CSE 421: Introduction to Algorithms

Bipartiteness - DFS

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Bipartite Graphs

Definition: An undirected graph G=(V,E) is bipartite

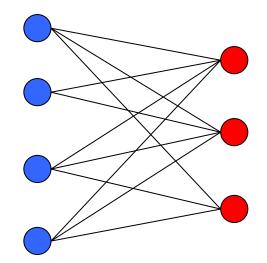
if you can partition the node set into 2 parts (say, blue/red or left/right) so that

all edges join nodes in different parts

i.e., no edge has both ends in the same part.

Application:

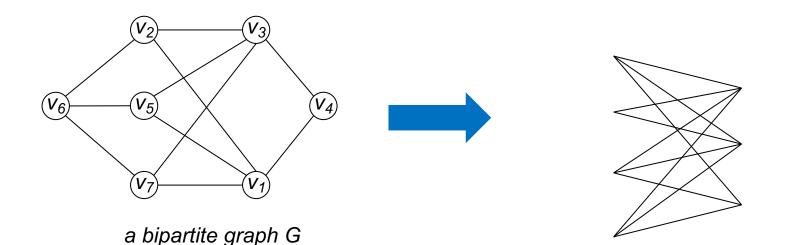
- Scheduling: machine=red, jobs=blue
- Stable Matching: men=blue, wom=red



a bipartite graph

Testing Bipartiteness

Problem: Given a graph G, is it bipartite?

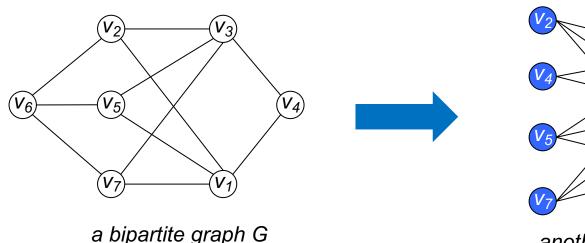


Testing Bipartiteness

Problem: Given a graph G, is it bipartite?

Many graph problems become:

- Easier if the underlying graph is bipartite (matching)
- Tractable if the underlying graph is bipartite (independent set)
 Before attempting to design an algorithm, we need to understand structure of bipartite graphs.

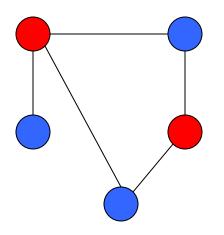


another drawing of G

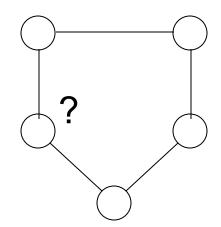
An Obstruction to Bipartiteness

Lemma: If G is bipartite, then it does not contain an odd length cycle.

Pf: We cannot 2-color an odd cycle, let alone G.



bipartite (2-colorable)

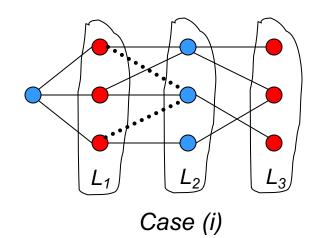


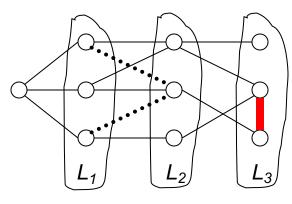
not bipartite (not 2-colorable)

A Characterization of Bipartite Graphs

Lemma: Let G be a connected graph, and let $L_0, ..., Lk$ be the layers produced by BFS(s). Exactly one of the following holds.

- (i) No edge of G joins two nodes of the same layer, and G is bipartite.
- (ii) An edge of G joins two nodes of the same layer, and G contains an odd-length cycle (and hence is not bipartite).





Case (ii)

A Characterization of Bipartite Graphs

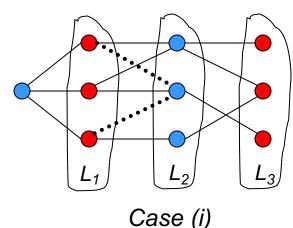
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- (i) No edge of G joins two nodes of the same layer, and G is bipartite.
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Pf. (i)

Suppose no edge joins two nodes in the same layer.

By previous lemma, all edges join nodes on adjacent levels.



Bipartition:

blue = nodes on odd levels, red = nodes on even levels.

A Characterization of Bipartite Graphs

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- (i) No edge of G joins two nodes of the same layer, and G is bipartite.
- (ii) An edge of G joins two nodes of the same layer, and G contains an odd-length cycle (and hence is not bipartite).

Pf. (ii)

Suppose (x, y) is an edge & x, y in same level L_j .

Let z = their lowest common ancestor in BFS tree.

Let L_i be level containing z.

Consider cycle that takes edge from x to y, then tree from y to z, then tree from z to x.

Its length is 1 + (j-i) + (j-i), which is odd.

z = lca(x, y)

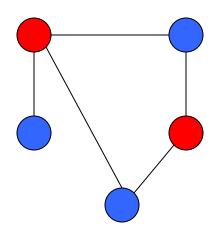
Layer L_i

Layer L_i

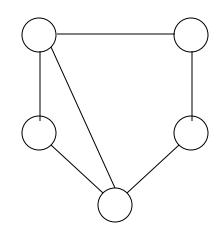
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Obstruction to Bipartiteness

Cor: A graph G is bipartite iff it contains no odd length cycles.



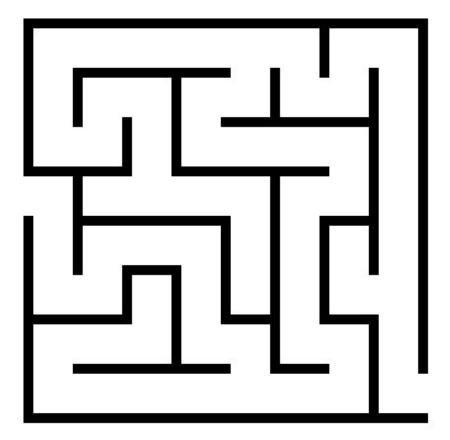
bipartite (2-colorable)



not bipartite (not 2-colorable)

Depth First Search

Follow the first path you find as far as you can go; back up to last unexplored edge when you reach a dead end, then go as far you can



Naturally implemented using recursive calls or a stack

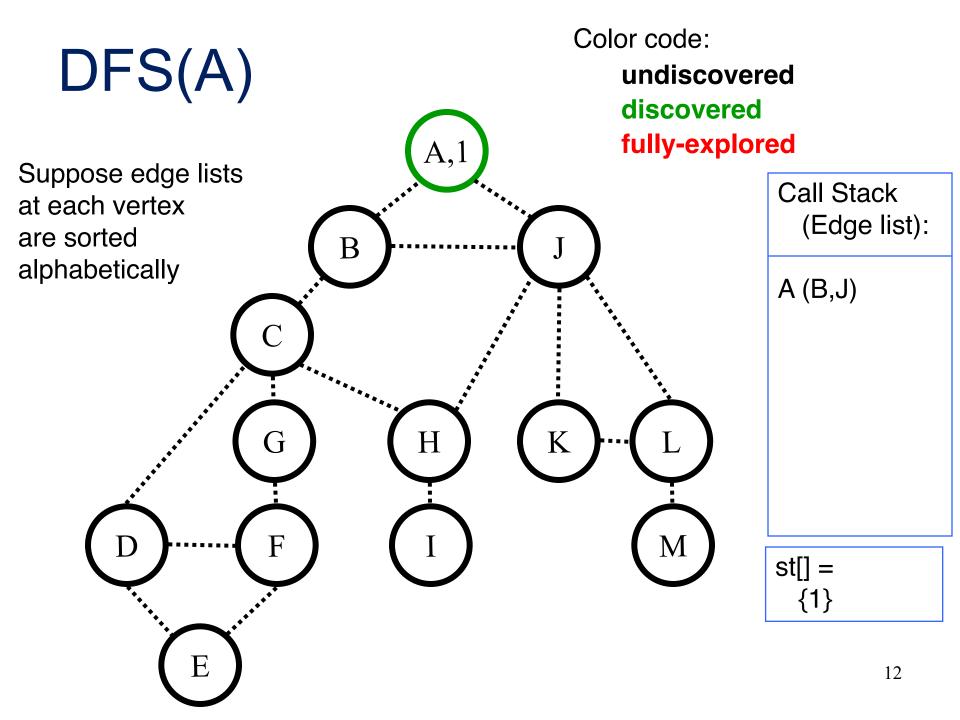
DFS(s) – Recursive version

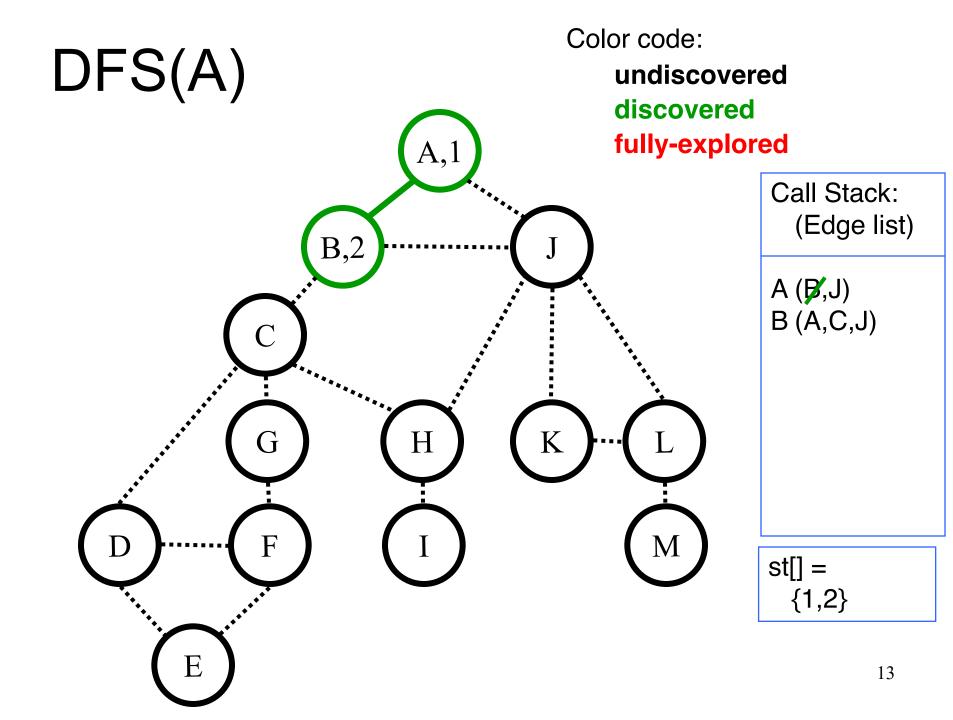
Global Initialization: mark all vertices undiscovered

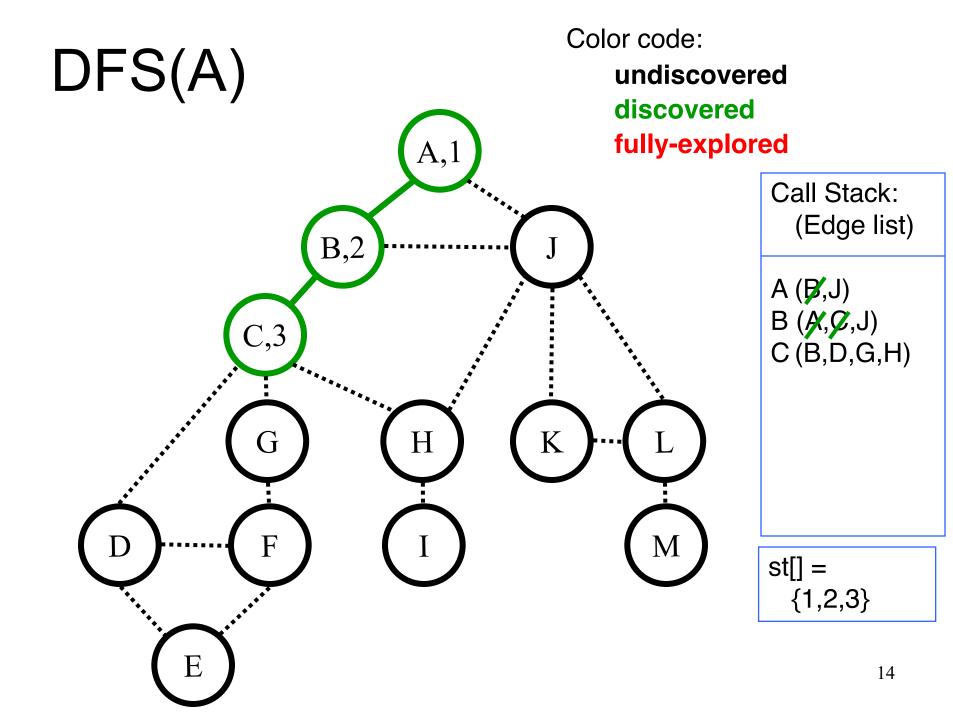
DFS(v) Mark v discovered

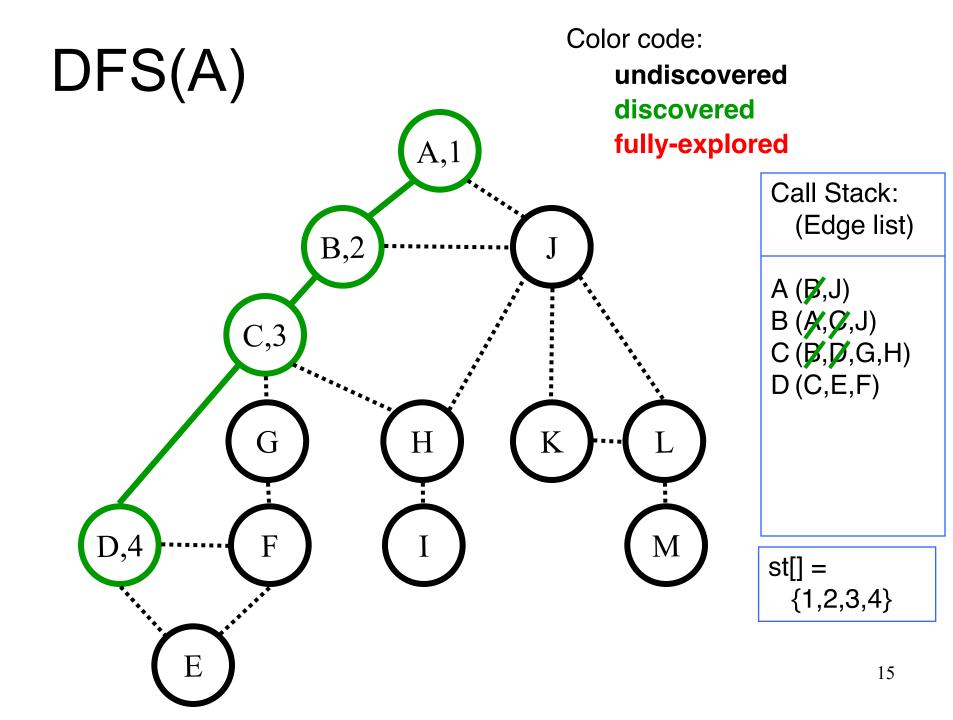
> for each edge {v,x} if (x is undiscovered) Mark x discovered DFS(x)

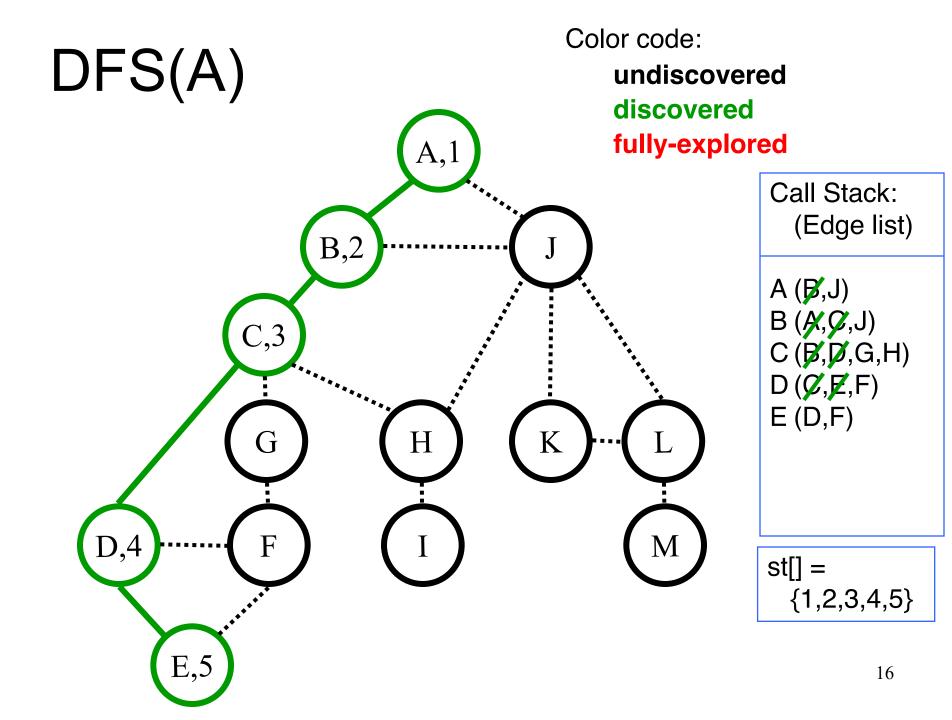
Mark v full-discovered

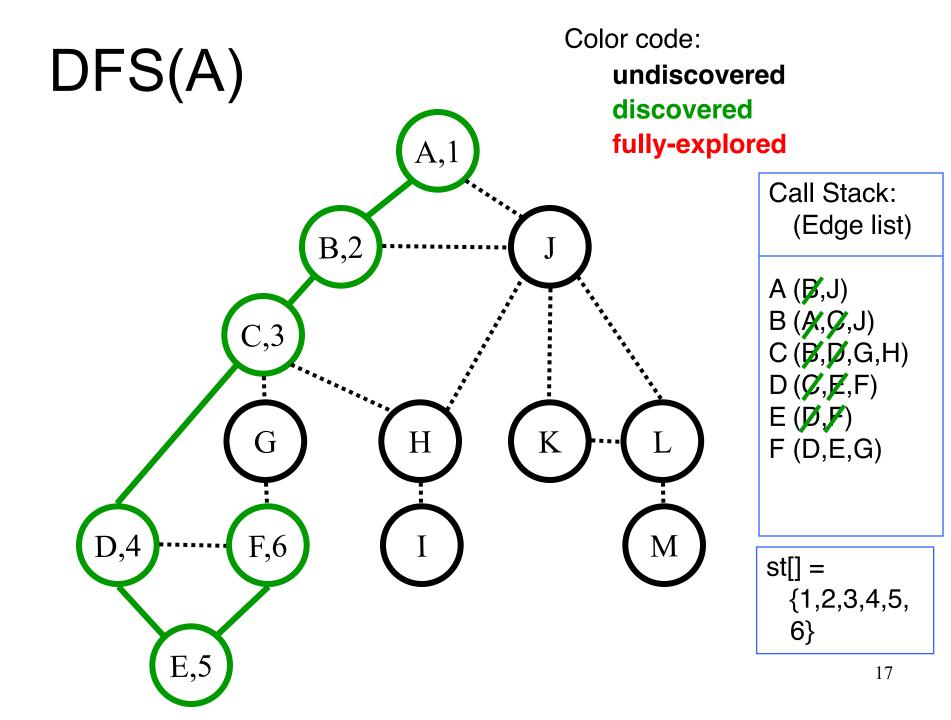


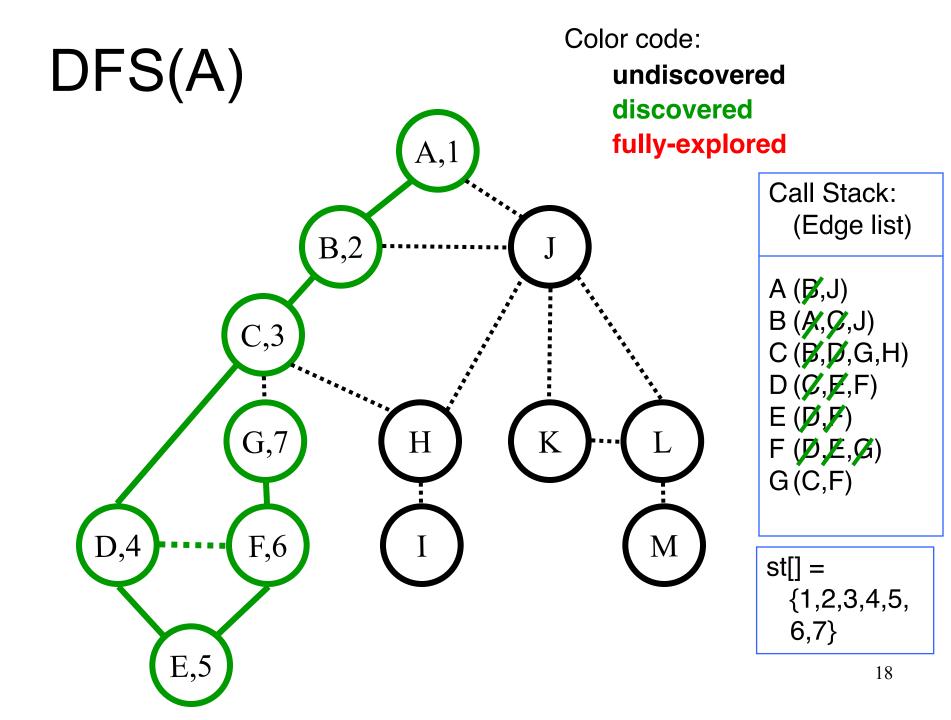


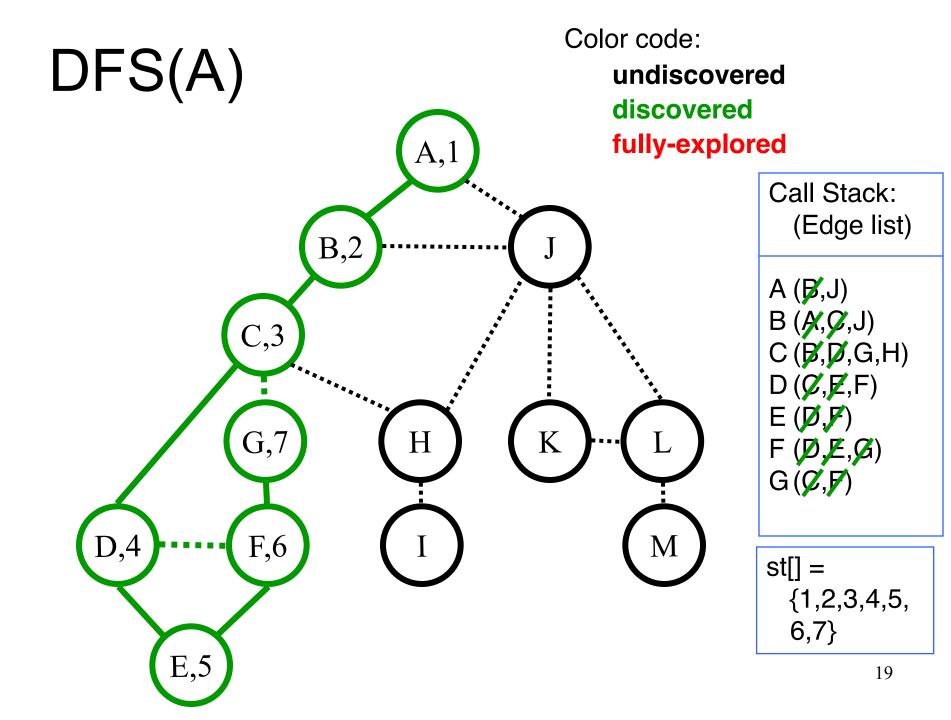


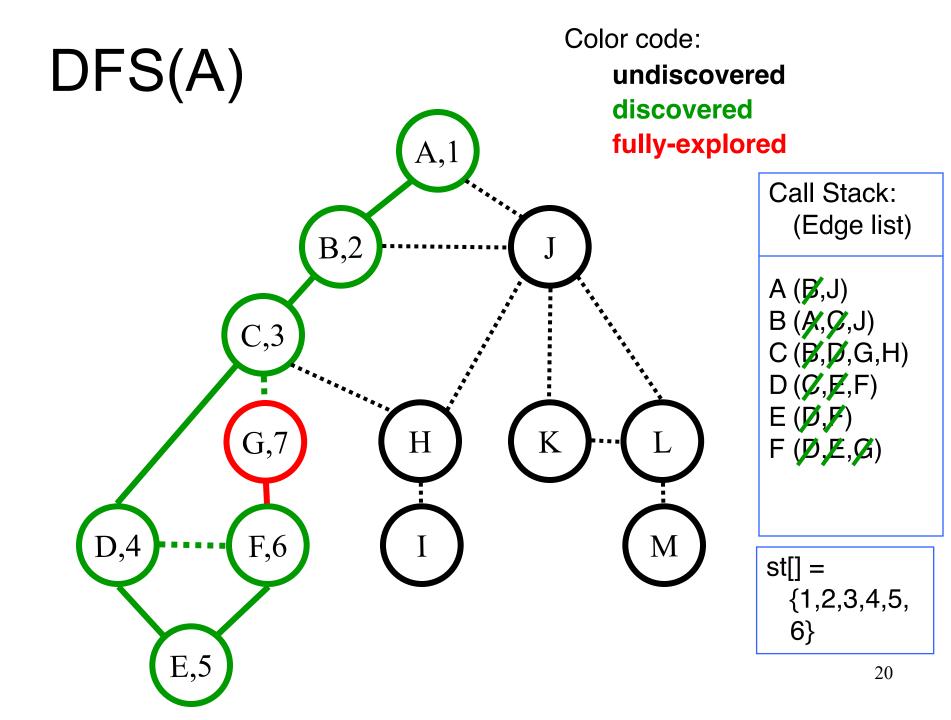


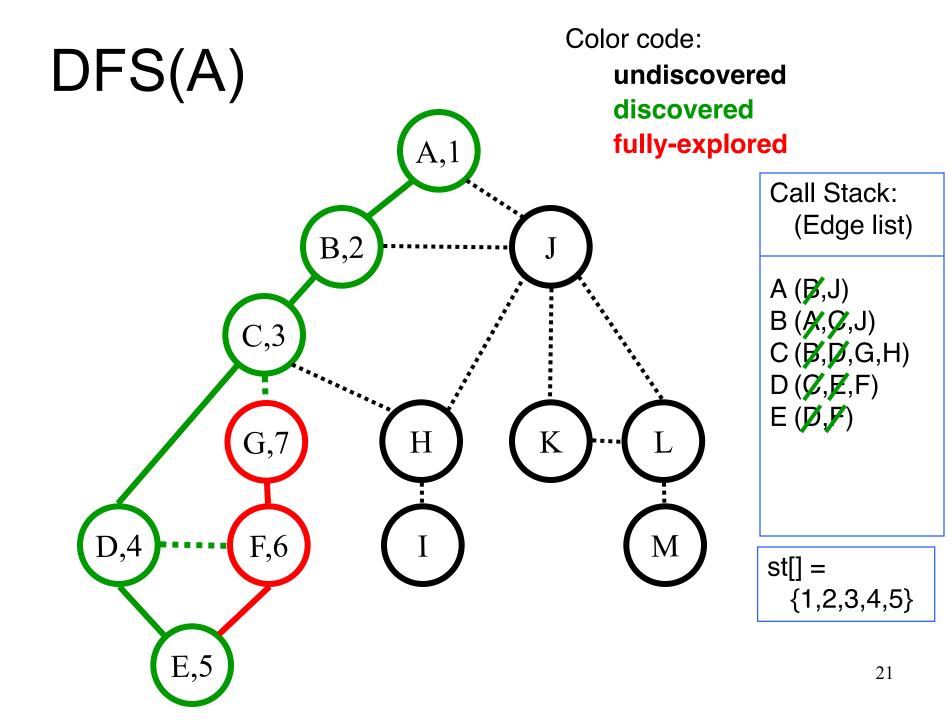


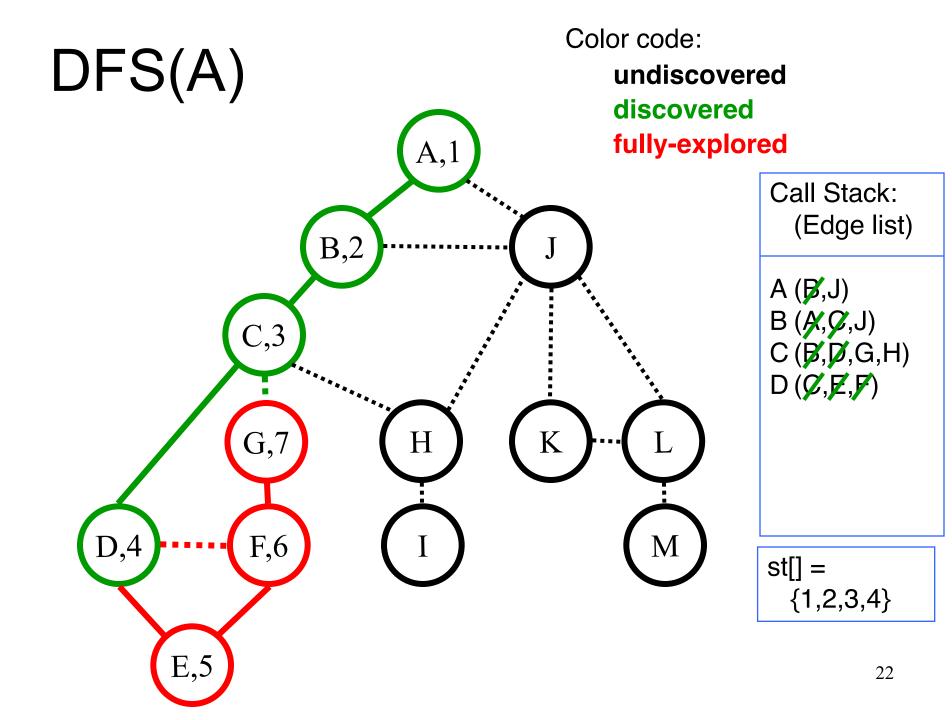


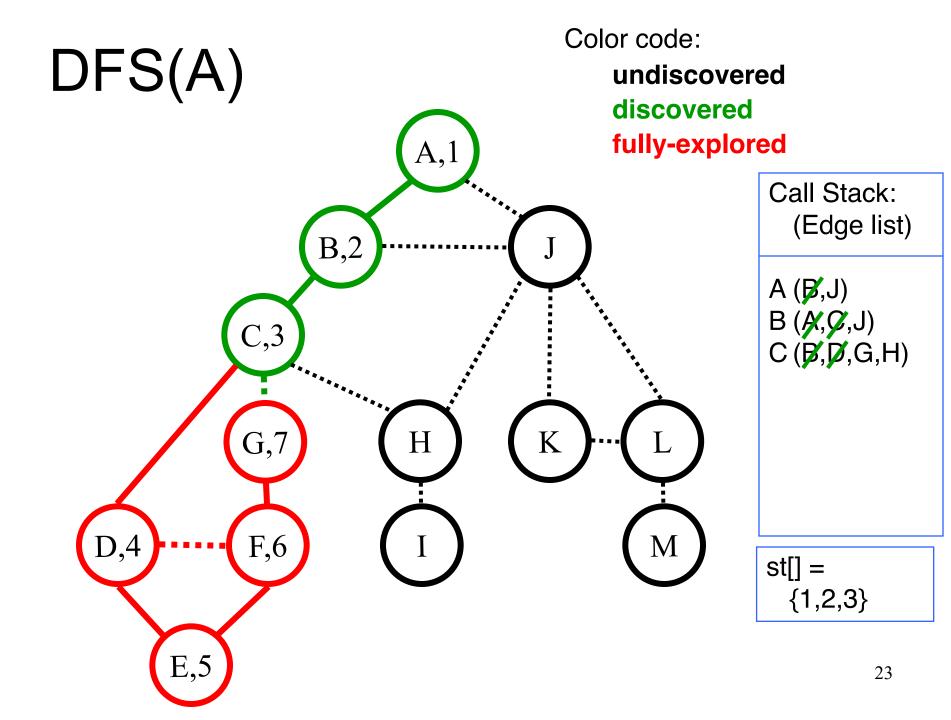


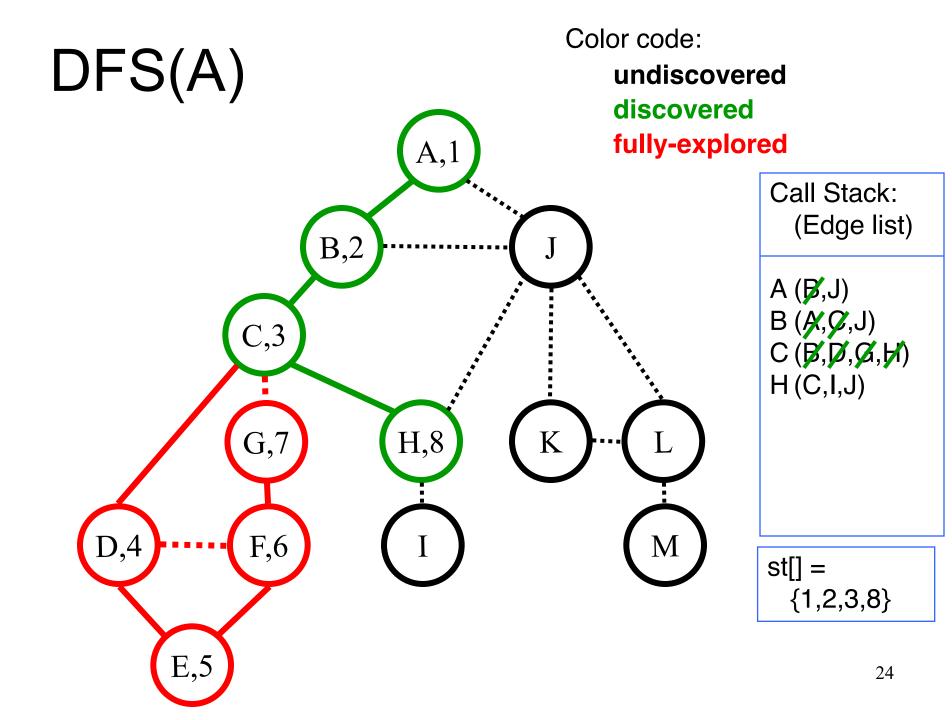


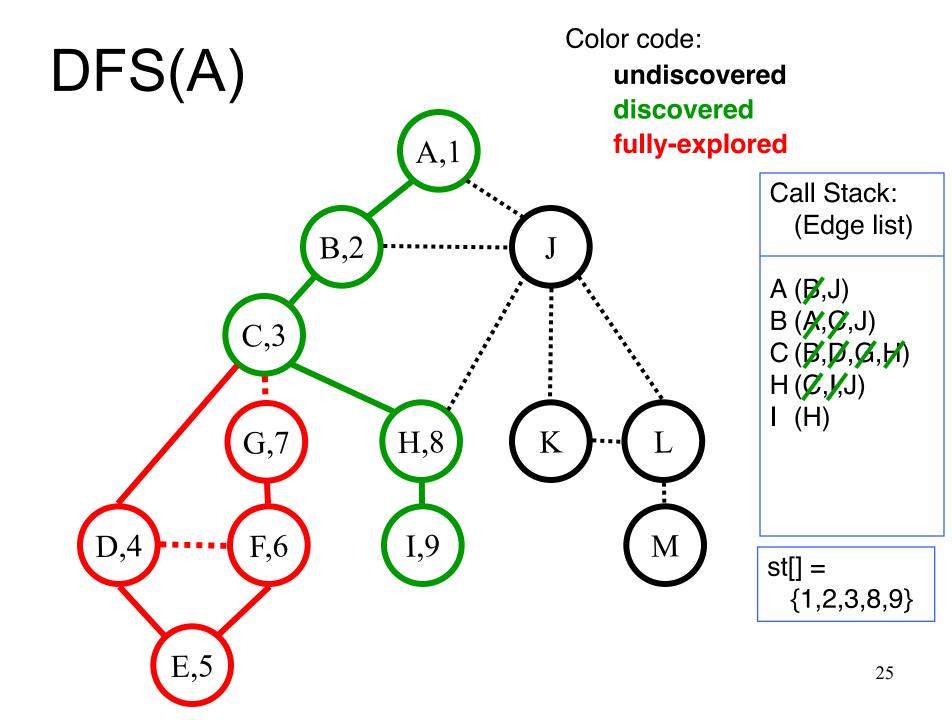


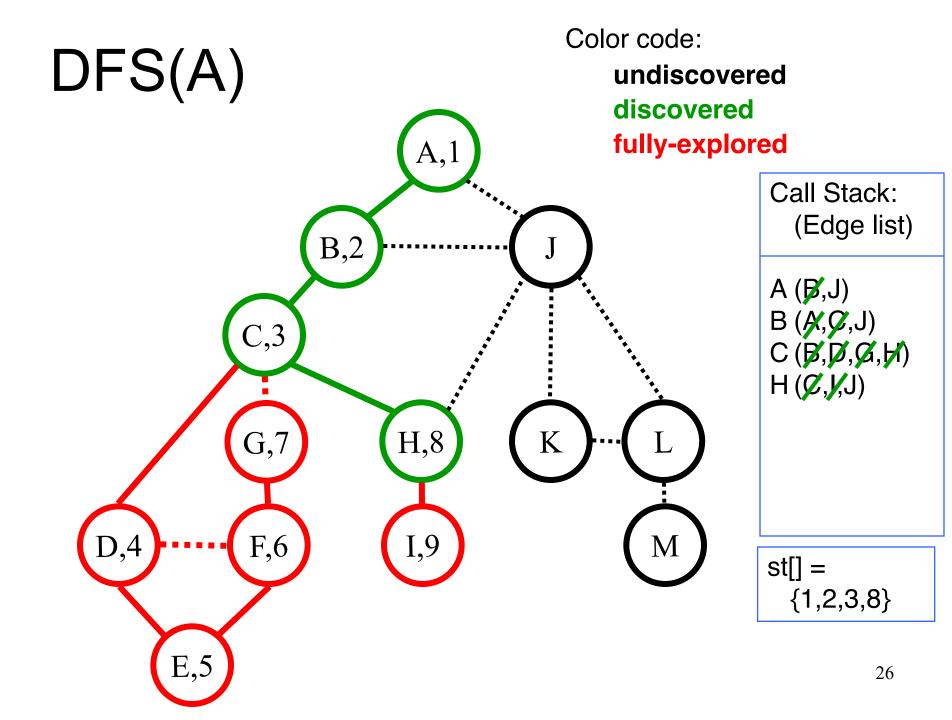


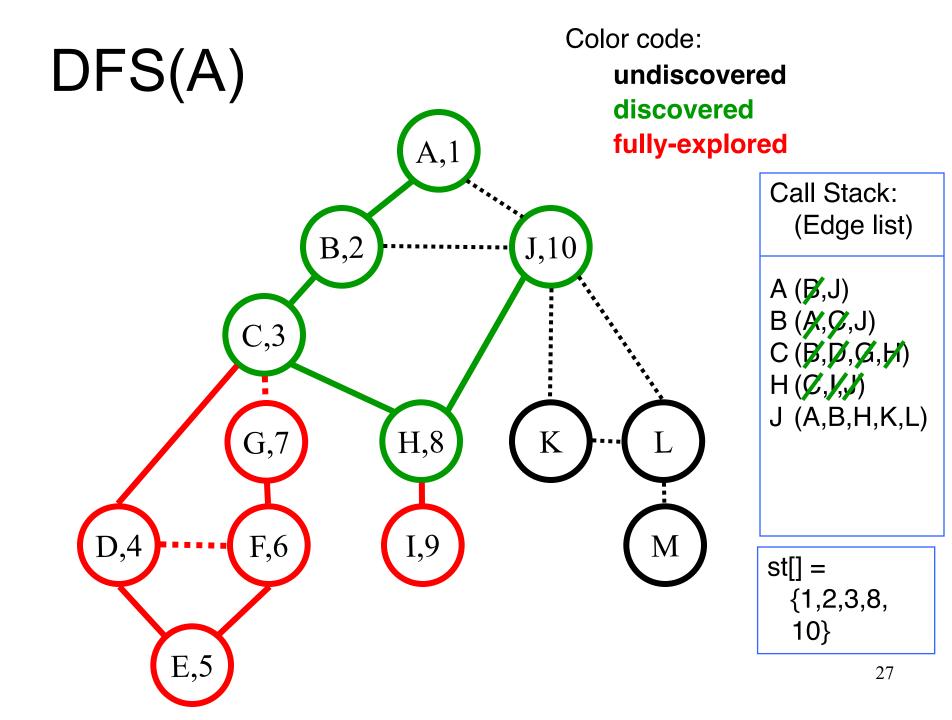


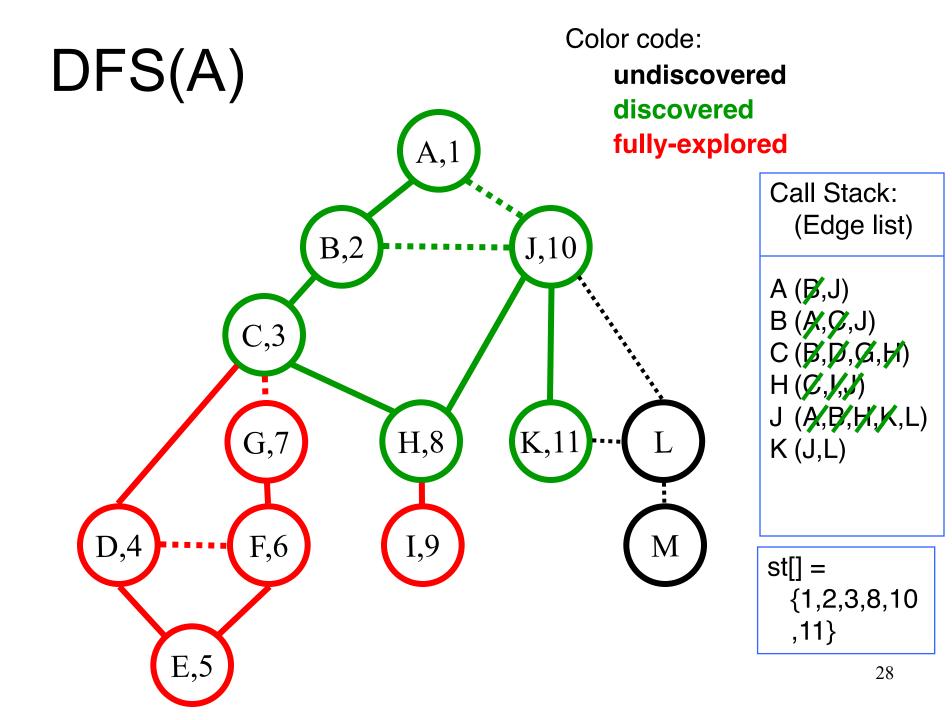


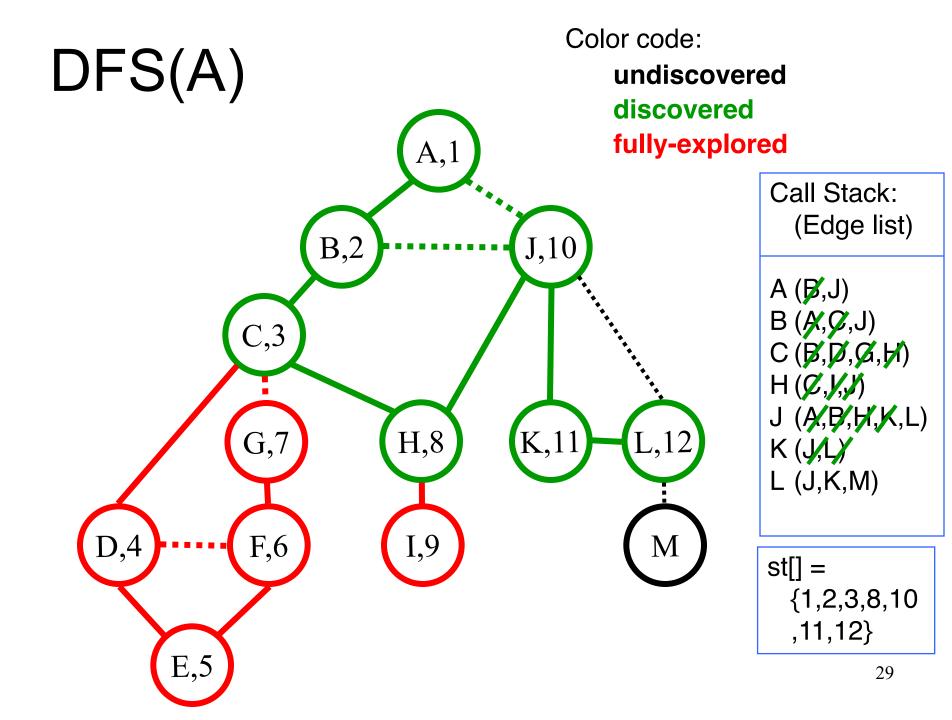


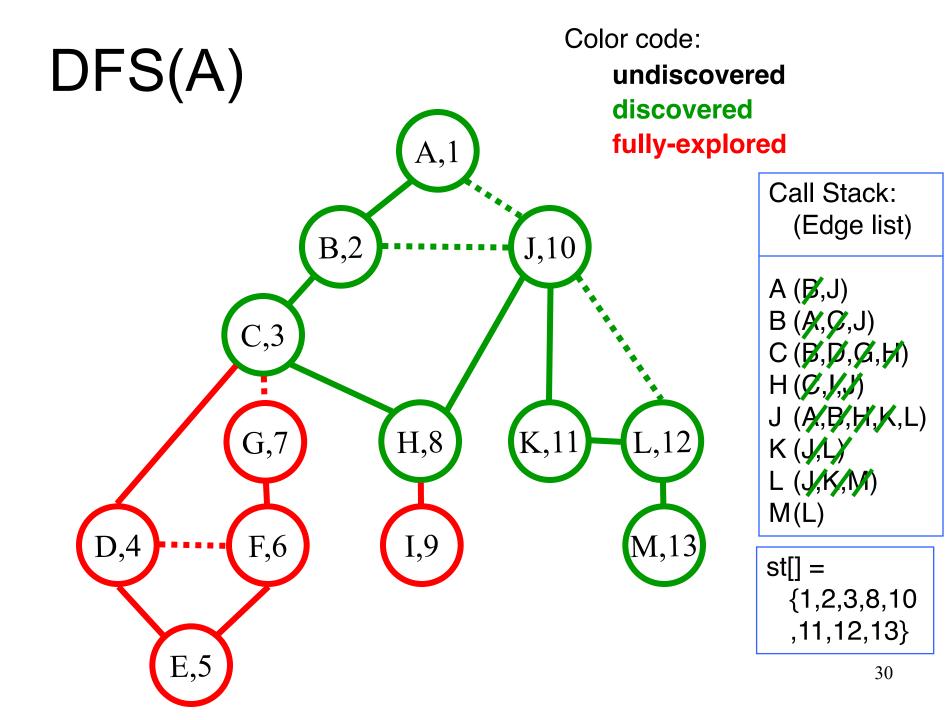


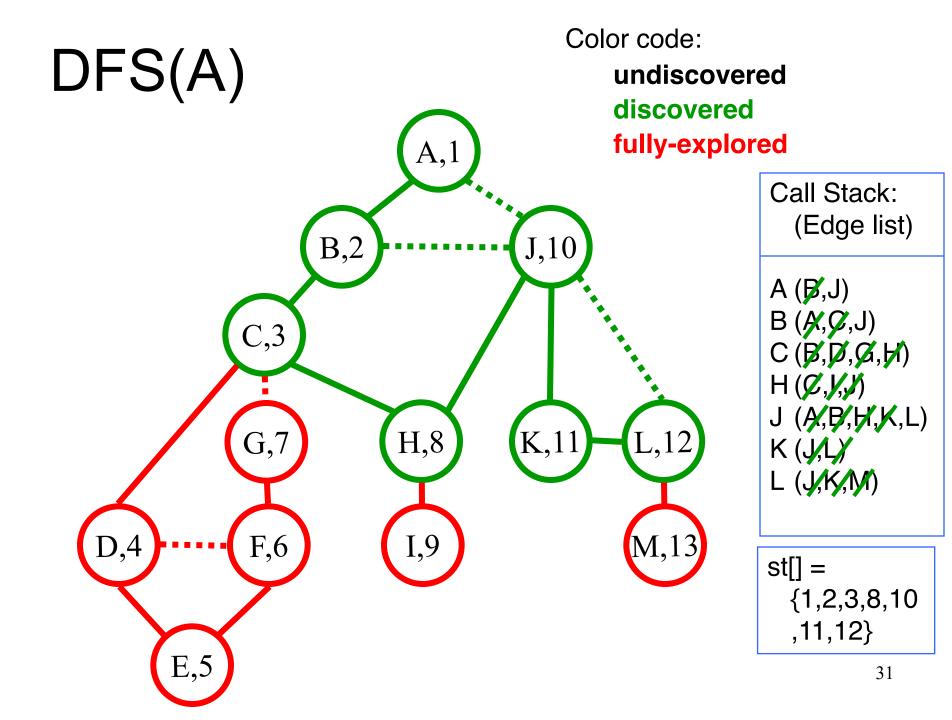


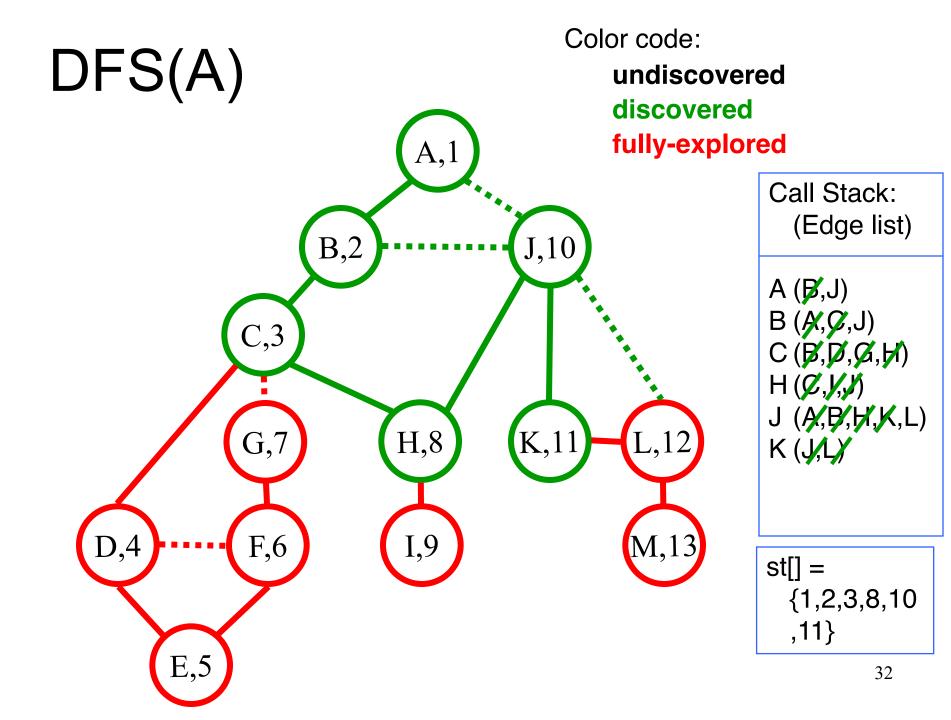


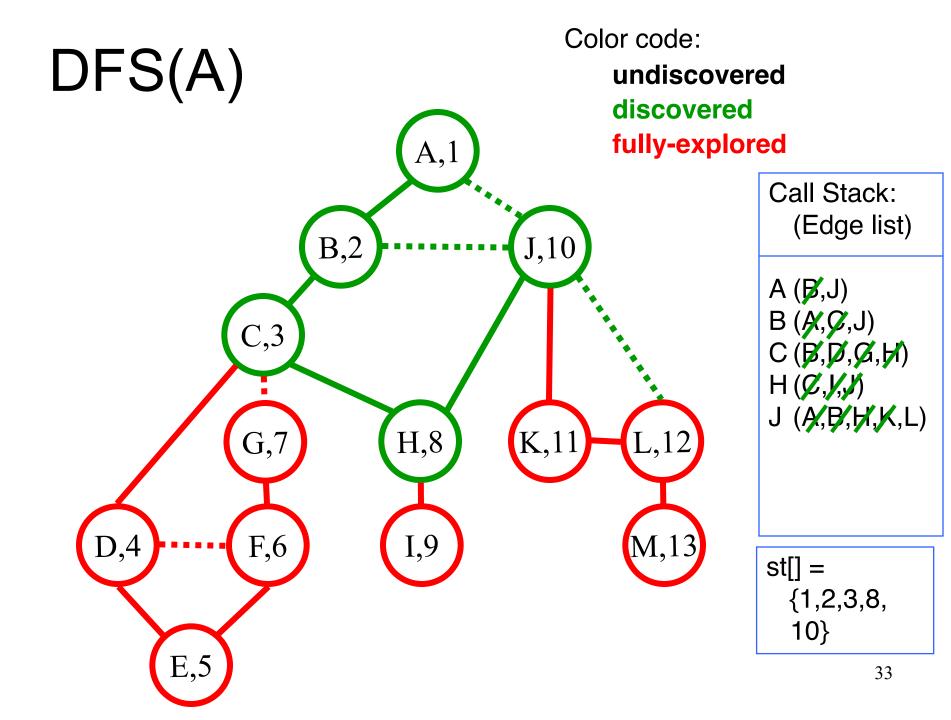


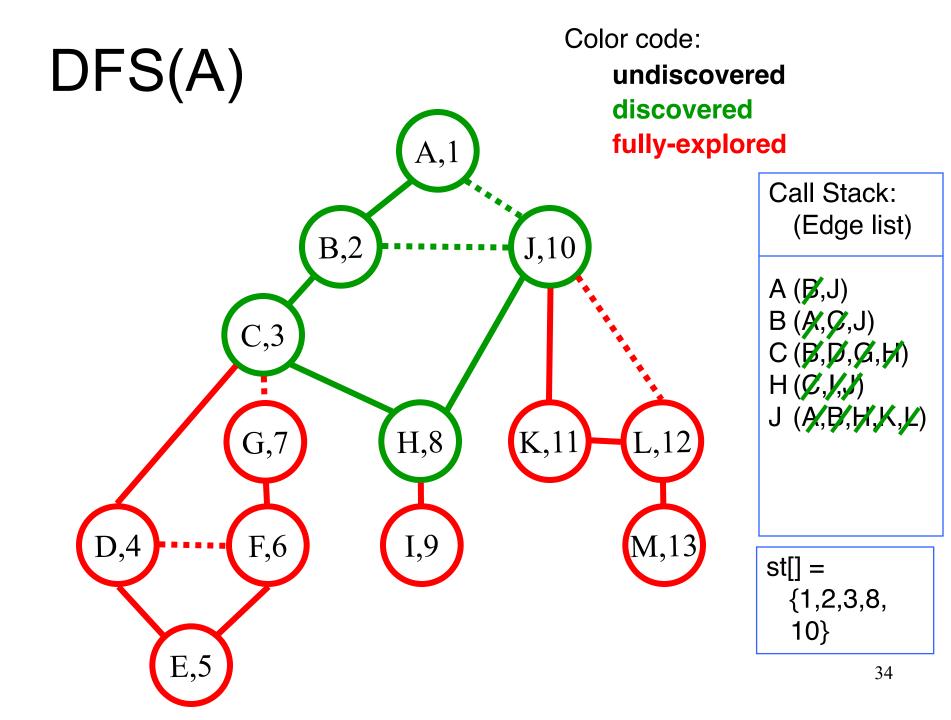


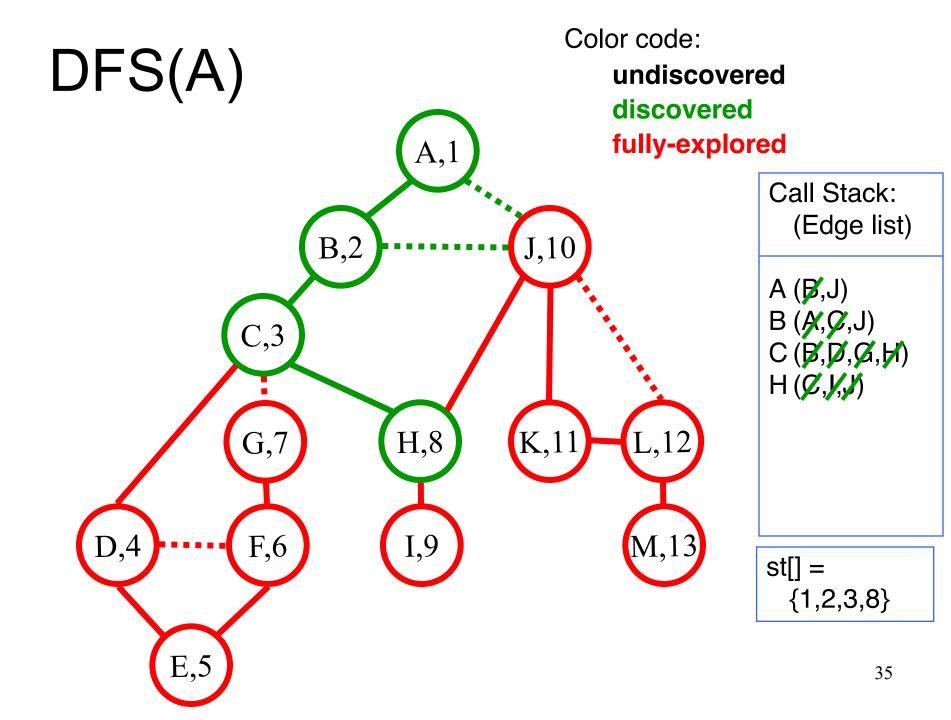


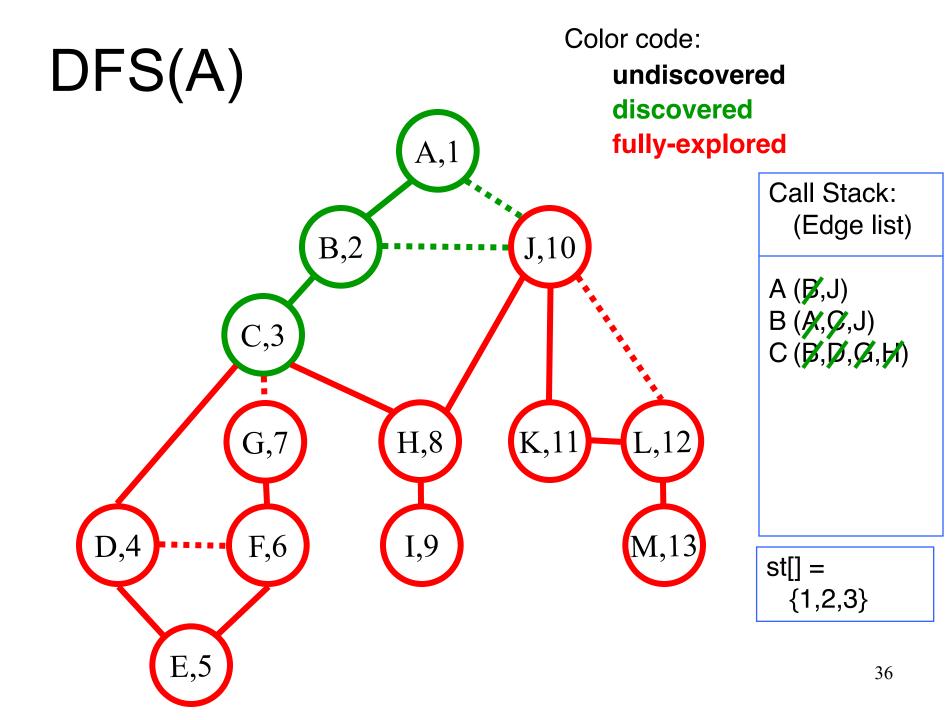


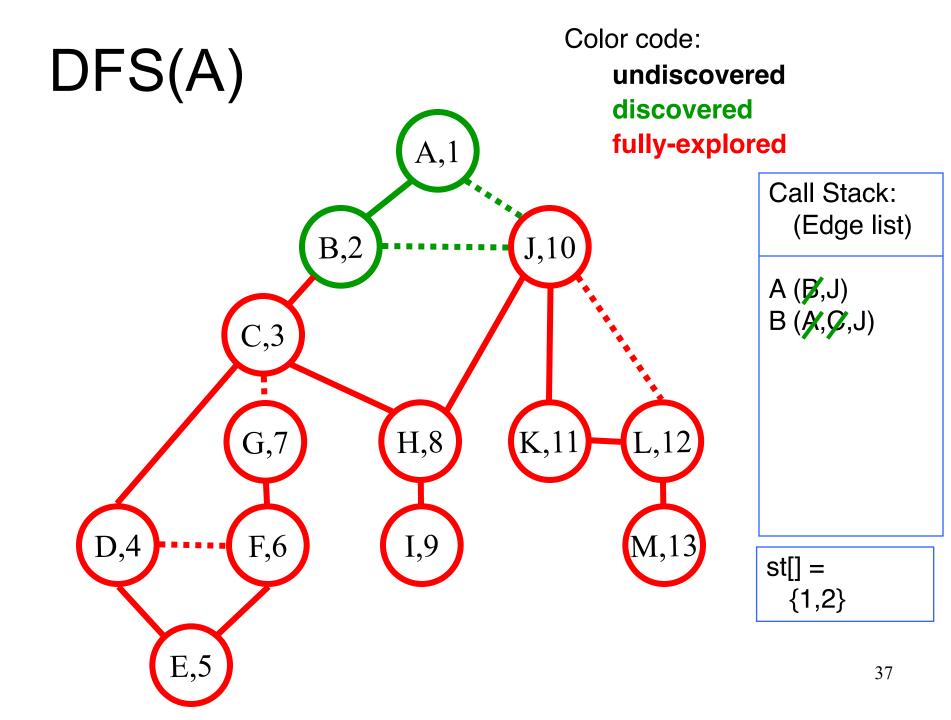


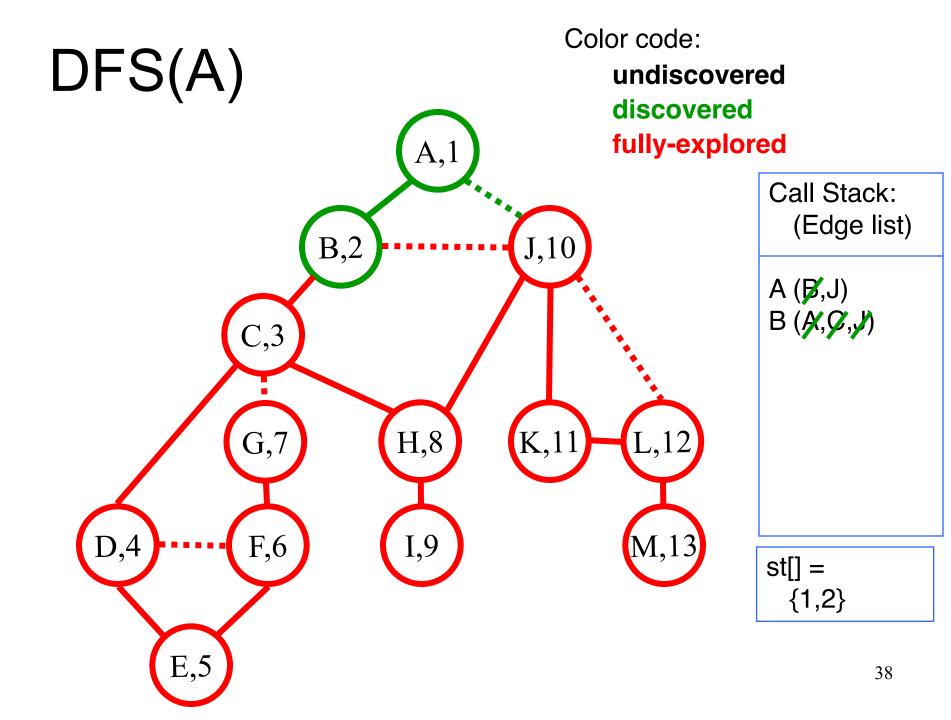


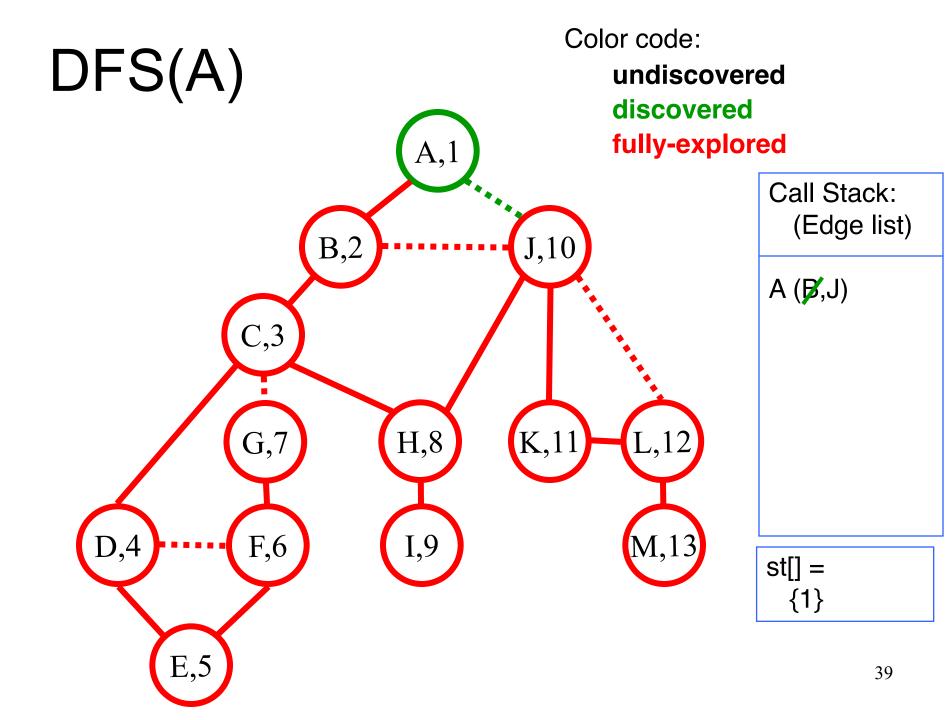


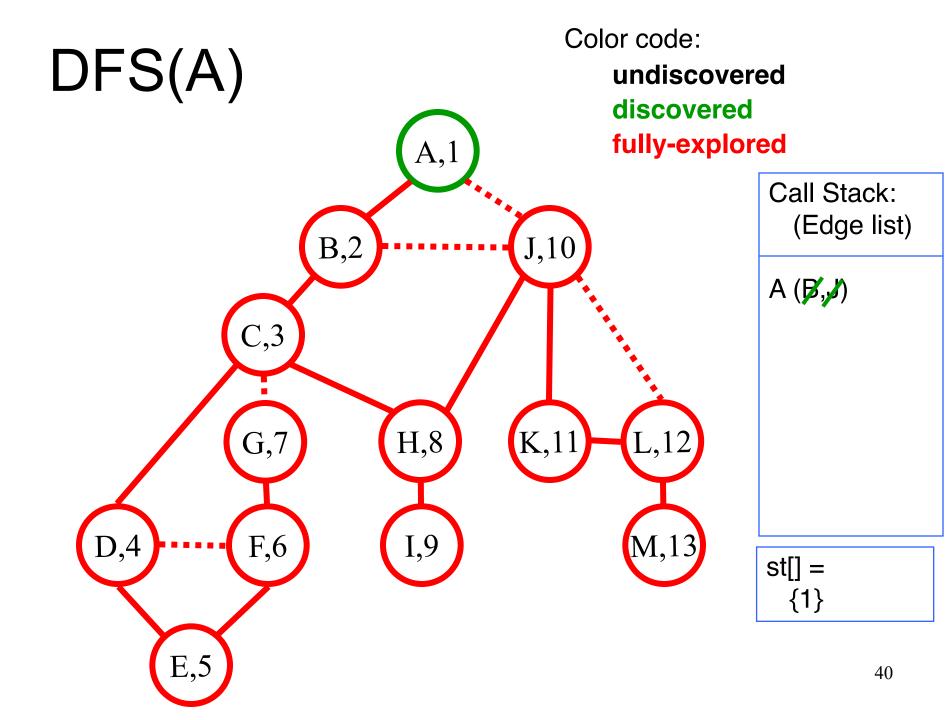


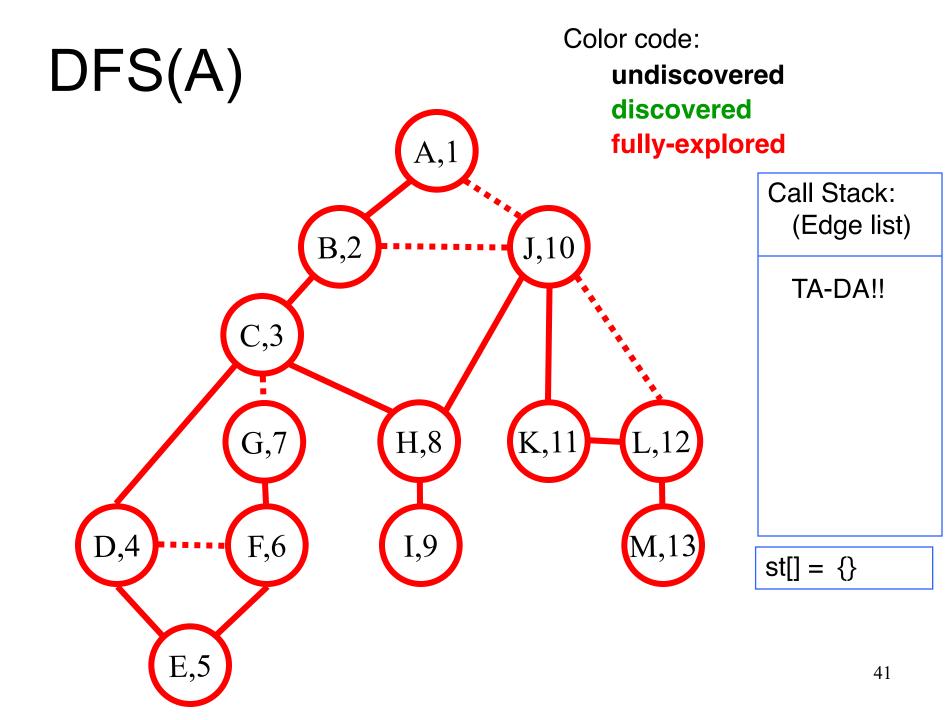


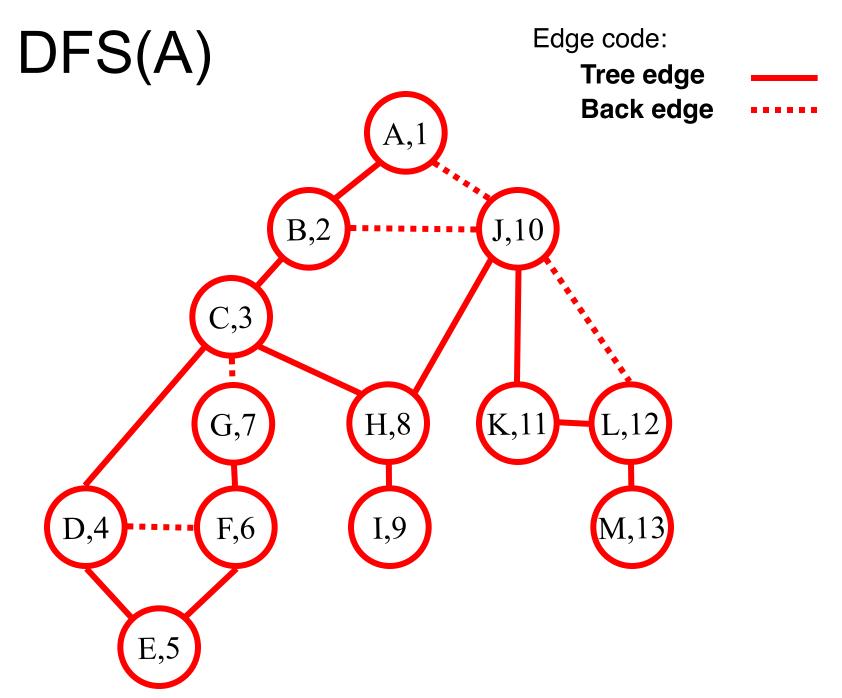


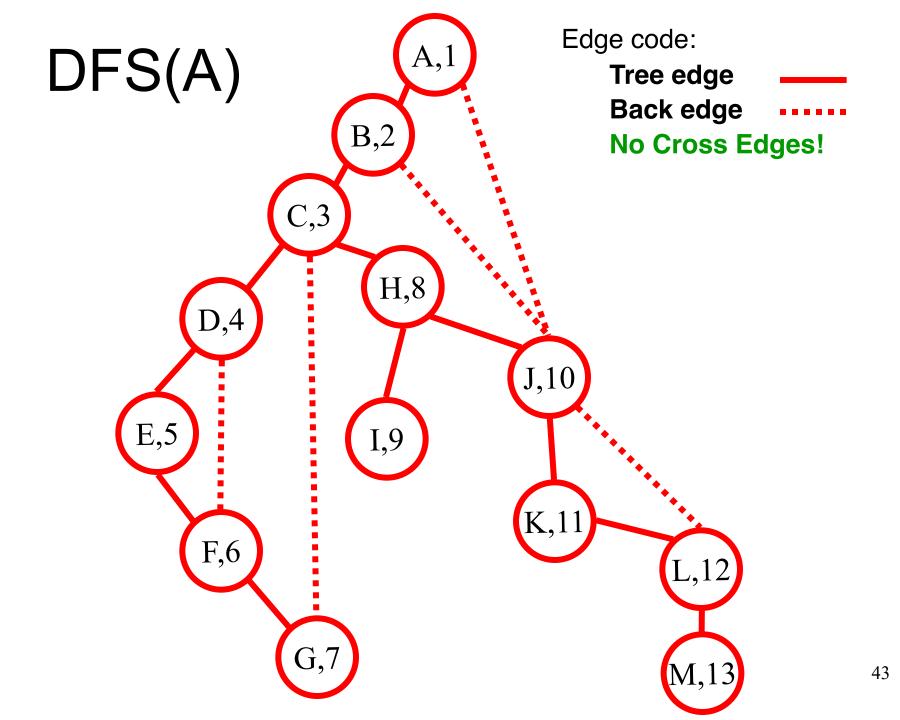












Properties of (undirected) DFS

Like BFS(s):

- DFS(s) visits x iff there is a path in G from s to x
 So, we can use DFS to find connected components
- Edges into then-undiscovered vertices define a tree the "depth first spanning tree" of G

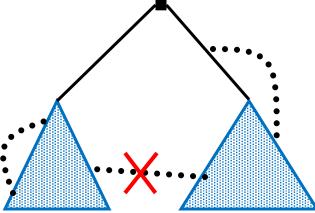
Unlike the BFS tree:

- The DF spanning tree isn't minimum depth
- Its levels don't reflect min distance from the root
- Non-tree edges never join vertices on the same or adjacent levels

Non-Tree Edges in DFS

All non-tree edges join a vertex and one of its descendants/ancestors in the DFS tree

BFS tree \neq DFS tree, but, as with BFS, DFS has found a tree in the graph s.t. non-tree edges are "simple" – only descendant/ancestor



Non-Tree Edges in DFS

Obs: During DFS(x) every vertex marked visited is a descendant of x in the DFS tree

Lemma: For every edge $\{x, y\}$, if $\{x, y\}$ is not in DFS tree, then one of x or y is an ancestor of the other in the tree.

Proof:

One of x or y is visited first, suppose WLOG that x is visited first and therefore DFS(x) was called before DFS(y)

Since $\{x, y\}$ is not in DFS tree, y was visited when the edge $\{x, y\}$ was examined during DFS(x)

Therefore y was visited during the call to DFS(x) so y is a descendant of x.

DAGs and Topological Ordering

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Precedence Constraints

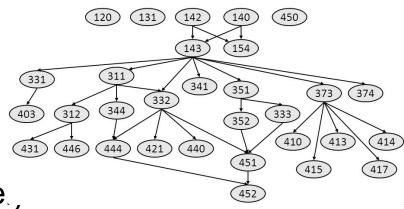
In a directed graph, an edge (i, j) means task *i* must occur before task *j*.

Applications

- Course prerequisite:
 course *i* must be taken before *j*
- Compilation:

must compile module *i* before.

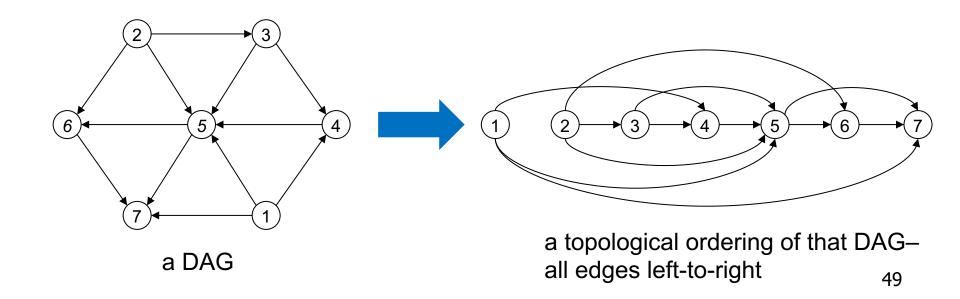
- Computing overflow:
 output of job *i* is part of input to job *j*
- Manufacturing or assembly: sand it before paint it



Directed Acyclic Graphs (DAG)

A DAG is a directed acyclic graph, i.e., one that contains no directed cycles.

Def: A topological order of a directed graph G = (V, E) is an ordering of its nodes as $v_1, v_2, ..., v_n$ so that for every edge (v_i, v_j) we have i < j.



DAGs: A Sufficient Condition

Lemma: If G has a topological order, then G is a DAG.

Pf. (by contradiction)

Suppose that G has a topological order 1, 2, ..., n and that G also has a directed cycle C.

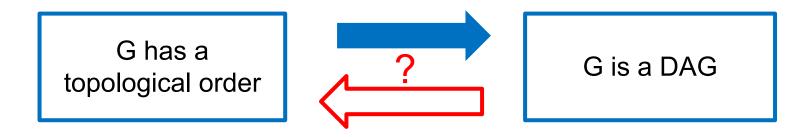
Let *i* be the lowest-indexed node in C, and let *j* be the node just before *i*; thus (j,i) is an (directed) edge.

By our choice of *i*, we have i < j.

On the other hand, since (j, i) is an edge and 1, ..., n is a topological order, we must have j < i, a contradiction

the directed cycle C

DAGs: A Sufficient Condition



Every DAG has a source node

Lemma: If G is a DAG, then G has a node with no incoming edges (i.e., a source).

Pf. (by contradiction)

Suppose that G is a DAG and and it has no source

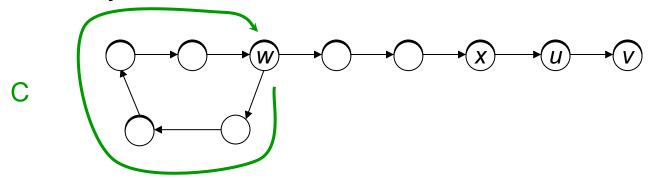
Pick any node v, and begin following edges backward from v. Since v has at least one incoming edge (u, v) we can walk backward to u.

Then, since u has at least one incoming edge (x, u), we can walk backward to x.

Repeat until we visit a node, say w, twice.

Is this similar to a previous proof?

Let C be the sequence of nodes encountered between successive visits to w. C is a cycle.



DAG => Topological Order

Lemma: If G is a DAG, then G has a topological order

```
Pf. (by induction on n)
Base case: true if n = 1.
```

IH: Every DAG with n-1 vertices has a topological ordering.

IS: Given DAG with n > 1 nodes, find a source node v.

 $G - \{v\}$ is a DAG, since deleting v cannot create cycles.

Reminder: Always remove vertices/edges to use IH

By IH, $G - \{v\}$ has a topological ordering.

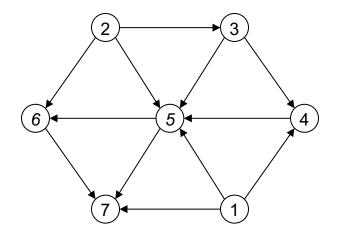
Place v first in topological ordering; then append nodes of G - { v }

in topological order. This is valid since v has no incoming edges.

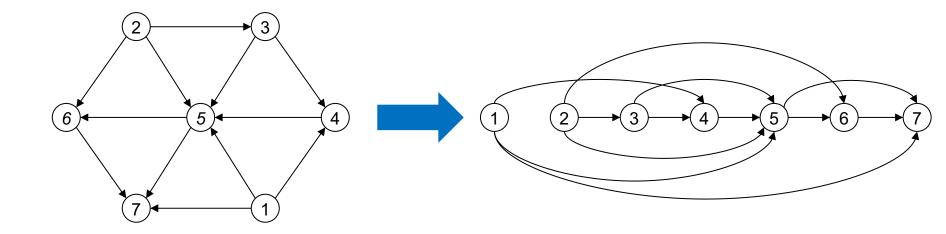
A Characterization of DAGs

G has a topological order

Topological Order Algorithm: Example



Topological Order Algorithm: Example



Topological order: 1, 2, 3, 4, 5, 6, 7

Topological Sorting Algorithm

Maintain the following:

count[w] = (remaining) number of incoming edges to node w

S = set of (remaining) nodes with no incoming edges

Initialization:

```
count[w] = 0 for all w
```

count[w]++ for all edges (v,w)

```
S = S \cup {w} for all w with count[w]=0
```

Main loop:

while S not empty

- remove some v from S
- make v next in topo order
- for all edges from v to some w
 - -decrement count[w]
 - -add w to S if count[w] hits 0

Correctness: clear, I hope

Time: O(m + n) (assuming edge-list representation of graph)

O(1) per node O(1) per edge

O(m + n)

Summary

- Graphs: abstract relationships among pairs of objects
- Terminology: node/vertex/vertices, edges, paths, multiedges, self-loops, connected
- Representation: Adjacency list, adjacency matrix
- Nodes vs Edges: $m = O(n^2)$, often less
- BFS: Layers, queue, shortest paths, all edges go to same or adjacent layer
- DFS: recursion/stack; all edges ancestor/descendant
- Algorithms: Connected Comp, bipartiteness, topological sort