# A PROPERTY OF MAXIMUM LIKELIHOOD ESTIMATOR

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SUMMARY. Roughly speaking our object in this note is to prove that under standard reporting conditions, with probability tending to one, the maximum iskelihood estimate lies in 100(1-a)% confidence set  $(0 < \alpha < 1)$  determined by the family of locally most powerful unbiased tests of  $H_0(\theta \to \theta_0)$  vs  $H_1(\theta \neq \theta_0)$ ; a sort of converse is also proved.

# 1. INTRODUCTION

Let  $X_1, X_2, \ldots$  be a sequence of i.i.d r.v.'s with a common d.f.  $F_{\theta}(x), \theta \in \Theta$ ;  $\Theta$  is an open subset of R. Let  $f(x, \theta)$  be the density of  $F_{\theta}(x)$  w.r.t. some dominating measure  $\mu$ .

We assume  $f(x, \theta)$  satisfies the regularity assumptions I to VI of the next section.

Roughly speaking our object in this note is to prove that under these conditions, with probability tending to one, the maximum likelihood estimate (m.l.e) lies in the  $100(1-\alpha)\%$  confidence set  $(0<\alpha<1)$  determined by the family of locally most powerful unbiased tests (LMPU tests) of  $H_0(\theta=\theta_0)$  vs.  $H_1(\theta\neq\theta_0)$ ; a sort of converse is also proved. A more precise statement is presented later.

We now proceed to a precise formulation of our result.

Our assumptions guarantee (see Lehmann, 1959, p. 83) the existence of a LMPU test of  $H_0(\theta = \theta_0)$  vs.  $H_1(\theta \neq \theta_0)$  with critical function

$$\phi_{\theta_0} = \left\{ \begin{array}{ccc} & \text{if } W_{n\theta_0} + Z_{n\theta_0}^2 > K_{1n\theta_0} + K_{3n\theta_0} Z_{n\theta_0} \\ & & \text{if} & \dots & < & \dots \\ & & & \text{arbitrary if} & \dots & = & \dots \end{array} \right.$$

where

$$Z_{n\theta_0} = n^{-1}I^{-1}(\theta_0) \sum_{i=1}^{n} \frac{d}{d\theta} \log f(x_i, \theta_0),$$

$$\Pi_{n\theta_0}^* = n^{-1}I^{-1}(\theta_0) \sum_{i=1}^{n} \frac{d^2}{d\theta^2} \log f(x_i, \theta_0),$$

and

$$K_{1n\theta_a}$$
 and  $K_{2n\theta_a}$  are such that

$$E_{\theta_a}(\phi_{\theta_a})$$
  $\alpha$  and  $E_{\theta_a}(\phi_{\theta_a}Z_{n\theta_a}) : 0$ .

Let  $V_n$  be the randomized confidence set arising from this family of tests i.e. it consists of all  $\theta$  accepted by the test  $\phi_\theta$ . The set  $V_n$  will depend on the randomising device in addition to  $X_1, X_2, ..., X_n$  but will contain

$$\omega_n = \{\theta : W_{n\theta} + Z_{n\theta}^2 < K_{1n\theta} + K_{2n\theta} Z_{n\theta}\}.$$

Similarly

$$V_n \subset \{0: W_{n\theta} + Z_{n\theta}^2 \leqslant K_{1n\theta} + K_{2n\theta} Z_{n\theta}\} = \omega_n' \text{ (say)}.$$

Now we state our result. Let  $\theta_n$  denote the maximum likelihood estimate.

Theorem: Under assumptions I to VI

- (a) For every  $\theta_0 \in \Theta$  and for every  $0 < \alpha < 1$ ,  $P_{\theta_n} \{ \hat{\theta}_n \in V_n \} \to 1$  as  $n \to \infty$ .
- (b) Let  $T_n$  be any other estimate of  $\theta$ ; then for  $\theta_0 \in \Theta$ .  $P_{\theta_0} \{T_n \in V_n\} \to 1$  for every  $0 < \alpha < 1$  if  $\sqrt{n}(\theta_n T_n) \xrightarrow{P_{\theta_0}} 0$ .
- (c) Let  $T_n$  be any consistent estimate of  $\theta$  such that for  $\theta_0 \in \Theta$ ,  $P_{\theta_0}(T_n \in V_n) \to 1$  for every  $0 < \alpha < 1$ . Then  $\sqrt{n}(\theta_n T_n) \stackrel{P_{\theta_0}}{\to} 0$ .

Remark: If instead of the randomized confidence set  $V_n$  one of the nonrandomized confidence sets  $\omega_n$  or  $\omega_n'$  be used, the resultant size of the test will be  $\alpha_n(\theta)$  which will eventually be  $\alpha$  as  $n\to\infty$  for every  $\theta\in\Theta$  (vide proof of Lemma 3). The theorem remains true if  $V_n$  is replaced by  $\omega_n$  or  $\omega_n'$  throughout. This is so because the proof of (a) and (b) uses  $\{\theta_n\in\omega_n\}$  and the proof of (c) uses  $\{\theta_n\in\omega_n'\}$ .

The proof of the theorem is deferred to Section 3. In Section 2 the assumptions are stated and some auxiliary results are proved.

### 2. ASSUMPTIONS AND LEMMAS

Assumption I: For each x,  $f(x, \theta)$  is twice continuously differentiable in  $\theta \in \Theta$ .

Assumption II: Lot

$$I(\theta) = E_{\theta} \left[ -\frac{d^2}{d\theta^2} \log f(x, \theta) \right];$$

then  $0 < I(\theta) < \infty$  for  $\theta \in \Theta$  and  $I(\theta)$  is continuous in  $\theta \in \Theta$ .

Assumption III: For every  $\theta_0 \in \Theta, \exists$  a neighbourhood (nhbd)  $C_{\theta_0}$  of  $\theta_0$  such that

$$\sup_{\theta \text{ is } C_{\theta_0}} E_{\theta} |\frac{d}{d\theta} \log f(X,\theta)|^3 < \infty.$$

Assumption IV: For every  $\theta_0 \in \Theta$ ,  $\exists$  a nhbd  $C_{\theta_0}$  of  $\theta_0$  such that

$$\left| \frac{d^2}{d\theta^2} \log f(x,\theta) \right| \leqslant H(x), \forall \theta \in C_{\theta_0}$$

$$\big| \begin{array}{l} \frac{d^2}{d\theta^2} \, \log f(x,\,\theta) - \frac{d^2}{d\theta^2} \, \, \log f(x,\,\theta') \big| \leqslant \big| \, \theta - \theta' \, \big| \, A(x) \\ \end{array}$$

for  $\forall \theta, \theta' \in C_{\theta_0}$ ; and for some  $\delta > 0$ 

$$\sup_{\theta \in C_{\theta_0}} E_{\theta} H^{2+\delta}(X) < \infty, \quad \sup_{\theta \in C_{\theta_0}} E_{\theta} A(X) < \infty.$$

Assumption V: If  $\phi_n$  is any test function based on n observations then  $E_\theta\phi_n$  is twice continuously differentiable in  $\theta\in\Theta$ ; moreover

$$\frac{d}{d\bar{\theta}} E_{\theta} \phi_n(X_1 \dots X_n) = \int \phi_n(x_1 \dots x_n) \frac{d}{d\bar{\theta}} \prod_{i=1}^n f(x_i, \theta) d\mu(x_1 \dots x_n)$$

$$\frac{d^2}{d\theta^2} \ E_{\theta} \phi_n(X_1 \dots X_n) = \int \phi_n(x_1 \dots x_n) \ \frac{d^2}{d\theta^2} \ \prod_{i=1}^n f(x_i \theta) d\mu(x_1 \dots x_n)$$

for every  $\theta \in \Theta$  and n > 1.

Assumption VI: The maximum likelihood estimate (mle)  $\theta_n$  of  $\theta$  exists and for every  $\theta_0 \in \Theta$  and  $\epsilon > 0$ .

$$P_{\boldsymbol{\theta}_0}\left\{\,\big|\,\boldsymbol{\theta}_n - \boldsymbol{\theta}_0\big| < \varepsilon, \frac{d}{d\theta}\,\log\,\prod_{i=1}^n f(\boldsymbol{x}_i,\,\boldsymbol{\theta}_n) = 0\right\} \to 1 \ \text{as} \ n \to \infty.$$

Remark: VI holds if conditions of Wald (1949) or Bahadur (1971, p. 34) hold.

We quote a lemma of Ghosh, Sinha and Wieand (1980) to be used later.

Lemma 1: Let C be a compact interval and let U(x,t) be a real valued function measurable in x for each  $t \in C$  and continuous in t for each x. Let  $X_1, X_2, \dots$  be a sequence of i.i.d r.v s having a common d.f.  $F_{\theta}$ ,  $\theta \in \Theta$  and H(x) and A(x) be measurable functions such that  $|U(x,t)| \leq |H(x)|$  for  $t \in C$   $|U(x,t) - U(x,t')| \leq ||t - t'| A(x)|$  for  $t, t' \in C$ , and for some  $\delta > 0$ 

$$\sup_{\theta \in \Theta} E_{\theta} H^{2+\delta}(X) < \infty, \quad \sup_{\theta \in \Theta} E_{\theta} A(X) < \infty.$$

Then for any  $\epsilon > 0 \exists n_0$  and  $K(0 < K < \infty)$  such that

$$P_{\theta}\left\{\sup_{t\in C}\left|\ n^{-1}\sum_{1}^{n}U(X_{i},t)-E_{\theta}U(X_{1},t)\right|<\varepsilon\right\}\geqslant1-Kn^{-\delta_{2}},\ \forall\theta\in\Theta$$

and  $\forall n \geqslant n_0$  and some  $\delta_2 > 0$ .

Note: The assumptions II and IV enable us to apply lemma ! to  $W_{n\theta}$ .

We also quote a version of Theorem 3 of Michel (1976) which will be needed in the sequel.

Lomma 2: Let  $X_1, X_2, \ldots$ , be a sequence of i.i.d r.v's having a common d.f.  $F_{\theta}$ ,  $\theta \in \Theta$  such that  $E_{\theta}(X_1) = 0$ ,  $E_{\theta}(X_1^2) = 1$ . If for some  $\delta > 0$ 

$$\sup_{\theta \in \Theta} |E_{\theta}| |X_1|^{2+\delta} < \infty$$

then there exists a constant f such that for  $n \geqslant 1$ ,  $\forall \theta \in \Theta$  and for all  $t \in R$ ,

$$\mid F_{n\theta}(t) - \Phi(t) \mid \leqslant f n^{-\delta \theta} [1 + \mid t \mid^{2+\delta}]^{-1}$$

where  $F_{n\beta}$  is the d.f. of  $n^{-1}\sum_{t}^{n}X_{t}$  under  $F_{\beta}$ .  $\Phi(t)=\int_{0}^{t}\frac{exp(-x^{2}/2)}{\sqrt{2\pi}}dx$  and  $\delta^{*}=\frac{1}{2}\min(\delta,1)$ .

Note: The assumptions II and III enable us to apply Lemma 2 to the d.f. of  $Z_{ns}$ .

Let

$$\begin{split} & X^{(n)} = (X_1, \dots, X_n), \\ & R_{n\theta} = \{X^{(n)} : Z_{n\theta}^2 + W_{n\theta} > K_{1n\theta} + K_{1n\theta} Z_{n\theta}\}, \\ & R_{n\theta} = \{X^{(n)} : Z_{n\theta}^2 + W_{n\theta} = K_{1n\theta} + K_{2n\theta} Z_{n\theta}\}, \\ & \widetilde{R}_{n\theta} = \{X^{(n)} : Z_{n\theta}^2 > K_{1n\theta} + K_{2n\theta} Z_{n\theta} + 1\} \end{split}$$

and  $A\Delta B = (A^c \cap B) \cup (A \cap B^c)$  for any two sets A and B, C being the usual notation for complement.

We fix a  $\theta_0 \in \Theta$  and a bounded open set  $\Omega$  containing  $\theta_0$ ,  $\Omega \subset \Theta$  such that assumptions III and IV hold on the closure of  $\Omega$ .

Lemma 3: Uniformly in  $\theta \in \Omega$ 

$$P_{\theta}\{\tilde{R}_{n\theta}\} \rightarrow \alpha$$
 ... (2.1)

and

$$E_{\theta}[I_{\tilde{R}_{0}\theta}Z_{n\theta}] \to 0. \tag{2.2}$$

Proof: Note

$$\mid E_{\theta}[I_{\widetilde{R}_{n\theta}}Z_{n\theta}] - E_{\theta}[\phi_{\theta}Z_{n\theta}] \mid \; \leqslant \; P_{\theta}^{1}(R_{n\theta}\Delta\widetilde{R}_{n\theta}) + P_{\theta}^{1}(R_{n\theta}^{'})$$

and

$$\mid P_{\theta}(\widetilde{R}_{n\theta}) - E_{\theta}\phi_{\theta} \mid \leqslant P_{\theta}(R_{n\theta}\Delta\widetilde{R}_{n\theta}) + P_{\theta}(R_{n\theta}')$$

Hence (2.1) and (2.2) are proved if we prove uniformly in

$$\theta \in \Omega$$
,  $P_{\theta}(R_{n\theta}\Delta \widetilde{R}_{n\theta}) \to 0$  and  $P_{\theta}(R'_{n\theta}) \to 0$ .

In view of assumptions II and IV it is clear that for every  $\epsilon > 0$ 

$$P_{\theta}(\mid W_{n\theta} + 1 \mid \leq \epsilon) \rightarrow 1 \text{ uniformly in } \theta \in \Omega.$$
 ... (2.3)

Note that,

$$\begin{split} A_{n,\,\epsilon,\,\theta} &\equiv \left\{ X^{(n)}: \ \frac{K_{200}^{\theta}}{4} + K_{1\,n\theta} + 1 - \epsilon \right. \\ \\ &\leqslant \left( Z_{n\theta} - \frac{K_{2\,n\theta}}{2} \right)^2 \leqslant \ \frac{K_{200}^{\theta}}{4} + K_{1\,n\theta} + 1 + \epsilon \right\} \end{split}$$

$$\supset (R_{n\theta} \Delta \widetilde{R}_{n\theta}) \cap \{|\mathfrak{B}_{n\theta}^* + 1| \leqslant \epsilon\}. \tag{2.4}$$

Also,

$$A_{n, \epsilon, \delta} \supset (|W_{n\theta} + 1| \leqslant \epsilon) \cap R'_{n\theta}$$

On the other hand

$$A_{n,\,\varepsilon,\,\theta} \subset \left\{ X^{(n)} : \left( Z_{n\theta} - \frac{K_{2\,n\theta}}{2} \right)^2 \leqslant 2\varepsilon \right\} \text{ if } \frac{K_{2\,n\theta}^2}{4} + K_{1\,n\theta} + 1 \, \leqslant \, \varepsilon \quad \dots \quad (2.5)$$

and

$$A_{n,\,\epsilon,\,\theta} = \left\{ X^{(n)} : (x-\epsilon)^{1} + \frac{K_{2\,n\theta}}{2} \leqslant Z_{n\theta} \leqslant \frac{K_{2\,n\theta}}{2} + (x+\epsilon)^{1} \right\}$$

$$\bigcup \left\{ X^{(n)} : \frac{K_{2n\theta}}{2} - (x+\epsilon)^{\mathsf{i}} \leqslant Z_{n\theta} \leqslant \frac{K_{2n\theta}}{2} - (x-\epsilon)^{\mathsf{i}} \right\}$$

if 
$$x = \frac{K_{2n\theta}^2}{4} + K_{1n\theta} + 1 \geqslant \epsilon$$
.

Now use the Berry-Essen theorem for  $Z_{ng}$  along with assumption III.

Since  $\alpha < 1$ , from Lemma 2 and (2.1) it is clear that  $\exists n_0$  such that

$$\frac{K_{2n\theta}^2}{4} + K_{1n\theta} - 1 \geqslant 0 \forall n \geqslant n_0, \ \forall \theta \in \Omega.$$

Let

$$C_{\text{ind}} = \frac{K_{2n\theta}}{2} + \left(\frac{1}{4}K_{2n\theta}^2 + K_{1n\theta} + 1\right)^{\frac{1}{2}}$$

and

$$C_{2n\theta} = \frac{1}{2} \; K_{2n\theta} - \left( \; \frac{1}{4} \; K_{2n\theta}^2 + K_{1\,n\theta} + 1 \right)^{\frac{1}{2}}$$

for  $n \geqslant n_0$ .

Lemma 4: Let Z be normal with zero mean and unit variance. Then uniformly in  $\theta \in \Omega$ 

$$E[I_{(C_{nne} < Z < C_{nne})}] \rightarrow 1 - \alpha \qquad ... (2.6)$$

and

$$E[I(O_{100} \le Z \le O_{100})Z] \to 0. \tag{2.7}$$

Proof: Note for  $n \geqslant n_0$ ,  $\widetilde{R}_{n\theta} = (Z_{n\theta} \leqslant C_{2n\theta}) \bigcup (Z_{n\theta} \geqslant C_{1n\theta})$  and hence  $P_{\theta}(C_{2n\theta} \leqslant Z_{n\theta} \leqslant C_{1n\theta}) \to 1-\alpha$  uniformly in  $\theta \in \Omega$  by (2.1). The Berry-Essen theorem along with assumption III completes the proof of (2.6).

Let  $F_{n\theta}$  be the d.f. of  $Z_{n\theta}$ . For a d.f. F(z) we have

$$\int\limits_{C_{2n\theta}}^{a_{1}n_{\theta}}z\;dF(z)=C_{1n\theta}F(C_{1n\theta})-C_{2n\theta}F(C_{2n\theta})-\int\limits_{C_{2n\theta}}^{a_{1}n_{\theta}}F(z)dz.$$

Hence we have, with  $\Phi(z) = P(Z \le z)$ ,

$$\begin{vmatrix} c_{1n\theta_{\theta}} & \int_{C_{2n\theta}}^{1} z \, d \, F_{n\theta}(z) - \int_{C_{2n\theta}}^{2} z d\Phi(z) \, dz \\ & \leq |C_{1n\theta_{\theta}}| |F_{n\theta}(C_{1n\theta}) - \Phi(C_{1n\theta})| \\ & + |C_{2n\theta_{\theta}}| |F_{n\theta}(C_{2n\theta}) - \Phi(C_{2n\theta_{\theta}})| \\ & + \int_{C_{2n\theta}}^{C_{2n\theta_{\theta}}} |F_{n\theta}(z) - \Phi(z)| \, dz. \qquad ... \quad (2.8)$$

Lemma 2 applied to F at implies,

$$\begin{split} \text{R.H.S. of } (2.8) &\leqslant b \, n^{-1} \{ \, | \, C_{1\,n\varphi} \, | \, (1 + | \, C_{1\,n\varphi} \, | \, ^3)^{-1} + | \, C_{2\,n\varphi} \, | \, (1 + | \, C_{2\,n\varphi} \, | \, ^3)^{-1} \\ &+ \int\limits_{C_{2\,n\varphi}}^{C_{1\,n\varphi}} (1 + | \, t \, | \, ^3)^{-1} dt \} \text{ for somo } b > 0. \end{split}$$

Lemma 5:  $K_{2n\theta} \rightarrow 0$  and  $K_{1n\theta} \rightarrow \xi_{n/2}^2 - 1$  both uniformly in  $\theta \in \Omega$ , where  $\Phi(\xi_{n}) = 1 - \alpha$ .

*Proof*: Note that  $0 < \alpha < 1$  and (2.6) imply existence of  $n_0$ ,  $0 < M < \infty$  and  $\delta > 0$  such that

$$C_{1n\theta} - C_{2n\theta} > \delta, \ \forall \ n \geqslant n_0, \ \forall \ \theta \in \Omega$$
 ... (2.9)

$$\min(|C_{1n0}|, |C_{2n0}|) < M, \forall n \ge n_0, \forall \theta \in \Omega.$$
 (2.10)

Hence  $\exists 0 < M' < \infty$  such that

$$\max\left(e^{-C_{1n\theta/2}^{\dagger}}, e^{-C_{2n\theta/2}^{\dagger}}\right) > M', \forall n > n_0, \forall \theta \in \Omega.$$
 (2.11)

Using (2.7) we get

$$\begin{vmatrix} c_{1n\theta} & z^{\frac{z^2}{2}} \\ c_{2n\theta} & z^{\frac{z^2}{2}} & dz \end{vmatrix} = e^{-C_{1n\theta}^2/2} \left| 1 - e^{\operatorname{t}(C_{1n\theta} + C_{2n\theta})(C_{1n\theta} - C_{2n\theta})} \right|$$

$$= e^{-C_{2n\theta}^2/2} \left| 1 - e^{-\operatorname{t}(C_{1n\theta} + C_{2n\theta})(C_{1n\theta} - C_{2n\theta})} \right|$$

$$\to 0 \text{ uniformly in } \theta \in \Omega. \qquad \dots (2.12)$$

(2.12) along with (2.11) implies

$$K_{2n\theta} = (C_{1n\theta} + C_{2n\theta}) \rightarrow 0$$
 uniformly in  $\theta \in \Omega$ . (2.13)

To prove the second part, note that if for every  $n \ni M > n$  and  $\theta' \in \Omega$  such that  $C_{2n\theta'}$  and  $C_{1n\theta'}$  are on the same side of zero then we get a contradiction to (2.7) using (2.9) and (2.10). Hence  $\ni n_0$  such that

$$C_{nna} \leq 0 \leq C_{nna}, \forall n \geq n_0, \forall \theta \in \Omega$$

Now

$$2\int_{0}^{c_{1R\theta}} d\Phi(z) = \left(\int_{0}^{c_{1R\theta}} d\Phi(z) + \int_{0}^{-c_{2R\theta}} d\Phi(z)\right)$$

$$+ \left(\int_{0}^{c_{1R\theta}} d\Phi(z) - \int_{0}^{-c_{2R\theta}} d\Phi(z)\right).$$

$$\rightarrow 1 - \alpha \text{ uniformly in } \theta \in \Omega.$$

because of (2.6) and (2.13); hence in view of (2.13) we have

$$K_{1n\theta} \rightarrow \frac{\xi_{\theta}^2}{2} - 1$$
 uniformly in  $\theta \in \Omega$ .

# 3. PROOF OF THE THEOREM

Choose  $\delta>0$  such that  $\{|\theta_0-\theta|<\delta\}\subset\Omega$ . By assumption VI for any  $\eta>0$  3  $n_0$  such that

$$P_{\theta_n}(|\hat{\theta}_n - \theta_0| < \delta, Z_n, \hat{\theta}_n = 0) > 1 - \eta, n > n_0.$$
 (3.1)

Choose  $\epsilon > 0$  such that  $\xi_{n/2}^2 > 2\epsilon$  and use Lemma 5 to get  $n_0$  such that

$$K_{1n\theta} > \xi_{o/2}^2 - 1 - \epsilon > -1 + \epsilon, n \ge n_0, \forall \theta \in \Omega.$$
 ... (3.2)

Using Lemma 1 we get no such that

$$P_{\theta_0}\{\sup_{n \in \Omega} W_{n\theta} \leqslant -1 + \epsilon\} \geqslant 1 - \eta, \forall n \geqslant n_0. \quad \dots \quad (3.3)$$

Note that

$$\hat{\theta}_n \in \omega_n \text{ iff } W_n \hat{\theta}_- < K_{1n} \hat{\theta}_n.$$
 (3.4)

Combining (3.1), (3.2), (3.3) and (3.4) we get part (a) of the theorem.

Under hypothesis of part (b) or (c) we have

$$P_{\theta_n}\{T_n \in \Omega\} \to 1.$$
 (3.5)

This along with Lemmas 2 and 3 implies

$$K_{2nT_n} \xrightarrow{P_{\theta_0}} 0, K_{1nT_n} \xrightarrow{P_{\theta_0}} \zeta_{1s}^2 - 1 \text{ and } W_{nT_n} \xrightarrow{P_{\theta_0}} 1, \dots$$
 (3.6)

so that

$$K_{\mathbf{In}T_n} - W_{\mathbf{n}T_n} \div \frac{1}{4} K_{\mathbf{S}_nT_n}^2 \xrightarrow{P_{\boldsymbol{\theta}_0}} \boldsymbol{\xi}_{\mathbf{I}\alpha}^* > 0, \ 0 < \alpha < 1. \dots (3.7)$$

Now expanding  $Z_{nT_n}$  around  $\theta_n$  and noting  $Z_{n\hat{\theta}_n} = 0$ , we have

$$\begin{split} \{T_n \in \omega_n\} &= \{ |V_{nT_n} + Z_{nT_n}^* < K_{1nT_n} + K_{1nT_n} Z_{nT_n} \} \\ &= \left[ \left\{ \sqrt{n} (T_n - \theta_n) \cdot \frac{1}{n} \cdot \Sigma \cdot \frac{d^2}{d\theta^2} \log f(x_i, \theta_n^*) \right\}^3 I^{-1}(T_n) \right. \\ &- K_{1nT_n} \left\{ \sqrt{n} (T_n - \theta_n) \cdot \frac{1}{n} \cdot \Sigma \cdot \frac{d^2}{d\theta^2} \cdot \log f(x_i, \theta_n^*) \right\} I^{-1}(T_n) \\ & \div |V_{nT_n} - K_{1nT_n} < 0 \right] \text{ where } \theta_n^* \text{ is botween } \theta_n \text{ and } T_n \\ &= \left\{ [\sqrt{n} (T_n - \theta_n) - \frac{1}{2} K_{2nT_n} |V_{n\theta_n^*}^{-1} I^{-1}(\theta_n^*) I^1(T_n)]^2 \right. \\ &< |V_{n\theta_n^*}^{-1} I^{-2}(\theta_n^*) I(T_n) \left[ K_{1nT_n} - W_{nT_n} + \frac{K_{1nT_n}^2}{4} \right] \right\}. \end{split}$$

Observe that by (3.5) and Lemma 1.

$$W_{n\hat{a}_{n}^{*}} I^{-2}(\hat{\theta}_{n}^{*})I(T_{n})$$
 ... (3.8)

is bounded away from zero and infinity with probability tending to one.

Let

$$\begin{split} & \Sigma_{n} = \left\{ \mathcal{W}_{n\hat{\theta}_{n}^{*}}^{-1} I^{-1}(\hat{\theta}_{n}^{*}) I^{i}(T_{n}) \left[ \frac{K_{1nT_{n}}}{2} - \left( K_{1nT_{n}} - i V_{nT_{n}} + \frac{K_{\frac{n}{4}nT_{n}}}{4} \right)^{\frac{1}{4}} \right] \\ & < \sqrt{n} (T_{n} - \hat{\theta}_{n}) < W_{n\hat{\theta}_{n}^{*}}^{-1} I^{-1}(\hat{\theta}_{n}^{*}) I^{i}(T_{n}) \left[ \frac{K_{2nT_{n}}}{2} + \left( K_{1nT_{n}} - i V_{nT_{n}} + \frac{K_{\frac{n}{2}nT_{n}}}{4} \right)^{\frac{1}{4}} \right] \right\} \text{ and } K_{1nT_{n}} - i V_{nT_{n}} + \frac{K_{\frac{n}{2}nT_{n}}}{4} > 0. \end{split}$$

$$(3.5)$$

Let  $\Sigma'_n$  denote the set with strict inequalities replaced by "  $\leq$ " in (3.9).

If  $P_{\theta_0}\left\{T_n \in \omega_n^n\right\} \to 1$  for each  $0 < \alpha < 1$  then, it is clear from (3.7) that  $P_{\theta_0}\left\{\Sigma_n^i\right\} \to 1$  for each  $0 < \alpha < 1$ . This in view of (3.6), (3.7), (3.8) and the fact  $\xi_{nt} \to 0$  as  $\alpha \to 1$  gives us part (e) of the theorem.

Finally, if  $\sqrt{n}(\hat{\theta}_n - T_n) \stackrel{P_{\theta_0}}{\longrightarrow} 0$ , clearly  $P_{\theta_0} (\Sigma_n) \to 1$  for every  $0 < \alpha < 1$  and hence  $P_{\theta_0} (T_n \varepsilon \omega_n) \to 1$  for every  $0 < \alpha < 1$ .

Since consistency of  $T_n$  was used only to derive (3.5), we have the following

Corollary: Suppose 
$$P_{\theta_0} \{T_n \in C_{\theta_0}\} \to 1$$
 and  $P_{\theta_0} \{T_n \in V_n\} \to 1$ , for all

$$\theta_0 \in \Theta$$
. Then  $\sqrt{n(T_n - \theta_n)} \stackrel{P_{\theta_0}}{\to} 0$ ,  $\forall \theta_0 \in \Theta$ .

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