

# MARTINGALE OPTIMAL TRANSPORT

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We are given

- ▶ two closed, convex sets  $X, Y \subset \mathbb{R}^d$ ,
- ▶ two probability measures  $\mu, \nu$  defined on Borel sets  $\mathcal{B}(X)$  and  $\mathcal{B}(Y)$ , respectively.

A **transport map** is a function  $T : X \rightarrow Y$ , satisfying

$$\nu = T_{\#}\mu, \quad \text{i.e.,} \quad \nu(A) = \mu(\{x \in X : T(x) \in A\}), \quad \forall A \in \mathcal{B}(Y).$$

In general, there may be no such maps.

Given the above structure and a **reward function**

$$\xi : \Omega := X \times Y \rightarrow \mathbb{R},$$

the **optimal transport** problem is to find the transport  $T$  that maximizes

$$J(T) := \int_X \xi(x, T(x)) \mu(dx).$$

This is a hard **nonlinear** problem!

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For a transport map, define a **probability measure on  $\Omega$**  by

$$\mathbb{P}(C) = \mu(\{x \in X : (x, T(x)) \in C\}), \quad C \in \mathcal{B}(\Omega).$$

Then, the definition of  $\mathbb{P}$  and  $\nu = T_{\#}\mu$  imply that

$$\mathbb{P}(A \times Y) = \mu(A), \quad A \in \mathcal{B}(X), \quad \text{and} \quad \mathbb{P}(X \times B) = \nu(B), \quad B \in \mathcal{B}(Y).$$

Let  $\mathcal{M}(\mu, \nu)$  is the set of all probability measures  $\mathbb{P}$  defined on Borel sets  $\mathcal{B}(\Omega)$  satisfying the above **marginal constraints**.



- ▶ For a given transport map  $T$  there is a generalized transport  $\mathbb{P}$ . Indeed, define  $\mathbb{P}$  by

$$\mathbb{P}(dx, dy) = \delta_{T(x)}(dy) \mu(dy).$$

Then,  $\mathbb{P} \in \mathcal{M}(\mu, \nu)$ .

- ▶ However, **the converse is not true**. There are many more generalized transports.
- ▶ Also,  $\mathcal{M}(\mu, \nu)$  is always nonempty.

$$\mu \times \nu \in \mathcal{M}(\mu, \nu).$$

The generalized map corresponds to the transport map  $T$  is

$$\mathbb{P}(dx, dy) = \delta_{T(x)}(dy) \mu(dx).$$

The objective functional is

$$\begin{aligned} J(\mathbb{P}) &= \int_{\mathcal{X}} \xi(x, T(x)) \mu(dx) \\ &= \int_{\Omega} \xi(x, y) \delta_{T(x)}(dy) \mu(dx) \\ &= \int_{\Omega} \xi(x, y) \mathbb{P}(dx, dy) \\ &= \mathbb{E}_{\mathbb{P}}[\xi]. \end{aligned}$$

The relaxation of the Monge problem is

$$\mathcal{P}(\xi) := \sup_{\mathbb{P} \in \mathcal{M}(\mu, \nu)} \mathbb{E}_{\mathbb{P}}[\xi].$$

- ▶ This is a *linear* problem and there always **exists an optimizer**.
- ▶ We are **given the marginals** of the measure  $\mathbb{P}$  and the optimal transport problem is to find the “**best**” **correlation structure**.
- ▶ It has many natural applications in risk management.

For  $h \in \mathcal{L}^1(X, \mu)$ ,  $g \in \mathcal{L}^1(Y, \nu)$ , set

$$(h \oplus g)(z) := h(x) + g(y), \quad z = (x, y) \in \Omega = X \times Y,$$

$$\mathcal{L}(h, g) := \int_X h d\mu + \int_Y g d\nu.$$

Then, for every  $\mathbb{P} \in \mathcal{M}(\mu, \nu)$ ,

$$\mathbb{E}_{\mathbb{P}}[h \oplus g] = \mathcal{L}(h, g).$$

This suggests that the dual is given by

$$\mathcal{D}(\xi) := \inf \{ \mathcal{L}(h, g) : h \oplus g \geq \xi \}.$$

Consider the subspace of  $C_b(\Omega)$ ,

$$\mathcal{A} := \{ f \in C_b(\Omega) : \exists g, h \text{ such that } f = h \oplus g - \mathcal{L}(h, g) \}.$$

Then,  $\mathcal{M}(\mu, \nu)$  is the set of all Radon measures that annihilate the set  $\mathcal{A}$ , i.e.

$$\mathbb{P} \in \mathcal{M}(\mu, \nu), \quad \Leftrightarrow \quad \mathbb{E}_{\mathbb{P}}[f] = 0, \quad \forall f \in \mathcal{A}.$$

This is the reason behind the duality.

## Theorem (Kantorovich 1940, Kellerer 1984)

For any bounded, Borel measurable  $\xi$ ,

$$\mathcal{P}(\xi) = \mathcal{D}(\xi).$$

- ▶ This follows immediately from Fenchel-Moreau theorem for **continuous** functions  $\xi$ .
- ▶ A straightforward approximation argument proves it for all **upper-semicontinuous**  $\xi$ 's.
- ▶ General result is due to Kellerer.
- ▶ The inequality  $\mathcal{D} \geq \mathcal{P}$  is easy.

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Suppose  $X = \{x_1, \dots, x_n\}$ ,  $Y = \{y_1, \dots, y_m\}$ . Then, the primal problem (in Kantorovich formulation) is

$$\text{maximize } \sum_{i,j} \xi(x_i, y_j) p_{ij}$$

over all  $p_{ij}$ 's satisfying

$$\mu_i = \sum_j p_{ij}, \quad i = 1, \dots, n,$$

$$\nu_j = \sum_i p_{ij}, \quad j = 1, \dots, m,$$

$$p_{ij} \geq 0, \quad i = 1, \dots, n, j = 1, \dots, m.$$



Let  $h = (h_1, \dots, h_n)$  be the dual variables for the constraint :

$$\mu_i = \sum_j p_{ij}, \quad i = 1, \dots, n,$$

Let  $g = (g_1, \dots, g_m)$  be the dual variables for the constraint :

$$\nu_j = \sum_i p_{ij}, \quad j = 1, \dots, m.$$

Then, the classical duality for linear program implies that the dual objective is to **minimize**

$$\sum_i h_i \mu_i + \sum_j g_j \nu_j.$$

Since the primal variables  $p_{ij} \geq 0$ , the dual constraint is

$$h_i + g_j \geq \xi(x_i, y_j), \quad \forall i, j.$$

A generalized transport  $q_{ij}$  is a regular transport map if

$$q_{ij} = \begin{cases} \mu_i & \text{if } j = T(i), \\ 0 & \text{if } j \neq T(i). \end{cases}$$

Then,  $T$  is a transport map if

$$\nu_j = \sum_{\{i : T(i)=j\}} \mu_i.$$

This is a difficult constraint and may not be possible. However, the generalized transport always exists :

$$\text{set } q_{ij} = \mu_i \nu_j, \quad \text{then } q \in \mathcal{M}(\mu, \nu).$$

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We restrict the class of generalized transports to be martingale as well. Indeed, let  $\mathcal{Q}(\mu, \nu)$  be the set of all  $\mathbb{Q} \in \mathcal{M}(\mu, \nu)$  satisfying

$$\mathbb{E}_{\mathbb{Q}}[\gamma(x) \cdot (y - x)] = 0,$$

for all bounded, Borel measurable  $\gamma$ . Strassen (1965) proved that  $\mathcal{Q}(\mu, \nu)$  is non-empty if and only if  $\mu$  and  $\nu$  are in convex order.

The primal problem is defined as

$$\mathcal{P}(\xi) := \sup_{\mathbb{Q} \in \mathcal{Q}(\mu, \nu)} \mathbb{E}_{\mathbb{Q}}[\xi].$$

Suppose there exists  $\mathbb{Q} \in \mathcal{Q}(\mu, \nu)$  and  $\phi : \mathbb{R}^d \rightarrow \mathbb{R}$  is convex. Recall that  $X, Y \subset \mathbb{R}^d$ . By conditioning (and abuse of notation)

$$\mathbb{E}_{\mathbb{Q}}[\phi(y)] = \mathbb{E}_{\mathbb{Q}}[\mathbb{E}_{\mathbb{Q}}[\phi(y) | x]].$$

Then, by Jensen's inequality and the fact that  $\mathbb{P}$  is a martingale measure,

$$\mathbb{E}_{\mathbb{Q}}[\phi(y) | x] \geq \phi(\mathbb{E}_{\mathbb{Q}}[y | x]) = \phi(x).$$

Combining and using marginals,

$$\int_Y \phi d\nu = \mathbb{E}_{\mathbb{Q}}[\phi(y)] \geq \mathbb{E}_{\mathbb{Q}}[\phi(x)] = \int_X \phi d\mu.$$

So  $\mu$  and  $\nu$  must be in convex order.

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Marginal constraints imply that for every  $\mathbb{Q} \in \mathcal{Q}(\mu, \nu)$ ,

$$\mathbb{E}_{\mathbb{Q}}[h \oplus g] = \mathcal{L}(h, g) = \int h d\mu + \int g d\nu.$$

Martingale condition is

$$\mathbb{E}_{\mathbb{Q}}[\gamma(x) \cdot (y - x)] = 0, \quad \gamma \in \mathcal{L}^{\infty}(X).$$

Combining all above we conclude that for all  $\mathbb{Q} \in \mathcal{Q}(\mu, \nu)$ ,

$$\mathbb{E}_{\mathbb{Q}}[H(h, g, \gamma)] = \mathcal{L}(h, g)$$

where

$$H(h, g, \gamma)(z) = (h \oplus g)(z) + \gamma(x) \cdot (y - x).$$



$$\mathbb{E}_{\mathbb{Q}} [H(h, g, \gamma)] = \mathcal{L}(h, g), \quad \forall \mathbb{Q} \in \mathcal{Q}(\mu, \nu),$$

$$H(h, g, \gamma)(z) = h(x) + g(y) + \gamma(x) \cdot (y - x).$$

This suggests that the dual is

$$\mathcal{D}(\xi) := \inf \{ \mathcal{L}(h, g) : H(h, g, \gamma) \geq \xi \}.$$

The main difference is the new term  $\gamma(x) \cdot (y - x)$ . This is due to the **martingale constraint** imposed on the probability measures  $\mathbb{Q}$ .

## Theorem

Suppose  $\mu$  and  $\nu$  are in convex order, i.e.,

$$\int \phi d\mu \leq \int \phi d\nu,$$

for every convex  $\phi$ . Then,

$$\mathcal{P}(\xi) = \mathcal{D}(\xi),$$

for every bounded and continuous  $\xi$ .

This follows directly from the Fenchel-Moreau theorem.

Let  $C_b(\Omega)$  be the set of all bounded, **continuous** functions.

Then,

$$\mathcal{D} : C_b(\Omega) \rightarrow \mathbb{R}.$$

It is clear that

- ▶  $\mathcal{D}$  is convex;
- ▶  $\mathcal{D}$  is Lipschitz in the sup-norm;
- ▶ For any  $\lambda > 0$  and  $\xi \in C_b(\Omega)$ ,  $\mathcal{D}(\lambda\xi) = \lambda\mathcal{D}(\xi)$ .

▶

$$|\mathcal{D}(\xi)| \leq \|\xi\|_\infty.$$

So  $\mathcal{D}$  is a proper convex function and Fenchel-Moreau theorem applies.

Fenchel-Moreau implies that,

$$\mathcal{D}(\xi) = \sup_{\mathbb{Q} \in (C_b(\Omega))^*} \{ \mathbb{Q}(\xi) - \mathcal{D}^*(\mathbb{Q}) \},$$

$$\mathcal{D}^*(\mathbb{Q}) = \sup_{\xi \in C_b(\Omega)} \{ \mathbb{Q}(\xi) - \mathcal{D}(\xi) \}.$$

- ▶  $\mathcal{D}^*(\mathbb{Q}) = 0$  if  $\mathbb{Q}$  satisfies  $\mathbb{Q}(H(h, g, \gamma)) = 0$  for all  $h, g, \gamma$ . And  $\mathcal{D}^*(\mathbb{Q}) = +\infty$ , otherwise. (will be discussed later).
- ▶ Marginal constraints imply that any  $\mathbb{Q} \in (C_b(\Omega))^*$  satisfying above is a probability measure.
- ▶ Hence  $\mathcal{D}^*(\mathbb{Q}) = 0$  if and only if  $\mathbb{Q} \in \mathcal{Q}(\mu, \nu)$ .

$$\mathcal{D}^*(Q) = \sup_{\xi \in C_b(\Omega)} \{ Q(\xi) - \mathcal{D}(\xi) \}.$$

- ▶  $\mathcal{D}(0) = 0$  implies that  $\mathcal{D}^* \geq 0$ .
- ▶ If  $Q(\xi) > \mathcal{D}(\xi)$  for some  $\xi$ , by scaling we conclude that  $\mathcal{D}^*(Q) = +\infty$ .
- ▶ Hence,  $\mathcal{D}^*(Q) = 0$  iff  $Q(\xi) \leq \mathcal{D}(\xi)$  for every  $\xi$ . And the definition of  $\mathcal{D}$  imply that  $Q(H(h, g, \gamma)) = 0$ .

Duality **does not hold** for general  $\xi$ . Indeed, consider the example :

- ▶  $X = Y = [0, 1]$ ,
- ▶  $\mu$  and  $\nu$  are the Lebesgue measures.
- ▶ Then,  $\mathcal{Q}(\mu, \nu) = \{\mathbb{Q}^*\}$  where  $\mathbb{Q}^*$  is the uniform distribution on the diagonal;  $T(x) = x$  is the only transport map.
- ▶ Set  $\xi^*$  to be zero on the diagonal and one off the diagonal.
- ▶ It is clear that  $\mathcal{P}(\xi^*) = 0$ .
- ▶ However,  $\mathcal{D}(\xi^*) = 1$ .

We modify the dual slightly by requiring the inequality to hold quasi-surely (q.s., in short) :

We say that  $\eta \geq \xi$ ,  $\mathcal{Q}$ -quasi surely if

$$\mathbb{Q}(\{\eta < \xi\}) = 0, \quad \forall \mathbb{Q} \in \mathcal{Q}(\mu, \nu).$$

Then, we define the dual by

$$\hat{\mathcal{D}}(\xi) := \inf \{ \mathcal{L}(h, g) : H(h, g, \gamma) \geq \xi, \mathcal{Q} - \text{q.s.} \}.$$

**Theorem (Beiglböck, Nutz & Touzi, 2015)**

$$\mathcal{P}(\xi) = \hat{\mathcal{D}}(\xi), \quad \forall \xi \in \mathcal{L}^\infty(\Omega).$$

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- ▶ Trading is only allowed in two future time points,  $0 < t_1 < t_2$ ;
- ▶ We can dynamically trade the stock whose future values will be  $S_1, S_2$ ;
- ▶ No assumptions on  $S_1, S_2$  are made but  $S_0$  is given;
- ▶ We can also now buy any option of the form  $h(S_1)$  and  $h(S_2)$  for a known price of  $\int h d\mu$ , respectively  $\int g d\nu$ . Here  $g, h$  are arbitrary and can be chosen by the investor. But probability measures  $\nu$  and  $\mu$  are given.
- ▶ We want to super-replicate a claim  $\xi(S_1, S_2)$ .

Mathematically, we want to minimize (with zero interest rate) the prices of the options  $h(\mathbb{S}_1)$  and  $g(\mathbb{S}_2)$ , i.e.,

$$\text{minimize } \left[ \int h(\mathbb{S}_1) \mu(d\mathbb{S}_1) + \int g(\mathbb{S}_2) \nu(d\mathbb{S}_2) \right],$$

over all options  $(g, h)$  that together with a dynamic trading strategy  $\theta_0, \theta_1(\mathbb{S}_1)$ , dominate the claim  $\xi(\mathbb{S}_1, \mathbb{S}_2)$ .

$\theta_0$  = shares held today,

$\theta_1(\mathbb{S}_1)$  = shares will be bought at time  $t_1$ , and this decision will depend on the value of the stock at time  $t_1$ .

More compactly (with  $S_0$  given),

$$\text{minimize } \left[ \int h \mu + \int g \nu \right],$$

over all  $h \in L^1(\mathbb{R}^d, \mu)$ ,  $g \in L^1(\mathbb{R}^d, \nu)$  so that there exists  $\theta_0 \in \mathbb{R}^d$ ,  $\theta_1 : \mathbb{R}^d \rightarrow \mathbb{R}^d$  satisfying,

$$\theta_0 \cdot (S_1 - S_0) + \theta_1(S_1) \cdot (S_2 - S_1) + h(S_1) + g(S_2) \geq \xi(S_1, S_2),$$

for all  $S_1, S_2 \in \mathbb{R}^d$ .

We make the identification,

$$x = \mathbb{S}_1, \quad y = \mathbb{S}_2, \quad H(x) = h(x) + \theta_0 \cdot (x - \mathbb{S}_0), \quad G(y) = g(y).$$

The problem is to minimize  $\int Hd\mu + \int Gd\nu$ , over all  $(G, H)$  so that

$$H(x) + G(y) + \theta_1(x) \cdot (y - x) \geq \xi(x, y),$$

for some  $\theta_1$ , for all  $(x, y)$ . The term  $\theta_1(x) \cdot (y - x)$  in the duality constraint is new.

The dual is given by

$$\text{maximize } \mathbb{E}_{\mathbb{Q}} [c(S_1, S_0)],$$

over all martingale measures  $\mathbb{Q} \in \mathcal{Q}(\nu, \mu)$ , i.e., for  $\Omega := \mathbb{R}^2$ , the canonical process

$$S_0((\omega_1, \omega_2)) = S_0, \quad S_1((\omega_1, \omega_2)) = \omega_1, \quad S_2((\omega_1, \omega_2)) = \omega_2,$$

with the canonical filtration is a  $\mathbb{Q}$  martingale, i.e.,

$$\mathbb{E}[S_2 | S_1] = S_1.$$

Assumptions are  $\int y \mu(dy) = \int x \nu(dx) = S_0$  and that they are in convex order. So that the set of martingale measures satisfying the constraint is non-empty.

Suppose that for some  $G, H$ ,

$$H(S_1) + G(S_2) + \theta_1(S_1) \cdot (S_2 - S_1) \geq \xi(S_1, S_2),$$

for some  $\theta_1$ , and for all  $(S_1, S_2)$ . Let  $\mathbb{Q} \in \mathcal{Q}(\nu, \mu)$  be a martingale measure. Then,

$$\mathbb{E}_{\mathbb{Q}} [\theta_1(S_1) \cdot (S_2 - S_1)] = 0,$$

and the marginals imply that

$$\mathbb{E}_{\mathbb{Q}} [H(S_1)] = \int H d\mu, \quad \mathbb{E}_{\mathbb{Q}} [G(S_2)] = \int G d\nu.$$

So if we integrate the first inequality with respect to  $\mathbb{Q}$ ,

$$\int H d\mu + \int G d\nu \geq \mathbb{E}_{\mathbb{Q}} [\xi(S_1, S_2)].$$

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Suppose that all right-continuous processes with left limits (namely càdlàg processes) are possible stock price processes :

$$\Omega := \mathbb{D} := \{S : [0, T] \rightarrow \mathbb{R}_+^d \mid S_0 = (1, \dots, 1), \text{ càdlàg} \}.$$

All **European options**  $g(S_T)$  maturing **only at the final time** are traded. Its price is assumed to be

$$\mathcal{L}(g) = \int_{\mathbb{R}_+^d} g d\mu, \quad \forall g \in \mathbb{L}^1(\mathbb{R}_+, \mu).$$

Technically, we assume that

$$\int_{\mathbb{R}_+} x_k \mu(dx) = 1, \quad \forall k = 1, \dots, d.$$

This allows for martingale measures.

A general European path-dependent claim,  $\xi$ , is given,

$$\xi = G(\mathbb{S}), \quad \text{where } G : \mathbb{D} \rightarrow \mathbb{R}.$$

The minimal super-replication cost also provides the upper bound of the price interval and is given by

$$\mathcal{D}(G) := \inf \left\{ \int g d\mu \mid \exists \gamma \text{ admissible such that} \right. \\ \left. \int_{[0,T]} \gamma_t(\mathbb{S}) d\mathbb{S}_t + g(\mathbb{S}_T) \geq G(\mathbb{S}), \forall \mathbb{S} \in \mathbb{D} \right\}.$$

Clearly, admissibility restrictions on  $\gamma$  are needed, such as a lower bound, predictability.

Dual refers to a probabilistic structure, which we now introduce.  $\Omega := \mathbb{D}$  is as before and let  $\mathbb{S}$  be the **canonical process** and  $\mathcal{F}_t$  be the **canonical filtration**.

As before  $\mathcal{Q}(\mu)$  is the set of all martingale measures such that the probability distribution of  $\mathbb{S}_T$  under  $\mathbb{Q}$  is  $\mu$ , i.e.,

$$\mathbb{Q}(\mathbb{S}_T \in B) = \mu(B).$$

Theorem (DS : continuous (2012) PTRF - càdlàg (2015) SPA)

Assume that  $G$  is bounded and uniformly continuous with respect to the Skorokhod metric. Then,

$$\mathcal{D}(G) = \sup_{\mathbb{Q} \in \mathcal{Q}(\mu)} \mathbb{E}_{\mathbb{Q}}[G(S)].$$

We could also consider finitely many in between marginals.

Continuity of the claim  $G$  is essential. For general claims, one needs the quasi-sure set-up that will be discussed later.

- ▶ Previous results fix a probability measure  $\mathbb{P}$  and the inequalities are understood  $\mathbb{P}$ -a.s.
- ▶ In the robust case, we fix the marginal  $\mu$ , but otherwise have **no dominating measure**. Inequalities are **pointwise**.
- ▶ There is also an in between case. We fix a **class of probability measures**  $\mathcal{P}$ . Then, the inequalities are to be understood  $\mathcal{P}$  quasi-surely.
- ▶ In discrete time **Bouchard & Nutz (2014)** proved the duality for **general claims** (not only continuous) and the fundamental theorem of asset pricing in this structure.
- ▶ Continuous time quasi-sure models are studied by **Biagini, Bouchard, Kardaras & Nutz** with continuous paths.

- ▶ Dupire 94, Local volatility and Dupire equation;
- ▶ Hobson 98, explicit hedges via Skorokhod embedding;
- ▶ Beiglböck, Henry-Labordère and Penkner, 2011, discrete time, duality and connection to optimal transport;
- ▶ Galichon, Henry-Labordère and Touzi, 2011, first result and formulation in continuous time and quasi-sure approach;
- ▶ Dolinsky and Soner, 2012, 2013, 2014, 2015 continuous time duality and discrete time market with transaction costs;
- ▶ Bouchard and Nutz, 2013, discrete time quasi-sure approach, fundamental paper studying the quasi-sure set-up;
- ▶ Biagini, Bouchard, Kardaras and Nutz, 2015, cont, time with continuous paths quasi sure FTAP;

FTAP= Fundamental Theorem of Asset Pricing.

MOT= Martingale Optimal Transport

SEP= Skorokhod Embedding Problem.

- ▶ [Acciaio, Beiglböck, Penkner and Schachermayer](#), 2013, FTAP in discrete time ;
- ▶ [Cox, Davis, Dobson, Huesmann, Klimmek, Obloj](#), 1998-2015 ; connections between SEP and MOT ;
- ▶ [Beiglböck, Cox and Huesmann](#), 2014, systematic construction of solutions to the SEP through MOT ;
- ▶ [Possamai, Royer and Touzi](#), 2013, Quasi-sure hedging of mbl claims.

- ▶ Henry-Labordère, Obloj, Spoida, and Touzi, 2012, explicit solutions to the max-max;
- ▶ Acciaio, Beiglböck, Penkner, Schachermayer and Temme, 2013, a proof of Doob's inequality;
- ▶ Beiglböck and Siorpaes, 2012, A new proof of the Biechtelier theorem;
- ▶ Riedel, 2012, decision under uncertainty in one-step setting;
- ▶ Burzoni, Frittelli and Maggis, 2014, arbitrage in multi-step.
- ▶ Soner, Touzi, Zhang 2012, 2013, quasi-sure hedging.



- ▶ Optimal transport can be used in **risk management** context to look for the **optimal correlation structure**.
- ▶ Martingale restriction is useful in applications where the **marginals are the distributions at different time points**.
- ▶ One may imagine restrictions other than martingality.

THANK YOU FOR YOUR ATTENTION.