

# Multimarginal optimal transportation: the one dimensional symmetric case

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# Introduction

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- Description of the multi marginal problem and physical motivation;
- Classical results with 2 marginals in dimension 1;
- Symmetric multimarginal case in dimension 1: existence of an optimal map and uniqueness of the symmetric optimal plan.

# General minimization problem

We are interested in the following minimum problem:

$$\min_{f \in L^2((\mathbb{R}^d)^n; \mathbb{C}), \|f\|_2=1} \int_{(\mathbb{R}^d)^n} (c|\nabla f|^2 + V_{ee}|f|^2 + V_{ext}|f|^2) dx,$$

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where

- $V_{ee}$  is the Coulombian interaction potential between the electrons:

$$V_{ee}(x_1, \dots, x_n) = \sum_{1 \leq i < j \leq n} \frac{1}{|x_i - x_j|};$$

- $V_{ext}$  is an external potential (due to the nucleus), which is the same for every electron:  $V_{ext}(x_1, \dots, x_n) = V(x_1) + \dots + V(x_n)$ ;

# Hohenberg e Kohn formulation(HK)

The wave function  $f$  gives us the density  $\rho$  of the (identical) electrons, that is, the marginals of the measure  $|f|^2 dx$  are always  $\rho$ .



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In order to solve the previous problem, the idea is to fix the density  $\rho$  and minimize only the (kinetic + interaction) part under the density constraint. Then one tries to solve the main problem as

$$\min_{\rho \in \mathcal{P}(\mathbb{R}^d)} \left\{ F(\rho) + n \int V(x) d\rho \right\},$$

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Another way is to see it as the strictly correlated regime, in which we neglect the kinetic part (Gori Giorgi).

# Mathematical formulation

We are interested in the problem

$$\min_{\pi \in \Gamma(\mu_1, \dots, \mu_n)} \int_{X^n} c(x_1, \dots, x_n) d\pi.$$

In our case  $X = \mathbb{R}^d$ ,  $\mu_1 = \dots = \mu_n = \rho$  and  $c$  is the sum of the repulsive Coulombian potentials.

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- When  $n = 2$  we obtain the Kantorovich formulation of optimal transport;
- this isn't exactly a transport problem but more a coupling problem;
- symmetries: impossibility for uniqueness.

# Multimarginal optimal transport

Notations:

- given  $\sigma \in S_n$  define  $\sigma : X^n \rightarrow X^n$  as

$$\sigma : (x_1, \dots, x_n) \mapsto (x_{\sigma(1)}, \dots, x_{\sigma(n)});$$

- $\Gamma_{sym}(\rho)$  is the set of probabilities  $\pi$  which have all marginals equal to  $\rho$  and such that  $\sigma_{\#}\pi = \pi$  for all  $\sigma \in S_n$ . The natural projection into symmetric plans is

$$\pi^S = \frac{1}{n!} \sum_{\sigma \in S_n} \sigma_{\#}\pi;$$

- $T_{sym}(\rho)$  is the set of Borel maps  $T : X \rightarrow X$  such that  $T_{\#}\rho = \rho$  e  $T^{(n)}(x) = x$  for  $\rho$ -almost every  $x$ .



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From now on we'll call  $(K_{sym})$  e  $(M_{sym})$  the following two problems:

$$\min_{\pi \in \Gamma_{sym}(\rho)} \int_{X^n} c(x_1, \dots, x_n) d\pi$$

$$\inf_{T \in T_{sym}(\rho)} \int_X c(x, T(x), T(T(x)), \dots, T^{(n-1)}(x)) d\rho$$

# Existence of the map

## Conjecture

*There exists an optimal symmetric map, in particular  $(M_{sym})$  is a minimum. Furthermore this minimum is equal to  $(K_{sym})$ .*

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The goal of this studies is to investigate the equality  $(K_{sym}) = (M_{sym})$ , the presence of an optimal map and eventually the characterization of optimal (symmetric) plans.

# Known results

- (Colombo - D.M.) In the symmetric case is true in general that  $(K_{sym}) = (M_{sym})$ , in every complete and separable metric space, with a symmetric l.s.c. cost, continuous in its finiteness domain, when  $\rho$  is without atoms;

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- (Pass, Pass - Kim ) in the non-symmetric case, if a modified twist condition on the cost holds true and  $\rho$  is absolutely continuous with respect to Lebesgue measure, then there exists a unique optimal plan, which is induced by an optimal map (as in the two marginal case);

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## Theorem (Buttazzo - De Pascale - Gori Giorgi)

*There exists a minimizer of  $(K_{sym})$ . Furthermore, the following duality formula holds true*

$$(K_{sym}) = n \sup \left\{ \int_X \phi \, d\rho : \phi(x_1) + \dots + \phi(x_n) \leq c(x_1, \dots, x_n) \right\}$$

Let  $X = \mathbb{R}$ , and let  $c(x_1, \dots, x_n) = \sum_{1 \leq i < j \leq n} \phi(|x_i - x_j|)$ , where  $\phi$  is a convex and decreasing function on  $\mathbb{R}^+$ .



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### Theorem

Let  $\rho$  be a diffuse probability measure on  $\mathbb{R}$  (such that  $(K) < \infty$ ). Let  $-\infty = d_0 < d_1 < \dots < d_N = +\infty$  be such that

$$\rho([d_i, d_{i+1}]) = 1/N \quad \forall i = 0, \dots, N-1. \quad (1)$$

Let  $T : \mathbb{R} \rightarrow \mathbb{R}$  be the unique function (up to  $\rho$ -null sets) that is increasing on every interval  $[d_i, d_{i+1}]$ ,  $i = 0, \dots, N-1$ , and such that

$$T_{\#} 1_{[d_i, d_{i+1}]} \rho = 1_{[d_{i+1}, d_{i+2}]} \rho \quad \forall i = 0, \dots, N-1. \quad (2)$$

Then  $T$  is an admissible map for  $(M_{\text{sym}})$  and

$$(K) = \int_{\mathbb{R}} c(x, T(x), T^2(x), \dots, T^{(N-1)}(x)) d\rho. \quad (3)$$

Moreover there exists a unique symmetric optimal plan, that is the symmetrization of the one induced by  $T$ .

# c-monotonicity

## Definition (*c*-monotonicity)

A set  $A \subset X \times X$  is *c-monotone* if for every two points  $(x_1, y_1), (x_2, y_2) \in A$  we have

$$c(x_1, y_1) + c(x_2, y_2) \leq c(x_1, y_2) + c(x_2, y_1)$$

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## Proof.

Consider neighborhoods of the points where inequality fails and rearrange the plan. □

# Geometric characterization in $\mathbb{R}$ , convex cost

In dimension 1, in the case  $c(x, y) = \phi(x - y)$ , with  $\phi$  strictly convex function we have: if  $A$  is  $c$ -monotone, then for every couple  $(x_1, y_1), (x_2, y_2) \in A$ , if  $x_1 < x_2$  then  $y_1 \leq y_2$ .

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Proof.

Suppose  $y_1 > y_2$ . Then we have that  $x_1 - y_1 < x_2 - y_1 < x_2 - y_2$ , so there exists  $t \in (0, 1)$  such that

$$x_2 - y_1 = t(x_1 - y_1) + (1 - t)(x_2 - y_2)$$

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$$\phi(x_2 - y_1) + \phi(x_1 - y_2) < \phi(x_1 - y_1) + \phi(x_2 - y_2)$$



## Corollary

*The support of an optimal plan is "monotone". If  $\rho_1$  and  $\rho_2$  are diffuse then an optimal plan is induced by a map.*

# multimarginal $c$ -monotonicity

We can adapt the definition to the multi marginal case. Given  $P \subseteq \{1, \dots, n\}$  and two points  $x, y \in X^n$  we define the  $P$ -mixing as

$$(P(x, y))_i = \begin{cases} x_i & \text{if } i \in P \\ y_i & \text{otherwise.} \end{cases} \quad (P^c(x, y))_i = \begin{cases} y_i & \text{if } i \in P \\ x_i & \text{otherwise.} \end{cases}$$

## Definition ( $c$ -monotonicity)

A set  $A \subset X^n$  is  $c$ -monotone if for every  $x, y \in A$  and every  $P \subseteq \{1, \dots, n\}$  one has

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## Theorem (Pass)

Let  $c$  be a nonnegative l.s.c. cost, continuous in its finiteness domain; then the support of an optimal plan with finite cost is necessarily  $c$ -monotone.

The proof uses Gangbo theorem applied to the spaces  $X^{|P|}$  e  $X^{|P^c|}$ .

# symmetric $c$ -monotonicity

If we add symmetry we can define  $c$ -monotonicity in the following equivalent way. We first define the "cumulative coordinate" of two points  $x, y \in X^n$  as

$$C(x, y) = \sum_{i=1}^n \delta_{x_i} + \delta_{y_i}$$

## Definition ( $c$ -monotonicity)

A symmetric set  $A \subset X^n$  is  $c$ -monotone if for every points  $x, y \in A$  and every points  $x', y'$  such that  $C(x, y) = C(x', y')$  we have that

$$c(x) + c(y) \leq c(x') + c(y')$$

From now on we'll have  $X = \mathbb{R}$  and  $c(x) = \sum_{i < j} \phi(|x_i - x_j|)$ . The crucial geometrical characterization is:

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### Lemma

*Given a set  $T$  consisting of  $2n$  coordinates counted with multiplicities, i.e., there exist  $t_1 \leq t_2 \leq \dots \leq t_{2n}$  such that  $T = \sum \delta_{t_i}$ , let  $x, y$  be points in  $X^n$  that minimize the problem*

$$\min\{c(x') + c(y') : C(x', y') = T\}.$$

*Then, up to swap  $x$  and  $y$  and up to re-arrange the coordinates increasingly, we have that  $x_i = t_{2i}$  and  $y_i = t_{2i-1}$ .*

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*Then, up to swap  $x$  and  $y$  and up to re-arrange the coordinates increasingly, we have that  $x_i = t_{2i}$  and  $y_i = t_{2i-1}$ .*

Equivalently,  $x, y$  are minimizers iff, defining  $x^*, y^*$  the increasing rearrangement of the coordinates, one has

$$x_1^* \leq y_1^* \leq x_2^* \leq \dots \leq x_n^* \leq y_n^*;$$

in this case the points are *well ordered*.

# How to conclude

Given the lemma, we can conclude the proof of the theorem:

- prove that the support of an optimal plan doesn't intersect  $x_i = x_j$ ;
- reduce the analysis to the zone  $x_1 < x_2 \cdots < x_n = O$ ;
- consider the numbers

$$d_i^+ = \max\{x_i : x \in O \cap \text{spt}\pi\} \quad d_i^- = \min\{x_i : x \in O \cap \text{spt}\pi\},$$

and prove  $d_i^+ \leq d_{i+1}^-$  and consequently  $\rho([d_i^+, d_{i+1}^-]) = 1/n$ ;

- conclude as in the 2 dimensional case, with monotone maps.

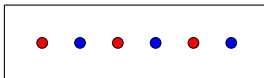
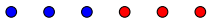


# Case with 3 marginals: by hands

We have 6 points on the real line (that represent  $T$ ). In what way I can group in two groups of three points such that the interaction potential is minimal?  $\binom{6}{3}/2 = 10$  cases.

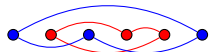
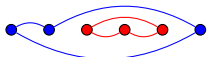
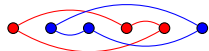
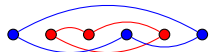
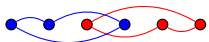
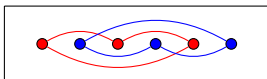
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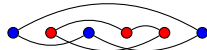
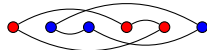
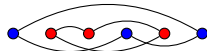
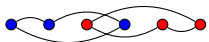
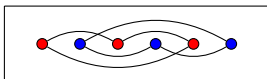
# Case with 3 marginals: by hands

We have 6 points on the real line (that represent  $T$ ). In what way I can group in two groups of three points such that the interaction potential is minimal?  $\binom{6}{3}/2 = 10$  cases.



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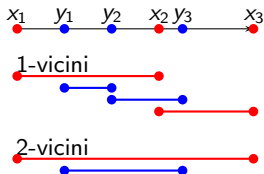


# $n$ -marginals symmetric case

Given  $2n$  points on the real line, what is the best way of separate them in two group of  $n$  points such that the energy interaction is minimal?

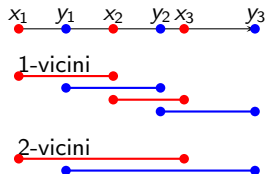
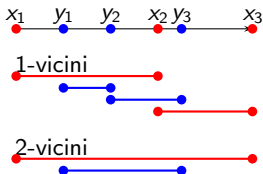
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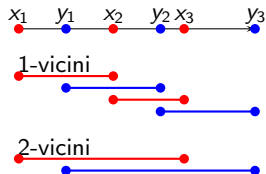
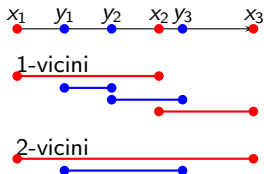
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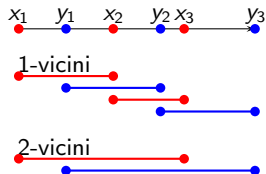
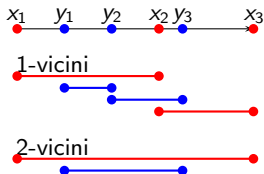


- Step 1 (convexity of  $\phi$ ): use the one-dimensional result for 2 marginals on  $k$ -neighbors (with fixed  $k$ );



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- Step 1 (convexity of  $\phi$ ): use the one-dimensional result for 2 marginals on  $k$ -neighbors (with fixed  $k$ );
- Step 2 (monotonicity of  $\phi$ ): stretch in the right way the segments to get the "well ordered" situation.

# Future development

- one-dimensional symmetric case with a cost that is concave or concave-convex;

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- symmetric case for  $X = \mathbb{R}^2$ , with radial marginal and Coulombian cost;
- sufficiency of multimarginal  $c$ -monotonicity (other possible definitions?).

Thanks for the attention