

## LOCALIZATION PROOF OF THE BAKRY–LEDOUX ISOPERIMETRIC INEQUALITY AND SOME APPLICATIONS\*

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(Translated by M. V. Khatuntseva)

**Abstract.** A Gaussian-type isoperimetric inequality due to D. Bakry and M. Ledoux is proved by means of the localization lemma of Lovász and Simonovits. Some applications on sharp large deviations are given.

**Key words.** isoperimetric inequality, localization, large deviations

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Let  $\gamma_n$  be a standard Gaussian measure in  $\mathbf{R}^n$ . Denote by  $\Phi$  the standard normal distribution function,

$$\Phi(x) = \gamma_1(-\infty, x] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-t^2/2} dt, \quad x \in \mathbf{R},$$

and by  $\Phi^{-1}: [0, 1] \rightarrow [-\infty, +\infty]$  the inverse function. As usual,  $\varphi(x) = (2\pi)^{-1/2} e^{-x^2/2}$ .

The isoperimetric theorem in the Gauss space  $(\mathbf{R}^n, \gamma_n)$  states that for all measurable  $A \subset \mathbf{R}^n$  and all  $h > 0$

$$(1) \quad \gamma_n(A^h) \geq \Phi(\Phi^{-1}(\gamma_n(A)) + h),$$

where  $A^h = \{x \in \mathbf{R}^n: \exists a \in A: |a-x| < h\}$  is an open  $h$ -neighborhood of the set  $A$  in a sense of ordinary Euclidean distance in  $\mathbf{R}^n$ . This important property of the Gaussian measure was discovered by Sudakov and Tsirel'son [26] and Borell [10]; they obtained inequality (1) as a corollary of the Lévy–Schmidt theorem on isoperimetric properties of balls on a sphere. Later Ehrhard [12] proposed a direct proof introducing the Gaussian symmetrization technique (see [9], [11], [22]). As was shown in [5], one can also derive (1) using an integral differential inequality for independent Bernoulli variables. Some abstract generalizations were studied in [6]; see also [3].

We consider the following natural generalization of the Gaussian isoperimetric inequality which was obtained by Bakry and Ledoux with use of a well developed semigroup technique [2], [21]. Let  $\mu$  be an absolutely continuous probability measure on  $\mathbf{R}^n$  with a density of the form  $d\mu(x)/dx = e^{-U(x)}$ , where  $U$  is a convex twice continuously differentiable function, whose second derivative satisfies (in the sense of comparing positively definite matrices in all points of the space) inequality  $cU''(x) \geq \text{Id}$  ( $c > 0$  is some constant). Then

$$(2) \quad \mu(A^h) \geq \Phi\left(\Phi^{-1}(\mu(A)) + \frac{h}{c}\right).$$

In frames of the more abstract scheme of diffusion generators, Bakry and Ledoux consider the statement of such type as an infinite dimensional analogue of the Gromov theorem [13], [14] generalizing the isoperimetric inequality on a sphere onto the Riemannian manifolds with the Ricci curve bounded from below. Considering the same measures on  $\mathbf{R}^n$  and, without loss

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of generality, the case  $c = 1$  in inequality (2), we can give the following equivalent statement of the Bakry-Ledoux result.

**THEOREM 1.** *If a probability measure  $\mu$  on  $\mathbf{R}^n$  has a log-concave density with respect to the measure  $\gamma_n$ , then for all measurable  $A \subset \mathbf{R}^n$  and all  $h > 0$*

$$\mu(A^h) \geq \Phi(\Phi^{-1}(\mu(A)) + h).$$

Recall that a nonnegative function  $\rho$  defined on  $\mathbf{R}^n$  is called log-concave if, for all  $x, y \in \mathbf{R}^n$  and all  $t \in (0, 1)$ , the inequality  $\rho((1 - t)x + ty) \geq \rho(x)^{1-t}\rho(y)^t$  holds. Up to the set of the Lebesgue measure zero, we can represent such a function in the form  $\rho(x) = e^{-U(x)}$ ,  $x \in K$ , where  $U$  is a convex function on some open convex set  $K \subset \mathbf{R}^n$ , setting  $\rho = 0$  outside of  $K$  (in this representation  $\rho$  is semicontinuous from below). In particular, Theorem 1 can be applied to the normalized contraction of the measure  $\gamma_n$  onto the arbitrary convex subset  $\mathbf{R}^n$  of the positive measure.

Note that using (1) we obtain inequality (2) for each image  $\mu = \gamma_n T^{-1}$  of the measure  $\gamma_n$  under the map  $T: \mathbf{R}^n \rightarrow \mathbf{R}^n$  having a finite Lipschitz seminorm  $c = \|T\|_{\text{Lip}}$ . For example, as was noted in [16], the uniform distribution on a cube  $[0, 1]^n$  can be obtained from  $\gamma_n$  with the help of the map  $T$  with the Lipschitz seminorm  $c = 1/\sqrt{2\pi}$ . The uniform distribution on a ball of volume 1 (hence of radius  $\sqrt{n}$ ) can also be obtained in the same way and the respective Lipschitz seminorm will be bounded by a constant independent of the dimensionality [8]. Continuing the analogy it will be interesting to see whether or not one can obtain the Bakry-Ledoux theorem in the same way (the problem of the intrinsic characterization of a class of all Lipschitz images of the Gaussian measure has not yet been solved).

There exists another way that permits us to obtain Theorem 1 and the Gaussian isoperimetric inequality as a direct corollary of the general localization lemma of Lovász and Simonovits [23]. The localization method in the form of [23] and [17] permits us to reduce some relations between multidimensional integrals to the relations between one-dimensional integrals, and, as we shall show in this paper, it particularly concerns inequality (2) (similar ideas were developed earlier by Gromov and Milman who investigated isoperimetric inequalities on a sphere; see [15], [1]). In the one-dimensional case probability measures having a log-concave density with respect to  $\gamma_1$  are indeed Lipschitz images of  $\gamma_1$ , and thus we can apply the one-dimensional inequality (1).

Denote by  $\mathcal{F}_n$  a family of all functions  $f: \mathbf{R}^n \rightarrow \mathbf{R}$  such that  $\|f\|_{\text{Lip}} \leq 1$ . Inequality (2) admits a series of equivalent statements in terms of distributions of Lipschitz functions in the same way as (1). For example, under the conditions of Theorem 1 we can equivalently affirm that for any  $f \in \mathcal{F}_n$  there exists a nondecreasing function  $f^* \in \mathcal{F}_1$  having the same distribution with respect to  $\gamma_1$  as  $f$  has with respect to  $\mu$ :  $\mu f^{-1} = \gamma_1(f^*)^{-1}$ . From here, in particular, follows an inequality for probabilities of deviations of  $f$  from its median  $m(f)$ :

$$\mu\{f - m(f) \geq h\} \leq 1 - \Phi(h), \quad h > 0.$$

We can write, with little degradation of the right-hand side, an analogous estimate for probability deviations of  $f$  from its mean value  $\mathbf{E}_\mu f = \int f d\mu$ . Cirel'son, Ibragimov, and Sudakov [16] were the first to study this problem in the Gaussian case  $\mu = \gamma_n$ ; they proved the inequality

$$(3) \quad \gamma_n\{f - \mathbf{E}_{\gamma_n} f \geq h\} \leq 2(1 - \Phi(h)), \quad h > 0.$$

The approach proposed in [16] is purely Gaussian. It is based on the representation of  $f - \mathbf{E}_{\gamma_n} f$  in the form of a functional of a Brownian process with random time and does not use Gaussian isoperimetric inequality. For large  $h$ , inequality (3) can be improved with the help of one Gaussian correlation identity (see [7]):

$$\gamma_n\{f - \mathbf{E}_{\gamma_n} f \geq h\} \leq \frac{\varphi(h)}{h}, \quad h > 0.$$

Indeed, relying on (1) and (2), one can get the exact information.

COROLLARY 1. Let  $\mu$  be a probability measure on  $\mathbf{R}^n$  having a log-concave density with respect to the measure  $\gamma_n$ . Then

$$\sup_{f \in \mathcal{F}_n} \mu\{f - \mathbf{E}_\mu f \geq h\} \leq \sup_{f \in \mathcal{F}_n} \gamma_n\{f - \mathbf{E}_{\gamma_n} f \geq h\} = 1 - \Phi(\alpha(h)), \quad h > 0,$$

where  $\alpha = \alpha(h)$  is a unique root of  $\alpha \Phi(\alpha) + \varphi(\alpha) = h$ .

Obviously,  $\alpha(h) \rightarrow -\infty$  for  $h \rightarrow 0+$ , so the constant 2 in (3) cannot be improved.

Let us note one more corollary of Theorem 1. Let us introduce the identity function  $f_1(x) = x$  on  $\mathbf{R}$ .

COROLLARY 2. Let  $\mu$  be a probability measure on  $\mathbf{R}^n$  having a log-concave density with respect to  $\gamma_n$ , and let  $\Psi$  be a convex function on  $\mathbf{R}$ . Then for all  $f \in \mathcal{F}_n$  with mean  $\mathbf{E}_\mu f = 0$

$$(4) \quad \mathbf{E}_\mu \Psi(f) \leq \mathbf{E}_{\gamma_1} \Psi(f_1).$$

In the Gaussian case  $\mu = \gamma_n$  such comparative estimates have also been studied by many authors. In the case  $\Psi(x) = x^2$  inequality (4) follows from an inequality of the Poincaré type for a Gaussian measure. In the case  $\Psi(x) = e^{\lambda x}$  we arrive at the other known inequality

$$(5) \quad \mathbf{E}_{\gamma_n} e^{\lambda f} \leq e^{\lambda^2/2}, \quad \lambda \in \mathbf{R}.$$

Although first obtained by Pisier and Maurey (see [25]), who applied, as in [16], functionals of a Brownian process, Ledoux [18] found another proof based on the properties of Ornstein–Uhlenbeck operators. Later it was discovered [19], [20] that (5) follows from the logarithmic Gross inequality. Under the additional assumption of a convexity of the function  $f$ , inequality (5) follows from one exponential inequality of Tsirel’son for a Gaussian random process [27], [28]. The case of an arbitrary convex function  $\Psi$  in the inequality

$$(6) \quad \mathbf{E}_{\gamma_n} \Psi(f) \leq \mathbf{E}_{\gamma_1} \Psi(f_1)$$

was considered by Pinelis [24]. He modified the approach proposed in [25] and obtained the more general inequality for the case where the Lipschitz function  $f$  takes values in  $\mathbf{R}^k$ , and  $\Psi$  is defined and convex on  $\mathbf{R}^k$  (respectively, the measure  $\gamma_1$  must be replaced with  $\gamma_k$ ). In [24] it is also shown that (6) implies (3) with a constant  $e$  instead of 2 and that this constant cannot be improved using (6). So inequalities for the deviations from the mean values are not equivalent to inequalities for moments in a class of convex functions  $\Psi$ . We give a simple proof of (6) using the Gaussian isoperimetric inequality.

Before we begin the proof let us formulate the localization Lovász–Simonovits theorem mentioned above.

LEMMA 1. Let continuous-from-below summable functions  $u, v$  on  $\mathbf{R}^n$  be given such that

$$(7) \quad \int_{\mathbf{R}^n} u(x) dx > 0, \quad \int_{\mathbf{R}^n} v(x) dx > 0.$$

Then there are  $a, b \in \mathbf{R}^n$  and an affine function  $\ell: (0, 1) \rightarrow (0, +\infty)$  such that

$$(8) \quad \int_0^1 u((1-t)a + tb) \ell(t)^{n-1} dt > 0, \quad \int_0^1 v((1-t)a + tb) \ell(t)^{n-1} dt > 0.$$

The last two integrals can be considered as normalized  $n$ -dimensional integrals of  $u$  and  $v$  over a frustum of a cone with an infinitely small base. In (7)–(8) (or, for example, only in (8)) one can replace symbols of strong inequalities with symbols of weak inequalities.

The proof of Lemma 1 given in [23] consists of two steps. In the first step the decreasing sequence of compacts  $K_i$  contracting in a point or interval  $[a, b] \subset \mathbf{R}^n$  is constructed so that the normalized integrals

$$\frac{1}{\text{Vol}_n(K_i)} \int_{K_i} u(x) dx, \quad \frac{1}{\text{Vol}_n(K_i)} \int_{K_i} v(x) dx$$

remain positive.

These integrals converge to one-dimensional integrals but, by the Brunn–Minkowski inequality, we pass in limit to (8) with some positive concave function  $\ell$  on  $(0,1)$ . Note that in this case the function  $\psi(t) = \ell(t)^{n-1}$  continued by zero outside of  $(0,1)$  is bounded and log-concave. We apply Lemma 1 in this weakened variant.

*Proof of Theorem 1.* First we show how Lemma 1 reduces Theorem 1 to the dimensionality  $n = 1$ . With respect to the Lebesgue measure, we can write the density of the measure  $\mu$  in the form

$$\frac{d\mu(x)}{dx} = \rho(x) \varphi_n(x), \quad x \in \mathbf{R}^n,$$

where  $\varphi_n(x) = d\gamma_n(x)/dx = (2\pi)^{-n/2} e^{-|x|^2/2}$  and  $\rho$  is the log-concave continuous-from-below function on  $\mathbf{R}^n$ . Fix  $p \in (0,1)$ ,  $h > 0$ , and choose an arbitrary open set  $A$  in  $\mathbf{R}^n$  of the measure  $\mu(A) > p$ . Using the one-dimensional inequality of Theorem 1, we show that  $\mu(A^h) \geq \Phi(\Phi^{-1}(p) + h)$ . Supposing the opposite we introduce the functions

$$u(x) = (\Phi(\Phi^{-1}(p) + h) - 1_{A^h}(x)) \rho(x) \varphi_n(x), \quad v(x) = (1_A(x) - p) \rho(x) \varphi_n(x),$$

where  $1_A$  denotes the indicator function. The functions  $u$  and  $v$  are continuous from below and  $\int_{\mathbf{R}^n} u(x) dx > 0$ ,  $\int_{\mathbf{R}^n} v(x) dx > 0$ . By Lemma 1 there exist  $a, b \in \mathbf{R}^n$  and a bounded log-concave function  $\psi: \mathbf{R} \rightarrow [0, +\infty)$  such that

$$\int_0^1 u((1-t)a + tb) \psi(t) dt > 0, \quad \int_0^1 v((1-t)a + tb) \psi(t) dt > 0.$$

The assumption  $a = b$  leads to a contradiction. In the case  $a \neq b$ , we set  $\theta = (b-a)/|b-a|$ ,  $r = |b-a|$  and after a change of variable  $z = rt$  we rewrite the last two inequalities in the form

$$(9) \quad \int_0^r u(a + z\theta) w(z) dz > 0, \quad \int_0^r v(a + z\theta) w(z) dz > 0,$$

where  $w(z) = \psi(z/r)$ .

Introduce a probability measure  $\nu$  on  $\mathbf{R}$  concentrated in  $[0, r]$  and having on it the density

$$\frac{d\nu(z)}{dz} = \frac{1}{c} w(z) \rho(a + z\theta) \varphi_n(a + z\theta), \quad z \in [0, r],$$

where  $c = \int_0^r w(y) \rho(a + y\theta) \varphi_n(a + y\theta) dy$  is a normalizing factor. Since  $|\theta| = 1$ , the function  $\varphi_n(a + z\theta)/\varphi_n(z)$  is log-concave with respect to  $z \in \mathbf{R}$ , and thus  $\nu$  has a log-concave density with respect to the measure  $\gamma_1$ . Now we introduce sets on the line

$$B = \{z \in \mathbf{R}: a + z\theta \in A\}, \quad C = \{z \in \mathbf{R}: a + z\theta \in A^h\}.$$

Then the inequalities in (9) become

$$(10) \quad \nu(C) < \Phi(\Phi^{-1}(\nu(B)) + h), \quad \nu(B) > p.$$

By construction,  $B^h \subset C$  (here the  $h$ -neighborhood is considered on the line); hence  $\nu(B^h) \leq \nu(C)$ . In view of (10) we obtain  $\nu(B^h) < \Phi(\Phi^{-1}(\nu(B)) + h)$ , which contradicts the one-dimensional inequality of Theorem 1. So

$$(11) \quad \mu(A^h) \geq \Phi(\Phi^{-1}(\mu(A)) + h)$$

for all open, and hence all measurable,  $A \subset \mathbf{R}^n$  as soon as we obtain that inequality for  $n = 1$ .

Now we consider the one-dimensional case. Let  $\mu$  be a probability measure on  $\mathbf{R}$  having with respect to the Lebesgue measure a density of the form

$$(12) \quad f(x) = \rho(x) \varphi(x),$$

where  $\rho$  is a log-concave function on  $\mathbf{R}$ . Let  $F(x) = \mu(-\infty, x]$  be a distribution function of the measure  $\mu$ . The measure  $\mu$  is concentrated on some interval  $(\alpha, \beta)$ , possibly infinite,

inside which  $\rho$  is positive and continuous. Let  $F^{-1}: (0, 1) \rightarrow (\alpha, \beta)$  be inverse to  $F$ . Show that the continuously differentiable increasing map  $T(x) = F^{-1}(\Phi(x))$  transforming the measure  $\gamma_1$  into  $\mu$  has the Lipschitz seminorm  $\|T\|_{\text{Lip}} \leq 1$ . The last is equivalent to the inequality  $f(F^{-1}(p)) \geq \varphi(\Phi^{-1}(p))$  for all  $p \in (0, 1)$ .

For this we prove the stronger statement: If a positive finite measure  $\mu$  on  $\mathbf{R}$  has a density of the form (12) with a log-concave function  $\rho$  and if a point  $x_0 \in \mathbf{R}$  is such that  $\mu(-\infty, x_0) \geq p$ ,  $\mu(x_0, +\infty) \geq q$  (where  $p \in (0, 1)$  is fixed and  $q = 1 - p$ ), then

$$(13) \quad f(x_0) \geq \varphi(\Phi^{-1}(p)).$$

Represent  $\rho$  in the form  $\rho(x) = e^{-U(x)}$ , where  $U$  is a convex function finite in some interval  $(\alpha, \beta)$  of the full  $\mu$ -measure. Then  $x_0 \in (\alpha, \beta)$  and we can construct a tangent  $\ell$  to  $U$  at point  $x_0$ . The function  $\rho_0(x) = e^{-\ell(x)}$  is log-concave and on the real axis satisfies inequality  $\rho_0 \geq \rho$ , and a measure  $\mu_0$  on  $\mathbf{R}$  defined with respect to the Lebesgue measure by the density  $f_0(x) = \rho_0(x)\varphi(x)$  satisfies  $\mu_0(-\infty, x_0) \geq p$ ,  $\mu_0(x_0, +\infty) \geq q$ . Moreover,  $f_0(x_0) = f(x_0)$ . Hence, without loss of generality we can assume in what follows (proving our stronger assumption) that  $U = \ell$  is an affine function on  $(\alpha, \beta) = \mathbf{R}$ .

So let  $f(x) = Ce^{\lambda x}\varphi(x)$  with parameters  $C > 0$ ,  $\lambda \in \mathbf{R}$ . We have

$$\begin{aligned} \mu(-\infty, x_0) &= \int_{-\infty}^{x_0} f(x) dx = Ce^{\lambda^2/2} \Phi(x_0 - \lambda) \geq p, \\ \mu(x_0, +\infty) &= \int_{x_0}^{+\infty} f(x) dx = Ce^{\lambda^2/2} (1 - \Phi(x_0 - \lambda)) \geq q \end{aligned}$$

if and only if  $C \geq e^{-\lambda^2/2} \max\{p/\Phi(x_0 - \lambda), q/(1 - \Phi(x_0 - \lambda))\}$ . Hence

$$\begin{aligned} f(x_0) &\geq e^{-\lambda^2/2} \max\left\{\frac{p}{\Phi(x_0 - \lambda)}, \frac{q}{1 - \Phi(x_0 - \lambda)}\right\} e^{\lambda x_0} \varphi(x_0) \\ &= \max\left\{\frac{p}{\Phi(x_0 - \lambda)}, \frac{q}{1 - \Phi(x_0 - \lambda)}\right\} \varphi(x_0 - \lambda). \end{aligned}$$

Since  $\lambda$  may be arbitrary in view of (13) we have to show that for all  $z \in \mathbf{R}$

$$(14) \quad \max\left\{\frac{p}{\Phi(z)}, \frac{q}{1 - \Phi(z)}\right\} \varphi(z) \geq \varphi(\Phi^{-1}(p)).$$

For  $z \leq \Phi^{-1}(p)$  the left-hand side of (14) is of the form  $(p/\Phi(z))\varphi(z)$ . Being a function of  $z$  it is decreasing (since  $\log \Phi$  is concave) and thus it takes the minimal value at the extreme point  $z = \Phi^{-1}(p)$ . Analogously, at the same point the left-hand side of (14) is minimizing on the interval  $z \geq \Phi^{-1}(p)$ . It remains to remark that for  $z = \Phi^{-1}(p)$  inequality (14) becomes the equality.

So the map  $T$  is a contraction of the measure  $\gamma_1$ , and thus one-dimensional isoperimetric inequality (11) holds for all  $\mu$  under consideration as soon as it holds for the measure  $\mu = \gamma_1$ . This concrete case can also be easily verified (see, for example, [4]). Theorem 1 is proved.

*Proof of Corollary 1.* As was noted, Theorem 1 can be formulated in the following way: For any function  $f \in \mathcal{F}_n$  there exists a nondecreasing function  $f^* \in \mathcal{F}_1$  such that  $\mu f^{-1} = \gamma_1(f^*)^{-1}$ . Thus

$$\sup_{f \in \mathcal{F}_n} \mu\{f - \mathbf{E}_\mu f \geq h\} \leq \sup_{f \in \mathcal{F}_n} \gamma_n\{f - \mathbf{E}_{\gamma_n} f \geq h\} = \sup_{f \in \mathcal{F}_1^+} \gamma_1\{f - \mathbf{E}_{\gamma_1} f \geq h\},$$

where  $\mathcal{F}_1^+$  denotes a family of all nondecreasing functions  $f$  on  $\mathbf{R}$  with the Lipschitz seminorm  $\|f\|_{\text{Lip}} \leq 1$ .

We maximize  $\gamma_1\{f - \mathbf{E}_{\gamma_1} f \geq h\}$  in the class  $\mathcal{F}_1^+$ . Let  $f \in \mathcal{F}_1^+$ ,  $a = \mathbf{E}_{\gamma_1} f$ . If  $h > f(x) - a$  for all  $x \in \mathbf{R}$ , then there is nothing to prove. Otherwise there exists a minimal  $\alpha \in \mathbf{R}$  such that  $f(\alpha) = a + h$ . Remark that the function  $g(x) = \min\{x - \alpha, 0\} + a + h$  belongs to  $\mathcal{F}_1^+$

and satisfies the inequality  $g(x) \leq f(x)$  for all  $x \in \mathbf{R}$ . In particular,  $\mathbf{E}_{\gamma_1} g \leq a$ . Moreover, since  $f(x) < f(\alpha)$  for  $x < \alpha$ , it follows that  $\{f \geq a + h\} = \{g \geq a + h\} = [\alpha, +\infty)$ . Hence,

$$\gamma_1\{f - a \geq h\} = \gamma_1\{g - a \geq h\} \leq \gamma_1\{g - \mathbf{E}_{\gamma_1} g \geq h\}.$$

So in view of the fact that the probabilities under consideration do not change after adding constants to the functions, we can restrict ourselves to a class of functions of the form  $g(x) = \min\{x - \alpha, 0\}$ . For such functions we have

$$\mathbf{E}_{\gamma_1} g = \int_{-\infty}^{\alpha} (x - \alpha) d\Phi(x) = -\varphi(\alpha) - \alpha\Phi(\alpha) \equiv -\xi(\alpha).$$

Thus

$$(15) \quad \gamma_1\{g - \mathbf{E}_{\gamma_1} g \geq h\} = 1 - \Phi(h + \alpha - \xi(\alpha)) \quad \text{if } h \leq \xi(\alpha),$$

and  $\gamma_1\{g - \mathbf{E}_{\gamma_1} g \geq h\} = 0$  in the case  $h > \xi(\alpha)$ . Further we have  $\xi'(\alpha) = \Phi(\alpha)$ , and hence the function  $\alpha - \xi(\alpha)$  is continuously increasing on the real axis changing in the interval  $(-\infty, +\infty)$ . Thus the right-hand side in (15) is maximal in the case  $\xi(\alpha) = h$ . This proves Corollary 1.

*Proof of Corollary 2.* As above, Theorem 1 reduces Corollary 2 to the case where  $n = 1$ ,  $\mu = \gamma_1$ , and  $f \in \mathcal{F}_1^+$ . As the following statement shows, here the Gaussian property is unessential.

LEMMA 2. *Let  $\mu$  be a probability measure on  $\mathbf{R}$  with a finite first moment and let  $\Psi$  be a convex function on  $\mathbf{R}$ . Then in the class of all nondecreasing functions  $f$  on  $\mathbf{R}$  with the Lipschitz seminorm  $\|f\|_{\text{Lip}} \leq 1$  the expression  $\mathbf{E}_{\mu}\Psi(f - \mathbf{E}_{\mu}f)$  achieves its maximal value (possibly, infinite) on the identity function  $f = f_1$ .*

*Proof.* Set  $u = f - \mathbf{E}_{\mu}f$ ,  $u_1 = f_1 - \mathbf{E}_{\mu}f_1$ . We can assume that the function  $\Psi$  is differentiable everywhere and its derivative  $\Psi'$  is bounded. In this case the function  $\psi(t) = \mathbf{E}_{\mu}\Psi((1 - t)u + tu_1)$  is finite everywhere, differentiable, and

$$\psi'(t) = \mathbf{E}_{\mu}\Psi'((1 - t)u + tu_1)(u_1 - u), \quad t \in \mathbf{R}.$$

Supposing the opposite we assume that  $\mathbf{E}_{\mu}\Psi(u_1) < \mathbf{E}_{\mu}\Psi(u)$ , i.e., that  $\psi(1) < \psi(0)$ . Since  $\psi$  is convex, this assumption implies the inequality  $\psi'(0) < 0$ , i.e.,

$$(16) \quad \mathbf{E}_{\mu}\Psi'(u)u_1 < \mathbf{E}_{\mu}\Psi'(u)u.$$

On the other hand, taking into account that  $\mathbf{E}_{\mu}u = \mathbf{E}_{\mu}u_1 = 0$ ,  $u$  satisfies  $0 \leq u(x) - u(y) \leq x - y = u_1(x) - u_1(y)$  for all  $x > y$ , and  $\Psi'$  is not decreasing, we obtain

$$\begin{aligned} \mathbf{E}_{\mu}\Psi'(u)u &= \iint_{x>y} (\Psi'(u(x)) - \Psi'(u(y))) (u(x) - u(y)) d\mu(x) d\mu(y) \\ &\leq \iint_{x>y} (\Psi'(u(x)) - \Psi'(u(y))) (u_1(x) - u_1(y)) d\mu(x) d\mu(y) = \mathbf{E}_{\mu}\Psi'(u)u_1. \end{aligned}$$

However, this contradicts (16). Lemma 2 and Corollary 2 are proved.

*Remark.* After this paper had been prepared, the author learned about the paper [29]. It specifically proves that all the probability measures considered in Theorem 1 can be represented as Lipschitz images of the standard Gaussian measure in  $\mathbf{R}^n$ .

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