

## A NOTE ON BAHADUR'S TRANSITIVITY

BY EITAN GREENSHTEIN

*Hebrew University and Ben Gurion University*

Let  $X_1, X_2, \dots$  be a sequence of random variables,  $(X_1, \dots, X_n) \sim F_\theta^n$ ,  $\theta \in \Theta$ . In a work by Bahadur it was shown that, for some sequential problems, an inference may be based on a sequence of sufficient and transitive statistics  $S_n = S_n(X_1, \dots, X_n)$  without any loss in statistical performance. A simple criterion for transitivity is given in Theorem 1.

**1. Introduction.** Let  $X_1, \dots, X_m$ ,  $m \leq \infty$ , be a sequence of random variables,  $(X_1, \dots, X_n) \sim F_\theta^n$ ,  $\theta \in \Theta$ . Bahadur [1] has shown that in a typical sequential decision problem it is enough to consider a sequence of sufficient statistics which is transitive under every  $\theta \in \Theta$ . The reason is that the risk function of any sequential procedure  $\Delta$  whose decision at the  $n$ th stage is a function of  $X_1, \dots, X_n$  can be achieved by a procedure  $\Delta'$  whose decision at the  $n$ th stage is a function of  $S_n(X_1, \dots, X_n)$ , if and only if  $S_1, S_2, \dots$  is a sequence of sufficient statistics which is transitive under every  $\theta \in \Theta$ .

**DEFINITION 1.** The sequence  $\{S_n\}$  is transitive if the conditional distribution of  $S_{n+1}$ , conditional upon  $X_1, \dots, X_n$ , is a function of  $X_1, \dots, X_n$  only through  $S_n$ .

A closely related concept is that the sequence  $S_1, S_2, \dots$  is a Markov sequence. When there is a one-to-one map from  $S_1, \dots, S_n$  to  $X_1, \dots, X_n$ , the sequence  $S_1, S_2, \dots$  is Markovian if and only if it is transitive.

An important work dealing with the concept of transitivity is by Ghosh, Hall and Wijsman [3]. Some criteria for transitivity are given there, using invariance considerations.

In Theorem 1 we give a criterion for a sequence of sufficient statistics to be transitive. The same proof can be used to show that such a sequence is a Markov sequence.

**2. Main result.** We assume that  $F_\theta^n \ll F_{\theta_0}^n$  for every  $n$  and for some  $\theta_0 \in \Theta$ , and that  $X_1, \dots, X_n$  is Euclidean for  $n = 1, 2, \dots$ . Hence conditional distributions are well defined.

Denote by  $F_\theta^{n+1}(x_{(n)}, s_{n+1}|s_n)$  the conditional distribution of  $(X_{(n)}, S_{n+1})$  conditional on  $S_n = s_n$ , where  $X_{(n)} = X_1, \dots, X_n$ . Let

$$f_\theta^n(s_n) = \frac{d\bar{F}_\theta^n(s_n)}{d\bar{F}_{\theta_0}^n(s_n)},$$

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where  $\bar{F}_\theta^n(s_n)$  is the distribution of  $S_n$  under  $\theta$ . Finally, we denote by  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n)$  the conditional distribution of  $S_{n+1}$  conditional on  $S_n = s_n$ , and by  $\bar{F}_\theta^n(x_{(n)}, s_n, s_{n+1})$  the joint distribution of  $X_{(n)}$ ,  $S_n$  and  $S_{n+1}$ .

LEMMA 1. Let  $S_n(X_1, \dots, X_n)$ ,  $n = 1, 2, \dots$ , be a sequence of sufficient statistics for  $F_\theta^n(x_1, \dots, x_n)$ ,  $\theta \in \Theta$ . Then the following hold:

- (i)  $dF_\theta^{n+1}(x_{(n)}, s_{n+1}|s_n) = \frac{f_\theta^{n+1}(s_{n+1})}{f_\theta^n(s_n)} dF_{\theta_0}^{n+1}(x_{(n)}, s_{n+1}|s_n) \quad a.e. (\bar{F}_{\theta_0}^n)$ .
- (ii)  $d\hat{F}_\theta^{n+1}(s_{n+1}|s_n) = \frac{f_\theta^{n+1}(s_{n+1})}{f_\theta^n(s_n)} d\hat{F}_{\theta_0}^{n+1}(s_{n+1}|s_n) \quad a.e. (\bar{F}_{\theta_0}^n)$ .

PROOF. (i)  $d\bar{F}_\theta^{n+1}(x_{(n)}, s_n, s_{n+1}) = f_\theta^{n+1}(s_{n+1}) d\bar{F}_{\theta_0}^{n+1}(x_{(n)}, s_n, s_{n+1})$  by sufficiency. Hence,

$$\begin{aligned} dF_\theta^{n+1}(x_{(n)}, s_{n+1}|s_n) &= \frac{f_\theta^{n+1}(s_{n+1}) dF_{\theta_0}^{n+1}(x_{(n)}, s_{n+1}|s_n)}{\int f_\theta^{n+1}(s_{n+1}) dF_{\theta_0}^{n+1}(x_{(n)}, s_{n+1}|s_n)} \\ &= \frac{f_\theta^{n+1}(s_{n+1})}{f_\theta^n(s_n)} dF_{\theta_0}^{n+1}(x_{(n)}, s_{n+1}|s_n). \end{aligned}$$

(ii) Immediate from (i).  $\square$

COROLLARY 1.  $S_{n+1}$  is sufficient for  $\theta$  with respect to the family  $F_\theta^{n+1}(x_{(n)}, s_{n+1}|s_n)$ ,  $\theta \in \Theta$ .

THEOREM 1. Let  $S_n(X_1, \dots, X_n)$ ,  $n = 1, 2, \dots$ , be a sequence of sufficient statistics for  $F_\theta^n(x_1, \dots, x_n)$ ,  $\theta \in \Theta$ . Suppose that, for  $n = 1, 2, \dots$  and for every value of  $S_n$ ,  $S_{n+1}$  is complete with respect to the family  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n)$ ,  $\theta \in \Theta$ . Then  $\{S_n\}$  is transitive under every  $\theta \in \Theta$ .

PROOF. It suffices to show that  $S_{n+1}$  and  $X_{(n)}$  are independent conditional upon  $S_n = s_n$ .

By Corollary 1,  $S_{n+1}$  is sufficient for  $\theta$  with respect to  $F_\theta^{n+1}(x_{(n)}, s_{n+1}|s_n)$ ,  $\theta \in \Theta$ , and by assumption it is complete. By sufficiency of  $S_n$ ,  $X_{(n)}$  is ancillary. Hence by Basu's lemma ([5], page 191),  $S_{n+1}$  and  $X_{(n)}$  are independent conditional upon  $S_n = s_n$ .  $\square$

In the following two propositions, we will consider the completeness assumption.

PROPOSITION 1. Suppose  $\{S_n\}$  is a sequence of complete and sufficient statistics for  $F_\theta^n(x_1, \dots, x_n)$ . Suppose  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n) \ll \bar{F}_{\theta_0}^{n+1}(s_{n+1})$  for every

value of  $s_n$ . Then  $S_{n+1}$  is complete with respect to  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n)$ ,  $\theta \in \Theta$  for every value of  $s_n$ .

PROOF. Let  $\varphi_{s_n}^{n+1}$  be such that

$$d\hat{F}_{\theta_0}^{n+1}(s_{n+1}|s_n) = \varphi_{s_n}^{n+1}(s_{n+1}) d\bar{F}_{\theta_0}^{n+1}(s_{n+1}).$$

Let  $h(s_{n+1})$  be any integrable real-valued function. Then

$$\int h(s_{n+1}) d\hat{F}_\theta^{n+1}(s_{n+1}|s_n) = \int h(s_{n+1}) \frac{f_\theta^{n+1}(s_{n+1})}{f_\theta^n(s_n)} d\hat{F}_{\theta_0}^{n+1}(s_{n+1}|s_n).$$

The equality is shown by using Lemma 1(ii). The last expression equals zero for every  $\theta$  if and only if

$$\int h(s_{n+1}) f_\theta^{n+1}(s_{n+1}) \varphi_{s_n}^{n+1}(s_{n+1}) d\bar{F}_{\theta_0}^{n+1}(s_{n+1}) = 0 \text{ for every } \theta.$$

By the completeness of  $S_{n+1}$  this implies  $h \varphi_{s_n}^{n+1} = 0$  a.e.  $(\bar{F}_{\theta_0}^{n+1})$ , which implies  $h = 0$  a.e.  $(\hat{F}_{\theta_0}^{n+1}(\cdot|s_n))$ , which by using Lemma 1(ii) implies  $h = 0$  a.e.  $(\hat{F}_\theta^{n+1}(\cdot|s_n))$  for all  $\theta$ . The proof now follows.  $\square$

PROPOSITION 2. Let  $S_n(X_1, \dots, X_n)$  be a sequence of sufficient statistics. Suppose that, for every  $n$ ,  $\bar{F}_\theta^n(s_n)$ ,  $\theta \in \Theta \subseteq \mathbb{R}^k$ , is a  $k$ -dimensional exponential family with a canonical parameter  $\theta$  and a canonical observation  $S_n$  (for a definition see [2], page 1). Suppose  $\Theta$  has a nonvoid interior. Then  $S_{n+1}$  is complete with respect to  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n)$ .

PROOF. Let  $\theta_0 = 0$ , w.l.o.g. Then  $d\bar{F}_\theta^n(s_n) = \exp(\theta \cdot s_n - \psi_n(\theta)) d\bar{F}_0^n(s_n)$  and

$$d\hat{F}_\theta^{n+1}(s_{n+1}|s_n) = \frac{\exp(\theta \cdot s_{n+1} - \psi_{n+1}(\theta)) d\hat{F}_0^{n+1}(s_{n+1}|s_n)}{\exp(\theta \cdot s_n - \psi_n(\theta))} \text{ a.e. } (\bar{F}_0^n),$$

by Lemma 1(ii).

Using this presentation we see that  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n)$ ,  $\theta \in \Theta$ , is an exponential family, where  $\Theta$  is a canonical (nonvoid) parameter set and  $S_{n+1}$  is a canonical observation; hence  $S_{n+1}$  is complete.  $\square$

**3. Examples.** Most of the examples in [3] can be derived by applying Theorem 1. We will consider two examples; the first is taken from [3].

EXAMPLE 1. Let  $\{X_i\}$  be i.i.d.  $X_1 \sim N(\mu, \sigma^2)$ . Let  $\theta = \mu/\sigma$ . Let

$$S_n = \bar{X}_n / \sqrt{\Sigma^n (X_i - \bar{X}_n)^2}.$$

Then  $S_n$  is sufficient for  $S_1, \dots, S_n$  when the parameter of interest is  $\theta$  (see

[4], Exercise 9, page 250, and [3]). Presenting  $N(\mu, \sigma^2)$  as an exponential family with  $(\sum^n X_i, \sum^n X_i^2)$  a minimal sufficient statistic for  $(\mu/\sigma^2, -1/(2\sigma^2))$ , we get that  $(\sum^n X_i, \sum^n X_i^2)$  is complete. Hence, any function of  $(\sum^n X_i, \sum^n X_i^2)$  is complete. In particular,

$$S_n\left(\sum^n X_i, \sum^n X_i^2\right) = \bar{X}_n / \sqrt{\sum^n (X_i - \bar{X}_n)^2}.$$

By Proposition 1 and Theorem 1,  $S_n$  is transitive.

In order to verify the dominance assumption of Proposition 1 note that

$$d\hat{F}_\theta^{n+1}(s_{n+1}|s_n) = \int dF_{(\mu, \sigma)}^{n+1}(s_{n+1}|s_n, \bar{x}_n) dG_{(\mu, \sigma)}^n(\bar{x}_n|s_n),$$

where  $F_{(\mu, \sigma)}^{n+1}(s_{n+1}|s_n, \bar{x}_n)$  is the conditional distribution of  $S_{n+1}$ , conditional on  $S_n$  and  $\bar{X}_n$ ; and where  $G_{(\mu, \sigma)}^n(\bar{x}_n|s_n)$  is the conditional distribution of  $\bar{X}_n$  conditional on  $S_n$ . The measure  $F_{(\mu, \sigma)}^{n+1}(s_{n+1}|s_n, \bar{x}_n)$  is dominated by the Lebesgue measure (using some algebra and the fact that  $X_{n+1}$  is normal); hence the same is true for  $\hat{F}_\theta^{n+1}(s_{n+1}|s_n)$ . The conclusion follows because of the equivalency of the Lebesgue measure and  $\bar{F}_{\theta_0}^{n+1}(s_{n+1})$ .

**EXAMPLE 2.** Let  $(X_1, \dots, X_m) \sim N(\theta \cdot \mathbf{1}_m, a\Sigma)$ , where  $\theta \cdot \mathbf{1}'_m = \theta \cdot (1, \dots, 1) = (\theta, \dots, \theta)$ ,  $a$  is a scalar and  $\Sigma$  is an  $m \times m$  covariance matrix. Consider the following cases:

- (i)  $\theta$  unknown,  $a\Sigma$  known;
- (ii)  $a$  unknown,  $\theta$  and  $\Sigma$  known;
- (iii)  $a$  and  $\theta$  unknown,  $\Sigma$  known.

In all three cases, a sequence of sufficient statistics that can be presented as an exponential family exists. Using Proposition 2 and Theorem 1, those sequences are transitive.

We will elaborate on part (i) of the example.

Let  $\Sigma_{(n)}$  be the covariance matrix of  $(X_1, \dots, X_n) = X'_{(n)}$ ,  $n \leq m$ , and  $\theta \cdot \mathbf{1}_{(n)}$  its expectation vector. Then

$$\begin{aligned} dF_\theta^n(x_1, \dots, x_n) &= \exp\left[-\frac{1}{2}(x_{(n)} - \theta \cdot \mathbf{1}_{(n)})' \cdot a^{-1} \cdot \Sigma_{(n)}^{-1}(x_{(n)} - \theta \cdot \mathbf{1}_{(n)})\right] d\mu_n \\ &= \exp\left[\theta \cdot \mathbf{1}'_{(n)} \cdot a^{-1} \cdot \Sigma_{(n)}^{-1} \cdot x_{(n)} - \psi_n(\theta)\right] d\nu_n, \end{aligned}$$

where  $\mu_n$ ,  $\nu_n$  and  $\psi_n$  are implicitly defined.

We see that  $S_n = \mathbf{1}'_{(n)} \cdot a^{-1} \cdot \Sigma_{(n)}^{-1} \cdot X_{(n)}$  is a canonical observation from a one-dimensional exponential family, with a canonical parameter  $\theta$ .

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DEPARTMENT OF INDUSTRIAL ENGINEERING  
AND MANAGEMENT  
BEN GURION UNIVERSITY  
P.O. BOX 653, BEER SHEVA 84120  
ISRAEL