

DYNAMIC MODELS OF MATCHING

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Slides available from
<http://alfredgalichon.com/>

- ▶ Dynamic aspects are crucial for matching problems
 - ▶ In labor economics (human capital formulation)
 - ▶ In family economics (fertility decisions)
 - ▶ In mergers and acquisitions
 - ▶ In school choice
 - ▶ Etc.
- ▶ We offer a framework for these dynamic matching problems:
 - ▶ with or without unobserved heterogeneity
 - ▶ with finite or infinite (stationary) horizon
 - ▶ with equilibrium prediction, structural estimation, comparative statics and welfare

- ▶ Large current literature on the estimation of **static transferable utility (TU) two-sided** (matching) models in the static case:
 - ▶ Choo and Siow (2006), Fox (2010), Galichon and Salanié (2011), Dupuy and Galichon (2014), Chiappori, Salanié and Weiss (2019), Fox et al. (2018)
- ▶ Dynamic discrete choice literature on one-sided models since Rust (1987) assumes the decision maker's type evolves stochastically depending on the choice made at the previous period.
- ▶ Today's goal: investigate the dynamic aspect of static matching models by assuming that the match has an effect on types *on both sides of the market*. And show how to take models to data on **changing relationships over time**.

- ▶ NTU case when matches are forever (e.g. kidney)
 - ▶ Unver (2010), Bloch and Cantala (2017), Doval (2021)
- ▶ Search and matching: the matching has no effect on partners, but match opportunities are scarce
 - ▶ NTU case: Burdett and Coles (1997); Eeckhout (1999), Peski (2021)
 - ▶ TU case: Shimer and Smith (2000) .
- ▶ TU case:
 - ▶ Erlinger, McCann, Shi, Siow and Wolthoff (2015), McCann, Shi, Siow and Wolthoff (2015) – 2 period sequential matching, with universities in a first period, then with firms.
 - ▶ Choo (2015) studies a dynamic matching problem with a focus on the age of marriage

Populations:

- ▶ $z \in \mathcal{Z}$ agents to be matched, $z = x$ (worker) or $z = y$ (firms)
- ▶ q_z = mass of agents of type z (fixed for now)

Matches:

- ▶ $a \in \mathcal{A}$ matches; $a = xy$ or $a = x$ (unassigned worker) or $a = y$ (unassigned firm)
- ▶ w_a = cardinality of the match (2 for pair, 1 for unassigned)
- ▶ \tilde{S}_a = joint (random utility) surplus of match a
 - ▶ Choo-Siow's separable random utility assumption:
 $\tilde{S}_a = S_a + \sum_{z \in a} \varepsilon_z$, where (ε_z) vector of random payoff shifters (Gumbel for simplicity)

Equilibrium quantities:

- ▶ p_z = payoff of z
- ▶ μ_a = mass of match a

Result 1 (Choo-Siow): μ_a and p_z are related by
 $\mu_a = \exp(w_a^{-1}(S_a + \sum_{z \in a}(\log q_z - p_z)))$ and p solves
 $\sum_{a \ni z} \exp(w_a^{-1}(S_a + \sum_{z \in a}(\log q_z - p_z))) = q_z$ for each z .

(Proof in the appendix at the end of these slides).

Note that at equilibrium, $\sum_{a \in \mathcal{A}} w_a \mu_a = \sum_{z \in \mathcal{Z}} q_z$. Hence, define

$$Z(q, p, S) = \sum_{a \in \mathcal{A}} w_a \exp \left(w_a^{-1} \left(S_a + \sum_{z \in a} (\log q_z - p_z) \right) \right) - \sum_{z \in \mathcal{Z}} q_z.$$

We have $\frac{\partial Z(p, q, S)}{\partial p_z} = \sum_{a \ni z} \mu_a$, with
 $\mu_a = \exp \left(w_a^{-1} (S_a + \sum_{z \in a} (\log q_z - p_z)) \right).$

Therefore:

Result 2 (Galichon-Salanié): The equilibrium p solves

$$\min_p \sum_{z \in \mathcal{Z}} q_z p_z + Z(p, q, S).$$

We now consider a two-sided Rust-type dynamic matching model with TU. Assume that individuals' types vary across periods, and that the transition depend on current period match.

Consider

$$\mathbb{R}_{za}$$

the mass of individuals z induced forward at next period by one unit of match a .

For instance, if $a = xy$, worker x 's type will transition to x' with proba. $\mathbb{P}_{x'|xy}$, and firm y 's type will transition to y' with proba. $\mathbb{Q}_{y'|xy}$. In that case,

$$\mathbb{R}_{za} = \sum_{x'} 1\{z = x'\} \mathbb{P}_{x'|xy} + \sum_{y'} 1\{z = y'\} \mathbb{Q}_{y'|xy}.$$

Note that (as in Rust) the transition are Markovian:

(x chooses $a = xy$ w.p. μ_a/q_x) and then (transitions to x' w.p. $\mathbb{R}_{x'|xy}$).

Hence, conditional transition probability $x \rightarrow x'$ equals to $\sum_y \mu_{xy} \mathbb{R}_{x'|xy} / q_x$.

In that case, S_a needs to accrue for future-period payoffs p' , in addition to short-term joint payoff Φ_a , and

$$S_a = \Phi_a + \beta \sum_z \mathbb{R}_{za} p'_z = (\Phi + \beta \mathbb{R}^\top p')_a.$$

Now redefine Z by inserting expression for S , we have

$$Z(q, p, p') = \sum_{a \in \mathcal{A}} w_a \exp \left(w_a^{-1} \left((\Phi + \beta \mathbb{R}^\top p')_a + \sum_{z \in \mathcal{Z}} (\log q_z - p_z) \right) \right) - \sum_{z \in \mathcal{Z}} q_z$$

Z is all we need to write the equilibrium equations of the model. Indeed,

- ▶ $\partial Z / \partial q_z = \sum_{a \ni z} \mu_a / q_z - 1$ excess share of demand for type z
- ▶ $-\partial Z / \partial p_z = \sum_{a \ni z} \mu_a =$ mass of z at current period
- ▶ $\beta^{-1} \partial Z / \partial p'_z = \sum_{a \in \mathcal{A}} \mathbb{R}_{za} \mu_a =$ mass of z at next period

A stationary equilibrium has

$$p = p' \text{ [rational expectations]}$$

and expresses as

$$\left\{ \begin{array}{l} \frac{\partial Z(q, p, p)}{\partial q_z} = 0 \text{ [market clearing for each type]} \\ \beta \frac{\partial Z(q, p, p)}{\partial p_z} + \frac{\partial Z(q, p, p)}{\partial p'_z} = 0 \text{ [stationarity]} \end{array} \right. .$$

Note that Z is concave in q and jointly convex in (p, p') .

When $\beta = 1$, set $F(q, p) = Z(q, p, p)$ is concave-convex and the equations of the model

$$\begin{cases} \partial F(q, p) / \partial q = 0 \\ \partial F(q, p) / \partial p = 0 \end{cases}$$

are obtained as the saddlepoint conditions for the min-max problem

$$\min_p \max_q F(q, p).$$

Computation using Chambolle-Pock's first order scheme:

$$\begin{cases} q^{t+1} = q^t - \epsilon \partial_q F(q^t, 2p^t - p^{t-1}) \\ p^{t+1} = p^t + \epsilon \partial_p F(q^t, p^t) \end{cases}$$

Surprising fact: algorithm works even for $\beta < 1$ although min-max interpretation is lost.

Now assume we want to solve the inverse problem: based on observed $\hat{\mu}_a$ recover information about Φ .

Parameterize $\Phi_a = \sum_k \phi_{ak} \lambda_k$ and look for λ .

Express

$$\begin{aligned} Z(q, p, p', \lambda) \\ &= \sum_{a \in \mathcal{A}} w_a \exp \left(w_a^{-1} \left(\left(\sum_k \phi_{ak} \lambda_k + \beta \mathbb{R}^\top p' \right)_a + \sum_{z \in \mathcal{A}} (\log q_z - p_z) \right) \right) \\ &\quad - \sum_{z \in \mathcal{Z}} q_z \end{aligned}$$

and note that the partial derivatives of Z with respect to the new variables λ_k also have a natural interpretation. Indeed,

$$\frac{\partial Z}{\partial \lambda_k} = \sum_{a \in \mathcal{A}} \mu_a \phi_{ak}$$

is the predicted k -th moments of ϕ .

Define a function H as

$$H(q, p, p', \lambda) = Z(q, p, p', \lambda) - \sum_{a \in \mathcal{A}} \hat{\mu}_a \phi_{ak} \lambda_k$$

which is jointly convex in (p, p', λ) , and note that the indentifying equations are now

$$\begin{cases} \frac{\partial H(q, p, p', \lambda)}{\partial q} = 0 \text{ [market clearing]} \\ \beta \frac{\partial H(q, p, p', \lambda)}{\partial p} + \frac{\partial H(q, p, p', \lambda)}{\partial p'} = 0 \text{ [stationarity]} \\ \frac{\partial H(q, p, p', \lambda)}{\partial \lambda} = 0 \text{ [moment matching]} \end{cases}$$

In the case $\beta = 1$ this is still a saddlepoint problem, now

$$\min_{p, \lambda} \max_q H(q, p, p, \lambda)$$

for which Chambolle-Pock's first order scheme still applies. It even (mysteriously) still applies when $\beta < 1$.

- ▶ Identification issues à la Kalouptsi, Scott & Souza-Rodrigues (2019) and Kalouptsi, Kitamura and Lima (2021).
- ▶ Theoretical convergence of the first order scheme outside of $\beta = 1$ (min-max).
- ▶ Empirical application: human capital accumulation on the labor market.
- ▶ With Dupuy, Ciscato and Weber: application to family economics (divorce, remarriage and the number of kids).
- ▶ Extension to imperfectly transferable utility (later).

- ▶ The next 'math+econ+code' masterclass on equilibrium transport and matching models in economics will take place June 21-25, 2021. More info on

http://alfredgalichon.com/mec_equil/

- ▶ Jules Baudet and I are organizing a **kidney transplant hackaton** for the math+econ+code. More info at

<http://alfredgalichon.com/kindey-transplant-hackaton/>

and email us: ag133@nyu.edu or jules.baudet99@gmail.com if you are interested!

We need to determine

- ▶ μ_a = mass of matches of type a is formed so that

$$\sum_{a \ni z} \mu_a = q_z$$
- ▶ U_{za} = z 's share of surplus in a match a so that

$$S_a = \sum_{z \in a} U_{za}$$
 and so that agent z in a match a gets $U_{za} + \varepsilon_a$
- ▶ p_z = average payoff of players of type z , so that

$$p_z = \log \sum_{a \ni z} \exp U_{za}$$

Logit model: probability that z chooses match a is

$$\mu_a / q_z = \frac{\exp U_{za}}{\sum_{a \ni z} \exp U_{za}} = \exp (U_{za} - p_z)$$

hence

$$\log (\mu_a / q_z) = U_{za} - p_z$$

Choo-Siow: summing over $z \in a$ yields

$$\mu_a = \exp (w_a^{-1} (S_a + \sum_{z \in a} (\log q_z - p_z))) .$$