

Chapter 6

de Haan and Ferreira(2006)

EVT: bivariate case

Suppose $(X_1, Y_1), (X_2, Y_2), \dots$ be i.i.d. random vectors with distribution function F . Suppose that there exist sequences of constants $a_n, c_n > 0, b_n, d_n \in \mathbb{R}$ a distribution function G with non-degenerate marginals such that for all continuity points (x, y) of G ,

$$\lim_{n \rightarrow \infty} P\left(\frac{\max(X_1, X_2, \dots, X_n) - b_n}{a_n} \leq x, \frac{\max(X_1, X_2, \dots, X_n) - d_n}{c_n} \leq y\right) = G(x, y). \quad (6.1.1)$$

Any limit distribution function G in (6.1.1) with non-degenerate marginals is called a multivariate extreme value distribution.

EVT: bivariate case

Let F_1, F_2 denote the marginal distribution of F . Define $U_i(t) := F_i^{\leftarrow}(1 - 1/t)$, $i = 1, 2$. Then

$$\begin{aligned}\lim_{t \rightarrow \infty} \frac{U_1(nx) - b_n}{a_n} &= \frac{x_1^\gamma - 1}{\gamma_1}, \\ \lim_{t \rightarrow \infty} \frac{U_2(nx) - d_n}{c_n} &= \frac{x_2^\gamma - 1}{\gamma_2},\end{aligned}\tag{1}$$

EVT: bivariate case

Now, we return to (6.1.1), which can be written as

$$\lim_{n \rightarrow \infty} F^n(a_n x + b_n, c_n y + d_n) = G(x, y). \quad (6.1.8)$$

If $x_n \rightarrow u, y_n \rightarrow v$, then

$$\lim_{n \rightarrow \infty} F^n(a_n x_n + b_n, c_n y_n + d_n) = G(u, v). \quad (6.1.9)$$

Apply (6.1.9) with

$$x_n = \frac{U_1(nx) - b_n}{a_n}, y_n = \frac{U_2(ny) - d_n}{c_n}$$

then

$$\lim_{n \rightarrow \infty} F^n(U_1(nx), U_2(ny)) = G\left(\frac{x^\gamma - 1}{\gamma}, \frac{y^\gamma - 1}{\gamma}\right) := G_0(x, y)$$

Corollary 6.1.3

For any (x, y) for which $0 < G_0(x, y) < 1$,

$$\lim_{n \rightarrow \infty} n \{1 - F(U_1(nx), U_2(ny))\} = -\log G_0(x, y) \quad (6.1.11)$$

This also holds by replacing n by t , where t runs through the real numbers.

Exponent Measure

There are set functions ν, ν_1, ν_2 defined for all Borel sets $A \subset \mathbb{R}_+^2$ with

$$\inf_{(x,y) \in A} \max(x, y) > 0$$

such that

1.

$$\nu_n \{ (s, t) \in \mathbb{R}_+^2 : s > x \text{ or } t > y \} = n(1 - F(U_1(nx), U_2(ny))),$$

$$\nu \{ (s, t) \in \mathbb{R}_+^2 : s > x \text{ or } t > y \} = -\log G_0(x, y)$$

2. for all $a > 0$ the set functions ν, ν_1, ν_2, \dots are finite measures on $\mathbb{R}_+^2 [0, a]^2$

3. for each Borel set $A \subset \mathbb{R}_+^2$ with $\inf_{(x,y) \in A} \max(x, y) > 0$ and $\nu(\partial A) = 0$,

$$\lim_{n \rightarrow \infty} \nu_n(A) = \nu(A).$$

The measure ν is sometimes called the exponent measure of the extreme value distribution G_0 .

Homogeneity of ν

For any Borel set $A \subset \mathbb{R}_+^2$, with $\inf_{(x,y) \in A} \max(x,y) > 0$ and $\nu(\partial A) = 0$,

$$\nu(aA) = a^{-1} \nu(A)$$

The Spectral Measure

The homogeneity property of the exponent measure v suggests a coordinate transformation in order to capitalize on that.

Examples are

$$\begin{cases} r(x, y) = \sqrt{x^2 + y^2} \\ d(x, y) = \arctan \frac{y}{x} \end{cases}$$

$$\begin{cases} r(x, y) = x + y \\ d(x, y) = \frac{x}{x+y} \end{cases}$$

$$\begin{cases} r(x, y) = x \vee y \\ d(x, y) = \arctan \frac{x}{y} \end{cases}$$

The Spectral Measure

Let us start with the first transformation. Define for constants $r > 0$ and $\theta \in [0, \pi/2]$ the set

$$B_{r,\theta} = \left\{ (x, y) \in \mathbb{R}_+^{2*} : \sqrt{x^2 + y^2} > r \text{ and } \arctan \frac{y}{x} \leq \theta \right\}$$

Clearly $B_{r,\theta} = rB_{1,\theta}$ and hence

$$\nu(B_{r,\theta}) := r^{-1} \nu(B_{1,\theta}).$$

Set for $0 \leq \theta \leq \pi/2$,

$$\Psi(\theta) := \nu(B_{1,\theta}).$$

Theorem 6.1.4

There exist a finite measure on $[0, \pi]$ such that for $x, y > 0$,

$$G_0(x, y) = \exp \left(- \int_0^{\pi/2} \left(\frac{\cos \theta}{x} \vee \frac{\sin \theta}{y} \right) \Psi(d\theta) \right)$$

with the side functions

$$\int_0^{\pi/2} \cos \theta \Psi(d\theta) = \int_0^{\pi/2} \sin \theta \Psi(d\theta) = 1.$$

Define

$$L(x, y) = v \{ (s, t) \in \mathbb{R}_+^2 : s > 1/x \text{ or } t > 1/y \}.$$

Properties of the function L ,

- $L(ax, ay) = aL(x, y)$
- $L(x, 0) = L(0, x) = x$
- $x \vee y \leq L(x, y) \leq x + y$
- If X, Y are independent, then $L(x, y) = x + y$. If X, Y are completely positive dependent, then $L(x, y) = x \vee y$.
- L is continuous.
- L is convex.

Define the set Q_1 by

$$Q_1 := \{(x, y) \in \mathbb{R}_+^2 : -\log G_0(1/x, 1/y) \leq 1\}$$

The function R is defined as

$$R(x, y) = x + y - L(x, y)$$

The function χ is defined as

$$\chi(t) = -R(t, 1)$$

Theorem 6.2.1

The followings are equivalent.

1.

$$\lim_{t \rightarrow \infty} \frac{1 - F(U_1(tx), U_2(ty))}{1 - F(U_1(t), U_2(t))} = S(x, y)$$

with $S(x, y) = \log G((x^{\gamma_1} - 1)/\gamma, (y^{\gamma_2} - 1)/\gamma) / \log G(0, 0)$.

2. For all $r > 1$ and all $\theta \in [0, \pi/2]$ that are continuity point of Ψ ,

$$P\left(V^2 + W^2 > t^2 r^2 \text{ and } \frac{W}{V} \leq \tan \theta \mid V^2 + W^2 > t^2\right) \rightarrow r^{-1} \frac{\Psi(\theta)}{\Psi(\pi/2)}$$

Asymptotic Independence

Let (X_1, \dots, X_d) be a random vector with distribution function F . If

$$\frac{P(X_i > U_i(t), X_j > U_j(t))}{P(X_i > U_i(t))} = 0$$

for all $1 \leq i < j \leq d$, then

$$\lim_{n \rightarrow \infty} F^n(a_n^{(1)}x_1 + b_n^{(1)}, \dots, a_n^{(d)}x_1 + b_n^{(d)}) = \exp\left(-\sum_{i=1}^d (1 + \gamma_i x_i)^{-1/\gamma_i}\right).$$