

ON THE CENTRAL LIMIT THEOREM FOR INDEPENDENT
RANDOM VARIABLES WITH ALMOST SURE CONVERGENCE

BY

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Abstract. We obtain an almost sure convergence limit theorem for independent nonidentically distributed random variables. Let S_n , $n \geq 1$, be the partial sums of independent random variables with zero means and finite variances and let $a(x)$ be a real function. We present sufficient conditions under which in logarithmic means $a(S_n/(ES_n^2)^{1/2})$ converges almost surely to $\int_{-\infty}^{\infty} a(x) d\Phi(x)$.

1. Introduction. Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, defined on some probability space (Ω, \mathcal{A}, P) , such that $EX_n = 0$ and $EX_n^2 = \sigma_n^2 < \infty$, $n \geq 1$.

Let us put

$$S_0 = 0, \quad S_n = X_1 + \dots + X_n, \quad V_n^2 = ES_n^2.$$

It is well known that under some additional assumption

$$S_n/V_n \xrightarrow{\mathcal{D}} \Phi \quad \text{as } n \rightarrow \infty,$$

where Φ denotes the standard normal distribution. But for mathematical statistics it may be of some interest whether assertions are possible for almost every realization of the random variables X_n , $n \geq 1$. Namely, for $x \in \mathbb{R}$ we denote by δ_x the probability measure on \mathbb{R} which assigns its total mass to x . Let us observe that the distribution function of S_n/V_n is just the average of the random measure $\delta_{S_n(\omega)/V_n}$ with respect to P , i.e., for every $A \in \mathcal{B}(\mathbb{R})$

$$P(S_n/V_n \in A) = \int \delta_{S_n(\omega)/V_n}(A) dP(\omega).$$

Of course, for every $\omega \in \Omega$, $\{\delta_{S_n(\omega)/V_n}, n \geq 1\}$ is a sequence of probability measures on the space $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$. Moreover, under the assumptions of Theorem 2 of Rodzik and Rychlik [8], P -a.s.

$$(1.1) \quad (\log V_n^2)^{-1} \sum_{k=1}^n (\sigma_{k+1}^2/V_k^2) \delta_{S_k(\omega)/V_k} \xrightarrow{\mathcal{D}} \Phi \quad \text{as } n \rightarrow \infty,$$

where $\xrightarrow{\mathcal{D}}$ denotes the weak convergence of measures on $(\mathbf{R}, \mathcal{B}(\mathbf{R}))$. Thus we form time averages with respect to a logarithmic scale and prove almost sure convergence for the resulting random measures.

In this paper we present sufficient conditions under which P -a.s.

$$(1.2) \quad (\log V_n^2)^{-1} \sum_{k=1}^n (\sigma_k^2/V_k^2) a(S_k(\omega)/V_k) \rightarrow \int_{-\infty}^{\infty} a(x) d\Phi(x) \quad \text{as } n \rightarrow \infty$$

for a real function $a(\cdot)$ which is almost everywhere continuous and $|a(x)| \leq \exp(\gamma x^2)$ for some $\gamma < 1/4$. Of course from (1.2) we easily get (1.1).

The almost sure version of the central limit theorem has been studied by many authors in the case where $\{X_n, n \geq 1\}$ is a sequence of independent or weakly dependent and identically distributed random variables. In this case, (1.1) and some extensions of (1.1) have been considered by Schatte [9], Brosamler [3], Lacey and Philipp [5], Atlagh and Weber [1], Berkes and Dehling [2], Peligrad and Shao [6]. The assertion (1.2), in the case of independent and identically distributed random variables, has been considered by Schatte [10]. Thus the main result presented in this paper extends Theorem 1 of [10] to the case of nonidentically distributed random variables. In the proofs we shall also follow the ideas of [10].

2. Results. We shall now state the main results of the paper.

THEOREM 1. *Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, defined on (Ω, \mathcal{A}, P) , with $EX_n = 0$ and $0 < EX_n^2 = \sigma_n^2 < \infty$, $n \geq 1$. Define, for $n \geq 1$, $S_n = X_1 + \dots + X_n$, $V_n^2 = ES_n^2$. Let $a(x)$ be a real function which is a.e. continuous and for which $|a(x)| \leq \exp(\gamma x^2)$ for some $\gamma < 1/4$. Assume*

$$(2.1) \quad (\max_{1 \leq k \leq n} \sigma_k^2)/V_n^2 \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

and for some positive, nondecreasing real function f on \mathbf{R}^+ , such that the function $f(x)/x$ is nonincreasing on \mathbf{R}^+ ,

$$(2.2) \quad \sup_n (\log V_n^2) (f(V_n^2)/V_n^2)^{1/4} < \infty,$$

$$(2.3) \quad \lim_{n \rightarrow \infty} (\log V_n^2)^{-1} \sum_{k=1}^n (\sigma_k^2/V_k^2) (f(V_k^2)/V_k^2)^{1/4} (\log V_k^2)^{(3\gamma \vee 2\gamma)+1} = 0,$$

and

$$(2.4) \quad \sum_{n=1}^{\infty} (f(V_n^2))^{-1} EX_n^2 I(X_n^2 \geq f(V_n^2)) < \infty.$$

Then

$$(2.5) \quad P\left(\lim_{n \rightarrow \infty} (\log V_n^2)^{-1} \sum_{k=1}^n (\sigma_k^2/V_k^2) a(S_k/V_k) = \int_{-\infty}^{\infty} a(x) d\Phi(x)\right) = 1.$$

Let us observe that, in general, (2.2) does not imply (2.3). For example, if $\gamma > 0$ and $V_n^2 = n, n \geq 2, f(x) = x/(\log x)^4$, then (2.2) holds. But

$$\begin{aligned} (\log N)^{-1} \sum_{n=2}^N n^{-1} (\log n)^{3\gamma} &= (\log N)^{-1} \sum_{n=2}^N N^{-1} (n/N)^{-1} \left\{ \log \left(N \frac{n}{N} \right) \right\}^{3\gamma} \\ &\leq (\log N)^{-1} \int_{1/N}^1 x^{-1} (\log Nx)^{3\gamma} dx = (3\gamma + 1)^{-1} (\log N)^{3\gamma+1} (\log N)^{-1} \rightarrow \infty \\ &\hspace{15em} \text{as } N \rightarrow \infty. \end{aligned}$$

On the other hand, in general, (2.3) does not imply (2.2) either. For example, if $-4 < \gamma < 0$ and $V_n^2 = n, n \geq 2, f(x) = x(\log x)^{-4-\gamma}$, then

$$(\log n)(f(n)/n)^{1/4} = (\log n)^{-\gamma/4} \rightarrow \infty \quad \text{as } n \rightarrow \infty,$$

and

$$\begin{aligned} (\log N)^{-1} \sum_{n=3}^N n^{-1} (\log n)^{7\gamma/8} &\leq (\log N)^{-1} \int_{\log 2}^{\log N} x^{7\gamma/8} dx \\ &= (7\gamma/8 + 1)^{-1} ((\log N)^{7\gamma/8+1} - (\log 2)^{7\gamma/8+1}) (\log N)^{-1} \rightarrow 0 \quad \text{as } N \rightarrow \infty. \end{aligned}$$

We also note that if $\gamma < 0$, then (2.2) implies (2.3). This is a consequence of the Toeplitz lemma.

Now let us observe that if (2.2) and (2.4) hold, then

$$(2.6) \quad X_n (2 \log \log V_n^2)^{1/2} / V_n \rightarrow 0 \text{ a.s. as } n \rightarrow \infty.$$

Namely, we have

$$\begin{aligned} \sum_{n=1}^{\infty} P(V_n^{-1} |X_n| (2 \log \log V_n^2)^{1/2} \geq (2 \log \log V_n^2)^{1/2} (f(V_n^2)/V_n^2)^{1/2}) \\ = \sum_{n=1}^{\infty} P(X_n^2 \geq f(V_n^2)) \leq \sum_{n=1}^{\infty} (f(V_n^2))^{-1} EX_n^2 I(X_n^2 \geq f(V_n^2)) < \infty. \end{aligned}$$

Since, by (2.2), $(2 \log \log V_n^2)^{1/2} (f(V_n^2)/V_n^2)^{1/2} \rightarrow 0$ as $n \rightarrow \infty$, it follows that (2.6) holds.

On the other hand, by (2.4) and Kronecker's lemma

$$(2.7) \quad (f(V_n^2))^{-1} \sum_{k=1}^n EX_k^2 I(X_k^2 \geq f(V_k^2)) \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Let us put

$$L_n(\varepsilon) = V_n^{-2} \sum_{k=1}^n EX_k^2 I(|X_k| > \varepsilon V_k).$$

Then, taking into account the assumptions concerning the function f ,

we get

$$\begin{aligned} L_n(\varepsilon) &= (f(V_n^2)/V_n^2)(f(V_n^2))^{-1} \sum_{k=1}^n EX_k^2 I(X_k^2 > \varepsilon^2 f(V_k^2)(V_k^2/f(V_k^2))) \\ &\leq (f(\sigma_1^2)/\sigma_1^2)(f(V_n^2))^{-1} \sum_{k=1}^n EX_k^2 I(X_k^2 \geq f(V_k^2)(\varepsilon\sigma_1^2/f(\sigma_1^2))). \end{aligned}$$

Hence, by (2.7), there exists an $\varepsilon > 0$ such that $L_n(\varepsilon) \rightarrow 0$ as $n \rightarrow \infty$; in fact, for every $\varepsilon \geq f(\sigma_1^2)/\sigma_1^2$

$$(2.8) \quad L_n(\varepsilon) \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Thus, by (2.6), (2.8) and Lemma 2 (ii), every sequence $\{X_n, n \geq 1\}$ satisfying the assumptions of Theorem 1 satisfies also the central limit theorem.

COROLLARY 1. *Under the assumptions of Theorem 1 with some $0 < \gamma < 1/4$, for every $q > 0$*

$$P\left(\lim_{n \rightarrow \infty} (\log V_n^2)^{-1} \sum_{k=1}^n (\sigma_k^2 S_k^{2q}/V_k^{2(q+1)}) = 2^q (\pi)^{-1/2} \Gamma(q + \frac{1}{2})\right) = 1$$

and

$$P\left(\lim_{n \rightarrow \infty} (\log V_n^2)^{-1} \sum_{k=1}^n (\sigma_k^2 S_k^2/V_k^4) = 1\right) = 1.$$

THEOREM 2. *Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, defined on (Ω, \mathcal{A}, P) , with $EX_n = 0$, $0 < EX_n^2 = \sigma^2 < \infty$, $n \geq 1$. If, for some $0 < r < 1$,*

$$(2.9) \quad \sum_{n=1}^{\infty} n^{-r} EX_n^2 I(|X_n| \geq \sigma n^{r/2}) < \infty,$$

then for every real function $a(x)$ which is a.e. continuous and $|a(x)| \leq \exp(\gamma x^2)$, $\gamma < 1/4$,

$$(2.10) \quad P\left(\lim_{n \rightarrow \infty} (\log n)^{-1} \sum_{k=1}^n k^{-1} a(S_k/\sigma k^{1/2}) = \int_{-\infty}^{\infty} a(x) d\Phi(x)\right) = 1.$$

COROLLARY 2. *Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables, defined on (Ω, \mathcal{A}, P) , with $EX_n = 0$, $0 < EX_n^2 = \sigma^2$ and $E|X_n|^{2+\delta} = \beta_{2+\delta} < \infty$, $n \geq 1$, for some $\delta > 0$. If $a(x)$ is a real function which is a.e. continuous and $|a(x)| \leq \exp(\gamma x^2)$ for some $\gamma < 1/4$, then (2.10) holds and for every $q > 0$*

$$P\left(\lim_{n \rightarrow \infty} (\log n)^{-1} \sum_{k=1}^n k^{-q-1} S_k^{2q} = (2\sigma^2)^q (\pi)^{-1/2} \Gamma(q + \frac{1}{2})\right) = 1.$$

3. Auxiliary lemmas. The proof of Theorem 1 is based on a martingale form of the Skorokhod representation theorem and on the Tomkins law of the iterated logarithm.

We present these results for the convenience of the reader.

LEMMA 1 (Strassen [11], Theorem 4.4). *Let $\{X_n, n \geq 1\}$ be a sequence of random variables such that, for all n , $E(X_n^2 | X_1, \dots, X_{n-1})$ is defined and $E(X_n | X_1, \dots, X_{n-1}) = 0$ P-a.s. Put*

$$S_n = \sum_{i \leq n} X_i \quad \text{and} \quad V_n = \sum_{i \leq n} E(X_i^2 | X_1, \dots, X_{i-1}),$$

where, in order to avoid trivial complications, we assume $V_1 = EX_1^2 > 0$. Let f be a positive, nondecreasing real function f on \mathbb{R}^+ such that the function $f(x)/x$ is nonincreasing on \mathbb{R}^+ . Assume that $V_n \rightarrow \infty$ P-a.s. as $n \rightarrow \infty$ and

$$(3.1) \quad \sum_{n=1}^{\infty} (f(V_n))^{-1} \int_{x^2 > f(V_n)} x^2 dP(X_n \leq x | X_1, \dots, X_{n-1}) < \infty \text{ P-a.s.}$$

Let S be the (random) function on $\mathbb{R}^+ \cup \{0\}$ obtained by interpolating S_n at V_n in such a way that $S(0) = 0$ and S is constant in each $\langle V_n, V_{n+1} \rangle$ (or, alternatively, is linear in each $\langle V_n, V_{n+1} \rangle$). Then without loss of generality there is a Brownian motion $\{W(t), t \geq 0\}$ such that, as $t \rightarrow \infty$,

$$(3.2) \quad S(t) = W(t) + o(\log t (tf(t))^{1/4}) \text{ P-a.s.}$$

LEMMA 2 (Tomkins [12], Theorem 3.1). *Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables such that $EX_n = 0$ and $EX_n^2 < \infty, n \geq 1$. Define, for $n \geq 1, S_n = X_1 + \dots + X_n, V_n^2 = ES_n^2, t_n^2 = 2 \log \log V_n^2$, and the Lindeberg function*

$$L_n(\varepsilon) = V_n^{-2} \sum_{k=1}^n EX_k^2 I(|X_k| > \varepsilon V_k), \quad \varepsilon > 0.$$

Suppose that

$$(3.3) \quad t_n X_n / V_n \rightarrow 0 \text{ a.s.} \quad \text{and} \quad V_n \rightarrow \infty \quad \text{as } n \rightarrow \infty.$$

(i) *The functions*

$$L_-(\varepsilon) = \liminf_{n \rightarrow \infty} L_n(\varepsilon) \quad \text{and} \quad L_+(\varepsilon) = \limsup_{n \rightarrow \infty} L_n(\varepsilon)$$

are both constant functions, and

$$(1 - L_+(\varepsilon))^{1/2} \leq \limsup_{n \rightarrow \infty} S_n / (V_n t_n) \leq (1 - L_-(\varepsilon))^{1/2} \text{ a.s. for every } \varepsilon > 0.$$

(ii) *If $\lim_{n \rightarrow \infty} L_n(\varepsilon) = 0$ for some $\varepsilon > 0$, then the CLT holds.*

(iii) *Let $EX_n^2 = o(V_n^2)$. If the CLT holds, then*

$$(3.4) \quad \limsup_{n \rightarrow \infty} S_n / (t_n V_n) = 1 \text{ P-a.s.}$$

4. Proofs of theorems. The symbol C , with or without subscripts, denotes a positive generic constant.

Proof of Theorem 1. Assume that $a(x) = I_{(-\infty, u)}(x)$ is the indicator function of the interval $(-\infty, u)$. Let $\{X_n, n \geq 1\}$ be a sequence of independent and normally distributed random variables with zero means and variance σ_n^2 , $n \geq 1$. Then S_n/V_n is normally distributed with zero mean and variance one.

Let, for $j < n$, $S_{j,n} = S_n - S_j$ and

$$(4.1) \quad g_{jn} = \mathbb{E} \{ (I_{(-\infty, u)}(S_j/V_j) - \Phi(u)) (I_{(-\infty, u)}(S_n/V_n) - \Phi(u)) \},$$

where, and in what follows, $\Phi(u) = \Phi((-\infty, u))$. Then $S_{j,n}$ is independent of S_j and normally distributed with zero mean and variance $V_{j,n}^2 = V_n^2 - V_j^2$. Furthermore,

$$(4.2) \quad S_n/V_n = (S_j/V_j)(V_j/V_n) + (S_{j,n}/V_{j,n})(V_{j,n}/V_n),$$

and so, by (4.1) and (4.2),

$$(4.3) \quad \begin{aligned} g_{jn} &= \mathbb{E} I_{(-\infty, u)}(S_j/V_j) I_{(-\infty, u)}(S_n/V_n) - \Phi^2(u) \\ &= P(S_j < uV_j; (S_{j,n}/V_{j,n}) < (uV_n - S_j)/V_{j,n}) - \Phi^2(u) \\ &= (2\pi)^{-1/2} \int_{-\infty}^u \exp(-x^2/2) \{ \Phi((uV_n - xV_j)/V_{j,n}) - \Phi(u) \} dx. \end{aligned}$$

On the other hand, by the inequalities (3.3) and (3.4) of Petrov [7], p. 161, for every x and u we get

$$(4.4) \quad \begin{aligned} &| \Phi((uV_n - xV_j)/V_{j,n}) - \Phi(u) | \\ &\leq (2\pi e)^{-1/2} (V_n/V_{j,n} - 1) + (2\pi)^{-1/2} |x| (V_j/V_{j,n}) \leq (2\pi)^{-1/2} (1 + |x|) (V_j/V_{j,n}) \end{aligned}$$

since $V_n/V_{j,n} - 1 \leq V_j/V_{j,n}$. Hence, by (4.3) and (4.4),

$$(4.5) \quad |g_{jn}| \leq (2\pi)^{-1} (V_j/V_{j,n}) \int_{-\infty}^u (1 + |x|) \exp(-x^2/2) dx \leq C (V_j/V_{j,n}),$$

where C is an absolute constant.

It is evident that $|g_{jn}| \leq 1$. Hence, by (4.5)

$$\begin{aligned} \mathbb{E} \left\{ \sum_{n=1}^N (\sigma_n^2/V_n^2) (I_{(-\infty, u)}(S_n/V_n) - \Phi(u)) \right\}^2 &\leq 2 \sum_{n=1}^N \sum_{j=1}^n (\sigma_n^2 \sigma_j^2 / V_n^2 V_j^2) |g_{jn}| \\ &\leq 2C \sum_{n=1}^N (\sigma_n^2/V_n^2) \sum_{j \in A_n} \{ \sigma_j^2 / (V_{j,n} V_j) \} \\ &\quad + 2 \sum_{n=1}^N (\sigma_n^2/V_n^2) \sum_{j \in B_n} \{ \sigma_j^2/V_j^2 \}, \end{aligned}$$

where $A_n = \{j: V_j^2 < V_n^2/2\}$ and $B_n = \{j \leq n: V_j^2 \geq V_n^2/2\}$.

Note also that

$$\sigma_j^2 / \{V_j^2 (V_n^2 - V_j^2)\}^{1/2} \leq \int_{v_{j-1}^2/V_n^2}^{v_j^2/V_n^2} \frac{dt}{\sqrt{t(1-t)}}$$

for every j such that $V_j^2 < V_n^2/2$. Hence

$$\sum_{j \in A_n} \{\sigma_j^2 / (V_{j,n} V_j)\} \leq \int_0^{1/2} \frac{1}{\sqrt{t(1-t)}} dt = \frac{\pi}{2}.$$

Now,

$$\sum_{j \in B_n} (\sigma_j^2 / V_j^2) \leq 2.$$

Consequently, combining the results from (4.3) down, we get

$$\begin{aligned} (4.6) \quad E \{(\log V_N^2)^{-1} \sum_{n=1}^N (\sigma_n^2 / V_n^2) (I_{(-\infty, u)}(S_n / V_n) - \Phi(u))\}^2 \\ \leq C_1 (\log V_N^2)^{-2} \sum_{n=1}^N (\sigma_n^2 / V_n^2) \leq C_1 (\log V_N^2)^{-2} (1 + \sum_{n=2}^N \int_{v_{n-1}^2/V_N^2}^{v_n^2/V_N^2} (1/x) dx) \\ = C_1 (\log V_N^2)^{-2} (1 + \int_{\sigma_1^2/V_N^2}^1 (1/x) dx) = C_1 (\log V_N^2)^{-2} (1 - \log \sigma_1^2 + \log V_N^2) \\ \leq C_2 (\log V_N^2)^{-1}. \end{aligned}$$

Define an increasing sequence of integers $\{N_k, k \geq 1\}$ by $V_{N_k}^2 \leq 2^{k^2} < V_{N_{k+1}}^2$. Since $\sigma_n^2 = o(V_n^2)$, as $n \rightarrow \infty$, entails $V_{n+1}^2 \sim V_n^2$, necessarily

$$V_{N_{k+1}}^2 = V_{N_k}^2 + \sigma_{N_{k+1}}^2 = V_{N_k}^2 + o(V_{N_k}^2),$$

and so $V_{N_k}^2 \sim 2^{k^2}$ as $k \rightarrow \infty$.

Hence we infer by combining Chebyshev's inequality, (4.6) and the Borel-Cantelli lemma that P -a.s.

$$(4.7) \quad (\log V_{N_k}^2)^{-1} \sum_{n=1}^{N_k} (\sigma_n^2 / V_n^2) (I_{(-\infty, u)}(S_n / V_n) - \Phi(u)) \rightarrow 0 \quad \text{as } k \rightarrow \infty.$$

On the other hand, for $N_k < N < N_{k+1}$ we have $(\log V_N^2)^{-1} \leq (\log V_{N_k}^2)^{-1}$ and

$$\begin{aligned} (\log V_N^2)^{-1} \left| \sum_{n=N_k+1}^N (\sigma_n^2 / V_n^2) (I_{(-\infty, u)}(S_n / V_n) - \Phi(u)) \right| \\ \leq (\log V_{N_k}^2)^{-1} \sum_{n=N_k+1}^{N_{k+1}} (\sigma_n^2 / V_n^2) \leq (\log V_{N_k}^2)^{-1} \sum_{n=N_k+1}^{N_{k+1}} \int_{v_{n-1}^2/V_{N_{k+1}}^2}^{v_n^2/V_{N_{k+1}}^2} (1/x) dx \\ = (\log V_{N_k}^2)^{-1} \{\log V_{N_{k+1}}^2 - \log V_{N_k}^2\} \leq Ck^{-1} \rightarrow 0 \quad \text{as } k \rightarrow \infty. \end{aligned}$$

Consequently, by (4.7), we get P -a.s.

$$(4.8) \quad (\log V_N^2)^{-1} \sum_{n=1}^N (\sigma_n^2/V_n^2) (I_{(-\infty, u)}(S_n/V_n) - \Phi(u)) \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Thus the case where $a(x) = I_{(-\infty, u)}(x)$ and X_n , $n \geq 1$, are normally distributed is considered.

If X_n , $n \geq 1$, are not normally distributed, then by Lemma 1

$$(4.9) \quad S_n - W(V_n^2) = \varepsilon_n(\omega) (V_n^2 f(V_n^2))^{1/4} (\log V_n^2) \quad P\text{-a.s.} \quad \text{as } n \rightarrow \infty,$$

where $\{W(t), t \geq 0\}$ is a standard Brownian motion and $\varepsilon_n(\omega) \rightarrow 0$ as $n \rightarrow \infty$ for almost all $\omega \in \Omega$.

Let

$$\eta_n = \eta_n(\omega) = \sup_{k \geq n} |\varepsilon_k(\omega)| (\log V_k^2) (f(V_k^2)/V_k^2)^{1/4}.$$

Then, by (4.9) and (2.2),

$$(4.10) \quad I_{(-\infty, u - \eta_n)}(W(V_n^2)/V_n) \leq I_{(-\infty, u)}(S_n/V_n) \leq I_{(-\infty, u + \eta_n)}(W(V_n^2)/V_n).$$

Let $\varepsilon > 0$ be given. Using (4.10) and (4.8), it is easy to see that

$$\begin{aligned} & (\log V_N^2)^{-1} \sum_{n=1}^N (\sigma_n^2/V_n^2) I_{(-\infty, u)}(S_n/V_n) \\ & \leq (\log V_N^2)^{-1} \left\{ \sum_{n=1}^M (\sigma_n^2/V_n^2) \sum_{n=M+1}^N (\sigma_n^2/V_n^2) I_{(-\infty, u + \eta_M)}(W(V_n^2)/V_n) \right\} \\ & \leq (1 + \log V_M^2 - \log \sigma_1^2) (\log V_N^2)^{-1} + \Phi(u + \eta_M) + \varepsilon \leq \Phi(u) + 2\varepsilon \end{aligned}$$

for sufficiently large N and suitable $M = M(N)$. Similarly, the left-hand sum can be bounded below, so that (4.8) is established for X_n , $n \geq 1$, not necessary normally distributed, too.

Let now $a(x) = \exp(\gamma x^2)$, $\gamma < 1/4$, and let $\{X_n, n \geq 1\}$ be a sequence of independent and normally distributed random variables with zero means and variance σ_n^2 , $n \geq 1$. Then

$$Ea(S_n/V_n) = (1 - 2\gamma)^{-1/2}.$$

We set

$$h_{jn} = E \left\{ (a(S_j/V_j) - (1 - 2\gamma)^{-1/2}) (a(S_n/V_n) - (1 - 2\gamma)^{-1/2}) \right\}.$$

Then, taking into account (4.2), we conclude that

$$(4.11) \quad h_{jn} = (2\pi)^{-1} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \exp \{ \gamma x^2 + \gamma [xV_j + yV_{j,n}]^2/V_n^2 - (x^2 + y^2)/2 \} dx dy - (1 - 2\gamma)^{-1}$$

$$\begin{aligned}
 &= (2\pi)^{-1} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \exp \left\{ \frac{\gamma}{V_n^2} (x^2 V_j^2 + 2xy V_j V_{j,n} - y^2 V_j^2) + (\gamma - \frac{1}{2})(x^2 + y^2) \right\} dx dy \\
 &\qquad\qquad\qquad - (1 - 2\gamma)^{-1} \\
 &= (2\pi)^{-1} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left[\exp \left\{ \frac{\gamma}{V_n^2} (x^2 V_j^2 + 2xy V_j V_{j,n} - y^2 V_j^2) \right\} - 1 \right] \\
 &\qquad\qquad\qquad \times \exp \left\{ (\gamma - \frac{1}{2})(x^2 + y^2) \right\} dx dy.
 \end{aligned}$$

Since $(xV_{j,n} - yV_j)^2 \geq 0$, we have

$$x^2 V_j^2 + 2xy V_j V_{j,n} - y^2 V_j^2 \leq x^2 V_n^2.$$

On the other hand, $|\exp(x) - 1| \leq |x|(\exp x + 1)$. Hence, for $j < n$

$$\begin{aligned}
 |h_{jn}| &\leq (2\pi)^{-1} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \left(\frac{|\gamma|}{V_n^2} \right) (x^2 V_j^2 + 2|x y| V_j V_{j,n} + y^2 V_j^2) \\
 &\qquad\qquad\qquad \times \exp \{ (2\gamma - 1/2)(x^2 + y^2) \} dx dy \\
 &\leq C(V_j^2/V_n^2 + V_j V_{j,n}/V_n^2),
 \end{aligned}$$

so that for every $1 \leq j \leq n, n \geq 1, |h_{jn}| \leq C$. Furthermore, by the same inequality we get $|h_{jn}| \leq C(V_j/V_{j,n})$.

Thus, as in the case $a(x) = I_{(-\infty, u)}(x)$, we get *P*-a.s.

$$(\log V_{N_k}^2)^{-1} \sum_{n=1}^{N_k} (\sigma_n^2/V_n^2) \{ \exp(\gamma S_n^2/V_n^2) - (1 - 2\gamma)^{-1/2} \} \rightarrow 0 \quad \text{as } k \rightarrow \infty,$$

where $\{N_k, k \geq 1\}$ is the sequence defined above. Moreover, by the law of the iterated logarithm, Theorem 1.106 of Freedman [4], we have *P*-a.s.

$$(4.12) \quad \exp(\gamma S_n^2/V_n^2) \leq (\log V_n^2)^{\gamma + 1/4}$$

for sufficiently large n . Hence for $N_k < N < N_{k+1}$ we have $(\log V_N^2)^{-1} \leq (\log V_{N_k}^2)^{-1}$ and

$$\begin{aligned}
 &(\log V_{N_k}^2)^{-1} \sum_{n=N_k+1}^{N_{k+1}} (\sigma_n^2/V_n^2) | \exp(\gamma S_n^2/V_n^2) - (1 - 2\gamma)^{-1/2} | \\
 &\qquad\qquad\qquad \leq C(\log V_{N_k}^2)^{\gamma - 3/4} \sum_{n=N_k+1}^{N_{k+1}} (\sigma_n^2/V_n^2) \\
 &\qquad\qquad\qquad \leq C(\log V_{N_k}^2)^{\gamma - 3/4} (\log V_{N_{k+1}}^2 - \log V_{N_k}^2) \leq C_1 k^{2\gamma - 1/2}.
 \end{aligned}$$

Thus *P*-a.s.

$$(\log V_N^2)^{-1} \sum_{n=1}^N (\sigma_n^2/V_n^2) \{ \exp(\gamma S_n^2/V_n^2) - (1 - 2\gamma)^{-1/2} \} \rightarrow 0 \quad \text{as } N \rightarrow \infty.$$

If X_n , $n \geq 1$, are random variables not normally distributed, then

$$(4.13) \quad \left| \exp(\gamma S_n^2/V_n^2) - \exp(\gamma W^2(V_n^2)/V_n^2) \right| \\ \leq (2|\gamma|/V_n^2) |S_n - W(V_n^2)| \max \{ |S_n| \exp(\gamma S_n^2/V_n^2), |W(V_n^2)| \exp(\gamma W^2(V_n^2)/V_n^2) \}.$$

But, under the assumptions of Theorem 1, (2.6) and (2.8) hold, so that by Lemma 2 (ii) and (iii)

$$(4.14) \quad \limsup_{n \rightarrow \infty} \frac{|S_n|}{(2V_n^2 \log \log V_n^2)^{1/2}} = 1 \text{ P-a.s.}$$

Thus, by (4.9), Theorem 1.106 of Freedman [4] and (4.14), the right-hand side of inequality (4.13) can be bounded by

$$4|\gamma\varepsilon_n| (f(V_n^2)/V_n^2)^{1/4} (\log V_n^2)^{(3\gamma \vee 2\gamma)+1}$$

for sufficiently large n . Since (2.3) holds and $\varepsilon_n \rightarrow 0$ P-a.s., as $n \rightarrow \infty$, so that (2.5) also holds for $a(x) = \exp(\gamma x^2)$, $\gamma < 1/4$.

Let now $a(x)$ be a function satisfying the assumptions of Theorem 1. Then, similarly to Schatte [10], we introduce an auxiliary function $a_1(x)$ which vanishes for $|x| > K$ and is in each of the intervals

$$-K + 2iK/L \leq x < -K + 2(i+1)K/L, \quad i = 0, 1, 2, \dots, L-1,$$

equal to the supremum of $a(x) - \exp(\gamma x^2)$ in these intervals. Let $a_2(x) = a_1(x) + \exp(\gamma x^2)$ and choose first K and then L large enough so that

$$\int_{-\infty}^{\infty} a_2(x) d\Phi(x) \leq \int_{-\infty}^{\infty} a(x) d\Phi(x) + \varepsilon/2.$$

This is possible since $a(x)$ is continuous a.e. and, therefore, Riemann-Stieltjes integrable with respect to $\Phi(x)$. Obviously, $a(x) \leq a_2(x)$ for every real number x . The function $a_2(x)$ is a finite linear combination of the special functions already considered in the proof. Thus

$$(\log V_N^2)^{-1} \sum_{n=1}^N (\sigma_n^2/V_n^2) a(S_n/V_n) \leq (\log V_N^2)^{-1} \sum_{n=1}^N (\sigma_n^2/V_n^2) a_2(S_n/V_n) \\ \leq \int_{-\infty}^{\infty} a_2(x) d\Phi(x) + \varepsilon/2 \leq \int_{-\infty}^{\infty} a(x) d\Phi(x) + \varepsilon$$

for sufficiently large N and almost all ω . Replacing $a(x)$ by $-a(x)$ we obtain the assertion of Theorem 1.

Proof of Theorem 2. Let us observe that, under assumptions of Theorem 2, $V_n^2 = \sigma^2 n$, $n \geq 1$. On the other hand, for every $0 < r < 1$, the function $f(x) = |x|^r$ satisfies the assumptions of Theorem 1. Thus Theorem 2 is a consequence of Theorem 1.

Proof of Corollary 2. Let us observe that for every $\delta > 0$ there exists an r such that $0 < r < 1$ and (2.9) holds. In fact, for every $n \geq 1$

$$EX_n^2 I(|X_n| \geq \sigma^r n^{r/2}) \leq \sigma^{-r\delta} n^{-r\delta/2} \beta_{2+\delta}.$$

Thus it is enough to take $2/(2+\delta) < r < 1$, so that Corollary 2 is a consequence of Theorem 2.

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