

Summer School on Optimal Transport

A variational regularity theory for optimal transportation

Felix Otto,
Max Planck Institute for Mathematics in the Sciences,
Leipzig, Germany

joint work with Michael Goldman (arXiv '17, Ann. ENS '20),
with MG & Martin Huesmann (to appear in CPAM),
with MH & Francesco Mattesini (arXiv)

version June 24st 2022

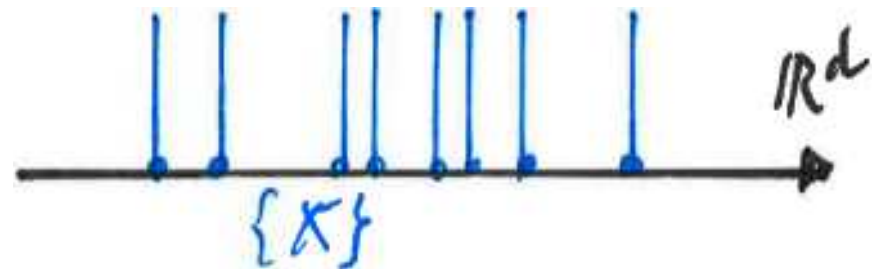
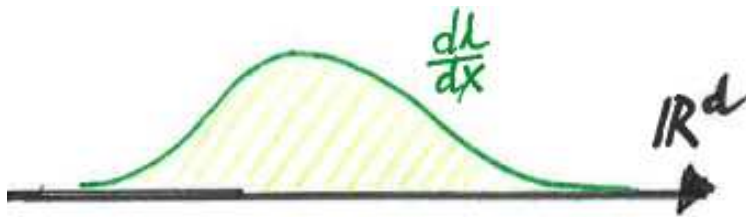
**An application to the matching problem
of our variational regularity theory
for optimal transportation**

A natural application for optimal transportation ...

Given λ probability measure/law on \mathbb{R}^d ,

draw N independent samples X_1, \dots, X_N .

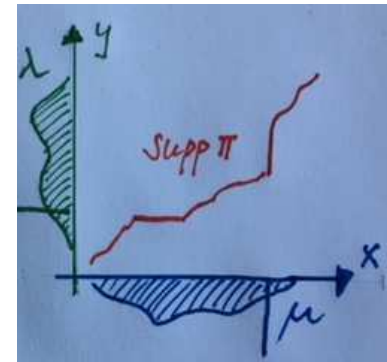
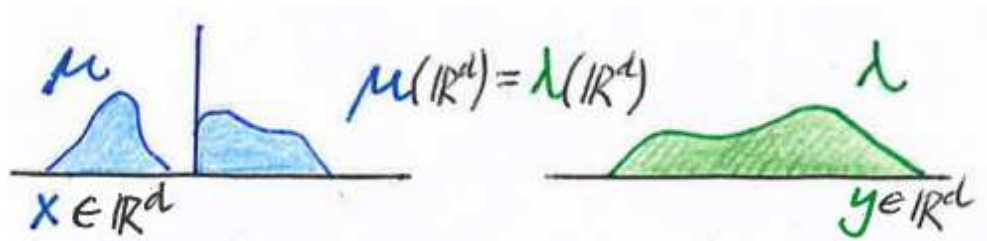
Consider empirical measure $\mu = \frac{1}{N} \sum_{n=1}^N \delta_{X_n}$.



How close are λ and μ ?

Optimal Transportation in Kantorowicz' formulation

Given two measures



seek transfer plan π , i. e. $\pi(U \times \mathbb{R}^d) = \mu(U)$, $\pi(\mathbb{R}^d \times V) = \lambda(V)$
that minimizes Euclidean transport cost $\int_{\mathbb{R}^d \times \mathbb{R}^d} |y-x|^2 \pi(dx dy)$;
= coupling π that maximizes covariance.

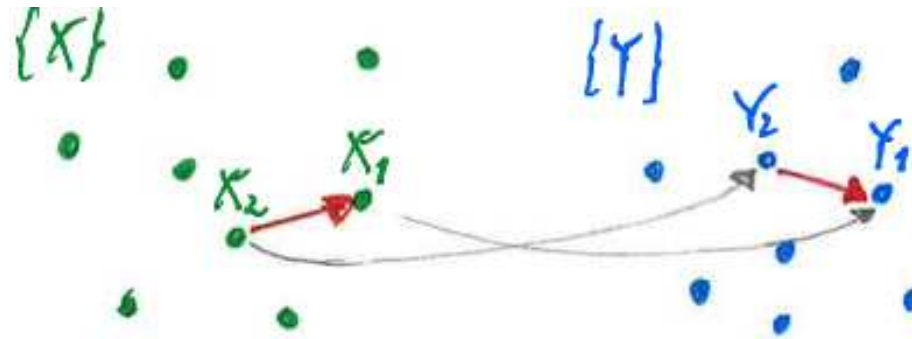
Minimum $=: W_2^2(\mu, \lambda)$ (squared) Wasserstein distance.

Optimal matching = Optimal transport

Two independent sets $\{X\}$, $\{Y\}$ of samples from λ .

Wasserstein distance between $\sum_{\{X\}} \delta_X$ and $\sum_{\{Y\}} \delta_Y$.

optimal transport
=optimal matching



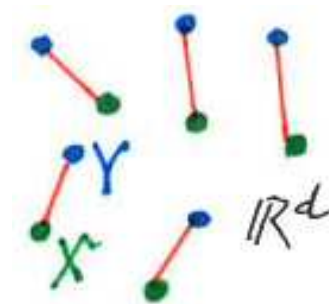
For the sake of discussion,

we also consider $|y - x|^p$ for $p \in (0, \infty)$ next to $|y - x|^2$.

Optimality for matching of infinite point clouds

Focus on mesoscopic behavior: $\lambda = \text{Lebesgue} \implies$
 $\{X\}, \{Y\}$ indep. samples of Poisson point processes.

Matching of two infinite locally finite
point clouds $\{X\}$ and $\{Y\} \subset \mathbb{R}^d$,
amounts to a **pairing** $\{(X, Y)\} \subset \mathbb{R}^d \times \mathbb{R}^d$.



Optimality means:

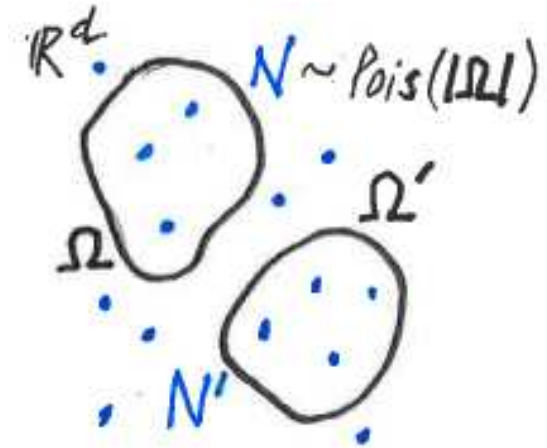
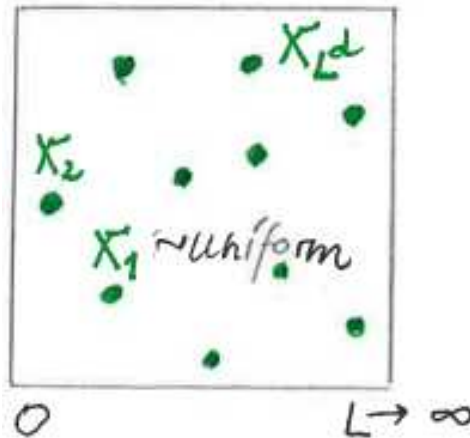
\forall matched finite subset
 $\{X_1, \dots, X_N\}$
 $\{Y_1, \dots, Y_N = Y_0\}$.

$$\begin{aligned} & \sum_{n=1}^N |Y_n - X_n|^2 \\ & \leq \sum_{n=1}^N |Y_{n-1} - X_n|^2 \end{aligned}$$

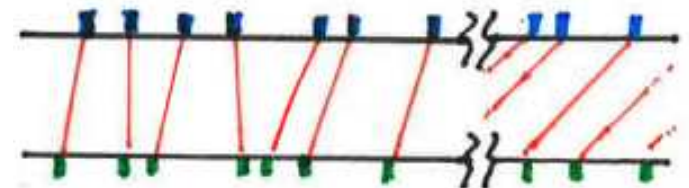
The Poisson point process

Locally finite point cloud via Poisson point process of unit intensity (means that distance between points $O(1)$)

canonical vs.
grand-canonical
definition

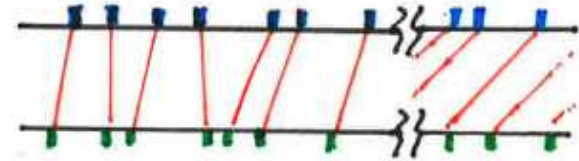


Seek optimal matching
of two independent
Poisson point processes
– divergent behavior
for $d = 1$ and $p > 1$
(implies monotonicity)

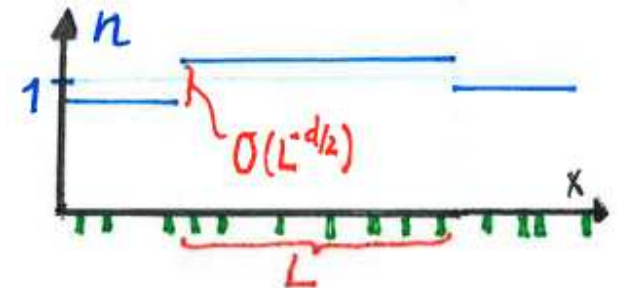


Matching depends on dimension d ...

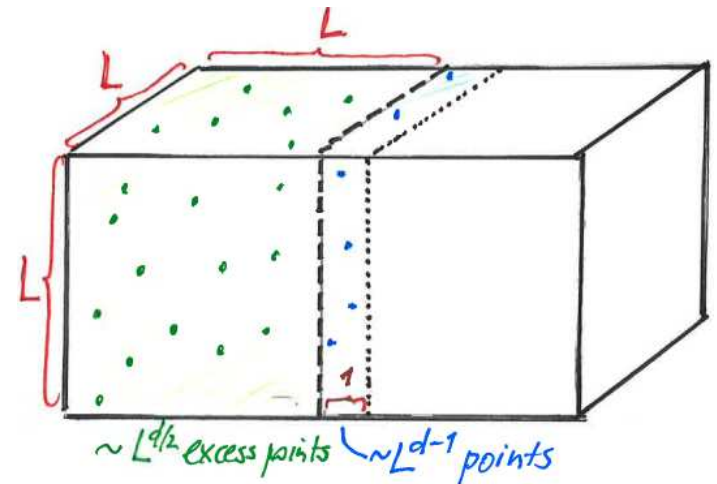
Cyclically monotone matching of two independent Poisson point processes
– distances diverge like square root for $d = 1$.



Fluctuations of number density n
 $= O(L^{-\frac{d}{2}})$; lower for higher d .



Number of excess points $= O(L^{\frac{d}{2}})$,
number of points in
(width one) boundary layer $= O(L^{d-1})$.



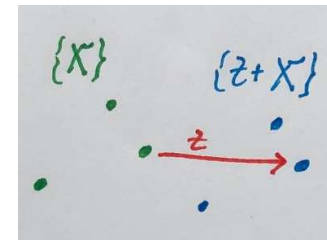
... critical dimension $d = 2$

Impose statistical translation invariance (“Stationarity”) of matching

Optimal matching of two independent Poisson point processes $\{X\}$, $\{Y\}$.

Poisson point process is *stationary*:

\forall shift vectors $z \in \mathbb{R}^d$ $\{z + X\} =_{\text{law}} \{X\}$.



Seek random point cloud $\{(X, Y)\}$ in $\mathbb{R}^d \times \mathbb{R}^d$ s. t. marginals are independent Poisson point processes, coupling is optimal almost surely,

and $\forall z \in \mathbb{R}^d$ $\{(z + X, z + Y)\} =_{\text{law}} \{(X, Y)\}$.

Critical dimension $d = 2$ rigorously captured

Interest in Combinatorics (eg. Ajtai et al. '84),
Probability Theory (Talagrand '92+, Holroyd-Peres '11+),
Physics (eg. Parisi et al. '14),
Analysis (eg. Ambrosio et al. '16+).

Seek random point cloud $\{(X, Y)\}$ in $\mathbb{R}^d \times \mathbb{R}^d$ s. t.
marginals are independent Poisson point processes,
coupling is cyclically monotone almost surely,
and $\forall z \in \mathbb{R}^d \quad \{(z + X, z + Y)\} =_{\text{law}} \{(X, Y)\}$.

Theorem (Huesmann&Sturm '13)

For $d > 2$ have existence.

Theorem (H.&Mattesini&0. '21)

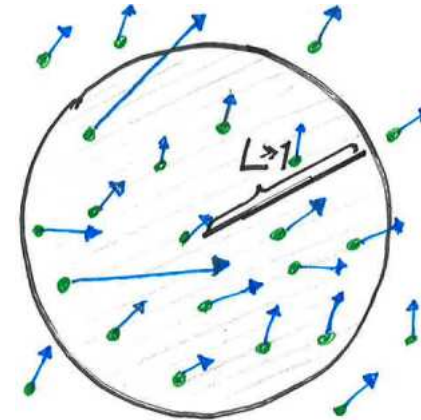
For $d \leq 2$ have non-existence.

**Mesoscopic analysis of the matching problem
via harmonic approximation of displacement,
arising from our variational regularity theory**

Indirect argument for non-existence in $d = 2$

Ergodicity: Most particles in B_L are moved at most $O(1)$.

Monotonicity: All particles in B_L are moved at most $o(L)$.



Local energy $E := L^{-d} \sum_{X \in B_L} |Y - X|^2 = o(L^2)$ is non-dimensionally small.

Local data $D := L^{-d} W_2^2(\sum_{X \in B_L} \delta_X, \text{uniform on } B_L) = O(\ln L)$ is non-dimensionally **much** smaller (same for $\sum_{Y \in B_L} \delta_Y$).

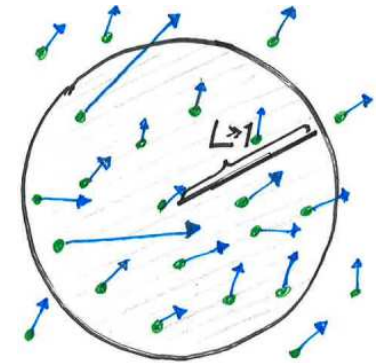
... want to transmit smallness of D to E

From D to E by harmonic approximation

Recall local energy and data size:

$$E := L^{-d} \sum_{X \in B_L} |Y - X|^2 = o(L^2),$$

$$D := L^{-d} W_2^2(\sum_{X \in B_L} \delta_X, \text{uniform on } B_L) = O(\ln L).$$



Harmonic approximation: $\exists \phi$ harmonic s. t.

$$L^{-d} \sum_{X \in B_L} |(Y - X) - \nabla \phi(X)|^2 \leq o(E) + O(D).$$

By $\sup_{B_L} |\nabla \phi|^2 \lesssim L^{-d} \int_{B_{2L}} |\nabla \phi|^2$ and iteration $L \uparrow \infty$:

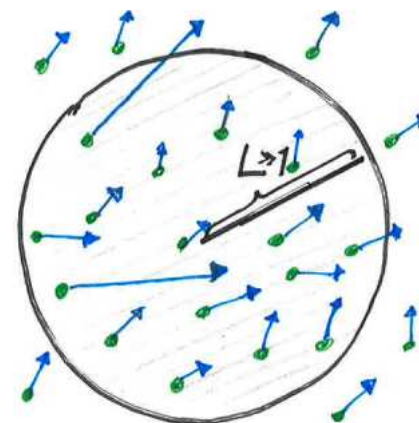
$$E = L^{-d} \sum_{X \in B_L} |Y - X|^2 \leq O(\ln L).$$

... smallness of D does transmit to E

The final contradiction

Most particles in B_L
are moved at most $O(1)$,

Mean *square*
transportation distance in B_L
at most $O(\ln L)$.



Combines to:

Mean transportation distance in B_L at most $o(\ln^{\frac{1}{2}} L)$.

On the other hand:

Mean transportation distance in B_L at least $O(\ln^{\frac{1}{2}} L)$,
by $W_1(\sum_{X \in B_L} \delta_X, \text{uniform in } B_L) \gtrsim \ln^{\frac{1}{2}} L$.

Why is the result not trivial?

Huesmann-Sturm existence applies to general p ;
have existence provided (morally speaking)

$$\lim_{R \uparrow \infty} \frac{1}{|B_R|} \mathbb{E} W_p^p(\mu_{\perp} B_R, \kappa dx_{\perp} B_R) < \infty,$$

i. e. the transportation cost per particle stays finite.

Our non-existence applies to $d = 2, p = 2$
($p \in (1, \infty)$ in preparation with L. Koch) and uses

$$\lim_{R \uparrow \infty} \frac{1}{|B_R| \ln^{\frac{1}{2}} R} \mathbb{E} W_1(\mu_{\perp} B_R, \kappa dx_{\perp} B_R) > 0.$$

Holroyd-Pemantle-Peres-Schramm '09 for $d = 1, p = \frac{1}{2}$
established existence of a stationary matching;

any stationary matching satisfies

$$\lim_{R \uparrow \infty} \frac{1}{|B_R|} \mathbb{E} W_{\frac{1}{2}}^{\frac{1}{2}}(\mu_{\perp} B_R, \kappa dx_{\perp} B_R) = \infty.$$

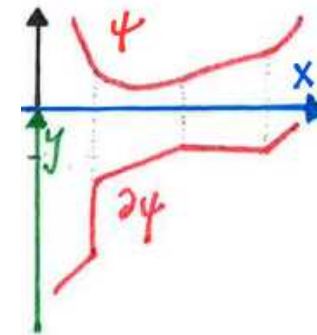
**A variational approach
to the regularity theory
for optimal transportation;
At its core: harmonic approximation**

From optimal transportation to Monge-Ampère

Minimize $\int_{\mathbb{R}^d \times \mathbb{R}^d} |y - x|^2 \pi(dx dy)$
among all $\pi(dx dy)$ with marginals $\mu(dx)$ and dy .

Support of optimal transfer plan π
is cyclically monotone;
hence \exists convex ψ

$\text{supp } \pi \subset \{ (x, y) \mid y \in \text{sub-gradient } \partial\psi(x) \}$.



\forall test functions ζ $\int \zeta(\nabla\psi(x)) \mu(dx) = \int \zeta(y) dy$.

In smooth case, this amounts to $\det D^2\psi = \mu$,
an instance of the Monge-Ampère equation.

Nature of the Monge-Ampère equation

Recall Monge-Ampère: $\det D^2\psi = 1$.

Fully non-linear with $F(A) := \det A - 1$.

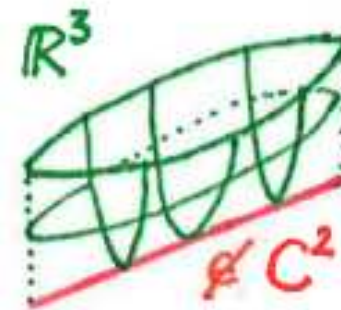
However elliptic: $F(A) > F(A')$ for $A > A' \geq 0$;
satisfies comparison principle.

However degenerate: \leftrightarrow affine invariant (non-compact $SL(d)$).

Cf. Laplacian $F(A) = \text{tr} A$: rotation invariant (compact $SO(d)$).

Caffarelli's '90 breakthrough:

comparison principle,
affine invariance,
compactness.



Pogorelov's
example is
worst case

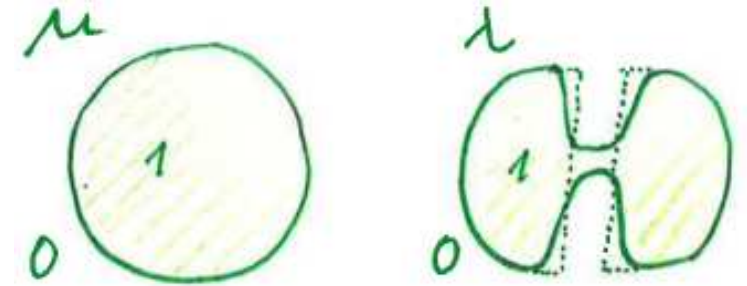
Monge-Ampère equation at crossroads
of fully nonlinear and variational.

Singularities in optimal transportation are generic

Caffarelli's example:

smooth data μ, λ

do not yield smooth $T = \nabla\psi$.



See also Loeper's example in Riemannian setting.

Thus ϵ -regularity is of interest (Figalli-Kim, DePhilippis-Figalli):

$$\int |y-x|^2 \pi(dx dy) \leq \epsilon \text{ locally, } \mu, \lambda \text{ smooth locally}$$

$$\implies \text{Kantorowicz potential } \psi \text{ smooth locally.}$$

Relies on harmonic approximation in our approach:

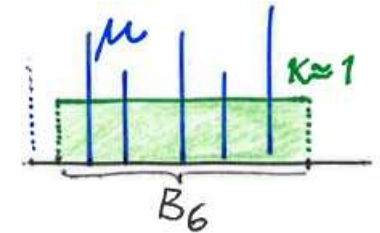
$$\int |y-x|^2 \pi(dx dy) \leq \epsilon \text{ locally, } \mu, \lambda \approx 1 \text{ locally}$$

$$\implies \text{displacement } (y-x)\pi(dx dy) \approx \nabla \text{harmonic locally.}$$

Statement of our harmonic approximation result

Local energy $E := \int_{(B_3 \times \mathbb{R}^d) \cup (\mathbb{R}^d \times B_3)} |y - x|^2 \pi(dx dy),$

Local data² $D := W_2^2(\mu \llcorner B_3, \kappa dx \llcorner B_3) + (\kappa - 1)^2$
 $+ \text{ same for } \lambda$



Proposition 1 (Goldman&Huesmann&O.)

$\forall \theta > 0 \quad \exists \epsilon(\theta, d) > 0, C(\theta, d) < \infty \quad \text{s. t.} \quad E + D \leq \epsilon \implies$

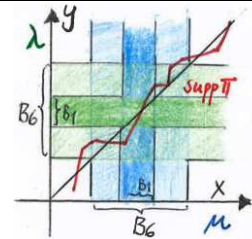
$\exists \nabla \phi$ harmonic, $\int_{B_1} |\nabla \phi|^2 \leq C(E + D),$

$\int_{(B_1 \times \mathbb{R}^d) \cup (\mathbb{R}^d \times B_1)} |y - x - \nabla \phi(x)|^2 \pi(dx dy) \leq \theta E + CD.$

Amounts to:

Displacement $y - x$

\approx harmonic gradient $\nabla \phi$



Plan for mini-course

Motivate connection between OT
and the Poisson equation

Elucidate the main ideas behind Proposition 1

Harmonic approximation: correct homogeneities ...

$$E := \int_{(B_3 \times \mathbb{R}^d) \cup (\mathbb{R}^d \times B_3)} |y - x|^2 \pi(dx dy), \text{ quadratic in solution,}$$

$$D := W_2(\mu \llcorner B_3, \kappa_\mu dy \llcorner B_3) + (\kappa_\mu - 1)^2 + \text{same for } \lambda, \text{ quadratic in data.}$$

Proposition 1

$$\forall \theta > 0 \quad \exists \epsilon(\theta, d) > 0, C(\tau, d) < \infty \quad \text{s. t.} \quad E + D \leq \epsilon \implies$$

$$\exists \nabla \phi \text{ harmonic,} \quad \int_{B_1} |\nabla \phi|^2 \leq C(E + D),$$

$$\int_{\{\exists t \in [0, 1] \ X(t) \in \bar{B}_1\}} \int_{\sigma}^{\tau} |\dot{X}(t) - \nabla \phi(X(t))|^2 dt \pi(dx dy)$$

$$\leq \theta E + CD.$$

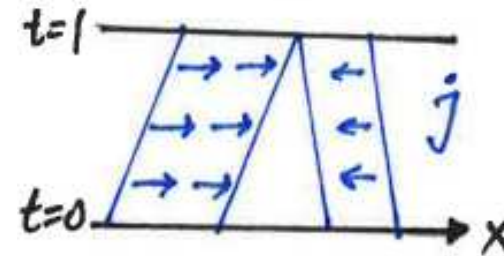
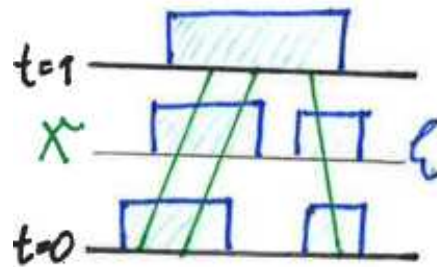
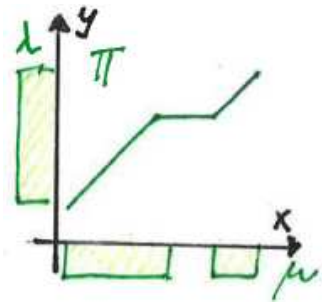
Compare to $\int_{B_1} L(\nabla u - \nabla \phi) \leq \tau \int_{B_6} L(\nabla u) + C \int_{B_6} |f|^2$

for $-\nabla \cdot DL(\nabla u) = \nabla \cdot f$ with uniformly convex L .

... and correct metric

From Lagrangian to Eulerian (Benamou-Brenier)

Transport plan π , trajectories $X(t)$, density/flux (ρ, j)



Continuity eqn. $\partial_t \rho + \nabla \cdot j = 0$, kinetic energy $\int_{\mathbb{R}^d \times (0,1)} \frac{1}{\rho} |j|^2$.

$$W^2(\mu, \lambda) = \inf \left\{ \int \frac{1}{\rho} |j|^2 \mid \partial_t \rho + \nabla \cdot j = 0, \rho(t=0) = \lambda, \rho(t=1) = \mu \right\}$$

Kinetic energy density $(\rho, j) \mapsto \frac{1}{\rho} |j|^2$ is mostly strictly CONVEX.

Linearization $\frac{1}{\rho} |j|^2 \rightsquigarrow |j|^2$ amounts to $W^2(\lambda, \mu) \rightsquigarrow \|\lambda - \mu\|_{H^{-1}}^2$.

Analogy for minimal surfaces: varifolds vs. currents.

Eulerian version of harmonic approximation

Density/flux $(\rho, j) = (\rho_t dt, j_t dt)$ where

$$\int \zeta d\rho_t = \int \zeta(ty + (1-t)x) \pi(dxdy), \quad \int \xi \cdot dj_t = \int \xi(ty + (1-t)x) \cdot (y-x) \pi(dxdy)$$

continuity
equation

$$\begin{array}{c} \xi = \lambda \\ \hline t=1 \\ \partial_t \rho + \nabla \cdot j = 0 \\ \hline t=0 \\ \xi = \mu \end{array}$$

$$E \rightsquigarrow \int_{B_5 \times (0,1)} \frac{1}{\rho} |j|^2$$

reveals strict convexity
of variational problem

Proposition 1'

$$\forall \theta > 0 \quad \exists \epsilon(\theta, d) > 0, \quad C(\tau, d) < \infty \quad \text{s. t.} \quad E + D \leq \epsilon \implies$$

$$\exists \nabla \phi \text{ harmonic,} \quad \int_{B_1} |\nabla \phi|^2 \leq C(E + D),$$

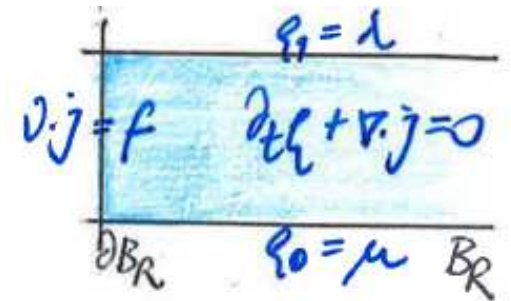
$$\int_{B_1 \times (0,1)} \frac{1}{\rho} |j - \rho \nabla \phi|^2 \leq \theta E + CD.$$

Amounts to: Eulerian velocity $\frac{j}{\rho} \approx \nabla \phi$ harmonic gradient

Construct $\nabla\phi$ via flux (Neumann) data

Normal flux $\nu \cdot j$ across bdry ∂B_R ,

its time integral $\int_0^1 \nu \cdot j dt$.



Proposition 1" $\forall \theta > 0 \exists \epsilon > 0, C < \infty : E + D \leq \epsilon \implies$

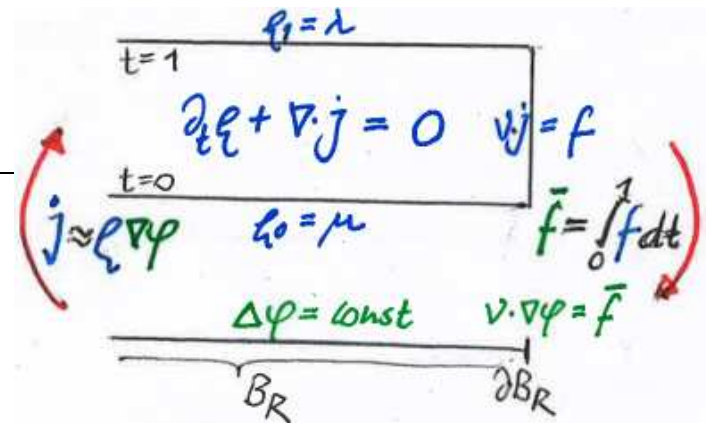
$\exists R \in [1, 2]$ s. t. $\Delta\phi = \text{const}$ in B_R , $\nu \cdot \nabla\phi = \int_0^1 \nu \cdot j dt$ on

∂B_R satisfies $\int_{B_1} |\nabla\phi|^2 \leq C(E + D),$

$$\int_{B_1 \times (0,1)} \frac{1}{\rho} |j - \rho \nabla\phi|^2 \leq \theta E + CD.$$

cf. Dacorogna-Moser.

Choice of "good" radius R .



Analogies to De Giorgi's approach to regularity for minimal surfaces (Schoen&Simon '82)

Approximate minimal surface by harmonic graph / approximate displacement by harmonic gradient.

Use: Object is minimizing under compact perturbations.
Don't use: Euler-Lagrange equation (= first variation).

Mismatch of type of boundary condition for construction of harmonic competitor:
graph vs. non-graph / time-averaged vs. time-resolved;

Lower-dimensional isoperimetric estimate:
error is of higher-order (choice of good radius).

Use of strict convexity to convert energy gap into distance ("approximate orthogonality");
need to smooth out boundary data.

ϵ -regularity as Schauder theory

Recall Hölder semi-norms $[u]_{\alpha, B} := \sup_{x \neq x' \in B} \frac{|u(x) - u(x')|}{|x - x'|^\alpha}$, $\alpha \in (0, 1)$.

Suppose $\lambda = \lambda dx$, $\mu = \mu dy$

with Hölder continuous λ, μ and $\lambda(0) = \mu(0) = 1$.

Monitor (the dimensionless) $E := \frac{1}{R^2 |B_R|} \int_{\Omega \cap B_{2R}} |T - x|^2 dx$

with $T = \nabla \psi$ Brenier map.

Monitor (the dimensionless) $D := R^{2\alpha} [\lambda]_{\alpha, B_{2R}}^2 + R^{2\alpha} [\mu]_{\alpha, B_{2R}}^2$.

Theorem 1 (Goldman&O., à la DePhilippis&Figalli)

If $E + D \ll 1$ then $R^{2\alpha} [\nabla T]_{\alpha, B_R}^2 \lesssim E + D$.

Amounts to $C^{2,\alpha}$ -regularity for Monge-Ampère $\det D^2 \psi = \lambda$ ($\mu \equiv 1$).

Comparison: DePhilippis&Figalli vs. Goldman&O.

Theorem 1 (Goldman&O., à la DePhilippis&Figalli)

$$E + [\lambda]_{\alpha, B_2}^2 + [\mu]_{\alpha, B_2}^2 \ll 1 \implies [D^2\psi - \text{id}]_{\alpha, B_1}^2 \lesssim E + [\lambda]_{\alpha, B_2}^2 + [\mu]_{\alpha, B_2}^2.$$

Perturbation around **linear** $\Delta\phi = \text{const}$,

as opposed to **nonlinear** $\det D^2\psi = 1$ (ϵ -regularity Figalli&Kim).

Get (immediately) linear homogeneities $[D^2\psi - \text{id}]_{\alpha, B_1} \lesssim [\lambda]_{\alpha, B_2} + [\mu]_{\alpha, B_2}$.

Get $\psi \in C^{2,\alpha}$ **in one step**,

as opposed to **three-step bootstrap** $C^{1,\alpha}, C^{1,1}, C^{2,\alpha}$.

Competitors in strictly convex variational problem,
instead of **comparison principle**.