

*Mathematics*

## On the Wolverton-Wagner Estimate of a Distribution Density

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**ABSTRACT.** The result of the work consists mainly in obtaining the limit distribution of an integral quadratic deviation of the Wolverton-Wagner nonparametric estimate of a multidimensional distribution density. © 2007 Bull. Georg. Natl. Acad. Sci.

**Key words:** nonparametric estimate, recursive estimate, convergence in distribution.

1. Let  $X_1, X_2, \dots, X_n$  be a sequence of independent, equally distributed random variables with values in a Euclidean  $p$ -dimensional space  $R_p$ ,  $p \geq 1$ , whose distribution density is  $f(x)$ ,  $x = (x_1, \dots, x_p)$ . As is known, Rosenblatt [1] and Parzen [2] gave the following definition of an empirical kernel density  $f_n^{(RP)}(x)$ , which is based on the sampling  $X_1, X_2, \dots, X_n$ :

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

where  $K(x)$  is a given kernel and  $\{a_n\}$  is a sequence of positive integers converging monotonically to infinity.

Wolverton and Wagner [3] introduce the following definition of an empirical density  $f_n^{(W)}(x)$  that differs little from  $f_n^{(RP)}(x)$  but is recurrent:

$$f_n^{(W)}(x) = \frac{1}{n} \sum_{j=1}^n a_j^p K(a_j(x - X_j)) = \frac{n-1}{n} f_{n-1}^{(W)}(x) + \frac{a_n^p}{n} K(a_n(x - X_n)).$$

The recurrent definition of probability density estimates  $f_n^{(W)}(x)$  has two obvious advantages: 1) there is no need to memorize data, i.e. if the estimate  $f_{n-1}^{(W)}(x)$  is known, then  $f_n^{(W)}(x)$  can be calculated by means of the last observation  $X_n$  only, without using the sampling  $X_1, X_2, \dots, X_{n-1}$ ; 2) the asymptotic dispersion of the estimate  $f_n^{(W)}(x)$  does not exceed the dispersion of the estimate  $f_n^{(RP)}(x)$ .

The aim of this work is to study the asymptotics of a mean value of the integral of the squared error

$$E \int [f_n^{(W)}(x) - f(x)]^2 r(x) dx$$

and the limit distribution of the functional

$$\int [f_n^{(W)}(x) - f(x)]^2 r(x) dx.$$

**Theorem 1.** *Let the kernel*

$$K(x) = \prod_{j=1}^p K_j(x_j)$$

and each of the kernels  $K_j(x)$  possess the following properties:

$$0 \leq K_j(x) \leq c < \infty, \quad K_j(-u) = K_j(u), \quad u^2 K_j(u) \in L_1(-\infty, \infty)$$

and

$$\int K_j(u) du = 1.$$

Assume that the density  $f(x)$  and its partial derivatives up to second order are bounded and belong to  $L_2(R_p)$ . If

$$\frac{a_n^p}{n} \rightarrow 0, \quad \frac{\gamma_n}{a_n^p} \rightarrow \gamma > 0 \quad \text{and} \quad \sum_{k=1}^{\infty} a_k^{-2} = \infty,$$

then we have

$$E \int [f_n^{(W)}(x) - f(x)]^2 r(x) dx = \frac{\gamma_n}{n} \int f(x) r(x) dx \int K^2(x) du + \left( \frac{1}{n} \sum_{k=1}^n a_k^{-2} \right) \frac{1}{4} \int \left[ \sum_{j=1}^p \alpha_j \frac{\partial^{(2)} f(x)}{\partial x_j^2} \right]^2 r(x) dx + o \left( \frac{\gamma_n}{n} + \left( \frac{1}{n} \sum_{k=1}^n a_k^{-2} \right)^2 \right), \quad \gamma_n = \frac{1}{n} \sum_{k=1}^n a_k^p, \quad \alpha_j = \int x_j^2 K_j(x_j) dx_j,$$

where  $r(x)$  is a bounded and piecewise-continuous function.

2. Let us formulate the assumptions with regard to  $K(x)$  and  $f(x)$ ,  $x = (x_1, \dots, x_p)$ , used in studying the limit distribution of an integral standard deviation  $f_n^{(W)}(x)$  from  $f(x)$ .

1<sup>0</sup>. The kernel  $K(x)$  satisfies the conditions of Theorem 1 and, moreover,  $K_j^0(ux) \geq K_j^0(x)$  for all  $u \in [0, 1]$  and for all  $x \in R_1$ , where  $K_j^0 = K_j * K_j$ ,  $*$  being the convolution operator.

2<sup>0</sup>. The density  $f(x)$  is bounded and has bounded partial derivatives up to second order.

Notation.

$$L_n = n \int (f_n^{(W)}(x) - E f_n^{(W)}(x))^2 r(x) dx,$$

$$U_n = \frac{n}{a_n^p} \int (f_n^{(W)}(x) - f(x))^2 r(x) dx,$$

$$\alpha_i(x, y) = a_i^p [K(a_i(x - y)) - EK(a_i(x - X_1))],$$

$$b_n^2 = \frac{4}{n^2} \sum_{i=2}^n \sum_{j=1}^{i-1} E \left( \int \alpha_i(x, X_i) \alpha_j(x, X_j) r(x) dx \right)^2,$$

$$d_n^2 = \frac{2}{n^2} \iint f^2(x) \left( \sum_{i=1}^n a_i^p K_0(a_i(x-y)) \right)^2 r(x)r(y) dx dy,$$

$$K_0 = K * K, \quad \gamma_s(n) = \frac{1}{n} \sum_{i=1}^n a_i^{ps}, \quad s = 1, 2, \dots,$$

$$\Theta_n^{(1)} = na_n^{-p} \int (Ef_n^{(W)}(x) - f(x))^2 r(x) dx, \quad \tilde{L}_n = a^{-p} L_n,$$

$$\Theta_n^{(2)} = n^{-1} a_n^{-p} \sum_{k=1}^n a_k^{2p} \iint K^2(a_k(x-u)) f(u) r(x) du dx.$$

**Lemma 1.** Let conditions 1<sup>0</sup> be fulfilled,  $f(x)$  and  $r(x) \geq 0$  be piecewise-continuous and bounded functions. If

$a_n \rightarrow \infty$ ,  $\frac{a_n^p}{n} \rightarrow 0$  and  $\frac{\gamma_s(n)}{a_n^{ps}} \rightarrow \gamma_s$ ,  $s = 1, 2$  ( $0 < \gamma_2 \leq \gamma_1 \leq 1$ ) 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

$$2\gamma_1^2 \leq \int f^2(x)r^2(x)dx \int K_0^2(u)du \leq \lim_{n \rightarrow \infty} \frac{d_n^2}{a_n^p} \leq \lim_{n \rightarrow \infty} \frac{d_n^2}{a_n^p} \leq 2\gamma_1 \int f^2(x)r^2(x)dx \int K_0^2(u)du.$$

**Theorem 2.** Let all conditions of Lemma 1 be fulfilled. Then

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

where  $d$  denotes convergence in distribution,  $N(0,1)$  is a random value having normal distribution with mean 0 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 1<sup>0</sup> and 2<sup>0</sup>, respectively,  $r(x) \geq 0$  be bounded and piecewise-continuous. If

$$n^{-1} a_n^{-p} \left( \sum_{k=1}^n a_k^{-2} \right)^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

then

$$a_n^{p/2} (U_n - \tilde{L}_n - \Theta_n^{(1)}) = o_p(1).$$

**Lemma 3.**  $E\tilde{L}_n = \Theta_n^{(2)} + O(a_n^{-p})$ .

**Theorem 3.** a) Let  $K(x)$ ,  $f(x)$  and  $r(x)$  satisfy the conditions of Lemma 2. If

$$a_n \uparrow \infty, \quad \frac{a_n^p}{n} \rightarrow 0, \quad \frac{\gamma_s(n)}{a_n^{sp}} \rightarrow \gamma_s, \quad s = 1, 2,$$

$$n^{-1} a_n^{-p} \left( \sum_{k=1}^n a_k^2 \right)^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

then

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

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

b) To the conditions with regard to  $f(x)$  we add the assumption: all partial derivatives of second order of the function  $f(x)$  belong to  $L_2(R_p)$ . If

$$a_n \uparrow \infty, \quad \frac{a_n^p}{n} \rightarrow 0, \quad \frac{\gamma_s(n)}{a_n^{sp}} \rightarrow \gamma_s, \quad s = 1, 2,$$

$$\frac{1}{n\sqrt{a_n^p}} \left( \sum_{j=1}^n a_j^{-2} \right)^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

then

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

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

$$\frac{a_n^p}{n} \rightarrow 0, \quad \frac{\gamma_s(n)}{a_n^{sp}} \rightarrow \gamma_s, \quad s = 1, 2,$$

and also

$$\frac{\gamma_1(n)}{a_n^p} = \gamma_1 + o(a_n^{-p/2}), \quad (na_n^{p/2})^{-1} \sum_{j=1}^n a_j^{p-2} \rightarrow 0$$

and

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

then

$$a_n^{p/2} \sigma_n^{-1} (U_n - \Theta) \xrightarrow{d} N(0,1), \quad \Theta = \gamma_1 \int f(x)r(x)dx \int K^2(u)du.$$

**Corollary.** If  $a_k = a_n$ ,  $k = 1, 2, \dots$ , then from Theorem 3 we obtain the well-known result on the limit distribution of a standard deviation of the estimate  $f_n^{(RP)}(x)$  ([4, 5]).

**Remark.** Theorems 2 and 3 generalize and refine the results obtained by one of the authors in [6]. The refinement of these results is done by a method differing from the method used in [6] and consists in the following: in this paper there is no integrability of the weight function  $r(x)$ ; the conditions on  $a_k$ ,  $k = 1, 2, \dots$  used in [6] are replaced by less restrictive requirements.

მათემატიკა

## განაწილების სიმკვრივის ვოლვერტონ-ვაგნერის არაპარამეტრული შეფასების შესახებ

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