

# STRASSEN'S MATRIX MULTIPLICATION

## ① Conventional matrix multiplication method

Let  $A$  and  $B$  be two  $n \times n$  matrices

$C = A \times B$  is also an  $n \times n$  matrix.

$$C(i, j) = \sum_{1 \leq k \leq n} A(i, k) \times B(k, j)$$

→ To compute  $C[i, j]$  using this formula, we need  $n$  multiplications.

matrix  $C$  has  $n \times n = n^2$  elements.

→ The time for the resulting matrix multiplication algorithm is  $O(n^3)$

② The divide-and-conquer strategy suggests another way to compute the product of two  $n \times n$  matrices.

→ For simplicity, we assume that  $n$  is a power of 2 (i.e.,  $n = 2^k$ ). In case if  $n$  is not power of 2,

Then enough rows and columns of zeros can be added to both  $A$  and  $B$  so that the resulting dimensions are a power of two.

$$\begin{bmatrix} 10 & 5 & 8 \\ 6 & 4 & 9 \\ 15 & 3 & 20 \end{bmatrix}_{3 \times 3} \longrightarrow \begin{bmatrix} 10 & 5 & 8 & 0 \\ 6 & 4 & 9 & 0 \\ 15 & 3 & 20 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}_{4 \times 4}$$

$2^2 \times 2^2$



→ Imagine that  $A$  and  $B$  are each partitioned into 4 square submatrices having dimensions  $\frac{n}{2} \times \frac{n}{2}$

→ If  $AB$  is  $\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix} = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$  Then

$$\begin{array}{l} C_{11} = A_{11}B_{11} + A_{12}B_{21} \\ C_{12} = A_{11}B_{12} + A_{12}B_{22} \\ C_{21} = A_{21}B_{11} + A_{22}B_{21} \\ C_{22} = A_{21}B_{12} + A_{22}B_{22} \end{array} \longrightarrow (1)$$

→ This algorithm will continue applying itself to smaller-sized submatrices until  $n$  becomes suitably small ( $2 \times 2$ ) so that the product can be computed directly.

→ To compute  $AB$  using (1) we need to perform  
8 multiplications of  $\frac{n}{2} \times \frac{n}{2}$  matrices and  
4 additions of  $\frac{n}{2} \times \frac{n}{2}$  matrices.

→ Two  $\frac{n}{2} \times \frac{n}{2}$  matrices can be added in time  $cn^2$  for some constant  $c$ .

→ The overall computing time  $T(n)$  of the resulting divide-and-conquer algorithm is given by the recurrence relation

$$T(n) = \begin{cases} b & \text{if } n \leq 2 \\ 8T(n/2) + cn^2 & \text{if } n > 2 \end{cases}$$



## Derivation of Time Complexity:-

$$\begin{aligned}T(n) &= 8T(n/2) + cn^{\sim}\\&= 8 \left[ 8T(n/4) + c \cdot \frac{n^{\sim}}{4} \right] + cn^{\sim}\\&= 8^2 T(n/4) + 3cn^{\sim}\\&= 8^2 \left[ 8T(n/8) + c \cdot \frac{n^{\sim}}{8} \right] + 3cn^{\sim}\\&= 8^3 T\left(\frac{n}{2^3}\right) + 7cn^{\sim}\end{aligned}$$

At  $k^{\text{th}}$  step, we can write

$$T(n) = 8^k T(n/2^k) + (2^k - 1)cn^{\sim}$$

Substitute  $n = 2^k \Rightarrow k = \log_2 n$

$$T(n) = 8^k T(n/n) + (n-1)cn^{\sim}$$

$$= (2^3)^k \times b + cn^3 - cn^{\sim}$$

$$= (2^k)^3 \times b + cn^3 - cn^{\sim}$$

$$T(n) = n^3 \times b + cn^3 - cn^{\sim}$$

$$T(n) \propto n^3$$

$\therefore T(n) = O(n^3)$

Again, we got same  $O(n^3)$ . No improvement over the conventional matrix multiplication method has been made. We can attempt to reformulate the equations for  $c_{ij}$  so as to have fewer multiplications and possibly more additions.



## Volker Strassen's method for matrix multiplication

Volker Strassen has discovered a way to compute the  $C_{ij}$  of equation (1) by using only 7 multiplications and 18 additions or subtractions. His method involves first computing the seven  $\frac{n}{2} \times \frac{n}{2}$  submatrices,  $P, Q, R, S, T, U$  and  $V$ .

$$P = (A_{11} + A_{22})(B_{11} + B_{22})$$

$$Q = (A_{21} + A_{22})B_{11}$$

$$R = A_{11}(B_{12} - B_{22})$$

$$S = A_{22}(B_{21} - B_{11})$$

$$T = (A_{11} + A_{12})B_{22}$$

$$U = (A_{22} - A_{11})(B_{11} + B_{12})$$

$$V = (A_{12} - A_{22})(B_{21} + B_{22})$$

$$C_{11} = P + S - T + V$$

$$C_{12} = R + T$$

$$C_{21} = Q + S$$

$$C_{22} = P + R - Q + U$$

The resulting recurrence relation for  $T(n)$  is

$$T(n) = \begin{cases} b & \text{if } n \leq 2 \\ 7T\left(\frac{n}{2}\right) + an^2 & \text{if } n > 2 \end{cases}$$

↑  
 $18 \times \frac{n}{2} \times \frac{n}{2}$



### Time Complexity derivation :-

$$T(n) = 7T(n/2) + an^{\tilde{v}}$$

$$= 7 \left[ 7T(n/4) + a \frac{n^{\tilde{v}}}{2^2} \right] + an^{\tilde{v}}$$

$$= 7^2 T(n/4) + an^{\tilde{v}} \left[ 1 + \frac{7}{4} \right]$$

$$= 7^3 \left[ 7T(n/8) + a \cdot \frac{n^{\tilde{v}}}{4^2} \right] + an^{\tilde{v}} \left[ 1 + \frac{7}{4} \right]$$

$$= 7^3 T(n/2^3) + an^{\tilde{v}} \left[ 1 + \frac{7}{4} + \frac{7^2}{4^2} \right]$$

⋮

Similarly at  $k^{\text{th}}$  step, we can write

$$T(n) = 7^k T(n/2^k) + an^{\tilde{v}} \left[ 1 + \frac{7}{4} + \frac{7^2}{4^2} + \dots + \frac{7^{k-1}}{4^{k-1}} \right]$$

$$\approx 7^k T(n/n) + an^{\tilde{v}} \left( \frac{7}{4} \right)^k$$

$$\approx 7^k T(1) + an^{\tilde{v}} \frac{7^k}{4^k}$$

$$\approx 7^k \times b + an^{\tilde{v}} \frac{7^k}{4^k} \quad (\because n = 2^k \quad k = \log_2 n)$$

$$\approx 7^{\log_2 n} \times b + an^{\tilde{v}} \times \frac{7^{\log_2 n}}{4^{\log_2 n}}$$

$$\approx 7^{\log_2 n} \times b + an^{\tilde{v}} \times \frac{7^{\log_2 n}}{n^{\log_2 4}}$$

$$\approx 7^{\log_2 n} \times b + a \cancel{n^{\tilde{v}}} \times \frac{7^{\log_2 n}}{\cancel{n^{\tilde{v}}}}$$

$$\approx 7^{\log_2 n} (a+b) = c \cdot n^{\log_2 7} = c \times n^{2.81}$$

$\therefore T(n) = O(n^{2.81})$

 Time Complexity reduced from  $O(n^3)$  to  $O(n^{2.81})$