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RESTRICTED PRÉKOPA-LEINDLER INEQUALITY

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We prove a functional version of the Brunn-Minkowski inequality for restricted sums obtained by Szarek and Voiculescu.

We only consider Lebesgue-measurable subsets of \mathbb{R}^n , and for $A \subset \mathbb{R}^n$, we denote its volume by |A|. If $A, B \subset \mathbb{R}^n$, their Minkowski sum is defined by

$$A + B = \{x + y, (x, y) \in A \times B\}.$$

The classical Brunn-Minkowski inequality provides a lower bound for its volume.

Theorem 1. Let A, B be compact, non void subsets of \mathbb{R}^n , one has

$$|A+B|^{\frac{1}{n}} \ge |A|^{\frac{1}{n}} + |B|^{\frac{1}{n}}.$$

In their study of the free analogue of the entropy power inequality [SV], Szarek and Voiculescu define the notion of restricted Minkowski sum of Aand B with respect to $\Theta \subset A \times B$:

$$A +_{\Theta} B = \{x + y, (x, y) \in \Theta\},\$$

and show that an analogue of the Brunn-Minkowski inequality holds:

Theorem 1'. There exists a positive constant c such that for all $\rho \in]0, 1[$, $n \in \mathbb{N}$, for all $A, B \subset \mathbb{R}^n$ and $\Theta \subset A \times B$ such that:

$$\rho \le \left(\frac{|A|}{|B|}\right)^{\frac{1}{n}} \le \rho^{-1} \quad and \quad \frac{|\Theta|}{|A| \cdot |B|} \ge 1 - c\min(\rho\sqrt{n}, 1),$$

one has

$$|A +_{\Theta} B|^{\frac{2}{n}} \ge |A|^{\frac{2}{n}} + |B|^{\frac{2}{n}}.$$

It is well known that the Brunn-Minkowski inequality can be derived from the Prékopa-Leindler inequality [**Pré**], [**Lei**], which we recall here: **Theorem 2.** Let f, g be non-negative functions in $L_1(\mathbb{R}^n)$ and $\lambda \in]0, 1[$, let H be a measurable function on \mathbb{R}^n such that

 $H(x) \ge \sup\{f^{\lambda}(u)g^{1-\lambda}(v), (u,v) \in \mathbb{R}^n \times \mathbb{R}^n \text{ and } x = \lambda u + (1-\lambda)v\},$

then

$$\int_{\mathbb{R}^n} H(x) \, dx \ge \left(\int f\right)^{\lambda} \left(\int g\right)^{1-\lambda}$$

We show that a corresponding restricted version of this statement holds.

Theorem 2'. There exist positive scalars c and n_0 such that for all $0 < \varepsilon \leq 1/2$, for all $\lambda \in [\varepsilon, 1 - \varepsilon]$ and for all $n \geq n_0$, if f, g are non-negative functions in $L_1(\mathbb{R}^n)$ and if Θ is a measurable subset of \mathbb{R}^{2n} such that

$$\frac{\int_{\Theta} f(x)g(y) \, dx \, dy}{\left(\int f\right) \left(\int g\right)} \ge \frac{1}{2} + \frac{c}{\sqrt{\varepsilon}} \cdot \frac{\log n}{\sqrt{n}}$$

then

$$\int_{\mathbb{R}^n} K(x) \, dx \ge \left(\int f\right)^{\lambda} \left(\int g\right)^{1-\lambda},$$

as soon as the function K satisfies:

$$K(x) \ge \sup\{f^{\lambda}(u)g^{1-\lambda}(v), (u,v) \in \Theta \text{ and } x = \sqrt{\lambda}u + \sqrt{1-\lambda}v\}.$$

Let us return to the example given in [SV] to show that the condition on the ratio

$$\theta = \frac{\int_{\Theta} f(x)g(y) \, dx \, dy}{\left(\int f\right) \left(\int g\right)}$$

is asymptotically optimal. Let B_2^n be the Euclidean unit ball in \mathbb{R}^n and let

$$\Theta = \{ (x, y) \in \mathbb{R}^n \times \mathbb{R}^n, \, \langle x, y \rangle \le 0 \}$$

then $|\Theta \cap (B_2^n \times B_2^n)| = 1/2 |B_2^n|^2$ and the Θ -restricted sum of a ball of radius r_1 and a ball of radius r_2 is a ball of radius $\sqrt{r_1^2 + r_2^2}$. In particular, for all $\lambda \in [0, 1]$,

$$\sqrt{\lambda}B_2^n +_{\Theta} \sqrt{1-\lambda}B_2^n = B_2^n.$$

The conclusion of Theorem 2' applied when f and g are the characteristic function of B_2^n would be

$$\left|\sqrt{\lambda}B_2^n +_{\Theta}\sqrt{1-\lambda}B_2^n\right| \ge |B_2^n|,$$

and actually the equality holds. It is then clear that the conclusion of Theorem 2' becomes false for ratios $\theta < 1/2$.

We shall first show that Theorem 2' implies Theorem 1', maybe with different conditions on the parameters. Let A, B be two subsets of \mathbb{R}^n , let $\Theta \subset A \times B$ such that

$$\rho := \left(\frac{|A|}{|B|}\right)^{\frac{1}{n}} \le 1$$

Assume that the ratio $\theta = \frac{|\Theta|}{|A| \cdot |B|}$ is larger than $\frac{1}{2} + c\sqrt{\frac{1+\rho^2}{\rho^2}} \cdot \frac{\log n}{\sqrt{n}}$. Let us define the set

$$\tilde{\Theta} = \left\{ \left(\frac{a}{|A|^{\frac{1}{n}}}, \frac{b}{|B|^{\frac{1}{n}}} \right) \in \mathbb{R}^{2n}, \, (a, b) \in \Theta \right\}.$$

Let

$$\tilde{A} = \frac{A}{|A|^{\frac{1}{n}}}$$
 and $\tilde{B} = \frac{B}{|B|^{\frac{1}{n}}}$

and let f and g be the characteristic functions of \tilde{A} and \tilde{B} . A simple change of variables gives that

$$\frac{\int_{\tilde{\Theta}} f(x)g(y)\,dx\,dy}{\left(\int f\right)\,\left(\int g\right)} = \frac{|\Theta|}{|A|.|B|} = \theta,$$

so we can apply Theorem 2' to f and g, with $\lambda = \frac{|A|^{\frac{2}{n}}}{|A|^{\frac{2}{n}} + |B|^{\frac{2}{n}}} = \frac{\rho^2}{1+\rho^2}$ and get

$$\left|\sqrt{\lambda}\,\tilde{A} +_{\tilde{\Theta}}\sqrt{1-\lambda}\,\tilde{B}\right| \ge 1,$$

where

$$\begin{split} \sqrt{\lambda}\,\tilde{A} +_{\tilde{\Theta}}\sqrt{1-\lambda}\,\tilde{B} &= \left\{\sqrt{\lambda}\frac{a}{|A|^{\frac{1}{n}}} + \sqrt{1-\lambda}\frac{b}{|B|^{\frac{1}{n}}},\,(a,b)\in\Theta\right\}\\ &= \left\{\frac{a+b}{\sqrt{|A|^{\frac{2}{n}} + |B|^{\frac{2}{n}}}},\,(a,b)\in\Theta\right\}\\ &= \frac{A+\Theta}{\sqrt{|A|^{\frac{2}{n}} + |B|^{\frac{2}{n}}}}\cdot\end{split}$$

Hence, we obtain

$$|A +_{\Theta} B|^{\frac{2}{n}} \ge |A|^{\frac{2}{n}} + |B|^{\frac{2}{n}}.$$

Our method is based on an observation of Brascamp and Lieb [**BL1**]: the Prékopa-Leindler inequality is a limit case of the reverse sharp form of Young's convolution inequality. We will first prove a restricted form of

Young's inequality and its converse, using a modification of the method we developed in [**Bar**], and then take the limits in certain parameters. Our proof of Young's inequality is based on measure-preserving mappings between measures. We use them in order to build a suitable change of variables which makes the problem simpler; then a simple arithmetico-geometric inequality gives the result. Now, we have to work with functions on \mathbb{R}^n , because the set Θ makes it difficult to use the classical tensorisation argument. In general, given two probability on \mathbb{R}^n , there are several measure-preserving mappings between them; for our purpose, the mapping built by Knothe in [**Kno**] fits:

Lemma 1. Let f, F be positive continuous functions on \mathbb{R}^n such that $\int f = \int F$. There exists a differentiable map $u : \mathbb{R}^n \to \mathbb{R}^n$ such that for $x \in \mathbb{R}^n$

(1) $\det(du(x)) \cdot f(u(x)) = F(x),$

and for all $i \leq n$ and $all(x_i)_{i=1}^n \in \mathbb{R}^n$,

$$u((x_i)_{i=1}^n) = (u_1(x_1), u_2(x_1, x_2), \dots, u_n(x_1, \dots, x_n)),$$

where for all x_1, \ldots, x_{i-1} , the function $u_i(x_1, \ldots, x_{i-1}, \cdot)$ is increasing on \mathbb{R} . In particular du(x) has always a lower triangular matrix with positive diagonal (in the canonical basis).

We also need a version of the arithmetico-geometric inequality for matrices of the previous form:

Lemma 2. Let M, N be lower triangular $n \times n$ -matrices with non-negative diagonal and let $t \in [0, 1]$, then

$$\det(tM + (1-t)N) \ge (\det M)^t (\det N)^{1-t}.$$

The first step of the proof is the following restricted version of Young's inequality. For t > 1, we denote by t' the real number such that 1/t + 1/t' = 1.

Lemma 3. Let f, F, g, G be positive continuous functions on \mathbb{R}^n , of integral 1 and dominated by some Gaussian function. Let u and v denote the measure preserving mappings obtained when applying Lemma 1 to (f, F) and (g, G) and let T be the bijective map of $\mathbb{R}^n \times \mathbb{R}^n$ defined by T(x, y) = (u(x), v(y)). Let $p, q, r \ge 1$ such that 1/p + 1/q = 1 + 1/r. We set

$$c = \sqrt{r'/q'}$$
 and $s = \sqrt{r'/p'}$,

and notice that $c^2 + s^2 = 1$. Then

(2)
$$\int f(x)g(y)\mathbf{1}_{T\Theta}(x,y)\,dx\,dy = \int F(X)G(Y)\mathbf{1}_{\Theta}(X,Y)\,dX\,dY,$$

and

$$\left(\int \left(\int f^{\frac{1}{p}}(cx-sy)g^{\frac{1}{q}}(sx+cy)\mathbf{1}_{T\Theta}(cx-sy,\,sx+cy)dx\right)^{r}\,dy\right)^{\frac{1}{r}}$$
$$\leq \int \left(\int F^{\frac{r}{p}}(cX-sY)G^{\frac{r}{q}}(sX+cY)\mathbf{1}_{\Theta}(cX-sY,sX+cY)\,dY\right)^{\frac{1}{r}}\,dX.$$

Proof. Equality (2) is a consequence of the measure-preserving properties of u and v. We give a detailed proof of the inequality. Let R be the rotation of matrix $\begin{pmatrix} c & -s \\ s & c \end{pmatrix}$ in the canonical basis. We are going to use the change of variable in \mathbb{R}^{2n} given by the function $\Phi = ({}^{t}R \otimes I_n)T(R \otimes I_n)$, where I_n is the identity map on \mathbb{R}^n . More precisely $(x, y) = \Phi(X, Y)$ means

$$x = c u(cX - sY) + s v(sX + cY)$$

$$y = -s u(cX - sY) + c v(sX + cY).$$

It is clear that Φ is a differentiable bijection of \mathbb{R}^{2n} . Its jacobian at the point (X, Y) is

$$J\Theta(X,Y) = \det(du(cX - sY))\det(dv(sX + cY)).$$

We want an upper estimate for the integral (finite by assumption)

$$I = \left(\int \left(\int f^{\frac{1}{p}} (cx - sy) g^{\frac{1}{q}} (sx + cy) \mathbf{1}_{T\Theta} (cx - sy, \, sx + cy) \, dx \right)^r \, dy \right)^{\frac{1}{r}}.$$

Using the $(L^r, L^{r'})$ -duality, there exists a positive function h on \mathbb{R}^n such that $\|h\|_{r'} = 1$ and

$$I = \iint f^{\frac{1}{p}}(cx - sy)g^{\frac{1}{q}}(sx + cy)\mathbf{1}_{T\Theta}(cx - sy, \, sx + cy)h(y)\,dx\,dy.$$

By the change of variable $(x, y) = \Phi(X, Y)$, we obtain that I is equal to

$$\iint f^{\frac{1}{p}}(u(cX-sY))g^{\frac{1}{q}}(v(sX+cY))h(-su(cX-sY)+cv(sX+cY))$$
$$\cdot \mathbf{1}_{T\Theta}(u(cX-sY),v(sX+cY))$$
$$\cdot \det(du(cX-sY))\det(dv(sX+cY))\,dXdY.$$

In order to shorten the formulas, denote

$$U = u(cX - sY), \qquad V = v(sX + cY),$$

$$U' = \det(du(cX - sY)), \qquad V' = \det(dv'(sX + cY)).$$

Noticing that the definition of T implies $\mathbf{1}_{T\Theta}(u(cX - sY), v(sX + cY)) = \mathbf{1}_{\Theta}(cX - sY, sX + cY)$, and using the differential formulas

$$\det(du(x)).f(u(x)) = F(x),$$

$$\det(dv(x)).g(u(x)) = G(x),$$

we get

$$\begin{split} I &= \iint f^{\frac{1}{p}} (u(cX - sY)) g^{\frac{1}{q}} (v(sX + cY)) \mathbf{1}_{\Theta} (cX - sY, sX + cY) \\ &\cdot h(-sU + cV) U'V' \, dX dY \\ &= \int \bigg(\int F^{\frac{1}{p}} (cX - sY) G^{\frac{1}{q}} (sX + cY) \mathbf{1}_{\Theta} (cX - sY, sX + cY) \\ &\cdot h(-sU + cV) (U')^{\frac{1}{p'}} (V')^{\frac{1}{q'}} \, dY \bigg) dX. \end{split}$$

Using Hölder's inequality for the integral in Y with parameters r and r', one has:

$$I \leq \int \left(\int F^{\frac{r}{p}}(cX - sY)G^{\frac{r}{q}}(sX + cY)\mathbf{1}_{\Theta}(cX - sY, sX + cY) \, dY \right)^{\frac{1}{r}} \\ \cdot \left(\int h^{r'}(-sU + cV)(U')^{\frac{r'}{p'}}(V')^{\frac{r'}{q'}} \, dY \right)^{\frac{1}{r'}} \, dX.$$

Let $H(X) = \int h^{r'}(-sU + cV)(U')^{\frac{r'}{p'}}(V')^{\frac{r'}{q'}} \, dY$, then
 $H(X) = \int h^{r'}(a(X, Y))(\det du(cX - sY))^{s^2} (\det dv(sX + cY))^{c^2} \, dY,$

where

$$a(X,Y) = -s u(cX - sY) + c v(sX + cY).$$

It is clear that the partial differential of a with respect to Y is

$$d_Y a(X, Y) = s^2 du(cX - sY) + c^2 dv(sX + cY).$$

By the arithmetico-geometric inequality stated in Lemma 2,

$$\det(d_Y a(X,Y)) \ge (\det du(cX-sY))^{s^2} (\det dv(sX+cY))^{c^2},$$

hence

$$H(X) \le \int h^{r'}(a(X,Y)) \det d_Y a(X,Y) \, dY \le \int h^{r'} = 1,$$

where we use the fact that a(X, Y) is an injective function of Y (indeed, u and v are by definition increasing for the lexicographic order on \mathbb{R}^n). This proves that

$$I \leq \int \left(\int F^{\frac{r}{p}}(cX - sY)G^{\frac{r}{q}}(sX + cY)\mathbf{1}_{\Theta}(cX - sY, sX + cY)\,dY \right)^{\frac{1}{r}}\,dX.$$

We are going to take a limit in r to obtain an inequality similar to the Prékopa-Leindler inequality. To simplify the notations, we set $\kappa = 1 - \lambda$.

Lemma 4. Let f, g, F, G be as in Lemma 3. Let $\Theta \subset \mathbb{R}^{2n}$ and denote $\theta = \int_{\Theta} F(X)G(Y) dX dY$. Then

$$\int \sup_{X=\sqrt{\lambda}u+\sqrt{\kappa}v} F^{\lambda}(u)G^{\kappa}(v)\mathbf{1}_{\Theta}(u,v) dX$$

$$\geq \inf_{A} \sup_{y\in\mathbb{R}^{n}} \int f^{\lambda} \left(\sqrt{\lambda}x - \sqrt{\kappa}y\right) g^{\kappa} \left(\sqrt{\kappa}x + \sqrt{\lambda}y\right)$$

$$\cdot \mathbf{1}_{A} \left(\sqrt{\lambda}x - \sqrt{\kappa}y, \sqrt{\kappa}x + \sqrt{\lambda}y\right) dx,$$

where the infimum is over the sets $A \subset \mathbb{R}^{2n}$ such that $\int_A f(x)g(y) \, dx dy \geq \theta$.

Proof. This lemma is a limit case of Lemma 3. For r > 1, we set

$$p_r = \frac{r}{\lambda(r+1)},$$
$$q_r = \frac{r}{\kappa(r+1)}.$$

Then $1/p_r + 1/q_r = 1 + 1/r$ and when r is large enough $p_r, q_r > 1$. We apply Lemma 3 with f, g, F, G for this triple and take the limit when r tends to $+\infty$. Notice that

$$\frac{1}{p_r} \to \lambda, \qquad \frac{1}{q_r} \to \kappa,$$

and

$$c_r = \sqrt{\frac{r'}{q'_r}} = \sqrt{\frac{1 - q_r^{-1}}{1 - r^{-1}}} \to \sqrt{\lambda}, \qquad s_r \to \sqrt{\kappa} \,.$$

Our strong domination hypothesis ensures that the r-norms tend to essential suprema when r tends to infinity. We get:

$$\begin{split} \sup_{y \in \mathbb{R}^n} & \int f^{\lambda} \left(\sqrt{\lambda} \, x - \sqrt{\kappa} \, y \right) g^{\kappa} \left(\sqrt{\kappa} \, x + \sqrt{\lambda} \, y \right) \\ & \cdot \mathbf{1}_{T\Theta} \left(\sqrt{\lambda} \, x - \sqrt{\kappa} \, y, \sqrt{\kappa} \, x + \sqrt{\lambda} \, y \right) \, dx \\ & \leq \int_{\mathbb{R}^n} \sup_{Y \in \mathbb{R}^n} F^{\lambda} \left(\sqrt{\lambda} \, X - \sqrt{\kappa} \, Y \right) G^{\kappa} \left(\sqrt{\kappa} \, X + \sqrt{\lambda} \, Y \right) \\ & \cdot \mathbf{1}_{\Theta} \left(\sqrt{\lambda} \, X - \sqrt{\kappa} \, Y, \sqrt{\kappa} \, X + \sqrt{\lambda} \, Y \right) \, dX. \end{split}$$

Noticing that $\begin{cases} u = \sqrt{\lambda} X - \sqrt{\kappa} Y\\ v = \sqrt{\kappa} X + \sqrt{\lambda} Y \end{cases}$ is equivalent to $\begin{cases} X = \sqrt{\lambda} u + \sqrt{\kappa} v\\ Y = -\sqrt{\kappa} u + \sqrt{\lambda} v \end{cases}$, we can rewrite the second member of the previous inequality as

$$\int \sup_{X=\sqrt{\lambda}\,u+\sqrt{\kappa}\,v} F^{\lambda}(u)G^{\kappa}(v)\mathbf{1}_{\Theta}(u,v)\,dX.$$

By equality (2) in Lemma 3, we have $\int_{T\Theta} f(x)g(y) dxdy = \theta$, which leads to the conclusion.

To finish the proof of Theorem 2', we have to estimate the infimum given in the previous lemma for two specific functions f and g.

Lemma 5. Let F, G be as in Lemma 3, then

$$\int \sup_{X=\sqrt{\lambda}u+\sqrt{\kappa}v} F^{\lambda}(u)G^{\kappa}(v)\mathbf{1}_{\Theta}(u,v) dX$$

$$\geq \mathbb{E}\left(\exp\left(\sqrt{\lambda\kappa}\sum_{i=1}^{n} X_{i}\right)\mathbf{1}_{\{\sum X_{i}\leq M_{n,\theta}\}}\right),$$

where $(X_i)_{i=1}^n$ is a sequence of i.i.d. random variables, their common law being the law of a difference of squares of two independent Gaussian variables $N(0, 1/\sqrt{2})$ and the number $M_{n,\theta}$ satisfies $\mathbb{P}(\sum X_i \leq M_{n,\theta}) = \theta$.

Proof. We apply Lemma 4 with

$$f(x) = g(x) = \pi^{-n/2} e^{-x^2}.$$

We denote by γ_n the probability measure on \mathbb{R}^n with the previous density. We want a lower estimate of

$$\mathcal{I} = \inf_{\gamma_{2n}(A)=\theta} \sup_{y\in\mathbb{R}^n} \int \exp\left(-\lambda \left(\sqrt{\lambda}\,x - \sqrt{\kappa}\,y\right)^2 - \kappa \left(\sqrt{\kappa}\,x + \sqrt{\lambda}\,y\right)^2\right) \\ \cdot \mathbf{1}_A \left(\sqrt{\lambda}\,x - \sqrt{\kappa}\,y, \sqrt{\kappa}\,x + \sqrt{\lambda}\,y\right) \pi^{-n/2} dx.$$

Since the condition on A is rotation invariant, we can replace A by B such that $(x, y) \in B$ if and only if $(\sqrt{\lambda} x - \sqrt{\kappa} y, \sqrt{\kappa} x + \sqrt{\lambda} y) \in A$. Hence

$$\begin{aligned} \mathcal{I} &= \inf_{\substack{\gamma_{2n}(B)=\theta}} \sup_{y\in\mathbb{R}^n} \int \exp\left(-\lambda \left(\sqrt{\lambda}\,x - \sqrt{\kappa}\,y\right)^2 - \kappa \left(\sqrt{\kappa}\,x + \sqrt{\lambda}\,y\right)^2\right) \\ &\cdot \mathbf{1}_B(x, y) \pi^{-n/2} \, dx \\ &= \inf_{\substack{\gamma_{2n}(B)=\theta}} \sup_{y\in\mathbb{R}^n} \int \exp\left(x^2 - \lambda \left(\sqrt{\lambda}\,x - \sqrt{\kappa}\,y\right)^2 - \kappa \left(\sqrt{\kappa}\,x + \sqrt{\lambda}\,y\right)^2\right) \\ &\cdot \mathbf{1}_B(x, y) \, d\gamma_n(x), \\ &\geq \inf_{\substack{\gamma_{2n}(B)=\theta}} \int \int \exp\left(x^2 - \lambda \left(\sqrt{\lambda}\,x - \sqrt{\kappa}\,y\right)^2 - \kappa \left(\sqrt{\kappa}\,x + \sqrt{\lambda}\,y\right)^2\right) \end{aligned}$$

$$\cdot \mathbf{1}_B(x,y) \, d\gamma_n(x) d\gamma_n(y)$$

The matrix of the quadratic form on \mathbb{R}^{2n} , $Q(x, y) = x^2 - \lambda(\sqrt{\lambda} x - \sqrt{\kappa} y)^2 - \kappa(\sqrt{\kappa} x + \sqrt{\lambda} y)^2$ in a suitable orthonormal basis is

$$\sqrt{\lambda\kappa} \left(\begin{array}{cc} I_n & 0\\ 0 & -I_n \end{array} \right),$$

where I_n is the identity $n \times n$ matrix. Hence, by rotation invariance of the Gaussian measure

$$\mathcal{I} \ge \inf_{\gamma_{2n}(B)=\theta} \iint_{B} \exp\left(\sqrt{\lambda\kappa}(x^2 - y^2)\right) \, d\gamma_n(x) d\gamma_n(y).$$

This is exactly

$$\mathcal{J} = \iint \exp\left(\sqrt{\lambda\kappa}(x^2 - y^2)\right) \mathbf{1}_{\{x^2 - y^2 \le M_{n,\theta}\}} d\gamma_n(x) d\gamma_n(y),$$

where $M_{n,\theta}$ is such that $\gamma_{2n}(\{x^2 - y^2 \leq M_{n,\theta}\}) = \theta$. We get the conclusion of the lemma by rewriting this with $X_i = x_i^2 - y_i^2$, where x_i and y_i are the i^{th} coordinates of x and y.

We are going to use the central-limit theorem in the rather precise form of the Berry-Essen theorem. [Fel].

Theorem 3. Let $(X_i)_{i \in \mathbb{N}}$ be a sequence of *i.i.d.* random variables, let

$$m = \mathbb{E}X_i, \qquad \sigma = \left(\mathbb{E}X_i^2\right)^{\frac{1}{2}} \quad and \quad \beta = \mathbb{E}|X_i|^3.$$

For all $t \in \mathbb{R}$, let

$$F_n(t) = \mathbb{P}\left(\frac{\sum_{i=1}^n X_i - nm}{\sigma\sqrt{n}} < t\right)$$

and

$$G(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{t} e^{-s^2/2} \, ds.$$

There exists a universal constant c > 0 such that for all t and for all n,

$$|F_n(t) - G(t)| \le \frac{c\beta}{\sigma^3 \sqrt{n}}$$

Proof of Theorem 2'. By homogeneity, we may assume $\int F = \int G = 1$. Comparing the assertions of Lemma 5 and Theorem 2', we see that to prove the latter, it is enough to show that the expectation from the former is ≥ 1 provided the parameter $\theta = \int_{\Theta} F(x)G(y)dxdy$ exceeds

$$\frac{1}{2} + \frac{c}{\sqrt{\varepsilon}} \frac{\log n}{\sqrt{n}}.$$

To this end, we apply Theorem 3 to the variables X_i defined in Lemma 5, and notice that m = 0 and β, σ and c are universal constants. We set

$$\xi_n = \frac{\log n}{\sigma \sqrt{\lambda \kappa n}}$$
 and $\alpha = \frac{c\beta}{\sigma^3}$.

We fix λ and prove that, for *n* large enough and for

$$\theta = G(\xi_n) + \frac{\alpha}{\sqrt{n}},$$

the quantity

$$\mathcal{J} = \mathbb{E}\left(\exp\left(\sqrt{\lambda\kappa}\sum_{i=1}^{n}X_{i}\right)\mathbf{1}_{\{\sum X_{i}\leq M_{n,\theta}\}}\right)$$

is larger than 1.

As $\mathbb{E}X_i = 0$, we get from the Berry-Essen theorem

$$\mathbb{P}\left(\sum_{i=1}^{n} X_i < \xi_n \sigma \sqrt{n}\right) \le G(\xi_n) + \frac{c\beta}{\sigma^3 \sqrt{n}} = \theta,$$

so $M_{n,\theta} \ge \xi_n \sigma \sqrt{n}$. We set $Z_n = \frac{\sum_{i=1}^n X_i}{\sigma \sqrt{n}}$, it is clear that $\mathcal{J} \ge \mathbb{E} \left(\exp \left(\sigma \sqrt{\lambda \kappa n} Z_n \right) \mathbf{1}_{\{Z_n \le \xi_n\}} \right).$

Let $n_1(\lambda)$ be the smallest integer n such that $\xi_n \leq 1$, notice that it is a non-increasing function of $\lambda \in]0, 1/2]$. We work with $n \geq n_1(\lambda)$. When n is large, Z_n behaves like a normal Gaussian g. So we can almost estimate this expectation by replacing Z_n by g.

More precisely, let $d = 2\alpha\sqrt{2\pi e}$ and let $n_2(\lambda)$ be the smallest integer such that $\xi_n/3 \ge d/\sqrt{n}$, it is a non-decreasing function of $\lambda \in]0, 1/2]$. Then for $n > \max(n_1(\lambda), n_2(\lambda))$, one has

$$\mathbb{P}_{Z_n}\left([t,\xi_n]\right) \ge \mathbb{P}_g\left(\left[t,\xi_n-\frac{d}{\sqrt{n}}\right]\right).$$

This comes from the Berry-Essen theorem and from the fact that ξ_n stays in [0, 1] where the density of the law of g is bounded from below:

$$\mathbb{P}_{Z_n}\left([t,\xi_n]\right) \geq \mathbb{P}_g\left([t,\xi_n]\right) - 2\frac{\alpha}{\sqrt{n}}$$

$$= \mathbb{P}_g\left(\left[t,\xi_n - \frac{d}{\sqrt{n}}\right]\right) + \frac{1}{\sqrt{2\pi}}\int_{\xi_n - \frac{d}{\sqrt{n}}}^{\xi_n} e^{-t^2/2}dt - 2\frac{\alpha}{\sqrt{n}}$$

$$\geq \mathbb{P}_g\left(\left[t,\xi_n - \frac{d}{\sqrt{n}}\right]\right) + \frac{1}{\sqrt{n}}\left(\frac{d}{\sqrt{2\pi}}e^{-1/2} - 2\alpha\right).$$

We are now able to compute a lower estimate of \mathcal{J} .

$$\mathcal{J} \geq \int_{-\infty}^{\xi_n} \exp\left(\sigma\sqrt{\lambda\kappa n}t\right) dP_{Z_n}(t)$$

=
$$\int_{-\infty}^{\xi_n} \sigma\sqrt{\lambda\kappa n} \exp\left(\sigma\sqrt{\lambda\kappa n}t\right) P_{Z_n}([t,\xi_n]) dt$$

$$\geq \int_{-\infty}^{\xi_n - \frac{d}{\sqrt{n}}} \sigma \sqrt{\lambda \kappa n} \exp\left(\sigma \sqrt{\lambda \kappa n}t\right) P_g\left(\left[t, \xi_n - \frac{d}{\sqrt{n}}\right]\right) dt$$
$$= \int_{-\infty}^{\xi_n - \frac{d}{\sqrt{n}}} \exp\left(\sigma \sqrt{\lambda \kappa n}t\right) e^{-t^2/2} \frac{dt}{\sqrt{2\pi}}.$$

Because of our assumptions on n, we can write:

$$\mathcal{J} \geq \int_{\xi_n/2}^{2\xi_n/3} \exp\left(\sigma\sqrt{\lambda\kappa n}t\right) e^{-t^2/2} \frac{dt}{\sqrt{2\pi}}$$
$$\geq \frac{\xi_n}{6} \exp\left(\sigma\sqrt{\lambda\kappa n}\xi_n/2\right) \frac{e^{-1/2}}{\sqrt{2\pi}}$$
$$= \frac{\log n}{6\sigma\sqrt{2\pi e\lambda\kappa n}} \exp\left(\frac{\log n}{2}\right)$$
$$= \frac{\log n}{6\sigma\sqrt{2\pi e\lambda\kappa}}.$$

We denote by $n_3(\lambda)$ the smallest integer n such that the previous quantity is larger than 1. It is a non-decreasing function of $\lambda \in [0, 1/2]$.

Eventually, if $\lambda \in [\varepsilon, 1/2]$, then for $n \ge \max(n_1(\varepsilon), n_2(1/2), n_3(1/2))$ the conclusion of Theorem 2' holds for

$$\theta = G(\xi_n) + \frac{\alpha}{\sqrt{n}}$$

As by concavity, $G(t) \leq \frac{1}{2} + \frac{t}{\sqrt{2\pi}}$ for all t positive, one easily deduces the theorem from the previous result. Theorem 2' gives some information only if the quantity $1/2 + c \log n/\sqrt{\varepsilon n}$ is smaller than one. So the condition $n \geq n_1(\varepsilon)$ is implicitely contained in Theorem 2'.

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