

Approximation of OT problems with marginal moments constraints

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Sommaire

- 1 Introduction
- 2 Characteriation of a minimizer of the MCOT Problem
- 3 Convergence of the MCOT problem towards the OT problem
- 4 Numerical simulations

The Moment Constraint Optimal Transport Problem

In the following presentation \mathcal{X} and \mathcal{Y} are compact subsets of \mathbb{R}^d , $d \in \mathbb{N}^*$.

- The Optimal Transport Problem

$$I^* = \inf_{\pi \in \Pi(\mu, \nu)} \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) d\pi(x, y), \quad (1)$$

where $\Pi(\mu, \nu) = \{\pi \in \mathcal{P}(\mathcal{X} \times \mathcal{Y}) \text{ s.t. } \int_{\mathcal{X}} d\pi = d\nu, \int_{\mathcal{Y}} d\pi = d\mu\}$.

- The Moment Constraint Optimal Transport (MCOT) Problem

$$I^N = \min_{\substack{\pi \in \mathcal{P}(\mathcal{X} \times \mathcal{Y}) \\ \forall 1 \leq m \leq N, \int_{\mathcal{X}} \phi_m(x) d\pi(x, y) = \int_{\mathcal{X}} \phi_m(x) d\mu(x) \\ \forall 1 \leq n \leq N, \int_{\mathcal{Y}} \psi_n(y) d\pi(x, y) = \int_{\mathcal{Y}} \psi_n(y) d\nu(y)}} \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) d\pi(x, y), \quad (2)$$

where for all $1 \leq m, n \leq N$, ϕ_m, ψ_n are given continuous integrable functions.

Main Results

- One can **characterize a minimizer of the MCOT Problem** (which is numerically computable).
- With well-chosen sets of **test functions** $(\phi_m)_{m \in \mathbb{N}}$ and $(\psi_n)_{n \in \mathbb{N}}$ the **MCOT** problem converges towards the **OT** problem.
- The **convergence speed** depends on the choice of **test functions**.

Sommaire

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- 3 Convergence of the MCOT problem towards the OT problem
- 4 Numerical simulations

The Tchakaloff Theorem

Proposition (Tchakaloff [Bayer & Teichmann, 2006], [Berschneider & Sasvári, 2012])

Let π be a positive measure on the space \mathbb{R}^d , with the Borel σ -algebra \mathcal{F} , concentrated in $A \in \mathcal{F}$, i.e. $\pi(\mathbb{R}^d \setminus A) = 0$, and $\Xi : \mathbb{R}^d \rightarrow \mathbb{R}^{N_0}$ a Borel measurable map.

Assume that the first moments of $\Xi \# \pi$ exist, i.e.

$$\int_{\mathbb{R}^N} \|u\| d\Xi \# \pi(u) < \infty.$$

Then, there exist an integer $1 \leq K \leq N_0$, points $z_1, \dots, z_K \in A$ and weights $w_1, \dots, w_K > 0$ such that

$$\int_{\Omega} \Xi_i(z) d\pi(z) = \sum_{k=1}^K w_k \Xi_i(z_k)$$

for all $1 \leq i \leq N$, where Ξ_i denotes the i -th component of Ξ .

Characterization of a minimizer of the MCOT Problem

Proposition

For any l.s.c. cost function $c : \mathcal{X} \times \mathcal{Y} \rightarrow \mathbb{R}^+ \cup \{\infty\}$, we consider problems of the form

$$I^N = \min_{\pi \in \mathcal{P}(\mathcal{X} \times \mathcal{Y})} \left\{ \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) d\pi(x, y) \right\}, \quad (3)$$

$$\forall 1 \leq m \leq N, \int_{\mathcal{X} \times \mathcal{Y}} \phi_m d\pi = \int_{\mathcal{X}} \phi_m(x) d\mu(x)$$

$$\forall 1 \leq n \leq N, \int_{\mathcal{X} \times \mathcal{Y}} \psi_n d\pi = \int_{\mathcal{Y}} \psi_n(y) d\nu(y)$$

With appropriate additional conditions on the test functions $(\phi_m)_{1 \leq m \leq N}$ and $(\psi_n)_{1 \leq n \leq N}$, I^N is finite and is a minimum.

Moreover, there exists a **finite discrete probability measure**

$\gamma = \sum_{k=1}^K w_k \delta_{x_k, y_k}$ (where for all k , $x_k \in \mathcal{X}$, $y_k \in \mathcal{Y}$ and $w_k \in \mathbb{R}_+^*$ and $\sum_{k=1}^K w_k = 1$, and all points (x_k, y_k) are different) with $0 < K \leq 2N + 2$, which is a minimizer.

Characterization of a minimizer of the MCOT Problem, remarks

Remark

One can formulate such a Moment Constraint Optimal Transport Problem even in the case where \mathcal{X} and \mathcal{Y} are non compact sets, with some additional technicalities. Thus it **can be applied to DFT**.

Remark

The numerical interest of the characterization of such a minimizer is that, for N given test functions on each set, it is computable by a particle algorithm needing only $2N + 2$ points and weights, and in a multimarginal case, with D marginal laws, $DN + 2$ points and weights.

Sommaire

- 1 Introduction
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- 3 Convergence of the MCOT problem towards the OT problem
- 4 Numerical simulations

Density Condition on test functions

The density condition needed to establish the convergence on the compact sets \mathcal{X} and \mathcal{Y} is for the **continuous bounded functions** and for the L^∞ norm:

$$\forall f \in C_c^0(\mathcal{X}), \forall \epsilon > 0, \exists M \in \mathbb{N}, \lambda_1, \dots, \lambda_M \in \mathbb{R} \mid \sup_{x \in \mathcal{X}} \left| f(x) - \sum_{i=1}^M \lambda_i \phi_i(x) \right| \leq \epsilon \quad (4)$$

and

$$\forall f \in C_c^0(\mathcal{Y}), \forall \epsilon > 0, \exists M \in \mathbb{N}, \lambda_1, \dots, \lambda_M \in \mathbb{R} \mid \sup_{y \in \mathcal{Y}} \left| f(y) - \sum_{i=1}^M \lambda_i \psi_i(y) \right| \leq \epsilon \quad (5)$$

Notations

Recall

- The Optimal Transport Problem

$$I^* = \inf_{\pi \in \Pi(\mu, \nu)} \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) d\pi(x, y). \quad (6)$$

- The Moment Constraint Optimal Transport (MCOT) Problem

$$I^N = \min_{\pi \in \mathcal{P}(\mathcal{X} \times \mathcal{Y})} \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) d\pi(x, y). \quad (7)$$

$$\forall 1 \leq m \leq N, \int_{\mathcal{X}} \phi_m(x) d\pi(x, y) = \int_{\mathcal{X}} \phi_m(x) d\mu(x)$$

$$\forall 1 \leq n \leq N, \int_{\mathcal{Y}} \psi_n(y) d\pi(x, y) = \int_{\mathcal{Y}} \psi_n(y) d\nu(y)$$

Convergence of the MCOT Pb towards the OT Pb

Proposition

Let us consider sequences of continuous test functions $(\phi_m)_{m \in \mathbb{N}}$ and $(\psi_n)_{n \in \mathbb{N}}$ defined on \mathcal{X} (resp. \mathcal{Y}) and valued on \mathbb{R} , and verifying the *density conditions (4) and (5)*.

Then, using the previous notations where c is a l.s.c. cost function valued on $\mathbb{R}_+ \cup \{+\infty\}$, one has that

$$I^N \xrightarrow{N \rightarrow \infty} I^*$$

and that from every sequence $(\pi^N)_{N \in \mathbb{N}}$ such that for all N , π^N is a minimizer of the MCOT Problem with N moments, one can extract a subsequence $(\pi^{\varphi(N)})_{N \in \mathbb{N}}$ which **converges towards**

$$\pi^* \in \arg \min_{\pi \in \Pi(\mu, \nu)} \left\{ \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) d\pi(x, y) \right\}.$$

Convergence of the MCOT Pb towards the OT Pb, remark

Remark

This result can be extended to non-compact sets \mathcal{X} and \mathcal{Y} with some more technical conditions on the test functions, for a DFT application.

Convergence speed for piecewise constant test functions

Let us define the intervals


$$\forall 1 \leq i \leq N-1, T_i^N = \left[\frac{i-1}{N}, \frac{i}{N} \right) \text{ and } T_N^N = \left[\frac{N-1}{N}, 1 \right]. \quad (8)$$

On compact sets in dimension 1, analogous MCOT problems¹ with piecewise constant test functions $\phi_i^N = \mathbf{1}_{T_i^N}$ converge towards the OT Problem at a $1/N$ speed.

Proposition

Let $\mu, \nu \in \mathcal{P}([0, 1])$ and $c : [0, 1]^2 \rightarrow \mathbb{R}_+$ a function with Lipschitz constant $K > 0$. Then, for all $N \in \mathbb{N}^*$,

$$I^N \leq I^* \leq I^N + \frac{K}{N}. \quad (9)$$

¹Here the MCOT Problem is a infimum and not a minimum. 

Convergence speed for piecewise affine test functions

On compact sets in dimension 1, **MCOT** problems with **continuous piecewise affine test functions**

$$\phi_i(x) = \begin{cases} N(x - \frac{i-1}{N}) & \text{if } x \in T_{i-1}^N \\ 1 - N(x - \frac{i}{N}) & \text{if } x \in T_i^N \\ 0 & \text{elsewhere.} \end{cases}$$

converge towards the **Wasserstein-1 distance** at a $1/N^2$ speed.

Proposition

Consider two marginal laws $\mu \in \mathcal{P}([0, 1])$ and $\nu \in \mathcal{P}([0, 1])$ with density ρ_μ and ρ_ν and cumulative distribution functions F_μ and F_ν respectively. Then

$$I^N \leq W_1(\mu, \nu) \leq I^N + 2 \sup_{[0,1]} |\rho_\mu - \rho_\nu| \frac{M}{N^2}, \quad (10)$$

where M is the number of intervals T_i^N ($1 \leq i \leq N$) on which $(F_\mu - F_\nu)$ changes of sign.

Sommaire

- 1 Introduction
- 2 Characteriation of a minimizer of the MCOT Problem
- 3 Convergence of the MCOT problem towards the OT problem
- 4 Numerical simulations

Convergence speed for piecewise affine test functions in dimension 1

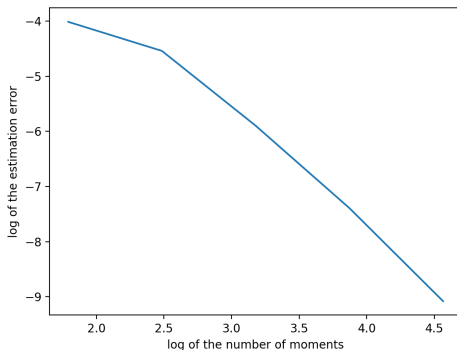
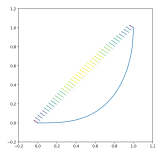
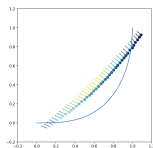


Figure: Convergence speed for piecewise affine test functions in dimension 1 in log-log scale.

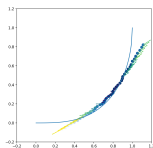
dimension 1



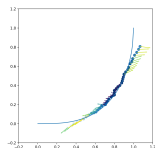
(a) iteration 0



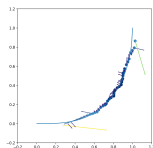
(b) iteration 20



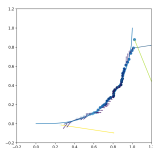
(c) iteration 140



(d) iteration 360



(e) iteration 3000



(f) iteration 5000

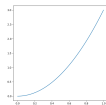
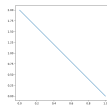
(a) μ (b) ν

Figure: Marginal laws

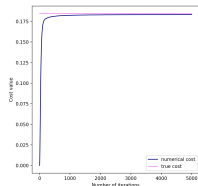
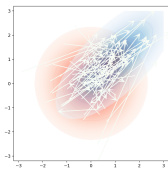


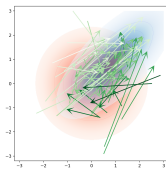
Figure: Cost

Figure: Convergence for two 1D marginal laws with 20 test functions on each set

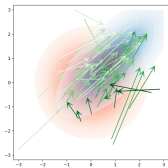
dimension 2



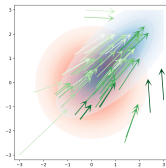
(a) iteration 0



(b) iteration 400



(c) iteration 1400



(d) iteration 9000

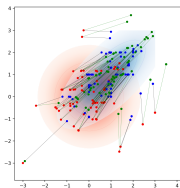


Figure: Transport map

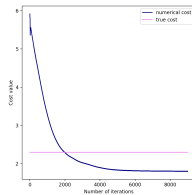


Figure: Cost

Figure: Convergence for two 2D marginal laws with 36 test functions on each set

Further work

- Explore other possibilities of particle algorithms.
- Study of a symmetric Tchakaloff theorem in order to treat the symmetrical multimarginal case more efficiently.
- Develop an efficient (perhaps multilevel) particle algorithm for dimension 3 in the multimarginal case for a Coulomb cost and in the martingale case.
- Proof of more general rates of convergences.

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Thank you for your attention.

Idea of proof of the Tchakaloof Theorem

- For a given measure $\pi \in \mathcal{P}(\mathbb{R}^d)$, $\int_{\mathbb{R}^d} z d\pi(z)$ lies in $\text{cone}(A)$, where A is such that $\pi(\mathbb{R}^d \setminus A) = 0$.
- The Caratheodory Theorem states that for a set B in dimension N , a point in $\text{cone}(B)$ a positive combination of at most N points of B . Thus, $\exists z_1, \dots, z_N \in \mathbb{R}^d$, $w_1, \dots, w_N > 0$,

$$\int_{\mathbb{R}^d} z d\pi(z) = \sum_{i=1}^N w_i z_i. \quad (11)$$

- One can apply the previous result to the measure $\Xi \# \pi$ wich yields to: $\exists z_1, \dots, z_N \in \mathbb{R}^d$, $w_1, \dots, w_N > 0$,

$$\int_{\mathbb{R}^d} z d\Xi \# \pi(z) = \sum_{i=1}^N w_i \Xi(z_i). \quad (12)$$