

# Dimensional Brunn–Minkowski inequalities for the relative entropy

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- Entropy and geometric inequalities
- The dimensional Brunn–Minkowski problem for measures (Main question) and our approach
- Log-concave measures with homogeneous potential
- the Gaussian measure

# The Brunn–Minkowski inequality

## Theorem (Brunn–Minkowski)

Let  $K_0, K_1 \subseteq \mathbb{R}^n$  be compact sets, then

$$\text{Vol}_n((1-t)K_0 + tK_1)^{1/n} \geq (1-t)\text{Vol}_n(K_0)^{1/n} + t\text{Vol}_n(K_1)^{1/n}$$

Here  $(1-t)K_0 + tK_1 = \{(1-t)x + ty : x \in K_0, y \in K_1\}$ .

- Says that the Lebesgue measure is “ $1/n$ -concave”.
- By scaling properties of volume, this is equivalent to the a priori weaker 0-concavity:  
$$\text{Vol}_n((1-t)K_0 + tK_1) \geq \text{Vol}_n(K_0)^{1-t}\text{Vol}_n(K_1)^t.$$
- How to prove this using “entropy”?

# Step 1: find a suitable notion of entropy to encode volume as maximum entropy

## Definition

Let  $\mu$  be a probability measure on  $\mathbb{R}^n$  having density  $f$  w.r.t. the Lebesgue measure. Then, the *Shannon-Boltzmann entropy* is defined by

$$h(\mu) = - \int f \log f \, dx.$$

- If  $\mu$  does not have density, can be meaningfully set to be  $-\infty$ .
- Sometimes we write  $h(f)$  for  $h(\mu)$ , as well as  $h(X)$ , if  $X$  is a random vector with distribution  $\mu$ .
- Thought of as a measure of how spread out  $\mu$  is.

## Lemma

Let  $K \subseteq \mathbb{R}^n$  be compact, then

$$\sup_{\mu \in \mathcal{P}(K)} h(\mu) = \log \text{Vol}_n(K).$$

### Proof:

- If  $X \sim \mu$  has density  $f$ ,

$$h(\mu) = - \int f \log f \, dx = \mathbb{E} \log \frac{1}{f(X)} \leq \log \mathbb{E} \frac{1}{f(X)} \leq \log \text{Vol}_n(K).$$

- Equality if  $X \sim \text{Uniform}(K)$ .

Therefore, if we want to prove

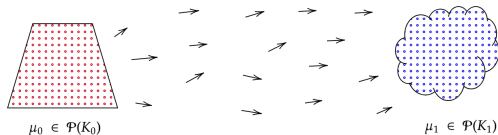
$$\log \text{Vol}_n((1-t)K_0 + tK_1) \geq (1-t) \log \text{Vol}_n(K_0) + t \log \text{Vol}_n(K_1),$$

we can do the following:

For every  $\mu_0 \in \mathcal{P}(K_0)$  and every  $\mu_1 \in \mathcal{P}(K_1)$  find an interpolation  $\mu_t \in \mathcal{P}((1-t)K_0 + tK_1)$ , such that

$$h(\mu_t) \geq (1-t)h(\mu_0) + th(\mu_1).$$

## Step 2: find a suitable interpolation to show concavity of entropy chosen earlier



- We think of probability distributions as distributions (or configurations) of the molecules of a gas in space.
- Look at all time-dependent vector fields  $v_t$  which displace  $\mu_0$  to  $\mu_1$  between  $t = 0$  and  $t = 1$ . Each choice of  $v_t$  corresponds to a family  $\mu_t$  representing the intermediate distribution at time  $t$  along the flow.
- Pick the one which minimises the action  $\int_0^1 \int |v_t|^2 d\mu_t dt$ , where  $\mu_t$  denotes the intermediate distribution at time  $t$  along the flow induced by  $v_t$ .

- The interpolation  $\{\mu_t\}$  corresponding to the action-minimising  $v_t$  is called the *displacement interpolation*.
- It turns out that the optimal vector field is of the form  $\nabla\theta_t$ , and displaces every molecule with constant speed along a straight line, leading to  $\mu_t \in \mathcal{P}((1-t)K_0 + tK_1)$ .

Two useful properties are:

- *Continuity equation*: For every test function  $\phi$ ,

$$\frac{d}{dt} \int \phi \, d\mu_t = \int \langle \nabla\phi, \nabla\theta_t \rangle \, d\mu_t.$$

- *The Hamilton–Jacobi equation*,

$$\frac{\partial\theta_t}{\partial t} + \frac{|\nabla\theta_t|^2}{2} = 0.$$



# An equivalent way to describe things

Let  $\mu_0, \mu_1 \in \mathcal{P}_{2,ac}(\mathbb{R}^n)$ , and consider the *Monge problem* of minimising

$$I = \int |x - T(x)|^2 d\mu_0(x),$$

over all maps  $T$  such that  $T_{\#}\mu_0 = \mu_1$ .

- (Brenier '91, McCann '97) The minimum is uniquely attained by a map  $T = \nabla\psi$ , where  $\psi$  is convex. The displacement interpolation is  $\mu_t = T_{t\#}\mu_0$ , where  $T_t = (1-t)I + tT$ .
- The time-dependent velocity field from before solves,

$$\nabla\theta_t(T_t(x)) = \frac{d}{dt}T_t(x).$$

# Displacement concavity of entropy

## Theorem (McCann '97)

*The Shannon–Boltzmann entropy is displacement concave:*

$$h(\mu_t) \geq (1 - t)h(\mu_0) + th(\mu_1),$$

*for all displacement interpolations.*

**Proof:** Change of variables obtained by the description of  $\mu_t$  as  $T_{t\#}\mu_0$ .

## Corollary

*For all compact  $K_0, K_1 \subseteq \mathbb{R}^n$ , we have*

$$\text{Vol}_n((1 - t)K_0 + tK_1) \geq \text{Vol}_n(K_0)^{1-t}\text{Vol}_n(K_1)^t.$$

# How to obtain the dimensional form directly?

- (McCann '97) The functional  $\mu \mapsto \int \frac{d\mu}{dx}^{1-\frac{1}{n}} dx$  is displacement concave. Or,

## Theorem (Erbar-Kuwada-Sturm, '15)

*The functional  $e^{h(\cdot)/n}$  is displacement concave:*

$$e^{h(\mu_t)/n} \geq (1-t)e^{h(\mu_0)/n} + te^{h(\mu_1)/n}.$$

- While they prove something much<sup>k</sup> more general, the argument specialised to obtain the above is very simple.
- In the language of random vectors, after some change of variables, this implies  $e^{h(X+Y)/n} \geq e^{h(X)/n} + e^{h(Y)/n}$ , when  $(X, Y)$  are coupled so that it minimise  $\mathbb{E}|X - Y|^2$ .
- Compare with the Entropy Power Inequality:  
 $e^{2h(X+Y)/n} \geq e^{2h(X)/n} + e^{2h(Y)/n}$ , when  $X, Y$  are independent.

# Which problem do we want approach using these ideas?

The question of  $a$ -concavity of Borel measures  $\nu$  on  $\mathbb{R}^n$ :

$$\nu((1-t)K_0 + tK_1) \geq ((1-t)\nu(K_0)^a + t\nu(K_1)^a)^{1/a},$$

is relatively well understood.

- (Borell, '75) We know this is equivalent to  $b$ -concavity of the density of  $\mu$ , where  $\frac{1}{a} = \frac{1}{b} + n$ .
- (EKS, '15) We know this is equivalent to  $(0, 1/a)$ -convexity of the relative entropy w.r.t.  $\nu$ .

**Broad question:** Do concavity properties of a measure improve when  $K_0, K_1$  are restricted to some sub-class of compact sets?

**In particular:** Can 0-concavity (i.e., log-concavity) for an even measure improve when restricted to sets with some convexity and symmetry properties?

## Some developments on this theme

- (Livshyts-Marsiglietti-Nayar-Zvavitch, '17) The influential Log-Brunn–Minkowski conjecture of Böröczky, Lutwak, Yang and Zhang, implies that

$$\nu((1-t)K_0 + tK_1)^{1/n} \geq (1-t)\nu(K_0)^{1/n} + t\nu(K_1)^{1/n},$$

holds for every even log-concave measure  $\nu$  and symmetric convex bodies  $K_0, K_1$ .

- The  $1/n$ -concavity question for the Gaussian measure specifically was asked by Gardner and Zvavitch ('10) when  $0 \in K_0 \cap K_1$ . Nayar and Tkocz ('12) showed that this is not true, and suggested central symmetry may be the correct condition.

General strategy: (initiated by Kolesnikov-Milman, '18, '22)

- Try to prove  $\frac{d^2}{dt^2} (\mu((1-t)K + tL)^a) \leq 0$  by directly computing this second derivative.
- Enough to check at  $t = 0$ , which gives a functional inequality for  $f : \partial K \rightarrow \mathbb{R}$ .
- Transform this inequality to an inequality for functions  $u : K \rightarrow \mathbb{R}$  by taking  $u$  to be the solution of a certain elliptic PDE with  $f$  as its Neumann boundary condition.
- As far as we can tell, the convexity of  $K, L$  is crucial.

- (Kolesnikov-Livshyts, '21) Obtained  $1/2n$ -concavity,  $\gamma_n((1-t)K_0 + tK_1)^{\frac{1}{2n}} \geq (1-t)\gamma_n(K_0)^{\frac{1}{2n}} + t\gamma_n(K_1)^{\frac{1}{2n}}$ , for all convex sets  $K_0, K_1 \subseteq \mathbb{R}^n$  containing 0.
- (Eskenazis-Moschidis, '21) Using a sufficiency criteria obtained in the above work, obtained  $1/n$ -concavity of the Gaussian measure when restricted to origin symmetric convex sets.
- (Cordero-Erasquin and Rotem, '23)  $1/n$ -concavity for Rotationally-invariant log-concave measures and symmetric convex sets
- (Livshyts, '21)  $\frac{1}{n^{4+o(1)}}$ -concavity for all even log-concave measures and symmetric convex sets.

# Our Approach

Let  $\nu$  be a log-concave measure on  $\mathbb{R}^n$ .

Step 1: Find the right notion of entropy to encode the  $\nu$ -measure as maximum entropy.

Step 2: Prove the right displacement concavity property for this entropy.

Main difference: Differentiate entropy instead of differentiating the  $\nu$ -measure of interpolating convex bodies.



## Step 1: Relative entropy

### Definition (Relative entropy)

Let  $\nu$  be a  $\sigma$ -additive Borel measure on  $\mathbb{R}^n$ . We define the *relative entropy of  $\mu$  with respect to  $\nu$*  by,

$$D(\mu\|\nu) = \begin{cases} \int f \log f \, d\nu, & \text{if } \mu \text{ has density } f \text{ w.r.t. } \nu, \\ +\infty, & \text{otherwise.} \end{cases}$$

- $D(\mu\|\nu)$  quantifies how much  $\mu$  is “spread out” from the viewpoint of  $\nu$ .
- We get an absolute measure of spread by looking at negative the amount  $\mu$  is spread out from the most spread out measure  $\text{Vol}_n$ ,  $h(\mu) = -D(\mu\|\text{Vol}_n)$ .

## Lemma

Let  $K \subseteq \mathbb{R}^n$  be a compact set, then

$$\sup_{\mu \in \mathcal{P}(K)} e^{-D(\mu \| \nu)} = \nu(K)$$

## Proof:

- If  $X \sim \mu \in \mathcal{P}(K)$  has density  $f$  w.r.t.  $\nu$ ,

$$-D(\mu \| \nu) = - \int f \log f \, d\nu = \mathbb{E} \log \frac{1}{f(X)} \leq \log \mathbb{E} \frac{1}{f(X)} \leq \log \nu(K).$$

- Equality if  $\frac{d\mu}{d\nu}$  is constant, that is, if  $\mu(\cdot) = \nu_K(\cdot) = \frac{\nu(\cdot \cap K)}{\nu(K)}$ .

- Suppose  $S \subseteq \mathcal{P}_2(\mathbb{R}^d)$  is a displacement convex set, that is,  $\mu_0, \mu_1 \in S$  implies  $\{\mu_t\} \subseteq S$ .
- Let  $\nu$  be a  $\sigma$ -finite Borel measure on  $\mathbb{R}^n$ . Assume that  $\{\nu_K : K \in \mathcal{K}\} \subseteq S$ , for some class  $\mathcal{K}$  of compact sets.
- If  $e^{-aD(\cdot\|\nu)}$  is displacement concave,  $a > 0$ , on  $S$ :

$$e^{-aD(\mu_t\|\nu)} \geq (1-t)e^{-aD(\mu_0\|\nu)} + te^{-aD(\mu_1\|\nu)};$$

then,

$$\nu((1-t)K_0 + tK_1)^a \geq (1-t)\nu(K_0)^a + t\nu(K_1)^a,$$

for all  $K_0, K_1 \in \mathcal{K}$ .

# Idea 1: Decompose relative entropy, work on pieces

- Suppose  $\nu = e^{-V} dx$ ,  $V$  convex. (that is,  $\nu$  is log-concave with potential  $V$ )

- $D(\mu\|\nu) = -h(\mu) + \mathcal{V}(\mu)$ , where  $\mathcal{V}(\mu) = \int V d\mu$ .

- Thus,

$$e^{-aD(\cdot\|\nu)} = \underbrace{e^{ah(\cdot)}}_{\text{know}} \underbrace{e^{-a\mathcal{V}(\cdot)}}_{\text{want}}.$$

- Concavity of  $e^{-a\mathcal{V}(\cdot)}$  is equivalent to

$$\frac{d^2}{dt^2} \mathcal{V}(\mu_t) \geq a \left( \frac{d}{dt} \mathcal{V}(\mu_t) \right)^2.$$

- These derivatives can be calculated in terms of the velocity-field  $\nabla\theta_t$  associated with the displacement interpolation  $\{\mu_t\}_{t\in[0,1]}$ :

$$\frac{d}{dt}\mathcal{V}(\mu_t) = \int \langle \nabla V, \nabla\theta_t \rangle d\mu_t,$$

$$\frac{d^2}{dt^2}\mathcal{V}(\mu_t) = \int \langle \nabla^2 V \cdot \nabla\theta_t, \nabla\theta_t \rangle d\mu_t.$$

- These quantities can be related by a Hölder-type inequality,

$$\begin{aligned} \left( \frac{d}{dt}\mathcal{V}(\mu_t) \right)^2 &= \left( \int \langle \nabla\theta_t, \nabla V \rangle d\mu_t \right)^2 \\ &\leq \int \langle \nabla^2 V \cdot \nabla\theta_t, \nabla\theta_t \rangle d\mu_t \cdot \int \langle (\nabla^2 V)^{-1} \cdot \nabla V, \nabla V \rangle d\mu_t \\ &= \left( \frac{d^2}{dt^2}\mathcal{V}(\mu_t) \right) \cdot \int \langle (\nabla^2 V)^{-1} \cdot \nabla V, \nabla V \rangle d\mu_t \end{aligned}$$

- Therefore, displacement concavity of  $e^{-aV}$  on  $S$  can be obtained by showing  $\int \langle (\nabla^2 V)^{-1} \cdot \nabla V, \nabla V \rangle d\mu \leq \frac{1}{a}$  on  $S$ .
- When  $V$  is  $p$ -homogeneous, that is, when  $V(\lambda x) = \lambda^p V(x)$  for all  $\lambda > 0$ , the quantity we need to bound becomes simpler allowing several computations.
- In this case, set  $S = \{\mu : \mathcal{V}(\mu) \leq \frac{n}{p}\}$ .
- This set is displacement concave and contains all measures whose density with respect to  $\nu$  is radially decreasing.

### Theorem (A-Rotem, '23+)

Let  $V : \mathbb{R}^n \rightarrow [0, \infty)$  be a  $p$ -homogeneous convex function,  $p \in (1, \infty)$ . Let  $d\nu = e^{-V+c} dx \in \mathcal{P}(\mathbb{R}^n)$ , for some constant  $c$ . Then, the functional  $e^{-\frac{p-1}{n}\mathcal{V}(\cdot)}$ , where  $\mathcal{V}(\mu) = \int V d\mu$ , is displacement concave on  $S$ .

# Putting things together

- So now we know  $e^{\frac{h(\cdot)}{n}}$  and  $e^{-\frac{p-1}{n}\mathcal{V}(\cdot)}$  are both displacement concave on  $S$ .
- $e^{-\frac{p-1}{pn}D(\cdot\|\nu)} = e^{\frac{p-1}{pn}c} \left(e^{\frac{h(\cdot)}{n}}\right)^{1-\frac{1}{p}} \left(e^{-\frac{p-1}{n}\mathcal{V}(\cdot)}\right)^{\frac{1}{p}}$  is a geometric-mean of concave functions, hence is itself concave.

## Theorem (A-Rotem, '23+)

*Let  $V : \mathbb{R}^n \rightarrow [0, \infty)$  be a  $p$ -homogeneous convex function,  $p \in (1, \infty)$ . Let  $d\nu = e^{-V+c} dx \in \mathcal{P}(\mathbb{R}^n)$ , for some constant  $c$ . Then, the functional  $e^{-\frac{p-1}{pn}D(\cdot\|\nu)}$  is displacement concave on the set  $S$ .*

# A dimensional Brunn–Minkowski for star bodies

Since every  $\nu_K$ , for a star-shaped body  $K$  (i.e.,  $[0, 1]K \subseteq K$ ), has a radially decreasing density with respect to  $\nu$ , we have:

## Theorem (A-Rotem, 23+)

Let  $V : \mathbb{R}^n \rightarrow [0, \infty)$  be a  $p$ -homogeneous convex function for  $1 < p < \infty$ , and let  $d\nu = e^{-V} dx$ . Suppose  $K_0, K_1 \subseteq \mathbb{R}^n$  are star bodies. Then for all  $0 \leq t \leq 1$  we have

$$\nu((1-t)K_0 + tK_1)^{\frac{p-1}{pn}} \geq (1-t)\nu(K_0)^{\frac{p-1}{pn}} + t\nu(K_1)^{\frac{p-1}{pn}}.$$

## Corollary

$\gamma_n((1-t)K_0 + tK_1)^{\frac{1}{2n}} \geq (1-t)\gamma_n(K_0)^{\frac{1}{2n}} + t\gamma_n(K_1)^{\frac{1}{2n}}$ , for all star-shaped compact sets  $K_0, K_1 \subseteq \mathbb{R}^n$ .



## Idea 2: work directly with the relative entropy

Formally, we can directly differentiate the relative entropy itself.

- Consider a measure  $\nu$  having density  $e^{-V}$  with respect to the Lebesgue measure, and the Markov semigroup generated by  $L = \Delta - \langle \nabla V, \nabla \rangle$ .
- Denote by  $\Gamma, \Gamma_2$  the carré du champ operator and its iteration, respectively, of this semigroup.

Then,

$$\begin{aligned}\frac{d}{dt} D(\mu_t \| \nu) &= - \int L\theta_t \, d\mu_t, \\ \frac{d^2}{dt^2} D(\mu_t \| \nu) &= \int \Gamma_2(\theta_t) \, d\mu_t,\end{aligned}$$

where  $\nabla\theta_t$  is the velocity-field of  $\{\mu_t\}$ .

To obtain the displacement concavity  $e^{-aD(\cdot\|\nu)}$  on a set  $S$ , we need to prove

$$\int \Gamma_2(\theta_t) \, d\mu_t \geq a \left( \int L\theta_t \, d\mu_t \right)^2.$$

- Using exponentiated relative entropy (as we do) is perhaps better than using Rényi entropy when working with a proper subset  $S$ .
- Remarkably similar to the Livshyts-Kolesnikov criteria for  $\nu = \gamma_n$ .

# The Gaussian case

Let  $S_n \subseteq \mathcal{P}_2(\mathbb{R}^n)$  be the collection of all even measures  $\mu$  satisfying the Poincaré inequality

$$\int f^2 \, d\mu \leq \int |\nabla f|^2 \, d\mu,$$

for all **odd** test functions  $f$ .

## Theorem (A-Rotem, '23+)

*Suppose the displacement interpolation  $\{\mu_t\}_{t \in [0,1]}$  completely lies in  $S_n$ , then  $e^{-D(\mu_t \parallel \gamma_n)}$  is concave in  $t$ . Moreover,  $S_1 \subseteq \mathcal{P}_2(\mathbb{R}^1)$  is itself displacement convex, thus  $e^{-D(\cdot \parallel \gamma)}$  is displacement concave on  $S_1$ .*

- The proof of the first part uses an idea very similar to the one used by Eskenazis and Moschidis.

# What do we know about the contents of $S_n$

We know that  $S_n$  contains:

- 1 All even measures whose density is log-concave with respect to the standard Gaussian (that is, even strongly log-concave measures). This class corresponds to origin-symmetric convex bodies.
- 2 The displacement convex set consisting of all Gaussians with covariance dominated by the identity matrix.
- 3 All displacement interpolation  $\{\mu_t\}_{t \in [0, \infty]}$  such that one endpoint is the Gaussian, and the other endpoint is even and strong log-concave.

- Is  $S_n \subseteq \mathcal{P}_2$  displacement convex in general?
- Does  $S_n$  contain a displacement convex set containing all even strongly log-concave probability measures?

THANK YOU!