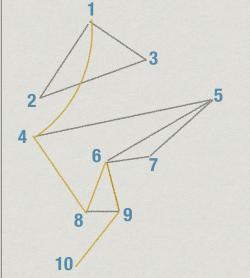
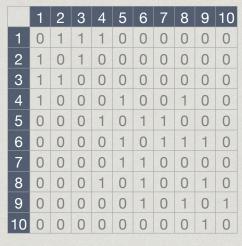
Graphs, formally

 $G=(V\!,\!E)$

- * Set of vertices V
- * Set of edges E
 - * E is a subset of pairs (v,v'): $E \subseteq V \times V$
 - * Undirected graph: (v,v') and (v',v) are the same edge
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 - * (v,v') is an edge from v to v'
 - * Does not guarantee that (v',v) is also an edge

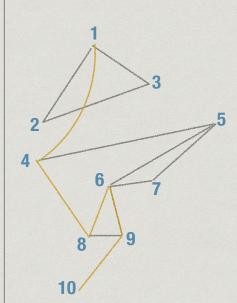
Adjacency matrix





Finding a route * Find a Vo sequence of vertices v₀, v₁, ..., v_k such that * vo is source V1 * Each (v_i, v_{i+1}) V₃ is an edge in F V4 V₂ * vk is target V5

Adjacency list



 For each vertex, maintain a list of its neighbours

1	2,3,4
2	1,3
3	1,2
4	1,5,8
5	4,6,7
6	5,7,8,9
7	5,6
8	4,6,9
9	6,8,10
10	9

Finding a path

- Mark vertices that have been visited
- Keep track of vertices whose neighbours have already been explored
 - Avoid going round indefinitely in circles
- Two fundamental strategies: breadth first and depth first

Breadth first search

- * Recall that V = {1,2,...,n}
- Array visited[i] records whether i has been visited
- When a vertex is visited for the first time, add it to a queue
 - Explore vertices in the order they reach the queue

Breadth first search

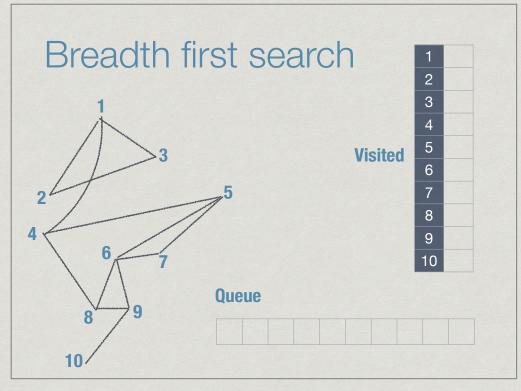
- * Explore the graph level by level
 - * First visit vertices one step away
 - * Then two steps away
 - * ...
- * Remember which vertices have been visited
- Also keep track of vertices visited, but whose neighbours are yet to be explored

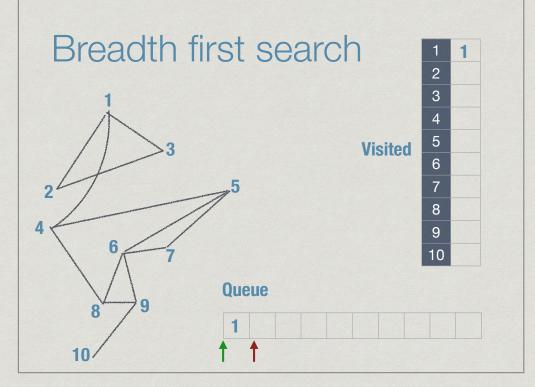
Breadth first search

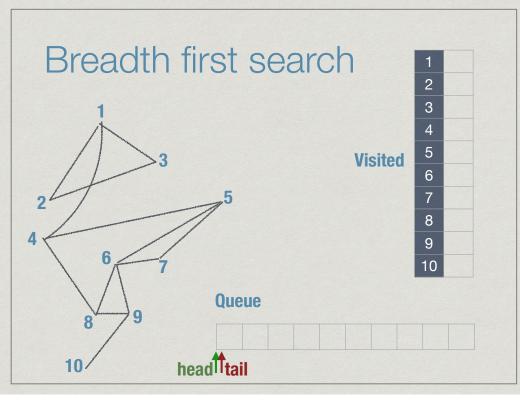
* Exploring a vertex i:

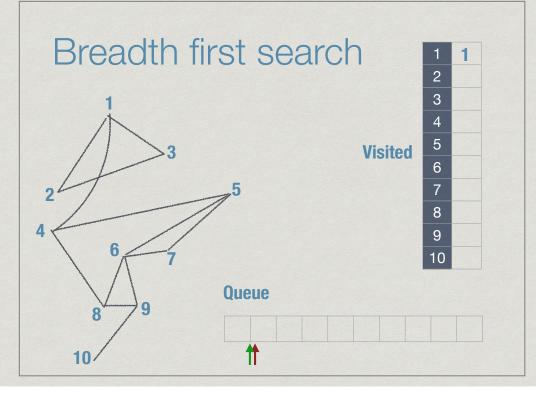
for each edge (i,j)
 if visited[j] == 0
 visited[j] = 1
 append j to queue

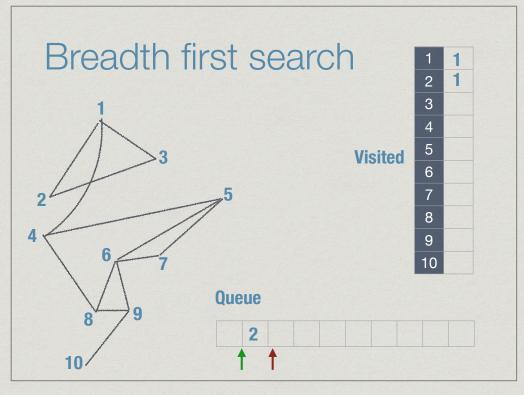
- Initially, queue contains only source vertex
- At each stage, explore vertex at the head of the queue
- * Stop when the queue becomes empty

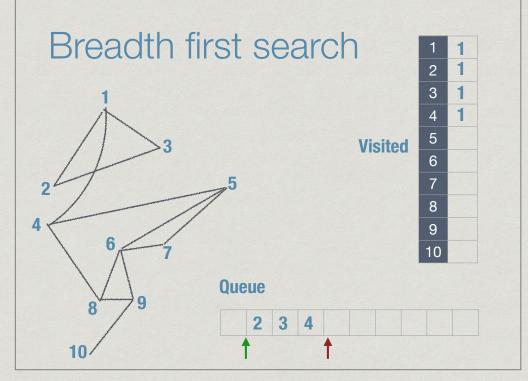


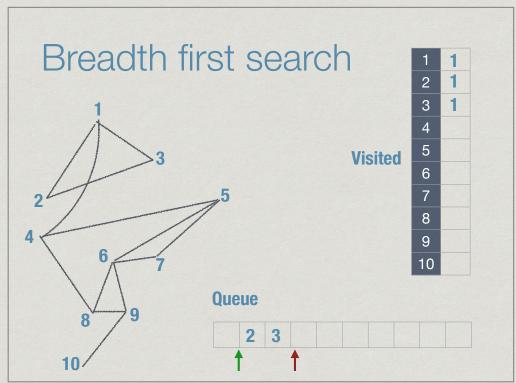


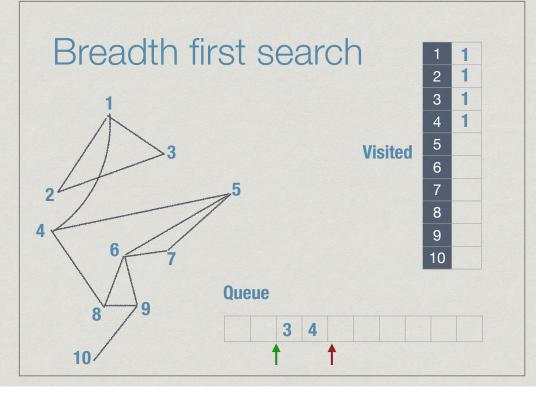


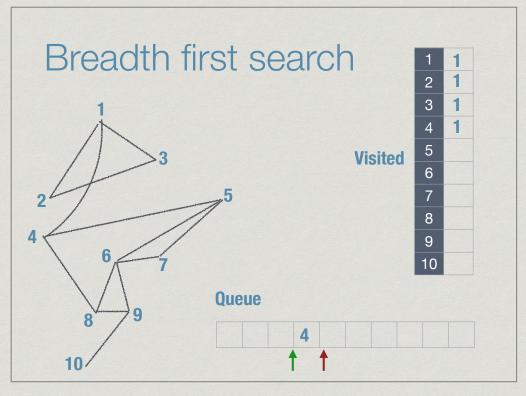


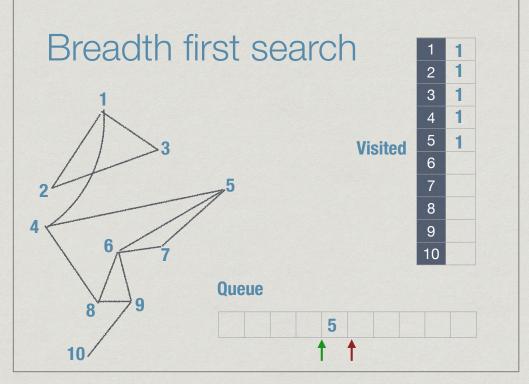


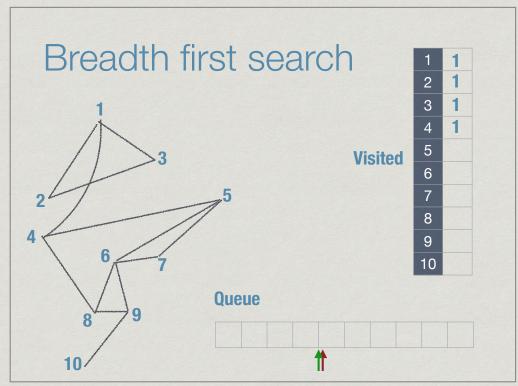


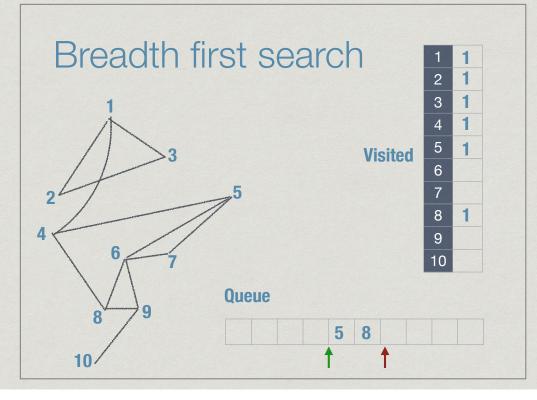


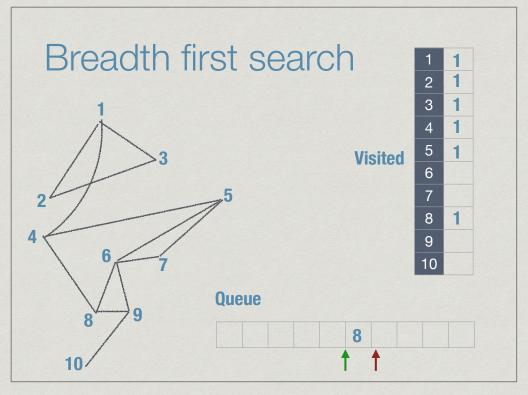


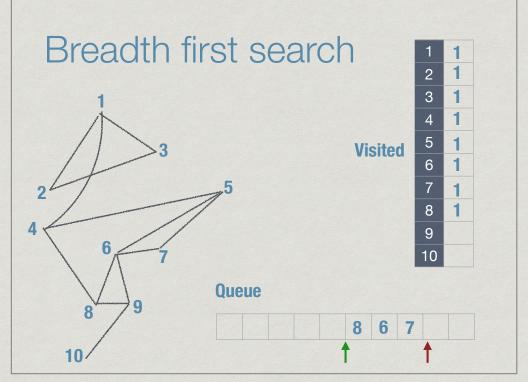


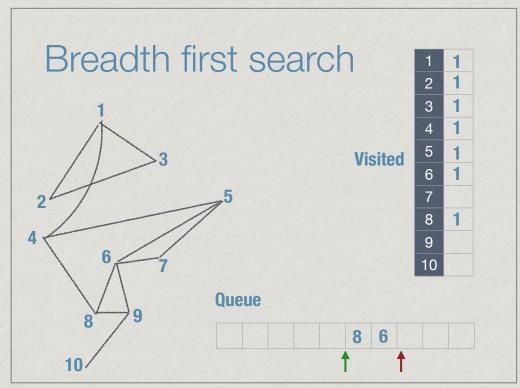


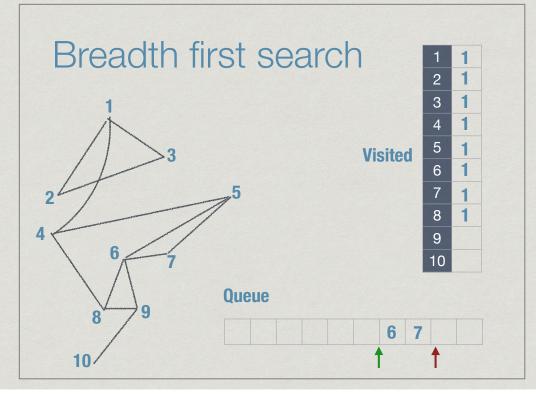


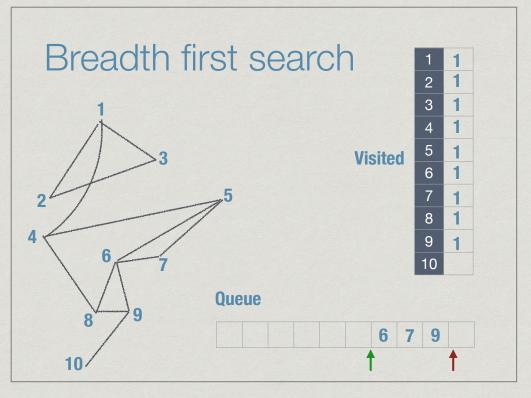


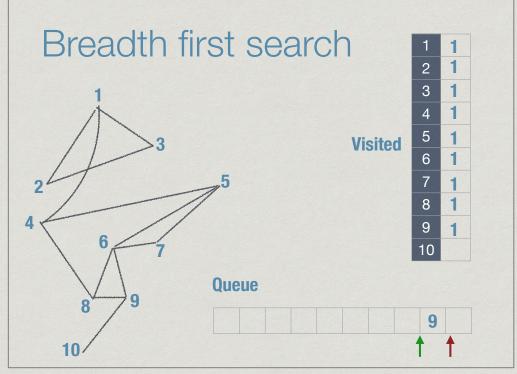


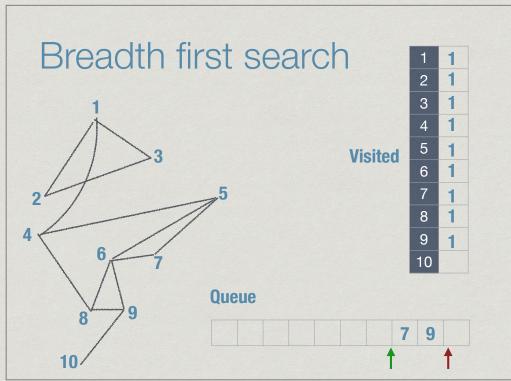


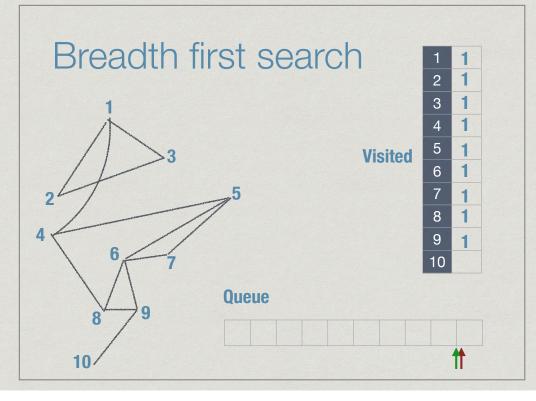


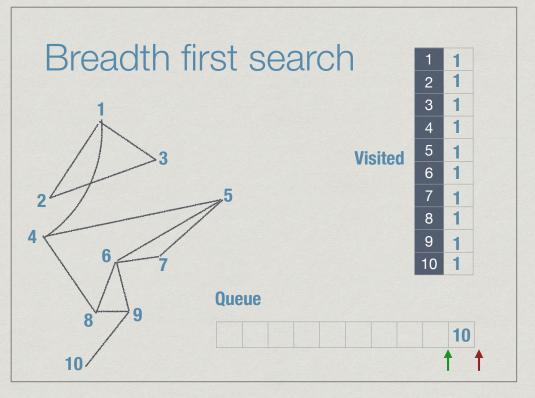












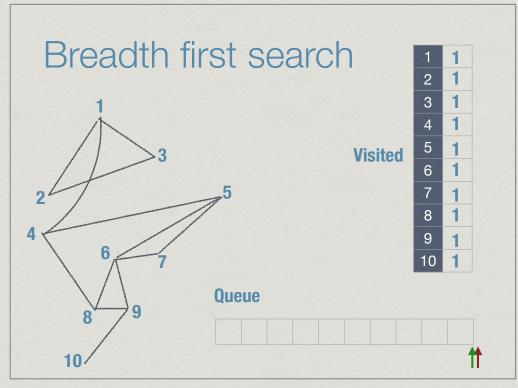
Breadth first search

function BFS(i) // BFS starting from vertex i

//Initialization
for j = 1..n {visited[j] = 0}; Q = []

//Start the exploration at i
visited[i] = 1; append(Q,i)

//Explore each vertex in Q
while Q is not empty
j = extract_head(Q)
for each (j,k) in E
 if visited[k] == 0
 visited[k] = 1; append(Q,k)



Complexity of BFS

- * Each vertex enters Q exactly once
- If graph is connected, loop to process Q iterated n times
 - For each j extracted from Q, need to examine all neighbours of j
 - * In adjacency matrix, scan row j: n entries
- Hence, overall O(n²)

Complexity of BFS

- * Let m be the number of edges in E. What if m << n²?
- Adjacency list: scanning neighbours of j takes time proportional to number of neighbours (degree of j)
- Across the loop, each edge (i,j) is scanned twice, once when exploring i and again when exploring j
 - * Overall, exploring neighbours takes time O(m)
- * Marking n vertices visited still takes O(n)
- * Overall, O(n+m)

Enhancements to BFS

- If BFS(i) sets visited[j] = 1, we know that i and j are connected
- * How do we identify a path from i to j
- When we mark visited[k] = 1, remember the neighbour from which we marked it
 - If exploring edge (j,k) visits k, set parent[k] = j

Complexity of BFS

- For graphs, O(m+n) is considered the best possible
 - * Need to see each edge and vertex at least once
- O(m+n) is considered to be linear in the size of the graph

Breadth first search

function BFS(i) // BFS starting from vertex i

```
//Initialization
for j = 1..n {visited[j] = 0; parent[j] = -1}
Q = []
```

//Start the exploration at i
visited[i] = 1; append(Q,i)

```
//Explore each vertex in Q
while Q is not empty
    j = extract_head(Q)
    for each (j,k) in E
        if visited[k] == 0
            visited[k] = 1; parent[k] = j; append(Q,k);
```

Reconstructing the path

- * BFS(i) sets visited[j] = 1
- * visited[j] = 1, so parent[j] = j' for some j'
- * visited[j'] = 1, so parent[j'] = j" for some j"
- * ...
- * Eventually, trace back path to k with parent[k] = i

Breadth first search

function BFS(i) // BFS starting from vertex i

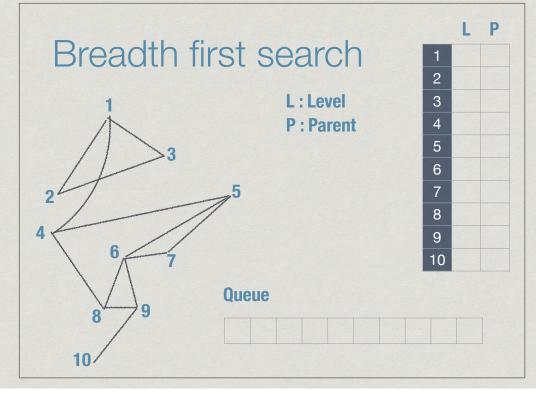
```
//Initialization
for j = 1..n {level[j] = -1; parent[j] = -1}
Q = []
```

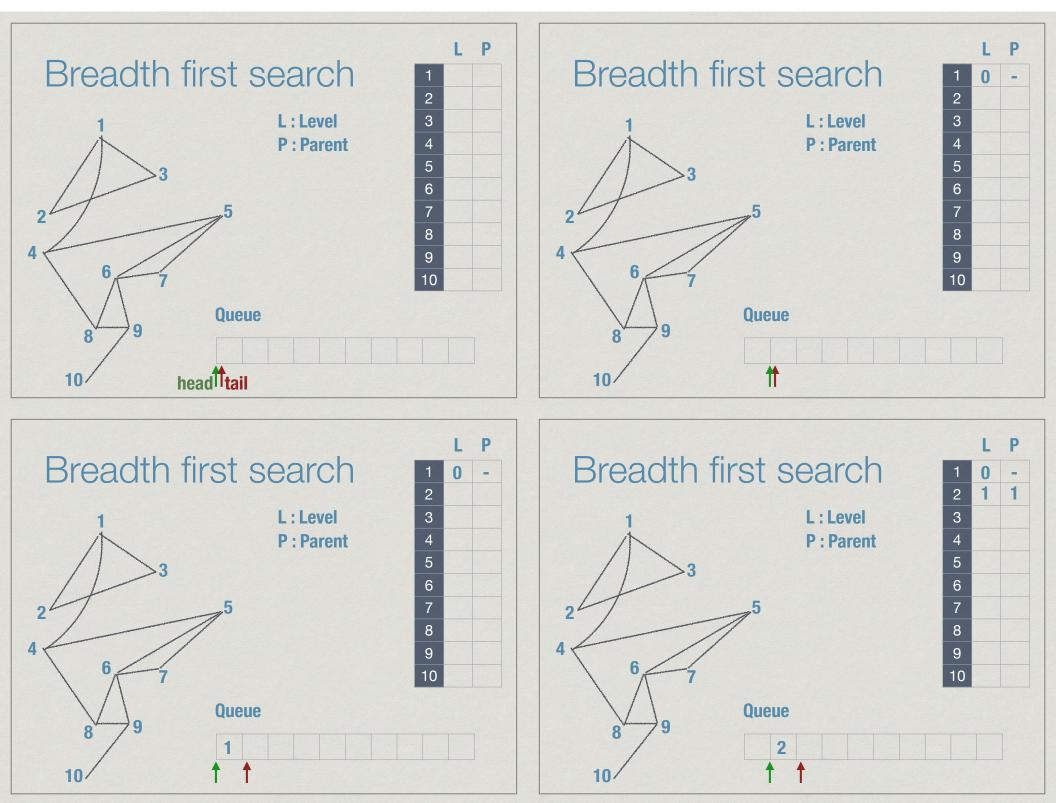
```
//Start the exploration at i, level[i] set to 0
level[i] = 0; append(Q,i)
```

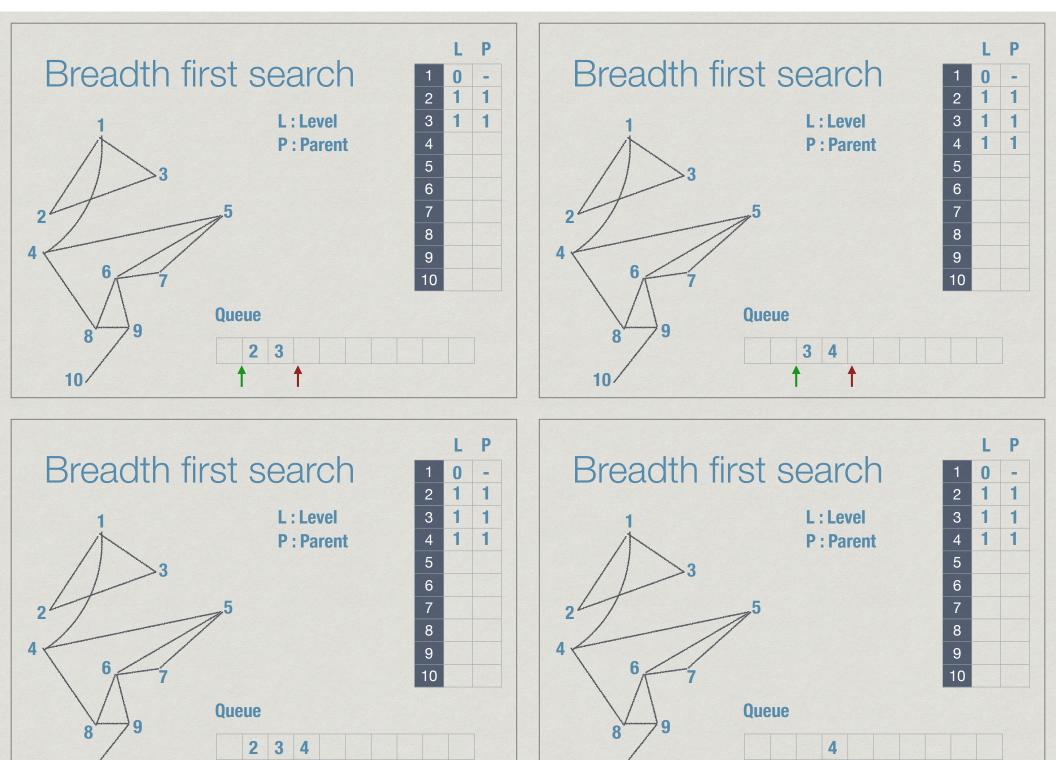
//Explore each vertex in Q, increment level for each new vertex
while Q is not empty
 j = extract_head(Q)
 for each (j,k) in E
 if level[k] == -1
 level[k] = 1+level[j]; parent[k] = j;
 append(Q,k);

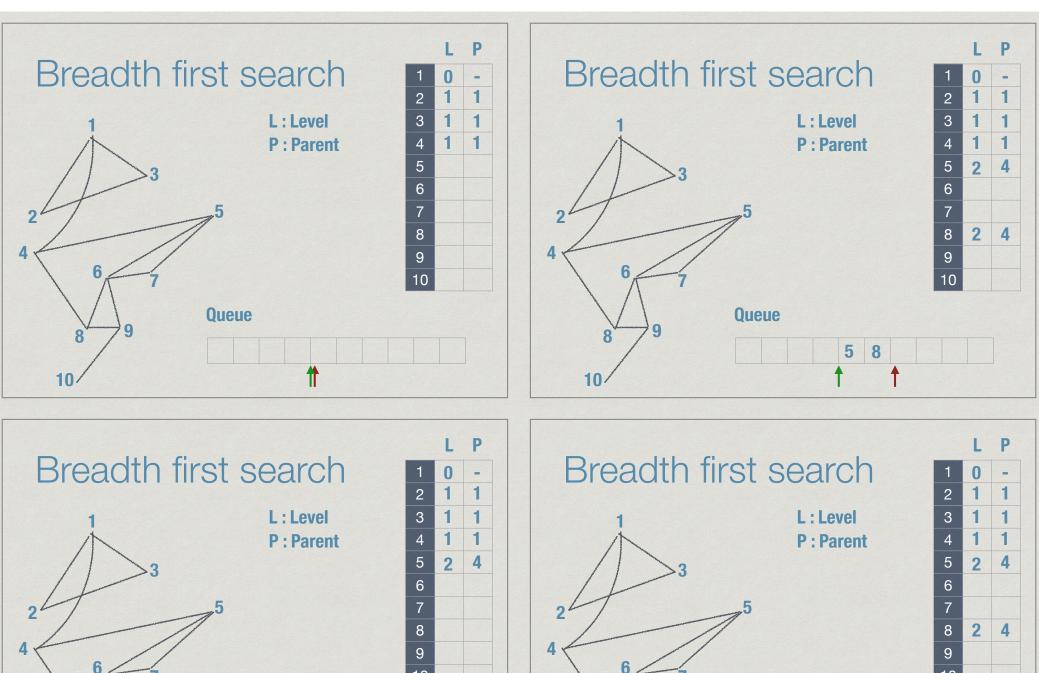
Recording distances

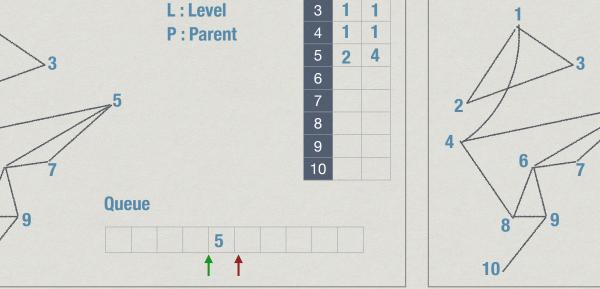
- BFS can record how long the path is to each vertex
- Instead of binary array visited[], keep integer array level[]
- * level[j] = -1 initially
- * level[j] = p means j is reached in p steps from i

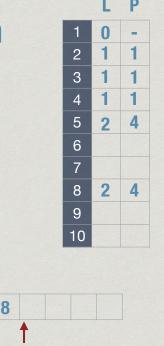




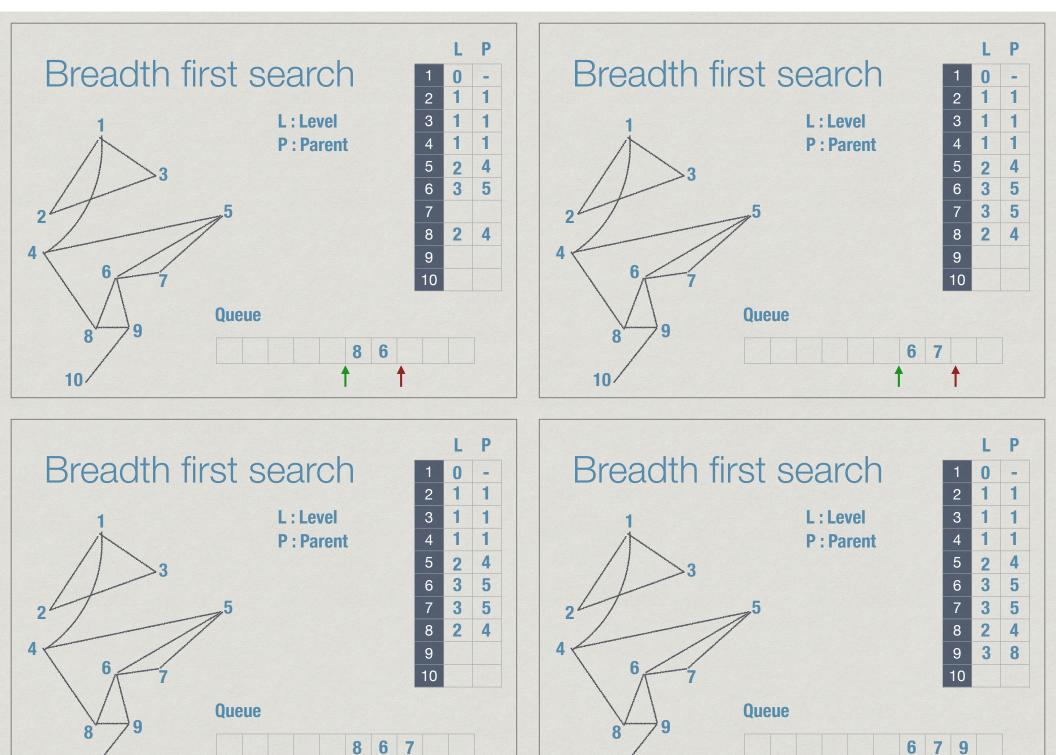


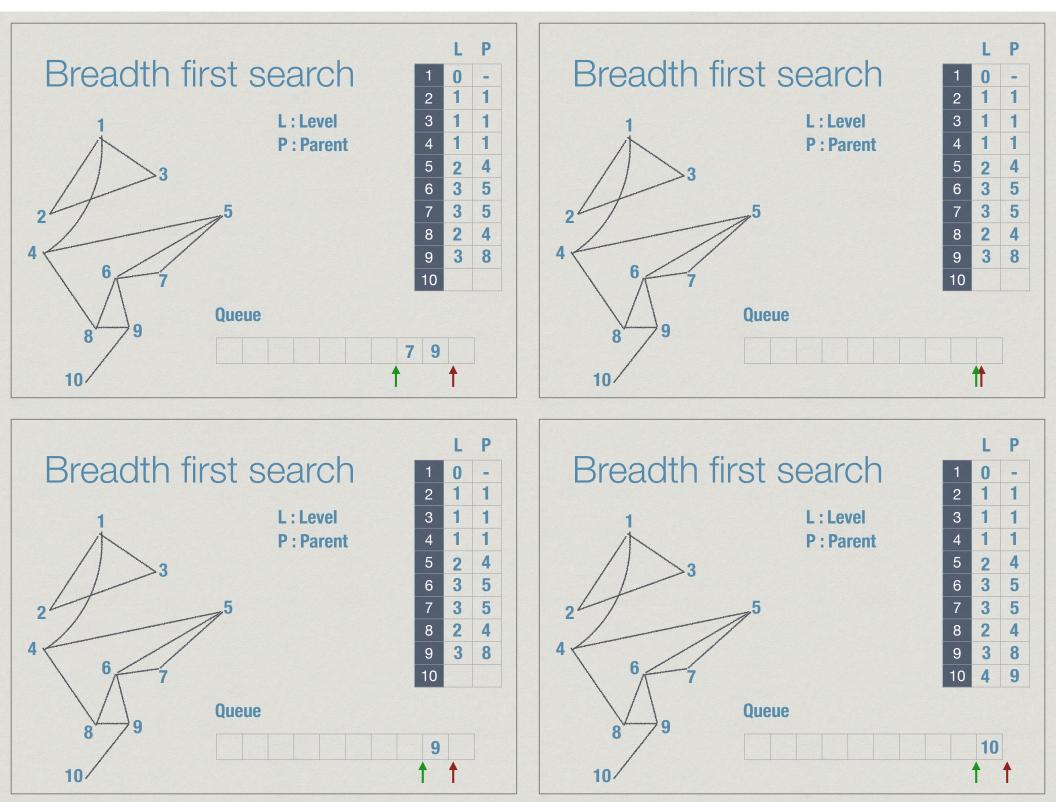


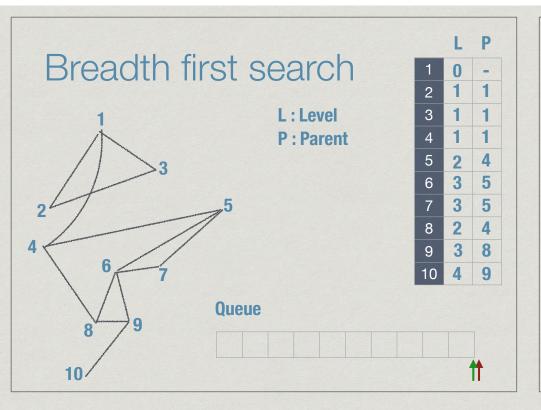




Queue





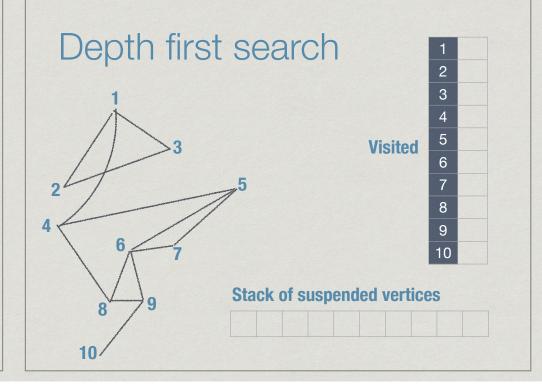


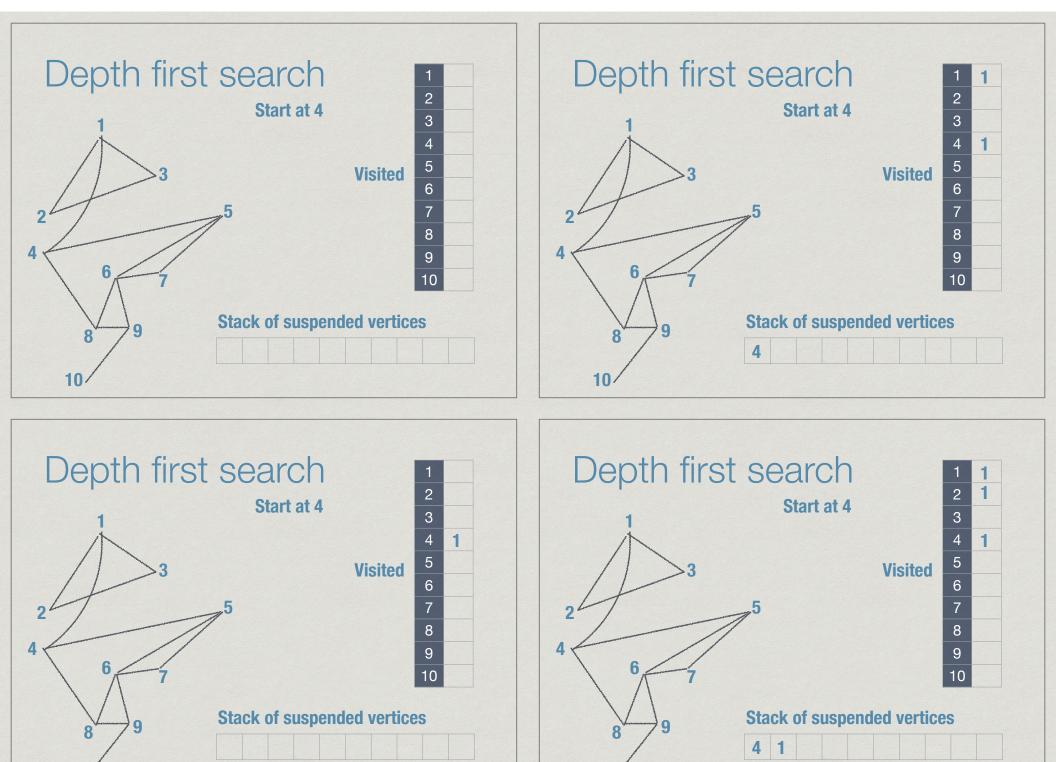
Depth first search

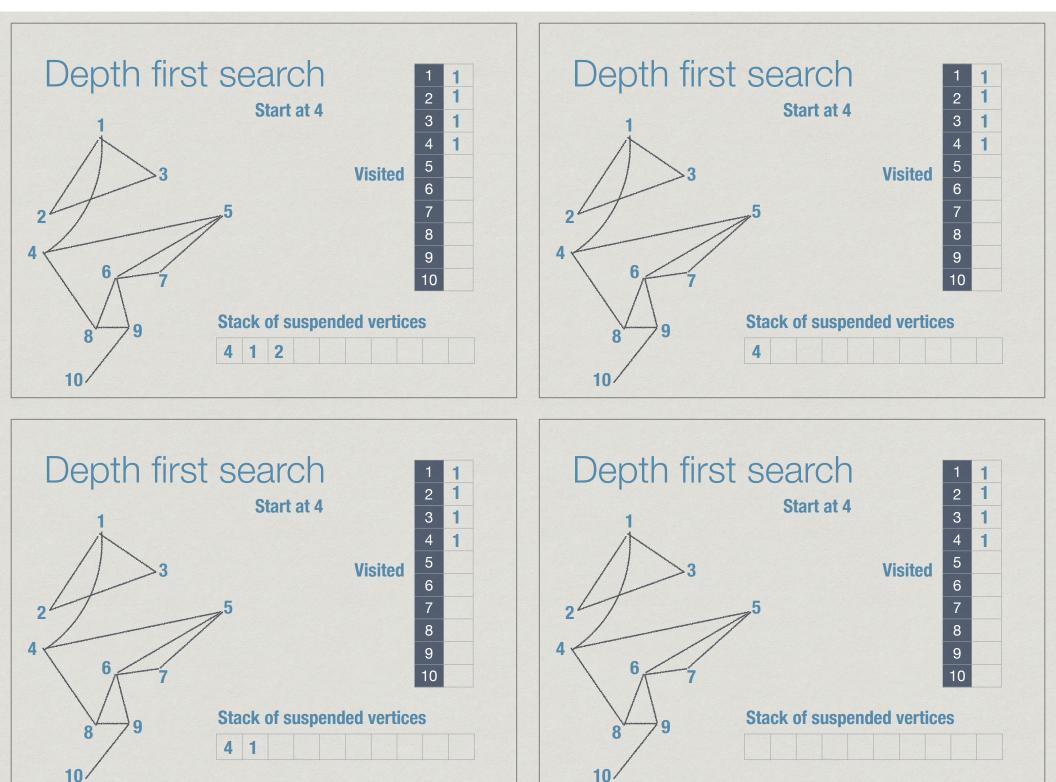
- * Start from i, visit a neighbour j
- * Suspend the exploration of i and explore j instead
- Continue till you reach a vertex with no unexplored neighbours
- Backtrack to nearest suspended vertex that still has an unexplored neighbour
- * Suspended vertices are stored in a stack
 - Last in, first out: most recently suspended is checked first

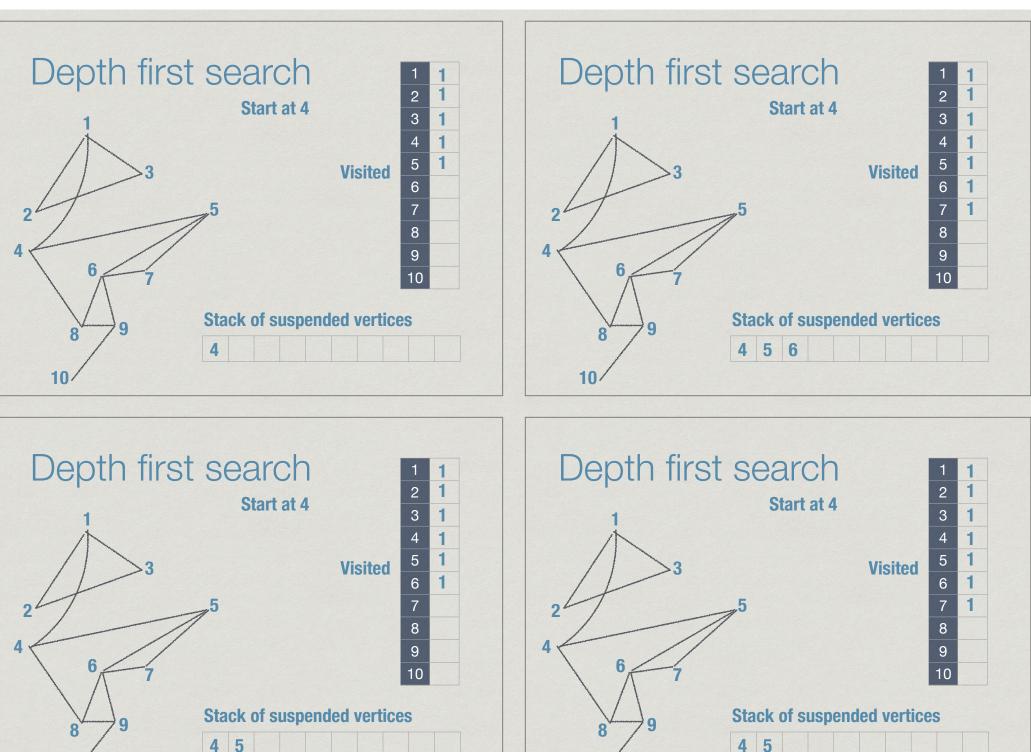
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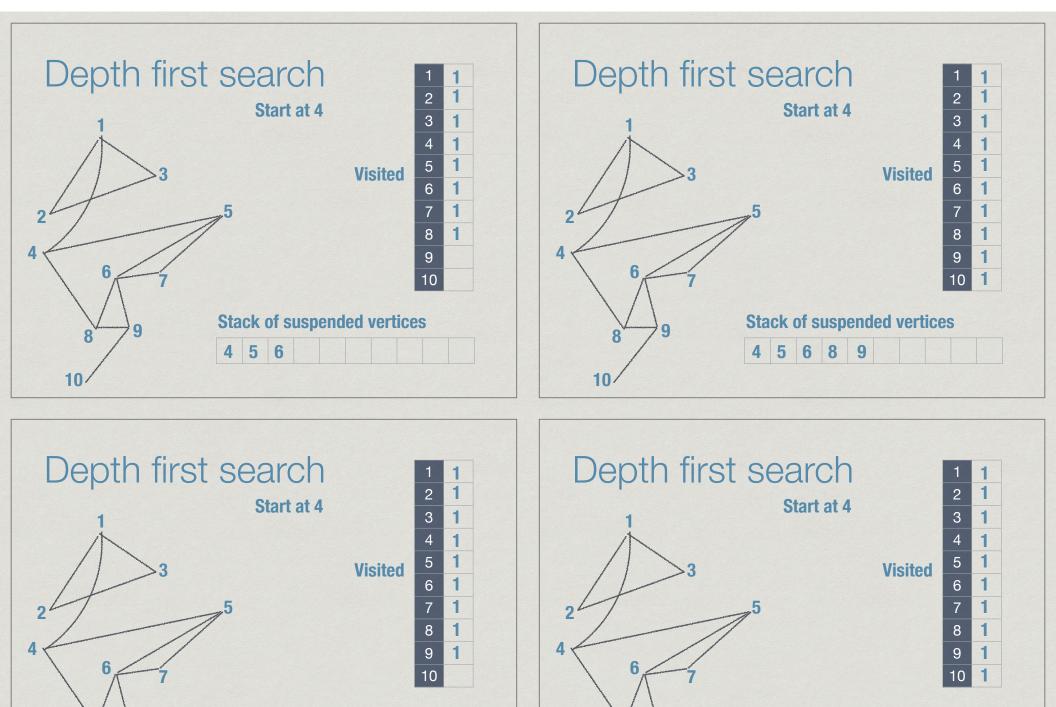
- BFS with level[] gives us the shortest path to each node in terms of number of edges
- In general, edges are labelled by a cost (money, time, distance ...)
 - Min cost path not same as fewest edges
- * Will look at shortest paths in weighted graphs later
 - * BFS computes shortest paths if all costs are 1











Stack of suspended vertices

8

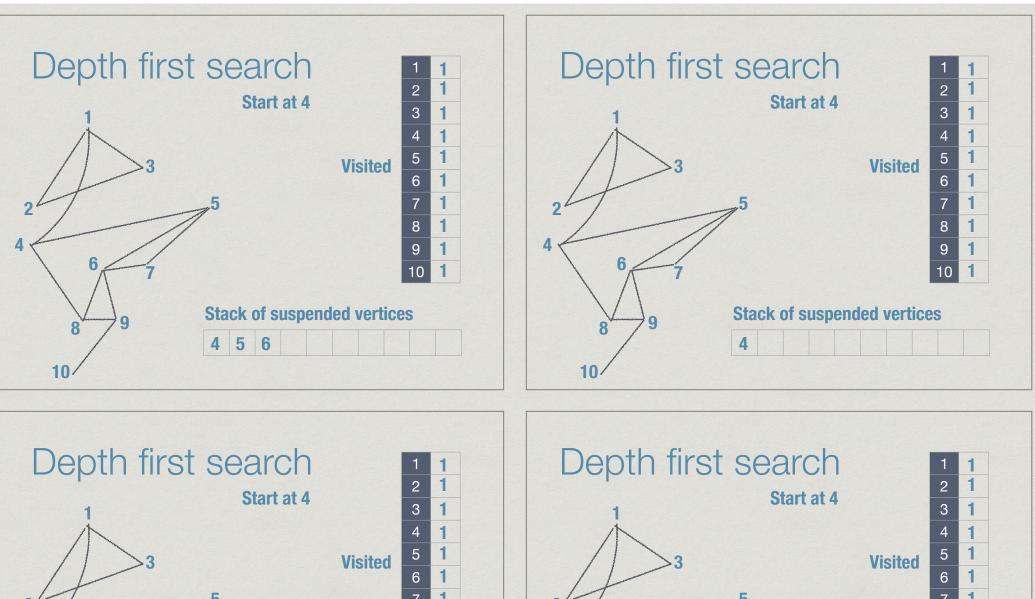
4 5 6

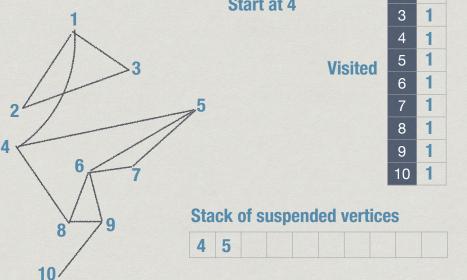
10

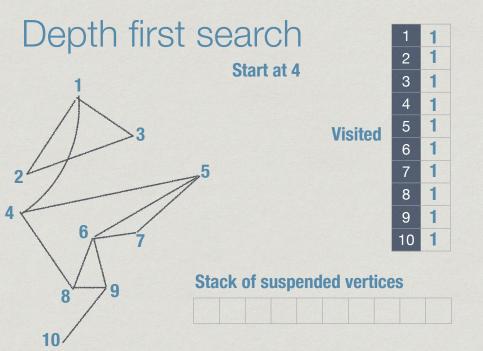
Stack of suspended vertices

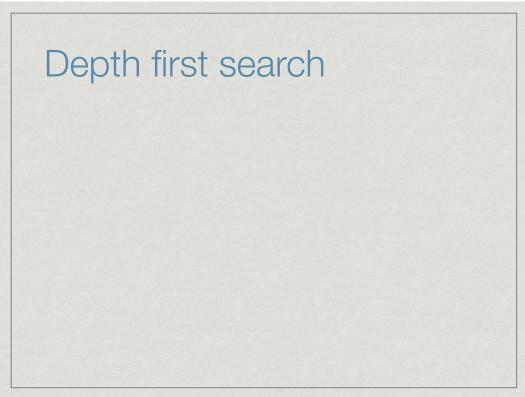
4 5

6









Depth first search

- * DFS is most natural to implement recursively
 - For each unvisited neighbour j of i, call DFS(j)
- * No need to explicitly maintain a stack
 - * Stack is maintained implicitly by recursive calls

Depth first search

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 - * For each unvisited neighbour j of i, call DFS(j)

Depth first search

```
//Initialization
```

```
for j = 1..n {visited[j] = 0; parent[j] = -1}
```

function DFS(i) // DFS starting from vertex i

```
//Mark i as visited
visited[i] = 1
```

```
//Explore each neighbour of i recursively
for each (i,j) in E
    if visited[j] == 0
        parent[j] = i
        DFS(j)
```



Complexity of DFS

- * Each vertex marked and explored exactly once
- * DFS(j) need to examine all neighbours of j

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Complexity of DFS

- * Each vertex marked and explored exactly once
- * DFS(j) need to examine all neighbours of j
- * In adjacency matrix, scan row j: n entries
 - * Overall O(n²)
- With adjacency list, scanning takes O(m) time across all vertices
 - * Total time is O(m+n), like BFS

Properties of DFS

 Paths discovered by DFS are not shortest paths, unlike BFS

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 - * DFS numbering
 - * Maintain a counter
 - * Increment and record counter value when entering and leaving a vertex.

Depth first search

```
//Initialization
for j = 1..n {visited[j] = 0; parent[j] = -1}
count = 0
```

```
function DFS(i) // DFS starting from vertex i
```

```
//Mark i as visited
visited[i] = 1; pre[i] = count; count++
```

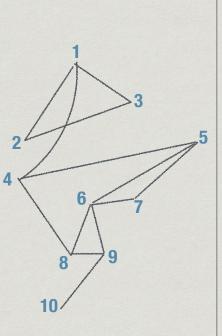
```
//Explore each neighbours of i recursively
for each (i,j) in E
    if visited[j] == 0
        parent[j] = i
        DFS(j)
        post[i] = count; count++
```

DFS numbering

pre[i] and post[i] can be used to find

- if the graph has a cycle —
 i.e., a loop
- * cut vertex removal disconnects the graph

*



DFS numbering

Summary

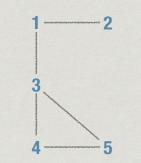
- BFS and DFS are two systematic ways to explore a graph
 - Both take time linear in the size of the graph with adjacency lists
- * Recover paths by keeping parent information
- BFS can compute shortest paths, in terms of number of edges
- * DFS numbering can reveal many interesting features

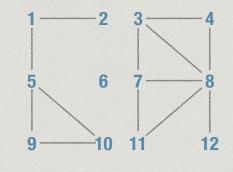
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Connectivity





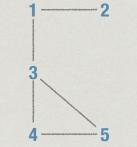
Connected graph

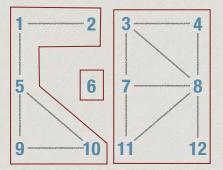
Disconnected graph

Exploring graph structure

- * Breadth first search
 - Level by level exploration
- * Depth first search
 - * Explore each vertex as soon as it is visited
 - * DFS numbering
- What can we find out about a graph using BFS/ DFS?

Connectivity





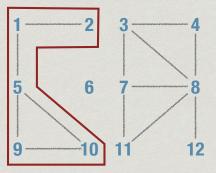
Connected graph

Disconnected graph Connected components

Identifying connected components

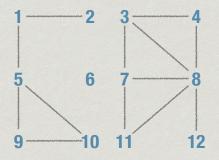
- * Vertices {1,2,...,N}
- * Start BFS or DFS from 1
 - * All nodes marked Visited form a connected component
 - Pick first unvisited node, say j, and run BFS or DFS from j
 - * Repeat till all nodes are visited
- Update BFS/DFS to label each visited node with component number

Connected components



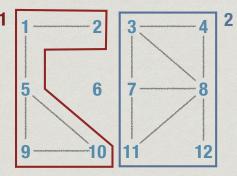
- * Add a counter comp to number components
- Increment counter each time a fresh BFS/DFS starts
- * Label each visited node j with component[j] = comp

Connected components



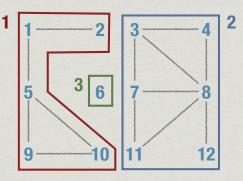
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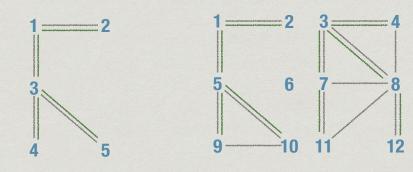
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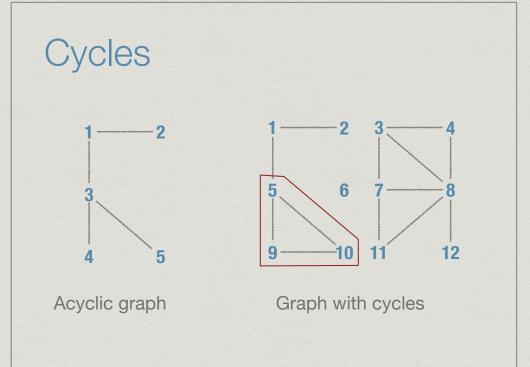
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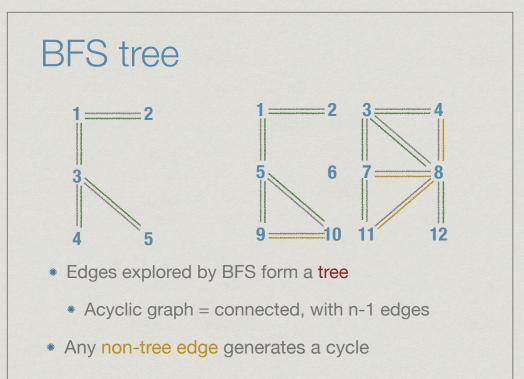
BFS tree

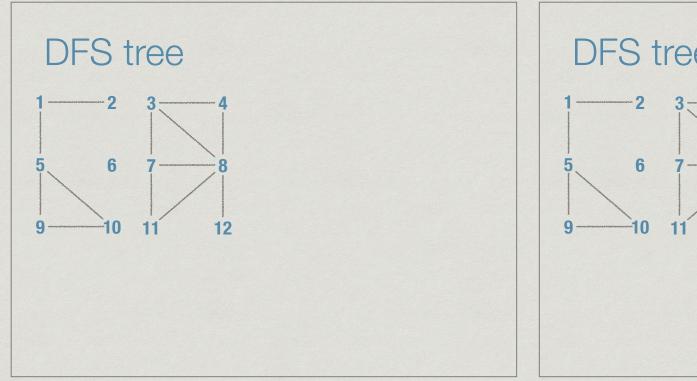


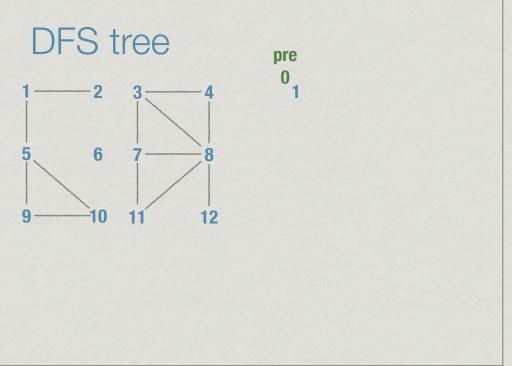
* Edges explored by BFS form a tree

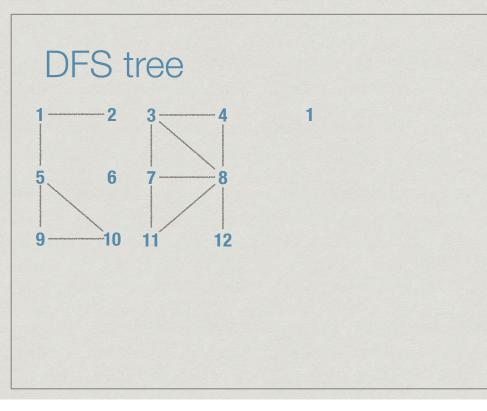
* Acyclic graph = connected, with n-1 edges

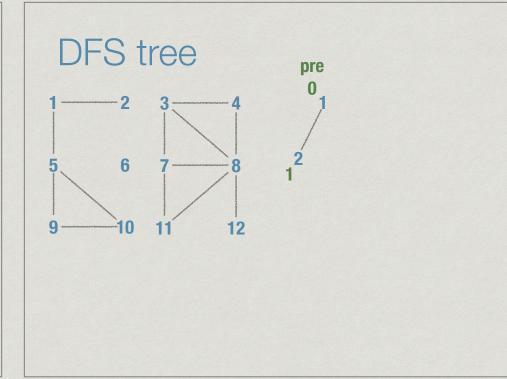


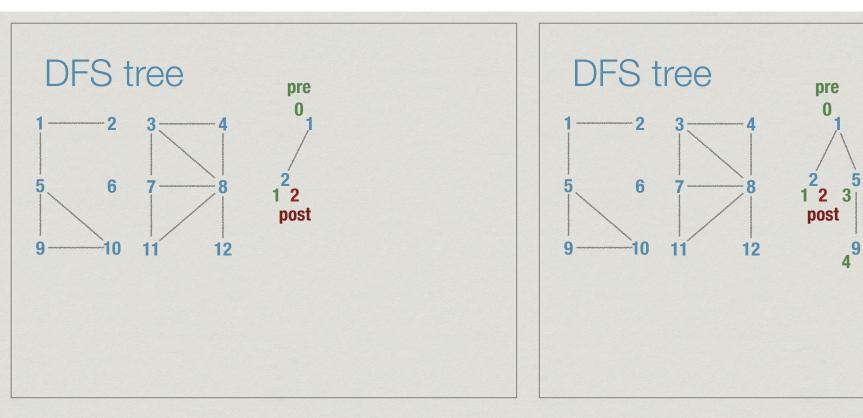


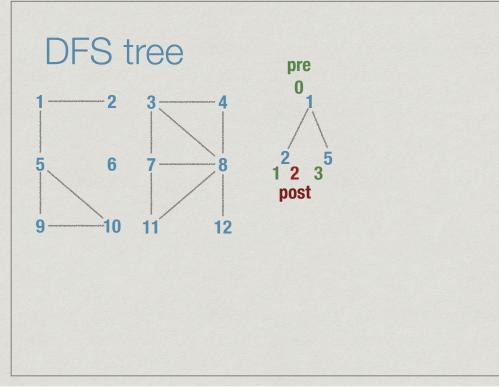


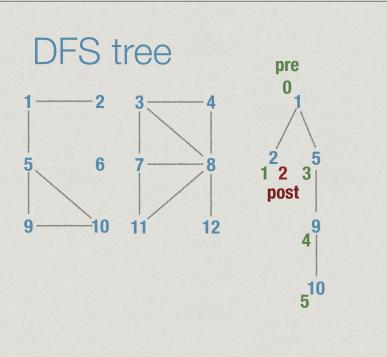


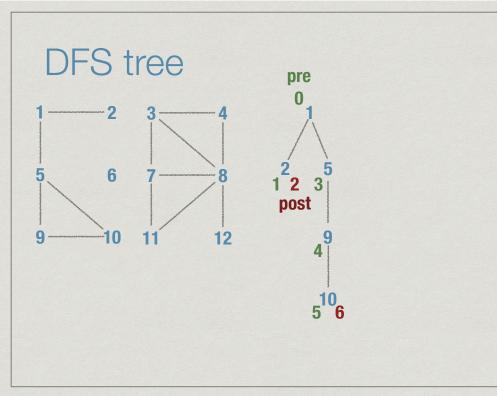


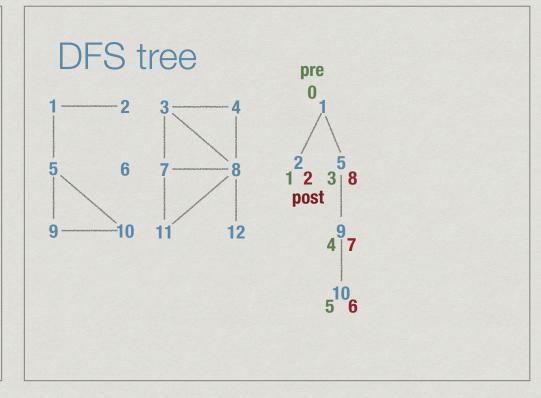


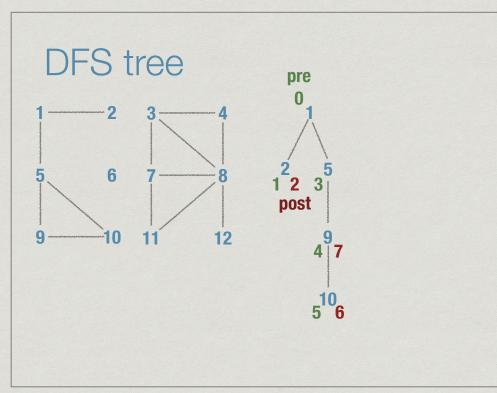


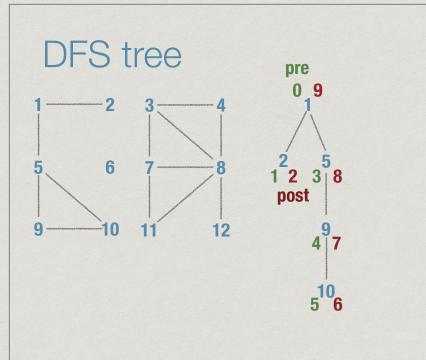


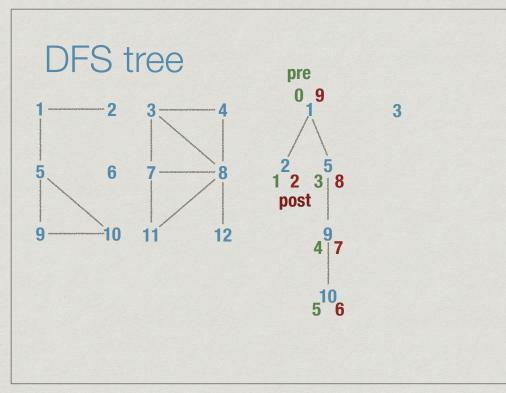


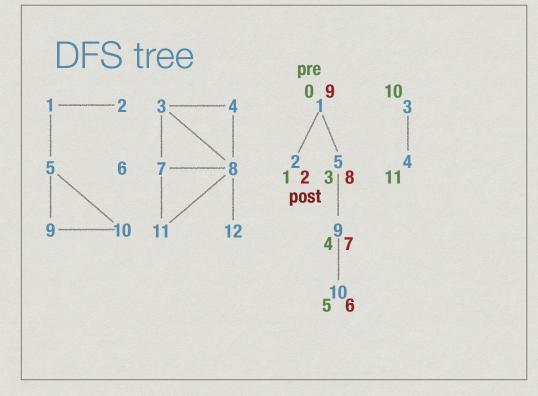


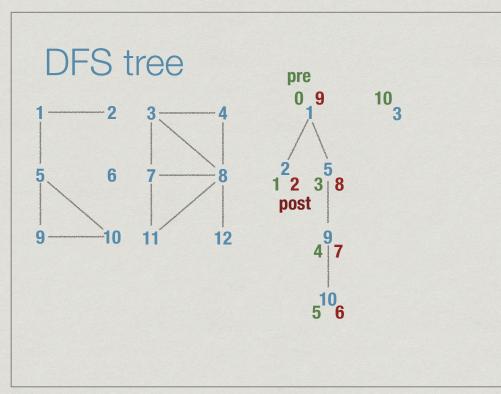


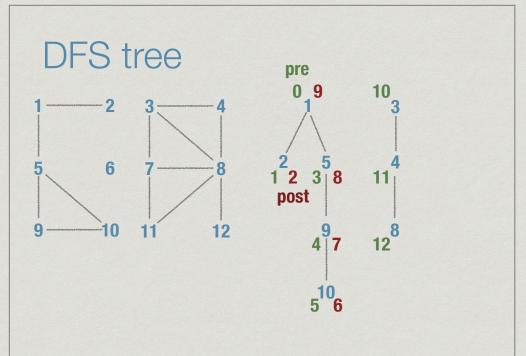


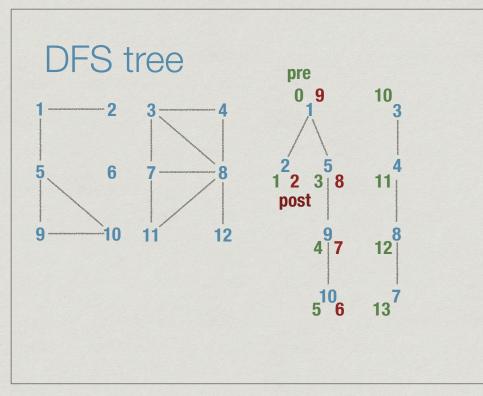


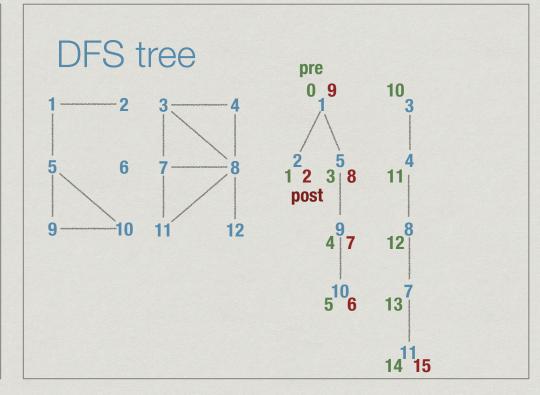


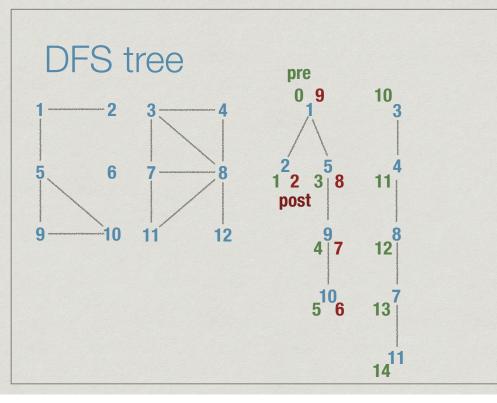


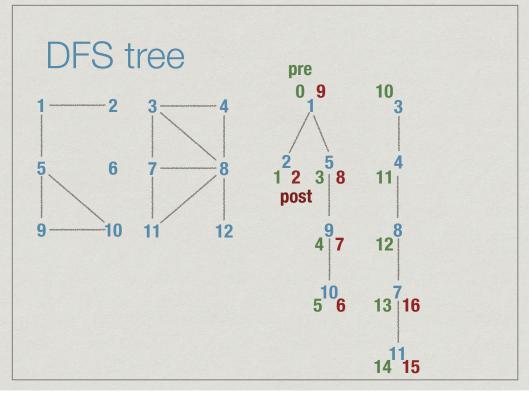


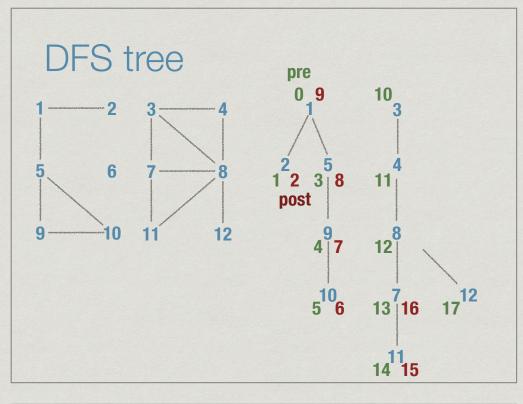


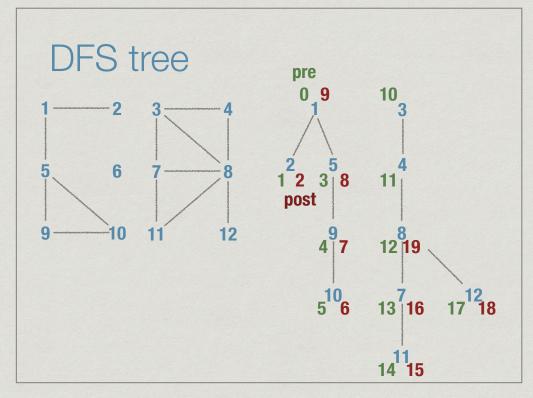


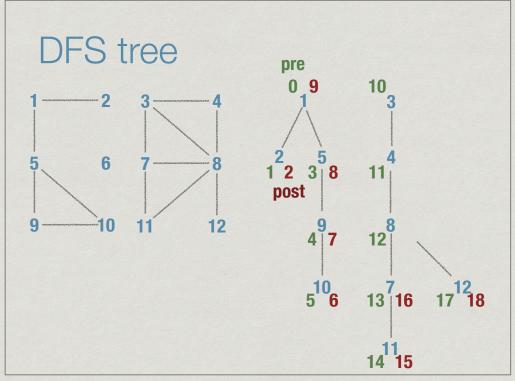


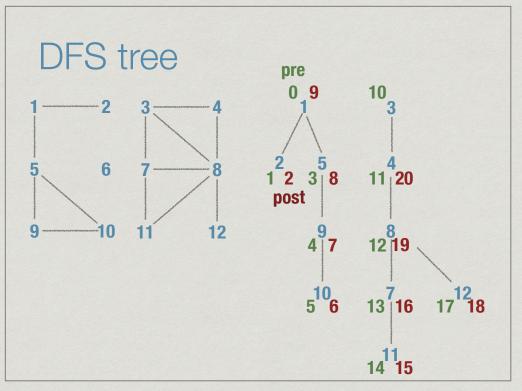


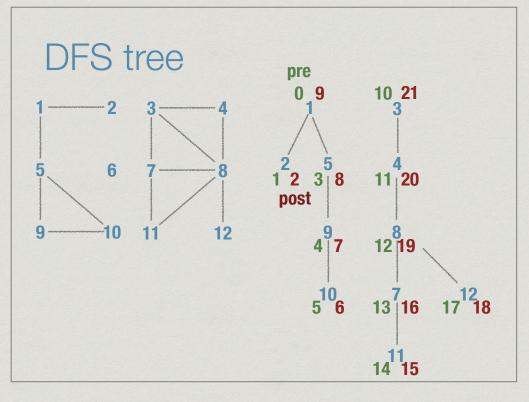


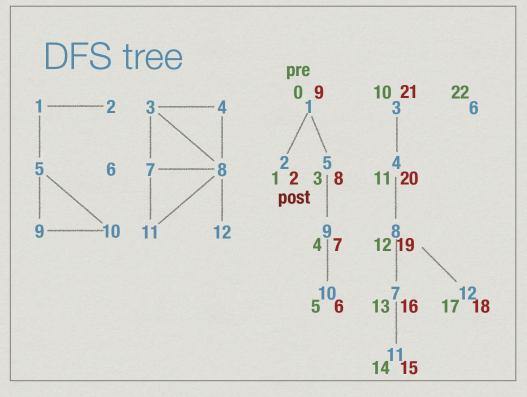


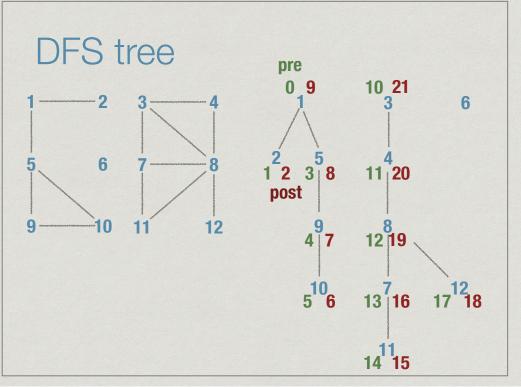


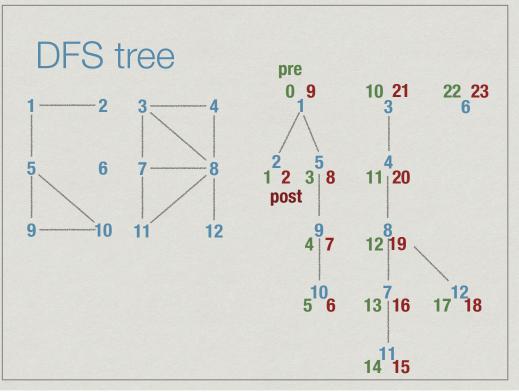


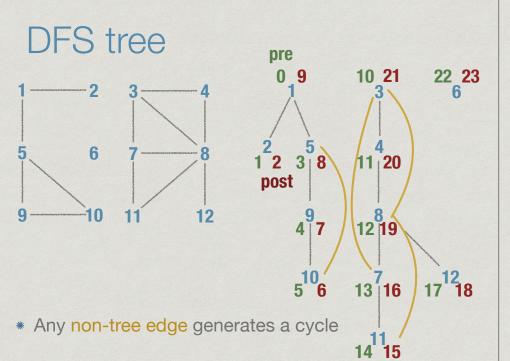


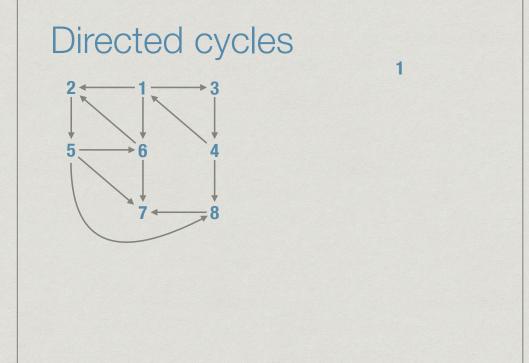


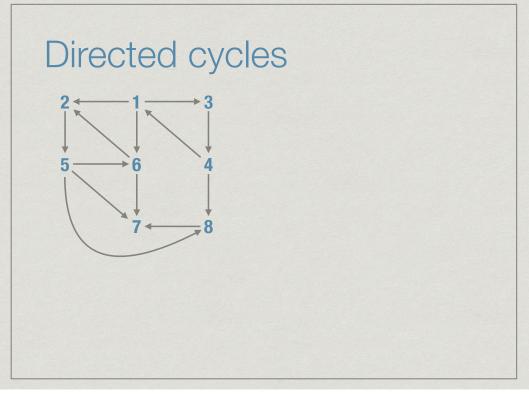


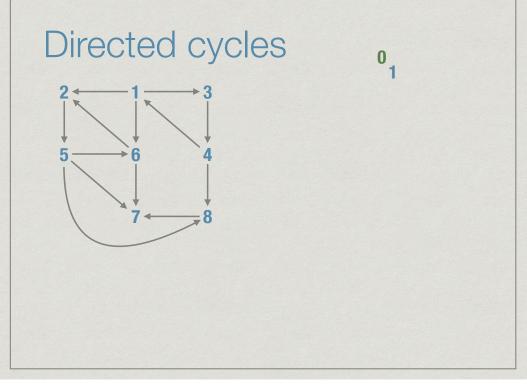


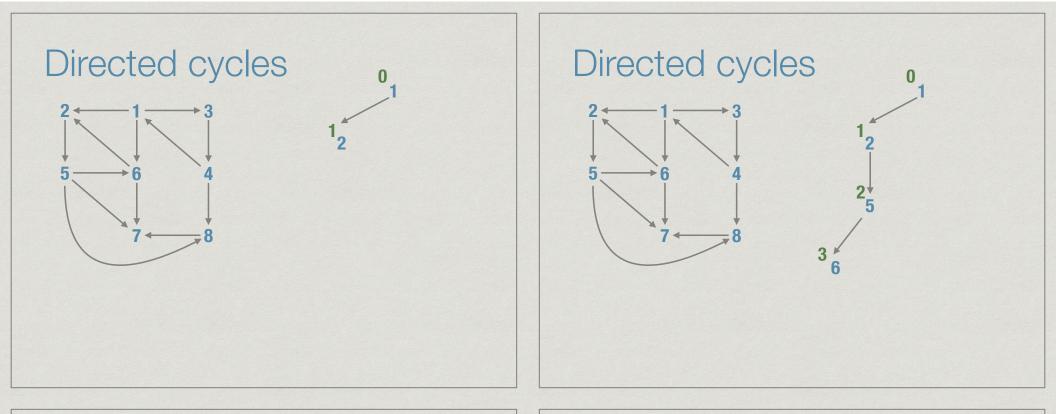


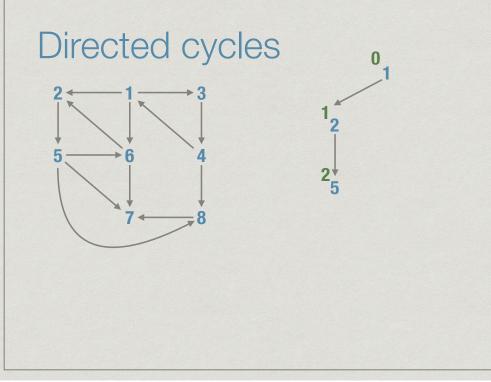


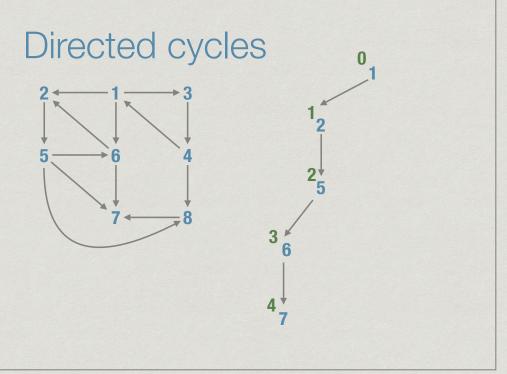


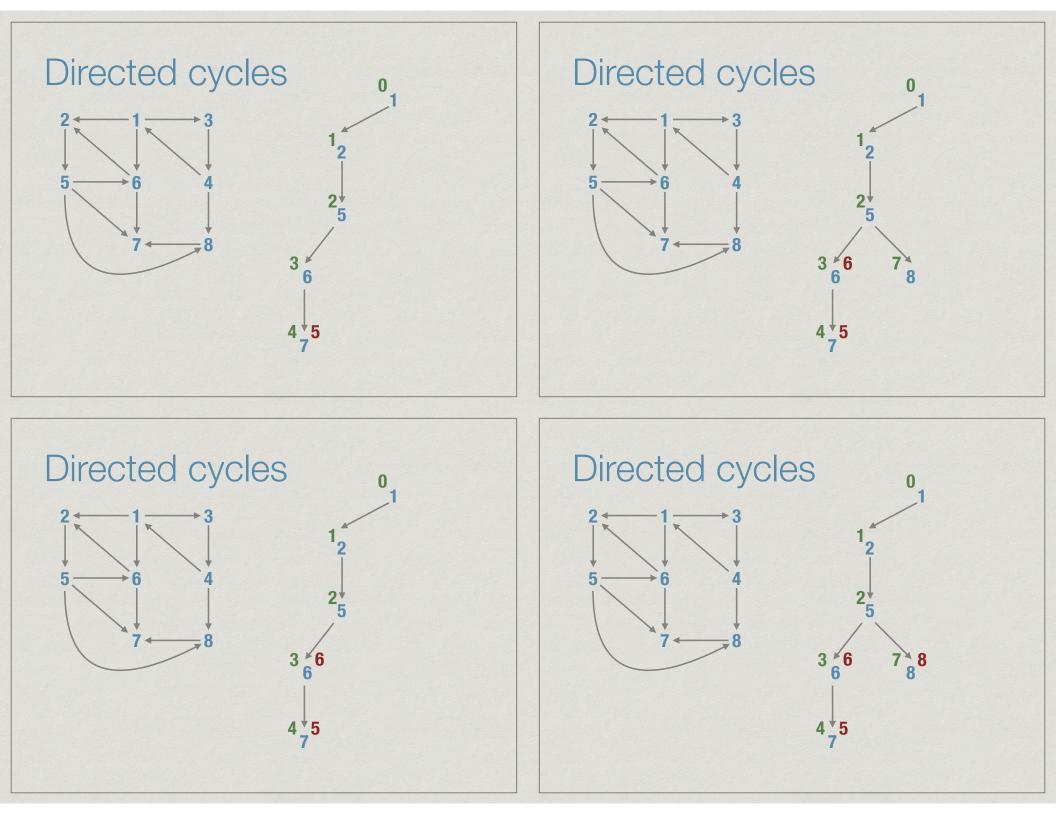


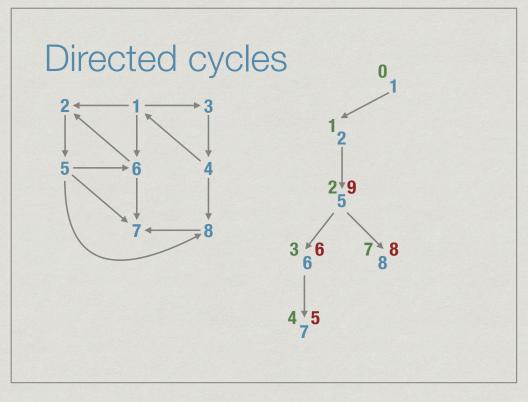


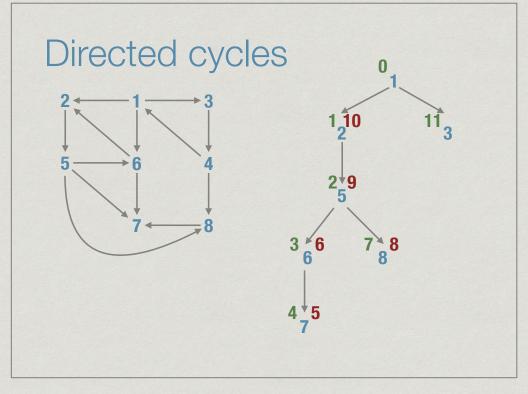


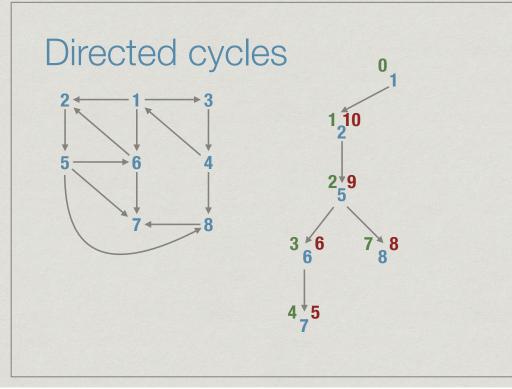


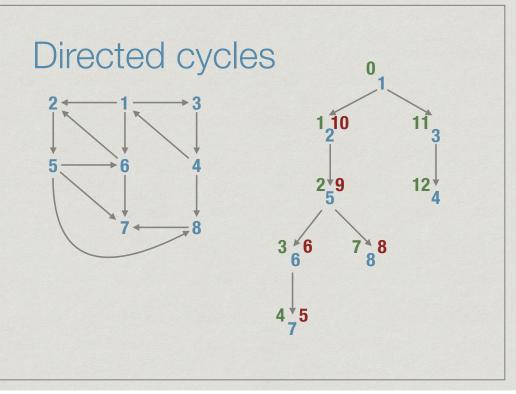


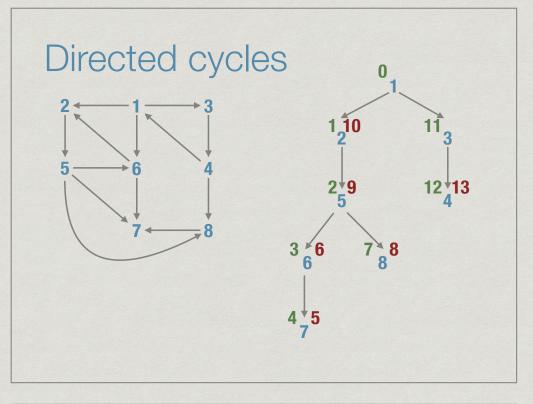


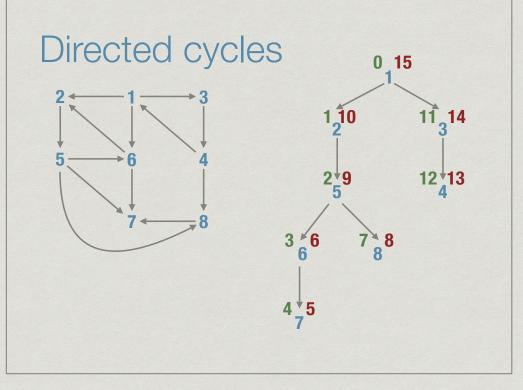


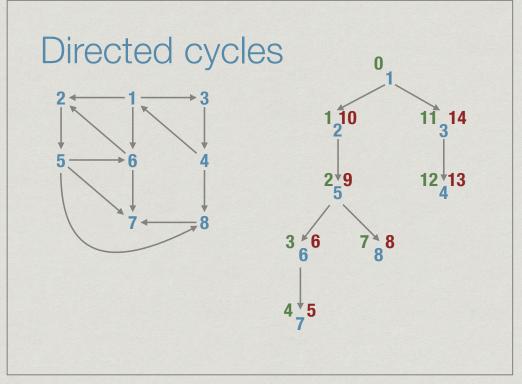


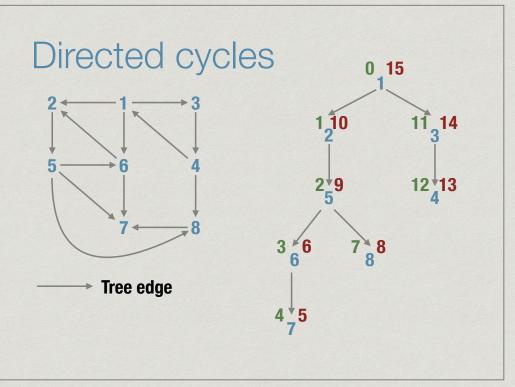


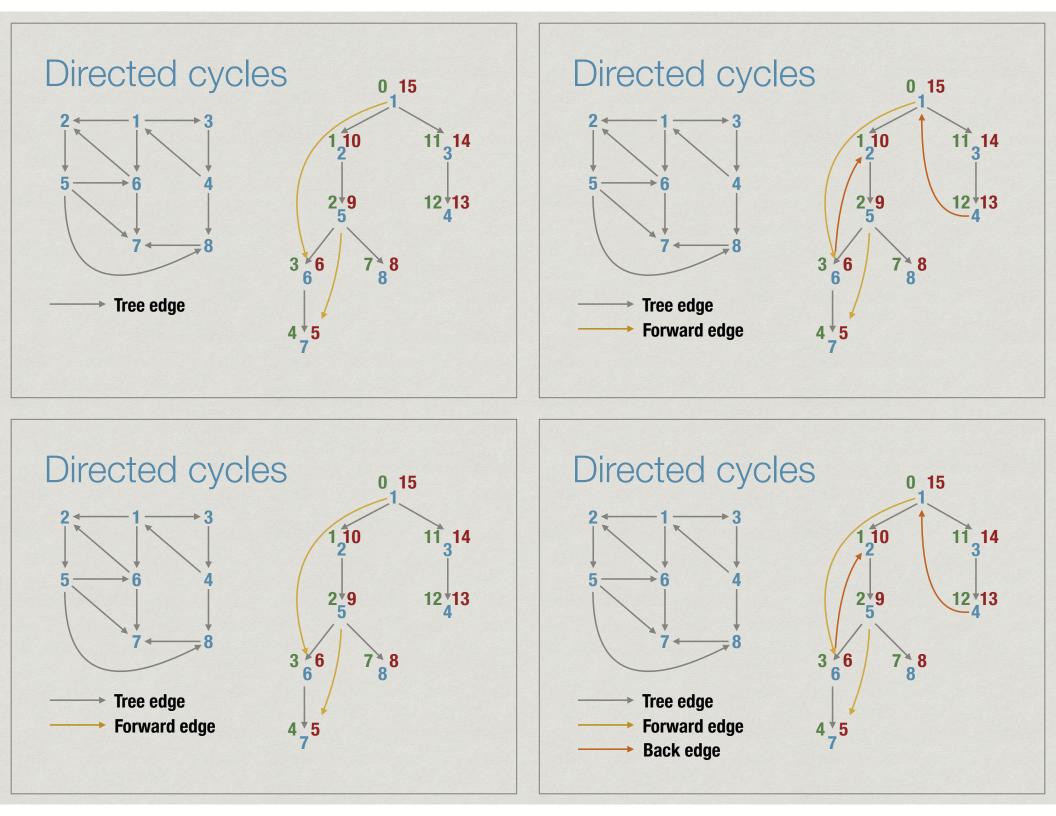


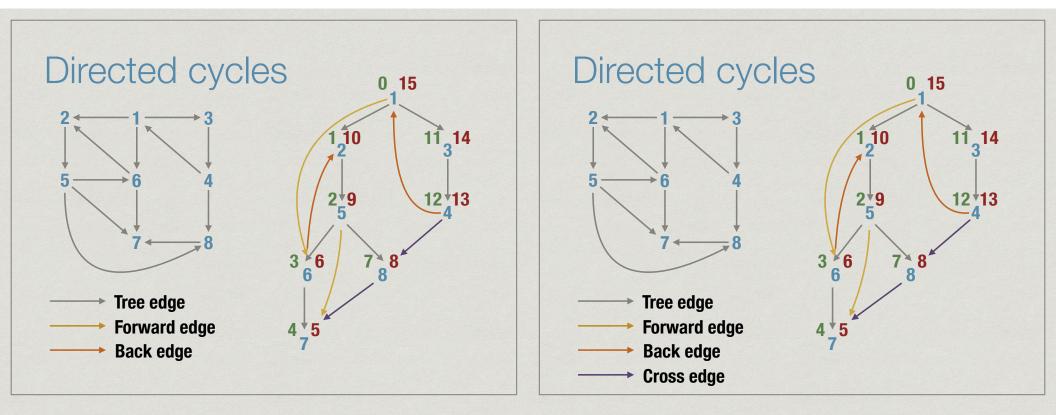


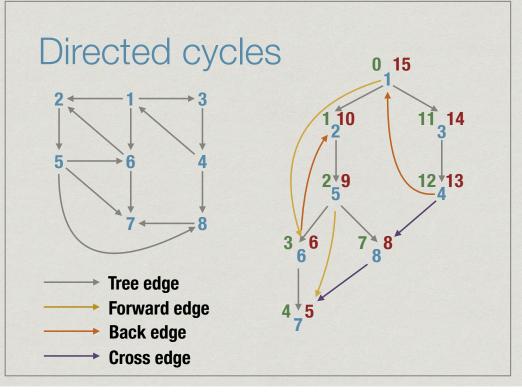












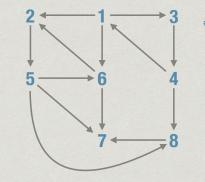
Directed cycles

- A directed graph has a cycle if and only if DFS reveals a back edge
- * Can classify edges using pre and post numbers
 - Tree/Forward edge (u,v) : Interval [pre(u),post(u)] contains [(pre(v),post(v)]
 - Backward edge (u,v): Interval [pre(v),post(v)] contains [(pre(u),post(u)]
 - Cross edge (u,v): Intervals [(pre(u),post(u)] and [(pre(v),post(v)] disjoint

Directed acyclic graphs

- Directed graphs without cycles are useful for modelling dependencies
 - Courses with prerequisites
 - Edge (Algebra, Calculus) indicates that Algebra is a prerequisite for Calculus
- * Will look at Directed Acyclic Graphs (DAGs) soon

Computing SCCs

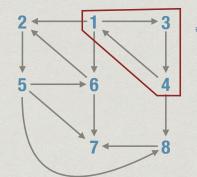


- DFS numbering (pre and post) can be used to compute SCCs
 - [Dasgupta, Papadimitriou,Vazirani]

Connectivity in directed graphs

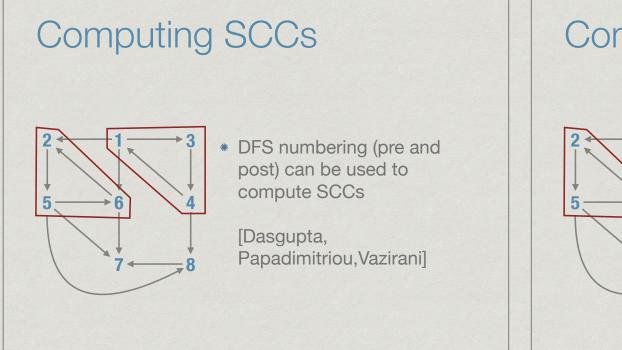
- * Need to take directions into account
- Nodes i and j are strongly connected if there is a path from i to j and a path from j to i
- Directed graph can be decomposed into strongly connected components (SCCs)
 - All pairs of nodes in an SCC are strongly connected

Computing SCCs

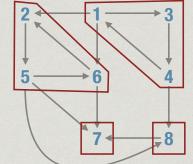


 DFS numbering (pre and post) can be used to compute SCCs

[Dasgupta, Papadimitriou,Vazirani]

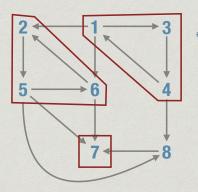


Computing SCCs



- DFS numbering (pre and post) can be used to compute SCCs
 - [Dasgupta, Papadimitriou,Vazirani]

Computing SCCs



 DFS numbering (pre and post) can be used to compute SCCs

[Dasgupta, Papadimitriou,Vazirani]

Other properties

- * A number of other structural properties can be inferred from DFS numbering
- * Articulation points (vertices)
 - Removing such a vertex disconnects the graph
- * Bridges (edges)
 - Removing such an edge disconnects the graph