# CSE 421: Introduction to Algorithms

3

## **Stable Matching**

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# 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?
- Q: How many stable matchings are there?

### Propose-And-Reject Algorithm [Gale-Shapley'62]

```
Initialize each person to be free.
while (some man is free and hasn't proposed to every woman) {
    Choose such a man m
    W = 1<sup>st</sup> woman on m's list to whom m has not yet proposed
    if (w is free)
        assign m and w to be engaged
    else if (w prefers m to her fiancé m')
        assign m and w to be engaged, and m' to be free
    else
        w rejects m
}
```

# Implementation of GS Algorithm

### Problem size

N=2n<sup>2</sup> words

2n people each with a preference list of length n

### Brute force algorithm

Try all **n**! possible matchings Do any of them work?

### **Gale-Shapley Algorithm**

n<sup>2</sup> iterations, each costing constant time as follows:

# **Efficient Implementation**

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

#### Representing men and women:

Assume men are named 1, ..., n. Assume women are named n+1, ..., 2n.

### 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**.

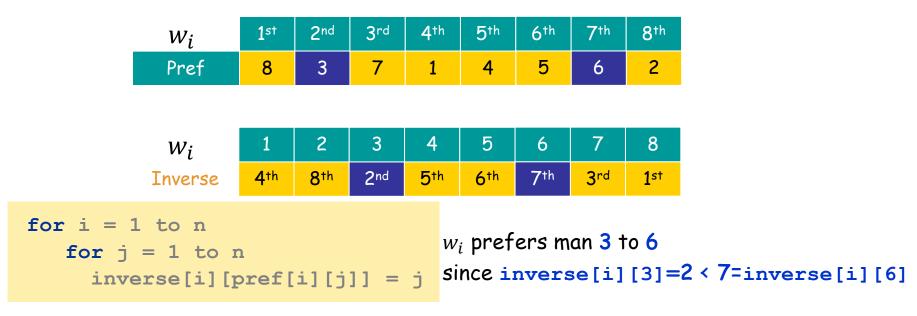
# **Efficient Implementation**

### Women rejecting/accepting.

Does woman w prefer man m to man m'?

For each woman, create inverse of preference list of men.

Constant time access for each query after O(n) preprocessing per woman. O(n<sup>2</sup>) total reprocessing cost.



# Summary

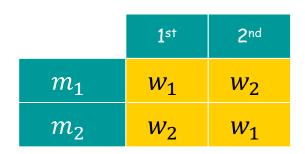
- Stable matching problem: Given n men and n women, and their preferences, find a stable matching if one exists.
- Gale-Shapley algorithm guarantees to find a stable matching for any problem instance.
- GS algorithm finds a stable matching in O(n<sup>2</sup>) time.
- Q: If there are multiple stable matchings, which one does GS find?
- Q: How many stable matchings are there?

# 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:

- $(m_1, w_1), (m_2, w_2).$
- $(m_1, w_2), (m_2, w_1).$



	1st	2 <sup>nd</sup>
<i>w</i> <sub>1</sub>	$m_2$	$m_1$
<i>w</i> <sub>2</sub>	$m_1$	$m_2$

# Man Optimal Assignments

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

Man-optimal matching: Each man receives the best valid partner (according to his preferences).

• Not that each man receives his most favorite woman.

# Example

### Here

Valid-partner $(m_1) = \{w_1, w_2\}$ Valid-partner $(m_2) = \{w_1, w_2\}$ Valid-partner $(m_3) = \{w_3\}$ .

### Man-optimal matching $\{m_1, w_1\}, \{m_2, w_2\}, \{m_3, w_3\}$

	favorite least favorite ↓ ↓		2	favorite ↓			least favorite ↓	
	1 <sup>s†</sup>	2 <sup>nd</sup>	3 <sup>rd</sup>			1 <sup>s†</sup>	2 <sup>nd</sup>	3rd
$m_1$	<i>w</i> <sub>1</sub>	<i>w</i> <sub>2</sub>	<i>W</i> <sub>3</sub>		<i>w</i> <sub>1</sub>	$m_2$	$m_1$	$m_3$
$m_2$	<i>w</i> <sub>2</sub>	<i>w</i> <sub>1</sub>	<i>W</i> <sub>3</sub>		<i>W</i> <sub>2</sub>	$m_1$	$m_2$	$m_3$
$m_3$	<i>w</i> <sub>1</sub>	<i>w</i> <sub>2</sub>	<i>w</i> <sub>3</sub>		<i>W</i> <sub>3</sub>	$m_1$	$m_2$	$m_3$

# Man Optimal Assignments

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

Man-optimal matching: Each man receives the best valid partner (according to his preferences).

• Not that each man receives his most favorite woman.

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

- So, output of GS is unique!!
- No reason a priori to believe that man-optimal matching is perfect, let alone stable.

# Man Optimality

### (*m*,*w*) (*m*',*w*') ...

S

Claim: GS matching **S**\* is man-optimal. **Proof:** (by contradiction)

Suppose some man is paired with someone other than his best partner. Men propose in decreasing order of preference  $\Rightarrow$  some man is rejected by a valid partner.

Let m be the man who is the first such rejection, and let w be the women who is first valid partner that rejects him.

Let **S** be a stable matching where m and w are matched. In building **S**\*, when m is rejected, w forms (or reaffirms) engagement with a man, say m' whom she prefers to m.

Let w' be m' partner in **S**.

In building S<sup>\*</sup>, m' is not rejected by any valid partner at the point when m is rejected by w. Thus, m' prefers w to w'.

But w prefers m' to m.

Thus (m', w) is unstable in **S**.

since this is the first rejection by a valid partner

# Man Optimality Summary

Man-optimality: In version of GS where men propose, each man receives the 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?

# Woman Pessimality

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

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

### Proof.

Suppose (m, w) matched in S\*, but m is not worst valid partner for w. There exists stable matching S in which w is paired with a man, say m', whom she likes less than m.

Let w' be m partner in **S**.

*m* prefers *w* to w'. *man-optimality* of S\*

Thus, (m, w) is an unstable in **S**.

# 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.
- GS algorithm finds a stable matching in O(n<sup>2</sup>) time.
- GS algorithm finds man-optimal woman pessimal matching
- Q: How many stable matching are there?

# How many stable Matchings?

We already show every instance has at least 1 stable matchings.

There are instances with about  $b^n$  stable matchings for b > 2

[Karlin-O-Weber'17]: Every instance has at most  $c^n$  stable matchings for some c > 2

[Research-Question]:

Is there an "efficient" algorithm that chooses a uniformly random stable matching of a given instance.

## Extensions: Matching Residents to Hospitals

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

- Variant 1: Some participants declare others as unacceptable.
- Variant 2: Unequal number of men and women.

e.g. A resident not interested in Cleveland

• Variant 3: Limited polygamy.

e.g. A hospital wants to hire 3 residents

Def: Matching **S** is unstable if there is hospital **h** and resident **r** s.t.

- 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.

## Lessons Learned

- Powerful ideas learned in course.
  - Isolate underlying structure of problem.
  - Create useful and efficient algorithms.
- Potentially deep social ramifications. [legal disclaimer]
  - Historically, men propose to women. Why not vice versa?
  - Men: propose early and often.
  - Men: be more honest.
  - Women: ask out the guys.
  - Theory can be socially enriching and fun!

## "The Match": Doctors and Medical Residences

- Each medical school graduate submits a ranked list of hospital where he wants to do a residency
- Each hospital submits a ranked list of newly minted doctors
- A computer runs stable matching algorithm (extended to handle polygamy)
- Until recently, it was hospital-optimal.



# History

### 1900

- Idea of hospital having residents (then called "interns")
   1900-1940s
- Intense competition among hospitals
  - Each hospital makes offers independently
  - Process degenerates into a race; hospitals advancing date at which they finalize binding contracts

### 1944

 Medical schools stop releasing info about students before a fixed date

### 1945-1949

- Hospitals started putting time limits on offers
  - Time limits down to 12 hours; lots of unhappy people

## "The Match"

### 1950

- NICI run a centralized algorithm for a trial run
- The pairing was not stable, Oops!!

### 1952

- The algorithm was modified and adopted. It was called the Match.
- The first matching produced in April 1952

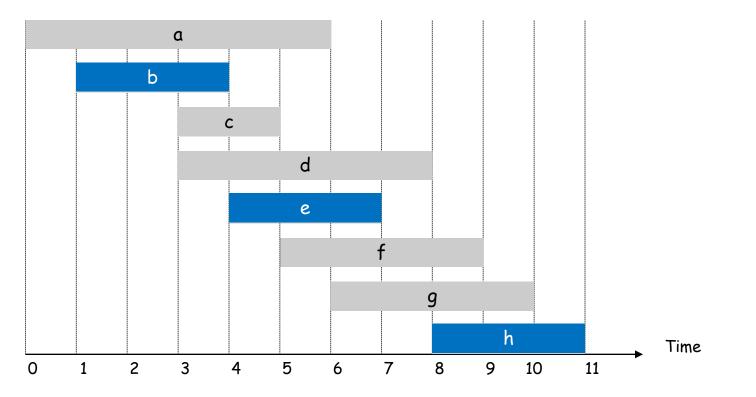
## **Five Representative Problems**

- 1. Interval Scheduling
- 2. Weighted Interval Scheduling
- 3. Bipartite Matching
- 4. Independent Set Problem
- 5. Competitive Facility Location

## **Interval Scheduling**

Input: Given a set of jobs with start/finish times

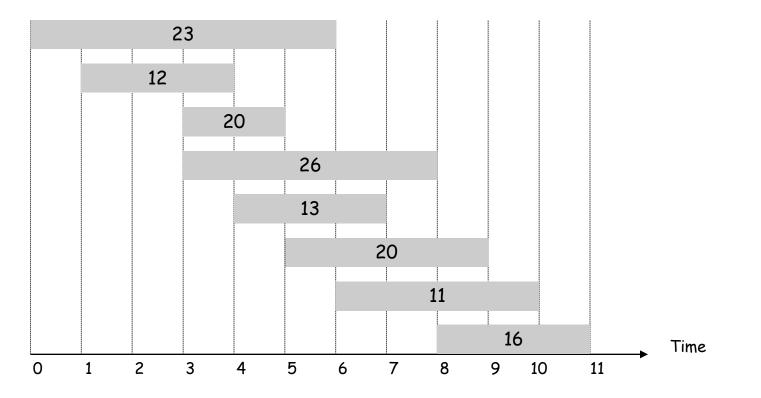
Goal: Find the maximum cardinality subset of jobs that can be run on a single machine.



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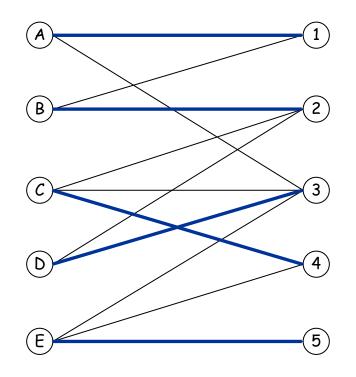
Goal: Find the maximum weight subset of jobs that can be run on a single machine.



## **Bipartite Matching**

Input: Given a bipartite graph

Goal: Find the maximum cardinality matching

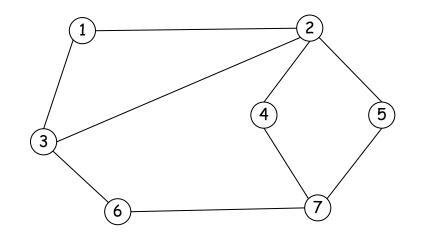


## Independent Set

Input: A graph

Goal: Find the maximum independent set

Subset of nodes that no two joined by an edge

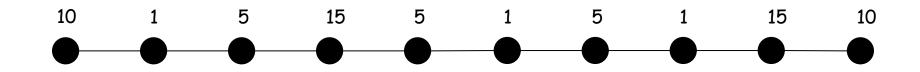


## **Competitive Facility Location**

Input: Graph with weight on 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. Does player 2 have a strategy which guarantees a total value of *V* no matter what player 1 does?



Second player can guarantee 20, but not 25.

## **Five Representative Problems**

Variation of a theme: Independent set Problem

- 1. Interval Scheduling *n* log *n* greedy algorithm
- 2. Weighted Interval Scheduling *n* log *n* dynamic programming algorithm
- 3. Bipartite Matching  $n^k$  maximum flow based algorithm
- 4. Independent Set Problem: NP-complete
- 5. Competitive Facility Location: PSPACE-complete