

ASYMPTOTIC NORMALITY OF NEAREST NEIGHBOR REGRESSION FUNCTION ESTIMATES BASED ON SOME DEPENDENT OBSERVATIONS

(Dedicated to Professor Yukihiro Kodama on his 60th Birthday)

By

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(Received December 13, 1987)

1. Introduction. For a random vector (X, Y) in R^2 with finite expectation $E|Y|$, the regression function $m(x) = E(Y|X=x)$, $x \in R$, of Y on X exists and is (μ -almost surely in x) uniquely defined in view of the equation

$$m(X) = E(Y|X)$$

where R^2 is the 2-dimensional Euclidean space and μ is the distribution of X .

Let $\{(X_i, Y_i)\}$ be independent random observations each of which has the same distributions as (X, Y) .

Stute (1984) considered a smoothed nearest neighborhood estimate

$$m_n(x_0) = (na_n)^{-1} \sum_{i=1}^n Y_i K\left(\frac{F(x_0) - F(X_i)}{a_n}\right) \quad (1.1)$$

of $m(x_0)$ and obtained the central limit theorem of $m_n(x_0)$ for μ -almost all $x_0 \in R$.

In this paper, we show that Stute's result (1984) remains true for some weakly dependent observations if $E|Y|^{2+\delta} < \infty$ for some $\delta > 0$.

2. Main results. Let $\{(X_n, Y_n), n=0, \pm 1, \pm 2, \dots\}$ be a strictly stationary sequence of random vectors in the plane with distribution function (d. f.) $H(x, y)$.

We consider the following two conditions:

(I) $\{(X_n, Y_n)\}$ satisfies the *-mixing condition, i. e.,

$$(2.1) \quad \phi(n) = \sup_{A \in \mathcal{M}_{-\infty}^0, B \in \mathcal{M}_n^\infty} \left| \frac{P(AB) - P(A)P(B)}{P(AB)} \right| \rightarrow 0 \text{ as } n \rightarrow \infty;$$

(II) $\{(X_n, Y_n)\}$ satisfies the ϕ -mixing condition, i. e.,

$$(2.2) \quad \phi(n) = \sup_{A \in \mathcal{M}_{-\infty}^0, B \in \mathcal{M}_n^\infty} \left| \frac{P(AB) - P(A)P(B)}{P(A)} \right| \rightarrow 0 \text{ as } n \rightarrow \infty.$$

Here, \mathcal{M}_a^b denotes the σ -algebra generated by $(X_a, Y_a), \dots, (X_b, Y_b)$ ($a \leq b$).

Let H_n denote the (bivariate) empirical d. f. of the sample $\{(X_1, Y_1), \dots,$

(X_n, Y_n) and F_n the empirical d.f. of $\{X_1, \dots, X_n\}$. Let K be a twice continuously differentiable kernel function vanishing outside some bounded interval. We assume without loss of generality that K vanishes outside $(-1, 1)$. Further, let $\{a_n\}$ be any bandsequence such that $a_n \downarrow 0$.

We consider the following nearest neighbor estimator of $m(x_0)$:

$$(2.3) \quad \begin{aligned} m_n(x_0) &= (na_n)^{-1} \sum_{i=1}^n Y_i K\left(\frac{F_n(x_0) - F_n(X_i)}{a_n}\right) \\ &= a_n^{-1} \int y K\left(\frac{F_n(x_0) - F_n(X_i)}{a_n}\right) H_n(dx, dy). \end{aligned}$$

Now, we state our main results, which are extensions of Stute's result (1984).

Theorem 1. Let $\{(X_i, Y_i)\}$ be a strictly stationary $*$ -mixing sequence of random vectors satisfying the following conditions;

(i) each X_i has a continuous d.f. F and

(ii) there exists a positive number δ for which $E|Y_i|^{2+\delta} < \infty$ and $\sum (n+1)^2 \phi^{1/4}(n) < \infty$. Let

$$(2.4) \quad \sigma_1^2 = \text{Var}(Y_1 | X_1 = x_0) \int_{-1}^1 K^2(u) du > 0.$$

Suppose $na_n^3 \rightarrow \infty$. Then, we have

$$(2.5) \quad (na_n)^{1/2} \{m_n(x_0) - \bar{m}_n(x_0)\} \xrightarrow{D} N(0, \sigma_1^2)$$

for μ -almost all $x_0 \in R$ where

$$(2.6) \quad \bar{m}_n(x_0) = a_n^{-1} \int y K\left(\frac{F(x_0) - F(x)}{a_n}\right) H(dx, dy).$$

Theorem 2. Let $\{(X_n, Y_n)\}$ be a strictly stationary ϕ -mixing sequence of random vectors satisfying the following conditions;

(i) each X_i has a continuous d.f. F and

(ii) there exists a positive number δ for which $E|Y_i|^{2+\delta} < \infty$ and

$$\sum_{n=1}^{\infty} (n+1)^2 \phi^{1/4}(n) < \infty.$$

Suppose $na_n^3 \rightarrow \infty$. Suppose further that

$$(2.7) \quad \begin{aligned} \sigma_0^2 &= \lim_{n \rightarrow \infty} \left(\frac{n}{a_n}\right) E \left[\int \{y - m(x_0)\} K\left(\frac{F(x_0) - F(x_1)}{a_n}\right) \right. \\ &\quad \left. \times \{H_n(dx, dy) - H(dx, dy)\} \right]^2 \end{aligned}$$

exists and is positive. Then, the conclusion of Theorem 1 holds.

Next, let

$$m \circ F^{-1}(u) = E(Y_1 | F(X_1) = u).$$

Corollary. Let K be a twice continuously differentiable kernel function such that $K(u) = 0$ for $|u| \geq 1$, $\int K(u) du = 1$ and $\int u K(u) du = 0$. Suppose the conditions in Theorem 2 hold. Suppose further $m \circ F^{-1}$ is twice continuously differentiable in a neighborhood of x_0 ($0 < F(x_0) < 1$) and $na_n^5 \rightarrow 0$. Then

$$(2.8) \quad (na_n)^{1/2}(\bar{m}_n(x_0) - m(x_0)) \rightarrow 0 \text{ as } n \rightarrow \infty$$

holds. Hence, we have

$$(2.9) \quad (an_n)^{1/2}\{m_n(x_0) - m(x_0)\} \xrightarrow{D} N(0, \sigma_0^2).$$

3. Auxiliary results. Let $\{\xi_i\}$ be a strictly stationary ϕ -mixing sequence of d -dimensional random vectors.

Let $i_1 < i_2 < \dots < i_k$ be arbitrary integers and put

$$(3.1) \quad P_j^{(k)}(E^{(j)} \times E^{(k-j)}) = P((\xi_{i_1}, \dots, \xi_{i_j}) \in E^{(j)}) P((\xi_{i_{j+1}}, \dots, \xi_{i_k}) \in E^{(k-j)})$$

($j = 1, \dots, k$).

$$P_0^{(k)}(E^{(k)}) = P((\xi_{i_1}, \dots, \xi_{i_k}) \in E^{(k)}).$$

where $E^{(j)}$ is a Borel set in R^{jd} . The following lemma is a special case of Lemma 1 in Yoshihara (1976) since a ϕ -mixing sequence is an absolutely regular sequence.

Lemma 3.1. Let $h(x_1, \dots, x_k)$ be a Borel function such that

$$(3.2) \quad \int \dots \int |h(x_1, \dots, x_k)|^{1+\delta} dP_j^{(k)} \leq M \quad (0 \leq j \leq k-1)$$

for some $\delta > 0$. Then

$$(3.3) \quad \left| \int \dots \int h(x_1, \dots, x_k) dP_0^{(k)} - \int \dots \int h(x_1, \dots, x_k) dP_j^{(k)} \right| \leq 4M^{1/(1+\delta)} \beta^{\delta/(1+\delta)}(i_{j+1} - i_j).$$

Next, let $\{X_n\}$ be a strictly stationary ϕ -mixing sequence, and put

$$(3.4) \quad \alpha_n(x) = n^{1/2}\{F_n(x) - F(x)\}, \quad x \in R$$

which denotes the empirical process pertaining to X_1, \dots, X_n .

In the following, we often use the well-known inequalities.

Lemma 3.2. Let $\{\xi_i\}$ be \ast -mixing with mixing coefficient $\phi(k)$. Let η be

$\mathcal{M}_{-\infty}^k$ -measurable and let ζ be $\mathcal{M}_{k+n}^{\infty}$ -measurable. If $E|\eta| < \infty$ and $E|\zeta| < \infty$, then

$$(3.5) \quad |E\xi\eta - E\xi E\eta| \leq \phi(n)E|\xi|E|\eta|.$$

(See, for example, Samur (1984)).

Lemma 3.3. Let $\{\xi_i\}$ be ϕ -mixing with mixing coefficient $\phi(k)$. Let η be $\mathcal{M}_{-\infty}^k$ -measurable and let ζ be $\mathcal{M}_{k+n}^{\infty}$ -measurable. If $E|\eta|^r < \infty$, $E|\zeta|^s < \infty$ and $r^{-1} + s^{-1} = 1$ ($r, s > 0$), then

$$(3.6) \quad |E\xi\eta - E\xi E\eta| \leq 2\{\phi(n)\}^{1/r}\{E|\eta|^r\}^{1/r}\{E|\zeta|^s\}^{1/s}.$$

(See, for example, Billingsley (1968)).

Lemma 3.4. Suppose $\{\xi_i\}$ is a strictly stationary ϕ -mixing sequence of zero-mean random variables such that $M_{\delta} = E|\xi_i|^{2+\delta}$ for some $\delta (> 0)$ and

$$c_0 = \sum_{k=0}^{\infty} (k+1)^2 \phi^{1/4}(k) \quad (\phi(0)=1).$$

Then

$$(3.7) \quad E \left| \sum_{i=1}^n \xi_i \right|^{2+\delta} \leq c_0 M_{\delta} n^{1+(\delta/2)}$$

(See, Corollary 2.1 in Utev (1984)).

Next, let $\{X_n\}$ be a strictly stationary ϕ -mixing sequence of random variables and put

$$(3.8) \quad \alpha_n(x) = n^{1/2}\{F_n(x) - x\}, \quad 0 \leq x \leq 1,$$

which denotes the uniform empirical process pertaining to X_1, \dots, X_n . In the following, we often use the well-known representation

$$\alpha_n(x) = \bar{\alpha}_n(F(x)), \quad x \in R$$

of α_n in terms of a uniform empirical process α_n .

From now on, we shall agree to denote by the letter c , with or without subscript, some quantity bounded in absolute value.

Lemma 3.5. Let $\{X_n\}$ be a strictly stationary ϕ -mixing sequence of random variables. If $\sum_{n=1}^{\infty} n\phi^{1/2}(n) < \infty$, then

$$(3.9) \quad P\left(\sup_{s \leq F(x) \leq s + c_1 a_n} |\bar{\alpha}_n(F(x)) - \bar{\alpha}_n(s)| \geq \lambda a_n^{1/2}\right) \leq c\lambda^{-4}$$

for all n sufficiently large.

Proof. In Sen (1971) it was proved that for every ε ($0 < \varepsilon < 1$), if $t-s > \varepsilon n^{-1}$, then for all n

$$E|\bar{\alpha}_n(t) - \bar{\alpha}_n(s)|^4 \leq c\varepsilon^{-1}(t-s)^2.$$

Hence, by the method of the proof of (22.20) in Billingsley (1968) we have

$$(1.10) \quad P\left(\sup_{s \leq t \leq s+m_p} |\bar{\alpha}_n(t) - \bar{\alpha}_n(s)| \geq \lambda a_n^{1/2}\right) \leq c\varepsilon^{-1} \lambda^{-4} a_n^{-2} (m_p)^2.$$

Thus, putting $m_p = C_1 a_n$, we have (3.9) from (3.10). \square

Lemma 3.6 *Let $\varepsilon > 0$ be given arbitrarily. Then, there is a constant C_1 such that, up to an event of probability less than or equal to ε , we have*

$$(3.11) \quad |F(x_0) - F(x)| \leq C_1 a_n$$

whenever

$$(3.12) \quad |F_n(x_0) - F_n(x)| \leq a_n.$$

Proof. Since by Theorem 22.1 in Billingsley (1968) (cf. Sen (1971))

$$\sup_{x: 0 \leq F(x) \leq 1} \sqrt{n} |F_n(x) - F(x)| \xrightarrow{D} \sup_{0 \leq t \leq 1} |Z(t)| \quad (n \rightarrow \infty),$$

so for an arbitrary number $\varepsilon > 0$ there exists some finite number C such that for all n sufficiently large

$$P(A_n(C)) \geq 1 - \varepsilon$$

where

$$A_n(C) = \left\{ \sup_{x: 0 \leq F(x) \leq 1} |F_n(x) - F(x)| \leq C n^{-1/2} \right\}.$$

Because of (3.12), on the set $A_n(C)$ we have that

$$\begin{aligned} |F(x_0) - F(x)| &\leq |F(x_0) - F_n(x_0)| + |F_n(x_0) - F_n(x)| + |F_n(x) - F(x)| \\ &\leq a_n + 2C n^{-1/2} \leq C_1 a_n \end{aligned}$$

for some $C_1 < \infty$, and the proof is completed. \square

4. Lemmas. In this and the following section, we always assume that conditions of Theorem 2 are satisfied. We note first that if K is twice differentiable, then by Taylor's expansion we have

$$(4.1) \quad \begin{aligned} m_n(x_0) &= a_n^{-1} \int y K\left(\frac{F(x_0) - F(x)}{a_n}\right) H_n(dx, dy) \\ &\quad + a_n^{-2} \int y \{F_n(x_0) - F_n(x) - F(x_0) + F(x)\} K'\left(\frac{F(x_0) - F(x)}{a_n}\right) H_n(dx, dy) \end{aligned}$$

$$\begin{aligned}
& + a_n^{-3} \int y \{F_n(x_0) - F_n(x) - F(x_0) + F(x)\}^2 \frac{K''(\Delta)}{2} H_n(dx, dy) \\
& = I_1 + I_2 + I_3 \text{ (say),}
\end{aligned}$$

where Δ is some number between $a_n^{-1}\{F_n(x_0) - F(x)\}$ and $a_n^{-1}\{F(x_0) - F(x)\}$.

Lemma 4.1. $(na_n)^{1/2} I_3 \rightarrow 0$ in probability as $n \rightarrow \infty$.

Proof. Since K vanishes outside $(-1, 1)$, the above expansion of $m_n(x_0)$ holds true with integration restricted to those x for which $|F_n(x) - F(x)| < a_n$. Let $b_n = (na_n^3)^{1/6}$. Then, by Lemma 3.5 and 3.7

$$\begin{aligned}
& P\left(\sup_{x: |F(x_0) - F(x)| \leq C_1 a_n} (na_n^{-1})^{1/2} |F_n(x_0) - F_n(x) - F(x_0) + F(x)| \geq b_n\right) \\
& = P\left(\sup_{x: |F(x_0) - F(x)| \leq C_1 a_n} |\bar{\alpha}_n(F(x_0)) - \bar{\alpha}_n(F(x))| \geq b_n a_n^{1/2}\right) \\
& \leq cb_n^{-4} = c(na_n^3)^{-2/3}
\end{aligned}$$

On the other hand, as $\{Y_i\}$ is strictly stationary and ϕ -mixing, and $E|Y_i|^{2+\delta} < \infty$, so by the pointwise ergodic theorem for stationary sequences

$$(4.2) \quad \limsup_{n \rightarrow \infty} \int |y| H_n(dx, dy) < \infty \quad \text{a. s.}$$

Hence, the assertion of the lemma follows from (4.2) and the fact that K'' is bounded, upon observing that $\epsilon > 0$ was arbitrary and $na_n^3 \rightarrow \infty$. \square

Next, to show that $(na_n)^{1/2} I_2$ is asymptotically equivalent to

$$-(na_n)^{1/2} a_n^{-1} m(x_0) \int K' \left(\frac{F(x_0) - F(x)}{a_n} \right) \{F_n(dx) - F(dx)\},$$

define $\{Z_n, n \geq 1\}$ by

$$(4.3) \quad Z_n = n^{-1} a_n^{-3/2} \sum_{i=1}^n \{Y_i - m(X_i)\} \{\alpha_n(x_0) - \alpha_n(X_i)\} K' \left(\frac{F(x_0) - F(X_i)}{a_n} \right).$$

Lemma 4.2. $Z_n \rightarrow 0$ in probability as $n \rightarrow \infty$.

Proof. We rewrite Z_n as follows;

$$(4.4) \quad Z_n = n^{-3/2} a_n^{-3/2} \sum_{i=1}^n \sum_{j=1}^n \{Y_i - m(X_i)\} g(X_j; x_0, X_i) K' \left(\frac{F(x_0) - F(X_i)}{a_n} \right),$$

where

$$(4.5) \quad g(X_j; x_0, X_i) = \{I(X_j < x_0) - F(x_0)\} - \{I(X_j < X_i) - F(X_i)\}$$

and

$$(4.6) \quad I(y < x) = I_{(-\infty, x)}(y) - F(x).$$

We note that if $\{(\tilde{X}_i, \tilde{Y}_i), i=1, 2, \dots, n\}$ are i.i.d. random vectors with d.f. $H(x, y)$, then

$$(4.7) \quad E\{\tilde{Y}_i - m(\tilde{X}_i) | \tilde{X}_i\} = 0 \quad \text{a. s.}$$

and

$$(4.8) \quad E\{g(\tilde{X}_j; x_0, \tilde{X}_i) | \tilde{X}_i\} = 0 \quad \text{a. s.}$$

Put

$$(4.9) \quad G((x_1, y_1); x_0, x_2) = (y_1 - m(x_1))g(x_2; x_0, x_1)K'\left(\frac{F(x_0) - F(x_1)}{a_n}\right)$$

and let $P(x_{i_1}, y_{i_1}, \dots, x_{i_k}, y_{i_k})$ be the d.f. of $(X_{i_1}, Y_{i_1}, \dots, X_{i_k}, Y_{i_k})$. Then it is obvious that

$$(4.10) \quad \int G((x_1, y_1); x_0, x_2) dP(x_2) = 0$$

and

$$(4.11) \quad \int G((x_1, y_1); x_0, x_2) dP(x_1, y_1) = 0.$$

For brevity, let

$$(4.12) \quad \theta(i; j) = G((X_i, Y_i); x_0, X_j).$$

To prove $\lim_{n \rightarrow \infty} EZ_n^2 = 0$, we consider the following quantity;

$$(4.13) \quad \begin{aligned} n^2 a_n^3 EZ_n^2 &= \sum_{i=1}^n \sum_{j=1}^n E\theta^2(i; j) + 2 \sum_{i=1}^n \sum_{1 \leq j_1 < j_2 \leq n} E\theta(i; j_1)\theta(i; j_2) \\ &\quad + 2 \sum_{1 \leq i_1 < i_2 \leq n} \sum_{j_1=1}^n \sum_{j_2=1}^n E\theta(i_1; j_1)\theta(i_2; j_2) \\ &= J_1 + 2J_2 + 2J_3, \quad (\text{say}). \end{aligned}$$

Since K' and g are bounded and

$$E|Y_i - m(X_i)|^{2+\delta} \leq cE|Y_i|^{2+\delta} < \infty,$$

so

$$(4.14) \quad E\theta^2(i; j) \leq c \quad \text{for all } i, j.$$

Thus, we have

$$(4.15) \quad |J_1| \leq cn^2.$$

Next, if $i < j_1$ and $j_i - i > j_2 - j_1$, then by Lemma 3.1 and (4.10)

$$\begin{aligned}
& |E\{\theta(i; j_1)\theta(i; j_2)\}| \\
& \leq \left| \int \cdots \int G((x_i, y_i); x_0, x_{j_1})G((x_i, y_i); x_0, x_{j_2})dP(x_i, y_i, x_{j_1})dP(x_{j_2}) \right. \\
& \quad \left. + c\phi^{2/(2+\delta)}(j_2-j_1) \right. \\
& \leq c\phi^{2/(2+\delta)}(j_2-j_1).
\end{aligned}$$

Hence, we have

$$\begin{aligned}
(4.16) \quad & \sum_{1 \leq i < j_1 < j_2 \leq n} |E\theta(i; j_1)\theta(i; j_2)| \\
& \leq \sum_{1 \leq i < j_1 < j_2 \leq n} c\phi^{2/(2+\delta)}(j_2-j_1) \leq cn^2.
\end{aligned}$$

Similarly, we have

$$(4.17) \quad \sum_{1 \leq j_1 < j_2 < i \leq n} |E\{\theta(i; j_1)\theta(i; j_2)\}| \leq c \sum_{1 \leq j_1 < j_2 < i \leq n} \phi^{2/(2+\delta)}(j_2-j_1) \leq cn^2$$

and

$$(4.18) \quad \sum_{1 \leq j_1 < i \leq j_2 \leq n} |E\{\theta(i; j_1)\theta(i; j_2)\}| \leq c \sum_{1 \leq j_1 < i < j_2 \leq n} \phi^{2/(2+\delta)}(j_2-i) \leq cn^2.$$

Further, by (4.15)

$$(4.19) \quad \left\{ \sum_{1 \leq i < j_1 < j_2 \leq n} + \sum_{1 \leq j_1 < j_2 = i \leq n} \right\} |E\{\theta(i; j_1)\theta(i; j_2)\}| \leq cn^2$$

Thus, from (4.16)-(4.19) we obtain

$$(4.20) \quad |J_2| \leq cn^2.$$

It remains to prove $|J_3| \leq cn^2$. We use the method of the proof of Lemma 2 in Yoshihara (1976)

By Lemma 3.1, (4.10) and (4.11), we have the following inequalities:

If $1 \leq i_1 < i_2 \leq j_1 < j_2 \leq n$ and $j_2 - j_1 \geq i_2 - i_1$, then

$$(4.21) \quad |E\{\theta(i_1; j_1)\theta(i_2; j_2)\}| \leq c\phi^{2/(2+\delta)}(j_2-j_1)$$

and similarly, if $1 \leq i_1 < i_2 \leq j_1 < j_2 \leq n$ and $i_2 - i_1 \geq j_2 - j_1$, then

$$(4.22) \quad |E\{\theta(i_1; j_1)\theta(i_2; j_2)\}| \leq c\phi^{2/(2+\delta)}(i_2-i_1).$$

Thus, from (4.21) and (4.22) we obtain

$$\begin{aligned}
(4.23) \quad & \left| \sum_{1 \leq i_1 < i_2 \leq j_1 < j_2 \leq n} E\{\theta(i_1; j_1)\theta(i_2; j_2)\} \right| \\
& \leq \left\{ \sum_{\substack{1 \leq i_1 < i_2 \leq j_1 < j_2 \leq n \\ i_2 - i_1 \geq j_2 - j_1}} + \sum_{\substack{1 \leq i_1 < i_2 \leq j_1 < j_2 \leq n \\ i_2 - i_1 \leq j_2 - j_1}} \right\} |E\theta(i_1; j_1)\theta(i_2; j_2)| \\
& \leq cn^2 \sum_{k=1}^n (k+1)\phi^{2/(2+\delta)}(k) \leq cn^2.
\end{aligned}$$

(ii) Similarly, we have

$$(4.24) \quad \begin{aligned} & \left| \sum_{1 \leq i_1 < j_1 \leq i_2 < j_2 \leq n} E\{\theta(i_1; j_1)\theta(i_2; j_2)\} \right| \\ & \leq \left\{ \sum_{\substack{1 \leq i_1 < j_1 \leq i_2 < j_2 \leq n \\ j_1 - i_1 \geq j_2 - i_2}} + \sum_{\substack{1 \leq i_1 < j_1 \leq i_2 < j_2 \leq n \\ j_1 - i_1 \leq j_2 - i_2}} \right\} |E\theta(i_1; j_1)\theta(i_2; j_2)| \\ & \leq cn^2, \end{aligned}$$

$$(4.25) \quad \begin{aligned} & \left| \sum_{1 \leq i_1 < j_1 < j_2 < i_2 \leq n} |E\theta(i_1; j_1)\theta(i_2; j_2)| \right| \\ & \leq \left\{ \sum_{\substack{1 \leq i_1 < j_1 < j_2 < i_2 \leq n \\ j_1 - i_1 \geq j_2 - i_2}} + \sum_{\substack{1 \leq i_1 < j_1 < j_2 < i_2 \leq n \\ j_1 - i_1 \leq j_2 - i_2}} \right\} |E\theta(i_1; j_1)\theta(i_2; j_2)| \\ & \leq cn^2, \end{aligned}$$

$$(4.26) \quad \begin{aligned} & \left| \sum_{1 \leq i_1, j_1 \leq n} \sum_{i_2=1}^n E\theta(i_1; j_1)\theta(i_2; j_1) \right| \\ & \leq \sum_{i_1=1}^n \sum_{i_2=1}^n |E\theta(i_1; j_1)\theta(i_2; i_1)| \\ & \quad + 2 \sum_{1 \leq i_1 < j_1 \leq n} \sum_{i_2=1}^n |E\theta(i_1; j_1)\theta(i_2; j_1)| \\ & \leq cn^2 \left(1 + \sum_{k=1}^n \phi^{2/(2+\delta)}(k) \right) \leq cn^2 \end{aligned}$$

$$(4.27) \quad \left| \sum_{1 \leq i_2, j_2 \leq n} \sum_{i_1=1}^n E\theta(i_1; j_1)\theta(i_2; j_2) \right| \leq cn^2 \sum_{k=1}^n \phi^{2/(2+\delta)}(k) \leq cn^2.$$

Hence, from (4.24)-(4.26) we have

$$(4.28) \quad |J_3| \leq cn^2.$$

Now, from (4.15), (4.20), (4.28) and the fact that $na_n^3 \rightarrow \infty$ we have $\lim EZ_n^2 = 0$, which completes the proof. \square

Next, consider the function

$$(4.29) \quad k(x, y) = m(x)K' \left(\frac{F(x_0) - F(x)}{a_n} \right) \{I_{(-\infty, x_0]}(y) - I_{(-\infty, x]}(y)\}$$

with corresponding "von Mises" statistic

$$\begin{aligned} T_n &= n \int k(x, y) \{F_n(dy) - F(dy)\} \{F_n(dx) - F(dx)\} \\ &= \int k(x, y) \alpha_n(dy) \alpha_n(dx). \end{aligned}$$

Then, using the method of the proof of Lemma 2 in Yoshihara and noting $\sum (k+1)^2 \phi^{1/4}(k) < \infty$, we have

$$ET_n^2 = O(1) \text{ as } n \rightarrow \infty,$$

which implies

$$a_n^{-3/2} \int k(x, y) \alpha_n(dy) \{F_n(dx) - F(dx)\} \rightarrow 0 \text{ in probability}$$

since $na_n^3 \rightarrow \infty$.

Thus, we obtain that $(na_n)^{1/2} I_2$ is asymptotically equivalent to

$$(4.30) \quad a_n^{-3/2} \int m(x) \{ \alpha_n(x_0) - \alpha_n(x) \} K' \left(\frac{F(x_0) - F(x)}{a_n} \right) F(dx).$$

Lemma 4.3. *If conditions of Theorem 2 hold, then*

$$(4.31) \quad I_4 = a_n^{-3/2} \int |m(x) - m(x_0)| \alpha_n(x_0) - \alpha_n(x) \left| K' \left(\frac{F(x_0) - F(x)}{a_n} \right) \right| F(dx) \rightarrow 0$$

in probability

for μ -almost all $x_0 \in R$.

Proof. In the proof of Lemma 3 in Stute (1984), it was proved that $F^{-1}(F(x_0)) = x_0$ for μ -almost all x_0 and that

$$\lim_{n \rightarrow \infty} \int_{-1}^1 |m(F^{-1}(s - u a_n)) - m(F^{-1}(s))| |K'(u)| du = 0$$

for μ -almost all s , say, for all $s \in A$, if F is continuous. Hence

$$\mu(\{x_0 : F(x_0) \in A, F^{-1}(F(x_0)) = x_0\}) = 1.$$

Let $F(x_0) \in A$ and put

$$e_n(x_0) = \left\{ a_n^{-1} \int_0^1 |m(F^{-1}(u)) - m(x_0)| \left| K' \left(\frac{F(x_0) - u}{a_n} \right) \right| du \right\}^{-1/6}$$

Then, by Lemma 3.5 we have that for any $\varepsilon > 0$

$$\begin{aligned} & P(I_4 > \varepsilon) \\ & \leq P(a_n^{-3/2} \sup_{u: |F(x_0) - u| \leq a_n} |\bar{\alpha}_n(F(x_0)) - \bar{\alpha}_n(u)| \\ & \quad \times \int_0^1 |m(F^{-1}(u)) - m(x_0)| \left| K' \left(\frac{F(x_0) - u}{a_n} \right) \right| du > \varepsilon) \\ & \leq P\left(\sup_{u: |F(x_0) - u| \leq a_n} |\bar{\alpha}_n(F(x_0)) - \bar{\alpha}_n(u)| \geq e_n(x_0) a_n^{1/2} \right) \\ & \leq c e_n^{-4}(x_0) = o(1) \end{aligned}$$

as $n \rightarrow \infty$. Hence, (4.31) is obtained. □

Lemma 4.4. $(na_n)^{1/2} I_2$ is asymptotically equivalent to

$$(4.32) \quad -a_n^{-1/2}m(x_0)\int K\left(\frac{F(x_0)-F(x)}{a_n}\right)\alpha_n(dx).$$

Proof. From Lemma 4.3 it follows that $(na_n)^{1/2}I_2$ is asymptotically equivalent to

$$\begin{aligned} & a_n^{-3/2}m(x_0)\int[\alpha_n(x_0)-\alpha_n(x)]K'\left(\frac{F(x_0)-F(x)}{a_n}\right)F(dx) \\ &= -a_n^{-3/2}m(x_0)\int\alpha_n(x)K'\left(\frac{F(x_0)-F(x)}{a_n}\right)F(dx) \\ &= -a_n^{-3/2}m(x_0)\int K\left(\frac{F(x_0)-F(x)}{a_n}\right)\alpha_n(dx), \end{aligned}$$

which completes the proof. \square

5. Proofs. Let

$$(5.1) \quad \begin{aligned} W_{ni} &= a_n^{-1/2}(Y_i - m(x_0))K\left(\frac{F(x_0) - F(X_i)}{a_n}\right) \\ & - a_n^{-1/2}E\left\{(Y_i - m(x_0))K\left(\frac{F(x_0) - F(X_i)}{a_n}\right)\right\} \\ & \quad (i=1, \dots, n). \end{aligned}$$

Then, it is clear that W_{n1}, \dots, W_{nn} are identically distributed and satisfy the ϕ -mixing condition with

$$\sum(k+1)^2\phi^{1/4}(k) < \infty.$$

Lemma 5.1. For any δ' ($0 \leq \delta' < 1$)

$$(5.2) \quad E\left|\sum_{i=1}^n W_{ni}\right|^{2+\delta'} \leq ck^{(2+\delta')/2}a_n^{-\delta'/2}.$$

Proof. By Lemma 3.3,

$$E\left|\sum_{i=1}^k W_{ni}\right|^{2+\delta'} \leq ck^{(2+\delta')}E|W_{n1}|^{2+\delta'}$$

Since K vanishes outside $(-1, 1)$ and is bounded, so for all n sufficiently large we have

$$\begin{aligned} E|W_{n1}| &\leq ca_n^{-(2+\delta')/2}\int|y-m(x_0)|^{2+\delta'}\left|K\left(\frac{F(x_0)-F(x)}{a_n}\right)\right|^{2+\delta'}H(dx, dy) \\ &= ca_n^{-(2+\delta')/2}\int h(x)\left|K\left(\frac{F(x_0)-F(x)}{a_n}\right)\right|^{2+\delta'}F(dx) \\ &\leq ca_n^{-\delta'/2}\int_{-1}^1 h(F^{-1}(F(x_0)-a_nu))|K(u)|^{2+\delta'}du \end{aligned}$$

$$\leq c a_n^{-\delta'/2}$$

where

$$h(x) = E\{|Y_1 - m(x_0)|^{2+\delta} | X = x\}.$$

Thus, (5.2) is obtained. \square

Proof of Theorem 2. According to Lemma 4.4, to prove Theorem 2 it remains to show

$$(5.3) \quad I_5 = \left(\frac{n}{a_n}\right)^{1/2} \int \{y - m(x_0)\} K\left(\frac{F(x_0) - F(x)}{a_n}\right) \{H_n(dx, dy) - H(dx, dy)\} \\ \xrightarrow{D} N(0, \sigma_0^2).$$

Let $p = [n^{2/3}]$, $q = [n^{1/3}]$ and $k = [n/(p+q)]$ where $[s]$ denotes the largest integer j such that $j \leq s$. Put

$$\eta_j = \sum_{i=(j-1)(p+q)+1}^{(j-1)(p+q)+p} W_{ni} \quad (j=1, \dots, k), \\ \zeta_j = \sum_{i=(j-1)(p+q)+p+1}^{j(p+q)} W_{ni} \quad (j=1, \dots, k), \\ \zeta_{k+1} = \sum_{i=k(p+q)+1}^n W_{ni}.$$

We rewrite I_5 as follows:

$$I_5 = \frac{1}{\sqrt{n}} \sum_{i=1}^n W_{ni} = \frac{1}{\sqrt{n}} \sum_{j=1}^k \eta_j + \frac{1}{\sqrt{n}} \sum_{j=1}^{k+1} \zeta_j.$$

By Lemma 5.1 and definitions of q and k we can easily show that

$$\frac{1}{\sqrt{n}} \sum_{j=1}^{k+1} \zeta_j \rightarrow 0 \quad \text{in probability.}$$

So, by Lemma 18.4.1 in Ibragimov and Linnik (1971) to show (5.3) it suffices to prove

$$\frac{1}{\sqrt{n}} \sum_{j=1}^k \eta_j \xrightarrow{D} N(0, \sigma_0^2).$$

Since $\{\eta_j\}$ is ϕ -mixing with $\sum (j+1)^2 \phi^{1/4}(j) < \infty$, so

$$\left| E\left\{\exp\left(it \frac{1}{\sqrt{n}} \sum_{j=1}^k \eta_j\right)\right\} - \left[E\left\{\exp\left(it \frac{\eta_i}{\sqrt{n}}\right)\right\} \right]^k \right| \leq ck\phi(q) = o(1)$$

(cf. Lemma 2 in Billingsley (1968) page 171).

Hence, to show (5.3), it is enough to prove

$$(5.4) \quad \left[E \left\{ \exp \left(it \frac{\eta_j}{\sqrt{n}} \right) \right\} \right]^k \longrightarrow e^{-t^2/2}$$

By Lemma 5.1 and the definition of p

$$\begin{aligned} E \left\{ \exp \left(it \frac{\eta_j}{\sqrt{n}} \right) \right\} &= 1 - \frac{t^2}{2n} E \eta_j^2 + O \left(\frac{|t|^{2+\delta}}{n^{(2+\delta)/2}} E |\eta_j|^{2+\delta} \right) \\ &= 1 - \frac{t^2}{2n} E \eta_j^2 + o \left(|t|^{2+\delta} n^{-(2+\delta)/2} a_n^{-\delta/2} \right). \end{aligned}$$

So, using the facts that $na_n^3 \rightarrow \infty$ and

$$\lim_{n \rightarrow \infty} \frac{k E \eta_j^2}{n} = \sigma_0^2,$$

we have (5.4). Thus, the proof is completed. \square

Proof of Theorem 1. Since any *-mixing sequence is ϕ -mixing, so by Theorem 2 to prove Theorem 1 it is enough to show

$$\lim_{n \rightarrow \infty} \frac{1}{n} E \left(\sum_{i=1}^n W_{ni} \right)^2 = \text{Var} (Y_1 | X = x_0) \int_{-1}^1 K^2(u) du = \sigma_1^2.$$

Now, we evaluate

$$E \left(\sum_{i=1}^n W_{ni} \right)^2 = \sum_{i=1}^n E W_{ni}^2 + 2 \sum_{1 \leq i < j \leq n} E W_{ni} W_{nj}.$$

By Lemma 3.4 and the fact that $E W_{ni} = 0$, we have

$$|E W_{ni} W_{nj}| \leq c \phi(|i-j|) E |W_{ni}| E |W_{nj}|.$$

As

$$\begin{aligned} E |W_{ni}| &\leq c a_n^{-1/2} \int \left| y - m(x_0) \right| \left| K \left(\frac{F(x_0) - F(x)}{a_n} \right) \right| H(dx, dy) \\ &\leq c a_n^{1/2} \int_{-1}^1 |m(F^{-1}(F(x_0) - u a_n)) - m(x_0)| |K(u)| du \\ &\leq c a_n^{1/2} \end{aligned}$$

for all n sufficiently large, so using the fact that $\sum \phi(k) < \infty$

$$\left| \frac{1}{n} E \left| \sum_{i=1}^n W_{ni} \right|^2 - E W_{n1}^2 \right| \leq c a_n.$$

On the other hand, by the method of the proof of Theorem 2

$$E W_{n1}^2 \longrightarrow \text{Var} (Y | X = x_0) \int_{-1}^1 K^2(u) du.$$

Henc, we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} E \left(\sum_{i=1}^n W_{ni} \right)^2 = \text{Var}(Y | X=x_0) \int_{-1}^1 K^2(u) du.$$

Thus, the proof of Theorem 1 is completed. \square

Proof of Corollary. The proof is identical to that of Corollary in Stute (1984) and so is omitted.

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