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THE BERRY-ESSEEN BOUND FOR MAXIMUM LIKELIHOOD ESTIMATES OF TRANSLATION PARAMETER OF TRUNCATED DISTRIBUTION

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1. Introduction. Let X_1, \dots, X_n be independent random variables with common density $f(x-\theta)$, $-\infty < x$, $\theta < \infty$, where θ is an unknown translation parameter. We shall consider here the case that $f(x)$ is a uniformly continuous density which vanishes on the interval $(-\infty, 0)$ and is positive on the interval $(0, \infty)$ and particularly

$$f(x) \sim \alpha x \quad \text{as } x \rightarrow +0$$

with $0 < \alpha < \infty$. Let $\hat{\theta}_n = \hat{\theta}_n(X_1, \dots, X_n)$ be a m.l.e. (maximum likelihood estimate) of θ for the sample size n . Woodroffe [1] showed that $(\frac{1}{2}\alpha n \log n)^{1/2} \times (\hat{\theta}_n - \theta)$ has an asymptotic standard normal distribution. The purpose of the present paper is to estimate the speed of convergence of $a_n(\hat{\theta}_n - \theta)$ to the standard normal distribution. Here $2a_n^2 = \alpha n(\log n + \log \log n)$. Similar results for minimum contrast estimates in the regular case were given by Michel and Pfanzagl [2] and Pfanzagl [3]. More precisely, Pfanzagl [3] showed that for every compact K there exists a constant c_K such that for all $\theta \in K$, $n \geq 1$ and $t \in R$

$$\left| P_\theta \left\{ \frac{n^{1/2}(\theta_n^* - \theta)}{\beta(\theta)} < t \right\} - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^t \exp\left(-\frac{r^2}{2}\right) dr \right| \leq c_K n^{-1/2},$$

where θ_n^* denotes a minimum contrast estimate for the sample size n .

2. Conditions and the main result. We shall impose the following regularity conditions on $f(x)$. These conditions are stronger than those made by Woodroffe [1].

CONDITIONS

(i) $f(x)$ is a uniformly continuous density which vanishes on $(-\infty, 0)$ and is positive on $(0, \infty)$.

(ii) $f(x)$ is continuously differentiable on $(0, \infty)$ with derivative $f'(x)$ and $f'(x)$ is absolutely continuous on every compact subinterval of $(0, \infty)$ with de-

rivative $f''(x)$.

(iii) For some α and r , $0 < \alpha$, $r < \infty$

$$f'(x) = \alpha + O(x^r) \text{ and } f''(x) = O(x^{r-1}) \quad \text{as } x \rightarrow +0.$$

Let $g(x) = \log f(x)$ for $x > 0$. Then $g(x)$ will be continuously differentiable on $(0, \infty)$ with derivative $g' = f'/f$ and $g'(x)$ will be absolutely continuous on every compact subinterval of $(0, \infty)$ with derivative $g'' = (ff'' - f'^2)/f^2$.

(iv) For every $t \geq 0$

$$\int_0^\infty \{g(x+t)\}^2 f(x) dx < \infty.$$

(v) For every $a > 0$, there is a $\delta > 0$, for which

$$\int_a^\infty \sup_{|u| \leq \delta} |g'(x+u)|^3 f(x) dx < \infty.$$

(vi) For every $a > 0$, there is a $\delta > 0$, for which

$$\int_a^\infty \sup_{|u| \leq \delta} \{g''(x+u)\}^2 f(x) dx < \infty.$$

REMARK. Under conditions (i) and (ii), condition (iii) is equivalent to the following condition (iii)'.

(iii)' For some α and r , $0 < \alpha$, $r < \infty$

$$f(x) = \alpha x + O(x^{1+r}), g'(x) = x^{-1} + O(x^{r-1}) \text{ and } g''(x) = -x^{-2} + O(x^{r-2}) \\ \text{as } x \rightarrow +0.$$

EXAMPLES ([1]). Let

$$f(x) = r \left[\Gamma\left(\frac{2}{r}\right) \right]^{-1} x \exp(-x^r), \quad x > 0, \text{ for some } r > 0,$$

$$\text{or } f(x) = \frac{1}{d(1+d)} \frac{x}{(1+x)^{2+d}}, \quad x > 0, \text{ for some } d > 0,$$

then conditions (i)–(vi) are all satisfied.

Let $M_n = \min(X_1, \dots, X_n)$ and $G_n(t) = \sum_{i=1}^n g(X_i - t)$ for $t < M_n$. Condition (i) insures that m.l.e.'s exist in the interval $(-\infty, M_n)$. Let $\hat{\theta}_n$, $n \geq 1$, be a sequence of m.l.e.'s. If conditions (i) and (ii) are satisfied, then

$$-\infty < \hat{\theta}_n < M_n \text{ and } G'_n(\hat{\theta}_n) = 0$$

with probability 1.

Theorem. Suppose that conditions (i)–(vi) are all satisfied. Let $\hat{\theta}_n$, $n \geq 1$, denote a sequence of m.l.e.'s for $\prod_{i=1}^n f(X_i - \theta)$ and let $2a_n^2 = \alpha n(\log n + \log \log n)$. Then there exists a constant c_1 such that for all $\theta \in R$, $n \geq 1$ and $t \leq 0$

$$(2.1) \quad |P_\theta \{a_n(\hat{\theta}_n - \theta) \leq t\} - \Phi(t)| \leq c_1(\log n)^{-1}.$$

Also, for every s , $0 < s < 1$, there exists a constant c_2 such that for all $\theta \in R$, $n \geq 1$ and $t > 0$

$$(2.2) \quad |P_\theta \{a_n(\hat{\theta}_n - \theta) \leq t\} - \Phi(t)| \leq c_2(\log n)^{s-1}.$$

Here

$$\Phi(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^t \exp\left(-\frac{x^2}{2}\right) dx.$$

REMARK. (1) The assertion of (2.2) holds with $(\log n)^{-1}$ instead of $(\log n)^{s-1}$ provided t is restricted to a finite interval $(0, M]$.

(2) We used $\{\frac{1}{2}\alpha n(\log n + \log \log n)\}^{1/2}$ as the convergence order of m.l.e. to the true parameter θ . However our result is true for any a_n , $n \geq 1$, satisfying that $\alpha n a_n^{-2} \log a_n = 1 + O((\log n)^{-1})$. Obviously, this condition includes the case that $a_n = \{\frac{1}{2}\alpha n(\log n + \log \log n)\}^{1/2}$ but excludes the case that $a_n = (\frac{1}{2}\alpha n \log n)^{1/2}$.

3. Some lemmas. Since θ is a translation parameter, it will suffice to prove our result in the special case that $\theta = 0$. Hereafter, suppose that $\theta = 0$. The following Lemma 1 refines the result of Woodroffe [1].

Lemma 1. *Let conditions (i)–(iii) and (vi) be satisfied. Then, for sufficiently small $\varepsilon > 0$, there exists $c \geq 0$ such that*

$$(3.1) \quad P\left\{\sup_{-\varepsilon \leq t < M_n} \frac{1}{n} G_n''(t) \geq -1\right\} \leq c n^{-1}$$

for all $n \geq 1$.

Proof. Let $a > 0$ be so small that $g''(x) \leq -\frac{1}{2}x^{-2}$ for $0 < x \leq 2a$. There is a sufficiently small number $0 < \varepsilon < a$ such that

$$\int_0^a (x + \varepsilon)^{-2} f(x) dx > 2 \int_a^\infty \sup_{|t| \leq \varepsilon} |g''(x+t)| f(x) dx + 5$$

because the left-hand integral diverges to ∞ as $\varepsilon \rightarrow 0$. Then the event $M_n \leq \varepsilon$ implies that

$$\sup_{-\varepsilon \leq t < M_n} \frac{1}{n} G_n''(t) \leq \frac{-1}{2n} \sum_u^a (X_i + \varepsilon)^{-2} + \frac{1}{n} \sum_a^\infty \sup_{|t| \leq \varepsilon} |g''(X_i + t)|$$

where \sum_u^a denotes summation over $i \leq n$ for which $u \leq X_i < v$. Hence the relations $M_n \leq \varepsilon$ and $\sup_{-\varepsilon \leq t < M_n} \frac{1}{n} G_n''(t) \geq -1$ imply

$$\left| \frac{1}{n} \sum_a^{\infty} (X_i + \varepsilon)^{-2} - \int_0^a (x + \varepsilon)^{-2} f(x) dx \right| \geq 1$$

or

$$\left| \frac{1}{n} \sum_a^{\infty} \sup_{|t| \leq \varepsilon} |g''(X_i + t)| - \int_a^{\infty} \sup_{|t| \leq \varepsilon} |g''(x + t)| f(x) dx \right| \geq 1.$$

Hence we have

$$\begin{aligned} & P \left\{ \sup_{-\varepsilon \leq t < M_n} \frac{1}{n} G_n''(t) \geq -1 \right\} \\ & \leq P \{M_n > \varepsilon\} + P \left\{ \left| \frac{1}{n} \sum_a^{\infty} (X_i + \varepsilon)^{-2} - \int_0^a (x + \varepsilon)^{-2} f(x) dx \right| \geq 1 \right\} \\ & \quad + P \left\{ \left| \frac{1}{n} \sum_a^{\infty} \sup_{|t| \leq \varepsilon} |g''(X_i + t)| - \int_a^{\infty} \sup_{|t| \leq \varepsilon} |g''(x + t)| f(x) dx \right| \geq 1 \right\}. \end{aligned}$$

Since $P\{M_n > \varepsilon\} = o(n^{-1})$, Lemma 1 follows from condition (vi) and Chebyshev's inequality.

Woodroffe [1] mentioned that condition (i) and

$$\int_0^{\infty} -g(x)f(x)dx < \infty,$$

which is a weaker condition than (iv), imply all assumptions of Wald [4]. Thus we can make use of his results.

Lemma 2. *Let $\hat{\theta}_n, n \geq 1$, be a sequence of m.l.e.'s. Suppose that conditions (i)–(iii) and (iv) hold. Then for every $\varepsilon > 0$ there exists $c \geq 0$ such that*

$$P\{|\hat{\theta}_n| \geq \varepsilon\} \leq cn^{-1}$$

for all $n \geq 1$.

Proof. Let M be a positive number chosen such that

$$E\{\log \sup_{t < -M} f(X - t)\} < E\{\log f(X)\}.$$

For every $t \in [-M, -\varepsilon]$ there exists an open neighborhood U_t of t such that

$$E\{\log \sup_{u \in U_t} f(X - u)\} < E\{\log f(X)\}.$$

The existence of such a positive number M and that of such a U_t follow from Wald [4]. As $\{U_t: t \in [-M, -\varepsilon]\}$ covers the compact set $[-M, -\varepsilon]$, there exists a finite subcover of this set $[-M, -\varepsilon]$ determined by $t_j \in [-M, -\varepsilon], j = 1, \dots, m$. For notational convenience, let $U_0 = (-\infty, -M)$ and $U_j = U_{t_j}, j = 1, \dots, m$. If $|\hat{\theta}_n| \geq \varepsilon$ and $M_n < \varepsilon$, then $-\infty < \hat{\theta}_n \leq -\varepsilon$ and therefore $\hat{\theta}_n \in U_j$ for some $j \in \{0, 1, \dots, m\}$, that is to say,

$$n^{-1} \sum_{i=1}^n \log \sup_{t \in \mathcal{U}_j} f(X_i - t) \geq n^{-1} \sum_{i=1}^n \log f(X_i)$$

for some $j \in \{0, 1, \dots, m\}$. Write

$$b_j = E \{\log f(X)\} - E \{\log \sup_{t \in \mathcal{U}_j} f(X - t)\} > 0, \quad j = 0, 1, \dots, m$$

and let $2b = \min \{b_j; j = 0, 1, \dots, m\} > 0$. Then

$$|n^{-1} \sum_{i=1}^n \log \sup_{t \in \mathcal{U}_j} f(X_i - t) - E \{\log \sup_{t \in \mathcal{U}_j} f(X - t)\}| < b, \quad j = 0, 1, \dots, m$$

and

$$|n^{-1} \sum_{i=1}^n \log f(X_i) - E \{\log f(X)\}| < b$$

imply

$$n^{-1} \sum_{i=1}^n \log \sup_{t \in \mathcal{U}_j} f(X_i - t) < n^{-1} \sum_{i=1}^n \log f(X_i), \quad j = 0, 1, \dots, m.$$

Hence we have

$$\begin{aligned} P\{|\hat{\theta}_n| \geq \varepsilon\} &\leq P\{|\hat{\theta}_n| \geq \varepsilon, M_n < \varepsilon\} + P\{M_n \geq \varepsilon\} \\ &\leq \sum_{j=0}^m P\{|n^{-1} \sum_{i=1}^n \log \sup_{t \in \mathcal{U}_j} f(X_i - t) - E \{\log \sup_{t \in \mathcal{U}_j} f(X - t)\}| \geq b\} \\ &\quad + P\{|n^{-1} \sum_{i=1}^n \log f(X_i) - E \{\log f(X)\}| \geq b\} + P\{M_n \geq \varepsilon\}. \end{aligned}$$

Now, by conditions (i)–(iii) and (iv), the assertion follows from Chebyshev's inequality.

For $i = 1, \dots, n$ and $0 \leq t \leq (\log n)^{1/2}$, let

$$Z_{ni} = Z_{ni}(X_i, t) = Y_{ni} - E\{Y_{ni}\},$$

where

$$\begin{aligned} Y_{ni} &= Y_{ni}(X_i, t) = g'(X_i + a_n^{-1}t), & \text{if } X_i > a_n^{-1}, \\ &= 0, & \text{if } X_i \leq a_n^{-1}. \end{aligned}$$

Here E denotes expectation. Moreover, let $b_n(t) = E\{Z_{n1}(X_1, t)\}^2$.

Lemma 3. *Let conditions (i)–(iii), (v) and (vi) be satisfied. Then there exists a constant c such that for all $x \in R$, $n \geq 1$ and $0 \leq t \leq (\log n)^{1/2}$*

$$|P\{(nb_n(t))^{-1/2} \sum_{i=1}^n Z_{ni}(X_i, t) < x\} - \Phi(x)| \leq c(\log n)^{-1}.$$

Proof. We shall first show that

$$(3.2) \quad E\{Y_{n1}\} = -t\alpha a_n^{-1} \log a_n(1+t)^{-1} + O(a_n^{-1}(1+t)),$$

$$(3.3) \quad E\{Y_{n1}^2\} = \alpha \log a_n(1+t)^{-1} + O(1),$$

$$(3.4) \quad E\{|Y_{n1}|^3\} = O(a_n(1+t)^{-1}).$$

By condition (iii)', choose $a > 0$ and $c_0 \geq 0$ such that

$$(3.5) \quad |f(x) - \alpha x| \leq c_0 x^{1+r}, \quad |g'(x) - x^{-1}| \leq c_0 x^{r-1} \text{ and } |g''(x) + x^{-2}| \leq c_0 x^{r-2}$$

for $0 < x \leq 2a$. Next choose $\delta > 0$ such that conditions (v) and (vi) hold. Then we may establish (3.2) as follows. Since

$$g'(x + a_n^{-1}t) = g'(x) + \int_0^{a_n^{-1}t} g''(x+u) du$$

we have

$$\begin{aligned} E\{Y_{n1}\} &= \int_{a_n^{-1}}^{\infty} g'(x) f(x) dx + \int_{a_n^{-1}}^{\infty} \left\{ \int_0^{a_n^{-1}t} g''(x+u) du \right\} f(x) dx \\ &= I_1 + I_2, \quad \text{say.} \end{aligned}$$

It is easily seen that

$$I_1 = - \int_0^{a_n^{-1}} g'(x) f(x) dx,$$

so that $I_1 = O(a_n^{-1})$ by (3.5). Next we put

$$\begin{aligned} I_2 &= \int_{a_n^{-1}}^a \left\{ \int_0^{a_n^{-1}t} g''(x+u) du \right\} f(x) dx + \int_a^{\infty} \left\{ \int_0^{a_n^{-1}t} g''(x+u) du \right\} f(x) dx \\ &= I_{21} + I_{22}, \quad \text{say.} \end{aligned}$$

By condition (vi), we have $I_{22} = O(a_n^{-1}t)$. Moreover let

$$I_{21} = I_{211} + I_{212} + I_{213},$$

where

$$\begin{aligned} I_{211} &= - \int_{a_n^{-1}}^a \left\{ \int_0^{a_n^{-1}t} (x+u)^{-2} du \right\} \alpha x dx, \\ I_{212} &= - \int_{a_n^{-1}}^a \left\{ \int_0^{a_n^{-1}t} (x+u)^{-2} du \right\} (f(x) - \alpha x) dx, \\ I_{213} &= \int_{a_n^{-1}}^a \left\{ \int_0^{a_n^{-1}t} [g''(x+u) + (x+u)^{-2}] du \right\} f(x) dx. \end{aligned}$$

By easy computation, (3.5) implies that

$$I_{211} = -t\alpha a_n^{-1} \log a_n(1+t)^{-1} + O(a_n^{-1}t),$$

$$I_{212} = O(a_n^{-1}t)$$

and

$$I_{213} = O(a_n^{-1}t),$$

so that (3.2) is established.

To establish (3.3), let

$$\begin{aligned} E\{Y_{n1}^2\} &= \int_{a_n^{-1}}^a \{g'(x+a_n^{-1}t)\}^2 f(x) dx + \int_a^\infty \{g'(x+a_n^{-1}t)\}^2 f(x) dx \\ &= J_1 + J_2, \quad \text{say.} \end{aligned}$$

Condition (v) implies that $J_2 = O(1)$. Divide J_1 into J_{11} , J_{12} , J_{13} and J_{14} as follows:

$$\begin{aligned} J_{11} &= \int_{a_n^{-1}}^a (x+a_n^{-1}t)^{-2} \alpha x dx, \\ J_{12} &= \int_{a_n^{-1}}^a (x+a_n^{-1}t)^{-2} (f(x) - \alpha x) dx, \\ J_{13} &= \int_{a_n^{-1}}^a 2(x+a_n^{-1}t)^{-1} \{g'(x+a_n^{-1}t) - (x+a_n^{-1}t)^{-1}\} f(x) dx, \\ J_{14} &= \int_{a_n^{-1}}^a \{g'(x+a_n^{-1}t) - (x+a_n^{-1}t)^{-1}\}^2 f(x) dx. \end{aligned}$$

Then, by (3.5), we have

$$J_{11} = \alpha \log a_n(1+t)^{-1} + O(1),$$

$$J_{12} = O(1),$$

$$J_{13} = O(1)$$

and

$$J_{14} = O(1),$$

so that (3.3) is established.

Finally, we shall establish (3.4). Let

$$\begin{aligned} E\{|Y_{n1}|^3\} &= \int_{a_n^{-1}}^a |g'(x+a_n^{-1}t)|^3 f(x) dx + \int_a^\infty |g'(x+a_n^{-1}t)|^3 f(x) dx \\ &= K_1 + K_2, \quad \text{say.} \end{aligned}$$

By condition (v), we have $K_2 = O(1)$. Also by (3.5) we have

$$\begin{aligned} K_1 &\leq \int_{a_n^{-1}}^a \{(1+(2a)^r c_0) (x+a_n^{-1}t)^{-1}\}^3 f(x) dx \\ &= O(a_n(1+t)^{-1}). \end{aligned}$$

This implies (3.4).

From (3.2), (3.3) and (3.4), we have

$$\begin{aligned} (3.6) \quad E\{Z_{n1}^2\} &= \alpha \log a_n(1+t)^{-1} + O(1), \\ E\{|Z_{n1}|^3\} &= O(a_n(1+t)^{-1}). \end{aligned}$$

Now, the assertion of Lemma 3 follows from the Berry-Esseen theorem ([5], Theorem 12.4).

In the rest of this section, we shall study the conditional distribution of

$a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t)$ given $M_n > a_n^{-1}t$ for $0 < t \leq (\log n)^{1/2}$. The conditional distribution of X_1, \dots, X_n , given $M_n > a_n^{-1}t$, is that of independent random variables with common density

$$\begin{aligned} f_n^*(x) &= c_n f(x), & x > a_n^{-1}t \\ &= 0, & \text{otherwise} \end{aligned}$$

where

$$c_n = \left[\int_{a_n^{-1}t}^{\infty} f(x) dx \right]^{-1}.$$

For $i=1, \dots, n$ and $0 < t \leq (\log n)^{1/2}$ let

$$Z_{ni}^* = Z_{ni}^*(X_i, t) = Y_{ni}^* - E^*\{Y_{ni}^*\}$$

where

$$\begin{aligned} Y_{ni}^* &= Y_{ni}^*(X_i, t) = g'(X_i - a_n^{-1}t), & \text{if } X_i > a_n^{-1}(1+2t), \\ &= 0, & \text{if } X_i \leq a_n^{-1}(1+2t). \end{aligned}$$

Here E^* denotes conditional expectation given $M_n > a_n^{-1}t$. It is easily seen that $c_n = 1 + O(n^{-1})$ for $0 < t \leq (\log n)^{1/2}$. Thus, in a similar way to Lemma 3, we obtain

$$E^*\{Y_{n1}^*\} = t\alpha a_n^{-1} \log a_n(1+t)^{-1} + O(a_n^{-1}(1+t)),$$

$$E^*\{Z_{n1}^{*2}\} = \alpha \log a_n(1+t)^{-1} + O(1)$$

and

$$E^*\{|Z_{n1}^*|^3\} = O(a_n(1+t)^{-1}),$$

which lead to the following lemma.

Lemma 4. *Let conditions (i)–(iii), (v) and (vi) be satisfied. Then there exists a constant c such that for all $x \in R$, $n \geq 1$ and $0 < t \leq (\log n)^{1/2}$*

$$|P\{(nb_n^*(t))^{-1/2} \sum_{i=1}^n Z_{ni}^*(X_i, t) < x | M_n > a_n^{-1}t\} - \Phi(x)| \leq c(\log n)^{-1},$$

where $b_n^*(t) = E^*\{Z_{n1}^*(X_1, t)\}^2$.

4. Proof of Theorem. As the left sides of (2.1) and (2.2) are uniformly bounded for $\theta \in R$ and $t \in R$, it suffices to prove the assertion for all sufficiently large n . To simplify our notations we shall use n_0 as a generic constant instead of the phrase “for all sufficiently large n ”. In the same manner we shall use c as a generic constant to denote factors occurring in the bounds.

We shall use ideas related to Woodroffe [1]. It follows from Lemma 1 and Lemma 2 that

$$(4.1) \quad P\{a_n \hat{\theta}_n \leq -t\} = P\{a_n^{-1} \sum_{i=1}^n g'(X_i + a_n^{-1}t) \geq 0\} + O(n^{-1}),$$

where $O(n^{-1})$ is uniform in $t \in [0, a_n \varepsilon)$. Here $\varepsilon > 0$ is chosen sufficiently small so that (3.1) of Lemma 1 holds. Similarly, it follows from Lemma 1 and Lemma 2 that

$$\begin{aligned} (4.2) \quad P\{a_n \hat{\theta}_n > t\} &= P\{a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t) < 0, M_n > a_n^{-1}t\} + O(n^{-1}) \\ &= P\{a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t) < 0 \mid M_n > a_n^{-1}t\} P\{M_n > a_n^{-1}t\} \\ &\quad + O(n^{-1}), \end{aligned}$$

where $O(n^{-1})$ is uniform in $t > 0$.

We shall first show the validity of (2.1). By condition (iii)'

$$\begin{aligned} (4.3) \quad |P\{a_n^{-1} \sum_{i=1}^n g'(X_i + a_n^{-1}t) \geq 0\} - P\{a_n^{-1} \sum_{i=1}^n Y_{ni} \geq 0\}| \\ \leq \sum_{i=1}^n P\{X_i \leq a_n^{-1}\} \\ \leq c(\log n)^{-1} \end{aligned}$$

for all $n \geq n_0$ and $0 \leq t \leq (\log n)^{1/2}$. Since

$$P\{a_n^{-1} \sum_{i=1}^n Y_{ni} \geq 0\} = P\{(nb_n(t))^{-1/2} \sum_{i=1}^n Z_{ni} \geq x_n(t)\},$$

where

$$x_n(t) = -n^{1/2}(b_n(t))^{-1/2}E\{Y_{n1}\},$$

it follows from Lemma 3 that

$$(4.4) \quad |P\{a_n^{-1} \sum_{i=1}^n Y_{ni} \geq 0\} - \Phi(-x_n(t))| \leq c(\log n)^{-1}$$

for all $n \geq 1$ and $0 \leq t \leq (\log n)^{1/2}$. According to (3.2) and (3.6)

$$\begin{aligned} -x_n(t) &= (na_n^{-1}E\{Y_{n1}\})(na_n^{-2}b_n(t))^{-1/2} \\ &= \{-t + 2t \log(1+t)(\log n)^{-1} + O((1+t)(\log n)^{-1})\} \\ &\quad \times \{1 - 2 \log(1+t)(\log n)^{-1} + O((\log n)^{-1})\}^{-1/2} \\ &= -t + t \log(1+t)(\log n)^{-1} + O((1+t)(\log n)^{-1}). \end{aligned}$$

Hence, for $n \geq n_0$ and $0 \leq t \leq (\log n)^{1/2}$

$$\begin{aligned} (4.5) \quad |\Phi(-x_n(t)) - \Phi(-t)| &\leq \frac{1}{\sqrt{2\pi}} |t - x_n(t)| \max \left\{ \exp\left(-\frac{t^2}{2}\right), \exp\left(-\frac{x_n(t)^2}{2}\right) \right\} \\ &\leq c(\log n)^{-1}. \end{aligned}$$

From (4.1), (4.3), (4.4) and (4.5), there exists a constant c such that

$$(4.6) \quad |P\{a_n \hat{\theta}_n \leq -t\} - \Phi(-t)| \leq c(\log n)^{-1}$$

for all $n \geq n_0$ and $0 \leq t \leq (\log n)^{1/2}$. For $t > (\log n)^{1/2}$ we have

$$|P\{a_n \hat{\theta}_n \leq -t\} - \Phi(-t)| \leq P\{a_n \hat{\theta}_n \leq -(\log n)^{1/2}\} + \Phi(-(\log n)^{1/2}).$$

Using (4.6) and Feller ([6], p. 166, Lemma 2), we obtain

$$(4.7) \quad |P\{a_n \hat{\theta}_n \leq -t\} - \Phi(-t)| \leq c(\log n)^{-1}$$

for $n \geq n_0$ and $t > (\log n)^{1/2}$. Hence (4.6) and (4.7) imply (2.1).

We next show the validity of (2.2). By condition (iii)'

$$\begin{aligned} & |P\{a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t) < 0 \mid M_n > a_n^{-1}t\} - P\{a_n^{-1} \sum_{i=1}^n Y_{ni}^* < 0 \mid M_n > a_n^{-1}t\}| \\ & \leq \sum_{i=1}^n P\{X_i \leq a_n^{-1}(1+2t) \mid M_n > a_n^{-1}t\} \\ & \leq cn a_n^{-2}(3t^2 + 4t + 1). \end{aligned}$$

Hence, for every s , $0 < s < 1$, there exists a constant c such that for $n \geq n_0$ and $0 < t \leq (\log n)^{s/2}$

$$(4.8) \quad |P\{a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t) < 0 \mid M_n > a_n^{-1}t\} - P\{a_n^{-1} \sum_{i=1}^n Y_{ni}^* < 0 \mid M_n > a_n^{-1}t\}| \leq c(\log n)^{s-1}.$$

Applying arguments similar to those used in (4.4) and (4.5), Lemma 4 implies

$$(4.9) \quad |P\{a_n^{-1} \sum_{i=1}^n Y_{ni}^* < 0 \mid M_n > a_n^{-1}t\} - \{1 - \Phi(t)\}| \leq c(\log n)^{-1}$$

for $n \geq n_0$ and $0 < t \leq (\log n)^{1/2}$. By (4.8) and (4.9) we have

$$\begin{aligned} & |P\{a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t) < 0 \mid M_n > a_n^{-1}t\} P\{M_n > a_n^{-1}t\} - \{1 - \Phi(t)\}| \\ & \leq |P\{a_n^{-1} \sum_{i=1}^n g'(X_i - a_n^{-1}t) < 0 \mid M_n > a_n^{-1}t\} - \{1 - \Phi(t)\}| P\{M_n > a_n^{-1}t\} \\ & \quad + \{1 - \Phi(t)\} P\{M_n \leq a_n^{-1}t\} \\ & \leq \{1 - \Phi(t)\} P\{M_n \leq a_n^{-1}t\} + c(\log n)^{s-1} \end{aligned}$$

for $n \geq n_0$ and $0 < t \leq (\log n)^{s/2}$. Using Feller ([6], p. 166, Lemma 2), we obtain

$$\begin{aligned} \{1 - \Phi(t)\} P\{M_n \leq a_n^{-1}t\} & \leq ct \exp\left(-\frac{t^2}{2}\right) (\log n)^{-1} \\ & \leq c(\log n)^{-1} \end{aligned}$$

for $n \geq n_0$ and $t > 0$. Hence, from this and (4.2) it follows that for $n \geq n_0$ and $0 < t \leq (\log n)^{s/2}$

$$|P\{a_n \hat{\theta}_n > t\} - \{1 - \Phi(t)\}| \leq c(\log n)^{s-1},$$

from which (2.2) is shown by a similar argument used in (4.7). Thus we complete the proof of the theorem.

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