

# A glance on optimal transport and (some of) its (many) applications

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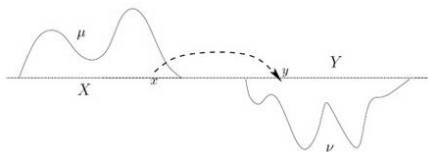
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# Monge and Kantorovich's theory

## Duality, existence, and economic interpretation

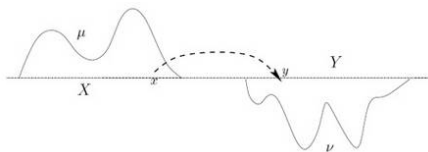
# The Monge problem



If two distributions of mass  $\mu, \nu \in \mathcal{P}(\Omega)$  are given on a compact domain of  $\mathbb{R}^d$ , the Monge problem reads:

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The problem can be generalized with  $\mu \in \mathcal{P}(X), \nu \in \mathcal{P}(Y)$  and  $c : X \times Y \rightarrow \mathbb{R}$ , thus becoming

$$\inf \left\{ \int c(x, T(x)) d\mu(x) : T : X \rightarrow Y, T_{\#}\mu = \nu \right\}.$$

This problem, proposed in the 18th century, has stayed with no solution for more than 200 years...

G. MONGE, *Mémoire sur la théorie des déblais et des remblais*, 1781

# The Kantorovich problem

L. Kantorovich proposed to reformulate the same problem by describing the “transport” from  $X$  to  $Y$  via a different language:

$$\inf \left\{ \int_{X \times Y} c(x, y) d\gamma(x, y) : \gamma \in \Pi(\mu, \nu) \right\},$$

where  $\Pi(\mu, \nu) := \{\gamma \in \mathcal{P}(X \times Y) : (\pi_x)_\# \gamma = \mu, (\pi_y)_\# \gamma = \nu\}$ . We have now a convex, infinite-dimensional, linear programming problem.

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L. KANTOROVICH, *On the transfer of masses*, 1942.



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$$\sup_{\varphi, \psi} \int \varphi d\mu + \int \psi d\nu : \varphi(x) + \psi(y) \leq c(x, y)$$

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# Existence of solutions

## Theorem

If  $X$  and  $Y$  are compact and  $c$  continuous, then

- there exists a solution  $\gamma$  to the primal problem  $\min \left\{ \int c \, d\gamma : \gamma \in \Pi(\mu, \nu) \right\}$  **(KP)**,
- there exist a solution  $(\varphi, \psi) \in C(X) \times C(Y)$  to the dual problem  $\max \left\{ \int \varphi \, d\mu + \int \psi \, d\nu : \varphi(x) + \psi(y) \leq c(x, y) \right\}$  **(DP)**,
- $\min(\mathbf{KP}) = \max(\mathbf{DP})$ ,
- given  $\gamma$  and  $(\varphi, \psi)$ , they are optimal in the primal and dual problems, respectively, if and only if we have  $\varphi(x) + \psi(y) = c(x, y)$  on  $\text{supp}(\gamma)$ .

If  $\mu$  has no atoms, the infimum in the Monge problem equals  $\min(\mathbf{KP})$

$$\inf \left\{ \int c(x, T(x)) \, d\mu(x) : T_{\#}\mu = \nu \right\} = \min \left\{ \int c \, d\gamma : \gamma \in \Pi(\mu, \nu) \right\}$$

and if the optimal  $\gamma$  is of the form  $\gamma = (id, T)_{\#}\mu$  (i.e. it is concentrated on the graph of a map  $T : X \rightarrow Y$ ), then  $T$  solves the Monge problem.

# Few references - monographs

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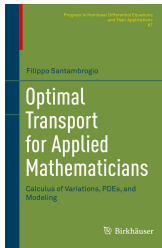
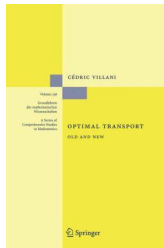
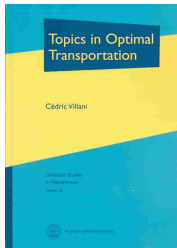
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# Stable Marriages

$X$  = types of women,  $Y$  = types of men,  $\mu$  and  $\nu$  their distributions.

$u_w(x, y)$  = the interest of Ms  $x$  for Mr  $y$ ,  $u_m(x, y)$  = that of Mr  $y$  for Ms  $x$ .

Problem: **finding a stable set of marriages**, i.e. a measure  $\gamma \in \Pi(\mu, \nu)$  (who marries whom), such that no new couple  $(x, y)$  will decide to divorce (each one from his/her current partner) to go together.

**Case of transferable utility:** once  $x$  and  $y$  get married, they decide how to split their total utility  $u_w(x, y) + u_m(x, y)$ , into a quantity  $\varphi(x)$  (utility surplus for Ms  $x$  - now Mrs  $y$ ), and  $\psi(y)$  for Mr  $y$ . Only the sum  $U(x, y) := u_w(x, y) + u_m(x, y)$  really plays a role.

A stable marriage is a triple  $(\gamma, \varphi, \psi)$  such that

- $U(x, y) = \varphi(x) + \psi(y)$   $\gamma$ -a.e.,
- $U(x, y) \leq \varphi(x) + \psi(y)$  for all  $(x, y)$ ,
- $\gamma \in \Pi(\mu, \nu)$ .

Just solve **(KP)** and **(DP)** for  $c = -U$  and change the sign to  $\varphi, \psi$ .

D. GALE, L. S. SHAPLEY, College Admissions and the Stability of Marriage, *Amer. Math. Month.* 1962

P.-A. CHIAPPORI, R. J. McCANN, L. P. NESHEIM Hedonic price equilibria, stable matching, and optimal transport: equivalence, topology, and uniqueness, *Economic Theory*, 2010.

# Prices

$X$  = types of goods on the market,  $Y$  = consumers,  $\mu, \nu$  their distributions.

$u(x, y)$  = the utility of the consumer  $y$  when he buys the good  $x$ .

The goal is to **determine the prices of the goods and who buys what**.

Suppose that the price  $p(x)$  of each good is known; then, each  $y$  will choose what to buy by solving  $\max_x u(x, y) - p(x)$ . Let us call  $p^{(u)}(y)$  the value of the max. We look for  $(\gamma, p)$  such that

- $\gamma \in \Pi(\mu, \nu)$ .
- $p^{(u)}(y) = u(x, y) - p(x)$  for  $(x, y) \in \text{supp}(\gamma)$ .

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**Wait, wait...** prices are only obtained up to an additive constant??

Consider a special good  $x_0$  corresponding to “not buying anything at all”, and impose  $p(x_0) = 0$  (if  $\mu(\{x_0\}) > 0$  then there are not enough goods for everybody and some consumers will stay out of the market).

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The happy ending of free market economy: the stable **(equilibrium)** solution is also the one which maximizes the social utility **(optimal)**.

## Monopolist pricing, principal-agent, contract theory...

A company has the monopoly on a given sector and can decide prices  $p$  and production  $\mu$ . Now  $X$  = set of feasible goods. We also have a production cost  $k : X \rightarrow \mathbb{R}$ .  $Y$ ,  $\nu$ ,  $u$  and  $x_0$  are as before (with  $k(x_0) = 0$ ).

**Goal:** find the optimal pricelist  $p$ . Every  $y$  selects its optimal good  $X_p(y) \in \operatorname{argmax}_x u(x, y) - p(x)$ , and that the total income of the company is

$$I(p) := \int (p - k)(X_p(y)) d\nu(y) = \int (p - k) d\mu,$$

where  $\mu = X_{\#}\nu$  (measure of the real production of goods). The price  $p$  will be chosen so as to maximize  $I(p)$ , with  $p(x_0) = 0$ .

Of course the formulation should be revisited in case the optimal point  $X_p(y)$  is not unique...

M. ARMSTRONG, Multiproduct nonlinear pricing, *Econometrica*, 1996

M. GHISI AND M. GOBBINO The monopolist's problem: existence, relaxation and approximation, *Calc. Var. PDE*, 2005

# The Rochet-Choné formulation of the principal-agent problem

Consider  $u(x, y) = x \cdot y$ , with  $x, y \in \mathbb{R}^d$  (which means: goods  $x$  have  $d$  characteristics, consumers are classified by their interest for each of them). Look at  $\max_x x \cdot y - p(x)$ : we are touching  $p$  from below by affine functions and  $p$  can be replaced by its convex envelop  $\bar{p}$  (if  $p(x) > \bar{p}(x)$  then no  $y$  will ever buy  $x$ , so it's better to decrease its price to  $\bar{p}(x)$ ). The maximizer  $x$  is characterized by  $y = \nabla \bar{p}(x)$ , i.e.  $X_p(y) = (\nabla \bar{p})^{-1}(y) = \nabla p^*(y)$ . Using  $\bar{p}(\nabla p^*(y)) = y \cdot \nabla p^*(y) - p^*(y)$ , the income maximization problem becomes a problem on  $p^*$ . The condition  $p(0) = 0$  will bring  $p^* \geq 0$ :

$$\min \left\{ \int (k(\nabla p^*(y)) + p^*(y) - y \cdot \nabla p^*(y)) d\nu(y) : p^* \text{ convex}, p^* \geq 0 \right\},$$

which is a calculus of variations problem with (non-standard) convexity constraints.

J.-C. ROCHET, P. CHONÉ Ironing, Sweeping, and Multidimensional Screening. *Econometrica*, 1998.

# Optimal transport maps, Brenier's theorem, and gradients of convex functions

Let's consider  $X = Y \subset \mathbb{R}^d$  and  $c$  smooth. Take  $\gamma, \varphi, \psi$  optimal and  $(x_0, y_0) \in \text{supp}(\gamma)$ . We get that  $x \mapsto c(x, y_0) - \varphi(x)$  is maximal at  $x = x_0$ , hence  $\nabla_x c(x_0, y_0) = \nabla \varphi(x_0)$ . If  $c$  satisfies the **twist condition** ( $\nabla_x c$  is injective in  $y$  for every  $x_0$ ), then  $y_0 = (\nabla_x c(x_0, \cdot))^{-1}(\nabla \varphi(x_0)) := T(x_0)$  is uniquely defined, and  $\gamma$  is unique and concentrated on the graph of  $T$  (of course, differentiability of  $\varphi$  must be guaranteed).

## Theorem

Suppose  $c(x, y) = -x \cdot y$  and  $\mu$  absolutely continuous. Then, given  $\nu$ , the optimal  $\gamma$  in **(KP)** is unique and concentrated on the graph of  $\nabla \varphi$ , where  $\varphi$  is a convex function and solves **(DP)**.

A map  $T$  is optimal for the Monge problem if and only if it is the gradient of a convex function (which is differentiable a.e., hence  $\mu$ -a.e.).

Y. BRENIER, Décomposition polaire et réarrangement monotone des champs de vecteurs, CRAS, 1987.



# Wasserstein spaces

Distances, curves, geodesics, and barycenters

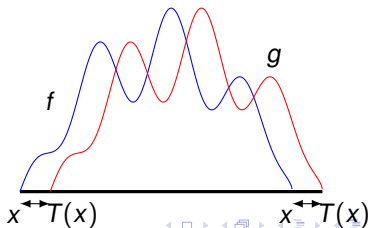
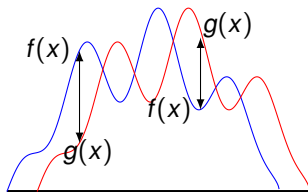
# Wasserstein distances and Wasserstein spaces - 1

Starting from the values **(KP)** we can define a set of distances over  $\mathcal{P}(X)$ , in the following way: for any  $p \in [1, +\infty[$  set

$$W_p(\mu, \nu) = \left( \min \text{ (KP) with } c(x, y) = |x - y|^p \right)^{1/p}.$$

Compared to  $L^p$  distances between densities we can say that they are “horizontal” instead of “vertical”.

**Topology and functional analysis:** if  $X$  is compact, then the convergence for any  $W_p$  is equivalent to the weak convergence in the dual of  $C(X)$ , the space of continuous functions on  $X$ .



## Wasserstein distances and Wasserstein spaces - 2

There is also a dynamical formulation whenever  $X \subset \mathbb{R}^d$  is convex:

$$\begin{aligned} W_p^p(\mu, \nu) &= \inf \left\{ \int_0^1 \int \rho_t |v_t|^p \, dx dt : \partial_t \rho + \nabla \cdot (\rho v) = 0, \rho_0 = \mu, \rho_1 = \nu \right\} \\ &= \inf \left\{ \int_0^1 \int \frac{|w_t|^p}{\rho_t^{p-1}} \, dx dt : \partial_t \rho + \nabla \cdot w = 0, \rho_0 = \mu, \rho_1 = \nu \right\} \end{aligned}$$

This kinetic energy minimization is the so-called *Benamou-Brenier* formulation, which amounts to a convex optimization problem, solvable by Augmented Lagrangian methods.

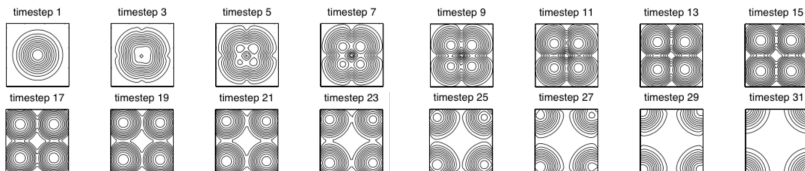
In the case  $p = 1$  the last problem gives an alternative formulation, which is dynamic but stationary, and is known as the Beckmann's **minimal flow** formulation of the Monge problem:

$$W_1(\mu, \nu) = \min \left\{ \int |w| : \nabla \cdot w = \mu - \nu \right\}.$$

J.-D. BENAMOU, Y. BRENIER A computational fluid mechanics solution to the Monge- Kantorovich mass transfer problem, *Numer. Math.*, 2000.

M. BECKMANN, A continuous model of transportation, *Econometrica*, 1952.

# Wasserstein distances and Wasserstein spaces - 3

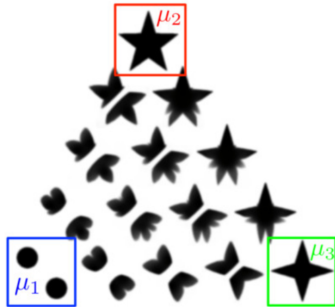


The measures  $\rho_t$  which minimize the Benamou-Brenier formulation are indeed constant-speed geodesics connecting  $\mu$  to  $\nu$  and they have an explicit expression

$$\rho_t = ((1 - t)id + tT)_{\#}\mu,$$

where  $T$  is the optimal transport map from  $\mu$  to  $\nu$ . Such a map exists and is unique, since  $|x - y|^p$  satisfies the twist condition (for  $p > 1$ ).

# Wasserstein barycenters



Picture and numerics by G. Peyré

Given some distributions of mass  $\mu_i$  on a metric space  $X$ , how can we “average” them, to get a typical distribution out of them? We can apply different weights  $\lambda_i$  with  $\sum_i \lambda_i = 1$ , but want to preserve qualitative properties (concentration, ...) which would disappear if we took  $\sum_i \lambda_i \mu_i$ . In a Hilbert space, a barycenter  $\sum_i \lambda_i x_i$  is the solution of  $\min_y \sum_i \lambda_i \|y - x_i\|^2$ .

Here solve  $\min_\rho \sum_i \lambda_i W_2^2(\rho, \mu_i)$ .

M. AGUEH, G. CARLIER, Barycenters in the Wasserstein space, *SIAM J. Math. An.*, 2011.  
J. RABIN, G. PEYRÉ, J. DELON, M. BERNOT. Wasserstein Barycenter and Its Application to Texture Mixing, *Scale Space and Variational Methods in Computer Vision*, 2012.

# Geodesic convexity

What does it mean that a function  $\mathcal{F} : X \rightarrow \mathbb{R}$  is convex, when  $X$  is a metric space? we can say that  $\mathcal{F}$  is **geodesically convex** if  $t \mapsto \mathcal{F}(x(t))$  is convex for every constant-speed geodesics connecting arbitrary points  $x(0), x(1)$ . In the case  $X = W_2(\mathbb{R}^d)$  this was introduced by McCann with the name of **displacement convexity** and the geodesics  $\rho_t$  are known. Three main functionals are considered

$$\mathcal{H}(\rho) := \int H(\rho(x)) dx, \quad \mathcal{V}(\rho) := \int V d\rho, \quad \mathcal{W}(\rho) := \int W(x-y) d\rho(x) d\rho(y).$$

## Theorem

$\mathcal{V}$  is geodesically convex if and only if  $V$  is convex,  $\mathcal{W}$  if  $W$  is convex, and  $\mathcal{H}$  if  $s \mapsto s^d H(s^{-d})$  is convex nondecreasing (this is satisfied by  $H(s) = s^p$ ,  $p > 1$ , and  $H(s) = s \log s$ ).

R. J. McCann A convexity principle for interacting gases. *Adv. Math.*, 1997.

# Equilibria and optimization with measures

## Non-atomic games and urban equilibria

## Nash equilibria with a continuum of players

Consider a game with infinitely many identical players, each one negligible compared to the others (*non-atomic games*), and a common space  $S$  of strategies; players choose their strategies and we look for the realized measure,  $\rho \in \mathcal{P}(S)$ , which induces a payoff function  $f_\rho : S \rightarrow \mathbb{R}$  and we want:  $f_\rho(x) = \min f_\rho$  for every  $x \in \text{supp}(\rho)$ .

**Definition/notation:** given a functional  $\mathcal{F} : \mathcal{P}(S) \rightarrow \mathbb{R}$  we define its first variation as the function  $\frac{\delta \mathcal{F}}{\delta \rho}$ , if it exists, such that

$$\frac{d}{d\varepsilon} \mathcal{F}(\rho + \varepsilon \chi)|_{\varepsilon=0} = \int \frac{\delta \mathcal{F}}{\delta \rho}(\rho) d\chi.$$

In the particular case where  $f_\rho = \frac{\delta \mathcal{F}}{\delta \rho}$ , then the above game is a **potential game**, and equilibria can be found by minimizing  $\mathcal{F}(\rho)$  among  $\rho \in \mathcal{P}(S)$ . If  $\mathcal{F}$  is convex, then necessary optimality conditions are also sufficient, and every equilibrium is a minimizer.

J. NASH, Equilibrium points in n-person games, *Proc. Natl. Acad. Sci.*, 1950.

J. NASH, Non-Cooperative Games *Ann. Math.*, 1951.

R. J. AUMANN, L. S. SHAPLEY, *Values of Non-Atomic Games*, 1968.



## An example from urban use of land

A number of agents must choose where to live in an urban region  $\Omega \subset \mathbb{R}^d$ ;  $\rho \in \mathcal{P}(\Omega)$  is their density. Every agent considers the sum of three costs:

- an exogenous cost, depending on the amenities of  $x$  only:  $V(x)$  (distance to the points of interest. . . );
- an interaction cost, depending on the distances with all the others; when living at  $x$  the cost is  $\int W(x - y)\rho(y) dy$  where  $W$  is usually an increasing function of the distance;
- a residential cost, increasing in the density at  $x$ ; at  $x$  the cost is  $h(\rho(x))$ , for  $h : \mathbb{R}_+ \rightarrow \mathbb{R}$  increasing (where more people live, the price of land is higher or, for the same price, they have less space).

The total cost that we consider is  $f_\rho(x) := V(x) + (W * \rho)(x) + h(\rho(x))$ .

M.J. BECKMANN. Spatial equilibrium and the dispersed city, *Mathematical Land Use Theory*, 1976.

M. FUJITA AND J. F. THISSE. *Economics of Agglomeration: Cities, Industrial Location, and Regional Growth*. 2002.

# Equilibrium and optimality conditions

Consider the following quantity

$$\mathcal{F}(\rho) := \int_{\Omega} V(x)\rho(x)dx + \frac{1}{2} \int_{\Omega} \int_{\Omega} W(x-y)\rho(x)\rho(y)dxdy + \int_{\Omega} H(\rho(x))dx,$$

where  $H$  is defined through  $H' = h$ .

We can see that  $f_{\rho} := V + W * \rho + h(\rho) = \frac{\delta \mathcal{F}}{\delta \rho}$ .

**Warning:** the energy  $\mathcal{F}$  is not the total cost for all the agents, which should be  $\int_{\Omega} f_{\rho}(x)\rho(x)dx$ .

Minimizers are equilibria. What about the converse? In general  $\mathcal{W}$  is not convex (think at  $W(z) = |z|^2$ , so that  $\frac{1}{2}\mathcal{W}(\rho) = \int |x|^2 d\rho(x) - (\int x d\rho(x))^2$ ).

## Theorem

*If  $\mathcal{F}$  is geodesically convex, then minimizers of  $\mathcal{F}$  and equilibria for  $f_{\rho}$  coincide.*

A. BLANCHET, P. MOSSAY, F. SANTAMBROGIO Existence and uniqueness of equilibrium for a spatial model of social interactions, *Int. Econ. Rev.*, 2015.

## About the residential cost

Suppose that agents have a certain budget to be divided into land consumption and money consumption, and that they have a concave and increasing utility function  $U$  for land. They solve a problem of the form

$$\max\{U(L) + m : pL + m \leq B\},$$

where  $p$  represents the price for land,  $L$  is the land consumption,  $m$  is the left-over of the money, and  $B$  the budget constraint. The optimal land consumption will be such that  $U'(L) = p$ . The optimal utility is  $B + U(L) - U'(L)L$  (relation between  $L$  and utility).

The land consumption is the reciprocal of the density, hence  $L = \frac{1}{\rho}$ , and the residential cost  $h(\rho)$ , which is the opposite of the utility, is

$$h(\rho) = \frac{1}{\rho} U' \left( \frac{1}{\rho} \right) - U \left( \frac{1}{\rho} \right) - B.$$

$\frac{1}{t} U' \left( \frac{1}{t} \right) - U \left( \frac{1}{t} \right) = \left( -t U \left( \frac{1}{t} \right) \right)'$ , hence  $h = H'$  with  $H(t) = -t U \left( \frac{1}{t} \right) - Bt$ .  $H$  satisfies McCann's condition if and only if  $s \mapsto U(s^d)$  is concave and increasing.

# Cournot-Nash urban equilibria

Cournot-Nash equilibria concern nonatomic games where agents are not indistinguishable. Consider an extra variable  $z \in Z$ , the origin of agents, and  $\mu \in \mathcal{P}(Z)$  its distribution. Agents must choose a location  $x \in \Omega$ , and their payoff is given by  $c(z, x) + f_\rho(x)$ , where  $\rho$  is the distribution of their choices on  $\Omega$ . We look for a triple  $(\gamma, \rho, \varphi)$  with  $\gamma \in \Pi(\mu, \rho)$ ,  $\varphi(z) = \min_x c(z, x) + f_\rho(x)$  and  $\varphi(z) = c(z, x) + f_\rho(x)$  on  $\text{spt}(\gamma)$ .

This condition may be obtained by solving

$$\min \left\{ \int c \, d\gamma + \mathcal{F}((\pi_x)_\# \gamma) : (\pi_z)_\# \gamma = \mu \right\}.$$

In the case  $c(x, y) = |x - y|^p$  this means  $\min_\rho W_\rho^p(\mu, \rho) + \mathcal{F}(\rho)$ . The Kantorovich potentials will be of the form  $(\varphi, -f_\rho)$ .

**Convexity:** the extra term  $W_\rho^p(\mu, \rho)$  is convex in  $\rho$ . It is in general non geodesically convex (!! ) but it is convex along generalized geodesics  $\rho_t = ((1 - t)T_0 + tT_1)_\# \mu$  (with  $T_i$  the optimal map from  $\mu$  to  $\rho_i$ ,  $i = 0, 1$ ).

A. BLANCHET, G. CARLIER, *Optimal transport and Cournot-Nash Equilibria*, *Math. Op. Res.*, 2016  
 L. AMBROSIO, N. GIGLI, G. SAVARÉ *Gradient Flows*, 2005

## Variants

### Congestion problems – Martingale transport

## Congestion and continuous Wardrop equilibria

Consider  $\min\{\int |w| : \nabla \cdot w = \mu - \nu\} = \min(\mathbf{KP})$ . If we insert a weight,  $\min\{\int k(x)|w| : \nabla \cdot w = \mu - \nu\}$  becomes equivalent to  $(\mathbf{KP})$  with the cost  $c$  given by the distance  $d_k$  ( $d_k(x, y) := \min\{\int_0^1 k(\omega)|\omega'| : \omega(0) = x, \omega(1) = y\}$ ).

## Congestion and continuous Wardrop equilibria

Consider  $\min\{\int |w| : \nabla \cdot w = \mu - \nu\} = \min(\mathbf{KP})$ . If we insert a weight,  $\min\{\int k(x)|w| : \nabla \cdot w = \mu - \nu\}$  becomes equivalent to  $(\mathbf{KP})$  with the cost  $c$  given by the distance  $d_k$  ( $d_k(x, y) := \min\{\int_0^1 k(\omega)|\omega'| : \omega(0) = x, \omega(1) = y\}$ ). But in congested urban regions,  $k$  mainly depends on traffic itself, i.e. on  $w$ . Hence, we should consider the superlinear Beckmann's problem

$$\min\left\{\int H(|w|) : \nabla \cdot w = \mu - \nu\right\},$$

which is connected to degenerate elliptic equations  $\nabla \cdot (H'(|\nabla\phi|)\frac{\nabla\phi}{|\nabla\phi|}) = \mu - \nu$ , and to equilibrium problems where the cost per unit length of a path is  $H'(i)$  and  $i$  is the traffic intensity.

- J. G. WARDROP, Some theoretical aspects of road traffic research, *Proc. Inst. Civ. Eng.*, 1952.  
 G. CARLIER, C. JIMENEZ, F. SANTAMBROGIO, Optimal transportation with traffic congestion and Wardrop equilibria, *SIAM J. Contr. Optim.* 2008.  
 L. BRASCO, G. CARLIER, F. SANTAMBROGIO, Congested traffic dynamics, weak flows and very degenerate elliptic equations, *J. Math. Pures et Appl.*, 2010.

# MFG with density penalization

A variant of the Benamou-Brenier problem: for  $G(x, \cdot)$  convex, minimize

$$\mathcal{A}(\rho, \nu) := \int_0^T \int_{\Omega} \left( \frac{1}{2} \rho_t |v_t|^2 + G(x, \rho_t) \right) + \int_{\Omega} \Psi \rho_T$$

among pairs  $(\rho, \nu)$  such that  $\partial_t \rho + \nabla \cdot (\rho \nu) = 0$ , with given  $\rho_0$ . The solution is given by  $\nu = -\nabla \phi$ , where  $(\rho, \phi)$  solves a coupled system:

$$\begin{cases} -\partial_t \phi + \frac{|\nabla \phi|^2}{2} = G'(x, \rho), & \phi(T, x) = \Psi(x), \\ \partial_t \rho - \nabla \cdot (\rho \nabla \phi) = 0, & \rho(0, x) = \rho_0(x). \end{cases}$$

Every agent minimizes  $\int_0^T \left[ \frac{1}{2} |\omega'(t)|^2 + G'(\omega(t), \rho_t(\omega(t))) \right] dt + \Psi(\omega(T))$ : the running cost depends on the density  $\rho_t$  realized by the agents themselves.

J.-M. LASRY, P.-L. LIONS, *Mean-Field Games*, *Japan. J. Math.* 2007

P.-L. LIONS, courses at Collège de France, 2006/12, videos available online

P. CARDALIAGUET, lecture notes, [www.ceremade.dauphine.fr/~cardalia/](http://www.ceremade.dauphine.fr/~cardalia/)

G. BUTTAZZO, C. JIMENEZ, E. OUDET An optimization problem for mass transportation with congested dynamics *SICON*, 2009

J.-D. BENAMOU, G. CARLIER, F. SANTAMBROGIO, *Variational Mean Field Games*, *Active Particles I*, 2016



## Option pricing

Let  $Z = f(S_T^1, S_T^2, \dots, S_T^N)$  be an option with payoff depending on several assets  $S_T^i$  at a same time  $T$ . Its value is  $\mathbb{E}^Q[Z]$  for a probability  $Q$  which makes each price a martingale. Suppose that  $Z$  is new on the market, but that the European Calls on each  $S^i$  already exist for each strike price  $K$ . We know the law  $\mu_i$  of each of the  $S_T^i$ , but not their joint law (under  $Q$ ): estimating the price of  $Z$  is a **multi-marginal transport problem**

$$\min / \max \left\{ \int f(x_1, x_2, \dots, x_N) d\gamma : (\pi_i)_\# \gamma = \mu_i \right\}$$

If on the contrary,  $Z = f(S_1, S_2, \dots, S_T)$  depends on the history of a same asset (and the corresponding call options  $(S_t - K)_+$  exist on the market for every strike and every maturity time), then the problem is different

$$\min / \max \left\{ \mathbb{E}^Q[f(S_1, \dots, S_N)] : (S_i)_\# Q = \mu^i, (S_t)_t \text{ is a } Q\text{-martingale} \right\}.$$

**Kantorovich problems with the martingale constraints are a new frontier of optimal transport!**

M. BEIGLBÖCK, N. JUILLET On a problem of optimal transport under marginal martingale constraints, *Ann. Prob.*, 2016

# Robust super-hedging

The same problem can be read in this way: suppose we want to hedge a possible loss of  $f(S_1, \dots, S_N)$ , depending on the prices of the asset  $S$ . What we can do:

- buy usual options, based on the value of the asset at time  $t$ , i.e. buy  $\phi(S_t)$ , paying  $\int \phi d\mu_t$ ;
- buy the asset itself at time  $t \leq N - 1$ , pay  $S_t$ , and re-sell at price  $S_{t+1}$  (but the number of assets to buy can only be chosen according to the information at time  $t$ , i.e. it must be of the form  $\psi(S_1, \dots, S_t)$ , which hedges an amount  $\psi(S_1, \dots, S_t)(S_{t+1} - S_t)$ .

The optimal hedging problem becomes

$$\min \left\{ \sum_{t=0}^N \int \varphi_t d\mu_t : \sum_{t=0}^N \varphi_t(x_t) + \sum_{t=0}^{N-1} \psi(x_1, \dots, x_t)(x_{t+1} - x_t) \geq f(x_1, \dots, x_N) \right\}.$$

This problem is exactly the dual of the martingale transport problem (the  $\varphi_t$  dualize the marginal constraints, the  $\psi_t$  the martingale constraints).

A. GALICHON, P. HENRY-LABORDÈRE, N. TOUZI, A stochastic control approach to no-arbitrage bounds given marginals, with an application to lookback options *Ann. Appl. Prob.*, 2014.

*That's all for this short presentation of OT and some of its applications*

Thanks for your attention