

**ON THE RATE OF CONVERGENCE TO NORMAL LAW IN
HILBERT SPACE**

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(Translated by Amal Ellis)

1. Introduction

Let X_1, X_2, \dots be independent identically distributed random variables (r.v.'s) in a real separable Hilbert space H , $\mathbf{E}X_1 = 0$, $\beta_3 = \mathbf{E}|X_1|^3 < \infty$. Here and below, the symbol $|x|$, $x \in H$, denotes the norm in H .

Set $\sigma^2 = \mathbf{E}|X_1|^2$. Let Λ denote the covariance operator of the r.v. X_1 . Let $\sigma_1^2 \geq \sigma_2^2 \geq \dots \geq \sigma_k^2 \geq \dots$ denote the eigenvalues of Λ .

Let Z_1, Z_2, \dots take values in H , be independent and have Gaussian distribution with covariance operator Λ , where $\mathbf{E}Z_1 = 0$.

The goal of the present paper is estimation of the distance

$$\Delta_n(a) = \sup_r \left| \mathbf{P} \left(\sum_1^n X_i / \sqrt{n} \in S(a, r) \right) - \mathbf{P} \left(\sum_1^n Z_i / \sqrt{n} \in S(a, r) \right) \right|,$$

where $S(a, r)$ is the sphere $\{x: |x + a| \leq r, x \in H\}$.

This problem has been taken up by many authors (see [1]–[18]). The estimates improved steadily; however, for a long time it was unclear whether the estimate $\Delta_n(a) = O(1/\sqrt{n})$ was valid—the best one in the sense of dependence on n . Such an estimate was first derived in 1978 (see [12]) under the condition $H = l_2$ (this of course is no loss in generality), $a = 0$, while the coordinates of the random vector X_1 are mutually independent. More precisely, the estimate derived in [12] has the form

$$\Delta_n(0) < c \left(\beta_3 / \left(\prod_1^4 \sigma_j \right)^{3/4} + \sum_1^4 \mathbf{E}|\xi_j|^3 / \sigma_j^3 \right) / \sqrt{n},$$

where ξ_j is the j th coordinate of X_1 , $\sigma_j^2 = \mathbf{E}|\xi_j|^2$, and c is an absolute constant.

In [13], [16] this estimate is generalized to the case when the first 7 coordinates are independent of the others and $a \neq 0$.

As for the general case, in 1979 Götze [14] obtained the estimate $\Delta_n(0) < c(\beta_6 / \sigma_{31}^6)^3 n^{-1/2}$, where $\beta_6 = \mathbf{E}|X_1|^6$.

Finally, in 1981, Yurinskii [17], combining the methods of [13] and [14], obtained the estimate

$$(1) \quad \Delta_n(a) < c(\Lambda) \beta_3 (1 + |a|^3 / \sigma^3) n^{-1/2},$$

where $c(\Lambda)$ is a constant depending only on Λ .

A somewhat less accurate estimate $O(n^{-1/2} \log n)$ was announced in [18].

The drawback of estimate (1) is the fact that the explicit dependence of $c(\Lambda)$ on Λ is not indicated in it.

In the present paper we study the dependence of $\Delta_n(a)$ on the covariance operator of the r.v. X_1 truncated on a certain level L .

Let the operator $\Lambda_0(L)$ be defined by the relation

$$(\Lambda_0(L)x, x) = \mathbf{E}\{(X_1 - \mathbf{E}\{X_1 | |X_1| \leq L\}, x)^2 | |X_1| \leq L\}.$$

Set $\Lambda(L) = p_L \Lambda_0(L)$, where $p_L = \mathbf{P}(|X_1| \leq L)$. Let $\sigma_1^2(L) \geq \sigma_2^2(L) \geq \dots \geq \sigma_k^2(L) \geq \dots$ be the eigenvalues of the operator $\Lambda(L)$. We shall now formulate the main result.

Theorem. *There exists an absolute constant c such that, for $0 < L < \infty$,*

$$\Delta_n(a) \leq c \left(\beta_3 / \left(\prod_1^7 \sigma_j(L) \right)^{6/7} + L / \sigma^2 \sigma_1 \sigma_2 \right) (\sigma^3 + |a|^3) n^{-1/2}.$$

The proof is based on the method of estimating the characteristic function of the r.v. $\sum_1^n X_j$ suggested in [14].

Now let us agree on some notation to be used throughout the paper. We shall denote one-dimensional r.v.'s by ξ, η, ζ, \dots , other r.v.'s by X, Y, Z, W, \dots . If X is a r.v., then X' denotes an independent copy of X , while X^s denotes the symmetrization of X , i.e., $X^s = X - X'$. The symbol η_0 denotes the r.v. with density $p(u) = c_0((60/u) \sin(u/60))^{60}$, where c_0 is a normalizing factor. It is important for us that $\mathbf{E} e^{it\eta_0} = 0$ for $|t| \geq 1$.

Let $\chi(t)$ denote the indicator of the set $\{t: |t| \leq 1\}$. For any r.v. X , set $X(L) = \chi(|X|/L)X$. The symbol c stands for various absolute constants. We shall also use Vinogradov's symbol \ll ($A \ll B$ if there exists an absolute constant c such that $|A| \leq cB$). If a constant depends on some parameter α , then we shall use the notation $c(\alpha)$. Whenever the symbol O is used, the corresponding constant is absolute.

2. Ancillary Results

First of all we shall prove a special variant of a smoothing lemma.

Lemma 1. *For $i = 1, 2$, let X, Y_i, η be mutually independent and $Z_i(a) = X + Y_i + a$, $V_i(a, v) = |Z_i(a)|^2 + (\eta - v)|X|$, $F_i(r, a) = \mathbf{P}(|Z_i(a)|^2 < r)$, $G_i(r, a, v) = \mathbf{P}(V_i(a, v) < r)$, $\Delta = \sup(|F_1(r, a) - F_2(r, a)| / (\beta + |a|^3))$,*

$$\Delta(v) = \sup_{r, a} (|G_1(r, a, v) - G_2(r, a, v)| / (\beta + |a|^3)),$$

$$\delta = \sup_{r, a} (|\mathbf{P}(|Y_1 + a|^2 < r) - \mathbf{P}(|Y_2 + a|^2 < r)| / (\beta + |a|^3)),$$

where $\beta = \mathbf{E}|X|^3$. Then $(\forall \varepsilon > 0)$

$$\Delta < \alpha^{-1} \max[\Delta(\varepsilon), \Delta(-\varepsilon)] + 8\delta\alpha^{-1}(1 - \alpha) + 2q\varepsilon\beta^{-1}\mathbf{E}|X|, \text{ where } \alpha = \mathbf{P}(|\eta| \leq \varepsilon),$$

$$q = \sup_{r, a, h > 0} h^{-1} \mathbf{P}(r \leq |Y_2 + a|^2 \leq r + h).$$

Lemma 1 is not entirely traditional in the sense that the operator carrying out the smoothing is not convolution.

PROOF. It is not hard to see that $\mathbf{P}(|Z_1(a)|^2 + (\eta - \varepsilon)|X| < r, |\eta| \leq \varepsilon) \geq \mathbf{P}(|Z_1(a)|^2 < r)\mathbf{P}(|\eta| \leq \varepsilon)$. On the other hand, $\mathbf{P}(|Z_2(a)|^2 + (\eta - \varepsilon)|X| < r, |\eta| \leq \varepsilon) \leq \mathbf{P}(|Z_2(a)|^2 < r + 2\varepsilon|X|)\mathbf{P}(|\eta| \leq \varepsilon) \leq \mathbf{P}(|Z_2(a)|^2 < r)\mathbf{P}(|\eta| \leq \varepsilon) + 2q\varepsilon\mathbf{E}|X|\mathbf{P}(|\eta| \leq \varepsilon)$. Thus,

$$(2) \quad \begin{aligned} & (\mathbf{P}(|Z_1(a)|^2 < r) - \mathbf{P}(|Z_2(a)|^2 < r))\mathbf{P}(|\eta| \leq \varepsilon) \\ & \leq \mathbf{P}(V_1(a, \varepsilon) < r, |\eta| \leq \varepsilon) - \mathbf{P}(V_2(a, \varepsilon) < r, |\eta| \leq \varepsilon) \\ & \quad + 2q\varepsilon\mathbf{E}|X|\mathbf{P}(|\eta| \leq \varepsilon). \end{aligned}$$

Further,

$$(3) \quad \mathbf{P}(V_i(a, \varepsilon) < r, |\eta| > \varepsilon|X) = \int_{|u| > \varepsilon} \mathbf{P}(|Z_i(a)|^2 < r + (\varepsilon - u)|X||X)Q(du),$$

where Q is the distribution of η . By definition of δ ,

$$(4) \quad \begin{aligned} & |\mathbf{P}(|Z_1(a)|^2 < r + (\varepsilon - u)|X||X) - \mathbf{P}(|Z_2(a)|^2 < r + (\varepsilon - u)|X||X)| \\ & \leq \delta(\beta + |a + X|^3) \leq 4\delta(\beta + |a|^3 + |X|^3). \end{aligned}$$

From (3) and (4) it follows that

$$(5) \quad |\mathbf{P}(V_1(a, \varepsilon) < r, |\eta| > \varepsilon) - \mathbf{P}(V_2(a, \varepsilon) < r, |\eta| > \varepsilon)| \leq 4\delta(2\beta + |a|^3)\mathbf{P}(|\eta| > \varepsilon).$$

Combining (2) and (5), we arrive at the inequality $\alpha(F_1(r, a) - F_2(r, a)) \leq G_1(r, a, \varepsilon) - G_2(r, a, \varepsilon) + 4\delta(1 - \alpha)(2\beta + |a|^3) + 2q\varepsilon\alpha\mathbf{E}|X|$.

Hence we have:

$$\begin{aligned} & \sup_{r, a} ((F_1(r, a) - F_2(r, a))/(\beta + |a|^3)) \\ & \leq \alpha^{-1}\Delta(\varepsilon) + 8\delta\alpha^{-1}(1 - \alpha) + 2q\varepsilon\mathbf{E}|X|/\beta. \end{aligned}$$

Analogous arguments show that

$$\begin{aligned} & \sup_{r, a} ((F_2(r, a) - F_1(r, a))/(\beta + |a|^3)) \\ & \leq \alpha^{-1}\Delta(-\varepsilon) + 8\delta\alpha^{-1}(1 - \alpha) + 2q\varepsilon\mathbf{E}|X|/\beta. \end{aligned}$$

The assertion of the lemma follows from the last two inequalities.

Lemma 2. Let X, Y, V, η be mutually independent, $Z = X + Y$, $\mathbf{E}e^{it\eta} = 0$, $\forall |t| \leq \omega$, $\text{Im } t = 0$. Then ($\forall \varepsilon > 0$)

$$\begin{aligned} & \mathbf{E} \exp \{it(|V + Z|^2 + (\eta + \varepsilon)|X|)\} \\ & = (1 + it\mathbf{E}|V|^2) \mathbf{E} \exp \{it(|Z|^2 + (\eta + \varepsilon)|X|)\} \\ & \quad + 2it \mathbf{E}(Z, V) \exp \{it(|Z|^2 + (\eta + \varepsilon)|X|)\} \\ & \quad - 2t^2 \mathbf{E}(Z, V)^2 \exp \{it(|Z|^2 + (\eta + \varepsilon)|X|)\} + R(t), \end{aligned}$$

where

$$\begin{aligned} R(t) & \ll \mathbf{E}|V|^3[\mathbf{E}^{1/2}f_{X^*}(2t)|\chi(t, X, X', \omega)|(|t|^{3/2} + \mathbf{E}^{1/2}|Y|^2t^2) \\ & \quad + t^2 \mathbf{E}^{1/2}|X||X'|f_{X^*}(2t)|\chi(t, X, X', \omega)] \end{aligned}$$

$$\begin{aligned}
& + |t|^3 \sum_{j+k=3} \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} |X|^k |X'|^k |f_{X^s}(2t)| \chi(t, X, X', \omega), \\
& \chi(t, X, X', \omega) = \chi(t \max[|X|, |X'|]/\omega), \\
& f_x(t) = \mathbf{E} \exp \{it(Y, x)\}.
\end{aligned}$$

PROOF. For brevity, we shall use the notation $s = it$. We start from the representation

$$\begin{aligned}
& \mathbf{E} \exp \{s(|Z + V|^2 + (\eta + \varepsilon)|X|)\} \\
(6) \quad & = \mathbf{E} \exp \{s(\Omega + |V|^2)\} \\
& = \mathbf{E} \exp \{s\Omega\} + s\mathbf{E}|V|^2 \exp \{s\Omega\} + \mathbf{E}r(s, V) \exp \{s\Omega\} \\
& = A_1 + A_2 + A_3,
\end{aligned}$$

where $\Omega = |Z|^2 + 2(Z, V) + (\eta + \varepsilon)|X|$, $r(s, V) = \exp \{s|V|^2\} - 1 - s|V|^2$.

It is not hard to see that

$$(7) \quad r(s, V) = O(|s|^{3/2}|V|^3).$$

Obviously,

$$(8) \quad A_3 = \mathbf{E}r(s, V) \mathbf{E} \{\exp \{s\Omega\} | V\}.$$

Further, ($\forall x \in H$)

$$\begin{aligned}
\mathbf{E}_1(x) & \equiv \mathbf{E} \exp \{s(|Z|^2 + 2(Z, x) + (\eta + \varepsilon)|X|)\} \\
& = \mathbf{E} \exp \{s(|Y|^2 + 2(Y, x))\} \mathbf{E} \{\exp \{s(|X|^2 + 2(X, x + Y) + (\eta + \varepsilon)|X|)\} | Y\}.
\end{aligned}$$

Using Cauchy's inequality, we have:

$$\begin{aligned}
|\mathbf{E}_1(x)| & \leq \mathbf{E}^{1/2} |\mathbf{E} \{\exp \{s(|X|^2 + 2(X, x + Y) + (\eta + \varepsilon)|X|)/Y\}|^2 \\
& = \mathbf{E}^{1/2} \mathbf{E} \{\exp \{s(|X|^2 - |X'|^2 + 2(X - X', x + Y) \\
& \quad + (\eta + \varepsilon)|X| - (\eta' + \varepsilon)|X'|)\} | Y\}.
\end{aligned}$$

Changing the order of integration, we get

$$\begin{aligned}
|\mathbf{E}_1(x)| & \leq \mathbf{E}^{1/2} |\mathbf{E} \{\exp \{s\eta|X|\} | X\} \mathbf{E} \exp \{-s\eta'|X'|\} | X'\} \\
& \quad \cdot \mathbf{E} \{\exp \{2s(Y + x, X - X')\} | X, X'\}.
\end{aligned}$$

Since $\mathbf{E} \exp \{s|y|\eta\} = 0$, $\mathbf{E} \exp \{s|y|\eta'\} = 0$, $\forall |s| > \omega/|y|$, $y \in H$, this means that

$$\begin{aligned}
(9) \quad |\mathbf{E}_1(x)| & \leq \mathbf{E}^{1/2} |f_{X^s}(2t)| \chi(t|X|/\omega) \chi(t|X'|/\omega) \\
& = \mathbf{E}^{1/2} |f_{X^s}(2t)| \chi(t, X, X', \omega).
\end{aligned}$$

From (7)-(9) it follows that

$$(10) \quad A_3 \ll |t|^{3/2} \mathbf{E} |V|^3 \mathbf{E}^{1/2} |f_{X^s}(2t)| \chi(t, X, X', \omega).$$

Let us now turn to the estimation of A_1 . It is not hard to see that

$$\begin{aligned}
(11) \quad A_1 & = \mathbf{E} e^{s\Omega_1} + 2s\mathbf{E}(Z, V) e^{s\Omega_1} + 2s^2\mathbf{E}(Z, V)^2 e^{s\Omega_1} \\
& \quad + \mathbf{E}(Z, V)^3 R_1(s, V, Z) e^{s\Omega_1} \equiv \sum_1^4 A_{1,j}
\end{aligned}$$

where

$$\Omega_1 = |Z|^2 + (\eta + \varepsilon)|X|,$$

$$R_1(s, V, Z) = (e^{2s(Z, V)} - 2(Z, V)s - 2s^2(Z, V)^2)/(Z, V)^3.$$

Writing $R_1(s, V, Z)$ in integral form, we have

$$R_1(s, V, Z) = -4i \int_0^t \exp\{2iu(Z, V)\}(t-u)^2 du.$$

Consequently,

$$(12) \quad \begin{aligned} A_{1,4} &= -4i \int_0^t \mathbf{E}(Z, V)^3 \exp\{s\Omega_1 + 2iu(Z, V)\}(t-u)^2 du \\ &= -4i \sum_0^3 C_3^k \int_0^t \mathbf{E}(X, V)^k (Y, V)^{3-k} \exp\{s\Omega_1 + 2iu(Z, V)\}(t-u)^2 du. \end{aligned}$$

Now let us estimate

$$\mathbf{E}_{kj}(x) \equiv \mathbf{E}(X, x)^k (Y, x)^j \exp\{s\Omega_1 + iu(Z, x)\}, x \in H.$$

Obviously,

$$\begin{aligned} \mathbf{E}_{kj}(x) &= \mathbf{E}\{(Y, x)^j \exp\{s|Y|^2 + iu(Y, x)\} \\ &\quad \cdot \mathbf{E} \exp\{s(|X|^2 + 2(X, Y) + (\eta + \varepsilon)|X|) + iu(X, x)\} (X, x)^k | Y\}. \end{aligned}$$

Whence

$$(13) \quad |\mathbf{E}_{kj}(x)| \leq \mathbf{E}^{1/2}|(Y, x)|^{2j} \mathbf{E}^{1/2} |\mathbf{E}\{(X, x)^k \exp\{s\Omega_2 + iu(X, x)\} | Y\}|^2,$$

where $\Omega_2 = |X|^2 + 2(X, Y) + (\eta + \varepsilon)|X|$.

Using the same arguments as in the derivation of (9), we arrive at the estimate

$$(14) \quad \begin{aligned} &\mathbf{E}|\mathbf{E}\{(X, x)^k \exp\{s\Omega_2 + iu(X, x)\} | Y\}|^2 \\ &= \mathbf{E}(X, x)^k (X', x)^k \exp\{s(|X|^2 - |X'|^2 \\ &\quad + 2(X - X', Y) + (\eta + \varepsilon)|X| - (\eta' + \varepsilon)|X'|) + iu(X - X', x)\} \\ &\leq |x|^{2k} \mathbf{E}|X|^k |X'|^k |f_{X'}(2t)| \chi(t, X, X', \omega) \equiv A(k)|x|^{2k}. \end{aligned}$$

From (13) and (14) it follows that

$$(15) \quad |\mathbf{E}_{kj}(x)| \leq |x|^{k+j} \mathbf{E}^{1/2} |Y|^{2j} A^{1/2}(k).$$

Combining (12) and (15), we obtain

$$(16) \quad A_{1,4} \ll |t|^3 \mathbf{E}|V|^3 \sum_{j+k=3} A^{1/2}(k) \mathbf{E}^{1/2} |Y|^{2j}.$$

From (11) and (16) it follows that

$$(17) \quad A_1 = \sum_1^3 A_{1,j} + O\left(|t|^3 \mathbf{E}|V|^3 \sum_{j+k=3} \mathbf{E}^{1/2} |Y|^{2j} A^{1/2}(k)\right).$$

Now let us get an estimate for A_2 . Evidently,

$$(18) \quad \begin{aligned} A_2 &= s\mathbf{E} \exp \{s(|Z|^2 + (\eta + \varepsilon)|X|)\} \mathbf{E}|V|^2 \\ &\quad + 2s\mathbf{E} \exp \{s\Omega_1\}(Z, V) |V|^2 R_2(s, V, Z) \\ &\equiv A_{2,1} + A_{2,2}, \end{aligned}$$

where

$$R_2(s, V, Z) = \int_0^t \exp \{2iu(Z, V)\} du.$$

The same arguments as in the derivation of (16) lead to

$$(19) \quad A_{2,2} \ll t^2 \mathbf{E}|V|^3 \sum_{j+k=1} \mathbf{E}^{1/2} |Y|^{2j} A^{1/2}(k).$$

From (6), (10), (11), (17)–(19) follows the assertion of the lemma.

Lemma 3. *Let X, Y, V, η satisfy the conditions of Lemma 2. Then ($\forall \varepsilon > 0, m > 0$)*

$$\begin{aligned} &|\mathbf{E}(Z, V)^m \exp \{it(|Z|^2 + (\eta + \varepsilon)|X|)\}| \\ &\leq c(m) \mathbf{E}|V|^m \sum_{j+k=m} \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} |X|^k |X'|^k |f_{X^s}(2t)|_{\mathcal{X}(t, X, X', \omega)}. \end{aligned}$$

PROOF. Introduce the notation

$$\mathbf{E}_m = \mathbf{E}(Z, V)^m \exp \{it(|Z|^2 + (\eta + \varepsilon)|X|)\},$$

$$\mathbf{E}_{kj}(x, X, Y) = \mathbf{E}(x, X)^k (x, Y)^j \exp \{it(|Z|^2 + (\eta + \varepsilon)|X|)\}.$$

Setting $s = it$, we have the following chain of inequalities:

$$\begin{aligned} |\mathbf{E}_{kj}(x, X, Y)| &= |\mathbf{E}(Y, x)^j \exp \{s|Y|^2\} \mathbf{E}\{(X, x)^k \\ &\quad \cdot \exp \{s(|X|^2 + 2(X, Y) + (\eta + \varepsilon)|X|)\} Y\}| \\ &\leq \mathbf{E}^{1/2}(Y, x)^{2j} \mathbf{E}^{1/2} |\mathbf{E}\{(X, x)^k \exp \{s(|X|^2 + 2(X, Y) \\ &\quad + (\eta + \varepsilon)|X|)\} Y\}|^2 \\ &\leq |x|^j \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} \mathbf{E}\{(X, x)^k (X', x)^k \\ &\quad \cdot \exp \{s(|X|^2 - |X'|^2 + 2(X^s, Y) \\ &\quad + \varepsilon(|X| - |X'|) + \eta|X| - \eta'|X'|)\} Y\}| \\ &= |x|^j \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} (X, x)^k (X', x)^k \\ &\quad \cdot \exp \{s(|X|^2 - |X'|^2 + \varepsilon(|X| - |X'|))\} \\ &\quad \cdot \mathbf{E}\{\exp \{s(2(X^s, Y) + \eta|X| - \eta'|X'|)\} |X, X'\} \\ &\leq |x|^{j+k} \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} |X|^k |X'|^k |\mathbf{E}\{\exp \{s\eta|X|\} |X\} \\ &\quad \cdot \mathbf{E}\{\exp \{s\eta'|X'| |X'\} \mathbf{E} \exp \{2s(X^s, Y)\} |X, X'\}| \\ &\leq |x|^{j+k} \mathbf{E}^{1/2} |X|^k |X'|^k |f_{X^s}(2t)|_{\mathcal{X}(t, X, X', \omega)} \mathbf{E}^{1/2} |Y|^{2j}. \end{aligned}$$

It remains to apply the identity

$$\mathbf{E}_m = \sum_{k+j=m} C_m^k \mathbf{E} \mathbf{E}_{kj}(V, X, Y).$$

Lemma 3, just as the preceding Lemma 2, is a variant of Lemma (3.37) in [14].

Lemma 4. *Let $X_i, Y, i = 1, 2$, be independent and*

$$\begin{aligned} \tilde{F}_i(r, a) &= \mathbf{P}(|X_i + a| < r), \\ \tilde{G}_i(r, a) &= \mathbf{P}(|X_i + Y + a| < r), \\ \tilde{\Delta}(a) &= \sup_r |\tilde{F}_1(r, a) - \tilde{F}_2(r, a)|, \\ \tilde{\delta} &= \sup_r |\tilde{G}_1(r, a) - \tilde{G}_2(r, a)|. \end{aligned}$$

Then $(\forall \varepsilon > 0, \alpha > \frac{1}{2})$

$$\tilde{\Delta}(a) < (\tilde{\delta} + 2\alpha\tilde{q}\varepsilon)/(2\alpha - 1),$$

where

$$\alpha = \mathbf{P}(|Y| \leq \varepsilon), \quad \tilde{q} = \sup_{r, h > 0} h^{-1}(\tilde{F}_2(r+h, a) - \tilde{F}_2(r, a)).$$

PROOF. By the independence of X_1 and Y ,

$$\begin{aligned} &\mathbf{P}(|X_1 + Y + a| < r + \varepsilon, |Y| \leq \varepsilon) \\ &\cong \mathbf{P}(|X_1 + a| < r, |Y| \leq \varepsilon) = \mathbf{P}(|X_1 + a| < r)\mathbf{P}(|Y| \leq \varepsilon). \end{aligned}$$

Analogously,

$$\begin{aligned} &\mathbf{P}(|X_2 + Y + a| < r + \varepsilon, |Y| \leq \varepsilon) \\ &\cong \mathbf{P}(|X_2 + a| < r + 2\varepsilon, |Y| \leq \varepsilon) \\ &= \mathbf{P}(|X_2 + a| < r + 2\varepsilon)\mathbf{P}(|Y| \leq \varepsilon) \\ &\cong (\mathbf{P}(|X_2 + a| < r) + 2\tilde{q}\varepsilon)\mathbf{P}(|Y| \leq \varepsilon). \end{aligned}$$

Thus,

$$\begin{aligned} (\tilde{F}_1(r, a) - \tilde{F}_2(r, a))\alpha &\cong \tilde{G}_1(r + \varepsilon, a) - \tilde{G}_2(r + \varepsilon, a) \\ &\quad + 2\tilde{q}\varepsilon\alpha + \mathbf{P}(|X_2 + Y + a| < r + \varepsilon, |Y| > \varepsilon) \\ &\quad - \mathbf{P}(|X_1 + Y + a| < r + \varepsilon, |Y| > \varepsilon). \end{aligned}$$

On the other hand,

$$\begin{aligned} &|\mathbf{P}(|X_1 + Y + a| < r, |Y| > \varepsilon) - \mathbf{P}(|X_2 + Y + a| < r, |Y| > \varepsilon)| \\ &= \left| \int_{|y| > \varepsilon} (\tilde{F}_1(r, a+y) - \tilde{F}_2(r, a+y))Q(dy) \right| < \tilde{\Delta}(a)(1 - \alpha), \end{aligned}$$

where Q is the distribution of Y . Consequently,

$$\sup_r (\tilde{F}_1(r, a) - \tilde{F}_2(r, a)) \leq \tilde{\delta}/\alpha + 2\tilde{q}\varepsilon + \tilde{\Delta}(a)(1 - \alpha)/\alpha.$$

Similarly,

$$\sup_r (\tilde{F}_2(r, a) - \tilde{F}_1(r, a)) \leq \tilde{\delta}/\alpha + 2\tilde{q}\varepsilon + \tilde{\Delta}(a)(1-\alpha)/\alpha.$$

The assertion of the lemma follows from the last two estimates.

Lemma 4 carries over the corresponding assertions for R^k to Hilbert space (e.g., see [6], [20]).

Lemma 5. *Let ξ_1 and ξ_2 be independent and normal, respectively, $N(a_1, \sigma_1)$ and $N(a_2, \sigma_2)$, let $p(r)$ be the distribution density of $\xi_1^2 + \xi_2^2$.*

Then $(\forall a_1, a_2, \sigma_1, \sigma_2)$

$$(20) \quad p(r) \ll \min [(1/\sigma_1 + 1/\sigma_2)/\sqrt{r}, 1/\sigma_1\sigma_2].$$

PROOF. It is not hard to see that

$$\begin{aligned} 2\pi\sigma_1\sigma_2 p(r) &\leq \int_0^r (r-\rho)^{-1/2} \rho^{-1/2} \exp\{-(r-\rho)^{1/2} - |a_1|\}^2 / 2\sigma_1^2 \\ &\quad - (\rho^{1/2} - |a_2|)^2 / 2\sigma_2^2\} d\rho = \int_0^{r/2} + \int_{r/2}^r. \end{aligned}$$

Further,

$$\int_0^{r/2} < (2/r)^{1/2} \int_0^{r/2} \rho^{-1/2} \exp\{-(\rho^{1/2} - |a_2|)^2 / 2\sigma_2^2\} d\rho \ll \sigma_2 / r^{1/2}.$$

Similarly,

$$\int_{r/2}^r \ll \sigma_1 / r^{1/2}.$$

Consequently,

$$p(r) \ll (1/\sigma_1 + 1/\sigma_2) / r^{1/2}.$$

On the other hand,

$$\int_0^r < \int_0^r (r-\rho)^{-1/2} \rho^{-1/2} d\rho \leq c.$$

This means that $p(r) \ll 1/\sigma_1\sigma_2$.

The lemma is completely proved.

Lemma 6. *Let $X = \{\xi_i\}_1^\infty$, $X \in l_2$, where the ξ_i are mutually independent and normal $N(0, \sigma_i)$, $p_k(r, a)$ is the distribution density of $|X + a|^k$, $a \in l_2$, $k = 1, 2$.*

Then

$$p_1(r, a) \ll (B + |a|) / \sigma_1\sigma_2, \quad p_2(r, a) \ll \min [(B + |a|) / \sqrt{r}, 1] / \sigma_1\sigma_2,$$

$$\text{where } B^2 = \sum_1^\infty \sigma_i^2.$$

PROOF. Let $a = \{a_i\}_1^\infty$. Let us represent $|X + a|^2$ in the form

$$|X + a|^2 = (\xi_1 + a_1)^2 + (\xi_2 + a_2)^2 + |Y|^2, \quad \text{where } Y = \{\xi_i + a_i\}_3^\infty.$$

Obviously,

$$p_2(r, a) = \int_0^r p(r - \rho) d\mathbf{P}(|Y|^2 < \rho),$$

where $p(r)$ is the density of $(\xi_1 + a_1)^2 + (\xi_2 + a_2)^2$. Using the estimate (see (20)) $p(r) \ll 1/\sigma_1\sigma_2$, we get from here that $p_2(r, a) \ll 1/\sigma_1\sigma_2$.

Further,

$$\int_0^{r/2} p(r - \rho) d\mathbf{P}(|Y|^2 < \rho) \ll (1/\sigma_1 + 1/\sigma_2)/\sqrt{r},$$

since, due to Lemma 5, $p(r) < (1/\sigma_1 + 1/\sigma_2)/\sqrt{r}$.

On the other hand, by the same Lemma 5,

$$\int_{r/2}^r \ll \sigma_1^{-1}\sigma_2^{-1}\mathbf{P}(|Y|^2 > r/2).$$

Obviously,

$$\mathbf{P}(|Y|^2 > r/2) \ll (B + |a|)r^{-1/2}.$$

Therefore,

$$p_2(r, a) \ll \min [(B + |a|)/\sqrt{r}, 1]/\sigma_1\sigma_2.$$

Finally,

$$p_1(r, a) = 2rp_2(r^2, a) \ll (B + |a|)/\sigma_1\sigma_2.$$

Lemma 7. Let the r.v.'s Y_1, Y_2, \dots, Y_m take values in \mathbb{R}^k and be identically distributed, $\theta = \{\theta_j\}_1^k$,

$$|\mathbf{E} \exp \{i(Y_1, \theta)\}| \leq \exp \left\{ -\sum_1^k \sigma_j^2 \theta_j^2 \right\},$$

if

$$\max_j |\theta_j| \leq \gamma, \quad S = \sum_1^m Y_j + Z,$$

where Z does not depend on $Y_j, j = 1, \dots, m$,

$$\mathbf{E} \exp \{i(Z, \theta)\} = \prod_1^k \varphi(\theta_j), \quad \text{where } \varphi_j(\theta_j) = 0 \text{ for } |\theta_j| > \gamma.$$

Then there exists a distribution density $p(u_1, u_2, \dots, u_k)$ of the r.v. S and $p(u_1, u_2, \dots, u_k) \leq c(k)/m^{k/2} \prod_1^k \sigma_j$.

PROOF. By the inversion formula,

$$(21) \quad p(u_1, u_2, \dots, u_k) = (2\pi i)^{-k} \int_{\mathbb{R}^k} \exp \{-i(u, \theta)\} f^m(\theta) \varphi(\theta) d\theta,$$

where

$$f(\theta) = \mathbf{E} \exp \{i(Y_1, \theta)\}, \quad \varphi(\theta) = \prod_1^k \varphi_j(\theta_j), \quad d\theta = \prod_1^k d\theta_j, \quad \theta = \{\theta_j\}_1^k, \\ u = \{u_j\}_1^k.$$

From the conditions of the lemma it follows that

$$(22) \quad \int_{\mathbb{R}^k} |f^m(\theta) \varphi(\theta)| d\theta \leq \int_{-\infty}^{\infty} \exp \left\{ -m \sum_1^k \sigma_j^2 \theta_j^2 \right\} d\theta \leq c(k)/m^{k/2} \prod_1^k \sigma_j.$$

Combining (21) and (22), we get the assertion of the lemma.

3. Proof of the Main Theorem

Let us choose in H an orthonormal basis $\{e_j\}_1^\infty$ so that in it the covariance operator $\Lambda(L)$ has diagonal form. Below, when referring to the coordinates of a vector $x \in H$, we shall always mean the coordinates with respect to this basis.

Introduce the notation $\bar{X}_j = X_j(\sigma\sqrt{n})$. Let P_k be the distribution of $n^{-1/2} \sum_1^k \bar{X}_j$, Φ_k be the distribution of $n^{-1/2} \sum_1^k Z_j$. For even n ,

$$(23) \quad P_n - \Phi_n = (P_{n/2} - \Phi_{n/2}) * P_{n/2} + (P_{n/2} - \Phi_{n/2}) * \Phi_{n/2}.$$

Now let us introduce a smoothing r.v. W not depending on $\{X_j\}_1^\infty$ and $\{Z_j\}_1^\infty$. Let $W = \sum_1^7 \eta_j e_j$, where the η_j are mutually independent and coincide in distribution with $4\eta_0 L/\sqrt{n}$, P_W is the distribution of the r.v. W .

Set

$$\rho_n = \sup_{r,a} |(P_{n/2} - \Phi_{n/2}) * P_{n/2} * P_W(S(a, r))| / (\sigma^3 + |a|^3).$$

Obviously, $P_n * P_W$ is the distribution of $n^{-1/2} \sum_1^n \bar{X}_j + W$, while $\Phi_{n/2} * P_{n/2} * P_W$ is the distribution of

$$n^{-1/2} \left(\sum_1^{n/2} \bar{X}_j + \sum_{n/2+1}^n Z_j \right) + W$$

under the condition that $\{X_j\}_1^\infty$ does not depend on $\{Z_j\}_1^\infty$.

Let us estimate ρ_n by means of Lemma 1, setting

$$X = n^{-1/2} \sum_{n/2+1}^n \bar{X}_j + W, \\ Y_1 = n^{-1/2} \sum_1^{n/2} \bar{X}_j, \quad Y_2 = n^{-1/2} \sum_1^{n/2} Z_j, \quad \eta = 6\eta_0 L/\sqrt{n}.$$

First let us estimate $\Delta(\varepsilon)$. Set

$$Q_j(r, a, \varepsilon) = \mathbf{P} \left(\left| \left(\sum_1^j Z_i + \sum_{j+1}^n \bar{X}_i \right) n^{-1/2} + W + a \right|^2 + (\eta - \varepsilon) |X| < r \right), \\ j = 0, \dots, n/2.$$

Obviously,

$$\begin{aligned} Q_0(r, a, \varepsilon) &= G_1(r, a, \varepsilon), \\ Q_{n/2}(r, a, \varepsilon) &= G_2(r, a, \varepsilon). \end{aligned}$$

Therefore,

$$(24) \quad G_1(r, a, \varepsilon) - G_2(r, a, \varepsilon) = \sum_0^{n/2-1} (Q_j(r, a, \varepsilon) - Q_{j+1}(r, a, \varepsilon)).$$

By the inversion formula,

$$(25) \quad Q_j(r, a, \varepsilon) - Q_{j+1}(r, a, \varepsilon) = (1/2\pi i) \int_{-\infty}^{\infty} (q_j(t) - q_{j+1}(t))(e^{-itr}/it) dt,$$

where

$$q_j(t) = \int_{-\infty}^{\infty} e^{itr} Q_j(dr, a, \varepsilon).$$

To estimate $q_j(t) - q_{j+1}(t)$ we apply Lemma 2, setting

$$\begin{aligned} X &= n^{-1/2} \sum_{n/2+1}^n \bar{X}_i + W, \\ Y &= n^{-1/2} \left(\sum_1^j Z_i + \sum_{j+2}^{n/2} \bar{X}_i \right) + a, \\ V &= n^{-1/2} \bar{X}_{j+1} \text{ or } n^{-1/2} Z_{j+1}. \end{aligned}$$

As a result we obtain:

$$\begin{aligned} (26) \quad q_j(t) - q_{j+1}(t) &= (\mathbf{E}|\bar{X}_{j+1}|^2 - \mathbf{E}|Z_{j+1}|^2) \mathbf{E}f(t, X, Y) itn^{-1} \\ &\quad + 2\mathbf{E}f(t, X, Y)((\bar{X}_{j+1}, Z) - (Z_{j+1}, Z)) itn^{-1/2} \\ &\quad + 2\mathbf{E}f(t, X, Y)((Z_{j+1}, Z)^2 - (\bar{X}_{j+1}, Z)^2) t^2 n^{-1} + R_j(t) - R_{j+1}(t) \\ &\equiv A(t) + R_j(t) - R_{j+1}(t), \end{aligned}$$

where

$$f(t, X, Y) = \exp \{ it(|Z|^2 + (\eta - \varepsilon)|X|) \},$$

and $R_j(t)$ and $R_{j+1}(t)$ are the remainder terms in the expansion for $q_j(t)$ and $q_{j+1}(t)$.

By Lemma 3,

$$\begin{aligned} (27) \quad |\mathbf{E}(\bar{X}_{j+1}, Z)f(t, X, Y)| &= |\mathbf{E}(X_{j+1} - \bar{X}_{j+1}, Z) \\ f(t, X, Y)| &\ll \mathbf{E}|X_{j+1} - \bar{X}_{j+1}| \sum_{j+k=1} \mathbf{E}^{1/2}|Y|^{2j} \\ &\quad \cdot \mathbf{E}^{1/2}|X|^k |X'|^k |f_{X'}(2t)| \chi(t, X, X', \omega), \\ |\mathbf{E}f(t, X, Y)| &\ll \mathbf{E}^{1/2}|f_{X'}(2t)| \chi(t, X, X', \omega). \end{aligned}$$

Further,

$$\begin{aligned} \mathbf{E}(\bar{X}_{j+1}, Z)^2 f(t, X, Y) &= \mathbf{E}(X_{j+1}, Z)^2 f(t, X, Y) \\ &\quad - \mathbf{E}(X_{j+1} - \bar{X}_{j+1}, Z)^2 f(t, X, Y). \end{aligned}$$

It is not hard to see that

$$\begin{aligned} \mathbf{E}(X_{j+1}, Z)^2 f(t, X, Y) &= \mathbf{E}f(t, X, Y) \mathbf{E}\{(X_{j+1}, Z)^2 | X, Y\} \\ &= \mathbf{E}f(t, X, Y) \mathbf{E}\{(Z_{j+1}, Z)^2 | X, Y\} \\ &= \mathbf{E}(Z_{j+1}, Z)^2 f(t, X, Y). \end{aligned}$$

On the other hand, by Lemma 3,

$$\begin{aligned} &\mathbf{E}(X_{j+1} - \bar{X}_{j+1}, Z)^2 f(t, X, Y) \\ &\ll \mathbf{E}|X_{j+1} - \bar{X}_{j+1}|^2 \sum_{j+k=2} \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} |X|^k \\ &\quad \cdot |X'|^k |f_{X'}(2t)| \chi(t, X, X', \omega). \end{aligned}$$

Hence,

$$\begin{aligned} &\mathbf{E}((\bar{X}_{j+1}, Z)^2 - (Z_{j+1}, Z)^2) f(t, X, Y) \\ (28) \quad &= \mathbf{E}((X_{j+1}, Z)^2 - (Z_{j+1}, Z)^2 - (X_{j+1} - \bar{X}_{j+1}, Z)^2) f(t, X, Y) \\ &\ll \sum_{j+k=2} \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} |f_{X'}(2t)| |X|^k |X'|^k \chi(t, X, X', \omega). \end{aligned}$$

Obviously,

$$\begin{aligned} (29) \quad &\mathbf{E}|X_j|^2 - \mathbf{E}|\bar{X}_j|^2 = \mathbf{E}\{|X_j|^2; |X_j| > \sigma\sqrt{n}\} \leq \beta_3/\sigma\sqrt{n}, \\ &\mathbf{E}|X_j - \bar{X}_j| \leq \beta_3/\sigma^2 n, \quad \mathbf{E}|X_j - \bar{X}_j|^2 \leq \beta_3/\sigma\sqrt{n}. \end{aligned}$$

Finally,

$$(30) \quad \mathbf{E}(Z_{j+1}, Z) f(t, X, Y) = \mathbf{E}f(t, X, Y) \mathbf{E}\{(Z_{j+1}, Z) | X, Y\} = 0.$$

From (27)–(30) it follows that

$$\begin{aligned} (31) \quad &A(t) \ll \beta_3 n^{-3/2} (\sigma^{-1} |t| \mathbf{E}^{1/2} |f_{X'}(2t)| \chi(t, X, X', \omega) \\ &\quad + \sigma^{-3} \sum_{1 \leq j+k \leq 2} \sigma^{j+k} |t|^{j+k} \mathbf{E}^{1/2} |Y|^{2j} \mathbf{E}^{1/2} |X|^k |X'|^k |f_{X'}(2t)| \chi(t, X, X', \omega)). \end{aligned}$$

Using the inequality for the moments of sums of independent r.v.'s with values in a Banach space (e.g., see [19]), we have

$$\mathbf{E}|X|^t \leq c(t) (\mathbf{E}|\bar{X}_1|^t n^{1-t/2} + \mathbf{E}^{t/2} |\bar{X}_1|^2 + \mathbf{E}|X - W| + \mathbf{E}|W|^t), \quad t \geq 1.$$

Further,

$$\mathbf{E}|\bar{X}_1|^t \leq \sigma^{t-2} n^{t/2-1} \mathbf{E}|X_1|^2 = \sigma^t n^{t/2-1}, \quad t \geq 2.$$

It is not hard to see that

$$\mathbf{E}|X - W| \leq (\mathbf{E}|X_1|^2/2)^{1/2} + \beta_3/2\sigma^2\sqrt{n} \leq 2\sigma,$$

since without loss of generality $\beta_3/\sigma^3\sqrt{n} \leq 2$. Finally,

$$\mathbf{E}|W|^t \ll (L/\sqrt{n})^t \leq \sigma^t, \quad 0 < t \leq 58,$$

since without loss of generality $L\sigma/\sigma_1\sigma_2\sqrt{n} \leq 1$.

As a result we have:

$$(32) \quad \mathbf{E}|X|^t \leq c(t)\sigma^t, \quad 1 \leq t \leq 58.$$

Similarly,

$$(33) \quad \mathbf{E}|Y|^t \leq c(t)(\sigma^t + |a|^t).$$

Let us turn to estimation of $f_x(t)$. In the case at hand,

$$(34) \quad f_x(t) = (\mathbf{E} \exp \{it(\bar{X}_1, x)/n^{1/2}\})^{n/2-j-1} \cdot (\mathbf{E} \exp \{it(Z_1, x)/n^{1/2}\})^j \exp \{it(a, x)\}.$$

From the identity

$$\begin{aligned} \mathbf{E} \exp \{it(\bar{X}_1, x)\} &= \mathbf{E}\{\exp \{it(\bar{X}_1, x)\}; |\bar{X}_1| \leq L\} \\ &\quad + \mathbf{E}\{\exp \{it(\bar{X}_1, x)\}; |\bar{X}_1| > L\} \end{aligned}$$

it follows that

$$|\mathbf{E} \exp \{it(\bar{X}_1, x)\}| \leq 1 + (|\varphi(t)| - p),$$

where

$$\varphi(t) = \mathbf{E}\{\exp \{it(\bar{X}_1, x)\}; |\bar{X}_1| \leq L\}, \quad p = \mathbf{P}(|\bar{X}_1| \leq L).$$

Obviously,

$$\begin{aligned} p - |\varphi| &= (p^2 - |\varphi|^2)/(p + |\varphi|) \geq (p^2 - |\varphi|^2)/2p, \\ |p^2 - |\varphi(t)|^2 - 2^{-1}t^2\mathbf{E}\{(\bar{X}_1^s, x)^2; |\bar{X}_1| \leq L, |\bar{X}'_1| \leq L\} \\ &\leq 6^{-1}|t|^3\mathbf{E}\{|\bar{X}_1^s, x|^3; |\bar{X}_1| \leq L, |\bar{X}'_1| \leq L\} \\ &\leq 3^{-1}|t|^3|x|L\mathbf{E}\{(\bar{X}_1^s, x)^2; |\bar{X}_1| \leq L, |\bar{X}'_1| \leq L\}. \end{aligned}$$

It is not hard to see that

$$\begin{aligned} \mathbf{E}\{(\bar{X}_1^s, x)^2; |\bar{X}_1| \leq L, |\bar{X}'_1| \leq L\} &= 2(p\mathbf{E}(X_1(L), x)^2 - (\mathbf{E}X_1(L), x)^2) \\ &\geq 2p(\Lambda(L)x, x). \end{aligned}$$

As a result we obtain that $(\forall |t| \leq \frac{3}{4}|x|L)$

$$(35) \quad \begin{aligned} |\mathbf{E} \exp \{it(\bar{X}_1, x)\}| &\leq 1 - t^2(\Lambda(L)x, x)/4 \\ &\leq \exp \{-t^2(\Lambda(L), x, x)/4\}. \end{aligned}$$

Further,

$$(36) \quad \begin{aligned} \mathbf{E} \exp \{it(Z_1, x)\} &= \exp \{-t^2\mathbf{E}(Z_1, x)^2/2\} \\ &\leq \exp \{-t^2\mathbf{E}(X_1(L), x)^2/2\} \\ &< \exp \{-t^2(\Lambda(L)x, x)/4\}. \end{aligned}$$

From (34)-(36) it follows that ($\forall |t| \leq 3\sqrt{n}/4|x|L, n \geq 4$)

$$(37) \quad |f_x(t)| \leq \exp \{-(\Lambda(L)x, x)^2 t^2 / 16\}.$$

Let us estimate

$$\mathbf{E}_k \equiv \mathbf{E}|X|^k |X'|^k |f_{X^s}(2t)|_{\mathcal{X}}(t, X, X', \omega),$$

setting $\omega = 3\sqrt{n}/16L$.

By (37),

$$\mathbf{E}_k \leq \mathbf{E}|X|^k |X'|^k \exp \{-t^2(\Lambda(L)X^s, X^s)^2 / 4\}.$$

Using Hölder's inequality now we have

$$(38) \quad \mathbf{E}_k \leq \mathbf{E}^{1/p} |X|^{kp} |X'|^{kq} \mathbf{E}^{1/q} \exp \{-4^{-1}qt^2(\Lambda(L)X^s, X^s)\}, \quad 1/p + 1/q = 1.$$

Further,

$$(39) \quad \mathbf{E} \exp \{-4^{-1}qt^2(\Lambda(L)X^s, X^s)\} \leq \mathbf{E} \exp \{-4^{-1}qt^2 \sum_1^7 \sigma_j^2(L) \zeta_j^2\},$$

where $\zeta_j = (X^s, e_j)$.

Let $\xi_j = (\bar{X}_1, e_j)$. Obviously,

$$\mathbf{E} \exp \left\{ i \sum_1^7 \xi_j \theta_j \right\} = \mathbf{E} \exp \{i(\bar{X}_1, \theta)\}, \text{ where } \theta = \sum_1^7 \theta_j e_j.$$

Hence by (35),

$$\left| \mathbf{E} \exp \left\{ i \sum_1^7 \xi_j \theta_j \right\} \right| < \exp \left\{ -\sum_1^7 \sigma_j^2(L) \theta_j^2 / 4 \right\}$$

for $(\sum_1^7 \theta_j^2)^{1/2} < \frac{3}{4}L$ and consequently for $\max_{1 \leq j \leq 7} |\theta_j| \leq \frac{1}{4}L$.

Applying now Lemma 7, we find that the density $p(u_1, u_2, \dots, u_7)$ of the distribution of the first seven coordinates of X satisfies the inequality

$$p(u_1, u_2, \dots, u_7) \ll \left(\prod_1^7 \sigma_j(L) \right)^{-1}.$$

Consequently,

$$(40) \quad p_s(u_1, u_2, \dots, u_7) \ll 1 / \prod_1^7 \sigma_j(L),$$

where $p_s(u_1, u_2, \dots, u_7)$ is the distribution density of the vector $\sum_1^7 \xi_j e_j$.

From (39) and (40) it follows that

$$(41) \quad \mathbf{E} \exp \{-4^{-1}qt^2(\Lambda(L)X^s, X^s)\} \ll 1/q^{7/2} |t|^7 \prod_1^7 \sigma_j^2(L).$$

Combining (32), (38) and (41), we get

$$(42) \quad \mathbf{E}_k \ll \sigma^{2k} \left(q^{7/2} |t|^7 \prod_1^7 \sigma_j^2(L) \right)^{-1/q}, \quad q > 19/18.$$

Estimates (31), (33) and (42) give us

$$|A(t)| \ll \beta_3 n^{-3/2} \frac{|t|/\sigma + \sum_{1 \leq j+k \leq 2} \sigma^{j+2k-3} |t|^{j+k} (\sigma^j + |a|^j)}{(q^{7/4} |t|^{7/2} \prod_1^7 \sigma_j(L))^{1/q} + 1}.$$

Similarly, using the estimate for the remainder in Lemma 2, we obtain

$$R_j(t) \ll \beta_3 n^{-3/2} \left(|t|^{3/2} + (\sigma + |a|) t^2 + |t|^3 \sum_{i+k=3} (\sigma^i + |a|^i) \sigma^k \right) / \left(\left(q^{7/4} |t|^{7/2} \prod_1^7 \sigma_i(L) \right)^{1/q} + 1 \right).$$

From the last two estimates, choosing $q < 8/7$, we obtain

$$(43) \quad \begin{aligned} \int_{-\infty}^{\infty} |A(t)/t| dt &\ll \beta_3 n^{-3/2} (\sigma^3 + |a|^3) \left(\prod_1^7 \sigma_j(L) \right)^{-6/7}, \\ \int_{-\infty}^{\infty} |R_j(t)/t| dt &\ll \beta_3 n^{-3/2} (\sigma^3 + |a|^3) \left(\prod_1^7 \sigma_j(L) \right)^{-6/7}. \end{aligned}$$

From (25), (26) and (43) it follows that

$$|Q_j(r, a, \varepsilon) - Q_{j+1}(r, a, \varepsilon)| \ll \beta_3 n^{-3/2} (\sigma^3 + |a|^3) \left(\prod_1^7 \sigma_j(L) \right)^{-6/7}.$$

Whence, by (24),

$$(44) \quad \Delta(\varepsilon) \ll \beta_3 \sigma^3 / \mathbf{E}|X|^3 \left(\prod_1^7 \sigma_j(L) \right)^{6/7} \sqrt{n}.$$

Let us estimate $\sigma^3 / \mathbf{E}|X|^3$. To this end we note that $\mathbf{E}|X|^3 \geq (\mathbf{E}|X|^2)^{3/2}$. Further, $\mathbf{E}|X|^2 > \mathbf{E}|\sum_{n/2+1}^n \bar{X}_i|^2 / n$, since $\mathbf{E}W = 0$.

On the other hand,

$$\begin{aligned} \mathbf{E} \left| \sum_{n/2+1}^n \bar{X}_i \right|^2 &\geq \mathbf{E} \left| \sum_{n/2+1}^n (\bar{X}_i - \mathbf{E}\bar{X}_i) \right|^2 + 4^{-1} n^2 |\mathbf{E}\bar{X}_1|^2 \\ &= 2^{-1} n \mathbf{E}|\bar{X}_1 - \mathbf{E}\bar{X}_1|^2 + 4^{-1} n^2 |\mathbf{E}\bar{X}_1|^2 \\ &= 2^{-1} n \mathbf{E}|\bar{X}_1|^2 + (n^2/4 - n/2) |\mathbf{E}\bar{X}_1|^2. \end{aligned}$$

It is easy to see that

$$\mathbf{E}|\bar{X}_1|^2 \geq \sigma^2 - \beta_3 / \sigma \sqrt{n}.$$

Without loss of generality, we may assume that $\beta_3 / \sigma^3 \sqrt{n} < 1/2$. Thus, $\mathbf{E}|\sum_{n/2+1}^n \bar{X}_i|^2 > n\sigma^2/4$ and consequently

$$(45) \quad \sigma^3 / \mathbf{E}|X|^3 < 8.$$

From Lemmas 1, 6 and inequalities (44), (45) it follows that

$$(46) \quad \begin{aligned} &\sup_{r, \alpha} (|(P_{n/2} - \Phi_{n/2}) * P_{n/2} * P_W(S(a, r))| / (\mathbf{E}|X|^3 + |a|^3)) \\ &\leq c\alpha^{-1} \left(\beta_3 / \left(\prod_1^7 \sigma_j(L) \right)^{6/7} \sqrt{n} + \varepsilon \mathbf{E}|X| / \mathbf{E}|X|^3 \sigma_1 \sigma_2 \right) \\ &\quad + 8\alpha^{-1} (1 - \alpha) \sup_{r, \alpha} |P_{n/2}(S(a, r)) - \Phi_{n/2}(S(a, r))| / (\mathbf{E}|X|^3 + |a|^3). \end{aligned}$$

Due to (32) and (45),

$$1/c(3) \leq (\sigma^3 + |a|^3)/(\mathbb{E}|X|^3 + |a|^3) \leq 8.$$

Hence (46) turns into the inequality

$$\begin{aligned} \rho_n \leq & c\alpha^{-1} \left(\beta_3 / \left(\prod_1^7 \sigma_j(L) \right)^{6/7} \sqrt{n} + \varepsilon / \sigma_1 \sigma_2 \sigma^2 \right) \\ & + 64c(3) \delta_n \alpha^{-1} (1 - \alpha), \end{aligned}$$

where

$$\delta_n = \sup_{r,a} (|P_{n/2}(S(a, r)) - \Phi_{n/2}(S(a, r))| / (\sigma^3 + |a|^3)).$$

Now let us choose ε so that $64c(3)\alpha^{-1}(1 - \alpha) = \frac{1}{16}$. It is not hard to see that for such a choice

$$(47) \quad \varepsilon \ll L/\sqrt{n}.$$

As a result we get

$$(48) \quad \rho_n \leq c\Gamma/\sqrt{n} + \delta_n/16,$$

where

$$\Gamma = \beta_3 / \left(\prod_1^7 \sigma_j(L) \right)^{6/7} + L/\sigma_1 \sigma_2 \sigma^2.$$

It is not hard to show that

$$\left| P_{n/2}(S(a, r)) - P \left(\sqrt{2/n} \sum_1^{n/2} X_j \in S(\alpha\sqrt{2}, r\sqrt{2}) \right) \right| < \sqrt{2}\beta_3/\sigma^3\sqrt{n}.$$

Consequently,

$$(49) \quad \delta_n 2^{3/2} \Delta_{n/2} + \sqrt{2}\beta_3/\sigma^6\sqrt{n},$$

where $\Delta_n = \sup_a (\Delta_n(a)/(\sigma^3 + |a|^3))$.

By (48) and (49),

$$(50) \quad \rho_n \leq c\Gamma/\sqrt{n} + \Delta_{n/2}/4.$$

Similarly,

$$(51) \quad \rho'_n \equiv \sup_{r,a} (|(P_{n/2} - \Phi_{n/2}) * \Phi_{n/2} * P_W(S(a, r))| / (\sigma^3 + |a|^3)) \leq c\Gamma/\sqrt{n} + \Delta_{n/2}/4.$$

By Lemmas 4 and 6,

$$\begin{aligned} & \sup_r |P_n(S(a, r)) - \Phi_n(S(a, r))| \\ & \ll \sup_r |(P_n - \Phi_n) * P_W(S(a, r))| + L(\sigma + |a|)/\sigma_1 \sigma_2 \sqrt{n}. \end{aligned}$$

This means in view of (23) that

$$(52) \quad \Delta'_n \ll \rho_n + \rho'_n + L/\sigma^2 \sigma_1 \sigma_2 \sqrt{n},$$

where

$$\Delta'_n = \sup_{r,a} (|P_n(S(a, r)) - \Phi_n(S(a, r))|) / (\sigma^3 + |a|^3).$$

On the other hand,

$$\sup_r |P_n(S(a, r)) - \Phi_n(S(a, r))| > \Delta_n(a) - \beta_3 / \sigma^3 \sqrt{n}.$$

Therefore,

$$(53) \quad \Delta'_n \geq \Delta_n - \beta_3 / \sigma^6 \sqrt{n}.$$

From (50)-(53) it follows that $\Delta_n < c\Gamma/\sqrt{n} + \Delta_{n/2}/2$.

Hence, for $n = 2^k$,

$$(54) \quad \Delta_n \ll \Gamma/\sqrt{n}.$$

Let $2^k < n < 2^{k+1}$. Denoting by Ψ_j the distribution of the r.v. $\sum_1^j X_m/\sqrt{n}$, we have

$$(55) \quad \begin{aligned} \Psi_n(S(a, r)) - \Phi_n(S(a, r)) &= (\Psi_{2^k} - \Phi_{2^k}) * \Psi_{n-2^k}(S(a, r)) \\ &+ (\Psi_{n-2^k} - \Phi_{n-2^k}) * \Phi_{2^k}(S(a, r)) = K_1 + K_2. \end{aligned}$$

It is not hard to see that

$$\left| \int_H (\Psi_{2^k} - \Phi_{2^k})(S(a, r) - x) \Psi_{n-2^k}(dx) \right| \ll \Delta_{2^k} \int_H (\sigma^3 + |x|^3 + |a|^3) \Psi_{n-2^k}(dx).$$

Using the moment inequality from [19], we obtain

$$\int_H |x|^3 \Psi_{n-2^k}(dx) = n^{-3/2} \mathbf{E} \left| \sum_1^{n-2^k} X_j \right|^3 \ll \sigma^3 + \beta_3/\sqrt{n}.$$

Since without loss of generality $\beta_3/\sqrt{n} < \sigma^3$, we conclude in view of the two preceding inequalities that

$$(56) \quad K_1 \ll \Delta_{2^k}(\sigma^3 + |a|^3).$$

It remains to estimate K_2 . We shall avail ourselves of the representation

$$\begin{aligned} &\mathbf{E} \exp \left\{ it \left| n^{-1/2} \left(\sum_1^{2^k} Z_m + \sum_{2^{k+1}}^m X_m \right) + a \right|^2 \right\} - \mathbf{E} \exp \left\{ it \left| n^{-1/2} \sum_1^n Z_m + a \right|^2 \right\} \\ &= \sum_{2^k}^{n-1} \left(\mathbf{E} \exp \left\{ it \left| n^{-1/2} \left(\sum_1^j Z_m + \sum_{j+1}^n X_m \right) + a \right|^2 \right\} \right. \\ &\quad \left. - \mathbf{E} \exp \left\{ it \left| n^{-1/2} \left(\sum_1^{j+1} Z_m + \sum_{j+2}^n X_m \right) + a \right|^2 \right\} \right) \equiv \sum_{2^k}^{n-1} \gamma_j(t). \end{aligned}$$

Now we apply Lemma 2, setting

$$\begin{aligned} X &= n^{-1/2} \sum_1^{2^{k-1}} Z_m, & Y &= n^{-1/2} \left(\sum_{2^{k-1}+1}^j Z_m + \sum_{j+2}^n X_m \right) + a, \\ V &= X_{j+1}/\sqrt{n}, \text{ or } Z'_{j+1}/\sqrt{n}, & \eta &= 0 \end{aligned}$$

(the latter implies that $\omega = \infty$, i.e., $\chi(t, \omega, X, X') \equiv 1$).

Due to the normality of Z_m we have

$$|f_x(t)| \leq \exp \{-t^2(\Lambda x, x)/8\} \leq \exp \{-t^2(\Lambda(L)x, x)/8\} \\ \leq \exp \left\{ -t^2 \sum_1^7 x_i^2 \sigma_i^2(L)/8 \right\},$$

where $x_i = (x, e_i)$.

On the other hand, $|X|$ has moments of all orders and $E|X|^t \leq c(t)\sigma^t$. Now repeating the arguments which led us to (43), we see that

$$\int_{-\infty}^{\infty} |\gamma_j(t)/t| dt \ll \beta_3(\sigma^3 + |a|^3) / \left(\prod_1^7 \sigma_j(L) \right)^{6/7} n^{3/2}.$$

Hence, by the inversion formula,

$$(57) \quad K_2 \ll \beta_3(\sigma^3 + |a|^3) / \left(\prod_1^7 \sigma_j(L) \right)^{6/7} n^{1/2}.$$

The assertion of the theorem follows from (54)-(57).

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