

Convergence rate of entropy-regularized multi-marginal optimal transport costs

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Paul Pegon

Joint work with L. Nenna

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CEREMADE, Université Paris-Dauphine & MOKAPLAN, Inria Paris

1. Entropic Multi-Marginal Optimal Transport
2. The upper bound
3. The lower bound

Entropic Multi-Marginal Optimal Transport

Definition of the problem

Consider

- $m \geq 2$ probability measures μ_i compactly supported on \mathcal{C}^2 submanifolds $X_i \subseteq \mathbb{R}^N$ of dim d_i ;
- a cost function $c : X \rightarrow \mathbb{R}_+$ (e.g. continuous or lsc) where $X := \times_i^m X_i$;

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It reads as:

$$\text{MOT}_\varepsilon := \inf_{\gamma \in \Pi(\mu_1, \dots, \mu_m)} \left\{ \int_X c(x_1, \dots, x_m) \, d\gamma(x_1, \dots, x_m) + \varepsilon \text{Ent}(\gamma \mid \otimes_{i=1}^m \mu_i) \right\},$$

where $\Pi(\mu_1, \dots, \mu_m)$ is the set of **couplings** $\gamma \in \mathcal{P}(X)$ having μ_i as marginals, $\text{Ent}(\gamma \mid P)$ is the Boltzmann-Shannon entropy ($= \int \rho \log \rho \, dP$ if $\gamma = \rho P$), and $\varepsilon > 0$ is a small noise parameter.

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- $\varepsilon = 0$ and $m = 2$. **Classical Optimal Transport problem.** Convex problem, but **may have several solutions γ , with or without finite entropy!**
- $\varepsilon > 0$. Strictly convex cost \implies **unique solution γ_ε with finite entropy.**

What are we interested in?

Asymptotics for $\varepsilon \rightarrow 0$ of

- the cost MOT_ε
- the optimal entropic plan γ_ε and optimal Schrödinger potentials $(\phi_\varepsilon, \psi_\varepsilon)$

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- Upper bound for the multi-marginal (Eckstein and Nutz 2022) **with a condition on the optimal transport plans** in terms of quantization dimension ;

Goal

- Asymptotics as $\varepsilon \rightarrow 0$ (lower and upper bounds) for multi-marginal optimal transport cost;
- Possibly degenerate ground costs c (for 2-marginal, $D_{xy}^2 c$ not necessarily invertible);
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In particular we obtain:

Theorem (Nenna - P.)

Let μ_i be compactly supported measures over X_i with L^∞ densities. Assume that $c \in \mathcal{C}^2(\mathbf{X})$ and satisfies a signature condition on second mixed derivatives. Then

$$\text{MOT}_\varepsilon = \text{MOT}_0 + \frac{1}{2} \left(\sum_{i=1}^m d_i - \max_i d_i \right) \varepsilon \log(1/\varepsilon) + O(\varepsilon).$$

Application for *Wasserstein barycenters* (more later).

The upper bound

Assumptions

- μ_i are compactly supported measures in $L^\infty(X_i)$ where X_i are \mathcal{C}^2 submanifolds of dimension d_i ;
- $c \in \mathcal{C}_{loc}^{1,1}(\mathbf{X})$ or more generally locally semi-concave (also, weaker upper bound $c \in \mathcal{C}^{0,1}(\mathbf{X})$);

Goal: get an upper bound of the form

$$\text{MOT}_\varepsilon - \text{MOT}_0 \leq \frac{1}{2} \left(\sum_{1 \leq i \leq m} d_i - \max_i d_i \right) \varepsilon \log(1/\varepsilon) + O(\varepsilon).$$

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Strategy. Straightforward (almost) generalization of the upper bound in (Carlier, Pegon, and Tamanini 2022) on \mathcal{C}^2 submanifolds:

- Build a suitable competitor for the entropic (primal) problem

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using an optimizer for (MOT_0) and a block-approximation of (Carlier, Duval, et al. 2017).

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- Show and use some integral variant of Alexandrov theorem on convex functions.

Upper bound: some details for $m = 2$, marginals μ^-, μ^+

For blocks $\bigsqcup_n A_n = \mathbb{R}^N$ of diameter $\leq \delta$, take as competitor

$$\gamma^\delta := \sum_{i,j \in \mathbb{N}} \gamma_0(A_i \times A_j) \frac{\mu^- \llcorner A_i}{\mu^-(A_i)} \otimes \frac{\mu^+ \llcorner A_j}{\mu^+(A_j)}.$$

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- Plug this competitor into the primal problem, write $E = c - \phi \oplus \psi$ the duality gap, then:

$$\text{MOT}_\varepsilon \leq \int_{\mathbb{R}^d \times \mathbb{R}^d} c \, d\gamma^\delta + \varepsilon \text{Ent}(\gamma^\delta | \mu^- \otimes \mu^+) = \text{MOT}_0 + \int_{\mathbb{R}^d \times \mathbb{R}^d} E \, d(\gamma^\delta - \gamma^0) + \varepsilon \text{Ent}(\gamma^\delta | \mu^- \otimes \mu^+)$$

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$$\begin{aligned} \text{Ent}(\gamma^\delta | \mu^- \otimes \mu^+) &= \sum_{i,j \in \mathbb{N}} \gamma_0(A_i \times A_j) \log \left(\frac{\gamma_0(A_i \times A_j)}{\mu^-(A_i) \mu^+(A_j)} \right) \\ &\leq \sum_{j \in \mathbb{N}} \mu^+(A_j) \log(1/\mu^+(A_j)) = d^+ \log(1/\delta) + O(1). \end{aligned}$$

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- Show that $\int_{\mathbb{R}^d \times \mathbb{R}^d} E \, d(\gamma^\delta - \gamma^0) = O(\delta^2)$ then take $\varepsilon = \delta^2$ (integral Alexandrov-type estimate):

$$\text{MOT}_\varepsilon \leq \text{MOT}_0 + O(\delta^2) + d^+ \varepsilon \log(1/\delta) + O(\varepsilon) = \text{MOT}_0 + \frac{d^+}{2} \varepsilon \log(1/\varepsilon) + O(\varepsilon).$$

The lower bound

A signature condition on the second mixed derivatives

- Consider a $c \in \mathcal{C}^2(\mathbf{X})$ and let P be the set of partitions of $\{1, \dots, m\}$ into two non empty disjoint subsets: $p := \{p_-, p_+\} \in P$;

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- For each $p \in P$ we denote by g_p the bilinear form on TX as

$$g_p = \sum_{i \in p_-, j \in p_+} D_{x_i, x_j}^2 c + D_{x_j, x_i}^2 c \quad \text{where} \quad D_{x_i, x_j}^2 c = \sum_{\alpha_i, \alpha_j} \frac{\partial^2 c}{\partial x_i^{\alpha_i} \partial x_j^{\alpha_j}} dx_i^{\alpha_i} \otimes dx_j^{\alpha_j}.$$

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Theorem (Upper bound on the dimension of the support of the optimal plan (Pass 2011))

Let γ_0 a solution to MOT_0 and suppose that at some point $x \in X$, the signature of some $g \in G$ is $(d^+(g), d^-(g), d^0(g))$. Then, there exists a neighbourhood N_x of x such that $N_x \cap \text{spt}(\gamma_0)$ is contained in a Lipschitz submanifold *with dimension no greater than $\sum_i d_i - d^+(g)$* .

- μ_i be compactly supported measures over X_i with L^∞ densities;
- $c \in \mathcal{C}^2(X)$;
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Strategy

- Use the dual regularized problem (in log form):
- Take Kantorovich potentials (solution to un-regularized dual) as competitors and show that the duality gap $E \doteq c - \bigoplus_{i=1}^m \phi_i$ grows enough near $\Sigma = \{E = 0\}$.

Lower bound: some details

Let $\{1, \dots, m\} = p_- \sqcup p_+$, we identify $x \in X$ with (x_-, x_+) and write $\phi_{\pm}(y) = \sum_{i \in p_{\pm}} \phi_i(y_i)$.

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- If (ϕ_i) are c-conjugate, for $\mathbf{x}, \mathbf{x}' \in X$, we have:

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- By Taylor's integral formula

$$E(\mathbf{x}') + E(\mathbf{x}) \geq \int_0^1 \int_0^1 D_{p_- p_+}^2 c(\mathbf{x}_{s,t})(x'_- - x_-, x'_+ - x_+) = \frac{1}{2} g_p(\bar{\mathbf{x}})(\mathbf{x}' - \mathbf{x}) + O_{\bar{\mathbf{x}}}(\|\mathbf{x}' - \mathbf{x}\|^2)$$

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and taking a convex combination $g = \sum t_p g_p$, for diagonalizing coordinates (u^+, u^-, u^0)

$$E(\mathbf{x}') + E(\mathbf{x}) \geq |u^+(\mathbf{x}') - u^+(\mathbf{x})|^2 - |u^-(\mathbf{x}') - u^-(\mathbf{x})|^2 + O(\|\mathbf{x}' - \mathbf{x}\|^2)$$

\implies quadratic detachment of the duality gap E in $d^+(g) \geq d^*$ dimensions.

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$$\begin{aligned} E(\mathbf{x}') &= c(x'_-, x'_+) - \phi_-(x'_-) - \phi_+(x'_+) \\ &\geq c(x'_-, x'_+) - (c(x'_-, x_+) - \phi_+(x_+)) - (c(x_-, x'_+) - \phi_-(x_-)) \\ &= c(x'_-, x'_+) - c(x'_-, x_+) - c(x_-, x'_+) + c(x_-, x_+) - E(\mathbf{x}). \end{aligned}$$

- By Taylor's integral formula

$$E(\mathbf{x}') + E(\mathbf{x}) \geq \int_0^1 \int_0^1 D_{p_- p_+}^2 c(\mathbf{x}_{s,t})(x'_- - x_-, x'_+ - x_+) = \frac{1}{2} g_p(\bar{\mathbf{x}})(\mathbf{x}' - \mathbf{x}) + O_{\bar{\mathbf{x}}}(\|\mathbf{x}' - \mathbf{x}\|^2)$$

and taking a convex combination $g = \sum t_p g_p$, for diagonalizing coordinates (u^+, u^-, u^0)

$$E(\mathbf{x}') + E(\mathbf{x}) \geq |u^+(\mathbf{x}') - u^+(\mathbf{x})|^2 - |u^-(\mathbf{x}') - u^-(\mathbf{x})|^2 + O(\|\mathbf{x}' - \mathbf{x}\|^2)$$

\implies quadratic detachment of the duality gap E in $d^+(g) \geq d^*$ dimensions.

- Taking $(\phi_i)_{1 \leq i \leq m}$ as competitor in the dual of the entropic MOT:

$$\text{MOT}_{\varepsilon} \geq \text{MOT}_0 - \varepsilon \log \left(\int_{\Pi_{1 \leq i \leq m} X_i} e^{-\frac{\varepsilon}{2} d} \otimes_{1 \leq i \leq m} \mu_i \right) \geq \text{MOT}_0 + \frac{d^*}{2} \varepsilon \log(1/\varepsilon) - O(\varepsilon).$$

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- Consider $d_i = d$ for all i and the cost $c = h(\sum_{i=1}^m x_i)$ with $D^2 h < 0$ then $d^* = (m - 1)d$ and

$$\text{MOT}_\varepsilon = \text{MOT}_0 + \frac{(m - 1)d}{2} \varepsilon \log(1/\varepsilon) + O(\varepsilon).$$

This is the case of Gangbo-Świąch cost, that is $\sum_{i < j} |x_i - x_j|^2$ which corresponds to the multi-marginal formulation of the [Wasserstein barycenter problem](#).

Thank You.