

# Chapter 1

Introduction: Some Representative Problems



## Matching Residents to Hospitals

Goal. Given a set of preferences among hospitals and medical school students, design a self-reinforcing admissions process.

Unstable pair: applicant x and hospital y are unstable if:

- x prefers y to its assigned hospital.
- y prefers x to one of its admitted students.

Stable assignment. Assignment with no unstable pairs.

- Natural and desirable condition.
- Individual self-interest will prevent any applicant/hospital deal from being made.

# 1.1 A First Problem: Stable Matching

## Stable Matching Problem

Goal. Given n men and n women, find a "suitable" matching.

- Participants rate members of opposite sex.
- Each man lists women in order of preference from best to worst.
- Each woman lists men in order of preference from best to worst.

	favorite ↓		least favori ↓	te		favorite ↓		least favorite ↓
	1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>			1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>
Xavier	Amy	Bertha	Clare		Amy	Yancey	Xavier	Zeus
Yancey	Bertha	Amy	Clare		Bertha	Xavier	Yancey	Zeus
Zeus	Amy	Bertha	Clare		Clare	Xavier	Yancey	Zeus
	Man's Drafa	namaa Drafili	_		И	laman'a Draf	ananaa Draf	1.

Men's Preference Profile

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Women's Preference Profile

## Stable Matching Problem

# Stable Matching Problem

Perfect matching: everyone is matched monogamously.

- Each man gets exactly one woman.
- Each woman gets exactly one man.

Stability: no incentive for some pair of participants to undermine assignment by joint action.

- In matching M, an unmatched pair m-w is unstable if man m and woman w prefer each other to current partners.
- Unstable pair m-w could each improve by eloping.

Stable matching: perfect matching with no unstable pairs.

Stable matching problem. Given the preference lists of n men and n women, find a stable matching if one exists.

## Q. Is assignment X-C, Y-B, Z-A stable?

	favorite ↓		least favorit ↓	e		favorite ↓		least favorite ↓
	1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>			1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>
Xavier	Amy	Bertha	Clare		Amy	Yancey	Xavier	Zeus
Yancey	Bertha	Amy	Clare		Bertha	Xavier	Yancey	Zeus
Zeus	Amy	Bertha	Clare		Clare	Xavier	Yancey	Zeus
		0 (1)						.,

Men's Preference Profile

Women's Preference Profile

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#### Stable Matching Problem

- Q. Is assignment X-C, Y-B, Z-A stable?
- A. No. Bertha and Xavier will hook up.

	favorite ↓		least favorite ↓	:		favorite ↓		least favorite ↓
	1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>			1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>
Xavier	Amy	Bertha	Clare		Amy	Yancey	Xavier	Zeus
Yancey	Bertha	Amy	Clare		Bertha	Xavier	Yancey	Zeus
Zeus	Amy	Bertha	Clare		Clare	Xavier	Yancey	Zeus

Men's Preference Profile

Women's Preference Profile

#### Stable Matching Problem

Q. Is assignment X-A, Y-B, Z-C stable?

A. Yes.

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	favorite ↓		least favorito ↓	e		favorite ↓		least favorite ↓
	1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>			1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>
Xavier	Amy	Bertha	Clare		Amy	Yancey	Xavier	Zeus
Yancey	Bertha	Amy	Clare		Bertha	Xavier	Yancey	Zeus
Zeus	Amy	Bertha	Clare		Clare	Xavier	Yancey	Zeus
,	Men's Prefer	rence Profile	2		W	'omen's Pref	erence Prot	ile

#### Stable Roommate Problem

Q. Do stable matchings always exist?

A. Not obvious a priori.

is core of market nonempty?

#### Stable roommate problem.

- 2n people; each person ranks others from 1 to 2n-1.
- . Assign roommate pairs so that no unstable pairs.

	1 <sup>st</sup>	$2^{nd}$	3 <sup>rd</sup>	
Adam	В	С	D	
Bob	С	А	D	$A-B, C-D \Rightarrow B-C$ unstable $A-C, B-D \Rightarrow A-B$ unstable
Chris	А	В	D	A-D, B-C $\Rightarrow$ A-C unstable
Doofus	А	В	С	

Observation. Stable matchings do not always exist for stable roommate problem.

## Proof of Correctness: Termination

Observation 1. Men propose to women in decreasing order of preference.

Observation 2. Once a woman is matched, she never becomes unmatched; she only "trades up."

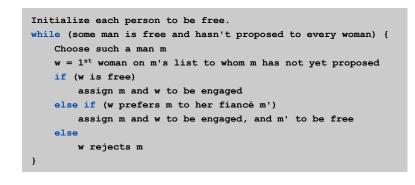
Claim. Algorithm terminates after at most  $n^2$  iterations of while loop. Pf. Each time through the while loop a man proposes to a new woman. There are only  $n^2$  possible proposals. •

	1st	2 <sup>nd</sup>	3 <sup>rd</sup>	4 <sup>th</sup>	5 <sup>th</sup>		1 <sup>st</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>	4 <sup>th</sup>	5 <sup>th</sup>
Victor	A	В	С	D	E	Amy	W	х	У	Z	v
Wyatt	В	С	D	Α	E	Bertha	х	У	z	V	W
Xavier	С	D	Α	В	E	Clare	У	Z	v	w	х
Yancey	D	Α	В	С	E	Diane	Z	v	W	х	У
Zeus	Α	В	С	D	E	Erika	V	W	х	У	Z

n(n-1) + 1 proposals required

## Propose-And-Reject Algorithm

Propose-and-reject algorithm. [Gale-Shapley 1962] Intuitive method that guarantees to find a stable matching.



Proof of Correctness: Perfection

Claim. All men and women get matched.

- Pf. (by contradiction)
- Suppose, for sake of contradiction, that Zeus is not matched upon termination of algorithm.
- Then some woman, say Amy, is not matched upon termination.
- By Observation 2, Amy was never proposed to.
- But, Zeus proposes to everyone, since he ends up unmatched.

## Proof of Correctness: Stability

/ order of preference

#### Claim. No unstable pairs.

- Pf. (by contradiction)
- Suppose A-Z is an unstable pair: each prefers each other to partner in Gale-Shapley matching S\*.

• Case 1: Z never proposed to A.  $\Rightarrow$  Z prefers his GS partner to A. men propose in decreasing **S**\* Amy-Yancey Bertha-Zeus

Case 2: Z proposed to A.

 $\Rightarrow$  A-Z is stable.

- $\Rightarrow$  A rejected Z (right away or later)
- ⇒ A prefers her GS partner to Z. ← women only trade up
- $\Rightarrow$  A-Z is stable.
- In either case A-Z is stable, a contradiction.

### **Efficient Implementation**

Efficient implementation. We describe  $O(n^2)$  time implementation.

#### Representing men and women.

- Assume men are named 1, ..., n.
- Assume women are named 1', ..., n'.

#### Engagements.

- Maintain a list of free men, e.g., in a queue.
- Maintain two arrays wife[m], and husband[w].
  - set entry to 0 if unmatched
  - if m matched to w then wife [m] =w and husband [w] =m

### Men proposing.

- For each man, maintain a list of women, ordered by preference.
- Maintain an array count [m] that counts the number of proposals made by man m.

#### Summary

Stable matching problem. Given n men and n women, and their preferences, find a stable matching if one exists.

Gale-Shapley algorithm. Guarantees to find a stable matching for any problem instance.

- Q. How to implement GS algorithm efficiently?
- Q. If there are multiple stable matchings, which one does GS find?

#### Efficient Implementation

#### Women rejecting/accepting.

Does woman w prefer man m to man m'?

for i

- For each woman, create inverse of preference list of men.
- Constant time access for each query after O(n) preprocessing.

Amy	1st	2 <sup>nd</sup>	3 <sup>rd</sup>	4 <sup>th</sup>	5 <sup>th</sup>	6 <sup>th</sup>	7 <sup>th</sup>	8 <sup>th</sup>
Pref	8	3	7	1	4	5	6	2
Amy	1	2	3	4	5	6	7	8
Inverse	4 <sup>th</sup>	8 <sup>th</sup>	2 <sup>nd</sup>	5 <sup>th</sup>	6 <sup>th</sup>	7 <sup>th</sup>	3 <sup>rd</sup>	1st

	,
ri=1ton	sir
<pre>inverse[pref[i]] = i</pre>	

Amy prefers man 3 to 6 nce inverse[3] < inverse[6] 2 7

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#### Understanding the Solution

Q. For a given problem instance, there may be several stable matchings. Do all executions of Gale-Shapley yield the same stable matching? If so, which one?

#### An instance with two stable matchings.

- A-X, B-Y, C-Z.
- A-Y, B-X, C-Z.

	1st	2 <sup>nd</sup>	3 <sup>rd</sup>		1st	2 <sup>nd</sup>	3 <sup>rd</sup>
Xavier	Α	В	С	Amy	У	Х	Z
Yancey	В	Α	С	Bertha	х	У	Z
Zeus	Α	В	С	Clare	х	У	Z

#### Understanding the Solution

Q. For a given problem instance, there may be several stable matchings. Do all executions of Gale-Shapley yield the same stable matching? If so, which one?

Def. Man m is a valid partner of woman w if there exists some stable matching in which they are matched.

Man-optimal assignment. Each man receives best valid partner.

Claim. All executions of GS yield man-optimal assignment, which is a stable matching!

- No reason a priori to believe that man-optimal assignment is perfect, let alone stable.
- Simultaneously best for each and every man.

#### Man Optimality

#### Claim. GS matching S\* is man-optimal.

#### Pf. (by contradiction)

- Suppose some man is paired with someone other than best partner. Men propose in decreasing order of preference  $\Rightarrow$  some man is rejected by valid partner. s
- Let Y be first such man, and let A be first valid woman that rejects him.
- Let S be a stable matching where A and Y are matched.
- When Y is rejected, A forms (or reaffirms) engagement with a man, say Z, whom she prefers to Y.
- Let B be Z's partner in S.
- . Z not rejected by any valid partner at the point when Y is rejected by A. Thus, Z prefers A to B. since this is first rejection
- But A prefers Z to Y.
- Thus A-Z is unstable in S.

#### Stable Matching Summary

Stable matching problem. Given preference profiles of n men and n women, find a stable matching.

> no man and woman prefer to be with each other than assigned partner

Gale-Shapley algorithm. Finds a stable matching in  $O(n^2)$  time.

Man-optimality. In version of GS where men propose, each man receives best valid partner.

> w is a valid partner of m if there exist some stable matching where m and w are paired

Q. Does man-optimality come at the expense of the women?

Amy-Yancey

Bertha-Zeus

by a valid partner

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#### Woman Pessimality

Woman-pessimal assignment. Each woman receives worst valid partner.

Claim. GS finds woman-pessimal stable matching S\*.

#### Pf.

- Suppose A-Z matched in S\*, but Z is not worst valid partner for A.
- There exists stable matching S in which A is paired with a man, say Y, whom she likes less than Z.
- Let B be Z's partner in S.
- Z prefers A to B. ← man-optimality
- Thus, A-Z is an unstable in S.



s

#### Extensions: Matching Residents to Hospitals

Ex: Men  $\approx$  hospitals, Women  $\approx$  med school residents.

Variant 1. Some participants declare others as unacceptable.

Variant 2. Unequal number of men and women.

Variant 3. Limited polygamy.

hospital X wants to hire 3 residents

resident A unwilling to

work in Cleveland

- Def. Matching S unstable if there is a hospital h and resident r such that:
- . h and r are acceptable to each other; and
- either r is unmatched, or r prefers h to her assigned hospital; and
- either h does not have all its places filled, or h prefers r to at least one of its assigned residents.

#### Application: Matching Residents to Hospitals

#### NRMP. (National Resident Matching Program)

- Original use just after WWII. ← predates computer usage
- Ides of March, 23,000+ residents.

#### Rural hospital dilemma.

- Certain hospitals (mainly in rural areas) were unpopular and declared unacceptable by many residents.
- Rural hospitals were under-subscribed in NRMP matching.
- . How can we find stable matching that benefits "rural hospitals"?

Rural Hospital Theorem. Rural hospitals get exactly same residents in every stable matching!

#### Lessons Learned

#### Powerful ideas learned in course.

- Isolate underlying structure of problem.
- Create useful and efficient algorithms.

Potentially deep social ramifications. [legal disclaimer]

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# Interval Scheduling

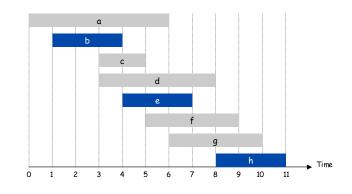
# 1.2 Five Representative Problems

Input. Set of jobs with start times and finish times. Goal. Find maximum cardinality subset of mutually compatible jobs.

jobs don't overlap

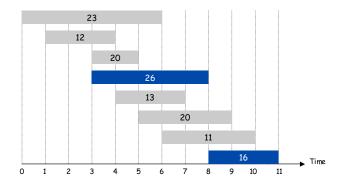
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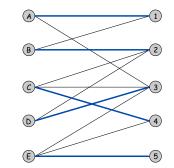
Weighted Interval Scheduling

Input. Set of jobs with start times, finish times, and weights. Goal. Find maximum weight subset of mutually compatible jobs.



# **Bipartite Matching**

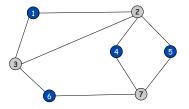
Input. Bipartite graph. Goal. Find maximum cardinality matching.



# Independent Set

Input. Graph. Goal. Find maximum cardinality independent set.

> subset of nodes such that no two joined by an edge



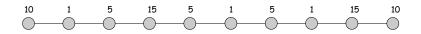
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# **Competitive Facility Location**

Input. Graph with weight on each each node. Game. Two competing players alternate in selecting nodes. Not allowed to select a node if any of its neighbors have been selected.

Goal. Select a maximum weight subset of nodes.



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Second player can guarantee 20, but not 25.

Five Representative Problems

Variations on a theme: independent set.

Interval scheduling: n log n greedy algorithm. Weighted interval scheduling: n log n dynamic programming algorithm. Bipartite matching: n<sup>k</sup> max-flow based algorithm. Independent set: NP-complete. Competitive facility location: PSPACE-complete.