# Towards multi-dimensional localisation 

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Payne and Weinberger approach to Poincaré inequality - hyperplane bisections Motivating example from 1960

- 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 $\mu$ on $K$.

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

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- It suffices to prove the inequality for $K \cap H_{+}$and $K \cap H_{-}$.
- Reduction can be done as long as $\operatorname{dim} K \geq 2$.
- This procedure produces a measurable partition and conditional measures of $\mu$ with respect to the partition are log-concave.
- Therefore, it suffices to prove the inequality in the one-dimensional case.


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- Ohta generalised the technique to Finsler manifolds and Cavalletti and Mondino generalised it to metric measure spaces.


## Optimal transport

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- The Kantorovich-Rubinstein duality gives us a more useful way to look at the problem. It tells that the minimal cost coincides with

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\max \left\{\int_{M} u f d \mu \mid u: M \rightarrow \mathbb{R} \text { is } 1 \text {-Lipschitz }\right\}
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## Partition and disintegration

One-dimensional localisation

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- Then $\int_{A} f d \mu=0$ for any transport set $A$.



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- It follows that $\int_{M} f d \mu_{\mathcal{T}}=0$ for $\nu$-almost every $\mathcal{T}$.


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## Curvature-dimension condition $C D(\kappa, N)$

Generalisation of log-concavity

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- $C D(\kappa, N)$ may be understood as a condition that the Ricci curvature is bounded from below and the dimension is bounded from above.
- If $\mu$ is satisfies $C D(\kappa, N)$, then $\nu$-almost every $\mu_{\mathcal{T}}$ is satisfies $C D(\kappa, N)$.


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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 $(M, d, \mu)$ ?


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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 $\eta$ be an $\mathbb{R}^{m}$-valued vector-measure on $\mathbb{R}^{n}$ with finite first moment and such that $\eta\left(\mathbb{R}^{n}\right)=0$.

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- Suppose $\eta \ll \lambda$. It was conjectured by Klartag that $\eta(A)=0$ for any transport set $A$.


## Extensions of 1-Lipschitz functions

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Theorem (C.)

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


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Proof.

- Let $\tilde{u}^{\prime}$ 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 $\tilde{u}=\tilde{u}^{\prime} \vee(v-\delta) \wedge(v+\delta)$.

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## Kirszbraun theorem

Theorem (Kirszbraun, '34)
Let $X \subset \mathbb{R}^{n}$ and let $v: X \rightarrow \mathbb{R}^{m}$ be a 1-Lipschitz map. Then there exists a 1 -Lipschitz extension of $v$ to $\mathbb{R}^{n}$.

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- Choose any pairwise distinct $x_{1}, x_{2}, x_{3} \in \mathbb{R}^{n}$ and $v_{1}, v_{2}, v_{3} \in \mathbb{R}^{m}$ in general position, with $\sum_{i=1}^{3} v_{i}=0$.

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- Let $\nu_{0}=\sum_{i=1}^{3} v_{i} \delta_{x_{i}}$. For $\epsilon>0$ set $\nu_{\epsilon}=\frac{1}{\lambda(B(0, \epsilon))} \sum_{i=1}^{3} v_{i} \lambda_{B\left(x_{i}, \epsilon\right)}$.

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- Hence, there exist $\left(x_{r s}^{\epsilon}, x_{s r}^{\epsilon}\right) \in B\left(x_{r}, \epsilon\right) \times B\left(x_{s}, \epsilon\right)$ such that $\left\|u_{\epsilon}\left(x_{r s}^{\epsilon}\right)-u_{\epsilon}\left(x_{s r}^{\epsilon}\right)\right\|=\left\|x_{r s}^{\epsilon}-x_{s r}^{\epsilon}\right\|$.


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- By uniform convergence, $u_{0}$ is an isometry on $\left\{x_{1}, x_{2}, x_{3}\right\}$.
- We may take $x_{2}=t x_{1}+(1-t) x_{3}$ for some $t \in(0,1)$. Then any 1-Lipschitz map $f$ that is isometric on $\left\{x_{1}, x_{2}, x_{3}\right\}$ has to satisfy $f\left(x_{2}\right)=t f\left(x_{1}\right)+(1-t) f\left(x_{3}\right)$.

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- Indeed, by the isometric property we have equality in the triangle inequality $\left\|f\left(x_{3}\right)-f\left(x_{1}\right)\right\| \leq\left\|f\left(x_{2}\right)-f\left(x_{1}\right)\right\|+\left\|f\left(x_{3}\right)-f\left(x_{2}\right)\right\|$. Strict convexity of balls in $\mathbb{R}^{m}$ 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 $u$

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\sum_{i=1}^{3}\left\langle u\left(x_{i}\right), v_{i}\right\rangle \leq \sup \left\{\sum_{i=1}^{3}\left\langle f\left(x_{i}\right), v_{i}\right\rangle \mid f \text { is affine and } 1 \text {-Lipschitz }\right\}
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- By the density of vectors in general position, we infer that for any 1 -Lipschitz $u$ and any $w_{1}, w_{2}, w_{3} \in \mathbb{R}^{3}$ 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 $\sup \left\{\int_{\mathbb{R}^{m}}\langle u, d \mu\rangle \mid u: \mathbb{R}^{n} \rightarrow \mathbb{R}^{m}\right.$ is 1 -Lipschitz $\}$, by maximisation over any uniformly closed subset of 1-Lipschitz maps.


## Optimal transport of vector measures (C.)

Primal problem and duality

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\mathcal{I}(\eta)=\sup \left\{\int_{\mathbb{R}^{n}}\langle u, d \eta\rangle \mid u: X \rightarrow \mathbb{R}^{m} \text { is 1-Lipschitz }\right\} .
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The same holds true for metric spaces $(X, d)$ in lieu of $\left(\mathbb{R}^{n},\|\cdot\|\right)$. This generalises the result for $m=1$, where one can take $\pi$ to be a non-negative measure with $\mathrm{P}_{1} \pi=\mu_{1}$ and $\mathrm{P}_{2} \pi=\mu_{2}, \mu=\mu_{1}-\mu_{2}$.

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1. there exists an optimal transport $\pi \in \Gamma(\eta)$,
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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- The idea of proof of $C D(\kappa, N)$ for conditional measures on leaves builds upon work of Caffarelli, Feldman and McCann.

Ghost subspaces (C.)
Work in progress

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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.


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- If the ghost subspace for $\mathcal{S}$ is trivial then $\eta_{\mathcal{S}} \ll \mathcal{H}_{\operatorname{dim} \mathcal{S}}$.
- If all the ghost subspaces for an optimal potential for a vector measure $\eta$ are trivial, then there exists an optimal transport and the mass-balance condition holds true.

Continuity of extensions
Multi-dimensional case

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Let $m \geq 2, v: \mathbb{R}^{n} \rightarrow \mathbb{R}^{m}$. The following conditions are equivalent:
i) for any $A \subset \mathbb{R}^{n}$ and for any 1-Lipschitz map $u: A \rightarrow \mathbb{R}^{m}$ there exists 1-Lipschitz extension $\tilde{u}: \mathbb{R}^{n} \rightarrow \mathbb{R}^{m}$ of $u$ such that

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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

## Mass balance condition

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

- Let $v$ attain $\max \left\{\int_{M} u f d \mu \mid u: M \rightarrow \mathbb{R}\right.$ is 1 - Lipschitz $\}$.


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- Observe that $A^{c}=\bigcup_{\delta>0} K_{\delta} \cup B$. Here $B$ is the set of points belonging to at least two transport rays of $v ; \lambda(B)=0-v$ is not differentiable on $B$.
- Moreover $\int_{M} \frac{v-u_{\delta}}{\delta} f d \mu \geq 0$. Sending $\delta$ to zero, we get $\int_{A} f d \mu \geq-2 \epsilon$. Similarly, $\int_{A}(-f) d \mu \geq 0$.


Idea of the proof (C.)

- The formula $F(a, b)=r(a)+\operatorname{Dv}(r(a))^{*}(b)$, where $a \in \mathbb{R}^{n-m}$ and $b \in \mathbb{R}^{m}$, 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

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- By the area formula and Fubini theorem, the density on the leaves needs to be multiplied by $\left|\operatorname{det}\left(\operatorname{Id}+P^{\perp} D^{2} v^{*}\left(P^{\perp}.\right)(b)\right)\right|$. Notice the relation to the second fundamental form of the fibre of $v$.


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- In the relative interiors of m-dimensional leaves $D v$ is Lipschitz:

$$
2 \sigma^{2}\left\|D v\left(x_{1}\right)-D v\left(x_{2}\right)\right\|^{2} \leq\left\|x_{1}-x_{2}\right\|^{2}-\left\|v\left(x_{1}\right)-v\left(x_{2}\right)\right\|^{2}
$$

where $\sigma=\min \left(\mathrm{d}\left(x_{1}, \partial \mathcal{S}\left(x_{1}\right)\right), \mathrm{d}\left(x_{2}, \partial \mathcal{S}\left(x_{2}\right)\right)\right)$.

## Continuity of extensions

Arbitrary subsets of $\mathbb{R}^{n}$

## Theorem (C.)

- Let $A \subset X \subset \mathbb{R}^{n}$. Suppose that $v: X \rightarrow \mathbb{R}^{m}$ 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\|v(x)-\sum_{i=1}^{m} t_{i} v\left(x_{i}\right)\right\| \leq\left\|x-\sum_{i=1}^{m} t_{i} x_{i}\right\|
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- Let $A \subset X \subset \mathbb{R}^{n}$. Suppose that $v: X \rightarrow \mathbb{R}^{m}$ 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

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\left\|v(x)-\sum_{i=1}^{m} t_{i} v\left(x_{i}\right)\right\| \leq\left\|x-\sum_{i=1}^{m} t_{i} x_{i}\right\|
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- Suppose that $u: A \rightarrow \mathbb{R}^{m}$ is 1-Lipschitz.


## Continuity of extensions

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

Rate of continuity

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

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Theorem (C.)

- Let $u: A \rightarrow \mathbb{R}^{m}$ be 1-Lipschitz.
- Assume that $\|u(x)-v(x)\| \leq \delta$ for $x \in A$.
- Then there exists a 1-Lipschitz map ũ: $B \rightarrow \mathbb{R}^{m}$ such that $\tilde{u}(x)=u(x)$ for $x \in A$ and

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\|v(x)-\tilde{u}(x)\| \leq \sqrt{\delta^{2}+2 \delta d_{v}(A, B)}
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for all $x \in B$.

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- If we take $\epsilon=\sqrt{\delta^{2}+2 \delta d_{v}(A, B)}$, then $w$ is 1-Lipschitz.
- 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 $x \in B,\|\tilde{u}(x)-v(x)\|=\|\tilde{w}(x, 0)-w(x, \epsilon)\| \leq \epsilon$.

Optimality of the rate of continuity (C.)
Example

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


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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 $\sqrt{\delta^{2}+2 \delta d_{v}(A, B)}$.

