



CS215 DISCRETE MATHEMATICS FOR COMPUTER SCIENCE

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Euler's Formula

- **Theorem** (Euler's Formula) Let G be a connected planar simple graph with e edges and v vertices. Let r be the number of regions in a planar representation of G . Then $r = e - v + 2$.



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- **Corollary 1** If G is a connected planar simple graph with e edges and v vertices, where $v \geq 3$, then $e \leq 3v - 6$.
- Corollary 2** If G is a connected planar simple graph, then G has a vertex of degree not exceeding 5.
- Corollary 3** In a connected planar simple graph has e edges and v vertices with $v \geq 3$ and no circuits of length three, then $e \leq 2v - 4$.



Graph Coloring

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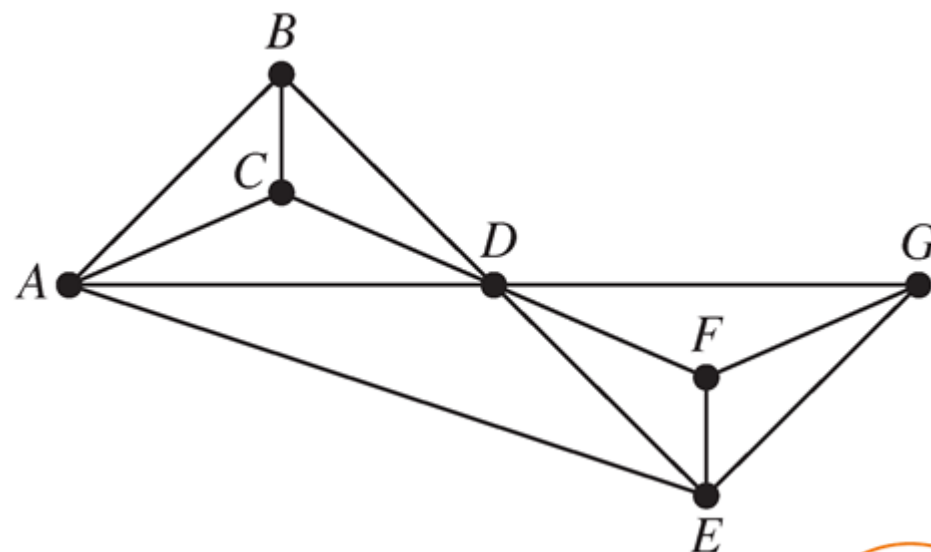
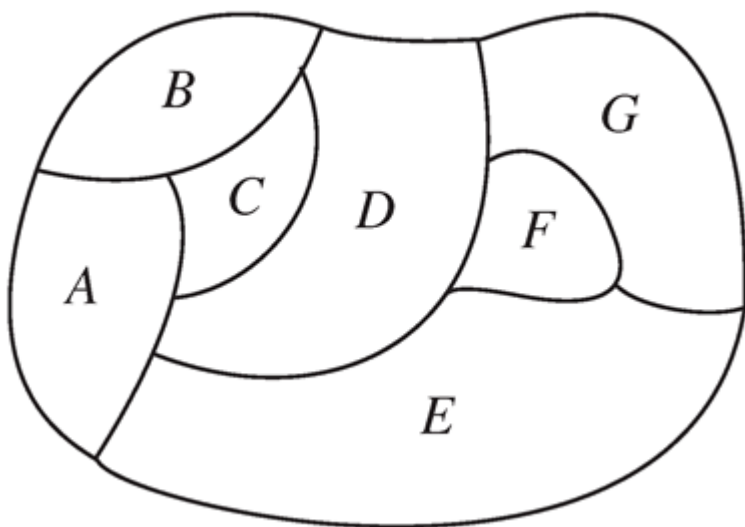
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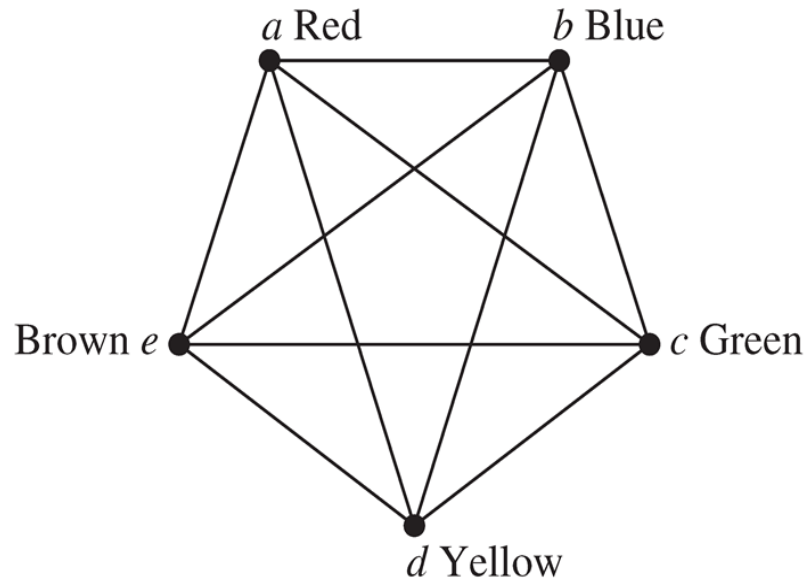
Examples

- What is the chromatic number of K_n , $K_{m,n}$, C_n ?



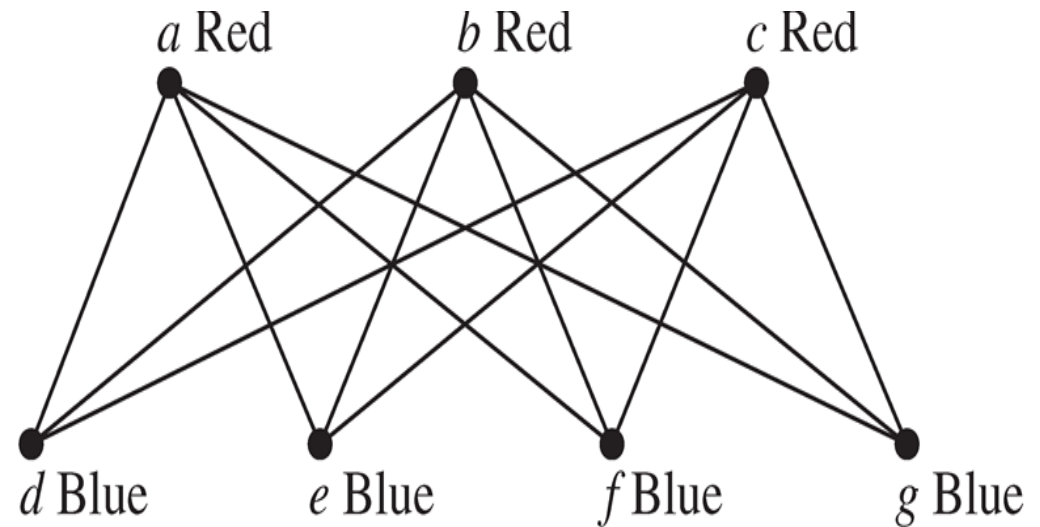
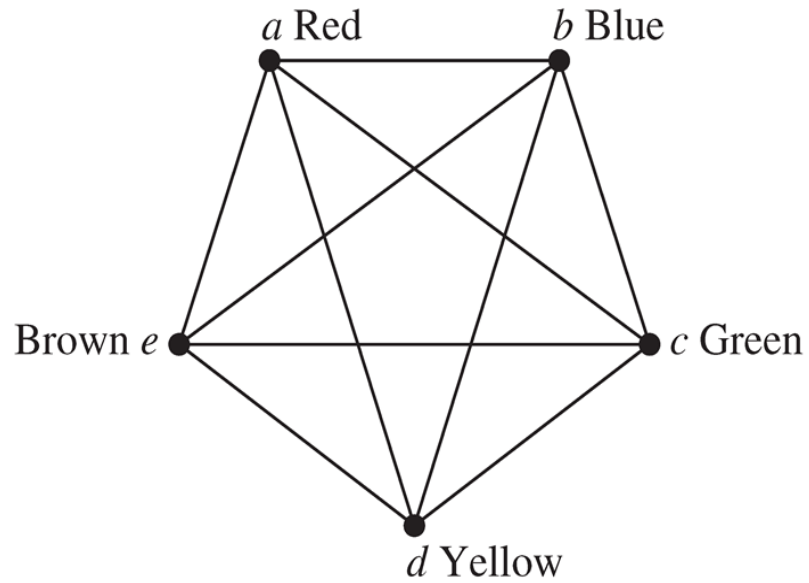
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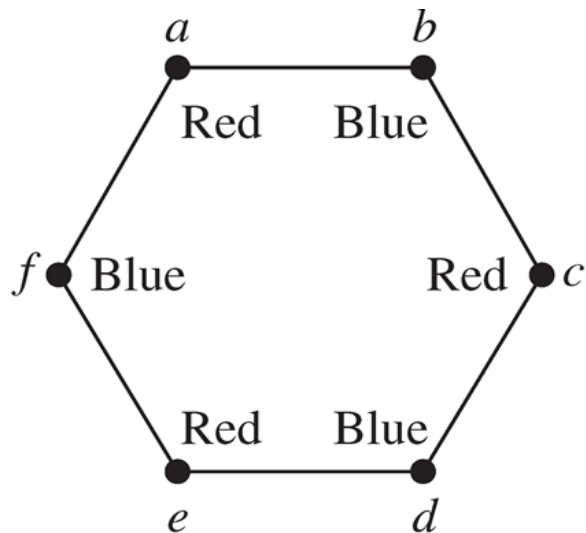
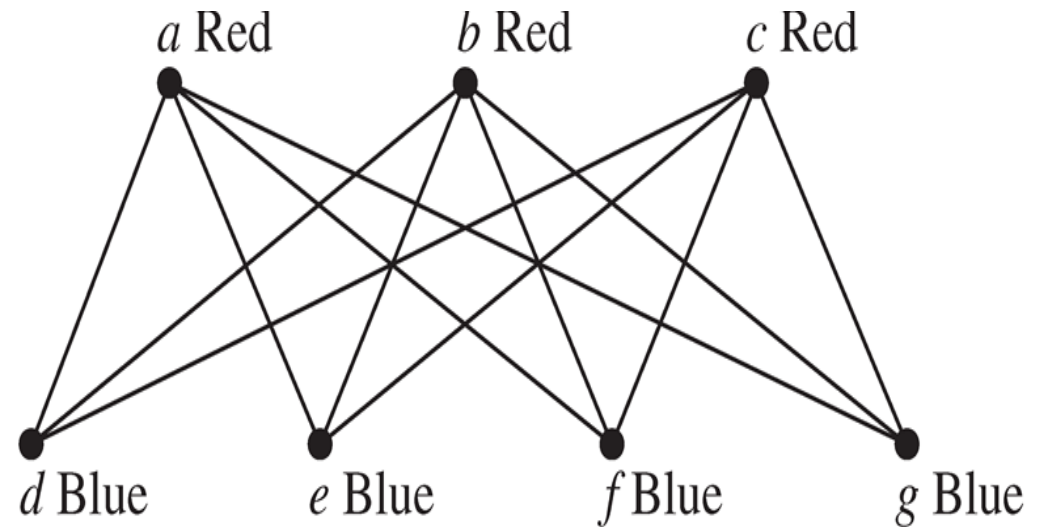
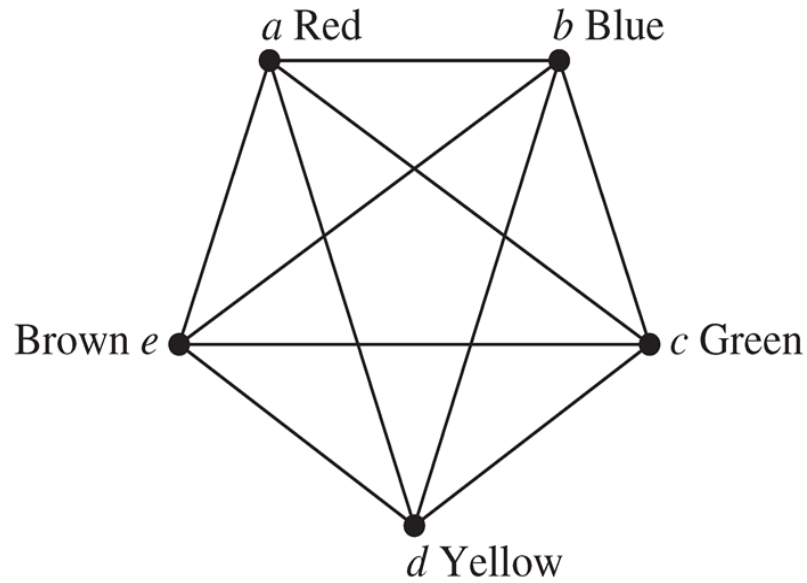
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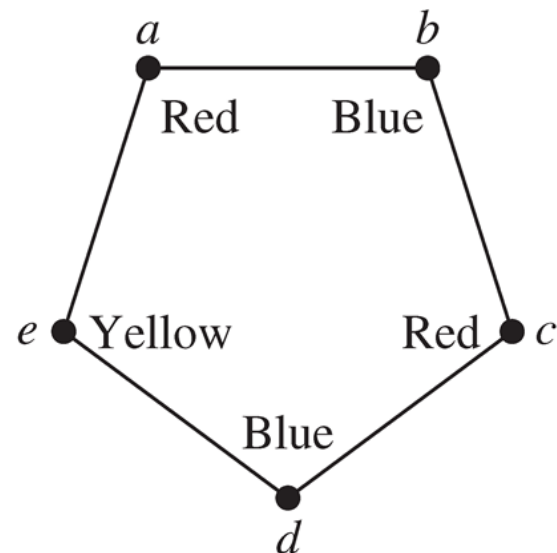
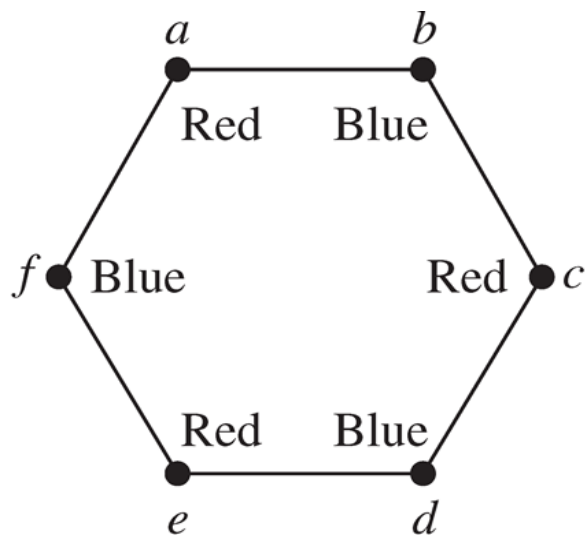
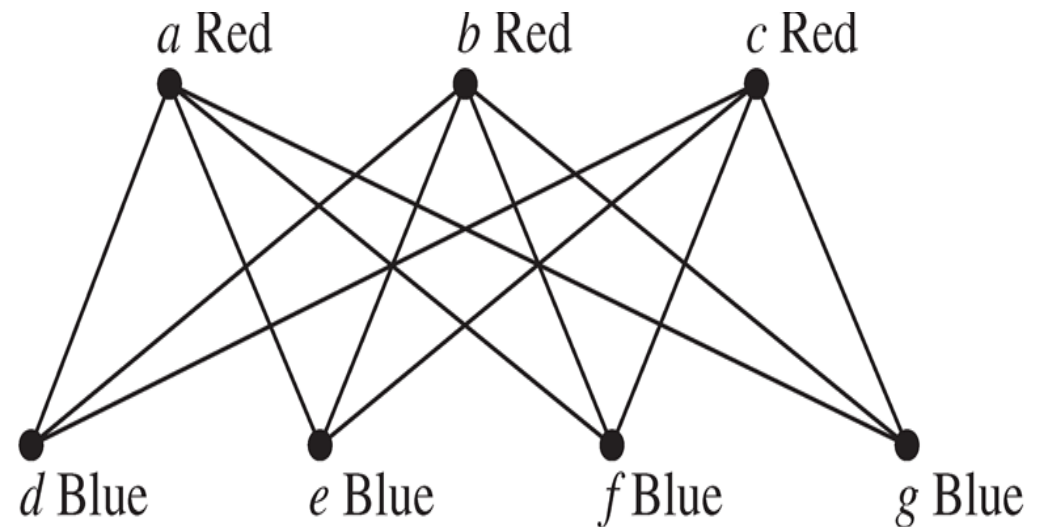
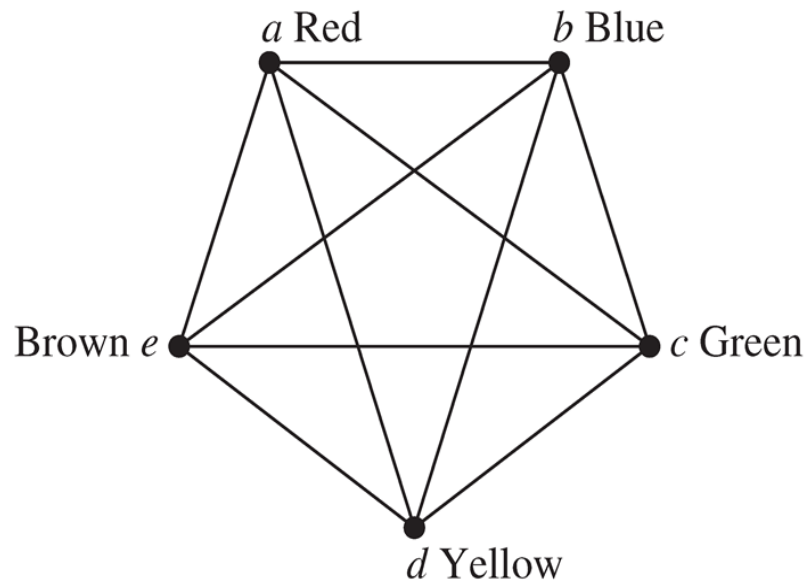
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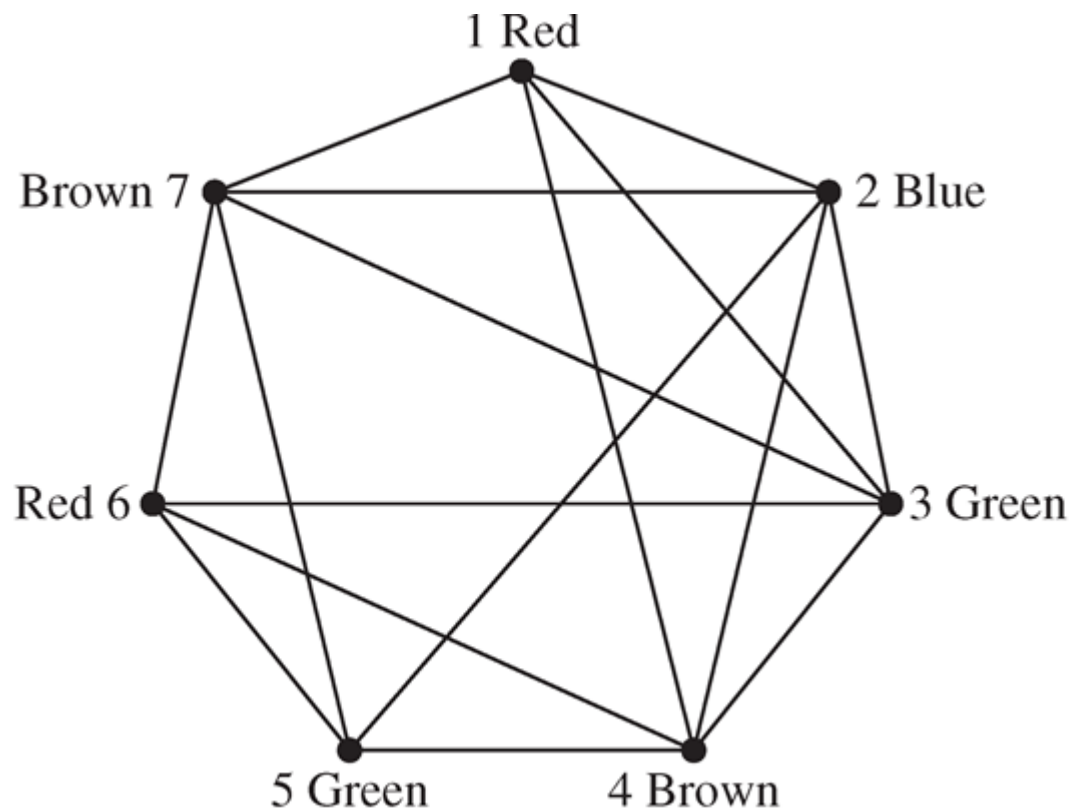
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Applications of Graph Coloring

■ Scheduling Final Exams

Vertices represent courses, and there is an edge between two vertices if there is a common student in the courses.



Time Period

I

Courses

1, 6

II

2

III

3, 5

IV

4, 7

Applications of Graph Coloring

■ Channel Assignments

Television channels 2 through 13 are assigned to stations in North America so that no two stations within 150 miles can operate on the same channel . How can the assignment of channels be modeled by graph coloring?



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Graph Coloring \in NPC



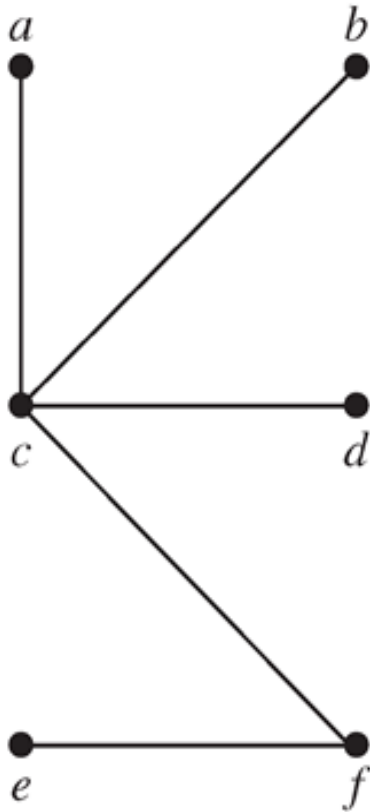
Trees

- **Definition** A *tree* is a connected undirected graph with no simple circuits.



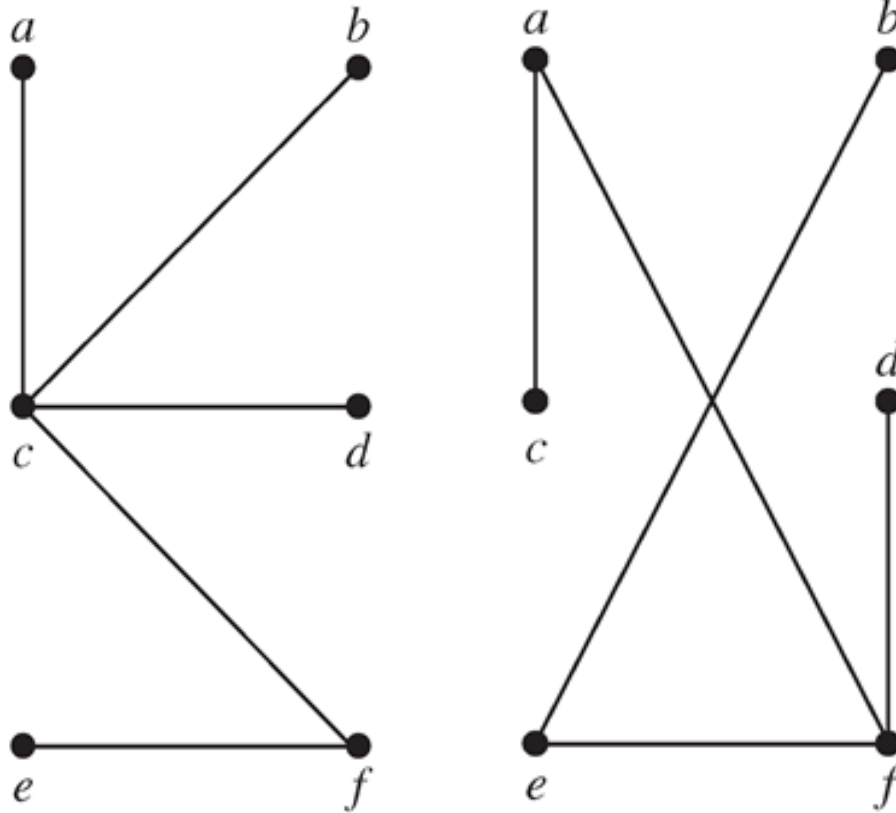
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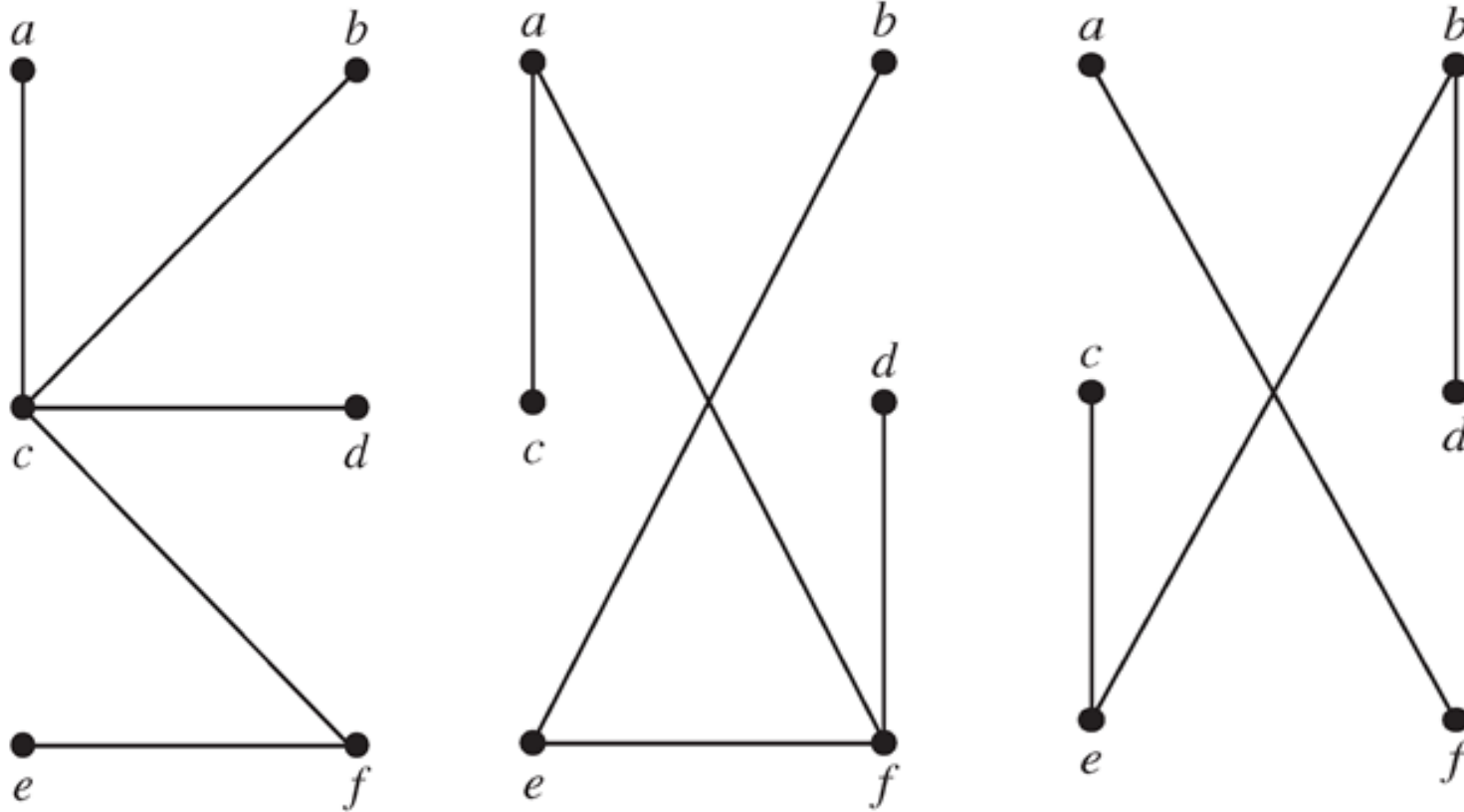
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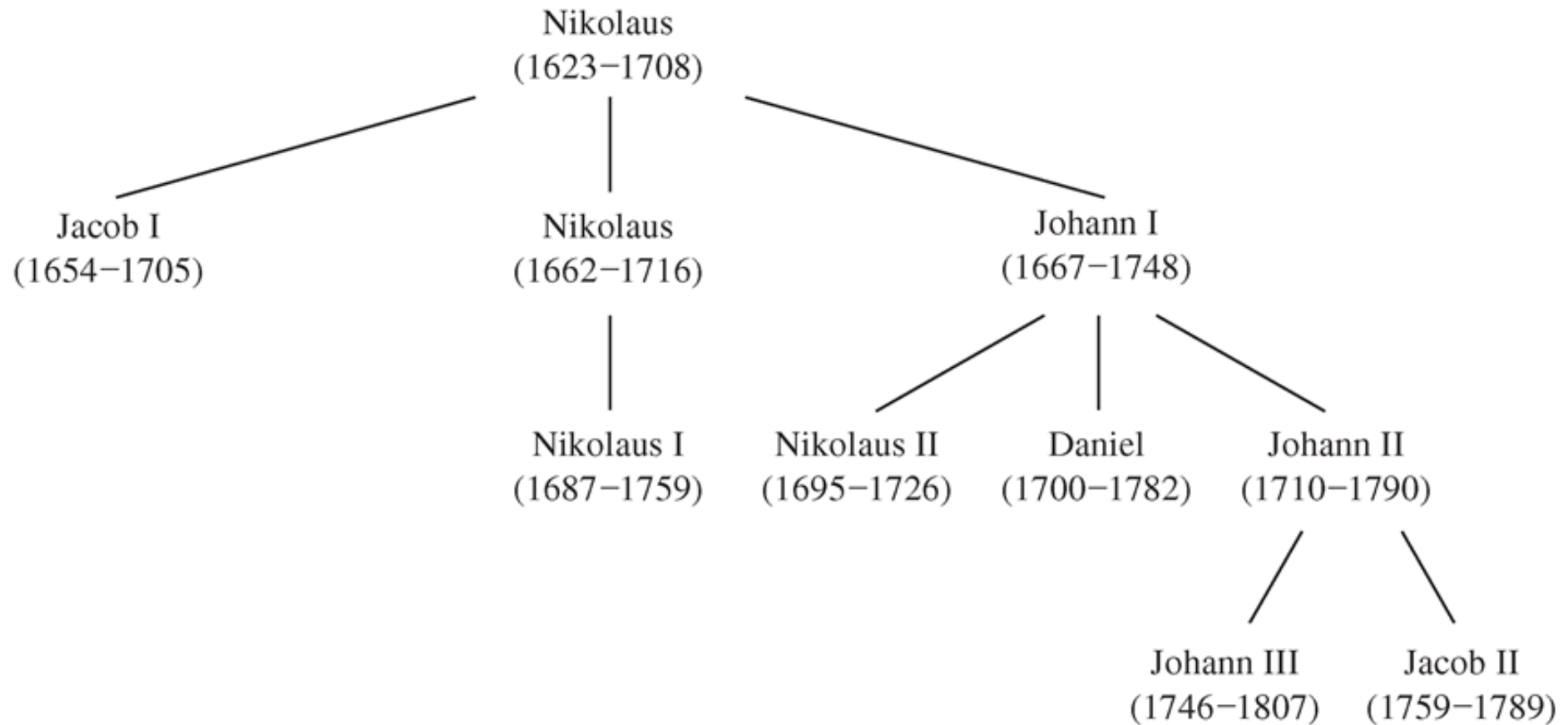
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Two properties of tree: **connected**, **no circuit**



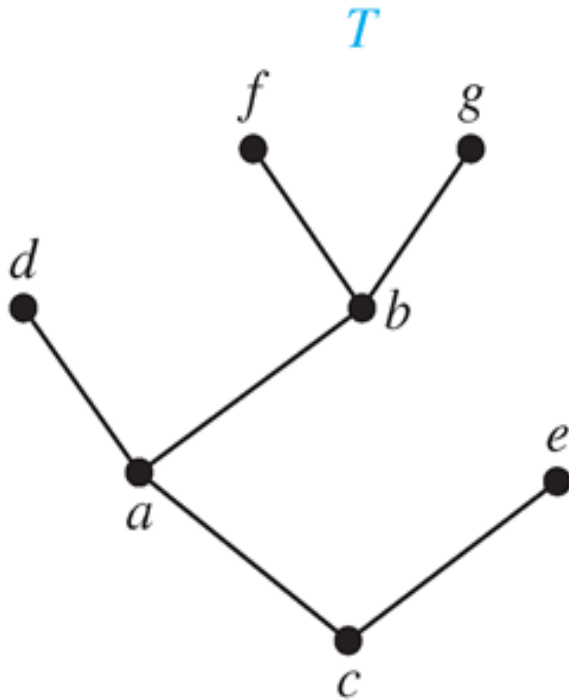
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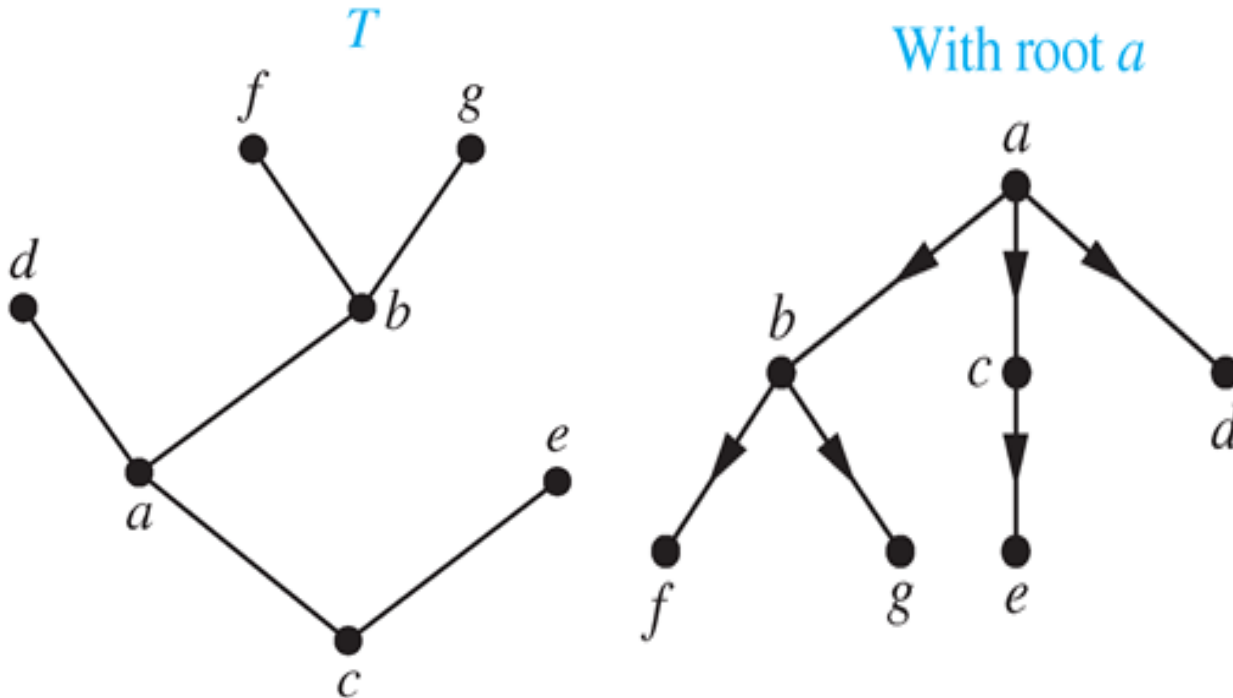
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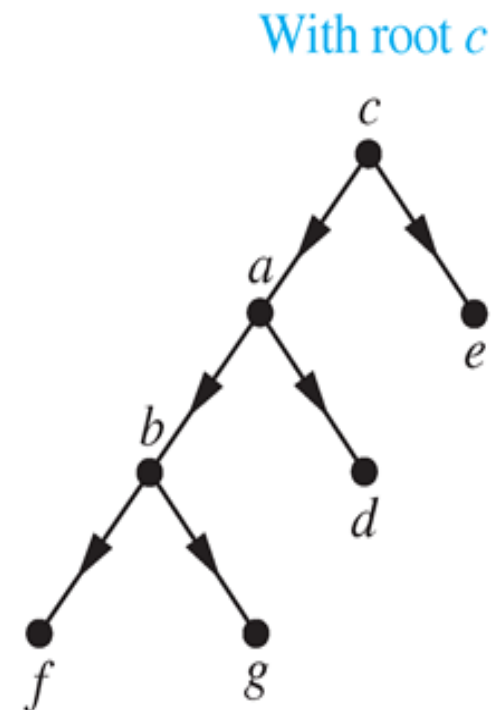
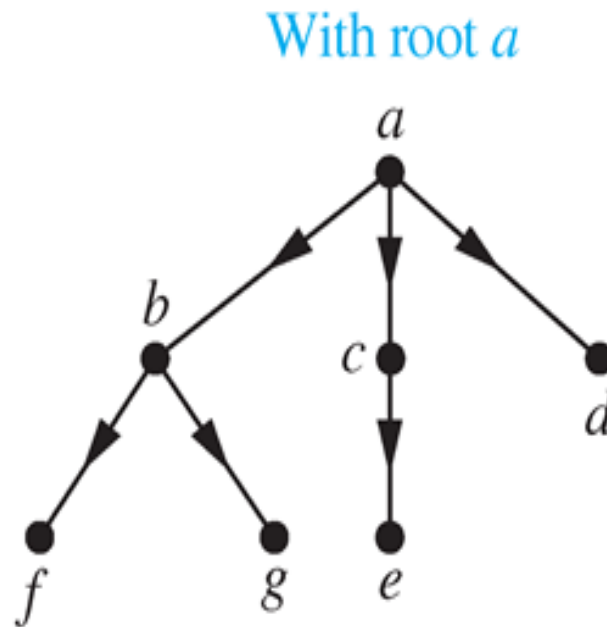
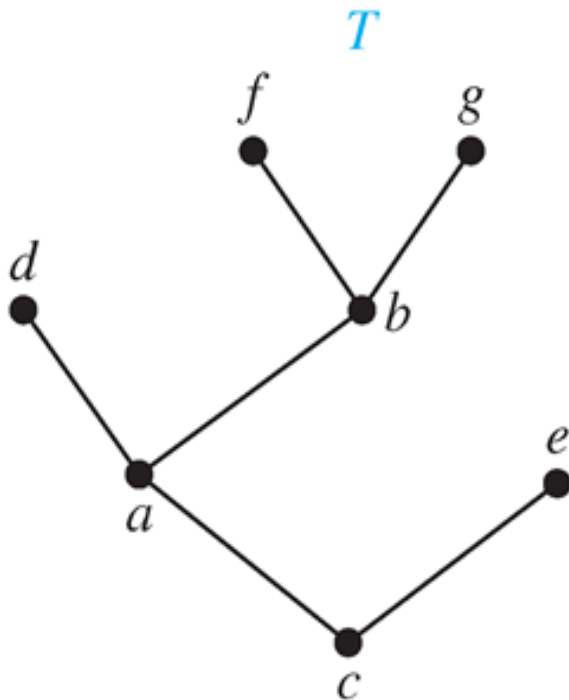
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ancestor, descendant



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leaf, internal vertex



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subtree with a as its root: consists of a and its descendants and all edges incident to these descendants



m -Ary Trees

- **Definition** A rooted tree is called an *m -ary tree* if every internal vertex has **no more than** m children. The tree is called a *full m -ary tree* if every internal vertex has **exactly** m children. In particular, an m -ary tree with $m = 2$ is called a *binary tree*.



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- (ii) i internal vertices, $n = mi + 1$, $\ell = (m - 1)i + 1$
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using $n = mi + 1$ and $n = i + \ell$



Level and Height

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Definition A rooted m -ary tree of height h is *balanced* if all leaves are at levels h or $h - 1$. (differ no greater than 1)



The Number of Leaves

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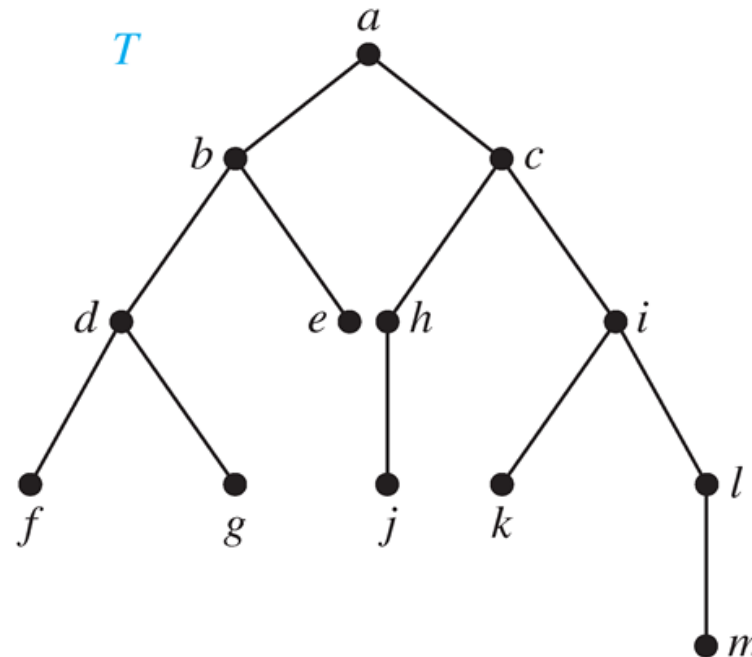
Binary Trees

- **Definition** A *binary tree* is an **ordered** rooted tree where each internal tree has **two children**, the first is called the *left child* and the second is the *right child*. The tree rooted at the left child of a vertex is called the *left subtree* of this vertex, and the tree rooted at the right child of a vertex is called the *right subtree* of this vertex.



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The three most commonly used traversals are *preorder traversal*, *inorder traversal*, *postorder traversal*.

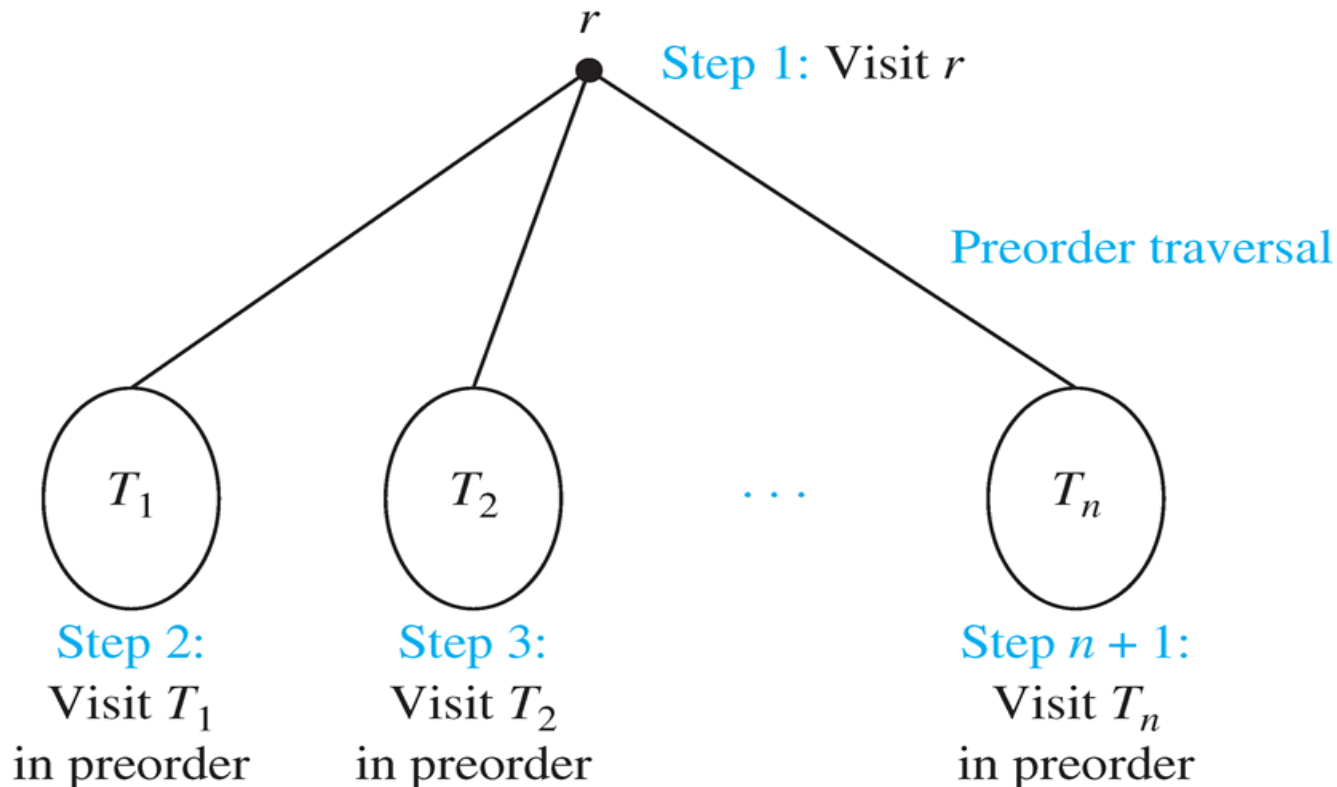


Preorder Traversal

- **Definition** Let T be an ordered rooted tree with root r . If T consists only of r , then r is the *preorder traversal* of T . Otherwise, suppose that T_1, T_2, \dots, T_n are the subtrees of r from left to right in T . The *preorder traversal* begins by *visiting* r , and continues by traversing T_1 in preorder, then T_2 in preorder, and so on, until T_n is traversed in preorder.

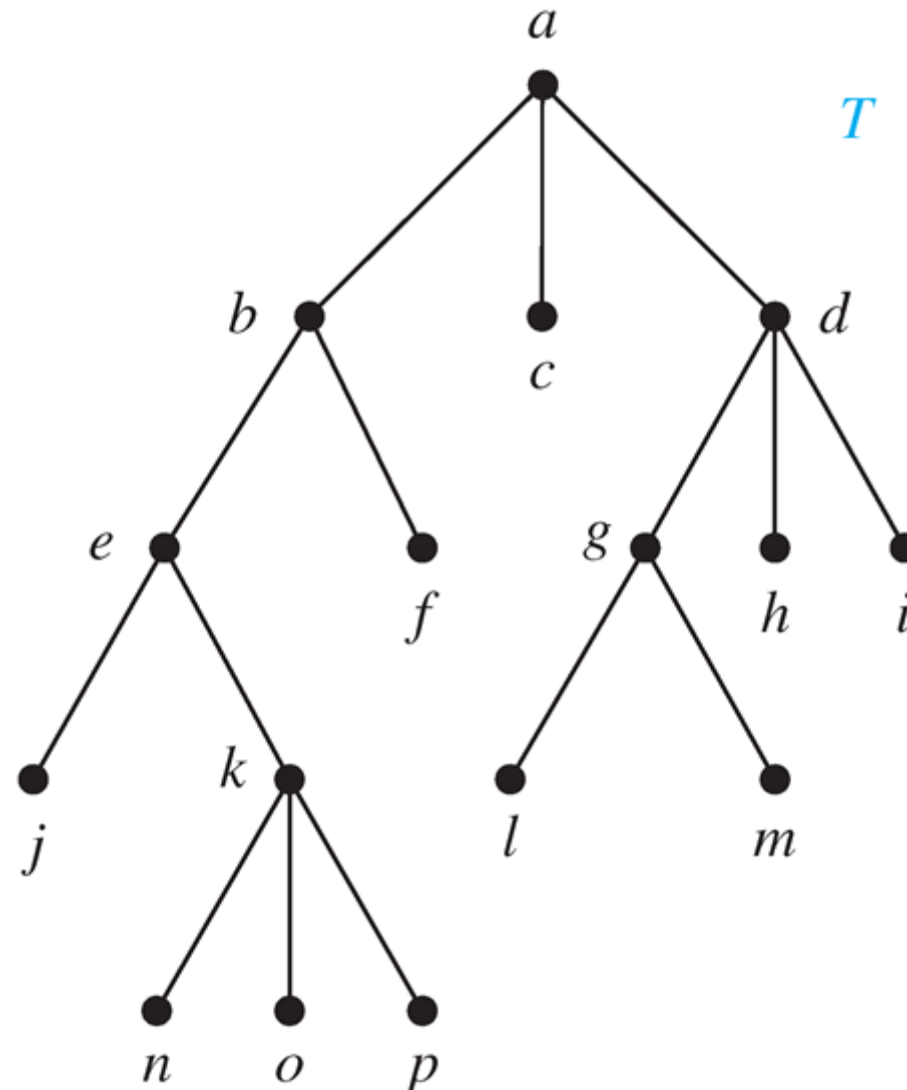
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Preorder Traversal

■ Example



Preorder Traversal

```
procedure preorder ( $T$ : ordered rooted tree)
 $r := \text{root of } T$ 
list  $r$ 
for each child  $c$  of  $r$  from left to right
     $T(c) := \text{subtree with } c \text{ as root}$ 
    preorder( $T(c)$ )
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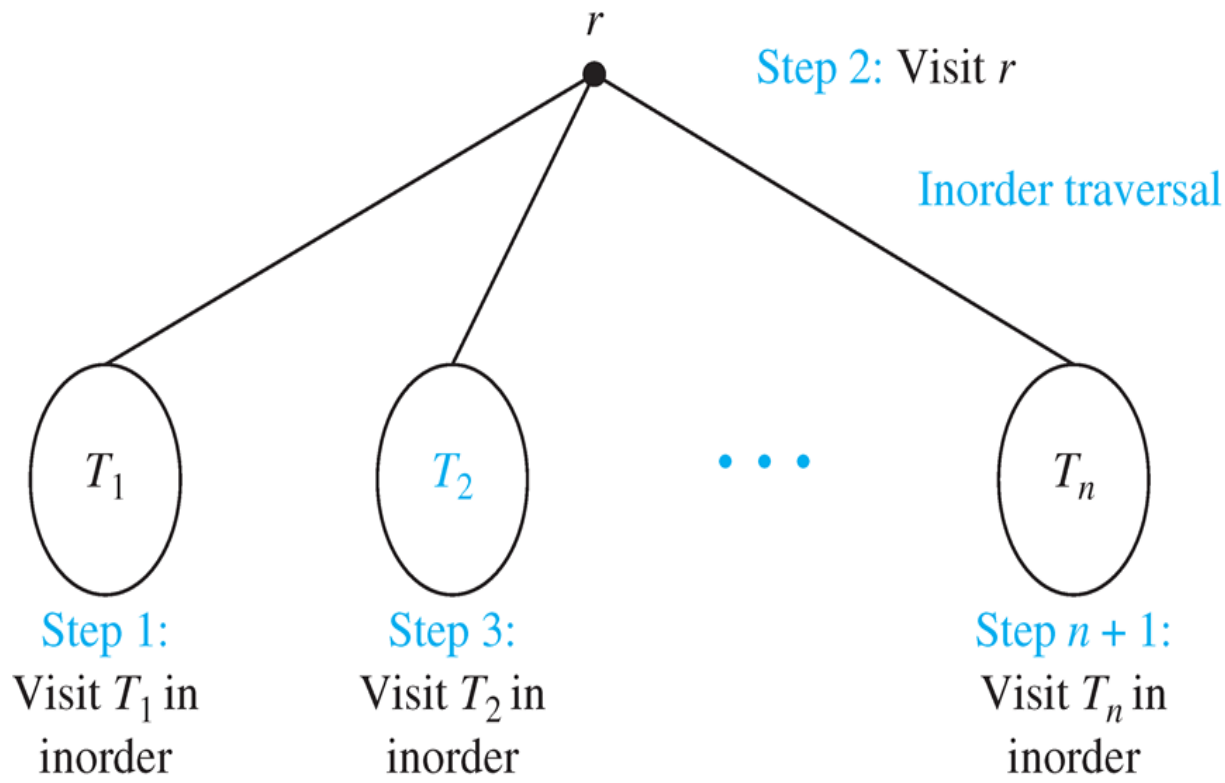
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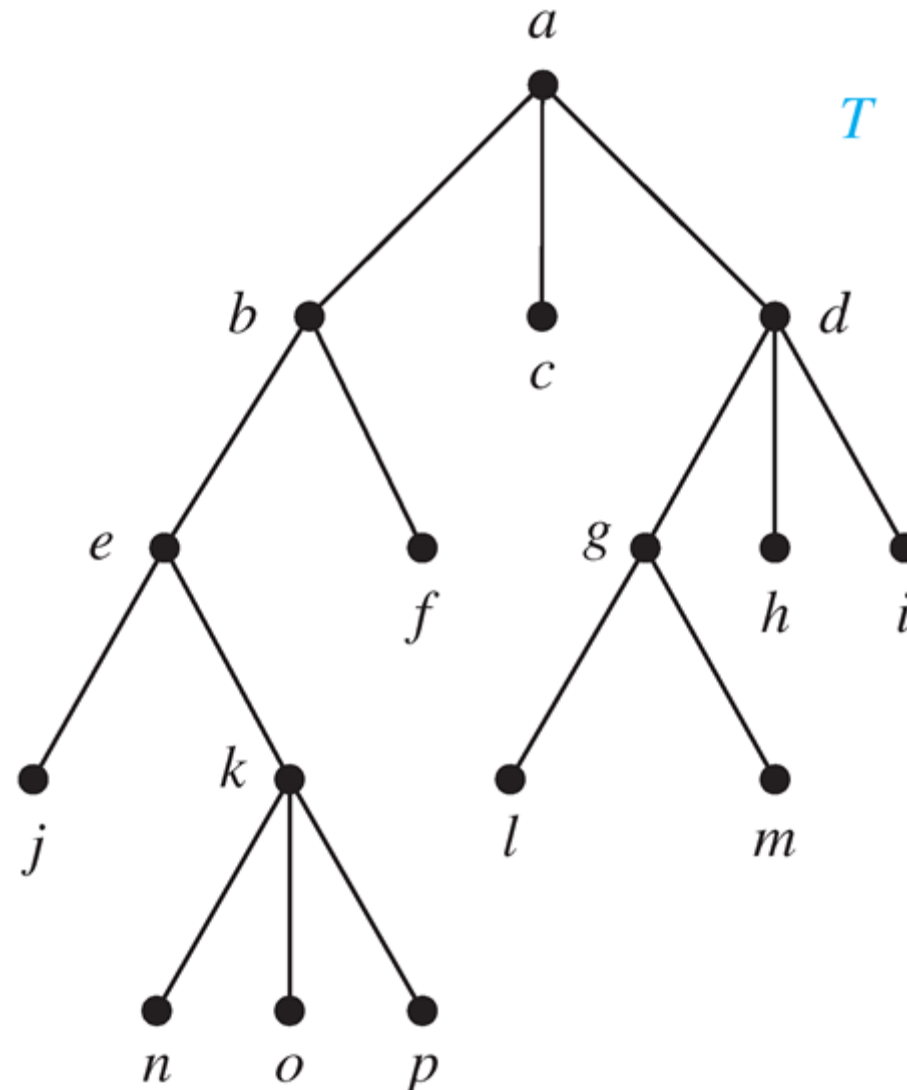
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Inorder Traversal

```
procedure inorder (T: ordered rooted tree)
  r := root of T
  if r is a leaf then list r
  else
    l := first child of r from left to right
    T(l) := subtree with l as its root
    inorder(T(l))
    list(r)
    for each child c of r from left to right
      T(c) := subtree with c as root
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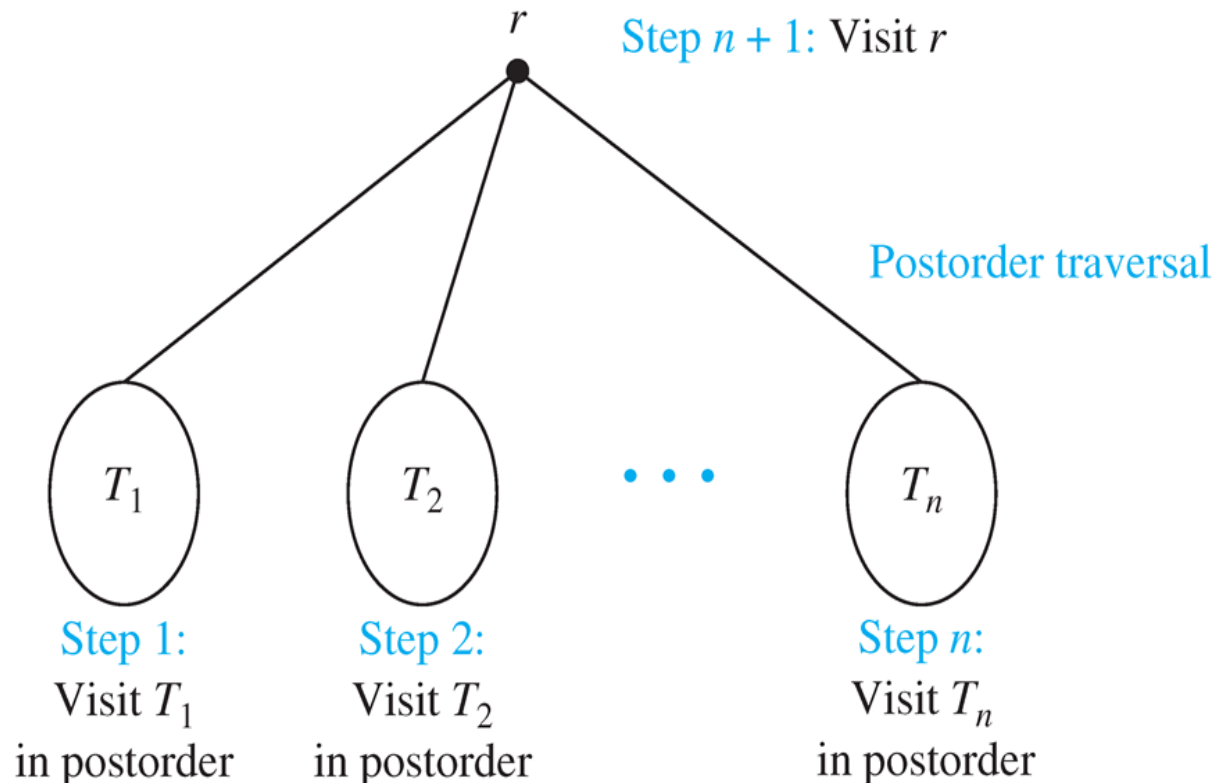

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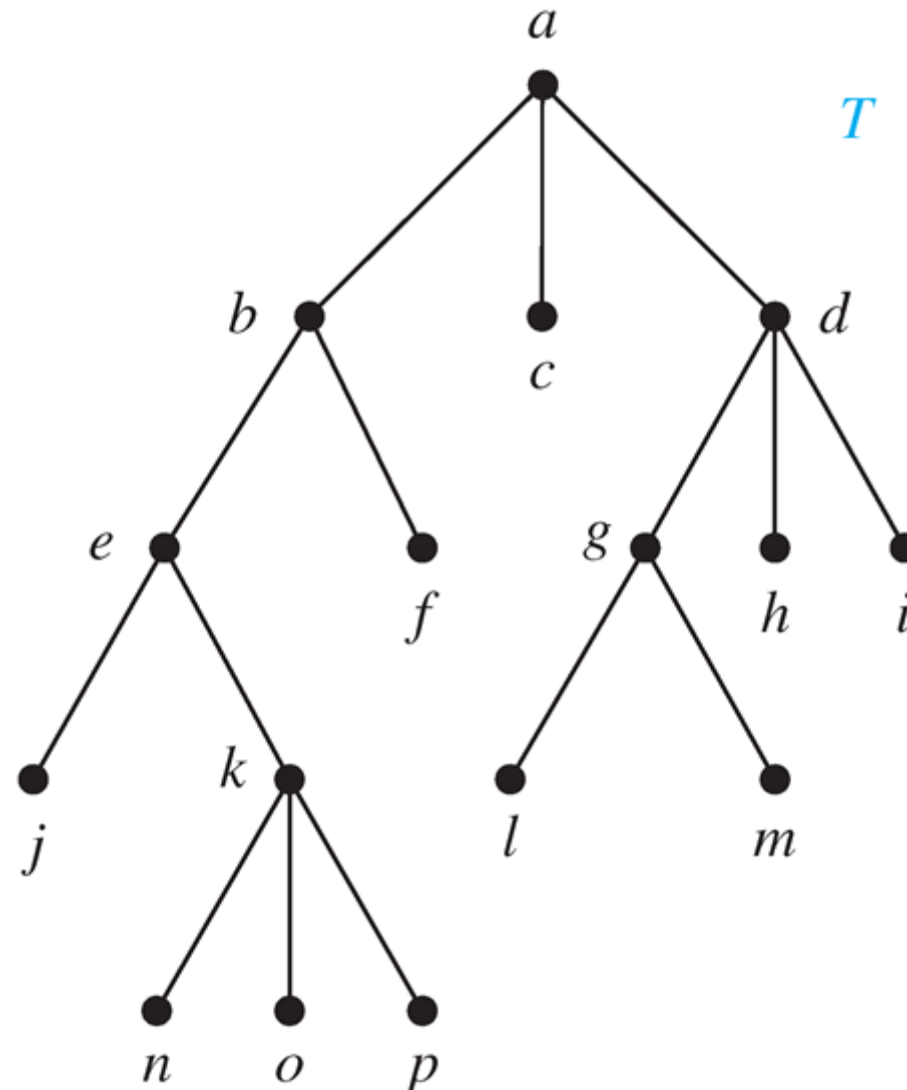
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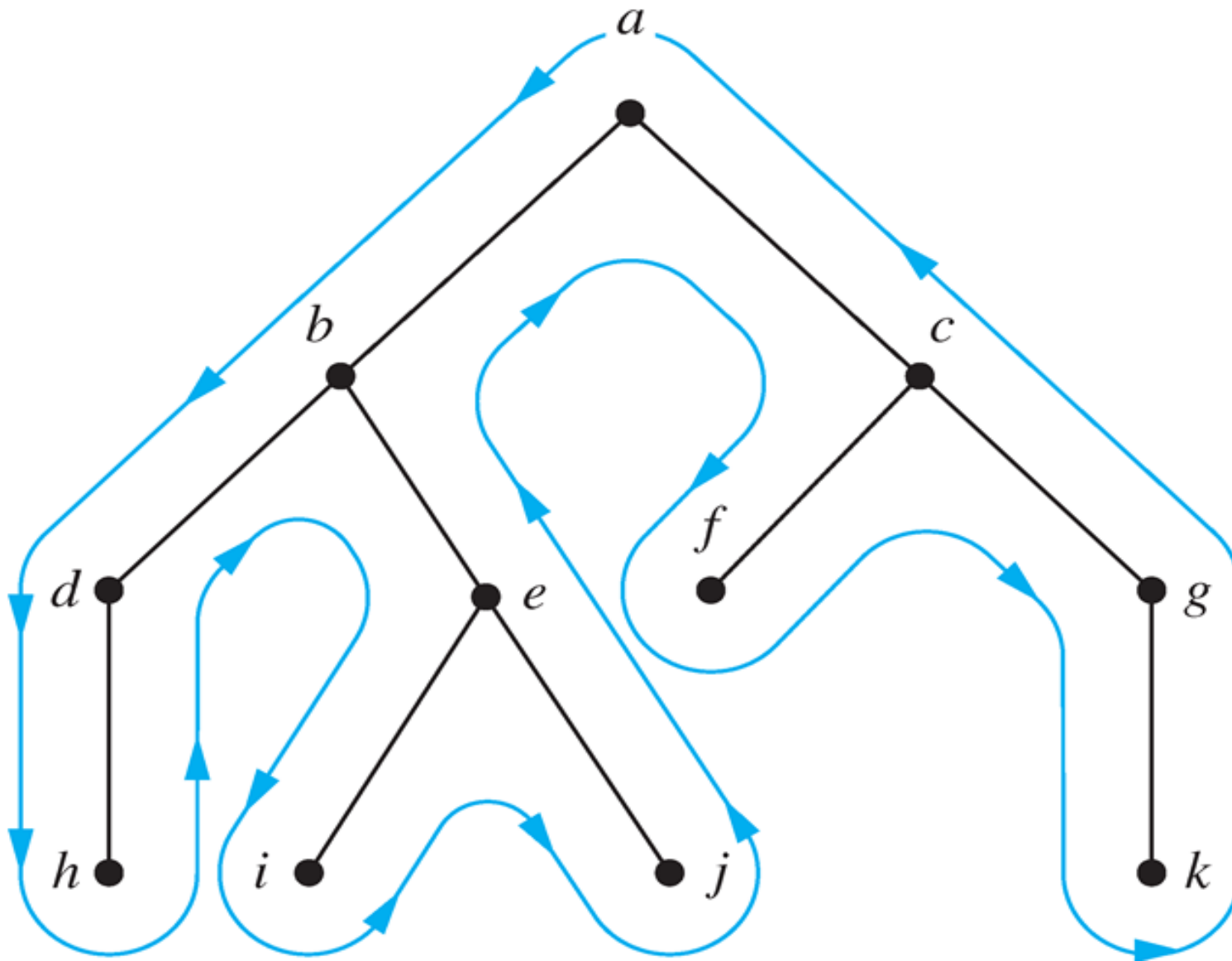


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Preorder, Inorder, Postorder Traversal



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Example

consider the expression $((x + y) \uparrow 2) + ((x - 4)/3)$

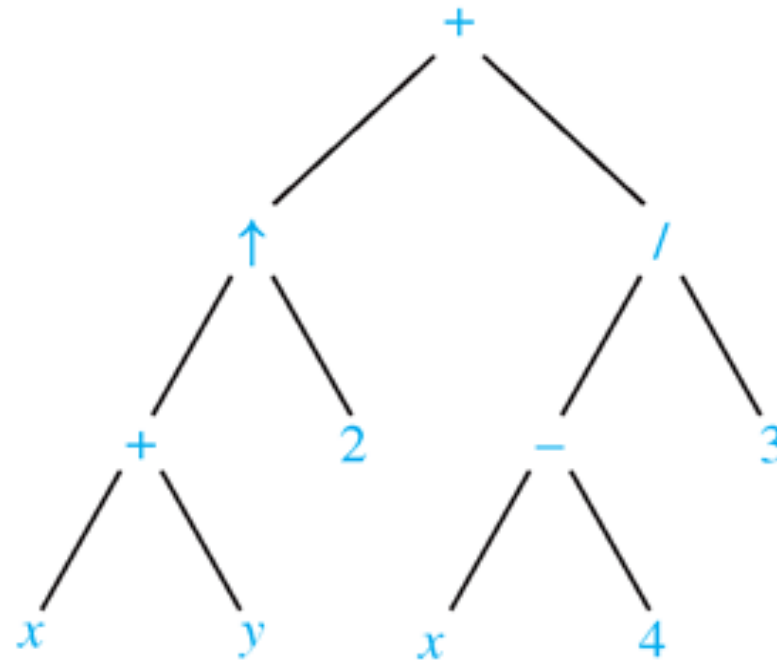


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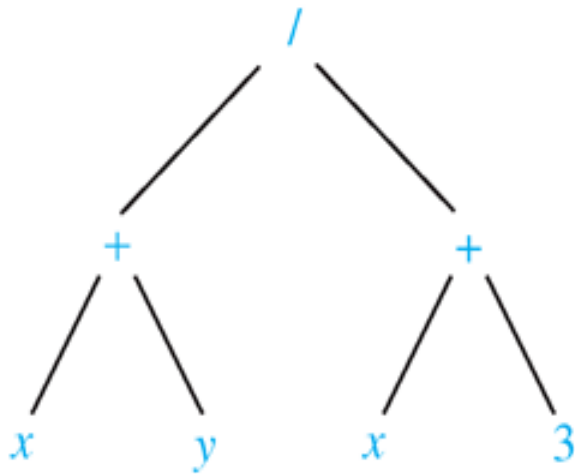
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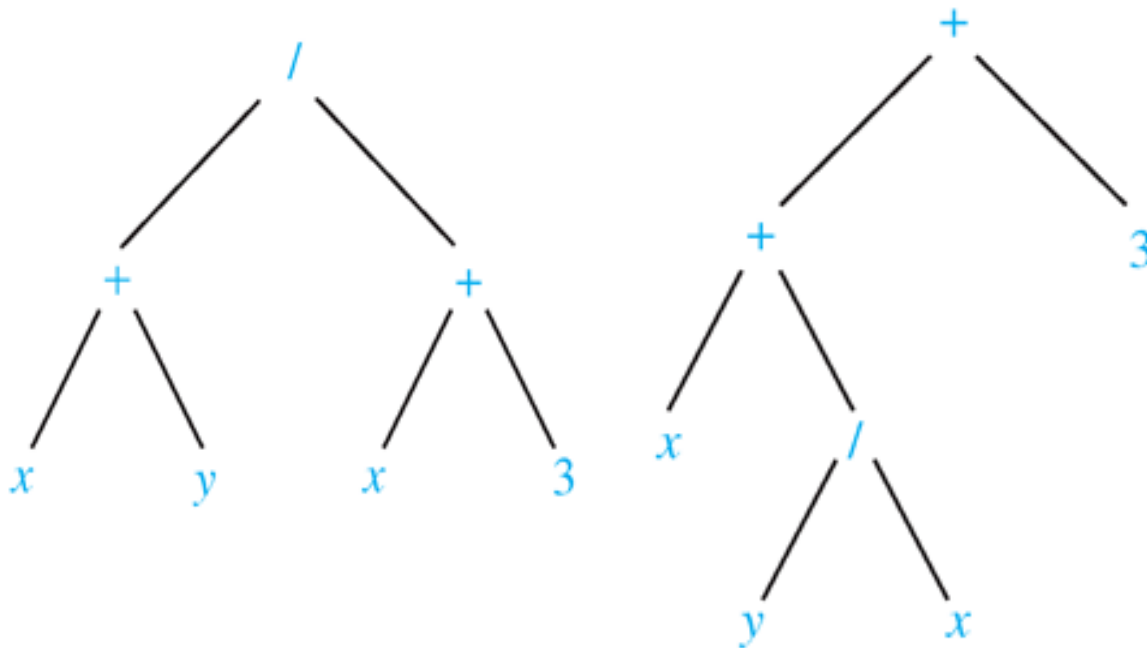
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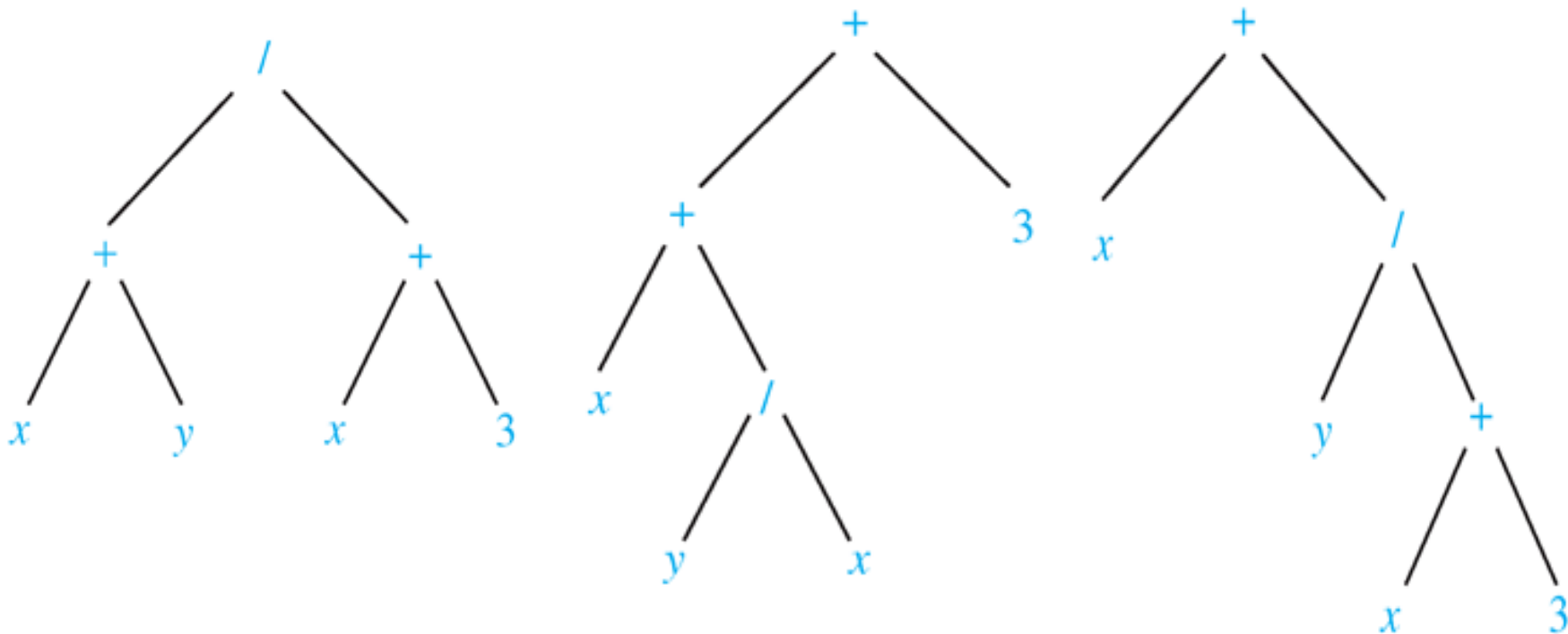
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Prefix expressions are evaluated by working *from right to left*. When we encounter an operator, we perform the operation with *the two operands to the right*.



Prefix Notation

■ Example

+ - * 2 3 5 / ↑ 2 3 4



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$$\begin{array}{ccccccccccc} + & - & * & 2 & 3 & 5 & / & \uparrow & 2 & 3 & 4 \\ & & & & & & & \underbrace{} & & & \\ & & & & & & & 2 \uparrow 3 = 8 & & & \end{array}$$

$$\begin{array}{ccccccccccc} + & - & * & 2 & 3 & 5 & / & 8 & 4 \\ & & & & & & \underbrace{} & & & & \\ & & & & & & 8 / 4 = 2 & & & & \end{array}$$

$$\begin{array}{ccccccccccc} + & - & * & 2 & 3 & 5 & 2 \\ & & \underbrace{} & & & & & & & & \\ & & 2 * 3 = 6 & & & & & & & & \end{array}$$

$$\begin{array}{ccccccccccc} + & - & 6 & 5 & 2 \\ & \underbrace{} & & & & & & & & & \\ & 6 - 5 = 1 & & & & & & & & & \end{array}$$

$$\begin{array}{ccccccc} + & 1 & 2 \\ \underbrace{} & & & & & & \\ 1 + 2 = 3 & & & & & & \end{array}$$

30 - 2



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Postfix Notation

■ Example

7 2 3 * - 4 ↑ 9 3 / +



Postfix Notation

■ Example

$$7\ 2\ 3\ * \ -\ 4\ \uparrow\ 9\ 3\ /\ +$$
$$7 \quad 2 \quad 3 \quad * \quad - \quad 4 \quad \uparrow \quad 9 \quad 3 \quad / \quad +$$

$$2 * 3 = 6$$

7 6 - 4 ↑ 9 3 / +

$$7 - 6 = 1$$

1 4 ↑ 9 3 / +

$$1^4 = 1$$

$$\begin{array}{r} 193 \\ \hline \end{array} +$$

$$9 / 3 = 3$$

$$\begin{array}{r} 1 \quad 3 \quad + \\ \hline \end{array}$$

$$1 + 3 = 4$$



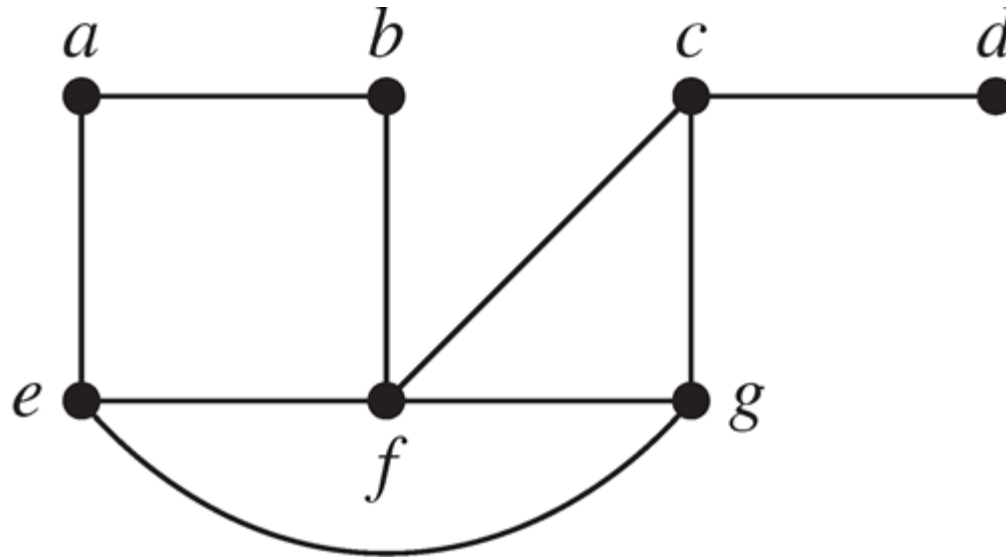
Spanning Trees

- **Definition** Let G be a simple graph. A *spanning tree* of G is a subgraph of G that is a tree containing every vertex of G .



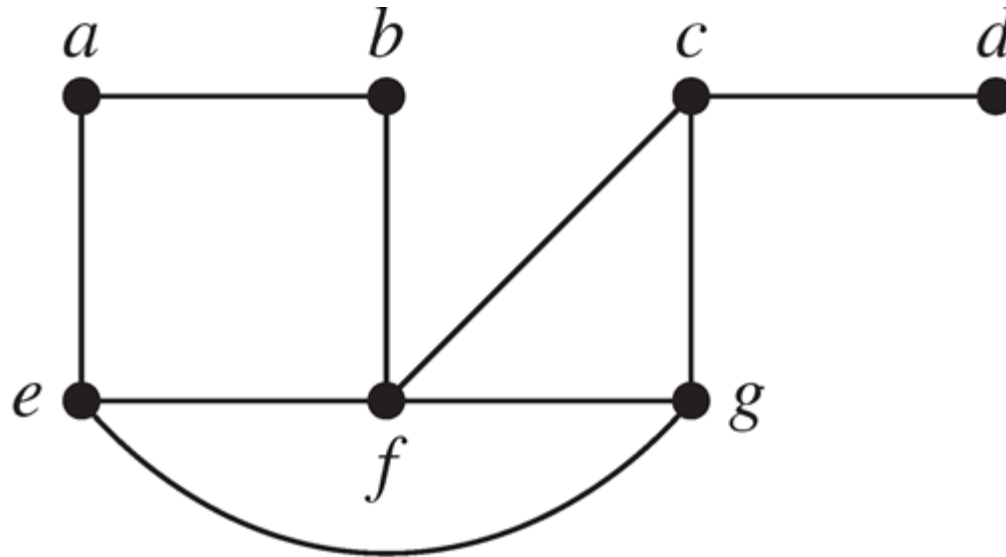
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remove edges to avoid circuits

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- **Theorem** A simple graph is **connected** if and only if it has a **spanning tree**.



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easy



Depth-First Search

- We can find **spanning trees** by **removing edges from simple circuits**.



Depth-First Search

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But, this is **inefficient**, since simple circuits should be identified **first**.



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- ◇ Form a path by **successively adding vertices and edges**. Continue adding to this path **as long as possible**.



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- ◇ If the path goes through all vertices of the graph, **the tree is a spanning tree**.



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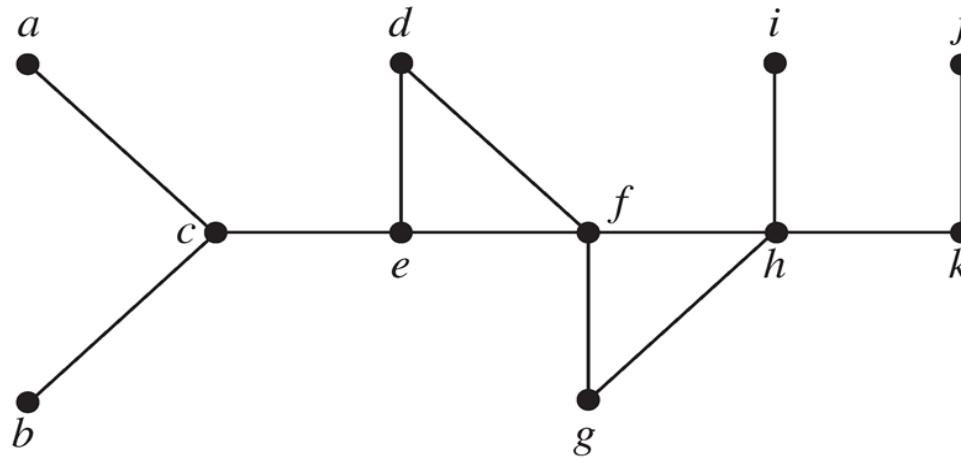
Instead, we build up **spanning trees** by **successively adding edges**.

- ◇ First arbitrarily choose a vertex of the graph as the root.
- ◇ Form a path by **successively adding vertices and edges**. Continue adding to this path **as long as possible**.
- ◇ If the path goes through all vertices of the graph, **the tree is a spanning tree**.
- ◇ Otherwise, **move back to some vertex** to repeat this procedure (*backtracking*)



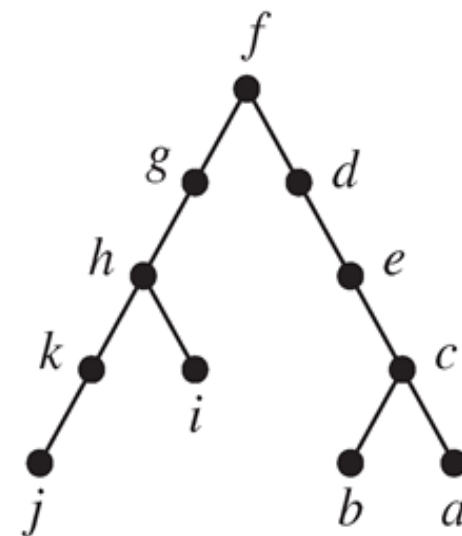
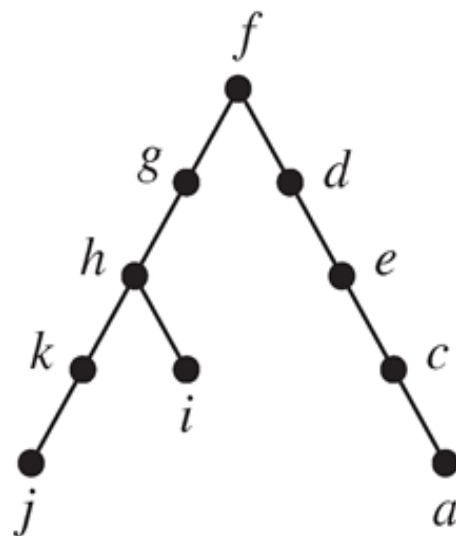
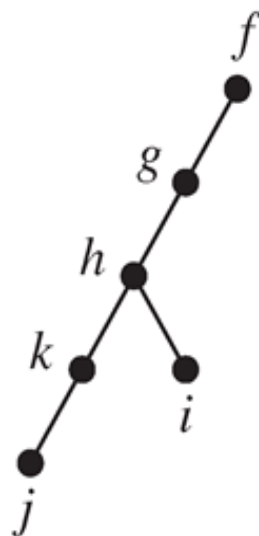
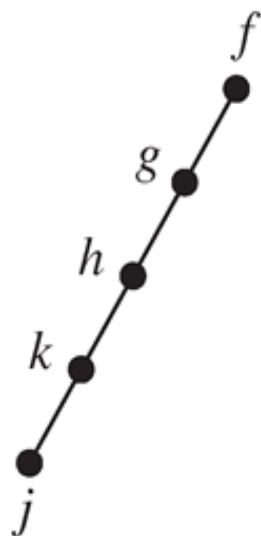
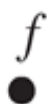
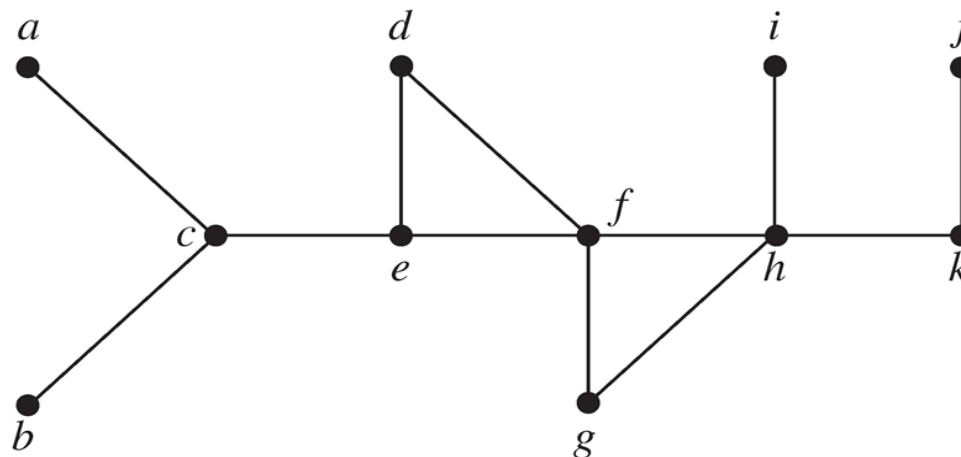
Depth-First Search

■ Example



Depth-First Search

■ Example



Depth-First Search Algorithm

```
procedure DFS( $G$ : connected graph with vertices  $v_1, v_2, \dots, v_n$ )  
 $T :=$  tree consisting only of the vertex  $v_1$   
visit( $v_1$ )
```

```
procedure visit( $v$ : vertex of  $G$ )  
for each vertex  $w$  adjacent to  $v$  and not yet in  $T$   
    add vertex  $w$  and edge  $\{v, w\}$  to  $T$   
    visit( $w$ )
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time complexity: $O(e)$



Breadth-First Search

- This is the **second** algorithm that we build up **spanning trees** by **successively adding edges**.



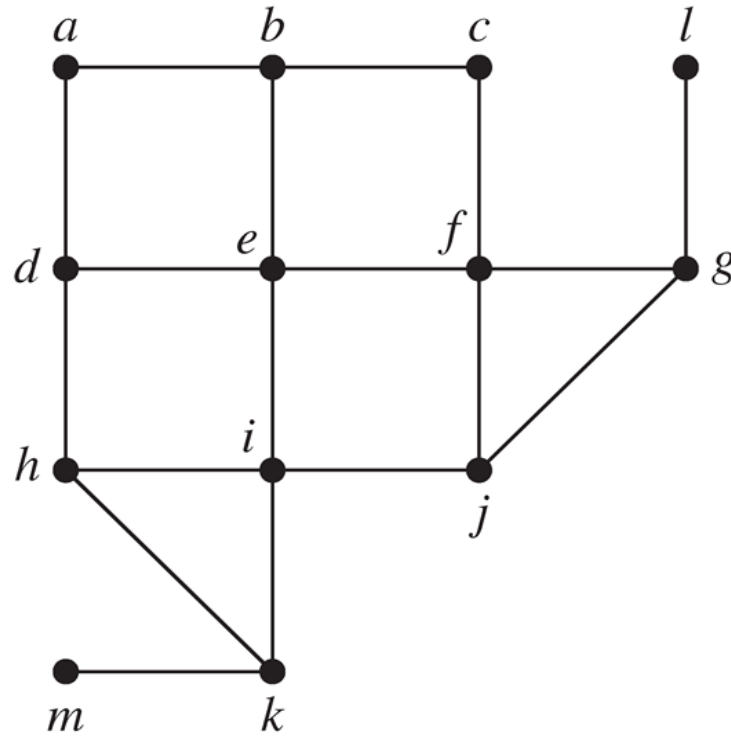
Breadth-First Search

- This is the **second** algorithm that we build up **spanning trees** by **successively adding edges**.
 - ◇ First arbitrarily choose a vertex of the graph as the root.
 - ◇ Form a path by **adding all edges incident to this vertex and the other endpoint of each of these edges**
 - ◇ For each vertex added at the **previous level**, **add edge incident to this vertex**, as long as it does **not** produce a simple circuit.
 - ◇ Continue in this manner until **all vertices have been added**.



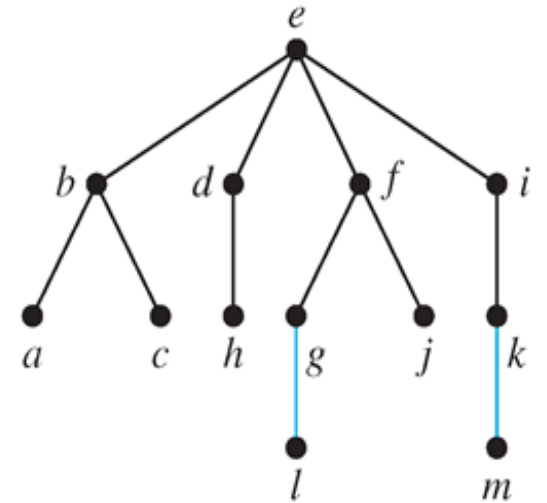
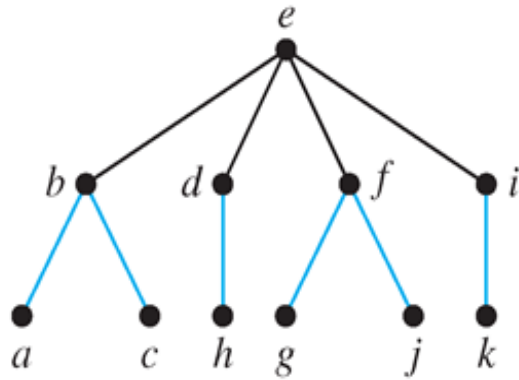
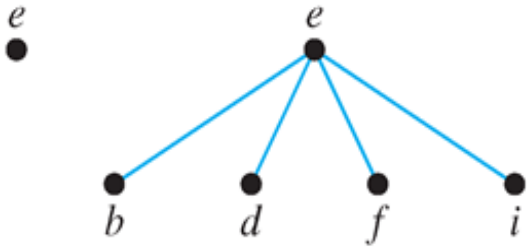
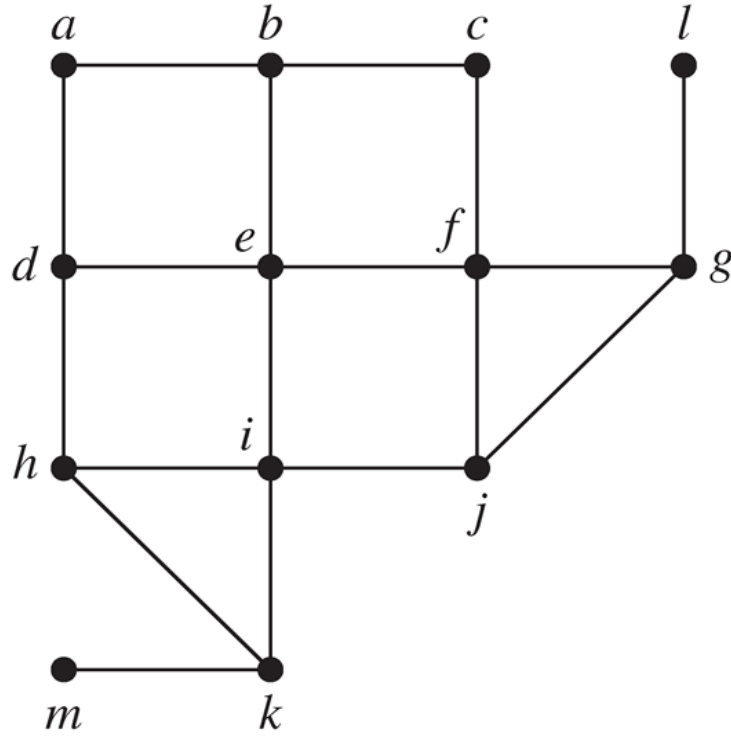
Breadth-First Search

■ Example



Breadth-First Search

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Breadth-First Search

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procedure BFS(G: connected graph with vertices  $v_1, v_2, \dots, v_n$ )  
   $T :=$  tree consisting only of the vertex  $v_1$   
   $L :=$  empty list visit( $v_1$ )  
  put  $v_1$  in the list  $L$  of unprocessed vertices  
  while  $L$  is not empty  
    remove the first vertex,  $v$ , from  $L$   
    for each neighbor  $w$  of  $v$   
      if  $w$  is not in  $L$  and not in  $T$  then  
        add  $w$  to the end of the list  $L$   
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Breadth-First Search

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Applications of DFS, BFS

- find paths, circuits, connected components, cut vertices, ...



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- find shortest paths, determine whether bipartite, ...

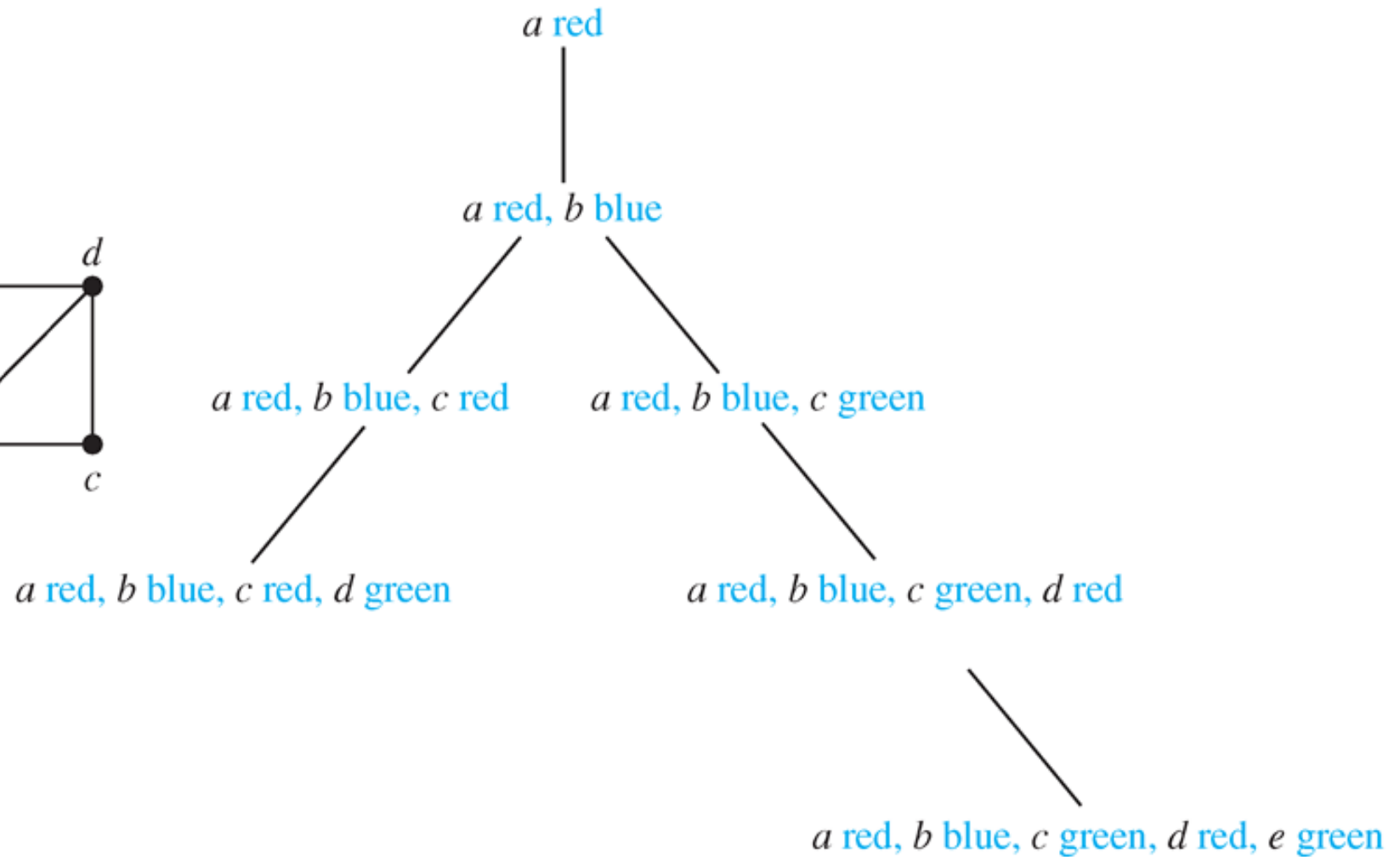
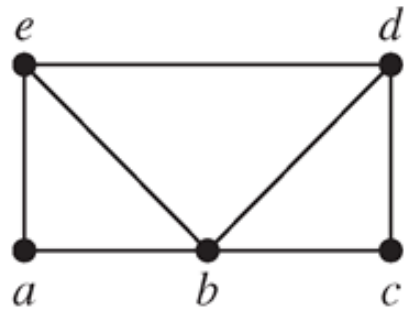


Applications of DFS, BFS

- find paths, circuits, connected components, cut vertices, ...
- find shortest paths, determine whether bipartite, ...
- graph coloring, sums of subsets, ...



Applications of DFS, BFS

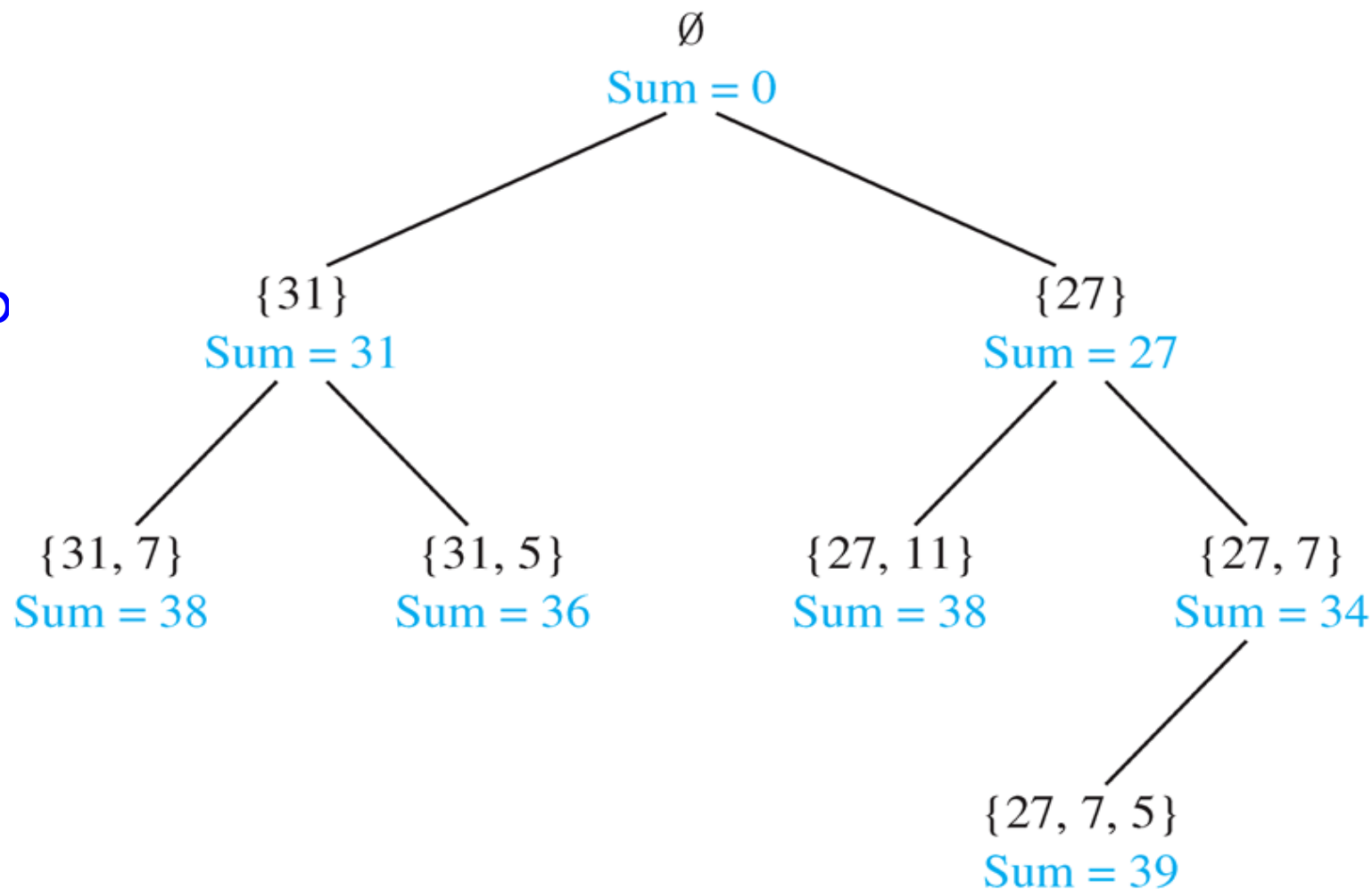


Applications of DFS, BFS

- find paths, circuits, connected components, cut vertices, ...

find

graph



find a subset of $\{31, 27, 15, 11, 7, 5\}$ with the sum 39



Minimum Spanning Trees

- **Definition** A *minimum spanning tree* in a connected weighted graph is a spanning tree that has the **smallest possible sum of weights** of its edges.



Minimum Spanning Trees

- **Definition** A *minimum spanning tree* in a connected weighted graph is a spanning tree that has the **smallest possible sum of weights** of its edges.

two **greedy algorithms**: Prim's Algorithm, Kruscal's Algorithm



Prim's Algorithm

ALGORITHM 1 Prim's Algorithm.

```
procedure Prim( $G$ : weighted connected undirected graph with  $n$  vertices)  
   $T :=$  a minimum-weight edge  
  for  $i := 1$  to  $n - 2$   
     $e :=$  an edge of minimum weight incident to a vertex in  $T$  and not forming a  
      simple circuit in  $T$  if added to  $T$   
     $T := T$  with  $e$  added  
  return  $T$  { $T$  is a minimum spanning tree of  $G$ }
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Prim's Algorithm

ALGORITHM 1 Prim's Algorithm.

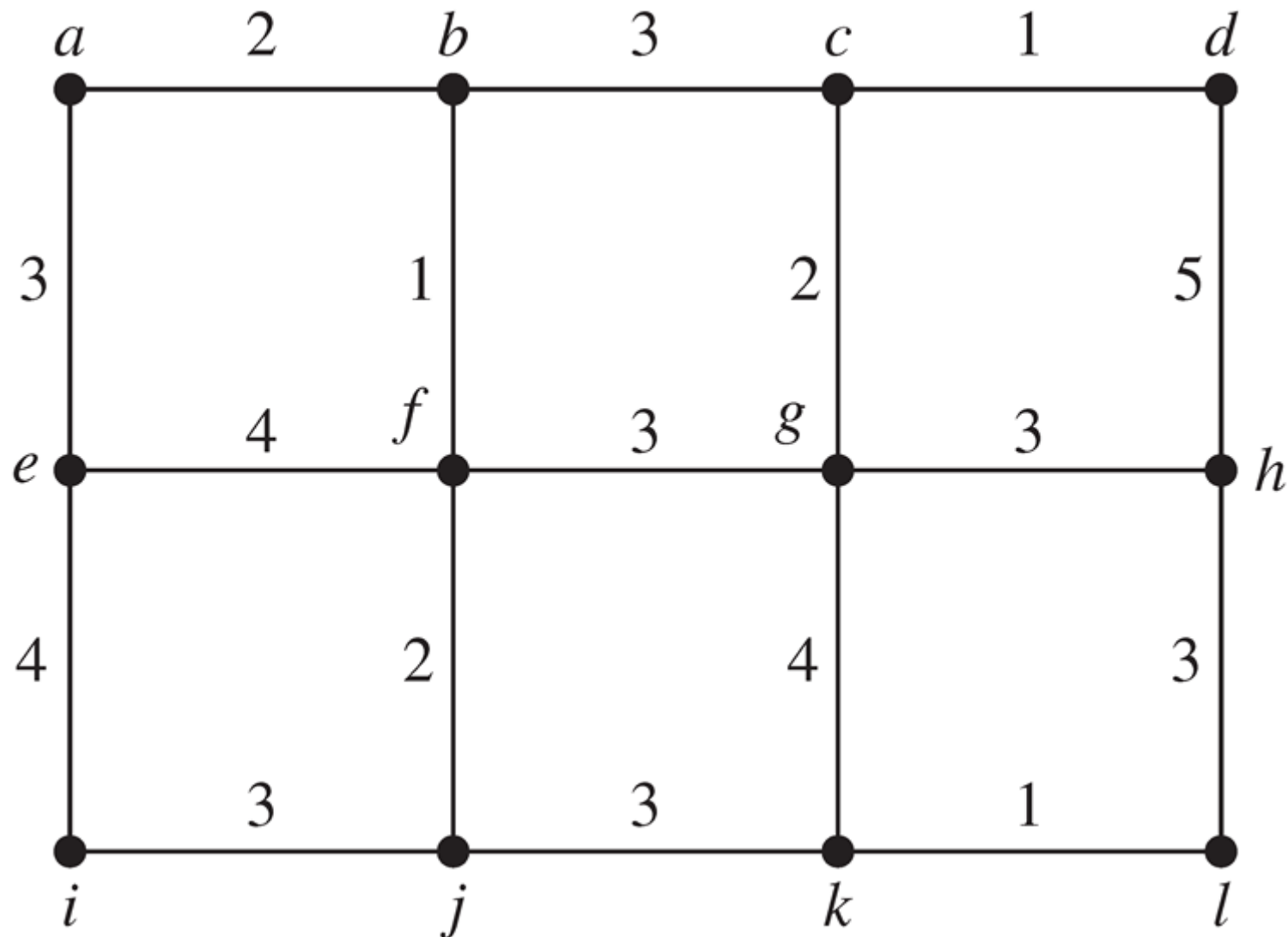
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time complexity: $e \log v$



Prim's Algorithm

■ Example



Kruskal's Algorithm

ALGORITHM 2 Kruskal's Algorithm.

```
procedure Kruskal( $G$ : weighted connected undirected graph with  $n$  vertices)  
 $T$  := empty graph  
for  $i$  := 1 to  $n - 1$   
     $e$  := any edge in  $G$  with smallest weight that does not form a simple circuit  
        when added to  $T$   
     $T$  :=  $T$  with  $e$  added  
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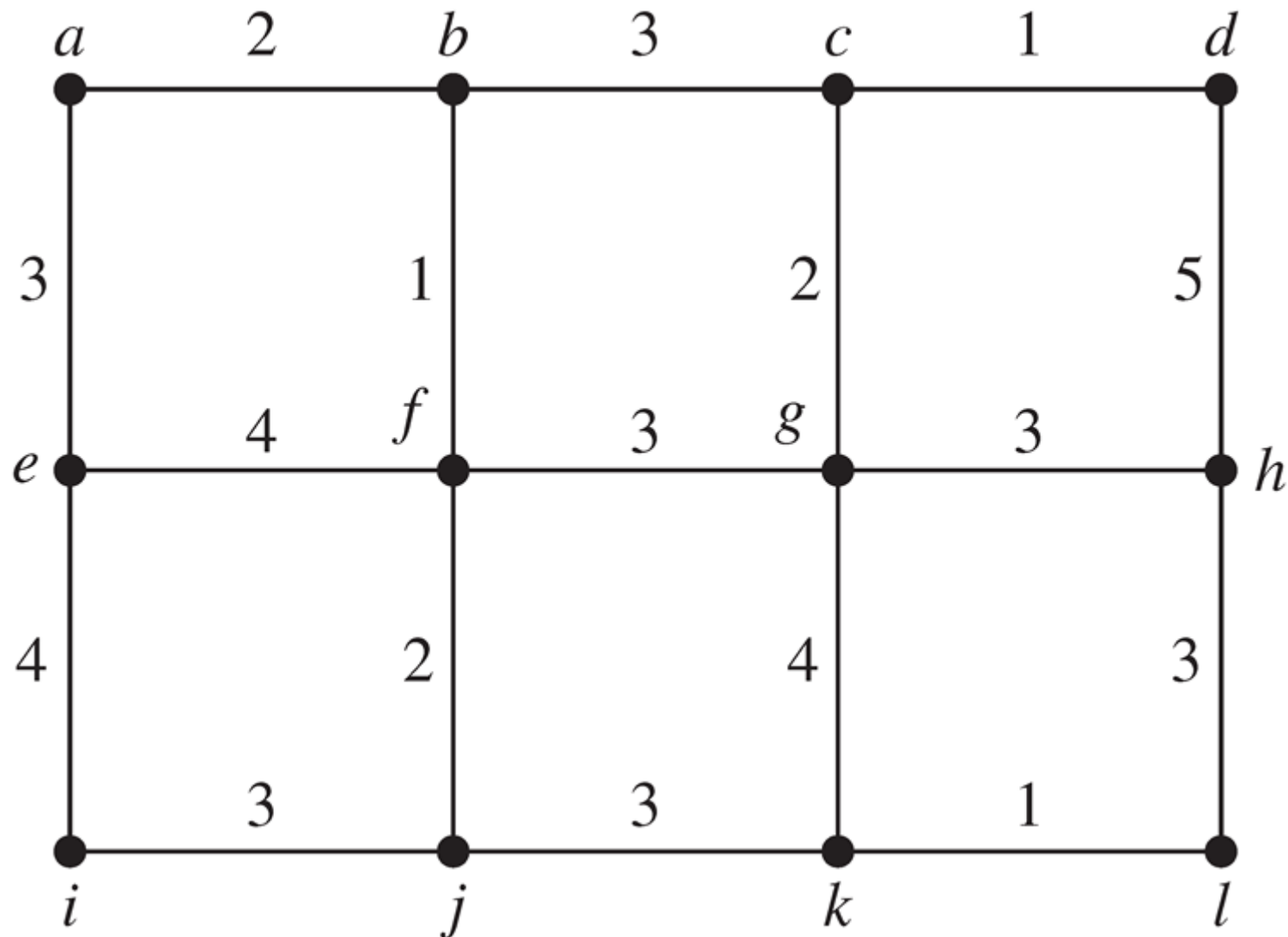
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see *CLRS / Algorithm Design*, J. Kleinberg, E. Tardos



Kruskal's Algorithm

■ Example



Next Lecture

- course review ...

