

A NOTE ON RECURSIVE ESTIMATOR OF THE DENSITY FUNCTION WHICH IS NOT NECESSARY CONTINUOUS

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Summary. Consider a sequence X_1, X_2, \dots, X_n of independent, identically distributed random variables with unknown density function f , which with its first derivative are not necessarily continuous, and let

$$f_n^*(x) = (n^2 h_n)^{-1/2} \sum_{j=1}^n h_j^{-1/2} K\left(\frac{x - X_j}{h_j}\right)$$

be the recursive kernel estimator of f . It will be shown, under certain additional regularity conditions on K and $|h(n)|$, that, $MISE [f_n^*(x)] = O(n^{-4/5})$ if f and f' are continuous, where as $MISE [f_{n_1}^*(x)] = O(n^{-3/4})$ if f is continuous and f' is not and $MISE [f_{n_2}^*(x)] = O(n^{-1/2})$ if f and f' are not continuous.

Introduction:

Consider a sequence X_1, X_2, \dots, X_n of n independent, identically distributed random variables (i.i.d.) with unknown density function (d.f.) $f(x)$. A class of density estimators, known as Kernel estimators, which has been widely studied by Parzen [1] and Rosenblatt [2] may be defined as follows:

$$\hat{f}_n(x) = (nh)^{-1} \sum_{j=1}^n K\left(\frac{x - X_j}{h}\right),$$

Parzen has shown that if the Kernel function $K(\cdot)$ satisfies certain conditions, including $\int_{-\infty}^{\infty} K(y)dy = 1$ and $h = h(n)$ is a bounded and nonincreasing function of n such that $h(n) = O(n^{-1})$, then $\hat{f}_n(x)$ is a mean squared error consistent estimator of $f(x)$ at all continuity points x of f .

E. J. Wegman and H. I. Davies [3] considered a modified estimator of the form

$$f_n^*(x) = (n^2 h_n)^{-1/2} \sum_{j=1}^n (h_j)^{-1/2} K\left(\frac{x - X_j}{h_j}\right). \quad (1)$$

They established laws of the iterated logarithm for density estimators as well as the asymptotic distribution results using the almost sure invariance principle.

In this article we shall discuss classes of estimators of the form $f_n^*(x)$ for an unknown d.f. $f(x)$ which with its first derivative are not necessarily continuous. As a measure of improvement, many authors use the mean integrated square error given by

$$MISE [f_n^*(x)] = E \int_{-\infty}^{\infty} [f_n^*(x) - f(x)]^2 dx \quad (2)$$

Let

$$M_1ISE [f_n^*(x)] = E \int_{-\infty}^{\infty} [f_n^*(x) - B(f(x))]^2 dx; \quad (3)$$

where

$$B(f(x)) = E [f_n^*(x)] - f(x).$$

Hence

$$MISE [f_n^*(x)] = M_1ISE [f_n^*(x)] + (1 - B)^2 \|f(x)\|^2.$$

Consider a recursive Kernel estimator of the form (1) of f based on the random sample. Assuming $K(\cdot)$ is a some suitable Kernel function satisfying the following conditions:

$$(A) \begin{cases} \sup_x |K(x)| < \infty, & \lim_{x \rightarrow \pm\infty} |xK(x)| = 0, \\ \int_{-\infty}^{\infty} |K(x)| dx < \infty \text{ and } \int_{-\infty}^{\infty} x^2 K(x) dx < \infty, \end{cases}$$

and $h(n)$ is a sequence of nonincreasing positive constants converging to zero as n tends to ∞ .

As symbols, we shall use $f_{n_0}^*(x)$ as an estimator for f when f and f' are continuous, if f is continuous but f' is not, then we take $f_{n_1}^*(x)$ as an estimator for f and $f_{n_2}^*(x)$ as an estimator for f when f and f' are not continuous.

Main results:

Theorem 1. *Assum that $K(\cdot)$ is a Borel function satisfies conditions (A) and $nh_n \rightarrow \infty$ as $n \rightarrow \infty$, if $f(x)$, $f'(x)$ are continous and f'' is square integrable, then*

$$MISE [f_{n_0}^*(x)] = C_1 \{ (nh_n)^{-1} \|K(t)\|^2 + 0((nh_n)^{-1}) + [1 - (n^2 h_n)^{-1/2} r_n]^2 \\ \times \|f(x)\|^2 + \frac{1}{2} (nh_n)^{-1} \left[\sum_{j=1}^n h_j^5 \|tK^{1/2}(t)\|^4 \|f''(x)\|^2 + 0(h_j^5) \right] \};$$

where $0 < c_1 < 1$, $\|g(t)\|^2 = \int g^2(t) dt$

and

$$r_n = \sum_{j=1}^n h_j^{1/2}.$$

If $(n^2 h_n)^{-1/2} r_n \rightarrow B < \infty$ and $(nh_n)^{-1} \sum_{j=1}^n h_j^5 \rightarrow 0$ as $n \rightarrow \infty$,

then

$$MISE [f_n^*(x)] \rightarrow (1 - B)^2 \|f(x)\|^2 \text{ as } n \rightarrow \infty,$$

and hence

$$M_1ISE [f_n^*(x)] \rightarrow 0 \text{ as } n \rightarrow \infty.$$

In particular if $h_j = 0 (n^{-r})$, $j = 1, 2, \dots, n$; $r > 0$, then

$$M_1ISE [f_n^*(x)] = 0(n^{r-1}) + 0(n^{-4r}),$$

and hence the asymptotic optimum value r_0 of r is $\frac{1}{5}$ and then we get minimum $M_1ISE [f_{n_0}^*(x)] = 0(n^{-4/5})$.

Theorem 2. Suppose $K(\cdot)$ and h_n satisfy conditions of theorem 1.

If $f(x)$ is continuous, f' has finite points of discontinuity $(b_1, b_2, \dots, b_\ell)$ and f'' is square integrable, then

$$M_1ISE [f_{n_1}^*(x)] = C_2 \{ M_1ISE [f_{n_0}^*(x)] + \frac{4\Delta}{n} (\sum_{j=1}^n h_j^3) \int_{-\infty}^0 [\int_{-\infty}^y (y-t)K(t)dt]^2 dy + O(\frac{1}{n} \sum_{j=1}^n h_j^3) \},$$

where $0 < c_2 < 1$, $\Delta = \sum_{j=1}^{\ell} \Delta_j^2$,

$$\Delta_j = f'(b_j^-) - f'(b_j^+), j = 1, 2, \dots, n,$$

If $\frac{1}{n} \sum_{j=1}^n h_j^3 \rightarrow 0$ as $n \rightarrow \infty$, then as $n \rightarrow \infty$, the estimation $f_{n_1}^*(x)$ is consistent in M_1ISE sense, specially if $h_j = 0(n^{-r})$, $j = 1, 2, \dots, n$, $r > 0$, then $M_1ISE [f_{n_1}^*(x)] = 0(n^{r-1}) + 0(n^{-3r})$ and we get the asymptotic optimum value r_0 of r is $\frac{1}{4}$, and hence minimum $M_1ISE [f_{n_1}^*(x)] = 0(n^{-3/4})$.

Theorem 3. Let $K(\cdot)$ and h_n satisfy conditions of theorem 1.

If f has $k(k \geq 0)$ points of discontinuity, $-\infty = a_0 < a_1 < a_2 < \dots < a_k < a_{k+1} = \infty$, and for each $i = 1, \dots, k + 1$, f' has $\ell_i(\ell_i \geq 0)$ points of discontinuity.

$$a_{i-1} = b_{i_0} < b_{i_1} < b_{i_2} < \dots < b_{i_{\ell_i}} < b_{i_{\ell_i+1}} = a_i, \text{ also}$$

$$\int_{-\infty}^{\infty} |f'(x)| dx < \infty, \quad \int_{-\infty}^{\infty} |f''(x)|^2 dx < \infty$$

then

$$M_1ISE [f_n^*(x)] < c_2 M_1ISE [f_{n_1}^*(x)] + \frac{4\delta}{n} (\sum_{j=1}^n h_j) \int_0^{\infty} | \int_y^{\infty} K(x)dx |^2 dy + O(\frac{1}{n} \sum_{j=1}^n h_j),$$

where $\delta = \sum_{i=1}^k \delta_i^2$, $\delta_i = f(a_i^-) - f(a_i^+)$,

$$\Delta = \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_i+1} \Delta_{ij}^2 \quad \text{and} \quad \Delta_{ij} = f'(b_{ij}^-) - f'(b_{ij}^+),$$

$$i = 1, \dots, k+1 \quad \text{and} \quad j = 1, \dots, \ell_i + 1.$$

If $\frac{1}{n} \sum_{j=1}^n h_j \rightarrow 0$ as $n \rightarrow \infty$, then the estimation $f_{n_2}^*(x)$ is consistent in M_1 ISE sense. In particular if $h_j = O(n^{-r})$, $j = 1, 2, \dots, n$, $r > 0$, then M_1 ISE $[f_{n_2}^*(x)] = O(n^{r-1}) + O(n^{-r})$ and also we get $r_0 = \frac{1}{2}$ and minimum M_1 ISE $[f_{n_2}^*(x)] = O(n^{-1/2})$.

Proof: It is sufficient to prove theorem 3 which implies theorem 1 and 2. We can see that

$$\begin{aligned} MISE [f_{n_2}^*(x)] &= E \int_{-\infty}^{\infty} [f_{n_2}^*(x) - f(x)]^2 dx \\ &= E \int_{-\infty}^{\infty} [(n^2 h_n)^{-1/2} \sum_{j=1}^n h_j^{-1/2} K(\frac{x - X_j}{h_j}) - f(x)]^2 dx \\ &= \int_{-\infty}^{\infty} [(h^2 h_n)^{-1} \sum_{j=1}^n h_j^{-1} EK^2(\frac{x - X_j}{h_j}) \\ &\quad + 2(n^2 h_n)^{-1} \sum_{1=i < j}^n (h_i h_j)^{-1/2} EK(\frac{x - X_i}{h_i}) K(\frac{x - X_j}{h_j}) \\ &\quad - 2f(x)(n^2 h_n)^{-1/2} \sum_{j=1}^n (h_j)^{-1/2} EK(\frac{x - X_j}{h_j}) + f^2(x)] dx. \end{aligned} \quad (3)$$

First

$$\begin{aligned} \int_{-\infty}^{\infty} h_j^{-1} EK^2(\frac{x - X_j}{h_j}) dx &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} h_j^{-1} K^2(\frac{x - y}{h_j}) f(y) dy dx \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} K^2(t) f(x - th_j) dt dx \\ &= \int_{-\infty}^{\infty} K^2(t) dt + O(1). \end{aligned} \quad (4)$$

Second

$$\int_{-\infty}^{\infty} h_j^{-1} f(x) EK(\frac{x - X_j}{h_j}) dx = \int_{-\infty}^{\infty} f(x) \int_{-\infty}^{\infty} K(t) f(x - th_j) dt dx. \quad (5)$$

At last

$$\begin{aligned}
 & \sum_{1=i < j}^n \int_{-\infty}^{\infty} (h_i h_j)^{-1/2} EK\left(\frac{x - X_i}{h_i}\right) EK\left(\frac{x - X_j}{h_j}\right) dx \\
 &= \sum_{1=i < j}^n (h_i h_j)^{1/2} \left\{ \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} K(t)K(s)[f(x - th_i) - f(x)] \right. \\
 & \quad \times [f(x - sh_j) - f(x)] dt ds dx + \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} K(t)K(s)f(x)[f(x - th_i) \\
 & \quad \left. + f(x - sh_j)] dt ds dx - \int_{-\infty}^{\infty} f^2(x) dx \right\}. \tag{6}
 \end{aligned}$$

It is easy to see that the second term in the last equation is equal to

$$r_n \sum_{j=1}^n h_j^{1/2} \int_{-\infty}^{\infty} f(x) \int_{-\infty}^{\infty} K(t)f(x - th_j) dt dx; \quad r_j = \sum_{j=1}^n h_j^{1/2}.$$

from (3), (4), (5) and (6) we get

$$\begin{aligned}
 MISE [f_{n_2}^*(x)] &= (n h_n)^{-1} \int_{-\infty}^{\infty} K^2(t) dt \\
 &+ 2(n^2 h_n)^{-1} \sum_{1=i < j}^n (h_i h_j)^{1/2} \int_{-\infty}^{\infty} J(x, h_i) J(x, h_j) dx \\
 &+ 2r_n(n^2 h_n)^{-1} \sum_{j=1}^n h_j^{1/2} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} K(t)f(x)f(x - th_j) dt dx \\
 &- (n^2 h_n)^{-1} r_n^2 \int_{-\infty}^{\infty} f^2(x) dx - 2(n^2 h_n)^{-1/2} \sum_{j=1}^n h_j^{1/2} \int_{-\infty}^{\infty} f(x) J(x, h_j) dx \\
 &- 2(n^2 h_n)^{-1/2} r_n \int_{-\infty}^{\infty} f^2(x) dx + \int_{-\infty}^{\infty} f^2(x) dx + o((n h_n)^{-1}),
 \end{aligned}$$

where $J(x, h) = \int_{-\infty}^{\infty} K(t)[f(x - th) - f(x)] dt$. Then

$$\begin{aligned}
 MISE [f_{n_2}^*(x)] &= (n h_n)^{-1} \int_{-\infty}^{\infty} K^2(t) dt \\
 &+ 2(n^2 h_n)^{-1} \sum_{1=i < j}^n (h_i h_j)^{-1/2} \int_{-\infty}^{\infty} J(x, h_i) J(x, h_j) dx \\
 &- 2(n^2 h_n)^{-1/2} [1 - r_n(n^2 h_n)^{-1/2}] \sum_{j=1}^n h_j^{1/2} \int_{-\infty}^{\infty} f(x) J(x, h_j) dx \\
 &+ [1 - (n^2 h_n)^{-1/2} r_n]^2 \int_{-\infty}^{\infty} f^2(x) dx + o((n h_n)^{-1}).
 \end{aligned}$$

By using Hölder's inequality we get

$$\begin{aligned} MISE [f_{n_2}^*(x)] &\leq (n h_n)^{-1} \|K(t)\|^2 + [1 - (n^2 h_n)^{-1/2} r_n]^2 \|f(x)\|^2 \\ &+ 2(n^2 h_n)^{-1} \sum_{1=i < j}^n (h_i h_j)^{+1/2} \|J(x, h_i)\| \|J(x, h_j)\| + O((n h_n)^{-1}). \end{aligned} \quad (7)$$

By using Taylor series-like expansions for f satisfying conditions of our theorem

- a) For $b_{i j_1 - 1} < x < b_{i j_1}$, $b_{i j_2 - 1} < x - th < b_{i j_2}$, $1 \leq j_1 \leq l_i + 1$, $1 \leq j_2 \leq l_i + 1$, $i = 1, 2, \dots, k + 1$

$$\begin{aligned} &f(x - th) - f(x) + th f'(x) - h^2 \int_0^t (t - s) f''(x - sh) ds \\ &= \begin{cases} 0 & , \quad j_1 = j_2 \\ \sum_{r=j_1}^{j_2-1} (b_{i r} - x + th) \Delta_{i r} & , \quad j_1 < j_2 \\ \sum_{r=j_2}^{j_1-1} (b_{i r} - x + th) \Delta_{i r} & , \quad j_1 > j_2 \end{cases} \end{aligned}$$

- b) For $b_{i_1 j_1 - 1} < x < b_{i_1 j_1}$, $b_{i_2 j_2 - 1} < x - th < b_{i_2 j_2}$, $1 \leq j_1 \leq l_{i_1} + 1$, $1 \leq j_2 \leq l_{i_2} + 1$, $1 \leq i_1 \leq k + 1$, $1 \leq i_2 \leq k + 1$, $i_1 \neq i_2$,

$$\begin{aligned} &f(x - th) - f(x) + th f'(x) - h^2 \int_0^t (t - s) f''(x - sh) ds \\ &= \begin{cases} \text{(i)} & \sum_{r=j_1}^{l_{i_1}+1} (b_{i_1 r} - x + th) \Delta_{i_1 r} + \sum_{r=i_1+1}^{i_2-1} \sum_{r'=1}^{l_{i_2}+1} (b_{r r'} - x + th) \Delta_{r r'}, \\ \text{(ii)} & + \sum_{r=j_2}^{j_2-1} (b_{i_2 r} - x - th) \Delta_{i_2 r} - \sum_{\mu=i_2}^{i_2-1} \delta_\mu \text{ if } i_1 < i_2 \\ \text{(iii)} & - \sum_{r=j_2}^{l_{i_2}+1} (b_{i_2 r} - x + th) \Delta_{i_2 r} - \sum_{r=i_2+1}^{i_1-1} \sum_{r'=1}^{l_{i_1}+1} (b_{r r'} - x + th) \Delta_{r r'} \\ & - \sum_{r=1}^{j_1-1} (b_{i_1 r} - x + th) \Delta_{i_1 r} + \sum_{\mu=i_2}^{i_1-1} \delta_\mu \text{ if } i_1 > i_2 \end{cases} \end{aligned}$$

Let

$$G(x, h) = h^2 \int_{-\infty}^{\infty} K(t) \int_0^t (t - s) f''(x - sh) ds dt, \quad -\infty < x < \infty,$$

$$H_{ij1}(x, h) = \sum_{r=j}^{l_i+1} \Delta_{ir} \int_{-\infty}^{\frac{x-b_{ir}}{h}} (b_{ir} - x + th)K(t)dt$$

$$+ \sum_{r=i+1}^{k+1} \sum_{r'=1}^{l_{i+1}} \Delta_{rr'} \int_{-\infty}^{\frac{x-b_{rr'}}{h}} (b_{rr'} - x + th)K(t)dt,$$

$$H_{ij2}(x, h) = - \sum_{r=1}^{j-1} \Delta_{ir} \int_{\frac{x-b_{ir}}{h}}^{\infty} (b_{ir} - x + th)K(t)dt$$

$$- \sum_{r=1}^{i-1} \sum_{r'=1}^{l_{i+1}} \Delta_{rr'} \int_{\frac{x-b_{rr'}}{h}}^{\infty} (b_{rr'} - x + th)K(t),$$

$$b_{ij-1} < x < b_{ij}; j = 1, 2, \dots, l_{i+1}, i = 1, 2, \dots, k + 1,$$

$$V_{i1}(x, h) = \sum_{\mu=1}^{i-1} \delta_{\mu} \{1 - W(\frac{x - a_{\mu}}{h})\}, a_{i-1} < x < a_i, i = 1, \dots, k + 1,$$

$$V_{i2}(x, h) = - \sum_{\mu=i}^k \delta_{\mu} W(\frac{x - a_{\mu}}{h}), a_{i-1} < x < a_i, i = 1, \dots, k.$$

Where $W(y) = \int_{-\infty}^y K(t)dt, -\infty < y < \infty$

Further let, for $j = 1, \dots, l_i + 1, i = 1, \dots, k + 1$

$$H_{ij}(x, h) = H_{ij1}(x, h) + H_{ij2}(x, h) \text{ and}$$

$$V_i(x, h) = V_{i1}(x, h) + V_{i2}(x, h).$$

Lemma 1.

$$\lim_{h \rightarrow 0} \frac{1}{h^4} \int_{-\infty}^{\infty} G(x, h)dx = \frac{1}{4} (\int_{-\infty}^{\infty} t^2 K(t)dt)^2 \cdot \int_{-\infty}^{\infty} (f''(x))^2 dx.$$

Lemma 2.

$$\lim_{h \rightarrow 0} \frac{1}{h^3} \int_{b_{ij-1}}^{b_{ij}} H_{ij}^2(x, h)dx = (\Delta_{ij}^2 + \Delta_{ij-1}^2) \int_{-\infty}^0 \{ \int_{-\infty}^y (y-t)K(t)dt \}^2 dy,$$

$$1 \leq j \leq l_{i+1}, i = 1, \dots, k + 1.$$

Lemma 3.

$$\lim_{h \rightarrow 0} \frac{1}{h} \int_{a_{i-1}}^{a_i} V_i^2(x, h)dx = (\delta_i^2 + \delta_{i-1}^2) \int_0^{\infty} (1 - W(y))^2 dy, i = 1, \dots, k + 1.$$

Lemma 4.

$$\lim_{h \rightarrow 0} \frac{1}{h^3} \int_{b_{ij-1}}^{b_{ij}} G(x, h)H_{ij}(x, h)dx = \lim_{h \rightarrow 0} \frac{1}{h} \int_{a_{i-1}}^{a_i} G(x, h)V_i(x, h)dx$$

$$= \lim_{h \rightarrow 0} \frac{1}{h} \int_{b_{ij-1}}^{b_{ij}} H_{ij}(x, h)V_i(x, h)dx = 0, 1 \leq j \leq l_{i+1}, i = 1, \dots, k + 1$$

Proofs of these lemmas given by Eden, C.V. [4].

We can see that

$$\begin{aligned}
& \int_{-\infty}^{\infty} J^2(x, h) dx \\
&= \int_{-\infty}^{\infty} \left\{ \int_{-\infty}^{\infty} K(t)[f(x - th) - f(x)] dt \right\}^2 dx \\
&= \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} \{G(x, h) + H_{ij}(x, h) + V_i(x, h)\}^2 dx \\
&= \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} G^2(x, h) dx + \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} H_{ij}^2(x, h) dx \\
&\quad + \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} V_i^2(x, h) dx + 2 \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} G(x, h) H_{ij}(x, h) dx \\
&\quad + 2 \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} G(x, h) V_i(x, h) dx + 2 \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} H_{ij}(x, h) V_i(x, h) dx \\
&= \int_{-\infty}^{\infty} G^2(x, h) dx + \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} H_{ij}^2(x, h) dx \\
&\quad + \sum_{i=1}^{k+1} \int_{a_{i-1}}^{a_i} V_i^2(x, h) dx + 2 \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} G(x, h) H_{ij}(x, h) dx \\
&\quad + 2 \sum_{i=1}^{k+1} \int_{a_{i-1}}^{a_i} G(x, h) V_i(x, h) dx + 2 \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \int_{b_{ij-1}}^{b_{ij}} H_{ij}(x, h) V_i(x, h) dx.
\end{aligned}$$

Then from lemmas (1),(2),(3) and (4), we get

$$\begin{aligned}
\int_{-\infty}^{\infty} J^2(x, h) dx &= \frac{h^4}{4} \left(\int_{-\infty}^{\infty} t^2 K(t) dt \right)^2 \int_{-\infty}^{\infty} (f''(x))^2 dx + 0(h^4) \\
&\quad + h^3 \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} (\Delta_{ij}^2 + \Delta_{ij-1}^2) \int_{-\infty}^0 \left\{ \int_{-\infty}^y (y-t) K(t) dt \right\}^2 dy \\
&\quad + 0(h^3) + h \sum_{i=1}^{k+1} (\delta_i^2 + \delta_{i-1}^2) \int_0^{\infty} (1 - W(y))^2 dy + 0(h).
\end{aligned}$$

But from the fact that

$$\sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \Delta_{ij-1}^2 = \sum_{i=1}^{k+1} \sum_{j=1}^{\ell_{i+1}} \Delta_{ij}^2 = \Delta,$$

and

$$\sum_{i=1}^{k+1} \delta_{i-1}^2 = \sum_{i=1}^{k+1} \delta_i^2 = \delta,$$

We have

$$\begin{aligned} \int_{-\infty}^{\infty} J^2(x, h_i) dx &= \frac{1}{4} h_j^4 \| tK^{1/2}(t) \|^4 \cdot \| f''(x) \|^2 \\ &+ 0(h_j^4) + 2h_j^3 \Delta \int_{-\infty}^0 \left(\int_{-\infty}^y (y-t)K(t)dt \right)^2 dy + 0(h_j^3) \\ &+ 2h_j \delta \int_0^{\infty} (1 - W(y))^2 dy + 0(h_j). \end{aligned} \tag{8}$$

From (7) and (8) we get

$$\begin{aligned} MISE [f_{n_2}^*(x)] &\leq (nh_n)^{-1} \| K(t) \|^2 + 0(nh_n)^{-1} + [1 - (n^2 h_n)^{-1/2} r_n]^2 \| f(x) \|^2 \\ &+ 2(nh_n)^{-1} \left[\frac{1}{4} \| tK^{1/2}(t) \|^4 \cdot \| f'' \|^2 \left(\sum_{j=1}^n h_j^5 \right) + 0(h_j^5) \right] \\ &+ 2\Delta \left(\sum_{j=1}^n h_j^4 \right) \int_{-\infty}^0 \left\{ \int_{-\infty}^y (y-t)K(t)dt \right\}^2 dy \\ &+ 0(h_j^4) + 2\delta \int_0^{\infty} \{1 - W(y)\}^2 dy \left(\sum_{j=1}^n h_j^2 \right) + 0(h_j^2). \end{aligned}$$

Which completes the proof of the theorem.

References

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