



# Graph Traversal Techniques: BFS and DFS

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MTL 776  
Graph Algorithms



# Graph Representation

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Two popular computer representations of a graph. Both represent the vertex set and the edge set, but in different ways.

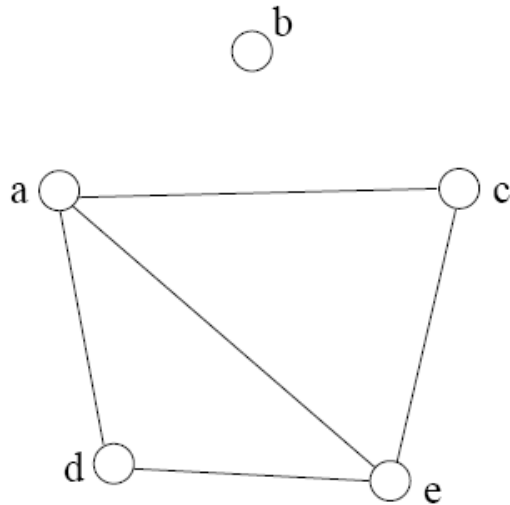
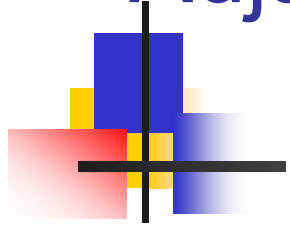
1. **Adjacency Matrix**

Use a 2D matrix to represent the graph

2. **Adjacency List**

Use a 1D array of linked lists

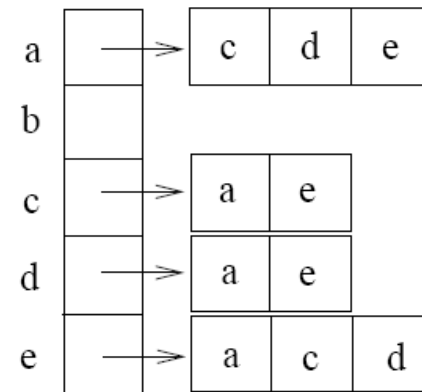
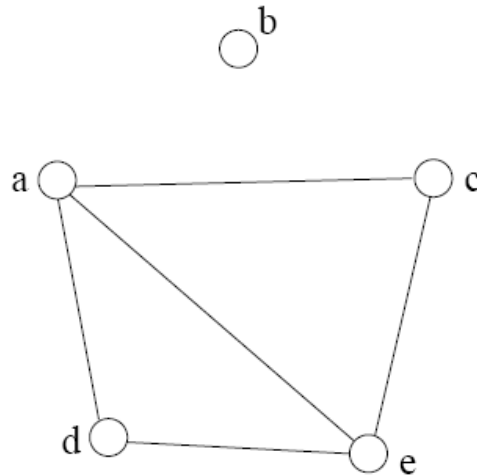
# Adjacency Matrix



	a	b	c	d	e
a	0	0	1	1	1
b	0	0	0	0	0
c	1	0	0	0	1
d	1	0	0	0	1
e	1	0	1	1	0

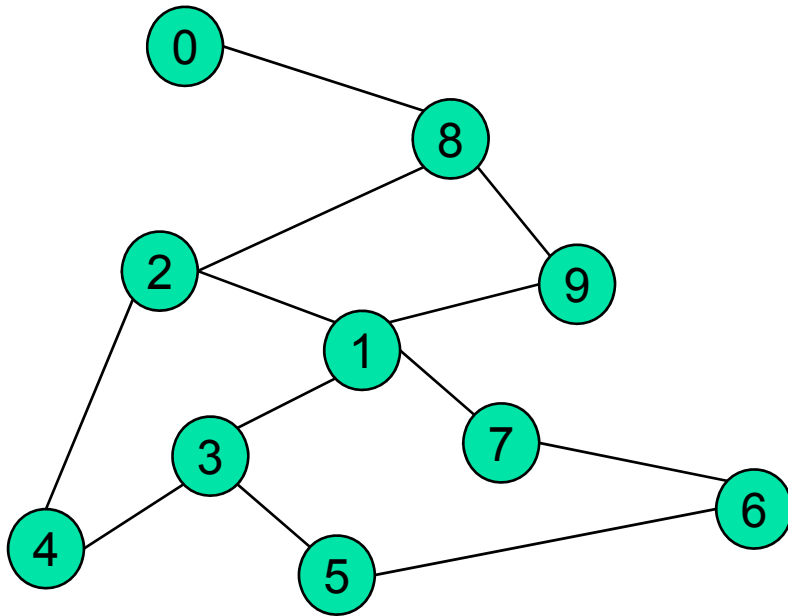
- 2D array  $A[0..n-1, 0..n-1]$ , where  $n$  is the number of vertices in the graph
- Each row and column is indexed by the vertex id
  - e.g a=0, b=1, c=2, d=3, e=4
- $A[i][j]=1$  if there is an edge connecting vertices  $i$  and  $j$ ; otherwise,  $A[i][j]=0$
- The storage requirement is  $\Theta(n^2)$ . It is not efficient if the graph has few edges. An adjacency matrix is an appropriate representation if the graph is dense:  $|E|=\Theta(|V|^2)$
- We can detect in  $O(1)$  time whether two vertices are connected.

# Adjacency List



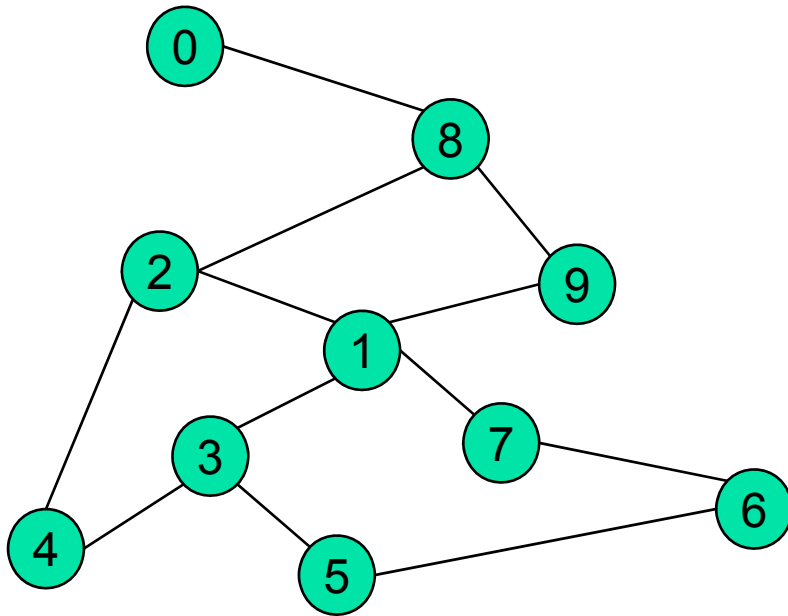
- If the graph is not dense, in other words, **sparse**, a better solution is an adjacency list
- The adjacency list is **an array  $A[0..n-1]$  of lists**, where  $n$  is the number of vertices in the graph.
- Each array entry is indexed by the vertex id
- Each **list  $A[i]$**  stores the **ids of the vertices adjacent to vertex  $i$**

# Adjacency Matrix Example



	0	1	2	3	4	5	6	7	8	9
0	0	0	0	0	0	0	0	0	1	0
1	0	0	1	1	0	0	0	1	0	1
2	0	1	0	0	1	0	0	0	1	0
3	0	1	0	0	1	1	0	0	0	0
4	0	0	1	1	0	0	0	0	0	0
5	0	0	0	1	0	0	1	0	0	0
6	0	0	0	0	0	1	0	1	0	0
7	0	1	0	0	0	0	1	0	0	0
8	1	0	1	0	0	0	0	0	0	1
9	0	1	0	0	0	0	0	0	1	0

# Adjacency List Example



0	→	8
1	→	2 3 7 9
2	→	1 4 8
3	→	1 4 5
4	→	2 3
5	→	3 6
6	→	5 7
7	→	1 6
8	→	0 2 9
9	→	1 8

# Storage of Adjacency List

The array takes up  $\Theta(n)$  space

- Define **degree** of  $v$ ,  $\deg(v)$ , to be the number of edges incident to  $v$ . Then, the total space to store the graph is proportional to:

$$\sum_{\text{vertex } v} \deg(v)$$

- An edge  $e=\{u,v\}$  of the graph contributes a count of 1 to  $\deg(u)$  and contributes a count 1 to  $\deg(v)$
- Therefore,  $\sum_{\text{vertex } v} \deg(v) = 2m$ , where  $m$  is the total number of edges
- In all, the **adjacency list takes up  $\Theta(n+m)$  space**
  - If  $m = O(n^2)$  (i.e. dense graphs), both adjacent matrix and adjacent lists use  $\Theta(n^2)$  space.
  - If  $m = O(n)$ , adjacent list outperform adjacent matrix
- However, one cannot tell in  $O(1)$  time whether two vertices are connected



# Adjacency List vs. Matrix

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## ■ Adjacency List

- More compact than adjacency matrices if graph has few edges
- Requires more time to find if an edge exists

## ■ Adjacency Matrix

- Always require  $n^2$  space
  - This can waste a lot of space if the number of edges are sparse
- Can quickly find if an edge exists





# Generic Search

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## Algorithm Generic GraphSearch

```
{ for i=1 to n do {Mark[i]=0; P[i]=0; Num[i]=0; Com[i]=0;} count=1; comp=0;
For i=1 to n do { if (Mark[i]==0) comp++;search(i);}
```

### **Search(i){**

```
Mark[i]=1; P[i]=-1; Num[i]=count; Count++; com[i]=comp;
```

```
S={i};
```

```
While ( S ≠ emptyset){
```

```
    select a vertex x from S;
```

```
    if ( x has an unmarked neighbor y)
```

```
        { Mark[y]=1; P[y]=x; Num[y]=count; count=count+1;
```

```
          S=S ∪ {y};com[y]=comp;
```

```
        }
```

```
    else S=S-{x}; }
```

```
}
```



# Queue implementation of Generic Search: BFS

## Algorithm BFS

```
{ for i=1 to n do {Mark[i]=0; P[i]=0; Num[i]=0;} count=1;
For i=1 to n do { if (Mark[i]==0) search(i); BFS(i);} Q=createQueue();
BFS(i) search(i){
    Mark[i]=1; P[i]=-1; Num[i]=count; Count++;
    S={i}; Enqueue(Q,i);
    While ( S !=empty lseemptyqueue(Q)!=0){
        select a vertex x from S; x=Front(Q);
        if ( x has an unmarked neighbor y)
            { Mark[y]=1; P[y]=x; Num[y]=count; count=count+1;
              S=S ∪ {y}; Enqueue(Q,y);
            }
        else S=S-{x}; Dequeue(Q); }
    }
}
```

} Note that green colored text is replaced with red colored text in generic search



# Stack implementation of Generic Search: DFS

## Algorithm DFS

```
{ for i=1 to n do {Mark[i]=0; P[i]=0; Num[i]=0;} count=1;
For i=1 to n do { if (Mark[i]==0) search(i); DFS(i);} S=createStack();
Search(i) DFS(i){
    Mark[i]=1; P[i]=-1; Num[i]=count; Count++;
    S={i}; Push(S,i);
    While ( S !=emptyset lseempty(S)!=0){
        select a vertex x from S;
        if ( x has an unmarked neighbor y)
            { Mark[y]=1; P[y]=x; Num[y]=count; count=count+1;
              S=S ∪ {y}; Push(S,y);
            }
        else S=S-{x}; Pop(S); }
    }
```

} Note that green colored text is replaced with red colored text in generic search



# BFS Algorithm

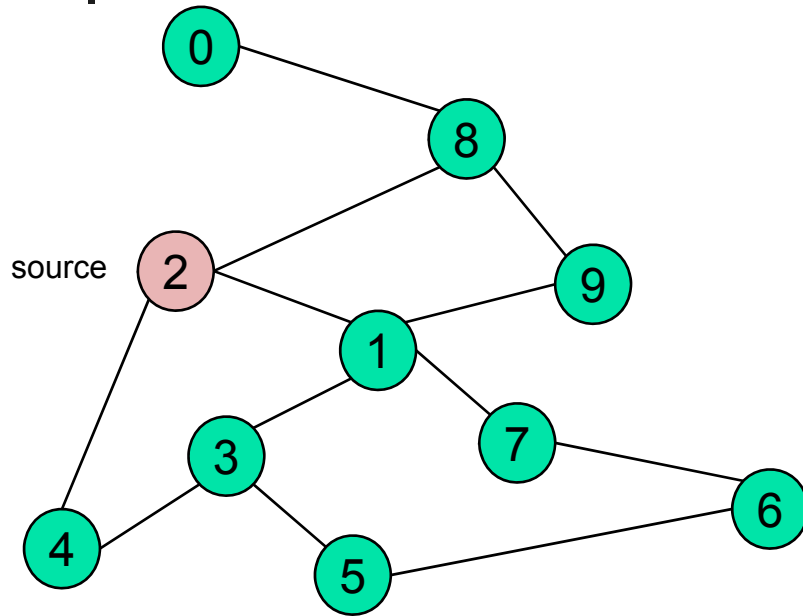
**Algorithm**  $BFS(s)$

**Input:**  $s$  is the source vertex

**Output:** Mark all vertices that can be visited from  $s$ .

1. **for** each vertex  $v$
2.     **do**  $flag[v] := \text{false}$ ;   **// flag[ ]: visited table**
3.      $Q = \text{empty queue}$ ;       **Why use queue? Need FIFO**
4.      $flag[s] := \text{true}$ ;
5.      $enqueue(Q, s)$ ;
6.     **while**  $Q$  is not empty
7.         **do**  $v := dequeue(Q)$ ;
8.         **for** each  $w$  adjacent to  $v$
9.             **do if**  $flag[w] = \text{false}$
10.                 **then**  $flag[w] := \text{true}$ ;
11.                  $enqueue(Q, w)$

# BFS Example



$Q = \{ \}$

Initialize **Q** to be empty

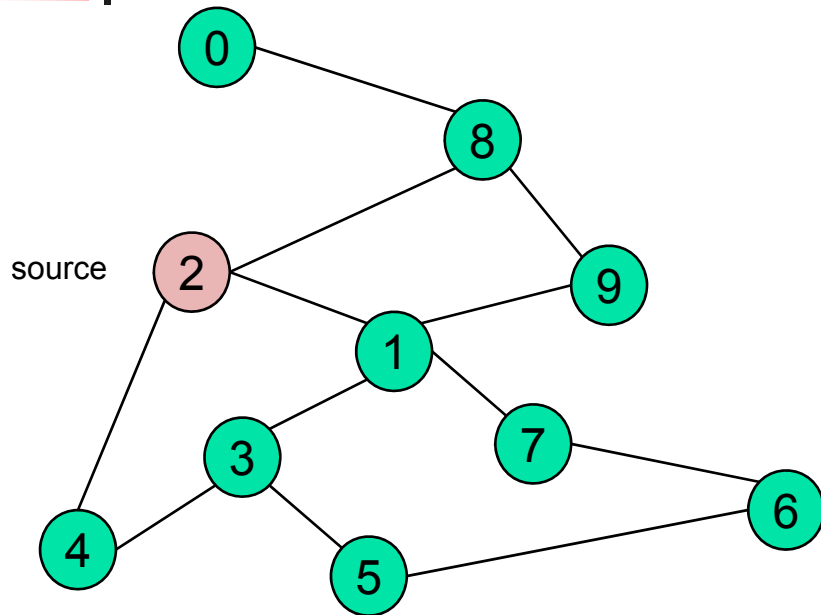
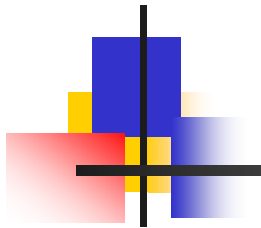
Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

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

Initialize visited  
table (all False)



$Q = \{ 2 \}$

Place source 2 on the queue

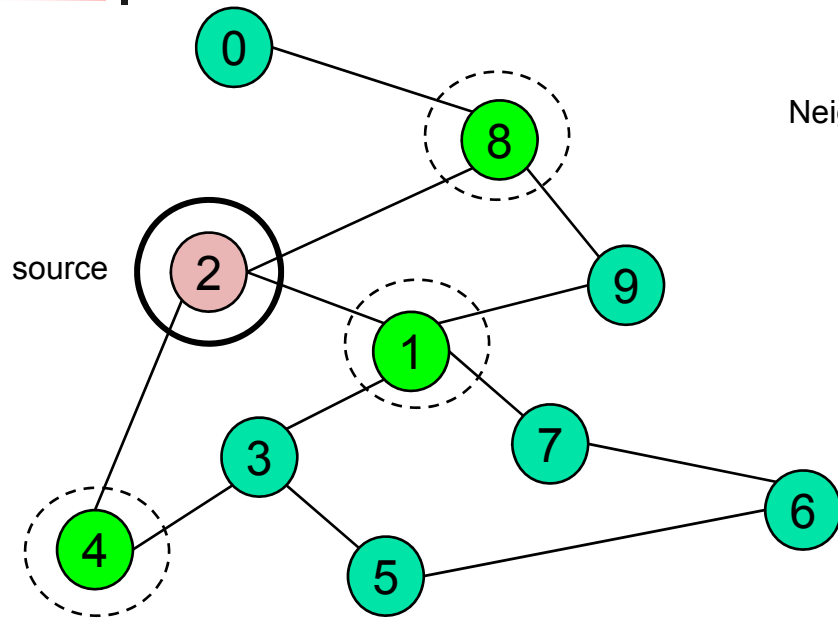
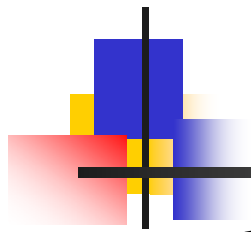
Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	F
1	F
2	T
3	F
4	F
5	F
6	F
7	F
8	F
9	F

Flag that 2 has  
been visited



Neighbors →

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

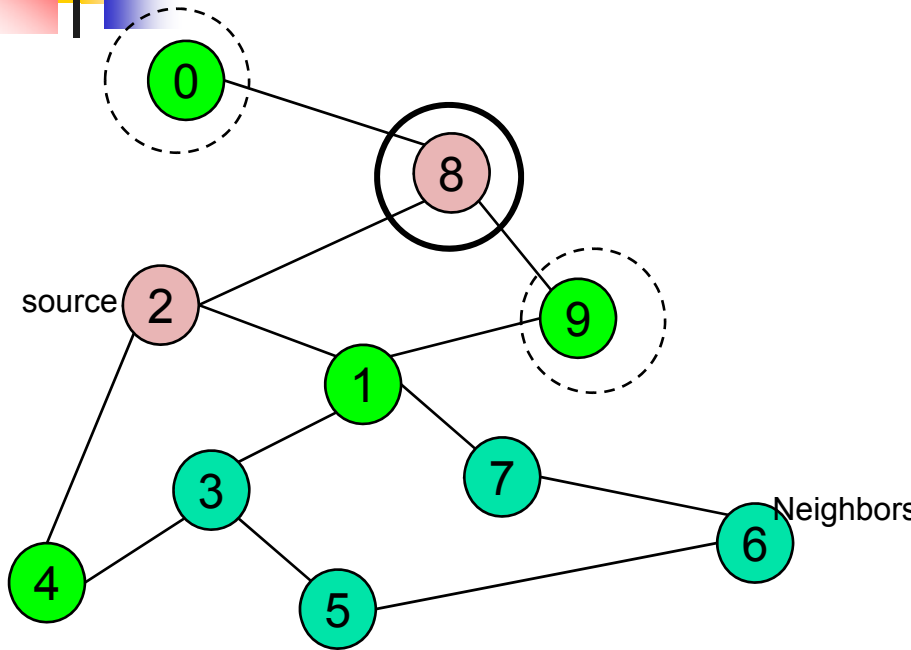
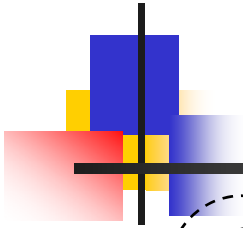
0	F
1	T
2	T
3	F
4	T
5	F
6	F
7	F
8	T
9	F

Mark neighbors  
as visited 1, 4, 8

$Q = \{2\} \rightarrow \{8, 1, 4\}$

Dequeue 2.

Place all **unvisited** neighbors of 2 on the queue



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T
1	T
2	T
3	F
4	T
5	F
6	F
7	F
8	T
9	T

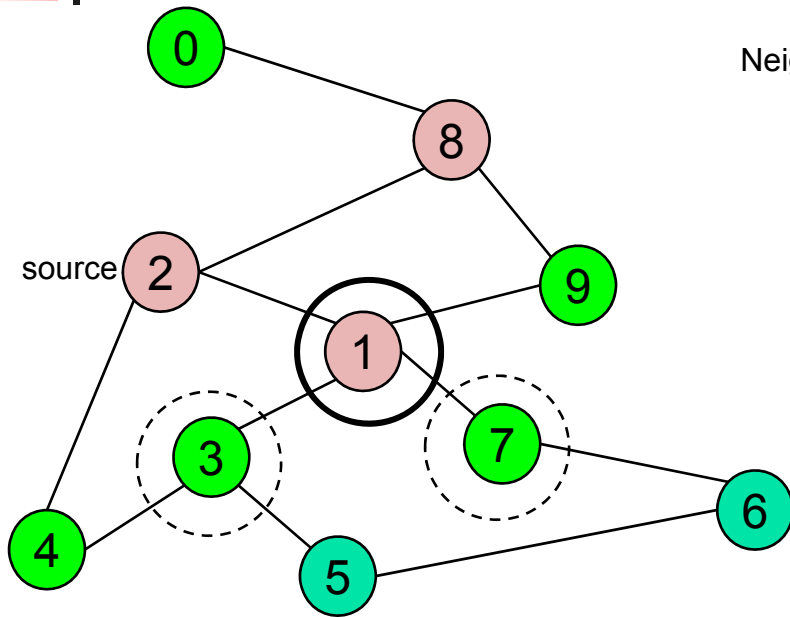
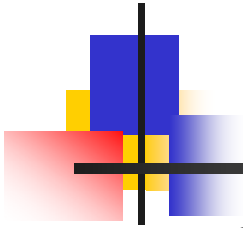
Mark new visited  
Neighbors 0, 9

$Q = \{ 8, 1, 4 \} \rightarrow \{ 1, 4, 0, 9 \}$

Dequeue 8.

- Place all unvisited neighbors of 8 on the queue.
- Notice that 2 is not placed on the queue again, it has been visited!





Adjacency List

Neighbors →

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

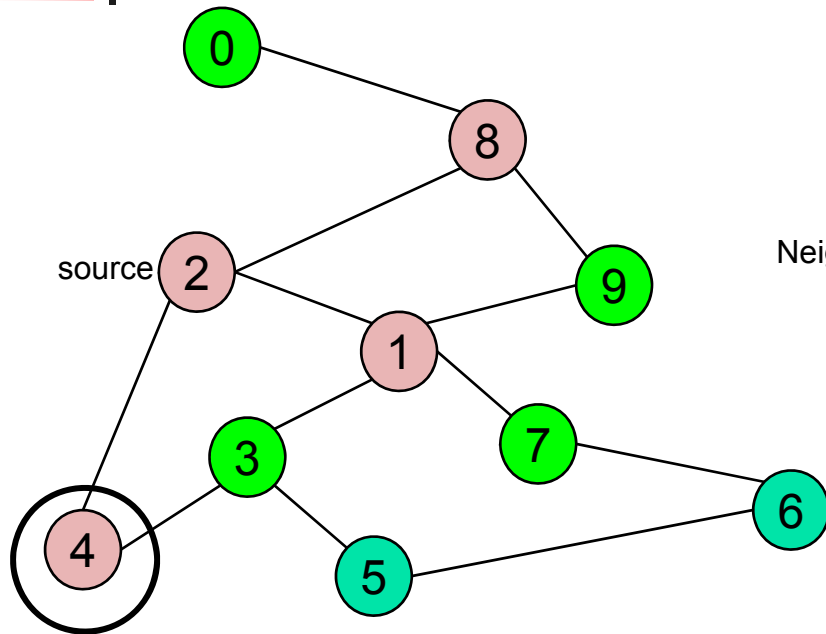
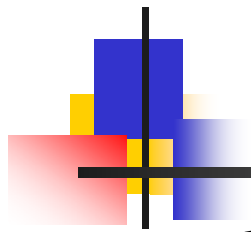
0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

Mark new visited  
Neighbors 3, 7

$Q = \{ 1, 4, 0, 9 \} \rightarrow \{ 4, 0, 9, 3, 7 \}$

Dequeue 1.

- Place all unvisited neighbors of 1 on the queue.
- Only nodes 3 and 7 haven't been visited yet.



Neighbors →

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

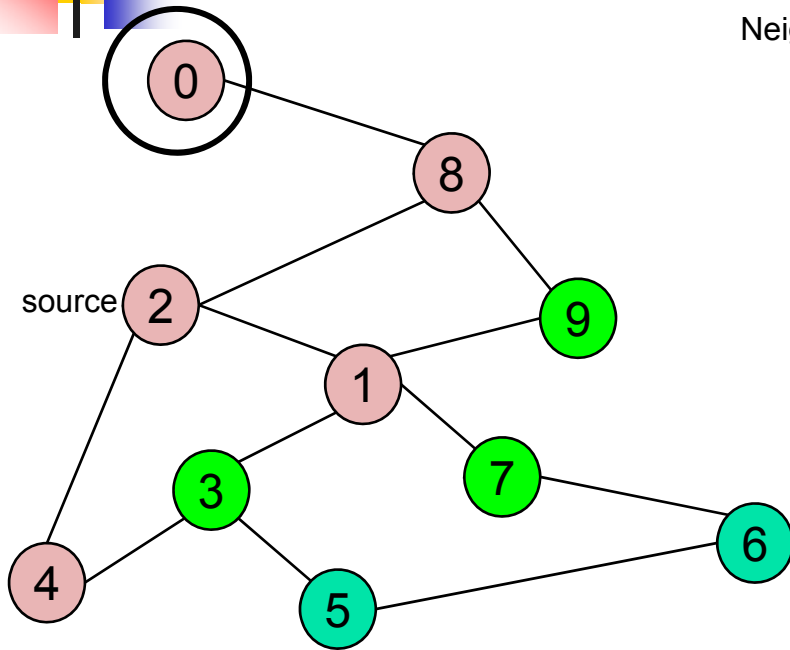
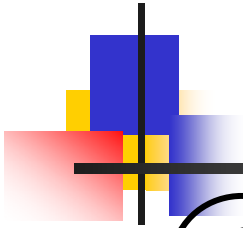
Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

$Q = \{4, 0, 9, 3, 7\} \rightarrow \{0, 9, 3, 7\}$

Dequeue 4.

-- 4 has no unvisited neighbors!



Adjacency List

Neighbors →

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

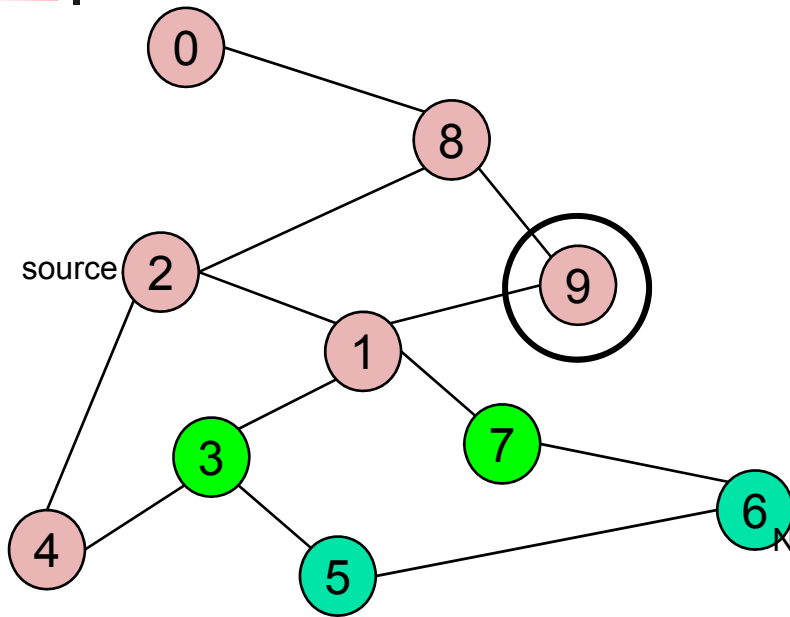
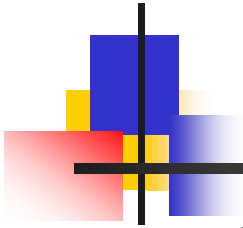
Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

$Q = \{0, 9, 3, 7\} \rightarrow \{9, 3, 7\}$

Dequeue 0.

-- 0 has no unvisited neighbors!



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

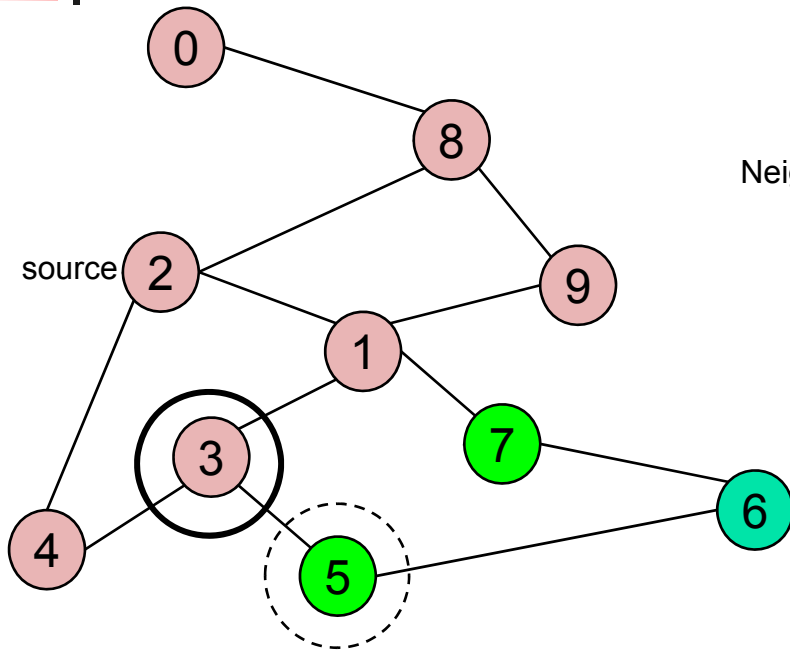
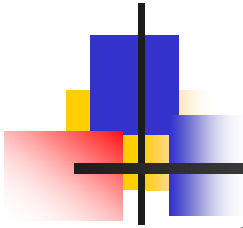
Visited Table (T/F)

0	T
1	T
2	T
3	T
4	T
5	F
6	F
7	T
8	T
9	T

$Q = \{9, 3, 7\} \rightarrow \{3, 7\}$

Dequeue 9.

-- 9 has no unvisited neighbors!



Neighbors →

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

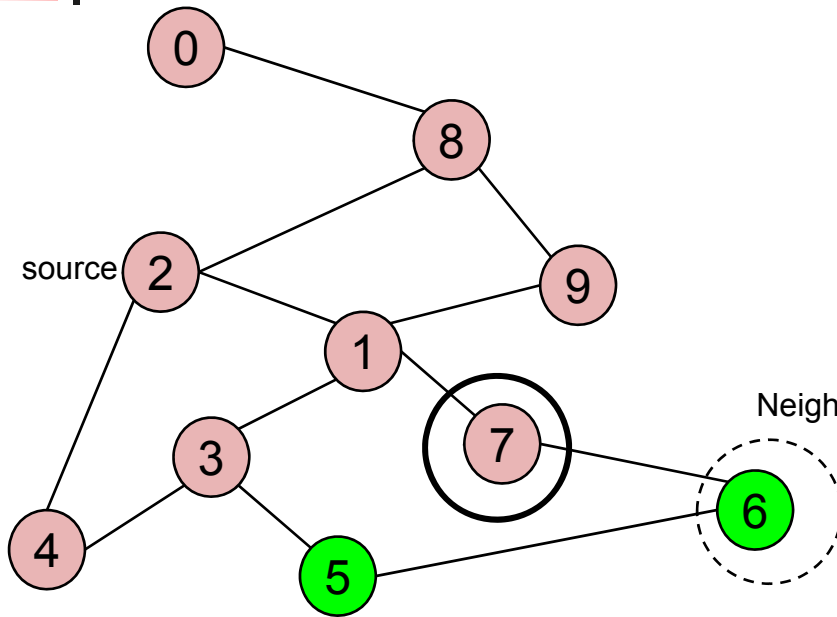
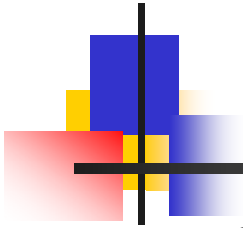
0	T
1	T
2	T
3	T
4	T
5	T
6	F
7	T
8	T
9	T

Mark new visited  
Vertex 5

$Q = \{3, 7\} \rightarrow \{7, 5\}$

Dequeue 3.

-- place neighbor 5 on the queue.



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

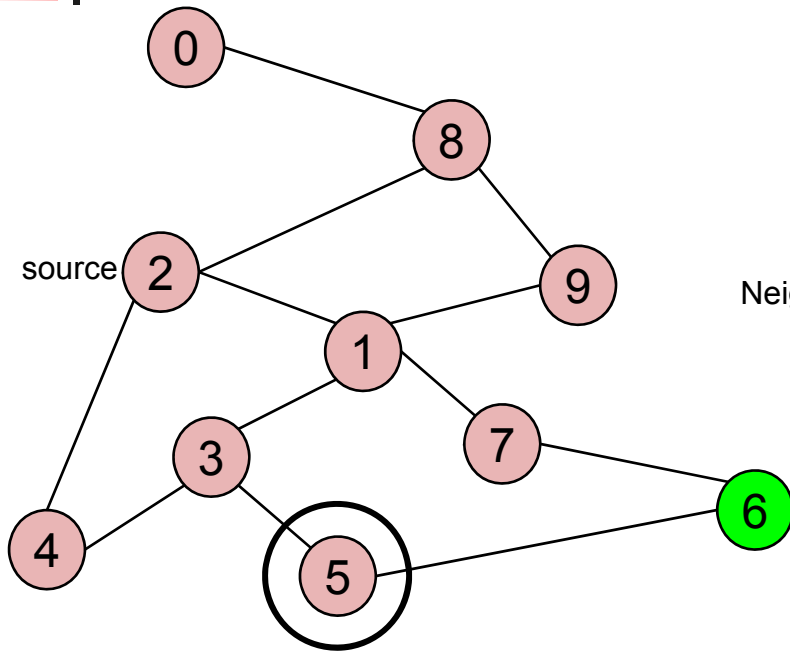
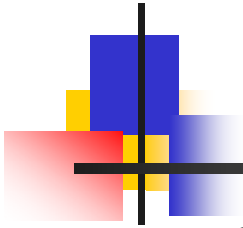
Visited Table (T/F)

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

Mark new visited  
Vertex 6

$Q = \{7, 5\} \rightarrow \{5, 6\}$

Dequeue 7.  
-- place neighbor 6 on the queue



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

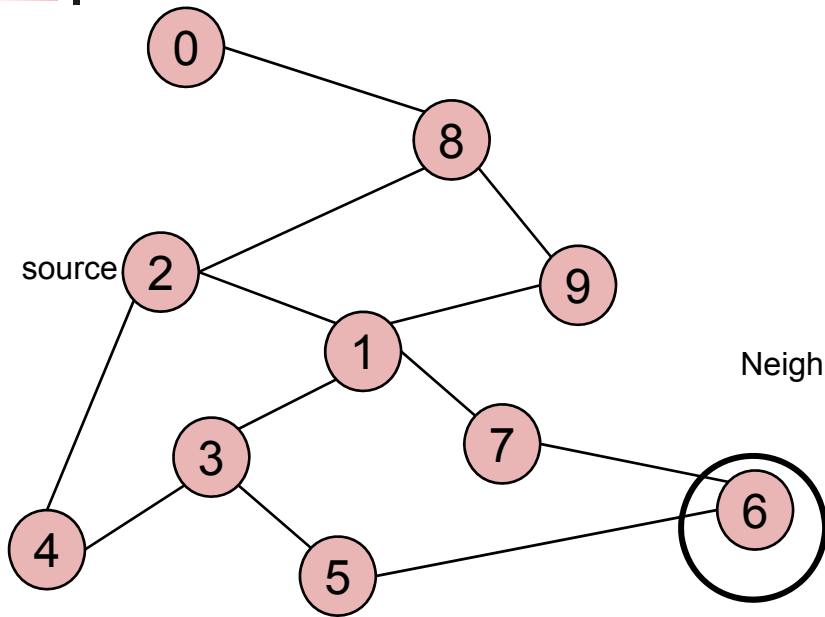
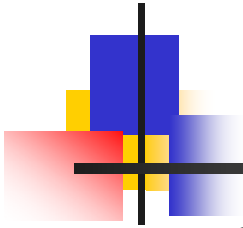
Visited Table (T/F)

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

$Q = \{ 5, 6 \} \rightarrow \{ 6 \}$

Dequeue 5.

-- no unvisited neighbors of 5



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

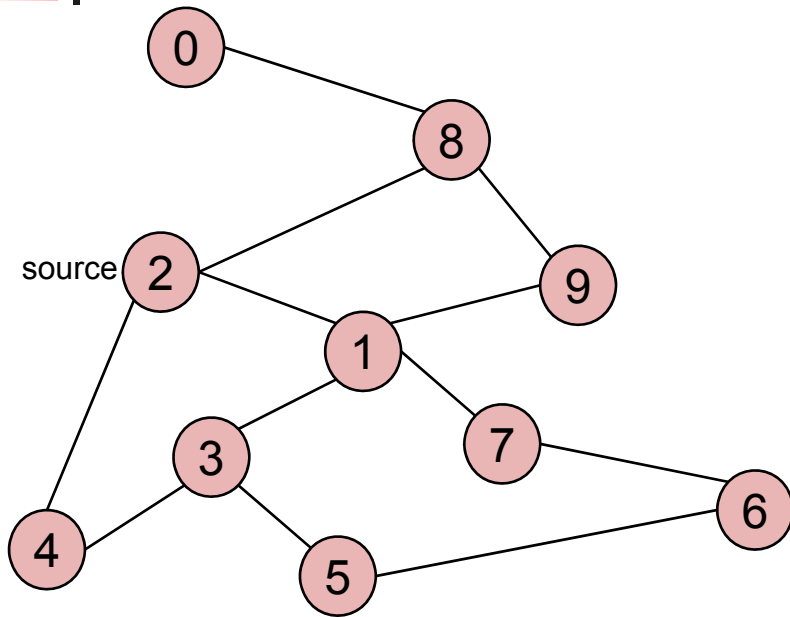
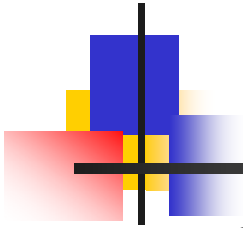
Visited Table (T/F)

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

$Q = \{6\} \rightarrow \{ \}$

Dequeue 6.  
-- no unvisited neighbors of 6





$Q = \{ \}$  **STOP!!!** **Q is empty!!!**

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

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

What did we discover?

Look at “visited” tables.

There exists a path from source vertex 2 to all vertices in the graph

# Time Complexity of BFS

(Using Adjacency List)

Assume adjacency list

- $n$  = number of vertices     $m$  = number of edges

**Algorithm**  $BFS(s)$

**Input:**  $s$  is the source vertex

**Output:** Mark all vertices that can be visited from  $s$ .

```
1.  for each vertex  $v$ 
2.      do  $flag[v] := false$ ;
3.   $Q =$  empty queue;
4.   $flag[s] := true$ ;
5.   $enqueue(Q, s)$ ;
6.  while  $Q$  is not empty
7.      do  $v := dequeue(Q)$ ;
8.      for each  $w$  adjacent to  $v$ 
9.          do if  $flag[w] = false$ 
10.             then  $flag[w] := true$ ;
11.                  $enqueue(Q, w)$ 
```

**$O(n + m)$**

Each vertex will enter  $Q$  at most once.

Each iteration takes time proportional to  $\deg(v) + 1$  (the number 1 is to account for the case where  $\deg(v) = 0$  --- the work required is 1, not 0).



# Running Time

---

- Recall: Given a graph with  $m$  edges, what is the total degree?

$$\sum_{\text{vertex } v} \deg(v) = 2m$$

- The **total** running time of the while loop is:

$$O\left(\sum_{\text{vertex } v} (\deg(v) + 1)\right) = O(n+m)$$

this is summing over all the iterations in the while loop!

# Time Complexity of BFS

(Using Adjacency Matrix)

Assume adjacency list

- $n$  = number of vertices     $m$  = number of edges

**Algorithm**  $BFS(s)$

**Input:**  $s$  is the source vertex

**Output:** Mark all vertices that can be visited from  $s$ .

```
1.  for each vertex  $v$ 
2.      do  $flag[v] := false$ ;
3.   $Q =$  empty queue;
4.   $flag[s] := true$ ;
5.   $enqueue(Q, s)$ ;
6.  while  $Q$  is not empty
7.      do  $v := dequeue(Q)$ ;
8.          for each  $w$  adjacent to  $v$ 
9.              do if  $flag[w] = false$ 
10.                  then  $flag[w] := true$ ;
11.                       $enqueue(Q, w)$ 
```

$O(n^2)$

Finding the adjacent vertices of  $v$  requires checking all elements in the row. This takes linear time  $O(n)$ .

Summing over all the  $n$  iterations, the total running time is  $O(n^2)$ .

So, with adjacency matrix, BFS is  $O(n^2)$  independent of the number of edges  $m$ . With adjacent lists, BFS is  $O(n+m)$ ; if  $m=O(n^2)$  like in a dense graph,  $O(n+m)=O(n^2)$ .



# Depth-First Search (DFS)

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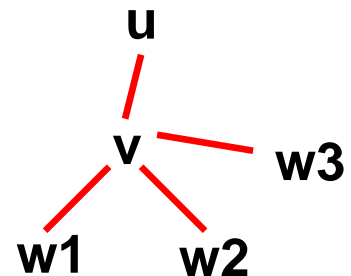
- DFS is another popular graph search strategy
  - Idea is similar to pre-order traversal (visit node, then visit children recursively)
- DFS can provide certain information about the graph that BFS cannot



# DFS Algorithm

---

- DFS will continue to visit **neighbors** in a recursive pattern
  - Whenever we visit  $v$  from  $u$ , we recursively visit all unvisited neighbors of  $v$ . Then we backtrack (return) to  $u$ .
  - Note: it is possible that  $w2$  was unvisited when we recursively visit  $w1$ , but became visited by the time we return from the recursive call.





# Stack implementation of Generic Search: DFS

## Algorithm DFS

```
{ for i=1 to n do {Mark[i]=0; P[i]=0; Num[i]=0;} count=1;
For i=1 to n do { if (Mark[i]==0) search(i); DFS(i);} S=createStack();
Search(i) DFS(i){
    Mark[i]=1; P[i]=-1; Num[i]=count; Count++;
    S={i}; Push(S,i);
    While ( S !=emptyset lseempty(S)!=0){
        select a vertex x from S;
        if ( x has an unmarked neighbor y)
            { Mark[y]=1; P[y]=x; Num[y]=count; count=count+1;
              S=S ∪ {y}; Push(S,y);
            }
        else S=S-{x}; Pop(S); }
    }
```

} Note that green colored text is replaced with red colored text in generic search



# DFS Algorithm

---

## Algorithm $DFS(s)$

1. **for** each vertex  $v$
2.     **do**  $flag[v] := \text{false};$
3.      $RDFS(s);$

Flag all vertices as not visited

## Algorithm $RDFS(v)$

1.  $flag[v] := \text{true};$
2.     **for** each neighbor  $w$  of  $v$
3.         **do if**  $flag[w] = \text{false}$
4.             **then**  $RDFS(w);$

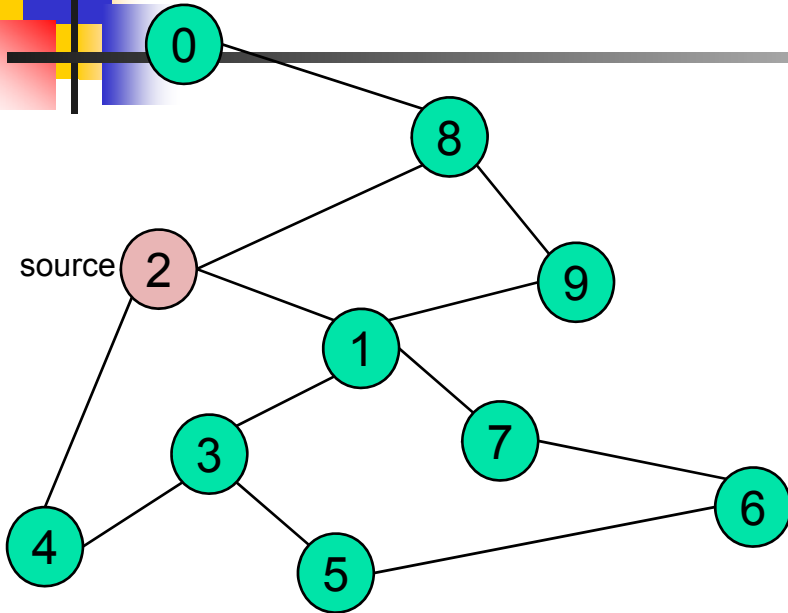
Flag yourself as visited

For unvisited neighbors,  
call  $RDFS(w)$  recursively

**We can also record the paths using  $pred[ ]$ .**



# Example



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

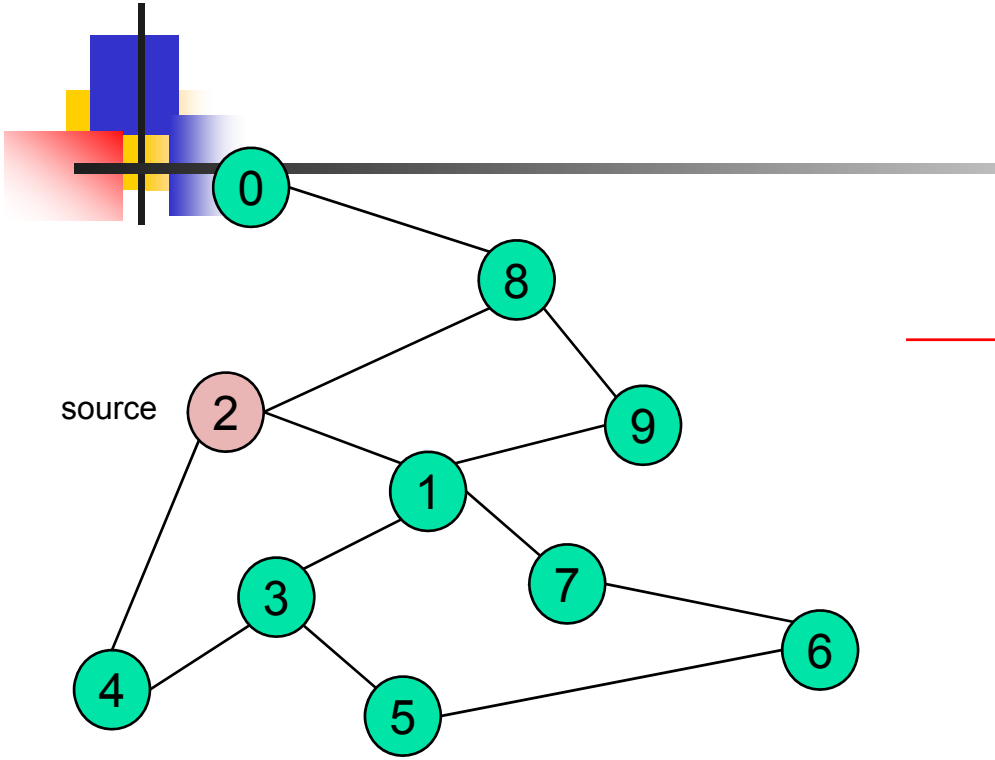
0	F
1	F
2	F
3	F
4	F
5	F
6	F
7	F
8	F
9	F

*Pred*

-
-
-
-
-
-
-
-
-
-

Initialize visited  
table (all False)

Initialize Pred to -1



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

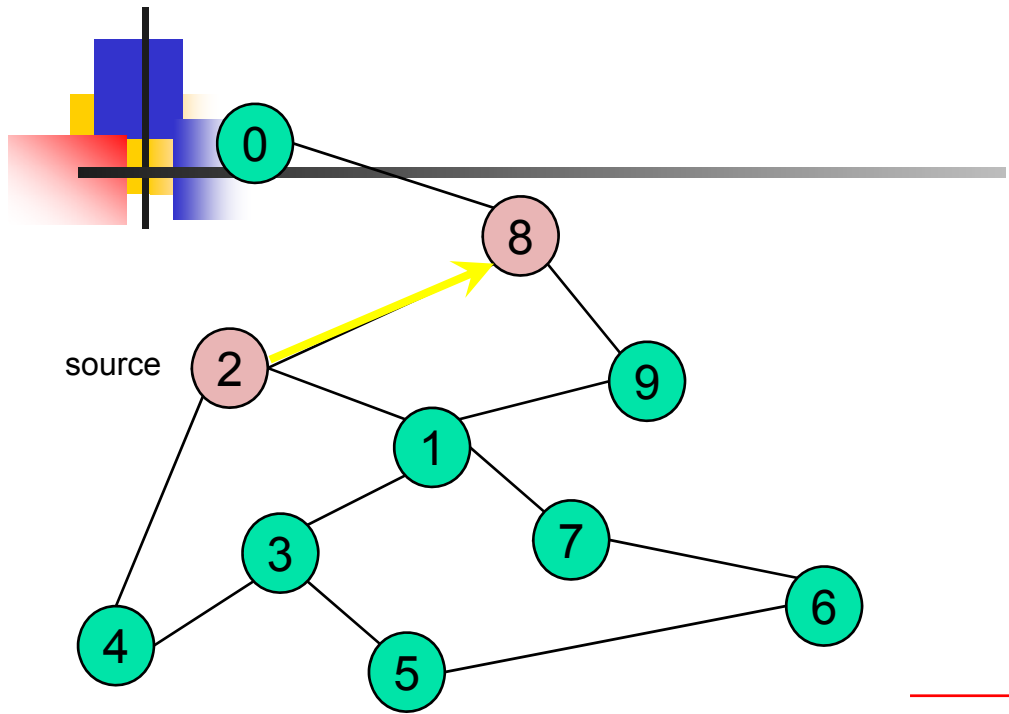
0	F	-
1	F	-
2	T	-
3	F	-
4	F	-
5	F	-
6	F	-
7	F	-
8	F	-
9	F	-

*Pred*

Mark 2 as visited

RDFS( 2 )

Now visit RDFS(8)



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	F	-
1	F	-
2	T	-
3	F	-
4	F	-
5	F	-
6	F	-
7	F	-
8	T	2
9	F	-

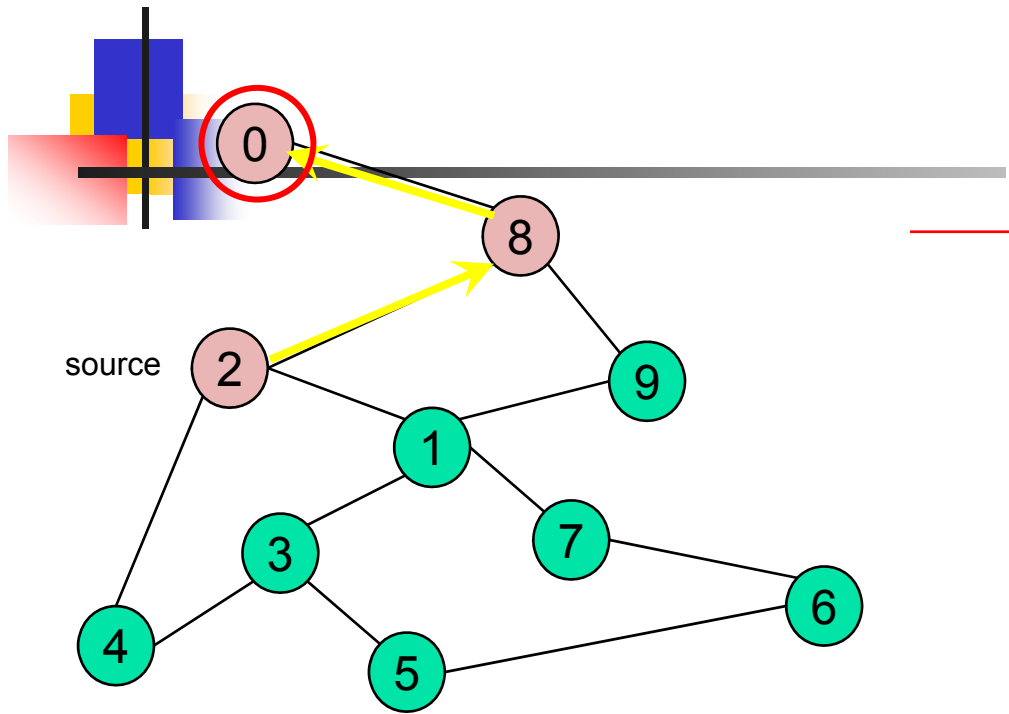
*Pred*

Mark 8 as visited

mark Pred[8]

Recursive calls

RDFS( 2 )  
 RDFS(8)  
 2 is already visited, so visit RDFS(0)



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	F	-
2	T	-
3	F	-
4	F	-
5	F	-
6	F	-
7	F	-
8	T	2
9	F	-

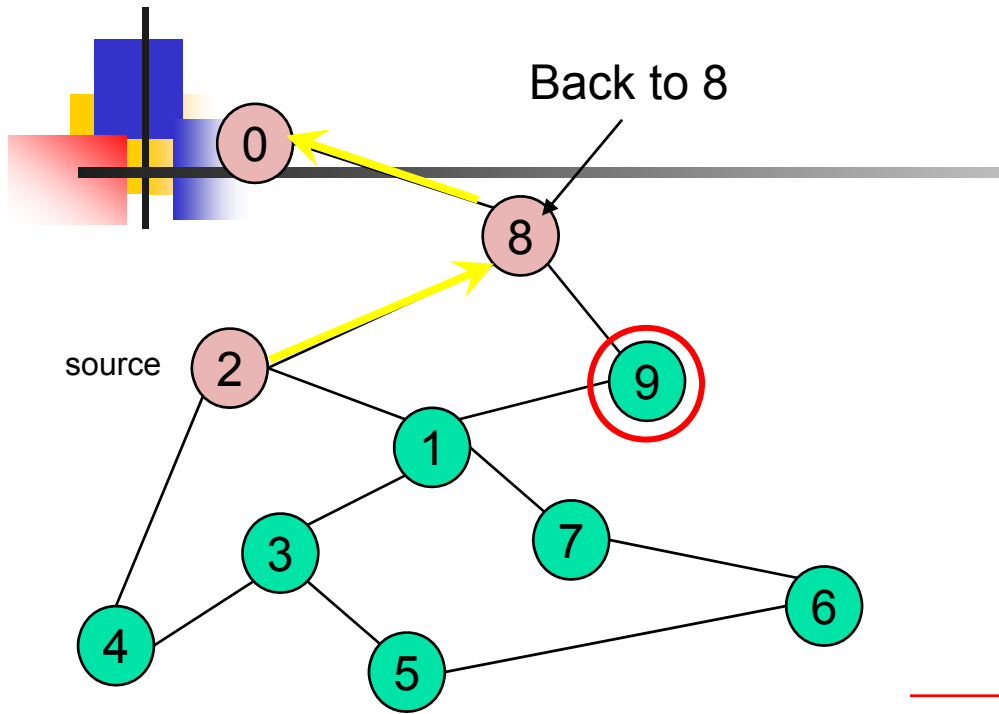
*Pred*

Mark 0 as visited

Mark Pred[0]

Recursive calls

RDFS( 2 )  
   RDFS(8)  
     RDFS(0) -> no unvisited neighbors, return  
       to call RDFS(8)



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	F	-
2	T	-
3	F	-
4	F	-
5	F	-
6	F	-
7	F	-
8	T	2
9	F	-

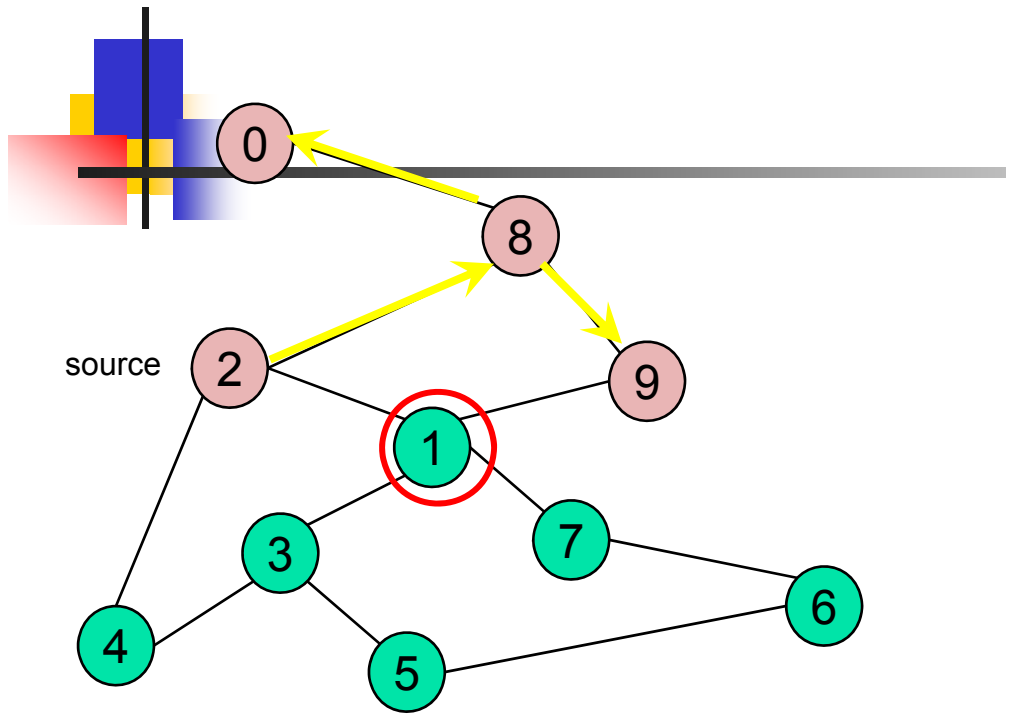
*Pred*

Recursive  
calls

RDFS( 2 )

RDFS(8)

Now visit 9 -> RDFS(9)



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	F	-
2	T	-
3	F	-
4	F	-
5	F	-
6	F	-
7	F	-
8	T	2
9	T	8

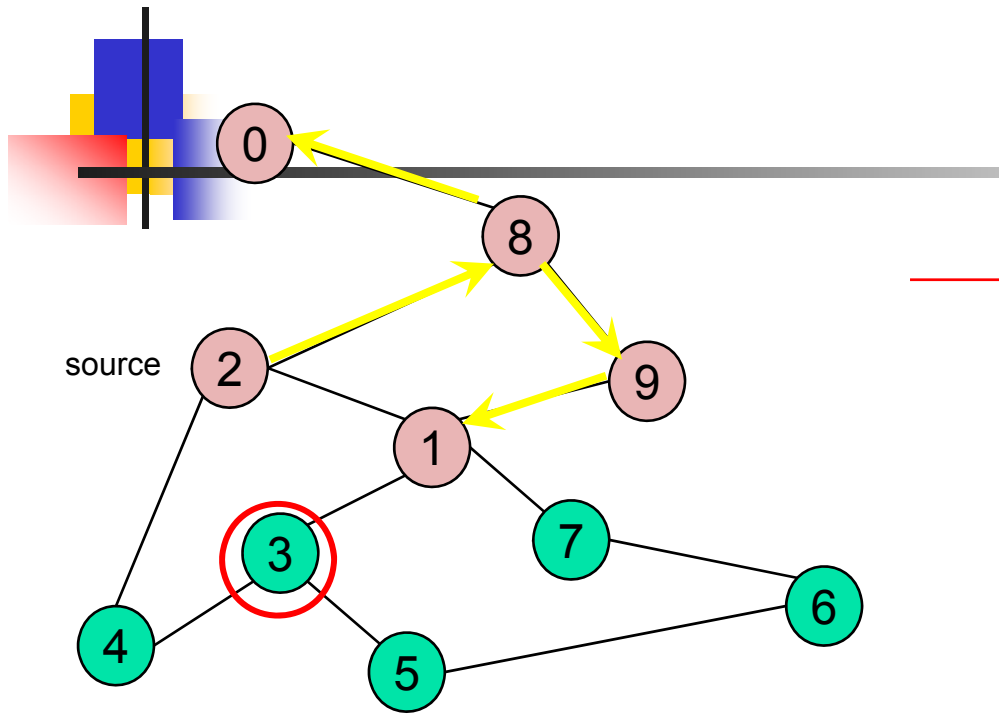
*Pred*

Mark 9 as visited

Mark Pred[9]

Recursive calls

RDFS( 2 )  
     RDFS(8)  
         RDFS(9)  
             -> visit 1, RDFS(1)



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

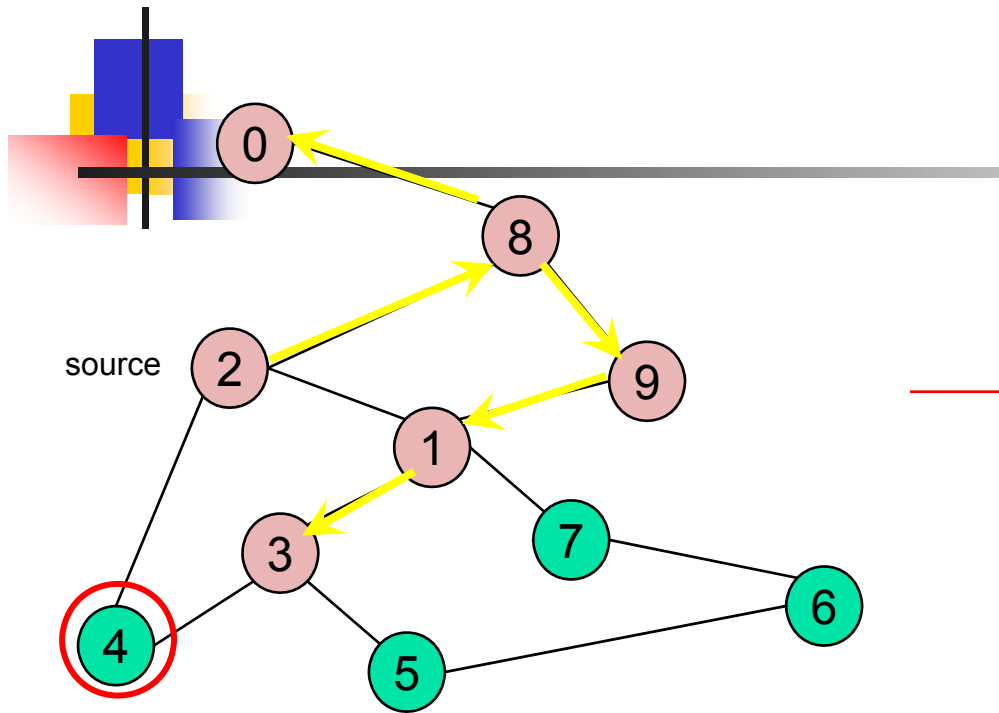
0	T	8
1	T	9
2	T	-
3	F	-
4	F	-
5	F	-
6	F	-
7	F	-
8	T	2
9	T	8

*Pred*

Recursive calls

RDFS( 2 )  
   RDFS(8)  
     RDFS(9)  
       RDFS(1)  
         visit RDFS(3)

Mark 1 as visited  
 Mark Pred[1]



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	F	-
5	F	-
6	F	-
7	F	-
8	T	2
9	T	8

*Pred*

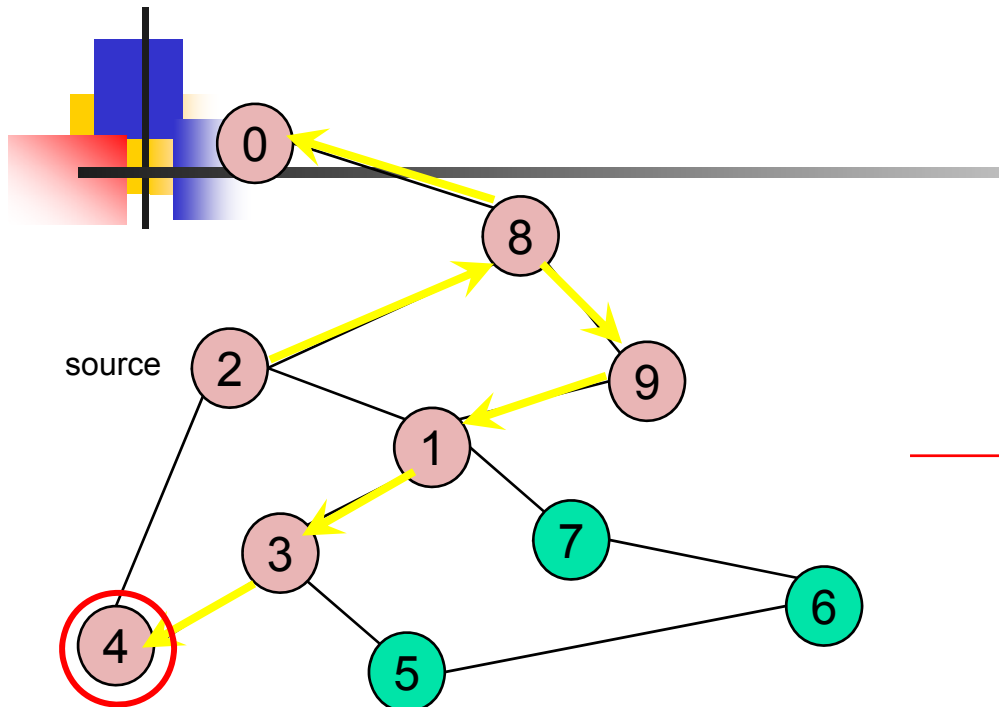
Mark 3 as visited

Mark Pred[3]

Recursive  
calls

RDFS( 2 )  
   RDFS(8)  
     RDFS(9)  
       RDFS(1)  
         RDFS(3)  
           visit RDFS(4)





Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	F	-
6	F	-
7	F	-
8	T	2
9	T	8

*Pred*

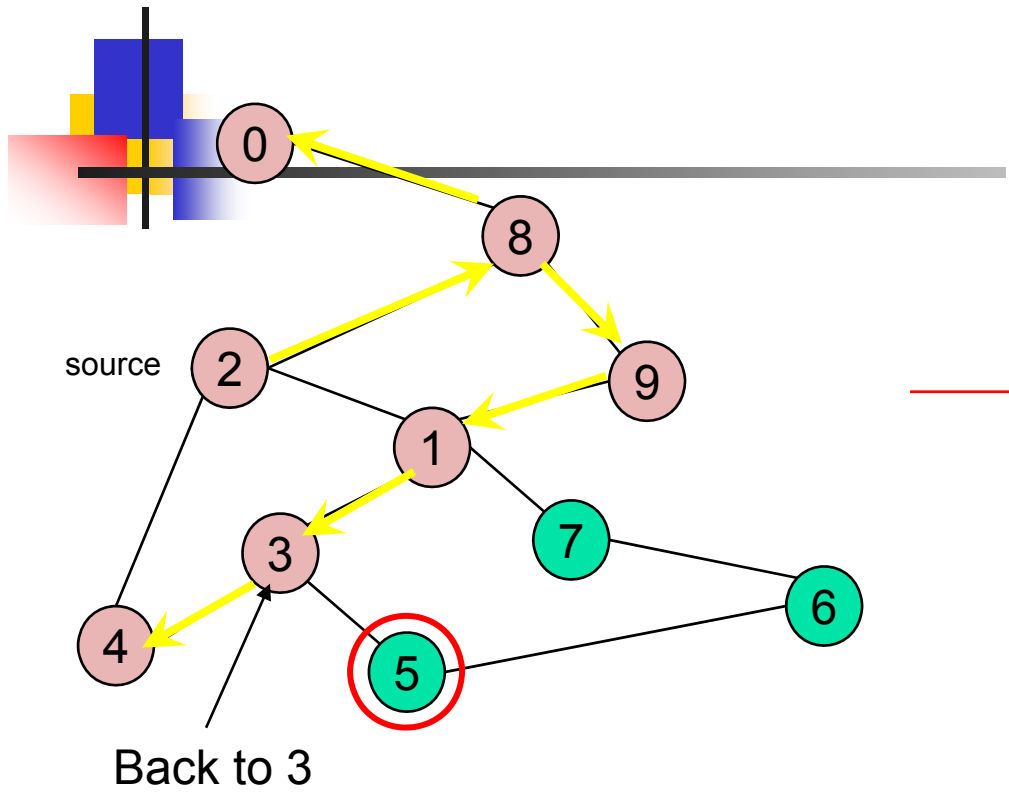
Mark 4 as visited

Mark Pred[4]

Recursive calls

RDFS( 2 )  
 RDFS(8)  
 RDFS(9)  
 RDFS(1)  
 RDFS(3)

RDFS(4) → STOP all of 4's neighbors have been visited  
 return back to call RDFS(3)



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

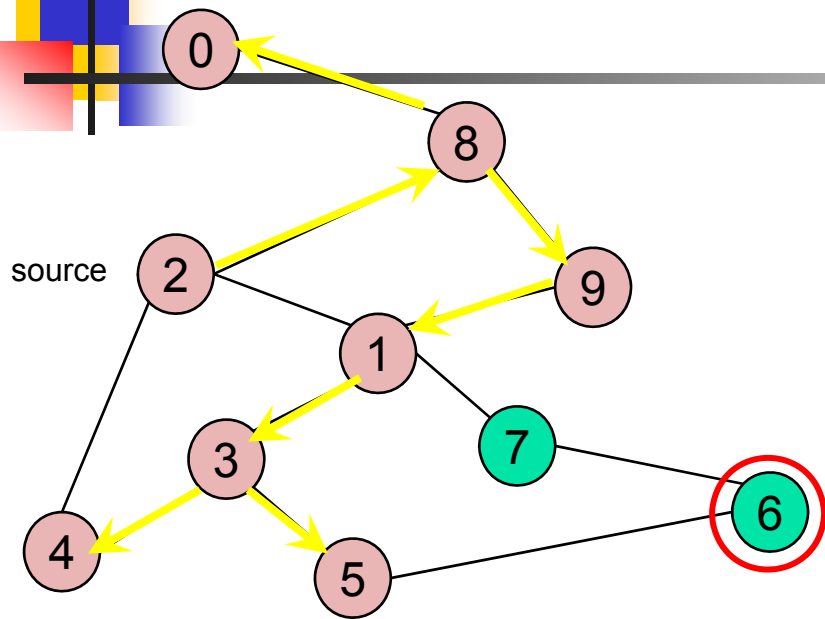
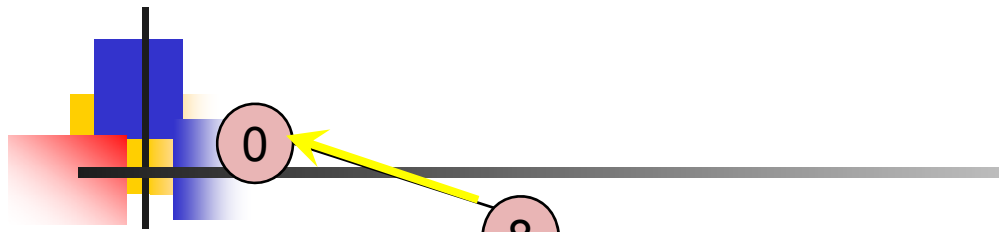
Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	F	-
6	F	-
7	F	-
8	T	2
9	T	8

*Pred*

RDFS( 2 )  
   RDFS(8)  
     RDFS(9)  
       RDFS(1)  
         RDFS(3)  
           visit 5 -> RDFS(5)

Recursive calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	F	-
7	F	-
8	T	2
9	T	8

*Pred*

RDFS( 2 )

RDFS(8)

RDFS(9)

RDFS(1)

RDFS(3)

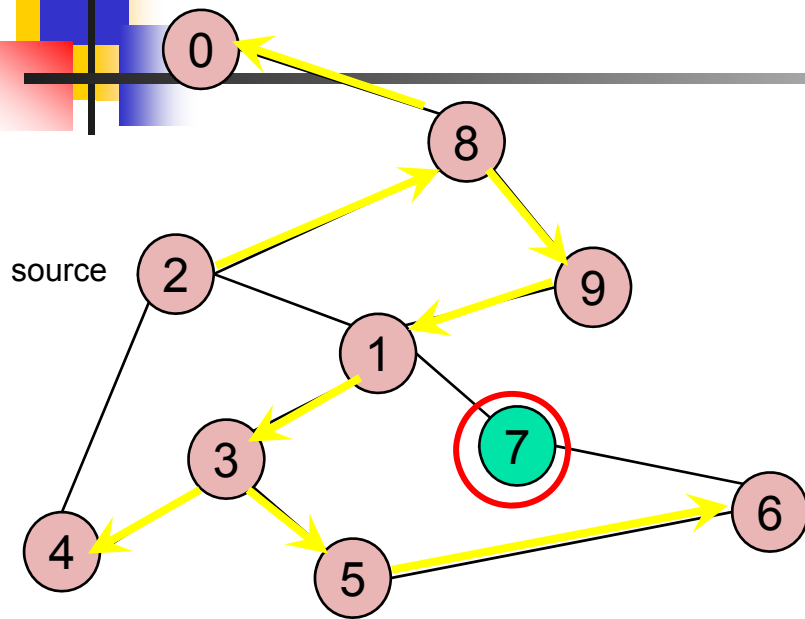
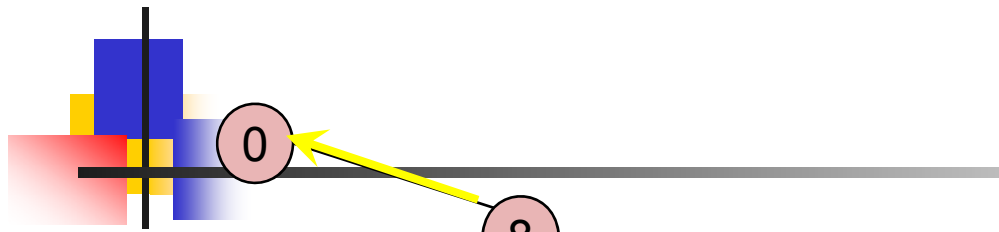
RDFS(5)

3 is already visited, so visit 6 -> RDFS(6)

Recursive  
calls

Mark 5 as visited

Mark Pred[5]



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	F	-
8	T	2
9	T	8

*Pred*

RDFS( 2 )

RDFS(8)

RDFS(9)

RDFS(1)

RDFS(3)

RDFS(5)

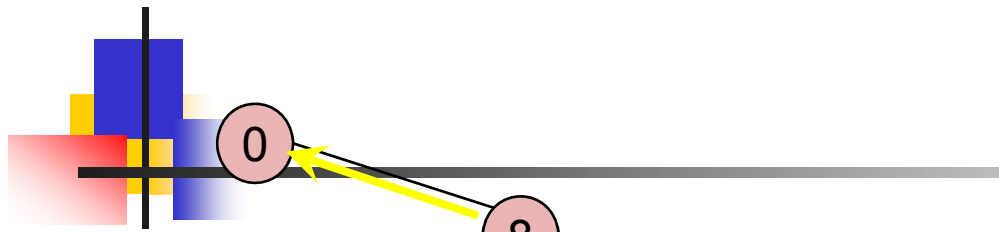
RDFS(6)

visit 7 -> RDFS(7)

Mark 6 as visited

Mark Pred[6]

Recursive  
calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

RDFS( 2 )

RDFS(8)

RDFS(9)

RDFS(1)

RDFS(3)

RDFS(5)

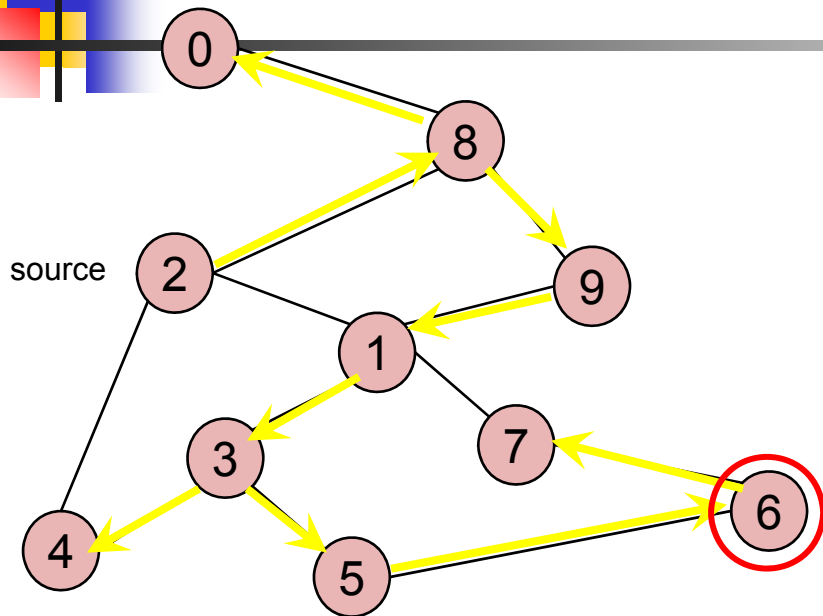
RDFS(6)

RDFS(7) -> Stop no more unvisited neighbors

Mark 7 as visited

Mark Pred[7]

Recursive  
calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

RDFS( 2 )

RDFS(8)

RDFS(9)

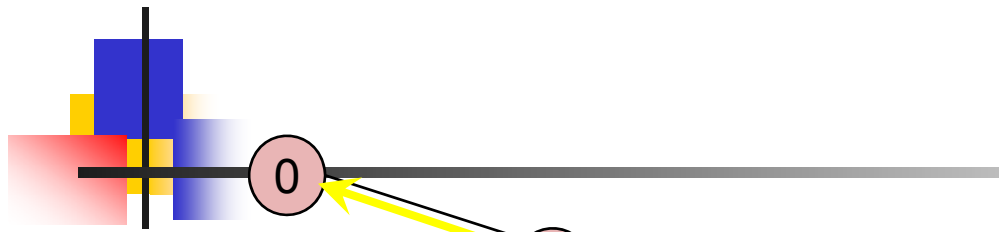
RDFS(1)

RDFS(3)

RDFS(5)

RDFS(6) -> Stop

Recursive  
calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

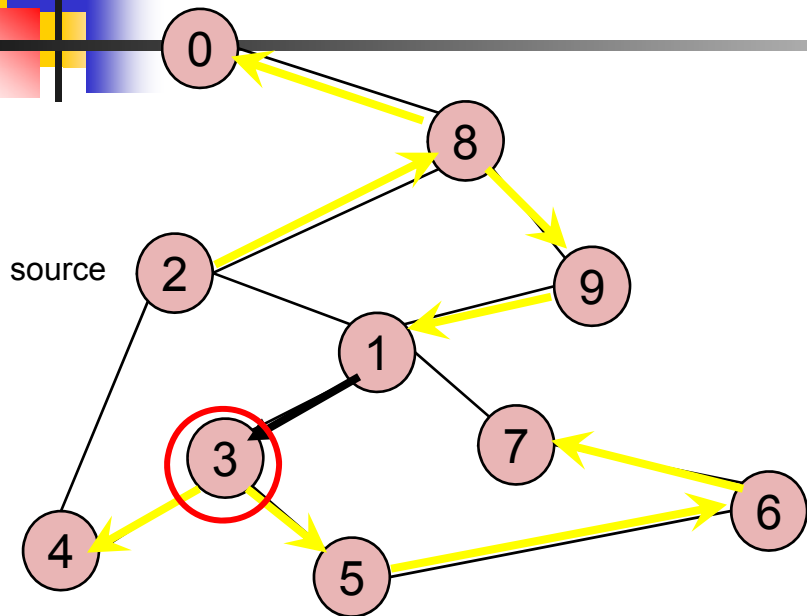
Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

RDFS( 2 )  
   RDFS(8)  
     RDFS(9)  
       RDFS(1)  
         RDFS(3)  
           RDFS(5) -> Stop

Recursive calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

RDFS( 2 )

RDFS(8)

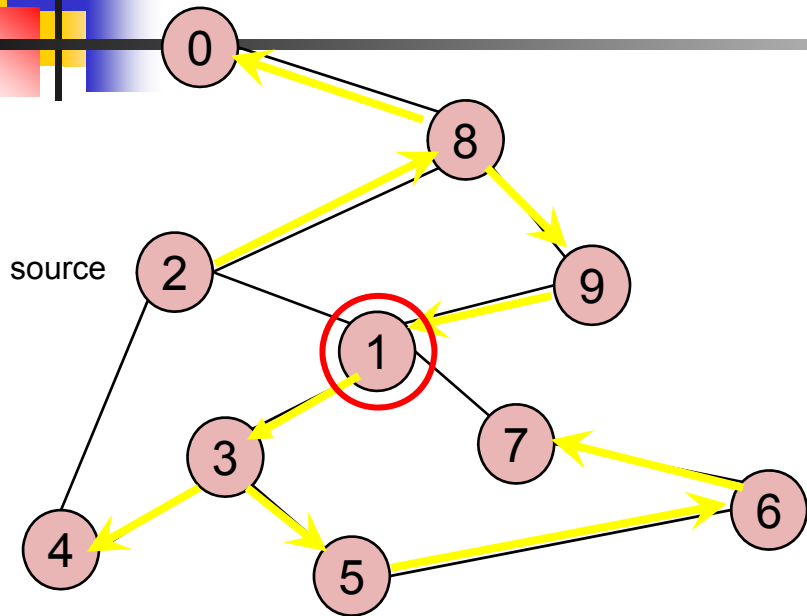
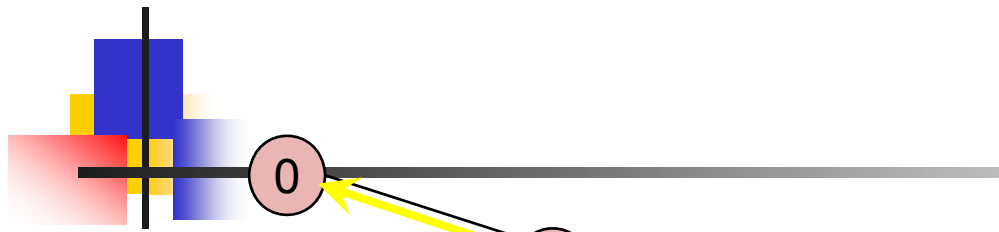
RDFS(9)

RDFS(1)

RDFS(3) -> Stop

Recursive  
calls





Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

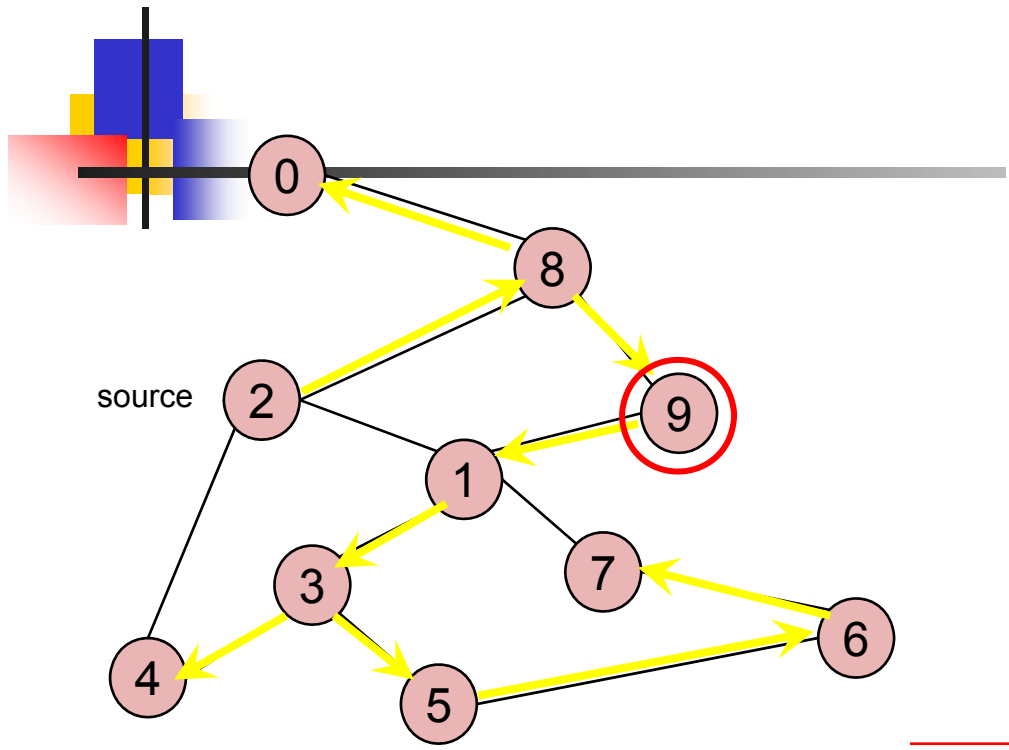
RDFS( 2 )

RDFS(8)

RDFS(9)

RDFS(1) -> Stop

Recursive  
calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

RDFS( 2 )  
 RDFS(8)  
 RDFS(9) -> Stop

Recursive calls



Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

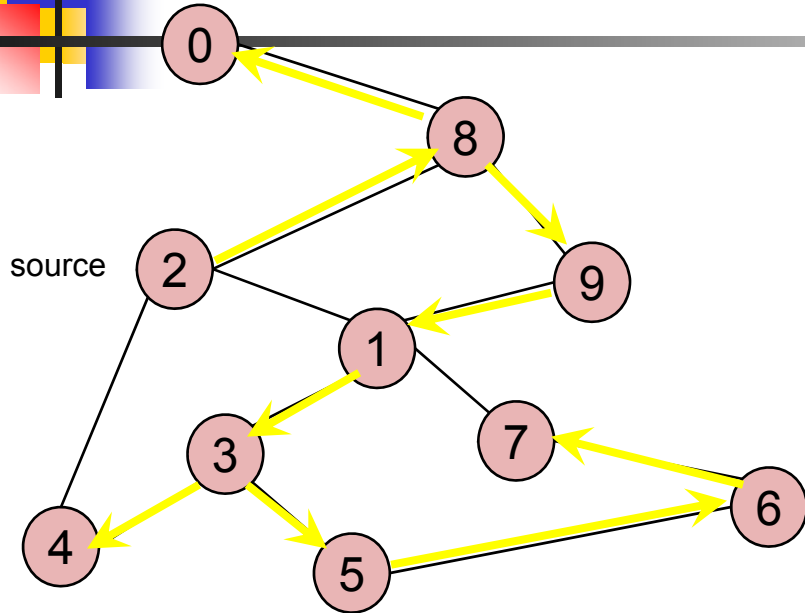
0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

RDFS( 2 )  
RDFS(8) -> Stop

Recursive  
calls

# Example Finished



RDFS( 2 ) -> Stop

Recursive calls finished

Adjacency List

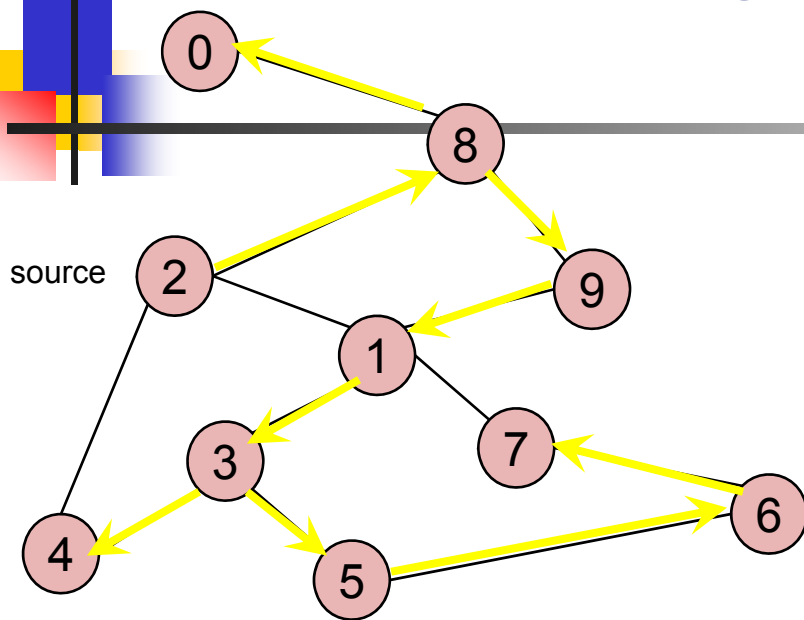
0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

0	T	8
1	T	9
2	T	-
3	T	1
4	T	3
5	T	3
6	T	5
7	T	6
8	T	2
9	T	8

*Pred*

# DFS Path Tracking



DFS find out path too

**Algorithm**  $Path(w)$

1. **if**  $pred[w] \neq -1$
2.     **then**
3.          $Path(pred[w]);$
4.     output  $w$

Adjacency List

0	8
1	3 7 9 2
2	8 1 4
3	4 5 1
4	2 3
5	3 6
6	7 5
7	1 6
8	2 0 9
9	1 8

Visited Table (T/F)

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

*Pred*

Try some examples.

Path(0) ->

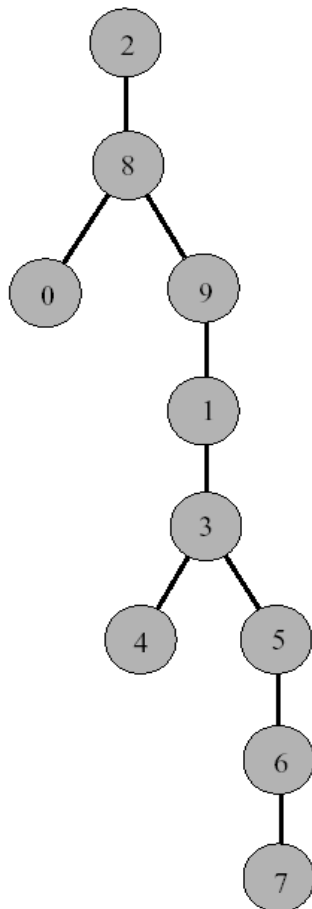
Path(6) ->

Path(7) ->

# DFS Tree

Resulting DFS-tree.

Notice it is much “deeper”  
than the BFS tree.



Captures the structure of the recursive calls

- when we visit a neighbor  $w$  of  $v$ , we add  $w$  as child of  $v$
- whenever DFS returns from a vertex  $v$ , we climb up in the tree from  $v$  to its parent



# Time Complexity of DFS

(Using adjacency list)

We never visited a vertex more than once

- We had to examine all edges of the vertices
  - We know  $\sum_{\text{vertex } v} \text{degree}(v) = 2m$  where  $m$  is the number of edges
- So, the running time of DFS is proportional to the number of edges and number of vertices (same as BFS)
  - $O(n + m)$
- You will also see this written as:
  - $O(|v| + |e|)$        $|v|$  = number of vertices ( $n$ )  
                                  $|e|$  = number of edges ( $m$ )