

The paper I am going to present is “Trends in Extreme Value Indices”. This paper is published on Journal of the American Statistical Association in 2020. It is written by Laurens de Haan and Chen Zhou.

This is the outline of My presentation. First, I will give a introduction to this study. And then I will present the main methodology. And Finally, a simulation study and a real data application will be presented.

Now, we turn to the first part.

Classic extreme value analysis assumes that the observations are independent and identically distributed. Recent studies aim at dealing with case when observations are drawn from different distribution. This paper considers a continuously changing extreme value index and try to estimate the functional extreme value index accurately.

Mathematically, consider a set of distributions  $F_s(x)$  for  $s \in [0, 1]$ . Now we have independent random variables  $X_1$  to  $X_n$  and the distribution function of  $X_i$  is  $F_{i/n}$ .

Here, assume that the function  $F_s(x)$  is continuous with respect to  $s$  and  $x$ . Also, assume that  $F_s$  belongs to the maximum domain of attraction with extreme value index  $\gamma(s)$ .

This paper considers the case that the function  $\gamma$  is positive. This means that for each  $s$ , the function  $F_s$  is a Frechet distribution. Also, the function  $\gamma$  is assumed to be continuous.

The goal of this paper is to estimate the function  $\gamma$  and test the hypothesis that  $\gamma = \gamma_0$  for some given function  $\gamma_0$ . In particular, it can be applied to test whether the extreme value index remains at a constant level across all observations.

We first discuss how to estimate the function  $\gamma$  locally. And then I will present how to obtain a global estimator and how to use this to do the hypothesis testing that  $\gamma = \gamma_0$  for some given function  $\gamma_0$ .

The idea for estimating  $\gamma(s)$  locally is similar to the kernel density estimation. More specifically, use only observations  $X_i$  in the  $h$ -neighborhood of  $s$ . And the mathematical definition of the  $h$ -neighborhood is displayed as this, where  $h$  is bandwidth and  $h$  satisfies that as  $n$  goes to infinity,  $h$  will go to infinity and  $n$  times  $h$  will go to infinity.

Correspondingly there will be  $[2nh]$  observations in the  $h$ -neighborhood for each  $s \in [h, 1 - h]$ . To apply any known estimators for extreme value index, such as the Hill estimator, we choose an intermediate sequence  $k = k(n)$  such that as  $n \rightarrow \infty, k \rightarrow \infty$  and  $k/n \rightarrow 0$ . This choice of  $k$  is very standard in the extreme value analysis.

Then, we will use the top  $[2kh]$  order statistics among the  $[2nh]$  local observation in the  $h$ -neighborhood of  $s$  to estimate  $\gamma(s)$ . Rank the  $[2nh]$  observations in the  $h$ -neighborhood of  $s$  as this. Then we can apply the Hill estimator to estimate

$\gamma(s)$  locally. And the mathematical definition of the local Hill estimator is defined as this.

This is the main idea of how to estimate the function  $\gamma$  locally. To obtain the asymptotic properties of this estimator, we need some conditions.

The first condition is second order condition. Suppose there exist a continuous negative function  $\rho(s)$  and a set of function  $A_s(t)$ , such that the equation (3) holds. Assume a second order condition is quite standard and sometimes necessary for the extreme value analysis. What is different here is that this is a functional form of the second order condition.