

Statistical Learning with Extreme Values

Master's program MVA, Université Paris-Saclay

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With the ubiquity of sensors, Big Data are now increasingly available in a wide variety of domains of human activity (science, industry, health, environment, commerce, security, ...) and rare/extreme phenomena are becoming observable in a significant manner. Before, such events were mainly 'out-of-sample' and Extreme Value Theory (EVT), the field of probability and statistics concerned with tails of distributions, tackled their study through (parametric) modelling essentially. With the need for analyzing extreme observations, carrying often the critical information to design solutions to applications (*e.g.* health monitoring of complex infrastructures) for which worst-case scenarios crucially matter, the most recent years have seen an increasing interest of the EVT research community towards novel machine learning algorithms and statistical learning theory, resonating with a continuing effort of the statistical community to address larger-dimensional problems with computationally feasible approaches (see *e.g.* the review [Engelke and Ivanovs \(2021\)](#)).

Let X be a random element (variable, vector, or function) of interest. One major goal of EVT is to provide probabilistic descriptions and statistical inference methods for the conditional distribution of $t^{-1}X$ given large $\|X\|$, where $\|\cdot\|$ is a semi-norm and t is a large threshold (see *e.g.* the monographs ([De Haan and Ferreira \(2007\)](#); [Resnick \(2008\)](#)). In applications, relevant thresholds t may be as high as the largest observation among n realizations of X . Probabilistic extrapolation is then needed to use the information brought by a subsample of size $k_n \ll n$ composed of the observations with the largest semi-norms. This requires sound theoretical assumptions pertaining to the theory of regular variation and maximum domains of attraction, ensuring that a limit distribution $\mu = \lim \text{Law}(t^{-1}X \mid \|X\| > t)$ exists as $t \rightarrow \infty$, up to suitable standardization. This stylized setting encompasses a wide range of applications in various scientific disciplines and risk management where extremes have tremendous impact, such as climate science, insurance, industrial monitoring systems ([Beirlant et al. \(2004\)](#)).

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This course aims at introducing the students with the most recent development of statistical learning viewpoints on EVT. From a theoretical perspective they will be offered an overview of recent statistical learning theory for rare events, in addition to the necessary probabilistic background on extreme value theory and regular variation. Theoretical development will be motivated and illustrated by recent successful algorithms for handling extreme values, be it for anomaly detection, extreme event classification or dimension reduction in distributional tails.

Course structure: Each session is approximately divided into a 2h lecture and a 1h tutorial. There will be short lab sessions during courses #2 and #4. Bring your laptops.

The second to last lecture is a, exercise Q&A session.

The last session is a written exam.

Grading

Written exam, including theoretical and practical questions related to the lab sessions.

Syllabus (9*3h)

1. Basics of EVT: learning from block maxima.

Context and applications in risk management and anomaly detection. Fisher and Tipett's theorem with elements of proof. Method of block maxima

Tutorial: derivation of norming constants, numerical illustration for the weak convergence of block maxima, case studies, choice of the block size.

2. Peaks-Over-Thresholds (POT) and Regular variation.

Link between POT and block maxima. Generalized Pareto distributions. Basics of regular variation and vague convergence. Informal introduction to the Hill estimator.

Tutorial: Lab session: EVT methods in univariate problems. POT modeling on case studies - threshold choice - Hill estimator in practice - Quantile estimation.

3. Regular Variation II, tail measures and weak convergence.

More on regular variation - Karamata representation theorem - weak consistency of the Hill estimator - Quantile estimation.

Tutorial: Visualization of weak convergence of tail measures - Elements of proof for Karamata.

4. Multivariate extremes.

Reduction to the standard case - characterizing max-id distributions - characterizing simple max-stable distributions - Angular measure - Multivariate Peaks-over-threshold

Tutorial: Lab session - Simulation - Basics of non-parametric estimation in moderate dimension (kernel and histogram methods).

5. **Statistical learning guarantees for extremes.**

A refresher on statistical learning theory - improved guarantees on rare classes - guarantees on empirical measures of dependence in multivariate settings. **Tutorial:** Elements of proof and numerical illustration of the error bounds

6. **Supervised learning with extreme values.**

Learning with extreme covariates (classification/regression)

Tutorial: Lab session.

7. **High dimensional extremes.**

Notions of sparsity in multivariate extremes - Applications to anomaly detection

Tutorial: PCA - multiple subspace clustering - Dimension reduction with extreme targets - Anomaly detection

8. **Exercises and Q&A session.**

Help with exercises, preparing the final exam.

9. **Written exam**

References

- Beirlant, J., Goegebeur, Y., Segers, J., and Teugels, J. L. (2004). *Statistics of extremes: theory and applications*, volume 558. John Wiley & Sons.
- De Haan, L. and Ferreira, A. (2007). *Extreme value theory: an introduction*. Springer Science & Business Media.
- Engelke, S. and Ivanovs, J. (2021). Sparse structures for multivariate extremes. *Annual Review of Statistics and Its Application*, 8:241–270.
- Resnick, S. I. (2008). *Extreme values, regular variation, and point processes*, volume 4. Springer Science & Business Media.