

Lec11: Graph Algorithms I

Algorithm I
COMP319-003
Spring 2023

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Kyungpook National University (KNU)

Course Overview

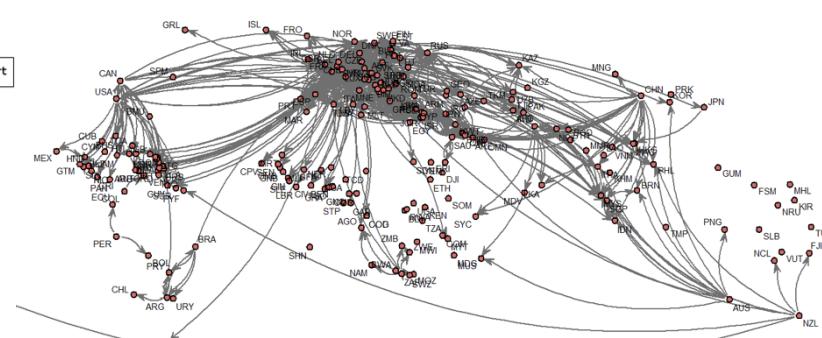
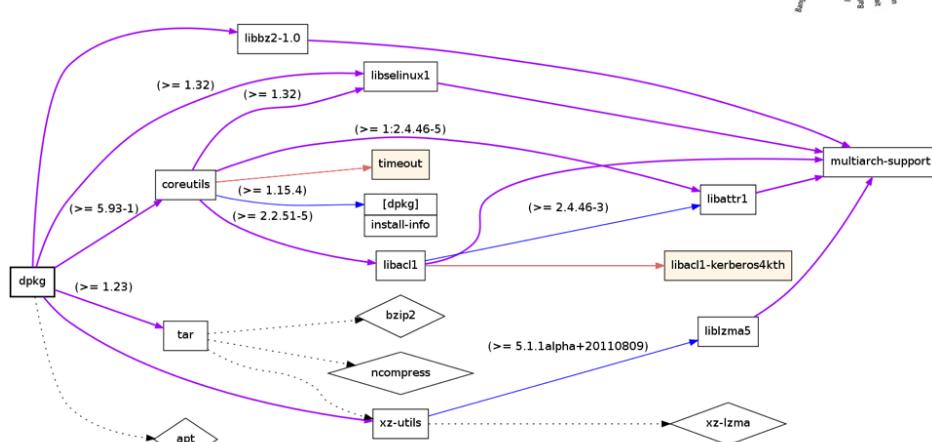
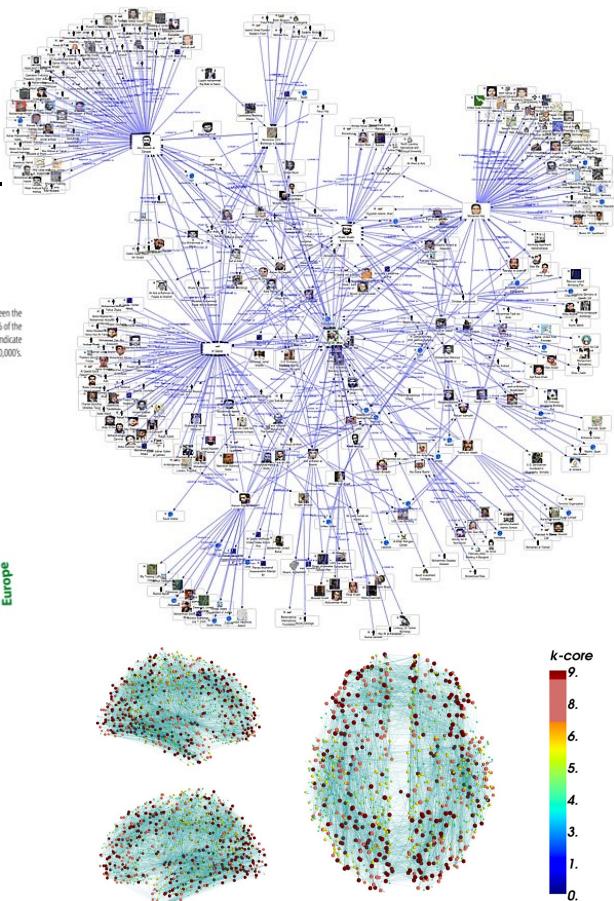
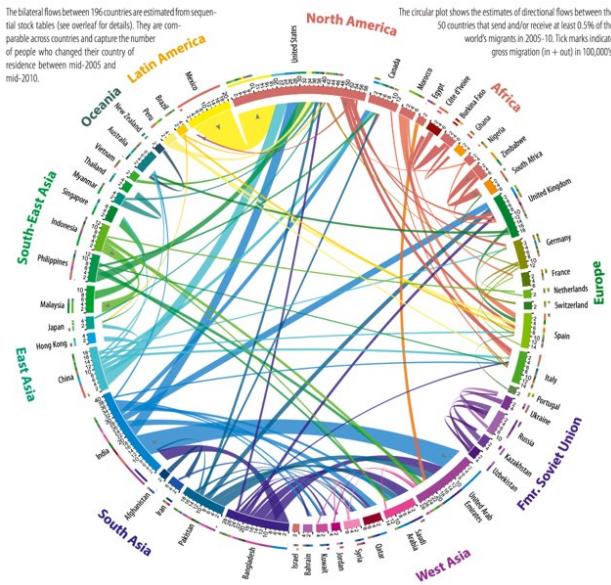
- Algorithmic Analysis
- Divide and Conquer
- Randomized Algorithms
- Tree Algorithms
- Hashing
- **Graph Algorithms**
- Dynamic Programming
- Greedy Algorithms
- NP-Completeness

| Outline

1. Graph Basics
 - Representations
 - DFS (Depth First Search)
 - BFS (Breadth First Search)
- *Reading: CLRS 22.1 – 22.4*

Examples of Graphs

- Web
 - Networks
 - dependencies
 - Neural connections
 - Social relations
 - ...

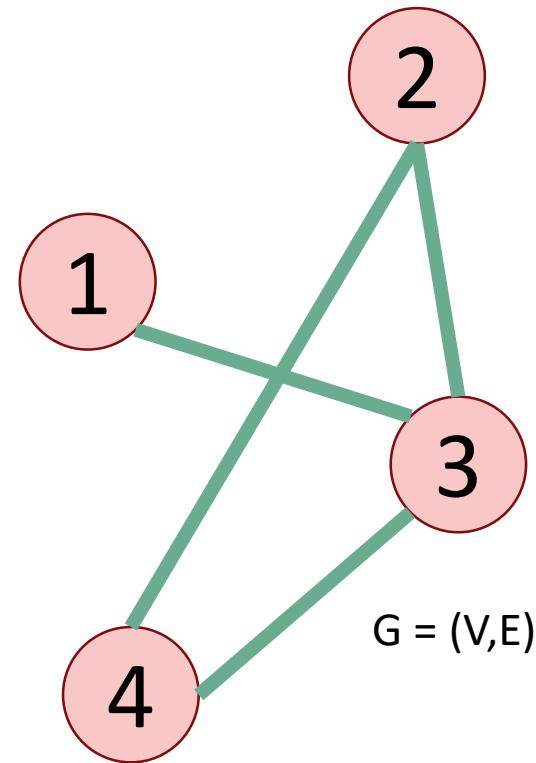


Graphs

- There are a lot of graphs.
- We want to answer questions about them.
 - **Efficient routing?**
 - Find the shortest path between two vertices
 - **Dependency?**
 - Find the topological ordering
 - **Community detection/clustering?**
 - Find strongly connected components
- This is what we'll do for the next several lectures.

Undirected Graphs

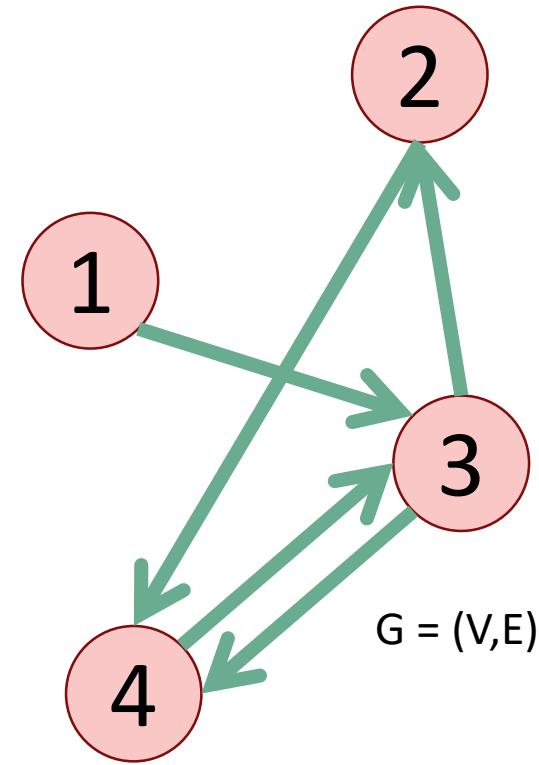
- Has vertices and edges
 - V is the set of vertices
 - E is the set of edges
 - Formally, a graph is $G = (V, E)$
- Example
 - $V = \{1, 2, 3, 4\}$
 - $E = \{ (1, 3), (2, 4), (3, 4), (2, 3) \}$



- The **degree** of vertex 4 is 2.
 - There are 2 edges coming out.
 - Vertex 4's **neighbors** are 2 and 3

Directed Graphs

- Has vertices and edges
 - V is the set of vertices
 - E is the set of **DIRECTED** edges
 - Formally, a graph is $G = (V, E)$
- Example
 - $V = \{1, 2, 3, 4\}$
 - $E = \{ (1, 3), (2, 4), (3, 4), (4, 3), (3, 2) \}$

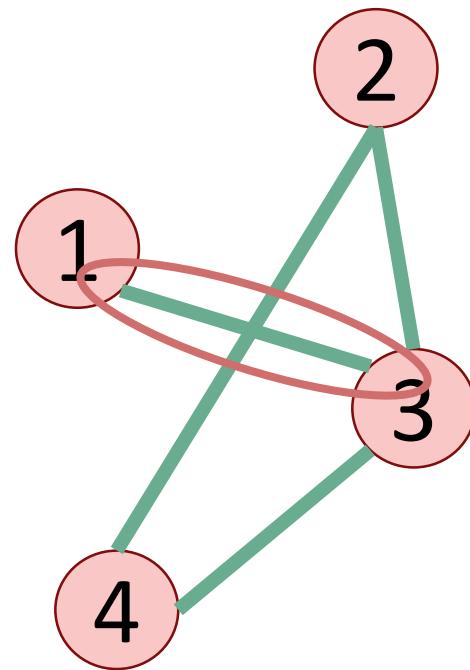


- The **in-degree** of vertex 4 is 2.
- The **out-degree** of vertex 4 is 1.
- Vertex 4's **incoming neighbors** are 2,3
- Vertex 4's **outgoing neighbor** is 3.

How do we represent graphs?

- Option 1: **adjacency matrix**

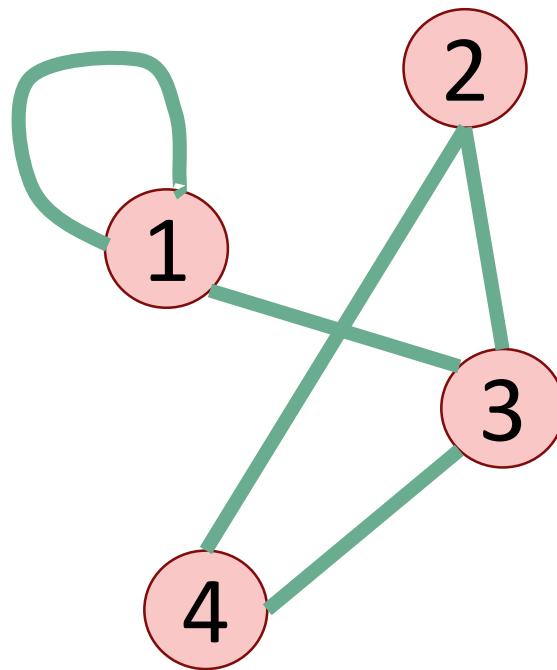
$$\begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix} \end{matrix}$$



How do we represent graphs?

- Option 1: **adjacency matrix**

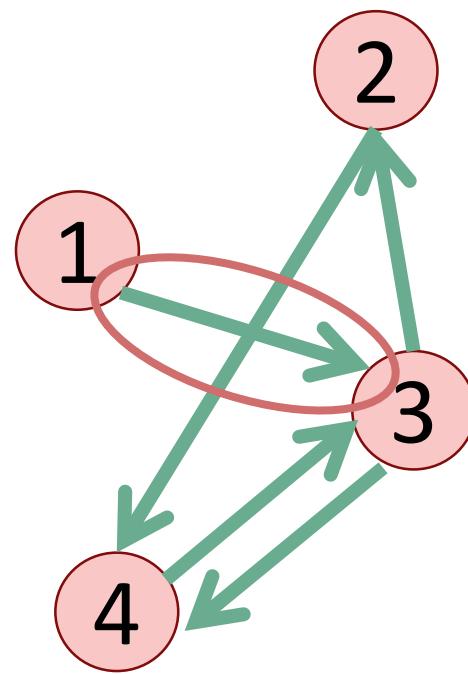
$$\begin{bmatrix} & 1 & 2 & 3 & 4 \\ 1 & 1 & 0 & 1 & 0 \\ 2 & 0 & 0 & 1 & 1 \\ 3 & 1 & 1 & 0 & 1 \\ 4 & 0 & 1 & 1 & 0 \end{bmatrix}$$



How do we represent graphs?

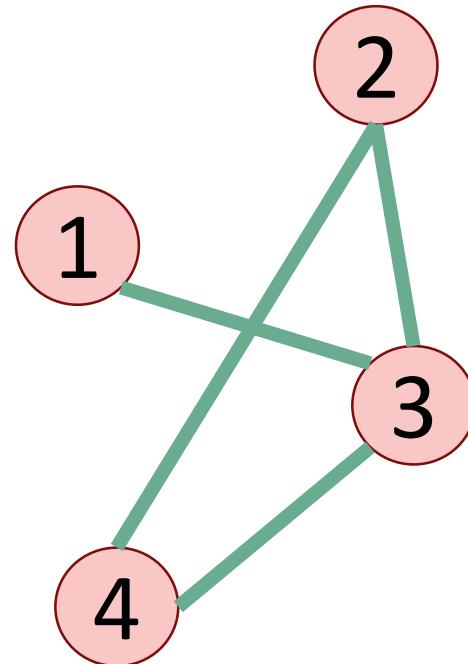
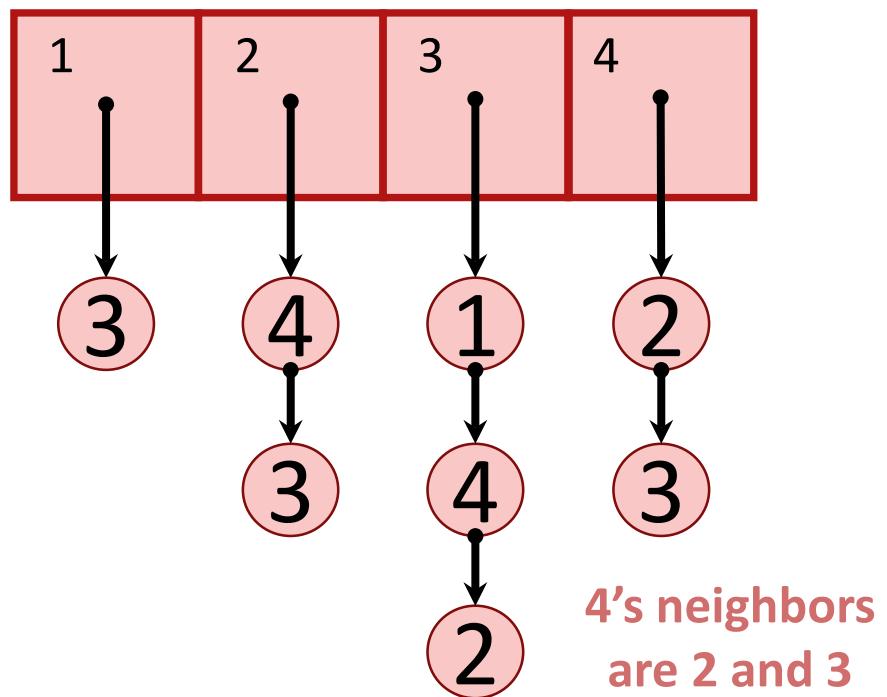
- Option 1: **adjacency matrix**

Destination					
		1	2	3	4
Source	1	0	0	1	0
	2	0	0	1	1
	3	0	1	0	1
	4	0	1	1	0



How do we represent graphs?

- Option 2: **adjacency lists**.



In either case

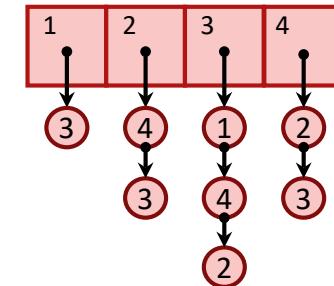
- Vertices can store other information
 - Attributes (name, IP address, ...)
 - Helper information for algorithms that we will perform on the graph
- Want to be able to do the following operations:
 - **Edge Membership**: Is edge e in E?
 - **Neighbor Query**: What are the neighbors of vertex v?

Trade-offs

Generally better for **sparse** graphs (where $m \ll n^2$)

Say there are n vertices and m edges.

$$\begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix}$$



Edge membership
Is $e = \{v,w\}$ in E ?

$O(1)$

$O(\deg(v))$ or
 $O(\deg(w))$

Neighbor query
Give me a list of v 's neighbors.

$O(n)$

$O(\deg(v))$

Space requirements

$O(n^2)$

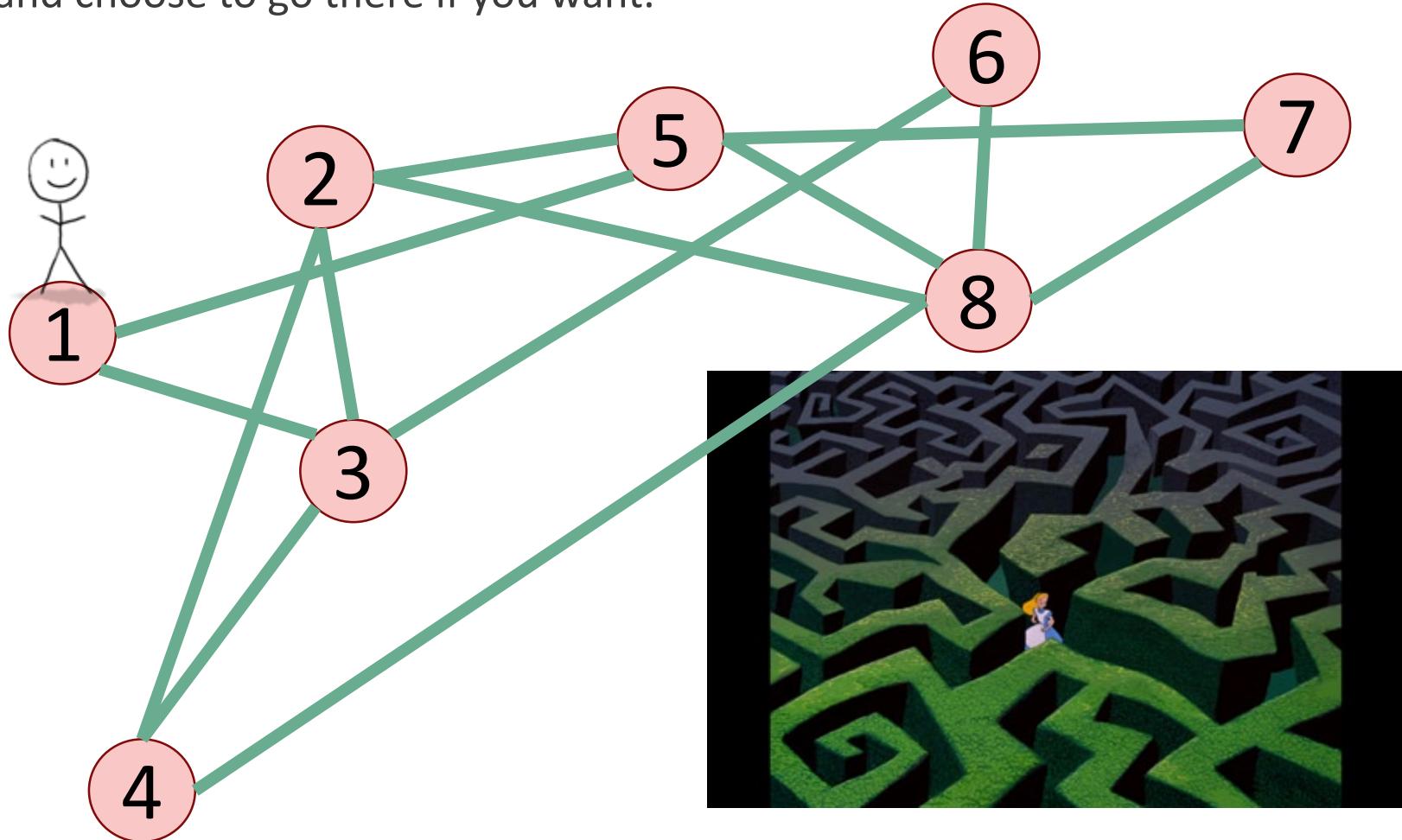
$O(n + m)$

We'll assume this representation for the rest of the class

Depth First Search

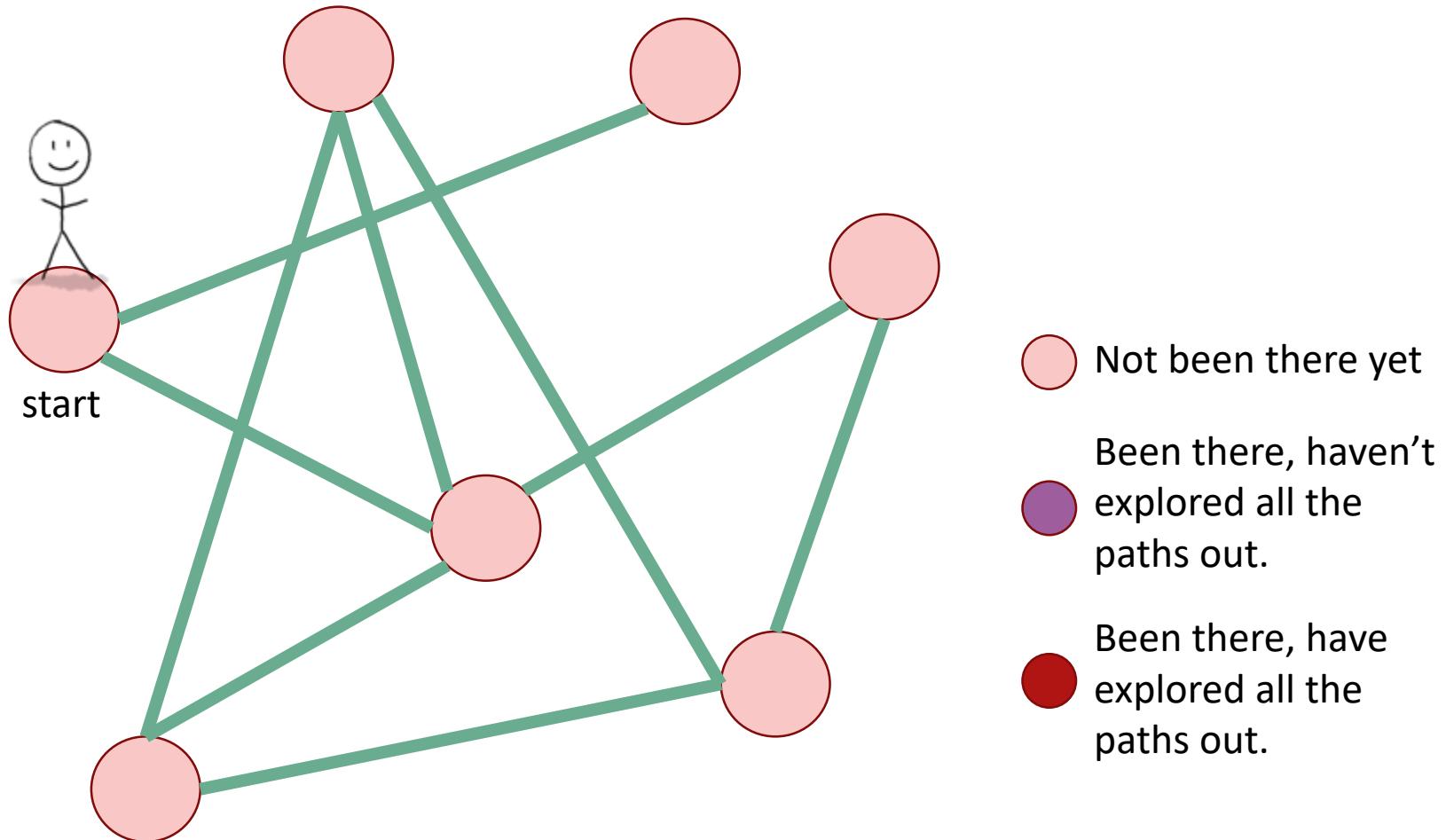
How do we explore a graph?

- At each node, you can get a list of neighbors, and choose to go there if you want.



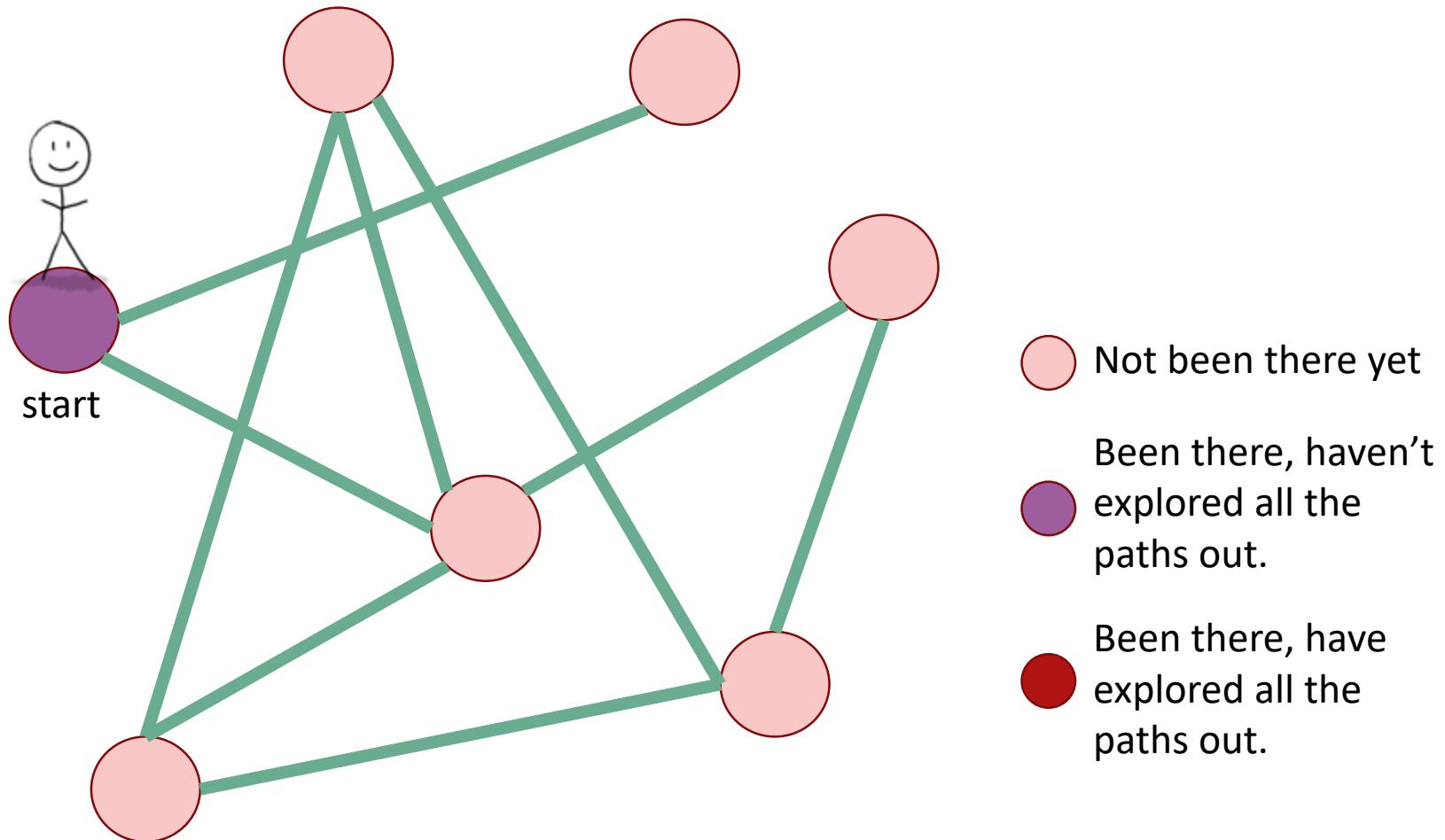
Depth First Search

- Searching “deeper” in the graph whenever possible.



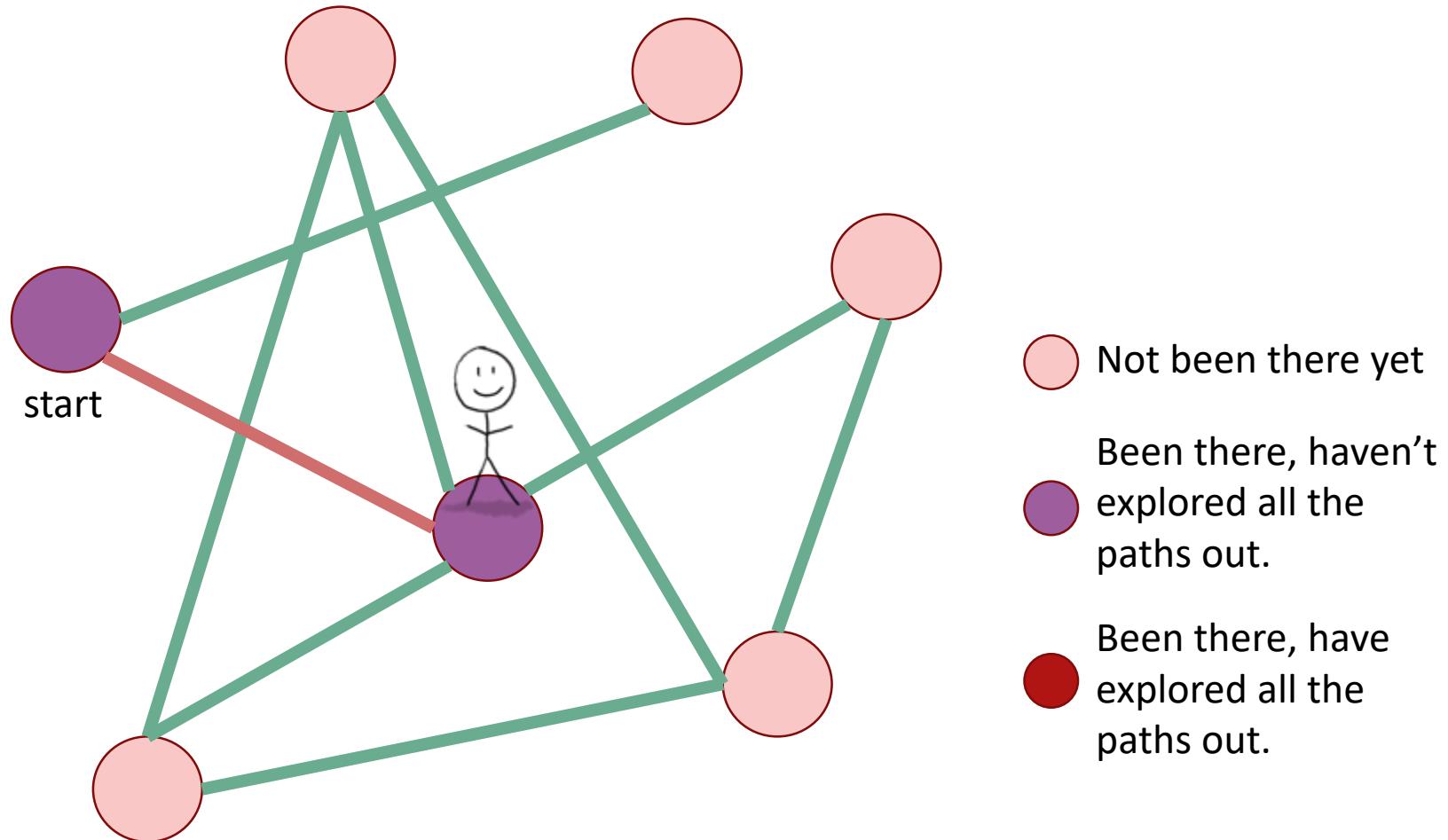
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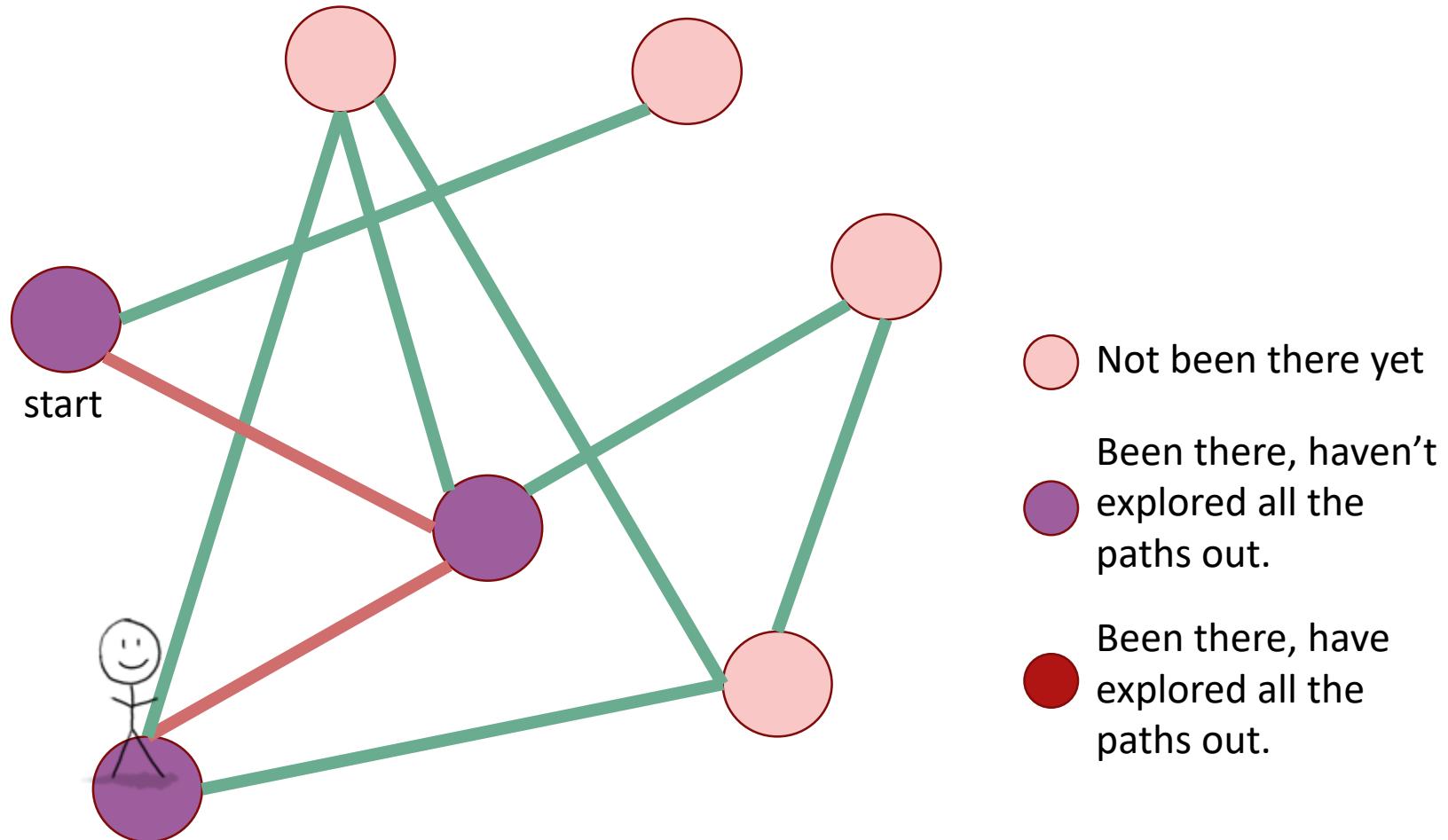
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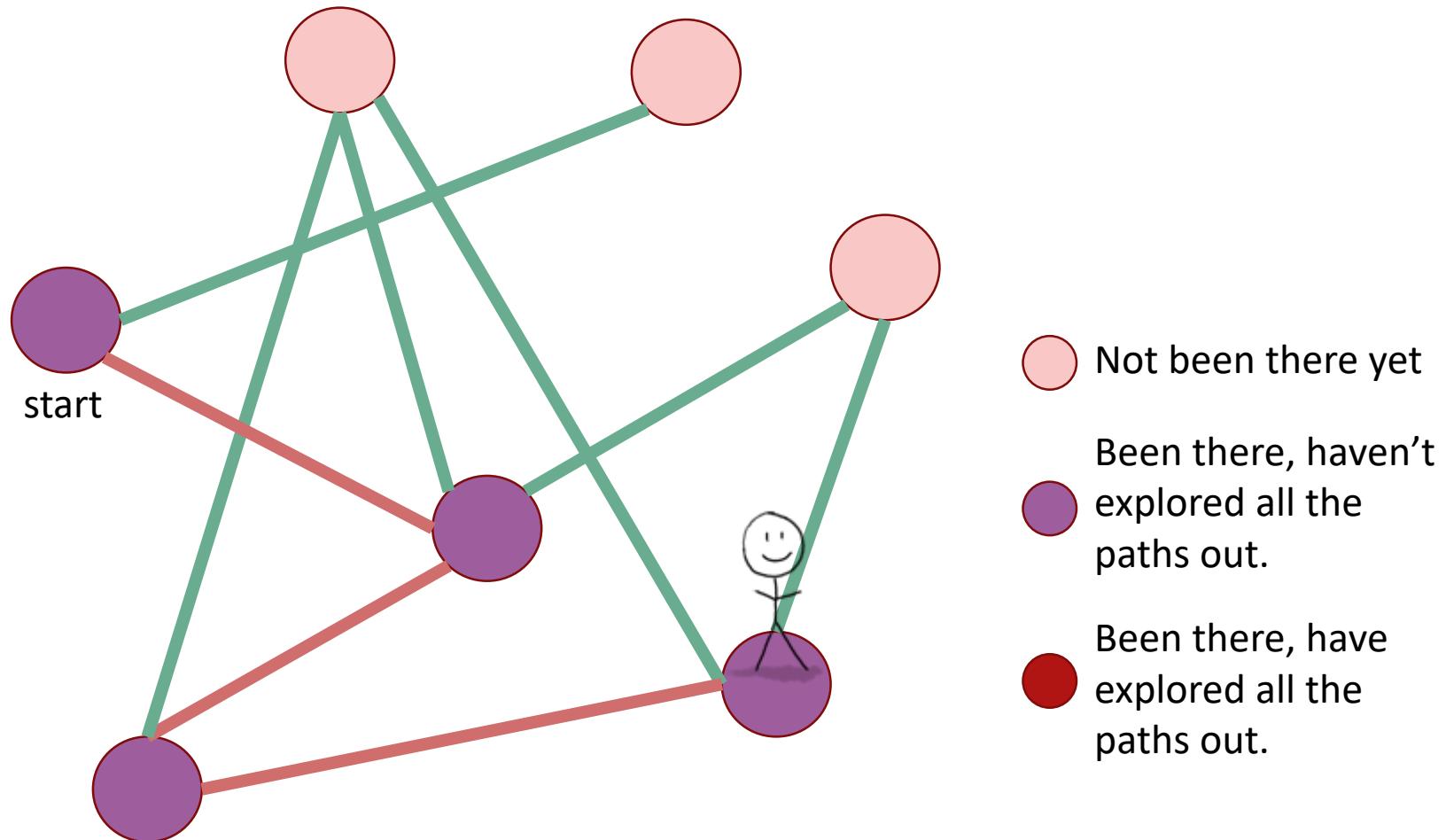
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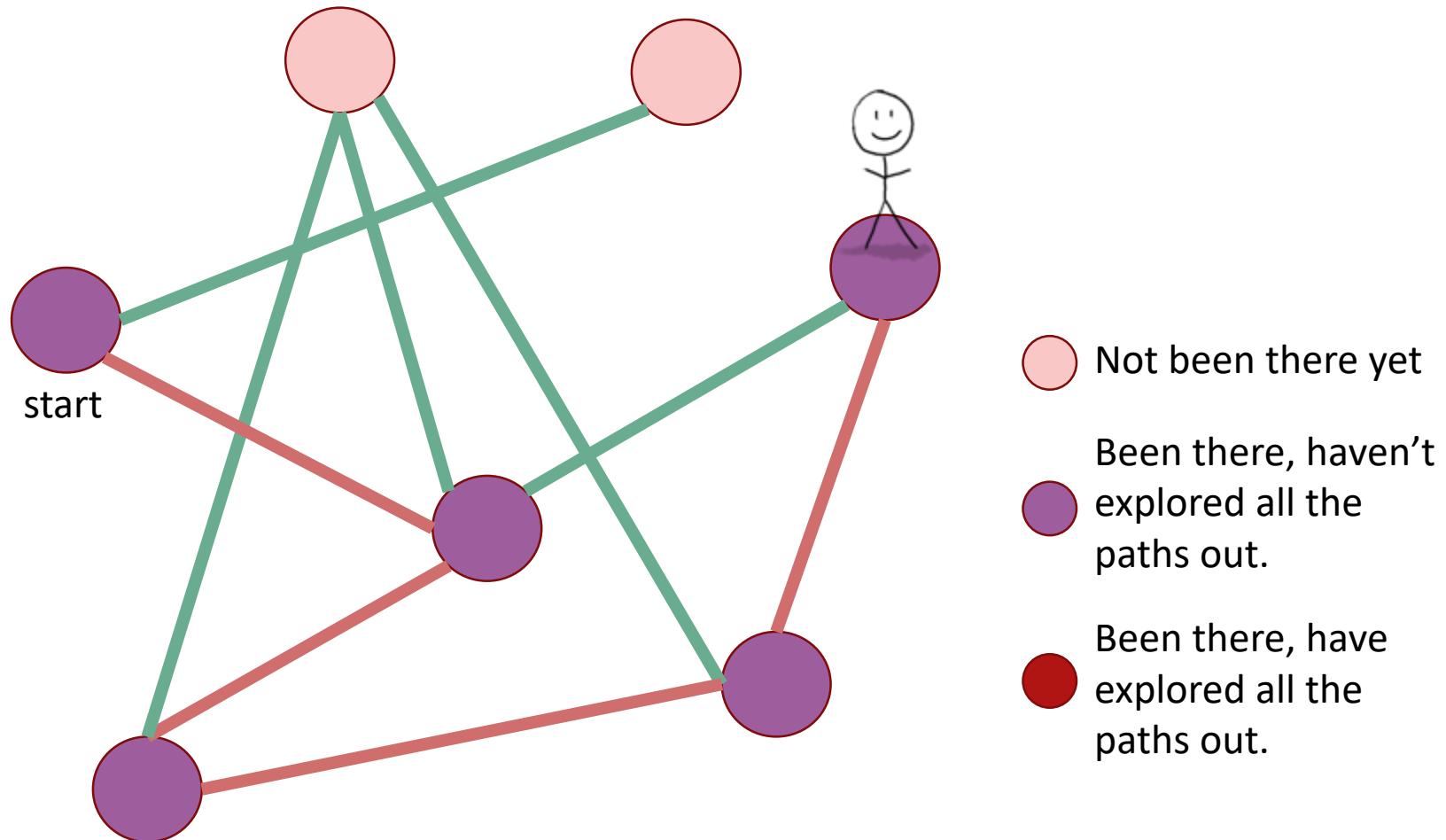
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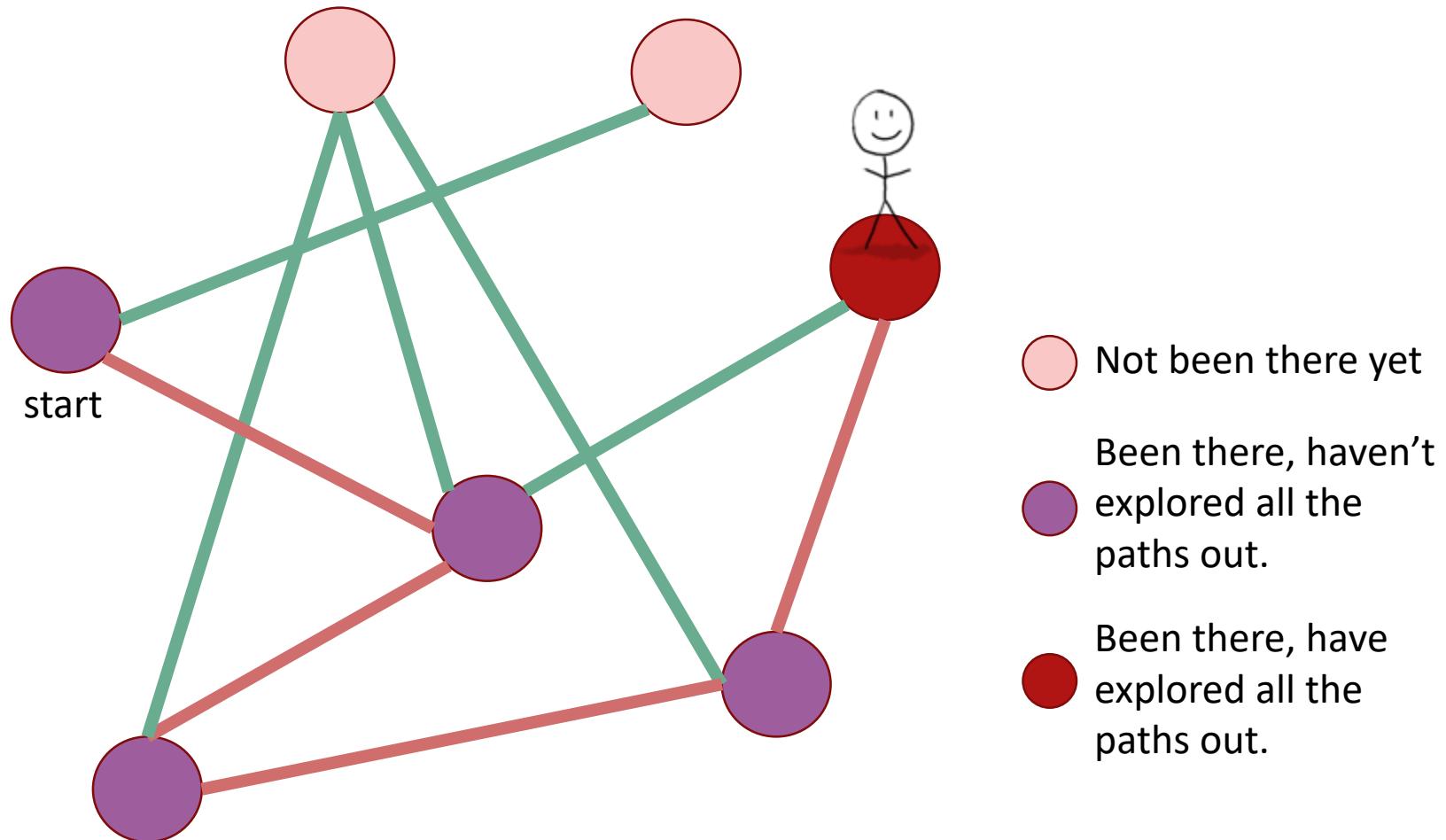
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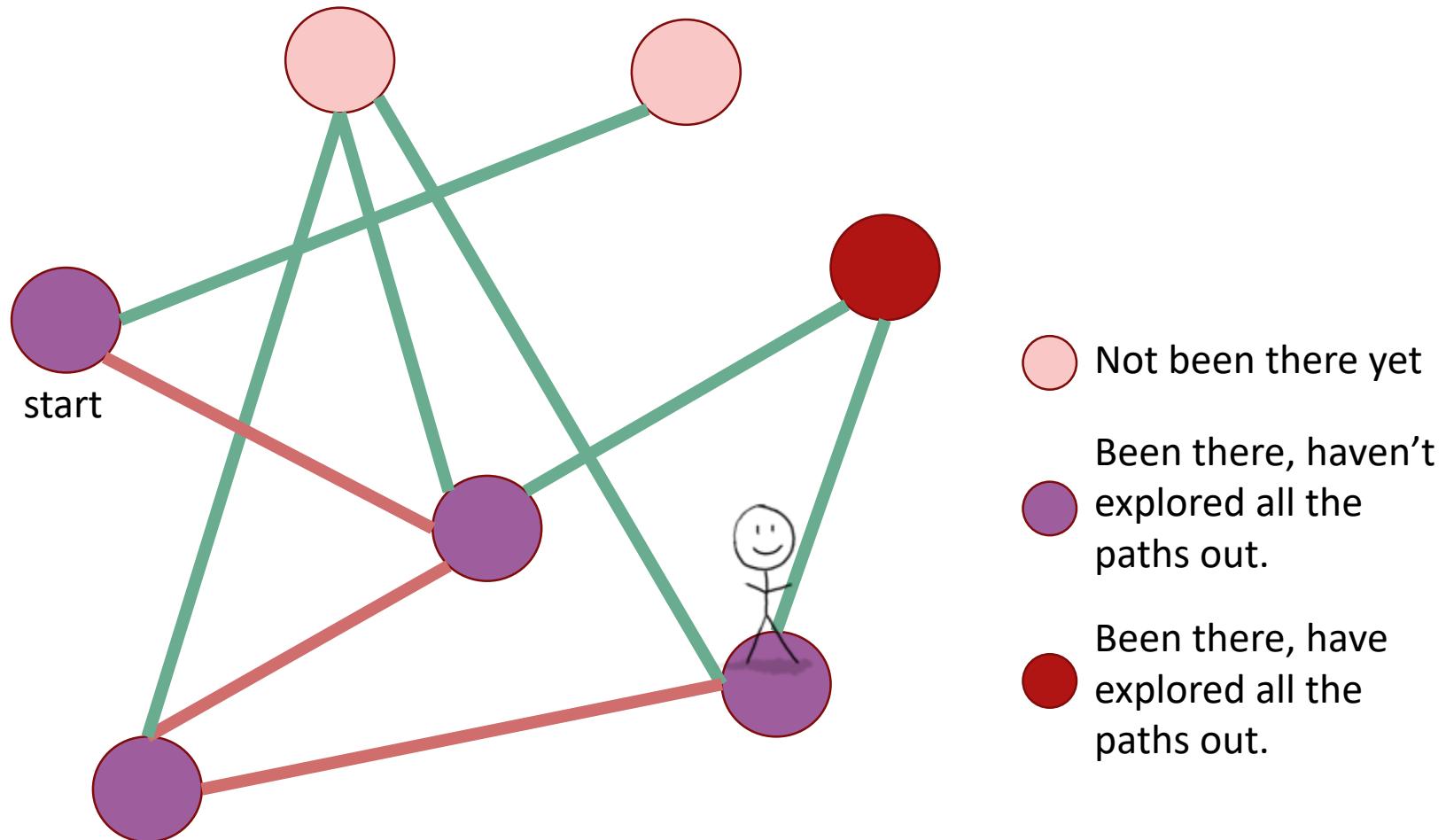
Depth First Search

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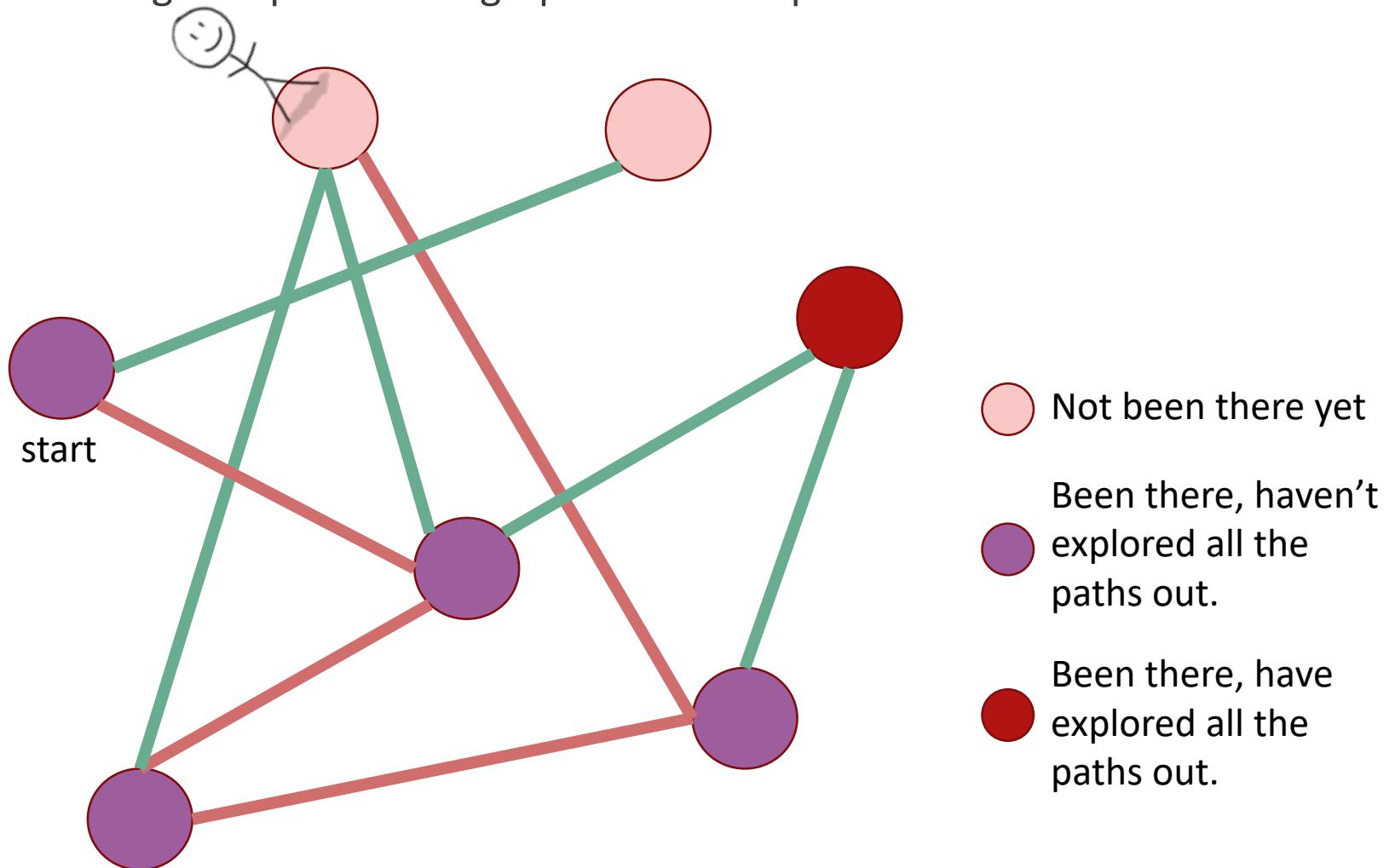
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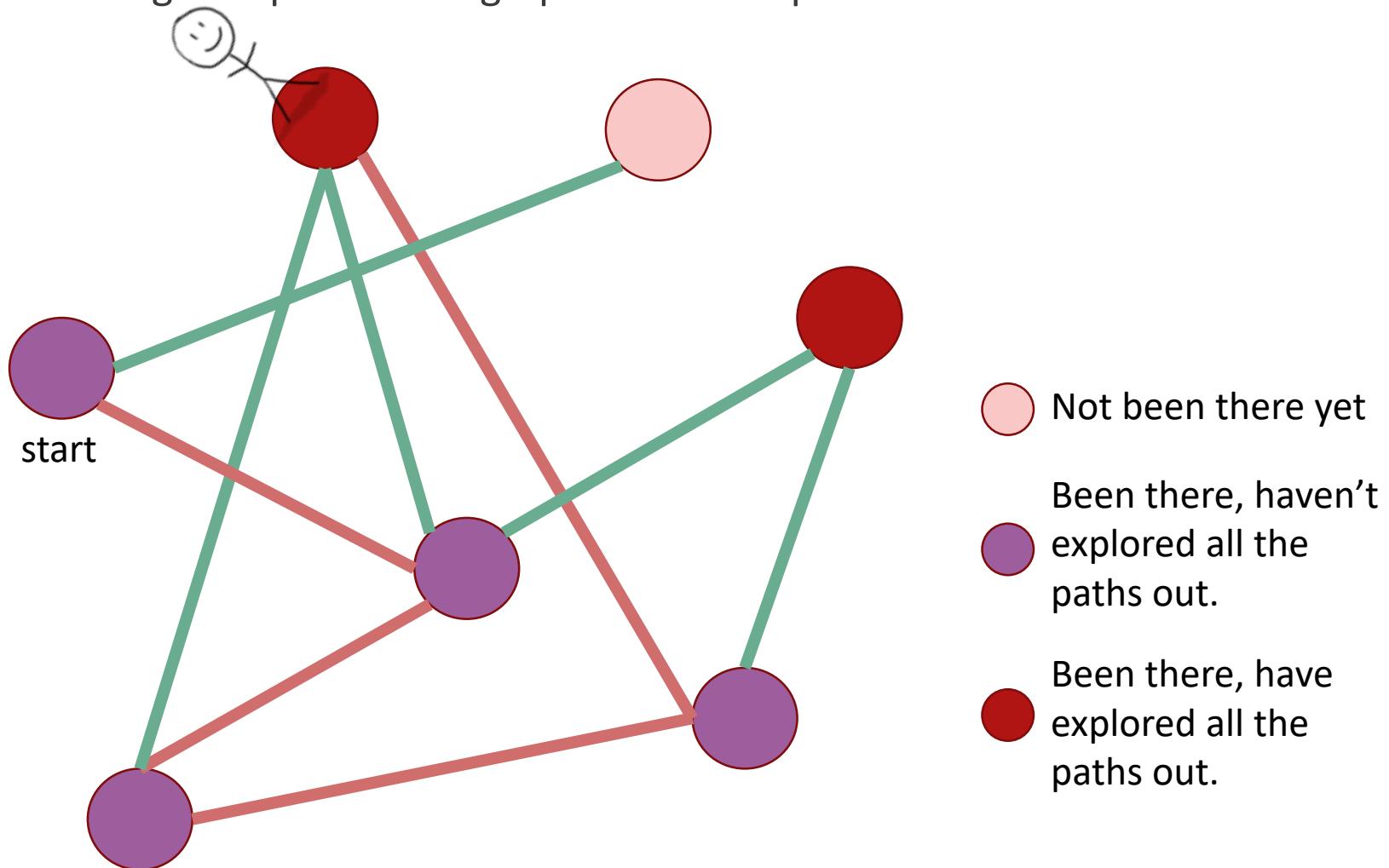
Depth First Search

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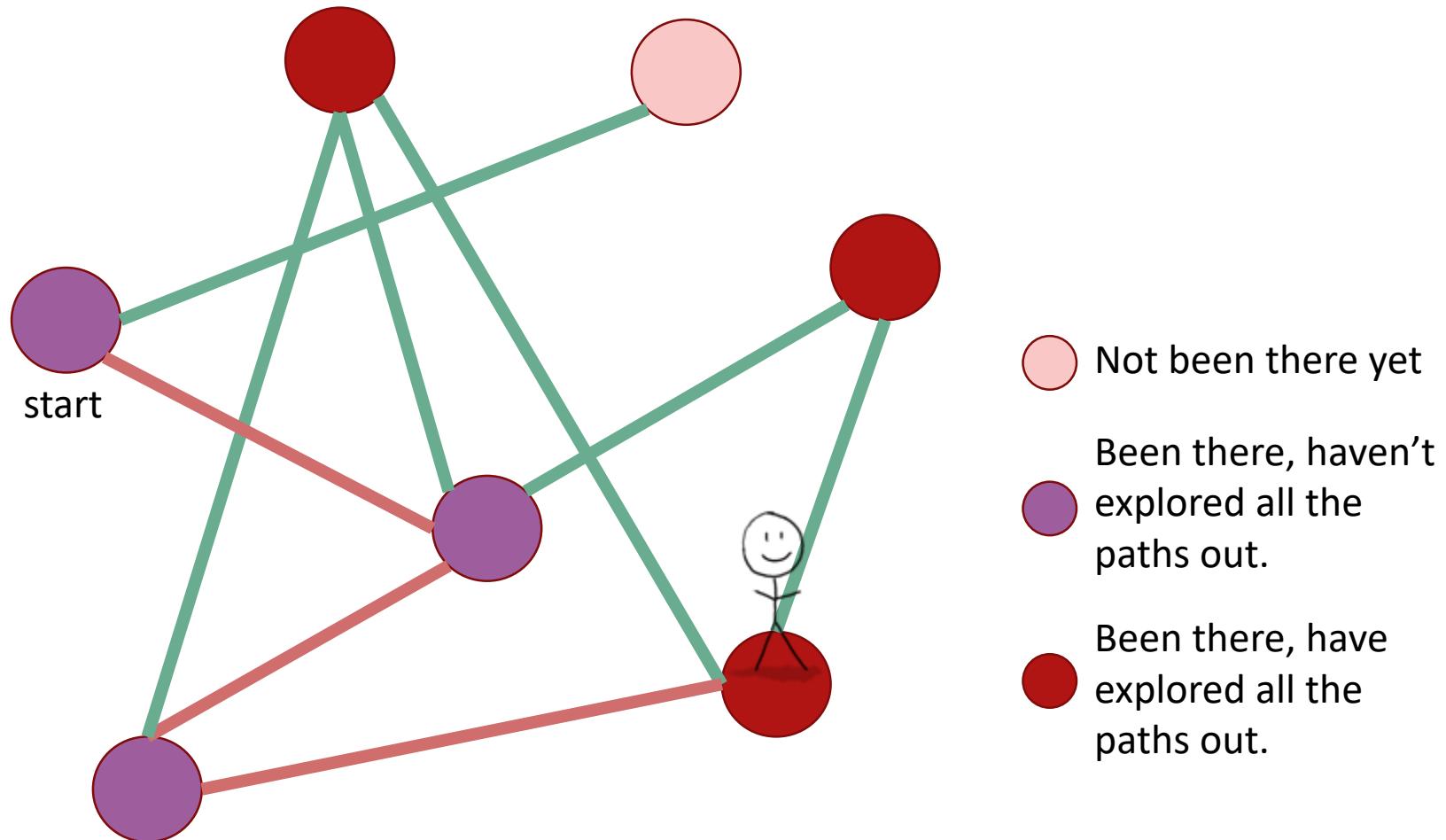
Depth First Search

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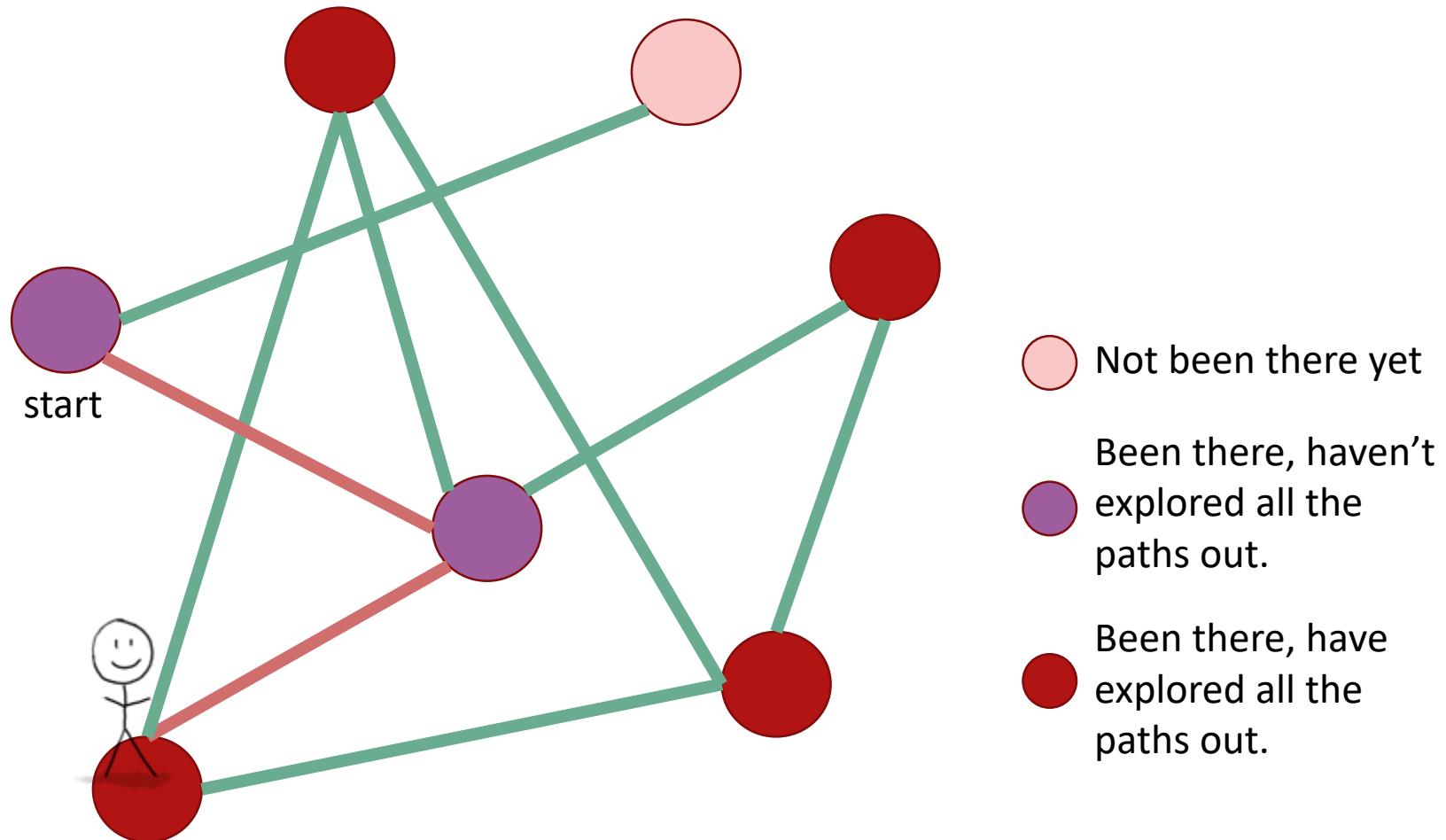
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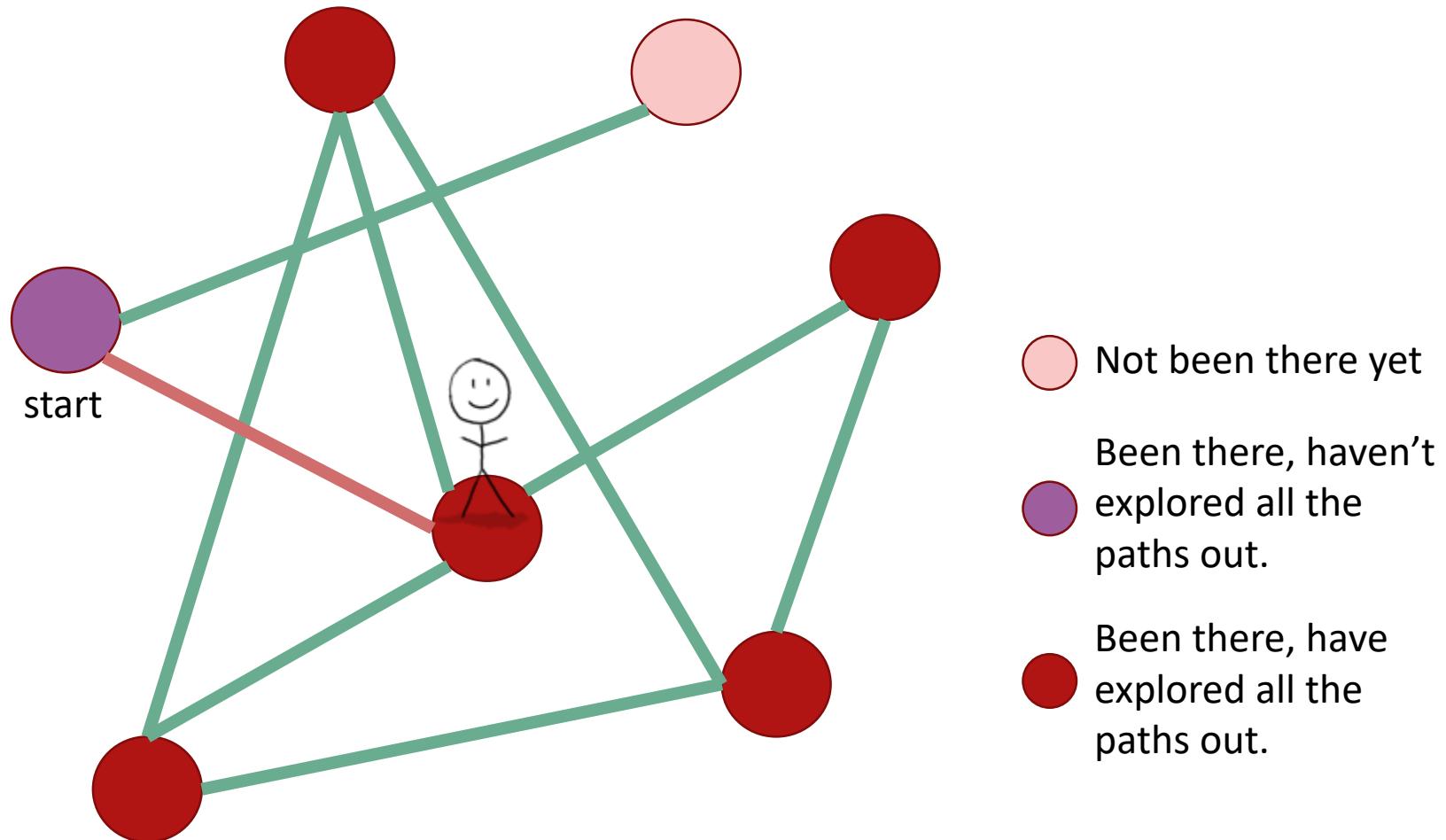
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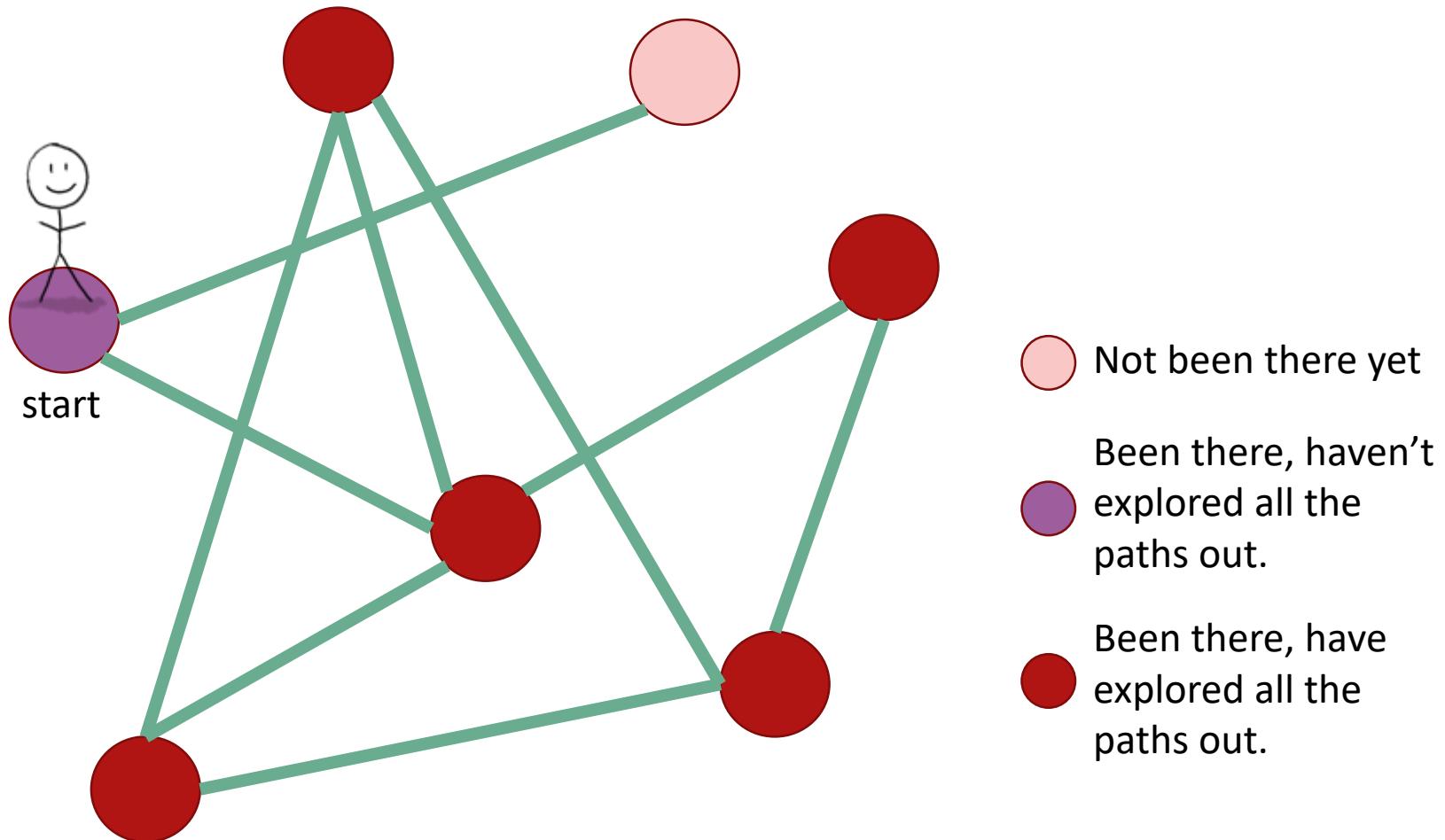
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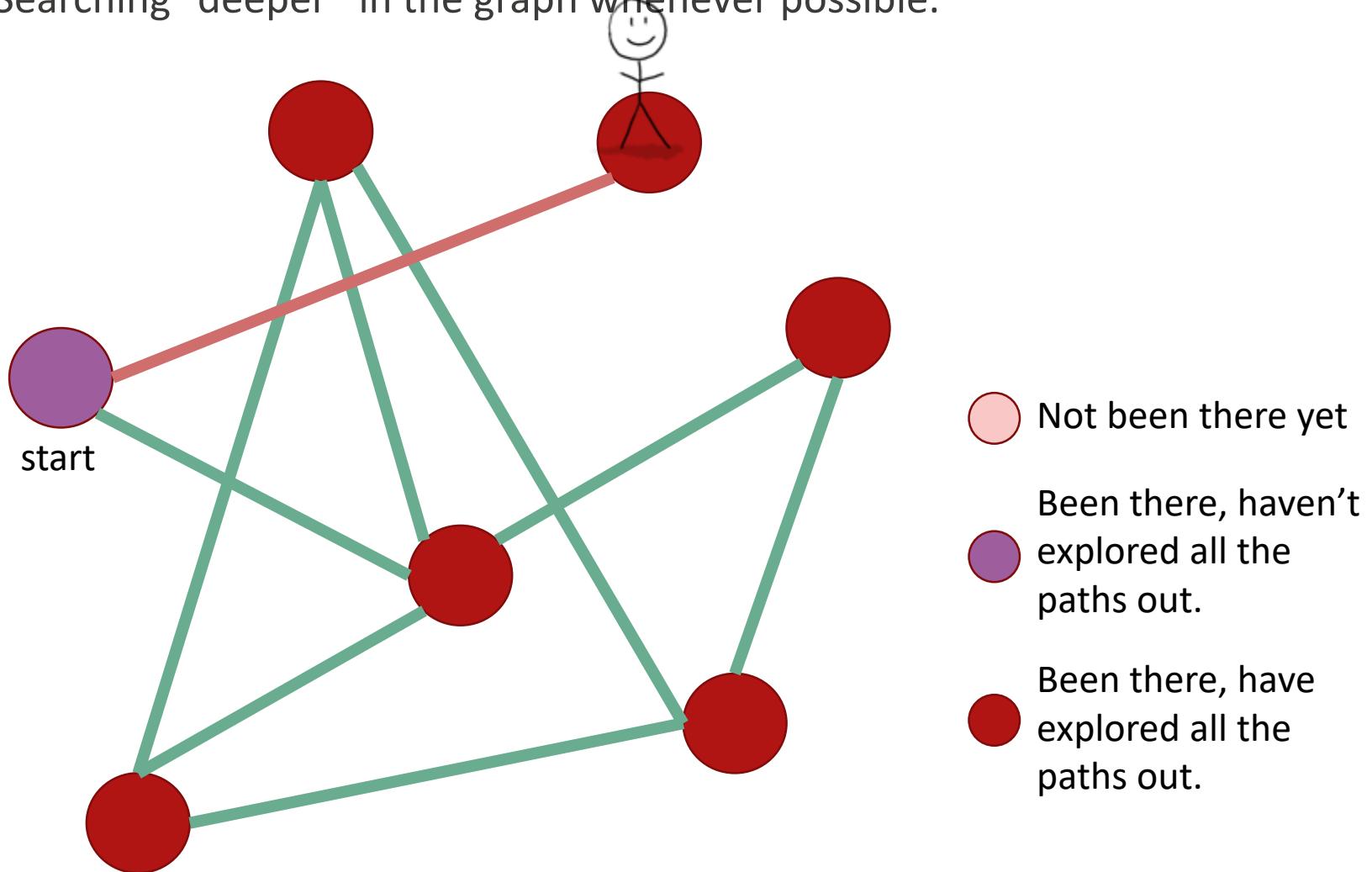
Depth First Search

- Searching “deeper” in the graph whenever possible.



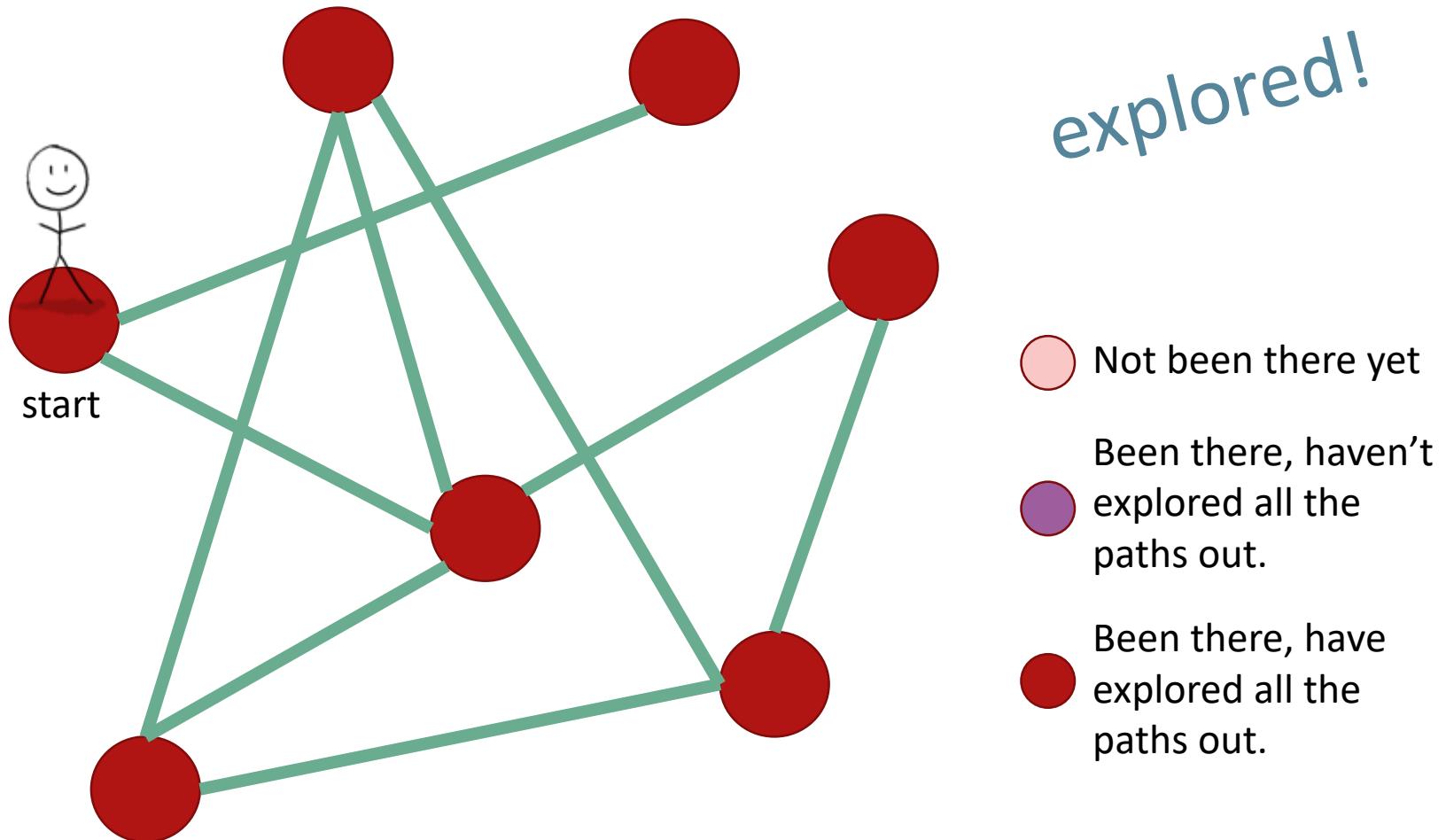
Depth First Search

- Searching “deeper” in the graph whenever possible.



Depth First Search

- Searching “deeper” in the graph whenever possible.



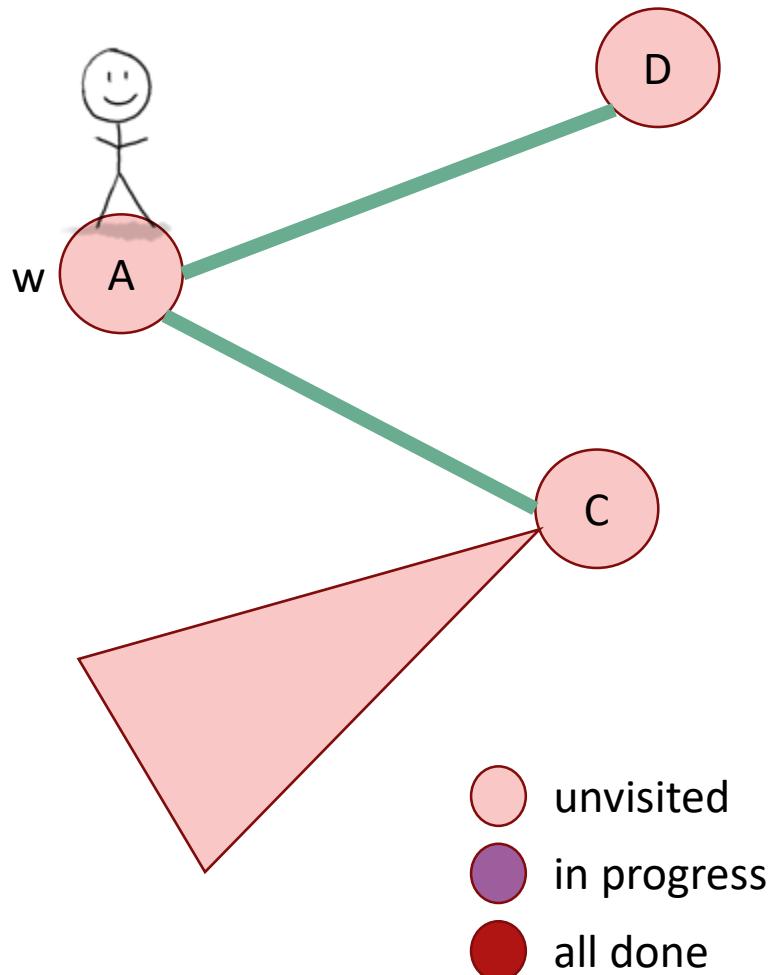
Depth First Search

- Searching “deeper” in the graph whenever possible.
- Each vertex keeps track of whether it is:
 - Unvisited 
 - In progress 
 - All done 
- Each vertex will also keep track of:
 - The time we **first enter it**.
 - The time we finish with it and mark it **all done**.
 - Note:
 - This is not the only way to write DFS.
 - This way has more bookkeeping. (It will be useful later!)



Depth First Search

currentTime = 0

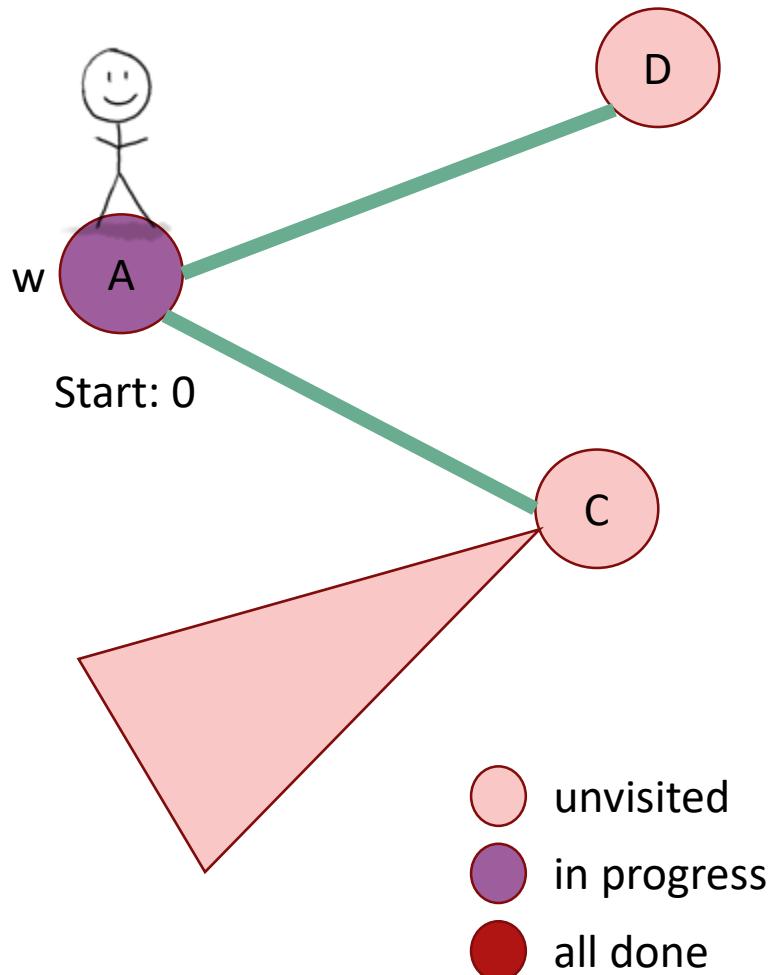


w: *input vertex*

```
DFS(w, currentTime):  
    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
    w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 0

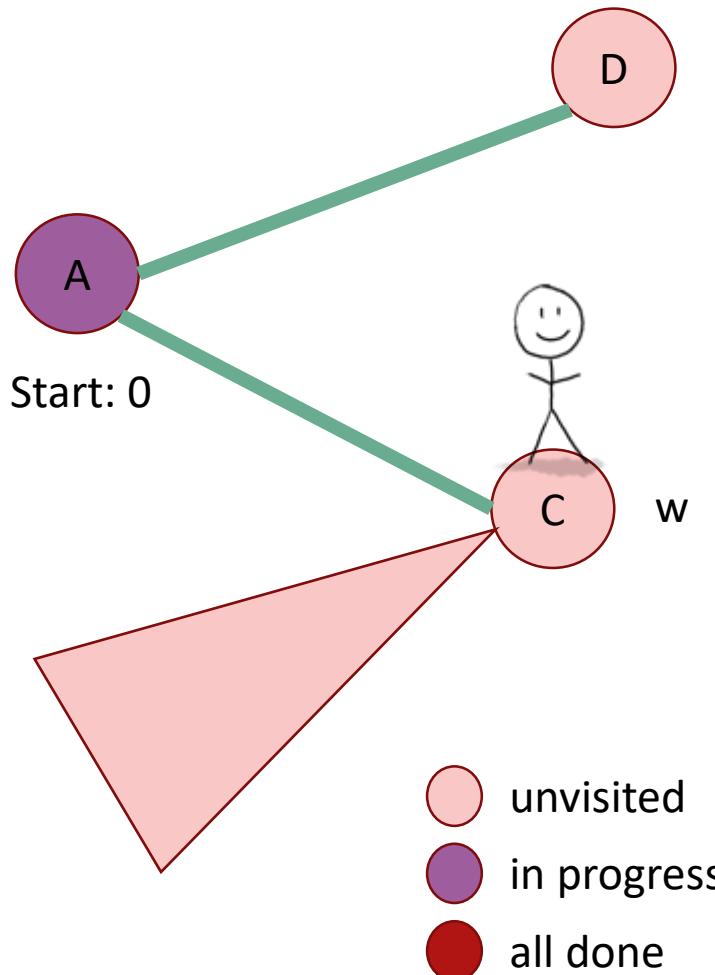


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    w.startTime = currentTime  
    currentTime += 1  
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        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
    w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 1

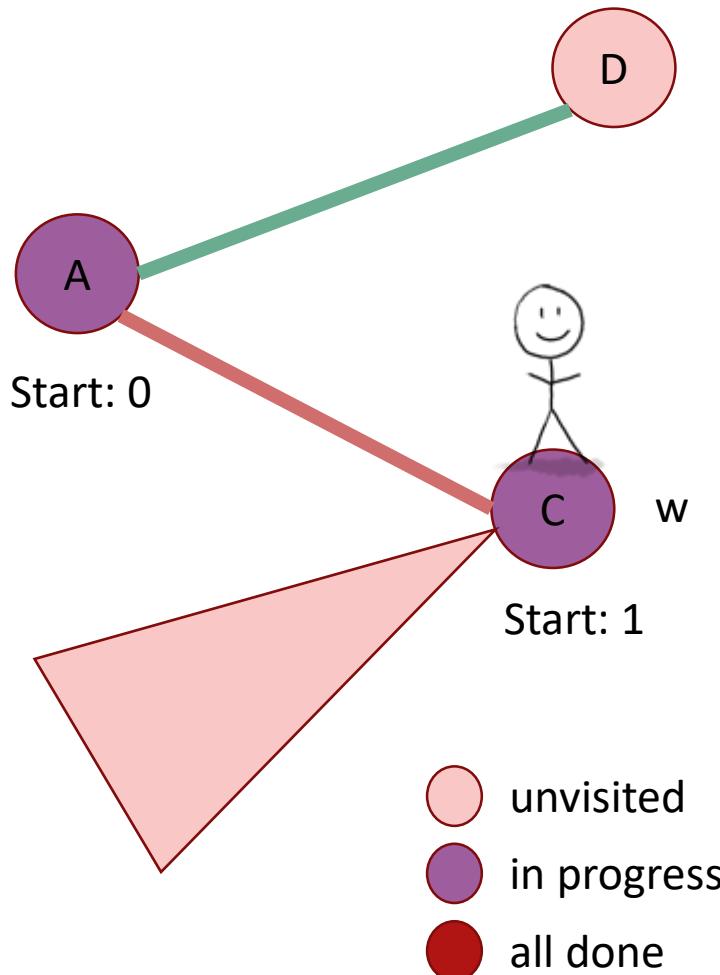


w: *input vertex*

```
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    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
        w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 1

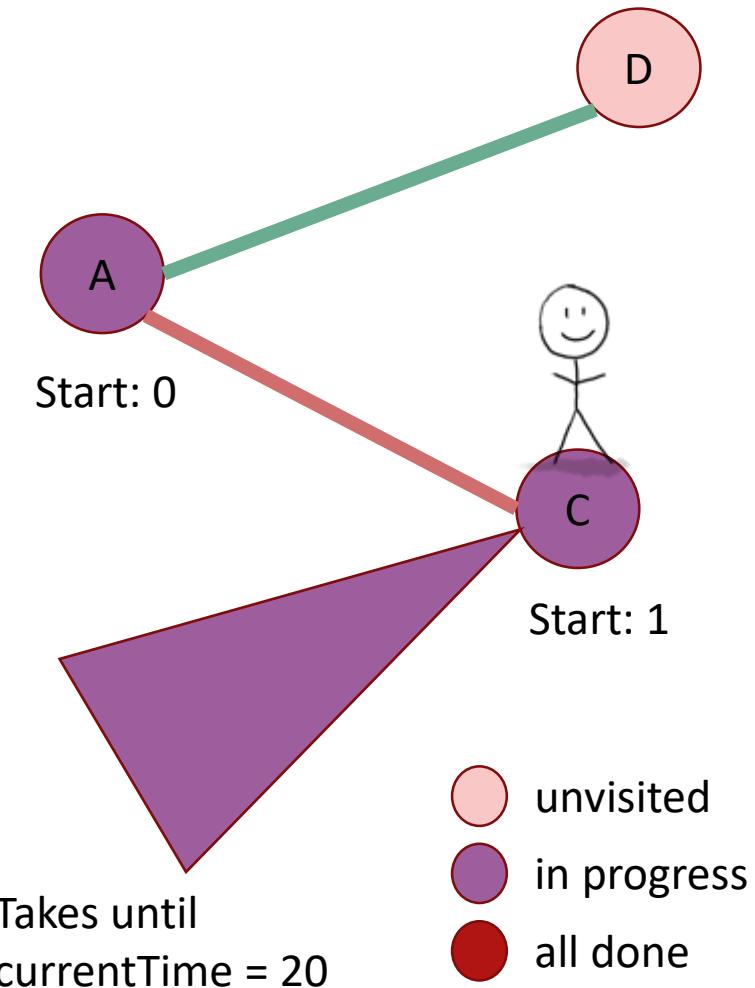


w: *input vertex*

```
DFS(w, currentTime):  
    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
        w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 20

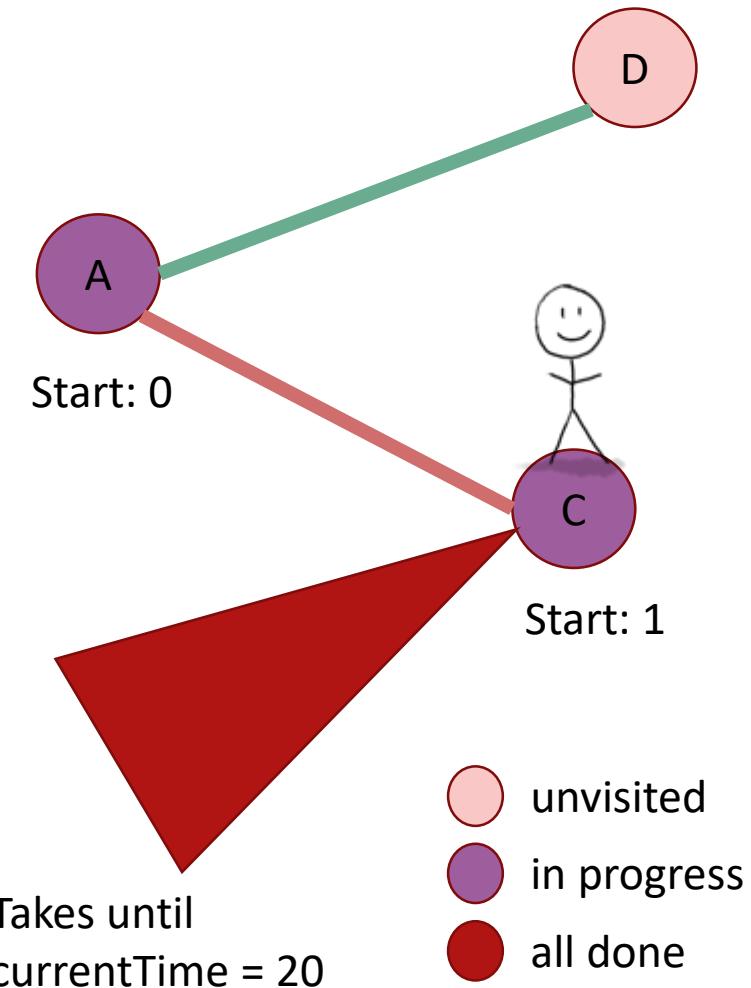


w: *input vertex*

```
DFS(w, currentTime):  
    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
        w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 21

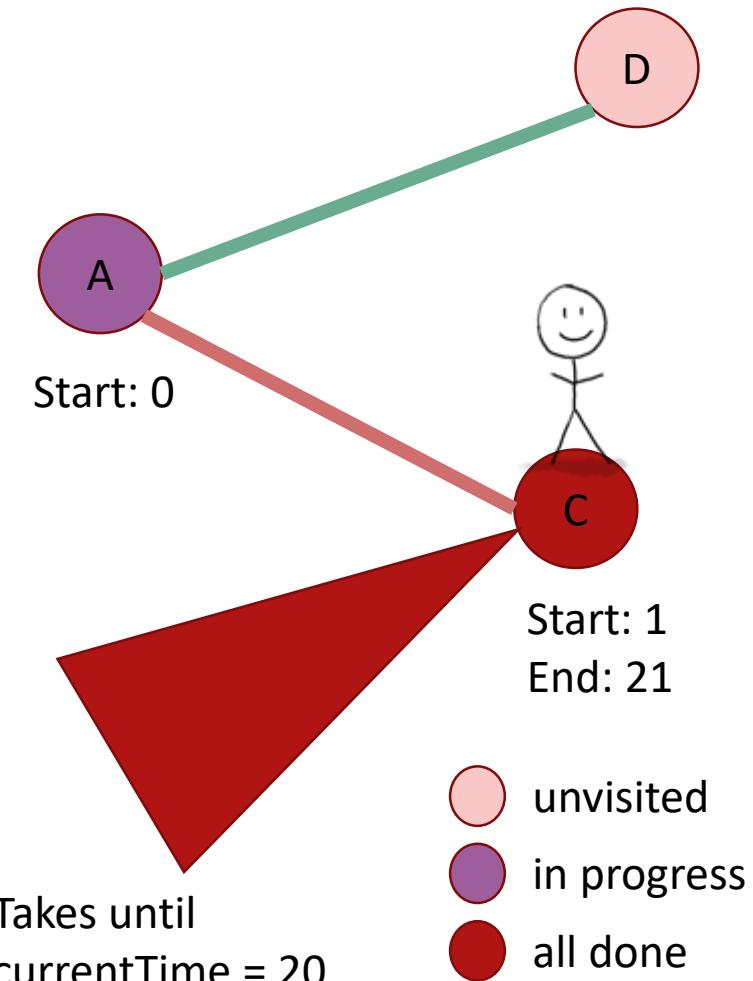


w: *input vertex*

```
DFS(w, currentTime):  
    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
        w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 21

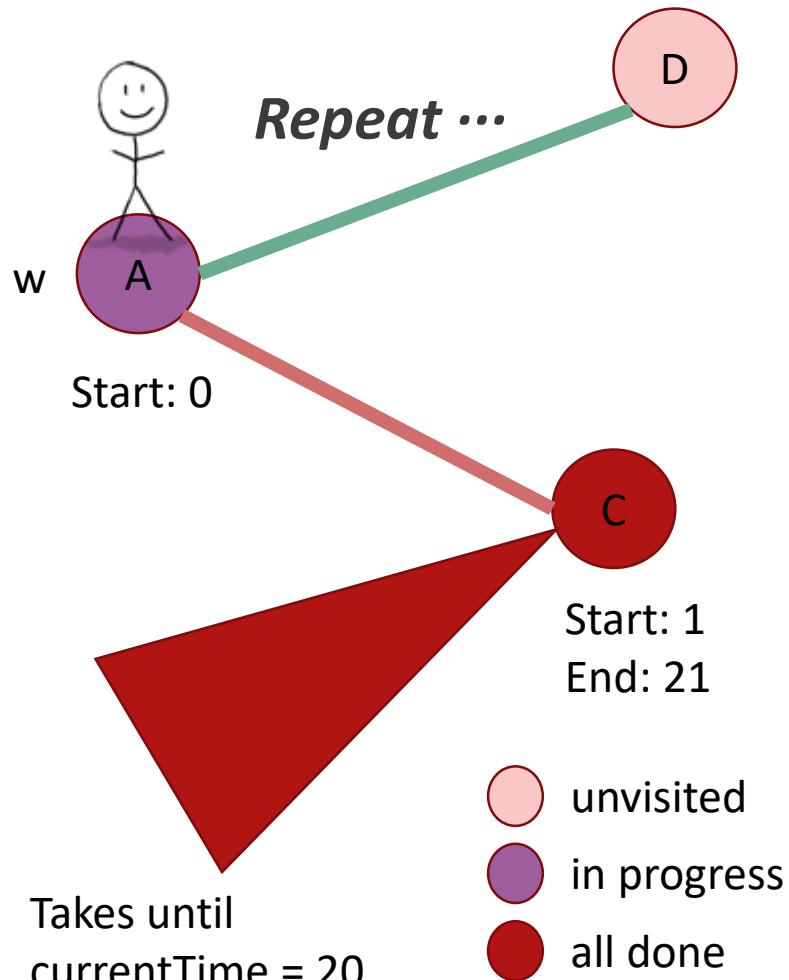


w: *input vertex*

```
DFS(w, currentTime):  
    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
        w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

Depth First Search

currentTime = 22

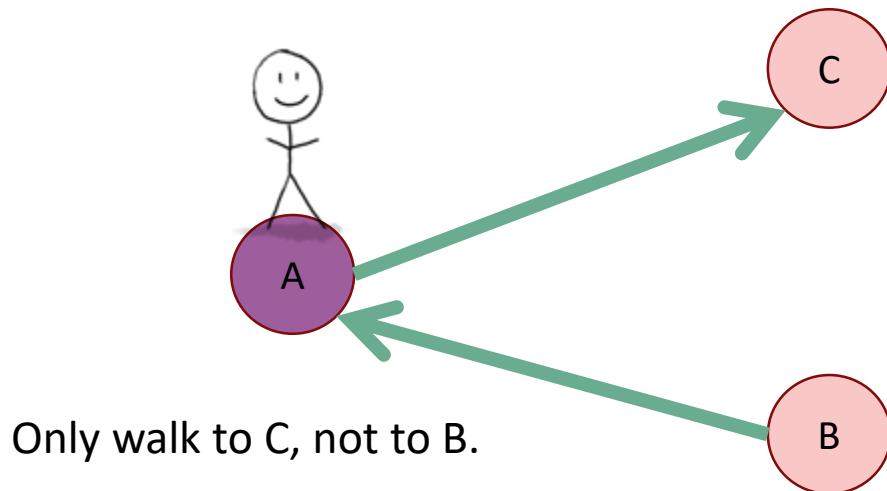


w: input vertex

```
DFS(w, currentTime):  
    w.startTime = currentTime  
    currentTime += 1  
    Mark w as in progress  
    for v in w.neighbors:  
        if v is unvisited:  
            currentTime =  
                DFS(v, currentTime)  
            currentTime += 1  
    w.finishTime = currentTime  
    Mark w as all done  
    return currentTime
```

You check:

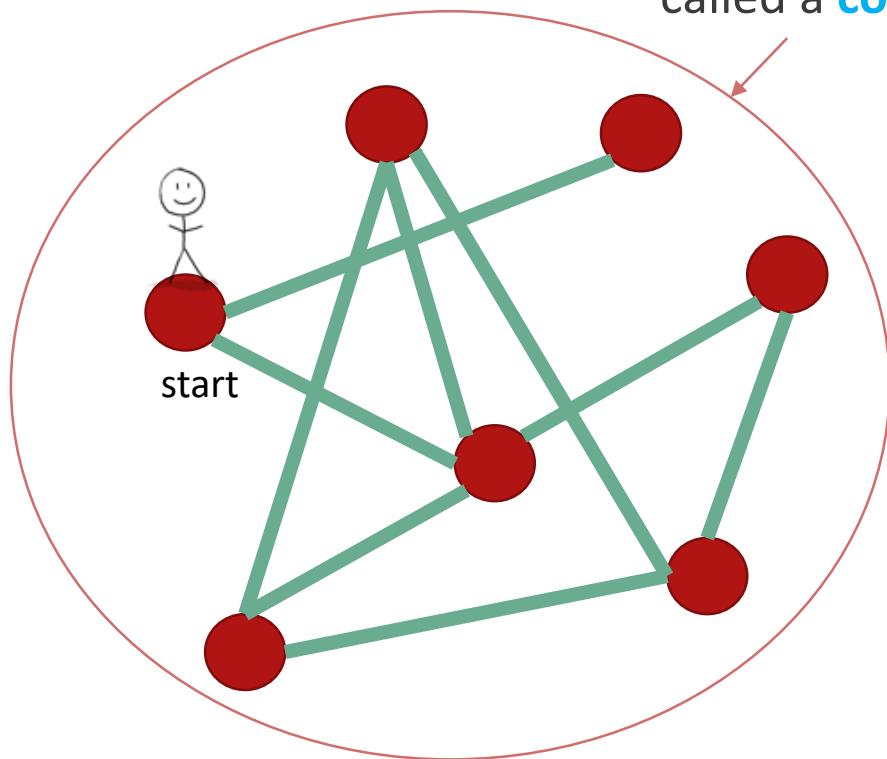
- DFS works fine on directed graphs too!



Depth First Search

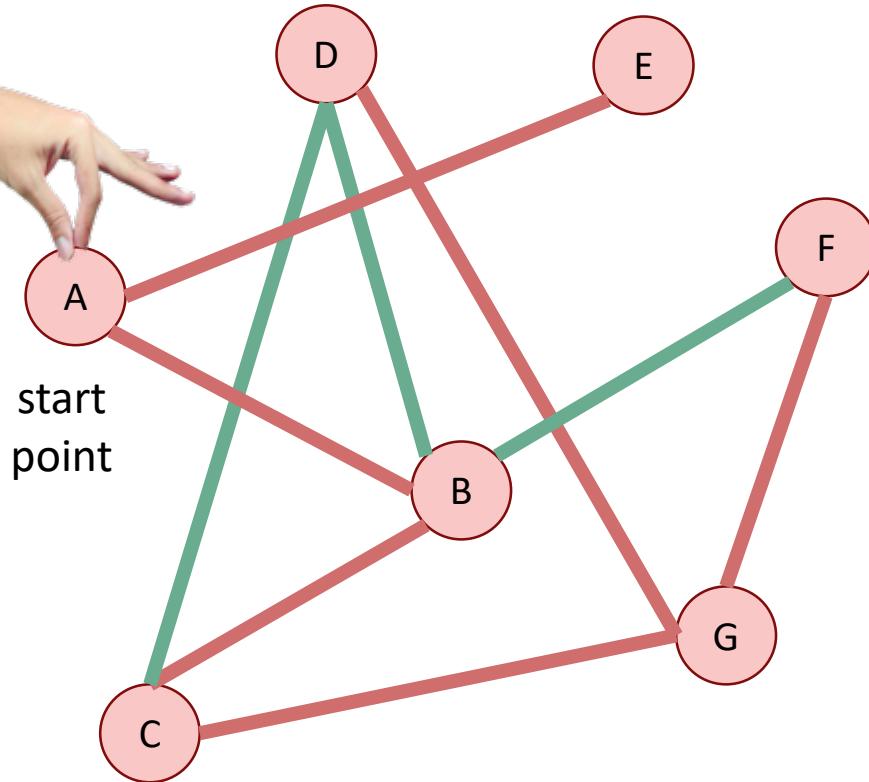
- DFS finds all the nodes reachable from the starting point

In an undirected graph, this is called a **connected component**.

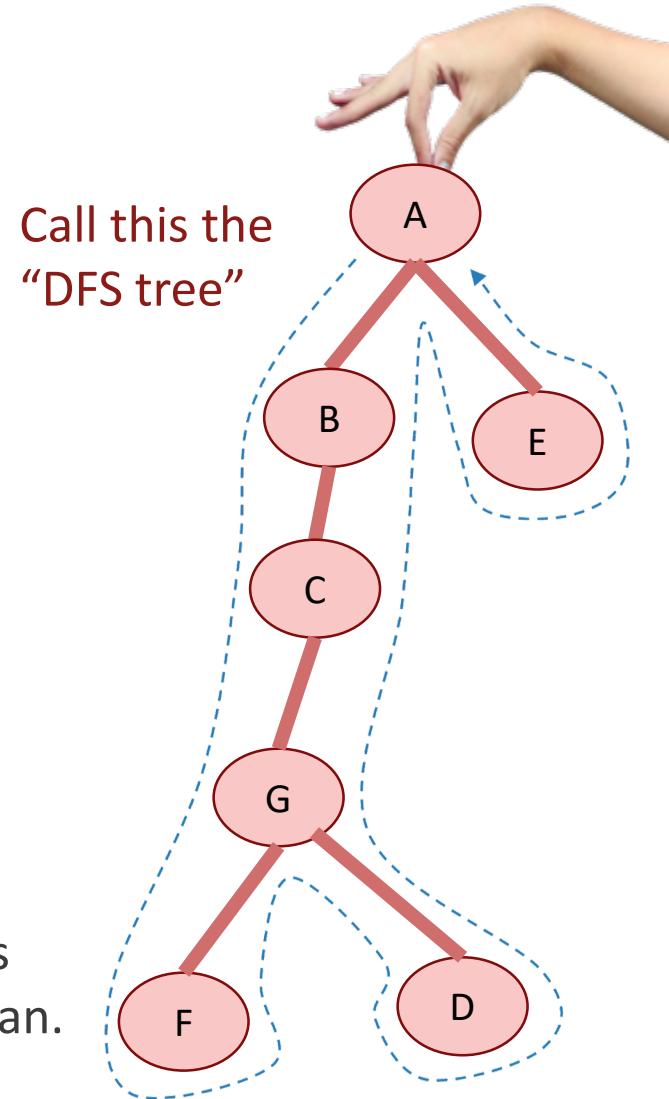


Why is it called depth-first?

- We are implicitly building a tree:



We first go as deep as we can.

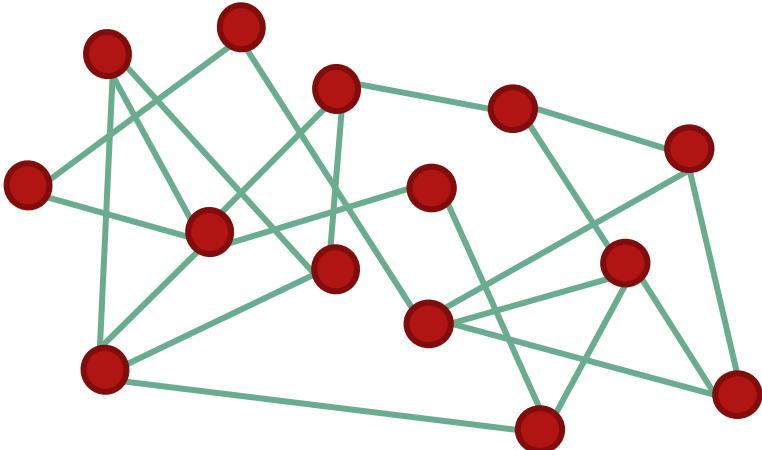


Running time

- To explore just the connected component we started in
 - We look at each edge at most twice.
 - Once from each of its endpoints
 - And basically, we don't do anything else.
 - So, **O(m)**

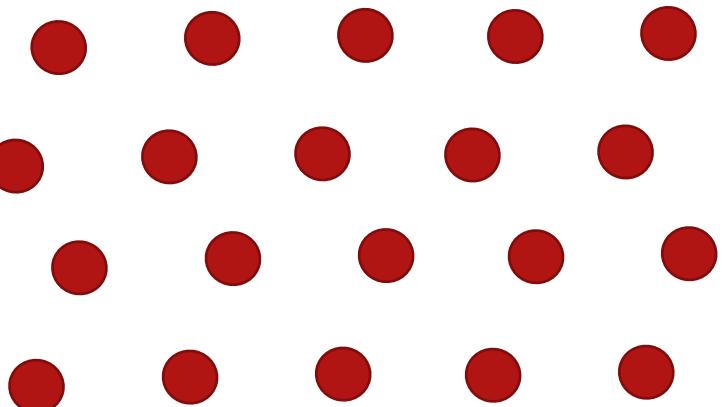
Running time

- To explore the whole graph
 - Explore the connected components one-by-one.
 - This takes time $O(n + m)$



Here the running time
is $O(m)$ like before

or



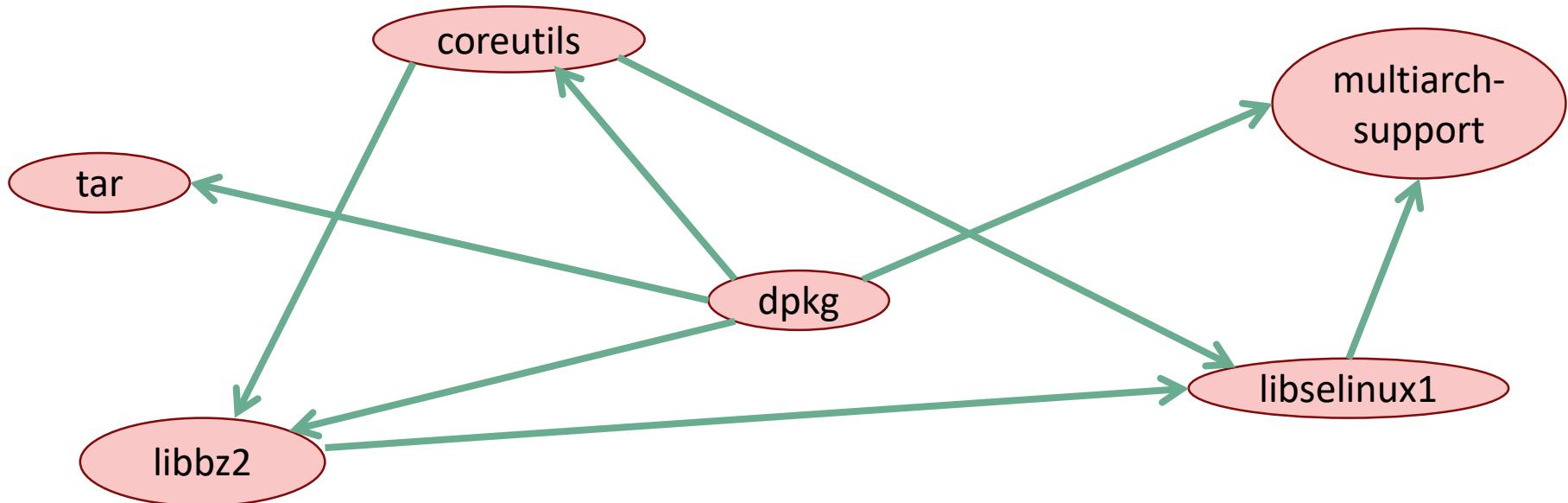
Here $m=0$ but it still takes time
 $O(n)$ to explore the graph.

Topological Ordering

Topological Ordering

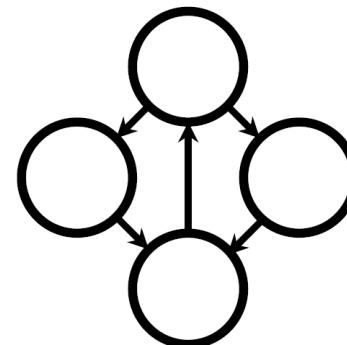
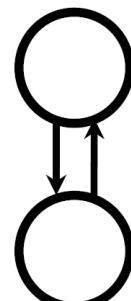
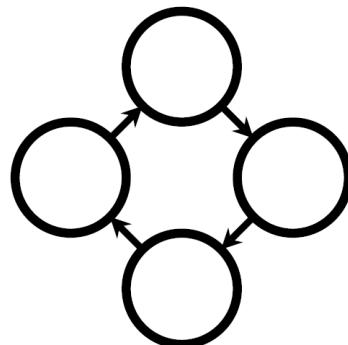
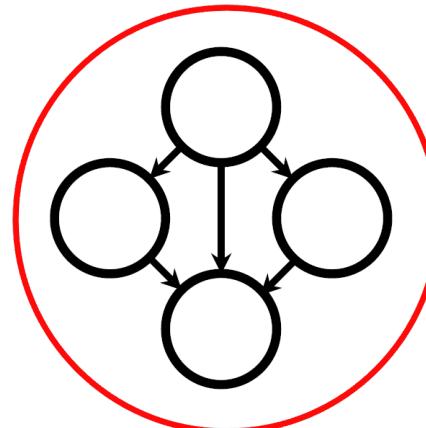
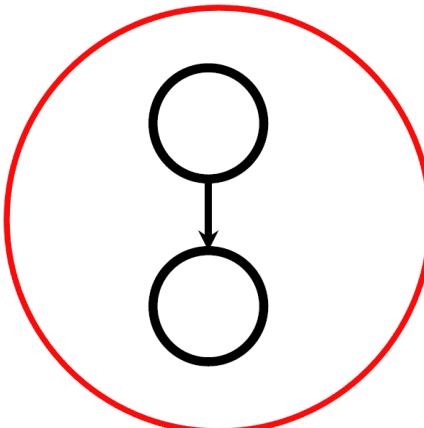
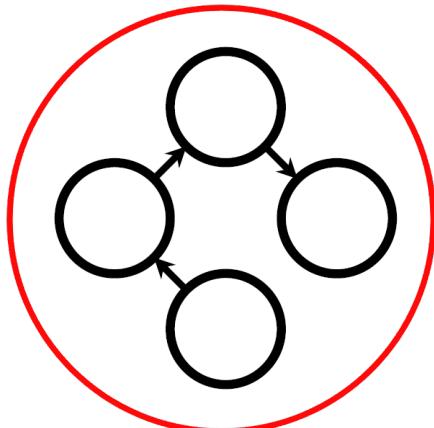
- Q. Given a package dependency graph, how do you install packages in the correct order?

Suppose the dependency graph has no cycles:
it is a **Directed Acyclic Graph (DAG)**



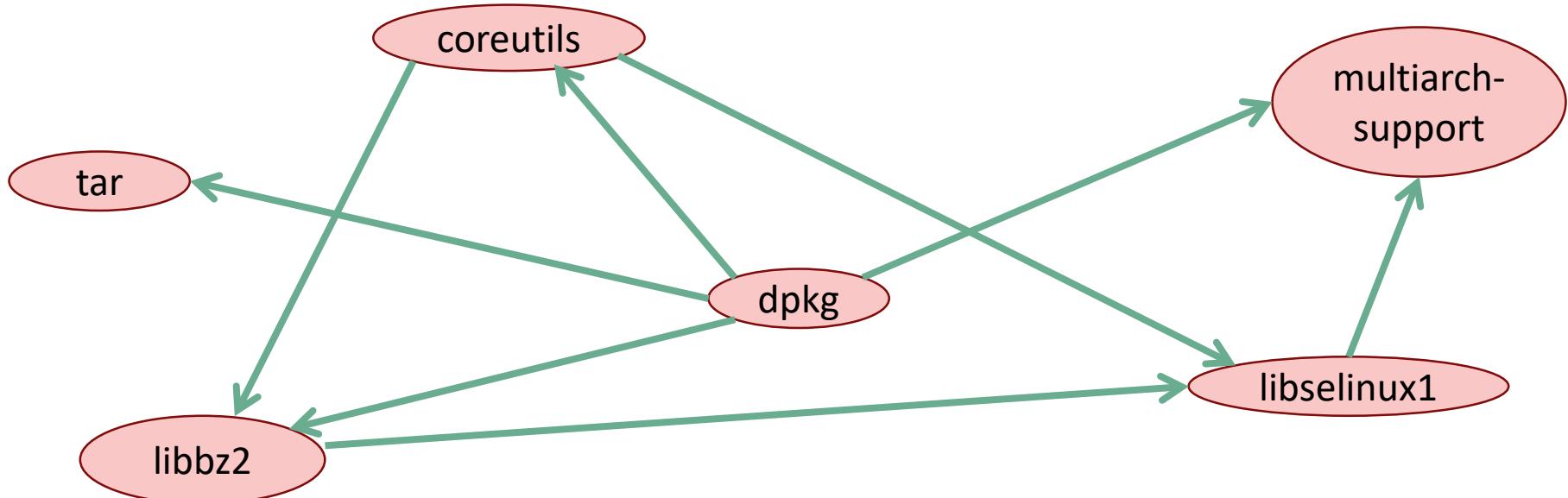
Wait.. What is a DAG?

- DAG is a **directed graph with no directed cycles**.
- Which of these graphs are valid DAGs?

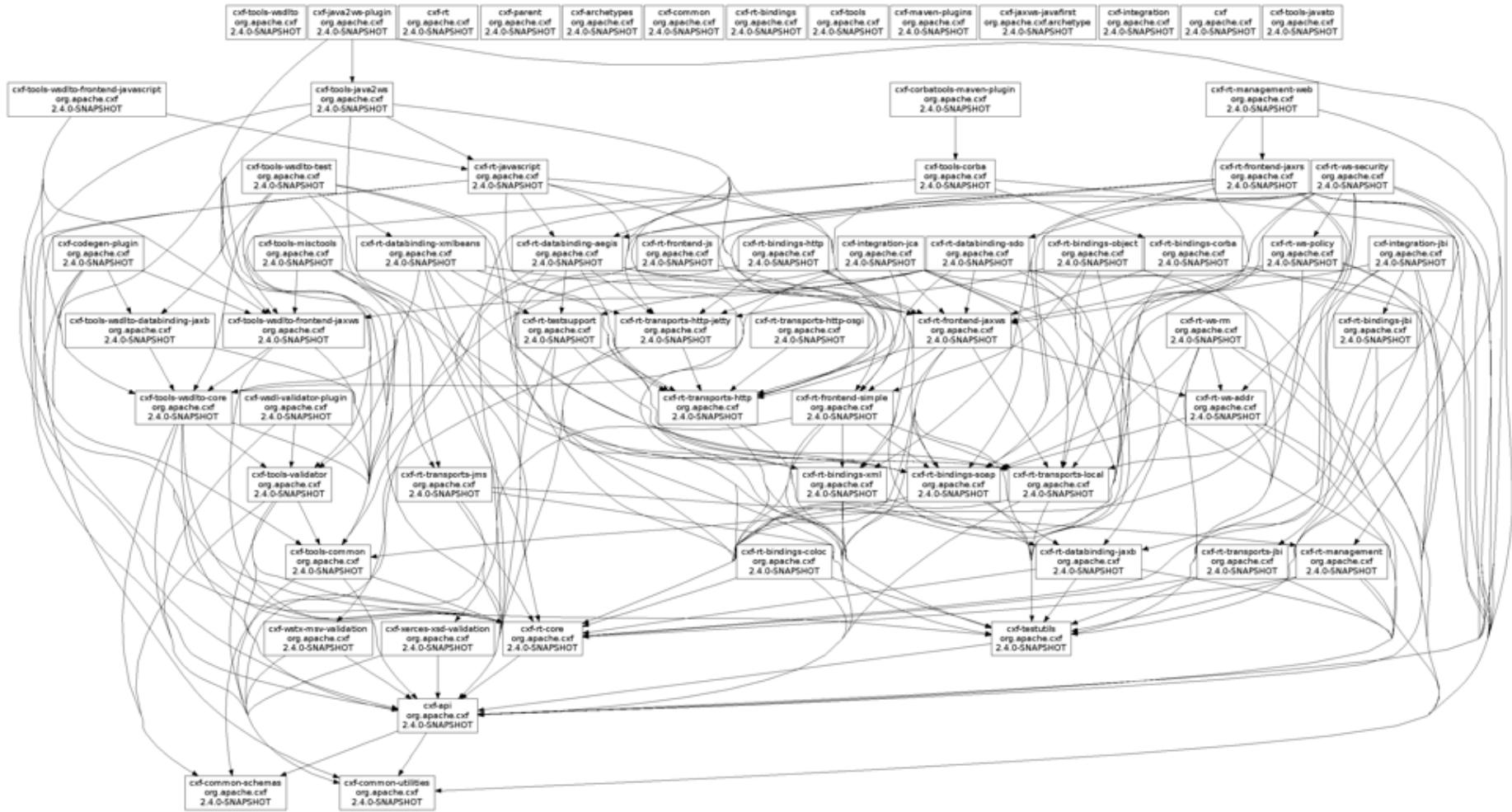


Topological Ordering

- Application of DFS
- Find an ordering of vertices in a DAG so that all of the dependency requirements are met.
 - If v comes before w in the ordering, there is not an edge from w to v.

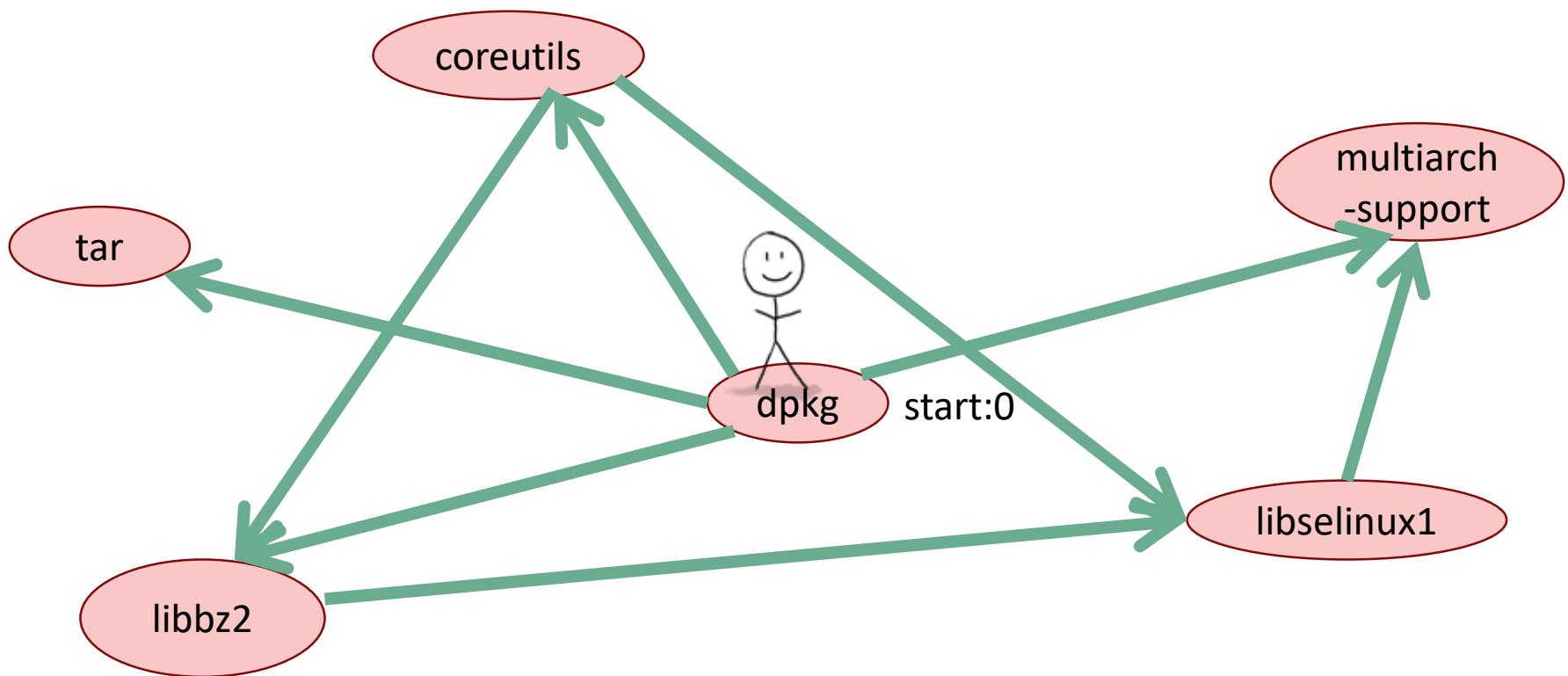


Can't always eyeball it.

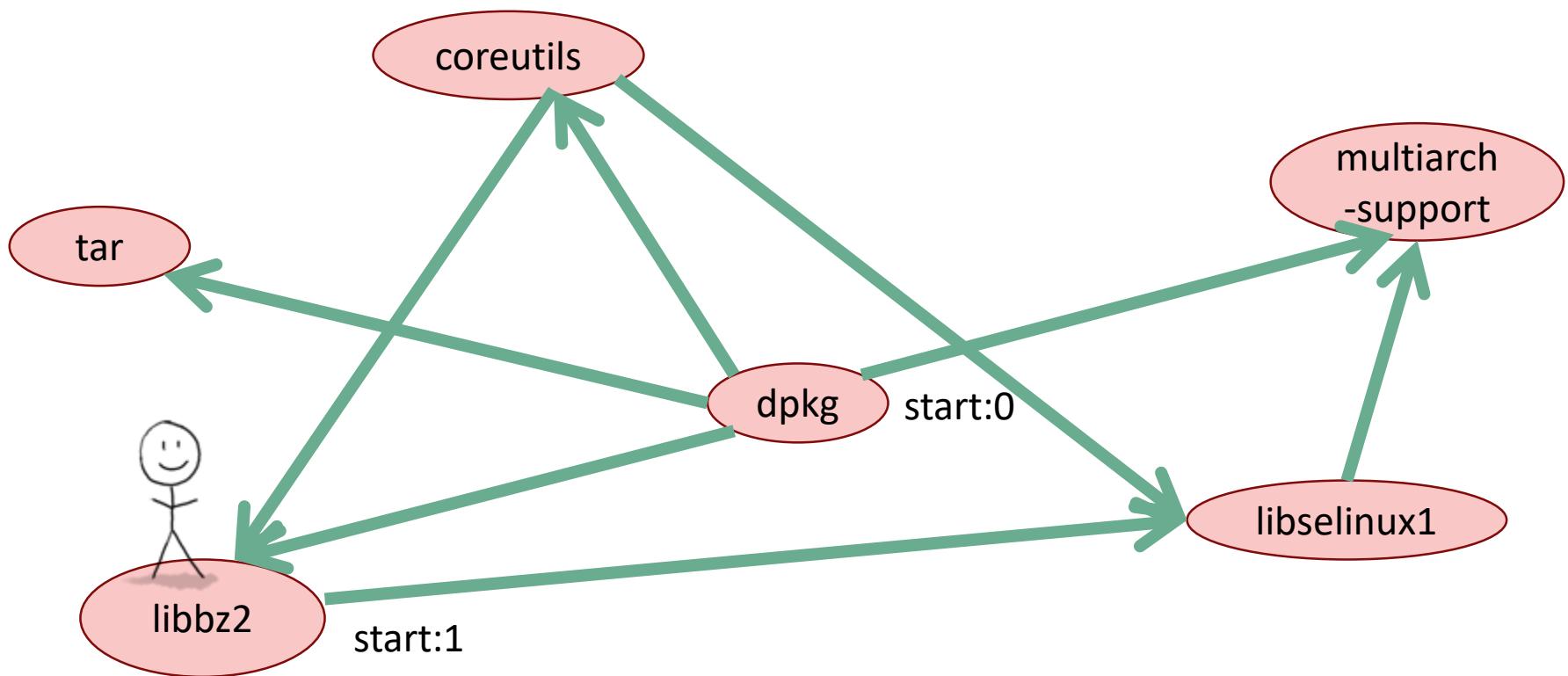


Apache CXF

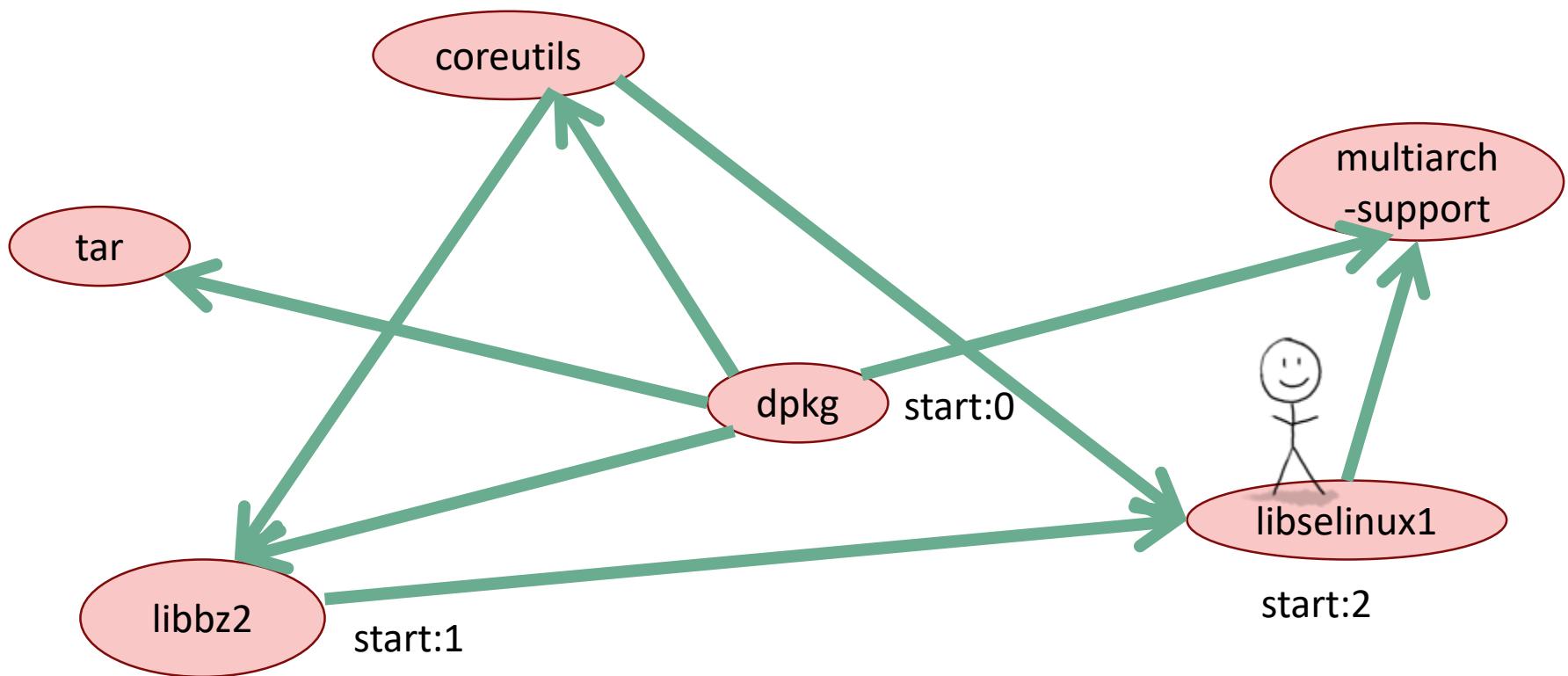
Let's do DFS



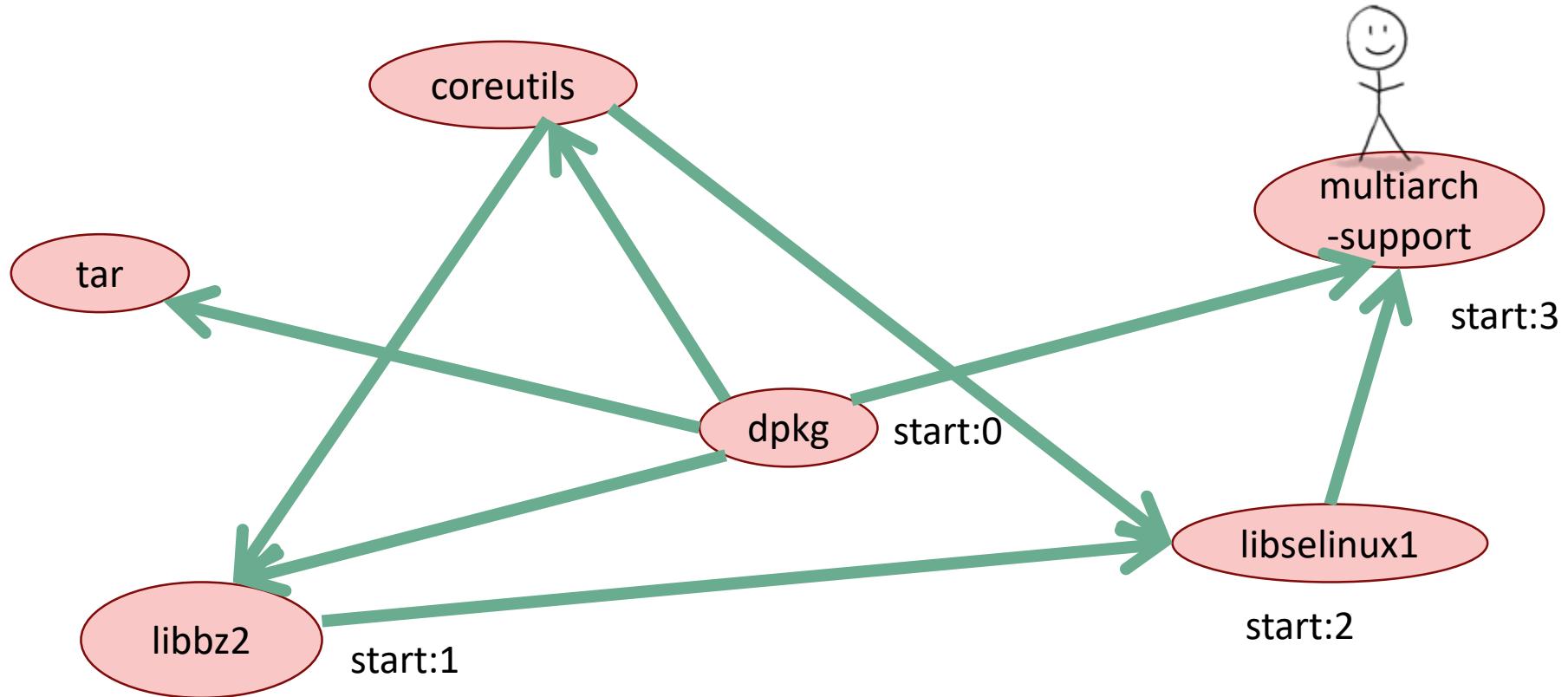
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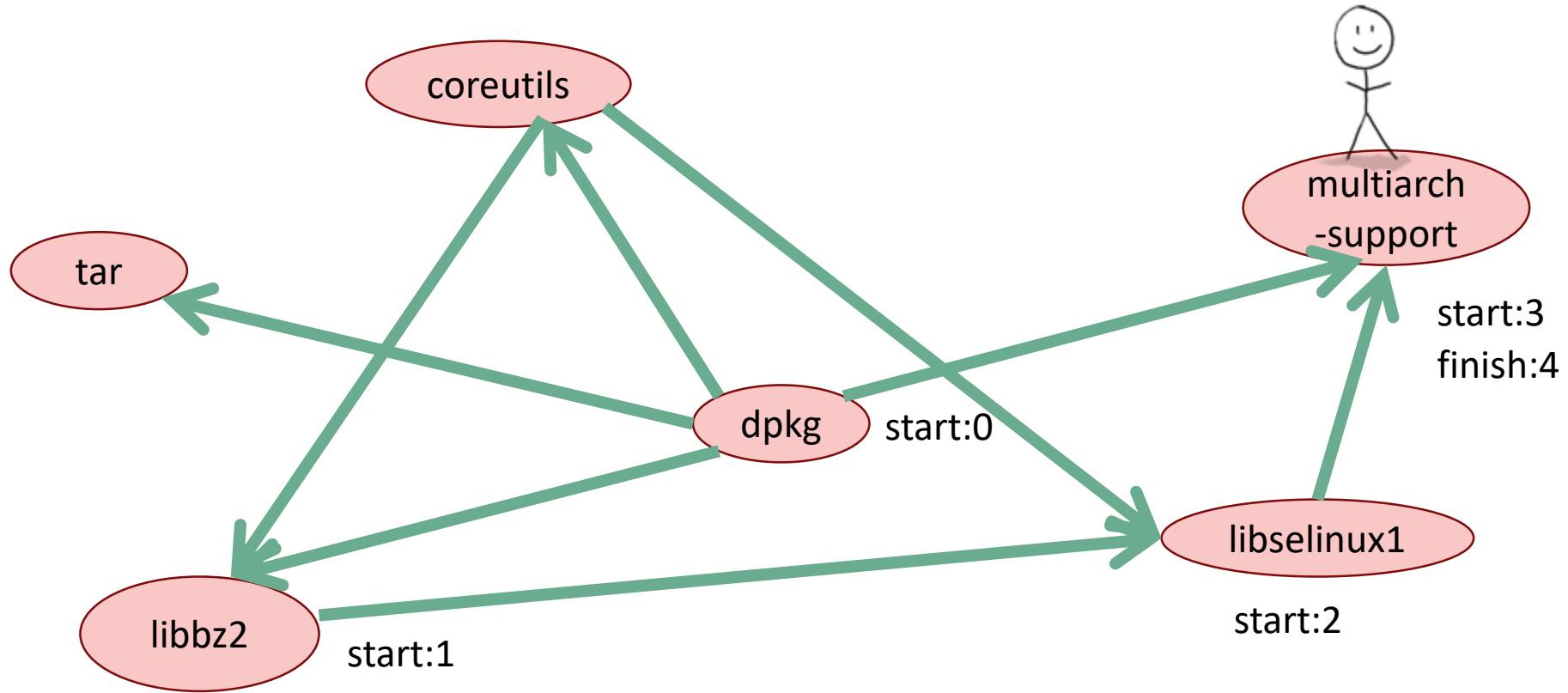
Let's do DFS



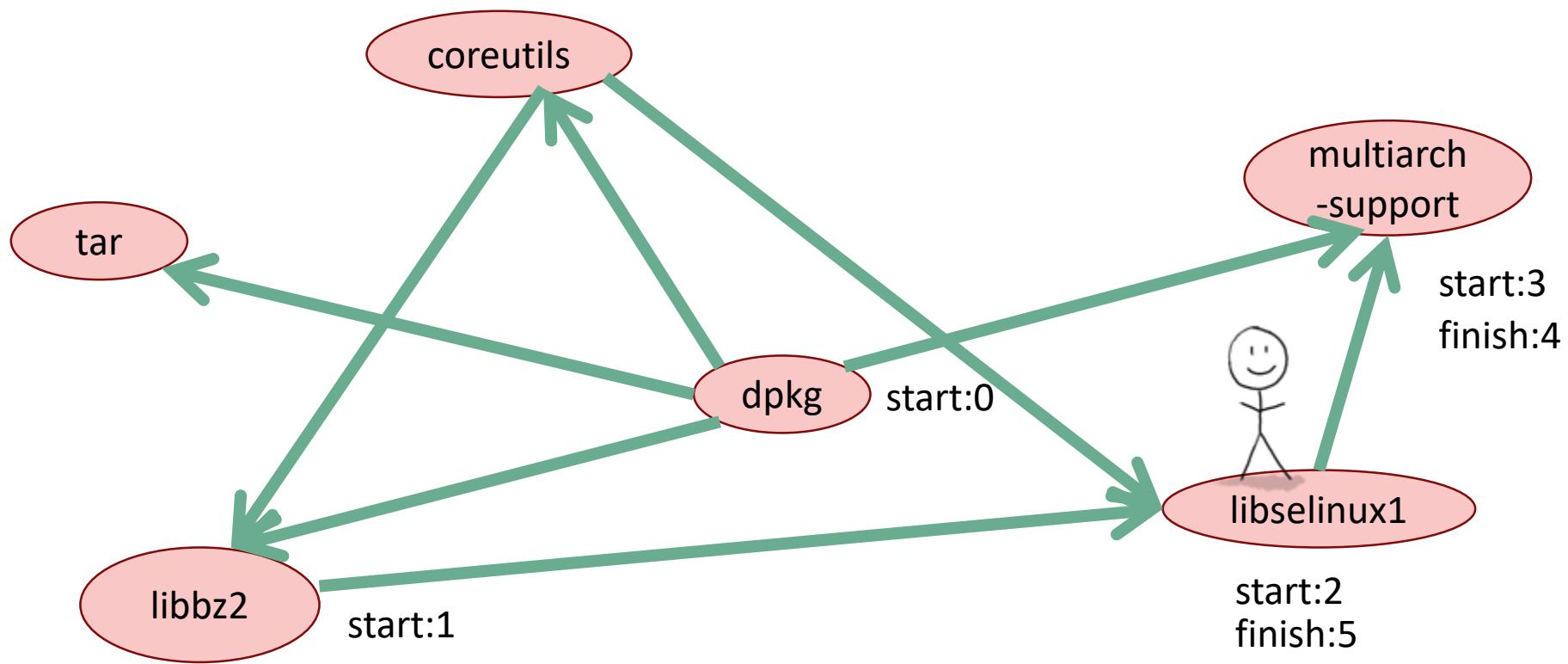
Let's do DFS



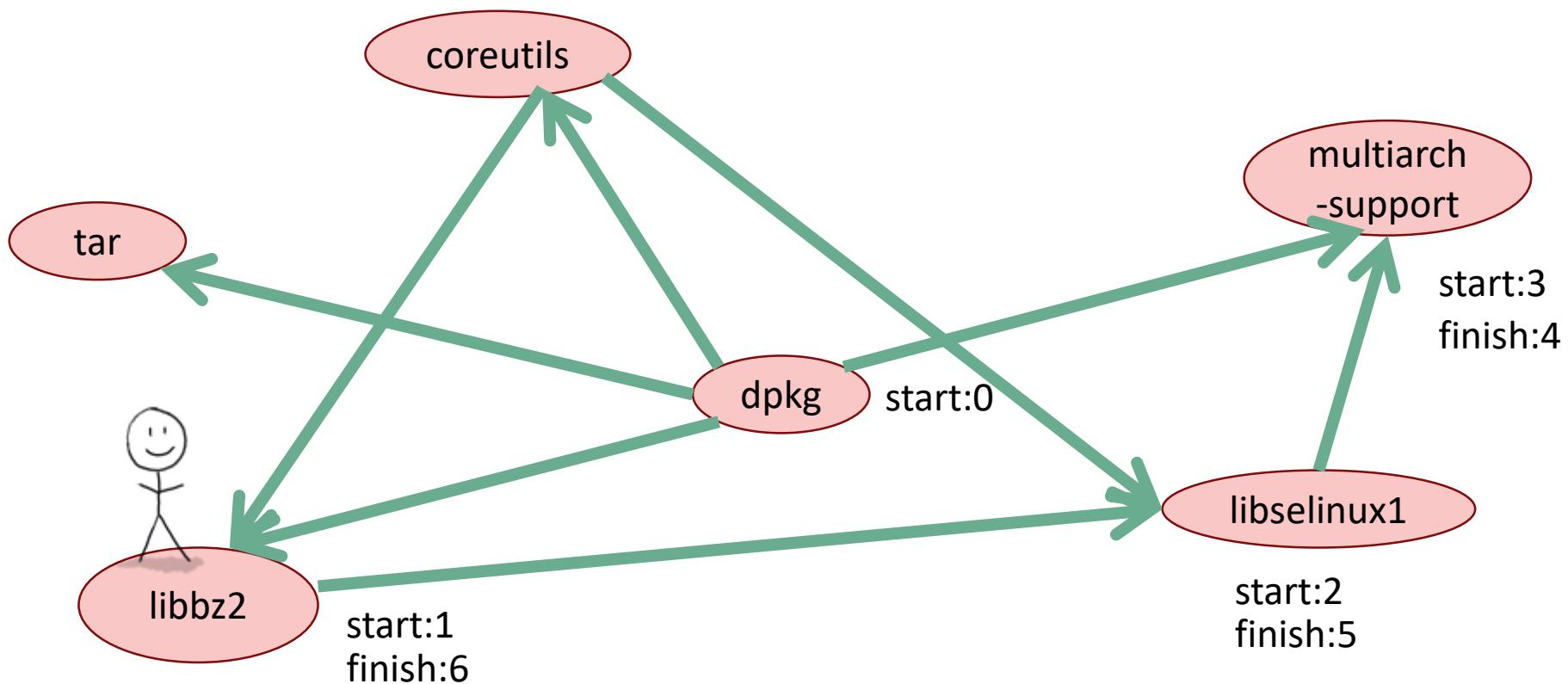
Let's do DFS



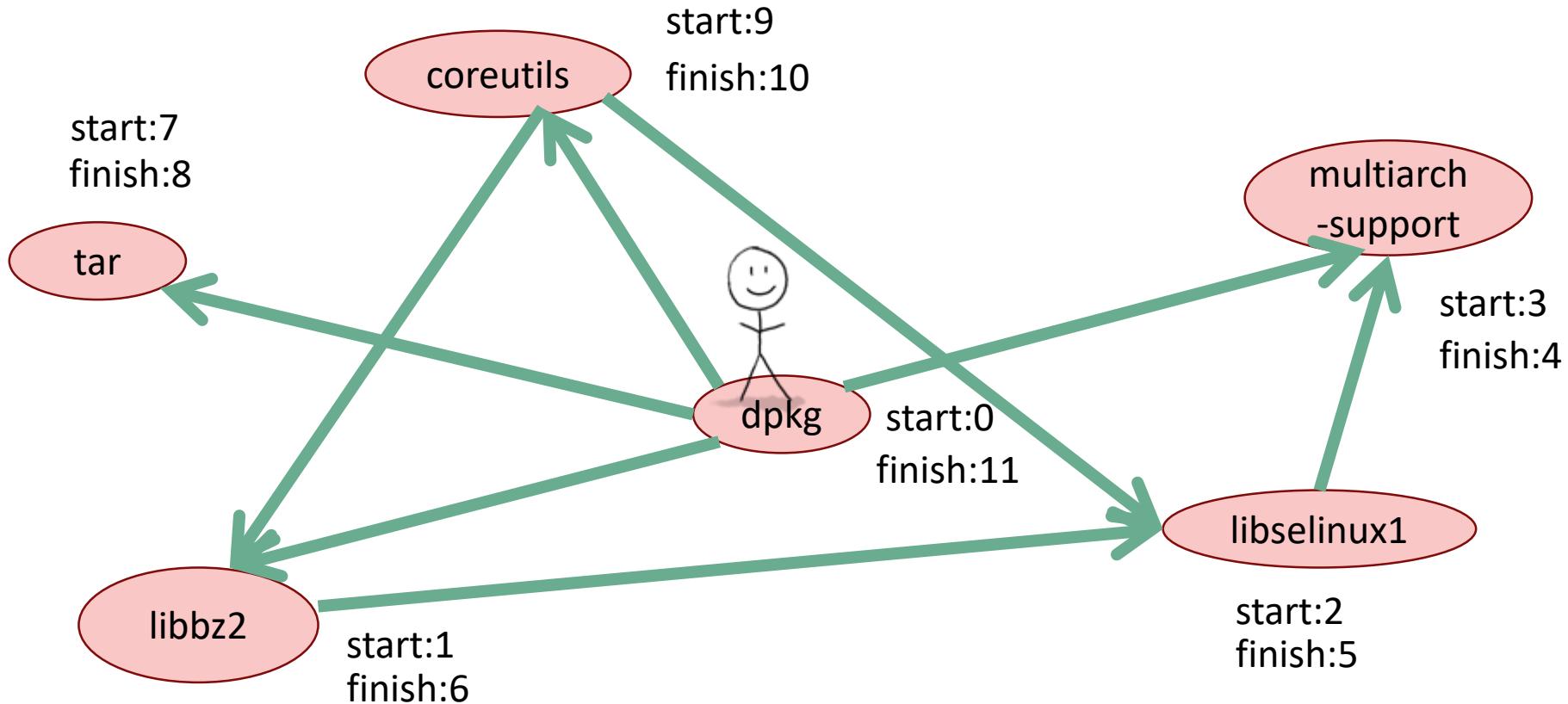
Let's do DFS



Let's do DFS



Let's do DFS



- What do you notice about **the finish times?**
- Any ideas for how we should do topological sort?

Finish times seem useful

- **Claim:** In general, we'll always have



- To understand why, let's go back to that DFS tree.

Proof the claim

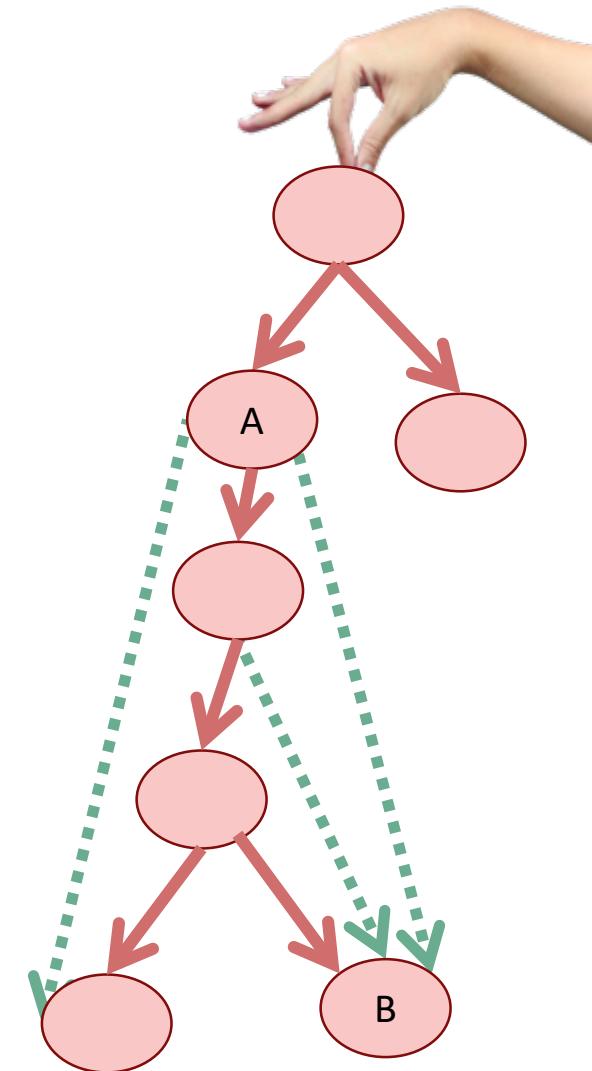
Suppose the underlying graph has no cycles

If $A \rightarrow B$, then $B.\text{finishTime} < A.\text{finishTime}$

- **Case 1:** B is a descendant of A in the DFS tree.
- Then



- aka, $B.\text{finishTime} < A.\text{finishTime}$.



Proof the claim

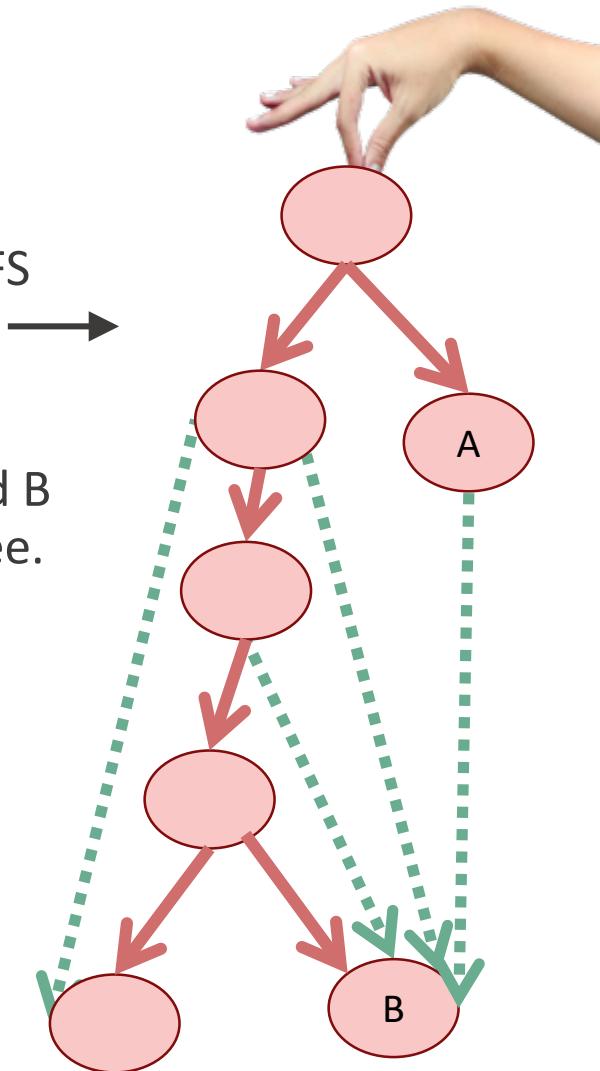
Suppose the underlying graph has no cycles

If $A \rightarrow B$, then $B.\text{finishTime} < A.\text{finishTime}$

- **Case 2:** B is a NOT descendant of A in the DFS tree.
 - Notice that A can't be a descendant of B in the DFS tree or else there'd be a cycle; so it looks like this →
- Then we must have explored B before A.
 - Otherwise we would have gotten to B from A, and B would have been a descendant of A in the DFS tree.
- Then

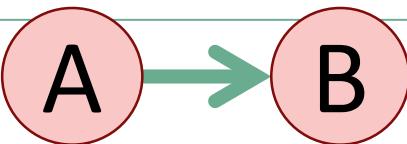


- aka, $B.\text{finishTime} < A.\text{finishTime}$.



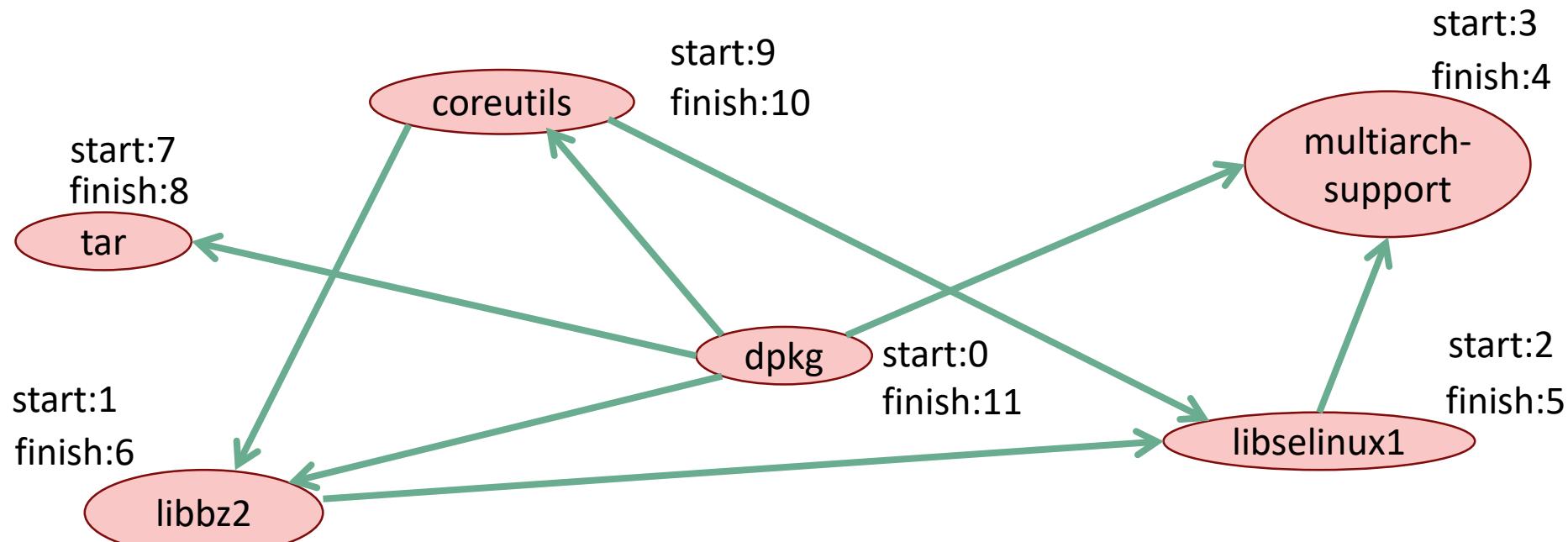
Theorem

- If we run DFS on a directed acyclic graph,

If  Then $B.\text{finishTime} < A.\text{finishTime}$

Back to topological sorting

- In what order should I install packages?

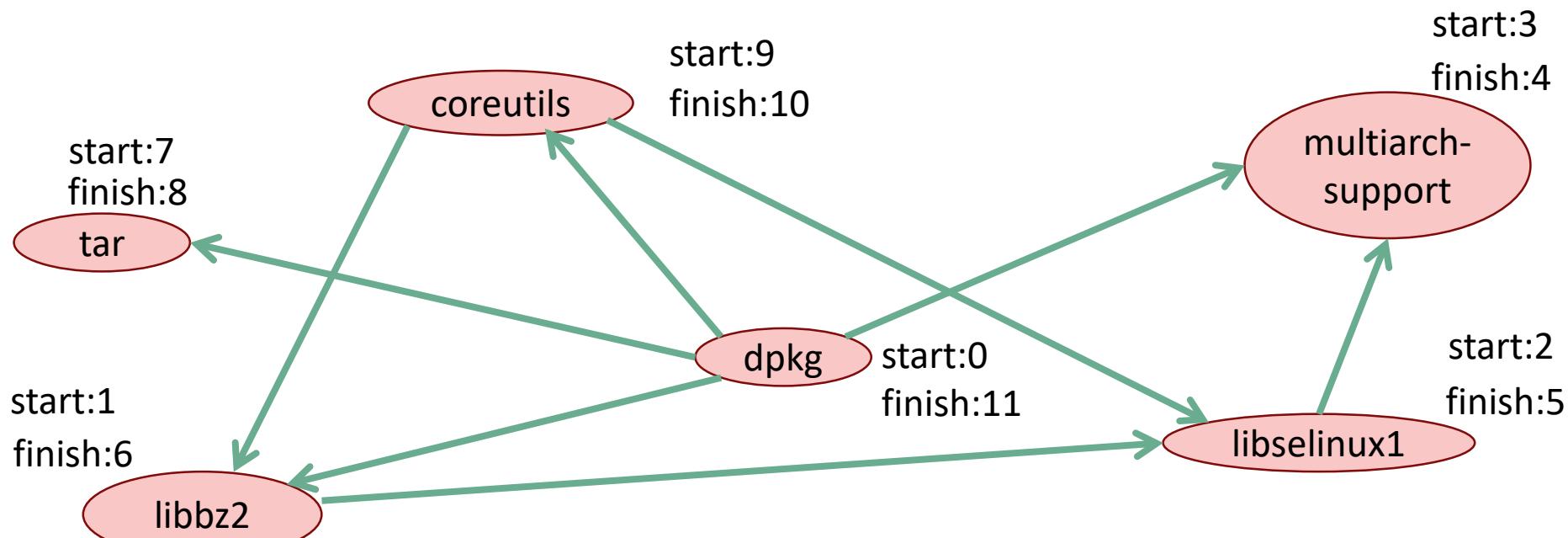


- In reverse order of finishing time in DFS!

Topological Sorting

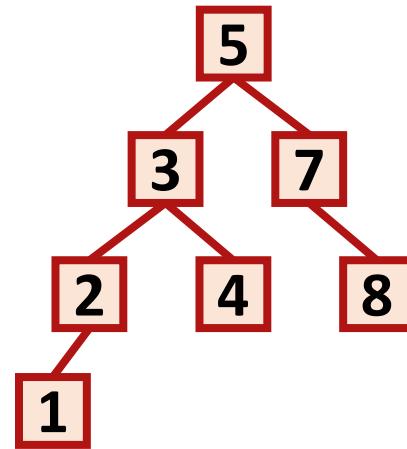
- Do DFS
- When you mark a vertex as **all done**, put it at the **beginning** of the list.

- dpkg
- coreutils
- tar
- libbz2
- libselinux1
- multiarch_support



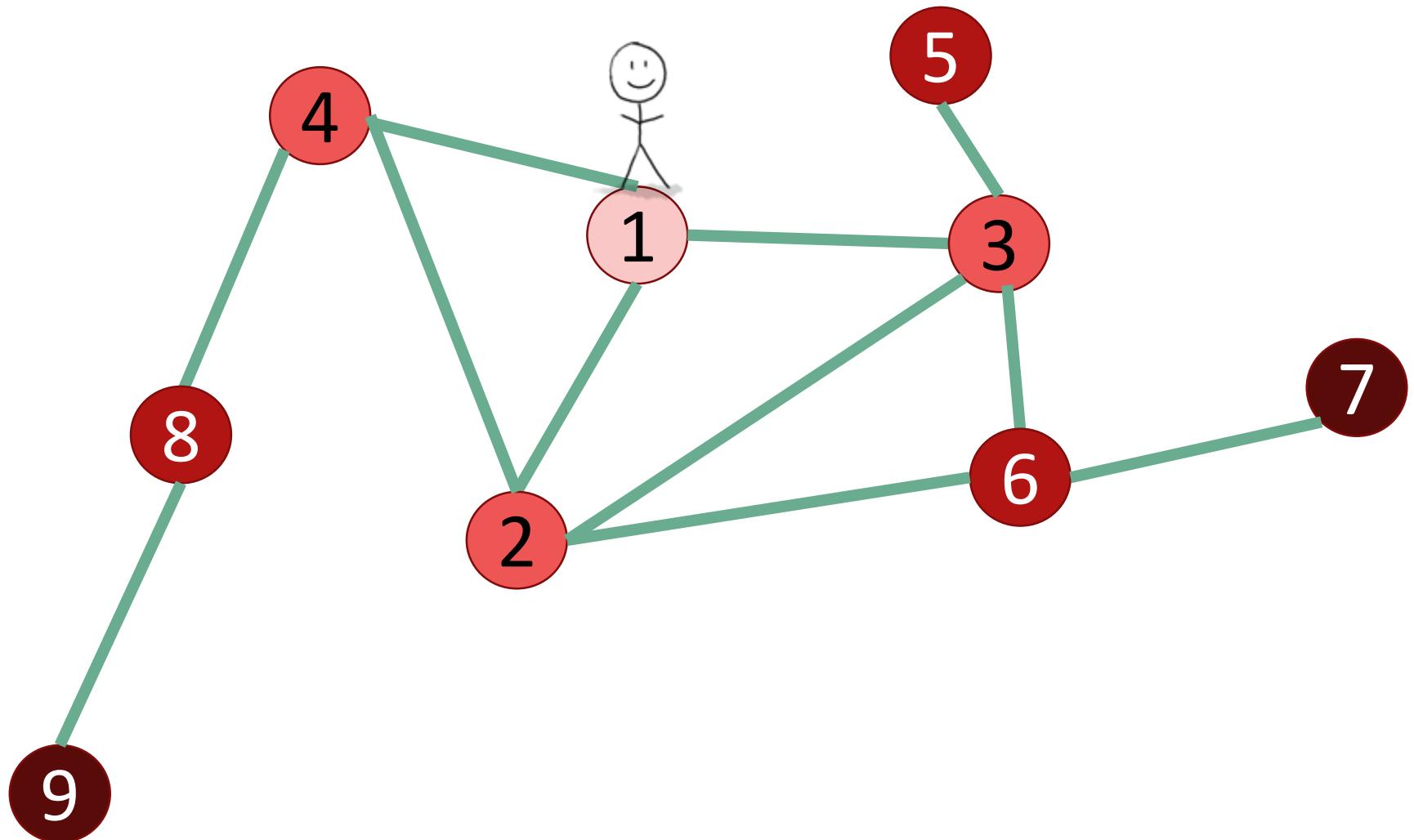
What did we just learn?

- DFS can help you solve the **topological sorting problem**
- Another use of DFS: **In-order enumeration** of binary search trees

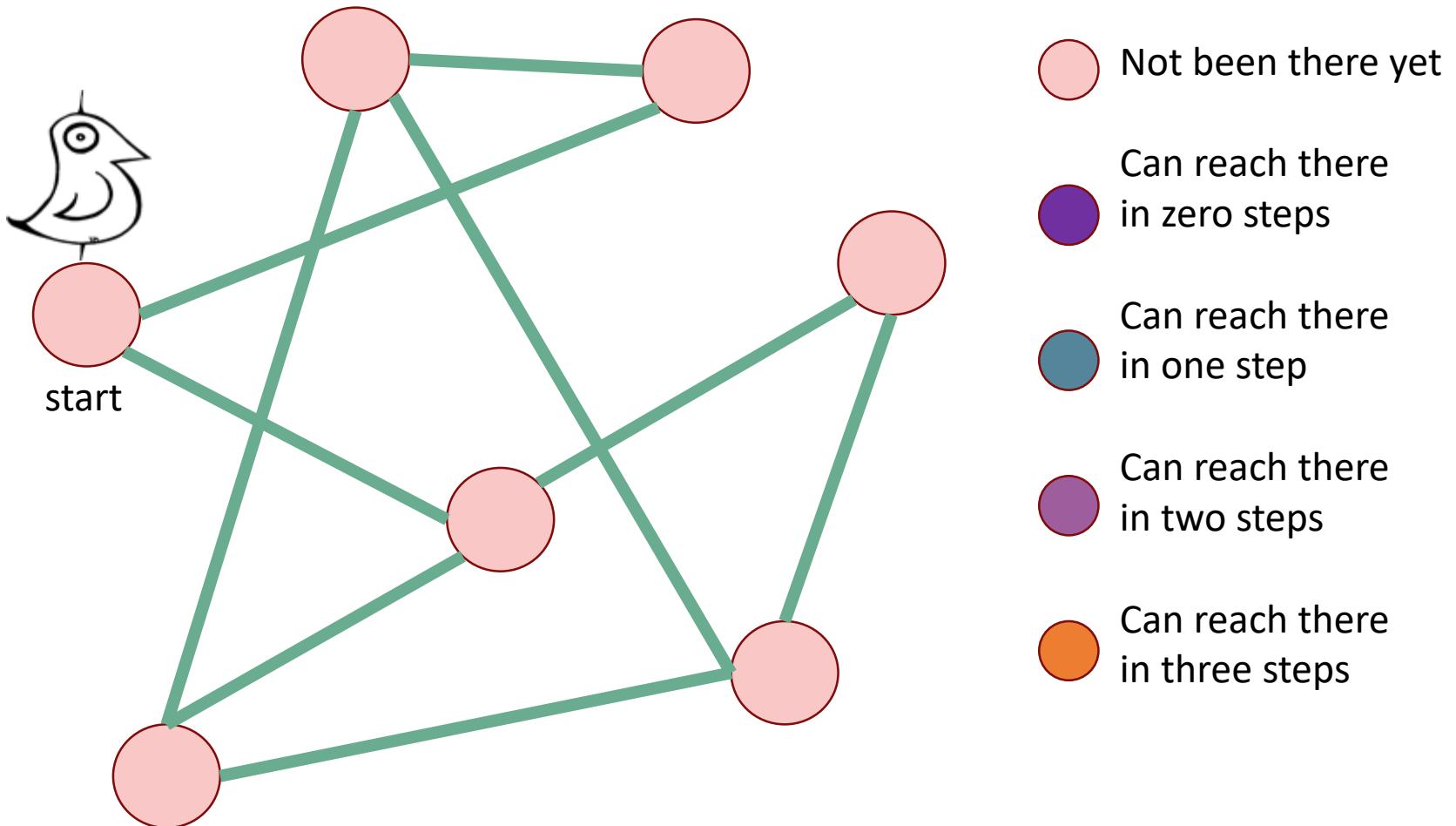


Breadth-First Search

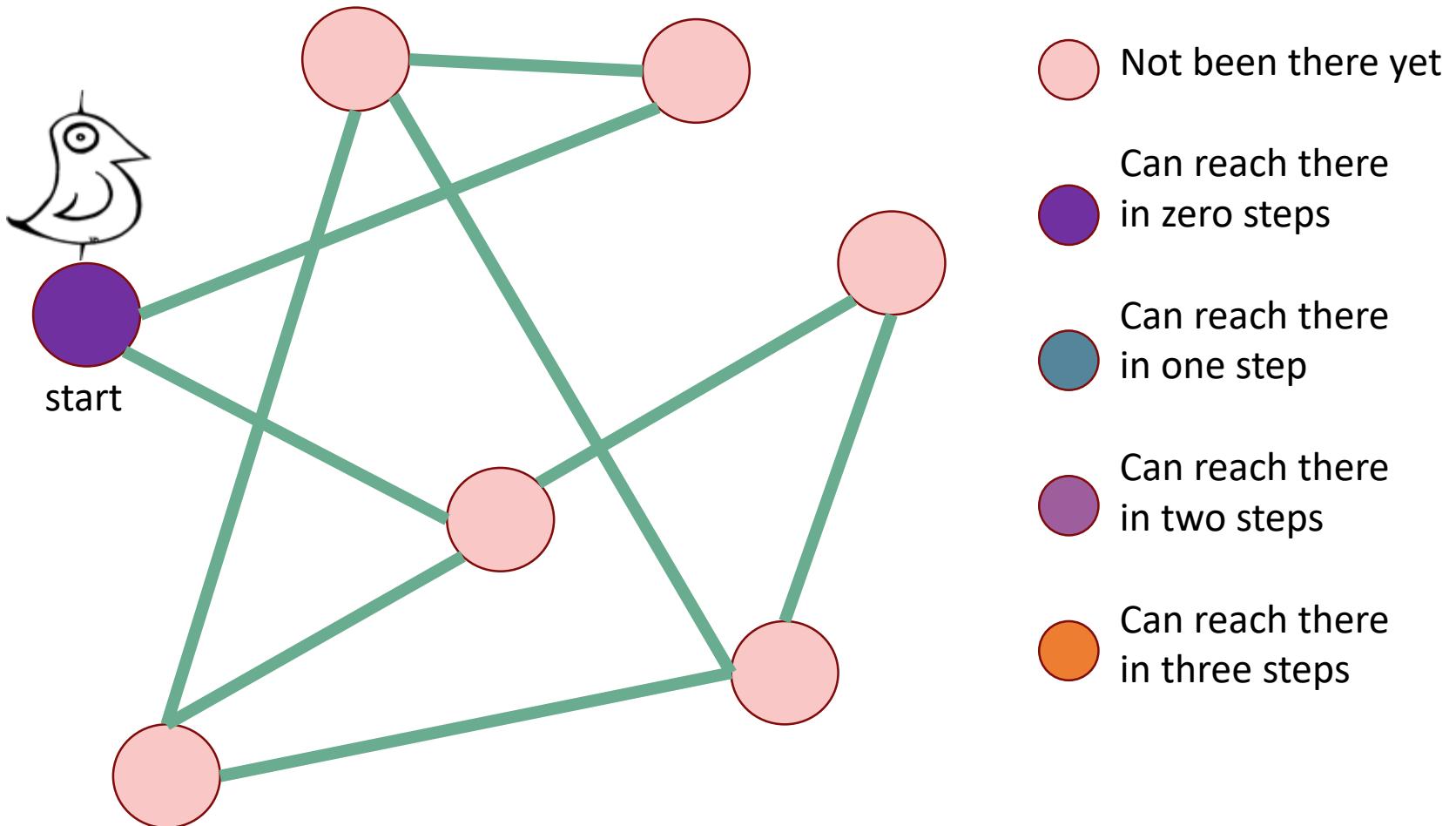
How do we explore a graph?



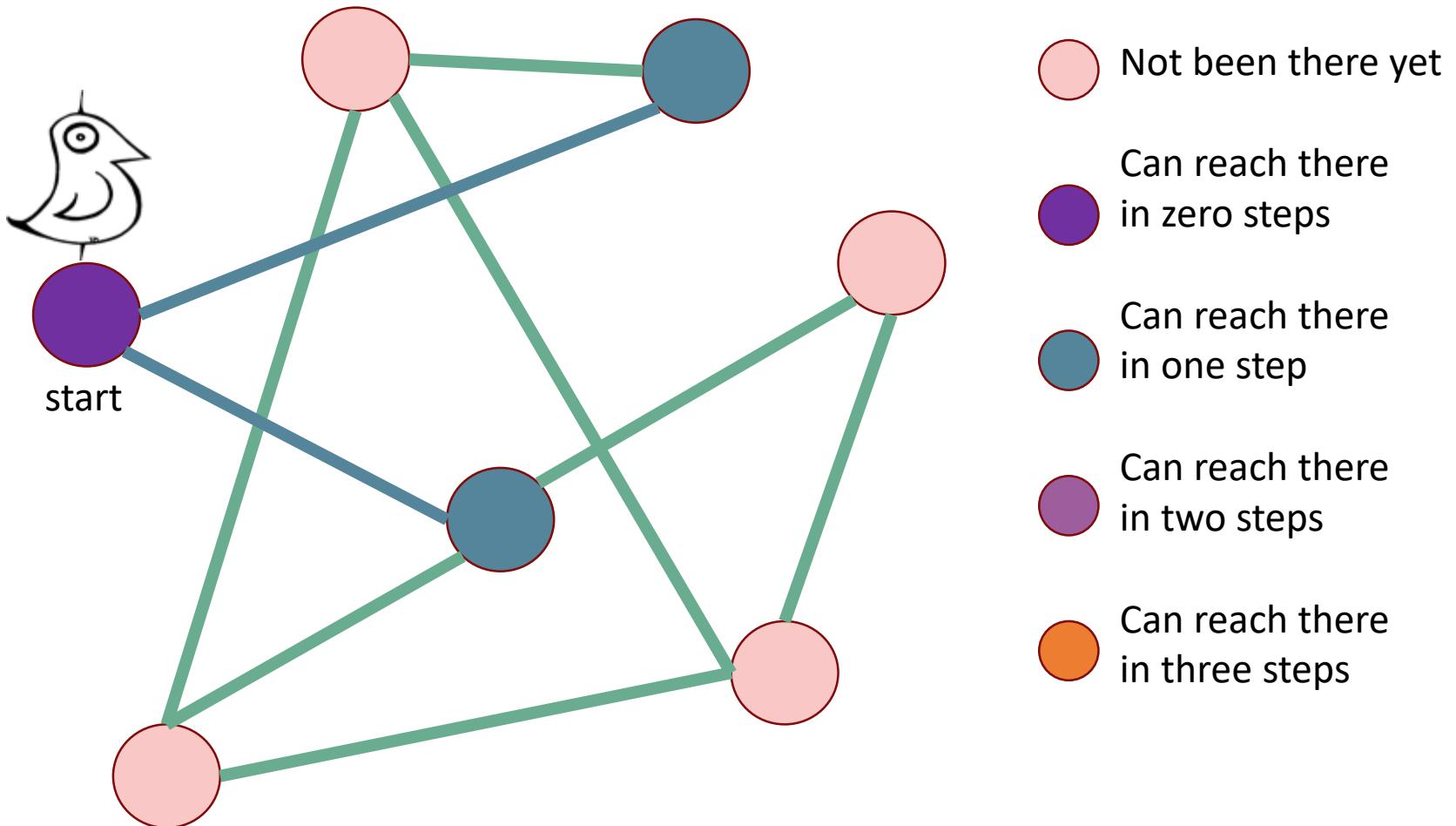
Breadth-First Search



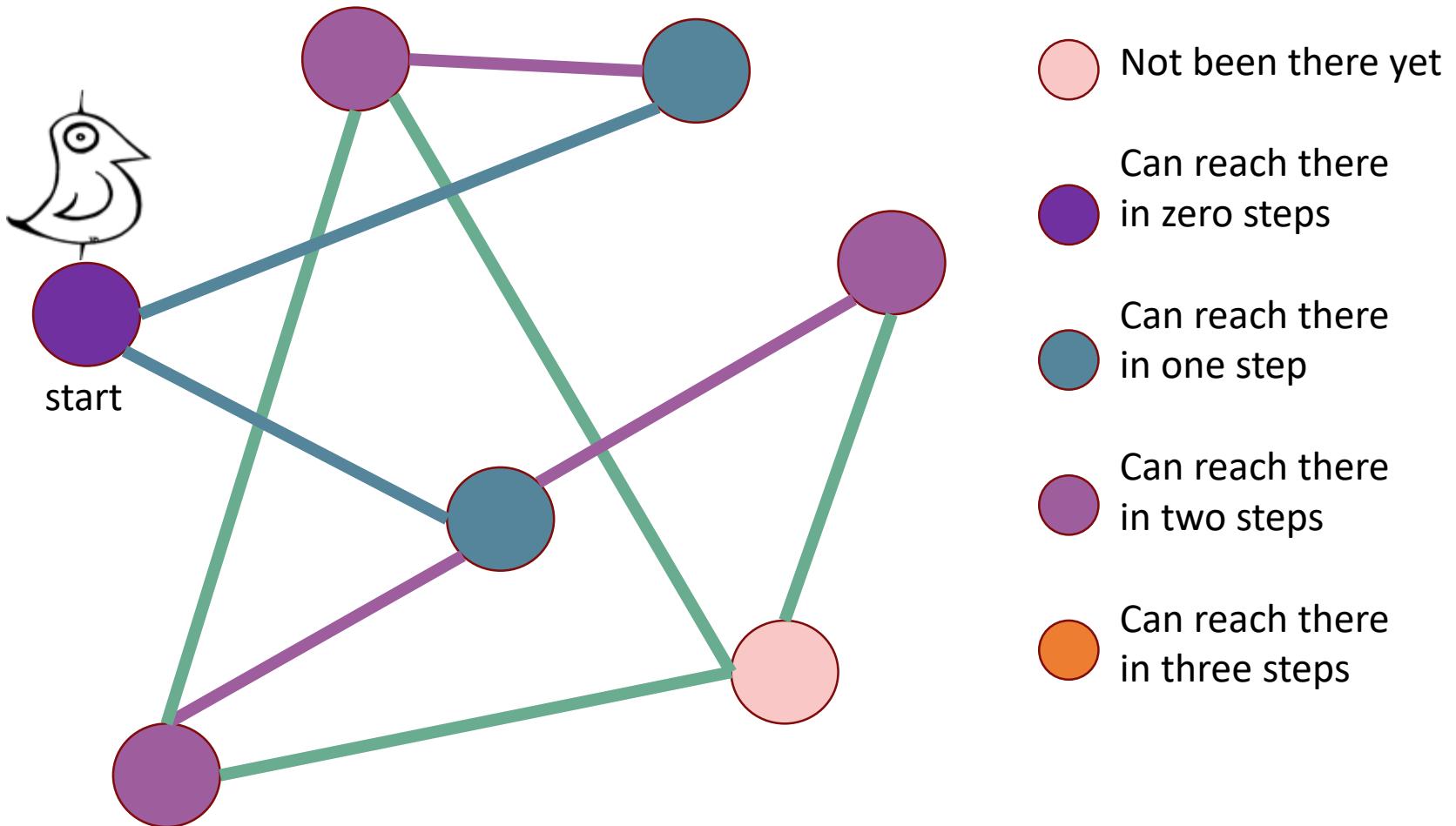
Breadth-First Search



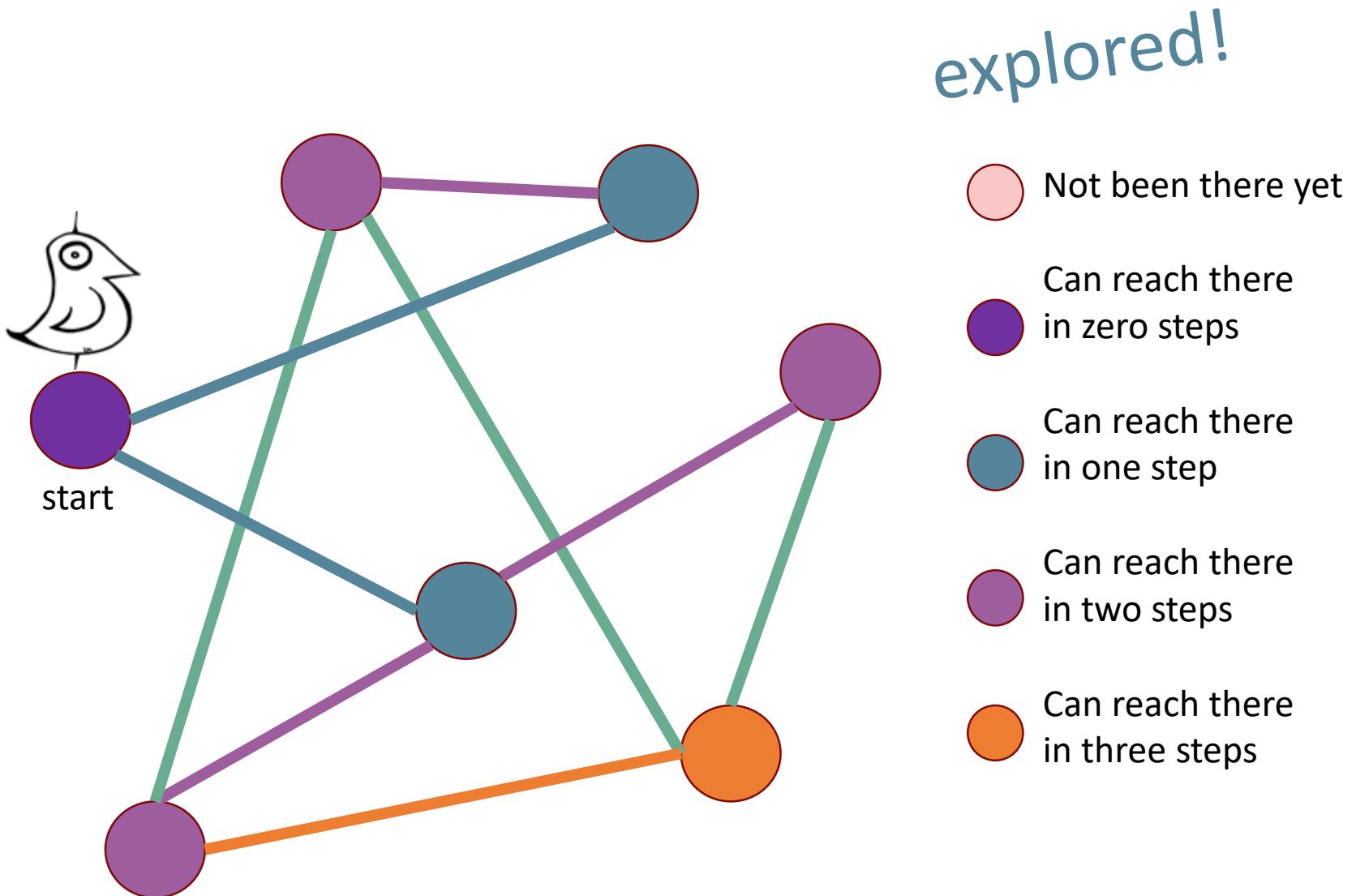
Breadth-First Search



Breadth-First Search



Breadth-First Search



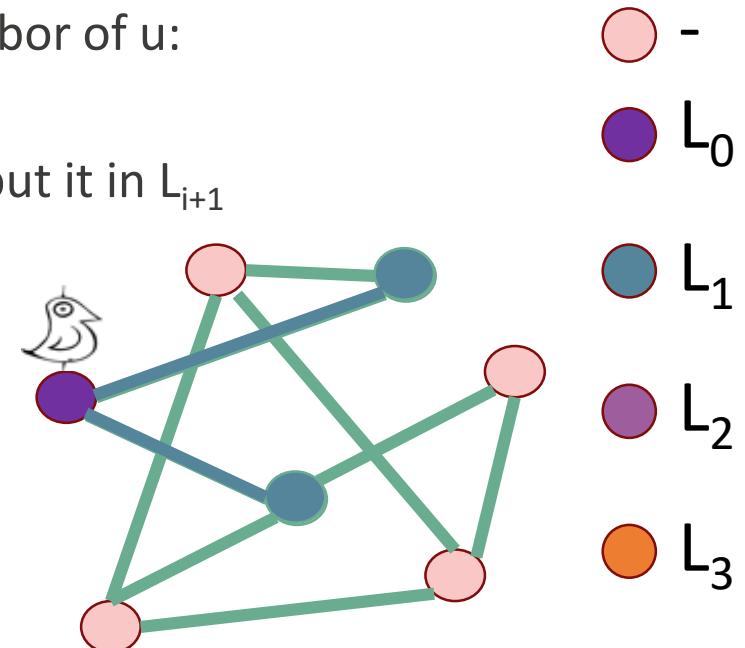
Breadth-First Search

- Set $L_i = []$ for $i=1,\dots,n$
- $L_0 = [w]$, where w is the start node
- Mark w as visited
- For $i = 0, \dots, n-1$:

- For u in L_i :
 - For each v which is a neighbor of u :
 - If v isn't yet visited:
 - mark v as visited, and put it in L_{i+1}

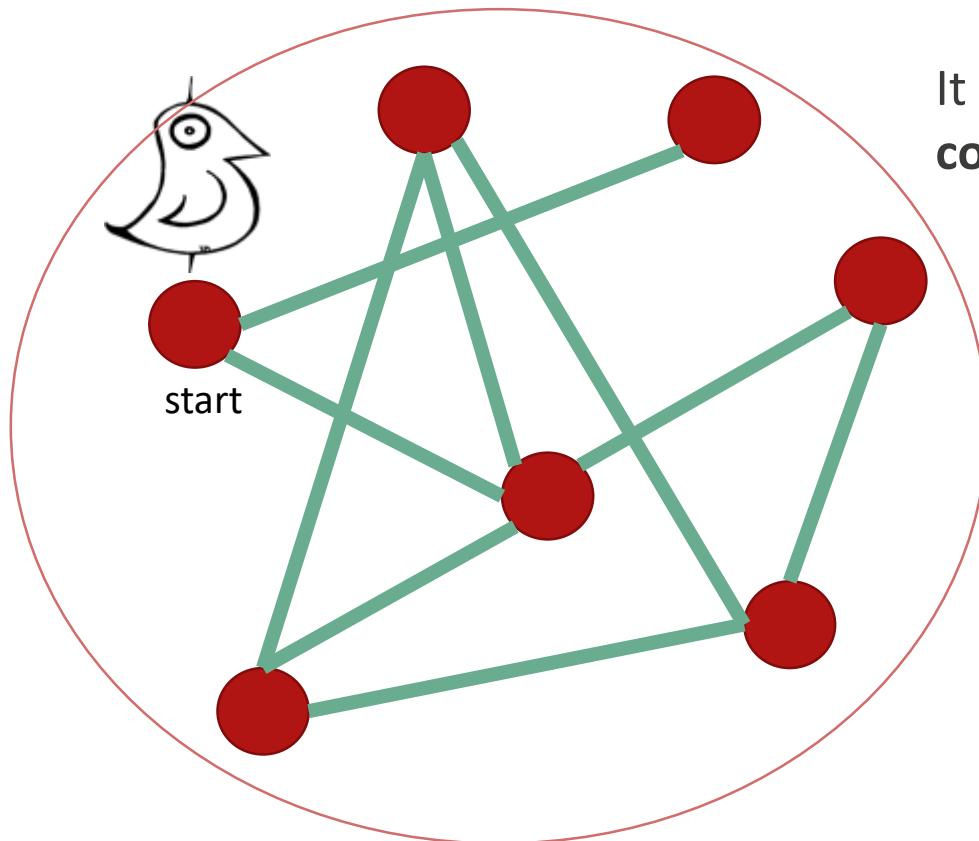
Go through all the nodes in L_i and add their unvisited neighbors to L_{i+1}

L_i is the set of nodes we can reach in i steps from w



Breadth-First Search

- BFS also finds all the nodes reachable from the starting point



It is also a good way to find all the connected components.

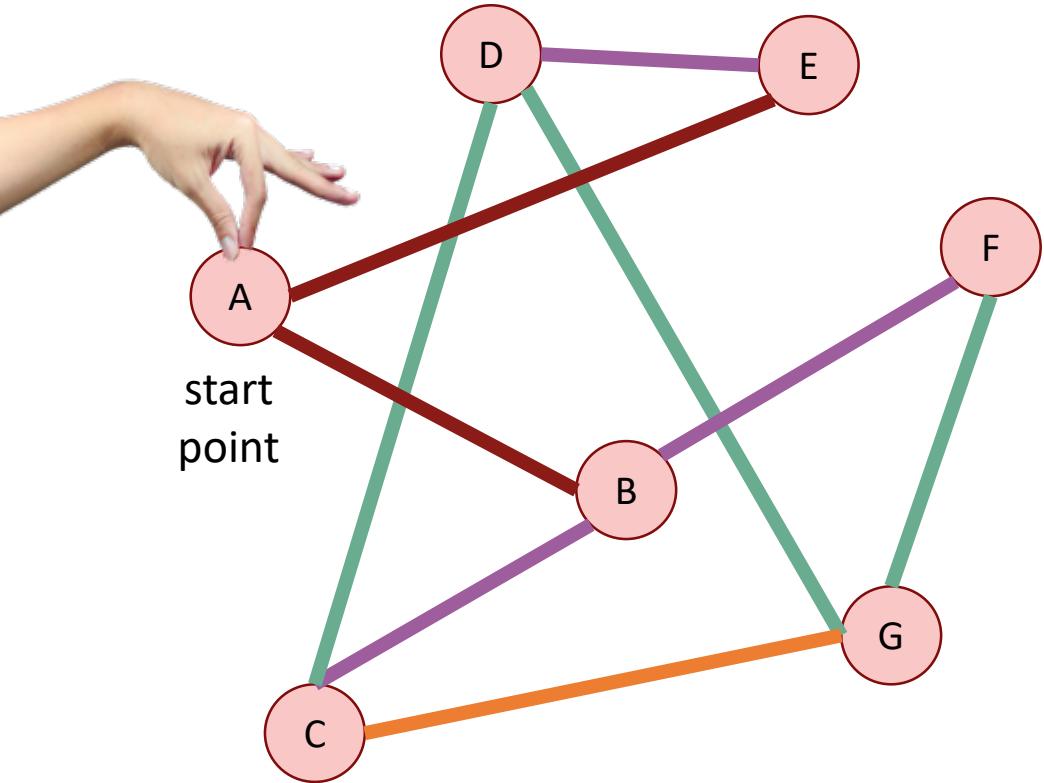
- Like DFS, BFS also works fine on directed graphs.

Running time

- To explore the whole graph, explore the connected components one-by-one.
- Same argument as DFS: BFS running time is $O(n + m)$

Why is it called breadth-first?

- We are implicitly building a tree:



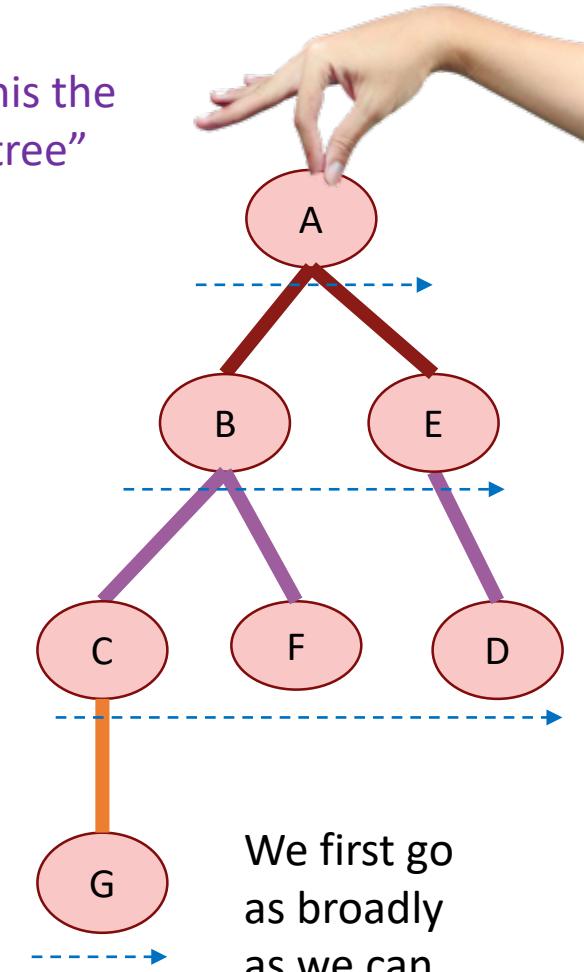
Call this the
“BFS tree”

L_0

L_1

L_2

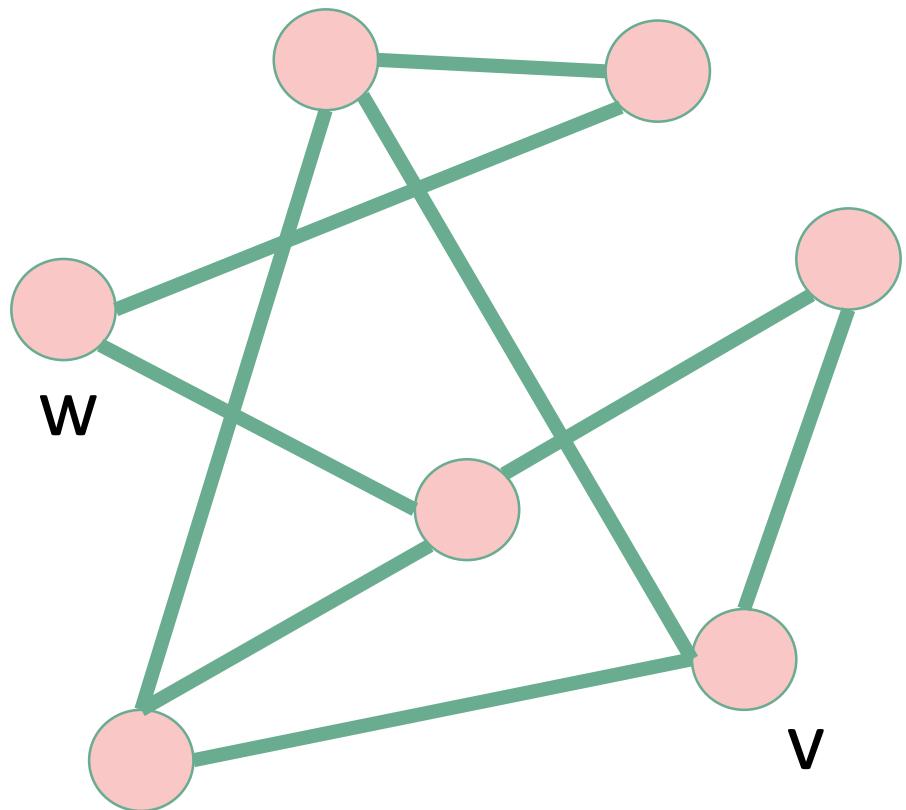
L_3



We first go
as broadly
as we can.

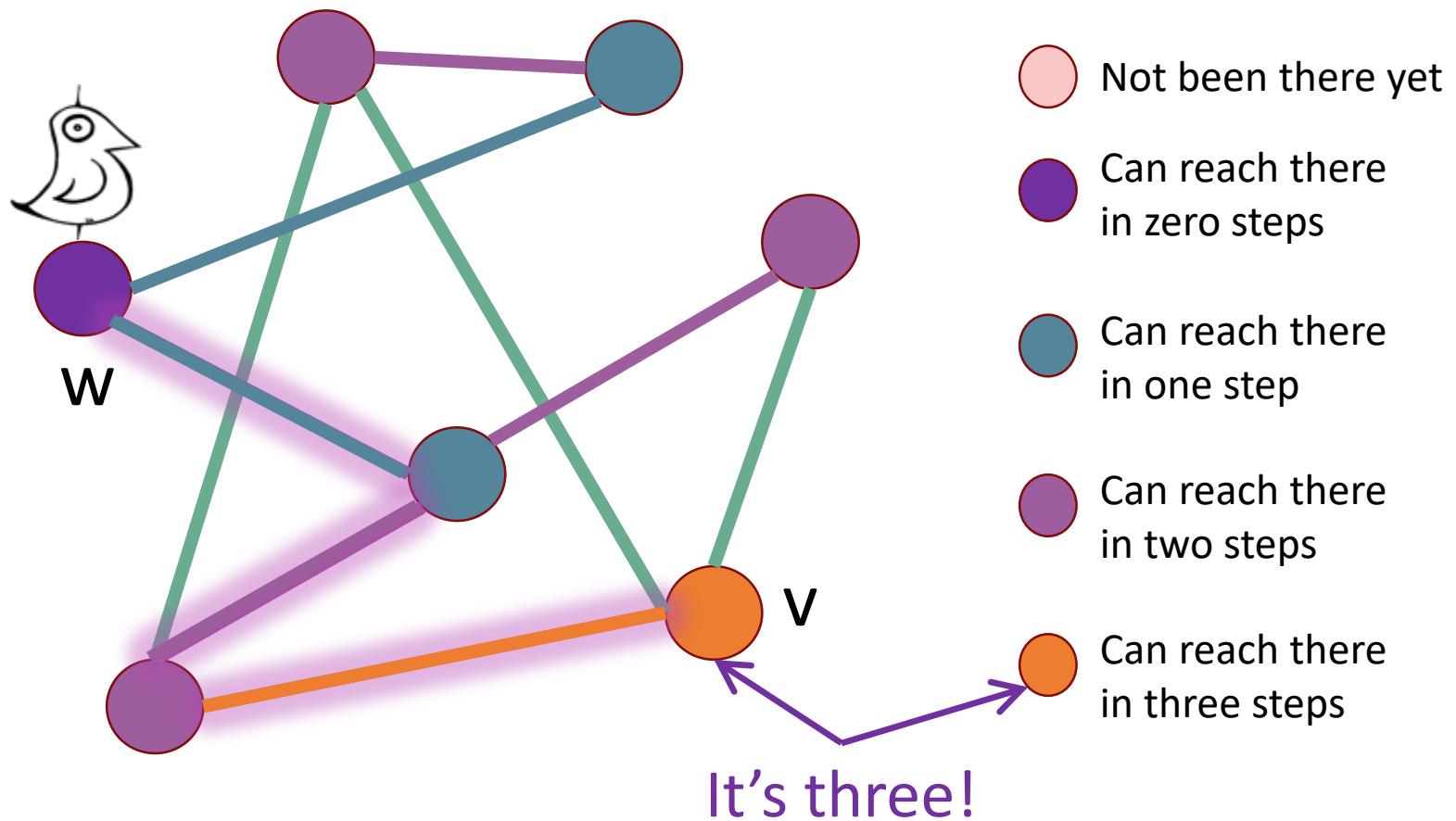
Application of BFS: shortest path

- How long is the shortest path between w and v?



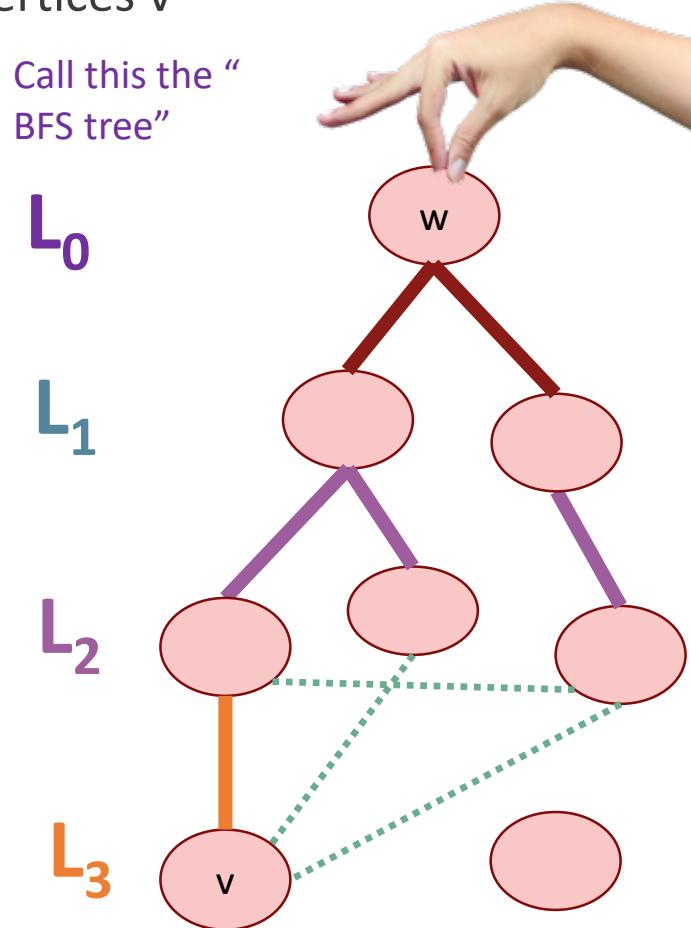
Application of BFS: shortest path

- How long is the shortest path between w and v?



Finding distance

- To find the **distance** between w and all other vertices v
 - Do a BFS starting at w
 - For all v in L_i
 - The shortest path between w and v has length i.
 - A shortest path between w and v is given by the path in the BFS tree.
 - If we never found v, the distance is infinite.



What have we learned?

- The BFS tree is useful for computing distances between pairs of vertices.
- We can find the shortest path between u and v in time $O(m)$.

Recap

- **Depth-first search**
 - Useful for **topological sorting**
 - Also **in-order traversals** of BSTs
- **Breadth-first search**
 - Useful for **finding shortest paths**
- **Both DFS, BFS:**
 - Useful for exploring graphs, finding connected components, etc.

Still open (next few lectures)

- We can now find components in undirected graphs.
 - What if we want to find strongly connected components in [directed graphs](#)?
- How can we find shortest paths in [weighted](#) graphs?

- Next Time:
 - Strongly Connected Components



Any Question?