

# TOPICS IN EQUILIBRIUM TRANSPORTATION

Alfred Galichon (NYU, Economics and Courant)

Statistics seminar,  
London School of Economics, March 7, 2018

- ▶ This talk is based on the following three working papers:
  - ▶ [GKW] AG, Kominers, S. and Weber, S. (2015). Costly Concessions: An Empirical Framework for Matching with Imperfectly Transferable Utility.
  - ▶ [CGJK] Dupuy, A., AG, Jaffe, S., and Kominers, S. On the incidence of taxation in matching markets.
  - ▶ [GH] Galichon, A., Hsieh, Y.-W. A theory of decentralized matching markets without transfers, with an application to surge pricing.

## Agenda:

1. Economic motivation
2. Equilibrium transportation
3. Galois connections and distance functions

## Agenda:

1. Economic motivation
2. Equilibrium transportation
3. Galois connections and distance functions

## Agenda:

1. Economic motivation
2. Equilibrium transportation
3. Galois connections and distance functions

# Section 1

## ECONOMIC MOTIVATION

- ▶ Consider a model of the labor market. Assume that a population of *workers* is characterized by their type  $x \in \mathcal{X}$ , where  $\mathcal{X} = \mathbb{R}^d$  for simplicity. There is a distribution  $P$  over the workers, which is assumed to sum to one.
- ▶ A population of *firms* is characterized by their types  $y \in \mathcal{Y}$  (say  $\mathcal{Y} = \mathbb{R}^d$ ), and their distribution  $Q$ . It is assumed that there is the same total mass of workers and firms, so  $Q$  sums to one.
- ▶ Each worker must work for one firm; each firm must hire one worker. Let  $\pi(x, y)$  be the probability of observing a matched  $(x, y)$  pair.  $\pi$  should have marginal  $P$  and  $Q$ , which is denoted

$$\pi \in \mathcal{M}(P, Q).$$

- In the simplest case, the utility of a worker  $x$  working for a firm  $y$  at wage  $w(x, y)$  will be

$$\alpha(x, y) + w(x, y)$$

while the corresponding profit of firm  $y$  is

$$\gamma(x, y) - w(x, y).$$

- In this case, the total surplus generated by a pair  $(x, y)$  is

$$\alpha(x, y) + w + \gamma(x, y) - w = \alpha(x, y) + \gamma(x, y) =: \Phi(x, y)$$

which does not depend on  $w$  (no transfer frictions). A central planner may thus like to choose assignment  $\pi \in \mathcal{M}(P, Q)$  so to

$$\max_{\pi \in \mathcal{M}(P, Q)} \int \Phi(x, y) d\pi(x, y).$$

But as it turns out, this is also **the equilibrium solution**.



- ▶ The equilibrium assignment is determined by an important quantity: the **wages**. Let  $w(x, y)$  be the wage of employee  $x$  working for firm of type  $y$ .
- ▶ Let the indirect surpluses of worker  $x$  and firm  $y$  be respectively

$$u(x) = \max_y \{ \alpha(x, y) + w(x, y) \}$$

$$v(y) = \max_x \{ \gamma(x, y) - w(x, y) \}$$

so that  $(\pi, w)$  is an equilibrium when

$$u(x) \geq \alpha(x, y) + w(x, y) \text{ with equality if } (x, y) \in \text{Supp}(\pi)$$

$$v(y) \geq \gamma(x, y) - w(x, y) \text{ with equality if } (x, y) \in \text{Supp}(\pi)$$

- ▶ By summation,

$$u(x) + v(y) \geq \Phi(x, y) \text{ with equality if } (x, y) \in \text{Supp}(\pi).$$

- One can show that the equilibrium outcome  $(\pi, u, v)$  is such that  $\pi$  is solution to the primal Monge-Kantorovich Optimal Transportation problem

$$\max_{\pi \in \mathcal{M}(P, Q)} \int \Phi(x, y) d\pi(x, y)$$

and  $(u, v)$  is solution to the dual OT problem

$$\begin{aligned} \min_{u, v} \int u(x) dP(x) + \int v(y) dQ(y) \\ \text{s.t. } u(x) + v(y) \geq \Phi(x, y) \end{aligned}$$

- Feasibility+Complementary slackness yield the desired equilibrium conditions

$$\begin{aligned} \pi &\in \mathcal{M}(P, Q) \\ u(x) + v(y) &\geq \Phi(x, y) \\ (x, y) \in \text{Supp}(\pi) &\implies u(x) + v(y) = \Phi(x, y) \end{aligned}$$

Here, **optimum=equilibrium**. “Second welfare theorem”, “invisible hand”, etc.

- ▶ Is equilibrium always the solution to an optimization problem?
- ▶ **It is not.** This is why this talk is about “Equilibrium Transportation,” which contains, but is strictly more general than “Optimal Transportation”.

- ▶ Is equilibrium always the solution to an optimization problem?
- ▶ **It is not.** This is why this talk is about “Equilibrium Transportation,” which contains, but is strictly more general than “Optimal Transportation”.

- Consider the same setting as above, but introduce (nonlinear) taxes.
- Instead of assuming that workers' and firm's payoffs are linear in wages, assume

$$u(x) = \max_y \{ \alpha_{xy} + N(w(x, y)) \}$$

$$v(y) = \max_x \{ \gamma_{xy} - w(x, y) \}$$

where  $N(w)$  is indecreasing and continuous, interpreted as the net wage if  $w$  if the gross wage.

- Of course, OT is recovered when  $N(w) = w$  (no tax).

- Let  $\mathcal{F}_{xy}$  be the set of feasible utilities that  $x$  and  $y$  can achieve through some wage  $w$ . One has

$$\mathcal{F}_{xy} = \{(u, v) : \exists w \in \mathbb{R}, u \leq \alpha_{xy} + N(w), v \leq \gamma_{xy} - w\},$$

which rewrites

$$\mathcal{F}_{xy} = \{(u, v) : u - \alpha_{xy} \leq N(\gamma_{xy} - v)\}.$$

- The interior of this set, denoted  $\mathcal{F}_{xy}^0$ , is the set such that this inequality holds true.
- In the case of OT,

$$\mathcal{F}_{xy} = \{(u, v) : u + v \leq \alpha_{xy} + \gamma_{xy}\}.$$

## Section 2

# EQUILIBRIUM TRANSPORTATION

- We have therefore that  $(\pi, u, v)$  is an equilibrium outcome when

$$\left\{ \begin{array}{l} (PF) : \pi \in \mathcal{M}(P, Q) \\ (DF) : (u(x), v(y)) \notin \mathcal{F}_{xy}^0 \\ (NC) : (x, y) \in \text{Supp}(\pi) \implies (u(x), v(y)) \in \mathcal{F}_{xy}. \end{array} \right.$$

- This is a **Nonlinear Complementarity Problem** (NCP) with much structure.
- Problem: existence of an equilibrium outcome? yes in the discrete case ( $\mathcal{X}$  and  $\mathcal{Y}$  finite): Kelso-Crawford, Alkan and Gale.



- Link with Galois connection, see Noeldeke and Samuelson (2015) and Larry's talk. Let

$$G_{xy}(v) = a_{xy} + N(\gamma_{xy} - v).$$

One has  $(u_x, v_y) \notin \mathcal{F}_{xy}^0$  if and only if  $u_x \geq G_{xy}(v_y)$  which is equivalent to  $v_y \geq G_{xy}^{-1}(u_x)$ .

- By condition (DF) and (NC), we get that if  $(\pi, u, v)$  is the solution to an Equilibrium transportation problem

$$u(x) = \max_{y \in \mathcal{Y}} G_{xy}(v(y)) \text{ and } v(y) = \max_{x \in \mathcal{X}} G_{xy}^{-1}(u(x))$$

- OT is a special case ( $\Phi$ -conjugacy, see Villani's book):

$$u(x) = \max_{y \in \mathcal{Y}} \{\Phi_{xy} - v(y)\} \text{ and } v(y) = \max_{x \in \mathcal{X}} \{\Phi_{xy} - u(x)\}$$

and even more special:  $\Phi_{xy} = x.y$  (Legendre-Fenchel conjugacy)

$$u(x) = \max_{y \in \mathcal{Y}} \{x.y - v(y)\} \text{ and } v(y) = \max_{x \in \mathcal{X}} \{x.y - u(x)\}.$$

- Link with Galois connection, see Noeldeke and Samuelson (2015) and Larry's talk. Let

$$G_{xy}(v) = \alpha_{xy} + N(\gamma_{xy} - v).$$

One has  $(u_x, v_y) \notin \mathcal{F}_{xy}^0$  if and only if  $u_x \geq G_{xy}(v_y)$  which is equivalent to  $v_y \geq G_{xy}^{-1}(u_x)$ .

- By condition (DF) and (NC), we get that if  $(\pi, u, v)$  is the solution to an Equilibrium transportation problem

$$u(x) = \max_{y \in \mathcal{Y}} G_{xy}(v(y)) \text{ and } v(y) = \max_{x \in \mathcal{X}} G_{xy}^{-1}(u(x))$$

- OT is a special case ( $\Phi$ -conjugacy, see Villani's book):

$$u(x) = \max_{y \in \mathcal{Y}} \{\Phi_{xy} - v(y)\} \text{ and } v(y) = \max_{x \in \mathcal{X}} \{\Phi_{xy} - u(x)\}$$

and even more special:  $\Phi_{xy} = x.y$  (Legendre-Fenchel conjugacy)

$$u(x) = \max_{y \in \mathcal{Y}} \{x.y - v(y)\} \text{ and } v(y) = \max_{x \in \mathcal{X}} \{x.y - u(x)\}.$$

- Link with Galois connection, see Noeldeke and Samuelson (2015) and Larry's talk. Let

$$G_{xy}(v) = \alpha_{xy} + N(\gamma_{xy} - v).$$

One has  $(u_x, v_y) \notin \mathcal{F}_{xy}^0$  if and only if  $u_x \geq G_{xy}(v_y)$  which is equivalent to  $v_y \geq G_{xy}^{-1}(u_x)$ .

- By condition (DF) and (NC), we get that if  $(\pi, u, v)$  is the solution to an Equilibrium transportation problem

$$u(x) = \max_{y \in \mathcal{Y}} G_{xy}(v(y)) \text{ and } v(y) = \max_{x \in \mathcal{X}} G_{xy}^{-1}(u(x))$$

- OT is a special case ( $\Phi$ -conjugacy, see Villani's book):

$$u(x) = \max_{y \in \mathcal{Y}} \{\Phi_{xy} - v(y)\} \text{ and } v(y) = \max_{x \in \mathcal{X}} \{\Phi_{xy} - u(x)\}$$

and even more special:  $\Phi_{xy} = x.y$  (Legendre-Fenchel conjugacy)

$$u(x) = \max_{y \in \mathcal{Y}} \{x.y - v(y)\} \text{ and } v(y) = \max_{x \in \mathcal{X}} \{x.y - u(x)\}.$$

- Assuming everything is smooth, and letting  $f_P$  and  $f_Q$  be the densities of  $P$  and  $Q$  we have under some conditions that the equilibrium transportation plan is given by  $y = T(x)$ , where mass balance yields

$$|\det DT(x)| = \frac{f_P(x)}{f_Q(T(x))}$$

and optimality yields

$$\partial_x G_{xT(x)}^{-1}(u(x)) + \partial_u G_{xT(x)}^{-1}(u(x)) \nabla u(x) = 0$$

which thus inverts into

$$T(x) = e(x, u(x), \nabla u(x)).$$

- Trudinger (2014) studies Monge-Ampere equations of the form

$$|\det De(., u, \nabla u)| = \frac{f_P}{f_Q(e(., u, \nabla u))}.$$

(more general than Optimal Transport where no dependence on  $u$ ).

- Assuming everything is smooth, and letting  $f_P$  and  $f_Q$  be the densities of  $P$  and  $Q$  we have under some conditions that the equilibrium transportation plan is given by  $y = T(x)$ , where mass balance yields

$$|\det DT(x)| = \frac{f_P(x)}{f_Q(T(x))}$$

and optimality yields

$$\partial_x G_{xT(x)}^{-1}(u(x)) + \partial_u G_{xT(x)}^{-1}(u(x)) \nabla u(x) = 0$$

which thus inverts into

$$T(x) = e(x, u(x), \nabla u(x)).$$

- Trudinger (2014) studies Monge-Ampere equations of the form

$$|\det De(., u, \nabla u)| = \frac{f_P}{f_Q(e(., u, \nabla u))}.$$

(more general than Optimal Transport where no dependence on  $u$ ).

- ▶ Galois connections and Monge-Ampère equations break the symmetry between  $x$  and  $y$ . Duality is an offspring of this broken symmetry.
- ▶ However, in contrast, the ET problem is symmetric in the role  $x$  and  $y$  play. Is there a way of restoring symmetry between  $x$  and  $y$ ?

- ▶ Galois connections and Monge-Ampère equations break the symmetry between  $x$  and  $y$ . Duality is an offspring of this broken symmetry.
- ▶ However, in contrast, the ET problem is symmetric in the role  $x$  and  $y$  play. Is there a way of restoring symmetry between  $x$  and  $y$ ?

## Section 3

# DISTANCE FUNCTIONS



- ▶ [GKW] introduce distance function: for  $(u, v) \in \mathbb{R}^2$ , let

$$\Psi_{xy}(u, v) = \min \{t \in \mathbb{R} : (u - t, v - t) \in \mathcal{F}_{xy}\}$$

which is the distance along the diagonal between  $(u, v)$  and the frontier of  $\mathcal{F}_{xy}$ , with a minus sign if  $(u, v)$  is in the set.

- ▶ Economic interpretation: what is the quantity of utility that we can give or remove to  $x$  and  $y$  *in the same amount* such that they reach the efficient frontier?
- ▶ This object has nice properties:
  - ▶  $\Psi_{xy}(u, v) \leq 0$  iff  $(u, v) \in \mathcal{F}_{xy}$
  - ▶  $\Psi_{xy}(u, v) < 0$  iff  $(u, v) \in \mathcal{F}_{xy}^0$
  - ▶  $\Psi_{xy}(u + t, v + t) = \Psi_{xy}(u, v) + t$
- ▶ Note that in the case of OT,

$$\Psi_{xy}(u, v) = \frac{u + v - (\alpha_{xy} + \gamma_{xy})}{2}.$$

- $(\pi, u, v)$  is an equilibrium outcome when

$$\left\{ \begin{array}{l} (PF) : \pi \in \mathcal{M}(P, Q) \\ (DF) : \Psi_{xy}(u(x), v(y)) \geq 0 \\ (NC) : (x, y) \in \text{Supp}(\pi) \implies \Psi_{xy}(u(x), v(y)) = 0. \end{array} \right.$$

- In the rest of the talk, I will argue that these objects can be useful in the study of the Equilibrium Transport problem.

- It is convenient to parameterize the frontier of  $\mathcal{F}_{xy}$  by a scalar parameter. To do this, consider a reparameterization by a  $45^\circ$  rotation: if  $u_x$  and  $v_y$  are such that  $\Psi_{xy}(u_x, v_y) = 0$ , let  $w_{xy} = u_x - v_y$ .
- More generally, the set of equations

$$\begin{cases} \Psi_{xy}(u_x, v_y) = 0 \\ u_x - v_y = w_{xy} \end{cases}$$

defines  $u_x = \mathcal{U}_{xy}(w_{xy})$  and  $v_y = \mathcal{V}_{xy}(w_{xy})$ .

- One has a very simple expression of  $\mathcal{U}_{xy}$  and  $\mathcal{V}_{xy}$  from  $\Psi$ :

$$\begin{aligned} \mathcal{U}_{xy}(w_{xy}) &= -\Psi_{xy}(0, -w_{xy}) \\ \mathcal{V}_{xy}(w_{xy}) &= -\Psi_{xy}(w_{xy}, 0) \end{aligned}$$

and

$$G_{xy}(v) = \mathcal{U}_{xy} \circ \mathcal{V}_{xy}^{-1}(v) \text{ and } G_{xy}^{-1}(u) = \mathcal{V}_{xy} \circ \mathcal{U}_{xy}^{-1}(u).$$

- Look at the individual rationality conditions:

$$\begin{cases} u_x \geq \mathcal{U}_{xy}(v_y) \\ \pi_{y|x} > 0 \implies u_x \geq \mathcal{U}_{xy}(v_y) \end{cases}$$

where  $\pi_{y|x} = \pi_{xy}/p_x$  is the conditional distribution of  $Y$  given  $X$  under  $\pi$ .

- A regularized version of these is provided by the Gibbs distribution

$$\pi_{y|x} = \frac{\exp \mathcal{U}_{xy}(w_{xy}) / T}{\exp u_x / T}$$

where the max has been replaced by the smooth-max. Hence, letting  $a_x = u_x - T \ln p_x$ , we get

$$T \ln \pi_{xy} + a_x = \mathcal{U}_{xy}(w_{xy}).$$

- Similarly, individual rationality on the side of firms relaxes into

$$\pi_{x|y} = \frac{\exp \mathcal{V}_{xy}(w_{xy}) / T}{\exp v_y / T}$$

and thus, after letting  $b_y = v_y - T \ln q_y$ ,

$$T \ln \pi_{xy} + b_y = \mathcal{V}_{xy}(w_{xy}).$$

- We get

$$\begin{cases} T \ln \pi_{xy} + a_x = \mathcal{U}_{xy}(w_{xy}) \\ T \ln \pi_{xy} + b_y = \mathcal{V}_{xy}(w_{xy}) \end{cases}$$

- Thus, applying  $\Psi$  term by term yields

$$\Psi_{xy}(T \ln \pi_{xy} + a_x, T \ln \pi_{xy} + b_y) = 0$$

hence

$$\pi_{xy} = \exp(-\Psi_{xy}(a_x, b_y) / T)$$

where  $a_x$  and  $b_y$  solve the system

$$\begin{cases} \sum_y \exp(-\Psi_{xy}(a_x, b_y) / T) = p_x \\ \sum_x \exp(-\Psi_{xy}(a_x, b_y) / T) = q_y \end{cases}.$$

- By subtraction, we have

$$a_x - b_y = w_{xy}.$$

- ▶ **Theorem:** when  $T \rightarrow 0$ ,  $u$  and  $v$  tend to the solution of the ET problem.
- ▶ **Open question:** when there are multiple solutions at  $T = 0$ , which solution is selected?
- ▶ In the case of OT, it is the maximal entrop solution. Does this result still hold here?

- In OT, a lot of attention recently on the Monge-Kantorovich problem with entropic regularization

$$\max_{\pi \in \mathcal{M}(P, Q)} \int \Phi(x, y) d\pi(x, y) - 2T \int \log \pi(x, y) d\pi(x, y)$$

which is useful for computation (via Iterated Bregman projections) and econometrics (link w the logit model).

- In the OT case, this boils down to

$$\pi_{xy} = \exp \left( \frac{\Phi_{xy} - a_x - b_y}{2T} \right)$$

where  $a$  and  $b$  are adjusted to meet the constraint  $\pi \in \mathcal{M}(P, Q)$ .

- In the ET case, no optimization formulation... but works just as well. Recall

$$\pi_{xy} = \exp \left( -\Psi_{xy}(a_x, b_y) / T \right).$$

## POINT 2: COMPUTATION (CTD)

- Take temperature parameter  $T > 0$  and look for  $\pi$  under the form

$$\pi_{xy} = \exp \left( -\frac{\Psi_{xy}(a_x, b_y)}{T} \right)$$

Note that when  $T \rightarrow 0$ , the limit of  $\Psi_{xy}(a_x, b_y)$  is nonnegative, and the limit of  $\pi_{xy} \Psi_{xy}(a_x, b_y)$  is zero.

- If  $\pi_{xy} = \exp(-\Psi_{xy}(a_x, b_y)/T)$ , condition  $\pi \in \mathcal{M}(P, Q)$  boils down to set of nonlinear equations in  $(u, v)$

$$\begin{cases} \sum_{y \in \mathcal{Y}} \exp \left( -\frac{\Psi_{xy}(a_x, b_y)}{T} \right) = p_x \\ \sum_{x \in \mathcal{X}} \exp \left( -\frac{\Psi_{xy}(a_x, b_y)}{T} \right) = q_y \end{cases}$$

which we call the *nonlinear Bernstein-Schrödinger* equation.

- In the optimal transportation case, this becomes the classical B-S equation

$$\begin{cases} \sum_{y \in \mathcal{Y}} \exp \left( \frac{\Phi_{xy} - a_x - b_y}{2T} \right) = p_x \\ \sum_{x \in \mathcal{X}} \exp \left( \frac{\Phi_{xy} - a_x - b_y}{2T} \right) = q_y \end{cases}$$



- Note that  $F_x : a_x \rightarrow \sum_{y \in \mathcal{Y}} \exp\left(-\frac{\Psi_{xy}(a_x, b_y)}{T}\right)$  is a decreasing and continuous function. Mild conditions on  $\Psi$  therefore ensure the existence of  $a_x$  so that  $F_x(a_x) = p_x$ .
- Our algorithm [GKW] is thus a nonlinear Jacobi algorithm:
  - Make an initial guess of  $b_y^0$
  - Determine the  $a_x^{k+1}$  to fit the  $p_x$  margins, based on the  $b_y^k$
  - Update the  $b_y^{k+1}$  to fit the  $q_y$  margins, based on the  $a_x^{k+1}$ .
  - Repeat until  $b^{k+1}$  is close enough to  $b^k$ .
- One can proof that  $b_y^k$  decrease to fixed point. Convergence is very fast in practice.

- The distance function formalism allows to make a connection with Gale-Shapley's theory of stable marriages. Introduce

$$\Psi_{xy}(u, v) = \max(u - \alpha_{xy}, v - \gamma_{xy}).$$

- This interprets as: no matter what happens,  $x$  can get at most  $\alpha_{xy}$  and  $y$  can get at most  $\gamma_{xy}$ .
- Assume  $n_x = 1$  and  $m_y = 1$  for all  $x$  and  $y$ . Then GH show that for any stable matching  $\pi$  in the Gale-Shapley sense, there exist  $u$  and  $v$  such that  $(\pi, u, v)$  is an equilibrium transport; and conversely, if  $(\pi, u, v)$  is an equilibrium transport, then  $\pi$  is stable in the Gale and Shapley sense.

## POINT 4: BACK TO NONLINEAR TAXES

- Back to the tax example [DGJK], assume that

$$N(w) = \min_{k=1 \dots K} \left\{ n^k + (1 - \theta^k) (w - w^k) \right\}$$

where  $\theta^1 = 0 < \theta^2 < \dots < \theta^K$ , and  $n^0 = 0$ ,  
 $n^{k+1} = n^k + (1 - \theta^k) (w^{k+1} - w^k)$ .

- Recall  $\mathcal{F}_{xy} = \{(u, v) : u - \alpha_{xy} \leq N(\gamma_{xy} - v)\}$ . Because the tax schedule is progressive ( $N$  concave),  $\mathcal{F}_{xy}$  can be written as the intersection of the  $\mathcal{F}_{xy}^k$ , where

$$\mathcal{F}_{xy}^k = \left\{ (u, v) : u - \alpha_{xy} - n^k \leq (1 - \theta^k) (\gamma_{xy} - v - w^k) \right\}.$$

- Note that  $\Psi$  associated with the intersection of  $\Psi^k$  is equal to the  $\max_k \Psi^k$ . One has

$$\Psi^k(u, v) = \frac{u - \alpha_{xy} - n^k - (1 - \theta^k) (\gamma_{xy} - v - w^k)}{2 - \theta^k},$$

and as a result

$$\Psi(u, v) = \max_{k=1, \dots, K} \left\{ \frac{u - \alpha_{xy} - n^k - (1 - \theta^k) (\gamma_{xy} - v - w^k)}{2 - \theta^k} \right\}.$$

Thank you!

- ▶ More generally, the following operations on distance functions correspond to geometric operations on feasible sets:
  - ▶  $\max \{ \Psi^1, \Psi^2 \}$ : intersection
  - ▶  $\min \{ \Psi^1, \Psi^2 \}$ : union
  - ▶  $\Psi(u - \alpha, v - \gamma)$ : translation
  - ▶  $T\Psi(u/T, v/T)$ : homothety
  - ▶  $\lambda\Psi^1 + (1 - \lambda)\Psi^2$ : interpolation
- ▶ These operations are exploited in the TraME project (<https://github.com/TraME-Project/>), a software for flexible computation of equilibrium transportation problems.