

Network Flow

6.06.22

Outline

1. Introduction
2. Flow network definition
3. Flow and capacity
4. The maximum flow problem
5. Cuts in FN
6. Residual network
7. Augmenting paths
8. Ford Fulkerson's Algo on FN
9. Example
10. Max-flow-min cuts theorem
11. Application of NF
 - (i) Matching (Bipartite matching)

Network Flow

Introduction

- Model of transportation
- Every vertex → is called switch
- Every Edge → is called Edge

- * Graph can be model as a transportation network
 - (i) whose edges carry some sort of traffic
 - (ii) whose nodes act as a switch passing traffic between the edge.

② Definition of F.N

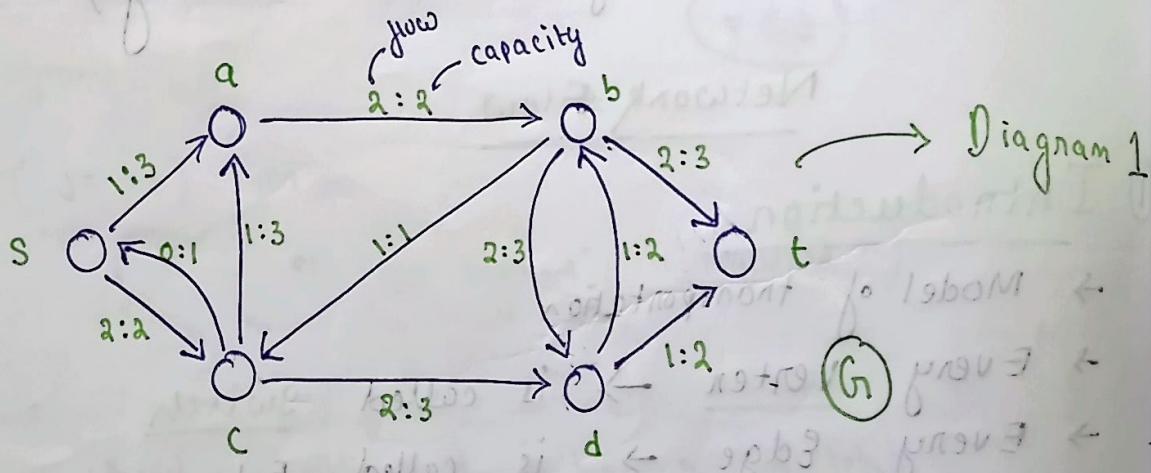
A flow network is a directed weighted graph $G(V, E)$ such that:

- (i) Each and every edge of the graph, associated with a non-negative capacity which is denoted by $C(u, v)$ if $(u, v) \in E$
if $(u, v) \notin E$ then $C(u, v) = 0$

- (ii) In this graph there are 2 distinguished vertices

- one is called source denoted by s
- another is sink denoted by t

- (iii) For each vertex $v \in V$ then they have they have a path $s \rightarrow v \rightarrow t$



Definition of flow

A flow in a flow network G , is a real valued function

$$f: V \times V \rightarrow \mathbb{R}$$

that satisfy the 2 property:-

(i) Capacity constraint [for all $u, v \in V$]
we require

$$0 \leq f(u, v) \leq c(u, v)$$

(ii) Flow conservation [$\forall u \in V - \{s, t\}$]

* not applicable for source and sink

incoming flow of a node $v \rightarrow \sum f(u, v) = \sum f(v, u)$ outgoing flow of a node v

$$\Rightarrow \sum f(u, v) - \sum f(v, u) = 0$$

Maximum Flow Problem

Problem statement: Given $f \in (s, t)$

Given a flow network G_f , with flow f and capacity c . Find a flow of max. value of G_f

Goal:

Arrange the traffic so as to make f as efficient use as possible of the available capacity

Value of flow

The value of a flow f in a flow network G_f is defined by:

$$|f| = \sum_{v \in V} f(s, v) + \sum_{v \in V} f(v, s)$$

flow conservation

$$\left| \begin{array}{l} \text{for source } (\text{incoming} = 0) \\ |f| = \sum_{v \in V} f(s, v) \end{array} \right| \quad \left| \begin{array}{l} \text{for destination } (\text{outgoing} = 0) \\ |f| = \sum_{v \in V} f(v, t) \end{array} \right.$$

Property of flow

Property 1

if $x \subseteq V$, then $f(x, x) = 0$

if $v \in V$, then $f(v, v) = 0$

Property 2

new symmetry if $(x, y) \subseteq V$ then $f(x, y) = -f(y, x)$

if $v, u \in V$ then $f(v, u) = -f(u, v)$

Property 3

if $x, y, z \subseteq V$, $x \cap y = \emptyset$

(i) $f(x \cup y, z) = f(x, z) + f(y, z)$

(ii) $f(z, x \cup y) = f(z, x) + f(z, y)$

Theorem 1

The flow value of a FN

$$\begin{aligned} |f| &= f(v, t) \\ &= \sum_{v \in V} f(v, t) \end{aligned}$$

The value of a flow in FN is equal to the flow into the sink.

$$(2, 0)f_3 + (0, 2)f_3 = 16$$

Theorem 1

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$$|\delta| = \delta(v, t)$$

Proof.

$$|\delta| = \sum_{v \in V} \delta(s, v) = \delta(s, v)$$

$$\text{by rule ①} = \delta(s \cup v - s, v) - \delta(v - s, v)$$

$$\text{of flow property} = \underbrace{\delta(v, v)}_0 - \delta(v - s, v)$$

$$= \delta(v, v - s)$$

$$= \delta(v, t \cup v - s - t)$$

$$= \delta(v, t) + \underbrace{\delta(v, v - s - t)}_0 \quad \text{- By applying flow conservation property}$$

$$= \delta(v, t)$$

proved

Cuts

A cut (S, T) of a FN or (V, E) is a partition of vertices V into 2 disjoint sets S and T where

$$T = V - S \quad \text{and} \quad s \in S \quad \text{and} \quad t \in T$$

$$S \cap T = \emptyset$$

Net flow across the cut (S, T)

If f is a flow on a flow network G , the net flow across

$$f(S, T) = f(u, v) - f(v, u), \forall u \in S, v \in T$$

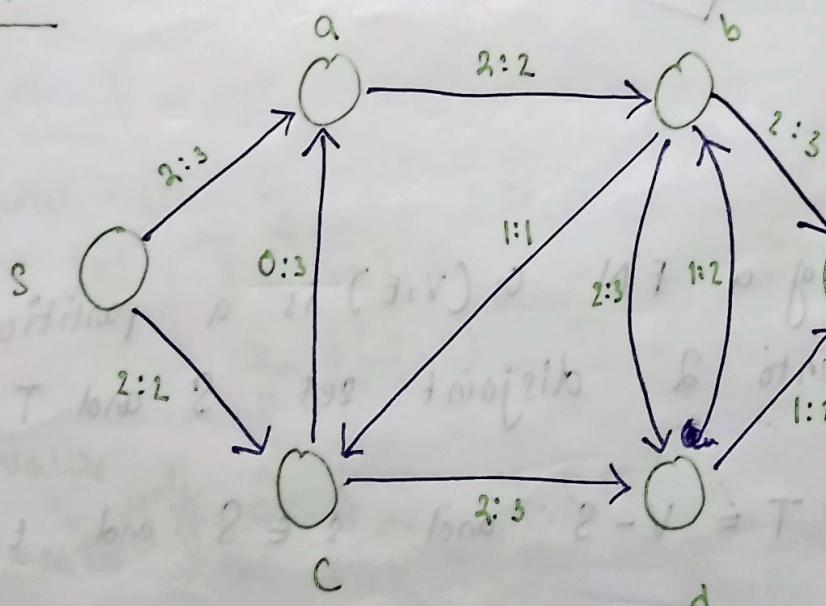
Capacity of a cut (S, T) (\sum sum of capacities of all edges coming out of S to T)

$$C(S, T) = \sum_{\forall u, v} C(u, v) \text{ when } u \in S \text{ and } v \in T$$

Minimum Cut

A minimum cut of a flow network G is a cut, whose capacity is minimum over all cuts of flow network G .

Example



① $S = \{a, b\}$ $T = \{s, c, d, t\}$

cut (S, T) is a valid cut in the FN G

$$\textcircled{2} \quad S = \{s, a, b, c\} \quad T = \{d, t\}$$

$$\textcircled{3} \quad S = \{s, t\} \quad T = \{a, b, c, d\}$$

$$\textcircled{4} \quad S = \{s, a\} \quad T = \{b, c, d, t\}$$

Cut Validity

$$\textcircled{1} \quad S = \{a, b\}, \quad T = \{s, c, d, t\}$$

here $s \notin S$, $t \notin T$, $S \cap T = \emptyset$
 \downarrow
 $\boxed{\text{Not valid cut}}$

$$\textcircled{2} \quad S = \{s, a, b, c\}, \quad T = \{d, t\}$$

here $s \in S$, $t \in T$, $S \cap T = \emptyset$
 \downarrow
 $\boxed{\text{Valid cut}}$

$$\textcircled{3} \quad S = \{s, t\}, \quad T = \{a, b, c, d\}$$

here $s \in S$, $t \notin T$, $S \cap T = \emptyset$
 \downarrow
 $\boxed{\text{not valid cut}}$

$$\textcircled{4} \quad S = \{s, a\}, \quad T = \{b, c, d, t\}$$

here $s \in S$, $t \in T$, $S \cap T = \emptyset$
 \downarrow
 $\boxed{\text{Valid cut}}$

Net flow across cut (S, T)

$$\textcircled{2} \quad S = \{s, a, b, c\} \quad T = \{d, t\}$$

$$f(S, T) = f(a, b, d) + f(b, t) + f(c, d) - f(d, b)$$
$$= 2 + 2 + 2 - 1$$

$$\boxed{f(S, T) = 5}$$

$$\textcircled{3} \quad \textcircled{4} \quad S = \{s, a\} \quad T = \{b, c, d, t\}$$

$$f(S, T) = f(s, c) + f(a, b) - f(c, a)$$
$$= 2 + 2 - 0$$

$$\boxed{f(S, T) = 4}$$

Capacity of a cut

$$\textcircled{2} \quad S = \{s, a, b, c\} \quad T = \{d, t\}$$

$$C(S, T) = C(b, d) + C(c, d) + C(b, t)$$
$$= 3 + 3 + 3$$

$$= \frac{9}{c}$$

$$\textcircled{3} \quad S = \{s, a\} \quad T = \{b, c, d, t\}$$

$$C(S, T) = C(s, c) + C(a, b)$$
$$= 2 + 2$$

$$= \underline{4}$$

For some case $f(S, T) = C(S, T)$

Theorem

For a given flow f in a flow network G , the net flow across the cut is same as

$$f(S, T) = |f|$$

The value of any flow f is a. FN G

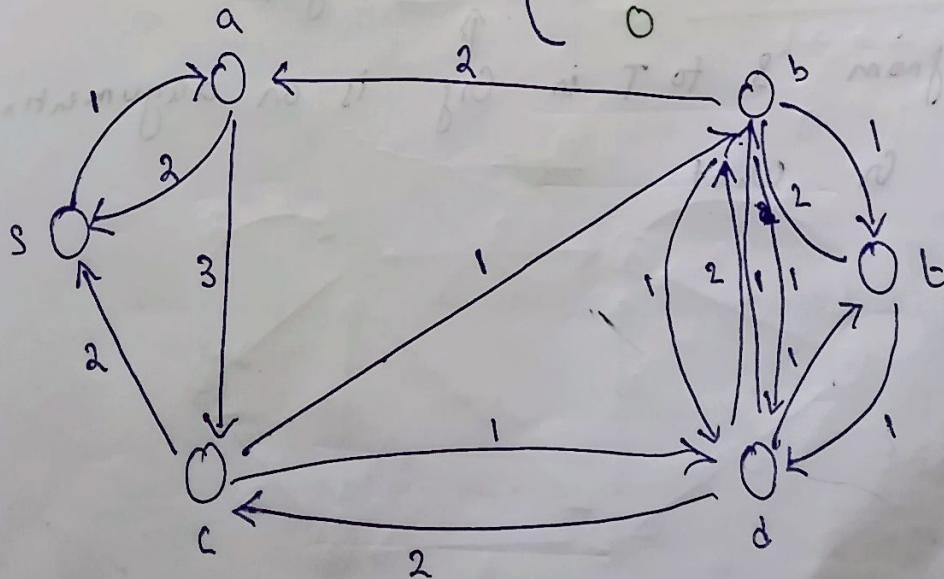
is bounded from the above theorem by the capacity on

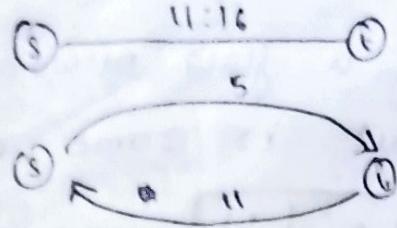
$$C(S, T) = |f|$$

Residual Networks

Let f be a flow on FN, G the residual network $G_f(V, E_f)$ is the network on graph with strictly positive residual capacity $C_f(u, v)$

$$C_f(u, v) = \begin{cases} C(u, v) - f(u, v) & \text{if } (u, v) \in E \\ f(v, u) & \text{if } (u, v) \notin E \\ (v, u) \in E \\ \text{otherwise} & \end{cases}$$





$$(i) |V| = |V|$$

$$(ii) |E_S| \leq 2|E|$$

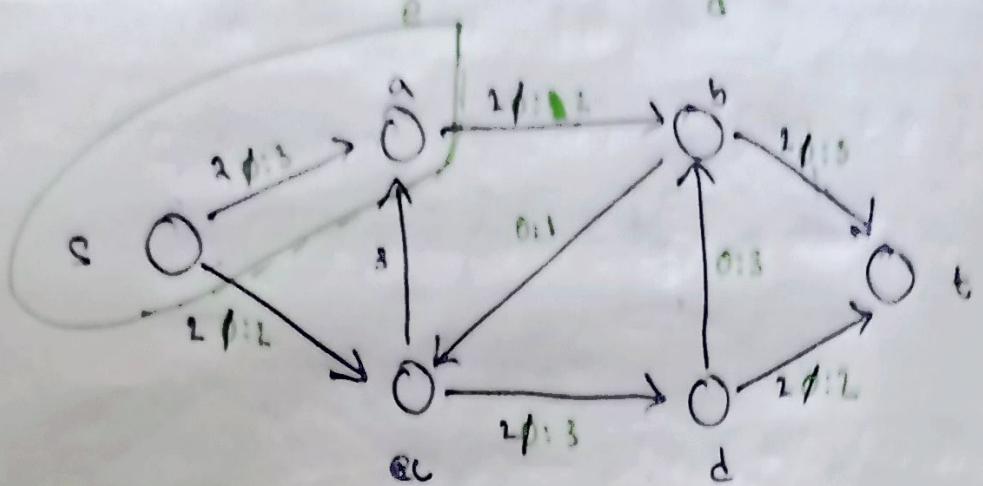
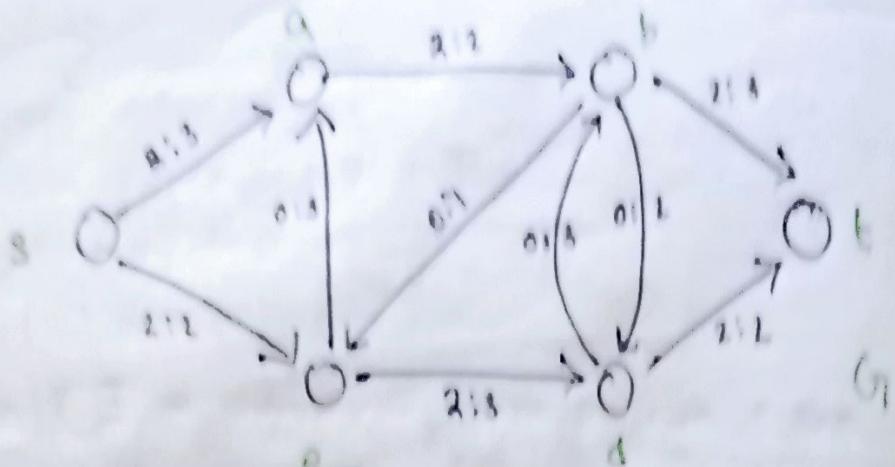
Augmenting Path (ρ)

Any path from S to T in G_f is an augmenting path (ρ) in G . W.n.t.

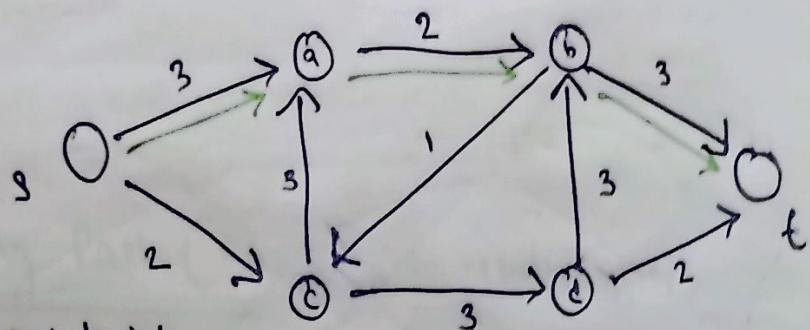
residual graph



Example



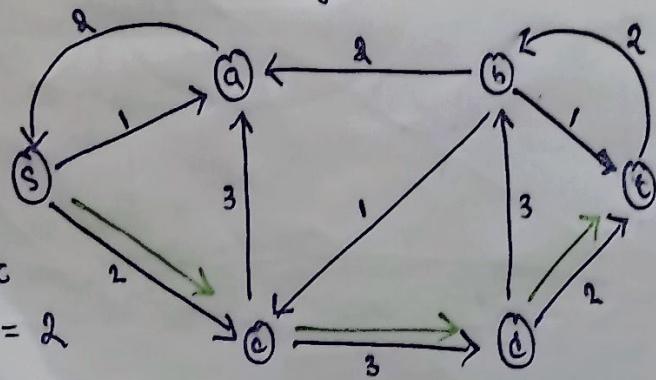
$\downarrow G_{f_1}$



augmenting path $\rightarrow s \rightarrow a \rightarrow b \rightarrow t$

bottle neck capacity = 2

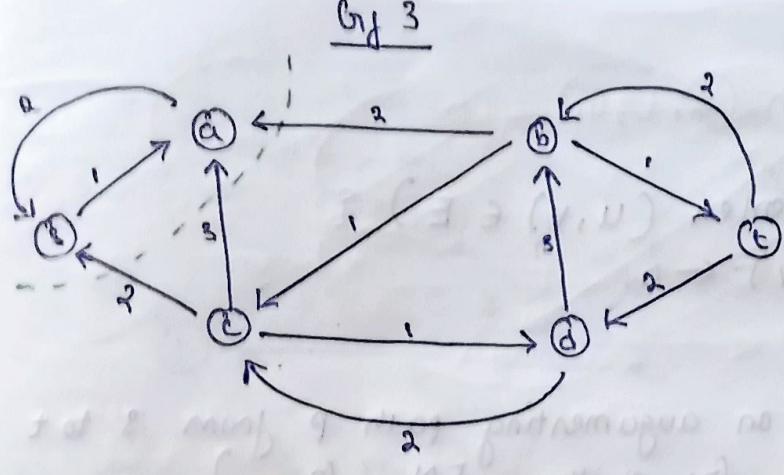
G_{f_2}



augmenting

path $\rightarrow s \rightarrow c \rightarrow d \rightarrow t$

bottle neck capacity = 2



Final Residual Graph

→ no more augmenting path.

$$\text{Maxflow} = 2 + 2 = 4$$

min cut (S, T)

$$S = \{s, a\}, T = \{b, c, d, t\}$$

$$\Leftrightarrow C(S, T) = C(s, c) + C(a, b) = 2 + 2 = 4$$

Ford-Fulkerson (s, t, G)

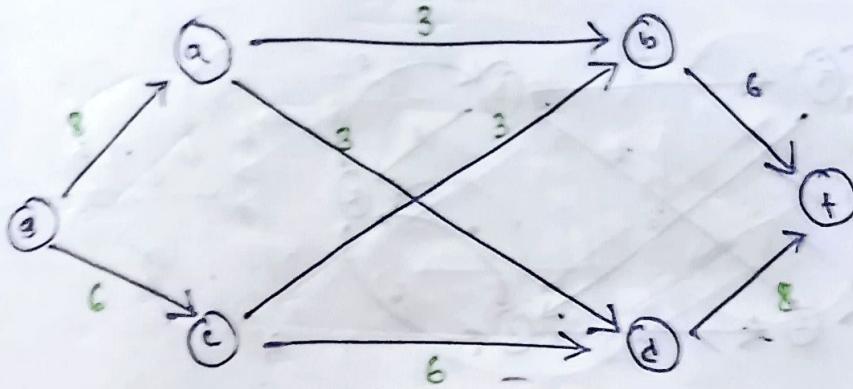
```
{  
    for (each vertex  $(u, v) \in E$ ) {  
         $f(u, v) \leftarrow 0$   
    }  
}
```

```
while ( $\exists$  an augmenting path  $P$  from  $s$  to  $t$   
      in  $G_f$  of the FN,  $G_f$ )
```

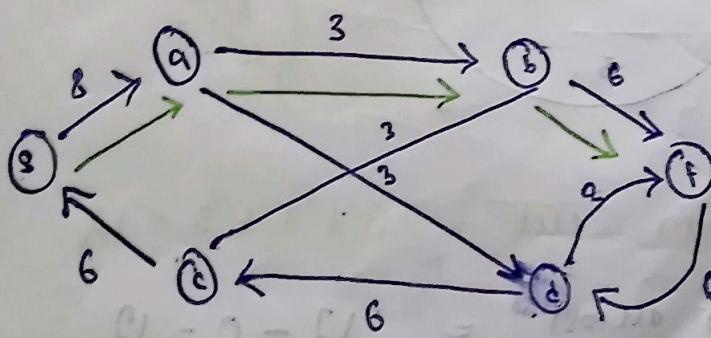
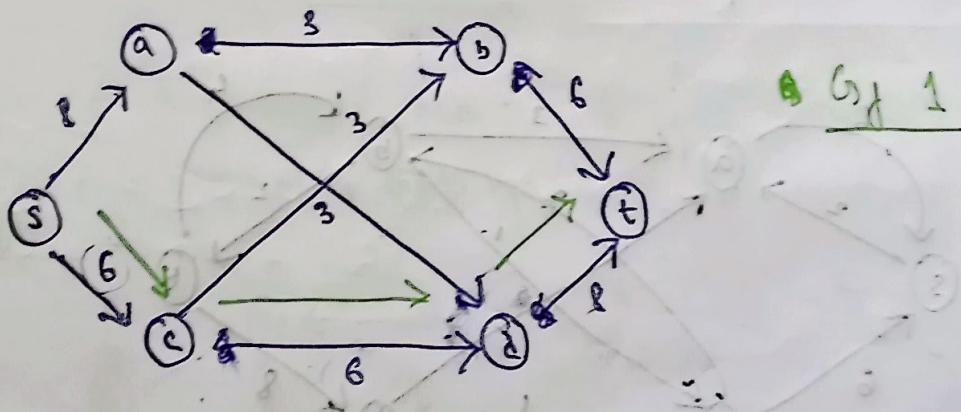
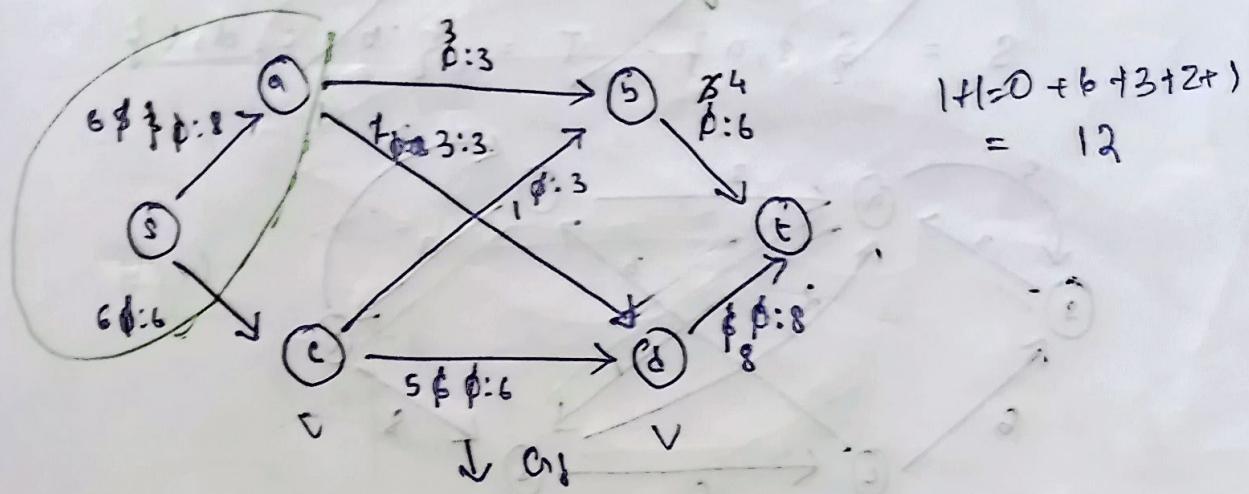
```
{  
     $c_f(P) = \min \{ c_f(u, v) ; (u, v) \in P \}$   
    for (each edge  $(u, v) \in E_f$  in P path  $P$ )
```

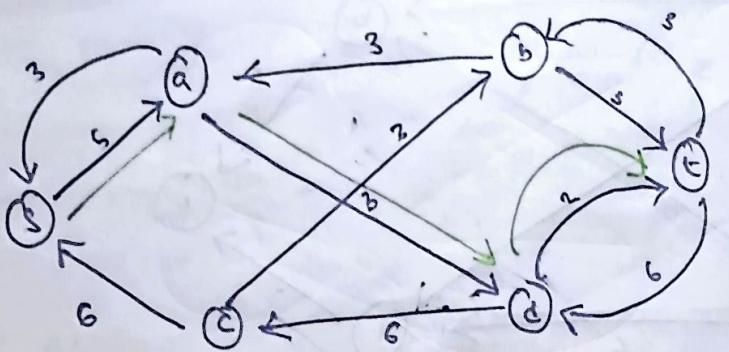
```
{  
    if  $(u, v) \in E$   
         $f(u, v) \leftarrow f(u, v) + c_f(u, v);$   
    else  
         $f(v, u) \leftarrow f(v, u) - c_f(u, v);$   
    }  
}
```

Time Complexity = $O(|f| \cdot E)$



Find the flow max flow of the flow network G by applying Ford-Fulkerson Algo.



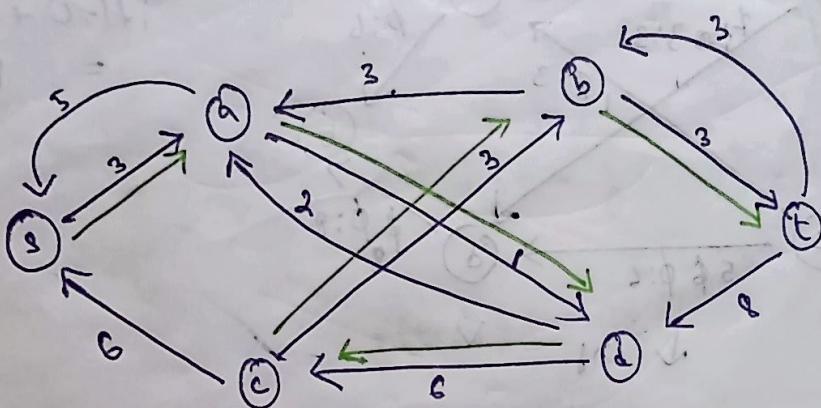


$G + 3$

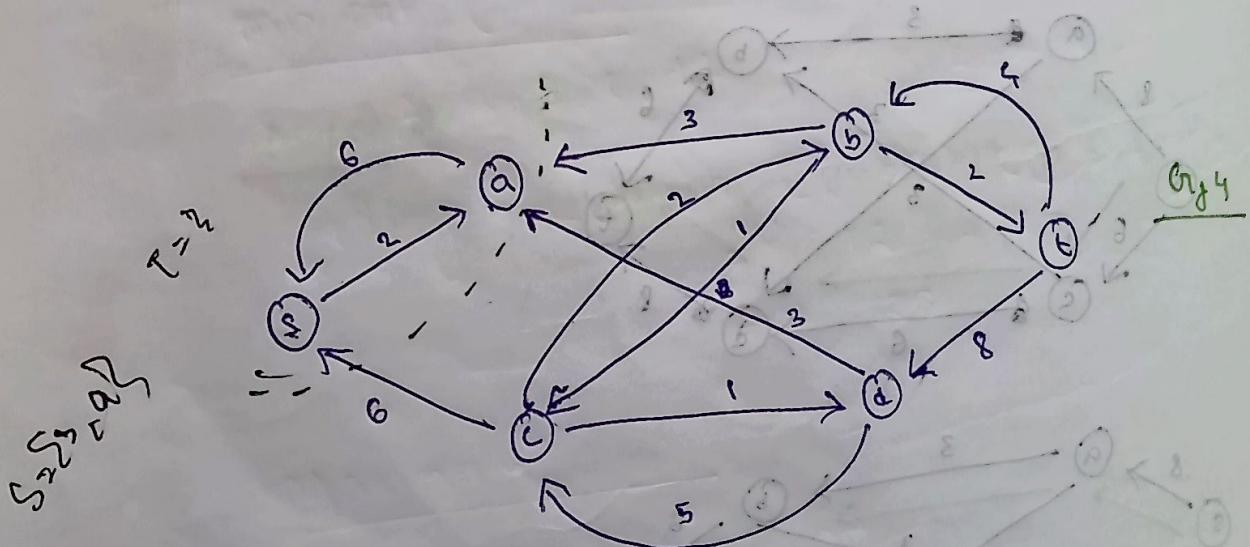
minimum cut (S, T)

$$S = \{s, a\}$$

$$T = \{b, c, d, t\}$$



G_f/g



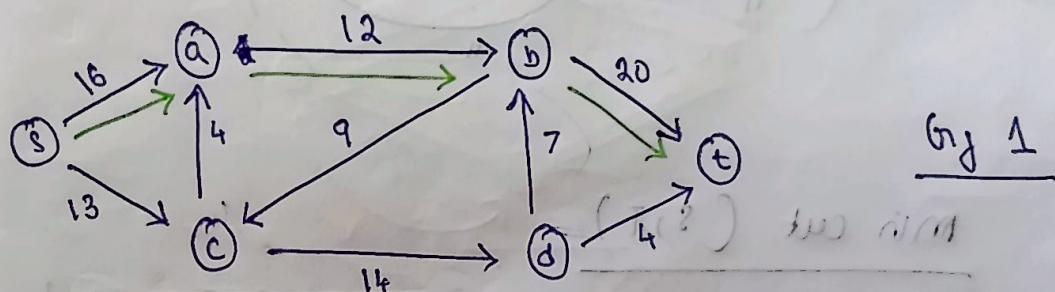
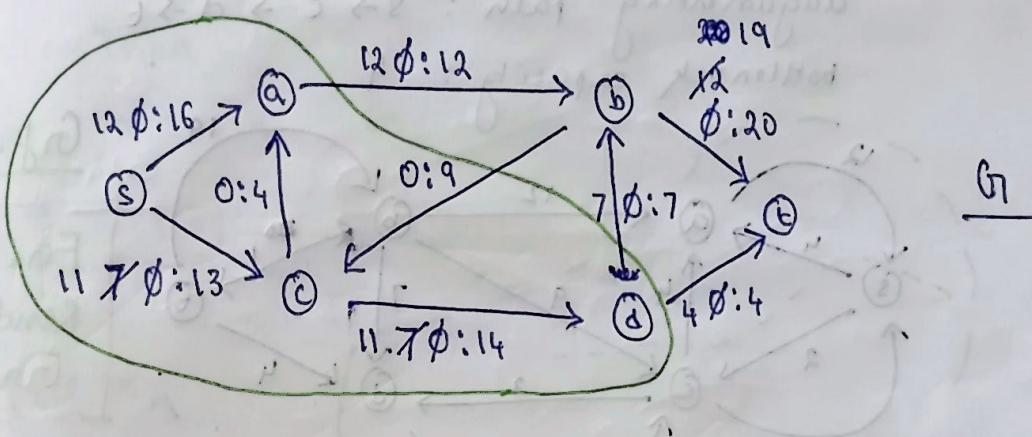
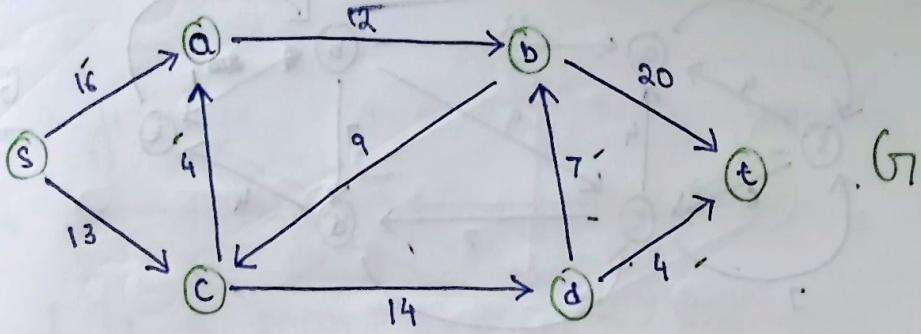
G_f/g

$S \cup \{a\} \cup \{g\}$

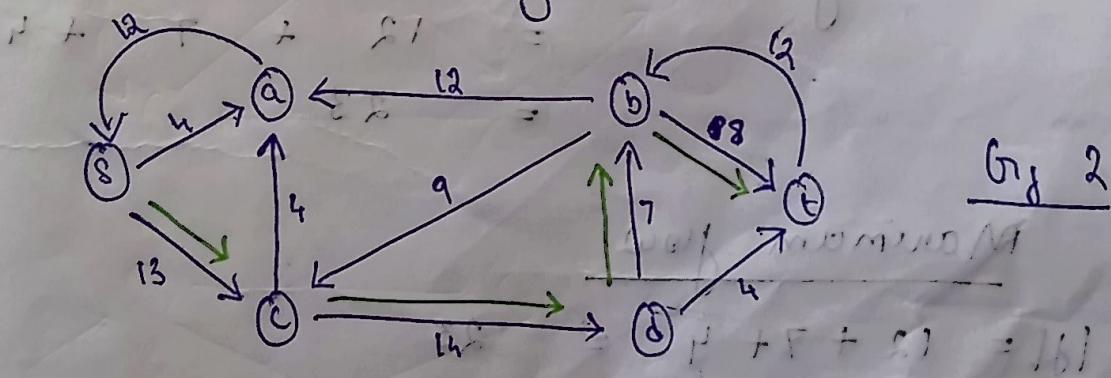
$T = \{e, f, g, t\}$

minimum cut $= 6 + 3 + 3 = 12$

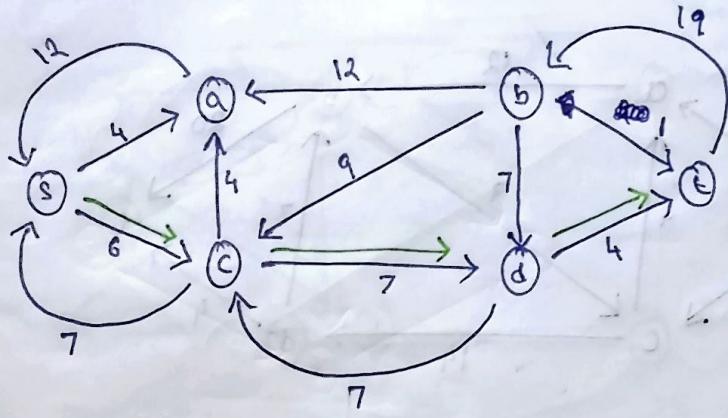
ret flow across cut $= 12 - 0 = 12$



Augmenting path: $S \rightarrow a \rightarrow b \rightarrow t$
 $(S \rightarrow a) + (a \rightarrow b)$ bottleneck capacity: $\min(12, 12) = 12$

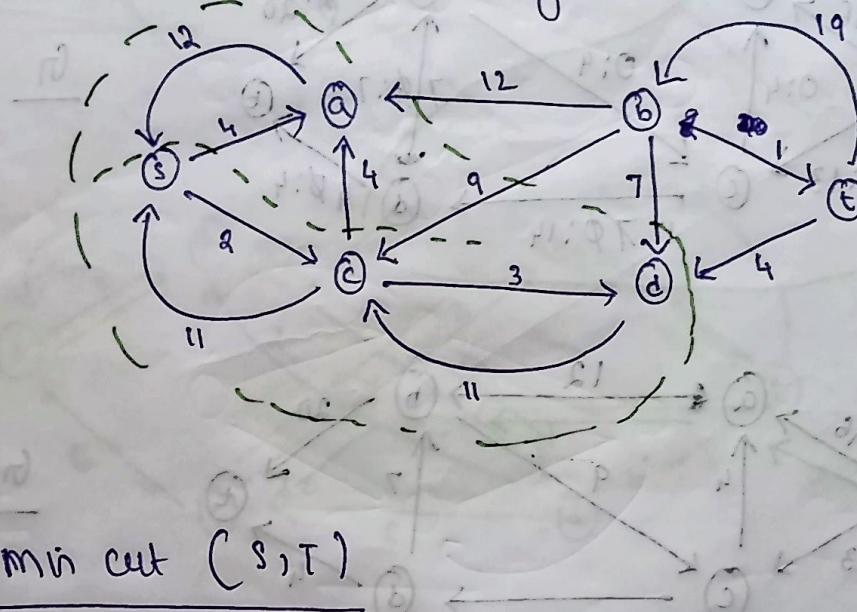


Augmenting path: $S \rightarrow c \rightarrow d \rightarrow b \rightarrow t$
 bottleneck capacity: 7



Gf3

augmenting path : $S \rightarrow C \rightarrow D \rightarrow E$
bottleneck capacity : 4



Gf4

Final
Residual
Graph

min cut (S, T)

$$S = \{a, s, c, d\} \quad T = \{b, e, t\}$$

$$\text{capacity } \Rightarrow C(S, T) = C(a, b) + C(d, b) + C(d, t)$$

$$= 12 + 7 + 4 \\ = 23$$

Maximum flow

$$|f| = 12 + 7 + 4 = 23$$

$s \leftarrow d \leftarrow b \leftarrow e \leftarrow t$: max path length
: progress permitted

Max-flow min-Cut theorem

let f is a flow in FN, $G(v, e)$ with capacity C

then the ~~given~~ following conditions are equivalent:

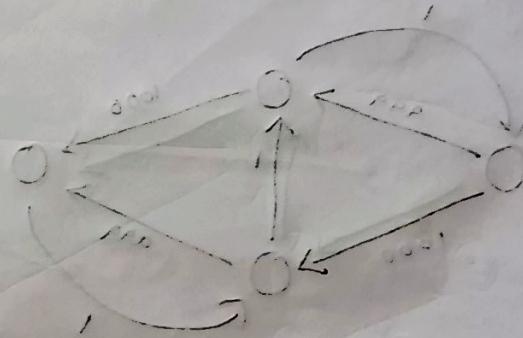
(i) f is a max flow in a FN, G

\Updownarrow Then there is no augmenting path in Residual

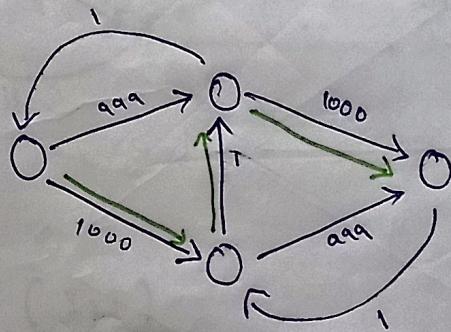
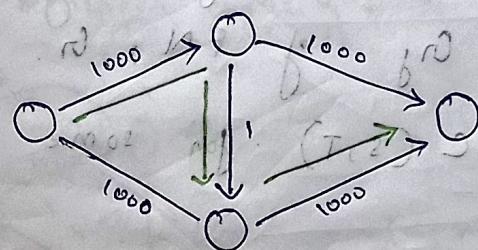
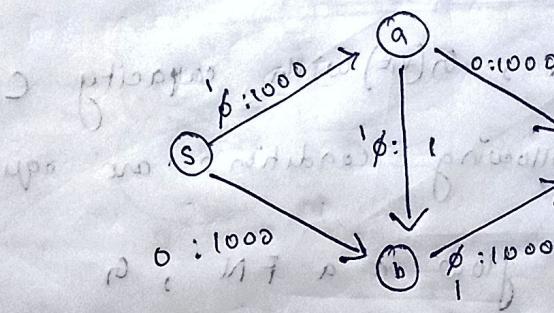
\Updownarrow network G_f of FN, G .

(ii) $|f| = C(s, T)$ for some cut (s, T) in FN G .

* PROVE



Drawback of Ford Fulkerson :



Application of Max Flow Problem

(i) Bipartite Matching

List of Job J_i

what are Rules

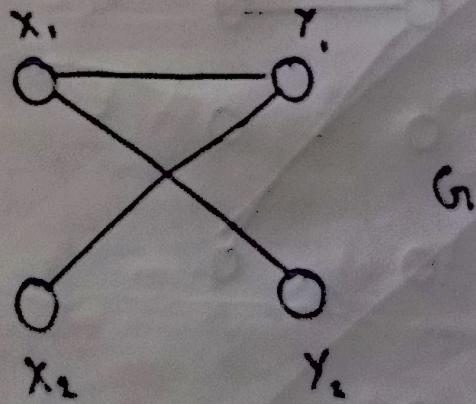
→ each candidate must be given at most 1 job and
each job must be given to one candidate

Goal

→ assign the candidate to job such that max no.
of job are filled.

Example!

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$$M_1 = \{(x_i, y_j)\}$$

Application of Max flow Problem

(i) Bipartite Matching

List of Job J_i

~~Rules~~

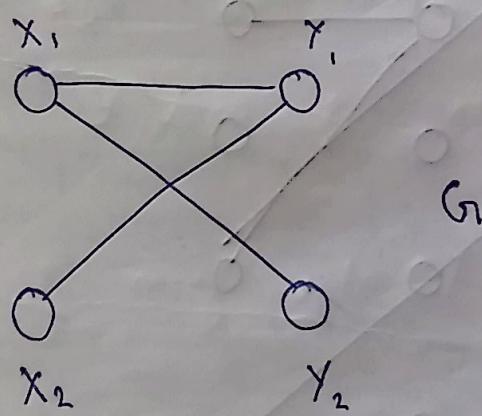
→ each candidate must be given atmost 1 job and each job must be give to one candidate

Goal

→ assign the candidate to job such that max no. of job are filled.

Example 1

13.06.22



$$M_1 = \{ (x_1, y_1), (x_1, y_2), (x_2, y_1), (x_2, y_2) \}$$

Matching

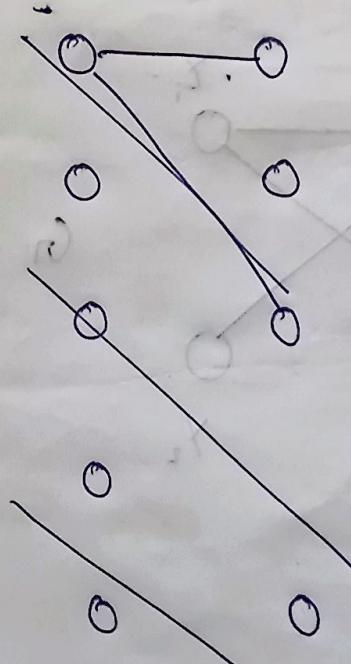
Def: A matching in Graph $G = (V, E)$ is a set of edges $M \subseteq E$ with the property that each node appears at most in one edge of M . The M is a matching.

- Perfect matching (ii) Each and every edge vertex should be connected to exactly one edge of M .

Greedy Method

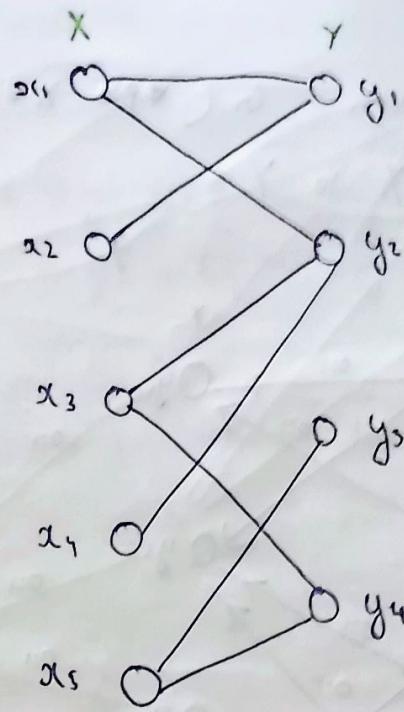
→ if ^{keep} adding edge into M , till no more edges can be added

Example 2.



$$M = \{g - e, a - c\}$$

(a, b)

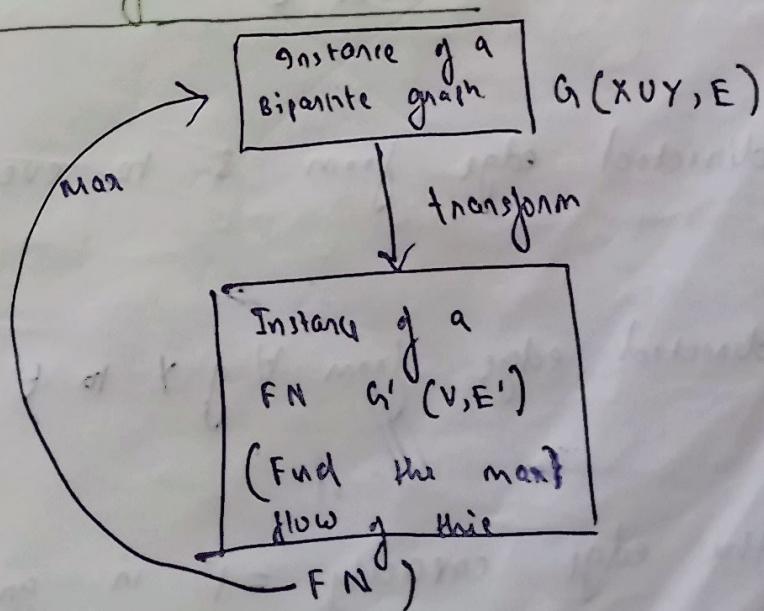


X ✗ $M = \{(x_1, y_1), (x_3, y_2), (x_5, y_3)\}$ $|M| = 3$

✓ $M = \{(x_1, y_2), (x_2, y_1), (x_3, y_1), (x_5, y_3)\}$ $|M| = 4$

✓ $M = \{(x_4, y_2), (x_1, y_1), (x_3, y_2), (x_5, y_3)\}$ $|M| = 4$

③ Maximum flow problem

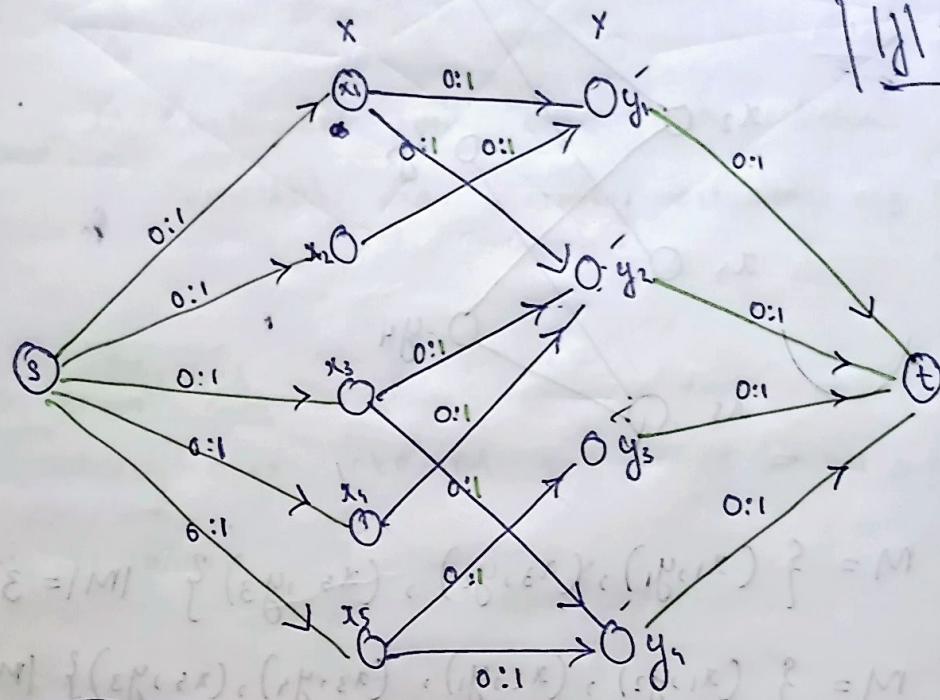


Step 1

Given Bipartite Graph $G(V \times U, E)$ direct
the edge from X to Y

$$|X| = 4 = |U|$$

$$|Y| = |V|$$



$$P_1: s \rightarrow x_2 \rightarrow y_1 \rightarrow t$$

$$P_2: s \rightarrow x_4 \rightarrow y_2 \rightarrow t$$

Step 2

Add to ~~the~~ vertex s and t

Step 3

Add a directed edge from s to every vertex in Y

Step 4

Add a directed edge from t of Y to t

Step 5

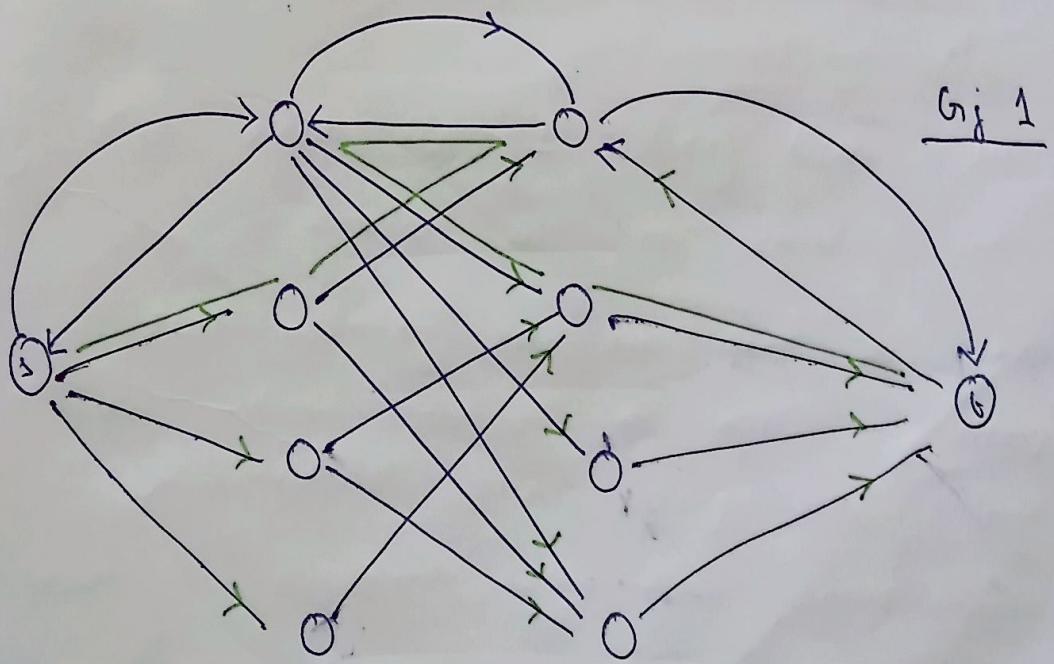
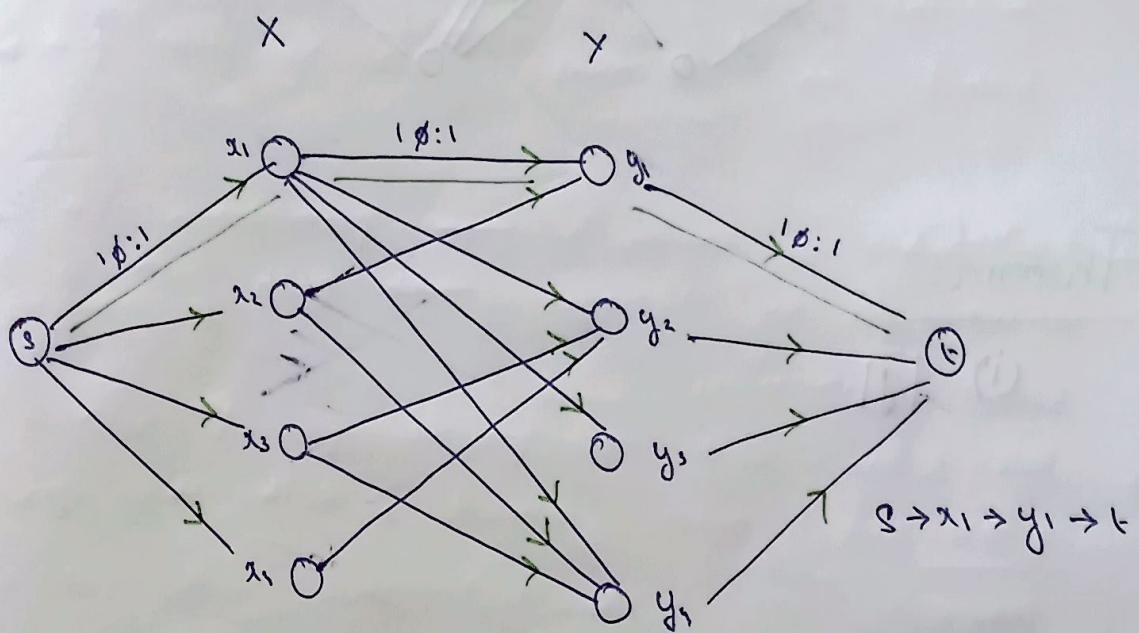
Set all the edge capacity = 1 in graph made $G'(V', E')$

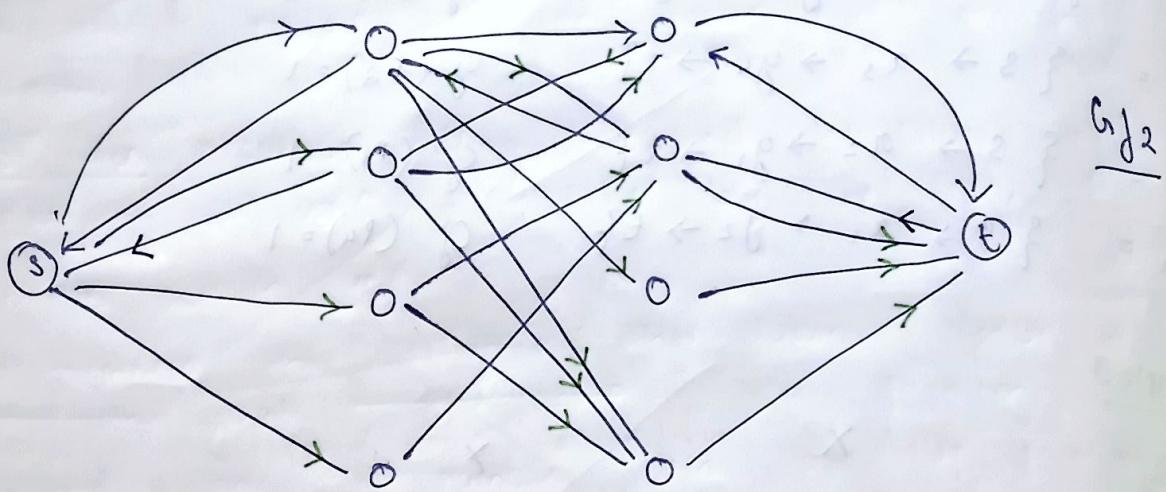
Step 6

Apply Ford - Fulkerson method

$$\begin{aligned}
 p_1 &= \{ s \rightarrow x_1 \rightarrow y_1 \rightarrow t \} & c_f(p_1) &= 1 \\
 p_2 &= \{ s \rightarrow x_2 \rightarrow y_2 \rightarrow t \} & c_f(p_2) &= 1 \\
 p_3 &= \{ s \rightarrow x_3 \rightarrow y_3 \rightarrow t \} & c_f(p_3) &= 1 \\
 p_4 &= \{ s \rightarrow x_4 \rightarrow y_4 \rightarrow t \} & c_f(p_4) &= 1
 \end{aligned}$$

Example 3



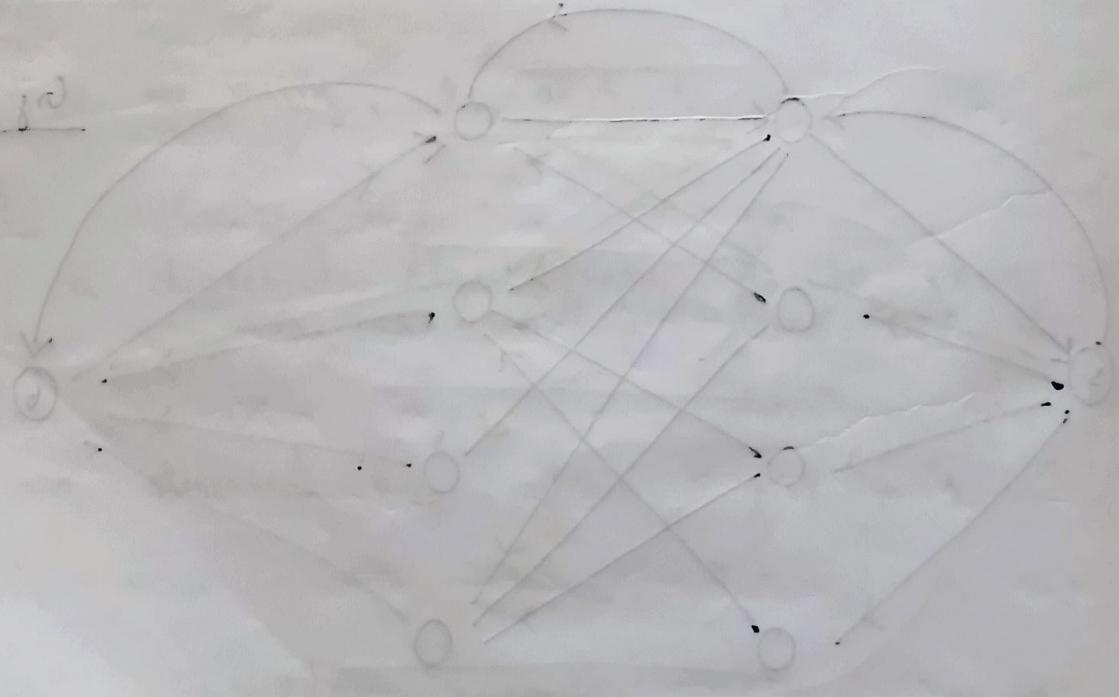


Theorem

181. (i)

\leftarrow \rightarrow \leftarrow \rightarrow

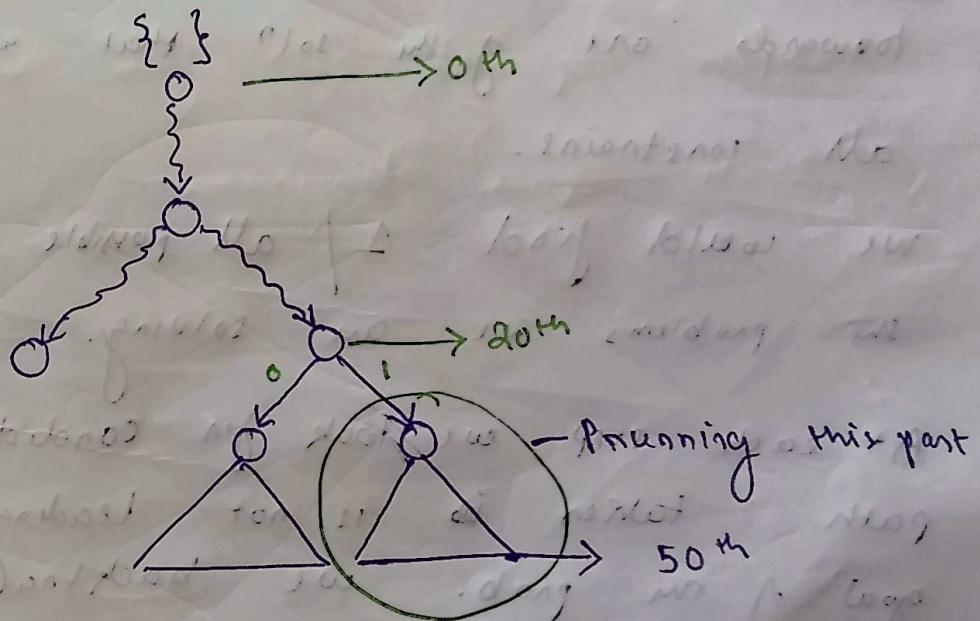
\leftarrow \rightarrow



Backtracking

- (i) Introduction
- (ii) Control Abstraction.
- (iii) General Algo for Backtracking , Method
- (iv) Constructing All subsets of a given set.
- (v) Constructing / All permutation of a given set
- (vi) Constructing all paths in a graph G.
- (vii) X Subset sum } by using backtracking.
- (viii) X N-Queen }
 {
- (ix) X Sudoku }

Introduction



- Backtracking is an algo. design technique for listing all possible solⁿ for combinational searching problem.
- It uses recursive calling to find the solution by building a solⁿ step by step increasing values with time.
- It removes the solution that does not give rise to the goal ~~of the~~ of the problem based on constraints given to solve problem.
- Backtracking is similar to prior permutation n combination where we try different ~~path~~ path for solⁿ of the same problem.
- We start with a possible partial solⁿ of the prob. and move ahead, with this approach towards one of the solⁿ that will satisfy all constraints.
- We would find 1/ all possible solⁿ's for the problems we are solving.
- At each step - we look for candidate if the path taken ~~is~~ is not leading us to goal of the prob. we backtrack 1 level back and start with new candidate if that level again doesn't lead to correct solⁿ we backtrack to 1 level further up till we reach the root, we say that solⁿ is not available if combination with said constraints.

→ we find a potential candidate which will lead us to goal of problem which will become part of partial soln and that will be used as a part of final soln.

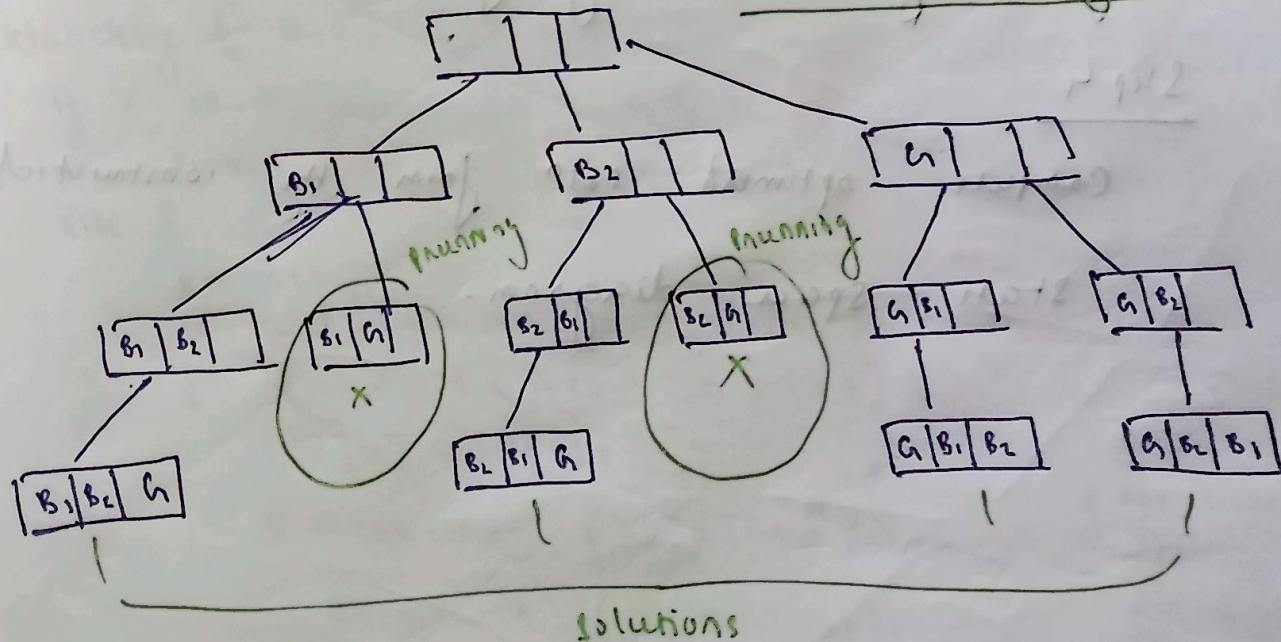
Problem 2

(B_1, B_2, G)

we want to find all combination 2 boys, 1 girl in 3 seat

constraints :- girls should not be in middle of 2 boys

State Space Diagram



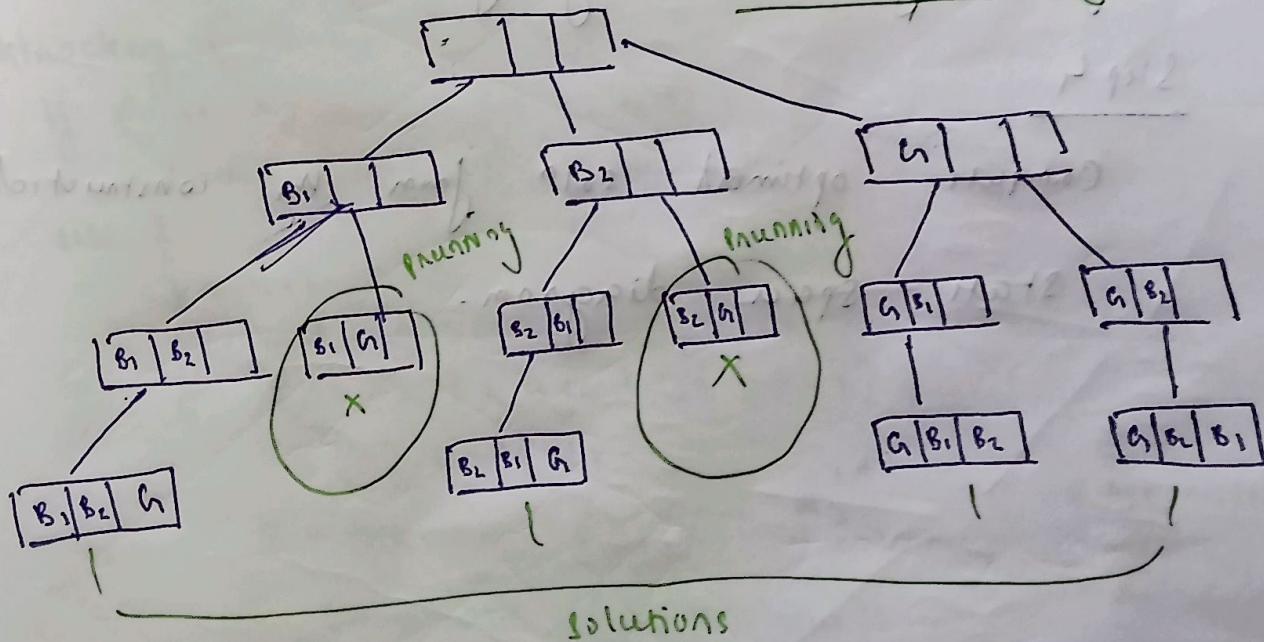
→ we find a potential candidate which will lead us to goal of problem which will become part of partial soln and that will be used as a part of final soln.

Problem 2 (B_1, B_2, G)

we want to find all combination 2 boys & 1 girl in 3 seat

constraints :- girls should not be in middle of 2 boys

State Space Diagram



Strategy

Step 1

Identify all the constraints (implicit, explicit) of given problem

Step 2

define an appropriate "bounding func" based on given constraints.

Step 3

Construct a state space diagram on them by using all bounding func

Step 4

Compute optimal sol from the constructed

state space diagram.

had best

```
Backtracking( a, k ) {  
    if ( a is a feasible sol'n g m prob.) then  
        print( a );  
    else {  
        K ← K+1  
        compute  $s_K$   
        while (  $s_K \neq q$  ) {  
             $s_K \leftarrow s_K - a_K$  ;  
        }  
        Backtracking( a, k );  
    }  
}
```

```
Backtracking( a, k ) {  
    if ( is-a-solution( a, k, input ) )  
        problem-solution( a, k );  
    else {  
        K ← K+1  
        construct-candidate( a, k, input, c, n_candidate );  
        for ( i < 0, i < n_candidate, i++ ) {  
            a[ k ] = c[ i ];  
            # make-move( a, k, input );  
            back-track( a, k, input );  
            # unmark-move( a, k, input );  
        }  
        if ( finished )  
            return true;  
    }  
}
```

Construct All Subsets

How many subsets are there of an 'n' element set
to construct all 2^n subsets of array size n /
vector of size n.

when the value of $a[i]$ is either true / false
it signifies whether i^{th} element is or is not in the
subset.

what order will this generate the subsets

boolean Is_A_Solution (int a[], int k, int input) {

if ($k == \text{input}$)
then return true;

void Process_Solution (int a[]; int k) {

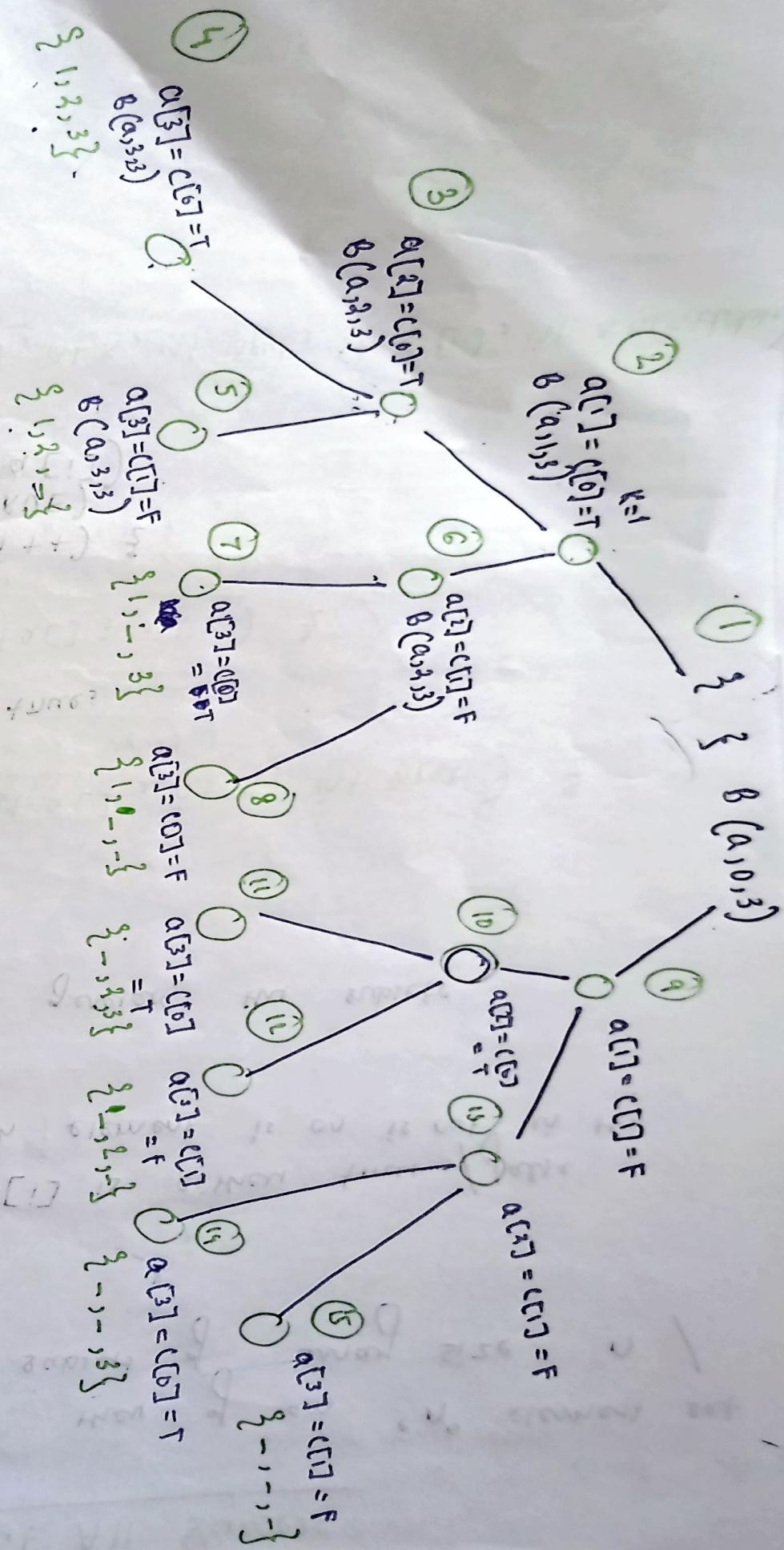
int i;
for ($i \leftarrow 1$; $i \leq k$; $i++$) {
if ($a[i] == \text{TRUE}$)
printf ("a[%d]",

Construct_Candidate (int a[], int k, int input, int c[], int *nCandidate)

$c[0] \leftarrow \text{TRUE}$

$c[1] \leftarrow \text{FALSE}$

*nCandidate $\leftarrow 2$



void PS - A - Solution (int a[], int k, int n)
if ($k = n$) then return T;
{

void process - solution (int a[], int k) {
 int i;
 for (i = 1; i $\leq k$; i++)
 printf ("%d", a[i]);

void construct - candidate (int a[], int k, int n, int c[], int *nCandidate)
{
 bool Pe - sol [MAX];
 for i
 for (i = 1; i $\leq n$; i++) {
 Pe - sol [a[i]] \leftarrow T
 }
 *nCandidate \leftarrow 0
 for (i = 1; i $\leq n$; i++) {
 if (Pe - sol [i] == F) {
 c [*nCandidate] \leftarrow i
 *nCandidate ++
 }
 }
}