# CSE 421 Introduction to Algorithms

Lecture 4: BFS, DFS Properties/Applications, Topological Sort

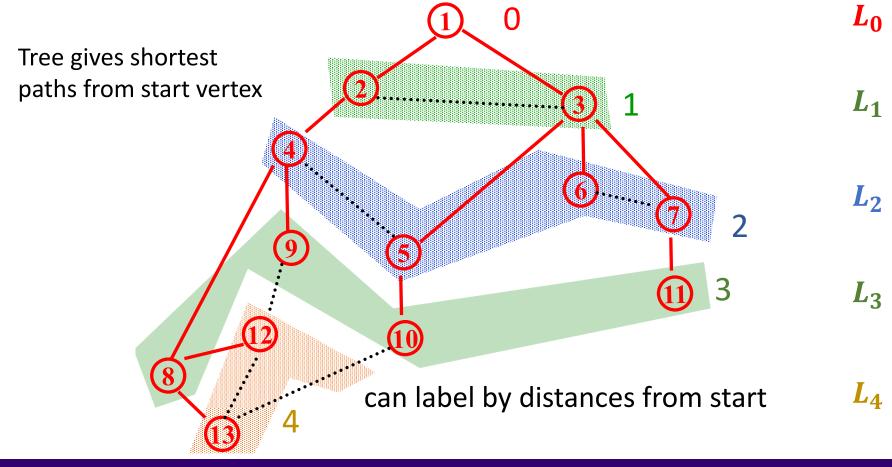
#### **Generic Graph Traversal Algorithm**

```
Given: Graph graph G = (V, E) vertex S \in V
```

Find: set R of vertices reachable from  $S \in V$ 

```
Reachable(s): R \leftarrow \{s\} while there is a (u, v) \in E where u \in R and v \notin R Add v to R return R
```

## **BFS Tree**



#### **Undirected Graph Search Application: Connected Components**

Want to answer questions of the form:

**Given**: vertices  $\boldsymbol{u}$  and  $\boldsymbol{v}$  in  $\boldsymbol{G}$ 

Is there a path from  $\boldsymbol{u}$  to  $\boldsymbol{v}$ ?

Idea: create array A s.t

A[u] = smallest numbered vertex connected to u

Answer is yes iff A[u] = A[v]

**Q:** Why is this better than an array Path[u, v]?

#### **Undirected Graph Search Application: Connected Components**

```
Initial state: all v unvisited for s \leftarrow 1 to n do if state(s) \neq fully-explored then BFS(s): setting A[u] \leftarrow s for each u found (and marking u visited/fully-explored) endfor
```

Total cost: O(n + m)

- Each vertex is touched once in outer procedure and edges examined in different BFS runs are disjoint
- Works also with Depth First Search ...

## $\mathsf{DFS}(u)$ – Recursive Procedure

Global Initialization: mark all vertices "unvisited"  $\mathsf{DFS}(u)$ mark u "visited" and add u to R for each edge (u,v) if (v) is "unvisited")

DFS(v)end for mark u "fully-explored"

## Properties of DFS(s)

#### Like BFS(s):

- DFS(s) visits x iff there is a path in G from s to x
- Edges into undiscovered vertices define depth-first spanning tree of G

#### Unlike the BFS tree:

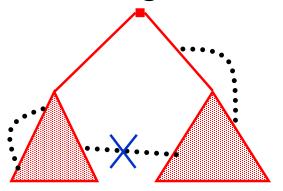
- the DFS spanning tree isn't minimum depth
- its levels don't reflect min distance from the root
- non-tree edges never join vertices on the same or adjacent levels

#### BUT...

## Non-tree edges in DFS tree of undirected graphs

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

• In other words ... No "cross edges".



#### No cross edges in DFS on undirected graphs

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

Claim: For every x, y in the DFS tree T, if (x, y) is an edge not in T then one of x or y is an ancestor of the other in T

#### **Proof:**

- One of  $\mathsf{DFS}(x)$  or  $\mathsf{DFS}(y)$  is called first, suppose WLOG that  $\mathsf{DFS}(x)$  was called before  $\mathsf{DFS}(y)$
- During DFS(x), the edge (x, y) is examined
- Since (x, y) is a not an edge of T, y was already visited when 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.

#### **Applications of Graph Traversal: Bipartiteness Testing**

**Definition:** An undirected graph *G* is bipartite iff we can color its vertices **red** and **green** so each edge has different color endpoints

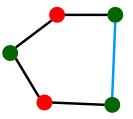
Input: Undirected graph G

**Goal:** If **G** is bipartite, output a coloring;

otherwise, output "NOT Bipartite".

**Fact:** Graph G contains an odd-length cycle  $\Rightarrow$  it is not bipartite

Just coloring the cycle part of *G* is impossible



On a cycle the two colors must alternate, so

- green every 2<sup>nd</sup> vertex
- red every 2<sup>nd</sup> vertex

Can't have either if length is not divisible by 2.

#### **Applications of Graph Traversal: Bipartiteness Testing**

**WLOG** ("without loss of generality"): Can assume that G is connected

Otherwise run on each component

Simple idea: start coloring nodes starting at a given node s

- Color s red
- Color all neighbors of s green
- Color all their neighbors red, etc.
- If you ever hit a node that was already colored
  - the **same** color as you want to color it, ignore it
  - the **opposite** color, output "NOT Bipartite" and halt

#### **BFS** gives Bipartiteness

Run BFS assigning all vertices from layer  $L_i$  the color  $i \mod 2$ 

- i.e., red if they are in an even layer, green if in an odd layer
- if no edge joining two vertices of the same color
  - then it is a good coloring
- otherwise
  - there is a bad edge; output "Not Bipartite"

Why is that "Not Bipartite" output correct?

#### Why does BFS work for Bipartiteness?

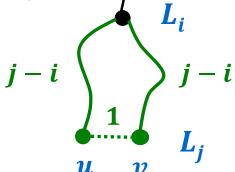
**Recall:** All edges join vertices on the same or adjacent BFS layers

 $\Rightarrow$  Any bad edge must join two vertices u and v in the same layer

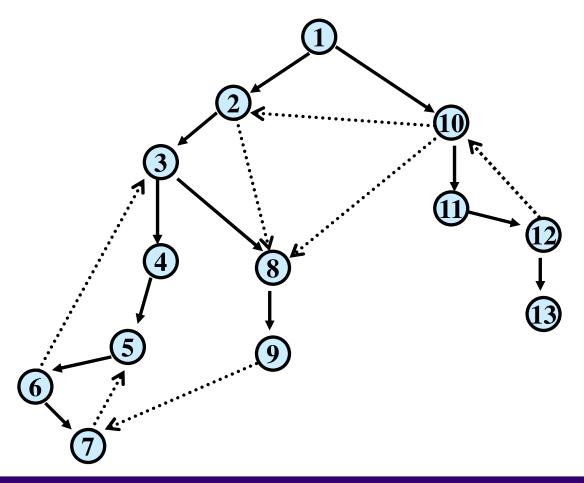
Say the layer with  $oldsymbol{u}$  and  $oldsymbol{v}$  is  $oldsymbol{L_j}$ 

 $oldsymbol{u}$  and  $oldsymbol{v}$  have common ancestor at some level  $oldsymbol{L_i}$  for  $oldsymbol{i} < oldsymbol{j}$ 

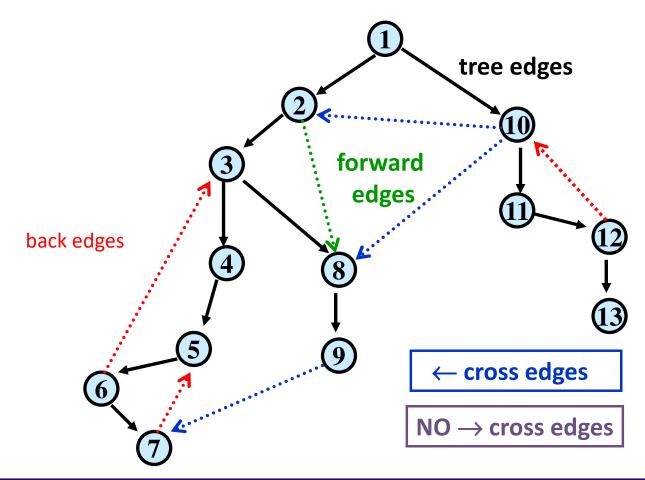
Odd cycle of length 2(j - i) + 1  $\Rightarrow$  Not Bipartite



# $\mathsf{DFS}(v)$ for a directed graph



# $\mathsf{DFS}(v)$



#### **Properties of Directed DFS**

 Before DFS(s) returns, it visits all previously unvisited vertices reachable via directed paths from s

Every cycle contains a back edge in the DFS tree

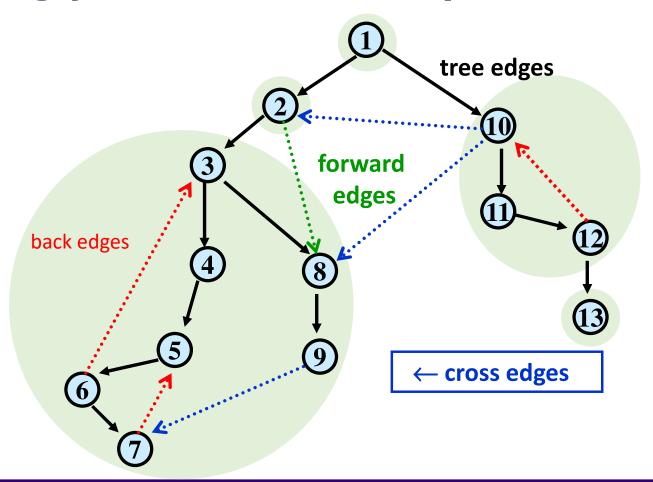
#### **Strongly Connected Components of Directed Graphs**

**Defn:** Vertices u and v are strongly connected iff they are on a directed cycle (there are paths from u to v and from v to u).

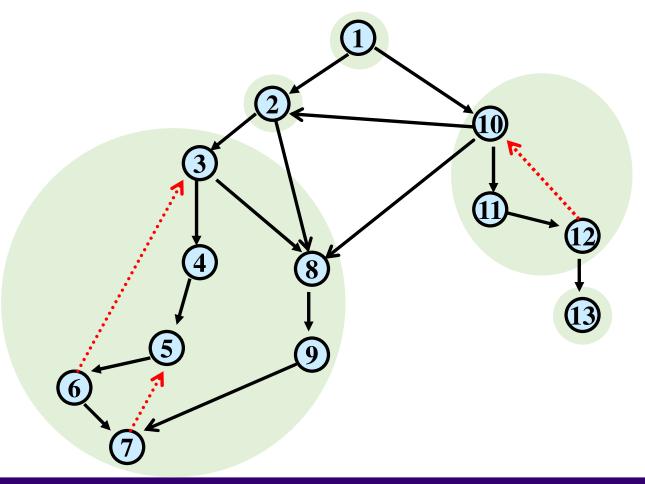
**Defn:** Can partition vertices of any directed graph into strongly connected components:

- 1. all pairs of vertices in the same component are strongly connected
- 2. can't merge components and keep property 1
- Strongly connected components can be stored efficiently just like connected components
- Can be found by extending DFS algorithm in O(n+m) time using extra bookkeeping
  - We won't cover the details

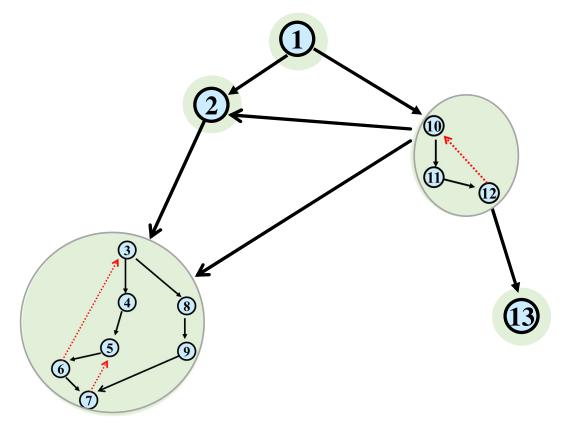
## **Strongly Connected Components**



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## **Strongly Connected Components**



## **Directed Acyclic Graphs**

A directed graph G = (V, E) is acyclic iff it has no directed cycles

Terminology: A directed acyclic graph is also called a DAG

After shrinking the strongly connected components of a directed graph to single vertices, the result is a DAG

**Given:** a directed acyclic graph (DAG) G = (V, E)

Output: numbering of the vertices of G with distinct numbers from  $\mathbf{1}$  to n so that edges only go from lower numbered to higher numbered vertices

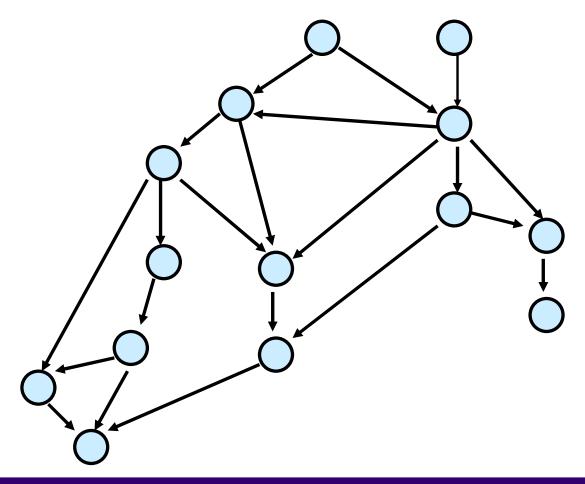
#### Applications:

- nodes represent tasks
- edges represent precedence between tasks
- topological sort gives a sequential schedule for solving them

Nice algorithmic paradigm for general directed graphs:

• Process strongly connected components one-by-one in the order given by topological sort of the DAG you get from shrinking them.

## **Directed Acyclic Graph**



## In-degree 0 vertices

Claim: Every DAG has a vertex of in-degree 0

**Proof:** By contradiction

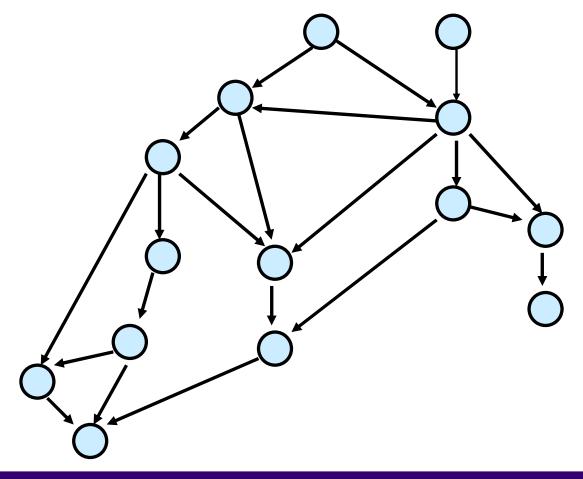
Suppose every vertex has some incoming edge Consider following procedure:

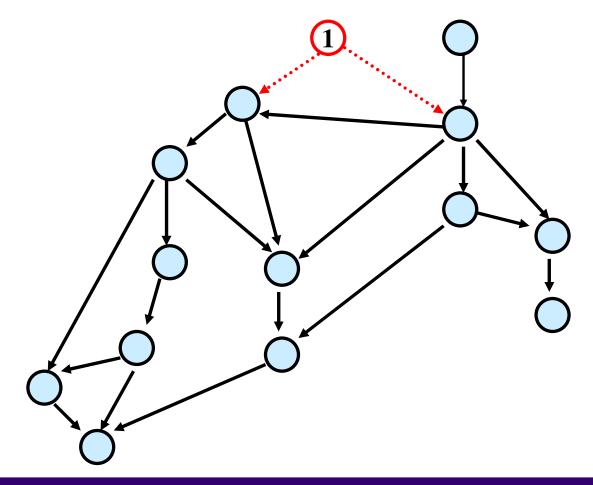
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while (true) do v \leftarrow some predecessor of v
```

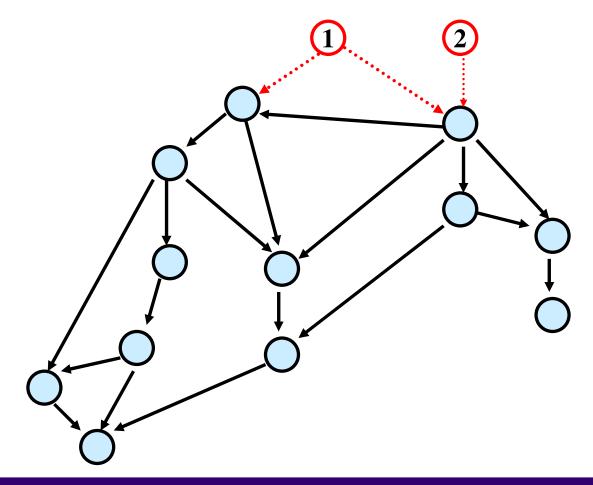
- After n + 1 steps where n = |V| there will be a repeated vertex
  - This yields a cycle, contradicting that it is a DAG.

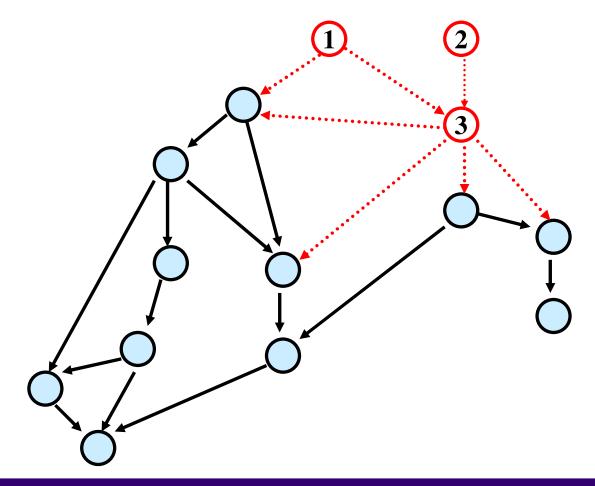
Can do using DFS

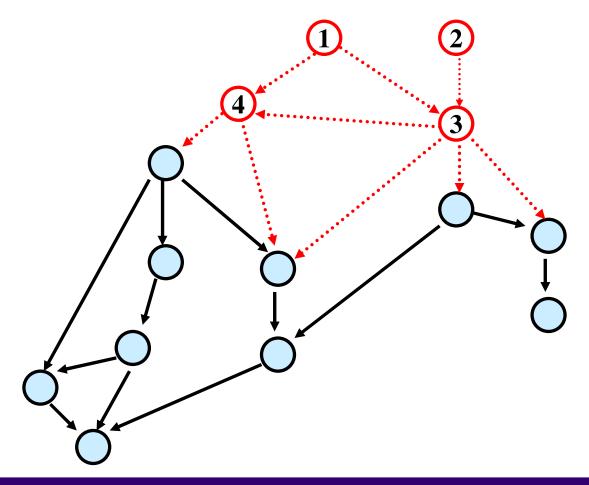
- Alternative simpler idea:
  - Any vertex of in-degree 0 can be given number 1 to start
  - Remove it from the graph
  - Then give a vertex of in-degree 0 number 2
  - Etc.

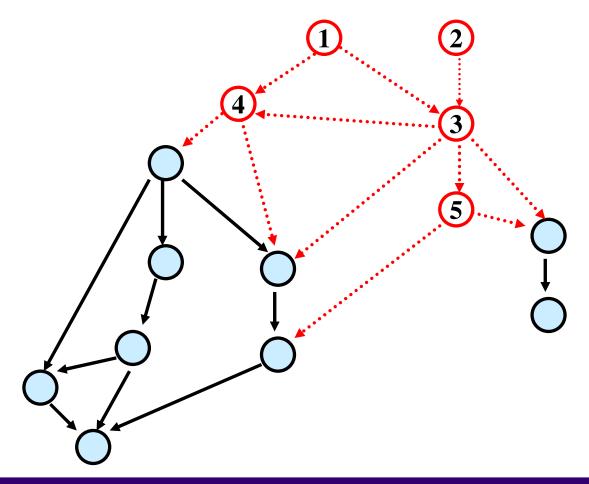


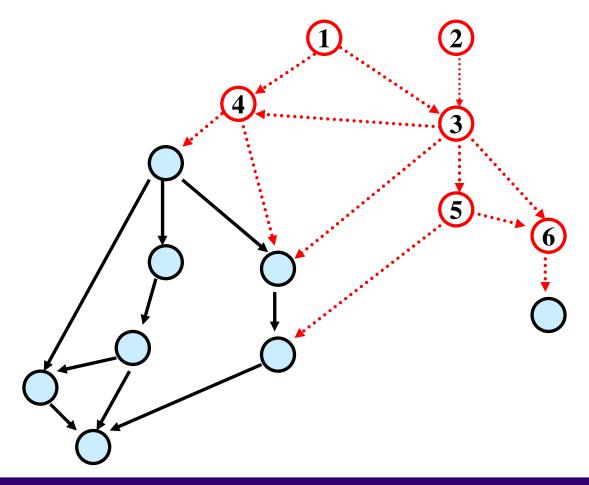


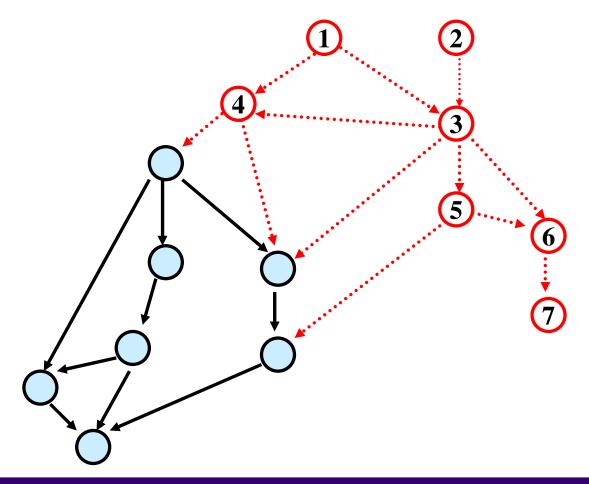


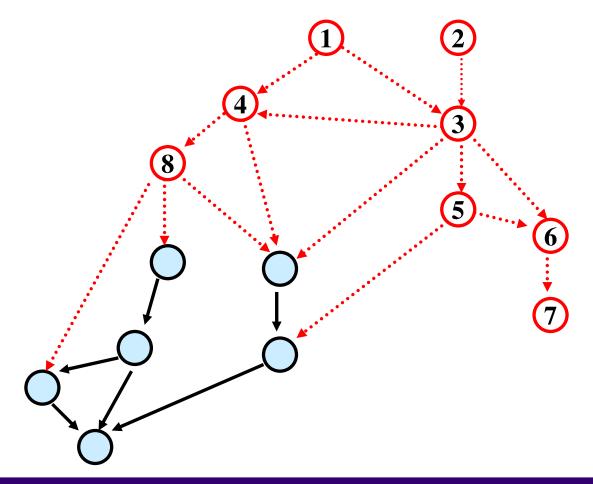


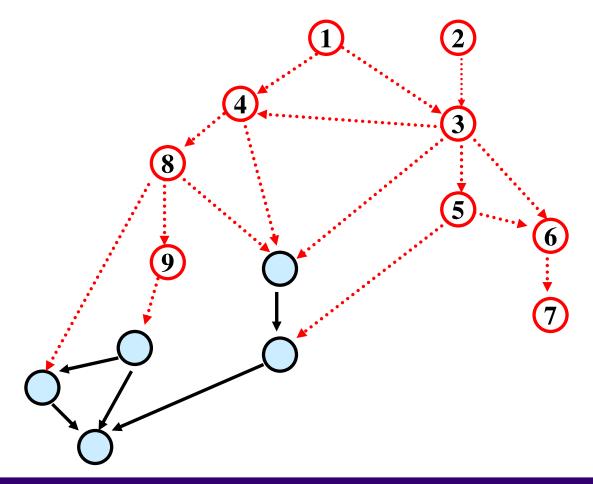


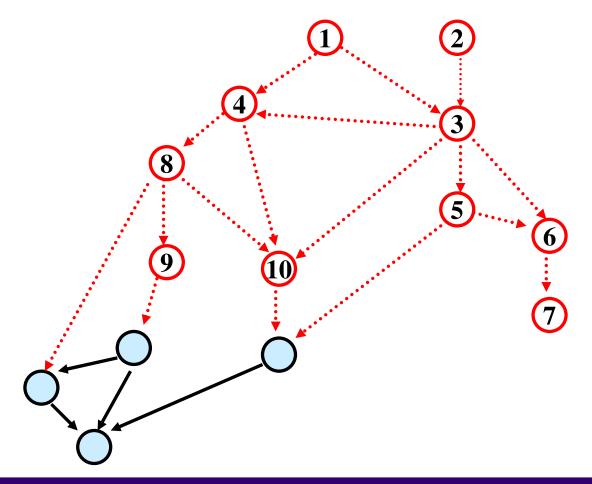


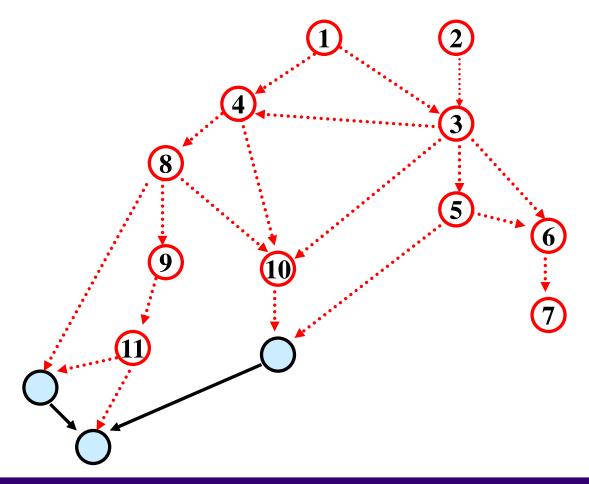


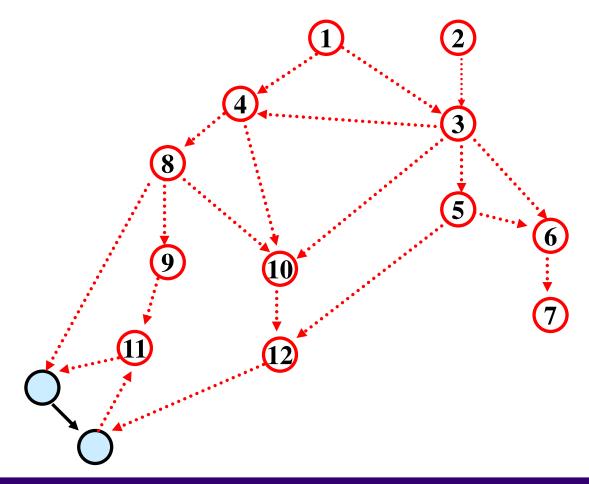


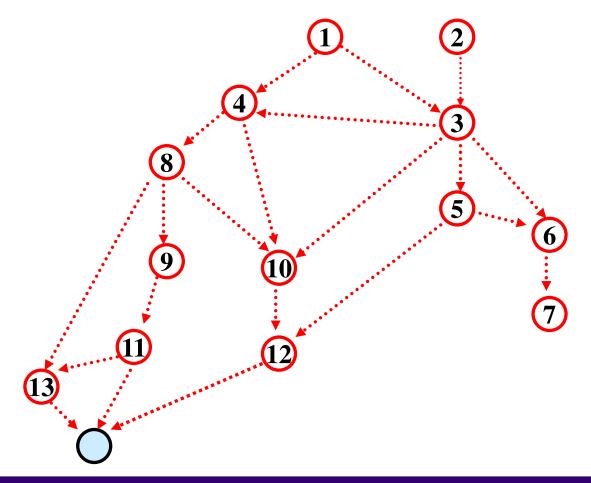


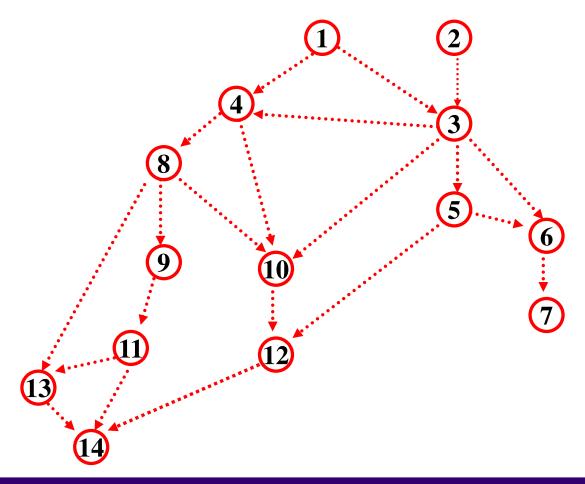












## **Implementing Topological Sort**

- Go through all edges, computing array with in-degree for each vertex O(m+n)
- Maintain a list of vertices of in-degree 0
- Remove any vertex in list and number it
- When a vertex is removed, decrease in-degree of each neighbor by 1
   and add them to the list if their degree drops to 0

Total cost: O(m+n)