

# ESTIMATION OF THE LOCATION PARAMETER OF $\alpha$ -STABLE DISTRIBUTIONS

TROUSH N.N. <sup>†</sup>, CHEN HAILONG <sup>†,‡</sup>

<sup>†</sup> *Department of Theory Probability and Mathematical Statistic, Belorussian State University, 4 Hezavisimosti - Minsk*

<sup>‡</sup> *Computer Science and Technology College, Harbin University of Science and Technology, China, Xuefu Road 52 - Harbin*

e-mail: TroughNN@bsu.by, hrbustchl@mail.ru

## Abstract

In this paper a method of estimating location parameter  $\mu$  of  $\alpha$ -stable distribution for  $\alpha \in (0, 2]$  is provided. Firstly, the estimates of parameters  $\alpha$  and  $\sigma$  with the characteristic function method (*CF*) are obtained, then the method of estimating of parameter  $\mu$  based on the means of non-skewed transformations.

## 1 Introduction

In practice the estimation of parameters  $\alpha$ -stable distributions is very important and now several estimating methods are known: maximum likelihood method [1,2], Hill method [3], quantile method [4], characteristic function method [5] etc. For estimating of parameters of symmetric distributions (*S $\alpha$ S*) fractional lower order moment method, extreme value method, logarithmic moment method [6,7] and others are used. But now in the literature only a few methods of estimating location parameter  $\mu$  exist. When  $\alpha \rightarrow 1$  and  $\beta \neq 0$ , it is difficult to estimate  $\mu$ . The estimation for parameter  $\mu$  in the case of  $\alpha \in (1, 2]$  was given in [8]. Estimation of parameter  $\mu$  for the case  $\alpha \in (0, 2]$ , based on the property of stable distributions and transformations for  $\alpha$ -stable random variables is given in this paper.

## 2 Method *CF* for estimation of parameters

From [9] we know that  $X$  is stable random variable if and only if the logarithm of its characteristic function  $\varphi_X(t)$  has the following representation:

$$\ln \varphi_X(t) = \begin{cases} -\sigma^\alpha |t|^\alpha (1 - i\beta \operatorname{sign}(t) \tan(\frac{\pi\alpha}{2})) + i\mu t, & \alpha \neq 1, \\ -\sigma |t| (1 + i\beta \operatorname{sign}(t) \frac{2}{\pi} \ln|t|) + i\mu t, & \alpha = 1, \end{cases} \quad (1)$$

where  $\alpha \in (0, 2]$ ,  $\beta \in [-1, 1]$ ,  $\sigma > 0$ ,  $\mu \in R$ ,  $t \in R$ . We will use a designation  $X \sim S_\alpha(\sigma, \beta, \mu)$ .

The basic idea of *CF* method is to estimate characteristic function from the sample data and then estimate parameters  $\alpha$ ,  $\sigma$  and  $\beta$ ,  $\mu$  by using the real and imaginary parts of the logarithm of its characteristic function. However, since characteristic function of  $\alpha$ -stable distributions is not continuous at  $\alpha = 1$ , it is difficult to obtain a good estimation of parameters  $\beta$  and  $\mu$ .

### 3 Method of the estimation of the location parameter

On the basis of properties of stable distributions, we establish the following result.

**Theorem 1** [10, 11]. Let  $X_k$  be independent identically distributed stable random variables with parameters  $\alpha, \sigma, \beta, \mu$ , i.e.  $X_k \sim S_\alpha(\sigma, \beta, \mu)$ , and  $a_k \in R, k = \overline{1, n}$ , then

$$Z = \sum_{k=1}^n a_k X_k \sim S_\alpha\left(\left(\sum_{k=1}^n |a_k|^\alpha\right)^{1/\alpha} \sigma, \frac{\sum_{k=1}^n a_k^{(\alpha)}}{\sum_{k=1}^n |a_k|^\alpha} \beta, \sum_{k=1}^n a_k \mu\right),$$

where  $a_k^{(\alpha)} = \text{sign}(a_k) |a_k|^\alpha$ .

Three types of transformations of  $\alpha$ -stable random variables can be derived from Theorem 1.

**Lemma 1** Let  $X_k \sim S_\alpha(\sigma, \beta, \mu), k = \overline{1, n}$  and  $X_k^C \sim X_{3k} + X_{3k-1} - 2X_{3k-2}$ , then

$$X_k^C \sim S_\alpha\left([2 + 2^\alpha]^{1/\alpha} \sigma, \left[\frac{2 - 2^\alpha}{2 + 2^\alpha}\right] \beta, 0\right).$$

**Lemma 2** Let  $X_k \sim S_\alpha(\sigma, \beta, \mu), k = \overline{1, n}$  and  $X_k^D \sim X_{3k} + X_{3k-1} - 2^{1/\alpha} X_{3k-2}$ , then

$$X_k^D \sim S_\alpha(4^{1/\alpha} \sigma, 0, [2 - 2^{1/\alpha}] \mu).$$

**Lemma 3** Let  $X_k \sim S_\alpha(\sigma, \beta, \mu), k = \overline{1, n}$  and  $X_k^S \sim X_{2k} - X_{2k-1}$ , then

$$X_k^S \sim S_\alpha(2^{1/\alpha} \sigma, 0, 0).$$

Here  $X^C$  denotes centered transformation,  $X^D$  - non-skewed transformation,  $X^S$  - symmetrized transformation. Observe that the lengths of sequences  $X^C$  and  $X^D$  are 1/3 length of initial sequence, and length of sequence  $X^S$  - 1/2 length of initial sequence. Since the given in lemma 2 transformation can transform random sequence  $X_k \sim S_\alpha(\sigma, \beta, \mu)$  to non-skewed distribution  $X_k^D \sim S_{\alpha^D}(\sigma^D, \beta^D, \mu^D)$  and  $\mu^D = \text{median}(X_k^D)$  [9], then,

$$\hat{\mu} = \hat{\mu}^D (2 - 2^{1/\alpha})^{-1}.$$

**Theorem 2** Estimation  $\hat{\mu}$  of parameter  $\mu$  is unbiased and consistent.

### 4 Simulation

Considering that the length of sequence should be a multiple of 3, we simulate 9999  $\alpha$ -stable random variables and estimate location parameter  $\mu$  for  $\alpha \in [0.2, 1.8]$  (see figure 1).

Comparison of estimates of parameter  $\mu$  by method in the paper and CF method in case of symmetric distribution  $S_\alpha(1, 0, 0)$  with various values of parameter  $\alpha$  is shown in figure 2.

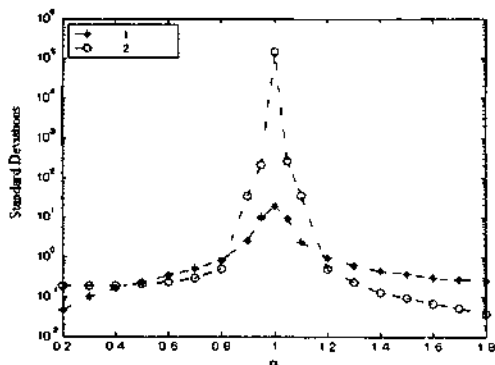
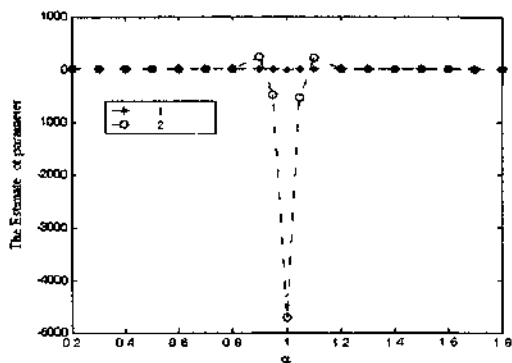


Figure 1: The estimate and standard deviations of estimation of parameter  $\mu$  for  $S_\alpha(2,0.9,1)$ : 1-method in paper; 2-CF method

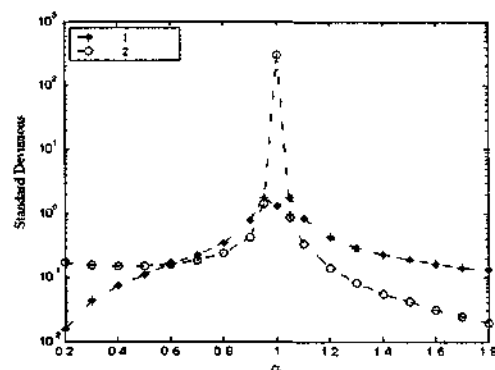
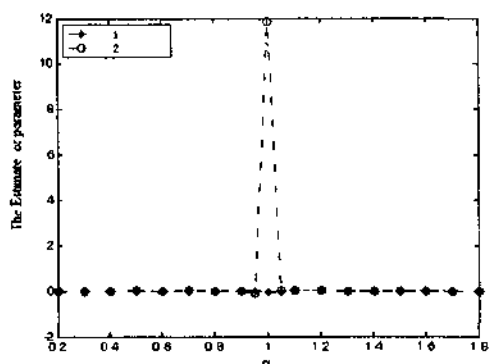


Figure 2: The estimate and standard deviations of estimation of parameter  $\mu$  for  $S_\alpha(1,0,0)$ : 1-method in paper; 2-CF method

Table 1: Estimation of parameters  $\mu$  with  $\alpha$  close to 1

$S_\alpha(2, -0.9, 1)(\mu = 1)$					
$\alpha$	0.90	0.95	1.00	1.05	1.10
Method in paper	1.2398 (2.4528)	2.0959 (9.7951)	-0.1201 (18.26)	0.4134 (9.2316)	0.8140 (2.3457)
CF Method	$2.39 \times 10^2$ (32.503)	$-4.78 \times 10^2$ ( $2.08 \times 10^2$ )	$-4.70 \times 10^3$ ( $1.489 \times 10^5$ )	$-5.30 \times 10^2$ ( $2.57 \times 10^2$ )	$2.25 \times 10^2$ (34.543)
$S_\alpha(1, 0, 0)(\mu = 0)$					
$\alpha$	0.90	0.95	1.00	1.05	1.10
Method in paper	$1.0295 \times 10^{-3}$ (0.7852)	$-2.4514 \times 10^{-2}$ (1.7413)	$-3.7650 \times 10^{-2}$ (1.3164)	$-3.6197 \times 10^{-2}$ (1.7552)	$2.2179 \times 10^{-2}$ (0.8410)
CF Method	$3.5412 \times 10^{-3}$ (0.4245)	$-7.6779 \times 10^{-2}$ (1.4505)	11.8660 ( $2.9999 \times 10^2$ )	$3.9201 \times 10^{-2}$ (0.8615)	$1.1582 \times 10^{-2}$ (0.3355)

## 5 Conclusion

We can compare the estimation of parameter  $\mu$  by the method offered in paper with the *CF* method at various  $\alpha$  on the basis of the results obtained from two simulations of  $\alpha$  - stable distributions which are shown table 1. The estimation of parameter  $\mu$  for symmetric distribution  $S_\alpha(1,0,0)$  by the *CF* method is better than the estimation for asymmetrical distribution. When estimating parameter  $\mu$  with  $\alpha$  close to 1, the method that proposed in article is better than the *CF* method. And for asymmetrical distribution  $S_\alpha(2,-0.9,1)$  the method of estimating parameter  $\mu$  proposed in paper is better than the *CF* method with  $\alpha$  close to 1.

## References

- [1] Bodenschatz J.S., Nikias C.L. (1999). Maximum-Likelihood symmetric  $\alpha$ -stable parameter estimation. *IEEE Transactions on Signal Processing*. Vol. 47, pp. 1382-1384.
- [2] Chen Hailong, Le Hong son. (2010). Simulation and estimation of parameter of stable random variables. *Journal of Belorussian State University*. Vol. 1, pp. 123-127.
- [3] Chen Hailong. (2009). About estimation of the index of heavy tails of distributions. *The collection of scientific papers of 66-th scientific conference of students and post-graduate students of Belorussian State University*. Vol. 2, pp. 252-255.
- [4] McCulloch J.H. (1986). Simple Consistent Estimations of Stable Distribution Parameters. *Communications in Statistics-Simulation and Computation*. Vol. 15, pp. 4, 1109-1136.
- [5] Koutrouvelis I.A. (1980). Regression-type estimation of the parameters of stable laws. *Journal of the American Statistical Association*. Vol. 75, pp. 918-928.
- [6] Ma Xinyu, Nikias C.L. (1995). Parameter estimation and blind channel identification in impulsive signal environments. *IEEE Transactions on Signal Processing*. Vol. 43, pp. 2884-2897.
- [7] Tsihrintzis G.A., Nikias C.L. (1996). Fast estimation of the parameters of alpha-stable impulsive interference. *IEEE Transactions on Signal Processing*. Vol. 44, pp. 1492-1503.
- [8] Samorodnitsky G., Naqqu M. (1994). *Stable Non-Gaussian Random Processes: Stochastic Models with Infinite Variance*. Chapman and Hall, New York, London.
- [9] Zolotarev V.M. (1983). *One - Dimensional Stable Distributions*. Science, Moscow.
- [10] Kuruoglu E.E. (2001). Density parameter estimation of skewed  $\alpha$ -stable distributions. *IEEE Transaction on Signal Processing*. . Vol. 49, pp. 2192-2201.
- [11] Trough N.N. (2008). *The statistical analysis of estimation of spectral density of stable random processes*. Belorussian State University, Minsk.