

Applications of the multivariate tail process for extremal inference

Anja Janssen

University of Copenhagen, Denmark

Multivariate regularly varying time series are a common tool for modelling the dynamics of heavy-tailed processes of dimension larger than one. Let $(X_t)_{t \in \mathbb{Z}}$ be a stationary d -dimensional regularly varying process. The extremal behavior of this process can be described by the index $\alpha > 0$ of regular variation and the law of the so-called spectral tail process $(\Theta_t)_{t \in \mathbb{Z}}$, for which

$$\mathcal{L}\left(\frac{X_{-n}}{x}, \dots, \frac{X_m}{x} \mid \|X_0\| > x\right) \xrightarrow{w} \mathcal{L}(Y \cdot \Theta_{-n}, \dots, Y \cdot \Theta_m), \quad x \rightarrow \infty,$$

with a Pareto(α)-distributed random variable Y which is independent of $(\Theta_t)_{t \in \mathbb{Z}}$, cf. Basrak & Segers (2009). The spectral tail process satisfies a certain property which is sometimes called the “time change formula” that describes its behavior when shifted in time, cf. Basrak & Segers (2009). We are interested in estimating the law of Θ_t for $t \in \mathbb{Z}$ with a focus on the cases $t = 0, 1$. These two quantities are of interest in particular for Markov processes $(X_t)_{t \in \mathbb{Z}}$ where their joint distribution (together with the value of α) already determines the whole distribution of $(\Theta_t)_{t \in \mathbb{Z}}$, cf. Janßen and Segers (2014). By extending an idea used in Drees, Segers & Warchoł (2015) from the univariate to the multivariate case we show that it may be helpful for the estimation of Θ_1 to make use of the time change formula that gives us

$$P(\Theta_1 \in A) = E\left(\mathbf{1}_A\left(\frac{\Theta_0}{\|\Theta_{-1}\|}\right) \|\Theta_{-1}\|^\alpha\right), \quad A \in \mathbb{B}^d \text{ with } \mathbf{0} \notin A,$$

and use an indirect estimator instead of a direct one.

Furthermore, we try to detect independence of $\|\Theta_1\|$ and $\Theta_1/\|\Theta_1\|$ and explore the implications of this fact for the structure of the spectral tail process.

References

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