Introduction to CS 4104

T. M. Murali

January 18, 2017

Course Information

- Instructor
  - T. M. Murali, 2160B Torgerson, 231-8534, murali@cs.vt.edu
  - Office Hours: 9:30am-11:30am, Mondays and by appointment
- Graduate teaching assistants
  - Wenjie Zhuang (GTA), kaito@vt.edu
  - Office Hours: to be announced
  - Peter Steele (UTA), peter707@vt.edu
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- Class meeting times
  - MW 2:30pm–3:45pm, SURGE 204C

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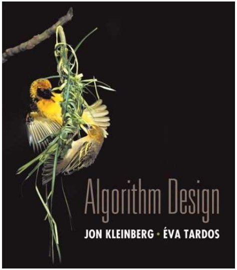
- Class meeting times
  - MW 2:30pm–3:45pm, SURGE 204C
- Prerequisite: Grade of C or better in CS 3114; P or better in MATH 3034 or MATH 3134
- Force-add: Visit https://www.cs.vt.edu/S17Force-Adds before 3:45pm today and use the password "4104tmm$"

Keeping in Touch

- Course web site http://courses.cs.vt.edu/~cs4104/murali/spring2017, updated regularly through the semester
- Canvas: grades and homework/exam solutions
- Piazza: announcements

Required Course Textbook



- Algorithm Design
- Jon Kleinberg and Éva Tardos
- Addison-Wesley
- 2006
- ISBN: 0-321-29535-8

Course Goals

- Learn methods and principles to construct algorithms.
- Learn techniques to analyze algorithms mathematically for correctness and efficiency (e.g., running time and space used).
- Course roughly follows the topics suggested in textbook
  - Stable matching
  - Measures of algorithm complexity
  - Graphs
  - Greedy algorithms
  - Divide and conquer
  - Dynamic programming
  - Network flow problems
  - NP-completeness
  - Coping with intractability
  - Approximation algorithms

Required Readings

- Reading assignment available on the website.
- Read **before** class.
- I strongly encourage you to keep up with the reading. Will make the class much easier.

Lecture Slides

- Will be available on class web site.
- Usually posted just before class.
- Class attendance is extremely important.

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- Usually posted just before class.
- Class attendance is extremely important. Lecture in class contains significant and substantial additions to material on the slides.

Homeworks

- Posted on the web site $\approx$ one week before due date.
- Announced via Canvas/Piazza.
- Prepare solutions digitally and upload on Canvas.
  - Solution preparation recommended in LaTeX.
  - Do not submit handwritten solutions!

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- Posted on the web site $\approx$ one week before due date.
- Announced via Canvas/Piazza.
- Prepare solutions digitally and upload on Canvas.
  - Solution preparation recommended in LaTeX.
  - Do not submit handwritten solutions!
- Homework grading: lenient at beginning but gradually become stricter over the semester.
- Essential that you read posted homework solutions to learn how to describe algorithms and write proofs.

Examinations

- Take-home midterm.
- Take-home final (comprehensive).
- Prepare digital solutions (recommend LaTeX).

Grades

- Homeworks: 7–8, 60% of the grade.
- Take-home midterm: 15% of the grade.
- Take-home final: 25% of the grade.

Honor Code

- Virginia Tech Honor Code applies to this class.
- Assistance from the internet or anyone else is a violation of the Honor Code.
- Your work and solutions to the examinations must be only your own.
- Special policy for homeworks:
  - Work on the homework in pairs. You can bounce ideas off your partner.
  - Prepare solutions individually. *Identical or similar solutions to any problem in a homework violate the Honor Code.*

What is an Algorithm?

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Chamber's
: A set of prescribed computational procedures for solving a problem; a step-by-step method for solving a problem.

Knuth, TAOCP
: An algorithm is a finite, definite, effective procedure, with some input and some output.

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Chamber's: A set of prescribed computational procedures for solving a problem; a step-by-step method for solving a problem.

Knuth, TAOCP: An algorithm is a finite, definite, effective procedure, with some input and some output.

Two other important aspects:

1. **Correct**: We will be able to rigourously prove that the algorithm does what it is supposed to do.
2. **Efficient**: We will also prove that the algorithm runs in polynomial time. We will try to make it as fast as we can.

Origin of the word "Algorithm"

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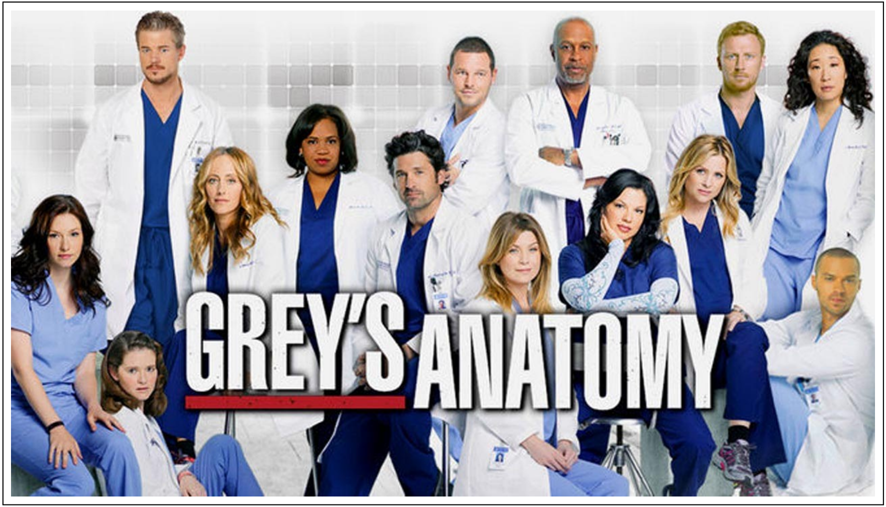
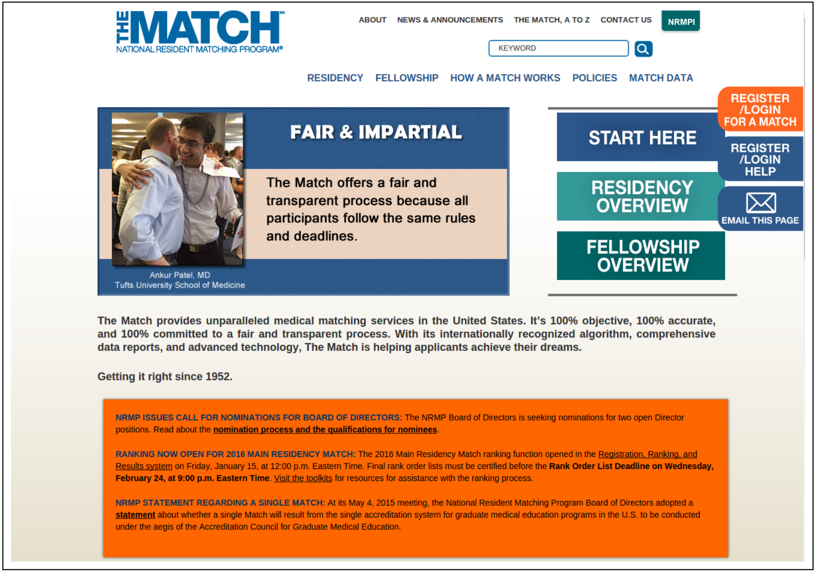
Origin of the word "Algorithm"

1. From the Arabic *al-Khwarizmi*, a native of Khwarazm, a name for the 9th century mathematician, Abu Ja'far Mohammed ben Musa. He wrote "Kitab al-jabr wa'l-muqabala," which evolved into today's high school algebra text.


GREY'S ANATOMY


THE MATCH
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FELLOWSHIP
HOW A MATCH WORKS
POLICIES
MATCH DATA
REGISTER /LOGIN FOR A MATCH
FAIR & IMPARTIAL
The Match offers a fair and transparent process because all participants follow the same rules and deadlines.
Ankur Patel, MD
Tufts University School of Medicine
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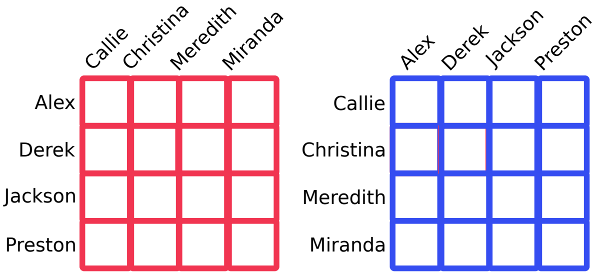
The Match provides unparalleled medical matching services in the United States. It's 100% objective, 100% accurate, and 100% committed to a fair and transparent process. With its internationally recognized algorithm, comprehensive data reports, and advanced technology, The Match is helping applicants achieve their dreams.
Getting it right since 1952.
NRMP ISSUES CALL FOR NOMINATIONS FOR BOARD OF DIRECTORS: The NRMP Board of Directors is seeking nominations for two open Director positions. Read about the nomination process and the qualifications for nominees.
RANKING NOW OPEN FOR 2016 MAIN RESIDENCY MATCH: The 2016 Main Residency Match ranking function opened in the Registration, Ranking, and Results system on Friday, January 15, at 12:00 p.m. Eastern Time. Final rank order lists must be certified before the Rank Order List Deadline on Wednesday, February 24, at 9:00 p.m. Eastern Time. Visit the toolkits for resources for assistance with the ranking process.
NRMP STATEMENT REGARDING A SINGLE MATCH: At its May 4, 2015 meeting, the National Resident Matching Program Board of Directors adopted a statement about whether a single Match will result from the single accreditation system for graduate medical education programs in the U.S. to be conducted under the aegis of the Accreditation Council for Graduate Medical Education.

Stable Matching Problem



There are n men and n women.

Stable Matching Problem

	Callie	Christina	Meredith	Miranda
Alex	1	2	3	4
Derek	4	3	1	2
Jackson	4	3	1	2
Preston	3	1	4	2

	Alex	Derek	Jackson	Preston
Callie	1	2	3	4
Christina	4	1	3	2
Meredith	4	1	2	3
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Each man ranks all the women in order of preference.
Each woman ranks all the men in order of preference.
Each person uses all ranks from 1 to n.

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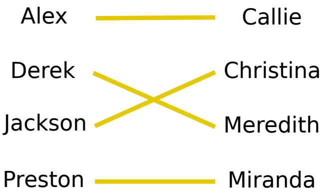

Perfect matching: each man is paired with exactly one woman and vice versa.

Stable Matching Problem

	Callie	Christina	Meredith	Miranda
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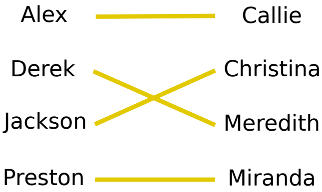

Is this matching good?

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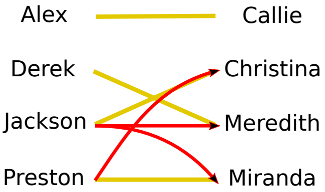

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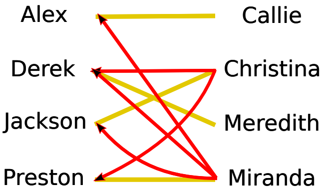

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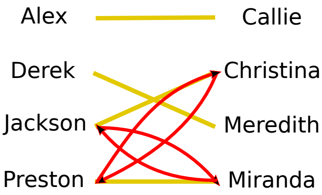

Rogue couple: a man and a woman who are not matched but prefer each other to their current partners.

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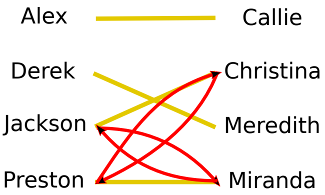

Stable matching: A perfect matching without any rogue couples.

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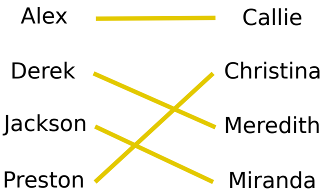

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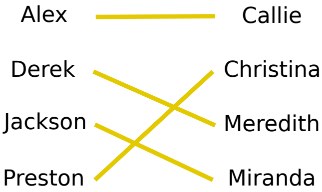

Questions

1. Given preferences for every woman and every man, does a stable matching exist?
2. If it does, can we compute it? How fast?

Examples

Example

m prefers w to w'

m' prefers w to w'

---

w prefers m to m'

w' prefers m to m'

Stable Matching

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Stable Matching

(m, w) and (m', w')

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(m, w) and (m', w')

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m prefers w to w' m' prefers w' to w ——— w prefers m' to m w' prefers m to m'	(m, w) and (m', w') or (m, w') and (m', w)

Gale-Shapley Algorithm

Each man proposes to each woman, in decreasing order of preference.
Woman accepts if she is free or prefers new prospect to current fiance.

```
Initially, all men and all women are free
While there is at least one free man who has not
   proposed to every woman
    Choose such a man m
    m proposes to the highest-ranked woman w on his list
    to whom he has not yet proposed
    If w is free, then
         she becomes engaged to m
    else if w is engaged to m' and she prefers m to m'
         she becomes engaged to m
         m' becomes free
    Otherwise, m remains free
Return set S of engaged pairs
```

Questions about the Algorithm

- Does the algorithm even terminate?
- If it does, how long does the algorithm take to run?
- If it does, is S a perfect matching? A stable matching?

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- Ranking of a woman's partner: Can never go down.

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Formal proof: Let $p(k), k \geq 1$ be the number of proposals made after k iterations. Clearly, $p(k) \leq n^2$ since there are n^2 man-woman pairs. Moreover, at least one proposal is made in every iteration. Hence, the algorithm terminates after n^2 iterations.

Running Time

Implement each iteration in constant time to get running time $\propto n^2$

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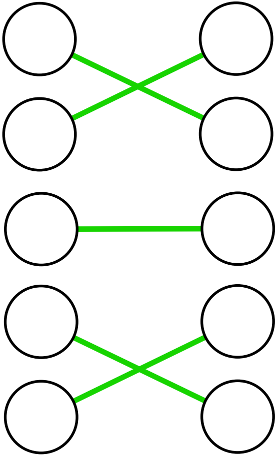
    else if w is engaged to m′ and she prefers m to m′
         she becomes engaged to m Rank[w,m] = rank of m in
         m′ becomes free                          w's list
    Otherwise, m remains free
Return set S of engaged pairs
```

Proof: Matching Computed is Perfect

- Suppose the set S of pairs returned by the Gale-Shapley algorithm is not perfect.
- S is a matching. Therefore, there must be at least one free man m.
- m has proposed to all the women (since algorithm terminated).
- Therefore, each woman must be engaged (since she remains engaged after the first proposal to her).
- Therefore, all men must be engaged, contradicting the assumption that m is free.

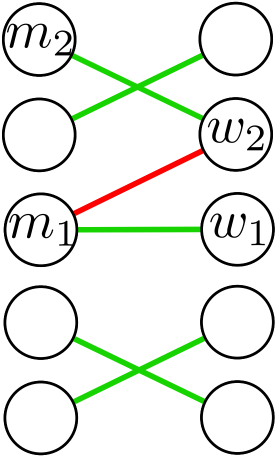
Proof: Matching Computed is Stable

Perfect matching S returned by algorithm



Proof: Matching Computed is Stable

Not stable: m_1 paired with w_1 but prefers w_2;
w_2 paired with m_2 but prefers m_1



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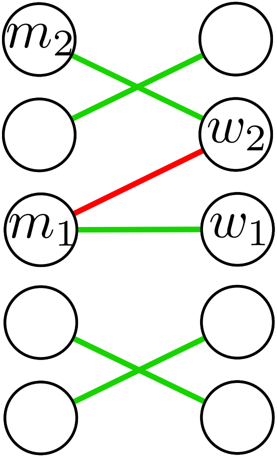
$\Rightarrow$ m_1 proposed to w_2 before proposing to w_1



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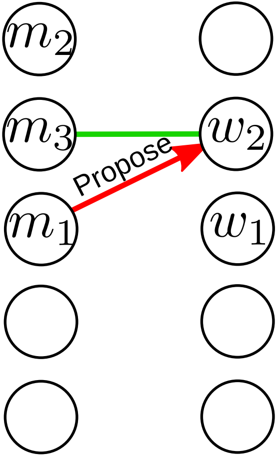
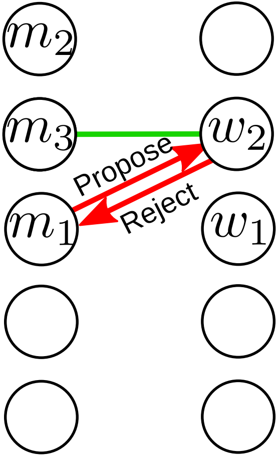
Rewind: What happened when m_1 proposed to w_2?



Proof: Matching Computed is Stable

Case 1: w_2 rejected m_1 because she preferred current partner m_3.



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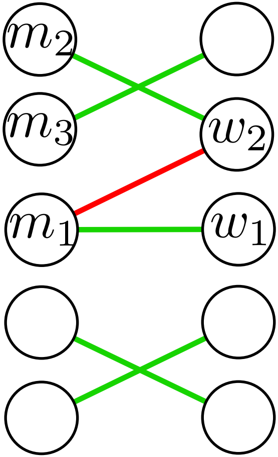
Case 1: At termination, w_2 must prefer her final partner m_2 to m_3.

Contradicts consequence of instability: w_2 prefers m_1 to m_2



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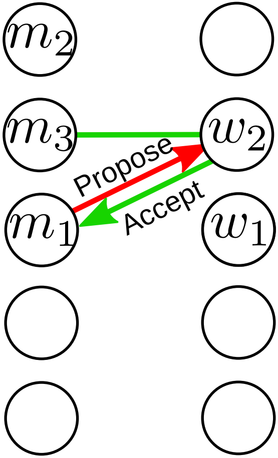
Case 2: w_2 accepted m_1 because she had no partner or preferred m_1 to current partner m_3.



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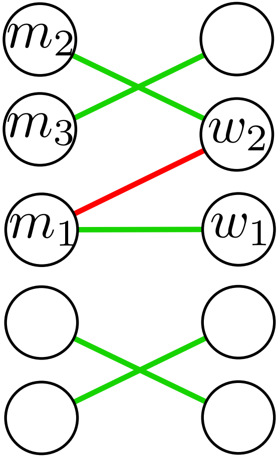
Case 2: By instability, we know w_2 prefers m_1 to m_2. But at termination, w_2 is matched with m_2, which contradicts property that a woman switches only to a better match.



Proof: Stable Matching (in Words)

- Suppose S is not stable, i.e., there are two pairs (m_1, w_1) and (m_2, w_2) in S such that m_1 prefers w_2 to w_1 and w_2 prefers m_1 to m_2.
- m_1 must have proposed to w_2 before w_1 because ....
- At that stage w_2 must have rejected m_1; otherwise, the algorithm would pair m_1 and w_2, which would prevent the pairing of m_2 and w_2 in a later iteration of the algorithm. (Why?)
- When w_2 rejected m_1, she must have been paired with some man, say m_3, whom she prefers to m_1.
- Since m_2 is paired with w_2 at termination, w_2 must prefer to m_2 to m_3 or $m_2 = m_3$ (Why?), which contradicts our conclusion (from instability) that w_2 prefers m_1 to m_2.

Further Reading and Viewing

- Gail-Shapley algorithm always produces the same matching in which each man is paired with his best valid partner but each woman is paired with her worst valid partner. Read pages 9–12 of the textbook.
- Video describing matching algorithm used by the National Resident Matching Program
- Description of research to Roth and Shapley that led to 2012 Nobel Prize in Economics

Variants of Stable Matching

- Hospitals and residents: Each hospital can take multiple residents.
- Hospitals and residents with couples: Each hospital can take multiple residents. A couple must be assigned together, either to the same hospital or to a specific pair of hospitals chosen by the couple.
- Stable roommates problem: there is only one "gender".
- Preferences may be incomplete or have ties or people may lie.

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- Preferences may be incomplete or have ties or people may lie. Several variants are NP-hard, even to approximate.