# CS 105: DIC on Discrete Structures 

## Graph theory

Stable matchings, and the end.

Lecture 35<br>Nov 092023

## Topic 3: Graph theory

## Topics in Graph theory

1. Basics concepts and definitions.
2. Eulerian graphs: Using degrees of vertices.
3. Bipartite graphs: Using odd length cycles.
4. Connected components: Using cycles.
5. Maximum matchings: Using augmenting paths.
6. Perfect matchings in bipartite graphs: Using neighbour sets. - Hall's theorem
7. Applications of Hall's theorem: Minimum vertex covers -Konig-Egervary's theorem
8. Stable matchings...

## Stable matchings

Boys

- 1
- 2
- 3
- 4
- 5

Girls

- $A$
- $B$
- $C$
- $D$
- $E$


## Stable matchings

$$
\begin{aligned}
\text { Boys } & \text { Girls } \\
C>B>E>A>D \cdot 1 & \bullet A: 35214 \\
A B E C D \cdot 2 & \bullet B: 52143 \\
D C B A E \cdot 3 & \bullet C: 43512 \\
A C D B E \cdot 4 & \bullet D: 12345 \\
A B D E C \cdot 5 & \bullet E: 23415
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$A B D E C \cdot 5 \longrightarrow E: 23415$

- Let us try a "greedy" marriage strategy for boys.
- Danger! 4 prefers $C$ to $B$ and $C$ prefers 4 to 1 . Divorce!
- Qn: Can you match everyone without such Rogue couples?!


## More than just a funny puzzle

- College admissions: Original Gale and Shapley paper, 1962.
- Matching hospitals and residents.
- Matching dancing partners.
- Matching students with jobs.


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- College admissions: Original Gale and Shapley paper, 1962.
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- Matching students with jobs.
- Matching (PG) TAs with courses.
- JEE algorithm...


## Stable matchings

## Definition

Given a matching $M$ in a graph with preference lists of nodes.

- Unstable pair: Two vertices $x, y$ such that $x$ prefers $y$ to its assigned vertex and vice versa.
- $x, y$ would be happier by eloping.
- Qn: Find a perfect matching with no unstable pairs. Such a matching is called a Stable Matching.


## Roommates Problem

- $A: B C D$
- $B: C A D$
- $C: A B D$

- $D$
- $D: A B C$
- What can you observe from this?


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- $A: B C D$
- $B: C A D$
- $C: A B D$

- D
- $D: A B C$
- What can you observe from this?
- Everybody hates $D$.


## Roommates Problem

- $A: B C D$
- $B: C A D$
- $C: A B D$

- $D$
- $D: A B C$
- What can you observe from this?
- Stable matchings don't always exist.


## Roommates Problem

- $A: B C D$
- B : CAD
- $C: A B D$

- $D$
- $D: A B C$
- What can you observe from this?
- Stable matchings don't always exist.
- So, do they exist for bipartite graphs and how can we prove this?


## The proposal algorithm

Given: bipartite graph, preference list for $n \mathrm{men} /$ women

- 8am: Every man goes to first woman on his list not yet crossed off, and proposes to her!


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- Does this algorithm terminate?
- If yes, does it produce a stable matching when it terminates?


## Termination and Correctness of the proposal algo

- Try out the algo on the example.


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- The algo terminates within $n^{2}$ days.
- For each day (except last), at least one woman is crossed off some man's list.
- As there are $n$ men and each has list of size $n$, algo must terminate in $n^{2}$ days.


## Termination and Correctness of the proposal algo

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- $W^{\prime}$ rejected him only because she preferred some $M^{\prime \prime}$ to $M$.
- By Lemma 2, she likes her final partner at least as much as $M^{\prime \prime}$, so better than $M$.


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- If $(M, W)$ is pair in current matching, s.t., $M$ prefers $W^{\prime}$.
- We will show that $W^{\prime}$ prefers some other $M^{\prime}$ and hence no unstable pair.
- Thus no man can be part of an unstable pair, implies stable matching.


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Conclusion: Propose first!

## Further reading

- Many questions, rich theory.
- How many stable marriages are possible?
- Can you do better by lying? Boys - no!, Girls - yes!
- What if there are brother-sisters (who should not be matched!)?


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- D. Gale and L.S. Shapley, College Admissions and the Stability of Marriage, American Mathematical Monthly 69(1962), pp. 9-14.
- D. Gusfield and R.W. Irving, The Stable Marriage Problem: Structure and Algorithms, MIT Press, 1989.
The 2012 Nobel prize in Economics to Shapley and Roth: "for the theory of stable allocations and the practice of market design".


## Summary

## What we covered in this course

1. Mathematical proofs and basic structures
2. Counting and combinatorics
3. Introduction to graph theory

## Summary: Part 1

- Mathematical proofs and reasoning
- Basic discrete structures


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- How to reason and write proofs formally
- Propositions, predicates
- Proof techniques: contradiction, contrapositive
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- Basic discrete structures
- Sets: finite and infinite sets, countable and uncountable sets
- Functions: bijections (from e.g., $\mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$ ), injections and surjections, Cantor's diagonalization technique
- Relations: equivalence relations and partitions; partial orders, chains, anti-chains, lattices


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- Applications: showing impossibility theorems for CS, parallel task scheduling algorithms.


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- Basic counting principles, double counting
- Binomial theorem, permutations and combinations, Estimating $n$ !
- Recurrence relations and generating functions
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- Pigeon-Hole Principle (PHP) and its applications.
- Some special numbers and sequences: Fibonacci, Catalan
- Introduction to Ramsey theory


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## Topics in Graph theory

- Basics: graphs, paths, cycles, walks, trails, ...
- Graph representations, isomorphisms and automorphisms.
- Matchings: maximal, maximum, perfect and stable.


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7. Stable matchings... and the Gale Shapley Algo

## Beyond this course

Topics we didn't cover in discrete structures

- Number theory and cryptographic applications
- Discrete Probability Theory
- Symbolic logic and its applications
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Many core courses and electives to choose from

