

## Lecture 13 Announcements:

◦ Select Wiki Article for revision by Mon 2/21.

◦ Make up class tomorrow:

Friday 2/18, 1:30-2:45pm, SH6635

From Kantorovich back to Monge...

Thm (Knott-Smith Optimality Criterion):  
Fix  $X \subset \mathbb{R}^d$ ,  $\mu, \nu \in \mathcal{P}(X)$ . Let  $c(x^1, x^2) = |x^1 - x^2|^2$ .

(i) There exists  $f_* \in L^1(\mu)$  proper, lsc, convex  
s.t.

$$(i.a) \sup_{\substack{\varphi, \psi \in C_b(\mathbb{R}^d) \\ \varphi \oplus \psi = c}} \int \varphi d\mu + \int \psi d\nu = -P_0 = \int |x|^2 - 2f_*(x) d\mu(x) \\ + \int |x|^2 - 2f_*(x) d\nu(x)$$

(i.b) For any optimal transport plan  $\gamma_*$ ,  
we have  $x^2 \in \partial f_*(x^1)$  for  $\gamma_*$ -a.e.  $(x^1, x^2)$

(ii) Conversely, if  $\gamma \in \Gamma(\mu, \nu)$  and  $f \in L^1(\mu)$   
proper, lsc, convex for which  
 $x^2 \in \partial f(x^1)$  for  $\gamma$ -a.e.  $(x^1, x^2)$  then...

(ii.a)  $\gamma$  is optimal

$$(ii.b) -P_0 = \int |x|^2 - f(x) d\mu(x) + \int |x|^2 - 2f_*(x) d\nu(x)$$

Now, we have what we need to not only solve the Monge problem, but also to characterize its unique solution.

Remark: Recall that  $t\#\mu = \nu$   
 $\Leftrightarrow \int \varphi(t(x)) d\mu(x) = \int \varphi(y) d\nu(y) \quad \forall \varphi \in L^1(\nu)$   
 $\int d(t\#\mu|_y)$

Note that if  $t=s$   $\mu$ -a.e., then  $t\#\mu = s\#\mu$ .

Thm (Brenier): Given  $\mu, \nu \in \mathcal{P}_2(\mathbb{R}^d)$ ,  $\mu \ll \mathcal{L}^d$

① For any optimal transport plan  $\gamma$ ,  $\exists$  an optimal transport map  $t_*$  s.t.  $\gamma = (\text{id} \times t_*)\#\mu$

In particular,  $\inf_{t\#\mu=\nu} M_2(t) = \inf_{\gamma \in \Pi(\mu, \nu)} K_2(\gamma)$ .

② Given  $t$  s.t.  $t\#\mu = \nu$ ,  $t$  is optimal  $\Leftrightarrow t = \nabla \varphi$  for  $\varphi \in L^1(\mu)$  convex, lsc  
 $\nabla$  defined everywhere on  $\mathbb{R}^d$   
 $\varphi$  defined  $\mu$ -a.e. on  $\mathbb{R}^d$       differentiable  $\mu$ -a.e.

③ the OT map is unique, up to  $\mu$ -a.e. equiv

The proof relies strongly on following theorem:

Thm: Given  $\mu \in \mathcal{P}_2(\mathbb{R}^d)$ ,  $\mu \ll \mathcal{L}^d$ ,  $\varphi \in L^1(\mu)$  convex.  
then

- $\varphi$  is differentiable  $\mu$ -a.e.
- where it is differentiable,  $\partial\varphi = \{\nabla\varphi\}$
- $\nabla\varphi$  coincides with the distributional gradient.

Now, we can prove Brenier's theorem!

Pf:

CLAIM: If  $\gamma_*$  is an optimal plan and  $\gamma_* = (\text{id} \times t) \# \mu$  for some  $t$  s.t.  $t \# \mu = \nu$ , then  $t$  is an optimal transport map.

Suppose  $\tilde{T}$  is another transport map from  $\mu$  to  $\nu$ , i.e.,  $\tilde{T} \# \mu = \nu$ . Then  $\tilde{\gamma} = (\text{id} \times \tilde{T}) \# \mu \in \Gamma(\mu, \nu)$ .

$$\begin{aligned} M_2(t) &= \int |t(x) - x|^2 d\mu(x) = \int |x^2 - x^1|^2 d\gamma_* \\ &= \int |x^2 - x^1|^2 d\tilde{\gamma} = \int |\tilde{T}(x) - x|^2 d\mu(x) \\ &= M_2(\tilde{T}) \end{aligned}$$

Now prove ①.

Let  $\gamma_*$  be an OT plan. By K-S Thm,  
 $\exists f_* \in L^1(\mu)$  proper, lsc, convex s.t.

$$x^2 \in \partial f_*(x^1) \quad \gamma_*\text{-a.e.}$$

$\Updownarrow$  by prev Thm

$$x^2 = \nabla f_*(x^1) \quad \gamma_*\text{-a.e.}$$

Consequently,  $\forall \varphi \in L^1(\gamma_*)$

$$\int \varphi(x^1, x^2) d\gamma_*(x^1, x^2) = \int \varphi(x^1, \nabla f_*(x^1)) d\gamma_*(x^1, x^2)$$

$$= \int \varphi(x^1, \nabla f_*(x^1)) d\mu(x^1)$$

Thus  $\gamma_* = (\text{id} \times \nabla f) \# \mu$ .  $r \# (s \# \sigma) = (r \circ s) \# \sigma$



In particular,  $\nu = \pi_2 \# \gamma_* = \nabla f \# \mu$ ,  
so  $t = \nabla f$  is a transport map from  $\mu$  to  $\nu$ .  
By our claim, it is an optimal  
transport map.

Now, prove ②:

" $\Leftarrow$ "

Given  $t = \nabla \varphi$  for such a  $\varphi$ , define  
 $\gamma = (\text{id} \times t) \# \mu \in \mathcal{P}(\mu, \nu)$ .

Then

$$\int |x^2 - \nabla \varphi(x^1)|^2 d\gamma(x^1, x^2) = \int |H(x^1) - \nabla \varphi(x^1)|^2 d\mu(x^1) = 0$$

so  $x^2 = \nabla \varphi(x^1) \in \partial \varphi(x^1)$   $\gamma$ -a.e..

By K-S thm,  $\gamma$  is optimal.  
So CLAIM ensures  $t$  optimal.

" $\Rightarrow$ "

Given  $t$  optimal,  $\gamma = (\text{id} \times t) \# \mu$   
is an optimal plan.

By K-S thm,  $\exists \varphi_*$  satisfying  
hypotheses s.t.  $x_2 = \nabla \varphi_*(x^1)$   $\gamma$ -a.e.

Thus,

$$0 = \int |x^2 - \nabla \varphi_*(x^1)|^2 d\gamma(x^1, x^2) = \int |H(x^1) - \nabla \varphi_*(x^1)|^2 d\mu(x^1)$$

Therefore,  $t = \nabla \varphi$   $\mu$ -a.e.

Now show ③

Suppose  $t, \tilde{t}$  are OT maps.

By ②,  $\exists \varphi, \tilde{\varphi} \in L^1(\mu)$  convex, lsc, proper  
s.t.  $t = \nabla \varphi$ ,  $\tilde{t} = \nabla \tilde{\varphi}$ .

Arguing as before,

$$\gamma := (\text{id} \times t) \# \mu, \quad \tilde{\gamma} := (\text{id} \times \tilde{t}) \# \mu$$

then  $x_2 \in \partial \varphi(x_1)$   $\gamma$ -a.e.,  $x_2 \in \partial \tilde{\varphi}(x_1)$   $\tilde{\gamma}$ -a.e.  
and

$$\begin{aligned} & \int |x_1|^2 - 2\varphi(x_1) + |x_2|^2 - 2\varphi^*(x_2) d\gamma(x_1, x_2) \\ & \stackrel{||}{=} \int |x|^2 - 2\varphi(x) d\mu(x) + \int |x|^2 - 2\varphi^*(x) d\nu(x) \\ & \quad - P_0 \\ & \stackrel{||}{=} \int |x|^2 - 2\tilde{\varphi}(x) d\mu(x) + \int |x|^2 - 2\tilde{\varphi}^*(x) d\nu(x) \\ & \stackrel{||}{=} \int |x_1|^2 - 2\tilde{\varphi}(x_1) + |x_2|^2 - 2\tilde{\varphi}^*(x_2) d\tilde{\gamma}(x_1, x_2) \end{aligned}$$

Rearranging and recalling that  $x^2 \in \partial \tilde{\varphi}(x^1)$  iff equality holds in Young's inequality,

$$\int \varphi(x^1) + \varphi^*(x^2) d\tilde{\gamma}(x^1, x^2) = \int \tilde{\varphi}(x^1) + \tilde{\varphi}^*(x^2) d\tilde{\gamma}(x^1, x^2)$$

$$= \int x^1 \cdot x^2 d\tilde{\gamma}(x^1, x^2)$$

Rearranging,

$$\int \varphi(x^1) + \varphi^*(x^2) - x^1 \cdot x^2 d\tilde{\gamma}(x^1, x^2) = 0$$

$$\int \varphi(x^1) + \varphi^*(\nabla \tilde{\varphi}(x^1)) - x^1 \cdot \nabla \tilde{\varphi}(x^1) d\tilde{\gamma}(x^1, x^2)$$

$$\int \varphi(x^1) + \varphi^*(\nabla \tilde{\varphi}(x^1)) - x^1 \cdot \nabla \tilde{\varphi}(x^1) d\mu(x^1)$$

$\geq 0$  by Young

Thus  $\varphi(x^1) + \varphi^*(\nabla \tilde{\varphi}(x^1)) - x^1 \cdot \nabla \tilde{\varphi}(x^1) = 0$   $\mu$ -a.e.

$$\Downarrow$$

$$\nabla \tilde{\varphi}(x^1) \in \partial \varphi(x^1) \quad \mu\text{-a.e.}$$

$$\Downarrow$$

$$\underbrace{\nabla \tilde{\varphi}(x^1)}_{\tilde{x}} = \underbrace{\nabla \varphi(x^1)}_x \quad \mu\text{-a.e.}$$

This shows OT maps are unique  $\mu$ -a.e.  $\square$

In this way, the dual of the Kantorovich problem, with  $c(x^1, x^2) = |x^1 - x^2|^2$  on  $\mathbb{R}^d$ , helped us solve the Monge problem in this case.

The solution of the Monge problem will then be a key component in proving

$$W_2(\mu, \nu) := \inf_{\gamma \in \Pi(\mu, \nu)} K_2(\gamma)$$

is a metric on the space  $\mathcal{P}_2(\mathbb{R}^d)$ .

But, before we leave duality behind, one last important application to the case  $c(x^1, x^2) = |x^1 - x^2|$ .

Recall...

Thm: For all  $\mu, \nu \in \mathcal{P}(X)$ ,  $c: X \times X \rightarrow [0, +\infty)$  cts,

$$\inf_{\gamma \in \Gamma(\mu, \nu)} K(\gamma) = \sup_{(\varphi, \psi) \in C(X) \times C(X)} \int \varphi d\mu + \int \psi d\nu$$



$$\underbrace{\quad}_{-D_0} \quad \underbrace{\varphi \oplus \psi \leq c}_{-P_0}$$

Furthermore, the maximum is attained.

In the particular case  $X \subset \mathbb{R}^d$ ,  $c(x^1, x^2) = |x^1 - x^2|$ , we can rewrite this in a nice way.

Thm: Given  $X \subset \mathbb{R}^d$ ,  $\mu, \nu \in \mathcal{P}(X)$ ,

$$\inf_{\delta \in \mathcal{P}(\mu, \nu)} K_1(\delta) = \sup_{\varphi \in C(X), \|\varphi\|_{\text{Lip}} \leq 1} \int \varphi d(\mu - \nu)$$

and  $\exists \varphi^*$  that achieves maximum.

This gives another important way to compare probability measures, known as the 1-Wasserstein or Earth Movers Distance,

$$W_1(\mu, \nu) := \inf_{\delta \in \mathcal{P}(\mu, \nu)} K_1(\delta).$$

Remark: The first part of the theorem continues to hold on any Polish space, under the additional constraint that  $\varphi \in L^1(|\mu - \nu|)$

Pf: By our duality theorem, it suffices to show

$$\sup_{\substack{\varphi, \psi \in C(X) \\ \varphi(x^1) + \psi(x^2) \leq |x^1 - x^2|}} \int \varphi d\mu + \int \psi d\nu = \sup_{\varphi \in C(X), \|\varphi\|_{Lip} \leq 1} \int \varphi d(\mu - \nu)$$

First " $\geq$ "

Note that  $\varphi \in C(X), \|\varphi\|_{Lip} \leq 1$ , then  $\varphi(x^1) - \varphi(x^2) \leq |x^1 - x^2|$ , so taking  $\psi = -\varphi$  gives  $(\varphi, \psi)$  satisfying constraints on LHS w/ values of objective functions equal.

Next " $\leq$ "

Take  $(\varphi^*, \psi^*)$  that attain maximum on LHS.