Z. Wahrscheinlichkeitstheorie verw. Gebiete 51, 27-38 (1980)

Zeitschrift für Wahrscheinlichkeitstheorie und verwandte Gebiete © by Springer-Verlag 1980

# Local Theorems for Euclidean Motions. I

V.M. Maximov

Institute of Chemical Physics, Acad. of Sciences of the USSR, Vorobyevskoe Chaussee 2<sup>6</sup>, Moscow 117334, USSR

### 1. Introduction

Let **G** be some topological group, v a probability measure on **G**,  $v^{*n}$  the *n*-fold convolution of v. The value  $v^{*n}(\mathcal{D})$  is interesting for some applications where  $\mathcal{D}$ is any bounded domain in **G**. Every estimate of the value  $v^{*n}(\mathcal{D})$ ,  $n \to \infty$  will be called a local theorem (l.t.) on the group **G**. Accordingly the estimate of the value  $v^{*n}(\mathcal{D}_1)/v^{*n}(\mathcal{D}_2)$  for any bounded domains  $\mathcal{D}_1$  and  $\mathcal{D}_2$  is called a relation local theorem (r.l.t.). Local theorems have a short history. They have been proved in the case of additive groups  $\mathbb{R}^1$  and  $\mathbb{R}^d$  by Shepp [11] and Stone [12] respectively in 1965. Kazdan [5] was the first who began to study r.l.t. for the group of Euclidean plane motions. This result refers to a particular case of a measure v whose support contains four motions  $\mathcal{A}$ ,  $\mathcal{A}^{-1}$ ,  $\mathcal{B}$ ,  $\mathcal{B}^{-1}$ . This result was obtained by functional methods and didn't possess a probability character. A r.l.t. has also been proved for local compact unimodal groups by Le Pege [6] and for a wide class of commutative groups by Stone [13].

The first l.t. for the group of Euclidean plane motions was reported in the 3rd Soviet Japanese symposium on probability in 1975 [10]. Independently P. Baldi, Ph. Bougerol, P. Crépel obtained analogous results under wider conditions for Euclidean motions of the *d*-dimensional space  $\mathbb{R}^d$  using the Plancherel formula for the Euclidean group of motions [1, 2].

By  $\mathbb{M}(d)$  we denote the group of Euclidean motions  $\mathbb{R}^d$ .

Some details from [10] were given by the author in [8] and [9], especially with reference to the group  $\mathbb{M}(2)$ . In particular the concept of "lattice motions" of  $\mathbb{M}(2)$  in [9] together with l.t. [1] or with the theorem of the present work give l.t. for any case of measures v on  $\mathbb{M}(2)$  with finite support. It is clear that Kazdan's theorem follows from the general cases of l.t. with a finite support of the measure v. The purpose of the present work is to obtain the same [1] by the direct probability method. This method uses a variant of Stone's l.t. [12] applied to independent and nonidentically distributed values and it is sensitive to complementary conditions. In particular, the method can give some other variants of l.t. that do not follow from [1]. They will be obtained in the second part of this work.

### 2. The Formulation of the Main Result

The group of Euclidean motions  $\mathbb{M}(d)$  is the cross-group of the additive group  $\mathbb{R}^d$  with the rotations group  $\mathbb{SO}(d)$ . Every element of this group is uniquely determined by the pair  $(x, \alpha)$ , where x is the shift and  $\alpha$  is the rotation. The product of the motions is defined by  $(x, \alpha)(y, \beta) = (x + y\alpha, \alpha\beta)$ , where  $\alpha\beta$  is the product of the rotations and  $\mathscr{Y}\alpha$  is the image  $\mathscr{Y}\in\mathbb{R}^d$  under  $\alpha\in\mathbb{SO}(d)$ . Let v be a measure defined on  $\mathbb{M}(d)$ . Taking its natural desintegration we obtain for every  $\alpha \in \mathbb{SO}(d)$  the measure  $v(\alpha)$  on  $\mathbb{R}^d$  and the measure v on  $\mathbb{SO}(d)$  which is the conditional distribution v. The random elements  $\xi_1, \xi_2, ..., \xi_n, ...$  in  $\mathbb{M}(d)$  correspond to the random independent pairs  $(x_1, \alpha_1), \ldots, (x_n, \alpha_n), \ldots$  For the pairs  $(x_k, \alpha_k), k = 1, 2, ...$  the random values  $x_k$  and  $\alpha_k$  may be dependent. Then the product  $\xi_1 \xi_2 \dots \xi_n$  corresponds to  $(x_1 + x_1 \alpha_1 + \dots + x_n \alpha_1 \dots \alpha_{n-1}, \alpha_1 \dots \alpha_n)$ . Consequently, the product  $\xi_1 \xi_2 \dots \xi_n$  may be reduced to the study of the joint distribution of  $\mathscr{G}_n = x_1 + x_2 \alpha_1 + \ldots + x_n \alpha_1 \ldots \alpha_{n-1} \in \mathbb{R}^d$  and  $\alpha_1 \alpha_2 \ldots \alpha_n \in \mathbb{SO}(d)$ . Under general conditions the product  $\alpha_1 \alpha_2 \dots \alpha_n$  converges to the uniform distribution on SO(d) and becomes asymptotically independent with  $\mathcal{G}_n$ . Consequently the l.t. in  $\mathbb{M}(d)$  is equivalent to the estimate of the probability  $\mathbb{P}\{\mathscr{G}_n \in \mathscr{D}\}$ . Therefore we will estimate this probability. Let  $\xi$  be arbitrary  $\mathbb{M}(d)$  - valued random variable having the distribution v determined by the pair  $(x, \alpha)$ . By  $z(\alpha)$ we denote the conditional variable with fixed  $\alpha$ , v - the distribution  $\alpha$ ,  $\mathscr{F}_{v}$  - the distribution x,  $v(\alpha)$  – the distribution  $x(\alpha)$  and  $v'(\alpha)$  – the distribution obtained by symmetrization of  $v(\alpha)$ .

**Theorem** (l.t.) Let  $\xi_1, \xi_2, ..., \xi_n, ...$  be independent  $\mathbb{M}(d)$  – valued random variables with identical distribution v and satisfying the conditions:

1. the n-fold convolution  $v^{*n}$  converges weakly as  $n \to \infty$  to the uniform distribution on SO(d).

2. The distribution  $\mathscr{F}_{v}$  on  $\mathbb{R}^{d}$  has finite second moments.

3. There exists some number  $n_0 > 0$  and a set K,  $K \subseteq SO(d)$  of measure  $\mathbf{v}(K) > 0$  so that for any  $\alpha \in K$  the support of distribution  $v^{*n_0}(\alpha)$  does not contain any hyperplanes. In what follows for the sake simplicity we shall suppose  $n_0 = 1$ .

Then for any bounded Jordan measurable domain  $\mathcal{D} \in \mathbb{R}^{d}$ ,

$$\mathbf{P}\left\{\mathscr{G}_{\mathbf{n}}\in\mathscr{D}\right\} = c\,\operatorname{mes}(\mathscr{D})\,n^{-d/2} + O(n^{-d/2}) \tag{1}$$

where  $mes(\mathcal{D})$  is the Lebesgue measure of the domain  $\mathcal{D}$ , c is a constant which does not depend on  $\mathcal{D}$  and will be evaluated at the end of this paper.

It is easy to see that the formulations of the proposition [1] and our theorem coincide.

The main tool of the direct probability method for studying local properties of the sum  $\mathscr{S}_n$  is the reduction to the conditional sums  $\mathscr{S}_n(\alpha_1, ..., \alpha_n)$  of independent but not identically distributed random variables where

$$\mathscr{G}_{n}(\alpha_{1},\ldots,\alpha_{n}) = \varepsilon(\alpha_{1}) + \varepsilon(\alpha_{1}) \alpha_{1} + \ldots + \varepsilon(\alpha_{n}) \alpha_{1} \ldots \alpha_{n-1}$$
<sup>(2)</sup>

for fixed collections  $(\alpha_1, \alpha_2, ..., \alpha_n)$ .

For fixed  $(\alpha_1, \alpha_2, ..., \alpha_n)$  the conditional variables are independent but not identically distributed. If for the "majority" of these sums the l.t. in  $\mathbb{R}^d$  may be

applied, i.e. the value  $\mathbf{P}\{\mathscr{G}_n(\alpha_1, ..., \alpha_n) \in \mathscr{D}\}\$  may be effectively estimated, then we have by integration

$$\mathbf{P}\left\{\mathscr{G}_{n}\in\mathscr{D}\right\} = \int \mathbf{P}\left\{\mathscr{G}_{n}(\alpha_{1},\ldots,\alpha_{n})\in\mathscr{D}\right\}\mathbf{v}(d\alpha_{1})\ldots\mathbf{v}(d\alpha_{n}).$$
(3)

In spite of the essential difference of the methods that have been applied here and in [1] they use effectively the same condition 3. For example whenever this condition or analogous ones are absent the l.t. has not yet been proved, if  $d \ge 3$  and the measure v has a finite support without additional conditions.

Finally we remark that by virtue of condition 2 we get the following condition 2': the variables  $z(\alpha)$ ,  $\alpha \in \mathbb{SO}(d)$  have second moments which are absolutely integrable. The expectation  $\mathbf{M} z(\alpha) = m(\alpha)$  has second moments also. We denote the characteristic function of  $z(\alpha)$  by  $f_{\alpha}(t)$ .

## **3.** A Special Formulation of the l.t. in $\mathbb{R}^d$

To apply the l.t. to the sum (2) we need the following variant of Stone's l.t.

**Lemma 1.** Let  $\xi_1(\tau)$ ,  $\xi_2(\tau), \ldots, \xi_n(\tau), \ldots$  be a sequence of  $\mathbb{R}^d$ -valued nonidentically distributed variables depending on the parameter  $\tau$  on a measurable compact  $\mathbb{T}$  with probability measure  $\mu$ . For any  $\tau \in \mathbb{T}$  the variables  $\xi_k(\tau)$  are independent and possess second moments with respect to the measure  $\mu$ . By  $f_n(t, \tau_n)$  we denote the characteristic function of the sum  $\xi_1(\tau_n) + \ldots + \xi_n(\tau_n)$ . If for all n, n > 0 there exists a set  $\mathscr{H}_n \subseteq \mathbb{T}, \mu(\mathscr{H}_n) \to 1$  and we have uniformly for any sequence  $\{\tau_n\}, \tau_n \in \mathscr{H}_n$ :

1) the distribution of the variable  $\frac{1}{\sqrt{n}} \{\xi_1(\tau_n) + \ldots + \xi_n(\tau_n) - m(n, \tau)\}$  converges

weakly as  $n \rightarrow \infty$  to the normal distribution with mean zero and covariance matrix  $\Omega$  where

$$m(n,\tau_n) = M \{ \xi_1(\tau_n) + \ldots + \xi_n(\tau_n) \},\$$

2) the integral  $\sqrt{n} \int_{b}^{B} |f_{n}(t,\tau_{n})| dt$  tends to zero as  $n \to \infty$  for any positive numbers b, B,

3) the integral  $\sqrt{n} \int_{\frac{A}{\sqrt{n}} \leq ||t|| \leq b} |f_n(t, \tau_n)| dt$  tends to zero as  $n \to \infty, A \to \infty, b \to 0$ ,

then for any bounded Jordan measurable domain  $\mathcal{D}$  and for all sequences  $\{\tau_n\}, \tau_n \in \mathscr{H}_n$  we have

$$\mathbf{P}\left\{\xi_{1}(\tau_{n})+\ldots+\xi_{n}(\tau_{n})\in\mathscr{D}\right\}=(\sqrt{n})^{-d}\int_{\mathscr{D}}\mathbf{N}_{\Omega}\left(\frac{\mathscr{Y}}{\sqrt{n}}-\frac{m(n,\tau_{n})}{\sqrt{n}}\right)d\mathscr{Y}$$
$$+o_{n}(A,b,\tau_{n})(\sqrt{n})^{-d},$$
(4)

where  $\mathbf{N}_{\Omega}(x)$  is the density of the normal distribution with mean zero and covariance matrix  $\Omega$ , and the value  $o_n(A, b, \tau_n)$  tends to zero uniformly as  $n \to \infty$ ,  $A \to \infty$ ,  $b \to 0$ .

The scheme of Stone's proof [12] remains valid when applied to the proof of Lemma 1. Since we intend to apply Lemma 1 to the sum  $\mathscr{G}_n(\alpha_1, ..., \alpha_n)$  it is necessary to represent the terms of the sum in accordance with Lemma 1. In order to do this we take the compact space  $\mathbb{T} = \mathbb{SO}(d)^{\infty}$  with measure  $\mu$ , where  $\mathbb{SO}(d)^{\infty}$  is equal to a countable product of the group  $\mathbb{SO}(d)$  and the measure  $\mu$  is the direct product of the measure  $\mathbf{v}$  on  $\mathbb{SO}(d)$ . For any  $\tau = (\alpha_1, \alpha_2, ...) \in \mathbb{T}$  we construct the sequence of  $\mathbb{R}^d$ -valued variables

$$\xi_1(\tau) = z(\alpha_1), \qquad \xi_2(\tau) = z(\alpha_2)\alpha_1, \dots, \xi_n(\tau) = z(\alpha_n)\alpha_1 \dots \alpha_{n-1}$$
(5)

which are independent and nonidentically distributed.

Then for any  $\tau \in \mathbf{T}$  we can define the random variables

$$\mathscr{G}_n(\tau) = \mathscr{G}_n(\alpha_1, \dots, \alpha_n) = \xi_1(\tau) + \dots + \xi_n(\tau)$$

We shall apply the equality (4) to the sum  $\mathscr{G}_n(\alpha_1, ..., \alpha_n)$ . For this it is enough to verify that the conditions 1), 2), 3) are valid for  $\mathscr{G}_n(\alpha_1, ..., \alpha_n)$ .

# 4. The Normal Convergence of the Sequence $\frac{1}{\sqrt{n}} \{\mathscr{S}_n(\tau_n) - \mathbf{M}(\mathscr{S}_n(\tau_n))\}$

Condition 1 will hold for  $\mathscr{G}_n(\alpha_1, ..., \alpha_n)$  if the distribution of the sequence  $\frac{1}{\sqrt{n}}$  $\{\mathscr{G}_n(\tau_n) - \mathbf{M}(\mathscr{G}_n(\tau_n))\}, \tau_n \in \mathscr{K}_n$  converges to the normal distribution with density  $\mathbf{N}_{\Omega_1}(x)$  where  $\mathscr{K}_n$  are some subsets of  $\mathbb{SO}(d)^{\infty}, \mu(\mathscr{K}_n) \to 1$ . We need the following auxilliary proposition:

**Lemma 2.** Let  $\alpha_1, \alpha_2, \ldots$  be independent  $\mathbb{SO}(d)$ -valued variables with the same distribution **v**. Let  $f(\alpha)$  be the linear sum of elements taken from the matrix of nontrivial irreducible representation of  $\mathbb{SO}(d)$ ,  $a(\alpha)$  a real function on  $\mathbb{SO}(d)$  with  $\int |a(\alpha)| d\mathbf{v} \leq \infty$ . Then

$$\frac{1}{n}\sum_{1}^{n}a(\alpha_{k})f(\alpha_{1}\ldots\alpha_{k-1})$$
(6)

tends to zero weakly.

*Proof.* At first we suppose that  $a(\alpha)$  has second moments and we shall prove that

$$\frac{1}{n^2} \mathbf{M} \left\{ \sum_{1}^{n} a(\alpha_k) f(\alpha_1 \dots \alpha_{k-1}) \right\}^2 \to 0.$$
(7)

The sum (7) is equal to

$$\frac{1}{n^2} \sum_{1}^{n} \mathbf{M} \alpha^2(\alpha_k) \mathbf{M} f^2(\alpha_1 \dots \alpha_{k-1}) \\ + \frac{1}{n^2} \sum_{1 \le k < m \le n} \mathbf{M} \alpha(\alpha_k) a(\alpha_m) f(\alpha_1 \dots \alpha_{k-1}) f(\alpha_1 \dots \alpha_{k-1} \alpha_k \dots \alpha_{m-1}).$$

Since the function  $f(\alpha)$  is continuous,  $|f(\alpha)| < c_1$  on  $\mathbb{SO}(d)$ . Let us set  $\mathbf{M} a^2(\alpha) = c_2$ ,  $\mathbf{M} |a(\alpha)| = c_3$ . Under condition 1) for the measure **v** it was proved in [7] that  $|\mathbf{M} f(\alpha \alpha_1 \dots \alpha_k)| < c_4 \lambda^k$  where  $c_4$ ,  $\lambda$  are constants not depending on  $\alpha$ , and in addition  $0 < \lambda < 1$ . Therefore the first sum remains less than  $\frac{1}{n} c_2 c_1^2$  and the second one is less than  $\frac{1}{n^2} \sum_{1 \le k < m \le n} c_3^2 c_1 c_4 \lambda^{m-k}$ . Consequently (7) is true, and  $\frac{1}{n} \sum a(\alpha_k) f(\alpha_1 \dots \alpha_{k-1})$  tends to zero weakly if  $\mathbf{M} a^2(\alpha) < \infty$ . If  $\mathbf{M} a^2(\alpha) = \infty$  we can write  $a(\alpha) = a_N(\alpha) + \varepsilon_N(\alpha)$  where  $|a_N(\alpha)| < N$  and  $\mathbf{M} |\xi_N| < \delta$ . Then the sum (6) is equal to

$$\frac{1}{n}\sum_{1}^{n}a_{N}(\alpha_{k})f(\alpha_{1}\ldots\alpha_{k-1})+\frac{1}{n}\sum_{1}^{n}\varepsilon_{N}(\alpha_{k})f(\alpha_{1}\ldots\alpha_{k-1}).$$

It is proved above that the first sum tends to zero and the second sum is less than  $c_1 \frac{1}{n} \sum_{i=1}^{n} |\varepsilon_N(\alpha_k)|$ . But  $c_1 \frac{1}{n} \sum_{i=1}^{n} |\varepsilon_N(\alpha_k)|$  tends to  $c_1 \mathbf{M} |\varepsilon_N|$  which is less than  $c_1 \delta$ . Since  $\delta$  may be chosen arbitrarily we get the required result.

**Lemma 3.** Under the conditions of Lemma 2 the value  $\frac{1}{n} \sum_{1}^{n} a(\alpha_k) f^2(\alpha_1 \dots \alpha_{k-1})$  converges weakly to  $\mathbf{M} a(\alpha) \int f^2(\alpha) d\alpha$  where the integral  $\int f^2(\alpha) d\alpha$  is taken with respect to the invariant measure on  $\mathbf{SO}(d)$ .

*Proof.* The function  $f^2(\alpha)$  is an element of the tensor square of the corresponding representation. Therefore

$$f^{2}(\alpha) = \gamma_{0} + \sum_{1}^{d} \gamma_{k} f_{k}(\alpha)$$
(8)

where  $\gamma_1$  is constant,  $f_k(\alpha)$  are elements of the some irreducible nontrivial representations  $\mathbb{SO}(d)$  and  $\int f^2(\alpha) d\alpha = \gamma_0$  because  $\int f_k(\alpha) d\alpha = 0$ . Consequently applying Lemma 2 to the sum (8) we obtain Lemma 3.

By  $(x_1(\alpha), ..., x_n(\alpha))$  we denote the coordinates of the random vector  $x(\alpha)$  with mean zero, and by  $g(w) = ||g_{ij}(w)||$  we denote the orthogonal matrix of the rotation  $w \in \mathbb{SO}(d)$ . The map  $w \to g(w)$  is continuous and  $g(w, w_2) = g(w_1) g(w_2)$ , g(w) is an irreducible representation of  $\mathbb{SO}(d)$ . Let  $\mathbf{M}(\alpha)$  be the covariance matrix of  $x(\alpha)$  and  $x(\alpha)w$  be the image of  $x(\alpha) \in \mathbb{R}^d$  under  $w \in \mathbb{SO}(d)$ . Then the covariance of  $x(\alpha)w$  is equal to  $g(w)^T M(\alpha) g(w)$ , where  $g(w)^T$  is the transposed matrix. However it follows that the covariance matrix for the random variable  $\frac{1}{\sqrt{n}} \{\mathscr{S}_n - \mathbf{M} \mathscr{S}_n(\tau)\}$ , where  $\tau = (\alpha_1, \alpha_2, ...)$  is equal to

$$\frac{1}{n}\sum_{k=1}^{n} g^{\mathbf{T}}(\alpha_1 \dots \alpha_{k-1}) \mathbf{M}(\alpha_k) g(\alpha_1 \dots \alpha_{k-1}) = \Omega(n; \alpha_1, \dots, \alpha_n).$$
(9)

By virtue of condition 2' the expectation  $\mathbf{M} \{ \mathbf{M}(\alpha) \} = \|\sigma_{ij}\|$  exists. Set  $\sigma = \frac{1}{d} (\sigma_{11} + \ldots + \sigma_{dd})$ . It follows from condition 1, that  $\sigma > 0$ .

**Lemma 4.** The value  $\Omega(n; \alpha_1, ..., \alpha_n)$  converges weakly to  $\sigma II$  where I is the identical matrix.

*Proof.* The coordinates of the vector  $z(\alpha)$  w are equal to

$$z_j(\alpha, w) = z_1(\alpha) g_{j1}(w) + \ldots + z_d(\alpha) g_{jd}(w), \quad j = 1, 2, \ldots, d.$$

Then the (i, j)-th element of the matrix  $\Omega$  is equal to

$$\frac{1}{n}\sum_{1}^{n}\left\{\sum_{m,r}M\varkappa_{m}(\alpha_{k})\varkappa_{r}(\alpha_{k})g_{mi}(\alpha_{1}\ldots\alpha_{k-1})g_{rj}(\alpha_{1}\ldots\alpha_{k-1})\right\}.$$
(10)

The functions  $g_{ij}(w)$  are the elements of an irreducible non-trivial unitary representation for which the following ortogonal relations are well-known:

$$\int g_{ij}^2(w) \, dw = \frac{1}{d}; \qquad \int g_{mi}(w) \, g_{rj}(w) = 0 \quad \text{if } (m, i) \neq (r, j).$$

The functions  $g_{ij}(w)g_{rj}(w)$  are the elements of the tensor square representation of g(w). Therefore to every sum of the kind

$$\frac{1}{n}\sum_{k=1}^{n}M z_{m}(\alpha_{k}) z_{r}(\alpha_{k}) g_{mi}(\alpha_{1} \ldots \alpha_{k-1}) g_{rj}(\alpha_{1} \ldots \alpha_{k-1})$$

Lemmas 2 or 3 may be applied. This proves Lemma 4.

Now we will indicate the set  $\mathscr{K}_n \subset \mathbb{SO}(d)^{\infty}$ ,  $\mu(\mathscr{K}_n) \to 1$ , for which the distribution of the variable  $\frac{1}{\sqrt{n}} \{\mathscr{S}_n(\tau_n) - \mathbf{M}\mathscr{S}_n(\tau_n)\}, \tau_n \in \mathscr{K}_n$  converges weakly and uniformly to the normal distribution  $\mathbf{N}_{\Omega_1}(\infty), \Omega_1 = \sigma \mathbf{I}$ .

At first we set  $\varphi_{\lambda}(x) = \sup_{\substack{w, x_0 \ \|x - x_0\| > 2}} \int_{\|x w\|^2} ||x w||^2 dv(\alpha)$  where  $\|x\| = \sqrt{x_1^2 + \ldots + x_d^2}$ . Evidently the function is measurable and  $\varphi_{\lambda}(\alpha) \leq \int ||x||^2 dv(\alpha)$ .

Since the right hand side of the inequality is integrable in v, we have for all  $\lambda > 0$ ,  $\int \varphi_{\lambda}(\alpha) d\mathbf{v} = \varphi(\lambda) < c$  where the constant c does not depend on  $\lambda$ . For any  $\alpha$  the function  $\varphi_{\lambda}(\alpha)$  decreases to zero as  $\lambda \to \infty$ . Consequently  $\varphi(\lambda) \to 0$  as  $\lambda \to \infty$ , and there exists a sequence  $\lambda_{N}$ ,  $\varphi(\lambda_{N}) = \varepsilon_{N} \downarrow 0$ . Since the sum  $\frac{1}{n} \sum_{1}^{n} \varphi_{\lambda}(\alpha) \to \varphi(\lambda)$  in  $\mu$  measure, we may define the sets  $\mathscr{K}_{N}$ ,  $\mathscr{K}_{N} \subset \mathscr{K}_{N+1} \subset SO(d)^{\infty}$ ,  $\mu(\mathscr{K}_{N}) \uparrow 1$  such that for any  $(\alpha_{1}, \alpha_{2}, \ldots) \in \mathscr{K}_{N}$  the following inequality holds:  $\limsup \frac{1}{n} \sum_{k=1}^{n} \varphi_{\lambda_{n}}(\alpha_{k}) \leq \varepsilon_{N-1}$ . In virtue of the decreasing  $\varphi_{\lambda}(\alpha) = \varepsilon_{N+m}$  for any  $\alpha$  and  $\tau = (\alpha_{1}, \alpha_{2}, \ldots) \in \mathscr{K}_{N}$  we have  $\limsup \frac{1}{n} \sum_{k=1}^{n} \varphi_{\ln n}(\alpha_{k}) < \varepsilon_{N+m}$  for any integer N, m > 0 or  $\limsup \frac{1}{n} \sum_{k=1}^{n} \varphi_{\ln n}(\alpha_{k}) = 0$ . This implies that

$$\lim \frac{1}{n} \sum_{1}^{n} \int_{\|x-m(\alpha_{k})\alpha_{1}\dots\alpha_{k-1}\| > \varepsilon \sqrt{n}} \|x\|^{2} d\nu(\alpha_{k})\alpha_{1}\dots\alpha_{k-1} \leq \lim \sup \frac{1}{n} \sum_{1}^{n} \varphi_{\ln n}(\alpha_{k}) = 0.$$
(11)

The last inequality holds for arbitrary but fixed  $\mathscr{K}_N$ , any  $\varepsilon > 0$  and from some positive integer  $N_{\varepsilon}$  on.

The right-hand side of the inequality (11) is a necessary and sufficient condition for the convergence in the weak sense of  $\frac{1}{\sqrt{n}} \{\mathscr{S}_n(\tau_n) - \mathbf{M}\mathscr{S}_n(\tau_n)\}$  to the normal distribution under  $\tau_n \in \mathscr{K}_N$ .

On the other hand the covariance matrix from Lemma 4  $\frac{1}{\sqrt{n}} \{\mathscr{S}_n(\tau_n) - \mathbf{M}\mathscr{S}_n(\tau_n)\}$  converges to  $\Omega_1 = \sigma \mathbf{II}$ . Since the normal distribution with zero is determined by the covariance matrix uniquely and  $\mu(\mathscr{K}_N) \to 1$  as  $N \to \infty$ , condition 1) of Lemma 1 for the variables (5) is satisfied.

*Remark.* It is well-known by [4] that the distribution  $\frac{1}{\sqrt{n}} \mathscr{S}_n$  converges weakly to the normal one with density  $N_{\Omega}(x)$ . The matrix  $\Omega$  was defined in [4], too. On other hand we have

$$\mathscr{S}_{n} = \frac{1}{\sqrt{n}} \{\mathscr{S}_{n}(\tau) - \mathbf{M}\mathscr{S}_{n}(\tau)\} + \frac{1}{\sqrt{n}} \mathbf{M}\mathscr{S}_{n}(\tau)$$

for any collection  $(\alpha_1, \alpha_2, ...) = \tau$ .

It was shown above that the distribution  $\frac{1}{\sqrt{n}} \{\mathscr{S}_n(\tau_n) - \mathbf{M}\mathscr{S}_n(\tau_n)\}, \tau_n \in \mathscr{K}_n$ converges to the normal one with density  $\mathbf{N}_{\Omega_1}(x)$ . The distribution of the variable  $\frac{1}{\sqrt{n}} \mathbf{M}\mathscr{S}_n(\tau) = \frac{1}{\sqrt{n}} \sum \mathfrak{M}(\alpha_k) \alpha_1 \dots \alpha_{k-1}$  converges to the normal distribution with the density  $\mathbf{N}_{\Omega_2}(x)$  [4] since by condition 2'. the value of  $\int ||\mathfrak{M}(\alpha)||^2 d\mathbf{v}$  is bounded, and the random variables  $\alpha_1, \alpha_2, \dots$  are independent with common distribution  $\mathbf{v}$ , satisfying condition 1. Consequently  $\mathbf{N}_{\Omega}(x) = \mathbf{N}_{\Omega_1 + \Omega_2}(x)$ . This equality will be used to calculate the constant c in the theorem.

5. Proof of Conditions 2), 3) for the Sum  $\mathscr{S}_r(\alpha_1, \ldots, \alpha_n)$ 

To check conditions 2), 3) for the sum  $\mathcal{G}_n(\alpha_1, ..., \alpha_n)$  we need the following proposition.

**Lemma 5.** Let  $f_{\alpha}(t)$  be the characteristic function of the variable  $\alpha(\alpha)$ . Put  $w_k = (\alpha_1 \dots \alpha_k)^{-1}$ ,  $w_0$  – the identity of  $\mathbb{SO}(d)$ . Then for any segment [b, B], 0 < b < B there exists  $\theta$ ,  $0 < \theta < 1$  such that the inequality

$$\mathbf{M}\left\{\prod_{k=1}^{n}|f_{\alpha_{k}}(tw_{k-1})|\right\} < \theta^{n}$$

holds uniformly for any  $t \in [b, B]$  as  $n \to \infty$ .

*Proof.* For  $\alpha \in K$  (see condition 3.) the support of the distribution  $v(\alpha)$  does not contain any hyperplane. Thus, if  $\alpha \in K$ , then the function  $|f_{\alpha}(t)|$  may be equal to  $\mathbb{1}$ 

only for the points laying on some lines  $t_{\alpha}$  passing through the origin  $\mathbb{O}$ . Let  $\beta$  be a random variable on  $\mathbb{SO}(d)$  with distribution **v**. The convergence of  $\mathbf{v}^{*n}$  to the uniform distribution on  $\mathbb{SO}(d)$  involves the convergence of  $(\mathbf{v}w)^{*n}$  to one under fixed  $w \in \mathbb{SO}(d)$ . Then  $\mathbf{M}_{\beta} | f_{\alpha}(t \beta w) | < 1$ , since the distribution  $v(\alpha)$  must be degenerate in the contrary case. Thus  $\mathbf{M}_{\beta} | f_{\alpha}(t \beta w) | = v(t, \alpha, w) < 1$  for  $t \neq 0$ ,  $\alpha \in K$ ,  $v(t, \alpha, w)$  is a continuous function and  $\sup_{w} v(t, \alpha, w) = v(t, \alpha) < 1$ ,  $\mathbf{M}_{\alpha} v(t, \alpha) = v(t) < 1$ 

for  $t \neq 0, \alpha \in K$ .

Hence we can find a  $\theta_1$ ,  $0 < \theta_1 < 1$  such that  $v(t) < \theta_1$  for all  $t \in [b, B]$ . Since  $|f_{\alpha}(t)| \leq 1$  we have the following inequalities:

$$\mathbf{M}\left\{\prod_{k=1}^{n}|f_{\alpha_{k}}(t\,w_{k-1})|\right\} \leq \mathbf{M}\left\{\prod_{k=1}^{\left[n/2\right]}|f_{\alpha_{2k}}(t\,w_{2k-1})|\right\}$$
(12)

and

$$\mathbf{M}_{\alpha_{2k}}(\mathbf{M}_{2k-1} | f_{\alpha_{2k}}(t \, \alpha_{2k-1}^{-1} \cdot \alpha_{2k-2}^{-1} \dots \alpha_{1}^{-1}) | \\
= \mathbf{M}_{\alpha_{2k}} v(t, \alpha_{2k}, w_{2k-2}) \leq \mathbf{M}_{\alpha_{2k}} v(t, \alpha_{2k}) = v(t) < \theta_{1}.$$
(13)

We remark that the random pairs  $(\alpha_{2k}, \alpha_{2k-1})$  and  $(\alpha_{2m}, \alpha_{2m-1})$  are independent for any integers k, m,  $k \neq m$ . Thus it follows from (13) and  $\mathbf{M}_{\alpha}\mathbf{M}_{\beta}|f_{\alpha}(t\beta w| < \theta_{1})$  that the right of (12) can be estimated recursively, i.e.

$$\mathbf{M}\left\{\prod_{k=1}^{[n/2]} |f_{\alpha_{2k}}(t \, w_{2k-1})|\right\} \leq \mathbf{M}\left\{\prod_{k=1}^{[n/2]-1} |f_{\alpha_{2k}}(t \, w_{2k-1})|\right\} \theta_{1} \leq \ldots \leq \theta_{1}^{[n/2]}.$$

This completes the proof.

To verify that condition 2) is fulfilled for the sum  $\mathscr{G}_n(\alpha_1, \ldots, \alpha_n)$  it suffices to show that

$$\mathbf{M}\left\{\sqrt{n}\int_{b\leq ||t||\leq B}\prod |f_{\alpha_k}(tw_{k-1})|\,dt\right\}\to 0.$$

But by Lemma 5 this value is less than  $\theta^n \sqrt{n(B-b)}$ ,  $0 < \theta < 1$  and consequently it tends to zero. To verify that condition 3) is fulfilled it is enough to show that

$$\sqrt{n} \mathbf{M} \left\{ \int_{\frac{A}{n} \leq ||t|| \leq b} \prod |f_{\alpha_k}(t w_{k-1})| dt \right\}$$

tends to zero as  $n, A \to \infty$ ,  $b \to 0$ . For this we put  $\mathbf{M}|f_{\alpha}(t)|=u(t)$ ,  $u(tw_{i}) = \max_{w} u(tw)$ . We know that the variables  $\alpha_{k}$ , k=1, 2, ... are independent and distributed identically, hence we have

$$\mathbf{M}\prod_{k=1}^{n}|f_{\alpha_{k}}(t\,w_{k-1})| \leq \left\{\mathbf{M}\prod_{k=1}^{n-1}|f_{\alpha_{k}}(t\,w_{k-1})|\right\}u(t\,w_{t}) \leq u^{n}(t\,w_{t}).$$
(14)

It is regrettable that the function u(t) cannot be a characteristic function. However,  $\{\mathbf{M}|f_{\alpha}(t)|\}^2 = u(t)^2 \leq \mathbf{M}|f_{\alpha}(t)|^2 = \mathbf{U}(t)$  where  $\mathbf{U}(t)$  is the characteristic function of the distribution  $\int v'(\alpha) d\mathbf{v}$ . This distribution has finite second moments and by condition 3) does not contain any hyperplanes. Hence it follows that for some neighbourhood of zero we have the estimate

$$\mathbf{U}(t) < 1 - c \|t\|^2$$
.

Since ||t|| = ||tw|| for any rotation  $w \in \mathbb{SO}(d)$ , we also have

$$\mathbf{U}(t\,w_t) < 1 - c \,\|t\|^2. \tag{15}$$

Consequently it follows from (14) and (15) that

$$\sqrt{n} \mathbf{M}_{\frac{A}{\sqrt{n}} \leq \|t\| \leq b} \prod |f_{\alpha}(t w_{k-1})| dt \leq \sqrt{n} \int_{\frac{A}{\sqrt{n}} \leq \|t\| \leq b} \mathbf{U}^{[n/2]}(t) dt.$$
(16)

According to (15) standard tools can be applied to estimate the integral of the power of the characteristic function in the segment  $[A/\sqrt{n}, b]$ . In that case we obtain  $\sqrt{n} \int \mathbf{U}^{[n/2]}(t) dt \to 0$  as  $n, A \to \infty, b \to 0$ .

Consequently, this completes the proof of condition 3) for the sum  $\mathscr{S}_n(\alpha_1, \ldots, \alpha_n)$ .

### 6. Essen's Inequality for the Sum $\mathcal{S}_n(\alpha_1, \ldots, \alpha_n)$ .

Let  $\xi_1, \xi_2, ..., \xi_n, ...$  be a sequence of random variables which are independent and identically distributed in  $\mathbb{R}^d$ . In the paper by Esseen [3] the following inequality was obtained for the sum  $\mathscr{S}_n = \xi_1 + ... + \xi_n$ 

$$\sup_{x \in \mathbb{R}^d} \mathbf{P}\{\mathscr{S}_n \in \Sigma_r + x\} = c(d, r) \left\{ \sup_{u \ge r} u^{-2} \sum_{1}^n \chi_k(u) \right\}^{-d/2}$$
(17)

where  $\Sigma_r$  is the sphere of radius *r* with the centre at the origin; c(d, r) is a constant depending only on *d* and *r*,  $\mathbf{P}_k$  is the distribution of  $\xi_k$ ,  $\chi_k(u) = \inf_{\|t\|=1} \int_{\|x\|| < u} (t, x)^2 d\mathbf{P}'_k$  where the distribution  $\mathbf{P}'_k$  is obtained by symmetrization of  $\mathbf{P}_k$ . Let us apply the inequality (17) to the sum

$$\mathscr{S}_n(w; \alpha_1, \ldots, \alpha_n) = x_1(\alpha_1) w_1 + \ldots + x_n(\alpha_n) w_n$$

where the rotations  $w_1, \ldots, w_n$  are arbitrary but fixed. Since for any rotation  $w \in \mathbb{SO}(d)$  the equality  $(v(\alpha) w)' = v'(\alpha) w$  holds, we shall have

$$\inf_{\|t\| = 1} \int_{\|x\| \le \lambda_0} (t, x)^2 d(v(\alpha) \tau)' = \inf_{\|t\| = 1} \int_{\|x\| \le \lambda_0} (t \tau^{-1}, x)^2 dv'(\alpha)$$

$$= \inf_{\|t\| = 1} \int_{\|x\| \le \lambda_0} (t, x)^2 dv'(\alpha).$$
(18)

Thus Esseen's inequality for the sum  $\mathscr{L}_n(w; \alpha_1, ..., \alpha_n)$  does not depend on the arbitrary rotations  $w_1, ..., w_n \in \mathbb{SO}(d)$ . The function  $\psi(\alpha, \lambda)$  $= \inf_{\|t\|=1} \int_{\|x\| \le \lambda} (t, x)^2 dv'(\alpha)$  is measurable with respect to  $\alpha$  and is decreasing with respect to  $\lambda$  for every fixed  $\alpha$ . By condition 3. if  $\alpha \in K$  then  $\lim_{\lambda \to \infty} \inf \psi(\alpha, \lambda) > 0$ . Consequently we can find a subset  $K_0 \subset K$  with positive measure  $\mathbf{v}(K_0) = p_0 > 0$ and some  $\lambda_0$  such that simultaneously for all  $\alpha \in K_0$ ,  $\lambda \ge 0$  the inequality

$$\inf_{\|t\|=1} \int_{\|x\|<\lambda} (t,x)^2 \, dv'(\alpha) \ge \delta_0 > 0$$
(19)

is fulfilled.

**Lemma 6.** Let us suppose that among  $\alpha_1, \ldots, \alpha_n$  the value  $\alpha_k \in K_0$  is taken on  $p_0 n/2$  times at least. Then there exists a sphere  $\Sigma_r$  and a constant  $c(p_0, \delta, d, r)$  such that

$$\sup_{x\in\mathbb{R}^d} \mathbf{P}\{\mathscr{S}_n(\alpha_1,\ldots,\alpha_n)\in\Sigma_r+x\} < c(p_0,\,\delta,\,d,\,r)\,n^{-d/2}.$$

In fact this can be obtained by employing (18) and (19) in the inequality (17) for the sum  $\mathcal{G}_n(\alpha_1, \ldots, \alpha_n)$ .

The next lemma follows from the preceding one.

**Lemma 7.** Under the conditions of Lemma 6 for any bounded domain  $\mathcal{D}$  in  $\mathbb{R}^d$  there exists a constant  $c(\mathcal{D}, p_0, \delta, d)$  such that

$$\sup_{x \in \mathbb{R}^d} \mathbf{P}\{\mathscr{S}_n(\alpha_1, \dots, \alpha_n) \in \mathscr{D} + x\} \leq c(\cdot) n^{-d/2}.$$
(20)

Furthermore the measure of all trajectories  $(\alpha_1, ..., \alpha_n)$  for which the inequality (20) may not be true is less than  $\exp\{-c(p_0)n\}$  and the constant  $c(p_0) > 0$  depends only on  $p_0$ .

*Proof.* In fact any bounded domain  $\mathscr{D}$  of diameter L may be covered by the finite number of spheres of the radius r. This number depends only on L and r. Consequently we can apply Lemma 6. The second part of Lemma 7 follows by elementary combinatorial estimations.

#### 7. The Proof of the Theorem

It follows from above that we can define some sequence of sets  $\mathscr{K}_n \subset \mathbb{SO}(d)^{\infty}$ ,  $\mu(\mathscr{K}_n) \to 1$  as  $n \to \infty$  such that the conditions of Lemma 1 are fulfilled for the sum  $\mathscr{S}_n(\alpha_1, \ldots, \alpha_n)$  if  $\tau_n = (\alpha_1, \ldots, \alpha_n, \ldots) \in \mathscr{K}_n$ . Then for every number *n* we shall represent the compact  $\mathbb{SO}(d)^{\infty}$  as the sum of the sets  $\mathscr{K}_n$ ,  $\mathscr{L}_n$  and  $\mathscr{E}_n$  where  $\mathscr{L}_n$  is a set of  $\tau$  such that  $\tau = (\alpha_1, \ldots, \alpha_n, \ldots) \in \mathbb{SO}(d)^{\infty}$ , so that the value  $\alpha_k \in K_0$  is taken on less than  $p_0 n/2$  times among the *n* first components,  $\mathscr{E}_n$  is the rest of the elements  $\mathbb{SO}(d)$ . It is obvious that  $\mu(\mathscr{E}_n) \leq \varepsilon_n$ . By virtue of Lemma 7 we have  $\mu(\mathscr{L}_n) < \exp(-c(p_0)n)$ . It is clear from the foregoing definition of the measure  $\mu$  on  $\mathbb{SO}(d)^{\infty}$  that integral (3) can be written

$$\mathbf{P}\left\{\mathscr{S}_{n}\in\mathscr{D}\right\} = \int \mathbf{P}\left\{\mathscr{S}_{n}(\tau)\in\mathscr{D}\right\} d\mu = \int_{\mathscr{K}_{n}} + \int_{\mathscr{L}_{n}} + \int_{\mathscr{S}_{n}} + \int_{\mathscr{S}_{n}}.$$
(21)

By Lemma 6 and the evident inequality  $0 \leq \mathbf{P}\{\mathscr{G}_n(\tau) \in \mathscr{D}\} \leq 1$  we have

$$\int_{\mathscr{K}_n} = (\sqrt{n})^{-d} \int_{\mathscr{K}_n} (\int_{\mathscr{D}} \mathbf{N}_{\Omega_1}(\cdot) \, dy) \, d\mu + (\int \theta_n(\cdot) \, d\mu) \, n^{-d/2}.$$

Since  $\theta_n$  tends to zero uniformly on  $\mathscr{K}_n$  as  $n \to \infty$  the second integral is less than  $\varepsilon'_n n^{-d/2}$  where  $\varepsilon'_n \to 0$ .

It is obvious that

$$\int_{\mathcal{X}_n} = n^{-d/2} \mathbf{M}_{\tau}(\cdot) - n^{-d/2} \int_{\mathcal{X}_n + \mathscr{E}_n}, \quad \text{where } \int_{\mathcal{X}_n + \mathscr{E}_n} \mathbf{N}_{\Omega_1}(0) \operatorname{mes} \mathscr{D} \cdot \varepsilon_n'', \varepsilon_n'' \to 0.$$

To evaluate  $M_{\tau}(\cdot)$  it suffices to change the integration order and to note that  $\operatorname{mes}(\mathscr{D})$  and  $\mathbf{N}_{\Omega_1}(x)$  are bounded. Therefore

$$\mathbf{M}_{\mathfrak{r}}(\,\cdot\,) = \operatorname{mes} \mathscr{D} \cdot \left\{ \lim_{n} \mathbf{M}_{\mathfrak{r}} \left[ N_{\Omega_{1}} \left( \frac{\mathbf{M} \mathscr{L}_{n}(\tau)}{\sqrt{n}} \right) \right] \right\} + \varepsilon_{n}^{\prime\prime\prime}, \qquad \varepsilon_{n}^{\prime\prime\prime} \to 0$$

Thus we obtain

$$\mathbf{P}\{\mathscr{G}_{n}\in\mathscr{D}\} = \operatorname{mes}\,\mathscr{D}\{\lim_{n}\mathbf{M}_{\tau}[\cdot]\}\,n^{-d/2} + o_{n}(1)\,n^{-d/2}.$$

According to the remark above the distribution of the variable  $\mathbf{M} \frac{\mathscr{G}_n(\tau)}{\sqrt{n}}$  converges uniformly to the normal distribution with the density  $\mathbf{N}_{\Omega}(x)$ . Consequently the constant  $\lim_{n} \mathbf{M}_{\tau} \left[ \mathbf{N}_{\Omega_1} \left( \frac{\mathbf{M}\mathscr{G}_n(\tau_n)}{\sqrt{n}} \right) \right] = c$  is equal to  $\int_{\mathbb{R}^d} \mathbf{N}_{\Omega_1}(x) \mathbf{N}_{\Omega_2}(x) dx = \mathbf{N}_{\Omega}(0)$ , where  $\mathbf{N}_{\Omega}(x)$  is the density of the limit distribution of  $\frac{1}{\sqrt{n}} \mathscr{G}_n$ . Thus the theorem is proved.

# References

- Baldi, P., Bougerol, Ph., Crepel, P.: Théorème central limit local sur les déplacements de R<sup>d</sup>. C.R. Acad. Sci. Paris, Sér. A 283, 53-55 (1976)
- 2. Bougerol, Ph.: Thèse 3<sup>ème</sup> cycle. Université Paris VII (1977)
- Esseen, C.G.: On the concentration function of a sum of independent random variables. Z. Wahrscheinlichkeitstheorie verw. Gebiete 9, 290-308 (1968)
- Gorostiza, L.: The central limit theorem for random motions of d-dimensional euclidean space. Ann. of Probability 1, 603-612 (1973)
- 5. Kazdan, D.A.: A uniform distribution in the plane. Trudy Moskov Mat. Obšč. 14, 299-305 (1965)
- Le Page, E.: Théorèmes quotients pour certaines marches aleatoires. C. R. Acad. Sci. Paris, Sér. A 279, 69-72 (1974)
- 7. Maximov, V.M.: On the applicability of the central Limit theorem to sums of the form  $\sum f(\xi_1...\xi_i)$ . Izv. Vysš. Učebn. Zaved. Matematika **12**, 61–71 (1970)
- Maximov, V.M.: Uniform distribution of points and a local theorem for random motions. Soviet Math. Dokl. 18, 67-70 (1977)
- 9. Maximov, V.M.: A local theorem for lattice motions of the euclidean plane. Soviet Math. Dokl. 18, 1545-1549 (1977)
- Maximov, V.M., Tutubalin, V.N.: On the integral and local theorem in groups of Motions. 3rd Soviet Japanese Symposium. Proceedings. Tastikent 91-92 (1975) (in Russian)

- 11. Shepp, L.A.: A local limit theorem. Ann. Math. Statist. 35, 419-423 (1964)
- Stone, Ch.J.: A local limit theorem for nonlattice multidimensional distribution functions. Ann. Math. Statist. 36, 546–551 (1965)
- 13. Stone, Ch.J.: Ratio limit theorems for random walks on groups. Trans. Amer. Math. Soc. 125, 86-100 (1966)

Received August 18, 1977; in revised form June 1, 1979