# Lecture 4: Integer and Matrix Multiplication

More complicated examples of divide-and-conquer

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# Integer Arithmetic

Add. Given two n-bit integers a and b, compute a + b.

 $\bullet$   $\Theta(n)$  time

Multiply. Given two n-bit integers a and b, compute  $a \cdot b$ .

■ Primary school method:  $\Theta(n^2)$  time.

- A.k.a. "long multiplication"

	1	1	0	1	0	1	0	1
+	0	1	1	1	1	1	0	1
1	0	1	0	1	0	0	1	0

							1	1	0	1	0	1	0	1
						×	0	1	1	1	1	1	0	1
							1	1	0	1	0	1	0	1
						0	0	0	0	0	0	0	0	
					1	1	0	1	0	1	0	1		
				1	1	0	1	0	1	0	1			
			1	1	0	1	0	1	0	1				
		1	1	0	1	0	1	0	1		ı			
	1	1	0	1	0	1	0	1						
0	0	0	0	0	0	0	0							
1	1	0	1	0	0	0	0	0	0	0	0	0	0	1

### Divide-and-Conquer Multiplication: First Attempt

#### Observation:

- Let X,a, b, c, d be integers
- Simple algebra says

$$(aX + b) (cX + d) = acX^2 + (ad + bc)X + bd$$

• If  $X = 2^{n/2}$  this becomes

$$(a2^{n/2} + b)(c2^{n/2} + d) = ac2^n + (ad + bc)2^{n/2} + bd$$

Example

$$163 97 = 15,811$$
$$(10 \cdot 2^4 + 3)(6 \cdot 2^4 + 1) = 60 \cdot 2^8 + (10 + 18)2^4 + 3$$

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### Divide-and-Conquer Multiplication: First Attempt

#### Recall:

• If  $X = 2^{n/2}$ 

$$(a2^{n/2} + b)(c2^{n/2} + d) = ac2^n + (ad + bc)2^{n/2} + bd$$

- Integers are stored in computers in binary format.
  - Multiplication by  $2^k$  can be done in one time unit by performing a left shift of k bits
- Example 10 = 00001010
  - $\rightarrow$  10 X  $2^3 = 80$
  - > is the same as left shift of 3
  - > 00001010 << 3 = 01010000 = 80

Note: In the sequel, for simplicity, we write  $\times 2^k$ . This should be read as an O(1) time left shift << k.

### Divide-and-Conquer Multiplication: First Attempt

```
(75)(218) = (4 \cdot 2^{4} + 11)(13 \cdot 2^{4} + 10)
= 4 \cdot 13 \cdot 2^{8} + (4 \cdot 10 + 11 \cdot 13)2^{4} + 11 \cdot 10
= 52 \cdot 2^{8} + 183 \cdot 2^{4} + 110
= 16,350
```

```
0100\ 1011 \times 1101\ 1010 = (0100 \cdot 2^{4} + 1011) \times (1101 \cdot 2^{4} + 1010)
= (0100 \times 1101) 2^{8}
+ ((0100 \times 1010) + (1011 \times 1101)) 2^{4}
+ 1011 \times 1010
```

### In general:

■ Let  $a = a_1 2^{n/2} + a_0$ , and  $b = b_1 2^{n/2} + b_0$ , where  $a_1, a_0, b_1, b_0$  are all (n/2)-bit integers.

$$\Rightarrow ab = a_1b_12^n + (a_1b_0 + a_0b_1)2^{n/2} + a_0b_0$$

The first divide-and-conquer algorithm for integer multiplication

Suppose the bits are stored in arrays A[1..n] and B[1..n], A[1] and B[1] are the least significant bits

```
Multiply (A, B):
n \leftarrow \text{size of } A
if n = 1 then return A[1] \cdot B[1]
mid \leftarrow \lfloor n/2 \rfloor
U \leftarrow \text{Multiply}(A[mid + 1..n], B[mid + 1..n])
                                                                   a_1b_1
V \leftarrow \text{Multiply}(A[mid + 1..n], B[1..mid])
                                                                   a_1b_0
W \leftarrow \text{Multiply}(A[1..mid], B[mid + 1..n])
                                                                  a_0b_1
Z \leftarrow Multiply(A[1..mid], B[1..mid])
                                                                   % a_0b_0
M[1..2n] \leftarrow 0
M[1..n] \leftarrow Z
                                                                   a_0b_0
M[mid + 1..] \leftarrow M[mid + 1..] \oplus V \oplus W
                                                                    * + (a_1b_0 + a_0b_1) \ll n/2 
M[2mid + 1..] \leftarrow M[2mid + 1..] \oplus U
                                                                     + a_1 b_1 \ll n 
return M
```

⊕: denotes the integer addition algorithm

### Analysis (Expansion Method)

#### Recurrence.

For, 
$$n > 1$$
,  $T(n) = 4T(n/2) + n$ .  $T(1) = 1$ 

$$T(n) = 4 T\left(\frac{n}{2}\right) + n$$
$$= 4\left(4T\left(\frac{n}{4}\right) + \frac{n}{2}\right) + n$$

$$=4^{2}\left(4T\left(\frac{n}{2^{3}}\right)+\frac{n}{2^{2}}\right)+\frac{4}{2}n+n$$

$$= 4^{3} \left(4T\left(\frac{n}{2^{4}}\right) + \frac{n}{2^{3}}\right) + \frac{4^{2}}{2^{2}}n + \frac{4}{2}n + n$$

$$=4^2 T\left(\frac{n}{2^2}\right) + \frac{4}{2}n + n$$

$$= 4^{3} T\left(\frac{n}{2^{3}}\right) + \frac{4^{2}}{2^{2}}n + \frac{4}{2}n + n$$

$$= 4^4 T\left(\frac{n}{2^4}\right) + \left(\frac{4^3}{2^3} + \frac{4^2}{2^2} + \frac{4}{2} + 1\right)n$$

····

$$=4^{i} T\left(\frac{n}{2^{i}}\right) + \left(\frac{4^{i-1}}{2^{i-1}} + \frac{4^{i-2}}{2^{i-2}} + \dots + \frac{4}{2} + 1\right)n$$

### Analysis (Expansion Method)

$$T(n) = 4^{i} T\left(\frac{n}{2^{i}}\right) + \left(\frac{4^{i-1}}{2^{i-1}} + \frac{4^{i-2}}{2^{i-2}} + \dots + \frac{4}{2} + 1\right) n$$

....

$$= 4^{h} T\left(\frac{n}{2^{h}}\right) + \left(\left(\frac{4}{2}\right)^{h-1} + \left(\frac{4}{2}\right)^{h-2} + \dots + \left(\frac{4}{2}\right) + 1\right) n$$

$$= n^2 T\left(\frac{n}{n}\right) + \left(2^{h-1} + 2^{h-2} + \dots + 2 + 1\right)n$$

$$= n^2 T(1) + \left(\frac{2^h - 1}{2 - 1}\right) n$$

$$=n^2+\left(\frac{n-1}{1}\right)n$$

$$= n^2 + n(n-1) = \Theta(n^2)$$

$$h = \log_2 n$$

$$2^h = n$$

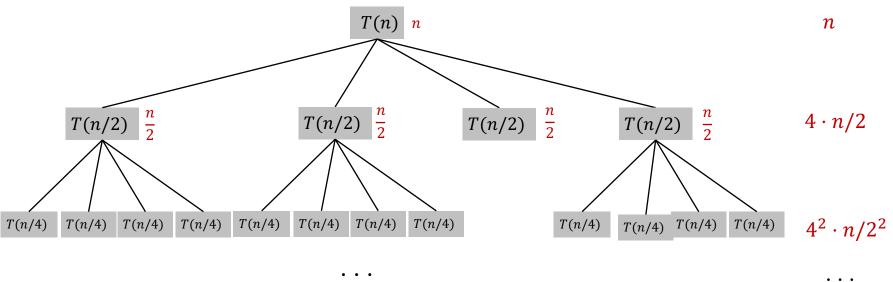
$$4^h = (2^2)^h = (2^h)^2 = n^2$$

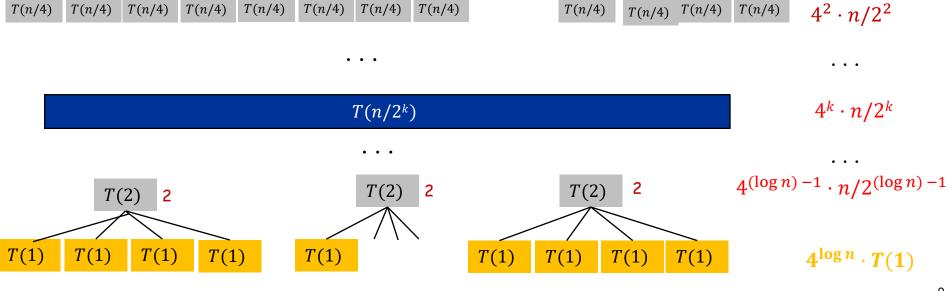
# Analysis (Tree Method)

#### Recurrence:

$$T(n) = 4T(n/2) + n;$$
  $T(1) = 1$ 

#### Solve the recurrence:





### Analysis (Tree Method)

$$n + \left(\frac{4}{2}\right)n + \left(\frac{4}{2}\right)^2n + \dots + + \left(\frac{4}{2}\right)^{(\log n)-1}n + 4^{\log n}T(1)$$

$$= n(1 + 2 + 2^2 + 2^3 + \dots + 2^{(\log n) - 1}) + 4^{\log n} T(1)$$

$$= n \left( \frac{2^{\log n} - 1}{2 - 1} \right) + 4^{\log n} T(1) = n(n - 1) + 2^{2 \log n} T(1)$$

$$= n(n-1) + n^2 = \Theta(n^2)$$

- The divide-and-conquer algorithm is as bad as the primary school method
- Essentially, the algorithm still multiplies every bit of A with every bit of B.
- Compared with merge sort, the key difference is that one problem generates 4 subproblems of size n/2.

### Karatsuba Multiplication

#### New Observation:

- Let X,a, b, c, d be integers
- Simple algebra said

$$(aX + b) (cX + d) = acX^2 + (ad + bc)X + bd$$

- This used 4 multiplications to find the three coefficients ac, (ad +bc), bd
- We will now see how to find these 3 coefficients using only 3 multiplications
- Calculate ac, bd, and A= (a+b) (c+d)
- Notice that ad + bc = A ac -bd
- So, we can calculate the three coefficients using only 3 multiplications
   (and one more addition and two subtractions)

# Karatsuba Multiplication

- Let  $a = a_1 2^{n/2} + a_0$ , and  $b = b_1 2^{n/2} + b_0$  where  $a_1, a_0, b_1, b_0$  are all (n/2)-bit integers.
- We already saw

$$ab = a_1 b_1 2^n + (a_1b_0 + a_0b_1)2^{n/2} + a_0 b_0$$

Use the trick from previous page:

$$a_1b_0 + a_0b_1 = (a_1 + a_0)(b_1 + b_0) - a_1b_1 - a_0b_0$$

Calculating ab now only requires performing 3 multiplication subproblems of size n/2!

# Karatsuba's multiplication algorithm

```
Multiply (A, B):
n \leftarrow \text{size of } A
if n = 1 then return A[1] \cdot B[1]
mid \leftarrow \lfloor n/2 \rfloor
U \leftarrow \text{Multiply}(A[mid + 1..n], B[mid + 1..n])
Z \leftarrow Multiply(A[1..mid], B[1..mid])
A' \leftarrow A[mid + 1..n] \oplus A[1..mid]
B' \leftarrow B[mid + 1..n] \oplus B[1..mid]
Y \leftarrow \text{Multiply}(A', B')
M[1..2n] \leftarrow 0
M[1..n] \leftarrow M[1..n] \oplus Z
M[mid + 1..] \leftarrow M[mid + 1..] \oplus Y \ominus U \ominus Z
M[2mid + 1..] \leftarrow M[2mid + 1..] \oplus U
return M
```

 $\oplus$   $\ominus$ : denotes the integer addition/subtraction algorithm

### Analysis (Expansion Method)

#### Recurrence.

For, 
$$n > 1$$
,  $T(n) = 3T(n/2) + n$ .  $T(1) = 1$ 

$$T(n) = 3 T\left(\frac{n}{2}\right) + n$$

$$= 3\left(3T\left(\frac{n}{4}\right) + \frac{n}{2}\right) + n$$

$$=3^{2}\left(3T\left(\frac{n}{2^{3}}\right)+\frac{n}{2^{2}}\right)+\frac{3}{2}n+n$$

$$= 3^{3} \left(3T\left(\frac{n}{2^{4}}\right) + \frac{n}{2^{3}}\right) + \frac{3^{2}}{2^{2}}n + \frac{3}{2}n + n$$

$$=3^{2} T\left(\frac{n}{2^{2}}\right) + \frac{3}{2}n + n$$

$$=3^{3} T\left(\frac{n}{2^{3}}\right) + \frac{3^{2}}{2^{2}}n + \frac{3}{2}n + n$$

$$= 3^4 T\left(\frac{n}{2^4}\right) + \left(\frac{3^3}{2^3} + \frac{3^2}{2^2} + \frac{3}{2} + 1\right)n$$

····

$$= 3^{i} T\left(\frac{n}{2^{i}}\right) + \left(\frac{3^{i-1}}{2^{i-1}} + \frac{3^{i-2}}{2^{i-2}} + \dots + \frac{3}{2} + 1\right)n$$

# Analysis (Expansion Method)

$$T(n) = 3^{i} T\left(\frac{n}{2^{i}}\right) + \left(\frac{3^{i-1}}{2^{i-1}} + \frac{3^{i-2}}{2^{i-2}} + \dots + \frac{3}{2} + 1\right)n$$

....

$$= 3^{h} T\left(\frac{n}{2^{h}}\right) + \left(\left(\frac{3}{2}\right)^{h-1} + \left(\frac{3}{2}\right)^{h-2} + \dots + \left(\frac{3}{2}\right) + 1\right) n = 3^{h} T\left(\frac{n}{n}\right) + \left(\frac{\left(\frac{3}{2}\right)^{h} - 1}{\frac{3}{2} - 1}\right) n$$

$$= 3^{h} T(1) + 2\left(\left(\frac{3}{2}\right)^{h} - 1\right)n$$

$$= 3^{h} + 2\left(\frac{3^{h}}{2^{h}} - 1\right)n$$

$$= 3^h + 2\left(\frac{3^h}{n} - 1\right)n = 3 \cdot 3^{\log_2 n} - 2n$$

$$= 3 \cdot n^{\log_2 3} - 2n$$

$$= \Theta(n^{1.585\dots})$$

$$h = \log_2 n$$

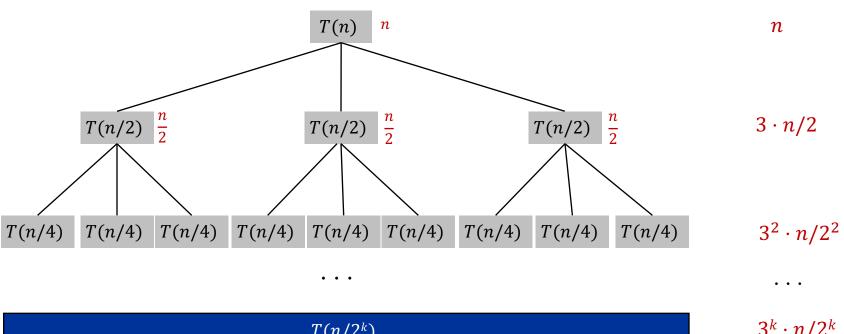
$$2^h = n$$

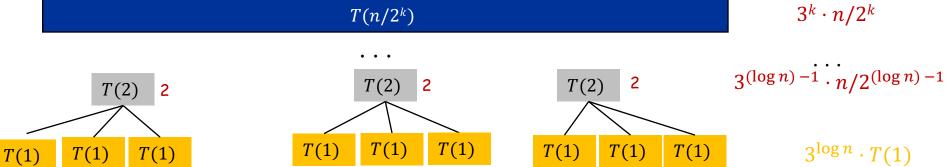
# **Analysis**

#### Recurrence:

$$T(n) = 3T(n/2) + n$$

#### Solve the recurrence:





# Analysis (continued)

$$T(n) = n + \left(\frac{3}{2}\right)^{1} n + \left(\frac{3}{2}\right)^{2} n + \dots + \left(\frac{3}{2}\right)^{(\log n) - 1} n + 3^{\log n} T(1)$$

$$= n \left(1 + \frac{3}{2} + \left(\frac{3}{2}\right)^2 + \left(\frac{3}{2}\right)^3 + \dots + \left(\frac{3}{2}\right)^{(\log n) - 1}\right) + 3^{\log n} T(1)$$

$$= n \left( \frac{\left(\frac{3}{2}\right)^{\log n} - 1}{\frac{3}{2} - 1} \right) + 3^{\log n} T(1)$$

$$= n \Theta\left(\frac{3^{\log n}}{2^{\log n}}\right) + \Theta\left(3^{\log n}\right)$$

$$= n \Theta\left(\frac{n^{\log 3}}{n}\right) + \Theta(n^{\log 3}) = \Theta(n^{\log 3})$$

$$=\Theta(n^{1.585...})$$

Recall

$$3^{\log n} = (2^{\log 3})^{\log n} = (2^{\log n})^{\log 3} = n^{\log 3}$$

# Analysis (continued)

### Recurrence For First D&C Algorithm

$$T(n) = 4T(n/2) + n;$$
  $T(1) = 1$ 

Solution:  $T(n) = \Theta(n^2)$ 

### Recurrence For Karatsuba Multiplication

$$T(n) = 3T(n/2) + n;$$
  $T(1) = 1$ 

Solution:  $T(n) = \Theta(n^{1.585...})$ 

### Analysis (continued)

### Karatsuba Multiplication:

- Dividing each integer into 2 parts, and solve 3 subproblems
  - T(n) = 3T(n/2) + n,  $T(n) = \Theta(n^{\log_2 3}) = \Theta(n^{1.585...})$

### Progressive improvements:

- Dividing each integer into 3 parts, and solve 5 subproblems
  - T(n) = 5T(n/3) + n,  $T(n) = \Theta(n^{\log_3 5}) = \Theta(n^{1.465})$
- Dividing each integer into 4 parts, and solve 7 subproblems

- 
$$T(n) = 7T(n/4) + n$$
,  $T(n) = \Theta(n^{\log_4 7}) = \Theta(n^{1.404})$ 

- **...**
- An  $\Theta(n \log n \log \log n)$  algorithm (based on FFT)
- An  $\Theta(n \log n \log \log \log n)$  algorithm
- The fastest algorithm runs in time  $O(n \log n \, 2^{\Theta(\log^* n)})$ 
  - $log^*n$  is a VERY slow growing function
- The conjecture is that the problem can be solved in  $\Theta(n \log n)$  time. This conjecture is still open.

### Integer Multiplication in Practice

#### Work on the word level

- Example (using 16-bit words):
  - Decimal: 1316103040073424382
  - Hexadecimal: 1243 BCBD EF63 5DFE
  - Stored using an array of 4 words

### In practice:

- Long multiplication: Best for < 20 words</li>
- Karatsuba's algorithm: Best for 20 ~ 2000 words
- FFT based algorithm: Best for > 2000 words

# The Master Theorem (proof coming soon)

Theorem: Let  $a \ge 1, b > 1, c \ge 0$  be constants. The recurrence  $T(n) = aT(n/b) + n^c$  have the following solutions.

- Case 1:  $c < \log_b a$ :  $T(n) = \Theta(n^{\log_b a})$ .
- Case 2:  $c = \log_b a$ :  $T(n) = \Theta(n^c \log n)$ .
- Case 3:  $c > \log_b a$ :  $T(n) = \Theta(n^c)$ .

Examples: We have already seen Cases 1 & 2. Case 3 will arise later

• Case 1: 
$$T(n) = 3T\left(\frac{n}{2}\right) + n$$
  $\Rightarrow$   $T(n) = \Theta(n^{\log_2 3})$ 

• Case 2: 
$$T(n) = 2T\left(\frac{n}{2}\right) + n$$
  $\Rightarrow$   $T(n) = \Theta(n \log n)$ 

• Case 3: 
$$T(n) = 2T\left(\frac{n}{3}\right) + n$$
  $\Rightarrow$   $T(n) = \Theta(n)$ 

### Matrix Multiplication

Matrix multiplication. Given two n-by-n matrices A and B, compute C = AB.

$$C_{ij} = \sum_{k=1}^{n} a_{ik} b_{kj} \qquad \begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \times \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{n1} & b_{n2} & \cdots & b_{nn} \end{bmatrix}$$

Brute force.  $\Theta(n^3)$  time.

Fundamental question. Can we improve upon brute force?

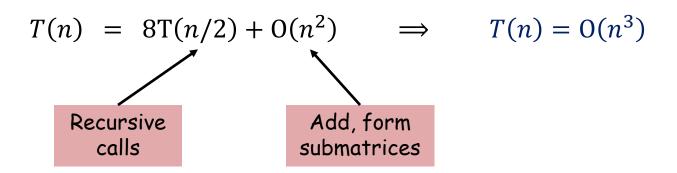
### Matrix Multiplication: First Attempt

### Divide-and-conquer.

- Divide: partition A and B into  $\frac{1}{2}n$ -by- $\frac{1}{2}n$  blocks.
- Conquer: multiply 8  $\frac{1}{2}n$ -by- $\frac{1}{2}n$  submatrices recursively.
- Combine: add appropriate products using 4 matrix additions.

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$\begin{bmatrix} C_{11} & = (A_{11} \times B_{11}) + (A_{12} \times B_{21}) \\ C_{12} & = (A_{11} \times B_{12}) + (A_{12} \times B_{22}) \\ C_{21} & = (A_{21} \times B_{11}) + (A_{22} \times B_{21}) \\ C_{22} & = (A_{21} \times B_{12}) + (A_{22} \times B_{22}) \end{bmatrix}$$



### Strassen's Matrix Multiplication Algorithm

Key idea. multiply 2-by-2 block matrices with only 7 multiplications.

$$\begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} \qquad P_1 = A_{11} \times (B_{12} - B_{22})$$

$$P_2 = (A_{11} + A_{12}) \times B_{22}$$

$$P_3 = (A_{21} + A_{22}) \times B_{11}$$

$$C_{11} = P_5 + P_4 - P_2 + P_6$$

$$C_{12} = P_1 + P_2$$

$$C_{21} = P_3 + P_4$$

$$C_{22} = P_5 + P_1 - P_3 - P_7$$

$$P_4 = A_{22} \times (B_{21} - B_{11})$$

$$P_5 = (A_{11} + A_{22}) \times (B_{11} + B_{22})$$

$$P_6 = (A_{12} - A_{22}) \times (B_{21} + B_{22})$$

$$P_7 = (A_{11} - A_{21}) \times (B_{11} + B_{12})$$

- 7 multiplications of  $\frac{1}{2}n$ -by- $\frac{1}{2}n$  submatrices.
- $\bullet$   $\Theta(n^2)$  additions and subtractions.

• 
$$T(n) = 7T(n/2) + n^2 \implies T(n) = \Theta(n^{\log_2 7}) = \Theta(n^{2.807})$$

In practice: Used to multiply large matrices (e.g., n > 100)

### Fast Matrix Multiplication in Theory

- Q. Multiply two 2-by-2 matrices with only 7 multiplications?
- **A.** Yes!  $\Theta(n^{2.807})$  [Strassen, 1969]
- Q. Multiply two 2-by-2 matrices with only 6 multiplications?
- A. Impossible.
- Q. Two 3-by-3 matrices with only 21 multiplications?
- A. Also impossible.
- Q. Two 70-by-70 matrices with only 143,640 multiplications?
- **A.** Yes!  $\Theta(n^{2.795})$

### The competition continues...

- $\Theta(n^{2.376})$  [Coppersmith-Winograd, 1990.]
- $\Theta(n^{2.374})$  [Stothers, 2010.]
- $\Theta(n^{2.3728642})$  [Williams, 2011.]
- $\Theta(n^{2.3728639})$  [Le Gall, 2014.]
- Conjecture: close to  $\Theta(n^2)$