

On the Speed of Convergence in the Random Central Limit Theorem for φ -Mixing Processes

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1. Introduction

Let X_n , $n \in \mathbb{N}$, be a sequence of random variables on a probability space (Ω, \mathcal{B}, P) . The σ -algebra generated by X_n with $a \leq n \leq b$ is denoted by \mathcal{M}_a^b .

Suppose there exists a sequence $\varphi(k)$, $k \in \mathbb{N}$, of real numbers with $1 \geq \varphi(k) \downarrow 0$ for $k \rightarrow \infty$ such that

$$|P(E_1 \cap E_2) - P(E_1)P(E_2)| \leq \varphi(k)P(E_1)$$

for all $l, k \in \mathbb{N}$, $E_1 \in \mathcal{M}_1^l$, $E_2 \in \mathcal{M}_{l+k}^\infty$, then the sequence X_n , $n \in \mathbb{N}$, is called φ -mixing.

In this paper uniform and non-uniform bounds in the random central limit theorem for φ -mixing processes are derived. The bounds are very near to those given by Landers and Rogge [9, 10] in the independent case. Of course the results are based on corresponding theorems for non-random summation. These theorems are obtained by modifying and developing methods of Tihomirov [16], Erickson [4], Babu, Ghosh and Singh [1].

The following notations are used:

If X is a random variable the measure induced by X is denoted by $P * X$.

Put $\sigma_n^2 := \text{Var} \sum_{i=1}^n X_i$, and for $x \in \mathbb{R}$

$$[x] := \max(1, \max\{n \in \mathbb{N} : n \leq x\}),$$
$$\psi(x) := (2\pi)^{-1/2} \exp(-x^2/2), \quad \Phi(x) := \int_{-\infty}^x \psi(t) dt.$$

Throughout the paper D denotes a generic constant.

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2. Non-Random Summation

For stationary processes the following theorem has been proved by A.N. Tihomirov ([16], Theorem 3). Therefore, we have only to show how in Tihomirov's paper stationarity can be substituted by our weaker conditions.

Theorem 1. *Let $X_n, n \in \mathbb{N}$, be a φ -mixing sequence with*

$$EX_n = 0, \quad n \in \mathbb{N} \tag{1}$$

$$\sup_{n \in \mathbb{N}} E|X_n|^3 < \infty \tag{2}$$

$$\liminf_{n \in \mathbb{N}} \sigma_n^2/n > 0 \tag{3}$$

$$\varphi(n) \leq D \exp(-\lambda n), \quad n \in \mathbb{N}, \quad \text{for some } \lambda > 0. \tag{4}$$

Then for all $n > 1$

$$\sup_{t \in \mathbb{R}} \left| P \left\{ \sum_{i=1}^n X_i < \sigma_n t \right\} - \Phi(t) \right| \leq D n^{-1/2} \log n. \tag{5}$$

Proof. Using our assumptions (2) and (3) and Lemma 2 we can renounce stationarity in Tihomirov's proof. Difficulties arise only in the proof of Tihomirov's Lemma 3.2. (Like in [16] we first assume m -dependence.)

We show now how we can avoid this Lemma. We adopt the notation of Tihomirov. Lemma 3.2 and 3.3 are used only to give a bound for

$$L := \left| \sum_{j=1}^n EX_j \prod_{l=1}^{r-1} \xi_j^{(l)} e^{itS_j^{(r)}} - \sum_{j=1}^n EX_j \prod_{l=1}^{r-1} \xi_j^{(l)} f_n(t) \right|.$$

But this can be achieved directly without using stationarity. If we set $a_j = EX_j \prod_{l=1}^{r-1} \xi_j^{(l)}$ we have by [17] Lemma 3.1

$$|a_j| \leq D \left(\frac{|t| D \sqrt{m}}{\sigma_n} \right)^{r-1}$$

and therefore

$$\begin{aligned} L &= \left| E \exp(itS_n) \sum_{j=1}^n a_j \eta_j^{(r)} \right| \\ &\leq \left| E \exp(itS_n) \sum_{j=1}^n a_j E \eta_j^{(r)} \right| + \left| E \exp(itS_n) \sum_{j=1}^n a_j (\eta_j^{(r)} - E \eta_j^{(r)}) \right| \\ &\leq D |f_n(t)| \cdot n \cdot \left(\frac{|t| D \sqrt{m}}{\sigma_n} \right)^{r-1} \cdot \left| \frac{t}{\sigma_n} \sqrt{rm} + E \left| \sum_{j=1}^n a_j (\eta_j^{(r)} - E \eta_j^{(r)}) \right| \right|. \end{aligned}$$

Here the second summand is bounded by (cf. [16], p. 806, 807)

$$\left(\sum_{j=1}^n \sum_{|p-j| \leq 3rm} \text{cov}(a_j \eta_j^{(r)}, a_p \eta_p^{(r)}) \right)^{1/2} \leq D \left(\frac{|t| D \sqrt{m}}{\sigma_n} \right)^{r-1} \sqrt{n} \cdot \left| \frac{t}{\sigma_n} \right| rm.$$

Like in [16] it is easy to deduce the corresponding bound for φ -mixing variables.

The next theorem is a specification of a result of Erickson [4].

Theorem 2. *Let $X_n, n \in \mathbb{N}$, be a φ -mixing sequence fulfilling (1) and (4). Further let*

$$\sup_{n \in \mathbb{N}} E|X_n|^s < \infty \quad \text{for some real } s > 2, \tag{6}$$

$$\inf_{n \in \mathbb{N}} \sigma_n^2/n > 0. \tag{7}$$

Set $X_i^{(n)} = X_i I\{|X_i| \leq \sqrt{n}\}$.

Then for all $n > 1, t \in \mathbb{R}$

$$\left| P\left\{ \sum_{i=1}^n X_i < \sigma_n t \right\} - P\left\{ \sum_{i=1}^n X_i^{(n)} < \sigma_n t \right\} \right| \leq D n^{1-s/2} (1+|t|)^{-s} (\log n)^s. \tag{8}$$

Erickson only gives a proof for the d -dependent case. It is not very difficult to extend his proof, if the following hints are observed. (cf. [4], Sect. 6).

Instead of [4] (2.2) set

$$K(x) := 1 - P(U > x, \cap B_k^c).$$

The terms T_2, T_4, T_6 in the bound given in [4] Proposition 3.1 disappear, if the last equation in the proof of the proposition is replaced by

$$|s_k|^m I(d_k) - |s_{k-1}|^m I(d_{k-1}) = |s_k|^m I(d_{k-1}^c) I(b_k) + (|s_k|^m - |s_{k-1}|^m) I(d_{k-1}).$$

This shortens the proof a great deal.

Use our Lemma 2(i) instead of Erickson's Proposition 5.2 and apply Lemma 1 where Erickson uses the d -dependence. A detailed proof of Theorem 2 can be found in [14].

With the aid of Theorem 2 we obtain a nonuniform bound in the central limit theorem by altering a proof of Babu, Ghosh, Singh [1].

Theorem 3. *Let $X_n, n \in \mathbb{N}$, be a φ -mixing sequence satisfying (1), (4), (6) and (7).*

Then there exists a constant $d > 0$ so that for all $n > 1, t \in \mathbb{R}$, with $t^2 \geq d \log n$

$$\left| P\left\{ \sum_{i=1}^n X_i < \sigma_n t \right\} - \Phi(t) \right| \leq D n^{1-s/2} |t|^{-s} (\log n)^s. \tag{9}$$

Proof. Let $c = s - 2, c' = \min(1, c)$ and define $X_i^{(n)}$ like in Theorem 2. Assume w.l.o.g. that $t > 0$.

We have by [5], p. 175, Lemma 2

$$\Phi(-t) \leq D n^{-c/2} t^{-2-c} \quad \text{if } t^2 > (c+1) \log n \tag{10}$$

and by (1) and (7)

$$\left| \sum_{i=1}^n E X_i^{(n)} / \sigma_n \right| \leq D n^{-c/2}. \tag{11}$$

Using these inequalities, Theorem 2 and (7) we see that it suffices to show

$$P \left\{ \sum_{i=1}^n (X_i^{(n)} - EX_i^{(n)}) \geq 3tn^{1/2} \right\} \leq Dn^{-c/2} t^{-2-c} \tag{12}$$

for all $n > 1$, $t^2 \geq d \log n$ where $d > 1$ can be chosen later. (The factor $(\log n)^s$ in (9) is only caused by Theorem 2.)

To prove (12) we use Lemma 2(i). We proceed in the same way as Babu, Ghosh and Singh in the proof of Lemma 3 [1]. So we adopt their notation and only indicate the changes to be made.

We set

$$\begin{aligned} X'_i &= X_i^{(n)} - EX_i^{(n)}, \\ y &= 12(c+1)c'^{-1}t^{-1}n^{-1/2}, \\ \xi_j^* &= \xi_j I\{\xi_j < 1/y\}. \end{aligned}$$

Since $\sum_{i=1}^n X'_i \leq n^{3/2}$ we can assume that $t \leq n$ and so we can replace [1] (3.5) by

$$\begin{aligned} P\{U_n^* > tn^{1/2}\} &\leq \exp(-(2c+2)\log n) E \exp(zU_n^*) \\ &\leq n^{-c/2} t^{-c-2} n^{-c/2} \prod_{j=1}^k s_j \quad (\text{use [1], Lemma 2}) \end{aligned}$$

where

$$z := z(n, t) = (2c+2)t^{-1}n^{-1/2}\log n$$

and

$$s_j := 2 \exp((c'/6)\log n) \varphi(p) + E \exp(z\xi_j^*).$$

Obviously the first summand of s_j is smaller than Dk^{-1} , and the second is according to Lemma 4 bounded by

$$\begin{aligned} P\{|\xi_j| \geq 1/y\} + E(I\{|\xi_j| < 1/y\} \exp(z\xi_j)) \\ \leq y^{2c+2} E|\xi_j|^{2c+2} + 1 + z^2 E\xi_j^2/2 + E|\xi_j|^{2+c'} y^{2+c'} \exp(2z/y). \end{aligned}$$

Here the first and the last summand are bounded by Dk^{-1} .

Since $t^2 \geq d \log n$ we have

$$z^2 \leq (D/d)n^{-1}\log n$$

yielding

$$z^2 E\xi_j^2 \leq (D/d)k^{-1}\log k.$$

Therefore (use $x^n \leq \exp(n(x-1))$)

$$\prod_{j=1}^k s_j \leq (1 + Dk^{-1} + (D/d)k^{-1}\log k)^k \leq k^{D/d} \leq Dn^{c/2}$$

if d is chosen large enough.

Now we have

$$P\{U_n^* > tn^{1/2}\} \leq Dn^{-c/2} t^{-c-2}$$

and this yields the assertion like in [1].

Combining Theorem 1 and Theorem 3 we get the following result (cf. [6], (2.4), (2.5) and [16], Theorem 4)

Theorem 4. *Let the assumptions of Theorem 3 be fulfilled for some $s \geq 3$. Then for all $t \in \mathbb{R}$, $n > 1$*

$$\left| P \left\{ \sum_{i=1}^n X_i < \sigma_n t \right\} - \Phi(t) \right| \leq D n^{-1/2} (1 + |t|)^{-s} (\log n)^{1+s'/2}$$

where $s' = s$ if $s > 3$ and $s' = s + 1$ if $s = 3$.

3. Random Summation

We are now ready to prove two theorems about the speed of convergence in the random central limit theorem for φ -mixing processes. The only result in this direction is due to B.L.S. Prakasa Rao [12]. His technique is different from ours and the order of convergence he reaches is far from the order in the independent case.

First we give a φ -mixing version of a theorem of Landers and Rogge [9]. Their result gives the exact rate of convergence under independence (see [8]). Examining their proof, we see that one of the main tools, namely Lemma 7 of [8], cannot be transferred to φ -mixing processes because the proof heavily uses independence and stationarity. (The last fact has not been noticed by Rychlik [13] and so his proof is not correct in this place, see [13], p. 233.)

Lemma 7 of [8] allows to replace $\max_{p < n \leq q} \sum_{i=p+1}^n X_i$ by $\sum_{i=p+1}^q X_i$ in a certain situation.

This lemma can be avoided by retaining the maximum and using Serfling's [15] inequality for maxima of sums in the proof of Landers' and Rogge's Lemma 8 (see [8], p. 282, (*)).

The order of approximation we obtain differs from the order in the independent case only by a logarithmic factor. If the assumptions are strengthened a little bit, the factor disappears.

Theorem 5. *Let the assumptions of Theorem 1 be fulfilled. Let ε_n , $n \in \mathbb{N}$, be a sequence with $n^{-1} \leq \varepsilon_n < 1$ and $\varepsilon_n \xrightarrow{n \rightarrow \infty} 0$.*

Let τ_n , $n \in \mathbb{N}$, be positive integer valued random variables. Assume that τ is a positive random variable independent of $(X_n)_{n \in \mathbb{N}}$ so that one of the two following conditions is fulfilled.

(i)
$$P\{\tau < c_0 / (n\varepsilon_n)\} \leq D \delta_n, \quad n \in \mathbb{N}, \tag{13}$$

for some constant $c_0 > 0$ where $\delta_n := \sqrt{\varepsilon_n} (\log \varepsilon_n)^2$.

(ii) *There exists $\varepsilon > 0$ so that $\varepsilon_n \geq n^{-1+\varepsilon}$, $n \in \mathbb{N}$, and*

$$P\{\tau < c_0 / (n\varepsilon_n^{1+\varepsilon})\} \leq D \delta_n, \quad n \in \mathbb{N}, \tag{14}$$

for some constant $c_0 > 0$ where $\delta_n := \sqrt{\varepsilon_n}$.

Suppose further that for some $c_1 > 0$

$$P\{|\tau_n/(n\tau) - 1| > c_1 \varepsilon_n\} \leq D \delta_n, \quad n \in \mathbb{N}.$$

Then

$$\sup_{t \in \mathbb{R}} \left| P \left\{ \sum_{i=1}^{\tau_n} X_i < \sigma_{[n\tau]} t \right\} - \Phi(t) \right| \leq D \delta_n, \quad n \in \mathbb{N}.$$

Proof. Let $n \in \mathbb{N}$ be sufficiently large and $t \in \mathbb{R}$. In case (i) we set $\gamma_n := [c_0/\varepsilon_n]$ and in case (ii) $\gamma_n := [c_0/\varepsilon_n^{1+\varepsilon}]$.

Therefore in either case

$$P\{n\tau < \gamma_n\} \leq D \delta_n.$$

In case (i) we have

$$\varepsilon_n^{1/2} \log(D/\varepsilon_n) \leq D \varepsilon_n^{1/2} |\log \varepsilon_n| \tag{15}$$

and in case (ii)

$$\varepsilon_n^{(1+\varepsilon)/2} \log(D/\varepsilon_n^{1+\varepsilon}) \leq D \varepsilon_n^{1/2}. \tag{16}$$

Thus

$$\gamma_n^{-1/2} \log \gamma_n \leq D \delta_n.$$

Using Theorem 1 one gets, like in [9], p. 1021,

$$\begin{aligned} & \left| P \left\{ \sum_{i=1}^{[n\tau]} X_i < \sigma_{[n\tau]} t \right\} - \Phi(t) \right| \\ & \leq P\{n\tau < \gamma_n\} + \sum_{l=\gamma_n}^{\infty} DP\{[n\tau]=l\} l^{-1/2} \log l \leq D \delta_n. \end{aligned} \tag{17}$$

Set $S_0 = 0$, $S_j = \sum_{i=1}^j X_i$, $j \in \mathbb{N}$,

$$p_x = [x(1 - c_1 \varepsilon_n)], \quad q_x = [x(1 + c_1 \varepsilon_n)], \quad x > 0.$$

(17) yields like in [9], the assertion, if it is shown that

$$P\left\{ \min_{p_{n\tau} \leq j \leq q_{n\tau}} S_j < \sigma_{[n\tau]} t \right\} - P\left\{ \max_{p_{n\tau} \leq j \leq q_{n\tau}} S_j < \sigma_{[n\tau]} t \right\} \leq D \delta_n.$$

This difference is bounded from above by

$$P\{n\tau < \gamma_n\} + \int_{\gamma_n}^{\infty} P\left\{ \min_{p_x \leq j \leq q_x} S_j < \sigma_{[x]} t \right\} - P\left\{ \max_{p_x \leq j \leq q_x} S_j < \sigma_{[x]} t \right\} P^*(n\tau) dx.$$

For $p < q$, $r \in \mathbb{R}$ we have

$$P\left\{ \min_{p \leq j \leq q} S_j < r \right\} - P\left\{ \max_{p \leq j \leq q} S_j < r \right\} = P\{S_p < r \leq \max_{p \leq j \leq q} S_j\} + P\left\{ \min_{p \leq j \leq q} S_j < r \leq S_p \right\}.$$

Since we can replace X_i by $-X_i$ it apparently suffices to show for $r \in \mathbb{R}$, $x \geq \gamma_n$, $p := p_x$, $q := q_x$ that

$$P\{S_p \leq r \leq \max_{p \leq j \leq q} S_j\} \leq D \delta_n,$$

where D does not depend on n, r, x .

Set

$$m := \lceil -(2\lambda)^{-1} \log \varepsilon_n \rceil, \quad \text{then } \varphi(m) \leq D \sqrt{\varepsilon_n}.$$

Let $k \in \mathbb{N}$, $\eta > 0$ to be specified later, fulfilling

$$(2k+1)m \leq p/2. \tag{18}$$

We now show

$$P\{S_p \leq r \leq \max_{p \leq j \leq q} S_j\} \leq Dk(\sqrt{\varepsilon_n} + p^{-1/2}(\eta + \log p)) + (D(-\log \varepsilon_n)^{3/2}/\eta^3)^{k+1} \tag{19}$$

where D is independent of η and k , too.

Setting

$$L_j := S_{p-(j-1)m} - S_{p-jm}, \quad H := \max_{p \leq j \leq q} (S_j - S_p),$$

one obtains

$$\begin{aligned} P\{S_p \leq r \leq \max_{p \leq j \leq q} S_j\} &\leq P\{r - H \leq S_p \leq r\} \\ &\leq P\{|L_j| > \eta \text{ for all } j=1, \dots, 2k+1\} + \sum_{j=1}^{2k+1} P\{r - H \leq S_p \leq r, |L_j| \leq \eta\} \\ &\leq P\left(\bigcap_{i=0}^k \{|L_{2i+1}| > \eta\}\right) + \sum_{j=1}^{2k+1} P\{r - \eta - H \leq S_{p-jm} + S_p - S_{p-(j-1)m} \leq r + \eta\}. \end{aligned}$$

Using the φ -mixing property and Lemma 2(ii), the first probability can be bounded by

$$k\varphi(m) + \prod_{i=0}^k P\{|L_{2i+1}| > \eta\} \leq Dk\sqrt{\varepsilon_n} + (D(-\log \varepsilon_n)^{3/2}/\eta^3)^{k+1}.$$

Let $j \in \{1, \dots, 2k+1\}$ and set $Z := S_p - S_{p-(j-1)m}$.

According to (3) and (18)

$$\sigma_{p-jm}^2 \geq D(p-jm) \geq Dp.$$

Since S_{p-jm} is \mathcal{M}_1^{p-jm} -measurable and (Z, H) is $\mathcal{M}_{p-(j-1)m+1}^q$ -measurable we obtain, using Lemma 3 and Theorem 1,

$$\begin{aligned} &P\{r - \eta - H \leq S_{p-jm} + Z \leq r + \eta\} - \varphi(m) \\ &\leq \int P\{r - \eta - h - z \leq S_{p-jm} \leq r + \eta - z\} P^*(Z, H) dz dh \\ &\leq D(p-jm)^{-1/2} \log(p-jm) \\ &\quad + \int |\Phi((r + \eta - z)/\sigma_{p-jm}) - \Phi((r - \eta - h - z)/\sigma_{p-jm})| P^*(Z, H) dz dh \\ &\leq D(p/2)^{-1/2} \log p + (2\eta + EH)/\sigma_{p-jm} \\ &\leq Dp^{-1/2}(\eta + \log p + EH). \end{aligned}$$

As $EH \leq D(q-p)^{1/2}$ according to Lemma 2(iii) and $(q-p)/p \leq D\varepsilon_n$ we get (19).

Now we choose k and η .

In case (i) we set $k = \lceil -\log \varepsilon_n \rceil$. Since

$$(2k+1)m \leq D(\log \varepsilon_n)^2 \quad \text{and} \quad p \geq D\gamma_n \geq D/\varepsilon_n,$$

(18) is valid.

Let $\eta := \alpha^{1/3}(-\log \varepsilon_n)^{1/2}$ where α is chosen such that

$$D(-\log \varepsilon_n)^{3/2}/\eta^3)^{k+1} \leq (D/\alpha)^{-\log \varepsilon_n} = \varepsilon_n^{-\log(D/\alpha)} = \varepsilon_n^{1/2}.$$

Using (15) it is now easy to see that the bound in (19) has the order $\sqrt{\varepsilon_n}(\log \varepsilon_n)^2$.

In case (ii) we choose $k \in \mathbb{N}$ so large that $3(k+1)\varepsilon > 1$ and set $\eta := \varepsilon_n^{-\varepsilon/2}$. Then (18) is fulfilled and

$$D(-\log \varepsilon_n)^{3/2}/\eta^3)^{k+1} \leq D(-\log \varepsilon_n)^{3(k+1)/2} \varepsilon_n^{3(k+1)\varepsilon/2} \leq D\varepsilon_n^{1/2}.$$

Using this and (16), we see that (19) yields the desired order of convergence.

Remark. a) If τ is constant, the condition (13) respectively (14) is fulfilled, and in case (i) we obtain the order

$$\delta_n = |\log \varepsilon_n|(\sqrt{\varepsilon_n} + n^{-1/2} \log n), \quad \text{if we set } \gamma_n = \lceil n\tau/2 \rceil.$$

b) With the additional assumption $\sup_{n \in \mathbb{N}} \|X_n\|_\infty =: M < \infty$ we obtain in case (i) the order $\sqrt{\varepsilon_n} |\log \varepsilon_n|$ if we set $k=0$ and $\eta = c|\log \varepsilon_n|$ where c is a constant so that $mM \leq \eta$. If τ is furthermore constant, we obtain the order $\sqrt{\varepsilon_n} + n^{-1/2} \log n$.

In the following theorem a non uniform bound is derived, corresponding to another result of Landers and Rogge [10]. For the sake of brevity it is assumed that τ is constant. It is not difficult to weaken this assumption like in the preceding theorem (cf. [7]). The moment condition required by Landers and Rogge is somewhat surprising. But it was shown by A. Klein [7] that the theorem becomes wrong, if only the existence of lower moments is assumed.

Theorem 6. Let $X_n, n \in \mathbb{N}$, be a φ -mixing sequence fulfilling (1), (4), (7) and

$$\sup_{n \in \mathbb{N}} E|X_n|^{s+1} < \infty \quad \text{for some } s \geq 2.$$

Let $\varepsilon_n, n \in \mathbb{N}$, be a sequence with $n^{-1} \leq \varepsilon_n < 1$ and $\varepsilon_n \xrightarrow[n \rightarrow \infty]{} 0$.

Set $\delta_n := (\sqrt{\varepsilon_n} + n^{-1/2}(\log n)^{1+(s'+1)/2}) |\log \varepsilon_n|$ where $s' = s$ if $s > 2$ and $s' = s+1$ if $s = 2$.

If $\varepsilon_n \geq n^{-1+\varepsilon}, n \in \mathbb{N}$, for some $\varepsilon > 0$ also $\delta_n := \sqrt{\varepsilon_n}$ is allowed.

Let $\tau_n, n \in \mathbb{N}$, be positive integer valued random variables with

$$P\{\tau_n/(n\tau) - 1 > t\varepsilon_n\} \leq D\delta_n t^{-s}, \quad n \in \mathbb{N}, t \geq t_0,$$

for some constants $\tau > 0, t_0 > 0$.

Then for all $n > 1, t \in \mathbb{R}$

$$\left| P \left\{ \sum_{i=1}^{\tau_n} X_i < t \sigma_{[n\tau]} \right\} - \Phi(t) \right| \leq D \delta_n (1 + |t|)^{-s} (\log(2 + |t|))^{s+1}. \quad (20)$$

Proof. Define S_n like before. Let

$$\begin{aligned} p &= p(n, t) = [n\tau(1 - \varepsilon_n |t|)] \\ q &= q(n, t) = [n\tau(1 + \varepsilon_n |t|)] \\ I_n(t) &= \{k \in \mathbb{N} : p \leq k \leq q\}. \end{aligned}$$

With a view to the remark after Theorem 5 we can assume that $|t| \geq t_1$ for some constant $t_1 \geq t_0$. Then

$$Dn\varepsilon_n |t| \leq q - p \leq Dn\varepsilon_n |t|. \quad (21)$$

Set $m := [\lambda^{-1} \log(n^{1/2} |t|^s)]$, then $\varphi(m) \leq Dn^{-1/2} |t|^{-s}$.

According to (7) and Lemma 2 we have $Dn \leq \sigma_n^2 \leq Dn$.

(i) Let $n > 1, t \in \mathbb{R}$ and $t_1 \leq |t| \leq n^{1/(2s)}$.

First we proceed like in the proof of Theorem 1 in [10] (2)–(6). Instead of Petrov's theorem we use Theorem 4. Like in the proof of Theorem 5 we see that it suffices to estimate

$$R := P \{ S_p \leq t \sigma_{[n\tau]} \leq \max_{p \leq j \leq q} S_j \}.$$

Since $|t| \leq n^{1/(2s)}$ we have

$$m \leq D \log n. \quad (22)$$

We now consider two cases.

Case 1. $|t| \leq \varepsilon_n^{-(s-1)/s}$.

Let $0 < \eta \leq |t| \sigma_{[n\tau]}/6, k \in \mathbb{N}$ to be specified later.

We show that there exists a constant D independent of n, t, η, k with

$$\begin{aligned} R &\leq Dk(n^{-1/2} (\log n)^{1+(s'+1)/2} + \sqrt{\varepsilon_n} + k^{s/2}/n^{1/2} \\ &\quad + \eta/(|t| \sqrt{n})/|t|^s + (D((\log n)^{1/2}/\eta)^s)^{k+1}. \end{aligned} \quad (23)$$

For $H := \max_{p \leq j \leq q} (S_j - S_p)$ we have according to Lemma 2(iii) and (21)

$$P \{ H \geq |t| \sigma_{[n\tau]}/6 \} \leq D(q-p)^{s/2}/(|t| \sigma_{[n\tau]})^s \leq D\sqrt{\varepsilon_n}/|t|^s \quad (24)$$

since $|t| \leq \varepsilon_n^{-(s-1)/s}$.

If $p \leq (2k+1)m$ we have by Lemma 2(ii) and (22)

$$P \{ |S_p| \geq |t| \sigma_{[n\tau]}/2 \} \leq Dp^{s/2}/(|t| \sigma_{[n\tau]})^s \leq Dk^{s/2}/(|t|^s n^{1/2}). \quad (25)$$

For $t < 0$ this yields (23).

For $t > 0$

$$R \leq P \{ |S_p| \geq |t| \sigma_{[n\tau]}/2 \} + P \{ H \geq |t| \sigma_{[n\tau]}/2 \}.$$

Then (23) follows from (24) and (25).

Let $p > (2k + 1)m$.

Setting $L_j := S_{p-(j-1)m} - S_{p-jm}$ one obtains

$$R \leq P \left(\bigcap_{i=0}^k \{ |L_{2i+1}| > \eta \} \right) + \sum_{j=1}^{2k+1} P \{ t \sigma_{[n\tau]} - H - \eta \leq S_{p-jm} + S_p - S_{p-(j-1)m} \leq t \sigma_{[n\tau]} + \eta \}. \tag{26}$$

Like in the foregoing proof we can bound the first probability by $Dkn^{-1/2} |t|^{-s} + (D((\log n)^{1/2}/\eta)^s)^{k+1}$.

Let $j \in \{1, \dots, 2k + 1\}$, $Z := S_p - S_{p-(j-1)m}$

$$F_n(t) := \{ \omega : H(\omega) \geq |t| \sigma_{[n\tau]}/6 \text{ or } |Z(\omega)| \geq |t| \sigma_{[n\tau]}/6 \} \\ h_1(\omega) := t \sigma_{[n\tau]} - H(\omega) - \eta - Z(\omega), \quad h_2(\omega) := t \sigma_{[n\tau]} + \eta - Z(\omega).$$

For $\omega \in F_n(t)^c$ (the complement of $F_n(t)$)

$$|h_1(\omega)| \geq |t| \sigma_{[n\tau]}/2, \quad |h_2(\omega)| \geq |t| \sigma_{[n\tau]}/2. \tag{27}$$

Thus in view of Lemma 3 and Theorem 4 the j -th summand in (26) is bounded by

$$\varphi(m) + P(F_n(t)) + \int_{F_n(t)^c} P \{ h_1(\omega) \leq S_{p-jm} \leq h_2(\omega) \} P d\omega \\ \leq Dn^{-1/2} |t|^{-s} + P(F_n(t)) \\ + D(p-jm)^{-1/2} (|t| \sigma_{[n\tau]}/\sigma_{p-jm})^{-s-1} (\log(p-jm))^{1+(s'+1)/2} \\ + \int_{F_n(t)^c} |\Phi(h_2(\omega)/\sigma_{p-jm}) - \Phi(h_1(\omega)/\sigma_{p-jm})| P d\omega.$$

It is easy to bound the first three summands here (use (24)). Now we estimate the integral. According to (27) the integrand is bounded by

$$D((h_2(\omega) - h_1(\omega))/\sigma_{p-jm}) \cdot \max \{ \exp(-h_1(\omega)^2/(2\sigma_{p-jm}^2)), \exp(-h_2(\omega)^2/(2\sigma_{p-jm}^2)) \} \\ \leq D((h_2(\omega) - h_1(\omega))/\sigma_{p-jm}) (\sigma_{p-jm}/(|t| \sigma_{[n\tau]}))^{s+1} \\ \leq D(\eta + H(\omega))/(|t|^{s+1} n^{1/2}).$$

By using Lemma 2(iii) and (21) the proof of (23) is accomplished.

(23) yields the desired bound if we set $k := \lceil -\log \varepsilon_n \rceil$, $\eta := \gamma |t| \log n$ where γ is chosen so that $0 < \gamma \leq \sigma_{[n\tau]}/(6 \log n)$ for all $n > 1$. If $\varepsilon_n \geq Dn^{-1+\varepsilon}$, $n \in \mathbb{N}$, we set $\eta := \gamma \sigma_{[n\tau]}^\varepsilon |t|$ with $0 < \gamma \leq \sigma_{[n\tau]}^{1-\varepsilon}/6$, $n \in \mathbb{N}$, and choose $k \in \mathbb{N}$ so large that $s(k+1)\varepsilon > 1$.

Case 2. $|t| \geq \varepsilon_n^{-(s-1)/s}$.

Using this inequality, Theorem 4, Lemma 2 of Feller [5], p. 175 and the fact that $q \leq Dn |t| \leq Dn^{1+1/(2s)}$ we obtain for all $j \leq q$

$$\begin{aligned}
 P\{|S_j| \geq |t| \sigma_{[n\tau]}/8\} &\leq 2(1 - \Phi(|t| \sigma_{[n\tau]}/(8\sigma_j))) \\
 &\quad + D j^{-1/2} (\log j)^{1+(s'+1)/2} (|t| \sigma_{[n\tau]}/8\sigma_j)^{-s-1} \\
 &\leq D \left(\frac{\sigma_j}{|t| \sigma_{[n\tau]}}\right)^{2s} + D (\log n)^{1+(s'+1)/2} \frac{\sqrt{n}^{s(1+1/(2s))}}{(|t| \sqrt{n})^{s+1}} \\
 &\leq D \left(\frac{\sqrt{q}}{|t| \sqrt{n}}\right)^{2s} + D/|t|^{s+1} \\
 &\leq D \sqrt{\varepsilon_n}/|t|^s.
 \end{aligned} \tag{28}$$

For $t < 0$ (28) implies directly

$$R \leq D \sqrt{\varepsilon_n}/|t|^s.$$

If $t > 0$ we set

$$A_k := \{\omega : S_j(\omega) < t \sigma_{[n\tau]}, p \leq j < k, S_k \geq t \sigma_{[n\tau]}\}.$$

Then

$$R \leq P\{t \sigma_{[n\tau]}/2 \leq S_q\} + \sum_{k=p}^{q-1} P(A_k \cap \{S_q < t \sigma_{[n\tau]}/2\}).$$

(28) yields the estimation for the first term.

Set $d = d(k) = \min(m, q - k)$. Then the second is bounded by

$$\begin{aligned}
 &\sum_{k=p}^{q-1} P(A_k \cap \{S_k - S_q > t \sigma_{[n\tau]}/2\}) \\
 &\leq P\left(\bigcup_{k=p}^{q-1} \{S_{k+d} - S_k < -t \sigma_{[n\tau]}/4\}\right) + \sum_{k=p}^{q-1} P(A_k \cap \{S_{k+d} - S_q > t \sigma_{[n\tau]}/4\}).
 \end{aligned}$$

The first term can be estimated by using Lemma 2(ii), (21) and (22). The second is smaller than

$$\sum_{k=p}^{q-1} \varphi(m) P(A_k) + P(A_k) P\{S_q - S_{k+d} < -t \sigma_{[n\tau]}/4\} \leq D \sqrt{\varepsilon_n}/|t|^s \quad \text{by (28).}$$

(ii) Let $n > 1$, $t \in \mathbb{R}$ with $|t| \geq t_1 \geq 2$, $|t| \geq n^{1/2s}$.

Then

$$q(n, t) \leq n\tau + n\tau \varepsilon_n |t| + 1 \leq Dn|t| \leq D|t|^{1+2s} \tag{29}$$

$$m \leq D \log |t|. \tag{30}$$

W.l.o.g. we assume $t > 0$ and then proceed like Landers and Rogge [10], p. 102, (21)–(23). Apparently it suffices to show

$$P\{\max_{p \leq k \leq q} S_k \geq t \sigma_{[n\tau]}\} \leq Dn^{-1/2} t^{-s} (\log t)^{s+1}.$$

For $k \in \mathbb{N}$, $p \leq k \leq q$ define $d = d(k)$ and A_k like in (i).

Since in view of Lemma 2 and (29)

$$P\{|S_q - S_{k+d}| \geq t \sigma_{[n\tau]}/4\} \leq D/t^{s/2}$$

we obtain (for t_1 sufficiently large)

$$P\{S_{k+d} - S_q \leq t\sigma_{[n\tau]}/4\} \geq 1/2.$$

Therefore by (29), (30)

$$\begin{aligned} P\{\max_{p \leq k \leq q} S_k \geq t\sigma_{[n\tau]}\}/2 &\leq \sum_{k=p}^q P(A_k) P\{S_{k+d} - S_q \leq t\sigma_{[n\tau]}/4\} \\ &\leq \sum_{k=p}^q \varphi(m) P(A_k) + P(A_k \cap \{S_{k+d} - S_q \leq t\sigma_{[n\tau]}/4\}) \\ &\leq \varphi(m) + \sum_{k=p}^q P\{S_k - S_{k+d} \geq t\sigma_{[n\tau]}/4\} + P(A_k \cap \{S_k - S_q \leq t\sigma_{[n\tau]}/2\}) \\ &\leq Dn^{-1/2} t^{-s} (\log t)^{(s+1)/2} + P\{S_q \geq t\sigma_{[n\tau]}/2\}. \end{aligned}$$

(29) implies

$$t\sigma_{[n\tau]}/(2\sigma_q) \geq Dt\sqrt{n}/\sqrt{q} \geq D\sqrt{t} \geq D\sqrt{q}^{1/(1+2s)}.$$

This shows that Theorem 3 is applicable. It yields with (29) and Feller [5], p. 175

$$\begin{aligned} P\{S_q \geq t\sigma_{[n\tau]}/2\} &\leq Dq^{1-(s+1)/2} (t\sigma_{[n\tau]}/(2\sigma_q))^{-s-1} (\log q)^{s+1} \\ &\quad + D(\sigma_q/(t\sigma_{[n\tau]}))^{4s} \leq Dn^{-1/2} t^{-s} (\log t)^{s+1}. \end{aligned}$$

Remark. If in Theorem 6 $s > 3$ we can apply in its proof Theorem 4 to s instead of $s + 1$. Then we can replace $(\log n)^{1+(s+1)/2}$ by $(\log n)^{1+s/2}$ in the definition of δ_n .

4. Lemmas

Lemma 1. Let $\mathcal{B}_1, \mathcal{B}_2$ be sub- σ -algebras of \mathcal{B} and $c > 0$.

If $|P(B_1 \cap B_2) - P(B_1)P(B_2)| \leq cP(B_1)$ for all $B_i \in \mathcal{B}_i, i = 1, 2$ then for all $r_1 > 1, r_2 > 1$ with $r_1^{-1} + r_2^{-1} = 1, f_i \in \mathcal{L}_{r_i}(\Omega, \mathcal{B}_i, P), i = 1, 2,$

$$|Ef_1 f_2 - Ef_1 Ef_2| \leq 2c^{1/r_1} \|f_1\|_{r_1} \|f_2\|_{r_2}.$$

Proof. See [2], p. 170, Lemma 1.

Lemma 2. Let $X_n, n \in \mathbb{N}$, be a φ -mixing sequence. Assume that

$$\begin{aligned} EX_n &= 0, \quad n \in \mathbb{N} \\ \sum_{n=1}^{\infty} \varphi(n)^{1/2} &< \infty \\ \sup_{n \in \mathbb{N}} E|X_n|^s &\leq N \quad \text{for some } s > 2 \quad \text{and } N > 1. \end{aligned}$$

For $d > 1$ set $Y_i := Y_{d,i} := X_i 1_{\{|X_i| \leq d\}}$.

(i) Then for any real number $v \geq 2$ there exists a constant $C(v) > 0$ depending only on φ, v, s and N such that for all positive integers $n \leq d^2$

$$E \left| \sum_{i=1}^n Y_i \right|^v \leq C(v)(n^{v/2} + nd^{v-s})$$

(ii) For all $v \in [2, s], n \in \mathbb{N}$

$$E \left| \sum_{i=1}^n X_i \right|^v \leq 2C(v) n^{v/2}.$$

(iii) For any $v \in [2, s]$ there exists $D(v) > 0$ such that for all $n \in \mathbb{N}, j \in \mathbb{N} \cup \{0\}$

$$E \max_{1 \leq l \leq n} \left| \sum_{i=j+1}^{j+l} X_i \right|^v \leq D(v) n^{v/2}.$$

Notice that the constant $C(v)$ does not alter, if we turn to a subsequence of $X_n, n \in \mathbb{N}$.

Proof. (i) see [1], Lemma 1.

(ii) follows from (i).

(iii) follows from (ii) on account of [15] Corollary B1.

Lemma 3. Let X_i be a random variable with values in a measurable space $(\Omega_i, \mathcal{B}_i), i = 1, 2$, and $0 \leq c \leq 1$.

If for all $B_1 \in \mathcal{B}_1, B_2 \in \mathcal{B}_2$

$$|P * (X_1, X_2)(B_1 \times B_2) - (P * X_1) \times (P * X_2)(B_1 \times B_2)| \leq c P * X_1(B_1)$$

then

$$|P * (X_1, X_2)(D) - (P * X_1) \times (P * X_2)(D)| \leq c$$

for every $D \in \mathcal{B}_1 \times \mathcal{B}_2$.

Proof. See [3] (3.5).

The last lemma is obtained by evaluation of the constant b in Michel's Lemma 3 [11].

Lemma 4. Let X be a random variable with $EX = 0$ and $E|X|^{2+c} < \infty$ for some $c \in (0, 1]$. Then for all $z > 0, h \geq \|X\|_2$

$$E(I_{\{|X| \leq h\}} \exp(zX)) \leq 1 + z^2 EX^2/2 + E|X|^{2+c} h^{-2-c} \exp(2hz).$$

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