



Algorithms:

COMP3121/3821/9101/9801

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INTRODUCTION

Introduction

What is this course about?

It is about **designing algorithms** for solving problems.

Why should you study algorithms design?

Can you find every algorithm you might need using Google?

Our goal:

To learn **techniques** which can be used to solve **new, unfamiliar** problems that arise in a rapidly changing field.

Course content:

- a survey of algorithms **design techniques**
- particular algorithms will be mostly used to illustrate design techniques
- emphasis on development of **algorithm design skills**

Textbook:

Kleinberg and Tardos: *Algorithm Design*

paperback edition available at the UNSW book store

good: very readable (and very pleasant to read!); an excellent textbook;

not so good: as a reference manual for later use.

An alternative textbook:

Cormen, Leiserson, Rivest and Stein: *Introduction to Algorithms*

preferably the third edition, should also be available at the bookstore

excellent: to be used later as a reference manual;

not so good: quite formalistic and written in a rather dry style.

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Only when this is not clear by common sense.

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Merge-Sort(A, p, r) *sorting $A[p..r]$ *

- ❶ **if** $p < r$
- ❷ **then** $q \leftarrow \lfloor \frac{p+r}{2} \rfloor$
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- ❺ The above is essentially a proof by induction, but we will never bother formalizing proofs of (essentially) obvious facts.

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- **However, be very careful what you call trivial!!**

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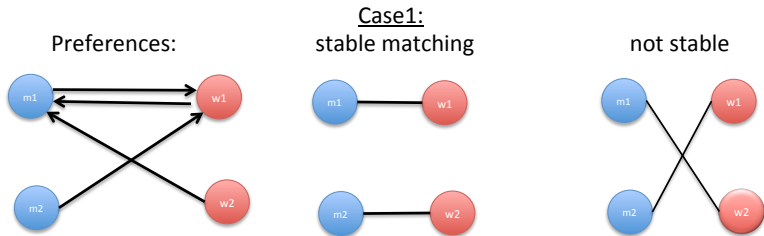
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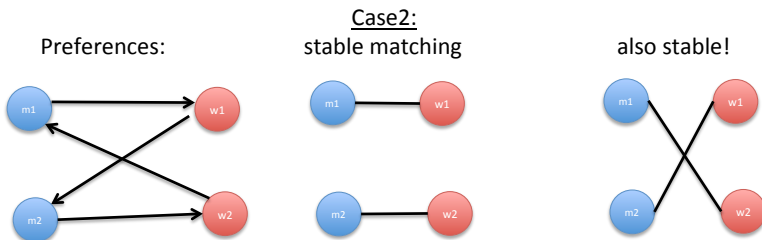
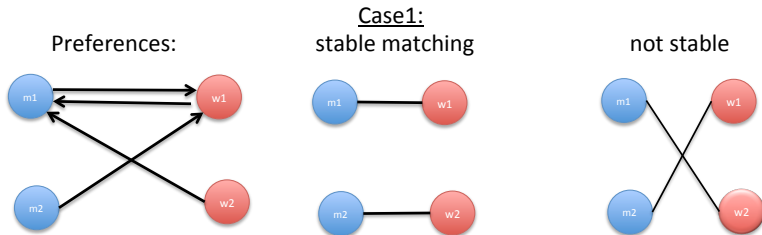
for two pairs $p = (m, w)$ and $p' = (m', w')$:

- man m prefers woman w' to woman w , **and**
- woman w' prefers man m to man m' .

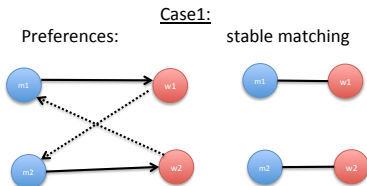
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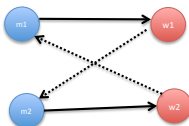
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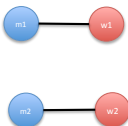
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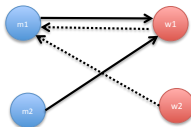


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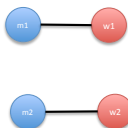


Case2:

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stable matching



Stable Matching Problem: Gale - Shapley algorithm (invented to pair newly graduated physicians with US hospitals for residency training)

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Answer: **YES**, using the **Gale - Shapley algorithm**.

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Else m is lower on her preference list than m' ;
the proposal is rejected and m remains free.

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- Thus, every woman is paired with a man, because a woman is not paired with anyone only if no one has made a proposal to her.
- But this would mean that n women are paired with all of n men so m cannot be free.

Stable Matching Problem: Gale - Shapley algorithm

Claim 1: Algorithm terminates after $\leq n^2$ rounds of the *While* loop

Proof:

- In every round of the *While* loop one man proposes to one woman;
- every man can propose to a woman at most once;
- thus, every man can make at most n proposals;
- there are n men, so in total they can make $\leq n^2$ proposals

Thus the *While* loop can be executed no more than n^2 many times.

Claim 2: Algorithm produces a matching, i.e., every man is eventually paired with a woman (and thus also every woman is paired to a man)

Proof:

- Assume that the while *While* loop has terminated, but m is still free.
- This means that m has already proposed to every woman.
- Thus, every woman is paired with a man, because a woman is not paired with anyone only if no one has made a proposal to her.
- But this would mean that n women are paired with all of n men so m cannot be free. **Contradiction!**

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Claim 3: The matching produced by the algorithm is stable.

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Note: this method is called “divide-and-conquer”.

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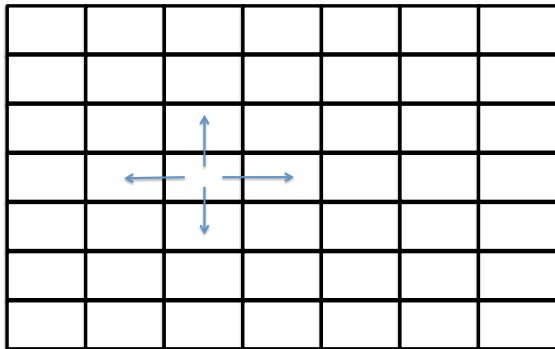
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- More ways to hide than the number of all possible weighting outcomes! Thus, it is impossible to do it!

Puzzles!!!

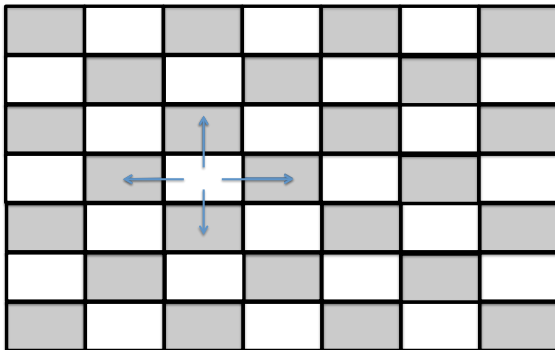
Problem: Consider a block of 7 X 7 houses:



The inhabitant of each house thinks that all four houses around him (to the left, right, top and bottom) are nicer than his house and would like to move to any of the four. Can you move the inhabitants around to make them all happier?

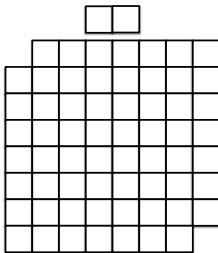
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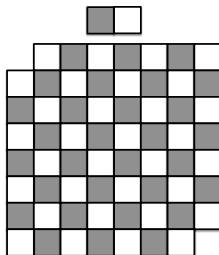
Problem: Consider an 8×8 board with two diagonal squares missing, and an 1×2 domino:



Can you cover the entire board with 31 such dominoes?

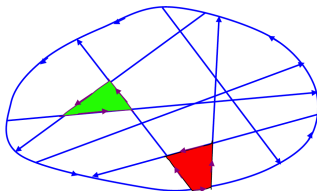
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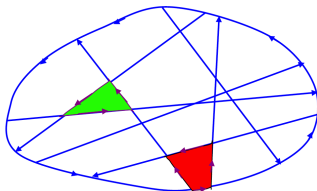
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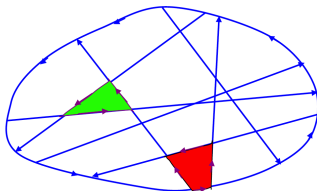
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- A block is a part of the city that is not intersected by any street.
- Design an algorithm that, given a map of a city, finds a block (just one such block, not all such blocks) that can be circumnavigated while respecting all one-way signs.

(for example, the green block has such property, but not the red one)

Basics revisited: how do we add two numbers?

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- no asymptotically faster algorithm

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- “Simple” problems can actually turn out to be difficult!

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$$\begin{array}{ll} A = A_1 2^{\frac{n}{2}} + A_0 & \underbrace{XX \dots X}_{\frac{n}{2}} \underbrace{XX \dots X}_{\frac{n}{2}} \\ B = B_1 2^{\frac{n}{2}} + B_0 & \end{array}$$

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- AB can now be calculated as follows:

$$AB = A_1 B_1 2^n + (A_1 B_0 + B_1 A_0) 2^{\frac{n}{2}} + A_0 B_0 \quad (1)$$

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What we mean is that the product AB can be calculated recursively by the following program:

```

1: function MULT( $A, B$ )
2:   if  $|A| = |B| = 1$  then return  $AB$ 
3:   else
4:      $A_1 \leftarrow \text{MoreSignificantPart}(A)$ ;
5:      $A_0 \leftarrow \text{LessSignificantPart}(A)$ ;
6:      $B_1 \leftarrow \text{MoreSignificantPart}(B)$ ;
7:      $B_0 \leftarrow \text{LessSignificantPart}(B)$ ;
8:      $X \leftarrow \text{MULT}(A_0, B_0)$ ;
9:      $Y \leftarrow \text{MULT}(A_0, B_1)$ ;
10:     $Z \leftarrow \text{MULT}(A_1, B_0)$ ;
11:     $W \leftarrow \text{MULT}(A_1, B_1)$ ;
12:    return  $W 2^n + (Y + Z) 2^{n/2} + X$ 
13:   end if
14: end function

```


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$$T(n) = 4T\left(\frac{n}{2}\right) + cn \quad (2)$$

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Claim: if $T(n)$ satisfies

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$$T(n) = 4T\left(\frac{n}{2}\right) + cn = 4\left(\left(\frac{n}{2}\right)^2 (c+1) - \frac{n}{2}c\right) + cn$$

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Claim: if $T(n)$ satisfies

$$T(n) = 4T\left(\frac{n}{2}\right) + cn \quad (3)$$

then

$$T(n) = n^2(c+1) - cn$$

Proof: By “fast” induction. We assume it is true for $\lfloor n/2 \rfloor$:

$$T\left(\frac{n}{2}\right) = \left(\frac{n}{2}\right)^2 (c+1) - c \frac{n}{2}$$

and prove that it is also true for n :

$$\begin{aligned} T(n) &= 4T\left(\frac{n}{2}\right) + cn = 4\left(\left(\frac{n}{2}\right)^2 (c+1) - \frac{n}{2}c\right) + cn \\ &= n^2(c+1) - 2cn + cn = n^2(c+1) - cn \end{aligned}$$

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Some history: In 1952, one of the most famous mathematicians of the 20th century, Andrey Kolmogorov, conjectured that you cannot multiply in less than $\Omega(n^2)$ elementary operations. In 1960, Karatsuba, then a 23-year-old student, found an algorithm (later it was called “divide and conquer”) that multiplies two n -digit numbers in $\Theta(n^{\log_2 3}) \approx \Theta(n^{1.58\dots})$ elementary steps, thus disproving the conjecture!! Kolmogorov was shocked!

The Karatsuba trick

How did Karatsuba do it??

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Take again our two input numbers A and B , and split them into two halves:

$$\begin{array}{lcl} A = A_1 2^{\frac{n}{2}} + A_0 & \underbrace{XX \dots X}_{\frac{n}{2}} & \underbrace{XX \dots X}_{\frac{n}{2}} \\ B = B_1 2^{\frac{n}{2}} + B_0 & & \end{array}$$

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$$\begin{aligned} AB &= A_1 B_1 2^n + (A_1 B_0 + A_0 B_1) 2^{\frac{n}{2}} + A_0 B_0 \\ &= A_1 B_1 2^n + ((A_1 + A_0)(B_1 + B_0) - A_1 B_1 - A_0 B_0) 2^{\frac{n}{2}} + A_0 B_0 \end{aligned}$$

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Thus, the algorithm will look like this:

```

1: function MULT( $A, B$ )
2:   if  $|A| = |B| = 1$  then return  $AB$ 
3:   else
4:      $A_1 \leftarrow \text{MoreSignificantPart}(A)$ ;
5:      $A_0 \leftarrow \text{LessSignificantPart}(A)$ ;
6:      $B_1 \leftarrow \text{MoreSignificantPart}(B)$ ;
7:      $B_0 \leftarrow \text{LessSignificantPart}(B)$ ;
8:      $U \leftarrow A_0 + A_1$ ;
9:      $V \leftarrow B_0 + B_1$ ;
10:     $X \leftarrow \text{MULT}(A_0, B_0)$ ;
11:     $W \leftarrow \text{MULT}(A_1, B_1)$ ;
12:     $Y \leftarrow \text{MULT}(U, V)$ ;
13:    return  $W 2^n + (Y - X - W) 2^{n/2} + X$ 
14:  end if
15: end function

```

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$$\begin{aligned}T(n) &= 3T\left(\frac{n}{2}\right) + cn = 3\left(3T\left(\frac{n}{2^2}\right) + c\frac{n}{2}\right) + cn = \underbrace{3^2 T\left(\frac{n}{2^2}\right)} + c\frac{3n}{2} + cn \\&= 3^2 \left(\underbrace{3T\left(\frac{n}{2^3}\right) + c\frac{n}{2^2}}\right) + c\frac{3n}{2} + cn = 3^3 T\left(\frac{n}{2^3}\right) + c\frac{3^2 n}{2^2} + c\frac{3n}{2} + cn \\&= 3^3 \underbrace{T\left(\frac{n}{2^3}\right)} + cn \left(\frac{3^2}{2^2} + \frac{3}{2} + 1\right) \\&= 3^3 \left(\underbrace{3T\left(\frac{n}{2^4}\right) + c\frac{n}{2^3}}\right) + cn \left(\frac{3^2}{2^2} + \frac{3}{2} + 1\right)\end{aligned}$$

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So we got

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$$T(n) \approx n^{\log_2 3} T(1) + 2cn \left(n^{\log_2 \frac{3}{2}} - 1 \right) = n^{\log_2 3} T(1) + 2cn \left(n^{\log_2 3 - 1} - 1 \right)$$

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Please review the basic properties of logarithms and the asymptotic notation from the review material (the first item at the class webpage under “class resources”).

Next time:

- 1 Can we multiply large integers faster than $O(n^{\log_2 3})$??

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- 1 Can we multiply large integers faster than $O(n^{\log_2 3})$??
- 2 Can we avoid messy computations like:

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That's All, Folks!!