

Lecture 12

Office Hours

Reminders

• Today 11-12

• Solutions for 2-3 exercises

• Tomorrow 10-11

• First wiki article due Fri Feb 14

• Select article to revise by Fri, Feb 21

Recall:

Thm: Suppose $(X, d_X), (Y, d_Y)$ are cpt Polish spaces and $c \in C(X \times Y), c \geq 0$. Then $\forall \mu \in \mathcal{P}(X), \nu \in \mathcal{P}(Y)$,

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

$\underbrace{\hspace{10em}}_{= -D_0} \qquad \underbrace{\varphi(x) + \psi(y) \leq c(x, y) \quad \forall x, y}_{= -P_0}$

Furthermore, the maximum is attained.

Exercise 24: Consider the problem

$$\inf_{\substack{x \in \mathbb{R}^m \\ Ax \geq b}} \sum_{i=1}^m x_i c_i$$

Find the dual problem

Exercise 25: Consider metric spaces X, Y and $\mu \in \mathcal{P}(X), \nu \in \mathcal{P}(Y)$ given by

$$\mu = \sum_{i=1}^m \delta_{x_i} u_i, \quad \nu = \sum_{j=1}^n \delta_{y_j} v_j.$$

If $c: X \times Y \rightarrow \mathbb{R}$ is the cost fn in KP, show that

$$C_{ij} := c(x_i, y_j)$$

$$KP = \inf_{P \in \mathcal{U}(\mu, \nu)} \sum_{i,j} C_{ij} P_{ij}$$

$$\mathcal{U}(\mu, \nu) = \left\{ P \in \mathbb{R}^{m \times n} : P_{ij} \geq 0, \sum_j P_{ij} = u_i, \sum_i P_{ij} = v_j \right\}$$

Next, prove that the optimum value coincides with the dual problem

$$\sup_{f \in \mathbb{R}^n, g \in \mathbb{R}^m} \sum_i f_i u_i + \sum_j g_j v_j$$

$$f_i + g_j \leq C_{ij} \quad \forall i, j$$

From Kantorovich back to Monge

Thm (Knott-Smith Optimality)
Fix $X \subseteq \mathbb{R}^d$ cpt, $c(x,y) = |x-y|^2$
 $\mu, \nu \in \mathcal{P}(X)$.

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

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

i.b For any OT plan γ_* , $y \in \partial_{\gamma_*} f_*(x)$ γ_* -a.e.
 \uparrow
same f_* for all γ_*

(ii) Conversely, if $\gamma \in \Gamma(\mu, \nu)$ and $f \in L^1(\mu)$ proper, lsc, convex for which $y \in \partial f(x)$ γ -a.e., then...
equality holds in Young's inequality γ -a.e.

ii.a γ is optimal

ii.b f satisfies \star

Remark: Though we prove the result in the case X cpt (so that solns of dual problem exist), the result continues to hold for $X = \mathbb{R}^d$ provided that $\mu, \nu \in \mathcal{P}_2(\mathbb{R}^d)$, where

$$\mathcal{P}_2(\mathbb{R}^d) = \{ \sigma \in \mathcal{P}(\mathbb{R}^d) : \int |x|^2 d\sigma(x) < +\infty \}$$

Pg:

Last time, we showed art (i).

Now, show (ii). Suppose δ and f are as in (ii).

For any $\tilde{\delta} \in \Gamma(\mu, \nu)$, we have

$$-\int |x|^2 - 2f(x) d\mu + \int |y|^2 - 2f^*(y) d\nu$$

$$= \int -|x|^2 + 2f(x) - |y|^2 + 2f^*(y) d\tilde{\delta}(x, y)$$

Cauchy's Inequality

$$\geq \int -|x|^2 + 2\langle x, y \rangle - |y|^2 d\tilde{\delta}(x, y)$$

$$= -\int |x - y|^2 d\tilde{\delta}(x, y)$$

However, for δ , equality holds throughout. Thus

$$\begin{aligned} -\int |x-y|^2 d\delta(x,y) \\ &= -\int |x|^2 - 2f(x) dx + \int |y|^2 - 2f^*(y) dy \\ &\geq -\int |x-y|^2 d\tilde{\delta}(x,y) \end{aligned}$$

Since $\tilde{\delta}$ was arbitrary, this shows δ is optimal.

This also shows ii.b. □

We now have what we need to characterize solns to Monge's Problem.

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

① If γ_* is an OT plan, $\exists t$ meas
s.t. $\gamma_* = (\text{id} \times t) \# \mu$.

"Any OT plan is induced by a transport map"

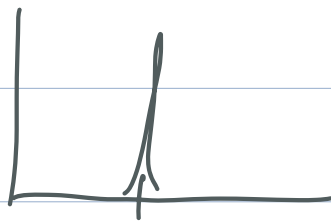
② t is an OT map $\iff (\text{id} \times t) \# \mu$ is an OT plan

③ Given t s.t. $t \# \mu = \nu$,
 t is optimal $\iff t = \nabla \varphi$ for $\varphi \in L^1(\mu)$
 μ -a.e. convex, lsc

④ the OT map from μ to ν is unique (μ -a.e.)

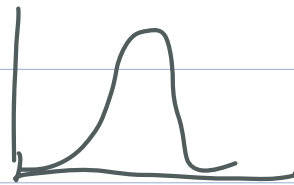
Remark: While we assume $\mu \ll \mathcal{L}^d$
for our proof, the result
continues to hold for any μ s.t.
 $\mu(S) = 0 \forall S \in \mathcal{B}(\mathbb{R}^d)$ of finite d -dim
Hausdorff measure

Recall:



x_0

$$\mu = \delta_{x_0}$$



ν

We know an OT plan exists.

It's clear that no OT map can exist.

This motivates why μ can't give mass to "small sets"

The proof relies strongly on the following theorems:

Thm (Rademacher): Given $U \subseteq \mathbb{R}^d$ open and $f: U \rightarrow \mathbb{R}$ Lipschitz, f is differentiable Lebesgue a.e. on U .

Pf: See, e.g., Folland p108 Ex 37

Thm: Given $\mu \in \mathcal{P}(\mathbb{R}^d)$, $\mu \ll \mathcal{L}^d$, $\varphi: \mathbb{R}^d \rightarrow \mathbb{R} \cup \{+\infty\}$ convex, $\varphi \in L^1(\mu)$,
(i) φ is differentiable μ -a.e.
(ii) where it is differentiable, $\partial\varphi = \{\nabla\varphi\}$

Exercise 27: Prove part (i).

Now, we can prove Brenier's theorem!

Pf:

First show (1). Suppose γ_* is an OT plan. By K-S thm, $\exists f_* \in L^1(\mu)$ proper, lsc, convex s.t.

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

Let $B := \{x \in \mathbb{R}^d : f_* \text{ is not differentiable}\}$

By previous thm

$$0 = \mu(B) = \gamma_*(B \times \mathbb{R}^d),$$

so $\partial f_*(x) = \{\nabla f_*(x)\} \quad \gamma_*\text{-a.e.}$

Thus $y = \nabla f_*(x) \quad \gamma_*\text{-a.e.}$ \star

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

$$\begin{aligned}
 \int g(x,y) d\gamma^*(x,y) &= \int g(x, \nabla f_*(x)) d\gamma^*(x,y) \\
 &= \int g(x, \nabla f_*(x)) d\mu(x) \\
 &= \int g(x,y) (\text{id} \times \nabla f_*) \# \mu
 \end{aligned}$$

Thus, $\gamma^* = (\text{id} \times \nabla f_*) \# \mu$.

Note: same f_* for any OT plan γ^*

Now show (2).

" \Leftarrow " is exercise 28

" \Rightarrow "

Let t be an OT map. Then

$$K((\text{id} \times t) \# \mu) \leq K((\text{id} \times s) \# \mu)$$

for any s s.t. $s \# \mu = \nu$.

Let γ_* be an OT plan. Then

$$\gamma_* = (\text{id} \times \nabla f_*) \# \mu. \quad \text{Thus,}$$

$$|K((id \times t) \# \mu) \leq |K(\gamma_*)$$

and $(id \times t) \# \mu$ is an OT plan.

Now show (3). Consider t s.t. $t \# \mu = \gamma$.

First, suppose t is optimal. By (2), $(id \times t) \# \mu$ is an OT plan. By (1) \star

$$y = \nabla f_*(x) \quad \gamma \text{-a.e.}$$

Thus,

$$0 = \int |y - \nabla f_*(x)| d\gamma(x, y)$$

$$= \int |t(x) - \nabla f_*(x)| d\mu(x),$$

$$\text{so } t = \nabla f_* \quad \mu \text{-a.e.}$$

(This shows (4).)

Finally, suppose $t = \nabla\varphi$ for $\varphi \in L^1(\mu)$
convex, lsc.

Define $\gamma := (\text{id} \times t) \# \mu \in \Gamma(\mu, \nu)$.

Then,

$$\int |y - \nabla\varphi(x)| d\gamma(x, y)$$

$$= \int |t(x) - \nabla\varphi(x)| d\mu(x) = 0,$$

so $y = \nabla\varphi(x) \in \{\partial\varphi(x)\}$ γ -a.e.

By K-S, γ is optimal. By (2),
 $t \circ \mu$ is an OT map. \square

Immediate consequence: Given $\mu \ll \mathcal{L}^d$.
For any $a > 0$, $b \in \mathbb{R}^d$, $t(x) = ax + b$,
 t is the unique OT map from μ to $t \# \mu$.

We just saw how the dual of KP helped us solve Monge's problem for $c(x,y) = |x-y|^2$.

More generally, similar arguments can be used to show...

Thm: Given $X \subseteq \mathbb{R}^d$ cpt, $\mu, \nu \in \mathcal{P}(X)$, $c(x,y) = h|x-y|^2$ for h strictly convex...

- \exists OT plan γ_*
- if $\mu \ll \mathcal{L}^d$, then γ_* is unique and $\gamma_* = (\text{id} \times t) \# \mu$ for

$$t(x) = x - \nabla h^{-1}(\nabla \varphi(x))$$

Pf: Santambrogio, Thm 1.17

One last important application of Kantorovich duality...

$$K_{\varphi}(\gamma) := \int d(x,y)^{\varphi} d\gamma(x,y)$$

Thm: Given X cpt Polish, $\mu, \nu \in \mathcal{P}(X)$,
 $c(x,y) = d(x,y)$

$$\inf_{\gamma \in \Pi(\mu, \nu)} K_1(\gamma) = \sup_{\varphi \in C(X)} \int \varphi d(\mu - \nu)$$

$$\|\varphi\|_{Lip} \leq 1$$

and $\exists \varphi^*$ that achieves \uparrow max on RHS.

$$|\varphi(x) - \varphi(y)| \leq d(x,y) \quad \forall x, y \in X$$

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

Pf: By Kantorovich duality,
it suffices to show

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

First " \geq "

Suppose $\varphi \in C(X)$, $\|\varphi\|_{\text{Lip}} \leq 1$. Then,
taking $\psi(y) := -\varphi(y)$, (φ, ψ)
satisfies constraint of LHS,
and the value of the objective
fn is the same.

Other direction next time $\ddot{\smile}$.