

THE ORDER OF NORMAL APPROXIMATION FOR SIGNED LINEAR RANK STATISTICS*

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Summary. The rate of convergence of the cdf (cumulative distribution function) of the signed linear rank statistics to the normal one is investigated. Under suitable assumptions, it is shown that the convergence rate is of order $O(N^{-1/2+\delta})$ for any $\delta > 0$.

1. Introduction. Let Y_{Ni} , $1 \leq i \leq N$, $N \geq 1$, be independent random variables distributed according to the cdf's $F_{Ni}(y) = F(y - \Delta d_{Ni})$, $1 \leq i \leq N$, $N \geq 1$, where Δ is an unknown real parameter, and d_{Ni} , $1 \leq i \leq N$, $N \geq 1$, are known real constants. Consider the signed linear rank statistics

$$(1.1) \quad S_{\Delta N} = \sum_{i=1}^N c_{Ni} \varphi \left(\frac{R_{Ni}^{\Delta}}{N+1} \right) \text{sgn} (Y_{Ni}),$$

where R_{Ni}^{Δ} is the rank of $|Y_{Ni}|$ among $|Y_{N1}|, \dots, |Y_{NN}|$, c_{Ni} 's are known real constants, $\text{sgn} (y) = 1$ or -1 according as $y \geq 0$ or < 0 , and φ is a score generating function defined over $(0, 1)$. Under suitable assumptions on the c 's and the score generating function φ , the asymptotic normality of $S_{\Delta N}$ has been established by Hušková [3]. However, the problem of determining the rate of convergence of the cdf of $S_{\Delta N}$ to the limiting normal one has not been investigated so far. The purpose of this paper is to study this problem. It is shown that the rate of convergence is of order $O(N^{-1/2+\delta})$ for any $\delta > 0$ for the case $\Delta = 0$ as well as $\Delta \neq 0$. For the corresponding study of the similar problem in the case of (unsigned) linear rank statistics of the form

$$S_N = \sum_{i=1}^N c_{Ni} \varphi \left(\frac{R_{Ni}}{N+1} \right),$$

where R_{Ni} is the rank of Y_{Ni} among Y_{N1}, \dots, Y_{NN} , the reader is referred to Jurečková and Puri [5], Bergström and Puri [1], and Hušková [4], among others. Our methods are mainly the adaptations of the ideas of Jurečková and Puri [5], Bergström and Puri [1] and Hušková [4]. Throughout the paper we make the following assumptions:

A1) $F(x)$ is continuous and symmetric about zero,

$$A2) \quad \sum_{i=1}^N c_{Ni}^2 = 1, \quad \max_{1 \leq i \leq N} |c_{Ni}| = O(N^{-1/2} \sqrt{\log N}),$$

A3) φ is not identically zero and satisfies the Lipschitz condition of order one on $(0, 1)$, that is, there exists a constant C such that

$$|\varphi(u) - \varphi(v)| \leq C|u - v| \text{ for all } u, v \in (0, 1).$$

Remark. Unlike in earlier papers referred to above where the derivatives of different orders on φ are assumed, here the Lipschitz condition on φ is the only assumption we are making.

For simplicity of notation, we shall suppress the subscript N in c_{Ni} , d_{Ni} , R_{Ni}^{Δ} etc., whenever this causes no confusion.

2. Rate of convergence for $\Delta = 0$. Here we assume that $\Delta = 0$, so that Y_{Ni}, \dots, Y_{NN} , $N \geq 1$, are i.i.d. random variables distributed according to a continuous and symmetric cdf $F(y)$. We shall need the following theorem (cf. Petrov [7, Chap. V, Thm. 6]).

THEOREM 2.1. Let V_1, \dots, V_N be independent random variables with $E V_j = 0$, $E|V_j|^{2+\delta} < \infty$ for some positive $\delta \leq 1$ ($j = 1, \dots, N$). Then

$$\sup_{x \in (-\infty, \infty)} |G_N(x) - \Phi(x)| \leq A B_N^{-1-\delta/2} \sum_{j=1}^N E|V_j|^{2+\delta},$$

where $A > 0$ is a constant independent of N , $B_N = \sum_{j=1}^N D V_j$, $G_N(x) = P[B_N^{-1/2} \sum_{j=1}^N V_j \leq x]$, and $\Phi(x)$ is the standard cumulative normal distribution function.

* Received by the editors July 13, 1982; in revised form May 20, 1985.

¹ Work supported by the Office of Naval Research under contract No. N00014-77-C-0659. Reproduction in whole or part is permitted for any purpose of the United States Government.

AMS 1970 subject classifications. Primary 62E20, Secondary 60F99.

Now let us denote

$$(2.1) \quad T_{\Delta N} = \sum_{i=1}^N c_{Ni} \varphi(F^*(|Y_{Ni}|)) \operatorname{sgn}(Y_{Ni}), \quad F^*(x) = P_0[|Y_{11}| \leq x],$$

where P_0 means that the probability is computed under $\Delta = 0$, and note that $T_{\Delta N}$ is a sum of independent random variables. We shall use $T_{\Delta N}$ to approximate $S_{\Delta N}$.

LEMMA 2.1. *Under assumptions A1-A3, we have*

$$\sup_{x \in (-\infty, \infty)} |G_{0N}(x) - \Phi(x)| = O(N^{-1/2} \sqrt{\log^3 N}),$$

where G_{0N} is the cdf of $T_{0N}/\|\varphi\|_2$ and T_{0N} is given by (2.1).

Proof. Since $\mathbf{E}[c_i \varphi(F^*(|Y_i|)) \operatorname{sgn}(Y_i)] = 0$, and $\mathbf{E}|c_i \varphi(F^*(|Y_i|)) \operatorname{sgn}(Y_i)|^3 = |c_i|^3 \int_0^1 |\varphi^3(u)| du < \infty$ for $1 \leq i \leq N$, it follows from Theorem 2.1 that

$$(2.2) \quad \sup_{x \in (-\infty, \infty)} |G_{0N}(x) - \Phi(x)| \leq AB_N^{-3/2} \sum_{i=1}^N |c_i|^3 \int_0^1 |\varphi^3(u)| du,$$

where $A > 0$ is a constant independent of N , and $B_N = \sum_{i=1}^N c_i^2 \int_0^1 \varphi^2(u) du$. The proof now follows from A2, A3 and (2.2).

LEMMA 2.2. *Under assumptions A1-A3,*

$$\mathbf{E}(S_{0N} - T_{0N})^{2K} = O[N^{-K} (\log N)^K]$$

for any fixed positive integer K , where S_{0N} and T_{0N} are given by (1.1) and (2.1) respectively.

Proof. Put $Z_i = [\varphi(R_i^0/(N+1)) - \varphi(F^*(|Y_i|))] \operatorname{sgn}(Y_i)$. Then, for arbitrary nonnegative integers p_1, p_2, \dots, p_N satisfying $\sum_{i=1}^N p_i = 2K$ we have

$$(2.3) \quad \mathbf{E}\left(\prod_{i=1}^N Z_i^{p_i}\right) = 0,$$

if at least one of the p_i 's is an odd number. Then using (2.3) we obtain

$$(2.4) \quad \begin{aligned} \mathbf{E}[(S_{0N} - T_{0N})]^{2K} &= \mathbf{E}\left[\left(\sum_{i=1}^N c_i Z_i\right)^{2K}\right] \\ &= \sum_{l_1=0}^K \sum_{l_2=0}^K \cdots \sum_{\substack{l_N=0 \\ l_1+l_2+\dots+l_N=K}}^K C_{2l_1}^{2K} c_1^{2l_1} c_2^{2l_2} \cdots c_N^{2l_N} \mathbf{E}\left(\prod_{i=1}^N Z_i^{2l_i}\right) \\ &= \sum_{m=1}^K \left[\sum_{i_1=1}^N \sum_{i_2=1}^N \cdots \sum_{\substack{i_m=1 \\ i_1 < i_2 < \dots < i_m}}^N \right. \\ &\quad \left. \cdot \left[\sum_{\substack{l_1=1 \\ l_1+l_2+\dots+l_m=K}}^K \sum_{l_2=1}^K \cdots \sum_{l_m=1}^K C_{2l_1}^{2K} c_{i_1}^{2l_1} \cdots c_{i_m}^{2l_m} \mathbf{E}(Z_{i_1}^{2l_1} Z_{i_2}^{2l_2} \cdots Z_{i_m}^{2l_m}) \right] \right], \end{aligned}$$

where

$$l_j = (l_1, l_2, \dots, l_j) \quad \text{and} \quad C_{2l_j}^{2K} = \frac{(2K)!}{(2l_1)!(2l_2)! \cdots (2l_j)!} \quad \text{for } j = N \text{ or } m.$$

Now, by the generalized Cauchy-Schwarz inequality and A3, we have for any $i_1 < i_2 < \dots < i_m$, m ; $l_j = 1, \dots, K$; $j = 1, \dots, m$, $\sum_{j=1}^m l_j = K$, that

$$(2.5) \quad \begin{aligned} \mathbf{E}(Z_{i_1}^{2l_1} Z_{i_2}^{2l_2} \cdots Z_{i_m}^{2l_m}) &\leq \left(\prod_{j=1}^m \mathbf{E}(Z_j^{2m l_j})\right)^{1/m} \\ &= \left\{ \prod_{j=1}^m \mathbf{E}[\varphi(R_{i_j}^0/(N+1)) - \varphi(F^*(|Y_{i_j}|))]^{2m l_j} \right\}^{1/m} \\ &\leq C^{2K} \left\{ \prod_{j=1}^m \mathbf{E}\left(\frac{R_{i_j}^0}{N+1} - F^*(|Y_{i_j}|)\right)^{2m l_j} \right\}^{1/m} \leq M_k N^{-K}, \end{aligned}$$

where M_K is a constant depending only on K . To obtain the last inequality in (2.5) we have used the well-known fact (see e.g. Kallenberg [6, p. 406]) that

$$\mathbf{E} \left(\frac{R_1^0}{N+1} - F^*(|Y_1|) \right)^{2p} = O(N^{-p})$$

for any positive integer p . Also, in view of A2, there exists a constant M'_K (depending only on K) such that

$$(2.6) \quad |c_{i_1}^{2l_1} c_{i_2}^{2l_2} \cdots c_{i_m}^{2l_m}| \leq M'_K N^{-K} (\log N)^K$$

for any $i_1 < \cdots < i_m$, $m = 1, \dots, K$, and any $l_j = 1, \dots, K$; $j = 1, \dots, m$, $\sum_{j=1}^m l_j = K$. The proof now follows from (2.4), (2.5) and (2.6).

The main theorem of this section is the following.

THEOREM 2.2. *Under assumptions A1-A3, we have for any $\delta > 0$ that*

$$\sup_{x \in (-\infty, \infty)} |E_{0N}(x) - \varphi(x)| = O(N^{-1/2+\delta}),$$

where F_{0N} is the cdf of $S_{0N}/\|\varphi\|_2$.

Proof. For any $\varepsilon > 0$ and any N , we have

$$(2.7) \quad \mathbf{P}[\|\Phi\|_2^{-1} S_{0N} \leq x] \leq \mathbf{P}[\|\Phi\|_2^{-1} T_{0N} \leq x + \varepsilon] + \mathbf{P}[\|\varphi\|_2^{-1} |S_{0N} - T_{0N}| \geq \varepsilon]$$

and

$$(2.8) \quad \mathbf{P}[\|\Phi\|_2^{-1} S_{0N} \leq x] \geq \mathbf{P}[\|\Phi\|_2^{-1} T_{0N} \leq x - \varepsilon] - \mathbf{P}[\|\Phi\|_2^{-1} |S_{0N} - T_{0N}| \geq \varepsilon].$$

It then follows from (2.7), (2.8), Lemma 2.1, Lemma 2.2 and Markov's inequality that for any $\delta > 0$

$$(2.9) \quad \sup_{x \in (-\infty, \infty)} |F_{0N}(x) - \Phi(x)| \leq O(N^{-1/2} \sqrt{\log^3 N}) + O(\varepsilon) + \frac{O(N^{-K} (\log N)^K)}{(\varepsilon \|\Phi\|_2)^{2K}}$$

holds for any $\varepsilon > 0$ and any $K \geq 1$. Now, pick K large enough such that $(2K + 1)^{-1} < 2\delta$ and put $\varepsilon = N^{-1/2+1/(2(2K+1))}$. Then using (2.9), the proof follows immediately.

3. Rate of convergence under near alternatives. In this section, we assume $\Delta \neq 0$ so that Y_{N1}, \dots, Y_{NN} , $N \geq 1$, are independently distributed with cdf's $F(y - \Delta d_{N1}), \dots, F(y - \Delta d_{NN})$, $N \geq 1$, respectively. Furthermore, besides the assumptions A1-A3, we assume:

B1) F is absolutely continuous, $f(x) = F'(x)$ is absolutely continuous, and there exists a constant $\theta_0 > 0$ such that

$$\sup_{\theta \in [-\theta_0, \theta_0]} \int_{-\infty}^{\infty} \frac{[f'(x - \theta)]^2}{f(x)} dx < \infty,$$

and

$$B2) \quad \sum_{i=1}^N d_{Ni}^2 = 1, \quad \sum_{i=1}^N d_{Ni} = 0, \quad N = 2, 3, \dots, \quad \max_{1 \leq i \leq N} |d_{Ni}| = O(N^{-1/2} \sqrt{\log N}).$$

In what follows \mathbf{E}_0 and \mathbf{E}_Δ will denote the expectations computed under $\Delta = 0$ and $\Delta \neq 0$ respectively. Furthermore, $P_{\Delta r}$ will denote the restriction of P_Δ to the set $D_N = \{\prod_{i=1}^N f(Y_{Ni}) \neq 0\}$ so that for any measurable function h of (Y_1, Y_2, \dots, Y_N)

$$(3.1) \quad \begin{aligned} \mathbf{E}_{\Delta r} h(Y_1, Y_2, \dots, Y_N) &= \int_{\{\prod_{i=1}^N f(y_i) \neq 0\}} h(y_1, y_2, \dots, y_N) \\ &\cdot \prod_{i=1}^N f(y_i - \Delta d_i) dy_1 dy_2 \cdots dy_N, \end{aligned}$$

i.e., $\mathbf{E}_{\Delta r} h(Y_1, Y_2, \dots, Y_N) = \mathbf{E}_\Delta \{h(Y_1, Y_2, \dots, Y_N) I_{D_N}(Y_1, Y_2, \dots, Y_N)\}$, where I denotes the indicator function.

LEMMA 3.1. *Under assumptions A1, B1 and B2,*

$$P_\Delta(D_N^c) = O(N^{-1/2} \sqrt{\log^3 N})$$

holds for any $\Delta \neq 0$, where D_N^c is the complement of D_N .

Remark. Our assumption B1 is equivalent to assumption V in Hušková [4]. In fact, Hušková's assumption V implies our assumption B1 and her Lemma 3.5 implies our Lemma 3.1.

Proof. By Lemma 3.5 of Hušková [4] and B2, there exists a constant C_1 (independent of N) and a positive integer N_Δ such that for all $N \geq N_\Delta$ we have

$$P_\Delta \left[\prod_{i=1}^N f(Y_{Ni}) = 0 \right] \leq C_1 |\Delta|^3 \sum_{i=1}^N |d_i|^3 = O(N^{-1/2} \sqrt{\log^3 N}).$$

The proof follows.

LEMMA 3.2. *Under assumptions A1 to A3, B1 and B2, it holds true for any positive integer K and any $\Delta \neq 0$ that*

$$(3.2) \quad \{E_\Delta[(S_{\Delta N} - T_{\Delta N})I_{D_N^\Delta}(Y_1, \dots, Y_N)]\}^{2K} = O[(N^{-K}(\log N)^{3K})^{1-1/(2p)}(\log N)^{2K}]$$

for all $p = 1, 2, \dots$.

Proof. By the Hölder inequality, the left-hand side of (3.2) is bounded by

$$(3.3) \quad [E_\Delta(S_{\Delta N} - T_{\Delta N})^{2p}]^{K/p} [P_\Delta(D_N^\Delta)]^{2K(1-1/(2p))}$$

for any $p = 1, 2, \dots$. Thus the assertion (3.2) will be proved in view of (3.3) and Lemma 3.1, if we can show that $E_\Delta(S_{\Delta N} - T_{\Delta N})^{2p} = O[(\log N)^{2p}]$ for any $p = 1, 2, \dots$. Now, put

$$Z_{i\Delta} = \left[\varphi \left(\frac{R_i^\Delta}{N+1} \right) - \varphi(F^*(|Y_i|)) \right] \text{sgn}(Y_i).$$

Then from (1.1), (2.1) and (2.4) we have

$$(3.4) \quad \begin{aligned} E_\Delta(S_{\Delta N} - T_{\Delta N})^{2p} &= E_\Delta \left[\left(\sum_{i=1}^N c_i Z_{i\Delta} \right)^{2p} \right] \\ &= \sum_{m=1}^{2p} \left[\sum_{i_1=1}^N \cdots \sum_{i_m=1}^N \left[\sum_{l_1=1}^{2p} \cdots \sum_{l_m=1}^{2p} C_{l_1}^{2p} c_{i_1}^{l_1} \cdots c_{i_m}^{l_m} E_\Delta(Z_{i_1\Delta}^{l_1} \cdots Z_{i_m\Delta}^{l_m}) \right] \right], \end{aligned}$$

where

$$C_{l_m}^{2p} = \frac{(2p)!}{(l_1!) \cdots (l_m!)}.$$

We now consider the expectation $E_\Delta Z_{i\Delta}^{2q}$ for any $q = 1, 2, \dots$ and any $i = 1, \dots, N$.

From A3 it follows that

$$(3.5) \quad \begin{aligned} E_\Delta Z_{i\Delta}^{2q} &\leq C^{2q} E_\Delta \left(\frac{R_i^\Delta}{N+1} - F^*(|Y_i|) \right)^{2q} \\ &\leq 2^{2q-1} C^{2q} \left[E_\Delta \left(\frac{R_i^\Delta}{N+1} - V_{ii\Delta} \right)^{2q} + E_\Delta (V_{ii\Delta} - F^*(|Y_i|))^{2q} \right], \end{aligned}$$

where $V_{ii\Delta} = E_\Delta(R_i^\Delta/(N+1)|Y_i)$. By Lemma 2.1 (after slight and trivial modification) of Bergström and Puri [1] we obtain

$$(3.6) \quad E_\Delta \left(\frac{R_i^\Delta}{N+1} - V_{ii\Delta} \right)^{2q} = O(N^{-q}), \quad q = 1, 2, \dots,$$

uniformly in $i = 1, 2, \dots, N$. Using the Mean Value Theorem and B1, we can easily show that

$$(3.7) \quad E_\Delta (V_{ii\Delta} - F^*(|Y_i|))^{2q} = O(\max_{i \leq j \leq N} |d_j|^{2q}), \quad q = 1, 2, \dots,$$

uniformly in $i = 1, 2, \dots, N$. It now follows from B2, (3.5), (3.6) and (3.7) that

$$(3.8) \quad E_\Delta Z_{i\Delta}^{2q} = O[N^{-q}(\log N)^q], \quad q = 1, 2, \dots,$$

holds uniformly in $i = 1, \dots, N$. In view of (3.4) and (3.8), it follows, by using arguments analogous to those in the proof of Lemma 2.2, that

$$(3.9) \quad E_\Delta(S_{\Delta N} - T_{\Delta N})^{2p} = O[(\log N)^{2p}], \quad p = 1, 2, \dots.$$

The proof follows.

LEMMA 3.3. *Under assumptions A1-A3, B1 and B2, it holds for any $K = 1, 2, \dots$ and any $\Delta \neq 0$ that*

$$(3.10) \quad E_{\Delta r}[(S_{\Delta N} - E_\Delta S_{\Delta N}) - (T_{\Delta N} - E_\Delta T_{\Delta N})]^{2K} = O[N^{-K(1-1/(2p))}(\log N)^{q_{Kp}}]$$

for all $p = 1, 2, \dots$, where $q_{Kp} = (5-3/(2p))K$.

Proof. The left-hand side of (3.10) is bounded by

$$(3.11) \quad 2^{2K-1} \mathbf{E}_{\Delta r}(S_{\Delta N} - T_{\Delta N})^{2K} + 2^{2K-1} [\mathbf{E}_{\Delta}(S_{\Delta N} - T_{\Delta N})]^2 K.$$

In view of Lemma 3.6 of Hušková [4], there exist a constant C_2 (independent of N) and a positive integer N'_Δ such that for all $N \geq N'_\Delta$

$$(3.12) \quad \mathbf{E}_{\Delta r}(S_{\Delta N} - T_{\Delta N})^{2K} \leq C_2 \{\mathbf{E}_0(S_{0N} - T_{0N})^{4K}\}^{1/2} = O[N^{-K}(\log N)^K],$$

where the last equality is by Lemma 2.2. Next, we have

$$(3.13) \quad \begin{aligned} [\mathbf{E}_{\Delta}(S_{\Delta N} - T_{\Delta N})]^{2K} &= \{\mathbf{E}_{\Delta r}(S_{\Delta N} - T_{\Delta N}) + \mathbf{E}_{\Delta}[(S_{\Delta N} - T_{\Delta N})I_{D_N^c}(Y_1, \dots, Y_N)]\}^{2K} \\ &\leq 2^{2K-1} [\mathbf{E}_{\Delta r}(S_{\Delta N} - T_{\Delta N})]^{2K} + 2^{2K-1} \{\mathbf{E}_{\Delta}[(S_{\Delta N} - T_{\Delta N})I_{D_N^c}(Y_1, \dots, Y_N)]\}^{2K}. \end{aligned}$$

Using (3.1) and Jensen's inequality, it follows that

$$(3.14) \quad [\mathbf{E}_{\Delta r}(S_{\Delta N} - T_{\Delta N})]^{2K} \leq \mathbf{E}_{\Delta r}(S_{\Delta N} - T_{\Delta N})^{2K}.$$

The proof now follows from (3.11)–(3.14) and Lemma 3.2.

The following lemma is an immediate consequence of the previous lemmas and the Markov inequality.

LEMMA 3.4. *Under assumptions A1–A3, B1 and B2, it holds true for any $\Delta \neq 0$ and $\varepsilon > 0$ that*

$$(3.15) \quad \begin{aligned} P_{\Delta}[(S_{\Delta N} - \mathbf{E}_{\Delta} S_{\Delta N}) - (T_{\Delta N} - \mathbf{E}_{\Delta} T_{\Delta N})] \|\varphi\|_2^{-1} > \varepsilon] \\ \leq M_{Kp} [N^{-1/2} \sqrt{\log^3 N} + (\|\varphi\|_2 \varepsilon)^{-2K} N^{-K(1-1/(2p))} (\log N)^{q_{Kp}}] \end{aligned}$$

for all $K, p = 1, 2, \dots$ and for all $N \geq N_{\Delta Kp}$ where $N_{\Delta Kp}$ is a positive integer, M_{Kp} is a constant independent of N and $q_{Kp} = (5-3/(2p))K$.

Proof. For any $\varepsilon > 0$, the left-hand of (3.15) is bounded by

$$(3.16) \quad P_{\Delta}(D_N^c) + P_{\Delta r}[(S_{\Delta N} - \mathbf{E}_{\Delta} S_{\Delta N}) - (T_{\Delta N} - \mathbf{E}_{\Delta} T_{\Delta N})] \|\varphi\|_2^{-1} > \varepsilon].$$

(Recall from (3.1) that $P_{\Delta r}(A) = P_{\Delta}(A \cap D_N)$ for any measurable set A .)

The first term in (3.16) is of order $O(N^{-1/2} \sqrt{\log^3 N})$ as can be seen from Lemma 3.1. The second term in (3.16), by Markov's inequality, is bounded by

$$\mathbf{E}_{\Delta r}[(S_{\Delta N} - \mathbf{E}_{\Delta} S_{\Delta N}) - (T_{\Delta N} - \mathbf{E}_{\Delta} T_{\Delta N})]^{2K} (\|\varphi\|_2 \varepsilon)^{-2K} \text{ for any } K = 1, 2, \dots$$

The lemma now follows from Lemma 3.3.

THEOREM 3.1. *Under assumptions A1–A3, B1 and B2, it holds true that*

$$\sup_{x \in (-\infty, \infty)} |G_{\Delta N}(x) - \Phi(x)| = O(N^{-1/2} \sqrt{\log^3 N}),$$

where $G_{\Delta N}$ is the cdf of $(T_{\Delta N} - \mathbf{E}_{\Delta} T_{\Delta N}) \|\varphi\|_2^{-1}$ and $T_{\Delta N}$ is given by (2.1).

Proof. Let us denote $W_{i\Delta} = \varphi(F^*(|Y_i|) \operatorname{sgn}(Y_i))$. Then using (2.1), we have $T_{\Delta N} - \mathbf{E}_{\Delta} T_{\Delta N} = \sum_{i=1}^N c_i (W_{i\Delta} - \mathbf{E}_{\Delta} W_{i\Delta})$ which is the sum of independent random variables, and by A2 we have

$$(3.17) \quad \begin{aligned} \|\mathbf{D}_{\Delta} T_{\Delta N} - \|\varphi\|_2\|^2 &= \left| \sum_{i=1}^N c_i^2 [\mathbf{E}_{\Delta} W_{i\Delta}^2 - \|\varphi\|_2^2 - (\mathbf{E}_{\Delta} W_{i\Delta})^2] \right| \\ &\leq \max_{1 \leq i \leq N} |\mathbf{E}_{\Delta} W_{i\Delta}^2 - \|\varphi\|_2^2| + \max_{1 \leq i \leq N} |\mathbf{E}_{\Delta} W_{i\Delta}|^2. \end{aligned}$$

Now, using A1, A3, B1 and the Mean Value Theorem we obtain

$$(3.18) \quad \begin{aligned} \max_{1 \leq i \leq N} |\mathbf{E}_{\Delta} E_{i\Delta}^2 - \|\varphi\|_2^2| &\leq \max_{1 \leq i \leq N} \int_{-\infty}^{\infty} |\varphi^2(F^*(|x + \Delta d_i|)) - \varphi^2(F^*(|x|))| dF(x) \\ &\leq 2\|\varphi\|_{\infty} C \max_{1 \leq i \leq N} \int_{-\infty}^{\infty} |F^*(|x + \Delta d_i|) - F^*(|x|)| dF(x) \\ &\leq 2\|\varphi\|_{\infty} C \int_{-\infty}^{\infty} [F^*(|x| + \max_{1 \leq i \leq N} |\Delta d_i|) \\ &\quad - F^*(|x| - \max_{1 \leq i \leq N} |\Delta d_i|)] dF(x) \\ &= O(\max_{1 \leq i \leq N} |d_i|) = O(N^{-1/2} \sqrt{\log N}) \end{aligned}$$

and similarly

$$\begin{aligned}
 \max_{1 \leq i \leq N} |\mathbf{E}_\Delta W_{i\Delta}| &= \max_{1 \leq i \leq N} |\mathbf{E}_\Delta W_{i\Delta} - \mathbf{E}_0 W_{i0}| \\
 &\leq \max_{1 \leq i \leq N} \int_{M_{i\Delta}}^{\infty} |\varphi(F^*(x + \Delta d_i)) - \varphi(F^*(x))| dF(x) \\
 (3.19) \quad &+ \max_{1 \leq i \leq N} \int_{-\infty}^{m_{i\Delta}} |\varphi(F^*(-x)) - \varphi(F^*(-x - \Delta d_i))| dF(x) \\
 &+ \max_{1 \leq i \leq N} \int_{m_{i\Delta}}^{M_{i\Delta}} |\varphi(F^*(|x + \Delta d_i|)) \operatorname{sgn}(x + \Delta d_i) \\
 &\quad - \varphi(F^*(|x|)) \operatorname{sgn}(x)| dF(x),
 \end{aligned}$$

where $M_{i\Delta} = \max(0, -\Delta d_i)$ and $m_{i\Delta} = \min(0, -\Delta d_i)$. The third integral in (3.19) is of order $O(\max_{1 \leq i \leq N} |d_i|)$ since $\|\Phi\|_\infty < \infty$ and $F(M_{i\Delta}) - F(m_{i\Delta}) = O(\max_{1 \leq i \leq N} |d_i|)$. The first and second integrals are also of order $O(\max_{1 \leq i \leq N} |d_i|)$, which can be proved as in deriving (3.18). It then follows from B2, (3.17), (3.18) and (3.19) that

$$(3.20) \quad \|\mathbf{D}_\Delta T_{\Delta N} - \|\varphi\|_2^2\| = O(N^{-1/2} \sqrt{\log N}).$$

But A3 implies that $\|\varphi\|_2 > 0$. Thus there exists a positive integer N'_Δ such that $\frac{1}{2}\|\varphi\|_2^2 < \mathbf{D}_\Delta T_{\Delta N} < \frac{3}{2}\|\varphi\|_2^2$ for all $N \geq N'_\Delta$. Consequently we have for all $N \geq N'_\Delta$,

$$\begin{aligned}
 \left| \frac{\|\varphi\|_2}{\sqrt{\mathbf{D}_\Delta T_{\Delta N}}} - 1 \right| &= \frac{|\|\varphi\|_2 - \sqrt{\mathbf{D}_\Delta T_{\Delta N}}| (\|\Phi\|_2 + \sqrt{\mathbf{D}_\Delta T_{\Delta N}})}{\sqrt{\mathbf{D}_\Delta T_{\Delta N}} (\|\varphi\|_2 + \sqrt{\mathbf{D}_\Delta T_{\Delta N}})} \\
 (3.21) \quad &\leq \frac{3\|\varphi\|_2^{-2}}{\sqrt{2+1}} \|\varphi\|_2^2 - \mathbf{D}_\Delta T_{\Delta N} = O(N^{-1/2} \sqrt{\log N}).
 \end{aligned}$$

Now let us denote the cdf of $(T_{\Delta N} - \mathbf{E}_\Delta T_{\Delta N})(\mathbf{D}_\Delta T_{\Delta N})^{-1/2}$ by $G_{\Delta N}^*$. Since $\mathbf{E}_\Delta |c_i(W_{i\Delta} - \mathbf{E}_\Delta W_{i\Delta})|^3 < \infty$, $i = 1, \dots, N$, we obtain from Theorem 2.1 and (3.20) that there exists a constant A (independent on N) such that for all $N \geq N'_\Delta$,

$$\begin{aligned}
 \sup_{x \in (-\infty, \infty)} |G_{\Delta N}^*(x) - \varphi(x)| &\leq A(\mathbf{D}_\Delta T_{\Delta N})^{-3/2} \sum_{i=1}^N |c_i|^3 \mathbf{E} W_{i\Delta} - \mathbf{E}_\Delta W_{i\Delta}|^3 \\
 (3.22) \quad &\leq 16\sqrt{2}A \|\varphi\|_2^{-3} \|\varphi\|_\infty^3 \sum_{i=1}^N |c_i|^3.
 \end{aligned}$$

We now consider the uniform distance between $G_{\Delta N}$ and Φ . It follows from the triangle inequality and (3.22) that

$$\begin{aligned}
 \sup_{x \in (-\infty, \infty)} |G_{\Delta N}(x) - \Phi(x)| &= \sup_{x \in (-\infty, \infty)} |G_{\Delta N}^*(x \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2}) - \Phi(x)| \\
 &\leq \sup_{x \in (-\infty, \infty)} |G_{\Delta N}^*(x \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2}) - \Phi(x \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2})| \\
 (3.23) \quad &+ \sup_{x \in (-\infty, \infty)} |\Phi(x \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2}) - \Phi(x)| \\
 &\leq O\left(\sum_{i=1}^N |c_i|^3\right) + \sup_{|x| \leq \sqrt{\log N}} \|\Phi(x \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2}) - \Phi(x)\| \\
 &\quad + \sup_{|x| > \sqrt{\log N}} |\Phi(x \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1}) - \Phi(x)|.
 \end{aligned}$$

On the right-hand side of the last inequality in (3.23), the second term, in view of (3.21) and the Mean Value Theorem, is bounded by

$$(3.24) \quad \|\Phi'\|_\infty \sqrt{\log N} \left\| \frac{\|\varphi\|_2^2}{\sqrt{\mathbf{D}_\Delta T_{\Delta N}}} - 1 \right\| = O(N^{-1/2} \log N),$$

and the third term, by the triangle inequality and the monotonicity of Φ , is bounded by

$$\begin{aligned}
 (3.25) \quad & \sup_{x > \sqrt{\log N}} |1 - \Phi(x) \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2}| + \sup_{x > \sqrt{\log N}} |1 - \Phi(x)| \\
 & = [1 - \Phi(\sqrt{\log N} \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2})] + [1 + \Phi(\sqrt{\log N})] \\
 & = [\Phi(\sqrt{\log N}) - \Phi(\sqrt{\log N} \|\varphi\|_2 (\mathbf{D}_\Delta T_{\Delta N})^{-1/2})] + 2[1 - \Phi(\sqrt{\log N})]
 \end{aligned}$$

which is of order $O(N^{-1/2} \log N)$ as can be seen from (3.21) and from Lemma 2, Chapter VII of Feller [2]. The rest of the proof follows from A2, (3.23), (3.24) and (3.25).

We now prove the main result of this section.

THEOREM 3.2. *Under assumptions A1-A3, B1 and B2, it holds true for any $\delta > 0$ that*

$$\sup_{x \in (-\infty, \infty)} |F_{\Delta N}(x) - \Phi(x)| = O(N^{-1/2+\delta}),$$

where $F_{\Delta N}$ is the cdf of $(S_{\Delta N} - \mathbf{E}_\Delta S_{\Delta N}) \|\Phi\|_2^{-1}$.

Proof. For any $\varepsilon > 0$ and any real number x , we have

$$\begin{aligned}
 (3.26) \quad & P_\Delta[(S_{\Delta N} - \mathbf{E}_\Delta S_{\Delta N}) \|\varphi\|_2^{-1} \leq x] \leq P_\Delta[(T_{\Delta N} - \mathbf{E}_\Delta T_{\Delta N}) \|\varphi\|_2^{-1} \leq x + \varepsilon] \\
 & \quad + P_\Delta[|(S_{\Delta N} - \mathbf{E}_\Delta S_{\Delta N}) - (T_{\Delta N} - \mathbf{E}_\Delta T_{\Delta N})| \|\varphi\|_2^{-1} > \varepsilon]
 \end{aligned}$$

and

$$\begin{aligned}
 (3.27) \quad & P_\Delta[(S_{\Delta N} - \mathbf{E}_\Delta S_{\Delta N}) \|\varphi\|_2^{-1} \leq x] \geq P_\Delta[(T_{\Delta N} - \mathbf{E}_\Delta T_{\Delta N}) \|\varphi\|_2^{-1} \leq x - \varepsilon] \\
 & \quad - P_\Delta[|(S_{\Delta N} - \mathbf{E}_\Delta S_{\Delta N}) - (T_{\Delta N} - \mathbf{E}_\Delta T_{\Delta N})| \|\varphi\|_2^{-1} > \varepsilon].
 \end{aligned}$$

It follows from (3.26), (3.27), Lemma 3.4 and Theorem 3.1 that for all large N

$$\begin{aligned}
 (3.28) \quad & \sup_{x \in (-\infty, \infty)} |F_{\Delta N}(x) - \Phi(x)| \leq O(N^{-1/2} \sqrt{\log^3 N}) + O(\varepsilon) \\
 & \quad + (\|\varphi\|_2 \varepsilon)^{-2K} O[N^{-(1-1/(2p))K} (\log N)^{(5-3/(2p))K}]
 \end{aligned}$$

holds for any $K, p = 1, 2, \dots$, and any $\varepsilon > 0$. Now, pick K large enough so that $(2K + 1)^{-1} < \delta$, let $p = K(2K + 1)$, and put $\varepsilon = N^{-1/2} + 1/(2(2K + 1))$. The theorem then follows from (3.28).

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