

# Design and Analysis of Algorithms.

## UNIT-1

### Properties of Algorithm:

1. Efficient Output
2. Finiteness.

### Analysing of Algorithms:

→ It can be done in two ways.

#### Time Complexity $T(n)$

- Machine independent
- Counting no. of instructions  
(or)  
no. of assignments  
(or)  
no. of operations.

#### Space complexity.

- Instruction space
- Data space
- Environment space.

#### Constant time complexity

→ not depends on input size

```
int sum(a,b) {  
    return a+b;  
}
```

$T(n)=1$

#### Linear time complexity.

```
int sum(A, n) {  
    sum=0; ①  
    for(i=0; i<n; i++)  
        sum=sum+A[i];  
    return sum; ②  
}
```

Step Count function  $T(n) = 3n + 4 = O(n)$

Cost	Repetition
1	1
1	1
1	$n+1$
1	$n$
1	$n$
1	1

$T(n) = 3n + 4$

⇒ Linear time complexity depends on input size.  $O(n)$ .

# Asymptotic Notations:

1. Big Oh ( $O$ )
2. Omega ( $\Omega$ )
3. Theta ( $\Theta$ )
4. Small oh ( $o$ )  $\Rightarrow f(n) < c \cdot g(n)$
5. small Omega ( $\omega$ )  $\Rightarrow f(n) > c \cdot g(n)$

## 1. Big Oh ( $O$ ):

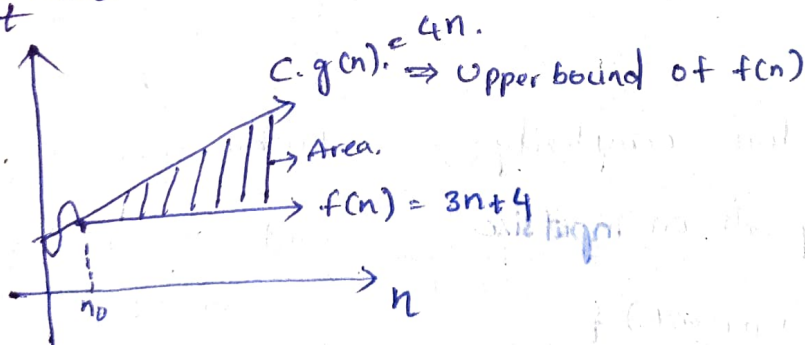
$\rightarrow$  Considering two functions.

$\uparrow$  step count function  
 $f(n)$  lesser than  $g(n) \rightarrow$  simple function.

$$f(n) = O(g(n)) \text{ iff}$$

$$f(n) \leq c \cdot g(n), \quad c > 0, \quad n \geq n_0 \geq 1$$

$n_0$  is  $\downarrow$  from which value the inequality is satisfying.  
the value  $\uparrow$



eg:-

$$T(n) = f(n) = 3n + 4$$

$$f(n) = O(g(n))$$

$$f(n) \leq c \cdot g(n)$$

$$3n + 4 \leq c \cdot n$$

$$3n + 4 \leq 4n$$

$$4 \leq n \Rightarrow n \geq 4$$

$$\boxed{n_0 = 4}$$

$g(n) = n$  (taking  $n$  because it is a simple linear function)

Assume  $c = 4, n_0 = 4$ .

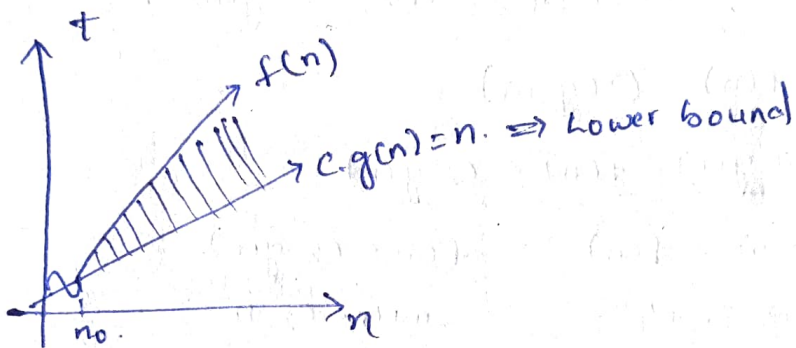
2.  $\Omega(n)$ :

→ Considering two functions.

$f(n)$  greater than  $g(n) \rightarrow$  Simple function

$$f(n) = \Omega(g(n)) \text{ iff}$$

$$f(n) \geq c \cdot g(n), \quad c > 0, n \geq n_0 \geq 1$$



eg:-

$$f(n) = T(n) = 3n + 4 = O(n), \quad g(n) = n.$$

$$f(n) = \Omega(g(n))$$

$$f(n) \geq c \cdot g(n)$$

$$3n + 4 \geq c \cdot n.$$

$$3n + 4 \geq n.$$

$$c = 1, n_0 = 1$$

$$2n + 4 \geq 0$$

$$n + 2 \geq 0$$

$$n \geq -2 \geq 1.$$

$$\boxed{n_0 = 1}$$

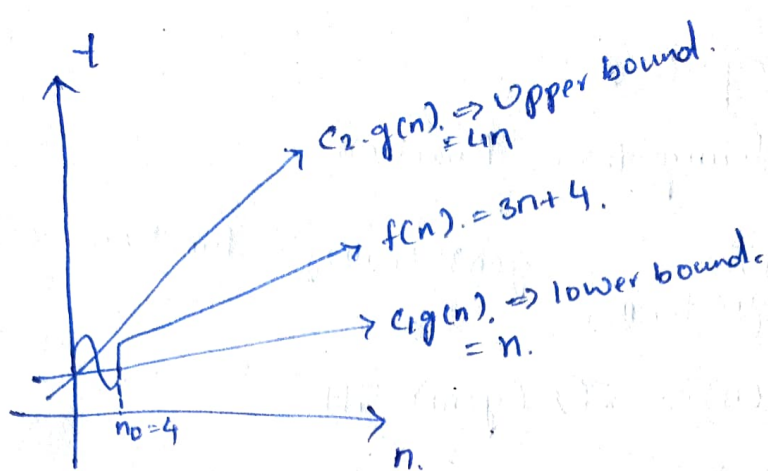
3.  $\Theta(n)$ :

→ Considering two functions.

$f(n)$  equal  $g(n) \rightarrow$  simple function,

$$f(n) = \Theta(g(n)) \text{ if } f$$

$$c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n), \quad c_1, c_2 > 0, n \geq n_0 \geq 1$$



eg:  $f(n) = T(n) = 3n + 4 = O(n)$ ,  $g(n) = n$ .

$$f(n) = O(g(n))$$

$$c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$$

$$c_1 \cdot g(n) \leq f(n)$$

$$c_1 \cdot n \leq 3n + 4$$

$$n \leq 3n + 4$$

$$c_1 = 1, n_0 = 1$$

$$f(n) \leq c_2 \cdot g(n)$$

$$3n + 4 \leq c_2 \cdot n$$

$$3n + 4 \leq 4n$$

$$n_0 = 4, c_2 = 4$$

$$T(n) = 3n + 4 = \Theta(n)$$

$$T(n) = 3n + 4 = \Omega(n)$$

eg:

①  $T(n) = 2n + 5$ ,  $g(n) = n$ .

$$f(n) = T(n) = 2n + 5$$

$$2n + 5 \leq c \cdot g(n)$$

$$2n + 5 \leq c \cdot n$$

$$2n + 5 \leq c \cdot n$$

$$2n + 5 \leq 3n \Rightarrow c = 3$$

$$\boxed{n_0 = 2} \quad 5 \leq n$$

$$n \geq 5$$

$$\Rightarrow \boxed{n_0 = 5}$$

$$\therefore T(n) = 2n + 5 = O(n)$$



$$② \quad f(n) = T(n) = 2n + 5$$

$$f(n) \geq c \cdot g(n)$$

$$2n + 5 \geq c \cdot n \Rightarrow c = 1$$

$$2n + 5 \geq n$$

$$n + 5 \geq 0$$

$$n \geq -5$$

$$n \geq -5 \geq 1 \Rightarrow \boxed{n_0 = 1}$$

$$\therefore T(n) = 2n + 5 = \Omega(n)$$

$$③ \quad f(n) = T(n) = 2n + 5$$

$$f(n) = 2n + 5$$

$$g(n) = n$$

$$c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$$

$$c_1 \cdot n \leq 2n + 5$$

$$n \leq 2n + 5$$

$$c_1 = 1$$

$$2n + 5 \geq 0$$

$$n \geq -5 \geq 1$$

$$\boxed{n_0 = 1}$$

$$f(n) \leq c_2 \cdot g(n)$$

$$2n + 5 \leq c_2 \cdot n$$

$$\boxed{c_2 = 3}$$

$$2n + 5 \leq 3n$$

$$5 \leq n$$

$$n \geq 5$$

$$\boxed{n_0 = 5}$$

$$\therefore T(n) = 2n + 5 = \Theta(n)$$

$$Rsum(A, n) \begin{cases} T(n) \end{cases}$$

$$\text{if } (n == 0) \text{ return } 0;$$

$$\text{return } Rsum(A, n-1) + A[n]$$

}

$$\frac{T(n-1) + 1}{T(n-1) + 2}$$

$$T(n) = \begin{cases} T(n-1) + 2 & n > 0 \\ 2 & n = 0 \end{cases}$$

Recursive Relation.

$$\begin{aligned}
 T(n) &= T(n-1) + 2 \\
 &= T(n-2) + 2 + 2 \\
 &= T(n-3) + 2 + 2 + 2 \\
 &\vdots
 \end{aligned}$$

$$n = T(n-n) + n(2).$$

$$T(n) = 2 + 2n = O(n)$$

1) Fibonacci Series:

	<u>Cost</u>	<u>Repetition</u>
Fib(n) {		
fib1 = 0;	1	1
fib2 = 1;	1	1
for(i=2 to n) do {	1	1
fib3 = fib1 + fib2	1	n
fib1 = fib2;	1	n-2
fib2 = fib3;	1	n-2
}	1	n-2
}		<hr/> 5n-5 = O(n)

$$T(n) = O(n)$$

2) Addition of two matrices:

Add(A, B, C, m, n) {	i = 0 — 1
for(i=0 to m) {	i < m — m+1
for(j=0 to n) {	i++ — m
C(i, j) = A(i, j) + B(i, j);	j = 0 — m
}	j < n — m(n+1)
}	j++ — mn
}	statement — mn
	<hr/> 3mn + 4m + 2

$$3mn + 4m + 2 = O(mn)$$

$$T(n) = O(mn).$$

### 3) Linear Search

```

Linear search (A, n, key) {
    for (i = 0 to n) do {
        if (key == A[i])
            return i;
    }
    return -1;
}

```

→ Best case:  $T(n) = O(1)$

→ Average case:  $T(n) = \frac{n+1}{2} = O(n)$   
middle for this case.  $A \cdot N$

→ Worst case:  $T(n) = O(n)$

### 4) Binary Search.

```

Binary search (A, start, end, key) {

```

```

    if (start > end) return -1;

```

```

    mid = (start + end) / 2;

```

```

    if (key == A[mid]) return mid;

```

```

    else if (key < A[mid]) {

```

```

        return Binary search (A, start, mid-1, key);
    }

```

```

}

```

```

return Binary search (A, mid+1, end, key);
}

```

$$T(n) = \begin{cases} 2, & n=0 \\ 4, & \text{found at mid} \\ 5+T(n/2), & n>0 \end{cases}$$

$$T(n) = \begin{cases} 2 & n=0 \\ 4 & n=1 \\ 5+T(n/2) & n>0 \end{cases}$$

$$T(n) = 5 + T(n/2)$$

$$= 5 + T(n/4) + 5$$

$$= 5 + T(n/8) + 5 + 5$$

$$\vdots$$

$$= T(n/2^k) + k(5).$$

$$2^{\frac{n}{k}} = 1$$

$$n = 2^k$$

$$k = \log n$$

$\log n$  times.

$$T(n) = T(1) + \log n(5)$$

$$= 5 \cdot \log n + 4$$

$$\boxed{T(n) = O(\log n)}$$

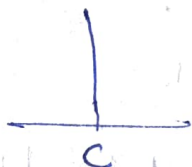
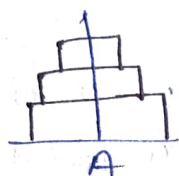
Q:-  $\log n, n, n^2, 1, 2^n, n^n, n^3$

A:-  $1 \leq \log n \leq n \leq n^2 \leq n^3 \leq 2^n \leq n^n$ . [Ascending order].

$$1 = O(n).$$

### Problem

#### Towers of Hanoi :



$\Rightarrow$  If the tower having  $n$  disks it requires  $2^n - 1$  steps to move from A to C.

Algorithm:

TOH( $n, A, C, B$ ) {

if ( $n == 1$ ) mov ( $A, C$ );

if ( $n > 1$ ) {

TOH( $n-1, A, B, C$ );

TOH( $1, A, C$ );

TOH( $n-1, B, C, A$ );

}

}

Eg:  $N=3$ .

TOH( $3, A, C, B$ ) {

TOH( $2, A, B, C$ ) {

TOH( $1, A, C, B$ )

TOH( $1, A, B, C$ )

TOH( $1, C, B, A$ )

}

TOH( $1, A, C, B$ )

TOH( $2, B, C, A$ )

TOH( $1, B, A, C$ )

TOH( $1, B, C, A$ )

TOH( $1, A, C, B$ )

1. mov ( $A, C$ ).

2.  $A \rightarrow B$

3.  $C \rightarrow B$

4.  $A \rightarrow C$

5.  $B \rightarrow A$

6.  $B \rightarrow C$

7.  $A \rightarrow C$



Time Complexity

$$T(n) = \begin{cases} 1 & n=1 \\ T(n-1)+1+T(n-1), & n>1 \end{cases}$$

$$T(n) = 2T(n-1) + 1$$

$$= 2[2T(n-2) + 1] + 1$$

$$= 2^2 T(n-2) + 2 + 1 \rightarrow 2^2 - 1$$

$$= 2^2 [2T(n-3) + 1] + 1$$

$$T(n) = 2^3 T(n-3) + 2^2 + 2 + 1 \rightarrow 2^3 - 1$$

After k steps,

$$T(n) = 2^k T(n-k) + 2^k - 1$$

$$n-k=1$$

$$k=n-1$$

$$\Rightarrow T(n) = 2^{n-1} \cdot T(1) + 2^{n-1} - 1$$

$$= 2 \cdot 2^{n-1} - 1 = 2^n - 1$$

$$\boxed{T(n) = O(2^n)}$$

Master's Theorem:- [dividing recurrence Relation].

$$T(n) = a \cdot T(n/b) + \Theta(n^k \log^p n) \quad a>0, b>0, k \geq 0$$

i)  $a > b^k$  :  $T(n) = \Theta(n \log^a_b)$

ii)  $a = b^k$  : a]  $p > -1$  :  $T(n) = \Theta(n^{\log_b a} \cdot \log^{p+1} n)$

b]  $p = -1$  :  $T(n) = \Theta(n^{\log_b a} \cdot \log \log n)$

c]  $p < -1$  :  $T(n) = \Theta(n^{\log_b a})$

iii)  $a < b^k$  : a]  $p \geq 0$  :  $T(n) = \Theta(n^k \log^p n)$

b]  $p < 0$  :  $T(n) = \Theta(n^k)$

Eg:

$$① T(n) = a \cdot T(n/2) + 2$$

$$a=1, b=2, k=0, p=0.$$

$$1 = 2^0 = 1 \Rightarrow a = b^k.$$

$$T(n) = n^{\log_2^1} \cdot \log^1 n.$$

$$= n^{\log_2^0} \cdot \log^1 n.$$

$$= n^0 \cdot \log n$$

$$= \Theta(\log n)$$

$$= O(\log n).$$

$$② T(n) = 2T(n/2) + n$$

$$③ T(n) = 2T(n/2) + n \log n.$$

$$a=2, b=2, k=1, p=1$$

$$ii) 2 = 2^1$$

$$T(n) = \Theta(n^{\log_2^2})$$

## For decreasing recurrence Relation:

$$T(n) = aT(n-b) + f(n)$$

$$f(n) = O(n^k), a > 0, b > 0, k > 0$$

1.  $a < 1: T(n) = f(n)$

2.  $a = 1: T(n) = O(n \cdot f(n))$

3.  $a > 1: T(n) = O(a^{n/b} \cdot f(n))$

Eg: ①  $T(n) = 2T(n-1) + 1$

$$a = 2, b = 1, f(n) = 1 = O(n^0)$$

$$T(n) = O(2^{n/1} \cdot 1)$$

$$T(n) = O(2^n)$$

②  $T(n) = T(n-1) + n$

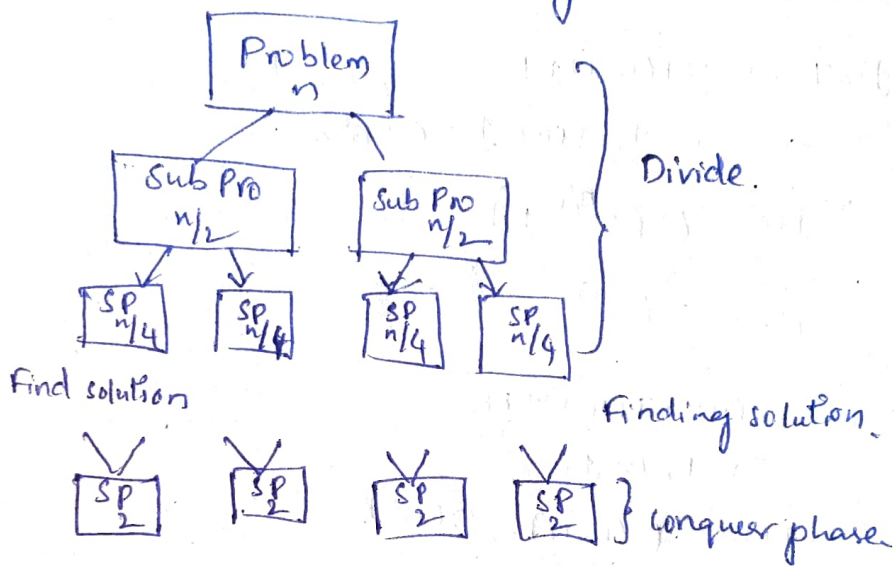
$$a = 1, b = 1,$$

$$T(n) = O(n \cdot n) = O(n^2)$$

# Divide and Conquer

Eg: 1) Merge sort 2) Quick sort 3) Binary search  
4) Tree traversals 5) Strassen's matrix multiplication

- Steps:
1. Dividing the problem into subparts.
  2. Finding the solution to subparts.
  3. Combining all these solutions to get overall solution.



1) Merge sort

Recurrence relation :  $T(n) = \begin{cases} 1 & n=1 \\ 2T(n/2) + n & n>1 \end{cases}$

Time complexity :  $O(n \log n)$ .

2) Quick sort

Recurrence relation for worst case [because of choosing fixed position of pivot].

i.e Pivot =  $A[\text{start}]$

Pivot =  $A[\text{end}]$

$T(n) = \begin{cases} T(n-1) + n & n>1 \\ 1 & n=1 \end{cases}$

Best Case:  $T(n) = \begin{cases} 2T(n/2) + n & n>1 \\ 1 & n=1 \end{cases}$

Time complexity for worst case:  $O(n^2)$

Best Case :  $O(n \log n)$ .



Merge

Quick.

1.  $O(n \log n)$

1.  $O(n^2)$ .

2. Extra space is required

2. No extra space is required

3. Spread sheets, programming languages.

Randomization.

→ Optimization technique for avoiding the worst case in Quick sort

⇒ Choose pivot randomly with uniform distribution.

3) Strassen's Matrix Multiplication.

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}_{2 \times 2}$$

$$B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}_{2 \times 2}$$

$$A \times B = \begin{bmatrix} A_{11} \times B_{11} + A_{11} \times B_{21} & A_{11} \times B_{12} + A_{11} \times B_{22} \\ A_{21} \times B_{11} + A_{21} \times B_{21} & A_{21} \times B_{12} + A_{22} \times B_{22} \end{bmatrix}_{2 \times 2}$$

⇒ Size no. of \* operators

$2 \times 2$  8

$4 \times 4$  64

$8 \times 8$  512

$$T(n) = \begin{cases} 8T(n/2) + 1 & n > 1 \\ 1 & n = 1 \end{cases} \Rightarrow \text{Recurrence Relation}$$

$$T(n) = O(n^3) \Rightarrow \text{Time complexity}$$

→ Add (same row)  
Multiply (same column),

$$C = \begin{bmatrix} C_{11} & C_{12} \\ C_{21} & C_{22} \end{bmatrix}$$

$$P = (A_{11} + A_{22}) \cdot (B_{21} + B_{22})$$

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

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

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

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

$$U = (A_{21} - A_{11}) \cdot (B_{11} + B_{12})$$

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

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

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

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

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

$$T(n) = \begin{cases} 7 \cdot T(n/2) + 1 \\ 1 & ; n=1 \end{cases}$$

$$T(n) = O(n^{2.81}) = O(n^{\log 7})$$

for 4x4

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}_{4 \times 4}$$

$$B = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}_{4 \times 4}$$

SMM

$$2 \times 2 = 7$$

$$4 \times 4 = 49$$

$$8 \times 8 = 343$$

NMM

$$2 \times 2 = 8$$

$$4 \times 4 = 64$$

$$8 \times 8 = 512$$

Pseudo Code

Strassen(A, B, n) {

  if (n == 1) {

    C ← new matrix n x n.

    C[0][0] = A[0][0] \* B[0][0];

  } else {

    C ← new matrix n x n.

    A<sub>11</sub>, A<sub>12</sub>, A<sub>21</sub>, A<sub>22</sub> ← A.split\_matrix();

    B<sub>11</sub>, B<sub>12</sub>, B<sub>21</sub>, B<sub>22</sub> ← B.split\_matrix();

$$P = \text{Strassen}(A_{11} + A_{22}, B_{11} + B_{22}, n/2);$$

$$Q = \text{Strassen}(A_{21} + A_{22}, B_{11}, n/2);$$

$$R = \text{Strassen}(A_{11}, B_{12} - B_{22}, n/2);$$

$$S = \text{Strassen}(A_{22}, B_{21} - B_{11}, n/2);$$

$$T = \text{Strassen}(A_{11} + A_{12}, B_{22}, n/2);$$

$$U = \text{Strassen}(A_{21} - A_{11}, B_{11} + B_{12}, n/2);$$

$$V = \text{Strassen}(A_{12} - A_{22}, B_{21} + B_{22}, n/2);$$

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

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

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

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

$$C = \text{combine\_quadrants}(C_{11}, C_{12}, C_{21}, C_{22});$$

}

return C;

}

Eg:-

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}_{2 \times 2}$$

$$B = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}_{2 \times 2}$$

$$\begin{matrix} 1 \times 1 + 2 \times 3 & 1 \times 2 + 2 \times 4 \\ 3 \times 1 + 4 \times 3 & 3 \times 2 + 4 \times 4 \end{matrix} \Rightarrow \begin{bmatrix} 7 & 10 \\ 15 & 22 \end{bmatrix}$$

$$P = (1+4) \cdot (1+4) = 25$$

$$Q = (3+4) \cdot 1 = 7$$

$$R = 1 \cdot (2-4) = -2$$

$$S = 4 \cdot (3-1) = 8$$

$$T = (1+2) \cdot 4 = 12$$

$$U = (3-1) \cdot (1+2) = 6$$

$$V = (2-4) \cdot (3+4) = -14$$

$$C_{11} = 25 + 8 - 12 - 14 = 7$$

$$C_{12} = -2 + 12 = 10$$

$$C_{21} = 7 + 8 = 15$$

$$C_{22} = 25 + (-2) - 7 + 6 = 22$$

$$A \times B = \begin{bmatrix} 7 & 10 \\ 15 & 22 \end{bmatrix}$$