

A Remark on Comparison with a Maximum Likelihood Estimator in Asymptotic Variances in a Non-Regular Case*

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Abstract

Let $X_1, X_2, \dots, X_n, \dots$ be a sequence of independent and identically distributed random variables with a truncated exponential density. Then it is shown that there exists an estimator whose asymptotic variance is smaller than that of a maximum likelihood estimator.

1. Introduction

Suppose that $X_1, X_2, \dots, X_n, \dots$ is a sequence of independent and identically distributed random variables with a density $f(x-\theta)$ satisfying

$$f(x) = \begin{cases} ce^{-x} & \text{for } 0 < x < 1; \\ 0 & \text{otherwise,} \end{cases} \quad (1)$$

where $c = 1/(1 - e^{-1})$.

It is easily seen that the maximum likelihood estimator (MLE) $\hat{\theta}_{ML}$ is given by $\hat{\theta}_{ML} = \min_{1 \leq i \leq n} X_i$. Then it is shown that there exists an estimator whose asymptotic variance is smaller than that of $\hat{\theta}_{ML}$.

Although the density of only the form (1) is treated in this paper, it is possible to extend it to the density $f(x)$ such that $f(x)$ is continuously differentiable in the interval (α, β) ,

$$\begin{aligned} f(x) &> 0 && \text{for } \alpha < x < \beta; \\ f(x) &= 0 && \text{otherwise,} \end{aligned}$$

and $0 < \lim_{x \rightarrow \alpha+0} f(x) = \lim_{x \rightarrow \beta-0} f(x) < \infty$.

2. Results

Let \mathcal{X} be an abstract sample space whose generic point is denoted by x , \mathcal{B} a σ -field of subsets of \mathcal{X} and $\{P_\theta : \theta \in \Theta\}$ a set of probability measures on \mathcal{B} , where Θ is called a parameter space. We suppose that $\mathcal{X} = \Theta = R^1$ and \mathcal{B} is a Borel σ -field and for each θ P_θ has the density $f(x-\theta)$ of the form (1). Consider n -fold direct products (R^n, \mathcal{B}^n) of (R^1, \mathcal{B}) and the corresponding product measure P_θ^n of P_θ . An estimator of θ is defined to be a sequence $\{\hat{\theta}_n\}$ of \mathcal{B}^n -measurable functions $\hat{\theta}_n$ on R^n into Θ . For simplicity we denote $\{\hat{\theta}_n\}$ by $\hat{\theta}$. A distribution function $F_{\theta, \hat{\theta}^C}(\cdot)$ is called to be the asymptotic distribution function of an estimator $\hat{\theta}$ of order $C = \{c_n\}$ if for each real number y , $F_{\theta, \hat{\theta}^C}(y)$ is continuous in θ and for

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any $\vartheta \in \Theta$ there exists a positive number d such that for any continuity point y of $F_{\vartheta, \hat{\vartheta}^C(y)}$,

$$\lim_{n \rightarrow \infty} \sup_{\theta: |\theta - \hat{\vartheta}| < d} |P_{\theta^n}(\{c_n(\hat{\theta}_n - \theta) \leq d\}) - F_{\vartheta, \hat{\vartheta}^C(y)}| = 0$$

(Akahira [1]).

Since

$$\begin{aligned} & \lim_{n \rightarrow \infty} P_{\theta^n}(\{n(\hat{\theta}_{ML} - \theta) \leq y\}) \\ &= \lim_{n \rightarrow \infty} P_{\theta^n}(\{\min_{1 \leq i \leq n} x_i \leq \theta + yn^{-1}\}) \\ &= \begin{cases} 1 - e^{-cy} & \text{for } y > 0; \\ 0 & \text{for } y \leq 0, \end{cases} \end{aligned}$$

it follows that the density $f_{\hat{\theta}_{ML}}(y)$ of the asymptotic distribution of $\hat{\theta}_{ML}$ of order $\{n\}$ is given by

$$f_{\hat{\theta}_{ML}}(y) = \begin{cases} ce^{-cy} & \text{for } y > 0; \\ 0 & \text{for } y \leq 0. \end{cases} \quad (2)$$

Let $\hat{\theta}^* = \max_{1 \leq i \leq n} X_i - 1$. Then we have

$$\begin{aligned} & \lim_{n \rightarrow \infty} P_{\theta^n}(\{n(\hat{\theta}^* - \theta) \leq y\}) \\ &= \lim_{n \rightarrow \infty} P_{\theta^n}(\{\max_{1 \leq i \leq n} x_i \leq \theta + 1 + yn^{-1}\}) \\ &= \begin{cases} 1 & \text{for } y \geq 0; \\ e^{ce^{-1}y} & \text{for } y < 0. \end{cases} \end{aligned}$$

Hence the density $g_{\hat{\theta}^*}(y)$ of the asymptotic distribution of $\hat{\theta}^*$ of order $\{n\}$ is given by

$$g_{\hat{\theta}^*}(y) = \begin{cases} 0 & \text{for } y \geq 0; \\ ce^{-1}e^{ce^{-1}y} & \text{for } y < 0. \end{cases} \quad (3)$$

We define an estimator $\hat{\theta}_a$ by $a\hat{\theta}_{ML} + (1-a)\hat{\theta}^*$, where $4/5 \leq a < 1$.

We remark that $(\min_{1 \leq i \leq n} X_i, \max_{1 \leq i \leq n} X_i)$ is asymptotically sufficient (Akahira [2]).

Since $\hat{\theta}_{ML}$ and $\hat{\theta}^*$ are asymptotically independent, the density $h_{\hat{\theta}_a}(y)$ of the asymptotic distribution of $\hat{\theta}_a$ of order $\{n\}$ is a convolution of $f_{\hat{\theta}_{ML}}(y)$ and $g_{\hat{\theta}^*}(y)$. It follows from (2) and (3) that

$$h_{\hat{\theta}_a}(y) = \begin{cases} K_a \exp\left\{\frac{c}{(1-a)e}y\right\} & \text{for } y \leq 0; \\ K_a \exp\left(-\frac{c}{a}y\right) & \text{for } y < 0, \end{cases}$$

where $K_a = c/\{a + (1-a)e\}$.

Next we shall calculate the asymptotic variances $V_a(Y)$ and $V_{ML}(Y)$ of $\hat{\theta}_a$ and $\hat{\theta}_{ML}$, respectively.

Since

$$\begin{aligned} E_a(Y) &= \frac{K_a \{a^2 - e^2(1-a)^2\}}{c^2}; \\ E_a(Y^2) &= \frac{2K_a \{a^3 + e^3(1-a)^3\}}{c^3}, \end{aligned}$$

it follows that

$$V_a(Y) = \frac{a^2 + e^2(1-a)^2}{c^2}.$$

Since $e < 3$ and $4/5 \leq a < 1$, we have

$$V_a(Y) < \frac{a^2 + 9(1-a)^2}{c^2} \leq \frac{1}{c^2}.$$

On the other hand it is easily seen that

$$V_{ML}(Y) = \frac{1}{c^2}.$$

Hence

$$V_a(Y) < V_{ML}(Y).$$

Therefore it is shown that the asymptotic variance of $\hat{\theta}_a$ is smaller than that of $\hat{\theta}_{ML}$.

References

- [1] Akahira, M., "Asymptotic theory for estimation of location in non-regular cases, II : Bounds of asymptotic distributions of consistent estimators," Rep. Stat. Appl. Res., JUSE, 22, 99-115 (1975)
- [2] Akahira, M., "A remark on asymptotic sufficiency of statistics in non-regular cases," Rep. Univ. Electro-Comm. 27-1, 125-128 (1976)