

## AN ORTHOGONAL SERIES ESTIMATE OF TIME-VARYING REGRESSION

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### Summary

Let  $(X_1, Y_1), (X_2, Y_2), \dots$  be independent pairs of random variables according to the model  $Y_n = t_n(X_n)R(X_n) + Z_n$ ,  $n=1, 2, \dots$ , where  $t_n$  and  $R$  are unknown functions.  $Z_n$ 's are i.i.d. random variables with zero mean and finite variance. The marginal density of  $X_n$  is independent of  $n$ . In the paper nonparametric estimates of a nonstationary regression function  $E\{Y_n | X_n = x\} = t_n(x)R(x)$  are proposed and their asymptotic properties are investigated.

### 1. Introduction

Let  $(X_1, Y_1), (X_2, Y_2), \dots$  be a sequence of independent pairs of random variables according to the model

$$(1) \quad Y_n = R_n(X_n) + Z_n, \quad n=1, 2, \dots,$$

where  $R_n$ 's are Borel-measurable functions and  $Z_n$ 's are i.i.d. random variables.  $Z_n$  is independent of  $X_n$  and

$$(2) \quad E Z_n = 0, \quad E Z_n^2 < \infty.$$

$Y_n$  takes values in  $R$ , while  $X_n$  in  $\mathcal{X}$ , where  $\mathcal{X}$  is a Borel subset of  $R^p$ . The marginal Lebesgue density  $f$  of  $X_n$  is independent of  $n$ . Our aim is to estimate the nonstationary regression function i.e. to track  $R_n(x) = E\{Y_n | X_n = x\}$ .

In stationary case several nonparametric methods have been proposed. We mention works of Nadaraya [13], Rosenblatt [15], Noda [14], Collomb [4], Grebicki and Krzyżak [10] as well as Devroye and Wagner [6] based on the Rosenblatt-Parzen density estimate. The nearest neighbor estimate is represented by Stone [18] and Devroye [5]. The orthog-

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onal series approach has been discussed by Mirzahmedov and Hašimov [12] and Greblicki [9].

The next section consists of assumptions and preliminaries. The main results of the paper i.e. Theorems 1 and 2 are given in Sections 3 and 4. Concluding Theorems 3 and 4 are in Section 5. In the closing section an example is considered in which restrictions made in this paper are satisfied even if the regression function converges to infinity as  $n$  tends to infinity.

## 2. Preliminaries and assumptions

Throughout this paper we assume that

$$(3) \quad R_n(x) = t_n(x)R(x) .$$

A sequence of functions  $\{t_n\}$  is unknown. It, however, becomes similar to some sequence of numbers; more precisely, there exists a sequence  $\{c_n\}$  such that

$$(4) \quad \sup_x |t_n(x) - c_n| \xrightarrow{n} 0 .$$

Further assumptions imposed on  $\{t_n\}$  will be given in the sequel. The functional form of  $R$  is completely unknown.

All integrals and supremums are taken over  $\mathcal{X}$ . Besides,  $K_1, K_2, \dots$  denote positive constants numbered in order of appearance.

Furthermore we assume that

$$(5) \quad \int R^2(x)f(x)dx < \infty ,$$

which implies

$$(6) \quad \int |R(x)|f(x)dx < \infty .$$

Note that from (1)–(5) it follows that

$$(7) \quad E Y_n^2 \leq K_1 + K_2 c_n^2 .$$

In the next parts of the paper we refer (7) rather than (1)–(5).

We also introduce the following notations:

$$h_n(x) = R_n(x)f(x) , \quad h(x) = R(x)f(x) .$$

We shall use a complete orthonormal system  $\{g_j\}$ ,  $j=0, 1, \dots$ , defined on  $\mathcal{X}$ , such that

$$(8) \quad |g_j(x)| \leq G_j$$

for all  $x \in \mathcal{X}$ , where  $\{G_j\}$  is a sequence of numbers.

From (4), (6) and (8) it follows that functions  $h_n$  can be expanded in the orthogonal series

$$(9) \quad h_n(x) \sim \sum_{j=0}^{\infty} a_{jn} g_j(x) .$$

It means that

$$(10) \quad a_{jn} = \int R_n(x) g_j(x) f(x) dx = E \{ Y_n g_j(X_n) \} .$$

In Section 5 unknown coefficients  $a_{jn}$ 's are estimated by the Robbins-Monro stochastic approximation method, see e.g. Wasan [21], i.e.

$$(11) \quad \hat{a}_{j,n+1} = \hat{a}_{jn} - \gamma_n (\hat{a}_{jn} - Y_{n+1} g_j(X_{n+1})) ,$$

where  $\hat{a}_{j0} = 0$  for all  $j$ , and  $\{\gamma_n\}$  is a sequence of positive numbers.

Let us expand  $f$  in the orthogonal series

$$(12) \quad f(x) \sim \sum_{j=0}^{\infty} b_j g_j(x) ,$$

where

$$(13) \quad b_j = \int g_j(x) f(x) dx = E g_j(X_1) .$$

Clearly

$$(14) \quad \hat{b}_{jn} = n^{-1} \sum_{i=1}^n g_j(X_i)$$

is an unbiased estimator of  $b_j$ .

As an estimator of  $R_n$  we take the statistics

$$(15) \quad \hat{R}_n(x) = \hat{h}_n(x) / \hat{f}_n(x) ,$$

where

$$(16) \quad \hat{h}_n(x) = \sum_{j=0}^{N(n)} \hat{a}_{jn} g_j(x) ,$$

$$(17) \quad \hat{f}_n(x) = \sum_{j=0}^{M(n)} \hat{b}_{jn} g_j(x) ,$$

and where  $\{N(n)\}$  and  $\{M(n)\}$  are sequences of integers.

It should be mentioned that estimator (17) of a density function was proposed by Čencov [2] and studied by Schwartz [17], Kronmal and Tarter [11] and Bosq [1] among others. For  $\gamma_n = 1/(n+1)$ ,  $\hat{a}_{jn}$  is equal to  $n^{-1} \sum_{i=1}^n Y_i g_j(X_i)$  and estimate (15) becomes that of studied by

Greblicki [9] for the stationary case.

In the paper we investigate asymptotic properties of (15), i.e. we show that, under suitable conditions,

$$|\hat{R}_n(x) - R_n(x)| \xrightarrow{n} 0$$

in probability and with probability one.

In order to prove convergence theorems we expand  $h$  in the orthogonal series

$$(18) \quad h(x) \sim \sum_{j=0}^{\infty} a_j g_j(x),$$

where

$$(19) \quad a_j = \int R(x) g_j(x) f(x) dx.$$

Finally, we define

$$(20) \quad d_n = \sup_{N_1, N_2} \left\{ \sum_{j=N_1}^{N_2} E(\hat{a}_{jn} - a_{jn})^2 \middle/ \sum_{j=N_1}^{N_2} G_j^2 \right\},$$

where  $N_1$  and  $N_2$  run over the set of all integers. In Section 5 it will be shown that  $\{d_n\}$  is bounded by a power sequence convergent to zero.

Herein we use the following two lemmas:

LEMMA A (Chung [3]). *Let  $p_1, p_2, \dots$  be real numbers such that for  $n \geq n_0$*

$$p_{n+1} \leq (1 - c/n^\omega) p_n + c'/n^t,$$

where  $0 < \omega < 1$ ,  $c > 0$ ,  $c' > 0$ ,  $t$  real. Then

$$\limsup_{n \rightarrow \infty} n^{t-\omega} p_n \leq c'/c.$$

LEMMA B (Van Ryzin [20]). *Let  $\{A_n\}$  and  $\{B_n\}$  be two sequences of random variables on a probability space  $(\Omega, F, P)$ . Let  $\{F_n\}$  be a sequence of Borel fields such that  $F_n \subset F_{n+1} \subset F$ , and let  $A_n$  and  $B_n$  be measurable with respect to  $F_n$ . If  $A_n \geq 0$  a.e.,  $E A_1$  is finite, and*

$$E \{A_{n+1} | F_n\} \leq A_n + B_n \quad \text{a.e.},$$

$$\sum_{n=1}^{\infty} E |B_n| < \infty,$$

then  $\{A_n\}$  converges almost surely to a finite limit as  $n$  tends to infinity.

### 3. Convergence in probability

First we state and prove two lemmas.

LEMMA 1. *If (6) is satisfied and*

$$(21) \quad d_n^{1/2} \sum_{j=0}^{N(n)} G_j^2 \xrightarrow{n} 0 ,$$

$$(22) \quad \sup_x |t_n(x) - c_n| \sum_{j=0}^{N(n)} G_j^2 \xrightarrow{n} 0 ,$$

then

$$(23) \quad E (\hat{h}_n(x) - h_n(x))^2 \xrightarrow{n} 0$$

at every point  $x \in \mathcal{X}$  at which

$$(24) \quad c_n \left[ \sum_{j=0}^{N(n)} a_j g_j(x) - h(x) \right] \xrightarrow{n} 0 .$$

PROOF. Observe

$$(25) \quad \begin{aligned} \hat{h}_n(x) - h_n(x) = & \sum_{j=0}^{N(n)} (\hat{a}_{jn} - a_{jn}) g_j(x) + \sum_{j=0}^{N(n)} (a_{jn} - c_n a_j) g_j(x) \\ & + c_n \left[ \sum_{j=0}^{N(n)} a_j g_j(x) - h(x) \right] + (c_n h(x) - h_n(x)) . \end{aligned}$$

By Cauchy's inequality, the expectation of the squared first term on the right-hand side in (25) is not greater than

$$(26) \quad \sum_{j=0}^{N(n)} E (\hat{a}_{jn} - a_{jn})^2 \sum_{j=0}^{N(n)} G_j^2 \leq d_n \left[ \sum_{j=0}^{N(n)} G_j^2 \right]^2 .$$

Since

$$\begin{aligned} |a_{jn} - c_n a_j| = & \left| \int (t_n(x) - c_n) R(x) g_j(x) f(x) dx \right| \\ \leq & G_j \sup_x |t_n(x) - c_n| \int |R(x)| f(x) dx , \end{aligned}$$

the absolute value of the second term in (25) does not exceed

$$(27) \quad (\sup_x |t_n(x) - c_n|) \sum_{j=0}^{N(n)} G_j^2 \int |R(x)| f(x) dx .$$

Moreover, the absolute value of the fourth term in (25) is majorized by

$$(28) \quad |R(x)| f(x) \sup_x |t_n(x) - c_n| .$$

In view of (25), (26), (27) and (28) the proof is complete.

LEMMA 2. *If*

$$c_n n^{-1/2} \sum_{j=0}^{M(n)} G_j^2 \xrightarrow{n} 0 ,$$

then

$$c_n^2 \mathbb{E} (\hat{f}_n(x) - f(x))^2 \xrightarrow{n} 0$$

at every point  $x \in \mathcal{X}$ , at which

$$c_n \left[ \sum_{j=0}^{M(n)} b_j g_j(x) - f(x) \right] \xrightarrow{n} 0 .$$

PROOF. Obviously

$$(29) \quad \hat{f}_n(x) - f(x) = \sum_{j=0}^{M(n)} (\hat{b}_{jn} - b_j) g_j(x) + \left[ \sum_{j=0}^{M(n)} b_j g_j(x) - f(x) \right] .$$

Since  $\mathbb{E} (\hat{b}_{jn} - b_j)^2 \leq G_j^2/n$ , the expectation of the squared first term in (29) is not greater than

$$\sum_{j=0}^{M(n)} \mathbb{E} (\hat{b}_{jn} - b_j)^2 \sum_{j=0}^{M(n)} G_j^2 \leq n^{-1} \left[ \sum_{j=0}^{M(n)} G_j^2 \right]^2 ,$$

which completes the proof.

Combining Lemmas 1 and 2, we get the main result of this section.

THEOREM 1. *Let (6), (21) and (22) be satisfied. Let, moreover,*

$$(30) \quad (|c_n| + 1) n^{-1/2} \sum_{j=0}^{M(n)} G_j^2 \xrightarrow{n} 0 .$$

Then

$$|\hat{R}_n(x) - R_n(x)| \xrightarrow{n} 0$$

in probability at every point  $x \in \mathcal{X}$  at which  $f(x) > 0$ , (24) holds and

$$(31) \quad (|c_n| + 1) \left[ \sum_{j=0}^{M(n)} b_j g_j(x) - f(x) \right] \xrightarrow{n} 0 .$$

PROOF. The result follows from the equality

$$(32) \quad \begin{aligned} \hat{R}_n(x) - R_n(x) &= (\hat{h}_n(x) - h_n(x)) / \hat{f}_n(x) \\ &\quad + h_n(x) (f(x) - \hat{f}_n(x)) / f(x) \hat{f}_n(x) . \end{aligned}$$

#### 4. Almost sure convergence

As in the previous section we start with two lemmas.

LEMMA 3. *Assume that conditions of Lemma 1 are satisfied. Furthermore, let*

$$(33) \quad \sum_{n=1}^{\infty} \gamma_n^2 (1 + c_n^2) \left[ \sum_{j=0}^{N(n)} G_j^2 \right]^2 < \infty ,$$

and let

$$(34) \quad \sum_{n=1}^{\infty} \gamma_n d_n \left[ \sum_{j=0}^{N(n)} G_j^2 \right]^2 < \infty ,$$

$$(35) \quad \sum_{n=1}^{\infty} d_n \sum_{j=0}^{N(n)} G_j^2 \sum_{k=N(n)}^{N(n+1)} G_k^2 < \infty .$$

Then

$$(36) \quad |\hat{h}_n(x) - h_n(x)| \xrightarrow{n} 0$$

with probability one, at every point  $x \in \mathcal{X}$  at which (24) holds.

PROOF. By (25), (27) and (28) it suffices to show that

$$(37) \quad \sum_{j=0}^{N(n)} (\hat{a}_{jn} - a_{jn}) g_j(x) = \sum_{j=0}^{N(n)} (\hat{a}_{jn} - E \hat{a}_{jn}) g_j(x) + \sum_{j=0}^{N(n)} (E \hat{a}_{jn} - a_{jn}) g_j(x)$$

converges to zero with probability one as  $n$  tends to infinity. Now we are concerned with the second term in (37). The absolute value of the term does not exceed

$$(38) \quad \left[ \sum_{j=0}^{N(n)} (E \hat{a}_{jn} - a_{jn})^2 \sum_{k=0}^{N(n)} G_k^2 \right]^{1/2} \leq \left[ \sum_{j=0}^{N(n)} E (\hat{a}_{jn} - a_{jn})^2 \sum_{k=0}^{N(n)} G_k^2 \right]^{1/2} \leq d_n^{1/2} \sum_{j=0}^{N(n)} G_j^2 .$$

By making use of Lemma B we prove the convergence of the first term in (37). Denote

$$V_n(x) = \sum_{j=0}^{N(n)} (\hat{a}_{jn} - E \hat{a}_{jn}) g_j(x) .$$

Observe

$$V_{n+1}(x) = V_n(x) + u_n(x) + w_n(x) ,$$

where

$$u_n(x) = \gamma_n \sum_{j=0}^{N(n+1)} [Y_{n+1} g_j(X_{n+1}) - E(Y_{n+1} g_j(X_{n+1}))] g_j(x) ,$$

$$w_n(x) = (1 - \gamma_n) \sum_{j=N(n)+1}^{N(n+1)} (\hat{a}_{jn} - E \hat{a}_{jn}) g_j(x) - \gamma_n \sum_{j=0}^{N(n)} (\hat{a}_{jn} - E \hat{a}_{jn}) g_j(x) .$$

Thus,

$$E(V_{n+1}^2(x) | X_1, Y_1, X_2, Y_2, \dots, X_n, Y_n) = V_n^2(x) + B_n(x) ,$$

where

$$B_n(x) = E u_n^2(x) + w_n^2(x) + 2V_n(x)w_n(x) .$$

Now it will be verified that  $\sum_{n=1}^{\infty} E|B_n(x)| < \infty$  for every  $x \in \mathcal{X}$ . By Cauchy's inequality and (7) we obtain

$$\begin{aligned}
 (39) \quad \mathbb{E} u_n^2(x) &\leq \gamma_n^2 \sum_{j=0}^{N(n+1)} \text{var} [Y_{n+1} g_j(X_{n+1})] \sum_{j=0}^{N(n+1)} G_j^2 \\
 &\leq \gamma_n^2 \mathbb{E} Y_{n+1}^2 \left[ \sum_{j=0}^{N(n+1)} G_j^2 \right]^2 \leq \gamma_n^2 (K_1 + K_2 c_{n+1}^2) \left[ \sum_{j=0}^{N(n+1)} G_j^2 \right]^2.
 \end{aligned}$$

Using Cauchy's inequality again we get

$$\begin{aligned}
 w_n^2(x) &\leq 2(1-\gamma_n)^2 \sum_{j=N(n)}^{N(n+1)} (\hat{a}_{jn} - \mathbb{E} \hat{a}_{jn})^2 \sum_{j=N(n)}^{N(n+1)} G_j^2 \\
 &\quad + 2\gamma_n^2 \sum_{j=0}^{N(n+1)} (\hat{a}_{jn} - \mathbb{E} \hat{a}_{jn})^2 \sum_{j=0}^{N(n+1)} G_j^2.
 \end{aligned}$$

Therefore

$$(40) \quad \mathbb{E} w_n^2(x) \leq 2(1-\gamma_n)^2 d_n \left[ \sum_{j=N(n)}^{N(n+1)} G_j^2 \right]^2 + 2\gamma_n^2 d_n \left[ \sum_{j=0}^{N(n+1)} G_j^2 \right]^2.$$

In turn

$$\begin{aligned}
 |V_n(x)w_n(x)| &\leq \gamma_n \left[ \sum_{j=0}^{N(n)} (\hat{a}_{jn} - \mathbb{E} \hat{a}_{jn}) g_j(x) \right]^2 \\
 &\quad + \left| (1-\gamma_n) \sum_{j=0}^{N(n)} (\hat{a}_{jn} - \mathbb{E} \hat{a}_{jn}) g_j(x) \sum_{j=N(n)}^{N(n+1)} (\hat{a}_{jn} - \mathbb{E} \hat{a}_{jn}) g_j(x) \right|.
 \end{aligned}$$

Applying Schwartz's and Cauchy's inequalities one gets

$$\begin{aligned}
 (41) \quad \mathbb{E} |V_n(x)w_n(x)| &\leq \gamma_n \sum_{j=0}^{N(n)} \mathbb{E} (\hat{a}_{jn} - a_{jn})^2 \sum_{j=0}^{N(n)} G_j^2 \\
 &\quad + |1-\gamma_n| \left[ \sum_{j=0}^{N(n)} \mathbb{E} (\hat{a}_{jn} - a_{jn})^2 \sum_{j=0}^{N(n)} G_j^2 \right. \\
 &\quad \left. \times \sum_{j=N(n)}^{N(n+1)} \mathbb{E} (\hat{a}_{jn} - a_{jn})^2 \sum_{j=N(n)}^{N(n+1)} G_j^2 \right]^{1/2} \\
 &\leq \gamma_n d_n \left[ \sum_{j=0}^{N(n)} G_j^2 \right]^2 + |1-\gamma_n| d_n \sum_{j=0}^{N(n)} G_j^2 \sum_{j=N(n)}^{N(n+1)} G_j^2.
 \end{aligned}$$

In view of assumptions (33), (34) and (35), inequalities (39), (40) and (41) imply that  $\sum_{n=1}^{\infty} \mathbb{E} |B_n(x)|$  is finite for every  $x \in \mathcal{X}$ . Consequently  $V_n^2(x)$  converges to a finite limit almost surely as  $n$  tends to infinity. By Lemma 1 the limit is zero. The proof has been completed.

LEMMA 4. *If*

$$(42) \quad \sum_{n=1}^{\infty} n^{-2} c_n^2 \left[ \sum_{j=0}^{M(n)} G_j^2 \right]^2 < \infty,$$

then

$$(43) \quad c_n |\hat{f}_n(x) - f(x)| \xrightarrow{n} 0$$



with probability one at every point at which (31) holds.

PROOF. By virtue of Lemma 2 it suffices to show that

$$c_n(\hat{f}_n(x) - E \hat{f}_n(x)) = c_n n^{-1} \sum_{i=1}^n \sum_{j=0}^{M(n)} (g_j(X_i) - E g_j(X_i)) g_j(x)$$

converges to zero with probability one as  $n$  tends to infinity. Let

$$\xi_n(x) = c_n \sum_{j=0}^{M(n)} (g_j(X_n) - E g_j(X_n)) g_j(x).$$

Obviously

$$E \xi_n^2(x) \leq c_n^2 \sum_{j=0}^{M(n)} \text{var } g_j(X_n) \sum_{j=0}^{M(n)} g_j^2(x) \leq c_n^2 \left[ \sum_{j=0}^{M(n)} G_j^2 \right]^2.$$

By virtue of the Kolmogorov strong law of large numbers (see Doob [7], p. 127) and (42), the proof is complete.

Finally we are able to establish the strong consistency of (15).

THEOREM 2. Let (6), (33), (34) and (35) be satisfied. Let, moreover,

$$(44) \quad \sum_{n=1}^{\infty} (c_n^2 + 1) n^{-2} \left[ \sum_{j=0}^{M(n)} G_j^2 \right]^2 < \infty.$$

Then

$$|\hat{R}_n(x) - R_n(x)| \xrightarrow{n} 0$$

w.p. 1 at every point  $x \in \mathcal{X}$  at which  $f(x) > 0$ , (24) and (31) hold.

## 5. The rate of the convergence of $\{d_n\}$

Here we show that the sequence  $\{d_n\}$  defined by (20) converges to zero and is bounded by a power sequence. While proving Theorem 3 we use arguments similar to those used in Dupač [8].

THEOREM 3. Let conditions (6) and (7) be satisfied and let

$$(45) \quad \gamma_n = \delta n^{-r}, \quad \delta > 0, \quad 0 < r < 1,$$

$$(46) \quad \sup_x |t_{n+1}(x) - t_n(x)| = O(n^{-p}), \quad r < p,$$

$$(47) \quad c_n = O(n^q), \quad 2q^+ < r,$$

where  $q^+ = \max(0, q)$ . Then

$$(48) \quad d_n = O(n^{-s}),$$

where

$$s = \begin{cases} 2(p-r) & \text{for } r \geq 2(p+q^+)/3 \\ r-2q^+ & \text{otherwise.} \end{cases}$$

PROOF. To begin with, let us observe that

$$(49) \quad E \{Y_{n+1}g_j(X_{n+1}) | \hat{a}_{j1}, \dots, \hat{a}_{jn}\} = a_{j,n+1}.$$

By (7), (8) and (47),

$$(50) \quad \text{var} [Y_{n+1}g_j(X_{n+1}) | \hat{a}_{j1}, \dots, \hat{a}_{jn}] \leq G_j^2(K_1 + K_3n^{2q}).$$

From (3), (6), (8), (10) and (46) it follows that

$$(51) \quad |a_{j,n+1} - a_{jn}| \leq G_j \sup_x |t_{n+1}(x) - t_n(x)| \int |R(x)|f(x)dx \leq K_4G_jn^{-p}.$$

Subtracting  $a_{j,n+1}$  on both sides of (11) we get

$$\begin{aligned} \hat{a}_{j,n+1} - a_{j,n+1} &= (1-\gamma_n)(\hat{a}_{jn} - a_{jn}) - \gamma_n(a_{j,n+1} - Y_{n+1}g_j(X_{n+1})) \\ &\quad - (1-\gamma_n)(a_{j,n+1} - a_{jn}). \end{aligned}$$

Now after squaring and taking conditional expectations, using (49), (50) and (51) we obtain

$$(52) \quad \begin{aligned} E \{(\hat{a}_{j,n+1} - a_{j,n+1})^2 | \hat{a}_{j1}, \dots, \hat{a}_{jn}\} \\ \leq (1-\gamma_n)(\hat{a}_{jn} - a_{jn})^2 + K_4G_jn^{-p}|\hat{a}_{jn} - a_{jn}| \\ + G_j^2(K_1\gamma_n^2 + K_3\gamma_n^2n^{2q} + K_5n^{-2p}), \end{aligned}$$

for sufficiently large  $n$  satisfying  $|1-\gamma_n| \leq 1$ . It is clear that for every  $\varepsilon > 0$  and every random variable  $Z$  with finite variance

$$2E|Z| \leq \varepsilon^{-1} + \varepsilon E Z^2.$$

Thus, choosing  $\varepsilon = K_6G_j^{-1}\gamma_n n^p$  (for some small  $K_6$ ) one gets

$$(53) \quad 2G_jn^{-p}E|\hat{a}_{jn} - a_{jn}| \leq K_6^{-1}G_j^2\gamma_n^{-1}n^{-2p} + K_6\gamma_n E(\hat{a}_{jn} - a_{jn})^2.$$

Now taking unconditional expectation on both sides of (52) and using (53) one obtains

$$\begin{aligned} E(\hat{a}_{j,n+1} - a_{j,n+1})^2 &\leq (1-K_7n^{-r})E(\hat{a}_{jn} - a_{jn})^2 \\ &\quad + G_j^2(K_8n^{-2r} + K_9n^{2(q-r)} + K_{10}n^{r-2p}) \end{aligned}$$

for sufficiently large  $n$ . Thus,

$$(54) \quad E(\hat{a}_{j,n+1} - a_{j,n+1})^2 \leq (1-K_7n^{-r})E(\hat{a}_{jn} - a_{jn})^2 + K_{11}G_j^2n^{-(r+s)}.$$

Hence

$$d_{n+1} \leq (1-K_7n^{-r})d_n + K_{11}n^{-(r+s)}.$$

Since  $\hat{a}_{j_0}=0$ ,  $d_1$  is finite. A straightforward application of Lemma A completes the proof.

From Theorems 1, 2 and 3 one easily gets the next two concluding ones.

**THEOREM 4.** *Let (6), (7) and (22) hold. Let, moreover, (45), (46) and (47) be satisfied. If*

$$(55) \quad n^{(q^+-1/2)} \sum_{j=0}^{M(n)} G_j^2 \xrightarrow{n} 0 ,$$

$$(56) \quad n^{-s/2} \sum_{j=0}^{N(n)} G_j^2 \xrightarrow{n} 0 ,$$

then

$$|\hat{R}_n(x) - R_n(x)| \xrightarrow{n} 0$$

in probability at every  $x \in \mathcal{X}$  at which  $f(x) > 0$ ,

$$(57) \quad \sum_{j=0}^{N(n)} a_j g_j(x) - h(x) = o(n^{-q})$$

and

$$(58) \quad \sum_{j=0}^{M(n)} b_j g_j(x) - f(x) = o(n^{-q^+}) .$$

**THEOREM 5.** *Let (6), (7) and (22) be satisfied. Let, moreover, (45), (46), (47) and (56) be fulfilled. If*

$$(59) \quad \sum_{n=1}^{\infty} n^{-2(1-q^+)} \left[ \sum_{j=0}^{M(n)} G_j^2 \right]^2 < \infty$$

and

$$(60) \quad \sum_{n=1}^{\infty} n^{-(s+r)} \left[ \sum_{j=0}^{N(n)} G_j^2 \right]^2 < \infty ,$$

then

$$|\hat{R}_n(x) - R_n(x)| \xrightarrow{n} 0$$

almost surely at every  $x \in \mathcal{X}$  at which  $f(x) > 0$  and both (57) and (58) hold.

**PROOF.** Verifying that (33) is implied by (60), (35) is implied by (56) and (60), one can easily complete the proof.

## 6. Examples

The following examples illustrate the fact that conclusions of Theorems 4 and 5 are valid even if  $R_n$  tends to infinity as  $n \rightarrow \infty$ .

Let

$$t_n(x) = (1 + \rho(x)/n)n^q,$$

where

$$\sup_x |\rho(x)| < \infty,$$

and let  $q$  be unknown despite the fact that  $0 < q \leq Q$ , where  $Q$  is a known number. Now  $c_n = n^q$ . One can select  $\gamma_n = \delta n^{-2/3}$  and sequences  $\{N(n)\}$  and  $\{M(n)\}$  of types  $\{n^\alpha\}$  and  $\{n^\beta\}$ , respectively, where  $\alpha$  and  $\beta$  are positive numbers. This choice is decided by examples given below. In this case (46) and (48) hold with  $p = 1 - q$  and  $s = (2 - 6q)/3$ .

We shall consider two examples of applicable orthogonal systems.

### *Hermite orthogonal system*

If  $\mathcal{X}$  is a real line, we can use a system

$$g_j(x) = (2^j j! \pi^{1/2})^{-1/2} e^{-x^2/2} H_j(x),$$

where

$$H_0(x) = 1, \quad H_j(x) = (-1)^j e^{x^2} (d^j e^{-x^2} / dx^j), \quad j = 1, 2, \dots$$

are Hermite polynomials. It can be found in Szegő ([19], p. 242) that  $G_j = K_{12} j^{-1/12}$ .

Suppose that series (12) and (18) converge at a point  $x$  to  $f(x)$  and  $h(x)$ , respectively. Various conditions for the pointwise convergence of orthogonal expansions with the Hermite system can be found in Sansone [16]. Nevertheless, we mention here that the series under consideration converge to  $f(x)$  and  $h(x)$  at every differentiability point of  $f$  and  $h$ , respectively.

One can verify that conditions (22), (55) and (56) of Theorem 4 imposed on sequences  $\{N(n)\}$  and  $\{M(n)\}$  are satisfied for  $\alpha < (2 - 6Q)/5$  and  $\beta < (3 - 6Q)/5$ . In turn, restrictions (22), (59) and (60) of Theorem 5 are fulfilled for  $\alpha < (1 - 6Q)/5$  and  $\beta < (3 - 6Q)/5$ .

We are now interested in assumption (57). Let us assume that the function

$$\zeta(x) = e^{x^2/2} d^m (e^{-x^2/2} h(x)) / dx^m$$

exists and is square integrable. By Schwartz's [17] result

$$|a_j| \leq K_{13} (2j)^{-m/2},$$

where  $K_{13}$  is the  $L_2$  norm of  $\zeta$ . Hence, at every point  $x$  at which the series in (18) converges to  $h(x)$ ,

$$\left| h(x) - \sum_{j=0}^{N(n)} a_j g_j(x) \right| = \left| \sum_{j=N(n)+1}^{\infty} a_j g_j(x) \right| \leq K_{14} \sum_{j=N(n)+1}^{\infty} j^{-(m+1/6)/2} \leq K_{15} n^{-(m-11/6)\alpha/2},$$

which leads to  $\alpha > 12Q/(6m-11)$ . Similar result can be given for (58).

### *Legendre orthogonal system*

If  $\mathcal{X} = [-1, 1]$  we can apply the Legendre system

$$g_j(x) = (j+1/2)^{1/2} P_j(x),$$

where

$$P_0(x) = 1, \quad P_j(x) = (2^j j!)^{-1} [d^j (x^2 - 1)^j / dx^j], \quad j = 1, 2, \dots$$

are Legendre polynomials. In this case  $G_j = K_{16} j^{1/2}$  (see Szegő ([19], p. 164)).

Criterion for the pointwise convergence of series (12) and (18) are given in Sansone [16]. In particular, the series converge to  $f(x)$  and  $h(x)$  at every point at which  $f$  and  $h$  satisfy the Lipschitz condition of a positive order.

By Jackson's theorem, see Sansone ([16], p. 206), if  $h$  is of bounded variation,

$$\left| h(x) - \sum_{j=0}^n a_j g_j(x) \right| = O(n^{-1}),$$

at every  $x$  in the interior of  $\mathcal{X}$ . In this case, (57) is satisfied for  $\alpha > Q$ . Similar result is true for (58).

The order restrictions of Theorems 4 and 5 are satisfied for  $\alpha < (1-3Q)/6$ ,  $\beta < (1-2Q)/4$  and  $\alpha < (1-6Q)/12$ ,  $\beta < (1-2Q)/4$ , respectively.

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