

# Mechanisms for Large-Scale Kidney Exchanges

(And More)

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October 13, 2010

# Chronic Kidney Disease

- ▶ Kidneys: filter wastes from the blood, among other things
- ▶ Chronic renal failure
  - ▶ life-threatening
  - ▶ usually irreversible
  - ▶ only treatment is replacement of kidney function
- ▶ Short run: dialysis
  - ▶ filtering outside the body, 3 to 7 times a week
  - ▶ low quality of life, low life expectancy
- ▶ Long run: transplantation of a healthy kidney

# Kidney Transplantation

data and more from a very nice talk by Al Roth, see <http://crcs.seas.harvard.edu/videos/#roth>

- ▶ Situation in the US:
  - ▶ more than 86,000 patients on the waiting list
  - ▶ 33,678 patients added in 2009
  - ▶ 10,441 transplants from deceased donors
  - ▶ 4,456 patients died while on the waiting list, 1,941 more removed as “too sick to transplant”

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  - ▶ 10,441 transplants from deceased donors
  - ▶ 4,456 patients died while on the waiting list, 1,941 more removed as “too sick to transplant”
- ▶ Good news:
  - ▶ live donation possible, without long-term effect on donor
  - ▶ many patients have somebody (usually a relative or friend) willing to donate them one of their kidneys

# Outline

Kidney Exchanges

Incentives in Large-Scale Kidney Exchanges

Results: Mechanisms for Large-Scale Kidney Exchanges

A Larger Agenda: Approximate Mechanisms without Payments

## Kidney Exchange

- ▶ Kidney exchange enables transplantations in cases where patient and potential donor are not compatible

patient 1  
blood type A

donor 1  
blood type B

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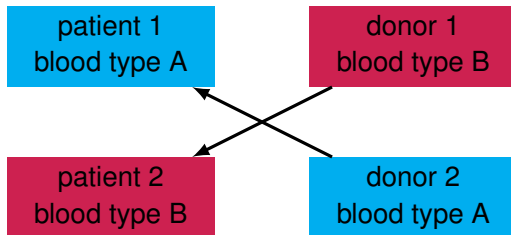
donor 1  
blood type B

patient 2  
blood type B

donor 2  
blood type A

## Kidney Exchange

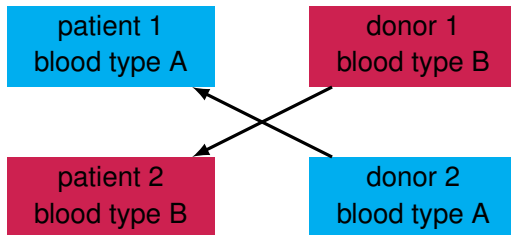
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## Kidney Exchange

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- ▶ 6,387 transplants from living donors in 2009 (first year with more living than deceased donors)

## Kidney Exchange

- ▶ One-to-one barter (item-to-item exchange) is hard
- ▶ Money would be one way to solve this, but buying and selling organs is illegal (National Organ Transplant Act, 1984)
- ▶ Amended in 2007 to allow kidney exchange, but we still need to bring (compatible) donors and patients together

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- ▶ New England Program for Kidney Exchange: since 2005, 14 transplant centers in New England
- ▶ Alliance for Paired Donation: since 2006, 81 transplant centers in 27 states
- ▶ A national exchange would clearly be desirable

## Incentives, Old and New

- ▶ Transplantations conducted simultaneously to control incentives (and other problems) within a single paired exchange
- ▶ In large-scale exchanges, incentives of hospitals become an issue (first pointed out by Roth, can be observed in practice)



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## The Formal Model

- ▶ Set  $N$  of agents, corresponding to hospitals
- ▶ Graph  $G = (V, E)$  with  $V = \biguplus_i V_i$ 
  - ▶ Each vertex  $v \in V$  corresponds to a donor-patient pair
  - ▶ Edge  $(u, v) \in E$  means donor of  $u$  is compatible with patient of  $v$ , and donor of  $v$  with patient of  $u$
- ▶ Agents report subsets  $V'_i \subseteq V_i$
- ▶ Mechanism produces matching  $M'$  of subgraph induced by  $\biguplus_i V'_i$
- ▶ Agent  $i$  adds matching  $\widehat{M}_i$  for hidden and unmatched vertices
- ▶ Utility of agent  $i$  is number of vertices in  $V_i$  matched in  $M'$  or  $\widehat{M}_i$

## Goals

- ▶ No payments
- ▶ Strategyproofness: agents cannot gain by hiding vertices
- ▶ Approximate optimality: call a mechanism  $\alpha$ -approximate if returned matching is at most by factor  $\alpha$  smaller than maximum cardinality matching ( $\alpha = 1$  is optimal, smaller is better)
- ▶ We are going to prove lower bounds (impossibility results) and upper bounds
  
- ▶ Complementary results: optimality on average with a weaker notion of stability (Ashlagi and Roth, 2010)

## Lower Bounds

**Theorem:** If there are at least two agents,  
no deterministic strategyproof mechanism can be  
 $\alpha$ -approximate for  $\alpha < 2$ , and  
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## A Deterministic Mechanism for Two Agents

- ▶ Choose matching that has
  - (i) maximum cardinality on both  $V_1$  and  $V_2$  and
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- ▶ Strategyproof (more on the next slide)
- ▶ 2-approximate, as returned matching is inclusion-maximal

## A Generalization

- ▶ Fix a bipartition  $\Pi = (\Pi_1, \Pi_2)$  of  $N$
- ▶  $\text{MATCH}_\Pi$ : choose matching that has
  - (i) maximum cardinality on  $V_i$  for all  $i \in N$
  - (ii) no edges between  $V_i$  and  $V_j$  if  $i, j \in \Pi_\ell$  for  $\ell \in \{1, 2\}$
  - (iii) maximum cardinality among all matchings satisfying (i) and (ii), breaking ties serially

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**Theorem:** For any number of agents and any bipartition  $\Pi$ ,  $\text{MATCH}_\Pi$  can be executed in polynomial time.

*Proof idea:* reduction to weighted matching

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- ▶  $\Pi = (\{1\}, \{2\})$  yields the two-agent mechanism we just saw
- ▶ Strategyproof and 2-approximate mechanism for two agents
- ▶ No finite approximation ratio for more than two agents

## Mix and Match

- ▶ **MIX-AND-MATCH:**
  1. Construct random bipartition  $\Pi = (\Pi_1, \Pi_2)$ : for each agent flip a fair coin to determine whether she goes to  $\Pi_1$  or  $\Pi_2$
  2. Execute  $\text{MATCH}_{\Pi}$

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**Theorem:** For any number of agents, **MIX-AND-MATCH** is (universally) strategyproof and 2-approximate in expectation.

## What We (Don't) Know

	deterministic		randomized	
	lower bound	upper bound	lower bound	upper bound
two agents	2	2	4/3	2
$n$ agents	2	$O(k)$	4/3	2

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FLIP-AND-MATCH: with probability 1/2 each

- ▶ execute MIX-AND-MATCH
- ▶ return a (specific) maximum cardinality matching

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- ▶ Ongoing work
  - ▶ simulations with true distribution of blood types
  - ▶ longer exchange sequences (three or four might be enough)



## Approximate Mechanisms without Payments

- ▶ VCG mechanism: socially optimal outcome, payments according to damage caused to respective other agents
- ▶ Legal, ethical, or technical reasons may prohibit payments
- ▶ Examples: kidney exchanges, voting
  
- ▶ Dekel, F, Procaccia (2008); Procaccia and Tennenholtz (2009): in the absence of payments, select outcome that is as close as possible to the optimal one in the worst case
- ▶ Same objective as in algorithmic mechanism design, but used due to incentive rather than computational constraints

## Strategyproof Selection from the Selectors

- ▶ Approval voting
  - ▶ each voter approves of set of candidates (of any size)
  - ▶ choose candidate (or committee of desired size) with largest number of votes
- ▶ Strategyproof (assuming dichotomous preferences)
  
- ▶ No longer the case when sets of candidates and voters coincide
  - ▶ scientific organizations (GTS, AMS, IEEE, IFAAMAS)
  - ▶ web graph, (directed) social networks, reputation systems

## The Model

- ▶ Set  $N = [n]$  of agents
- ▶ Directed graph  $G = (N, E) \in \mathcal{G}$ , no self-loops
- ▶ Ideally: select  $S \in \mathcal{S}_k = \{T \subseteq N : |T| = k\}$  to maximize
$$\sum_{i \in S} \deg(i) = \sum_{i \in S} |\{j \in N : (j, i) \in E\}|$$
- ▶ Mechanism  $M : \mathcal{G} \rightarrow \Delta(\mathcal{S}_k)$
- ▶ Strategyproofness: probability of selecting  $i$  independent of edges  $(i, j)$  for  $j \in N$

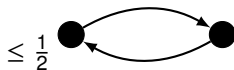
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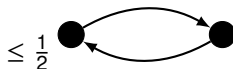
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upper bounds:  
mechanisms

lower bounds:  
impossibility results

## Bad News

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*Proof:* very cute



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Particularly surprising for  $k = n - 1$ : cannot guarantee to select unique agent receiving any votes

## Bounds for Randomized Mechanisms

**Theorem:** There exists a class of mechanisms, parameterized by  $m \geq 2$ , that is (universally) strategyproof for all  $n, k, m$ ,

- ▶ 4-approximate for  $m = 2$ , and
- ▶  $1 + O(1/k^{\frac{1}{3}})$ -approximate for  $m \approx k^{\frac{1}{3}}$ .

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Thank you!