

**MA 515: Introduction to Algorithms &
MA353 : Design and Analysis of Algorithms
[3-0-0-6]**

Lecture 6

http://www.iitg.ernet.in/psm/indexing_ma353/y09/index.html

Partha Sarathi Manal

psm@iitg.ernet.in

Dept. of Mathematics, IIT Guwahati

Mon 10:00-10:55 Tue 11:00-11:55 Fri 9:00-9:55

Class Room : 2101

Binary search

Find an element in a sorted array:

Divide: Check middle element.

Conquer: Recursively search 1 subarray.

Combine: Trivial.

Example: Find 9

3 5 7 8 9 12 15

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Recurrence for binary search

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

subproblems *subproblem size* *work dividing and combining*

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

Powering a number

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

Naive algorithm: $\Theta(n)$.

Divide-and-conquer algorithm:

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

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

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 ...

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

Fibonacci numbers

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is the *golden ratio*.

Golden Ratio

- Two quantities are in the **golden ratio** if the ratio of the sum of the quantities to the larger one equals the ratio of the larger one to the smaller.
- Two quantities a and b are said to be in the **golden ratio** ϕ if: $a+b/a = a/b = \phi$, if $a = \phi b$ then
- $\phi^2 - \phi - 1 = 0 \Rightarrow \phi = 1 + \sqrt{5} / 2$
 $= 1.6180339887\dots$
- $F(n) = \phi^n - (1-\phi)^n / \sqrt{5}$.

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:

$F_n = \phi^n / \text{Sqrt}(5)$ 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.

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  $c_{ij} \leftarrow c_{ij} + a_{ik} \cdot b_{kj}$ 
```

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

Divide-and-conquer algorithm

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

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

$$C = A \cdot B$$

$$\begin{aligned} r &= ae + bg \\ s &= af + bh \\ t &= ce + dg \\ u &= cf + dh \end{aligned}$$

- 8 mults of $(n/2) \times (n/2)$ sub-matrices
- 4 adds of $(n/2) \times (n/2)$ sub-matrices

recursively

Analysis of D&C algorithm

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

submatrices *submatrix size* *work adding submatrices*

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

- No better than the ordinary algorithm.

Strassen's idea

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

$$C = A \cdot B$$

- 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

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

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

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

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

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

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

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

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

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

$$\begin{aligned} r &= P_5 + P_4 - P_2 + P_6 \\ &= (a + d)(e + h) \\ &\quad + d(g - e) - (a + b)h \\ &\quad + (b - d)(g + h) \\ &= ae + ah + de + dh \\ &\quad + dg - de - ah - bh \\ &\quad + bg + bh - dg - dh \\ &= ae + bg \end{aligned}$$

Strassen's algorithm (1969)

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

$C = A \cdot B$

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

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

Analysis of Strassen

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

- $n^{\log_b a} = n^{\log_2 7} = n^{2.807} \Rightarrow \text{CASE 1} \Rightarrow T(n) = \Theta(n^{\lg 7})$.
- The number **2.807** 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.

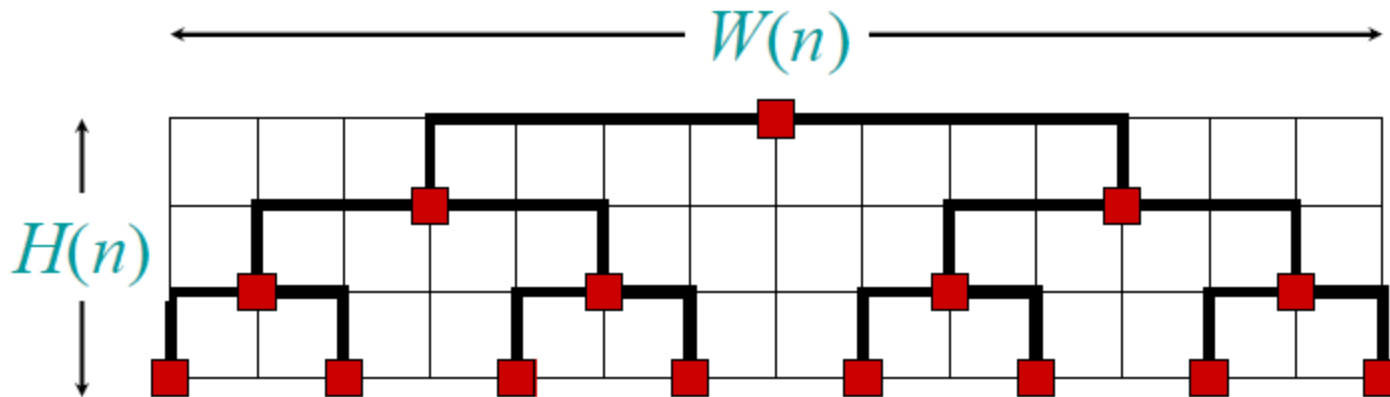
Coppersmith–Winograd algorithm (1987)

- **Best to date (of theoretical interest only): $\Theta(n^{2.376\dots})$.**
 - asymptotically fastest known algorithm for **square matrix multiplication** as of 2008.
 - It is frequently used as a building block in other algorithms to prove theoretical time bounds.
 - it is not used in practice because it only provides an advantage for **matrices so large** that they cannot be processed by modern hardware.

“Toward an Optimal Algorithm for Matrix Multiplication”,
SIAM News, Volume 38, Number 9, November 2005

VLSI layout

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



$$\begin{aligned} H(n) &= H(n/2) + \Theta(1) & W(n) &= 2W(n/2) + \Theta(1) \\ &= \Theta(\lg n) & &= \Theta(n) \end{aligned}$$

$$\text{Area} = \Theta(n \lg n)$$