

On KLS conjecture for certain classes of convex sets

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Poincaré inequality

We say that a probability measure μ satisfies the Poincaré inequality if

$$\mathrm{Var}_\mu f = \int f^2 d\mu - \left(\int f d\mu \right)^2 \leq C_\mu \int |\nabla f|^2 d\mu.$$

Poincaré (spectral gap) constant = best value of C_μ

Motivating problem: KLS conjecture

Conjecture

Kannan, Lovasz, Simonovits (1995). There exists a universal number c such that for every uniform distribution μ on a convex body K satisfying

$$\mathbb{E}_{\mu} x_i = 0, \quad \mathbb{E}_{\mu}(x_i x_j) = \delta_{ij}.$$

one has

$$C_K := C_{\mu} \leq c.$$

We call such bodies **isotropic**.

Equivalently:

there exists a universal number c such that for every uniform distribution μ on a convex body K

$$C_K \leq c \cdot C_K^{lin},$$

where

$$C_K^{lin} = \sup_{\text{linear } f} \frac{\text{Var}_\mu(f)}{\int_K |\nabla f|^2 d\mu}.$$

Other related conjectures and results

- 1) **Hyperplane conjecture.** (**Bourgain**). There exists universal $c > 0$ such that for any convex set $K \subset \mathbb{R}^d$ of $\text{Vol}(K) = 1$ there exists a hyperplane L such that $\text{Vol}(K \cap L) > c$.
- 2) **Thin-shell conjecture.** This is KLS conjecture for concrete function $f = |X|$. Can be put in the following way:

$$\mathbb{E}_\mu((|X| - \sqrt{d})^2) \leq c$$

for some universal constant c and every uniform distribution μ on an isotropic convex body $K \subset \mathbb{R}^d$.

- 3) **Central limit theorem for convex bodies.** Estimates of the thin-shell type \implies central limit theorem for convex bodies (initiated by **Sudakov** (70's), recent result: **Klartag, 2006**).

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What is known?

1. Special bodies

l_p -balls

$$B_p = \{x : |x_1|^p + |x_2|^p + \cdots + |x_d|^p \leq 1\}, \quad 1 \leq p \leq \infty.$$

It is known that

$$C_{B_p} \leq \frac{c(p)}{d^{\frac{2}{p}}}$$

and B_p satisfies the KLS conjecture.

Case $1 \leq p \leq 2$: S. Sodin [2008]

Case $2 \leq p \leq \infty$: R. Latała, J. O. Wojtaszczyk [2008]

Simplex: F. Barthe, P. Wolff [2009].

2. Unconditional convex bodies

Unconditional bodies are bodies which are invariant with respect to the mappings

$$x \rightarrow (\pm x_1, \dots, \pm x_d).$$

K. Klartag [2009]: KLS conjecture holds for unconditional convex bodies up to the logarithmic factor

$$C_K \leq C \log d.$$

3. General bodies

O. Guédon, E. Milman [2011] and **R. Eldan** [2013] :

$$C_K \leq Cd^{\frac{1}{3}} \log d.$$

Best known result:

Y.T. Lee, S. Vempala [2016]

$$C_K \leq Cd^{\frac{1}{4}}.$$

Main result

Level set

$$K_E = \{V \leq E\}.$$

Theorem

Let $\mu = e^{-V} dx$ be a log-concave probability measure with $\min(V) = 0$. There exists a universal constant $C > 1$ and $E \leq d$ such that

$$\frac{1}{C} \leq \text{Vol}^{\frac{1}{d}}(K_E) \leq C$$

and

$$C_{K_E} \leq C \cdot C_\mu \log(e + \sqrt{d}C_\mu).$$

Removing logarithmic factor

Theorem

Let $V_i: \mathbb{R} \mapsto \mathbb{R}$ be convex functions, $1 \leq i \leq d$. Assume that $\min(V_i) = 0$ and every $\mu_i = e^{-V_i} dx_i$ is a probability measure with barycenter at the origin. There exists a universal constant $C > 1$ and $E \leq d$ such that

$$\frac{1}{C} \leq \text{Vol}^{\frac{1}{d}}(K_E) \leq C$$

and

$$C_{K_E} \leq C \log(e + \min(A_2, d)) C_{K_E}^{\text{lin}},$$

where

$$A_2 = \frac{1}{\sqrt{d}} \|(\alpha_i)\|,$$
$$\alpha_i = \inf \left\{ t > 0: \int_{\mathbb{R}} e^{\frac{|V_i'(y)y-1|}{t}} dy \leq e \right\}.$$

Example

Let $p_i^\pm \in [1, P]$, $i = 1, \dots, d$, for a fixed arbitrary constant $P \geq 1$. Then there exists $E \leq d$ so that the generalized Orlicz ball:

$$K_E := \left\{ x \in \mathbb{R}^d ; \sum_{i=1}^d ((x_i)_+^{p_i^+} + (x_i)_-^{p_i^-}) \leq E \right\},$$

satisfies

$$\frac{1}{C} \leq \text{Vol}^{\frac{1}{d}}(K_E) \leq C$$

and

$$C_{K_E} \leq C \log(e + \min(P, d)) C_{K_E}^{\text{lin}}$$

for some universal $C > 0$.

Proof. Step 1. From μ to linearized measure on annulus

Probability measure

$$\mu = e^{-V} dx.$$

Generalized ball

$$K_E = \{V \leq E\}.$$

Linearized probability measure

$$\mu_{K_E} = \frac{1}{Z_E} e^{-(E+n(\|x\|_{K_E}-1))} dx$$

Linearized probability measure on annulus

$$\mu_{K_{E,\omega}} = \frac{1}{Z_{E,\omega}} e^{-(E+n(\|x\|_{K_E}-1))} I_{1-\omega \leq \|x\|_{K_E} \leq 1} dx.$$

Ratio of normalizing constants

Note that $\|X_{K_E}\|$ has Gamma distribution with respect to μ_{K_E} . In particular

$$Z_E = \frac{d! e^d}{d^d} e^{-E} \text{Vol}(K_E),$$

$$\frac{Z_{E,\omega}}{Z_E} = \frac{d^d}{(d-1)!} \int_{1-\omega}^1 e^{-dr} r^{d-1} dr \geq c\sqrt{d}\omega$$

provided $0 \leq \omega \leq \frac{1}{\sqrt{d}}$.

Uniform estimate

Corollary

$$\frac{d\mu_{K_{E,\omega}}}{d\mu} \leq \frac{e^{d\omega}}{c\sqrt{d}\omega Z_E}$$

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We apply estimate $\frac{d\mu_{K_{E,\omega}}}{d\mu} \leq \frac{e^{d\omega}}{c\sqrt{d\omega}Z_E}$ and the following result

Theorem

[Barthe–Milman] *Let ν_1, ν_2 be probability measures. Assume that $\|\frac{d\nu_2}{d\nu_1}\|_{L^p(\nu_1)} \leq L$ for some $p \in (1, \infty]$. Then setting $q = p^* = \frac{p}{p-1}$, we have:*

$$K_2(r) \leq 2LK_1^{1/q}(r/2) \quad \forall r > 0,$$

where K_1, K_2 are concentration functions of ν_1, ν_2 .

Thus we transfer concentration estimates from μ to $\mu_{E,K}$.

Step 2. From annulus to cone measure

Note that the mapping

$$T(x) = \frac{x}{\|x\|_{K_E}}$$

pushes forward the annulus measure $\mu_{K_E, \omega}$ onto the cone measure

$$\sigma = \frac{1}{\text{Vol}(K_E)} \frac{\langle x, \nu_{\partial K_E} \rangle}{d} \cdot \mathcal{H}^{d-1}|_{\partial K_E}.$$

To transfer concentration from the annulus measure to the cone measure we apply

Lemma

[E. Milman] For every 1-Lipshitz function f

$$\int |f - \text{med}_{\mu_2} f| d\mu_2 \leq \int |f - \text{med}_{\mu_1} f| d\mu_1 + W_1(\mu_1, \mu_2),$$

where W_1 is the Kantorovich (Wasserstein) distance.

and

Lemma

$$W_1(\mu_{K_E, \omega}, \sigma) \leq \frac{d+1}{d} \omega \int \|x\|_{K_E} d\lambda_{K_E},$$

where λ_{K_E} is the Lebesgue probability measure on K_E .

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Step 3. From cone measure to uniform measure

Hardy-type inequality. For any convex Ω

$$\int_{\Omega} |f - med_{\lambda_{\Omega}}| d\lambda_{\Omega} \leq \int_{\partial\Omega} |f - med_{\sigma_{\Omega}}| d\sigma_{\Omega} + \frac{1}{d} \int |\nabla f| |x| d\lambda_{\Omega},$$

where σ_{Ω} is the corresponding cone measure.

Proof:

$$\begin{aligned} d \int_{\Omega} g dx &= \int_{\Omega} \operatorname{div}(x) g dx = - \int_{\Omega} \langle x, \nabla g \rangle dx + \int_{\partial\Omega} \langle x, \nu_{\Omega} \rangle g d\mathcal{H}^{d-1} \\ &\leq \int_{\Omega} |x| |\nabla g| dx + \int_{\partial\Omega} \langle x, \nu_{\Omega} \rangle g d\mathcal{H}^{d-1}. \end{aligned}$$

Setting $g = |f - med_{\sigma_{\Omega}}|$ we complete the proof.

Last step

Collecting everything together we get that for every 1-Lipschitz function f

$$\int_{\Omega} |f - \text{med}_{\lambda_{\Omega}}| d\lambda_{\Omega} \leq C(\omega, d, Z_E, C_{\mu}).$$

Optimize parameters: choosing $\omega = \frac{1}{d}$ and applying lemma below we get the result.

Lemma

Let $\mu = e^{-V} dx$ be a log-concave probability measure with $\min(V) = 0$. Set

$$\text{Level}(V) = \left\{ E \geq 0 : e^{-E} \text{Vol}(K_E) \geq \frac{1}{e} \frac{d^d e^{-d}}{d!} \right\}.$$

Then $\text{Level}(V) = [E_{\min}, E_{\max}]$ is a non-empty interval such that

1.

$$E_{\min} \leq d$$

2.

$$1 \leq E_{\min} - E_{\max} \leq e\sqrt{2\pi d}(1 + o(1))$$

3.

$$c(d) \leq \text{Vol}^{\frac{1}{d}}(E_{\min}) \leq \text{Vol}^{\frac{1}{d}}(E_{\max}) \leq ec(d)(1 + o(1)),$$

where $c(d) \rightarrow 1$ as $d \rightarrow \infty$.

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