

# Martingale Transports, Monge Maps, and Strassen's Theorem

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Optimal Transport and Distributional Robustness  
Banff, March 25–29, 2024

# Agenda

- 1 Martingale transport
- 2 Monge martingale transport maps
- 3 A refinement of Strassen's theorem
- 4 Other results
- 5 Open questions

Based on joint work with Marcel Nutz (Columbia) and Zhenyuan Zhang (Stanford)

# Main result

$X, Y \in L^1$ :  $X \leq_{\text{cx}} Y$  means  $\mathbb{E}[\phi(X)] \leq \mathbb{E}[\phi(Y)]$  for all convex  $\phi$

We will prove

A refinement of Strassen's theorem

On an atomless probability space, for  $X, Y \in L^1$ ,

$$X \leq_{\text{cx}} Y \iff X \stackrel{\text{law}}{=} \mathbb{E}[Y|\mathcal{G}] \text{ for some } \sigma\text{-field } \mathcal{G}$$

... and some other related results

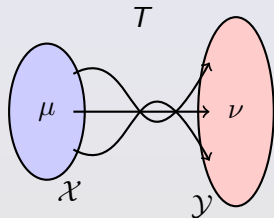
# Monge's problem

**Monge's problem:** find a transport map  $T : \mathcal{X} \rightarrow \mathcal{Y}$  that minimizes

$$\int_{\mathcal{X}} c(x, T(x)) d\mu(x) : T_{\#}\mu = \nu$$

where

- ▶  $\mathcal{X}$  and  $\mathcal{Y}$  are two Polish spaces (e.g.,  $\mathbb{R}^d$ )
- ▶ **Cost function**  $c : \mathcal{X} \times \mathcal{Y} \rightarrow [0, \infty]$  or  $(-\infty, \infty]$
- ▶ Probabilities  $\mu$  on  $\mathcal{X}$  and  $\nu$  on  $\mathcal{Y}$  are given
- ▶  $T_{\#}\mu = \mu \circ T^{-1}$  is the push forward of  $\mu$  by  $T$
- ▶ Such  $T$  is an **optimal transport map**



# Kantorovich's problem

## Kantorovich's problem

$$\min \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) \pi(dx, dy) : \pi \in \Pi(\mu, \nu)$$

- ▶  $\Pi(\mu, \nu)$ : probabilities on  $\mathcal{X} \times \mathcal{Y}$  with marginals  $\mu$  and  $\nu$

## Kernel formulation

$$\min \int_{\mathcal{X} \times \mathcal{Y}} c(x, y) (\mu \otimes \kappa)(dx, dy) : \kappa \in \mathcal{K}(\mu, \nu)$$

- ▶  $\mathcal{K}(\mu, \nu)$ : kernels  $\kappa$  satisfying  $\kappa_{\#}\mu := \int_{\mathcal{X}} \kappa_x \mu(dx) = \nu$

## Probabilistic formulation

$$\min \mathbb{E}[c(X, Y)] : X \stackrel{\text{law}}{\sim} \mu; Y \stackrel{\text{law}}{\sim} \nu$$

# Martingale transport

- ▶ **Martingale** optimal transport

$$\min \mathbb{E}[c(X, Y)] : X \stackrel{\text{law}}{\sim} \mu; Y \stackrel{\text{law}}{\sim} \nu; X = \mathbb{E}[Y|X]$$

- ▶ Equivalently

$$\min \int c(x, y)(\mu \otimes \kappa)(dx, dy) : \kappa \in \mathcal{K}(\mu, \nu); e[\kappa_x] = x \ (\mu\text{-a.e.})$$

where  $e$  is the mean of a probability

- ▶  $\mathcal{M}(\mu, \nu)$ : martingale transports (MT) in  $\Pi(\mu, \nu)$

# Martingale transport

- ▶ Motivated by (Asian) option pricing: sup/inf of

$$\left\{ \mathbb{E}[(X_1 + X_2 - K)_+] : X_1 \stackrel{\text{law}}{\sim} \mu_1; X_2 \stackrel{\text{law}}{\sim} \mu_2; \mathbb{E}[X_2|X_1] = X_1 \right\}$$

where  $K \in \mathbb{R}$  and  $\mu_1, \mu_2$  are calibrated from option prices

- ▶ On  $\mathbb{R}$ :
  - $c(x, y) = (x - y)^2$ : all MT have the same cost
  - $c(x, y) = h(x - y)$ : the first two derivatives do not matter, but the third does





# Monge Martingale transport

- ▶  $X = \mathbb{E}[Y|X]$  and  $Y$  is function of  $X \implies Y = X$
- ▶ MT cannot be Monge in the forward direction unless trivial
- ▶ Backward direction is possible:
  - $X = \mathbb{E}[Y|X]$  and  $X$  is function of  $Y$
  - Example:  $Y \stackrel{\text{law}}{\sim} U[-2, 2]$  and  $X = \text{sign}(Y)$
- ▶ Monge martingale transports (MMT; omit “backward”)
- ▶  $\mathcal{M}_M(\mu, \nu)$ : set of MMT
- ▶ MMT is useful for:
  - identifying worst-case probabilities
  - some settings of matching problems

# Strassen's theorem

- ▶ From now on  $\mathcal{X} = \mathcal{Y} = \mathbb{R}$
- ▶  $\mathcal{P}(\mathbb{R})$ : Borel probabilities on  $\mathbb{R}$  with finite first moment
- ▶  $\mu \leq_{\text{cx}} \nu$ :  $\int \phi d\mu \leq \int \phi d\nu$  for all convex  $\phi$

## Strassen's Theorem

*For  $\mu, \nu \in \mathcal{P}(\mathbb{R})$ ,  $\mu \leq_{\text{cx}} \nu$  if and only if  $\mathcal{M}(\mu, \nu)$  is non-empty.*

- ▶ Increasing in risk

Rothschild-Stiglitz'70 JET

# Monge martingale transport

## Theorem 1 (Existence)

Let  $\mu, \nu \in \mathcal{P}(\mathbb{R})$  satisfy  $\mu \leq_{\text{cx}} \nu$ . There exists  $\pi \in \mathcal{M}(\mu, \nu)$  and a Borel function  $h : \mathbb{R} \rightarrow \mathbb{R}$  such that  $\pi(T_{\text{rg}} \cup T_{\text{atom}}) = 1$ , where

- (i)  $T_{\text{rg}} = \{(h(y), y) : y \in \mathbb{R}\}$ ;
- (ii)  $T_{\text{atom}} = \{(x, y) : \nu(\{y\}) > 0\}$ .

In particular, if  $\nu$  is atomless,  $\pi$  is a Monge martingale transport.

$\nu$  atomless and  $\mu \leq_{\text{cx}} \nu \implies$  MMT exists

# Main idea of the proof

- ▶ We need the **left-curtain transport**  $\pi_{lc}$  Beiglöck/Juillet'16 AOP
- ▶ Write  $\mu \leq_E \nu$  for finite measures  $\mu, \nu$  with finite first moment if  $\int \phi d\mu \leq \int \phi d\nu$  for any nonnegative convex  $\phi$ 
  - if  $\mu(\mathbb{R}) = \nu(\mathbb{R})$ , this is  $\mu \leq_{cx} \nu$
  - if  $\mu \leq \nu$  (set-wise) then  $\mu \leq_E \nu$
- ▶ Given  $\mu \leq_E \nu$ , the **shadow**  $S^\nu(\mu)$  of  $\mu$  in  $\nu$  is defined as

$$S^\nu(\mu) = \min_{\leq_{cx}} \{ \eta : \mu \leq_{cx} \eta \leq \nu \},$$

- always well-posed
- ▶ Given  $\mu \leq_{cx} \nu$ , the **left-curtain (LC) transport**  $\pi_{lc} \in \mathcal{M}(\mu, \nu)$  is uniquely defined by the property that it transports  $\mu|_{(-\infty, x]}$  to its shadow  $S^\nu(\mu|_{(-\infty, x]})$  for every  $x \in \mathbb{R}$

# Main idea of the proof

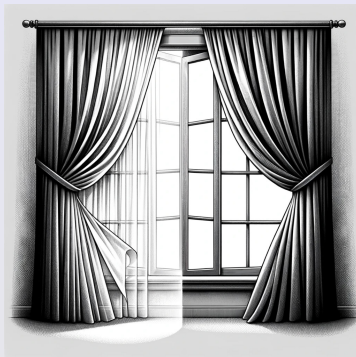
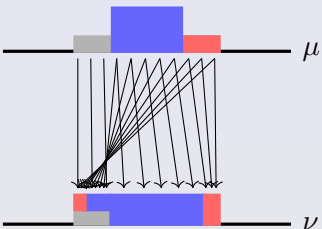


Figure: An example of the left-curtain transport (not MMT)

- ▶ The LC transport uniquely minimizes  $\int c d\pi$ :  $\pi \in \mathcal{M}(\mu, \nu)$  for  $c(x, y) = h(y - x)$  with  $h'$  strictly convex

# Main idea of the proof

- ▶ We will create a **barcode transport** by decomposing  $\mu$  and  $\nu$  into countably many mutually singular Monge parts
- ▶ Define the densities

$$d_\mu = \frac{d\mu}{d(\mu + \nu)} \quad \text{and} \quad d_\nu = \frac{d\nu}{d(\mu + \nu)}$$

The **barcode transport** is defined by a sequential construction:

- ▶ Take the part on  $\mathbb{R}$  with  $d_\mu \geq 1/2$  “ $d_\mu \geq d_\nu$ ”
- ▶ Apply the left-curtain transport on this part with its shadow
- ▶ Remove the matched parts from both  $\mu$  and  $\nu$
- ▶ Repeat on the rest
- ▶ This procedure converges

# Main idea of the proof

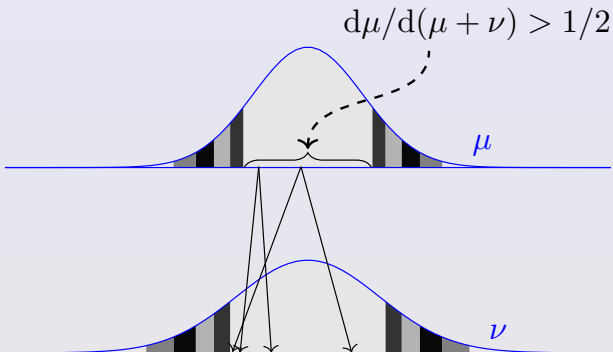


Figure: An example of the barcode transport between Gaussian marginals

# Main idea of the proof

## Proposition 1 (Structure of $\pi_{lc}$ )

Let  $\mu \leq_{cx} \nu$ . There exists a Borel function  $h : \mathbb{R} \rightarrow \mathbb{R}$  such that the LC transport  $\pi_{lc}$  satisfies  $\pi_{lc}(S_{rg} \cup S_{diag} \cup S_{atom}) = 1$ , where

- (i)  $S_{rg} = \{(h(y), y) : y \in \mathbb{R}\}$ ;
- (ii)  $S_{diag} = \{(x, x) : x \in \mathbb{R}\}$ ;
- (iii)  $S_{atom} = \{(x, y) : \nu(\{y\}) > 0\}$ .

If  $d_\mu \geq 1/2$   $\mu$ -a.e., then  $\pi_{lc}(S_{rg} \cup S_{atom}) = 1$ . In particular, if in addition  $\nu$  is atomless, then  $\pi_{lc} \in \mathcal{M}_M(\mu, \nu)$ .

$\nu$  atomless and  $d_\mu \geq 1/2 \implies \pi_{lc}$  is MMT



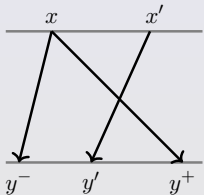
# Let us prove Proposition 1

**Lemma 1** (LC is left-monotone; Beiglböck/Juillet'16 AOP)

The LC transport  $\pi_{\text{lc}} \in \mathcal{M}(\mu, \nu)$  satisfies  $\pi_{\text{lc}}(\Gamma) = 1$  for some  $\Gamma \subseteq \mathbb{R} \times \mathbb{R}$  that is a *left-monotone* set; i.e.,

$(x, y^-), (x, y^+), (x', y') \in \Gamma$  with  $x < x'$ : it *forbids*  $y^- < y' < y^+$ .

Moreover,  $\pi_{\text{lc}} \in \mathcal{M}(\mu, \nu)$  is uniquely characterized by that property.



**Figure:** Forbidden configuration for left-monotonicity: the legs of a point  $x'$  cannot step into the legs of another point  $x$  to the left of  $x'$

# Let us prove Proposition 1

## Lemma 2 (Support of LC; Beiglböck/Juillet'16 AOP)

*There exist two functions  $T_d, T_u : \mathbb{R} \rightarrow \mathbb{R}$  such that*

*$\pi_{lc}(R_{\text{legs}} \cup R_{\text{atom}}) = 1$ , where*

- (a)  $R_{\text{legs}}$  is the union of the graphs of  $T_d, T_u$  over the first marginal;*
- (b)  $R_{\text{atom}} = \{(x, y) : \mu(\{x\}) > 0\}$ .*

# Let us prove Proposition 1

Denote by  $\kappa_x(dy)$  the disintegration of  $\pi_{1c}$  by  $\mu$ :

$$\pi_{1c}(dx, dy) = \mu(dx) \otimes \kappa_x(dy)$$

## Lemma 3

We have  $d_\mu \leq d_\nu$   $\mu$ -a.e. on  $\{x \in \mathbb{R} : \kappa_x = \delta_x\}$ .



Figure: For the left-curtain transport,  $d_\mu \leq d_\nu$   $\mu$ -a.e. on  $\{x \in \mathbb{R} : \kappa_x = \delta_x\}$

# Let us prove Proposition 1

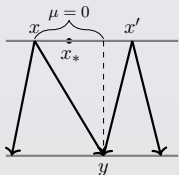
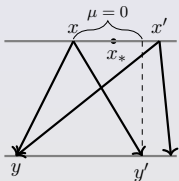
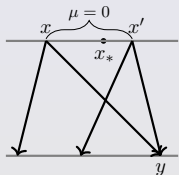
Let us prove

$\pi_{1c}$  is supported on  $S_{\text{rg}}$  if  $d_\mu \geq d_\nu$   $\mu$ -a.e. and  $\nu$  is atomless

- ▶  $\nu$  is atomless  $\implies$  if  $\mu(\{x\}) > 0$  then  $\kappa_x(\{x\}) = 0$
- ▶ Lemma 2  $\implies$   $\mu$ -a.e. if  $\mu(\{x\}) = 0$  then either  $\kappa_x = \delta_x$  or  $\kappa_x$  is supported on two points  $T_d(x) < x < T_u(x)$ .
- ▶ Lemma 3  $\implies$   $\mu$ -a.e. if  $\kappa_x(\{x\}) > 0$  then  $d_\mu(x) \leq d_\nu(x)$
- ▶  $d_\mu \geq d_\nu$   $\mu$ -a.e.  $\implies$   $\mu$ -a.e. if  $\kappa_x(\{x\}) > 0$  then  $d_\mu(x) = d_\nu(x)$
- ▶  $\pi_{1c}$  is id on  $S := \{x \in \mathbb{R} : \kappa_x(\{x\}) > 0\}$  and Monge
- ▶ Safely remove  $S$  and assume  $\kappa_x(\{x\}) = 0$   $\mu$ -a.e. to continue

# Let us prove Proposition 1

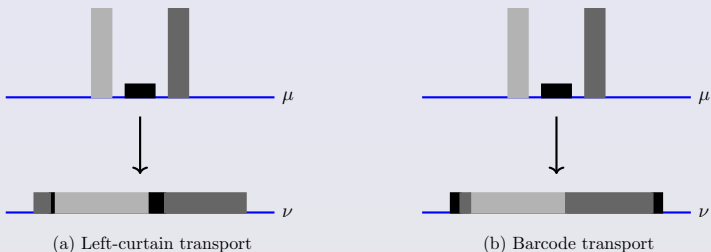
- ▶ Lemma 1  $\implies \pi_{1c}$  is supported on a left-monotone set  $\Gamma$
- ▶  $\Gamma \subseteq \text{supp}(\mu) \times \text{supp}(\nu)$  and  $\Gamma \subseteq R_{\text{legs}} \cup R_{\text{atom}}$
- ▶ Previous step  $\implies \Gamma \cap \{(x, x) : x \in \mathbb{R}\} = \emptyset$
- ▶ Suppose  $(x, y), (x', y) \in \Gamma$  with  $y \notin \{x, x'\}$ 
  - $x < y < x' \implies \mu((x, y)) = 0$
  - $y < x < x' \implies \mu((x, y' \wedge x')) = 0$  for  $y'$  the right leg of  $x$
  - $x < x' < y \implies \mu((x, x')) = 0$

(a) The case  $x < y < x'$ (b) The case  $y < x < x'$ (c) The case  $x < x' < y$

# Let us prove Proposition 1

- ▶ As  $\text{supp}(\mu)$  is closed, its complement can be written as a countable disjoint union of open intervals
- ▶ Each pair of  $(x, y), (x', y) \in \Gamma$  with  $x \neq x'$  corresponds to an endpoint of one of the open intervals
- ▶ There are at most countably many  $y$  that do this
- ▶  $\nu$  atomless  $\implies$  such  $y$  has  $\nu$ -measure 0  $\implies \pi_{1c}(S_{\text{rg}}) = 1$
- ▶ Verify that the transport map in  $S_{\text{rg}}$  can be required Borel

# Main idea of the proof



**Figure:** It is possible that the left-curtain transport is MMT, but it is not necessarily equal to the barcode transport





# Strassen's theorem on random variables

Consider the statement

$$X \leq_{\text{cx}} Y \iff X \stackrel{\text{law}}{=} \mathbb{E}[Y|\mathcal{G}] \text{ for some } \sigma\text{-field } \mathcal{G}?$$

- ▶ Jensen's inequality gives  $\Leftarrow$
- ▶ Is  $\Rightarrow$  true?

## Example 1

Let  $\Omega = \{1, 2\}$ , uniform probability,  $X = (1, 2)$  and  $Y = (0, 3)$

- ▶  $X \leq_{\text{cx}} Y$  but  $X \stackrel{\text{law}}{=} \mathbb{E}[Y|\mathcal{G}]$  does not hold for any  $\mathcal{G}$

# Strassen's theorem on random variables

Reverse Jensen's:

$$X \leq_{\text{cx}} Y \implies X \stackrel{\text{law}}{=} \mathbb{E}[Y|\mathcal{G}] \text{ for some } \sigma\text{-field } \mathcal{G} \quad (\star)$$

What about an atomless space?

- ▶ If  $(\star)$  holds true for  $\sigma(Y) = \mathcal{F}$  then necessarily

$$X \leq_{\text{cx}} Y \implies X \stackrel{\text{law}}{=} f(Y) \text{ for some measurable } f$$

- ▶  $(\star)$  requires a Monge martingale transport to exist!

# A refinement of Strassen's theorem

## Theorem 2 (Refinement of Strassen's Theorem)

For random variables  $X$  and  $Y$  on  $(\Omega, \mathcal{F}, \mathbb{P})$  that is *atomless*,  
 $X \leq_{\text{cx}} Y$  if and only if  $X \stackrel{\text{law}}{=} \mathbb{E}[Y|\mathcal{G}]$  for some  $\sigma$ -field  $\mathcal{G} \subseteq \mathcal{F}$ .

- ▶ It is a refinement to the following version

## Strassen's in the form of Theorem 3.A.4 of Shaked/Shantikumar'07

Two random variables  $X$  and  $Y$  satisfy  $X \leq_{\text{cx}} Y$  if and only if there exist random variables  $X', Y'$  on an atomless probability space  $(\Omega, \mathcal{F}, \mathbb{P})$  satisfying  $X' \stackrel{\text{law}}{=} X$ ,  $Y' \stackrel{\text{law}}{=} Y$  and  $X' = \mathbb{E}[Y'|X']$ .



# Denseness

## Theorem 3 (MMTs are dense)

Let  $\mu \leq_{\text{cx}} \nu$  with  $\nu$  atomless. Then  $\mathcal{M}_M(\mu, \nu)$  is weakly dense in  $\mathcal{M}(\mu, \nu)$ . If  $\mu$  is discrete, it is also dense for the  $\infty$ -Wasserstein topology.

## Corollary 1 (Optimal MT cost = optimal MMT cost)

Let  $\mu \leq_{\text{cx}} \nu$  with  $\nu$  atomless. If  $c : \mathbb{R}^2 \rightarrow \mathbb{R}$  is continuous with  $|c(x, y)| \leq a(x) + b(y)$  for some  $a \in L^1(\mu)$  and  $b \in L^1(\nu)$ , then

$$\inf_{\pi \in \mathcal{M}_M(\mu, \nu)} \int c d\pi = \inf_{\pi \in \mathcal{M}(\mu, \nu)} \int c d\pi.$$

# Denseness

## Lemma 4

Let  $\nu \in \mathcal{P}(\mathbb{R})$  be atomless. Given any decomposition  $\nu = \sum_{i=1}^{\infty} \nu_i$  of  $\nu$ , there exist mutually singular  $\hat{\nu}_i$ ,  $i \in \mathbb{N}$  such that  $\nu = \sum_{i=1}^{\infty} \hat{\nu}_i$  and  $\nu_1 \leq_{\text{cx}} \hat{\nu}_1$  and  $\text{bary}(\nu_i) = \text{bary}(\hat{\nu}_i)$  for  $i \geq 2$ .

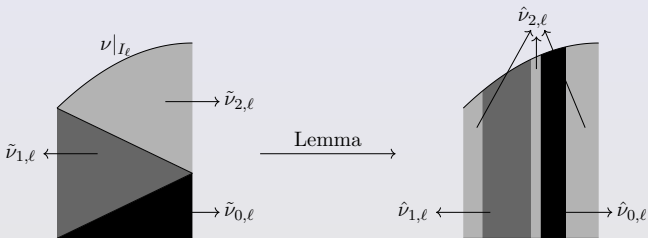


Figure: An illustration of the main idea to prove Theorem 3

# Uniqueness

## Theorem 4 (Uniqueness)

Let  $\mu \leq_{\text{cx}} \nu$  with  $\nu$  **atomless**. The following are equivalent:

- (i) The MT from  $\mu$  to  $\nu$  is unique.
- (ii) The MMT from  $\mu$  to  $\nu$  is unique.
- (iii) Let  $\mu_a := \sum_{j \in \mathbb{N}} a_j \delta_{x_j}$  be the atomic part of  $\mu$ , where  $\{x_j\}_{j \in \mathbb{N}}$  are distinct. Then the shadows  $S^\nu(a_j \delta_{x_j})$ ,  $j \in \mathbb{N}$  are mutually singular and  $\mu - \mu_a = \nu - \sum_{j \in \mathbb{N}} S^\nu(a_j \delta_{x_j})$ .

►  $\mu$  and  $\nu$  are both atomless: uniqueness  $\iff \mu = \nu$

# Uniqueness

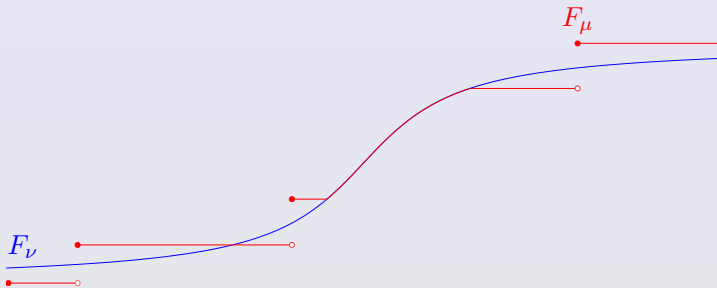


Figure: Distribution functions of  $\mu, \nu$  where the MMT (and MT) from  $\mu$  to  $\nu$  is unique



# Uniqueness

## Example 2 (MT exists; MMT does not)

Let  $\mu$  and  $\nu$  be two-point distributions satisfying  $\mu \leq_{\text{cx}} \nu$ . Then there is a unique MT but there is no MMT unless  $\mu = \nu$ .

- ▶ In general, if  $\mu, \nu$  are discrete and  $\text{card}(\cdot)$  denotes the cardinality of the support, the existence of an MMT implies  $(2 \text{card}((\mu - \nu)_+)) \vee \text{card}(\mu) \leq \text{card}(\nu)$

## Example 3 (MMT is unique; MT is not)

Let  $\mu$  be uniform on  $\{2, 5\}$  and  $\nu$  be uniform on  $\{0, 3, 4, 7\}$ . The unique MMT is given by transporting  $\{2\}$  to  $\{0, 4\}$  and  $\{5\}$  to  $\{3, 7\}$ , while it is easy to see that there exist many MTs.

# Optimizer

## Proposition 2

Consider  $\mu \leq_{\text{cx}} \nu$  with  $\nu$  atomless. For any strictly convex  $f, g : \mathbb{R} \rightarrow \mathbb{R}$ ,

$$\mathcal{M}_M(\mu, \nu) = \arg \min_{(X, Y) \in \Pi(\mu, \nu)} \mathbb{E}[f(\mathbb{E}[Y|X] - X) - g(\mathbb{E}[X|Y])]. \quad (\diamond)$$

- ▶  $f(x) = g(x) = x^2$ :  $(\diamond)$  is equivalent to

$$\mathbb{E}[\mathbb{E}[Y|X]^2 - \mathbb{E}[X|Y]^2 - 2\mathbb{E}[XY]]$$

- ▶ This cost is not symmetric in  $X$  and  $Y$
- ▶ The term  $-2\mathbb{E}[XY]$  is essential:  $\mathcal{M}_M(\mu, \nu)$  **does not** minimize  $\mathbb{E}[\mathbb{E}[Y|X]^2 - \mathbb{E}[X|Y]^2]$  unless  $X$  is a constant

# Backward deterministic martingales

## Definition 1

A stochastic process  $(X_n)_{n \in \mathbb{N}}$  is **backward deterministic** if  $(X_j)_{j=1}^n$  is  $\sigma(X_n)$ -measurable for all  $n \in \mathbb{N}$ .

- ▶  $\sigma(X_n)$  is non-decreasing in  $n$ .
- ▶ A backward deterministic process is Markovian
- ▶ Perfect memory: its time- $n$  value records all its history up to time  $n$

# Backward deterministic martingales

## Corollary 2

Given any martingale  $(Y_n)_{n \in \mathbb{N}}$  with atomless marginals, there exists a backward deterministic martingale  $(X_n)_{n \in \mathbb{N}}$  such that  $X_n \stackrel{\text{law}}{=} Y_n$  for all  $n \in \mathbb{N}$ .

- ▶ Although rare, the class of backward deterministic martingale is surprisingly “rich”

- ① Martingale transport
- ② Monge martingale transport maps
- ③ A refinement of Strassen's theorem
- ④ Other results
- ⑤ Open questions

# Open questions

## Backward deterministic fake Brownian motions

- ▶ Let  $(W_t)_{t \in [0, T]}$  be a martingale with marginal distribution  $W_t \stackrel{\text{law}}{=} N(0, t)$
- ▶ Fake Brownian motions
  - (continuous path) [Beiglböck/Lowther/Pammer/Schachermayer'23 FS](#)
- ▶ Does there exist such  $W$  that is also **backward deterministic**?
  - $(W_s)_{s \in [0, t]}$  is  $\sigma(W_t)$ -measurable for each  $t \in [0, T]$

# Open questions

Generalization to  $\mathbb{R}^d$

- ▶ Seems highly challenging

## Conjecture 1

*Let  $\mu, \nu$  be probability measures on  $\mathbb{R}^d$  satisfying  $\mu \leq_{cx} \nu$ , and  $\{\nu_x : x \in \mathbb{R}^d\}$  is a decomposition into irreducible components of  $(\mu, \nu)$ . Suppose that  $\nu_x$  is atomless for  $\mu$ -a.e.  $x \in \mathbb{R}^d$ . Then  $\mathcal{M}_M(\mu, \nu)$  is non-empty and weakly dense in  $\mathcal{M}(\mu, \nu)$ . If  $\mu$  is discrete, it is also dense for the  $\infty$ -Wasserstein topology.*

# Open questions

## Monge supermartingale/directional transport

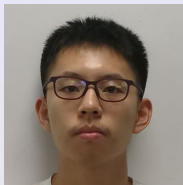
- ▶ Existence of Monge directional transport: **clear** Nutz-W.'22 AAP
- ▶ Existence of Monge supermartingale transport: **unclear**
- ▶ Denseness: **unclear**



# Thank you for your kind attention



Marcel Nutz  
(Columbia)



Zhenyuan Zhang  
(Stanford)

**Takeaway 1.** On an atomless probability space,

$$X \leq_{cx} Y \iff X \stackrel{\text{law}}{=} \mathbb{E}[Y|\mathcal{G}] \text{ for some } \sigma\text{-field } \mathcal{G}$$

**Takeaway 2.** For atomless  $\nu$  and continuous and “non-exploding”  $c$ ,

optimal MT cost = optimal MMT cost