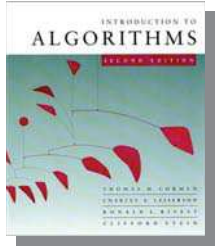


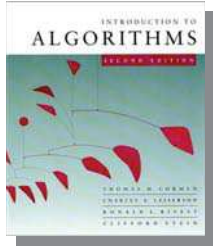
The divide-and-conquer design paradigm

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



Merge sort

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



Merge sort

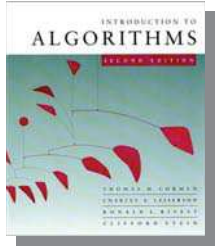
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3. **Combine:** Linear-time merge.

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

subproblems

subproblem size

work dividing and combining



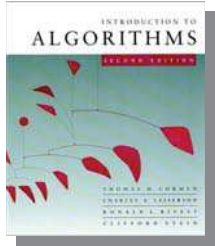
Master theorem (reprise)

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

CASE 1: $f(n) = O(n^{\log_b a - \varepsilon})$, constant $\varepsilon > 0$
 $\Rightarrow T(n) = \Theta(n^{\log_b a})$.

CASE 2: $f(n) = \Theta(n^{\log_b a} \lg^k n)$, constant $k \geq 0$
 $\Rightarrow T(n) = \Theta(n^{\log_b a} \lg^{k+1} n)$.

CASE 3: $f(n) = \Omega(n^{\log_b a + \varepsilon})$, constant $\varepsilon > 0$,
and regularity condition
 $\Rightarrow T(n) = \Theta(f(n))$.



Master theorem (reprise)

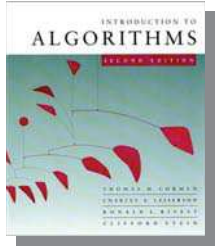
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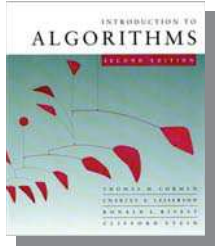
Merge sort: $a = 2, b = 2 \Rightarrow n^{\log_b a} = n^{\log_2 2} = n$
 \Rightarrow **CASE 2** ($k = 0$) $\Rightarrow T(n) = \Theta(n \lg n)$.



Binary search

Find an element in a sorted array:

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



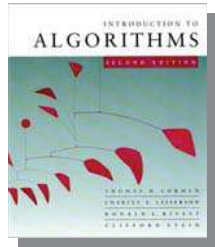
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- 1. *Divide:*** Check middle element.
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- 3. *Combine:*** Trivial.

Example: Find **9**

3 5 7 8 9 12 15



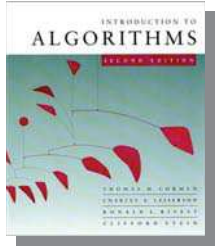
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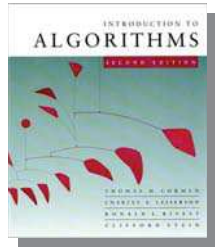
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Example: Find **9**

3

5

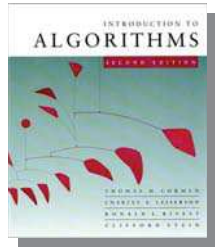
7

8

9

12

15



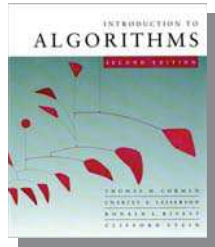
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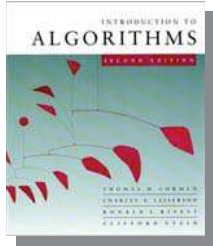
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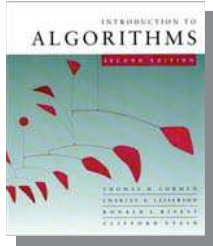
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Recurrence for binary search

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

subproblems *subproblem size* *work dividing and combining*

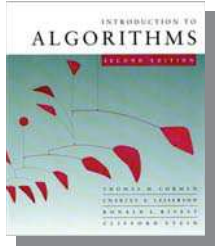


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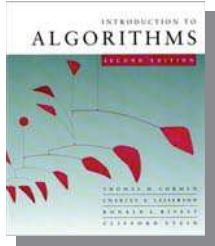
$$n^{\log_b a} = n^{\log_2 1} = n^0 = 1 \Rightarrow \text{CASE 2 } (k = 0) \\ \Rightarrow T(n) = \Theta(\lg n) .$$



Powering a number

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

Naive algorithm: $\Theta(n)$.



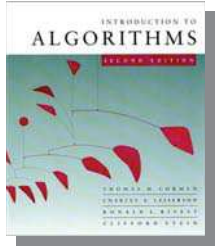
Powering a number

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Divide-and-conquer algorithm:

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



Powering a number

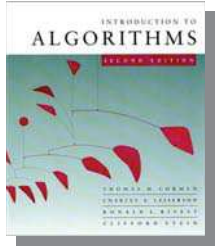
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$$T(n) = T(n/2) + \Theta(1) \Rightarrow T(n) = \Theta(\lg n) .$$

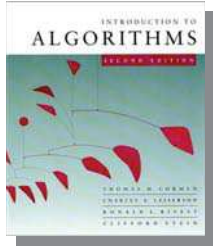


Fibonacci numbers

Recursive definition:

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

0 1 1 2 3 5 8 13 21 34 ...



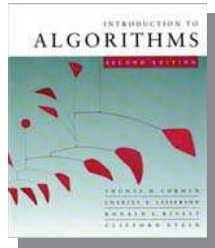
Fibonacci numbers

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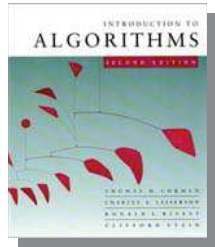
Naive recursive algorithm: $\Omega(\phi^n)$
(exponential time), where $\phi = (1 + \sqrt{5})/2$
is the *golden ratio*.



Computing Fibonacci numbers

Bottom-up:

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



Computing Fibonacci numbers

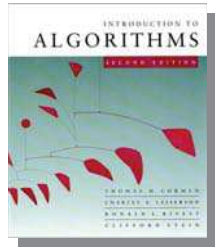
Bottom-up:

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

Naive recursive squaring:

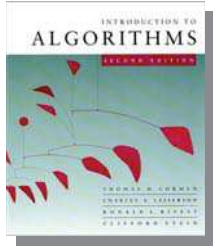
$F_n = \phi^n / \sqrt{5}$ rounded to the nearest integer.

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



Recursive squaring

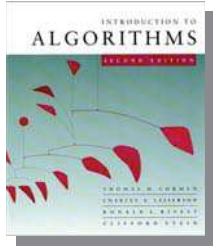
Theorem:
$$\begin{bmatrix} F_{n+1} & F_n \\ F_n & F_{n-1} \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^n.$$



Recursive squaring

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

Algorithm: Recursive squaring.
Time = $\Theta(\lg n)$.



Recursive squaring

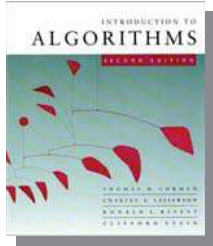
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Algorithm: Recursive squaring.

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


Proof of theorem. (Induction on n .)

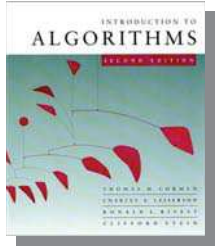
Base ($n = 1$):
$$\begin{bmatrix} F_2 & F_1 \\ F_1 & F_0 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & 0 \end{bmatrix}^1 .$$



Recursive squaring

Inductive step ($n \geq 2$):

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


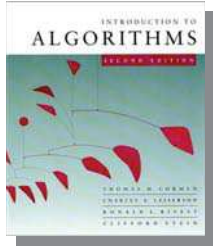


Matrix multiplication

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

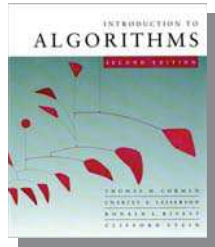
$$\begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots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

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



Standard algorithm

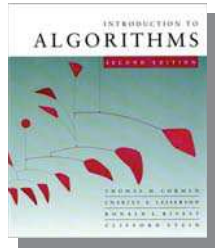
```
for  $i \leftarrow 1$  to  $n$ 
  do for  $j \leftarrow 1$  to  $n$ 
    do  $c_{ij} \leftarrow 0$ 
      for  $k \leftarrow 1$  to  $n$ 
        do  $c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}$ 
```



Standard algorithm

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```

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



Divide-and-conquer algorithm

IDEA:

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

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

$$C = A \cdot B$$

$$r = ae + bg$$

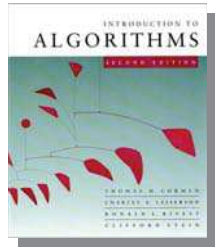
$$s = af + bh$$

$$t = ce + dg$$

$$u = cf + dh$$

8 mults of $(n/2) \times (n/2)$ submatrices

4 adds of $(n/2) \times (n/2)$ submatrices



Divide-and-conquer algorithm

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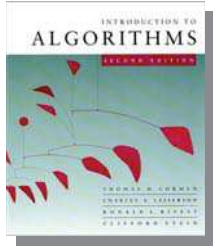
$$t = ce + dh$$

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recursive

8 mults of $(n/2) \times (n/2)$ submatrices

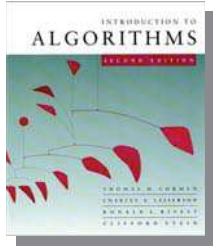
4 adds of $(n/2) \times (n/2)$ submatrices



Analysis of D&C algorithm

$$T(n) = 8T(n/2) + \Theta(n^2)$$

submatrices *submatrix size* *work adding submatrices*

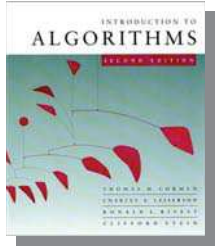


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$$n^{\log_b a} = n^{\log_2 8} = n^3 \Rightarrow \text{CASE 1} \Rightarrow T(n) = \Theta(n^3).$$



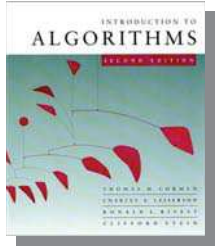
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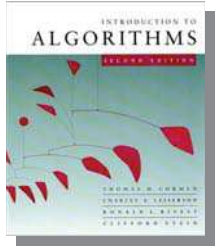
$$n^{\log_b a} = n^{\log_2 8} = n^3 \Rightarrow \text{CASE 1} \Rightarrow T(n) = \Theta(n^3).$$

No better than the ordinary algorithm.



Strassen's idea

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



Strassen's idea

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$$P_1 = a \cdot (f - h)$$

$$P_2 = (a + b) \cdot h$$

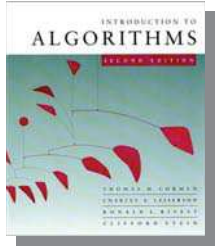
$$P_3 = (c + d) \cdot e$$

$$P_4 = d \cdot (g - e)$$

$$P_5 = (a + d) \cdot (e + h)$$

$$P_6 = (b - d) \cdot (g + h)$$

$$P_7 = (a - c) \cdot (e + f)$$



Strassen's idea

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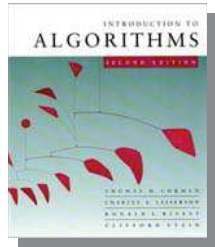
$$P_7 = (a - c) \cdot (e + f)$$

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

$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

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



Strassen's idea

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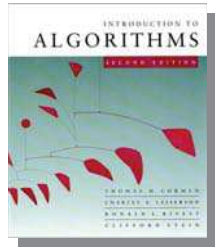
$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

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

7 mults, 18 adds/subs.

Note: No reliance on commutativity of mult!



Strassen's idea

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

$$P_1 = a \cdot (f - h)$$

$$P_2 = (a + b) \cdot h$$

$$P_3 = (c + d) \cdot e$$

$$P_4 = d \cdot (g - e)$$

$$P_5 = (a + d) \cdot (e + h)$$

$$P_6 = (b - d) \cdot (g + h)$$

$$P_7 = (a - c) \cdot (e + f)$$

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

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

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

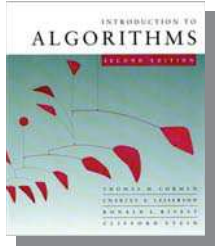
$$+ (b - d)(g + h)$$

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

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

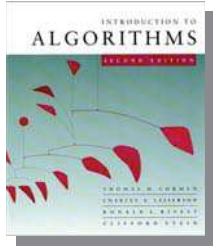
$$+ bg + bh - dg - dh$$

$$= ae + bg$$



Strassen's algorithm

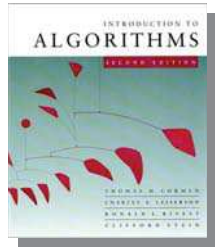
1. **Divide:** Partition A and B into $(n/2) \times (n/2)$ submatrices. Form terms to be multiplied using $+$ and $-$.
2. **Conquer:** Perform 7 multiplications of $(n/2) \times (n/2)$ submatrices recursively.
3. **Combine:** Form C using $+$ and $-$ on $(n/2) \times (n/2)$ submatrices.



Strassen's algorithm

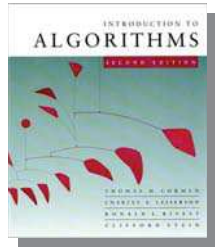
1. **Divide:** Partition A and B into $(n/2) \times (n/2)$ submatrices. Form terms to be multiplied using $+$ and $-$.
2. **Conquer:** Perform 7 multiplications of $(n/2) \times (n/2)$ submatrices recursively.
3. **Combine:** Form C using $+$ and $-$ on $(n/2) \times (n/2)$ submatrices.

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



Analysis of Strassen

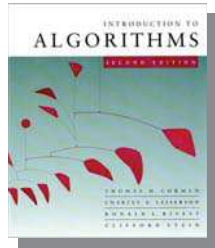
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Analysis of Strassen

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

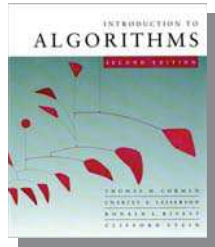


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The number 2.81 may not seem much smaller than 3, but because the difference is in the exponent, the impact on running time is significant. In fact, Strassen's algorithm beats the ordinary algorithm on today's machines for $n \geq 32$ or so.



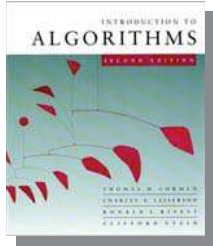
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Best to date (of theoretical interest only): $\Theta(n^{2.376\dots})$.



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).
- The divide-and-conquer strategy often leads to efficient algorithms.