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# A remark on the optimal transport between two probability measures sharing the same copula

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### ABSTRACT

We study the optimal transport between two probability measures on  $\mathbb{R}^n$  sharing the same copula *C*. We investigate the optimality of the image of the probability measure *dC* by the vectors of pseudo-inverses of marginal distributions.

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### 1. Optimal transport between two probability measures sharing the same copula

Given two probability measures  $\mu$  and  $\rho$ , the optimal transport theory aims at minimizing  $\int c(x, y)\nu(dx, dy)$  over all couplings  $\nu$  with first marginal  $\nu \circ ((x, y) \mapsto x)^{-1} = \mu$  and second marginal  $\nu \circ ((x, y) \mapsto y)^{-1} = \rho$  for a measurable non-negative cost function c. We use the notation  $\nu <_{\rho}^{\mu}$  for such couplings. In the present note, we are interested in the particular case of the so-called Wasserstein distance between two probability measures  $\mu$  and  $\rho$  on  $\mathbb{R}^n$ :

$$W_{p,q}(\mu,\rho) = \inf_{\nu < \frac{\mu}{\rho}} \left( \int_{\mathbb{R}^n \times \mathbb{R}^n} \|x - y\|_q^p \nu(dx, dy) \right)^{1/p}$$
(1.1)

obtained for the choice  $c(x, y) = ||x - y||_q^p$ . Here  $\mathbb{R}^n$  is endowed with the norm  $||(x_1, \dots, x_n)||_q = (\sum_{i=1}^n |x_i|^q)^{1/q}$  for  $q \in [1, +\infty)$  whereas  $p \in [1, +\infty)$  is the power of this norm in the cost function.

In dimension n = 1,  $||x||_q = |x|$  so that the Wasserstein distance does not depend on q and is simply denoted by  $W_p$ . Moreover, the optimal transport is given by the inversion of the cumulative distribution functions: whatever  $p \in [1, +\infty)$ , an optimal coupling is the image of the Lebesgue measure on (0, 1) by  $u \mapsto (F_{\mu}^{-1}(u), F_{\rho}^{-1}(u))$  where for  $u \in (0, 1), F_{\mu}^{-1}(u) = \inf\{x \in \mathbb{R} : \mu((-\infty, x]) \ge u\}$  and  $F_{\rho}^{-1}(u) = \inf\{x \in \mathbb{R} : \rho((-\infty, x]) \ge u\}$  (see for instance Theorem 3.1.2 in Rachev and Rüschendorf (1998)). This implies that  $W_p^p(\mu, \rho) = \int_{(0,1)} |F_{\mu}^{-1}(u) - F_{\rho}^{-1}(u)|^p du$ .

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In higher dimensions, according to Sklar's theorem (see for instance Theorem 2.10.11 in Nelsen (2006)),

$$\mu\left(\prod_{i=1}^{n}(-\infty,x_i]\right)=C\left(\mu_1((-\infty,x_1]),\ldots,\mu_n((-\infty,x_n])\right)$$

where we denote by  $\mu_i = \mu \circ ((x_1, \ldots, x_n) \mapsto x_i)^{-1}$  the *i*th marginal of  $\mu$  and *C* is a copula function i.e.  $C(u_1, \ldots, u_n) = m \left(\prod_{i=1}^n [0, u_i]\right)$  for some probability measure *m* on  $[0, 1]^n$  with all marginals equal to the Lebesgue measure on [0, 1]. The copula function *C* is uniquely determined on the product of the ranges of the marginal cumulative distribution functions  $x_i \mapsto \mu_i((-\infty, x_i])$ . In particular, when the marginals  $\mu_i$  do not weight points, the copula *C* is uniquely determined. Sklar's theorem shows that the dependence structure associated with  $\mu$  is encoded in the copula function *C*. Last, we give the well-known Fréchet–Hoeffding bounds

$$\forall u_1, \ldots, u_n \in [0, 1], \quad C_n^-(u_1, \ldots, u_n) \le C(u_1, \ldots, u_n) \le C_n^+(u_1, \ldots, u_n)$$

that hold for any copula function *C* with  $C_n^+(u_1, \ldots, u_n) = \min(u_1, \ldots, u_n)$  and  $C_n^-(u_1, \ldots, u_n) = (u_1 + \cdots + u_n - n + 1)^+$ (see Nelsen (2006), Theorem 2.10.12 or Rachev and Rüschendorf (1998), section 3.6). We recall that the copula  $C_n^+$  is the *n*-dimensional cumulative distribution function of the image of the Lebesgue measure on [0, 1] by  $\mathbb{R} \ni x \mapsto (x, \ldots, x) \in \mathbb{R}^n$ . Also the copula  $C_2^-$  is the two-dimensional cumulative distribution function of the image of the Lebesgue measure on [0, 1] by  $\mathbb{R} \ni x \mapsto (x, 1 - x) \in \mathbb{R}^2$  and, for  $n \ge 3$ ,  $C_n^-$  is not a copula.

In dimension n = 1, the unique copula function is C(u) = u and therefore the optimal coupling between  $\mu$  and  $\rho$ , which necessarily share this copula, is the image of the probability measure dC by  $u \mapsto (F_{\mu}^{-1}(u), F_{\rho}^{-1}(u))$ . It is therefore natural to wonder whether, when  $\mu$  and  $\rho$  share the same copula C in higher dimensions, the optimal coupling is still the image of the probability measure dC by  $(u_1, \ldots, u_n) \mapsto (F_{\mu_1}^{-1}(u_1), \ldots, F_{\mu_n}^{-1}(u_1), \ldots, F_{\rho_n}^{-1}(u_n))$ . We denote by  $\mu \diamond \rho$  this probability law on  $\mathbb{R}^{2n}$ . It turns out that the picture is more complicated than in dimension one because of the choice of the index q of the norm: optimality is guaranteed only when p = q i.e. when the cost  $||x - y||^{p_q}$  in (1.1) may be decomposed as the sum of coordinatewise costs.

**Proposition 1.1.** Let  $n \ge 2$ ,  $\mu$  and  $\rho$  be two probability measures on  $\mathbb{R}^n$  sharing the same copula C and  $\mathcal{W}_{p,q}(\mu, \rho) = \inf_{\nu < q} \left( \int_{\mathbb{R}^n \times \mathbb{R}^n} \|x - y\|_q^p \nu(dx, dy) \right)^{1/p}$ .

• If p = q, then an optimal coupling between  $\mu$  and  $\rho$  is given by  $\nu = \mu \diamond \rho$  and

$$\mathcal{W}_{p,p}^{p}(\mu,\rho) = \int_{[0,1]^{n}} \sum_{i=1}^{n} |F_{\mu_{i}}^{-1}(u_{i}) - F_{\rho_{i}}^{-1}(u_{i})|^{p} dC(u_{1},\ldots,u_{n}) = \int_{[0,1]} \sum_{i=1}^{n} |F_{\mu_{i}}^{-1}(u) - F_{\rho_{i}}^{-1}(u)|^{p} du.$$

• If  $p \neq q$ , the coupling  $\mu \diamond \rho$  is in general no longer optimal. For p < q, if  $C \neq C_n^+$ , we can construct probability measures  $\mu$  and  $\rho$  on  $\mathbb{R}^n$  admitting C as their unique copula such that

$$\left(\int_{\mathbb{R}^n\times\mathbb{R}^n}\|x-y\|_q^p\mu\diamond\rho(dx,dy)\right)^{1/p}>\mathcal{W}_{p,q}(\mu,\rho)$$

For p > q, the same conclusion holds if  $n \ge 3$  or n = 2 and  $C \ne C_2^-$ .

**Remark 1.2.** Let  $\mu$  and  $\rho$  be two probability measures on  $\mathbb{R}^n$  and  $\nu <_{\rho}^{\mu}$ . For n = 1,  $\nu$  is said to be comonotonic if  $\nu((-\infty, x], (-\infty, y]) = C_2^+(\mu((-\infty, x]), \rho((-\infty, y]))$ . Puccetti and Scarsini (2010) investigate several extensions of this notion for  $n \ge 2$ . In particular, they say that  $\nu$  is  $\pi$ -comonotonic (resp. *c*-comonotonic) if  $\mu$  and  $\rho$  have a common copula and  $\nu = \mu \diamond \rho$  (resp.  $\nu$  maximizes  $\int_{\mathbb{R}^n \times \mathbb{R}^n} \langle x, y \rangle \tilde{\nu}(dx, dy)$  over all the coupling measures  $\tilde{\nu} <_{\rho}^{\mu}$ ). Looking at some connections between their different definitions of comonotonicity, they show in Lemma 4.4 that  $\pi$ -comonotonicity implies *c*-comonotonicity. Since

$$\int_{\mathbb{R}^n \times \mathbb{R}^n} \|x - y\|_2^2 \tilde{\nu}(dx, dy) = \int_{\mathbb{R}^n} \|x\|_2^2 \mu(dx) + \int_{\mathbb{R}^n} \|y\|_2^2 \rho(dy) - 2 \int_{\mathbb{R}^n \times \mathbb{R}^n} \langle x, y \rangle \tilde{\nu}(dx, dy),$$

this yields our result in the case p = q = 2.

#### 2. Proof of Proposition 1.1

The optimality in the case q = p, follows by choosing  $d_1 = \cdots = d_n = d'_1 = \cdots = d'_n = d''_1 = \cdots = d''_n = 1$ ,  $c_i(y_i, z_i) = |y_i - z_i|^p$ ,  $\alpha = dC$ , and  $\varphi_i = F_{\mu_i}^{-1}$ ,  $\psi_i = F_{\rho_i}^{-1}$  in the following Lemma.

**Lemma 2.1.** Let  $\alpha$  be a probability measure on  $\mathbb{R}^{d_1} \times \mathbb{R}^{d_2} \times \cdots \times \mathbb{R}^{d_n}$  with respective marginals  $\alpha_1, \ldots, \alpha_n$  on  $\mathbb{R}^{d_1}, \ldots, \mathbb{R}^{d_n}$  and  $\varphi_i : \mathbb{R}^{d_i} \to \mathbb{R}^{d'_i}, \psi_i : \mathbb{R}^{d_i} \to \mathbb{R}^{d''_i}$  and  $c_i : \mathbb{R}^{d'_i} \times \mathbb{R}^{d''_i} \to \mathbb{R}_+$  be measurable functions such that

$$\forall i \in \{1, \dots, n\}, \quad \inf_{\substack{\nu_i < \varphi_i^{-1} \\ \alpha_i \circ \psi_i^{-1}}} \int_{\mathbb{R}^{d'_i} \times \mathbb{R}^{d''_i}} c_i(y_i, z_i) \nu_i(dy_i, dz_i) = \int_{\mathbb{R}^{d_i}} c_i(\varphi_i(x_i), \psi_i(x_i)) \alpha_i(dx_i).$$
(2.1)

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