

# RATES OF A. S. CONVERGENCE OF THE ESTIMATION OF ERROR VARIANCE IN LINEAR MODELS

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## Abstract

Under appropriate conditions we obtain the best rates of a. s. convergence of estimates  $\hat{\sigma}_n^2$  of error variance  $\sigma^2$  and establish the law of iterated logarithm about  $\hat{\sigma}_n^2$ .

## § 1. Introduction and Main Results

Consider the usual linear model

$$y_i = x_i' \beta + e_i, \quad i = 1, \dots, n, \dots, \quad (1)$$

where  $\{x_i\}$  is a known sequence of  $p$ -vectors,  $\beta$  is an unknown  $p$ -vector of regression coefficients,  $\{e_i\}$  is an independent random error sequence, satisfying the following conditions

$$\int_0^\infty x^{2\lambda-1} q(x) dx < \infty \text{ for some } \lambda > 0, \text{ where } q(x) = \sup_i P(|e_i| > x), \quad (2)$$

$$Ee_i = 0, \quad i = 1, 2, \dots, \text{ when } \lambda > \frac{1}{2}. \quad (3)$$

$$0 < Ee_i^2 = \sigma^2 < \infty, \quad i = 1, 2, \dots, \text{ when } \lambda \geq 1. \quad (4)$$

When  $\lambda \geq 1$ , on the basis of the first  $n$  observations of sequence (1), one may calculate the estimate  $\hat{\sigma}_n^2$  of  $\sigma^2$ , based on the residual sum of squares, as follows

$$\hat{\sigma}_n^2 = \frac{1}{n - r_n} \left\{ \sum_{k=1}^n \hat{\sigma}_k^2 - \sum_{j=1}^{r_n} \left( \sum_{k=1}^n c_{nj k} e_k \right)^2 \right\}, \quad (5)$$

where  $r_n = r_k(x_1 | \dots | x_n)$ ,  $\{c_{nj k}, j = 1, \dots, r_n; k = 1, \dots, n\}$  is a group of real numbers determined by  $x_1, \dots, x_n$ , satisfying

$$\sum_{k=1}^n c_{njk} c_{njk} = \delta_{ij}, \quad (6)$$

where  $\delta_{ij}$  is Kronecker sign. Even in case  $\lambda < 1$ , we may still define  $\hat{\sigma}_n^2$  by (5). For rates of convergence of  $\hat{\sigma}_n^2$ , we prove the following theorem:

**Theorem 1.** *Suppose  $e_1, e_2, \dots$  are mutually independently distributed, we have*

(i) *If (2) is satisfied with  $0 < \lambda \leq \frac{1}{2}$ , or (2) — (3) are satisfied with  $\frac{1}{2} < \lambda < 1$ , then*

$$\frac{1}{n^{\frac{1}{\lambda}-1}} \hat{\sigma}_n^2 \xrightarrow{\text{a.s.}} 0 \text{ as } n \rightarrow \infty. \tag{7}$$

(ii) If (2)—(4) are satisfied with  $1 \leq \lambda < 2$ , then

$$n^{1-\frac{1}{\lambda}} (\hat{\sigma}_n^2 - \sigma^2) \xrightarrow{\text{a.s.}} 0 \text{ as } n \rightarrow \infty. \tag{8}$$

when  $1 < \lambda < 2$ , this theorem gives the rates of convergence of  $\hat{\sigma}_n^2 \xrightarrow{\text{a.s.}} \sigma^2$ , that is

$$\hat{\sigma}_n^2 - \sigma^2 = o(n^{-(1-\frac{1}{\lambda})}) \text{ a. s.} \tag{8'}$$

When  $\lambda \geq 2$ , (8') is not true, in this case the rates of convergence are described by the following law of iterated logarithm. For  $\lambda = 2$ , we have

**Theorem 2.** Suppose  $e_1, e_2, \dots$  are mutually independently distributed, and (2)—(4) are satisfied with  $\lambda = 2$ , and

$$\liminf_{n \rightarrow \infty} B_n/n > 0, \tag{9}$$

where  $B_n = \sum_{k=1}^n \text{Var}(e_k^2)$ , then

$$\left. \begin{aligned} \limsup_{n \rightarrow \infty} \frac{n(\hat{\sigma}_n^2 - \sigma^2)}{\sqrt{2B_n \log \log B_n}} &= 1 \text{ a. s.} \\ \liminf_{n \rightarrow \infty} \frac{n(\hat{\sigma}_n^2 - \sigma^2)}{\sqrt{2B_n \log \log B_n}} &= -1 \text{ a. s.} \end{aligned} \right\} \tag{10}$$

When (2) is not satisfied, we have

**Theorem 3.** Suppose  $e_1, e_2, \dots$  are mutually independently distributed,  $Ee_i = 0, Ee_i^2 = \sigma^2$  ( $i = 1, 2, \dots$ ), write  $Z_i = |e_i^2 - \sigma^2|$ , then under the condition (9) and

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n E(Z_k^2 |\log Z_k|^{1+\delta}) < \infty \text{ for some } \delta > 0, \tag{11}$$

(10) holds true.

We remark that the condition (2) is equivalent to the following condition:

There exists a r. v.  $\tilde{\epsilon}$  such that,  $E|\tilde{\epsilon}|^{2\lambda} < \infty$  and  $P(|e_i| > x) \leq P(|\tilde{\epsilon}| > x)$  for any  $x \geq 0$  and any  $i$ . (2')

In this case we often say that  $|e_i|$  is stochastically large than  $|\tilde{\epsilon}|$  for all  $i$ . If  $\{e_i\}$  is an i.i.d. sequence, and  $E|e_1|^{2\lambda} < \infty$ , then (2) holds trivially.

It is well known that, if  $X_1, X_2, \dots$  i.i.d.,  $E|X_1|^\lambda < \infty, \bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$ , then the Marcinkiewicz strong law of large numbers is true, that is

$$\lim_{n \rightarrow \infty} \bar{X}_n/n^{\frac{1}{\lambda}-1} = 0 \text{ a. s., when } 0 < \lambda < 1, \tag{12}$$

$$\lim_{n \rightarrow \infty} n^{1-\frac{1}{\lambda}} (\bar{X}_n - EX_1) = 0 \text{ a. s., when } 1 \leq \lambda < 2. \tag{13}$$

Conversely, if

$$n^{1-\frac{1}{\lambda}} (\bar{X}_n - b_n) \xrightarrow{\text{a.s.}} 0 \text{ as } n \rightarrow \infty \tag{14}$$

for some  $0 < \lambda < 2$  and centering constants  $\{b_n\}$ , then  $E|X_1|^\lambda < \infty$ . This shows that, under the conditions of Theorem 1, the orders given by (7) and (8) are the best possible.

## § 2. Some Lemmas

**Lemma 1.** Suppose  $y \leq 1$ , then

$$e^y \leq 1 + 2|y|^{2q}, \text{ when } 0 < q \leq \frac{1}{2}, \quad (15)$$

$$e^y \leq 1 + y + |y|^{2q}, \text{ when } \frac{1}{2} < q \leq 1. \quad (16)$$

The proof of the Lemma is easy. For example, when  $|y| \leq 1$  and  $0 < q \leq \frac{1}{2}$ , we have

$$e^y = 1 + y + \frac{y^2}{2!} + \frac{y^3}{3!} + \dots \leq 1 + |y| \left( 1 + \frac{1}{1 \times 2} + \frac{1}{2 \times 3} + \dots \right) = 1 + 2|y| \leq 1 + 2|y|^{2q}. \quad (17)$$

Other cases can be dealt with similarly.

**Lemma 2.** Suppose  $e_1, e_2, \dots$  are mutually independently distributed,  $E|e_k|^{2q} \leq M < \infty$  ( $k=1, 2, \dots$ ) for some  $0 < q \leq 1$ , and  $Ee_k = 0$  for each  $k$  when  $q > \frac{1}{2}$ . Also, for some  $\beta > 0$

$$\sum_{k=1}^{\infty} P(|e_k| > k^\beta \varepsilon) < \infty \text{ for every } \varepsilon > 0. \quad (18)$$

Let  $\{a_{nk}, k, n=1, 2, \dots\}$  be a double sequence of real number, satisfying

$$|a_{nk}| \leq Dk^{-\beta} \text{ for all } k, n, \quad (19)$$

and for some  $\alpha > 0$

$$a_n \triangleq \sum_k |a_{nk}|^{2q} \leq Dn^{-\alpha} \text{ for all } n, \quad (20)$$

where  $D > 0$  is a constant. Then  $T_n = \sum_{k=1}^{\infty} a_{nk} e_k$  are a. s. finite, and

$$\lim_{n \rightarrow \infty} T_n = 0 \text{ a. s.} \quad (21)$$

*Proof* It is well known that, if  $X_1, X_2, \dots$  are mutually independently distributed, and  $\sum_k E|X_k|^\delta < \infty$  for some  $0 < \delta \leq 2$ , then  $\sum_{k=1}^{\infty} X_k$  or  $\sum_{k=1}^{\infty} (X_k - EX_k)$  converges a. s. according to  $0 < \delta \leq 1$  or  $1 < \delta \leq 2$  respectively. For any fixed  $n$ , under the conditions of the Lemma,  $\sum_k |a_{nk}|^{2q} E|e_k|^{2q} < \infty$  with  $0 < 2q \leq 2$ , and  $Ee_k = 0$  when  $1 < 2q \leq 2$ , therefore  $T_n$  are all a. s. finite.

Without loss of generality, we assume that all  $a_{nk} \geq 0$ . Choose  $N = \left[ \frac{2}{\alpha} + 1 \right]$ , where  $[x]$  denotes the maximum integral number not exceeding  $x$ . By (18), there exists  $\varepsilon_k \downarrow 0$  such that

$$\sum_{k=1}^{\infty} P\left(|e_k| > \frac{\varepsilon_k}{ND} k^\beta\right) < \infty. \quad (22)$$

By Borel-Cantelli's lemma

$$P\left(|e_k| > \frac{\varepsilon_k}{ND} k^\beta, \text{ i. o.}\right) = 0, \quad (23)$$

therefore for any  $t > 0$

$$\sum_{k=1}^{\infty} |e_k|^t I\left(|e_k| > \frac{\varepsilon_k}{ND} k^\beta\right) < \infty \quad \text{a. s.}, \quad (24)$$

where  $I(A)$  denotes the indicator of the event  $A$ .

Let

$$\left. \begin{aligned} e'_k &= e_k I\left(e_k > \frac{\varepsilon_k}{ND} k^\beta\right), & T'_n &= \sum_k a_{nk} e'_k, \\ e''_{nk} &= e_k I\left(a_{nk} e_k \leq n^{-\alpha/2q}\right), & T''_n &= \sum_k a_{nk} e''_{nk}, \\ e'''_{nk} &= e_k - e'_k - e''_{nk}, & T'''_n &= \sum_k a_{nk} e'''_{nk}. \end{aligned} \right\} \quad (25)$$

For any fixed  $n$ , when  $T_n = \sum_k a_{nk} e_k$  is finite,  $e_k = e''_{nk}$  for  $k$  large enough, hence  $|T_n| < \infty$  a. s. implies  $|T''_n| < \infty$  a. s., and  $|T'_n| < \infty$  a. s. in view of the inequality (26) to be deduced in the following, we have  $|T'''_n| < \infty$  a. s. also.

When  $0 < q \leq \frac{1}{2}$ , let  $q' = 2/3$ , otherwise let  $q' = q$ , then

$$a'_n \triangleq \sum_k |a_{nk}|^{2q'} \leq D' n^{-\alpha} \quad \text{for all } n,$$

where  $D' > 0$  is a constant. By Hölder's inequality

$$\begin{aligned} |T'_n| &\leq |a'_n|^{\frac{1}{2q'}} \left\{ \sum_k |e_k|^{\frac{2q'}{2q'-1}} I\left(|e_k| > \frac{\varepsilon_k}{ND} k^\beta\right) \right\}^{\frac{2q'-1}{2q'}} \rightarrow 0 \quad \text{a. s.} \\ &\text{as } n \rightarrow \infty. \end{aligned} \quad (26)$$

Let  $y_{nk} = n^{\frac{\alpha}{2q}} a_{nk} e''_{nk}$ , then  $y_{nk} \leq 1$ , and  $E y_{nk} \leq 0$  when  $\frac{1}{2} < q \leq 1$ . By Lemma 1

$$\exp(y_{nk}) \leq \begin{cases} 1 + 2|y_{nk}|^{2q}, & \text{when } 0 < q \leq \frac{1}{2}, \\ 1 + y_{nk} + |y_{nk}|^{2q}, & \text{when } \frac{1}{2} < q \leq 1. \end{cases}$$

Thus for  $0 < q \leq 1$

$$E \exp(y_{nk}) \leq 1 + 2E|y_{nk}|^{2q} \leq \exp(2E|Y_{nk}|^{2q}). \quad (27)$$

Therefore by Fatou's lemma, noticing (20) and  $E|e''_{nk}|^{2q} \leq M$ , we have

$$\begin{aligned} E \left\{ \exp\left(n^{\frac{\alpha}{2q}} T'''_n\right) \right\} &= E \left\{ \liminf_{K \rightarrow \infty} \exp\left(\sum_{k=1}^K y_{nk}\right) \right\} \\ &\leq \liminf_{K \rightarrow \infty} E \left\{ \exp\left(\sum_{k=1}^K y_{nk}\right) \right\} = \liminf_{K \rightarrow \infty} \prod_{k=1}^K E \left\{ \exp(y_{nk}) \right\} \\ &\leq \liminf_{K \rightarrow \infty} \prod_{k=1}^K \exp(2E|y_{nk}|^{2q}) \leq \liminf_{K \rightarrow \infty} \exp\left(2n^{\alpha/2} \sum_{k=1}^K |a_{nk}|^{2q} E|e''_{nk}|^{2q}\right) \\ &\leq \exp(2n^{\alpha/2} \cdot Dn^{-\alpha} \cdot M) \leq c, \end{aligned} \quad (28)$$

where  $c > 0$  is a constant. For any  $\varepsilon > 0$

$$P(T_n'' \geq \varepsilon) \leq \exp(-n^{\frac{\alpha}{4q}} \varepsilon) E \{ \exp(n^{\frac{\alpha}{4q}} T_n'') \} \leq c \exp(-n^{\frac{\alpha}{4q}} \varepsilon), \quad (29)$$

hence

$$\sum_{n=1}^{\infty} P(T_n'' \geq \varepsilon) < \infty, \quad (30)$$

by Borel-Cantelli's lemma

$$P(\limsup_{n \rightarrow \infty} T_n'' > \varepsilon) \leq P(T_n'' \geq \varepsilon, \text{ i. o.}) = 0, \quad (31)$$

since  $\varepsilon$  is arbitrary, we have

$$P(\limsup_{n \rightarrow \infty} T_n'' > 0) = 0,$$

i. e.

$$\limsup_{n \rightarrow \infty} T_n'' \leq 0 \text{ a. s.} \quad (32)$$

For any given  $\varepsilon > 0$ , choose a fixed integral number  $K$  such that  $\varepsilon_k < \varepsilon/2$  for all  $k > K$ . By (20), there exists  $n_0$  such that, when  $n \geq n_0$

$$\sum_{k=1}^K a_{nk} e_{nk}'' \leq (Dn^{-\alpha})^{\frac{1}{2q}} \sum_{k=1}^K \frac{\varepsilon_k}{ND} k^\beta < \frac{\varepsilon}{2}. \quad (33)$$

For  $n \geq n_0$ , let

$$D_n = \{k | k > K, a_{nk} e_{nk} \geq n^{-\frac{\alpha}{4q}}\}, \quad (34)$$

then  $D_n$  is a finite set with probability one since  $|T_n| < \infty$  a. s., denoting  $d_n$  the number of elements in  $D_n$ , we have

$$\sum_{k \in D_n} a_{nk} e_{nk}'' \leq \sum_{k \in D_n} D k^{-\beta} \cdot \frac{\varepsilon_k}{ND} k^\beta \leq d_n \frac{\varepsilon}{2N}, \quad (35)$$

thus  $\sum_{k < K} a_{nk} e_{nk}'' \geq \varepsilon/2$  implies  $d_n \geq N$ . So we have

$$\begin{aligned} P(T_n''' \geq \varepsilon) &\leq P\left\{\sum_{k < K} a_{nk} e_{nk}'' \geq \frac{\varepsilon}{2}\right\} \\ &\leq P\{\text{There exist at least } N \text{ of subscripts } k > K \\ &\quad \text{such that } |e_k| \geq (a_{nk} n^{\frac{\alpha}{4q}})^{-1}\} \\ &\leq \left\{\sum_k P[|e_k| \geq (a_{nk} n^{\frac{\alpha}{4q}})^{-1}]\right\}^N \leq \left(\sum_k |a_{nk}|^{2q} n^{\alpha/2} E|e_k|^{2q}\right)^N \\ &\leq (DM n^{-\alpha + \frac{\alpha}{2}})^N = (DM)^N n^{-\frac{\alpha}{2}N}, \end{aligned} \quad (36)$$

by the choice of  $N$ ,  $\frac{\alpha}{2} N > 1$ , therefore

$$\sum_{n=1}^{\infty} P(T_n''' \geq \varepsilon) < \infty \text{ for any given } \varepsilon > 0. \quad (37)$$

The same reasoning as before gives

$$\limsup_{n \rightarrow \infty} T_n''' \leq 0 \text{ a. s.} \quad (38)$$

From (26), (32) and (38), it follows that

$$\limsup_{n \rightarrow \infty} T_n \leq 0 \text{ a. s.} \quad (39)$$

From (39), replacing  $e_k$  by  $-e_k$ , one gets

$$\liminf_{n \rightarrow \infty} T_n \geq 0 \text{ a. s.} \quad (40)$$

Now (21) follows from (39) and (40). The Lemma is proved.

**Lemma 3.** Suppose  $e_1, e_2, \dots$  are mutually independently distributed  $E|e_k|^{2\lambda} \leq M < \infty$  for some  $0 < \lambda \leq 1$ ,  $Ee_k = 0$  when  $\frac{1}{2} < \lambda \leq 1$  ( $k=1, 2, \dots$ ), and

$$\sum_{k=1}^{\infty} P(|e_k| < k^{-\frac{1}{2\lambda}} \varepsilon) < \infty \text{ for every } \varepsilon > 0. \tag{41}$$

Then we have

$$T_n \triangleq n^{-\frac{1}{2\lambda}} \sum_{k=1}^n c_{nk} e_k \xrightarrow{\text{a.s.}} 0 \text{ as } n \rightarrow \infty, \tag{42}$$

where  $\{c_{nk}, k=1, \dots, n; n=1, 2, \dots\}$  is any triangle sequence of real numbers, satisfying the condition

$$\sum_{k=1}^n c_{nk}^2 \leq 1. \tag{43}$$

*Proof* Let

$$a_{nk} = \begin{cases} n^{-\frac{1}{2\lambda}} c_{nk}, & \text{when } k \leq n, \\ 0, & \text{when } k > n, \end{cases}$$

then

$$\left. \begin{aligned} |a_{nk}| &\leq n^{-\frac{1}{2\lambda}} \leq k^{-\frac{1}{2\lambda}}, \text{ when } k \leq n, \\ a_n &\triangleq \sum_k |a_{nk}|^{2\lambda} = \sum_{k=1}^n |c_{nk}|^{2\lambda} / n \leq \left( \frac{1}{n} \sum_{k=1}^n c_{nk}^2 \right)^\lambda \leq n^{-\lambda} \text{ for all } n, \end{aligned} \right\} \tag{44}$$

so that, by Lemma 2 with  $q = \lambda$ ,  $\beta = \frac{1}{2\lambda}$  and  $\alpha = \lambda$ , (42) holds.

Similarly, we have

**Lemma 4.** Suppose  $e_1, e_2, \dots$  are mutually independently distributed,  $Ee_k = 0$ ,  $Ee_k^2 \leq M < \infty$  ( $k=1, 2, \dots$ ), and (41) holds with  $\lambda \geq 1$ , then (43) implies (42).

### § 3. Proof of the Theorems

*Proof of Theorem 1.* First of all, we prove that the condition (2) is equivalent to (2'). To do this, write  $\tilde{q}(x) = P(|\tilde{\varepsilon}| > x)$ ,  $q_t(x) = P(|e_t| > x)$ ,  $\tilde{F}(x) = 1 - \tilde{q}(x)$ ,  $F_t(x) = 1 - q_t(x)$ . Assume that (2') holds, then for  $\forall A > 0, \lambda > 0$

$$\int_0^A x^{2\lambda} d\tilde{F}(x) = - \int_0^A x^{2\lambda} d\tilde{q}(x) = -A^{2\lambda} \tilde{q}(A) + 2\lambda \int_0^A x^{2\lambda-1} \tilde{q}(x) dx, \tag{45}$$

$$A^{2\lambda} \tilde{q}(A) \leq \int_{x \geq A} x^{2\lambda} d\tilde{F}(x) \rightarrow 0 \text{ as } A \rightarrow \infty, \tag{46}$$

therefore

$$E|\tilde{\varepsilon}|^{2\lambda} = \int_0^\infty x^{2\lambda} d\tilde{F}(x) = 2\lambda \int_0^\infty x^{2\lambda-1} \tilde{q}(x) dx. \tag{47}$$

By  $q(x) = \sup_t q_t(x) \leq \tilde{q}(x)$ , we have

$$\int_0^\infty x^{2\lambda-1} q(x) dx \leq \int_0^\infty x^{2\lambda-1} \tilde{q}(x) dx < \infty, \tag{48}$$

i. e., condition (2) is satisfied. Conversely, suppose that (2) holds, then  $q(x)$  is a decreasing function with  $q(0_-) = 1$  and  $q(\infty) = 0$ . In fact, if  $q(\infty) = c > 0$ , then

$$\int_0^\infty x^{2\lambda-1}q(x)dx \geq \int_0^\infty x^{2\lambda-1}dx = \infty \text{ for } \lambda > 0, \quad (49)$$

this contradicts (2). Choose any  $x_n \downarrow x$ , then  $q(x_n) \leq q(x)$ ; on the other hand, for  $\forall \varepsilon > 0$ , there exists  $i_0$  such that  $q_{i_0}(x) \geq q(x) - \varepsilon/2$ , and  $q_{i_0}(x_n) \geq q_{i_0}(x) - \varepsilon/2$  for all sufficiently large  $n$  by the right continuity of  $q_{i_0}(x)$ , thus  $q(x_n) \geq q_{i_0}(x_n) \geq q_{i_0}(x) - \varepsilon/2 \geq q(x) - \varepsilon$  for sufficiently large  $n$ , from this, right continuity of  $q(x)$  is obtained. Therefore,  $F(x) = 1 - q(x)$  furnishes the distribution function of some random variable  $\epsilon$ , and by (45)

$$\int_0^\infty x^{2\lambda}dF(x) \leq 2\lambda \int_0^\infty x^{2\lambda-1}q(x)dx < \infty, \quad (50)$$

i. e.,  $E\epsilon^{2\lambda} < \infty$ , thus the condition (2') is satisfied.

It is well known that, if  $\{X_n\}$  is a sequence of independent r. v.'s, and there exists a r. v.  $X$  such that  $E|X|^p < \infty$  ( $0 < p < 2$ ), and  $P(|X_n| > x) \leq P(|X| > x)$  for  $\forall x \geq 0$  and  $\forall n$ , then

$$\lim_{n \rightarrow \infty} n^{-\frac{1}{p}} \sum_{k=1}^n (X_k - a_k) = 0 \text{ a. s.,}$$

where  $a_k = 0$  or  $EX_k$ , according to  $0 < p < 1$  or  $1 \leq p < 2$  respectively (see [1], p. 242). Therefore

$$\lim_{n \rightarrow \infty} n^{-\frac{1}{\lambda}} \sum_{k=1}^n (\theta_k^2 - a) = 0 \text{ a. s.,} \quad (51)$$

where  $a = 0$  or  $\sigma^2$ , according to  $0 < \lambda < 1$  or  $1 \leq \lambda < 2$  respectively.

By (45) and (47), noticing  $q_k(x) \leq \tilde{q}(x)$ , we have

$$\int_0^A x^{2\lambda}dF_k(x) \leq 2\lambda \int_0^\infty x^{2\lambda-1}q_k(x)dx \leq 2\lambda \int_0^\infty x^{2\lambda-1}\tilde{q}(x)dx = E|\tilde{\epsilon}|^{2\lambda},$$

i. e.

$$E|\theta_k|^{2\lambda} \leq E|\tilde{\epsilon}|^{2\lambda} \triangleq M < \infty \text{ for all } k. \quad (52)$$

On the other hand, for any  $\varepsilon > 0$ ,

$$\begin{aligned} \sum_{k=1}^\infty P(|\theta_k| > k^{\frac{1}{2\lambda}}\varepsilon) &\leq \sum_{k=1}^\infty P(|\tilde{\epsilon}| > k^{\frac{1}{2\lambda}}\varepsilon) = \sum_{k=1}^\infty P(|\tilde{\epsilon}|^{2\lambda} > k\varepsilon^{2\lambda}) \\ &= \sum_{k=1}^\infty \sum_{j=k}^\infty P(j\varepsilon^{2\lambda} < |\tilde{\epsilon}|^{2\lambda} \leq (j+1)\varepsilon^{2\lambda}) \\ &= \sum_{j=1}^\infty jP(j\varepsilon^{2\lambda} < |\tilde{\epsilon}|^{2\lambda} \leq (j+1)\varepsilon^{2\lambda}) \leq \frac{1}{\varepsilon^{2\lambda}} E|\tilde{\epsilon}|^{2\lambda} < \infty. \end{aligned} \quad (53)$$

Applying Lemma 3 (when  $0 < \lambda \leq 1$ ) or Lemma 4 (when  $1 < \lambda < \infty$ ), noticing  $r_n \leq p$ , we have

$$n^{-\frac{1}{\lambda}} \sum_{j=1}^{r_n} \left( \sum_{k=1}^n c_{njk}\theta_k \right)^2 \xrightarrow{\text{a.s.}} 0 \text{ as } n \rightarrow \infty. \quad (54)$$

By (5), (51) and (54), Theorem 1 is proved.

*Proof of Theorem 2.* To begin with, we quote a result in [2] (refer to [3], p. 317): Let  $\{X_n\}$  be a sequence of independent r. v.'s with  $EX_n = 0$ , there exists a r. v.  $X$  such that  $EX^2 < \infty$  and

$$\frac{1}{n} \sum_{k=1}^n P(|X_k| > x) \leq P(|X| > x) \text{ for all sufficiently large } x \text{ and } n, \tag{55}$$

write  $B_n = \sum_{k=1}^n V_{ar}(X_k)$ , then when

$$\liminf_{n \rightarrow \infty} B_n/n > 0, \tag{56}$$

we have

$$\limsup_{n \rightarrow \infty} \frac{\sum_{k=1}^n X_k}{\sqrt{2B_n \log \log B_n}} = 1 \text{ a. s.} \tag{57}$$

Suppose that the conditions of Theorem 2 are satisfied, then for  $x > \sigma^2$  and all  $k$

$$P(|\theta_k^2 - \sigma^2| > x) = P(\theta_k^2 > \sigma^2 + x) \leq P(\tilde{\epsilon}^2 > \sigma^2 + x) \leq P(|\tilde{\epsilon}^2 - \sigma^2| > x). \tag{58}$$

Thus  $\{\theta_k^2 - \sigma^2\}$  and  $\tilde{\epsilon}^2 - \sigma^2$  can serve as  $\{X_k\}$  and  $X$  respectively in (55). Writing  $h_n = \sqrt{2B_n \log \log B_n}$ , we have

$$\left. \begin{aligned} \limsup_{n \rightarrow \infty} \frac{1}{h_n} \sum_{k=1}^n (\theta_k^2 - \sigma^2) &= 1 \text{ a. s.} \\ \liminf_{n \rightarrow \infty} \frac{1}{h_n} \sum_{k=1}^n (\theta_k^2 - \sigma^2) &= -1 \text{ a. s.} \end{aligned} \right\} \tag{59}$$

By (9), there exists a constant  $A > 0$  such that

$$\frac{B_n}{n} > \frac{1}{2A^2} \text{ for all sufficiently large } n, \tag{60}$$

so that, by (54) with  $\lambda = 2$

$$\lim_{n \rightarrow \infty} \frac{1}{h_n} \sum_{j=1}^{r_n} \left( \sum_{k=1}^n c_{njk} \theta_k \right)^2 \leq \lim_{n \rightarrow \infty} A n^{-\frac{1}{2}} \sum_{j=1}^{r_n} \left( \sum_{k=1}^n c_{njk} \theta_k \right)^2 = 0 \text{ a. s.} \tag{61}$$

On the other hand, we have

$$\begin{aligned} \frac{n(\hat{\sigma}_n^2 - \sigma^2)}{\sqrt{2B_n \log \log B_n}} &= \frac{n}{(n-r_n)h_n} \sum_{k=1}^n (\theta_k^2 - \sigma^2) + \frac{nr_n \sigma^2}{(n-r_n)h_n} \\ &\quad + \frac{n}{(n-r_n)h_n} \sum_{j=1}^{r_n} \left( \sum_{k=1}^n c_{njk} \theta_k \right)^2, \end{aligned} \tag{62}$$

and

$$\lim_{n \rightarrow \infty} \frac{nr_n \sigma^2}{(n-r_n)h_n} \leq \lim_{n \rightarrow \infty} A n^{-\frac{1}{2}} p \sigma^2 = 0, \tag{63}$$

therefore (10) holds, and Theorem 2 is proved.

*Proof of Theorem 3.* Without loss of generality, assume that  $\sigma^2 = 1$ . By a well known theorem (see [3], p. 306), under conditions (9) and (11) we have

$$\left. \begin{aligned} \limsup_{n \rightarrow \infty} \frac{1}{h_n} \sum_{k=1}^n (\theta_k^2 - 1) &= 1 \text{ a. s.} \\ \liminf_{n \rightarrow \infty} \frac{1}{h_n} \sum_{k=1}^n (\theta_k^2 - 1) &= -1 \text{ a. s.} \end{aligned} \right\} \tag{64}$$

To simplify the writing, let  $s_n = \sum_{k=1}^n E(Z_k^2 | \log Z_k|^{1+\delta})$ , and use  $c$  to denote a positive constant which may take different value in each appearance. By (11), we have

$$s_n \leq cn, \tag{65}$$

so that, when  $n > 16$  we have

$$\begin{aligned} \sum_{k=16}^n P(|\theta_k| > k^{\frac{1}{4}}) &\leq \sum_{k=16}^n \frac{E(Z_k^2 |\log Z_k|^{1+\delta})}{(\sqrt{k-1})^2 [\log(\sqrt{k-1})]^{1+\delta}} \leq c \sum_{k=16}^n \frac{s_k - s_{k-1}}{k (\log k)^{1+\delta}} \\ &\leq c \left\{ \sum_{k=16}^n s_k \left[ \frac{1}{k (\log k)^{1+\delta}} - \frac{1}{(k+1) (\log(k+1))^{1+\delta}} \right] + \frac{s_n}{(n+1) (\log(n+1))^{1+\delta}} \right\} \\ &\leq c \left\{ \sum_{k=16}^n k \left[ \frac{1}{k (\log k)^{1+\delta}} - \frac{1}{(k+1) (\log(k+1))^{1+\delta}} \right] + \frac{n}{(n+1) (\log(n+1))^{1+\delta}} \right\} \\ &\leq c \sum_{k=16}^{\infty} \frac{1}{k (\log k)^{1+\delta}} < \infty, \end{aligned} \tag{66}$$

hence

$$\sum_{k=1}^{\infty} P(|\theta_k| > k^{\frac{1}{4}}) < \infty, \tag{67}$$

arguing in the same way, for any  $\varepsilon > 0$ , we have

$$\sum_{k=1}^{\infty} P(|\theta_k| > k^{\frac{1}{4}} \varepsilon) < \infty. \tag{68}$$

Therefore, By Lemma 4, (61) holds, and (10) is proved.

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#### References

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