

Mathematics

On Deheuvels' Nonparametric Estimation of Distribution Density

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ABSTRACT. We establish the limiting distribution of an integral quadratic deviation of Deheuvels' nonparametric estimation of a multidimensional distribution density. © 2007 Bull. Georg. Natl. Acad. Sci.

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1. Let X_1, X_2, \dots, X_n be a sequence of independent, equally distributed random values in a p -dimensional Euclidean space R_p , $p \geq 1$, that have the distribution density $f(x)$, $x = (x_1, x_2, \dots, x_p)$. As is known, Rosenblatt [1] and Parzen [2] gave the following definition of the nonparametric kernel estimation of the distribution density $f(x)$:

$$f_n^{(RP)}(x) = \frac{1}{nh_n^p} \sum_{i=1}^n K\left(\frac{x - X_i}{h_n}\right),$$

where $K(x)$ is a given kernel, while $\{h_n\}$ is a sequence of positive integers monotonically converging to zero.

Deheuvels [3] proposed a recurrent estimation of the density $f(x)$ of the following form:

$$f_n(x) = \frac{1}{H_n} \sum_{i=1}^n K\left(\frac{x - X_i}{h_i}\right) = \frac{H_{n-1}}{H_n} f_{n-1}(x) + \frac{1}{H_n} K\left(\frac{x - X_n}{h_n}\right),$$
$$H_n = \sum_{j=1}^n h_j.$$

The recurrent definition of estimations $f_n(x)$ of the probability density $f(x)$ has two advantages: a) there is no need to memorize data, i.e. if the estimation $f_{n-1}(x)$ is known, then to calculate $f_n(x)$ we use only the last observation $f(x)$; b) Deheuvels [3] shows that the asymptotic dispersion of the estimation $f_n(x)$ is less than that of estimations $f_n^{(RP)}(x)$. He also obtained asymptotically optimal values of h_j for the criterion of an integral mean-square error.

The aim of this paper is to obtain a limit distribution of the quadratic deviation $\int (f_n(x) - f(x))^2 r(x) dx$, where

$r(x)$ is a weight function which we assume to be piecewise-continuous and bounded. To obtain our results we use the functional limit theorem for a sequence of semimartingales [4], which was previously also used in [5] to solve an analogous problem for the estimation $f_n^{(RP)}(x)$.

2. Let us formulate the assumptions with regard to $f(x)$ and $K(x)$ to be used below in investigating the problem posed.

1°. The kernel

$$K(x) = \prod_{j=1}^p K_j(x), \quad x = (x_1, \dots, x_p),$$

and each of the kernels $K_j(u)$, $u \in R_1$, possesses the following properties:

$$\begin{aligned} 0 \leq K_j(u) \leq c < \infty, \quad K_j(-u) = K_j(u), \quad u^2 K_j(u) \in L_1(R_1), \\ \int K_j(u) du = 1, \quad K_j^0(ux) \geq K_j^0(x) \quad \text{for all } u \in [0,1] \text{ and all } x \in R_1, \end{aligned}$$

where $K_j^0 = K_j * K_j$, $*$ being a convolution operator.

2°. The distribution density $f(x)$ is bounded and has bounded partial derivatives up to the second order.

Notation.

$$\begin{aligned} L_n &= nh_n^p \int (f_n(x) - Ef_n(x))^2 r(x) dx, \\ U_n &= nh_n^p \int (f_n(x) - f(x))^2 r(x) dx, \\ \alpha_i(x, y) &= K\left(\frac{x - X_i}{h_i}\right) - EK\left(\frac{x - X_i}{h_i}\right), \\ b_n^2 &= 4n^2 h_n^{2p} H_n^{-4} \sum_{i=2}^n \sum_{j=1}^{i-1} E \left(\int \alpha_i(x, X_i) \alpha_j(x, X_j) r(x) dx \right)^2, \\ d_n^2 &= 2n^2 h_n^{2p} H_n^{-4} \iint f^2(x) \left(\sum_{j=1}^n h_j^p K_0\left(\frac{x-y}{h_j}\right) \right)^2 r(x) r(y) dx dy, \\ \Theta_n^{(1)} &= nh_n^p \int (Ef_n(x) - f(x))^2 r(x) dx, \\ \Theta_n^{(2)} &= nh_n^p H_n^{-2} \sum_{k=1}^n \iint K^2\left(\frac{x-u}{h_k}\right) f(u) r(x) du dx, \\ K_0 &= K * K, \quad \gamma_s(n) = \frac{1}{n} \sum_{i=1}^n h_i^{ps}, \quad s = 1, 2, \dots \end{aligned}$$

Lemma 1. Let condition 1° be fulfilled, $f(x)$ and $r(x) \geq 0$ be piecewise-continuous and bounded functions. If $h_n \rightarrow 0$, $nh_n^p \rightarrow \infty$ and $h_n^{ps} / \gamma_s(n) \rightarrow \gamma_s$, $s = 1, 2$, $0 < \gamma_2 \leq \gamma_1 \leq 1$ as $n \rightarrow \infty$, then

$$b_n^2 = d_n^2 + \gamma_1(n)o(1) + O(1) + O\left(\frac{\gamma_1(n)}{n}\right)$$

and also

$$2\gamma_1^2 \int f^2(x)r^2(x)dx \int K_0^2(u)du \leq \\ \leq \liminf_{n \rightarrow \infty} h_n^{-p} d_n \leq \overline{\lim}_{n \rightarrow \infty} h_n^{-p} d_n \leq 2 \frac{\gamma_1^3}{\gamma_2} \int f^2(x)r^2(x)dx \int K_0^2(u)du.$$

Theorem 1. *Let the conditions of Lemma 1 be fulfilled. Then*

$$b_n^{-1}(L_n - EL_n) \xrightarrow{d} N(0,1),$$

where d is the convergence in distribution, and $N(0,1)$ is a random value having a normal distribution with zero mean value and dispersion 1.

Corollary. $d_n^{-1}(L_n - EL_n) \xrightarrow{d} N(0,1)$ as $n \rightarrow \infty$.

Lemma 2. *Let $K(x)$ and $f(x)$ satisfy conditions 1⁰ and 2⁰, respectively; $r(x) \geq 0$ be a bounded and piecewise-continuous function. If*

$$n^{-1}h_n^{-p} \left(\sum_{j=1}^n h_j^{p+2} \right)^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

then

$$h_n^{-p/2} (U_n - L_n - \Theta_n^{(1)}) = o_p(1).$$

Theorem 2. a) *Let $K(x)$, $f(x)$ and $r(x)$ satisfy the conditions of Lemma 2. If*

$$nh_n^p \rightarrow \infty, \quad \frac{h_n^{ps}}{\gamma_s(n)} \rightarrow \gamma_s, \quad s=1,2, \quad n^{-1}h_n^{-p} \left(\sum_{j=1}^n h_j^{p+2} \right)^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

then

$$h_n^{-p/2} \sigma_n^{-1} (U_n - \Theta_n^{(1)} - \Theta_n^{(2)}) \xrightarrow{d} N(0,1),$$

where $\sigma_n^2 = h_n^{-p} d_n^2$.

b) *Let the following condition be added to the conditions with regard to $f(x)$ and all second order partial derivatives of the function $f(x)$ belong to $L_2(R_p)$. If*

$$nh_n^p \rightarrow \infty, \quad \frac{h_n^{ps}}{\gamma_s(n)} \rightarrow \gamma_s, \quad s=1,2, \quad n^{-1}h_n^{-\frac{3p}{2}} \left(\sum_{j=1}^n h_j^{p+2} \right)^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

then

$$h_n^{-p/2} \sigma_n^{-1} (U_n - \Theta_n^{(2)}) \xrightarrow{d} N(0,1).$$

c) *Let $K(x)$, $f(x)$ and $r(x)$ satisfy the conditions of Lemma 2 and, in addition to this, the second order partial derivatives of the function $f(x)$ belong to $L_1(R_p)$. If*

$$nh_n^p \rightarrow \infty, \quad \frac{h_n^{sp}}{\gamma_s(n)} \rightarrow \gamma_s, \quad s=1,2,$$

also

$$\frac{h_n^p}{\gamma_1(n)} = \gamma_1 + o(h_n^{p/2}) \quad \text{and} \quad n^{-1} h_n^{-\frac{3p}{2}} \left(\sum_{j=1}^n h_j^{p+2} \right)^s \rightarrow 0, \quad s=1,2 \quad \text{as } n \rightarrow \infty, \quad 0 < p \leq 3,$$

then

$$h_n^{-p/2} \sigma_n^{-1} (U_n - \Theta) \xrightarrow{d} N(0,1),$$

$$\Theta = \gamma_1 \int f(x) r(x) dx \int K^2(u) du.$$

Corollary. If $h_k = h_n$, $k=1, \dots, n$, then from Theorem 2 we obtain the well-known result on the limit distribution of an integral quadratic deviation of the estimation $f_n^{(RP)}(x)$ [6-8].

The conditions c) are fulfilled, for instance, if we assume $h_k = k^{-\alpha}$ for $\frac{2}{p+8} < \alpha < \frac{1}{p+2}$, $\frac{1}{p+2} < \alpha < \frac{2}{3p}$,

$p < 4$, noting that $\gamma_1 = 1 - \alpha p > 0$.

Remark. Theorems 1 and 2 define more precisely the results obtained in [9]. A more precise definition of these results consists in the following: in the present paper there is no integrability of the weight function $r(x)$; the conditions on h_k , $k=1,2, \dots$, used in [10] are replaced by less restrictive ones.

მათემატიკა

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