

# Serie 7

## Optimal transport, Fall semester

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**Exercise 7.1** (Counterexamples). For any of the following statements, find two probability measures  $\mu, \nu \in \mathcal{P}(\mathbb{R}^d)$  with compact support such that the statement holds (you can choose also the dimension  $d \in \mathbb{N}$ ). Each of the statements should be treated independently.

- (i) There is more than one<sup>1</sup> optimal transport map from  $\mu$  to  $\nu$  with respect to the linear cost  $|x - y|$ .
- (ii) There is more than one optimal transport map from  $\mu$  to  $\nu$  with respect to the quadratic cost  $\frac{1}{2}|x - y|^2$ .
- (iii) There is not an optimal transport plan between  $\mu$  and  $\nu$  with respect to the cost  $c(x, y) = \lfloor |x - y| \rfloor$  (the floor function<sup>2</sup> of the distance).
- (iv) There is an optimal transport map from  $\mu$  to  $\nu$  with respect to the linear cost, but there is none with respect to the quadratic cost.

*Hint:* To solve (iii), show that the infimum of the Kantorovich problem for  $\mu = \chi_{[0,1]}\mathcal{L}^1$ ,  $\nu = \chi_{[1,2]}\mathcal{L}^1$  is 0 but any transport plan has *strictly* positive cost.

### Solution:

(i) Let  $d = 1$  and  $\mu := \frac{1}{2}(\delta_0 + \delta_1)$ ,  $\nu := \frac{1}{2}(\delta_1 + \delta_2)$ . Every transport map from  $\mu$  to  $\nu$  has the same cost (cf. Remark 2.7.5), thus the two maps  $T_1, T_2 : \{0, 1\} \rightarrow \{1, 2\}$  given by

$$T_1(0) = 1, \quad T_1(1) = 2, \quad T_2(0) = 2, \quad T_2(1) = 1,$$

both send  $\mu$  to  $\nu$  and are both optimal.

(ii) Let  $d = 2$  and let  $p_1, p_2, p_3 \subset \mathbb{R}^2$  be the three vertices of an equilateral triangle. Let  $\mu := \frac{1}{2}(\delta_{p_1} + \delta_{p_2})$  and  $\nu := \frac{1}{2}(\delta_{p_3} + \delta_{\frac{p_1+p_2+p_3}{3}})$ . One can explicitly check that every transport plan from  $\mu$  to  $\nu$  has the same quadratic cost. Thus the two maps from  $\mu$  to  $\nu$  (namely,  $T_1(p_1) = p_3$ ,  $T_1(p_2) = \frac{p_1+p_2+p_3}{3}$ , and  $T_2(p_1) = \frac{p_1+p_2+p_3}{3}$ ,  $T_2(p_2) = p_3$ ) are both optimal.

(iii) Let  $d = 1$  and  $\mu := dx|_{[0,1]}$ ,  $\nu := dy|_{[1,2]}$ . We will show that the infimum of the Kantorovich problem is 0, but any transport plan has *strictly* positive cost.

Given  $\varepsilon > 0$ , consider the map  $T_\varepsilon : [0, 1] \rightarrow [1, 2]$  defined as

$$T_\varepsilon(x) := \begin{cases} x + 2 - \varepsilon & \text{if } 0 \leq x \leq \varepsilon, \\ x + 1 - \varepsilon & \text{if } \varepsilon < x \leq 1. \end{cases}$$

<sup>1</sup>Uniqueness should be understood in the  $\mu$ -almost everywhere sense.

<sup>2</sup>Given  $t \geq 0$ ,  $\lfloor t \rfloor$  is the largest integer  $n$  such that  $n \leq t$ .

One can check that  $(T_\varepsilon)_\# \mu = \nu$ . Also, it holds

$$\int_{[0,1]} c(x, T_\varepsilon(x)) d\mu(x) = \int_0^\varepsilon \lfloor 2 - \varepsilon \rfloor dx + \int_\varepsilon^1 \lfloor 1 - \varepsilon \rfloor dx = \varepsilon.$$

Since  $\varepsilon$  can be chosen arbitrarily small, we have proven that the infimum of the Kantorovich problem is 0.

Let us assume by contradiction that  $\gamma \in \Gamma(\mu, \nu)$  achieves cost 0. Hence  $\gamma$  must have all his mass on the set

$$\{(x, y) \in \mathbb{R}^2 : c(x, y) = 0\} = \{(x, y) \in \mathbb{R}^2 : |x - y| < 1\}. \quad (1)$$

However, since  $\gamma \in \Gamma(\mu, \nu)$ , by the marginal condition we get

$$1 = \int_{\mathbb{R}^2} (y - x) d\gamma(x, y) = \int_{\mathbb{R}^2} |y - x| d\gamma(x, y) < 1,$$

a contradiction.

(iv) Let  $d = 1$  and  $\mu := dx|_{[0, \frac{1}{2}]} + \frac{1}{2}\delta_1$ ,  $\nu := dy|_{[\frac{5}{2}, 3]} + \frac{1}{2}\delta_2$ . Since  $\mu$  is supported on  $\{x \leq 2\}$ , whereas  $\nu$  is supported on  $\{x \geq 2\}$ , Remark 2.7.5 implies that any admissible plan has the same cost with respect to the linear cost, and this cost is given by

$$\int_{\mathbb{R}} x d\nu - \int_{\mathbb{R}} x d\mu = \frac{7}{4}.$$

In particular, any admissible map is optimal with respect to the linear cost (and it is clear that there is at least one, take for instance  $T(x) = x + 5/2$  for  $x \in [0, 1/2]$  and  $T(1/2) = 2$ ).

On the other hand, the optimal map with respect to the quadratic cost (if it exists) must be nondecreasing. Let us assume by contradiction that there is an admissible transport map  $T : \mathbb{R} \rightarrow \mathbb{R}$  that is nondecreasing. Since it is an admissible map, it must hold  $T(1) = 2$ . The monotonicity then implies that  $T(x) \leq 2$  for any  $x \leq 1$  and thus  $\nu = T_\# \mu$  is supported on  $x \leq 2$ , that is a contradiction.

**Exercise 7.2.** Let  $\mu, \nu \in \mathcal{P}(\mathbb{R}^d)$  be two compactly supported probability measures invariant under rotations (that is,  $\mu(L(E)) = \mu(E)$  and  $\nu(L(E)) = \nu(E)$  for any Borel set  $E$  and any orthogonal transformation  $L \in O(d)$ ). Assume that  $\mu \ll \mathcal{L}^d$ , and let  $T$  be the unique optimal transport map from  $\mu$  to  $\nu$  with respect to the quadratic cost (see Theorem 2.5.9). Show that  $T$  can be written as  $x \rightarrow \tau(|x|) \frac{x}{|x|}$ , where  $\tau : [0, +\infty) \rightarrow [0, +\infty)$  is a nondecreasing function.

*Hint:* The function  $\tau$  is the monotone transport map between two suitable 1-dimensional measures. Also, one may want to use (and prove) the following lemma:

**Lemma 1.** Let  $\mu^0, \mu^1 \in \mathcal{P}(\mathbb{R}^d)$  be two rotationally invariant probability measures, and let  $\Phi(x) := |x|$ . If  $\Phi_\# \mu^0 = \Phi_\# \mu^1$  then  $\mu^0 = \mu^1$ .

**Solution:**

*Sketch of the proof of Lemma 1.* We need to show that, given any smooth function  $\varphi \in C_c^\infty(\mathbb{R}^d)$ ,

$$\int \varphi \mu^0 = \int \varphi \mu^1.$$

Since  $\mu_0$  is rotationally symmetric, we know that  $\int_{\mathbb{R}^d} \varphi(Lx) d\mu^0(x) = \int_{\mathbb{R}^d} \varphi(x) d\mu^0(x)$  for any orthonormal transformation  $L \in O(d)$ . In particular, we can take all symmetries  $L \in O(d)$  and we have

$$\int_{\mathbb{R}^d} \varphi(x) d\mu^0(x) = \frac{1}{|O(d)|} \int_{O(d)} \int_{\mathbb{R}^d} \varphi(Lx) d\mu^0(x) dL = \int_{\mathbb{R}^d} \bar{\varphi}(x) d\mu^0(x)$$

where

$$\bar{\varphi}(x) = \frac{1}{|O(d)|} \int_{O(d)} \varphi(Lx) dL$$

is rotationally invariant. That is,  $\bar{\varphi}(x) = \phi(|x|)$  for some  $\phi : [0, \infty) \rightarrow \mathbb{R}$ , and since  $\Phi_\# \mu^0 = \Phi_\# \mu^1$  we have

$$\int \varphi \mu^0 = \int_{\mathbb{R}^d} \phi(|x|) d\mu^0(x) = \int_{\mathbb{R}} \phi(x) d\Phi_\# \mu^0(x) = \int_{\mathbb{R}} \phi(x) d\Phi_\# \mu^1(x) = \int_{\mathbb{R}^d} \phi(|x|) d\mu^1(x) = \int \varphi \mu^1$$

as we wanted to see.  $\square$

Let us now solve the exercise. Let  $\Phi(x) := |x|$ , and define the one-dimensional compactly supported measures

$$\tilde{\mu} := \Phi_\# \mu \in \mathcal{P}([0, +\infty)), \quad \tilde{\nu} := \Phi_\# \nu \in \mathcal{P}([0, +\infty)).$$

From the identity  $\Phi_\# dx = \omega_d r^{d-1} dr|_{[0, +\infty)}$  (use polar coordinates), where  $\omega_d$  is the measure of the unit sphere in  $\mathbb{R}^d$ , and the fact that  $\mu \ll \mathcal{L}^d$ , it follows that  $\tilde{\mu} \ll \Phi_\# \mathcal{L}^d = \omega_d r^{d-1} dr|_{[0, +\infty)} \ll dr$ . Hence, applying Theorem 2.5.9 from  $\tilde{\mu}$  to  $\tilde{\nu}$ , there exists a convex function  $\psi : [0, +\infty) \rightarrow \mathbb{R}$  such that  $\tau := \psi'$  is the optimal transport map from  $\tilde{\mu}$  to  $\tilde{\nu}$ . Notice that  $\psi$  is nondecreasing ( $\psi' \geq 0$ ).

Let us show that  $\mathcal{T}(x) := \tau(|x|) \frac{x}{|x|}$  is the optimal transport map from  $\mu$  to  $\nu$ , so  $\mathcal{T} = T$ . We begin by noticing that  $\mathcal{T}$  is the gradient of the function  $\Psi(x) := \psi(|x|)$ ,  $\mathcal{T} = \nabla \Psi$ , and that  $\Psi$  is convex (since  $\psi$  is convex and non-decreasing). By Corollary 2.5.10,  $\mathcal{T}$  is an optimal transport map. It remains only to prove  $\mathcal{T}_\# \mu = \nu$ .

First, we prove that  $\mathcal{T}_\# \mu$  is rotationally invariant and  $\Phi_\# \mathcal{T}_\# \mu = \Phi_\# \nu$ . Given  $L \in O(d)$ ,

$$L \circ \mathcal{T}(x) = L\left(\tau(|x|) \frac{x}{|x|}\right) = \tau(|x|) \frac{Lx}{|x|} = \tau(|Lx|) \frac{Lx}{|Lx|} = \mathcal{T} \circ Lx.$$

Since  $\mu$  is rotationally invariant, we get

$$L_\# \mathcal{T}_\# \mu = (L \circ \mathcal{T})_\# \mu = (\mathcal{T} \circ L)_\# \mu = \mathcal{T}_\# L_\# \mu = \mathcal{T}_\# \mu,$$

and thus  $\mathcal{T}_\# \mu$  is rotationally invariant. Also, from the identity  $\Phi \circ \mathcal{T} = \tau \circ \Phi$ , we deduce that

$$\Phi_\# \mathcal{T}_\# \mu = (\Phi \circ \mathcal{T})_\# \mu = (\tau \circ \Phi)_\# \mu = \tau_\# \tilde{\mu} = \tilde{\nu} = \Phi_\# \nu.$$

Hence, applying the Lemma stated in the hint we conclude that  $T_{\#}\mu = \nu$ , as desired.

**Exercise 7.3.** Find the optimal transport map for the quadratic cost  $c(x, y) = \frac{1}{2}|x - y|^2$  between  $\mu = f \cdot \mathcal{L}^2$  and  $\nu = g \cdot \mathcal{L}^2$  in  $\mathbb{R}^2$ , where  $f(x) = \frac{1}{\pi} \mathbb{1}_{B_1}(x)$  and  $g(x) = \frac{1}{8\pi}(4 - |x|^2) \mathbb{1}_{B_2}(x)$ .

**Solution:** Let us use Exercise 7.2. We know that the optimal transport map will be of the form  $T(x) = \tau(|x|) \frac{x}{|x|}$  for some  $\tau$  nondecreasing. We need to choose  $\tau$  such that  $T_{\#}\mu = \nu$ . It is enough to check this condition on balls. Indeed, we need to find  $\tau$  such that for any  $r \in (0, 1)$ ,

$$\mu(B_r) = \nu(T(B_r)) = \nu(B_{\tau(r)}).$$

We have

$$\mu(B_r) = \frac{1}{\pi} \int_{B_r} dx = r^2$$

and, knowing that  $\tau(r) < 2$ ,

$$\nu(B_{\tau(r)}) = \frac{1}{8\pi} \int_{B_{\tau(r)}} (4 - |x|^2) dx = \frac{1}{4} \int_0^{\tau(r)} (4 - t^2) t dt = \frac{1}{2}(\tau(r))^2 - \frac{1}{16}(\tau(r))^4.$$

That is, for all  $r \in (0, 1)$  solving  $\frac{1}{2}(\tau(r))^2 - \frac{1}{16}(\tau(r))^4 = r^2$  for  $\tau(r)$  positive we obtain  $\tau(r) = 2\sqrt{1 + \sqrt{1 - r^2}}$ . That is,

$$T(x) = 2\sqrt{1 + \sqrt{1 - |x|^2}} \frac{x}{|x|} \mathbb{1}_{B(0,1)}$$

is the optimal transport map we are looking for.

**Remark 7.1.** We gave in class a counterexample for the following statement: let us fix a cost  $c$  lower semicontinuous and  $\gamma \in \Gamma(\mu, \nu)$  such that  $\text{supp}\gamma$  is  $c$ -cyclically monotone, then  $\gamma$  is optimal.

The example is the following: let  $\mu = \mathcal{L}$  be the lebesgue measure on the one dimensional torus  $\mathbb{T} = \mathbb{R}/\mathbb{Z}$ ,  $\alpha \in \mathbb{R} \setminus \mathbb{Q}$ ,  $c : \mathbb{T} \times \mathbb{T} \rightarrow \mathbb{R}$  be defined as follows

$$c(x, y) = \begin{cases} 1 & \text{if and only if } y = x - \alpha, \\ 0 & \text{if and only if } y = x, \\ \infty & \text{otherwise,} \end{cases}$$

$\gamma = (Id, T)_{\#}\mu$  and  $T(x) = x - \alpha$ . We proved in class that  $\text{supp}\gamma$  is  $c$ -cyclically monotone but  $\gamma$  is not optimal.

**Exercise 7.4.** (✿) In this exercise we want to go deeper in the understanding of the counterexample in Remark 7.1.

- (i) Find as many points as you can in the proof of the statement “ $\text{supp}\gamma$  is  $c$ -cyclically monotone, then  $\gamma$  is optimal for a continuous cost” where it could fail.

(ii) If we define for any  $n \in \mathbb{N}$  the cost

$$c_n(x, y) = \begin{cases} 1 & \text{if and only if } y = x - \alpha, \\ 0 & \text{if and only if } y = x, \\ n & \text{otherwise.} \end{cases}$$

Is it true that  $\gamma$  is optimal for the cost  $c_n$  (where  $\gamma$  was defined above  $\gamma = (Id, T)_{\#}\mu$ )? Is it true that  $\text{supp}\gamma$  is  $c_n$ -cyclically monotone for the cost  $c_n$ ?

**Solution:**

(i) One point is when we defined

$$\varphi(x) = \sup_{N \geq 1, (x_i, y_i) \in S} \{-c(x, y_N) + c(x_N, y_N) - c(x_N, y_{N-1}) + c(x_{N-1}, y_{N-1}) + \cdots - c(x_1, y_0)\}$$

we should verify that  $\varphi(x) \not\equiv -\infty$  (which is direct if we use the quadratic cost, because  $\varphi$  is a convex function which is finite in a point) otherwise the definition of

$$\varphi^c(y) = \sup_z \{-c(y, z) - \varphi(z)\} \equiv +\infty$$

(or is not well defined, what is  $-\infty + \infty = ?$ ) which compromises the proof of

$$c(x, y) + \varphi(x) + \varphi^c(y) = 0 \iff y \in \partial_c \varphi(x)$$

used in the proof of “ $\text{supp}\gamma$  c-cyclically monotone implies  $\gamma$  optimal”. Indeed if  $x$  is such that  $\varphi(x) = -\infty$  then  $\partial\varphi(x) = \mathbb{R}^d$ , but  $c(x, y) + \varphi(x) + \varphi^c(y)$  is not well defined and in particular cannot be equal to 0.

**Remark 7.2.** As an extra exercise you can verify that  $\varphi$  defined above for the cost function  $c$  and the c-cyclically monotone set  $S = \text{supp}\gamma$  defined in Remark 7.1 we do not have the following property:  $\varphi(x) \neq -\infty$  for any point  $x \in \mathbb{T}$ . This is one point where the proof for a continuous cost function fails in the particular case of the example in Remark 7.1.

(ii) For any  $n \in \mathbb{N}$ ,  $\gamma$  is not optimal because  $(KP)=0$ , but  $\int_{\mathbb{T}^2} c(x, y) d\gamma(x, y) = 1$ .

For any  $n \in \mathbb{N}$  the support of  $\gamma$  is not  $c_n$ -cyclically monotone.

Construct the same sequence as in class, but now the closing condition cost  $n$  instead of  $+\infty$ , more precisely, let us fix  $n \in \mathbb{N}$  and we prove that  $\text{supp}\gamma$  is not  $c_n$ -cyclically monotone: let  $x_1 \in \mathbb{T}$ , for any  $i \in \mathbb{N}$  we iteratively define

$$x_{i+1} = y_i = x_i - \alpha.$$

It is clear that the sequence  $(x_i, y_i) \in \text{supp}(\gamma)$  for any  $i$  and if we take the sequence  $\{x_i\}_{i=1, \dots, n+1}$  we have that

$$n+1 = \sum_{i=1}^{n+1} c(x_i, y_i) > \sum_{i=1}^{n+1} c(x_{i+1}, y_i) = c(x_{n+2}, y_{n+1}) = c(x_1, y_{n+1}) = c(x_1, x_1 - (n+1)\alpha) = n$$

where we imposed the closing condition  $x_{n+2} = x_1$ .