

Lecture 9

Graphs, BFS and DFS

Announcements!

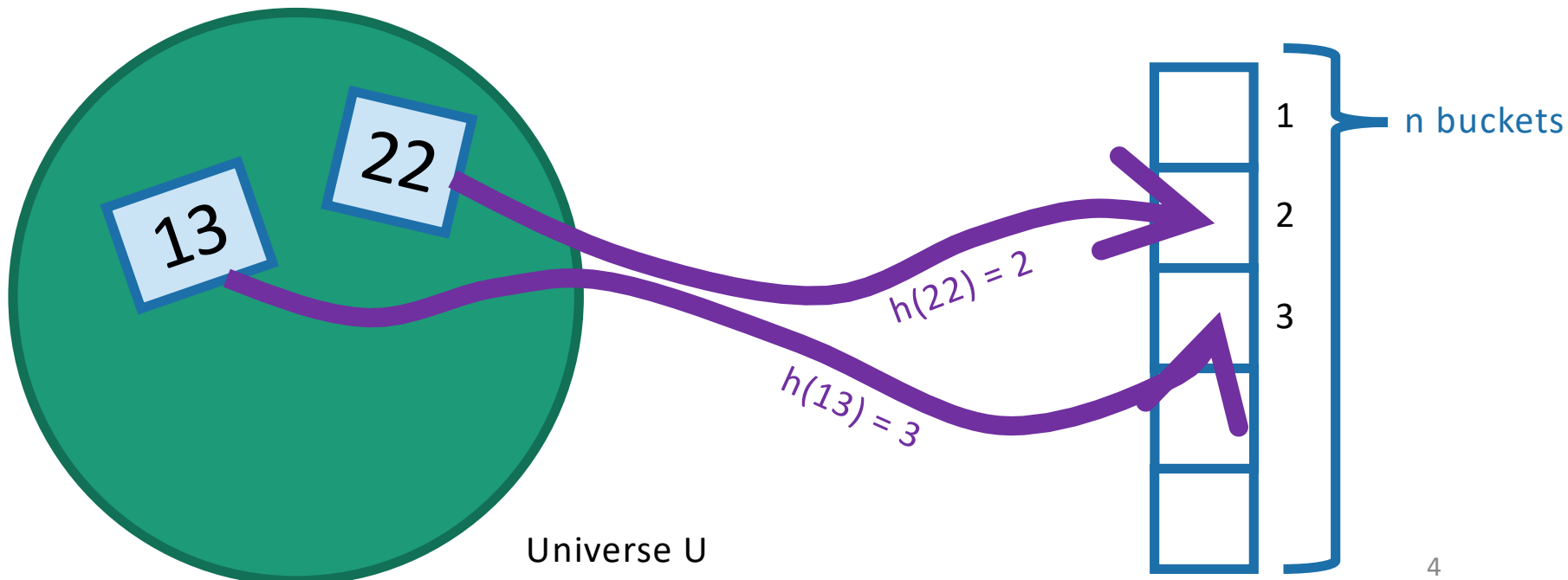
- HW3 due Friday
 - Follow the Piazza HW3 post for updates – we fixed some typos and also added a hint.
- HW4 released Friday
- I goofed up on the slides for Monday's lecture – see website for updated slides.
 - *Sorry!*

Prologue

- Real quick recap of hash functions!
- Go to section this week for a longer recap!

Hash tables

- U is the **universe**. It has size M.
- As **hash table** stores items of U in n buckets.
- A hash table comes with a **hash function** $h: U \rightarrow \{1, \dots, n\}$ which says what element goes in what bucket.



Hash families

- A hash family is a collection of hash functions.
- For example:
 - $H = \{ \text{all of the functions } h:U \rightarrow \{1,..,n\} \}$
 - $H = \{ \text{least-significant-digit, most-significant-digit} \}$
- We can build a hash table by drawing a random function uniformly at random from a hash family and using that hash function.

Example

h_0 = Most_significant_digit
 h_1 = Least_significant_digit
 $H = \{h_0, h_1\}$

2. You, the algorithm, chooses a **random** hash function $h: U \rightarrow \{0, \dots, 9\}$. Choose it randomly from H .



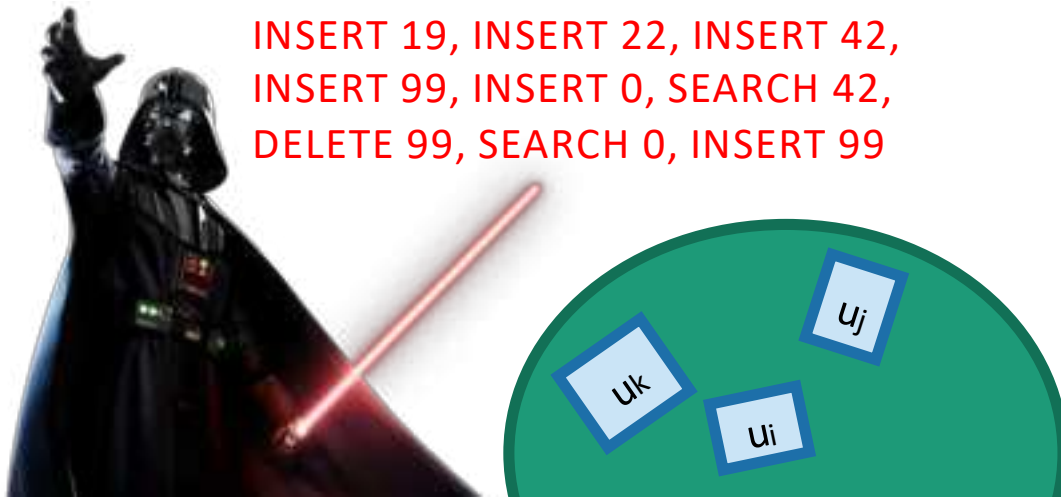
I picked h_1



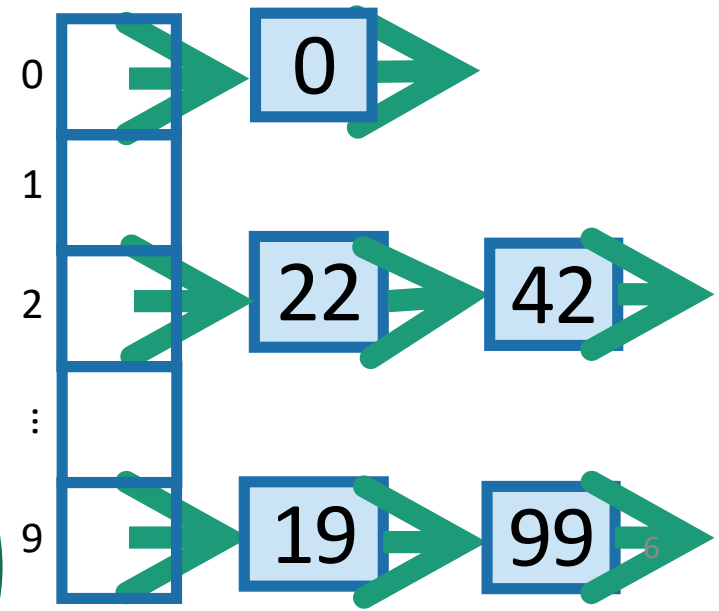
1. An adversary (who knows H) chooses any n items $u_1, u_2, \dots, u_n \in U$, and any sequence of INSERT/DELETE/SEARCH operations on those items.



INSERT 19, INSERT 22, INSERT 42,
INSERT 99, INSERT 0, SEARCH 42,
DELETE 99, SEARCH 0, INSERT 99



3. HASH IT OUT #hashpuns



Example

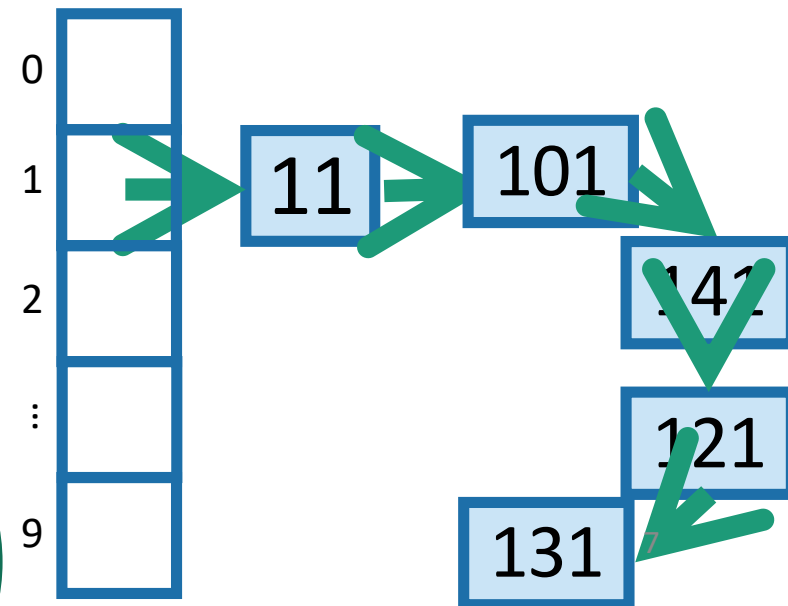
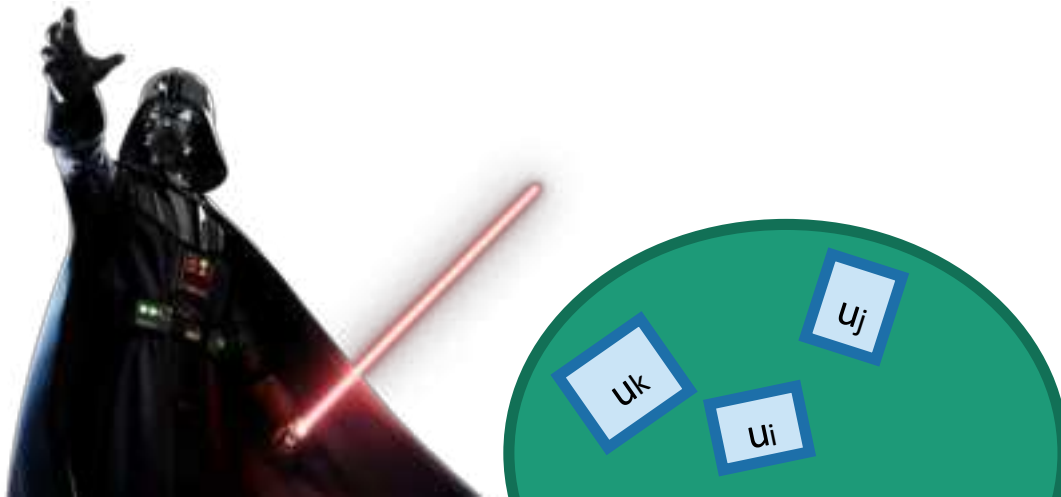
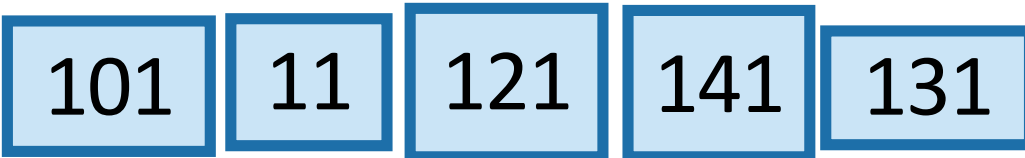
h_0 = Most_significant_digit
 h_1 = Least_significant_digit
 $H = \{h_0, h_1\}$

2. You, the algorithm, chooses a **random** hash function $h: U \rightarrow \{0, \dots, 9\}$. Choose it randomly from H .



1. An adversary (who knows H) chooses any n items $u_1, u_2, \dots, u_n \in U$, and any sequence of INSERT/DELETE/SEARCH operations on those items.

3. HASH IT OUT #hashpuns



Example

h_0 = Most_significant_digit
 h_1 = Least_significant_digit
 $H = \{h_0, h_1\}$

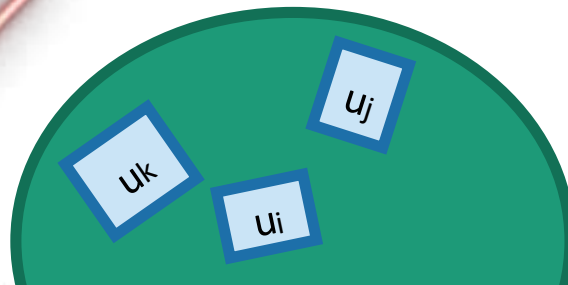
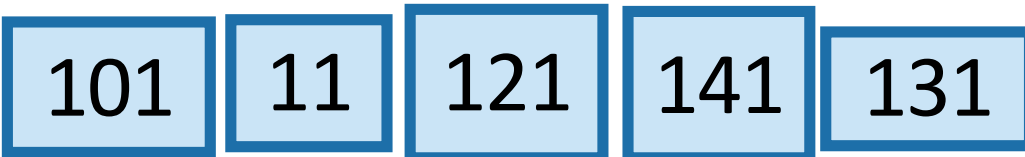
2. You, the algorithm, chooses a **random** hash function $h: U \rightarrow \{0, \dots, 9\}$. Choose it randomly from H .



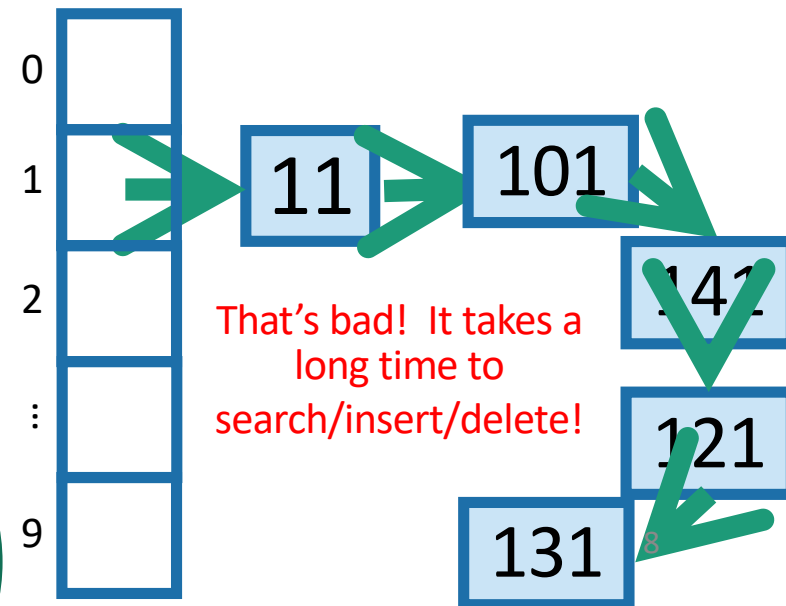
I picked h_0



1. An adversary (who knows H) chooses any n items $u_1, u_2, \dots, u_n \in U$, and any sequence of INSERT/DELETE/SEARCH operations on those items.



3. HASH IT OUT #hashpuns



Universal hash family

- H is a ***universal hash family*** if, when h is chosen uniformly at random from H,

for all $u_i, u_j \in U$ with $u_i \neq u_j$,

$$P_{h \in H} \{ h(u_i) = h(u_j) \} \leq \frac{1}{n}$$

- In English: For any pair of items in the universe, the probability that they end up in the same bucket is small.
- We came up with this definition because it implied that (in expectation) not too many items land in any given bucket.
- There exist universal hash families of size $O(M^2)$.

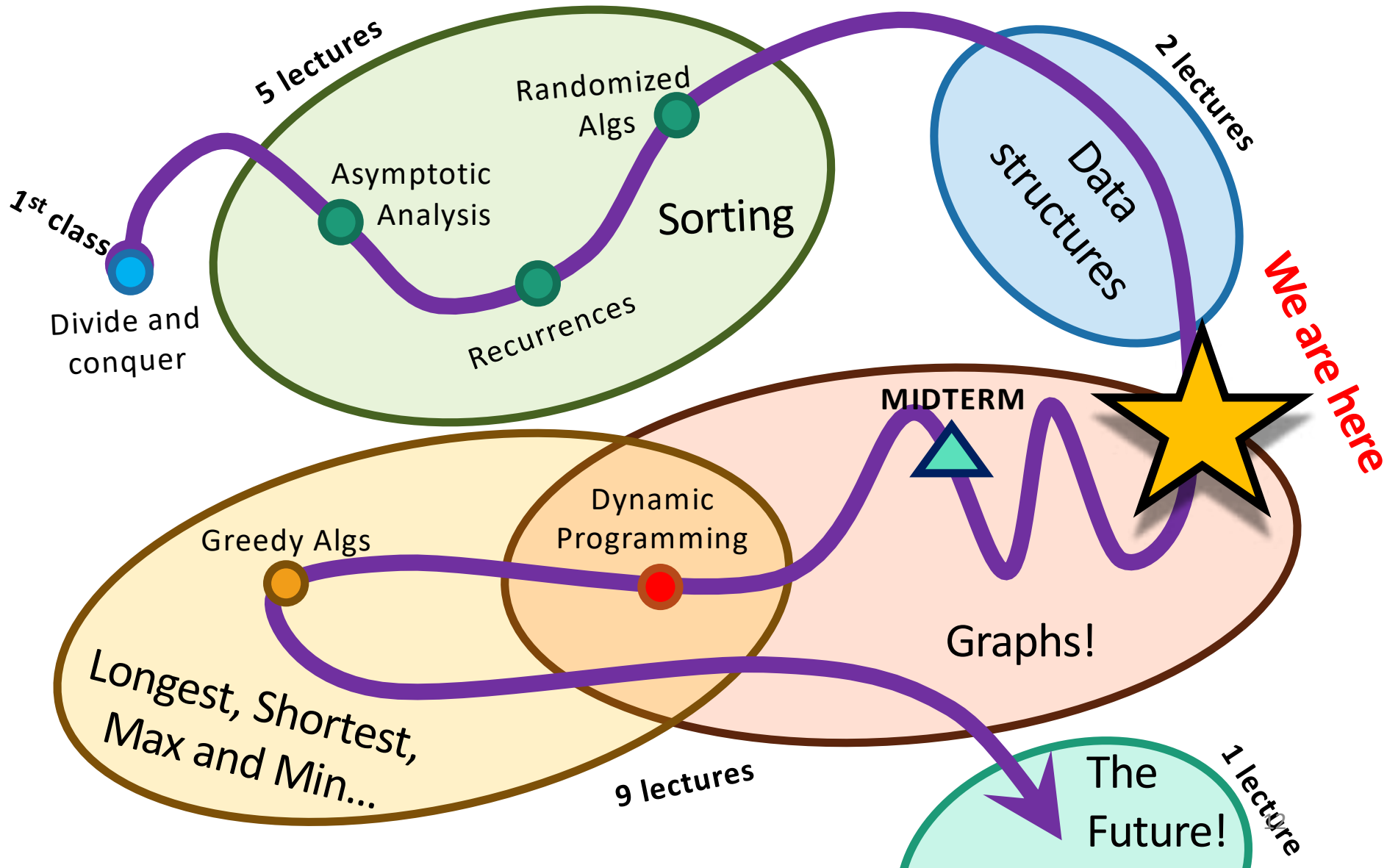
Conclusion

- We can build a hash table that:
 - supports INSERT/DELETE/SEARCH in $O(1)$ *expected* time
 - requires $O(n \log(M))$ bits of space.
 - $O(n)$ buckets
 - $O(n)$ items with $\log(M)$ bits per item
 - $O(\log(M))$ bits to store the hash function

Questions about hash functions?

- Section this week
- Office hours
- CLRS
- (Optional) lecture notes posted on website

Roadmap

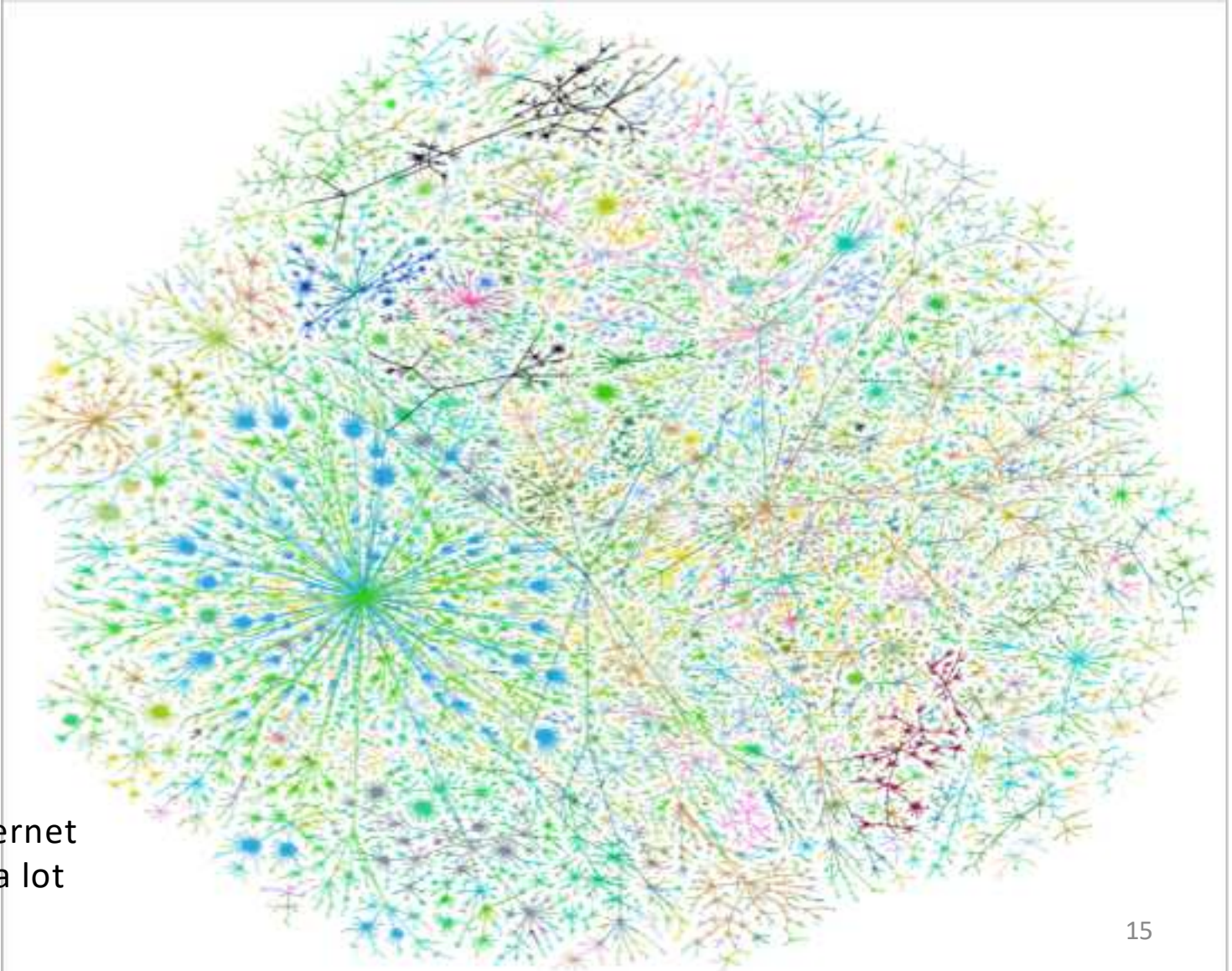


Outline

- Part 0: Graphs and terminology
- Part 1: Depth-first search
 - Application: topological sorting
 - Application: in-order traversal of BSTs
- Part 2: Breadth-first search
 - Application: shortest paths
 - Application (if time): is a graph bipartite?

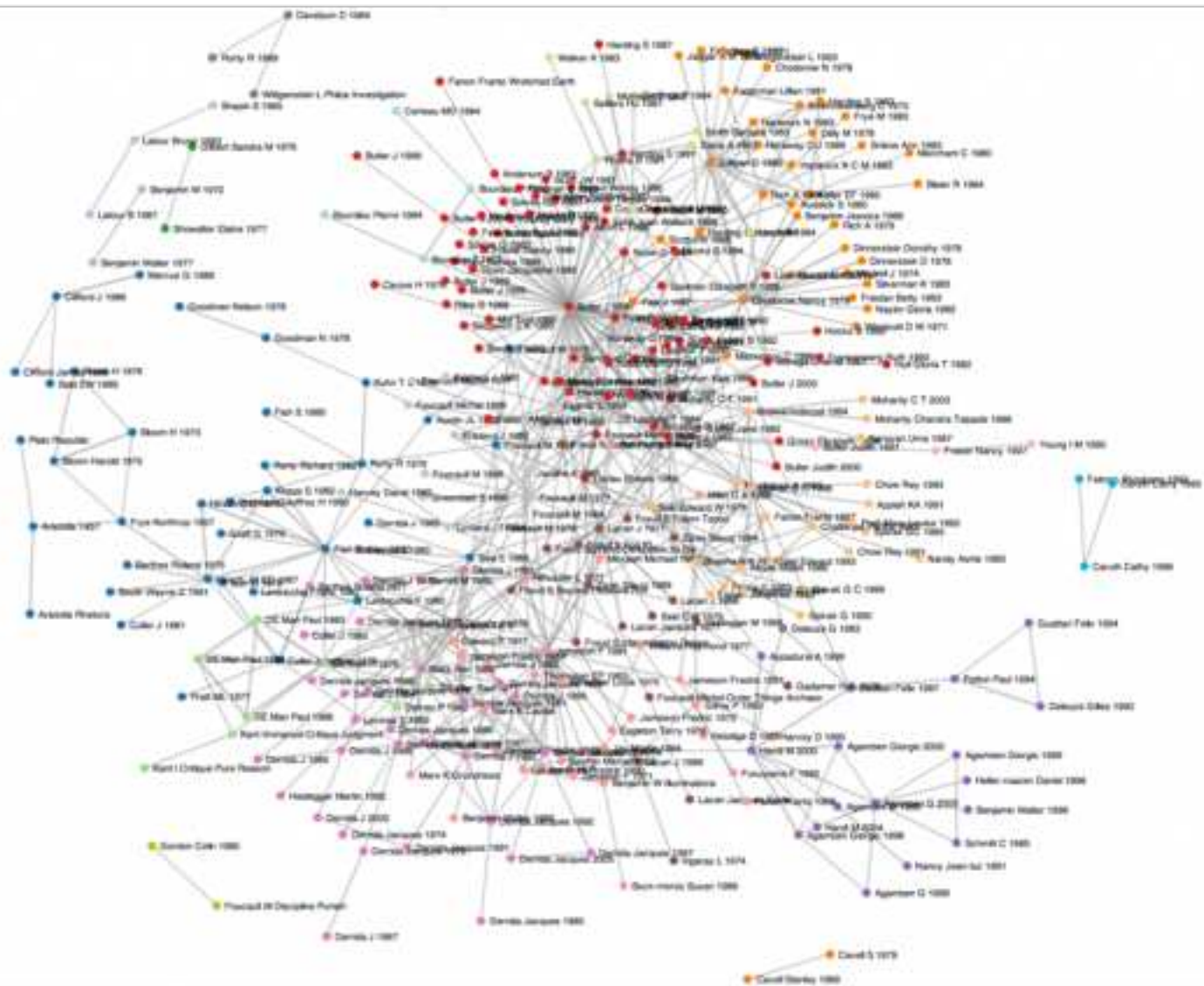
Part 0: Graphs

Graphs



Graph of the internet
(circa 1999...it's a lot
bigger now...)

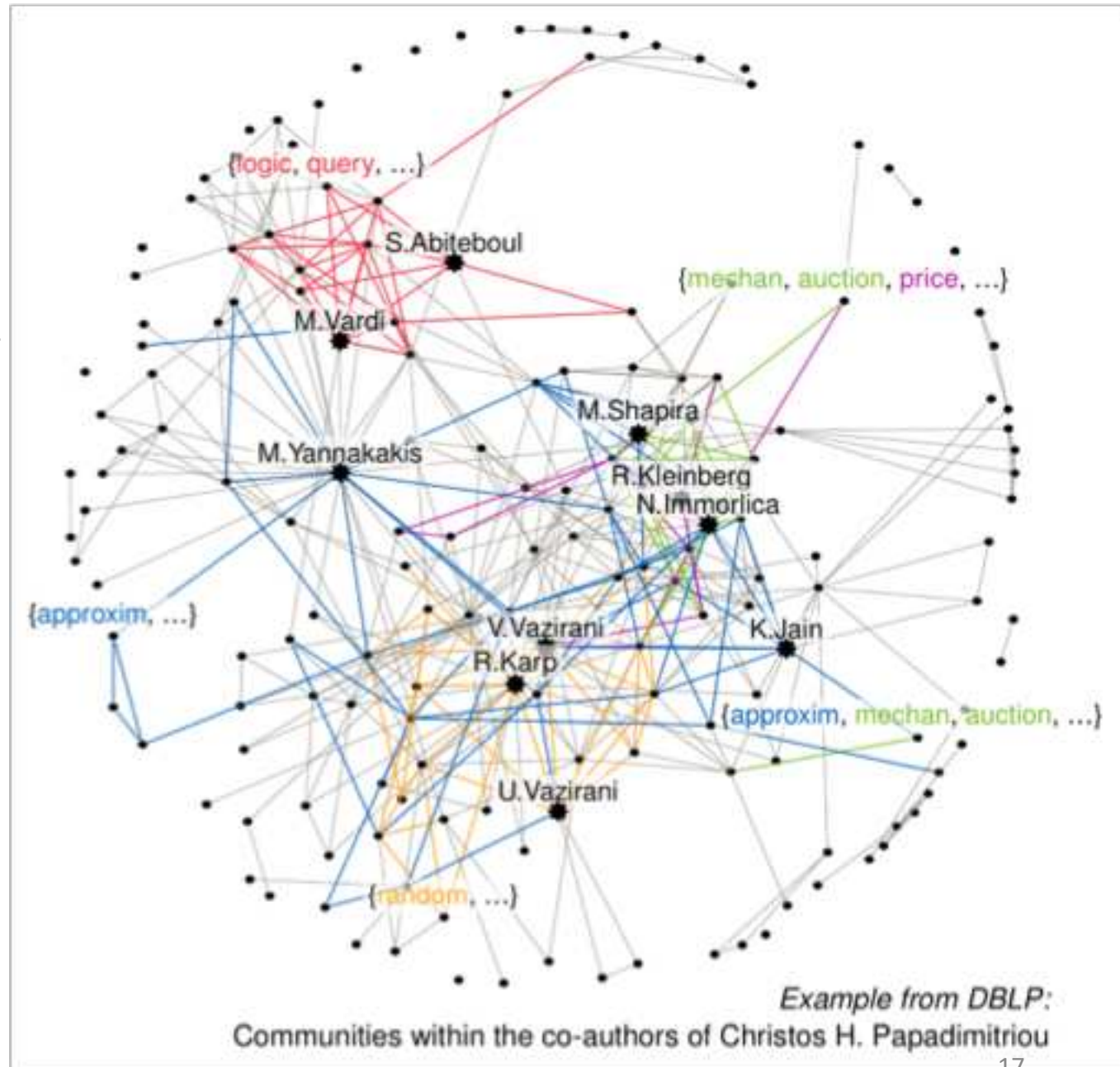
Graphs



Citation graph of
literary theory
academic papers

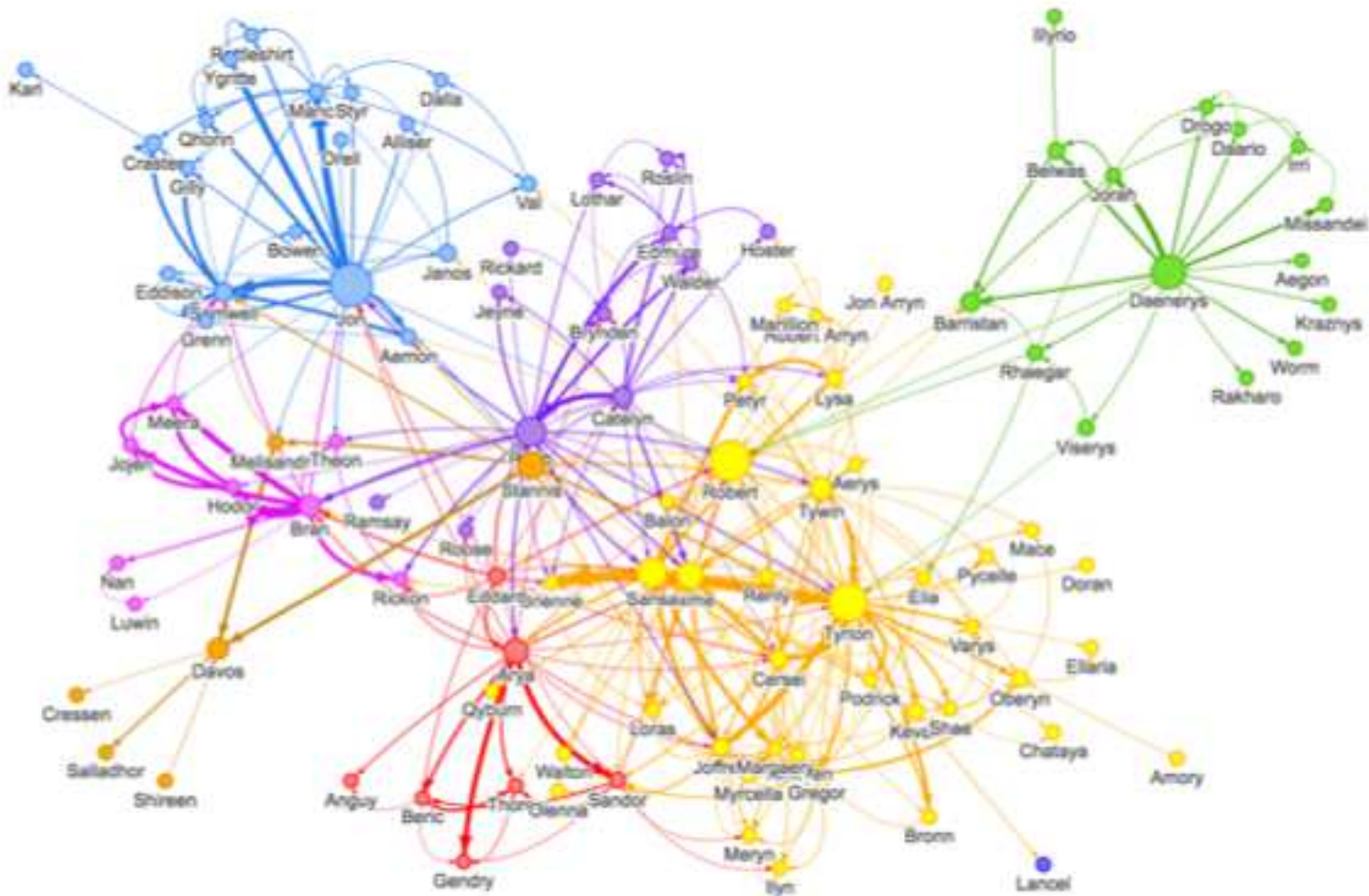
Graphs

Theoretical Computer
Science academic
communities



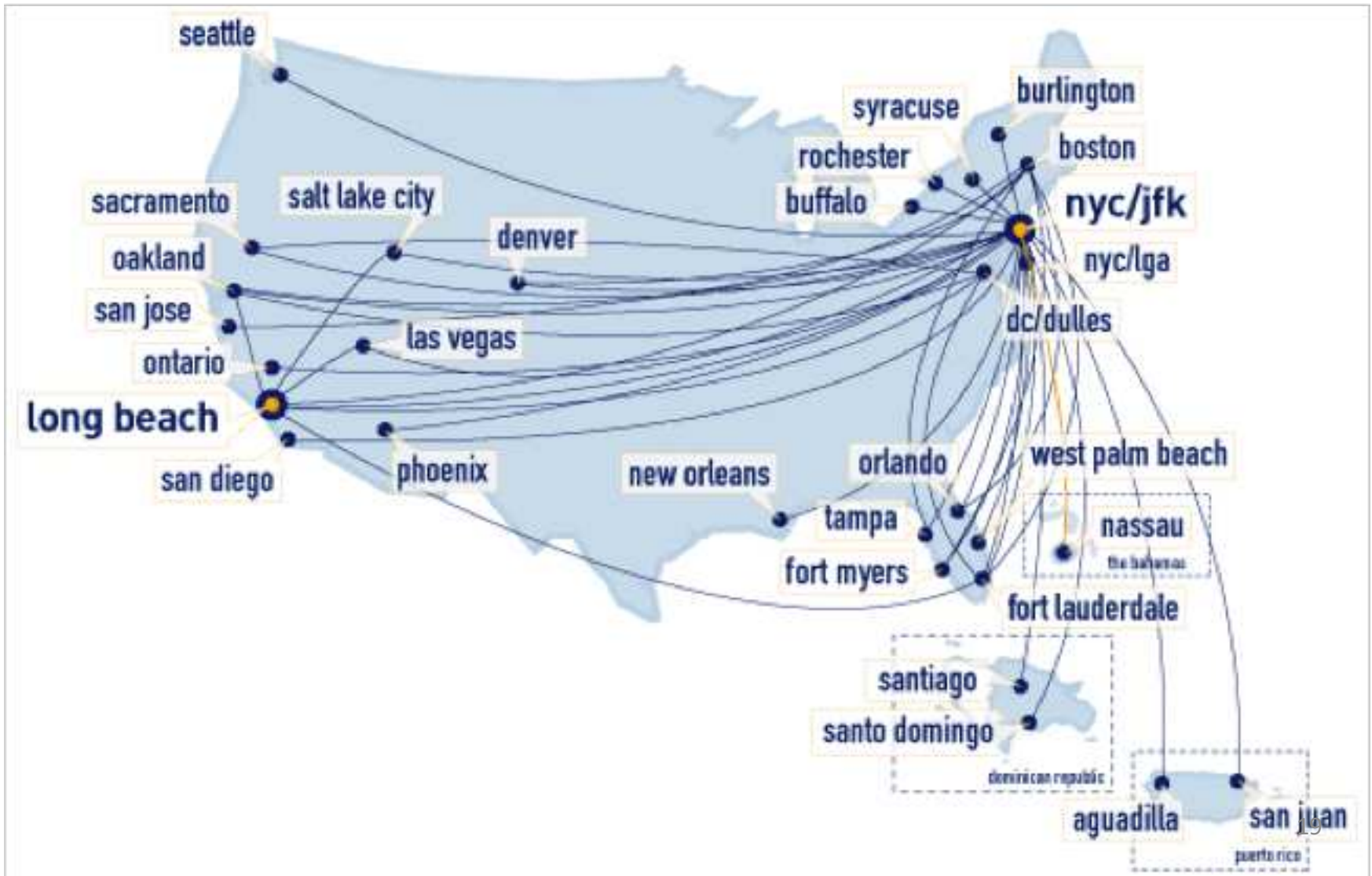
Graphs

Game of Thrones Character Interaction Network



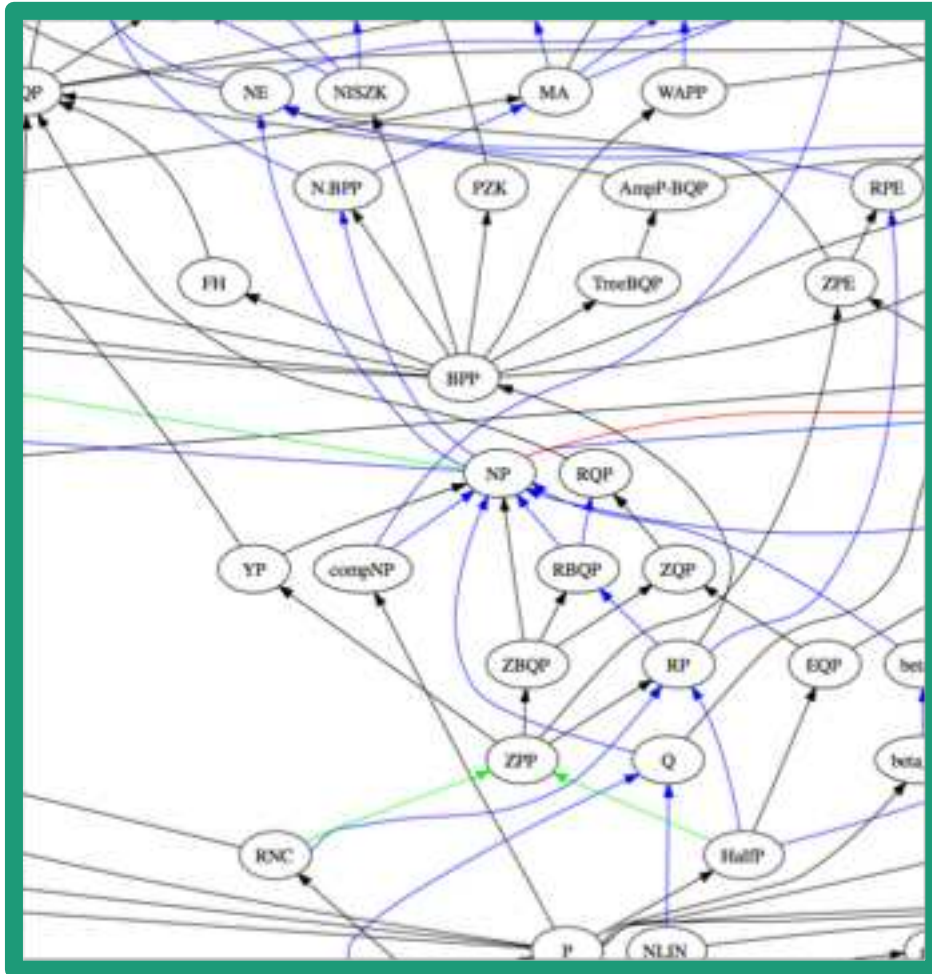
Graphs

jetblue flights



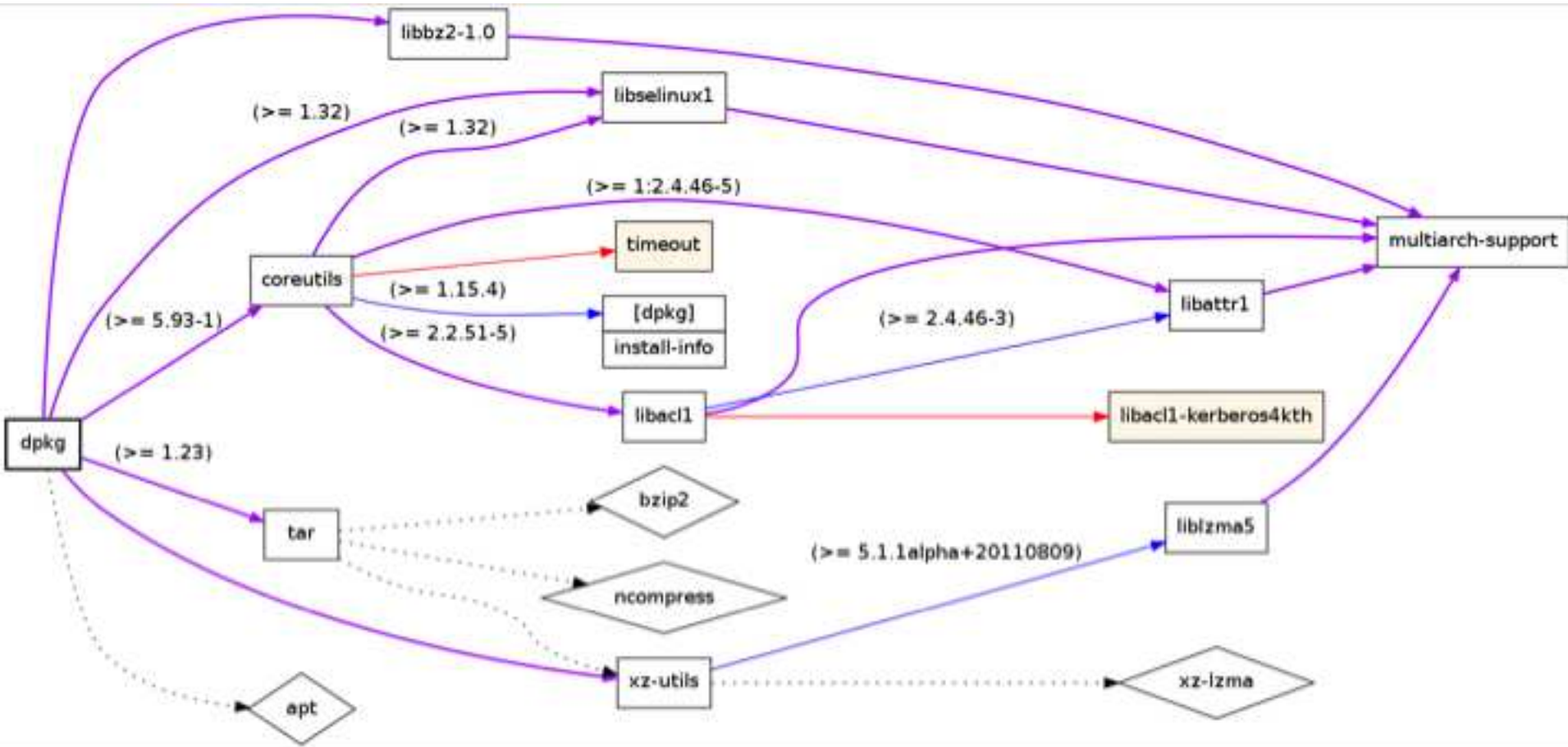
Graphs

Complexity Zoo
containment graph



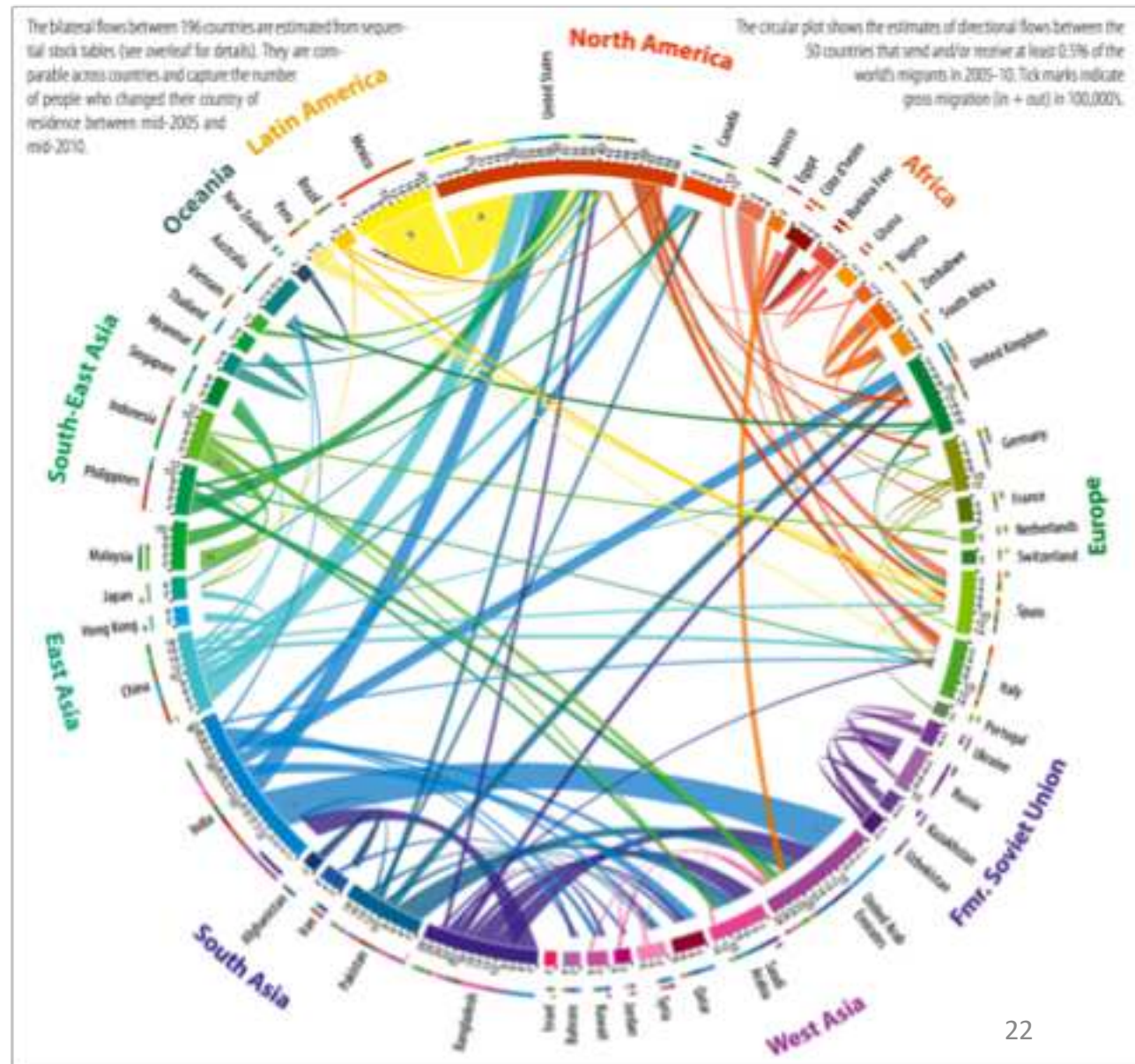
Graphs

debian dependency (sub)graph



Graphs

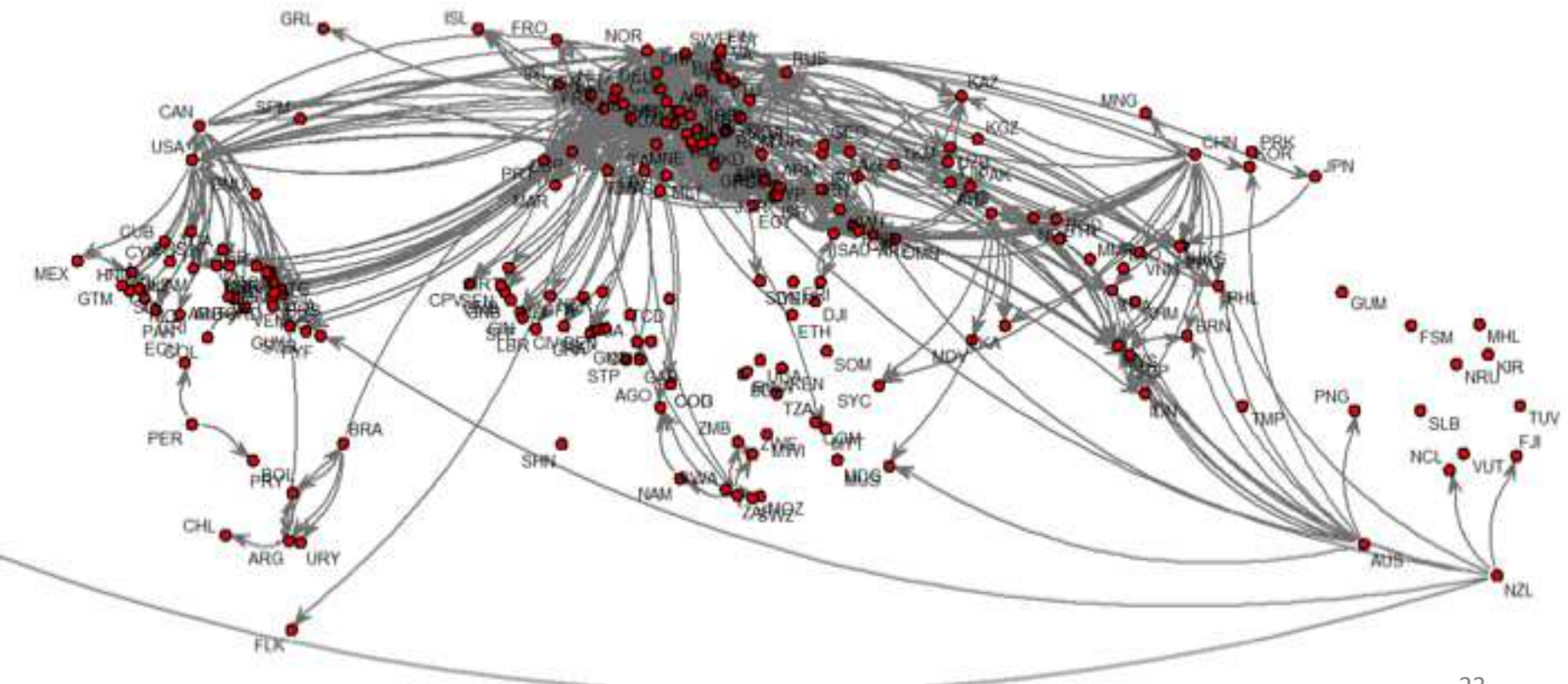
Immigration
flows



Graphs

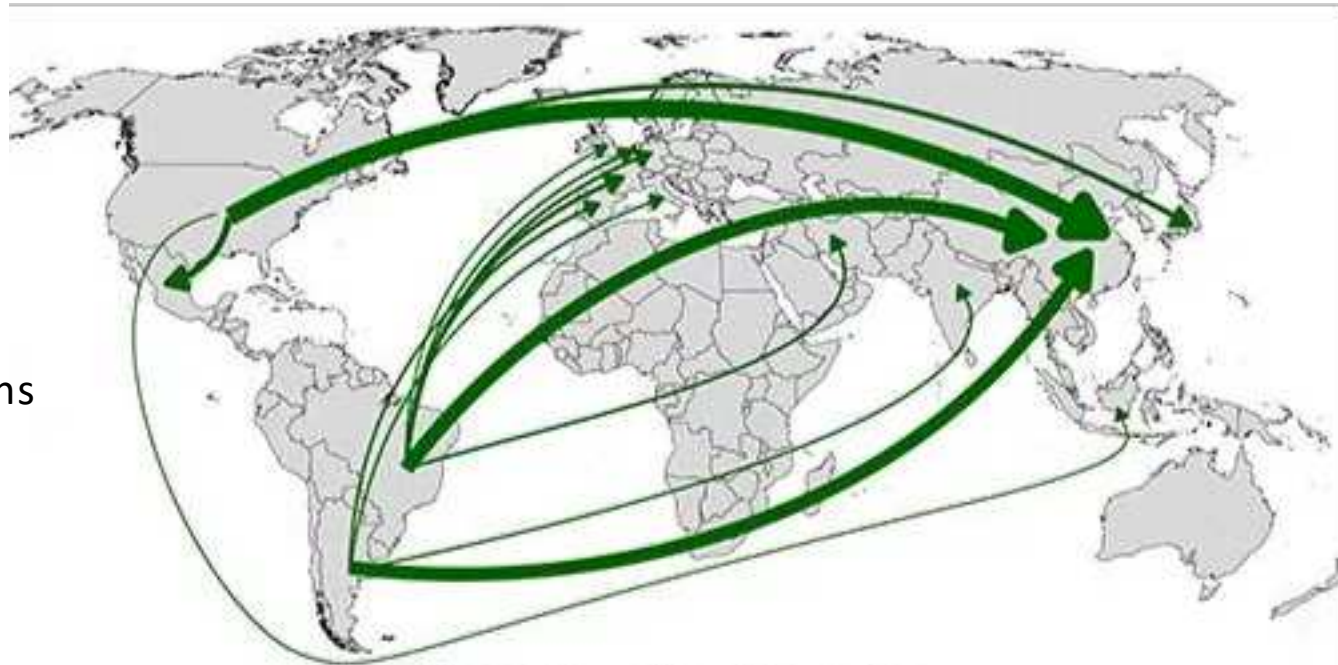
Potato trade

World trade in fresh potatoes, flows over 0.1 m US\$ average 2005-2009

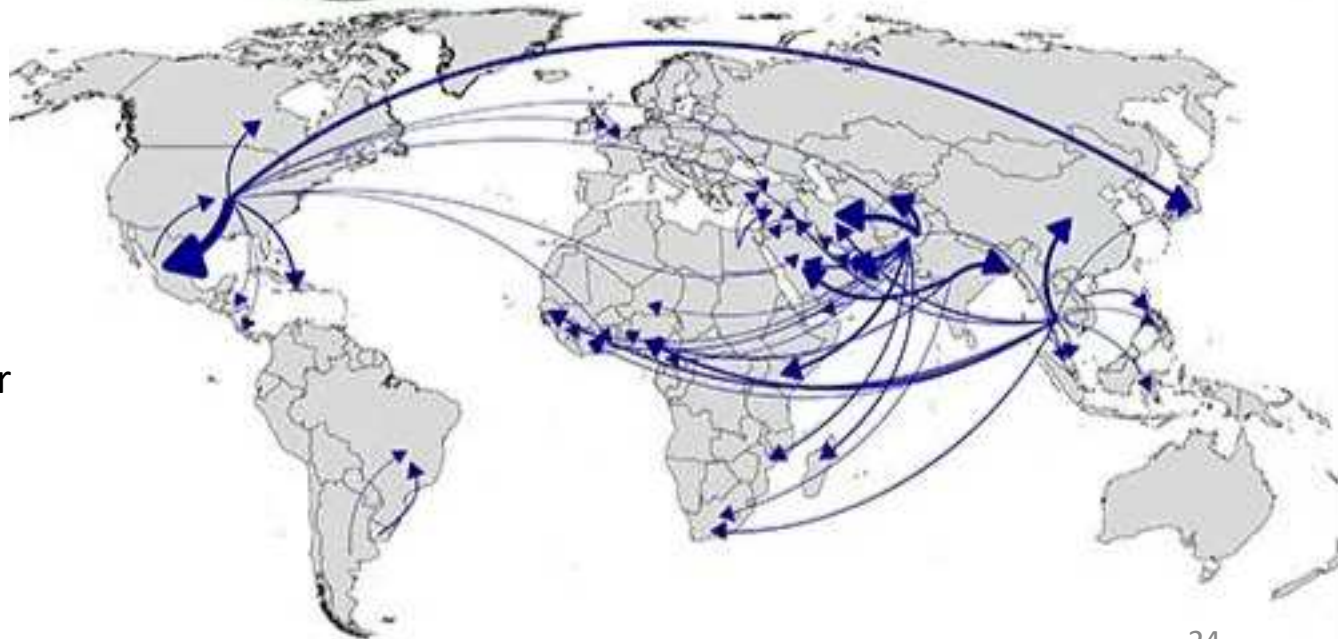


Graphs

Soybeans

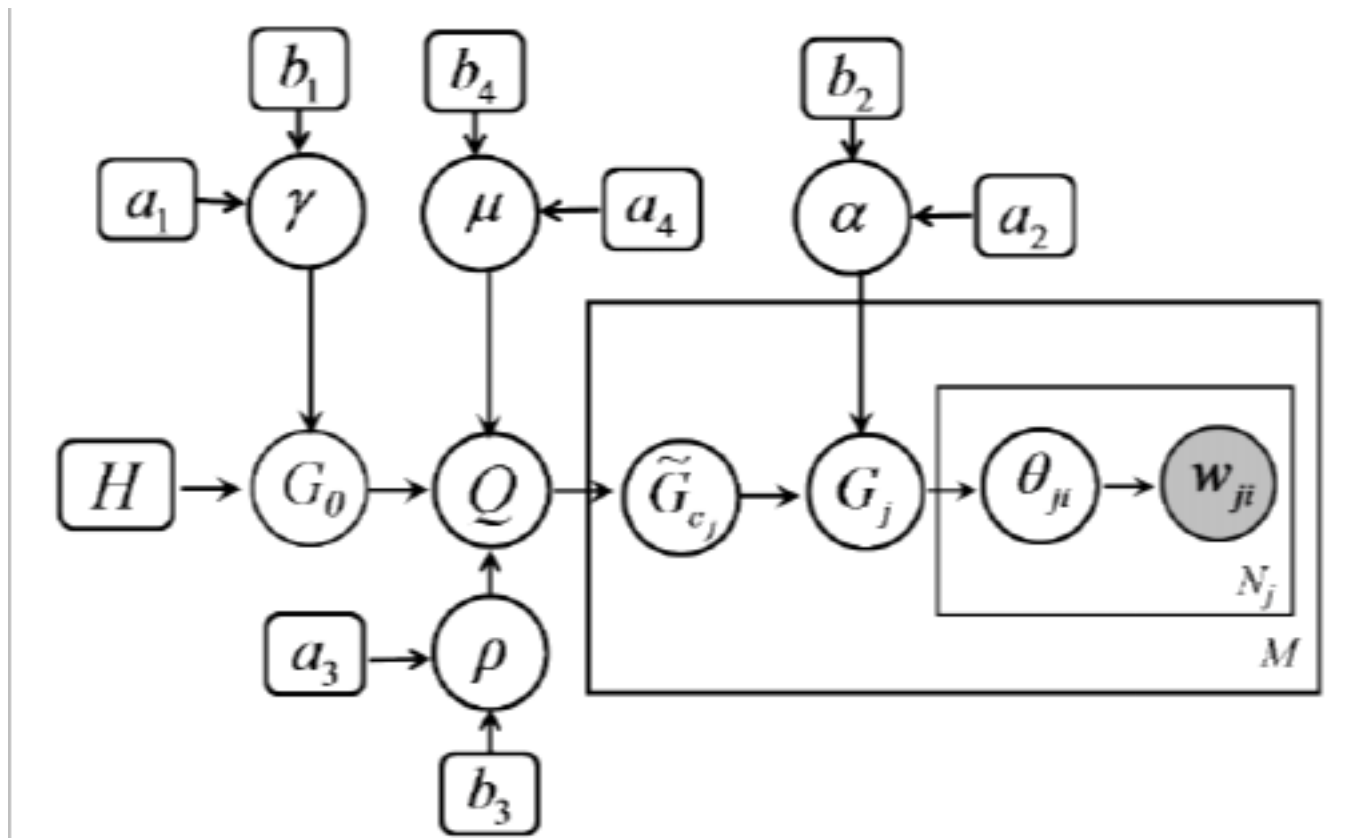


Water



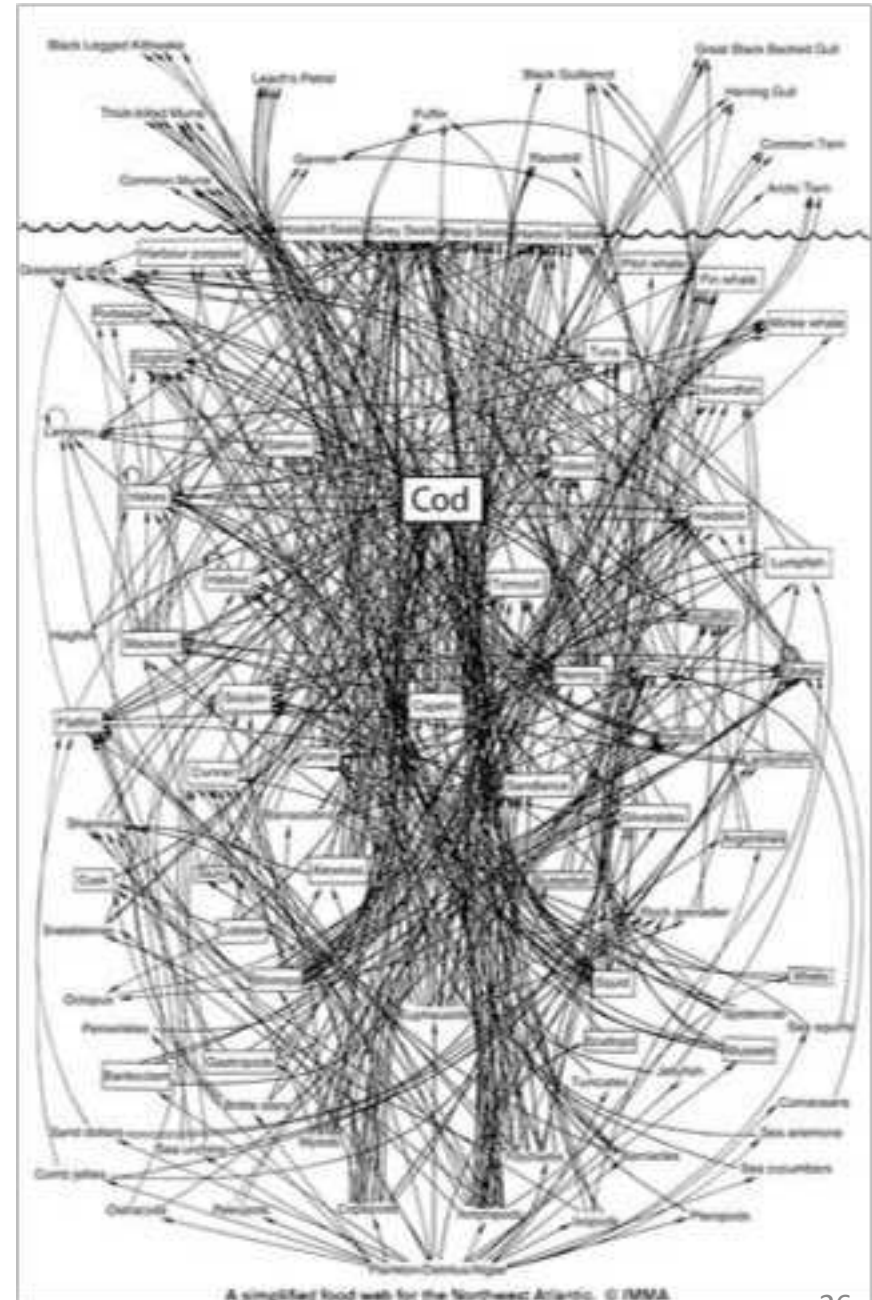
Graphs

Graphical models



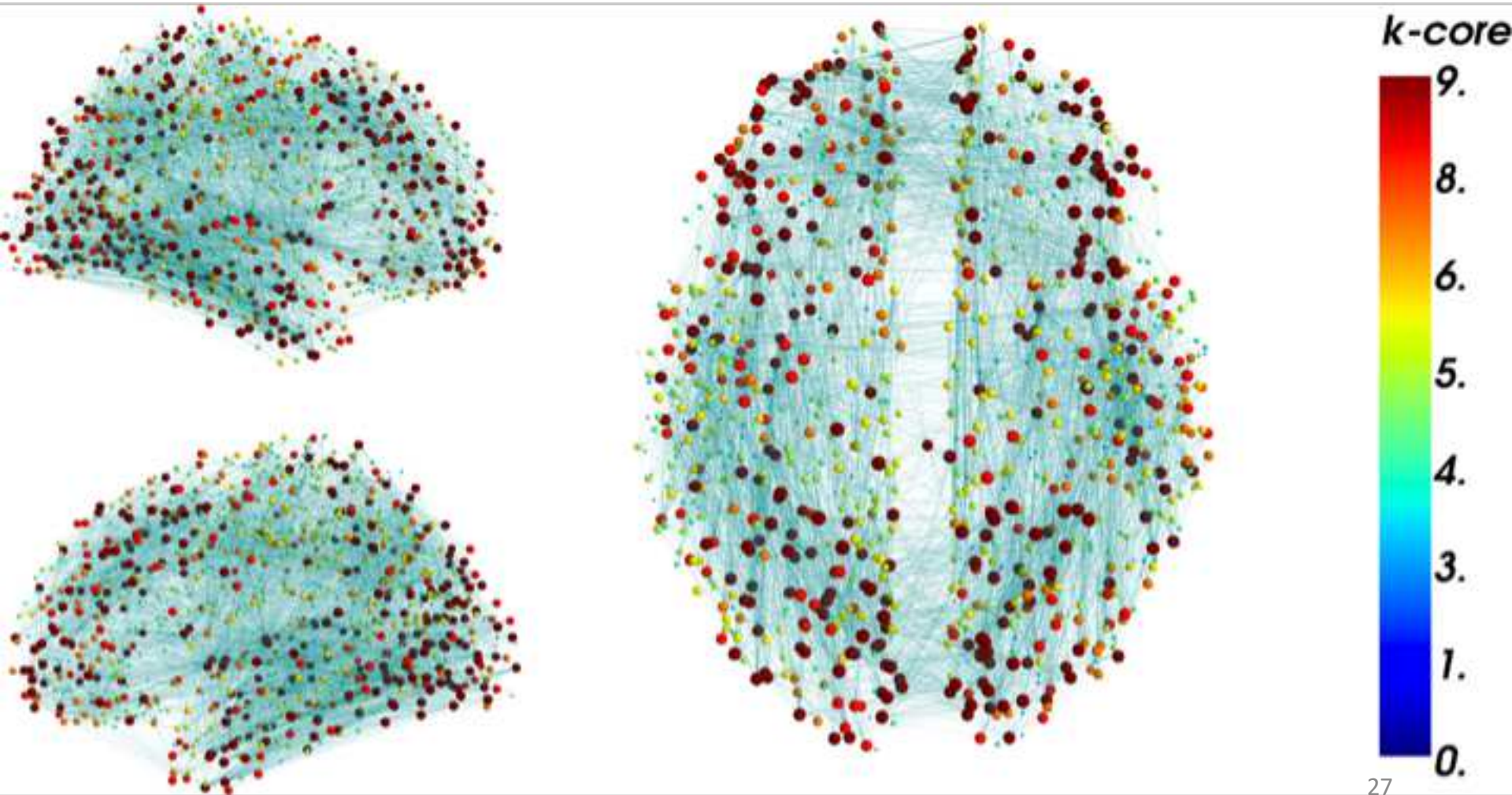
Graphs

What eats what in
the Atlantic ocean?



Graphs

Neural connections
in the brain

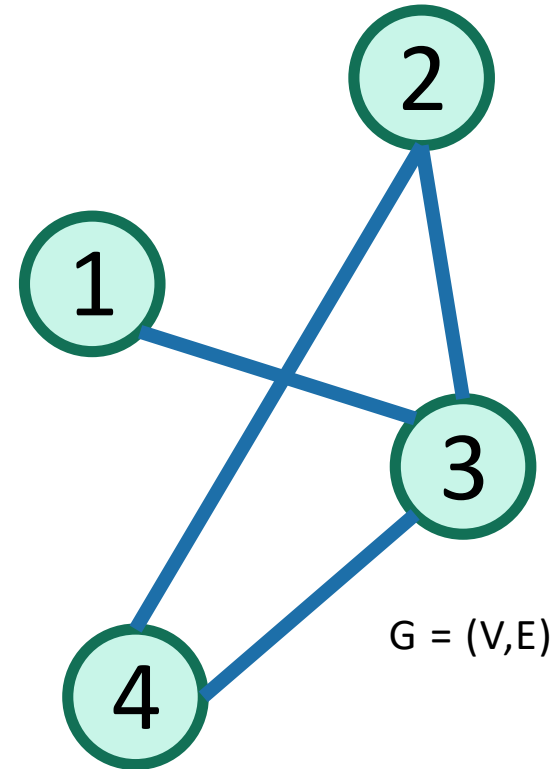


Graphs

- **There are a lot of graphs.**
- We want to answer questions about them.
 - Efficient routing?
 - Community detection/clustering?
 - From pre-lecture exercise:
 - Computing Bacon numbers
 - Signing up for classes without violating pre-req constraints
 - How to distribute fish in tanks so that none of them will fight.
- This is what we'll do for the next several lectures.

Undirected Graphs

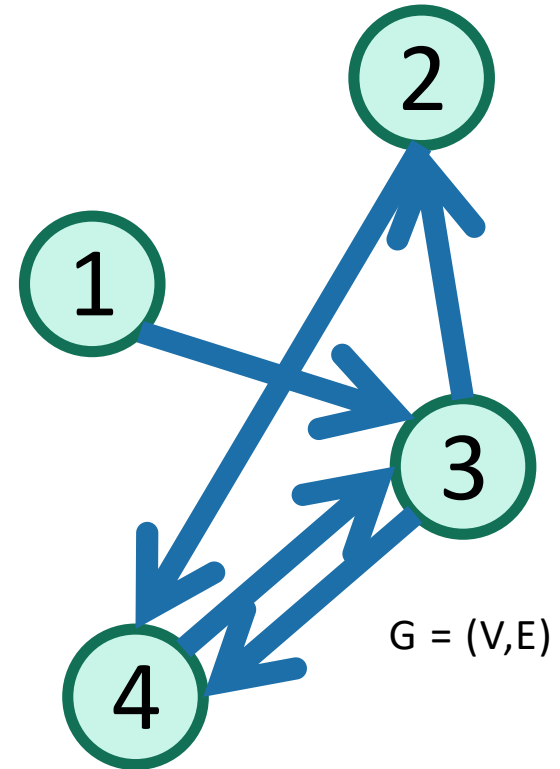
- Has vertices and edges
 - V is the set of vertices
 - E is the set of edges
 - Formally, a graph is $G = (V, E)$
- Example
 - $V = \{1, 2, 3, 4\}$
 - $E = \{ \{1, 3\}, \{2, 4\}, \{3, 4\}, \{2, 3\} \}$



- 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
 - $V = \{1, 2, 3, 4\}$
 - $E = \{ (1, 3), (2, 4), (3, 4), (4, 3), (3, 2) \}$

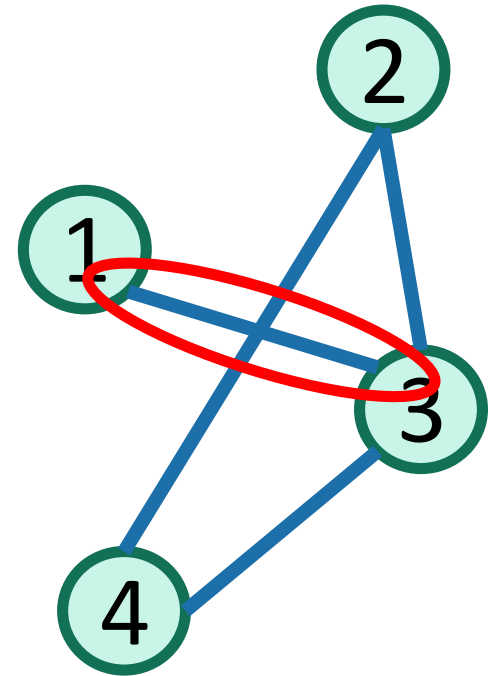


- The **in-degree** of vertex 4 is 2.
- The **out-degree** of vertex 4 is 1.
- Vertex 4's **incoming neighbors** are 2, 3.
- Vertex 4's **outgoing neighbor** is 3.

How do we represent graphs?

- Option 1: adjacency matrix

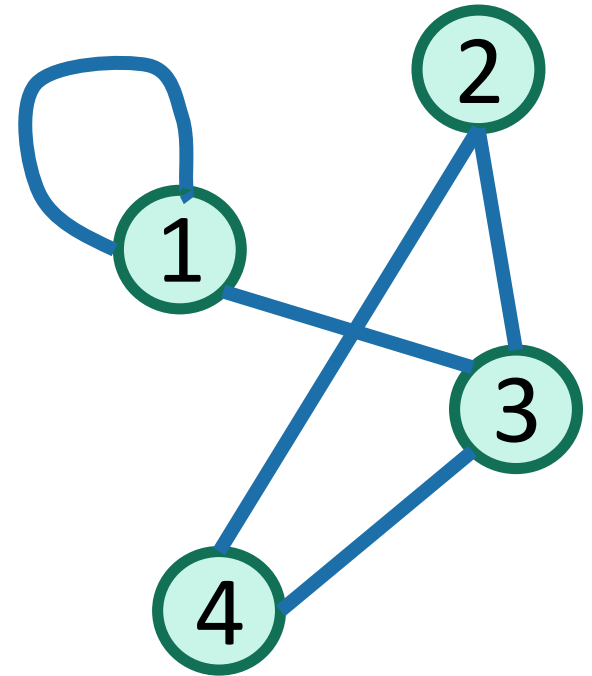
	1	2	3	4
1	0	0	1	0
2	0	0	1	1
3	1	1	0	1
4	0	1	1	0



How do we represent graphs?

- Option 1: adjacency matrix

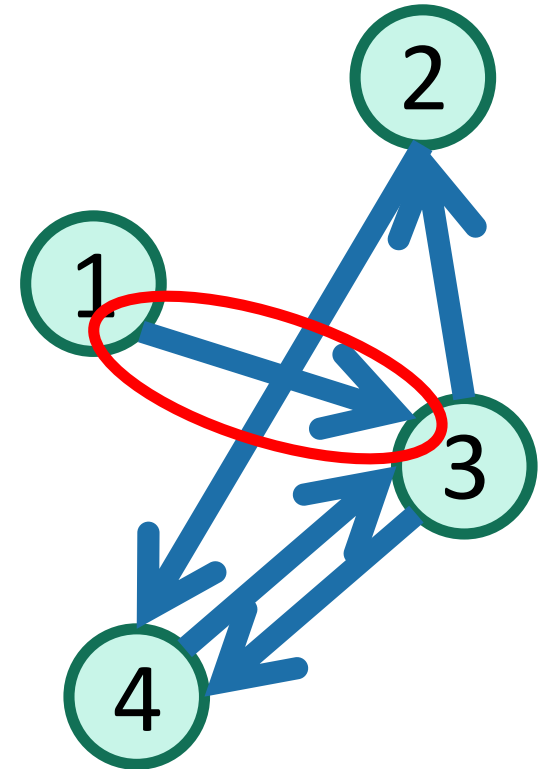
	1	2	3	4
1	1	0	1	0
2	0	0	1	1
3	1	1	0	1
4	0	1	1	0



How do we represent graphs?

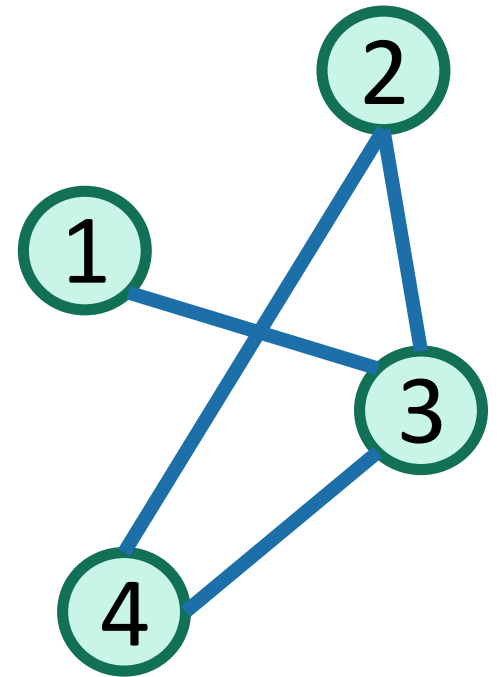
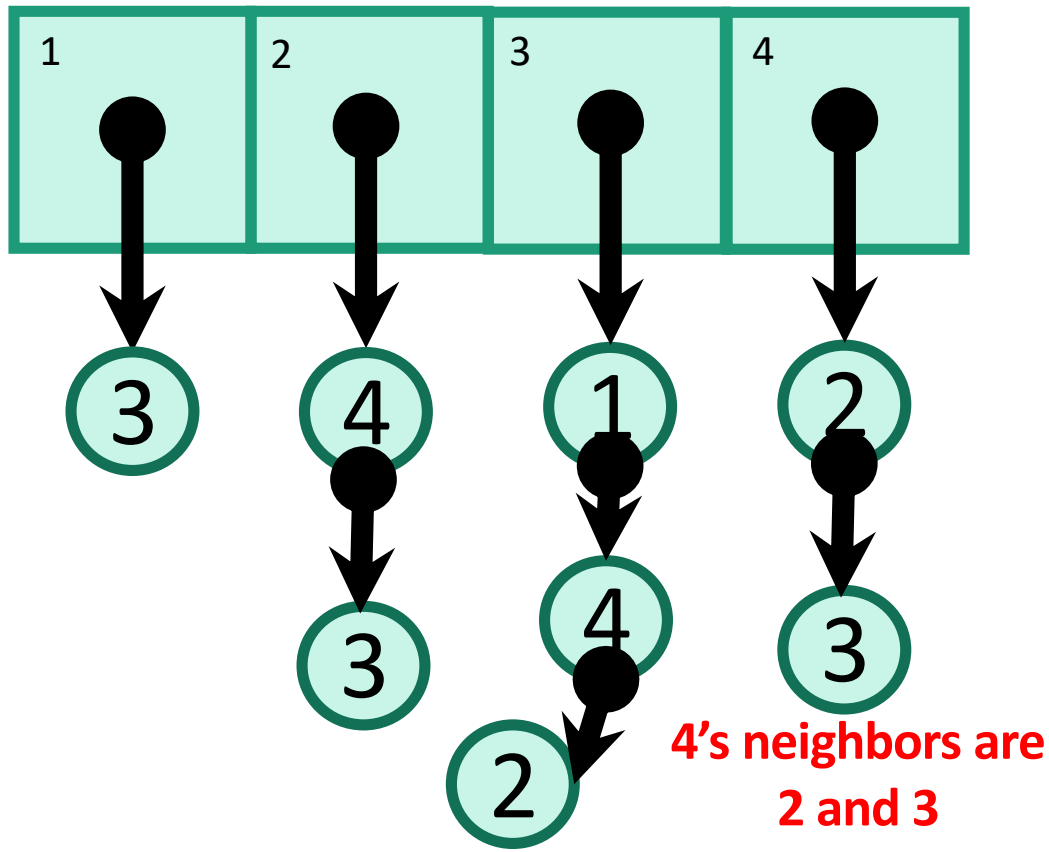
- Option 1: adjacency matrix

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



How do we represent graphs?

- Option 2: adjacency lists.



How would you modify this for directed graphs?



In either case

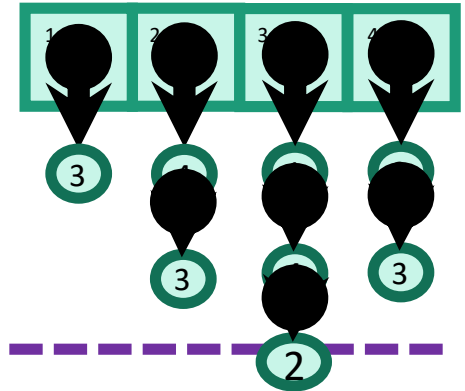
- Vertices can store other information
 - Attributes (name, IP address, ...)
 - helper info for algorithms that we will perform on the graph
- Want to be able to do the following operations:
 - **Edge Membership**: Is edge e in E ?
 - **Neighbor Query**: What are the neighbors of vertex v ?

Trade-offs

Say there are n vertices
and m edges.

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

Generally better
for **sparse** graphs



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

$O(1)$

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

Neighbor query
Give me v 's neighbors.

$O(n)$

$O(\deg(v))$

Space requirements

$O(n^2)$

$O(n + m)$

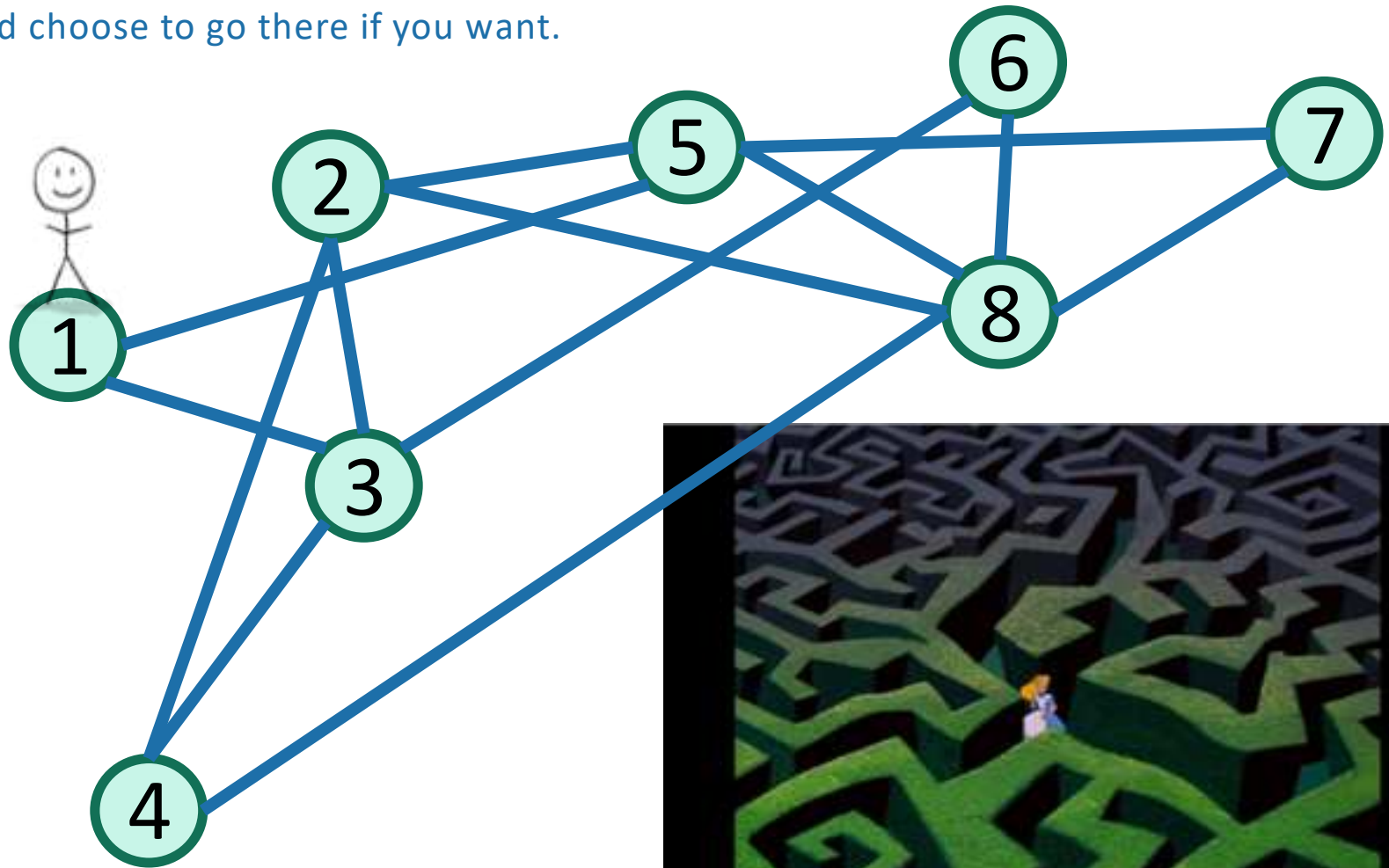
See Lecture 9 IPython notebook for an actual
implementation!

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

Part 1: Depth-first search

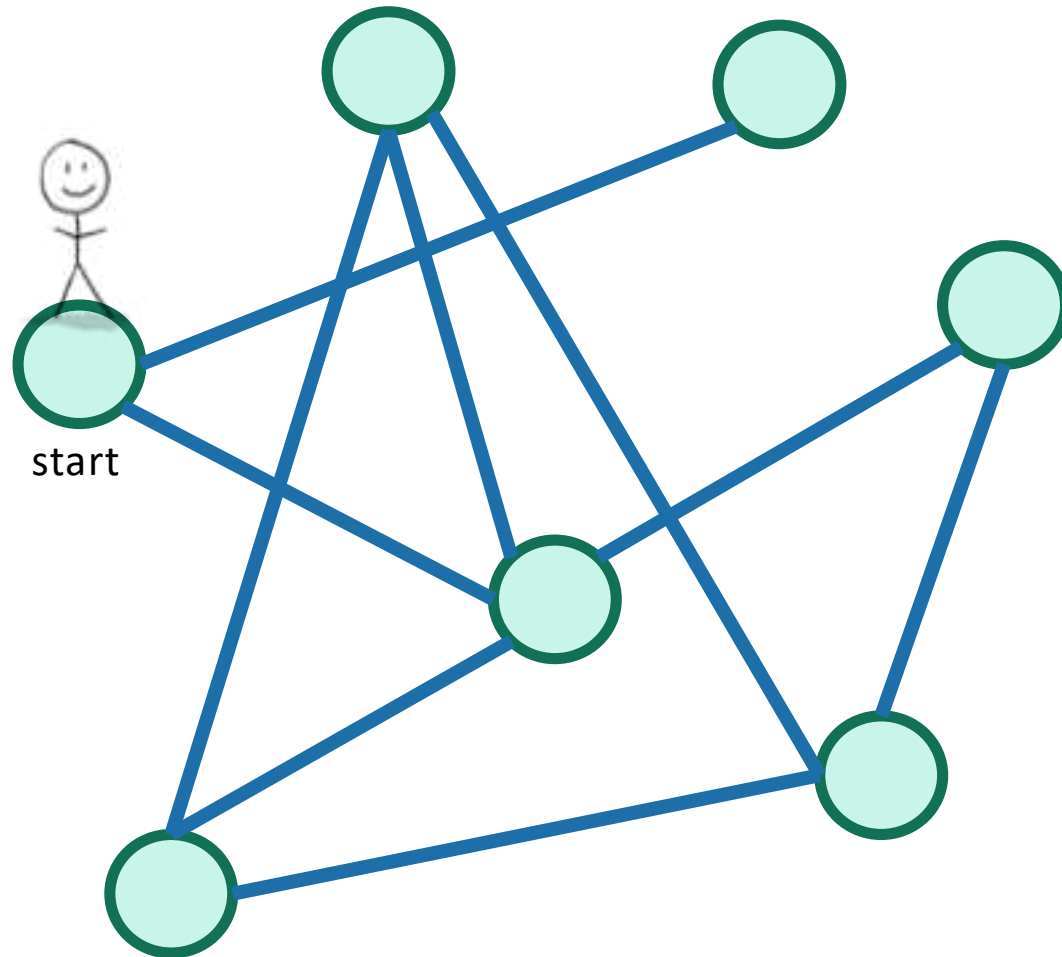
How do we explore a graph?

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



Depth First Search

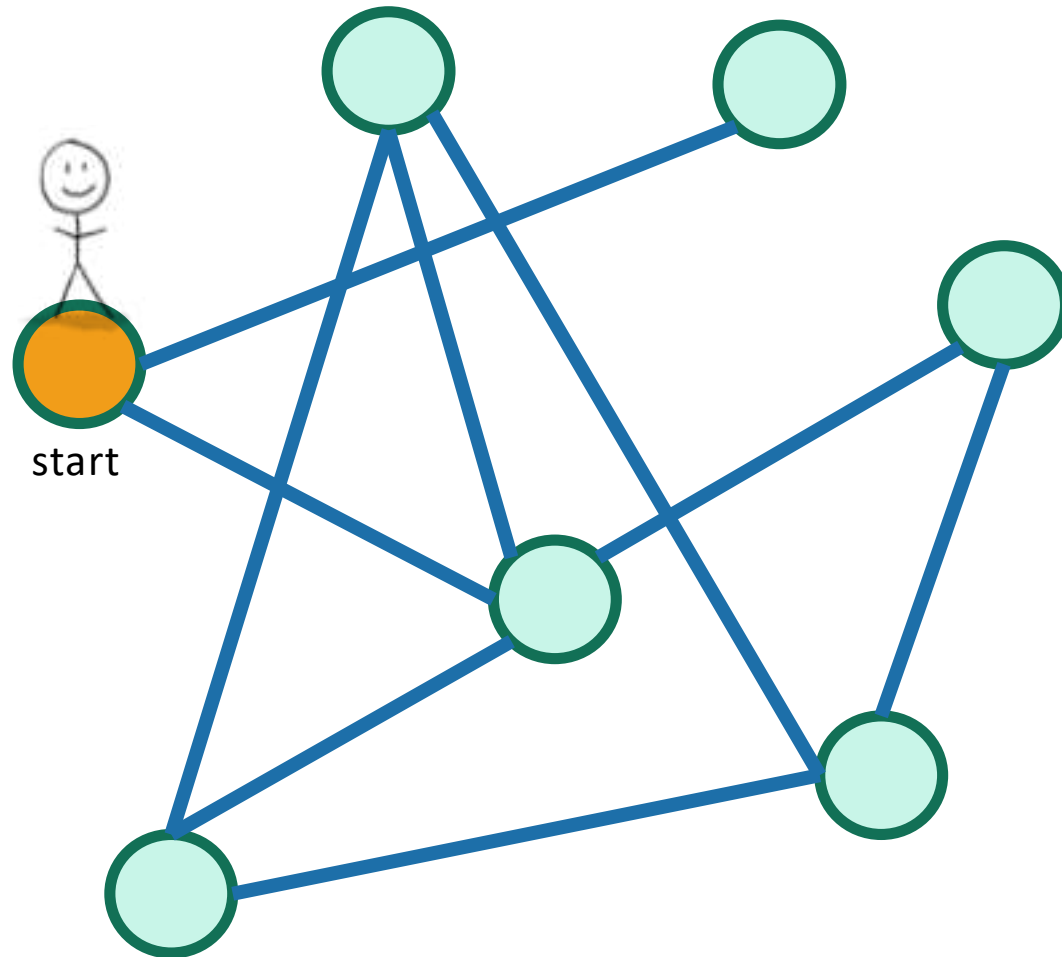
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Depth First Search

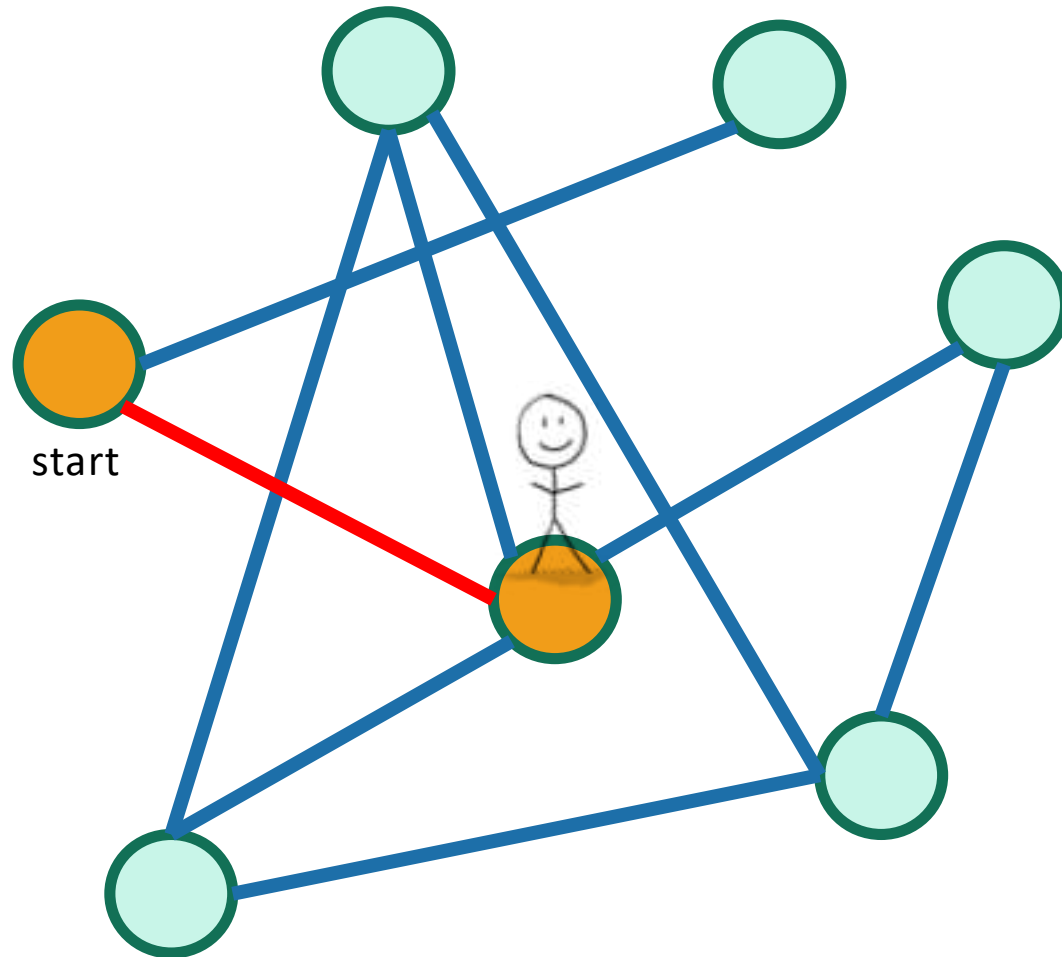
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Depth First Search

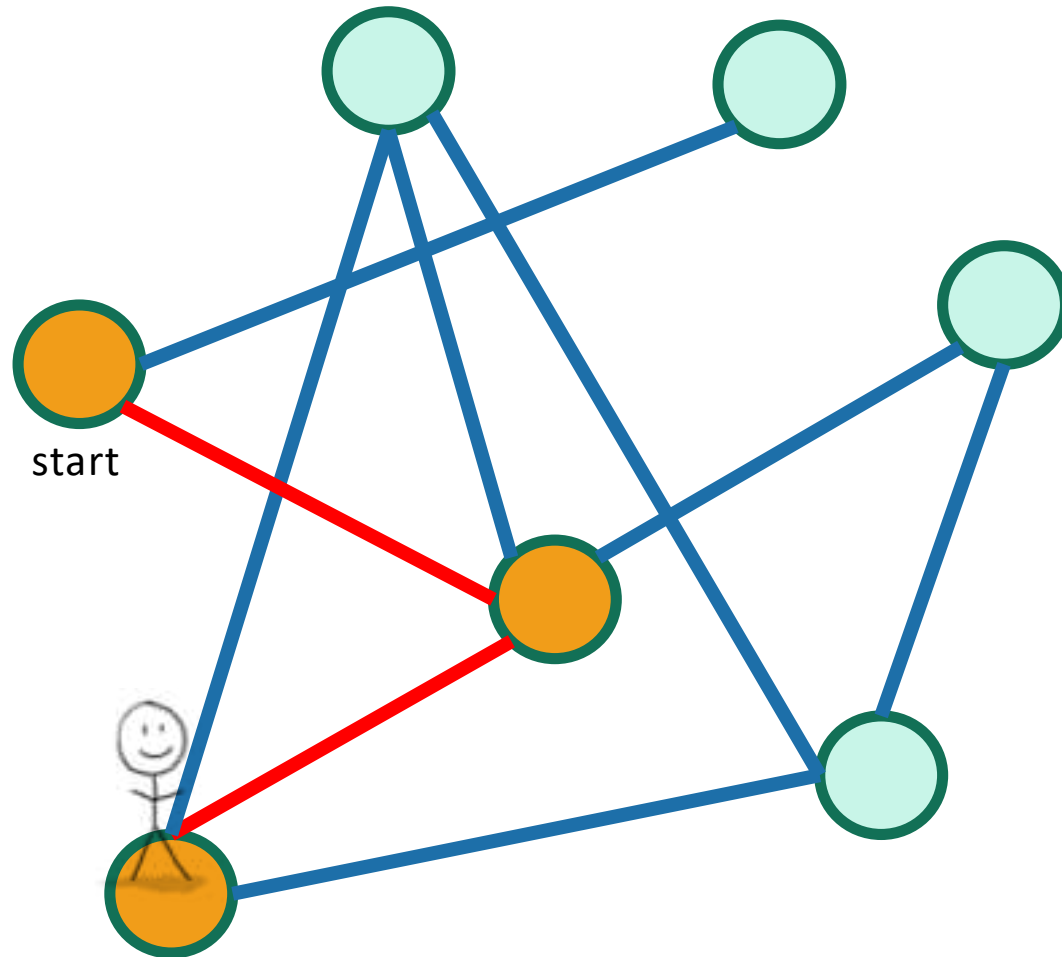
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Depth First Search

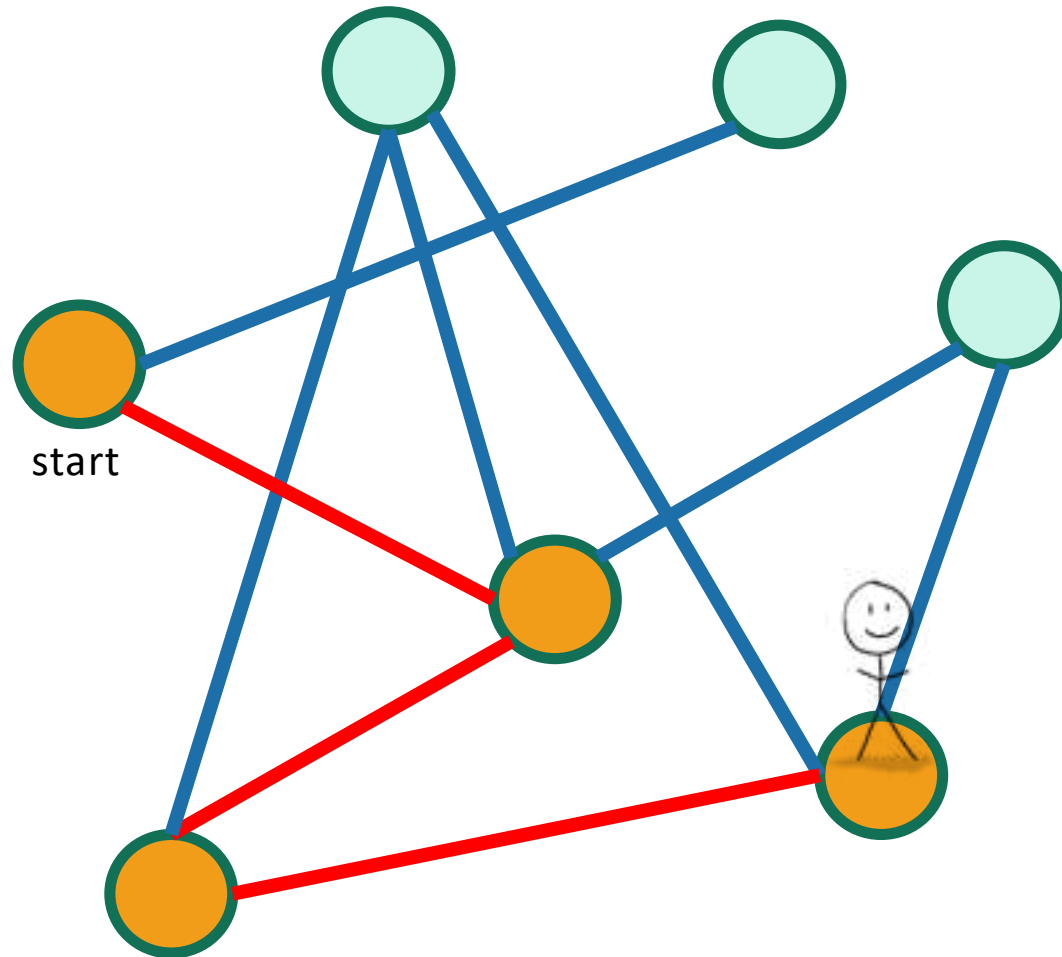
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Depth First Search

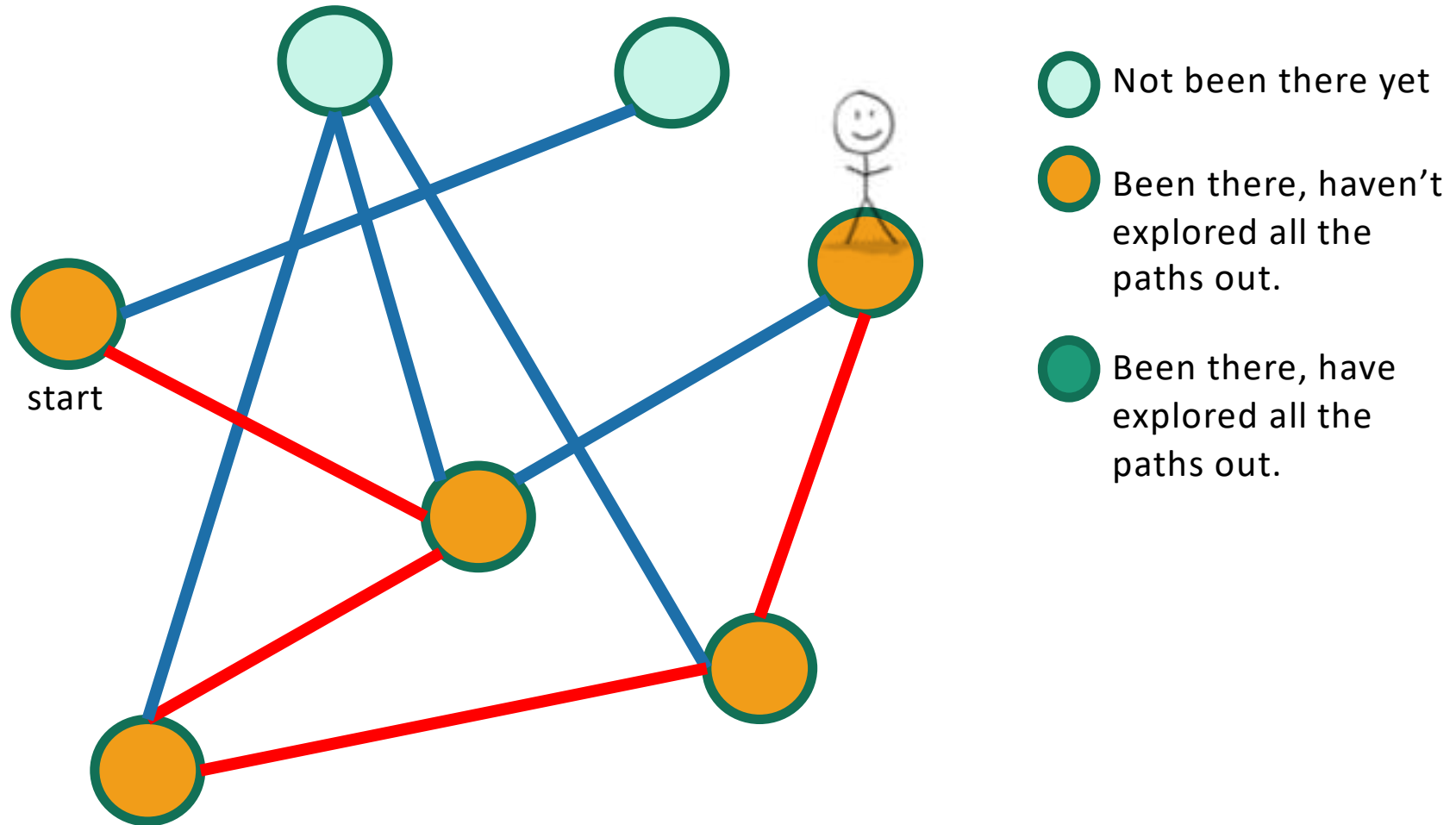
Exploring a labyrinth with chalk and a piece of string



- Not been there yet
- Been there, haven't explored all the paths out.
- Been there, have explored all the paths out.

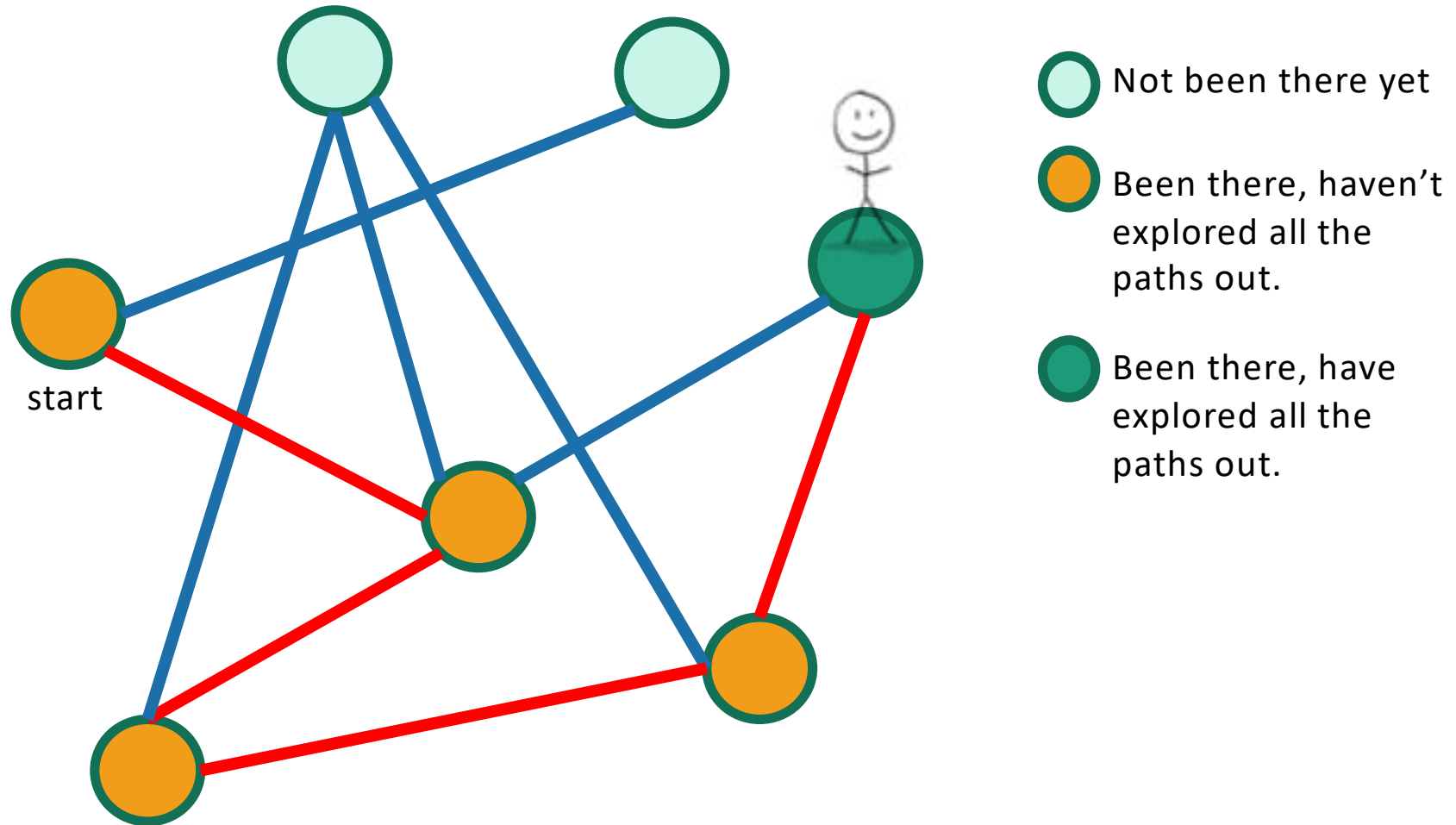
Depth First Search

Exploring a labyrinth with chalk and a piece of string



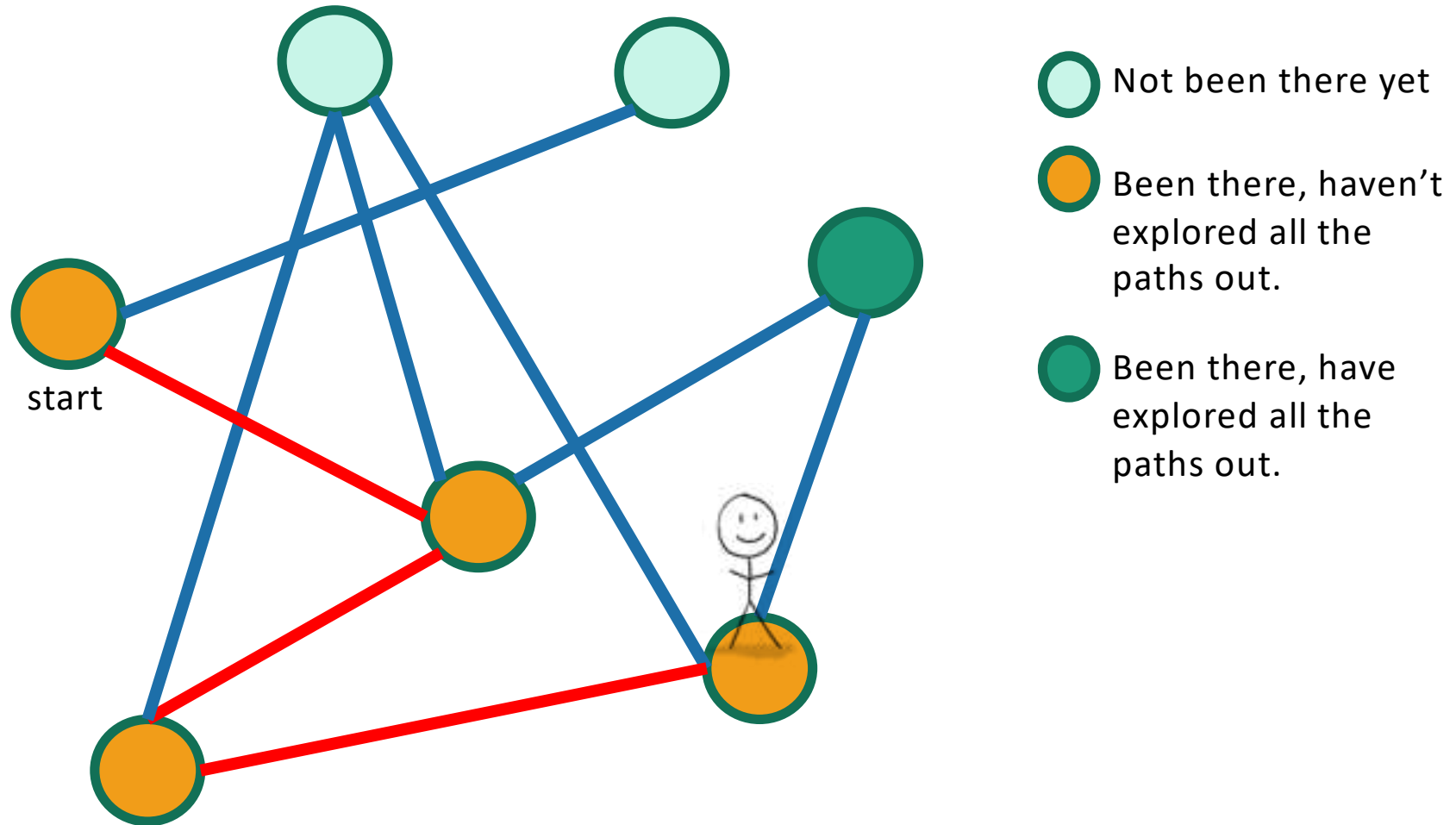
Depth First Search

Exploring a labyrinth with chalk and a piece of string



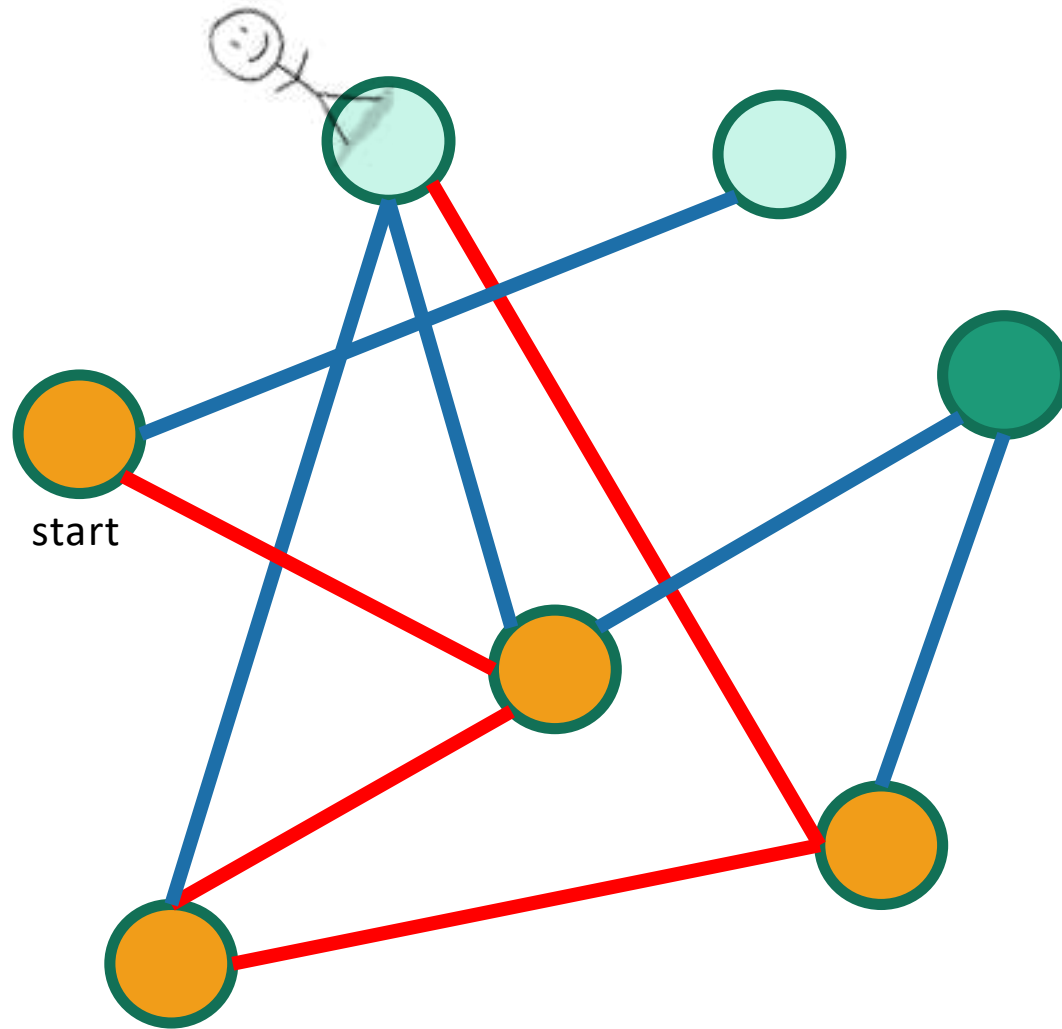
Depth First Search

Exploring a labyrinth with chalk and a piece of string



Depth First Search

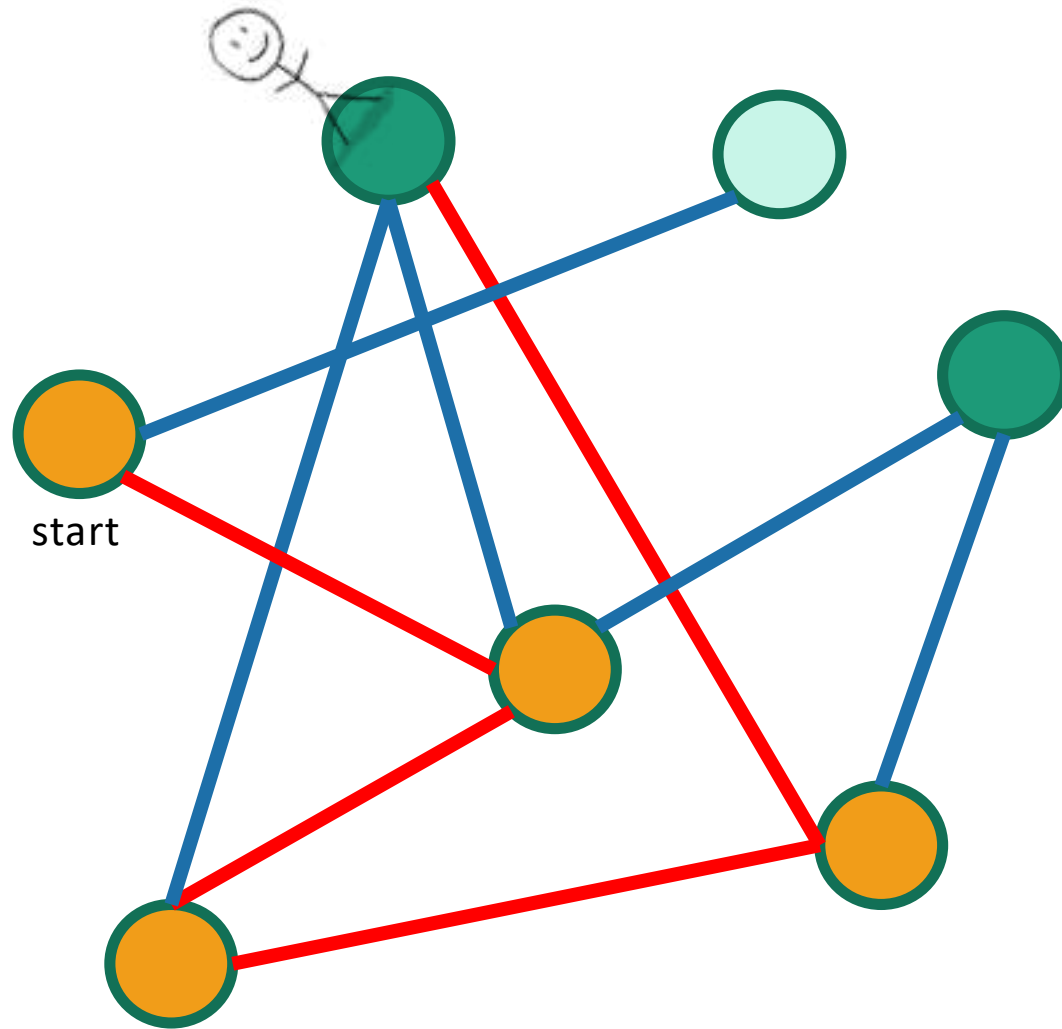
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Depth First Search

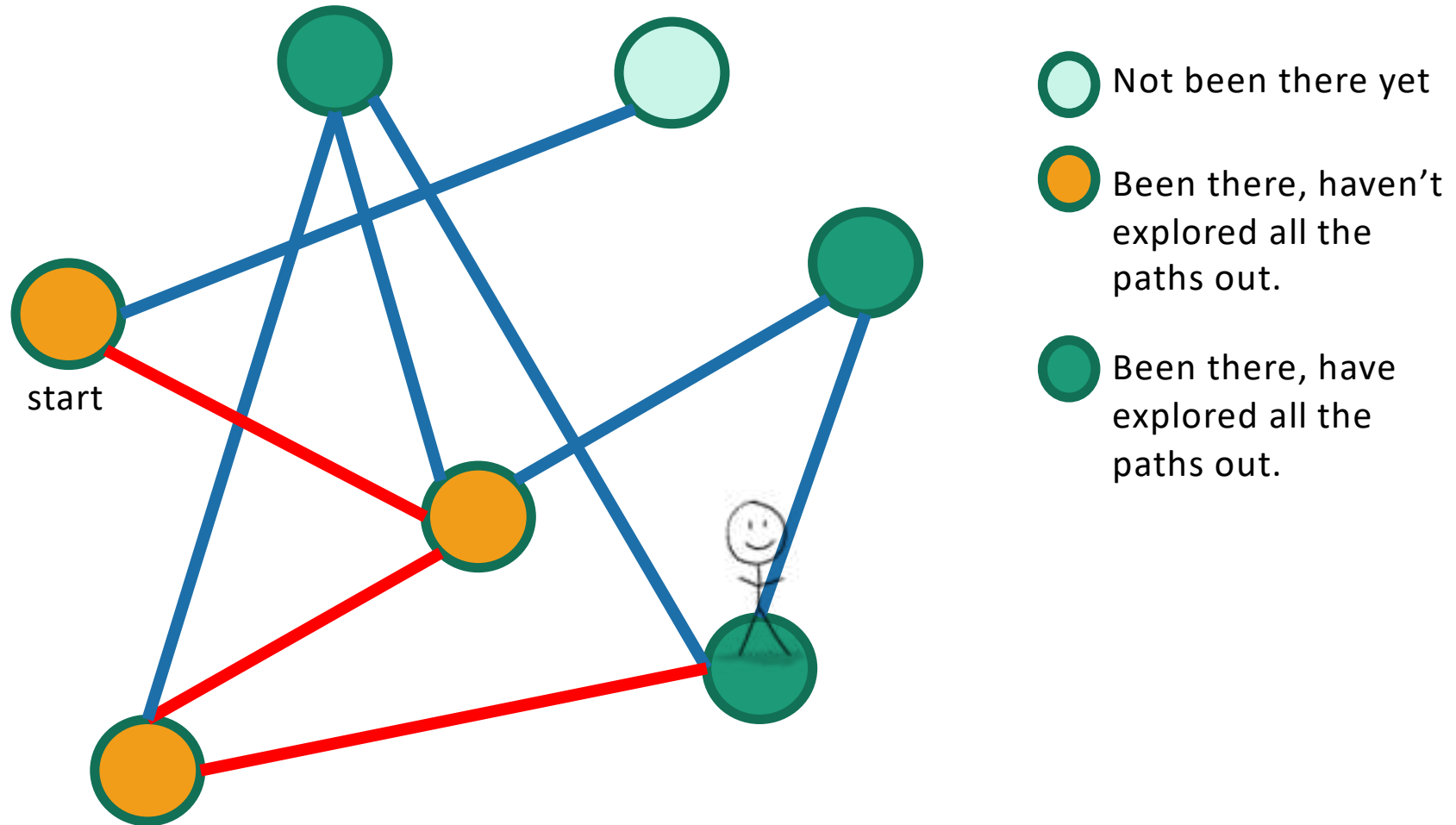
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

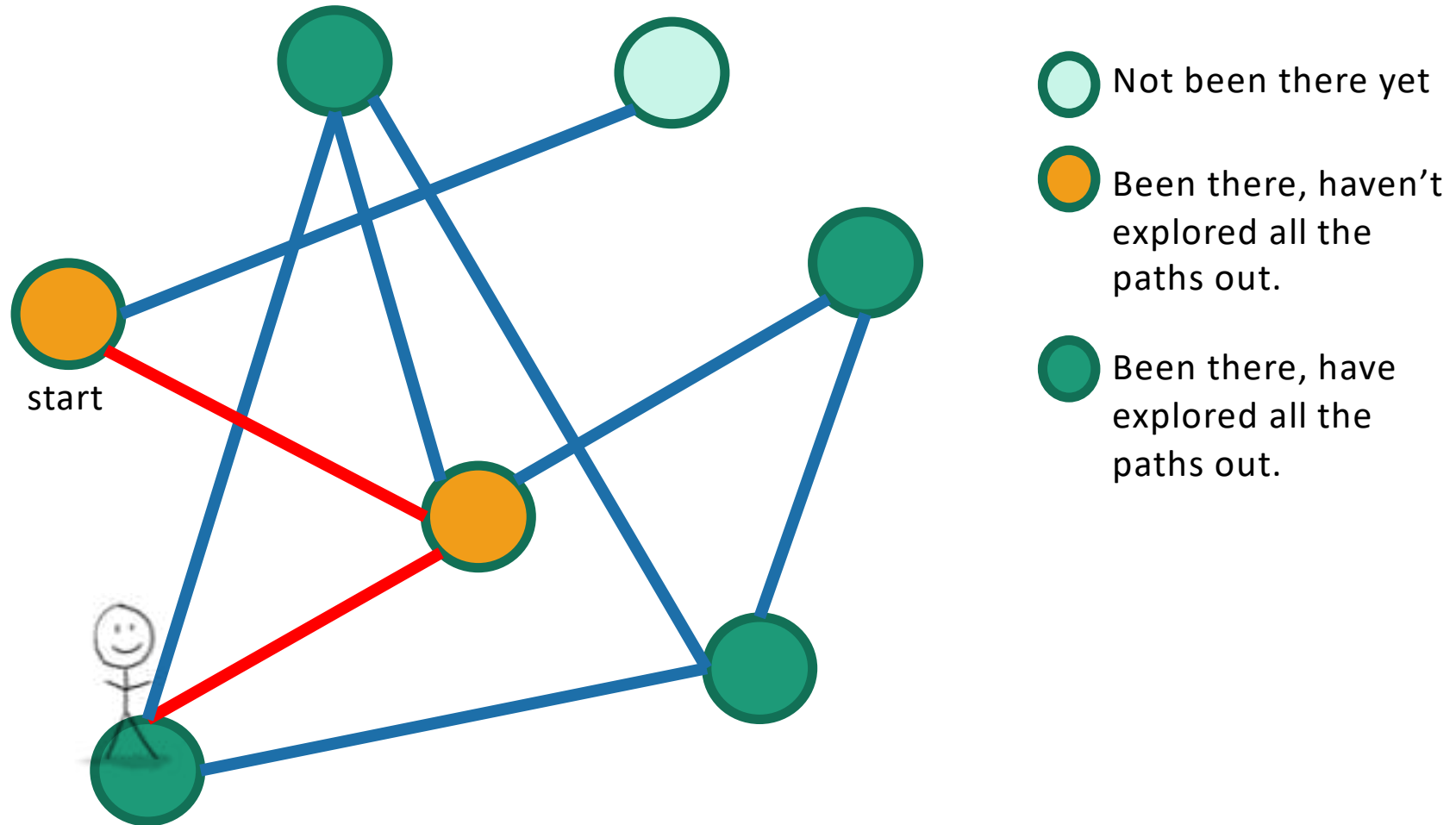
Depth First Search

Exploring a labyrinth with chalk and a piece of string



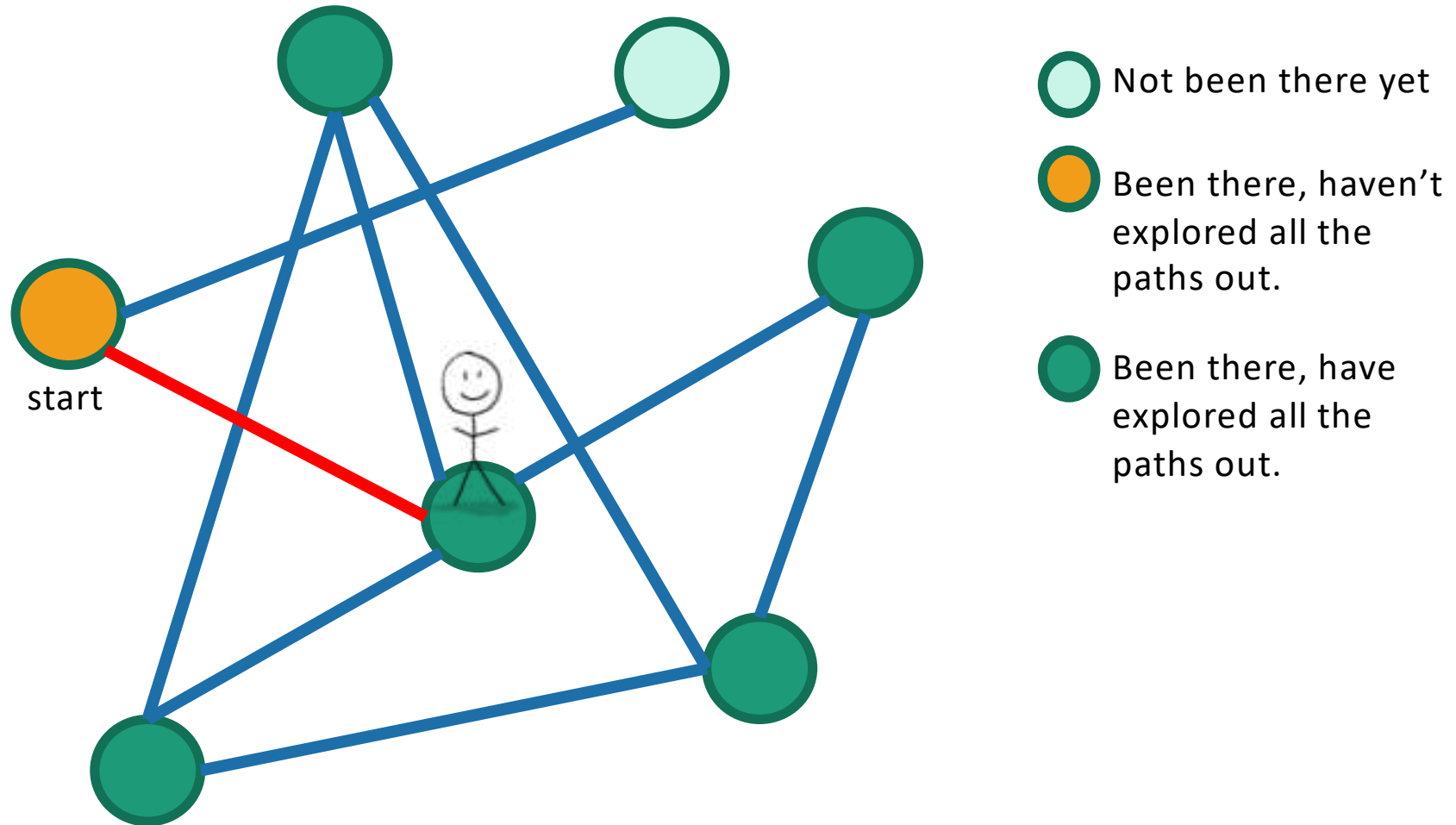
Depth First Search

Exploring a labyrinth with chalk and a piece of string



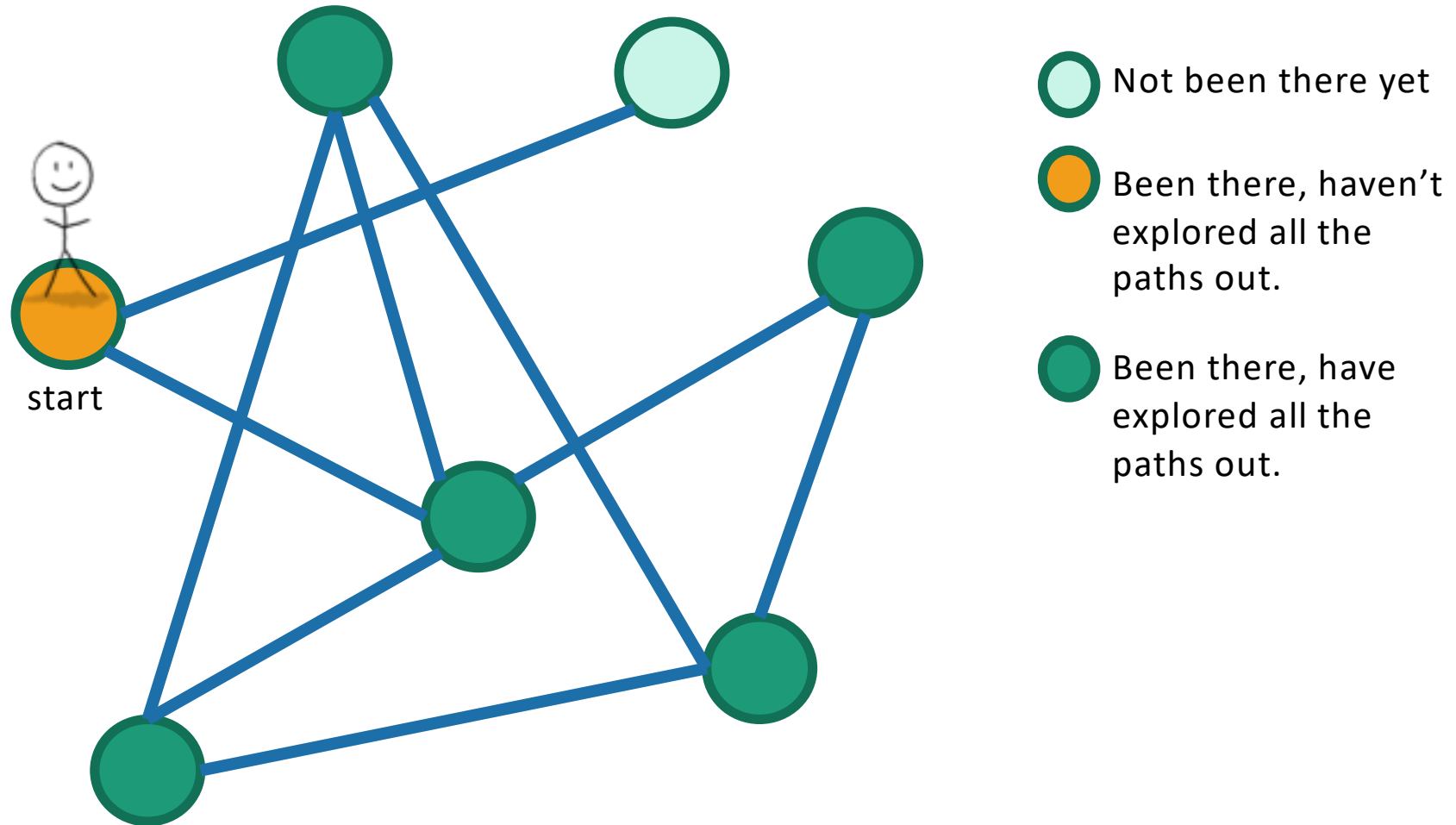
Depth First Search

Exploring a labyrinth with chalk and a piece of string



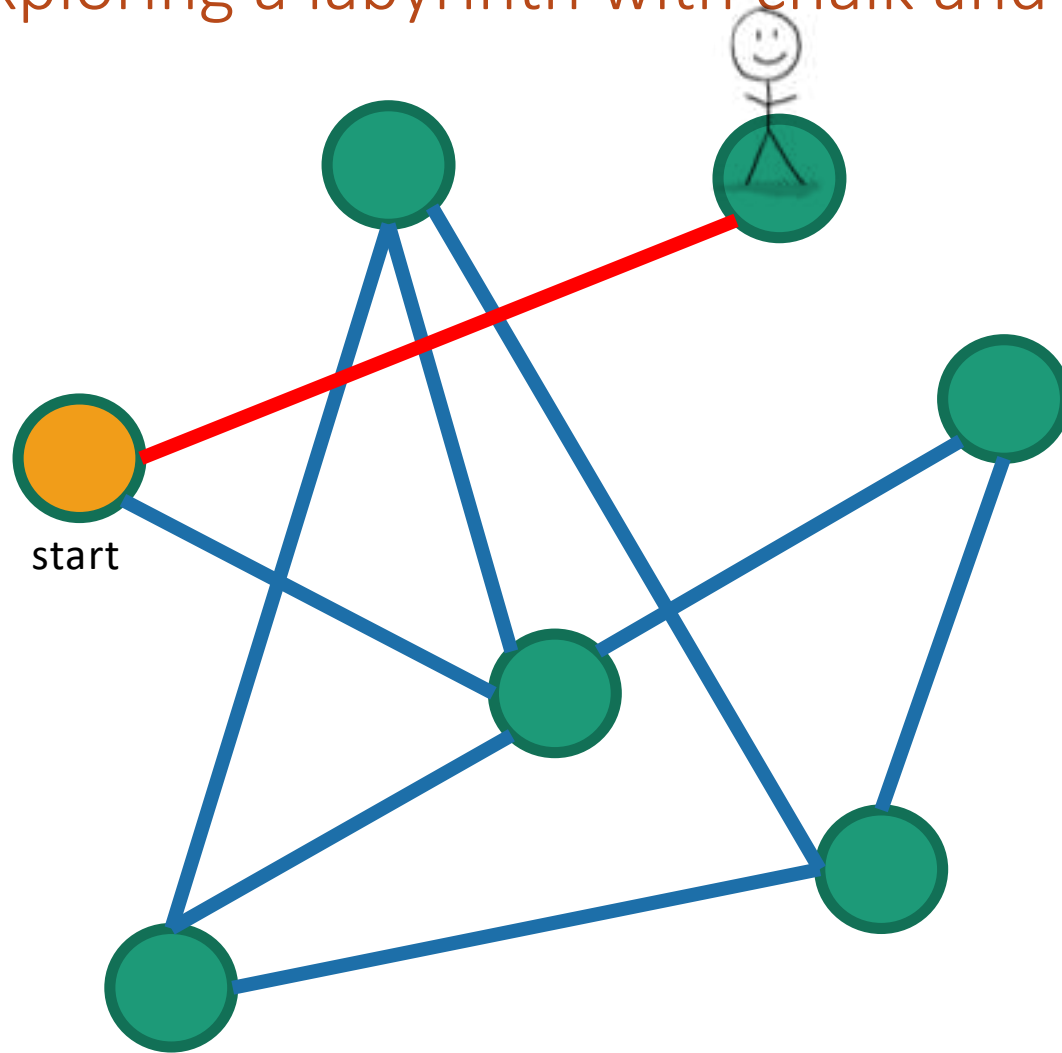
Depth First Search

Exploring a labyrinth with chalk and a piece of string



Depth First Search

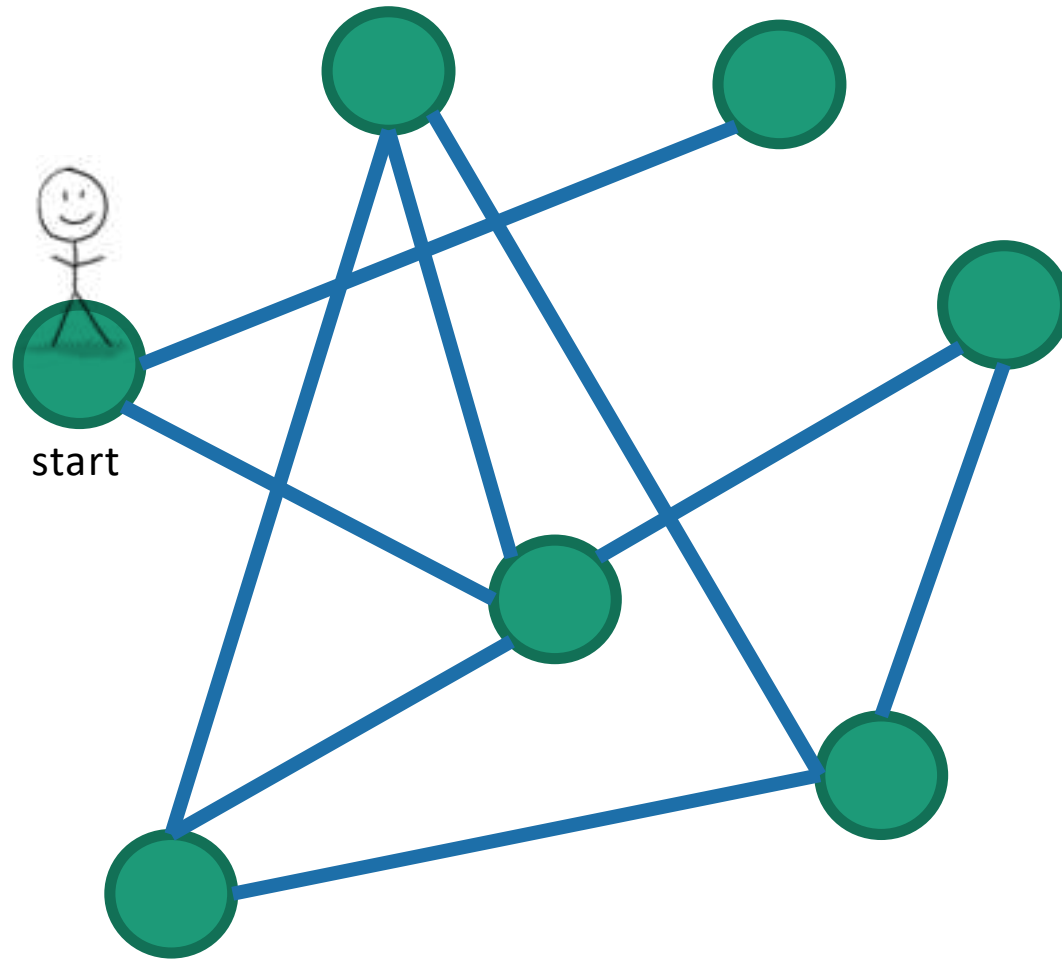
Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Depth First Search

Exploring a labyrinth with chalk and a piece of string



-  Not been there yet
-  Been there, haven't explored all the paths out.
-  Been there, have explored all the paths out.

Labyrinth:
EXPLORED!

Depth First Search

Exploring a labyrinth with pseudocode

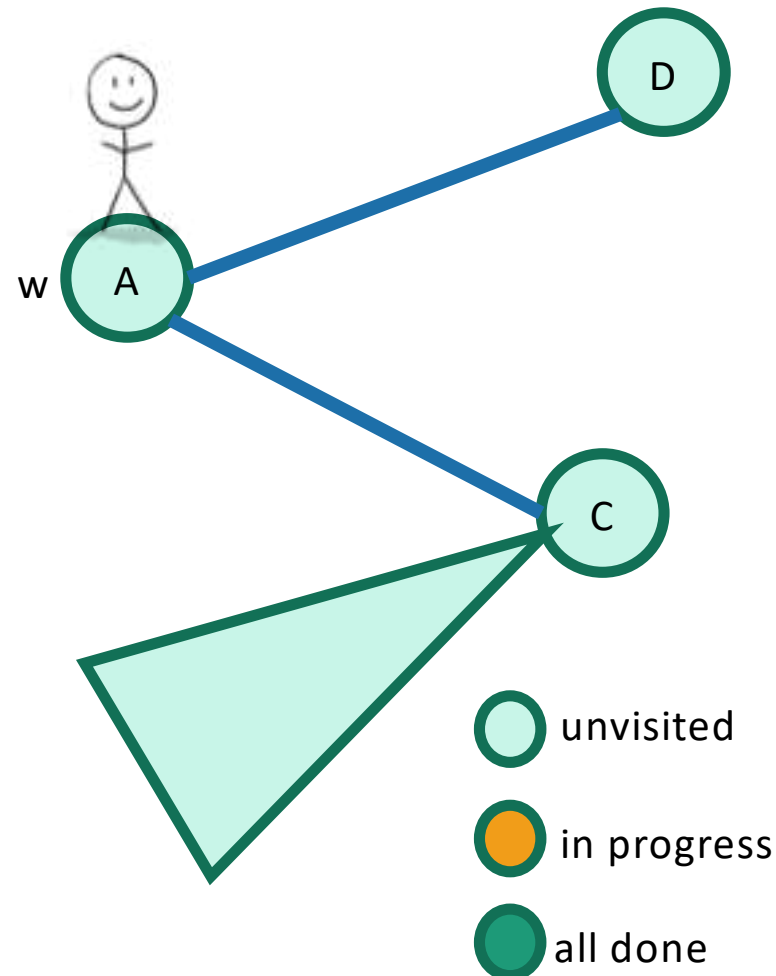
- Each vertex keeps track of whether it is:
 - Unvisited 
 - In progress 
 - All done 
- Each vertex will also keep track of:
 - The time we **first enter it**.
 - The time we finish with it and mark it **all done**.



You might have seen other ways to implement DFS than what we are about to go through. This way has more bookkeeping – the bookkeeping will be useful later!

Depth First Search

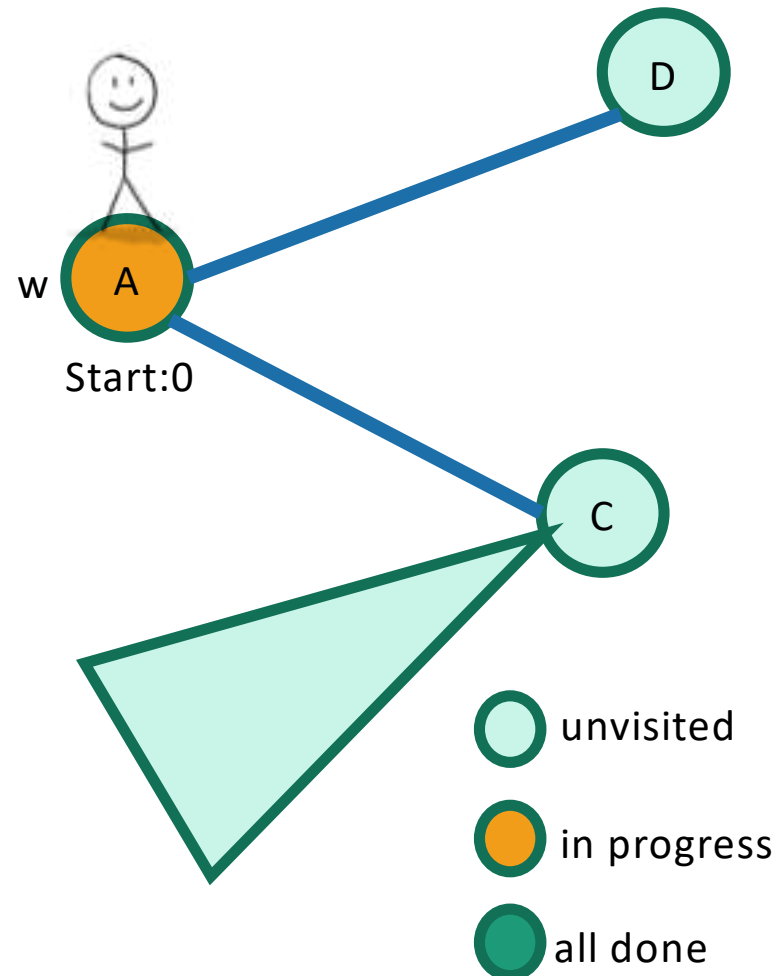
currentTime = 0



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

Depth First Search

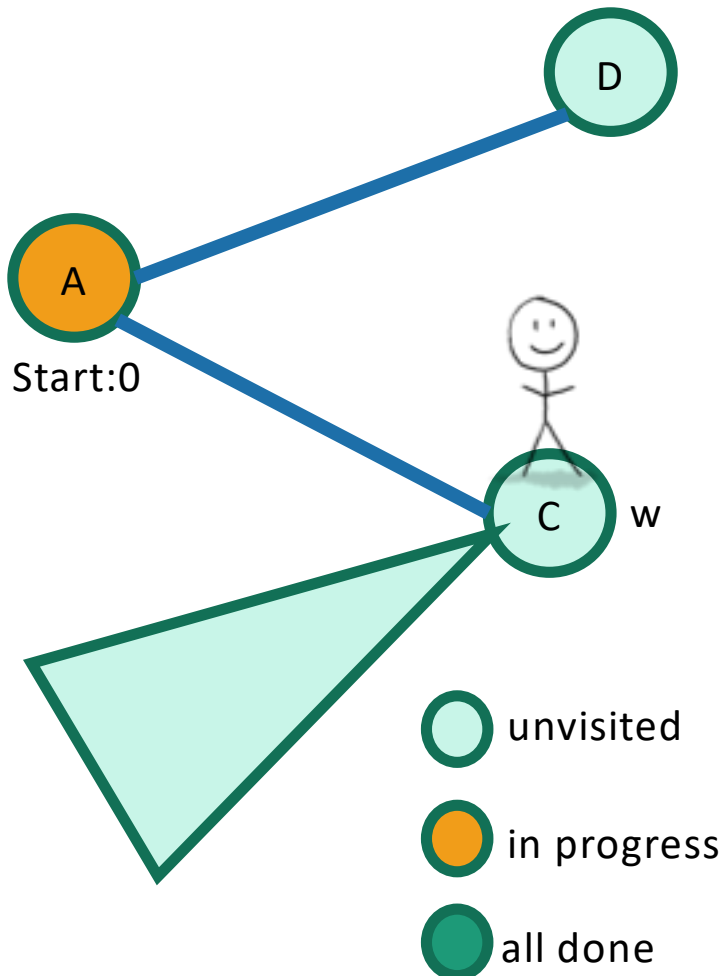
currentTime = 1



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

Depth First Search

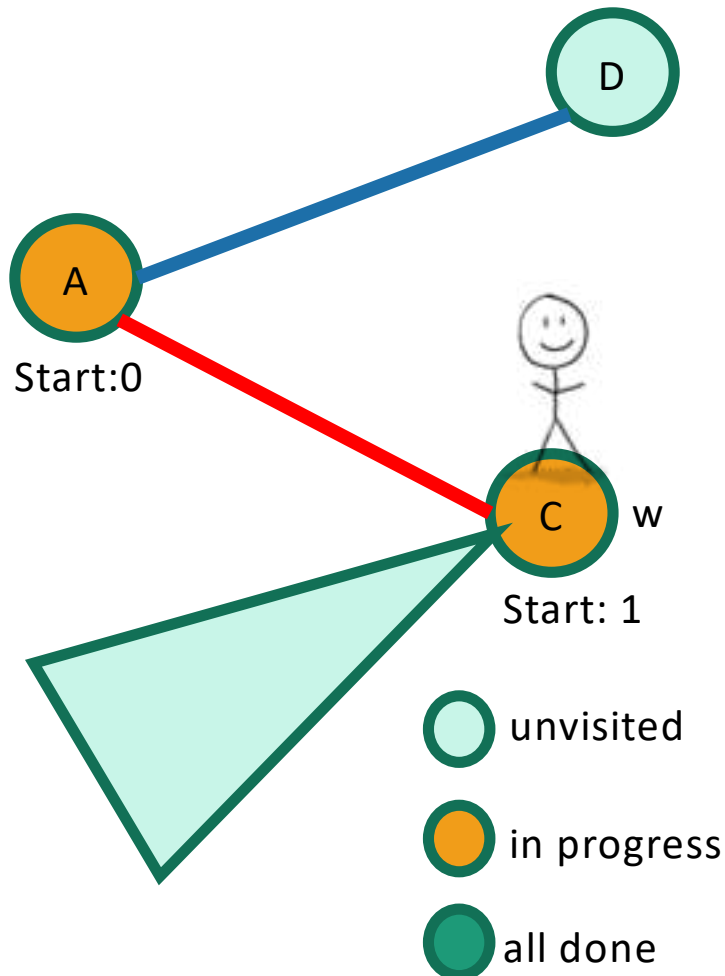
currentTime = 1



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

Depth First Search

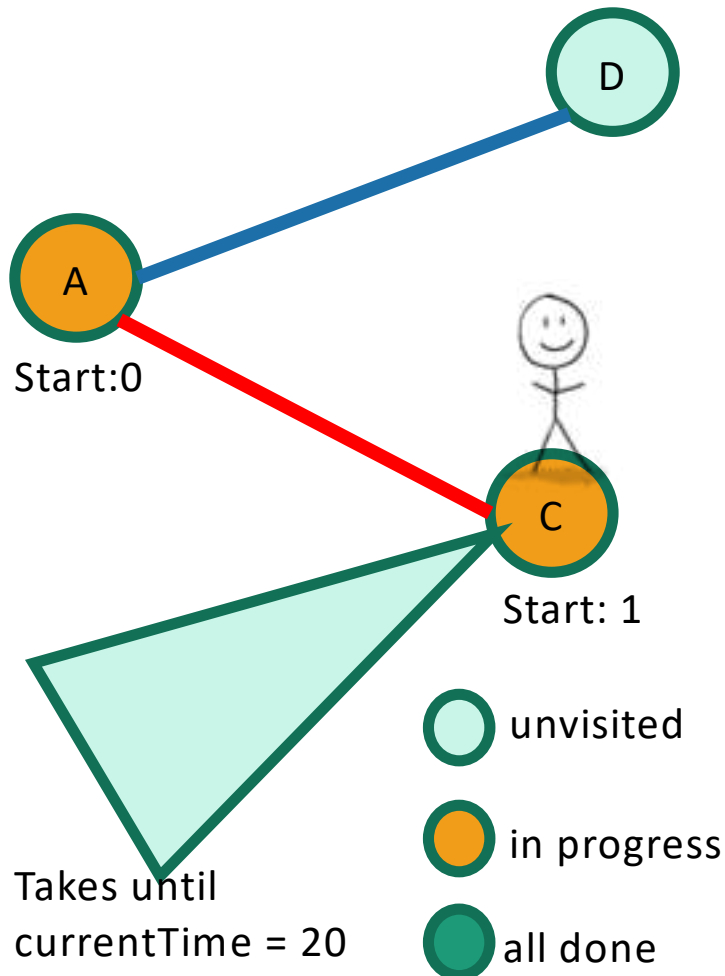
currentTime = 2



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

Depth First Search

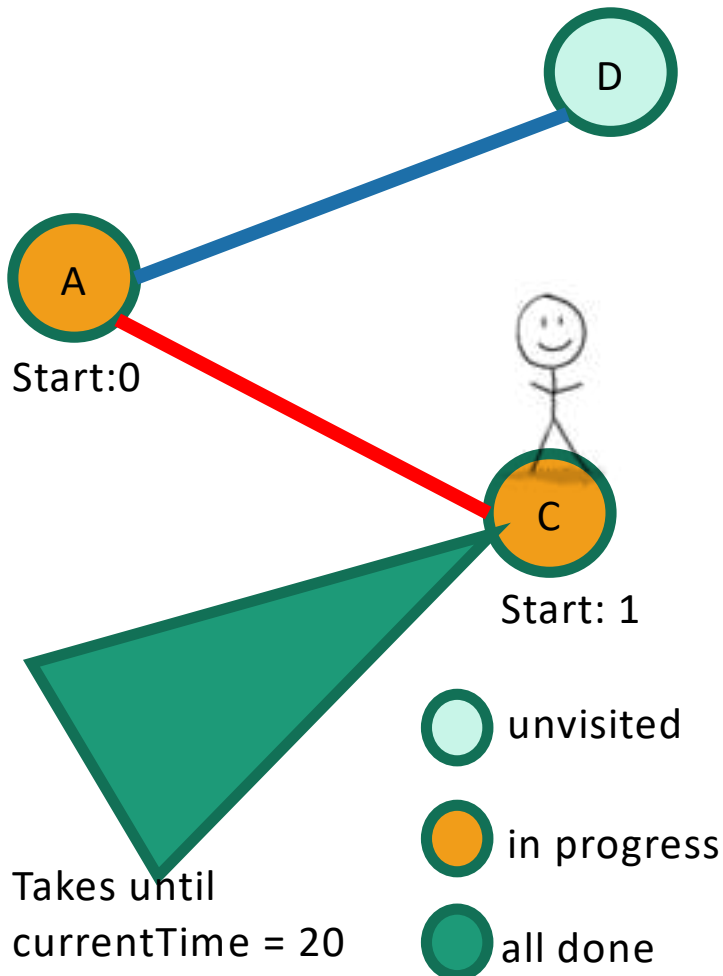
currentTime = 20



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

Depth First Search

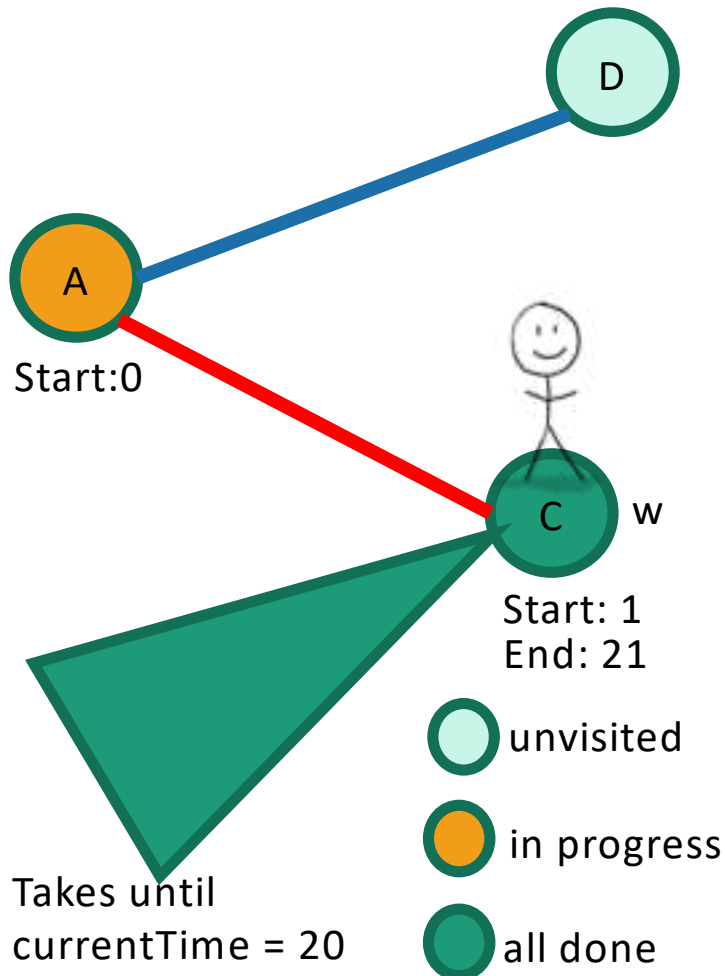
currentTime = 21



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

Depth First Search

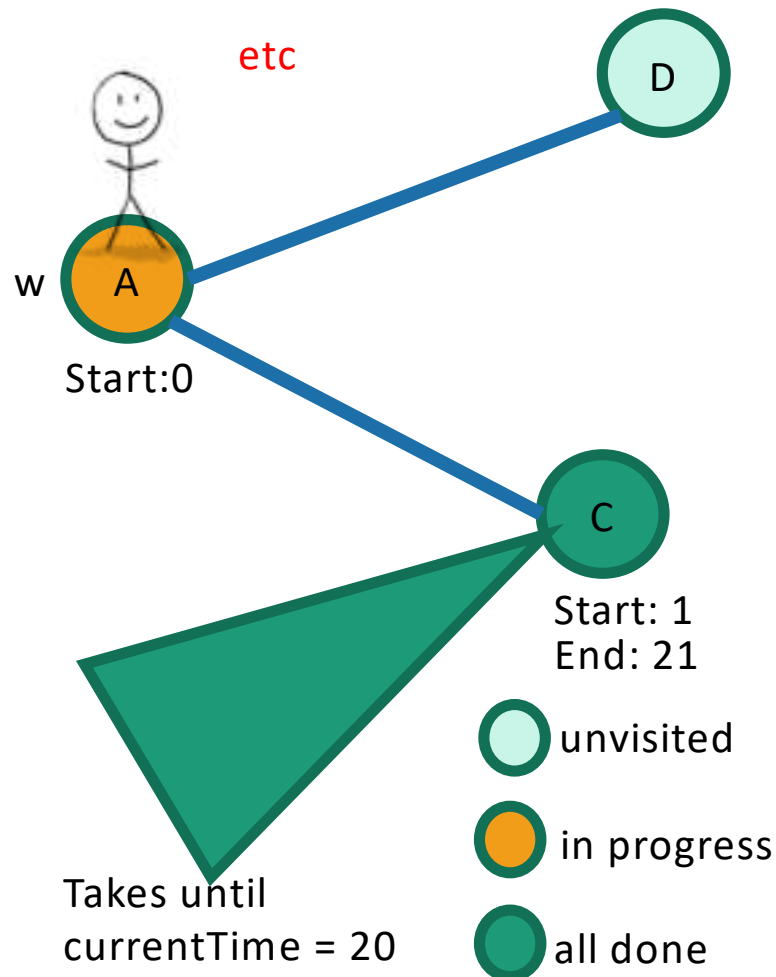
currentTime = 21



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

Depth First Search

currentTime = 22



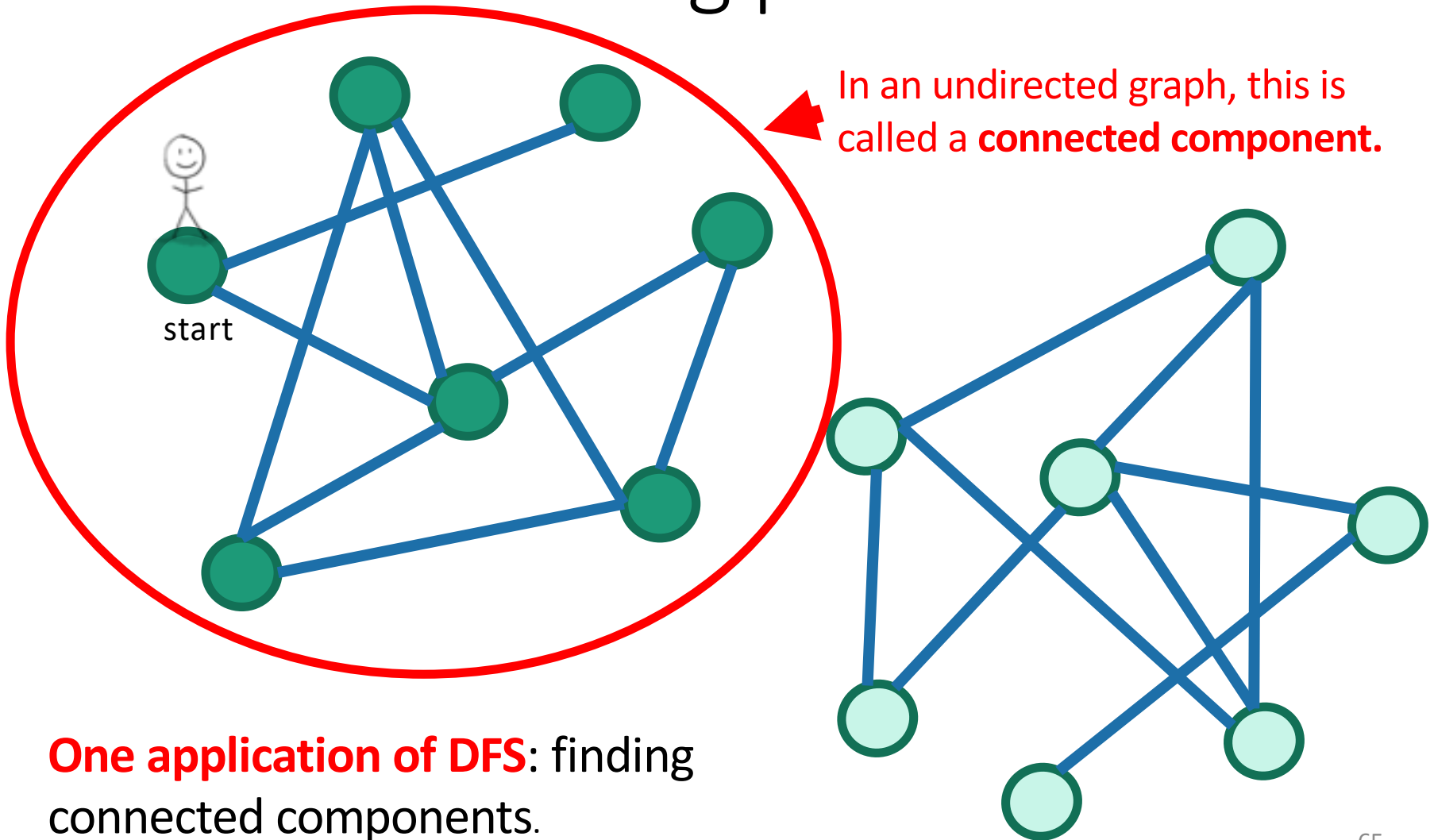
- **DFS**(w, currentTime):
 - w.startTime = currentTime
 - currentTime ++
 - Mark w as **in progress**.
 - **for** v in w.neighbors:
 - **if** v is **unvisited**:
 - currentTime
 - = DFS**(v, currentTime)
 - currentTime ++
 - w.finishTime = currentTime
 - Mark w as **all done**
 - **return** currentTime

Fun exercise

- Write pseudocode for an iterative version of DFS.

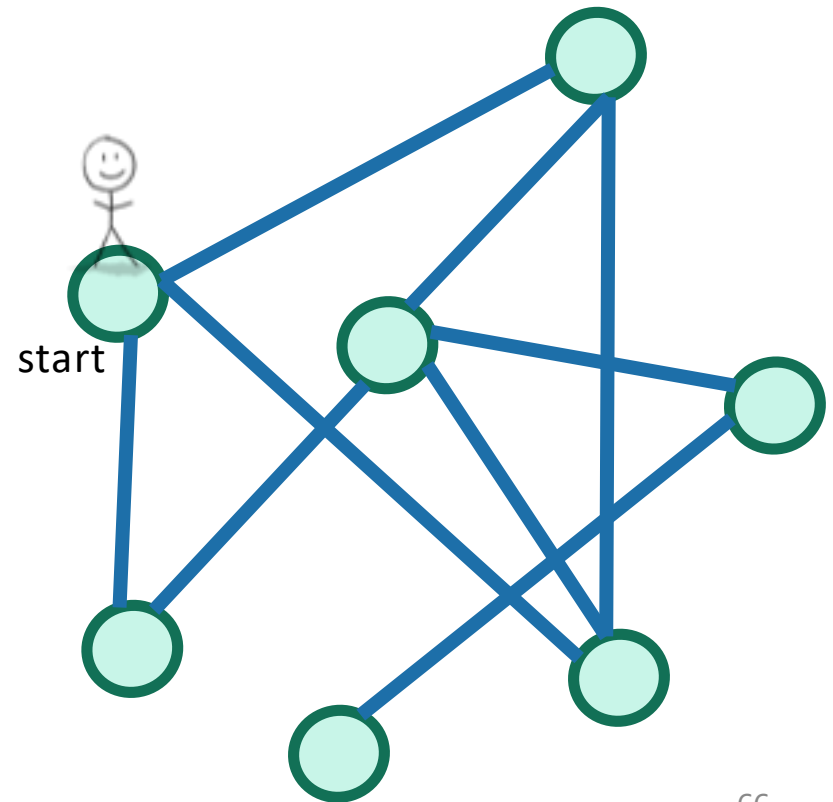
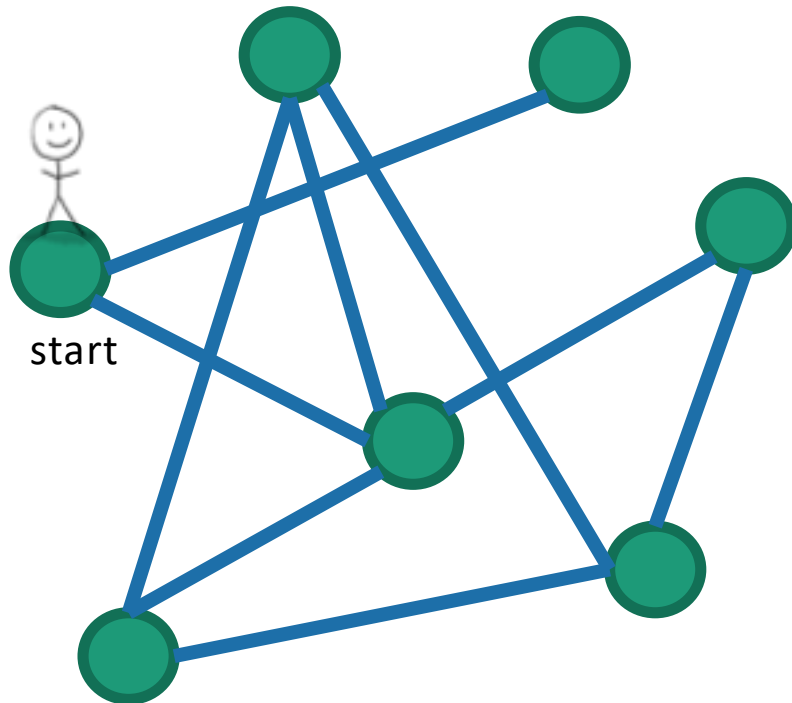


DFS finds all the nodes reachable from the starting point



To explore the whole graph

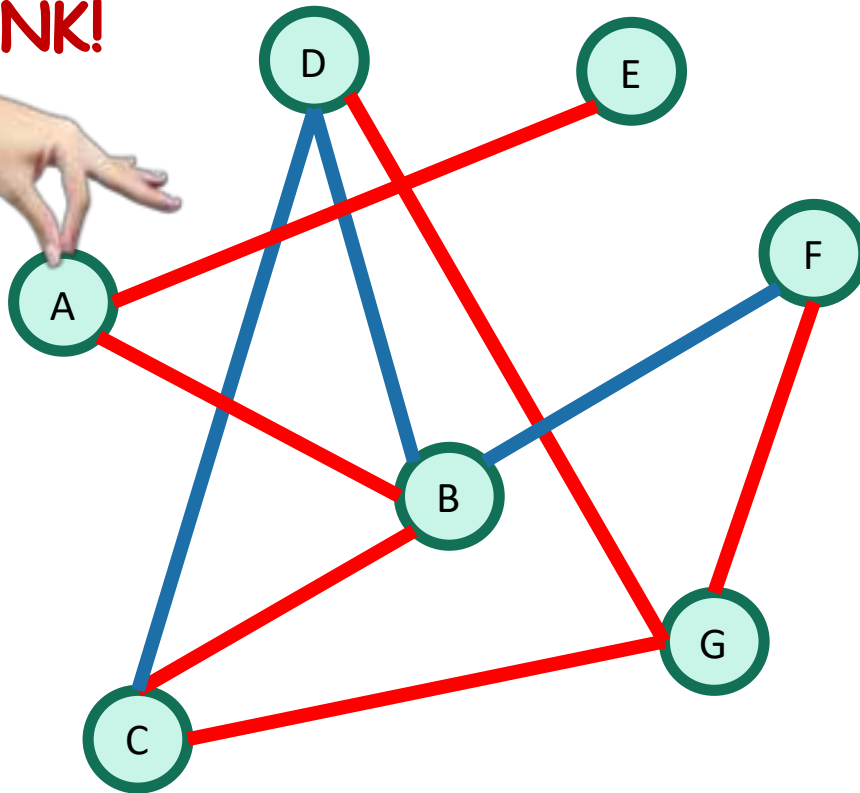
- Do it repeatedly!



Why is it called depth-first?

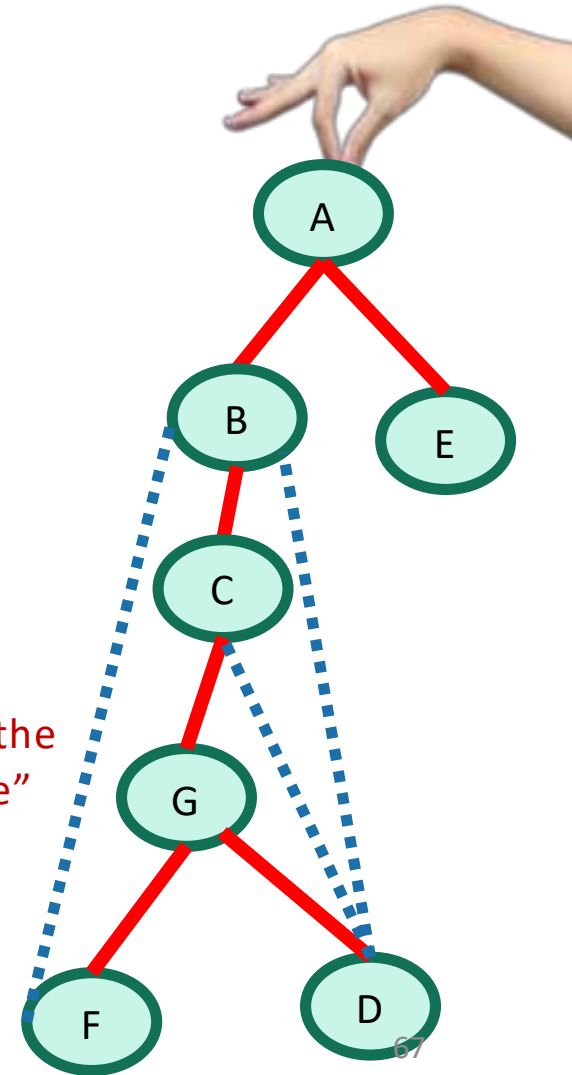
- We are implicitly building a tree:

YOINK!



- First, we go as deep as we can.

Call this the
“DFS tree”



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 just the connected component we started in

- Assume we are using the linked-list format for G .
- Say $C = (V', E')$ is a connected component.
- We visit each vertex in C exactly once.
 - Here, “visit” means “call DFS on”
- At each vertex w , we:
 - Do some book-keeping: $O(1)$
 - Loop over w 's neighbors and check if they are visited (and then potentially make a recursive call): $O(1)$ per neighbor or $O(\deg(w))$ total.
- Total time:
 - $\sum_{w \in V'} (O(\deg(w)) + O(1))$
 - $= O(|E'| + |V'|)$
 - $= O(|E'|)$



In a connected graph,
 $|V'| \leq |E'| + 1$.

Running time

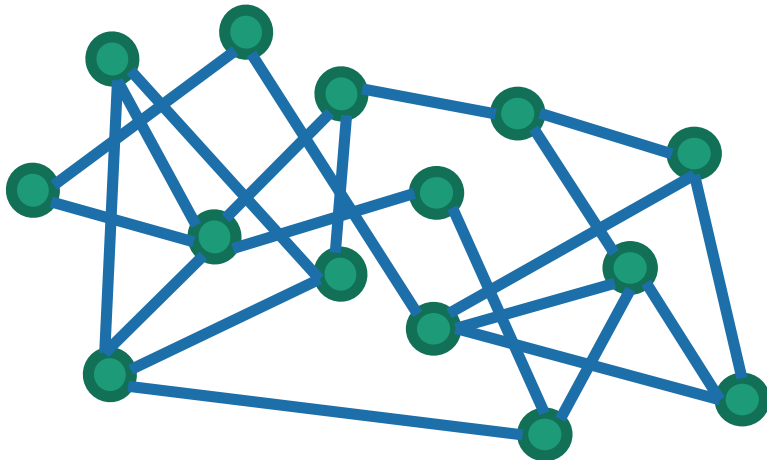
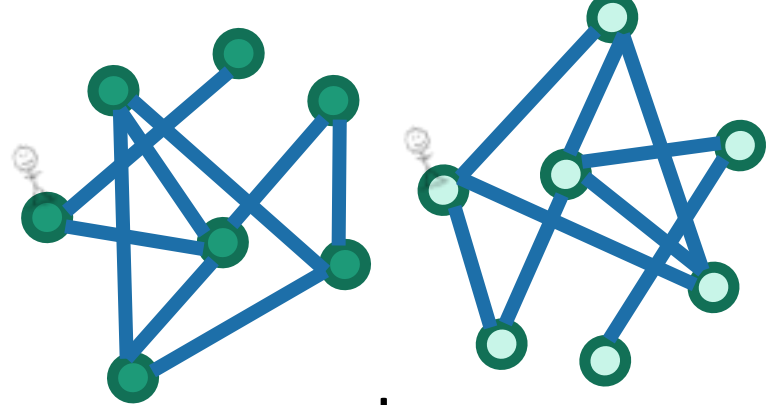
To explore **the whole graph**

- Explore the connected components one-by-one.

- This takes time $O(n + m)$

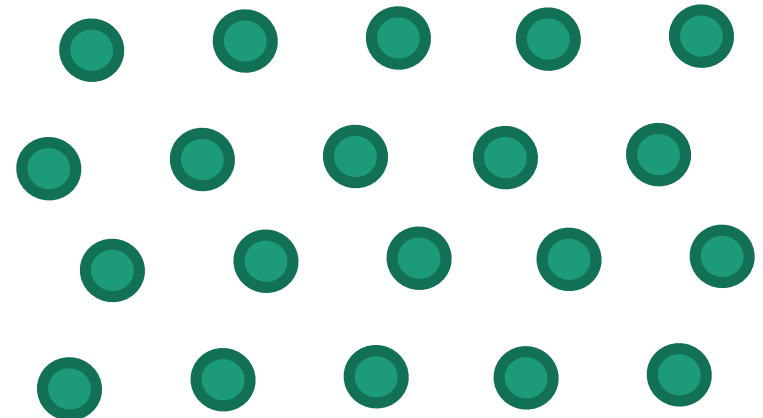
- Same computation as before:

$$\sum_{w \in V} (O(\deg(w)) + O(1)) = O(|E| + |V|) = O(n + m)$$



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

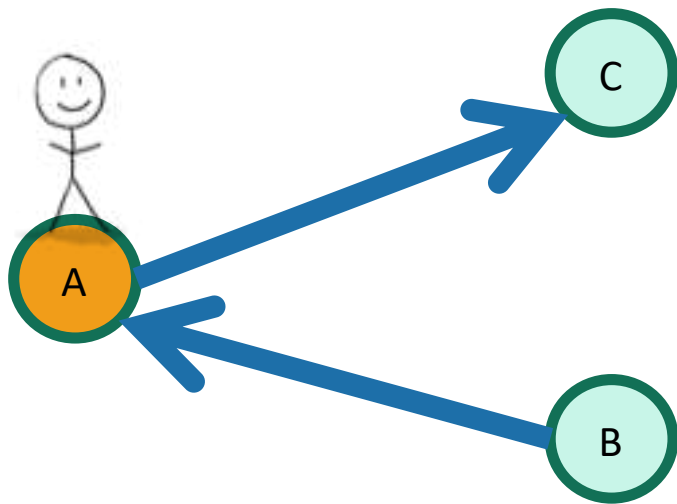
or



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

You check:

DFS works fine on directed graphs too!



Only walk to C, not to B.



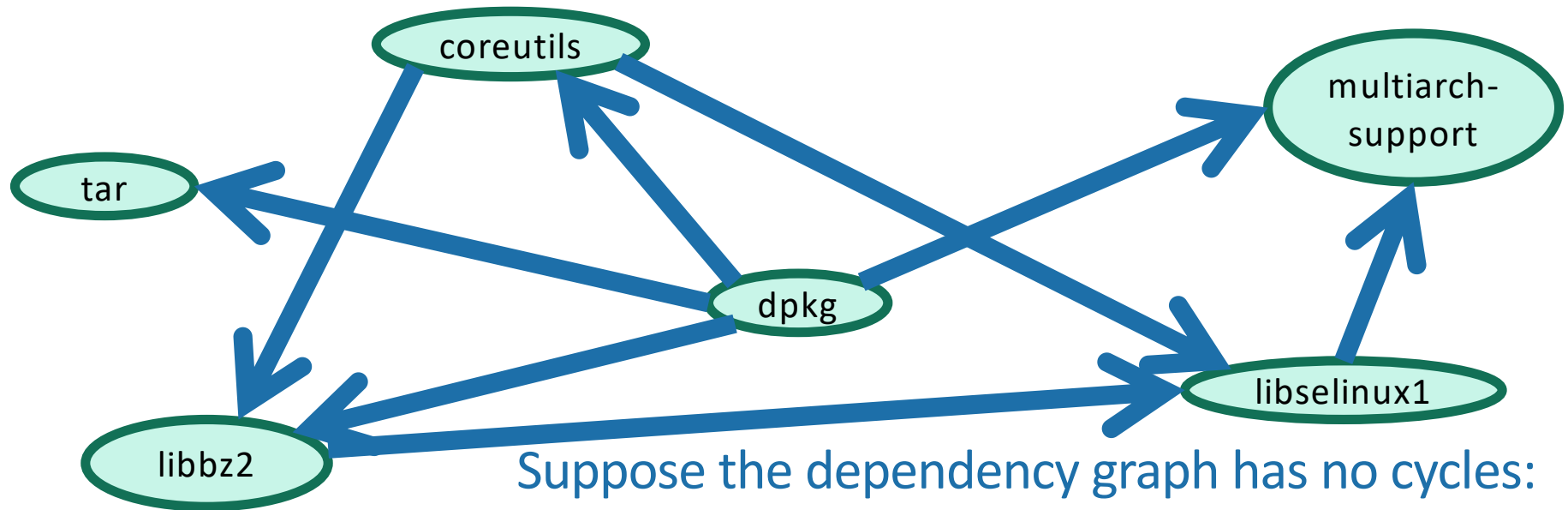
Siggi the studios stork

Pre-lecture exercise

- How can you sign up for classes so that you never violate the pre-req requirements?
- More practically, how can you install packages without violating dependency requirements?

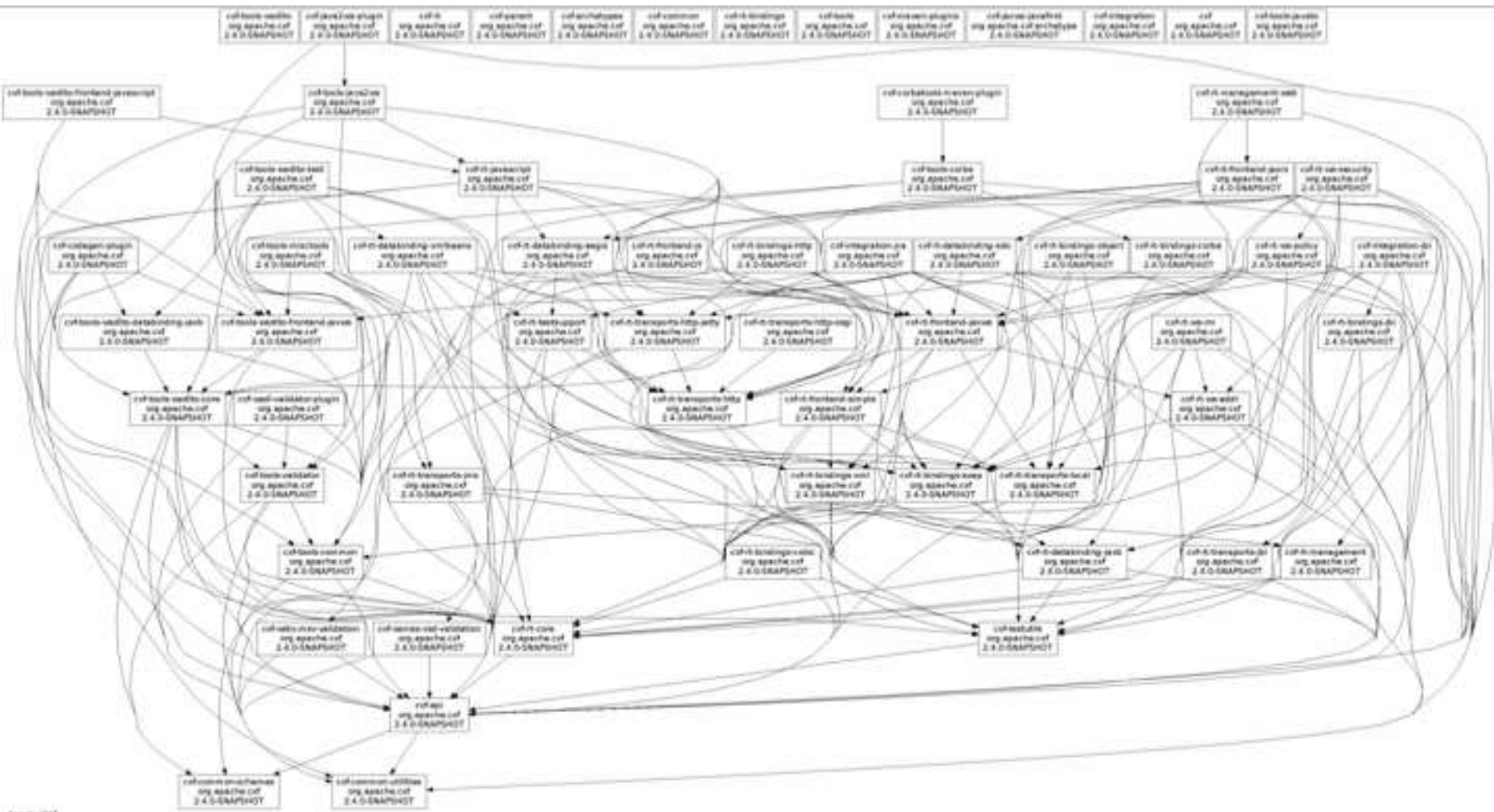
Application of DFS: topological sorting

- Find an ordering of vertices so that all of the dependency requirements are met.
 - Aka, if v comes before w in the ordering, there is not an edge from w to v .



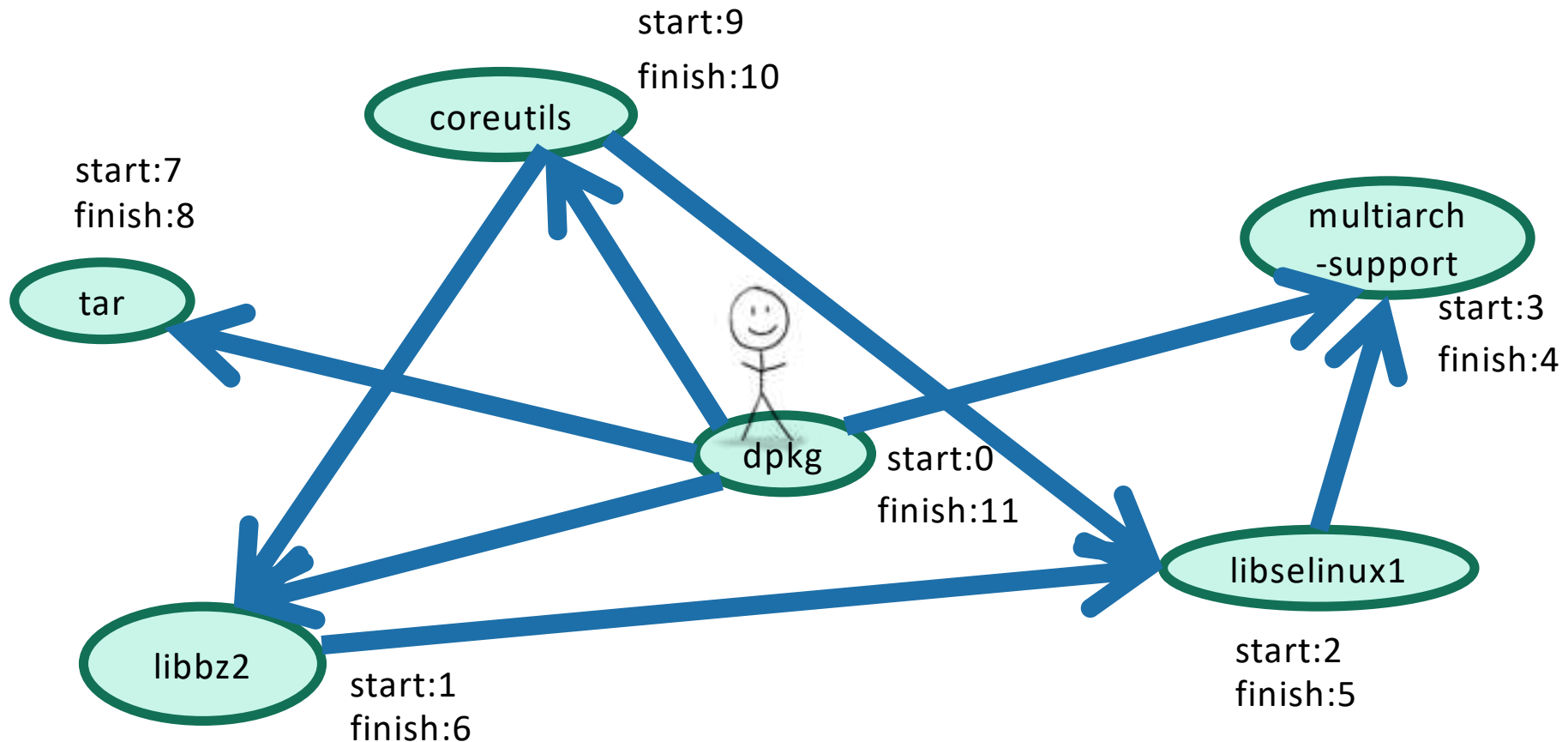
Suppose the dependency graph has no cycles:
it is a **Directed Acyclic Graph (DAG)**

Can't always eyeball it.



Let's do DFS

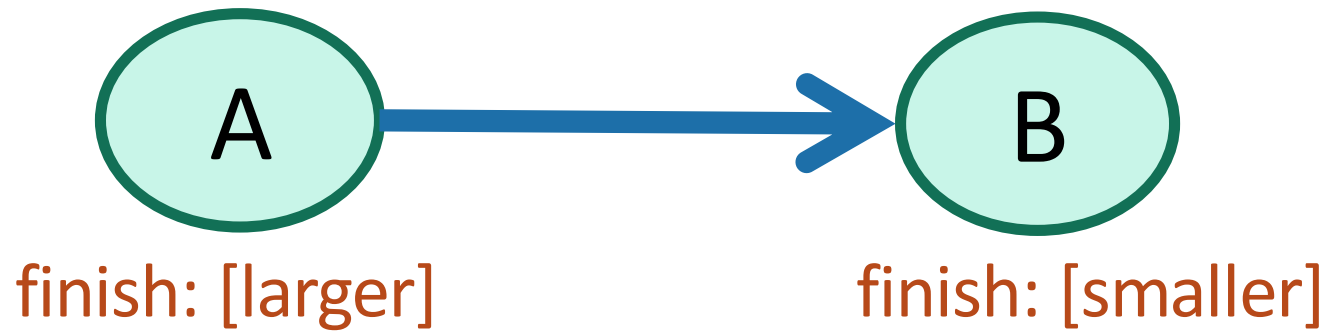
What do you notice about the finish times? Any ideas for how we should do topological sort?



Suppose the underlying
graph has no cycles

Finish times seem useful

Claim: In general, we'll always have:



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

A more general statement

(this holds even if there are cycles)

This is called the “parentheses theorem” in CLRS

(check this statement carefully!)



- If v is a descendant of w in this tree:



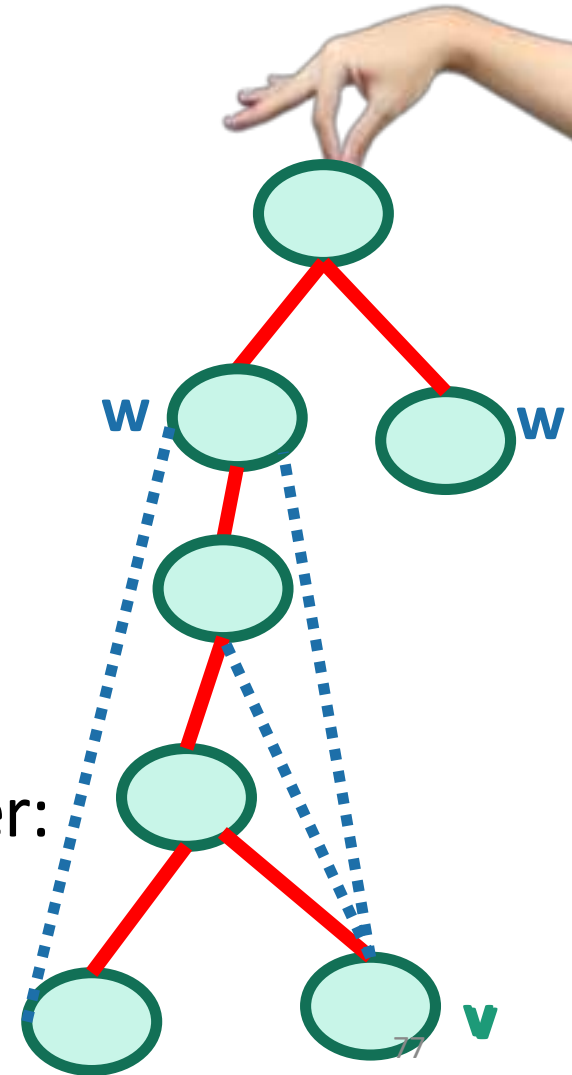
- If w is a descendant of v in this tree:



- If neither are descendants of each other:

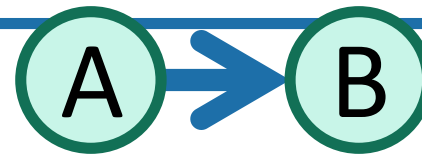


(or the other way around)



So to prove this →

If



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

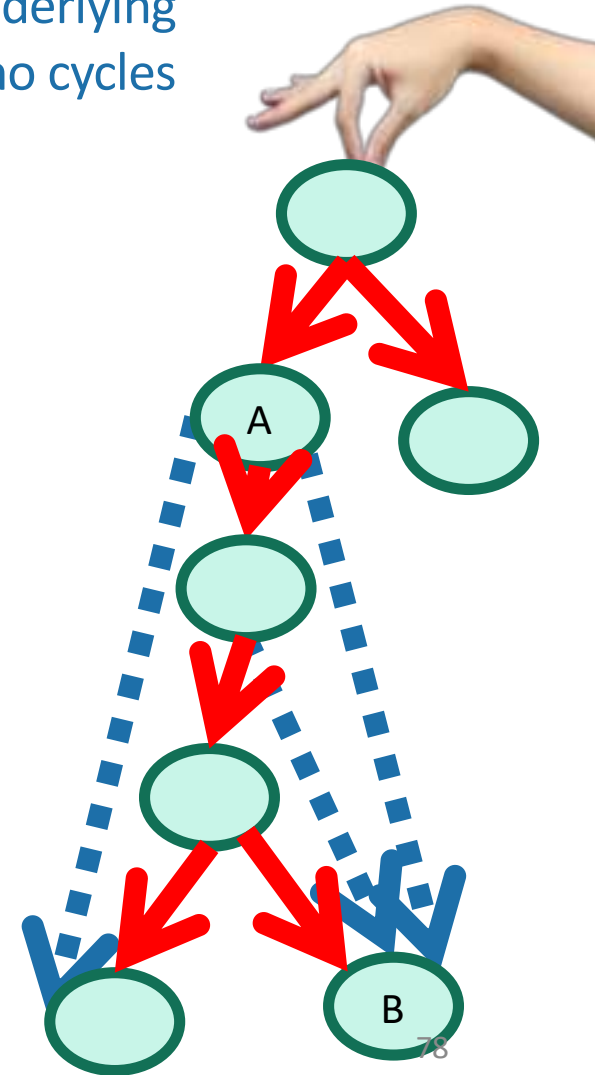
Suppose the underlying
graph has no cycles

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

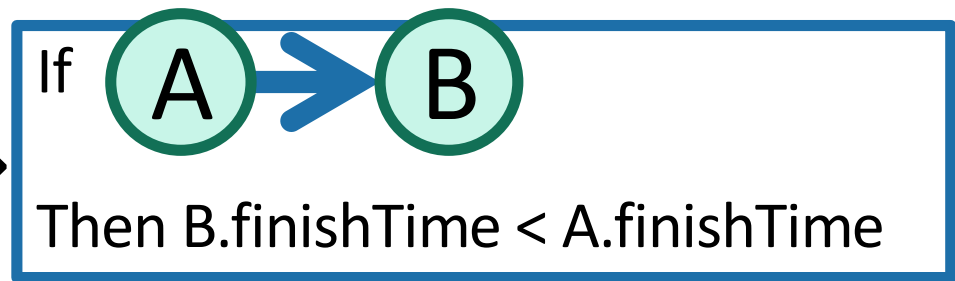
- Then



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



So to prove this →



Suppose the underlying graph has no cycles

- **Case 2:** B is a **NOT** descendant of A in the DFS tree.

- Notice that A can't be a descendant of B or else there'd be a cycle; so it looks like this →

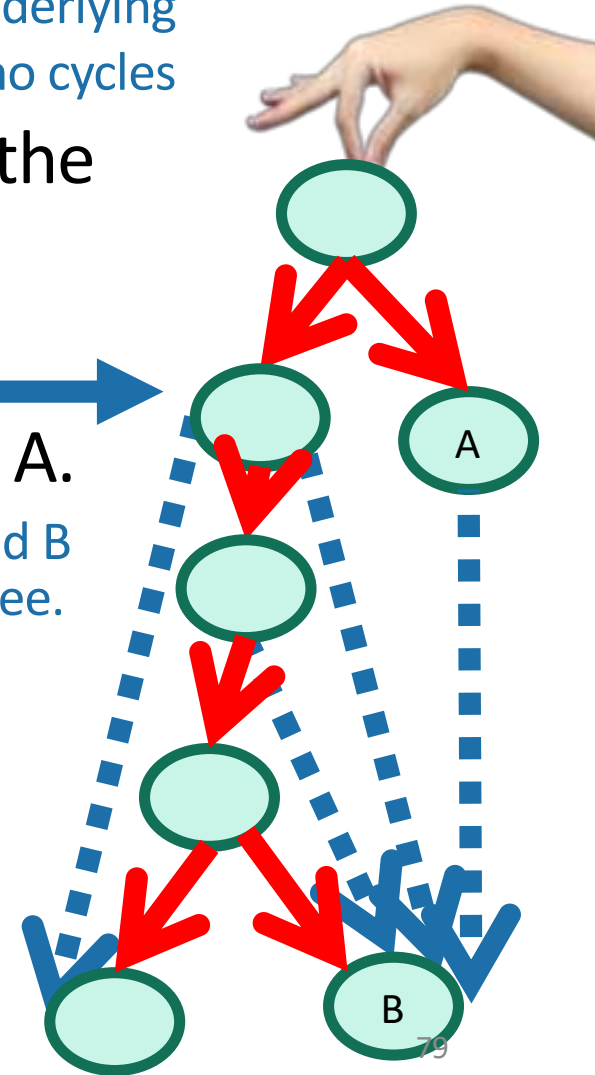
- Then we must have explored B before A.

- Otherwise we would have gotten to B from A, and B would have been a descendant of A in the DFS tree.

- Then

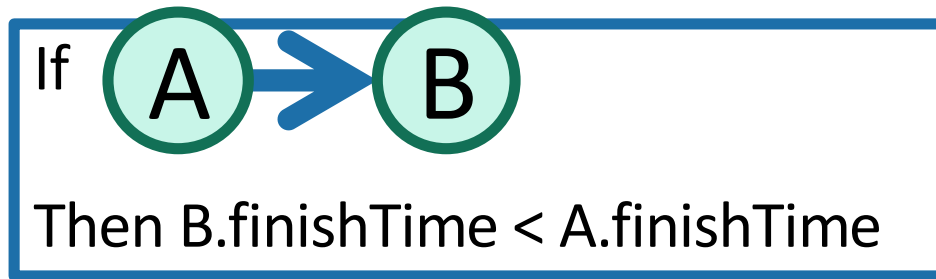


- aka, $B.\text{finishTime} < A.\text{finishTime}$.

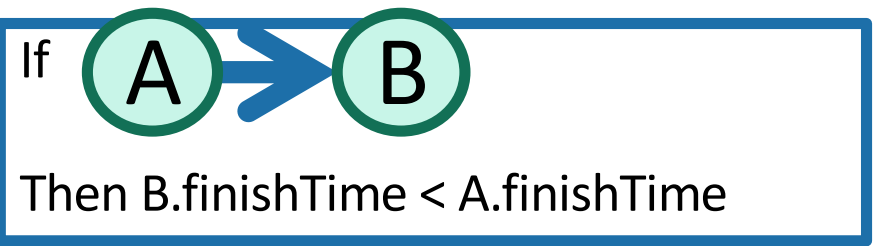


Theorem

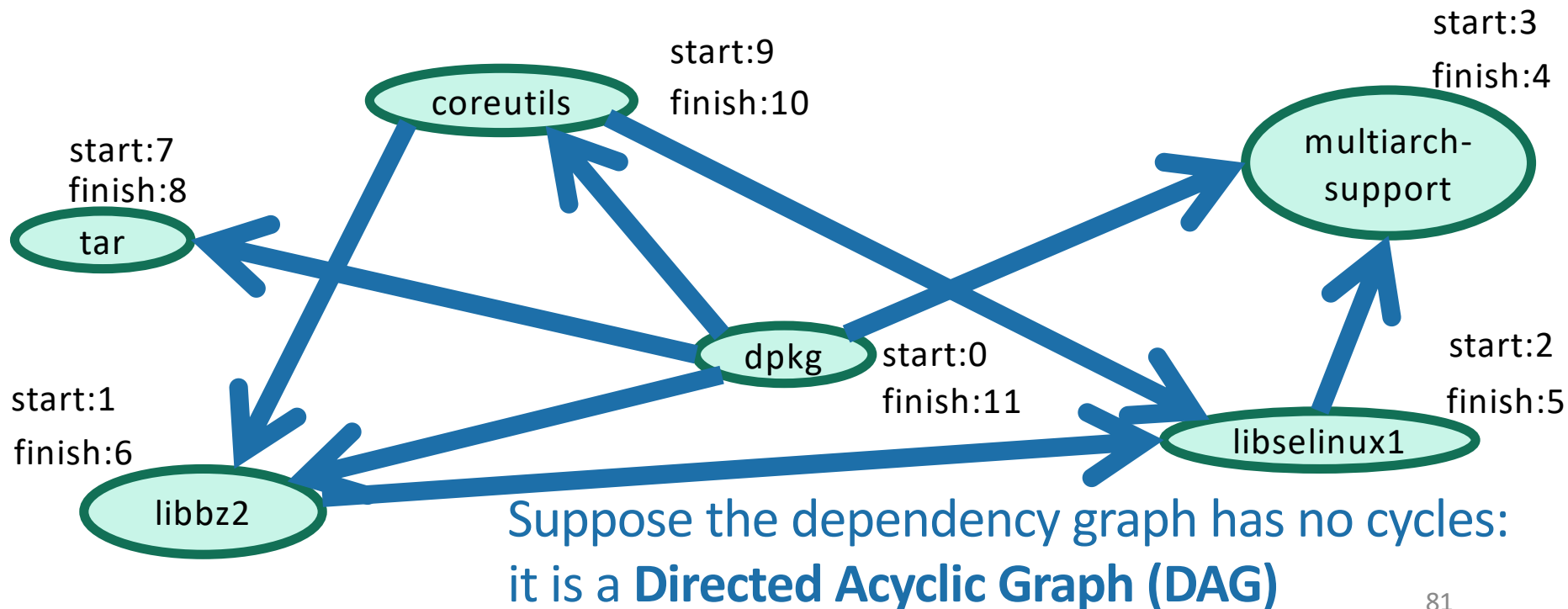
- If we run DFS on a directed acyclic graph,



Back to topological sorting



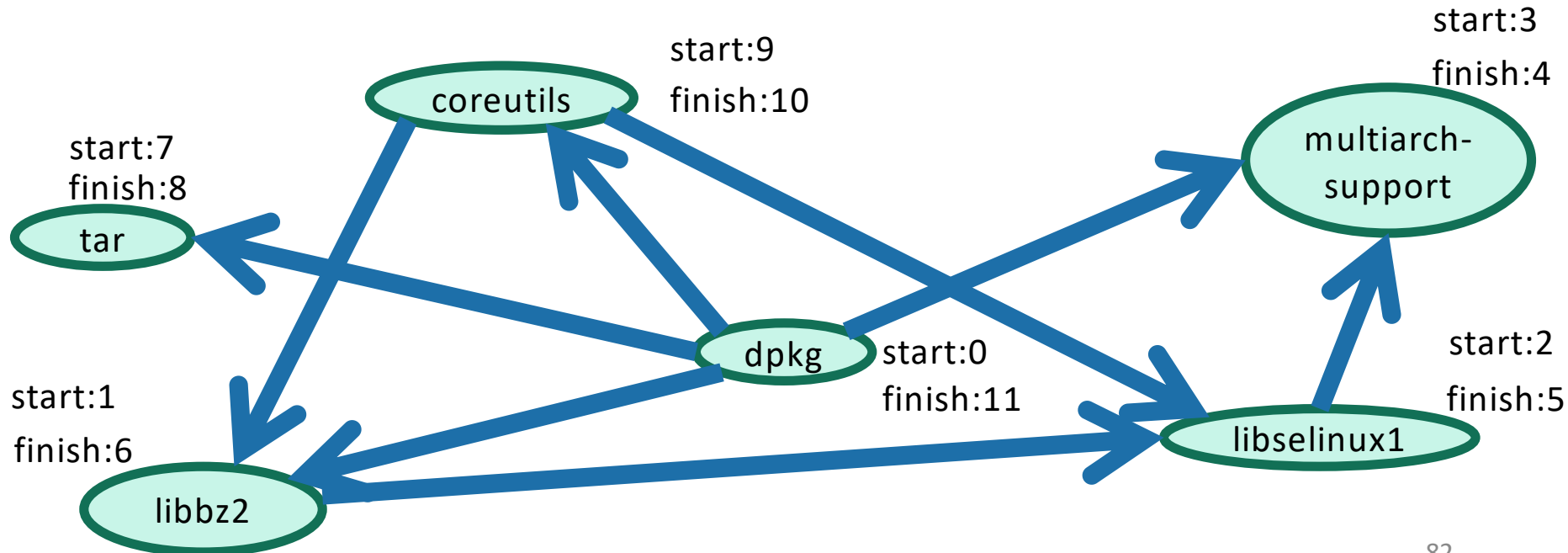
- In what order should I install packages?
- In reverse order of finishing time in DFS!



Topological Sorting (on a DAG)

- 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



For implementation, see IPython notebook

```
In [69]: print(G)
```

```
CS161Graph with:
  Vertices:
    dkpg,coreutils,multiarch_support,libselinux1,libbz2,tar,
  Edges:
    (dkpg,multiarch_support) (dkpg,coreutils) (dkpg,tar) (dkpg,libbz2
) (coreutils,libbz2) (coreutils,libselinux1) (libselinux1,multiarch_suppo
rt) (libbz2,libselinux1)
```

```
In [71]: V = topoSort(G)
for v in V:
    print(v)
```

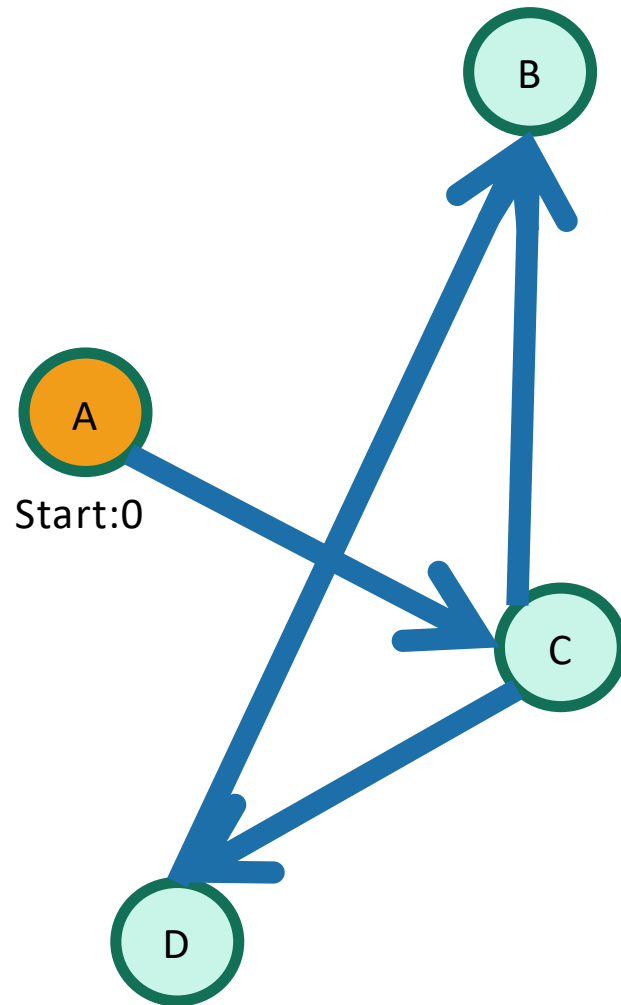
```
dkpg
tar
coreutils
libbz2
libselinux1
multiarch_support
```

What did we just learn?

- DFS can help you solve the **topological sorting problem**
 - That's the fancy name for the problem of finding an ordering that respects all the dependencies
- Thinking about the DFS tree is helpful.

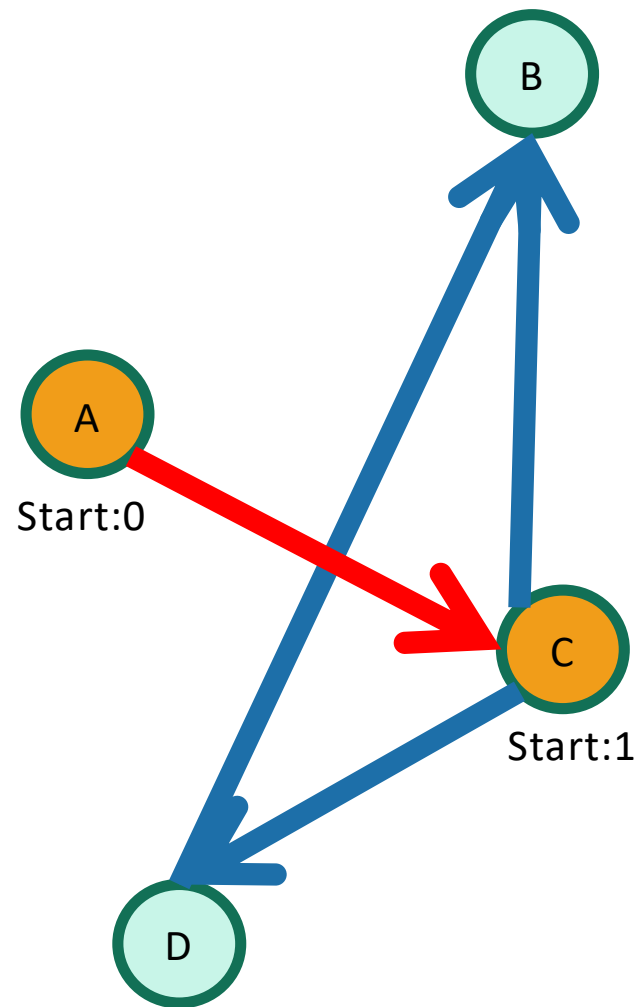
Example:

This example skipped in class – here for reference.



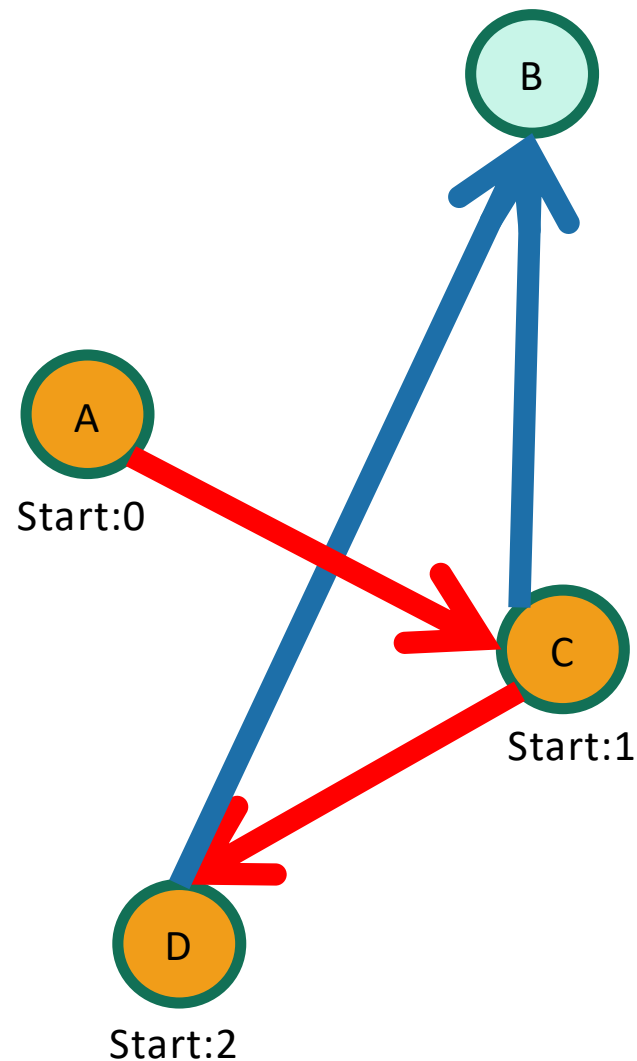
Example

This example skipped in class – here for reference.



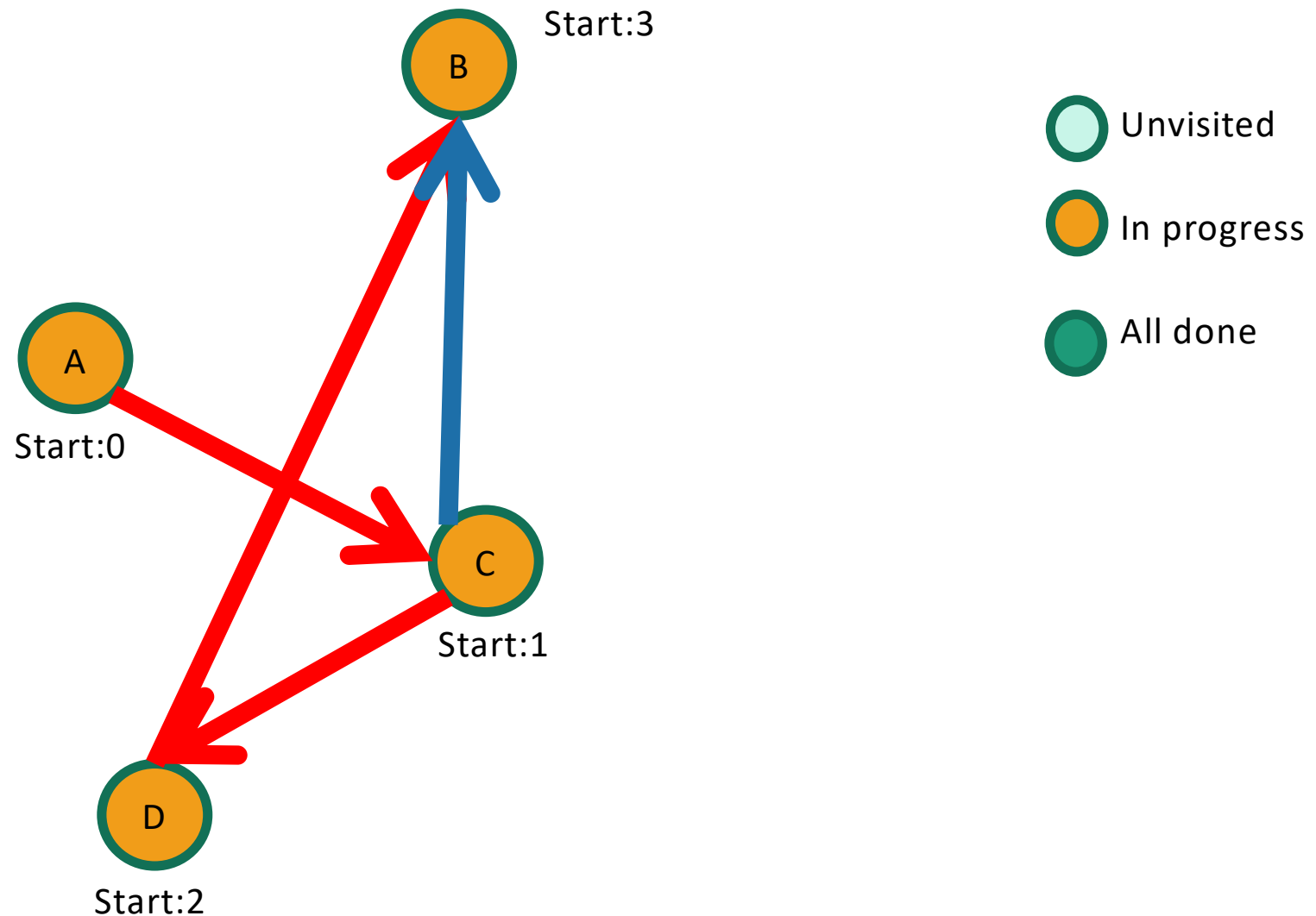
Example

This example skipped in class – here for reference.



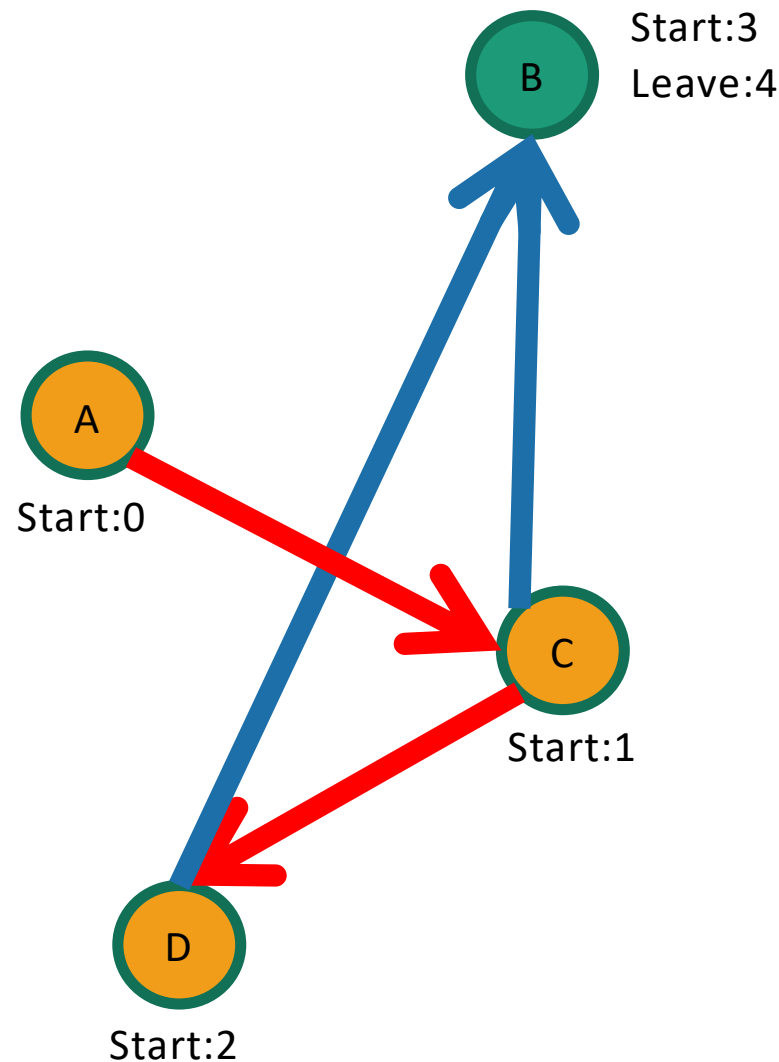
Example

This example skipped in class – here for reference.



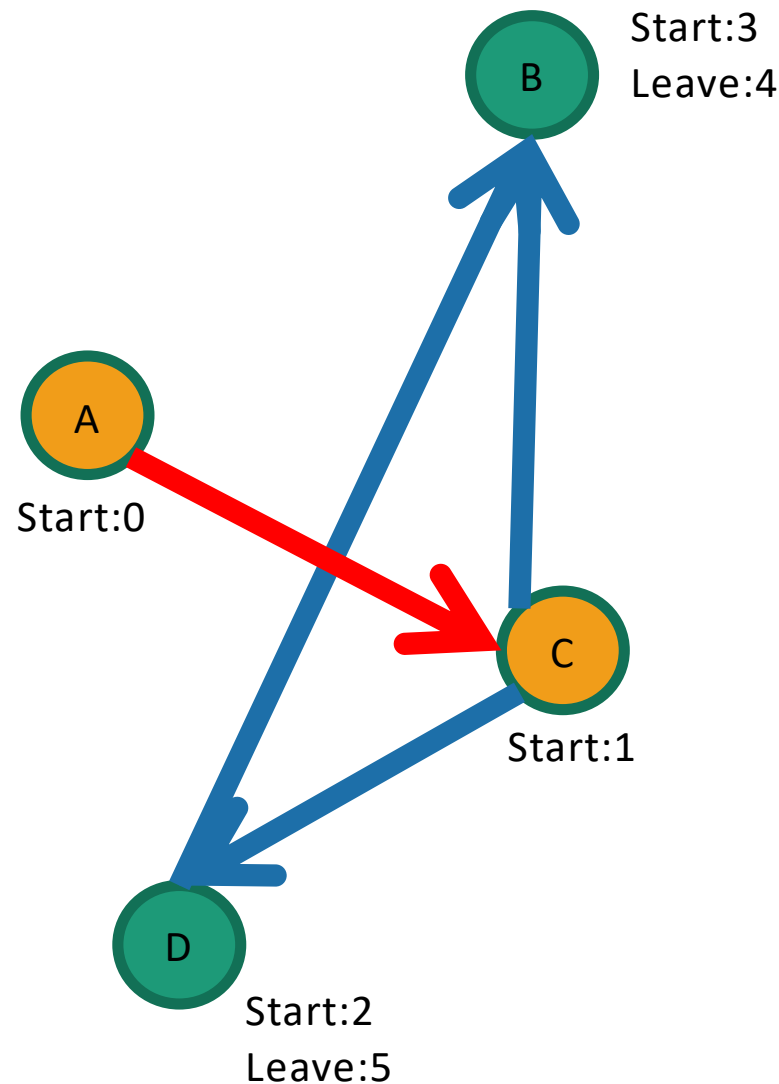
Example

This example skipped in class – here for reference.



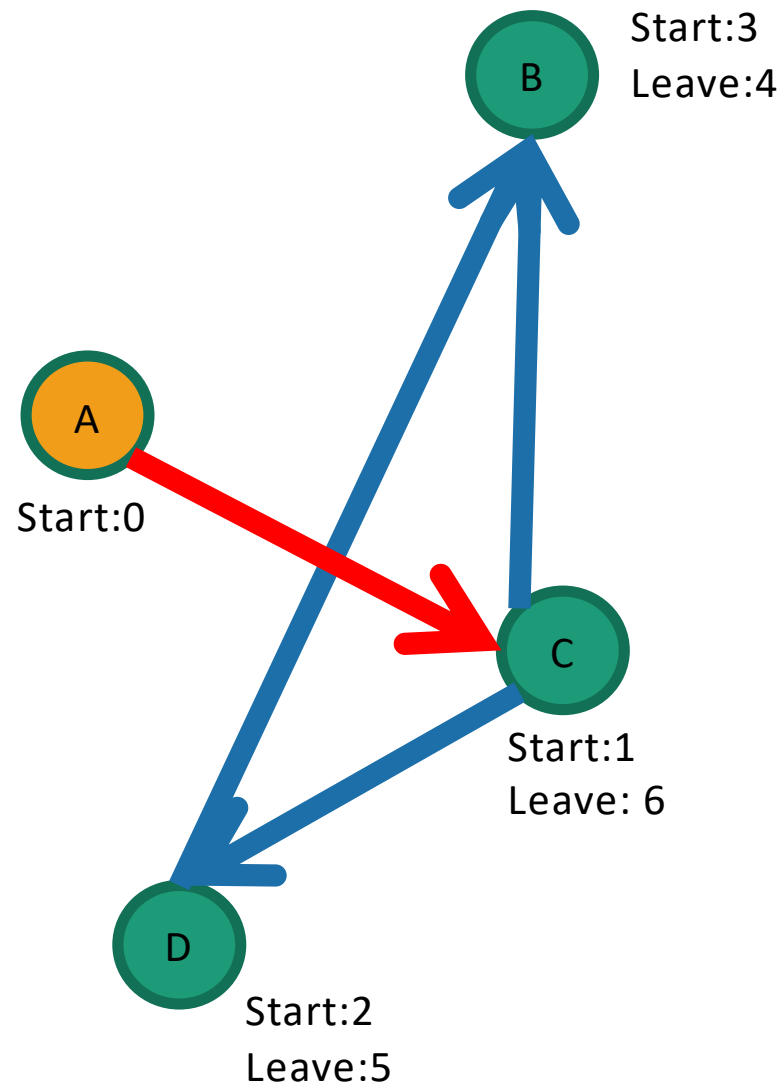
Example

This example skipped in class – here for reference.



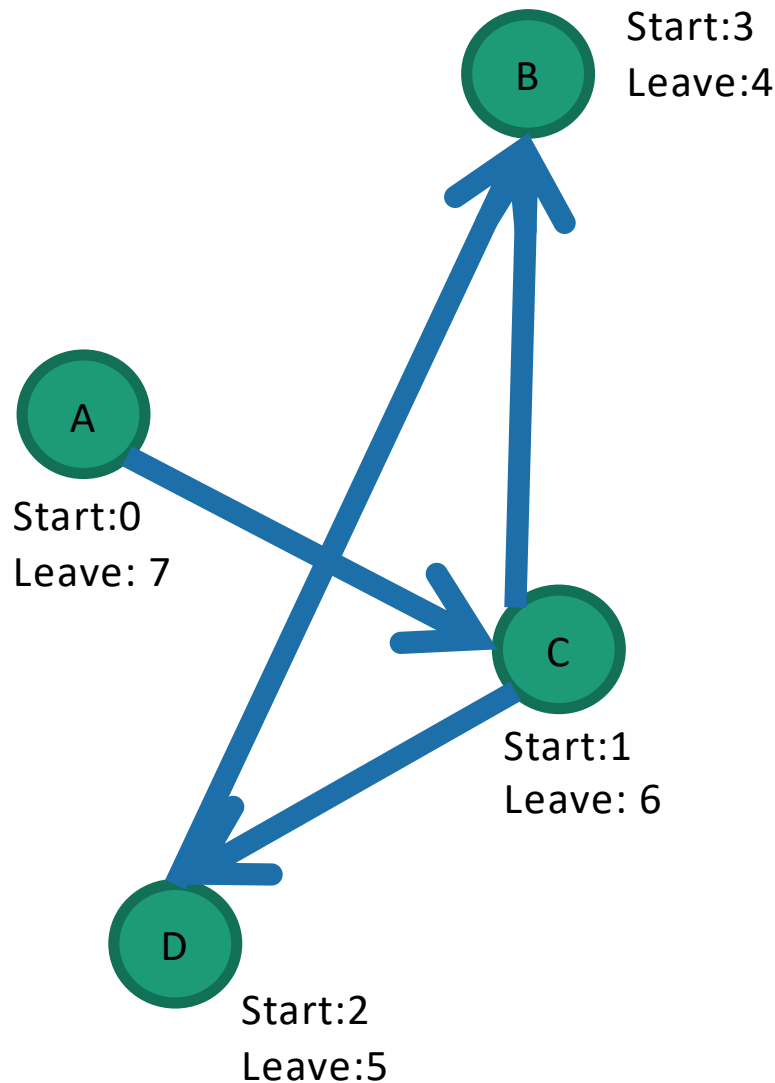
Example

This example skipped in class – here for reference.



Example

This example skipped in class – here for reference.



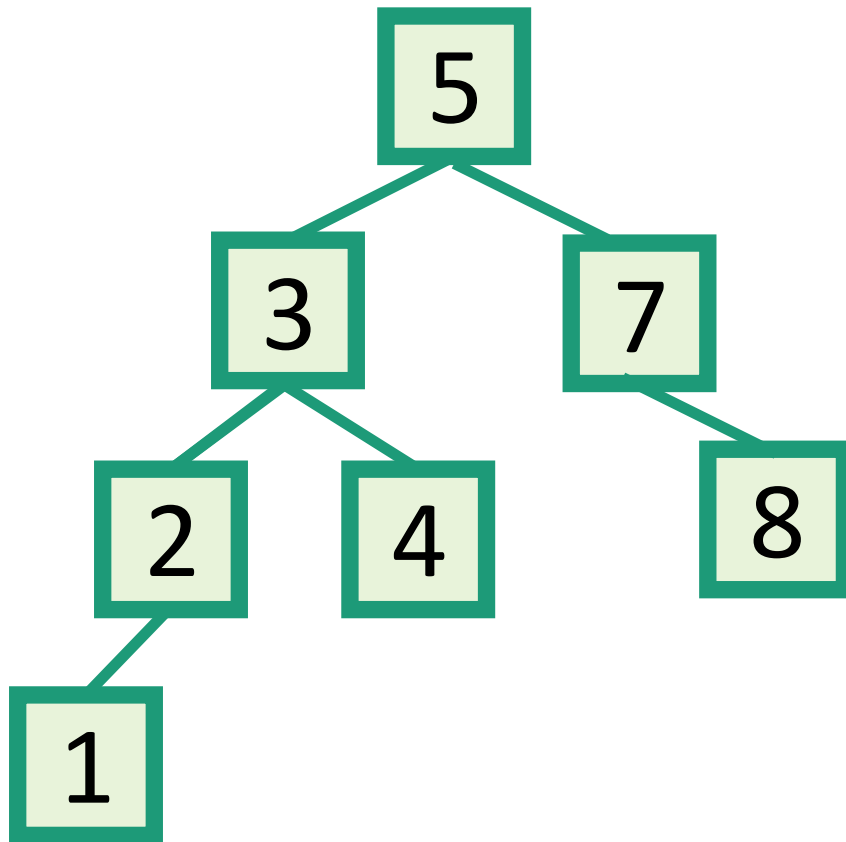
- Unvisited
- In progress
- All done

Do them in this order:



Another use of DFS that we've already seen

- In-order enumeration of binary search trees



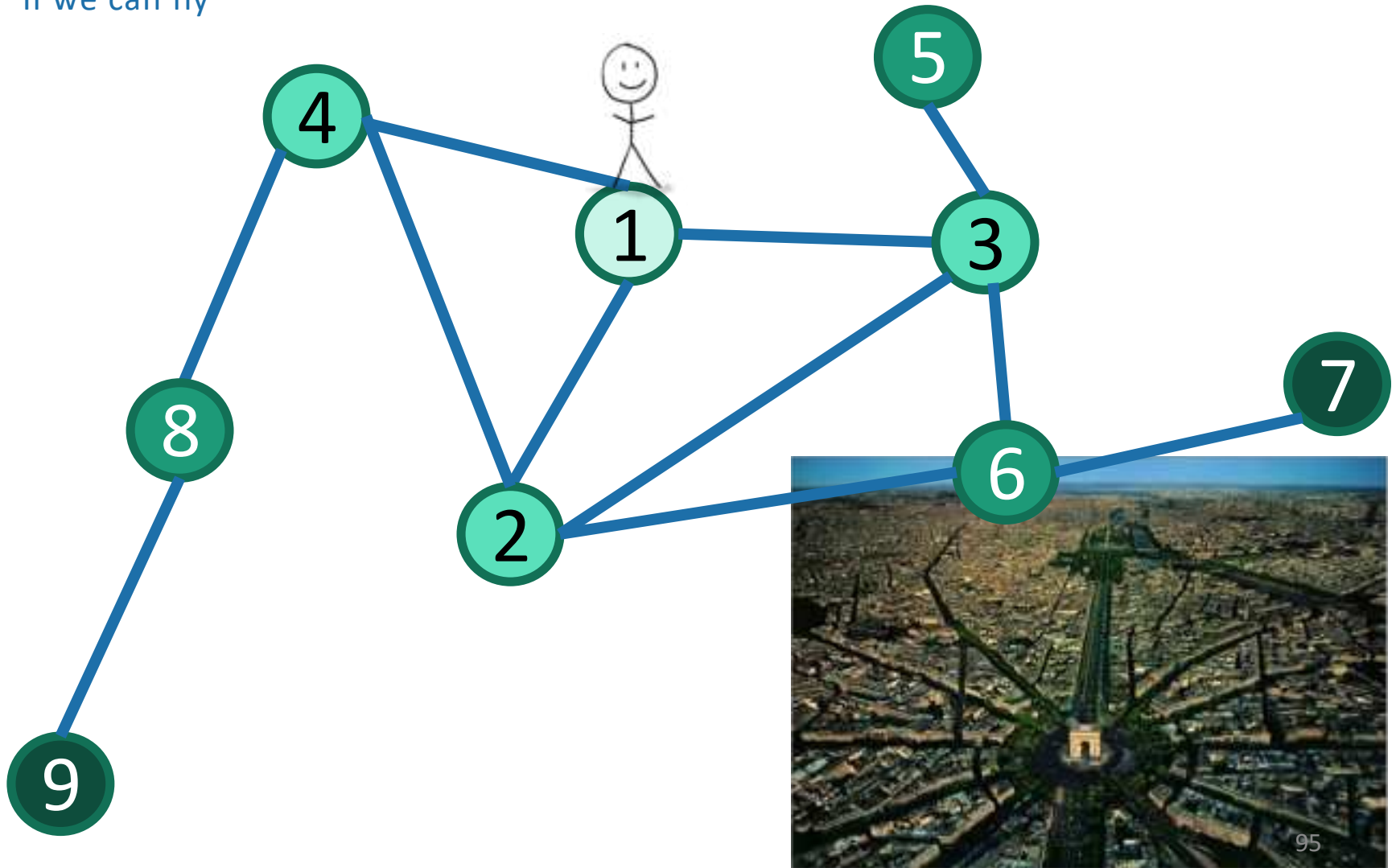
**HINT for Problem 6 on the
current homework! You
can do this for any BST,
even a *gooseTree*!**

Do DFS and print a node's
label when you are done with
the left child and before you
begin the right child.

Part 2: breadth-first search

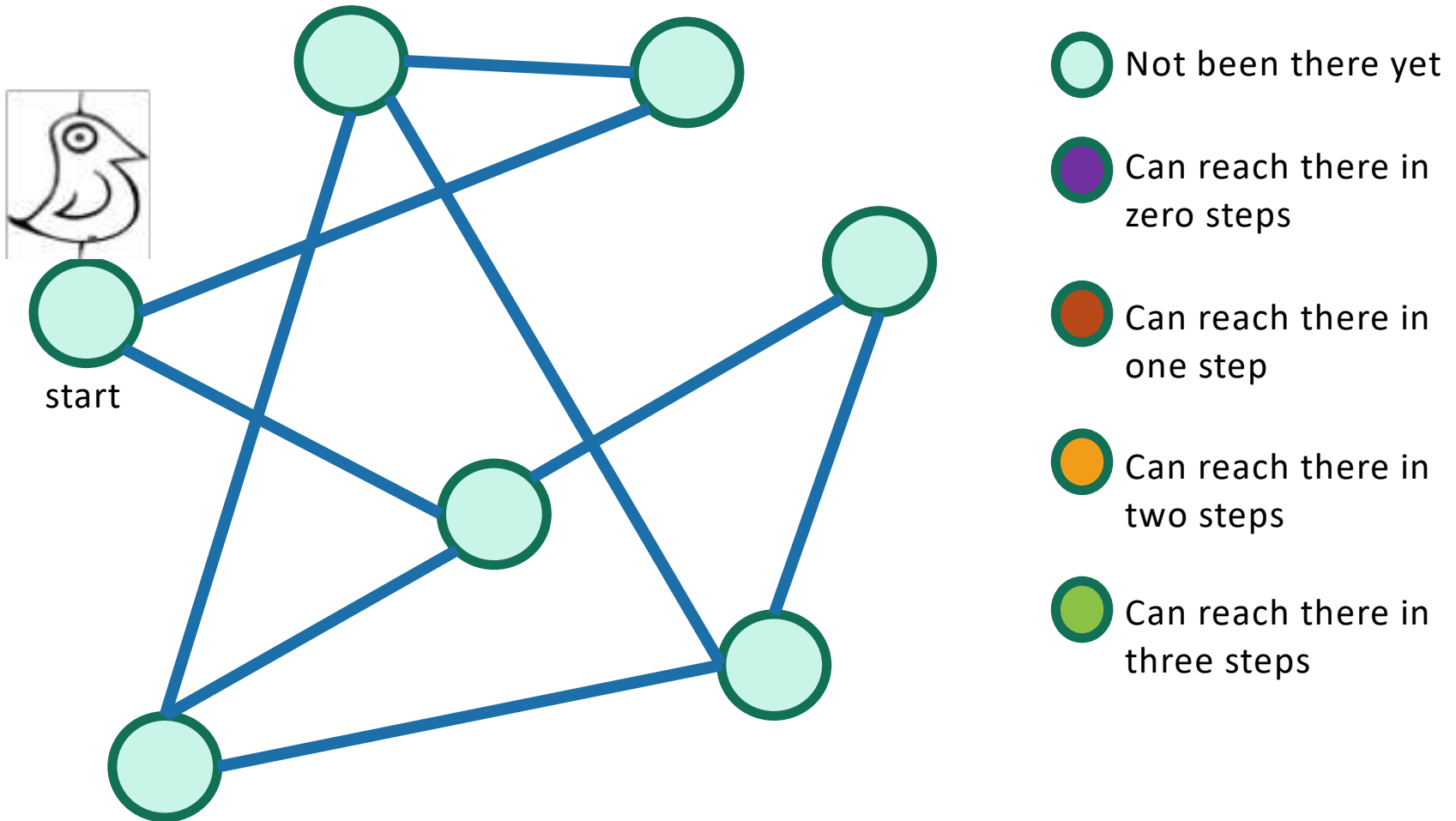
How do we explore a graph?

If we can fly



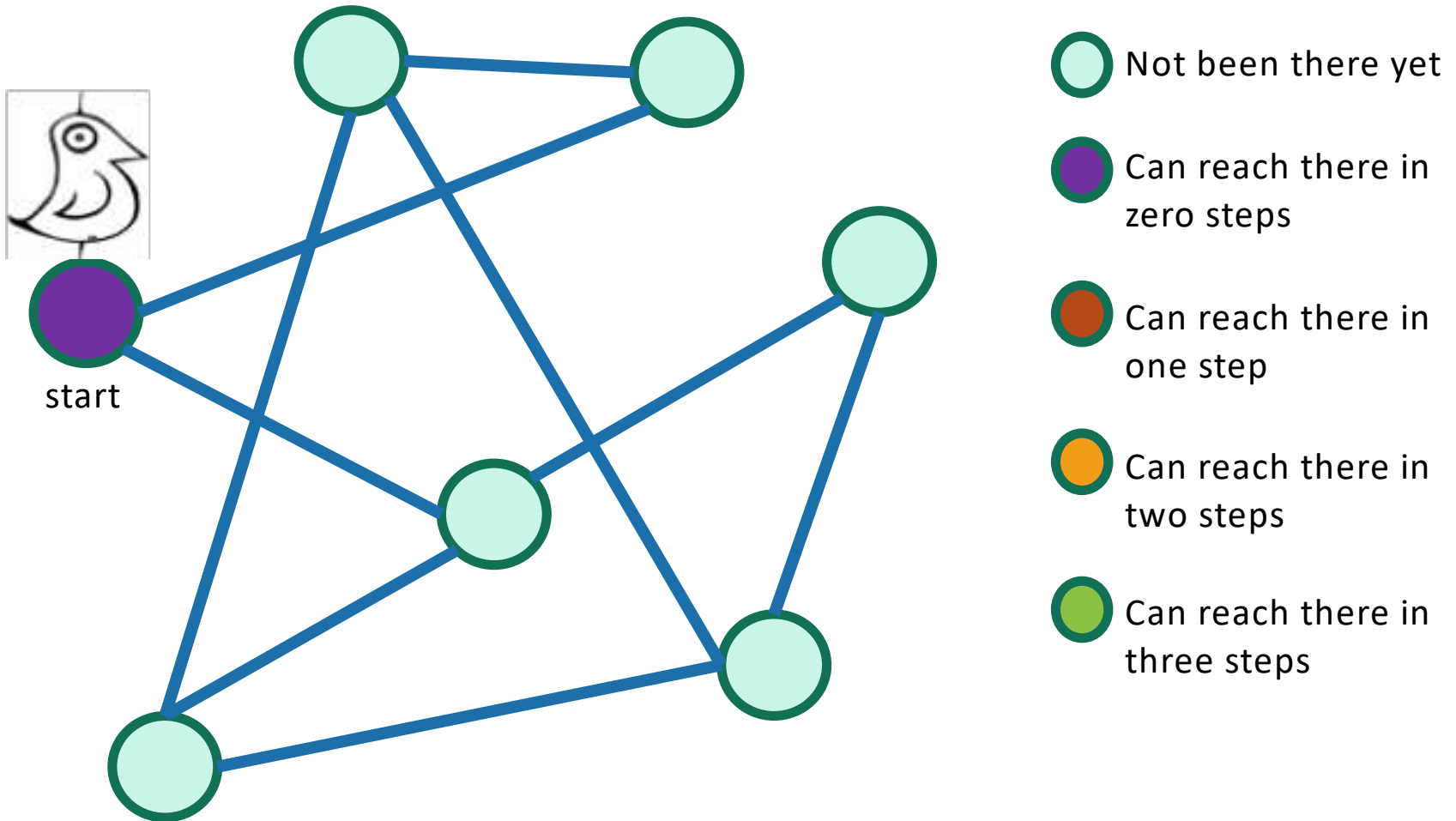
Breadth-First Search

Exploring the world with a bird's-eye view



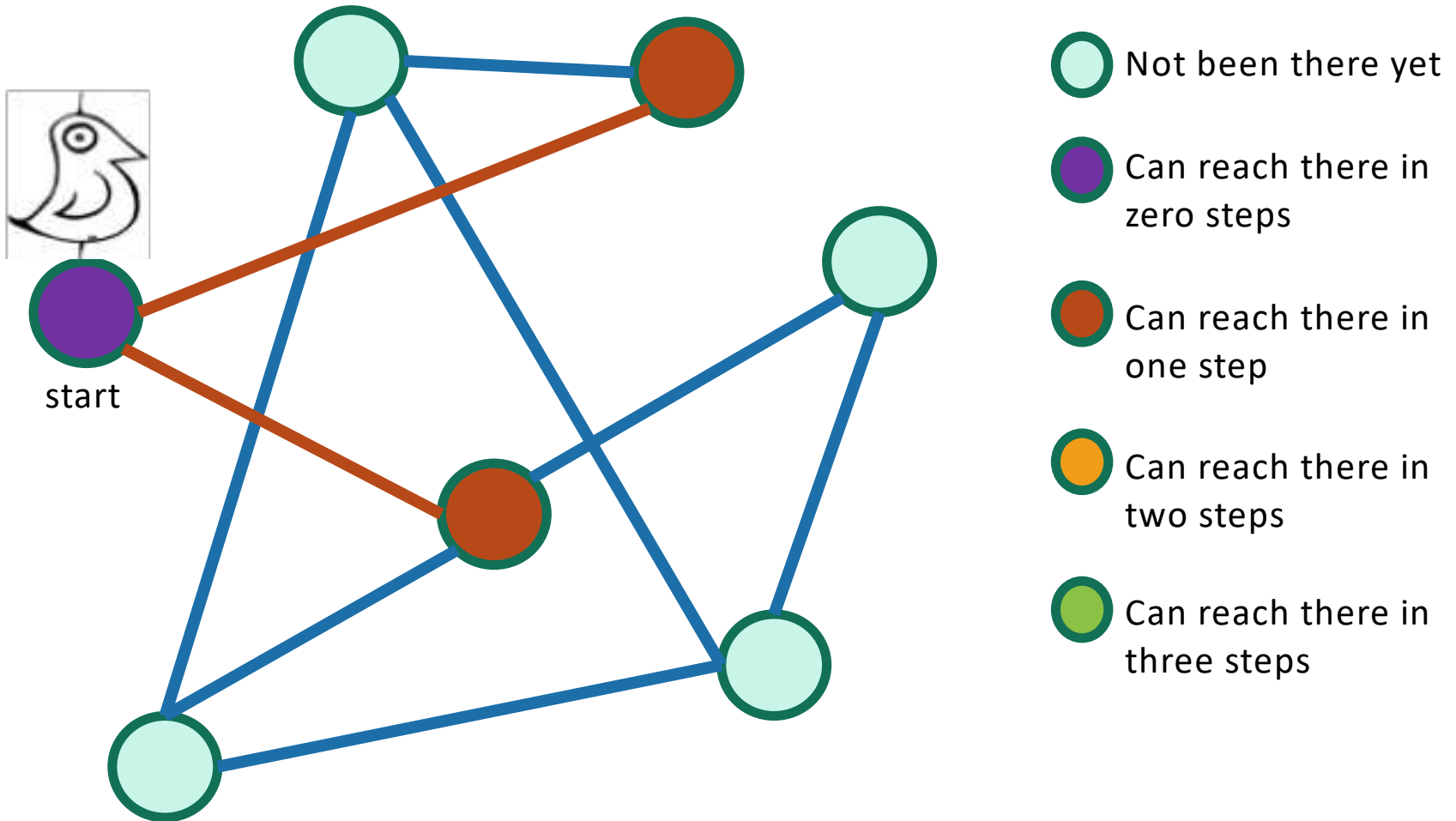
Breadth-First Search

Exploring the world with a bird's-eye view



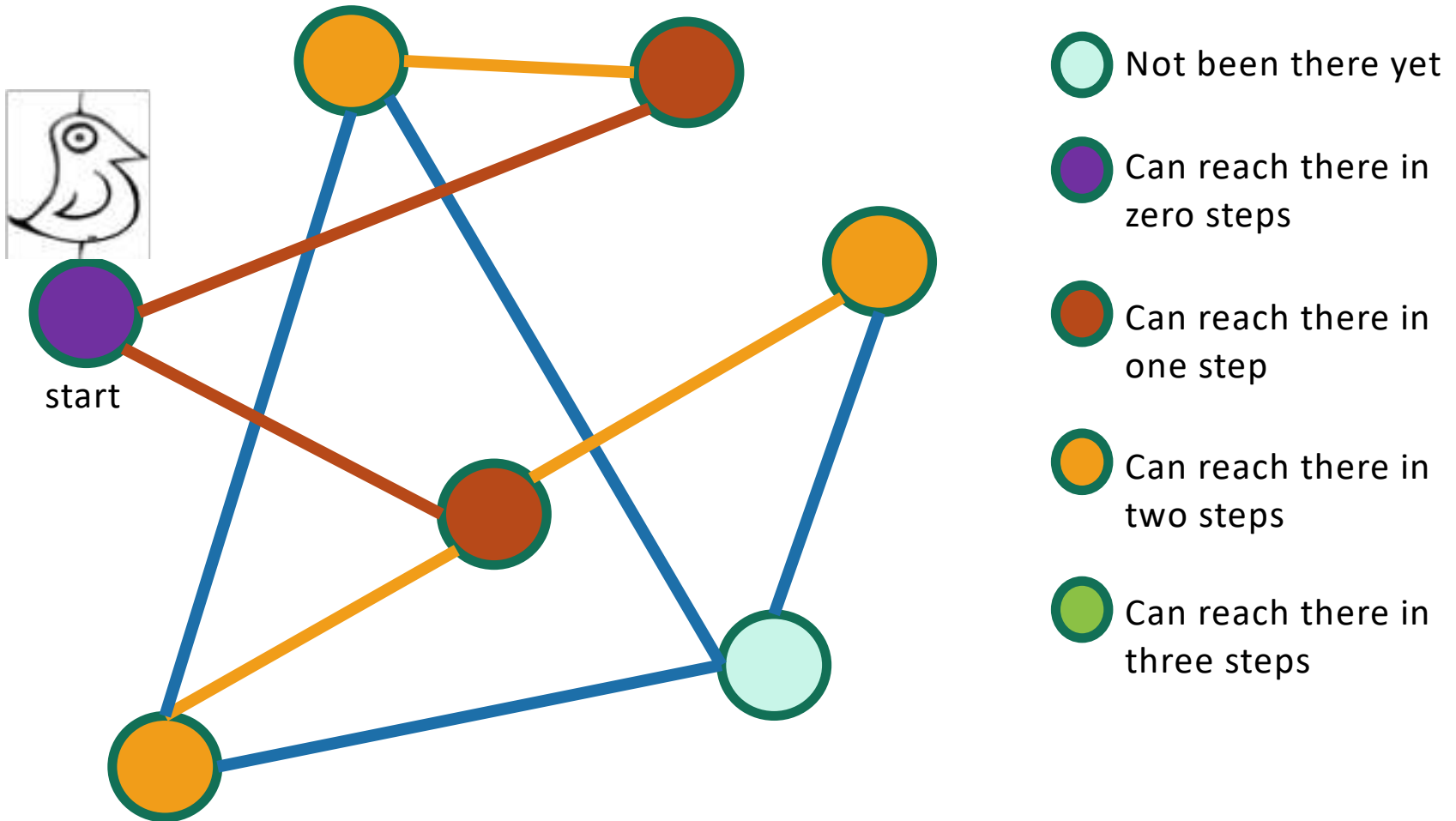
Breadth-First Search

Exploring the world with a bird's-eye view



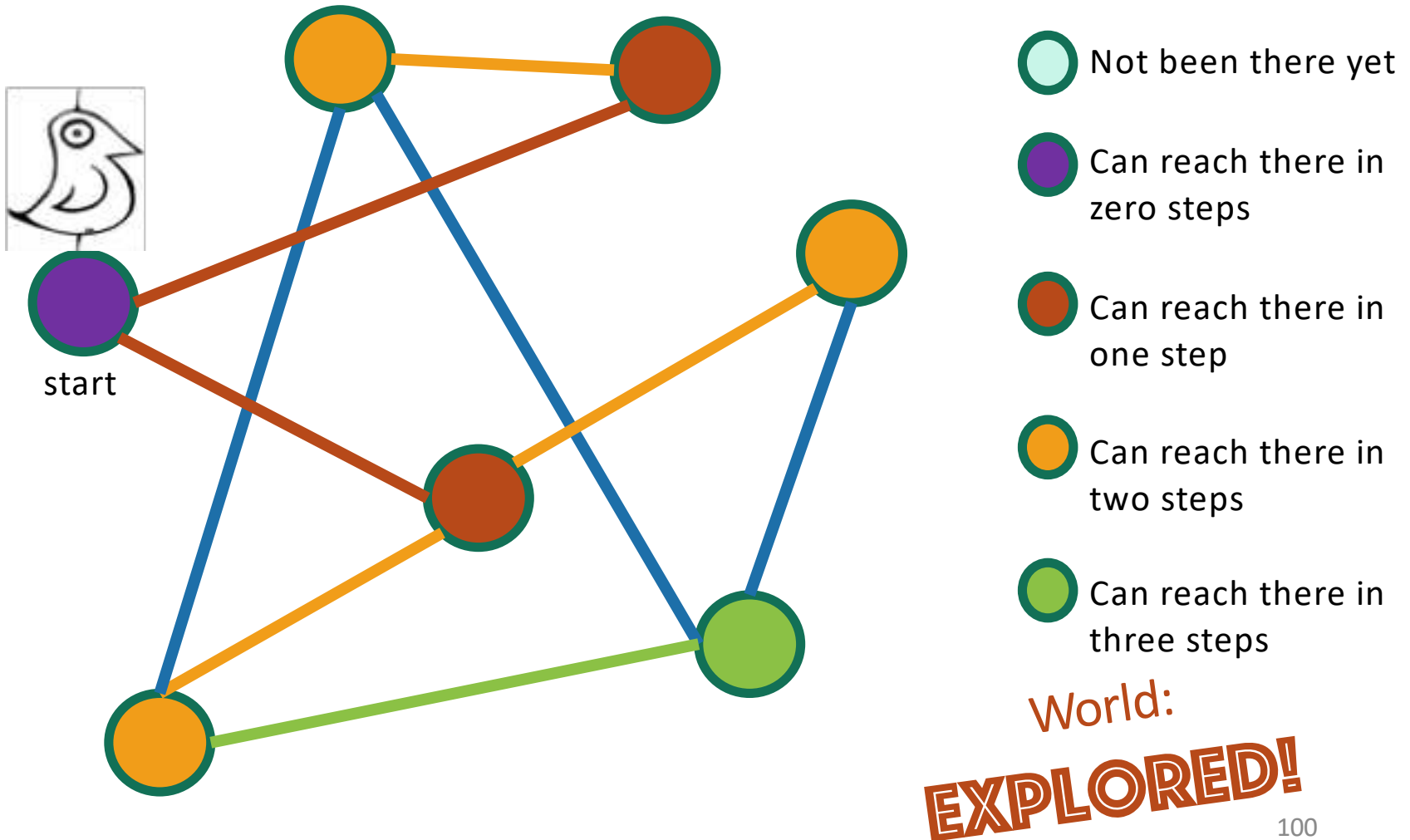
Breadth-First Search

Exploring the world with a bird's-eye view



Breadth-First Search

Exploring the world with a bird's-eye view



Breadth-First Search

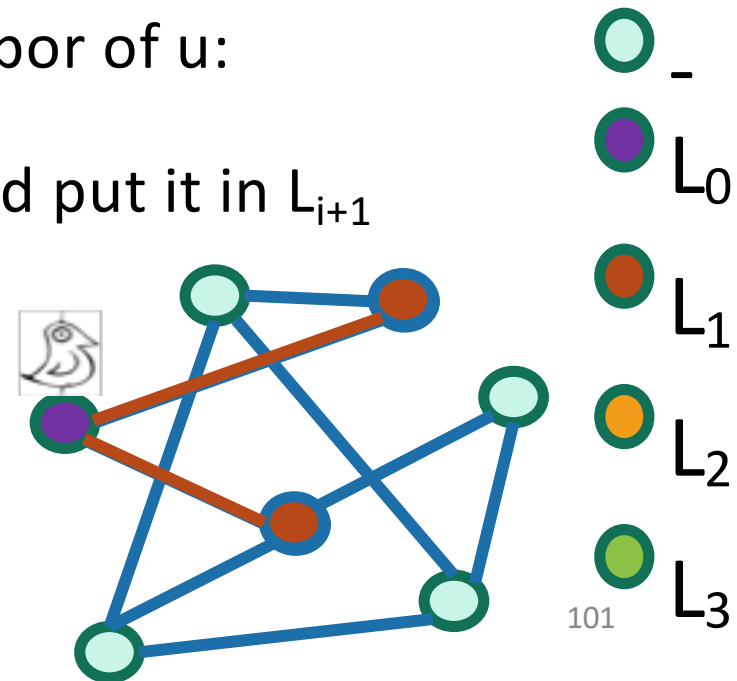
Exploring the world with pseudocode

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

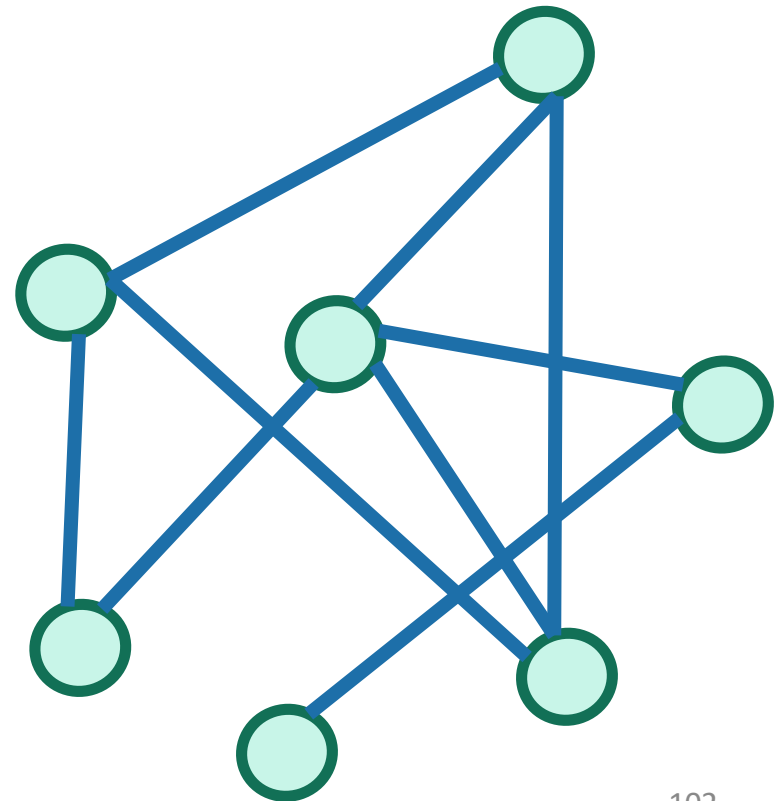
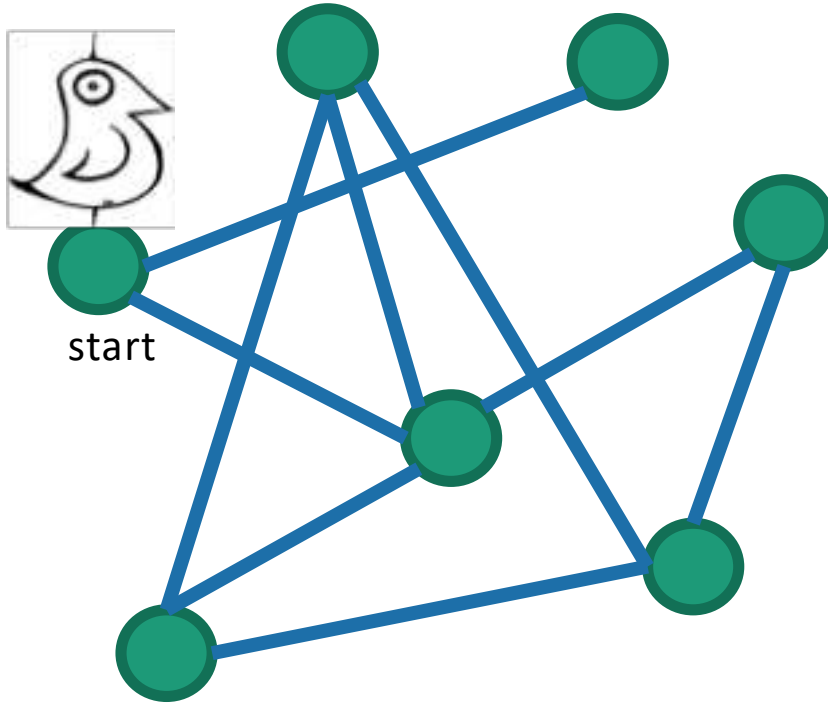
L_i is the set of nodes
we can reach in i
steps from w

- **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}



BFS also finds all the nodes reachable from the starting point



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

Running time and extension to directed graphs

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

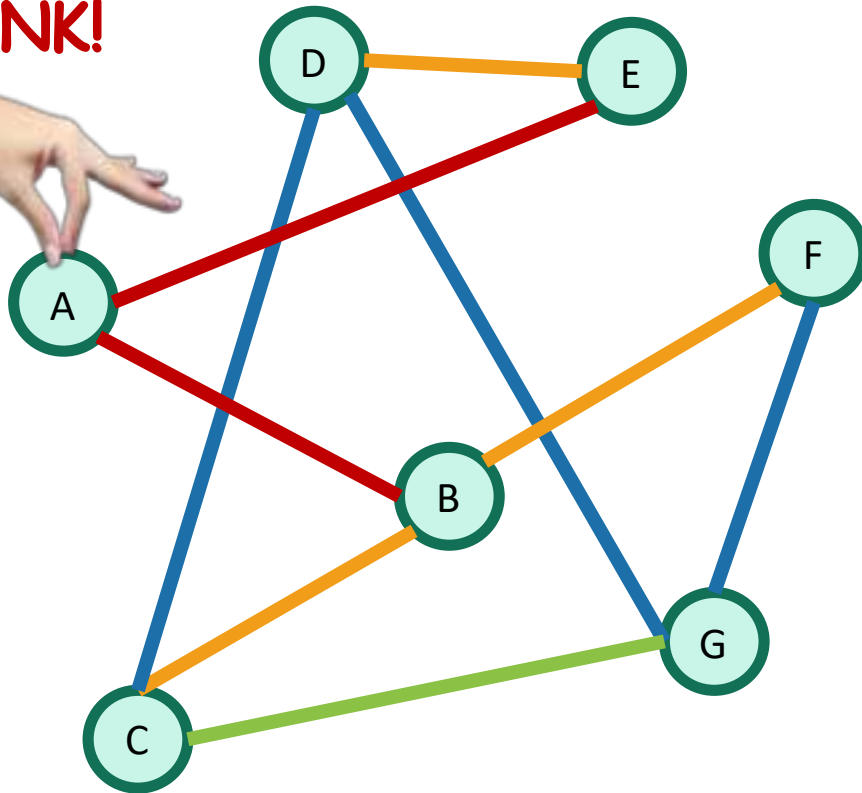
Verify these!



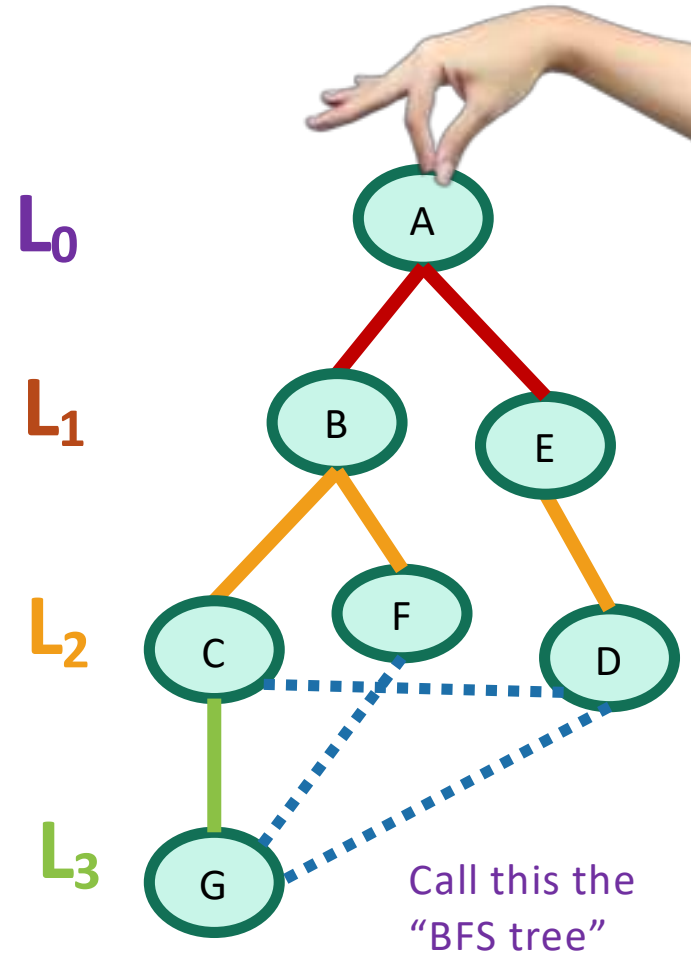
Why is it called breadth-first?

- We are implicitly building a tree:

YOINK!

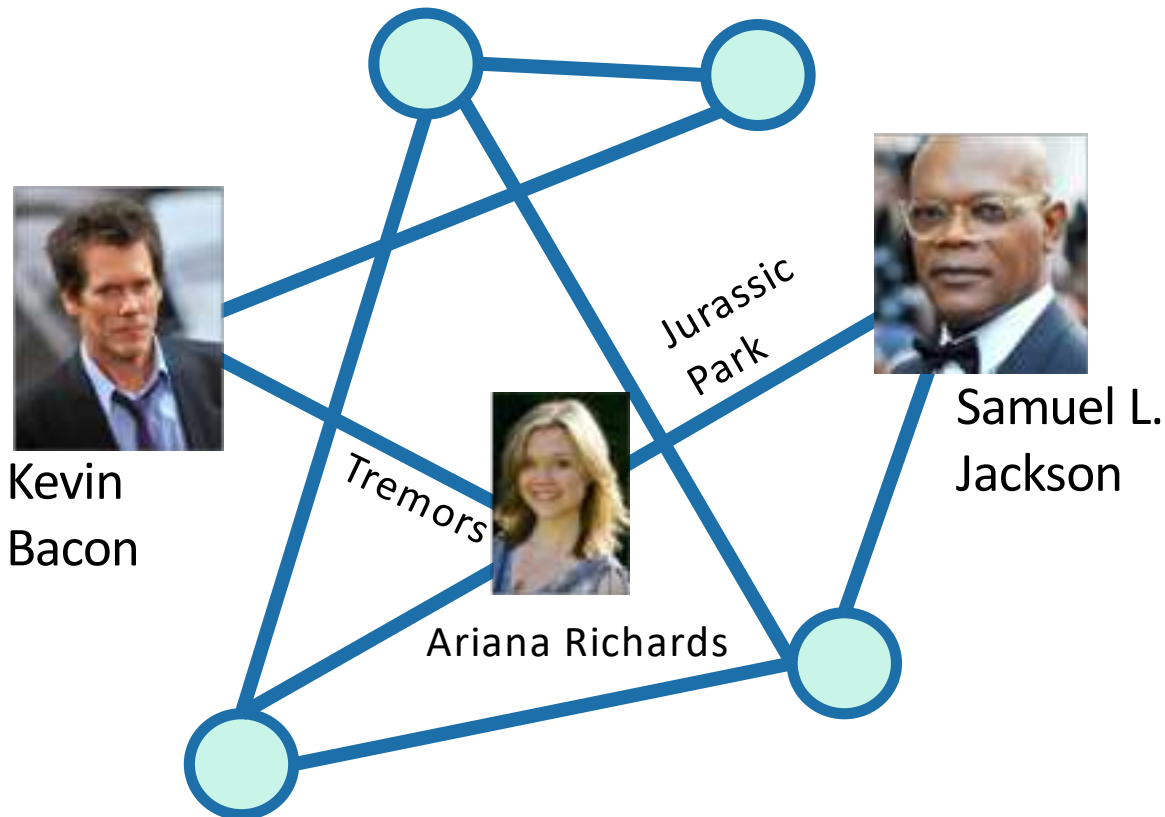


- First we go as broadly as we can.



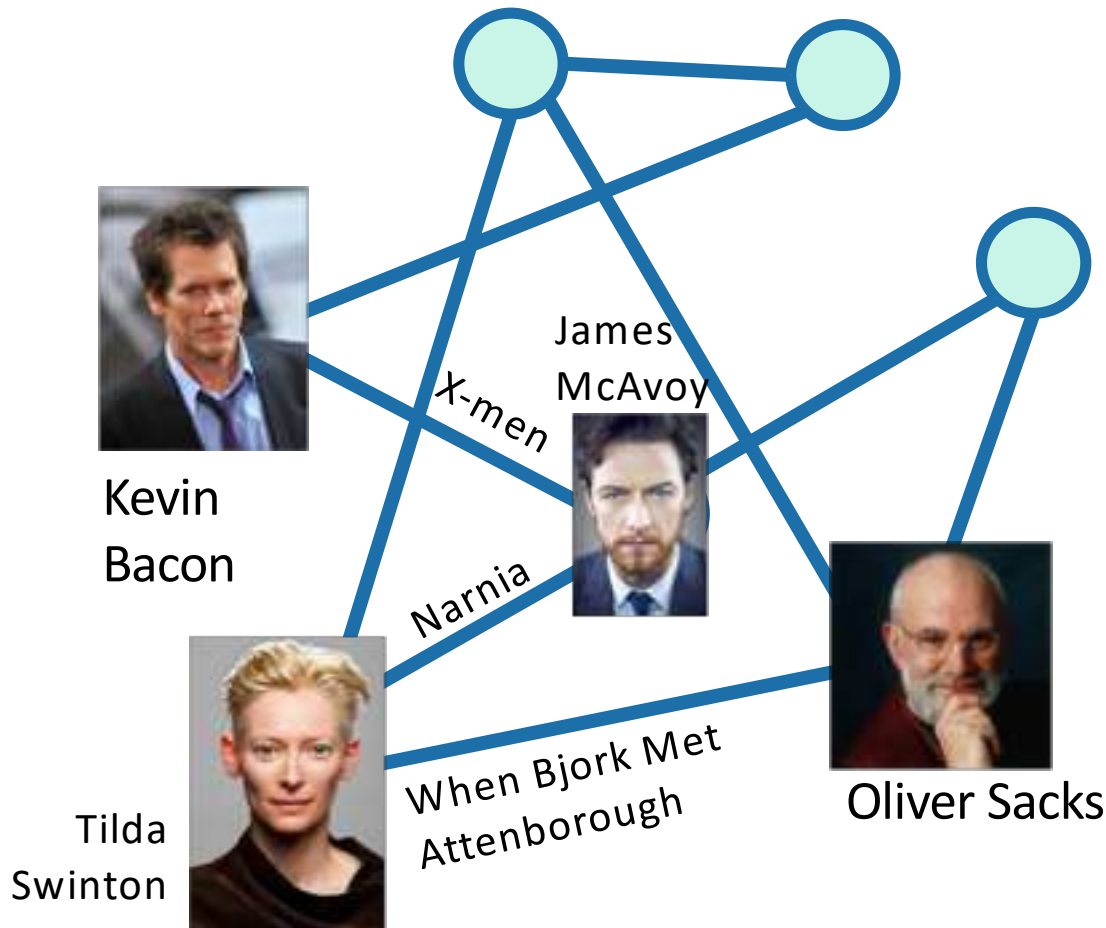
Pre-lecture exercise

- What Samuel L. Jackson's Bacon number?



(Answer: 2)

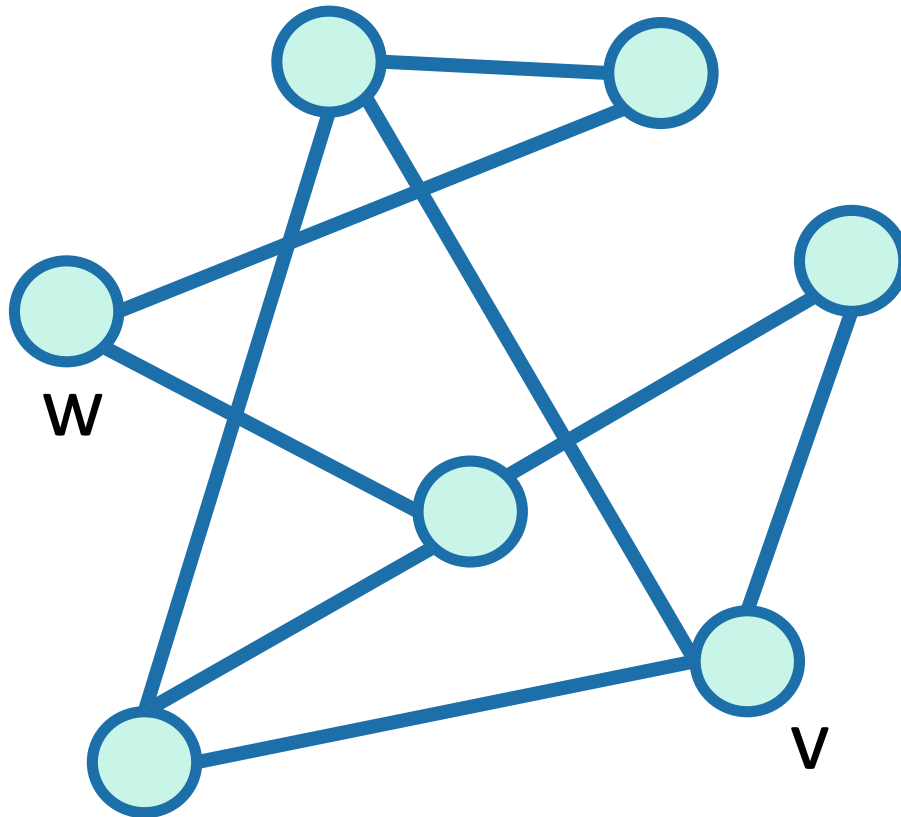
I wrote the pre-lecture exercise before I realized that I really wanted an example with distance 3



It is really hard to find
people with Bacon
number 3!

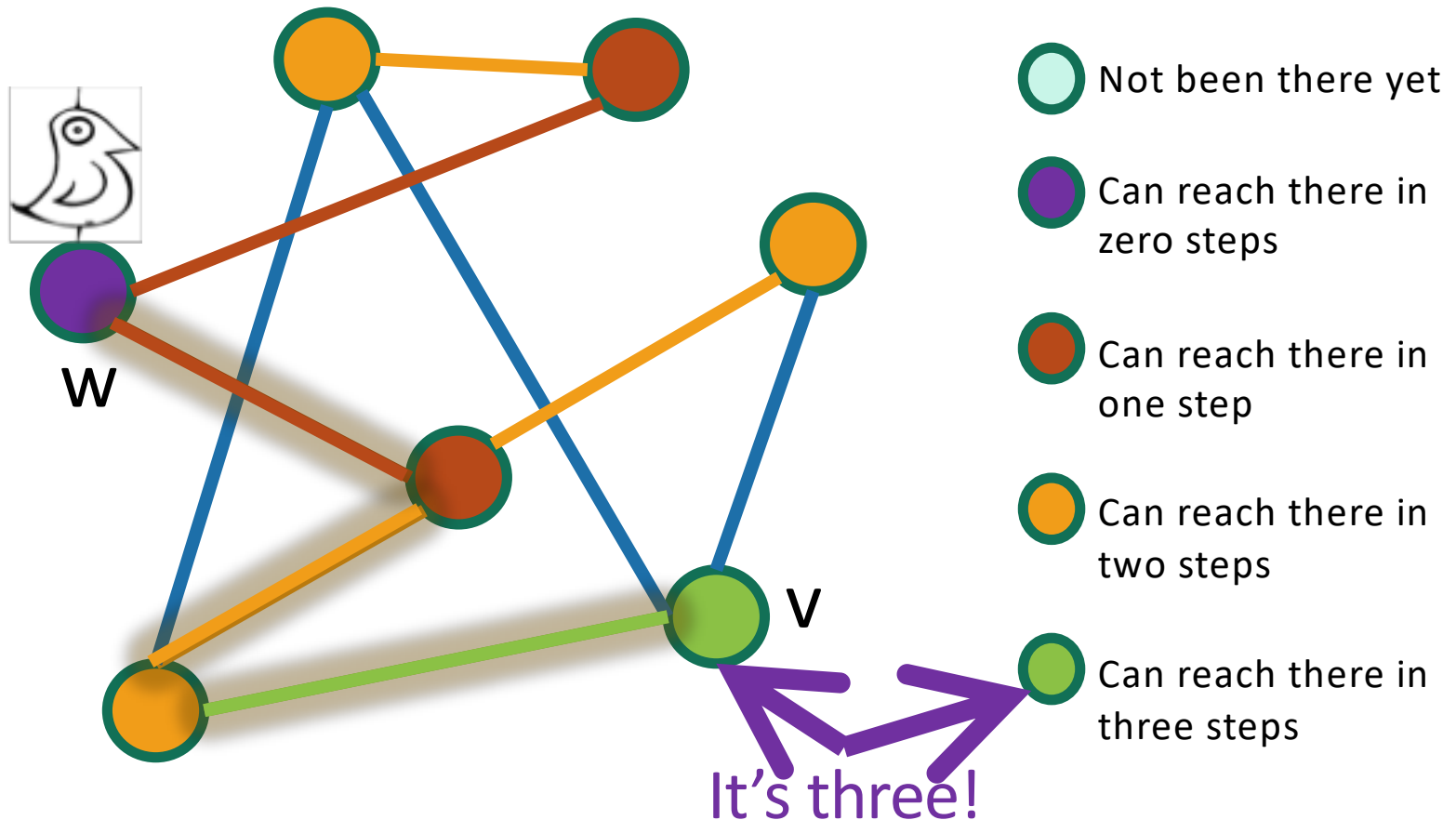
Application of BFS: shortest path

- 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?



To find the **distance** between w and all other vertices v

The **distance** between two vertices is the number of edges in the shortest path between them.

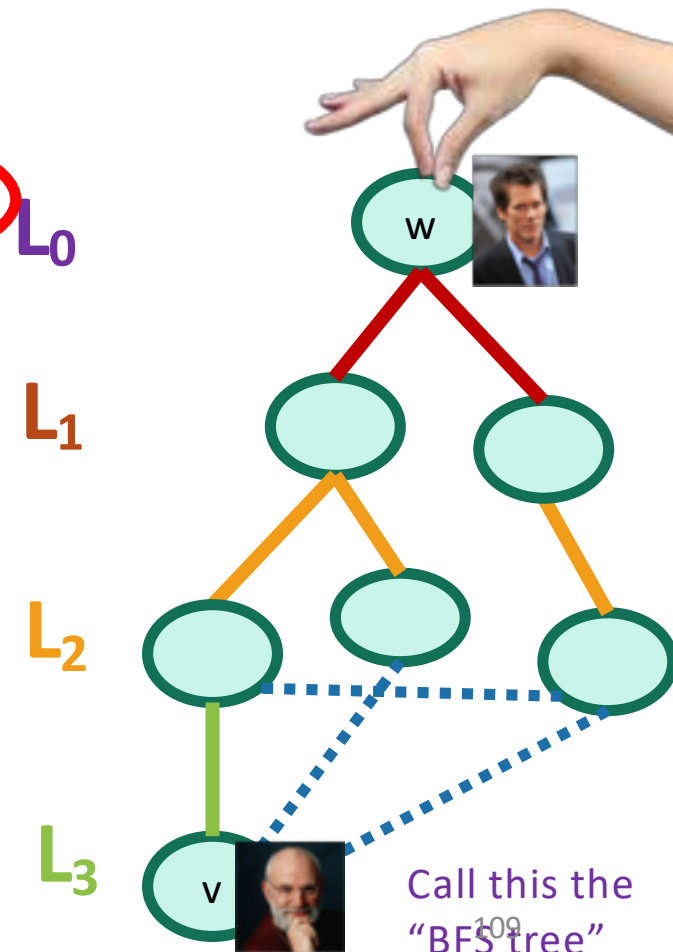
- Do a BFS starting at w
- For all v in L_i
 - The shortest path between w and v has length i
 - A shortest path between w and v is given by the path in the BFS tree.
- If we never found v , the distance is infinite.

This requires some proof! See skipped slide.

Modify the BFS pseudocode to return shortest paths!



Gauss has no Bacon number



Call this the
"BFS tree"

Proof overview

THIS SLIDE SKIPPED IN CLASS

See CLRS for details!



that the BFS tree behaves like it should

- Proof by induction.
- Inductive hypothesis for j :
 - For all $i < j$ the vertices in L_i have distance i from v .
- Base case:
 - $L_0 = \{v\}$, so we're good.
- Inductive step:
 - Let w be in L_j . Want to show $\text{dist}(v, w) = j$.
 - We know $\text{dist}(v, w) \leq j$, since $\text{dist}(v, w\text{'s parent in } L_{j-1}) = j-1$ by induction, so that gives a path of length j from v to w .
 - On the other hand, $\text{dist}(v, w) \geq j$, since if $\text{dist}(v, w) < j$, w would have shown up in an earlier layer.
 - Thus, $\text{dist}(v, w) = j$.
- Conclusion:
 - For each vertex w in V , if w is in L_j , then $\text{dist}(v, w) = j$.

What have we learned?

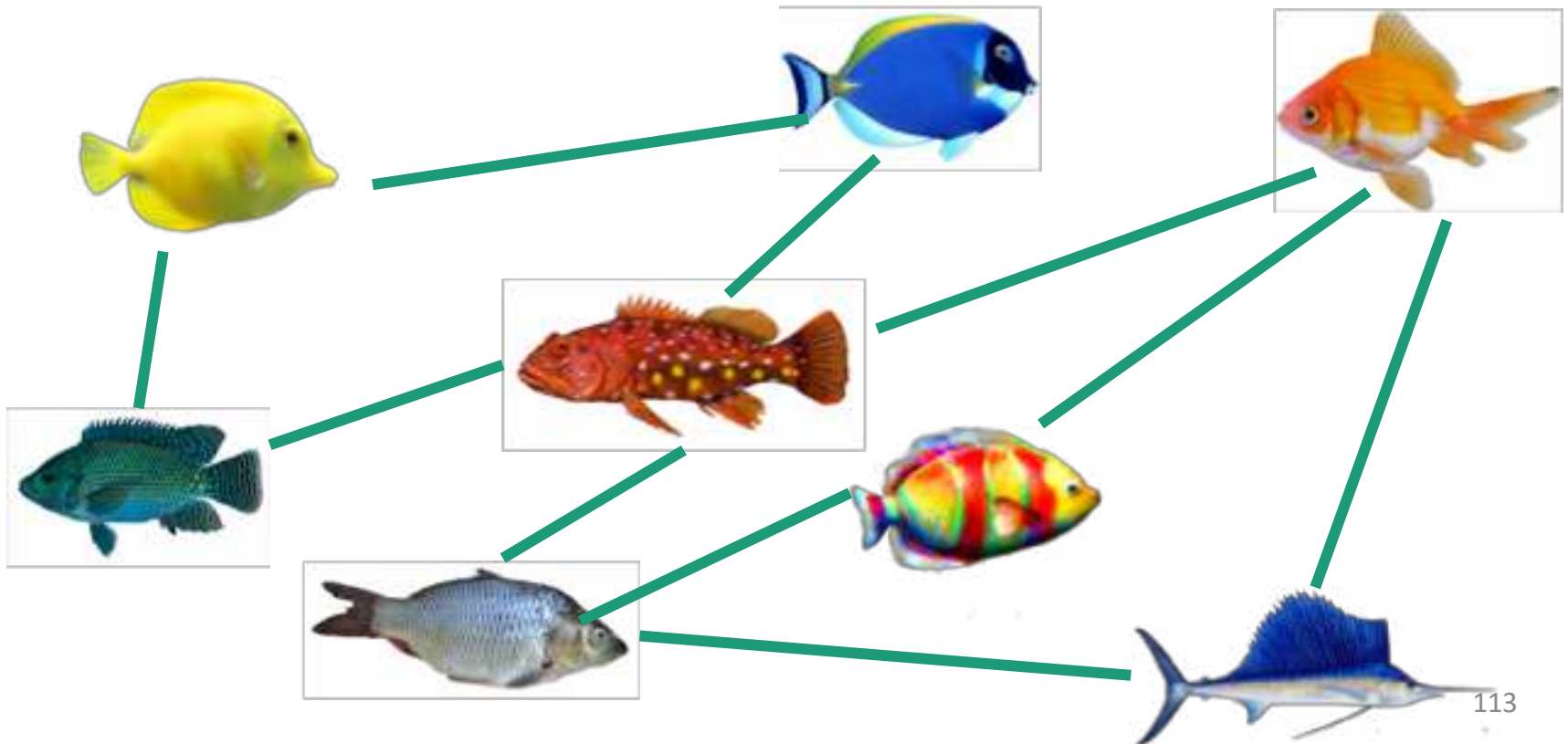
- The BFS tree is useful for computing distances between pairs of vertices.
- We can find the shortest path between u and v in time $O(m)$.

Another application of BFS

- Testing bipartite-ness

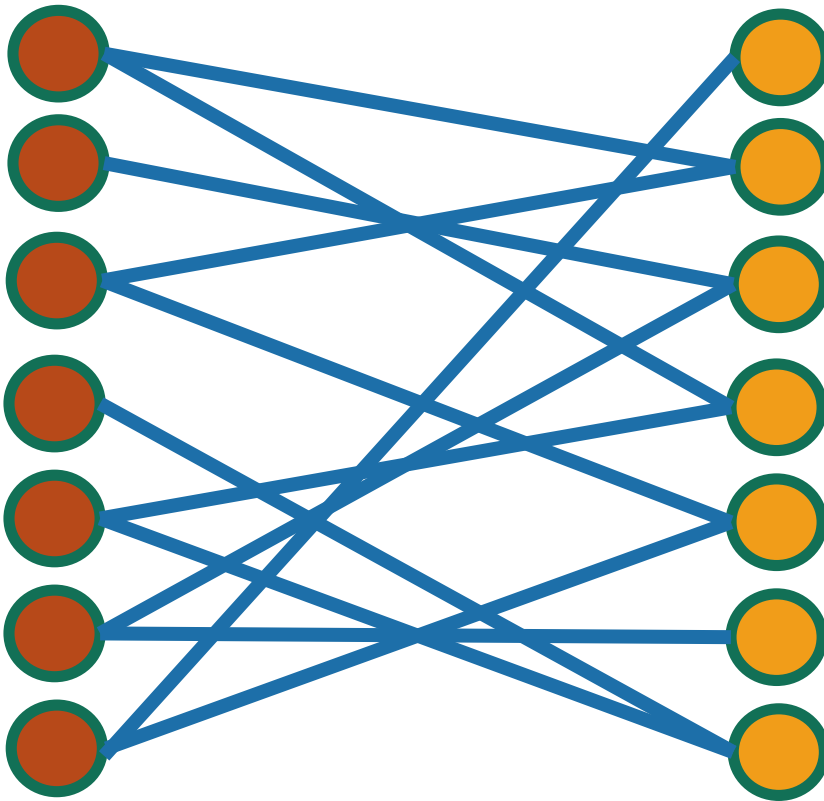
Pre-lecture exercise: fish

- You have a bunch of fish and two fish tanks.
- Some pairs of fish will fight if put in the same tank.
 - Model this as a graph: connected fish will fight.
- Can you put the fish in the two tanks so that there is no fighting?



Bipartite graphs

- A bipartite graph looks like this:



Can color the vertices red and orange so that there are no edges between any same-colored vertices

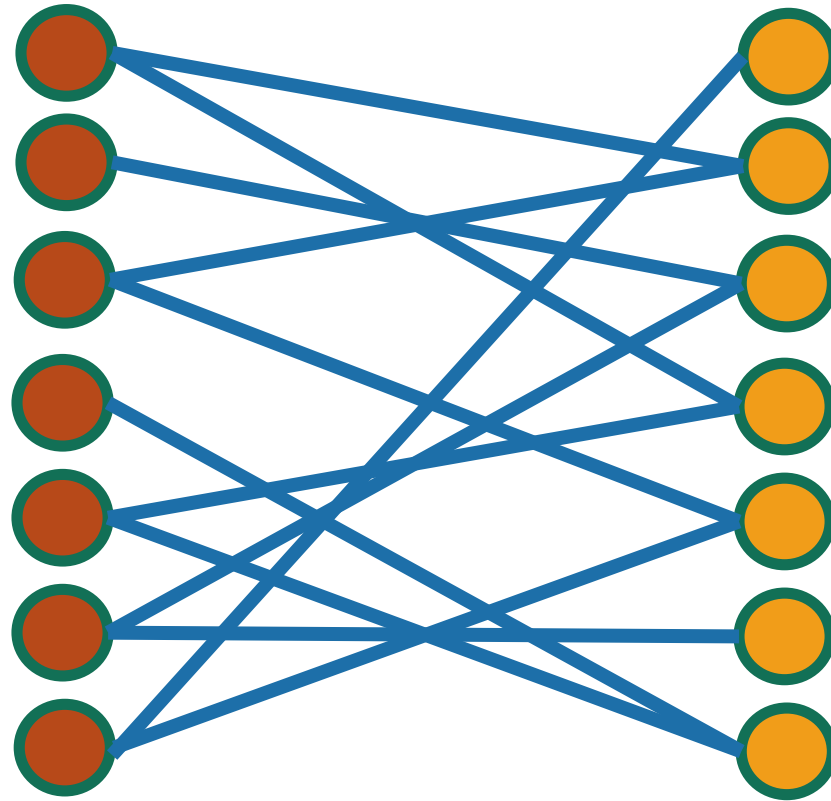
Example:

- are in tank A
- are in tank B
- if the fish fight

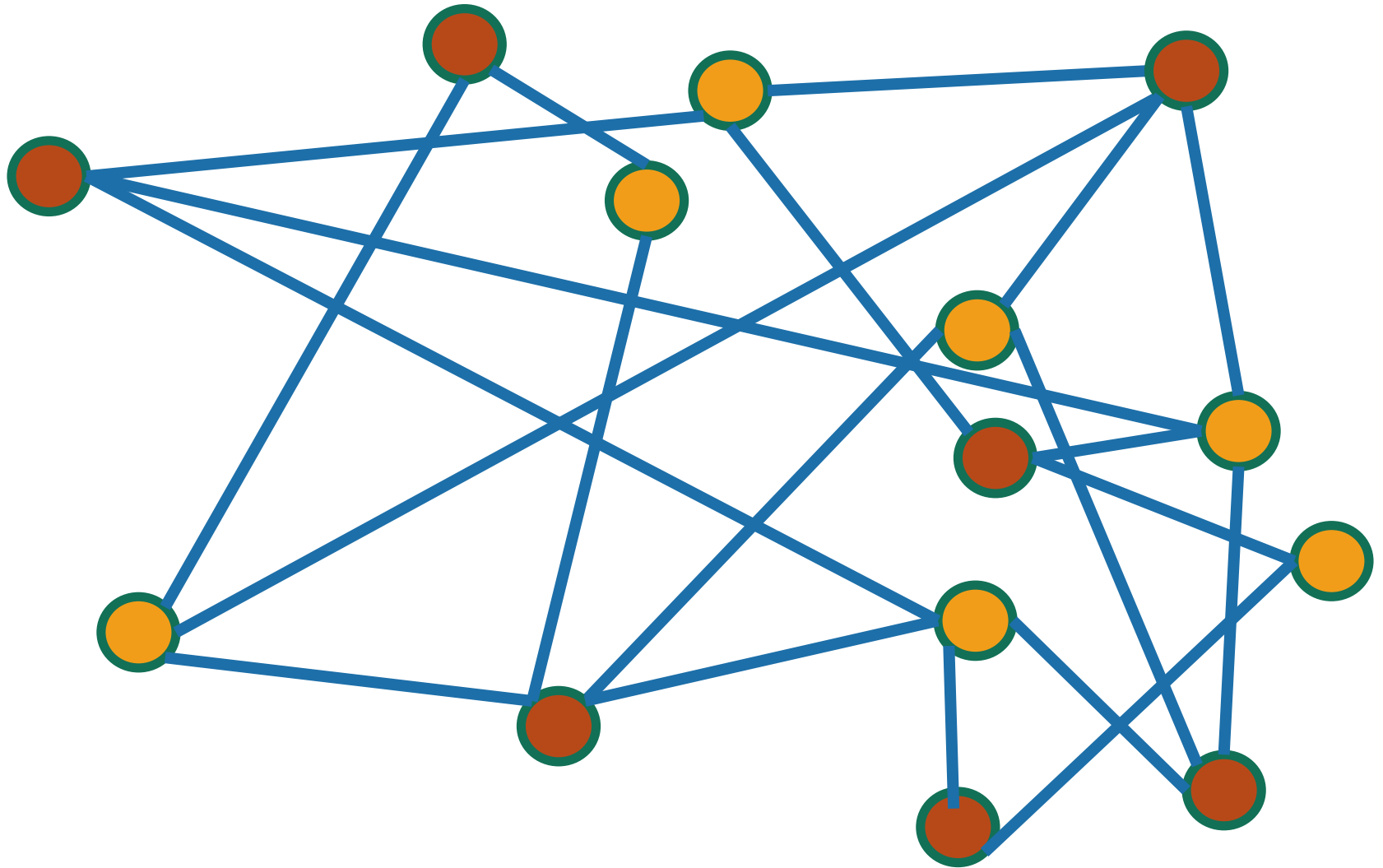
Example:

- are students
- are classes
- if the student is enrolled in the class

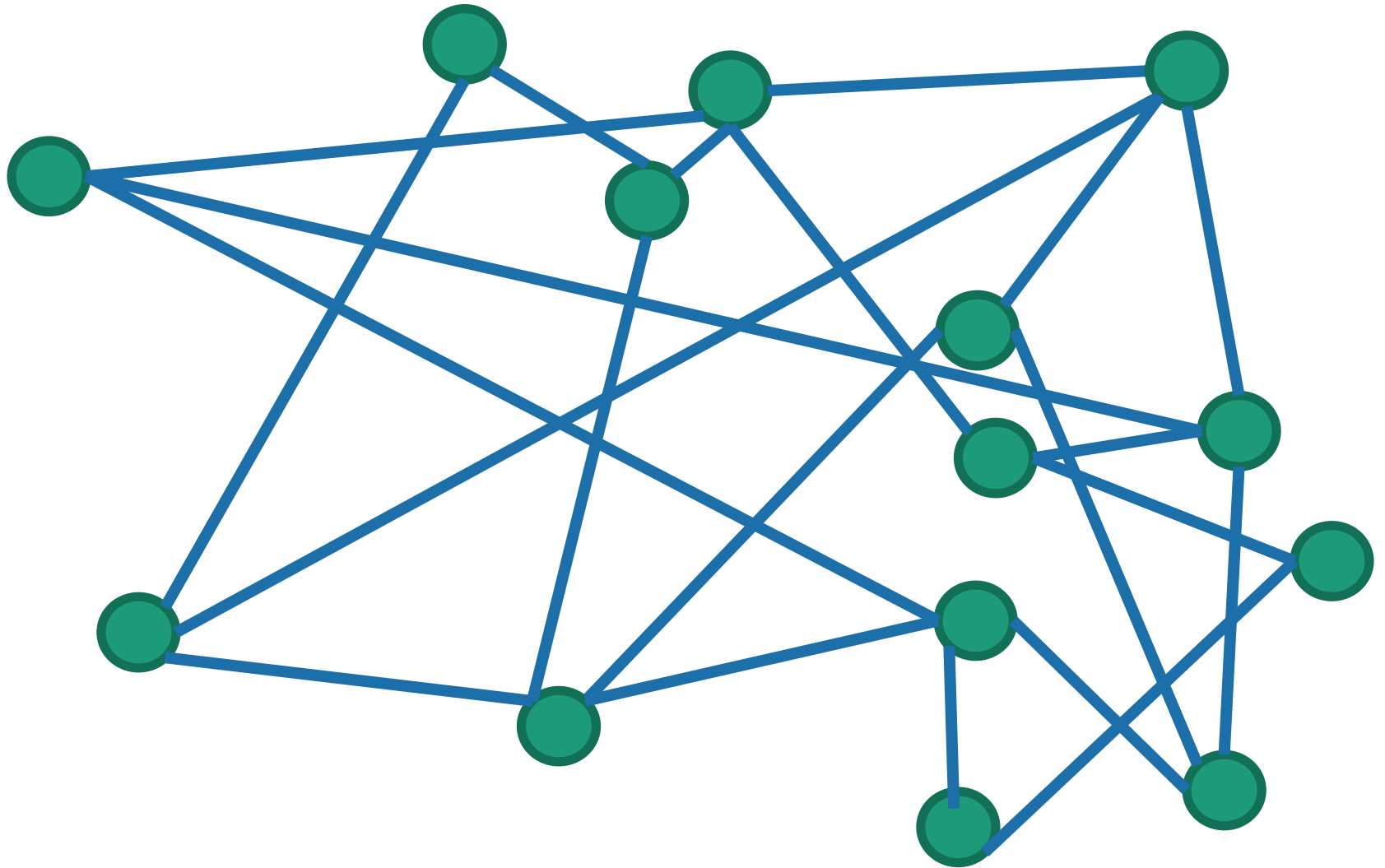
Is this graph bipartite?



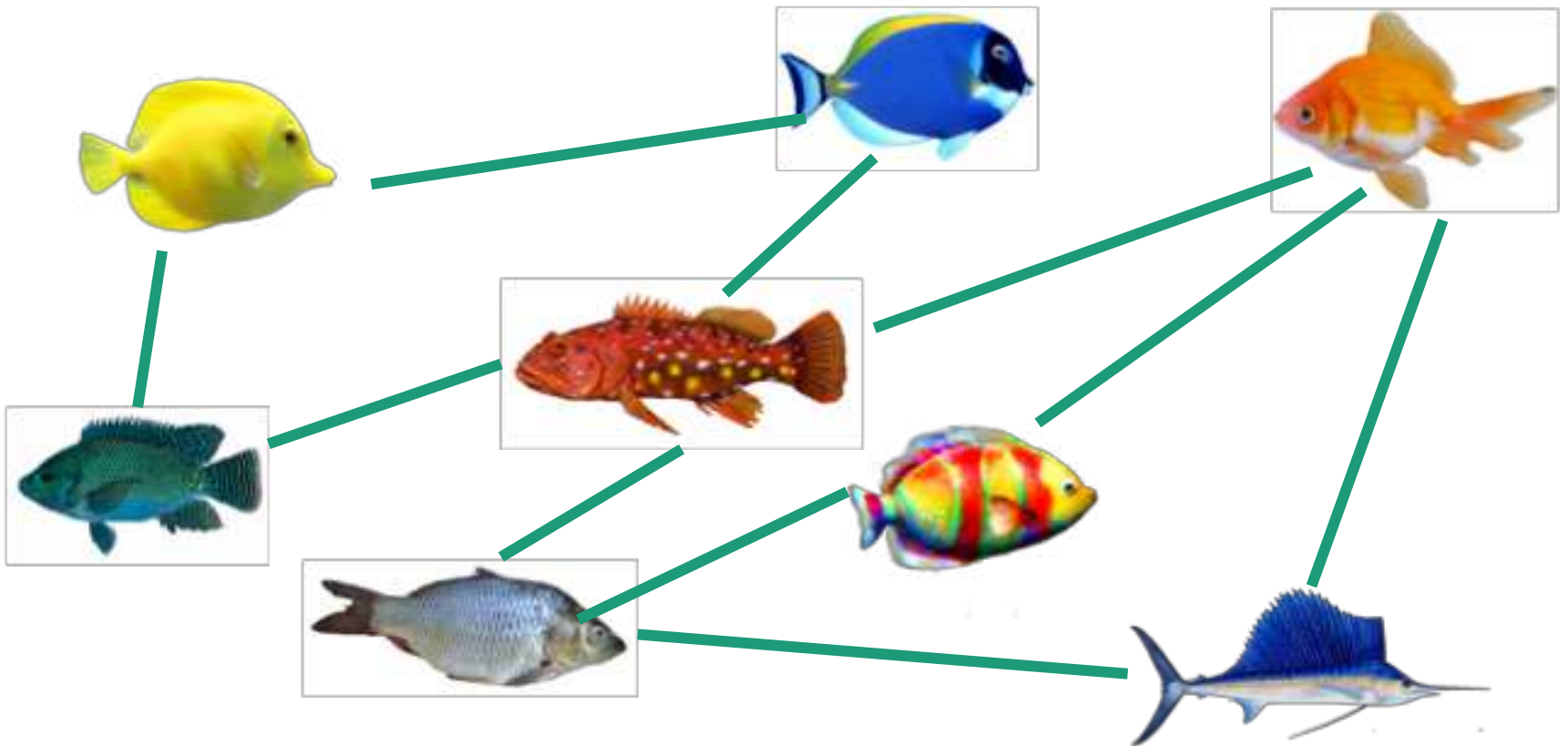
How about this one?



How about this one?

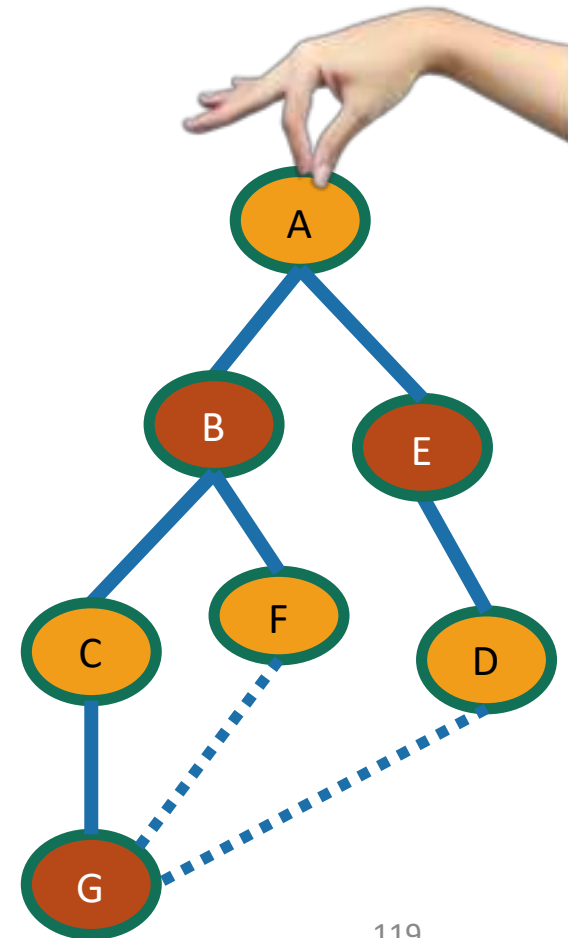


This one?



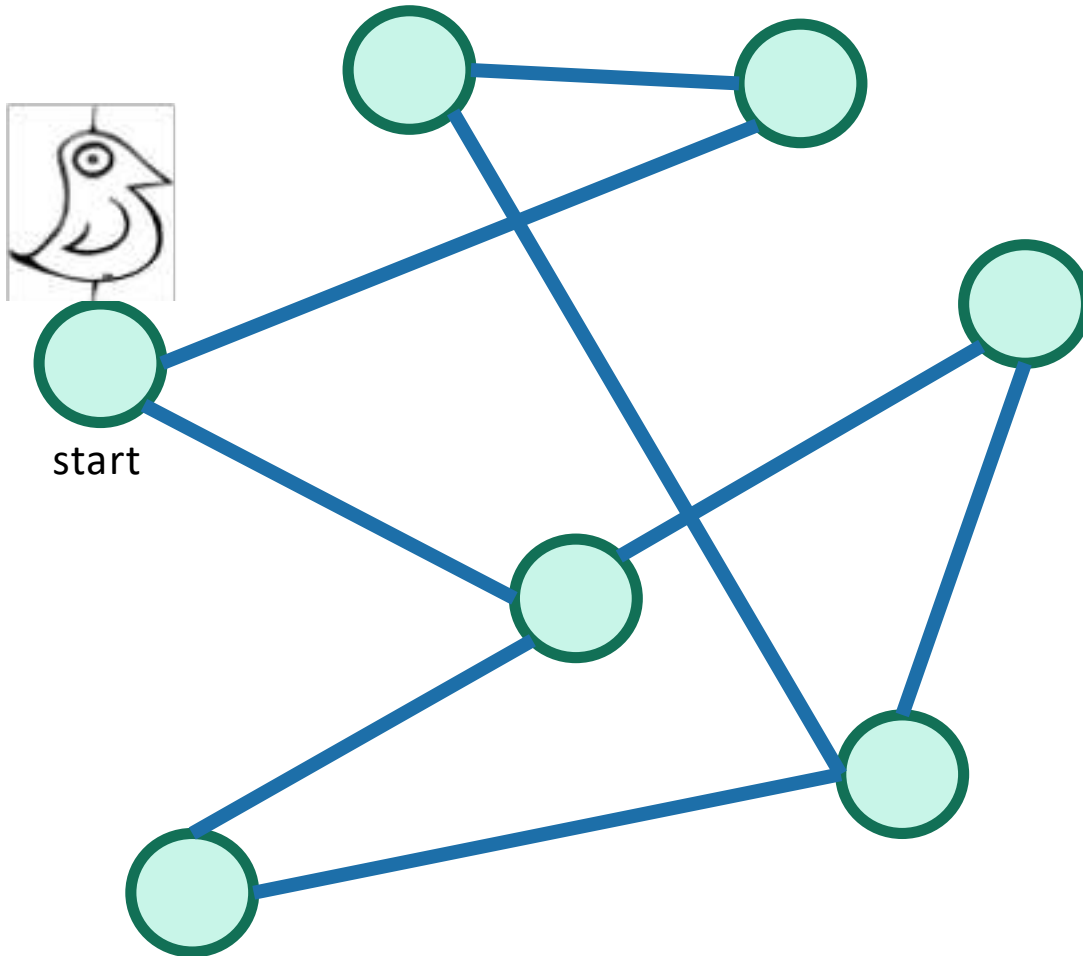
Application of BFS: Testing Bipartiteness

- Color the levels of the BFS tree in alternating colors.
- If you never color two connected nodes the same color, then it is bipartite.
- Otherwise, it's not.



Breadth-First Search

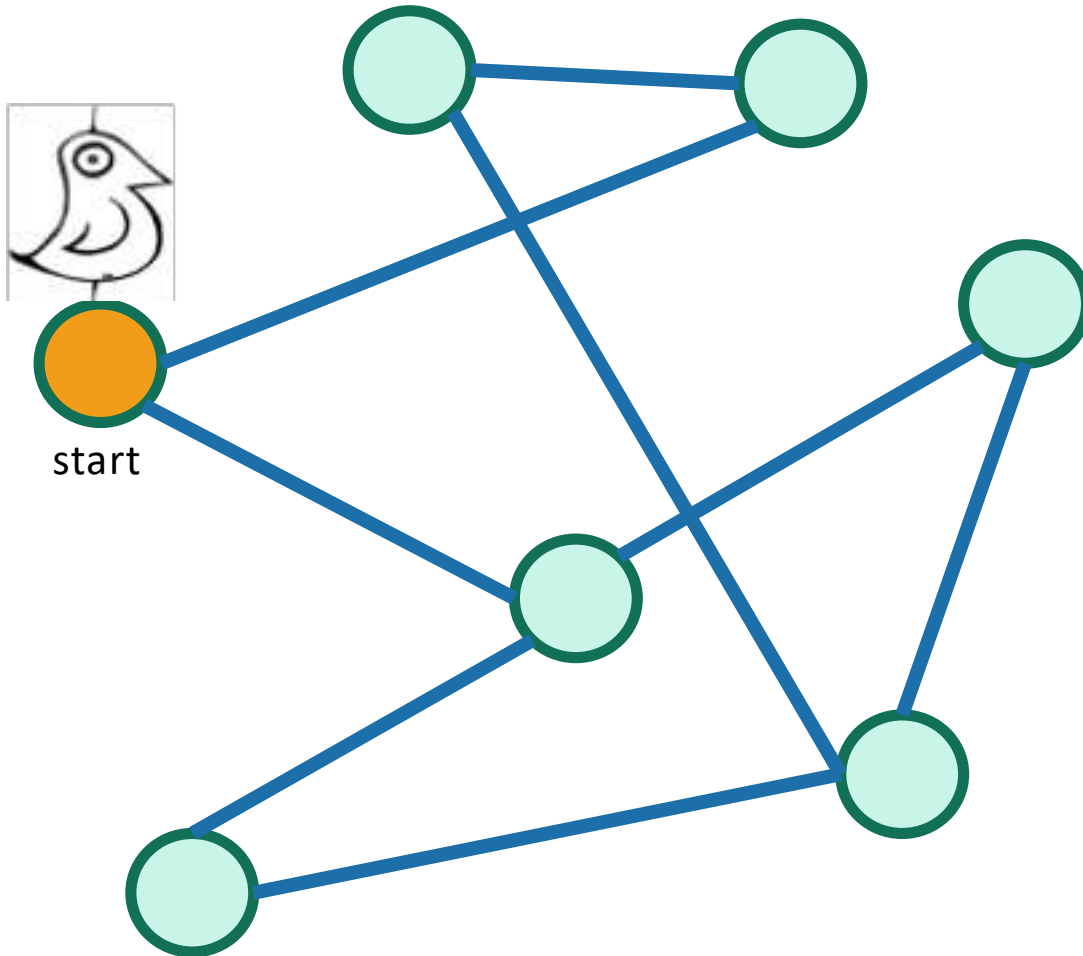
For testing bipartite-ness



-  Not been there yet
-  Can reach there in zero steps
-  Can reach there in one step
-  Can reach there in two steps
-  Can reach there in three steps

Breadth-First Search

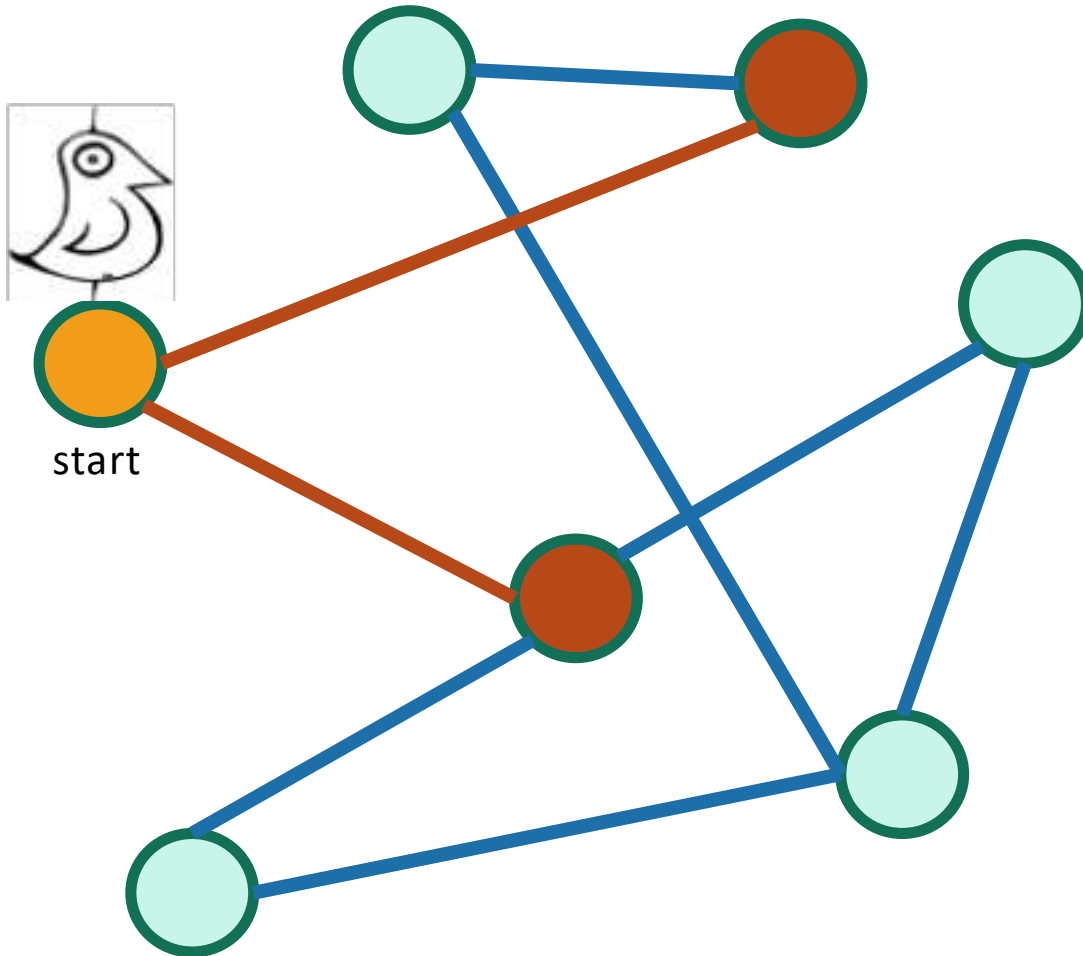
For testing bipartite-ness



- Not been there yet
- Can reach there in zero steps
- Can reach there in one step
- Can reach there in two steps
- Can reach there in three steps

Breadth-First Search

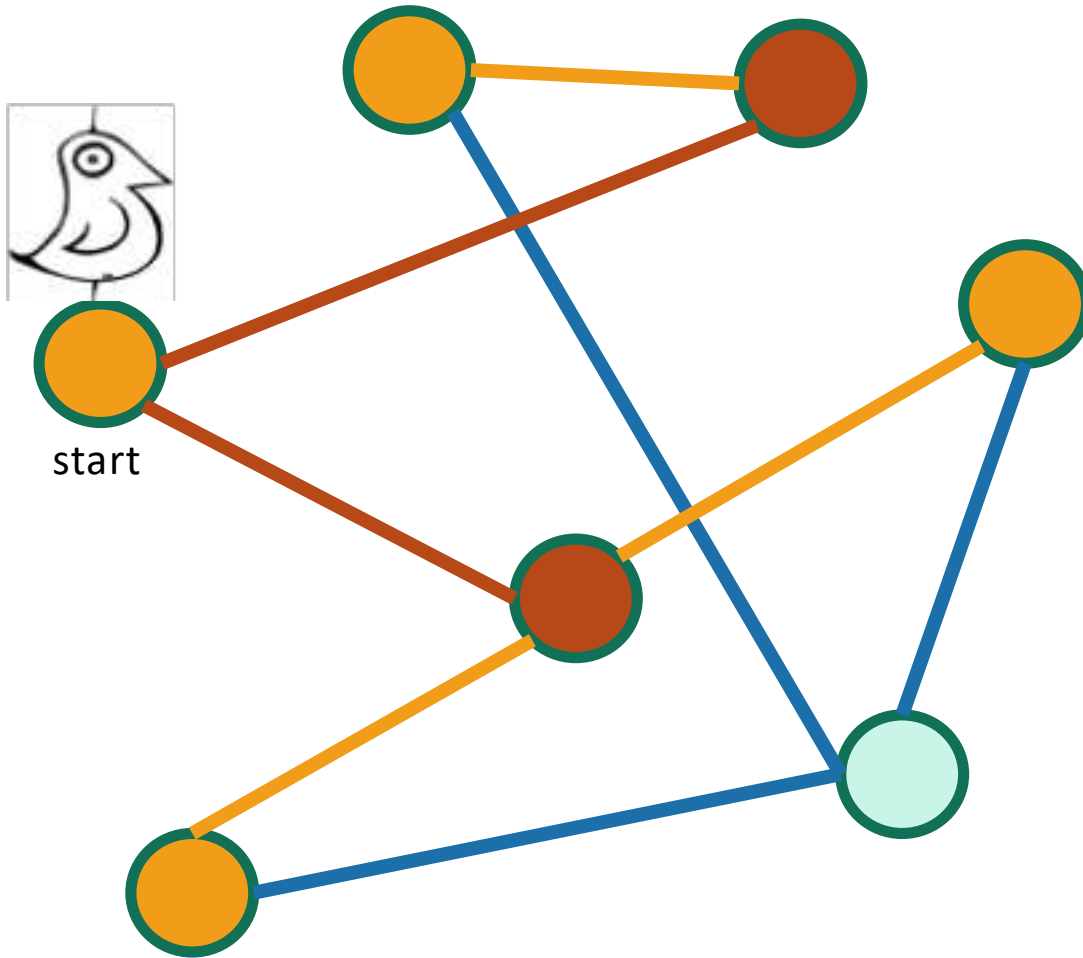
For testing bipartite-ness



- Not been there yet
- Can reach there in zero steps
- Can reach there in one step
- Can reach there in two steps
- Can reach there in three steps

Breadth-First Search

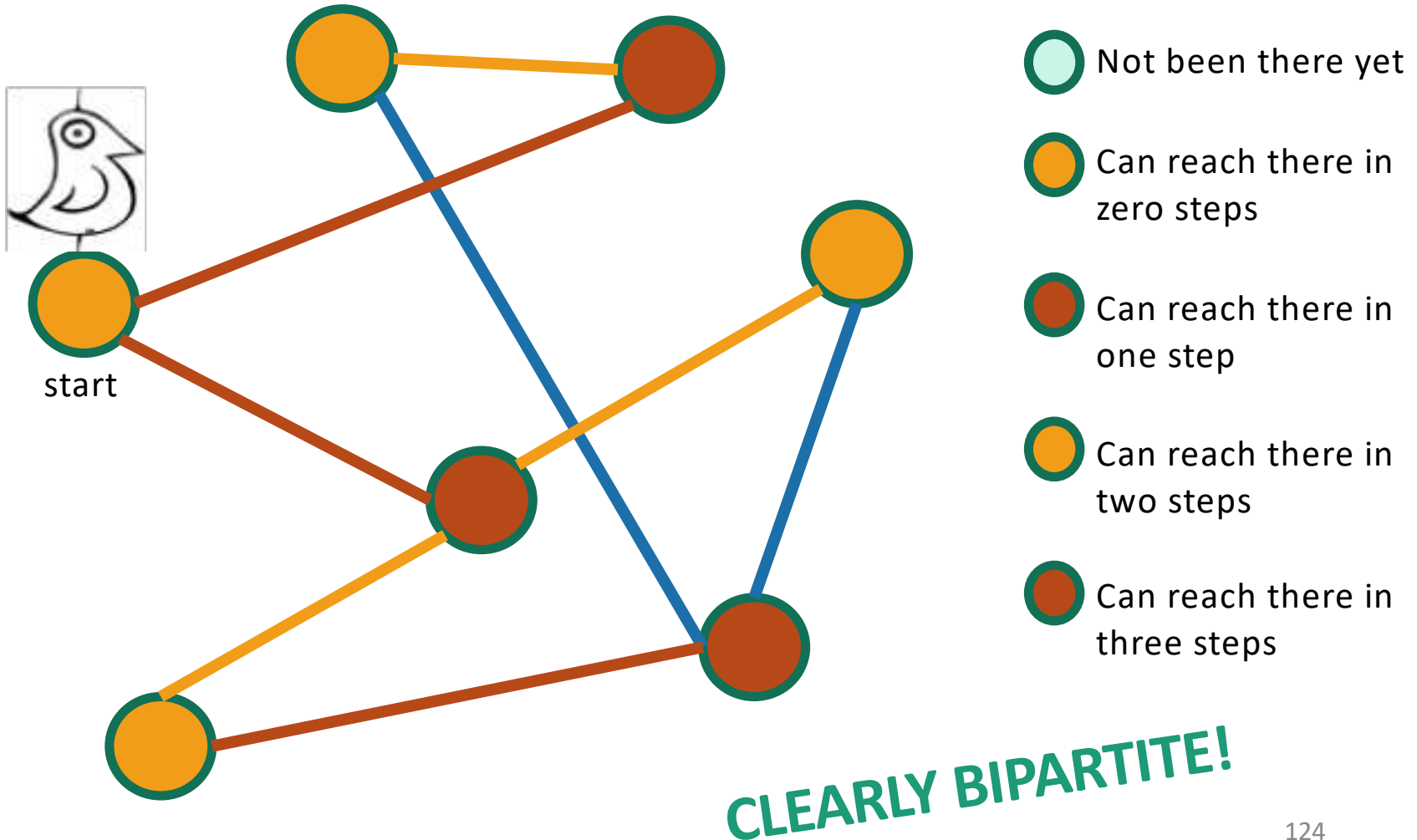
For testing bipartite-ness



-  Not been there yet
-  Can reach there in zero steps
-  Can reach there in one step
-  Can reach there in two steps
-  Can reach there in three steps

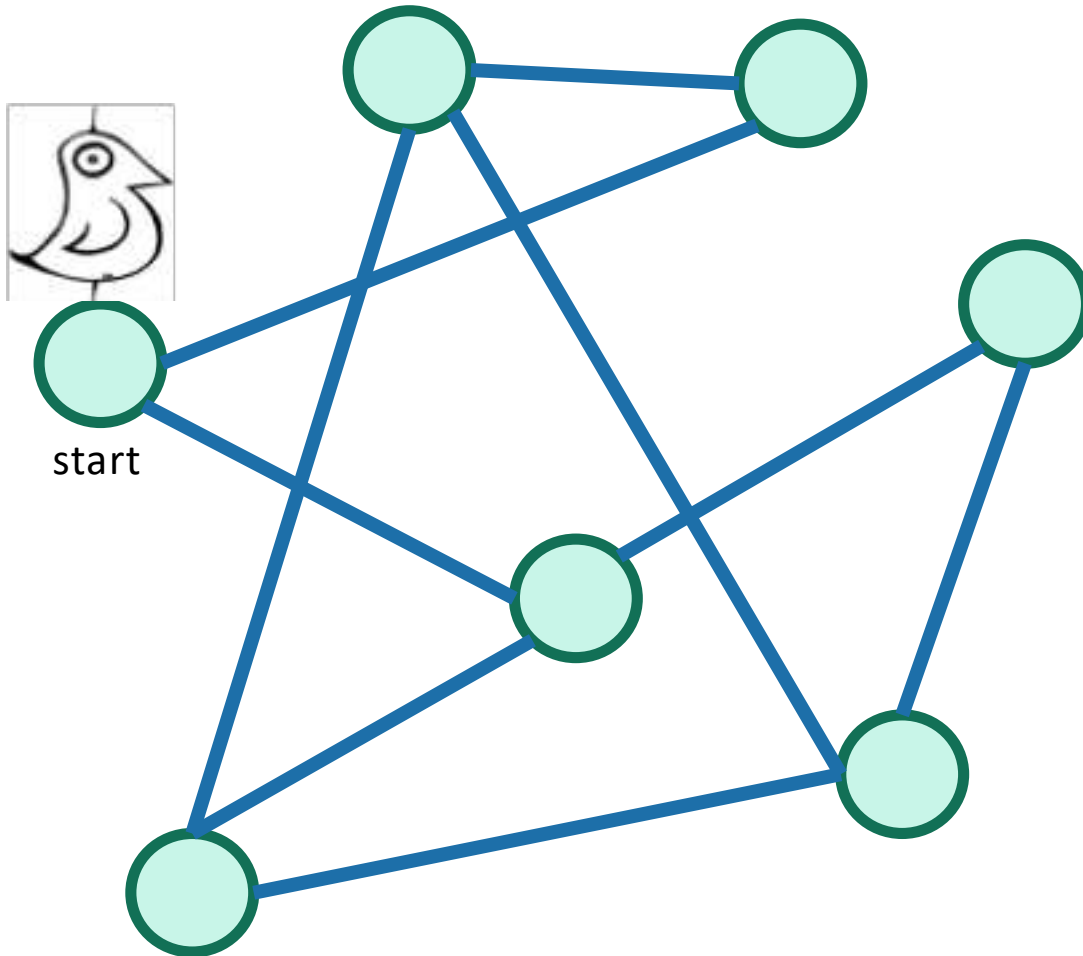
Breadth-First Search

For testing bipartite-ness



Breadth-First Search

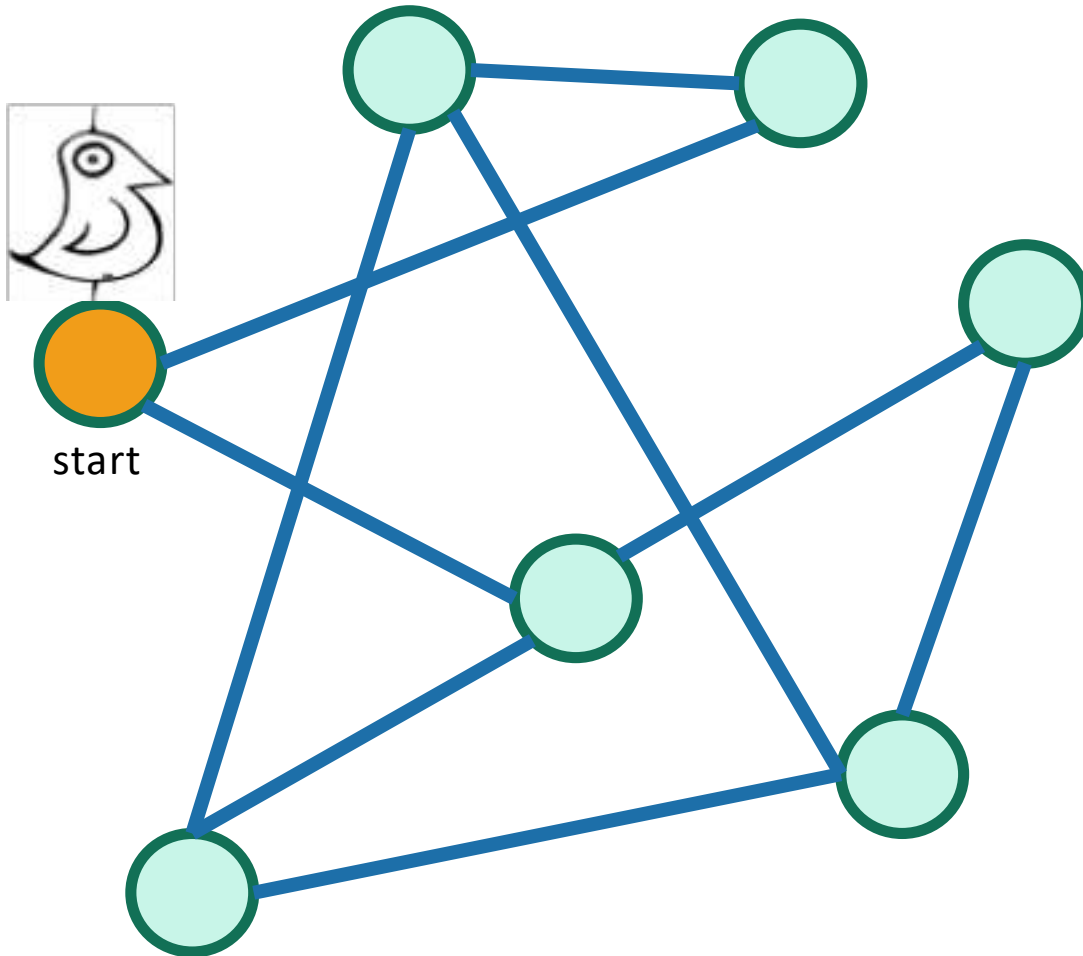
For testing bipartite-ness



-  Not been there yet
-  Can reach there in zero steps
-  Can reach there in one step
-  Can reach there in two steps
-  Can reach there in three steps

Breadth-First Search

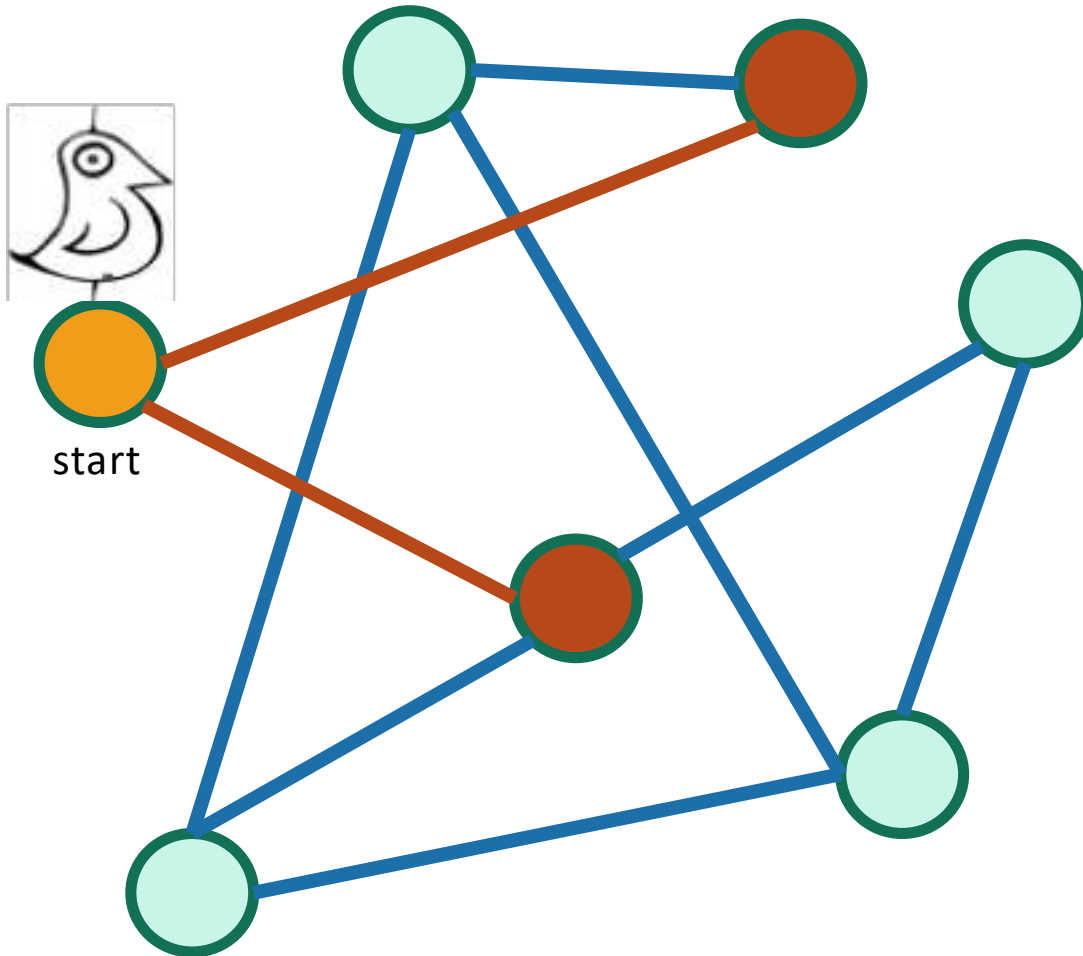
For testing bipartite-ness



- Not been there yet
- Can reach there in zero steps
- Can reach there in one step
- Can reach there in two steps
- Can reach there in three steps

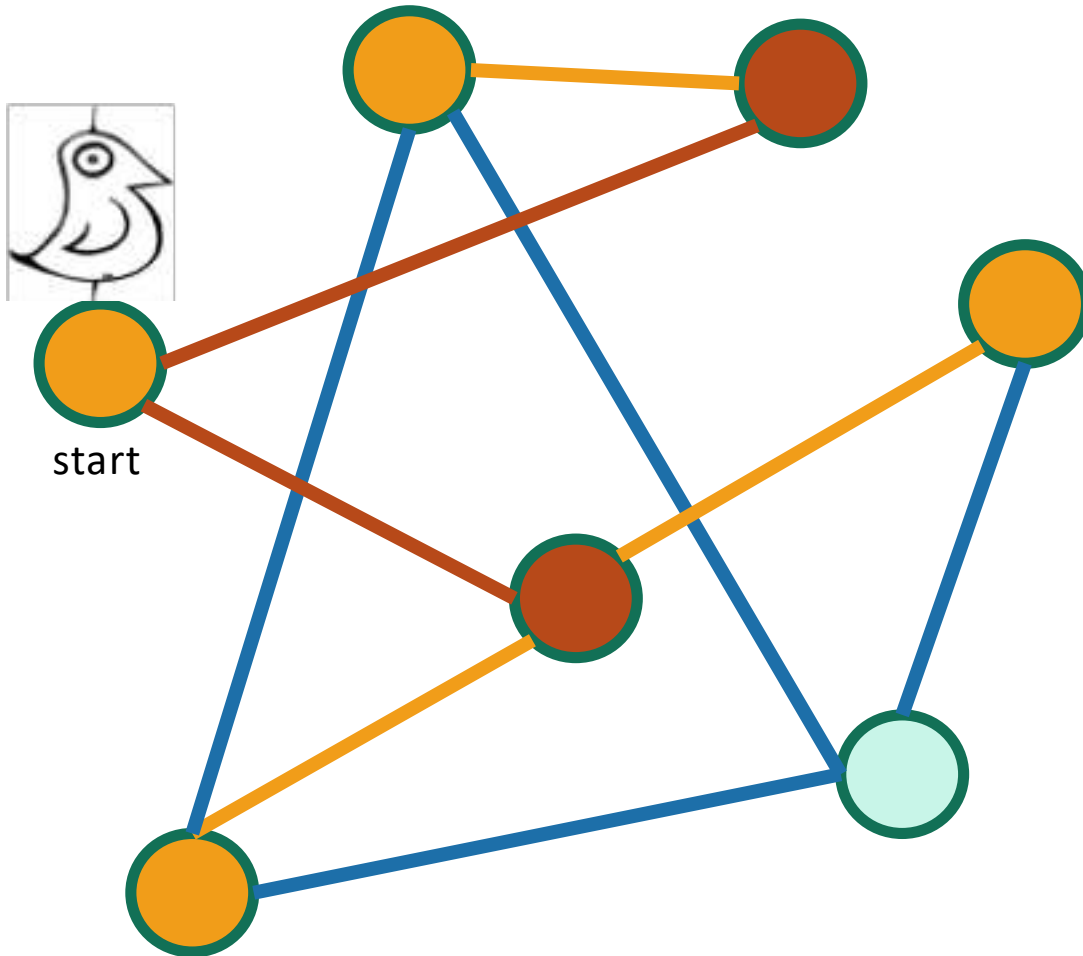
Breadth-First Search

For testing bipartite-ness



Breadth-First Search

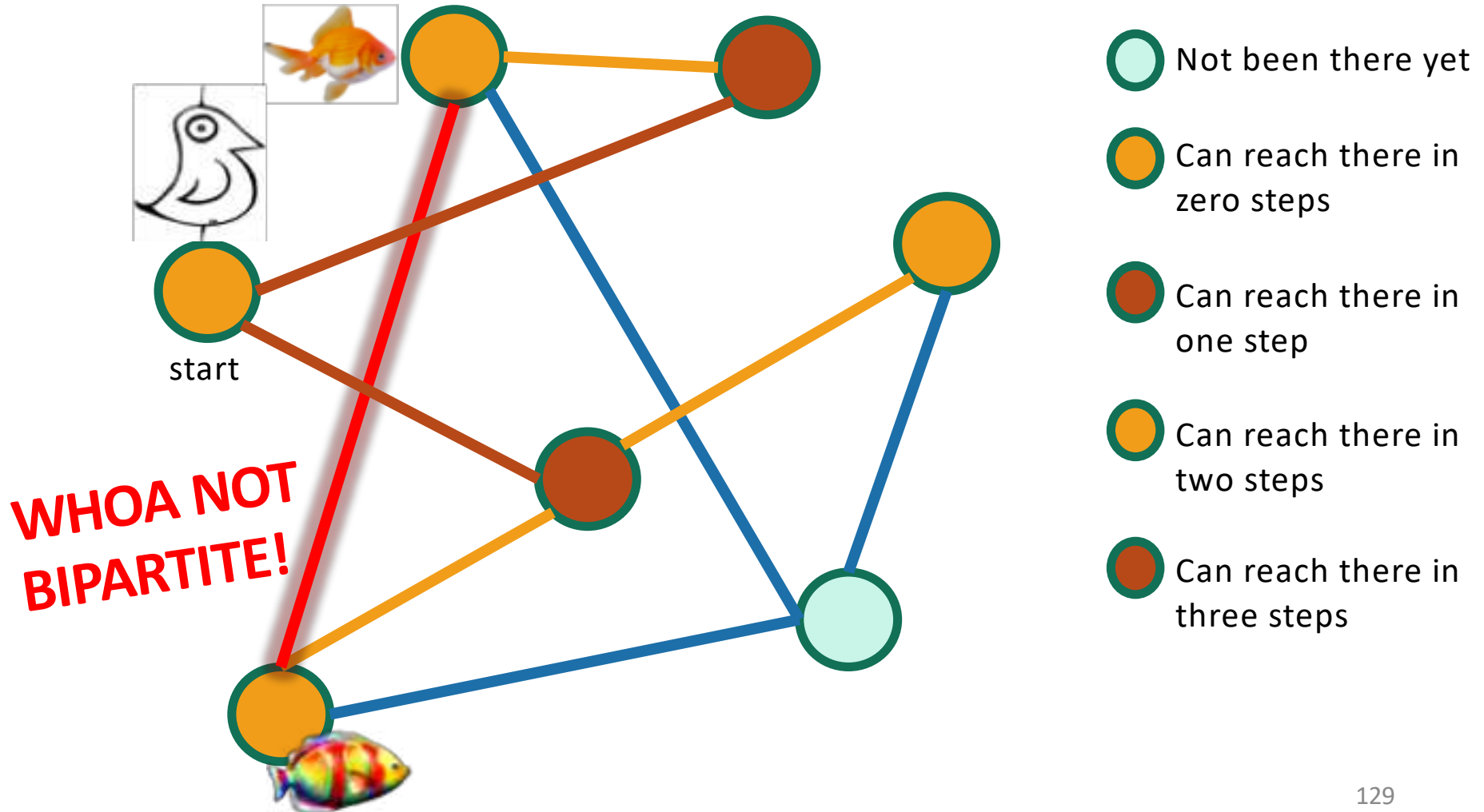
For testing bipartite-ness



- Not been there yet
- Can reach there in zero steps
- Can reach there in one step
- Can reach there in two steps
- Can reach there in three steps

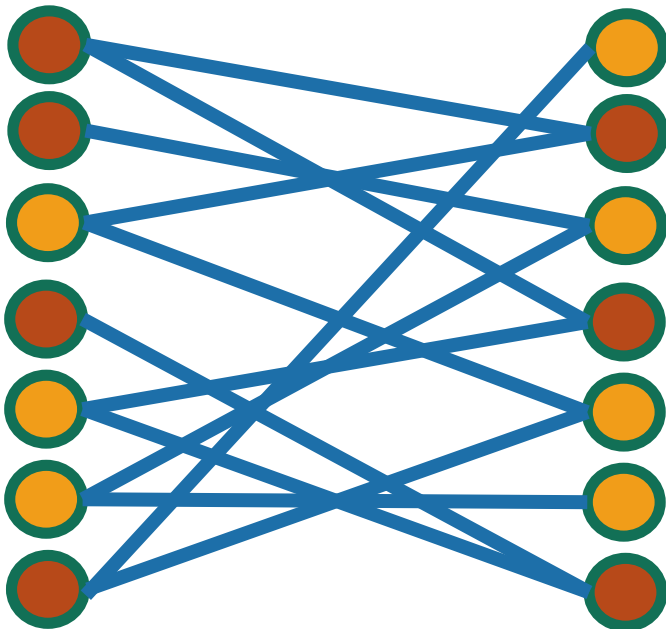
Breadth-First Search

For testing bipartite-ness



Hang on now.

- Just because **this** coloring doesn't work, why does that mean that there is **no** coloring that works?



I can come up
with plenty of bad
colorings on this
legitimately
bipartite graph...



Plucky the
pedantic penguin

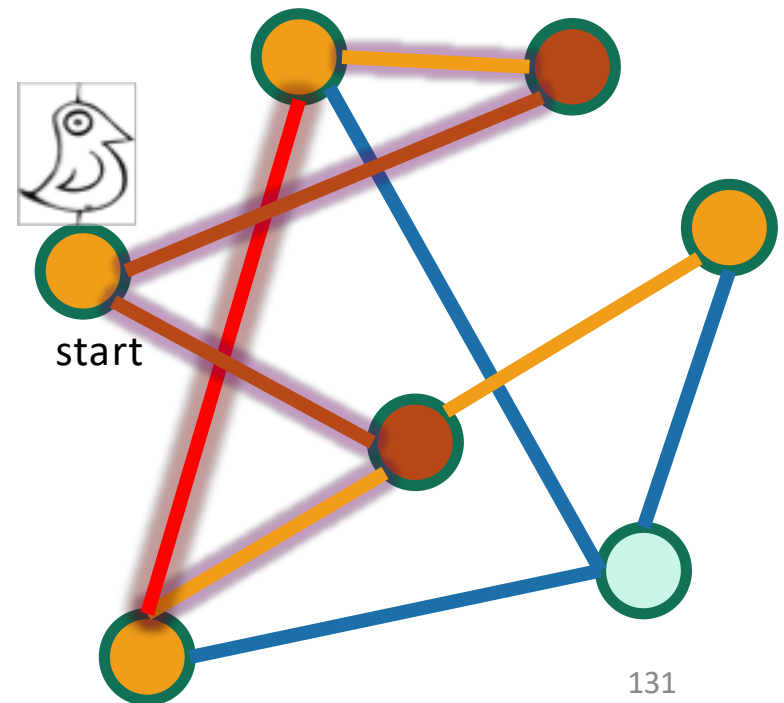
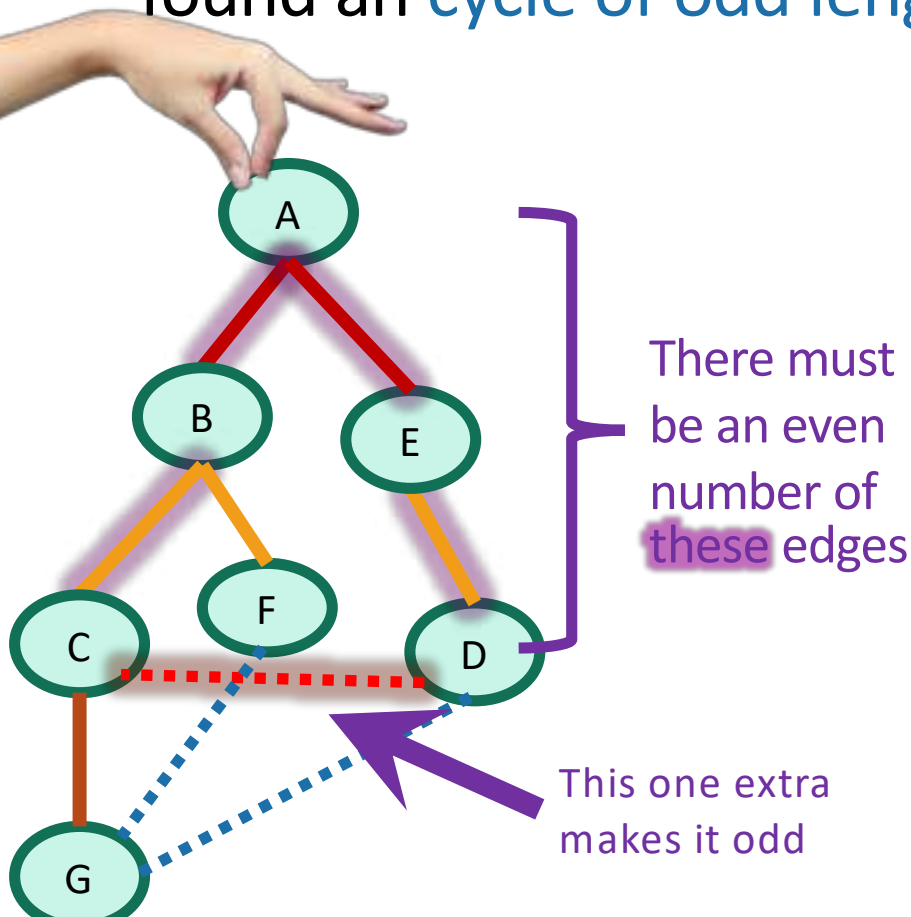
Make this proof
sketch formal!



Ollie the over-achieving ostrich

Some proof required

- If BFS colors two neighbors the same color, then it's found an **cycle of odd length** in the graph.



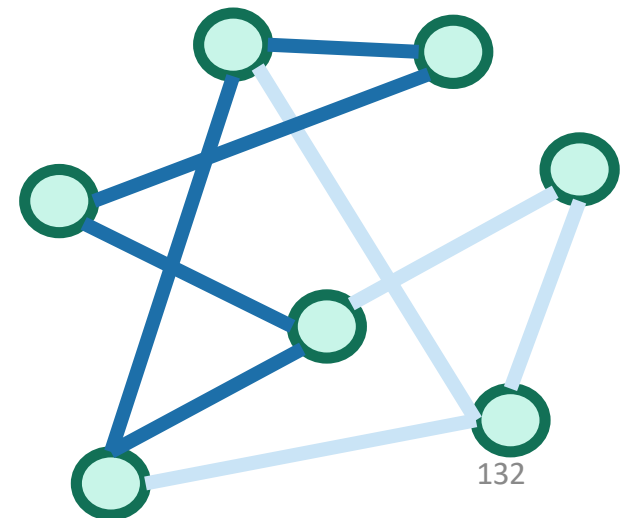
Make this proof
sketch formal!



Ollie the over-achieving ostrich

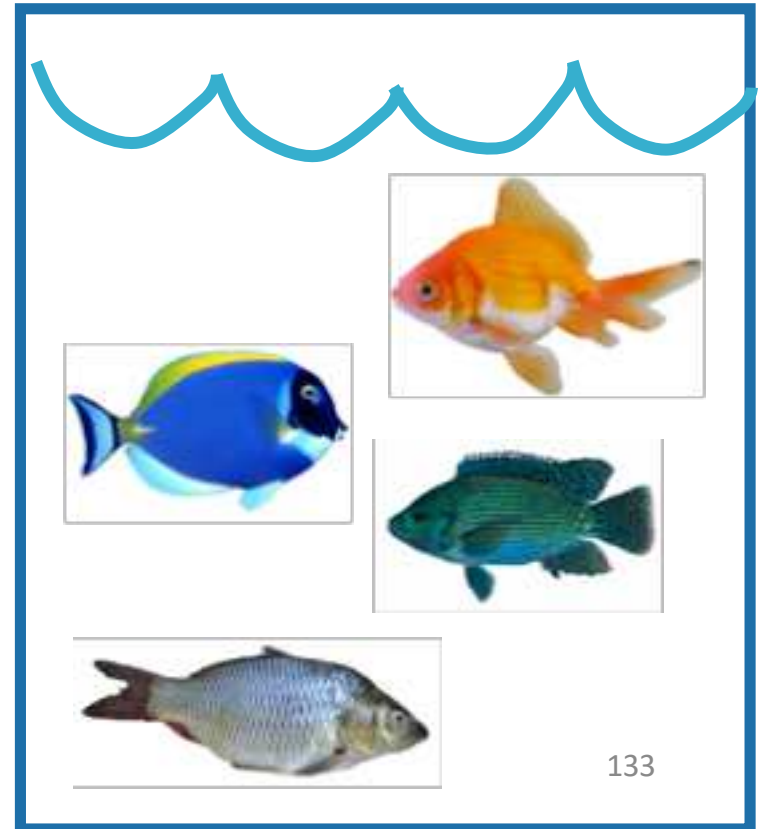
Some proof required

- If BFS colors two neighbors the same color, then it's found an **cycle of odd length** in the graph.
- But you can **never** color an odd cycle with two colors so that no two neighbors have the same color.
 - [Fun exercise!]
- So you can't legitimately color the whole graph either.
- **Thus it's not bipartite.**



What have we learned?

BFS can be used to detect bipartite-ness in time $O(n + m)$.



Outline

- Part 0: Graphs and terminology
- Part 1: Depth-first search
 - Application: topological sorting
 - Application: in-order traversal of BSTs
- Part 2: Breadth-first search
 - Application: shortest paths
 - Application (if time): is a graph bipartite?

Recap



Recap

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

Still open (next few classes)

- 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?
- What is Samuel L. Jackson's Erdos number?
 - (Or, what if I want everyone's everyone-else number?)

Next Time

- Strongly Connected Components

Before Next Time

- Pre-lecture exercise: Strongly Connected What-Now?