

Unit 4.1 Breadth First Search

Algorithms

EE3980

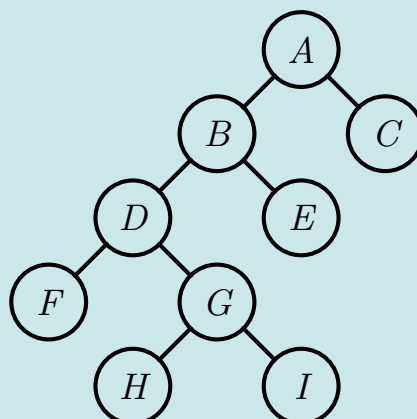
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Binary Tree Traversal

- Given a binary tree, some applications need to visit every node of the tree.
- It is assumed that each node of the tree has the underlying structure as

```
1 struct treenode {  
2     Type data; // store data of specified Type  
3     treenode *lchild, *rchild;  
4 }
```

- Example

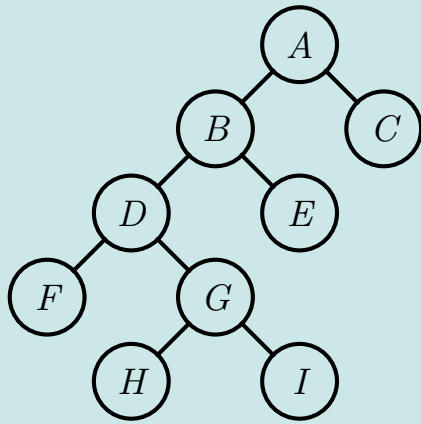


- Three ways to traverse a binary tree

Binary Tree — In-order Traversal

Algorithm 4.1.1. In Order Traversal

```
1 Algorithm InOrder(T)
2 // To visit every node of the binary tree in-order.
3 {
4     if (T ≠ NULL) then {
5         InOrder(T → lchild);
6         Visit(T);
7         InOrder(T → rchild);
8     }
9 }
```



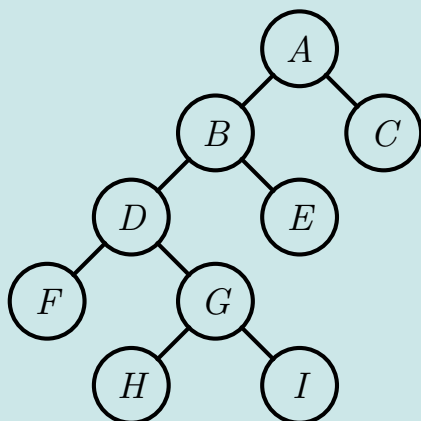
• Execution sequence

InOrder A	visit G
InOrder B	InOrder I
InOrder D	visit I
InOrder F	visit B
visit F	InOrder E
visit D	visit E
InOrder G	visit A
InOrder H	InOrder C
visit H	visit C

Binary Tree — Pre-order Traversal

Algorithm 4.1.2. Pre-Order Traversal

```
1 Algorithm PreOrder(T)
2 // To visit every node of the binary tree pre-order.
3 {
4     if (T ≠ NULL) then {
5         Visit(T);
6         PreOrder(T → lchild);
7         PreOrder(T → rchild);
8     }
9 }
```



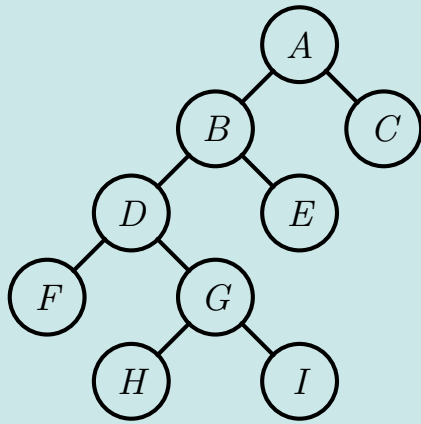
• Execution sequence

PreOrder A	visit G
visit A	PreOrder H
PreOrder B	visit H
visit B	PreOrder I
PreOrder D	visit I
visit D	PreOrder E
PreOrder F	visit E
visit F	PreOrder C
PreOrder G	visit C

Binary Tree — Post-order Traversal

Algorithm 4.1.3. Post-Order Traversal

```
1 Algorithm PostOrder(T)
2 // To visit every node of the binary tree post-order.
3 {
4   if (T ≠ NULL) then {
5     PostOrder(T → lchild);
6     PostOrder(T → rchild);
7     Visit(T);
8   }
9 }
```



• Execution sequence

PostOrder A	visit I
PostOrder B	visit G
PostOrder D	visit D
PostOrder F	PostOrder E
visit F	visit E
PostOrder G	visit B
PostOrder H	PostOrder C
visit H	visit C
PostOrder I	visit A

Binary Tree Traversal — Complexities

- In traversing the tree, each node is reached three times
 - From its root; when returning from *lchild* and *rchild*
- Thus, the time complexity is $T(n) = \Theta(n)$ for an n -node binary tree.
- The space needed for an n -node binary tree is $\Theta(n)$.
- Traversing the tree using recursive calls would need a heap space proportional to the depth, d , of the tree.
- Since $d \leq n$, the space complexity is $\mathcal{O}(n)$.

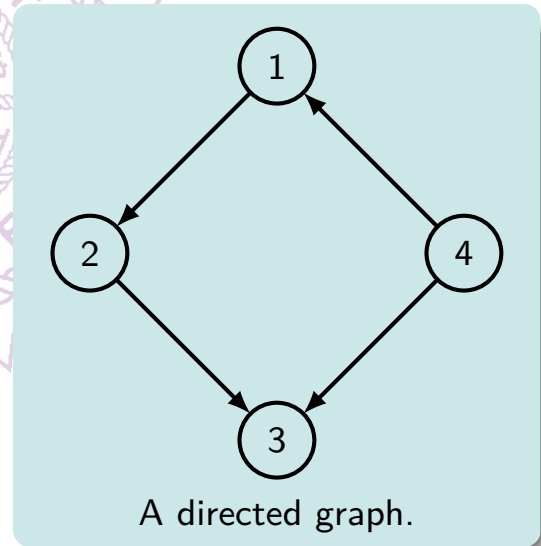
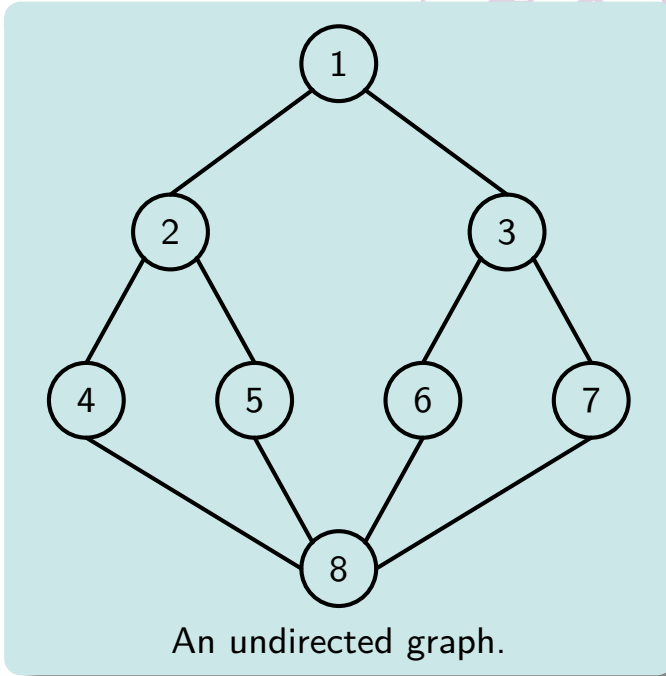
Theorem 4.1.4. Binary Tree Traversal

Let $T(n)$ and $S(n)$ be the time and space complexities of any of the binary traversing algorithms above, then $T(n) = \Theta(n)$ and $S(n) = \mathcal{O}(n)$.

- Proof, please see textbood [Horowitz], pp. 335-337.

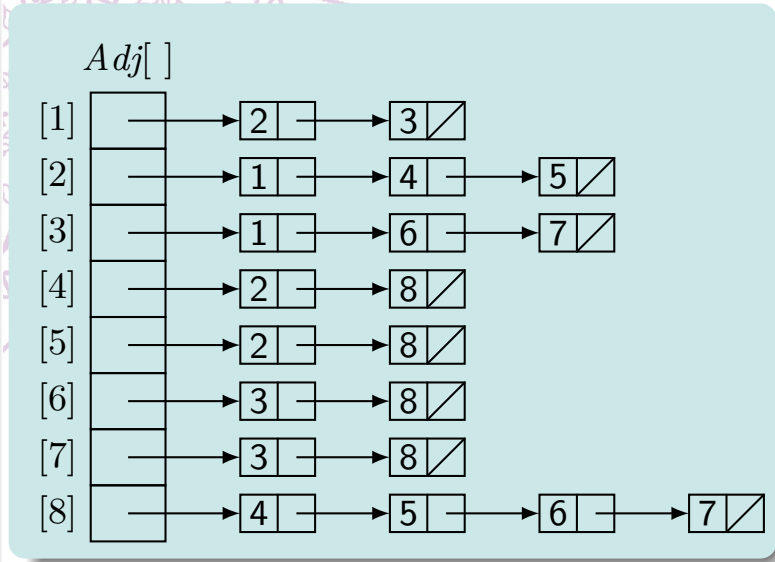
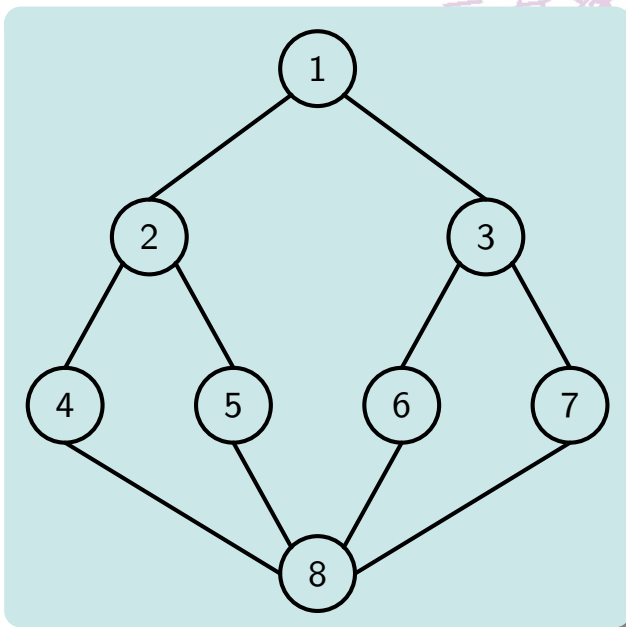
Graph Traversal

- Given a graph $G = (V, E)$ with vertex set V and edge set E , a typical graph traversal problem is to find all vertices that is reachable from a particular vertex, for example $v \in V$.
 - Note that G can be either a directed graph or undirected graph.



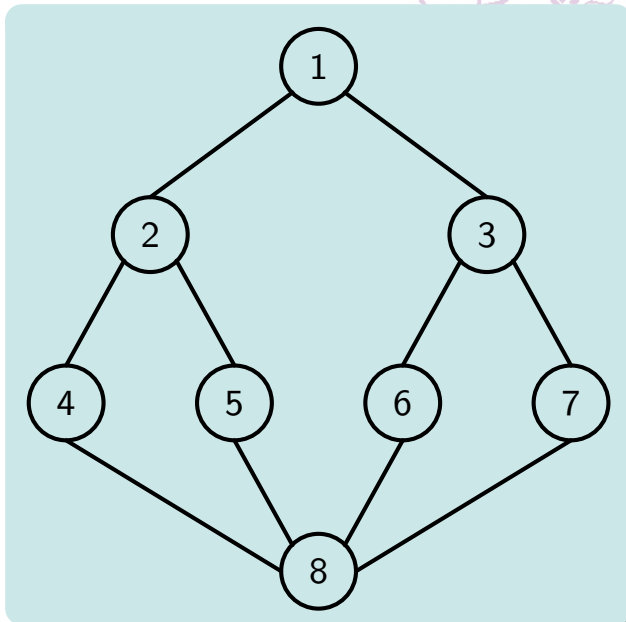
Graph and Adjacency Lists

- One way to represent the adjacency information of a graph $G = (V, E)$ is the **adjacency list**.
 - Both directed and undirected graphs can be represented.
 - In an undirected graph, each edge should appear twice.
 - More efficient if the graph is sparse, $|E| \ll |V|^2$.
 - Weighted graphs can also be represented with more space for each edge.



Graph and Adjacency Matrix

- The other way to keep the adjacent information of a graph $G = (V, E)$ is the **adjacency matrix**.
 - For undirected graphs, symmetric matrices are obtained.
 - Asymmetric matrices for directed graphs.
 - Weighted graphs can also be represented.
 - More applicable when the graph is dense, $|E| \approx |V|^2$, or faster search of an edge (i, j) is needed.



$$A[i, j] = \begin{cases} 1 & \text{if } (i, j) \in E, \\ 0 & \text{otherwise.} \end{cases} \quad (4.1.1)$$

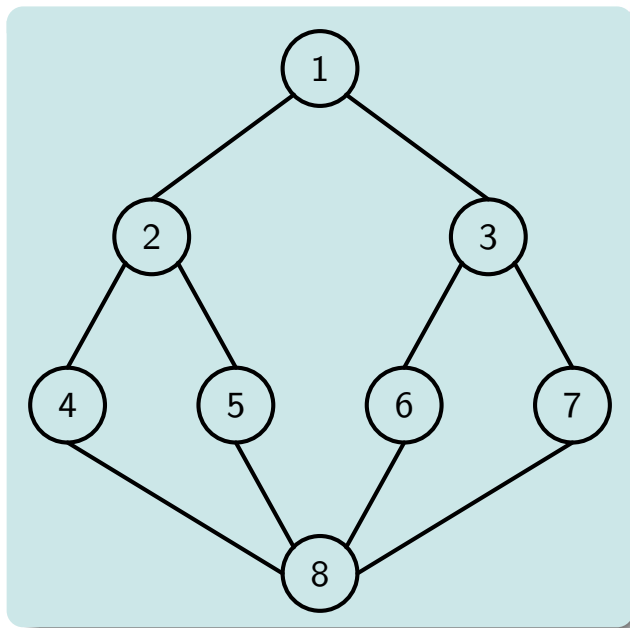
	1	2	3	4	5	6	7	8
1	0	1	1	0	0	0	0	0
2	1	0	0	1	1	0	0	0
3	1	0	0	0	0	1	1	0
4	0	1	0	0	0	0	0	1
5	0	1	0	0	0	0	0	1
6	0	0	1	0	0	0	0	1
7	0	0	1	0	0	0	0	1
8	0	0	0	1	1	1	1	0

Breadth First Search

- A popular graph traversal algorithm for both directed and undirected graphs is

Algorithm 4.1.5. Breadth First Search

```
1 Algorithm BFS( $v$ )
2 // Breadth first search starting from vertex  $v$  of graph  $G$ .
3 //  $Q$  is assume to be a queue. Array  $visited$  is initialized to 0.
4 {
5      $u := v$ ;  $visited[v] := 1$ ;
6     repeat {
7         for all vertices  $w$  adjacent to  $u$  do {
8             if ( $visited[w] = 0$ ) then {
9                 Enqueue( $w$ );  $visited[w] := 1$ ;
10            }
11        }
12        if not Qempty() then  $u := Dequeue()$ ; // get the next vertex.
13    } until ( Qempty());
14 }
```



- **BFS** calling sequence

visit 1	Queue = (2, 3)
visit 2	Queue = (3, 4, 5)
visit 3	Queue = (4, 5, 6, 7)
visit 4	Queue = (5, 6, 7, 8)
visit 5	Queue = (6, 7, 8)
visit 6	Queue = (7, 8)
visit 7	Queue = (8)
visit 8	Queue = ()

Breadth First Search – Properties

Theorem 4.1.6. BFS Complexities

Let $T(n, e)$ and $S(n, e)$ be the maximum time and maximum *additional* space taken by algorithm **BFS** on any graph G with n vertices and e edges.

1. $T(n, e) = \Theta(n + e)$ and $S(n, e) = \Theta(n)$ if G is represented by its adjacency lists,
2. $T(n, e) = \Theta(n^2)$ and $S(n, e) = \Theta(n)$ if G is represented by its adjacency matrix.

- Proof please see textbook [Horowitz], pp. 341-343.
 - The additional space refers to array $v[1, n]$, $\Theta(n)$, and memory needed for the queue, $\mathcal{O}(n)$.

Theorem 4.1.7. BFS Reachability

Algorithm **BFS** visits all vertices of G reachable from v .

- Proof please see textbook [Horowitz], p. 340.

Definition 4.1.8. Shortest Path.

Given a graph $G = (V, E)$, the **shortest-path distance**, $\delta(s, v)$, between any two vertices, $s, v \in V$, is the minimum number of edges in any path from s to v . If there is no path from s to v then $\delta(s, v) = \infty$. A path of length $\delta(s, v)$ from s to v is a **shortest path** from s to v .

Lemma 4.1.9.

Given a directed or undirected graph $G = (V, E)$ and an arbitrary vertex $s \in V$, then for any edge $(u, v) \in E$ we have

$$\delta(s, v) \leq \delta(s, u) + 1. \quad (4.1.2)$$

- Proof please see textbook [Cormen], p. 598.

Shortest Path and Breadth First Search

- The breadth first search algorithm can be modified to find the shortest distance to other vertices.

Algorithm 4.1.10. Shortest path – Breadth First Search

```
1 Algorithm BFS_d(v, d)
2 // Breadth first search starting with path length.
3 // Array d records the shortest path length from vertex v.
4 // Array p records the preceding vertex of the shortest path.
5 {
6     u := v; visited[v] := 1; d[v] := 0; p[v] := 0;
7     repeat {
8         for all vertices w adjacent to u do {
9             if (visited[w] = 0) then {
10                Enqueue(w); visited[w] := 1; d[w] := d[u] + 1; p[w] := u;
11            }
12        }
13        if not Qempty() then u := Dequeue(); // Get the next vertex.
14    } until ( Qempty());
15 }
```

Shortest Path and Breadth First Search, II

Lemma 4.1.11.

Given a graph $G = (V, E)$, if the `BFS_d`(s, d) is called for a source vertex $s \in V$, then upon the termination of the algorithm we have for any $v \in V$, $d[v] \geq \delta(s, v)$.

- Proof please see textbook [Cormen], p. 598.

Lemma 4.1.12.

Suppose that during the execution of the `BFS_d`(s, d) algorithm on a graph $G = (V, E)$, the queue Q contains the vertices $\langle v_1, v_2, \dots, v_r \rangle$, where v_1 is the head of the queue and v_r is the tail. Then, we have

$$d[v_r] \leq d[v_1] + 1, \quad (4.1.3)$$

$$d[v_i] \leq d[v_{i+1}] \quad \text{for } i = 1, 2, \dots, r-1. \quad (4.1.4)$$

- Proof please see textbook [Cormen], p. 599.

Shortest Path and Breadth First Search, III

Corollary 4.1.13.

Suppose that during the execution of the `BFS_d`(s, d) algorithm on a graph $G = (V, E)$, both vertices v_i and v_j are enqueue and v_i is enqueued before v_j , then $d[v_i] \leq d[v_j]$.

- Proof please see textbook [Cormen], p. 599.

Theorem 4.1.14.

Given a graph $G = (V, E)$ and a source vertex $s \in V$, if the algorithm `BFS_d`(s, d) is called, then for every vertex $v \in V$ reachable from s , upon termination we have $d[v] = \delta(s, v)$.

- Proof please see textbook [Cormen], p. 600.

Shortest Path and Breadth First Search – Print Path

- A shortest path from source s to any vertex $v \in V$ can be printed using the array p .
 - Note that array p records the predecessor information.
 - $p[w]$ is the vertex preceding vertex w in the shortest path.
 - For source vertex v , $p[v] = 0$.

Algorithm 4.1.15. Print Shortest Path

```
1 Algorithm BFSpath( $w$ )
2 // To print the shortest path that ends at  $w$ .a
3 //  $p$  is the predecessor array.
4 {
5     if ( $p[w] \neq 0$ ) BFSpath( $p[w]$ );
6     write ( " w " );
7 }
```

Spanning Trees of Connected Graphs

- The **BFS** algorithm can be modified to find the spanning tree of a connected graph.

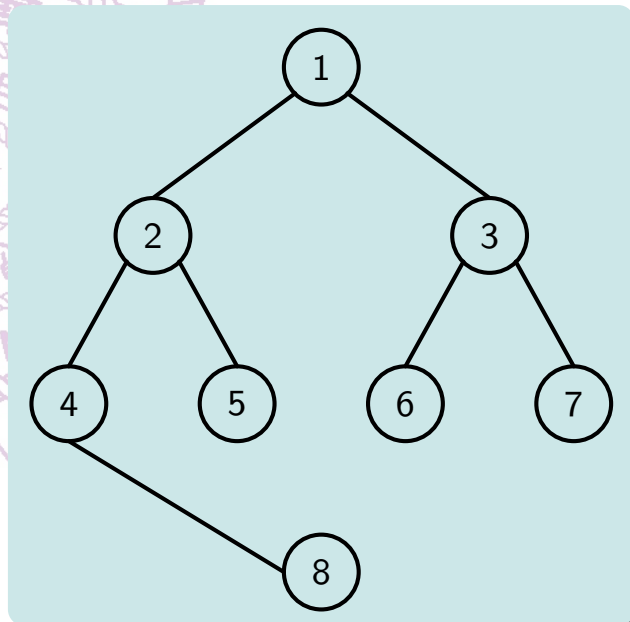
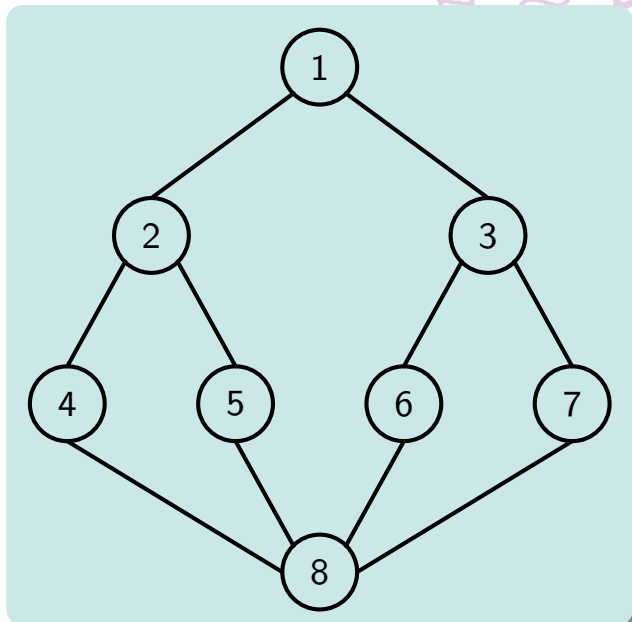
Algorithm 4.1.16. BFS to find a spanning tree

```
1 Algorithm BFS*( $v$ )
2 // Breadth first search to find the spanning tree from vertex  $v$ .
3 {
4      $u := v$ ;  $visited[v] := 1$ ;  $t := \emptyset$ ;
5     repeat {
6         for all vertices  $w$  adjacent to  $u$  do {
7             if ( $visited[w] = 0$ ) then {
8                 Enqueue( $w$ );  $visited[w] := 1$ ;  $t := t \cup \{(u, w)\}$ ;
9             }
10        }
11        if not Qempty() then  $u := Dequeue(u)$ ; // Get the next vertex.
12    } until ( Qempty());
13 }
```

- On termination, t is the set of edges that forms a spanning tree of G .

BFS Spanning Tree

- The spanning tree found by Algorithm **BFS*** can be called **BFS spanning tree**.
- This tree has the property that the path from the root s to any vertex $v \in V$ is a shortest path.
- Example



- The time and space complexity of **BFS*** is the same as **BFS**.

Summary

- Binary tree traversal
- Graph traversal
- Breadth first search
- Spanning tree

