

# A CENTRAL LIMIT THEOREM FOR THE SUM OF A RANDOM NUMBER OF INDEPENDENT RANDOM VARIABLES

by  
J. MOGYORÓDI<sup>1</sup>

## §. 1. Introduction

Let  $\xi_1, \xi_2, \dots, \xi_n, \dots$  denote throughout the present paper a sequence of independent and identically distributed random variables with mean value 0 and variance 1. Let us put

$$(1) \quad \zeta_n = \xi_1 + \xi_2 + \dots + \xi_n$$

and

$$(2) \quad \eta_n = \frac{\zeta_n}{\sqrt{n}}.$$

Then by the simplest case of the central limit theorem

$$(3) \quad \lim_{n \rightarrow +\infty} \mathbf{P}(\eta_n < x) = \Phi(x) \quad (-\infty < x < +\infty).$$

Here and in what follows  $\mathbf{P}(\cdot)$  denotes the probability of the event in the brackets and  $\Phi(x)$  the standard form of the normal distribution function, i.e.

$$(4) \quad \Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{y^2}{2}} dy.$$

In the present paper we shall investigate the limiting distribution of the random variables  $\eta_{v_n}$  for  $n \rightarrow +\infty$ , where  $v_n$  ( $n = 1, 2, \dots$ ) is a sequence of positive integer-valued random variables. We mention that in this paper nothing is supposed about the dependence of  $v_n$  on the random variables  $\xi_k$ .

The first results of this kind have been obtained by F. J. ANSCOMBE [1]. We mention here a special case of a more general result of the mentioned author.

**Theorem 1.** *If  $v_n$  is a sequence of positive integer-valued random variables such that  $v_n/n$  converges for  $n \rightarrow +\infty$  in probability to a constant  $c > 0$ , then*

$$(5) \quad \lim_{n \rightarrow +\infty} \mathbf{P}(\eta_{v_n} < x) = \Phi(x)$$

*holds.*

---

<sup>1</sup> Eötvös Loránd University, Mathematical Institute, Budapest.

Recently A. RÉNYI [2] generalized Theorem 1. He investigated the case when  $\frac{v_n}{n}$  tends in probability to a positive random variable  $\lambda$  having a discrete distribution. He proved the following

**Theorem 2.** *If  $v_n$  ( $n = 1, 2, \dots$ ) is a sequence of positive integer-valued random variables such that  $v_n/n$  converges in probability to a positive random variable  $\lambda$  having a discrete distribution, then (5) holds.*

In paper [6] we set ourself as an aim to investigate in a following paper the case when  $v_n/n$  converges in probability to a positive variable  $\lambda$  having arbitrary distribution.

According to this program, the aim of the present paper is to generalize RÉNYI's result by omitting the restriction that the positive random variable  $\lambda$  has a distribution of the discrete type.

We prove the following

**Theorem 3.<sup>2</sup>** *Let  $\xi_1, \xi_2, \dots, \xi_n, \dots$  be a sequence of independent and identically distributed random variables with mean value 0 and variance 1.*

*If  $v_n$  ( $n = 1, 2, \dots$ ) is a sequence of positive integer-valued random variables such that  $v_n/n$  converges in probability to a positive random variable  $\lambda$ , then (5) holds.*

## § 2. Some theorems and an inequality of Kolmogorov type

We denote by  $A \circ B$  the symmetric difference  $(A - B) + (B - A)$  of the random events  $A$  and  $B$  in the following.

**Lemma 1.** *Let  $\tau_1, \tau_2, \dots, \tau_n, \dots$  be a sequence of random variables and suppose that  $\tau_n$  converges in probability to a random variable  $\tau$ . Let further  $a_1$  and  $a_2$  ( $a_1 < a_2$ ) be continuity points of the distribution function  $F(x)$  of the random variable  $\tau$ . Let  $A$  denote the event  $\{a_1 \leq \tau < a_2\}$  and  $A_n$  the event  $\{a_1 \leq \tau_n < a_2\}$ . Then  $\mathbf{P}(A_n \circ A) \rightarrow 0$  if  $n \rightarrow +\infty$ .*

**Proof.** Given any  $\varepsilon > 0$  we can choose (because of the continuity of  $F(x)$  at the points  $a_1$  and  $a_2$ ) a positive number  $\delta$  such that  $|F(x) - F(a_i)| < \varepsilon$  if  $|x - a_i| \leq \delta$  ( $i = 1, 2$ ).

We have the following sequence of equalities and inequalities

$$\begin{aligned} \mathbf{P}(A_n \bar{A}) &= \mathbf{P}(A_n \bar{A}, \tau < a_1 - \delta) + \mathbf{P}(A_n \bar{A}, \tau \geq a_2 + \delta) + \\ &\quad + \mathbf{P}(A_n \bar{A}, a_1 - \delta \leq \tau < a_1) + \mathbf{P}(A_n \bar{A}, a_2 \leq \tau \leq a_2 + \delta) \leq \\ &\leq \mathbf{P}(|\tau_n - \tau| > \delta) + F(a_1) - F(a_1 - \delta) + F(a_2 + \delta) - F(a_2) \end{aligned}$$

and

$$\begin{aligned} \mathbf{P}(\bar{A}_n A) &= \mathbf{P}(\bar{A}_n A, a_1 + \delta \leq \tau < a_2 - \delta) + \mathbf{P}(\bar{A}_n A, a_1 \leq \tau < a_1 + \delta) + \\ &\quad + \mathbf{P}(\bar{A}_n A, a_2 - \delta \leq \tau < a_2) \leq \mathbf{P}(|\tau_n - \tau| > \delta) + \\ &\quad + F(a_1 + \delta) - F(a_1) + F(a_2) - F(a_2 - \delta). \end{aligned}$$

Taking into account that  $\tau_n$  converges in probability to  $\tau$  we can choose the positive integer  $n_0$  such that if  $n \geq n_0$  we have  $\mathbf{P}(|\tau_n - \tau| > \delta) < \varepsilon$ .

<sup>2</sup> Professor RÉNYI kindly informed me that in the paper "On the strong law of large numbers for a class of stochastic processes" written by J. R. BLUM, D. L. HANSON, and L. H. KOOPMANS the same theorem is proved independently of me. This paper is to be appear in "Zeitschrift für Wahrscheinlichkeitstheorie und verwandte Gebiete". (Springer-verlag, Berlin — Göttingen — Heidelberg, 1962).



Thus we have  $\mathbf{P}(A_n \bar{A}) < 3\varepsilon$  and  $\mathbf{P}(\bar{A}_n A) < 3\varepsilon$  and so  $\mathbf{P}(A_n \circ A) < 6\varepsilon$ . This gives our assertion.

We shall use this Lemma 1 in the following form:

**Consequence.** If  $[a_1, b_1), [a_2, b_2), \dots, [a_k, b_k)$  are disjoint intervals ( $k = 1, 2, \dots$ ),  $a_i, b_i$  ( $i = 1, 2, \dots, k$ ) are continuity points of the distribution function of  $\tau$  and  $A^{(i)}$  denotes the event  $\{a_i \leq \tau < b_i\}$  and  $A_n^{(i)}$  the event  $\{a_i \leq \tau_n < b_i\}$  ( $i = 1, 2, \dots, k$ ) then for any fixed positive integer  $k$

$$\sum_{i=1}^k \mathbf{P}(A_n^{(i)} \circ A^{(i)}) \rightarrow 0, \text{ if } n \rightarrow +\infty.$$

The following Lemma is almost trivial.

**Lemma 2.** If  $\tau_1, \tau_2, \dots, \tau_n, \dots$  is a sequence of random variables which converges in probability to a random variable  $\tau$  and  $a$  is a continuity point of the distribution function of  $\tau$  such that  $\mathbf{P}(\tau < a) < \varepsilon$  (resp.  $1 - \mathbf{P}(\tau < a) < \varepsilon$ ) then there exists a positive integer  $n_0$  such that for  $n \geq n_0$

$$\mathbf{P}(\tau_n < a) < 2\varepsilon. \quad (\text{resp. } 1 - \mathbf{P}(\tau_n < a) < 2\varepsilon).$$

**Proof.** The assertion follows at once from the fact that  $a$  is a continuity point of the distribution function of  $\tau$  and from the following equalities and inequalities:

$$\begin{aligned} \mathbf{P}(\tau_n < a) &= \mathbf{P}(\tau_n < a, |\tau_n - \tau| \leq \delta) + \mathbf{P}(\tau_n < a, |\tau_n - \tau| > \delta) \leq \\ &\leq \mathbf{P}(\tau - \delta < a) + \varepsilon, \text{ if } n \geq n_0. \end{aligned}$$

The case when  $1 - \mathbf{P}(\tau < a) < \varepsilon$ , can be treated similarly.

We formulate a theorem of A. RÉNYI [4].

**Theorem 4.** If  $\tau_n$  is a sequence of independent random variables such that putting

$$\sigma_n = \frac{1}{B_n} \sum_{k=1}^n \tau_k, \quad \text{where } B_n \rightarrow +\infty$$

the distribution of the random variable  $\sigma_n$  tends to a limiting distribution, then the conditional distribution of  $\sigma_n$  under any condition having positive probability, tends to the same limiting distribution.

**Theorem 5.** Let  $\tau_1, \tau_2, \dots, \tau_n, \dots$  be a sequence of random variables such that for  $n = 1, 2, \dots$   $\mathbf{M}(\tau_n)$  exists<sup>3</sup> and

$$\mathbf{M}(\tau_n) \rightarrow M \quad (|M| < +\infty)$$

as  $n \rightarrow +\infty$ .

Let us suppose that

a)

$$\lim_{N \rightarrow +\infty} \int_{|\tau_n| > N} |\tau_n| d\mathbf{P} = 0$$

<sup>3</sup> Here and in what follows  $\mathbf{M}(\cdot)$  and  $\mathbf{M}(\cdot|A)$  denote the mathematical expectation and the conditional mathematical expectation resp. of the random variable in the brackets.

uniformly in  $n$ .

$$\text{b) } \lim_{n \rightarrow +\infty} |\mathbf{P}(\tau_n < x | A) - \mathbf{P}(\tau_n < x)| = 0$$

where  $A$  is an event such that  $1 > \mathbf{P}(A) > 0$ .

Then

$$\lim_{n \rightarrow +\infty} \mathbf{M}(\tau_n | A) = M.$$

**Proof.** We have

$$\mathbf{M}(\tau_n | A) = \frac{1}{\mathbf{P}(A)} \left\{ \int_{(A \cap \{|\tau_n| \leq N\})} \tau_n d\mathbf{P} + \int_{(A \cap \{|\tau_n| > N\})} \tau_n d\mathbf{P} \right\}$$

where  $N$  is an arbitrary positive number.

Condition a) ensures that for any  $\varepsilon > 0$  we can choose a number  $N = N(\varepsilon, A)$  such that

$$(6) \quad \left| \frac{1}{\mathbf{P}(A)} \int_{(A \cap \{|\tau_n| > N(\varepsilon)\})} \tau_n d\mathbf{P} \right| \leq \frac{1}{\mathbf{P}(A)} \int_{(|\tau_n| > N(\varepsilon))} |\tau_n| d\mathbf{P} < \varepsilon$$

holds for  $n = 1, 2, \dots$ .

Thus we have

$$(7) \quad \left| \mathbf{M}(\tau_n | A) - \frac{1}{\mathbf{P}(A)} \int_{(A \cap \{|\tau_n| \leq N(\varepsilon)\})} \tau_n d\mathbf{P} \right| < \varepsilon. \quad (n = 1, 2, \dots).$$

It is easy to see that for  $n = 1, 2, \dots$

$$\frac{1}{\mathbf{P}(A)} \int_{\{A \cap \{|\tau_n| \leq N(\varepsilon)\}\}} \tau_n d\mathbf{P} \geq \sum_{|kh_0| \leq N(\varepsilon)} kh_0 \mathbf{P}(kh_0 \leq \tau_n < (k+1)h_0 | A)$$

where  $h_0$  is a fixed positive number and  $h_0 < \varepsilon$ .

Let

$$\varepsilon' = \frac{\varepsilon}{N(\varepsilon) \left[ \frac{N(\varepsilon)}{h_0} \right]},$$

Then it follows from condition b) that we can choose an index  $n_0 = n_0(\varepsilon') = n_0(\varepsilon)$  such that for  $n \geq n_0$  the following inequality holds:

$$|\mathbf{P}(kh_0 \leq \tau_n < (k+1)h_0 | A) - \mathbf{P}(kh_0 \leq \tau_n < (k+1)h_0)| < \varepsilon'$$

for any integer  $k$  such that  $|kh_0| \leq N(\varepsilon)$ .

Using this inequality we obtain for  $n \geq n_0$

$$\frac{1}{\mathbf{P}(A)} \int_{\{A \cap \{|\tau_n| \leq N(\varepsilon)\}\}} \tau_n d\mathbf{P} \geq \sum_{|kh_0| \leq N(\varepsilon)} kh_0 \mathbf{P}(kh_0 \leq \tau_n < (k+1)h_0) - \varepsilon.$$



Thus from (7) we see that for  $n \geq n_0$

$$(8) \quad \mathbf{M}(\tau_n | A) \geq \sum_{|kh_0| \leq N(\varepsilon)} kh_0 \mathbf{P}(kh_0 \leq \tau_n < (k+1)h_0) - 2\varepsilon.$$

On the other hand it follows from (6) that

$$|\mathbf{M}(\tau_n) - \int_{(|\tau_n| \leq N(\varepsilon))} \tau_n d\mathbf{P}| < \varepsilon \quad (n = 1, 2, \dots)$$

and we have for  $n = 1, 2, \dots$

$$\left| \int_{(|\tau_n| \leq N(\varepsilon))} \tau_n d\mathbf{P} - \sum_{|kh_0| \leq N(\varepsilon)} kh_0 \mathbf{P}(kh_0 \leq \tau_n < (k+1)h_0) \right| \leq h_0 < \varepsilon.$$

Confering these last two inequalities with (8) we obtain for  $n \geq n_0$

$$(9) \quad \mathbf{M}(\tau_n | A) \geq \mathbf{M}(\tau_n) - 4\varepsilon.$$

It follows from (9) that

$$(10) \quad \liminf_{n \rightarrow +\infty} \mathbf{M}(\tau_n | A) \geq M.$$

Now condition b) ensures that

$$\lim_{n \rightarrow +\infty} |\mathbf{P}(\tau_n < x | \bar{A}) - \mathbf{P}(\tau_n < x)| = 0.$$

Similarly, as in the preceding argumentation, one shows that

$$(11) \quad \liminf_{n \rightarrow +\infty} \mathbf{M}(\tau_n | \bar{A}) \geq M.$$

Now from (11) and from the equality

$$\mathbf{M}(\tau_n) = \mathbf{M}(\tau_n | A) \mathbf{P}(A) + \mathbf{M}(\tau_n | \bar{A}) \mathbf{P}(\bar{A})$$

we deduce that

$$(12) \quad \limsup_{n \rightarrow +\infty} \mathbf{M}(\tau_n | A) \leq M.$$

(10) and (12) together give our assertion.

**Theorem 6.** Let  $\vartheta_1, \vartheta_2, \dots$  be a sequence of independent random variables with mean value 0 and variances  $D_1, D_2, \dots$ . Put

$$\sigma_n = \frac{\vartheta_1 + \vartheta_2 + \dots + \vartheta_n}{B_n}$$

(where  $B_n^2 = D_1^2 + D_2^2 + \dots + D_n^2$  and  $B_n \rightarrow +\infty$ ). Let us suppose that the distribution function  $F_n(x)$  of the random variable  $\sigma_n$  tends to a non-degenerate limiting distribution function  $F(x)$ . Let us suppose further that

$$\mathbf{M}(\sigma_n^2) = \int_{-\infty}^{+\infty} x^2 dF(x).$$

Then under any condition  $A$ , having positive probability, we have

$$\mathbf{M}(\sigma_n^2 | A) \rightarrow 1 \quad (\text{as } n \rightarrow +\infty).$$

**Proof.** We prove that under conditions of our theorem

$$\lim_{N \rightarrow +\infty} \int_{|x| \geq N} x^2 dF_n(x) = 0$$

holds uniformly in  $n$ .

For this purpose let us choose the positive number  $N(\varepsilon)$  according to the given positive number  $\varepsilon$  such that for  $N \geq N(\varepsilon)$

$$\int_{|x| \geq N} x^2 dF(x) < \frac{\varepsilon}{4}$$

be satisfied. Then we have for these  $N$

$$\left| 1 - \int_{|x| < N} x^2 dF(x) \right| < \frac{\varepsilon}{4}.$$

On the other hand, since  $F_n(x)$  converges to  $F(x)$  at all continuity points of  $F(x)$ , we have for  $n \geq n_0(\varepsilon)$

$$\left| \int_{|x| \leq N} x^2 dF_n(x) - \int_{|x| \leq N} x^2 dF(x) \right| < \frac{\varepsilon}{4}.$$

Thus for  $n \geq n_0(\varepsilon)$  and for  $N \geq N(\varepsilon)$  the following inequality holds

$$(13) \quad \left| 1 - \int_{|x| \leq N} x^2 dF_n(x) \right| < \frac{\varepsilon}{2}.$$

Further from the condition

$$\int_{-\infty}^{+\infty} x^2 dF_n(x) = \int_{-\infty}^{+\infty} x^2 dF(x)$$

we deduce that

$$(14) \quad \left| \int_{-\infty}^{+\infty} x^2 dF_n(x) - \int_{-\infty}^{+\infty} x^2 dF(x) \right| < \frac{\varepsilon}{2},$$

if  $n \geq n_1(\varepsilon)$ . Let  $n(\varepsilon) = \max(n_0(\varepsilon), n_1(\varepsilon))$ . Confering (13) with (14) we conclude

$$0 \leq \int_{|x| \geq N} x^2 dF_n(x) < \varepsilon,$$

if  $n \geq n(\varepsilon)$  and  $N \geq N(\varepsilon)$ . This essentially means the asserted uniform integrability.

We are now in the position to prove the assertion of our theorem. According to the preceding note we can choose for any  $n$  and for any positive  $\varepsilon$  a



positive number  $N(\varepsilon)$  such that

$$0 \leq \int_{|x| \geq N(\varepsilon)} x^2 dF_n(x) \leq \varepsilon \quad (n = 1, 2, \dots)$$

and

$$0 \leq \int_{|x| \geq N(\varepsilon)} x^2 dF(x) \leq \varepsilon.$$

Using this inequality we obtain for any fixed condition  $A$ , having positive probability,

$$(15) \quad 0 \leq \int_{|x| \geq N(\varepsilon)} x^2 dF_n(x | A) \leq \frac{1}{P(A)} \int_{|x| \geq N(\varepsilon)} x^2 dF_n(x) \leq \frac{\varepsilon}{P(A)}, \quad (n = 1, 2, \dots).$$

Thus it is sufficient to deal with the convergence of the integrals

$$\int_{|x| \leq N(\varepsilon)} x^2 dF_n(x | A).$$

Now according to the assertion of Theorem 4 we have

$$F_n(x | A) \rightarrow F(x)$$

at all continuity points of  $F(x)$ .

This means that

$$(16) \quad \int_{|x| \leq N(\varepsilon)} x^2 dF_n(x | A) \rightarrow \int_{|x| \leq N(\varepsilon)} x^2 dF(x).$$

Taking into account the choosing of  $N(\varepsilon)$  and conferring (15) with (16) we obtain the assertion of our Theorem.

**Theorem 7.** Let  $\tau_1, \tau_2, \dots, \tau_n, \dots$  be a sequence of independent random variables with mean values  $\mathbf{M}(\tau_i) = M_i$  and variances  $\mathbf{D}(\tau_i) = D_i$ . Let further

$$\sigma_n = \frac{\tau_1 + \dots + \tau_n - (M_1 + \dots + M_n)}{S_n}$$

where

$$S_n = \sqrt{D_1^2 + D_2^2 + \dots + D_n^2}.$$

Let us suppose that the distribution function  $F_n(x)$  of random variable  $\sigma_n$  converges to the non-degenerate distribution function  $F(x)$  with variance 1. Let further  $C$  be an arbitrary random event having positive probability. Then there exists an integer  $n_0 = n_0(C)$  such that for  $n \geq n_0$

$$P\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k (\tau_i - M_i) \right| \geq \lambda S_n, C\right) \leq \frac{3\sqrt{P(C)}}{\lambda^2}$$

where  $\lambda$  is an arbitrary positive number.

**Proof.** Let  $\vartheta_k = \sum_{i=1}^k (\tau_i - M_i)$  and let  $\gamma$  denote the indicator of the event  $C$ , i.e.

$$\gamma = \begin{cases} 1, & \text{if } C \text{ occurs} \\ 0, & \text{otherwise.} \end{cases}$$

Let further  $A_k$  ( $k = 2, \dots, n$ ) denote the following event:

$$|\vartheta_i| \leq \lambda S_n, \quad (i = 1, 2, \dots, k-1) \text{ and } |\vartheta_k| \geq \lambda S_n,$$

and  $A_1$  the following event:  $|\vartheta_1| \geq \lambda S_n$ ,

and let  $\alpha_k$  be the indicator of the event  $A_k$ . Then

$$\alpha_i \alpha_j = 0, \quad \text{if } i \neq j, \quad 0 \leq \sum_{k=1}^n \alpha_k \leq 1 \text{ and}$$

$$\mathbf{M} \left[ \left( \sum_{k=1}^n \alpha_k \right) \gamma \right] = \mathbf{P} \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k (\tau_i - M_i) \right| \geq \lambda S_n, C \right).$$

Let us investigate the mean value of  $\gamma \vartheta_n^2$ . It is easy to see that

$$\mathbf{M}(\gamma \vartheta_n^2) \geq \sum_{k=1}^n \mathbf{M}(\gamma \alpha_k \vartheta_k^2) + 2 \sum_{k=1}^{n-1} \mathbf{M}(\gamma \alpha_k \vartheta_k (\vartheta_n - \vartheta_k)).$$

The second sum of the right-hand side can be written as follows

$$\mathbf{M} \left( \gamma \sum_{k=1}^{n-1} \alpha_k \vartheta_k (\vartheta_n - \vartheta_k) \right) = \mathbf{M} \left( \gamma \sum_{j=2}^n \beta_j \right),$$

where

$$\beta_j = (\tau_j - M_j) \sum_{k=1}^{j-1} \alpha_k \vartheta_k. \quad (j = 2, 3, \dots, n).$$

The system  $\{\beta_j\}$  is suborthonormal ( $j = 2, 3, \dots, n$ ).

For, if  $j \neq i$  and  $j < i$ , then  $(\tau_i - M_i)$  is independent of all the random variables playing role in the product  $\beta_j \beta_i$ , and thus we have  $\mathbf{M}(\beta_j \beta_i) = 0$ .

On the other hand

$$\mathbf{M}(\beta_j^2) = \mathbf{M}((\tau_j - M_j)^2) \mathbf{M} \left( \left( \sum_{k=1}^{j-1} \alpha_k \vartheta_k \right)^2 \right) \leq D_j^2 S_n^2 < +\infty.$$

Thus applying the inequality of Cauchy and Bessel resp. we obtain

$$\begin{aligned} \left| \mathbf{M} \left( \gamma \sum_{j=2}^n \beta_j \right) \right| &= \left| \sum_{j=2}^n \mathbf{M} \left( \gamma \frac{\beta_j}{\sqrt{\mathbf{M}(\beta_j^2)}} \sqrt{\mathbf{M}(\beta_j^2)} \right) \right| \leq \\ &\leq \sqrt{\sum_{j=2}^n \mathbf{M}^2 \left( \gamma \frac{\beta_j}{\sqrt{\mathbf{M}(\beta_j^2)}} \right) \sum_{j=2}^n \mathbf{M}(\beta_j^2)} \leq \sqrt{\mathbf{M}(\gamma^2)} \sqrt{S_n^2 \sum_{j=1}^n D_j^2} \leq \sqrt{\mathbf{P}(C)} S_n^2. \end{aligned}$$



Taking into account this inequality we can write

$$(17) \quad \mathbf{M}(\gamma \vartheta_n^2) \geq \sum_{k=1}^n \mathbf{M}(\gamma \alpha_k) \lambda^2 S_n^2 - 2 \sqrt{\mathbf{P}(C)} S_n^2.$$

On the other hand according to Theorem 6 we have

$$\lim_{n \rightarrow +\infty} \mathbf{M}(\gamma \sigma_n^2) = \lim_{n \rightarrow +\infty} \mathbf{M}(\sigma_n^2 | \gamma = 1) \mathbf{P}(C) = \mathbf{P}(C).$$

Thus there is an index  $n_0 = n_0(C)$  such that if  $n \geq n_0$

$$\mathbf{M}(\gamma \sigma_n^2) \leq \sqrt{\mathbf{P}(C)}$$

or

$$(18) \quad \mathbf{M}(\gamma \vartheta_n^2) \leq S_n^2 \sqrt{\mathbf{P}(C)}.$$

Confering (17) with (18) we get the assertion.

**Remark.** Theorem 7 is in some sense an extension of the celebrated Kolmogorov inequality. Its deficiency however is that we postulate the convergence of the distribution functions  $\mathbf{P}(\sigma_n < x)$  to a non-degenerate distribution function  $F(x)$ . This requirement is eliminated in a theorem of A. RÉNYI [5] but on the other hand the existence of the fourth moment of random variables  $\tau_n$  is postulated in his theorem. We give here a generalized form of the theorem of A. RÉNYI, where only the existence of the  $2 + p$ -adic ( $p$  is an arbitrary positive number) moment of the random variables  $\tau_n$  is required.

**Theorem 8.** Let  $\tau_1, \tau_2, \dots, \tau_n$  be independent random variables and suppose that their moment of order  $2 + p$  ( $p$  is an arbitrary positive number) exists. Let further  $C$  be any event of positive probability. If  $M_i, S_n, \lambda$  denote the same quantities as in the preceding theorem then the following inequality holds for any  $n = 1, 2, \dots$

$$\begin{aligned} & \mathbf{P} \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k (\tau_i - M_i) \right| \geq \lambda S_n, C \right) \leq \\ & \leq \frac{\mathbf{P}(C)^{\frac{p}{2+p}}}{\lambda^2} \left[ \frac{B_p \sum_{j=1}^n (|\tau_j - M_j|^{2+p})^{\frac{2}{2+p}}}{S_n^2} + 2 \right] \end{aligned}$$

where  $B_p$  is a positive constant depending only on the parameter  $p$ .

The proof of this theorem can be performed in somewhat similar way as that of the theorem of A. RÉNYI.

### § 3. Proof of theorem 3

Let us consider the distribution function of the positive random variable  $\lambda$ . Given any  $\varepsilon > 0$  we can choose a sufficiently small  $a > 0$  and a sufficiently large  $b > 0$  such that  $a$  and  $b$  are continuity points of the distribution function of  $\lambda$  and

$$\mathbf{P}(a \leq \lambda < b) > 1 - \varepsilon.$$

Then taking into account that  $v_n/n$  converges in probability to  $\lambda$ , there exists by virtue of Lemma 2 a positive integer  $n_0$  such that for  $n \geq n_0$

$$\mathbf{P}\left(a \leq \frac{v_n}{n} < b\right) > 1 - 2\varepsilon.$$

Denoting by  $A_n$  the event  $a \leq \frac{v_n}{n} < b$  we can write

$$\mathbf{P}(\eta_{v_n} < x) = \mathbf{P}(\eta_{v_n} < x, A_n) + \mathbf{P}(\eta_{v_n} < x, \bar{A}_n).$$

The second member of the right-hand side is smaller than  $2\varepsilon$ , and thus we have only to deal with the first member. Let us divide the interval  $[a, b)$  into  $k$  subintervals by the points  $a_1 < a_2 < \dots < a_{k-1}$  ( $a < a_1, a_{k-1} < b$ ) and suppose that they are continuity points of the distribution function of  $\lambda$ . We introduce the following notations:

$$a_0 = a, a_k = b, \text{ and } A_n^{(i)} = \left\{a_{i-1} \leq \frac{v_n}{n} < a_i\right\} \quad (i = 1, 2, \dots, k).$$

Then  $\sum_{i=1}^k A_n^{(i)} = A_n$  and  $A_n^{(i)} \cdot A_n^{(j)} = \emptyset$ , if  $i \neq j$ .

We can write

$$(19) \quad \mathbf{P}(\eta_{v_n} < x, A_n) = \sum_{i=1}^k \mathbf{P}(\eta_{v_n} < x, A_n^{(i)}).$$

We have clearly

$$(20) \quad \eta_{v_n} = \eta_{[na_{i-1}]} + \sqrt{\frac{[na_{i-1}]}{v_n}} \left( \frac{\zeta_{v_n} - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right) + \eta_{[na_{i-1}]} \left( \sqrt{\frac{[na_{i-1}]}{v_n}} - 1 \right).$$

( $i = 1, 2, \dots, k$ ).

Let us denote by  $C(n, k, i, \varrho)$  the event that

$$\left| \sqrt{\frac{[na_{i-1}]}{v_n}} \left( \frac{\zeta_{v_n} - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right) + \eta_{[na_{i-1}]} \left( \sqrt{\frac{[na_{i-1}]}{v_n}} - 1 \right) \right| < 2\varrho \quad (i = 1, 2, \dots, k).$$

Let us choose the positive number  $\varrho$  such that the following inequality

$$|\Phi(x) - \Phi(x \pm 2\varrho)| < \varepsilon$$

be satisfied.

Let further the positive integer  $k$  be chosen such that

$$\frac{12(b-a)}{\varrho^2 \sqrt{k}} < \frac{\varepsilon}{2}$$

be satisfied.

On the basis of (19) we can write

$$(21) \quad \begin{aligned} \mathbf{P}(\eta_{v_n} < x, A_n) &= \sum_{i=1}^k \mathbf{P}(\eta_{v_n} < x, A_n^{(i)}, C(n, k, i, \varrho)) + \\ &+ \sum_{i=1}^k \mathbf{P}(\eta_{v_n} < x, A_n^{(i)}, \overline{C(n, k, i, \varrho)}). \end{aligned}$$



Applying now the inequalities

$$\mathbf{P}(ABC) \geq \mathbf{P}(AB) + \mathbf{P}(AC) - \mathbf{P}(A) = \mathbf{P}(AB) - \mathbf{P}(A\bar{C}), \quad \text{and} \quad \mathbf{P}(AB) \leq \mathbf{P}(A)$$

and taking into account (20) we have

$$(22) \quad \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x - 2\varrho, A_n^{(i)}) - \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) \leq \\ \leq \mathbf{P}(\eta_{v_n} < x, A_n) \leq \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x + 2\varrho, A_n^{(i)}) + \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}).$$

It follows from the consequence of Lemma 1 that for any  $\varepsilon > 0$  there exists a positive integer  $n_1$  ( $n_1 \geq n_0$ ) such that if  $A^{(i)}$  denotes the event  $(a_{i-1} \leq \lambda < a_i)$  ( $i = 1, 2, \dots, k$ ) then

$$(23) \quad \sum_{i=1}^k \mathbf{P}(A_n^{(i)} \circ A^{(i)}) < \frac{\varepsilon}{2} \quad \text{if } n \geq n_1.$$

Obviously we have for any three event

$$|\mathbf{P}(AB) - \mathbf{P}(AC)| \leq \mathbf{P}(B \circ C)$$

and thus for  $n \geq n_1$

$$(24) \quad \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x \pm 2\varrho, A^{(i)}) - \varepsilon \leq \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x \pm 2\varrho, A_n^{(i)}) \leq \\ \leq \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x \pm 2\varrho, A^{(i)}) + \varepsilon.$$

We obtain thus the following estimate for  $\mathbf{P}(\eta_{v_n} < x, A_n)$

$$(25) \quad \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x - 2\varrho, A^{(i)}) - \varepsilon - \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) \leq \mathbf{P}(\eta_{v_n} < x, A_n) \leq \\ \leq \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x + 2\varrho, A^{(i)}) + \varepsilon + \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)})$$

if  $n \geq n_1$ .

Since the distribution of random variables  $\eta_{[na_{i-1}]}$  converges to (4), we have by Theorem 4

$$(26) \quad \lim_{n \rightarrow +\infty} \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x \pm 2\varrho, A^{(i)}) = \Phi(x \pm 2\varrho) \sum_{i=1}^k \mathbf{P}(A^{(i)}) = \\ = \Phi(x \pm 2\varrho) \mathbf{P}(a \leq \lambda < b)$$

if  $k$  is the fixed integer.

This means that there exists a positive integer  $n_2$  ( $n_2 \geq n_1$ ) such that if  $n \geq n_2$ ,

$$\left| \sum_{i=1}^k \mathbf{P}(\eta_{[na_{i-1}]} < x \pm 2\varrho, A^{(i)}) - \Phi(x \pm 2\varrho) \mathbf{P}(a \leq \lambda < b) \right| < \varepsilon.$$

From this inequality and from (25) we have

$$(27) \quad \begin{aligned} & \Phi(x - 2\varrho) \mathbf{P}(a \leq \lambda < b) - 2\varepsilon - \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) \leq \mathbf{P}(\eta_{v_n} < x, A_n) \leq \\ & \leq \Phi(x + 2\varrho) \mathbf{P}(a \leq \lambda < b) + 2\varepsilon + \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, C(n, k, i, \varrho)) \end{aligned}$$

if  $n \geq n_2$ . We have

$$\mathbf{P}(a \leq \lambda < b) > 1 - \varepsilon$$

and thus for  $n \geq n_2$  and the chosen  $\varrho$

$$(28) \quad \begin{aligned} & \Phi(x) - 4\varepsilon - \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) \leq \mathbf{P}(\eta_{v_n} < x, A_n) \leq \\ & \leq \Phi(x) + 3\varepsilon + \sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) . \end{aligned}$$

The proof will be completed if we prove that for any  $\varepsilon > 0$  there exists  $n_3$  ( $n_3 \geq n_2$ ) such that for  $n \geq n_3$

$$\sum_{i=1}^k \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) < \varepsilon .$$

For this purpose we remark that the following inequality holds

$$(29) \quad \begin{aligned} & \mathbf{P}(A_n^{(i)}, \overline{C(n, k, i, \varrho)}) \leq \mathbf{P}\left(A_n^{(i)}, \left| \frac{\sqrt{[na_{i-1}]}}{v_n} \left( \frac{\zeta_{v_n} - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right) \right| \geq \varrho \right) + \\ & + \mathbf{P}\left(A_n^{(i)}, \left| \eta_{[na_{i-1}]} \left( \frac{\sqrt{[na_{i-1}]}}{v_n} - 1 \right) \right| \geq \varrho \right) . \end{aligned}$$

First of all we remark that if the event  $A_n^{(i)}$  takes place then

$$0 \leq \frac{\sqrt{[na_{i-1}]}}{v_n} \leq 1 .$$

Thus we have only to estimate the sum of the probabilities

$$(30) \quad \mathbf{P}\left(\left| \frac{\zeta_{v_n} - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A_n^{(i)}\right)$$

where  $\varrho$  is the fixed positive number, ( $i = 1, 2, \dots, k$ ). It can be easily seen that (30) is smaller than

$$(31) \quad \mathbf{P}\left(\max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A_n^{(i)}\right) .$$

It follows from (31) by the aid of the inequality

$$|\mathbf{P}(AB) - \mathbf{P}(AC)| \leq \mathbf{P}(B \circ C)$$



and by (23) that for  $n \geq n_3(\varepsilon)$

$$(32) \quad \sum_{i=1}^k \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A_n^{(i)} \right) \leq \\ \leq \sum_{i=1}^k \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A^{(i)} \right) + \frac{\varepsilon}{2}.$$

Applying Theorem 7 to the terms of the right-hand side of (32) we obtain for  $n \geq n_4$  ( $n_4 \geq n_3$ )

$$(33) \quad \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[n(a_i - a_{i-1})]}} \right| \geq \varrho \sqrt{\frac{[na_{i-1}]}{[n(a_i - a_{i-1})]}}, A^{(i)} \right) \leq \\ \leq 3 \sqrt{\mathbf{P}(A^{(i)})} \frac{[n(a_i - a_{i-1})]}{\varrho^2 [na_{i-1}]}.$$

Since the set of the continuity points of the distribution function of  $\lambda$  is everywhere dense, we can choose the points of subdivision such that for any  $\delta > 0$

$$a_i - a_{i-1} = (b - a) \frac{1 + \varepsilon_i \delta}{k} \quad (i = 1, 2, \dots, k)$$

holds with some  $\varepsilon_i$ ,  $|\varepsilon_i| \leq 1$ .

Putting now this subdivision of  $[a, b]$  we obtain from (33) the following inequality

$$(34) \quad \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A^{(i)} \right) \leq 3 \sqrt{\mathbf{P}(A^{(i)})} \frac{n(b - a)}{\varrho^2 [na]} \frac{1 + \delta}{k}.$$

Let us choose  $n_5$  ( $n_5 \geq n_4$ ) such that for  $n \geq n_5$  the inequality

$$\frac{n}{[na]} < \frac{1}{a - \delta}$$

be satisfied.

Then (34) gives

$$(35) \quad \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A^{(i)} \right) \leq \frac{K \sqrt{\mathbf{P}(A^{(i)})}}{\varrho^2 k}$$

where  $K$  is a positive constant independent of  $n$ . Substituting (35) in (32) we have for

$$(36) \quad \sum_{i=1}^k \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A_n^{(i)} \right) \leq \frac{K}{\varrho^2 k} \left( \sum_{i=1}^k \sqrt{\mathbf{P}(A^{(i)})} \right) + \frac{\varepsilon}{2}.$$

The Cauchy inequality gives

$$\sum_{i=1}^k \sqrt{\mathbf{P}(A^{(i)})} \leq \sqrt{\sum_{i=1}^k \mathbf{P}(A^{(i)})} \sqrt{k} \leq \sqrt{k}.$$

Taking into account the choice of  $k$  and the preceding estimate we obtain for  $n \geq n_5$

$$(37) \quad \sum_{i=1}^k \mathbf{P} \left( \max_{na_{i-1} \leq l \leq na_i} \left| \frac{\zeta_l - \zeta_{[na_{i-1}]}}{\sqrt{[na_{i-1}]}} \right| \geq \varrho, A_n^{(i)} \right) \leq \varepsilon.$$

Our next aim is to estimate the sum of the probabilities

$$(38) \quad \mathbf{P} \left( \left| \eta_{[na_{i-1}]} \left( \sqrt{\frac{[na_{i-1}]}{v_n}} - 1 \right) \right| \geq \varrho, A_n^{(i)} \right), \quad (i = 1, 2, \dots, k).$$

Arguing again similarly as in the preceding estimation (37) we obtain that for  $n \geq n_6$  ( $n_6 \geq n_5$ ) and for the chosen  $k$

$$(38) \quad \sum_{i=1}^k \mathbf{P} \left( \left| \eta_{[na_{i-1}]} \left( \sqrt{\frac{[na_{i-1}]}{v_n}} - 1 \right) \right| \geq \varrho, A_n^{(i)} \right) \leq \varepsilon.$$

Conferring now (28), (29), (37) and (38) we obtain the assertion of Theorem 3.

#### § 4. Some additional remarks

It is easy to see that the above method of proof of Theorem 3 can be applied to the case also when the random variables  $\xi_k$  are not identically distributed, but are such that the distribution of the normed sums of random variables  $\xi_k$  tends to a non-degenerate limiting distribution. Especially the following generalization of Theorem 3 is valid.

**Theorem 9.** Let  $\xi_1, \xi_2, \dots, \xi_n, \dots$  be a sequence of independent random variables with mean value 0 (this is not an essential restriction) and variance  $\text{Var } \xi_k = D_k$ . Let us denote by  $\eta_n$  the expression

$$\frac{\xi_1 + \dots + \xi_n}{B_n}$$

where

$$B_n = \sqrt{\sum_{k=1}^n D_k^2}.$$

Let us suppose that the distribution of  $\eta_n$  tends to a non-degenerate limiting distribution and that the sequence  $B_n$  is in the sense of KARAMATA „slowly oscillating” i. e.

$$B_n = n^\alpha L(n), \quad (\alpha > 0)$$

where for any  $c > 0$ ,  $L(cn)/L(n) \rightarrow 1$ , if  $n \rightarrow +\infty$ . Let further  $v_n$  be a sequence of positive integer-valued random variables such that  $\frac{v_n}{n}$  converges in probability to  $\lambda$ . Then  $\eta_{v_n}$  has the same limiting distribution as  $\eta_n$ .

The proof of this theorem can be performed in the same way as that of Theorem 3.



We remark also that the supposition that  $\frac{v_n}{n}$  converges in probability to  $\lambda$  can be replaced by the more general supposition that  $\frac{v_n}{\omega(n)}$  converges in probability to  $\lambda$  where  $\omega(n)$  is an arbitrary positive function tending to infinity for  $n \rightarrow +\infty$ .

(Received October 20, 1961; in revised from August 10, 1962)

## REFERENCES

- [1] ANSCOMBE, F. J.: „Large sample theory of sequential estimation”. *Proc. Cambridge Phil. Soc.* **48** (1952) 600.
- [2] RÉNYI, A.: „On the central limit theorem for the sum of a random number of independent random variables”. *Acta Math. Acad. Sci. Hung.* **11** (1960) 97—102.
- [3] RÉNYI, A.: „On the asymptotic distribution of the sum of a random number of independent random variables”. *Acta Math. Acad. Sci. Hung.* **8** (1957) 193—199.
- [4] RÉNYI, A.: „On mixing sequences of sets.” *Acta Math. Acad. Sci. Hung.* **9** (1958) 215—228.
- [5] RÉNYI, A.: „On Kolmogoroff's inequality.” *Publications of the Mathematical Institute of the Hungarian Academy of Sciences* **6** (1961) Series A, 411—415.
- [6] MOGYORÓDI, J.: „On limiting distributions for sums of a random number of independent random variables”. *Publications of the Mathematical Institute of the Hungarian Academy of Sciences* **6** (1961) Series A, 365—371.

## ЦЕНТРАЛЬНАЯ ПРЕДЕЛЬНАЯ ТЕОРЕМА ДЛЯ СУММ СЛУЧАЙНОГО ЧИСЛА НЕЗАВИСИМЫХ СЛУЧАЙНЫХ ВЕЛИЧИН

J. MOGYORÓDI

### Резюме

Доказывается следующая

**Теорема 3.** Пусть  $\xi_1, \xi_2, \dots, \xi_n, \dots$  — последовательность независимых одинаково распределенных случайных величин, с математическим ожиданием 0 и дисперсией 1.

Если  $v_n (n = 1, 2, \dots)$  — последовательность положительных целочисленных случайных величин, таких что  $\frac{v_n}{n}$  стремится по вероятности к некоторой положительной случайной величине  $\lambda$ , то мы имеем:

$$\lim_{n \rightarrow +\infty} \mathbf{P} \left( \frac{\xi_1 + \dots + \xi_{v_n}}{\sqrt{v_n}} < x \right) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{y^2}{2}} dy.$$

Теорема 3 является обобщением результатов ANSCOMBE-а [1], RÉNYI [2], [3] и автора [6]. Именно в работе ANSCOMBE-а и автора предполагается, что  $\frac{v_n}{n}$  стремится по вероятности к некоторой положительной константе, а в работах RÉNYI к некоторой положительной случайной величине  $\lambda$ , имеющей дискретное распределение.

Теорема 9, является легким обобщением Теоремы 3. Метод доказательства Теоремы 3 переносится почти тривиальным образом на доказательство Теоремы 9.

Теорема 3 доказывается с помощью Теоремы 7, которая основывается на следующей теореме:

**Теорема 6.** Пусть  $\vartheta_1, \vartheta_2, \dots$  последовательность независимых случайных величин с математическим ожиданием 0 и дисперсией  $D_1, D_2, \dots$ , таких, что функция распределения  $F_n(x)$  случайной величины

$$\sigma_n = \frac{\vartheta_1 + \vartheta_2 + \dots + \vartheta_n}{B_n} \quad (B_n \rightarrow +\infty)$$

стремится к некоторой несобственной функции распределения  $F(x)$ . Пусть

$$\lim_{n \rightarrow +\infty} \mathbf{M}(\sigma_n^2) = M, \quad (|M| < +\infty, M = \int_{-\infty}^{+\infty} x^2 dF(x))$$

то мы имеем для условных математических ожиданий

$$\lim_{n \rightarrow +\infty} \mathbf{M}(\sigma_n^2 | A) = M,$$

где  $A$  — произвольное случайное событие, имеющее положительную вероятность.

**Теорема 7.** Пусть  $\vartheta_1, \vartheta_2, \dots, \vartheta_n, \dots$  — последовательность независимых случайных величин. Предполагается, что математическое ожидание  $\mathbf{M}(\vartheta_i) = M_i$  и дисперсия  $\mathbf{D}(\vartheta_i) = D_i$  ( $i = 1, 2, \dots$ ) существуют для всех  $i$ . Возьмем

$$\sigma_n = \frac{\vartheta_1 + \dots + \vartheta_n - (M_1 + \dots + M_n)}{S_n}$$

где

$$S_n = \sqrt{D_1^2 + \dots + D_n^2}.$$

Предположим, что функция распределения  $F_n(x)$  случайной величины  $\sigma_n$  стремится к некоторой несобственной функции распределения. Если  $C$  — случайное событие, имеющее положительную вероятность, то существует целое число  $n_0 = n_0(C)$ , такое что для чисел  $n$ , больших  $n_0$  мы имеем:

$$\mathbf{P} \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k (\vartheta_i - M_i) \right| \geq \lambda S_n, C \right) \leq \frac{3\sqrt{P(C)}}{\lambda^2}$$

где  $\lambda$  произвольное положительное число.