

Aggregate Matchings

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What we do:

Revealed preference exercise for matching theory.

Reconcile:

- ▶ Theory of stable **individual** matchings.
- ▶ Data on **aggregate** matchings.

What we do.



vs.



What we do.

$$\begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

What we do.

$$\begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 8 & 0 & 0 \\ 0 & 4 & 3 & 0 \\ 7 & 3 & 0 & 0 \\ 0 & 0 & 9 & 5 \end{pmatrix}$$

Marriage Data (Michigan)

Age	12-20	21-25	26-30	31-35	36-40	41-50	51-94
12-20	231	47	8	0	0	1	0
21-25	329	798	156	32	11	7	0
26-30	71	477	443	136	27	8	0
31-35	11	148	249	196	83	21	0
36-40	2	41	105	144	114	51	1
41-50	0	15	42	118	121	162	25
51-94	0	2	11	11	35	137	158

Question:

- ▶ Given an “aggregate matching table” (data), when are there preferences for individuals s.t. the matching is stable?
- ▶ In other words, what are the **testable implications of stability** for aggregate matchings.

General motivation: two sided decision problems

- ▶ Standard revealed preference:
Alice buys tomatoes when carrots are available
 $\Rightarrow (T \succ_A C)$.

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- ▶ Standard revealed preference:
Alice buys tomatoes when carrots are available
 $\Rightarrow (T \succ_A C)$.
- ▶ Two sided decision:
Alice chooses Tomás over Carlos
 $\Rightarrow (T \succ_A C)$ or (C prefers its match to A).

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- ▶ Important problem: rationalizing preferences can explain revealed preference and “available sets” (budgets).
- ▶ Hence direction of revealed preference is affected by the hypothesized rationalizing preferences.
- ▶ Literature mostly deals with the problem by assuming **transferable utility**.

Main results

Revealed preference exercise:

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- ▶ Characterization of rationalizable agg. match.
- ▶ Characterization under TU: strictly more restrictive.

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Ex:

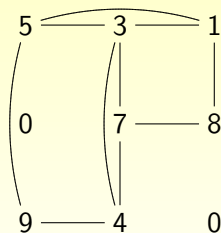
5	3	1
0	7	8
9	4	0

Main results

Revealed preference exercise:

- ▶ Characterization of rationalizable agg. match.
- ▶ Characterization under TU: strictly more restrictive.

Ex:



Main results

Econometric estimation strategy:

- ▶ Moment inequalities
- ▶ Set identification parameters in “index” utility model.
- ▶ Empirical illustration to US marriage data.

Other results

- ▶ Stability for aggregate match. is substantially different from individual match.
- ▶ Structure of stable aggregate matchings.

Model

An *aggregate matching market* is described by a triple $\langle M, W, \succ \rangle$, where:

- ▶ M and W are disjoint, finite sets. We call the elements of M *types of men* and the elements of W *types of women*.
- ▶ $\succ = ((\succ_m)_{m \in M}, (\succ_w)_{w \in W})$ is a profile of strict preferences: for each m and w , \succ_m is a linear order over $W \cup \{m\}$ and \succ_w is a linear order over $M \cup \{w\}$.

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Note: identical preferences within type.

We show that relaxing this assumption in our framework leads to a vacuous theory.

Model

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Model

- ▶ An *aggregate matching* is a $K \times L$ matrix $X = (X_{ij})$ with $X_{ij} \in \mathbf{N}$.
- ▶ An aggregate matching X is *canonical* if $X_{ij} \in \{0, 1\}$.
- ▶ A canonical matching X is a *simple matching* if for each i there is at most one j with $X_{ij} = 1$, and for each j there is at most one i with $X_{ij} = 1$.

Model

- ▶ X is *individually rational* if

$$X_{ij} > 0 \Rightarrow w_j >_{m_i} m_i \text{ and } m_i >_{w_j} w_j.$$

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$$X_{ij} > 0 \Rightarrow w_j >_{m_i} m_i \text{ and } m_i >_{w_j} w_j.$$

- ▶ (m_i, w_j) is a *blocking pair* if \exists
 - ▶ $w_k \in W$ with $X_{ik} > 0$, and $m_l \in M$ with $X_{jl} > 0$,
 - ▶ s.t. $w_j >_{m_i} w_k$ and $m_i >_{w_j} m_l$.

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 - ▶ s.t. $w_j >_{m_i} w_k$ and $m_i >_{w_j} m_l$.
- ▶ X is *stable* if it is individually rational and there are no blocking pairs for X .

Model

Given X , construct a canonical aggregate matching X^c by:

- ▶ $X_{ij}^c = 0$ when $X_{ij} = 0$ and
- ▶ $X_{ij}^c = 1$ when $X_{ij} > 0$.

Observation

An aggregate matching X is stable if and only if X^c is stable.

Example: simple vs. aggregate matching

Let $\langle M, W, \succ \rangle$ with $M = \{m_1, m_2, m_3\}$, $W = \{w_1, w_2, w_3\}$, and

m_1	m_2	m_3	w_1	w_2	w_3
<hr/>			<hr/>		
w_1	w_2	w_3	m_2	m_3	m_1
w_2	w_3	w_1	m_3	m_1	m_2
w_3	w_1	w_2	m_1	m_2	m_3

Model

The following simple matchings are stable:

$$X^1 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad X^2 = \begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix}$$

Sum of X^1 and X^2 :

$$\hat{X} = X^1 + X^2 = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \\ 0 & 1 & 1 \end{pmatrix}.$$

(m_1, w_2) is a blocking pair.

Stability

$\langle M, W, \succ \rangle$ defines a graph (V, E) where

- ▶ V is the set of pairs (i, j)
- ▶ $((i, j), (k, l)) \in E$ if
 - ▶ $w_l \succ_{m_i} w_j$ and $m_i \succ_{w_l} m_k$ or
 - ▶ $w_j \succ_{m_k} w_l$ and $m_k \succ_{w_j} m_i$.

X is stable iff

$$((i, j), (k, l)) \in E \Rightarrow X_{ij}X_{kl} = 0. \quad (1)$$

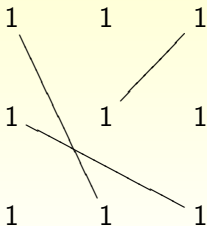
Otherwise (ie. $X_{ij} = X_{kl} = 1$), either (i, j) or (k, j) is blocking pair.

Stability – Example

3 men and women:

\succ_{m_1}	\succ_{m_2}	\succ_{m_3}	\succ_{w_1}	\succ_{w_2}	\succ_{w_3}
w_1	w_2	w_3	m_2	m_3	m_1
w_2	w_3	w_1	m_3	m_1	m_2
w_3	w_1	w_2	m_1	m_2	m_3

Graph:

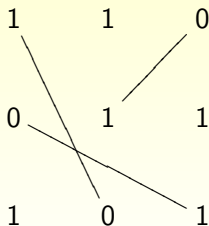


Stability – Example

3 men and women:

\succ_{m_1}	\succ_{m_2}	\succ_{m_3}	\succ_{w_1}	\succ_{w_2}	\succ_{w_3}
w_1	w_2	w_3	m_2	m_3	m_1
w_2	w_3	w_1	m_3	m_1	m_2
w_3	w_1	w_2	m_1	m_2	m_3

Stable matching:



Stability – contrapositive

An *antiedge* is a pair $(i, j), (k, l)$ with $i \neq k \in M; j \neq l \in W$ s.t.
 $X_{ij} = X_{kl} = 1$.

Then X is stable iff

$$(ij), (kl) \text{ is anti-edge} \Rightarrow \begin{cases} d_{ij}d_{lik} = 0 \\ d_{jki}d_{kjl} = 0 \end{cases} \quad (2)$$

Define: $d_{ij} = \mathbb{1}(w_l >_{m_i} w_j)$

Structure of Aggregate Stable Matchings

X *dominates* X' if

$$X_{ij} = 0 \Rightarrow X'_{ij} = 0.$$

Proposition

Let X be a stable aggregate matching. If X' is an aggregate matching, and X dominates X' , then X' is stable.

So all stable matchings are described by set of maximal stable matchings.

(Trivial) Algorithm for maximal stable matching.

Given (V, E)

- ▶ Enumerate vertices, $V = \{1, 2, \dots, N\}$.
- ▶ $X^0 =$ identically zero.
- ▶ For $v \in V$, X^{v-1} , define X^v by changing entry v .
 - ▶ $X_v^v = 1$ if 1 is not violated
 - ▶ $X_v^v = 0$ o/w.
- ▶ Let $X = X^N$.

Model

Proposition

Let X be an individual stable matching.

- 1. If $K = L = 3$ then X is not a maximal stable matching.*
- 2. If $K > 3$, $L > 3$ and X is a maximal stable matching, then one of the following two possibilities must hold:*
 - 2.1 For all (i, j) , the submatching $X^{-(i,j)}$ is a maximal stable matching in the $-(i, j)$ submarket.*
 - 2.2 There is (h, l) with $X_{hl} = 1$, and a maximal stable matching \tilde{x} , for which $\tilde{x}_{h,j} = \tilde{x}_{i,l} = 0$ for all i and j .*

Rationalizable Matchings

Given: $M = \{m_1, \dots, m_K\}$ and $W = \{w_1, \dots, w_L\}$.

X is *rationalizable* if \exists preference profile \succ s.t. X is a stable aggregate matching in $\langle M, W, \succ \rangle$.

Rationalizable Matchings

Given X :

Define a “lattice graph” (V, L) on the matrix X .

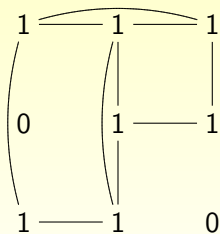
- ▶ Vertices: (i, j) s.t. $X_{i,j} = 1$
- ▶ Edge $(i, j) - (i', j')$ if share a column or a row.

Example

Let X be

$$\begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{pmatrix}.$$

(V, L) is:



Rationalizable Matchings

Theorem

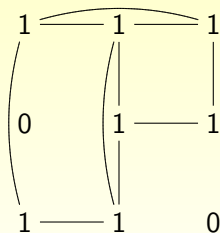
An aggregate matching X is rationalizable if and only if the associated graph (V, L) has not two connected distinct minimal cycles.

Rationalizable Matchings

Let X be

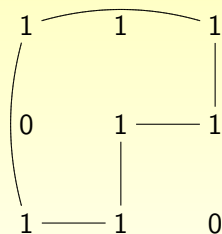
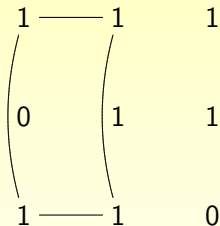
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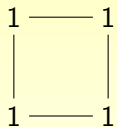
Rationalizable Matchings

The following are two minimal cycles that are connected.



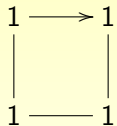
Idea: necessity.

Canonical cycle:



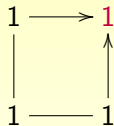
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Preferences \Rightarrow orientation of edges:



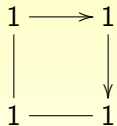
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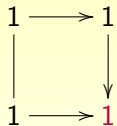
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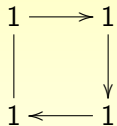
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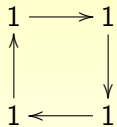
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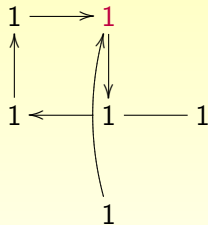
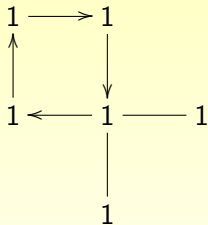
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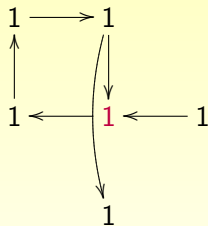
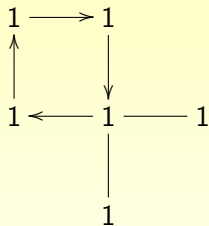
Idea: necessity

So a cycle must be oriented as a flow.

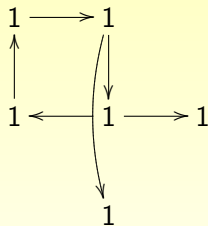
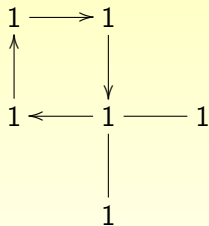
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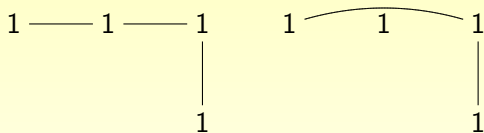
- ▶ Orientation of a minimal path must then point **away** from a cycle.

Idea: necessity

- ▶ Orientation of a minimal path must then point **away** from a cycle.
- ▶ Two connected cycles \Rightarrow connecting path must point away from both.

Idea: necessity

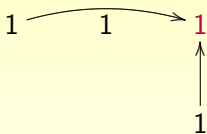
Subsequent edges in a minimal path must be at a right angle:



Idea: necessity

Two connected cycles \Rightarrow connecting path must point away from both.

So connected path does (at some point):



\Rightarrow no two connected cycles.

Idea: sufficiency

- ▶ Given X , construct an orientation of (V, L) .
- ▶ Use orientation to define preferences.

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- ▶ Given X , construct an orientation of (V, L) .
- ▶ Use orientation to define preferences.
- ▶ Decompose (V, L) in connected components. At most one cycle in each.
- ▶ Orient cycle as a “flow,” and paths as “flows” pointing away from cycle.
- ▶ Uniqueness of cycle within a component ensures transitivity.

TU rationalization

Surplus: $\alpha_{i,j} \in \mathbb{R}$.

Surplus generated by matchings of types i and j in X is $X_{i,j}\alpha_{i,j}$.

TU rationalization

X is *TU-rationalizable* by a matrix of surplus α if X is unique sol. to:

$$\begin{aligned} & \max_{\tilde{X}} \sum_{i,j} \alpha_{i,j} \tilde{X}_{i,j} \\ & \text{s.t.} \begin{cases} \forall j \sum_i \tilde{X}_{i,j} = \sum_i X_{i,j} \\ \forall i \sum_j \tilde{X}_{i,j} = \sum_j X_{i,j} \end{cases} \end{aligned} \quad (3)$$

TU rationalization

Theorem

An aggregate matching X is TU-rationalizable if and only if the associated graph (V, L) contains no minimal cycles.

Corollary

If an aggregate matching X is TU-rationalizable, then it is rationalizable.

Estimation

Parametrized preferences:

$$u_{ij} = Z_{ij}\beta + \varepsilon_{ij}, \quad (4)$$

$$d_{ijk} \equiv \mathbf{1}(u_{ij} \geq u_{ik}).$$

Recall:

An *antiedge* is a pair $(i, j), (k, l)$ with $i \neq k \in M; j \neq l \in W$ s.t.
 $X_{ij} = X_{kl} = 1$.

Then X is stable iff

$$(ij), (kl) \text{ is anti-edge} \Rightarrow \begin{cases} d_{ij}d_{lik} = 0 \\ d_{jki}d_{kjl} = 0 \end{cases} \quad (5)$$

Estimation

$$\begin{aligned} Pr((ij), (kl) \text{ antiedge}) &\leq (1 - Pr(d_{ij}d_{lik} = 1))(1 - Pr(d_{jki}d_{kjl} = 1)) \\ &= Pr(d_{ij}d_{lik} = 0, d_{jki}d_{kjl} = 0). \end{aligned}$$

Estimation

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Gives a moment inequality:

$$\mathbb{E} \left[\underbrace{\mathbb{1}((ij), (kl) \text{ antiedge}) - Pr(d_{ij}d_{lik} = 0, d_{jki}d_{kjl} = 0; \beta)}_{g_{ijkl}(X_t; \beta)} \right] \leq 0.$$

The identified set is defined as

$$\mathbb{B}_0 = \{\beta : \mathbb{E} g_{ijkl}(X_t; \beta) \leq 0, \forall i, j, k, l\}.$$

Estimation

Sample analog

$$\begin{aligned} & \frac{1}{T} \sum_t \mathbb{1}((ij), (kl) \text{ is antiedge in } X_t) - 1 \\ & \qquad \qquad \qquad + Pr(d_{ij}d_{lik} = 0, d_{jki}d_{kjl} = 0; \beta) \\ & = \frac{1}{T} \sum_t g_{ijkl}(X_t; \beta). \end{aligned}$$

Estimation

Problem: condition in the theorem is violated.

Hence **no** preferences (no betas) rationalize data.

Estimation

Problem: condition in the theorem is violated.
Hence **no** preferences (no betas) rationalize data.

We relax the model (\exists other solutions).

Estimation – Relaxation of the model

A blocking pair may not form.

$$\delta_{ijkl} = P(\text{types } (i, j), (k, l) \text{ communicate}).$$

Idea: a BP forms only when types (i, j) , (k, l) communicate.

Estimation – Relaxation of the model

Stability inequalities become:

$$\left(\begin{array}{l} (ij), (kl) \text{ is anti-edge} \\ (ij), (kl) \text{ meet} \end{array} \right) \Rightarrow \begin{cases} d_{ij}d_{lik} = 0 \\ d_{jki}d_{kjl} = 0 \end{cases}$$

Assume: two events are independent.

Estimation – Relaxation of the model

Modified moment inequality:

$$Pr((ij), (kl) \text{ antiedge}) \leq \frac{Pr(d_{ij}d_{lik} = 0, d_{jki}d_{kjl} = 0; \beta)}{\delta_{ijkl}}$$

Estimation – Relaxation of the model

We suppose δ_{ijkl} depends on the number of agents in each type.

$$\delta_{ijkl}^t = 1 - (1 - p)^{x_{T_i^M, T_j^W} + x_{T_k^M, T_l^W}}$$

where p is the prob. that two agents meet.

Estimation – Relaxation of the model

As a result: we are **weighting anti-edges** by how many agents are involved.

Sample moment inequalities

$$\begin{aligned} \frac{1}{T} \sum_t g_{ijkl}(X_t; \beta) = & \left(\frac{1}{T} \sum_t \mathbb{1}((ij), (kl) \text{ is antiedge in } X_t) * \delta_{ijkl}^t \right) \\ & - (1 - Pr(d_{ilj} = 1; \beta_{1,2}) Pr(d_{lik} = 1; \beta_{3,4})) \\ & (1 - Pr(d_{jki} = 1; \beta_{3,4}) Pr(d_{kjl} = 1; \beta_{1,2})) \end{aligned}$$

for all combinations of pairs, (i, j) and (k, l) .

Specification of Utilities

$$Utility^m = \beta_1 |Age^m - Age^w|^- + \beta_2 |Age^m - Age^w|^+ + \varepsilon^m$$

$$Utility^w = \beta_3 |Age^m - Age^w|^- + \beta_4 |Age^m - Age^w|^+ + \varepsilon^w$$

Results: Identified set.

We describe the identified set for different values of p .

if p is too high \Rightarrow identified set = \emptyset .

if p is too low \Rightarrow identified set is everything.

Idea: choose high p as discipline on our estimates.

Results: Identified set.

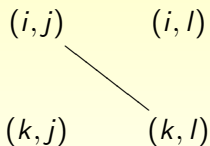
Table: Unconditional Bounds of β .

p	β_1		β_2		β_3		β_4	
	min	max	min	max	min	max	min	max
0.0006	-2	2	-2	2	-2	2	-2	2
0.0007	-2	2	-2	1.6	-2	2	-2	1.2
0.0008	-2	1.2	-1.6	1	-2	1	-1.4	0.6
0.0009	-1.2	0.4	-0.6	0.6	-2	0.4	-0.6	0.4

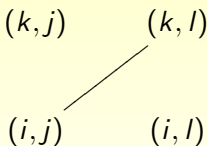
Joint identified sets

- ▶ More anti-edges below the diagonal, where $age^m > age^w$.
- ▶ More “downward-sloping” anti-edges than “upward-sloping” ones.

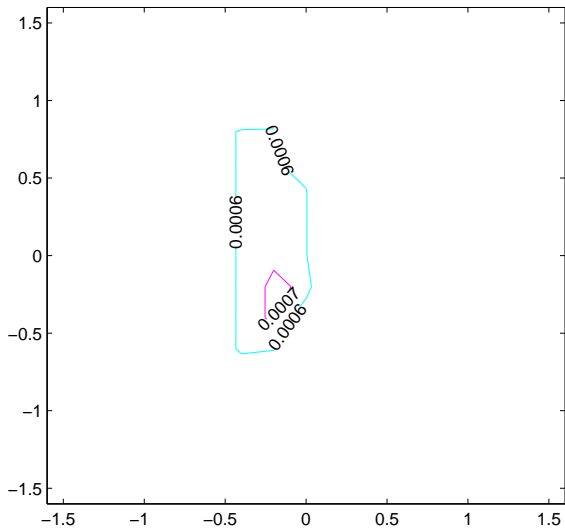
Downward-sloping anti-edge:



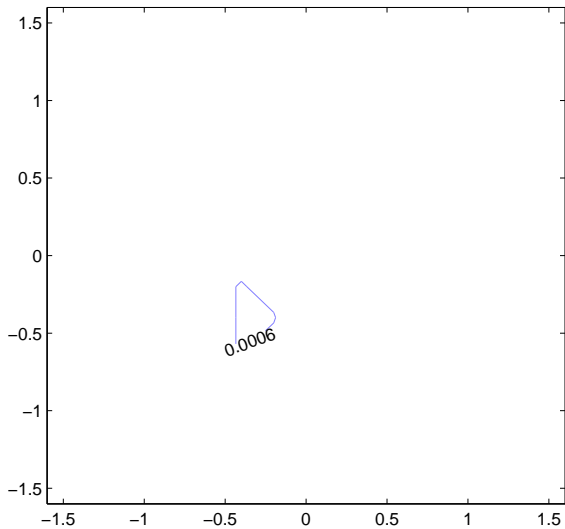
Upward-sloping anti-edge:



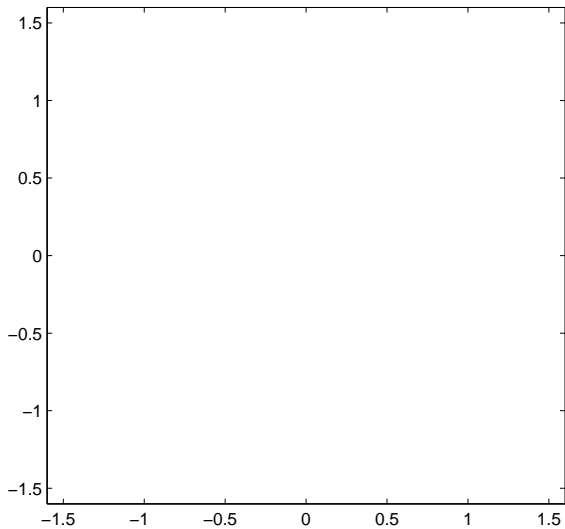
$$\beta_3 = -1 \text{ and } \beta_4 = 1$$



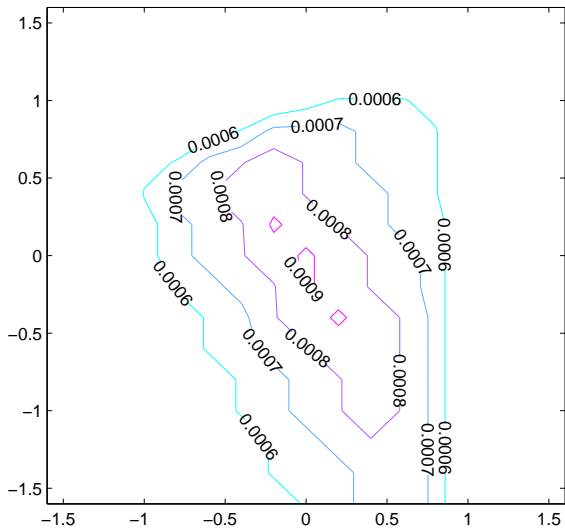
$$\beta_3 = 0 \text{ and } \beta_4 = 1$$



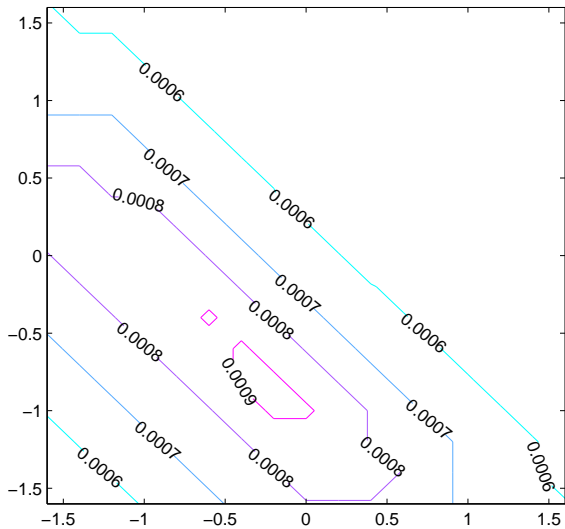
$\beta_3 = 1$ and $\beta_4 = 1$



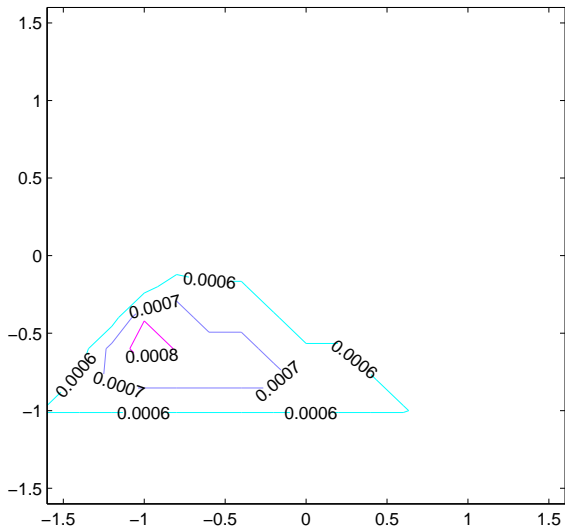
$$\beta_3 = -1 \text{ and } \beta_4 = 0$$



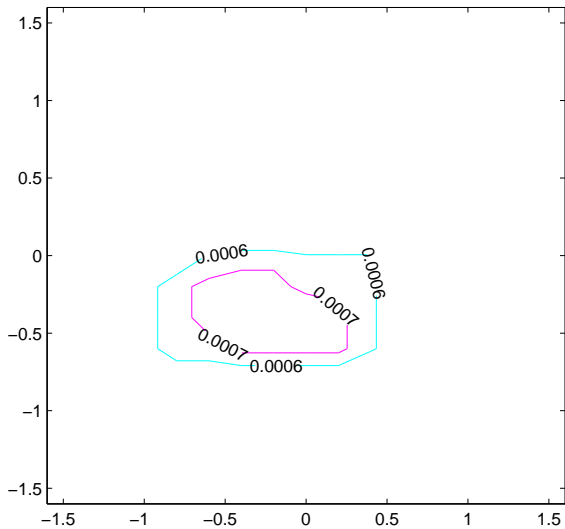
$\beta_3 = 0$ and $\beta_4 = 0$



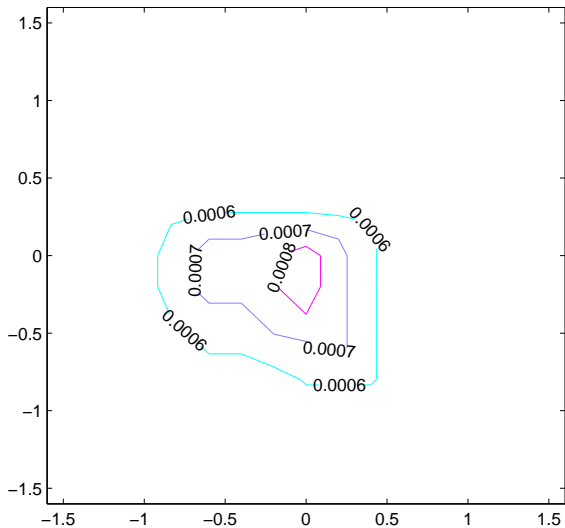
$$\beta_3 = 1 \text{ and } \beta_4 = 0$$



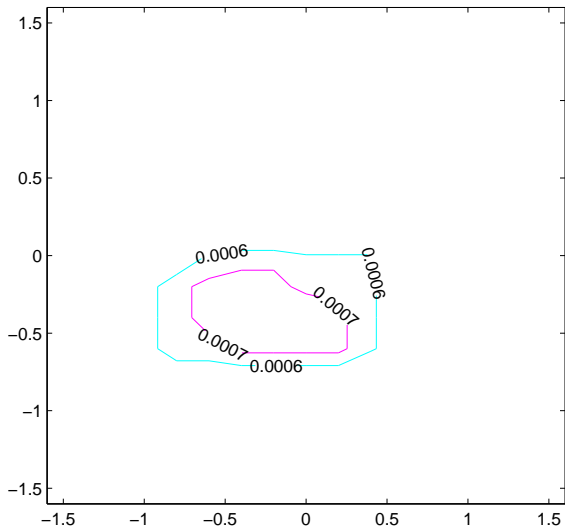
$$\beta_3 = -1 \text{ and } \beta_4 = -1$$



$$\beta_3 = 0 \text{ and } \beta_4 = -1$$



$$\beta_3 = 1 \text{ and } \beta_4 = -1$$



Related literature

TU model:

- ▶ Theory: Shapley-Shubik (1971)
- ▶ Econometrics: Choo-Siow (2006), Fox (2007), Bajari-Fox (2008)

NTU model:

- ▶ Theory: Gale-Shapley (1967), Knuth (1971)
- ▶ Econometrics: Dagsvik (2000)

Aggregate NTU matching: no theoretical results; Dagsvik develops estimation techniques.

Conclusions

- ▶ First theoretical characterization of stable aggregate matchings.
- ▶ Testable implications of stability for aggregate matchings.
- ▶ Econometric estimation technique ← based on moment inequalities implied by stability.
- ▶ Illustration to US marriage data.