

# Optimal Dynamic Matching under Local Compatibility: An Application to Kidney Exchange

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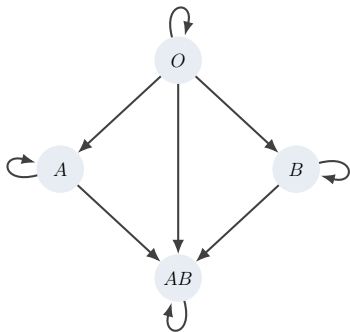
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# Kidney Exchange

- **Kidney transplantation** is the best treatment for end-stage renal disease, particularly when the kidney comes from a **living donor**.
- Patients often arrive with a willing donor who is **incompatible**.
  - **Compatibility** = ABO blood-type *and* tissue-type (cross-match).
- **Kidney exchange**: incompatible patient–donor pairs swap donors so that each patient gets a compatible kidney.
- Operating in the US, UK, Netherlands, Korea, Turkey, ...

# Compatibility

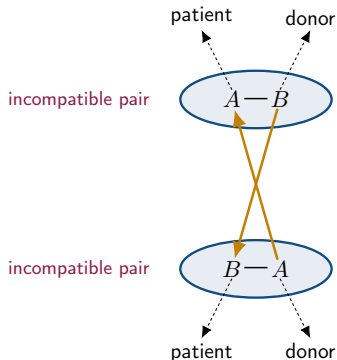
- Blood-type Compatibility:



$O \succ A, B \succ AB$  in donor-receiver compatibility.

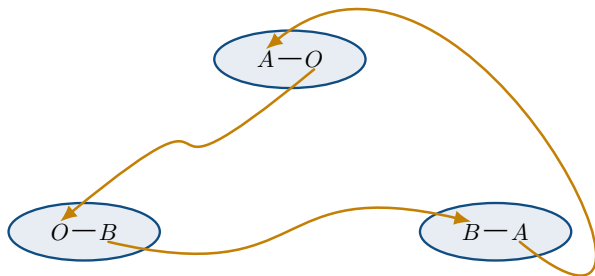
- Tissue-type Compatibility: Patient does not have pre-formed antibodies against donor's tissue-type DNA.

# Two-way Exchange



Denoted  $(A - B, B - A)$ . Surplus: 2 transplants.

# Three-way Exchange



Denoted  $(A - O, O - B, B - A)$ . Surplus: 3 transplants.

# Dynamic Nature of Kidney Exchange

- Pairs arrive with a Poisson process; arrival rates  $\lambda_i$ , total rate  $\lambda$ .
- Patients incur a waiting cost; continuous discount rate  $\rho$ .
- **Question.** How should a centralized clearinghouse match dynamically to maximize total **expected discounted surplus**?

**Core trade-off:** match *now* for a smaller, sooner surplus vs. wait for a *larger but later* (discounted) surplus.

# An Illustrative Trade-off

Suppose the pool contains a  $B - A$  pair and an  $A - O$  pair just arrived. A future  $A - B$  arrives stochastically.

Option	Exchange now	Exchange later	Surplus
1	$(B - A, A - O, O - B)$	—	3 now
2	$(A - O, O - A)$	$(B - A, A - B)$	2 now + 2 later

**Intuition.** Hold the  $B - A$  if and only if the marginal value of an additional  $B - A$  in the pool exceeds 1. **But that marginal value depends on the entire state.**

# Matching Environment

- $O - A, O - B$  are **abundant** (**underdemanded** pairs — matched on the spot at surplus 2, treated as use-it-or-lose-it).
- WLOG: no  $AB$  donor/patient (rare); no self-demanded pair ( $A - A, B - B, O - O$ ).

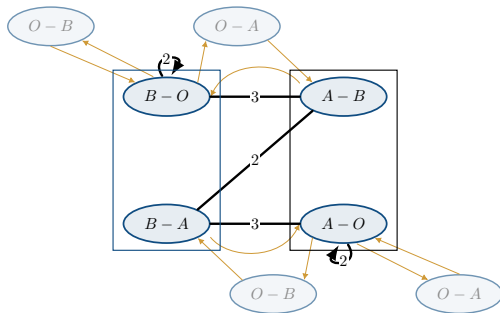
Four pair types control the state:

- $A - B$  and  $B - A$ : **reciprocal** pairs.
- $\mathbb{O}_{A-B} = \{B - O, AB - O, AB - A\}$ : 3-way ( $A - B, \mathbb{O}_{A-B}, O - A$ ), surplus 3.
- $\mathbb{O}_{B-A} = \{A - O, AB - O, AB - B\}$ : symmetric.

State

$$s = (s_1, s_2, s_3, s_4) \in \mathbb{N}^4 = (|\mathbb{O}_{A-B}|, |A - B|, |B - A|, |\mathbb{O}_{B-A}|).$$

# Linear Compatibility Structure



A local linear compatibility graph with surplus weights  $(2, 3, 2, 3, 2)$ .

# 1d Was Easy. 4d Is Not.

**1d case** (Ünver, 2010): overdemanded pairs matched on arrival, so  $s_1 = s_4 = 0$  and the state collapses to  $s_2 - s_3 \in \mathbb{Z}$ .

- A threshold mechanism is optimal. Driver: concavity in a one-dim state.

**General 4d problem:** all four state variables matter; state space is  $\mathbb{N}^4$ .

- Concavity alone is not propagated by the optimality operator.
- Even concavity plus sub-/supermodularity / strong superconcavity is not propagated.
- Prior queueing-theory methods (Koole, 1998) don't apply: their queueing basis  $e_j - e_{j-1}$  has no economic meaning in matching.

# Our Contributions

- **Methodology**: a tailored notion of  $\mathcal{D}^M$ -multimodularity (with a new “matching basis”  $\mathcal{D}^M$ ) that
  - is propagated by the optimality operator,
  - is just strong enough to imply all required second-order properties.
- **Structural result**: characterize the second-order shape of  $V^*$ .
- **Optimal mechanism**: in the (numerically generic) **unbalanced** regime, the optimal policy is a **multi-dimensional threshold mechanism** with three monotone thresholds.
- **Extension**: methodology applies to any dynamic matching with a **local linear compatibility graph**.

- 1 Introduction
- 2 Dynamic Model**
- 3 Second-order Properties
- 4 The Propagation Puzzle and  $\mathcal{D}^M$ -Multimodularity
- 5 Structure of the Optimal Mechanism
- 6 Concluding Remarks

# Model

- Types  $\mathcal{T} = \{1, \dots, n\}$ ,  
 local linear compatibility:  
 type  $i$  can match  $i \pm 1$  & endpoints with themselves.
- Poisson arrival rates  $\lambda_i$  in continuous time; let  
 overall arrival rate  $\lambda := \sum \lambda_i$ ,  
 prob. for next pair's type  $i$   $p_i := \lambda_i / \lambda$ ,  
 (discrete time  $\equiv$ ) discount  $\delta := \lambda / (\lambda + \rho) \in (0, 1)$ .
- Surplus  $a_{ij}$   
 $\in \{2, 3\}$  for kidney exchange,  
 $\in \mathbb{R}$  for other problems.
- State  $s \in \mathbb{N}^n$ .

# Bellman Equation

$\mathcal{M}(s, s')$ : max surplus along any exchange sequence  $s \rightarrow s'$ .

Bellman equation for optimal value function  $V^*$ :

$$V^*(s) = \delta \sum_i p_i \max_{s+e_i \rightarrow s'} \left\{ \mathcal{M}(s + e_i, s') + V^*(s') \right\}.$$

# Matching Vectors

A single feasible exchange transitions  $s$  by  $-w$  where

$$w \in \mathcal{D}^* := \{e_1, e_1 + e_2, e_2 + e_3, e_3 + e_4, e_4\}.$$

- $w = e_1$ :  $\mathbb{O}_{A-B}$  matched with itself (surplus 2).
- $w = e_1 + e_2$ : 3-way  $(\mathbb{O}_{A-B}, A - B)$  (surplus 3).
- $w = e_2 + e_3$ : 2-way  $(A - B, B - A)$  (surplus 2).
- $w = e_3 + e_4$ : 3-way  $(B - A, \mathbb{O}_{B-A})$  (surplus 3).
- $w = e_4$ :  $\mathbb{O}_{B-A}$  matched with itself (surplus 2).

# The Matching Operator and Other Event Operators

- Joint matching operator  $T_M : \mathcal{V} \rightarrow \mathcal{V}$ :

$$T_M f(s) = \max_{\substack{(\#_w)_{w \in \mathcal{D}^*} \in \{0, \dots, q\}^{|\mathcal{D}^*|} \\ s - \sum_w \#_w w \geq 0}} f\left(s - \sum_{w \in \mathcal{D}^*} \#_w w\right) + \sum_{w \in \mathcal{D}^*} \#_w a_w.$$

$\mathcal{V}$ : set of functions from  $\mathbb{N}^n$  to  $\mathbb{R}$ .

- Discount:  $(T_\delta f)(s) := \delta f(s)$ .
- Arrival of  $i$ :  $(T_{\text{arr}, i} f)(s) := f(s + e_i)$ .
- Uniformization:  $T_p(f_1, \dots, f_m)(s) := \sum_i p_i f_i(s)$ .

## Decomposition and Propagation

Observation (decomposition of dynamic programming operator  $T^*$ )

$$\begin{aligned} V^*(s) &= T^*V^*(s) \\ &= \left( T_\delta \circ T_p(T_M \circ T_{\text{arr},1}V^*, \dots, T_M \circ T_{\text{arr},n}V^*) \right)(s). \end{aligned}$$

Theorem (Fundamental Theorem of Dynamic Programming)

Define for each state  $s$ ,  $V_k(s) := T^*V_{k-1}(s)$  from arbitrary  $V_0 : \mathcal{V} \rightarrow \mathcal{V}$ . Then for each  $s$

$$V^*(s) = \lim_{k \rightarrow \infty} V_k(s),$$

such that  $V^*$  exists and is unique.

**Our approach.** We only need to propagate a property called “multimodularity” through  $T_M$ ; the other operators are linear and propagate any  $P(u, v)$ -style property. So we can **characterize properties** of  $V^*$  at the limit and **the optimal policy**.

- 1 Introduction
- 2 Dynamic Model
- 3 Second-order Properties**
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- 6 Concluding Remarks

# From Concavity to a Family of Properties

## Example

For a 1-d operator  $(Tf)(x) = \max\{f(x), f(x - e_i) + 1\}$ :

$$\text{match} \iff f(x) - f(x - e_i) \leq 1.$$

- LHS is the **marginal value** of a type- $i$  agent.
- Concavity in  $e_i \implies$  LHS is decreasing in  $s_i \implies$  **threshold**.

In **higher dimensions** we need more: substitution / complementarity between types, and “second-derivative” comparisons across coordinates.

# A Unified Framework: Property $P(u, v)$

## Definition

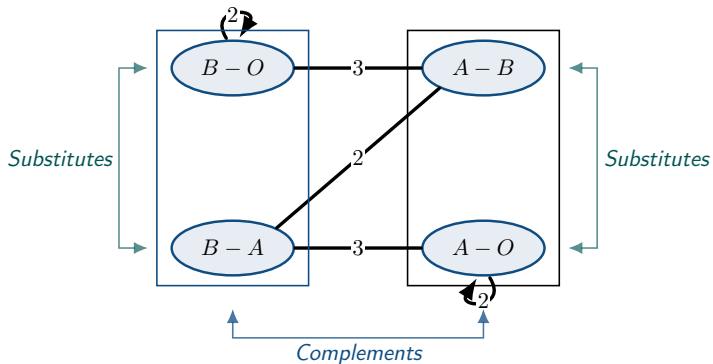
$f \in \mathcal{V}$  satisfies  $P(u, v)$  if, for every  $x$  with  $x + u, x + v, x + u + v \in \mathbb{N}^n$ ,

$$f(x + u) + f(x + v) \leq f(x) + f(x + u + v).$$

Named special cases (set  $u = e_i$ ):

- $v = -e_i$ : concavity in  $i$ .
- $v = -e_j$ :  $ij$ -submodularity —  $i$  and  $j$  are substitutes.
- $v = e_j$ :  $ij$ -supermodularity —  $i$  and  $j$  are complements.
- $v = \pm e_j - e_i$ : strong  $ij$ -superconcavity — nothing is a closer substitute to  $i$  than  $i$  itself.

## Substitutes and Complements in Kidney Exchange



**Side B**                      **Side A**  
 ('B' referring to  $B$  patients) ('A' referring to  $A$  patients)

- 1 Introduction
- 2 Dynamic Model
- 3 Second-order Properties
- 4 The Propagation Puzzle and  $\mathcal{D}^M$ -Multimodularity**
- 5 Structure of the Optimal Mechanism
- 6 Concluding Remarks

# Closedness

**Closedness condition:**  $\mathcal{P} = \{P(u, v) : (u, v) \in U\}$  is closed under  $T_w$  if

$$P(u, v) \in \mathcal{P} \implies P(u, v + w) \in \mathcal{P}.$$

**In 1d/2d:** {concavity} alone is not closed under matching operators; closing it requires strong superconcavity (and, with  $w = e_1 + e_2$ , also 12-supermodularity).

**Beyond 3 dimensions:** closure requires properties that are not concavity, sub-, super-, or strong superconcavity.

# $\mathcal{D}$ -Multimodularity (Hajek, 1985)

## Definition

- A set  $\mathcal{D} \subseteq \mathbb{Z}^n$  is a **multimodular basis** if  $\sum_{v \in \mathcal{D}} v = 0$  and  $\text{span}(\mathcal{D}) = \mathbb{Z}^n$ .
- $f$  is  **$\mathcal{D}$ -multimodular** if  $P(u, v)$  holds for every  $u, v \in \mathcal{D}$ .

## Lemma (disjoint-sum lemma)

for any two disjoint  $U, V \subseteq \mathcal{D}$  and admissible  $x$ ,

$$f(x + \sum_{u \in U} u) + f(x + \sum_{v \in V} v) \leq f(x) + f(x + \sum_{u \in U \cup V} u).$$

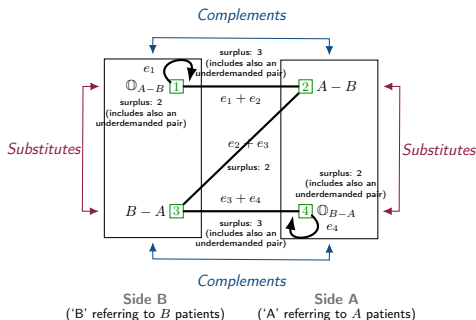
**Sums of disjoint subsets of  $\mathcal{D}$  are complements** — exactly what closedness needs.

# The Matching Basis $\mathcal{D}^M$

## Definition (matching basis)

$$\mathcal{D}^M := \{ e_1, -(e_1 + e_2), e_2 + e_3, -(e_3 + e_4), e_4 \}.$$

- Same vectors as  $\mathcal{D}^*$ , with **alternating signs** so they sum to 0.
- Each element corresponds to a feasible matching transition.
- Concavity, sub-/supermodularity and strong superconcavity all follow as corollaries.



# Main Propagation Result

Proposition (propagation under matching operator)

*If  $f \in \mathcal{V}$  is  $\mathcal{D}^M$ -multimodular, then so is  $T_M f$ .*

# Properties of the Value Function

## Theorem

The optimal value function  $V^*$  is  $\mathcal{D}^M$ -multimodular.

## Corollary

- *componentwise concave;*
- *strongly superconcave;*
- *$ij$ -submodular for  $i, j$  of the same parity:  $(\mathbb{O}_{A-B}, B - A)$  and  $(A - B, \mathbb{O}_{B-A})$  are *substitutes*;*
- *$ij$ -supermodular for  $i, j$  of different parity:  $(\mathbb{O}_{A-B}, A - B)$ ,  $(A - B, B - A)$ ,  $(B - A, \mathbb{O}_{B-A})$ ,  $(\mathbb{O}_{A-B}, \mathbb{O}_{B-A})$  are *complements*.*

- 1 Introduction
- 2 Dynamic Model
- 3 Second-order Properties
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# Unbalanced Dynamic Exchange

**Unbalanced:**  $A - B$  pairs arrive sufficiently more often than  $B - A$  pairs  $\implies$  at every reachable state,

- no  $\mathbb{O}_{B-A}$  pair waits in the pool ( $s_4 = 0$ ),
- the marginal value of an  $A - B$  pair is below 1.

**Numerically generic:** holds whenever  $p_{A-B} \neq p_{B-A}$  in the simulations of Ünver (2010).

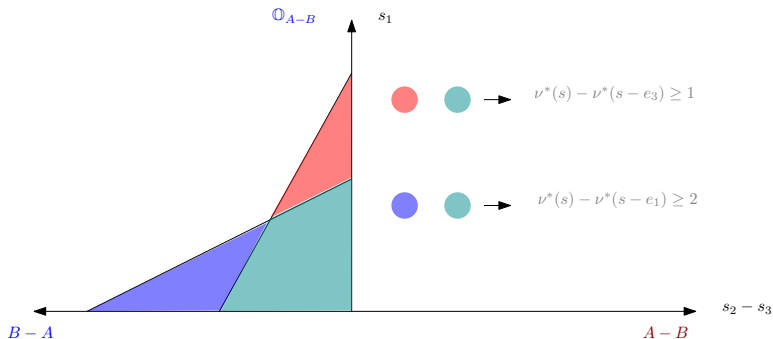
Reachable state space collapses to a 2-D picture in  $(s_2 - s_3, s_1)$ .

## Four Decisions, Three Thresholds

When the pool has no  $A - B$  pairs, the optimal mechanism faces **four decisions**, governed by three thresholds  $t^{1,3}$ ,  $t^2$ ,  $t^4$ :

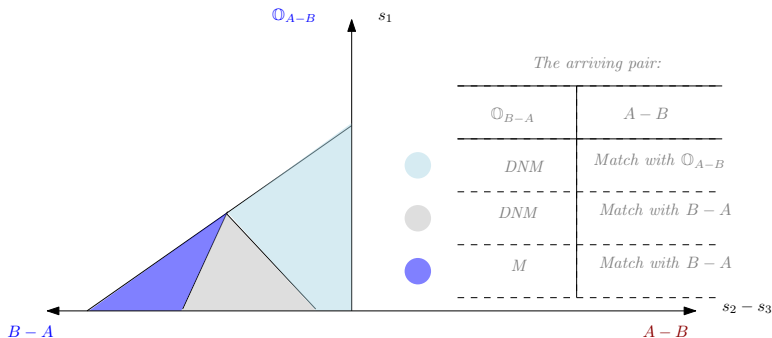
- $\mathbb{O}_{A-B}$  arrives. Keep if  $s_1 + 1 \leq t^{1,3}(s_3)$ , else match 2-way.
- $B - A$  arrives. Keep if  $s_1 \leq t^{1,3}(s_3 + 1)$ , else match an existing  $\mathbb{O}_{A-B}$  (keeping the  $B - A$ ).
- $\mathbb{O}_{B-A}$  arrives. 2-way if  $s_3 \leq t^4(s_1)$ , else 3-way with  $B - A$ .
- $A - B$  arrives with  $s_1, s_3 > 0$ . Match  $\mathbb{O}_{A-B}$  if  $s_3 \leq t^2(s_1)$ , else match  $B - A$ .

## Optimal Mechanism: Picture



- Triangular reachable region: existence follows from **concavity**.
- Boundary slope decreases: 1-3 **submodularity**.
- Slope in  $(-1, 0)$ : **strong superconcavity**.

## All Three Thresholds in One Picture



# Main Theorem (Optimal Mechanism)

## Theorem

For any unbalanced kidney exchange there exist threshold functions  $t^{1,3}, t^2, t^4 : \mathbb{N} \rightarrow \mathbb{N}$  such that the induced multi-dimensional threshold mechanism is dynamically optimal. Moreover,

- $t^{1,3}$  is *non-increasing*, with  $t^{1,3}(k+1) \geq t^{1,3}(k) - 1$ ;
- $t^2$  is *non-decreasing*;
- $t^4$  is *non-increasing*, with  $t^4(k+1) \geq t^4(k) - 1$ .





Each shape property maps to a second-order property of  $V^*$ , hence to  $\mathcal{D}^M$ -multimodularity.

- 1 Introduction
- 2 Dynamic Model
- 3 Second-order Properties
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- 5 Structure of the Optimal Mechanism
- 6 Concluding Remarks**

## Concluding Remarks

- **Methodological contribution:**  $\mathcal{D}^M$ -multimodularity — a closed family of second-order properties for dynamic matching with a local linear compatibility graph.
- **Core technical insight:** define  $T_M$  as a **single joint maximization**, not a composition of  $T_w$ 's.  
⇒ propagation of all required second-order properties *simultaneously*.
- **Economic structure:** in the (empirically generic) **unbalanced** regime, the optimal mechanism is a **multi-dimensional threshold mechanism** with three monotone thresholds.
- **Generality:** framework extends to any dynamic matching with a **local linear compatibility graph**.

# References

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