

Matrix Multiplication

Advanced Programming and Algorithmic Design

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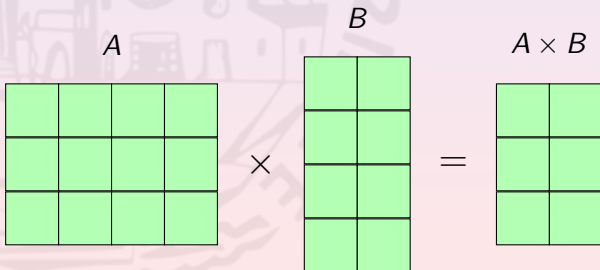
a.a. 2018/2019

Matrix Multiplication

Definition (Row-Column Multiplication)

Let A be a $n \times m$ matrix and let B be a $m \times l$ matrix. $A \times B$ is a $n \times l$ matrix s.t.

$$(A \times B)[i, j] = \sum_k A[i, k] * B[k, j]$$

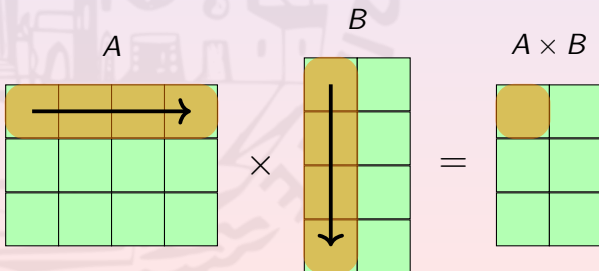


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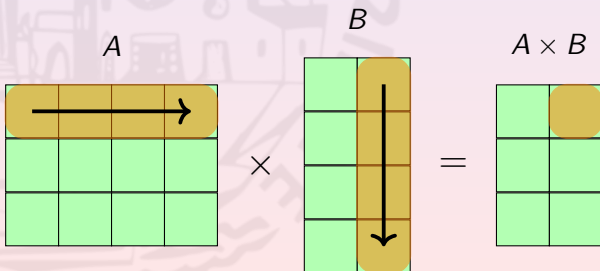


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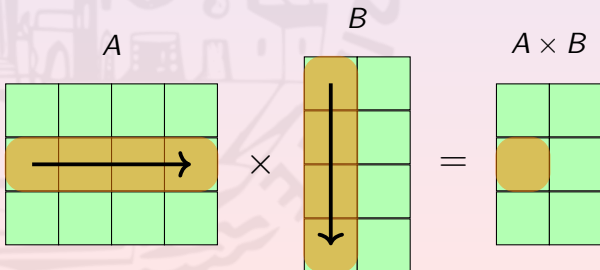


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Problem Definition

Input: Two $n \times n$ matrices A and B

Output: The $n \times n$ matrix $A \times B$

E.g.,

A				B				$A \times B$		
1	-1	2		4	-2	2		0	-4	2
2	0	3	,	2	0	0	\Rightarrow	5	5	4
0	-1	2		-1	3	0		-4	6	0

Square matrices solution can easily be generalized

Naïve Solution: Code

```
def naive_mult(C, A, B):  
    for i ← 1..rows(A):  
        for j ← 1..cols(B):  
            a ← 0  
            for k ← 1..cols(A):  
                a ← a + A[i,k] * B[k,j]  
            endfor  
            C[i,j] ← a  
        endfor  
    endfor  
  
    return C  
enddef
```

Naïve Solution: Complexity

The naïve solution mimes row-column definition

3 nested loops with indexes in $[1, n]$

The inner-block takes time $O(1)$

The overall execution takes time $\Theta(n^3)$

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The overall execution takes time $\Theta(n^3)$

Can we find a better algorithm?

Divide-and-Conquer Strategy

What about splitting A and B in blocks?

$$\begin{array}{|c|c|} \hline A & \\ \hline A_{11} & A_{12} \\ \hline A_{21} & A_{22} \\ \hline \end{array} \times \begin{array}{|c|c|} \hline B & \\ \hline B_{11} & B_{12} \\ \hline B_{21} & B_{22} \\ \hline \end{array} = \begin{array}{|c|c|} \hline A \times B & \\ \hline C_{11} & C_{12} \\ \hline C_{21} & C_{22} \\ \hline \end{array}$$

where

$$C_{ij} = (A_{i1} \times B_{1j}) + (A_{i2} \times B_{2j})$$

Divide-and-Conquer Strategy (Cont'd)

+ is the elements-wise matrix sum (time complexity $\Theta(n^2)$)

× is the usual row-column multiplication

A_{ik} and B_{kj} are $\frac{n}{2} \times \frac{n}{2}$ matrices

Divide-and-Conquer Strategy (Cont'd)

$+$ is the elements-wise matrix sum (time complexity $\Theta(n^2)$)

\times is the usual row-column multiplication

A_{ik} and B_{kj} are $\frac{n}{2} \times \frac{n}{2}$ matrices

We can define a recursive algorithm:

- if $\text{rows}(A) < 2$, return `naive_mult(C, A, B)`
- for $i, j, k \in [12]$ recursively compute $D_{ijk} = A_{ik} \times B_{kj}$
- for $i, j \in [12]$ compute $C_{ij} = D_{ij1} + D_{ij2}$
- return C

Divide-and-Conquer Strategy: Complexity

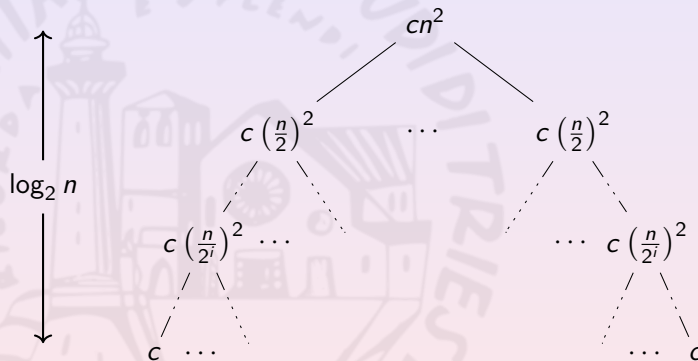
The recursive algorithm requires:

- 8 multiplications between $\frac{n}{2} \times \frac{n}{2}$ matrices
- 4 sums between $\frac{n}{2} \times \frac{n}{2}$ matrices

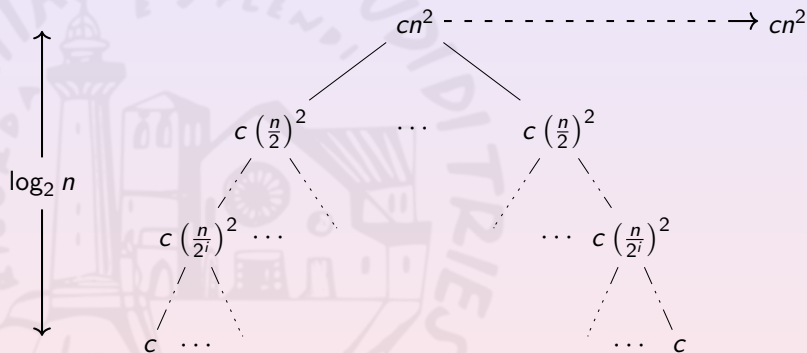
If T_M is the complexity of the algorithm

$$\begin{aligned}T_M(n) &= 8 * T_M(n/2) + 4 * \Theta(n^2) \\ &= 8 * T_M(n/2) + \Theta(n^2)\end{aligned}$$

Divide-and-Conquer Strategy: Complexity (Recursion Tree)

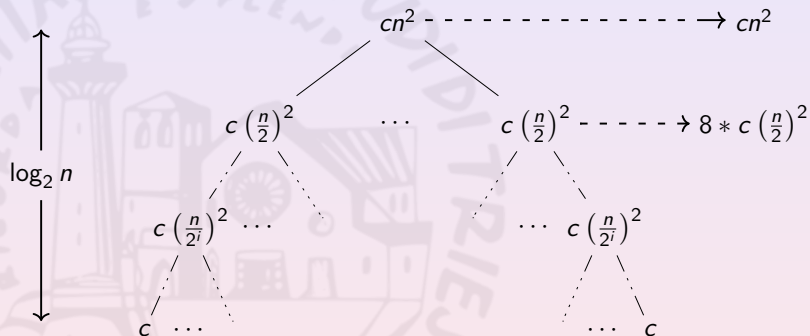


Divide-and-Conquer Strategy: Complexity (Recursion Tree)



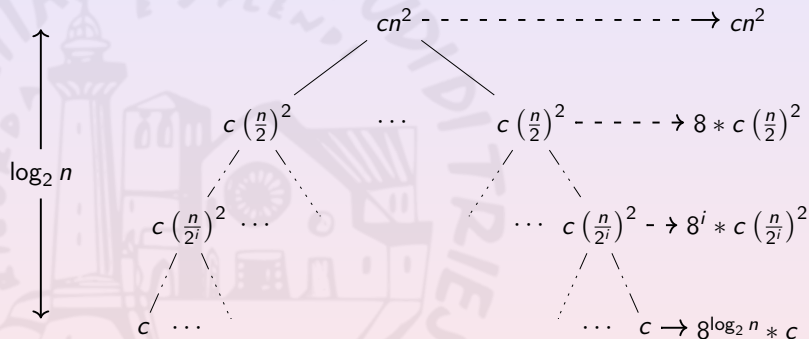
$$T_M(n) = cn^2 \left(1 + \right)$$

Divide-and-Conquer Strategy: Complexity (Recursion Tree)



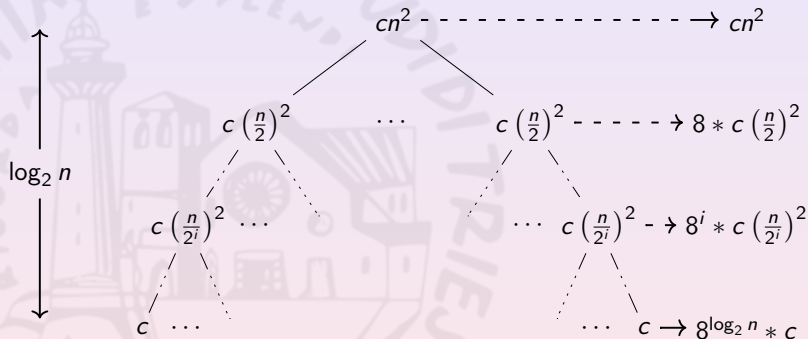
$$T_M(n) = cn^2 \left(1 + 2 + \dots + 2^{\log_2 n} \right)$$

Divide-and-Conquer Strategy: Complexity (Recursion Tree)



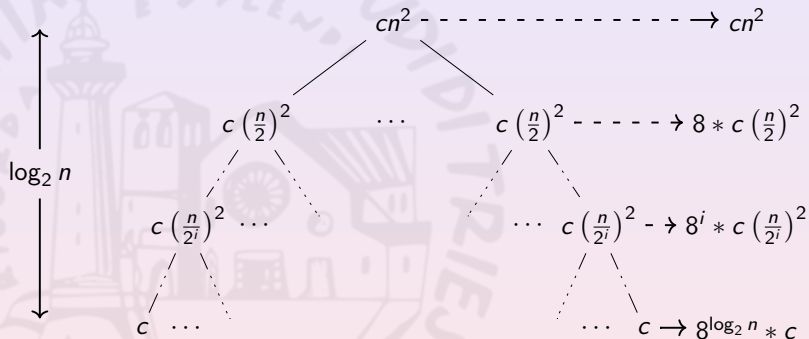
$$T_M(n) = cn^2 \left(1 + 2 + \dots + 2^i + \dots + 2^{\log_2 n} \right)$$

Divide-and-Conquer Strategy: Complexity (Recursion Tree)



$$\begin{aligned}
 T_M(n) &= cn^2 \left(1 + 2 + \dots + 2^i + \dots + 2^{\log_2 n} \right) \\
 &= cn^2 \left(2^{\log_2 n + 1} - 1 \right) = cn^2 (2n - 1)
 \end{aligned}$$

Divide-and-Conquer Strategy: Complexity (Recursion Tree)



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 T_M(n) &= cn^2 \left(1 + 2 + \dots + 2^i + \dots + 2^{\log_2 n} \right) \\
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 \end{aligned}$$

Some Further Thoughts

The divide-and-conquer approach **had too many recursive calls**

Can it be “rephrased” to decrease them?

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Can it be “rephrased” to decrease them?

Yes, it can!!!

Strassen's Algorithm

Sums ($\Theta(n^2)$)

$$S_1 = B_{12} - B_{22}$$

$$S_2 = A_{11} + A_{12}$$

$$S_3 = A_{21} + A_{22}$$

$$S_4 = B_{21} - B_{11}$$

$$S_5 = A_{11} + A_{22}$$

$$S_6 = B_{11} + B_{22}$$

$$S_7 = A_{12} - A_{22}$$

$$S_8 = B_{21} + B_{22}$$

$$S_9 = A_{11} - A_{21}$$

$$S_{10} = B_{11} + B_{12}$$

\Rightarrow

**Recursion
Calls**

$$P_1 = A_{11} \times S_1$$

$$P_2 = S_2 \times B_{22}$$

$$P_3 = S_3 \times B_{11}$$

$$P_4 = A_{22} \times S_4$$

$$P_5 = S_5 \times S_6$$

$$P_6 = S_7 \times S_8$$

$$P_7 = S_9 \times S_{10}$$

Strassen's Algorithm

Sums ($\Theta(n^2)$)

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

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

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

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

Strassen's Algorithm

Sums ($\Theta(n^2)$)

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

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

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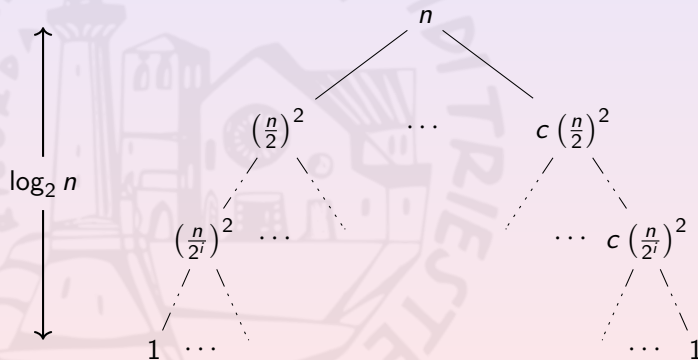
$$C_{22} = P_5 + P_1 - P_3 - P_7$$

The complexity equation is:

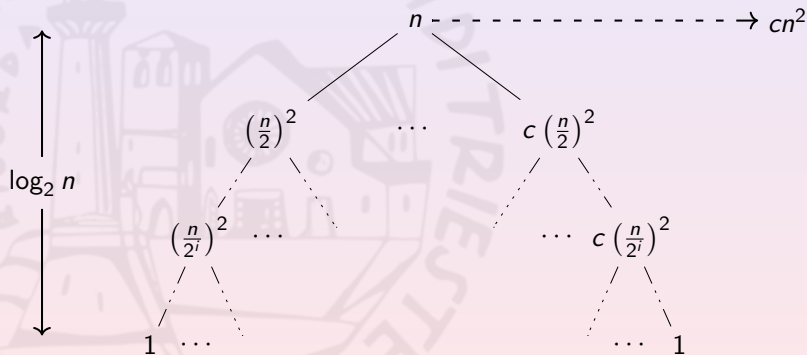
$$T_M(n) = 7 * T_M(n/2) + \Theta(n^2)$$

because the S 's and the C 's are computed by sums

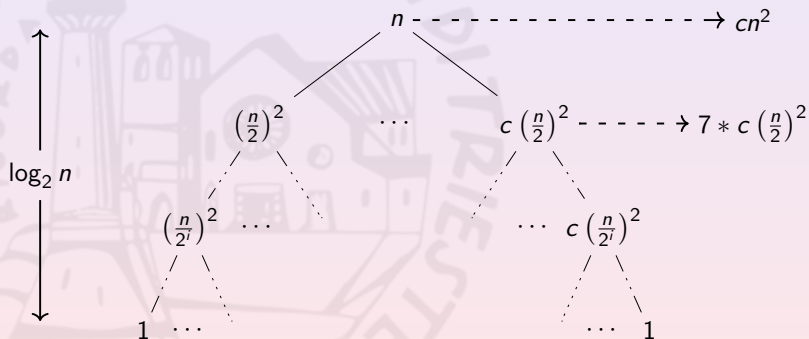
Strassen's Algorithm: Complexity (Recursion Tree)



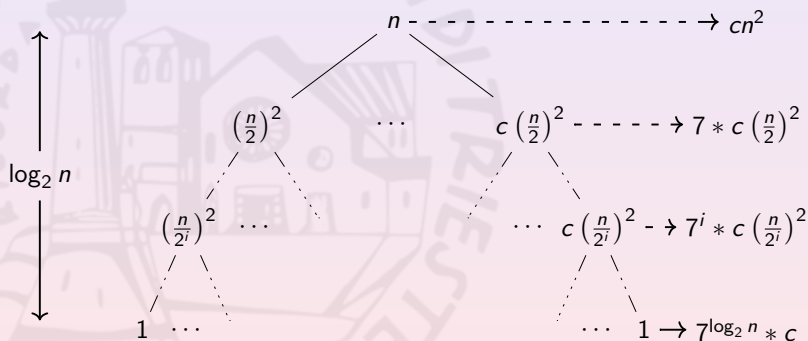
Strassen's Algorithm: Complexity (Recursion Tree)



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Strassen's Algorithm: Complexity (Recursion Tree)



Strassen's Algorithm: Complexity (Cont'd)

$$T_M(n) = cn^2 \left(1 + \frac{7}{4} + \dots + \left(\frac{7}{4}\right)^i + \dots + \left(\frac{7}{4}\right)^{\log_2 n} \right)$$

$$= c'n^2 \left(\left(\frac{7}{4}\right)^{\log_2 n + 1} - 1 \right)$$

$$c' = \frac{4}{3}c$$

$$= c'n^2 \left(\frac{7}{4}\right)^{\log_2 n + 1} - c'n^2$$

$$= c''4^{\log_2 n} \left(\frac{7}{4}\right)^{\log_2 n} - c'n^2$$

$$c'' = \frac{7}{4}c'$$

$$= c''7^{\log_2 n} - c'n^2$$

$$= c''7^{\frac{\log_7 n}{\log_7 2}} - c'n^2 = c''n^{\log_2 7} - c'n^2$$

Strassen's Algorithm: Complexity (Cont'd)

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