

# Convexity in High Dimensions

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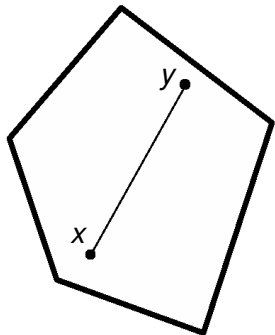
# The Poincaré inequality

## Theorem (Poincaré, 1890 and 1894)

Let  $K \subseteq \mathbb{R}^3$  be convex and open.  
Let  $f : K \rightarrow \mathbb{R}$  be  $C^1$ -smooth, with  
 $\int_K f = 0$ . Then,

$$\lambda_K \int_K f^2 \leq \int_K |\nabla f|^2$$

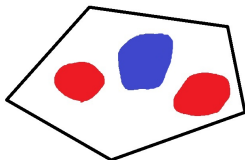
where  $\lambda_K \geq (16/9) \cdot \text{Diam}^{-2}(K)$ .



- In 2D, Poincaré got a better constant,  $24/7$ .
- Related to Wirtinger's inequality on periodic functions in one dimension (sharp constant, roughly a decade later).
- The largest possible  $\lambda_K$  is the Neumann spectral gap or the inverse **Poincaré constant** of  $K$ .
- Proof: Estimate  $\int_{K \times K} |f(x) - f(y)|^2 dx dy$  via segments.

# Motivation: The heat equation

- Suppose  $K \subseteq \mathbb{R}^3$  with  $\partial K$  an 'insulator', i.e., heat does not escape/enter  $K$ .
- Write  $u_t(x)$  for the temperature at the point  $x \in K$  at time  $t \geq 0$ .



## Heat equation (Neumann boundary conditions)

$$\begin{cases} \dot{u}_t = \Delta u_t & \text{in } K \\ \frac{\partial u_t}{\partial n} = 0 & \text{on } \partial K \end{cases}$$

Fourier's law: Heat flux is proportional to the temp. gradient.

## Rate of convergence to equilibrium

$$\frac{1}{|K|} \int_K u_0 = 1 \quad \implies \quad \|u_t - 1\|_{L^2(K)} \leq e^{-t\lambda_K} \|u_0 - 1\|_{L^2(K)}$$

The Poincaré inequality was generalized to all dimensions:

**Theorem (Payne-Weinberger, 1960 – precursor for convex localization)**

Let  $K \subseteq \mathbb{R}^n$  be convex and open, let  $\mu$  be the Lebesgue measure on  $K$ . If  $f : K \rightarrow \mathbb{R}$  is  $C^1$ -smooth with  $\int_K f d\mu = 0$ , then,

$$\frac{\pi^2}{\text{Diam}^2(K)} \int_K f^2 d\mu \leq \int_K |\nabla f|^2 d\mu.$$

- The constant  $\pi^2$  is best possible in every dimension  $n$ .  
E.g.,

$$K = (-\pi/2, \pi/2), \quad f(x) = \sin(x).$$

- In contrast, Poincaré's proof would lead to an exponential dependence on the dimension.
- Not only the Lebesgue measure on  $K$ , we may consider any log-concave measure.

The Poincaré inequality was generalized to all dimensions:

**Theorem (Payne-Weinberger, 1960 – precursor for convex localization)**

Let  $K \subseteq \mathbb{R}^n$  be convex and open, let  $\mu$  be any **log-concave** measure on  $K$ . If  $f : K \rightarrow \mathbb{R}$  is  $C^1$ -smooth with  $\int_K f d\mu = 0$ , then,

$$\frac{\pi^2}{\text{Diam}^2(K)} \int_K f^2 d\mu \leq \int_K |\nabla f|^2 d\mu.$$

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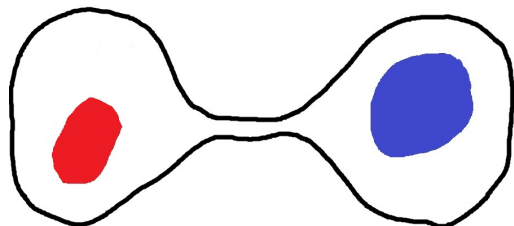
- In contrast, Poincaré's proof would lead to an exponential dependence on the dimension.
- A **log-concave** measure  $\mu$  on  $K$  is a measure with density of the form  $e^{-H}$ , where the function  $H$  is **convex**.

# The role of convexity / log-concavity

- For  $\Omega \subseteq \mathbb{R}^n$ , the Poincaré coefficient  $\lambda_\Omega$  measures the connectivity or conductance of  $\Omega$ .

Convexity is a strong form of connectedness

Without convexity/log-concavity assumptions:



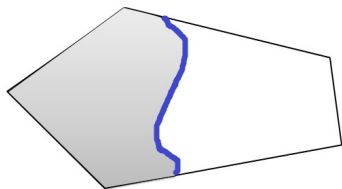
*long time to reach equilibrium,  
regardless of the diameter*

# Many other ways to measure connectivity

## The isoperimetric constant

For an open set  $K \subset \mathbb{R}^n$  define

$$h_K = \inf_{A \subset K} \frac{|\partial A \cap K|}{\min\{|A|, |K \setminus A|\}}$$



- If  $K$  is strictly-convex with smooth boundary, the infimum is attained when  $|A| = |K|/2$  (Sternberg-Zumbrun, 1999).

## Theorem (Cheeger '70, Buser '82, Ledoux '04)

For any open, convex set  $K \subseteq \mathbb{R}^n$ ,

$$\frac{h_K^2}{4} \leq \lambda_K \leq 9h_K^2.$$

- Mixing time of Markov chains, algorithms for estimating volumes of convex bodies (Dyer-Freeze-Kannan '89, ...)

# The study of uniform distributions on convex domains

What is this mathematical field about?

(relevant to linear programming, statistics, computer science, slightly to stat. physics)

- The questions are non-trivial in high dimensions, when

$$n \rightarrow \infty.$$

## The theme

Convexity in high dimension is a source of regularity, comparable to statistical independence. It may compensate for lack of structure or symmetry.

- Playground for various techniques that transcend convex geometry: Convex Localization, Optimal Transport, Bochner identities and curvature, heat flow and Eldan's stochastic localization, geometric measure theory.



# The simplest example: the Euclidean ball

- Consider the Euclidean ball  $B^n = \{x \in \mathbb{R}^n; |x| \leq 1\}$  or the sphere  $S^{n-1} = \{x \in \mathbb{R}^n; |x| = 1\}$ .
- Write  $\sigma_{n-1}$  for the uniform probability measure on  $S^{n-1}$ .

For a set  $A \subseteq S^{n-1}$  and for  $\varepsilon > 0$  denote

$$A_\varepsilon = \left\{ x \in S^{n-1}; \exists y \in A, d(x, y) \leq \varepsilon \right\},$$

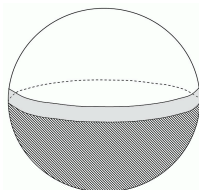
which is the  $\varepsilon$ -neighborhood of  $A$ .

- Consider the hemisphere  $H = \{x \in S^{n-1}; x_1 \leq 0\}$ .

## A well-known observation:

$$\sigma_{n-1}(H_\varepsilon) = \mathbb{P}(Y_1 \leq \sin \varepsilon) \approx \mathbb{P}(\Gamma \leq \varepsilon\sqrt{n})$$

where  $Y = (Y_1, \dots, Y_n)$  is distributed according to  $\sigma_{n-1}$ , and  $\Gamma$  is a standard normal random variable.



*The sphere's marginals are approximately Gaussian*

# Concentration of measure on the sphere

The amount of volume of a distance at least  $1/10$  from the equator is at most

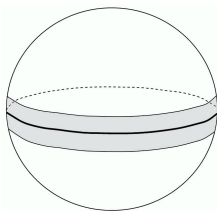
$$C \exp(-cn)$$

of the sphere, for universal constants  $c, C > 0$ .

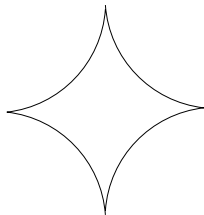
- Most of the mass of the sphere  $S^{n-1}$  in high dimensions, is concentrated at a narrow strip near the equator  $[x_1 = 0]$
- or any other equator.

**“Concentration of Measure”**

(à la V. Milman)



$\dim \rightarrow \infty$



# The isoperimetric inequality on the sphere

## The isoperimetric inequality (Lévy, Schmidt, '50s).

For any Borel set  $A \subset S^{n-1}$  and  $t, \varepsilon > 0$ ,

$$\sigma_{n-1}(A) = t \implies \sigma_{n-1}(A_\varepsilon) \geq \sigma_{n-1}(C_\varepsilon),$$

where  $C \subseteq S^{n-1}$  is a spherical cap of measure  $t$  (when  $t = 1/2$  it is a hemisphere).

- For any set  $A \subset S^{n-1}$  with  $\sigma_{n-1}(A) = 1/2$ ,

$$\sigma_{n-1}(A_\varepsilon) \geq 1 - 2 \exp(-\varepsilon^2 n/2).$$

Therefore, for any subset  $A \subset S^{n-1}$  of measure  $1/2$ , its  $\varepsilon$ -neighborhood covers almost the entire sphere.

- The isoperimetric inequality is “accompanied” by functional inequalities on the sphere: Sobolev, log-Sobolev, transport-cost, Poincaré and Brunn-Minkowski type.

## Corollary (“Lévy’s lemma”)

Let  $f : S^{n-1} \rightarrow \mathbb{R}$  be a 1-Lipschitz function. Denote

$$E = \int_{S^{n-1}} f(x) d\sigma_{n-1}(x).$$

Then, for any  $\varepsilon > 0$ ,

$$\sigma_{n-1} \left( \left\{ x \in S^{n-1}; |f(x) - E| \geq \varepsilon \right\} \right) \leq C \exp(-c\varepsilon^2 n),$$

where  $c, C > 0$  are universal constants.

- Lipschitz functions on the high-dimensional sphere are “effectively constant”.

# Concentration phenomena for general convex bodies?

- Does the uniform probability measure on an arbitrary convex body  $K \subseteq \mathbb{R}^n$  exhibit similar effects?

## A short answer

Pretty much, yes.

(Currently known up to polylogarithmic factors).

- There are rather strong **expansion properties**, and the isoperimetric inequality is nearly saturated by **half-spaces**.

## Caveat: exponential tail rather than subgaussian tail

There is no strong small-set-expansion as in the sphere.

The tail distribution of a 1-Lipschitz function is sub-exponential, rather than sub-gaussian.

# How would you normalize a convex body?

There is a unit ball  $B^n$  and a unit cube  $[0, 1]^n$ .

- How would you choose the correct “units” for a general convex body or a log-concave density?

## Definition

A convex body  $K \subseteq \mathbb{R}^n$  of volume one is **isotropic** if for the random vector  $X$  distributed uniformly on  $K$ ,

- 1  $\mathbb{E}X = 0$
  - 2 The covariance matrix  $\text{Cov}(X) \in \mathbb{R}^{n \times n}$  is a scalar matrix, where  $\text{Cov}(X)_{ij} = \mathbb{E}X_i X_j - (\mathbb{E}X_i)(\mathbb{E}X_j)$ .
- Any convex body can be made isotropic after applying a linear-affine transformation.
  - Another common, similar normalization: Rather than requiring  $\text{Vol}_n(K) = 1$ , require  $\text{Cov}(X) = \text{Id}$ .

# The Kannan-Lovász-Simonovits (KLS '95) conjecture

Let  $K \subseteq \mathbb{R}^n$  be an isotropic convex body of volume one (or of identity covariance),  $X$  distributed uniformly over  $K$ .

A conjecture (KLS), which has three equivalent formulations:

- 1 **Concentration:** For any 1-Lipschitz function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  with  $\mathbb{E}f(X) = 0$ ,

$$\mathbb{P}(|f(X)| \geq t) \leq Ce^{-ct} \quad \text{for all } t > 0.$$

- 2 **Poincaré inequality:** For any smooth  $f$  with  $\int_K f = 0$ ,

$$\int_K f^2 \leq C \int_K |\nabla f|^2$$

- 3 **Isoperimetry:** For any open  $A \subseteq \mathbb{R}^n$ , smooth boundary,

$$\text{Vol}_{n-1}(\partial A \cap K) \geq c \cdot \min\{\text{Vol}_n(K \cap A), \text{Vol}_n(K \setminus A)\}.$$

Here  $c, C > 0$  are universal constants.

# The Kannan-Lovász-Simonovits (KLS '95) conjecture

Equivalences in our formulation of the KLS conjecture due to Ball-Nguyen '13, Buser '82, Cheeger '70, Eldan-K. '11, Gromov and V. Milman '83, Ledoux '04, E. Milman '09.

Theorem (K.-Lehec '22, building upon Eldan '13, Lee-Vempala '16 and the breakthrough by Chen '20)

*KLS conjecture is true up to polylogarithmic factors.*

*i.e., replace the universal constants by  $C \log^\alpha n$ .*

- Bounds for the  $\alpha$ 's were improved a few weeks ago by LV+Jambulapati.
- Works for log-concave random vectors too.

Example: Apply the KLS Poincaré inequality with  $f(x) = |x| - E$

If  $X$  is an isotropic log-concave random vector in  $\mathbb{R}^n$ , then

$$\text{Var}(|X|) \leq C \log^\alpha n$$



# Thin shell phenomenon

- When  $X$  is isotropic,  $\mathbb{E}|X|^2 \approx n$  and by reverse Hölder inequalities

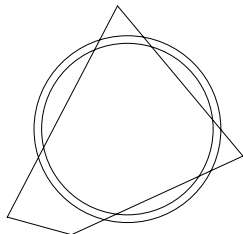
$$\mathbb{E}|X| \approx \sqrt{n}.$$

However, the variance of  $|X|$  is  $\leq C \log^\alpha n$ , much smaller!

Most of the mass of  $X$  is contained in a **thin spherical shell**.

- **Sudakov '76,**  
**Diaconis-Freedman '84:**

When most of the mass of the isotropic random vector  $X$  is contained in a thin spherical shell, we have **approx. Gaussian marginals**.



# Convexity as good as Independence

Current state of the art:

**Theorem (Bobkov, Chistyakov, Götze '19 using K.-Lehec '22)**

Let  $X$  be an isotropic, log-concave random vector in  $\mathbb{R}^n$ . Then there exists  $\Theta \subseteq S^{n-1}$  with  $\sigma_{n-1}(\Theta) \geq 9/10$  such that for any  $\theta \in \Theta$ ,

$$\sup_{t \in \mathbb{R}} \left| \mathbb{P}(X \cdot \theta \leq t) - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^t e^{-s^2/2} ds \right| \leq \frac{C \log^\alpha n}{n},$$

where  $C, \alpha > 0$  are universal constants.

- An optimal result for the Kolmogorov metric, up to the polylogarithmic factor. (No log's for independent random variables with bounded  $4^{\text{th}}$  moments, K.-Sodin '11)
- This **Central Limit Theorem for Convex Sets** was originally proven in K. '07, confirming a conjecture by Anttila, Ball and Perissinaki '03 and Brehm-Voigt '00.

# No log-factors under symmetry assumptions

- A random vector  $X = (X_1, \dots, X_n)$  is said to be invariant under coordinate reflections if it is equidistributed with  $(\pm X_1, \dots, \pm X_n)$  for any choice of signs.
- For instance, the uniform measure on  $\ell_p^n$ -balls.

## Theorem (K. '09)

Let  $X = (X_1, \dots, X_n)$  be a log-concave random vector in  $\mathbb{R}^n$ , invariant under coordinate reflections, with  $\mathbb{E}X_i^2 = 1$  for all  $i$ . Then for any  $\theta_1, \dots, \theta_n \in \mathbb{R}$  with  $\sum_i \theta_i^2 = 1$ ,

$$\sup_{\alpha \leq \beta} \left| \mathbb{P} \left( \alpha \leq \sum_{i=1}^n \theta_i X_i \leq \beta \right) - \frac{1}{\sqrt{2\pi}} \int_{\alpha}^{\beta} e^{-t^2/2} dt \right| \leq C \sum_{i=1}^n \theta_i^4.$$

- The bound is typically  $O(1/n)$ , since for a random  $\theta \in S^{n-1}$ , we have  $\sum_i \theta_i^4 \leq C/n$  with prob. at least  $1 - C \exp(-c\sqrt{n})$ .

We have the **thin-shell parameter**

$$\sigma_n^2 = \sup_X \mathbb{E} (|X| - \sqrt{n})^2,$$

where the sup is over isotropic, log-concave random vectors in  $\mathbb{R}^n$ . A trivial bound is  $\sigma_n \leq C\sqrt{n}$ . Better estimates:

$$\sigma_n \leq Cn^{2/5+o(1)} \quad (\text{K. '07})$$

$$\sigma_n \leq Cn^{3/8} \quad (\text{Fleury '09})$$

$$\sigma_n \leq Cn^{1/3} \quad (\text{Guédon-E. Milman '11})$$

$$\sigma_n \leq Cn^{1/4} \quad (\text{Lee-Vempala '16})$$

$$\sigma_n \leq C \exp\left((\log n)^{1/2+o(1)}\right) = n^{o(1)} \quad (\text{Chen '20})$$

$$\sigma_n \leq C \log^4 n \quad (\text{K.-Lehec '22})$$

$$\sigma_n \leq C \log^{2.23} n \quad (\text{Jambulapati-L.-V. '22+})$$

- Concentration of measure (hinted by Paouris LDP '06).

## Definition

Write  $\psi_n$  for the minimal number such that for any log-concave random vector  $X$ ,  $\text{Cov}(X) = \text{Id}$ , and any 1-Lipschitz function  $\varphi$ ,

$$\text{Var} \varphi(X) \leq \psi_n^2.$$

- Alternatively, if  $\rho$  is the density of  $X$ , then for any  $A \subseteq \mathbb{R}^n$  with smooth boundary,

$$\int_{\partial A} \rho \geq \frac{C}{\psi_n} \cdot \min\{\mathbb{P}(X \in A), \mathbb{P}(X \notin A)\}.$$

- We actually explained that

$$\sigma_n \leq C\psi_n.$$

Theorem (Eldan '13 – breakthrough towards KLS conjecture)

$$\psi_n \leq C \log n \cdot \sigma_n$$

# Thin Shell and Slicing

In the 1980s, Bourgain considered the following:

## Definition

Write  $L_n$  for the minimal number such that for any convex body  $K \subseteq \mathbb{R}^n$  of volume one, there exists a hyperplane  $H \subseteq \mathbb{R}^n$  such that

$$\text{Vol}_{n-1}(K \cap H) \geq \frac{1}{L_n}.$$

- **The hyperplane conjecture:**  $L_n \leq C$ .  
Any convex body of volume one should have a hyperplane section whose volume is at least  $c$ .
- We are still waiting for a short and sweet proof, but it is not coming...

## Theorem (Eldan-K. '11)

$$L_n \leq C\sigma_n \leq C'\psi_n$$

# The slicing problem

- The relation between Slicing and KLS was promoted by K. Ball, and Ball-Nguyen '13 proved  $L_n \leq \exp(C\psi_n^2)$ .

Corollary (because “thin shell implies slicing”)

$L_n \leq C \log^\alpha n$  for a universal constant  $C > 0$  and  $\alpha \leq 2.23$

- For many years, the best bound in Bourgain's slicing problem was  $Cn^{1/4} \log n$  (Bourgain '89) or  $n^{1/4}$  (K. '06). With Lee-Vempala '16, we reached three completely different proofs for  $1/4$  bound. Yet it was non-optimal.

An equivalent formulation of Bourgain's slicing problem

Suppose that  $K \subseteq \mathbb{R}^n$  with  $\text{Vol}_n(K) = 1$  is isotropic, so

$$\text{Cov}(K) = L_K^2 \cdot \text{Id}.$$

It is known that  $c < L_K$ . Is it true that  $c < L_K < C$ ?  
i.e., do the two normalizations essentially coincide?

# Using the Brunn-Minkowski inequality

This is related to Hensley's theorem:

**Theorem (Volumes of slices – Hensley '80, Fradelizi '99)**

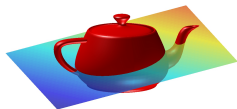
Let  $K \subseteq \mathbb{R}^n$  be an isotropic convex body. Then for any hyperplanes  $H_1, H_2 \subseteq \mathbb{R}^n$  through the origin,

$$\frac{\text{Vol}_{n-1}(K \cap H_2)}{\text{Vol}_{n-1}(K \cap H_1)} \leq \sqrt{6}.$$

- In fact,  $\text{Vol}_{n-1}(K \cap H) \sim 1/L_K$ .
- Proven using the **Brunn-Minkowski inequality** (1887):  
For any Borel sets  $A, B \subset \mathbb{R}^n$ ,

$$\text{Vol}\left(\frac{A+B}{2}\right) \geq \sqrt{\text{Vol}(A)\text{Vol}(B)}$$

Here  $(A+B)/2 = \{(a+b)/2; a \in A, b \in B\}$ .





# Corrected Busemann-Petty conjecture

In the 1950s, Busemann-Petty conjectured the following:

- Let  $K, T \subseteq \mathbb{R}^n$  be centrally-symmetric convex bodies such that

$$\text{Vol}_{n-1}(K \cap \theta^\perp) \leq \text{Vol}_{n-1}(T \cap \theta^\perp) \quad \text{for all } \theta \in S^{n-1}.$$

Does it follow that  $\text{Vol}_n(K) \leq \text{Vol}_n(T)$ ?

True if  $K$  is a Euclidean ball or a cross-polytope.

In general, true if  $n \leq 4$  and false if  $n \geq 5$ .

(Lutwak '88, Zhang, Gardner-Koldobsky-Schlumprecht '90s)

- Fails already for the cube and Euclidean ball in high dimensions! We lose a factor of  $\approx \sqrt{e/2}$  (Ball '86).

Another equivalent formulation of the slicing problem

Does it follow that  $\text{Vol}_n(K) \leq C \cdot \text{Vol}_n(T)$  for some universal constant  $C > 0$ ?

# Sharpened Milman ellipsoid?

## An equivalent formulation of the slicing problem

Let  $K \subset \mathbb{R}^n$  be a convex body. Does there exist an ellipsoid  $\mathcal{E} \subset \mathbb{R}^n$ , with  $\text{Vol}_n(\mathcal{E}) = \text{Vol}_n(K)$ , such that

$$\text{Vol}_n(K \cap C\mathcal{E}) / \text{Vol}_n(K) \geq 1/2.$$

## Theorem (V. Milman, '80s)

$$\text{Vol}_n(K \cap C\mathcal{E}) / \text{Vol}_n(K) \geq c^n,$$

for universal constants  $C, c > 0$ .

- This suffices for proving the reverse Brunn-Minkowski inequality as well as the Bourgain-Milman inequality

$$\text{Vol}_n(K) \text{Vol}_n(K^\circ) \geq c^n \text{Vol}_n(B^n)^2$$

where  $K^\circ$  is the polar body of the convex  $K \subseteq \mathbb{R}^n$ .

# Two more equivalent formulations of the slicing problem

- 1 **Sylvester problem.** Select  $n + 2$  independent, random points according to the uniform measure in a convex body  $K$ . Let  $p(K)$  be the probability that these  $n + 2$  points are in convex position. Is it true that

$$(1 - p(K))^{1/n} \simeq 1/\sqrt{n}?$$

- 2 **Entropy and covariance.** Is it true that for any log-concave random vector  $X$  in  $\mathbb{R}^n$ ,

$$\text{Ent}(X) = \frac{1}{2} \cdot \log \det \text{Cov}(X) + O(n)?$$

# Could the cube and simplex be the extreme cases?

- The isotropic constant of a convex body  $K \subseteq \mathbb{R}^n$  is an affine-invariant defined via

$$L_K^{2n} = \frac{\det \text{Cov}(K)}{\text{Vol}_n(K)^2}.$$

- We know that  $L_K > c$ , and is minimized for the Euclidean ball. The conjecture is that  $L_K < C$ . For example,

$$L_{[0,1]^n} = \frac{1}{\sqrt{12}}, \quad L_{\Delta^n} = \frac{(n!)^{1/n}}{(n+1)^{(n+1)/(2n)} \sqrt{n+2}} \approx \frac{1}{e}.$$

## Relations to classical conjectures

- 1 If  $L_K$  is maximized for the simplex  $\Delta^n$ , then the Mahler volume-product conjecture follows (the non-symmetric case, proven in 2D by Mahler, 1939). See K. '18.
- 2 If among centrally-symmetric bodies,  $L_K$  is maximized for the cube, then the Minkowski lattice conjecture follows (proven in 2D by Minkowski, 1901). See Magazinov '18.

Thank you!