

A Remark on Asymptotic Sufficiency of Statistics in Non-Regular Cases*

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Abstract

Suppose that $X_1, X_2, \dots, X_n, \dots$ is a sequence of independent identically distributed random variables with the density $f(x: \theta)$ with a compact support, where θ is a real valued parameter. We suppose that a strongly $\{c_n\}$ -consistent estimator of θ exists. Then we show that a statistic $(\min_{1 \leq i \leq n} X_i, \max_{1 \leq i \leq n} X_i)$ is asymptotically sufficient in non-regular cases.

1. Introduction

A consistent estimator with order $\{c_n\}$ (or a $\{c_n\}$ -consistent estimator) is defined and discussed in Akahira [1], where the necessary conditions for the existence of such an estimator are established and the bounds of the orders of convergence of consistent estimators are obtained for non-regular cases. Further the asymptotic accuracies of $\{c_n\}$ -consistent estimators are discussed in Akahira [2].

Asymptotic sufficiency has been discussed under regularity conditions by LeCam [4]. In this paper we extend a similar approach to non-regular cases.

Let $X_1, X_2, \dots, X_n, \dots$ be a sequence of independent identically distributed random variables with the density $f(x: \theta)$ with a compact support, where θ is a real valued parameter. We suppose that a strongly $\{c_n\}$ -consistent estimator of θ exists. Then we shall obtain that a statistic $(\min_{1 \leq i \leq n} X_i, \max_{1 \leq i \leq n} X_i)$ is asymptotically sufficient in non-regular cases.

2. Notations and definitions

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 Θ is an open set in a Euclidean 1-space R^1 . Consider n -fold direct products $(\mathcal{X}^{(n)}, \mathcal{B}^{(n)})$ of $(\mathcal{X}, \mathcal{B})$ and the corresponding product measure $P_\theta^{(n)}$ of P_θ . For each $n=1, 2, \dots$, the points of $\mathcal{X}^{(n)}$ will be denoted by $\tilde{x}_n = (x_1, \dots, x_n)$ and the corresponding random variable by \tilde{x}_n . An estimator of θ is defined to be a sequence $\{\hat{\theta}_n\}$ of $\mathcal{B}^{(n)}$ -measurable function $\hat{\theta}_n$ on $\mathcal{X}^{(n)}$ into Θ . For a sequence of positive numbers $\{c_n\}$ (c_n tending to infinity) an estimator $\{\hat{\theta}_n\}$ is called strongly consistent with order $\{c_n\}$ (or strongly $\{c_n\}$ -consistent for short) if for every $\varepsilon > 0$ and for every compact subset K of Θ , there exists a sufficiently large positive number L satisfying the following:

$$\lim_{n \rightarrow \infty} \sup_{\theta \in K} P_\theta^{(n)}(\{c_n | \hat{\theta}_n - \theta | \geq L\}) < \varepsilon.$$

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A weaker definition of a $\{c_n\}$ -consistent estimator than that of the above form has been given in Akahira [1].

We suppose that every $P_\theta(\cdot)$ ($\theta \in \Theta$) is absolutely continuous with respect to a σ -finite measure μ . Then we denote the density $dP_\theta/d\mu$ by $f(\cdot : \theta)$. If the distribution of \tilde{x}_n is the product measure $P_\theta^{(n)}$, then the corresponding density with respect to the product measure $\mu^{(n)}$ will be denoted by $\prod_{i=1}^n f(x_i : \theta)$. A statistic $T_n(\tilde{X}_n)$ is called asymptotically sufficient if there exist a nonnegative function $p_n(\tilde{x}_n : \theta)$, each the product of a function of \tilde{x}_n only by a function of T_n and θ only such that

$$\lim_{n \rightarrow \infty} \sup_{\theta \in K} \int_{\mathcal{X}^{(n)}} \left| \prod_{i=1}^n f(x_i : \theta) - p_n(\tilde{x}_n : \theta) \right| d\mu^{(n)} = 0$$

for any compact subset K of Θ (LeCam [4]).

3. Asymptotically sufficient statistics

Before discussing the asymptotic sufficiency in detail we shall give a definition and a lemma.

Definition. (Generalized from Gnedenko and Kolmogorov [3]) For each $\theta \in \Theta$ the sums

$$Y_n(\theta) = X_1(\theta) + X_2(\theta) + \cdots + X_n(\theta)$$

of positive independent random variables $X_1(\theta), X_2(\theta), \dots, X_n(\theta), \dots$ are said to be uniformly relatively stable for constants $B_n(\theta)$ if there exist positive constants $B_n(\theta)$ such that for any $\varepsilon > 0$

$$P_\theta^{(n)} \left(\left| \frac{Y_n(\theta)}{B_n(\theta)} - 1 \right| > \varepsilon \right) \rightarrow 0$$

as $n \rightarrow \infty$ uniformly in any compact subset of Θ .

In the subsequent lemma we use the notation that for each k and each $\theta \in \Theta$, $F_{\theta k}(x)$ is the distribution function of $X_k(\theta)$.

Lemma. (Gnedenko and Kolmogorov [3]).

For each $\theta \in \Theta$, let $X_1(\theta), X_2(\theta), \dots, X_n(\theta), \dots$ be a sequence of positive independent random variables. The sums

$$Y_n(\theta) = X_1(\theta) + X_2(\theta) + \cdots + X_n(\theta)$$

are uniformly relatively stable for constants $B_n(\theta)$, if there exists a sequence of positive constants $B_1(\theta), B_2(\theta), \dots, B_n(\theta), \dots$ such that for any $\varepsilon > 0$

$$\sum_{k=1}^n \int_{\varepsilon B_n}^{\infty} dF_{\theta k}(x) \rightarrow 0$$

as $n \rightarrow \infty$ uniformly in any compact subset of Θ ,

$$\frac{1}{B_n(\theta)} \sum_{k=1}^n \int_0^{\varepsilon B_n(\theta)} x dF_{\theta k}(x) \rightarrow 1$$

as $n \rightarrow \infty$ uniformly in any compact subset of Θ .

Let $\mathcal{X} = R^1$. Now we suppose that every $P_\theta(\cdot)$ ($\theta \in \Theta$) is absolutely continuous with respect to a Lebesgue measure m . Then we denote the density dP_θ/dm by $f(\cdot : \theta)$ and by $A(\theta) \subset \mathcal{X}$ the set of points in the space of \mathcal{X} for which $f(x : \theta) > 0$ and suppose $f(x : \theta) = f(x - \theta)$. We make the following assumptions (A), (B) and (C).

Assumption (A). $f(x) > 0$ for $a \leq x \leq b$;
 $f(x) = 0$ for $x < a, x > b$,

and $f(a)$ and $f(b)$ are finite.

Assumption (B). $f(x)$ is twice continuously differentiable in the interval (a, b) .

Define

$$\varphi(\theta) = \int_0^\infty w dF(w : \theta),$$

where $F(w : \theta)$ is the distribution function of

$$W(X : \theta) = \chi_{(a, b) \cap A(\theta)}(X) \left| \log \frac{f(x - \theta)}{f(x)} \right|$$

($\chi_{(a, b) \cap A(\theta)}(\cdot)$ denotes the indicator of $(a, b) \cap A(\theta)$).

Let $T_n = (Y, Z)$, where $Y = \min_{1 \leq i \leq n} X_i$ and $Z = \max_{1 \leq i \leq n} X_i$. We suppose that $\{\hat{\theta}_n(T_n)\}$ is a $\{c_n\}$ -

consistent estimator. The existence of the estimator is guaranteed (See Theorem 4.1 of [1]).

Then for any $\delta > 0$ and any compact subset K of Θ there exists a sufficiently large positive number L satisfying the following:

$$\overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in K} P_{\theta^{(n)}}(\{|\hat{\theta}_n(T_n) - \theta| > Lc_n^{-1}\}) < \delta. \quad (3.1)$$

Assumption (C). The following (3.2)~(3.4) hold:

$$\lim_{n \rightarrow \infty} n\varphi(Lc_n^{-1}) = 0 \quad (3.2)$$

$$\lim_{n \rightarrow \infty} \sup_{\theta \in K} \int_{\varepsilon n\varphi(Lc_n^{-1})}^\infty dF(w : \theta) = 0 \quad (3.3)$$

for any $\varepsilon > 0$ and any compact subset K of Θ ;

$$\lim_{n \rightarrow \infty} \sup_{\theta \in K} \frac{1}{\varphi(Lc_n^{-1})} \int_0^{\varepsilon n\varphi(Lc_n^{-1})} w dF(w : \theta) = 1 \quad (3.4)$$

for any $\varepsilon > 0$ and any compact subset K of Θ .

Theorem. Under Assumptions (A), (B) and (C), the statistic T_n , i.e. $(\min_{1 \leq i \leq n} X_i, \max_{1 \leq i \leq n} X_i)$, is asymptotically sufficient.

Proof. Let ε be an arbitrary positive number. We define $h(T_n, \theta)$ and $g(\bar{x}_n, \hat{\theta}_n(T_n))$ as follows:

$$h(T_n, \theta) = \chi_{\theta}(y, z) = \begin{cases} 1, & \text{if } z - b < \theta < y - a; \\ 0, & \text{otherwise} \end{cases} \quad (3.5)$$

$$g(\bar{x}_n, \hat{\theta}_n(T_n)) = \prod_{i=1}^n f(x_i - \hat{\theta}_n(T_n)). \quad (3.6)$$

It follows from (3.3), (3.4) and Lemma that $\sum_{i=1}^n W(X_i : \hat{\theta}_n(T_n) - \theta)$ is uniformly relatively stable for $n\varphi(Lc_n^{-1})$. Hence we have for any compact subset K of Θ

$$\lim_{n \rightarrow \infty} \inf_{\theta \in K} P_{\theta^{(n)}}(A_n(\hat{\theta}_n(T_n) - \theta : \varepsilon)) = 1,$$

where $A_n(\hat{\theta}_n(T_n) - \theta : \varepsilon) = \left\{ \bar{x}_n : \left| \frac{1}{n\varphi(Lc_n^{-1})} \sum_{i=1}^n W(x_i : \hat{\theta}_n(T_n) - \theta) - 1 \right| < \varepsilon \right\}$.

It follows from (3.1), (3.2) and (3.5)~(3.7) that for any compact subset K of Θ

$$\begin{aligned} & \overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in K} \int_{\mathcal{X}^{(n)}} \left| \prod_{i=1}^n f(x_i - \theta) - h(T_n, \theta) g(\bar{x}_n, \hat{\theta}_n(T_n)) \right| \prod_{i=1}^n dx_i \\ & \leq \overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in K} \left(\int_{\{|\hat{\theta}_n(T_n) - \theta| \leq Lc_n^{-1}\} \cap A_n(\hat{\theta}_n(T_n) - \theta : \varepsilon)} + \int_{\{|\hat{\theta}_n(T_n) - \theta| > Lc_n^{-1}\}} + \int_{A_n(\hat{\theta}_n(T_n) - \theta : \varepsilon)^c} \right) \\ & \quad \cdot \left| \prod_{i=1}^n f(x_i - \theta) - h(T_n, \theta) g(\bar{x}_n, \hat{\theta}_n(T_n)) \right| \prod_{i=1}^n dx_i \end{aligned}$$

$$\begin{aligned}
&= \overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in \bar{K}} \int_{\{|\hat{\theta}_n(T_n) - \theta| \leq Lc_n^{-1}\} \cap A_n(\hat{\theta}_n(T_n) - \theta : \varepsilon) \cap \{z - b < \theta < y - a\}} \left| \prod_{i=1}^n f(x_i - \theta) - \prod_{i=1}^n f(x_i - \hat{\theta}_n(T_n)) \right| \prod_{i=1}^n dx_i \\
&\quad + \overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in \bar{K}} 2 \int_{\{|\hat{\theta}_n(T_n) - \theta| > Lc_n^{-1}\}} \prod_{i=1}^n f(x_i - \theta) \prod_{i=1}^n dx_i \\
&\quad + \overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in \bar{K}} \{1 - P_{\theta^{(n)}}(A_n(\hat{\theta}_n) - \theta : \varepsilon)\} \\
&\leq \overline{\lim}_{n \rightarrow \infty} \sup_{\theta \in \bar{K}} \int_{\{|\hat{\theta}_n(T_n) - \theta| \leq Lc_n^{-1}\} \cap A_n(\hat{\theta}_n(T_n) - \theta : \varepsilon) \cap \{z - b < \theta < y - a\}} \left| \prod_{i=1}^n \frac{f(x_i - \hat{\theta}_n(T_n))}{f(x_i - \theta)} - 1 \right| \prod_{i=1}^n f(x_i - \theta) \prod_{i=1}^n dx_i + 2\delta \\
&\leq \overline{\lim}_{n \rightarrow \infty} \int_{\mathcal{X}^{(n)}} \left\{ \exp \sum_{i=1}^n W(x_i : Lc_n^{-1}) \right\} - 1 \left| \prod_{i=1}^n f(x_i) \prod_{i=1}^n dx_i + 2\delta \right. \\
&\leq \overline{\lim}_{n \rightarrow \infty} |\exp(1 + \varepsilon)n\varphi(Lc_n^{-1}) - 1| + 2\delta \\
&= 2\delta
\end{aligned}$$

Letting $\delta \rightarrow 0$, we complete the proof of the theorem.

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