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# MATRIX PARAMETER ESTIMATION IN AN AUTOREGRESSION MODEL

The vector difference equation  $\xi_k = Af(\xi_{k-1}) + \varepsilon_k$ , where  $(\varepsilon_k)$  is a square integrable difference martingale, is considered. A family of estimators  $\check{A}_n$  depending, besides the sample size n, on a bounded Lipschitz function is constructed. Convergence in distribution of  $\sqrt{n} (\check{A}_n - A)$  as  $n \to \infty$  is proved with the use of stochastic calculus. Ergodicity and even stationarity of  $(\varepsilon_k)$  is not assumed, so the limiting distribution may be, as the example shows, other than normal.

#### INTRODUCTON

We consider the vector autoregression process

(1) 
$$\xi_k = Af(\xi_{k-1}) + \varepsilon_k, \quad k \in \mathbb{N}.$$

Here, A is an unknown square matrix, f is a prescribed function, and  $(\varepsilon_k)$  is a square integrable difference martingale with respect to some flow  $(\mathcal{F}_k, k \in \mathbb{Z}_+)$  of  $\sigma$ -algebras such that the random variable  $\xi_0$  is  $\mathcal{F}_0$ -measurable. In the detailed form, the assumption about  $(\varepsilon_k)$  means that for any  $k \varepsilon_k$  is  $\mathcal{F}_k$ -measurable,

(2) 
$$E|\varepsilon_k|^2 < \infty$$

and

(3) 
$$\mathrm{E}(\varepsilon_k | \mathcal{F}_{k-1}) = 0.$$

All vectors are regarded, unless otherwise stated, as columns. Then  $a^{\top}b$  and  $ab^{\top}$  signify scalar and tensor product respectively. The latter is otherwise denoted  $a \otimes b$  (this is a (0, 2)-tensor), in particular  $a^{\otimes 2} = aa^{\top}$ . We use the Euclidean norm of vectors, denoting it  $|\cdot|$ , and the operator norm of matrices. Other notation:  $B^{\dagger}$  – the pseudoinverse to B; O – the null matrix; l.i.p. – limit in probability;  $\stackrel{d}{\rightarrow}$  – the weak convergence of the finite-dimensional distributions of random functions, in particular the convergence in distribution of random vectors.

Let h be a vector function such that for some n

$$E(|\xi_n| + |Af(\xi_n)|) |h(\xi_{n-1})| + E|h(\xi_{n-1})| < \infty.$$

Then from (1) – (3) we have  $E(\xi_n - Af(\xi_{n-1})) \otimes h(\xi_{n-1}) = O$ , whence

$$A = \left(\mathrm{E}\xi_n \otimes h(\xi_{n-1})\right) \left(\mathrm{E}f(\xi_{n-1}) \otimes h(\xi_{n-1})\right)^{-1}$$

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provided the inverse exists. This prompts the estimator

(4) 
$$\check{A}_n = \left(\sum_{k=1}^n \xi_k \otimes h(\xi_{k-1})\right) \left(\sum_{k=1}^n f(\xi_{k-1}) \otimes h(\xi_{k-1})\right)^{\dagger}$$

coinciding in the case f(x) = x with the LSE.

The goal of the article is to study the asymptotic behaviour of the normalized deviation  $\sqrt{n} (A_n - A)$  as  $n \to \infty$ . The use of stochastic calculus underlying our approach allows us to dispense with the assumptions of ergodicity and even asymptotic stationarity of the sequence  $(\varepsilon_k)$ , there t the limiting distribution of the studied statistic may be other than normal. This is the main distinction of our results from A.Ya. Dorogovtsev's ones [1] essentially based on the ergodicity assumption.

# Preliminaries

Let  $E^0$  denote  $E(\cdots | \mathcal{F}_0)$ .

**Lemma 1.** Let conditions (2) and (3) be fulfilled and there exist a number q such that for all x

$$(5) |Af(x)| \le q|x|.$$

Then, for any k,

$$\mathbf{E}^{0}|\xi_{k}|^{2} \leq q^{2k}|\xi_{0}|^{2} + \sum_{i=0}^{k-1} q^{i}\mathbf{E}^{0}|\varepsilon_{k-i}|^{2}.$$

*Proof.* Writing, on the basis of (1),

(6) 
$$|\xi_k|^2 = |Af(\xi_{k-1})|^2 + 2Af(\xi_{k-1})^\top \varepsilon_k + |\varepsilon_k|^2$$

we deduce our assertion from (2), (3) and (5) by induction.

Denote further  $\sigma_k^2 = \mathcal{E}\left(\varepsilon_k^{\otimes 2} | \mathcal{F}_{k-1}\right), \ \chi_k^N = I\{|\xi_k| > N\}, \ I_k^N = I\{|\varepsilon_k| > (1-q)N\}, \ b_k^N = \mathcal{E}^0|\xi_k|^2\chi_k^N.$  Obviously,

(7) 
$$\mathrm{E}\left(|\varepsilon_k|^2 |\mathcal{F}_{k-1}\right) = \mathrm{tr}\sigma_k^2.$$

Lemma 2. Let conditions (2), (3) and (5) be fulfilled and

$$(8) q < 1.$$

Then for any k

$$b_k^N \le q^2 b_{k-1}^N + \mathcal{E}^0 |\varepsilon_k|^2 \chi_{k-1}^N + 2(q/(1-q))^2 N^{-2} \mathcal{E}^0 |\xi_{k-1}|^2 \mathrm{tr}\sigma_k^2 + 2\mathcal{E}^0 |\varepsilon_k|^2 I_k^N.$$

*Proof.* Due to (1) and (5),

$$\chi_k^N \le \chi_{k-1}^N + I_k^N,$$

which together with (6), (5) and the obvious inequality  $|a^{\top}b| \leq |a|^2 + |b|^2$  yields

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$$|\xi_k|^2 \chi_k^N \le q^2 |\xi_{k-1}|^2 \chi_{k-1}^N + 2Af(\xi_{k-1})^\top \varepsilon_k \chi_{k-1}^N + |\varepsilon_k|^2 \chi_{k-1}^N + 2(q^2 |\xi_{k-1}|^2 + |\varepsilon_k|^2) I_k^N.$$

By Lemma 1 and condition (5),  $E^0 |Af(\xi_{k-1})|^2 < \infty$ . Hence, because of (2) and (3),  $\mathbf{E}^{0}\left(Af(\xi_{k-1})^{\top}\varepsilon_{k}|\mathcal{F}_{k-1}\right) = 0.$  The equality

$$\mathbf{E}^{0}|\xi_{k-1}|^{2}I_{k}^{N} = \mathbf{E}^{0}\left(|\xi_{k-1}|^{2}\mathbf{P}\{|\varepsilon_{k}| > (1-q)N|\mathcal{F}_{k-1}\}\right),$$

together with condition (8), Chebyshev's inequality, and equality (7) completes the proof. In what follows, C is a generic constant.

Obviously,  $E^0 \chi_i^N \leq N^{-2} b_i^N$ . Hence and from the previous lemmas we deduce (the details can be found in the proof of Theorem 2[2])

Corollary 1. Let conditions (2), (3), (5), and (8) be fulfilled,

(9) 
$$\lim_{N \to \infty} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=1}^{n} \mathbf{E} |\varepsilon_k|^2 I\{|\varepsilon_k| > N\} = 0$$

and let there exist an  $\mathcal{F}_0$ -measurable random variable v such that for all k

(10) 
$$\operatorname{E}\left(|\varepsilon_k|^2 | \mathcal{F}_{k-1}\right) \le v.$$

Then with probability 1

$$\lim_{N \to \infty} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=1}^{n} b_k^N = 0.$$

## THE MAIN RESULTS

Let h be a Borel function such that

$$(11) |h(x)| \le C|x|.$$

Denote  $\eta_k = h(\xi_k), K_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n \varepsilon_k \otimes \eta_{k-1}, Q_n = \frac{1}{n} \sum_{k=0}^{n-1} f(\xi_k) \otimes \eta_k,$  $T_n = \sqrt{n} (AQ_n Q_n^{\dagger} - A), G_n = \frac{1}{n} \sum_{k=1}^n \sigma_k^2 \otimes \eta_{k-1}^{\otimes 2},$ 

(12) 
$$Y_n(t) = \frac{1}{\sqrt{n}} \sum_{k=1}^{[nt]} \varepsilon_k \otimes \eta_{k-1}.$$

Then, because of (4),

(13) 
$$\sqrt{n}(\check{A}_n - A) = K_n Q_n^{\dagger} + T_n.$$

By construction and conditions (2), (3) and (5),  $Y_n$  is a locally square integrable martingale with quadratic characteristic

(14) 
$$\langle Y_n \rangle(t) = n^{-1} [nt] G^*_{[nt]},$$

where \* is a linear operation in the space of 4-valent tensors such that

$$(a \otimes b \otimes c \otimes d)^* = a \otimes c \otimes b \otimes d.$$

**Theorem 1.** Let conditions (2), (3), (5) and (8) – (11) be fulfilled, and let there exist a random (0, 4)-tensor G such that

(15) 
$$G_n \xrightarrow{\mathrm{d}} G$$

Then  $Y_n \xrightarrow{d} Y$ , where Y is a continuous local martingale with quadratic characteristic  $\langle Y \rangle(t) = G^* t$ .

*Proof.* According to Corollary in [3] and in view of (14) and (15), it suffices to show that for any t

(16) 
$$\operatorname{E}\sup_{s \le t} \|Y_n(s) - Y_n(s-)\|^2 \to 0.$$

The argument in [3] does not change if the expectation is taken conditioned on  $\mathcal{F}_0$ , so in (16) E and  $\rightarrow$  may be substituted by E<sup>0</sup> and  $\xrightarrow{P}$ , respectively. This weakened version of (16) is equivalent, because of (12), to the following relation:

$$n^{-1} \mathbf{E}^0 \max_{k \le nt} \rho_k \xrightarrow{\mathbf{P}} 0,$$

where  $\rho_k = |\varepsilon_k|^2 |\eta_{k-1}|^2$ . Since for any  $\delta > 0$ 

$$\max_{k} \rho_k \le \delta n + \sum_{k} \rho_k I\{\rho_k > \delta n\},\$$

it remains to prove that the random variables  $\sqrt{\rho_k/n}$  satisfy the Lindeberg condition: for any  $\delta > 0$ 

(17) 
$$\frac{1}{n} \sum_{k \le nt} \mathbf{E}^0 \rho_k I\{\rho_k > \delta n\} \xrightarrow{\mathbf{P}} 0.$$

Writing on the basis of (11)

$$\rho_k I\{\rho_k > \delta n\} \left( I\{|\xi_{k-1}| \le N\} + I\{|\xi_{k-1}| > N\} \right)$$
  
$$\le C^2 \left( N^2 |\varepsilon_k|^2 I\{|\varepsilon_k|^2 > (CN)^{-2} \delta n\} + |\varepsilon_k|^2 |\xi_{k-1}|^2 \chi_{k-1}^N \right),$$

we deduce (17) from both the conditions and the conclusion of Corollary 1.

Applying Theorem 1 to the compound processes  $(Y_n, Q_n)$  where the second component does not depend on t, we obtain

**Corollary 2.** Let conditions (2), (3), (5), and (8) – (11) be fulfilled, and let there exist given on a common probability space random (0, 4)-tensor G and (0, 2)-tensor Q such that

(18) 
$$(G_n, Q_n) \xrightarrow{\mathbf{d}} (G, Q).$$

Then  $(Y_n, Q_n) \xrightarrow{d} (Y, Q)$ , where Y is a continuous local martingale w. r. t. some flow  $(\mathcal{F}(t), t \in \mathbb{R}_+)$  such that  $\langle Y \rangle(t) = G^*t$  and the tensor-valued r. v. Q is  $\mathcal{F}(0)$ -measurable.

**Theorem 2.** Let the conditions of Corollary 2 be fulfilled and  $detQ \neq 0$  a. s. Then

(19) 
$$\sqrt{n}(\check{A}_n - A) \stackrel{\mathrm{d}}{\to} Y(1)Q^{-1}$$

Proof. By Corollary 2,

$$(Y_n(1), Q_n) \xrightarrow{\mathrm{d}} (Y(1), Q).$$

But  $Y_n(1) = K_n$ , which together with the nondegeneracy of Q implies that  $K_n Q_n^{\dagger} \xrightarrow{d} K Q^{-1}$ . Now, to obtain the assertion of the theorem from (13), it remains to note that

$$P\{T_n \neq O\} \le P\{\det Q_n \neq O\} \to 0$$

SIMPLER VERSIONS OF CONDITION (18)

Denote  $f_0(x) = x$  and, for  $r \ge 1$ ,

(20) 
$$f_r(x_0,\ldots,x_r) = Af(f_{r-1}(x_0,\ldots,x_{r-1})) + x_r.$$

Then

(21) 
$$\xi_k = f_r(\xi_{k-r}, \varepsilon_{k-r+1}, \dots, \varepsilon_k), \quad r < k,$$

and

(22) 
$$|f_r(x_0,\ldots,x_r)| \le \sum_{i=0}^r q^i |x_{r-i}|.$$

Below  $X_r$  stands for  $(x_1, \ldots, x_r)$ , and d is the dimensionality of each  $x_j$ .

**Lemma 3.** Let for all x, y

$$|Af(x) - Af(y)| \le q|x - y|$$

Then for all  $x, y, r, X_r$ 

$$|f_r(x, X_r) - f_r(y, X_r)| \le q^r |x - y|.$$

*Proof.* Due to (20) and (23),

$$|f_r(x, X_r) - f_r(y, X_r)| \le q |f_{r-1}(x, X_{r-1}) - f_{r-1}(y, X_{r-1})|$$

so it remains to apply the induction.

Corollary 3. Under the conditions of Lemma 3, for any N

$$\lim_{r \to \infty} \sup_{|x| \le N, X_r \in \mathbb{R}^{rd}} |f_r(x, X_r) - f_r(0, X_r)| = 0.$$

**Corollary 4.** Let conditions (5), (8), (11) and (23) be fulfilled and for any x, y

(24) 
$$|h(x) - h(y)| \le C|x - y|.$$

Then for any N > 0

(25)  $\lim_{r \to \infty} \sup_{|x| \le N, X_r \in \mathbb{R}^{rd}} \|f(f_r(x, X_r)) \otimes h(f_r(x, X_r)) - f(f_r(0, X_r)) \otimes h(f_r(0, X_r))\| = 0,$ 

(26) 
$$\lim_{r \to \infty} \sup_{|x| \le N, X_r \in \mathbb{R}^{rd}} \|h(f_r(x, X_r))^{\otimes 2} - h(f_r(0, X_r))^{\otimes 2}\| = 0.$$

Denote further  $\xi_k^r = f_r(0, \varepsilon_{k-r+1}, \dots, \varepsilon_k), \ \eta_k^r = h(\xi_k^r), \ Q_n^r = \frac{1}{n} \sum_{k=r}^{n-1} f(\xi_k^r) \otimes \eta_k^r,$  $G_n^r = \frac{1}{n} \sum_{k=r}^n \sigma_k^2 \otimes (\eta_{k-1}^r)^{\otimes 2}.$  We endow the space of (0, 4)-tensors with such a norm that for any (0, 2)-tensors  $A_1$  and  $A_2$ ,  $||A_1 \otimes A_2|| = ||A_1|| ||A_2||.$ 

Lemma 4. Let conditions (2), (3), (5), (8) - (11), (23) and (24) be fulfilled and

$$(27) |f(x)| \le C|x|.$$

Then almost surely

(28) 
$$\lim_{r \to \infty} \overline{\lim_{n \to \infty}} \mathbf{E}^0 \|Q_n - Q_n^r\| = 0,$$

(29) 
$$\lim_{r \to \infty} \overline{\lim_{n \to \infty}} \mathbb{E}^0 \| G_n - G_n^r \| = 0$$

*Proof.* By Corollary 4 for any N > 0,

(30) 
$$\lim_{r \to \infty} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=r}^{n-1} \mathbb{E} \| f(\xi_k) \otimes \eta_k - f(\xi_k^r) \otimes \eta_k^r \| I\{ |\xi_k| \le N \} = 0.$$

Due to (11) and (27),

$$\mathbf{E}^0 \| f(\xi_k) \otimes \eta_k \| \chi_k^N \le C^2 b_k^N,$$

so, by Corollary 1,

(31) 
$$\lim_{N \to \infty} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=0}^{n-1} \mathbf{E}^0 \| f(\xi_k) \otimes \eta_k \| \chi_k^N = 0.$$

Further, for  $k \ge r$ 

$$\mathbf{E}^{0} \| f(\xi_{k}^{r}) \otimes \eta_{k}^{r} \| = \mathbf{E}^{0} | f(f_{r}(0, \varepsilon_{k-r+1}, \dots, \varepsilon_{k})) | | h(f_{r}(0, \varepsilon_{k-r+1}, \dots, \varepsilon_{k})) |$$

(23)

whence, in view of (22), (27), and (11),

(32) 
$$\mathbf{E} \| f(\xi_k^r) \otimes \eta_k^r \| \chi_k^N \le C^2 \mathbf{E} \left( \sum_{i=0}^{r-1} q^i |\varepsilon_{k-i}| \right)^2 \chi_k^N.$$

Writing the Cauchy–Buniakowsky inequality

$$\left(\sum_{i=0}^{r-1} q^i |\varepsilon_{k-i}|\right)^2 \le \sum_{j=0}^{r-1} q^j \sum_{i=0}^{r-1} q^i |\varepsilon_{k-i}|^2,$$

we get for an arbitrary L > 0

(33)

$$E\left(\sum_{i=0}^{r-1} q^{i} |\varepsilon_{k-i}|\right)^{2} \chi_{k}^{N}$$

$$\leq (1-q)^{-1} \left( E\sum_{i=0}^{r-1} q^{i} |\varepsilon_{k-i}|^{2} I\{|\varepsilon_{k-i}| > L\} + L^{2} P\{|\xi_{k}| > N\} \sum_{i=0}^{r-1} q^{i} \right)$$

Lemma 1 together with (8) and (10) implies that

(34) 
$$\lim_{N \to \infty} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=0}^{n} \mathbb{P}\{|\xi_k| > N\} = 0$$

Obviously, for arbitrary nonnegative numbers  $u_0, \ldots, u_{r-1}, v_1, \ldots, v_{n-1}$ ,

$$\sum_{k=r}^{n-1} \sum_{i=0}^{r-1} u_i v_{k-i} \le \sum_{i=0}^{r-1} u_i \sum_{j=1}^{n-1} v_j,$$

so conditions (8) and (9) imply that

$$\lim_{L \to \infty} \sup_{r} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=r}^{n-1} \mathbb{E} \sum_{i=0}^{r-1} q^{i} |\varepsilon_{k-i}|^{2} I\{|\varepsilon_{k-i}| > L\} = 0,$$

whence, in view of (32) - (34),

$$\lim_{N \to \infty} \sup_{r} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=r}^{n-1} \mathbf{E} \| f(\xi_k^r) \otimes \eta_k^r \| \chi_k^N = 0.$$

Combining this with (30) and (31), we arrive at (28).

The proof of (29) is similar.

**Corollary 5.** Let the conditions of Lemma 4 be fulfilled and for any  $r \in \mathbb{N}$  there exist a pair  $(Q^r, G^r)$  of tensors such that

$$(Q_n^r, G_n^r) \xrightarrow{\mathrm{d}} (Q^r, G^r) \text{ as } n \to \infty.$$

Then the sequence  $((Q^r, G^r), r \in \mathbb{N})$  converges in distribution to some limit (Q, G) and relation (18) holds.

**Lemma 5.** Let the sequence  $(\varepsilon_k)$  satisfy conditions (9) and (10) and for any uniformly bounded sequence  $(\alpha_k)$  of Borel functions on  $\mathbb{R}^{rd}$ 

(36) 
$$\frac{1}{n}\sum_{k=r}^{n-1} \left( \alpha_k(\varepsilon_{k-r+1},\ldots,\varepsilon_k) - \mathbf{E}^0 \alpha_k(\varepsilon_{k-r+1},\ldots,\varepsilon_k) \right) \xrightarrow{\mathbf{P}} \mathbf{O}.$$

Then this relation holds for any sequence  $(\alpha_k)$  of Borel functions such that

(37) 
$$|\alpha_k(x_1,\ldots,x_r)| \le C\left(\sum_{i=1}^r |x_i|^2 + 1\right).$$

*Proof.* Denote  $\zeta_k = \alpha_k(\varepsilon_{k-r+1}, \ldots, \varepsilon_k)$ . Then for any N > 0

$$\frac{1}{n}\sum_{k=r}^{n-1} \left( \zeta_k I\{ |\varepsilon_{k-r+1}| \le N, \dots, |\varepsilon_k| \le N \} - \mathcal{E}^0 \zeta_k I\{ |\varepsilon_{k-r+1}| \le N, \dots, |\varepsilon_k| \le N \} \right) \xrightarrow{\mathcal{P}} 0,$$

so it suffices to prove that, for  $j = 0, \ldots, r - 1$ ,

(38) 
$$\lim_{N \to \infty} \overline{\lim_{n \to \infty} \frac{1}{n}} \sum_{k=1}^{n} \mathbf{E}^{0} |\zeta_{k}| I\{|\varepsilon_{k-j}| > N\} = 0 \quad \text{a.s}$$

By assumption,

(39) 
$$E^{0}|\zeta_{k}|I\{|\varepsilon_{k-j}|>N\} \leq C\left(P\{|\varepsilon_{k-j}|>N|\mathcal{F}_{0}\} + \sum_{i=0}^{k-1} E^{0}|\varepsilon_{k-i}|^{2}I\{|\varepsilon_{k-j}|>N\}\right).$$

Due to (10),  $P\{|\varepsilon_i| > N|\mathcal{F}_0\} \le v^2 N^{-2}$  and  $E^0|\varepsilon_{k-i}|^2 I\{|\varepsilon_{k-j}| > N\} \le v^4 N^{-2}$  as  $i \ne j$ , which together with (39) and (9) implies (38).

*Remark.* Obviously, if relation (36) holds for any sequence of  $\mathbb{R}$ -valued functions (uniformly bounded or satisfying (37)), then for any  $m \in \mathbb{N}$  it is valid for any sequence of  $\mathbb{R}^m$ -valued functions with the same property.

The proof of the following statement is similar.

**Lemma 6.** Let the sequence  $(\varepsilon_k)$  satisfy conditions (9) and (10) and for any uniformly bounded sequence  $(\alpha_k)$  of  $\mathbb{R}$ -valued Borel functions on  $\mathbb{R}^{rd}$ 

(40) 
$$\frac{1}{n} \sum_{k=r}^{n-1} \left( \sigma_k^2 \otimes \alpha_k(\varepsilon_{k-r+1}, \dots, \varepsilon_k) - \mathcal{E}^0(\sigma_k^2 \otimes \alpha_k(\varepsilon_{k-r+1}, \dots, \varepsilon_k)) \right) \xrightarrow{\mathbf{P}} \mathbf{O}$$

(here  $\otimes$  signifies the multiplication of a tensor by a real number). Then this relation holds for any sequence  $(\alpha_k)$  of tensor-valued functions satisfying (37) (with  $\|\cdot\|$  instead of  $|\cdot|$  on the left-hand side).

**Corollary 6.** Let the conditions of Lemmas 4 and 5 be fulfilled and for any uniformly bounded sequence  $(\alpha_k)$  of  $\mathbb{R}$ -valued Borel functions on  $\mathbb{R}^{rd}$  the sequence

$$\left(\frac{1}{n}\sum_{k=r}^{n-1} \mathbf{E}^0 \alpha_k(\varepsilon_{k-r+1},\ldots,\varepsilon_k), \quad n=r,r+1,\ldots\right)$$

converge in probability. Then the sequence  $(Q_n^r, n = r, r + 1...)$  converges in probability. **Corollary 7.** Let the conditions of Lemmas 4 and 6 be fulfilled and for any uniformly bounded sequence of  $\mathbb{R}$ -valued functions the sequence

$$\left(\frac{1}{n}\sum_{k=r}^{n-1} \mathbf{E}^0 \sigma_k^2 \alpha_k(\varepsilon_{k-r+1},\ldots,\varepsilon_k), \quad n=r,r+1,\ldots\right)$$

converge in probability. Then the sequence  $(G_n^r, n = r, r+1...)$  converges in probability.

# AN EXAMPLE

Suppose that conditions (5), (8), (11), (23) and (24) are fulfilled. Let also  $\varepsilon_k = \gamma_k \chi_k$ , where  $(\gamma_k)$  and  $(\chi_k)$  are independent sequences of random variables and i.i.d. random vectors, respectively,  $|\gamma_k| \leq C$ , and let for any  $r \in \mathbb{N}$  and bounded Borel function g the sequence

$$\left(\frac{1}{n}\sum_{k=r}^{n-1}g(\gamma_{k-r+1},\ldots,\gamma_k),\quad n=r,r+1,\ldots\right)$$

converge in probability;  $E\chi_1 = 0$ ,  $E\chi_1^{\otimes 2} = I$ .

For  $\mathcal{F}_k$ , we take the  $\sigma$ -algebra generated by  $\xi_0; \chi_1, \ldots, \chi_k; \gamma_1, \gamma_2, \ldots$  (so that the whole sequence  $(\gamma_k)$  is  $\mathcal{F}_0$ -measurable). Then  $\sigma_k^2 = \gamma_k^2 \mathbf{I}$ ,

(41) 
$$G_n^r = \mathbf{I} \otimes \frac{1}{n} \sum_{k=r}^n (\gamma_k \eta_k^r)^{\otimes 2}$$

and conditions (2), (3), and (10) are fulfilled. So is (9), because the  $\gamma_k$ 's are uniformly bounded and  $\chi_k$ 's are identically distributed.

To deduce (19) from Theorem 2 and Corollary 5 it suffices to verify the conditions of Corollaries 6 and 7. In view of (41) and the expressions for  $Q_n^r$  and  $\eta_k^r$ , we may confine ourselves with the case  $\alpha_k = \alpha$ .

By the Stone – Weierstrass theorem,  $\alpha$  can be approximated uniformly on compacta with finite linear combinations of functions of the kind  $h(y)h_1(x_1) \dots h_r(x_r)$  ( $y \in \mathbb{R}^r, x_j \in \mathbb{R}^d$ ). By the choice of  $\mathcal{F}_k$  and the assumptions on  $(\gamma_k)$  and  $(\chi_k)$ ,

$$\mathbf{E}^{0}h(\overline{\gamma}_{k})h_{1}(\chi_{k-r+1})\dots h_{r}(\chi_{k}) = h(\overline{\gamma}_{k})\prod_{i=1}^{r}\mathbf{E}h_{i}(\chi_{1})$$

where  $\overline{\gamma}_k = (\gamma_{k-r+1}, \dots, \gamma_k).$ 

Hence and from the Chebyshev's inequality, (36) emerges. The last condition of Corollary 6 follows from (41) and the above assumption on  $(\gamma_k)$ .

If det  $Q \neq 0$ , then Theorem 2 asserts (19). If herein l.i.p. $_{n\to\infty} \frac{1}{n} \sum_{k=r}^{n} g(\overline{\gamma}_k)$  is random, then the limiting distribution will not be Gaussian.

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