

DM826 – Spring 2014  
Modeling and Solving Constrained Optimization Problems

## Exercises 6

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1. Assignment 1
2. Filtering in Scheduling

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2. Filtering in Scheduling

# Matching under preferences

Matching agents one to another, subject to various criteria.

Examples:

- junior doctors to hospitals
- pupils to schools
- kidney patients to donors

subject to **ordinal preferences** over a subset of the others. That is, there is a ranking or list of preferences with first choice, second choice, etc. The list need not be **strictly** ordered.

Typically other constraints: such as capacity

Relevant and large applications:

- in Hungary in 2011, 140 953 students applied for admission at universities
- In US National Resident Matching Program in 2012, 38 777 aspiring residents, 26 772 available positions.
- Free-for-all markets: free negotiations: issues of unraveling, congestion, exploiting offers

# Centralized Matching Schemes

## Centralized clearinghouses

Third party computes (automatically) **optimal matching**. Eg. maximizing number of places filled at hospitals, giving the maximum number of school-leavers their first-choice university, or ensuring no junior doctor and hospital have an incentive to reject their assignees and become matched together.

# Classification

- Bipartite matching problems with two-sided preferences
  - Stable Marriage problem (SM)
  - Hospitals Resident problem (HR) (many-one SM generalization)
  - Workers Firm problem, Student-Project Allocation problem

Optimality criteria: **Stability**: no two agents prefer another to one of their current assignees

- Bipartite matching problems with one-sided preferences
  - House Allocation problem (HA)
  - Capacited House Allocation Problem (CHA) (many-one HA generalization)

Optimality criteria: **Pareto optimality, popularity, profile-based optimality**

- Non-bipartite matching problems with preferences
  - Stable Roommates problem (SR)  
chess players, kidney exchanges patient-donor, P2P network
  - Stable Fixtures, S. Multiple Activities, S. Allocation (many-many)
  - Coalition Formation Game (partnerships of size  $> 2$ )

Optimality criteria: **Stability**

- Indifference in agents' lists, ie ties
- Incomplete/bounded lists
- Exchange stability: no pair of residents who could exchange one another's assigned hospitals so as to improve their outcome
- tripartite matching problem with preferences
- find all stable matchings
- find stable matching with other properties

- Seminal paper by Gale and Shapley (1962). polytime algorithm for SM
- D.E. Knuth. Stable Marriage and its Relation to Other Combinatorial Problems. American Mathematical Society, 1976 (translated to English 1997)
- D. Gusfield and R. Irving. The Stable Marriage Problem: Structure and Algorithms. MIT Press, 1989
- K. Iwama and S. Miyazaki. A Survey of the Stable Marriage Problem and Its Variants. International Conference on Informatics Education and Research for Knowledge-Circulating Society (ICKS), 2008
- D.F. Manlove. Algorithmics of Matching Under Preferences. World Scientific, 2013
- <http://optimalmatching.com/>



In economics (game-theory):

- matching theory as part of **market design** in microeconomics
- matching algorithm as a mechanism
- interest in **strategy-proof** or **truthful** mechanisms: make a **dominant strategy** for the agents to **reveal** their true preferences
- A.E. Roth and M. Sotomayor. Two-Sided Matching: A Study in Game-Theoretic Modeling and Analysis. Cambridge University Press, 1990

In CS

- Computational social choice theory (collective decisions) -> Algorithmic mechanism design (**social welfare**)
- Algorithmic Game Theory concerned with computational questions

# Stable Marriages

- A set of men  $U = \{m_1, \dots, m_{n_1}\}$
- A set of women  $W = \{w_1, \dots, w_{n_2}\}$
- $n_1 = n_2 = n$  (if not add men or women with empty preference list)
- In SM  $E = U \times W$ . In SMI  $E \subset U \times W$
- Each agent (man or woman)  $a_k \in U \cup W$  has a **preference list** in which it ranks agents from the other set in **strict order**
- Given any man  $m_i \in U$  and given any women  $w_j, w_k \in W$ ,  $w_j$  is said to **prefer**  $w_j$  to  $w_k$  ( $w_j \succ_{m_i} w_k$ ) if  $(m_i, w_j) \in E$ ,  $(m_i, w_k) \in E$  and  $w_j$  precedes  $w_k$  on  $m_i$ 's preference list (same mutatis mutandis for any man)
- $rank(m_i, w_j)$  is 1 plus the number of women that  $m_i$  prefers to  $w_j$ . (similarly for  $rank(w_j, m_i)$ )
- An **assignment** / **matching**  $M$  is a subset of  $E$ . For each  $a_k \in U \cup W$  the set of assignees of  $a_k$  is denoted by  $M(a_k)$ .  $|M(a_k)| = 1$  (else **unassigned**)
- **Underlying graph** of an instance  $\Pi$  of SM is a bipartite graph  $G = (U \cup W, E)$

### Definition (Stable Matching)

A pair  $(m_i, w_j) \in E \setminus M$  blocks a matching  $M$ , or is a blocking pair for  $M$ , if

- $m_i$  is unassigned or prefers  $w_j$  to  $M(m_i)$
- $w_j$  is unassigned or prefers  $m_i$  to  $M(w_j)$

A matching  $M$  is said to be stable if it admits no blocking pair.

# Stability Checking Algorithm

```
for  $m := 1$  to  $n$  do
  for each  $w$  such that  $m$  prefers  $w$  to  $M(m)$  do
    if  $w$  prefers  $m$  to  $M(w)$  then
      return unstable;
return stable;
```

# Gale Shapley algorithm (1962)

assign each person to be free

**while** some man  $m$  is free **do**

$w :=$  first woman on  $m$ 's list to whom  $m$  has not yet proposed

**if**  $w$  is free **then**

        | assign  $m$  and  $w$  to be engaged (to each other)

**else**

        | **if**  $w$  prefers  $m$  to her fiancé'  $m'$  **then**

            | assign  $m$  and  $w$  to be engaged and  $m'$  to be free

        | **else**

            |  $w$  rejects  $m$  (and  $m$  remains free)

Let  $P(m, w)$  be the set of women whom  $m$  strictly prefers to  $w$

Let  $\hat{P}(w, m)$  be the set of men such that  $w$  strictly prefers  $m$  to each man in  $\hat{P}(w, m)$

$$\sum_w x(m, w) = 1 \quad \text{for each man } m \quad (1)$$

$$\sum_m x(m, w) = 1 \quad \text{for each woman } w \quad (2)$$

$$x(m, w) \geq 0 \quad \text{for each pair } (m, w) \quad (3)$$

$$\sum_{m' \in \hat{P}(w, m)} x(m', w) - \sum_{w' \in \hat{P}(m, w)} x(m, w') \leq 0 \quad \text{for each pair } (m, w) \quad (4)$$

Length of preference lists  $l(m_i)$  and  $l(w_j)$ , respectively.

$$m = |E|$$

Variables:

- $2n$  variables:

$$x_i, i \in U, \text{ dom}(x_i) = \{1, 2, \dots, l(m_i)\} \cup \{n+1\}$$

$$y_j, j \in W, \text{ dom}(y_j) = \{1, 2, \dots, l(w_j)\} \cup \{n+1\}$$

( $x_i = p, 1 \leq p \leq l(m_i)$ ) then  $m_i$  marries the woman  $w_j$  such that  
 $\text{rank}(m_i, w_j) = p$ )

Constraints:

$$1. \quad x_i \geq p \implies y_j \leq p \quad 1 \leq i \leq n, 1 \leq p \leq l(m_i)$$

$$2. \quad y_j \geq q \implies x_i \leq q \quad 1 \leq j \leq n, 1 \leq q \leq l(w_j)$$

$$3. \quad y_j \neq q \implies x_i \neq p \quad 1 \leq j \leq n, 1 \leq q \leq l(w_j)$$

$$4. \quad x_i \neq p \implies y_j \neq q \quad 1 \leq i \leq n, 1 \leq p \leq l(m_i)$$

# Extensions

- Incomplete lists: handled by a slight modification of Gale Shapley algorithm.
- Ties: three stability notions:
  - **super-stability**  
blocking pair is defined as a pair  $(m, w)$  such that  $M(m) \neq w$ ,  $w \succeq_m M(m)$ , and  $m \succeq_w M(w)$ .
  - **strong stability**,  $(x, y)$  is a blocking pair if  $M(x) \neq y$ ,  $y \succ_x M(x)$ , and  $x \succeq_y M(y)$
  - **weak stability**. a blocking pair is defined as  $(m, w)$  such that  $M(m) \neq w$ ,  $w \succ_m M(m)$ , and  $m \succ_w M(w)$ .

super-stable matching  $\implies$  strongly stable, strongly stable  $\implies$  weakly stable.

Weakly stable matching always exists and can be found in polynomial time. In contrast, there are instances that have no super-stable nor strongly stable matching. Nevertheless, there is a polynomial time algorithm that decides if a super-stable (strongly stable, resp.) matching exists and finds one if any, whose running time is  $O(n^2)$   $O(n^3)$

- MAX SMTI (SM with ties + incomplete lists): finding a largest



# Course Instructor Problem

Extension of SM: find a stable matching that matches as many agents as possible, given an instance of SM where the preference lists may involve ties and may be incomplete and constraints on the number of assignees that agents can must obtain in a stable matching.

- A set of instructors  $S = \{s_1, \dots, s_{m_S}\}$
- A set of courses  $C = \{c_1, \dots, c_{m_C}\}$
- $m_C \leq m_S$
- Each course  $c_j \in C$  requires a number of assistants  $n_j$  (posts or capacity)
- Each instructor  $s_i \in S$  has a required minimum  $p_i$  and a required maximum  $q_i$  number of courses to receive
- There is a set  $E \subseteq S \times C$  of acceptable course-instructor pairs  
Each assistant has an acceptable set of courses  
 $A(s_i) = \{c_j \in C : (s_i, c_j) \in E\}$   
Each course has an acceptable set of instructors  
 $A(c_j) = \{s_i \in S : (s_i, c_j) \in E\}$  (this handles the semester issue and other issues)
- Each agent (instructors and courses)  $a_k \in S \cup C$  have a preference list in which it ranks  $A(a_k)$  in partial order

- Given any instructor  $s_i \in S$  and given any courses  $c_j, c_k \in C$ ,  $s_i$  is said to prefer  $c_j$  to  $c_k$  ( $c_j \succ_{s_i} c_k$ ) if  $(s_i, c_j) \in E$ ,  $(s_i, c_k) \in E$  and  $c_j$  precedes  $c_k$  on  $s_i$ 's preference list (same mutatis mutandis for any course)
- $rank(s_i, c_j)$  is 1 plus the number of courses that  $s_i$  prefers to  $c_j$ . (similarly for  $rank(c_j, s_i)$ )
- An assignment  $M$  is a subset of  $E$ . For each  $a_k \in S \cup C$  the set of assignees of  $a_k$  is denoted by  $M(a_k)$ . (unassigned, undersubscribed, full, oversubscribed)
- A matching is an assignment such that  $p_j \leq |M(s_i)| \leq q_j$  for all  $s_i \in S$  and  $|M(c_j)| = n_j$  for each  $c_j \in C$

## Definition (Stable Matching)

Let  $\Pi$  be an instance of the problem and let  $M$  be a matching in  $\Pi$ . A pair  $(s_i, c_j) \in E \setminus M$  blocks  $M$ , or is a **blocking pair** for  $M$  if:

1.  $s_i$  is undersubscribed or (strictly) prefers  $c_j$  to at least one member of  $M(s_i)$
2.  $c_j$  is undersubscribed or (strictly) prefers  $s_i$  to at least one member of  $M(c_j)$  (or both)

$M$  is said to be **stable** if it admits no blocking pair.

## Variables

- $x_i, 1 \leq i \leq \sum_{I \in C} n_I,$

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If  $L_J - E_{J \cup \{i\}} < p_i + p_J$ , then  $i \gg J$  (a)

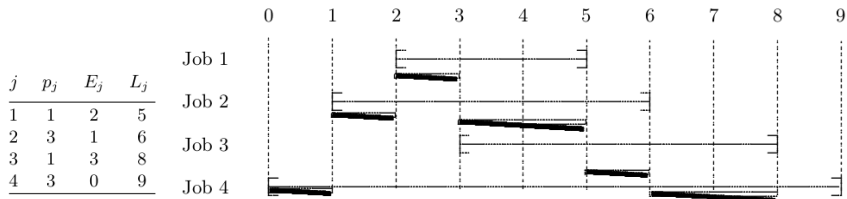
If  $L_{J \cup \{i\}} - E_J < p_i + p_J$ , then  $i \ll J$  (b)

If  $i \gg J$ , then update  $E_i$  to  $\max \left\{ E_i, \max_{J' \subset J} \{ E_{J'} + p_{J'} \} \right\}$ .

If  $i \ll J$ , then update  $L_i$  to  $\min \left\{ L_i, \min_{J' \subset J} \{ L_{J'} - p_{J'} \} \right\}$ .

$j$	$p_j$	$E_j$	$L_j$
1	1	2	5
2	3	1	6
3	1	3	8
4	3	0	9

# $O(n^2)$ algorithm



$i$	$J_i$	$\bar{p}$	$k$	$J_{ik}$	$L_k - E_i$	$p_i + \bar{p}J_{ik}$
1	{1, 2, 3, 4}	(1, 2, 1, 2)	4	{2, 3, 4}	9 - 2	1 + 5
			3	{2, 3}	8 - 2	1 + 3
			2	{2}	6 - 2	1 + 3
2	{1, 2, 3, 4}	(1, 3, 1, 2)	4	{1, 3, 4}	9 - 1	3 + 4
			3	{1, 3}	8 - 1	3 + 2
			1	{3}	5 - 1	3 + 1
3	{2, 3, 4}	(0, 2, 1, 2)	4	{2, 4}	9 - 3	1 + 4
			2	{2}	6 - 3	1 + 2
4	{1, 2, 3, 4}	(1, 3, 1, 3)	3	{1, 2, 3}	8 - 0	3 + 5
			2	{1, 2}	6 - 0	3 + 4

Conclude that  $4 \gg \{1, 2\}$  and update  $E_4$  from 0 to 5

If  $L_J - E_i < p_i + p_J$ , then  $\neg(i \ll J)$ . (a)

If  $L_i - E_J < p_i + p_J$ , then  $\neg(i \gg J)$ . (b)

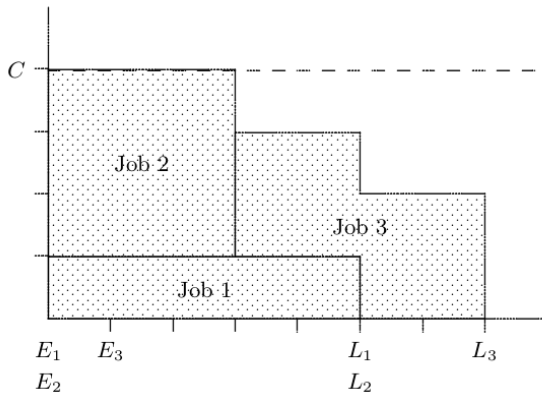
If  $\neg(i \ll J)$ , then update  $E_i$  to  $\max \left\{ E_i, \min_{j \in J} \{ E_j + p_j \} \right\}$  (a)

If  $\neg(i \gg J)$ , then update  $L_i$  to  $\min \left\{ L_i, \max_{j \in J} \{ L_j - p_j \} \right\}$  (b)



# Cumulative Scheduling

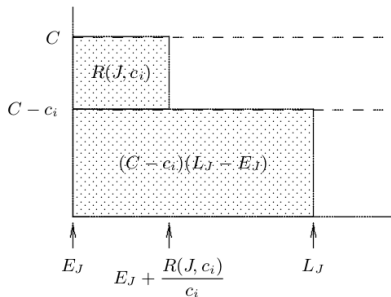
$j$	$p_j$	$c_j$	$E_j$	$L_j$
1	5	1	0	5
2	3	3	0	5
3	4	2	1	7



# Edge Finding

If  $e_i + e_J > C \cdot (L_J - E_{J \cup \{i\}})$ , then  $i > J$ . (a)

If  $e_i + e_J > C \cdot (L_{J \cup \{i\}} - E_J)$ , then  $i < J$ . (b)



If  $i > J$  and  $R(J, c_i) > 0$ , update  $E_i$  to  $\max \left\{ E_i, E_J + \frac{R(J, c_i)}{c_i} \right\}$ .

If  $i < J$  and  $R(J, c_i) > 0$ , update  $L_i$  to  $\min \left\{ L_i, L_J - \frac{R(J, c_i)}{c_i} \right\}$ .

# References