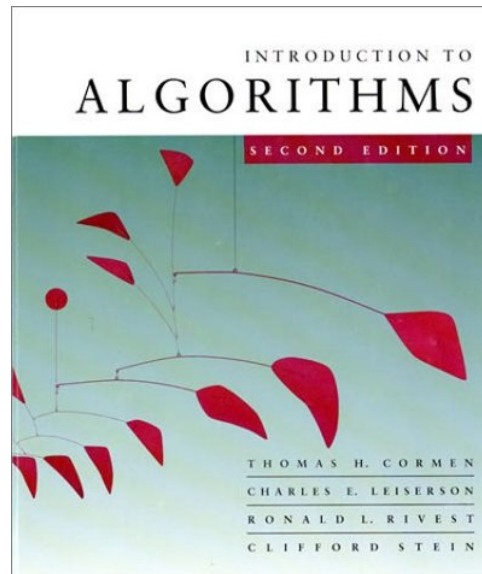


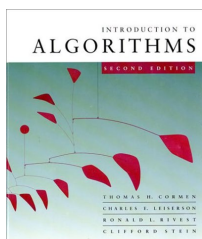
Introduction to Algorithms

6.006



Lecture 23

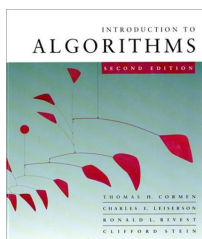
Jelani Nelson



Menu

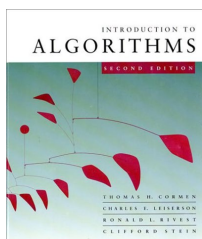
- “Numerics” - algorithms for operations on **large** numbers
 - Cryptography, simulations, etc
- Operations
 - Addition
 - Multiplication
 - – Division
 - Matrix multiplication

3.14159265358979323846264338327950288419
7169399375105820974944592307816406286208
9986280348253421170679821480865132823066
4709384460955058223172535940812848111745
0284102701938521105559644622948954930381
9644288109756659334461284756482337867831
6527120190914564856692346034861045432664
8213393607260249141273724587006...



Division a/b

- Inversion:
 - Given positive n -digit number b (radix r)
 - Compute $1/b$
 - I.e., for some $R=r^k$, compute $\lfloor R/b \rfloor$
- Iterative algorithm:
 - Start from some x_0
 - Keep computing x_{i+1} from x_i
 - Show this converges to the answer

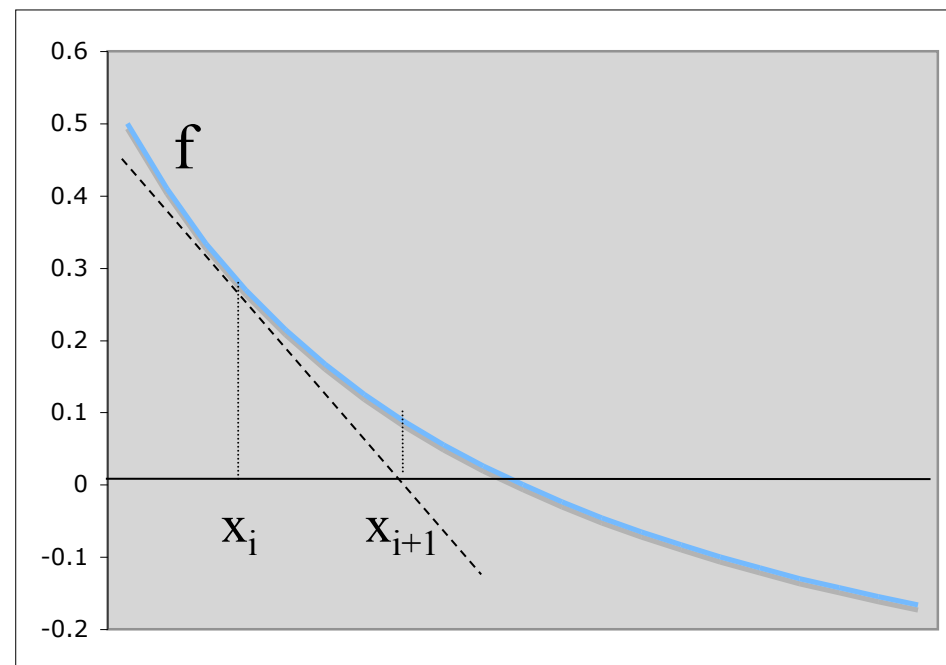


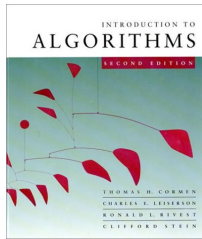
Newton's method

- Iterative approach to solving $f(x)=0$
 - In our case $f(x)=1/x-b/R$
- Iterative step:
 - Find a line
 $y = f(x_i) + f'(x_i)(x - x_i)$
tangent to $f(x)$ at x_i
 - Set x_{i+1} to the solution of

$$f(x_i) + f'(x_i)(x - x_i) = 0$$

$$\text{I.e., } x_{i+1} = x_i - f(x_i) / f'(x_i)$$





Division: algorithm

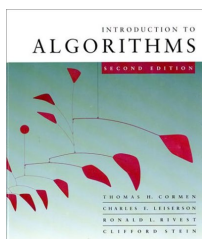
- Want to solve $f(x) = 1/x - b/R = 0$
- We have $f'(x) = -1/x^2$
- Iterative step $x_{i+1} = x_i - f(x_i)/f'(x_i)$ yields

$$x_{i+1} = x_i + x_i^2 (1/x_i - b/R)$$

I.e.,

$$x_{i+1} = 2x_i - x_i^2 b/R$$

- Only $O(1)$ multiplications, shifts and subtractions
- Convergence ?

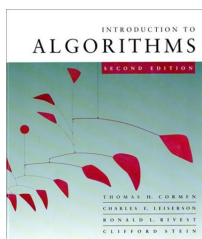


Convergence of $x_{i+1} = 2x_i - x_i^2 b/R$

- Assume $x_i = R/b (1+e_i)$, e_i = error
- Assumptions:
 - $|e_i|$ is small
 - Ignore the round-off errors caused by the “/R” operation
- How does each iteration affect e_i ?

$$\begin{aligned}x_{i+1} &= 2x_i - x_i^2 b/R \\ &= 2R/b (1+e_i) - [R/b (1+e_i)]^2 b/R \\ &= R/b [2+2e_i - 1 - 2e_i - e_i^2] \\ &= R/b [1 - e_i^2] \\ &= R/b [1 + e_{i+1}], \text{ where } e_{i+1} = -e_i^2\end{aligned}$$

- We have $|e_{i+1}| = |e_i|^2$ - “quadratic” convergence



Quadratic convergence $|e_{i+1}| = |e_i|^2$

- If we start close enough (say, $|e_0| < 1/2$), then the convergence is very fast:

$$|e_1| \leq |e_0|^2$$

$$|e_2| \leq |e_0|^{2^2}$$

...

$$|e_i| \leq |e_0|^{2^i} < 1/2^{2^i}$$

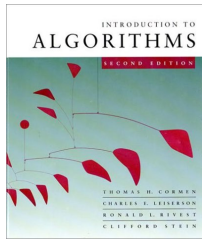
- To make sure we get k digits of precision, we need

$$|e_i| < 1/r^k$$

$$1/2^{2^i} < 1/r^k$$

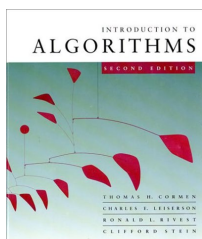
$$i > \log_2 (k \log_2 r)$$

- What if $|e_0| > 1/2$?
 - Heuristically, can apply the same algorithm
 - Analysis much more complicated
 - Does not always converge



General method

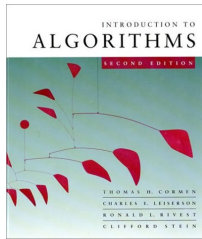
- E.g., square roots
 - $f(x) = x^2 - a$
 - $x_{i+1} = x_i - (x_i^2 - a) / 2 x_i$
 $= (x_i + a/x_i) / 2$
- Only $O(1)$ multiplications, subtractions and divisions



Matrix multiplication

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

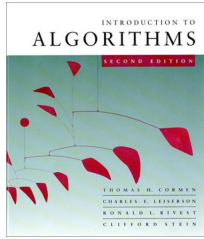
$$c_{ij} = \sum_k a_{ik} b_{kj}$$



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 $(n/2) \times (n/2)$ submatrices:

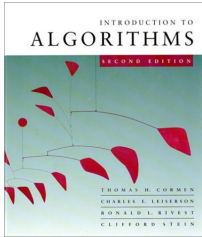
$$\begin{pmatrix} r & s \\ t & u \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} e & f \\ g & h \end{pmatrix}$$

$$C = A \cdot B$$

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

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

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



Analysis of D&C algorithm

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

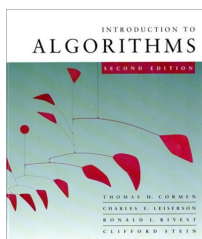
submatrices

submatrix size

work adding
submatrices

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

No better than the ordinary algorithm.



Strassen's idea (1969)

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

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

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

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

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

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

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

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

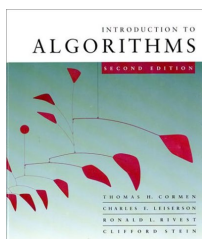
$$P_7 = (a - c) \cdot (e + f)$$

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

$$s = P_1 + P_2$$

$$t = P_3 + P_4$$

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



Strassen's idea

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

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

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

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

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

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

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

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

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

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

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

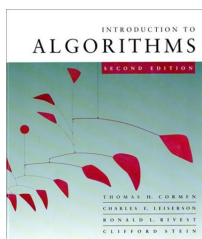
$$+ (b - d)(g + h)$$

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

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

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

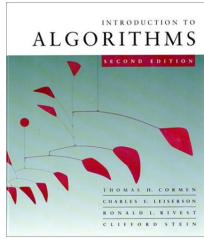
$$= ae + bg$$



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.

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



Analysis of Strassen

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

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

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