# Andrea Colesanti <br> Università di Firenze 

# Around the log-Brunn-Minkowski inequality 

61 probability encounters - In honour of Sergey Bobkov

Toulouse
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## The Brunn-Minkowski inequality

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Minkowski addition. Given $A$ and $B$ in $\mathbb{R}^{n}$, we set:

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Here $|\cdot|$ denotes the Lebesgue measure.

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Remark. $h_{k}$ is a 1-homogeneous convex function. Viceversa, every 1-homogeneous convex function $h: \mathbb{R}^{n} \rightarrow \mathbb{R}$ is the support function of a (unique) convex body.

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If $K$ and $L$ are convex bodies in $\mathbb{R}^{n}$, then

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h_{K+L}=h_{K}+h_{L} \quad \text { and } \quad h_{\alpha K}=\alpha h_{K}
$$

for every $\alpha \geq 0$.

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through:

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Theorem (Firey, 1962) Let $p \geq 1$. For every $K_{0}, K_{1}$, containing the origin, and for every $t \in[0,1]$ :

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- Ingredient 3: a standard argument based on homogeneity of the Lebesgue measure.

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- The main problem when passing to $p \leq 1$ (and $p \geq 0$ ), is that, given two convex bodies $K$ and $L$,

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\left(h_{K}^{p}+h_{L}^{p}\right)^{1 / p}
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is not convex, in general (and then it is not a support function).

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$h_{K[f]}$ is the largest 1-homogeneous convex function smaller than $f$.

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& \underset{\text { log }}{K_{t, 0} \subset \underset{0<p<1}{ } \subset \underset{\text { Minkowski }}{K_{t, p}} \subset \underset{t, 1}{K_{t-p}}} .
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Conjecture (Böröczky, Lutwak, Yang and Zhang, 2012). Let $K_{0}, K_{1} \in \mathcal{K}_{0}^{n}$, be centrally symmetric convex bodies, and let $t \in[0,1]$.

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Remarks

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Remarks

- (log-BM) implies (BM).


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- (log-BM) implies (BM).
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## The log-Brunn-Minkowski conjecture

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- A similar inequality (we will call it ( $p-\mathrm{BM}$ )) can be conjectured for $0<p<1$.

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- $p \in\left[1-\frac{c}{-0.75}, 1\right]$ under an inclusion assumption - Hosle, Kolesnikov, Livshyts, 2020 (with a unifying approach including other inequalities, like Gardner-Zvavitch).

See also:

- Kolesnikov-Livshyts, 2020: On the local version of the log-Brunn-Minkowski conjecture and some related geometric inequalities;
- E. Milman, 2021: A sharp centro-affine isospectral inequality of Szegö-Weinberger type and the $L^{p}$-Minkowski problem.

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See also an unpublished note of Dario Cordero-Erausquin \& Galyna Livshyts.

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verify the assumption of (PL); therefore

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or, equivalently,

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Remark. (*) is stronger than Prékopa-Leindler inequality.

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$\int_{\mathbb{R}^{n}} e^{-\mu_{t}} d z \geq\left(\int_{\mathbb{R}^{n}} e^{-u_{0}}\right)^{1-t}\left(\int_{\mathbb{R}^{n}} e^{-u_{1}}\right)^{t} \Leftrightarrow \underbrace{\left|K_{t}\right| \geq\left|K_{0}\right|^{1-t}\left|K_{1}\right|^{t}}_{\text {(log-BM) }}$.

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\left|K_{0}^{\circ} \cap K_{1}^{\circ}\right| \geq\left|K_{0}^{\circ}\right|^{1-t}\left|K_{1}^{\circ}\right|^{t} & \text { is in general false. }
\end{aligned}
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## Back to the inequality

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\begin{equation*}
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- We compute the infinitesimal form of $\left(^{*}\right)$.


## The first variation (under strong regularity assumptions

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because the r.h.s. and I.h.s. coincide for $t=0$, and the derivative of the r.h.s. is finite at $t=0$.

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(BLYZ, 2012). This suggests that the measure $\mu_{\mu_{0}}$ with density:

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v_{0} e^{v_{0}-\left\langle\nabla v_{0}, y\right\rangle} \operatorname{det}\left(D^{2} v_{0}\right) \quad \text { (in the regular case) }
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could be considered as a functional counterpart of the cone volume measure. Note that

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\mu_{u_{0}}=v_{0} \mu_{\mu_{0}}^{c k}
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where $\mu_{\mu_{0}}^{c k}$ is the moment measure, considered by Cordero-Erasquin and Klartag.

Infinitesimal version - I

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implies that $f$ is log-concave; in particular

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& \int_{\mathbb{R}^{n}} \psi^{2} d \mu+\int_{\mathbb{R}^{n}} \frac{\psi^{2}}{\langle\nabla u, x\rangle-u} d \mu \leq \int_{\mathbb{R}^{n}}\left\langle\left(D^{2} u\right)^{-1} \nabla \psi, \nabla \psi\right\rangle d \mu \\
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where

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Compare with the infinitesimal form of the Prékopa-Leindler inequality (Brascamp-Lieb, 1976):

$$
\int_{\mathbb{R}^{n}} \psi^{2} d \mu \leq \int_{\mathbb{R}^{n}}\left\langle\left(D^{2} u\right)^{-1} \nabla \psi, \nabla \psi\right\rangle d \mu+\left(\int_{\mathbb{R}^{n}} \psi d \mu\right)^{2}
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## The infinitesimal form at the Gaussian

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- Checked for $n=1$.
- Checked for $n$ large enough, with the crucial help of Yaozhong Qiu.

See also: S. Bobkov, M. Ledoux, From Brunn-Minkowski to Brascamp-Lieb and to logarithmic Sobolev inequalities, 2000.

