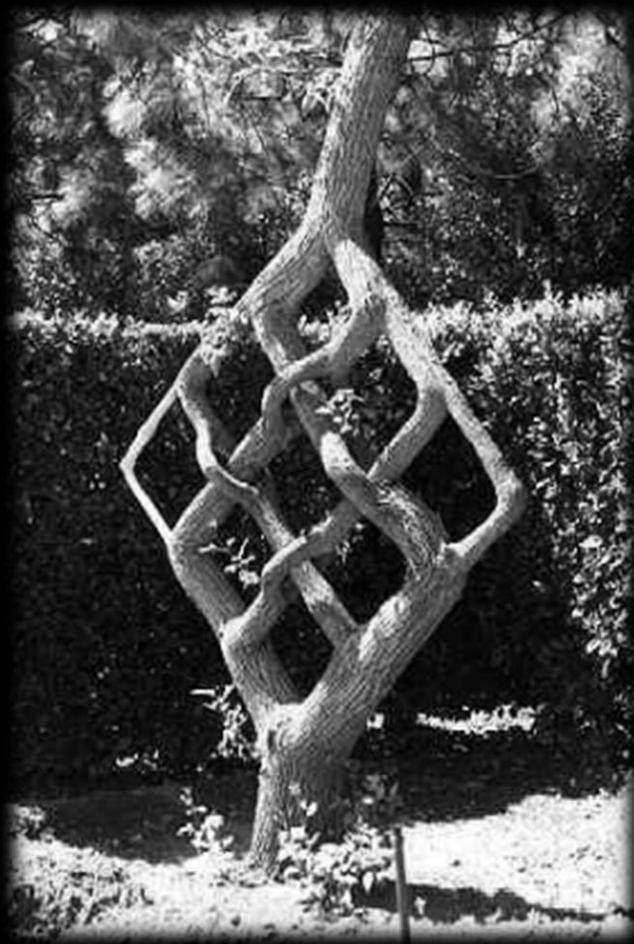


# INTRODUCTION TO TREES



**COMP 251**

# Topic Goals

- To learn the basic definitions and properties of trees.

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- To develop skill at representing trees as graphical objects and as ordered triples.
- **To discuss graphs and trees as data structures.**

# Recall: What is Graph Theory?

- The study of graphs.
  - Informally, graphs are mathematical structures used to model binary relations between elements of a given set.

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- The origin of the theory is credited to Euler, 1736 – The Königsberg Bridge Problem.

---

**Definition 11.1** Let  $V$  be a finite nonempty set, and let  $E \subseteq V \times V$ . The pair  $(V, E)$  is then called a *directed graph* (on  $V$ ), or *digraph*<sup>†</sup> (on  $V$ ), where  $V$  is the set of *vertices*, or *nodes*, and  $E$  is its set of *edges*. We write  $G = (V, E)$  to denote such a digraph.

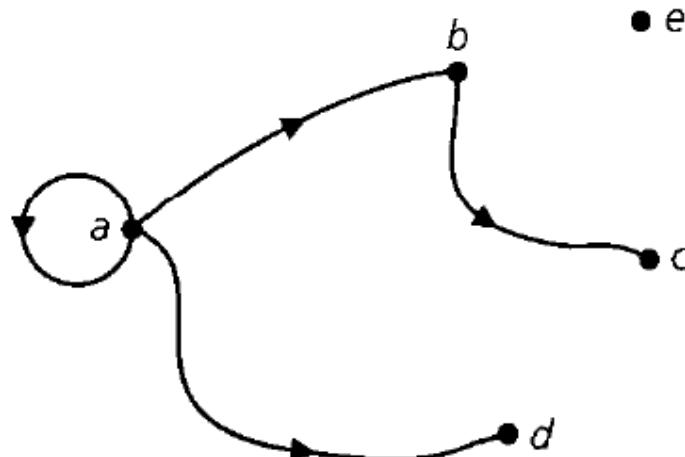
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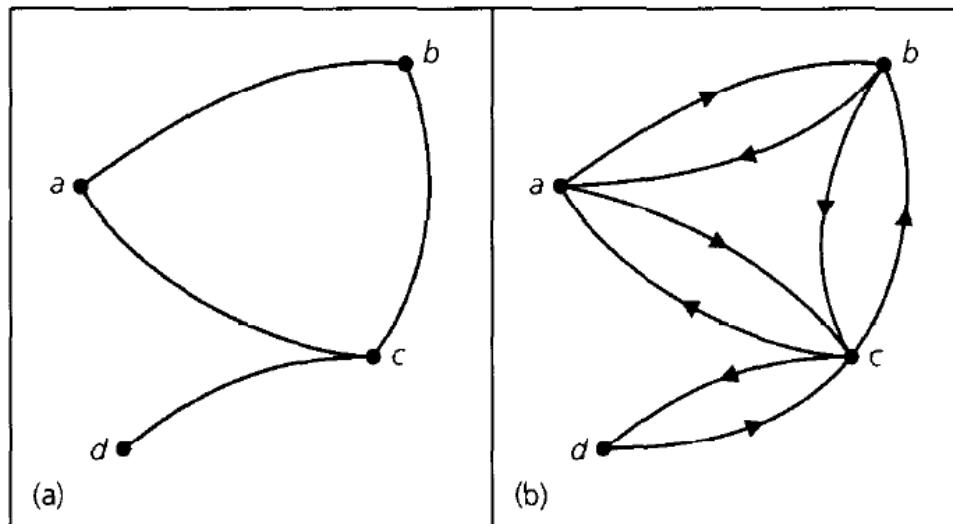
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Figure 11.2 provides an example of a directed graph on  $V = \{a, b, c, d, e\}$  with  $E = \{(a, a), (a, b), (a, d), (b, c)\}$ . The direction of an edge is indicated by placing a directed arrow on the edge, as shown here. For any edge, such as  $(b, c)$ , we say that the edge is *incident* with the vertices  $b, c$ ;  $b$  is said to be *adjacent to*  $c$ , whereas  $c$  is *adjacent from*  $b$ . In addition, vertex  $b$  is called the *origin*, or *source*, of the edge  $(b, c)$ , and vertex  $c$  is the *terminus*, or *terminating vertex*. The edge  $(a, a)$  is an example of a *loop*, and the vertex  $e$  that has no incident edges is called an *isolated vertex*.



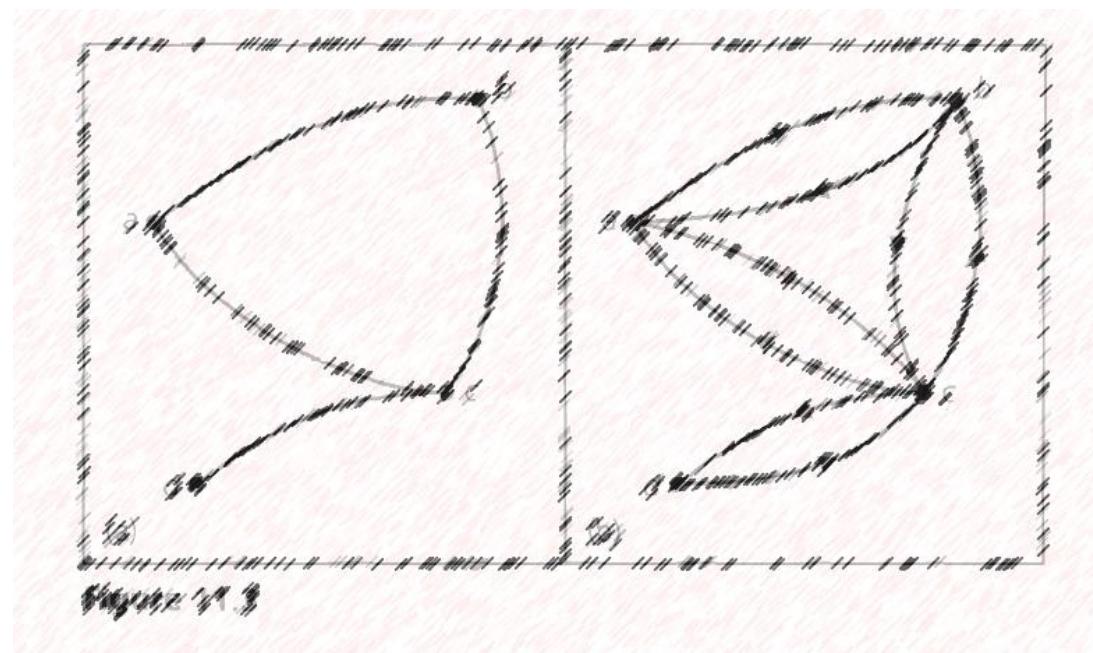
**Figure 11.2**

When there is no concern about the direction of any edge, the structure  $G = (V, E)$ , where  $E$  is now a set of unordered pairs from  $V$ , is called an *undirected graph*. An undirected graph is shown in Fig. 11.3(a). This graph is a more compact way of describing the directed graph given in Fig. 11.3(b). In an undirected graph, there are undirected edges such as  $\{a, b\}, \{b, c\}, \{a, c\}, \{c, d\}$  in Fig. 11.3(a). An edge such as  $\{a, b\}$  stands for  $\{(a, b), (b, a)\}$ . Although  $(a, b) = (b, a)$  only when  $a = b$ , we do have  $\{a, b\} = \{b, a\}$  for any  $a, b$ . We can write  $\{a, a\}$  to denote a loop in an undirected graph, but  $\{a, a\}$  is considered the same as  $(a, a)$ .



**Figure 11.3**

In general, if a graph  $G$  is not specified as directed or undirected, it is assumed to be undirected. When it contains no loops it is called *loop-free*.



# Basic Terminology

A **graph**  $G(V, E)$  is a set  $V$  of **vertices** and a set  $E$  of **edges**. In an **undirected graph**, an edge is an unordered pair of vertices. An ordered pair of vertices is called a **directed edge**. If we allow **multi-sets** of edges, i.e. multiple edges between two vertices, we obtain a **multigraph**. A **self-loop** or **loop** is an edge between a vertex and itself. An undirected graph without loops or multiple edges is known as a **simple graph**. In this class we will assume graphs to be simple unless otherwise stated.

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If vertices  $a$  and  $b$  are endpoints of an edge, we say that they are **adjacent** and write  $a \sim b$ . If vertex  $a$  is one of edge  $e$ 's endpoints,  $a$  is **incident** to  $e$  and we write  $a \in e$ . The **degree** of a vertex is the number of edges incident to it.

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A **walk** is a sequence of vertices  $v_1, v_2, \dots, v_k$  such that  $\forall i \in 1, 2, \dots, k - 1, v_i \sim v_{i+1}$ . A **path** is a walk where  $v_i \neq v_j, \forall i \neq j$ . In other words, a path is a walk that visits each vertex at most once. A **closed walk** is a walk where  $v_1 = v_k$ . A **cycle** is a closed path, i.e. a path combined with the edge  $(v_k, v_1)$ . A graph is **connected** if there exists a path between each pair of vertices. A **tree** is a connected graph with no cycles. A **forest** is a graph where each connected component is a tree. A node in a forest with degree 1 is called a **leaf**.

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The size of a graph is the number of vertices of that graph. We usually denote the number of vertices with  $n$  and the number edges with  $m$ .

# Basic Terminology

---

**Definition 11.2** Let  $x, y$  be (not necessarily distinct) vertices in an undirected graph  $G = (V, E)$ . An  $x$ - $y$  walk in  $G$  is a (loop-free) finite alternating sequence

$$x = x_0, e_1, x_1, e_2, x_2, e_3, \dots, e_{n-1}, x_{n-1}, e_n, x_n = y$$

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- a) If no edge in the  $x$ - $y$  walk is repeated, then the walk is called an  $x$ - $y$  trail. A closed  $x$ - $x$  trail is called a circuit.
- b) When no vertex of the  $x$ - $y$  walk occurs more than once, the walk is called an  $x$ - $y$  path. The term cycle is used to describe a closed  $x$ - $x$  path.

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**Convention:** In dealing with circuits, we shall always understand the presence of at least one edge. When there is only one edge, then the circuit is a loop (and the graph is no longer loop-free). Circuits with two edges arise in multigraphs, a concept we shall define shortly.

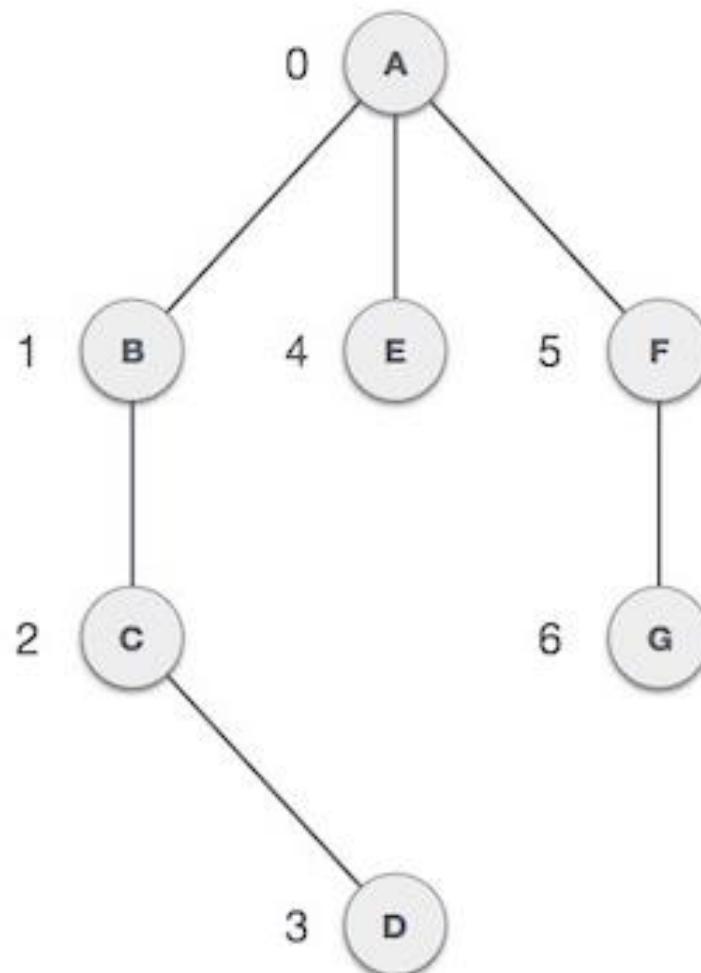
The term cycle will always imply the presence of at least three distinct edges (from the graph).

# Graph Data Structure

Mathematical graphs can be represented in data structure. We can represent a graph using an array of vertices and a two-dimensional array of edges. Before we proceed further, let's familiarize ourselves with some important terms:

- **Vertex** – Each node of the graph is represented as a vertex. In the following example, the labeled circle represents vertices. Thus, A to G are vertices. We can represent them using an array as shown in the following image. Here A can be identified by index 0. B can be identified using index 1 and so on.
- **Edge** – Edge represents a path between two vertices or a line between two vertices. In the following example, the lines from A to B, B to C, and so on represents edges. We can use a two-dimensional array to represent an array as shown in the following image. Here AB can be represented as 1 at row 0, column 1, BC as 1 at row 1, column 2 and so on, keeping other combinations as 0.
- **Adjacency** – Two node or vertices are adjacent if they are connected to each other through an edge. In the following example, B is adjacent to A, C is adjacent to B, and so on.
- **Path** – Path represents a sequence of edges between the two vertices. In the following example, ABCD represents a path from A to D.

# Graph Data Structure



# Graph Data Structure

The following are the basic primary operations of a Graph:

- **Add Vertex** – Adds a vertex to the graph.
- **Add Edge** – Adds an edge between the two vertices of the graph.
- **Display Vertex** – Displays a vertex of the graph.

A photograph of a dense forest of tall evergreen trees, likely Douglas firs, growing on a hillside. The trees are closely packed, with their dark green needles contrasting against the bright sunlight. The forest floor is covered in smaller shrubs and ferns. The perspective is from a low angle, looking up at the towering trunks.

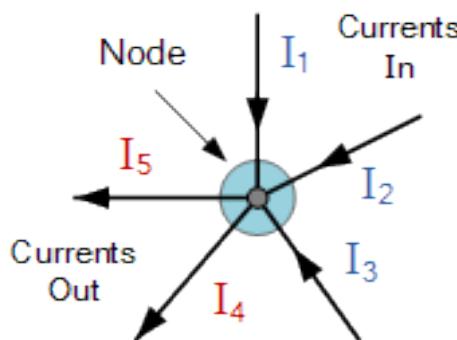
# Trees

# Trees

Continuing our study of graph theory, we shall now focus on a special type of graph called a tree. First used in 1847 by Gustav Kirchhoff (1824–1887) in his work on electrical networks, trees were later redeveloped and named by Arthur Cayley (1821–1895). In 1857 Cayley used these special graphs in order to enumerate the different isomers of the saturated hydrocarbons  $C_n H_{2n+2}$ ,  $n \in \mathbb{Z}^+$ .

With the advent of digital computers, many new applications were found for trees. Special types of trees are prominent in the study of data structures, sorting, and coding theory, and in the solution of certain optimization problems.

Currents Entering the Node  
Equals  
Currents Leaving the Node



$$I_1 + I_2 + I_3 + (-I_4 + -I_5) = 0$$

# Recall the Game Trees of Computational Complexity

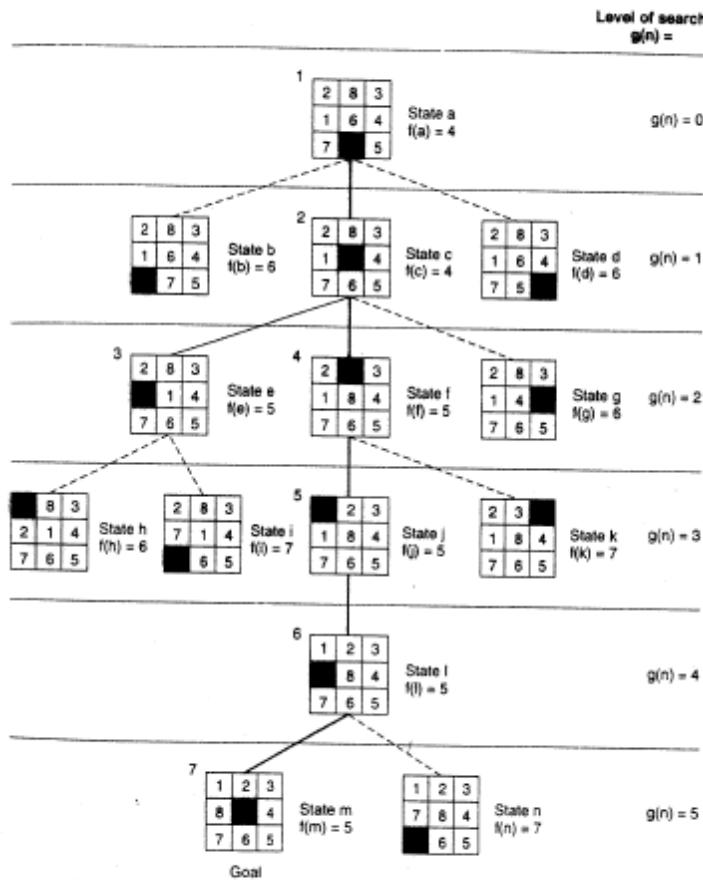
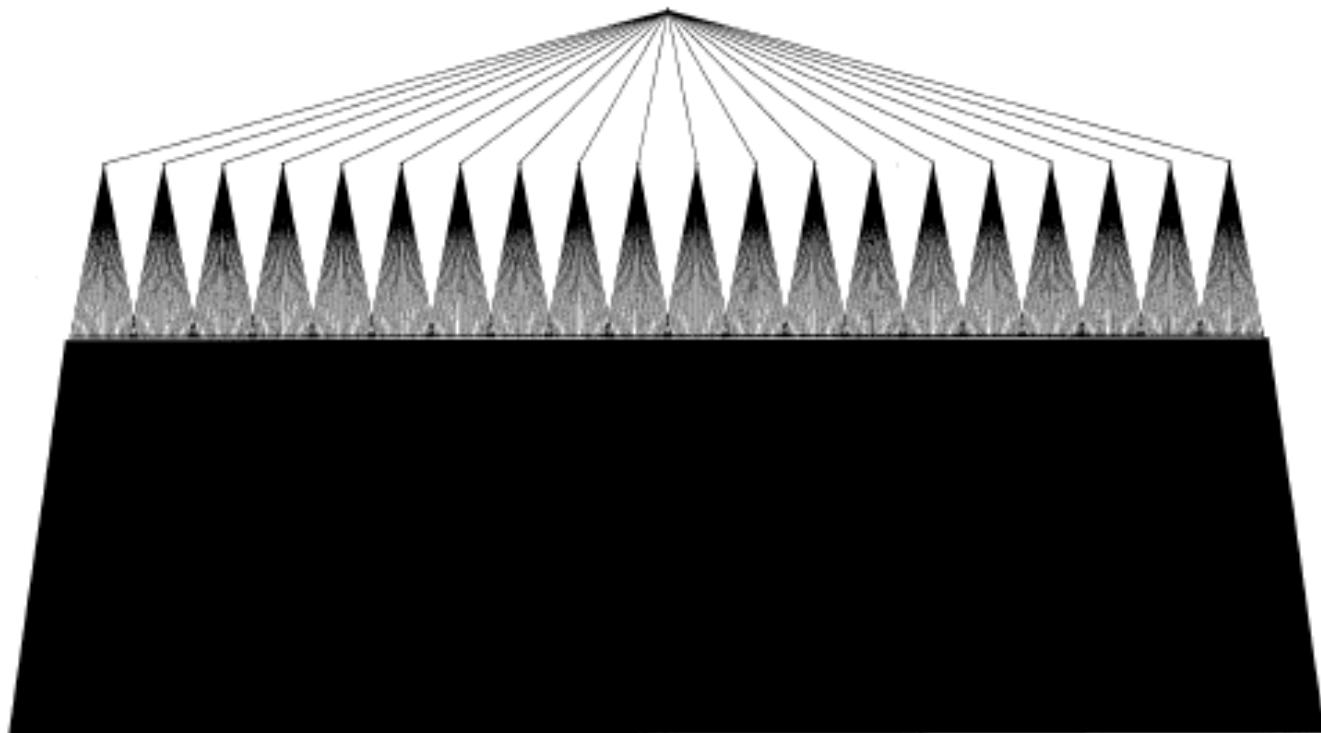


Figure 4.10 State space generated in heuristic search of the 8-puzzle graph.

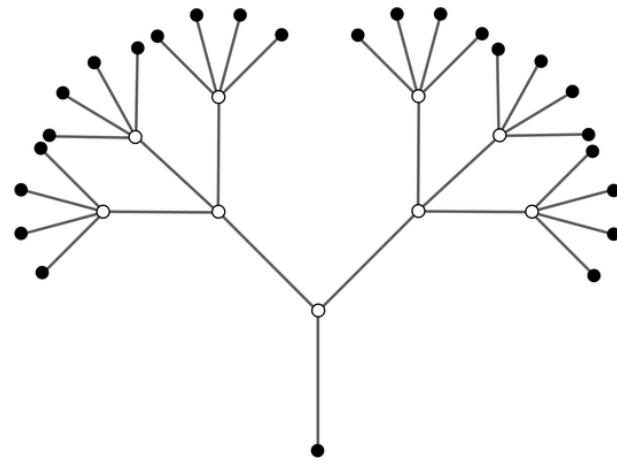
# Recall the Game Trees of Computational Complexity

A Graph of the States in the First Three Moves of a game of Chess



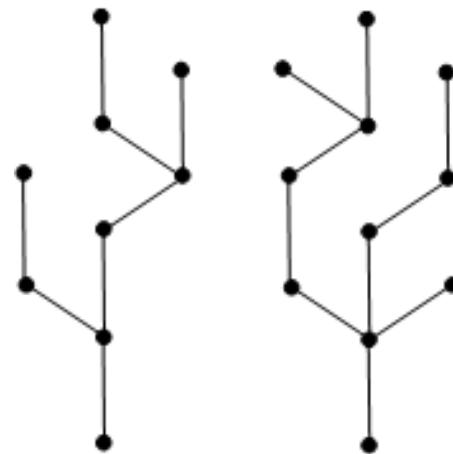
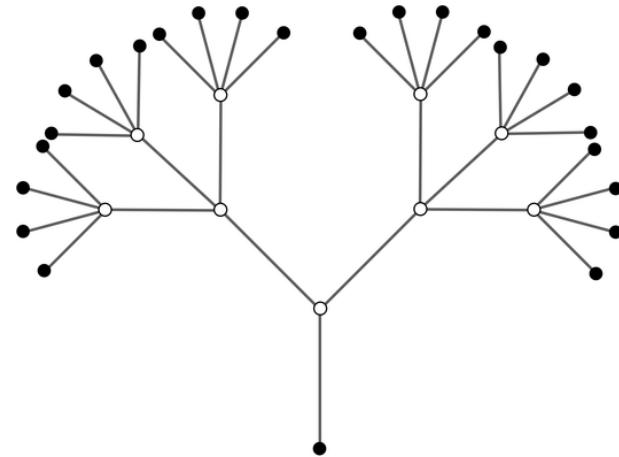
## Trees

In mathematics, and more specifically in graph theory, a **tree** is an undirected graph in which any two vertices are connected by exactly one simple path. In other words, any connected graph without simple cycles is a tree.



# Trees

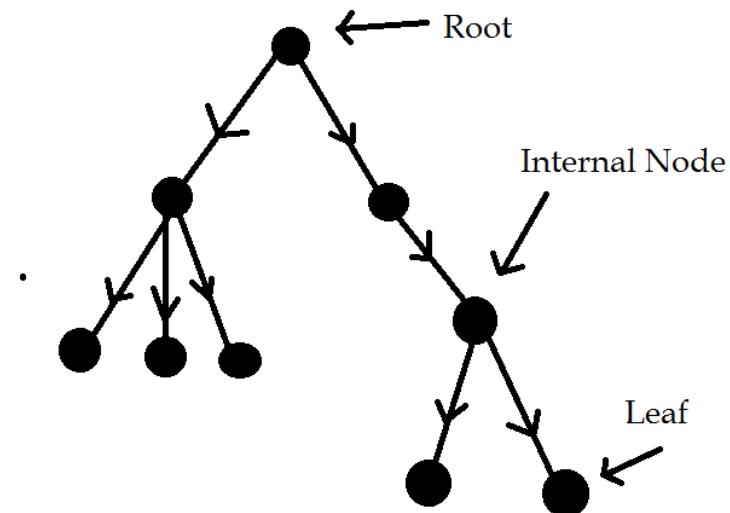
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The various kinds of data structures referred to as trees in computer science are equivalent as undirected graphs to trees in graph theory, although such data structures are generally **rooted trees**, thus in fact being directed graphs, and may also have additional ordering of branches.

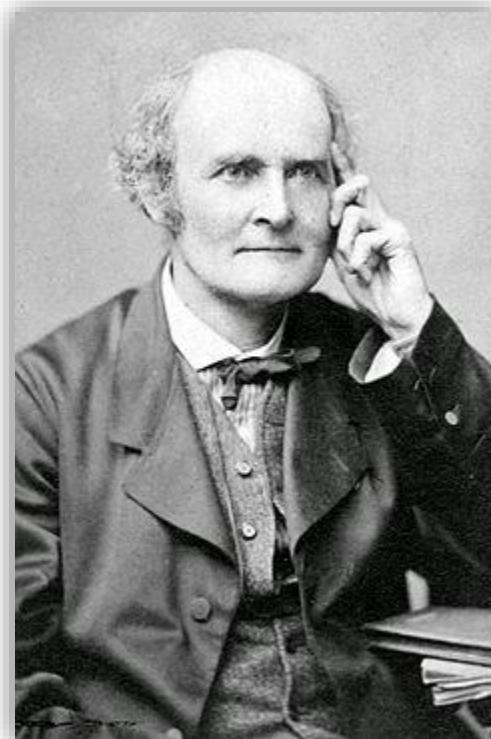


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The term "tree" was coined in 1857 by the British mathematician Arthur Cayley.<sup>[1]</sup>



# [http://en.wikipedia.org/wiki/Glossary\\_of\\_graph\\_theory](http://en.wikipedia.org/wiki/Glossary_of_graph_theory)

## Trees [edit]

A **tree** is a connected acyclic simple graph. For directed graphs, each vertex has at most one incoming edge. A vertex of degree 1 is called a **leaf**, or *pendant vertex*. An edge incident to a leaf is a **leaf edge**, or *pendant edge*. (Some people define a leaf edge as a *leaf* and then define a *leaf vertex* on top of it. These two sets of definitions are often used interchangeably.) A non-leaf vertex is an **internal vertex**. Sometimes, one vertex of the tree is distinguished, and called the **root**; in this case, the tree is called **rooted**. Rooted trees are often treated as **directed acyclic graphs** with the edges pointing away from the root.

A **subtree** of the tree  $T$  is a connected subgraph of  $T$ .

A **forest** is an acyclic simple graph. For directed graphs, each vertex has at most one incoming edge. (That is, a tree with the connectivity requirement removed; a graph containing multiple disconnected trees.)

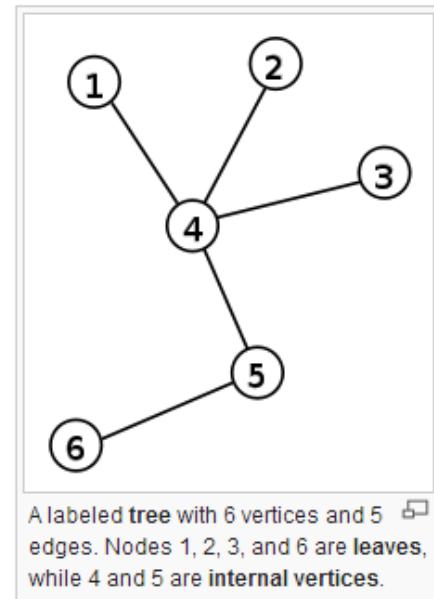
A **subforest** of the forest  $F$  is a subgraph of  $F$ .

A **spanning tree** is a spanning subgraph that is a tree. Every graph has a spanning forest. But only a connected graph has a spanning tree.

A special kind of tree called a **star** is  $K_{1,k}$ . An induced star with 3 edges is a **claw**.

A **caterpillar** is a tree in which all non-leaf nodes form a single path.

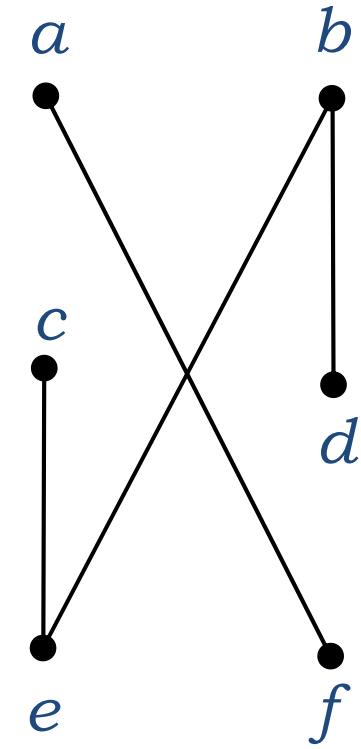
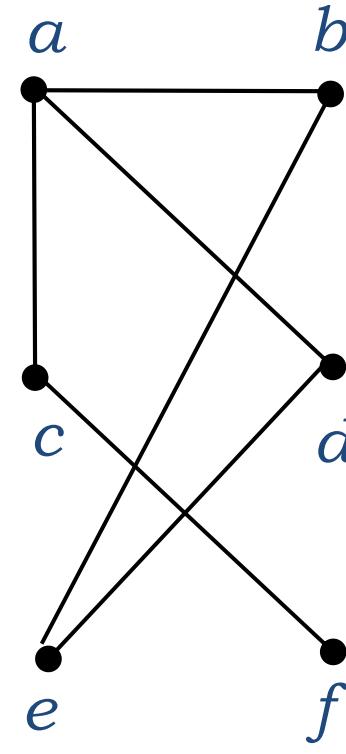
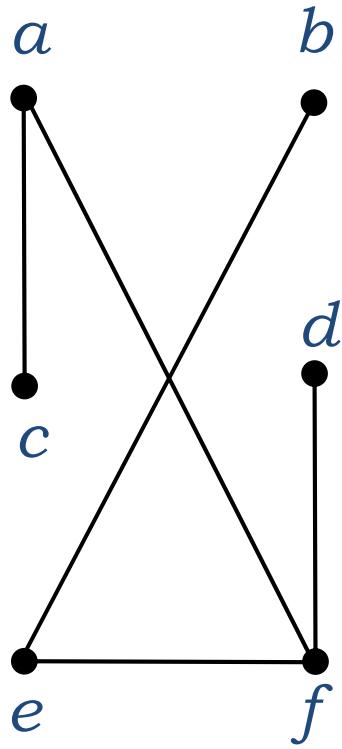
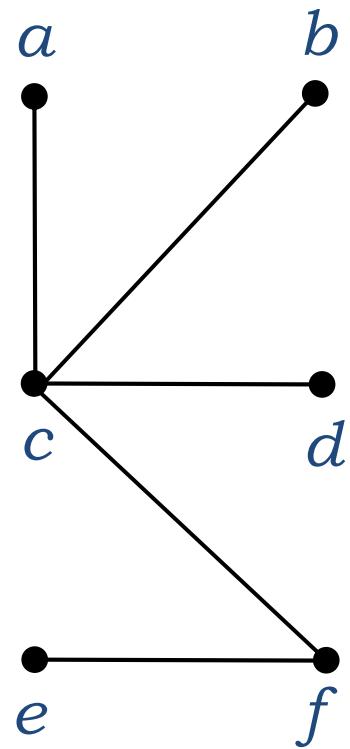
A  **$k$ -ary tree** is a rooted tree in which every internal vertex has no more than  $k$  children. A 1-ary tree is just a path. A 2-ary tree is also called a **binary tree**.



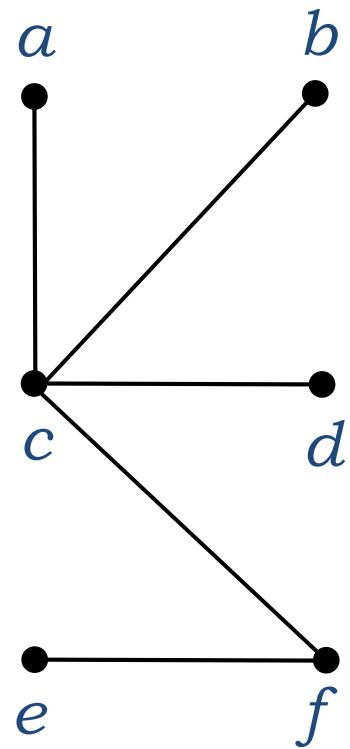
# Trees

- A *tree* is a connected undirected graph with
  - No simple circuits
  - No multiple edges
  - No loops
- An undirected graph is a tree if and only if there is a unique simple path between any two of its vertices.

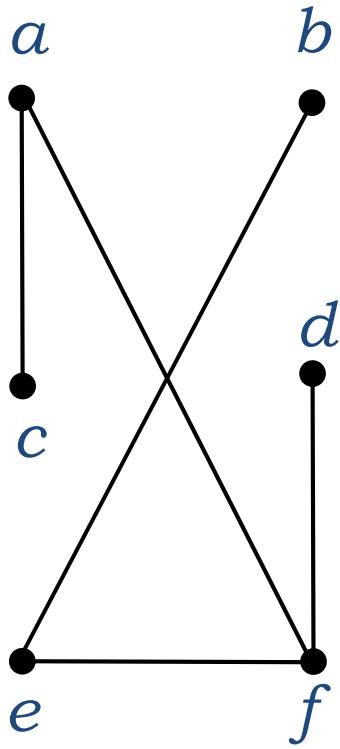
# Which graphs are trees?



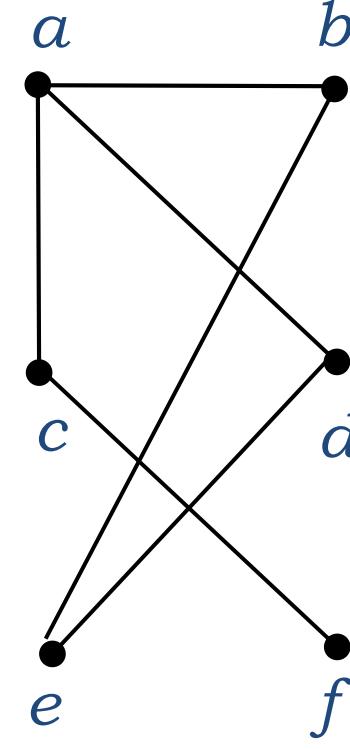
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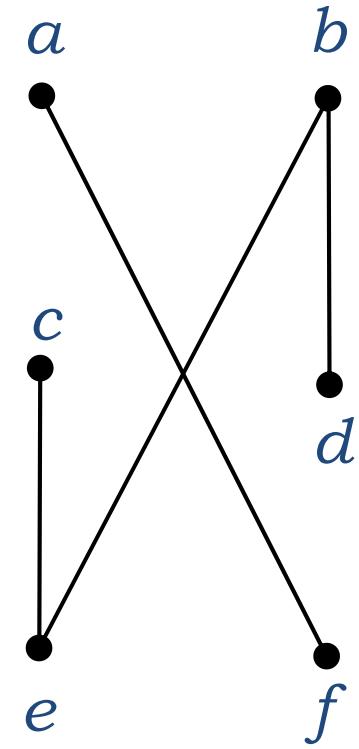
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YES



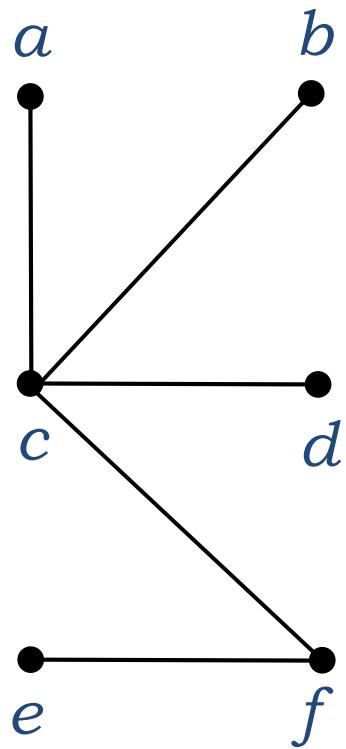
NO



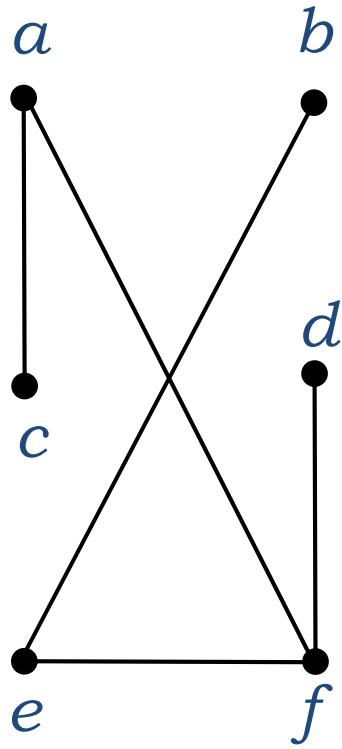
NO

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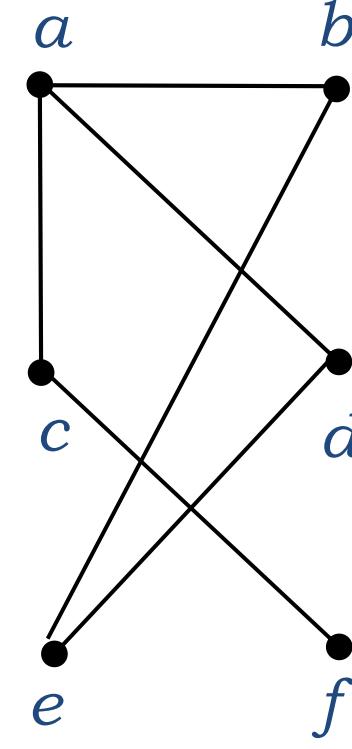
(e, b, a, d, e) is a simple circuit



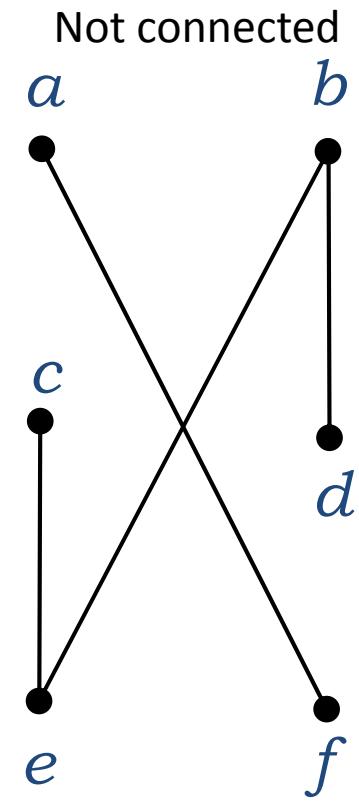
YES



YES



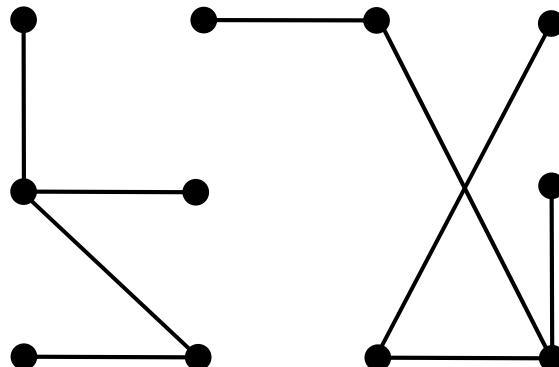
NO



NO

# Forest

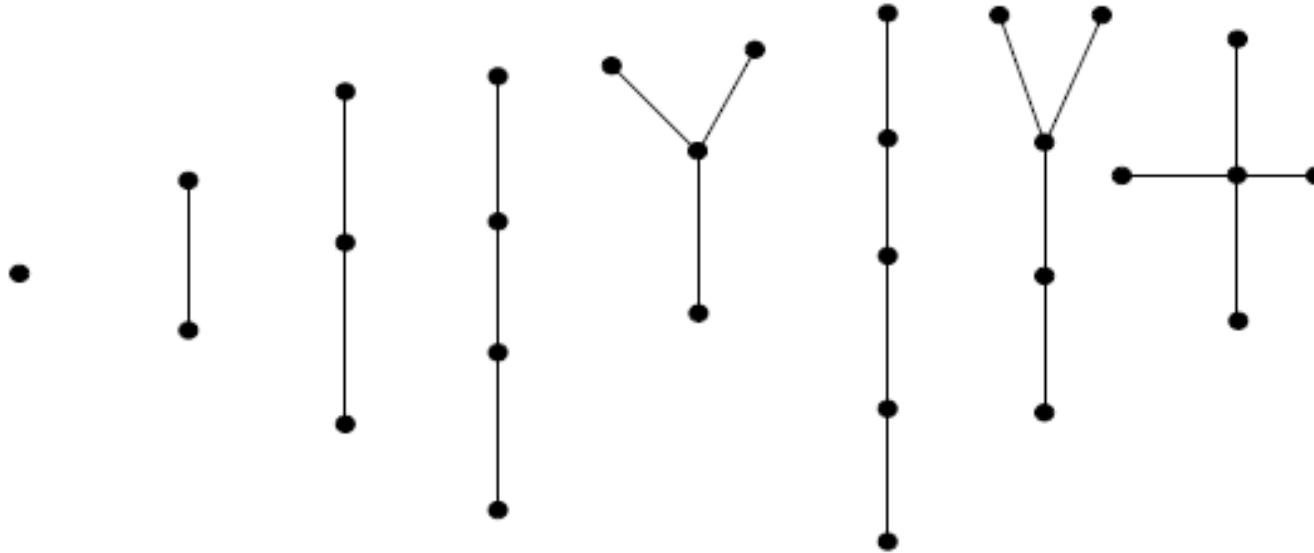
- What if there are no simple circuits but the graph is not connected?
- Each of the connected components is a tree
- The collection is called a *forest*.



**Definition:** A graph having no cycles is said to be *acyclic*. A *forest* is an acyclic graph.

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**Definition:** A *tree* is a connected graph without any cycles, or a tree is a connected acyclic graph. The edges of a tree are called *branches*. It follows immediately from the definition that a tree has to be a simple graph (because self-loops and parallel edges both form cycles).



All trees with fewer than six vertices.

# Notes

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    - A graph is a tree *iff* there is exactly one path between every pair of its vertices.
- Proof:**
- If we have a graph  $T$  which is a tree, then it must be connected with no cycles.
  - Since  $T$  is connected, there must be at least one simple path between each pair of vertices.
  - If there is more than one path between two vertices, then parts of those paths could be joined to form a cycle.
  - Thus, there must be exactly one path.
  - Now suppose we have a graph  $G$  where there exactly one simple path between vertices.
  - This graph is clearly connected.
  - If  $G$  contains a simple circuit, then there are two paths between any vertices on that circuit.
  - Thus contradicts the assumption, so there can be no circuits, and  $G$  is a tree. ■

# Another Definition of a Tree

Any connected graph with  $n$  vertices and  $n-1$  edges is a tree.

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**Proof** Let  $G$  be a connected graph with  $n$  vertices and  $n-1$  edges. We show that  $G$  contains no cycles. Assume to the contrary that  $G$  contains cycles.

Remove an edge from a cycle so that the resulting graph is again connected. Continue this process of removing one edge from one cycle at a time till the resulting graph  $H$  is a tree. As  $H$  has  $n$  vertices, so number of edges in  $H$  is  $n-1$ . Now, the number of edges in  $G$  is greater than the number of edges in  $H$ . So  $n-1 > n-1$ , which is not possible. Hence,  $G$  has no cycles and therefore is a tree.  $\square$

# What is a Tree?

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- Furthermore, some binary relations could be classified according to the existence or absence of certain pairs. ***In particular a tree is a binary relation that is “antireflexive” and symmetric.*** Therefore its digraph will possess **no loops** on any vertex, and **for every edge,  $(e_i, e_j)$ , there will be a matching edge,  $(e_j, e_i)$ .** Such a digraph is more commonly drawn by combining the pairs of directed edges that connect each pair of vertices by a single undirected edge. The resulting diagram is called an undirected graph or, more simply a graph. What distinguishes a tree from any other graph is that there are **no circuits**; that is there is no way, beginning at any vertex and by visiting a sequence of two or more distinct vertices (i.e., no repetitions) to return to the same vertex.

Aside: What is an antireflexive relation?

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Let  $\mathcal{R} \subseteq S \times S$  be a relation in  $S$ .

$\mathcal{R}$  is **antireflexive** iff:

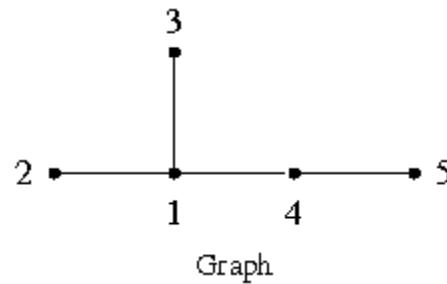
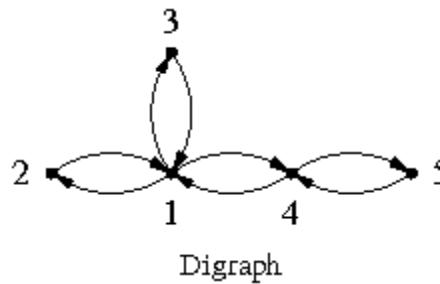
$$\forall x \in S : (x, x) \notin \mathcal{R}$$

# Tree Example

Example:

Given the set  $A = \{1, 2, 3, 4, 5\}$ , consider the binary relation  $R = \{(1, 2), (2, 1), (1, 3), (3, 1), (1, 4), (4, 1), (4, 5), (5, 4)\}$ .

Then the digraph and corresponding graph are:



The graph has no circuits and therefore is a tree.

# Equivalent Definitions of a Tree

Let  $T$  be a graph with  $n$  vertices.

Then the following statements are equivalent.

- $T$  is connected and has no cycles.
- $T$  has  $n - 1$  edges and has no cycles.
- $T$  is connected and has  $n - 1$  edges.
- $T$  is connected and the removal of any edge disconnects  $T$ .
- Any two vertices of  $T$  are connected by exactly one path.
- $T$  contains no cycles, but the addition of any new edge creates a cycle.

A photograph of a dense forest, likely Redwood National Park. The foreground is filled with the dark green, needle-covered branches of smaller coniferous trees. In the background, massive, towering redwood trees rise into the sky. Their trunks are thick and covered in a layer of bright green moss. Sunlight filters down from the canopy above, creating bright highlights on the tree trunks and casting deep shadows in the forest floor. The overall atmosphere is mysterious and ancient.

# Trees and Forests

# Trees and Forests

A **simple path** of length  $n$  in a graph is a sequence of distinct vertices  $v_1, v_2, \dots, v_n$  such that  $v_i$  is connected by an edge to  $v_{i+1}$  for  $i = 1$  to  $n-1$ . If a tree is a graph with no circuits, then there is at most one path between any two vertices. Intuitively, if the same two vertices  $v_a$  and  $v_b$  are connected by two paths, then one can leave  $v_a$  via one path to get to  $v_b$  and then return to  $v_a$  by the other path, completing a circuit of distinct vertices.

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Of course, there may be no path between two vertices,  $v_a$  and  $v_b$ , of a graph. When this is the case then the graph is said to be **disconnected**. In that case all the vertices of the graph can be divided into three sets:

1. Those for which there is a path to  $v_a$ ;
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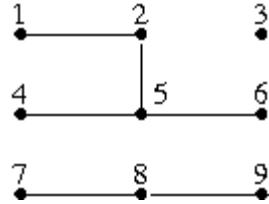
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The graph defined on vertices not connected to  $v_a$  or  $v_b$  may or may not be connected. However, one can decompose that graph into connected components in a similar manner. Thus, every graph can be decomposed into a set of one or more connected components. When those components are all trees, then the set of trees is called a **forest**.

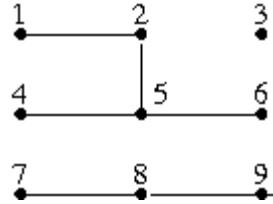
# Example

- Consider the following disconnected graph, defined on nine vertices numbered 1 through 9:



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- The vertices can be partitioned into three sets of connected vertices:

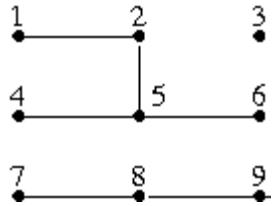
$$\{1, 2, 4, 5, 6\}$$

$$\{3\}$$

$$\{7, 8, 9\}$$

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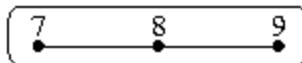
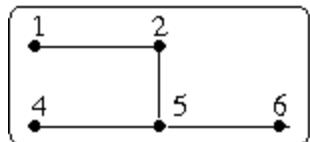
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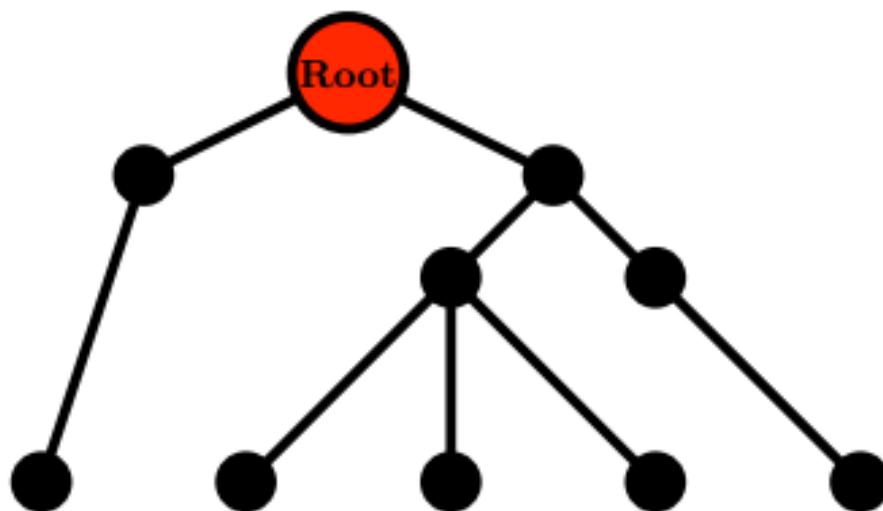
- For each edge in the original, disconnected graph we draw a corresponding edge between the same two vertices in the appropriate component:



a forest of trees

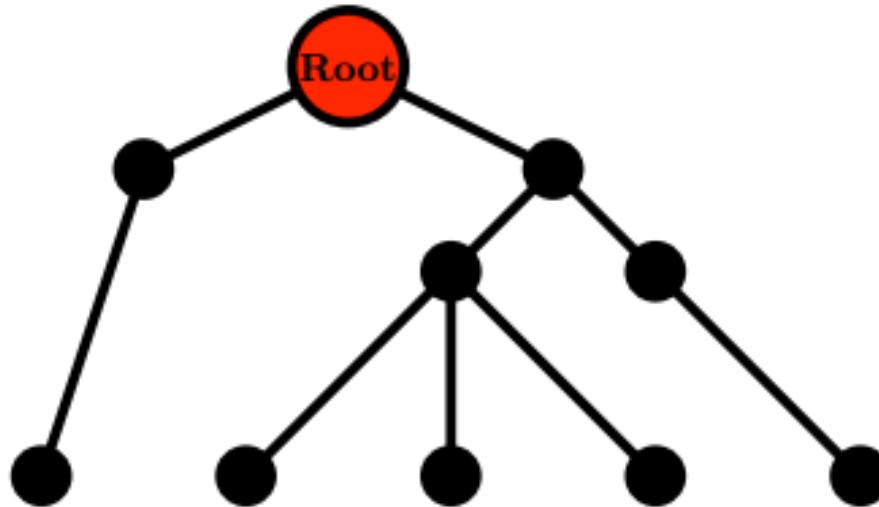
Each component defined on a set of connected vertices is a tree.

# Rooted Trees



# Rooted Trees

If one vertex of a tree is *singled out as a starting point* and *all the branches fan out from this vertex*, we call such a tree a **rooted tree**.

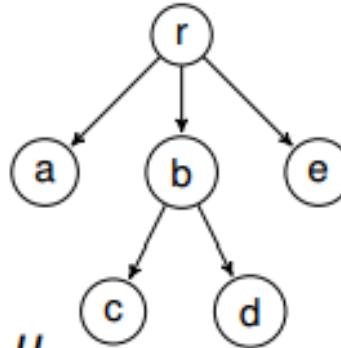


# Rooted Tree Summary

## Terminology for rooted trees

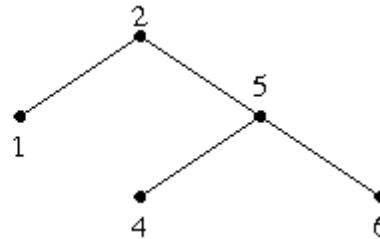
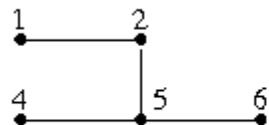
For a rooted tree  $(T, r)$ , with root  $r$ ,

- For each node  $v \neq r$  the **parent**, is the unique vertex  $u$  such that  $(u, v) \in E$ .  $v$  is then called a **child** of  $u$ .  
Two vertices with the same parent are called **siblings**.
- A **leaf** is a vertex with no children. Non-leaves are called **internal vertices**.
- The **height** of a rooted tree is the length of the longest directed path from the root to any leaf.
- The **ancestors** (**descendants**, respectively) of a vertex  $v$  are all vertices  $u \neq v$  such that there is a directed path from  $u$  to  $v$  (from  $v$  to  $u$ , respectively).
- The **subtree** rooted at  $v$ , is the subgraph containing  $v$  and all its descendants, and all directed edges between them.



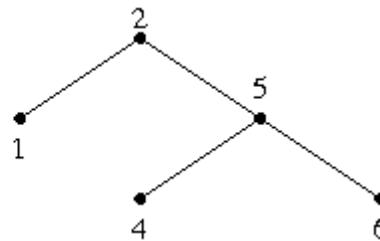
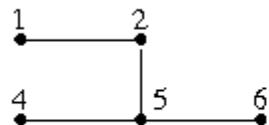
# Example

- If the tree given on the left-hand side below has vertex 2 designated as its **root**, then by convention it is often drawn in the manner illustrated on the right-hand side. That is, a **rooted tree** is typically drawn with the root at the top, then all its children are drawn immediately below it, and then all the children of the children are drawn below the children, and so on:



# Example

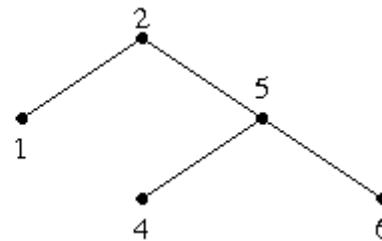
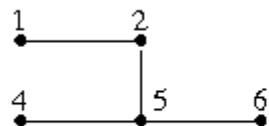
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- In this case, 2 is the root, 1 and 5 are its children (making 2 a parent), and 4 and 6 are the children of 5 (making 5 a parent). Vertices 2 and 5 are internal nodes, and vertices 1, 4, and 6 are leaves.

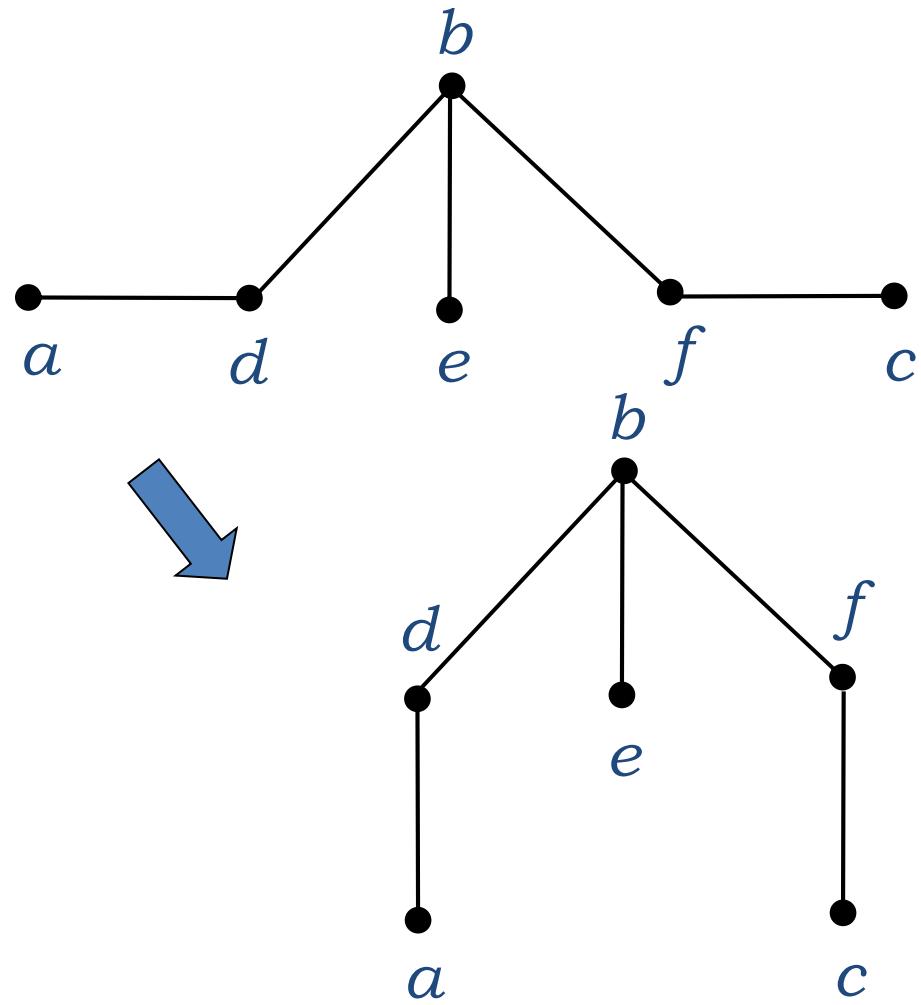
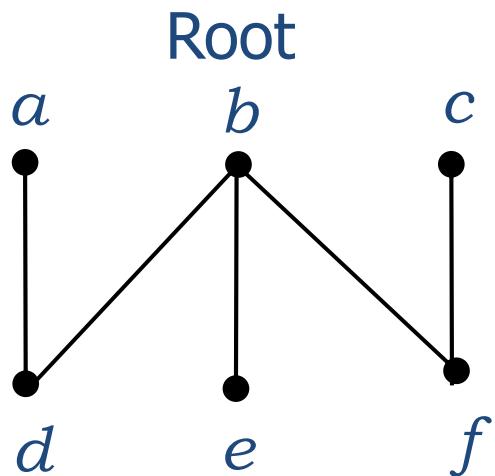
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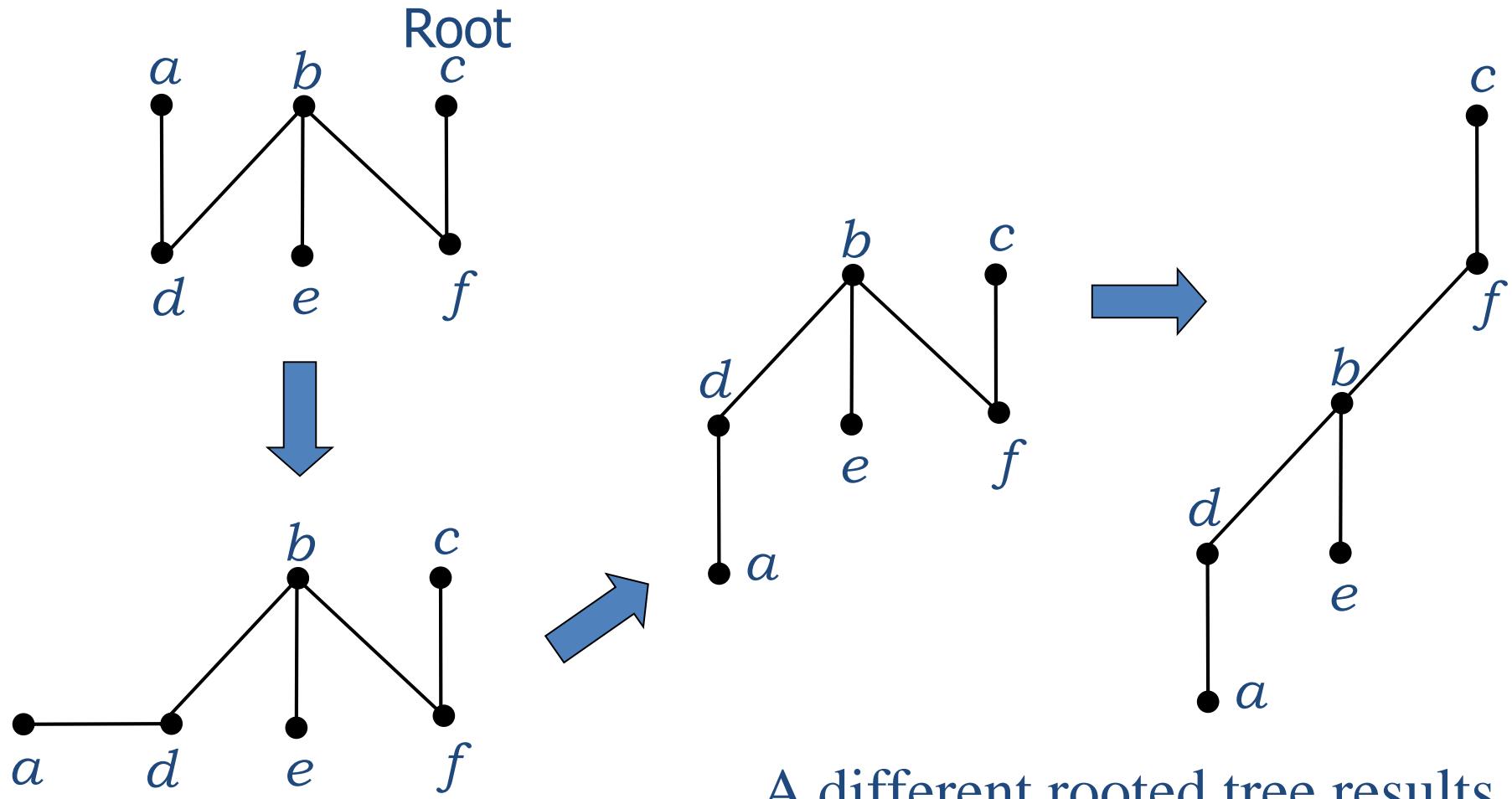


- In this case, 2 is the root, 1 and 5 are its children (making 2 a parent), and 4 and 6 are the children of 5 (making 5 a parent). Vertices 2 and 5 are internal nodes, and vertices 1, 4, and 6 are leaves.
- Except for the leaves, every vertex of a rooted tree has children. If the maximum number of children any vertex has is  $m$ , then the rooted tree is said to be an  **$m$ -ary rooted tree**. If all internal vertices have  $m$  children, the tree is called a **full  $m$ -ary rooted tree**. When  $m = 2$  such a tree is called a full binary rooted tree (rather than a full 2-ary tree). The rooted tree with root 2 given in the diagram above is an example of a full binary rooted tree.

# Example



# What if a different root is chosen?



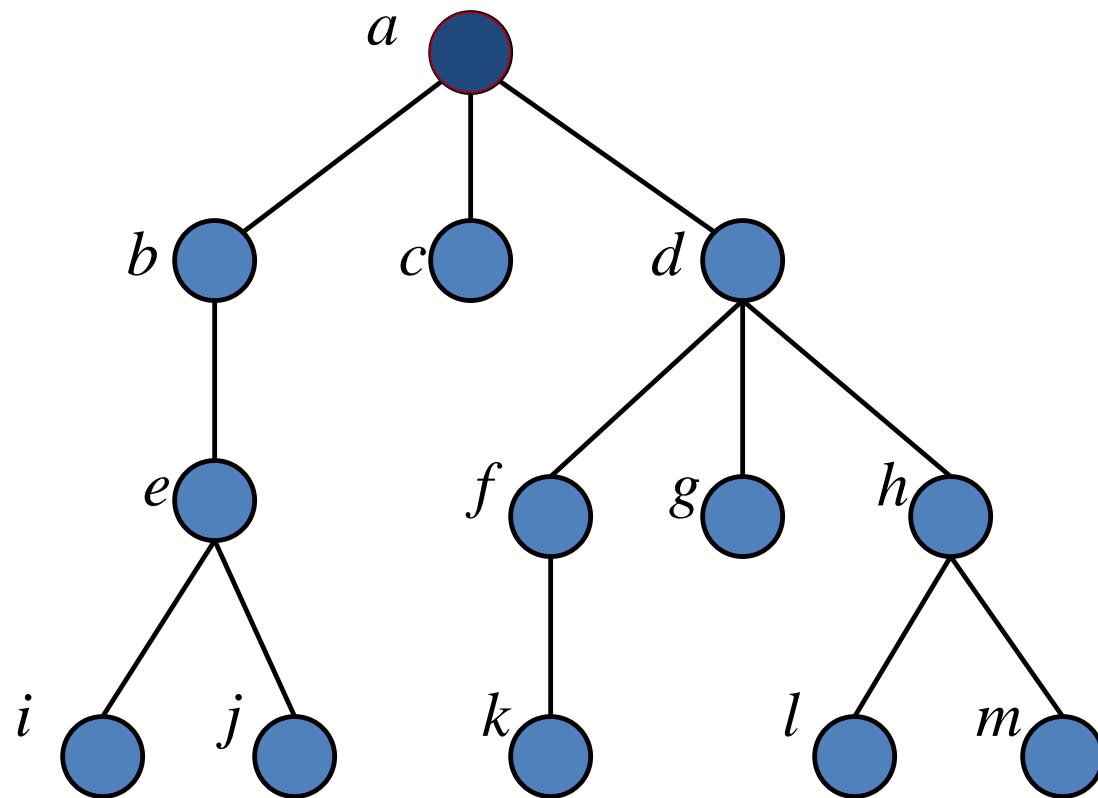
# Review of Tree Terminology

# Tree Terminology

- If  $v$  is a vertex of tree  $T$  other than the root, the *parent* of  $v$  is the unique vertex  $u$  such that there is a directed edge from  $u$  to  $v$ .
- When  $u$  is the parent of  $v$ ,  $v$  is called the *child* of  $u$ .
- If two vertices share the same parent, then they are called *siblings*.

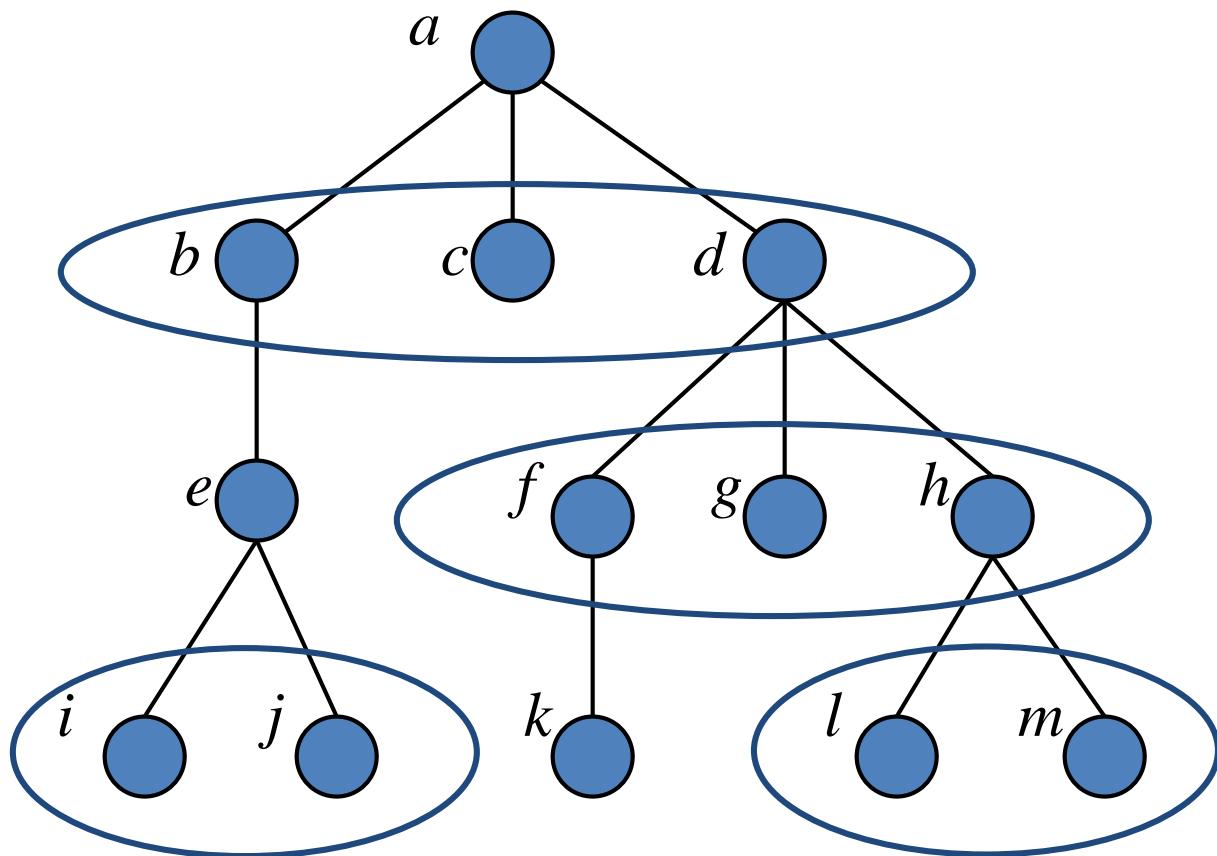
# Example

Root



# Example

Siblings

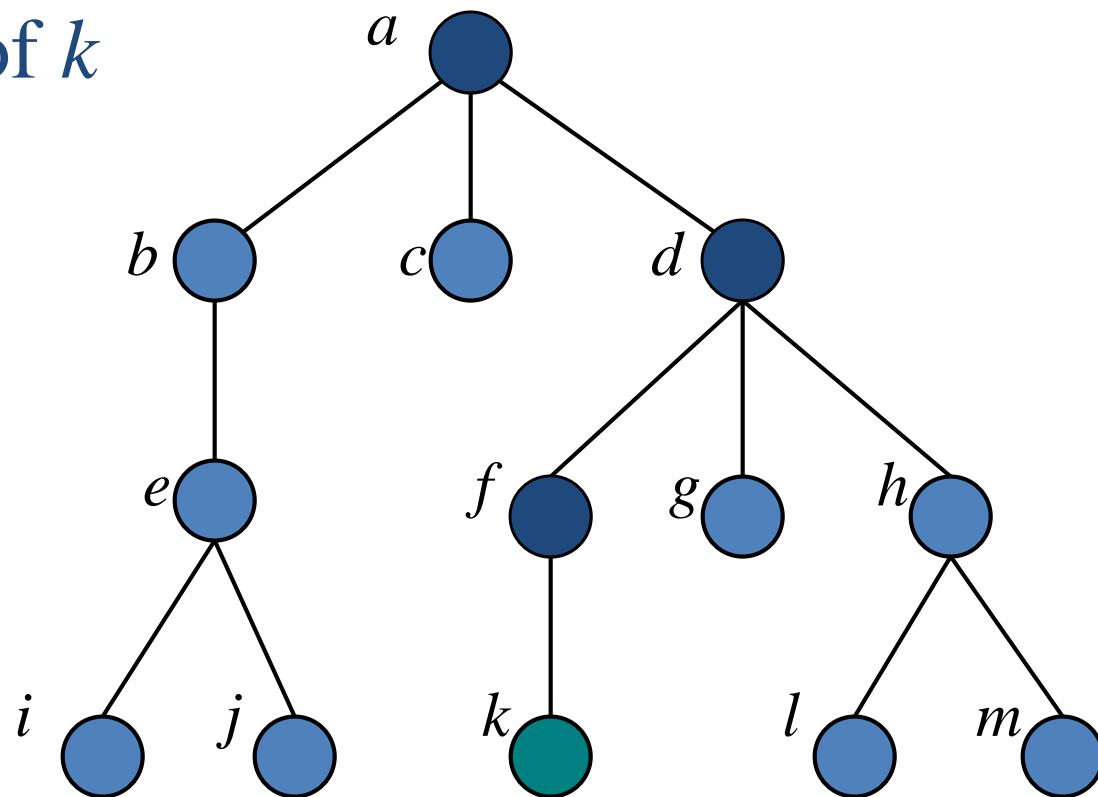


# Tree Terminology (Cont.)

- The *ancestors* of a vertex other than the root are the vertices in the path from the root to this vertex, excluding the vertex itself and including the root.
- The *descendants* of a vertex  $v$  are those vertices that have  $v$  as an ancestor.

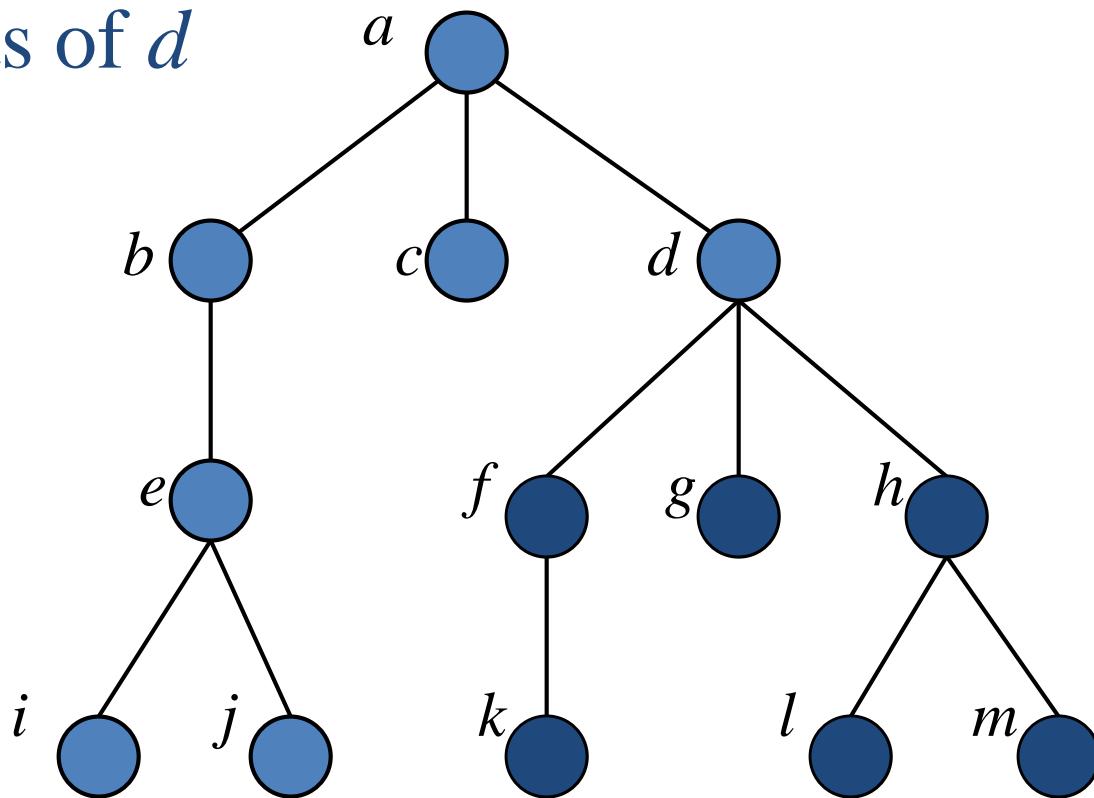
# Example

Ancestors of  $k$



# Example

Descendants of  $d$

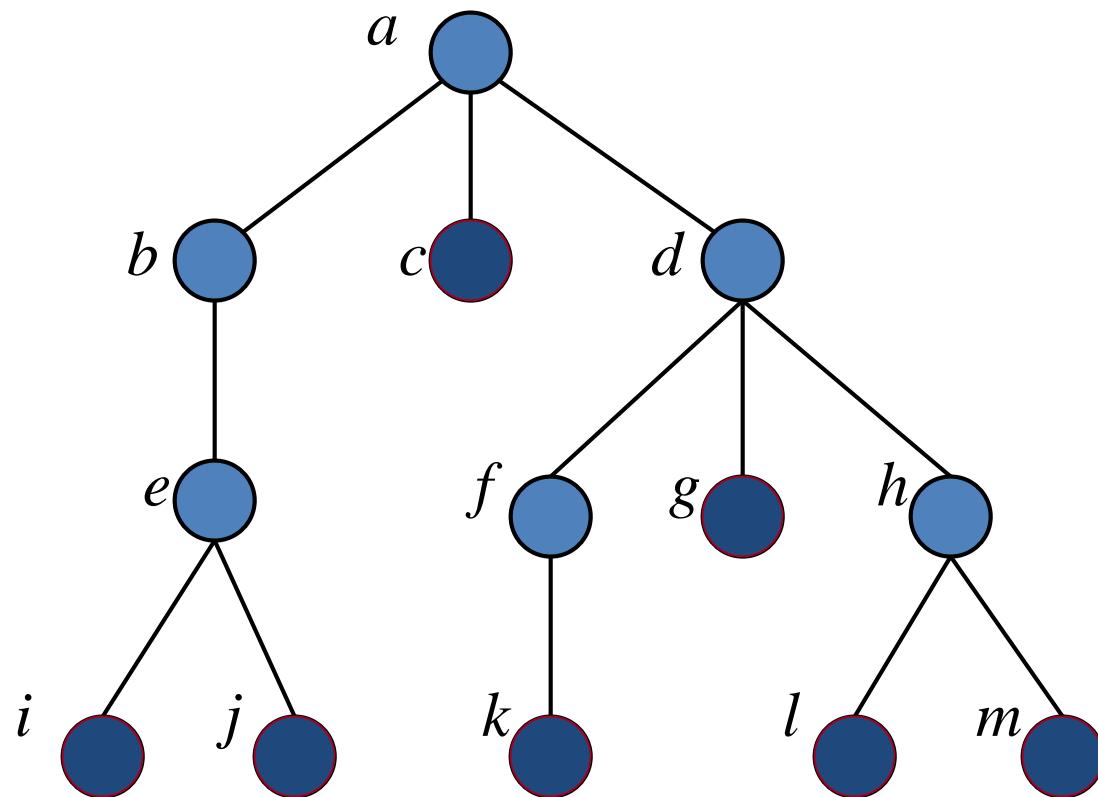


# Tree Terminology (Cont.)

- A vertex with no children is called a *leaf*.
- Vertices with children are called *internal vertices*.

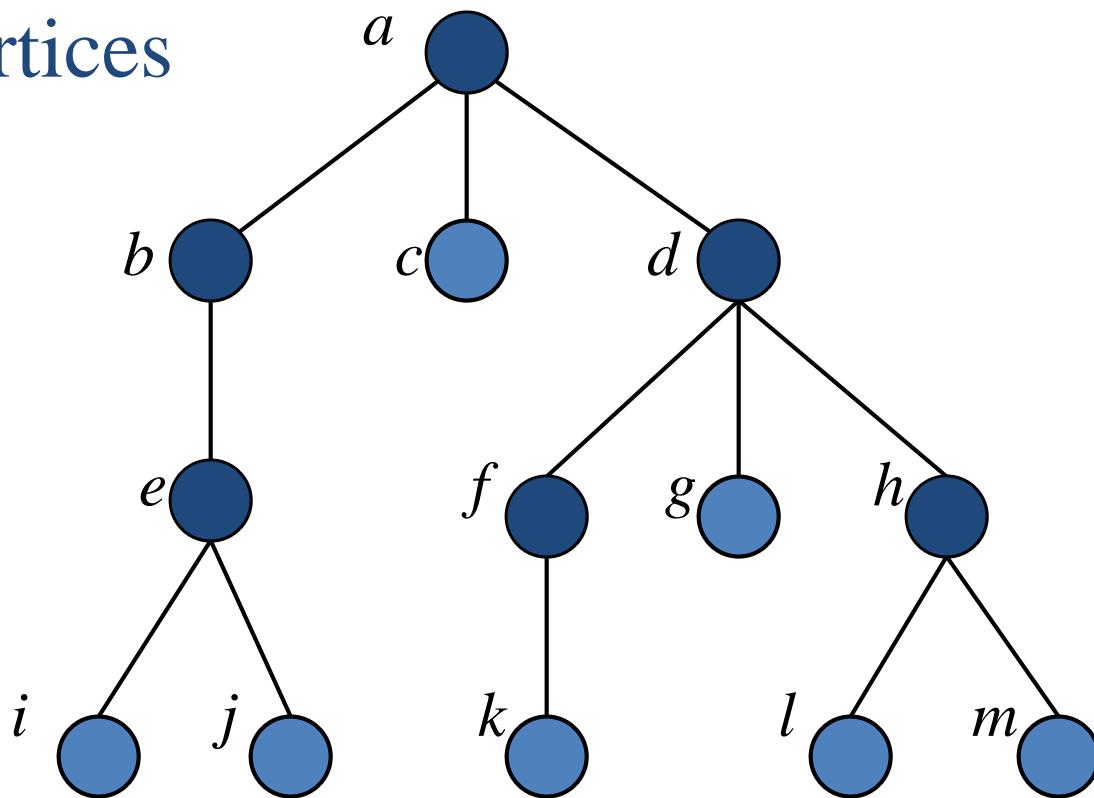
# Example

Leaves

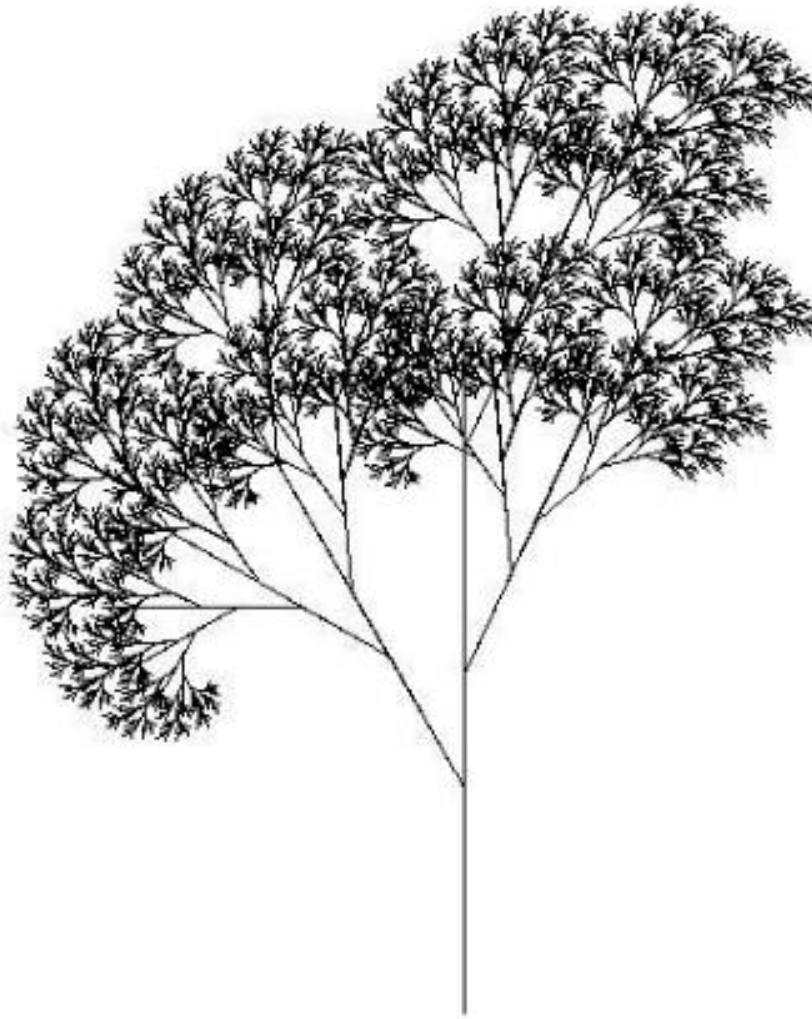


# Example

Internal vertices



# Recursive Definition of a Rooted Tree



# Recursive Definition of a Rooted Tree

The set of *rooted trees*, where a rooted tree consists of a set of vertices containing a distinguished vertex called the *root*, and edges connecting these vertices, can be defined recursively by these steps:

**BASIS STEP:** A single vertex  $r$  is a rooted tree.

**RECURSIVE STEP:** Suppose that  $T_1, T_2, \dots, T_n$  are disjoint rooted trees with roots  $r_1, r_2, \dots, r_n$ , respectively. Then the graph formed by starting with a root  $r$ , which is not in any of the rooted trees  $T_1, T_2, \dots, T_n$ , and adding an edge from  $r$  to each of the vertices  $r_1, r_2, \dots, r_n$ , is also a rooted tree.

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Basis step



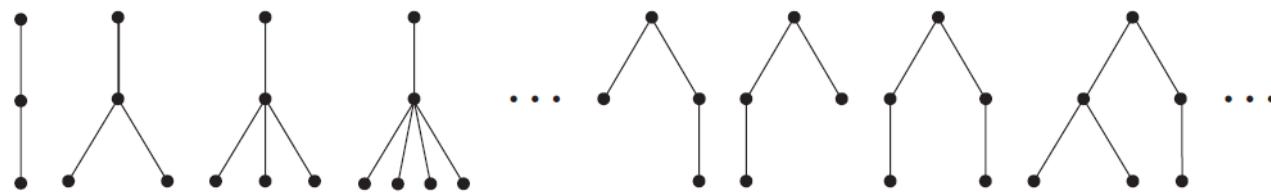
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Step 1



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Step 2



# Forests, Subforests, Subtrees and Spanning Trees

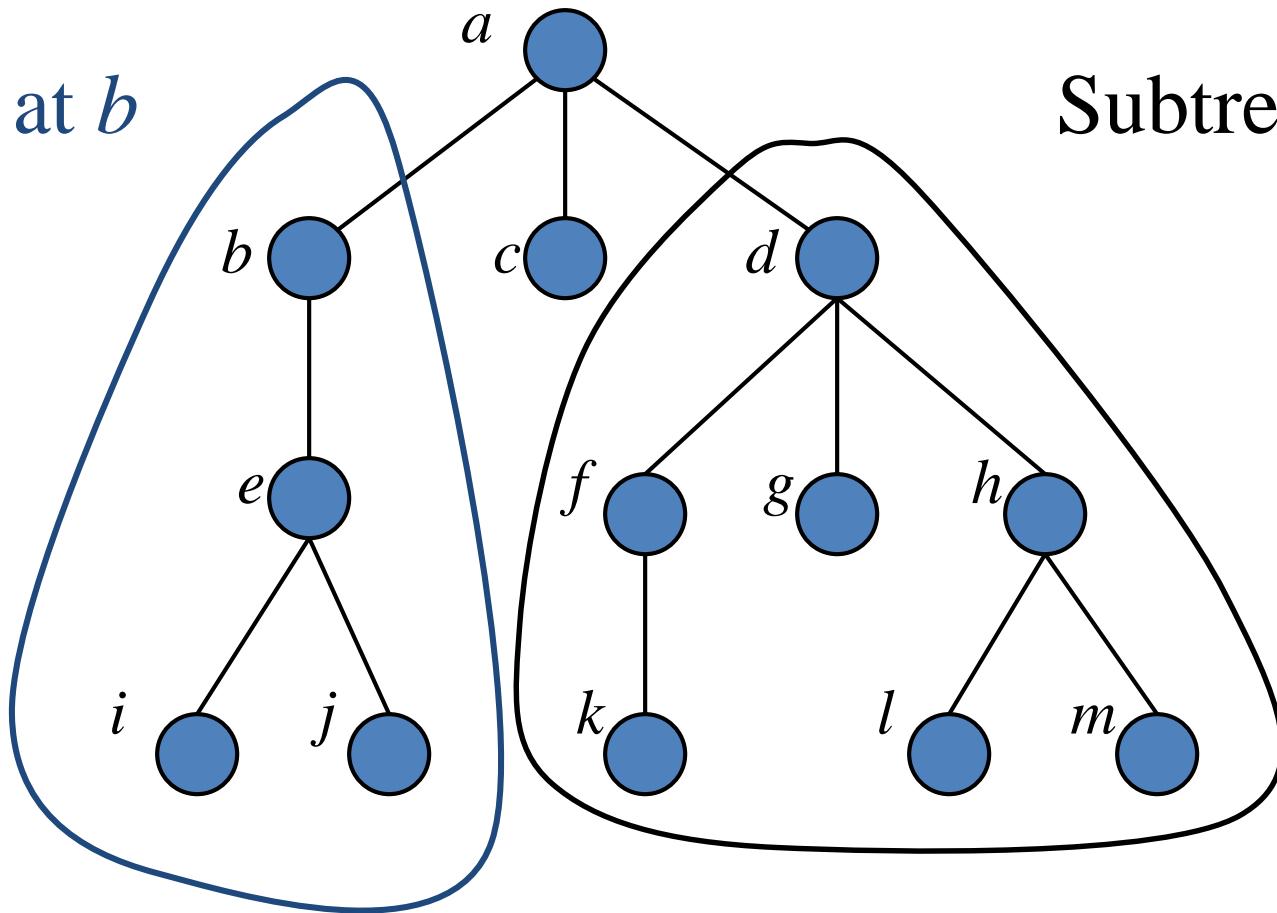
# Tree Terminology (Cont.)

- If  $a$  is a vertex in a tree, the *subtree* with  $a$  as its root is:
  - the subgraph of the tree consisting of  $a$  and its descendants, and
  - all edges incident to these descendants.

# Example

Subtree at  $b$

Subtree at  $d$



# Forests, Subforests, Subtrees and Spanning Trees

A *forest* is a circuitless graph. A *tree* is a connected forest. A *subforest* is a subgraph of a forest. A connected subgraph of a tree is a *subtree*. Generally speaking, a subforest (respectively subtree) of a graph is its subgraph, which is also a forest (respectively tree).

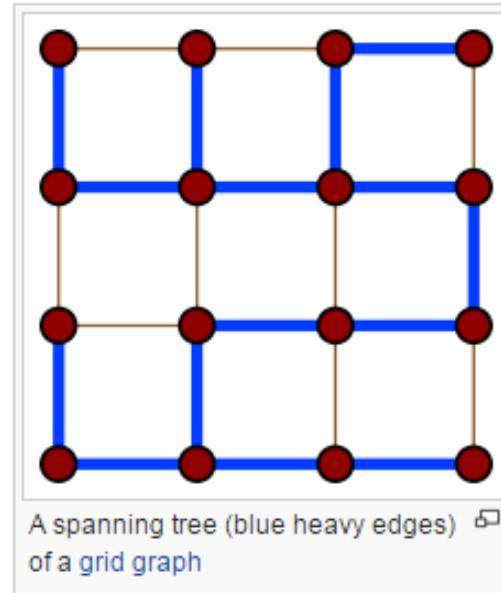
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From Wikipedia, the free encyclopedia

*For the protocol used to prevent forwarding loops on a LAN, see [Spanning Tree Protocol](#).*

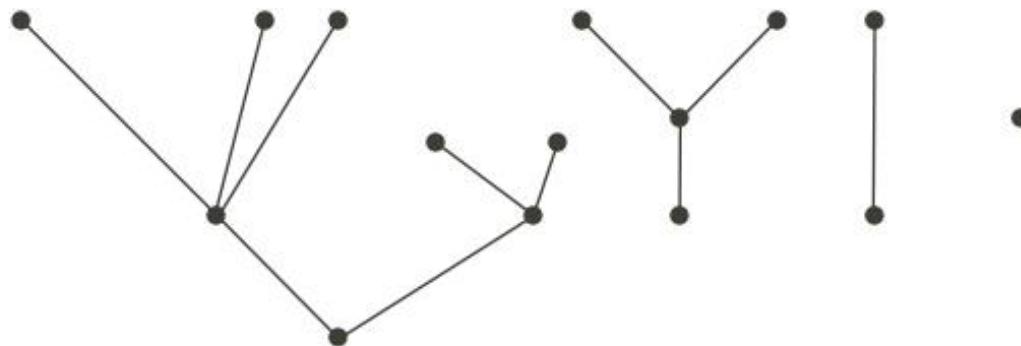
In the mathematical field of graph theory, a **spanning tree**  $T$  of an undirected graph  $G$  is a subgraph that is a tree which includes all of the vertices of  $G$ . In general, a graph may have several spanning trees, but a graph that is not connected will not contain a spanning tree (but see Spanning forests below). If all of the edges of  $G$  are also edges of a spanning tree  $T$  of  $G$ , then  $G$  is a tree and is identical to  $T$  (that is, a tree has a unique spanning tree and it is itself).



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**Example.** Four trees which together form a forest:



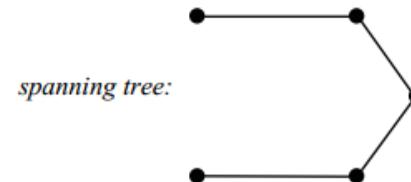
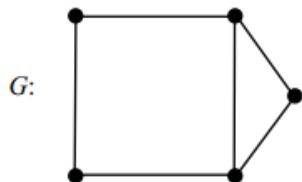
A *spanning tree* of a connected graph is a subtree that includes all the vertices of that graph. If  $T$  is a spanning tree of the graph  $G$ , then

$$G - T =_{\text{def.}} T^*$$

is the *cospanning tree*.

# Spanning Trees

**Example.**



*cospanning tree*



The edges of a spanning tree are called *branches* and the edges of the corresponding cospanning tree are called *links* or *chords*.

# Spanning Trees

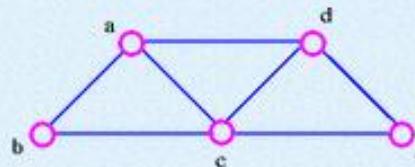
We can find a spanning tree systematically by using either of two methods.

- **Cutting-down Method**

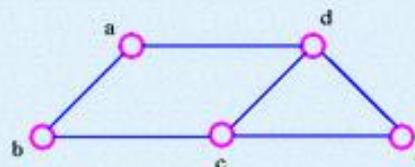
- Start choosing any cycle in  $G$ .
- Remove one of cycle's edges.
- Repeat this procedure until there are no cycles left.

# Cutting-down Method

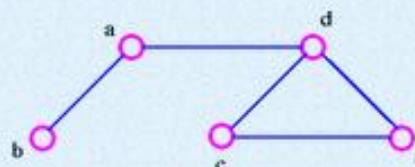
For example, given the graph G



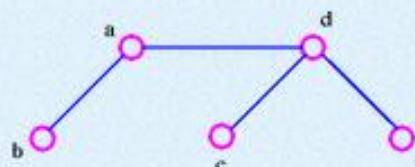
1. We remove the edge **ac**, which destroy the cycle **adcea** in the above graph and we get



2. We remove the edge **cb**, which destroy the cycle **adcba** in the above graph and we get



3. We remove the edge **ec**, which destroy the cycle **dcd** in the above graph and thus obtained the following spanning tree.



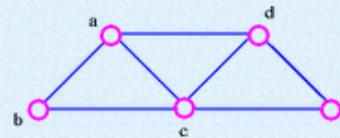
# Spanning Trees

We can find a spanning tree systematically by using either of two methods.

- **Cutting-down Method**
  - Start choosing any cycle in  $G$ .
  - Remove one of cycle's edges.
  - Repeat this procedure until there are no cycles left.
- **Building-up Method**
  - Select edges of  $G$  one at a time in such a way that no cycles are created.
  - Repeat this procedure until all vertices are included.

# Building-up Method

For example, for the following graph G



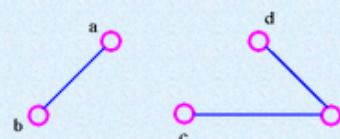
1. Choose the edge ab



2. Next choose the edge de as follows:



3. After that choose the edge ec as follows:

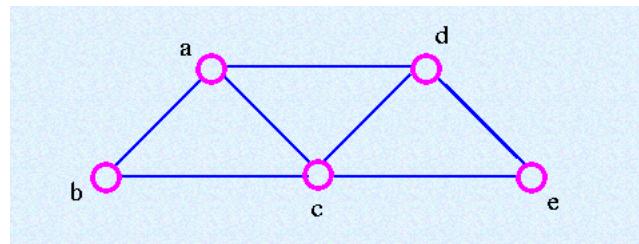


4. Finally, we choose the edge cb and thus obtain the following spanning tree.



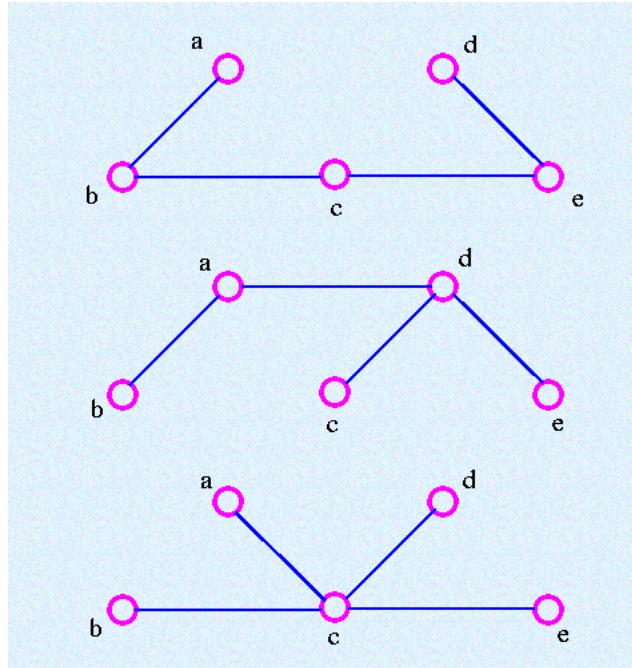
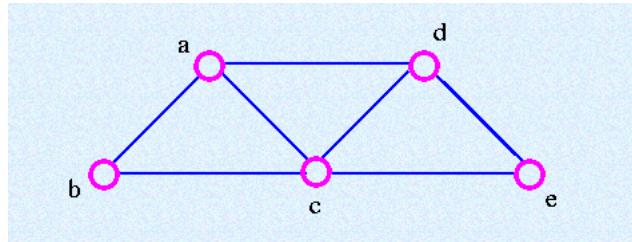
# Exercise

What are the three spanning trees associated with this graph?



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# Theorem

A graph is connected if and only if it has a spanning tree.

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## Proof

- Let  $G$  be a connected graph.
- Delete edges from  $G$  that are not bridges until we get a connected subgraph  $H$  in which each edge is a bridge.
- Then  $H$  is a spanning tree.
- Conversely, if there is a spanning tree in  $G$ , there is a path between any pair of vertices in  $G$ ; thus  $G$  is connected.

**QED**

# Counting Edges

- When trees are used in algorithms or to model data structures, it is important to be able to compute various properties possessed by a tree possibly as a function of the number of vertices, number of edges, or number of siblings for example. These parameters and others associated with trees are inter-related; that is, each is often constrained and therefore determined by the values of the others.

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- We already proved this theorem (Slide #44).

# Tree Terminology (Cont.)

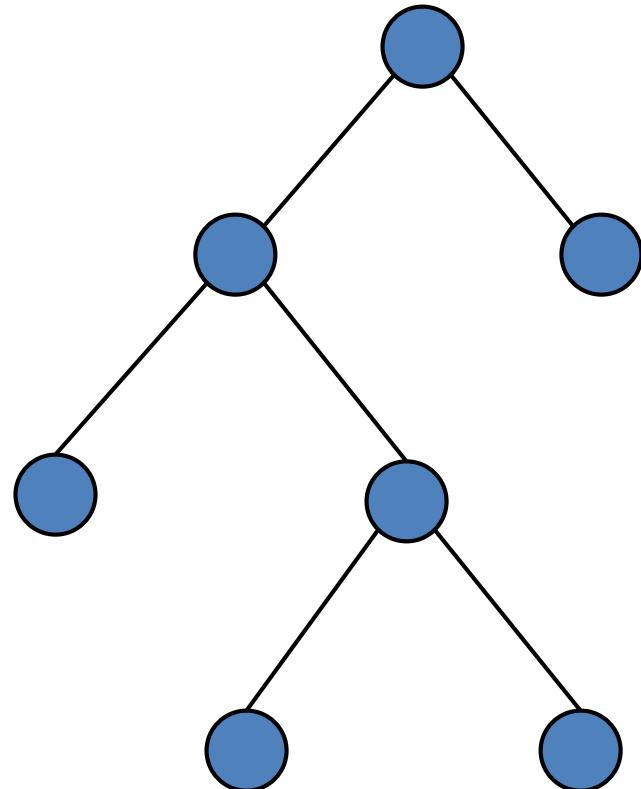
- A rooted tree is called an *m-ary tree* if every internal vertex has no more than  $m$  children.
- A tree is called a *full m-ary tree* if every internal vertex has exactly  $m$  children.
- An *m-ary tree* with  $m = 2$  is called a *binary tree*.

# Example

- What is the *arity* of this tree?
- Is this a full  $m$ -ary tree?

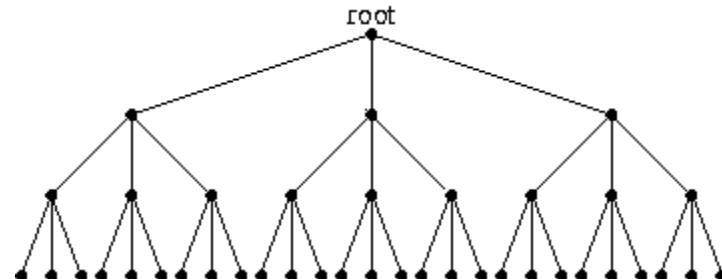
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- This is a 2-ary, or *binary*, tree.
- Yes, this is a full binary tree, since every internal vertex has exactly 2 children.



# Counting Vertices

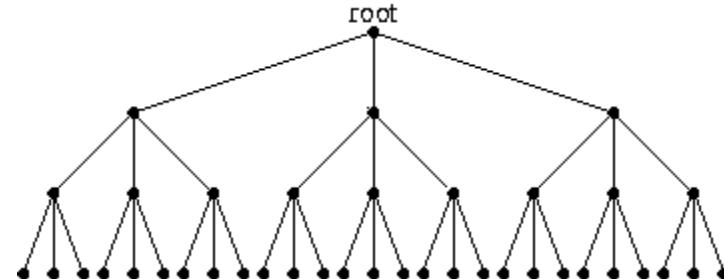
- In rooted trees there are two kinds of vertices: internal vertices and leaves. Many algorithms that employ trees in their implementation use the two types for different purposes. So it is important to recognize the distinction and this includes being able to determine or estimate the number of vertices of each kind that exist in a given tree.



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- The number of leaves:  $\lambda = n - i = i(m - 1) + 1$ .

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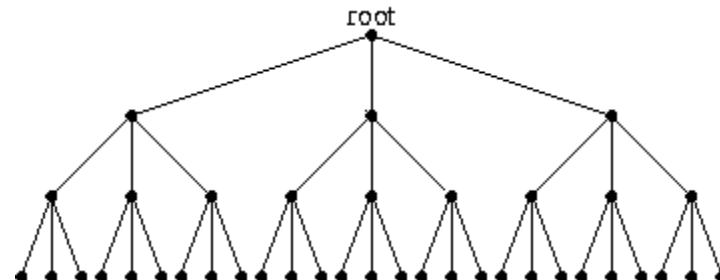
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- **By combining formula (3) with the edge counting formula, we can obtain a formula for the number of edges, e, in a full m-ary rooted tree with i internal vertices:**

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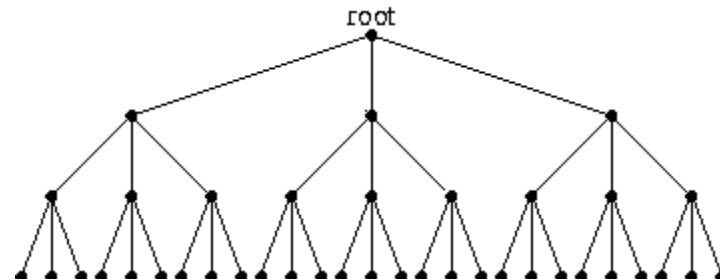
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- The following tree is an example of a full ternary (i.e. 3-ary) rooted tree:

- By inspection, there are:
  - $e = 39$  edges;
  - $n = 40$  vertices;
  - $m = 3$  children for every internal vertex;
  - $i = 13$  internal vertices;
  - $\lambda = 27$  leaves
- Then,  $m \times i = 3 \times 13 = 39 = n - 1$



# Balanced Trees

- A common activity in algorithms that employ trees to represent data structures is traversal of the tree from its root to some designated vertex, either an internal one or a leaf. In a rooted tree there is a unique path from the root to any vertex. The **length** of any path from  $v_a$  to  $v_b$  is the number of edges in the path connecting  $v_a$  to  $v_b$ .

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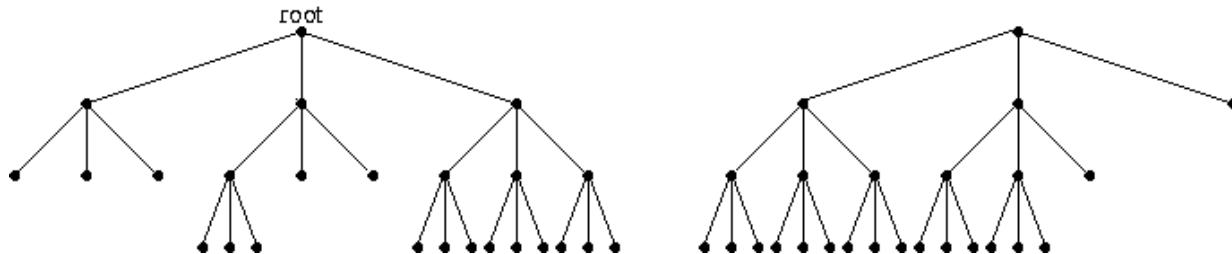
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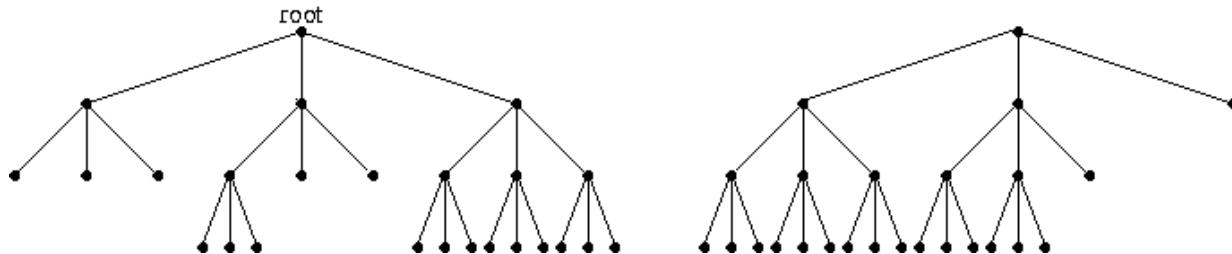
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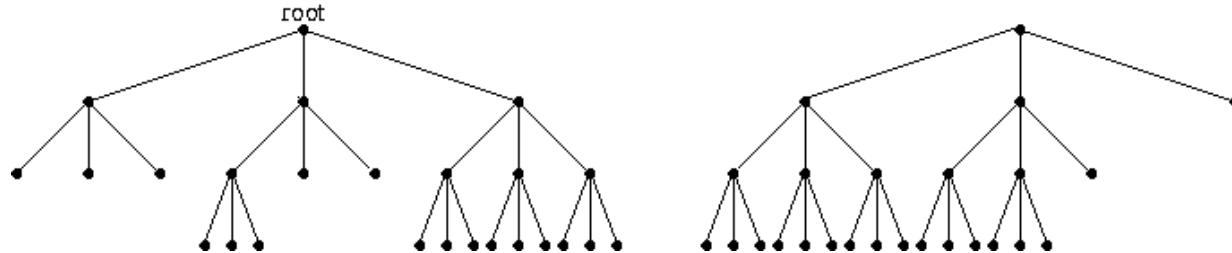
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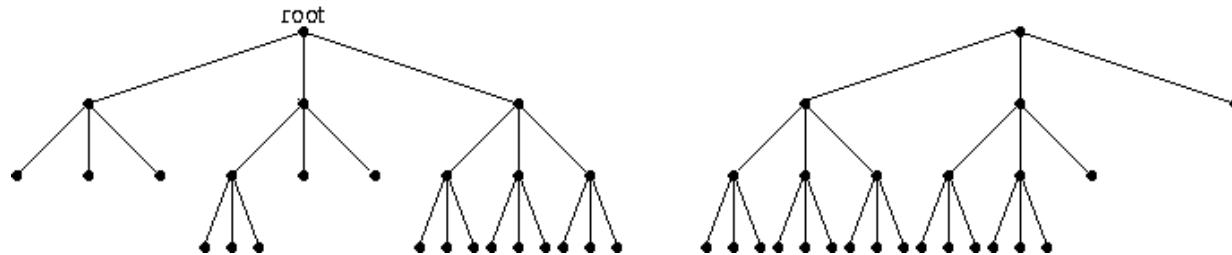
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- Each tree has height = 3. However, the left tree has five level-2 leaves whereas the right tree has one level-2 leaf and one level-1 leaf.
- In some algorithms that employ trees, it is important that all paths from the root to the leaves be nearly the same length. In particular, a tree of height  $h$  where the level of all leaves is either  $h$  or  $h - 1$  is called a **balanced** tree. Therefore the left tree above is a balanced tree, whereas the right tree is not.

# Binary Trees

- Deciding between two alternatives is a very common activity. This is because logical reasoning is the basis of formal decision making. Given a declarative statement, we must decide whether it is true or false. When a sequence of such decisions must be made, the choices and outcomes can be represented by a particular type of rooted tree, called a binary tree.

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- Denote by  $\lambda_h$  the number of leaves in a complete binary tree of height  $h$ .
- **Base Rule:**  $\lambda_0 = 1$
- **Recursive Rule:** For  $h > 0$ ,  $\lambda_h = 2 \times \lambda_{h-1}$

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- The number of leaves in the complete binary trees of heights 1, 2, and 3 are:  $\lambda_1 = 2^1 = 2$ ,  $\lambda_2 = 2^2 = 4$ , and  $\lambda_3 = 2^3 = 8$ . Using informal inductive reasoning, we can hypothesize that  $\lambda_h = 2^h$ . Once again, we can apply mathematical induction to formally prove this hypothesis:

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- Therefore the number of leaves in a complete binary tree of height  $h$  is given by:

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- The importance of this result is that it places a cap on the number of leaves that can occur in a binary tree, with that cap determined by the height of the tree. In effect, for a binary tree of height  $h$ , there can be at most  $2^h$  leaves. To illustrate the significance, consider the problem of growing a binary tree; that is, adding new vertices to it. There are two ways to do this:

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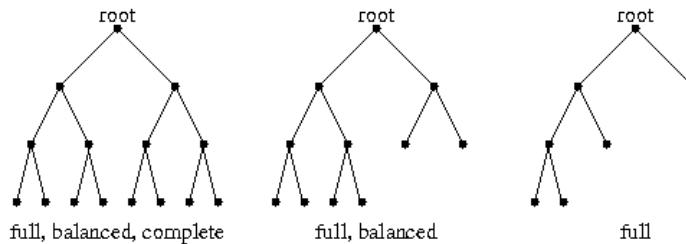
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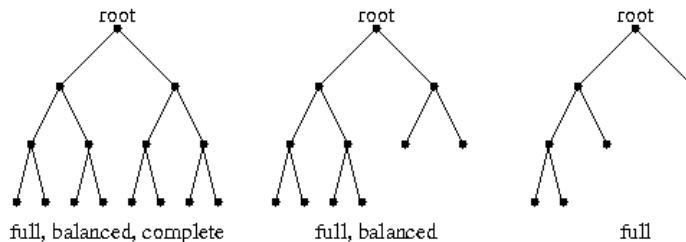
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- The following example shows three full binary trees; however, only the left-most tree is complete.



- Therefore no vertices can be added to the left binary tree; four vertices can be added to the centre binary tree; and eight vertices can be added to the right binary tree without the height of the tree being increased. As a result, except for the complete binary tree, for all other binary trees of height  $h$ :

$$\lambda < 2^h$$

FULL



**Theorem:** Every binary tree has an odd number of vertices.

**Theorem:** Every binary tree has an odd number of vertices.

**HINT:** Use the Handshaking Lemma, that “Every graph has an even number of vertices of odd degree” (from the Graph Theory PPT)

**Theorem:** Every binary tree has an odd number of vertices.

**Proof** Apart from the root, every vertex in a binary tree is of odd degree. We know that there are even number of such odd vertices. Therefore when the root (which is of even degree) is added to this number, the total number of vertices is odd.

# Proof of the Lemma

**Proposition:** Every graph has an even number of vertices of odd degree.

**Proof:**

The sum of all the degrees is equal to twice the number of edges. Since the sum of the degrees is even and the sum of the degrees of vertices with even degree is even, the sum of the degrees of vertices with odd degree must be even. If the sum of the degrees of vertices with odd degree is even, there must be an even number of those vertices.

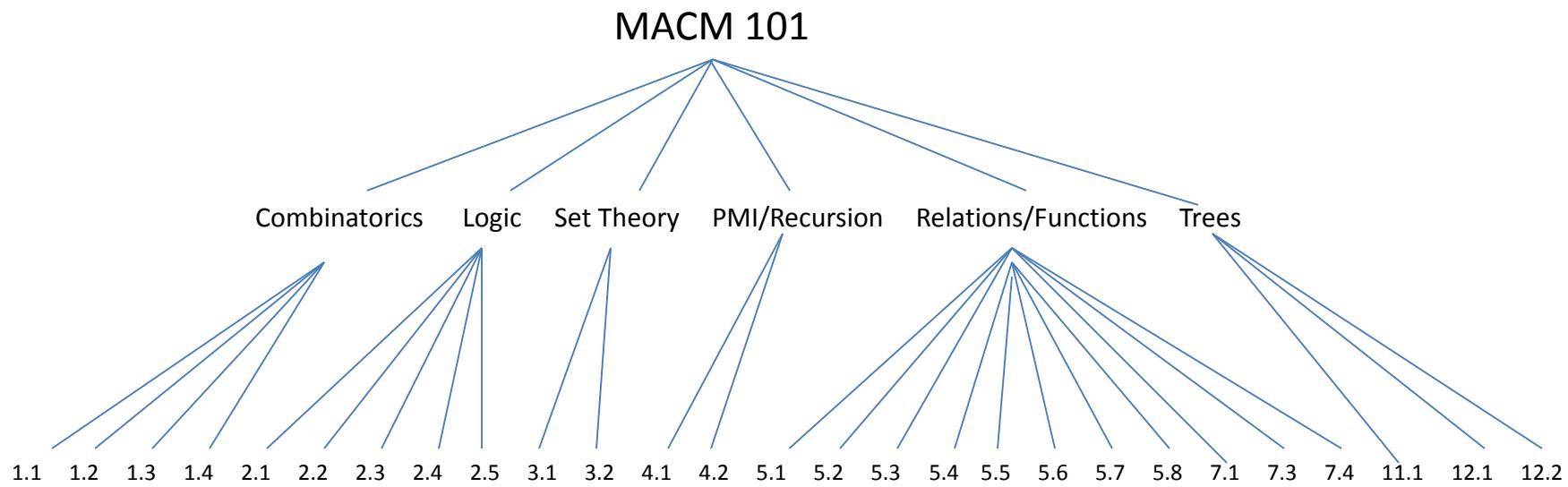
**QED**

# Application of Trees: Tree Searching

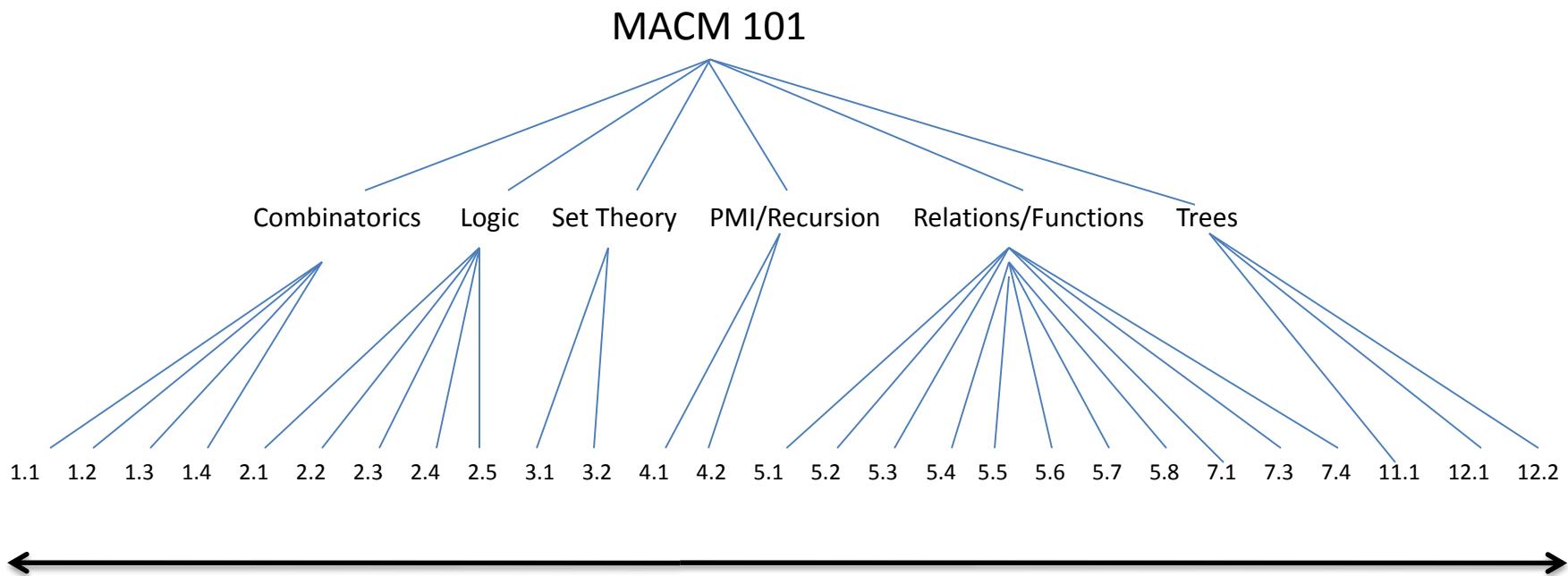
There are two well-known search methods.

- They are known as depth-first search(DFS) and breadth-first search (BFS).
- Each of these methods lists the vertices as they are encountered, and indicates the direction in which each edge is first traversed.
- The methods differ only in the way in which the vertex-lists are constructed.

# Example: Studying for the Final

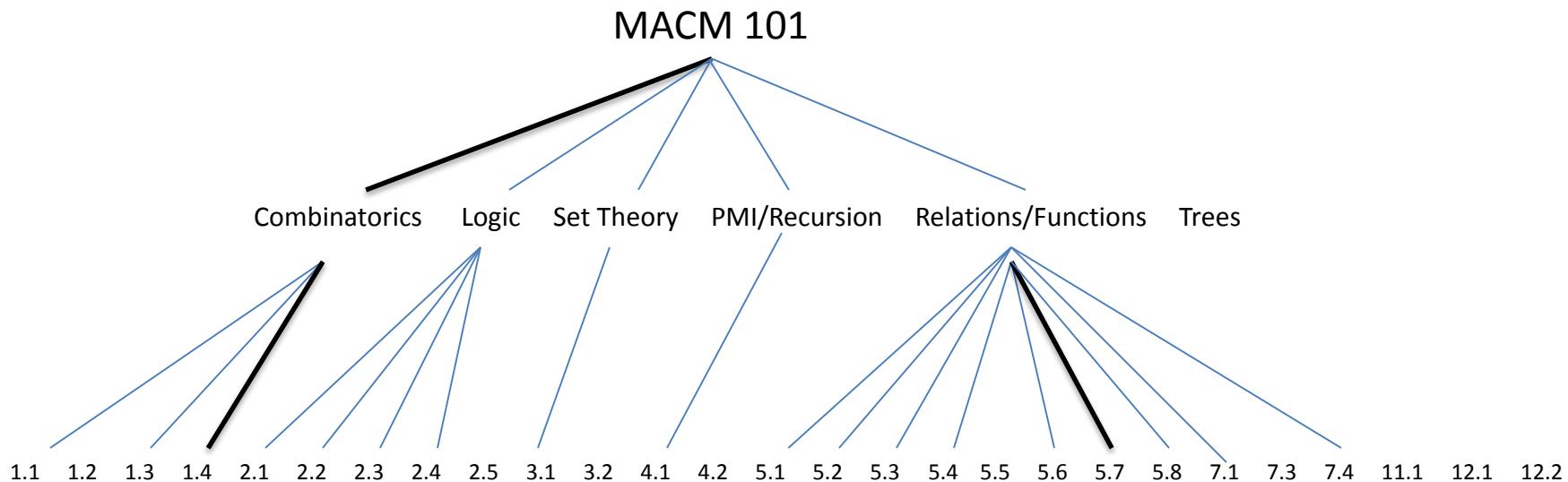


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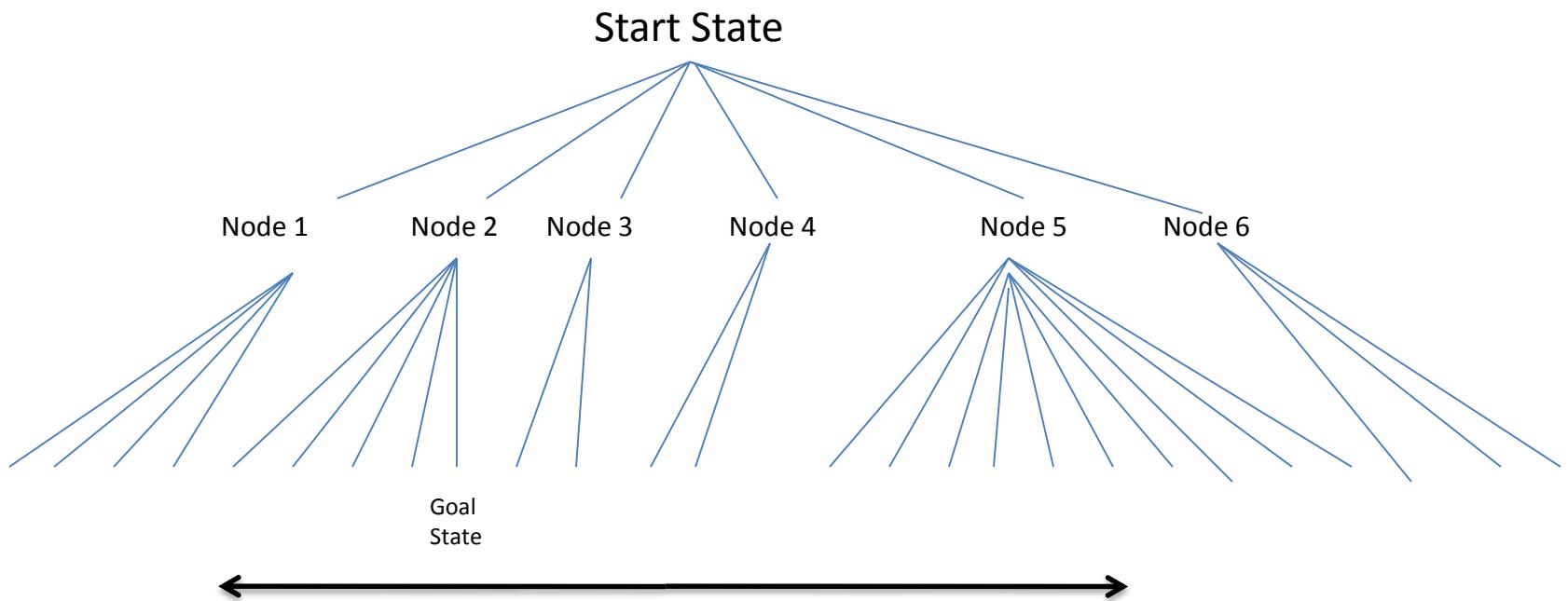
A Breadth First approach to studying would involve a cursory overview of all topics.

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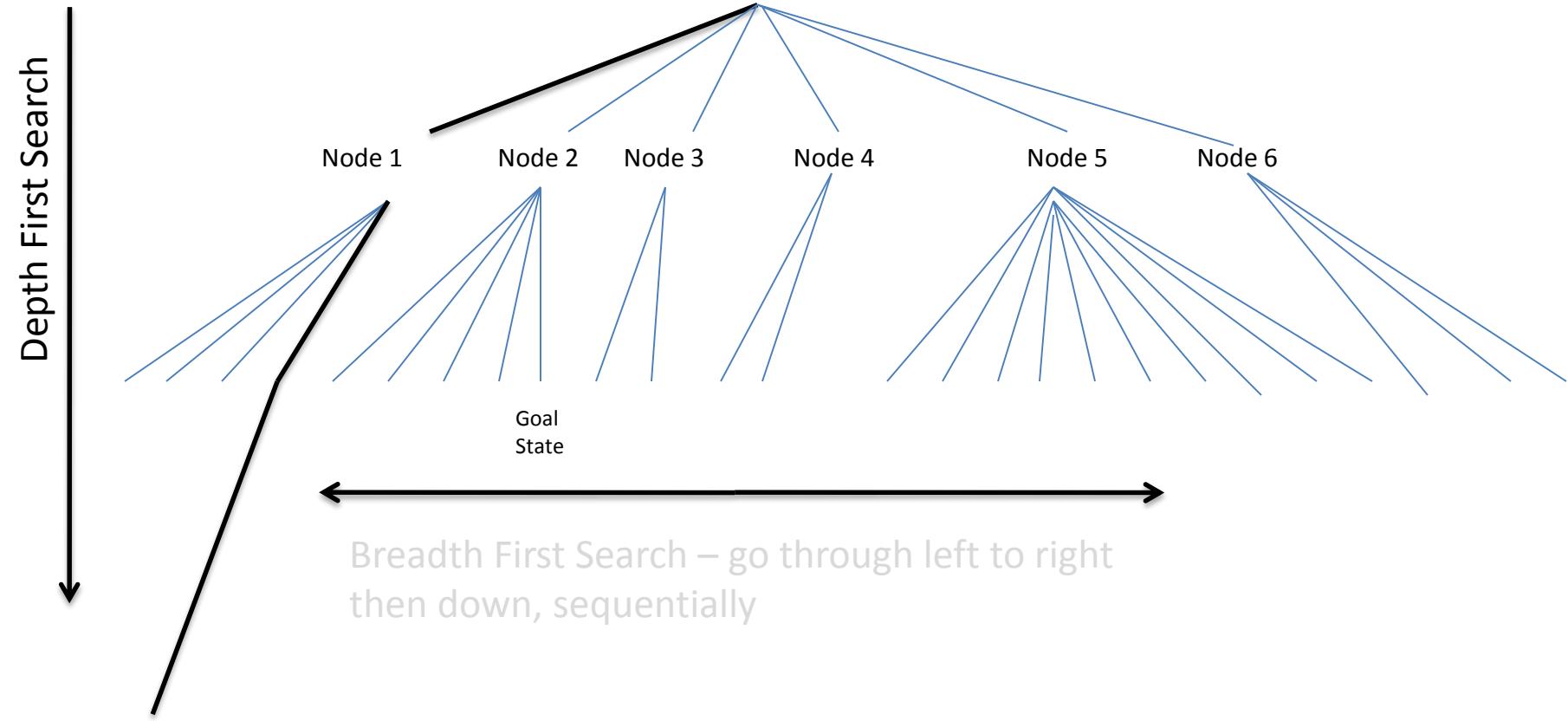
A Breadth First approach to studying would involve a cursory overview of all topics.  
A Depth First approach would involve an in-depth overview of some topics.

# Example: Search Trees



Breadth First Search – go through left to right  
then down, sequentially

# Example: Search Trees



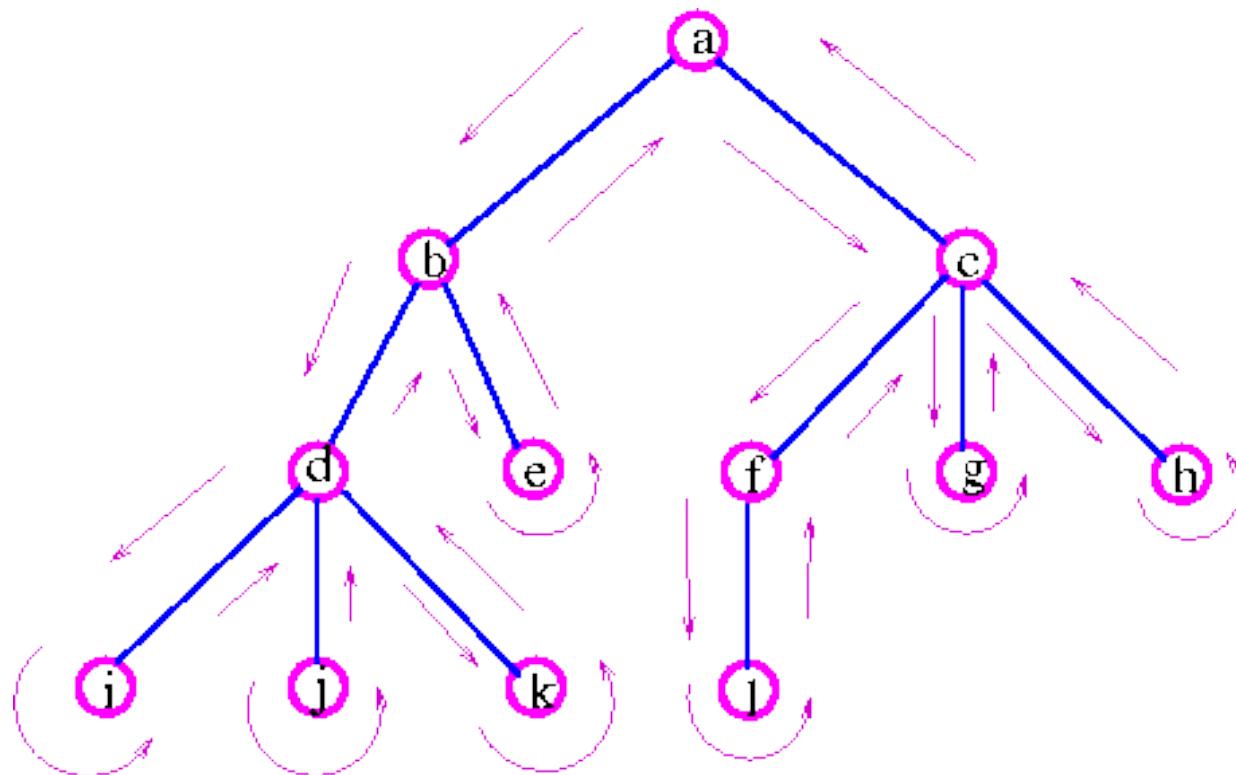
# Which Search Strategy?

- Depends on *a priori* knowledge – heuristics.

# Which Search Strategy?

- Depends on *a priori* knowledge – heuristics.
- You should do both for this final examination – know everything in-depth.

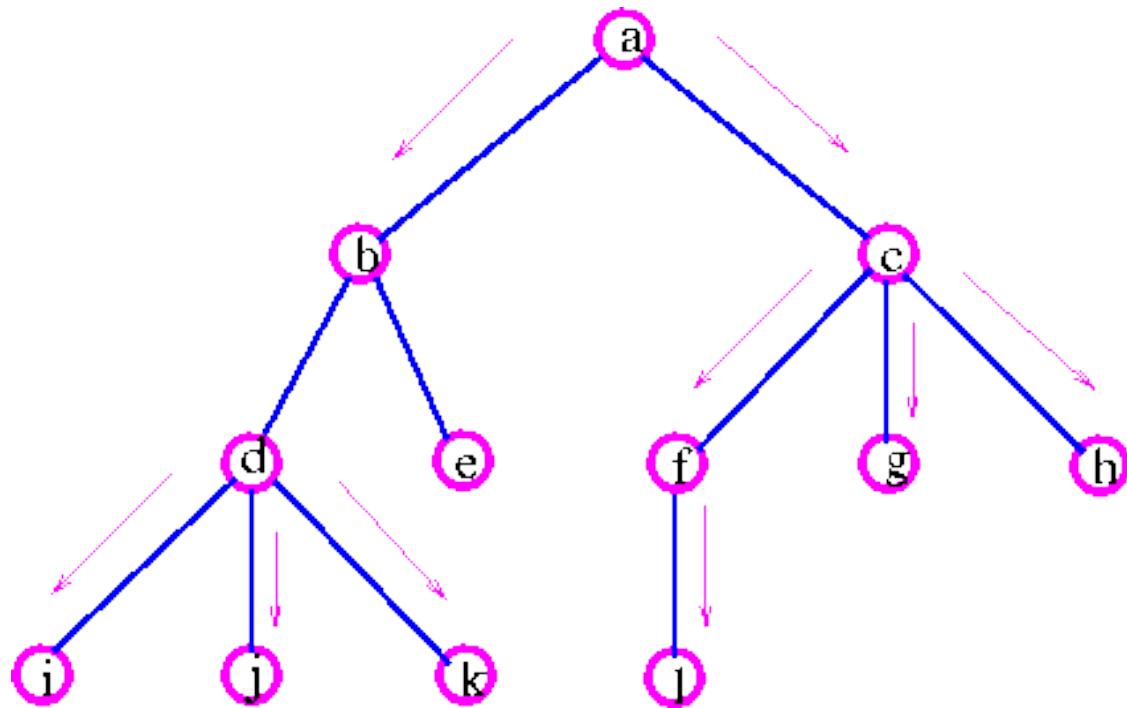
# DFS



# DFS

- **Rule 1** – Visit the adjacent unvisited vertex. Mark it as visited. Display it. Push it in a stack.
- **Rule 2** – If no adjacent vertex is found, pop up a vertex from the stack. (It will pop up all the vertices from the stack, which do not have adjacent vertices.)
- **Rule 3** – Repeat Rule 1 and Rule 2 until the stack is empty.

# BFS



# BFS

- **Rule 1** – Visit the adjacent unvisited vertex. Mark it as visited. Display it. Insert it in a queue.
- **Rule 2** – If no adjacent vertex is found, remove the first vertex from the queue.
- **Rule 3** – Repeat Rule 1 and Rule 2 until the queue is empty.

# Tree Traversal

Traversal is a process to visit all the nodes of a tree and may print their values too. Because, all nodes are connected via edges (links) we always start from the root (head) node. That is, we cannot randomly access a node in a tree. There are three ways which we use to traverse a tree –

- In-order Traversal
- Pre-order Traversal
- Post-order Traversal

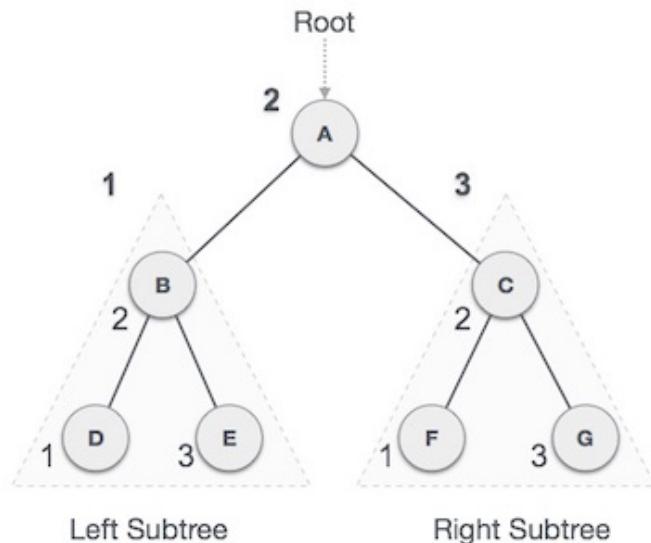
Generally, we traverse a tree to search or locate a given item or key in the tree or to print all the values it contains.

# Tree Traversal

## In-order Traversal

In this traversal method, the left subtree is visited first, then the root and later the right sub-tree. We should always remember that every node may represent a subtree itself.

If a binary tree is traversed **in-order**, the output will produce sorted key values in an ascending order.



We start from **A**, and following in-order traversal, we move to its left subtree **B**. **B** is also traversed in-order. The process goes on until all the nodes are visited. The output of inorder traversal of this tree will be –

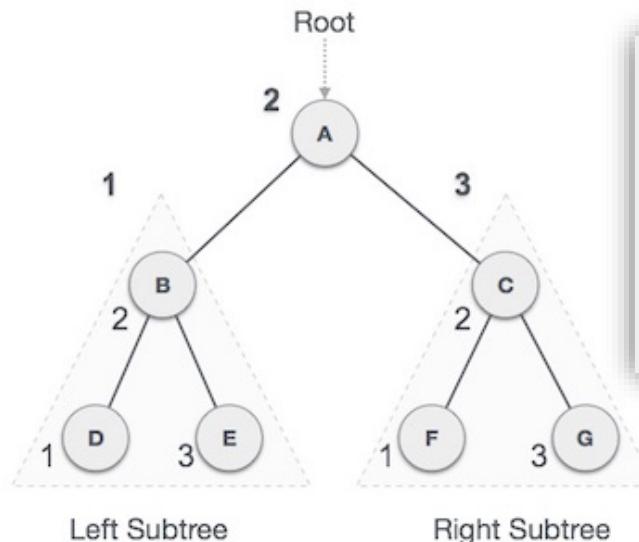
$$D \rightarrow B \rightarrow E \rightarrow A \rightarrow F \rightarrow C \rightarrow G$$

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If a binary tree is traversed **in-order**, the output will produce sorted key values in an ascending order.



## Algorithm

```
Until all nodes are traversed -  
Step 1 - Recursively traverse left subtree.  
Step 2 - Visit root node.  
Step 3 - Recursively traverse right subtree.
```

We start from **A**, and following in-order traversal, we move to its left subtree **B**.

**B** is also traversed in-order. The process goes on until all the nodes are visited.

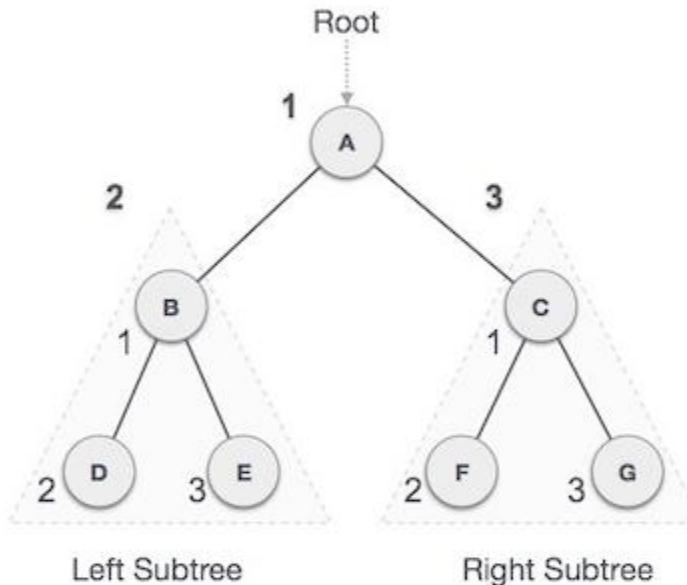
The output of inorder traversal of this tree will be –

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# Tree Traversal

## Pre-order Traversal

In this traversal method, the root node is visited first, then the left subtree and finally the right subtree.



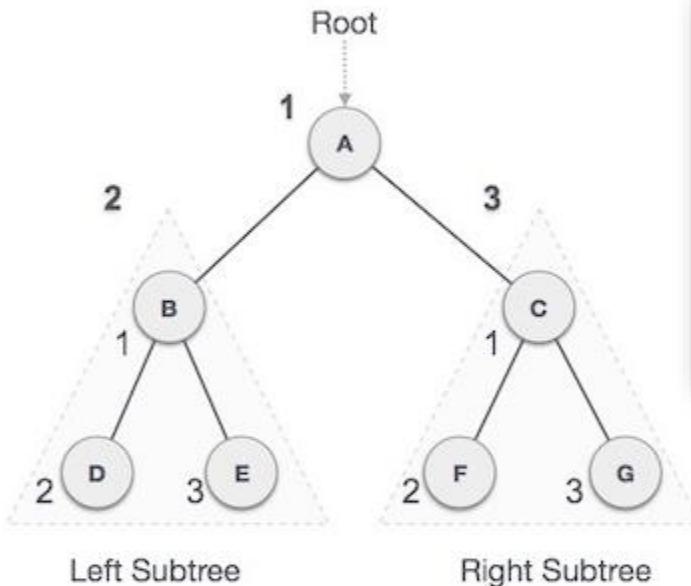
We start from **A**, and following pre-order traversal, we first visit **A** itself and then move to its left subtree **B**. **B** is also traversed pre-order. The process goes on until all the nodes are visited. The output of pre-order traversal of this tree will be –

**A → B → D → E → C → F → G**

# Tree Traversal

## Pre-order Traversal

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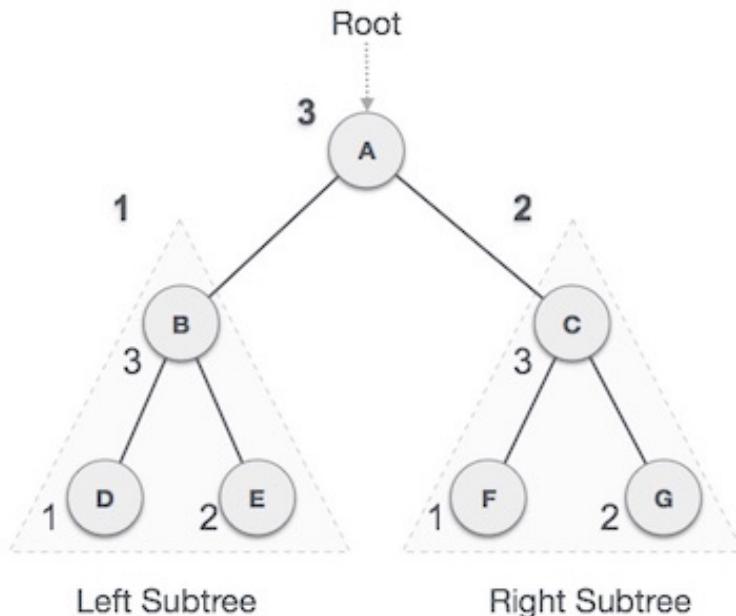
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**A → B → D → E → C → F → G**

# Tree Traversal

## Post-order Traversal

In this traversal method, the root node is visited last, hence the name. First we traverse the left subtree, then the right subtree and finally the root node.



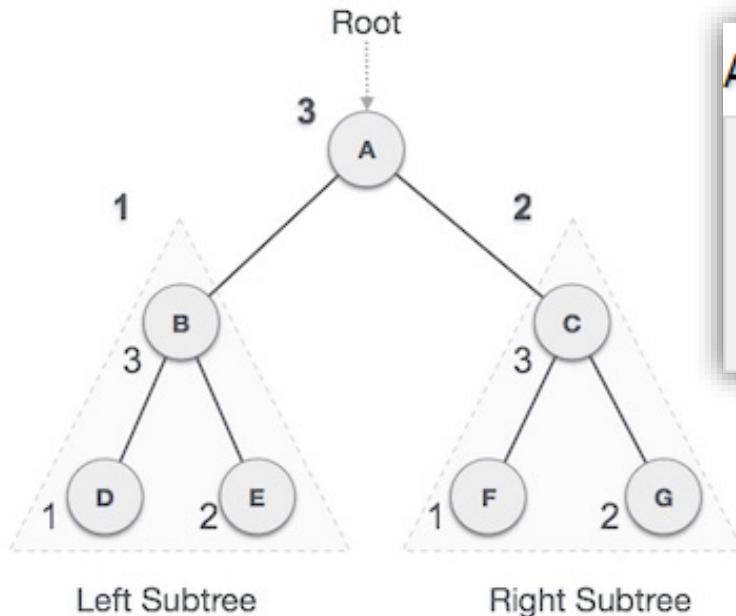
We start from **A**, and following pre-order traversal, we first visit the left subtree **B**. **B** is also traversed post-order. The process goes on until all the nodes are visited. The output of post-order traversal of this tree will be –

$$D \rightarrow E \rightarrow B \rightarrow F \rightarrow G \rightarrow C \rightarrow A$$

# Tree Traversal

## Post-order Traversal

In this traversal method, the root node is visited last, hence the name. First we traverse the left subtree, then the right subtree and finally the root node.



### Algorithm

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Until all nodes are traversed -  
Step 1 - Recursively traverse left subtree.  
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Step 3 - Visit root node.
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We start from **A**, and following pre-order traversal, we first visit the left subtree **B**. **B** is also traversed post-order. The process goes on until all the nodes are visited. The output of post-order traversal of this tree will be –

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# Binary Search Trees (BSTs) as Data Structures

A Binary Search Tree (BST) is a tree in which all the nodes follow the below-mentioned properties –

- The left sub-tree of a node has a key less than or equal to its parent node's key.
- The right sub-tree of a node has a key greater than to its parent node's key.

Thus, BST divides all its sub-trees into two segments; the left sub-tree and the right sub-tree and can be defined as –

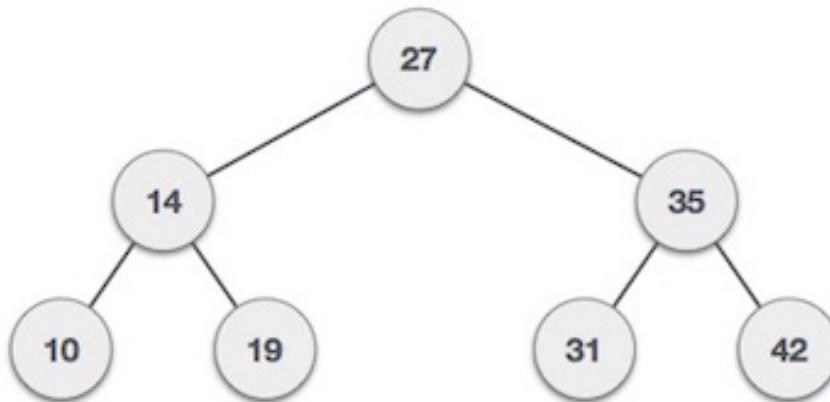
```
left_subtree (keys) ≤ node (key) ≤ right_subtree (keys)
```

# Binary Search Trees (BSTs) as Data Structures

## Representation

BST is a collection of nodes arranged in a way where they maintain BST properties. Each node has a key and an associated value. While searching, the desired key is compared to the keys in BST and if found, the associated value is retrieved.

Following is a pictorial representation of BST –



We observe that the root node key (27) has all less-valued keys on the left sub-tree and the higher valued keys on the right sub-tree.

# Binary Search Trees (BSTs) as Data Structures

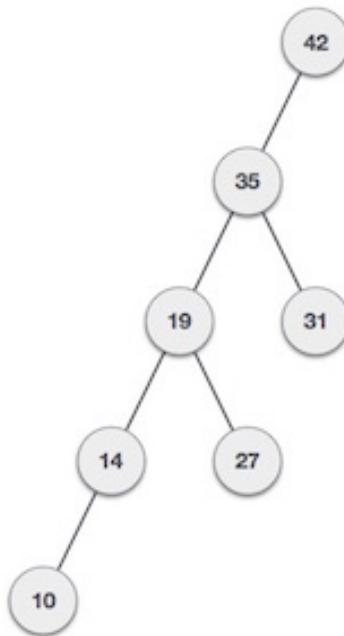
## Basic Operations

Following are the basic operations of a tree –

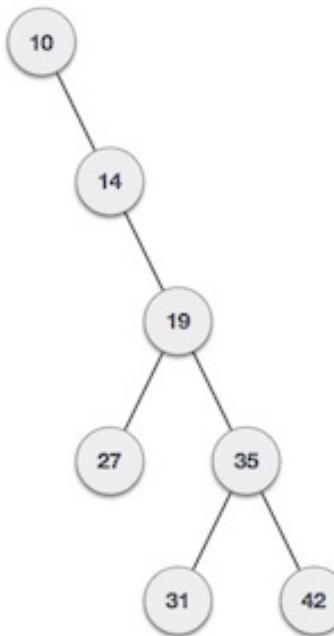
- **Search** – Searches an element in a tree.
- **Insert** – Inserts an element in a tree.
- **Pre-order Traversal** – Traverses a tree in a pre-order manner.
- **In-order Traversal** – Traverses a tree in an in-order manner.
- **Post-order Traversal** – Traverses a tree in a post-order manner.

# Question Regarding BSTs

What if the input to binary search tree comes in a sorted (ascending or descending) manner? It will then look like this –



If input 'appears' non-increasing manner



If input 'appears' in non-decreasing manner

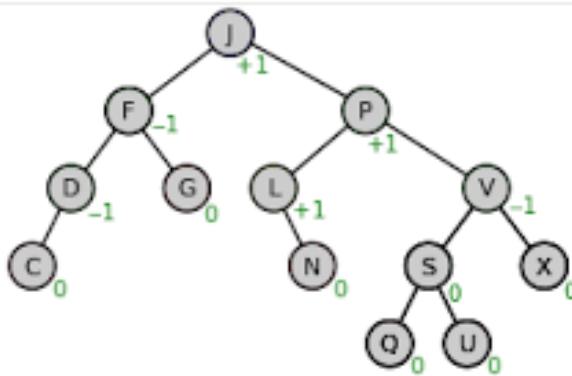
It is observed that BST's worst-case performance is closest to linear search algorithms, that is  $O(n)$ . In real-time data, we cannot predict data pattern and their frequencies. So, a need arises to balance out the existing BST.

# Definition

In computer science, an **AVL tree** is a self-balancing binary search **tree**. It was the first such data structure to be invented. In an **AVL tree**, the heights of the two child subtrees of any node differ by at most one; if at any time they differ by more than one, rebalancing is done to restore this property.

[AVL tree - Wikipedia](#)

[https://en.wikipedia.org/wiki/AVL\\_tree](https://en.wikipedia.org/wiki/AVL_tree)



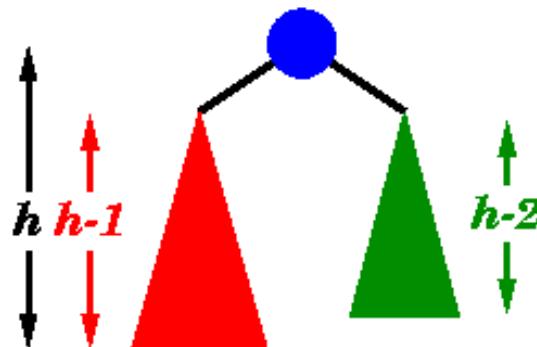
An **AVL tree** is another balanced binary search tree. Named after their inventors, Adelson-Velskii and Landis, they were the first dynamically balanced trees to be proposed. Like red-black trees, they are not perfectly balanced, but pairs of sub-trees differ in height by at most 1, maintaining an  $O(\log n)$  search time. Addition and deletion operations also take  $O(\log n)$  time.

### Definition of an AVL tree

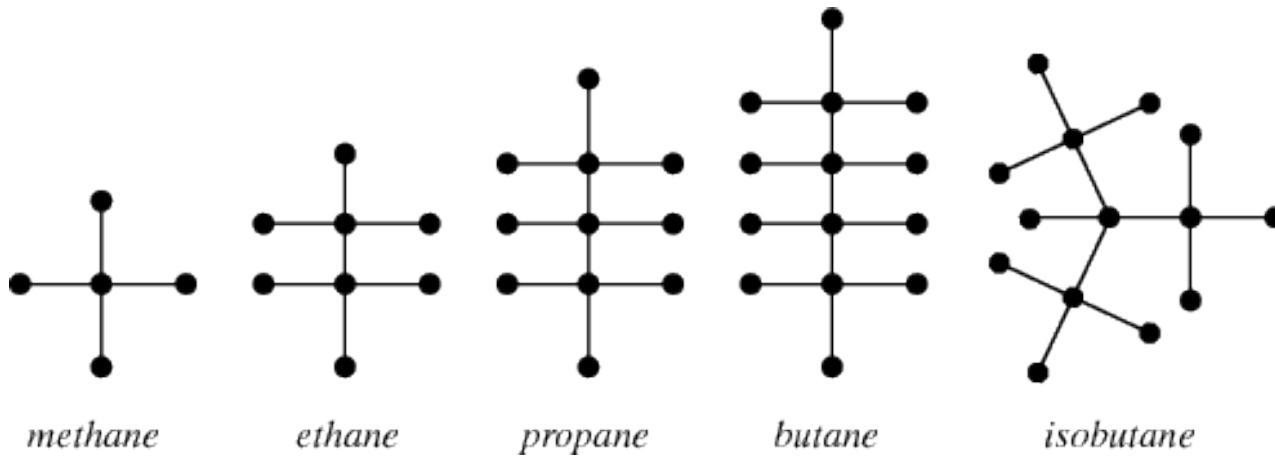
An AVL tree is a binary search tree which has the following properties:

1. The sub-trees of every node differ in height by at most one.
2. Every sub-tree is an AVL tree.

Balance requirement for an AVL tree: the left and right sub-trees differ by at most 1 in height.

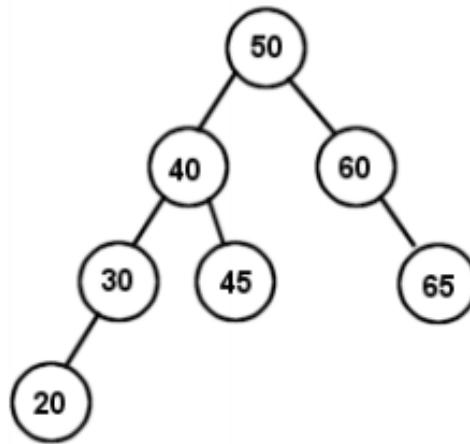


# Application of Trees: Biochemistry



The enumeration of saturated hydrocarbons (saturation just refers to the fact that each carbon has its maximum number of bonds to hydrogen).

# Application of Trees: Sorting



*Create tree via:*

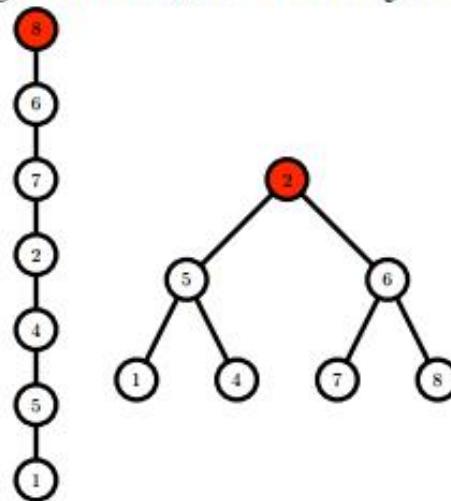
- *First number is the root.*
- *Put number in tree by traversing to an end vertex*
  - *If number less than or equal vertex number go left branch*
  - *If number greater than vertex number go right branch*

*Tree above for the sequence: 50 40 60 30 20 65 45*

# Application of Trees: Data Structures

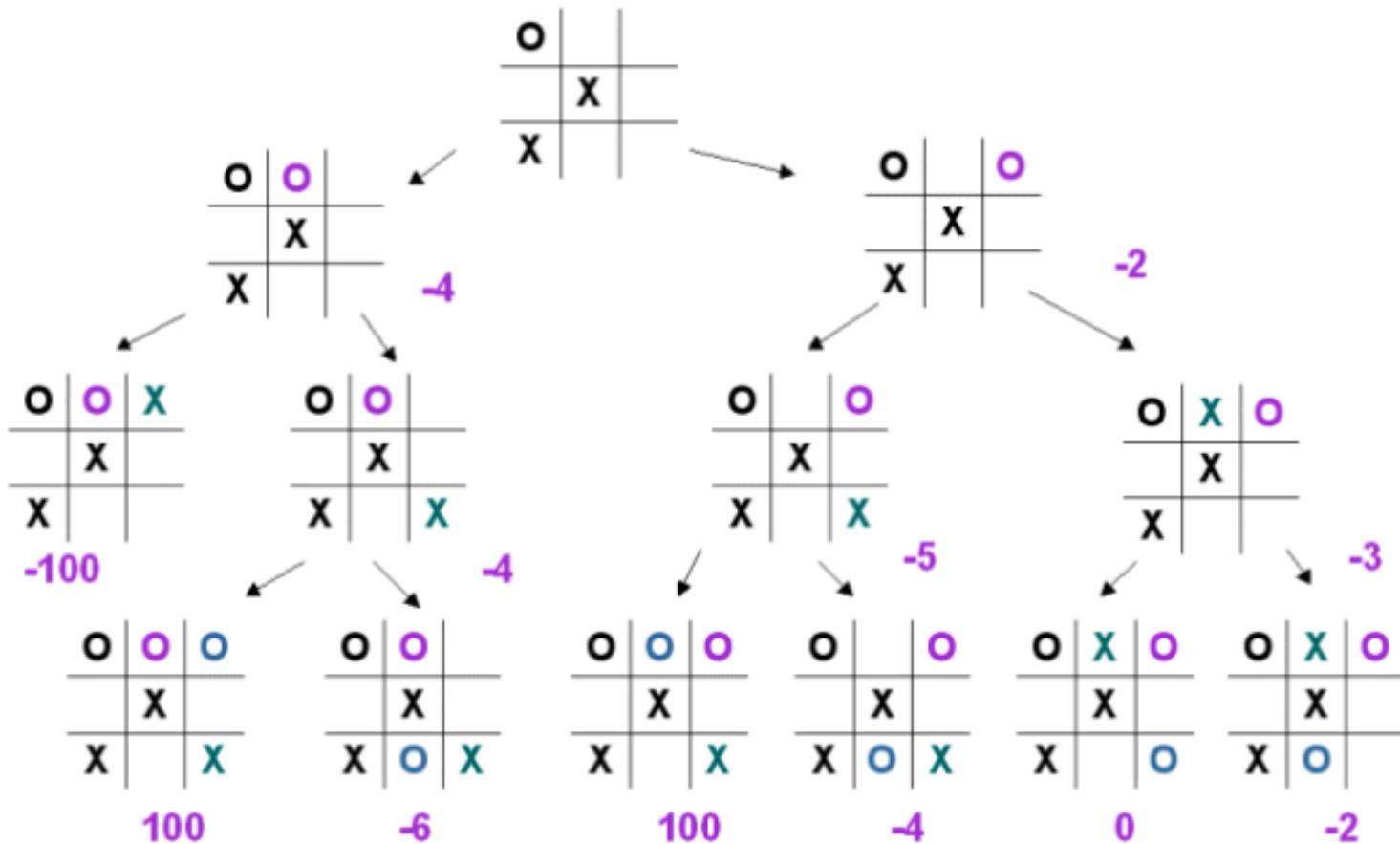
Rooted trees can be used to store data in a computer's memory in many different ways.

*Consider a list of seven numbers 1, 5, 4, 2, 7, 6, 8. The following trees show two ways of storing this data, as a binary tree and as a stack.*



*Both trees are rooted trees and both representations have their advantages. However it is important in both cases to know the starting point of the data, i.e. **the root**.*

# Application of Trees: Games



# Presentation Terminated

