

Mini course GSSI

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**A variational regularity theory for optimal transportation,
and applications to the matching problem**

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Kantorowicz' formulation of Monge's optimal transp., direct method of calculus of variations

Given: (locally finite Borel) measures

$\lambda \geq 0$ on $\mathbb{R}^d \ni x$ and $\mu \geq 0$ on $\mathbb{R}^d \ni y$.

A measure $\pi \geq 0$ on $\mathbb{R}^d \times \mathbb{R}^d \ni (x, y)$ is “admissible”
iff it has marginals λ and μ :

$$\int \zeta(x) d\pi = \int \zeta d\lambda \quad \text{and} \quad \int \zeta(y) d\pi = \int \zeta d\mu.$$

Provided mass is finite and equal

$$\lambda(\mathbb{R}^d) = \mu(\mathbb{R}^d) \in (0, \infty),$$

the product measure $\pi = \frac{1}{\lambda(\mathbb{R}^d)} \lambda \otimes \mu = \frac{1}{\mu(\mathbb{R}^d)} \lambda \otimes \mu$ is admissible.

Note that for any admissible π

$$\pi(\mathbb{R}^d \times \mathbb{R}^d) = \lambda(\mathbb{R}^d) = \mu(\mathbb{R}^d) < \infty.$$

Consider squared transport distance and

$$\text{minimize } \int |y - x|^2 d\pi \quad \text{among all } \pi \text{ admissible.}$$

Provided λ, μ have finite second moments,

$$\int |x|^2 d\lambda < \infty \quad \text{and} \quad \int |y|^2 d\mu < \infty,$$

any admissible π satisfies (monotone convergence)

$$\frac{1}{2} \int |x - y|^2 d\pi \leq \int |x|^2 + |y|^2 d\pi = \int |x|^2 d\lambda + \int |y|^2 d\mu < \infty.$$

In particular, infimum $\in [0, \infty)$, and any minimizing sequence of π 's is tight, so that marginals are preserved in the limit. Since functional is lower semi-continuous (Fatou), get minimizer by direct method. We fix a minimizer π .

Kantorowicz potential and Brenier's map; convex duality

By convex duality \exists convex function $\psi: \mathbb{R}^d \rightarrow (-\infty, +\infty]$
(not $\equiv +\infty$) such that

$$\text{supp}\pi \subset \partial\psi,$$

where the subgradient $\partial\psi \subset \mathbb{R}^d \times \mathbb{R}^d$ is defined by

$$(x, y) \in \partial\psi \iff \forall x' \in \mathbb{R}^d \psi(x') \geq \psi(x) + (x' - x) \cdot y.$$

Informally $\text{supp}\pi$ is d -dimensional, as opposed to $2d$ -dimensional for product measure. In particular, we have

$$\text{supp}\lambda \subset \{x \mid \exists y (x, y) \in \partial\psi\} =: \mathcal{D}(\psi).$$

Suppose \exists an (open) ball $B \subset \mathbb{R}^d$ such that

$$B \subset \text{supp} \lambda.$$

Then have $B \subset \mathcal{D}(\psi) \subset \{x \mid \psi(x) < \infty\}$. As convex function, ψ is locally bounded and locally Lipschitz on B . As locally Lipschitz function, ψ is Lebesgue-almost everywhere differentiable on B . If ψ is differentiable in $x \in B$, then by definition $\{y \mid (x, y) \in \partial\psi\} = \{\nabla\psi(x)\}$. Hence there exists a Lebesgue null set $N \subset B$ such that

$$(x, y) \in \partial\psi \text{ and } x \in B - N \quad \implies \quad y = \nabla\psi(x).$$

If we suppose in addition

$$\lambda \ll dx \text{ on } B$$

then we obtain

$$\int_{B \times \mathbb{R}^d} \zeta(x, y) d\pi = \int_B \zeta(x, \nabla\psi(x)) d\lambda.$$

Eulerian perspective: trajectories X and flux q

We identify pairs $(x, y) \in \mathbb{R}^d \times \mathbb{R}^d$
with straight trajectories $X: [0, 1] \rightarrow \mathbb{R}^d$ via (the Borel map)

$$X_t = ty + (1 - t)x \quad \text{so that} \quad \dot{X} = y - x.$$

Let the vectorial Borel measure q be defined through

$$\int \xi \cdot dq = \int \int_0^1 \xi(X_t) \cdot \dot{X} dt d\pi,$$

where ξ is a bounded smooth vector field on \mathbb{R}^d . Note q has finite total variation since integrand $\leq \sup |\xi|$ times $|y - x| \leq |x| + |y| \leq 1 + \frac{1}{2}|x|^2 + \frac{1}{2}|y|^2$.

Applying definition to gradient fields $\xi = \nabla\zeta$, appealing to the chain rule $\nabla\zeta(X_t) \cdot \dot{X} = \frac{d}{dt}\zeta(X_t)$, and to $\int_0^1 \nabla\zeta(X_t) \cdot \dot{X} dt = \zeta(y) - \zeta(x)$, we obtain by the admissibility of π

$$\int \nabla\zeta \cdot dq = \int \zeta(d\mu - d\lambda).$$

Incidentally, this means

$$-\nabla \cdot dq = d\mu - d\lambda \quad \text{distributionally.}$$

In view of this we think of q as a flux.

Entering and exiting times σ, τ and measures f, g for a ball

Given a closed ball $\bar{B} \subset \mathbb{R}^d$,

define $\Omega_{\bar{B}}$ to be the set of trajectories that spend time in \bar{B} :

$$\Omega_{\bar{B}} \stackrel{\text{short}}{=} \Omega := \{ X = (x, y) \mid \exists t \in [0, 1] X_t \in \bar{B} \}.$$

Define the two Borel functions $\sigma_{\bar{B}}, \tau_{\bar{B}}$ or short $\sigma, \tau: \Omega \rightarrow [0, 1]$ to be the times X enters/exits \bar{B} :

$$\begin{aligned} \sigma(X) &:= \min\{t \in [0, 1] \mid X_t \in \bar{B}\} \\ &\leq \max\{t \in [0, 1] \mid X_t \in \bar{B}\} =: \tau(X). \end{aligned}$$

Define the two Borel measures $f_{\bar{B}}, g_{\bar{B}} \geq 0$, or short f, g , where the trajectories enter or exit:

$$\int \zeta df = \int_{\Omega \cap \{\sigma > 0\}} \zeta(X_\sigma) d\pi \quad \text{and} \quad \int \zeta dg = \int_{\Omega \cap \{\tau < 1\}} \zeta(X_\tau) d\pi;$$

well-defined because of $\pi(\mathbb{R}^d \times \mathbb{R}^d) < \infty$.

Since by definition,

$$\left\{ \begin{array}{l} \sigma(X) > 0 \iff X_{\sigma(X)} \in \partial B \\ \tau(X) < 1 \iff X_{\tau(X)} \in \partial B \end{array} \right\}$$

we have

f, g are supported on ∂B .

Claim

$$\int_{\bar{B}} \nabla \zeta \cdot dq = \int_{\bar{B}} \zeta(d\mu - d\lambda) + \int_{\partial B} \zeta(dg - df).$$

Apply definition of q to $\xi = I(\bar{B})\nabla\zeta$,
 use $\int_0^1 \xi(X_t) \cdot \dot{X} dt = \int_\sigma^\tau \nabla\zeta(X_t) \cdot \dot{X} dt = \zeta(X_{\tau(X)}) - \zeta(X_{\sigma(X)})$. Since

$$\sigma(X) = 0 \iff x \in \bar{B} \quad \text{and} \quad \tau(X) = 1 \iff y \in \bar{B},$$

we get

$$\begin{aligned} \int_0^1 \xi(X_t) \cdot \dot{X} dt &= I(y \in \bar{B})\zeta(y) - I(x \in \bar{B})\zeta(x) \\ &\quad + I(\tau(X) < 1)\zeta(X_{\tau(X)}) - I(\sigma(X) > 0)\zeta(X_{\sigma(X)}). \end{aligned}$$

Integrate against π , use admissibility of π and definition of f, g .

Incidentally,

$$\begin{aligned} \text{normal trace of } q \text{ on } \partial B &= g - f \\ \text{provided } |q|(\partial B) &= \lambda(\partial B) = \mu(\partial B) = 0. \end{aligned}$$

Electrostatics: the Helmholtz projection ∇u of q on B ; some regularity theory

Helmholtz projection $\mathcal{H}_B = \mathcal{H}$ on B is $L^2(B, \mathbb{R}^d)$ -orthogonal projection onto closed subspace of gradient fields. By singular integral theory, if ξ is smooth on \bar{B} , then $\mathcal{H}\xi$ is smooth on \bar{B} , and the $C^k(\bar{B})$ -norm of $\mathcal{H}\xi$ is controlled by the $C^{k+1}(\bar{B})$ -norm of ξ . Moreover, \mathcal{H} is characterized by how it acts on smooth fields, namely

$$\mathcal{H}\nabla\zeta = \nabla\zeta \quad \text{for smooth } \zeta \text{ on } \bar{B},$$

$$\mathcal{H}\xi = 0 \quad \text{for smooth divergence-free } \xi \text{ supported in } B.$$

Hence to every distribution f on \bar{B} , we can associate its Helmholtz projection $\mathcal{H}f$ by duality via $\mathcal{H}f.\xi = f.\mathcal{H}\xi$. It is characterized by

$$\begin{cases} \mathcal{H}f.\nabla\zeta = f.\nabla\zeta & \text{for smooth } \zeta \text{ on } \bar{B}, \\ \mathcal{H}f.\xi = 0 & \text{for smooth divergence-free } \xi \text{ supported in } B. \end{cases}$$

As finite measure, $f = q|_{\bar{B}}$ is a distribution.

Claim: $\mathcal{H}f$ is absolutely continuous w. r. t. Lebesgue:

$$\mathcal{H}q|_{\bar{B}} \ll dx|_B.$$

Enough to construct a $u_B = u \in H^{1,1}(B)$ such that

$$\int_B \nabla \zeta \cdot \nabla u dx = \int_{\bar{B}} \nabla \zeta \cdot dq;$$

then we have $\mathcal{H}q|_{\bar{B}} = \nabla u dx|_B$. Enough to establish

$$\int_B \nabla \zeta \cdot \nabla u dx = \int_{\bar{B}} \zeta (d\mu - d\lambda) + \int_{\partial B} \zeta (dg - df).$$

Consider $\int_{\bar{B}} \zeta(d\mu - d\lambda) + \int_{\partial B} \zeta(dg - df)$ as a linear form in ζ .
 It is bounded w. r. t. $\sup_{\bar{B}} |\zeta|$; it vanishes for constant ζ .
 By Sobolev embedding

$$\sup_{x,y \in \bar{B}} \frac{|\zeta(y) - \zeta(x)|}{|y - x|^\alpha} \lesssim \left(\int_B |\nabla \zeta|^p dx \right)^{\frac{1}{p}}$$

form is bounded w. r. t. $\nabla \zeta \in L^p(B, \mathbb{R}^d)$ for $p \in (d, \infty)$. By duality
 it can be represented by $\int_B \nabla \zeta \cdot \tilde{q} dx$ for some $\tilde{q} \in L^{p'}(B, \mathbb{R}^d)$ with
 $p' \in (1, \frac{d}{d-1})$. Then ∇u is the Helmholtz projection of \tilde{q} , which
 by singular integral theory is bounded in $L^{p'}(B, \mathbb{R}^d)$. In particular
 $u \in H^{1,p'}(B) \subset H^{1,1}(B)$.

Incidentally, u satisfies the Poisson equation with Neumann b. c.:

$$-\Delta u = \mu - \lambda \text{ in } B \quad \text{and} \quad \nu \cdot \nabla u = \nu \cdot q \text{ on } \partial B \quad \text{in a weak sense.}$$

**Relating the Eulerian flux q
to the Lagrangian displacement $(y - x)d\pi$, locally**

From definition of q

$$\begin{aligned} & \int \xi(x) \cdot (dq - (y - x)d\pi) \\ &= \int_0^1 dt \int \left(\xi(ty + (1 - t)x) - \xi(x) \right) \cdot (y - x)d\pi, \end{aligned}$$

we obtain the inequality

$$\left| \int \xi(x) \cdot (dq - (y - x)d\pi) \right| \leq \sup |\nabla \xi| \int_0^1 dt \int t|y - x|^2 d\pi,$$

which entails

$$\left| \int \xi(x) \cdot (dq - (y - x)d\pi) \right| \leq \frac{1}{2} \sup |\nabla \xi| \int |y - x|^2 d\pi.$$

Seek version with transportation cost localized to a ball B ;

$$E(B) := \int_{\Omega(B)} |y - x|^2 d\pi.$$

Replace ξ by $I(\bar{B})\xi$ in definition of q , split difference into

$$\begin{aligned} I(X_t \in \bar{B})\xi(X_t) - I(x \in \bar{B})\xi(x) &= I(X_t \in \bar{B})I(x \in \bar{B})(\xi(X_t) - \xi(x)) \\ &\quad + I(X_t \in \bar{B}, x \notin \bar{B})\xi(X_t) - I(X_t \notin \bar{B}, x \in \bar{B})\xi(x). \end{aligned}$$

First contribution as before:

$$\begin{aligned} &\left| \int_0^1 dt \int I(X_t \in \bar{B})I(x \in \bar{B})(\xi(X_t) - \xi(x)) \cdot (y - x) d\pi \right| \\ &\leq \sup_{\bar{B}} |\nabla \xi| \int_0^1 dt \int I(X_t \in \bar{B}) |X_t - x| |y - x| d\pi \leq \sup_{\bar{B}} |\nabla \xi| \frac{1}{2} E(\bar{B}). \end{aligned}$$

Second contribution:

$$\begin{aligned}
& \left| \int_0^1 dt \int \left(I(X_t \in \bar{B}, x \notin \bar{B}) \xi(X_t) \right. \right. \\
& \quad \left. \left. - I(X_t \notin \bar{B}, x \in \bar{B}) \xi(x) \right) \cdot (y - x) d\pi \right| \\
& \leq \sup_B |\xi| \int_0^1 dt \int |I(X_t \in \bar{B}) - I(x \in \bar{B})| |y - x| d\pi.
\end{aligned}$$

Specify to a ball $\bar{B} = \bar{B}_R$ with radius R and write $|I(X_t \in \bar{B}) - I(x \in \bar{B})| = |I(R \geq |X_t|) - I(R \geq |x|)|$. Hence integral in R is estimated by $||X_t| - |x|| \leq |X_t - x|$ to the effect of

$$\begin{aligned}
& \int_0^{\bar{R}} dR \sup_{\xi} \frac{1}{\sup_{\bar{B}_R} |\xi|} \left| \int_0^1 dt \int \left(I(X_t \in \bar{B}_R, x \notin \bar{B}_R) \xi(X_t) \right. \right. \\
& \quad \left. \left. - I(X_t \notin \bar{B}_R, x \in \bar{B}_R) \xi(x) \right) \cdot (y - x) d\pi \right| \leq \frac{1}{2} E(B_{\bar{R}}).
\end{aligned}$$

We summarize these findings on the average-in- R estimate of a dual norm of $dq - (y - x)d\pi$ in

Lemma 1.

$$\int_0^{\bar{R}} dR \sup_{\xi} \frac{\left| \int_{\bar{B}_R} \xi(x) \cdot (dq - (y - x)d\pi) \right|}{\max\{\sup_{\bar{B}_R} |\xi|, \bar{R} \sup_{\bar{B}_R} |\nabla \xi|\}} \leq E(B_{\bar{R}}).$$

We now comment on the regime in which Lemma 1 is not vacuous. Note that the l. h. s. compares $dq|_{\bar{B}_R}$ to the marginal in x of $(y - x)d\pi|_{(\bar{B}_R \times \mathbb{R}^d)}$, in a norm that scales like the total variation (but is weaker more like the flat norm). Hence Lemma 1 is meaningful if and only if $\int_0^{\bar{R}} dR \int_{\bar{B}_R \times \mathbb{R}^d} |y - x| d\pi$ is small compared to the r. h. s. that by definition dominates $\int_{B_{\bar{R}} \times \mathbb{R}^d} |y - x|^2 d\pi$. This is the case if

$$|y - x| \ll \bar{R} \quad \text{on average w. r. t. } \pi|_{(B_{\bar{R}} \times \mathbb{R}^d)}.$$

Loosely speaking, this means

$$\text{transportation distance} \ll \text{localization scale}.$$

**The flux q is close to its Helmholtz projection ∇u ;
almost in total variation norm**

From now on we need

$$\lambda = dx \quad \text{in } B_{\bar{R}}.$$

In this case

$$\int_{\bar{B} \times \mathbb{R}^d} \zeta(x, y) d\pi = \int_B \zeta(x, \nabla \psi(x)) dx.$$

Hence expression in Lemma 1 turns into

$$\int_{\bar{B} \times \mathbb{R}^d} \xi(x) \cdot (dq - (y - x) d\pi) = \int_{\bar{B}} \xi(x) \cdot (dq - (\nabla \psi(x) - x) dx).$$

Note that by definition of Helmholtz projection on B (on $L^2(B, \mathbb{R}^d)$) we have $\mathcal{H}(\nabla \psi - \text{id}) = \nabla \psi - \text{id}$. Together with $\nabla u dx|_B = \mathcal{H}q|_{\bar{B}}$ we have in terms of the Helmholtz projection (on distributions)

$$dq|_{\bar{B}} - \nabla u dx|_B = (\text{id} - \mathcal{H})(dq|_{\bar{B}} - (\nabla \psi - \text{id}) dx|_B).$$

Note that like \mathcal{H} , the “Leray projection” $\text{id} - \mathcal{H}$ is bounded in the Hölder space $C^{1,\alpha}(\bar{B}, \mathbb{R}^d)$ for $\alpha \in (0, 1)$; more precisely, it is uniformly in B bounded w. r. t. the norm

$$\sup_{\bar{B}} |\xi| + R^{1+\alpha} \sup_{x,y \in \bar{B}} \frac{|\nabla \xi(x) - \nabla \xi(y)|}{|x - y|^\alpha},$$

where R is the radius of B . We appeal to the embeddings

$$\sup_{\bar{B}_R} |\xi| + R \sup_{\bar{B}_R} |\nabla \xi| \lesssim \text{above norm on } B_R \lesssim \sup_{\bar{B}_R} |\xi| + R^2 \sup_{\bar{B}_R} |\nabla^2 \xi|.$$

Corollary 1. of Lemma 1

$$\int_{\frac{\bar{R}}{2}}^{\bar{R}} dR \sup_{\xi} \frac{\left| \int_{\bar{B}_R} \xi \cdot (dq - \nabla u_R dx) \right|}{\sup_{\bar{B}_R} |\xi| + R^2 \sup_{\bar{B}_R} |\nabla^2 \xi|} \lesssim E(B_{\bar{R}}).$$

Corollary 1 expresses closeness in a norm that is weaker than the total variation norm; it is even weaker than the flat norm.

In particular, we cannot take $\xi = I(\hat{B})e$ some some unit vector $e \in \mathbb{R}^d$ and some ball \hat{B} . However, we will obtain an estimate as if we had control in the total variation norm, provided we average in the radius r of such a ball \hat{B}_r . This follows from a more subtle statement on the boundedness of the Leray projection:

$$\xi_{rR} := \text{Leray projection of } I(\hat{B}_r)e \text{ in } B_R$$

can be (not quite uniquely) written in form of

$$\xi_{rR} = I(\hat{B}_r)\xi_{rR}^{in} + I(B_R)\xi_{rR}^{out},$$

where both $\xi_{rR}^{in/out}$ are smooth provided \hat{B}_r is compactly contained in B_R . This allows us to apply Lemma 1 to

$$\begin{aligned} \int_{\hat{B}_r} e \cdot (dq - \nabla u_R dx) &= \int_{B_R} \xi_{rR} \cdot (dq - (\nabla \psi(x) - x)dx) \\ &= \int_{B_r} \xi_{rR}^{in} \cdot (dq - (\nabla \psi(x) - x)dx) + \int_{B_R} \xi_{rR}^{out} \cdot (dq - (\nabla \psi(x) - x)dx). \end{aligned}$$

In order to quantify smoothness, fix center of $\hat{B}_r \in B_{\frac{\bar{R}}{8}}$; then

$$\text{dist}(\hat{B}_r, B_R^c) \geq \frac{R}{4} \quad \text{as} \quad r \leq \frac{\bar{R}}{8} \quad \text{and} \quad \frac{\bar{R}}{2} \leq R \leq \bar{R}.$$

By translation invariance, center of \hat{B}_r fixed; by scaling invariance, $r = 1$. Then $\xi_{R,r=1}^{in/out}$ converge as $R \uparrow \infty$; hence smoothness is uniform in R . This (informally) establishes the estimates

$$\left. \begin{aligned} & \max\{\sup_{B_r} |\xi_{rR}^{in}|, r \sup_{B_r} |\nabla \xi_{rR}^{in}|\} \\ & \max\{\sup_{B_R} |\xi_{rR}^{out}|, r \sup_{B_R} |\nabla \xi_{rR}^{out}|\} \end{aligned} \right\} \lesssim 1.$$

Proposition 1.

$$\int_{\frac{\bar{R}}{2}}^{\bar{R}} dR \int_0^{\frac{\bar{R}}{8}} dr \left| \int_{\hat{B}_r} (dq - \nabla u_R dx) \right| \lesssim \bar{R} E(B_{\bar{R}}).$$

Optimal semidiscrete matching, heuristics, main result

Matching a law λ to its empirical measure μ

Scaling of mean-square Wasserstein distance $W(\lambda, \mu)$
by Ajtai-Komlòs-Tusnády

Approximation by Helmholtz projection, small-scale divergence.

A cut-off on scales \ll particle distance

Implementation by Ambrosio-Stra-Trevisan, on macroscopic scales

Heuristics by Carraciolo-Lucibello-Parisi-Sicuro,
on mesoscopic scales

Comparison of the Parisi-et-al. heuristics to ours

Heuristics made rigorous by Goldman-Huesmann-O.,
on mesoscopic scales

Matching a law λ to its empirical measure μ

Specify to $\lambda(\mathbb{R}^d) = 1$, i. e. to a probability measure.

Given $N \in \mathbb{N}$, draw $Y_1, \dots, Y_N \in \mathbb{R}^d$ be N independent samples distributed according to λ .

Consider $\mu := \frac{1}{N} \sum_{n=1}^N \delta_{Y_n}$, “empirical measure”.

The probability measure μ on \mathbb{R}^d is random.

As $N \uparrow \infty$, μ weakly converges to λ , almost surely.

Monitor the (squared) Wasserstein distance

$W^2(\lambda, \mu) := \inf \{ \int |y - x|^2 d\pi \mid \pi \text{ admissible for } \lambda, \mu \}$.

“Semi-discrete matching”.

Scaling of mean-square Wasserstein distance $W(\lambda, \mu)$ by Ajtai-Komlòs-Tusnády

Simplest case:

λ = uniform distribution on a cube Q_L of side length L .

Ignore probability normalization: $\lambda = dx|_{Q_L}$;

use number density normalization: $N = L^d \in \mathbb{N}$ and $\mu = \sum_{n=1}^N \delta_{Y_n}$.

Monitor $\sqrt{\mathbb{E} \frac{1}{N} W^2(\lambda, \mu)}$

= (mean-square) expected transportation distance per point.

Theorem 1 (Ajtai, Komlós, Tusnády '84).

$$\sqrt{\mathbb{E} \frac{1}{N} W^2(\lambda, \mu)} \sim \left\{ \begin{array}{ll} 1 & \text{for } d > 2, \\ \sqrt{\ln N} & \text{for } d = 2, \\ \sqrt{N} & \text{for } d = 1 \end{array} \right\}$$

Hence transportation distance \ll system size ($= L$) for all d ,
but transportation distance \sim particle distance ($= 1$) iff $d > 2$.
Hence $d = 2$ is the critical dimension.

Approximation by Helmholtz projection

Consider the distributional Helmholtz projection on Q_L of $\mu - dx$; given by $\nabla u dx|_{Q_L}$ characterized through

$$\int_{Q_L} \nabla \zeta \cdot \nabla u dx = \int_{Q_L} \zeta (d\mu - dx).$$

Informally, ∇u is solution of Neumann-Poisson problem

$$-\Delta u = \mu - dx \text{ in } Q_L \quad \text{and} \quad \nu \cdot \nabla u = 0 \text{ in } \partial Q_L.$$

By Section 1 $\int_{B \times \mathbb{R}^d} (y - x) d\pi \approx \int_B \nabla u dx$ for most balls $B \subset Q_L$ of (localization) radius $R \gg$ transportation distance $\sim \sqrt{\ln N}$.

Ignoring contribution of scales $\lesssim \sqrt{\ln N}$ to macroscopic output

naively expect $W^2(\lambda, \mu) = \int |y - x|^2 d\pi \approx \int_{Q_L} |\nabla u|^2 dx$;

use in averaged form of $\frac{1}{N} W^2(\lambda, \mu) \approx \frac{1}{|Q_L|} \int_{Q_L} |\nabla u|^2 dx$.

Small scale divergence in $d \geq 2$

However, since points have capacity zero in $d \geq 2$, meaning that Dirac $\delta \notin H^{-1}(Q_L)$, we have $\int_{Q_L} |\nabla u|^2 dx =: \int_{Q_L} ||\nabla|^{-1}(\mu - dx)|^2 dx = +\infty$.

Need to cut off scales $\lesssim \sqrt{\ln N}$; via spectral implementation:

$L^2(Q_L)$ -normalized eigenfunctions/-values of Neumann-Laplacian:

$$e_m(x) := \left(\frac{2}{L}\right)^{\frac{d}{2}} \prod_{i=1}^d \cos\left(\frac{\pi m_i x_i}{L}\right), \quad \lambda_m = \left(\frac{\pi|m|}{L}\right)^2 \text{ for } m \in \mathbb{N}_0^d - \{0\}.$$

$$\text{Plancherel: } \int_{Q_L} |\nabla u|^2 = \sum_{m \neq 0} \frac{1}{\lambda_m} \left(\int_{Q_L} e_m(d\mu - dx) \right)^2.$$

Second moments of shot noise $\mu - dx$ as if it were white noise:

$$\mathbb{E} \left(\int_{Q_L} e_m(d\mu - dx) \right)^2 = \int_{Q_L} e_m^2 dx = 1.$$

We recover $\mathbb{E} \int_{Q_L} |\nabla u|^2 = \sum_{m \neq 0} \left(\frac{L}{\pi|m|} \right)^2 = +\infty$ iff $d \geq 2$.

Implementation by Ambrosio-Stra-Trevisan, on macroscopic scale

Define the cut-off version $\nabla \bar{u}$ of ∇u
by removing the wavelengths $\frac{\pi|m|}{L} > \sqrt{\ln N}$,
i. e. by projecting on the wave numbers $|m| \leq \frac{L\sqrt{\ln N}}{\pi}$.

$$\begin{aligned} \text{Get } \mathbb{E} \frac{1}{|Q_L|} \int_{Q_L} |\nabla \bar{u}|^2 dx &= \sum_{m \in \mathbb{N}_0^2, 0 < |m| \leq \frac{L\sqrt{\ln N}}{\pi}} \left(\frac{1}{\pi|m|} \right)^2 \\ &\approx \frac{1}{4} \frac{2\pi}{\pi^2} \ln \frac{L\sqrt{\ln N}}{\pi} \approx \frac{1}{4\pi} \ln N \quad \text{since } L = \sqrt{N}. \end{aligned}$$

Theorem 2 (Ambrosio, Stra, Trevisan '19). *For $d = 2$*

$$\mathbb{E} \frac{1}{N} W^2(\lambda, \mu) \approx \frac{1}{4\pi} \ln N \quad \text{for } N \gg 1.$$

Heuristics by Carraciolo-Lucibello-Parisi-Sicuro '14, on mesoscopic level

Recall convex duality from Section 1:

\exists convex $\phi: \mathbb{R}^d \rightarrow (-\infty, \infty]$ such that $\text{supp}\phi \subset \partial\phi$.

Assume momentarily that $\text{supp}\mu = [0, L]^d$ and $\mu \ll dy$.

Then $\int_{\mathbb{R}^d \times [0, L]^d} \zeta(x, y) d\pi = \int_{[0, L]^d} \zeta(\nabla\phi(y), y) d\mu$,

by admissibility of π $\int_{[0, L]^d} \zeta(x) dx = \int_{[0, L]^d} \zeta(\nabla\phi(y)) d\mu$.

Assume momentarily $\nabla\phi$ is diffeomorphism of $[0, L]^d$.

Then $\int_{[0, L]^d} \zeta(\nabla\phi(y)) \det D^2\phi(y) dy = \int_{[0, L]^d} \zeta(\nabla\phi(y)) \frac{d\mu}{dy} dy$.

Get Monge-Ampère equation $\det D^2\phi = \frac{d\mu}{dy}$.

Expect Section 1: $\nabla\phi \approx \text{id}$

when averaged over scales \gg transportation distance.

Writing $\nabla v := \nabla\phi - \text{id}$, naively expect $\det D^2\phi \approx 1 + \text{tr} D^2v$

when averaged over scales \gg transportation distance.

To leading order, ∇v would be characterized by $\text{tr} D^2v = \frac{d\mu}{dy} - 1$.

Comparison of the Parisi-et-al. heuristics to ours

Parisi et al.'s heuristics predicts $(x - y)d\pi \approx \nabla v$
when averaged on scales \gg transportation distance
where $\Delta v dy = \mu - dy$.

Our heuristics predicts $(y - x)d\pi \approx q \approx \nabla u$
when averaged on scales \gg transportation distance
where $-\Delta u dx = \mu - dx$.

The two predictions are identical.

The two heuristics reflect the two faces of the Laplace operator:
divergence-form $-\nabla \cdot \nabla u$ vs. non-divergence form $\text{tr} D^2 v$.

Also reflect the two faces of the Monge-Ampère equation:
Euler-Lagrange equation of $\inf \{ \int |y - x|^2 d\pi \mid \pi \text{admissible} \}$
vs. fully non-linear equation $\det D^2 \phi = \frac{d\mu}{dy}$ (Caffarelli, Figalli).

Heuristics made rigorous by Goldman-Huesmann-O.

Mesoscopic vs. macroscopic closeness of $(y - x)d\pi$ to ∇u .

“Mesoscopic” means \gg particle distance,

“macroscopic” means \ll system size.

As before, $\lambda = dx|_{Q_L}$. More generally as before:

(deterministic) μ with $\text{supp}\mu \subset Q_L$ and $\mu(Q_L) = |Q_L|$

Monitor closeness of μ to Lebesgue measure dy on set $B \subset \mathbb{R}^d$:

$$D(B) := W^2(\mu|_B, \frac{\mu(B)}{|B|}dy|_B) + |B|R^2(\frac{\mu(B)}{|B|} - 1)^2.$$

Back to $\mu = \sum_{n=1}^N \delta_{Y_n}$ with $N = L^d$ and Y_n 's indep. and uniform in Q_L .

Consider all concentric balls $B = B_R \subset Q_L$, center = origin. Have for $d = 2$ by Ajtai et al. with overwhelming probability

$$D(Q_L) \sim |Q_L| \ln L \text{ and } D(B_R) \sim |B_R| \ln R \text{ for } 1 \ll R.$$

Convenient to introduce a rate functional $T: [0, \infty) \rightarrow [0, \infty)$, called sublinear iff $\forall 0 < \theta \ll 1 \ \forall R \ \theta T(R) \ll T(\theta R)$.

Recall: π is the minimizer of $\{ \int |y - x|^2 d\pi \mid \pi \text{ admissible for } \lambda, \mu \}$, $\nabla u dx|_{Q_L}$ is distributional Helmholtz projection on Q_L of $\mu - dx|_{Q_L}$.

Theorem 3 (Goldman, Huesmann, O. '21). *If μ is such that there exists a sublinear rate function T and a radius $\bar{r} \leq \frac{L}{4}$ with*

$$D(Q_L) \leq |Q_L| T^2(L) \text{ and } D(B) \leq |B_R| T^2(R) \text{ for } \bar{r} \leq R \leq \frac{L}{4}$$

then π and ∇u are related by

$$\frac{1}{\bar{r}} \int_{\frac{\bar{r}}{2}}^{\bar{r}} dr \left| \frac{1}{\pi(\mathbb{R}^d \times B_r)} \int_{\mathbb{R}^d \times B_r} (y - x) d\pi - \frac{1}{|B_r|} \int_{B_r} \nabla u dx \right| \lesssim \int_{\bar{r}}^L \frac{dr}{r} \frac{T^2(r)}{r}.$$

Theorem 4 (Goldman, Huesmann, O. '21). *If μ is such that there exists a sublinear rate function T and a radius $\bar{r} \leq \frac{L}{4}$ with*

$$D(Q_L) \leq |Q_L|T^2(L) \text{ and } D(B) \leq |B_R|T^2(R) \text{ for } \bar{r} \leq R \leq \frac{L}{4}$$

then π and ∇u are related by

$$\frac{1}{\bar{r}} \int_{\frac{\bar{r}}{2}}^{\bar{r}} dr \left| \frac{1}{\pi(\mathbb{R}^d \times B_r)} \int_{\mathbb{R}^d \times B_r} (y - x) d\pi - \frac{1}{|B_r|} \int_{B_r} \nabla u dx \right| \lesssim \int_{\bar{r}}^L \frac{dr}{r} \frac{T^2(r)}{r}.$$

Displacement $y - x$ π -averaged over $\mathbb{R}^d \times B_r$

\approx electrostatic field ∇u dx -averaged over $x \in B_r$.

R. h. s. is average over radii $r \sim \bar{r}$, cf. Proposition 1.

For empirical measure in $d = 2$:

$\bar{r} \sim 1$ and $T(r) \sim \sqrt{\ln r}$ (which clearly is sublinear),
so that l. h. s. $\lesssim 1 =$ particle distance.

Hence on scales \sim particle distance, have

displacement = electrostatic field + $O(\text{particle distance})$.

Best possible closeness.