

## Optimal Transport: Lecture #6

①

Let's return to the continuum formulation of the (KP). Let  $\mu, \nu \in \mathcal{P}(\mathbb{R}^d)$ ,  
$$\Pi(\mu, \nu) = \left\{ \gamma: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}_{\geq 0} \mid \begin{aligned} \gamma(A \times \mathbb{R}^d) &= \mu(A) \\ \gamma(\mathbb{R}^d \times B) &= \nu(B) \end{aligned} \forall A, B \text{ Borel} \right\}.$$

For a general cost function  $c: \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}_{\geq 0}$ , the (KP) solves

$$\inf_{\pi \in \Pi(\mu, \nu)} \int_{\mathbb{R}^d \times \mathbb{R}^d} c(x, y) d\pi(x, y). \quad (\text{KP})$$

So far, we have established (KP) admits a dual formulation, and that it can be solved exactly or approximately when  $\mu, \nu$  are supported on finite discrete point clouds.

Our next major task is to establish the existence of a unique solution to (KP).

In fact, we will establish a much stronger qualitative property of the solution to (KP). We will show the famous Brenier's Theorem: "The solution to (KP) — under certain mild conditions on  $\mu, \nu, c$  — is unique and in fact is supported on the gradient of a convex function."

We will become more explicit about these ideas, but an important remark is that when the above holds, the optimal  $\pi$  is not diffuse, but

concentrates on a set of the form  $\{(x, [\nabla u](x)) \mid x \in \mathbb{R}^d\}$  for some convex function  $u$ . In this case,  $\nu$  is the Monge map, so the solution to the (relaxed) Kantorovich problem coincides with the solution to the (hard) Monge problem. Amazing!

To ease the exposition, we'll assume  $c(x,y) = \|x-y\|_2^2$ .

//  
 • We will analyze  $\inf_{\pi \in \Pi(\mu, \nu)} \int_{\mathbb{R}^d \times \mathbb{R}^d} \|x-y\|_2^2 d\pi(x,y)$  through duality.  $\star$

We'll set up what this looks like, to develop some background on convex analysis; that may or may not be familiar from PDE. Our dim is:

Theorem (OT for Quadratic Cost): Let  $\mu, \nu \in \mathcal{P}(\mathbb{R}^d)$  have finite second moments, that is,  $\int_{\mathbb{R}^d} \|x\|_2^2 d\mu(x), \int_{\mathbb{R}^d} \|y\|_2^2 d\nu(y) < \infty$ .

(i)  $\pi \in \Pi(\mu, \nu)$  is optimal for  $\star$   
 $\iff$

$\exists \ell$ , convex and lower semi-continuous, such that  $\text{Supp}(\pi) \subset \text{Graph}(\partial \ell)$ .

For such a  $\ell$ , the pair  $(\ell, \ell^\#)$  is a minimizer to

$$\inf_{(\ell, \psi)} \left\{ \int_{\mathbb{R}^d} \ell(x) d\mu(x) + \int_{\mathbb{R}^d} \psi(y) d\nu(y) \mid x \cdot y \leq \ell(x) + \psi(y) \right\}$$

(ii) If  $\mu$  does not give mass to small sets (e.g., is a.c. in  $\mathbb{R}^d$ ), then there

exists a unique  $\hat{\Pi}$ , optimal for  $\textcircled{A}$ , which has the form

$$d\hat{\Pi}(x, y) = d\mu(x) \cdot \delta_{\{y = \nabla\ell(x)\}}(y)$$

where  $\nabla\ell$  is ~~maximizing~~ (unique up to  $\mu$ -a.e. equivalence) the gradient of a convex function that pushes  $\mu$  onto  $\nu$ :  $[\nabla\ell]_{\#}\mu = \nu$ . Moreover,  $\text{supp}(\nu) = \nabla\ell(\text{supp}(\mu))$ .

(iii) Under the assumptions of (ii),  $\nabla\ell$  is the unique solution to the Monge problem

$$\begin{aligned} \inf & \int_{\mathbb{R}^d} \|T(x) - x\|_2^2 d\mu(x) \quad (\text{MP}). \\ \text{T s.t.} & \\ T_{\#}\mu = \nu & \end{aligned}$$

(iv) If  $\nu$  also does not give mass to small sets, then for  $\mu$ -a.e.  $x$  and  $\nu$ -a.e.  $y$ ,

$$\begin{aligned} [\nabla\ell^{\#} \circ \nabla\ell](x) &= x \\ [\nabla\ell \circ \nabla\ell^{\#}](y) &= y \end{aligned}$$

and  $\nabla\ell^{\#}$  solves (MP) for pushing  $\nu$  onto  $\mu$ , i.e.

$$\begin{aligned} \inf & \int_{\mathbb{R}^d} \|T(x) - x\|_2^2 d\nu(x) \\ \text{T s.t.} & \\ T_{\#}\nu = \mu & \end{aligned}$$

• The first part (i) is known as "Krotz-Smith optimality," while the second part is usually called "Brenier's Theorem."



• We need not require  $\mu, \nu$  be a.c. only that they do not give mass to small sets, that is,  $\mu(A) = 0$  for any set  $A \subset \mathbb{R}^d$  of Hausdorff dimension  $\leq d-1$  (i.e. hyperplanes). Of course, if  $\mu, \nu$  have densities, this holds immediately. (4)

• Let's look at the primal-dual problem for our specific case  $(\star)$ .

• Following Villani, we'll scale  $c(x,y)$  by  $\frac{1}{2}$  for elegance of certain calculations:  $c(x,y) = \frac{1}{2} \|x-y\|_2^2$ . So, our functional in the primal

problem is

$$J[\pi] := \int_{\mathbb{R}^d \times \mathbb{R}^d} \frac{1}{2} \|x-y\|_2^2 d\pi(x,y).$$

Assumption:  $M_2 = \int_{\mathbb{R}^d} \frac{1}{2} \|x\|_2^2 d\mu(x) + \int_{\mathbb{R}^d} \frac{1}{2} \|y\|_2^2 d\nu(y) < +\infty$ . This is the condition that  $\mu, \nu$  have finite second moments. This ensures that for any

$\pi \in \Pi(\mu, \nu)$ ,

$$\begin{aligned} J[\pi] &= \int_{\mathbb{R}^d \times \mathbb{R}^d} \frac{1}{2} \|x-y\|_2^2 d\pi(x,y) \leq \int [\|x\|_2^2 + \|y\|_2^2] d\pi(x,y) \\ &= 2M_2 \\ &< \infty. \end{aligned}$$

• We have established, through Kantorovich Duality, that

$$\inf_{\pi \in \Pi(\mu, \nu)} J[\pi] = \sup_{(\ell, \psi) \in \Phi_c} J(\ell, \psi),$$

and moreover that the infimum on LHS is realized and is actually a minimum.

Let's recall the dual problem:

$$\sup_{(\ell, \psi) \in \mathcal{D}_c} \left\{ \int_{\mathbb{R}^d} \ell(x) d\mu(x) + \int_{\mathbb{R}^d} \psi(y) d\nu(y) \right\}$$

The constraint  $(\ell, \psi) \in \mathcal{D}_c$  is  $\ell(x) + \psi(y) \leq \frac{\|x-y\|_2^2}{2}$ ,  $\mu, \nu$ -a.e.

Breaking this down using  $\|x-y\|_2^2$

$$\begin{aligned}
 &= \langle x-y, x-y \rangle_{\mathbb{R}^d} \\
 &= \|x\|_2^2 + \|y\|_2^2 - 2 \langle x, y \rangle_{\mathbb{R}^d} \\
 &= \|x\|_2^2 + \|y\|_2^2 - 2[x \cdot y],
 \end{aligned}$$

we get  $\ell(x) + \psi(y) \leq \frac{\|x-y\|_2^2}{2}$



$$\ell(x) + \psi(y) \leq \frac{1}{2} \|x\|_2^2 + \frac{1}{2} \|y\|_2^2 - x \cdot y$$



$$x \cdot y \leq \underbrace{\left[ \frac{1}{2} \|x\|_2^2 - \ell(x) \right]}_{:= \tilde{\ell}(x)} + \underbrace{\left[ \frac{1}{2} \|y\|_2^2 - \psi(y) \right]}_{:= \tilde{\psi}(y)}$$

Let's just use the new dual variables  $(\tilde{\ell}, \tilde{\psi})$  and for convenience drop  $(\ell, \psi)$ .

Then the condition  $(\ell, \psi) \in \mathcal{D}_c$  is that  $x \cdot y \leq \tilde{\ell}(x) + \tilde{\psi}(y)$ .

So,  $\inf_{\pi \in \Pi(\mu, \nu)}$   $J[\pi] = M_2 - \sup_{\pi \in \Pi(\mu, \nu)} \int_{\mathbb{R}^d \times \mathbb{R}^d} x \cdot y d\pi(x, y)$

and  $\sup_{(\varphi, \psi) \in \tilde{\Phi}} J[\varphi, \psi] = M_2 - \inf_{\substack{(\varphi, \psi) \\ \in \tilde{\Phi}}} \{ J[\varphi, \psi] \}$ ,

where as usual  $J[\varphi, \psi] = \int_{\mathbb{R}^d} \varphi(x) d\mu(x) + \int_{\mathbb{R}^d} \psi(y) d\nu(y)$  but now

$\tilde{\Phi} = \{ (\varphi, \psi) \mid \varphi \in L^1(d\mu), \psi \in L^1(d\nu), x \cdot y \leq \varphi(x) + \psi(y) \text{ a.e.} \}$ .