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Towards multi-dimensional localisation

KRZYSZTOF CIOSMAK Fields Institute University of Toronto

11th-16th of June, 2023, High Dimensional Probability, Banach Centre

Motivating example from 1960

Suppose K ⊂ ℝⁿ is a closed convex set and let f ∈ C¹(K) be such that ∫_K f dµ = 0 for a log-concave measure µ on K.

Motivating example from 1960

Suppose $K \subset \mathbb{R}^n$ is a closed convex set and let $f \in C^1(K)$ be such that $\int_K f d\mu = 0$ for a log-concave measure μ on K.

Then

$$\lambda_{\mathcal{K}} \int_{\mathcal{K}} f^2 \, d\mu \leq \int_{\mathcal{K}} \|\nabla f\|^2 \, d\mu$$

for some $\lambda_{\mathcal{K}} \geq \frac{\pi^2}{(\operatorname{diam} \mathcal{K})^2}$.

$\label{eq:payne} Payne \mbox{ and Weinberger approach to Poincaré inequality} - hyperplane \mbox{ bisections}$

Motivating example from 1960

Suppose K ⊂ ℝⁿ is a closed convex set and let f ∈ C¹(K) be such that ∫_K f dµ = 0 for a log-concave measure µ on K. H = v[⊥]/
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for some $\lambda_K \geq \frac{\pi^2}{(\operatorname{diam} K)^2}$.

▶ By the Borsuk–Ulam theorem, if $\int_{K} f d\mu = 0$, then there exists a hyperplane H such that $\int_{K \cap H_{+}} f d\mu = \int_{K \cap H_{-}} f d\mu = 0$.

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Suppose $K \subset \mathbb{R}^n$ is a closed convex set and let $f \in \mathcal{C}^1(K)$ be such that $\int_{K} f d\mu = 0$ for a log-concave measure μ on K. $H = v^{\perp}$ Then v $\lambda_{\mathcal{K}} \int_{\mathcal{K}} f^2 d\mu \leq \int_{\mathcal{K}} \|\nabla f\|^2 d\mu$

for some $\lambda_K \geq \frac{\pi^2}{(\operatorname{diam} K)^2}$.

 $F(v) = \int_{H_{\perp} \cap K} f d\mu$

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- By the Borsuk–Ulam theorem, if ∫_K f dµ = 0, then there exists a hyperplane H such that ∫_{K∩H₊} f dµ = ∫_{K∩H_−} f dµ = 0.
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- Reduction can be done as long as $\dim K \ge 2$.
- This procedure produces a measurable partition and conditional measures of µ with respect to the partition are log-concave.
- Therefore, it suffices to prove the inequality in the one-dimensional case.



The localisation technique in convex geometry reduces an *n*-dimensional problem to a collection of one-dimensional problems.

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- It has been conjectured by Klartag that it can be generalised to multiple constraints setting.
- Ohta generalised the technique to Finsler manifolds and Cavalletti and Mondino generalised it to metric measure spaces.



One-dimensional localisation

▶ Let (M, d, μ) be an *n*-dimensional weighted Riemannian manifold. Assume that $f: M \to \mathbb{R}$ has null-integral and that $d(\cdot, x_0)f(\cdot)$ is μ -integrable.

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- Consider the problem of optimal transport of dη₁ = f₊ dμ to dη₂ = f_− dμ with metric cost.
- The Kantorovich-Rubinstein duality gives us a more useful way to look at the problem. It tells that the minimal cost coincides with

$$\max\Big\{\int_{M} ufd\mu\big|u\colon M\to\mathbb{R} \text{ is } 1\text{-Lipschitz}\Big\}.$$

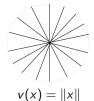
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One-dimensional localisation

Let v: M → ℝ be a maximiser.

 T ⊂ M − is a *transport ray* if it is a maximal set such that v|_T is an isometry.

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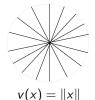
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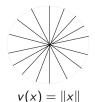
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• It follows that $\int_M f d\mu_T = 0$ for ν -almost every T.

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Generalisation of log-concavity

If µ is defined on ℝⁿ and is log-concave, then ν-almost every µ_T is log-concave. Moreover it is concentrated on intT.

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- If μ is satisfies $CD(\kappa, N)$, then ν -almost every $\mu_{\mathcal{T}}$ is satisfies $CD(\kappa, N)$.

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- Suppose that for some functions f_1, \ldots, f_m we have $\int_M f_i d\mu = 0$ for $i = 1, \ldots, m$. Does there exist a partition of M into m-dimensional pieces for which the related conditional measures would have zero integrals against f_1, \ldots, f_m ? Would it be possible for the pieces to retain the curvature-dimension properties of (M, d, μ) ?

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- Possible applications would include bounds for higher-order eigenvalues of Laplacian, multi-bubble problems, etc.
- Such generalisation was conjectured by Klartag to hold true in Euclidean spaces.

Conjecture - mass-balance condition

Multi-dimensional case

Let η be an ℝ^m-valued vector-measure on ℝⁿ with finite first moment and such that η(ℝⁿ) = 0.

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- Suppose η ≪ λ. It was conjectured by Klartag that η(A) = 0 for any transport set A.

Extensions of 1-Lipschitz functions

One-dimensional case

Theorem (C.)

▶ Let $A \subset \mathbb{R}^n$ and let $v : \mathbb{R}^n \to \mathbb{R}$ and $u : A \to \mathbb{R}$ be 1-Lipschitz functions.

One-dimensional case

Theorem (C.)

- Let $A \subset \mathbb{R}^n$ and let $v : \mathbb{R}^n \to \mathbb{R}$ and $u : A \to \mathbb{R}$ be 1-Lipschitz functions.
- Then there exists a 1-Lipschitz extension ũ of u such that ||u − v||_{A,∞} = ||ũ − v||_{ℝⁿ,∞}.

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Proof.

▶ Let \tilde{u}' be any 1-Lipschitz extension of u to \mathbb{R}^n . Let $\delta = \|u - v\|_{A,\infty}$.

Using the above theorem, one may show that if v is an optimal potential for $f d\mu$, then for any transport set $A \subset M$, $\int_A f d\mu = 0$.

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• Take
$$\tilde{u} = \tilde{u}' \vee (v - \delta) \wedge (v + \delta)$$
.

Using the above theorem, one may show that if v is an optimal potential for $f d\mu$, then for any transport set $A \subset M$, $\int_A f d\mu = 0$.

Multi-dimensional case

Theorem (Kirszbraun, '34)

Let $X \subset \mathbb{R}^n$ and let $v \colon X \to \mathbb{R}^m$ be a 1-Lipschitz map. Then there exists a 1-Lipschitz extension of v to \mathbb{R}^n .

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• Let
$$\nu_0 = \sum_{i=1}^3 v_i \delta_{x_i}$$
. For $\epsilon > 0$ set $\nu_{\epsilon} = \frac{1}{\lambda(B(0,\epsilon))} \sum_{i=1}^3 v_i \lambda_{B(x_i,\epsilon)}$.

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- Choose maximisers u_ε for ν_ε. Then ν_ε(B_ε) = 0 for any transport set associated with u_ε.
- ν_ε has to be concentrated on transport set of u_ε consisting of leaves of dimension at least one.
- ► For i = 1, 2, 3 let $N_{i\epsilon}$ be the set of all non-trivial leaves intersecting $B(x_i, \epsilon)$. Then $\nu_{\epsilon}(N_{i\epsilon}) = 0$.

Multi-dimensional case

- We shall show that the mass balance condition does not hold in multi-dimensional case m ≥ 2. Suppose on the contrary that it does.
- ► Choose any pairwise distinct x₁, x₂, x₃ ∈ ℝⁿ and v₁, v₂, v₃ ∈ ℝ^m in general position, with ∑³_{i=1} v_i = 0.

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$$\nu_0 = \sum_{i=1}^3 v_i \delta_{x_i}$$
. For $\epsilon > 0$ set $\nu_{\epsilon} = \frac{1}{\lambda(B(0,\epsilon))} \sum_{i=1}^3 v_i \lambda_{B(x_i,\epsilon)}$.

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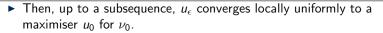
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- This implies that $\lambda(B(x_j, \epsilon) \cap N_{i\epsilon}) \neq 0$ for i, j = 1, 2, 3.
- ► Hence, there exist $(x_{rs}^{\epsilon}, x_{sr}^{\epsilon}) \in B(x_r, \epsilon) \times B(x_s, \epsilon)$ such that $||u_{\epsilon}(x_{rs}^{\epsilon}) u_{\epsilon}(x_{sr}^{\epsilon})|| = ||x_{rs}^{\epsilon} x_{sr}^{\epsilon}||.$

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- Indeed, by the isometric property we have equality in the triangle inequality || f(x₃) − f(x₁)|| ≤ || f(x₂) − f(x₁)|| + || f(x₃) − f(x₂)||. Strict convexity of balls in ℝ^m implies that assertion.

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- We extend u_0 to an affine 1-Lipschitz map on \mathbb{R}^n .
- ▶ We infer that for any 1-Lipschitz *u*

$$\sum_{i=1}^{3} \langle u(x_i), v_i \rangle \leq \sup \Big\{ \sum_{i=1}^{3} \langle f(x_i), v_i \rangle \mid f \text{ is affine and 1-Lipschitz} \Big\}.$$

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By the density of vectors in general position, we infer that for any 1-Lipschitz u and any w₁, w₂, w₃ ∈ ℝ³ that sum up to zero there is

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- ▶ The same ideas allow to show that similar result for any norm on \mathbb{R}^n and a strictly convex norm on \mathbb{R}^m . Also, similar conclusion will follow if we replace sup $\left\{ \int_{\mathbb{R}^m} \langle u, d\mu \rangle \mid u \colon \mathbb{R}^n \to \mathbb{R}^m \text{ is 1-Lipschitz} \right\}$, by maximisation over any uniformly closed subset of 1-Lipschitz maps.

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Theorem (C.)

$$\mathcal{I}(\eta) = \sup \Big\{ \int_{\mathbb{R}^n} \langle u, d\eta \rangle \big| u \colon X \to \mathbb{R}^m \text{ is } 1\text{-Lipschitz} \Big\}.$$

The same holds true for metric spaces (X, d) in lieu of $(\mathbb{R}^n, \|\cdot\|)$. This generalises the result for m = 1, where one can take π to be a non-negative measure with $P_1\pi = \mu_1$ and $P_2\pi = \mu_2$, $\mu = \mu_1 - \mu_2$.

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Suppose that η is absolutely continuous and has finite first moment, $\eta(\mathbb{R}^n) = 0$.

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 - 2. for any transport set A associated to v there is $\eta(A) = 0$.

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Curvature-dimension condition

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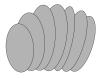
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• If η satisfies $CD(\kappa, N)$, then for ν -almost every S of dimension m, η_S satisfies $CD(\kappa, N)$. Moreover it is concentrated on intS.

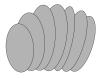
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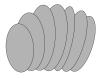
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- The idea of proof of CD(κ, N) for conditional measures on leaves builds upon work of Caffarelli, Feldman and McCann.

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Ghost subspaces (C.)

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- Such a subspace is termed *ghost subspace*. A ghost subspace is called trivial whenever it is equal to the tangent space to a leaf.
- If the ghost subspace for S is trivial then $\eta_S \ll \mathcal{H}_{\dim S}$.
- If all the ghost subspaces for an optimal potential for a vector measure η are trivial, then there exists an optimal transport and the mass-balance condition holds true.

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i) for any $A \subset \mathbb{R}^n$ and for any 1-Lipschitz map $u: A \to \mathbb{R}^m$ there exists 1-Lipschitz extension $\tilde{u}: \mathbb{R}^n \to \mathbb{R}^m$ of u such that

$$\|v-\tilde{u}\|_{\mathbb{R}^n,\infty}=\|v-u\|_{A,\infty}.$$

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ii) v is affine and 1-Lipschitz.

Thank you for your attention.

C. *Leaves decompositions in Euclidean spaces*, Journal de Mathématiques Pures et Appliquées, 2021

C. Leaves decompositions in Euclidean spaces II: ghost subspaces, in preparation, 2023

C. Optimal transport of vector measures, Calculus of Variations and Partial Differential Equations, 2021

C. Conitnuity of extensions of Lipschitz maps, Israel Journal of Mathematics, 2021

C. Optimal transport and 1-Lipschitz maps, D.Phil. dissertation, University of Oxford, 2020

One-dimensional case, relation to extensions of 1-Lipschitz maps

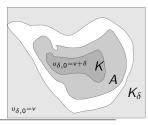
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- Let v attain max $\left\{ \int_M u f d\mu \mid u \colon M \to \mathbb{R} \text{ is } 1 \text{Lipschitz } \right\}$.
- ▶ Take a transport set $A \subset M$, $\epsilon > 0$, and a compact $K \subset A$ with $\int_{K^c \cap A} |f| d\mu \le \epsilon$. Pick $\delta > 0$ and set $u_{\delta,0} = v + \delta$ on K and $u_{\delta,0} = v$ on

$$\mathcal{K}_{\delta} = \{x \in M \mid \delta \leq \|x - y\| - |v(x) - v(y)| ext{ for all } y \in \mathcal{K}\}.$$

Then it is 1-Lipschitz and within δ -distance to v. We may extend it to M to u_{δ} , which is 1-Lipschitz and within δ -distance to v.



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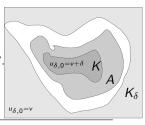
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Observe that A^c = ⋃_{δ>0} K_δ ∪ B. Here B is the set of points belonging to at least two transport rays of v; λ(B) = 0 − v is not differentiable on B.



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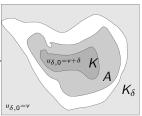
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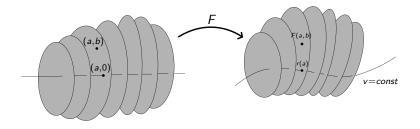
- Observe that A^c = ⋃_{δ>0} K_δ ∪ B. Here B is the set of points belonging to at least two transport rays of v; λ(B) = 0 − v is not differentiable on B.
- ► Moreover $\int_{M} \frac{v-u_{\delta}}{\delta} f d\mu \ge 0$. Sending δ to zero, we get $\int_{A} f d\mu \ge -2\epsilon$. Similarly, $\int_{A} (-f) d\mu \ge 0$.



Idea of the proof (C.)

Foliation

The formula F(a, b) = r(a) + Dv(r(a))*(b), where a ∈ ℝ^{n-m} and b ∈ ℝ^m, provides a local diffeomorphism, which is linear on the images of leaves of v. Here r is a local parametrisation of a fibre.



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Foliation, continuation

► Let P(x) denote the orthogonal projection onto the tangent space to the leaf containing x.

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Foliation, continuation

- Let P(x) denote the orthogonal projection onto the tangent space to the leaf containing x.
- By the area formula and Fubini theorem, the density on the leaves needs to be multiplied by |det(Id + P[⊥]D²v^{*}(P[⊥]·)(b))|. Notice the relation to the second fundamental form of the fibre of v.

Idea of the proof (C.)

Foliation, continuation

- Let P(x) denote the orthogonal projection onto the tangent space to the leaf containing x.
- By the area formula and Fubini theorem, the density on the leaves needs to be multiplied by |det(Id + P[⊥]D²v^{*}(P[⊥]·)(b))|. Notice the relation to the second fundamental form of the fibre of v.
- ▶ In the relative interiors of *m*-dimensional leaves *Dv* is Lipschitz:

$$2\sigma^2 \|Dv(x_1) - Dv(x_2)\|^2 \le \|x_1 - x_2\|^2 - \|v(x_1) - v(x_2)\|^2$$

where $\sigma = \min(d(x_1, \partial S(x_1)), d(x_2, \partial S(x_2))).$

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• Let $A \subset X \subset \mathbb{R}^n$. Suppose that $v \colon X \to \mathbb{R}^m$ is such that for all non-negative t_1, \ldots, t_m that sum up to one and all $x, x_1, \ldots, x_m \in X$ there is

$$\left\| \mathbf{v}(\mathbf{x}) - \sum_{i=1}^m t_i \mathbf{v}(\mathbf{x}_i) \right\| \leq \left\| \mathbf{x} - \sum_{i=1}^m t_i \mathbf{x}_i \right\|.$$

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The converse is proven for m = 1, 2, 3.

Rate of continuity

Let $A \subset B \subset \mathbb{R}^n$, $v \colon \mathbb{R}^n \to \mathbb{R}^m$ be 1-Lipschitz. Set $d_v(A, B) = \sup\{\|v(x) - v(y)\| | x \in A, y \in B\}.$

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Rate of continuity

Let $A \subset B \subset \mathbb{R}^n$, $v \colon \mathbb{R}^n \to \mathbb{R}^m$ be 1-Lipschitz. Set

$$d_{v}(A,B) = \sup\{\|v(x) - v(y)\| | x \in A, y \in B\}.$$

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- Let $u: A \to \mathbb{R}^m$ be 1-Lipschitz.
- Assume that $||u(x) v(x)|| \le \delta$ for $x \in A$.

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$$\|v(x) - \tilde{u}(x)\| \leq \sqrt{\delta^2 + 2\delta d_v(A, B)}$$

for all $x \in B$.

Proof.

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- On $A \times \{0\} \cup B \times \{\epsilon\} \subset \mathbb{R}^{n+1}$ we define w so that w(x, 0) = u(x) for $x \in A$ and $w(x, \epsilon) = v(x)$ for $x \in B$.

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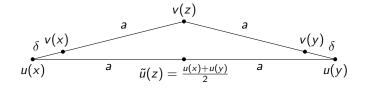
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- Let \tilde{w} be its 1-Lipschitz extension to \mathbb{R}^{n+1} .
- Define $\tilde{u}(x) = \tilde{w}(x, 0)$ for $x \in \mathbb{R}^n$.
- ► Then for $x \in B$, $\|\tilde{u}(x) v(x)\| = \|\tilde{w}(x,0) w(x,\epsilon)\| \le \epsilon$.

Optimality of the rate of continuity (C.) $_{\mbox{\sc Example}}$

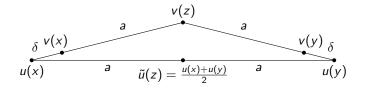
Let m > 1. Define u, v as in the picture, ||x - y|| = 2a, $z = \frac{x+y}{2}$. Set $A = \{x, y\}$, $B = \{x, y, z\}$.



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Let m > 1. Define u, v as in the picture, ||x - y|| = 2a, $z = \frac{x+y}{2}$. Set $A = \{x, y\}$, $B = \{x, y, z\}$.



Then, for any 1-Lipschitz extension \tilde{u} of u to B we have

$$\|v(z)-\tilde{u}(z)\|=\sqrt{\delta^2+2\delta a}.$$

If $\delta \geq a$, this is equal to $\sqrt{\delta^2 + 2\delta d_v(A, B)}$.

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