

THE EQUIVALENCE BETWEEN (MODIFIED) BAYES ESTIMATOR
AND MAXIMUM LIKELIHOOD ESTIMATOR
FOR MARKOV PROCESSES

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1. Introduction

In an earlier paper (Cf. Prakasa Rao [6]), the author has obtained the speed of convergence of Bernstein-von Mises approximation for Markov processes improving over the Bernstein-von Mises theorem for Markov processes proved in Borwanker, Kallianpur and Prakasa Rao [1] and generalizing the results of Strasser [7] and Hipp and Michel [4] in the independent identically distributed case.

Suppose $\{\theta_n\}$ is a sequence of maximum likelihood estimators and $\{\beta_n\}$ a sequence of Bayes estimators for a sufficiently regular prior distribution and smooth loss function. It was shown in Borwanker, Kallianpur and Prakasa Rao [1] that $n^{1/2}(\theta_n - \beta_n) \rightarrow 0$ a.s. under some regularity conditions. We shall now prove that $n(\beta_n - \theta_n)$ is bounded in probability over compact subsets of the parameter space. This result generalizes recent results of Strasser [8] to the Markov case. The present result implies that the (i) Berry-Esseen type bound for $\{\theta_n\}$, proved in Prakasa Rao [5], holds for $\{\beta_n\}$, (ii) $\{\beta_n\}$ is approximately median unbiased of order $n^{-1/2}$ and (iii) $\{\beta_n\}$ is asymptotically efficient of order $n^{-1/2}$. The proof is based on a result concerning rate of convergence of Bernstein-von Mises approximation. The rate obtained is the same as the rate obtained in Prakasa Rao [6] for a less general case and the result is an extension of a theorem of Strasser [8] even in the independent and identically distributed case. Since most of the proofs are similar to those in Strasser [8] and Prakasa Rao [5], we do not give them in detail.

2. Preliminaries

Consider a measurable space $(\mathcal{X}, \mathcal{A})$ and for each $\theta \in H$, let P_θ be a probability measure on $(\mathcal{X}, \mathcal{A})$. Suppose that H is an open interval contained in R . Let \bar{H} denote the closure of H in \bar{R} and \mathcal{B} be the

σ -field of Borel sets of H . Assume that, for every $\theta \in H$, $\{X_n, n \geq 1\}$ is a Markov process taking values in the space $(\mathcal{X}, \mathcal{A}, P_\theta)$ with stationary transition measures $p_\theta(\xi, A) = P_\theta(X_{n+1} \in A | X_n = \xi)$. We assume that for each $\theta \in H$, $p_\theta(\xi, A)$ is a measurable function of ξ for fixed A and a probability measure on \mathcal{A} for fixed ξ . Such a set of transition measures along with an initial probability measure give rise to a Markov process by Doob [2].

A family of $\mathcal{A} \times \mathcal{A}$ -measurable functions $f_\theta: \mathcal{X} \times \mathcal{X} \rightarrow \bar{R}$, $\theta \in \bar{H}$ is said to be a family of *contrast functions* for $\{P_\theta, \theta \in H\}$ if $E_\theta(f_\tau)$ exists for all $\theta \in H$ and $\tau \in \bar{H}$ and if

$$E_\theta(f_\theta) < E_\theta(f_\tau)$$

for all $\theta \in H$, $\tau \in \bar{H}$, $\theta \neq \tau$. Let (x_1, \dots, x_{n+1}) be an observation on the process. Any \mathcal{A}^{n+1} -measurable function $\theta_n: \mathcal{X}^{n+1} \rightarrow \bar{H}$ depending only on x_1, \dots, x_{n+1} is called an estimator. A *minimum contrast estimator* (MCE) is an estimator $\theta_n(\mathcal{X}^{n+1}) \subset \bar{H}$ and

$$\sum_{i=1}^n f_{\theta_n}(x_i, x_{i+1}) = \inf_{\theta \in \bar{H}} \sum_{i=1}^n f_\theta(x_i, x_{i+1}).$$

MCE's for Markov processes were studied by Prakasa Rao [5] and Gänsler [3]. Prakasa Rao [5] studied the rate of convergence of distributions of these estimators by obtaining Berry-Esseen type bound and Gänsler [3] studied measurability, consistency, and asymptotic normality of these estimators.

Unless otherwise stated, we shall assume that the process $\{X_n, n \geq 1\}$ satisfies Doeblin's condition (D_θ) , as given in Doob [2], for every $\theta \in H$. This implies in particular that there exist positive constants $r_\theta \geq 1$ and $\rho_\theta < 1$ and a stationary probability distribution $p_\theta(\cdot)$ such that, for every $\theta \in H$,

$$|p_\theta^{(n)}(\xi, E) - p_\theta(E)| \leq r_\theta \rho_\theta^n$$

for all measurable sets E , for all $\xi \in \mathcal{X}$ and for every $n \geq 1$. Here $p_\theta^{(n)}(\cdot, \cdot)$ denotes the n -step transition function. We shall suppose that the initial distribution is the stationary distribution of the process under consideration. Then the process $\{X_n, n \geq 1\}$ will be a stationary Markov process for each $\theta \in H$. Let P_θ denote the measure on $(\mathcal{X}^\infty, \mathcal{A}^\infty)$ determined by $p_\theta(\cdot, \cdot)$ and $p_\theta(\cdot)$.

Let λ be a prior distribution on (H, \mathcal{B}) . For those $\mathbf{x} \in \mathcal{X}^{n+1}$ for which it is possible, define the probability measure

$$R_{n, \mathbf{x}}(B) = \frac{\int_B \exp\left(-\sum_{i=1}^n f_\theta(x_i, x_{i+1})\right) \lambda(d\theta)}{\int_H \exp\left(-\sum_{i=1}^n f_\theta(x_i, x_{i+1})\right) \lambda(d\theta)}, \quad B \in \mathcal{B}.$$

Under the regularity conditions stated at the end of the paper, it can be shown that for every compact $K \subset H$

$$\sup_{\theta \in K} P_{\theta}(M_n^c) = O(n^{-1})$$

where

$$M_n = \left\{ \mathbf{x} \in \mathcal{X}^{n+1} : \sup_{\theta \in \bar{H}} \exp \left[- \sum_{i=1}^n f_{\theta}(x_i, x_{i+1}) \right] < \infty \right\} .$$

(Here M_n^c denotes complement of M_n). Clearly $R_{n,\mathbf{x}}(B)$ is well defined for $\mathbf{x} \in M_n$, $n \geq 1$. Let $L(\cdot, \cdot)$ be a loss function satisfying the conditions given in Section 6.

A (modified) Bayes estimator relative to prior λ and loss function $L(\cdot, \cdot)$ is an \mathcal{A}^{n+1} -measurable function $\beta_n : \mathcal{X}^{n+1} \rightarrow \bar{H}$ such that

$$\int_H L(\beta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) = \inf_{\theta \in \bar{H}} \int_H L(\theta, \sigma) R_{n,\mathbf{x}}(d\sigma) .$$

for $\mathbf{x} \in M_n$, $n \geq 1$.

3. The Bernstein-von Mises approximation

For every compact $K \subset H$, let $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ such that

$$(3.1) \quad \sup_{\theta \in K} P_{\theta}(M_{n,\theta}^c) = O(n^{-1/2}) .$$

Lemma 4.1 of Prakasa Rao [5] implies that, we can assume $\theta_n(\mathbf{x}) \in H$ if $\mathbf{x} \in M_{n,\theta}$ where θ_n is the MCE. Let

$$\alpha(\theta) = [E_{\theta}(f_{\theta}^{(2)}(X_1, X_2))]^{-1}, \quad \theta \in H .$$

Note that $f_{\theta}^{(2)}$ is defined by the regularity condition (iv) of Section 6.

For every Borel set $B \in \mathcal{B}$, let

$$B_n(\mathbf{x}, \theta) = \left\{ \sigma \in R : \frac{\sigma - \theta_n(\mathbf{x})}{\alpha(\theta)^{1/2}} \in B \right\}, \quad \mathbf{x} \in M_{n,\theta}$$

and

$$Z_n^{\theta}(\mathbf{x}) = \alpha(\theta) \frac{1}{n} \sum_{i=1}^n f_{\theta_n(\mathbf{x})}^{(2)}(x_i, x_{i+1}) - 1, \quad \mathbf{x} \in M_{n,\theta} .$$

Further, assumption (xi) in Section 6 implies that

$$\int_R |t|^k |J(t)| e^{-t^2/2} dt < \infty, \quad k = 0, 1, 2, 3 .$$

Define the signed measures

$$\Phi_k(B) = \int_B t^k J(t) \frac{1}{(2\pi)^{1/2}} e^{-t^2/2} dt, \quad B \in \mathcal{B}$$

for $k=0, 1, 2$ and 3 . Let

$$T_{n,\mathbf{x}}^\theta(\sigma) = \left\{ \frac{\sigma - \theta_n(\mathbf{x})}{\alpha(\theta)^{1/2}} \right\} n^{1/2}, \quad \sigma \in H,$$

$$W_n^\theta(\mathbf{x}, s) = \{ \sigma \in R : |\sigma - \theta_n(\mathbf{x})| \leq (s\alpha(\theta))^{1/2} (\log n)^{1/2} n^{-1/2} \}$$

and

$$A_n(\mathbf{x}, \sigma) = (2\pi\alpha(\theta))^{-1/2} n^{1/2} \exp \left\{ -\sum_{i=1}^n f_\sigma(x_i, x_{i+1}) + \sum_{i=1}^n f_{\theta_n(\mathbf{x})}(x_i, x_{i+1}) \right\}$$

for $s > 0$, $\mathbf{x} \in M_{n,\theta}$, $n \geq 1$, $\theta \in H$. In the following, c_K denotes a generic constant depending on compact set K not necessarily the same throughout the discussion.

THEOREM 3.1. *Assume that regularity conditions (i)-(xi) stated in Section 6 are satisfied. Then, for every compact $K \subset H$ there exist sets $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ satisfying (3.1) such that $\mathbf{x} \in M_{n,\theta}$ implies*

$$(3.2) \quad \sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap H} J(T_{n,\mathbf{x}}^\theta(\sigma)) R_{n,\mathbf{x}}(d\sigma) - \Phi_0(B) \right. \\ \left. + \frac{1}{2} Z_n^\theta(\mathbf{x}) [\Phi_0(B) + \Phi_2(B)] \right| \leq c_K n^{-1/2}$$

where $c_K > 0$ depends on the compact set K only.

PROOF. Let $K \subset H$ be compact and let $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ satisfy (3.1). Lemma 5.1 implies that we can assume that

$$W_n^\theta(\mathbf{x}, s_K) \subset H \quad \text{for } \mathbf{x} \in M_{n,\theta}.$$

It can now be shown by method similar to that given in Proposition 1 of Strasser [8] that

$$\int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) d\sigma \\ = \int_{B_n} J(t) \exp \left\{ -\frac{t^2}{2} (Z_n^\theta(\mathbf{x}) + \eta_n(\mathbf{x}, \theta, t)) \right\} \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt \\ = \int_{B_n} \exp \left\{ -\frac{t^2}{2} (Z_n^\theta(\mathbf{x}) + \eta_n(\mathbf{x}, \theta, t)) \right\} \Phi_0(dt)$$

where

$$B_n = \{ t \in R : t \in B, t^2 \leq s_K \log n \},$$

and

$$|\eta_n(\mathbf{x}, \theta, t)| \leq c_K |t| n^{-1/2} \quad \text{if } t \in B_n, \mathbf{x} \in M_{n,\theta},$$

where $c_K > 0$ depends on the compact set K only. Applying the inequality

$$|e^{-\alpha} - 1 + \alpha| \leq \alpha^2 e^{|\alpha|}$$

for

$$\alpha = \frac{t^2}{2} (Z_n^\theta(\mathbf{x}) + \eta_n(\mathbf{x}, \theta, t))$$

and using Lemma 5.7, we obtain that

$$\left| \int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) d\sigma - \int_{B_n} \left[1 - \frac{t^2}{2} (Z_n^\theta(\mathbf{x}) + \eta_n(\mathbf{x}, \theta, t)) \right] \Phi_0(dt) \right| \leq c_K n^{-1/2}$$

for $\mathbf{x} \in M_{n,\theta}$ and uniformly in $B \in \mathcal{B}$. Since

$$\sup_{B \in \mathcal{B}} \left| \int_{B_n} \frac{t^2}{2} \eta_n(\mathbf{x}, \theta, t) \Phi_0(dt) \right| \leq c_K n^{-1/2} \int_{\mathcal{R}} |t|^3 |\Phi_0(dt)|$$

for $\mathbf{x} \in M_{n,\theta}$, it follows that

$$\sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) d\sigma - \Phi_0(B_n) + \frac{1}{2} Z_n^\theta(\mathbf{x}) \Phi_2(B_n) \right| \leq c_K n^{-1/2}$$

for $\mathbf{x} \in M_{n,\theta}$. Applying the assumption (xi) of Section 6 we obtain that

$$\sup_B |\Phi_k(B) - \Phi_k(B_n)| \leq c_k n^{-1/2}, \quad k=0, 1, 2$$

and hence

$$(3.3) \quad \sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) d\sigma - \Phi_0(B) + \frac{1}{2} Z_n^\theta(\mathbf{x}) \Phi_2(B) \right| \leq c_K n^{-1/2}$$

for $\mathbf{x} \in M_{n,\theta}$.

According to the uniform cover theorem, for every $\varepsilon > 0$ satisfying $\bar{K}^\varepsilon \subset H$ there exists $e_K > 0$ such that $|\sigma - \sigma'| < e_K$ implies $|p(\sigma) - p(\sigma')| \leq c_K \cdot |\sigma - \sigma'|$ for every $\sigma \in K^\varepsilon$. We can also assume that $W_n^\theta(\mathbf{x}, s_K)$ is contained in an ε -neighbourhood of θ for every $\mathbf{x} \in M_{n,\theta}$ and $\theta \in K$. Then it follows that $\mathbf{x} \in M_{n,\theta}$ implies that

$$\begin{aligned}
 (3.4) \quad & \sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) p(\sigma) d\sigma \right. \\
 & \quad \left. - p(\theta_n(\mathbf{x})) \int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) d\sigma \right| \\
 & \leq c_K p(\theta_n(\mathbf{x})) \int_{W_n^\theta(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^\theta(\sigma))| A_n(\mathbf{x}, \sigma) |\sigma - \theta_n(\mathbf{x})| d\sigma .
 \end{aligned}$$

Using an argument similar to the one given earlier, it can be shown that

$$\begin{aligned}
 (3.5) \quad & \int_{W_n^\theta(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^\theta(\sigma))| A_n(\mathbf{x}, \sigma) |\sigma - \theta_n(\mathbf{x})| d\sigma \\
 & \leq c_K n^{-1/2} \int_{\{t^2 \leq s_K \log n\}} |tJ(t)| e^{-t^2/2(1-\delta_K)} dt
 \end{aligned}$$

for $\mathbf{x} \in M_{n,\theta}$ since we may assume that there exists a constant $0 < \delta_K < 1$ such that

$$Z_n^\theta(\mathbf{x}) - \eta_n(\mathbf{x}, \theta, t) \geq -\delta_K > -\infty \quad \text{if } t^2 \leq s_K \log n$$

and $\mathbf{x} \in M_{n,\theta}$. Under the assumption (xi) of Section 6, it follows that

$$\int_{\mathcal{R}} |tJ(t)| \exp\left(-\frac{1}{2} t^2(1-\delta_K)\right) dt < \infty .$$

Hence the term in the R.H.S. of (3.5) is $O(n^{-1/2})$ uniformly for $\theta \in K$. (3.3), (3.4) and (3.5) together show that, for all $\mathbf{x} \in M_{n,\theta}$,

$$\begin{aligned}
 (3.6) \quad & \sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap W_n^\theta(\mathbf{x}, s_K)} J(T_{n,\mathbf{x}}^\theta(\sigma)) A_n(\mathbf{x}, \sigma) p(\sigma) d\sigma \right. \\
 & \quad \left. - p(\theta_n(\mathbf{x})) \left[\Phi_0(B) - \frac{1}{2} Z_n^\theta(\mathbf{x}) \Phi_2(B) \right] \right| \leq c_K n^{-1/2}
 \end{aligned}$$

since one can assume that

$$p(\theta_n(\mathbf{x})) \leq c_K < \infty \quad \text{for } \mathbf{x} \in M_{n,\theta}$$

by conditions assumed and Lemma 4.1 of Prakasa Rao [5]. In view of (3.6), arguments similar to those in Strasser [8] show that

$$\begin{aligned}
 (3.7) \quad & \sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap H} J(T_{n,\mathbf{x}}^\theta(\sigma)) R_{n,\mathbf{x}}(d\sigma) \right. \\
 & \quad \left. - \Phi_0(B) + \frac{1}{2} Z_n^\theta(\mathbf{x}) [\Phi_0(B) + \Phi_2(B)] \right| \leq c_K n^{-1/2}
 \end{aligned}$$

as was to be proved.

Remarks. Observe that Theorem 3.1 is not only a generalization of Theorem 1 of Strasser [8] to the Markov case but it is also an im-

provement even in the i.i.d. case over Theorem 1 of Strasser [8] in that we have considered a general function $J(t)$ instead of $|t|^k$. This result generalizes Theorem 3.1 of Borwanker, Kallianpur and Prakasa Rao [1]. Clearly assumption (xi) of Section 6 is satisfied when $J(t) \equiv |t|^k$ and we obtain Strasser's theorem in the Markov case as a special result. This theorem is also a generalization of result in Prakasa Rao [6].

4. Bounds for the difference between MCE and Bayes estimator

We shall prove the following theorem giving a bound for the difference between MCE and Bayes estimator. The proof is only sketched as it is similar to Theorem 2 of Strasser [8].

THEOREM 4.1. *Assume that regularity conditions (i)–(xvi) are satisfied and that λ possesses second absolute moments on H . Then for every compact $K \subset H$ there exist sets $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ satisfying (3.1) such that $\mathbf{x} \in M_{n,\theta}$ implies*

$$(4.1) \quad |\beta_n(\mathbf{x}) - \theta_n(\mathbf{x})| \leq C_K n^{-1}$$

where $C_K > 0$ is a constant depending only on K .

PROOF. In view of the assumptions, it can be shown that there exist sets $M_{n,\theta}$ satisfying (3.1) such that $\mathbf{x} \in M_{n,\theta}$ implies that

$$(4.2) \quad \int_H L_{10}(\beta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) = 0$$

and

$$(4.3) \quad \left| \int_H L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) - \int_{W_n^g(\mathbf{x}, s_K)} L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) \right| \leq C_K n^{-1}$$

for some constant $C_K > 0$ depending on the compact K by Lemma 5.5. Define $T_{n,\mathbf{x}}^g(\sigma)$ as before for $\sigma \in W_n^g(\mathbf{x}, s_K)$ and let $F_{n,\mathbf{x}}^g$ be the probability measure on $\{t \in R: t^2 \leq s_K \log n\}$ induced by $R_{n,\mathbf{x}}$ and $T_{n,\mathbf{x}}^g$. Using the fact

$$L_{10}(\theta_n(\mathbf{x}), \theta_n(\mathbf{x})) = 0$$

and Lemma 5.6, it can be now shown that

$$(4.4) \quad \left| \int_{W_n^g(\mathbf{x}, s_K)} L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) - \alpha(\theta)^{1/2} L_{11}(\theta_n(\mathbf{x}), \theta_n(\mathbf{x})) n^{-1/2} \int_{\{t^2 \leq s_K \log n\}} t F_{n,\mathbf{x}}^g(dt) \right| \leq C_K n^{-1} \int_{\{t^2 \leq s_K \log n\}} t^2 F_{n,\mathbf{x}}^g(dt)$$

for all $\mathbf{x} \in M_{n,\theta}$ by using Taylor's expansion and arguments given in Proposition 2 of Strasser [8]. Theorem 3.1 implies that

$$(4.5) \quad \left| \int_{\{t^2 \leq s_K \log n\}} t F_{n,\mathbf{x}}^\theta(dt) \right| \leq C_K n^{-1}$$

and

$$(4.6) \quad \left| \int_{\{t^2 \leq s_K \log n\}} t^2 F_{n,\mathbf{x}}^\theta(dt) \right| \leq C_K n^{-1/2}$$

for all $\mathbf{x} \in M_{n,\theta}$ by taking $J(t) = t$ and $J(t) = t^2$. (4.3)–(4.6) show that

$$(4.7) \quad \left| \int_H L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) \right| \leq C_K n^{-1}$$

for all $\mathbf{x} \in M_{n,\theta}$. We can assume that $\beta_n(\mathbf{x}) \in H$ if $\mathbf{x} \in M_{n,\theta}$ and in this case, relation (4.2) implies that

$$(4.8) \quad 0 = \int_H L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) + (\beta_n(\mathbf{x}) - \theta_n(\mathbf{x})) \int_H L_{20}(\hat{\beta}_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma)$$

where $|\hat{\beta}_n(\mathbf{x}) - \theta_n(\mathbf{x})| \leq |\beta_n(\mathbf{x}) - \theta_n(\mathbf{x})|$. Arguments in Strasser [7] imply that

$$(4.9) \quad \left| \int_H L_{20}(\hat{\beta}_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) \right| \geq \varepsilon_K$$

for suitable $\varepsilon_K > 0$, for all $\mathbf{x} \in M_{n,\theta}$, $\theta \in K$, $n \geq 1$. (4.7), (4.8) and (4.9) prove the assertion.

Remarks. The result obtained above is an improvement over Theorem 4.1 in Borwanker, Kallianpur and Prakasa Rao [1] where in it was shown that $n^{1/2}(\beta_n(\mathbf{x}) - \theta_n(\mathbf{x})) \rightarrow 0$ a.s. when θ_n is an MLE. Since the conditions assumed imply a Berry-Esseen type bound for the normal approximation of MCE θ_n , as was shown in Prakasa Rao [5], the theorem implies that Berry-Esseen bound holds for normal approximation of Bayes estimate β_n for smooth loss functions and priors. It also follows that β_n is approximate median unbiased of order $n^{-1/2}$ and it is asymptotically efficient of order $n^{-1/2}$.

5. Some lemmas

We assume that regularity conditions stated in Section 6 are satisfied.

LEMMA 5.1. *Let $s > 0$ and $K \subset H$ be compact. Choose $\varepsilon > 0$ such that $\bar{K}' \subset H$ where $K' = \{\sigma \in R: |\sigma - \theta| < \varepsilon, \theta \in K\}$. Then there exist sets $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ such that*

$$(5.0) \quad \sup_{\theta \in K} P_{\theta}(M_{n,\theta}^c) = O(n^{-1/2})$$

and

$$(5.1) \quad W_n^{\theta}(\mathbf{x}, \mathbf{s}) \subset K^* \quad \text{if } \mathbf{x} \in M_{n,\theta}.$$

PROOF. Lemma 4.1 of Prakasa Rao [5] implies that there exist sets $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ such that

$$(5.2) \quad \mathbf{x} \in M_{n,\theta} \Rightarrow |\theta_n - \theta| < \varepsilon/2$$

and

$$\sup_{\theta \in K} P_{\theta}(M_{n,\theta}^c) = O(n^{-1/2}).$$

Relation (5.2) together with assumption (v) and the definition of $W_n^{\theta}(\mathbf{x}, \mathbf{s})$ imply (5.1).

LEMMA 5.2. For every compact $K \subset H$, there exist sets $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ satisfying (5.0) such that $\mathbf{x} \in M_{n,\theta}$ implies that

$$\left| \int_H |J(T_{n,\mathbf{x}}^{\theta}(\sigma))| R_{n,\mathbf{x}}(d\sigma) - \int_{W_n^{\theta}(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^{\theta}(\sigma))| R_{n,\mathbf{x}}(d\sigma) \right| \leq C_K n^{-1/2},$$

for some constant $C_K > 0$ depending on compact K .

PROOF. Lemma 5.1 implies that $\theta_n(\mathbf{x}) \in K^*$ for \mathbf{x} in $M_{n,\theta}$, $\varepsilon > 0$ and $\bar{K}^* \subset H$. By the assumptions made above there exists $\gamma > 0$ such that

$$|J(t)| \leq \gamma |t|^p \quad \text{for sufficiently large } |t| \text{ (say) for } |t| > \delta_0.$$

Note that for all $\theta \in K$

$$\frac{n^{1/2} |\sigma - \theta_n|}{\alpha(\theta)^{1/2}} \geq \frac{n^{1/2} |\sigma - \theta|}{\alpha(\theta)^{1/2}} - \frac{n^{1/2} |\theta_n - \theta|}{\alpha(\theta)^{1/2}} \geq C_K (n^{1/2} \delta - n^{1/2} \varepsilon) = n^{1/2} C_K (\delta - \varepsilon)$$

provided $|\sigma - \theta| \geq \delta$ for sufficiently small ε by Lemma 4.1 of Prakasa Rao [5]. Hence

$$|J(T_{n,\mathbf{x}}^{\theta}(\sigma))| \leq \gamma |T_{n,\mathbf{x}}^{\theta}(\sigma)|^p$$

for large n whenever $|\sigma - \theta| \geq \delta$, $\theta \in K$. Therefore

$$\begin{aligned} & \int_{W_n^{\theta}(\mathbf{x}, s_K)^c} |J(T_{n,\mathbf{x}}^{\theta}(\sigma))| R_{n,\mathbf{x}}(d\sigma) \\ & \leq \gamma \int_{[|\sigma - \theta| \geq \delta] \cap W_n^{\theta}(\mathbf{x}, s_K)^c} |T_{n,\mathbf{x}}^{\theta}(\sigma)|^p R_{n,\mathbf{x}}(d\sigma) \\ & \quad + \int_{[|\sigma - \theta| < \delta] \cap W_n^{\theta}(\mathbf{x}, s_K)^c} |J(T_{n,\mathbf{x}}^{\theta}(\sigma))| R_{n,\mathbf{x}}(d\sigma) \\ & \leq \gamma \int_{[|\sigma - \theta| \geq \delta]} |T_{n,\mathbf{x}}^{\theta}(\sigma)|^p R_{n,\mathbf{x}}(d\sigma) \end{aligned}$$

$$+ \int_{W_n^{\theta}(\mathbf{x}, s_K)^c \cap \{|\sigma - \theta| < \delta\}} |J(T_n^{\theta}, \mathbf{x}(\sigma))| R_{n, \mathbf{x}}(d\sigma).$$

The first term is bounded by $C_K n^{p/2} \exp(-\eta_K n)$ for suitable choice of $\delta = \delta_K$, C_K and $\eta_K > 0$. The second term is bounded by

$$(5.3) \quad \int_{W_n^{\theta}(\mathbf{x}, s_K)^c} \sup\{|J(T_n^{\theta}, \mathbf{x}(\sigma))| : |\sigma - \theta| < \delta\} R_{n, \mathbf{x}}(d\sigma).$$

Continuity of J and condition (xi) (e) of Section 6 imply that the integrand is of the order $n^{p/2}$, uniformly in $\theta \in K$ by Lemma 4.1 of Prakasa Rao [5]. Hence the term in (5.3) is bounded by

$$C_K n^{p/2} R_{n, \mathbf{x}}(W_n^{\theta}(\mathbf{x}, s_K)^c).$$

An analogue of Theorem 3 of Strasser [7] completes the proof.

LEMMA 5.3. For every compact $K \subset H$ there exist sets $M_{n, \theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ satisfying (5.0) such that $\mathbf{x} \in M_{n, \theta}$ implies that

$$(5.4) \quad \sup_{B \in \mathcal{B}} \left| \int_{B_n(\mathbf{x}, \theta) \cap H} J(T_n^{\theta}, \mathbf{x}(\sigma)) R_{n, \mathbf{x}}(d\sigma) - \frac{\int_{B_n(\mathbf{x}, \theta) \cap W_n^{\theta}(\mathbf{x}, s_K)} J(T_n^{\theta}, \mathbf{x}(\sigma)) A_n(\mathbf{x}, \sigma) p(\sigma) d\sigma}{\int_{W_n^{\theta}(\mathbf{x}, s_K)} A_n(\mathbf{x}, \sigma) p(\sigma) d\sigma} \right| \leq C_K n^{-1/2}.$$

PROOF. Lemma 5.1 implies that we can assume that $W_n^{\theta}(\mathbf{x}, s_K) \subset H$. Difference on the L.H.S. of (5.4) is bounded by

$$(5.5) \quad \left| \int_{W_n^{\theta}(\mathbf{x}, s_K)^c} |J(T_n^{\theta}, \mathbf{x}(\sigma))| R_{n, \mathbf{x}}(d\sigma) + |R_{n, \mathbf{x}}(W_n^{\theta}(\mathbf{x}, s_K)) - 1| \cdot \frac{\int_{W_n^{\theta}(\mathbf{x}, s_K)} |J(T_n^{\theta}, \mathbf{x}(\sigma))| R_{n, \mathbf{x}}(d\sigma)}{R_{n, \mathbf{x}}(W_n^{\theta}(\mathbf{x}, s_K))} \right|.$$

Suppose we prove that

$$(5.6) \quad (i) \quad \int_{W_n^{\theta}(\mathbf{x}, s_K)} |J(T_n^{\theta}, \mathbf{x}(\sigma))| R_{n, \mathbf{x}}(d\sigma) \leq C_K < \infty,$$

$$(5.7) \quad (ii) \quad R_{n, \mathbf{x}}(W_n^{\theta}(\mathbf{x}, s_K)) \geq \varepsilon_K > 0,$$

and

$$(5.8) \quad (iii) \quad |R_{n, \mathbf{x}}(W_n^{\theta}(\mathbf{x}, s_K)) - 1| \leq C_K n^{-1/2}$$

for all $\mathbf{x} \in M_{n, \theta}$, $\theta \in K$. Then (5.4) holds in view of Lemma 5.2 and bound (5.5). (ii) and (iii) can be proved by proving analogues of Theo-

rems 2 and 4 of Strasser [7] or from the results in Prakasa Rao [6]. We shall now prove (i). Consider

$$\begin{aligned}
 (5.9) \quad & \int_{W_n^\theta(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^\theta(\sigma))| R_{n,\mathbf{x}}(d\sigma) \\
 & \leq \int_{W_n^\theta(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^\theta(\sigma))| |R_{n,\mathbf{x}}(d\sigma) - Q_{n,\mathbf{x}}(d\sigma)| \\
 & \quad + \int_{W_n^\theta(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^\theta(\sigma))| Q_{n,\mathbf{x}}(d\sigma)
 \end{aligned}$$

where $Q_{n,\mathbf{x}}$ is the normal distribution with mean $\theta_n(\mathbf{x})$ and variance $n^{-1}\alpha(\theta)$. By an analogue of Theorem 4 of Strasser [7] or by a Theorem of Prakasa Rao [6], it follows that the first term on the R.H.S. of (5.9) is bounded by

$$\begin{aligned}
 C_K \int_{W_n^\theta(\mathbf{x}, s_K)} |J(T_{n,\mathbf{x}}^\theta(\sigma))| d\sigma n^{-1/2}(\log n)^{1/2} \\
 = C_K \left\{ \int_{\{|t| \leq s^{1/2}(\log n)^{1/2}\}} |J(t)| \alpha(\theta)^{1/2} n^{-1/2} dt \right\} n^{-1/2}(\log n)^{1/2} \\
 \leq C_K n^{-1}(\log n)^{1/2} \int_{\{|t| \leq s^{1/2}(\log n)^{1/2}\}} |J(t)| dt \leq C_K
 \end{aligned}$$

uniformly for $\theta \in K$ by assumption (xi) of Section 6. The second term is equal to

$$n^{-1/2}\alpha(\theta)^{1/2} \int_{\{|t| \leq s^{1/2}(\log n)^{1/2}\}} |J(t)| e^{-t^2/2} dt \leq C_K < \infty$$

by assumption (xi) of Section 6. This completes the proof of Lemma 5.3.

LEMMA 5.4. *For every compact $K \subset H$, there exist $\varepsilon > 0$ and $k_K: \mathcal{X} \times \mathcal{X} \rightarrow \bar{R}$ such that*

- (i) $\bar{K} \subset H$
- (ii) $\sup_{\theta \in \bar{K}} E_\theta(k_K^2) < \infty$
- (iii) $|f_{\sigma'}^{(2)}(x_1, x_2) - f_\sigma^{(2)}(x_1, x_2)| \leq k_K(x_1, x_2) |\sigma - \sigma'|$
 for all $x_1, x_2 \in \mathcal{X}$ and $\sigma, \sigma' \in K$ such that $|\sigma - \sigma'| < \varepsilon$.

PROOF. Similar to Lemma 3 of Strasser [8].

LEMMA 5.5. *For every compact $K \subset H$, there exist sets $M_{n,\theta} \in \mathcal{A}^{n+1}$, $n \geq 1$, $\theta \in K$ satisfying (5.1) such that $\mathbf{x} \in M_{n,\theta}$ implies*

$$\left| \int_H L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) - \int_{W_n^\theta(\mathbf{x}, s_K)} L_{10}(\theta_n(\mathbf{x}), \sigma) R_{n,\mathbf{x}}(d\sigma) \right| \leq C_K n^{-1}.$$

LEMMA 5.6 *For every compact $K \subset H$ and every $\varepsilon > 0$ with $\bar{K} \subset H$, there exists $e_K > 0$ such that*

$$|L_{11}(\sigma, \sigma) - L_{11}(\sigma, \sigma')| \leq C_K |\sigma - \sigma'|$$

for all $\sigma, \sigma' \in K^c$ such that $|\sigma - \sigma'| \leq e_K$.

PROOF. Similar to Lemma 3 of Strasser [8].

LEMMA 5.7. Let $\{\theta_n, n \geq 1\}$ be a sequence of MCE. For every compact $K \subset H$ there exists $a_K, C_K > 0$ such that

$$P_\theta \left\{ \left| n^{-1} \sum_{i=1}^n \alpha(\theta) f_\theta^{(2)}(X_i, X_{i+1}) - 1 \right| \geq a_K n^{-1/2} (\log n)^{1/2} \right\} \leq C_K n^{-1/2}$$

for all $\theta \in K, n \geq 1$.

Proof of this lemma is similar to Lemma 4.2 of Prakasa Rao [5] and it makes use of Lemmas 4.1 and 4.3 and the Berry-Esseen bound given in Lemmas 3.4 and 3.5 of the reference cited. We omit the details.

6. Regularity conditions

We shall now state the regularity conditions.

- (i) $\theta \rightarrow P_\theta$ is continuous in H with respect to the supremum metric on $\{P_\theta: \theta \in H\}$.
- (ii) For each pair x_1, x_2 in \mathcal{X} , $\theta \rightarrow f_\theta(x_1, x_2)$ is continuous in \bar{H} .
- (iii) For every $\theta \in H$, there exists a neighbourhood W_θ of θ such that

$$\sup_{\tau \in W_\theta} E_\tau [\sup_{\sigma \in W_\theta} f_\sigma^2] < \infty .$$

- (iv) For every pair $x_1, x_2 \in \mathcal{X}$, $\theta \rightarrow f_\theta(x_1, x_2)$ is twice differentiable in H and for all $\theta \in H$ and for all $x \in \mathcal{X}$

$$E_\theta [f_\theta^{(1)}(X_1, X_2) | X_1 = x] = 0$$

where $f_\theta^{(i)}$ denote the i th derivative with respect to θ .

- (v) For every compact $K \subset H$
 - (a) $\inf_{\theta \in K} E_\theta [f_\theta^{(1)}(X_1, X_2)]^2 > 0$,
 - (b) $\inf_{\theta \in K} E_\theta [f_\theta^{(2)}(X_1, X_2)] > 0$.
- (vi) For every compact $K \subset H$ there exists $b_K > 0$ such that for all $x \in \mathcal{X}$,
 - (a) $\sup_{\theta \in K} E_\theta (|f_\theta^{(1)}(X_1, X_2)|^3 | X_1 = x) \leq b_K$
 - (b) $\sup_{\theta \in K} E_\theta (|f_\theta^{(2)}(X_1, X_2)|^3 | X_1 = x) \leq b_K$.
- (vii) For every $\theta \in \bar{H}$, there exists a neighbourhood U_θ of θ such that for every neighbourhood U of $\theta, U \subset U_\theta$ and every compact $K \subset H$

$$\sup_{\tau \in K} E_\tau [\inf_{\sigma \in U} f_\sigma]^2 < \infty .$$

(viii) For every $\theta \in H$, there exists an open neighbourhood V_θ of θ and an $\mathcal{A} \times \mathcal{A}$ -measurable function $k_\theta : \mathcal{X} \times \mathcal{X} \rightarrow \bar{R}$ such that

$$\sup_{\tau \in K} E_\tau [k_\theta^2] < \infty$$

for every compact $K \subset H$ and

$$|f_\sigma^{(2)}(x_1, x_2) - f_\tau^{(2)}(x_1, x_2)| \leq |\sigma - \tau| k_\theta(x_1, x_2)$$

for all $\sigma, \tau \in V_\theta, x_1, x_2 \in \mathcal{X}$.

(ix) For every compact $K \subset H$

(a) $\sup_{\theta \in K} r_\theta < \infty,$

(b) $\sup_{\theta \in K} \rho_\theta < 1,$ and

(c) $\sup_{\theta \in K} ((\rho_\theta + \sqrt{2r_\theta}) / (1 + \sqrt{2r_\theta})) < 1.$

(x) λ has a continuous positive density $p(\cdot)$ on H with respect to the Lebesgue measure satisfying the following conditions: for every $\theta \in H$, there exists a neighbourhood W_θ of θ and a constant $C_\theta \geq 0$ such that

$$\left| \frac{p(\sigma)}{p(\sigma')} - 1 \right| \leq C_\theta |\sigma - \sigma'| \quad \text{for all } \sigma, \sigma' \in W_\theta.$$

(xi) Let $J(\cdot)$ be a real valued continuous function such that

$$\limsup_{|t| \rightarrow \infty} \frac{|J(t)|}{|t|^p} < \infty$$

for some $p > 0$.

Remark. Observe that assumption (xi) implies that

(a) $\int_{-\infty}^{\infty} |t|^k |J(t)| e^{-t^2/2} dt < \infty, \quad k = 0, 1, 2, 3,$

(b) $\sup_B \left| \int_B t^k J(t) e^{-t^2/2} dt - \int_{B_n} t^k J(t) e^{-t^2/2} dt \right| \leq C n^{-1/2}, \quad k = 0, 1, 2$

for some constant $C > 0$ where B is a Borel set in R and $B_n = \{t \in B : t^2 \leq s \log n\}$ for sufficiently large $s > 0$,

(c) for every $0 < \delta < 1$,

$$\int_{-\infty}^{\infty} |tJ(t)| \exp \left\{ -\frac{1}{2} t^2 (1 - \delta) \right\} dt < \infty,$$

(d) and for any fixed $s > 0$

$$\int_{\{|t| \leq s(\log n)^{1/2}\}} |J(t)| dt = O(n(\log n)^{-1/2}).$$

A $\mathcal{B} \times \mathcal{B}$ -measurable function $L : \bar{H} \times \bar{H} \rightarrow \bar{R}$ is called loss function if

$$L(\theta, \theta) < L(\tau, \theta) \quad \text{for all } \theta \in H, \tau \in \bar{H}, \theta \neq \tau.$$

We assume that the following regularity conditions given by Strasser [8] are satisfied.

- (xii) $L: \bar{H} \times H \rightarrow \bar{R}$ is continuous.
- (xiii) (a) $L(\theta, \sigma)$ is extended λ -integrable in σ for all $\theta \in \bar{H}$ and has finite λ -expectation if $\theta \in H$.
 (b) For every $\theta \in \bar{H}$ there exists a neighbourhood U_θ^1 of θ such that for every neighbourhood $U \subset U_\theta^1$ of θ , $\inf_{\theta \in U} L(\theta, \sigma)$ is λ -integrable.
- (xiv) (a) For every $\sigma \in H$, $L(\theta, \sigma)$ is twice differentiable in H . We denote $L_{10}(\theta, \sigma) = (\partial/\partial\theta)L(\theta, \sigma)$; $L_{20}(\theta, \sigma) = (\partial^2/\partial\theta^2)L(\theta, \sigma)$.
 (b) For every $\theta \in H$, $L_{10}(\theta, \sigma)$ is differentiable in H . We denote $L_{11}(\theta, \sigma) = (\partial/\partial\sigma)L_{10}(\theta, \sigma)$.
- (xv) For every compact $K \subset H$:
 (a) $\sup_{\theta \in K} |L_{10}(\theta, \theta)| < \infty$,
 (b) $\sup_{\theta \in K} |L_{20}(\theta, \theta)| < \infty$,
 (c) $\inf_{\theta \in K} L_{20}(\theta, \theta) > 0$,
 (d) $\sup_{\theta \in K} \int |L_{10}(\theta, \sigma)| \lambda(d\sigma) < \infty$,
 (e) $\sup_{\theta \in K} \int |L_{20}(\theta, \sigma)| \lambda(d\sigma) < \infty$.
- (xvi) (a) For every $\theta \in H$, there exists a neighbourhood \tilde{W}_θ of θ and a constant $\tilde{C}_\theta \geq 0$ such that

$$|L_{20}(\sigma, \sigma') - L_{20}(\sigma, \sigma'')| \leq \tilde{C}_\theta |\sigma' - \sigma''|$$

for all $\sigma, \sigma', \sigma''$ in \tilde{W}_θ .

- (b) For every $\theta \in H$, there exists a neighbourhood $\tilde{\tilde{W}}_\theta$ of θ and a constant $\tilde{\tilde{C}}_\theta \geq 0$ such that

$$|L_{11}(\sigma, \sigma') - L_{11}(\sigma, \sigma'')| \leq \tilde{\tilde{C}}_\theta |\sigma' - \sigma''|$$

for all $\sigma, \sigma', \sigma''$ in $\tilde{\tilde{W}}_\theta$.

- (c) For every $\theta \in H$, there exists a neighbourhood V_θ^1 of θ and a continuous function $k_\theta: H \rightarrow R$ such that

$$|L_{20}(\sigma', \sigma) - L_{20}(\sigma'', \sigma)| \leq k_\theta(\sigma) |\sigma' - \sigma''|$$

for all σ', σ'' in V_θ^1 and all $\sigma \in H$ and

$$\int k_\theta(\sigma) \lambda(d\sigma) < \infty \quad \text{for all } \theta \in H.$$

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