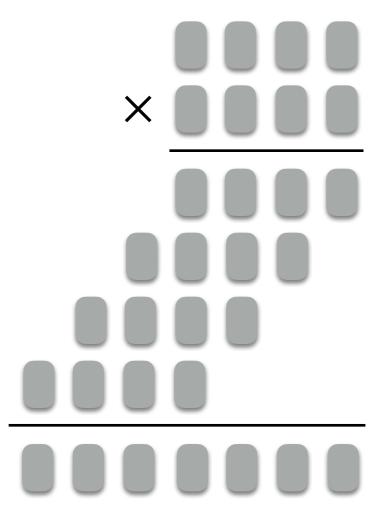
# CSE202 Design and Analysis of Algorithms

Week 2 — Divide & Conquer 1: How fast can we multiply?

# Naive Multiplication



Input: two *n*-digit integers

*n* multiplications + O(n) carries

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*n* multiplications + O(n) carries

 $O(n^2)$  additions + O(n) carries

Output:  $\leq 2n$  digits

Total:  $O(n^2)$  digit (or bit) operations

For integers  $\leq N$ this is  $O(\log^2 N)$ 

Quadratic algorithm: #operations  $O(n^2)$  for an input size n

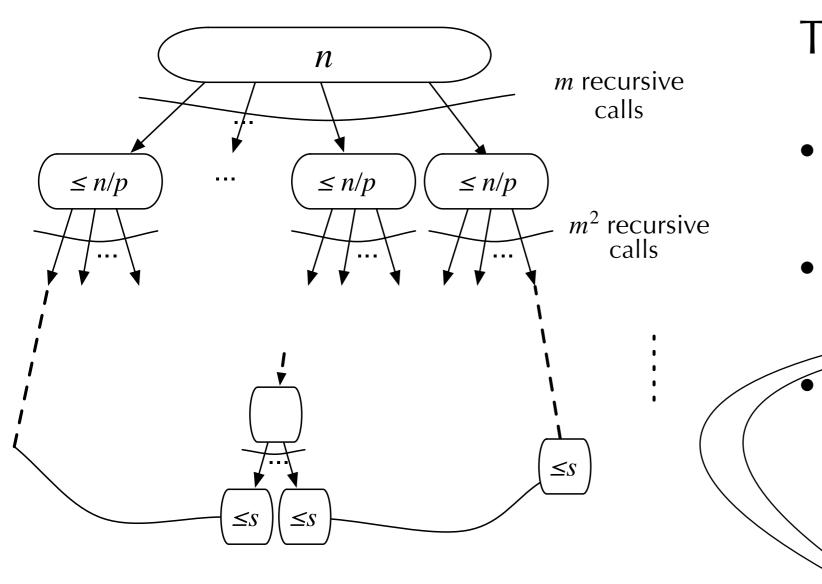
# One Second of Computation

With a good polynomial or integer library, 1 sec. is sufficient to

multiply two integers with 30,000,000 digits; multiply two polynomials of degree 650,000; multiply two matrices of size 850x850; (but factor an integer with 42 digits only).

1 sec. is in the asymptotic regime of the algorithms

# **Divide and Conquer**



The total cost may be

- mostly at the top (quickselect)
- mostly at the leaves

Karatsuba, Strassen

balanced along the levels (binary powering, mergesort)

this week

# I. Polynomials

# Polynomials and Integers

Polynomials Integers
$$(7x^{2} + 3x + 5) \times (2x + 3) = (7x^{2} + 3x + 5) \times 2x + (7x^{2} + 3x + 5) \times 2x = 21x^{2} + 9x + 15 \times 10 \times 100 \times 100$$

Polynomials behave like integers, without carries

# Divide & Conquer by Itself Doesn't Help

F and G of degree  $\langle n \mapsto H := FG$ 

- 1. If n = 1 return FG
- 2. Let  $k := \lceil n/2 \rceil$
- 3. Split  $F = F_0 + x^k F_1$ ,  $G = G_0 + x^k G_1$  $F_0, F_1, G_0, G_1$  of degree < k
- 4. Compute recursively

$$H_0 := F_0G_0, H_1 := F_0G_1, H_2 := F_1G_0, H_3 := F_1G_1$$

5. Return  $H_0 + x^k(H_1 + H_2) + x^{2k}H_3$ 

Complexity:  $C(n) \le 4C(\lceil n/2 \rceil) + \lambda n$  coefficient operations

# Complexity of the Naive DAC

$$C(n) \le 4C(\lceil n/2 \rceil) + \lambda n$$

iterate once

$$\leq \lambda n + 4\lambda \lceil n/2 \rceil + 16C(\lceil n/2 \rceil_2)$$

**Notation:** 

$$\lceil x/2 \rceil_1 = \lceil x/2 \rceil$$

$$\lceil x/2 \rceil_{k+1} = \lceil \lceil x/2 \rceil_k / 2 \rceil$$

N: power of 2 s.t.

$$n \le N < 2n$$

$$\leq \lambda \left(n + 2(2\lceil n/2 \rceil_1) + \dots + 2^{k-1}(2^{k-1}\lceil n/2 \rceil_{k-1})\right) + 2^{k+1}C(\lceil n/2 \rceil_k)$$

use N

$$\leq \lambda N \left( 1 + 2 + \dots + 2^{k-1} \right) + 4^k C(\lceil n/2 \rceil_k)$$

bound geometric series

$$\leq 4^k \left( \lambda \frac{N}{2^k} + C(\lceil n/2 \rceil_k) \right)$$

use 
$$k = \lceil \log_2 n \rceil$$

$$= O(n^2)$$

An extra idea is needed to beat the naive algorithm

# Polynomials of Degree 1

$$F = f_0 + f_1 T$$
,  $G = g_0 + g_1 T$   $\mapsto$   $H := FG = h_0 + h_1 T + h_2 T^2$ 

Naive algorithm:

$$H = (\underbrace{f_0 g_0}) + (\underbrace{f_0 g_1} + \underbrace{f_1 g_0}) T + \underbrace{f_1 g_1} T^2 \qquad \text{4 multiplications}$$
 & 1 addition

Interpolation from 3 values:

$$h_0 = F(0)G(0) = f_0 g_0$$
 1 mult.  
 $h_2 = {}^{\shortparallel}F(\infty)G(\infty){}^{\shortparallel} = f_1 g_1$  1 mult.  
 $\tilde{h}_1 = h_0 + h_1 + h_2 = F(1)G(1) = (f_0 + f_1)(g_0 + g_1)$  1 mult.

$$FG = h_0 + (\tilde{h}_1 - h_0 - h_2)T + h_2T^2$$

3 multiplications, 2 additions, 2 subtractions

# Karatsuba's Algorithm

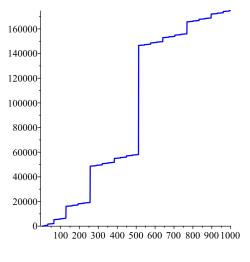
F and G of degree  $\langle n \mapsto H := FG$ 

Idea: Evaluate  $FG = h_0 + (\tilde{h}_1 - h_0 - h_2)T + h_2T^2$  at  $T = x^k$ .

- 1. If *n* is small, use naive multiplication
- 2. Let k := [n/2]
- 3. Split  $F = F_0 + x^k F_1$ ,  $G = G_0 + x^k G_1$  $F_0, F_1, G_0, G_1$  of degree < k
- 4. Compute recursively

$$H_0 := F_0 G_0, H_2 := F_1 G_1, \tilde{H}_1 := (F_0 + F_1)(G_0 + G_1)$$

5. Return 
$$H_0 + x^k(\tilde{H}_1 - H_0 - H_2) + x^{2k}H_2$$



$$u_n = n + 3u_{\lceil n/2 \rceil}, u_1 = 1$$

Complexity:  $C(n) \le 3C(\lceil n/2 \rceil) + \lambda n$  coefficient operations

# Complexity of Karatsuba's Algorithm

$$C(n) \le 3C(\lceil n/2 \rceil) + \lambda n$$

iterate once

$$\leq \lambda n + 3\lambda \lceil n/2 \rceil + 9C(\lceil n/2 \rceil_2)$$

iterate k-1 times, use N

$$\leq \lambda N \left( 1 + \frac{3}{2} + \dots + \left( \frac{3}{2} \right)^{k-1} \right) + 3^k C(\lceil n/2 \rceil_k)$$

reorder sum

$$\leq \lambda N \left(\frac{3}{2}\right)^{k-1} \left(1 + 2/3 + \dots + (2/3)^{k-1}\right) + 3^k C(\lceil n/2 \rceil_k)$$

bound geometric series

$$\leq 3^k \left( 2\lambda \frac{N}{2^k} + C(\lceil n/2 \rceil_k) \right)$$

 $use \\ k = \lceil \log_2 n \rceil$ 

$$\leq (2\lambda + 1)3^{\lceil \log_2 n \rceil} = O(n^{\log_2 3})$$

$$\approx 1.58$$

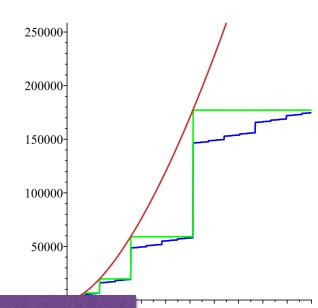
#### **Notation:**

$$\lceil x/2 \rceil_1 = \lceil x/2 \rceil$$

$$\lceil x/2 \rceil_{k+1} = \lceil \lceil x/2 \rceil_k / 2 \rceil$$

$$N$$
: power of 2 s.t.

$$n \le N < 2n$$



Not the final word. See assignment.

00 400 500 600 700 800 900 1000

# II. Integers

# From Polynomials to Integers

Recall: Polynomials behave like integers, without carries

No theorem of complexity equivalence exists, but the algorithms over polynomials can often be adapted to integers, with the same complexity.

# Karatsuba's Algorithm for Integers

$$F$$
 and  $G$  integers  $< 2^n \mapsto H := FG$ 

- 1. If *n* is small, use naive multiplication
- 2. Let  $k := \lceil n/2 \rceil$
- 3. Split  $F = F_0 + 2^k F_1$ ,  $G = G_0 + 2^k G_1$  $F_0, F_1, G_0, G_1 < 2^k$

x into 2 in the polynomial version.

Obtained by changing

4. Compute recursively

$$H_0 := F_0 G_0, \quad H_2 := F_1 G_1, \quad \tilde{H_1} := (F_0 + F_1)(G_0 + G_1)$$

5. Return  $H_0 + 2^k (\tilde{H}_1 - H_0 - H_2) + 2^{2k} H_2$ 

Same algorithm as for polynomials, similar (not exactly the same) complexity analysis.

$$\rightarrow O(n^{\log_2 3})$$
 bit operations

$$\approx 1.58$$

# Which of these Algorithms is Best?

#### None of them!

#### GMP (the Gnu Multiprecision Library) uses:

# 64-bit words	approx # digits	Algorithm		
0	0	Naive		
26	500	Karatsuba		
73	1,400	Toom - 3		
208	4,000	Toom - 4		
4736	90,000	FFT		



## III. Matrices

## **Matrix Multiplication**

Input: two  $n \times n$  matrices A, X with  $n = 2^k$ 

Output: AX

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

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

Naive algorithm: O(n<sup>3</sup>)

# Matrix Multiplication: Strassen's Algorithm

Input: two  $n \times n$  matrices A, X with  $n = 2^k$ 

Output: AX

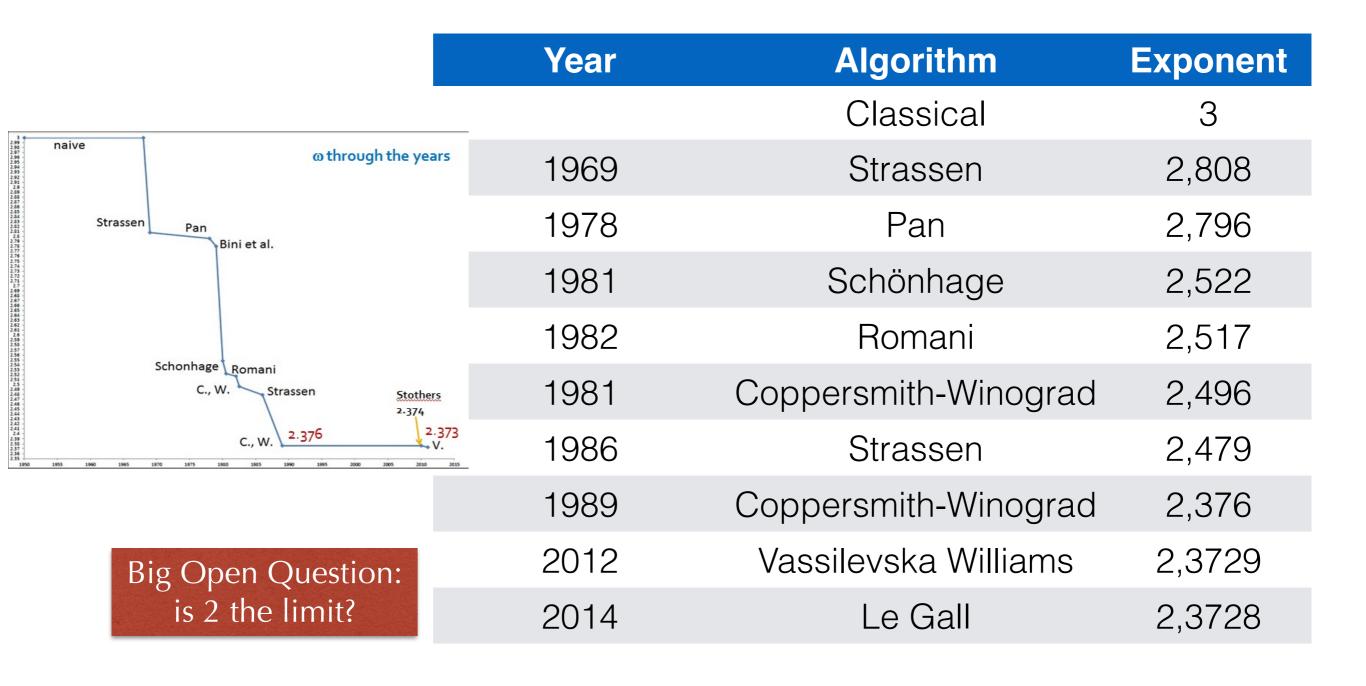
- 1. If n = 1, return AX
- 2. Split  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}, X = \begin{pmatrix} x & y \\ z & t \end{pmatrix}$ , with  $(n/2) \times (n/2)$  blocks
- 3. Compute recursively the 7 products

$$q_1 = a(x+z), \ q_2 = d(y+t), \ q_3 = (d-a)(z-y),$$
  $q_4 = (b-d)(z+t), \ q_5 = (b-a)z,$  Exercise: prove the complexity  $q_6 = (c-a)(x+y), \ q_7 = (c-d)y$  in  $O(n^{\log_2 7})$  operations.

4. Return  $\begin{pmatrix} q_1 + q_5 & q_2 + q_3 + q_4 - q_5 \\ q_1 + q_3 + q_6 - q_7 & q_2 + q_7 \end{pmatrix}$ 

## **World Records**

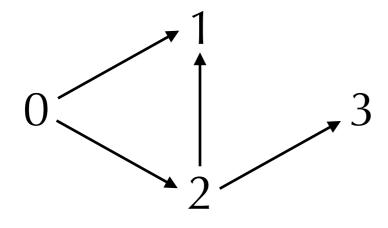
### Best algorithm for matrix multiplication



# **Application: Graph Transitive Closure**

**Def.** A graph is a pair (V,E) where

- 1. V is a finite set of nodes/vertices
- 2.  $E \subseteq V \times V$  is a finite set of edges



$$(\{0,1,2,3\},\{(0,1),(0,2),(2,1),(2,3)\})$$

#### Adjacency matrix

Nodes				
0	0	1	1 0 0 0	0
1	0	0	0	0
2	0	1	0	1
3	0	0	0	0

Boolean matrix

# **Boolean Matrices: Multiplication**

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

$$c_{ij} = \bigvee_{k=1}^{n} a_{ik} \wedge b_{kj}$$

Boolean ops (and, or) instead of algebraic ops (sum, product)

Algebraic multiplication + flatten all non-zero entries to 1

# **Graph Transitive Closure**

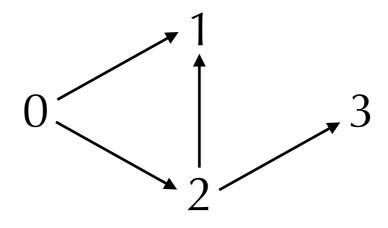
Let G = (V,E) be a graph

A path from i to j is a sequence of edges  $e_1,...,e_n$  such that:

- the source of e<sub>1</sub> is i and the target of e<sub>n</sub> is j
- for every  $1 \le k < n$ , the target of  $e_k$  is the source of  $e_{k+1}$

The transitive closure  $G^* = (V, E^*)$  is the graph where (u,v)  $\in E^*$  iff there is a path from u to v

# **Application: Graph Transitive Closure**



$$(\{0,1,2,3\},\{(0,1),(0,2),(2,1),(2,3)\})$$

#### Adjacency matrix

If A is the adjacency matrix of a graph, then

- $(A^k)_{ij} = 1$  iff there is a path of length k from i to j
- if I is the identity matrix, then  $(A \lor I)_{ij}^{k} = 1$  iff there exists a path of length at most k from i to j

# **Application: Graph Transitive Closure**

Let G = (V,E) be a graph with n vertices

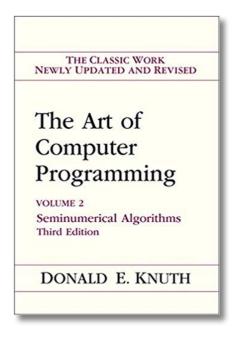
If A is the adjacency matrix of G, then  $(A \lor I)^{n-1}$  is the adjacency matrix of G\*

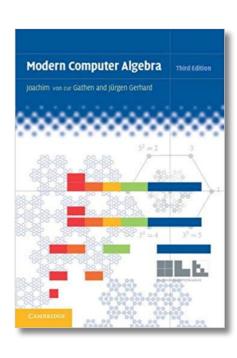
The matrix  $(A \lor I)^{n-1}$  can be computed by  $\log n$  squaring operations/multiplications  $O(n^{\log_2 7} \cdot \log_2 n)$ 

## References for this lecture

The slides are designed to be self-contained.

They were prepared using the following books that I recommend if you want to learn more:





## Next

Assignment this week: generalisation of Karatsuba's algorithm

Next tutorial: multiplication, transitive closure

Next week: algorithms related to statistics

## **Feedback**

**Moodle** 

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