

ABELIAN THEOREMS FOR A CLASS OF PROBABILITY DISTRIBUTIONS IN R^d AND THEIR APPLICATION

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A class of multidimensional absolutely continuous distributions is considered. Each of them has a moment-generating function that is finite in a bounded set S and, therefore, generates a family of so-called conjugate or associated distributions. At the focus of our attention are the limiting distributions for this family that appear as the conjugating parameter tends to the boundary of S . As in the one-dimensional case, each such limiting distribution can be obtained as a consequence of an Abelian theorem.

1. Introduction

Let P be a probability measure defined on the Borel sets of R^d , $d > 1$, and $f(s)$ be its moment-generating function, that is,

$$f(s) = \int_{R^d} e^{(s,x)} P(dx).$$

By $\langle \cdot, \cdot \rangle$ we denote the inner product. Suppose that the set

$$S = \{s \in R^d: f(s) < \infty\}$$

is not empty and its dimensionality equals d .

The moment-generating function plays a role of great importance in the large-deviation theory. Its basic properties are discussed in [1–5]. The present paper aims to make a contribution toward the further development of this theory. At the focus of our attention is the case when S is, being always convex, bounded.

Let S_0 be the interior of S . If S is bounded and $0 \in S_0$, then S_0 can be represented as

$$S_0 = \{s: s = te, 0 \leq t < h(e), e \in S^{d-1}\}.$$

It is convenient to call $h(e)$ the *shape function* of S or simply the *shape* of S .

Obviously, for $0 < t < h(e)$, $u > 0$ the Markov inequality holds, that is,

$$P(x: \langle e, x \rangle \geq u) \leq f(te)e^{-tu}. \quad (1.1)$$

In what follows, we assume that P is absolutely continuous. Denote its density by $p(x)$.

Further, assume that

$$p(x) = b(x)e^{-|x|a(e_x)}, \quad (1.2)$$

where $e_x = |x|^{-1}x$ and

$$0 < \inf_{e \in S^{d-1}} a(e) \leq \sup_{e \in S^{d-1}} a(e) < \infty.$$

If $p(x)$ is of the form (1.2) and $b(x)$ does not grow too fast as $|x| \rightarrow \infty$, then $f(s)$ is finite for some S with $0 \in S_0$. Intuitively, it is $a(e)$ that determines the shape of S . The following proposition justifies this conjecture.

PROPOSITION 1.1. Assume that in (1.2)

$$c_-(1 + |x|)^{-\beta} \leq b(x) \leq c_+(1 + |x|)^\beta, \quad \beta > 0, \quad c_\pm > 0. \quad (1.3)$$

Then

1°.

$$h(e) = \inf_{(\varepsilon \in S^{d-1}: \langle e, \varepsilon \rangle > 0)} \frac{a(\varepsilon)}{\langle e, \varepsilon \rangle}.$$

2°. For the shape function $h(e)$ of any bounded open convex set S_0 that contains 0, there exists $p(x)$ of the form (1.2) such that the interior of $S = \{s: f(s) < \infty\}$ is S_0 . As $a(e)$ in (1.2) one may take

$$a(e) = \sup_{(\varepsilon \in S^{d-1}: \langle e, \varepsilon \rangle > 0)} h(\varepsilon)\langle e, \varepsilon \rangle.$$

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The question arises: *What can be said about the asymptotic behavior of $f(s)$ as $s \rightarrow \partial S$?* The answer requires additional restrictions imposed on both $a(e)$ and $b(x)$ in (1.1). Our goal is to establish a multidimensional analog of the following fact.

Let $d = 1$ and

$$p(x) = e^{-s_+ x} r_\alpha(x), \quad (1.4)$$

where $s_+ > 0$ and $r_\alpha(x)$ is of regular variation as $x \rightarrow \infty$ with the exponent $\alpha > -1$. When $\tau \downarrow 0$,

$$f(s_+ - \tau) \sim \Gamma(1 + \alpha) \tau^{-1} r_\alpha(\tau^{-1}). \quad (1.5)$$

This is one of the simplest forms of the so-called Abelian theorem (see, e.g., [12]).

First, we need a relevant multidimensional analog of (1.4) and (1.5). Having it in mind, we introduce the following notion of regular variation that, in essence, coincides with that given in [11, Sec. 5.4.2].

Let $\lambda(e)$ be a nonnegative function defined on S^{d-1} .

Definition 1.2. We say that $b(x)$, $x \in R^d$, is the function of (α, λ) -regular variation in the cone $C_\lambda = \{x \in R^d : \lambda(e_x) > 0\}$ if

$$b(x) = r_\alpha(|x|)(\lambda(e_x) + u(x)), \quad (1.6)$$

where $r_\alpha(t)$ is of regular, in Karamata's sense, variation as $t \rightarrow \infty$ with the exponent α while

$$\lim_{|x| \rightarrow \infty} \sup_{e_x \in C_\lambda} |u(x)| = 0.$$

Denote $\Delta(\varepsilon) = a(\varepsilon) - h(e)\langle e, \varepsilon \rangle$. We need the following assumptions:

- (A) For a given direction e , the set $\arg \min_{\langle e, \varepsilon \rangle > 0} a(\varepsilon)/\langle e, \varepsilon \rangle$ consists of a single point $\varepsilon' = \varepsilon'(e)$.
- (B) $\Delta(\varepsilon)$ in a neighborhood of ε' admits the representation

$$\Delta(\varepsilon) = \frac{1}{2}(\varepsilon - \varepsilon')^T \Lambda (\varepsilon - \varepsilon') + o(|\varepsilon - \varepsilon'|^2).$$

Here Λ is a nonnegative definite matrix, and its rank equals $d - 1$. Furthermore, $\Lambda \varepsilon' = 0$.

- (C) For all sufficiently small δ ,

$$\inf_{|\varepsilon - \varepsilon'| > \delta} \Delta(\varepsilon) = c(\delta) > 0.$$

By λ_j , $j = 1, \dots, d - 1$, we denote the nonzero eigenvalues of Λ .

Consider the class of densities of the form (1.2), where $b(x)$ is of (α, λ) -regular variation in $(x : |e_x - e| < \delta)$, while in $(x : |e_x - e| \geq \delta)$ we have

$$b(x) \leq (1 + |x|)^\beta$$

for some $\beta > 0$.

THEOREM 1.3. Let $p(x)$ be of the form (1.2) with $\alpha > -(d + 1)/2$. Assume that $\lambda(e)$ is continuous for $|e - \varepsilon'| < \delta$ and (A), (B), and (C) hold. Then as $\tau \downarrow 0$ (cf. (1.5)),

$$f((h(e) - \tau)e) = c_\alpha g(e) \tau^{-(d+1)/2} r_\alpha(\tau^{-1})(1 + o(1)),$$

where

$$c_\alpha = \Gamma\left(\alpha + \frac{d+1}{2}\right) (2\pi)^{(d-1)/2}$$

and

$$g(e) = \lambda(\varepsilon'(e)) \langle \varepsilon'(e), e \rangle^{-\alpha - (d+1)/2} (\lambda_1 \dots \lambda_{d-1})^{-1/2}.$$

The question arises: *How does $f(s)$ behave as s approaches ∂S alongside some other direction?* It turns out that there exists a cone of admissible directions in which the form of the Abelian theorem is, in essence, preserved.

THEOREM 1.4. If the conditions of Theorem 1.3 hold, then, for any arbitrarily small $\eta > 0$ and $\tau \downarrow 0$,

$$f(h(e)e - \tau \hat{e}) = c_\alpha g(e, \hat{e}) \tau^{-(d+1)/2} r_\alpha(\tau^{-1})(1 + o(1))$$

uniformly in \hat{e} , $\langle \varepsilon', \hat{e} \rangle \geq \eta$. Here

$$g(e, \hat{e}) = \lambda(\varepsilon'(e)) \langle \varepsilon'(e), \hat{e} \rangle^{-\alpha - (d+1)/2} (\lambda_1 \dots \lambda_{d-1})^{-1/2}.$$

Consider the measures $P_s(A)$, $s \in S$, defined as

$$P_s(A) = \frac{\int_{R^d} \chi_A(x) e^{(s, x)} p(x) dx}{f(s)}.$$

These measures are called *conjugate* or *associated* to P . They proved to be useful in the large-deviation theory. When one deals with large deviations of arbitrarily large order it is required to learn much about the asymptotic properties of P_s as $s \rightarrow \partial S$ (see, e.g., [8–10]). The following theorem is devoted to such a case. Before stating it, we consider an orthogonal matrix C that reduces Λ to a diagonal matrix that is $C^T \Lambda C = \Lambda_0$, where

$$\Lambda_0 = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 & 0 \\ 0 & \lambda_2 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & \lambda_{d-1} & 0 \\ 0 & 0 & \dots & 0 & 0 \end{pmatrix}.$$

Set

$$\bar{\Lambda}_0 = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_{d-1} \end{pmatrix}$$

and $\bar{x} = (x_1, \dots, x_{d-1})$ for $x = (x_1, \dots, x_d) \in R^d$.

THEOREM 1.5. Assume that the conditions of Theorem 1.3 hold and $s = (h(e) - \tau)e$, $\tau \rightarrow 0$. Then $(\tau|\xi|, \tau^{-1/2} C^T(e_\xi - e'))$ converges in P_s -distribution to a random vector (ρ, ζ) such that $P(\zeta_d = 0) = 1$ and

$$P(\rho \in A; \bar{\zeta} \in B) = \int_A q_\alpha(r) \left(\int_B r^{(d-1)/2} \varphi_{\bar{\Lambda}_0}(z r^{1/2}) dz \right) dr,$$

where

$$q_\alpha(r) = \frac{r^{\alpha+(d-1)/2} \langle e', e \rangle^{\alpha+(d+1)/2} e^{-r \langle e', e \rangle}}{\Gamma(\alpha + (d+1)/2)}, \quad r > 0,$$

while

$$\varphi_{\bar{\Lambda}_0}(z) = (2\pi)^{-(d-1)/2} (\lambda_1 \dots \lambda_{d-1})^{1/2} \exp\left(-\frac{1}{2} z^T \bar{\Lambda}_0 z\right), \quad z \in R^{d-1}.$$

Thus $|\xi|$ as a limiting, with respect to P_s , distribution has the gamma distribution, while the limiting distribution for $e_\xi - e'$ is a mixture of normal distributions. That mixture can also be represented in the following form:

$$P(\bar{\zeta} \in B) = (2\pi)^{-(d-1)/2} \frac{\Gamma(\alpha + d)}{\Gamma(\alpha + (d+1)/2)} (\lambda_1 \dots \lambda_{d-1})^{1/2} \int_B \left(1 + \frac{1}{2} z^T \bar{\Lambda}_0 z\right)^{-\alpha-d} dz.$$

It is a nonstandard multidimensional Student distribution (see, e.g., [7, p. 134]).

The paper is organized as follows. Sections 2 and 3 contain the proofs of Proposition 1.1 and Theorem 1.3, respectively. The proofs of Theorems 1.4 and 1.5 are sketched in Sec. 4.

2. Proof of Proposition 1.1

From now on, c denotes any positive constant whose concrete value is of no importance. This means that $c + c = c$, $c^2 = c$, etc. By $\omega(t)$, we denote any nonnegative function such that $\lim_{t \rightarrow \infty} \omega(t) = 0$ while θ varies within $[-1, 1]$.

Set for $e \in S^{d-1}$

$$P^{(e)}(u) = P(\langle \xi, e \rangle \geq u), \quad u \in R^1.$$

We should show that

$$\lim_{u \rightarrow \infty} \frac{-\log P^{(e)}(u)}{u} = h(e) \quad (2.7)$$

as $u \rightarrow \infty$ (see (1.1)).

From (1.2) and (1.3) it follows that for all sufficiently large u

$$P_- = \frac{c_-}{2} \int_{|x| \langle e, e_x \rangle \geq u} |x|^{-\beta} e^{-|x|a(e_x)} dx \leq P^{(e)}(u) \leq 2c_+ \int_{|x| \langle e, e_x \rangle \geq u} |x|^\beta e^{-|x|a(e_x)} dx = P_+.$$

The change of variables

$$x_i = r\varepsilon_i, \quad i = 1, \dots, d-1, \quad x_d = \text{sign}(x_d)r(1 - \varepsilon_1^2 - \dots - \varepsilon_{d-1}^2)^{1/2},$$

having Jacobian $r^{d-1}|\varepsilon_d|^{-1}$ with $|\varepsilon_d| = (1 - \varepsilon_1^2 - \dots - \varepsilon_{d-1}^2)^{1/2}$, leads to

$$P_\pm = c \int_{r(\varepsilon, e) \geq u} r^{\pm\beta+d-1} e^{-ra(\varepsilon)} dr \chi_{d-1}(d\varepsilon),$$

where $\varepsilon = (\varepsilon_1, \dots, \varepsilon_d)^T$, $|\varepsilon| = 1$. From now on, χ_{d-1} is the normalized Haar measure on S^{d-1} .

Since for any $c > 0$ and β as $z \rightarrow \infty$

$$\int_z^\infty r^\beta e^{-cr} dr \sim c^{-1} z^\beta e^{-cz},$$

we have as $u \rightarrow \infty$

$$P_\pm \sim cu^{\pm\beta+d-1} \int_{\langle \varepsilon, e \rangle > 0} \frac{e^{-ua(\varepsilon)/\langle \varepsilon, e \rangle}}{a(\varepsilon)\langle \varepsilon, e \rangle^{\pm\beta+d-1}} \chi_{d-1}(d\varepsilon). \quad (2.8)$$

It is readily seen that both integrals converge. Denote them, respectively, by I_\pm .

Let $\delta' > 0$ be arbitrarily small. Set

$$E' = \left(\varepsilon \in S^{d-1} : h(e) \leq \frac{a(\varepsilon)}{\langle \varepsilon, e \rangle} \leq h(e) + \delta' \right).$$

Obviously,

$$I_- \geq e^{-u(h(e)+\delta')} \int_{E'} a(\varepsilon)^{-1} \langle \varepsilon, e \rangle^{\beta-d+1} \chi_{d-1}(d\varepsilon). \quad (2.9)$$

Estimate I_+ as

$$I_+ \leq \int_{E'} + \int_{\langle \varepsilon, e \rangle > 0, \varepsilon \notin E'} = I_1 + I_2.$$

It is clear that

$$I_1 \leq e^{-uh(e)} \int_{E'} a(\varepsilon)^{-1} \langle \varepsilon, e \rangle^{-\beta-d+1} \chi_{d-1}(d\varepsilon) = ce^{-uh(e)}.$$

Since for $\varepsilon \notin E'$

$$\inf_{\varepsilon \notin E'} (a(\varepsilon) - h(e)\langle \varepsilon, e \rangle) \geq c \inf_{\varepsilon \notin E'} (1 - h(e)\langle \varepsilon, e \rangle/a(\varepsilon)) \geq c \left(1 - \frac{h(e)}{h(e) + \delta'} \right),$$

we obtain for all sufficiently large u

$$I_2 \leq e^{-uh(e)} \int_{\langle \varepsilon, e \rangle > 0, \varepsilon \notin E'} a(\varepsilon)^{-1} \langle \varepsilon, e \rangle^{-\beta-d+1} \exp(-(\langle \varepsilon, e \rangle)^{-1}) \chi_{d-1}(d\varepsilon) < ce^{-uh(e)} \sup_{t>0} e^{-t} t^{\beta+d-1} = ce^{-uh(e)}.$$

Thus

$$I_+ < ce^{-uh(e)}. \quad (2.10)$$

Uniting (2.8)–(2.10) yields

$$cu^{-\beta+d-1} e^{-uh(e)} \leq P^{(e)}(u) \leq cu^{\beta+d-1} e^{-uh(e)},$$

whence (2.7) follows immediately. The first statement of Proposition 1.1 is proved.

Let S be a bounded open convex set. Consider its support function

$$\sigma(x) = \sup_{y \in S} \langle x, y \rangle, \quad x \in R^d,$$

and the so-called Minkowski function

$$m(x) = \inf\{t: t > 0, t^{-1}x \in S\}, \quad x \in R^d.$$

It is well known that (see, e.g., [12])

$$m(x) = \sup_{\langle x, y \rangle > 0} \frac{\langle x, y \rangle}{\sigma(y)}$$

and, therefore,

$$m(e) = \sup_{\langle e, \varepsilon \rangle > 0} \frac{\langle e, \varepsilon \rangle}{\sigma(\varepsilon)}, \quad e \in S_{d-1}.$$

Since $m(e) = 1/h(e)$, it follows that

$$h(e) = \inf_{\langle e, \varepsilon \rangle > 0} \frac{\sigma(\varepsilon)}{\langle e, \varepsilon \rangle}.$$

In other words, in order to come to the form $h(e)$, we should choose $a(e) = \sigma(e)$ in (1.2). Thus, the second statement is also proved.

3. Proof of Theorem 1.3

Assume for the time being that $\varepsilon' = \varepsilon'(e) = (0, \dots, 0, 1)^T$ and $\Lambda = \Lambda_0$.

Set for brevity $s = (h(e) - \tau)e$, and let

$$X_1 = (x: |x| > L, |e_x - \varepsilon'| < M\tau^{1/2}), \quad X_2 = (x: |x| > L, M\tau^{1/2} \leq |e_x - \varepsilon'| < \delta'),$$

$$X_3 = (x: |x| > L, |e_x - \varepsilon'| \geq \delta'), \quad X_4 = (x: |x| \leq L),$$

where $L > 0$ and $M > 0$ are arbitrarily large, while $0 < \delta' < \delta$ is arbitrarily small. Obviously,

$$f(s) = \sum_{i=1}^4 f_i(s) \tag{3.11}$$

with

$$f_i(s) = \int_{X_i} e^{\langle s, x \rangle} p(x) dx.$$

It is worth recalling that in $X_1 \cup X_2$ the function $b(x)$ admits representation (1.6). Consider the function $r_\alpha(t)$. It is of regular variation and, therefore, $r_\alpha(t) = t^\alpha l(t)$, where $l(t)$ slowly varies as $t \rightarrow \infty$. From the well-known Karamata representation (see, e.g., [6, Chap. VIII]) it follows that for any $\eta > 0$ there exists $L > 0$ such that for $\min(\rho, r) > L$

$$\frac{1}{2} \min((r/\rho)^\eta, (r/\rho)^{-\eta}) \leq \frac{l(r)}{l(\rho)} \leq 2 \max((r/\rho)^\eta, (r/\rho)^{-\eta}). \tag{3.12}$$

Let us estimate $f_i(s)$ in (3.11) one after another. Represent X_1 as follows:

$$X_1 = X_{11} \cup X_{12} \cup X_{13}, \tag{3.13}$$

where

$$X_{11} = X_1 \cap (x: |x| < N^{-1}\tau^{-1}),$$

$$X_{12} = X_1 \cap (x: N^{-1}\tau^{-1} \leq |x| < N\tau^{-1}),$$

$$X_{13} = X_1 \cap (x: |x| \geq N\tau^{-1}),$$

and $N > 0$ is arbitrarily large. Denote

$$f_{ik}(s) = \int_{X_{ik}} e^{(s,x)} p(x) dx. \quad (3.14)$$

Set $\bar{\varepsilon} = (\varepsilon_1, \dots, \varepsilon_{d-1})^T$. If $x \in X_1$, then for $\varepsilon = e_x$ we have $|\bar{\varepsilon}| = O(\tau^{1/2})$ and $|\varepsilon_d - 1| = O(\tau)$. Furthermore,

$$(\varepsilon - \varepsilon')^T \Lambda_0 (\varepsilon - \varepsilon') = \bar{\varepsilon}^T \bar{\Lambda}_0 \bar{\varepsilon}, \quad (3.15)$$

where as above $\bar{\Lambda}_0$ is the $(d-1) \times (d-1)$ diagonal matrix with the diagonal $(\lambda_1, \dots, \lambda_{d-1})$.

In view of (3.13) and (3.14), we have

$$f_{12}(s) = \lambda(\varepsilon') l(1/\tau) \int_{X_{12}} |x|^\alpha \exp(-|x|(\Delta(\varepsilon) + \tau \langle e, \varepsilon \rangle)) dx (1 + o(1)).$$

From (B) and (3.15), taking into account that $\varepsilon_d = 1 + O(\tau)$, we easily obtain

$$f_{12}(s) = \lambda(\varepsilon') l(1/\tau) \int_{N^{-1} < r\tau < N} \int_{|\bar{\varepsilon}| < M\tau^{1/2}} r^{\alpha+d-1} \exp\left(-\frac{r}{2} \bar{\varepsilon}^T \bar{\Lambda}_0 \bar{\varepsilon} - r\tau e_d\right) dr d\bar{\varepsilon} (1 + o(1)).$$

Making the change of variables $r\tau = u$, $\bar{z} = \tau^{-1/2} \bar{\varepsilon}$ yields

$$f_{12}(s) = \tau^{-\alpha-(d+1)/2} l(1/\tau) (c_\alpha g(e) + \theta \omega(\min(M, N))), \quad (3.16)$$

where

$$c_\alpha = \Gamma\left(\alpha + \frac{d+1}{2}\right) (2\pi)^{(d-1)/2}$$

and

$$g(e) = \lambda(\varepsilon') \langle \varepsilon', e \rangle^{-\alpha-(d+1)/2} (\det \bar{\Lambda}_0)^{-1/2} = \lambda(\varepsilon') \langle \varepsilon', e \rangle^{-\alpha-(d+1)/2} (\lambda_1 \dots \lambda_{d-1})^{-1/2}.$$

From (B), taking into account that $|\langle e, \varepsilon \rangle - e_d| = O(\tau^{1/2})$, we obtain

$$f_{11}(s) \leq 2 \sup_{\varepsilon} \lambda(\varepsilon) \int_{X_{11}} r_\alpha(|x|) \exp(-|x|(\Delta(\varepsilon) + \tau \langle e, \varepsilon \rangle)) dx \leq c \int_L^{N^{-1}\tau^{-1}} r^{\alpha+d-1} l(r) \left(\int_{|\bar{\varepsilon}| < M\tau^{1/2}} e^{-cr|\bar{\varepsilon}|^2} d\bar{\varepsilon} \right) dr.$$

In view of (3.12), we continue

$$f_{11}(s) \leq c\tau^{-\alpha-(d+1)/2} \int_0^{N^{-1}} r^{\alpha+(d-1)/2} l(r/\tau) dr \leq c\tau^{-\alpha-(d+1)/2} l(1/\tau) \int_0^{N^{-1}} r^{\alpha+(d-1)/2-\eta} dr \leq c\tau^{-\alpha-(d+1)/2} l(1/\tau) \omega(N)$$

provided that $0 < \eta < \min(1, \alpha + (d+1)/2)$. Similarly,

$$f_{13}(s) \leq 2 \sup_{\varepsilon} \lambda(\varepsilon) \int_{X_{13}} r_\alpha(|x|) \exp(-|x|(\Delta(\varepsilon) + \tau \langle e, \varepsilon \rangle)) dx \leq c \int_{N/\tau}^{\infty} r^{\alpha+d-1} l(r) e^{-cr\tau} \left(\int_{|\bar{\varepsilon}| < M\tau^{1/2}} e^{-cr|\bar{\varepsilon}|^2} d\bar{\varepsilon} \right) dr.$$

Taking advantage of (3.12), we get

$$\begin{aligned} f_{13}(s) &\leq c\tau^{-\alpha-(d+1)/2} \int_N^{\infty} r^{\alpha+(d-1)/2} l(r/\tau) e^{-cr} dr \\ &\leq c\tau^{-\alpha-(d+1)/2} l(1/\tau) \int_N^{\infty} r^{\alpha+(d-1)/2+\eta} e^{-cr} dr \leq c\tau^{-\alpha-(d+1)/2} l(1/\tau) \omega(N). \end{aligned}$$

Thus,

$$f_1(s) = \tau^{-\alpha-(d+1)/2} l(1/\tau) (c_\alpha g(e) + \theta \omega(\min(M, N))). \quad (3.17)$$

Before estimating $f_2(s)$, note that for $x \in X_2$ we have $|\bar{\varepsilon}| < \delta'$ and $|\varepsilon_d - 1| < c\delta'^2$. Moreover,

$$(\varepsilon - \varepsilon')^T \Lambda_0 (\varepsilon - \varepsilon') \geq c|\bar{\varepsilon}|^2.$$

That is why

$$f_2(s) \leq 2 \sup_{\varepsilon} \lambda(\varepsilon) \int_{X_2} r_\alpha(|x|) e^{-c|x||\bar{\varepsilon}|^2 - c|x|\tau} dx \leq c \int_L^\infty r^{\alpha+d-1} l(r) e^{-cr\tau} \left(\int_{|\bar{\varepsilon}| \geq (1/2)M\tau^{1/2}} e^{-cr|\bar{\varepsilon}|^2} d\bar{\varepsilon} \right) dr.$$

Represent

$$\int_L^\infty r^{\alpha+d-1} l(r) e^{-cr\tau} \left(\int_{|\bar{\varepsilon}| \geq (1/2)M\tau^{1/2}} e^{-cr|\bar{\varepsilon}|^2} d\bar{\varepsilon} \right) dr = \int_L^{N^{-1}\tau^{-1}} + \int_{N^{-1}\tau^{-1}}^\infty.$$

The first integral on the right-hand side is estimated as $f_{11}(s)$. As to the second one, it is easily seen that

$$\begin{aligned} & \int_{N^{-1}\tau^{-1}}^\infty r^{\alpha+d-1} l(r) e^{-cr\tau} \left(\int_{|\bar{\varepsilon}| \geq (1/2)M\tau^{1/2}} e^{-cr|\bar{\varepsilon}|^2} d\bar{\varepsilon} \right) dr \\ & \leq \tau^{-\alpha-(d+1)/2} \int_{N^{-1}}^\infty r^{\alpha+d-1} l(r/\tau) e^{-cr} \left(\int_{|\bar{\varepsilon}| \geq (1/2)M} e^{-(c/N)|\bar{\varepsilon}|^2} d\bar{\varepsilon} \right) dr. \end{aligned}$$

Obviously,

$$\int_{N^{-1}}^\infty r^{\alpha+d-1} l(r/\tau) e^{-cr} dr \sim l(1/\tau),$$

while

$$\int_{|\bar{\varepsilon}| \geq (1/2)M} e^{-(c/N)|\bar{\varepsilon}|^2} d\bar{\varepsilon} = M^{d-1} \omega(M^2/N).$$

Therefore,

$$f_2(s) = \tau^{-\alpha-(d+1)/2} l(1/\tau) \theta M^{d-1} \omega(M^2/N). \quad (3.18)$$

From (C) it follows that

$$f_3(s) \leq \int_{R^d} b(x) e^{-|x|(c(\delta) + \tau(e, \varepsilon))} dx \leq \int_{R^d} |x|^{\beta_1} e^{-c|x|} dx, \quad \beta_1 > \beta,$$

provided τ is sufficiently small. Therefore,

$$f_3(s) = O(1). \quad (3.19)$$

Finally,

$$f_4(s) = O(1). \quad (3.20)$$

Since L , M , N , and δ' are arbitrary from (3.11) and (3.17)–(3.20), it follows that

$$f(s) \sim c_\alpha g(e) \tau^{-\alpha-(d+1)/2} l(1/\tau). \quad (3.21)$$

Let us turn to the general case. Denote by C an orthogonal matrix such that $C^T \Lambda C = \Lambda_0$. It is obvious that $\varepsilon'(e) = C\varepsilon_0$ with $\varepsilon_0 = (0, \dots, 0, 1)^T$. Set for $0 < \tau < h(e)$, $s = (h(e) - \tau)e$

$$f(s) = \int_{R^d} e^{(s, Cx)} p(Cx) dx.$$

From (B) it follows that

$$\Delta(C\varepsilon) = \frac{1}{2}(\varepsilon - \varepsilon_0)^T \Lambda_0 (\varepsilon - \varepsilon_0) + o(|\varepsilon - \varepsilon_0|^2).$$

It remains to apply (3.21). Theorem is proved.

4. On the Proofs of Theorems 1.4 and 1.5

Here we give a sketch of the proofs of these statements.

Proof of Theorem 1.4. Let $s = h(e)e - \tau \hat{e}$. Then

$$\langle s, x \rangle - |x|a(\varepsilon) = -|x|(\Delta(\varepsilon) + \tau \langle \hat{e}, \varepsilon \rangle).$$

Suppose that $\varepsilon' = (0, \dots, 0, 1)^T$ and $\Lambda = \Lambda_0$. If $x \in X_{12}$, then

$$\langle \hat{e}, \varepsilon \rangle = \langle \hat{e}, \varepsilon' \rangle + O(\tau^{1/2}) = \hat{e}_d + O(\tau^{1/2}).$$

Therefore, $f_{12}(s)$ acquires the form

$$f_{12}(s) = \lambda(\varepsilon') l(1/\tau) \int_{N^{-1} < r\tau < N} \int_{|\bar{\varepsilon}| < M\tau^{1/2}} r^{\alpha+d-1} \exp\left(-\frac{r}{2} \bar{\varepsilon}^T \bar{\Lambda}_0 \bar{\varepsilon} - \hat{e}_d r \tau\right) dr d\bar{\varepsilon} (1 + o(1)),$$

and instead of (3.16) we obtain

$$f_{12}(s) = \tau^{-\alpha-(d+1)/2} l(1/\tau) (c_\alpha g(e, \hat{e}) + \theta \omega(\min(M, N))) \quad (4.22)$$

uniformly in $\hat{e}_d \geq \eta > 0$. The rest of the proof needs, in essence, no alteration. So, we get

$$f(s) = c_\alpha g(e, \hat{e}) \tau^{-\alpha-(d+1)/2} l(1/\tau) (1 + o(1)).$$

If $\varepsilon' \neq (0, \dots, 0, 1)^T$, we should argue as in the proof of Theorem 1.3.

Proof of Theorem 1.5. As in the proof of Theorem 1.3, consider, first, the simplest case $\varepsilon' = (0, \dots, 0, 1)^T$, $\Lambda = \Lambda_0$. Let $t > 0$, $r_2 > r_1 > 0$ be arbitrary and

$$X_\tau = \{x: r_1 < \tau|x| < r_2, |e_x - \varepsilon'| < t\tau^{1/2}\}.$$

Repeating the argument that led to (3.21), one easily obtains

$$\int_{X_\tau} e^{\langle s, x \rangle} p(x) dx = \lambda(\varepsilon') \tau^{-\alpha-(d+1)/2} l(1/\tau) \int_{r_1}^{r_2} r^{\alpha+d-1} e^{-re_d} \left(\int_{|\bar{\varepsilon}| < t} e^{-(r/2) \bar{\varepsilon}^T \bar{\Lambda}_0 \bar{\varepsilon}} d\bar{\varepsilon} \right) dr (1 + o(1))$$

or

$$\begin{aligned} \int_{X_\tau} e^{\langle s, x \rangle} p(x) dx &= (2\pi)^{(d-1)/2} (\lambda_1 \dots \lambda_{d-1})^{-1/2} \lambda(\varepsilon') \tau^{-\alpha-(d+1)/2} l(1/\tau) \\ &\times \int_{r_1}^{r_2} r^{\alpha+(d-1)/2} e^{-re_d} \left(\int_{|\bar{\varepsilon}| < t} r^{(d-1)/2} \varphi_{\bar{\Lambda}_0}(\bar{\varepsilon} r^{1/2}) d\bar{\varepsilon} \right) dr (1 + o(1)). \end{aligned}$$

It remains to recall the definition of the conjugate measure and to apply Theorem 1.3. It is easily seen that instead of the cubes $|e_x - \varepsilon'| < t\tau^{1/2}$ we could take, say, parallelograms or ellipses.

The general case $\varepsilon' \neq (0, \dots, 0, 1)^T$ requires obvious alterations (cf. the proof of Theorem 1.4).

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