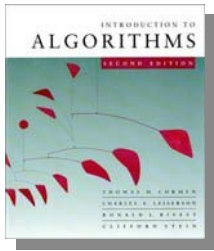


CS60020: Foundations of Algorithm Design and Machine Learning

Sourangshu Bhattacharya

DIVIDE AND CONQUER

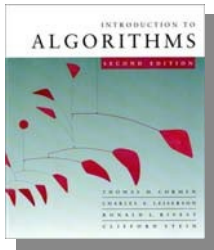


Fibonacci numbers

Recursive definition:

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

0 1 1 2 3 5 8 13 21 34 L



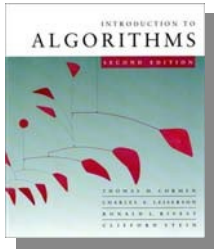
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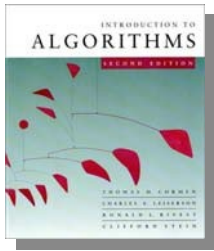
Naive recursive algorithm: $\Omega(\phi^n)$
(exponential time), where $\phi = (1 + \sqrt{5})/2$
is the *golden ratio*.



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

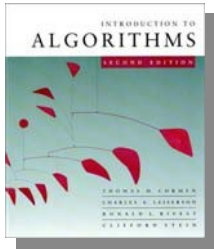
Bottom-up:

- 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:

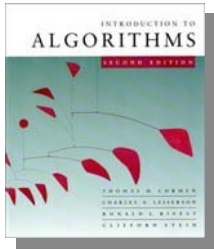
$F_n = \phi^n / \sqrt{5}$ rounded to the nearest integer.

- 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

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

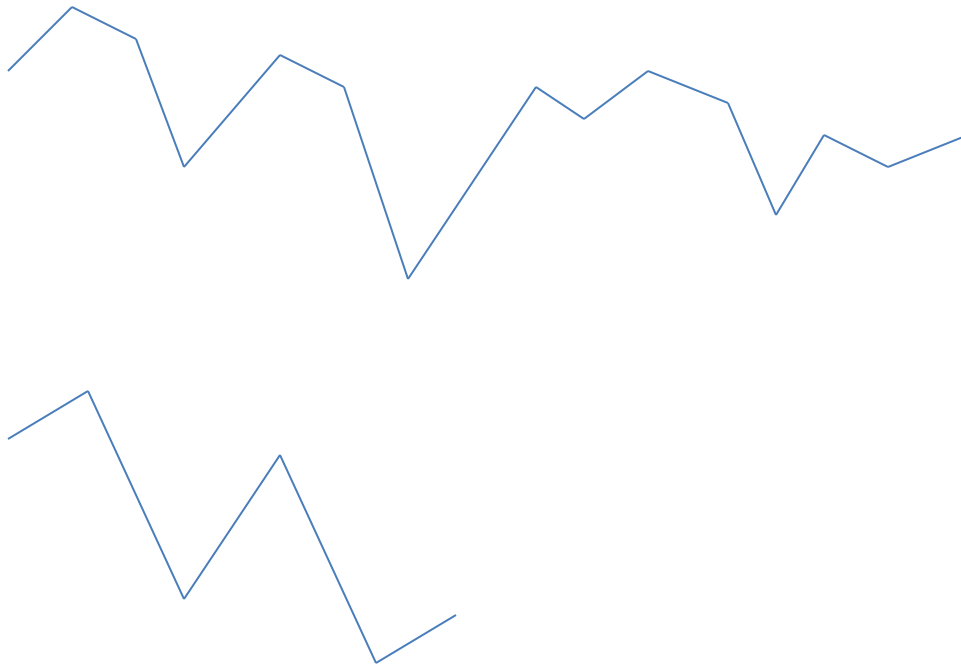
Algorithm: Recursive squaring.

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

Maximum Subarray Problem

- You can buy a unit of stock, only *one* time, then sell it at a later date
 - Buy/sell at end of day
- Strategy: buy low, sell high
 - The lowest price may appear after the highest price
- Assume you know future prices
- Can you maximize profit by buying at lowest price and selling at highest price?

Buy lowest sell highest



Brute force

- How many buy/sell pairs are possible over n days?
- Evaluate each pair and keep track of maximum
- Can we do better?

Transformation

- Find sequence of days so that:
 - the net change from last to first is maximized
- Look at the daily change in price
 - Change on day i : price day i minus price day $i-1$
 - We now have an array of changes (numbers), e.g.
12,-3,-24,20,-3,-16,-23,18,20,-7,12,-5,-22,14,-4,6
 - Find contiguous subarray with largest sum
 - **maximum subarray**
 - E.g.: buy after day 7, sell after day 11

Brute force again

- Trivial if only positive numbers (assume not)
- Need to check $O(n^2)$ pairs
- For each pair, find the sum
- Thus total time is ...

Divide-and-Conquer

- $A[\text{low}..\text{high}]$
- Divide in the middle:
 - $A[\text{low},\text{mid}]$, $A[\text{mid}+1,\text{high}]$
- Any subarray $A[i,..j]$ is
 - (1) Entirely in $A[\text{low},\text{mid}]$
 - (2) Entirely in $A[\text{mid}+1,\text{high}]$
 - (3) In both
- (1) and (2) can be found recursively

Divide-and-Conquer (cont.)

- (3) find maximum subarray that crosses midpoint
 - Need to find maximum subarrays of the form $A[i..mid]$, $A[mid+1..j]$, $low \leq i$, $j \leq high$
- Take subarray with largest sum of (1), (2), (3)

Divide-and-Conquer (cont.)

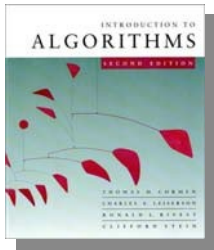
```
Find-Max-Cross-Subarray(A,low,mid,high)
  left-sum =  $-\infty$ 
  sum = 0
  for i = mid downto low
    sum = sum + A[i]
    if sum > left-sum then
      left-sum = sum
      max-left = i
  right-sum =  $-\infty$ 
  sum = 0
  for j = mid+1 to high
    sum = sum + A[j]
    if sum > right-sum then
      right-sum = sum
      max-right = j
  return (max-left, max-right, left-sum + right-sum)
```


Time analysis

- Find-Max-Cross-Subarray: $O(n)$ time
- Two recursive calls on input size $n/2$
- Thus:

$$T(n) = 2T(n/2) + O(n)$$

$$T(n) = O(n \log n)$$

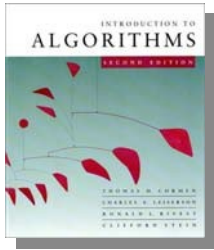


Matrix multiplication

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

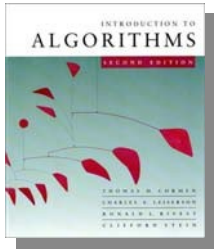
$$\begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots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

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



Standard algorithm

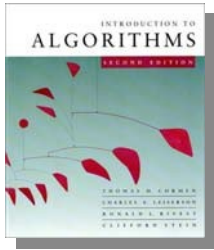
```
for  $i \leftarrow 1$  to  $n$   
  do for  $j \leftarrow 1$  to  $n$   
    do  $c_{ij} \leftarrow 0$   
      for  $k \leftarrow 1$  to  $n$   
        do  $c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}$ 
```



Standard algorithm

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for  $i \leftarrow 1$  to  $n$ 
  do for  $j \leftarrow 1$  to  $n$ 
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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:

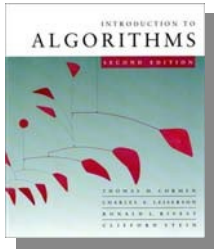
$$\begin{bmatrix} r & s \\ t & u \end{bmatrix} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \cdot \begin{bmatrix} e & f \\ g & h \end{bmatrix}$$

$$C = A \cdot B$$

$$\left. \begin{array}{l} r = ae + bg \\ s = af + bh \\ t = ce + dg \\ u = cf + dh \end{array} \right\}$$

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

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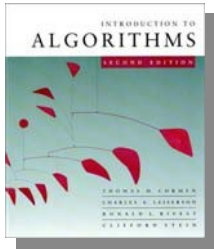
$$C = A \cdot B$$

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recursive

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



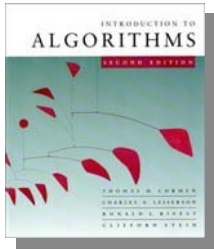
Analysis of D&C algorithm

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

submatrices

submatrix size

*work adding
submatrices*



Analysis of D&C algorithm

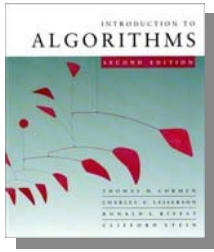
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$$n^{\log_b a} = n^{\log_2 8} = n^3 \quad \Rightarrow \quad \text{CASE 1} \quad \Rightarrow \quad T(n) = \Theta(n^3).$$



Analysis of D&C algorithm

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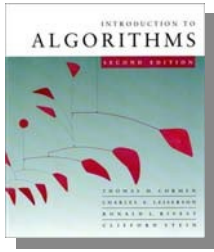
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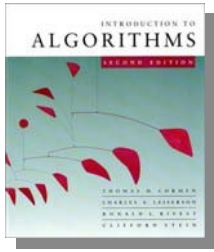
$$n^{\log_b a} = n^{\log_2 8} = n^3 \Rightarrow \text{CASE 1} \Rightarrow T(n) = \Theta(n^3).$$

No better than the ordinary algorithm.



Strassen's idea

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



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

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

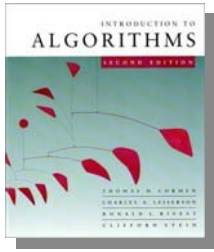
$$P_3 = (c + d) \cdot e$$

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

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

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

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



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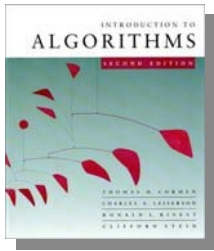
$$P_7 = (a - c) \cdot (e + f)$$

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

$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

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



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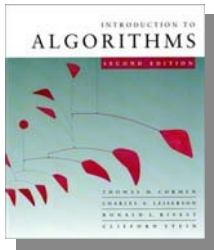
$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

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

7 mults, 18 adds/subs.

Note: No reliance on commutativity of mult!



Strassen's idea

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

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

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

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

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

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

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

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

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

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

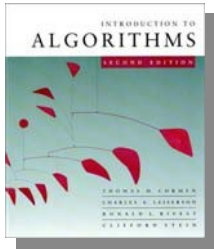
$$+ (b - d)(g + h)$$

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

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

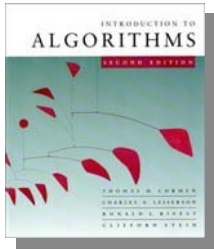
$$+ bg + bh - dg - dh$$

$$= ae + bg$$



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.



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