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Linear aggregation of frontier estimators

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Abstract

In this paper, we propose a linear aggregation technique to estimate the frontier of a dataset. Our approach is based on recent techniques of Conditional Extreme Value Analysis.

Let X be a random variable and $p(x)$ its probability density function, $F(x) = P(X \leq x)$ its distribution function, and $\Psi(x) = 1 - F(x)$ its complementary distribution function. We can define a new random variable, Y_n , as the maximum of n copies of the random variable X : $Y_n = \max\{X_1, X_2, \dots, X_n\}$. Y_n is the n -sample maximum of the random variable X . If the events generating the realizations of X are independent, the cumulative distribution of Y_n may be expressed as $[F(y)]^n$. Upon definition of a renormalized variable $S_n = (Y_n - b_n)/a_n$, the extreme value theorem establishes that

Theorem *If*

$$\lim_{n \rightarrow \infty} P(S_n < s) = \lim_{n \rightarrow \infty} F^n(a_n s + b_n) = H(s) \quad (1)$$

where $a_n > 0$ and b_n are normalization constants, then the function $H(s)$ in Eq. (1) must be one of the three following types:

- *EV1 or Gumbel:* $H(s) = \exp(-\exp(-s))$
- *EV2 or Fréchet:* $H(s) = \exp(-s^{-\alpha})$
- *EV3 or Weibull:* $H(s) = \exp(-|s|^\alpha)$

The three asymptotic types, EV1-EV3, can be thought of as special cases of a single Generalized Extreme Value distribution (GEV) :

$$H_{GEV}(s) = \exp \left\{ - \left(1 + \gamma \frac{s - \mu}{\sigma} \right)_+^{-1/\gamma} \right\} \quad (2)$$

where $(\cdot)_+ = \max(\cdot, 0)$, μ is the location parameter, $\sigma > 0$ is the scale parameter, and γ is a shape parameter. The limit $\gamma = 0$ corresponds to the EV1 distribution, $\gamma > 0$ to the EV2 distribution (with $\alpha = 1/\gamma$) and $\gamma < 0$ to the EV3 distribution (with $\alpha = -1/\gamma$). The function $H_{GEV}(s)$ is usually fitted

to the cumulative distribution of non-normalized maxima, so that the location parameter μ and the scale parameter σ are the renormalization parameters b_n and a_n respectively. However, it is important to note that the distribution describing the n -sample maximum will strictly be a GEV only for large values of n . How large the value of n needs to be should be determined by analyzing the convergence properties based on the observed realizations of X .

When the studied phenomena depends on a covariate, one has to deal with Conditional Extreme Value Analysis (CEVA). This branch of statistics has become very active these past ten years, the main contributions to this domain are listed below:

- Theoretical issues: [65, 63, 64, 26, 32, 20, 31, 33, 30, 5, 27, 13, 51, 25, 4, 1, 82, 81]
- Quantile regression: [56, 19, 66]
- Application to finance: [10, 3, 35, 11, 58, 80]

Edge estimation [53, 52], frontier estimation [62, 78, 75, 77, 72, 16, 14, 15, 9, 2, 39, 8, 61, 17, 70, 69, 21, 79, 60, 22, 48, 36, 59, 23, 24, 68, 6, 73, 12, 55, 57, 67, 54, 43, 45, 18, 29, 71, 74, 76, 38] and boundary estimation [44, 47, 46, 42, 41, 40, 49, 28, 50, 7, 34, 37] are particular cases of CEVA. They are embedded in the situation where the conditional extreme-value index is negative.

Our main result is the following:

Theorem *Let $\hat{\phi}_1(x), \dots, \hat{\phi}_K(x)$ be K consistent estimators of a frontier $\phi(x)$. Let $w_1, \dots, w_K$ be K weights summing to one. Then,*

$$\hat{\phi}(x) := \sum_{i=1}^K w_i \hat{\phi}_i(x)$$

is a consistent estimator of the frontier $\phi(x)$.

Proof. The proof is based on the continuity of the sum function and on the property that $\sum_{i=1}^K w_i = 1$. Let us highlight that the positivity of the weights is not required.

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