

Matrix Multiplication

Advanced Programming and Algorithmic Design

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The background of the slide features a large, faint watermark of the University of Trieste logo. The logo is circular and contains the text "UNIVERSITA' DEGLI STUDI DI TRIESTE" around the perimeter and "E SPLENDI" in the center. In the middle of the logo is a detailed illustration of a building with a dome and a tower.

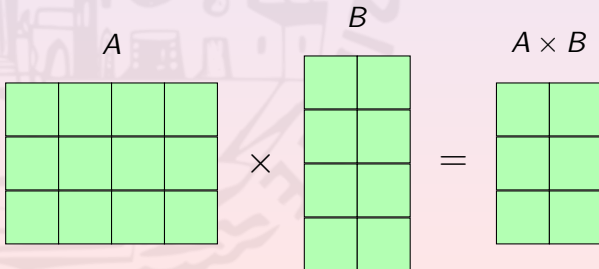
Problem Definition

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.

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

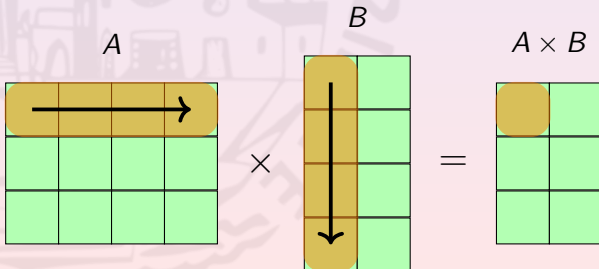


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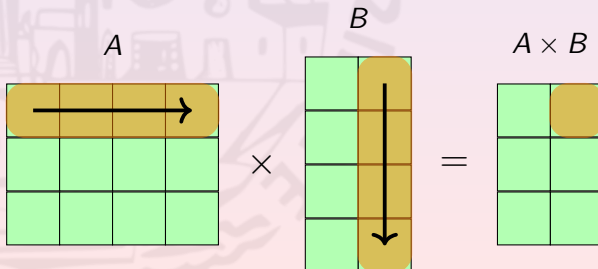


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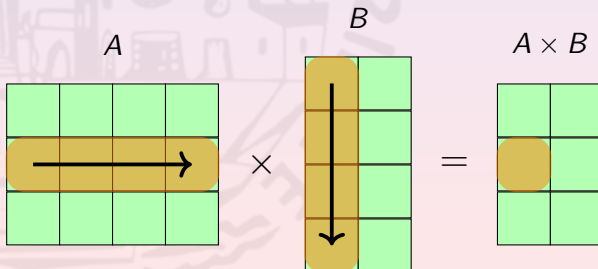


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Problem Definition

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

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

E.g.,

$$\begin{array}{c} A \\ \begin{array}{|c|c|c|} \hline 1 & -1 & 2 \\ \hline 2 & 0 & 3 \\ \hline 0 & -1 & 2 \\ \hline \end{array} \end{array}, \begin{array}{c} B \\ \begin{array}{|c|c|c|} \hline 4 & -2 & 2 \\ \hline 2 & 0 & 0 \\ \hline -1 & 3 & 0 \\ \hline \end{array} \end{array} \Rightarrow \begin{array}{c} A \times B \\ \begin{array}{|c|c|c|} \hline 0 & -4 & 2 \\ \hline 5 & 5 & 4 \\ \hline -4 & 6 & 0 \\ \hline \end{array} \end{array}$$

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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Can we find a better algorithm?

Divide-and-Conquer Strategy

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What about splitting A and B in blocks?

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

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 $i, j, k \in [1, 2]$ recursively compute $D_{ijk} = A_{ik} \times B_{kj}$
- for $i, j \in [1, 2]$ 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

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

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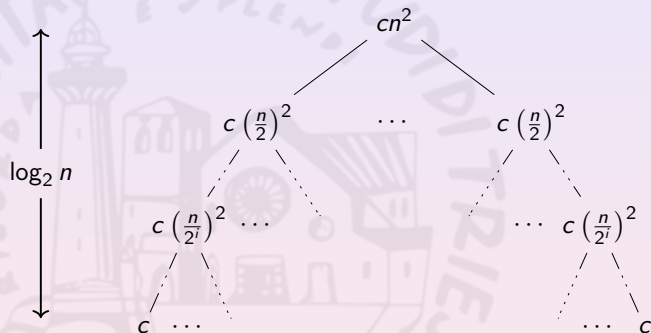
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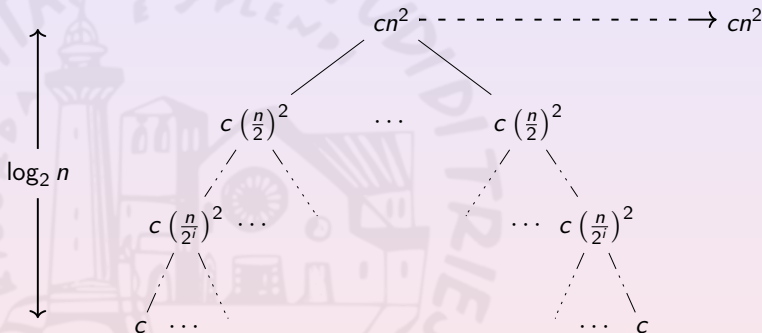
This is a **recursive equation**. How to solve it?

Let cn^2 be the cost of the four $n \times n$ -sums.

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

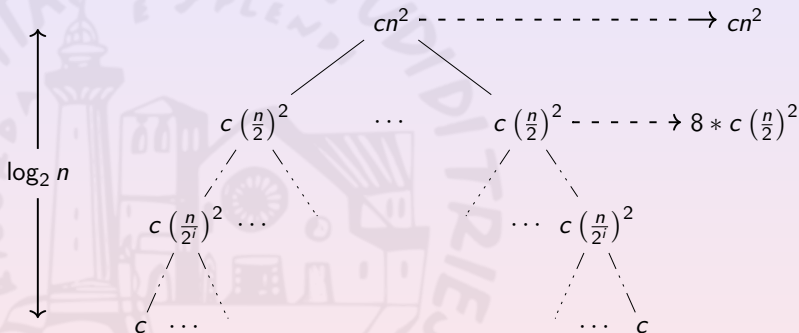


Divide-and-Conquer Strategy: Complexity (Recursion Tree)



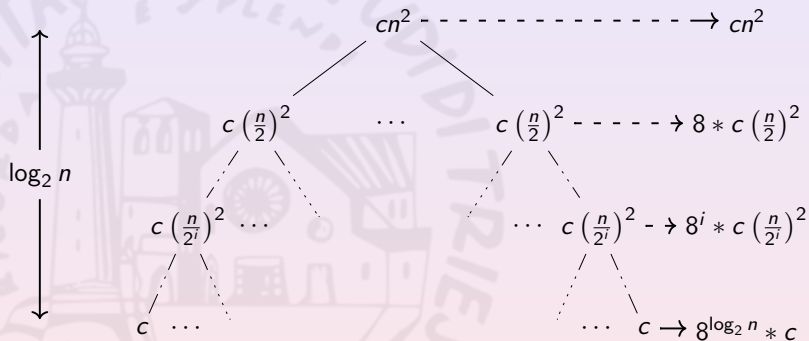
$$T_M(n) = cn^2 \left(1 + \right)$$

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



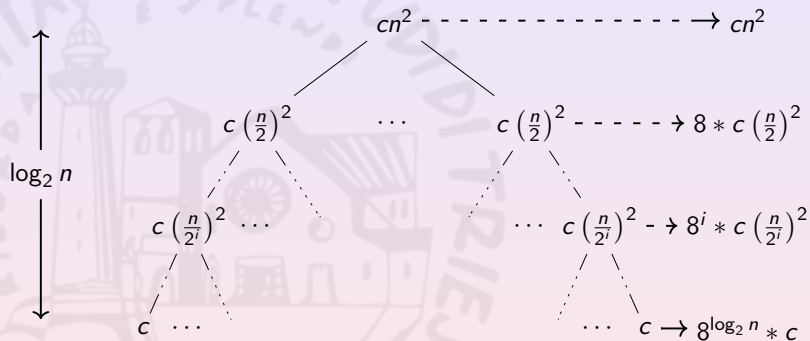
$$T_M(n) = cn^2 \left(1 + 2 + \dots + 2^{\log_2 n - 1} \right)$$

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)



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

Diagram illustrating the recursion tree for Merge Sort, showing the recurrence relation $T_M(n) = 2T_M(n/2) + cn$.

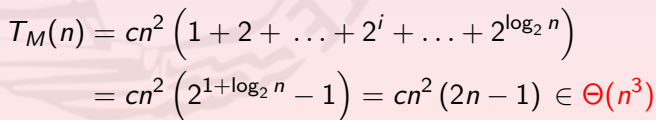
The tree structure shows the recurrence relation at each level:

- Root node: cn^2
- Level 1: $cn^2/2$ (two nodes)
- Level 2: $cn^2/4$ (four nodes)
- Level i : $cn^2/2^i$ (2^i nodes)
- Level $\log_2 n$: c (n nodes)

The total work done at each level is cn^2 . The total work done is the sum of work done at each level, which is $cn^2 \log_2 n$.

The recurrence relation is given by:

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

$$= cn^2 \left(2^{1+\log_2 n} - 1 \right) = cn^2 (2n - 1) \in \Theta(n^3)$$


Some Further Thoughts

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

Can it be “rephrased” to decrease them?

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

Can it be “rephrased” to decrease them?

Yes, it can!!!

Strassen's Algorithm

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

**Recursive
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 ($\Theta(n^2)$)

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

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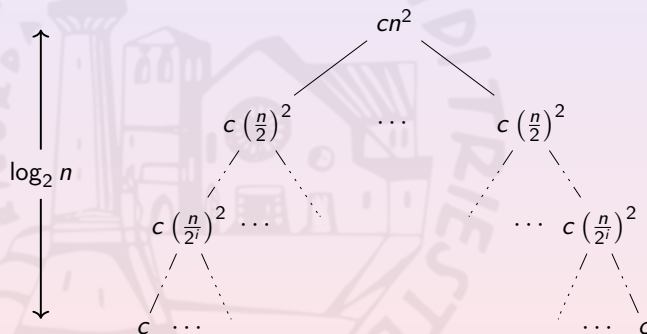
$$C_{22} = P_5 + P_1 - P_3 - P_7$$

The complexity equation is:

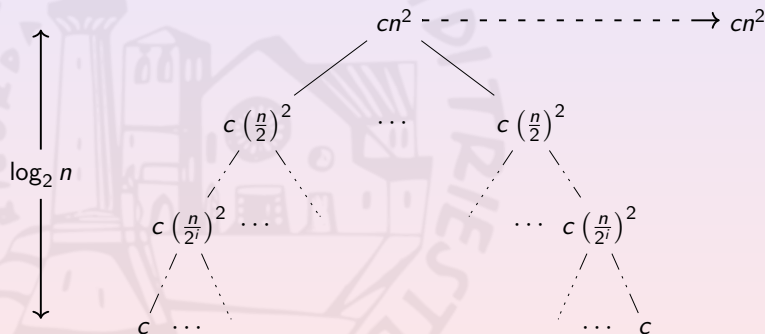
$$T_M(n) = 7 * T_M(n/2) + \Theta(n^2)$$

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

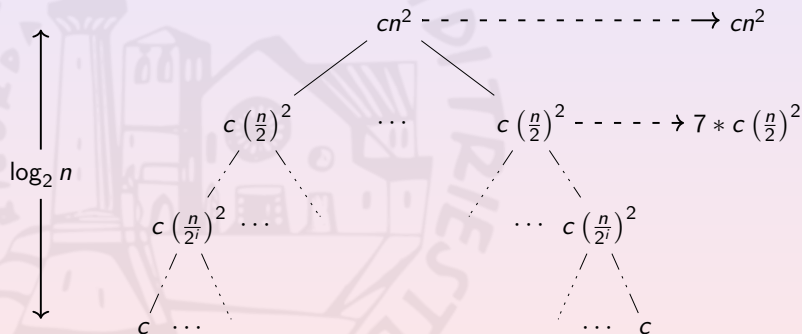
Strassen's Algorithm: Complexity (Recursion Tree)



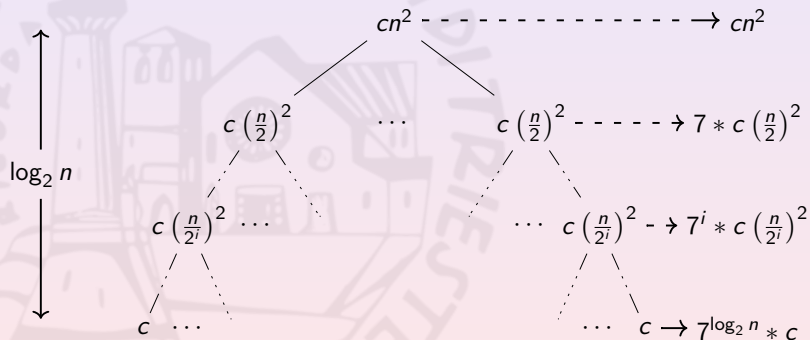
Strassen's Algorithm: Complexity (Recursion Tree)



Strassen's Algorithm: Complexity (Recursion Tree)



Strassen's Algorithm: Complexity (Recursion Tree)



Strassen's Algorithm: Complexity (Cont'd)

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

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

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

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

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

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

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

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

Strassen's Algorithm: Complexity (Cont'd)

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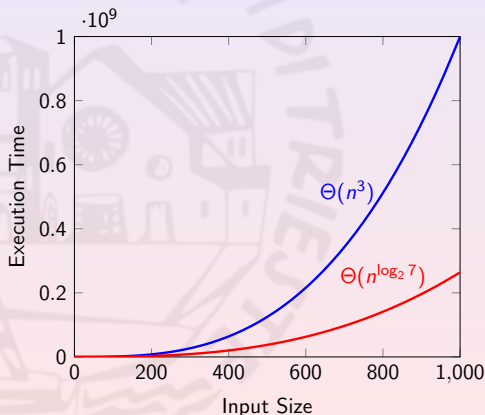
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$$= c''7^{\frac{\log_7 n}{\log_7 2}} - c'n^2 = c''n^{\log_2 7} - c'n^2 \in \Theta \left(n^{\log_2 7} \right)$$

Final Considerations

Strassen's algorithm ($\Theta(n^{\log_2 7})$) improves asymptotic complexity of naïve algorithm ($\Theta(n^3) = \Theta(n^{\log_2 8})$)



Final Considerations

However, it is not **in-place** i.e., it requires a non-constant amount of additional memory

A careful handling of the auxiliary memory may make the difference in implementation.