## Existence and consistency of Wasserstein barycenters

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## Barycenter: Definition

#### Geodesic space:

- $\bullet$  (E, d) is a complete metric space
- Every two points  $x, y \in E$  have a mid-point  $z \in E$ . (Mid-point: Given two points x, y in a metric space (E, d), their mid-point is any point  $z \in E$  such that d(x, z) = d(z, y) = 0.5 \* d(x, y).)

#### Barycenter:

• Set  $p \ge 1$  and let (E, d) be a geodesic space and  $\mu$  a probability measure on (E, d) such that

$$\int d^p(x, x_0) d\mu(x) < \infty \tag{1}$$

for some(and thus any)  $x_0 \in E$ .

• A point  $x_0 \in E$  is called a p-barycenter of  $\mu$  if

$$\int d^p(x, x_0) d\mu(x) = \inf \left\{ \int d^p(x, y) d\mu(x); y \in E \right\}$$
 (2)

• The set of all probability measures satisfying (1) is denoted  $W_p(E)$ .

## Barycenter: Properties

- Barycenters do not always exist.
- Hopf–Rinow–Cohn–Vossen theorem states that, on locally compact geodesic spaces, every closed ball is compact. Consequently, the infimum in (2) can be taken on a compact ball, and thus existence of a barycenter is ensured.

#### 定理 1.1

Set  $p \ge 1$  and let (E, d) be a locally compact geodesic space and  $\mu \in \mathcal{W}_p(E)$ Then, there exists a barycenter of  $\mu$ .

## Wasserstein space: Definition

- Set  $p \ge 1$  and let (E,d) be a metric space. Given two measures  $\mu,v$  in  $\mathcal{W}_p(E)$ , we denote by  $\Gamma(\mu,v)$  the set of all probability measures  $\pi$  over the product set EE with first, resp. second, marginal , resp. .
- The transportation cost with cost function  $d^p$  between two measures  $\mu, v$  in  $\mathcal{W}_p(E)$ , is defined as  $\mathcal{T}_p(\mu, v) = \inf_{\pi \in \Gamma(\mu, v)} \int d^p(x, y) d\pi$ .
- The transportation cost allows to endow the set  $W_p(E)$  with a metric  $W_p$  defined by  $W_p(\mu, v) = \mathcal{T}_p(\mu, v)^{1/p}$ .

This metric is known as the p-Wasserstein distance and the metric space  $(W_p(E), W_p)$  is called the Wassertein space of (E, d).

## Wasserstein space: Properties

- NPC spaces: A complete metric space (E,d) is called a global NPC space if for each pair of points  $x_0, x_1 \in E$ , there exists  $y \in E$  such that for all  $z \in E$ ,  $d^2(z, y) \le \frac{1}{2}d^2(z, x_0) + \frac{1}{2}d^2(z, x_1) \frac{1}{4}d^2(x_0, x_1)$ .
- NPC spaces are geodesic spaces and every probability measure on such spaces that satisfies  $\int d^2(x, x_0) d\mu(x) < \infty$  for some  $x_0 \in E$  has a unique 2-barycenter.
- Wasserstein spaces are not NPC spaces in general. Two probability measures  $\mu_0, \mu_1$  can have more than one mid-point in  $(\mathcal{W}_p(E), \mathcal{W}_p)$ : each mid-point is a barycenter of  $\frac{1}{2} \left( \delta_{\mu_0} + \delta_{\mu_1} \right) \in \mathcal{W}_p(\mathcal{W}_p(E))$ .

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## Wasserstein space: Properties

- Consider a random probability measure  $\tilde{\mu}$  in  $\mathcal{W}_p(E)$ , following a distribution  $\mathbb{P}$ .
- $\mathbb{P}$  is chosen in  $\mathcal{W}_p(\mathcal{W}_p(E))$  endowed with the metric  $W_p$ .
- For all  $v \in \mathcal{W}_p(E)$ ,

$$W_p^p(\delta_v, \mathbb{P}) = \mathbb{E}\left(W_p^p(v, \tilde{\mu})\right) = \int W_p^p(v, \mu) d\mathbb{P}(\mu)$$
(3)

- For a probability  $\mathbb{P} \in \mathcal{W}_p(\mathcal{W}_p(E))$ , consider a minimizer over  $v \in \mathcal{W}_p(E)$  of  $v \mapsto \mathbb{E}\left[W_p^p(v, \tilde{\mu})\right] = W_p^p(\delta_v, \mathbb{P})$ , where  $\tilde{\mu}$  is a random probability of  $\mathcal{W}_p(E)$  with distribution  $\mathbb{P}$ .
- If exists, this probability measure is a barycenter of  $\mathbb{P}$ .

## Existence of a Wasserstein Barycenter

#### 定理 2.1

Set  $p \ge 1$  and let (E, d) be a separable locally compact geodesic space. Hence, for  $\mathbb{P} \in \mathcal{W}_p(\mathcal{W}_p(E))$ , there exists a barycenter  $\bar{\mu}_{\mathbb{P}}$  defined as

$$\bar{\mu}_{\mathbb{P}} \in \arg\min_{v \in \mathcal{W}_p(E)} \mathbb{E}\left[W_p^p(v, \tilde{\mu})\right],$$
 (4)

for  $\tilde{\mu}$  a random measure with distribution  $\mathbb{P}$ .

Using the expression (3), we can see that Theorem 2.1 can be reformulated as stating the existence of the metric projection of  $\mathbb{P}$  onto the subset of  $\mathcal{W}_p(\mathcal{W}_p(E))$  of Dirac measures.

# Consistency of the barycenter of a sequence of measures

#### 定理 2.2

Set  $p \ge 1$  and let (E, d) be a separable locally compact geodesic space. Let  $(\mathbb{P}_j)_{j\ge 1} \subset W_p(W_p(E))$  be a sequence of probability measures on  $W_p(E)$  and set  $\mu_j$  a barycenter of  $\mathbb{P}_j$ , for all  $j \in \mathbb{N}$ . Suppose that for some  $\mathbb{P} \in W_p(W_p(E))$ , we have that  $W_p(\mathbb{P}, \mathbb{P}_j) \stackrel{j \to +\infty}{\longrightarrow} 0$ . Then, the sequence  $(\mu_j)_{j\ge 1}$  is precompact in  $W_p(E)$  and any limit is a barycenter of  $\mathbb{P}$ . Corollary:

- The set of all barycenters of a given measure  $\mathbb{P} \in \mathcal{W}_p(\mathcal{W}_p(E))$  is compact.
- Suppose  $\mathbb{P} \in \mathcal{W}_p(\mathcal{W}_p(E))$  has a unique barycenter. Then for any sequence  $(\mathbb{P}_j)_{j\geq 1} \subset \mathcal{W}_p(\mathcal{W}_p(E))$  converging to  $\mathbb{P}$ , any sequence  $(\mu_j)_{j\geq 1}$  of their barycenters converges to the barycenter of  $\mathbb{P}$ .

# Example, $E = \mathbb{R}^d$ and p = 2

• Let  $\mathbb{P} \in \mathcal{W}_2(\mathcal{W}_2(\mathbb{R}))$  such that there exists a set  $A \subset \mathcal{P}_2(\mathbb{R}^d)$  of measures such that for all  $\mu \in A$ ,

$$B \in \mathcal{B}(\mathbb{R}^d), \dim(B) \le d - 1 \Longrightarrow \mu(B) = 0,$$
 (5)

and  $\mathbb{P}(A) > 0$ , then,  $\mathbb{P}$  admits a unique barycenter.

- For any sequence  $(\mathbb{P}_j)_{j\geq 1}$  converging to  $\mathbb{P}$  in  $\mathcal{W}_2(\mathcal{W}_2(\mathbb{R}))$ , the barycenters of Pjconverge to the barycenter of P.
- Proof: if v satisfies (5), then  $\mu \mapsto W_2(\mu, v)$  is strictly convex, so is  $\mu \mapsto \mathbb{E} W_2^2(\mu, \tilde{\mu})$ .

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#### Two statistical frameworks

- When confronted to the statistical analysis of a collection of probability measures in  $\mathcal{W}_p(E), \mu_1, ..., \mu_J$ , it is natural to define a notion of variability as  $V_J(\mu_1, ..., \mu_J) = \inf_{v \in \mathcal{W}_p(E)} \frac{1}{J} \sum_{j=1}^J W_p^p(v, \mu_j)$ .
- In this work, we can extend this definition.  $V(\mu) = \inf_{v \in \mathcal{W}_p(E)} \mathbf{E} \left( W_p^p(v, \tilde{\mu}) \right)$ , where  $\tilde{\mu}$  is a random probability measure in  $\mathcal{W}_p(E)$ .
- Two different frameworks: whether the number of probabilities goes to infinity or whether the probabilities are not observed directly but through empirical samples. Theorem (2.2) handles both of these settings.

#### Two statistical frameworks

- First: The distribution  $\mathbb{P} \in \mathcal{W}_p(\mathcal{W}_p(E))$  is approximated by a growing discrete distribution  $\mathbb{P}_J$  supported on J elements, with J growing to infinity. Assume that  $\mathbb{P}_J$  converges to some measure  $\mathbb{P}$  with respect to Wasserstein distance. Hence Theorem (2.2) states that the barycenter (or any barycenter if not unique) of  $\mathbb{P}_J$  converges to the barycenter of  $\mathbb{P}$  (provided  $\mathbb{P}$  has a unique barycenter).
- Second: The measures  $\mu_j$  are unknown but approximated by a sequence of measures  $\mu_j^n$  converging with respect to the Wasserstein distance to measures  $\mu_j$  when n grows to infinity. Extracting a subsequence  $\mathbb{P}_n = \sum_{j=1}^J \lambda_j \delta_{\mu_j^n}$ , where  $\mu_j^n = \frac{1}{n} \sum_{i=1}^n \delta_{X_{i,j}}$  is empirical measure. We can still get the similar conclusion.