# Strassen's Algorithm for Matrix Multiplication

#### ESO207

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#### Introduction

 We will consider the problem of multiplying two n × n matrices. For example

$$\begin{pmatrix} 1 & 4 & 6 \\ 2 & 7 & 5 \\ -1 & 3 & 4 \end{pmatrix} \times \begin{pmatrix} 3 & -1 & 2 \\ 2 & 5 & 3 \\ -2 & -1 & 0 \end{pmatrix} = \begin{pmatrix} -1 & 13 & 14 \\ 10 & 28 & 25 \\ -5 & 13 & 7 \end{pmatrix}$$

• Consider multiplication of two  $n \times n$  matrices  $A \times B$ . If  $A = (a_{ij})$ , and  $B = (b_{ij})$ , where,  $1 \le i, j \le n$ , then, the product  $C = A \times B$  is the  $n \times n$  matrix  $C = (c_{ij})$ , with

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

### **Classical Matrix Multiplication**

```
SQUARE-MATRIX-MULTIPLY (A, B, n) {
// A, B are n \times n matrices
1. Let C be a new n \times n matrix
2. for i = 1 to n {
3. for j = 1 to n \{
        c_{ii}=0
5. for k = 1 to n \{
6.
              c_{ii} = c_{ij} + a_{ik} \cdot b_{ki}
7.
8.
9. }
10. return C
```

### Square Matrix Multiplication: Classical

- The c<sub>ij</sub> entry is the dot product of the ith row of A and the jth column of B.
- Each  $c_{ij}$  entry is computed as a dot product involving n multiplications and n additions. Time taken is O(n).
- There are  $n^2$  entries  $(c_{ij})$  for  $1 \le i, j \le n$  to compute.
- Total time required is  $n^2 \times O(n) = n^3$ .
- Alternatively: this is a triple loop with indices running from 1 to n each, giving  $O(n^3)$  total time.

### Summary

- It might appear that square matrix multiplication should take time  $\Omega(n^3)$ .
- We will see Strassen's algorithm that solves square matrix multiplication in time  $O(n^{2.81...}) = O(n^{\log_2 7})$ .
- We will first look at a simple divide-and-conquer algorithm and then understand Strassen's algorithm.

## Simple Divide and Conquer Algorithm

- Goal: To compute  $C = A \times B$ .
- For simplicity, assume that n is an exact power of 2.
- View A, B, C as 2 × 2 block matrices where, each block is of size n/2 × n/2.

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}, \quad B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}, C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

• Rewrite equation  $C = A \cdot B$  as

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

### Block matrix multiplication

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

gives four equations.

$$C_{11} = A_{11} \times B_{11} + A_{12} \times B_{21}$$
 $C_{12} = A_{11} \times B_{12} + A_{12} \times B_{22}$ 
 $C_{21} = A_{21} \times B_{11} + A_{22} \times B_{21}$ 
 $C_{22} = A_{21} \times B_{12} + A_{22} \times B_{22}$ 

• In the above  $\times$  denotes multiplication of  $n/2 \times n/2$  matrices and + denotes sum of  $n/2 \times n/2$  matrices.



## Translating into Recursive matrix multiplication

```
SQ-MAT-MULT-RECURSIVE(A, B, n)
   Let C be a new n \times n matrix
   if n==1
3
       c_{11} = a_{11} \cdot b_{11}
   else partition A, B, C into block sub-matrices as done earlier
5
       C_{11} = \text{SQ-MAT-MULT-RECURSIVE}(A_{11}, B_{11})
           + SQ-MAT-MULT-RECURSIVE(A_{12}, B_{21})
       C_{12} = \text{SQ-MAT-MULT-RECURSIVE}(A_{11}, B_{12})
6
           + SQ-MAT-MULT-RECURSIVE(A_{12}, B_{22})
7
       C_{21} = \text{SQ-MAT-MULT-RECURSIVE}(A_{21}, B_{11})
           + SQ-MAT-MULT-RECURSIVE(A_{22}, B_{21})
8
       C_{22} = \text{SQ-MAT-MULT-RECURSIVE}(A_{21}, B_{12})
           + SQ-MAT-MULT-RECURSIVE(A_{22}, B_{22})
```

#### **Index Calculations**

- To partition A, B, C into the  $n/2 \times n/2$  blocks, there is no need to copy the sub-matrices.
- This overhead is not needed, we represent a  $2^k \times 2^k$  sub-matrix A starting at row position i and column j as  $(A, i, j, 2^k)$ .
- Represent submatrices by index calculations.
- That is, we represent sub-matrices slightly more generally than we represent the original matrix.

### **Analysis**

- Let T(n) be the time to multiply two n × n matrices using the block-multiplication algorithm.
- Base Case:  $T(1) = \Theta(1)$ .
- Consider one of the four block multiplications:

$$C_{11} = A_{11} \times B_{11} + A_{12} \times B_{21}$$

where,  $A_{11}, \dots A_{22}$  and  $B_{11}, \dots, B_{22}$  are each  $n/2 \times n/2$  matrices.

- The addition operation requires Θ(n²) steps, since it adds two n/2 × n/2 matrices.
- The multiplication operation is done recursively, and this requires T(n/2) + T(n/2) = 2T(n/2) time.
- There are three other block equalities for  $C_{12}$ ,  $C_{21}$  and  $C_{22}$ . Each of them requires  $2T(n/2) + \Theta(n^2)$  time.

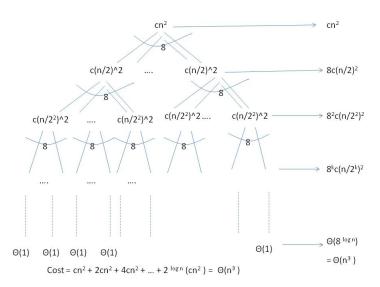


### Analysis: Basic recursive algorithm

 The recurrence equation is (assuming n is an exact power of 2).

$$T(n) = 8T(n/2) + \Theta(n^2)$$
$$T(1) = \Theta(1)$$

#### Recursion tree



#### Recurrence solution

• Thus solution to the recurrence equation

$$T(n) = 8T(n/2) + \Theta(n^2)$$
  
$$T(1) = \Theta(1)$$

is 
$$T(n) = \Theta(n^3)$$
.

# Master Theorem: a cookbook for recurrence equations

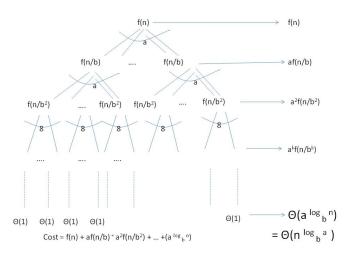
Master Theorem (part-I). Consider the recurrence equation

$$T(n) = \begin{cases} aT(n/b) + f(n) & \text{if } n \ge b \\ \Theta(1) & \text{if } n = 1 \end{cases}.$$

where,  $a \ge 1$  and b > 1 and n/b means either  $\lfloor n/b \rfloor$  or  $\lceil n/b \rceil$ .

• If  $f(n) = O(n^{\log_b a - \epsilon})$  for some constant  $\epsilon > 0$ , then,  $T(n) = \Theta(n^{\log_b a})$ .

#### Master Theorem -I: Recurrence tree



### Master Theorem -I: Cost

- Assume n is a power of b.
- The total cost is

$$= f(n) + af(n/b) + a^2f(n/b^2) + \ldots + a^{\log_b n}\Theta(1)$$

· Since,

$$a^{\log_b n} = b^{(\log_b a)(\log_b n)} = \left(b^{\log_b n}\right)^{\log_b a} = n^{\log_b a}$$

• total cost is =  $\sum_{j=0}^{\log_b n-1} a^j f(n/b^j)$ 

$$=\Theta\left(n^{\log_b a}+\frac{n^{\log_b a}}{a}f(b)+\frac{n^{\log_b a}}{a^2}f(b^2)+\ldots+f(n)\right)$$



### Master theorem-I: Cost

• The condition  $f(n) = O(n^{\log_b a - \epsilon})$  implies that

$$f(b^{j}) = O(b^{j(\log_b a - \epsilon)}) = a^{j} \cdot b^{-j\epsilon}$$

## Solution to recurrence equation

· Therefore, total cost is

$$= \Theta\left(n^{\log_b a} + \frac{n^{\log_b a}}{a}f(b) + \frac{n^{\log_b a}}{a^2}f(b^2) + \dots + f(n)\right)$$

$$= \Theta\left(\sum_{j=0}^{\log_b n} \frac{n^{\log_b a}}{a^j}f(b^j)\right)$$

$$= O\left(\sum_{j=0}^{\log_b n} n^{\log_b a}b^{-j\epsilon}\right), \text{ since, } f(b^j) = O(a^jb^{-j\epsilon})$$

$$= O\left(n^{\log_b a}\right)\sum_{j=0}^{\log_b n} b^{-j\epsilon}$$

$$= O\left(n^{\log_b a} \cdot \frac{1}{1 - b^{-\epsilon}}\right) = O\left(n^{\log_b a}\right)$$

#### **Total Cost**

- Since the first term is  $\Theta(n^{\log_b a})$ , the total cost is  $\Theta(n^{\log_b a})$ .
- The recursion tree is cost-wise bottom-heavy.

### Application of Master Theorem-I

Consider the recurrence equation

$$T(n) = \begin{cases} 8T(n/2) + \Theta(n^2) & \text{if } n = 2^k, k \ge 1\\ \Theta(1) & \text{if } n = 1 \end{cases}$$

• Applying Master Theorem-I, a = 8, b = 2 and  $f(n) = \Theta(n^2)$ , we have,

$$n^2 = O(n^{\log_2 8 - \epsilon}) = O(n^{3 - \epsilon})$$

for every  $0 < \epsilon < 1$ .

Hence, by Master theorem, the solution is

$$T(n) = \Theta(n^{\log_2 8}) = \Theta(n^3) .$$



### Strassen's Algorithm: Overview

- Strassen's algorithm performs only seven recursive multiplications of  $n/2 \times n/2$  matrices instead of eight.
- It performs several new additions of  $n/2 \times n/2$  matrices, but only a constant number. Total cost of additions remains  $\Theta(n^2)$ .

### Strassen's Algorithm: Overview

- 1. Divide the input matrices A and B and output matrix C into  $n/2 \times n/2$  submatrices, as before. This takes O(1) time by index calculation.
- 2. Create 10 matrices  $S_1, S_2, \ldots, S_{10}$  each of which is  $n/2 \times n/2$  and is the sum or difference of two matrices created in Step 1. This takes time  $\Theta(n^2)$ .
- 3. Using the submatrices of A, B and C, and  $S_1, \ldots, S_{10}$ , recursively compute seven matrix products  $P_1, P_2, \ldots, P_7$ . Each matrix  $P_i$  is  $n/2 \times n/2$ .
- 4. The output sub-matrices  $C_{11}$ ,  $C_{12}$ ,  $C_{21}$ ,  $C_{22}$  of the result matrix C is computed by adding and subtracting various combinations of the  $P_i$  matrices. This step can be done in time  $\Theta(n^2)$  time.

### Recurrence for Strassen's Algorithm

- We will see the exact calculations later, but we can write the recurrence relation for Strassen's method.
- Steps 1,2, and 4 require time  $\Theta(n^2)$ .
- Step 3 uses 7 recursive matrix multiplications of n/2 × n/2 matrices. This gives

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1\\ 7T(n/2) + \Theta(n^2) & \text{if } n > 1 \end{cases}$$

· By Master-theorem-I, the solution is

$$T(n) = \Theta(n^{\log_2 7}) = \Theta(n^{2.81...})$$
.

 Recall that we have four submatrices for each of A, B and C.

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}, \quad B = \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix}, C = \begin{pmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{pmatrix}$$

• Create the following ten matrices  $S_1, \ldots, S_{10}$ .

$$S_1 = B_{12} - B_{22}$$
  $S_2 = A_{11} + A_{12}$   
 $S_3 = A_{21} + A_{22}$   $S_4 = B_{21} - B_{11}$   
 $S_5 = A_{11} + A_{22}$   $S_6 = B_{11} + B_{22}$   
 $S_7 = A_{12} - A_{22}$   $S_8 = B_{21} + B_{22}$   
 $S_9 = A_{11} - A_{21}$   $S_{10} = B_{11} + B_{12}$ 

• These addition and subtraction of  $n/2 \times n/2$  matrices can be done in time  $\Theta(n^2)$ .



Recursively multiply n/2 × n/2 matrices seven times.

$$\begin{split} P_1 &= A_{11} \cdot S_1 = A_{11} \cdot B_{12} - A_{11} \cdot B_{22} \\ P_2 &= S_2 \cdot B_{22} = A_{11} \cdot B_{22} + A_{12} \cdot B_{22} \\ P_3 &= S_3 \cdot B_{11} = A_{21} \cdot B_{11} + A_{22} \cdot B_{11} \\ P_4 &= A_{22} \cdot S_4 = A_{22} \cdot B_{21} - A_{22} \cdot B_{11} \\ P_5 &= S_5 \cdot S_6 = A_{11} \cdot B_{11} + A_{11} \cdot B_{22} + A_{22} \cdot B_{11} + A_{22} \cdot B_{22} \\ P_6 &= S_7 \cdot S_8 = A_{12} \cdot B_{21} + A_{12} \cdot B_{22} - A_{22} \cdot B_{21} - A_{22} \cdot B_{22} \\ P_7 &= S_9 \cdot S_{10} = A_{11} \cdot B_{11} + A_{11} \cdot B_{12} - A_{21} \cdot B_{11} - A_{21} \cdot B_{12} \end{split}$$

 Note that the only multiplications are those in the middle column. The RHS shows the equivalent products.



 Step 4 adds and subtracts various P<sub>i</sub> matrices to construct the four n/2 × n/2 sub-matrices of the product C.

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

$$A_{11} \cdot B_{11} + A_{11} \cdot B_{22} + A_{22} \cdot B_{11} + A_{22} \cdot B_{22}$$

$$- A_{22} \cdot B_{11} + A_{22} \cdot B_{21}$$

$$- A_{11} \cdot B_{22} - A_{12} \cdot B_{22}$$

$$- A_{22} \cdot B_{22} - A_{22} \cdot B_{21} + A_{12} \cdot B_{22} + A_{12} \cdot B_{21}$$

$$A_{11} \cdot B_{11} + A_{12} \cdot B_{21}$$

which is the expression for  $C_{11}$ .



Set

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

 $C_{12}$  equals

$$A_{11} \cdot B_{12} - A_{11} \cdot B_{22} + A_{11} \cdot B_{22} + A_{12} \cdot B_{22}$$

$$A_{11} \cdot B_{12} + A_{12} \cdot B_{22}$$

which corresponds to the expression for  $C_{12}$ .

#### Setting

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

makes  $C_{21}$  equal to

$$A_{21} \cdot B_{11} + A_{22} \cdot B_{11} - A_{22} \cdot B_{11} + A_{22} \cdot B_{21}$$

$$A_{21} \cdot B_{11} + A_{22} \cdot B_{21}$$

which is the expression for  $C_{21}$ .

Finally, set

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

so that  $C_{22}$  equals

$$A_{11} \cdot B_{11} + A_{11} \cdot B_{22} + A_{22} \cdot B_{11} + A_{22} \cdot B_{22} \\ - A_{11} \cdot B_{22} + A_{11} \cdot B_{12} \\ - A_{22} \cdot B_{11} - A_{21} \cdot B_{11} \\ - A_{11} \cdot B_{11} - A_{11} \cdot B_{12} + A_{21} \cdot B_{11} + A_{21} \cdot B_{12}$$

$$+ A_{22} \cdot B_{22} + A_{21} \cdot B_{12}$$

which is the expression for  $C_{22}$ .