### Stochastic Processes and their Statistics in Finance

# A numerical characteristic of extreme values

### Takaaki SHIMURA

(The Institute of Statistical Mathematics)

# 0.1 Plan of Presentation

- 1. Introduction
- 2. Main result
- 3. Limit distributions
- 4. Summary

# 1 Introduction

1.1 Motivation and Problem

We consider a numerical characteristic of random numbers. Especially,

" Extremely large random numbers or small random numbers".

## Classification of normal random numbers by the first figure

 $egin{aligned} [0,1): 0.59922, \ 0.39319, \ 0.11950, \ 0.01336 \ [1,2): 1.04194, \ 1.43943, \ 1.38955, \ 1.66662 \ [2,3): 2.19377, \ 2.40794, \ 2.14139, \ 2.32582 \ [3,4): 3.06956, \ 3.86446, \ 3.20402, \ 3.04337, \ 3.07787, \ 3.16713, \ 3.45392, \ 3.04813 \end{aligned}$ 

# 1.2 Mathematical setting

[Transformation on  $[1, \infty)$  to [0, 1)] We consider a transformation from a large number to a number in [0, 1), which moves the decimal point and excludes the first figure.

$$d_1 d_2 d_3 \dots d_n \dots d_{n+1} \dots$$
 in  $[10^{n-1}, 10^n)$   
 $\rightarrow 0.d_2 d_3 \dots$  in  $[0, 1)$ ,

where n is a natural number. We call  $d_m$  the mth figure. Let F be a distribution on real line with infinite end point :  $\sup\{x : F(x) < 1\} = \infty$ and X be a random variable with distribution F.

If  $X = d_1 d_2 d_3 \dots d_n d_{n+1} \dots$  on  $[1, \infty)$ , then  $Y = 0.d_2 d_3 \dots d_n d_{n+1} \dots$  is a random variable on [0, 1).

We consider the distribution of Y for large X, which implies the behavior of the large random number except the first figure.

### Define N and K as

N : the number of figures before the decimal point of X :  $10^{N-1} \leq X < 10^N,$ 

K : the first figure of X :  $K10^{N-1} \leq X < (K+1)10^{N-1}$ 

Then previous transformation is written as

$$Y = X/10^{N-1} - K.$$

Let us consider the conditional distribution.

$$F^{k,n}(y) = P(Y \le y | K = k, N = n),$$
  
for  $k = 1, 2, ..., 9.$ 

Our main interest is in the behavior of  $F^{k,n}$  for each k as  $n \to \infty$ .

1.3 Classification of distributions Denote the tail of a distribution F by

$$\bar{F}(x) = 1 - F(x).$$

F is said to have regularly varying tail with index  $\alpha>0$  if

 $\lim_{x \to \infty} \bar{F}(\lambda x) / \bar{F}(x) = \lambda^{-\alpha} \text{ for each } \lambda > 0.$ 

For example, the Cauchy, the Pareto, the F and the Ziph distributions have regularly varying tail.

## [Regularly varying function]

A positive measurable function f(x) is said to be regularly varying with exponent (index)  $\rho \ (\in \mathbf{R}) \ (f \in \mathbf{RV}_{\rho})$  if for each  $\lambda > 0$ 

$$\lim_{x \to \infty} \frac{f(\lambda x)}{f(x)} = \lambda^{\rho}.$$

In the case of  $\rho = 0$ , it is called slowly varying.  $f(x) \in \mathbf{RV}_{\rho}$  is written as  $f(x) = x^{\rho}l(x)$  with a slowly varying l(x). e.g.  $f(x) = x^2 \log x$  [Rapidly varying function] f(x) is said to be a rapidly varying with exponent  $\infty$  ( $f \in \mathbf{RV}_{\infty}$ ) if for each  $\lambda > 1$ 

$$\lim_{x \to \infty} \frac{f(\lambda x)}{f(x)} = \infty.$$

For example,  $f(x) = \exp x$  is rapidly varying.

In the same way,  $f \in \mathbf{RV}_{-\infty}$  if for each  $\lambda > 1$  $\lim_{x\to\infty} \frac{f(\lambda x)}{f(x)} = 0.$  Rapidly varying tail distributions are various.

• Very rapid tail decay : the normal distribution and the Rayleigh distribution.

 Middle tail decay: the exponential type, i.e. the exponential distribution, the Gamma distribution, the Chi-square distribution, the generalized inverse Gaussian distribution.

• Little bit heavy tail : the log-normal distribution.

 $[\Pi$ -varying function]

A positive measurable function f(x) on  $(0,\infty)$ is  $\Pi$ -varying if there exists a positive function a(x) on  $(0,\infty)$  such that for  $\lambda > 0$ ,

$$\lim_{x \to \infty} \frac{f(\lambda x) - f(x)}{a(x)} = \log \lambda.$$

We write  $f \in \Pi$  or  $f \in \Pi(a)$ . a(x) is called an auxiliary function of f(x).

### For example,

 $f(x) = \log x$  is  $\Pi$ -varying with a(x) = 1.

 $f(x) = \log x + 2^{-1} \sin(\log x)$  is NOT  $\Pi$ -varying.

Roughly speaking, a  $\Pi$ -varying function is nondecreasing slowly varying with good (local) property.

Distribution with  $1/\Pi$ -varying tail is so to speak "super heavy" and not so familiar. The log Cauchy distribution is the case. Thus we have three kind of tail behaviors :

- $\cdot \bar{F}(x) \in \mathbf{RV}_{-\alpha}(\alpha > 0),$
- $\cdot F(x) \in \mathbf{RV}_{-\infty}$ ,
- $\cdot 1/\overline{F}(x) \in \Pi \ (\subset \mathbf{RV_0}).$

These classes cover most well-known distributions with infinite endpoint.

Distributions with finite endpoint are classified in the same way.

Let F be a distribution on  $(-\infty, 0)$  with the endpoint 0 :  $\sup\{x : F(x) < 1\} = 0$ .

We say that F has regularly varying tail at 0 if

$$\bar{F}(-1/x) \in \mathbf{RV}_{\alpha}(\alpha < 0).$$

Rapidly varying and  $1/\Pi$  varying tail at 0 are defined in a similar way. The definition for a general finite endpoint is also done.

# [Examples]

- (i) Regularly varying tail at their finite endpoint: The Beta distribution and the Pareto distribution.
- (ii) Rapidly varying tail at their finite endpoint : The exponential distribution and the log-normal distribution.

# 2 Main result2.1 Large random numbers

Remember

$$F^{k,n}(y) = P(Y \le y | K = k, N = n),$$
  
for  $k = 1, 2, ..., 9.$ 

Our main interest is in the behavior of  $F^{k,n}$  for each k as  $n \to \infty$ .

$$\begin{aligned} F^{k,n}(y) &= P(Y \le y | K = k, N = n) \\ &= \frac{P(k10^{n-1} \le X \le (k+y)10^{n-1})}{P(k10^{n-1} \le X < (k+1)10^{n-1})} \\ &= \frac{\bar{F}(k10^{n-1}) - \bar{F}((k+y)10^{n-1})}{\bar{F}(k10^{n-1}) - \bar{F}((k+1)10^{n-1})} \\ &= \frac{1 - \bar{F}((k+y)10^{n-1}) / \bar{F}(k10^{n-1})}{1 - \bar{F}((k+1)10^{n-1}) / \bar{F}(k10^{n-1})}.\end{aligned}$$

# The third equality holds for continuous F, but it is not essential.

$$F^{k,n}(y) = \frac{1 - \bar{F}((k+y)10^{n-1})/\bar{F}(k10^{n-1})}{1 - \bar{F}((k+1)10^{n-1})/\bar{F}(k10^{n-1})}.$$

If  $\overline{F}(x) \in \mathbf{RV}_{-\infty}$ , for x > 0

$$\lim_{n \to \infty} \bar{F}((k+y)10^{n-1})/\bar{F}(k10^{n-1}) = 0.$$

If  $F(x) \in \mathbf{RV}_{-\alpha}(\alpha > 0)$ ,

 $\lim_{n \to \infty} \bar{F}((k+y)10^{n-1})/\bar{F}(k10^{n-1}) = (1+y/k)^{-\alpha}.$ 

If  $1/\bar{F}(x) \in \Pi$ ,  $\bar{F}(k10^{n-1}) - \bar{F}((k+y)10^{n-1})$  $\sim \log(1+\frac{y}{k})a(k10^{n-1}).$ 

### **Theorem 1**

(i) If  $\overline{F}(x) \in \mathbf{RV}_{-\infty}$ , for every k,

$$\lim_{n \to \infty} F^{k,n}(y) = 1_{\{y \ge 0\}},$$

where  $1_A$  denotes the indicator function of a set A. Namely,  $F^{k,n}$  converges to  $\delta_0$  (a distribution concentrates at  $\{0\}$  as  $n \to \infty$ .

## (ii) If $F(x) \in \mathbf{RV}_{-\alpha}(\alpha > \mathbf{0})$ , for $0 \le y \le 1$ ,

$$\lim_{n \to \infty} F^{k,n}(y) = \frac{1 - (1 + \frac{y}{k})^{-\alpha}}{1 - (1 + \frac{1}{k})^{-\alpha}}$$

(iii) If  $1/\bar{F}(x) \in \Pi$ , for  $0 \le x \le 1$ ,

$$\lim_{n \to \infty} F^{k,n}(y) = \frac{\log(1 + \frac{y}{k})}{\log(1 + \frac{1}{k})}.$$

Let

$$G_{\alpha}^{k}(y) = \frac{1 - (1 + \frac{y}{k})^{-\alpha}}{1 - (1 + \frac{1}{k})^{-\alpha}},$$
$$G_{0}^{k}(y) = \frac{\log(1 + \frac{y}{k})}{\log(1 + \frac{1}{k})}.$$

(i) and (iii) are regarded as the limit of (ii) :  $G^k_{\alpha}(y)$  converges to  $\delta_0$  and  $G^k_0$  as  $\alpha \to \infty$  $\alpha \to 0$ , respectively. We add some secondary results.

First, the tail condition in the case (iii) is  $1/\Pi$ -varying, not general slowly varying. The following shows that this restriction is

significant.

**Theorem 2** For any distribution F with slowly varying tail and any distribution G on [0, 1), there exists a distribution  $F_G$  such that

$$\lim_{x \to \infty} \bar{F}_G(x) / \bar{F}(x) = 1 \text{ and } F_G^{k,n} = G.$$

### **Proof**.

Let  $X_1 = K10^{N-1}$  and  $X_2 = X - X_1$ . Since  $X_1 \le X < 2X_1$ , we have  $P(X > x) \sim P(X_1 > x)$ . For  $Z \sim G$ , set  $Y = X_1 + 10^{N-1}Z$ .  $P(X > x) \sim P(X_1 > x) \sim P(Y > x)$ . The rate of converge to  $\delta_0$  in (i) is as follows.

**Theorem 3**  $F(x) \in \mathbf{RV}_{-\infty}$  Moreover, assume that F is absolutely continuous and its hazard function h(t) belongs to  $\mathbf{RV}_{\rho}(\rho \ge -1)$ . For  $0 \le y < 1$ ,

$$\lim_{n \to \infty} \frac{1}{10^{n-1}h(10^{n-1})} \log \overline{F^{k,n}}(y) = -c(\rho, k, y),$$

#### where

$$\begin{split} c(\rho,k,y) &= \begin{cases} & (\rho+1)^{-1}\{(k+y)^{\rho+1}-k^{\rho+1}\} & \rho > -1 \\ & & \log(1+\frac{y}{k}) & \rho = -1 \end{cases} \end{split}$$

 $c(\rho,k,y)$  expresses the rate of convergence to  $\delta_0.$ 

[Property of  $c(\rho, k, y)$  as a function of k]

(i) If 
$$-1 \le \rho < 0$$
,  $c(\rho, k, y)$  is a decreasing function of  $k$ .

(ii)  $c(0, k, y) = (\rho + 1)^{-1}y$  does not depend on k Especially,  $F^{k,n}$  does not depend on k if F is an exponential distribution.

(iii) If  $\rho > 0$ ,  $c(\rho, k, y)$  is an increasing function of k.

## 2.2 Small random numbers

From now, we deal with distributions with finite endpoint. We assume F has finite end point :  $x_F = \sup\{x : F(x) < 1\} < \infty.$ 

Let X be a random variable with distribution F and consider the length until the endpoint :

$$x_F - X$$
.

Let  $x_F = 0$  for simplicity and transform from  $(-\infty, 0)$  to [0, 1):

 $X = -0.0 \cdots 0d_1 d_2 d_3 \ldots \in [-\infty, 0), \ d_1 \neq 0$  $\to Y = 0.d_2 d_3 \ldots \in [0, 1).$  Define a normalized random variable  $\boldsymbol{Y}$  as

$$Y = -10^N X - K,$$

where K is the first non-zero figure of X and N is the number of zeros before K:

$$-10^{-N+1} < X \le -10^{-N}$$
$$-10^{N}X - 1 < K \le -10^{N}X.$$

Y expresses the behavior of X except the first non-zero figure.

# The following conditional distribution is considered.

$$F^{k,n}(y) = P(Y \le y | K = k, N = n),$$
  
for  $k = 1, 2, ..., 9.$ 

As the large case, the behavior of  $F^{k,n}$  as  $n \to \infty$  for each k is investigated and similar results are given.

### Theorem 4

(i) If 
$$\overline{F}(-1/x) \in \mathbf{RV}_{-\infty}$$
, then for every  $k = 1, 2, \dots, 9$ ,

$$\lim_{n \to \infty} F^{k,n}(y) = 1_{\{y \ge 1\}},$$

where  $1_A$  is the indicate function of a set A.

# (ii) If $\overline{F}(-1/x) \in \mathbf{RV}_{\alpha}(\alpha < \mathbf{0})$ , then for $0 \le y \le 1$ ,

$$\lim_{n \to \infty} F^{k,n}(y) = \frac{(1 + \frac{y}{k})^{-\alpha} - 1}{(1 + \frac{1}{k})^{-\alpha} - 1}$$

(iii) If 
$$1/\bar{F}(-1/x) \in \Pi$$
, then for  $0 \le y \le 1$ ,

$$\lim_{n \to \infty} F^{k,n}(y) = \frac{\log(1 + \frac{y}{k})}{\log(1 + \frac{1}{k})}.$$

These limit distributions for small random numbers have the same form as ones for large case.

But the parameter range is different. While the large case is in  $\alpha \ge 0$ , the small case moves in  $\alpha(\le 0)$ .

Thus we get two parameter distribution class :

$$G_{\alpha}^{k} (\alpha \in (-\infty, \infty), k = 1, 2, \dots, 9.),$$

**Theorem 5** For arbitrary distribution F with slowly varying tail at 0 and arbitrary distribution G on [0, 1), there exists a distribution  $F_G$  such that

$$\lim_{x \uparrow 0} \overline{F}_G(x) / \overline{F}(x) = 1 \text{ and } F_G^{k,n} = G.$$

**Theorem 6** Assume that  $\overline{F}(-1/x) \in \mathbf{RV}_{-\infty}$ is absolutely continuous and its hazard function satisfies  $h(-1/t) \in \mathbf{R}_{\rho} (\rho \geq \mathbf{1})$ . For  $0 < y \leq 1$ ,

$$\lim_{n \to \infty} \frac{10^n}{h(-10^{-n})} \log F^{k,n}(y) = -\tilde{c}(\rho, k, y),$$

#### where

$$\begin{split} \tilde{c}(\rho,k,y) \\ = \begin{cases} & (\rho-1)^{-1}\{(k+y)^{1-\rho} - (k+1)^{1-\rho}\} & \rho > 1 \\ & \log(\frac{k+1}{k+y}) & \rho = 1 \end{cases} \end{split}$$

In this case,  $\tilde{c}(\rho, k, y)$  is decreasing on k.

# 3 Property of limit distributions 3.1 Property variety of $\alpha$ and k $G_{\alpha}^{k}$ $(k = 1, 2, ..., 9, \alpha \in (-\infty, \infty)).$

$$G_{\alpha}^{k}(x) = \frac{1 - (1 + \frac{x}{k})^{-\alpha}}{1 - (1 + \frac{1}{k})^{-\alpha}} \quad (\alpha \neq 0),$$
$$G_{0}^{k}(x) = \frac{\log(1 + \frac{x}{k})}{\log(1 + \frac{1}{k})} \quad (\alpha = 0).$$

Note  $\alpha \ge 0$  : large case.  $\alpha \le 0$  : small case.

For each k,  $G_{\alpha}^{k}$  moves between  $\delta_{1}$  and  $\delta_{0}$ . **Proposition 1** 

(i)  $G_{\alpha}^{k}$  converges to  $\delta_{0}$  ad  $\alpha \to \infty$ . (ii)  $G_{\alpha}^{k}$  converges to  $\delta_{1}$  as  $\alpha \to -\infty$ . (iii)  $G_{-1}^{k}$  is the uniform distribution on [0, 1].

# The density functions of $G^k_{\alpha}(\alpha \ge 0)$ denoted by $p^k_{\alpha}(y)$ are given as

$$p_{\alpha}^{k}(y) = \frac{\alpha k^{-1} (1 + \frac{y}{k})^{-\alpha - 1}}{1 - (1 + \frac{1}{k})^{-\alpha}},$$
$$p_{0}^{k}(y) = \frac{1}{\log(1 + \frac{1}{k})} \frac{1}{k + y}$$

for  $0 \le y \le 1$ .

## **Proposition 2**

(i) For each k, p<sup>k</sup><sub>α</sub>(y) is a decreasing (resp. constant, increasing) function of y and α > -1 (resp. α = -1, α < -1).</li>
(ii) p<sup>k</sup><sub>α</sub>(0) is an increasing function of α for each k. While p<sup>k</sup><sub>α</sub>(1) is decreasing function of α for each k.

(iii)  $p_{\alpha}^{k}(0)$  is a decreasing (resp. constant, increasing) function of k for  $\alpha > -1$  (resp.  $\alpha = -1, \alpha < -1$ ).  $p_{\alpha}^{k}(1)$  is an increasing (resp. constant, decreasing) function of k for  $\alpha > -1$  (resp.  $\alpha = -1, \alpha < -1$ ).

The probability density tends to be flat as k increases.

The distribution function  $G^k_{\alpha}$  has the following monotonicity.

## **Proposition 3**

- (i)  $G_{\alpha}^{k}(y)$  is an increasing function of  $\alpha$  for each k and y.
- (ii)  $G_{\alpha}^{k}(y)$  is is a decreasing (resp. constant, increasing) function of k for  $\alpha > -1$  (resp.  $\alpha = -1, \alpha < -1$ ).

## $M_{\alpha}^{k}$ denotes the mean of $G_{\alpha}^{k}$ . **Corollary 1**

(i) M<sup>k</sup><sub>α</sub> is a decreasing function of α for each k.
(ii) M<sup>k</sup><sub>α</sub> is an increasing (resp. constant, decreasing) function of k for α > −1 (resp. α = −1, α < −1).</li>

## 3.2 The limit distribution of m th figure

 $H_m^k$ : the distribution of the (m-1)th figure after the decimal point of  $G_{\alpha}^k$  (m = 2, 3, ...).  $H_m^k$  is a distribution on  $\{0, 1, \dots, 9\}$ Originally,  $H_m^k$  implies the distribution of mth figure of a original random number.

$$X = d_1 d_2 d_3 \dots d_m \dots$$

with  $d_1 = k$ .

Although  $H_m^k(j)$  is decreasing for j from Proposition 2, this property disappears as mgoes to  $\infty$ .

**Proposition 4** For each k,  $H_m^k$  converges to the uniform distribution on  $\{0, 1, \ldots, 9\}$  as  $m \to \infty$ .

This suggest the distribution of the second figure expresses the original distribution.

[Distribution of the second figure  $\alpha = 1$ ]

$$k = 1: H_2^1(0) = 2/11$$
  $H_2^1(9) = 1/19$   
 $k = 9: H_2^9(0) = 10/91$   $H_2^9(9) = 1/11$ 

 $H_2^1(0)/H_2^1(9) = 38/11 = 3.45...$  $H_2^9(0)/H_2^9(9) = 110/91 = 1.20...$ 

The ratio is  $2.857\ldots$ 

[Distribution of the second figure  $\alpha = 0$ ]

$$k = 1 : H_2^1(0) = \log(11/10) / \log 2$$
$$H_2^1(9) = \log(20/19) / \log 2$$
$$k = 9 : H_2^9(0) = \log(91/90) / \log(10/9)$$
$$H_2^9(9) = \log(100/99) / \log(10/9)$$

 $H_2^1(0)/H_2^1(9) = 1.85...$  $H_2^9(0)/H_2^9(9) = 1.10...$ 

The ratio is 1.69...

## 4 Summary

 Random numbers have a numerical characteristic. Especially, it is remarkable in extreme values.

 An extreme value (conditioned by the first figure) converges to a limit distribution depends on each tail behavior.

• The limit distribution depends on the tail behavior and the first figure.

### References

Bingham, N.H., Goldie, C.M. and Teugels, J.L.(1987). Regular Variation. Cambridge. Cambridge University press.

Shimura, T.(2012). Limit distribution of a roundoff error, Statistics and Probability Letters
82, 713-719.
Shimura, T.(2013). A numerical characteristic of

extreme values, submitted.

## Thank you for your attention!