# Introduction to Algorithms 6.046J/18.401J/SMA5503

## Lecture 3

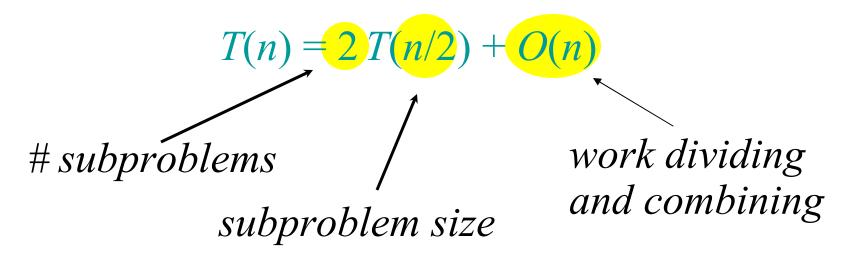
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# The divide-and-conquer design paradigm

- 1. Divide the problem (instance) into subproblems.
- 2. Conquer the subproblems by solving them recursively.
- 3. Combine subproblem solutions.

## Example: merge sort

- 1. Divide: Trivial.
- 2. Conquer: Recursively sort 2 subarrays.
- 3. Combine: Linear-time merge.



## Master theorem (reprise)

$$T(n) = a T(n/b) + f(n)$$

CASE 1: 
$$f(n) = O(n^{\log_b a - \varepsilon})$$
  
 $\Rightarrow T(n) = \Theta(n^{\log_b a})$ .

CASE 2: 
$$f(n) = \Theta(n^{\log_b a} \lg^k n)$$
  
 $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$ .

CASE 3: 
$$f(n) = \Omega(n^{\log_b a + \varepsilon})$$
 and  $af(n/b) \le cf(n)$   
 $\Rightarrow T(n) = \Theta(f(n))$ .

Merge sort: 
$$a = 2$$
,  $b = 2 \implies n^{\log_b a} = n$   
 $\Rightarrow$  Case 2  $(k = 0) \implies T(n) = \Theta(n \lg n)$ .

Find an element in a sorted array:

- 1. Divide: Check middle element.
- 2. Conquer: Recursively search 1 subarray.
- 3. Combine: Trivial.

Example: Find 9

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3 5 7 8

9 12 15

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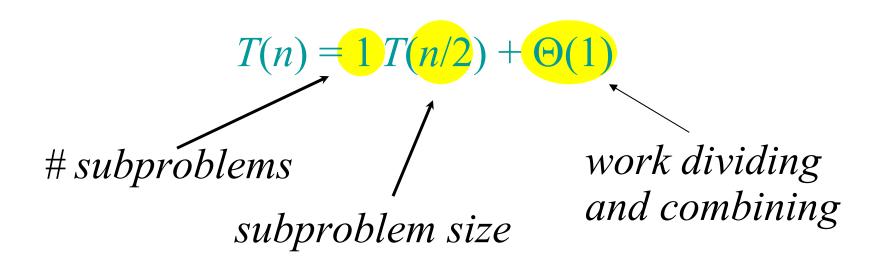
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## Recurrence for binary search



$$n^{\log_b a} = n^{\log_2 1} = n^0 = 1 \implies \text{CASE 2}(k = 0)$$
  
  $\implies T(n) = \Theta(\lg n)$ .

## Powering a number

**Problem:** Compute  $a^n$ , where  $n \in \mathbb{N}$ .

Naive algorithm:  $\Theta(n)$ .

#### Divide-and-conquer algorithm:

$$a^{n} = \begin{cases} a^{n/2} \cdot a^{n/2} & \text{if } n \text{ is even;} \\ a^{(n-1)/2} \cdot a^{(n-1)/2} \cdot a & \text{if } n \text{ is odd.} \end{cases}$$

$$T(n) = T(n/2) + \Theta(1) \implies T(n) = \Theta(\lg n)$$
.

#### Fibonacci numbers

#### **Recursive definition:**

$$F_{n} = \begin{cases} 0 & \text{if } n = 0; \\ 1 & \text{if } n = 1; \\ F_{n-1} + F_{n-2} & \text{if } n \ge 2. \end{cases}$$

$$0 \quad 1 \quad 1 \quad 2 \quad 3 \quad 5 \quad 8 \quad 13 \quad 21 \quad 34 \quad \cdots$$

Naive recursive algorithm:  $\Omega(\phi^n)$  (exponential time), where  $\phi = (1+\sqrt{5})/2$  is the *golden ratio*.

## **Computing Fibonacci numbers**

#### Naive recursive squaring:

 $F_n = \phi^n / \sqrt{5}$  rounded to the nearest integer.

- Recursive squaring:  $\Theta(\lg n)$  time.
- This method is unreliable, since floating-point arithmetic is prone to round-off errors.

#### **Bottom-up:**

- Compute  $F_0, F_1, F_2, ..., F_n$  in order, forming each number by summing the two previous.
- Running time:  $\Theta(n)$ .

## Recursive squaring

**Theorem:** 
$$\begin{bmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n.$$

Algorithm: Recursive squaring.

Time = 
$$\Theta(\lg n)$$
.

*Proof of theorem.* (Induction on *n*.)

Base 
$$(n = 1)$$
:  $\begin{bmatrix} F_2 & F_1 \\ F_1 & F_0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^1$ .

## Recursive squaring

Inductive step  $(n \ge 2)$ :

$$\begin{bmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{bmatrix} = \begin{bmatrix} F_n & F_{n-1} \\ F_{n-1} & F_{n-2} \end{bmatrix} \cdot \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^{n-1} \cdot \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}$$
$$= \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n$$

## Matrix multiplication

Input: 
$$A = [a_{ij}], B = [b_{ij}].$$
  
Output:  $C = [c_{ij}] = A \cdot B.$   $i, j = 1, 2, ..., n.$ 

$$\begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \cdot \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{nn} \end{bmatrix}$$

$$c_{ij} = \sum_{k=1}^{n} a_{ik} \cdot b_{kj}$$

## Standard algorithm

```
for i \leftarrow 1 to n
do for j \leftarrow 1 to n
do c_{ij} \leftarrow 0
for k \leftarrow 1 to n
do c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}

Running time = \Theta(n^3)
```

## Divide-and-conquer algorithm

#### **DEA:**

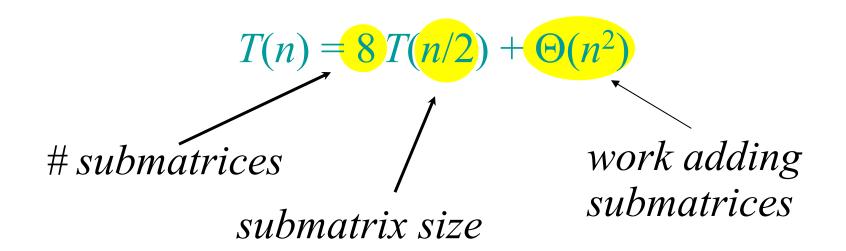
 $n \times n$  matrix = 2×2 matrix of  $(n/2) \times (n/2)$  submatrices:

$$\begin{bmatrix} r \mid s \\ -+- \\ t \mid u \end{bmatrix} = \begin{bmatrix} a \mid b \\ -+- \\ c \mid d \end{bmatrix} \cdot \begin{bmatrix} e \mid f \\ ---- \\ g \mid h \end{bmatrix}$$

$$C = A \cdot B$$

$$r = ae + bg$$
  
 $s = af + bh$   
 $t = ce + dh$   
 $u = cf + dg$   
8 mults of  $(n/2) \times (n/2)$  submatrices  
4 adds of  $(n/2) \times (n/2)$  submatrices

## Analysis of D&C algorithm



$$n^{\log_b a} = n^{\log_2 8} = n^3 \implies \text{Case } 1 \implies T(n) = \Theta(n^3).$$

No better than the ordinary algorithm.

### Strassen's idea

• Multiply 2×2 matrices with only 7 recursive mults.

$$P_1 = a \cdot (f - h)$$
  
 $P_2 = (a + b) \cdot h$   
 $P_3 = (c + d) \cdot e$   
 $P_4 = d \cdot (g - e)$   
 $P_5 = (a + d) \cdot (e + h)$   
 $P_6 = (b - d) \cdot (g + h)$   
 $P_7 = (a - c) \cdot (e + f)$ 

$$r = P_5 + P_4 - P_2 + P_6$$

$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

$$u = P_5 + P_1 - P_3 - P_7$$

7 mults, 18 adds/subs.
Note: No reliance on commutativity of mult!

### Strassen's idea

• Multiply  $2\times2$  matrices with only 7 recursive mults.

$$P_{1} = a \cdot (f - h)$$
  
 $P_{2} = (a + b) \cdot h$   
 $P_{3} = (c + d) \cdot e$   
 $P_{4} = d \cdot (g - e)$   
 $P_{5} = (a + d) \cdot (e + h)$   
 $P_{6} = (b - d) \cdot (g + h)$   
 $P_{7} = (a - c) \cdot (e + f)$ 

$$r = P_{5} + P_{4} - P_{2} + P_{6}$$

$$= (a + d)(e + h)$$

$$+ d(g - e) - (a + b)h$$

$$+ (b - d)(g + h)$$

$$= ae + ah + de + dh$$

$$+ dg - de - ah - bh$$

$$+ bg + bh - dg - dh$$

$$= ae + bg$$

## Strassen's algorithm

- 1. Divide: Partition A and B into  $(n/2) \times (n/2)$  submatrices. Form terms to be multiplied using + and -.
- 2. Conquer: Perform 7 multiplications of  $(n/2)\times(n/2)$  submatrices recursively.
- 3. Combine: Form C using + and on  $(n/2)\times(n/2)$  submatrices.

$$T(n) = 7 T(n/2) + \Theta(n^2)$$

## **Analysis of Strassen**

$$T(n) = 7 T(n/2) + \Theta(n^2)$$

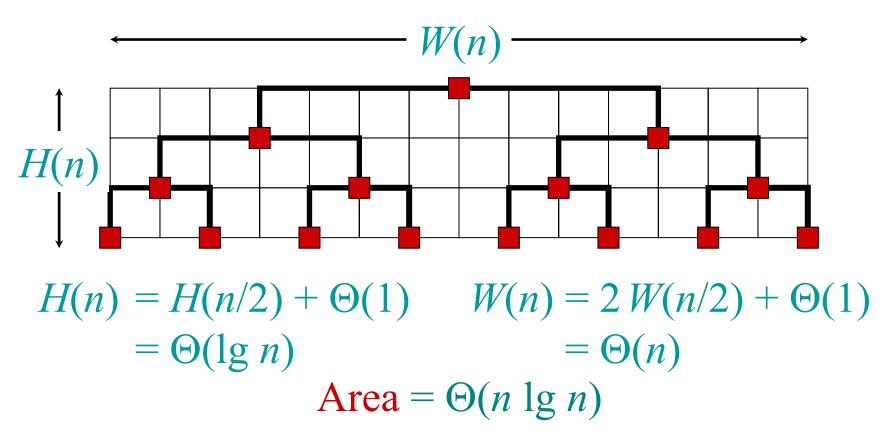
$$n^{\log_b a} = n^{\log_2 7} \approx n^{2.81} \implies \text{Case } 1 \implies T(n) = \Theta(n^{\lg 7}).$$

The number 2.81 may not seem much smaller than 3, but because the difference is in the exponent, the impact on running time is significant. In fact, Strassen's algorithm beats the ordinary algorithm on today's machines for  $n \ge 30$  or so.

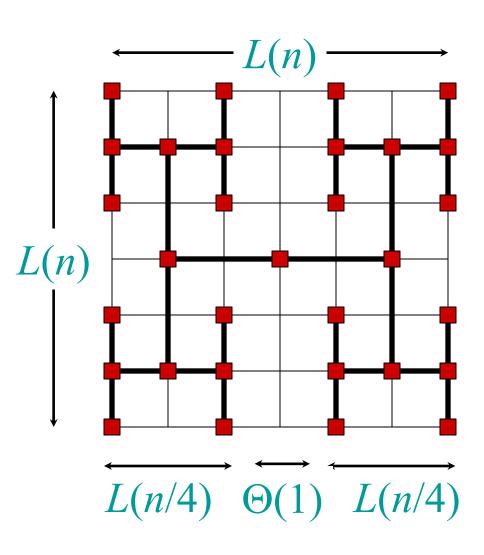
Best to date (of theoretical interest only):  $\Theta(n^{2.376\cdots})$ .

## VLSI layout

**Problem:** Embed a complete binary tree with *n* leaves in a grid using minimal area.



## H-tree embedding



$$L(n) = 2L(n/4) + \Theta(1)$$
$$= \Theta(\sqrt{n})$$

Area = 
$$\Theta(n)$$

### **Conclusion**

- Divide and conquer is just one of several powerful techniques for algorithm design.
- Divide-and-conquer algorithms can be analyzed using recurrences and the master method (so practice this math).
- Can lead to more efficient algorithms