

Lecture 12 Announcements:

- Recall:
- First wiki articles due this Friday 2/11
 - No class Tuesday 2/15, rescheduled to Friday 2/18, 1:30-2:45pm, SH6635
- Suppose X cpt Polish space.

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

$$\inf_{\gamma \in \Gamma(\mu, \nu)} \int \varphi d\mu + \int \psi d\nu$$

$\underbrace{\hspace{10em}}_{-D_0 \text{ "Primal"}}$ $\underbrace{\hspace{10em}}_{-P_0}$ $\underbrace{\hspace{10em}}_{\text{"Dual"}}$

$(\varphi, \psi) \in C(X) \times C(X)$
 $\varphi \oplus \psi \leq c$

Furthermore, the maximum is attained.

From Kantorovich back to Monge

Questions:

- ① When does an OT map $t(x)$ exist?
- ② When do the optima of Monge and Kantorovich's problems coincide?
- ③ When does $t(x) = \nabla \varphi(x)$ for φ convex?

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 γ_* ,
 we have $x^2 \in \partial f_*(x^1)$ for γ_* -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 γ -a.e. (x^1, x^2) then...

(ii.a) γ is optimal

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

Remark: More generally, the result
 continues to hold for $\mu, \nu \in \mathcal{P}_2(\mathbb{R}^d)$, i.e.,

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

Lemma (Double convexification for quadratic cost on \mathbb{R}^d)

Given

$$\{(\varphi, \psi)\} \in C(\mathbb{R}^d) \times C(\mathbb{R}^d), \quad F((\varphi, \psi), 0) < +\infty$$

$$-\int_{\mathbb{R}^d} \varphi d\mu - \int_{\mathbb{R}^d} \psi d\nu + \lambda \int_{\mathbb{R}^d} \{\varphi(x^1) + \psi(x^2)\} \leq |x^1 - x^2|^2$$

define

$$\tilde{\varphi}(x^1) = \inf_{x^2 \in \mathbb{R}^d} |x^1 - x^2|^2 - \psi(x^2),$$

$$\tilde{\psi}(x^2) = \inf_{x^1 \in \mathbb{R}^d} |x^1 - x^2|^2 - \tilde{\varphi}(x^1).$$

Then,

(i) $f(x^1) = \frac{1}{2}(|x^1|^2 - \tilde{\varphi}(x^1)) \in L^1(\mu)$ is

(ii) proper, lsc, and convex

$f^*(x^2) = \frac{1}{2}(|x^2|^2 - \tilde{\psi}(x^2))$

(iii) $F((\tilde{\varphi}, \tilde{\psi}), 0) \leq F((\varphi, \psi), 0)$

Remark: $\tilde{\varphi}_i, \tilde{\psi}_i$ are the Moreau-Yosida regularizations of $-\psi_i, -\varphi_i$ with respect to the square distance

Pl:

First, note that

$$\varphi(x^1) + \psi(x^2) \leq |x^1 - x^2|^2$$

$$\Downarrow$$

$$\varphi(x^1) \leq |x^1 - x^2|^2 - \psi(x^2)$$

$$\Updownarrow$$

$$\varphi(x^1) \leq \tilde{\varphi}(x^1)$$

Likewise, note that

$$\tilde{\varphi}(x^1) + \psi(x^2) \leq |x^1 - x^2|^2$$

$$\Downarrow$$

$$\psi(x^2) \leq |x^1 - x^2|^2 - \tilde{\varphi}(x^1)$$

$$\Updownarrow$$

$$\psi(x^2) \leq \tilde{\psi}(x^2)$$

Since, by definition
 $\tilde{\varphi}(x^1) + \tilde{\psi}(x^2) \leq |x^1 - x^2|^2$, we
 have,

$$F((\varphi, \psi), 0) = -\int \varphi d\mu - \int \psi d\nu$$

$$\geq -\int \tilde{\varphi} d\mu - \int \tilde{\psi} d\nu$$

$$= F((\tilde{\varphi}, \tilde{\psi}), 0)$$

Furthermore,

$$\tilde{\varphi}(x^1) = \inf_{x^2 \in X} |x^1 - x^2|^2 - \psi(x^2)$$

\Downarrow

$$\underbrace{\frac{1}{2}(|x^1|^2 - \tilde{\varphi}(x^1))}_{g^*(x^1)} = \sup_{x^2 \in X} \left(\langle x^1, x^2 \rangle - \underbrace{\frac{1}{2}(|x^2|^2 - \psi(x^2))}_{g(x^2)} \right)$$

By defn of convex conjugate,

$f(x^1) = g^*(x^1)$ is proper, lsc, convex.

Likewise,

$$\underbrace{\frac{1}{2}(|x^2|^2 - \tilde{\Psi}(x^2))}_{= g^{**}(x^2) = f^*(x^1)} = \sup_{x^1 \in X} \langle x^1, x^2 \rangle - \underbrace{\frac{1}{2}(|x^1|^2 - \tilde{\Phi}(x^1))}_{= g^*(x^1) = f(x^1)}$$

To see $f \in L^1(\mu)$, note that Young's inequality implies

$$g^*(x^1) + g(x^2) \geq \langle x^1, x^2 \rangle$$

Hence, $g^*(x^1) \geq \langle x^1, x^2 \rangle - \frac{1}{2}(|x^2|^2 - \Psi(x^2))$

OTOH, $g^*(x^1) = \frac{1}{2}(|x^1|^2 - \tilde{\Phi}(x^1)) \leq \frac{1}{2}(|x^1|^2 - \Phi(x^1))$

Thus, $f = g^* \in L^1(\mu)$. \square

Now, we have everything we need to prove the Knott-Smith optimality criterion.

Pf:

Part (i)

By Kantorovich Duality Thm, \exists
 $\varphi_0, \psi_0 \in C(X)$ s.t. $\varphi_0(x^1) + \psi_0(x^2) \leq |x^1 - x^2|^2$
with

$$-P_0 = \int \varphi_0 d\mu + \int \psi_0 d\mu = -F((\varphi_0, \psi_0), 0).$$

By Double Convexification Lemma,
 $\exists f \in L^1(\mu)$ proper, lsc, convex where

$$F((\varphi_0, \psi_0), 0) \geq F((\tilde{\varphi}, \tilde{\psi}), 0)$$

$$\text{for } f(x^1) = \frac{1}{2}(|x^1|^2 - \tilde{\varphi}(x^1)) \\ f^*(x^2) = \frac{1}{2}(|x^2|^2 - \tilde{\psi}(x^2)).$$

Thus if γ_* is an OT plan,

$$P_0 = F((\varphi_0, \psi_0), 0)$$

$$\geq F((\tilde{\varphi}, \tilde{\psi}), 0)$$

$$= F((|x^1|^2 - 2f(x^1), |x^2|^2 - 2f^*(x^2)), 0)$$

$$\star = -\int |x^1|^2 - 2f(x^1) d\mu(x^1) - \int |x^2|^2 - 2f^*(x^2) d\nu(x^2)$$

$$\odot = -\int |x^1|^2 - 2f(x^1) + |x^2|^2 - 2f^*(x^2) d\gamma_{\star}(x^1, x^2)$$

$$\stackrel{\text{Young}}{\geq} -\int |x^1|^2 + |x^2|^2 - 2\langle x^1, x^2 \rangle d\gamma_{\star}(x^1, x^2)$$

$$= -\int |x^1 - x^2|^2 d\gamma_{\star}(x^1, x^2)$$

$$= D_0$$

$$= P_0$$

Thus, equality must hold throughout.

\star ensures (i.a).

Subtracting eqn below \odot from \odot ,

$$\int (f(x^1) + f^*(x^2) - \langle x^1, x^2 \rangle) d\gamma_{\star}(x^1, x^2) = 0$$

Since Young's inequality guarantees the integrand is nonnegative, it must vanish γ_{\star} -a.e.

Thus $x^2 \in \partial f(x^1)$ γ -a.e.

Part (ii)

Suppose

- $\gamma \in \Gamma(\mu, \nu)$
- $f \in L^1(\mu)$ proper, lsc, convex
- $x^2 \in \partial f(x^1)$ γ -a.e.

WTS

(ii.a) γ is optimal

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

Since equality holds in Young's inequality γ -a.e.

$$- \int |x^1|^2 - 2f(x^1) d\mu(x^1) - \int |x^2|^2 - 2f^*(x^2) d\nu(x^2)$$

$$= - \int |x^1|^2 - 2f(x^1) + |x^2|^2 - 2f^*(x^2) d\gamma(x^1, x^2)$$

u \downarrow Young equality

$$= - \int |x^1|^2 - 2\langle x^1, x^2 \rangle + |x^2|^2 d\gamma(x^1, x^2)$$

$$= - \int |x^1 - x^2|^2 d\gamma(x^1, x^2)$$

For arbitrary $\gamma \in \Gamma(\mu, \nu)$ we have " \geq " at μ .

Thus

$$-\int |x^1 - x^2|^2 d\gamma(x^1, x^2)$$

$$= -\int |x^1|^2 - 2f(x^1) d\mu(x^1) - \int |x^2|^2 - 2f^*(x^2) d\nu(x^2)$$

$$\geq -\int |x^1 - x^2|^2 d\gamma'(x^1, x^2)$$

Thus γ is optimal.

Finally, since γ is optimal, which shows (ii.b)

Applying \rightsquigarrow again shows (ii.a). \square

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



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

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

"Any OT plan is induced by an OT map"

② Given t s.t. $t \# \mu = \nu$, t is optimal $\Leftrightarrow t = \nabla \varphi$ for $\varphi \in L^1(\mu)$ convex, lsc, differentiable μ -a.e.

\swarrow defined everywhere on \mathbb{R}^d
 \uparrow defined μ -a.e. on \mathbb{R}^d

③ the OT map is unique, up to μ -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

- φ is differentiable μ -a.e.
- where it is differentiable, $\partial \varphi = \{\nabla \varphi\}$

• $\nabla \varphi$ coincides with the distributional gradient.

Sketch of Proof:

- Any convex function is locally Lip on $\text{Int}(D(\varphi))$.
- $D(\varphi)$ is convex, so $\partial D(\varphi)$ has Lebesgue meas 0.
- Thus φ is differentiable a.e. on $D(\varphi)$.
- $\varphi \in L^1(\mu) \Rightarrow \mu(D(\varphi)) = 1 \Rightarrow \varphi$ is differentiable μ -a.e.

Now solve Monge's problem!