

ON A LOWER BOUND FOR MOMENTS OF POINT ESTIMATORS¹

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We consider the problem of estimating an unknown parameter θ on the basis of independent identically distributed observations with a common density $f(x, \theta)$ and give some lower bounds for the accuracy of estimates of θ expressed in terms of the Hellinger distance

$$\rho(\theta; \theta') = \int_{\mathcal{X}} (f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta'))^2 d\nu.$$

1. Introduction and results. Let X_1, X_2, \dots be a sequence of independent identically distributed random variables (observations) taking their values in a measurable space $(\mathcal{X}, \mathcal{R})$ with common distribution \mathcal{P}_θ . We suppose that \mathcal{P}_θ depends upon an unknown parameter $\theta \in \Theta$ and that Θ is an open subset of R^k . Denote the n -fold Cartesian product space by $(\mathcal{X}^n, \mathcal{R}^n)$ and the n -fold product measure $\mathcal{P}_\theta \times \dots \times \mathcal{P}_\theta$, by \mathcal{P}_θ^n . We write P_θ instead of $\mathcal{P}_\theta^\infty$; $E_\theta(\cdot)$ denotes mathematical expectation relative to P_θ .

Suppose that there exists a measure ν on \mathcal{R} such that all \mathcal{P}_θ are absolutely continuous relative to the measure ν and

$$\frac{d\mathcal{P}_\theta}{d\nu} = f(x; \theta), \quad x \in \mathcal{X}, \theta \in \Theta.$$

Let $\nu^n \equiv \nu \times \dots \times \nu$ and

$$\frac{d\mathcal{P}_\theta^n}{d\nu^n} = f_n(x^n; \theta) = \prod_{j=1}^n f(x_j; \theta), \quad x^n = (x_1, \dots, x_n) \in \mathcal{X}^n.$$

The Hellinger distance

$$\tilde{\rho}(\mathcal{P}_\theta, \mathcal{P}_{\theta'}) = \left(\int_{\mathcal{X}} |f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta')|^2 d\nu \right)^{\frac{1}{2}}$$

between the measures \mathcal{P}_θ and $\mathcal{P}_{\theta'}$, induces the distance

$$\tilde{\rho}(\theta; \theta') = \tilde{\rho}(\mathcal{P}_\theta; \mathcal{P}_{\theta'})$$

between the parametric points θ and θ' . Let $|\theta - \theta'|$ be a distance between θ and θ' in R^k . Assuming certain regularity conditions it is proved in [2] that if for some $0 < \alpha \leq \beta$,

$$(1) \quad K_1(\theta)|\theta - \theta'|^\alpha \geq \rho(\theta; \theta') \geq K_2(\theta)|\theta - \theta'|^\beta, \quad \rho(\theta; \theta') = \tilde{\rho}^2(\theta; \theta'),$$

then there exist estimators t_n of θ such that

$$(2) \quad \lim_{n \rightarrow \infty} n^{\lambda m} E_\theta |t_n - \theta|^m < \infty$$

for all $m > 0$ and $0 < \lambda < 1/\beta$. If $\alpha = \beta$ it is possible to let $\lambda = 1/\beta$.

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In this paper we prove inequalities which are complementary to (2).

THEOREM 1. *Suppose that*

$$\rho(\theta; \theta') \leq K(\theta)|\theta - \theta'|^\alpha, \quad \alpha > 0,$$

and let T_n denote an estimate of θ satisfying

$$\begin{aligned} S_n^{(m)}(\theta; T_n) &= S_n^{(m)}(\theta) = E_\theta |T_n - \theta|^m, \\ \bar{S}_n^{(m)}(\theta; T_n) &= \bar{S}_n^{(m)}(\theta) = \inf_{|u|=1} S_n^{(m)}(\theta + (8K(\theta)n)^{-1/\alpha}u). \end{aligned}$$

Then

$$(3) \quad \liminf_{n \rightarrow \infty} n^{m/\alpha} (S_n^{(m)}(\theta) + \bar{S}_n^{(m)}(\theta)) \geq 2^{-4m-1} (8K(\theta))^{-m/\alpha}.$$

The analogous result also holds in a more general situation of sequential estimation. This estimation procedure is as follows. We are given: 1) a stopping time τ , a random variable with positive integer values such that the events $\{\tau = n\}$ are measurable with respect to the σ -algebra generated by (X_1, \dots, X_n) and $P_\theta\{\tau < \infty\} = 1$ (we shall suppose for the sake of brevity that the σ -algebra of events generated by (X_1, \dots, X_n) coincides with \mathcal{F}^n); 2) a sequence of statistics $T = \{T_n(X_1, \dots, X_n)\}$. As an estimate of the parameter θ we use the random variable $T_\tau(X_1, \dots, X_\tau)$. Following Ju. V. Linnik, we call the pair $d = [T, \tau]$ a sequential estimation plan.

Let $d = [T, \tau]$ be a sequential estimation plan. Define

$$\begin{aligned} S^{(m)}(\theta; d) &= S^{(m)}(\theta) = E_\theta |T_\tau - \theta|^m, \\ \bar{S}^{(m)}(\theta; d) &= \bar{S}^{(m)}(\theta) = \inf_{|u|=1} S^{(m)}(\theta + (200K(\theta)n)^{-1/\alpha}u). \end{aligned}$$

THEOREM 2. *Under the condition of Theorem 1, for all sequential estimation plans $d = [T, \tau]$ with $E_\theta \tau \leq n, \theta \in \Theta$,*

$$(4) \quad \liminf_{n \rightarrow \infty} n^{m/\alpha} (S^{(m)}(\theta) + \bar{S}^{(m)}(\theta)) \geq 2^{-4m-1} (200K(\theta))^{-m/\alpha}.$$

Both Theorems 1 and 2 are almost immediate consequences of the following:

THEOREM 3. *Let $d = [T, \tau]$ be a sequential estimation plan. Then for all $\theta, \theta' \in \Theta$ with $\rho(\theta; \theta') \cdot \max\{E_\theta \tau, E_{\theta'} \tau\} \leq 200^{-1}$ and all $m \geq 1$*

$$(5) \quad S^{(m)}(\theta) + S^{(m)}(\theta') \geq 2^{-4m-1} |\theta - \theta'|^m.$$

If in addition $\tau \equiv n$, then for all $\theta, \theta' \in \Theta$ with $n\rho(\theta; \theta') \leq \frac{1}{8}$ and all $m \geq 1$

$$(6) \quad S^{(m)}(\theta) + S^{(m)}(\theta') \geq 2^{-4m-1} |\theta - \theta'|^m.$$

2. Proof of Theorem 3. If $m > 1$ then

$$\begin{aligned} S^{(m)}(\theta) + S^{(m)}(\theta') &\geq (S^{(1)}(\theta)) + (S^{(1)}(\theta'))^m \\ &\geq 2^{m-1} (S^{(1)}(\theta) + S^{(1)}(\theta'))^m \end{aligned}$$

so we need only prove the theorem for the case $m = 1$.

Let

$$S^{(1)}(\theta) = S(\theta), \quad E_\theta T_\tau = M(\theta), \quad M(\theta) - \theta = b(\theta).$$

Taking into account the measurability of $\{\tau = n\}$ relative to the σ -algebra generated by X_1, \dots, X_n we may consider $\{\tau = n\}$ as a subset of \mathcal{L}^n . Then, by the Schwarz inequality

$$\begin{aligned}
 & |M(\theta) - M(\theta')|^2 \\
 &= |E_\theta(T_\tau - \frac{1}{2}(M(\theta) + M(\theta'))) - E_{\theta'}(T_\tau - \frac{1}{2}(M(\theta) + M(\theta')))|^2 \\
 (7) \quad &= |\sum_{n=1}^\infty \int_{\{\tau=n\}} (T_n - \frac{1}{2}(M(\theta) + M(\theta')))(f_n(x^n; \theta) - f_n(x^n; \theta')) d\nu^n|^2 \\
 &\leq \sum_{n=1}^\infty \int_{\{\tau=n\}} |T_n - \frac{1}{2}(M(\theta) + M(\theta'))|(f_n^{\frac{1}{2}}(x^n; \theta) + f_n^{\frac{1}{2}}(x^n; \theta'))^2 d\nu^n \\
 &\quad \times \sum_{n=1}^\infty \int_{\{\tau=n\}} |T_n - \frac{1}{2}(M(\theta) + M(\theta'))| \\
 &\quad \times (f_n^{\frac{1}{2}}(x^n; \theta) - f_n^{\frac{1}{2}}(x^n; \theta'))^2 d\nu^n.
 \end{aligned}$$

It is easy to see that

$$\begin{aligned}
 & \int_{\{\tau=n\}} |T_n - \frac{1}{2}(M(\theta) + M(\theta'))|(f_n^{\frac{1}{2}}(x^n; \theta) + f_n^{\frac{1}{2}}(x^n; \theta'))^2 d\nu^n \\
 & \leq 2(\int_{\{\tau=n\}} |T_n - \theta| f_n(x^n; \theta) d\nu^n + \int_{\{\tau=n\}} |T_n - \theta'| f_n(x^n; \theta) d\nu^n) \\
 & \quad + 2(|M(\theta) - \theta| \cdot P_\theta\{\tau = n\} + |M(\theta') - \theta'| \cdot P_{\theta'}\{\tau = n\}) \\
 & \quad + |M(\theta) - M(\theta')|(P_\theta\{\tau = n\} + P_{\theta'}\{\tau = n\})
 \end{aligned}$$

and so the first multiplier on the right side of (7) is less than

$$(8) \quad 4(S(\theta) + S(\theta')) + 2|M(\theta) - M(\theta')|.$$

Upon setting $A_n = \{x^n : f_n(x^n; \theta) \geq f_n(x^n; \theta')\}$, we observe that the second multiplier in (7) is less than

$$\begin{aligned}
 & \sum_{n=1}^\infty (\int_{\{\tau=n\}A_n} |T_n - M(\theta)| f_n(x^n; \theta) d\nu^n \\
 & \quad + \int_{\{\tau=n\}\bar{A}_n} |T_n - M(\theta')| f_n(x^n; \theta') d\nu^n) \\
 (9) \quad & + |M(\theta) - M(\theta')| \sum_{n=1}^\infty \int_{\{\tau=n\}} (f_n^{\frac{1}{2}}(x^n; \theta) - f_n^{\frac{1}{2}}(x^n; \theta'))^2 d\nu^n \\
 & \leq 2(S(\theta) + S(\theta')) + 2|M(\theta) - M(\theta')| \\
 & \quad \times \left[1 - E_\theta \left\{ \prod_{j=1}^\tau \left(\frac{f(X_j; \theta')}{f(X_j; \theta)} \right)^{\frac{1}{2}} \right\} \right].
 \end{aligned}$$

To finish the proof we need the following:

LEMMA 1. For any stopping rule τ

$$(10) \quad 0 \leq 1 - E_\theta \prod_{i=1}^\tau \left(\frac{f(X_j; \theta')}{f(X_j; \theta)} \right)^{\frac{1}{2}} \leq 50 \cdot \rho(\theta; \theta') E_\theta \tau.$$

If, in addition $\tau \equiv n$, then

$$(11) \quad 0 \leq 1 - E_\theta \prod_{i=1}^n \left(\frac{f(X_j; \theta')}{f(X_j; \theta)} \right)^{\frac{1}{2}} \leq 2n\rho(\theta; \theta').$$

We postpone the proof of the lemma until the next section. The inequality (10) of Lemma 1 together with (7)—(9) implies that if $50 \cdot \rho(\theta; \theta') \cdot \max\{E_\theta \tau, E_{\theta'} \tau\} \leq \frac{1}{4}$ then

$$\mu^2 \leq (4\sigma + 2\mu)(2\sigma + \frac{1}{4}\mu),$$

where $\mu = |M(\theta) - M(\theta')|^2$, $\sigma = S(\theta) + S(\theta')$, and hence that

$$(12) \quad \sigma \geq \mu/16.$$

Now if $|b(\theta)| + |b(\theta')| \leq \frac{1}{2}|\theta - \theta'|$, then $|M(\theta) - M(\theta')| \geq \frac{1}{2}|\theta - \theta'|$ and (12) implies (5). If $|b(\theta)| + |b(\theta')| \geq \frac{1}{2}|\theta - \theta'|$, then

$$\sigma \geq |b(\theta)| + |b(\theta')| \geq \frac{1}{2}|\theta - \theta'|$$

and again (5) holds. The proof of (6) on the basis of (11) is the same.

3. Proof of Lemma 1. Consider first the simpler case $\tau \equiv n$. We have

$$\begin{aligned} 1 - E_\theta \prod_{i=1}^n \left(\frac{f(X_j; \theta')}{f(X_j; \theta)} \right)^{\frac{1}{2}} &= 1 - \int_{\mathcal{X}} (f(x; \theta)f(x; \theta'))^{\frac{1}{2}} d\nu^n \\ &\leq n(1 - \int_{\mathcal{X}} (f(x; \theta)f(x; \theta'))^{\frac{1}{2}} d\nu) = 2n\rho(\theta; \theta') \end{aligned}$$

and (11) is proved.

To prove (10) we establish a few lemmas.

LEMMA 2. (Wald). *Let τ be a stopping time relative to a sequence of independent identically distributed random variables $\{\xi_j\}$ with $E\xi_j^2 < \infty$. Then*

$$(13) \quad E \sum_{i=1}^{\tau} \xi_j = E\xi_1 \cdot E\tau, \quad \text{Var} (\sum_{i=1}^{\tau} (\xi_j - E\xi_j)) = \text{Var} \xi_1 \cdot E\tau.$$

For the proof see [1], page 350.

LEMMA 3. *Let $\eta \geq 0$ and ξ be random variables. Then*

$$(14) \quad E\eta e^\xi \geq E\eta \cdot \exp \left\{ \frac{E\xi\eta}{E\eta} \right\} \geq E\eta + E\xi\eta.$$

The first part of (14) is a consequence of Jensen's well-known inequality (see [3], page 159); the second part follows from the elementary inequality

$$e^y \geq 1 + y, \quad y \in R'.$$

Let $B = \{x: \frac{2}{3} \geq (f(x; \theta')/f(x; \theta))^{\frac{1}{2}} \geq \frac{2}{3}\}$. Denote by χ_j the indicator of the random event $X_j \in B$. Define random variables Z_j in the following way:

$$\begin{aligned} Z_j &= \frac{1}{2} \ln \frac{f(X_j; \theta')}{f(X_j; \theta)}, & X_j \in B \\ &= 0, & X_j \notin B \end{aligned}$$

LEMMA 4. *The following inequalities hold:*

$$(15) \quad \begin{aligned} P_\theta\{X_j \in B\} &= E_\theta(1 - \chi_j) \leq 9\rho(\theta; \theta'), \\ P_{\theta'}\{X_j \in B\} &= E_{\theta'}(1 - \chi_j) \leq 9\rho(\theta; \theta'). \end{aligned}$$

PROOF. If $x \in B$ then either $f^{\frac{1}{2}}(x; \theta') - f^{\frac{1}{2}}(x; \theta) > \frac{1}{2}f^{\frac{1}{2}}(x; \theta)$ or $f^{\frac{1}{2}}(x; \theta') - f^{\frac{1}{2}}(x; \theta) < -\frac{1}{3}f^{\frac{1}{2}}(x; \theta)$. In both cases

$$P_\theta\{X_j \in B\} = \int_B f(x; \theta) d\nu \leq 9 \int_B (f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta'))^2 d\nu \leq 9\rho(\theta; \theta').$$

The proof of the second part of (15) is the same.

LEMMA 5. *The following inequalities hold:*

$$(16) \quad |E_\theta Z_j| \leq 23\rho(\theta; \theta'), \quad E_\theta Z_j^2 \leq 7\rho(\theta; \theta'), \quad \text{Var}_\theta Z_j \leq 7\rho(\theta; \theta').$$

PROOF. For $x \in B$

$$\frac{1}{2} \ln \frac{f(x; \theta')}{f(x; \theta)} = -(f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta')) \cdot f^{-\frac{1}{2}}(x; \theta) + R,$$

where $|R| \leq 3(f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta'))^2 \cdot f^{-1}(x; \theta)$.

Using (15) we have

$$\begin{aligned} |E_\theta Z_j| &\leq |\int_B (f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta')) f^{\frac{1}{2}}(x; \theta) \, d\nu| \\ &\quad + 3\rho(\theta; \theta') \leq [1 - \int_{\mathcal{X}} (f(x; \theta) f(x; \theta'))^{\frac{1}{2}} \, d\nu] \\ &\quad + 21\rho(\theta; \theta') = 23\rho(\theta; \theta'). \end{aligned}$$

Further, for $x \in B$ $|f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta')| \leq \frac{1}{2} f^{\frac{1}{2}}(x; \theta)$, so that

$$E_\theta Z_j^2 \leq 2 \int_B [(f^{\frac{1}{2}}(x; \theta) - f^{\frac{1}{2}}(x; \theta'))^2 + R^2 \cdot f(x; \theta)] \, d\nu \leq 7\rho(\theta; \theta').$$

We are now ready to prove Lemma 1. We have

$$(17) \quad 1 - E_\theta \prod_i \left(\frac{f(X_j; \theta')}{f(X_j; \theta)} \right)^{\frac{1}{2}} \leq 1 - E_\theta (\prod_i \chi_j \cdot \exp\{\sum_i Z_j\}).$$

By Lemmas 3—5

$$\begin{aligned} &1 - E_\theta (\prod_i \chi_j \cdot \exp\{\sum_i Z_j\}) \\ &\leq 1 - E_\theta \prod_i \chi_j - E_\theta [\prod_i \chi_j \sum_i Z_j] \\ &= E_\theta (1 - \prod_i \chi_j) - E_\theta (\tau \prod_i \chi_j) \cdot E_\theta Z_1 \\ (18) \quad &\quad + E_\theta (1 - \prod_i \chi_j) (\sum_i Z_j - EZ_j) \\ &\leq E_\theta \sum_i (1 - \chi_j) + E_\theta \tau |E_\theta Z_1| + E_\theta^{\frac{1}{2}} (1 - \prod_i \chi_j)^2 \cdot \text{Var}^{\frac{1}{2}} \sum_i Z_j \\ &= E_\theta \tau \cdot E_\theta (1 - \chi_j) + E_\theta \tau |E_\theta Z_1| + E_\theta^{\frac{1}{2}} (1 - \prod_i \chi_j) \cdot E_\theta^{\frac{1}{2}} \tau \cdot \text{Var}^{\frac{1}{2}} Z_1 \\ &\leq 41\rho(\theta; \theta') E_\theta \tau. \end{aligned}$$

Thus Theorem 3 is proved.

4. **Remarks.** 1. It is easy to see that it will be sufficient to suppose that Θ is a subset of a normed space B . Theorems 1—3 are valid in this case if $|\theta|$ is the norm in B .

2. The requirement of absolute continuity of all measures \mathcal{P}_θ relative to some common measure ν is unnecessary. It is sufficient to take

$$f(x; \theta) = \frac{d\mathcal{P}_\theta}{d(\mathcal{P}_\theta + \mathcal{P}_{\theta'})}, \quad f(x; \theta') = \frac{d\mathcal{P}_{\theta'}}{d(\mathcal{P}_\theta + \mathcal{P}_{\theta'})}$$

when points θ, θ' are being considered.

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