

An Invitation to Statistics in Wasserstein Space

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- 1 Optimal Transportation Problem
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1 Optimal Transportation Problem

2 Kantorovich Duality

Definition (measure-preserving map)

A map $T : X \rightarrow Y$ is measure-preserving if for any measurable set $B \subset Y$, we have $\mu(T^{-1}(B)) = \nu(B)$. The measure-preserving condition is denoted as $T_{\#}\mu = \nu$, where $T_{\#}\mu$ is the push-forward measure induced by T .

Definition (OT map)

The solution to Monge's problem are called the OT maps, i.e. the OT maps are

$$\operatorname{argmin}_{T_{\#}\mu=\nu} \int_X c(x, T(x)) d\mu(x)$$

The total transportation cost of an OT map is called the Wasserstein distance between μ and ν , denoted as $W_c(\mu, \nu)$,

$$W_c(\mu, \nu) = \inf_{T_{\#}\mu=\nu} \int_X c(x, T(x)) d\mu(x)$$

Kantorovich Problem

Let the projection maps formally be $\pi_x(x, y) = x, \pi_y(x, y) = y$, then define the joint measure class as follows:

$$\Pi(\mu, \nu) := \{\rho(x, y) : X \times Y \rightarrow \mathbb{R} : (\pi_x)_\# \rho = \mu, (\pi_y)_\# \rho = \nu\}$$

Kantorovich's Problem (KP)

Given a transport cost function $c(x, y) : X \times Y \rightarrow \mathbb{R}^+$, find the joint probability measure $\rho(x, y) : X \times Y \rightarrow \mathbb{R}^+$ that minimizes the total transport cost.

$$(KP) \quad W_c(\mu, \nu) = \inf_{\rho \in \Pi(\mu, \nu)} \int_{X \times Y} c(x, y) d\rho(x, y)$$

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(Prime Problem)

$$\begin{aligned} \min \quad & b^T y \\ \text{s.t.} \quad & A^T y = c \\ & y \geq 0 \end{aligned}$$

$$L(y, \lambda) = b^T y - \lambda^T (A^T y - c)$$

$$\begin{aligned} g(\lambda) &= \inf_{y \geq 0} L(y, \lambda) \\ &= \inf_{y \geq 0} (b^T y - \lambda^T (A^T y - c)) \\ &= \inf_{y \geq 0} ((b^T - \lambda^T A^T)y + \lambda^T c) \end{aligned}$$

$$= \begin{cases} +\infty, & A^T \lambda > b \\ c^T \lambda, & A^T \lambda \leq b \end{cases}$$

(Duality Problem)

$$\begin{aligned} \max \quad & c^T x \\ \text{s.t.} \quad & A^T x \leq b \end{aligned}$$

(Discrete KP)

$$\begin{aligned} \min \quad & \sum_{i=1}^n \sum_{j=1}^m \pi_{ij} c_{ij} \\ \text{s.t.} \quad & \sum_{j=1}^m \pi_{ij} = \mu_i, i = 1, \dots, n \\ & \sum_{i=1}^n \pi_{ij} = \nu_j, j = 1, \dots, n \\ & \pi_{ij} \geq 0 \end{aligned}$$

(Discrete DP)

$$\begin{aligned} \max \quad & \sum_{i=1}^n \varphi_i \mu_i + \sum_{j=1}^m \psi_j \nu_j \\ \text{s.t.} \quad & \varphi_i + \psi_j \leq c_{ij} \end{aligned}$$

Duality of KP

Given a transport cost function $c(x, y) : X \times Y \rightarrow \mathbb{R}^+$, find the real integrable functions $\varphi : X \rightarrow \mathbb{R}$ and $\psi : Y \rightarrow \mathbb{R}$ such that

$$(DP) \quad \sup_{\varphi, \psi} \left[\int_X \varphi(x) d\mu + \int_Y \psi(y) d\nu : \varphi(x) + \psi(y) \leq c(x, y) \right]$$

Kantorovich Duality

We can directly derive the weak duality by

$$\int_X \varphi(x) d\mu(x) + \int_Y \psi(y) dv = \int_{X \times Y} c(x, y) d\rho(x, y) \leq W_c(\mu, \nu)$$

Theorem (Kantorovich Duality)

Let μ and ν be probability measures on complete separable metric space X and Y , respectively, and let $c(x, y) : X \times Y \rightarrow \mathbb{R}^+$ be a measurable function. Then

$$\inf_{\rho \in \Pi(\mu, \nu)} \int_{X \times Y} c(x, y) d\rho(x, y) = \sup_{\varphi, \psi} \left[\int_X \varphi(x) d\mu + \int_Y \psi(y) dv : \varphi(x) + \psi(y) \leq c(x, y) \right]$$

Kantorovich Duality

Next, we consider the case that μ are absolutely continuous with respect to Lebesgue measure and the cost takes form as follows:

$$c(x, y) = \frac{\|x - y\|^2}{2}$$

Definition (Legendre transform)

Given a function $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$, its Legendre transform is defined as follows:

$$\varphi^*(y) = \sup_x [\langle x, y \rangle - \varphi(x)]$$

Legendre transform is of fundamental importance in convex analysis. Suppose that an oracle tells us that some φ is a good candidate. Then the largest possible ψ satisfying $(\varphi, \psi) \in \{(\varphi, \psi) : \varphi(x) + \psi(y) \leq c(x, y)\}$ is

$$\psi(y) = \inf_{x \in \mathbb{R}^d} \left[\frac{\|x - y\|^2}{2} - \varphi(x) \right] = \frac{\|y\|^2}{2} + \inf_{x \in \mathbb{R}^d} \left[\frac{\|x\|^2}{2} - \varphi(x) - \langle x, y \rangle \right]$$

Kantorovich Duality

In other words,

$$\tilde{\psi} := \frac{\|y\|^2}{2} - \psi(y) = \tilde{\varphi}^*, \quad \tilde{\varphi} = \frac{\|x\|^2}{2} - \varphi(x).$$

Now by symmetry, one can also replace $\tilde{\varphi}$ by $(\tilde{\psi})^* = (\tilde{\varphi})^{**}$, so it is reasonable to expect that an optimal dual pair should take the form $(\|\cdot\|^2/2 - \tilde{\varphi}, \|\cdot\|^2/2 - (\tilde{\varphi})^*)$. The alternative representation of the dual objective value as

$$\int_{\mathbb{R}^d} \varphi d\mu + \int_{\mathbb{R}^d} \psi dv = \frac{1}{2} \int_{\mathbb{R}^d} \|x\|^2 d\mu + \frac{1}{2} \int_{\mathbb{R}^d} \|y\|^2 dv - \int_{\mathbb{R}^d} \tilde{\varphi} d\mu - \int_{\mathbb{R}^d} (\tilde{\varphi})^* dv.$$

Suppose that an optimal φ is found, according to the duality, a necessary and sufficient condition is that

$$\int_{\mathbb{R}^d \times \mathbb{R}^d} \frac{\|x - y\|^2}{2} d\rho(x, y) = \int_{\mathbb{R}^d} \varphi(x) d\mu + \int_{\mathbb{R}^d} \psi(y) dv$$

Kantorovich Duality

Equivalently,

$$\int_{\mathbb{R}^d \times \mathbb{R}^d} [\tilde{\varphi}(x) + (\tilde{\varphi})^*(y) - \langle x, y \rangle] d\rho(x, y) = 0.$$

Since we have $\tilde{\varphi}(x) + (\tilde{\varphi})^*(y) \geq \langle x, y \rangle$ everywhere, the integrand is nonnegative. Hence, the integral vanishes iff ρ is concentrated on the set of (x, y) such that $\tilde{\varphi}(x) + (\tilde{\varphi})^*(y) = \langle x, y \rangle$. By the definition of the Legendre transform as a supremum, this happens if and only if the supremum defining $(\tilde{\varphi})^*(y)$ is attained at x . Equivalently,

$$\tilde{\varphi}(z) - \tilde{\varphi}(x) \geq \langle z - x, y \rangle, \quad z \in \mathbb{R}^d.$$

This condition is precisely the definition of y being a subgradient of $\tilde{\varphi}$ at x . When $\tilde{\varphi}$ is differentiable at x , its unique subgradient is the gradient $y = \nabla \tilde{\varphi}(x)$. The problem, of course, is that $\tilde{\varphi}$ may fail to be differentiable μ -almost surely. This is remedied by assuming some regularity on the source measure μ in order to make sure that any convex function be differentiable μ -almost surely, and is done via the following regularity

result, which, roughly speaking, states that convex functions are differentiable almost surely.

Theorem 1.2 (Differentiability of Convex Functions)

Let $f : \mathbb{R}^d \rightarrow \mathbb{R} \cup \{\infty\}$ be a convex function with domain $\text{dom} f = \{x \in \mathbb{R}^d : f(x) < \infty\}$ and let \mathcal{N} be the set of points at which f is not differentiable. Then $\mathcal{N} \cap \text{dom} f$ has Lebesgue measure 0.

Kantorovich Duality

Based on the argument above, we can give a fundamental existence and uniqueness result for the Monge-Kantorovich problem.

Theorem 1.3 (Quadratic Cost in Euclidean Space)

Let μ and ν be probability measures on \mathbb{R}^d with finite second moments, and suppose that μ is absolutely continuous with respect to Lebesgue measure. Then the solution to the Kantorovich problem is unique, and is induced from a transport map T that equals μ -almost surely the gradient of a convex function φ . Furthermore, the pair $(\|x\|^2/2 - \phi, \|y\|^2/2 - \phi^*)$ is optimal for the dual problem.