

# A transport problem for two measures in the convex order

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# Outline

- 1 Notations, convex order
- 2 The model theorem
- 3 Different tools
- 4 The martingale theorem, examples

**Definition: martingale transport plan**

A probability measure  $\pi$  on  $\mathbb{R} \times \mathbb{R}$  is termed a martingale transport plan if  $\pi = \text{Law}(X, Y)$  where  $(X, Y)$  is a two-times martingale process:  $\mathbb{E}(Y|X) = X$ .

Equivalently if  $(\pi_x)_{x \in \mathbb{R}}$  is a disintegration of  $\pi$  (or “conditional laws”, or “Markov kernel”), it has to satisfy

$$\text{Barycenter}(\pi_x) = \int y \, d\pi_x(y) = x$$

for  $(\text{proj}_{\#}^x \pi)$ -almost every  $x$ .

## Some examples

- $\pi = \text{Law}(x, Y)$  where  $x = \mathbb{E}(Y)$ .
- $\pi = \text{Law}(\mathbb{E}(Y|\mathcal{F}), Y)$  for some  $\mathcal{F} \subseteq \sigma(Y)$ .
- $\pi = \text{Law}(B, B + B')$  where  $B'$  is independent from  $B$  and  $\mathbb{E}(B') = 0$ .
- $\pi_{1/2} = \frac{\pi_0 + \pi_1}{2}$  where  $\pi_0, \pi_1$  are martingale transport plans.
- $\pi_t = \sum_{i=1}^2 \sum_{j=1}^3 a_{i,j}^t \delta_{(x_i, y_j)}$  where  $x \in \{-1, 1\}$  and  $y \in \{-2, 0, 2\}$  and

$$(a_{i,j}^t) = \begin{pmatrix} 1/4 & 1/4 & 0 \\ 1/12 & 1/12 & 1/3 \end{pmatrix} + t \begin{pmatrix} 1/12 & -1/6 & 1/12 \\ -1/12 & 1/6 & -1/12 \end{pmatrix}$$

for some  $t \in [0, 1]$ .

## The convex order

### Theorem of Strassen

For two probability measures  $\mu$  and  $\nu$ ,

$$\Pi_M(\mu, \nu) \neq \emptyset \Leftrightarrow \mu \preceq_C \nu.$$

The convex order  $\mu \preceq_C \nu$  is defined by

$$\forall \varphi \text{ convex}, \int \varphi d\mu \leq \int \varphi d\nu$$

# The model theorem

## Model theorem in the classical setting

For  $\mu$  and  $\nu$  in  $\mathcal{P}_2$  in the convex order and  $\pi$  a transport plan from  $\mu$  to  $\nu$ . The following statements are equivalent:

1. The plan  $\pi$  is optimal for the transport problem with  $c(x, y) = (y - x)^2$ ,
2. The plan  $\pi$  is concentrated on a monotone set  $\Gamma$ ,
3. The plan  $\pi$  is the quantile coupling.

We (Mathias and I) have proved a theorem similar to this one in the martingale setting.

## The extended order and the shadows

### Proposition - Definition

- We write  $\mu \preceq_E \nu$  and say that  $\mu$  is smaller than  $\nu$  in the extended order if

$$F_\mu^\nu := \{\theta : \mu \preceq_C \theta \text{ and } \theta \leq \nu\}$$

is not empty.

- The partially ordered set  $(F_\mu^\nu, \preceq_C)$  has a minimum. We call it the shadow of  $\mu$  in  $\nu$  and denote it by  $S^\nu(\mu)$ .

## The extended order and the shadows

## Proposition

If  $\gamma_1 + \gamma_2 \preceq_E \nu$ , then

$$S^\nu(\gamma_1 + \gamma_2) = S^\nu(\gamma_1) + S^{\nu - S^\nu(\gamma_1)}(\gamma_2).$$

As an important consequence, if  $\gamma \leq \gamma'$  and  $\gamma' \preceq_E \nu$ ,

$$S^\nu(\gamma) \leq S^\nu(\gamma').$$

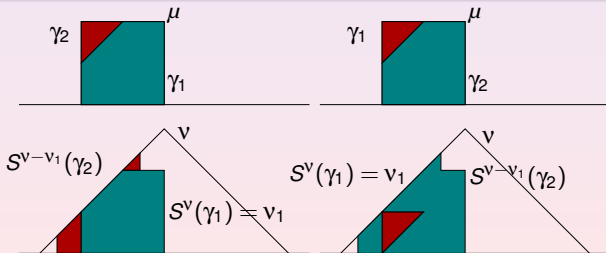


Figure : Shadow of  $\mu$  in  $\nu$  and associativity of the shadow projection.



## The variational lemma

This lemma is a kind of  $c$ -cyclical monotonicity lemma for the martingale setting.

### Variational Lemma

Let  $P$  be optimal, there exists  $\Gamma \subseteq \mathbb{R} \times \mathbb{R}$  such that for any finitely supported measure  $\alpha$  with  $\alpha(\Gamma) = 1$ , the minimum of  $\alpha' \in \text{Competitor}(\alpha) \mapsto \int c(x, y) d\alpha'(x, y)$  over

$$\text{Competitor}(\alpha) = \left\{ \alpha' : \begin{array}{l} \alpha' \text{ has the same marginals as } \alpha \\ \text{and } \forall x \in \mathbb{R}, \int y d\alpha_x = \int y d\alpha'_x \end{array} \right\}$$

is obtained in  $\alpha$ .

## The model theorem

### Model theorem in the classical setting

For  $\mu$  and  $\nu$  in  $\mathcal{P}_2$  in the convex order and  $\pi$  a transport plan from  $\mu$  to  $\nu$ . The following statements are equivalent:

1. The plan  $\pi$  is optimal for the transport problem with  $c(x, y) = (y - x)^2$ ,
2. The plan  $\pi$  is concentrated on a monotone set  $\Gamma$ ,
3. The plan  $\pi$  is the quantile coupling.

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## The general problem

### The problem

Minimize

$$f : \pi \in \Pi_M(\mu, \nu) \mapsto \int c(x, y) d\pi(x, y) \in \mathbb{R}$$

among the martingale transport plans from  $\mu$  to  $\nu$ .

For different cost functions  $c$  we would like to know:

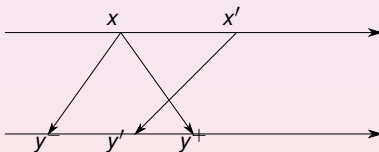
- How do the minimizers look like?
- What are their properties?
- Is there a unique minimizer?

## The martingale theorem

### Theorem (Beiglboeck–J.)

For  $\mu$  and  $\nu$  in  $\mathcal{P}_3$  in the convex order and  $\pi$  a martingale transport plan from  $\mu$  to  $\nu$ . The following statements are equivalent:

1. The plan  $\pi$  is optimal for the martingale transport problem with cost  $c(x, y) = (y - x)^3$ ,
2. The plan  $\pi$  is concentrated on a martingale-monotone set  $\Gamma$  (see the figure),
3. The plan  $\pi$  is the left-curtain coupling (i.e., transports  $\mu|_{]-\infty, x]} to its shadow)$



**Figure :** This configuration of three points  $(x, y)$ ,  $(x', y^-)$  and  $(x', y^+)$  is forbidden on martingale-monotone sets  $\Gamma$ .

## One example

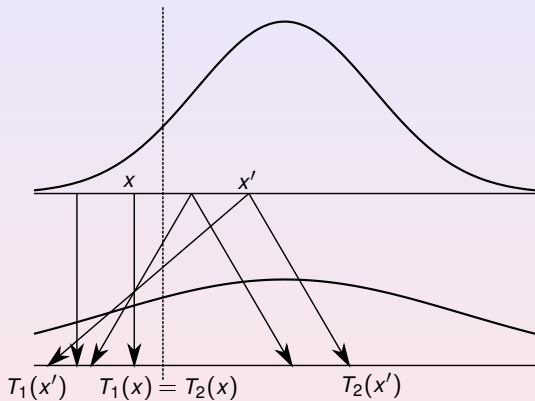


Figure : Optimal transport plan between Gaussian measures.

## Corollary

### Corollary for $\mu$ continuous

If  $\mu$  is continuous (=no atom), there are  $T_1, T_2 : \mathbb{R} \rightarrow \mathbb{R}$  such that the optimal  $\pi$  is concentrated on  $\text{graph}(T_1) \cup \text{graph}(T_2)$ .

The variational lemma is of general use, especially when  $\mu$  is continuous.

### Examples

- $c(x, y) = -|y - x|$
- $c(x, y) = |y - x|$
- $c(x, y) = (y - x)^n$
- $c(x, y) = h(y - x)$  where  $h'(x) = ax + b$  has less than  $n$  solutions.