

# Poincaré Inequalities and Moment Maps

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# Poincaré-type inequalities

- Poincaré-type inequalities (in this lecture): Bounds for the **variance** of a function in terms of the **gradient**.

## Wirtinger inequality (1904)

$$\int_0^\pi f = 0 \quad \Rightarrow \quad \int_0^\pi f^2 \leq \int_0^\pi (f')^2.$$

Here it is easy to describe the eigenfunctions. More generally,

## Poincaré inequality for Convex Domains (Payne & Weinberger, 1960)

For any convex body  $K \subset \mathbb{R}^n$  and  $f : K \rightarrow \mathbb{R}$ ,

$$\int_K f = 0 \quad \Rightarrow \quad \int_K f^2 \leq \frac{D^2}{\pi^2} \int_K |\nabla f|^2,$$

where  $D = \text{diameter}(K) = \sup_{x,y \in K} |x - y|$ .

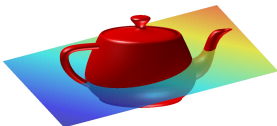
- Consider the **Laplacian** (with Neumann boundary conditions) on a convex body  $K \subset \mathbb{R}^n$ . Then  $\lambda_0 = 0$ , and according to Payne-Weinberger,

$$\lambda_1 = \lambda_1(K) \geq \pi^2/D^2$$

for  $D = \text{diameter}(K)$ .

- The Payne-Weinberger bound is optimal for a thin cylinder, but not for a Euclidean ball or a cube (off by a factor of  $n$ ).

For an arbitrary convex body, it is difficult to describe the eigenfunction(s). Original proof uses **bisections** of convex bodies. The idea was further developed by Gromov and Milman, and by Kannan, Lovász and Simonovits.



# Isoperimetry and spectral gap

- Intimately related to geometry, and to the **isoperimetric coefficient**: For an open set  $K \subset \mathbb{R}^n$  define

$$h(K) = \inf \{ \text{Vol}_{n-1}(\partial A \cap K); A \subseteq K, \text{Vol}_n(A) = \text{Vol}_n(K)/2 \}.$$

Theorem (Maz'ya/Cheeger '70, Buser/Ledoux '82/'04)

For any open, convex body  $K \subset \mathbb{R}^n$ ,

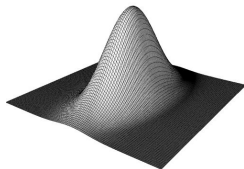
$$\frac{h^2(K)}{4} \leq \lambda_1(K) \leq 10h^2(K).$$

- Lower bound is from Maz'ya-Cheeger inequality (no convexity needed), upper bound is due to Buser and to Ledoux.
- May be used to estimate the **rate of mixing** of Brownian motion and various Markov chains on a convex domain.

# Our motivation

- A probability measure  $\mu$  on  $\mathbb{R}^n$  is **log-concave** if its density is  $\exp(-\Psi)$  for a convex function  $\Psi : \mathbb{R}^n \rightarrow \mathbb{R} \cup \{+\infty\}$ .
- The uniform measure on a convex body is log-concave, as is the Gaussian density.
- We say that  $\mu$  is **isotropic** when

$$\int_{\mathbb{R}^n} x d\mu(x) = 0 \quad \text{and} \quad \text{Cov}(\mu) = Id.$$



Thin shell problem (Antilla, Ball & Perissinaki '03, Bobkov & Koldobsky '03)

Suppose  $\mu$  is log-concave and isotropic. Is it true that

$$\int_{\mathbb{R}^n} (|x| - \sqrt{n})^2 d\mu(x) \leq C$$

for a universal constant  $C > 0$ ?

# Why care about Thin Shell?

A positive answer would imply...

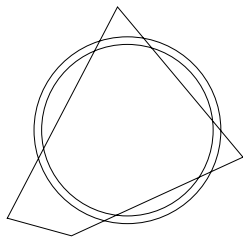
- 1 An optimal rate of convergence in the convex CLT.
- 2 K.-Eldan '11: The hyperplane conjecture.
- 3 Eldan '12: The Kannan-Lovasz-Simonovits conjecture, up to logarithmic factors! (work in progress)

What is the width of the **thin spherical shell** that contains most of the volume of an isotropic convex body?

- The best available thin shell bound is

$$\mathbb{E} \left( \frac{|X|}{\sqrt{n}} - 1 \right)^2 \leq \frac{C}{n^\alpha}$$

with  $\alpha = 1/3$  (Guédon-Milman '11),  $\alpha = 1/4$  (Fleury '10) and  $\alpha = 1/6$  (K., '07). Ideas of Paouris are relevant.



# How would you prove thin shell bound?

Suppose  $\mu$  is a log-concave measure on  $\mathbb{R}^n$ , isotropic. Need:

$$\int_{\mathbb{R}^n} \left( \frac{|x|^2}{n} - 1 \right)^2 d\mu(x) \ll 1.$$

All current proofs rely on the following idea: Projecting the measure to a random subspace, it becomes approximately the uniform measure on a sphere (à la Dvoretzky's theorem).

- Another possible approach:  
Try to prove a Poincaré-type inequality

$$\int_K \varphi^2 \leq \int_K Q(\nabla \varphi) d\mu$$

for all functions  $\varphi$  with  $\int \varphi d\mu = 0$ , for some quadratic expression  $Q$ . We only need  $\varphi(x) = |x|^2/n - 1$ .

- The Poincaré inequalities mentioned above are not sufficient for obtaining non-trivial bounds.

# Unconditional convex bodies

A convex body  $K \subset \mathbb{R}^n$  is called **unconditional** when

$$(x_1, \dots, x_n) \in K \iff (|x_1|, \dots, |x_n|) \in K.$$

(i.e., invariant under coordinate reflections)

**Theorem (Poincaré Inequality for unconditional convex bodies, K. '09+'11)**

Suppose  $K \subset \mathbb{R}^n$  is convex and unconditional. Then for any  $f$ ,

$$\int_K f = 0 \implies \int_K f^2 \leq \int_K \sum_{i=1}^n (4x_i^2 + V_i) |\partial^i f|^2 dx$$

where  $V_i = \int_K x_i^2 / \text{Vol}_n(K)$ .

- Generalizes directly to log-concave measures that are **unconditional** (i.e., invariant under coordinate reflections).
- Convexity assumptions are essential (coordinate-wise monotonicity does not suffice).



# Unconditional convex bodies

Suppose  $K$  is either the **Euclidean ball** of volume one (and radius  $c\sqrt{n}$ ), or else **the cube**  $[-1/2, 1/2]^n$ . The bound

$$\int_K f^2 \leq \int_K \sum_{i=1}^n (4x_i^2 + 1) |\partial^i f|^2 dx, \quad (1)$$

for  $f$  with  $\int_K f = 0$  is typically better than Payne-Weinberger's bound

$$\int_K f^2 \leq \frac{D^2(K)}{\pi^2} \int_K \sum_{i=1}^n |\partial^i f|^2 dx, \quad (2)$$

because  $D(K) = \text{diameter}(K) \sim \sqrt{n}$ , while typically  $x_i^2 \approx 1$ .

- Substituting  $f(x) = |x|^2/n - 1$  into (1), we completely solve the “thin shell” problem for unconditional convex bodies.

# Convexity and symmetry

- Other types of symmetries work as well, including those of the simplex (Barthe and Cordero-Erausquin '11).

## Theorem (unconditional log-concave measures)

Suppose  $\mu$  is a log-concave, unconditional probability measure on  $\mathbb{R}^n$ . Then for any  $f$ , denoting  $\text{Var}_\mu(f) = \int f^2 d\mu - (\int f d\mu)^2$ ,

$$\text{Var}_\mu(f) \leq \int_{\mathbb{R}^n} \sum_{i=1}^n (4x_i^2 + V_i) |\partial^i f(x)|^2 d\mu(x)$$

where  $V_i = \int x_i^2 d\mu$ .

Let  $\exp(-\Psi)$  be the density of  $\mu$ . Then  $\Delta_\mu u := \Delta u - \nabla u \cdot \nabla \Psi$  is the “ $\mu$ -Laplacian”, as it satisfies

$$\int_{\mathbb{R}^n} (u \Delta_\mu v) d\mu = - \int_{\mathbb{R}^n} (\nabla u \cdot \nabla v) d\mu.$$

# Dualizing Bochner's formula

- The proof uses the **Bochner-Lichnerowicz-Weitzenböck identity**: For any  $\varphi$ ,

$$\int_{\mathbb{R}^n} (\Delta_{\mu}\varphi)^2 d\mu = \int_{\mathbb{R}^n} |\nabla^2\varphi|^2 d\mu + \int_{\mathbb{R}^n} (\nabla^2\Psi)(\nabla\varphi) \cdot \nabla\varphi d\mu$$

where  $|\nabla^2\varphi|$  is the Hilbert-Schmidt norm of the Hessian.

- Our only use of convexity: The second term is non-negative as  $\Psi$  is **convex**. Hence,

$$\int_{\mathbb{R}^n} (\Delta_{\mu}\varphi)^2 d\mu \geq \int_{\mathbb{R}^n} |\nabla^2\varphi|^2 d\mu.$$

## Proposition (“dual Bochner”)

For any log-concave probability measure  $\mu$  on  $\mathbb{R}^n$  and a function  $f$ ,

$$\text{Var}_{\mu}(f) \leq \sum_{i=1}^n \|\partial^i f\|_{H^{-1}(\mu)}^2.$$

# What is the $H^{-1}(\mu)$ norm?

- We define, for a function  $g$ ,

$$\|g\|_{H^{-1}(\mu)} = \sup \left\{ \int_{\mathbb{R}^n} g\varphi d\mu ; \int_{\mathbb{R}^n} |\nabla\varphi|^2 d\mu \leq 1 \right\}.$$

- This is the dual of the  $H^1(\mu)$ -norm.

Observe that  $\|g\|_{H^{-1}(\mu)} = +\infty$  when  $\int g d\mu \neq 0$ .

- Hence, the bound

$$\text{Var}_\mu(f) \leq \int_{\mathbb{R}^n} \sum_{i=1}^n \|\partial^i f\|_{H^{-1}(\mu)}^2$$

is useless when  $\int \nabla f d\mu \neq 0$ .

# Transportation of measure

- The  $H^{-1}(\mu)$ -norm has a **geometric interpretation**.  
Suppose  $\mu, \nu$  are prob. measures on  $\mathbb{R}^n$ . Define

$$W_2^2(\mu, \nu) = \inf_T \int_{\mathbb{R}^n} |x - T(x)|^2 d\mu(x)$$

where the infimum runs over  $T$  that pushes forward  $\mu$  to  $\nu$ .  
This is the  $L^2$ -Monge-Kantorovich-Wasserstein distance.

From Brenier, Otto and Villani we learn:

**Proposition (“ $H^{-1}$ -norm is the infinitesimal transportation cost”)**

If  $\int g d\mu = 0$ , then (under mild regularity assumptions)

$$\|g\|_{H^{-1}(\mu)} = \lim_{\varepsilon \rightarrow 0} \frac{W_2(\mu_\varepsilon, \mu)}{\varepsilon}$$

where  $d\mu_\varepsilon/d\mu = 1 + \varepsilon g$ .

**Decompositions:** If  $\mu = \int \mu_\alpha d\lambda(\alpha)$ , then,

$$\|g\|_{H^{-1}(\mu)}^2 \leq \int \|g\|_{H^{-1}(\mu_\alpha)}^2 d\lambda(\alpha).$$

- Recall “dual Bochner”: When  $\mu$  is log-concave, for any  $f$ ,

$$\text{Var}_\mu(f) \leq \int_{\mathbb{R}^n} \sum_{i=1}^n \|\partial^i f\|_{H^{-1}(\mu)}^2 = \lim_{\varepsilon \rightarrow 0} \sum_{i=1}^n \frac{W_2^2((1 + \varepsilon \partial^i f)\mu, \mu)}{\varepsilon^2}.$$

Suppose  $\mu$  is uniform on  $K$ . Then we need to efficiently transport the uniform measure on  $K$  to the measure, supported on  $K$ , whose density on  $K$  is  $1 + \varepsilon \partial^i f$ .

- Assume that both  $f$  and  $K$  are **unconditional**. Then  $\partial^i f$  is **odd** with respect to the  $i^{\text{th}}$  coordinate. A naïve transportation that comes to one’s mind is...



# Transportation along parallel segments

Note that for any line  $\ell$  parallel to the  $i^{\text{th}}$  unit vector,

$$\int_{K \cap \ell} \partial^i f = 0.$$

We may thus apply the (non-optimal) transportation of  $1_K dx$  to  $(1 + \varepsilon \partial^i f) 1_K dx$ , separately on each line segment!

- Simple computation: For  $d\mu_R = 1_{[-R,R]} dx$  on the line,

$$\|g\|_{H^{-1}(\mu_R)}^2 \leq 4 \int_{-R}^R x^2 g^2(x) dx$$

for any odd function  $g$ .

- Hence, when  $K$  and  $f$  are unconditional (e.g.,  $f(x) = |x|^2$ ),

$$\|\partial^i f\|_{H^{-1}(K)}^2 \leq 4 \int_K x_i^2 |\partial^i f(x)|^2 dx.$$

# Symmetry seems crucial

- To conclude, we obtained the following Poincaré inequality for unconditional  $K$  and  $f$ :

$$\text{Var}_K(f) \leq \sum_{i=1}^n \|\partial^i f\|_{H^{-1}(K)}^2 \leq 4 \int_K \sum_{i=1}^n x_i^2 |\partial^i f(x)|^2 dx.$$

- Standard tricks reduce the case of any function  $f$  to an unconditional function  $f$ .
- The most important case is  $f(x) = |x|^2$ . Even here, we must utilize the symmetries of  $K$ . We obtain

$$\int_K \left( \frac{|x|^2}{n} - 1 \right)^2 \frac{dx}{\text{Vol}_n(K)} \leq \frac{C}{n^2} \sum_{i=1}^n \int_K x_i^4 \frac{dx}{\text{Vol}_n(K)}$$

when  $K$  is convex, unconditional and isotropic (e.g., the cube  $[-\sqrt{3}, \sqrt{3}]^n$ ).



# A digression: multiplicity of the Neumann eigenvalue

Proposition (K. '09 – was it known before?)

Let  $K \subset \mathbb{R}^n$  be a convex body. Let  $\varphi \neq 0$  be an eigenfunction of the Neumann Laplacian, corresponding to  $\lambda_1$ . Then,

$$\int_K \nabla \varphi \neq 0.$$

Consequently, the multiplicity of  $\lambda_1$  is at most  $n$ .

**Proof:** If  $\int_K \nabla \varphi = 0$ , then by Bochner,

$$\lambda_1^2 \int_K \varphi^2 = \int_K |\Delta \varphi|^2 > \int_K \sum_{i=1}^n |\nabla \partial^i \varphi|^2 \geq \lambda_1 \int_K \sum_{i=1}^n |\partial^i \varphi|^2.$$

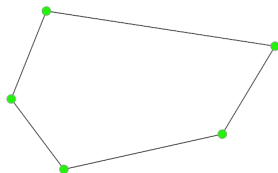
- A deeper fact (Nadirashvili '88): For any simply-connected planar domain, the multiplicity of  $\lambda_1$  is at most 2.

# Convexity without symmetry

- In the central case where  $f(x) = |x|^2$ , our method boils down to transporting

$$(1 + \varepsilon x_i) 1_K(x) dx \quad \text{to} \quad 1_K(x) dx.$$

How can we bound the transportation cost without symmetries?



While trying to find a solution, I came across unfamiliar territory: Moment maps and Kähler geometry.

- The basic idea: Introduce additional symmetries by considering a certain transportation of measure from a space of twice or thrice the dimension.

# The simplest example

- Suppose  $K = [0, 1] \subset \mathbb{R}$ . Denote  $D = \{z \in \mathbb{R}^2; |z|^2 \leq 1\}$ .  
The map

$$\pi(z) = |z|^2$$

pushes forward the uniform probability measure on  $D$  to the uniform probability measure on  $K$ .

Now, for  $f : K \rightarrow \mathbb{R}$  with  $\int_K f = 0$ , set

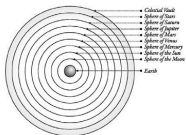
$$g(z) = f(\pi(z)) \quad (z \in \mathbb{R}^2).$$

- Since  $D$  is convex, we can use “dual Bochner”:

$$\text{Var}_K(f) = \text{Var}_D(g) \leq \left\| \frac{\partial g}{\partial x} \right\|_{H^{-1}(D)}^2 + \left\| \frac{\partial g}{\partial y} \right\|_{H^{-1}(D)}^2$$

# The example of the interval

- The integral of  $\nabla g$  is zero on each **circle** of the form  $\pi^{-1}(x)$  (because  $g$  is even).
- Transport



$$\left(1 + \varepsilon \frac{\partial g}{\partial x}\right) 1_D(x) dx \quad \text{to} \quad 1_D(x) dx$$

separately along each circle. Observe that

$$\nabla g(re^{i\theta}) = 2f'(r^2)re^{i\theta},$$

which is (up to a constant) one specific function on the circle.

This leads to the inequality: For any  $f : [0, 1] \rightarrow \mathbb{R}$ ,

$$\int_0^1 f = 0 \quad \implies \quad \int_0^1 f^2 \leq 4 \int_0^1 x^2 (f'(x))^2 dx$$

# A straightforward generalization

Fix integers  $\ell, n \geq 1$ . For  $z = (z_1, \dots, z_n) \in (\mathbb{R}^\ell)^n$  set

$$\pi(z) = (|z_1|^\ell, \dots, |z_n|^\ell).$$

Then  $\pi : \mathbb{R}^{\ell n} \rightarrow \mathbb{R}^n$  is a proper map, that pushes forward the Lebesgue measure on  $\mathbb{R}^{\ell n}$  to a (multiple of) the Lebesgue measure on  $\mathbb{R}_+^n$ .

- The case where  $\ell = 2$  is the **moment map** from  $\mathbb{C}^n$  to  $\mathbb{R}_+^n$ .

An example: Suppose  $\ell = 2$  and  $\Omega \subset \mathbb{C}^n$  is the unit ball

Then  $\pi$  pushes forward the uniform measure on  $\Omega$  to the uniform measure on the simplex, with  $\pi^{-1}(x)$  being a torus.

- The actually argument works for prob. measures in  $\mathbb{R}_+^n$  with certain **convexity properties**. No symmetries required!

# The convexity assumptions

Therefore, if we have enough **convexity** for the Bochner argument, then we obtain an interesting Poincaré inequality.

## Theorem (K. '11)

Let  $\ell, n > 1$  be integers and let  $\nu$  be a probability measure on  $\mathbb{R}_+^n$  whose density is  $(1/\ell)$ -**log-concave**. Then, for any  $f$ ,

$$\text{Var}_\nu(f) \leq \frac{\ell^2}{\ell - 1} \sum_{i=1}^n \int_{\mathbb{R}_+^n} x_i^2 |\partial^i f|^2 d\nu(x).$$

- The probability density  $\rho : \mathbb{R}_+^n \rightarrow \mathbb{R}$  is  $q$ -log-concave, for  $0 < q \leq 1$ , if the function

$$(x_1, \dots, x_n) \mapsto \rho(x_1^q, \dots, x_n^q)$$

is log-concave on  $\mathbb{R}_+^n$ .

- $(1/\ell)$ -log-concavity seems **weaker** than log-concavity (it is certainly so in the unconditional case).

# CLT for non-convex bodies

- The theorem holds also for non-integer  $\ell > 1$ .
- The uniform measure on

$$B_p^n = \left\{ x \in \mathbb{R}_+^n; \sum_{i=1}^n |x_i|^p \leq 1 \right\}$$

for  $0 < p \leq 1$  is  $p$ -log-concave.

## Corollary

Let  $n \geq 1$  be an integer, let  $\ell > 1$  and let  $K \subset \mathbb{R}_+^n$  be such that

$$\left\{ (x_1^{1/\ell}, \dots, x_n^{1/\ell}); (x_1, \dots, x_n) \in K \right\}$$

is a convex set. Then, for any function  $f$

$$\int_K f = 0 \quad \Rightarrow \quad \int_K f^2 \leq \frac{\ell^2}{\ell - 1} \sum_{i=1}^n \int_K x_i^2 |\partial^i f(x)|^2 dx.$$

# Remarks about our “weak convexity”

- We obtain a good “thin shell estimate” for convex bodies such as  $B_p^n$  for  $p = 1/2$ , or  $p > 0$  in general.
- The uniform measure on a convex body in  $\mathbb{R}_+^n$  is  $p$ -log-concave for  $p = 1$  (the useless case in our corollary).
- If  $K \subset \mathbb{R}^n$  is **unconditional** and convex, then the uniform measure on  $K \cap \mathbb{R}_+^n$  is  $1/2$ -log-concave (and also  $p$ -log-concave for any  $0 < p < 1$ ).
- We thus recover the result discussed earlier for **unconditional, convex sets**.

Unconditionality is used only to deduce  $p$ -log-concavity from  $q$ -log-concavity when  $p < q$ .

The symmetry assumptions return through the back door.



# Toric Kähler manifolds

- The most important case where  $\ell = 2$  corresponds to the moment map

$$\mathbb{C}^n \ni (z_1, \dots, z_n) \mapsto (|z_1|^2, \dots, |z_n|^2) \in \mathbb{R}_+^n.$$

- Be wise, generalise?!

Let  $\psi : \mathbb{R}^n \rightarrow \mathbb{R}$  be a smooth, convex function. Then,

$$K = \nabla\psi(\mathbb{R}^n)$$

is a convex set. Consider the complex torus

$\mathbb{T}_{\mathbb{C}}^n = \mathbb{C}^n / (\sqrt{-1}\mathbb{Z}^n)$ , and the Kähler form on  $\mathbb{T}_{\mathbb{C}}^n$

$$\omega_{\psi} = 2\sqrt{-1}\partial\bar{\partial}\psi = \frac{\sqrt{-1}}{2} \sum_{i,j=1}^n \psi_{ij} dz_i \wedge d\bar{z}_j.$$

# Kähler manifolds for alphabets

- Hmm... a Kähler form? This means that on  $\mathbb{T}_{\mathbb{C}}^n = \mathbb{C}^n / (\sqrt{-1}\mathbb{Z}^n)$  we have a Riemannian metric  $g_{\psi}$  with

$$g_{\psi} \left( \frac{\partial}{\partial x_i}, \frac{\partial}{\partial x_j} \right) = g_{\psi} \left( \frac{\partial}{\partial y_i}, \frac{\partial}{\partial y_j} \right) = \psi_{ij} \quad (i, j = 1, \dots, n)$$

while  $g_{\psi} \left( \frac{\partial}{\partial x_i}, \frac{\partial}{\partial y_j} \right) = 0$  for any  $i, j$ .

- The associated moment map

$$\mathbb{T}_{\mathbb{C}}^n \ni (x + \sqrt{-1}y) \mapsto \nabla\psi(x) \in \mathbb{R}^n$$

pushes forward the Riemannian volume to the uniform measure on the convex body  $K \subset \mathbb{R}^n$ .

- For  $x \in \mathbb{R}^n$ , the real torus

$$\left\{ x + \sqrt{-1}y; y \in \mathbb{R}^n / \mathbb{Z}^n \right\}$$

is the pre-image of a point in  $K$ .

# Simple formula for Ricci curvature

We would like to apply the technique demonstrated earlier. Two ingredients:

- 1 We need “enough convexity/curvature” to apply Bochner.
- 2 We need to estimate the  $H^{-1}$ -norm on each torus.

Regarding convexity: Bochner's identity has a **Ricci** term. Our Riemannian manifold has a non-negative Ricci form when

$$x \mapsto \det \nabla^2 \psi(x)$$

is log-concave on  $\mathbb{R}^n$ . This is the condition we will assume.

- In other words, we need to transport an arbitrary log-concave measure on  $\mathbb{R}^n$ , to the uniform measure on  $K$ .

For each such  $\mu$ , we get a Poincaré-type inequality on  $K$ .

# Messy expressions

The infinitesimal **transportation cost** along each torus is easily computable, yet the formula is messy.

- 1 For a smooth, convex function  $\psi : \mathbb{R}^n \rightarrow \mathbb{R}$  and  $x \in \mathbb{R}^n$  set

$$Q_{\psi,x}^*(V) = \sum_{i,j,k,\ell,m,p=1}^n V^i V^j \psi^{\ell m} \psi_{jkm} \psi^{kp} \psi_{i\ell p} \quad (V \in \mathbb{R}^n).$$

- 2 The dual quadratic form is

$$Q_{\psi,x}(U) = \sup \left\{ 4 \left( \sum_{i,j=1}^n \psi_{ij} U^i V^j \right)^2 ; V \in \mathbb{R}^n, Q_{\psi,x}^*(V) \leq 1 \right\},$$

for  $x \in K$  and  $U \in \mathbb{R}^n$ , where  $\psi_{ij}$  is evaluated at  $\nabla \psi^*(x)$ .

# An almost-inapplicable, too general theorem

## Theorem

Let  $K \subset \mathbb{R}^n$  be a convex set. Suppose  $\psi : \mathbb{R}^n \rightarrow \mathbb{R}$  is a smooth, convex function, with

- 1  $\nabla\psi(\mathbb{R}^n) = K$ .
- 2  $x \mapsto \det \nabla^2\psi(x)$  is log-concave on  $\mathbb{R}^n$ .
- 3 The function  $\psi$  should be “regular at infinity”, a mild technical assumption.

Then, for any function  $f : K \rightarrow \mathbb{R}$ ,

$$\int_K f = 0 \quad \Rightarrow \quad \int_K f^2 \leq \int_K Q_{\psi,x}(\nabla f) dx.$$

- We may thus take the infimum over all such functions  $\psi$ .
- We may replace the convex body  $K \subset \mathbb{R}^n$  by a log-concave probability measure  $\mu$ . Condition 2 is replaced by “ $\nabla\psi$  transports a log-concave measure to  $\mu$ .” (Kolesnikov, '11)

## Well, it's not entirely inapplicable...

Say, for the case where  $K$  is a Euclidean ball, we may construct reasonable functions  $\psi$  with  $\det \nabla^2 \psi$  log-concave...

- We may test our general theorem in the case where  $K \subset \mathbb{R}^n$  is a simplex, and

$$\psi(x) = \log(1 + e^{x_1} + \dots + e^{x_n}).$$

- The Kähler manifold obtained is  $\mathbb{C}\mathbb{P}^n$  (more precisely, it is the open, dense subset of a full measure in  $\mathbb{C}\mathbb{P}^n$ , where the toric action is free).
- In this case, we get wonderful Poincaré type inequalities.

We were not able to write down reasonable, explicit formulae in any case, other than  $\mathbb{C}^n$  and  $\mathbb{C}\mathbb{P}^n$ .

## Another approach, without complexification

Let us inspect again the “dual Bochner” inequality, in the context of a Riemannian manifold with a measure  $(M, g, \mu)$ .

- The  $L^2(\mu)$ -norm is defined also for vector fields:

$$\langle V, W \rangle_{L^2(\mu)} = \int_M \langle V, W \rangle_g d\mu.$$

- The operator  $\Delta_\mu$  satisfies  $\langle f, \Delta_\mu g \rangle_{L^2(\mu)} = -\langle \nabla f, \nabla g \rangle_{L^2(\mu)}$ .

For any Banach (semi-)norm  $\| \cdot \|$  on vector fields:

$$\forall u, \int (\Delta_\mu u)^2 d\mu \geq \|\nabla u\|^2 \iff \forall f, \text{Var}_\mu(f) \leq \|\nabla f\|_*^2.$$

Recall that Bochner's formula tells us:

$$\int_M (\Delta_\mu u)^2 d\mu = \int_M \|\nabla_M^2 u\|_{L^2(\mu)}^2 d\mu + \int_M \text{Ric}_{M,\mu}(\nabla u, \nabla u) d\mu$$

# Anti-climax?

- The toric Kähler manifold  $(\mathbb{T}_{\mathbb{C}}^n, \omega_{\psi})$  is rather similar to the measure-metric space

$$X = (\mathbb{R}^n, \nabla^2\psi, \mu)$$

where  $d\mu = \det \nabla^2\psi(x) dx$ .

(the Riemannian volume is  $d\mu = \sqrt{\det \nabla^2\psi(x)} dx$ ).

Simply apply the Bochner formula for  $X$ , play with the various terms, and deduce the dual version of our “too general theorem”.

Our results are thus trivially “realized” (i.e., the inverse operation to “complexify”).

- In particular, the amusing theorem for measures on  $\mathbb{R}_+^n$  follows directly by Brascamp-Lieb and change of variables (this proof works for non-integer  $\ell$  as well...)



# Many questions remain

- Is this point of view of “symmetries” and “transportation cost” helpful? Does it reduce to trivial manipulations of Bochner’s formula?

Another idea whose origin lies in complex geometry:

## **Kähler-Einstein manifolds.**

Theorem (Wang-Zhu '04, Donaldson '08, E. Legendre '11, Berman-Berndtsson '12)

*Suppose  $K \subset \mathbb{R}^n$  is a convex body, barycenter at the origin.*

*Then there exists convex function  $\psi : \mathbb{R}^n \rightarrow \mathbb{R}$  such that*

- (a)  $\det \nabla^2 \psi(x) = C_K e^{-\psi(x)}.$
- (b)  $\nabla \psi(\mathbb{R}^n) = K.$

- The Kähler manifold whose potential is  $\psi$ , has Ricci tensor which equals half of the metric. (Hence we have log Sobolev inequality, Gaussian concentration etc.)
- Related to the Mongé-Ampère eigenfunction (Lions '86).

# A bit more on Kähler-Einstein transportation

In the language of measure-probability spaces: Suppose  $K \subset \mathbb{R}^n$  is convex, barycenter at the origin.

- Then there is a Riemannian metric-measure space

$$X_K = (K, \nabla^2 \varphi, \mu_K)$$

where  $\mu_K$  is the uniform measure on  $K$ , and the Riemannian metric is induced by  $\varphi$  (the Legendre transform of  $\psi$  from the previous theorem).

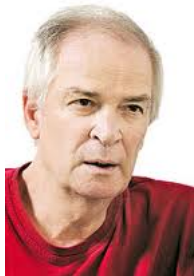
- The Laplacian's spectral gap is one, with multiplicity  $n$ . The eigenspace consists of all linear functionals!
- The Ricci tensor equals half the metric tensor. This space “resembles the Gaussian” in so many respects.

This space  $X_K$  must be good for something, isn't it?

Thank you!



Kähler



Einstein

Thank you!



Kähler



Einstein