

# Optimal Transport for Applied Mathematicians

## Reading notes

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### Introduction

Original Monge Problem:

$$\min\{M(T) := \int_X c(x, T(x)) d\mu(x) : T_{\#}\mu = \nu\} \quad (\text{MP})$$

$\implies$  Non linear in  $T$  &  $x$  cannot be sent to different places in  $Y$  through  $T(x)$

Kantorovich formulation:

$$\min\{K(\gamma) := \int_{X \times Y} c d\gamma : (\pi_x)_{\#}\gamma = \mu, (\pi_y)_{\#}\gamma = \nu\} \quad (\text{KP})$$

$\implies$  Linear in  $\gamma$  &  $\gamma(x, y)$  = "number of particles going from  $x$  to  $y$ "

# 1 Primal and dual problems

## 1.1 Kantorovich & Monge problem

**Theorem 1.4** Let  $X$  and  $Y$  be compact metric spaces,  $\mu \in \mathcal{P}(X)$ ,  $\nu \in \mathcal{P}(Y)$  and  $c : X \times Y \rightarrow \mathbb{R}$  a continuous function. Then (KP) admits a solution.

**Theorem 1.5** Let  $X$  and  $Y$  be compact metric spaces,  $\mu \in \mathcal{P}(X)$ ,  $\nu \in \mathcal{P}(Y)$  and  $c : X \times Y \rightarrow \mathbb{R} \cup \{+\infty\}$  a lower semi-continuous and bounded from below function. Then (KP) admits a solution.

**Theorem 1.7** Let  $X$  and  $Y$  be Polish spaces, i.e. complete and separable metric spaces,  $\mu \in \mathcal{P}(X)$ ,  $\nu \in \mathcal{P}(Y)$  and  $c : X \times Y \rightarrow [0, +\infty]$  a lower semi-continuous function. Then (KP) admits a solution.

When cost functions are not continuous:

**Lemma 1.8** Let  $\gamma_n, \gamma \in \Pi(\mu, \nu)$  be probabilities on  $X \times Y$  and  $a : X \rightarrow \tilde{X}$  and  $b : Y \rightarrow \tilde{Y}$  be measurable maps valued in two separable metric spaces  $\tilde{X}$  and  $\tilde{Y}$ . Let  $c : \tilde{X} \times \tilde{Y} \rightarrow \mathbb{R}_+$  be a continuous function with  $c(a, b) \leq f(a) + g(b)$  with  $f, g$  continuous and  $\int (f \circ a) d\mu, \int (g \circ b) d\nu < +\infty$ . Then:

$$\gamma_n \rightarrow \gamma \implies \int_{X \times Y} c(a(x), b(y)) d\gamma_n \rightarrow \int_{X \times Y} c(a(x), b(y)) d\gamma$$

## 1.2 Duality

(KP) can be rewritten as:

$$\min_{\gamma} \int_{X \times Y} c d\gamma + \sup_{\phi, \psi} \int_X \phi d\mu + \int_Y \psi d\nu - \int_{X \times Y} (\phi(x) + \psi(y)) d\gamma$$

with  $\phi, \psi$  continuous bounded functions. inf and sup can actually be interchanged (not classical Rockafellar duality since not finite-dimensional spaces).  $\implies$  dual problem:

$$\sup_{\phi, \psi} \int_X \phi d\mu + \int_Y \psi d\nu + \inf_{\gamma} \int_{X \times Y} (c(x, y) - \phi(x) - \psi(y)) d\gamma$$

Let  $\phi \oplus \psi(x, y) = \phi(x) + \psi(y)$ . Then

$$\inf_{\gamma} \int_{X \times Y} (c - \phi \oplus \psi) d\gamma = \begin{cases} 0 & \text{if } \phi \oplus \psi \leq c \text{ on } X \times Y \\ -\infty & \text{otherwise} \end{cases}$$

Thus we get the following Dual Problem:

$$\max \left\{ \int_X \phi d\mu + \int_Y \psi d\nu : \phi \in C_b(X), \psi \in C_b(Y) : \phi \oplus \psi \leq c \right\} \quad (\text{DP})$$

Obviously:

$$\sup (\text{DP}) \leq \inf (\text{KP})$$

**Definition**

- *c-transform* For  $\chi : X \rightarrow \bar{\mathbb{R}}$ ,  $\chi^c : Y \rightarrow \bar{\mathbb{R}}$  with  $\chi^c(y) = \inf_{x \in X} c(x, y) - \chi(x)$
- *$\bar{c}$ -transform* For  $\zeta : Y \rightarrow \bar{\mathbb{R}}$ ,  $\zeta^{\bar{c}} : X \rightarrow \bar{\mathbb{R}}$  with  $\zeta^{\bar{c}}(x) = \inf_{y \in Y} c(x, y) - \zeta(y)$
- *$\bar{c}$ -concavity*  $\psi$  defined on  $Y$  is  $\bar{c}$ -concave if there exists  $\chi$  such that  $\psi = \chi^c$
- *c-concavity*  $\psi$  defined on  $Y$  is  $c$ -concave if there exists  $\chi$  such that  $\psi = \chi^{\bar{c}}$

**Proposition 1.11** Suppose that  $X$  and  $Y$  are compact and  $c$  is continuous. Then there exists a solution  $(\phi, \psi)$  to problem (DP) and it has the form  $\phi \in c - \text{conc}(X)$ ,  $\psi \in \bar{c} - \text{conc}(X)$ , and  $\psi = \phi^c$ . In particular

$$\max(\text{DP}) = \max_{\phi \in c - \text{conc}(X)} \int_X \phi d\mu + \int_Y \phi^c d\nu$$

By admitting  $\min(\text{KP}) = \max(\text{DP})$  we have:

$$\min(\text{KP}) = \max_{\phi \in c - \text{conc}(X)} \int_X \phi d\mu + \int_Y \phi^c d\nu$$

which also shows that the minimum value of (KP) is a convex function of  $(\mu, \nu)$ , as it is a supremum of linear functionals. The functions  $\phi$  realizing this maximum are called *Kantorovich potentials*.

### 1.3 The case $c(x, y) = h(x - y)$ for $h$ strictly convex, and the existence of an optimal $T$

Here,  $X = Y = \Omega \subset \mathbb{R}^d$  and the cost  $c$  is of the form  $c(x, y) = h(x - y)$ , for a strictly convex function  $h$ . We will also assume  $\Omega$  to be compact for simplicity.

**Proposition 1.15** If  $c$  is  $C^1$ ,  $\phi$  is a Kantorovich potential for the cost  $c$  in the transport from  $\mu$  to  $\nu$ , and  $(x_0, y_0)$  belongs to the support of an optimal transport plan  $\gamma$ , then  $\nabla\phi(x_0) = \nabla_x c(x_0, y_0)$ , provided  $\phi$  is differentiable at  $x_0$ . In particular, the gradients of two different Kantorovich potentials coincide on every point  $x_0 \in \text{spt}(\mu)$  where both the potentials are differentiable.

*Twist condition*  $c : \Omega \times \Omega \rightarrow \mathbb{R}$  differentiable w.r.t.  $x$  at every point and map  $y \mapsto \nabla_x c(x_0, y)$  is injective for every  $x_0$ .

$\implies$  goal of this condition = deduce from  $(x_0, y_0) \in \text{spt}(\gamma)$  that  $y_0$  is indeed uniquely defined from  $x_0 \implies \gamma$  is concentrated on a graph, that of the map associating  $y_0$  to each  $x_0$ : this map will be the optimal transport plan! Since this map has been constructed using  $\phi$  and  $c$  only, and not  $\gamma$ , it also provides uniqueness for the optimal  $\gamma$ . For  $c(x, y) = h(x - y)$  and  $h$  differentiable,  $\nabla\phi(x_0) = \nabla h(x_0 - y_0)$  (use subdifferential if not differentiable). For  $h$  strictly convex, analytic transport map defined by:

$$x_0 - y_0 = (\nabla h)^{-1}(\nabla\phi(x_0))$$

$\implies$  **Theorem 1.17** Given  $\mu$  and  $\nu$  probability measures on a compact domain  $\Omega \subset \mathbb{R}^d$  there exists an optimal transport plan  $\gamma$  for the cost  $c(x, y) = h(x - y)$  with  $h$  strictly convex. It is unique and of the form  $(\text{id}, T)_\# \mu$ , provided  $\mu$  is absolutely continuous and  $\partial\Omega$  is negligible. Moreover, there exists a Kantorovich potential  $\phi$ , and  $T$  and the potentials  $\phi$  are linked by

$$T(x) = x - (\nabla h)^{-1}(\nabla\phi(x))$$

**Quadratic case**  $c(x, y) = \frac{1}{2}|x - y|^2 \implies T(x) = \nabla u(x)$  with  $u$  a convex (and l.s.c.) function. Actually,  $u$  exists and is unique (Brenier Theorem).

**Theorem 1.22** Let  $\mu, \nu$  be probabilities over  $\mathbb{R}^d$  and  $c(x, y) = \frac{1}{2}|x - y|^2$ . Suppose  $\int |x|^2 dx, \int |y|^2 dy < +\infty$ , which implies  $\min(\text{KP}) < +\infty$  and suppose that  $\mu$  gives no mass to  $(d - 1)$ -surfaces of class  $C^2$ . Then there exists, unique, an optimal transport map  $T$  from  $\mu$  to  $\nu$ , and it is of the form  $T = \nabla u$  for a convex function  $u$ . See *Remark 1.23* for nice consequence in 1D.

**Quadratic case on the flat torus - Theorem 1.25** Take  $\mu, \nu \in \mathcal{P}(\mathbb{T}^d)$ , with  $\mu \ll \mathcal{L}^d$  and  $c(x, y) = \frac{1}{2}||x - y||^2$ . Then there exists a unique optimal transport plan  $\gamma \in \Pi(\mu, \nu)$ , it has the form  $\gamma = \gamma_T$  and the optimal map  $T$  is given by  $T(x) = x - \nabla\phi(x)$  for a.e.  $x$ , where the sum  $x - \nabla\phi(x)$  is to be intended modulo  $\mathbb{Z}^d$ . Here the function  $\phi$  is a Kantorovich potential, solution to the dual problem, in the transport from  $\mu$  to  $\nu$  for the cost  $c$ . Moreover, for a.e.  $x \in \mathbb{T}^d$ , the point  $T(x)$  does not belong to  $\text{cut}(x)$ .

## 1.4 Counter-examples to existence

(Counter examples to Theorem 1.17)

- If  $\mu = \delta_a, a \in X$  and  $\nu$  is not a Dirac mass. Then with  $T_{\#}\delta_a = \delta_{T(a)}$ , no transport map  $T : X \rightarrow Y$  can exist if  $\nu$  is not of the form  $\delta_b$  for some  $b \in Y$ . In fact,  $\forall$  atom  $a$  of  $\mu$ , there exist an atom of  $\nu$  of mass *at least*  $\mu(\{a\})$ .  $\implies$  "absence of atom for source measure  $\mu$ " = typical assumption before solving (MP). But for (KP), atomistic case = OK and solution is unique:  $\gamma = \mu \otimes \nu = \delta_a \otimes \nu$
- Optimal transport map/plan may not exist while transportation is possible (see pages 19/20)

## 1.5 Kantorovich as a relaxation of Monge

Let  $K(\gamma) = \int_{\Omega \times \Omega} c d\gamma$  and  $J(\gamma) = \begin{cases} K(\gamma) = M(T) & \text{if } \gamma = \gamma_T \\ +\infty & \text{otherwise} \end{cases}$ . (MP) is equivalent to:

$$\min\{J(\gamma) : \gamma \in \Pi(\mu, \nu)\}$$

We will see here that  $K$  can be seen as the *relaxation* of  $J$  (see Memo Box 1.10 page 20 for relaxation definition).

**Lemma 1.27** If  $\mu, \nu$  are two probability measures on the real line  $\mathbb{R}$  and  $\mu$  is atomless, then there exists at least a transport map  $T$  such that  $T_{\#}\mu = \nu$ . This is actually also true for measures on  $\mathbb{R}^d$  (corollary 1.29).

**Theorem 1.32** On a compact subset  $\Omega \subset \mathbb{R}^d$ , the set of plans  $\gamma_T$  induced by a transport is dense in the set of plans  $\Pi(\mu, \nu)$  whenever  $\mu$  is atomless.

**Theorem 1.33** For  $\Omega \subset \mathbb{R}^d$  compact,  $K$  is the relaxation of  $J$ . In particular,  $\inf J = \min K$ , and hence Monge and Kantorovich problems have the same infimum.

## 1.6 Convexity, $c$ -concavity, cyclical monotonicity, duality and optimality

**Theorem 1.37.** If  $\Gamma \neq \emptyset$  is a  $c$ -CM set in  $X \times Y$  and  $c : X \times Y \rightarrow \mathbb{R}$  (note that  $c$  is required not to take the value  $+\infty$ ), then there exists a  $c$ -concave function  $\phi : X \rightarrow \mathbb{R} \cup \{-\infty\}$  (different from the constant  $-\infty$  function) such that  $\Gamma \subset \{(x, y) \in X \times Y : \phi(x) + \phi^c(y) = c(x, y)\}$ .

**Theorem 1.38** If  $\gamma$  is an optimal transport plan for the cost  $c$  and  $c$  is continuous, then  $spt(\gamma)$  is a  $c$ -CM set.

$\implies$  from the previous two theorems, we get the following **duality result**:

**Theorem 1.39** Suppose that  $X$  and  $Y$  are Polish spaces and that  $c : X \times Y \rightarrow \mathbb{R}$  is uniformly continuous and bounded. Then the problem (DP) admits a solution  $(\phi, \phi^c)$  and we have  $\max(\text{DP}) = \min(\text{KP})$  ( $\phi$  and  $\phi^c$  are continuous and bounded).

**Extension to quadratic case** ( $c(x, y) = \frac{1}{2}|x - y|^2$  is not bounded neither uniformly continuous):

**Theorem 1.40** Let  $\mu, \nu$  be probabilities over  $\mathbb{R}$  and  $c(x, y) = \frac{1}{2}|x - y|^2$ . Suppose  $\int |x|^2 dx, \int |y|^2 dy < +\infty$ . Consider the following variant of (DP):

$$\sup\left\{\int_{\mathbb{R}^d} \phi d\mu + \int_{\mathbb{R}^d} \psi d\nu : \phi \in L^1(\mu), \psi \in L^1(\nu), \phi \oplus \psi \leq c\right\} \quad (\text{DP - var})$$

Then (DP-var) admits a solution  $(\phi, \psi)$ , and the functions  $x \mapsto \frac{1}{2}|x|^2 - \phi(x)$  and  $y \mapsto \frac{1}{2}|y|^2 - \psi(y)$  are convex and conjugate to each other for the Legendre transform. Moreover, we have  $\max(\text{DP - var}) = \min(\text{KP})$ .

**Extension of duality in the l.s.c. case**

**Theorem 1.42** If  $X, Y$  are Polish spaces and  $c : X \times Y \rightarrow \mathbb{R} \cup \{+\infty\}$  is l.s.c. and bounded from below, then the duality formula  $\min(\text{KP}) = \sup(\text{DP})$  holds. *Note that for the cost  $c$  we cannot guarantee the existence of a maximizing pair  $(\phi, \psi)$ .*

**Theorem 1.43** If  $c$  is l.s.c. and  $\gamma$  is an optimal transport plan, then  $\gamma$  is concentrated on a  $c$ -CM set  $\Gamma$  (which will not be closed in general).

#### Sufficient conditions for optimality and stability

**Theorem 1.47** Let  $\Omega \subset \mathbb{R}^d$  be compact, and  $c$  be a  $C^1$  cost function satisfying the twist condition on  $\Omega \times \Omega$ . Suppose that  $\mu \in \mathcal{P}(\Omega)$  and  $\phi \in c - \text{conc}(\Omega)$  are given, that  $\phi$  is differentiable  $\mu$ -a.e. and that  $\mu(\partial\Omega) = 0$ . Suppose that the map  $T$  satisfies  $\nabla_x c(x, T(x)) = \nabla \phi(x)$ . Then  $T$  is optimal for the transport cost  $c$  between the measures  $\mu$  and  $\nu := T_{\#}\mu$ .

**Quadratic case - Theorem 1.48** Suppose that  $\mu \in \mathcal{P}(\mathbb{R}^d)$  is such that  $\int |x|^2 d\mu(x) < +\infty$ , that  $u : \mathbb{R}^d \rightarrow \mathbb{R} \cup \{+\infty\}$  is convex and differentiable  $\mu$ -a.e., and set  $T = \nabla u$  and suppose  $\int |T(x)|^2 d\mu(x) < +\infty$ . Then  $T$  is optimal for the transport cost  $c(x, y) := \frac{1}{2}|x - y|^2$  between the measures  $\mu$  and  $\nu := T_{\#}\mu$ .

We know that the support of optimal plan are  $c$ -CM. Actually, the reverse is true: plan that are  $c$ -CM are optimal:

**Theorem 1.49** Suppose that  $\gamma \in \mathcal{P}(X \times Y)$  is given, that  $X$  and  $Y$  are Polish spaces, that  $c : X \times Y \rightarrow \mathbb{R}$  is uniformly continuous and bounded, and that  $\text{spt}(\gamma)$  is  $c$ -CM. Then  $\gamma$  is an optimal transport plan between its marginals  $\mu = (\pi_x)_{\#}\gamma$  and  $\nu = (\pi_y)_{\#}\gamma$  for the cost  $c$ .

Stability result:

**Theorem 1.50** Suppose that  $X$  and  $Y$  are compact metric spaces and that  $c : X \times Y \rightarrow \mathbb{R}$  is continuous. Suppose that  $(\gamma_n) \in \mathcal{P}(X \times Y)$  is a sequence of transport plan which are optimal for the cost  $c$  between their own marginals  $\mu_n = (\pi_x)_{\#}\gamma_n$  and  $\nu_n = (\pi_y)_{\#}\gamma_n$ , and suppose  $\gamma_n \rightarrow \gamma$ . Then  $\mu_n \rightarrow \mu := (\pi_x)_{\#}\gamma$  and  $\nu_n \rightarrow \nu := (\pi_y)_{\#}\gamma$  and  $\gamma$  is optimal in the transport between  $\mu$  and  $\nu$ .

$\implies$  Useful consequence: let's define for the cost  $c : X \times Y \rightarrow \mathbb{R}$  and  $\mu \in \mathcal{P}(X)$ ,  $\nu \in \mathcal{P}(Y)$ ,  $\mathcal{T}_c(\mu, \nu) := \min\{\int_{X \times Y} c d\gamma : \gamma \in \Pi(\mu, \nu)\}$ . Then:

**Theorem 1.51** Suppose that  $X$  and  $Y$  are compact metric spaces and that  $c : X \times Y \rightarrow \mathbb{R}$  is continuous. Suppose that  $\mu_n \in \mathcal{P}(X)$  and  $\nu_n \in \mathcal{P}(Y)$  are two sequences of probability measures, with  $\mu_n \rightarrow \mu$  and  $\nu_n \rightarrow \nu$ . Then we have  $\mathcal{T}_c(\mu_n, \nu_n) \rightarrow \mathcal{T}_c(\mu, \nu)$ .

Stability of the Kantorovich potentials:

**Theorem 1.52** Suppose that  $X$  and  $Y$  are compact metric spaces and that  $c : X \times Y \rightarrow \mathbb{R}$  is continuous. Suppose that  $\mu_n \in \mathcal{P}(X)$  and  $\nu_n \in \mathcal{P}(Y)$  are two sequences of probability measures, with  $\mu_n \rightarrow \mu$  and  $\nu_n \rightarrow \nu$ . Let  $(\phi_n, \psi_n)$  be, for each  $n$ , a pair of  $c$ -concave Kantorovich potentials for the cost  $c$  in the transport from  $\mu_n$  to  $\nu_n$ . Then, up to subsequences, we have  $\phi_n \rightarrow \phi$ ,  $\psi_n \rightarrow \psi$ , where the convergence is uniform and  $(\phi, \psi)$  is a pair of Kantorovich potentials for  $\mu$  and  $\nu$ .

## 1.7 Discussion

### 1.7.1 Probabilistic interpretation

(MP) can be seen as the following optimization problem:

$$\min\{\mathbb{E}[c(X, Y)] : X \sim \mu, Y \sim \nu\}$$

For the case  $c(X, Y) = |X - Y|^2$ , this problem becomes:

$$\max\{\mathbb{E}[(X - x_0) \cdot (Y - y_0)] : X \sim \mu, Y \sim \nu\}$$

where  $x_0 = \mathbb{E}[X]$  and  $y_0 = \mathbb{E}[Y]$ .  $\implies$  covariance maximization.

### 1.7.2 Polar Factorization

A classical result in linear algebra states that every matrix  $A \in \mathcal{M}^{N \times N}$  can be decomposed as a product  $A = SU$ , where  $S$  is symmetric and positive-semidefinite, and  $U$  is a unitary matrix, i.e.  $UU^t = I$ . The decomposition is unique if  $A$  is non-singular (otherwise  $U$  is not uniquely defined), and

in such a case  $S$  is positive definite. Also, one can see that the matrix  $U$  of this decomposition is also a solution (the unique one if  $A$  is non singular) of

$$\max\{A : R : RR^t = I\}$$

where  $A : R$  stands for the scalar product between matrices, defined as  $A : R := \text{Tr}(AR^t)$ .

Analogously, in his first works about the quadratic optimal transport, Y. Brenier noted that Monge-Kantorovich theory allowed to provide a similar decomposition for vector fields instead of linear maps:

**Theorem 1.53** Given a vector map  $\xi : \Omega \rightarrow \mathbb{R}^d$  with  $\Omega \subset \mathbb{R}^d$ , consider the rescaled Lebesgue measure  $\mathcal{L}_\Omega$  on  $\Omega$  and suppose that  $\xi_\# \mathcal{L}_\Omega$  is absolutely continuous; then, one can find a convex function  $u : \Omega \rightarrow \mathbb{R}$  and a measure-preserving map  $s : \Omega \rightarrow \Omega$  (i.e. such that  $s_\# \mathcal{L}_\Omega = \mathcal{L}_\Omega$ ) such that  $\xi = (\nabla u) \circ s$ . Moreover, both  $s$  and  $\nabla u$  are uniquely defined a.e. and  $s$  solves

$$\max\left\{\int \xi(x) \cdot r(x) dx : r_\# \mathcal{L}_\Omega = \mathcal{L}_\Omega\right\}$$

## 2 One-dimensional issues

### 2.1 Monotone transport maps and plans in 1D