

# Graphs, Priority Queues

Quote of the Week: “As I walked out the door toward the gate that would lead to my freedom, I knew if I didn't leave my bitterness and hatred behind, I'd still be in prison.”

# Project 2 group evaluations

- ◆ They're due Thursday, with Monday / Tuesday lab
- ◆ **You will get a 0 on the project unless you complete this**
- ◆ Please be honest and fair. These may affect your group members' scores

# Midterm 2 on Friday

- ◆ Same time, place
  - ▶ 7 - 9 pm
  - ▶ Sections 101 - 103: 144 Dwinelle
  - ▶ Sections 104 - 109: 155 Dwinelle
- ◆ Cheat sheet: one 8.5 x 11 sheet, **two sides**

# Project 3 released on Friday

- ◆ Also a 3 - 4 group project
- ◆ Project is about **speed** (actual time, not theoretical)

# Project 3

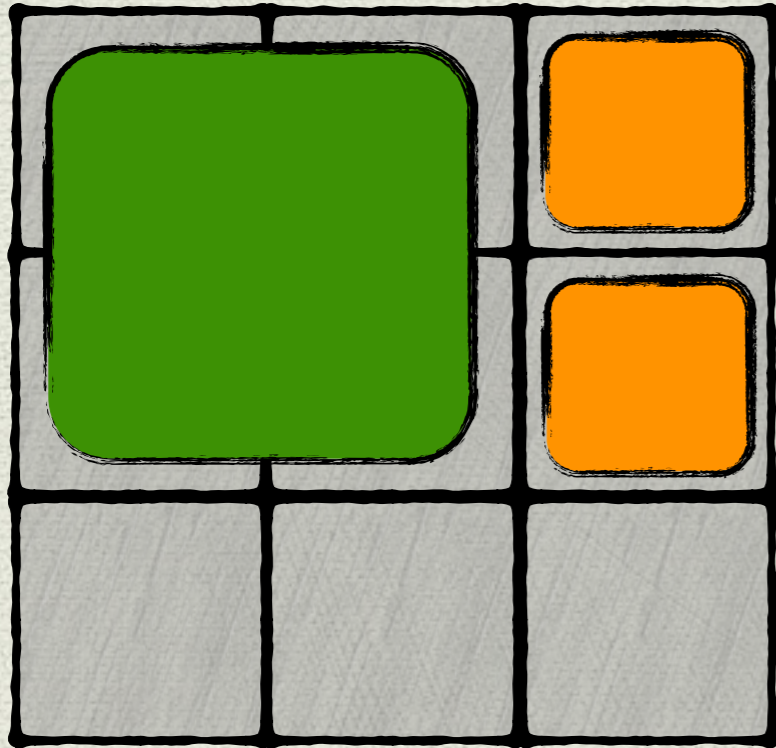
- ◆ Write a program to solve puzzles like this:



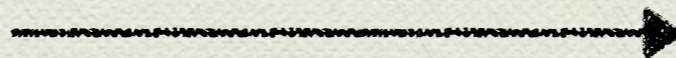
Credit: <http://magicpuzzles.org/>

# Your project 3

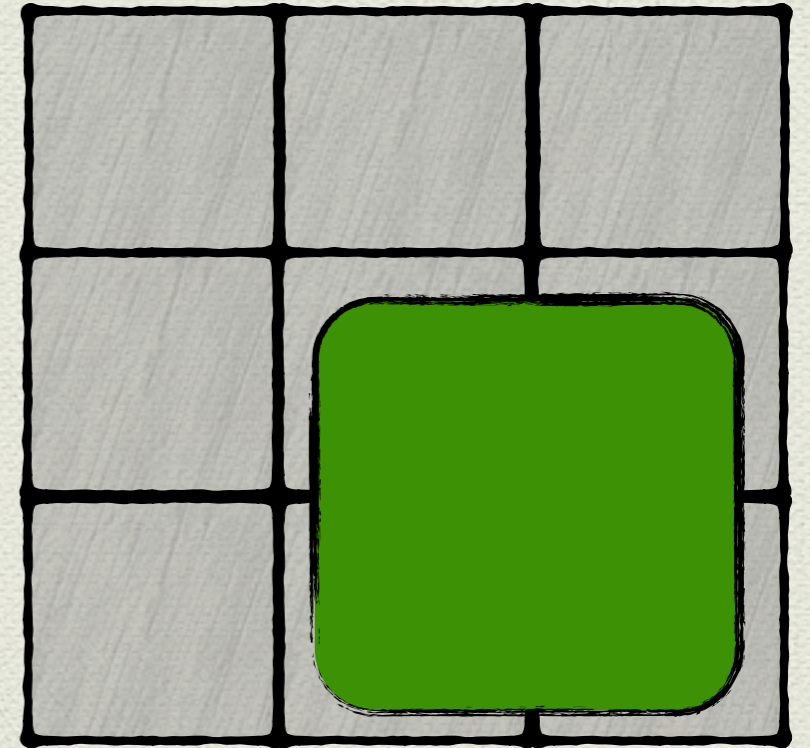
- ◆ Motivated by a simple problem: How to figure out the steps required to solve a puzzle like this?



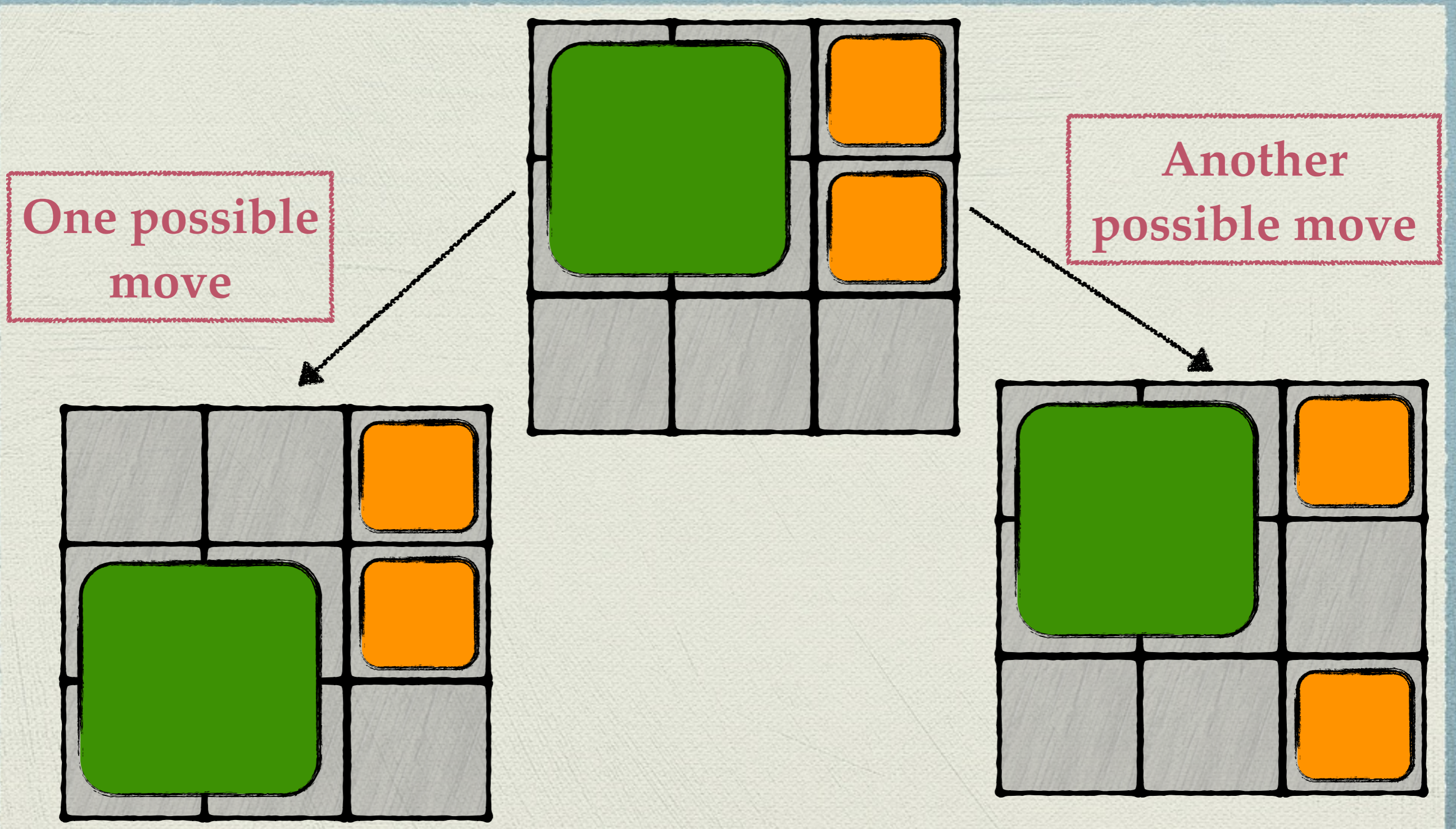
A tray with  
blocks you can  
slide around



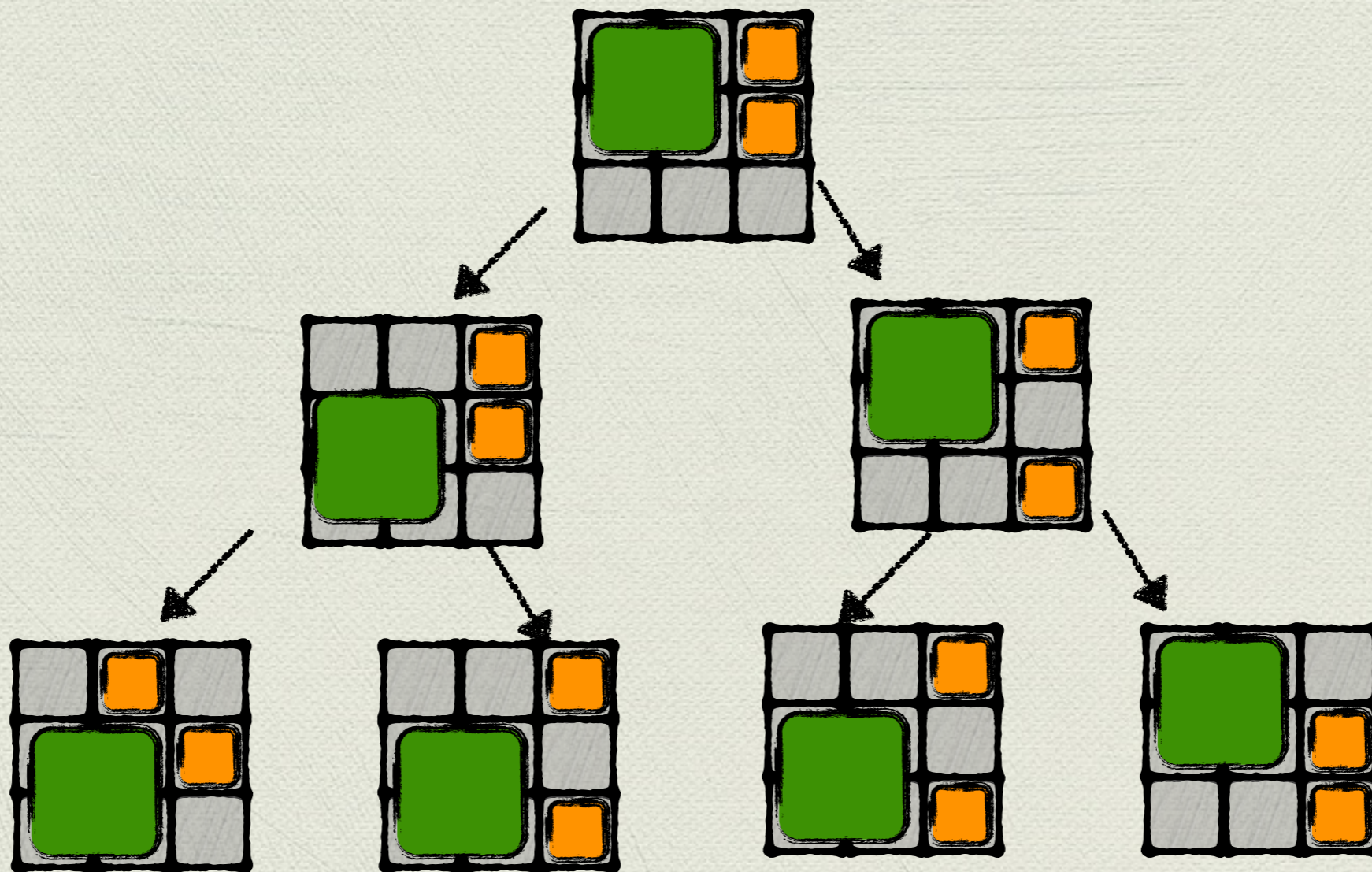
Goal: move green  
block to bottom-  
right corner



# Visualizing the problem

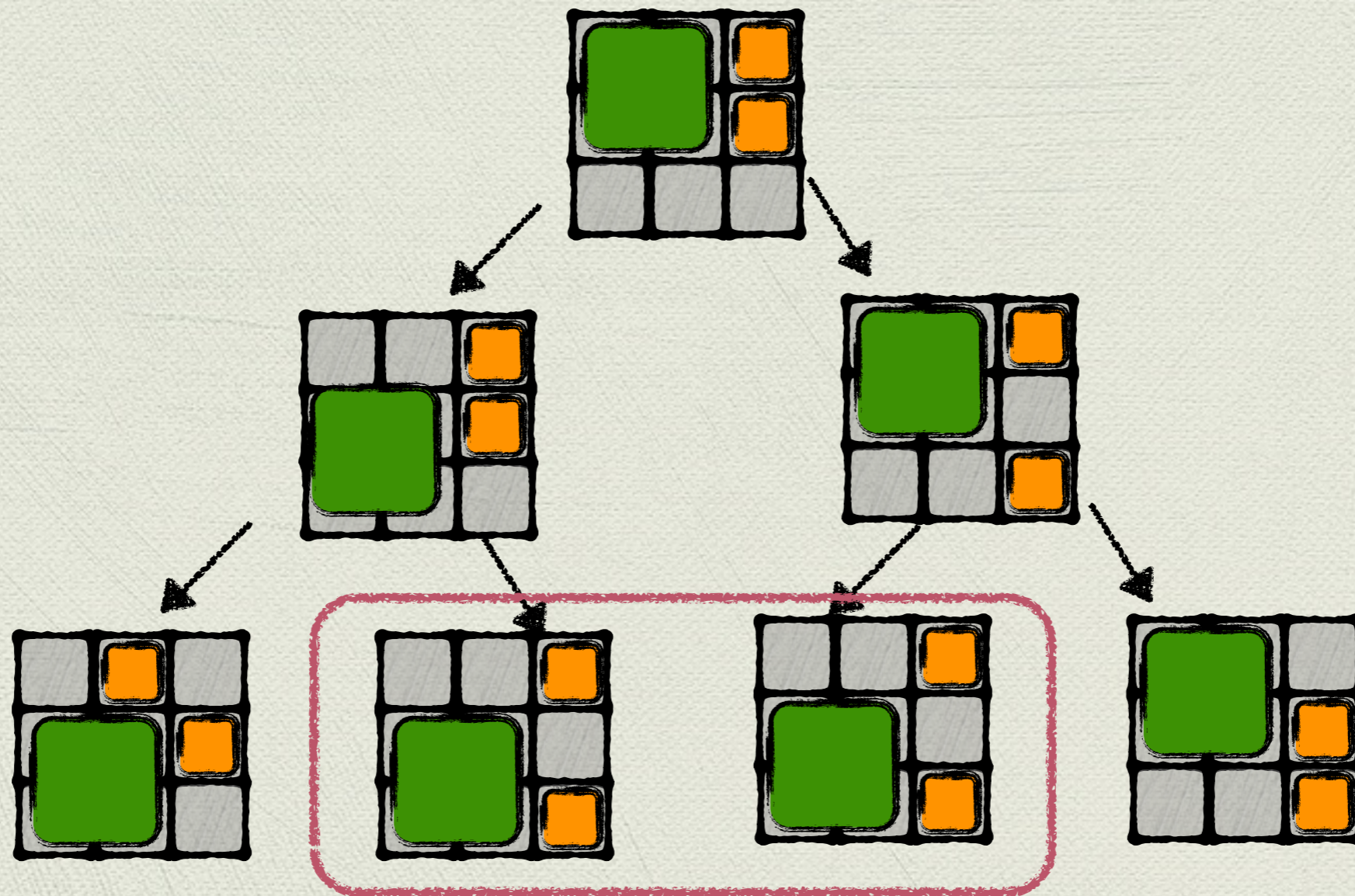


It looks kinda like a tree



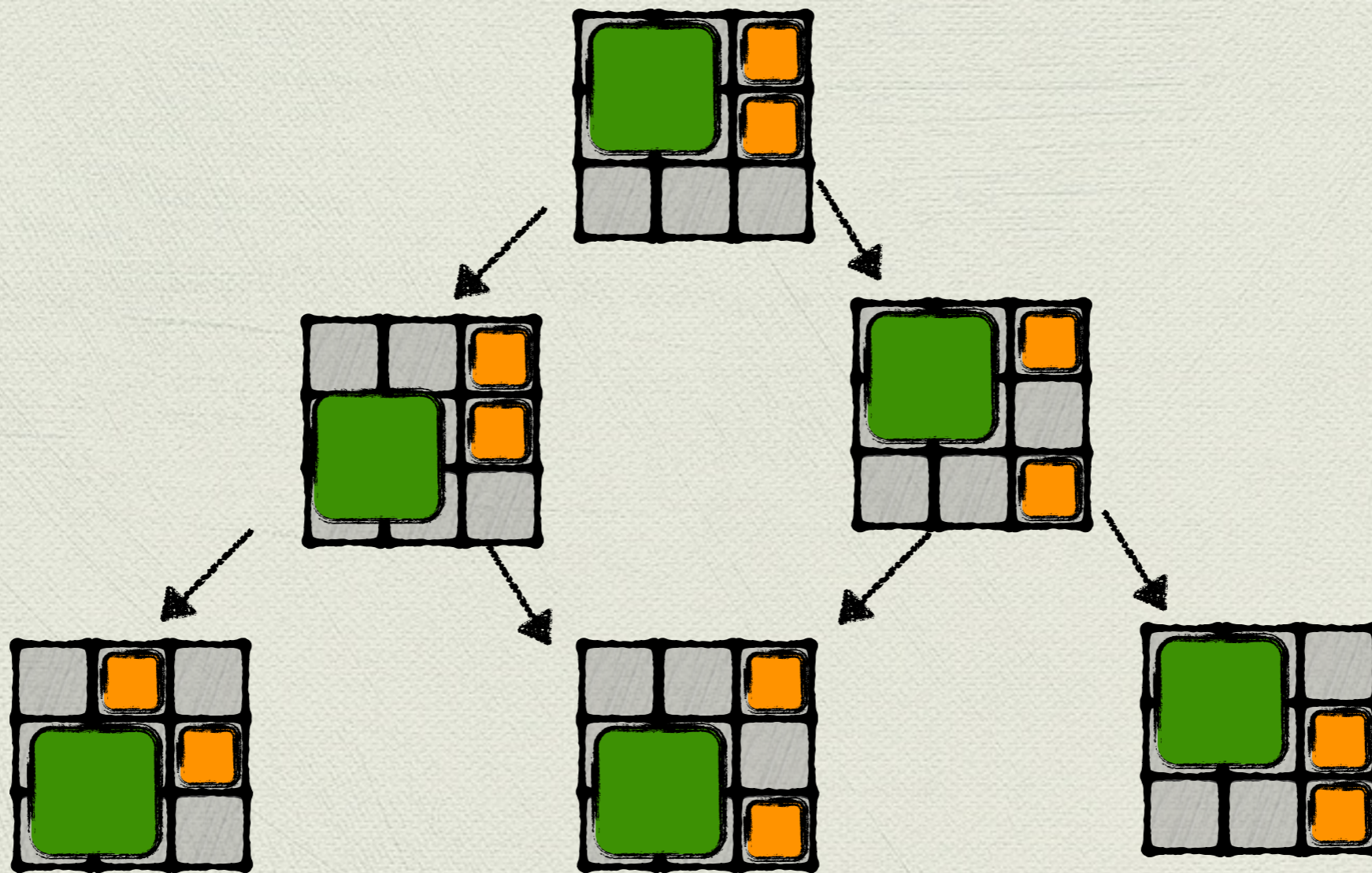


# But wait...!



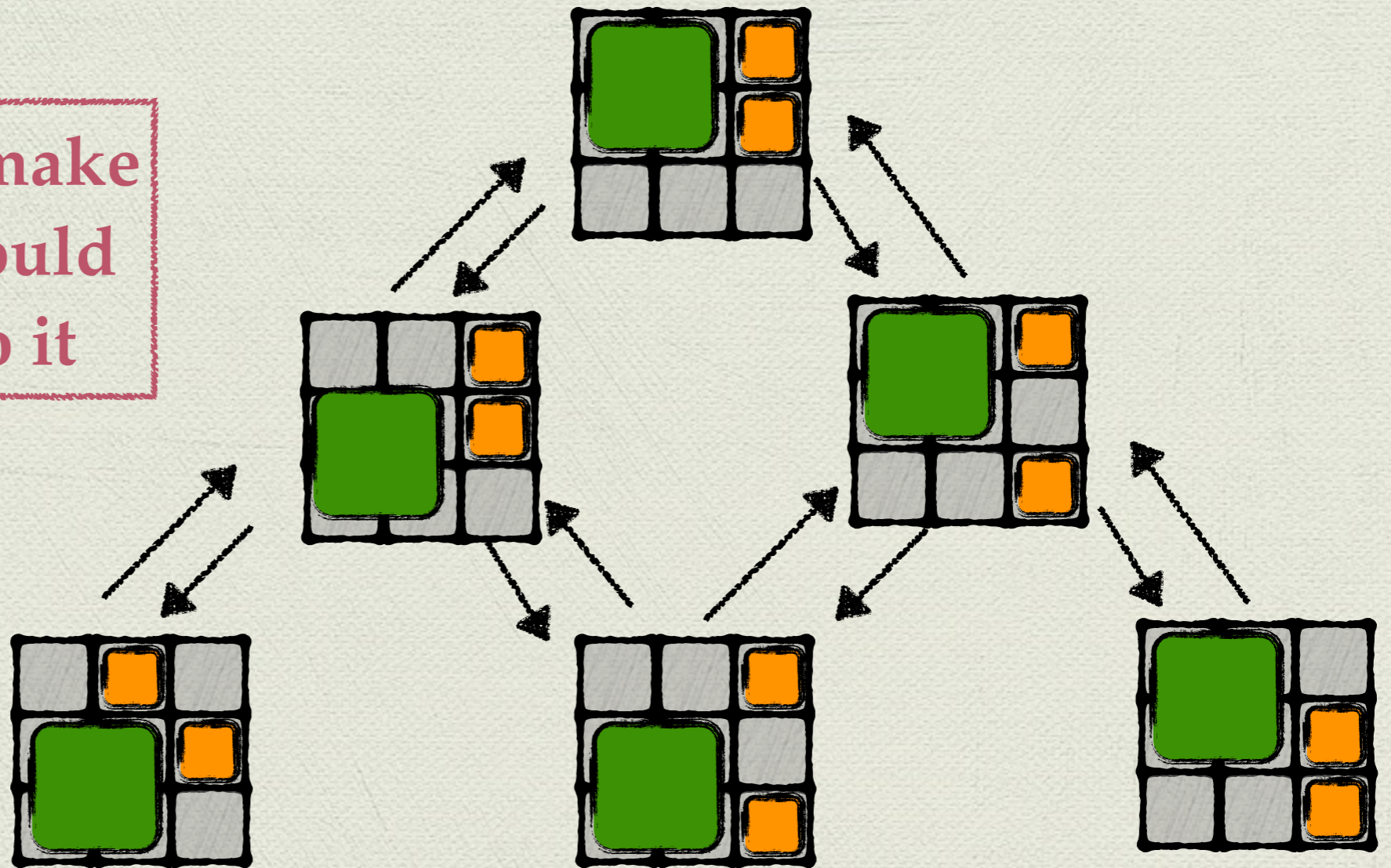
The same thing!

But wait...!



# Also

Any time we make  
a move, we could  
always undo it



# It looks kinda like a tree

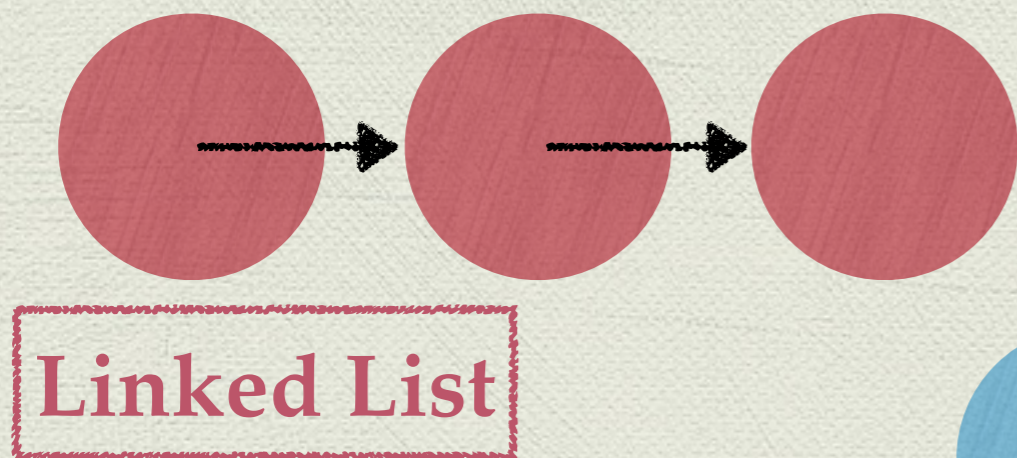
- ◆ But it violates the rules of trees
  - ▶ No edges point back up the tree
  - ▶ No node is descended from two nodes

# So it's not a tree

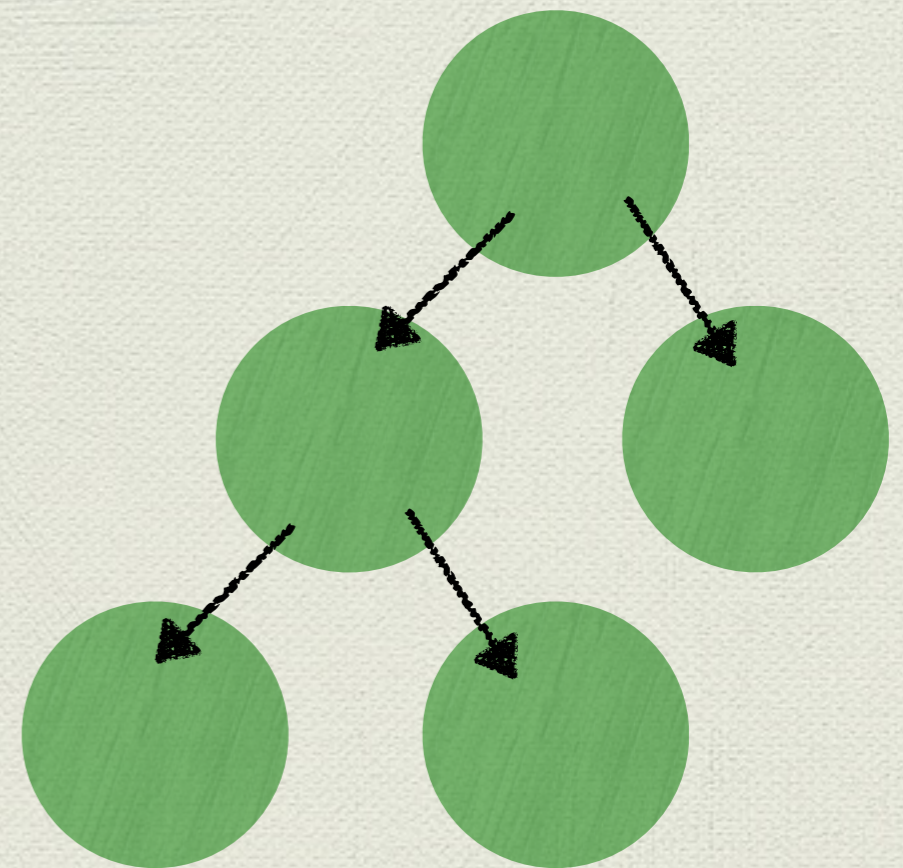
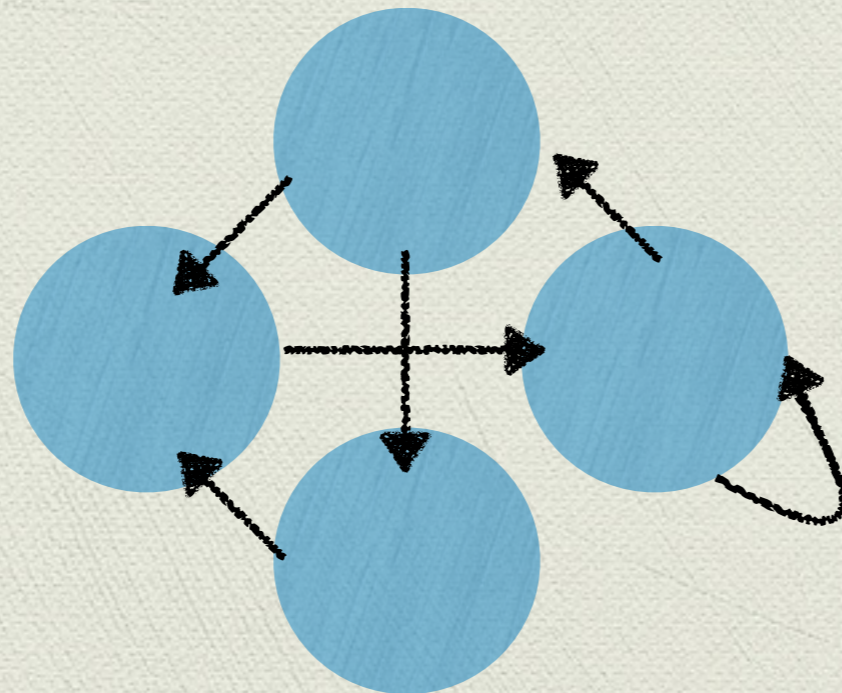
- ◆ We call it a graph

# Graphs

- ◆ A graph is a collection of nodes that can be connected in any which way



Graph



Tree

# Graph Traversals

# Sliding block puzzle

- ◆ To solve this problem, we must find a **path** through the graph from our initial tray to our goal tray
- ◆ Essentially, this boils down to iterating through our graph, starting from the initial tray, until we come across the goal tray



# Graph traversal

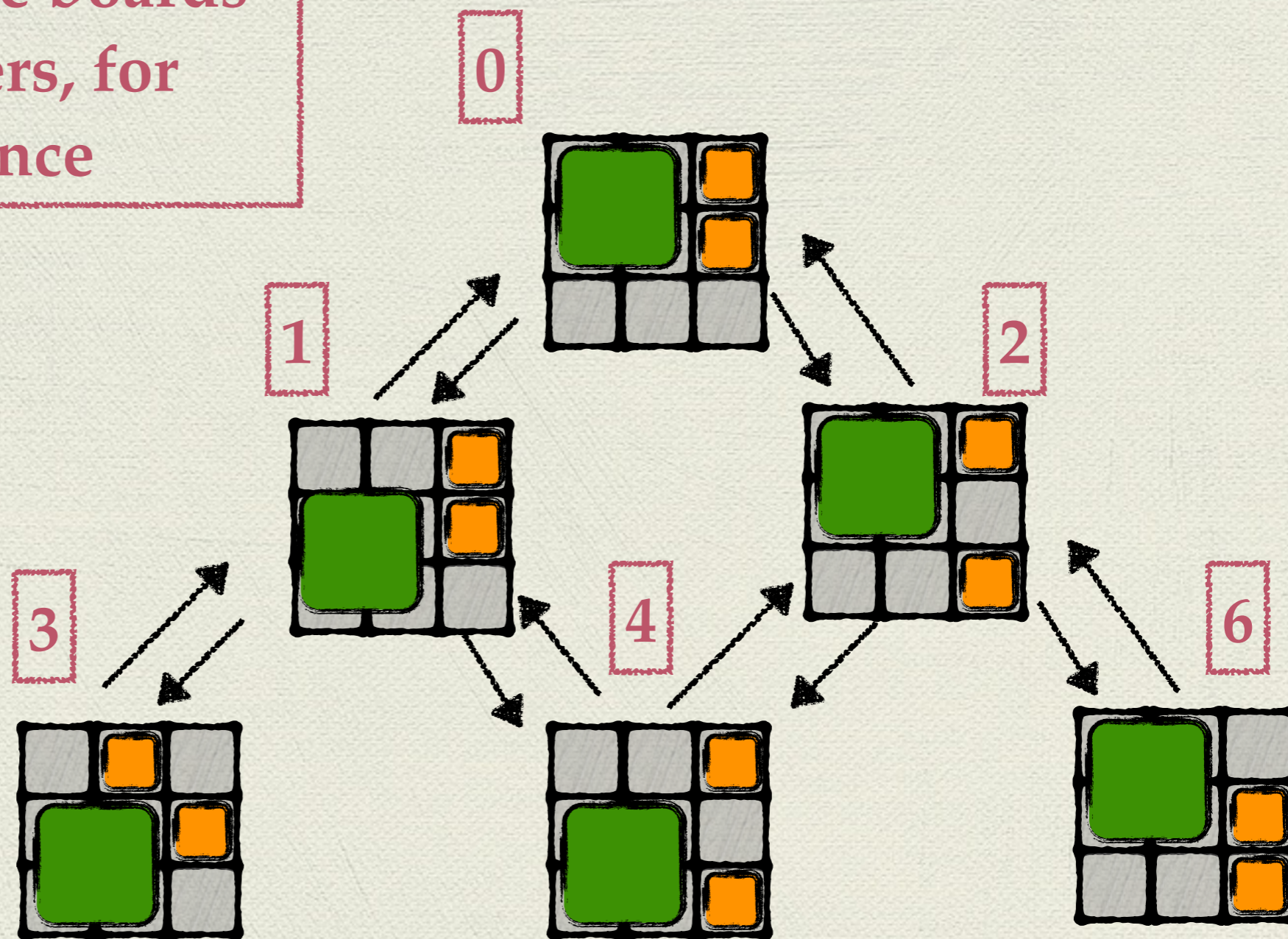
- ◆ How do we iterate over the nodes of a graph?
  - ▶ A graph isn't much different from a tree, so let's try tree traversal!

# Traversing a graph like a tree

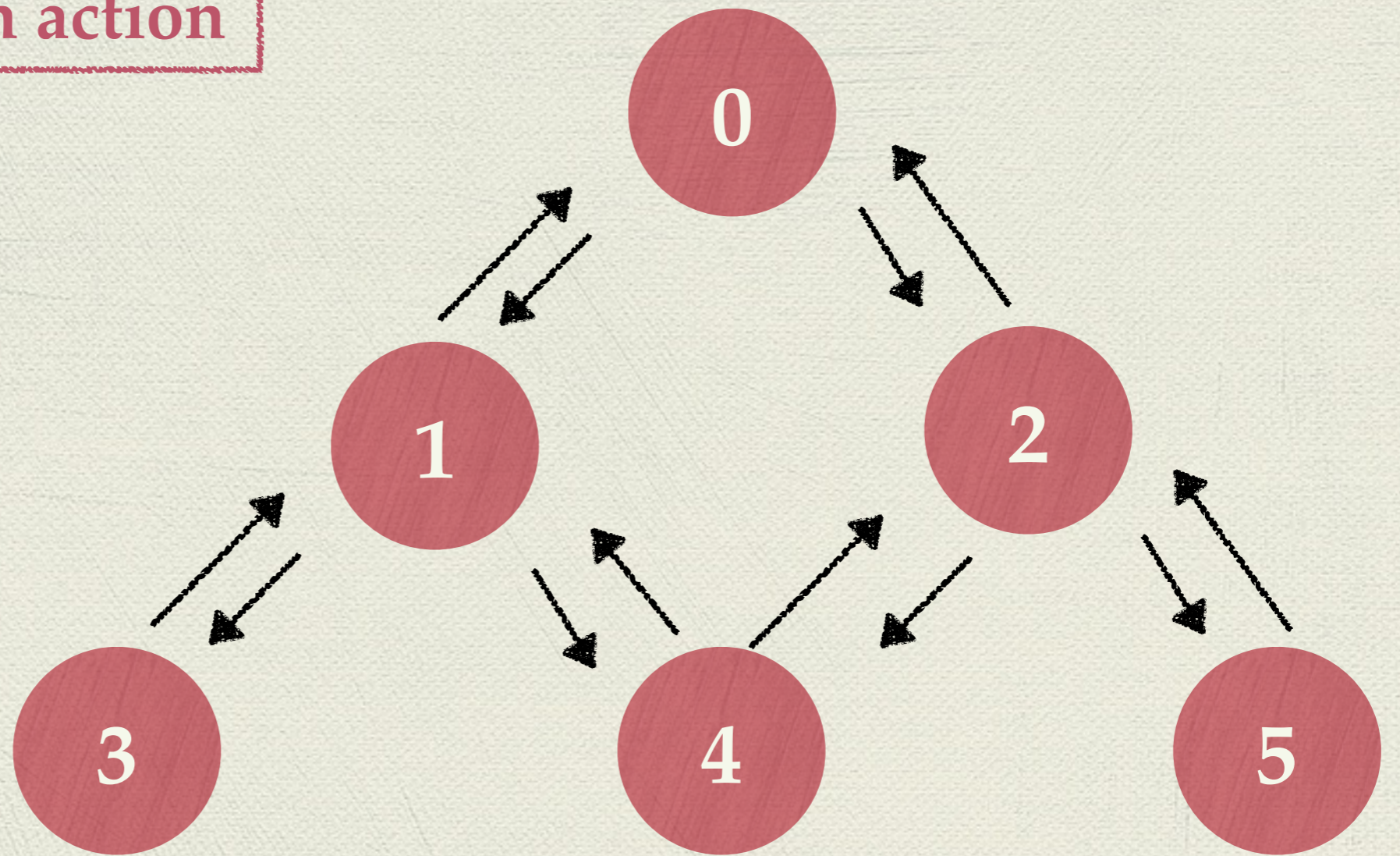
◆ Kinda works...?

```
Stack<Tray> fringe = new Stack<>();
fringe.push(initialTray);
while (!fringe.isEmpty()) {
    Tray currentTray = fringe.pop();
    // do stuff
    for (Tray t : currentTray.nextTrays()) {
        fringe.push(t);
    }
}
```

I've labeled the boards with numbers, for convenience

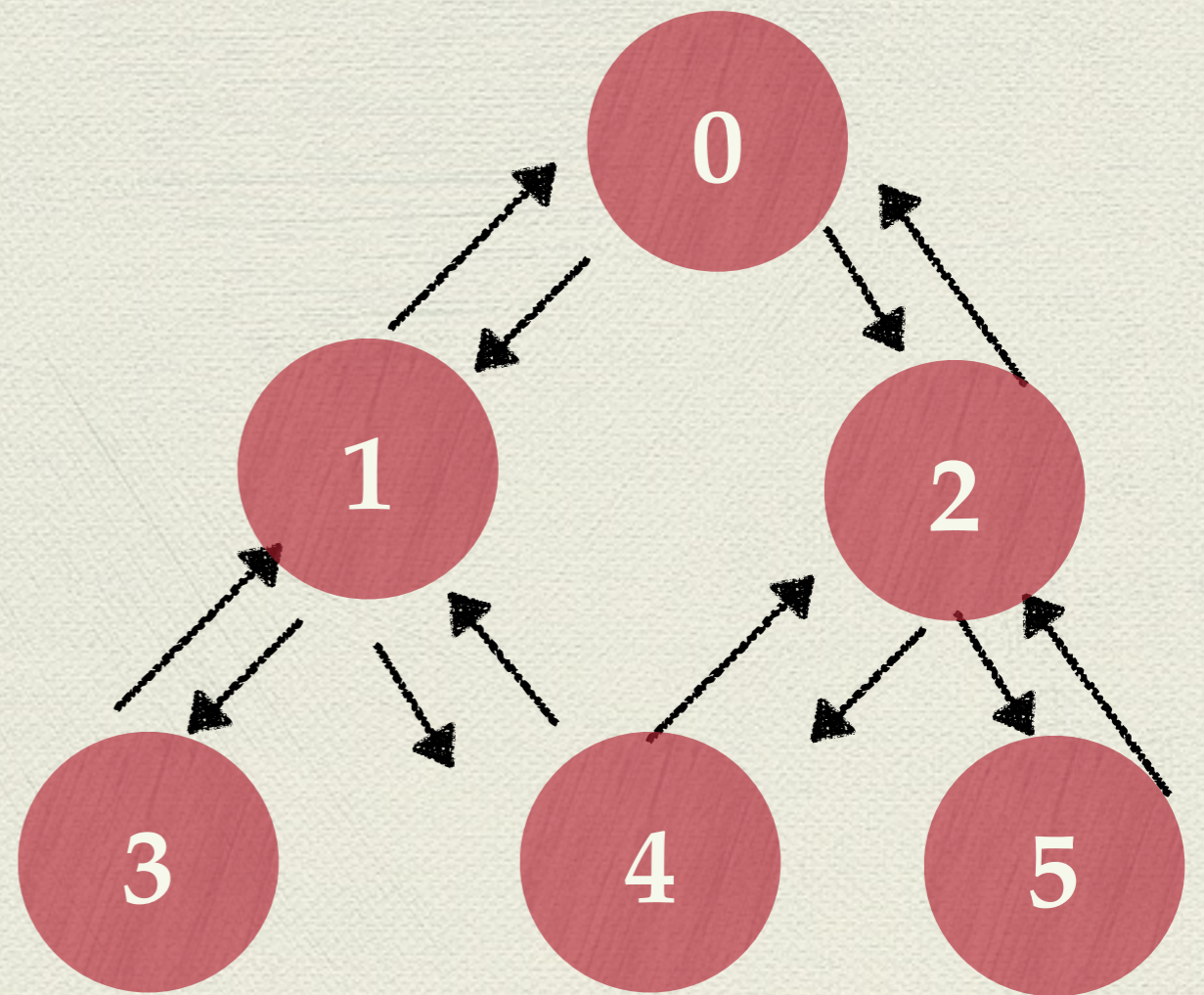


Let's see the traversal in action

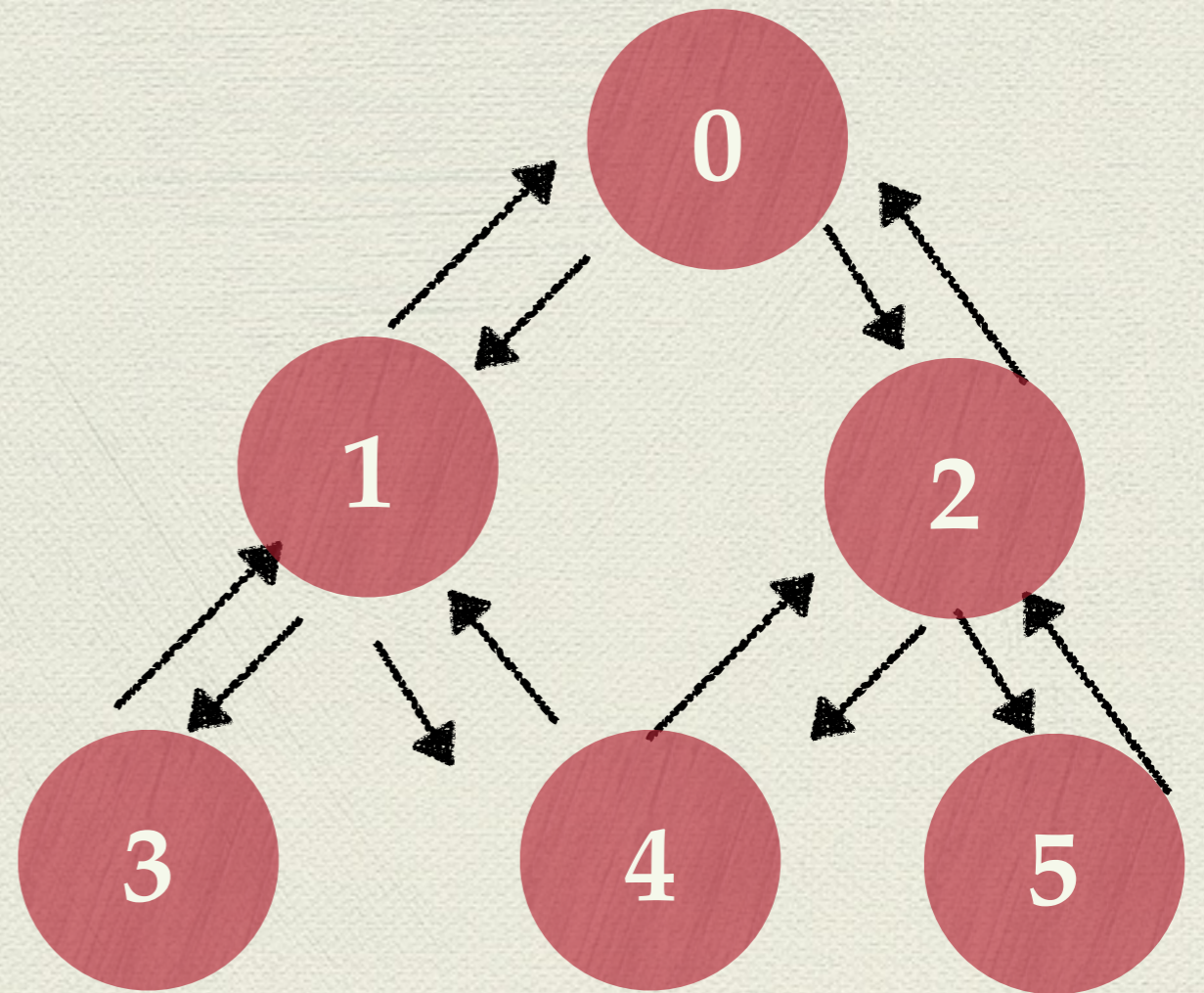


# Step 1: Create a fringe

The fringe (a Stack)

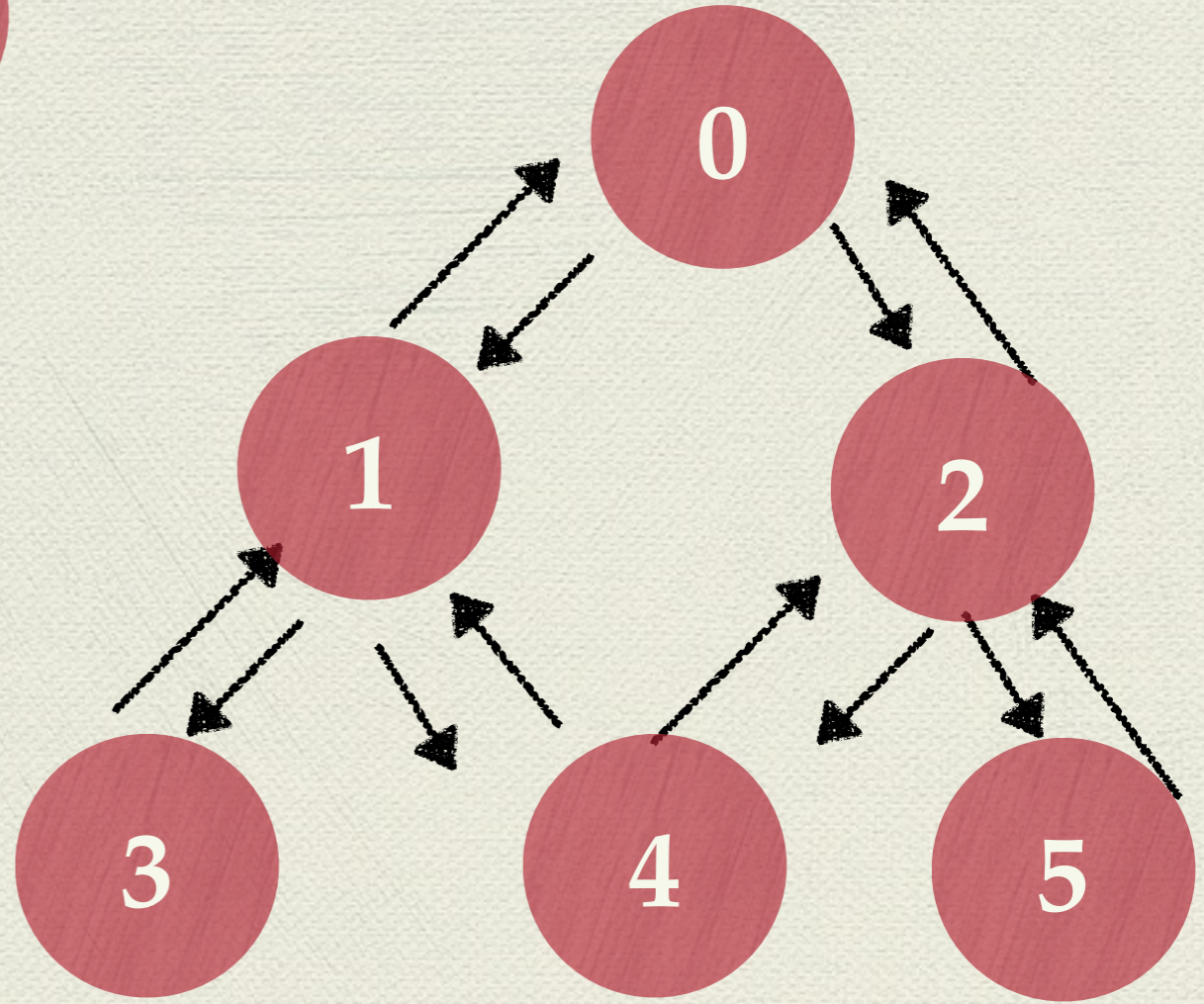


Step 2: Put  
initial tray in  
fringe



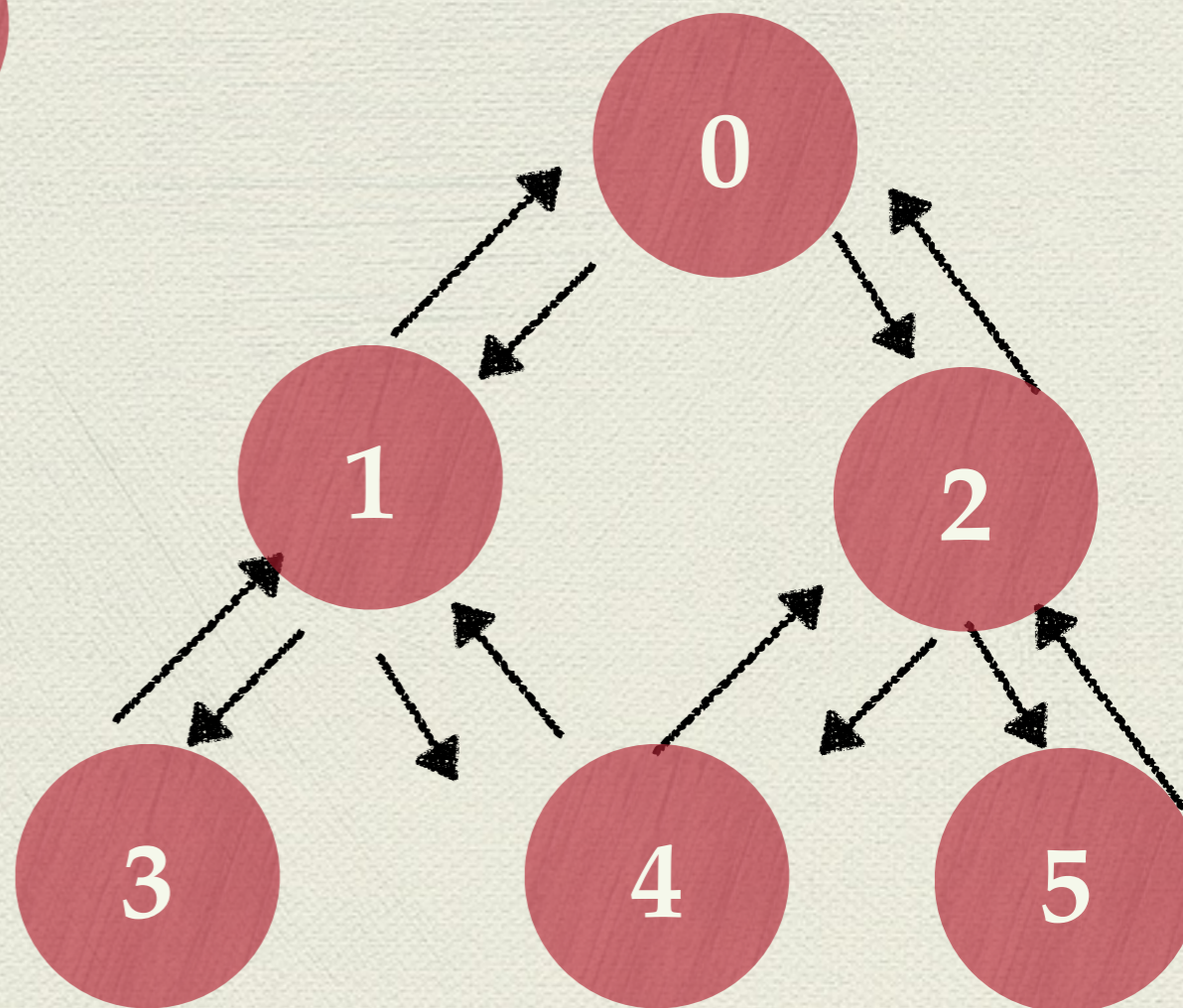
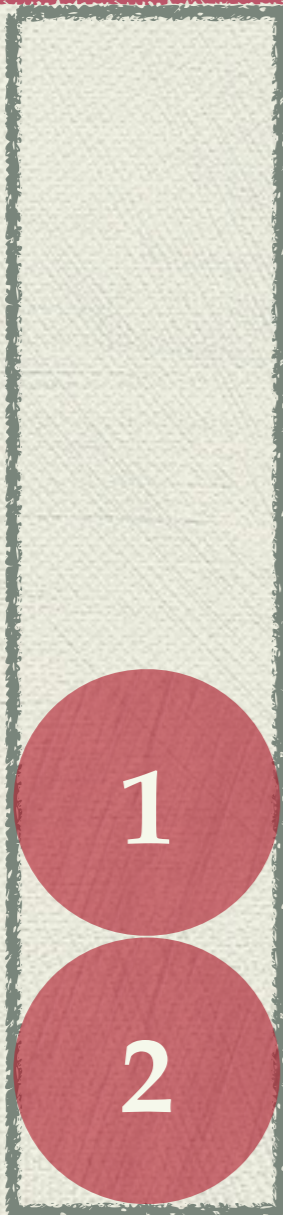
Step 3: Take something from the fringe, make it "current"

current



Check if it's the goal (it's not), so add adjacents to fringe

current



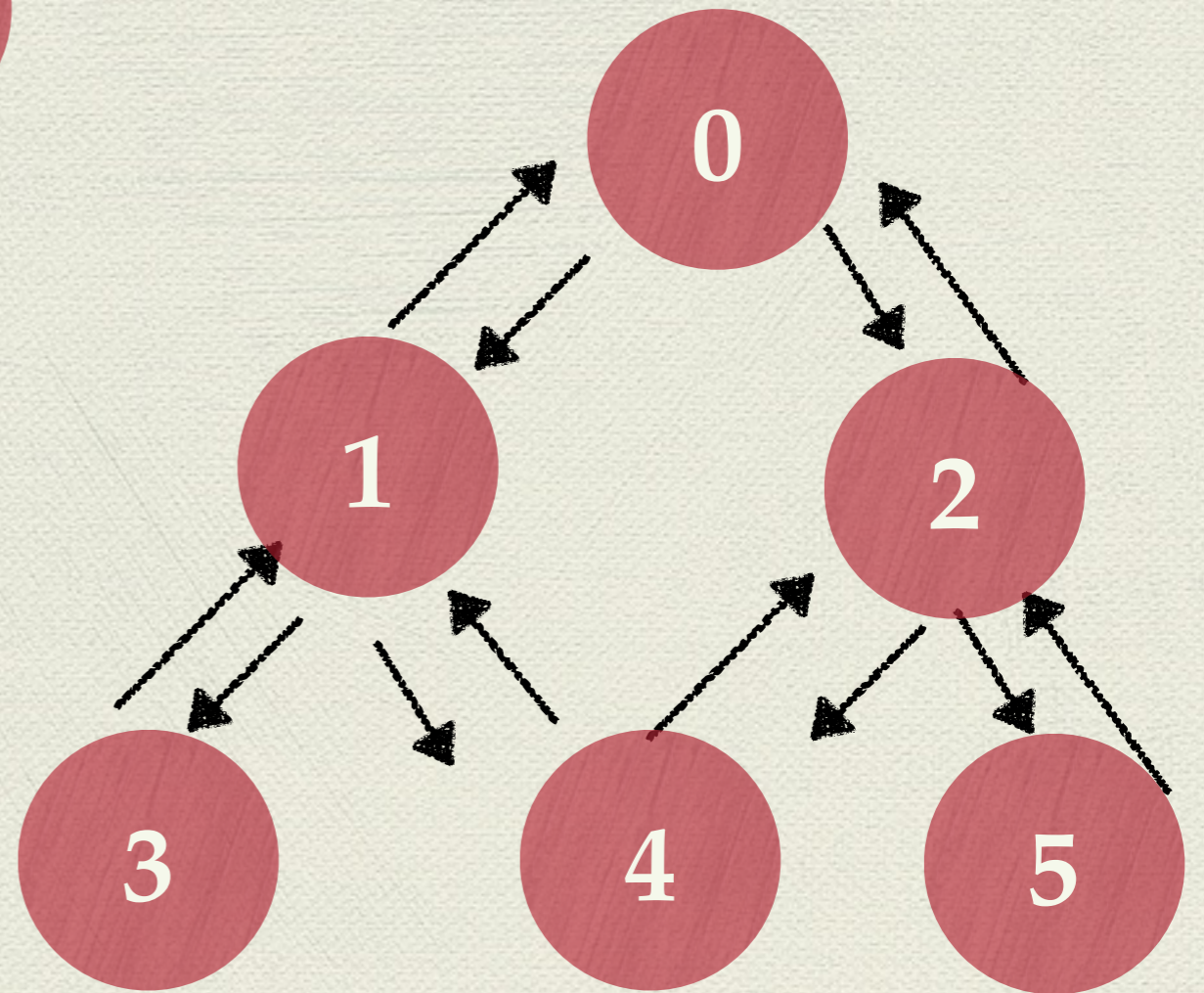


Take something,  
check if goal.  
It's not.

current

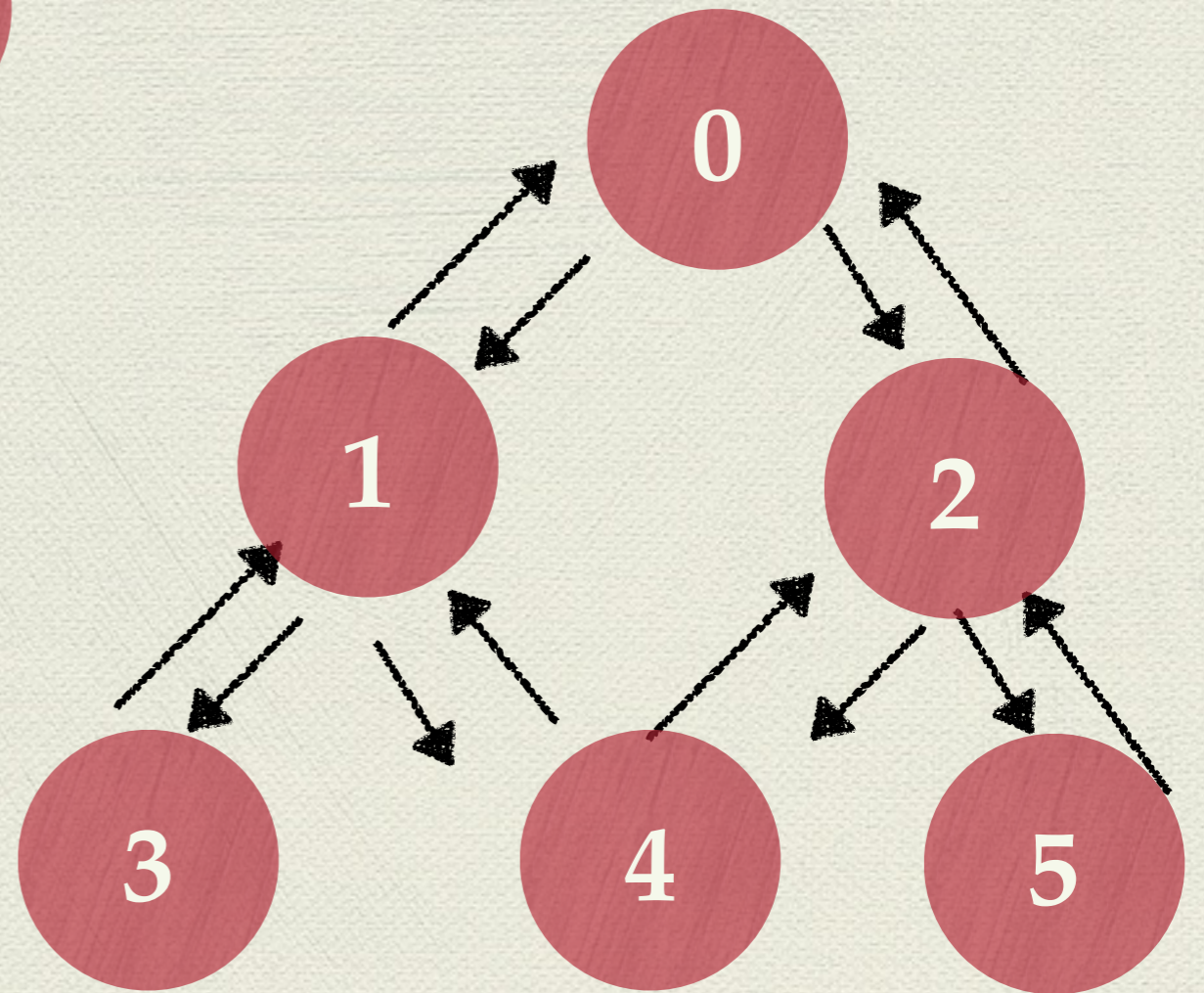
1

2



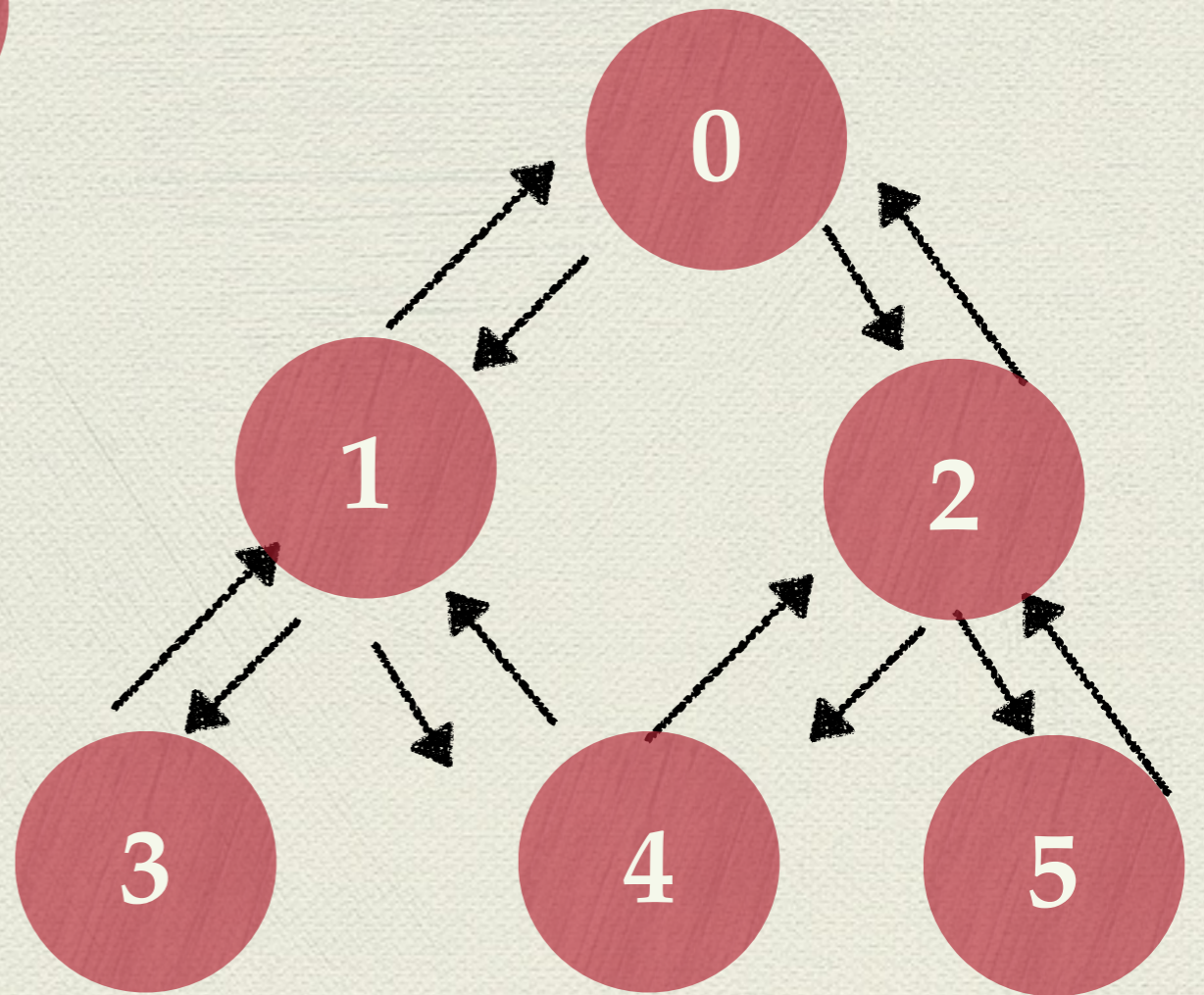
Add adjacent to  
fringe

current



Take something,  
make it current

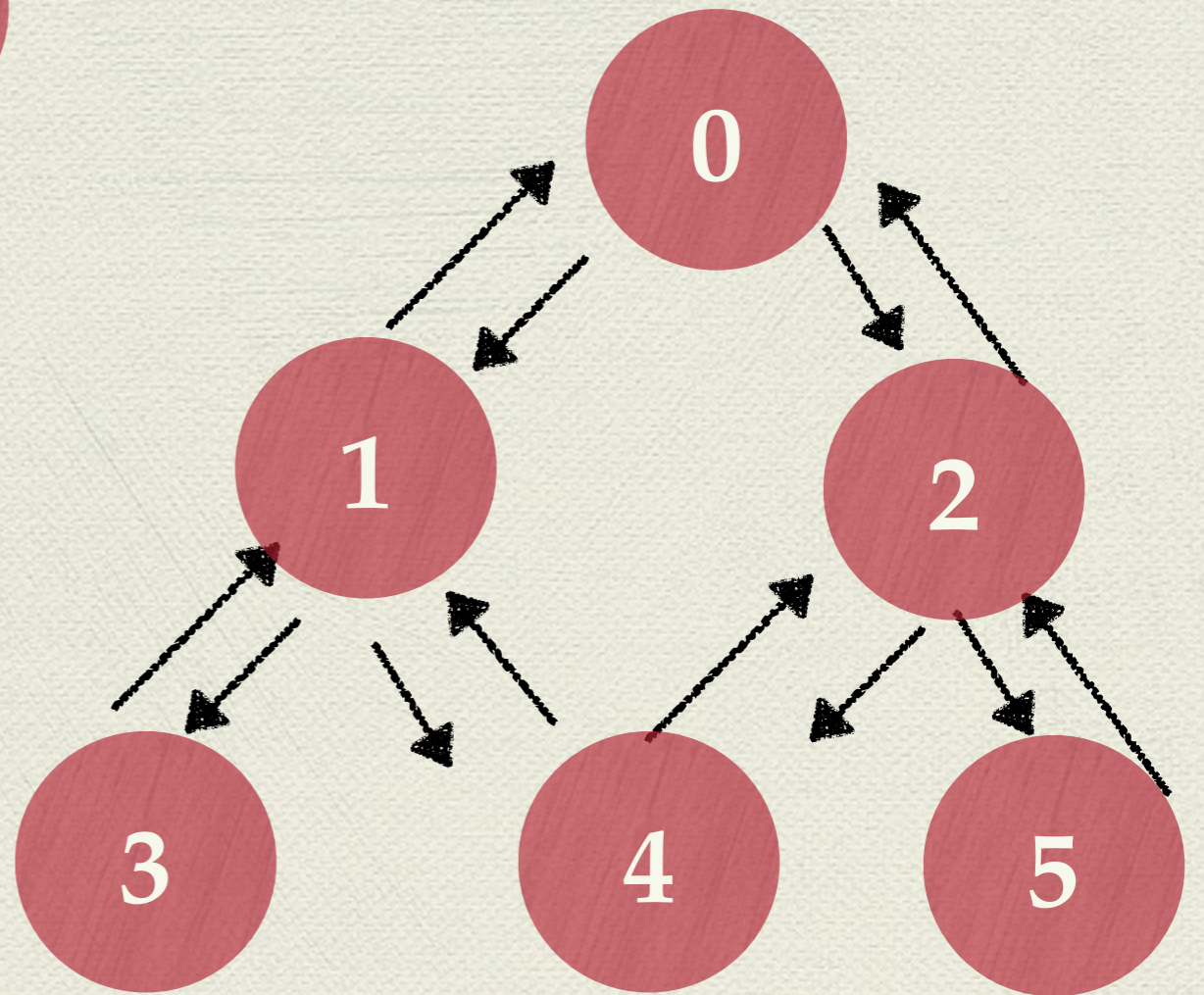
current



Check if goal.  
It's not.

current

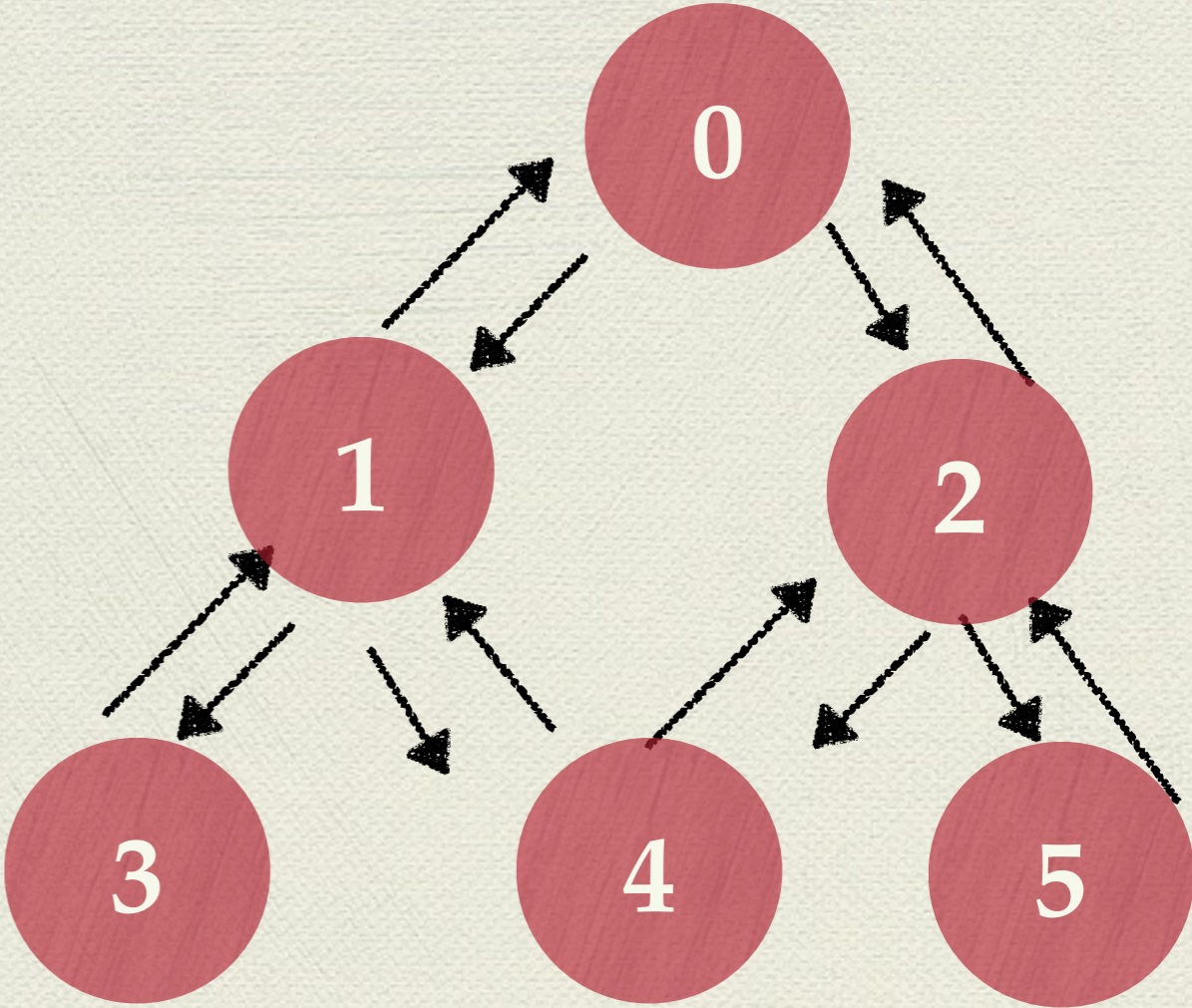
0



Add adjacent to fringe



current



Wait a minute!!

# Traversing a graph like a tree

- ◆ We end up going in circles!
- ◆ **What went wrong?**
  - ▶ The rules of trees ensure that, starting from root, there is only one possible path to each node
  - ▶ But for graphs, we can keep finding the same node over-and-over again

# Graph traversal

- ◆ Solution?
- ◆ Recall the **fringe** is meant to be a set of nodes we've temporarily passed by and intend to return to later
- ◆ So, let's not put something in the fringe if we've already visited it



# Graph traversal

```
Stack<Tray> fringe = new Stack<>();
fringe.push(initialTray);
while (!fringe.isEmpty()) {
    Tray currentTray = fringe.pop();
    // do stuff
    for (Tray t : currentTray.nextTrays()) {
        if (!alreadyVisited(t)) {
            fringe.push(t);
        }
    }
}
```

# What is this really?

```
Stack<Tray> fringe = new Stack<>();
Set<Tray> visited = new HashSet<>();
fringe.push(initialTray);
while (!fringe.isEmpty()) {
    Tray currentTray = fringe.pop();
    // do stuff
    visited.add(currentTray);
    for (Tray t : currentTray.nextTrays()) {
        if (!visited.contains(t)) {
            fringe.push(t);
        }
    }
}
```

# Graph traversal — the full story

- ◆ The same as tree traversal
- ◆ Except we make sure to not repeat ourselves

# Quiz part 1: path finding

- ◆ I claimed that **finding** the goal board during the traversal is essentially the same problem as figuring out the **path** to the goal board
- ◆ Is it really?

# Quiz part 1: path finding

```
public class GraphNode {
    String myItem;
    List<GraphNode> myAdjacents;
    /**
     * Prints out the items of the nodes you have
     * to follow from this node until you find a
     * you find a node with target item
     */
    public void printPathTo(String target) {
        // TODO your code here
    }
}
```

# Quiz part 1: path finding

```
public void printPathTo(String target) {
    Set<GraphNode> visited = new HashSet<>();
    Stack<GraphNode> fringe = new Stack<>();
    Map<String, String> steps = new HashMap<>();
    fringe.push(this);
    while (!fringe.isEmpty()) {
        GraphNode currentNode = fringe.pop();
        if (currentNode.myItem.equals(target)) {
            break;
        }
        visited.add(currentNode);
        for (GraphNode g : currentNode.myAdjacents) {
            if (!visited.contains(g)) {
                steps.put(g.myItem, currentNode.myItem);
                fringe.push(g);
            }
        }
    }
    Stack<String> reversePath = new Stack<>();
    String currentStep = target;
    while (currentStep != null) {
        String previousStep = steps.get(currentStep);
        if (previousStep != null) {
            reversePath.push(previousStep);
        }
        currentStep = previousStep;
    }
    while (!reversePath.isEmpty()) {
        System.out.println(reversePath.pop());
    }
}
```

# Our problem: an implicit graph

- ◆ Here again is our traversal code

```
Stack<Tray> fringe = new Stack<>();
Set<Tray> visited = new HashSet<>();
fringe.push(initialTray);
while (!fringe.isEmpty()) {
    Tray currentTray = fringe.pop();
    // do stuff
    visited.add(currentTray);
    for (Tray t : currentTray.nextTrays()) {
        if (!visited.contains(t)) {
            fringe.push(t);
        }
    }
}
```

- ◆ Notice we **don't** have one object that stores the entire graph of possible tray configurations

# Our problem: an implicit graph

- ◆ Instead, if each tray just knows about the trays that can follow it, then we **implicitly** have a graph
- ◆ We never actually have a variable of type **Graph<Tray>** that stores all the trays



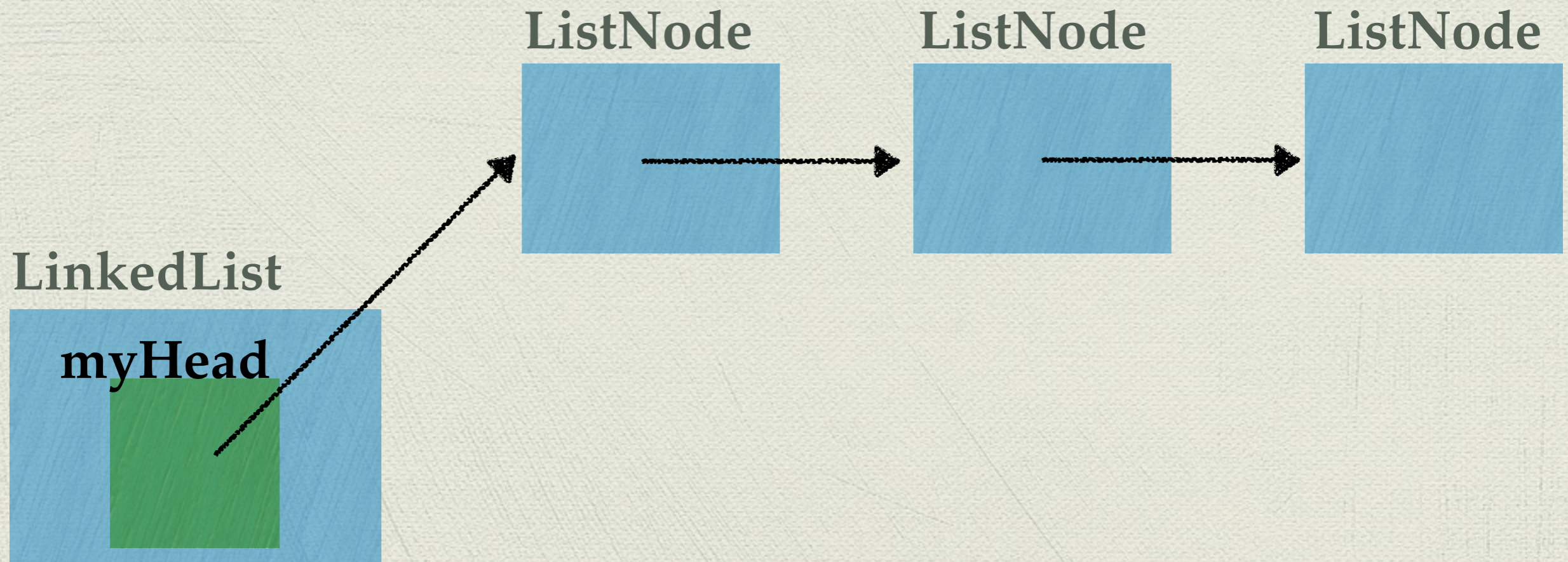
# Lucky us, because...

- ◆ ...the graph of possible trays is usually **far too big** for us to store in memory at once
- ◆ Good thing we only have to look at one local part at a time
- ◆ For completeness, though: what if we wanted to store the whole explicit graph?

# Digression: explicit graph representations

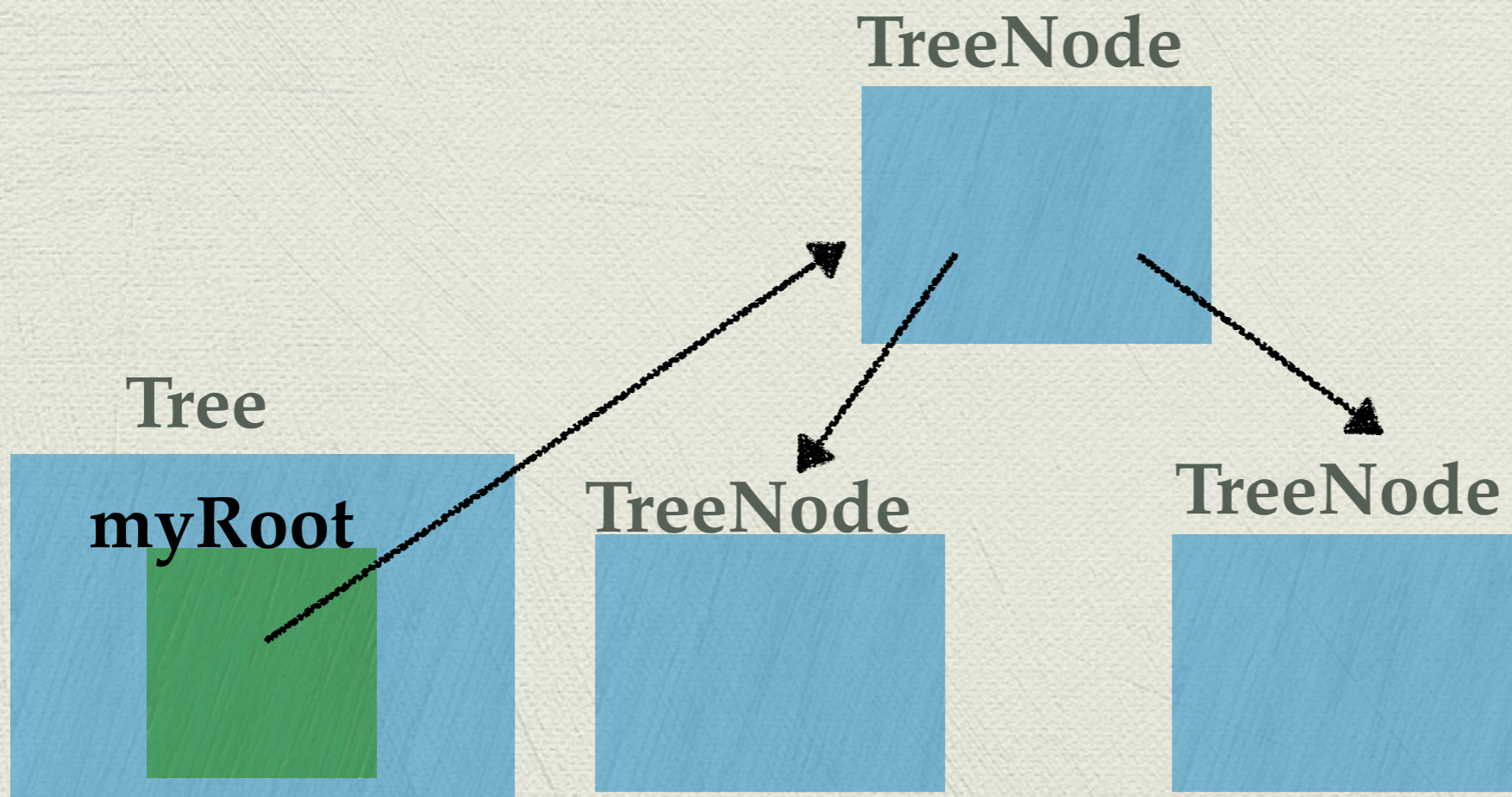
# Data structures with nodes

- ◆ For **linked lists**, we had a `LinkedList` class, that stored a reference to the first node, from which could be found all the other nodes



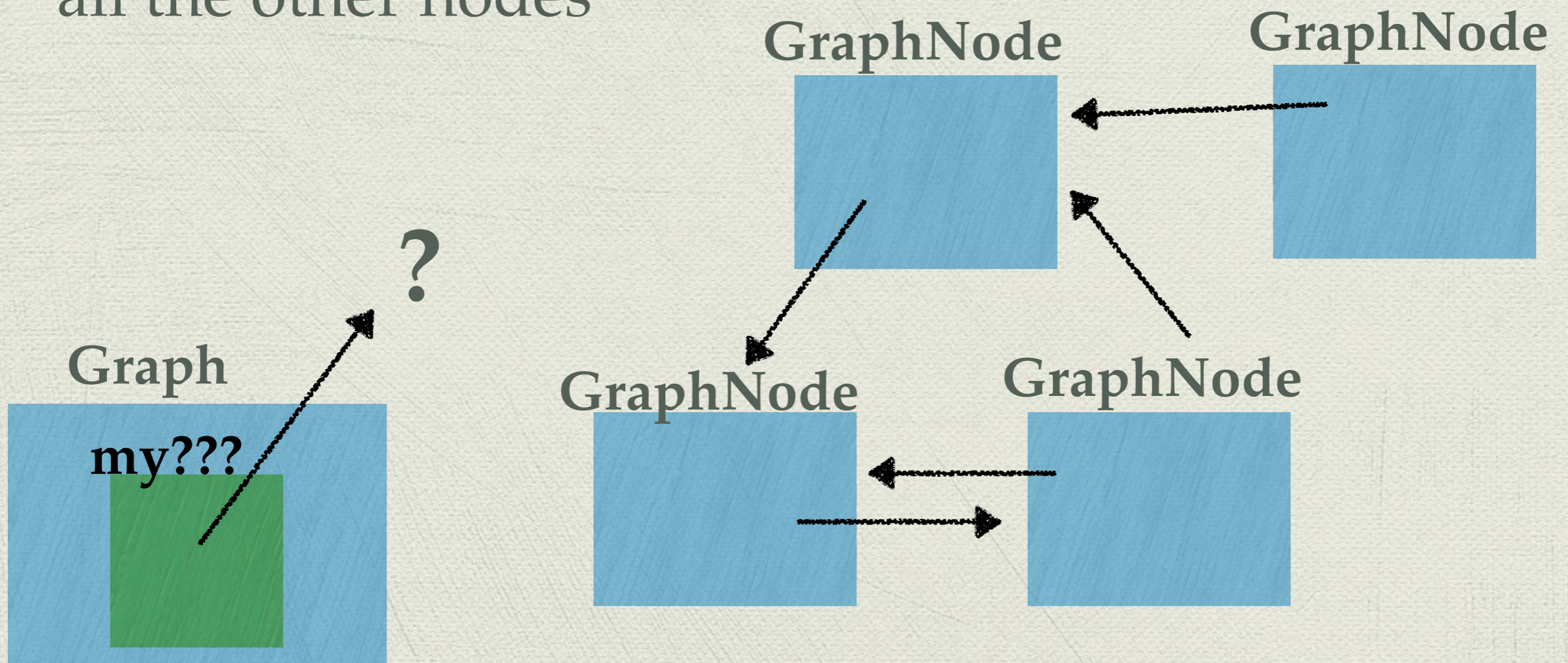
# Data structures with nodes

- ◆ For **trees**, we had a **Tree** class, that stored a reference to the **root**, from which could be found all the other nodes



# Data structures with nodes

- ◆ For **graphs**, we could have a **Graph** class, that stores a reference to **???**, from which could be found all the other nodes



# Graph data structure

- ◆ What would the Graph object store a reference to?
- ◆ Because a graph can have any structure, there isn't an obvious "first" or "starting" node in general

I guess we just have to store all of them

- ◆ The Graph object will store an array, one spot for each node

Warning: strange  
assumption!!

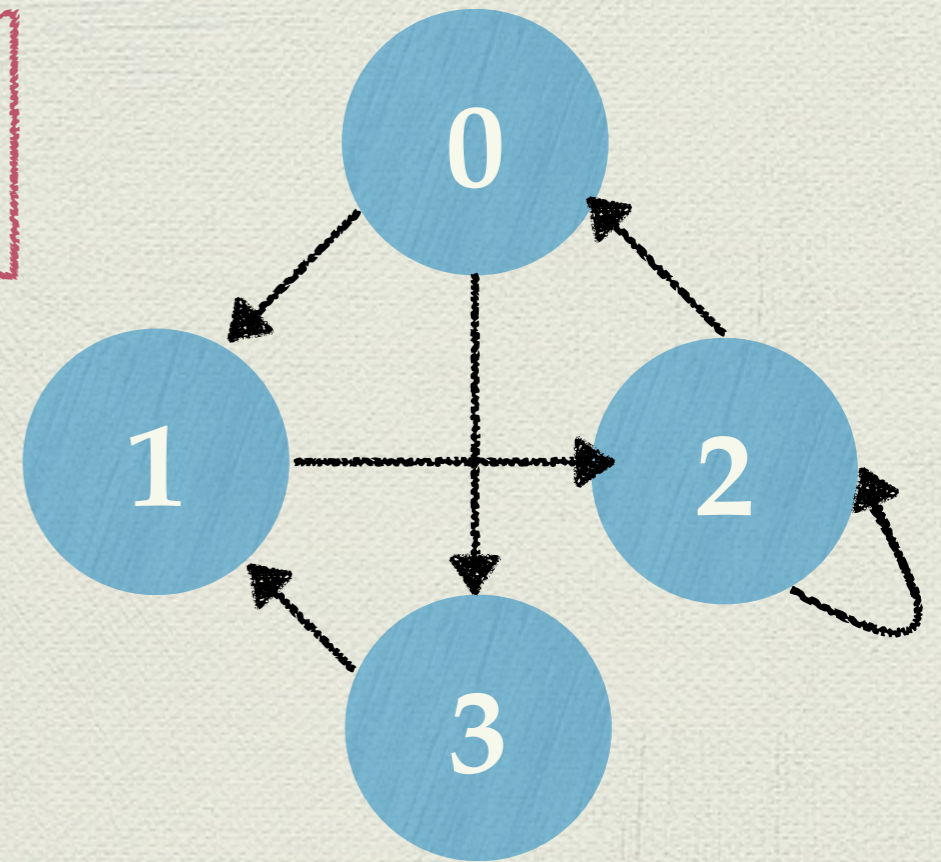


# But first, an assumption

- ◆ Before discussing the graph representations, I will first introduce an assumption
- ◆ The graph does not store arbitrary objects (like Strings, Trays, etc.). **Instead, it can only store the integers  $0 \dots N$  (if there are  $N+1$  vertices).**
- ◆ *Wha...? Why?*
- ◆ Will be justified later!

# Example graph we want to represent in Java

The theoretical  
(conceptual) graph

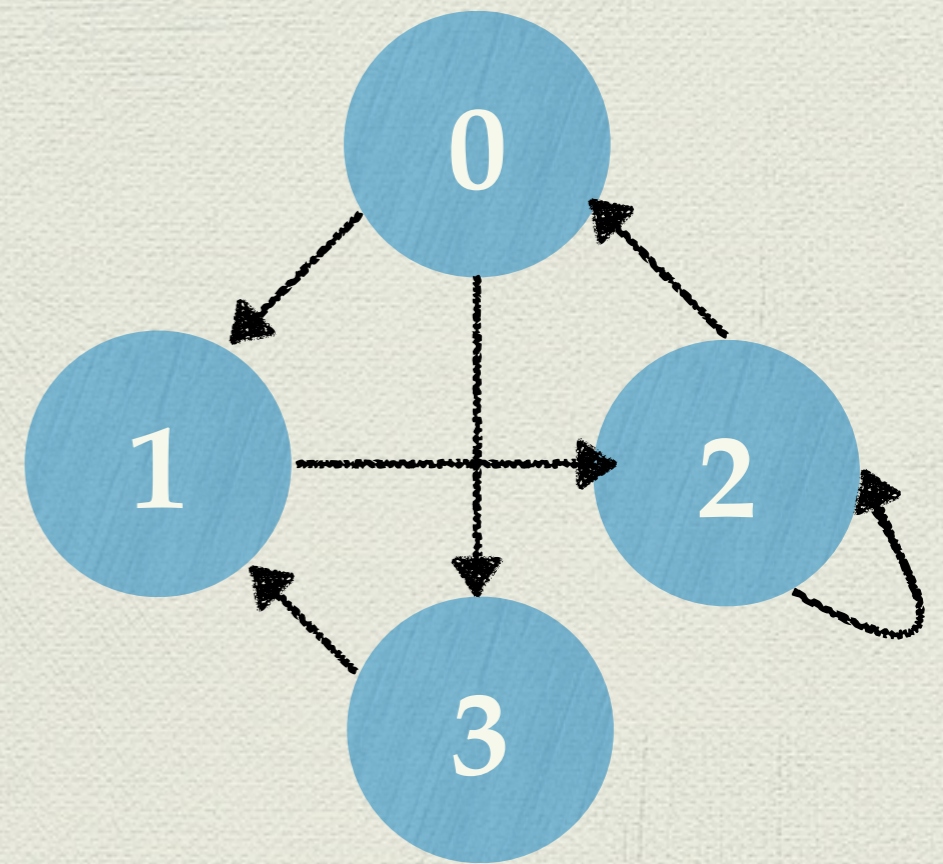
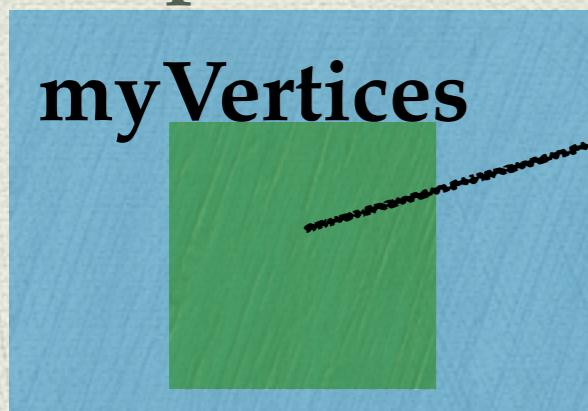


# The graph data structure

- ◆ The graph object will store an array, one spot for each vertex

Indices of the array  
match the vertices

Graph



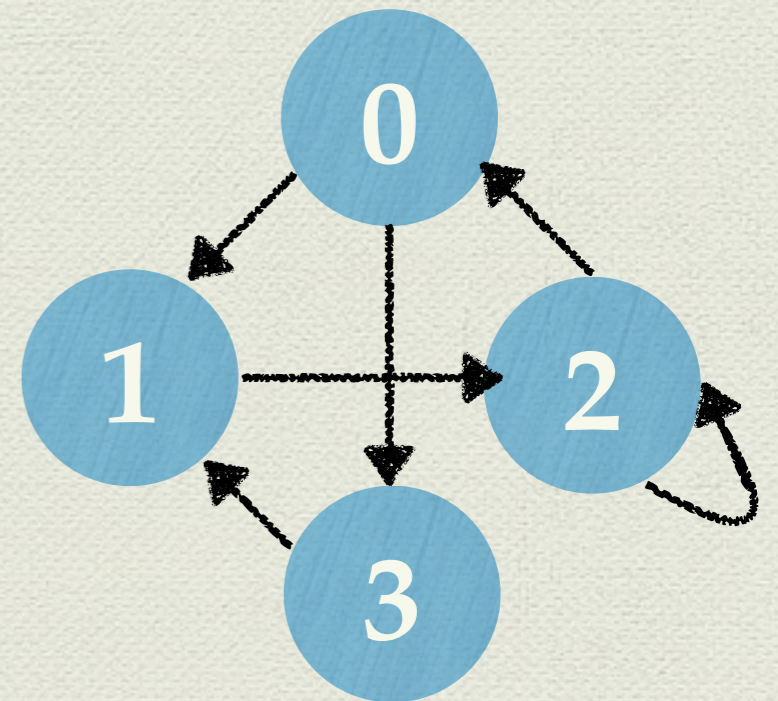
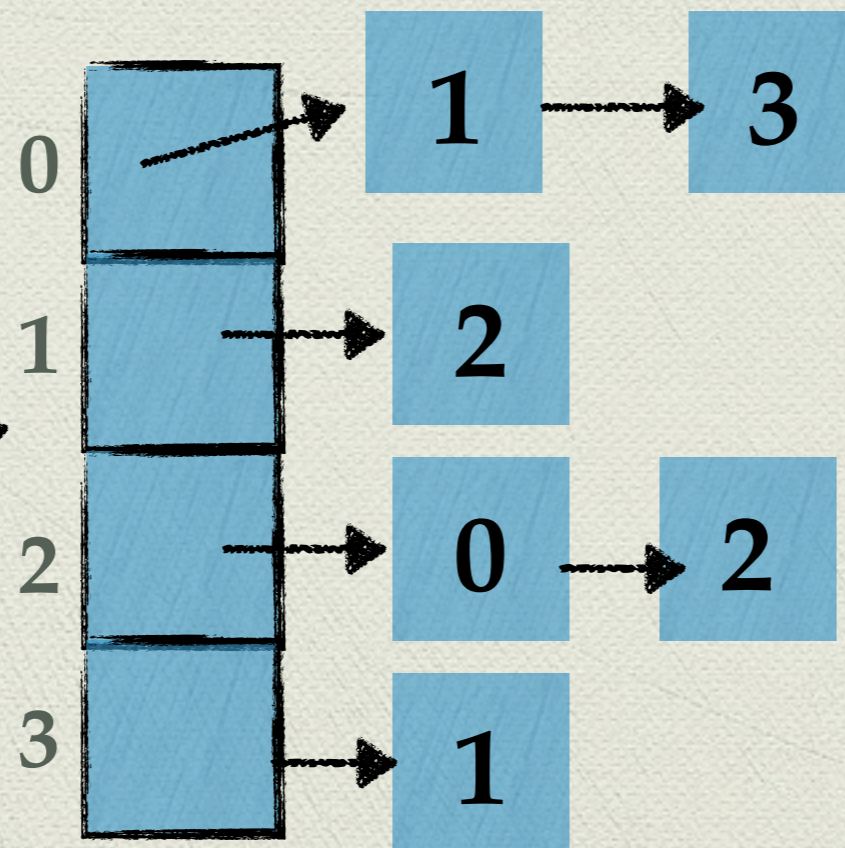
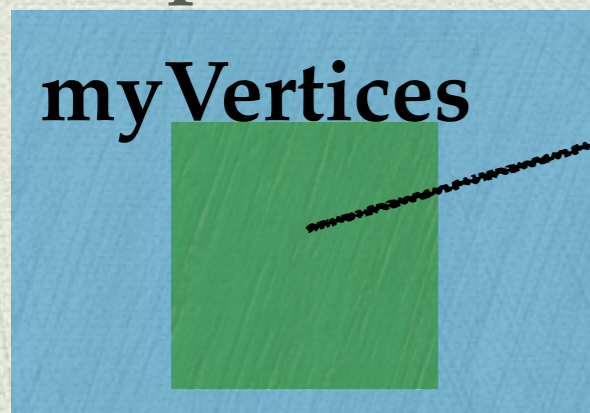
# The graph data structure

- ◆ The only thing that matters about a graph is which vertices have edges between them

# The graph data structure

- So each array will contain a **list** of vertices that are adjacent

Graph

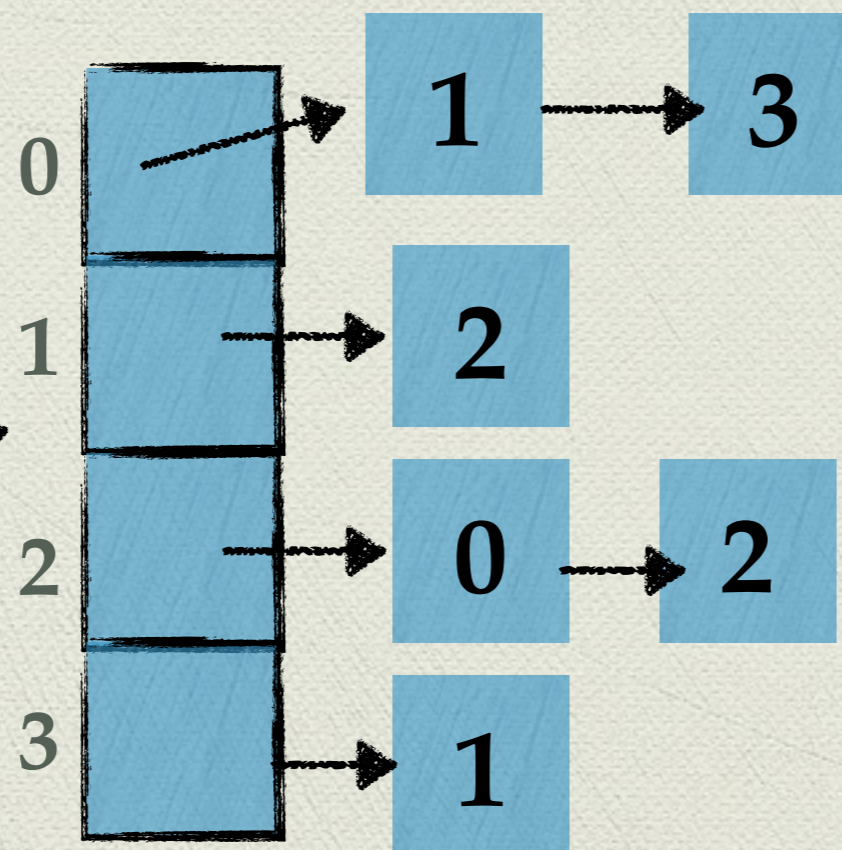
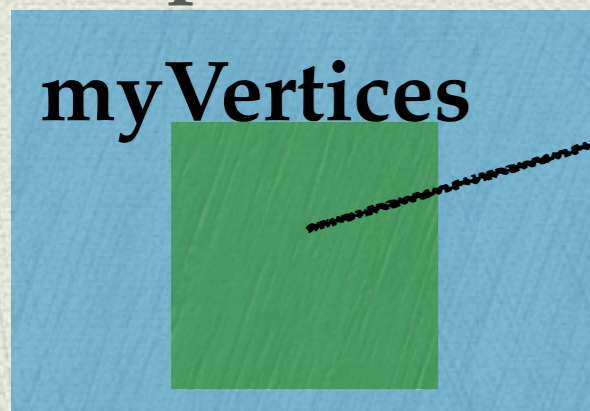


# The graph data structure

- ◆ So each array will contain a **list** of vertices that are adjacent

The connections between these nodes are NOT the edges in our graph!

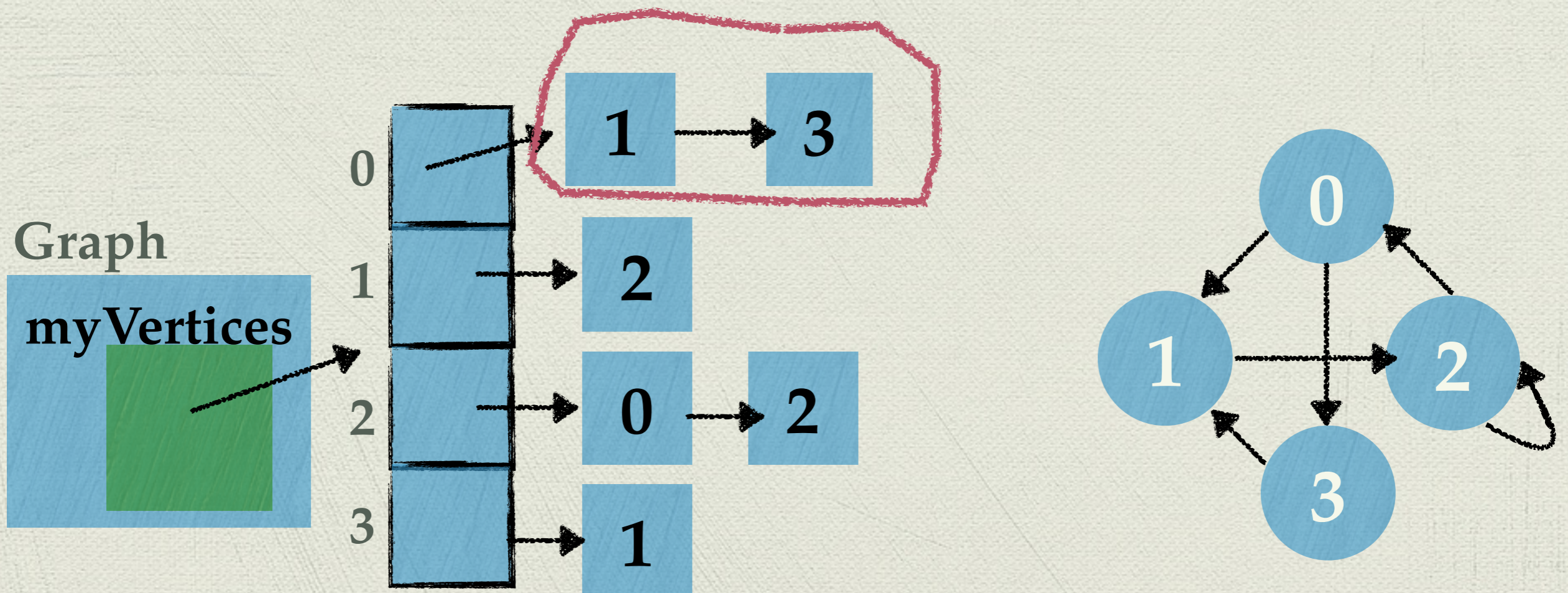
Graph



These are just a list of adjacent nodes

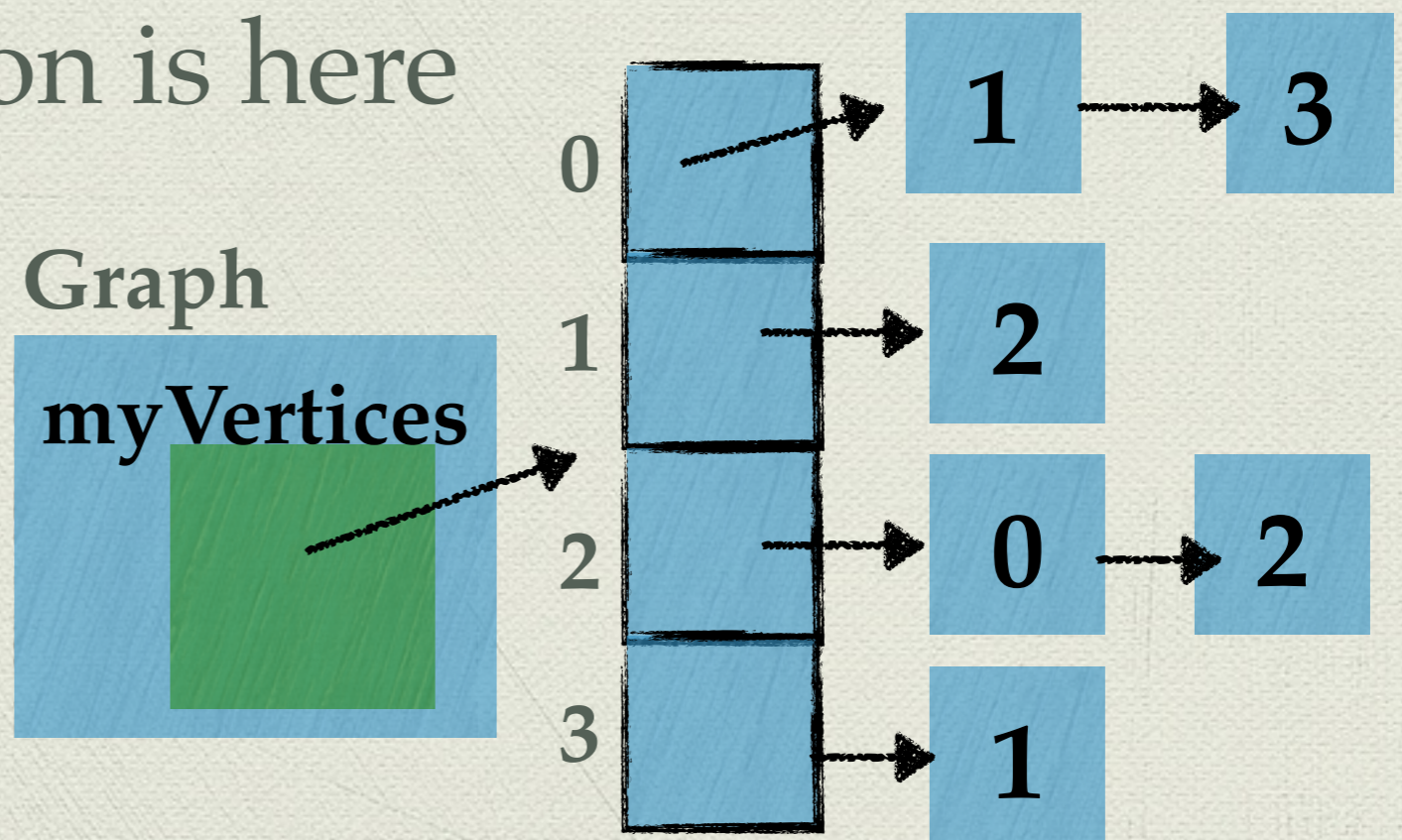
# The graph data structure

This is called the adjacency list associated with node 0. It tells you the vertices adjacent to 0



# Graph data structure: a picture

- ◆ A graph is nothing more than a set of vertices and connections between them
- ◆ All the information is here





# Story of a beautiful partnership: the sequel

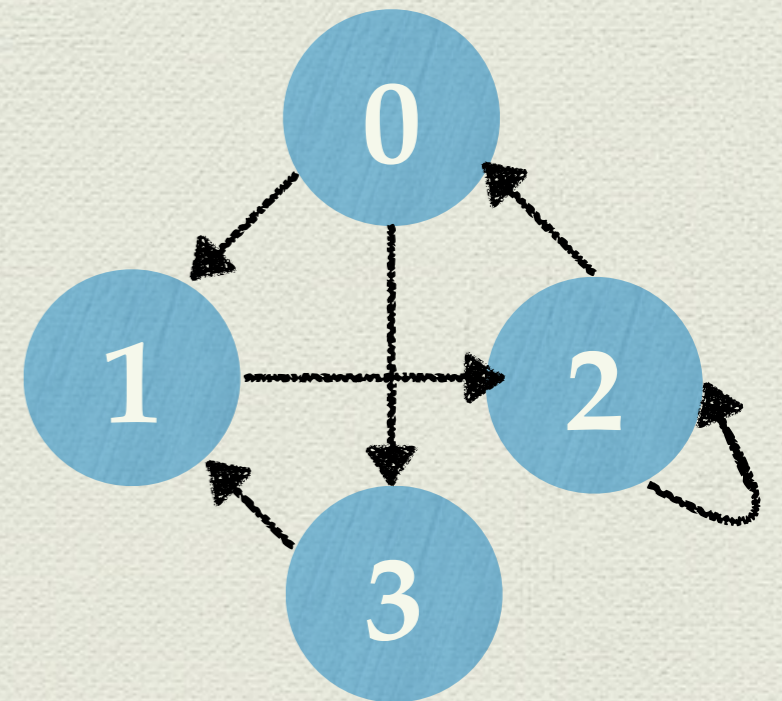
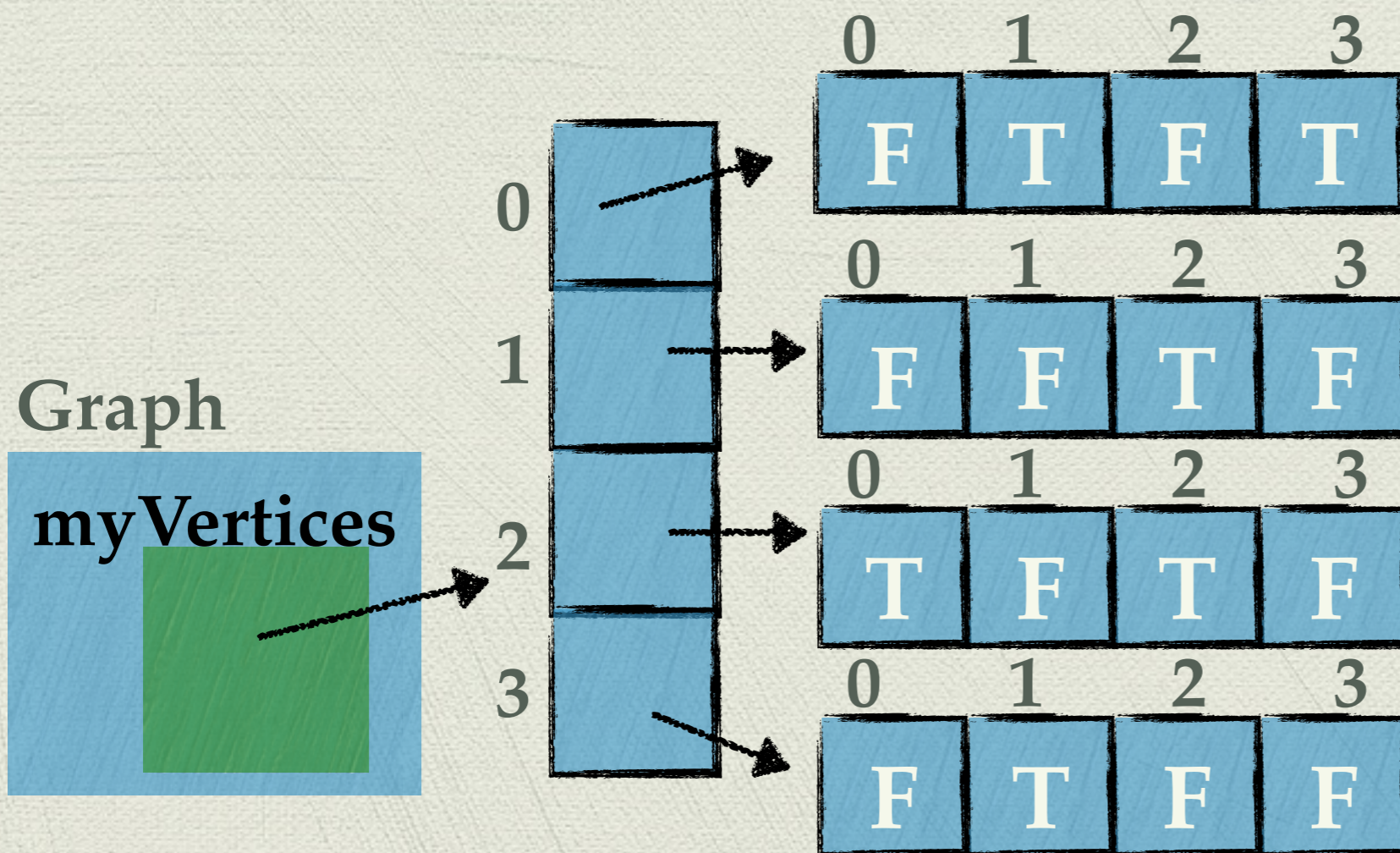
- ◆ A graph can be represented as an array of linked lists!

# Why linked lists?

- ◆ Linked lists make it easy to append new items to the end, make it easy to **add edges**
- ◆ But what if we want to check if an edge exists? e.g. check if  $0 \rightarrow 2$ ?
  - ▶ Must iterate through list at 0 until we find 2... not cool!

# Alternative: array of arrays

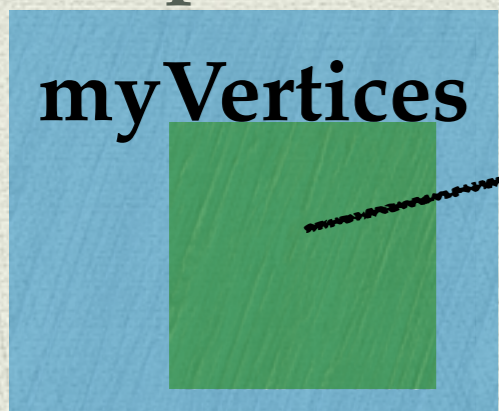
- Store an array rather than a list, tells you whether it is adjacent



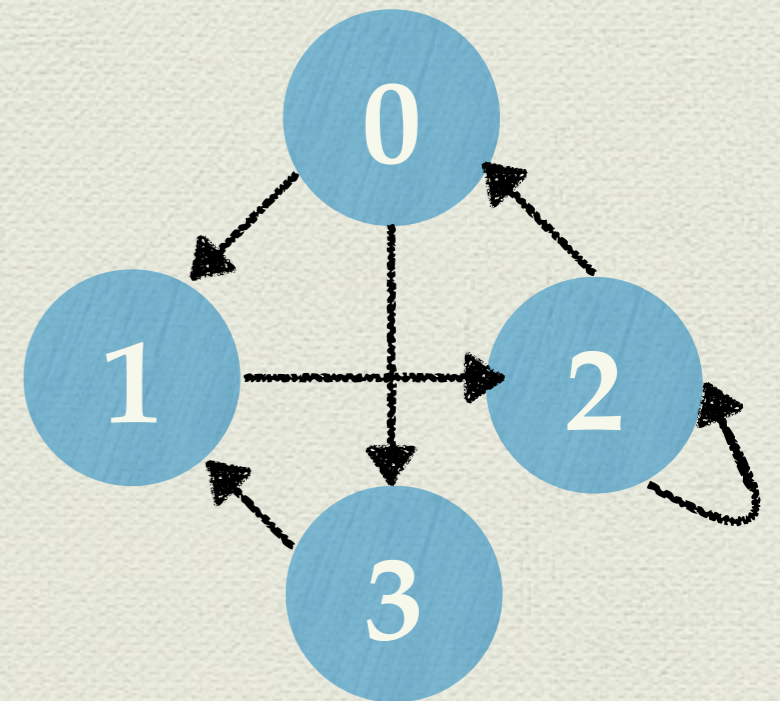
# Alternative: array of arrays

- Commonly drawn like this:

Graph

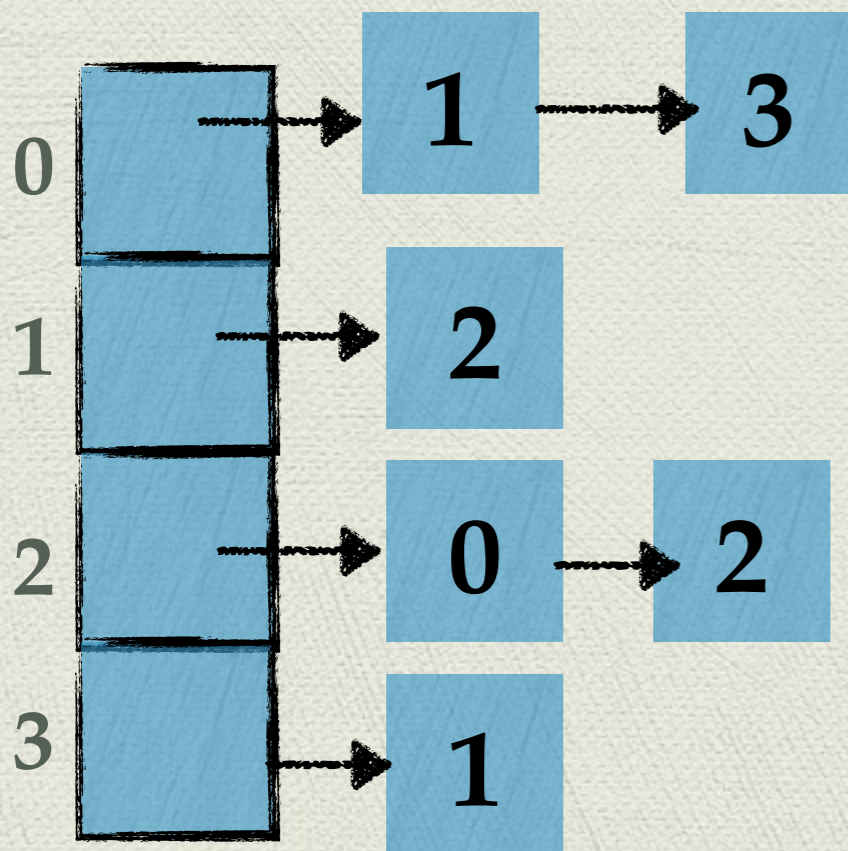


	0	1	2	3
0	F	T	F	T
1	F	F	T	F
2	T	F	T	F
3	F	T	F	F



# Two graph representations

## Array of adjacency lists



## Adjacency matrix

	0	1	2	3
0	F	T	F	T
1	F	F	T	F
2	T	F	T	F
3	F	T	F	F

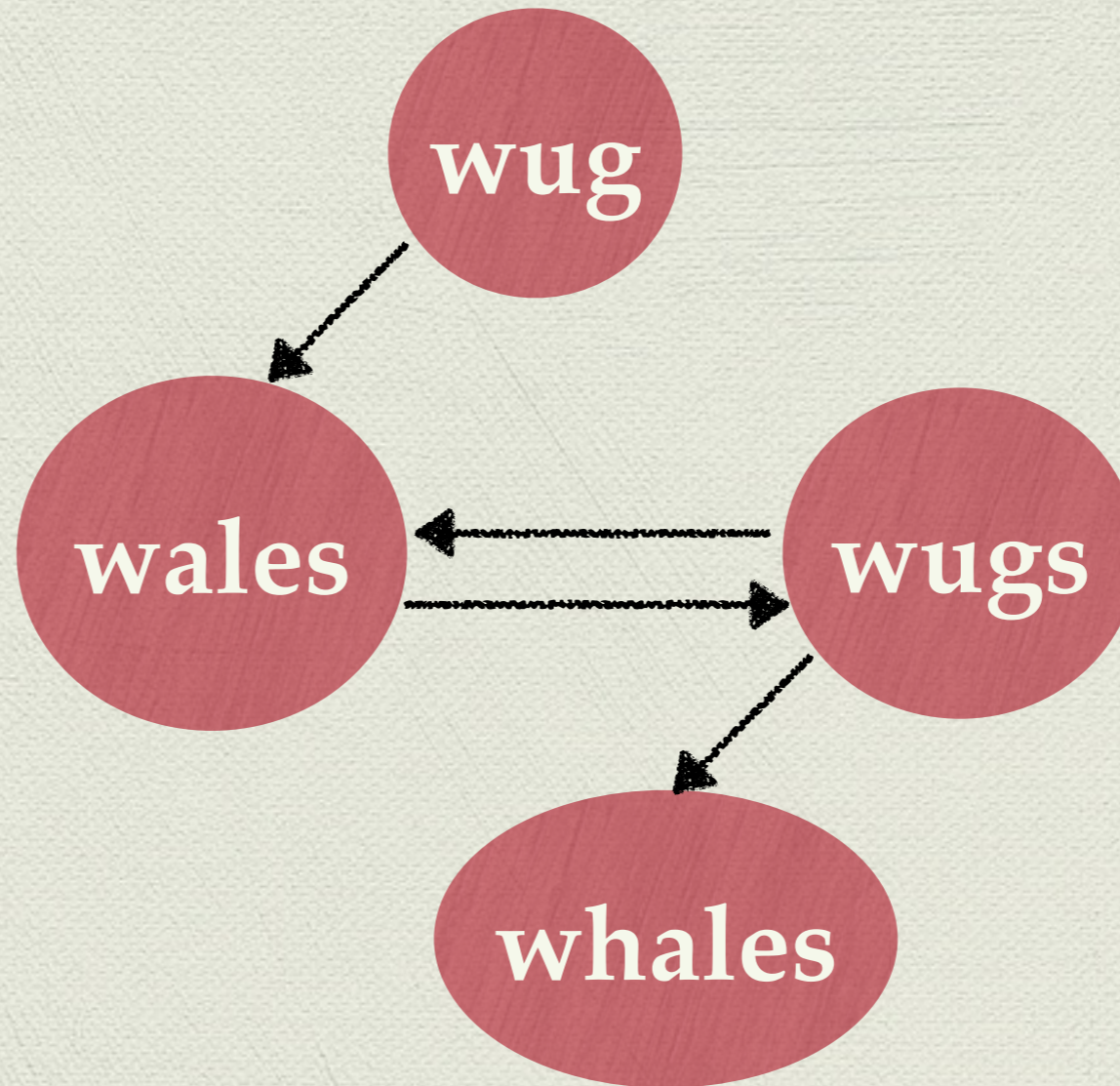
# Pros / cons of two graph representations

- ◆ The matrix is faster to check if there exists an edge (just index into the array)
- ◆ But, the matrix is a waste of space if there are lots of false values

# Graphs of other things

- ◆ So, we can now store a graph of integers  $0 \dots N$
- ◆ What if we want to store a graph of something else? Like Strings, or tray objects?
- ◆ **Easy!** We'll have a map from object to number, and back

Say we want to represent this  
conceptual graph





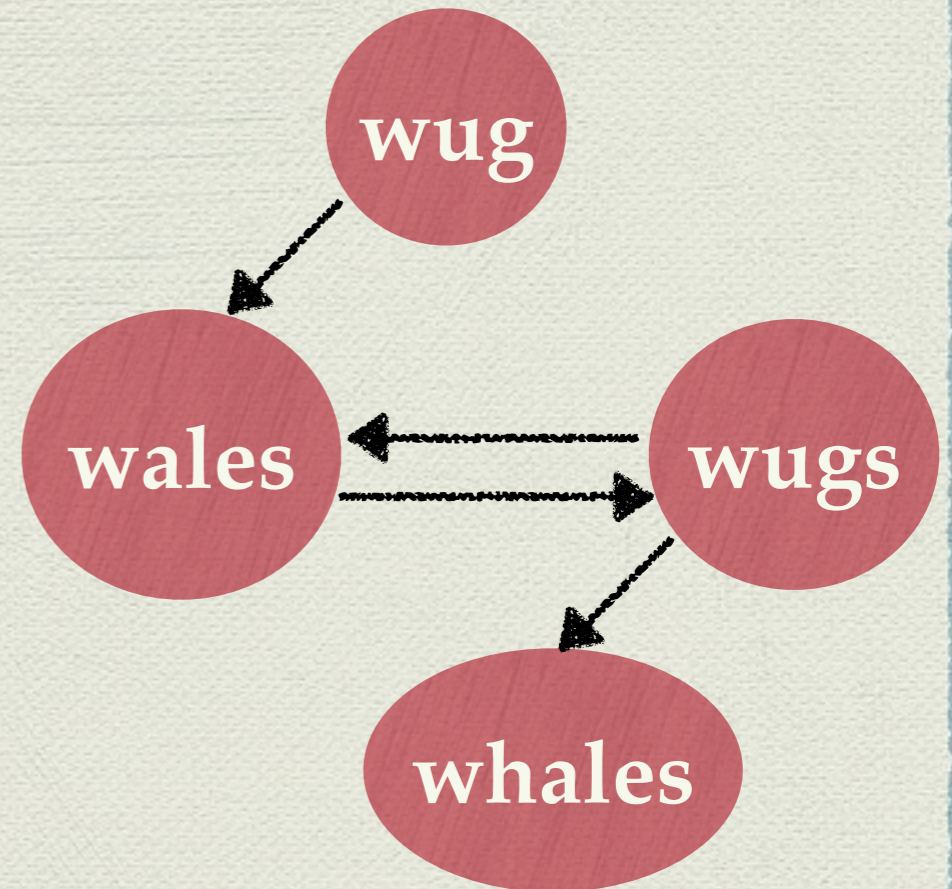
# Here's the picture

Object	Index
"wug"	0
"wugs"	1
"wales"	2
"whales"	3

Adjacency structure

	0	1	2	3
0	F	F	T	F
1	F	F	T	T
2	F	T	F	F
3	F	F	F	F

Map from object to index



The conceptual graph

# Graphs of other things

- ◆ **Consist of two parts:**
  - ▶ **The indexer**, which associates each item with an index
  - ▶ **And the adjacency structure**, which keeps track of all the connections

# What's this indexer thing?

- ◆ The reason we separate them out is so that graphs of all different kinds of things can have essentially *the same core structure*
- ◆ This means someone could write 1 graph processing function that could work on all sorts of graphs

Digression complete!

BREAK

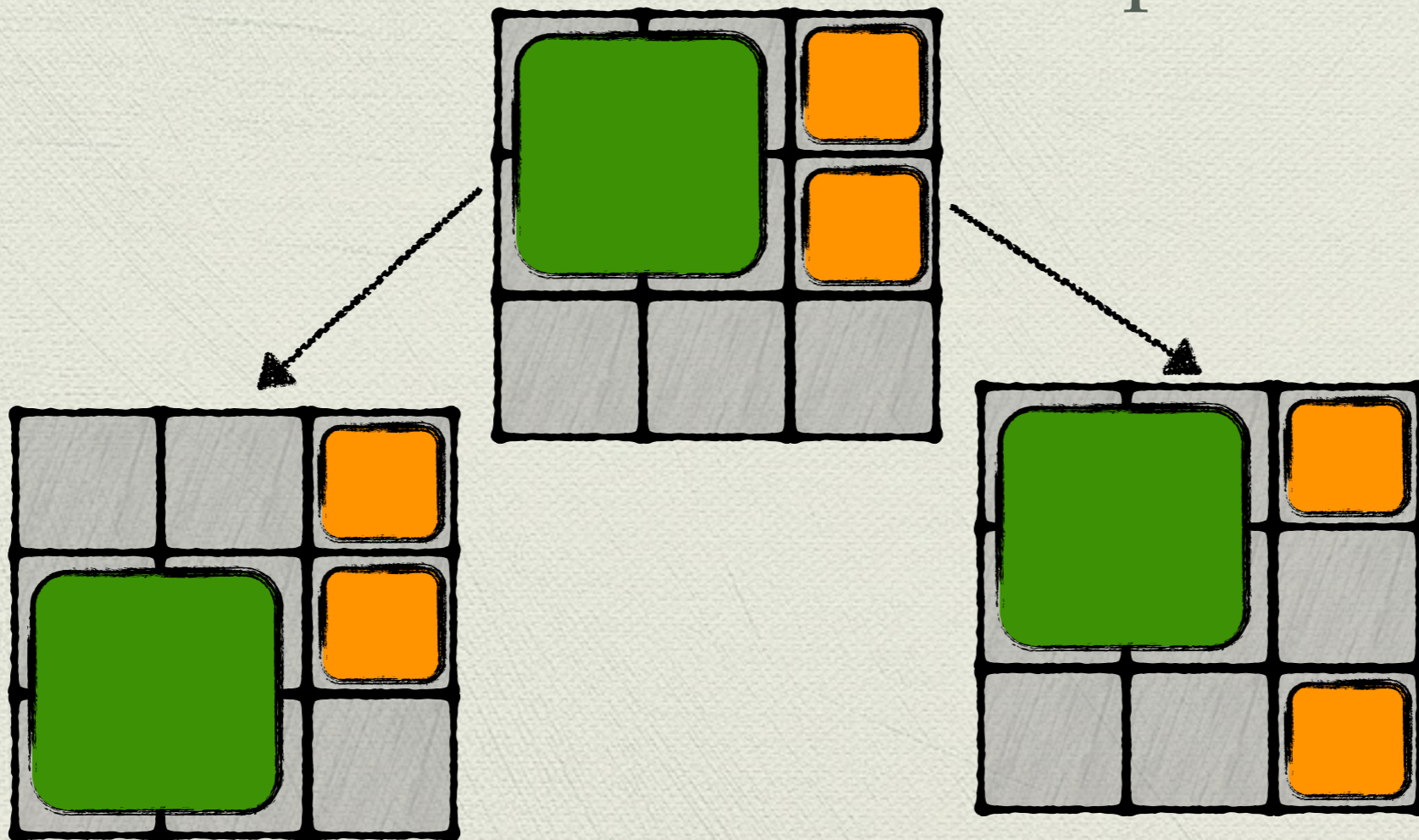
Back to our problem...

# Sliding block puzzle

- ◆ Yay we learned about explicit graphs~
- ◆ But these were irrelevant for our block sliding problem, which is an implicit graph
- ◆ We decided to solve this problem using traversals. Are we done?
- ◆ Not yet! We want a faster solution

# A new idea

- ◆ When we have a choice which way to go...
- ◆ Shouldn't we choose the *best* option?





# Introducing heuristics (your new best friend)

- ◆ **Idea:** Choose the one that moves the block closer to the goal position. Quantify this with a number.
- ◆ This number is called a **heuristic**, just a guess
- ◆ It might be wrong. Sometimes, moving the block toward the goal is the wrong thing to do. But it's a reasonable guess

# Change our code!

```
Stack<Tray> fringe = new Stack<>();
Set<Tray> visited = new HashSet<>();
fringe.push(initialTray);
while (!fringe.isEmpty()) {
    Tray currentTray = fringe.getBestItem();
    // do stuff
    visited.add(currentTray);
    for (Tray t : currentTray.nextTrays()) {
        if (!visited.contains(t)) {
            fringe.push(t);
        }
    }
}
```

# Introducing the priority queue ADT

- ◆ What is this `getBestItem` method?
- ◆ This is the method of a **priority queue**, not a stack!

# Priority queue

- ◆ Supports operations

- ▶ `void add(Comparable c)` (in Java, this is `offer`)

- ▶ `Comparable extractMin()` (in Java, this is `poll`)

# Min or max priority queue?

- ◆ Would we have extractMin or extractMax?
- ◆ Tradition is extractMin, but it's very easy to change to extractMax
  - ▶ Just flip everything I'm about to say for the remainder of lecture

# How to use priority queue

- ◆ `import java.util.PriorityQueue;`
- ◆ `PriorityQueue<Tray> q = new  
PriorityQueue<Tray>();`
- ◆ `q.offer(new Tray());`
- ◆ `Tray t = q.poll();`

# Let's make a priority queue!

- ◆ How?
- ◆ First idea: Use a **sorted linked list of items**

# Priority queue with a sorted list

- ◆ `add(Comparable c)` find the correct spot for the item in the sorted list, and put it there
- ◆ `Comparable extractMin()` remove and return the first item of the list



# Runtimes?

- ◆ `add(Comparable c)` find the correct spot for the item in the sorted list, and put it there:  $O(N)$  time in the worst case, where there are  $N$  items in the queue
- ◆ `Comparable extractMin()` remove and return the first item of the list:  $O(1)$

# Priority queue with sorted list

- ◆ Surely we can do better?

# Let's make a priority queue!

- ◆ Second idea: Use a **binary search tree** (already kinda sorted)

# Priority queue with a BST

- ◆ `add(Comparable c)` add to the BST like normal
- ◆ `Comparable extractMin()` go down to the far left, and return/remove the item there

# Runtimes?

- ◆ `add(Comparable c)` add to the BST like normal:  $O(\log N)$  time, with  $N$  items in queue
- ◆ `Comparable extractMin()` go down to the far left, and return/remove the item there:  $O(\log N)$  time

# The BST priority queue

- ◆ Both operations run in **log** time
- ◆ This is basically as good as it gets
- ◆ But there's still something a little unsatisfying

# The beefy BST

- ◆ The BST can find any item in log time
- ◆ But the priority queue only needs to find the best item quickly
  - ▶ The BST is *more powerful* than we need

# The beefy BST

- ◆ The BST is a complicated structure (did you have fun coding AVL tree rotations?)
- ◆ Since it is complicated **and** more powerful than we need, we might wonder if there's a simpler data structure that does the job just as well



# Simpler is better

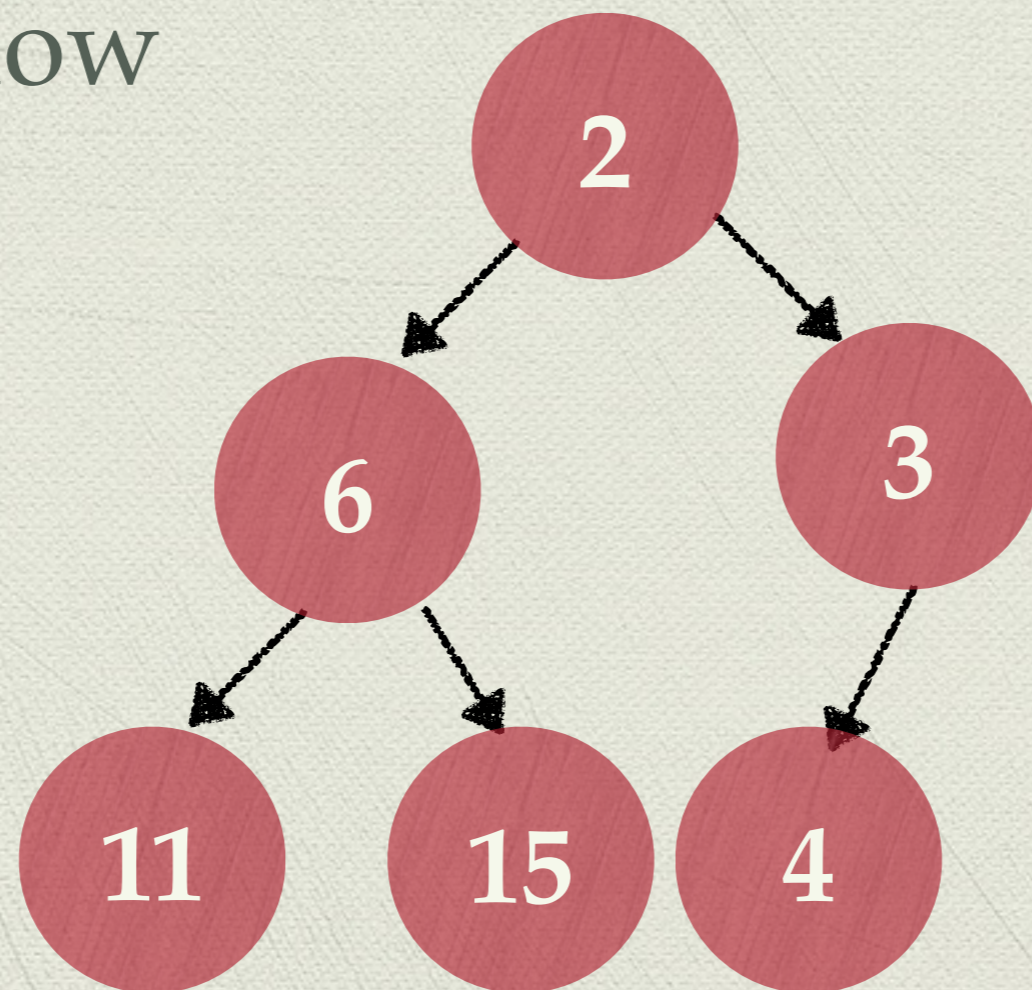
- ◆ Simpler data structures can be faster than complicated ones, due to constant factors, even if the asymptotic runtimes are the same
- ◆ This turns out to be the case for priority queue

# A simpler idea for a tree

- ◆ Why not just store the min item at the top of the tree? That's the easiest place to look

# Introducing the binary heap

- ◆ The binary (min) heap is a tree structure that stores items with smallest at the top, and bigger below



# Heap vs. BST

small items



large items

medium items



small items

large items

# Heap invariants

- ◆ Specifically, a heap is a tree with an extra **invariant**:
  - ▶ Every child is bigger than (or the same as) its parent
- ◆ Notice: **left-right order in heap does not matter** at all! This makes it simpler than BST

# Heap invariants

- ◆ Remember a BST also had an **almost balanced invariant**
- ◆ Heaps will also have a balance invariant, but it will be the **maximally balanced property**

# Maximally balanced — why now?

- ◆ For BSTs, maintaining maximal balance requires a lot of work — even maintaining almost balance required some wacky rotations
- ◆ Because the heap is simpler overall, it offsets the extra work required to maintain maximal balance

# Recall: maximal balance

- ◆ Was equivalent to the condition that the array tree has no holes in it



# Recall: maximal balance

- ◆ We agreed the array tree would be more memory efficient if the tree was maximally balanced
- ◆ So we'll implement the heap with an array, usually!

# Heap properties

- ◆ To sum up, a heap has additional two properties over a normal tree:
  - ▶ The **content property**: Each child is bigger than the parent
  - ▶ The **structure property**: Tree is maximally balanced

# Heap operations

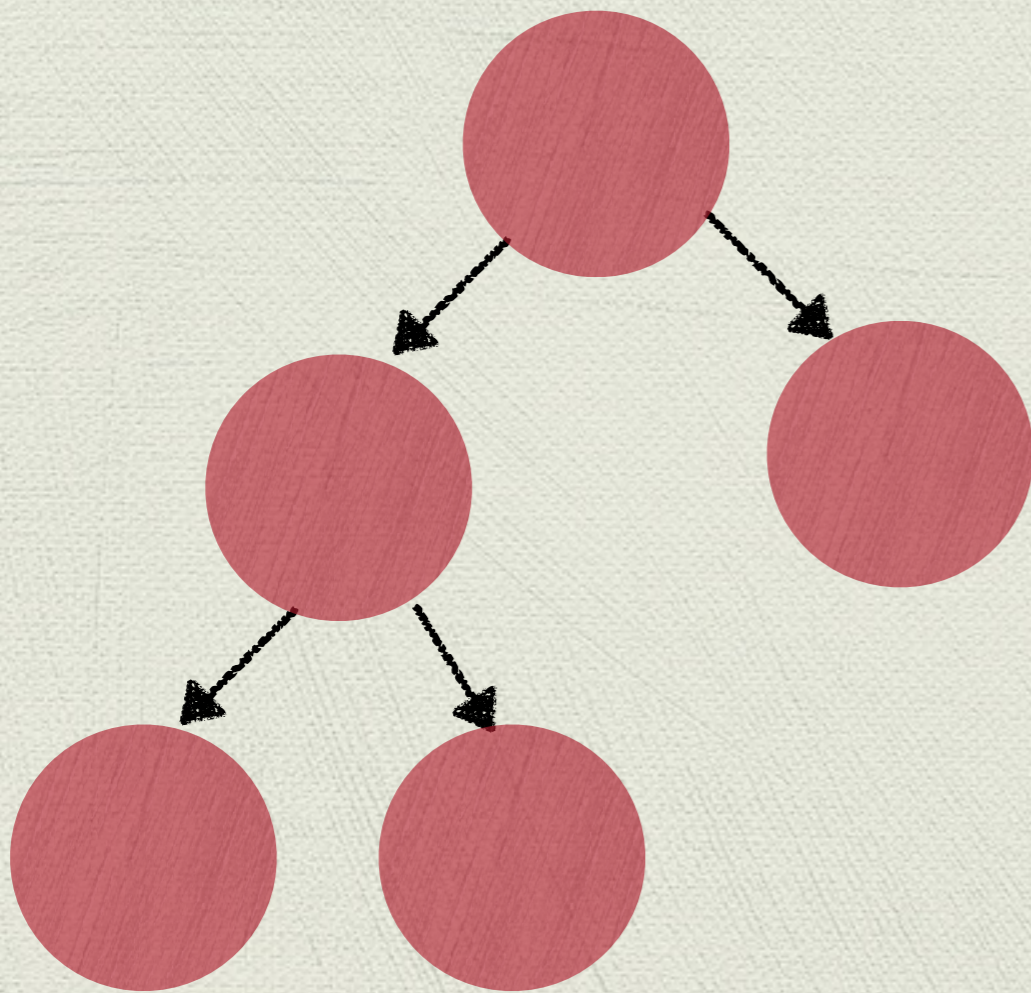
- ◆ Heap needs two operations
  - ▶ `add(Comparable c)`
  - ▶ `Comparable extractMin()`

# Adding to a heap

- ◆ When we add, we have to make sure to maintain the properties
- ◆ The structural property is the stricter of the two policies, so let's start with it

# Adding to a heap

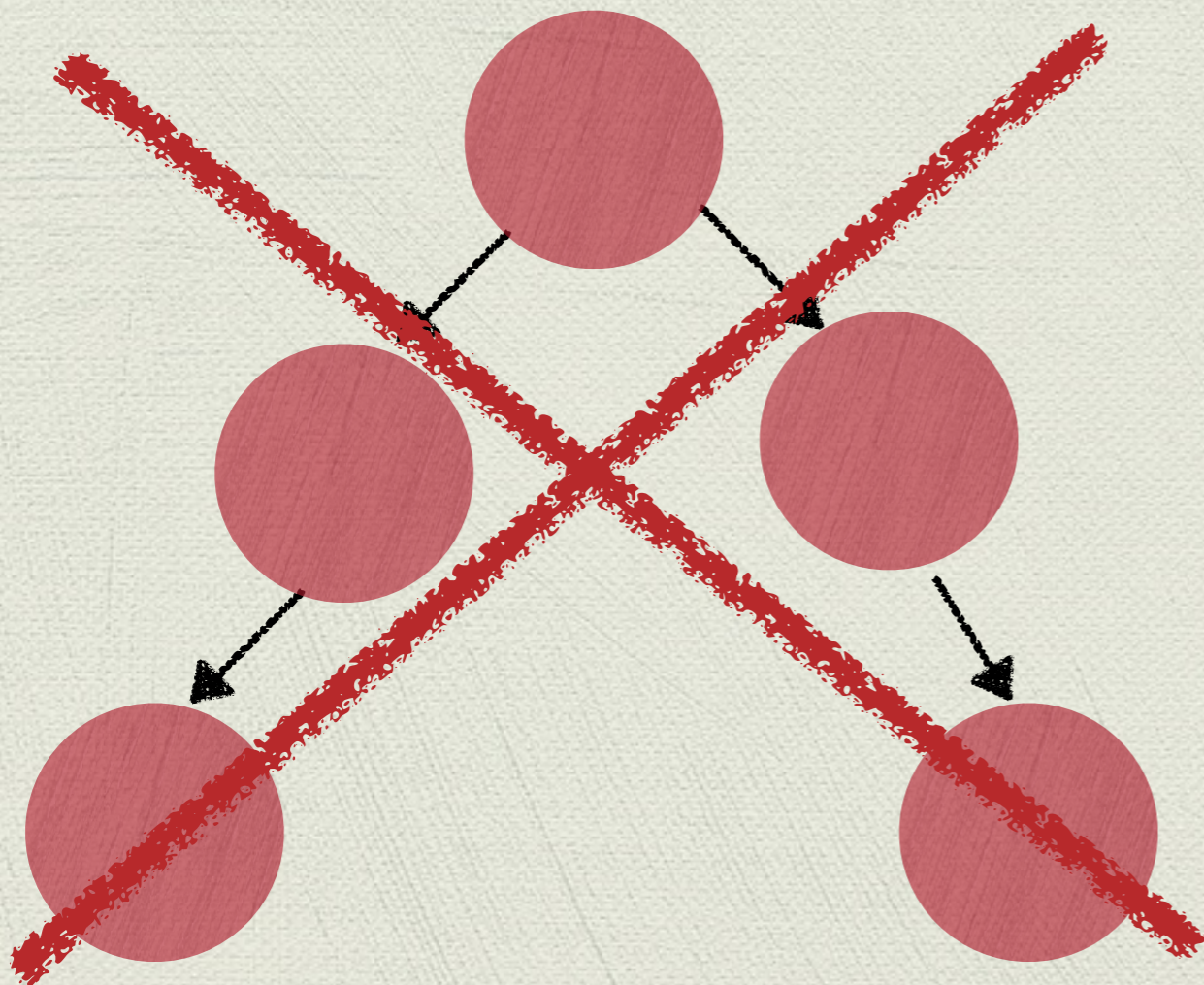
- ◆ There is only one possible shape for a heap with  $N$  nodes



The heap with 5 nodes always looks like this, regardless of content

# Adding to a heap

- ◆ There is only one possible shape for a heap with  $N$  nodes



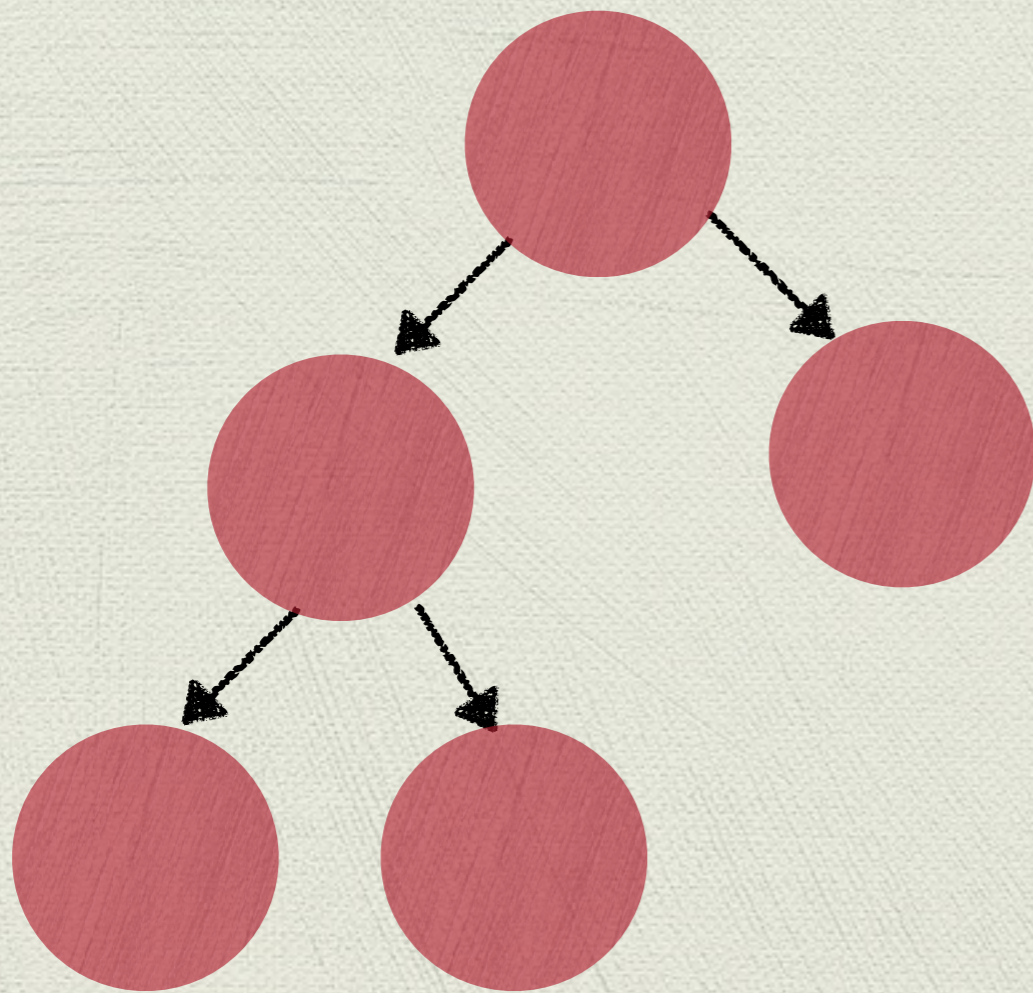
**Not possible! Not maximally balanced!**

# Adding to a heap

- ◆ Hence, the shape of a heap with  $N + 1$  nodes is completely predictable
- ◆ A new node always appears in the next open spot (gets appended to the end of the array)

# Adding to a heap

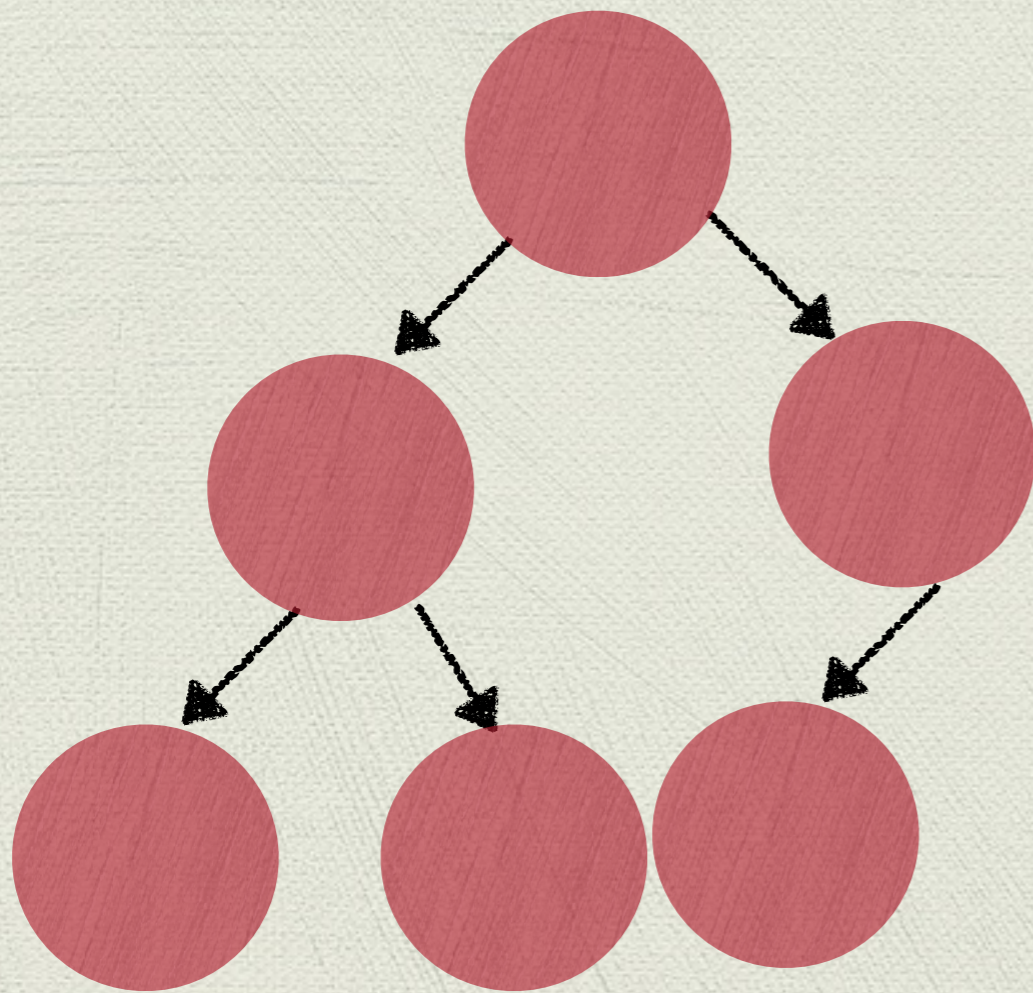
Heap with 5





# Adding to a heap

Heap with 6

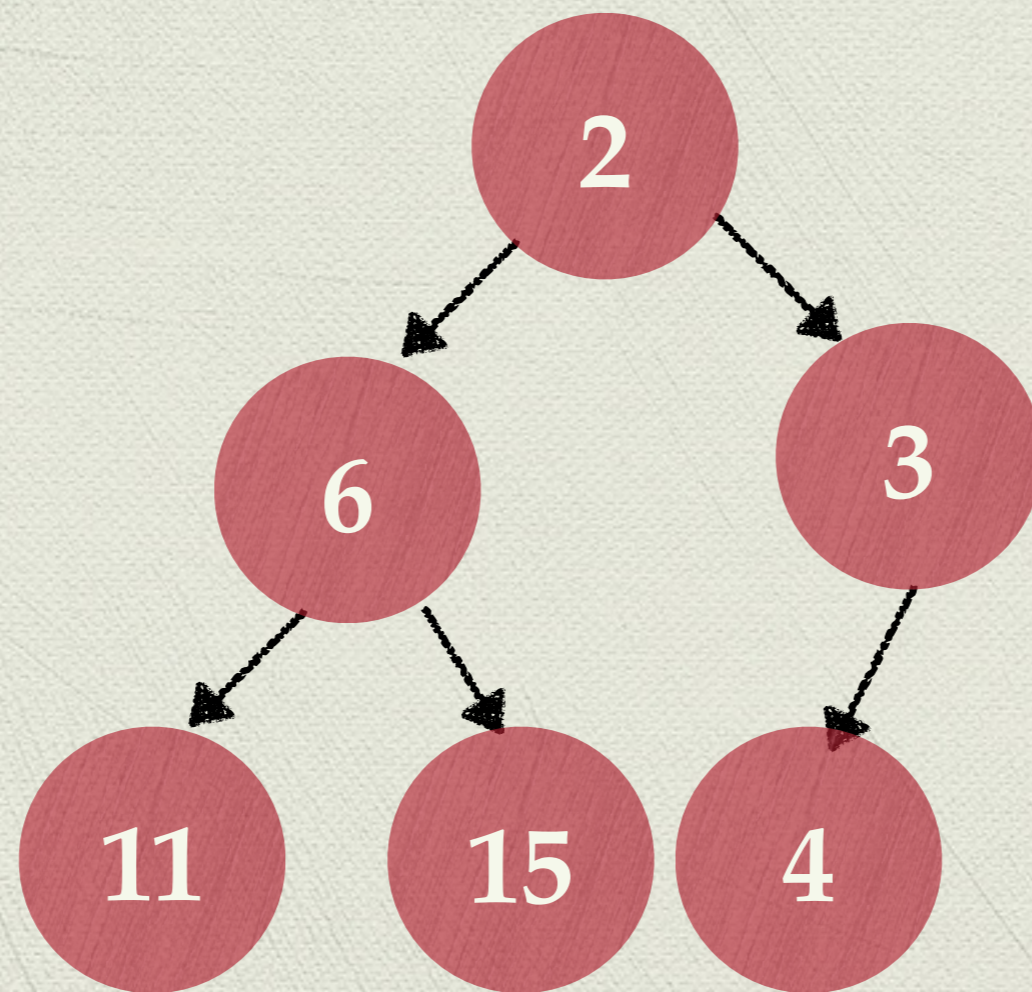


# Adding to a heap

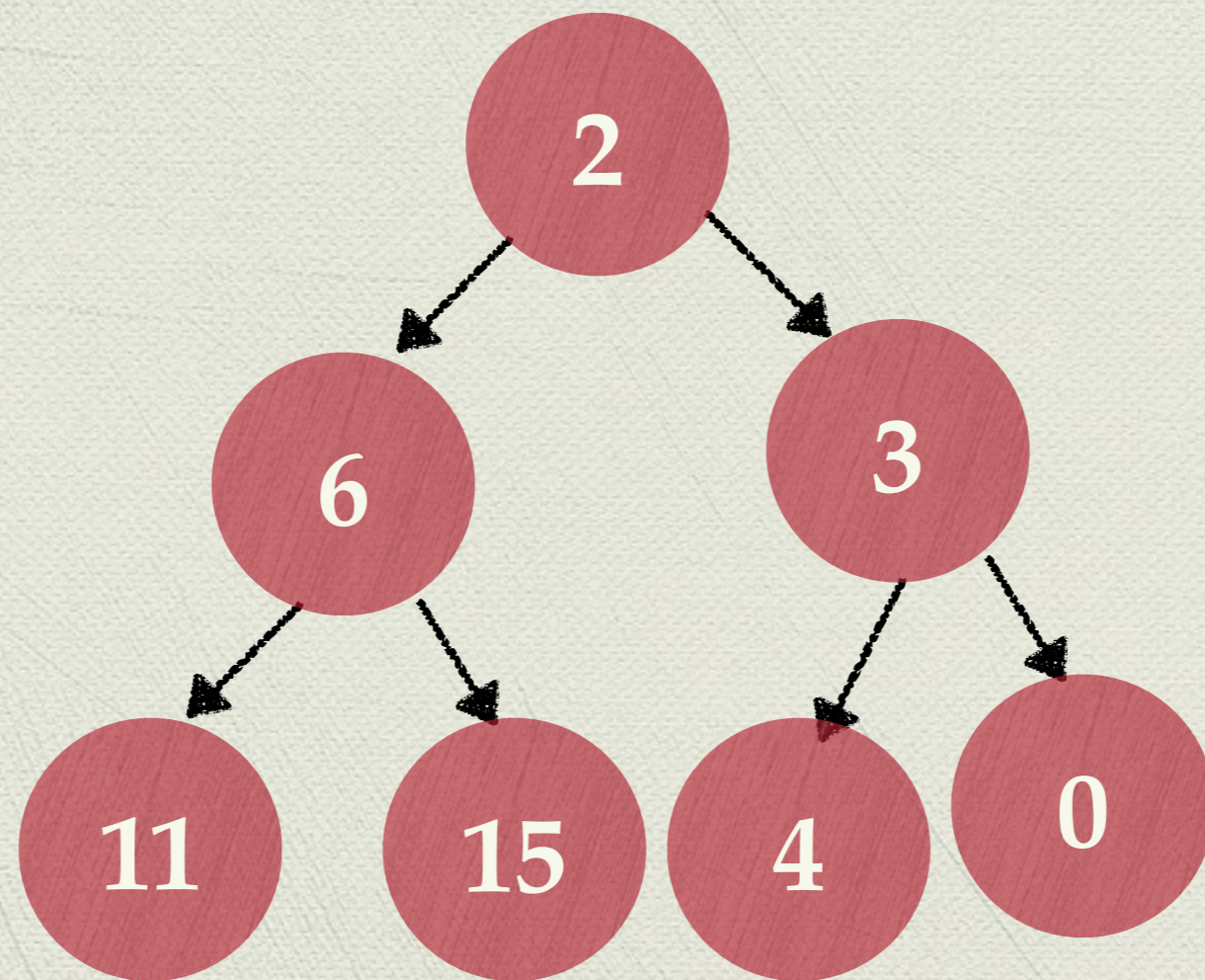
- ◆ So, when we add a new item to the heap, **we have no choice except to put it in the bottom right location**

# Adding to a heap

- ◆ Say we add 0 to this heap



We have no choice...!



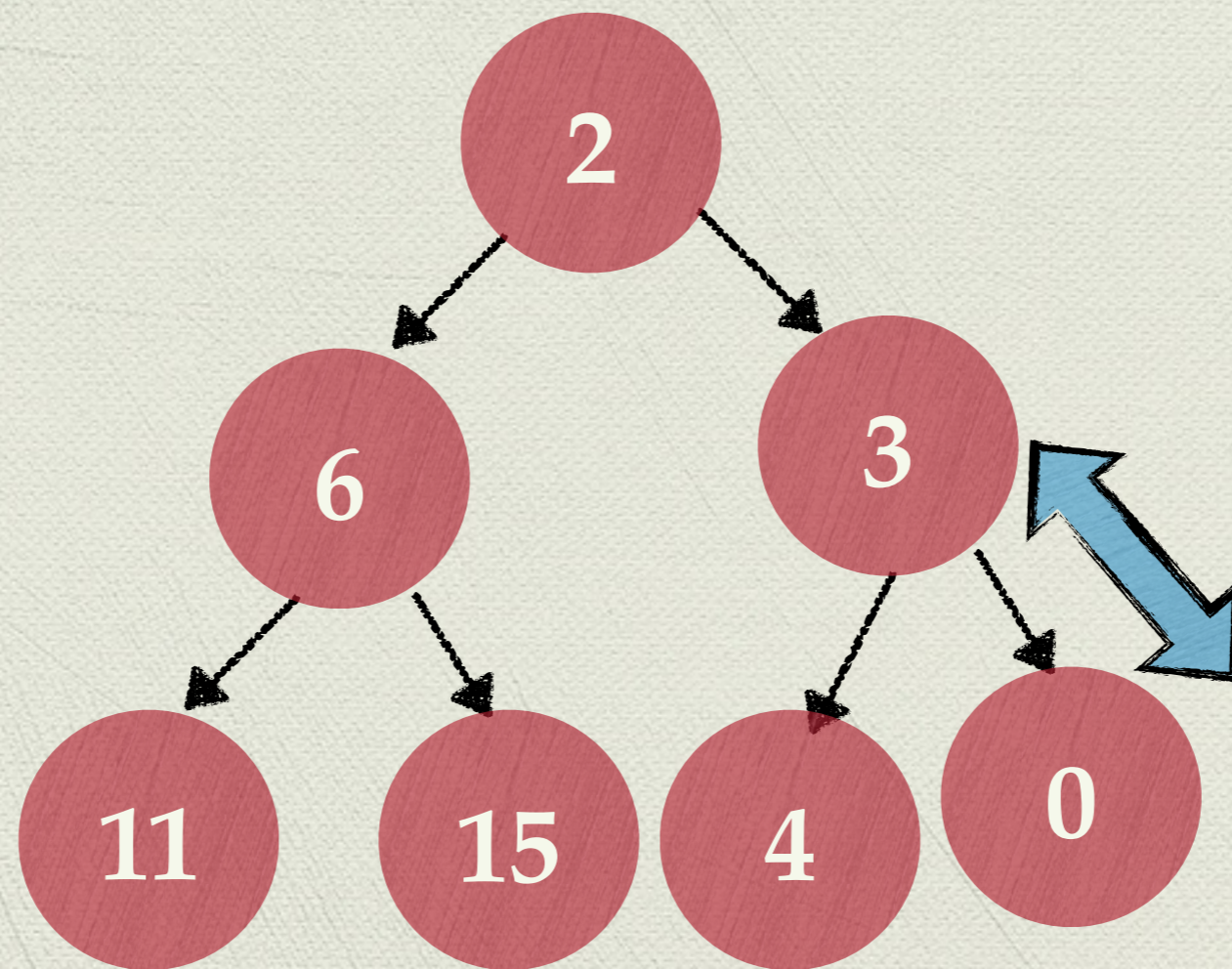
# Okay, but

- ◆ Now the content property is messed up

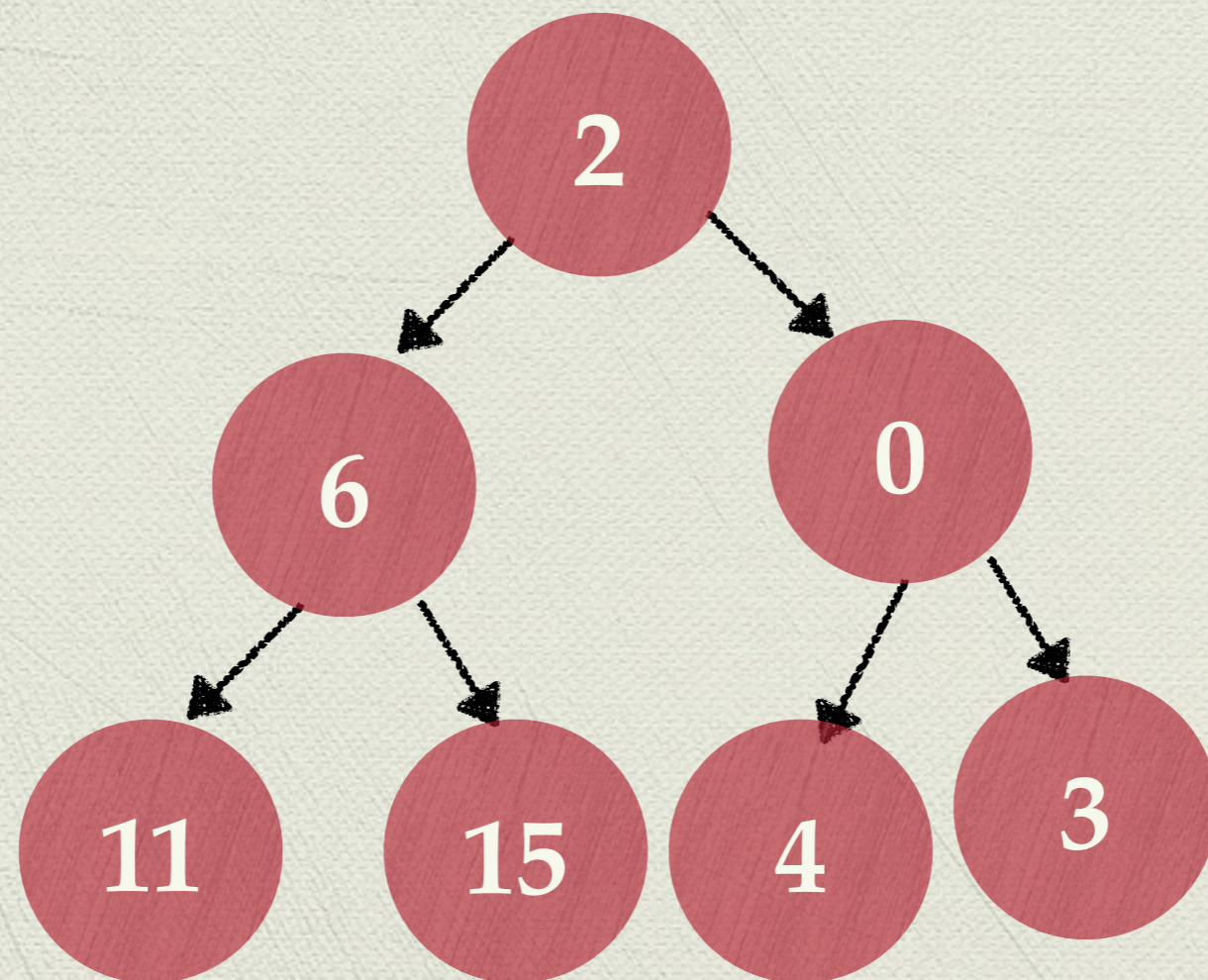
# Adding to a heap

- ◆ We want to fix the content property **without messing with the structure of the tree**
- ◆ We can do this by *swapping* the values of nodes until we're good

# Swap!

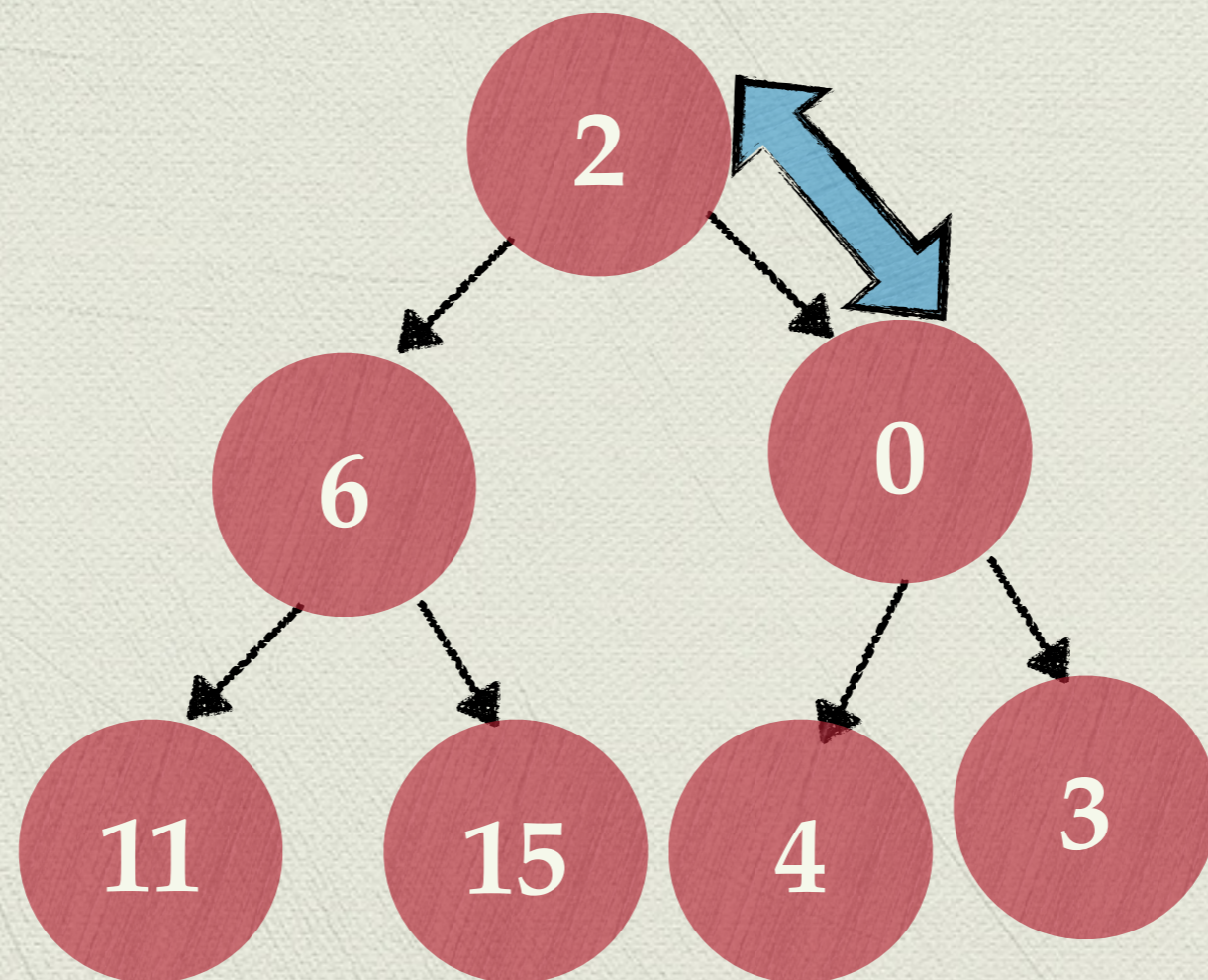


Okay, but still not good enough...

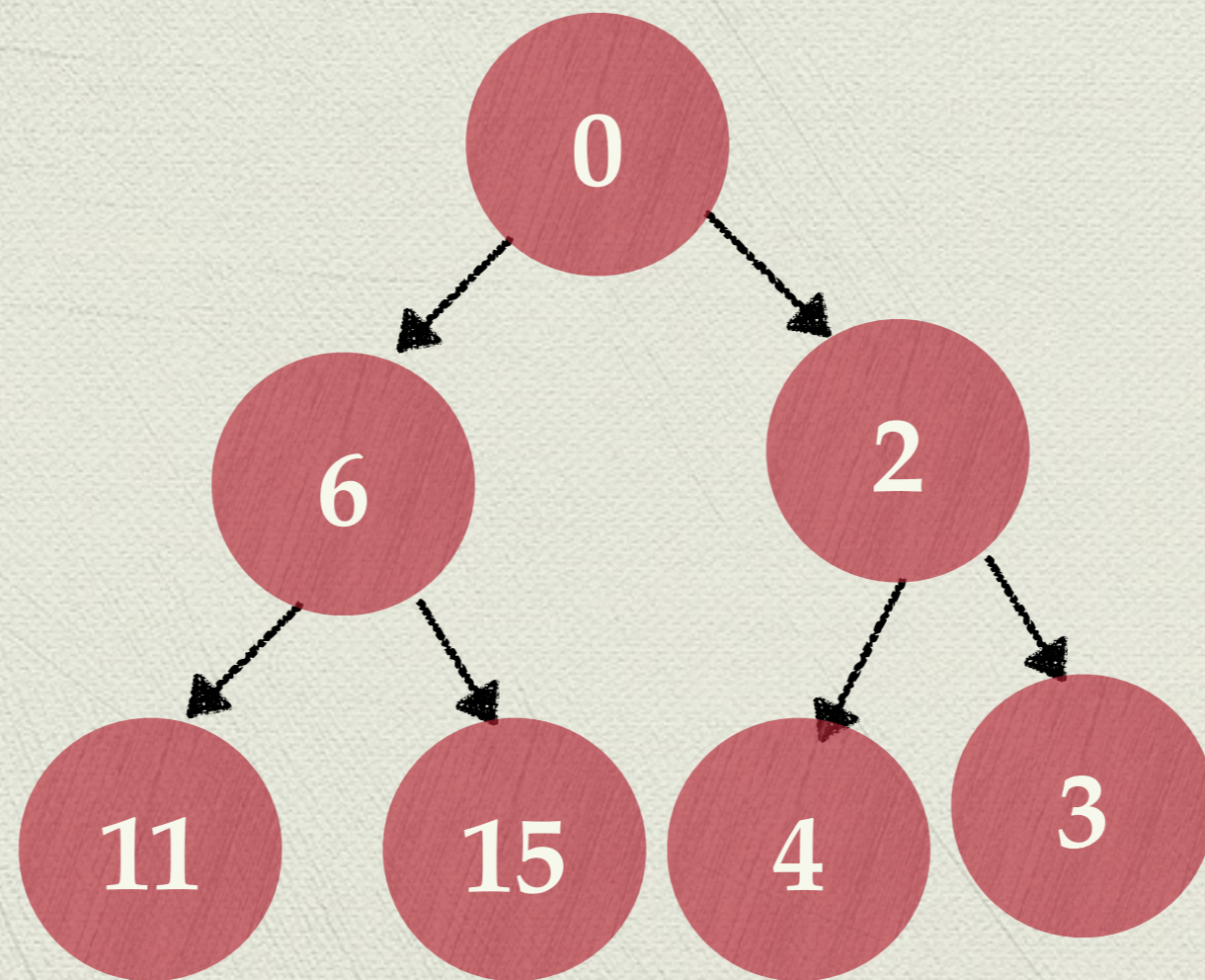




# Swap!!



# Now we're good



# Bubbling up

- ◆ After appending to the end, swap the value of the node up the tree until the content property is satisfied (this is called **bubbling up**)
- ◆ This is the full story with add
- ◆ (Easier than an AVL tree, right?)

# Heap operations

- ◆ `void add(Comparable c):` append an item to the end, then swap it up the tree until okay:  $O(\log N)$ , potentially have to swap all the way to the top
- ◆ `Comparable extractMin()`

# Extract min

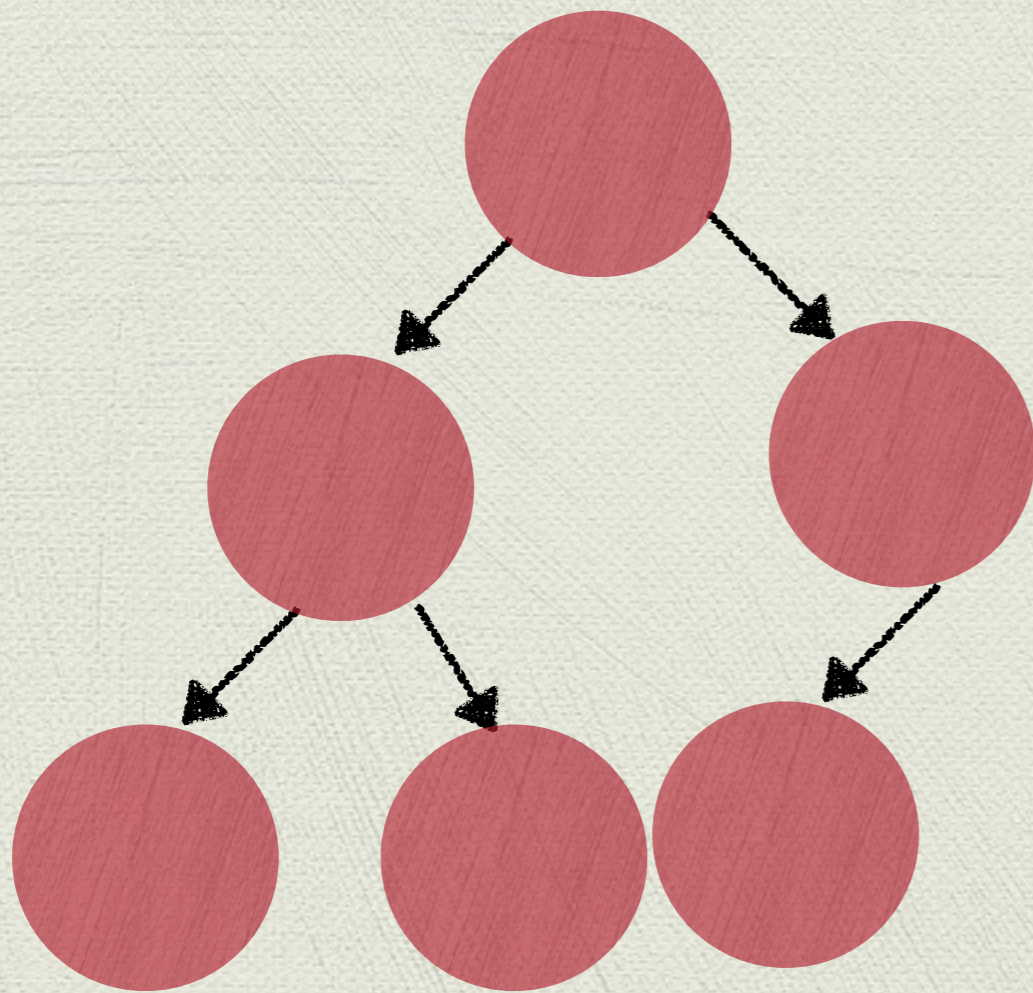
- ◆ It's easy to figure out the value of the min item in the heap
- ◆ It's just the root (position 1 in the array)
- ◆ But how do we take it out?

# Extract min

- ◆ The shape of a heap with  $N - 1$  nodes is completely predictable

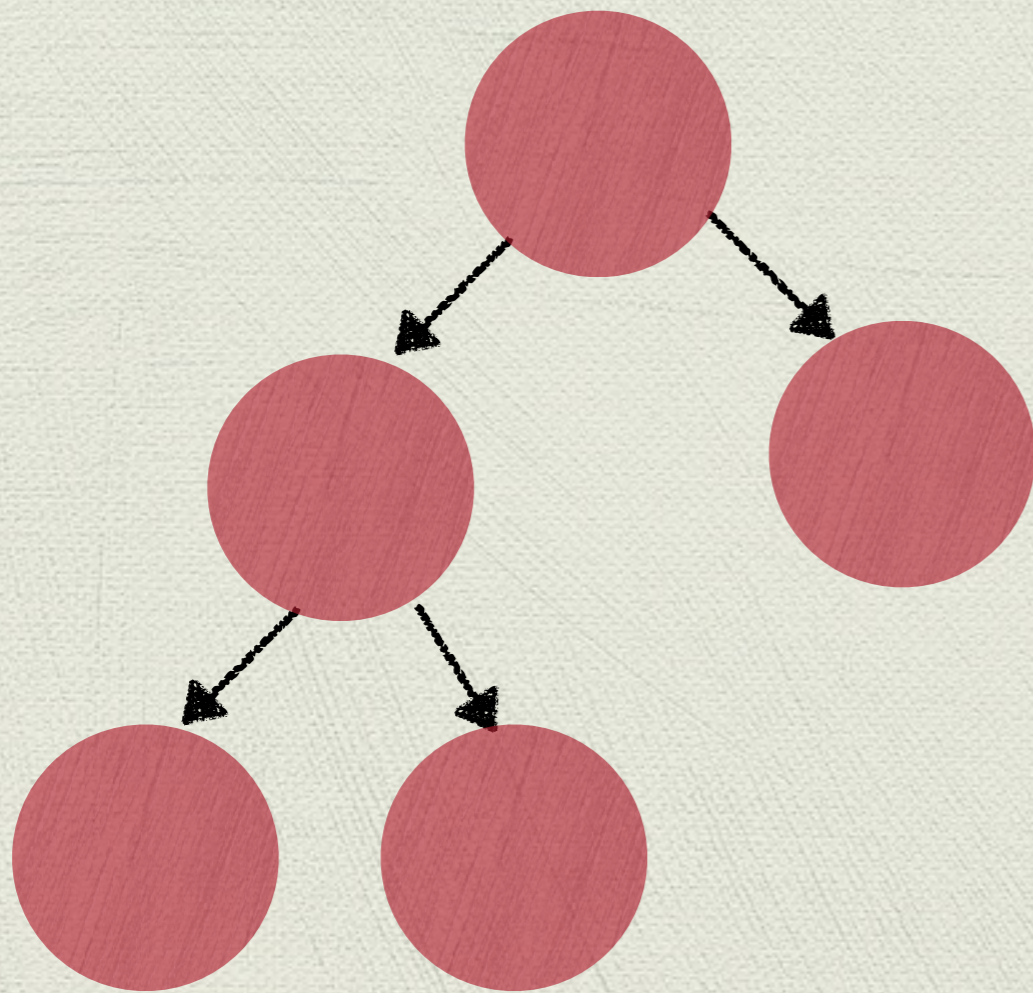
# Adding to a heap

Heap with 6



# Adding to a heap

Heap with 5

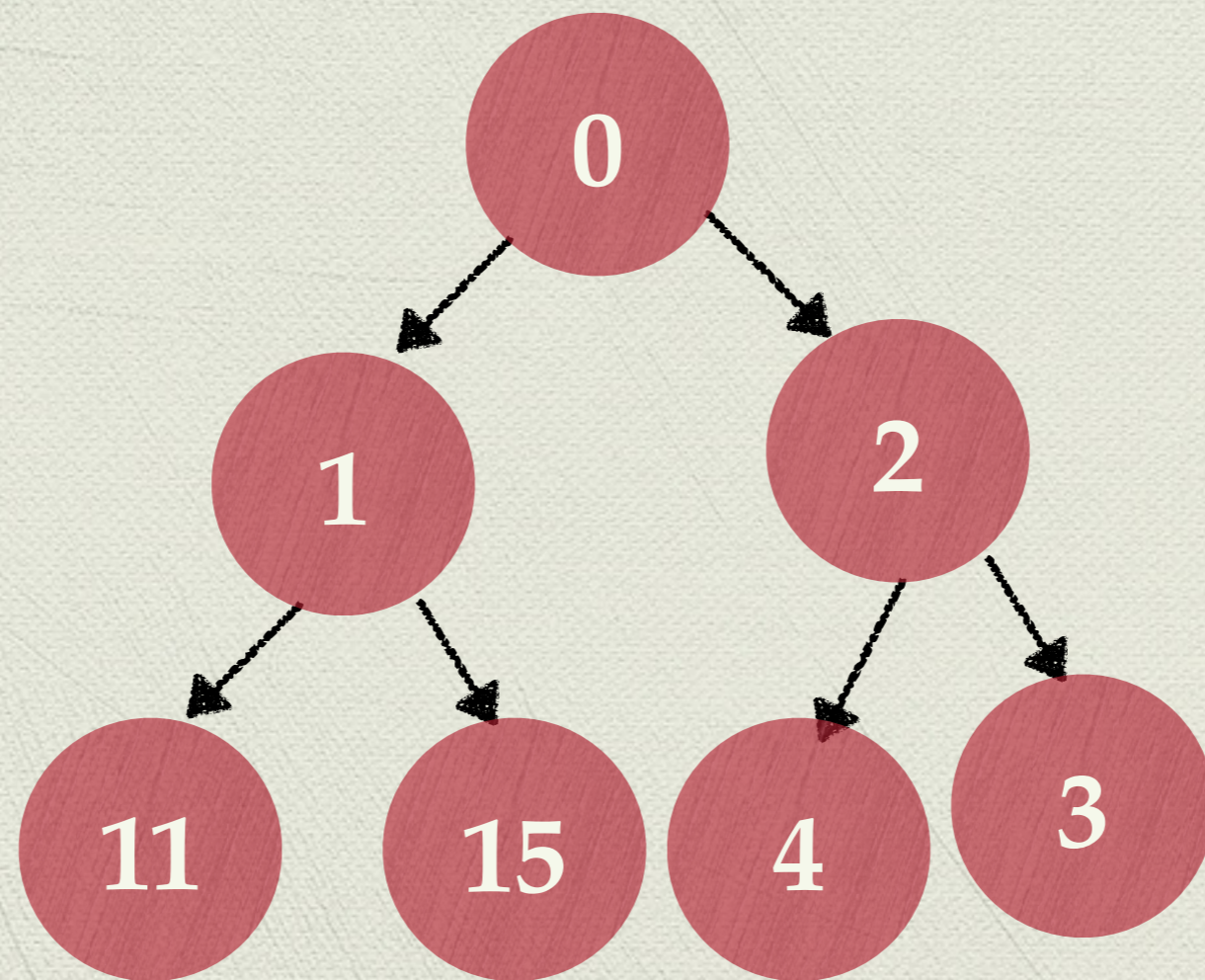




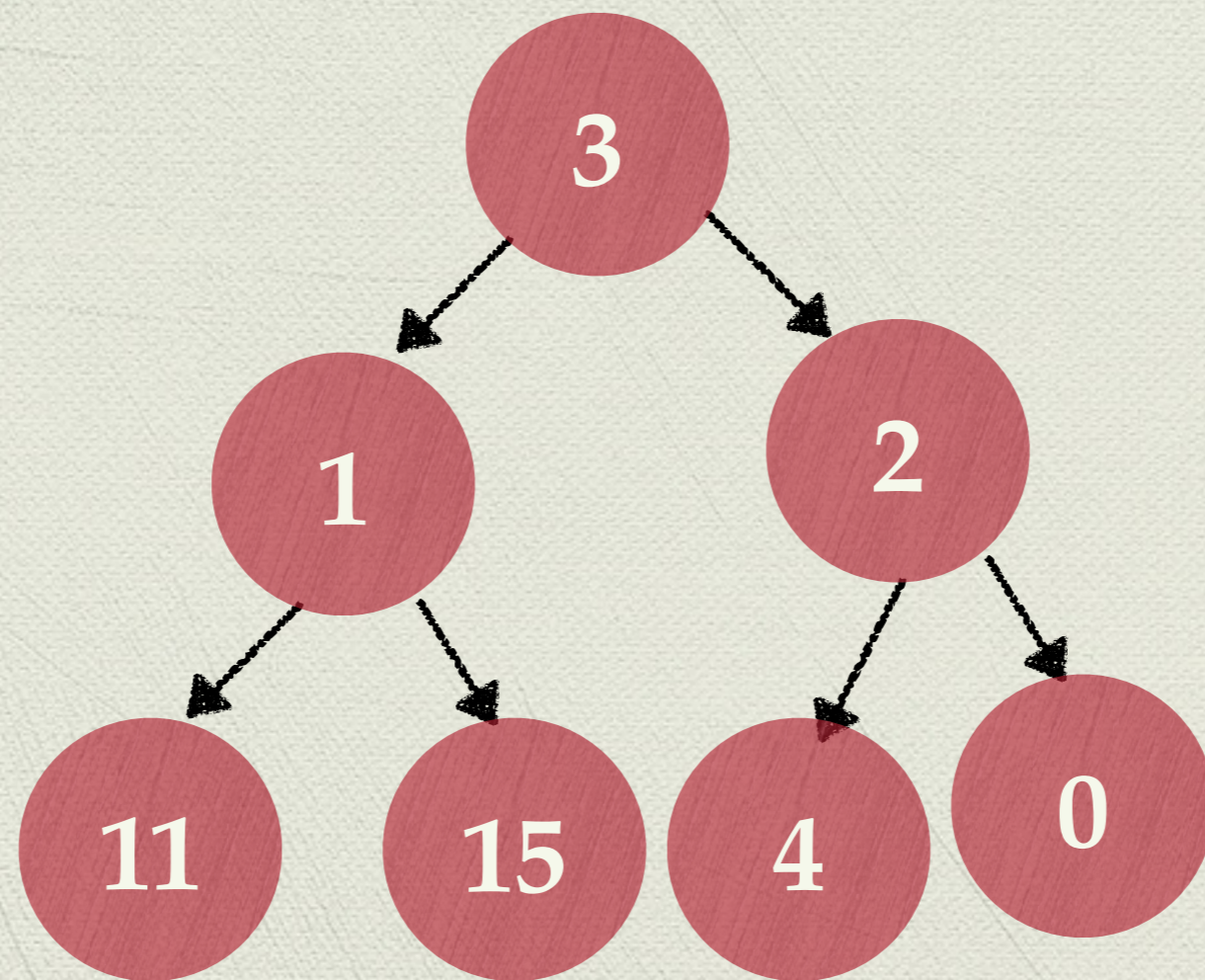
# Extract min

- ◆ We have no choice except to remove the bottom right element
- ◆ But that's not the one we want to remove! We want to remove the top!
- ◆ So: First swap the top element with the bottom, then remove

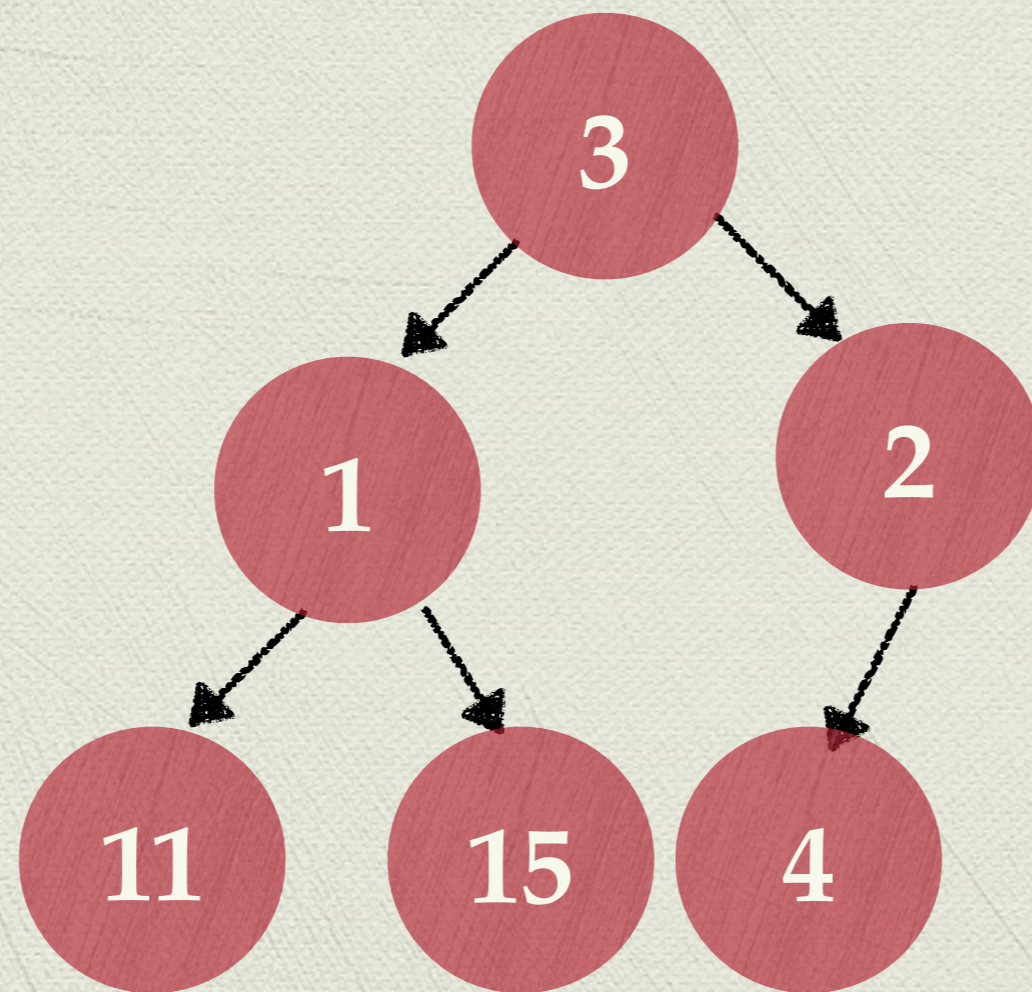
# Let's remove min



# First swap

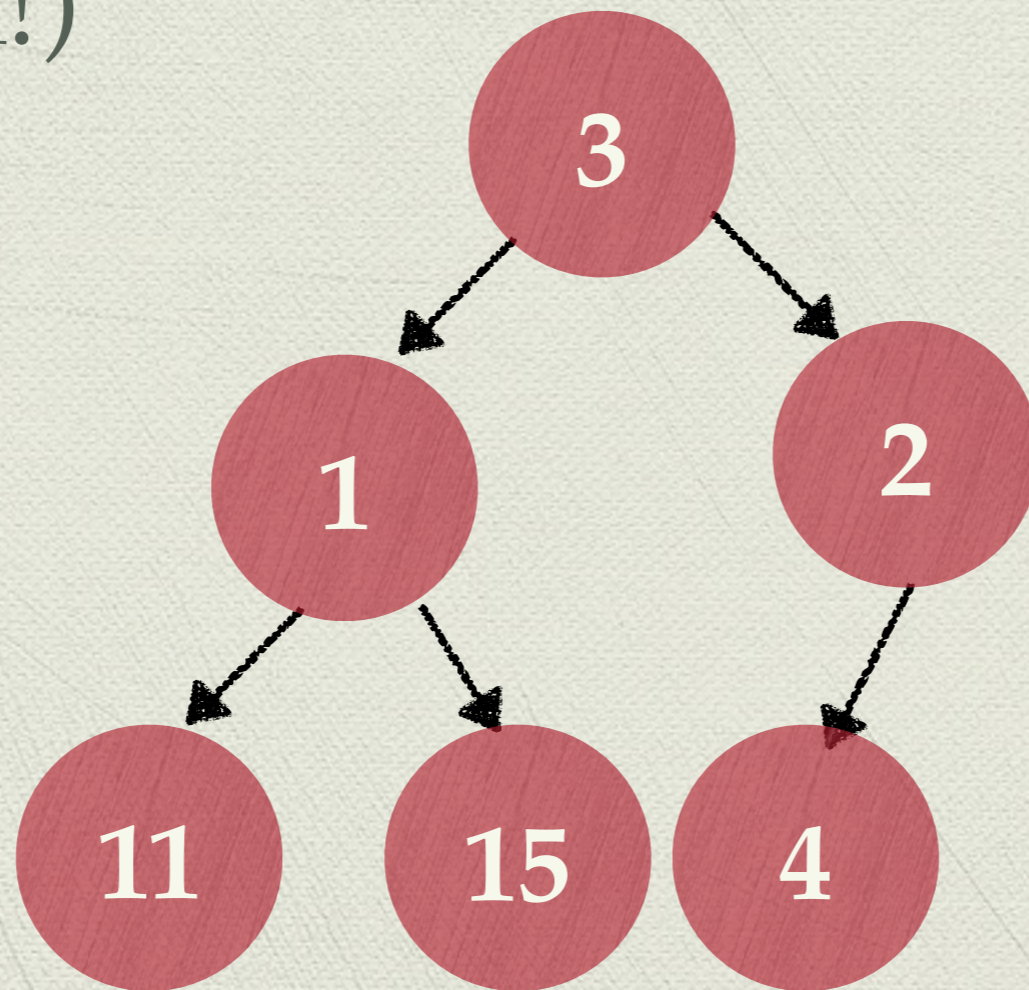


# Then take off



# Great, but

- ◆ Now the content property is messed up
- ◆ (not again!)

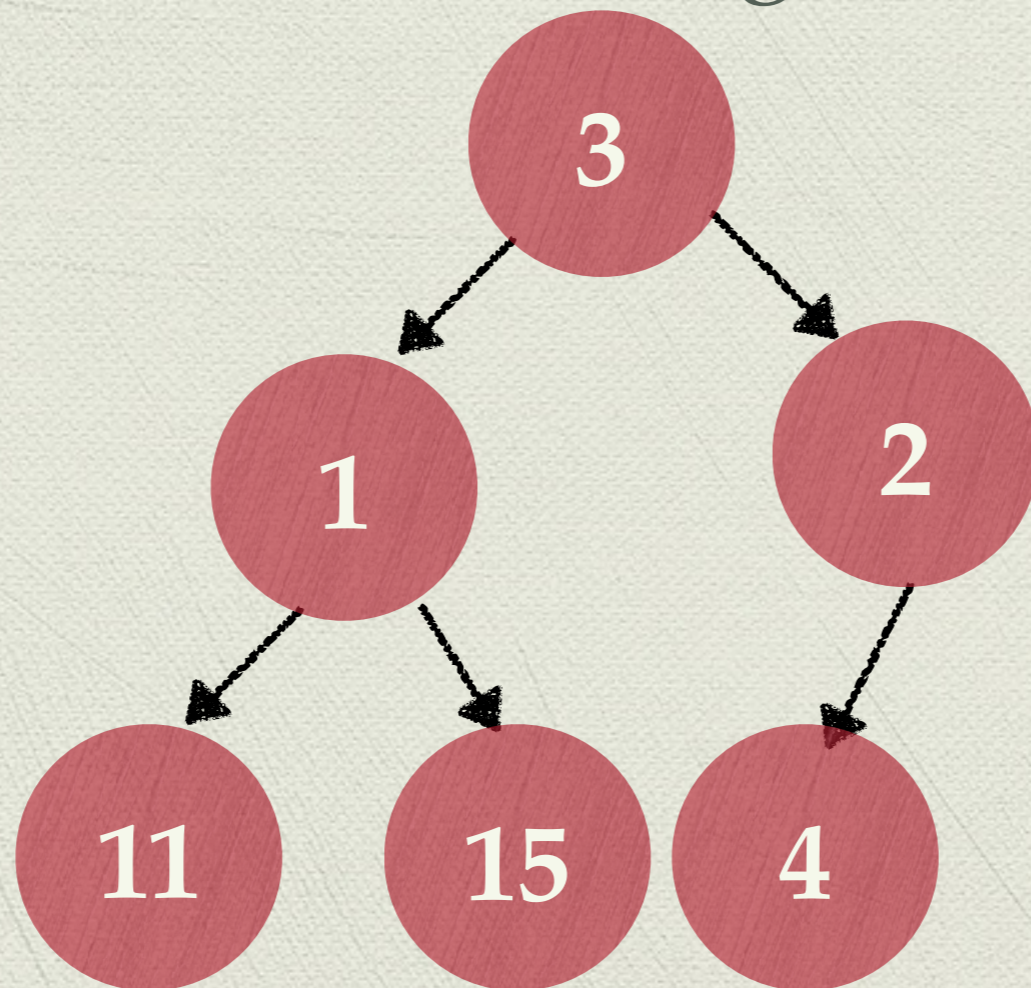


# Extract min

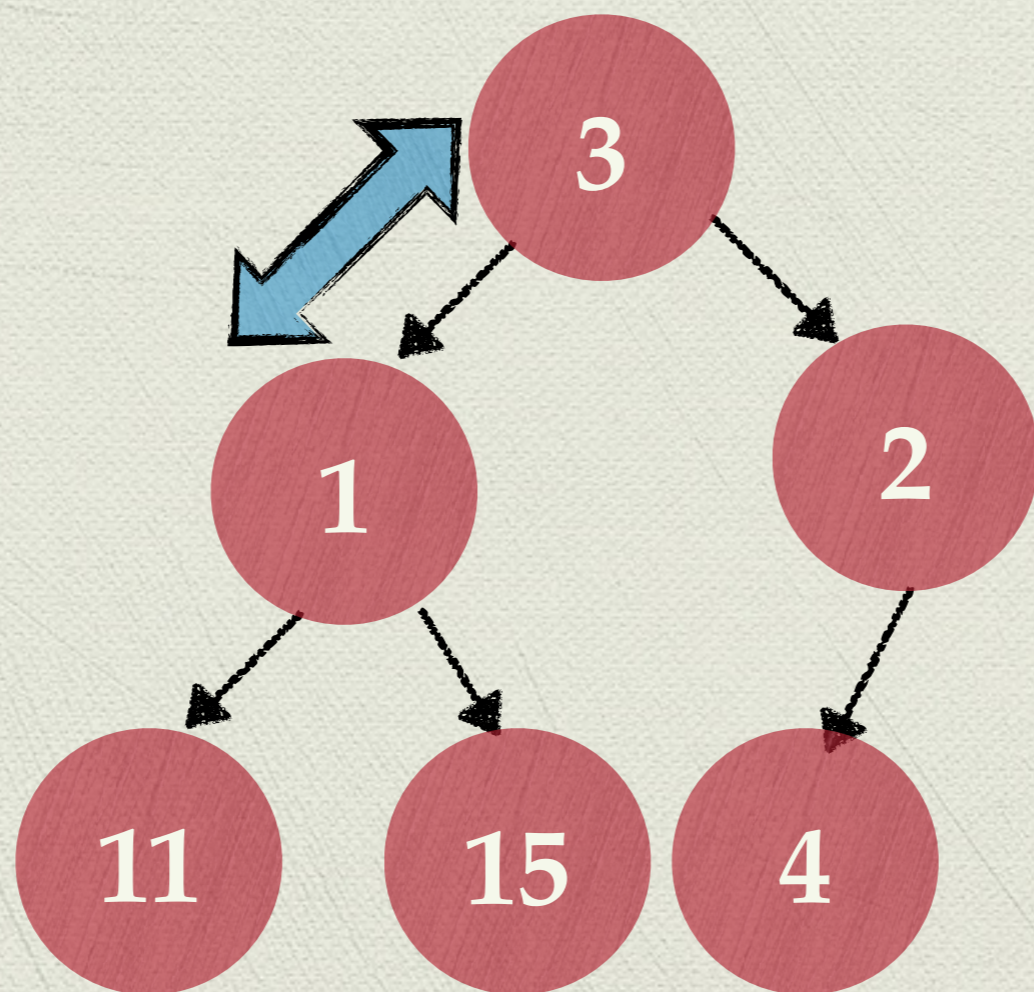
- ◆ We have to fix the content property without messing with the structure
- ◆ Back to swapping!
- ◆ This time, we'll swap the new top down the tree

# Extract min

- ◆ Which way do we swap the 3 down?
- ◆ Swap so the smallest thing ends up top

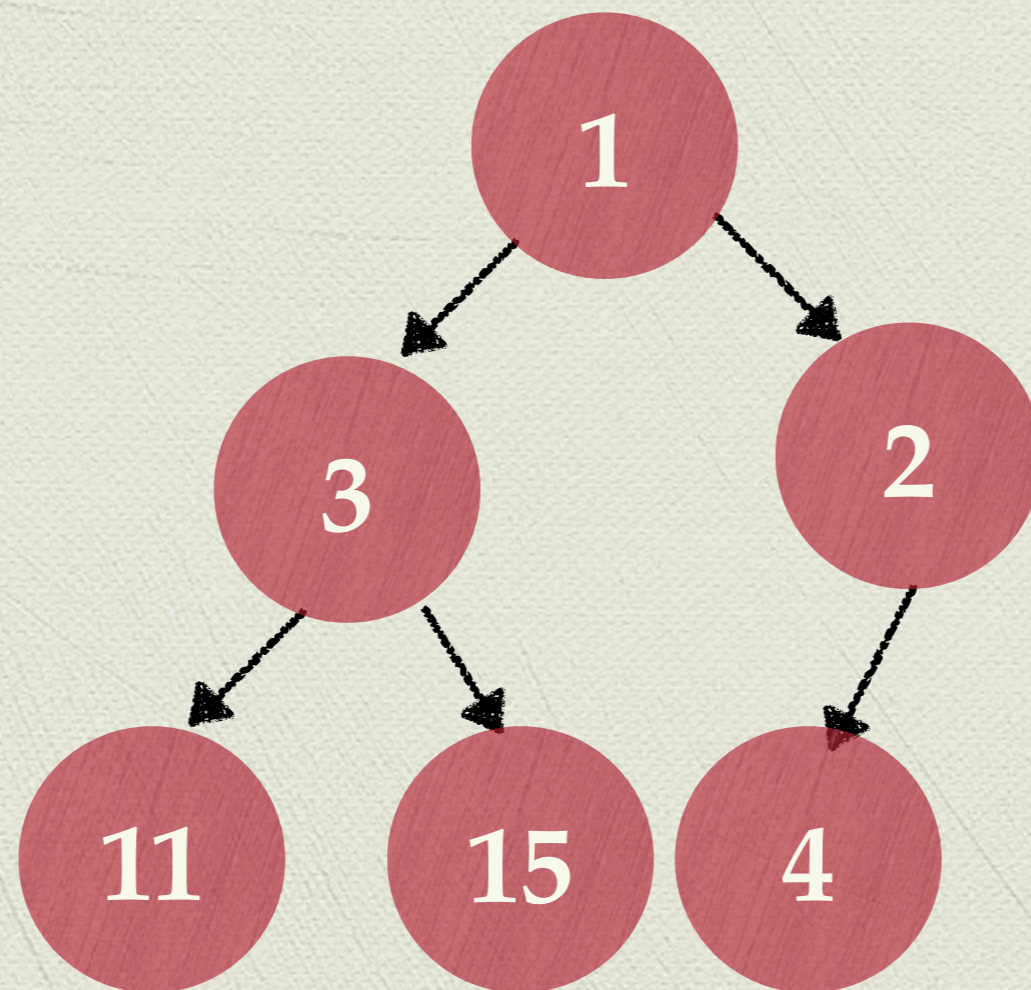


# Swap!!!





# Now we're good



# Bubbling down

- ◆ After swapping the top with the bottom, and taking off the bottom, **bubble down** the new top until it hits the correct spot
- ◆ This is the full story with extract min
- ◆ (Easier than an BST remove, right?)

# Heap operations

- ◆ `void add(Comparable c):` append an item to the end, then swap it up the tree until okay:  $O(\log N)$ , potentially have to swap all the way to the top
- ◆ `Comparable extractMin():` swap top and bottom, take off bottom, bubble down new top:  $O(\log N)$ , potentially have to swap all the way down

# Heap vs. BST

- ◆ Heap ultimately has the same asymptotic runtimes as BST for representing a priority queue
- ◆ But a heap is implemented with an array, which is far more memory efficient than nodes with tons of pointers
- ◆ And the most complicated operations in a heap are swapping values at array indices, which is super fast

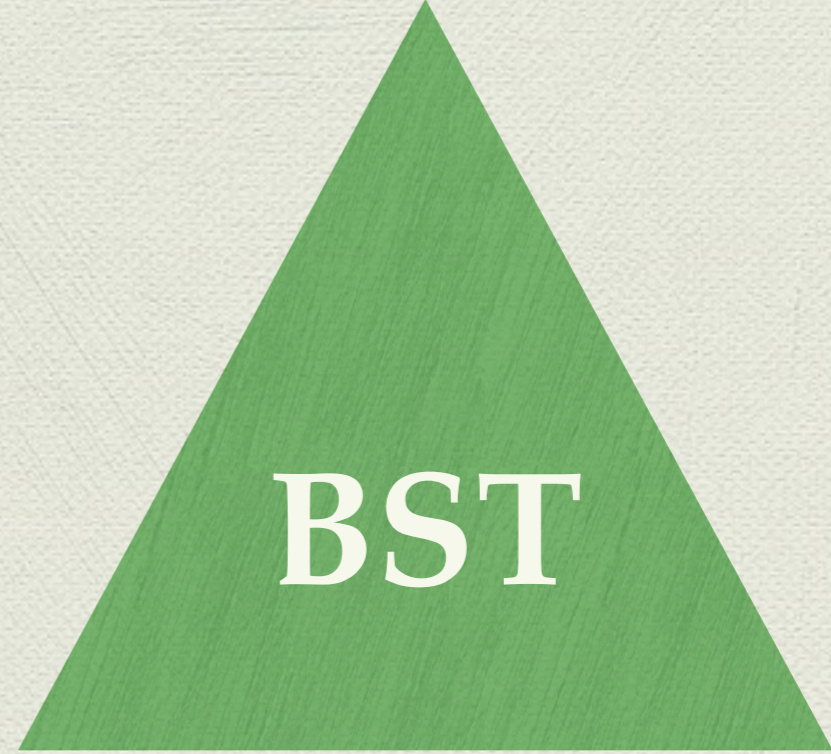
# Heap vs. BST: fun facts

small items



large items

medium items



small items

large items

# Heap vs. BST: fun facts

- ◆ **Heap is maximally balanced**
- ◆ **BST is usually almost balanced** (well, AVL tree is)

# Heap vs. BST: fun facts

- ◆ Heap is usually implemented as an **array tree**
- ◆ BST is usually implemented with **nodes**

# Heap vs. BST: fun facts

- ◆ Both can get/remove the min element in log time
- ◆ Both can add new items in log time
- ◆ Heap is slightly faster for priority queue



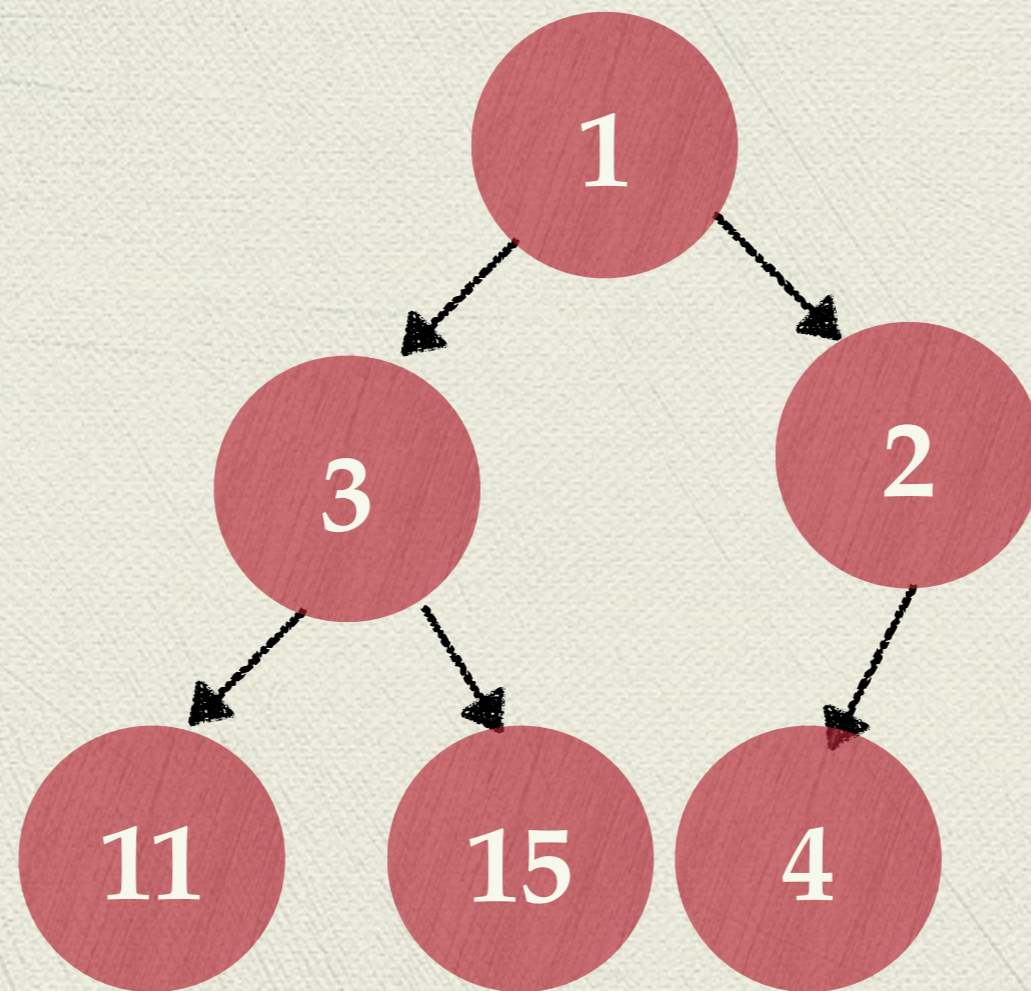
# Heap vs. BST: fun facts

- ◆ BST can find any item in the tree in log time
- ◆ Heap can find any item in the tree in...?

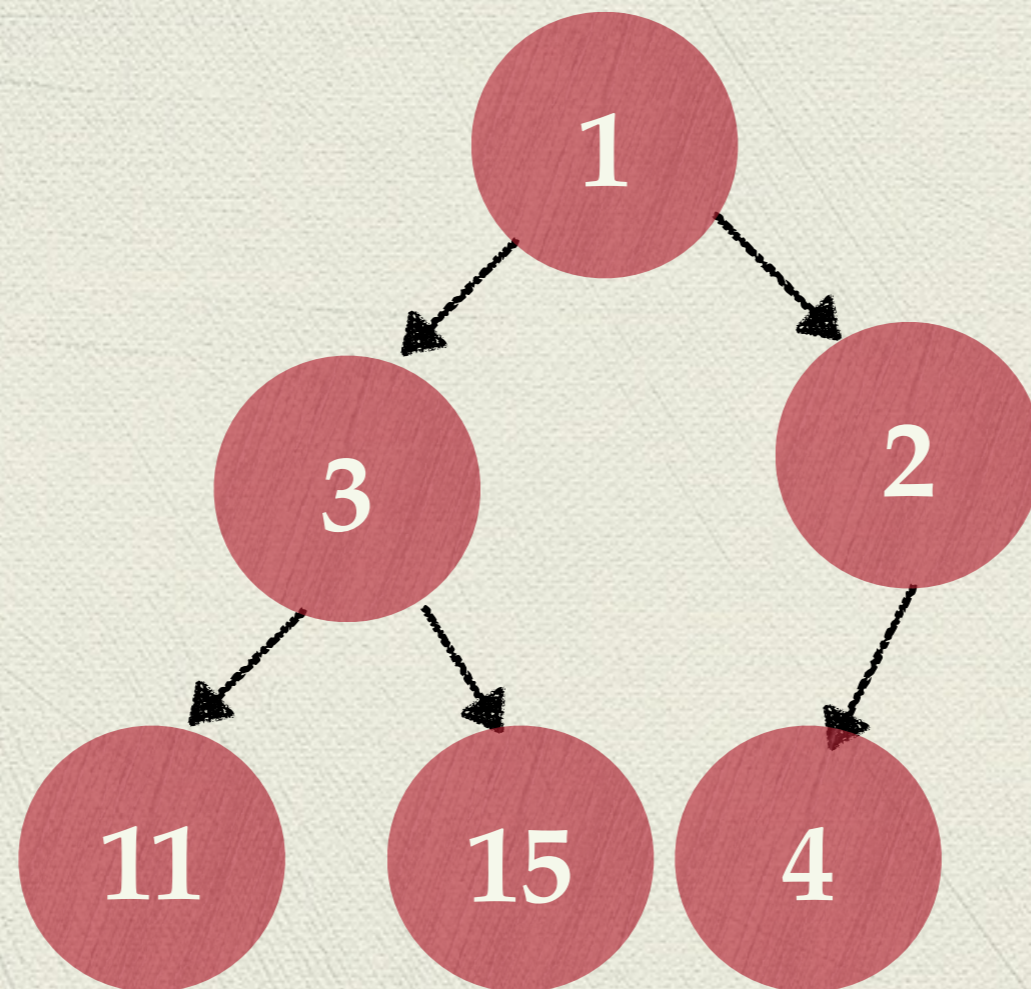
# Heap contains

- ◆ How do we write this method for a heap?
- ◆ `boolean contains(int item)`
- ◆ Uh oh.

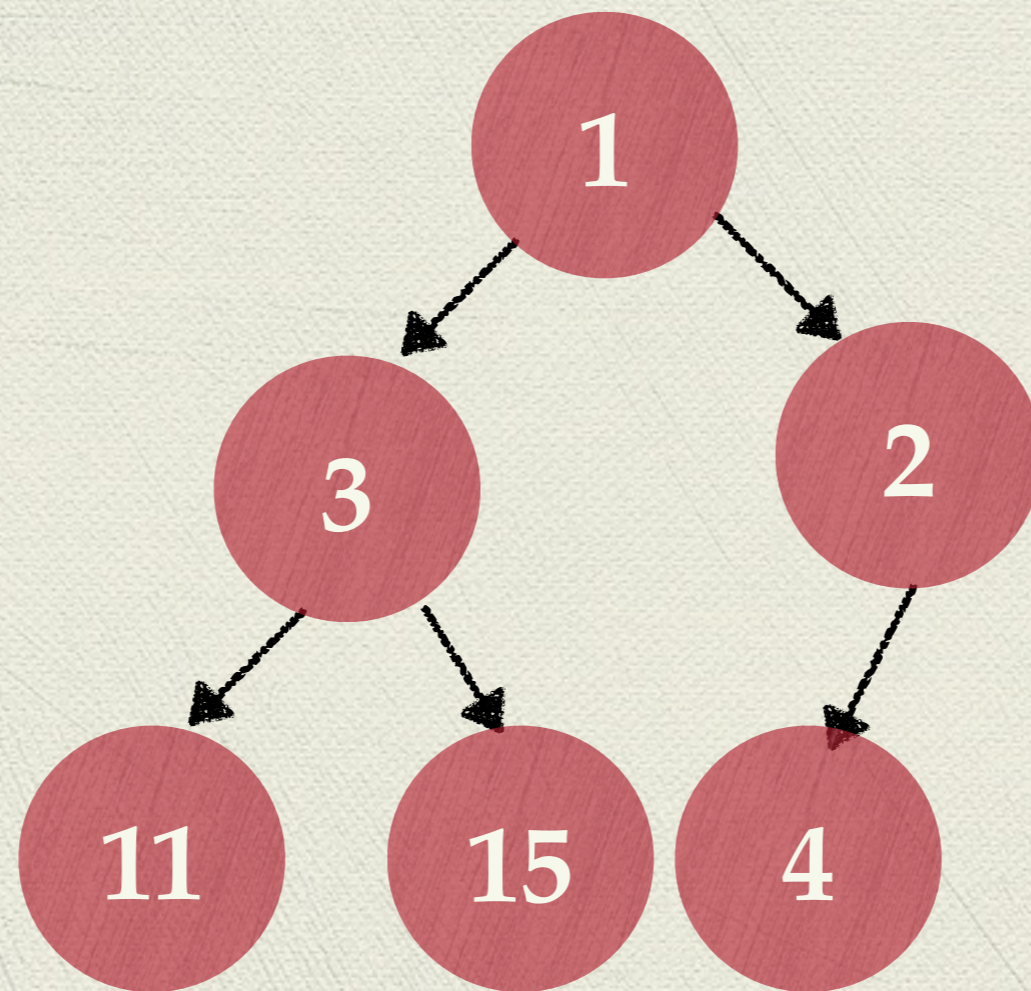
◆ Does the heap contain 4?



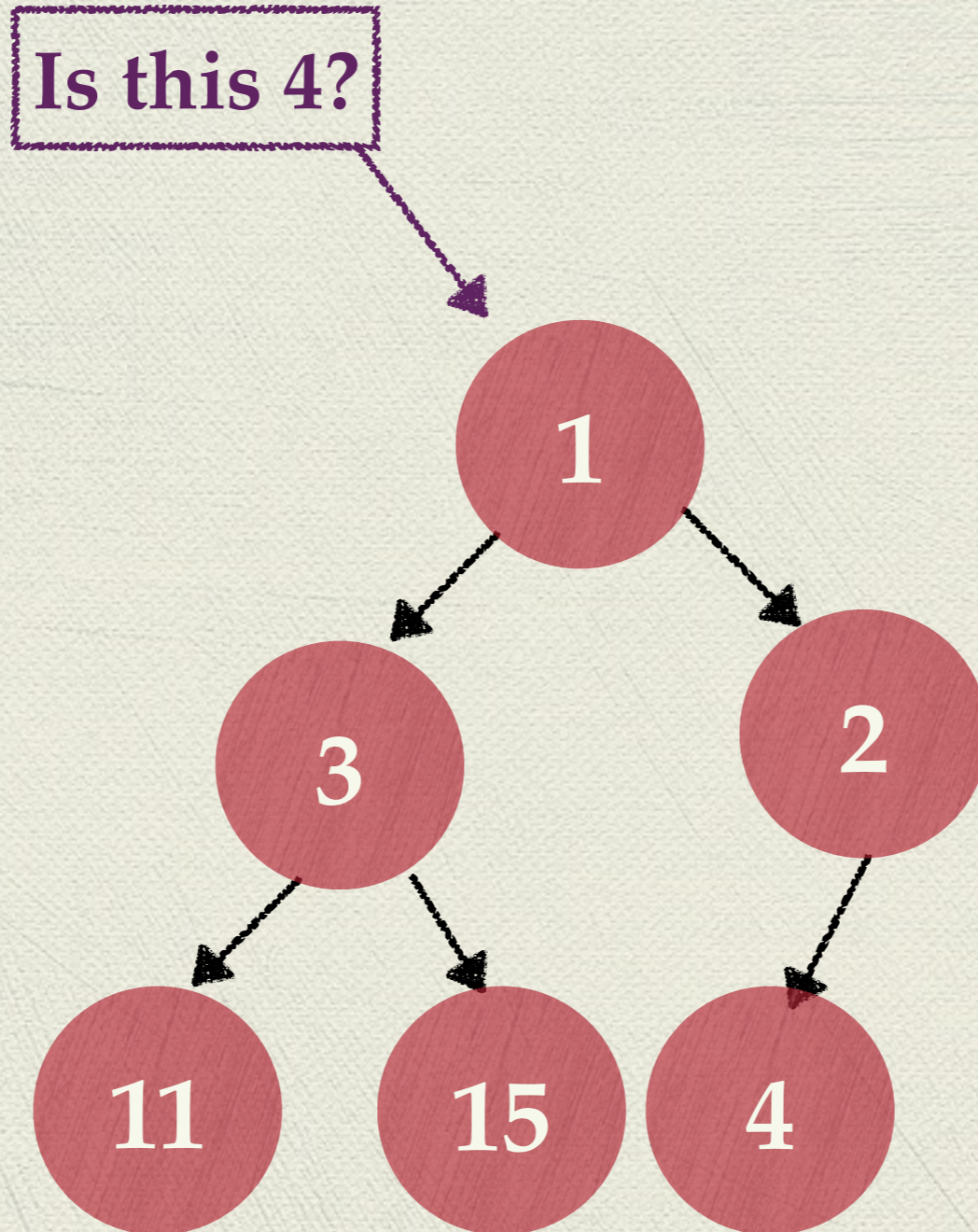
- ◆ Does the heap contain 4?
- ◆ Good question.



- ◆ Does the heap contain 4?
- ◆ Good question.
- ◆ Might as well start by iterating

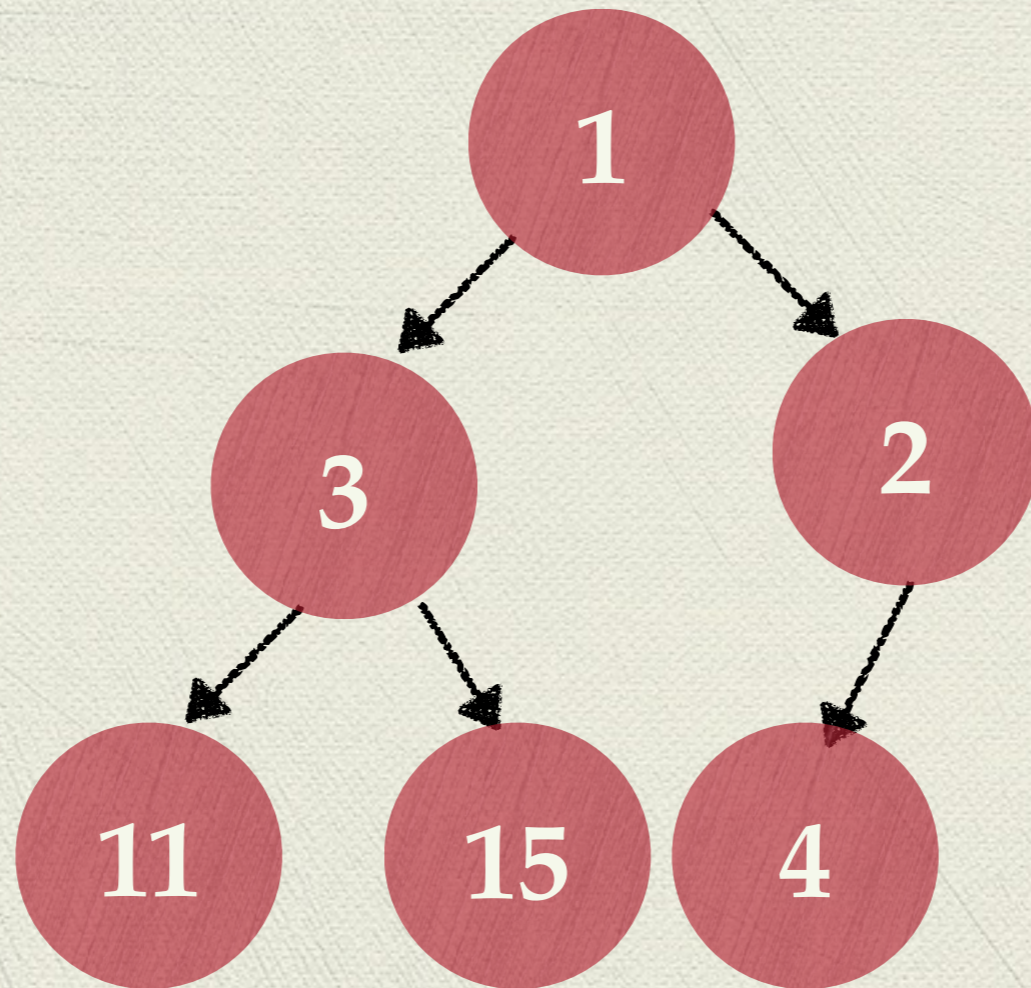


◆ Does the heap contain 4?



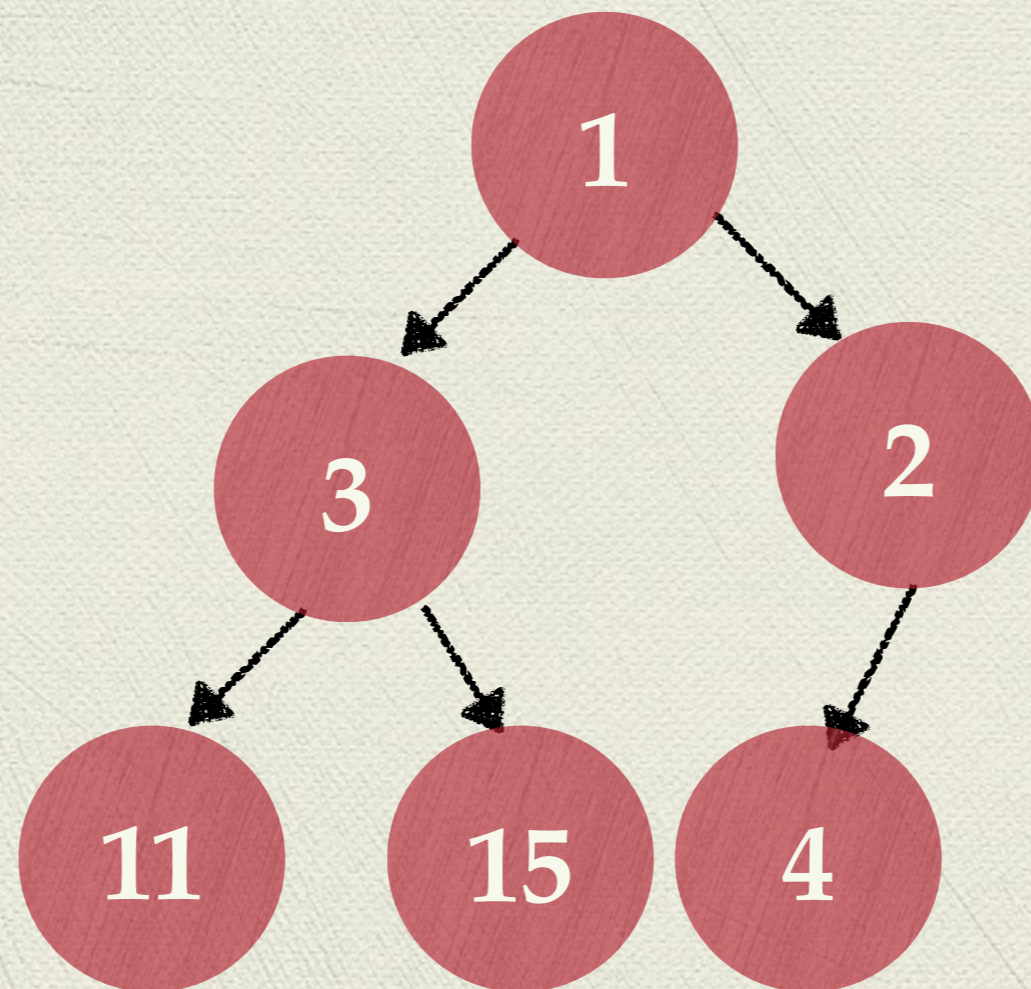
◆ Does the heap contain 4?

Nope



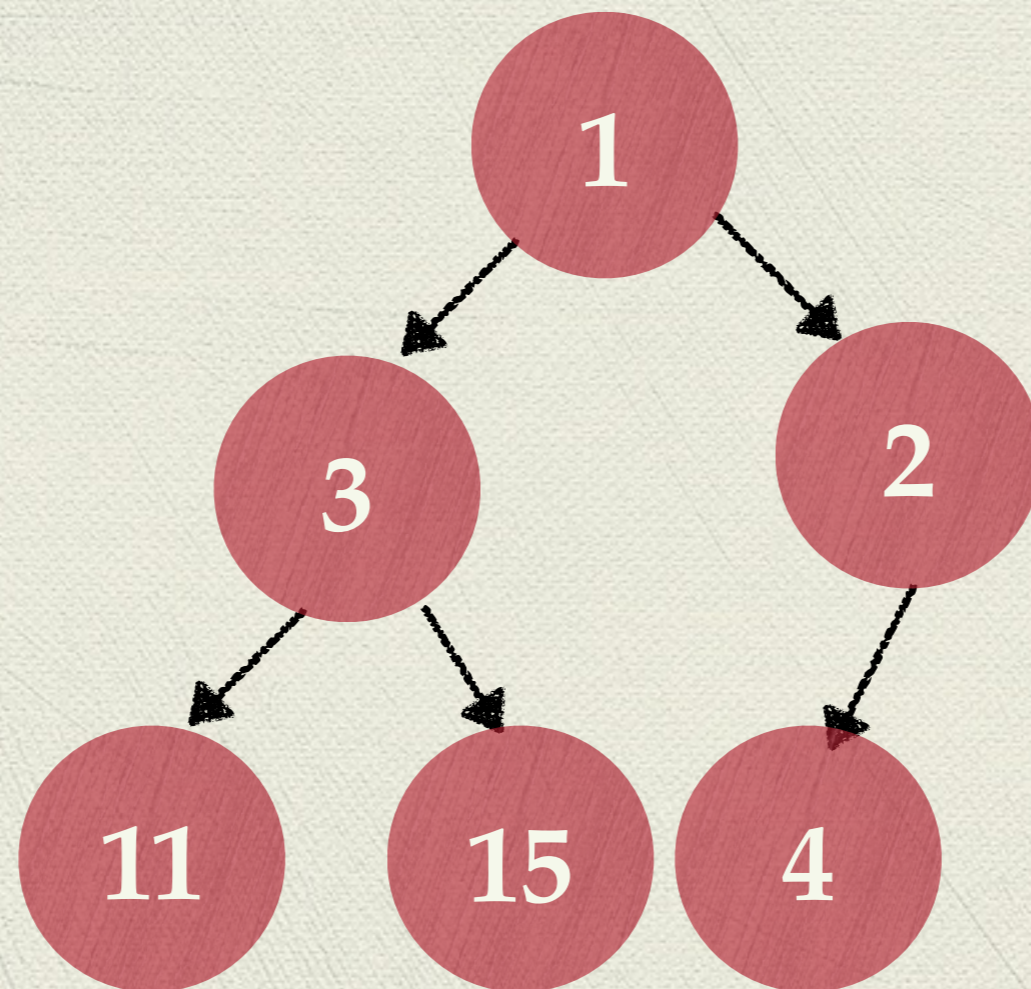
◆ Does the heap contain 4?

Now what?

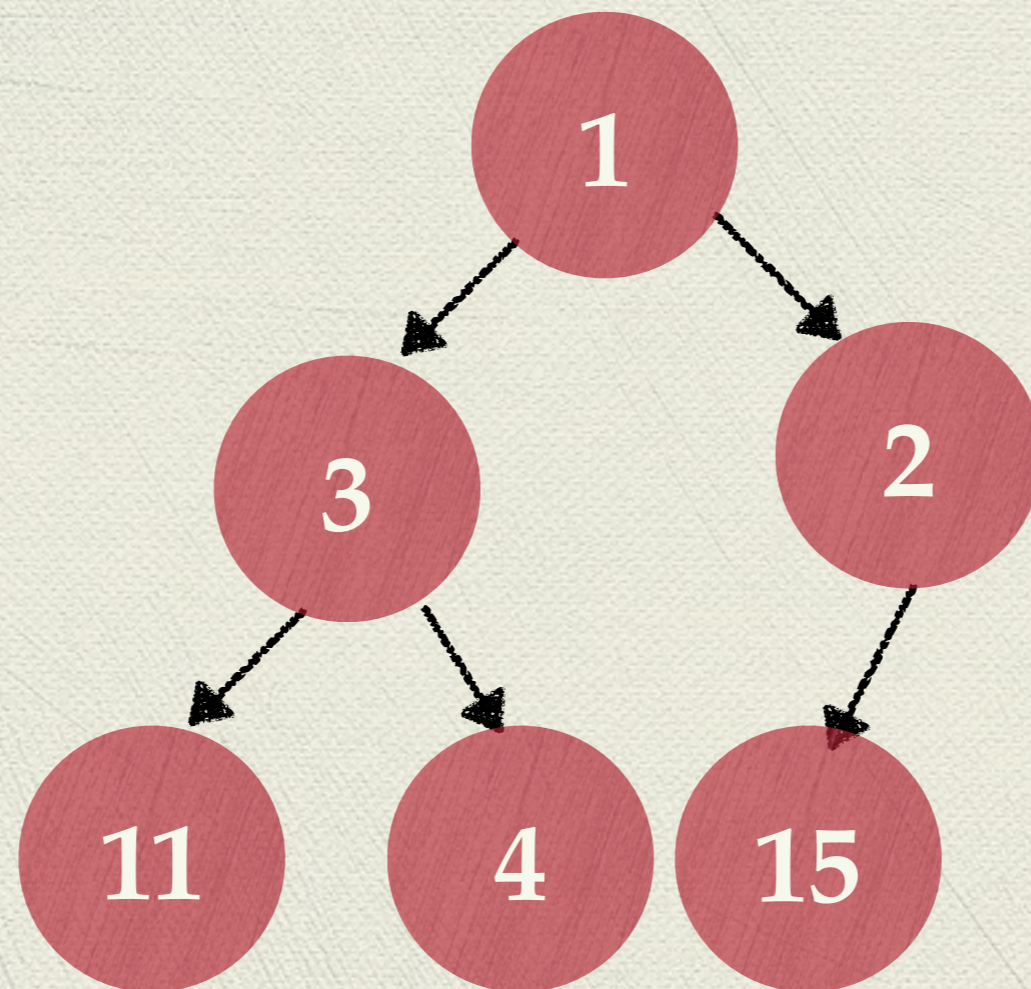




- ◆ If this were a BST, we could tell if we should go look down left or right
- ◆ But there's no hint at the direction in heap



◆ This would also have been a valid heap



# Heap contains

- ◆ In general, there's no way to know *where* an item is in a heap
- ◆ There are cute optimizations you can make, but asymptotically it's not better than just searching through every item in the heap

# Heap vs. BST: fun facts

- ◆ BST: Can check if contains any item in log time
- ◆ Heap: Can check if contains any item in linear time (how sad!) — Heap is really only good as a priority queue

# Quiz time!

- ◆ This is more of a midterm review question, and doesn't necessarily have to do with anything we just learned

# Quiz — Median maintainer

- ◆ Invent a data structure that supports the following operations:
  - ▶ `void add(int item)`
  - ▶ `int getMedian()`
- ◆ Your goal is to have as fast asymptotic runtimes as possible

# Quiz — Median maintainer

◆ Assume you have the following structures at your disposal to help

- ▶ LinkedList
- ▶ ArrayList
- ▶ HashSet
- ▶ HashMap

- ▶ BST
- ▶ Stack
- ▶ Queue
- ▶ PriorityQueue
- ▶ Graph

# Quiz — Median maintainer

- ◆ Hint 1: You should be able to have `getMedian` at constant time, and `add` at  $\log$  time



# Quiz — Median maintainer

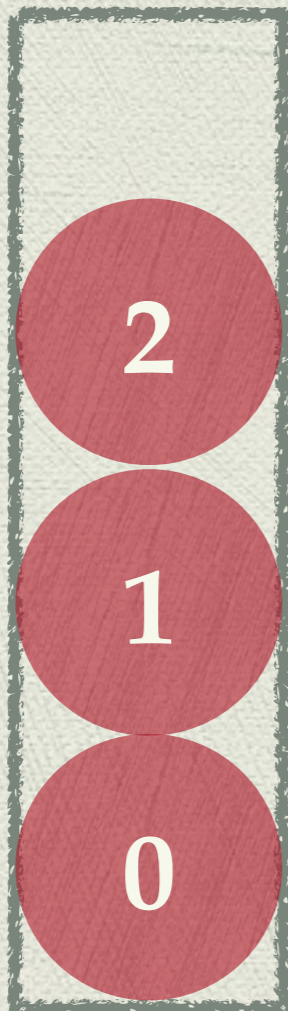
- ◆ Hint 2: Half the items are smaller than the median, and half of them are larger than the median
- ◆ Maybe have two separate collections: the items that are smaller, and the items that are larger...

# Quiz — Median maintainer

- ◆ Solution: Maintain two priority queues, one of which is a max priority queue, and the other of which is a min priority queue

# Main idea

Items less than  
median (max  
priority queue)



The median

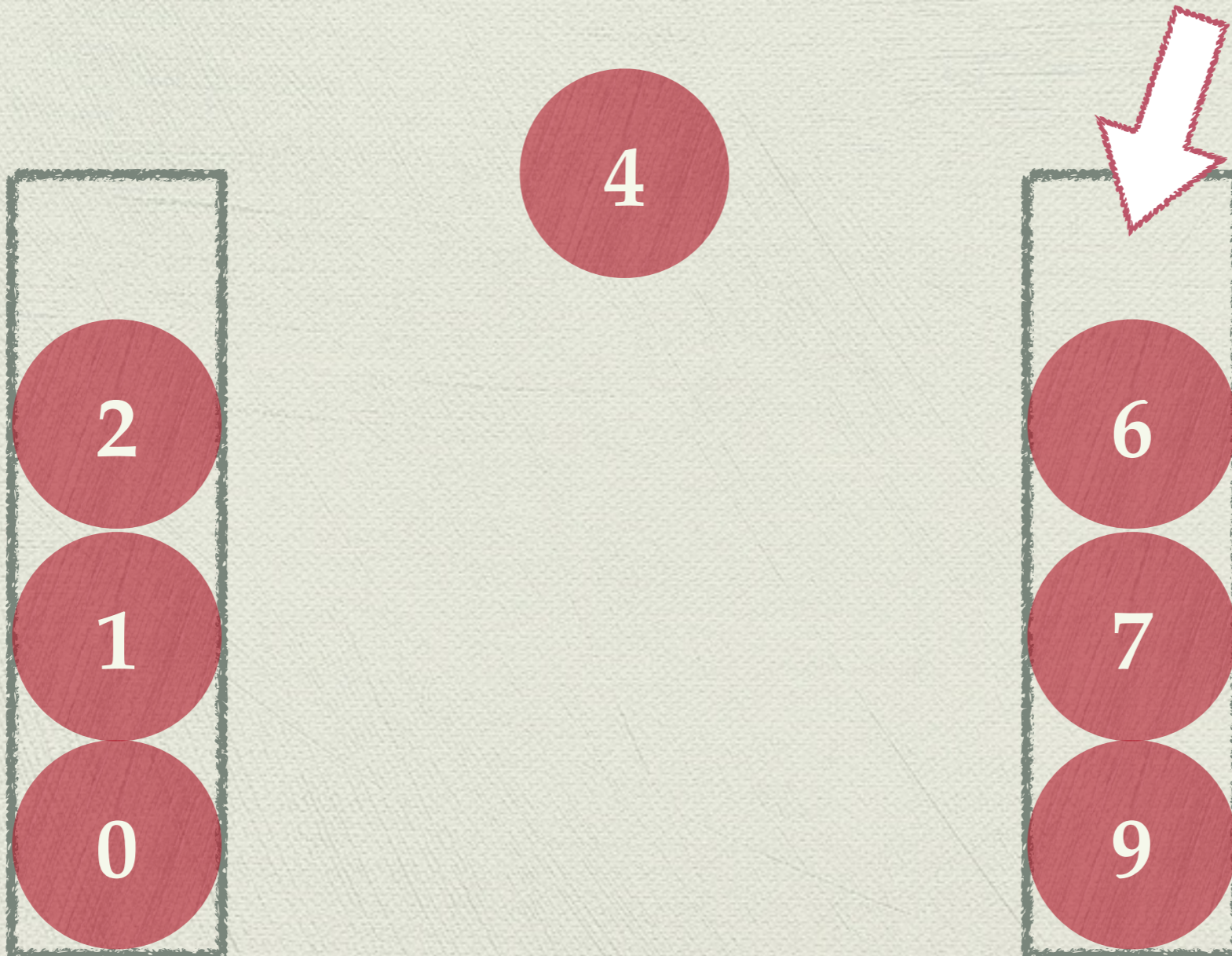


Items greater  
than median (min  
priority queue)



# Main idea

If we add items here, the new median will be the smallest thing in this collection (so we want min pq)



# Main idea

If we add items here, the new median will be the largest thing in this collection (so we want max pq)

