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# Sequential Interval Estimation of a Location Parameter with the Fixed Width in the Non-regular Case

*Dedicated to Professor Masafumi Akahira on his 60th birthday*

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**Abstract:** For a location-scale parameter family of distributions with a finite support, a sequential confidence interval with a fixed width is obtained for the location parameter, and its asymptotic consistency and efficiency are shown. Some comparisons with the Chow-Robbins procedure are also done.

**Keywords:** Coverage probability; Extreme value; Non-regular case; Sequential interval estimation.

**Subject Classifications:** 62L12; 62F25.

## 1. INTRODUCTION

Suppose that we are to estimate a location parameter  $\theta$  of a sequence of random observations  $X_1, X_2, \dots, X_n, \dots$  with unknown scale  $\xi$ . We would like to obtain sequentially a confidence interval of fixed width  $2d$  with confidence coefficient  $1 - \alpha$ . Obviously we can not obtain a fixed sample size procedure if  $\xi$  is unknown. There are many works on the fixed-width interval estimation of normal mean (see, e.g. Ghosh et al. (1997)).

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Suppose that  $X_1, X_2, \dots, X_n, \dots$  is a sequence of independent and identically distributed (i.i.d.) random variables according to the uniform distribution on the interval  $(\theta - (\xi/2), \theta + (\xi/2))$ , where  $\theta (\in \mathbb{R}^1)$  and  $\xi (> 0)$  are unknown. Let  $X_{(1)} := \min_{1 \leq i \leq n} X_i$ ,  $X_{(n)} := \max_{1 \leq i \leq n} X_i$ . Then the midrange and the range are  $M_n := (X_{(1)} + X_{(n)})/2$ ,  $R_n := X_{(n)} - X_{(1)}$ , respectively. Akahira and Koike (2005) considered a stopping rule:

$$\tau_1 := \inf \left\{ n \geq n_0 \mid \frac{R_n}{n-1} \leq -\frac{2d}{\log \alpha} \right\},$$

where  $n_0 (\geq 2)$  is an initial size of sample. They showed the asymptotic consistency and efficiency of the estimation procedure  $(\tau_1, [M_{\tau_1} - d, M_{\tau_1} + d])$ .

In this paper, we consider the case of a location-scale parameter family of distributions with a finite support on the interval  $(\theta - \xi a, \theta + \xi a)$ , where  $\theta$  and  $\xi$  are unknown, and obtain a sequential confidence interval of  $\theta$  with fixed width  $2d$  and confidence coefficient  $1 - \alpha$ , and show its asymptotic consistency and efficiency. Some comparisons with the Chow-Robbins procedure are also done.

## 2. ASYMPTOTIC DISTRIBUTIONS OF THE EXTREME VALUES

In this section we consider the asymptotic distributions of the extreme values for distributions with a finite support, in a similar way to Akahira (1991) and Akahira and Takeuchi (1995).

Let  $Z_1, Z_2, \dots$ , be a sequence of independent and identically distributed (i.i.d.) random variables according to the density function  $f_0(x - \theta)$  ( $\theta \in \mathbb{R}^1$ ) with respect to the Lebesgue measure. We assume the following conditions: (A1)  $f_0(x)$  has a finite support  $(-a, a)^1$  ( $a > 0$ ), i.e.,  $f_0(x) > 0$  for  $-a < x < a$ , and  $f_0(x) = 0$  otherwise.

(A2)  $f_0(x)$  is continuously differentiable in the open interval  $(-a, a)$  and

$$\lim_{x \rightarrow -a+0} f_0(x) = c, \quad \lim_{x \rightarrow a-0} f_0(x) = c',$$

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<sup>1</sup>If the support of  $f_0$  is  $(-a, b)$  ( $a \neq b$ ), then the normalized midrange does not converge to  $\theta$  in probability as  $n \rightarrow \infty$ .

where  $c$  and  $c'$  are some positive constants.

(A3)  $f_0(x)$  satisfies

$$\begin{aligned} f_0(x) &\approx g(x+a)^\gamma \quad (x \rightarrow -a+0), \\ f_0(x) &\approx g'|x-a|^\gamma \quad (x \rightarrow a-0), \end{aligned}$$

where  $\gamma, g$  and  $g'$  are some positive constants<sup>2</sup>.

Putting  $Z_{(1)} := \min_{1 \leq i \leq n} Z_i$ ,  $Z_{(n)} := \max_{1 \leq i \leq n} Z_i$ ,  $U := n(Z_{(1)} + a - \theta)$  and  $V := n(Z_{(n)} - a - \theta)$ , we have the following lemma (cf. Akahira (1991), Akahira and Takeuchi (1995)).

**Lemma 1.** *Under the conditions (A1) and (A2), the joint(j.) p.d.f.*

$f_{U,V}^{(n)}(u, v)$  *of*  $(U, V)$  *satisfies*

$$f_{U,V}^{(n)}(u, v) \rightarrow \begin{cases} cc' \exp\{c'v - cu\} & (v < 0 < u), \\ 0 & (\text{otherwise}). \end{cases} \quad (2.1)$$

as  $n \rightarrow \infty$ .

*Proof.* The j.p.d.f.  $f_{U,V}^{(n)}(u, v)$  of  $(U, V)$  is

$$\begin{aligned} &f_{U,V}^{(n)}(u, v) \\ &= \begin{cases} \frac{n-1}{n} \left\{ F\left(a + \frac{v}{n}\right) - F\left(-a + \frac{u}{n}\right) \right\}^{n-2} f_0\left(-a + \frac{u}{n}\right) f_0\left(a + \frac{v}{n}\right) & (v < 0 < u), \\ 0 & (\text{otherwise}), \end{cases} \end{aligned}$$

where  $F(x) = \int_{-\infty}^x f_0(u) du$ . Hence, by its expansion, we have the desired result.  $\square$

Next, we consider the location-scale parameter family of distributions with a finite support  $(\theta - \xi a, \theta + \xi a)$ . Suppose that  $X_1, X_2, \dots, X_n, \dots$  is a sequence of i.i.d. random variables with the p.d.f.  $(1/\xi)f_0((x - \theta)/\xi)$ , where  $\theta \in \mathbb{R}$  and  $\xi > 0$ . Put  $Y_i := (X_i - \theta)/\xi$  for each  $i = 1, 2, \dots$ , and  $Y_{(1)} := \min_{1 \leq i \leq n} Y_i$ ,  $Y_{(n)} := \max_{1 \leq i \leq n} Y_i$ . Letting  $S := n(Y_{(1)} + Y_{(n)})/2$  and  $T = n(Y_{(1)} - Y_{(n)} + 2a)/2$ , we have the asymptotic (as.) j.p.d.f. of  $(S, T)$

$$f_{S,T}(s, t) = \begin{cases} 2cc' \exp\{-(c - c')s - (c + c')t\} & (t > |s|), \\ 0 & (\text{otherwise}). \end{cases}$$

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<sup>2</sup>If the converging order  $\gamma$  is different, then the normalized midrange does not converge to  $\theta$  in probability as  $n \rightarrow \infty$ .

Then the as. marginal(m.) p.d.f.'s of  $S$  and  $T$  are given by

$$f_S(s) = \begin{cases} Ke^{-2cs} & (s \geq 0), \\ Ke^{2c's} & (s < 0), \end{cases} \quad (2.2)$$

$$f_T(t) = \begin{cases} \frac{2cc'}{c'-c} (e^{-2ct} - e^{-2c't}) & (t > 0 \text{ and } c \neq c'), \\ 4c^2te^{-2ct} & (t > 0 \text{ and } c = c'), \\ 0 & (\text{otherwise}), \end{cases}$$

respectively, where  $K = 2cc'/(c + c')$ .

In the case when  $\lim_{x \rightarrow -a+0} f_0(x) = \lim_{x \rightarrow a-0} f_0(x) = 0$ , we need another lemma. Putting  $U' := n^{1/(\gamma+1)}(Z_{(1)} + a - \theta)$  and  $V' := n^{1/(\gamma+1)}(Z_{(n)} - a - \theta)$ , we have the following lemma in a similar way to Lemma 1.

**Lemma 2.** *Under the conditions (A1) and (A3), the j.p.d.f.  $f_{U',V'}^{(n)}(u, v)$  of  $(U', V')$  satisfies*

$$f_{U',V'}^{(n)}(u, v) \rightarrow \begin{cases} gg'(-uv)^\gamma \exp\left\{-\frac{g'}{\gamma+1}(-v)^{\gamma+1} - \frac{g}{\gamma+1}u^{\gamma+1}\right\} & (v < 0 < u), \\ 0 & (\text{otherwise}). \end{cases}$$

as  $n \rightarrow \infty$ .

The proof is omitted since it is similar to the one of Lemma 1.

From Lemma 2,  $U'$  and  $(-V')$  are asymptotically, independently distributed according to Weibull distributions.

### 3. CONSTRUCTING CONFIDENCE INTERVAL

In this section we construct a sequential confidence interval for  $\theta$ . In the first place, we consider the case under the conditions (A1) and (A2). For  $0 < \alpha < 1$ , let  $l_0$  be the solution<sup>3</sup> of  $l$  for the equation

$$\frac{c + c'}{cc'}\alpha = \frac{e^{-2cl}}{c} + \frac{e^{-2c'l}}{c'}.$$

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<sup>3</sup>It can be shown easily that such  $l_0$  exists uniquely.

If  $\xi$  is known, we have from (2.2) that

$$\begin{aligned} P\{|M_n - \theta| \leq d\} &= P\{n|M_n - \theta|/\xi \leq dn/\xi\} \\ &\approx \int_{-dn/\xi}^{dn/\xi} f_S(s) ds \\ &= 1 - \frac{cc'}{c+c'} \left( \frac{e^{-2cnd/\xi}}{c} + \frac{e^{-2c'nd/\xi}}{c'} \right), \end{aligned}$$

where “ $\approx$ ” means that the distribution of  $n|M_n - \theta|/\xi$  is approximated by the asymptotic distribution. Letting  $n^* = l_0\xi/d$ , we have for  $n \geq n^*$

$$1 - \frac{cc'}{c+c'} \left( \frac{e^{-2cnd/\xi}}{c} + \frac{e^{-2c'nd/\xi}}{c'} \right) \geq 1 - \alpha.$$

$n^*$  is referred as the asymptotically *optimal* size of samples if  $\xi$  is known. Note that  $n(M_n - \theta)/\xi = S$  and  $R_n/\xi = -(T/n) + 2a$ . Now we take as the stopping rule

$$\tau_2 := \inf \left\{ n \geq n_0 \mid \frac{R_n}{n-1} \leq \frac{2ad}{l_0} \right\}, \quad (3.1)$$

where  $n_0 (\geq 2)$  is an initial size of sample. Then we obtain the asymptotic properties of the estimation procedure  $(\tau_2, [M_{\tau_2} - d, M_{\tau_2} + d])$  as follows.

**Theorem 1.** *For the sequential estimation procedure  $(\tau_2, [M_{\tau_2} - d, M_{\tau_2} + d])$ , the following hold.*

- (i)  $\lim_{d \rightarrow 0+} P\{|M_{\tau_2} - \theta| \leq d\} = 1 - \alpha$  (asymptotic consistency).
- (ii)  $\tau_2/n^* \xrightarrow{\text{a.s.}} 1$  ( $d \rightarrow 0+$ ).
- (iii)  $E(\tau_2)/n^* \rightarrow 1$  ( $d \rightarrow 0+$ ) (asymptotic efficiency).

*Proof.* (i) From Lemma 1 of Chow and Robbins (1965), the stopping rule  $\tau_2$  given by (3.1) satisfies

$$\lim_{d \rightarrow 0+} \frac{d\tau_2}{\xi l_0} = 1 \quad \text{a.s.} \quad (3.2)$$

Since  $S = n(M_n - \theta)/\xi$  converges in distribution to a distribution with the density given by (2.2) as  $n \rightarrow \infty$ , it follows from Theorem 1 of Anscombe (1952) that  $\tau_2(M_{\tau_2} - \theta)$  converges in distribution to the same distribution as  $d \rightarrow 0+$ . Hence, since  $d\tau_2/\xi \xrightarrow{\text{a.s.}} l_0$  as  $d \rightarrow 0+$  from (3.2), it follows that

$$\begin{aligned} \lim_{d \rightarrow 0+} P\{|M_{\tau_2} - \theta| \leq d\} &= \lim_{d \rightarrow 0+} P\{\tau_2|M_{\tau_2} - \theta|/\xi \leq d\tau_2/\xi\} \\ &= \int_{-l_0}^{l_0} f_S(s) ds = 1 - \alpha. \end{aligned} \quad (3.3)$$

(ii) From (3.2) and the definition of  $l_0$ , we have  $\tau_2/n^* = \tau_2 d/(l_0 \xi) \xrightarrow{\text{a.s.}} 1$  as  $d \rightarrow 0+$ .

(iii) From Lemma 2 of Chow and Robbins (1965), we have the desired result.  $\square$

**Remark.** In particular, if  $c = c'$ , then  $l_0 = -\log \alpha/(2c)$  and  $\tau_2$  given in (3.1) is expressed as

$$\tau_2 = \inf \left\{ n \geq n_0 \mid \frac{R_n}{n-1} \leq -\frac{4acd}{\log \alpha} \right\},$$

which is equal to  $\tau_1$  when the underlying distribution is uniform distribution on the interval  $(\theta - (\xi/2), \theta + (\xi/2))$ .

In the second place, we compare this with the Chow-Robbins procedure. Let  $X_1, X_2, \dots$  be a sequence of i.i.d. random variables with the mean  $\theta$  and the variance  $\sigma^2$ . Let  $\bar{X}_n := \sum_{i=1}^n X_i/n$ ,  $s_n^2 = \sum_{i=1}^n (X_i - \bar{X}_n)^2/(n-1)$ . Chow and Robbins (1965) considered a stopping rule defined by

$$\tau_{CR} := \inf \{ n \geq n_0 \mid n \geq u_{\alpha/2}^2 d^{-2} s_n^2 \},$$

where  $u_{\alpha/2}$  is the upper  $\alpha/2$  point of  $N(0, 1)$  and  $n_0 (\geq 2)$  is an initial size of samples. They showed the asymptotic consistency and efficiency of the estimation procedure  $(\tau_{CR}, [\bar{X}_{\tau_{CR}} - d, \bar{X}_{\tau_{CR}} + d])$ .

Since, from Theorem 2.2 of Akahira and Koike (2005), Theorem 1 and Theorem of Chow and Robbins (1965),

$$\tau_1 \approx \frac{\log \alpha}{\log(1 - (2d/\xi))} \approx \frac{-\xi \log \alpha}{2d}, \quad \tau_2 \approx l_0 \xi/d, \quad \tau_{CR} \approx u_{\alpha/2}^2 \sigma^2/d^2,$$

as  $d \rightarrow 0+$ , we have  $\tau_1/\tau_{CR}, \tau_2/\tau_{CR} \rightarrow 0$  ( $d \rightarrow 0+$ ). Therefore  $\tau_1, \tau_2$  is asymptotically better than  $\tau_{CR}$  in the sense of the average size of sample.

Furthermore, we consider the case under the conditions (A1) and (A3). By putting  $S' := n^{1/(\gamma+1)}(Y_{(1)} + Y_{(n)})/2$  and  $T' := n^{1/(\gamma+1)}(Y_{(1)} - Y_{(n)} + 2a)/2$ , the as.j.p.d.f. of  $(S', T')$  and the as.m.p.d.f.'s of  $S'$  and  $T'$  are obtained from Lemma 2. In a similar way to (3.3), we take  $l_0$  satisfying  $\int_{-l_0}^{l_0} f_{S'}(s) ds = 1 - \alpha$  for the as.m.p.d.f.  $f_{S'}(s)$  of  $S'$ .

If  $\xi$  is known, we have

$$\begin{aligned} P\{|M_n - \theta| \leq d\} &= P\{n^{1/(\gamma+1)}|M_n - \theta|/\xi \leq dn^{1/(\gamma+1)}/\xi\} \\ &\approx \int_{-dn^{1/(\gamma+1)}/\xi}^{dn^{1/(\gamma+1)}/\xi} f_{S'}(s) ds, \end{aligned}$$

where “ $\approx$ ” means that the distribution of  $n^{1/(\gamma+1)}|M_n - \theta|/\xi$  is approximated by the asymptotic distribution. The optimal size of sample required for attaining the preassigned coverage probability  $1 - \alpha$  is the smallest positive integer  $\geq (l_0\xi/d)^{\gamma+1} =: n^{**}$  (say). Define a stopping rule as

$$\tau_3 := \inf \left\{ n \geq n_0 \mid \frac{R_n}{n^{1/(\gamma+1)}} \leq \frac{2ad}{l_0} \right\},$$

where  $n_0 (\geq 2)$  is an initial size of samples. Then the next theorem follows.

**Theorem 2.** *For the sequential estimation procedure  $(\tau_3, [M_{\tau_3} - d, M_{\tau_3} + d])$ , the following hold.*

(i)  $\lim_{d \rightarrow 0+} P\{|M_{\tau_3} - \theta| \leq d\} = 1 - \alpha$  (asymptotic consistency).

(ii)  $\tau_3/n^{**} \xrightarrow{\text{a.s.}} 1$  ( $d \rightarrow 0+$ ).

(iii)  $E(\tau_3)/n^{**} \rightarrow 1$  ( $d \rightarrow 0+$ ) (asymptotic efficiency).

*Proof.* The proof for (i) is similar to the one of Theorem 1 (i). (ii) follows from  $(\tau_3/n^{**})^{1/(\gamma+1)} \xrightarrow{\text{a.s.}} 1$  as  $d \rightarrow 0+$ .

(iii) From (ii), by Fatou’s lemma,

$$\liminf_{d \rightarrow 0+} \frac{E(\tau_3)}{n^{**}} \geq E \left( \liminf_{d \rightarrow 0+} \frac{\tau_3}{n^{**}} \right) = 1. \quad (3.4)$$

On the other hand, since  $0 \leq R_n \leq 2a\xi$  with probability 1 for any  $n \in \mathbb{N}$ , we have  $0 \leq (R_n l_0 / (2ad))^{\gamma+1} \leq (2a\xi l_0 / (2ad))^{\gamma+1} = (l_0\xi/d)^{\gamma+1}$  with probability 1 for any  $n \in \mathbb{N}$ . So,  $0 \leq (R_n l_0 / (2ad))^{\gamma+1} \leq n$  with probability 1 for  $n$  satisfying  $n \geq (l_0\xi/d)^{\gamma+1} + 1$ . Therefore, since  $\tau_3 = \inf \{n \geq n_0 \mid (R_n l_0 / (2ad))^{\gamma+1} \leq n\}$ , we have  $\tau_3 \leq (l_0\xi/d)^{\gamma+1} + 1$ . Then, using the definition of  $n^{**}$ , we have

$$\frac{E(\tau_3)}{n^{**}} \leq \left\{ \left( \frac{l_0\xi}{d} \right)^{\gamma+1} + 1 \right\} \left( \frac{l_0\xi}{d} \right)^{-(\gamma+1)} = 1 + \left( \frac{d}{l_0\xi} \right)^{\gamma+1},$$



hence

$$\limsup_{d \rightarrow 0+} \frac{E(\tau_3)}{n^{**}} \leq 1. \quad (3.5)$$

Combining (3.4) and (3.5), we obtain (iii).  $\square$

From Theorem 2 and Theorem of Chow and Robbins (1965),  $\tau_3 \approx (l_0 \xi / d)^{\gamma+1}$  and  $\tau_{CR} \approx u_{\alpha/2}^2 \sigma^2 / d^2$  as  $d \rightarrow 0+$ . Therefore,

$$\tau_3 / \tau_{CR} \begin{cases} = o(1) & (0 < \gamma < 1), \\ = O(1) & (\gamma = 1), \\ \rightarrow \infty & (\gamma > 1) \end{cases}$$

as  $d \rightarrow 0+$ . Therefore,  $\tau_3$  is asymptotically better than  $\tau_{CR}$  in the sense of the average size of sample if  $0 < \gamma < 1$ .

In this paper, we considered the cases when the values at the endpoints of the support of the p.d.f. are positive simultaneously, or tend to 0 at the same speed. In the meantime, if the either value at the endpoints of the support of the p.d.f. is positive, or tend to 0 at a different speed, then the coefficients of  $n^\gamma(X_{(1)} - a - \theta)$  and  $n^\delta(X_{(n)} - b - \theta)$  converging to nontrivial random variables are different and estimation by using the midrange  $M_n$  is inappropriate.

#### 4. NUMERICAL EXAMPLE

In this section we examine the coverage probability of the procedure  $[M_{\tau_2} - d, M_{\tau_2} + d]$  by simulation based on 100000 repetitions. Suppose that  $X_1, X_2, \dots, X_n, \dots$  is a sequence of i.i.d. random variables with the p.d.f.  $(1/\xi)f_0((x - \theta)/\xi)$ , where  $\theta \in \mathbb{R}$ ,  $\xi > 0$  and  $f_0(\cdot)$  is a trapezoid-shape p.d.f. given by

$$f_0(x) = \begin{cases} (\frac{1}{2} - c)x + \frac{1}{2} & (x \in (-1, 1)), \\ 0 & (\text{otherwise}) \end{cases}$$

with  $0 < c < 1$ . Note that,  $f_0$  is the p.d.f. of the uniform distribution over  $(-1, 1)$  and an asymmetric p.d.f. over  $(-1, 1)$  for  $c = 0.5$  and a sufficiently small  $c > 0$ , respectively. Since  $M_{\tau_2}$  is location equivariant, we may assume  $\theta = 0$  without loss of generality.

When  $\alpha = 0.10$ ,  $d = 0.01(0.01)0.05$ ,  $\xi = 1(1)5$  and  $n_0 = 5$ , Tables 1 and 2 show the values of coverage probabilities of the sequential estimation

procedure  $(\tau_2, [M_{\tau_2} - d, M_{\tau_2} + d])$  for  $c = 0.1$  and  $c = 0.5$ , respectively. The result suggests that the estimation procedure is consistent for this case.

**Table 1.** Coverage probabilities of  $[M_{\tau_2} - d, M_{\tau_2} + d]$  for  $c = 0.1$

$\xi \setminus d$	0.01	0.02	0.03	0.04	0.05
1	0.90637	0.91545	0.92348	0.93092	0.93758
2	0.89830	0.90544	0.90960	0.91424	0.92017
3	0.90123	0.90313	0.90713	0.90832	0.91030
4	0.89926	0.90117	0.90333	0.90615	0.90804
5	0.89817	0.89952	0.90318	0.90421	0.90561

**Table 2.** Coverage probabilities of  $[M_{\tau_2} - d, M_{\tau_2} + d]$  for  $c = 0.5$

$\xi \setminus d$	0.01	0.02	0.03	0.04	0.05
1	0.90210	0.90727	0.91183	0.91328	0.91988
2	0.89929	0.90131	0.90330	0.90628	0.91176
3	0.89849	0.89947	0.90221	0.90235	0.90525
4	0.89729	0.89729	0.89982	0.90169	0.90322
5	0.89785	0.8998	0.89906	0.89862	0.90054

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