)}$$

with  $r(n) = (1 - (1 - \theta)^n)/\theta$ , where  $r(n) \rightarrow 1/\theta$  as  $n \rightarrow \infty$ .

Clearly, for all  $x \in \mathbb{R}$  and  $n \in \mathbb{N}$ ,

$$\Pr(M_n \leq x) = \delta_n(F(x)) = \{F(x)\}^{r(n)} \rightarrow \{F(x)\}^{1/\theta}.$$

# There is a limit to what we can do

Let  $X_1, X_2, \dots$  be identically distributed according to a continuous  $F$ .

If there exists a function  $r: \mathbb{N} \rightarrow (0, \infty)$  such that

$$r(n) \rightarrow \varrho \in (0, \infty)$$

as  $n \rightarrow \infty$  and for each  $u \in [0, 1]$ ,

$$\delta_n(u^{1/r(n)}) \rightarrow D(u)$$

for some continuous  $D$ , then for all  $x \in \mathbb{R}$ , as  $n \rightarrow \infty$ ,

$$\Pr(M_n \leq x) \rightarrow D(F^\varrho(x)).$$

# Fisher–Tippett–Gnedenko Theorem: Version I

Let  $X_1, X_2, \dots$  be a sequence of identically distributed rvs with continuous marginal  $F$ , and suppose that the following conditions hold:

- (a)  $F$  is in the maximum domain of attraction of  $H_{\xi, \mu, \sigma}$ ;
- (b) There exists a rate function  $r : \mathbb{N} \rightarrow (0, \infty)$  with  $r(n) \rightarrow \infty$  as  $n \rightarrow \infty$  and a continuous function  $D$  such that for all  $u \in [0, 1]$ ,

$$\delta_n \{u^{1/r(n)}\} \rightarrow D(u).$$

If there exist sequences  $(b_n)$  and  $(a_n)$  such that  $a_n > 0$  for all  $n \in \mathbb{N}$ ,

$$\lim_{n \rightarrow \infty} \Pr \left( \frac{M_n - b_n}{a_n} \leq x \right) = G(x),$$

for all continuity points of a non-degenerate  $G$ , then there exist  $a > 0$  and  $b \in \mathbb{R}$  so that  $G = D \circ H_{\xi, \tilde{\mu}, \tilde{\sigma}}$  where  $\tilde{\mu} = (\mu - b)/a$  and  $\tilde{\sigma} = \sigma/a$ .

## Fisher–Tippett–Gnedenko Theorem: Version II

Let  $X_1, X_2, \dots$  be identically distributed according to a continuous  $F$ .

Suppose that there exists  $r: \mathbb{N} \rightarrow (0, \infty)$  and a bijection  $\lambda: (0, \infty) \rightarrow (0, \infty)$  such that the following conditions hold:

- (a)  $r(n) \rightarrow \infty$  as  $n \rightarrow \infty$  and for all  $t > 0$ ,  $r(\lceil tn \rceil)/r(n) = \lambda(t)$ ;
- (b)  $\delta_n$  is strictly increasing and  $\delta_n\{u^{1/r(n)}\} \rightarrow D$  pointwise for a *continuous and strictly increasing*  $D: [0, 1] \rightarrow [0, 1]$ .

If there exist sequences  $(b_n)$  and  $(a_n)$  such that  $a_n > 0$  for all  $n \in \mathbb{N}$ ,

$$\lim_{n \rightarrow \infty} \Pr \left( \frac{M_n - b_n}{a_n} \leq x \right) = G(x),$$

for all continuity points of a *non-degenerate*  $G$ , then  $G = D \circ H$  and  $H$  is GEV. If  $n/r(n) \rightarrow \theta$  for  $\theta > 0$ ,  $F$  is in the domain of attraction of  $H^\theta$ .

# Power diagonals

Suppose that for all  $n$ , some  $\eta_n$  and all  $u \in [0, 1]$ ,

$$\delta_n(u) = u^{\eta_n}.$$

Then if  $\eta_n \rightarrow \infty$  as  $n \rightarrow \infty$ , we can set  $r(n) = \eta_n$ .

If  $F$  is in the maximum domain of attraction of  $H_{\xi, \mu, \sigma}$  with norming constants  $(a_n)$ ,  $a_n > 0$  and  $(b_n)$ ,

$$\Pr\left(\frac{M_n - b_{\lceil r(n) \rceil}}{a_{\lceil r(n) \rceil}} \leq x\right) \rightarrow H_{\xi, \mu, \sigma}(x).$$

Note also that if  $n/\eta_n \rightarrow \theta$  for  $\theta > 0$  as  $n \rightarrow \infty$ , we get

$$\Pr\left(\frac{M_n - b_n}{a_n} \leq x\right) \rightarrow H_{\xi, \mu, \sigma}^{1/\theta}(x) = H_{\xi, \mu\theta, \sigma\theta}$$

upon setting  $r(n) = n$  so that  $D(u) = u^{1/\theta}$ ,  $u \in [0, 1]$ .

# A super easy example

Consider an **i.i.d.** sequence  $X_1, X_2, \dots$

Clearly, for each  $n \in \mathbb{N}$ ,

$$C_n(u_1, \dots, u_n) = \Pi(u_1, \dots, u_n) = u_1 \times \dots \times u_n.$$

is the **independence copula**.

Obviously,  $\delta_n(u) = u^n$  and we can set  $r(n) = n$ .

## Meta max-stable sequences

Consider a **simple max-stable sequence**  $Z_1, Z_2, \dots$ . This means that  $Z_i$  is **unit Fréchet** and for each  $n \in \mathbb{N}$ ,

$$C_n(u_1, \dots, u_n) = \exp\{-\ell_n(-\ln u_1, \dots, -\ln u_n)\}.$$

is an **extreme-value** copula with stdf  $\ell_n$ . For more flexibility, set

$$X_i = F^{-1}\{\exp(-1/Z_i)\}.$$

In this case,  $\delta_n(u) = u^{\eta_n}$ ,  $\eta_n = \ell_n(1, \dots, 1)$ . For a  $D$ -norm construction

$$\ell_n(x_1, \dots, x_n) = \mathbb{E}\left\{\max_{1 \leq i \leq n}(x_i W_i)\right\}$$

where  $W_1, W_2, \dots$  is a sequence of positive rvs. with unit mean,  $\eta_n$  is the **extremal coefficient** (Smith, 1990; Falk, 2019).

## Meta max-stable sequences (cont'd)

Our theory applies as soon as

$$\lim_{n \rightarrow \infty} \ell_n(1, \dots, 1) = \infty.$$

A classical example where this works is the **logistic stdf** with

$$\ell_n(x_1, \dots, x_n) = (|x_1|^\theta + \dots + |x_n|^\theta)^{1/\theta}$$

for  $\theta \geq 1$ . Here,

$$\eta_n = \ell_n(1, \dots, 1) = n^{1/\theta}.$$

However, the Cuadras–Augé copula is also extreme-value, and yet

$$\ell_n(1, \dots, 1) = (1 - (1 - \theta)^n)/\theta \rightarrow 1/\theta.$$



# Characterizing transformations leading to GEV limits

Suppose that  $D: [0, 1] \rightarrow [0, 1]$  is a **continuous and strictly increasing** distribution function. Then

$$D \circ H$$

is GEV for all GEV  $H$  **if and only if** there exist  $\alpha > 0$  and  $c > 0$  so that

$$D(u) = \exp\{-\alpha(-\ln u)^c\}$$

for all  $u \in [0, 1]$ .

## Convergence of the copula diagonal is necessary

Let  $X_1, X_2, \dots$  be a sequence of identically distributed rvs with continuous marginal  $F$ , and suppose that the following conditions hold:

- (a) There exist sequences  $(a_n)$ ,  $a_n > 0$  and  $(b_n)$  such that for all  $x \in \mathbb{R}$ ,

$$F^n(a_n x + b_n) \rightarrow H_{\xi, \mu, \sigma}(x).$$

- (b) There exists a rate function  $r : \mathbb{N} \rightarrow (0, \infty)$  with  $r(n) \rightarrow \infty$  as  $n \rightarrow \infty$  and  $1/r(n) = O(1/n)$ , and a distribution function  $D$  on  $[0, 1]$  so that

$$\Pr\left(\frac{M_n - b_{\lceil r(n) \rceil}}{a_{\lceil r(n) \rceil}} \leq x\right) \rightarrow D\{H_{\xi, \mu, \sigma}(x)\}.$$

for all continuity points  $x$  of  $D \circ H$ ,

Then for all continuity points  $u \in [0, 1]$  of  $D$ ,

$$\delta_n\{u^{1/r(n)}\} \rightarrow D(u).$$

## Example

Consider a Gaussian AR(1) process

$$X_n = \phi X_{n-1} + Z_n,$$

with  $X_0 = 0$ ,  $\phi \in [0, 1)$  and iid  $Z_n \sim \mathcal{N}(0, \sigma^2)$ . Here,  $C_n$  is Gaussian with

$$\delta_n(u) = \Phi_{\Sigma_n} \left( \sigma \Phi^{-1}(u) / \sqrt{1 - \phi^2}, \dots, \sigma \Phi^{-1}(u) / \sqrt{1 - \phi^2} \right)$$

where  $(\Sigma_n)_{ij} = \phi^{|i-j|} \sigma^2 / (1 - \phi^2)$ . Because  $\ln(n) \text{cov}(X_1, X_n) \rightarrow 0$ ,

$$\Pr \left( \frac{M_n - b_n}{a_n} \leq x \right) \rightarrow \Lambda(x).$$

where  $(a_n)$ ,  $a_n > 0$  and  $(b_n)$  are the normalizing sequences of the corresponding iid series. This means that for all  $u \in [0, 1]$ ,

$$\delta_n \{ u^{1/n} \} \rightarrow u.$$

# Time series with limited long-range dependence

Suppose that  $X_1, X_2, \dots$  is a stationary sequence and  $X_1^*, X_2^*, \dots$  iid with the same (marginal) distribution  $F$ .

Set  $M_n = \max(X_1, \dots, X_n)$  and  $M_n^* = \max(X_1^*, \dots, X_n^*)$ . If

$$\frac{M_n^* - b_n}{a_n} \rightsquigarrow H$$

and

$$\frac{M_n - b_n}{a_n} \rightsquigarrow H^\theta$$

for some  $\theta \in (0, 1]$ , then necessarily

$$\delta_n(u^{1/n}) \rightarrow u^\theta$$

as  $n \rightarrow \infty$ .

# Archimax diagonals

Take a **simple max-stable sequence**  $Z_1, Z_2, \dots$ , an independent positive rv.  $V$  with **Laplace transform**  $\psi$ , and set

$$Y_i = VZ_i.$$

We can obtain a sequence  $X_1, X_2, \dots$  with common distribution  $F$ , viz.

$$X_i = F^{-1}(\psi(1/Y_i))$$

$C_n$  is then an **Archimax copula**, i.e., for all  $n$  and  $u_1, \dots, u_n \in [0, 1]$ ,

$$C_n(u_1, \dots, u_n) = \psi[\ell_n\{\psi^{-1}(u_1), \dots, \psi^{-1}(u_n)\}];$$

where  $\psi$  is a completely monotone **Archimedean generator** and  $\ell_n$  an **stdf**. In particular, if  $Z_1, Z_2, \dots$  are i.i.d.,  $C_n$  is an **Archimedean copula**.

## Archimax diagonals (cont'd)

The diagonal of an Archimax copula  $C_n$  has the form

$$\delta_n(u) = \psi\{\eta_n \psi^{-1}(u)\}, \quad \eta_n = \ell_n(1, \dots, 1).$$

If  $\eta_n \rightarrow \infty$  as  $n \rightarrow \infty$  and  $1 - \psi(1/\cdot) \in \text{RV}_{-\rho}$ , then

$$r(n) = \left\lfloor \frac{1}{1 - \psi(1/\eta_n)} \right\rfloor, \quad D(u) = \psi[\{-\ln(u)\}^{1/\rho}]$$

so that

$$\Pr\left(\frac{M_n - b_{\lceil r(n) \rceil}}{a_{\lceil r(n) \rceil}} \leq x\right) \rightarrow \psi[\{-\ln H_{\xi, \mu, \sigma}(x)\}^{1/\rho}]$$

provided that  $F$  is in the maximum domain of attraction of  $H_{\xi, \mu, \sigma}$  with normalizing constants  $(a_n)$ ,  $a_n > 0$  and  $(b_n)$ .

## A bit more on Archimax diagonals

- It is worth noting that

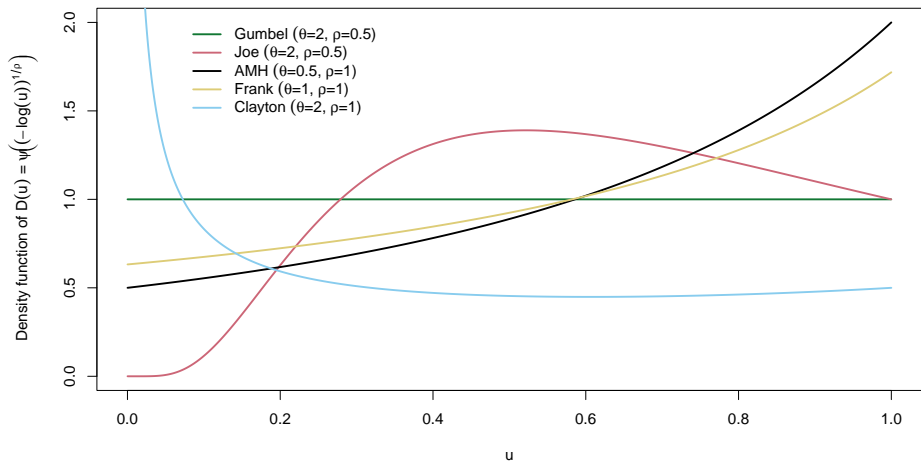
$$\psi[\{-\ln H_{\xi,\mu,\sigma}(x)\}^{1/\rho}] = \psi[-\ln H_{\rho\xi,\mu,\rho\sigma}(x)].$$

- When  $\ell_n(x_1, \dots, x_n) = x_1 + \dots + x_n$ ,  $X_1, X_2, \dots$  is an [exchangeable sequence](#) and  $C_n$  is [Archimedean](#). One then recovers the results of Berman (1962b), Ballerini (1994), and Wüthrich (2004).
- When  $-\psi'(0) \in (0, \infty)$  and  $\eta_n/n \rightarrow \theta$ , we get that

$$\Pr\left(\frac{M_n - b_n}{a_n} \leq x\right) \rightarrow \psi\{-\ln H_{\xi,\mu,\sigma}^{-\theta/\psi'(0)}(x)\}$$

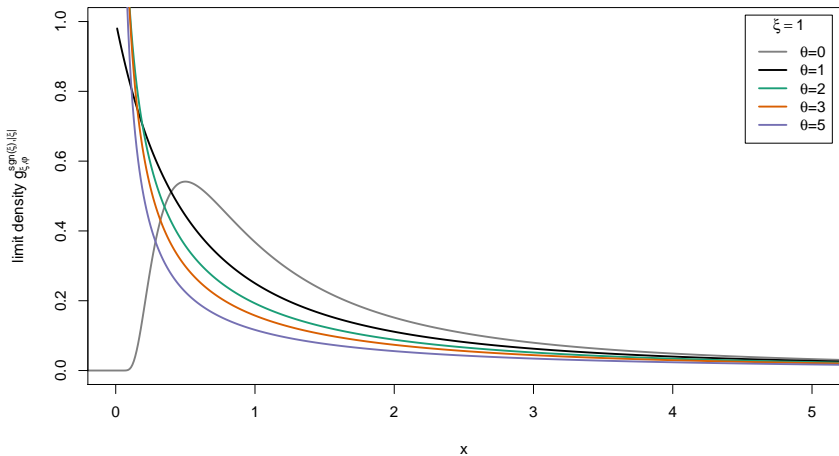
where  $(a_n)$ ,  $a_n > 0$  and  $(b_n)$  are the constants from the iid case.

# Illustration of $D$ for various $\psi$

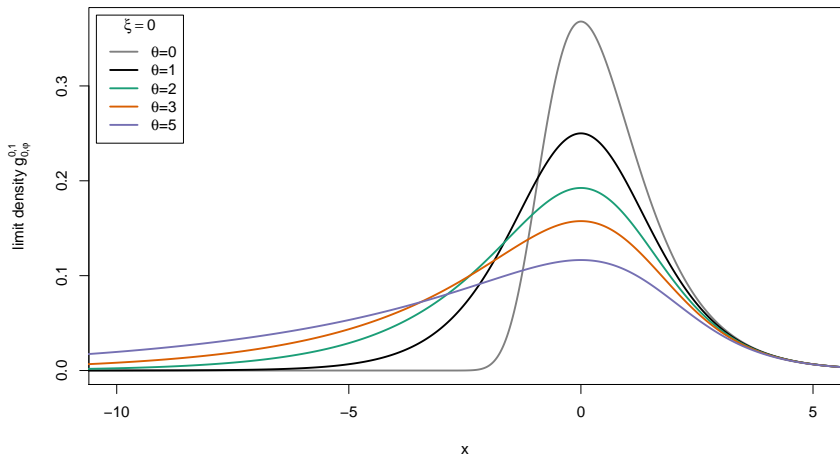




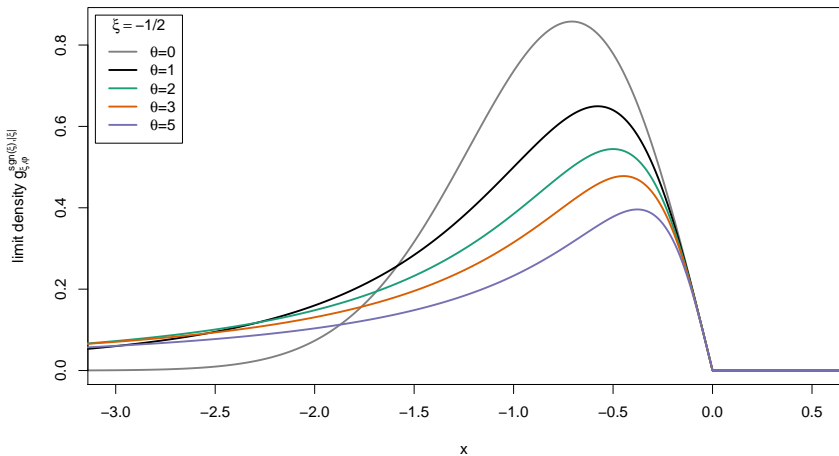
# Illustration when $\psi$ is Clayton and $\xi > 0$



# Illustration when $\psi$ is Clayton and $\xi = 0$



# Illustration when $\psi$ is Clayton and $\xi < 0$



# Convergence rates

Consider a sequence  $X_1, X_2, \dots$  of **identically distributed** random variables with continuous common distribution  $F$  such that

$$\sup_{x \in \mathbb{R}} |F^n(a_n x + b_n) - H(x)| \leq \beta(n),$$

for some EVD  $H$ . Suppose also that there exists  $r: \mathbb{N} \rightarrow (0, \infty)$  so that  $r(n) \rightarrow \infty$  as  $n \rightarrow \infty$  and

$$\sup_{u \in [0,1]} |\delta_n(u^{1/r(n)}) - D(u)| \leq s(n)$$

If  $D$  is Hölder continuous with constant  $K$  and parameter  $0 < \kappa \leq 1$ , then

$$\sup_{x \in \mathbb{R}} \left| \Pr \left( \frac{M_n - b_{\lceil r(n) \rceil}}{a_{\lceil r(n) \rceil}} \leq x \right) - D \circ H(x) \right| \leq K (\beta(\lceil r(n) \rceil) + 3e^{-1}/r(n))^\kappa + s(n)$$

## Moving maxima one last time

Consider  $X_1, X_2, \dots$ , such that  $X_i \sim \mathcal{N}(0, 1)$  and the dependence of the **moving maxima process**. We saw that for  $r(n) = n$  and  $u \in [0, 1]$ ,

$$\delta_n(u^{1/r(n)}) = \delta_n(u^{1/n}) = u^{(k+n)/(n(k+1))} \rightarrow u^{1/(k+1)}.$$

The limit is Hölder continuous with  $K = 1$  and  $\kappa = 1/(k+1)$ . Also,

$$\sup_{u \in [0,1]} |\delta_n(u^{1/r(n)}) - D(u)| = \frac{k}{n+k} \left(1 + \frac{k}{n}\right)^{-n/k}.$$

Hall (1979) provides sequences of constants so that

$$\sup_{x \in \mathbb{R}} |\Phi^n(a_n x + b_n) - \Lambda(x)| \leq 3/\ln(n).$$

The previous result shows that

$$\sup_{x \in \mathbb{R}} \left| \Pr(M_n \leq a_n x + b_n) - \Lambda(x)^{1/(k+1)} \right| \leq \frac{k}{n+k} \left(1 + \frac{k}{n}\right)^{-n/k} + \left(\frac{3}{\ln(n)}\right)^{1/(k+1)}$$

# Outlook

- Better understand the constraints on  $C_n$  (and its diagonal) when  $C_n$  is extendible.
- The development of inferential tools based on these results.
- Generalizations to the multivariate case.

Thank you for your attention!



# References I

- Ballerini, R. (1994). Archimedean copulas, exchangeability, and max-stability. *Journal of applied probability*, 31(2):383–390.
- Berman, S. M. (1962a). Equally correlated random variables. *Sankhya, Ser. A*, 24:155–156.
- Berman, S. M. (1962b). Limiting distribution of the maximum term in sequences of dependent random variables. *The Annals of Mathematical Statistics*, 33(3):894–908.
- Falk, M. (2019). *Multivariate Extreme Value Theory and D-Norms*. Springer, 1st edition.
- Hall, P. (1979). On the rate of convergence of normal extremes. *Journal of Applied Probability*, 16(2):433–439.
- Jaworski, P. (2009). On copulas and their diagonals. *Information Sciences*, 179:2863–2871.
- Leadbetter, M., Lindgren, G., and Rootzén, H. (1983). *Extremes and Related Properties of Random Sequences and Processes*. Springer-Verlag New York Inc., 1st edition.

## References II

- Mai, J.-F. (2018). Extreme-value copulas associated with the expected scaled maximum of independent random variables. *Journal of Multivariate Analysis*, 166:50 – 61.
- Smith, R. (1990). Max-stable processes and spatial extremes. Preprint Univ. North Carolina.
- Wüthrich, M. V. (2004). Extreme value theory and archimedean copulas. *Scandinavian Actuarial Journal*, 2004(3):211–228.