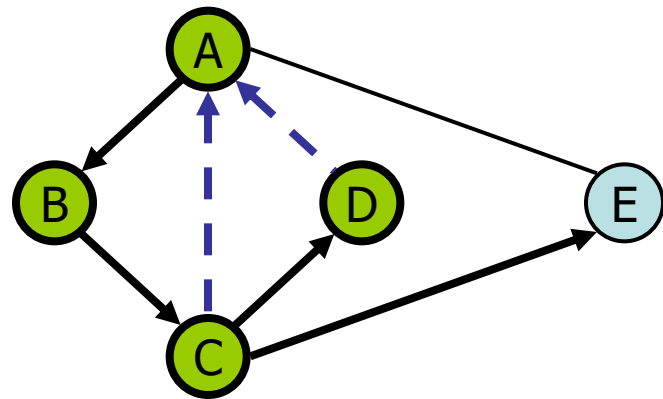
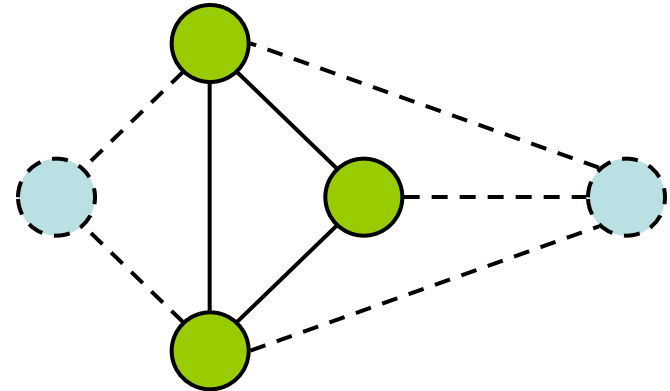


Graph Traversals

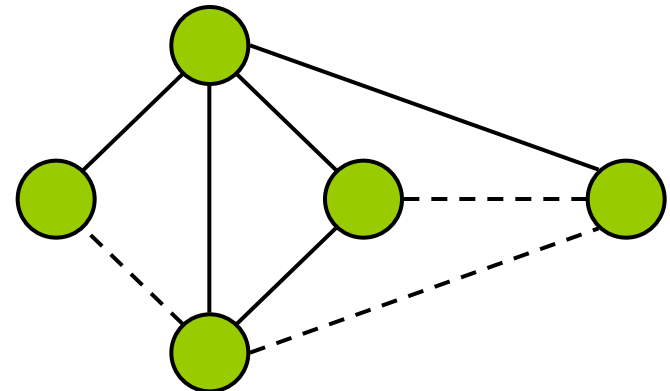


Subgraphs

- A subgraph S of a graph G is a graph such that
 - The vertices of S are a subset of the vertices of G
 - The edges of S are a subset of the edges of G
- A spanning subgraph of G is a subgraph that contains all the vertices of G



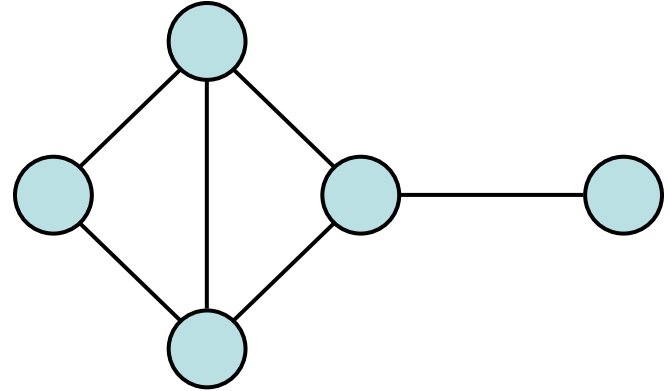
Subgraph



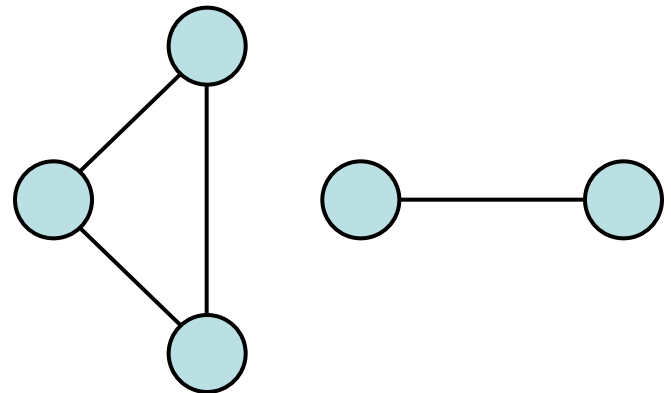
Spanning subgraph

Connectivity

- A graph is connected if there is a path between every pair of vertices
- A connected component of a graph G is a maximal connected subgraph of G



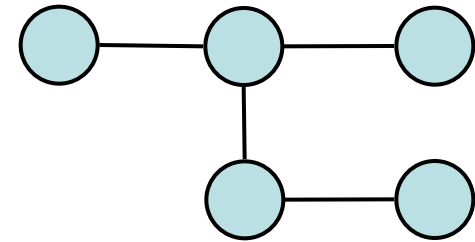
Connected graph



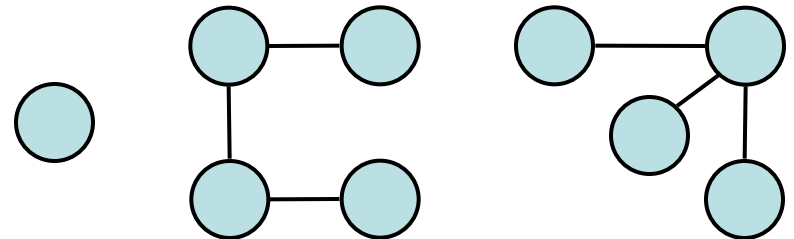
Non connected graph with two connected components

Trees and Forests

- A tree is an undirected graph T such that
 - T is connected
 - T has no cycles
- A forest is an undirected graph without cycles (a collection of trees).
- The connected components of a forest are trees



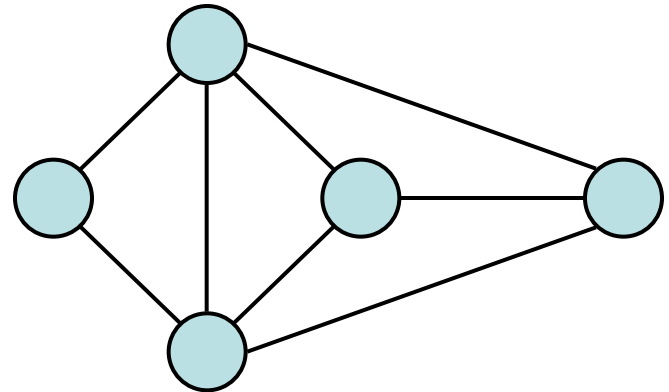
Tree



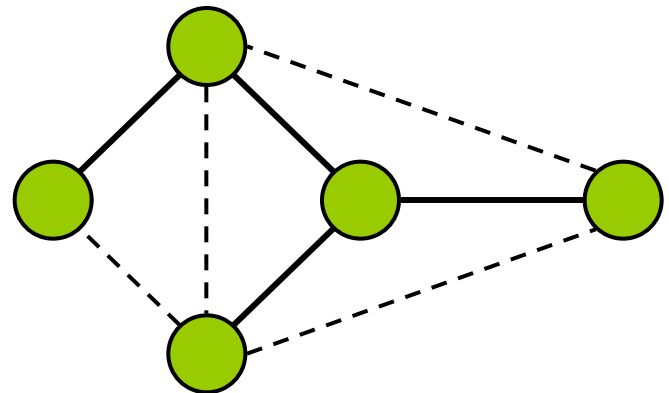
Forest

Spanning Trees and Forests

- A spanning tree of a connected graph is a spanning subgraph that is a tree
- A spanning tree is not unique unless the graph is a tree
- Spanning trees have applications to the design of communication networks
- A spanning forest of a graph is a spanning subgraph that is a forest



Graph



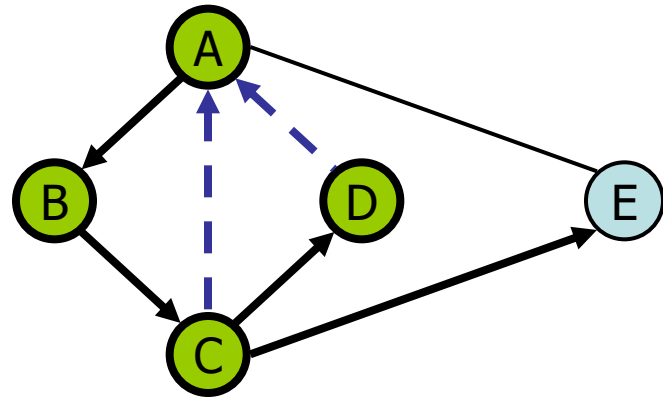
Spanning tree

Graph Traversals

A traversal of a graph G :

- Visits all the vertices and edges of G
- Determines whether G is connected
- Computes the connected components of G
- Computes a spanning forest of G
- Build a spanning tree in a connected graph

Depth-First Search



Outline and Reading

- Depth-first search
 - With a Stack
 - Recursive
 - Examples
 - Properties
 - Complexity
- Applications of DFS
 - Path finding
 - Cycle finding

Depth-First Search

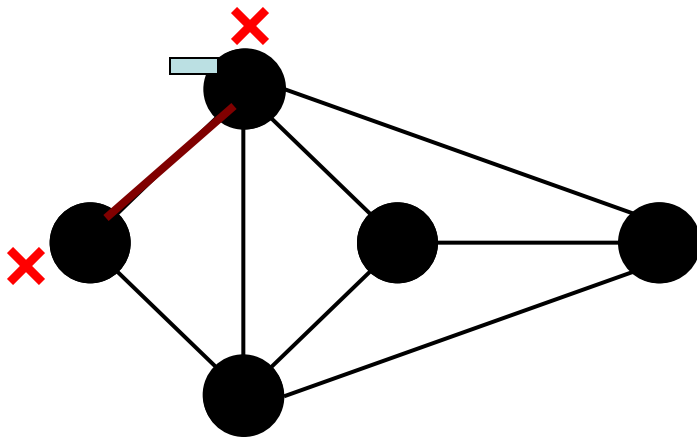
Depth-First Search is a graph traversal technique that:

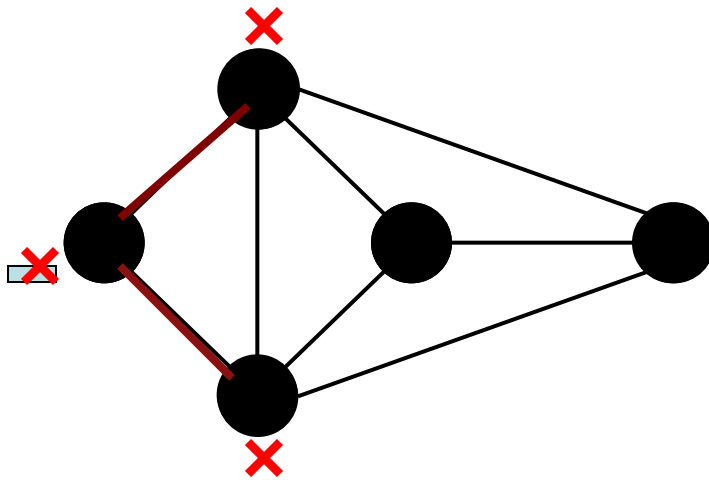
- on a graph with n vertices and m edges takes $O(n + m)$ time (which is $O(m)$)
- can be further extended to solve other graph problems
 - Find and report a path between two given vertices
 - Find a cycle in the graph

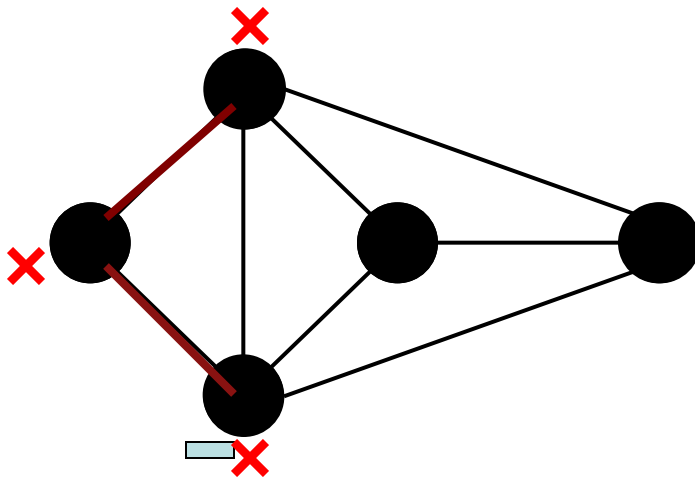
The idea:

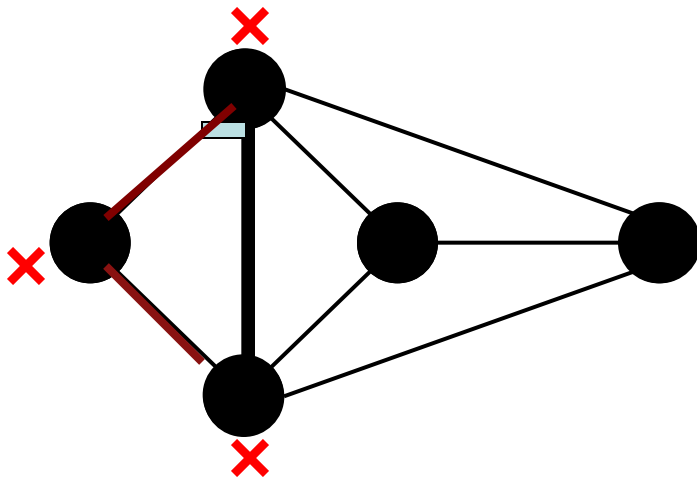
Starting at an arbitrary vertex, follow along a simple path until you get to a vertex which has no unvisited adjacent vertices.

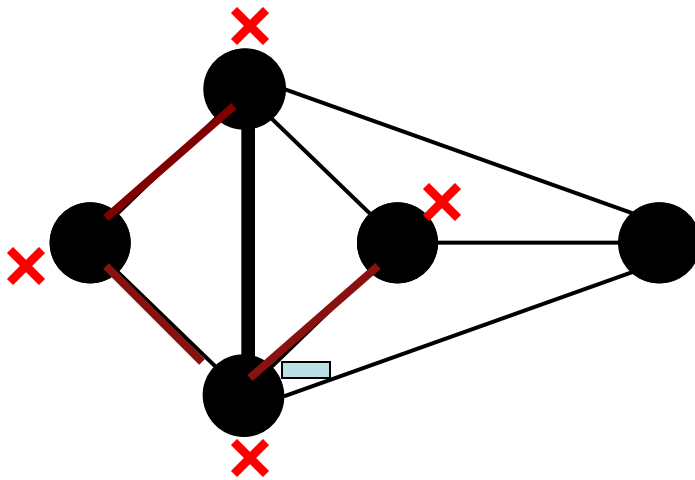
Then start tracing back up the path, one vertex at a time, to find a vertex with unvisited adjacent vertices.

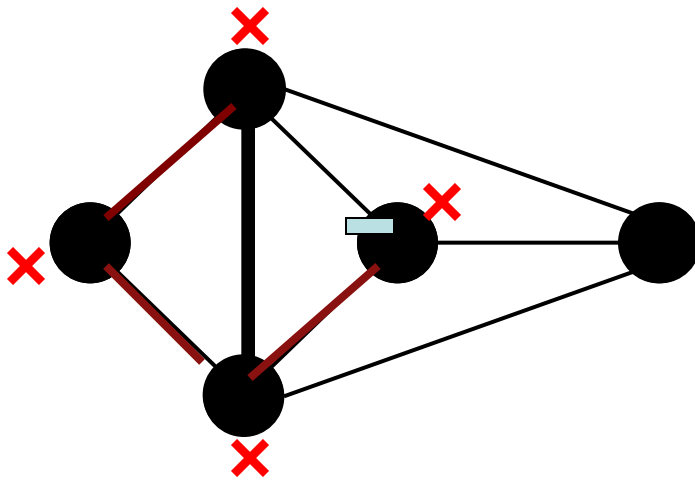


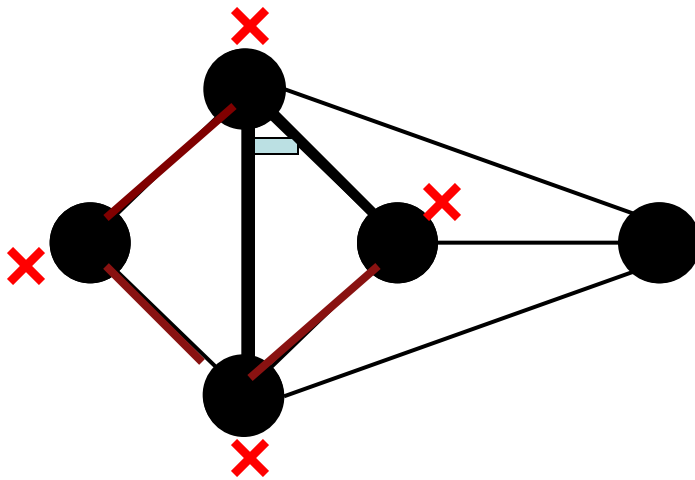


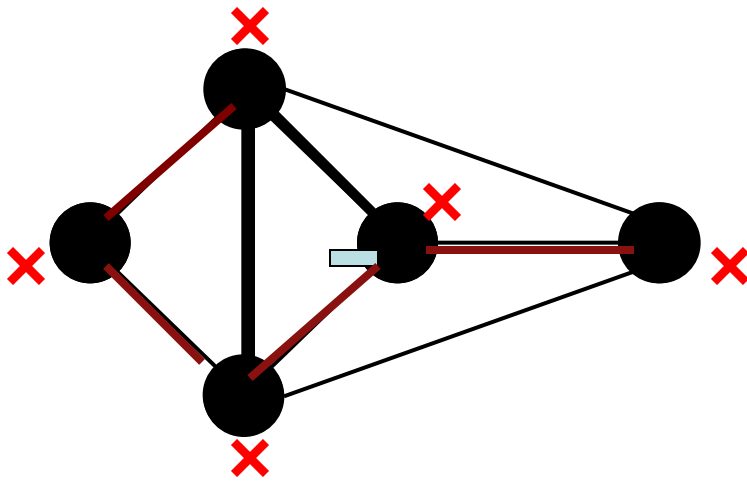


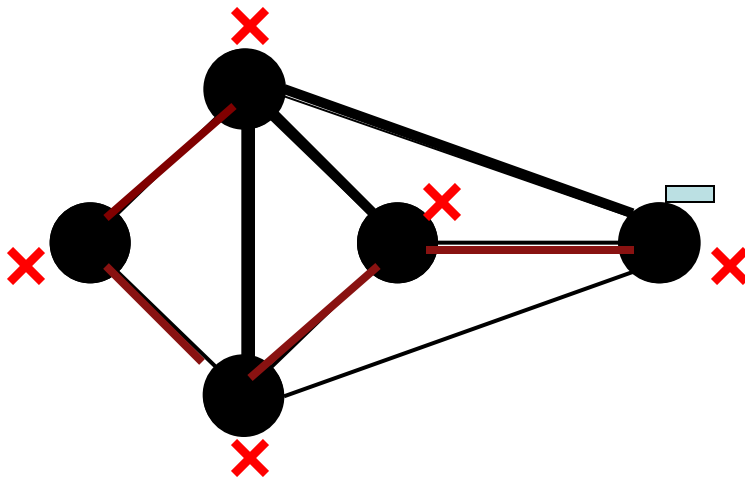


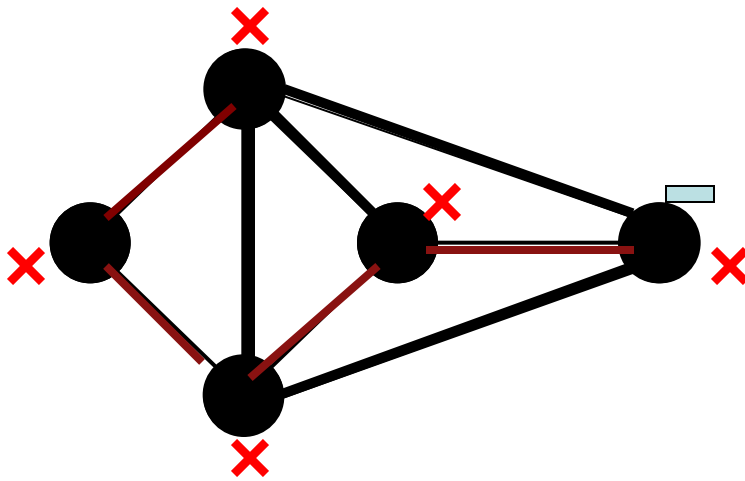


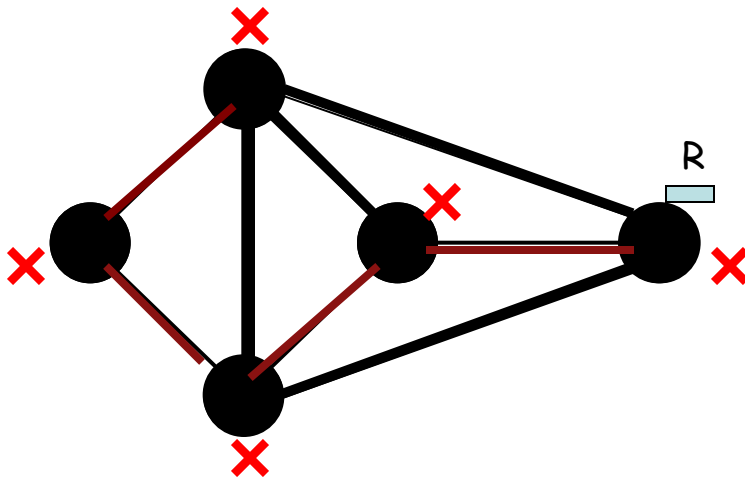


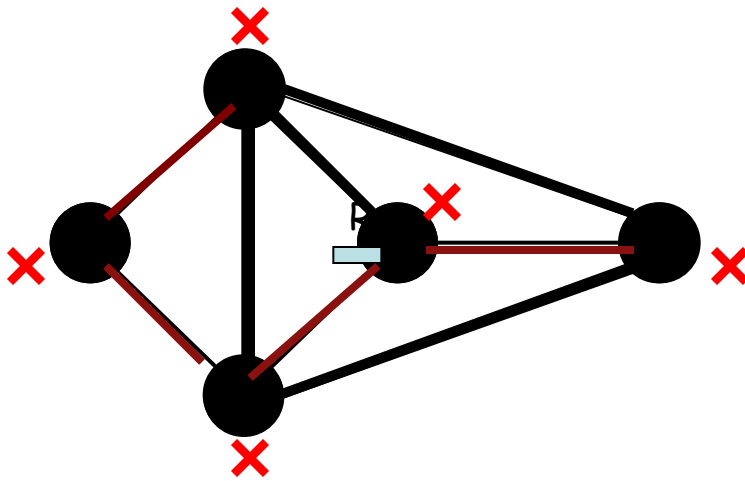




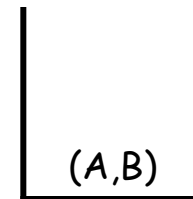
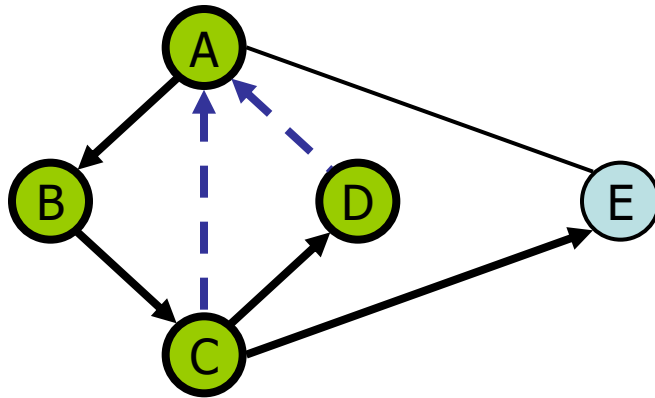


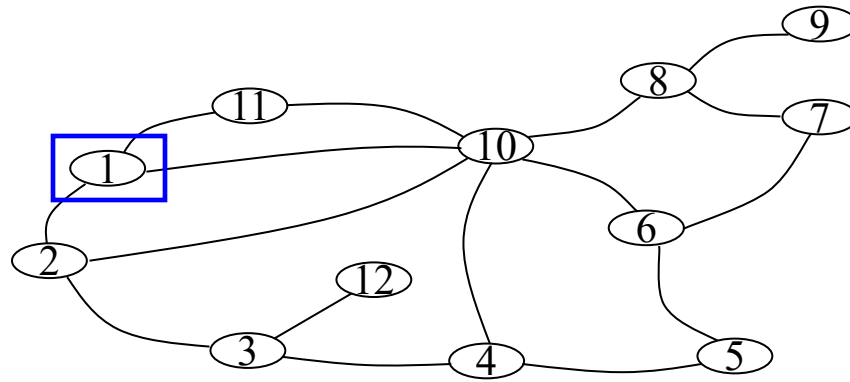




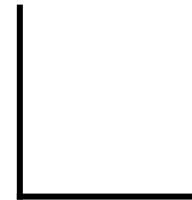


DFS Algorithm - With a Stack

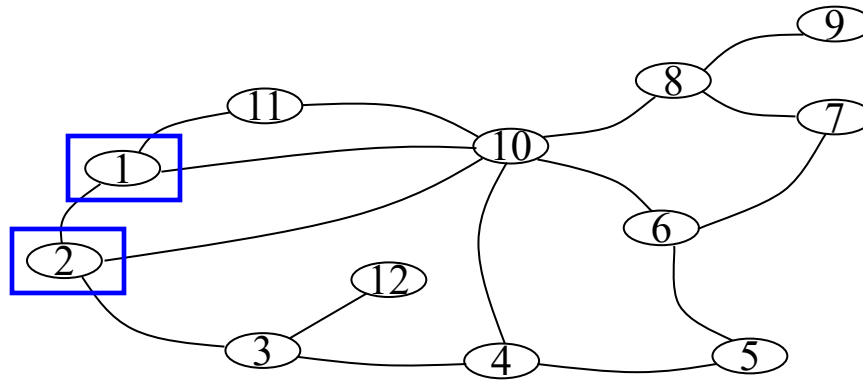




Visited : {1}



to visit



Visited : {1}

POP: (1,2)

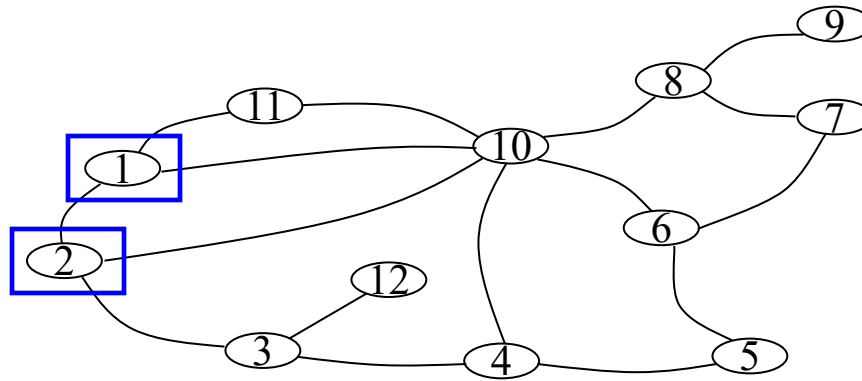
Visited : {1,2}

T={{(1,2)}}

PUSH

(1,2)
(1,11)
(1,10)

to visit



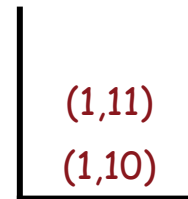
Visited : {1}

POP: (1,2)

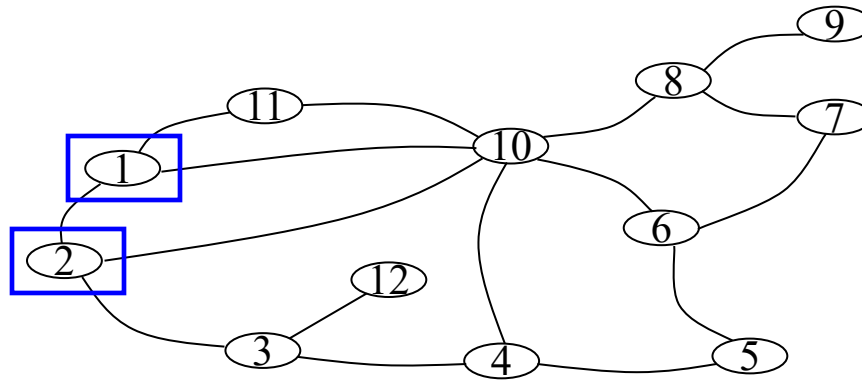
Visited : {1,2}

T={{(1,2)}}

PUSH



to visit



POP: (2,3)

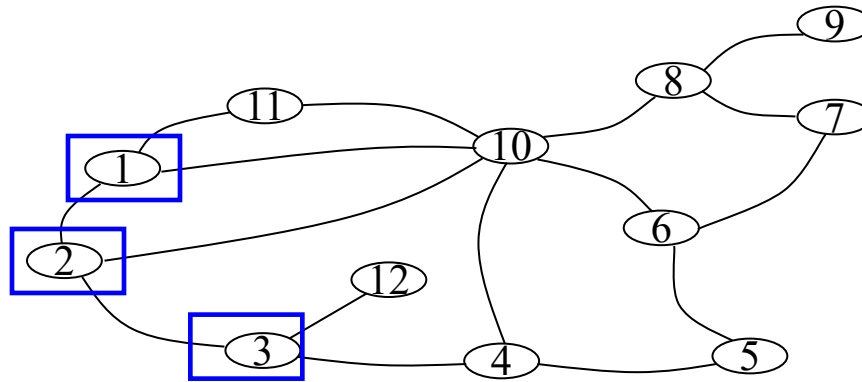
Visited : {1,2,3}

$T = \{(1,2), (2,3)\}$

PUSH

(2,3)
(2,10)
(1,11)
(1,10)

to visit



POP: (2,3)

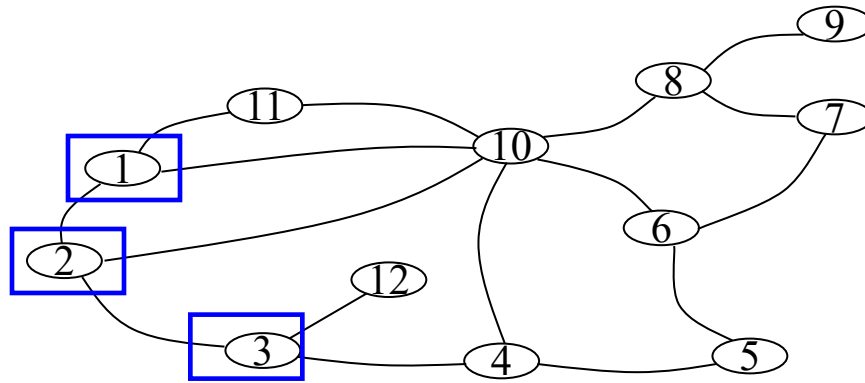
Visited : {1,2,3}

$T = \{(1,2), (2,3)\}$

PUSH

(2,10)
(1,11)
(1,10)

to visit

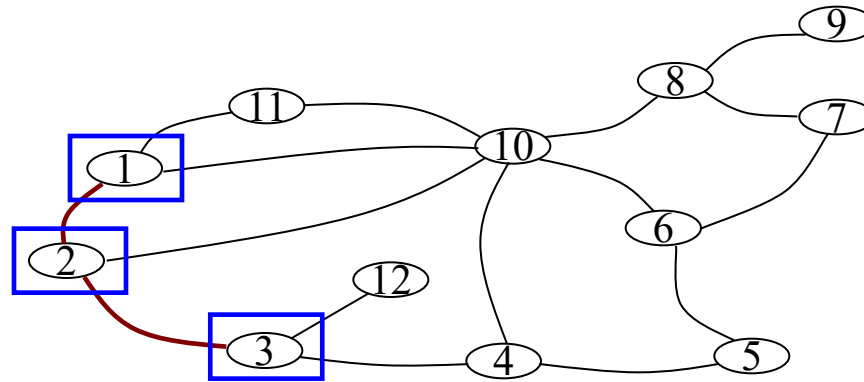


PUSH

(3,12)
(3,4)
(2,10)
(1,11)
(1,10)

to visit

$$T=\{(1,2), (2,3)\}$$



(3,12)

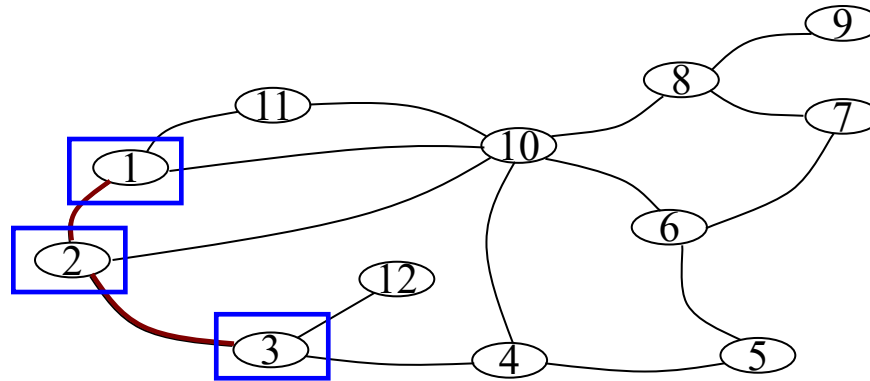
(3,4)

(2,10)

(1,11)

(1,10)

(3,12)
(3,4)
(2,10)
(1,11)
(1,10)



POP \rightarrow (3,12)

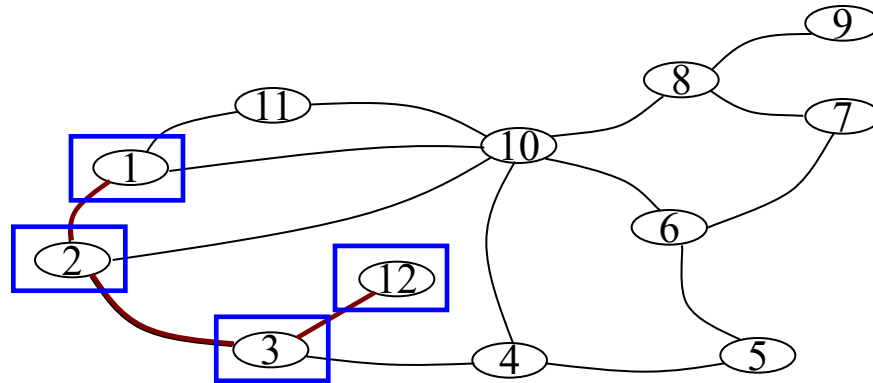
Visited : {1,2,3,12}

$T = \{(1,2), (2,3), (3,12)\}$

PUSH
(nothing
in this case)

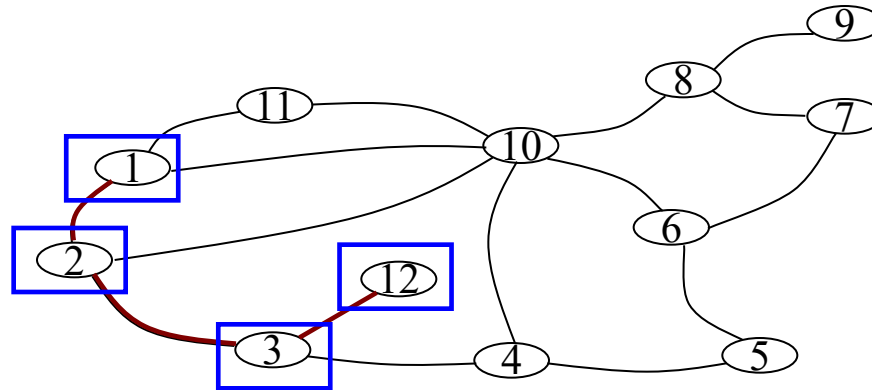
(3,4)
(2,10)
(1,11)
(1,10)

(3,4)
(2,10)
(1,11)
(1,10)



$$T = \{(1,2), (2,3), (3,12)\}$$

(3,4)
(2,10)
(1,11)
(1,10)



POP \rightarrow (3,4)

Visited : {1,2,3,12,4}

$T = \{(1,2), (2,3), (3,12), (3,4)\}$

PUSH

(4,4) and (4,10)

(4,10)
(4,5)
(2,10)
(1,11)
(1,10)

POP \rightarrow (4,10)

Visited : {1,2,3,12,4,10}

$T = \{(1,2), (2,3), (3,12), (3,4), (4,10)\}$

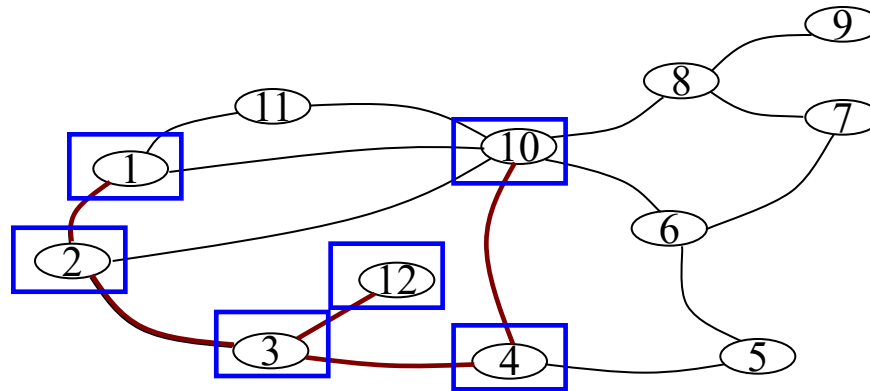
PUSH

(10,6) and (10,8)

(10,8)
(10,6)
(4,5)
(2,10)
(1,11)
(1,10)

• • •

(10,8)
(10,6)
(4,5)
(2,10)
(1,11)
(1,10)



$$T=\{(1,2), (2,3), (3,12), (3,4), (4,10)\}$$

• • •

Complexity

Elementary operations: Pop, Push, and visits

Number of PUSH:

$$\sum_{v \in V} d(v) = 2m$$

Number of POP:

$$\sum_{v \in V} d(v) = 2m$$

Visit of a node: n

$$O(n+m) = O(m)$$

DFS Algorithm - Recursive version

```
DFS(v)
Mark v visited
   $\forall w \in \text{Adjacent}(v)$ 
    if w not visited
      visit w
      DFS(w)
```

DFS Again - More Detail...

- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

Algorithm *DFS*(*G*)

Input graph *G*

Output labeling of the edges of *G*
as discovery edges and
back edges

for all *u* ∈ *G.vertices*()

setLabel(*u*, *UNEXPLORED*)

for all *e* ∈ *G.edges*()

setLabel(*e*, *UNEXPLORED*)

for all *v* ∈ *G.vertices*()

if *getLabel*(*v*) = *UNEXPLORED*
DFS(*G*, *v*)

Algorithm *DFS*(*G*, *v*)

Input graph *G* and a start vertex *v* of *G*
Output labeling of the edges of *G*
in the connected component of *v*
as discovery edges and back edges

setLabel(*v*, *VISITED*)

for all *e* ∈ *G.incidentEdges*(*v*)

if *getLabel*(*e*) = *UNEXPLORED*

w ← *opposite*(*v*, *e*)

if *getLabel*(*w*) = *UNEXPLORED*

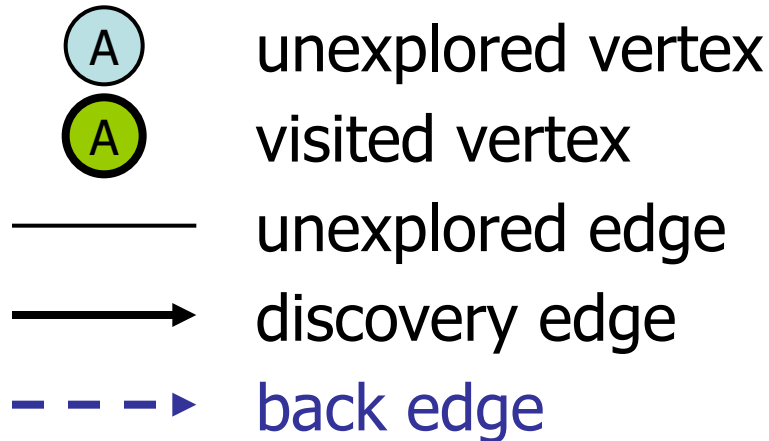
setLabel(*e*, *DISCOVERY*)

DFS(*G*, *w*)

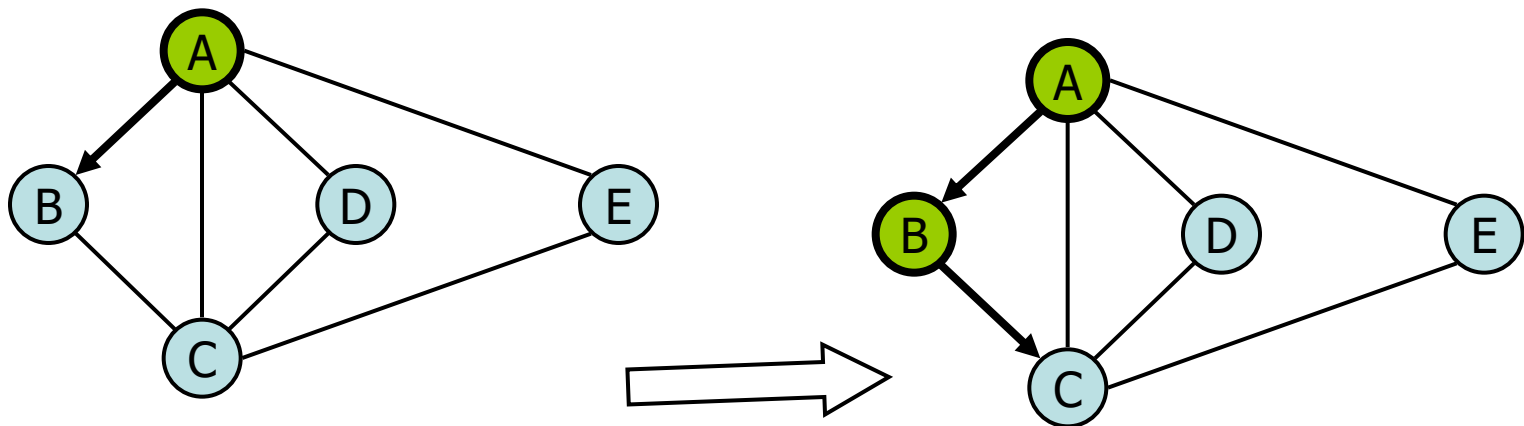
else

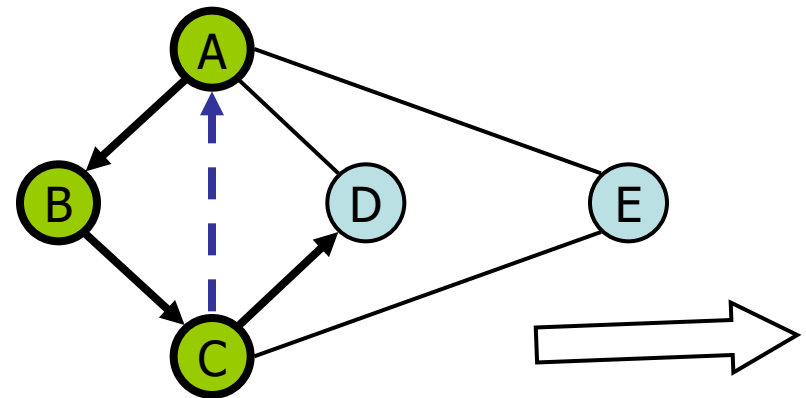
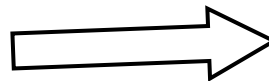
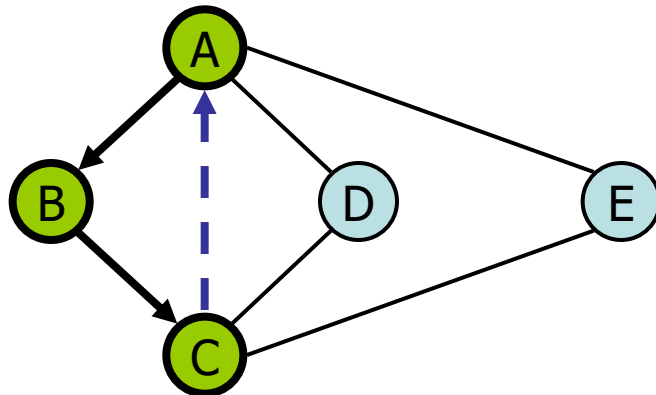
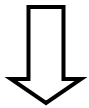
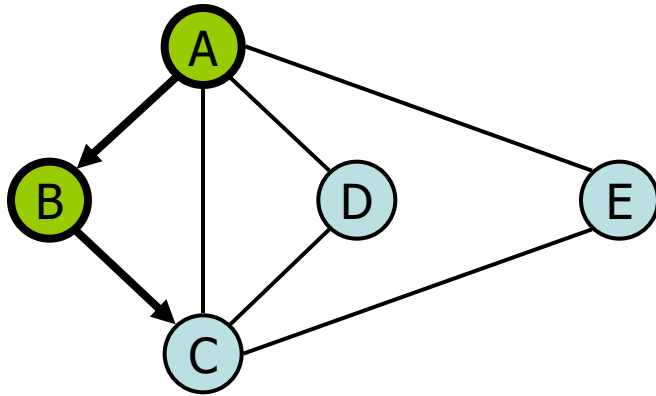
setLabel(*e*, *BACK*)

Example



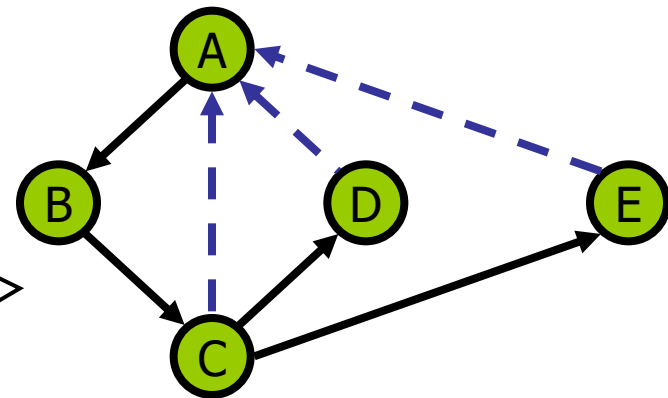
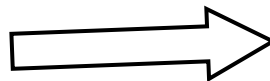
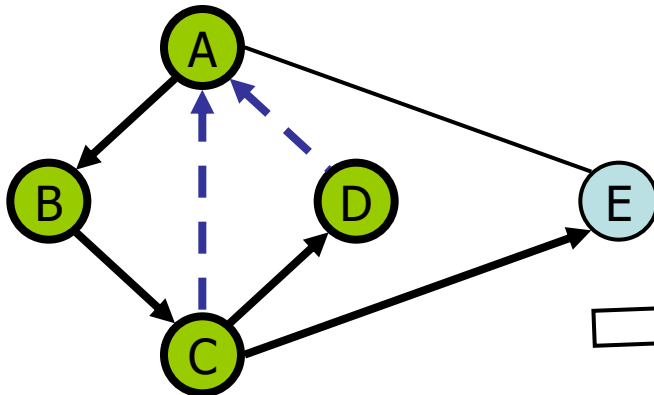
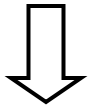
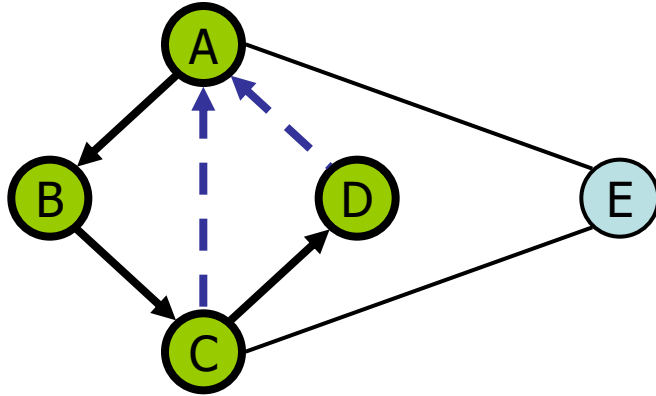
```
Algorithm DFS( $G, v$ )  
  setLabel( $v, VISITED$ )  
  for all  $e \in G.incidentEdges(v)$   
    if getLabel( $e$ ) = UNEXPLORED  
       $w \leftarrow opposite(v, e)$   
      if getLabel( $w$ ) = UNEXPLORED  
        setLabel( $e, DISCOVERY$ )  
        DFS( $G, w$ )  
      else  
        setLabel( $e, BACK$ )
```





```

Algorithm DFS(G, v)
  setLabel(v, VISITED)
  for all e ∈ G.incidentEdges(v)
    if getLabel(e) = UNEXPLORED
      w ← opposite(v,e)
      if getLabel(w) = UNEXPLORED
        setLabel(e, DISCOVERY)
        DFS(G, w)
      else
        setLabel(e, BACK)
  
```



Algorithm *DFS*(*G*, *v*)

setLabel(*v*, *VISITED*)

for all *e* ∈ *G.incidentEdges*(*v*)

if *getLabel*(*e*) = *UNEXPLORED*

w ← *opposite*(*v*, *e*)

if *getLabel*(*w*) = *UNEXPLORED*

setLabel(*e*, *DISCOVERY*)

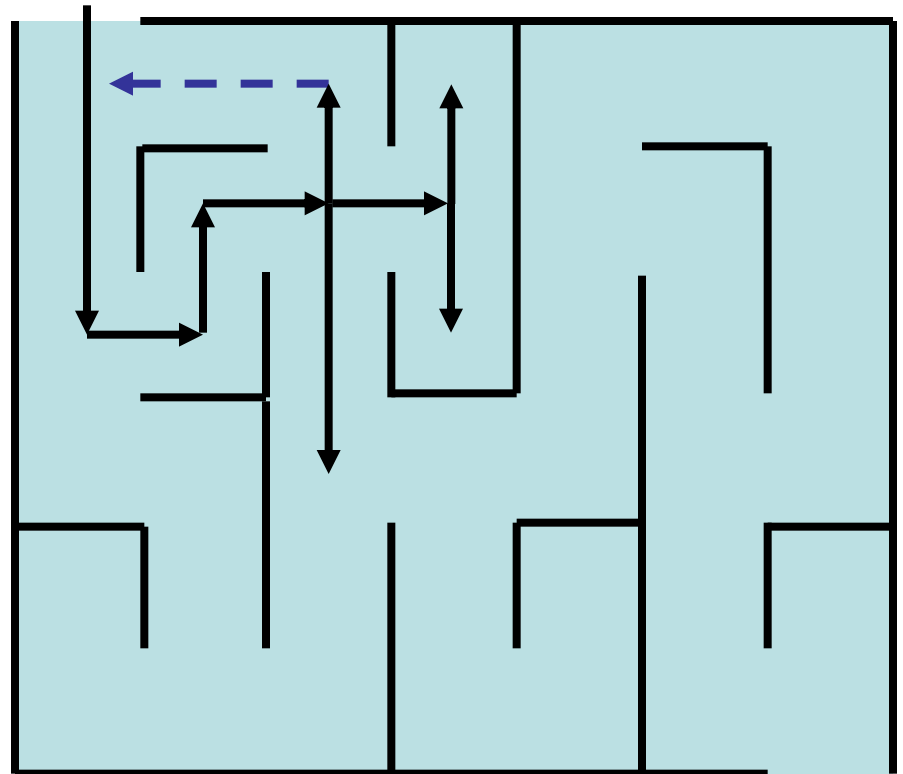
DFS(*G*, *w*)

else

setLabel(*e*, *BACK*)

DFS and Maze Traversal

- The DFS algorithm is similar to a classic strategy for exploring a maze
 - We mark each intersection, corner and dead end (vertex) visited
 - We mark each corridor (edge) traversed
 - We keep track of the path back to the entrance (start vertex) by means of a rope (recursion stack)



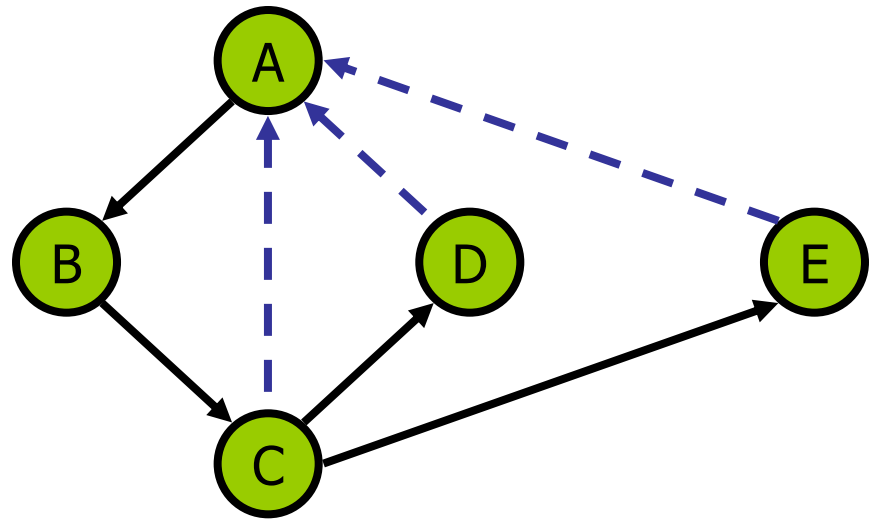
Properties of DFS

Property 1

$DFS(G, v)$ visits all the vertices and edges in the connected component of v

Property 2

The discovery edges labeled by $DFS(G, v)$ form a **spanning tree** of the connected component of v



Analysis of DFS + labeling

- Setting/getting a vertex/edge label takes $O(1)$ time
- Each vertex is labeled twice
 - once as UNEXPLORED
 - once as VISITED $2n$
- Each edge is labeled twice
 - once as UNEXPLORED
 - once as DISCOVERY or BACK $2m$
- Method incidentEdges is called once for each vertex

$$\sum_{v \in V} d(v) = 2m$$

If the graph is implemented with adjacency list

- DFS runs in $O(n + m)$ time provided the graph is represented by the adjacency list structure
- If the graph is connected ($m \geq n-1$) then
 $O(n + m) = O(m)$

Conclusion

If we represent the graph with an adjacency list

Complexity of DFS is $O(n+m)$

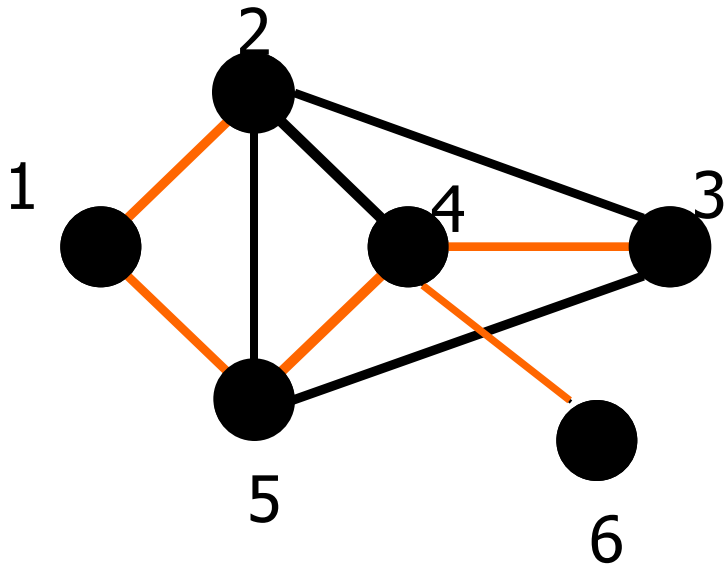
WORST CASE: $m = O(n^2)$.

Question: With adjacency matrix ?

With adjacency matrix DFS is always $O(n^2)$, even if m is much smaller than n^2 .

Path Finding

- We can specialize the DFS algorithm to find a path between two given vertices u and z using the template method pattern
- We call $DFS(G, u)$ with u as the start vertex
- We use a stack S to keep track of the path between the start vertex and the current vertex
- As soon as destination vertex z is encountered, we return the path as the contents of the stack



2 -- 6

$2_{(2,1)}$ 1 $(1,5)$ 5 $(5,4)$ 4 ~~$(4,3)$~~ ~~3~~ $(4,6)$ 6

```

Algorithm pathDFS( $G, v, z$ )
  setLabel( $v, VISITED$ )
  S.push( $v$ )
  if  $v = z$ 
    return S.elements()
  for all  $e \in G.incidentEdges(v)$ 
    if getLabel( $e$ ) = UNEXPLORED
       $w \leftarrow opposite(v, e)$ 
      if getLabel( $w$ ) = UNEXPLORED
        setLabel( $e, DISCOVERY$ )
        S.push( $e$ )
        pathDFS( $G, w, z$ )
        S.pop()
      else
        setLabel( $e, BACK$ )
  S.pop()
  
```

Cycle Finding

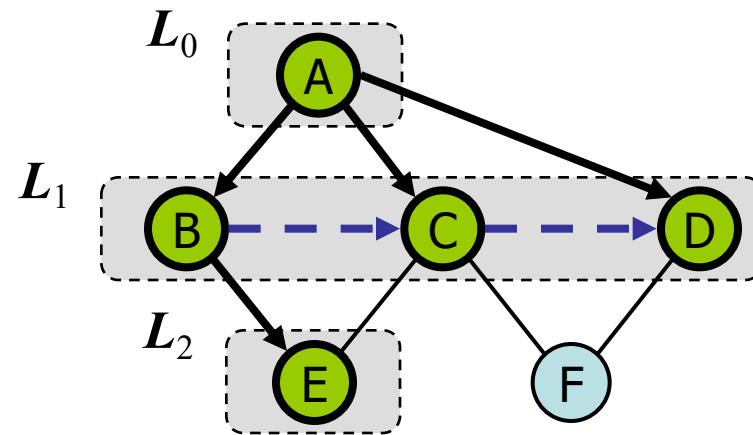
- We can specialize the DFS algorithm to find a simple cycle using the template method pattern
- We use a stack S to keep track of the path between the start vertex and the current vertex
- As soon as a back edge (v, w) is encountered, we return the cycle as the portion of the stack from the top to vertex w

```

Algorithm cycleDFS(G, v)
  setLabel(v, VISITED)
  S.push(v)
  for all e ∈ G.incidentEdges(v)
    if getLabel(e) = UNEXPLORED
      w ← opposite(v,e)
      S.push(e)
      if getLabel(w) = UNEXPLORED
        setLabel(e, DISCOVERY)
        cycleDFS(G, w)
        S.pop(e)
      else
        T ← new empty stack
        repeat
          o ← S.pop()
          T.push(o)
        until o = w
        return T.elements()
  S.pop(v)

```


Breadth-First Search



Outline and Reading

- Breadth-first search
 - Algorithm
 - Example
 - Properties
 - Analysis
 - Applications
- DFS vs. BFS
 - Comparison of applications
 - Comparison of edge labels

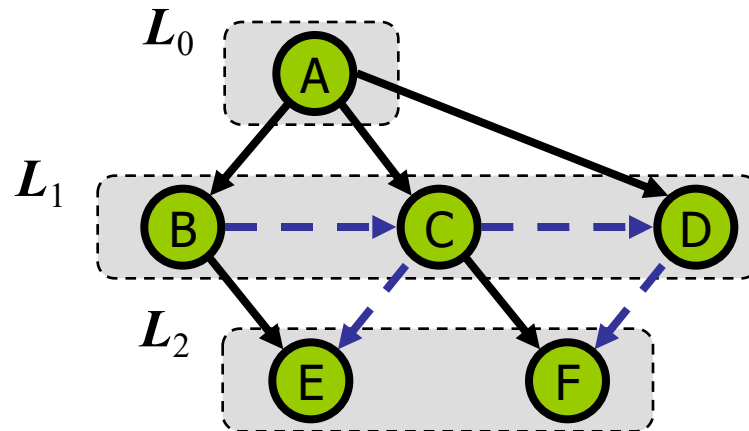
Breadth-First Search

Breadth-First Search (BFS) is a graph traversal technique that:

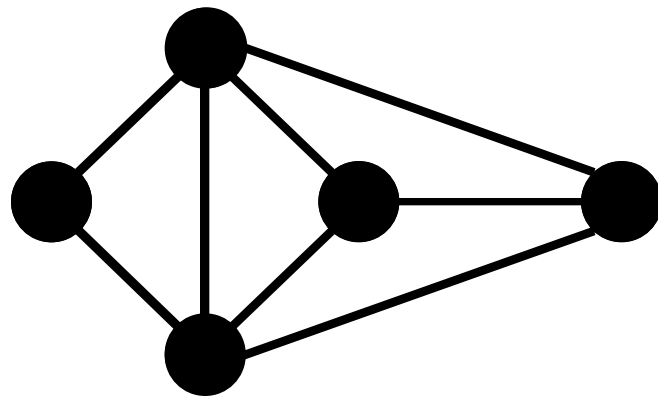
- on a graph with n vertices and m edges, takes $O(n + m)$ time
- can be further extended to solve other graph problems
 - Find and report a path with the minimum number of edges between two given vertices
 - Find a simple cycle, if there is one

The idea:

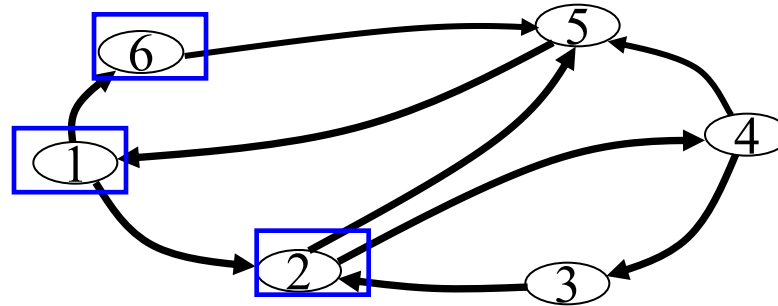
Visit a vertex and then visit all unvisited vertices that are adjacent to it before visiting a vertex which is 2 away from it.



level by level



Breadth First Search with a Queue



visited : {1}

$T = \phi$

to visit: {(1,2), (1,6)}

(1,2) – 2 visited?

visited : {1,2}

$T = \{(1,2)\}$

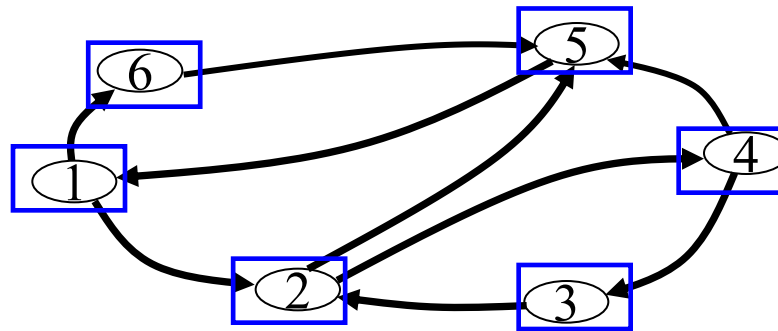
to visit: {(1,6), (2,4), (2,5)}

(1,6) - 6 visited ?

visited : {1,2,6}

$T = \{(1,2), (1,6)\}$

to visit: {(2,4), (2,5), (6,5)}



(2,4) - 4 visited?

visited: {1,2,6,4}

$T = \{(1,2), (1,6), (2,4)\}$

to visit: $\{(2,5), (6,5), (4,5), (4,3)\}$

(2,5) - 5 visited ?

visited: {1,2,6,4,5}

$T = \{(1,2), (1,6), (2,4), (2,5)\}$

to visit: $\{(6,5), (4,5), (4,3)\}$

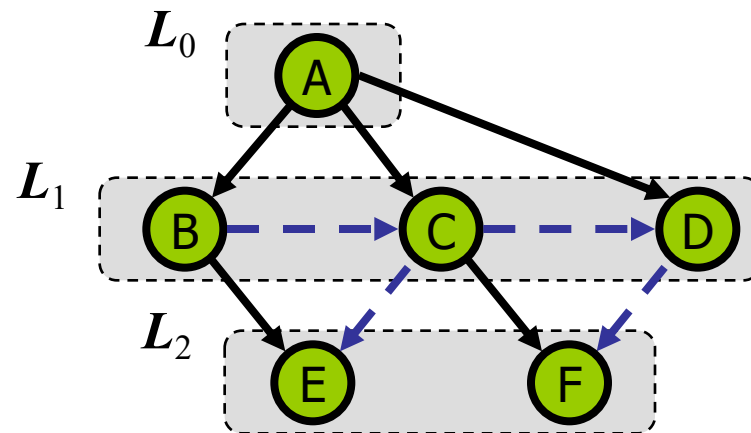
(6,5) - 5? already visited!

(4,5) - 5? already visited!

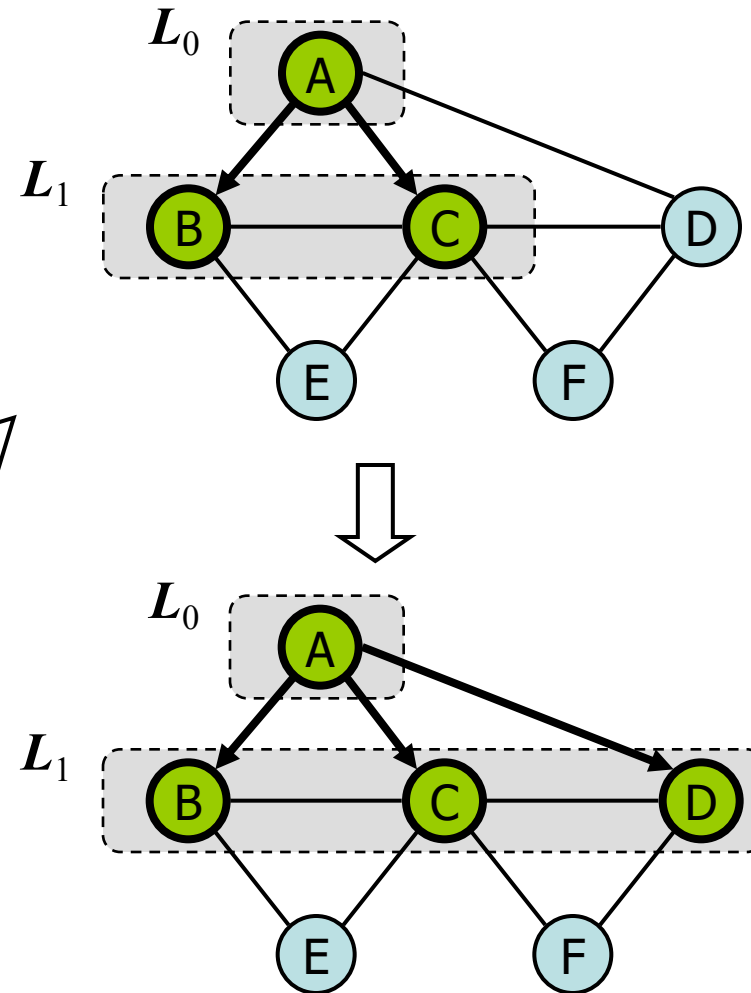
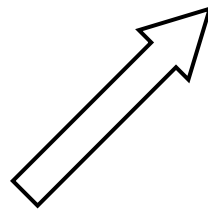
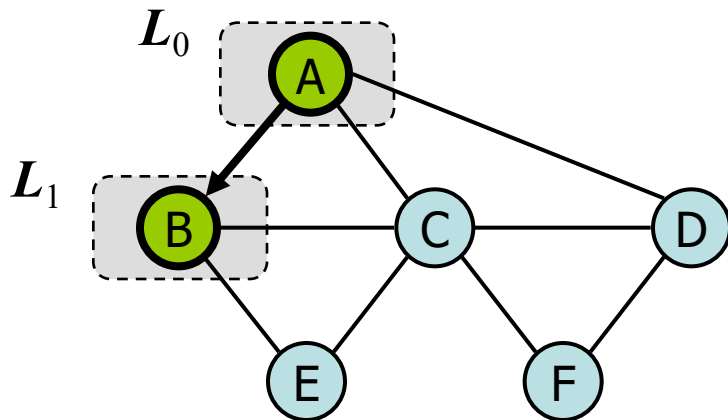
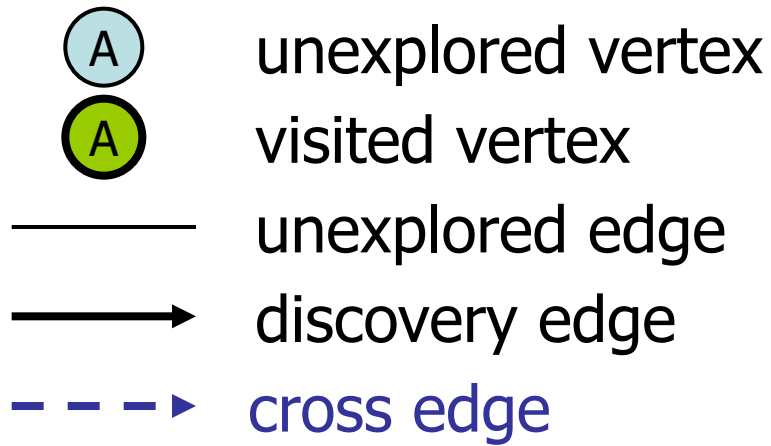
(4,3) - 3?

BFS with labeling

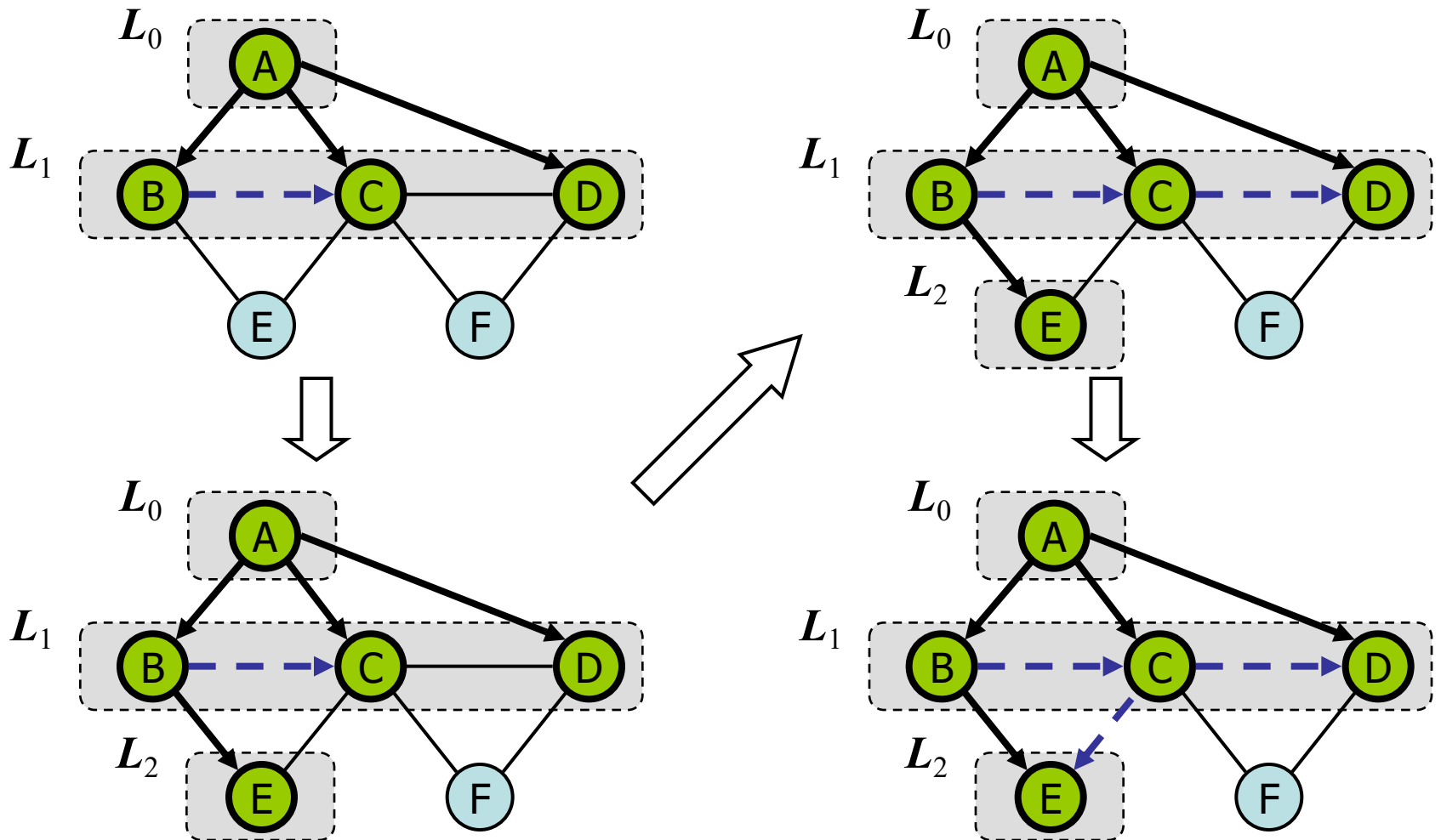
Using a sequence for each level



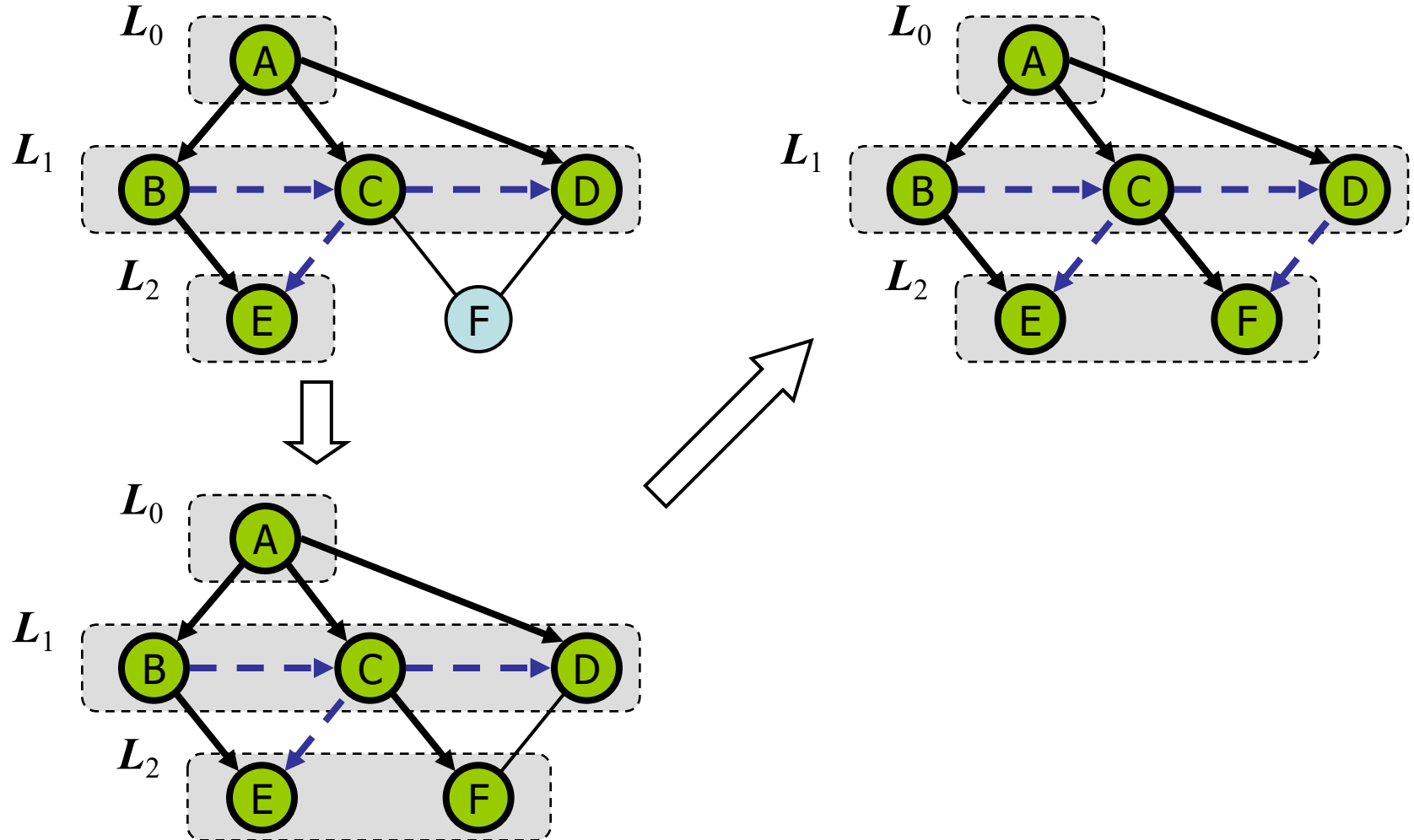
Example



Example (cont.)



Example (cont.)



BFS Again - more details

- The algorithm uses a mechanism for setting and getting “labels” of vertices and edges

Algorithm *BFS*(*G*)

Input graph *G*

Output labeling of the edges and partition of the vertices of *G*

```
for all u ∈ G.vertices()  
    setLabel(u, UNEXPLORED)  
for all e ∈ G.edges()  
    setLabel(e, UNEXPLORED)  
for all v ∈ G.vertices()  
    if getLabel(v) = UNEXPLORED  
        BFS(G, v)
```

Algorithm *BFS*(*G*, *s*)

*L*₀ ← new empty sequence

*L*₀.*insertLast*(*s*)

setLabel(*s*, *VISITED*)

i ← 0

while ! *L*_{*i*}.*isEmpty*()

*L*_{*i*+1} ← new empty sequence

for all *v* ∈ *L*_{*i*}.*elements*()

for all *e* ∈ *G.incidentEdges*(*v*)

if *getLabel*(*e*) = *UNEXPLORED*

w ← *opposite*(*v*, *e*)

if *getLabel*(*w*) = *UNEXPLORED*

setLabel(*e*, *DISCOVERY*)

setLabel(*w*, *VISITED*)

*L*_{*i*+1}.*insertLast*(*w*)

else

setLabel(*e*, *CROSS*)

i ← *i* + 1

Properties

Notation

G_s : connected component of s

Property 1

$BFS(G, s)$ visits all the vertices and edges of G_s

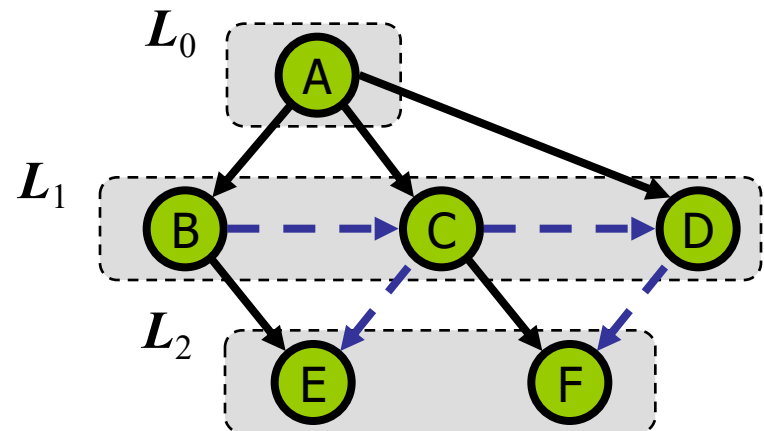
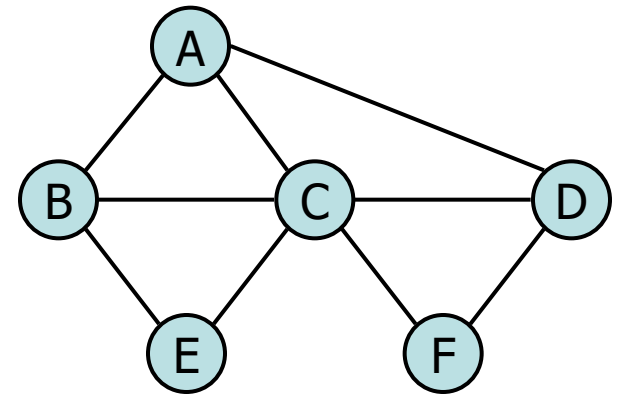
Property 2

The discovery edges labeled by $BFS(G, s)$ form a spanning tree T_s of G_s

Property 3

For each vertex v in L_i

- The path of T_s from s to v has i edges
- Every path from s to v in G_s has at least i edges



Analysis

- Setting/getting a vertex/edge label takes $O(1)$ time
- Each vertex is labeled twice
 - once as UNEXPLORED
 - once as VISITED
- Each edge is labeled twice
 - once as UNEXPLORED
 - once as DISCOVERY or CROSS
- Each vertex is inserted once into a sequence L_i
- Method incidentEdges is called once for each vertex
- BFS runs in $O(n + m)$ time provided the graph is represented by the adjacency list structure
 - Recall that $\sum_v \deg(v) = 2m$

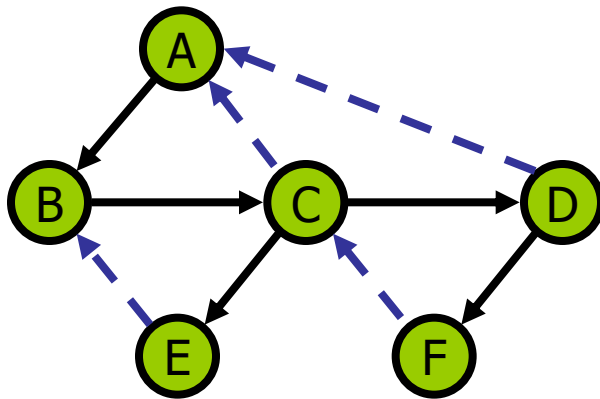
Question: what if your graph is represented with adjacency matrix? See DFS.

Applications

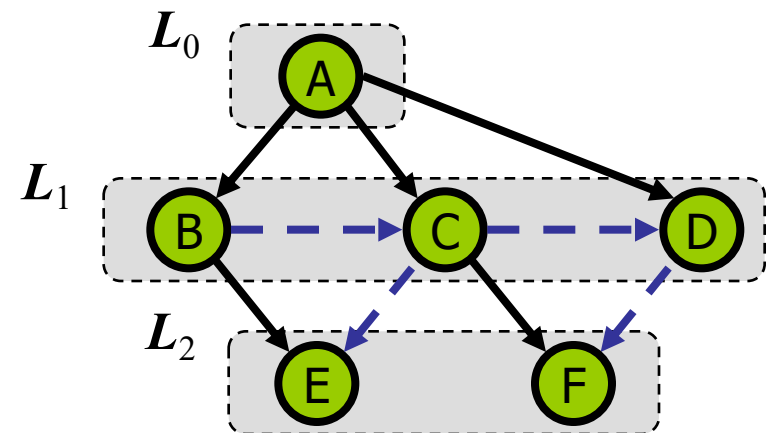
- Using the template method pattern, we can specialize the BFS traversal of a graph G to solve the following problems in $O(n + m)$ time
 - Compute the connected components of G
 - Compute a spanning forest of G
 - Find a simple cycle in G , or report that G is a forest
 - Given two vertices of G , find a path in G between them with the minimum number of edges, or report that no such path exists

DFS vs. BFS

Applications	DFS	BFS
Spanning forest, connected components, paths, cycles	✓	✓
Shortest paths		✓
*) Biconnected components	✓	



DFS



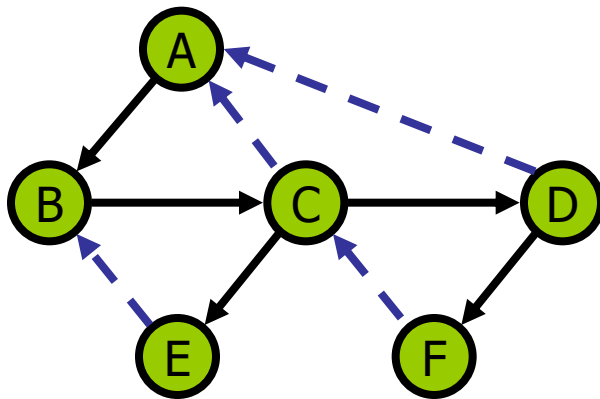
BFS

*) “A connected graph is *biconnected* if the removal of any single vertex (and all edges incident on that vertex) can not disconnect the graph.”

DFS vs. BFS (cont.)

Back edge (v, w)

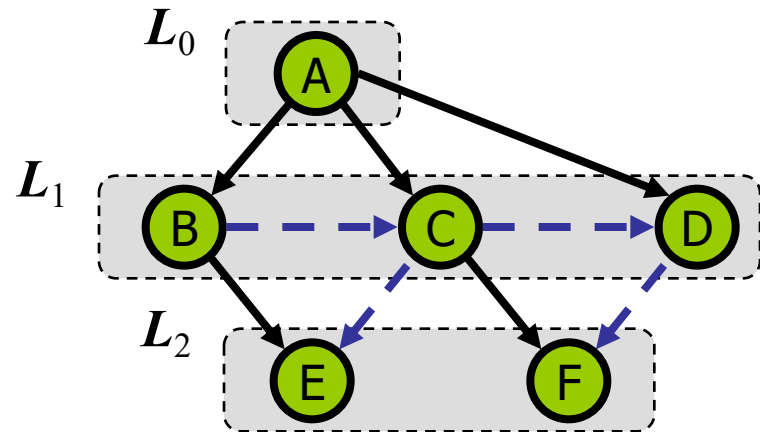
- w is an ancestor of v in the tree of discovery edges



DFS

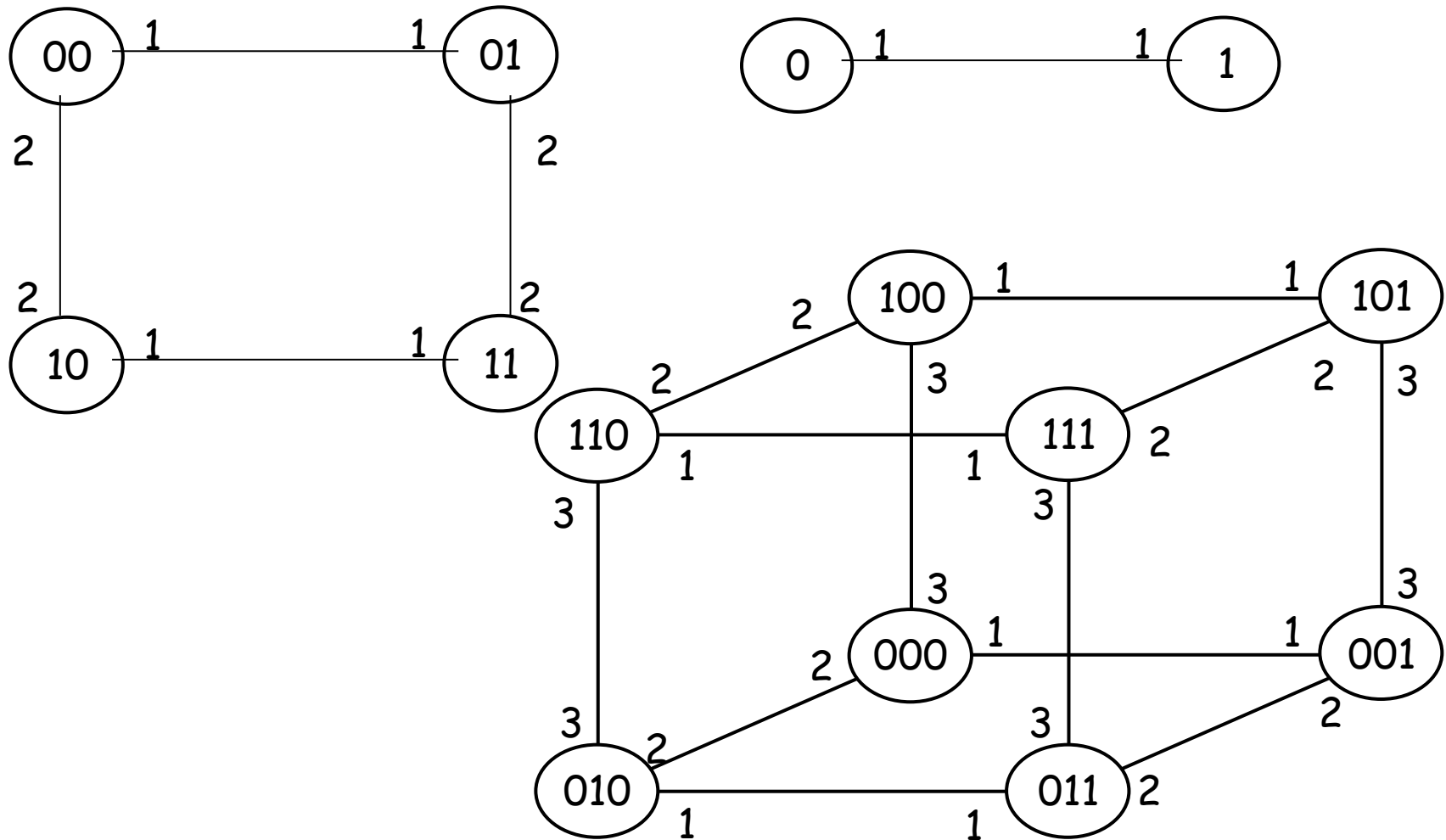
Cross edge (v, w)

- w is in the same level as v or in the next level in the tree of discovery edges

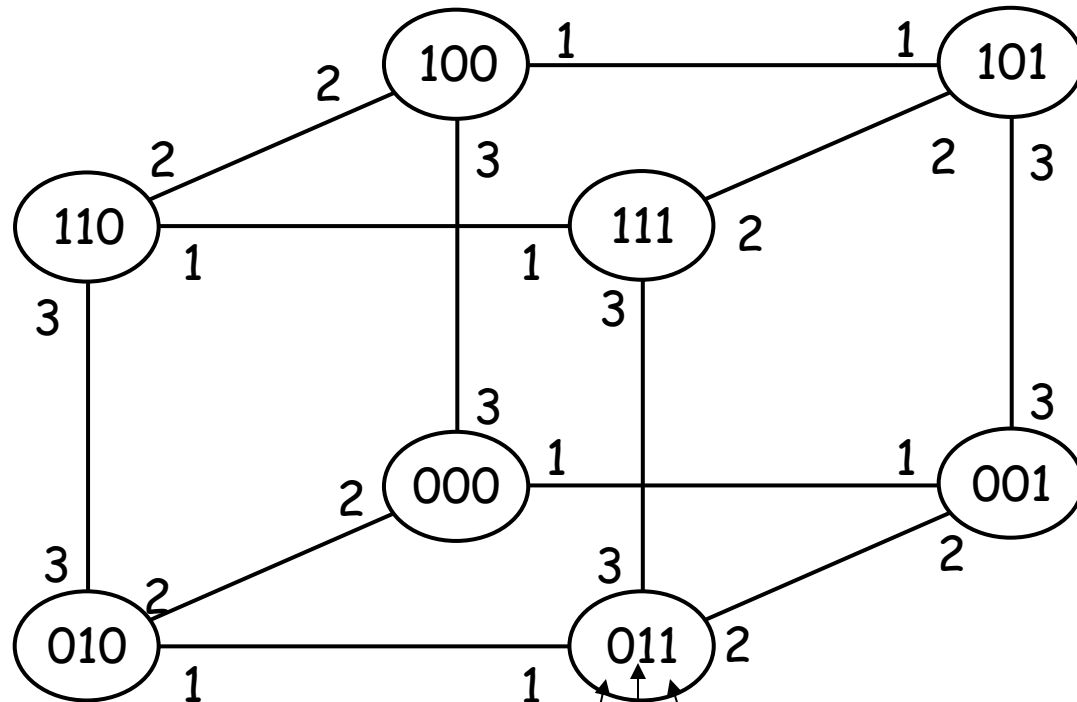


BFS

Example: DFS/BFS in the hypercube



Each link between two nodes is labeled by the dimension of the bit₆₆ by which the nodes' name differ.

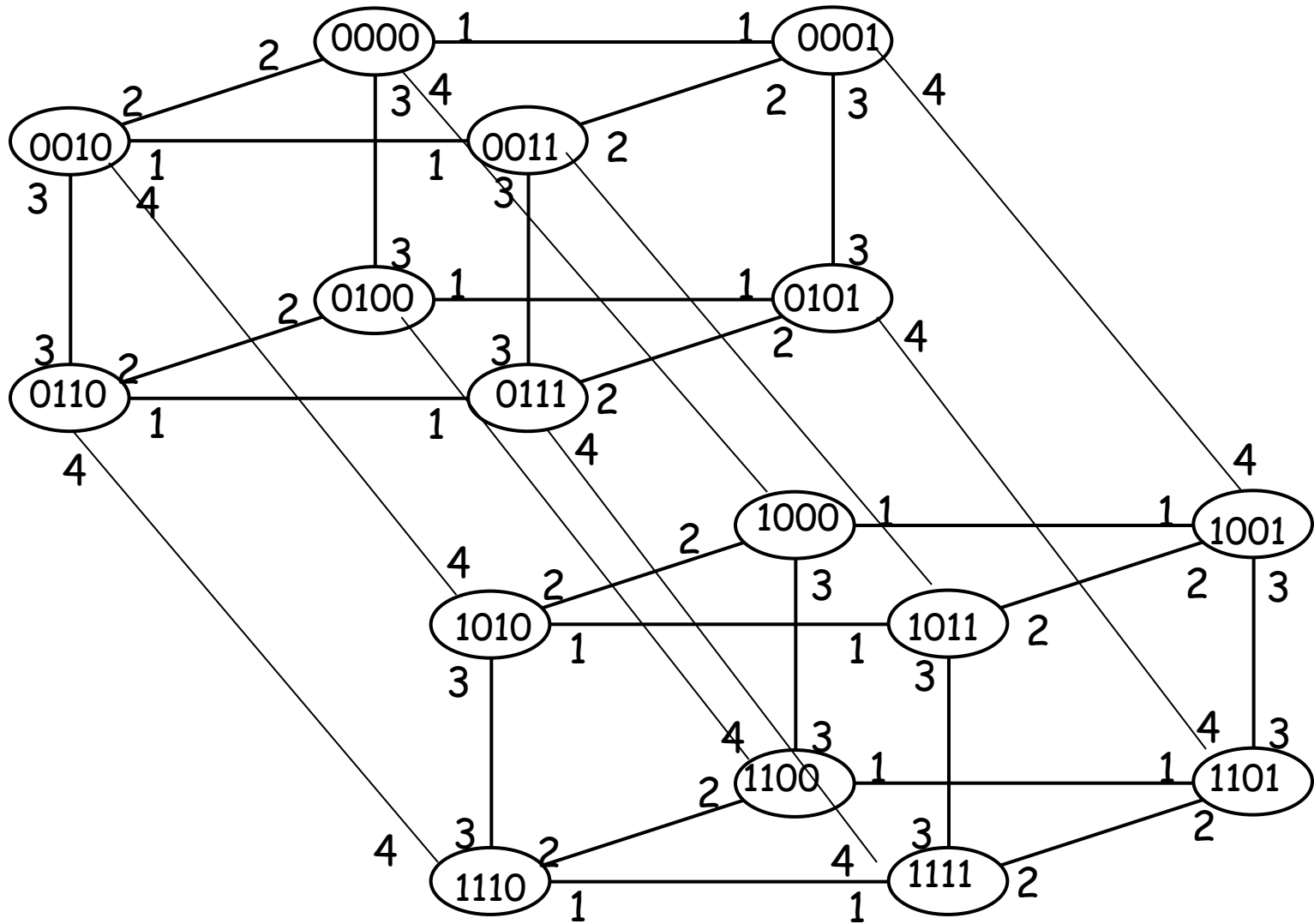


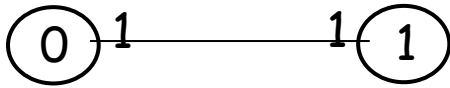
$X = x_k x_{k-1} \dots x_1 x_0$

K-bit name

$x_2 x_1 x_0$

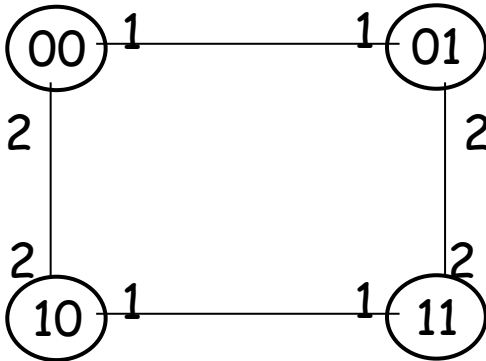
first bit





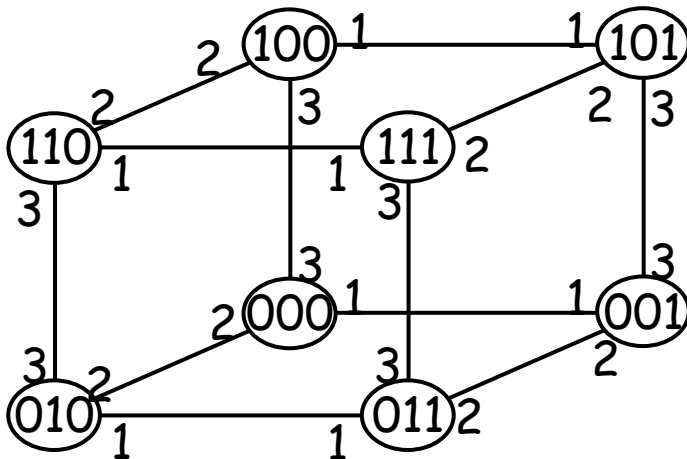
$n=2$

dimensions=1



$n=4$

dimensions=2



$n=8$

dimensions=3

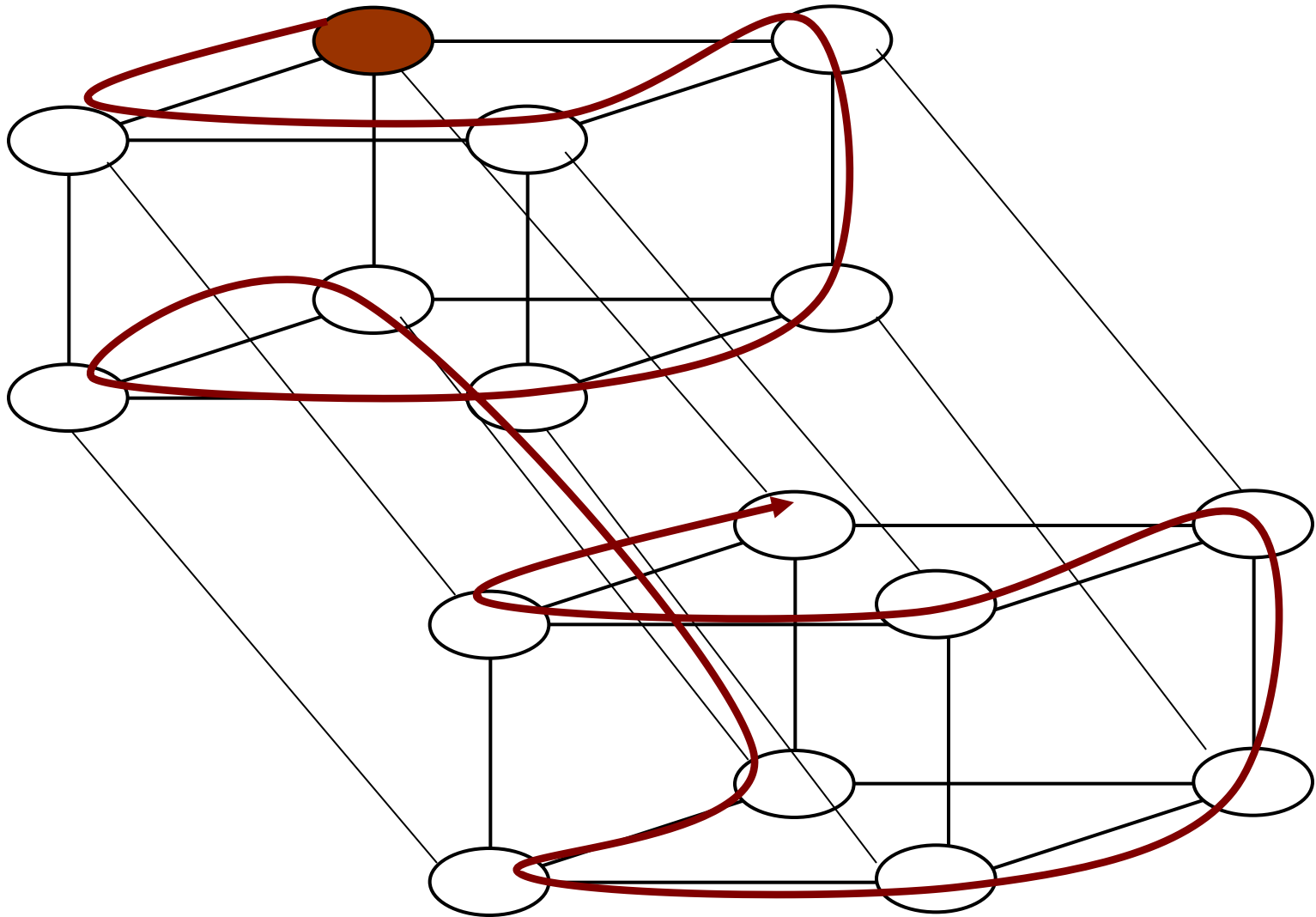
dimensions = $\log n$

A hypercube of dimension d has $n = 2^d$ nodes

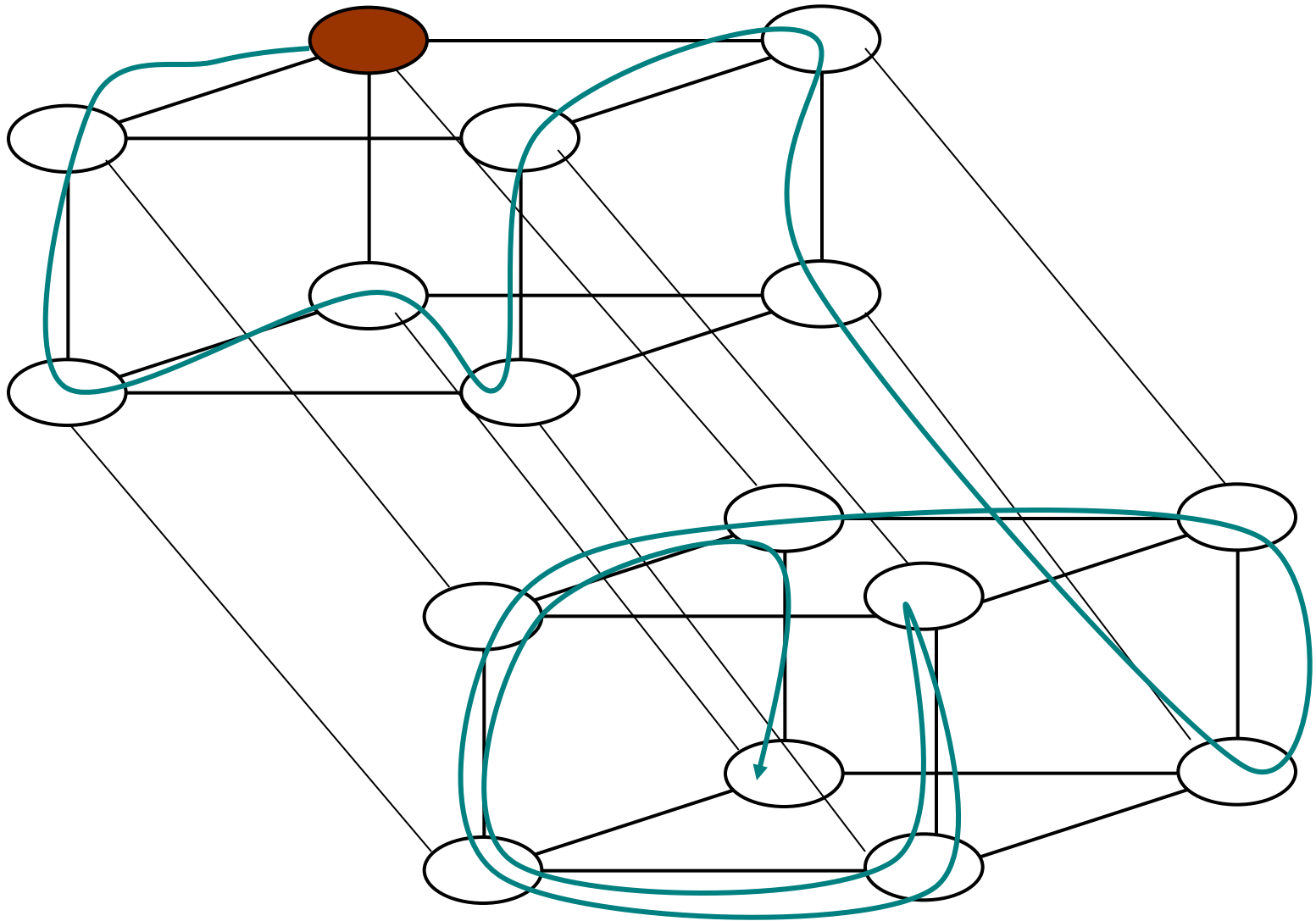
Each node has d links

$$\rightarrow m = n d / 2 = O(n \log n)$$

A Depth-first traversal



Another Depth-first traversal



A Breadth-first traversal

