

Constrained Wasserstein Fitting Problems

Or: “How Best to Fill a Region with a Curve?”

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The Problem

Given:

1. A region $\Omega \subseteq \mathbb{R}^2$,
2. A finite length ℓ of rope,

Find:

- The rope shape that best “fills” Ω .

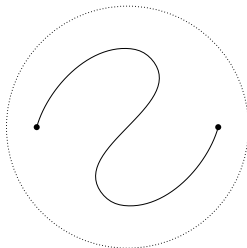
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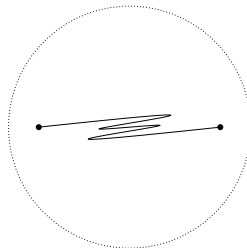
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(a) A “good” filling



(b) A “bad” filling

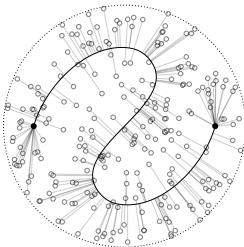
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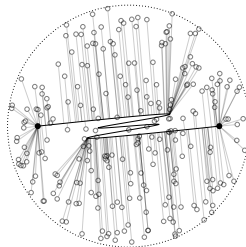
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Takeaway

Objective: Find a curve $f : [0, 1] \rightarrow \Omega$ minimizing

$$\mathcal{G}(f) = \int_{\Omega} d(\omega, f([0, 1])) \, d\omega$$

subject to $\text{length}(f) \leq \ell$.

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- ▶ Can we work with $f : X \rightarrow \Omega$ where $X \subseteq \mathbb{R}^m$, $\Omega \subseteq \mathbb{R}^n$, $m < n$?

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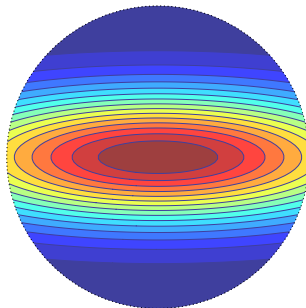
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- ▶ Can we work with $f : X \rightarrow \Omega$ where $X \subseteq \mathbb{R}^m$, $\Omega \subseteq \mathbb{R}^n$, $m < n$?
✓ Sure, but some care required.

Other Measures

Example: Choose $\rho \in \mathcal{P}(\Omega)$ more concentrated on the horizontal:

$$\rho \sim \mathcal{N}(0; [\sigma_X^2 \ 0; \ 0 \ \sigma_Y^2]) \quad \sigma_Y \ll \sigma_X$$

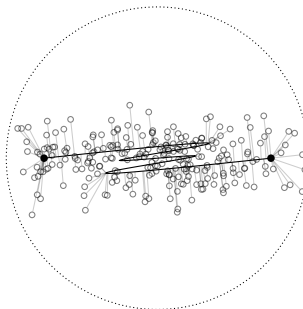
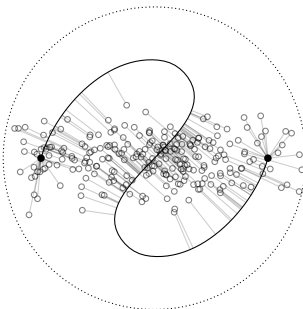


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Other Dimensions

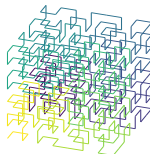
Example: If $f : [0, 1]^2 \rightarrow [0, 1]^3$, what should we pick to be the constraint?

- ▶ Maybe surface area?

Other Dimensions

Example: If $f : [0, 1]^2 \rightarrow [0, 1]^3$, what should we pick to be the constraint?

- ▶ Maybe surface area? **X** Unfortunately, no.
- ▶ Let $g : [0, 1] \rightarrow [0, 1]^3$ and take $f(x_1, x_2) = g(x_1)$
- ▶ g can still “fill” space while $\text{Area}(f) = 0$:



Choosing a Constraint $\mathcal{C}(f)$

Want $\mathcal{C}(f)$ to measure “complexity” of f :

- ▶ If f nonconstant, $\mathcal{C}(f) > 0$
 - prev slide: m -dimensional Hausdorff measure fails

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 - relevance in applications

Choice:

$$\mathcal{C}(f) = \|f\|_{W^{k,q}(X;\Omega)}$$

(assume $kq > m$ for Sobolev inequality)

Revised Problem Statement

Given:

- ▶ $X \subseteq \mathbb{R}^m, \Omega \subseteq \mathbb{R}^n$
- ▶ $\rho \in \mathcal{P}(\Omega)$ (measure to approximate)
- ▶ $\ell \geq 0$ (budget)
- ▶ $p \geq 1$ (weight of far-away points)
- ▶ Some technical hypotheses

Minimize:

$$\mathcal{I}_p(f) = \int_{\Omega} d^p(\omega, f) \, d\rho(\omega)$$

Subject to:

$$\|f\|_{W^{k,q}(X;\Omega)} \leq \ell$$

Related Work

- ▶ Quantization problem for measures (e.g. [Iac17])
 - Optimal approximation of ρ by Dirac masses
 - $m = 0$, $\mathcal{C} = \#$ of points
- ▶ Principal Curves ([KS17], [LS21], [DF20])
 - Like our problem for $m = 1$, $\mathcal{C}(f) = \text{length}(f)$
 - Common to use soft penalty:

$$\mathcal{G}_p^\lambda(f) = \mathcal{G}_p(f) + \lambda \mathcal{C}(f)$$

- ▶ Unequal-dimensional OT ([MP19])

Connection to OT

- ▶ Wasserstein Cost: Given $\mu \in \mathcal{P}(X), \nu \in \mathcal{P}(Y)$ ($X, Y \subseteq \mathbb{R}^n$)

$$\mathbb{W}_p(\mu, \nu) = \left(\inf_{\gamma \in \Gamma(\mu, \nu)} \int_{X \times Y} |x - y|^p d\gamma(x, y) \right)^{1/p}$$

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- ▶ $\mathbb{W}_p(\mu, \nu)$ gives metric that respects “structure” of \mathbb{R}^n
- ▶ $\mathcal{G}_p \approx$ measures closeness of f and ρ :

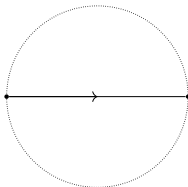
Proposition: For all f ,

$$\mathcal{G}_p(f) = \inf_{\nu \in \mathcal{P}(f(X))} \mathbb{W}_p^p(\rho, \nu).$$

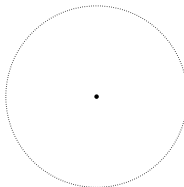
- ▶ Analogous to how $d(p, A) = \inf_{a \in A} d(p, a)$.

Flavor of Problem: \mathcal{J}_p

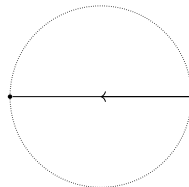
1. \mathcal{J}_p nonconvex



(a) f_0



(b) $f_{.5}$



(c) f_1

2. \mathcal{J}_p nonconcave as well
3. But, jointly weakly continuous in f and ρ

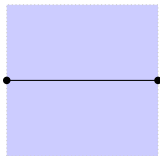
Flavor of Problem: $\mathcal{C}(f)$

Given ε extra budget, can we improve J ?

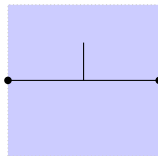
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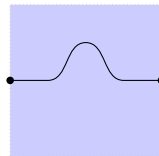
- ▶ 2nd-order term makes local modifications very hard.



(a) f



(b) Desired perturbation

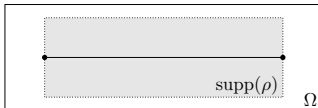


(c) Best we can do

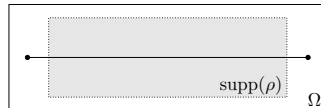
Flavor of Problem: $\mathcal{C}(f)$

Given ε extra budget, can we improve J ?

- ▶ 2nd-order term makes local modifications very hard.
- ▶ Can't just “extend” f past $\partial f(X)$...



(a) $f(X)$



(b) Extended $f(X)$

Summary of Main Results

1. Optimizers?

- Exist under mild hypotheses
- Generally nonunique (e.g. if X, Ω have symmetries)

2. Relationship between J and ℓ ?

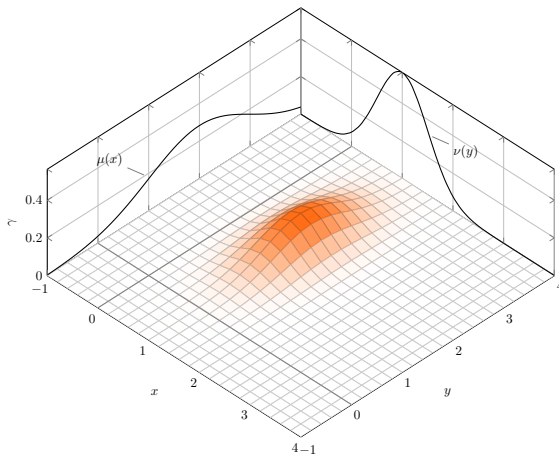
- J continuous in ℓ
- J (trivially) nonincreasing...more than this, hard to say.
- Coarse asymptotic estimates from covering numbers
- Important tool: Can find directional derivative of \mathcal{G}_p in $C(f(X); \mathbb{R}^n)$.

3. Discretization?

- Yes; nice consistency results

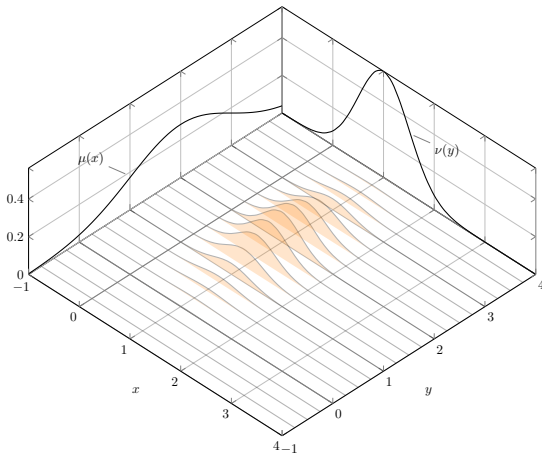
Defining a “Gradient:” Disintegration of Measures

- Given $\gamma \in \mathcal{P}(X \times Y)$ with marginals μ, ν :



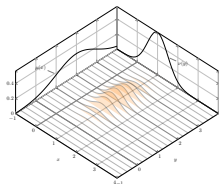
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- “Disintegrate” γ via projection onto x -axis:



Defining a “Gradient:” Disintegration of Measures

- ▶ “Disintegrate” γ via projection onto x -axis:



- ▶ Write μ as $(\pi_X)_\# \gamma$ (“projection” of γ onto x axis)
- ▶ For each x , let γ_x be the “slice” of γ living on $\pi_X^{-1}(x)$. Then

$$\int_{X \times Y} f(x, y) \, d\gamma(x, y) = \int_X \left(\int_{\pi_X^{-1}(x)} f(x, y) \, d\gamma_x(y) \right) d(\pi_X)_\# \gamma(x)$$

Defining a “Gradient”

Define a Gradient: Let $\pi_f : \Omega \rightarrow f(X)$ be closest point projection. Disintegrate ρ under π_f into $(\{\rho_y\}_{y \in f(X)}, \nu_\rho)$ and let

$$F_p(y) = p \int_{\pi_f^{-1}(\{y\})} (\omega - y) |\omega - y|^{p-2} d\rho_y(\omega).$$

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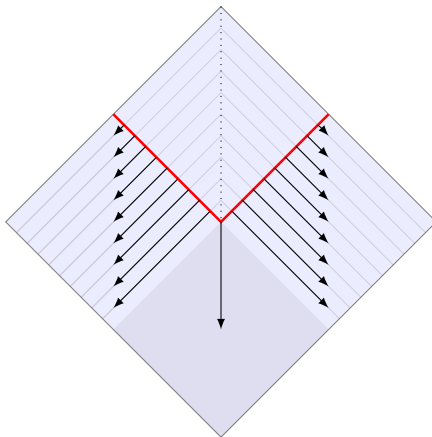
- ▶ Trivially also a function of x by taking $F_p(x) = F_p(f(y))$.
- ▶ $\langle F_p, \xi \rangle_{L^2}$ gives first variation in direction ξ .
- ▶ F_p can be discontinuous even when
 - $f \in C^\infty(X; \Omega)$
 - $f(X)$ is a C^1 manifold

Example of F_p discontinuous with $f \in C^\infty(X; \Omega)$

Let f smoothly parametrize the graph of $|x|$ with Ω a diamond.

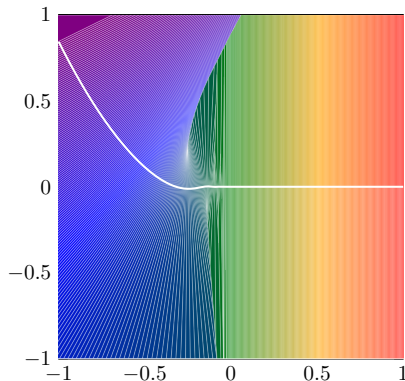
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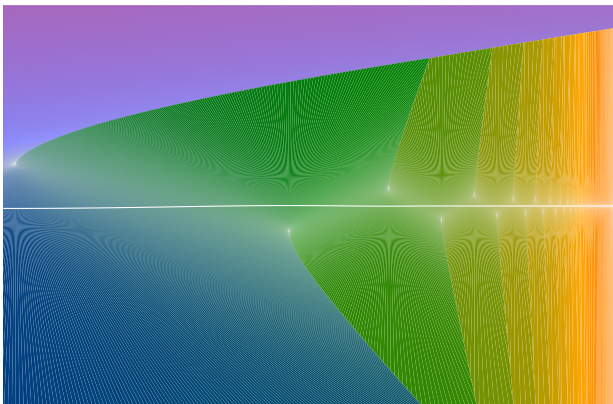
Example of F_p discontinuous with $f(X)$ a C^1 -manifold

Let $f(x) = (x, x^3 \sin(1/x))$ with Ω a square.



Example of F_p discontinuous with $f(X)$ a C^1 -manifold

Let $f(x) = (x, x^3 \sin(1/x))$ with Ω a square(not to scale!).



Local Modification Theorems

Theorem (A): Suppose $f \in C(X; \Omega)$ is ν_ρ -a.e. injective and $\xi : f(X) \rightarrow \mathbb{R}^n$ is cont. except perhaps on a ν_ρ -null closed set. Then

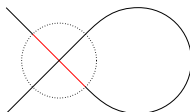
$$\lim_{\varepsilon \rightarrow 0} \frac{\mathcal{I}_p(f + \varepsilon \xi) - \mathcal{I}_p(f)}{\varepsilon} = - \int_X \langle F_p(x), \xi(x) \rangle d\nu_\rho(f(x))$$

Theorem (B): Suppose $f \in C(X; \Omega)$ and $\xi \in C(f(X); \Omega)$. Then

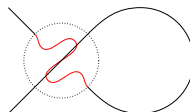
$$\lim_{\varepsilon \rightarrow 0} \frac{\mathcal{I}_p(f + \varepsilon \xi) - \mathcal{I}_p(f)}{\varepsilon} = - \int_Y \langle F_p(y), \xi(y) \rangle d\nu_\rho(y)$$

Local Modification Theorems

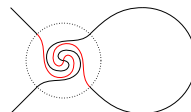
- ▶ Version A: can change topology, injectivity hypotheses on f
- ▶ Version B: can't change topology, no injectivity hypotheses on f



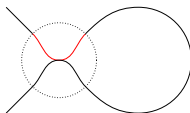
(a) Initial f



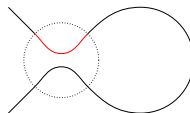
(b) An f_ϵ for Version A



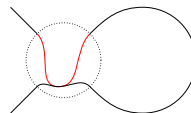
(c) An f_ϵ for Version B



(d) Another initial f



(e) An f_ϵ for Version A



(f) An f_ϵ for Version B

Local Modifications, Continued

- ▶ Challenge to gradient flows: Since F_p not even guaranteed to be continuous, $f + \varepsilon F_p$ might not be $W^{k,q}(X; \Omega)$
- ▶ However...

Theorem: Suppose $\nu_p(\{F_p \neq 0\}) > 0$. Then there exists $\xi \in C^\infty(f(X); \Omega)$ and $\varepsilon > 0$ such that $f + \varepsilon \xi \in W^{k,q}(X; \Omega)$ and

$$\lim_{\varepsilon \rightarrow 0} \frac{\mathcal{G}_p(f + \varepsilon \xi) - \mathcal{G}_p(f)}{\varepsilon} > 0.$$

Special Case: Strict Monotonicity!

Proposition: With very mild assumptions, when $m = 1$, $n = 2$, and $k = 1$, then if f is an optimizer for budget ℓ , then

$$\nu_\rho(\{F_p \neq 0\}) > 0.$$

Sketch (Contradiction):

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Sketch (Contradiction):

- ▶ Consider perturbation $-f$.
- ▶ $\mathcal{C}((1 - \varepsilon)f) = (1 - \varepsilon)\mathcal{C}(f)$. So we recover $O(\varepsilon)$ budget.

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Sketch (Contradiction):

- ▶ Consider perturbation $-f$.
- ▶ $\mathcal{C}((1 - \varepsilon)f) = (1 - \varepsilon)\mathcal{C}(f)$. So we recover $O(\varepsilon)$ budget.
- ▶ Perturbation is continuous on $f(X)$, so local modification theorem gives $\mathcal{J}_p(f - \varepsilon f) - \mathcal{J}_p(f) = O(\varepsilon^2)$
- ▶ Adding a “spike” in the right place changes objective by $O(\varepsilon^{3/2})$

Consistency Results

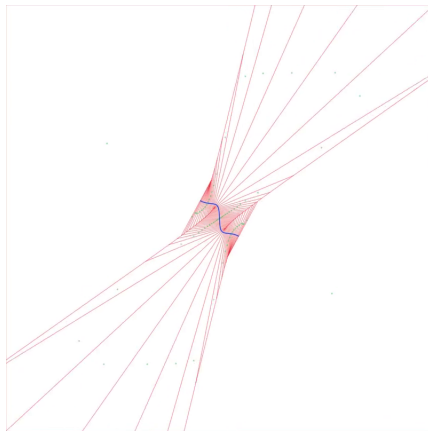
► Discretizing ρ :

Proposition: Let $\hat{\rho}_N$ be an empirical measure for ρ , and let f_N be associated ℓ -optimizers. Then almost surely, every limit point of $\{f_N\}$ is an ℓ -optimizer for ρ .

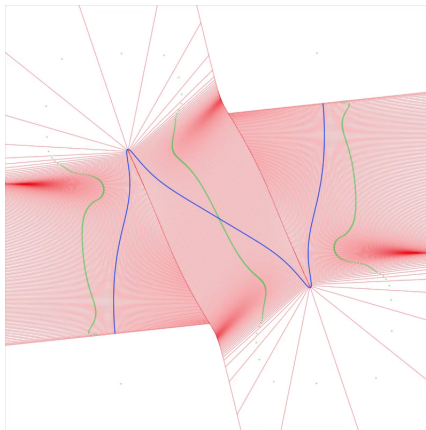
► Discretizing f :

- “Discrete” version of F_p given by Voronoi cells
- Somewhat intricate simulation algorithms (not included in paper)

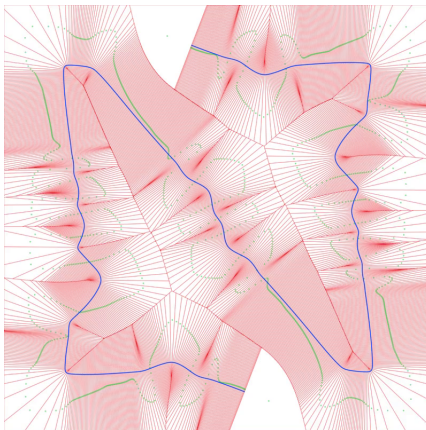
Some Simulations



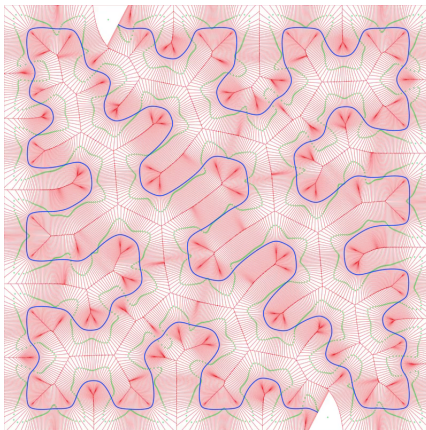
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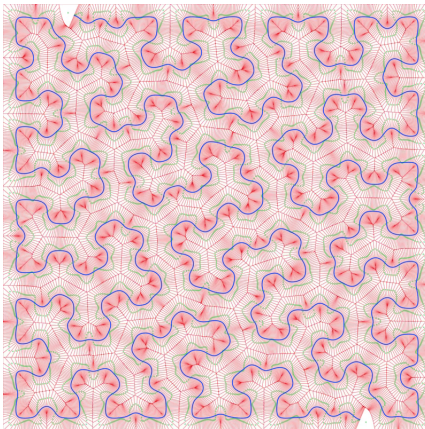
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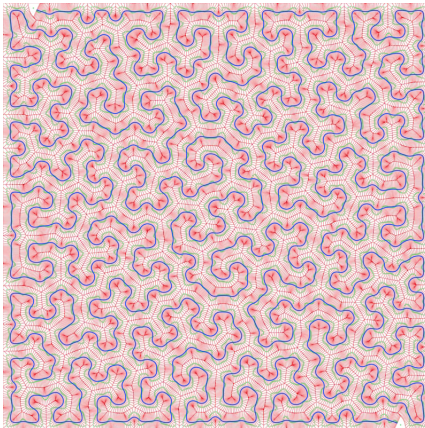
Some Simulations



Some Simulations



Some Simulations



Applications

- ▶ Routing problems
- ▶ Catalyst design
- ▶ Nonlinear Dimensional Reduction
- ▶ Generative Learning

Thank you!

References I



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