

FORD FULKERSON ALGORITHM:

(for maximum flow prob).

PROBLEM:

Given a graph which represents a flow n/w where every edge has a capacity. Also, given two vertices source 's' and sink 't' in the graph. Find out the maximum possible flow from 's' to 't' with following constraints.

- Flow on an edge does not exceed given capacity of the edge.
- In-flow is equal to out-flow for every vertex except 's' and 't'.

ALGORITHM:

FORD FULKERSON ALGORITHM (G, s, t)

initialize flow f to 0

while there exists an augmenting path p in the residual n/w G_f :

 augment flow f along p .

 update residual network

return f .

TERMINOLOGY:

Residual graph:

It's a graph which indicates additional possible flow. If there is such path from source to sink, then there is possibility to add flow.

Residual Capacity:

It's mathematically defined as:

$$\text{residual capacity} = \text{original capacity} - \text{current flow}$$

Minimum Cut:

* Also known as bottleneck capacity.

* It is the minimum capacity of any edge on the graph

Augmenting path:

* A path from 's' to 't', no cycle & only five weights

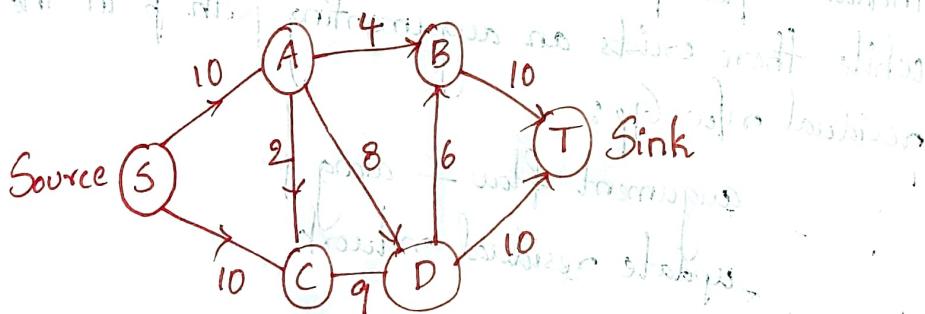
Two ways:

- ① Non-full forward edges
- ② Non-empty backward edges

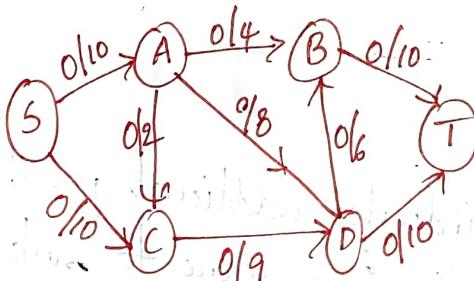
Time Complexity

$$O(E f^k l)$$

EXAMPLE:

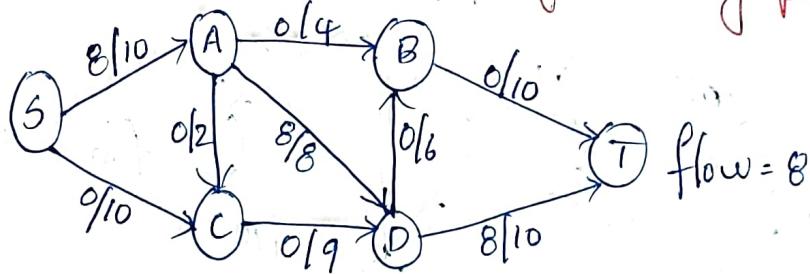


Step 1: Initialize flow as zero at all edges.



The process is to select a path, calculate a bottleneck, assign it to each edge of the path, calculate residual capacity & iterate through entire process till all edges are exhausted & maximum flow is attained.

Path $\rightarrow S \rightarrow A \rightarrow D \rightarrow T$, Bottleneck-8, Assign '8' along path.



Path

Bottleneck

$S \rightarrow A \rightarrow D \rightarrow T$

8

$S \rightarrow C \rightarrow D \rightarrow T$

2

$S \rightarrow C \rightarrow D \rightarrow A \rightarrow B \rightarrow T$

4

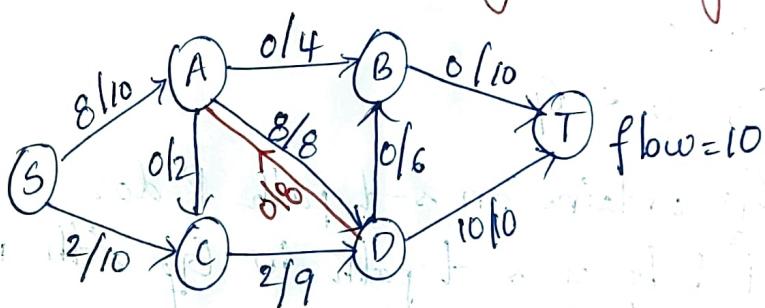
$S \rightarrow C \rightarrow D \rightarrow B \rightarrow T$

3

$S \rightarrow A \rightarrow D \rightarrow B \rightarrow T$

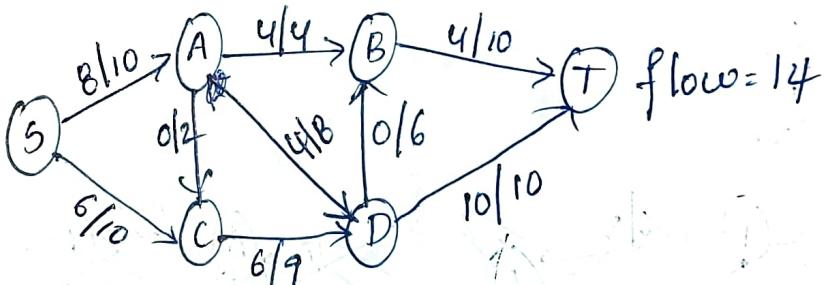
2

Path $\rightarrow S \rightarrow C \rightarrow D \rightarrow T$, Bottleneck-2, Assign '2' along path.

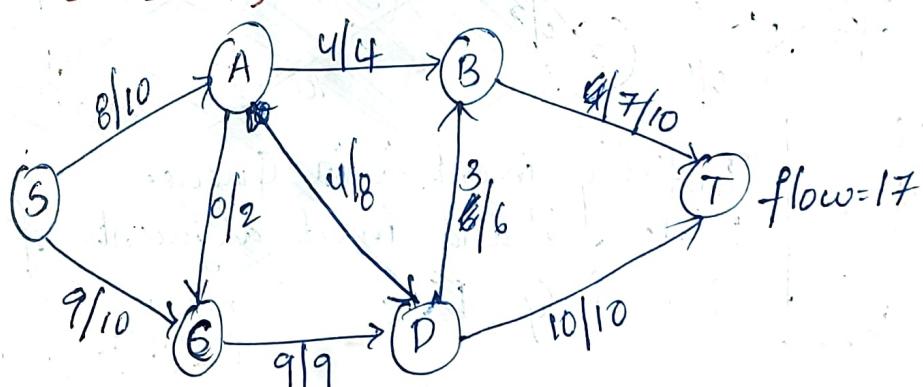


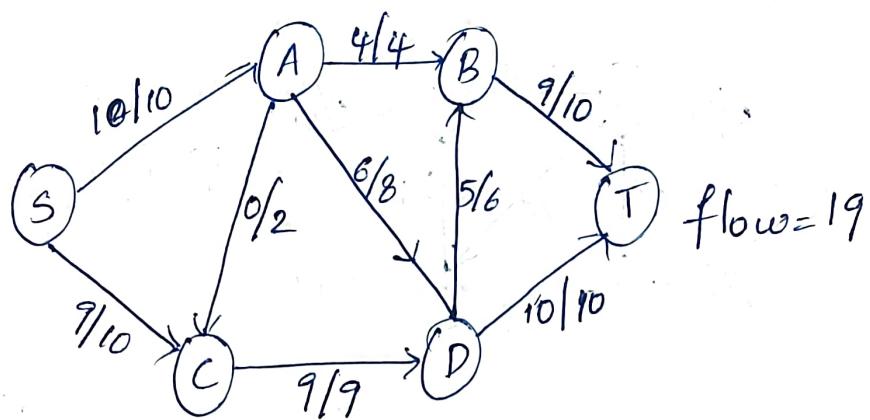
Path $\rightarrow S \rightarrow C \rightarrow D \rightarrow A \rightarrow B \rightarrow T$, [DA \rightarrow Non-empty backward edge \Rightarrow Taking reverse edge]

Bottleneck $\rightarrow 4$



Path $\rightarrow S \rightarrow C \rightarrow D \rightarrow B \rightarrow T$, Bottleneck $\rightarrow 3$.





EDMOND-KARP ALGORITHM:

An implementation of Ford Fulkerson that specifically mentions the use of BFS to find maximum flow.

ALGORITHM:

$$f \leftarrow 0; G_f \leftarrow G$$

while G_f contains an $s-t$ path P , do:

Let P be an $s-t$ path in G_f with min no. of edges
Augment f using P .

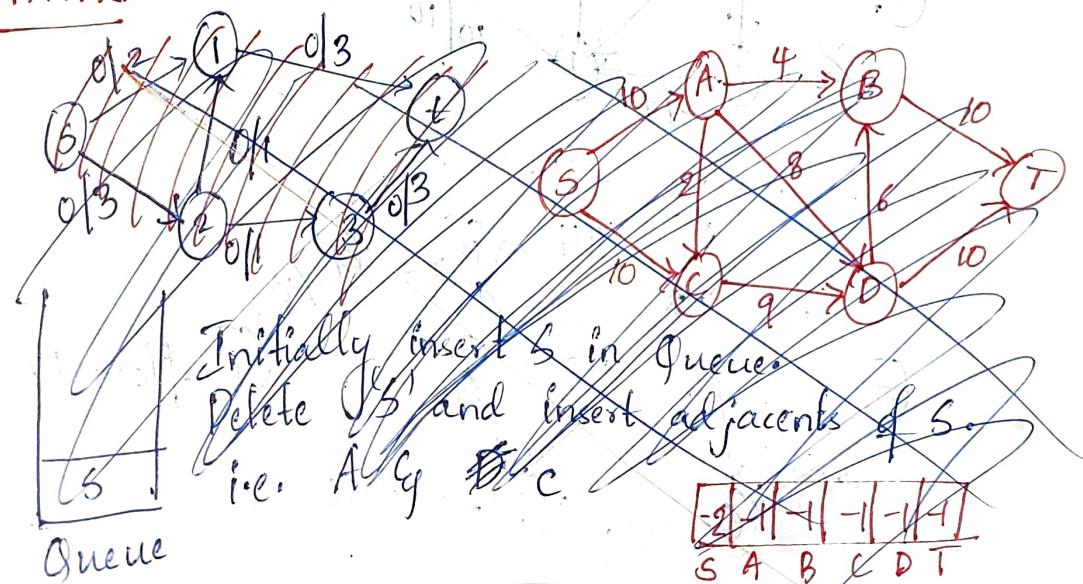
Update G_f

end while

return f .

PROBLEM:

PATH:

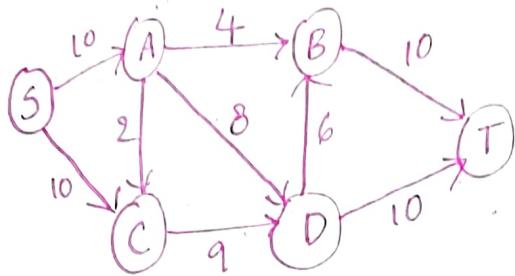


IMPLEMENTATION:

① Find the shortest path.

② Increment the flow monotonically.

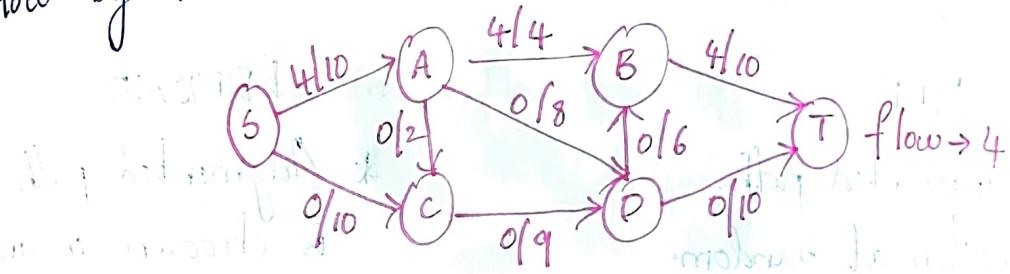
} DONE Using
BFS.



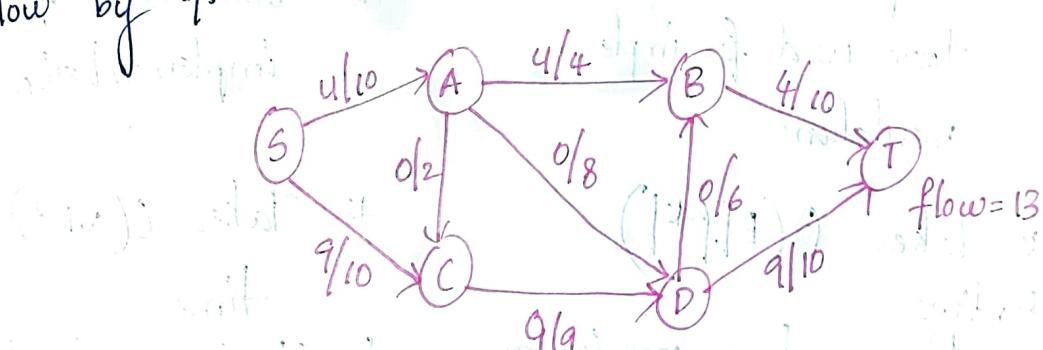
In the above graph, we have two shortest paths:

$S \rightarrow A \rightarrow B \rightarrow T$ & $S \rightarrow C \rightarrow D \rightarrow T$.

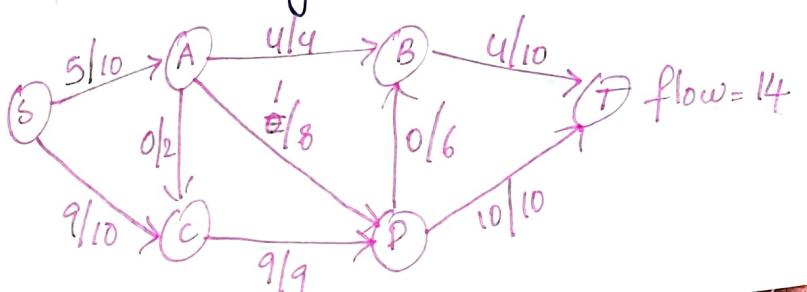
We consider $S \rightarrow A \rightarrow B \rightarrow T$. Bottleneck $\rightarrow 4$. Increment flow by 4.



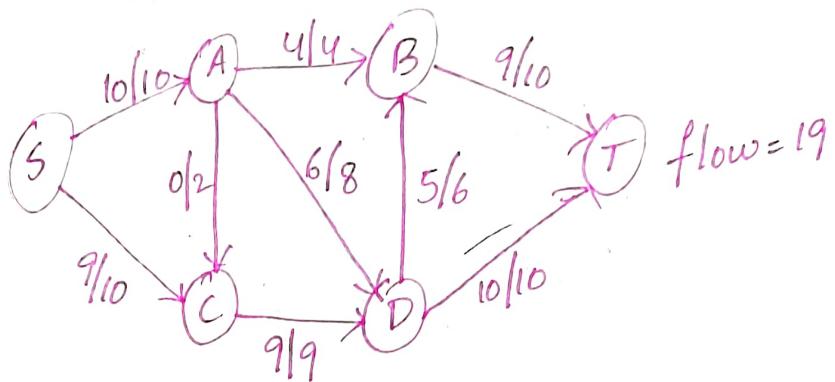
Now, we consider $S \rightarrow C \rightarrow D \rightarrow T$. Bottleneck $\rightarrow 9$. Increment flow by 9.



The next shortest path is: $S \rightarrow A \rightarrow D \rightarrow T$. Bottleneck is 1. So increment flow by 1.



The next shortest path is $S \rightarrow A \rightarrow D \rightarrow B \rightarrow T$.
 Bottleneck is 5. Increment flow by 5.



As the paths are now exhausted, we attain maximum flow = 19

MAIN DIFFERENCES BETWEEN FORD & EDMOND

FORD

- * Augmented paths chosen at random.
- * No special datastructure - closure used for implementation.
- * Takes $O(Elf^*)$ runtime
- * Dependent on maximum flow of network, f^* .

EDMOND

- * Augmented path is chosen in an ordered manner.
- * Uses BFS for implementation
- * Takes $O(VE^2)$ time
- * Independent of maximum flow of network, f^* .

DESCRIBE MAX-FLOW MIN-CUT THEOREM AND IT'S APPLICATIONS:

MAX-FLOW MIN-CUT:

Max flow Min Cut theorem states that, in a flow n/w, the max amount of flow passing through the source to the sink is equal to total weight of edges in a minimum cut (smallest total weight of the edges which if removed would disconnect source from the sink)

Max-flow Min-Cut Conditions:

If f is a flow in a flow n/w $G = (V, E)$ with source ' s ' and sink ' t ', then following conditions are equivalent:

1. f is a maximum flow in G .
2. the residual n/w G_f contains no augmenting paths.
3. $|f| = c(S, T)$ for some cut (S, T) of G .

Min-cut:

Set of edges whose removal divides n/w into 2 halves $X \cup Y$ such that

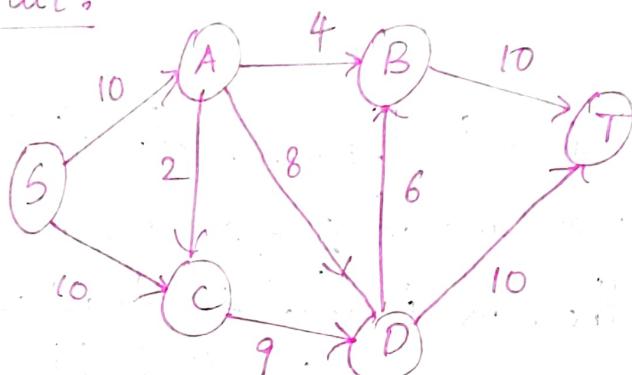
$$s \text{ (source)} \in X$$

$$t \text{ (sink)} \in Y$$

Example Illustration:

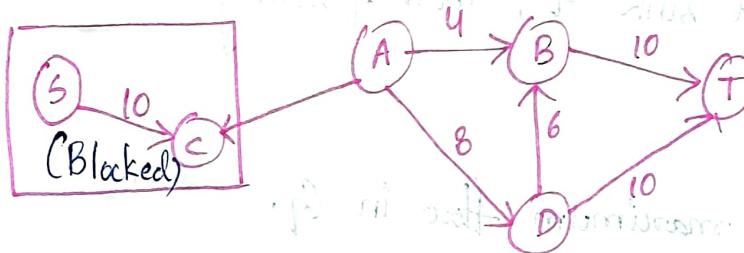
(For Max-flow refer to ford fulkerson)

For Min-cut:



When disconnecting the graph, it should be ensured that sum of weights of removed edges = max flow
Max-flow = 19.

If we remove $S \rightarrow A$ and $C \rightarrow D$



$$S \rightarrow A = 10$$

$$C \rightarrow D = 9$$

$$\text{Min-cut} = 19$$

weight

APPLICATIONS:

- ① Network reliability, availability & connectivity
- ② Matching in graphs (bipartite matching)
- ③ Airlines use this to decide when to allow planes to leave airports to maximize flow of flights.

DEFINE DIVIDE AND CONQUER METHOD AND MENTION ADV & DISADV.

- * Many Algorithms are recursive in nature
- * These Algorithms typically follow a divide-and-conquer approach

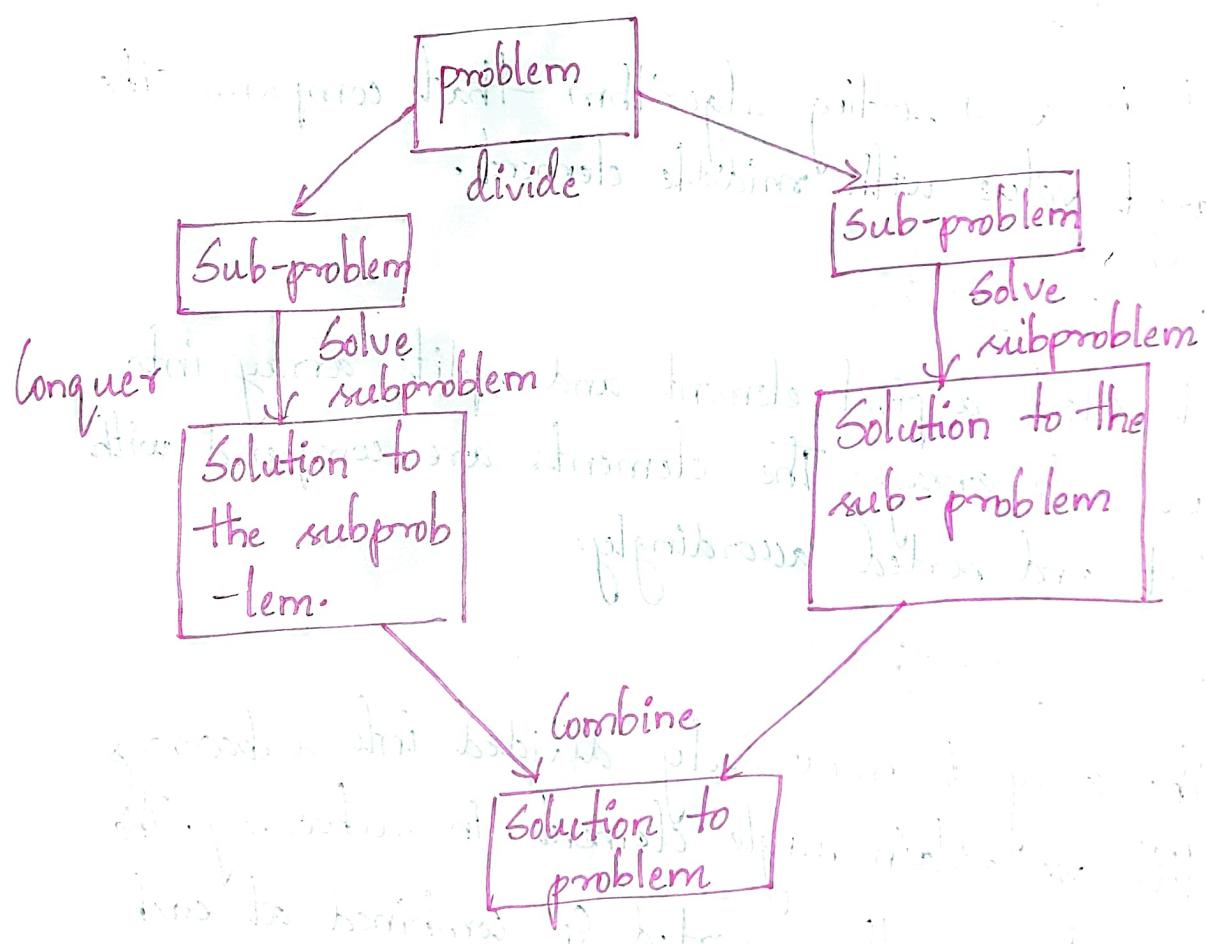
There are three steps at each level of recursion:

Divide & Conquer: Steps:

Divide: the problem is divided into subproblems

Conquer: The subproblems are solved recursively.

Combine: Merge the solutions to the subproblems into solution for original problem.



There are two fundamentals to Divide & Conquer Strategy:

1. Relational formula:

Formula that is developed and used to perform D & C strategy.

2. Stopping Condition:

Condition that specifies where the D & C strategy is to be stopped.

APPLICATIONS:

1. Binary Search:

It is a sorting algorithm that compares the target value with middle element.

2. Quick Sort:

It selects a pivot element and splits array into two sub-arrays. The elements are compared with pivot and sorted accordingly.

3. Merge Sort:

The array is recursively divided into subarrays till we attain single element in subarray. The elements are then sorted & combined at each level to attain sorted array.

4. Strassen's Algorithm:

An algorithm for matrix multiplication. It is faster than traditional algorithm.

Advantages:

1. Used to solve complicated problems such as towers of Hanoi.
2. Efficient use of cache memory
3. proficient than brute force technique
4. System can easily handle D & C by parallel processing.

Disadvantages:

1. Requires high memory management
2. Explicit stack is required resulting in space overuse
3. System may crash due to rigorous recursion.

EXPLAIN STRASSEN'S MATRIX MULTIPLICATION ALGORITHM

ALGORITHM

Given 2 square matrices A and B of size $n \times n$ each, find multiplication matrix.

NAIVE METHOD:

```
void multiply (int A[N][N], int B[N][N], int C[N][N])  
{  
    for (int i=0; i<N; i++)  
    {  
        int (j=0, j<N; j++)  
    }  
}
```

$$C[i][j] = 0;$$

for (int k=0; k < n; k++)

$$C[i][j] += A[i][k] * B[k][j];$$

Time Complexity is $O(N^3)$

DIVIDE AND CONQUER:

- * Matrices are divided into sub-matrices of size $n/2 \times n/2$.

$$\left[\begin{array}{c|c} a & b \\ \hline c & d \end{array} \right] \times \left[\begin{array}{c|c} e & f \\ \hline g & h \end{array} \right] = \left[\begin{array}{c|c} aetbg & af+bh \\ \hline cetdg & cf+dh \end{array} \right]$$

$A, B, C \rightarrow n \times n$ matrices

$a, b, c, d \rightarrow$ submatrices of A of size $N/2 \times N/2$

e,f,g,h → " " $\text{with } B_{\text{out}} \text{ " } N/2 \times N/2$

Time Complexity is $O(N^3)$

A better approach is to use strassen's matrix multiplication that reduces recursive calls

from 8 to 7.

STRASSEN'S METHOD

- * Uses divide and conquer.
- * Uses 7 calls and not 8.

$$\begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \times \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

$$M_1 = (A_{11} + A_{22})(B_{11} + B_{22})$$

$$M_2 = (A_{21} + A_{22})B_{11}$$

$$M_3 = (A_{11})(B_{12} - B_{22})$$

$$M_4 = A_{22}(B_{21} - B_{11})$$

$$M_5 = (A_{11} + A_{12})B_{22}$$

$$M_6 = (A_{21} - A_{11})(B_{11} + B_{22})$$

$$M_7 = (A_{12} - A_{22})(B_{21} + B_{22})$$

$$C_{11} = M_1 + M_4 - M_5 + M_7$$

$$C_{12} = M_3 + M_5$$

$$C_{21} = M_2 + M_4$$

$$C_{22} = M_1 - M_2 + M_3 + M_6$$

Time Complexity: $O(n^{\log 7})$