CS 105: DIC on Discrete Structures

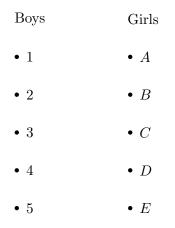
Graph theory Stable matchings, and the end.

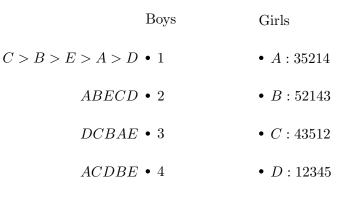
> Lecture 39 Nov 05 2024

Topic 3: Graph theory

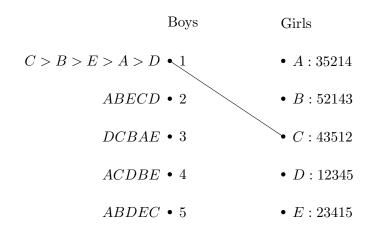
Topics in Graph theory

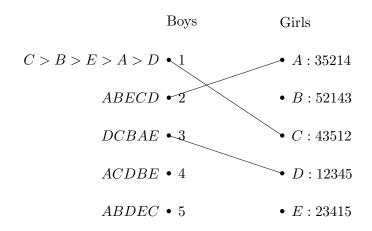
- 1. Basics concepts and definitions.
- 2. Eulerian graphs: Using degrees of vertices.
- 3. Bipartite graphs: Using odd length cycles.
- 4. Connected components: Using cycles.
- 5. Maximum matchings: Using augmenting paths.
- 6. Perfect matchings in bipartite graphs: Using neighbour sets. Hall's theorem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

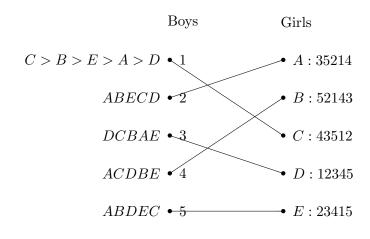


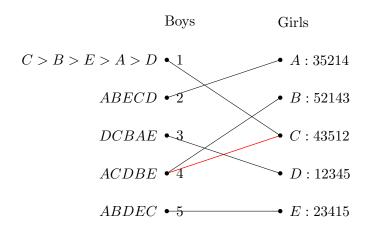


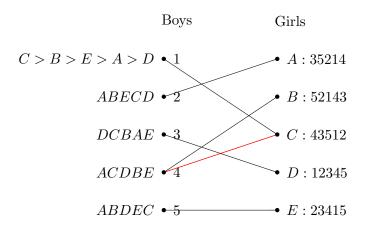
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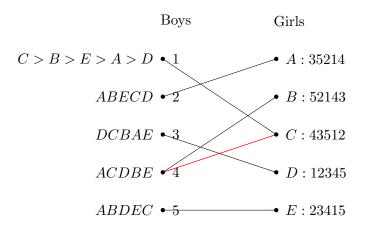






► Let us try a "greedy" marriage strategy for boys.

▶ Danger! 4 prefers C to B and C prefers 4 to 1. Divorce!



- ▶ Danger! 4 prefers C to B and C prefers 4 to 1. Divorce!
- ▶ Qn: Can you match everyone without such <u>Rogue</u> 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.

More than just a funny puzzle

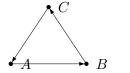
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- ▶ Matching students with jobs.
- ▶ Matching (PG) TAs with courses.
- ▶ JEE algorithm...

Definition

Given a matching ${\cal M}$ 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.
- \blacktriangleright x, y would be happier by eloping.
- Qn: Find a perfect matching with no unstable pairs. Such a matching is called a Stable Matching.

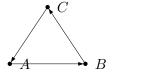
- $\bullet \ A:BCD$
- B:CAD
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• D

- D:ABC
- ▶ What can you observe from this?

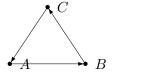
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- \blacktriangleright Everybody hates D.

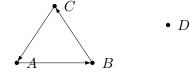
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- ▶ What can you observe from this?
- ▶ Stable matchings don't always exist.
- So, do they exist for bipartite graphs and how can we prove this?

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

▶ Try out the algo on the example.

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 - 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 n² days.

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- W' rejected him only because she preferred some M'' to M.
- By Lemma 2, she likes her final partner at least as much as M", so better than M.

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Theorem

- ▶ If (M, W) is pair in current matching, s.t., M prefers W'.
- We will show that W' prefers some other M' and hence no unstable pair.
- Thus no man can be part of an unstable pair, implies stable matching.

The proposal algorithm: who does better?

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Conclusion: Propose first!

Further reading

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- ▶ How many stable marriages are possible?
- ▶ Can you do better by lying? Boys no!, Girls yes!
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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".

What we covered in this course

- 1. Mathematical proofs and basic reasoning
- 2. Basic discrete structures
- 3. Counting and combinatorics
- 4. Introduction to graph theory

▶ Part 1: Mathematical proofs and reasoning



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- ▶ How to reason and write proofs formally
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- ▶ Proof techniques: contradiction, contrapositive
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- ▶ Sets: finite and infinite sets, countable and uncountable sets
- ▶ Functions: bijections (from e.g., $\mathbb{N} \times \mathbb{N} \to \mathbb{N}$), injections and surjections, Cantor's diagonalization technique
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- Applications: showing impossibility theorems for CS, parallel task scheduling algorithms.

Summary Contd.

▶ Part 3: Counting and Combinatorics

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- ▶ Basic counting principles, double counting
- Binomial theorem, permutations and combinations, Estimating n!

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- ▶ Principle of Inclusion-Exclusion (PIE) and its applications.
- ▶ Pigeon-Hole Principle (PHP) and its applications.
- Some special numbers and sequences: Fibonacci, Catalan
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- Applications: Bounding the complexity of algorithms, proving existence of structures.

Summary (Contd.)

Part 4: 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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Graph theory: Characterizations

- 1. Eulerian graphs: Using degrees of vertices.
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Core: CS 228 (Logic), CS213 (DSA), CS215 (DAI), CS 218 (DAA), CS 310 (Automata)

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