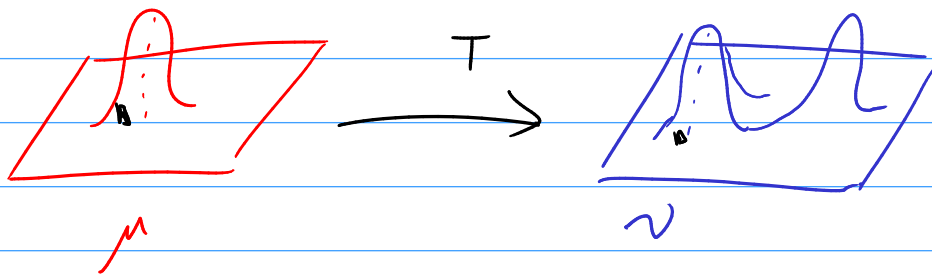


A (rough) introduction to OT

- $\Omega \subset \mathbb{R}^d$, compact
- $\mu, \nu \in \mathcal{P}(\Omega)$ are prob. measures (usually w/ 2+ moments)



T is a **map** that turns one pile of sand (μ) into another (ν).

$$T: \Omega \rightarrow \Omega, \quad T_{\#}\mu = \nu; \quad \forall A \subseteq \Omega \text{ msble,}$$

pushforward

$$\nu(A) = \mu(T^{-1}(A))$$

==

$$\min_T \int c(x, T(x)) d\mu(x) \quad \text{s.t.} \quad T_{\#}\mu = \nu \quad (\mu)$$

$$\text{w/ } c(x, y) = h(x - y) \quad (\text{e.g. } h(\cdot) = \|\cdot\|^2)$$

- Hard to solve numerically
- Existence of maps is not guaranteed
- Modulo details on cost (strictly convex), if $\mu, \nu \ll \text{Leb}$, then $T_{\#}$ exists (as does $T_{\#}^{-1}$)

==

"optimal plans"

$$\pi \in \Pi(\mu, \nu) \Leftrightarrow \int \pi(x, y) dy = \mu(x), \quad \int \pi(x, y) dx = \nu(y)$$

$$\inf_{\pi} \int c(x,y) d\pi(x,y) \quad \text{s.t. } \pi \in \Pi(\mu, \nu) \quad (k)$$

→ much more reasonable problem & significantly more tractable.
 ↳ now, solving for a "distribution"

Discrete form of (k)

$$\mu, \nu \in \mathbb{R}^n, \text{ w/ } \sum_{i=1}^n \mu_i = 1$$

$$(k) \quad \min \langle \pi, C \rangle \quad \pi^T \mathbb{1}_n = \mu_n, \quad \pi \mathbb{1}_n^T = \nu_n$$

is a linear program

$\mathbb{R}^{n \times n}$ ←
 cost matrix
 $C(x,y) = \|x-y\|_2^2$
 $\in \mathbb{R}^{n \times n}$

$$\Rightarrow \min_{\pi} \sum_{i,j} \pi_{ij} C_{ij} \quad \text{s.t. } \pi^T \mathbb{1}_n = \mu_n, \quad \pi \mathbb{1}_n^T = \nu_n$$

↳ $\tilde{O}(n^3)$ time to solve, where \tilde{O} hides polylog factors of n .

Duality in OT

Recall

$$OT(\mu, \nu) = \min_T \int c(x, T(x)) d\mu(x) \quad \text{s.t. } T\# \mu = \nu \quad (M)$$

$$T_{\#}\mu = \nu \Leftrightarrow \forall \varphi \in C(\Sigma), \int \varphi(y) d\nu(y) = \int \varphi(T(x)) d\mu(x)$$

$$OT(\mu, \nu) = \min_T \sup_{\varphi} \int c(x, T(x)) d\mu(x) + \int \varphi d\nu - \int \varphi(T(x)) d\mu(x)$$

$$= \min_T \sup_{\varphi} \int c(x, T(x)) - \varphi(T(x)) d\mu(x) + \int \varphi d\nu$$

$\mu \ll \text{Leb}$

$$\stackrel{''}{=} \sup_{\varphi} \min_T \int c(x, T(x)) - \varphi(T(x)) d\mu(x) + \int \varphi d\nu$$

$$= \sup_{\varphi} \int \underbrace{\min_T c(x, T(x)) - \varphi(T(x))}_{=: \varphi^c} d\mu + \int \varphi d\nu$$

$$= \sup_{\varphi} \int \varphi^c d\mu + \int \varphi d\nu$$

↳ "Kantorovich potentials"

→ Possible papers/topics

- Brenier's Theorem w/ applications
- "Fast OT w/ back & forth method" $\sim O(n \log n)$



What's the deal w/ Entropy?

For $\mu, \nu \in \mathcal{P}(\Omega)$,

$$H(\mu|\nu) = \begin{cases} \int \log\left(\frac{d\mu}{d\nu}\right) d\mu & \mu \ll \nu \\ +\infty & \text{o.w.} \end{cases}$$

Entropy

$$OT(\mu, \nu) = \min_{\pi \in \Pi(\mu, \nu)} \int c \cdot d\pi \quad (K) \sim \tilde{O}(n^3)$$

$$OT_\varepsilon(\mu, \nu) = \min_{\pi \in \Pi(\mu, \nu)} \int c d\pi + \varepsilon H(\pi | \mu \otimes \nu)$$

↳ 1-strongly convex, as $\varepsilon \rightarrow 0$, we recover π^*

- 2013, Cuturi published "Sinkhorn" $\sim \tilde{O}(n^2)$
- Since then, this has been sped up to $\tilde{O}(n)$ w/ some extra assumption