

ON THE MONGE PROBLEM

The first thing to point out about the MONGE problem is that, given arbitrary probability measures μ on (X, \mathcal{F}) and ν on (Y, \mathcal{G}) , a map $T: X \rightarrow Y$ that transports μ to ν may not exist.

Indeed, the transport property $T \# \mu = \nu$ is equivalent to

$$(1) \quad \int_Y \varphi(y) \nu(dy) = \int_X \varphi(T(x)) \mu(dx), \quad \forall \varphi \text{ bdd, meas.}$$

So, if $\mu = \delta_{x_0}$ is a DIRAC measure for some x_0 , this becomes $\int_Y \varphi(y) \nu(dy) = \varphi(T(x_0))$ and forces $\nu = \delta_{T(x_0)}$. In other words, unless the given ν is precisely this DIRAC measure on (Y, \mathcal{G}) , a transport map does not exist.

THE REAL LINE : Let us concentrate for a moment on the case $X = Y = \mathbb{R}$ and on two probability measures μ, ν on $\mathcal{B}(\mathbb{R})$ with respective probability distribution functions

$$F(\cdot) \triangleq \mu((0, \cdot]), \quad G(\cdot) \triangleq \nu((0, \cdot])$$

and right-continuous inverses $F^{-1}(\cdot)$, $G^{-1}(\cdot)$, respectively.

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When $\mu = \mathcal{L}|_{[0,1]}$ is LEBESGUE measure (uniform distribution) on the unit interval, then the map $T(x) = G^{-1}(x)$, $0 \leq x \leq 1$ transports μ to ν . This is the familiar SKOROKHOD construction.

A bit more generally, when μ has no atoms (i.e., $F(\cdot)$ is continuous), the mapping

$$(2) \quad T(x) \triangleq G^{-1}(F(x)), \quad x \in \mathbb{R}$$

transports μ to ν : i.e., $T \# \mu = \nu$. This can be verified easily. The mapping in (2) above is called monotone re-arrangement.

We shall extend now this idea.

SEVERAL DIMENSIONS : Let us start with the Euclidean plane $X = Y = \mathbb{R}^2$

THEOREM : DISINTEGRATION OF MEASURE .

Fix $\mu \in \mathcal{P}(\mathbb{R}^2)$ and define $\mu_1 \triangleq P_1 \# \mu$, the "1-marginal" of μ , where

$$P_1(x_1, x_2) \triangleq x_1, \quad P_2(x_1, x_2) = x_2.$$

There exists then a family of probability measures $(\mu_{x_1})_{x_1 \in \mathbb{R}}$ on the real line, with

$$(3) \quad \mu(dx_1, dx_2) = \mu_1(dx_1) \otimes \mu_{x_1}(dx_2).$$

Illustration: Conditional Distributions.

Suppose $\mu(A) = \iint_A f(x_1, x_2) dx_1 dx_2$, $A \in \mathcal{B}(\mathbb{R}^2)$ for some

$f > 0$ with $\iint_{\mathbb{R}^2} f = 1$. Define the "marginal density"

$$f_1(x_1) \triangleq \int_{\mathbb{R}} f(x_1, x_2) dx_2, x_1 \in \mathbb{R}.$$

Then μ is the

"marginal probability measure" $\mu(\cdot) = (P_1 \# \mu)(\cdot) = \int_{\mathbb{R}} f_1(x_1) dx_1$

and

$$\mu(\cdot) = \int_{x_1} \frac{f(x_1, x_2)}{f_1(x_1)} dx_2 \quad \text{the corresponding}$$

"conditional distribution".

Now, consider two such absolutely continuous measures
on $\mathcal{B}(\mathbb{R}^2)$:

$$\mu(dx_1, dx_2) = f(x_1, x_2) dx_1 dx_2 = \underbrace{\frac{f(x_1, x_2)}{f_1(x_1)} dx_2}_{\mu(dx_2)} \otimes \underbrace{f_1(x_1) dx_1}_{\mu(dx_1)}$$

(4)

$$\nu(dy_1, dy_2) = g(y_1, y_2) dy_1 dy_2 = \underbrace{\frac{g(y_1, y_2)}{g_1(y_1)} dy_2}_{\nu(dy_2)} \otimes \underbrace{g_1(y_1) dy_1}_{\nu(dy_1)}$$

(5)

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From our discussion about monotone re-arrangements, consider the mapping $T_1: \mathbb{R} \rightarrow \mathbb{R}$ that transports μ to $(6) \quad \nu_1$, i.e., $T_1 \# \mu = \nu_1$; and, for μ -a.e. $x_1 \in \mathbb{R}$,

↓ ↓
the monotone re-arrangement

$T_2(x_1, \cdot): \mathbb{R} \rightarrow \mathbb{R}$ that transports μ to ν_2 : $T_2(x_1, \cdot) \# \mu = \nu_2$.

↓ ↓
 x_1 x_1

THEOREM (KNOTHE, 1957): The map

$(8) \quad \mathbb{R}^2 \ni (x_1, x_2) \mapsto T(x_1, x_2) = (T_1(x_1), T_2(x_1, x_2)) \in \mathbb{R}^2$

transports μ to ν_2 .

Now iterate this procedure, to extend to \mathbb{R}^3, \dots et cetera.

Here is BRENIER's theorem, the first general result about the solvability of the MONGE problem.

THEOREM (BRENIER, 1987): Let $X = Y = \mathbb{R}^n$ and consider two probability measures μ, ν with

$$\int_{\mathbb{R}^n} |x|^2 \mu(dx) + \int_{\mathbb{R}^n} |y|^2 \nu(dy) < \infty.$$

There exists then a (unique) optimal coupling $\bar{\pi} \in \mathcal{P}(\mu, \nu)$ for the KANTOROVICH problem with $c(x, y) = |x - y|^2$, and this is of the form

$$(9) \quad \bar{\pi} = (\text{Id} \times T_*) \# \mu, \quad T_* \triangleq \nabla F$$

for some convex $F: \mathbb{R}^n \rightarrow \mathbb{R}$. In particular, the map T_* in (9) is then optimal for the corresponding MONGE problem, whose value is then the same as that of the KANTOROVICH problem:

$$I(\bar{\pi}) = \mathcal{T}_c(\mu, \nu) = \mathcal{M}_c(\mu, \nu) = G(T_*) = \int_{\mathbb{R}^n} |x - \nabla F(x)|^2 \mu(dx)$$

↑
optimality of $\bar{\pi}$
for KANTOROVICH

↑
equality
of the two
values

↑
 \mathbb{R}^n
optimality of
 T_* for MONGE

MONGE - AMPÈRE EQUATION

Consider the setting of BRENIER's Theorem (first sentence) with the added feature that both measures are absolutely continuous w.r.t. LEBESGUE with densities $f > 0, g > 0$:

$$\mu(\cdot) = \int f(x) dx, \quad \nu(\cdot) = \int g(x) dx.$$

We know from that result that μ can be transported to ν via the gradient $T = DF$ of a convex function $F: \mathbb{R}^n \rightarrow \mathbb{R}$, i.e., $T^\# \mu = \nu$; equivalently,

$$(10) \quad \int_{\mathbb{R}^n} \varphi(y) \underbrace{g(y) dy}_{\nu(dy)} = \int_{\mathbb{R}^n} \varphi(DF(x)) \underbrace{f(x) dx}_{\mu(dx)}$$

for every $\varphi \in C_b(\mathbb{R}^n)$.

And let us assume, for concreteness, that the BRENIER function F is smooth, say of class $C^1(\mathbb{R}^n)$, and strictly convex; then we can write the LHS of (10), via the change-of-variable $y = DF(x)$, as

$$\int_{\mathbb{R}^n} \varphi(y) g(y) dy = \int_{\mathbb{R}^n} \varphi(DF(x)) g(DF(x)) \det(D^2 F(x)) dx.$$

Comparing with (10), and because φ is arbitrary, we get

$$(11) \quad \det(D^2 F(x)) = \frac{f(x)}{g(DF(x))}, \quad x \in \mathbb{R}^d,$$

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a partial differential (non-linear) equation
for the BRENIER map F . It is a special case of the
famous MONGE - AMPÈRE equation

$$\det(D^2F(x)) = H(x, F(x), DF(x)), \quad x \in \mathbb{R}^n.$$