

## A Method for the Derivation of Limit Theorems for Sums of $m$ -dependent Random Variables

Lothar Heinrich

Bergakademie Freiberg, Sektion Mathematik, Bernhard-von-Cotta-Str. 2, DDR-9200 Freiberg

### 1. Introduction

Let  $(X_k)_{k=1,2,\dots}$  be a sequence of random variables ( $rv$ 's) which are defined on a probability space  $(\Omega, \mathfrak{A}, P)$ . For  $a \leq b$ , let  $\mathfrak{A}_a^b = \sigma(X_a, \dots, X_b)$  denote the  $\sigma$ -algebra of events generated by  $X_a, \dots, X_b$ . We put  $EX_k = 0$ ,  $k=1, 2, \dots$ , and write  $S_N = X_1 + \dots + X_N$ ,  $B_N^2 = ES_N^2$ ,  $F_N(x) = P(S_N < x B_N)$ ,  $f_N(z) = Ee^{zS_N}$  and  $\Delta_N(x) = |F_N(x) - \Phi(x)|$ , where  $\Phi(x)$  is the standard normal distribution. The sequence  $(X_k)_{k=1,2,\dots}$  is called  $m$ -dependent if for  $1 \leq s < t < \infty$ ,  $t-s > m$ , the  $\sigma$ -algebras  $\mathfrak{A}_1^s$  and  $\mathfrak{A}_t^\infty$  are independent (see [12]). For example, the sequence  $X_k = h(\xi_k, \dots, \xi_{k+m})$ ,  $k=1, 2, \dots$ , where  $(\xi_j)_{j=1,2,\dots}$  are independent  $rv$ 's and  $h|R^{m+1} \rightarrow R^1$  Borel-measurable, is  $m$ -dependent.

Sequences of  $m$ -dependent  $rv$ 's are special cases of  $\phi$ -mixing sequences (see [12, 13]). In general the estimates of the error  $\Delta_N(x)$  for  $\phi$ -mixing sequences are cruder than in the case of  $m$ -dependence (see [13]). Recently, many authors studied the error  $\Delta_N(x)$  for  $m$ -dependent  $rv$ 's. The reader is referred to [1, 2, 8, 9, 12, 13]. Perhaps the best published result on uniform bounds of  $\Delta_N(x)$  is that of V.V. Shergin [9], who shows that under the assumption  $E|X_k|^3 < \infty$ ,  $k=1, 2, \dots, N$ , the discrepancy  $\Delta_N(x)$  is less than

$$c_1(m+1)^2 \sum_{k=1}^N E|X_k|^3 B_N^{-3}.$$

Here and below,  $c_1, c_2, \dots$  denote positive constants (not depending on  $N$ ).  $\theta$  stands for a complex number with  $|\theta| \leq 1$  which may differ from one expression to another and  $\chi_A$  denotes the indicator function of the event  $A$ .  $\square$  indicates the end of a proof.

The main purpose of this paper is to provide a general method for the derivation of limit theorems for sums of  $m$ -dependent  $rv$ 's. This method is based on a factorization of the characteristic (moment-generating) function  $f_N(it)$ ,  $t \in R^1$ , ( $f_N(z)$ ,  $z \in C^1$ ) in some neighbourhood of  $t=0$  ( $z=0$ ). Section 4 contains some results concerning asymptotic expansions of  $F_N(x)$  and large

deviations of  $S_N$ . In Sect. 5, these theorems are proved by using the basic relations of Sect. 3 and the standard procedures from the summation theory of independent  $rv$ 's (see [7]). By the given factorization of  $f_N(it)$  it is possible to derive further results for sums of  $m$ -dependent  $rv$ 's and random vectors, for example: convergence to infinitely divisible distributions, non-uniform bounds of  $\Delta_N(x)$  and so on. The factorization of the characteristic function of a sum of dependent  $rv$ 's was first used by the author in order to investigate the limit behaviour of sums of  $rv$ 's connected in a Markov chain (see [3-6]).

**2. Preliminary Lemmas**

For a sequence of arbitrary complex-valued  $rv$ 's  $Y_1, Y_2, \dots$  the symbol  $\widehat{E}Y_1 Y_2 \dots Y_k$  is defined recursively by:

$$\widehat{E}Y_1 Y_2 \dots Y_k = EY_1 Y_2 \dots Y_k - \sum_{j=1}^{k-1} \widehat{E}Y_1 \dots Y_j EY_{j+1} \dots Y_k \quad \text{for } k \geq 2 \quad (2.1)$$

and  $\widehat{E}Y_1 = EY_1$ .

This symbol was at first defined (in another way) by V.A. Statulevičius [11]. The above equivalent definition was given by the author [3].

Applying (2.1) successively, we get

$$\begin{aligned} \widehat{E}Y_1 Y_2 \dots Y_k &= \sum_{l=1}^k (-1)^{l-1} \sum_{\substack{k_1 + \dots + k_l = k \\ k_i \geq 1}} EY_1 \dots Y_{k_1} EY_{k_1+1} \dots Y_{k_1+k_2} \\ &\dots EY_{k_1+\dots+k_{l-1}+1} \dots Y_k. \end{aligned} \quad (2.2)$$

Similarly to (2.2) we can show (see [3])

$$\begin{aligned} &\widehat{E}(Y_1 + a_1)(Y_2 + a_2) \dots (Y_k + a_k) \\ &= \sum_{l=2}^k \sum_{1=q_1 < q_2 < \dots < q_l = k} \widehat{E}Y_{q_1} Y_{q_2} \dots Y_{q_l} \prod_{\substack{j=2 \\ j \neq q_2, \dots, q_{l-1}}}^{k-1} a_j, \end{aligned} \quad (2.3)$$

where  $a_1, a_2, \dots, a_k$  are arbitrary complex numbers.

Another result is concerned with the differentiation of  $\widehat{E}Y_1(t) \dots Y_k(t)$ , where  $(Y_j(t))_{j=1, \dots, k}$  are differentiable  $rv$ 's (with respect to the parameter  $t$ ). If differentiation and expectation are exchangeable, then the following analogue to Leibniz's rule holds (see [3]):<sup>2</sup>

$$\begin{aligned} \frac{d^p}{dt^p} \widehat{E}Y_1(t) \dots Y_k(t) &= \sum_{\substack{p_1 + \dots + p_k = p \\ p_k \geq 0}} \binom{p}{p_1 \dots p_k} \widehat{E}Y_1^{(p_1)}(t) \dots Y_k^{(p_k)}(t) \\ &= \sum_{l=1}^{\min(p, k)} \sum_{\substack{p_1 + \dots + p_l = p \\ p_i \geq 1}} \binom{p}{p_1 \dots p_l} \sum_{1 \leq i_1 < \dots < i_l \leq k} \widehat{E}Y_1(t) \\ &\dots Y_{i_1}^{(p_1)}(t) \dots Y_{i_l}^{(p_l)}(t) \dots Y_k(t). \end{aligned} \quad (2.4)$$

<sup>1</sup> We suppose  $E|Y_j|^k < \infty, j=1, 2, \dots, k$ .

<sup>2</sup>  $\binom{p}{p_1 p_2 \dots p_k} = \frac{p!}{p_1! p_2! \dots p_k!}$

**Lemma 2.1.** Let  $U_1, \dots, U_p$  and  $V_1, \dots, V_q$  be arbitrary rv's with  $E|U_i|^p < \infty$ ,  $i=1, \dots, p$  and  $E|V_j|^q < \infty$ ,  $j=1, \dots, q$ . If the  $\sigma$ -algebras  $\sigma(U_1, \dots, U_p)$  and  $\sigma(V_1, \dots, V_q)$  are independent, then

$$\widehat{E}U_1 \dots U_p V_1 \dots V_q = 0.$$

*Proof.* Because of the Definition (2.1) and the assumptions of Lemma 2.1 we have

$$\begin{aligned} \widehat{E}U_1 \dots U_p V_1 \dots V_q &= EV_1 \dots V_q \left( EU_1 \dots U_p - \sum_{i=1}^{p-1} \widehat{E}U_1 \dots U_i EU_{i+1} \dots U_p \right) \\ &\quad - \widehat{E}U_1 \dots U_p EV_1 \dots V_q - \sum_{j=1}^{q-1} \widehat{E}U_1 \dots U_p V_1 \dots V_j EV_{j+1} \dots V_q \\ &= - \sum_{j=1}^{q-1} \widehat{E}U_1 \dots U_p V_1 \dots V_j EV_{j+1} \dots V_q. \end{aligned}$$

In the same way we get

$$\widehat{E}U_1 \dots U_p V_1 = EV_1 \left( EU_1 \dots U_p - \sum_{i=1}^{p-1} \widehat{E}U_1 \dots U_i EU_{i+1} \dots U_p - \widehat{E}U_1 \dots U_p \right) = 0.$$

Repeating the above procedure we obtain the assertion of Lemma 2.1.  $\square$

**Corollary 2.1.** If  $Y_1, \dots, Y_k$  are 1-dependent rv's with  $E|Y_j|^k < \infty$ ,  $j=1, \dots, k$ , then we have

$$\widehat{E}(Y_1 + a_1)(Y_2 + a_2) \dots (Y_k + a_k) = \widehat{E}Y_1 Y_2 \dots Y_k, \tag{2.5}$$

where  $a_1, \dots, a_k$  are arbitrary complex numbers.

*Proof.* By Lemma 2.1 and the 1-dependence we have  $\widehat{E}Y_{q_1} Y_{q_2} \dots Y_{q_l} = 0$  for  $1 = q_1 < q_2 < \dots < q_l = k$ ,  $2 \leq l < k$ . Using (2.3) we immediately get (2.5).  $\square$

**Lemma 2.2.** Let  $Y_1, \dots, Y_k$  be 1-dependent rv's with  $E|Y_j|^2 < \infty$ ,  $j=1, \dots, k$ . Then the following estimate holds:

$$|\widehat{E}Y_1 Y_2 \dots Y_k| \leq 2^{k-1} \prod_{j=1}^k (E|Y_j|^2)^{1/2}. \tag{2.6}$$

*Proof.* We prove relation (2.6) by induction. Clearly (2.6) holds for  $k=1$ . By Schwarz's inequality and the 1-dependence we have

$$\begin{aligned} |EY_{j+1} \dots Y_p| &\leq (E|Y_{j+1} Y_{j+3} \dots|^2)^{1/2} (E|Y_{j+2} Y_{j+4} \dots|^2)^{1/2} \\ &= \prod_{i=j+1}^p (E|Y_i|^2)^{1/2}. \end{aligned} \tag{2.7}$$

Let us assume that (2.6) is valid for  $k=1, \dots, p-1$ . Using the Definition (2.1) for  $k=p$  and (2.7), it is easy to see that (2.6) also holds for  $k=p$ .  $\square$

Let there be given a sequence of  $m$ -dependent  $rv$ 's  $X_1, X_2, \dots, X_N$ . We can construct the following sequence of 1-dependent  $rv$ 's:

$$Y_k = \sum_{j=1}^m X_{m(k-1)+j}, \quad k=1, \dots, \left[ \frac{N}{m} \right] \quad \text{and} \quad Y_{\left[ \frac{N}{m} \right]+1} = S_n - \sum_{k=1}^{\left[ \frac{N}{m} \right]} Y_k. \quad (2.8)$$

Therefore, it suffices to consider the case  $m=1$ . We use the notations of Section 1 and put  $M_{pN} = \max_{1 \leq k \leq N} E|X_k|^p$ ,

$$w_N(z) = \max_{1 \leq k \leq N} (E|e^{zX_k} - 1|^2)^{1/2}, \quad K_N = \{z \in C^1: w_N(z) \leq 1/6\}.$$

**Lemma 2.3.** *If  $X_1, \dots, X_N$  are 1-dependent  $rv$ 's with  $M_{pN} < \infty$ , then the following estimate holds:*

$$\left| \frac{d^p}{dt^p} \widehat{E}(e^{itX_j} - 1) \dots (e^{itX_k} - 1) \right| \leq \begin{cases} M_{pN}(k-j+1)^p 2^{3p+1} (2w_N(it))^{k-j-2p}, & k-j \geq 2p, \\ M_{pN}(k-j+1)^p 2^{2(k-j)}, & k-j < 2p \end{cases} \quad (2.9)$$

where  $1 \leq j < k \leq N$ .

*Proof.* To prove relation (2.9) we replace the  $rv$ 's  $(Y_{i=1, \dots, k})$  in (2.4) by  $e^{itX_i} - 1$ ,  $l=j, \dots, k$ . Using (2.2), the well-known inequality  $|e^{ix} - 1| \leq 2$ , Hölder's inequality and keeping in mind the 1-dependence, it follows easily that

$$|\widehat{E} \dots Y_{i_1}^{p_1} e^{itX_{i_1}} \dots Y_{i_l}^{p_l} e^{itX_{i_l}} \dots| \leq \begin{cases} M_{pN} 2^{3l+1} (2w_N(it))^{k-j-2l}, & k-j \geq 2l \\ M_{pN} 2^{2(k-j)-l+1}, & k-j < 2l \end{cases}$$

for  $j \leq i_1 < \dots < i_l \leq k$ .

Therefore the identity

$$\sum_{l=1}^{\min(p, k-j+1)} \sum_{\substack{p_1 + \dots + p_l = p \\ p_i \geq 1}} \binom{p}{p_1 \dots p_l} \binom{k-j+1}{l} = (k-j+1)^p$$

implies the desired estimate (2.9).  $\square$

### 3. Factorization of the Characteristic Function of a Sum of 1-dependent Random Variables

We are now in position to formulate the main lemmas of this paper.

**Lemma 3.1.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent  $rv$ 's. Then the product representation*

$$E e^{zS_N} = \varphi_1(z) \varphi_2(z) \dots \varphi_N(z) \quad (3.1)$$

holds for each  $z \in K_N$ , where  $\varphi_1(z) = E e^{zX_1}$  and for  $k=2, \dots, N$

$$\varphi_k(z) = \frac{E e^{zS_k}}{E e^{zS_{k-1}}} = E e^{zX_k} + \sum_{j=1}^{k-1} \frac{\widehat{E}(e^{zX_j} - 1)(e^{zX_{j+1}} - 1) \dots (e^{zX_k} - 1)}{\varphi_j(z) \varphi_{j+1}(z) \dots \varphi_{k-1}(z)}. \quad (3.2)$$

<sup>3</sup>  $[x]$  denotes the integer part of  $x$ .

Furthermore, the following estimates are true for each  $z \in K_N$  and  $k = 1, 2, \dots, N$ :

$$|\varphi_k(z) - 1| \underset{(\geq)}{\leq} |E e^{z X_k} - 1| + \frac{2(E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2}}{1 - 4w_N(z)} \tag{3.3}$$

or

$$|\varphi_k(z) - 1| \underset{(\geq)}{\leq} |E e^{z X_k} - 1| + 6(w_N(z))^2.$$

*Proof.* Applying the elementary symmetry property  $\widehat{E} Y_1 Y_2 \dots Y_k = \widehat{E} Y_k \dots Y_2 Y_1$  for arbitrary  $r v$ 's, we get from (2.1)

$$E e^{z S_k} = \widehat{E} e^{z X_1} \dots e^{z X_k} + \sum_{j=1}^{k-1} E e^{z S_j} \widehat{E} e^{z X_{j+1}} \dots e^{z X_k}.$$

Taking into account that  $\varphi_1(z) = E e^{z X_1}$  and  $\varphi_j(z) = E e^{z S_j} / E e^{z S_{j-1}}$ ,  $j = 2, \dots, k-1$ , we have

$$\frac{E e^{z S_k}}{E e^{z S_{k-1}}} = E e^{z X_k} + \sum_{j=1}^{k-1} \frac{\widehat{E} e^{z X_j} \dots e^{z X_k}}{\varphi_j(z) \dots \varphi_{k-1}(z)}.$$

Finally, using Corollary 2.1, we obtain the representation (3.2). Now, our problem is to estimate  $|\varphi_k(z) - 1|$  in certain neighbourhood of  $z = 0$ . We shall prove (3.3) by induction.

Clearly  $|\varphi_1(z) - 1| = |E e^{z X_1} - 1|$ . By Lemma 2.2 we conclude that

$$\begin{aligned} & |\widehat{E}(e^{z X_j} - 1) \dots (e^{z X_k} - 1)| \\ & \leq 2^{k-j} (w_N(z))^{k-j-1} (E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2}. \end{aligned} \tag{3.4}$$

Let us assume that (3.3) holds for  $|\varphi_j(z) - 1|$ ,  $j = 1, \dots, k-1$ . Then  $z \in K_N$  implies that

$$\max_{1 \leq j \leq k-1} |\varphi_j(z) - 1| \leq w_N(z) + 6(w_N(z))^2 \leq 2w_N(z). \tag{3.5}$$

Note that

$$\frac{2w_N(z)}{1 - \max_{1 \leq j \leq k-1} |\varphi_j(z) - 1|} \leq \frac{2w_N(z)}{1 - 2w_N(z)} \leq \frac{1}{2}$$

and hence combining (3.4) and (3.5), we get

$$\begin{aligned} |\varphi_k(z) - 1| & \leq |E e^{z X_k} - 1| + \frac{2(E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2}}{|\varphi_{k-1}(z)|} \\ & \times \sum_{j=1}^{k-1} \left( \frac{2w_N(z)}{1 - \max_{1 \leq i \leq k-1} |\varphi_i(z) - 1|} \right)^{k-j-1} \leq |E e^{z X_k} - 1| \\ & + \frac{2(E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2} (1 - 2w_N(z))}{(1 - |\varphi_{k-1}(z) - 1|)(1 - 4w_N(z))} \end{aligned}$$

which proves (3.6). In the same way one can show the validity of “ $\geq$ ” in (3.6).  $\square$

**Lemma 3.2.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's. Then the estimate*

$$\begin{aligned} & \left| \log E e^{z S_N} - \sum_{k=1}^N E(e^{z X_k} - 1) - \sum_{k=2}^N \widehat{E}(e^{z X_{k-1}} - 1)(e^{z X_k} - 1) \right| \\ & \leq 2w_N(z) \left( \sum_{k=1}^N |E e^{z X_k} - 1| + 22 \sum_{k=1}^N E |e^{z X_k} - 1|^2 \right) \end{aligned} \tag{3.6}$$

holds for each  $z \in K_N$ .

*Proof.* Using (3.1) and (3.2) we decompose the cumulant-generating function of  $S_N$  as follows:

$$\log E e^{z S_N} = \sum_{k=1}^N E(e^{z X_k} - 1) + \sum_{k=2}^N \widehat{E}(e^{z X_{k-1}} - 1)(e^{z X_k} - 1) + R_N(z),$$

where

$$\begin{aligned} R_N(z) &= \sum_{k=1}^N (\log \varphi_k(z) - (\varphi_k(z) - 1)) + \sum_{k=2}^N \sum_{j=1}^{k-1} \left( \frac{1}{\varphi_j(z) \dots \varphi_{k-1}(z)} - 1 \right) \\ & \quad \times \widehat{E}(e^{z X_j} - 1) \dots (e^{z X_k} - 1) + \sum_{k=3}^N \sum_{j=1}^{k-2} \widehat{E}(e^{z X_j} - 1) \dots (e^{z X_k} - 1). \end{aligned}$$

As an immediate consequence of (3.4), (3.5) and  $z \in K_N$  we obtain the following estimates:

$$\begin{aligned} \left| \frac{1}{\varphi_j(z) \dots \varphi_{k-1}(z)} - 1 \right| & \leq (k-j) \frac{2w_N(z)}{1-2w_N(z)} \left( 1 + \frac{2w_N(z)}{1-2w_N(z)} \right)^{k-j-1} \\ & \leq 3(k-j)w_N(z)(1+3w_N(z))^{k-j-1}, \\ \left| \sum_{j=1}^{k-2} \widehat{E}(e^{z X_j} - 1) \dots (e^{z X_k} - 1) \right| & \leq 6w_N(z)(E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2}, \\ \left| \sum_{j=1}^{k-1} \left( \frac{1}{\varphi_j(z) \dots \varphi_{k-1}(z)} - 1 \right) \widehat{E}(e^{z X_j} - 1) \dots (e^{z X_k} - 1) \right| \\ & \leq 6w_N(z) \sum_{j=1}^{k-1} (k-j)(3w_N(z))^{k-j-1} (E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2} \\ & \leq 24w_N(z)(E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2}. \end{aligned}$$

By these estimates we get

$$\begin{aligned} \sum_{k=1}^N |\varphi_k(z) - 1| & \leq \sum_{k=1}^N |E e^{z X_k} - 1| \\ & \quad + (2 + 30w_N(z)) \sum_{k=2}^N (E|e^{z X_{k-1}} - 1|^2 E|e^{z X_k} - 1|^2)^{1/2}. \end{aligned}$$

Using the above estimates again and the elementary inequality

$$|\log z - (z - 1)| \leq |z - 1|^2 \quad \text{for } |z - 1| \leq 1/2,$$

we have

$$|R_N(z)| \leq 2w_N(z) \sum_{k=1}^N |E e^{zX_k} - 1| + 44w_N(z) \sum_{k=2}^N (E|e^{zX_{k-1}} - 1|^2 E|e^{zX_k} - 1|^2)^{1/2}.$$

This yields (3.6).  $\square$

**Corollary 3.1.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's. If  $t \in \mathbb{R}^1$  satisfies  $w_N(it) \leq 1/6$ , then*

$$\left| \log E e^{itS_N} - \sum_{k=1}^N E(e^{itX_k} - 1) - \sum_{k=2}^N \widehat{E}(e^{itX_{k-1}} - 1)(e^{itX_k} - 1) \right| \leq 90w_N(it) \sum_{k=1}^N |E e^{itX_k} - 1|.$$

*Proof.* In fact, Lemma 3.2 and the well-known inequality

$$E|e^{itX_k} - 1|^2 \leq 2|E e^{itX_k} - 1|$$

yield the desired result.  $\square$

**Corollary 3.2.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's with  $EX_k = 0$  and  $E|X_k|^3 < \infty$ . If  $|t| \leq B_N^3/192NM_{3N}$ , then*

$$|f_N(it/B_N)| \leq e^{-t^2/4}. \tag{3.7}$$

*Proof.* Noting that  $\left| e^{itx} - \sum_{k=0}^{p-1} \frac{(itx)^k}{k!} \right| \leq \frac{|tx|^p}{p!}$ ,  $p = 1, 2, \dots$  and  $B_N^2 \leq 3 \sum_{k=1}^N EX_k^2$ , we have the following estimates:  $w_N(it/B_N) \leq 1/64$ ,

$$w_N(it/B_N) \sum_{k=1}^N |E e^{itX_k/B_N} - 1| \leq \frac{1}{2}|t|^3 \frac{NM_{3N}}{B_N^3},$$

$$\sum_{k=2}^N \left| \widehat{E}(e^{itX_{k-1}/B_N} - 1)(e^{itX_k/B_N} - 1) - \frac{(it)^2}{B_N^2} \widehat{E}X_{k-1}X_k \right| \leq 2 \left( \frac{|t|}{B_N} \right)^3 \sum_{k=1}^N E|X_k|^3,$$

$$\sum_{k=1}^N \left| E e^{itX_k/B_N} - 1 - it \frac{EX_k}{B_N} - (it)^2 \frac{EX_k^2}{2B_N^2} \right| \leq \frac{1}{6} \left( \frac{|t|}{B_N} \right)^3 \sum_{k=1}^N E|X_k|^3.$$

We conclude from Corollary 3.1 that

$$\left| \log E e^{itS_N/B_N} + \frac{t^2}{2} \right| \leq \frac{283}{6} \left( \frac{|t|}{B_N} \right)^3 NM_{3N} \leq \frac{t^2}{4},$$

which implies (3.7).  $\square$

A more detailed study of the terms  $\log \varphi_k(it)$  leads to the following extension of Lemma 3.2.

**Lemma 3.3.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's. If  $w_N(it) \leq 1/6$ , then*

$$\left| \log E e^{itS_N} - \sum_{k=1}^N ((E e^{itX_k} - 1) - \frac{1}{2}(E e^{itX_k} - 1)^2 + \frac{1}{3}(E e^{itX_k} - 1)^3) \right. \\ \left. + \sum_{k=2}^N ((E e^{itX_{k-1}} - 1) + (E e^{itX_k} - 1) - 1) \widehat{E}(e^{itX_{k-1}} - 1)(e^{itX_k} - 1) \right. \\ \left. - \sum_{k=3}^N \widehat{E}(e^{itX_{k-2}} - 1)(e^{itX_{k-1}} - 1)(e^{itX_k} - 1) \right| \leq c_2(w_N(it))^2 \sum_{k=1}^N |E e^{itX_k} - 1|. \quad (3.8)$$

One can verify the assertion of Lemma 3.3 by simple refinements in the proof of Lemma 3.2. To avoid tedious, but not difficult calculations we omit this proof.

**Lemma 3.4.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's with  $M_{pN} < \infty$ . If  $w_N(it) \leq 1/6$  then*

$$\max_{1 \leq k \leq N} \left| \frac{d^p \varphi_k(it)}{dt^p} \right| \leq C(p) M_{pN}, \quad (3.9)$$

where  $C(p)$  is a constant only depending on  $p$ .

*Proof.* For brevity we put

$$g_{jk}^{(p)} = \frac{d^p}{dt^p} (\varphi_j(it) \dots \varphi_{k-1}(it))$$

and

$$f_{jk}^{(p)} = \frac{d^p}{dt^p} \widehat{E}(e^{itX_j} - 1) \dots (e^{itX_k} - 1)$$

for  $p=0, 1, 2, \dots$ . It is easily seen from (2.9) and (3.5) that

$$\left| \frac{f_{jk}^{(p)}}{g_{jk}^{(p)}} \right| \leq 12^{2p} M_{pN} (k-j+1)^p 2^{-(k-j)}, \quad 1 \leq j \leq k. \quad (3.10)$$

Next we prove (3.9) for  $p=1$ .

Assume that  $|\varphi_q^{(1)}(it)| \leq C(1)M_{1N}$  with  $C(1) = 866$  holds for  $q=1, \dots, k-1$ . In view of the representation formula (3.2) the first derivative of  $\varphi_k(it)$  has the form

$$\frac{d}{dt} \varphi_k(it) = iE X_k e^{itX_k} + \sum_{j=1}^{k-1} \frac{f_{jk}^{(1)}}{g_{jk}^{(1)}} - \sum_{j=1}^{k-1} \frac{f_{jk}^{(1)}}{g_{jk}^{(1)}} \sum_{q=j}^{k-1} \frac{\varphi_q^{(1)}(it)}{\varphi_q(it)}.$$

Then by (3.4), (3.10) and our assumption

$$\left| \frac{d}{dt} \varphi_k(it) \right| \leq M_{1N} + \sum_{j=1}^{k-1} M_{1N} \frac{(k-j+1)12^2}{2^{k-j}} + 3w_N(it) \sum_{j=1}^{k-1} \frac{(k-j)C(1)}{2^{k-j+1}} M_{1N} \\ \leq M_{1N} \left( 1 + 144 \sum_{k=1}^{\infty} (k+1)2^{-k} + 3C(1)w_N(it) \sum_{k=0}^{\infty} (k+1)2^{-k-2} \right) \\ \leq M_{1N}(433 + C(1)/2) \leq C(1)M_{1N}.$$

Thus the last relation implies  $|\varphi_k^{(1)}(it)| \leq 866 M_{1N}$  for every  $k \geq 1$ .



To prove (3.9) for  $p \geq 2$  we state the following differentiation formulae:

$$\frac{d^p \varphi_k(it)}{dt^p} = i^p E X_k^p e^{itX_k} + \sum_{j=1}^{k-1} \frac{d^p}{dt^p} \left( \frac{f_{jk}(it)}{g_{jk}(it)} \right)$$

and

$$\begin{aligned} \frac{d^p}{dt^p} \left( \frac{f_{jk}(it)}{g_{jk}(it)} \right) &= \frac{f_{jk}^{(p)}}{g_{jk}} - \sum_{q=1}^p \binom{p}{q} \frac{f_{jk}^{(q)} g_{jk}^{(p-q)}}{g_{jk} g_{jk}} - \frac{f_{jk}}{g_{jk}} \sum_{q=j}^{k-1} \frac{\varphi_q^{(p)}(it)}{\varphi_q(it)} \\ &\quad - \frac{f_{jk}}{g_{jk}} \sum_{q=2}^{\min(p, k-j)} \sum_{\substack{p_1 + \dots + p_q = p \\ p_i \geq 1}} \binom{p}{p_1, \dots, p_q} \sum_{j \leq j_1 < \dots < j_q \leq k-1} \frac{\varphi_{j_1}^{(p_1)}(it) \dots \varphi_{j_q}^{(p_q)}(it)}{\varphi_{j_1}(it) \dots \varphi_{j_q}(it)} \\ &\quad - \sum_{l=3}^{p+1} (-1)^{l-1} \sum_{\substack{p_1 + (p_2+1) + \dots + (p_l+1) = p \\ p_i \geq 0}} \binom{p}{p_1, p_2+1, \dots, p_l+1} \\ &\quad \times \frac{f_{jk}^{(p_1)} g_{jk}^{(p_2+1)} \dots g_{jk}^{(p_l+1)}}{g_{jk} g_{jk} \dots g_{jk}}. \end{aligned}$$

Now one can verify (3.9) in the following way: It is assumed that (3.9) holds for  $p=1, \dots, s-1$  and  $|\varphi_q^{(s)}(it)| \leq C(s)M_{sN}$  for  $q=1, \dots, k-1$ . Then by the above formulae one can show that  $|\varphi_k^{(s)}(it)| \leq C(s)M_{sN}$  and therefore (3.9) holds for  $p=s$ .

This yields (3.9) for every  $p \geq 1$ . We remark that in accordance with the above differentiation formulae the constants  $C(p)$ ,  $p=1, 2, \dots$  are successively calculated.  $\square$

**Corollary 3.3.** Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's with  $M_{pN} < \infty$ . If  $w_N(it) \leq 1/6$ , then

$$\max_{1 \leq k \leq N} \left| \frac{d^p}{dt^p} \log \varphi_k(it) \right| \leq K(p)M_{pN}, \tag{3.11}$$

where

$$K(p) = p! \sum_{q=1}^p \left( \frac{3}{2} \right)^q \frac{1}{q} \sum_{\substack{p_1 + \dots + p_q = p \\ p_i \geq 1}} \frac{C(p_1)}{p_1!} \dots \frac{C(p_q)}{p_q!}.$$

*Proof.* The derivatives of  $\log \varphi_k(it)$  are given by

$$\frac{d^p}{dt^p} \log \varphi_k(it) = p! \sum_{q=1}^p \frac{(-1)^{q-1}}{(\varphi_k(it))^q} \sum_{\substack{p_1 + \dots + p_q = p \\ p_i \geq 1}} \frac{\varphi_k^{(p_1)}(it)}{p_1!} \dots \frac{\varphi_k^{(p_q)}(it)}{p_q!}.$$

A simple application of Hölder's inequality shows the validity of (3.11).  $\square$

*Remark 3.1.* If  $X_1, \dots, X_N$  are  $m$ -dependent rv's with  $M_{pN} < \infty$ , then using (2.8), (3.11) and the well-known inequality

$$|a_1 + \dots + a_m|^p \leq m^{p-1} (|a_1|^p + \dots + |a_m|^p)$$

we get for  $|t| \leq B_N/6m \max_{1 \leq k \leq N} (E X_k^2)^{1/2}$

$$\left| \frac{d^p}{dt^p} \log E e^{itS_N/B_N} \right| \leq K(p) \frac{m^{p-1} N M_{pN}}{B_N^p}. \tag{3.12}$$

**4. Results**

**Theorem 4.1.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's with  $EX_k=0$ . Suppose that*

$$E|X_k|^p \leq H \frac{EX_k^2 p!}{\delta_N^{p-2}}, \quad p=2, 3, \dots; \quad k=1, \dots, N, \tag{4.1}$$

with  $H \geq 1/2$  and  $0 < \delta_N \leq (18HM_{2N})^{-1/2}$ .

Then in the interval  $0 \leq x \leq c_3 \Delta_N/H_N$  the following relation for large deviations holds:

$$\frac{P(S_N \geq x B_N)}{1 - \Phi(x)} = \exp \left\{ x^3 \sum_{k=0}^{\infty} \lambda_{kN} x^k \right\} \left( 1 + O \left( \frac{H_N(1+x)}{\Delta_N} \right) \right)^4 \quad \text{as } N \rightarrow \infty,$$

where

$$\Delta_N = \delta_N B_N / 4, \quad H_N = 80H \sum_{k=1}^N EX_k^2 / B_N^2$$

and

$$\lambda_{kN} = \frac{1}{(k+2)(k+3)} \sum_{l=1}^{k+1} (-1)^{l-1} \binom{k+l+1}{l} \sum_{\substack{k_1 + \dots + k_l = k+1 \\ k_i \geq 1}} \prod_{i=1}^l \frac{\Gamma_{k_i+2}(S_N/B_N)}{(k_i+1)!}$$

and

$$\Gamma_k(S_N) = \frac{1}{i^k} \frac{d^k}{dt^k} \log f_N(it)|_{t=0}.$$

**Corollary 4.1.** *Let  $X_1, \dots, X_N$  be a sequence of  $m$ -dependent ( $m \geq 1$ ) rv's with  $EX_k=0$ . Suppose that (4.1) is satisfied. Then the assertion of Theorem 4.1 remains valid if we replace  $H_N$  and  $\Delta_N$  by*

$$\bar{H}_N = 80Hm \sum_{k=1}^N EX_k^2 / B_N^2 \quad \text{and} \quad \bar{\Delta}_N = \delta_N B_N / 4m,$$

respectively.

**Theorem 4.2.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent rv's with  $EX_k=0$  and  $E|X_k|^3 < \infty$ . Put  $L_{3N} = NM_{3N}/B_N^3$ . Suppose that the following conditions are satisfied:*

- (A)  $0 < c_4 \leq M_{3N}$ ,  $B_N \rightarrow \infty$  and  $L_{3N} \rightarrow 0$  as  $N \rightarrow \infty$ ,
- (B) there exists a real number  $\alpha$ ,  $0 < \alpha < 1$ , such that

$$\max_{1 \leq k \leq N} E|X_k|^3 \chi_{\{|X_k| \geq L_{3N}^{-\alpha}\}} = o(M_{3N}) \quad \text{as } N \rightarrow \infty,$$

$$(C) \quad \int_{|t| \geq \frac{1}{192L_{3N}}} \frac{|E e^{itS_N/B_N}|}{|t|} dt = o(L_{3N}) \quad \text{as } N \rightarrow \infty.$$

<sup>4</sup>  $A_N(x) = \sum_{k=0}^{\infty} \lambda_{kN} x^k$  is the Cramér's power series.

Then the asymptotic expansion

$$F_N(x) - \Phi(x) - \frac{(1-x^2)}{6\sqrt{2\pi}} \frac{\Gamma_3(S_N)}{B_N^3} e^{-x^2/2} = o(L_{3N}) \quad \text{as } N \rightarrow \infty$$

holds uniformly in  $x$ .

**Corollary 4.2.** Let  $X_1, \dots, X_N$  be a sequence of  $m$ -dependent rv's with  $EX_k = 0$  and  $E|X_k|^3 < \infty$ . Then the whole formulation of Theorem 4.2 remains valid if we replace  $L_{3N}$  by  $\bar{L}_{3N} = m^2 NM_{3N}/B_N^3$ .

**5. Proofs of the Results**

*Proof of Theorem 4.1.* (2.6) implies

$$|\widehat{E}X_j^{p_j} \dots X_k^{p_k}| \leq 2^{k-j} (E|X_j|^{2p_j} \dots E|X_k|^{2p_k})^{1/2},$$

and therefore by a short calculation one can show the analyticity of

$$\widehat{E}(e^{zX_j} - 1) \dots (e^{zX_k} - 1) = \sum_{p=k-j+1}^{\infty} z^p \sum_{\substack{p_j + \dots + p_k = p \\ p_i \geq 1}} \frac{\widehat{E}X_j^{p_j} \dots X_k^{p_k}}{p_j! \dots p_k!},$$

$1 \leq j \leq k, 1 \leq k \leq N$ , for  $|z| < \delta_N/2$ .

The analyticity of  $\varphi_1(z), \dots, \varphi_N(z)$  for  $|z| \leq \delta_N/4$  can be verified inductively by the defining formula (3.2). Thus  $\log Ee^{zS_N}$  is an analytic function for  $|z| \leq \delta_N/4$ . Again using (2.6) we have for  $|z| \leq \delta_N/4$

$$|Ee^{zX_k} - 1| \leq \sum_{p=2}^{\infty} \frac{|z|^p}{p!} E|X_k|^p \leq \frac{4}{3} H E X_k^2 |z|^2$$

and

$$E|e^{zX_k} - 1|^2 \leq \sum_{p=2}^{\infty} \frac{|z|^p}{p!} 2^p E|X_k|^p \leq 8 H E X_k^2 |z|^2.$$

$\delta_N$  was chosen in such way that each  $z$  with  $|z| \leq \delta_N/4$  belongs to  $K_N$ . Next we apply Lemma 3.2 and use the above estimates:

$$\begin{aligned} |\log Ee^{zS_N}| &\leq \sum_{k=1}^N |Ee^{zX_k} - 1| + 2 \sum_{k=1}^N E|e^{zX_k} - 1|^2 \\ &\quad + 2w_N(z) \left( \sum_{k=1}^N |Ee^{zX_k} - 1| + 22 \sum_{k=1}^N E|e^{zX_k} - 1|^2 \right) \\ &\leq 80H \sum_{k=1}^N E X_k^2 |z|^2 \quad \text{for } |z| \leq \delta_N/4. \end{aligned}$$

Together with the introduced notations we can state the following result. The cumulant-generating function  $\log Ee^{zS_N/B_N}$  is an analytical function for  $|z| \leq \delta_N$  such that

$$|\log Ee^{zS_N/B_N}|_{|z|=\delta_N} \leq H_N \Delta_N^2.$$

Now we are in position to apply a lemma on large deviations which is due to V.A. Statulevičius [10, 11]. This lemma yields the assertion of Theorem 4.1.  $\square$

*Proof of Corollary 4.1.* In order to proof Corollary 4.1 we only repeat the proof of Theorem 4.1 for the 1-dependent  $rv$ 's (2.8). Taking into account the inequality

$$E|Y_k|^p \leq m^{p-1} \sum_{j=1}^m E|X_{(k-1)m+j}|^p, \quad k=1, \dots, \left\lfloor \frac{N}{m} \right\rfloor \quad (5.1)$$

and the corresponding inequality for  $k = \left\lfloor \frac{N}{m} \right\rfloor + 1$ , it is easy to show the analyticity of  $\log E e^{zS_N}$  for  $|z| \leq \delta_N/4m$  and the estimate

$$|\log E e^{zS_N}|_{|z|=\delta_N} \leq \bar{H}_N \bar{\Delta}_N^2. \quad \square$$

*Proof of Theorem 4.2.* We first prove the following

**Lemma 5.1.** *Let  $X_1, \dots, X_N$  be a sequence of 1-dependent  $rv$ 's with  $EX_k = 0$  and  $E|X_k|^3 < \infty$ . Suppose that the conditions (A) and (B) are satisfied. Then, for  $|t| \leq L_{3N}^{-\beta}$ ,  $0 < \beta < 1 - \alpha$ ,*

$$\left| \log f_N(it/B_N) - \left( -\frac{t^2}{2} + \frac{(it)^3}{6} \Gamma_3(S_N) B_N^{-3} \right) \right| \leq \varepsilon_N L_{3N} |t|^3.$$

Here and below,  $\varepsilon_N$  denotes a positive null sequence which may be differ from one expression to another.

*Proof of Lemma 5.1.* Since  $|L_{3N}^{-\alpha} t| \leq \varepsilon_N B_N$  by our assumptions, we have for  $p=1, 2, 3$

$$\begin{aligned} E(e^{itX_k/B_N} - 1) &= \sum_{q=1}^p \left( \frac{it}{B_N} \right)^q \frac{EX_k^q}{q!} + \theta \frac{|t|^p}{p! B_N^p} (E|X_k|^p \chi_{\{|X_k| \geq L_{3N}^{-\alpha}\}}) \\ &+ \frac{|t|}{(p+1)B_N} E|X_k|^{p+1} \chi_{\{|X_k| < L_{3N}^{-\alpha}\}} = \sum_{q=1}^p \left( \frac{it}{B_N} \right)^q \frac{EX_k^q}{q!} + \theta \varepsilon_N \frac{|t|^p (M_{3N})^{p/3}}{p! B_N^p}. \end{aligned}$$

The following estimates may be derived in like manner:

$$\begin{aligned} \widehat{E}(e^{itX_{k-1}/B_N} - 1)(e^{itX_k/B_N} - 1) &= \left( \frac{it}{B_N} \right)^2 \widehat{E}X_{k-1}X_k + \frac{1}{2} \left( \frac{it}{B_N} \right)^3 \\ &\times (\widehat{E}X_{k-1}^2X_k + \widehat{E}X_{k-1}X_k^2) + \theta \varepsilon_N |t|^3 M_{3N} B_N^{-3}, \\ \widehat{E}(e^{itX_{k-1}/B_N} - 1)(e^{itX_k/B_N} - 1)(E(e^{itX_{k-1}/B_N} - 1) + E(e^{itX_k/B_N} - 1)) \\ &= \left( \frac{it}{B_N} \right)^3 (EX_{k-1} + EX_k) \widehat{E}X_{k-1}X_k + \theta \varepsilon_N |t|^3 M_{3N} B_N^{-3}, \\ \widehat{E}(e^{itX_{k-2}/B_N} - 1)(e^{itX_{k-1}/B_N} - 1)(e^{itX_k/B_N} - 1) \\ &= \left( \frac{it}{B_N} \right)^3 \widehat{E}X_{k-2}X_{k-1}X_k + \theta \varepsilon_N |t|^3 M_{3N} B_N^{-3}, \end{aligned}$$

$$w_N^2 \left(\frac{it}{B_N}\right) \sum_{k=1}^N |E e^{itX_k/B_N} - 1| \leq \varepsilon_N \left(\frac{|t|}{B_N}\right)^3 M_{3N}.$$

A short calculation shows that

$$\begin{aligned} \Gamma_3(S_N) = & \sum_{k=1}^N (EX_k^3 - 3EX_k EX_k^2 + 2(EX_k)^3) + 3 \sum_{k=2}^N (\widehat{E}X_{k-1}^2 X_k + \widehat{E}X_{k-1} X_k^2) \\ & - 6 \sum_{k=2}^N (EX_{k-1} + EX_k) \widehat{E}X_{k-1} X_k + 6 \sum_{k=3}^N \widehat{E}X_{k-2} X_{k-1} X_k \end{aligned}$$

and by Hölder’s inequality

$$|\Gamma_3(S_N/B_N)| \leq 13 \sum_{k=1}^N \frac{E|X_k|^3}{B_N^3} \leq 13 L_{3N}. \tag{5.2}$$

Using Lemma 3.3 and the above estimates, it is rapidly seen that

$$\begin{aligned} \log f_N(it/B_N) = & \frac{it}{B_N} ES_N + \frac{1}{2} \left(\frac{it}{B_N}\right)^2 D^2 S_N + \frac{1}{6} \left(\frac{it}{B_N}\right)^3 \Gamma_3(S_N) \\ & + \theta \varepsilon_N |t|^3 L_{3N}. \quad \square \end{aligned}$$

We return to the proof of Theorem 4.2.

Making use of Lemma 5.1 and (5.2) we see that for  $|t| \leq L_{3N}^{-\beta}$

$$\begin{aligned} f_N(it/B_N) = & e^{-t^2/2} \left\{ 1 + \frac{(it)^3}{6} \Gamma_3(S_N/B_N) + \theta_1 \varepsilon_N L_{3N} |t|^3 \right. \\ & \left. + \theta_2 \left| \frac{(it)^3}{6} \Gamma_3(S_N/B_N) + \varepsilon_N |t|^3 L_{3N} \right|^2 \exp \left( \frac{|t|^3}{6} |\Gamma_3(S_N/B_N)| + \varepsilon_N L_{3N} |t|^3 \right) \right\} \\ = & e^{-t^2/2} \left( 1 + \frac{(it)^3}{6} \Gamma_3(S_N/B_N) \right) + \theta \varepsilon_N L_{3N} (|t|^2 + |t|^5) e^{-t^2/4}. \end{aligned} \tag{5.3}$$

Now we quote a lemma which is needed for the proof of Theorem 4.2.

**Lemma 5.2.** (see [8], p.13). *Let  $F_1(x), F_2(x)$  be functions of bounded variation on the real line,*

$$\int_{-\infty}^{\infty} |x| |dF_k(x)| < \infty, \quad k=1, 2.$$

Further, suppose that  $R(-\infty) = R(+\infty) = 0$  and  $\int_{-\infty}^{\infty} \left| \frac{r(t)}{t} \right| dt < \infty$ , where  $R(x) = F_1(x) - F_2(x)$  and  $r(t) = \int_{-\infty}^{\infty} e^{itx} dR(x)$ . Then

$$\frac{1}{2} (R(x-0) - R(x+0)) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \frac{e^{-itx}}{-it} r(t) dt$$

for every  $x$ .

If we take in Lemma 5.2

$$F_1(x) = F_N(x) \quad \text{and} \quad F_2(x) = \Phi(x) + \frac{(1-x^2)\Gamma_3(S_N)}{6\sqrt{2\pi}B_N^3} e^{-x^2/2},$$

it remains to show that the integral

$$I_N = \int_{-\infty}^{\infty} \left| f_N(it/B_N) - e^{-t^2/2} \left( 1 + \frac{(it)^3}{6} \Gamma_3(S_N/B_N) \right) \right| |t|^{-1} dt$$

equals to the order  $o(L_{3N})$  as  $N \rightarrow \infty$ .

Splitting the domain of integration we obtain

$$\begin{aligned} I_N \leq & \int_{|t| \leq L_{3N}^{-\beta}} \left| f_N(it/B_N) - e^{-t^2/2} \left( 1 + \frac{(it)^3}{6} \Gamma_3(S_N/B_N) \right) \right| |t|^{-1} dt \\ & + \int_{|t| \geq (192 L_{3N})^{-1}} \left| \frac{f_N(it/B_N)}{t} \right| dt + \int_{L_{3N}^{-\beta} < |t| < (192 L_{3N})^{-1}} \left| \frac{f_N(it/B_N)}{t} \right| dt \\ & + \int_{|t| > L_{3N}^{-\beta}} e^{-t^2/2} \left| 1 + \frac{(it)^3}{6} \Gamma_3(S_N/B_N) \right| |t|^{-1} dt. \end{aligned}$$

By (3.7), (5.3) and (C) it is easily seen that  $I_N = o(L_{3N})$  as  $N \rightarrow \infty$ . This completes the proof of Theorem 4.2.  $\square$ .

For the proof of Corollary 4.2 we write  $S_N = \sum_{k=1}^{[N/m]+1} Y_k$  with the 1-dependent  $r\nu$ 's (2.8). Then the assertion of Corollary 4.2 can be proved by using (5.1) and the technique in the proof of Theorem 4.2.

**References**

1. Berk, K.N.: A central limit theorem for  $m$ -dependent random variables with unbounded  $m$ . Ann. Probab. **1**, 352-354 (1973)
2. Erickson, R.V.:  $L_1$  bounds for asymptotic normality of  $m$ -dependent sums using Stein's technique. Ann. Probab. **2**, 522-529 (1974)
3. Heinrich, L.: On a factorization of the characteristic function of a sum of dependent random variables (German). Liet. mat. rink. **22**, 190-202 (1982)
4. Heinrich, L.: On probabilities of large deviations for sums of random variables connected in a Markov chain under a non-Cramérian type condition (German). Liet. mat. rink. (To appear)
5. Heinrich, L.: A new approach to the summation theory of random variables connected in a Markov chain. Math. Nachr. (To appear, 1982)
6. Heinrich, L.: Infinitely divisible distributions as limit laws for sums of random variables connected in a Markov chain. Math. Nachr. (To appear, 1982)
7. Petrov, V.V.: On the central limit theorem for  $m$ -dependent random variables (Russian). In: Proc. All-Union-Conf. Prob. Theory and Math. Statistics, Erevan 1958, pp. 138-144. Erevan: Izdat. Akad. Nauk Armjan. SSR 1960
8. Petrov, V.V.: Sums of independent random variables. Berlin-Heidelberg-New York: Springer 1975
9. Shergin, V.V.: On the speed of convergence in the central limit theorem for  $m$ -dependent random variables (Russian). Teor. verojatn. primen. **24**, 781-794 (1979)

10. Statulevičius, V.A.: On large deviations. *Z. Wahrscheinlichkeitstheorie verw. Gebiete* **6**, 133–144 (1966)
11. Statulevičius, V.A.: On limit theorems for random functions I (Russian). *Liet. mat. rink.* **10**, 583–592 (1970)
12. Stein, C.: A bound for the error in the normal approximation to the distribution of a sum of dependent random variables. In: *Proc. 6th Berkeley Sympos. Math. Statistics Probab.*, Vol. 2, pp. 583–602. Berkeley: Univ. Calif. Press 1972
13. Tichomirov, A.N.: On the speed of convergence in the central limit theorem for weakly dependent random variables (Russian). *Teor. verojatn. primen.* **25**, 800–818 (1980)

Received December 1, 1981