

AN EXACT RATE OF CONVERGENCE IN THE INVARIANCE PRINCIPLE FOR MARTINGALE DIFFERENCE ARRAYS

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ABSTRACT. A convergence rate in the invariance principle for martingale difference arrays is obtained. The rate obtained depends essentially on the asymptotic behavior of the conditional variances. An example which shows the rate obtained is exact is shown. It will be also remarked, by an example, that the convergence rate in the central limit theorem is not always faster than the one in the invariance principle.

1. Introduction and results. Let $X_n = (X_{n,1}, X_{n,2}, \dots, X_{n,n})$, $n \geq 1$, be a triangular array of real valued and square integrable random variables satisfying

$$(1.1) \quad E(X_{n,i} | \mathcal{F}_{n,i-1}) = 0 \text{ a.s.}, \quad 1 \leq i \leq n,$$

where $\mathcal{F}_{n,i}$ is the σ -field generated by $X_{n,1}, X_{n,2}, \dots, X_{n,i}$ and $\mathcal{F}_{n,0}$ is the trivial σ -field. We write

$$S_{n,k} = \sum_{i=1}^k X_{n,i}, \quad 1 \leq k \leq n, \quad V_n^2 = \sum_{i=1}^n E(X_{n,i}^2 | \mathcal{F}_{n,i-1}).$$

Under some assumptions for X_n (see e.g. Theorem 4.1 in Hall and Heyde [1]), it follows from Donsker's invariance principle that, for each $\lambda \geq 0$,

$$(1.2) \quad \lim_{n \rightarrow \infty} P\left(\max_{1 \leq k \leq n} |S_{n,k}| \leq \lambda\right) = T(\lambda),$$

$$(1.3) \quad T(\lambda) = P\left(\max_{0 \leq t \leq 1} |B(t)| \leq \lambda\right) = \frac{4}{\pi} \sum_{k=0}^{\infty} \frac{(-1)^k}{2k+1} \exp\left\{-\frac{(2k+1)^2 \pi^2}{8\lambda^2}\right\},$$

$\{B(t), t \geq 0\}$ being a standard Wiener process.

On the other hand, as a special case of the result due to Hall and Heyde (Theorem 4.6 in [1]), we have the following result: If we assume that

$$(1.4) \quad \max_{1 \leq k \leq n} |X_{n,k}| \leq K_1 n^{-1/2} \text{ a.s.}, \quad n \geq 1,$$

$$(1.5) \quad P(|V_n^2 - 1| \geq 20K_1^2 n^{-1/2} (\log n)^2) = O(n^{-1/4} (\log n)^{3/2}), \quad n \rightarrow \infty,$$

then it holds that

$$(1.6) \quad \Delta_n \equiv \sup_{\lambda} \left| P\left(\max_{1 \leq k \leq n} |S_{n,k}| \leq \lambda\right) - T(\lambda) \right| = O(n^{-1/4}(\log n)^{3/2}), \quad n \rightarrow \infty.$$

K_1, K_2, \dots denote some positive constants independent of n throughout the paper.

We shall in this paper obtain a better rate $O(n^{-1/2} \log n)$ in (1.6) under the assumption (1.4) and a condition which is weaker than (1.5) as a special case of the following theorem.

Theorem 1. *Let $\{X_n\}_{n \geq 1}$ be a triangular array of martingale difference sequences satisfying (1.1). Let $g(\cdot)$ be a positive function such that*

$$(1.7) \quad \liminf_{n \rightarrow \infty} g(n)n^{1/2}(\log n)^{-1} \geq 1.$$

If we assume (1.4) and that

$$(1.8) \quad P(|V_n^2 - 1| \geq g(n)) = O(g(n)), \quad n \rightarrow \infty,$$

then we have

$$(1.9) \quad \Delta_n = O(g(n)), \quad n \rightarrow \infty.$$

Our rate is sharp in the sense that there exists an example of a martingale difference array (see the example in the last section) which satisfies the condition (1.4) and

$$(1.10) \quad |V_n^2 - 1| = g(n)/2 + O(n^{-1}) \text{ a.s.}, \quad n \rightarrow \infty,$$

and hence $\Delta_n \leq K_2 g(n)$ by Theorem 1, while

$$(1.11) \quad \Delta_n \geq K_3 g(n),$$

for some constant $K_3 > 0$.

We now point out about the relationship between our result and the rate of convergence in the central limit theorem. It is plausible that for independent random variables the ordinary rate of convergence to the normal law of single sums is much faster than that for maximums of sums (see Sawyer [3]). However this situation does not keep true for the case of martingale difference arrays. Actually, in an example of Hall and Heyde which satisfies all assumptions of Theorem 1 with $g(n) = n^{-1/2}(\log n)^2$,

$$(1.12) \quad \sup_{-\infty < \lambda < \infty} \left| P(S_{n,n} \leq \lambda) - (2\pi)^{-1/2} \int_{-\infty}^{\lambda} e^{-t^2/2} dt \right| \geq K_4 n^{-1/4} \log n$$

holds (see Remarks in [1], p. 84), but by our present result the convergence rate of (1.2) is given by

$$\Delta_n = O(n^{-1/2}(\log n)^2), \quad n \rightarrow \infty,$$

which is faster than (1.12).

2. Proof of Theorem 1. In proving the theorem we shall use the following lemma which is due to Skorokhod (Lemma 1 in [4], p. 150) and Kato (Lemma 2 in [2]).

Lemma 1. *Suppose that $\{\mathcal{M}_1, \mathcal{M}_2, \dots, \mathcal{M}_n\}$ is a nondecreasing sequence of σ -fields and that $\eta_1, \eta_2, \dots, \eta_n$ are random variables such that η_i is \mathcal{M}_i -measurable, $E(\eta_i | \mathcal{M}_{i-1}) = 0$ a.s., and*

$$(2.1) \quad E(\eta_i^2 | \mathcal{M}_{i-1}) \leq K_5 n^{-1} \text{ a.s.}, \quad E(\eta_i^4) \leq K_6 n^{-2},$$

for each i . Then we have

$$(2.2) \quad P\left(\left|\sum_{i=1}^n \eta_i\right| \geq \log n\right) \leq n^{-1} \{2 \exp(K_5(1+e^2/2)) + K_6\}.$$

Write

$$F_n(\lambda) = P\left(\max_{1 \leq k \leq n} |S_{n,k}| \leq \lambda\right).$$

By the Skorokhod embedding theorem due to Strassen [5] applied to the martingale difference sequence X_n , there is a standard Wiener process $\{B(t), t \geq 0\}$ together with a sequence of nonnegative random variables $\tau_1, \tau_2, \dots, \tau_n$ on a new probability space such that

$$(2.3) \quad \{S_{n,1}, S_{n,2}, \dots, S_{n,n}\} \stackrel{d}{=} \left\{B(\tau_1), B(\tau_1 + \tau_2), \dots, B\left(\sum_{i=1}^n \tau_i\right)\right\}$$

("d" means the equivalence in joint distributions),

$$(2.4) \quad E(\tau_i | \mathcal{B}_{n,i-1}) = E(Y_{n,i}^2 | \mathcal{B}_{n,i-1}) = E(Y_{n,i}^2 | \mathcal{G}_{n,i-1}) \text{ a.s.},$$

$$(2.5) \quad E(\tau_i^k | \mathcal{B}_{n,i-1}) \leq K_7 E(Y_{n,i}^{2k} | \mathcal{B}_{n,i-1}) = K_7 E(Y_{n,i}^{2k} | \mathcal{G}_{n,i-1}) \text{ a.s.},$$

for each positive integer k and $1 \leq i \leq n$, where

$$Y_{n,i} \equiv B\left(\sum_{k=1}^i \tau_k\right) - B\left(\sum_{k=1}^{i-1} \tau_k\right)$$

and $\mathcal{G}_{n,i}$ (respectively $\mathcal{B}_{n,i}$) is the σ -field generated by $Y_{n,1}, Y_{n,2}, \dots, Y_{n,i}$ (respectively $Y_{n,1}, Y_{n,2}, \dots, Y_{n,i}$ and $B(t)$ for $0 \leq t \leq \sum_{k=1}^i \tau_k$) for each $1 \leq i \leq n$. From the assumption (1.4) and the construction of $\{\tau_i\}$, we easily find that

$$(2.6) \quad \max_{0 \leq k \leq n-1} \max_{0 \leq t \leq \tau_{k+1}} \left|B\left(t + \sum_{i=1}^k \tau_i\right) - B\left(\sum_{i=1}^k \tau_i\right)\right| \leq K_1 n^{-1/2} \text{ a.s.},$$

and hence

$$(2.7) \quad F_n(\lambda) \leq P\left(\max_{0 \leq t \leq \sum_{i=1}^n \tau_i} |\mathbf{B}(t)| \leq \lambda + K_1 n^{-1/2}\right),$$

$$(2.8) \quad F_n(\lambda) \geq P\left(\max_{0 \leq t \leq \sum_{i=1}^n \tau_i} |\mathbf{B}(t)| \leq \lambda - K_1 n^{-1/2}\right).$$

Write $\xi_i = n^{1/2}(\tau_i - E(\tau_i | \mathcal{S}_{n,i-1}))$ for each i . From the inequality (2.7), we have

$$(2.9) \quad \begin{aligned} F_n(\lambda) &\leq P\left(\max_{0 \leq t \leq \sum_{i=1}^n \tau_i} |\mathbf{B}(t)| \leq \lambda + K_1 n^{-1/2}, \left|\sum_{i=1}^n \tau_i - 1\right| < 2g(n)\right) \\ &\quad + P\left(\left|\sum_{i=1}^n \tau_i - 1\right| \geq 2g(n)\right) \\ &\leq P\left(\max_{0 \leq t \leq 1-2g(n)} |\mathbf{B}(t)| \leq \lambda + K_1 n^{-1/2}\right) + P\left(\left|\sum_{i=1}^n n^{-1/2} \xi_i\right| \geq g(n)\right) \\ &\quad + P\left(\left|\sum_{i=1}^n E(\tau_i | \mathcal{S}_{n,i-1}) - 1\right| \geq g(n)\right). \end{aligned}$$

We first estimate the second part of the extreme right hand side of (2.9). From the definition of ξ_i and the inequalities (1.4), (2.4) and (2.5), we have

$$\begin{aligned} E(\xi_i | \mathcal{S}_{n,i-1}) &= 0 \text{ a.s.}, \\ E(\xi_i^2 | \mathcal{S}_{n,i-1}) &\leq n E(\tau_i^2 | \mathcal{S}_{n,i-1}) \leq K_8 n E(X_{n,i-1}^2 | \mathcal{S}_{n,i-1}) \leq K_9 n^{-1} \text{ a.s.}, \\ E(\xi_i^4) &\leq n^2 E(\tau_i^4) \leq K_{10} n^2 E(Y_{n,i-1}^8) \leq K_{11} n^{-2}, \end{aligned}$$

for each i . Thus we can use Lemma 1 to obtain, from the assumption (1.7), that

$$(2.10) \quad P\left(\left|\sum_{i=1}^n n^{-1/2} \xi_i\right| \geq g(n)\right) \leq P\left(\left|\sum_{i=1}^n \xi_i\right| \geq \log n\right) \leq n^{-1} \{2 \exp(K_9(1+e^2/2)) + K_{11}\}.$$

As to the last term of the right hand side of (2.9), we have from (2.4) and (1.8) that

$$(2.11) \quad P\left(\left|\sum_{i=1}^n E(\tau_i | \mathcal{S}_{n,i-1}) - 1\right| \geq g(n)\right) = P\left(\left|\sum_{i=1}^n E(Y_{n,i}^2 | \mathcal{S}_{n,i-1}) - 1\right| \geq g(n)\right) = O(g(n)),$$

and as to the first term we have, from the scaling property of the Wiener process, that

$$(2.12) \quad \begin{aligned} &P\left(\max_{0 \leq t \leq 1-2g(n)} |\mathbf{B}(t)| \leq \lambda + K_1 n^{-1/2}\right) \\ &= P\left(\max_{0 \leq t \leq 1} |\mathbf{B}((1-2g(n))t)| \leq \lambda + K_1 n^{-1/2}\right) \end{aligned}$$

$$\begin{aligned}
 &= P\left(\max_{0 \leq t \leq 1} |\mathbf{B}(t)| \leq (\lambda + K_1 n^{-1/2})(1 - 2g(n))^{-1/2}\right) \\
 &= T((\lambda + K_1 n^{-1/2})(1 - 2g(n))^{-1/2}) .
 \end{aligned}$$

Thus we have from (2.9)-(2.12) that

$$(2.13) \quad F_n(\lambda) \leq T((\lambda + K_1 n^{-1/2})(1 - 2g(n))^{-1/2}) + O(g(n)) .$$

We can also obtain, by a similar argument, that

$$(2.14) \quad \begin{aligned} F_n(\lambda) &\geq P\left(\sup_{0 \leq t \leq 1+2g(n)} |\mathbf{B}(t)| \leq \lambda - K_1 n^{-1/2}\right) - P\left(\left|\sum_{i=1}^n \tau_i - 1\right| \geq 2g(n)\right) \\ &\geq T((\lambda - K_1 n^{-1/2})(1 + 2g(n))^{-1/2}) - O(g(n)) . \end{aligned}$$

On the other hand, we have, from the inequality (3.1) in Sawyer [4], that

$$(2.15) \quad \begin{aligned} \sup_{\lambda} |T((\lambda + K_1 n^{-1/2})(1 - 2g(n))^{-1/2}) - T((\lambda - K_1 n^{-1/2})(1 + 2g(n))^{-1/2})| \\ \leq 8\{3/(\pi e)\}^{1/2} g(n) + 8(1 + g(n))K_1 n^{-1/2} = O(g(n)) . \end{aligned}$$

The combination of this with (2.13) and (2.14) yields the theorem.

3. Example. With a slight modification of the Hall and Heyde example (Example 5 in [1], p. 82), we can give an example of a triangular array of independent random variables which satisfies all assumptions of Theorem 1 and the lower bound (1.11).

Example. Let $\{Z_i, i \geq 1\}$ be i.i.d. random variables with $P(Z_1 = 1) = P(Z_1 = -1) = 1/2$. Given a function $g(\cdot)$ satisfying (1.7), we set $m_n = [n(1 - g(n)/2)]$ and

$$X_{n,t} = \begin{cases} n^{-1/2} Z_i, & 1 \leq i \leq m_n, \\ 0, & m_n < i \leq n, \end{cases}$$

here $[a]$ denotes the integral part of a .

Then for each n , $\mathbf{X}_n = \{X_{n,t}, 1 \leq i \leq n\}$ is a martingale difference satisfying (1.4). It also satisfies (1.8), because

$$V_n^2 = \sum_{t=1}^n E(X_{n,t}^2 | \mathcal{F}_{n,t-1}) = m_n/n = 1 - g(n)/2 + O(n^{-1}) \text{ a.s.}$$

We now prove that \mathbf{X}_n satisfies (1.11). Since the derivative $T'(\lambda)$ is continuous and strictly positive for $\lambda > 0$, we have $T(1 + \varepsilon) - T(1) \sim K_{12}\varepsilon$, $\varepsilon \downarrow 0$, where $K_{12} = T'(1) > 0$. By the scaling property of the Wiener process we have

$$\begin{aligned}
 \max_{0 \leq t \leq m_n/n} |\mathbf{B}(t)| &\stackrel{d}{=} (m_n/n)^{1/2} \max_{0 \leq t \leq 1} |\mathbf{B}(t)| \\
 &= \{1 - g(n)/2 + O(n^{-1})\}^{1/2} \max_{0 \leq t \leq 1} |\mathbf{B}(t)| \\
 &= \{1 + g(n)/4 + O(n^{-1})\}^{-1} \max_{0 \leq t \leq 1} |\mathbf{B}(t)| ,
 \end{aligned}$$

and hence

$$\begin{aligned}
 (3.1) \quad \Delta_n^* &\equiv \sup_{\lambda} \left| P\left(\max_{0 \leq t \leq m_n/n} |\mathbf{B}(t)| \leq \lambda \right) - T(\lambda) \right| \\
 &\geq \left| P\left(\max_{0 \leq t \leq m_n/n} |\mathbf{B}(t)| \leq 1 \right) - T(1) \right| \\
 &= T(1 + g(n)/4 + O(n^{-1})) - T(1) \sim K_{12}g(n).
 \end{aligned}$$

But from Theorem 1 in Sawyer [3] we know that

$$\begin{aligned}
 (3.2) \quad \tilde{\Delta}_n &\equiv \sup_{\lambda} \left| P\left(\max_{1 \leq k \leq m_n} |S_{n,k}| \leq \lambda \right) - P\left(\max_{0 \leq t \leq m_n/n} |\mathbf{B}(t)| \leq \lambda \right) \right| \\
 &= O(n^{-1/2}(\log n)^{1/2}) = o(g(n)),
 \end{aligned}$$

and hence from (3.1) and (3.2) we have

$$\Delta_n \geq \Delta_n^* - \tilde{\Delta}_n \geq K_{12}g(n) - o(g(n)),$$

proving (1.11).

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