

Backtracking, Branch & Bound

Module 2

Sem IV

AoA Even 2021-22

Introduction

- Backtracking and Branch and Bound are two graph based methods for design of algorithms.
- In both cases we explore a search tree.
- In Backtracking, we start with a node and explore the nodes in **Depth First manner**.
- All the nodes need not to be explored. Cut the branches of the tree based on the constraint of the problem. This reduces the time complexity of the algorithm.
- Branch and Bound explores the search tree in a **Breadth First** manner.

Backtracking : Introduction

- It is modified form of Depth First Search.
- Here solution vector is of form $x_1, x_2, x_3, \dots, x_n$, n tuple $(x_1, x_2, x_3, \dots, x_n)$, where x_i is chosen from finite set of S_i , such that constraint of the problem is satisfied.
- Backtracking algorithm solves the problem using two types of constraints:
 1. Explicit Constraint
 2. Implicit Constraint

Backtracking : Introduction

Terminologies Used:

Backtracking algorithms determine problem solutions by systematically searching for solution using tree structure.

1. **State space tree:** The solution space is organized as a tree called the state space tree.
2. **Explicit constraint:** These are the rules that restrict each component x_i of the solution vector to take values only from a given set S .
3. **Implicit constraint:** These are the rules that describe the way in which the x_i 's must relate to each other or which of the components of the solution vector satisfy the criteria function.
4. **Solution space:** It is the set of all tuples that satisfy the explicit constraints.
5. **Live node:** It is the node that has been generated, but none of its descendants are yet generated.
6. **Bounding function or criteria:** It is a function created that is used to kill live nodes without generating all its children.

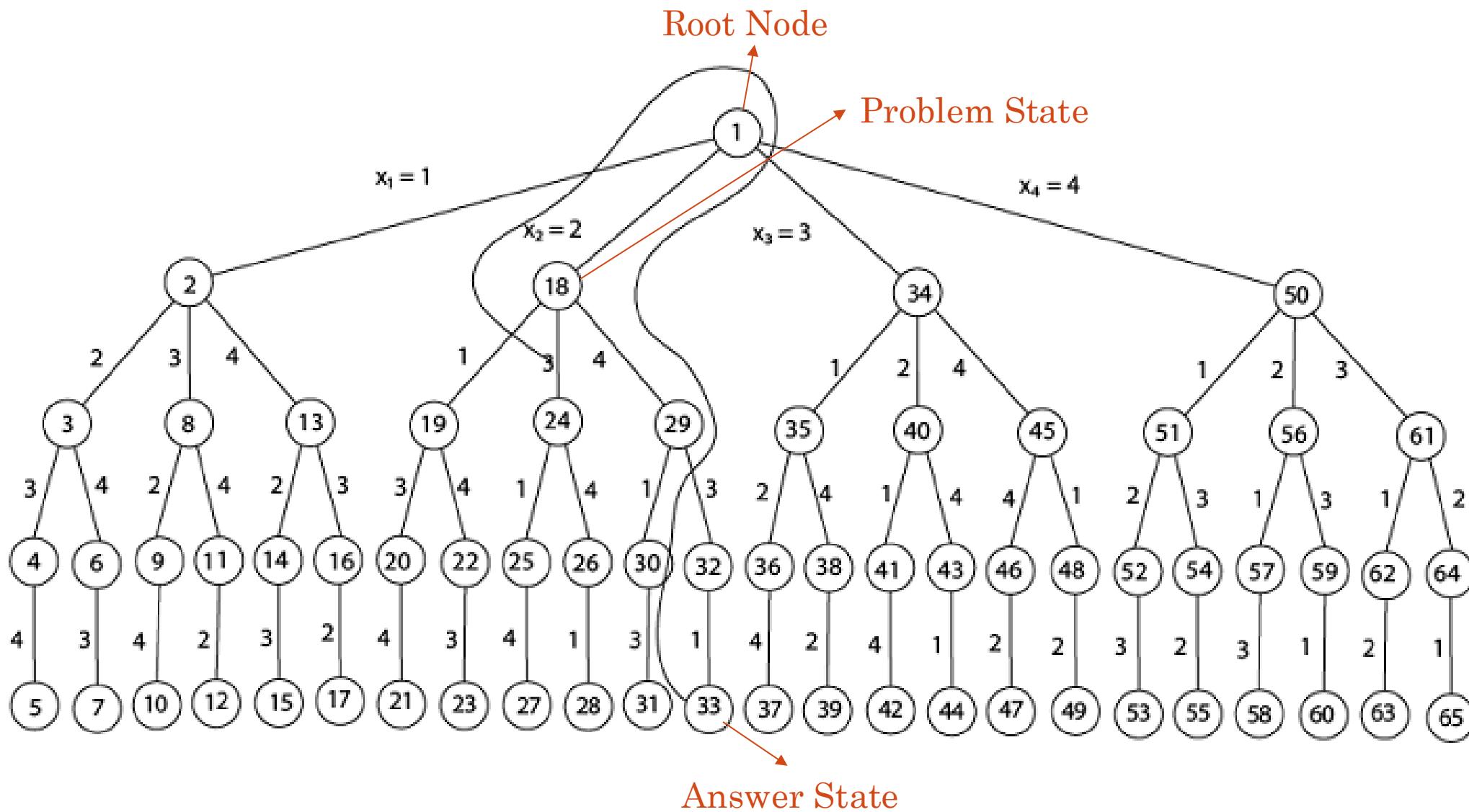
Backtracking : Introduction

Terminologies Used:

Backtracking algorithms determine problem solutions by systematically searching for solution using tree structure.

7. **Extended node or E-node:** It is the live node whose children are currently being generated.
8. **Dead node:** It is the node that is not to be extended further or all of whose children have already been generated.
9. **Answer node:** It is the node that represents the answer of the problem that means the node at which the criteria functions are maximized, minimized or satisfied.
10. **Solution node:** It is the node that has the possibility to become the answer node.

Backtracking : Introduction



Backtracking : N-Queen Problem

1. 2-Queen problem
2. 3-Queen Problem
3. 4-Queen Problem
4. 8-Queen problem

2-Queen problem:

Q	X
X	X

X	Q
X	X

Therefore, No Solution

Backtracking : N-Queen Problem

1. 2-Queen problem
2. 3-Queen Problem
3. 4-Queen Problem
4. 8-Queen problem

2-Queen problem:

Q	X
X	X

X	Q
X	X

Therefore, No Solution

Similarly, No Solution for 3-Queen problem

Backtracking : N-Queen Problem

4 Queen Problem:

State space Tree:

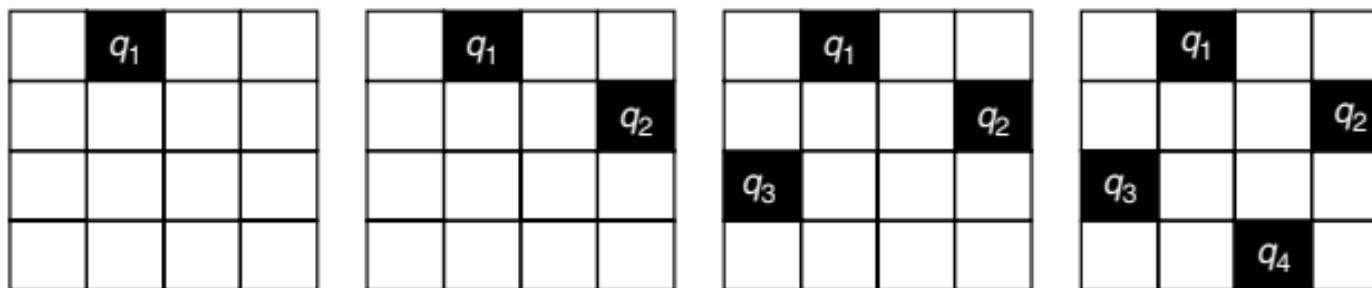
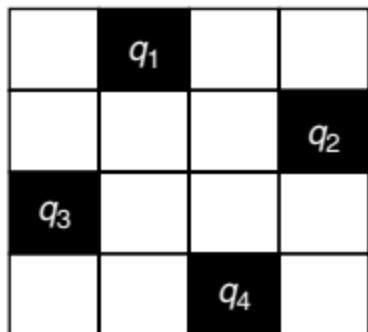


Figure 2 Solution of four-queens problem.

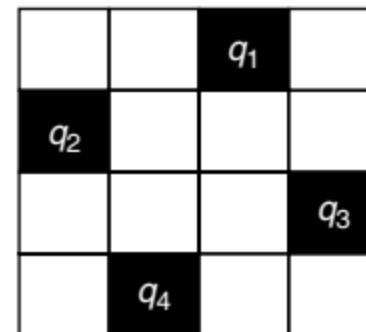
Backtracking : N-Queen Problem

4 Queen Problem:

Solution can be represented as four-tuple (x_1, x_2, x_3, x_4) where x_1 is column value in row 1 for placement of Q_1 and so on.



Solution 2, 4, 1, 3



Solution 3, 1, 4, 2

Figure 7 Two possible solutions to four-queens problem.

Backtracking : N-Queen Problem

N Queen Problem:

```
1  Algorithm NQueens( $k, n$ )
2  // Using backtracking, this procedure prints all
3  // possible placements of  $n$  queens on an  $n \times n$ 
4  // chessboard so that they are nonattacking.
5  {
6      for  $i := 1$  to  $n$  do
7          {
8              if Place( $k, i$ ) then
9                  {
10                      $x[k] := i;$ 
11                     if ( $k = n$ ) then write ( $x[1 : n]$ );
12                     else NQueens( $k + 1, n$ );
13                 }
14             }
15 }
```

Algorithm 7.5 All solutions to the n -queens problem

Backtracking : N-Queen Problem

N Queen Problem:

```
1  Algorithm Place( $k, i$ )
2  // Returns true if a queen can be placed in  $k$ th row and
3  //  $i$ th column. Otherwise it returns false.  $x[ ]$  is a
4  // global array whose first  $(k - 1)$  values have been set.
5  //  $\text{Abs}(r)$  returns the absolute value of  $r$ .
6  {
7      for  $j := 1$  to  $k - 1$  do
8          if  $((x[j] = i) \text{ // Two in the same column}$ 
9              or  $(\text{Abs}(x[j] - i) = \text{Abs}(j - k)))$ 
10             // or in the same diagonal
11             then return false;
12     return true;
13 }
```

Algorithm 7.4 Can a new queen be placed?

Backtracking : N-Queen Problem

N Queen Problem:

- (4,1), (5,2), (6,3), (7,4), (8,5)

Let (i, j) and (k,l) be two cells in the chessboard.

If $i-j = k-l$

e.g. $4-1 = 6-3 = 3$

Rearranging above equation, we have

$$i-k = j-l \Rightarrow |i - k| = |j - l|$$

- (5,8), (6,7), (7,6), (8,5)

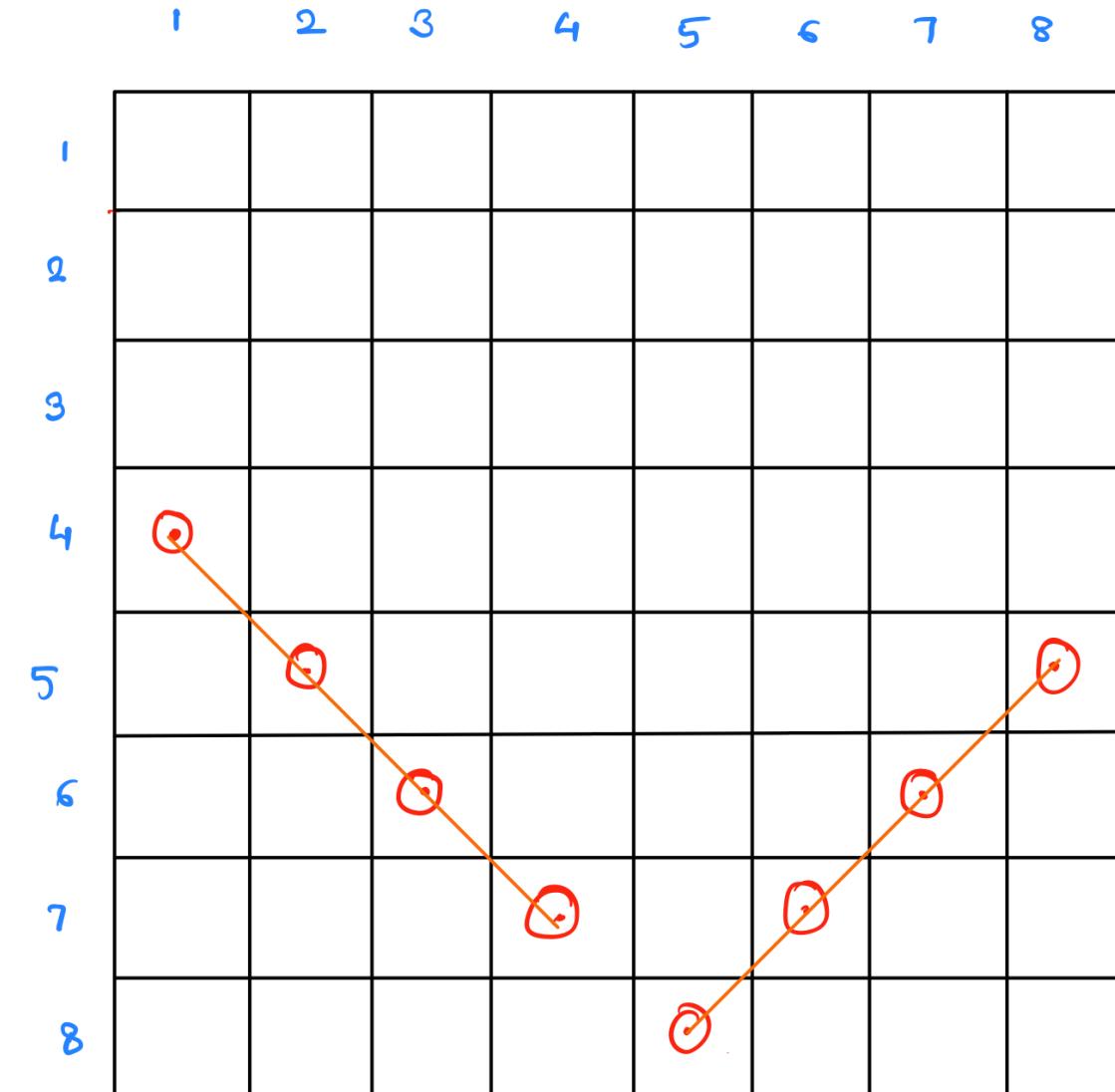
Again let (i, j) and (k,l) be two cells in the chessboard.

If $i+j = k+l$

e.g. $5+8 = 6+7 = 13$

Rearranging above equation, we have

$$i-k = l-j \Rightarrow |i - k| = |j - l|$$



Backtracking : Sum of Subsets

Given:

1. n distinct positive numbers (called weights w_i), where $1 \leq i \leq n$
2. Sum (m)

We need to find all possible subsets of given numbers (w_i) having sum equal to the target Sum (m).

Backtracking : Sum of Subsets

Example:

$$n=4; (w_1, w_2, w_3, w_4) = (11, 13, 24, 7) \text{ & } m=31$$

Desired subsets are (11,13,7) & (24,7)

Solution vector (1, 2, 4) & (3, 4) → [Variable Length]

In general, all solution vectors are k-tuples, $(x_1, x_2, x_3, x_4); 1 \leq k \leq n$

Implicit Constraints:

1. No two subsets should be same & sum of corresponding w_i 's be m
2. $x_i < x_{i+1}$ such that $1 \leq i \leq k$, to avoid generating multiple instances of same subset e.g. (1,2,4) & (1,4,2)

Backtracking : Sum of Subsets

Example:

$$n=4; (w_1, w_2, w_3, w_4) = (11, 13, 24, 7) \text{ & } m=31$$

Another Approach: [Fixed Length]

Each solution set is represented by n-tuple (x_1, x_2, x_3, x_4) such that

$x_i \in \{0,1\}$ where $1 \leq i \leq n$

$x_i = 0 \rightarrow w_i$ not selected, and $x_i = 1 \rightarrow w_i$ selected

Therefore, Solution space of above instance are $(1,1,0,1)$ & $(0,0,1,1)$

Backtracking : Sum of Subsets

Example:

$$n=4; (w_1, w_2, w_3, w_4) = (11, 13, 24, 7) \text{ & } m=31$$

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Therefore, Solution space of above instance are $(1,1,0,1)$ & $(0,0,1,1)$

Backtracking : Sum of Subsets

Example:

$$S = \{2, 3, 5, 6, 7, 9, 10\} \text{ & } M= 15$$

In state space tree of solution, node list values of sumSoFar, k & remWeight

Initialize root node with values, sumSoFar = 0, k= 1 & remWeight= 42

Possible solutions:

1. {2,3,10}
2. {2,6,7}
3. {3,5,7}
4. {5, 10}
5. {6,9}

0	1	42
---	---	----

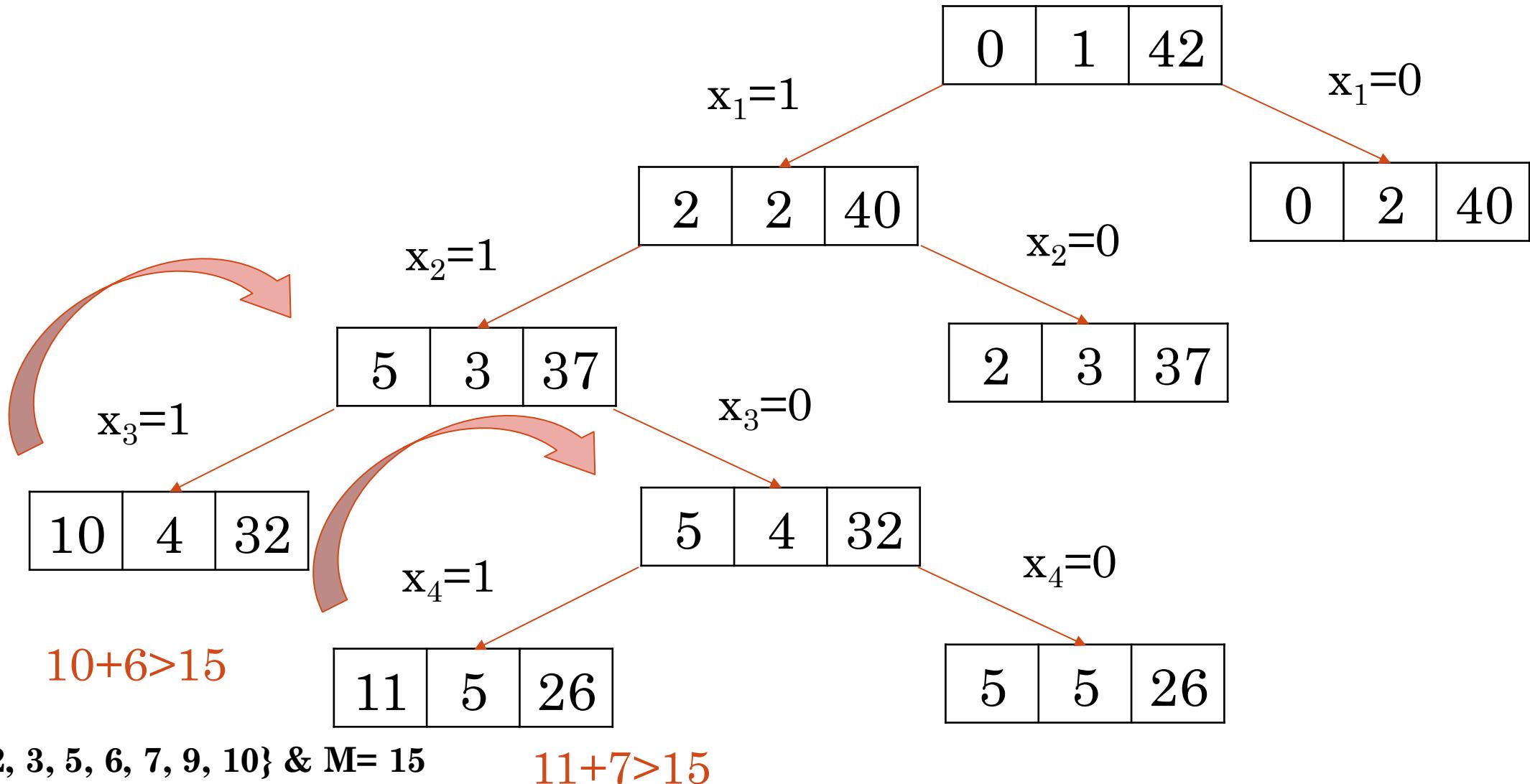
sumSoFar

k

remWeight

Backtracking : Sum of Subsets

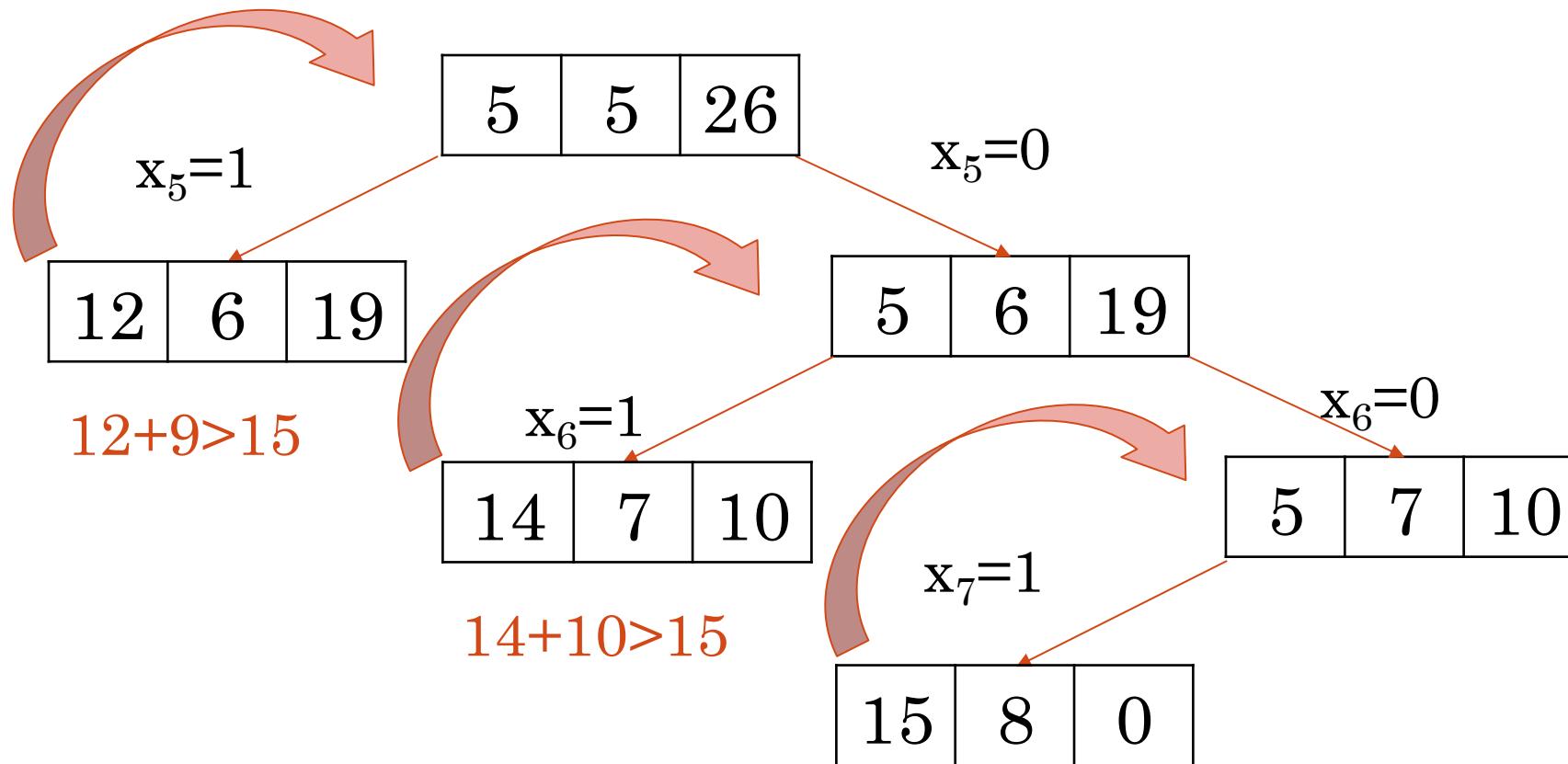
Example:



$$S = \{2, 3, 5, 6, 7, 9, 10\} \text{ & } M= 15$$

$$11+7>15$$

Backtracking : Sum of Subsets



$$S = \{2, 3, 5, 6, 7, 9, 10\} \text{ & } M = 15$$

Backtracking : Sum of Subset

Algorithm 4 SUMOFSUBSETS (Sumsofar, k, remweight)

This algorithm is used to find all the solutions of the sum of subsets problem. The X [] is the solution vector.

1. Set $X[k]=1$
2. if $(\text{Sumsofar}+w[k]=M)$ then
print $X[1..k]$

// solution is found
else
 if $(\text{Sumsofar}+w[k]+w[k+1] \leq M)$ then
 // Generate Left child
 SUMOFSUBSETS(Sumsofar+w[k], k+1, remweight-w[k])
 Endif
3. Endif
4. if $((\text{Sumsofar}+\text{remweight}-w[k] \geq M) \text{ and } (\text{Sumsofar}+w[k+1] \geq M))$ then
 // Generate Right child
 {
 $X[k]=0$
 SUMOFSUBSETS(Sumsofar, k+1, remweight-w[k])
 }
5. Stop

Backtracking : Graph Coloring

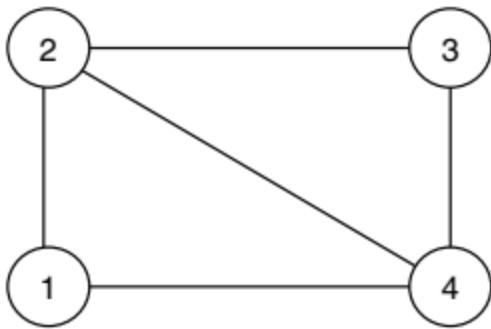
- It's a classic combinatorial Problem
- It's a problem of coloring N vertices of a given graph G in such a way that no two adjacent vertices share the same color and yet M colors are used.
- The problem is called as M coloring problem.
- M coloring Decision problem: M is given, whether graph can be colored using M colors
- M coloring optimization problem: smallest number of colors (M) required to color the graph.

Backtracking : Graph Coloring Algorithm

- Suppose we have graph $G=(V,E)$ with N vertices and M is given number of colors.
- We represent Graph G by adjacency matrix $G[n,n]$ where,
 - $G[i,j]=1$ if (i,j) is an edge of G and
 - $G[i,j]=0$ otherwise.
- If d is degree of given graph, then it can be colored with $d+1$ colors [m is referred to as **chromatic number**].
- Here colors are represented as integers $1,2,3,\dots,M$ and coloring solution will be a vector $x[1\dots N]$.
- So, solutions are given by n -tuple $(x_1, x_2, x_3, \dots, x_n)$ where, $1 \leq x_i \leq M$ and $1 \leq i \leq N$ and x_i is color of node i .

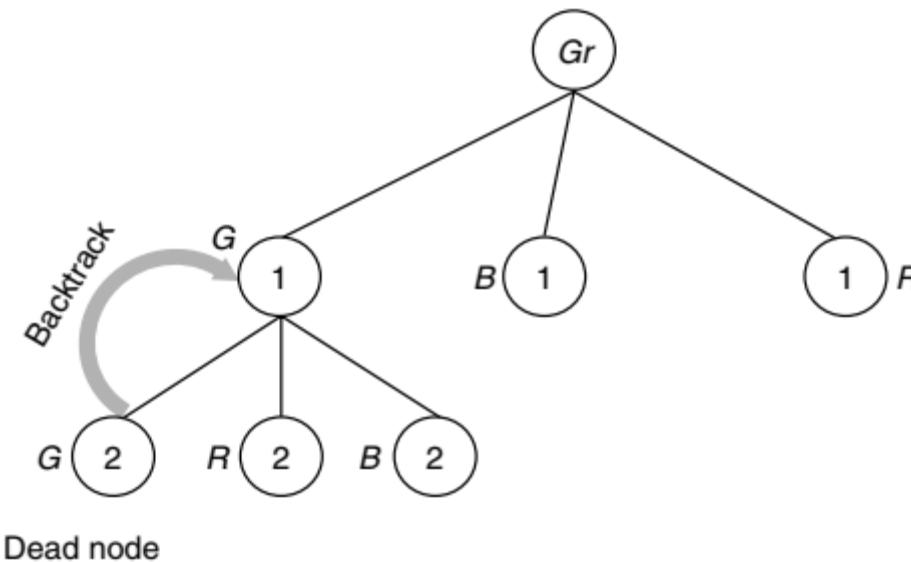
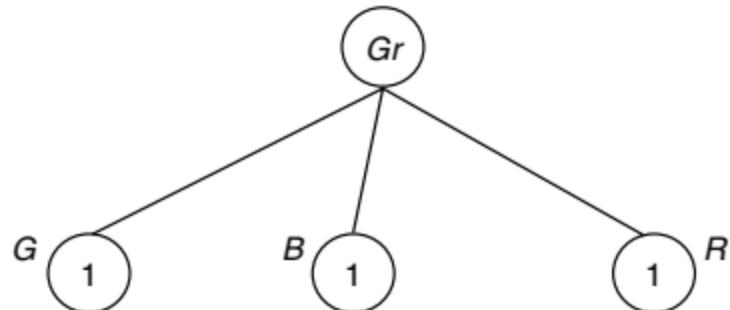
Backtracking : Graph Coloring Example

Example 1:

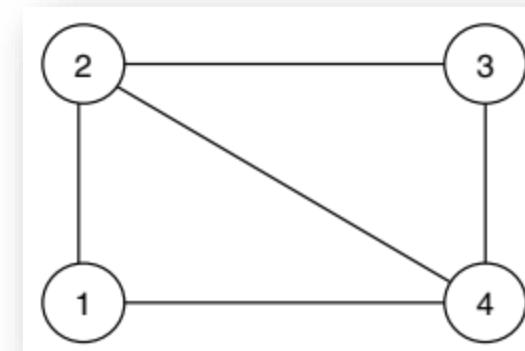
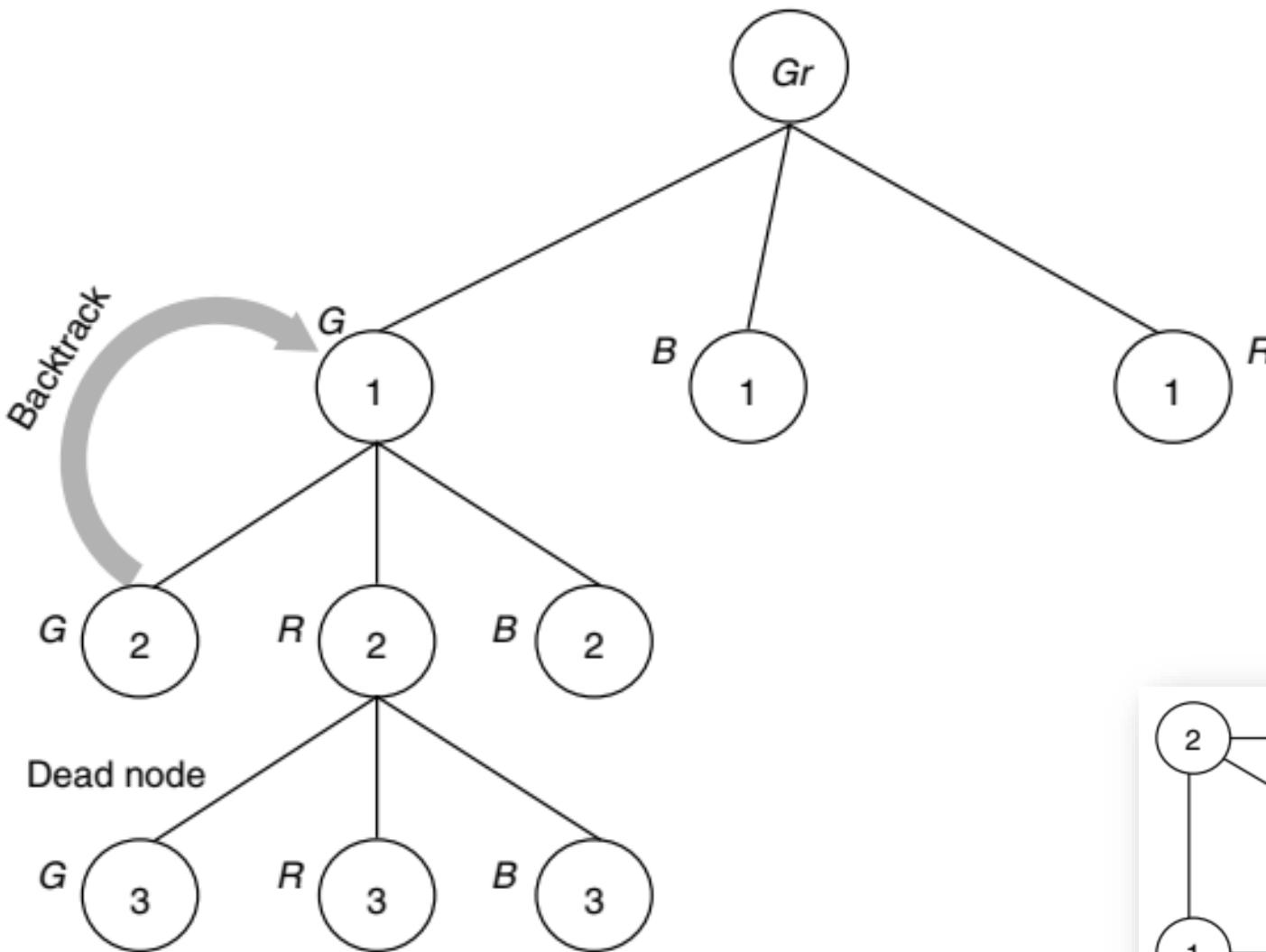


	1	2	3	4
1	0	1	0	1

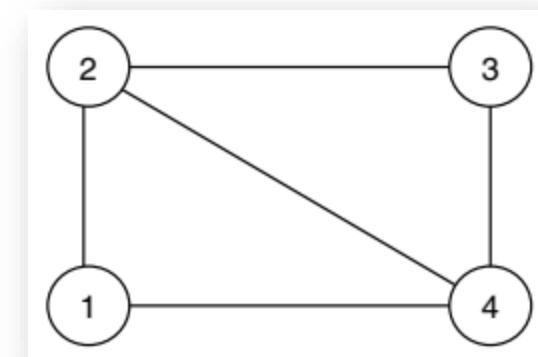
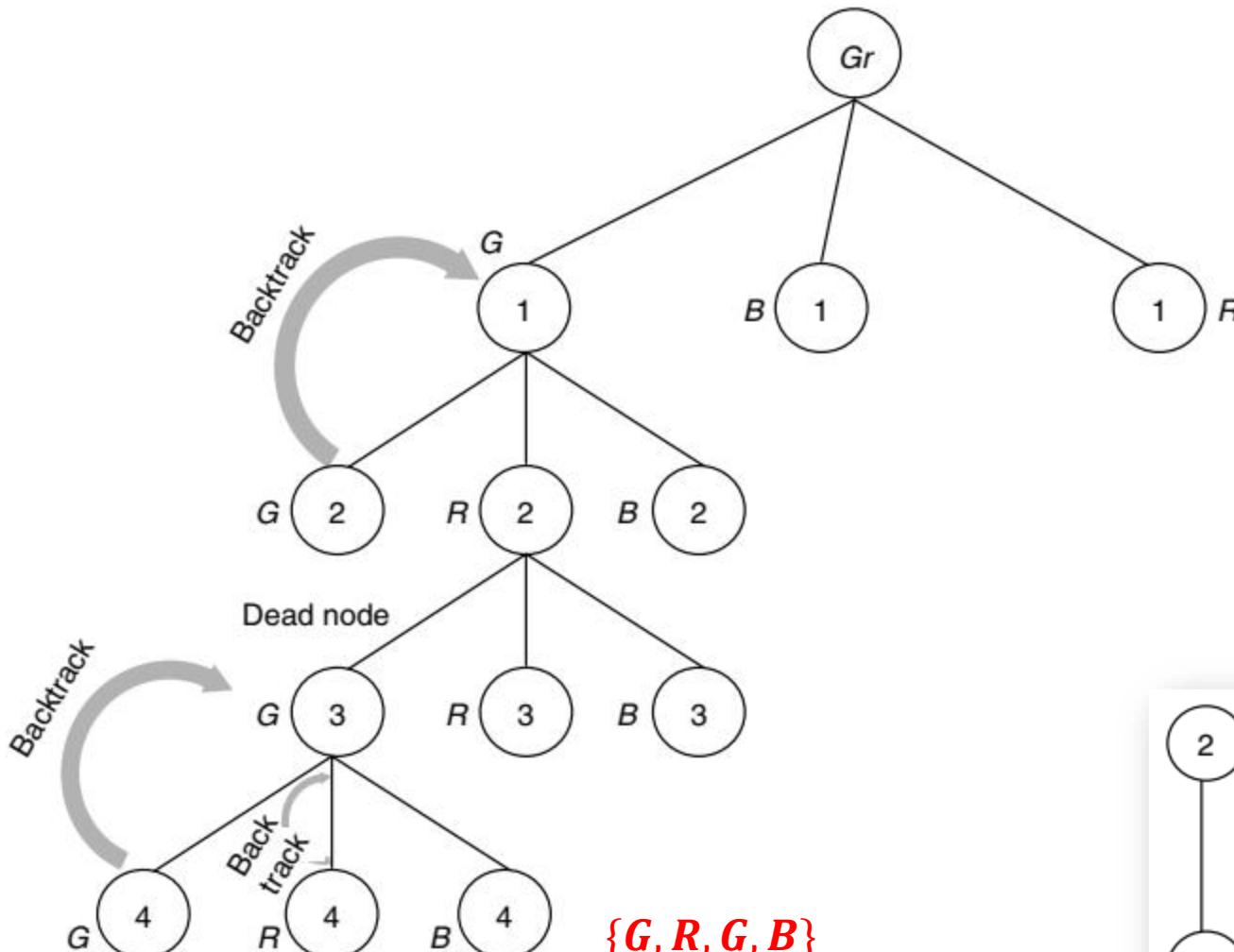
State Space Tree:



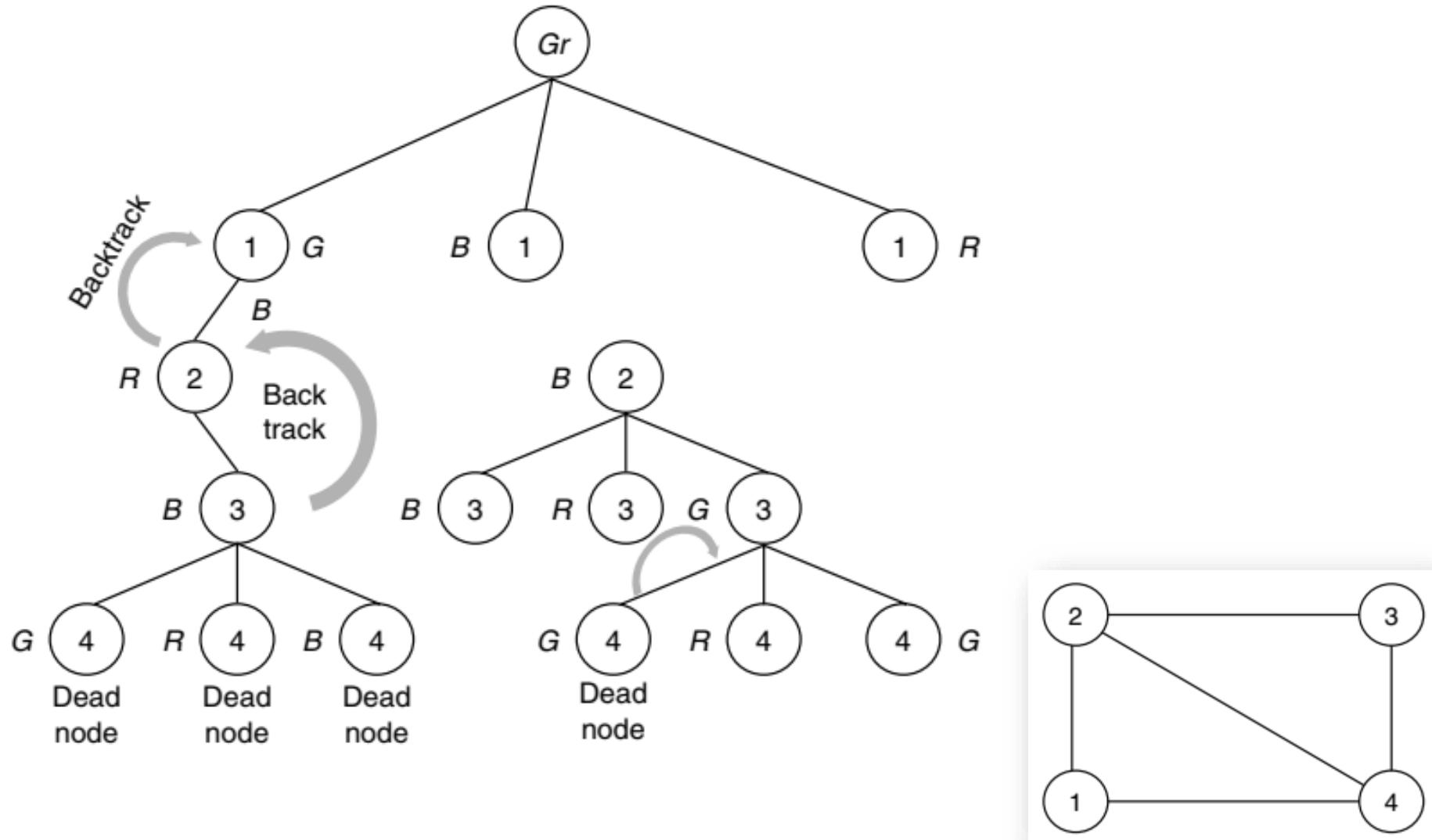
Backtracking : Graph Coloring Example



Backtracking : Graph Coloring Example

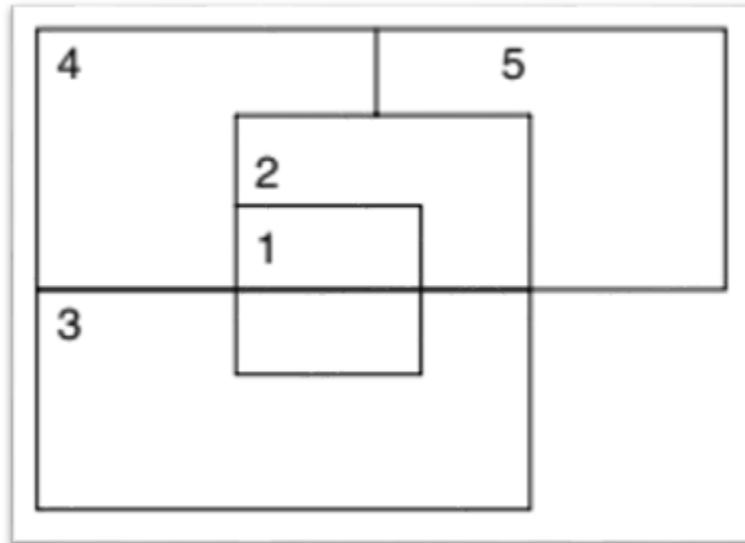


Backtracking : Graph Coloring Example

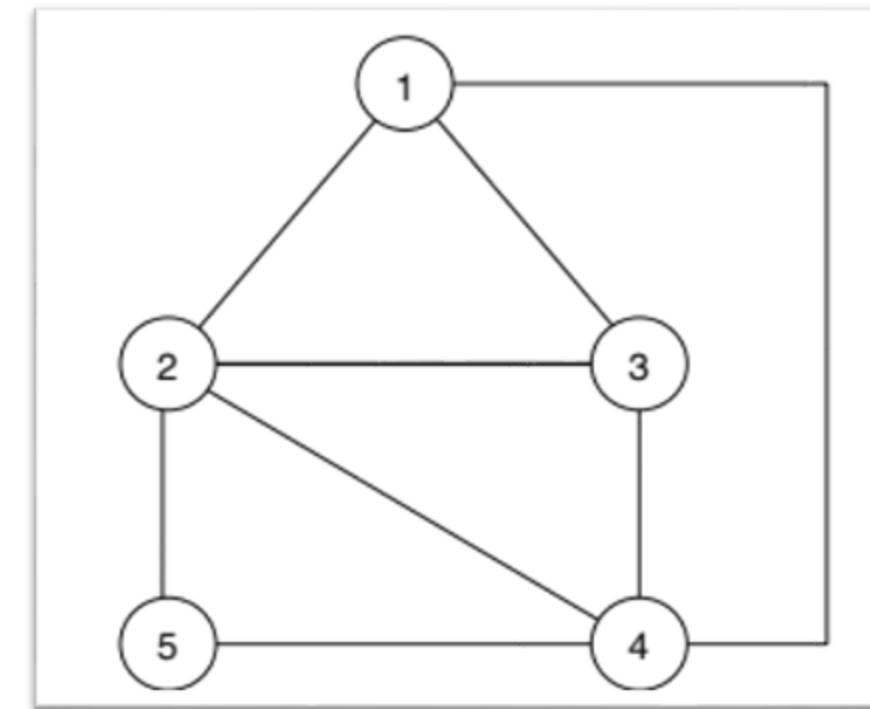
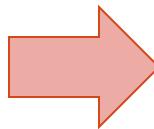


Backtracking : Graph Coloring Example

Example 2:



Map



Planer Graph

Branch & Bound : Introduction

- **Branch:** using State Space Tree (Similar to Backtracking)
- **Bound:** using Upper and Lower bounds
- Branch and Bound differs from backtracking in the sense that all the children of the E-Node are generated before any other live node becomes the E-Node.
- Branch and Bound is the generalization of both graph search strategies, BFS and DFS.
- The state space tree of the branch and bound method can be constructed using following three strategies:
 - FIFO (First In First Out) search (Implemented using QUEUE)
 - LIFO (Last In First Out) search (Implemented using STACK)
 - LC (Least Cost) search (Implemented using PRIORITY QUEUE)

Branch & Bound : Introduction

FIFO (First In First Out) Branch and Bound

- In FIFO search, queue data structure is used.
- Initially node 1 is taken as the E-node.
- The child nodes of node 1 are generated. All these live nodes are placed in a queue.
- Next the first element in the queue is deleted, i.e. node 2, the child nodes of node 2 are generated and placed in the queue.
- This continues until the answer node is found.

Branch & Bound : FIFO

LIFO (Last In First Out) Branch and Bound

Example: Job sequencing with deadlines problem

- Jobs = {J1, J2, J3, J4}; P = {10, 5, 8, 3}; d = {1, 2, 1, 2}
- Node 1 is the E-node. Child nodes of node 1 are generated and placed in the queue.

2	3	4	5	
---	---	---	---	--

- First element in the queue is deleted, ie., 2 is deleted and its child nodes are generated.

3	4	5	6	7	8	
---	---	---	---	---	---	--

- Similarly, the next element is deleted, ie., 3 and its child nodes are generated and placed in the queue. This is continued until an answer node is reached.

Branch & Bound : FIFO

FIFO (First In First Out) Branch and Bound

Example: Job sequencing with deadlines problem

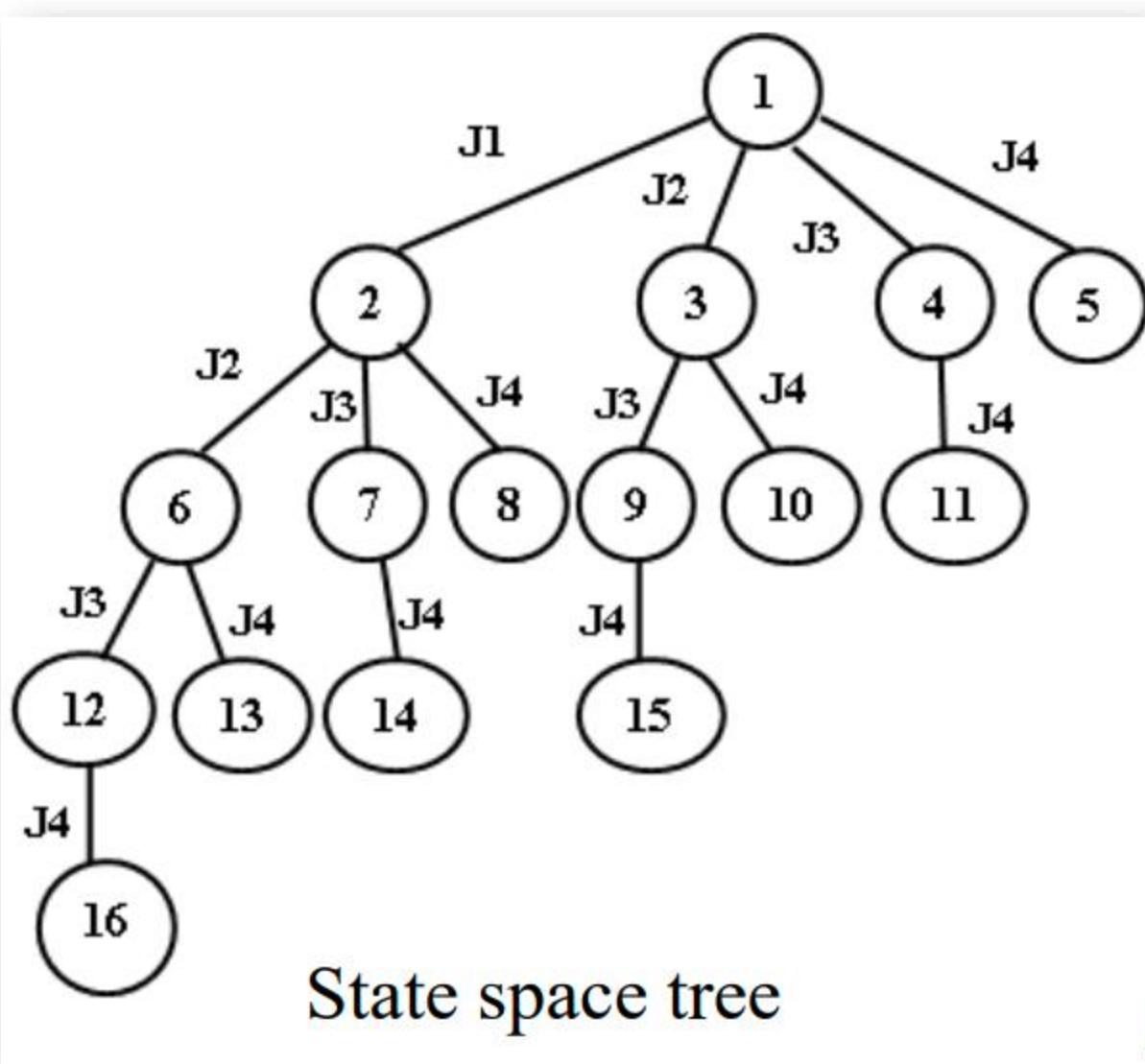
- Jobs = {J1, J2, J3, J4}; P = {10, 5, 8, 3}; d = {1, 2, 1, 2}
- Node 1 is the E-node. Child nodes of node 1 are generated and placed in the queue.



- First element in the queue is deleted, ie., 2 is deleted and its child nodes are generated.



- Similarly, the next element is deleted, ie., 3 and its child nodes are generated and placed in the queue. This is continued until an answer node is reached.



Branch & Bound : LIFO

LIFO (Last In First Out) Branch and Bound

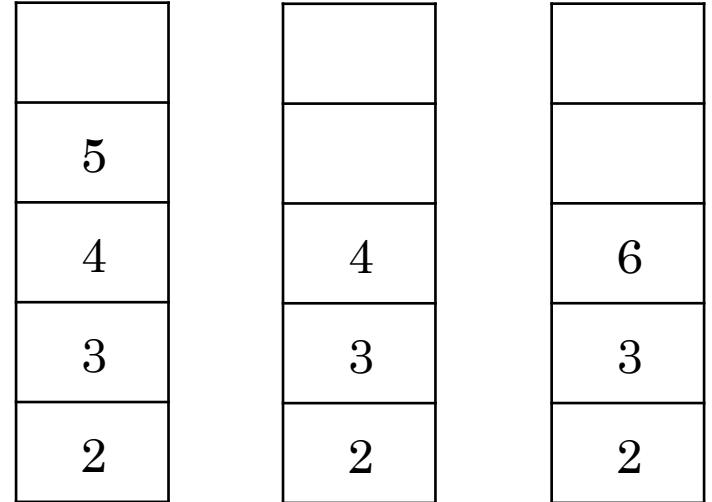
- In LIFO search, stack data structure is used.
- Initially node 1 is taken as the E-node.
- The child nodes of node 1 are generated. All these live nodes are placed in a stack.
- Next the first element in the stack is deleted, i.e. node 5, the child nodes of node 5 are generated and placed in the stack.
- This continues until the answer node is found.

Branch & Bound : LIFO

FIFO (First In First Out) Branch and Bound

Example: Job sequencing with deadlines problem

- Jobs = {J₁, J₂, J₃, J₄}; P = {10, 5, 8, 3}; d = {1, 2, 1, 2}
- Node 1 is the E-node. Child nodes of node 1 are generated and placed in the stack.
- First element in the stack is deleted, i.e., 5 is deleted and its child nodes are generated.
- Similarly, the next element is deleted, i.e., 4 and its child nodes are generated and placed in the queue. This is continued until an answer node is reached.

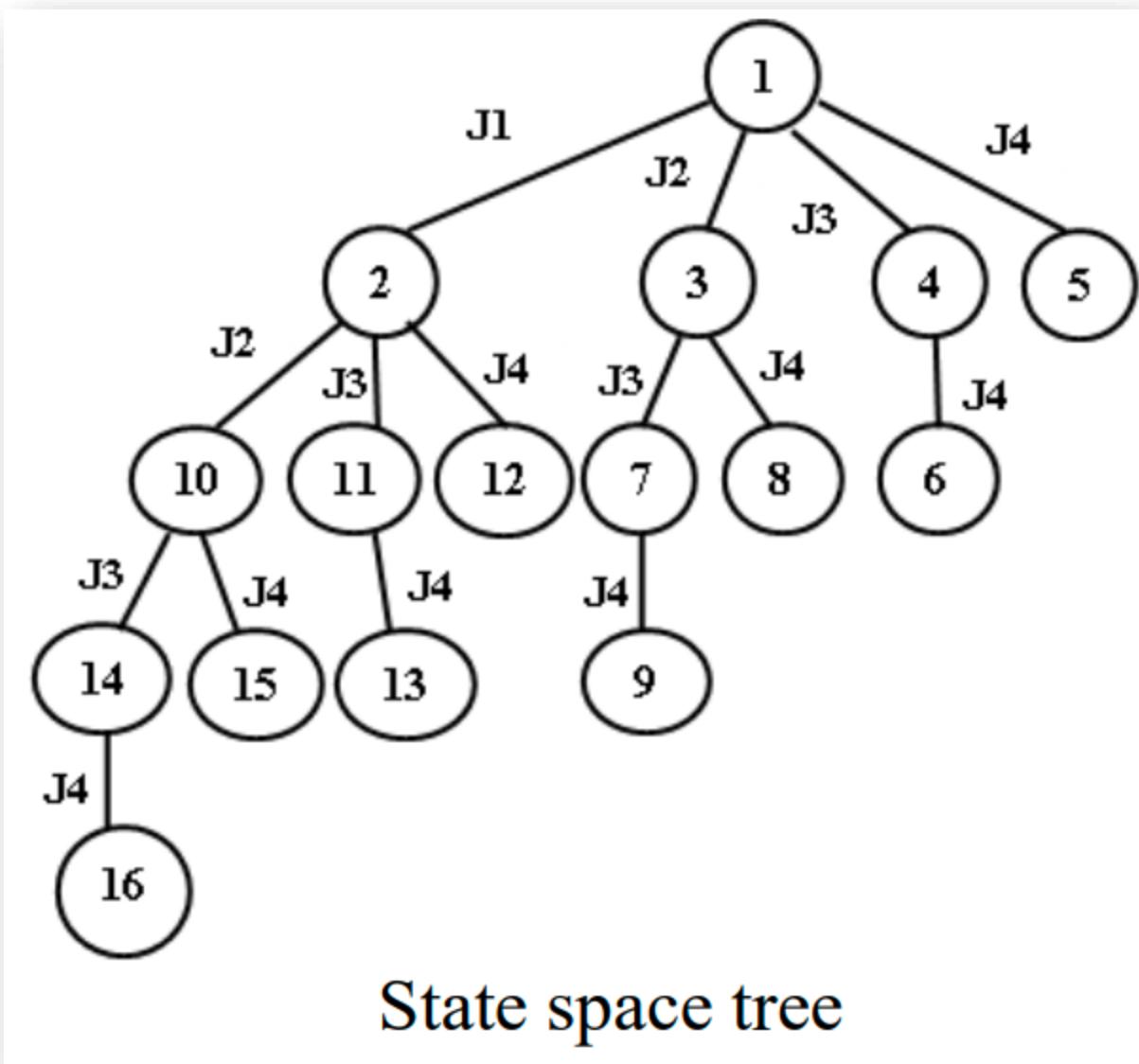


Branch & Bound : LIFO

FIFO (First In First Out) Branch and Bound

Example: Job sequencing with deadlines problem

- Jobs = {J1, J2, J3, J4}; P = {10, 5, 8, 3}; d = {1, 2, 1, 2}
- Node 1 is the E-node. Child nodes of node 1 are generated and placed in the stack.
- First element in the stack is deleted, i.e., 5 is deleted and its child nodes are generated.
- Similarly, the next element is deleted, i.e., 4 and its child nodes are generated and placed in the queue. This is continued until an answer node is reached.



Branch & Bound : LCBB

LC (Least Count) Branch and Bound:

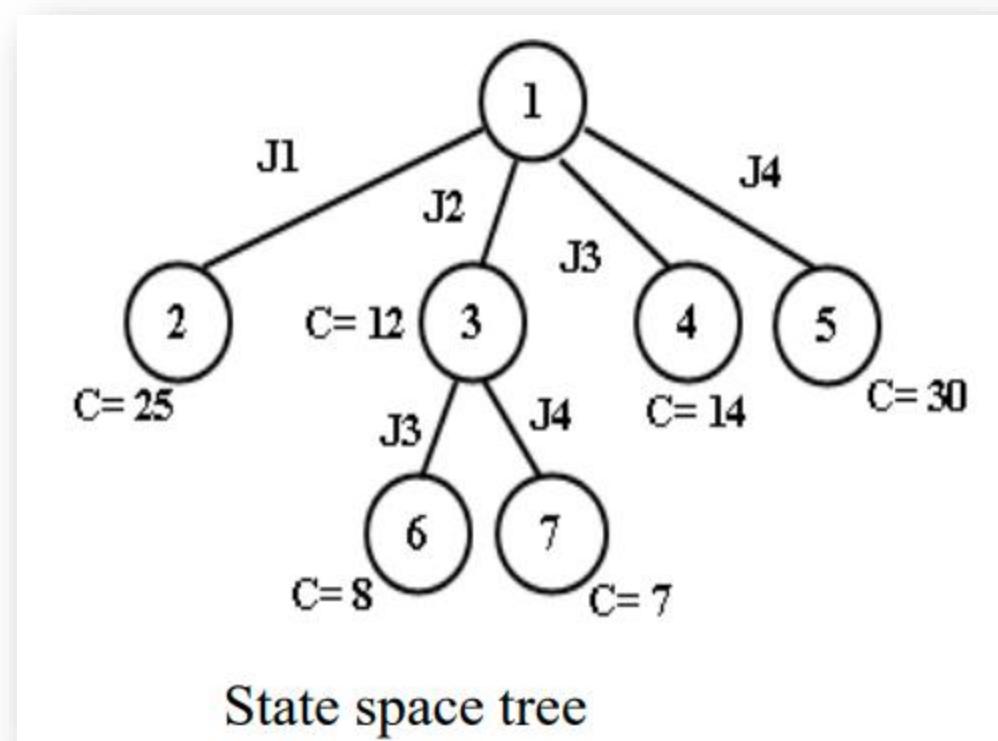
- In both FIFO and LIFO Branch and Bound the selection rules for the next E-node is rigid and blind.
- The selection rule for the next E-node does not give any preferences to a node that has a very good chance of getting the search to an answer node quickly.
- In this method ranking function or cost function is used.
- The child nodes of the E-node are generated, among these live nodes; a node which has minimum cost is selected. By using ranking function, the cost of each node is calculated.

Branch & Bound : LCBB

LC (Least Count) Branch and Bound:

Example: Job sequencing with deadlines problem

- Jobs = {J₁, J₂, J₃, J₄}; P = {10, 5, 8, 3}; d = {1, 2, 1, 2}
- Initially we will take node 1 as E-node. Generate children of node 1, the children are 2, 3, 4, 5. By using ranking function we will calculate the cost of 2, 3, 4, 5 nodes is $\hat{c} = 25$, $\hat{c} = 12$, $\hat{c} = 14$, $\hat{c} = 30$ respectively.
- Now we will select a node which has minimum cost i.e., node 3. For node 3, the children are 6, 7.



Branch & Bound : LCBB

LC (Least Count) Branch and Bound:

- All the live nodes are stored in a PRIORITY QUEUE or HEAP.
- The live nodes are not selected according to the order in which they have been queued or stacked but according to their heuristic value.
- The heuristic value is calculated for each live node and then the node with the highest heuristic value is chosen as the E-node.

Branch & Bound : 0/1 Knapsack Problem

LC (Least Count) Branch and Bound:

- The 0/1 knapsack problem is to

$$\text{Maximize } \sum_{i=1}^n p_i x_i \text{ subject to } \sum_{i=1}^n w_i x_i \leq M$$

- objective of this problem is to fill the knapsack in order to maximize the profit subject to its capacity.
- But Branch & Bound is used for minimization problem.

Branch & Bound : 0/1 Knapsack Problem

LC (Least Count) Branch and Bound:

- This modified knapsack problem is stated as,
- The 0/1 knapsack problem is the maximization problem where the value of the objective function $\hat{c}(x) = \sum p_i x_i$ is maximized subjected to $\sum w_i x_i \leq M$,
- Now our aim is minimization, so we take the objective function $\hat{c}(x) = - \sum p_i x_i$ subjected to $\sum w_i x_i \leq M$ in order to convert the 0/1 knapsack problem as the minimization problem where $x_i = 0$ or 1 , $1 \leq i \leq n$
- The two functions $\hat{c}(x)$ and $U(x)$ are defined using two algorithms Bound and UBound .

Branch & Bound : 0/1 Knapsack Problem

LC (Least Count) Branch and Bound:

- UBound computes the weights of the list of objects placed in the knapsack as a whole and their sum $\leq m$, and the profit is correspondingly decremented from initial profit and returned.
- Bound is similar to UBound but it considers fractional objects to use the entire capacity of the sack $\sum w_i x_i = m$.

Branch & Bound : 0/1 Knapsack Problem

$$n = 4; m = 15;$$

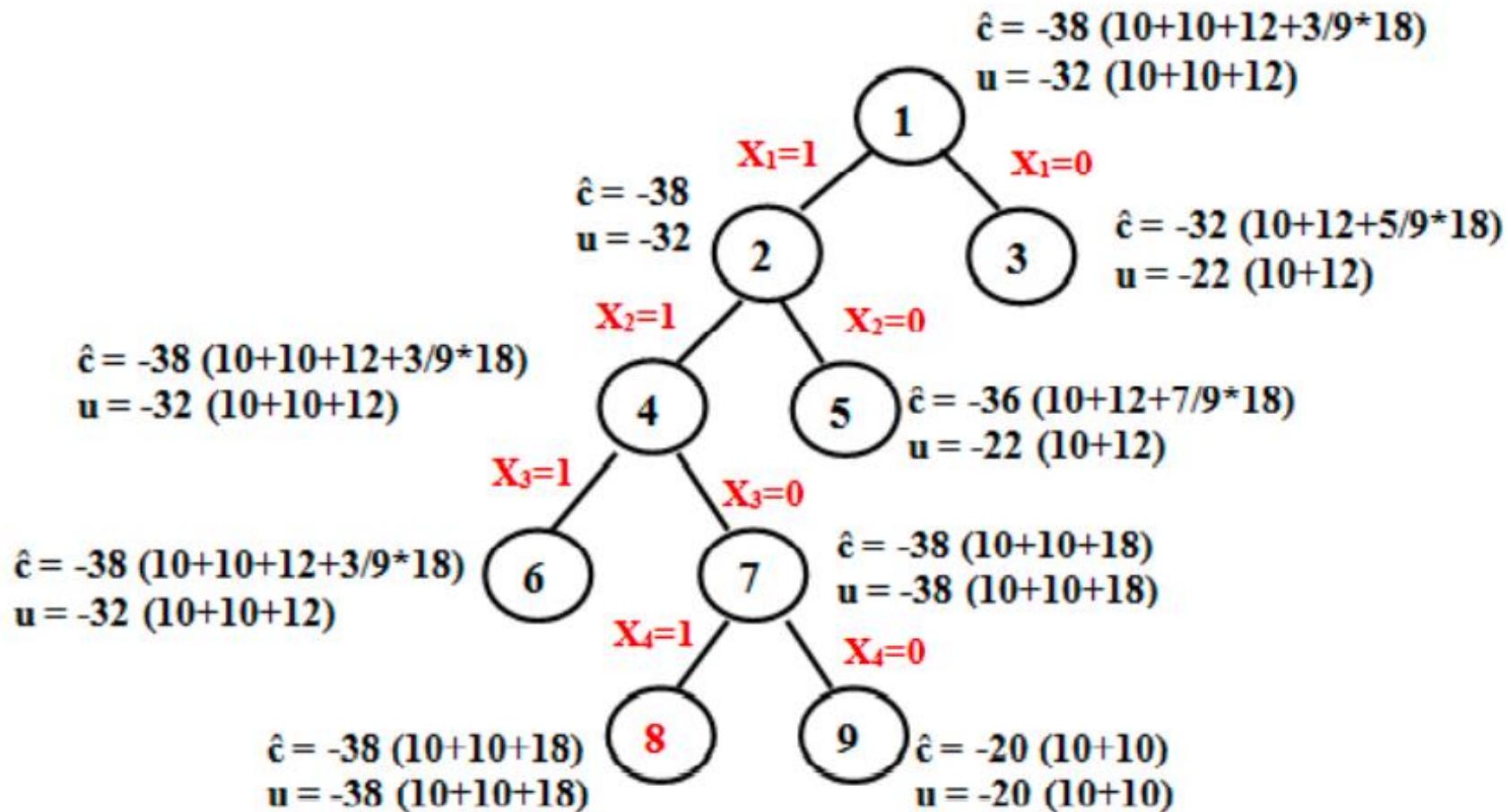
$$(p_1, p_2, p_3, p_4) = \{10, 10, 12, 18\}; (w_1, w_2, w_3, w_4) = \{2, 4, 6, 9\}$$

$$x_1 = 1,$$

$$x_2 = 1,$$

$$x_3 = 0,$$

$$x_4 = 1$$



Branch & Bound : 15 Puzzle Problem

- The 15 Puzzle problem is invented by Sam Loyd in 1878.
- The problem consist of 15 numbered (0-15) tiles on a square box with 16 tiles(one tile is blank or empty).
- The objective of this problem is to change the arrangement of initial node to goal node by using series of legal moves.
- The Initial and Goal node arrangement is shown by following figure

Initial state				Goal state			
1	2	3	4	1	2	3	4
5	6		8	5	6	7	8
9	10	7	11	9	10	11	12
13	14	15	12	13	14	15	

Figure 19 Initial and goal states for 15-puzzle problem.

Branch & Bound : 15 Puzzle Problem

- In initial node four moves are possible. User can move any one of the tile like 2, or 3, or 5, or 6 to the empty tile. From this we have four possibilities to move from initial node.
- The legal moves are for adjacent tile number is left, right, up, down, ones at a time.
- Each and every move creates a new arrangement, and this arrangement is called state of puzzle problem.
- By using different states, a state space tree diagram is created, in which edges are labeled according to the direction in which the empty space moves.
- The LCBB method is the general method used to solve the 15-puzzle problem so that the goal state can be achieved in minimum number of tile movement.

Branch & Bound : 15 Puzzle Problem

- In state space tree, nodes are numbered as per the level. In each level we must calculate the value or cost of each node by using given formula:

$$C(x) = f(x) + g(x),$$

- $f(x)$ is length of path from root or initial node to node x ,
- $g(x)$ is estimated length of path from x downward to the goal node. Number of non-blank tile not in their correct position.
- $C(x) < \text{Infinity}$. (initially set bound).
- Each time node with smallest cost is selected for further expansion towards goal node. This node become the e-node.

Branch & Bound : 15 Puzzle Problem

- Example:

Solve the given 15-puzzle problem using LCBB.

Initial state

1	2	3	4
5	6		8
9	10	7	11
13	14	15	12

Goal state

1	2	3	4
5	6	7	8
9	10	11	12
13	14	15	

Branch & Bound : 15 Puzzle Problem

- Example:

