

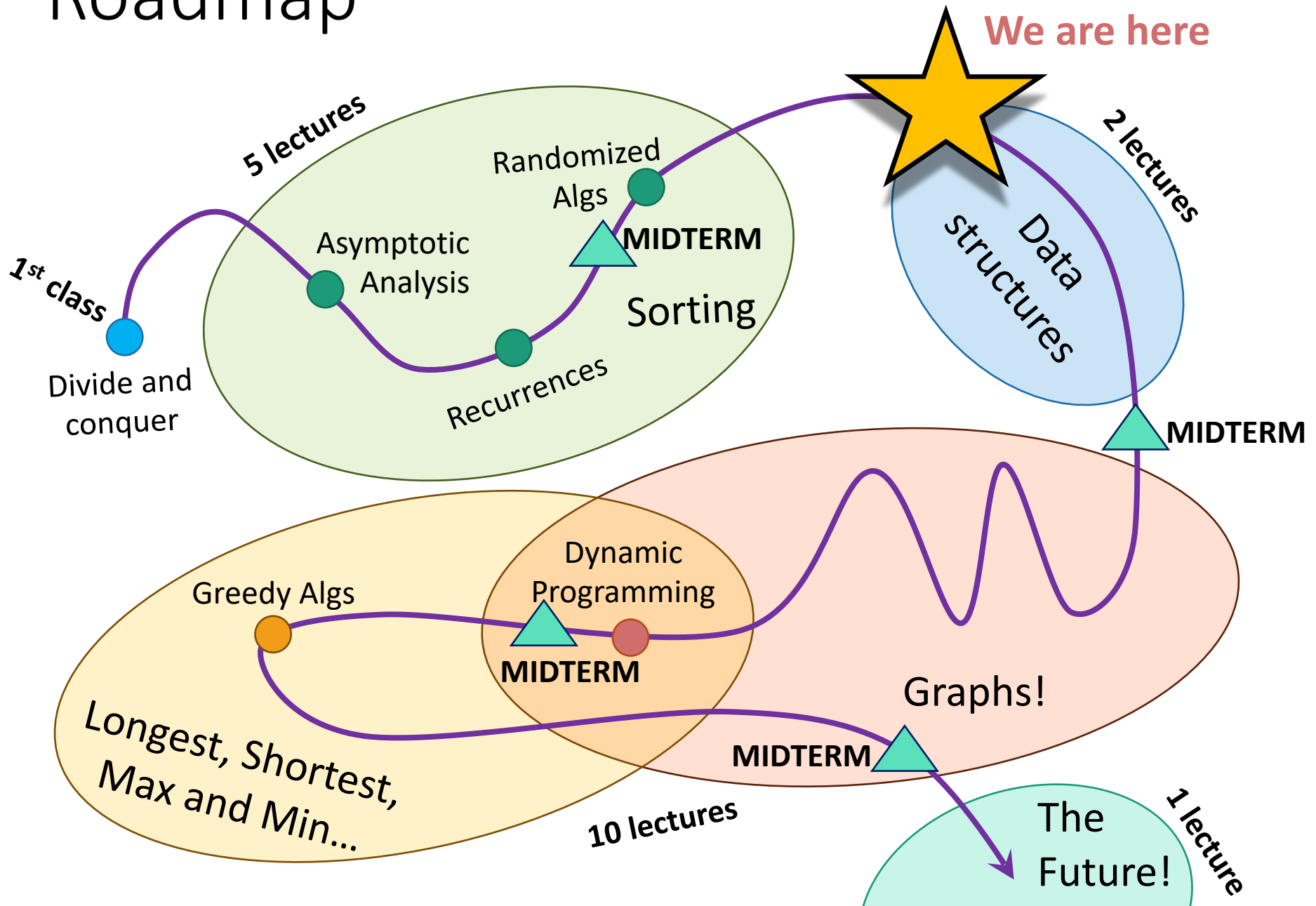
# Lecture 7

Binary Search Trees and Red-Black Trees

# Announcements

- HW4 is out today
- HW partners
- Please tag all pages of a question on Gradescope, not just first one!

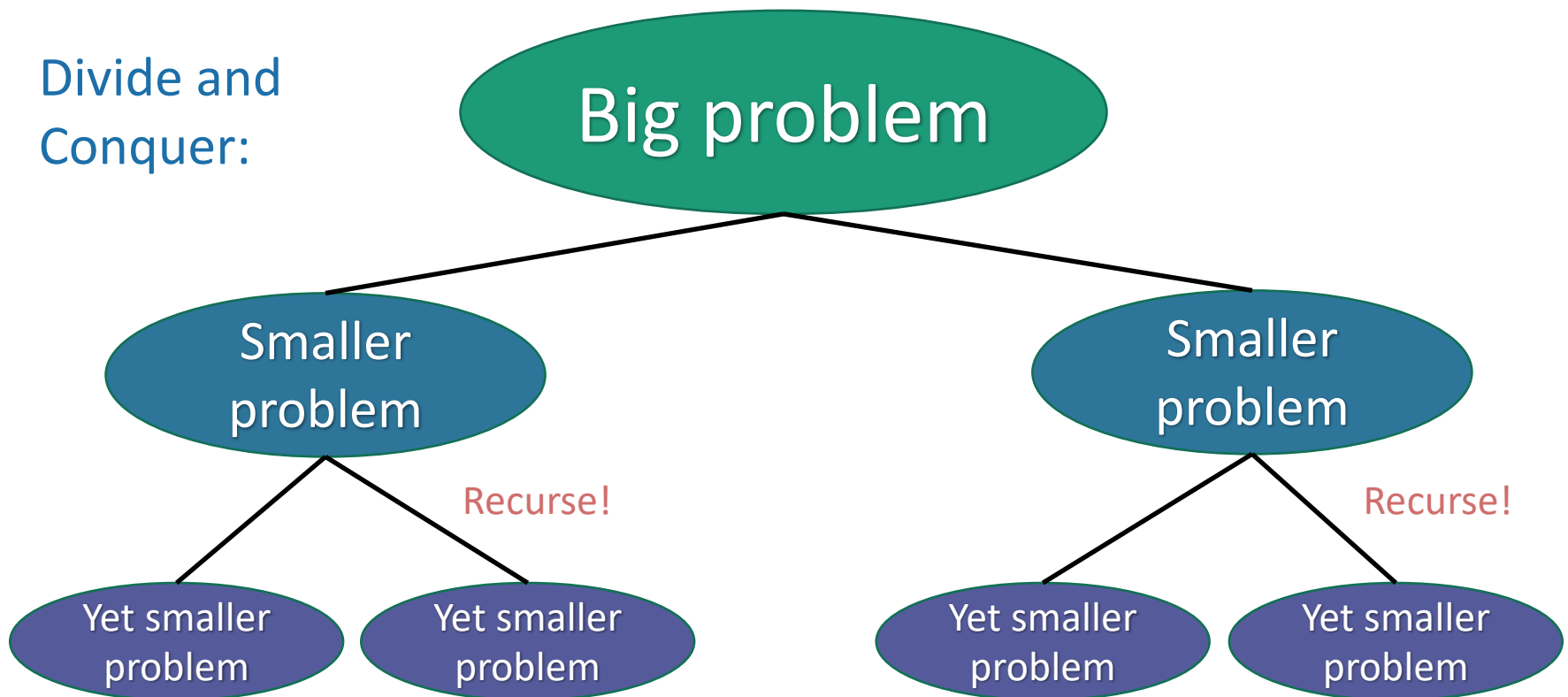
# Roadmap



# But first!

- A brief wrap-up of divide and conquer.

Divide and  
Conquer:



# How do we design divide-and-conquer algorithms?

- So far we've seen lots of examples.
  - Karatsuba (and Alien Multiplication)
  - MergeSort
  - Select
  - QuickSort
  - Matrix Multiplication (HW2)
  - Majority Element (HW3)
- Let's take a minute to zoom out and look at some general strategies.



# One Strategy

1. Identify natural sub-problems
  - Arrays of half the size
  - Things smaller/larger than a pivot
2. Imagine you had the magical ability to solve those natural sub-problems...what would you do?
  - Just try it with all of the natural sub-problems you can come up with! Anything look helpful?
3. Work out the details
  - Write down pseudocode, etc.

Think about how you could  
arrive at MergeSort or  
QuickSort via this strategy!





# Other tips

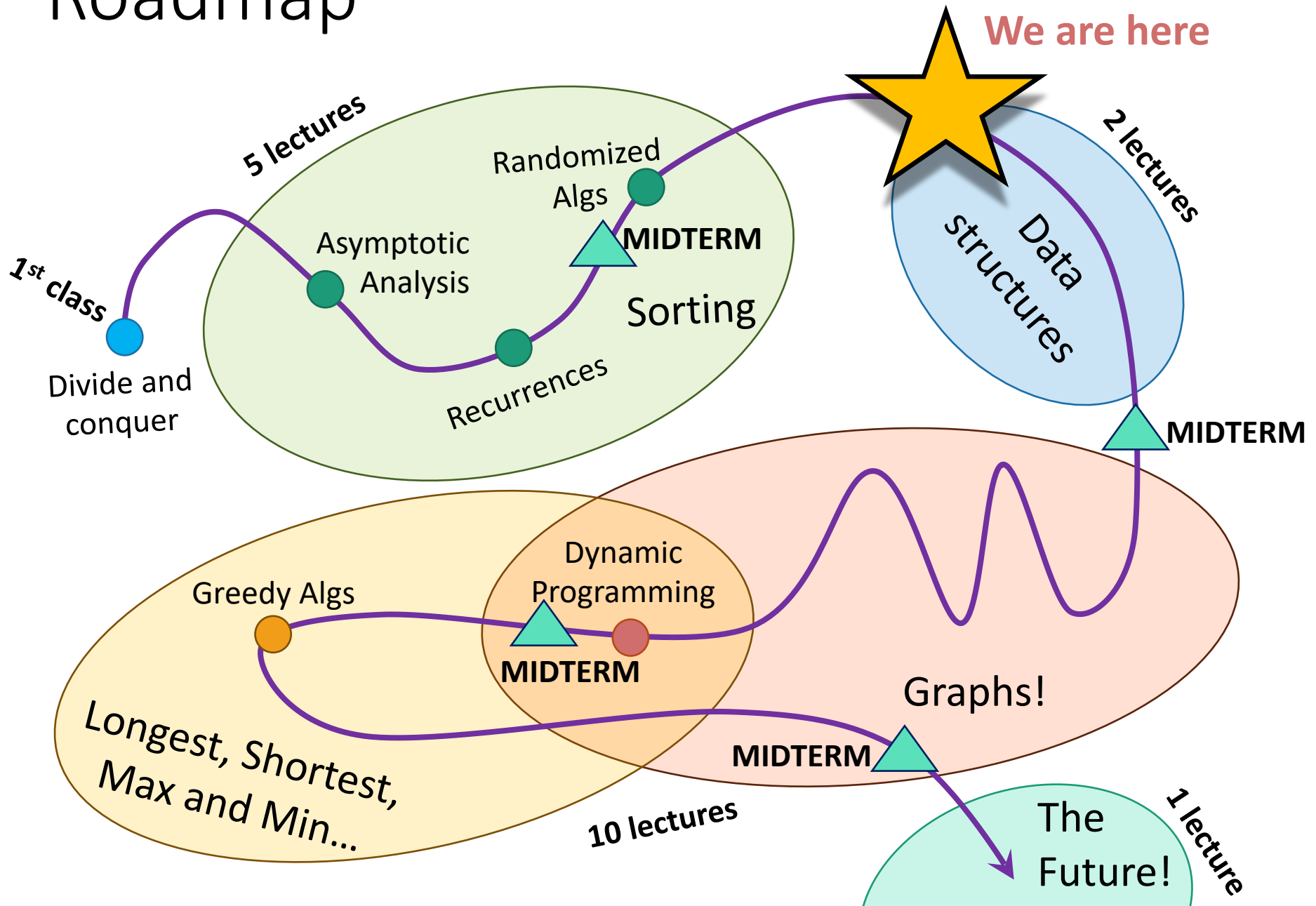
- Small examples.
  - If you have an idea but are having trouble working out the details, try it on a small example by hand.
- Gee, that looks familiar...
  - The more algorithms you see, the easier it will get to come up with new algorithms!
- Bring in your analysis tools.
  - E.g., if I'm doing divide-and-conquer with 2 subproblems of size  $n/2$  and I want an  $O(n \log n)$  time algorithm, I know that I can afford  $O(n)$  work combining my sub-problems.
- Iterate.
  - Darn, that approach didn't work! But, if I tweaked this aspect of it, maybe it works better?
- Everyone approaches problem-solving differently...find the way that works best for you.

# There is no universal recipe for designing algorithms.

- This can be frustrating on HW and midterms ....
  - P vs NP: much easier to understand a proof than to come up with one!
- Practice helps!
  - The examples we see in Lecture and in HW are meant to help you practice this skill.
- There are even more algorithms in the optional texts!
  - Check out CLRS Chapter 4, or Algorithms Illuminated Chapter 3 for even more examples of divide and conquer algorithms.

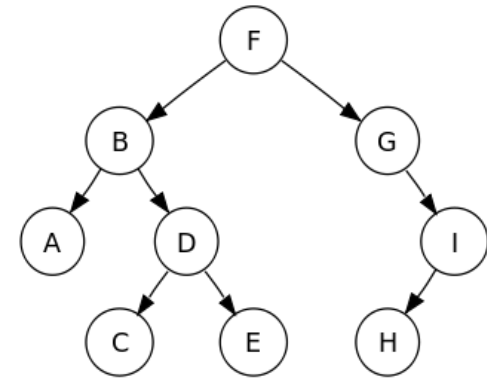


# Roadmap



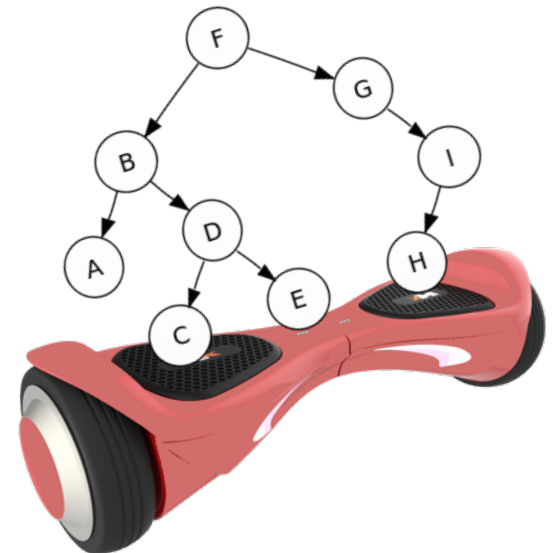
# Today

- Begin a brief foray into data structures!
  - See CS 166 for more!
- Binary search trees
  - You may remember these from CS 106B
  - They are better when they're balanced.



this will lead us to...

- Self-Balancing Binary Search Trees
  - **Red-Black** trees.



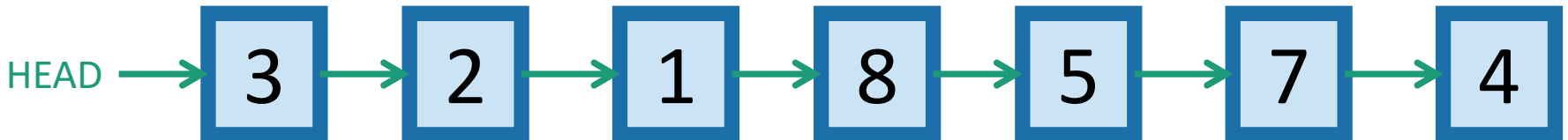
# Some data structures

for storing objects like **5** (aka, **nodes** with **keys**)

- (Sorted) arrays:



- Linked lists:



- Some basic operations:
  - INSERT, DELETE, SEARCH

# Sorted Arrays

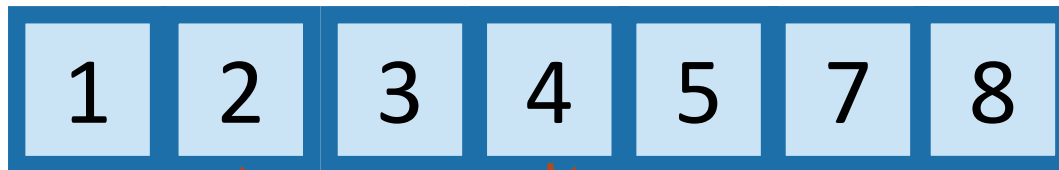


- $O(n)$  INSERT/DELETE:

- First, find the relevant element (we'll see how below), and then move a bunch of elements in the array:



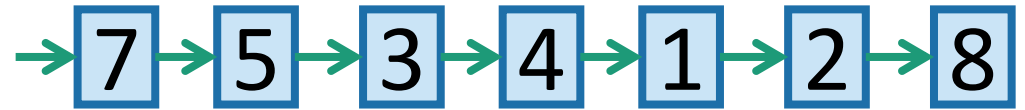
- $O(\log(n))$  SEARCH: eg, insert 4.5



eg, Binary search to see if 3 is in A.

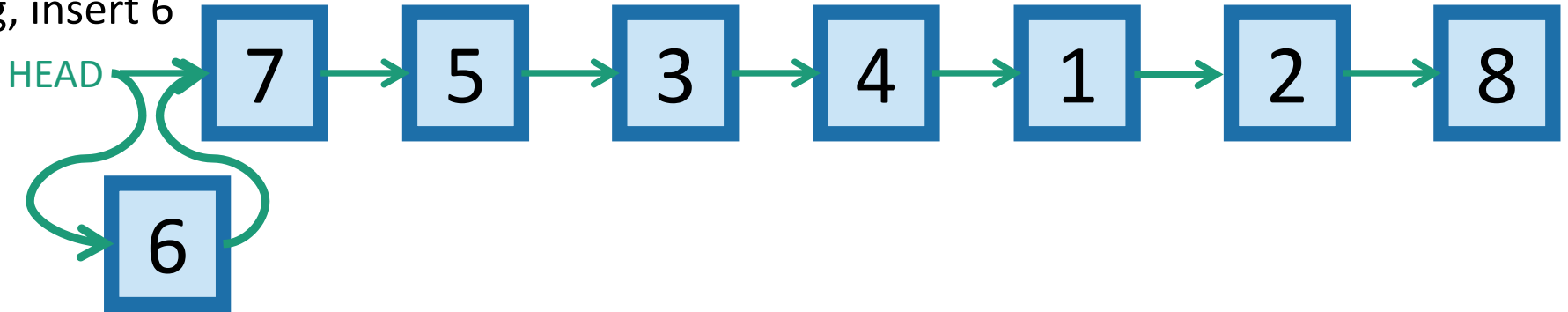
(Not necessarily sorted)

# Linked lists

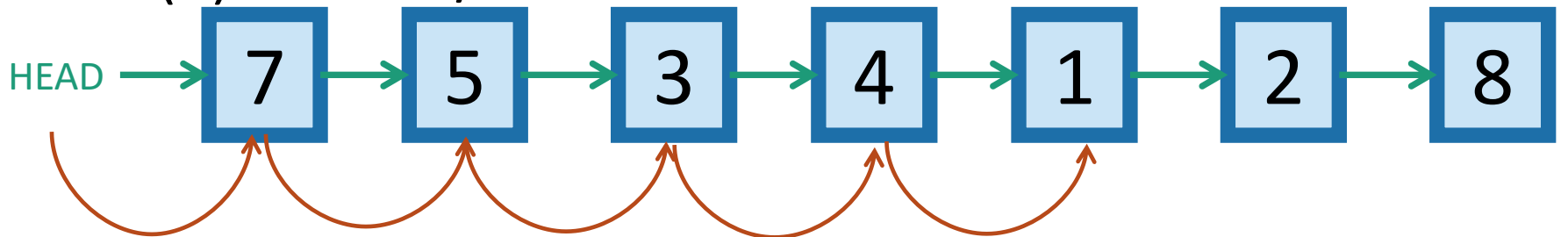


- $O(1)$  INSERT:

eg, insert 6



- $O(n)$  SEARCH/DELETE:



eg, search for 1 (and then you could delete it by manipulating pointers).

# Motivation for Binary Search Trees

*TODAY!*

	Sorted Arrays	Linked Lists	(Balanced) Binary Search Trees
Search	$O(\log(n))$ 😊	$O(n)$ 😞	$O(\log(n))$ 😊
Delete	$O(n)$ 😞	$O(n)$ 😞	$O(\log(n))$ 😊
Insert	$O(n)$ 😞	$O(1)$ 😊	$O(\log(n))$ 😊

For today all keys are distinct.

# Binary tree terminology [poll]

Each node has two **children**.

The **left child** of **3** is **2**

The **right child** of **3** is **4**

The **parent** of **3** is **5**

**2** is a **descendant** of **5**

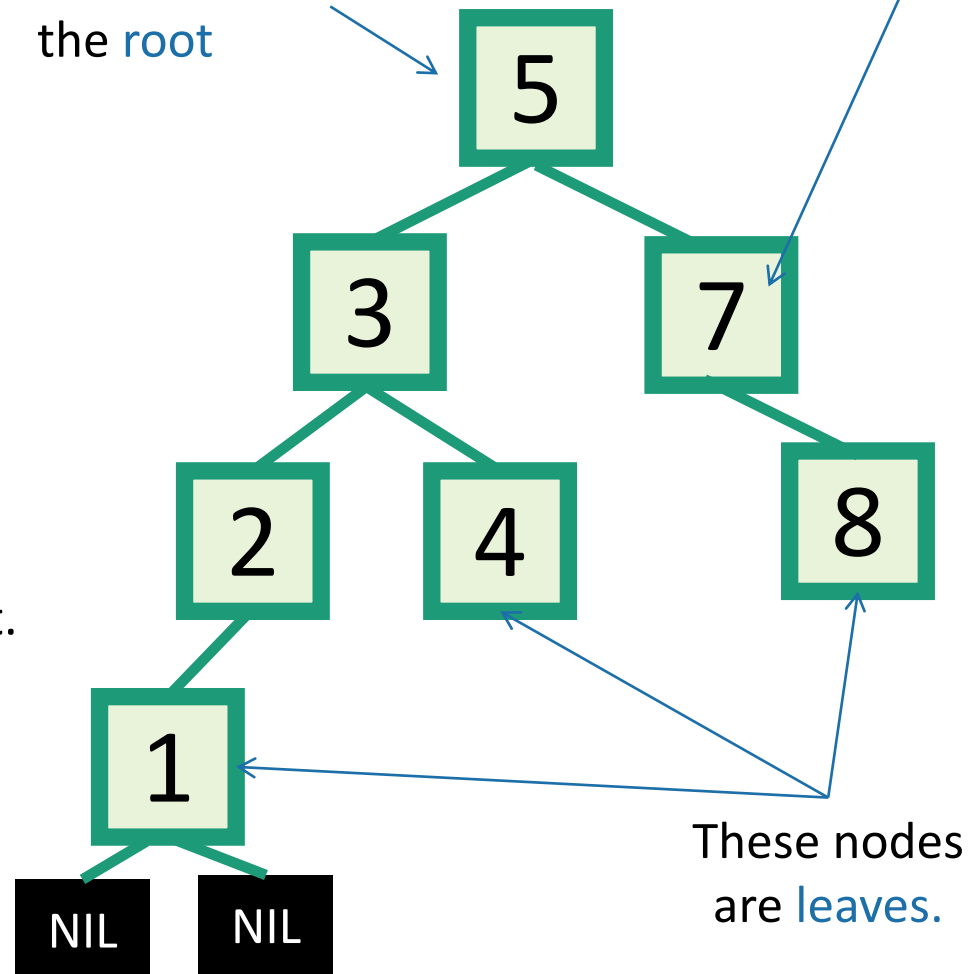
Each node has a pointer to its left child, right child, and parent.

Both **children** of **1** are NIL.  
(I won't usually draw them).

The **height** of this tree is 3.  
(Max number of edges from the root to a leaf).

This node is the **root**

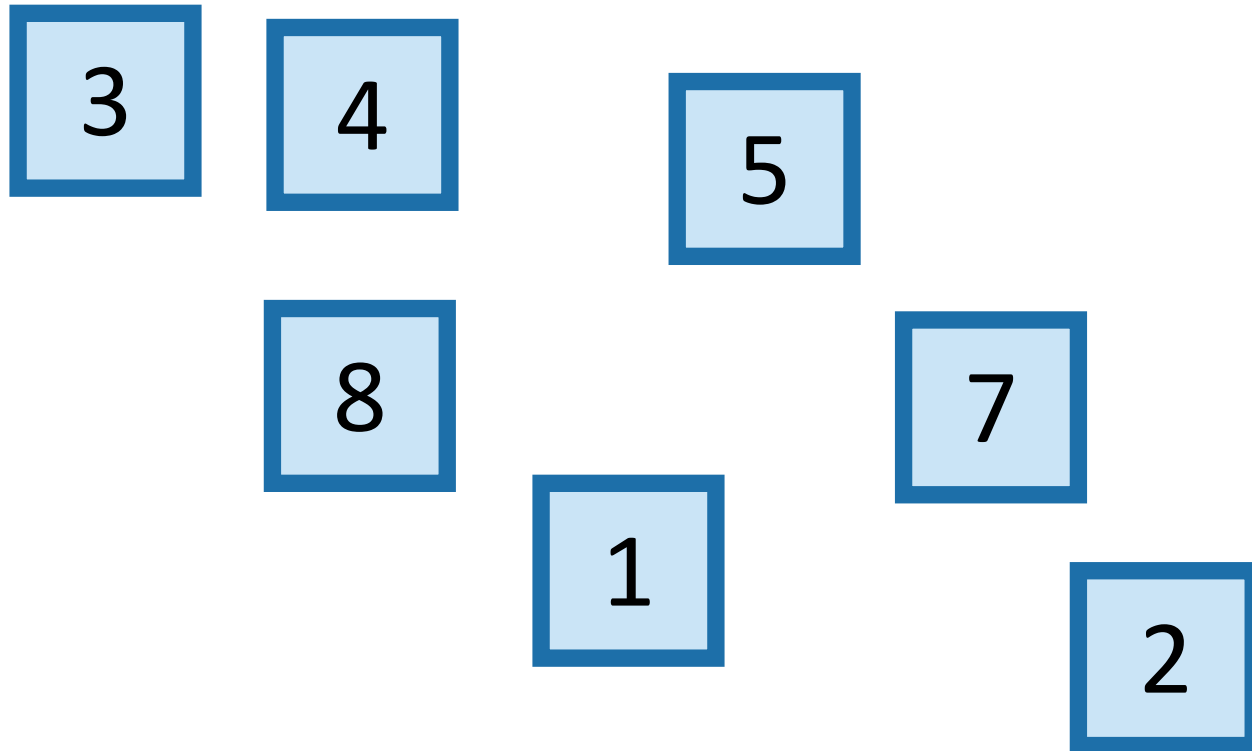
This is a **node**.  
It has a **key** (7).



From your pre-lecture exercise...

# Binary Search Trees

- A BST is a binary tree so that:
  - Every LEFT descendant of a node has key less than that node.
  - Every RIGHT descendant of a node has key larger than that node.
- Example of building a binary search tree:

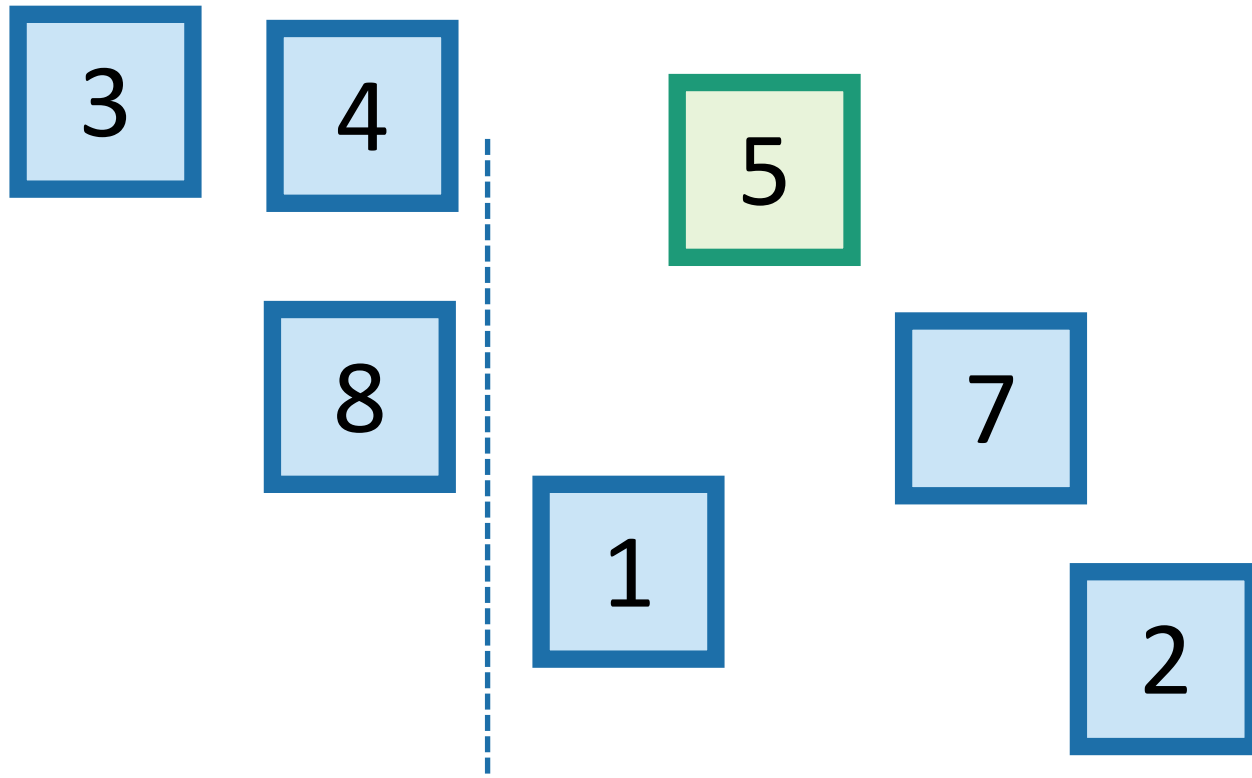




From your pre-lecture exercise...

# Binary Search Trees

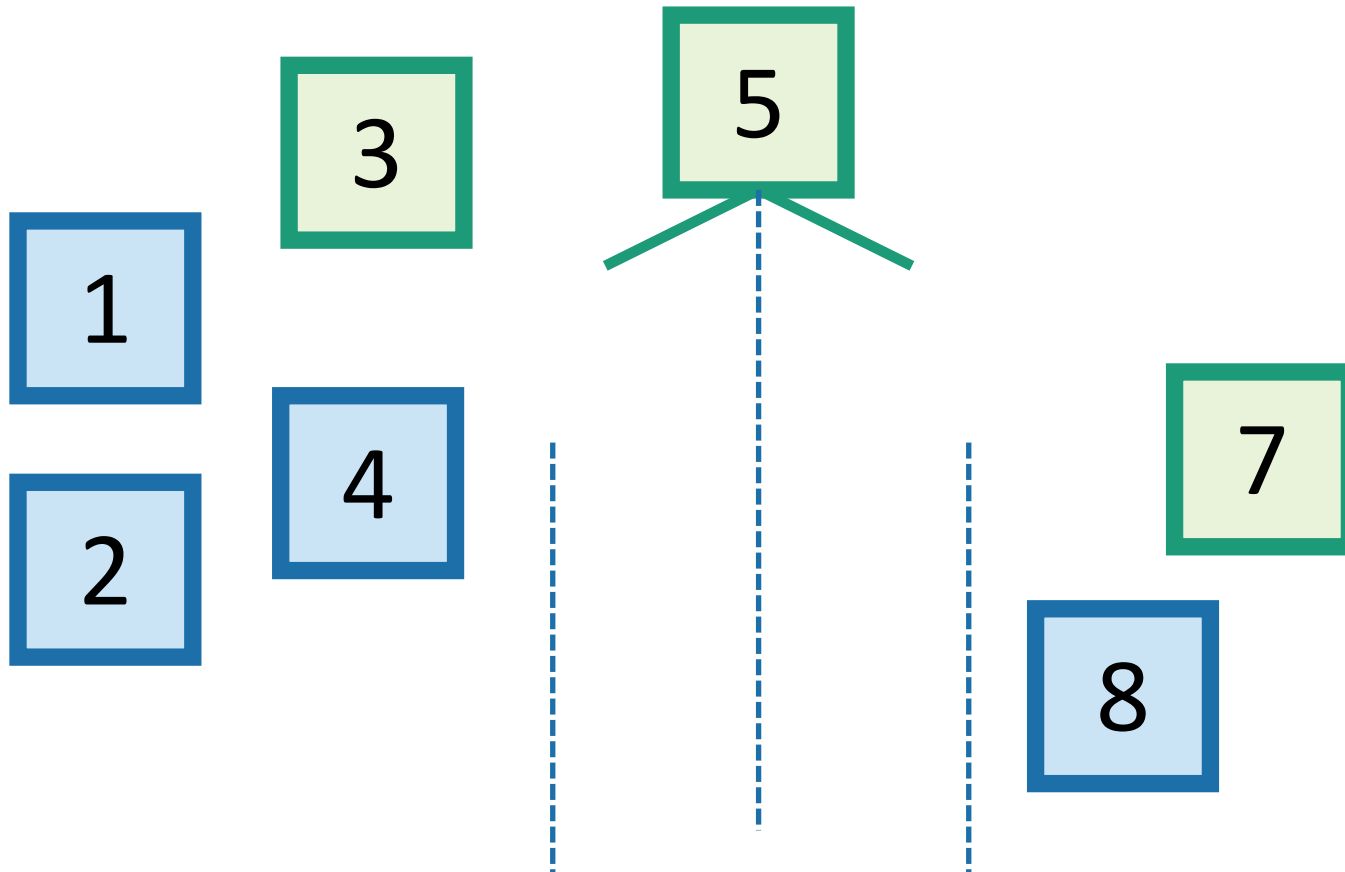
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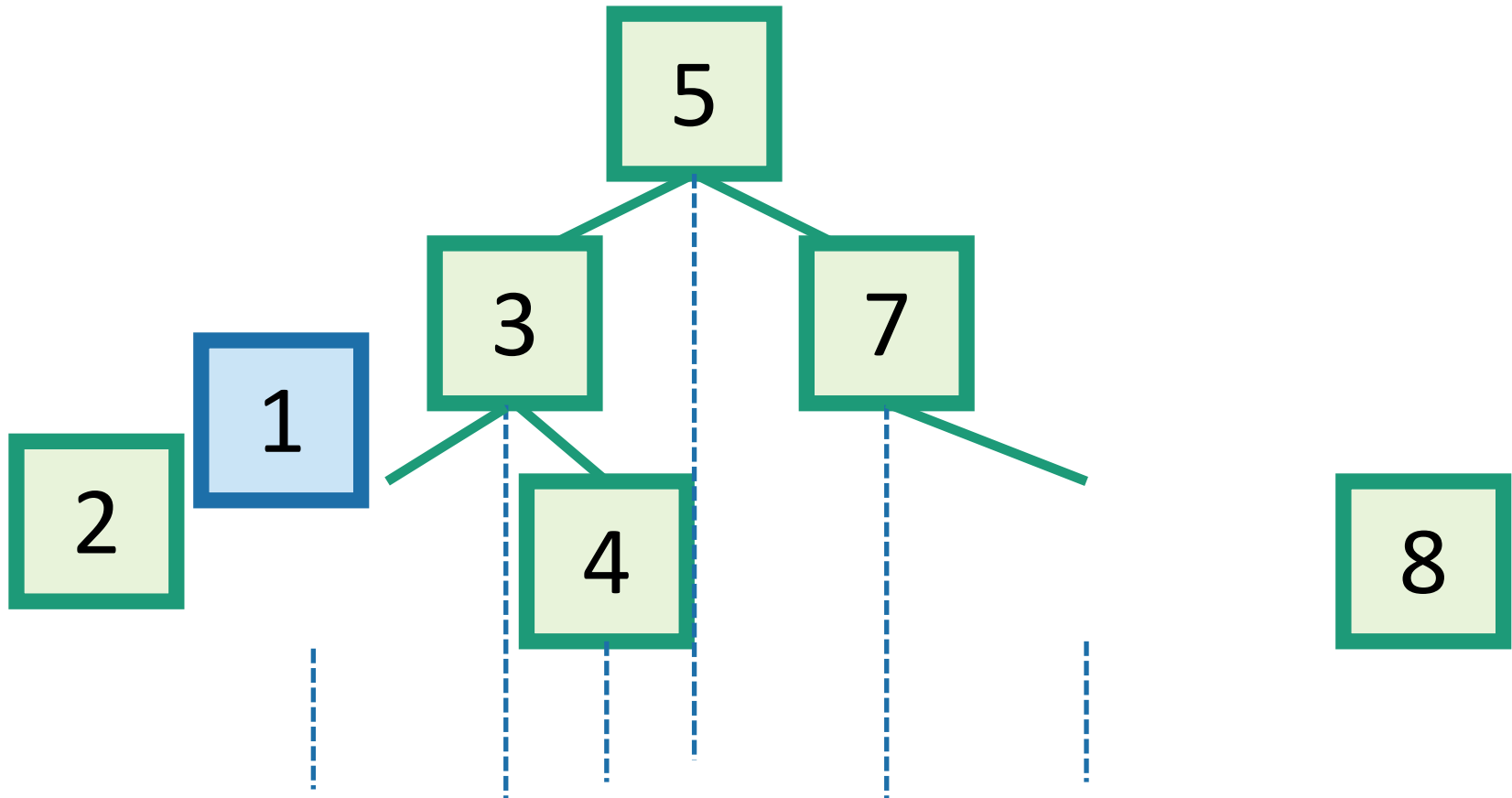
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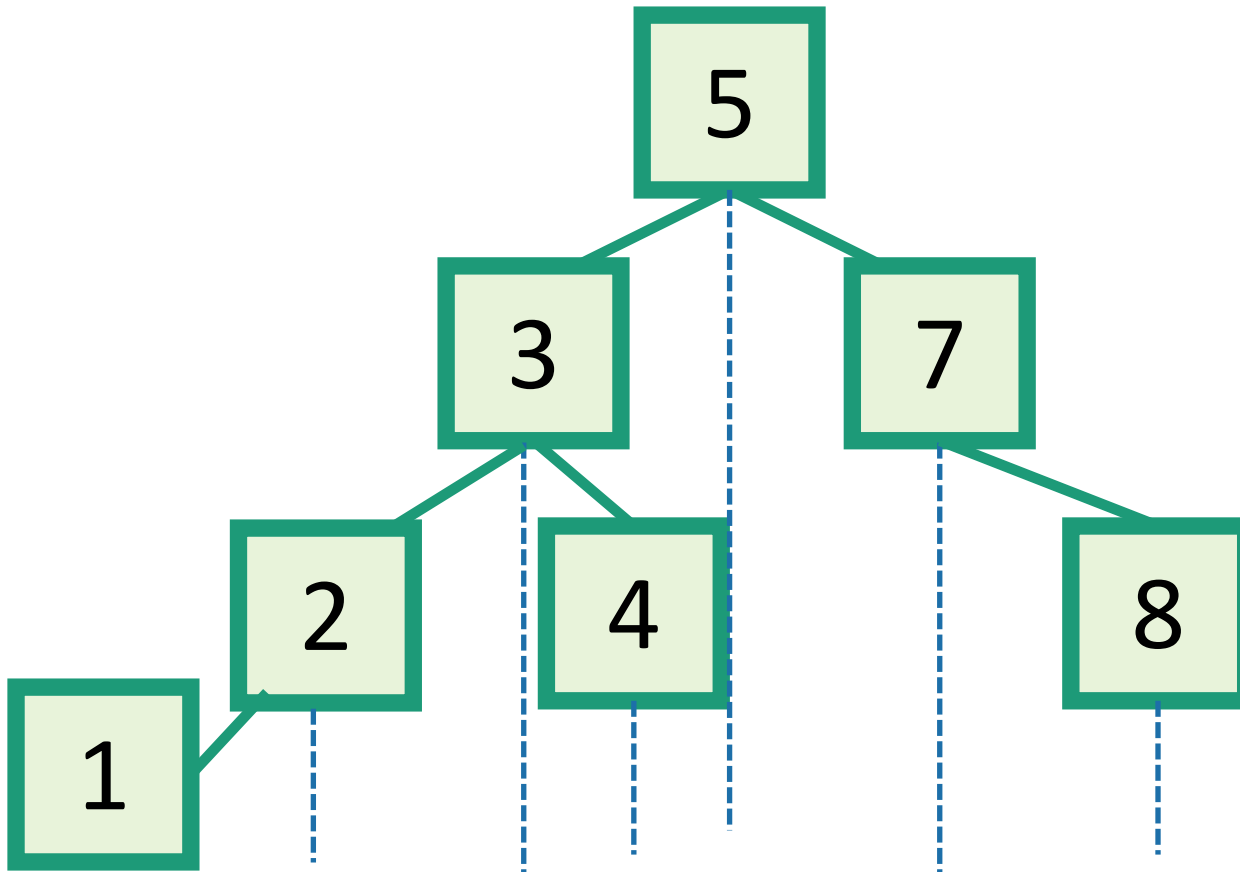
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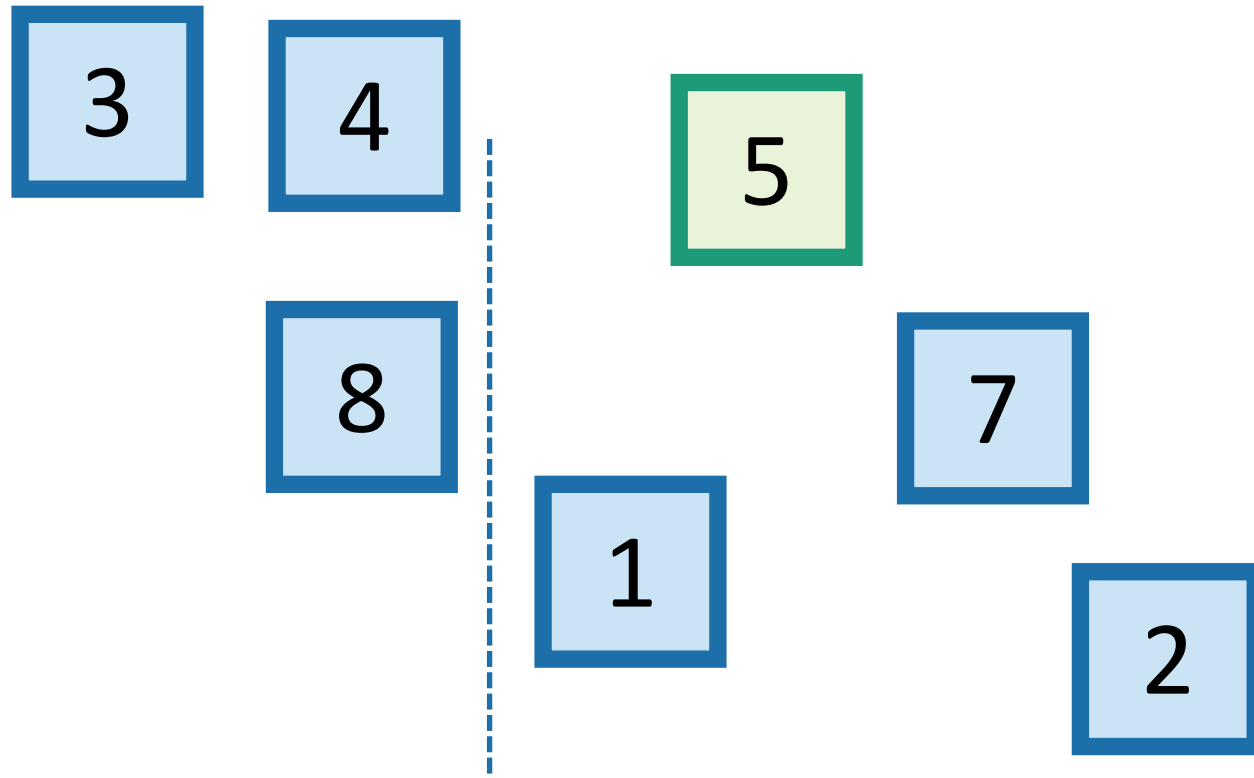
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- Example of building a binary search tree:



Q: Is this the only binary search tree I could possibly build with these values?

A: **No.** I made choices about which nodes to choose when. Any choices would have been fine.

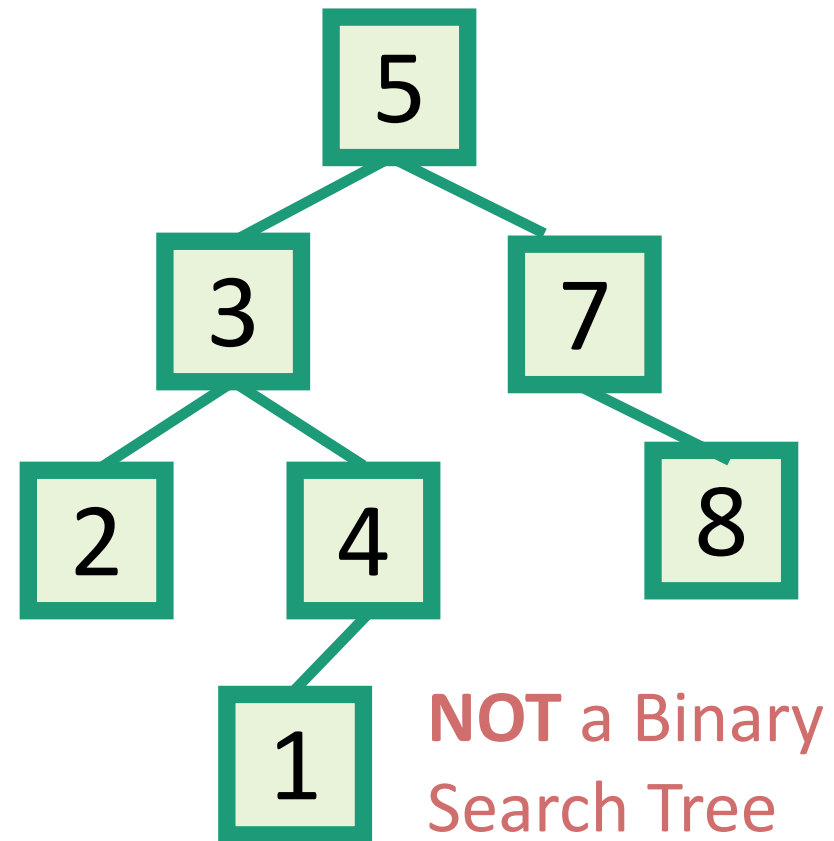
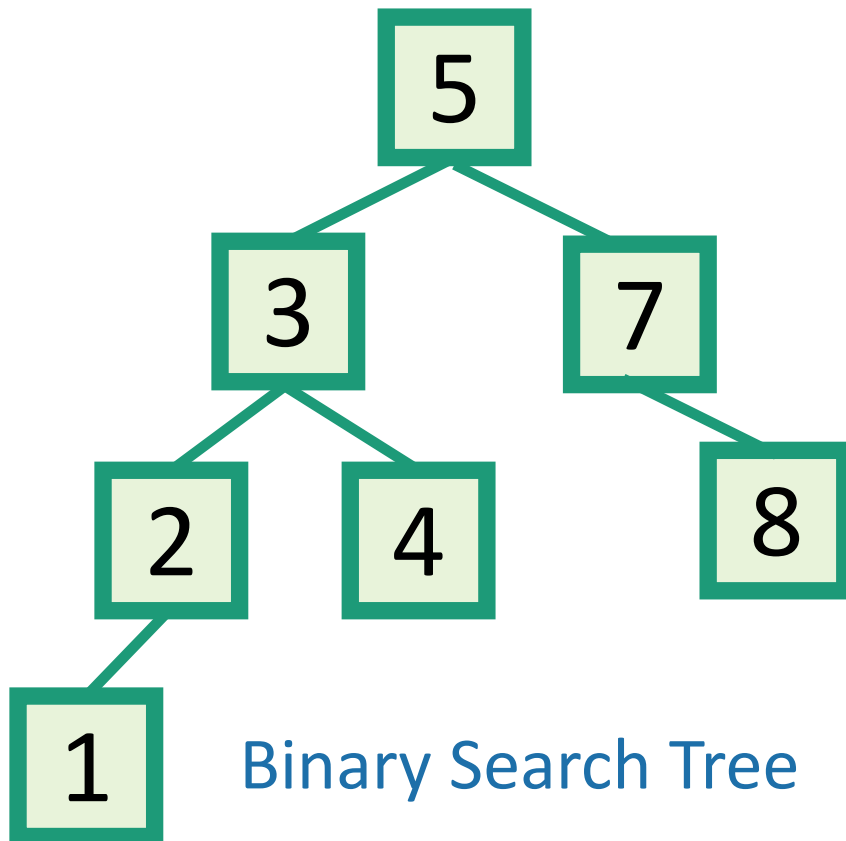
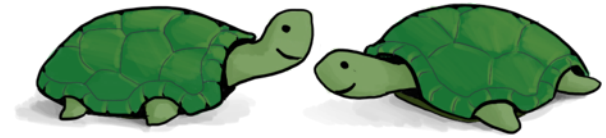
Aside: this should look familiar  
kinda like QuickSort



# Binary Search Trees

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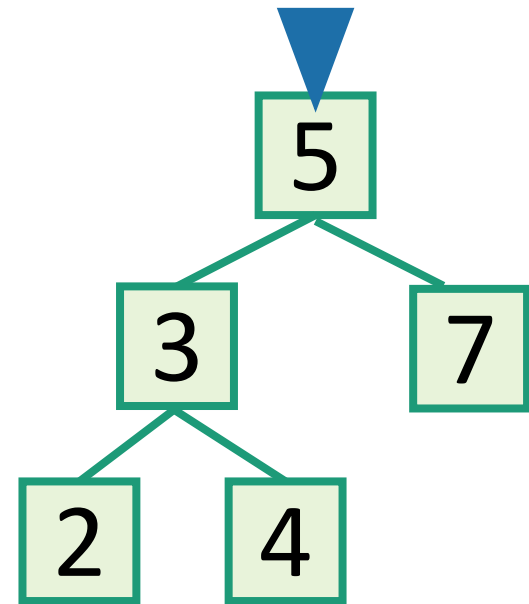
Which of these is a BST?  
1 minute Think-Pair-Share



# Aside: In-Order Traversal of BSTs

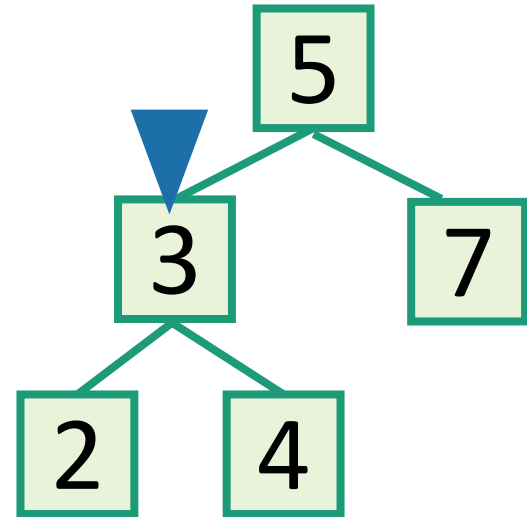
- Output all the elements in sorted order!

- inOrderTraversal(x):
  - if  $x \neq \text{NIL}$ :
    - inOrderTraversal( x.left )
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    - inOrderTraversal( x.right )



# Aside: In-Order Traversal of BSTs

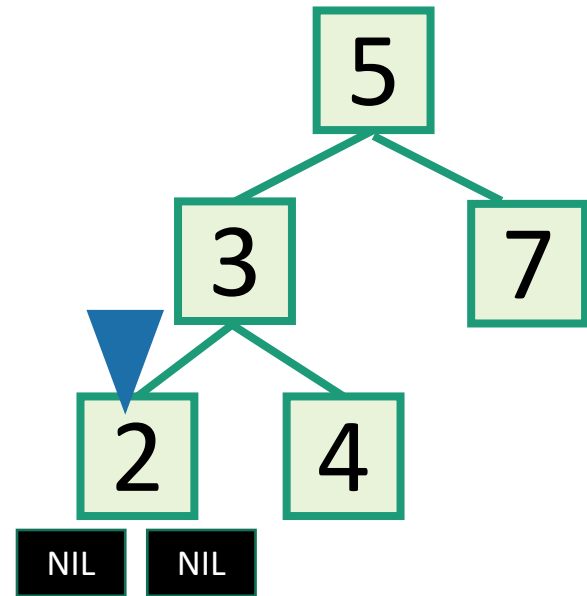
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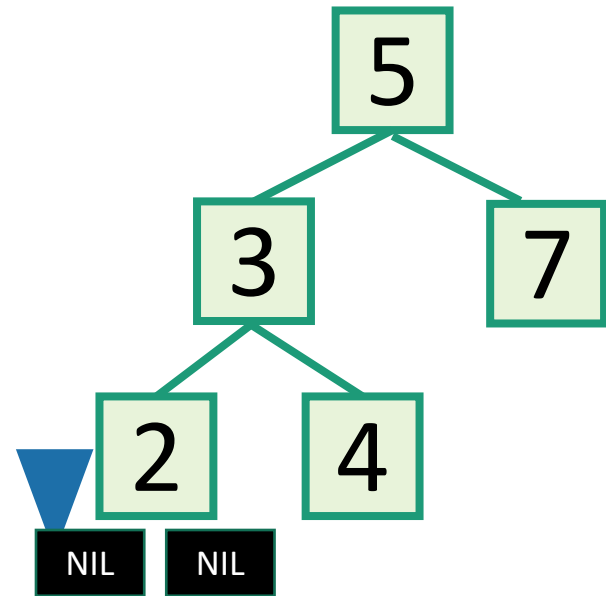
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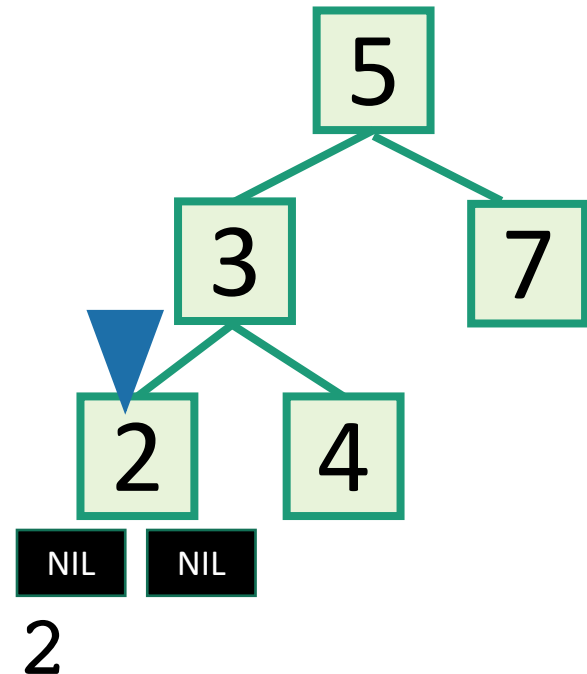
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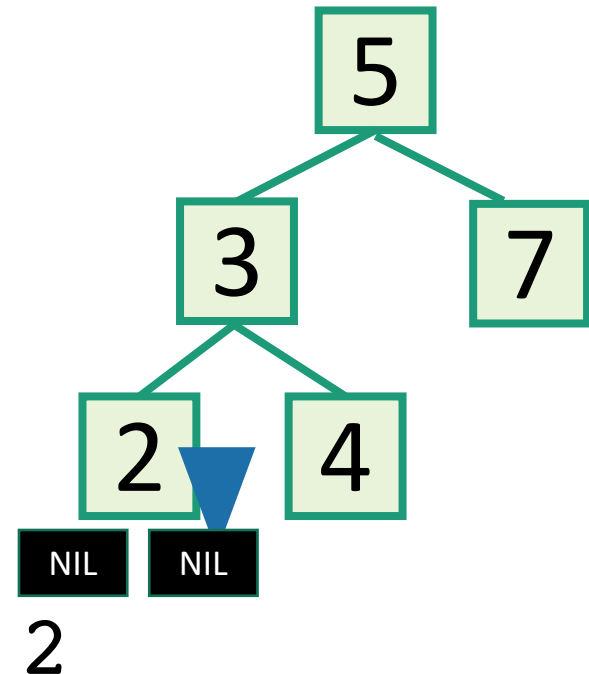
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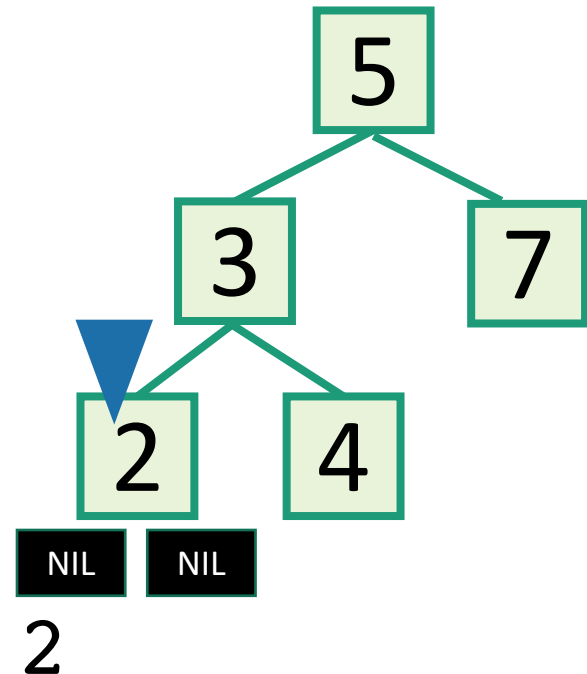
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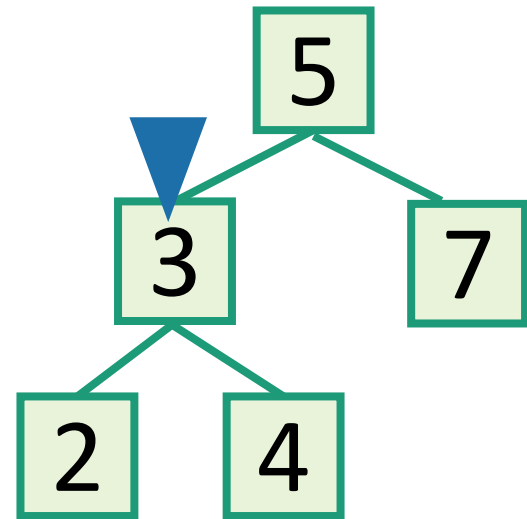
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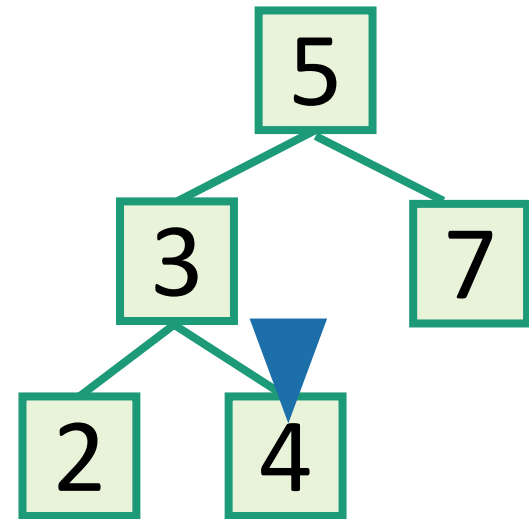
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2 3

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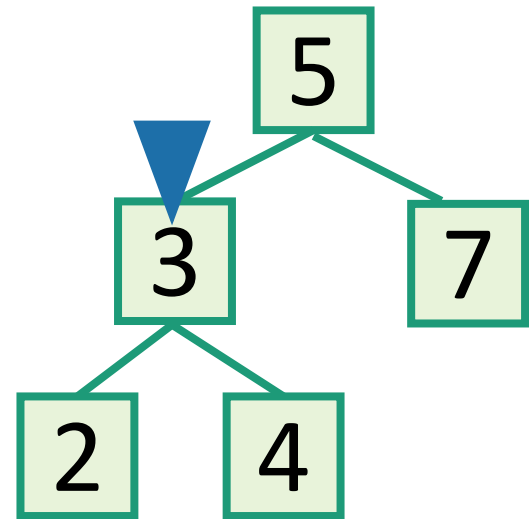
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2 3 4

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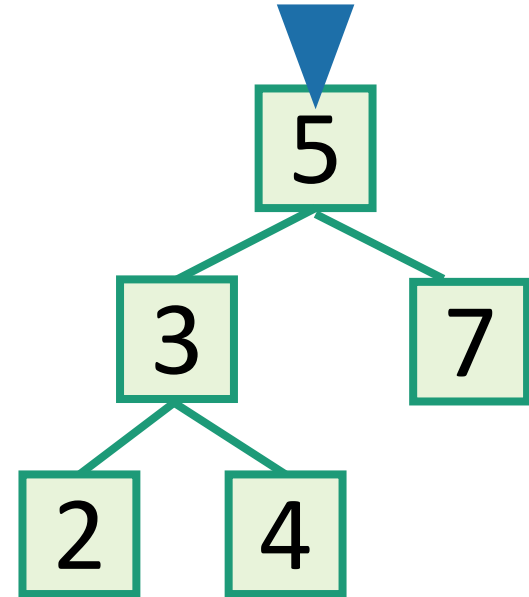
2 3 4



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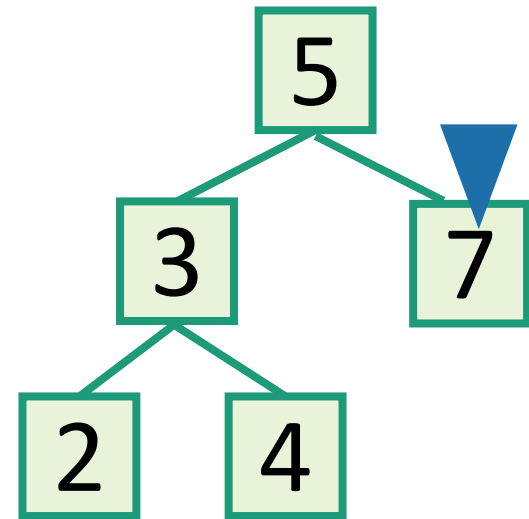


2 3 4 5

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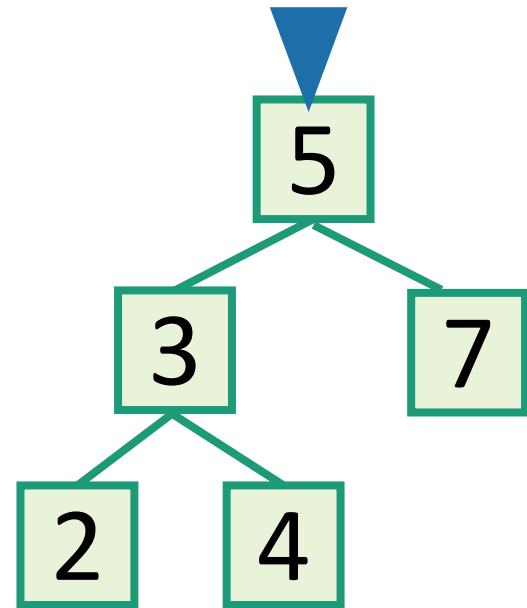


2 3 4 5 7

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- Runs in time  $O(n)$ .

2 3 4 5 7 Sorted!

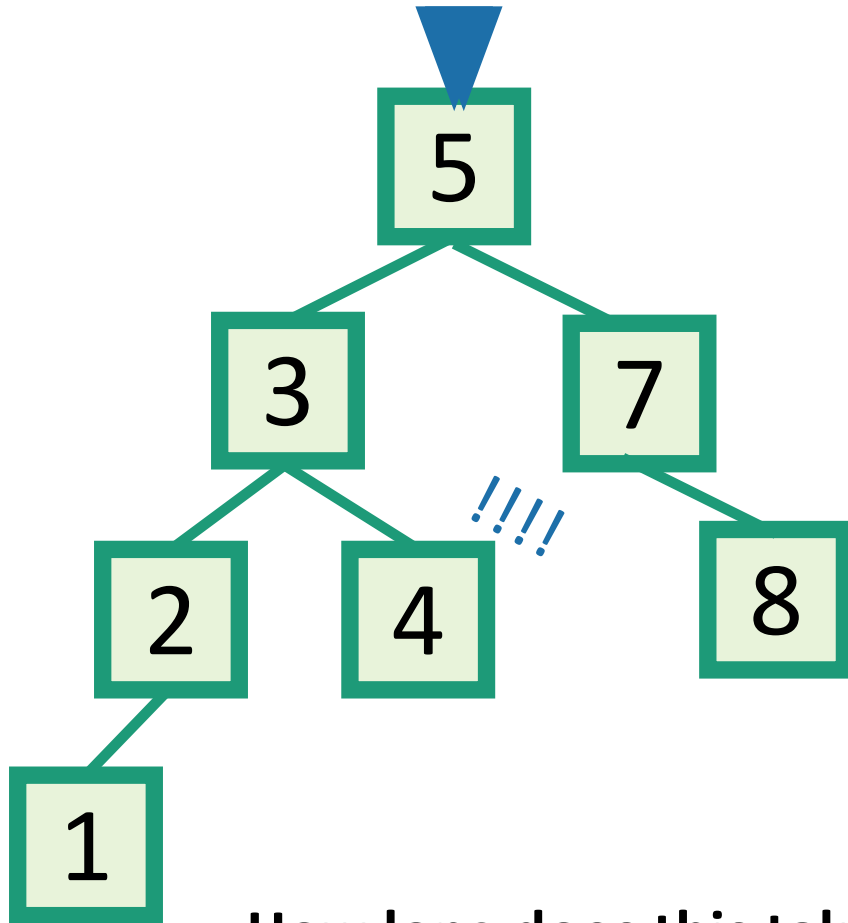
# Back to the goal

## Fast SEARCH/INSERT/DELETE

Can we do these?

# SEARCH in a Binary Search Tree

definition by example



How long does this take?

$O(\text{length of longest path}) = O(\text{height})$

**EXAMPLE:** Search for 4.

**EXAMPLE:** Search for 4.5

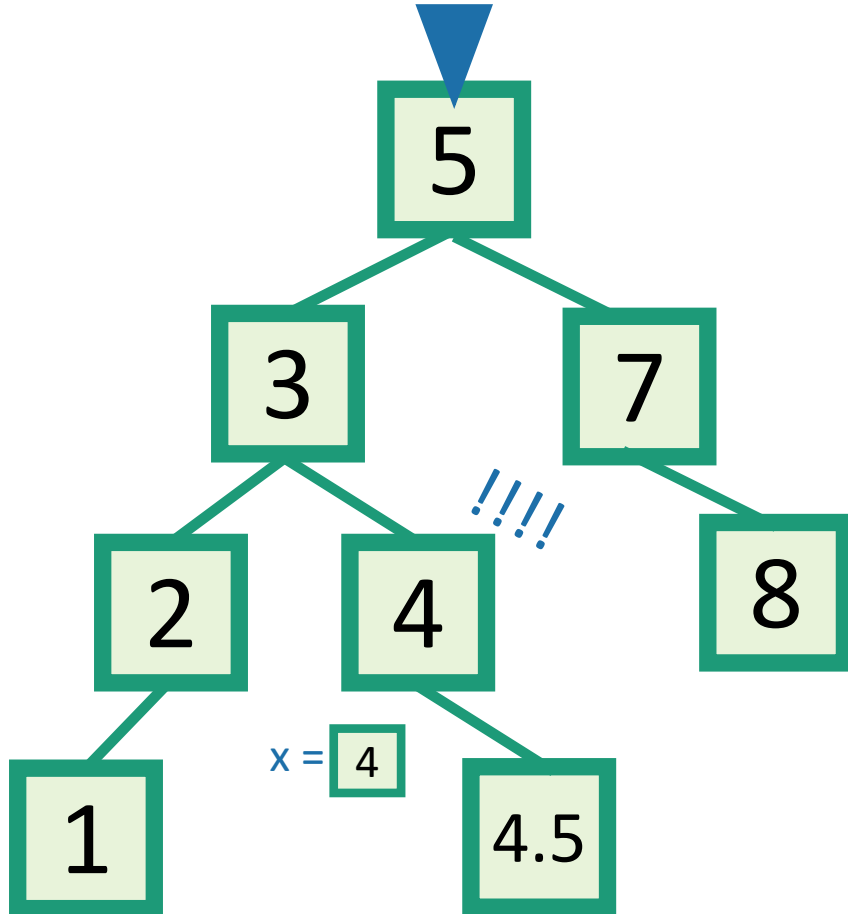
- It turns out it will be convenient to **return 4** in this case
- (that is, **return** the last node before we went off the tree)

Write pseudocode  
(or actual code) to  
implement this!



Ollie the over-achieving ostrich

# INSERT in a Binary Search Tree



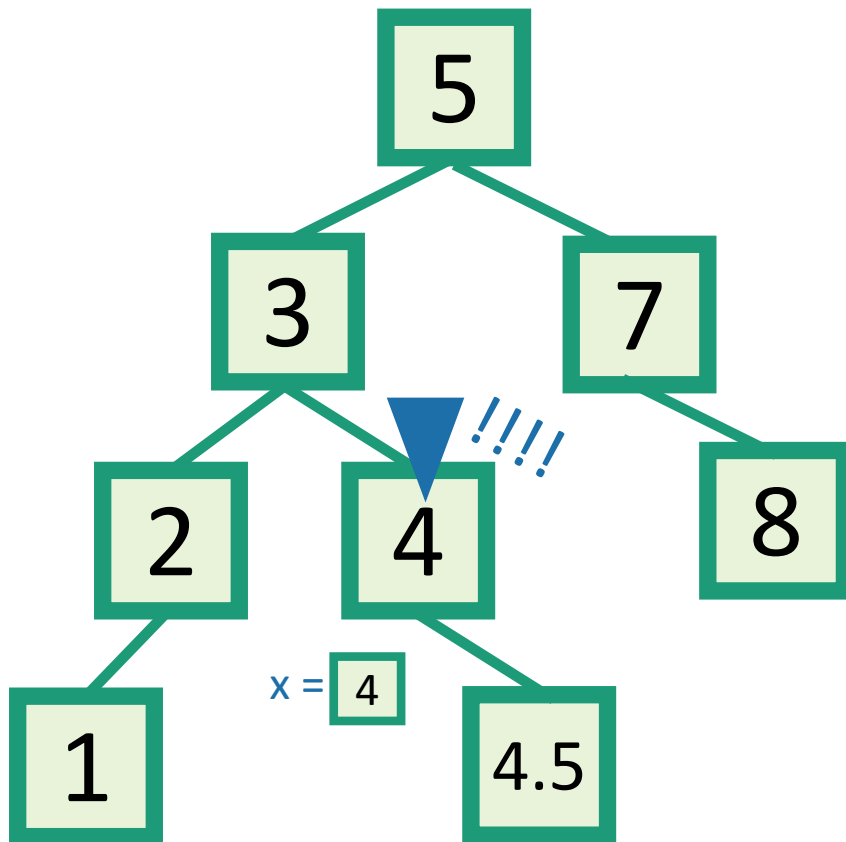
## EXAMPLE: Insert 4.5

- **INSERT**(key):
  - $x = \text{SEARCH}(\text{key})$
  - **Insert** a new node with desired key at  $x$ ...

You thought about this on  
your pre-lecture exercise!  
(See skipped slide for  
pseudocode.)

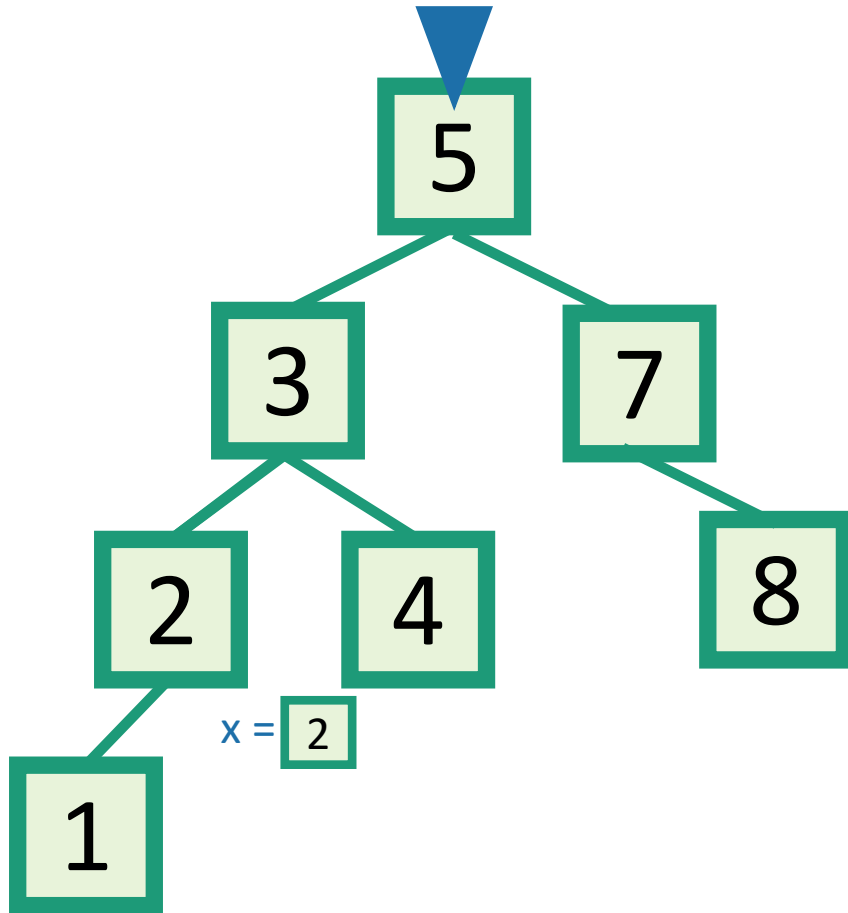
# INSERT in a Binary Search Tree

## EXAMPLE: Insert 4.5



- **INSERT**(key):
  - $x = \text{SEARCH}(\text{key})$
  - **if**  $\text{key} > x.\text{key}$ :
    - Make a new node with the correct key, and put it as the right child of  $x$ .
  - **if**  $\text{key} < x.\text{key}$ :
    - Make a new node with the correct key, and put it as the left child of  $x$ .
  - **if**  $x.\text{key} == \text{key}$ :
    - **return**

# DELETE in a Binary Search Tree



## EXAMPLE: Delete 2

- **DELETE**(key):
  - $x = \text{SEARCH}(\text{key})$
  - **if**  $x.\text{key} == \text{key}$ :
    - ....delete  $x$ ....

You thought about this in your pre-lecture exercise too!

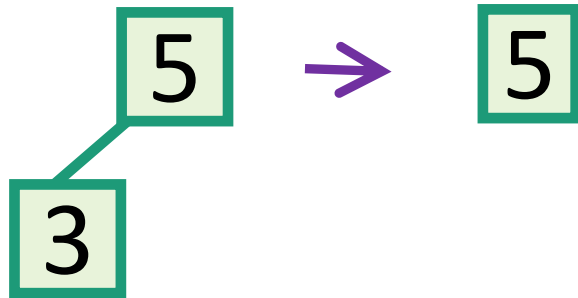
This is a bit more complicated...see the skipped slides for some pictures of the different cases.



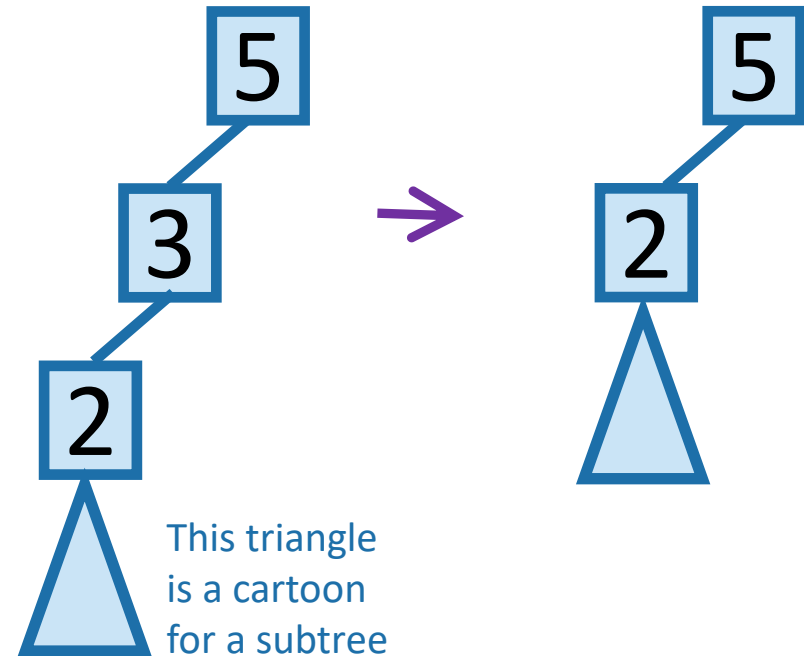
# DELETE in a Binary Search Tree

several cases (by example)  
say we want to delete 3

This slide skipped  
in class – here for  
reference!



**Case 1:** if 3 is a leaf,  
just delete it.



**Case 2:** if 3 has just one child,  
move that up.

Write pseudocode for all of these!

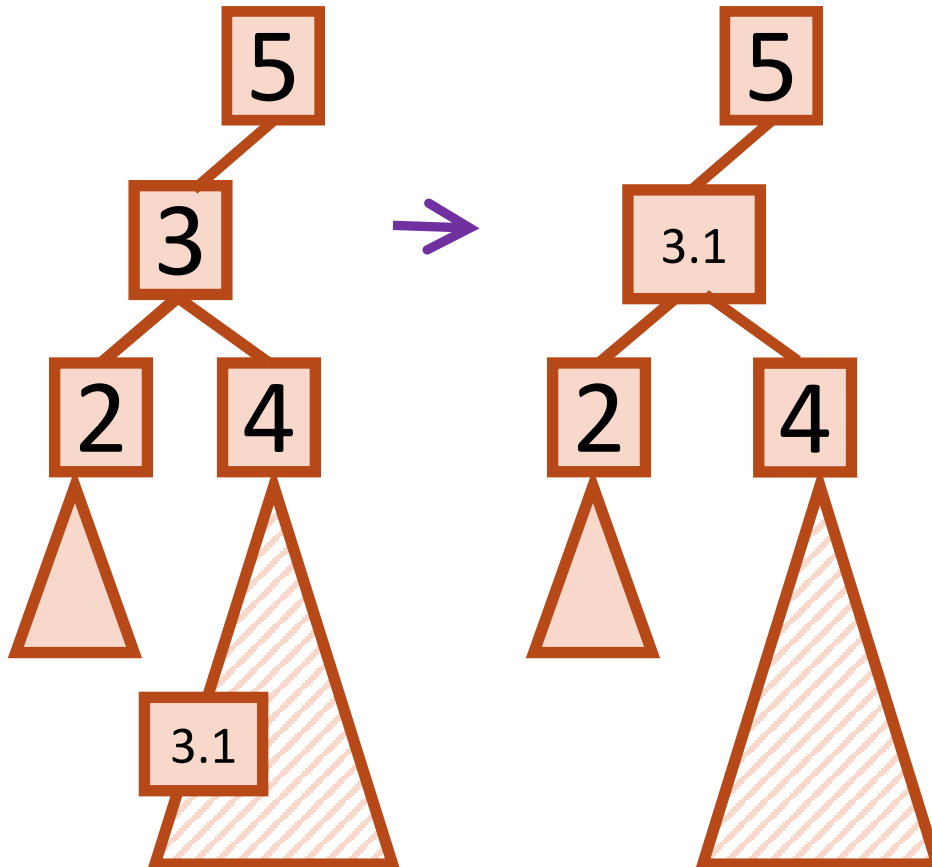


# DELETE in a Binary Search Tree

ctd.

This slide skipped  
in class – here for  
reference!

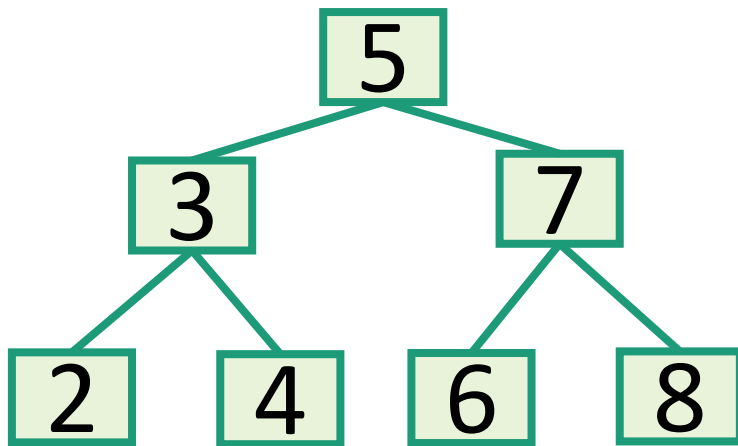
**Case 3:** if 3 has two children,  
replace 3 with its **immediate successor**.  
(aka, next biggest thing after 3)



- Does this maintain the BST property?
  - Yes.
- How do we find the immediate successor?
  - SEARCH for 3 in the subtree under 3.right
- How do we remove it when we find it?
  - If [3.1] has 0 or 1 children, do one of the previous cases.
- What if [3.1] has two children?
  - It doesn't.

# How long do these operations take?

- **SEARCH** is the big one.
  - Everything else just calls **SEARCH** and then does some small  $O(1)$ -time operation.



Time =  $O(\text{height of tree})$

Trees have depth  $O(\log(n))$ . **Done!**

Wait a second...



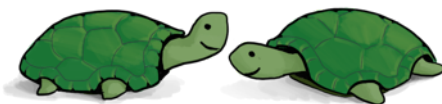
Lucky the  
lackadaisical lemur.



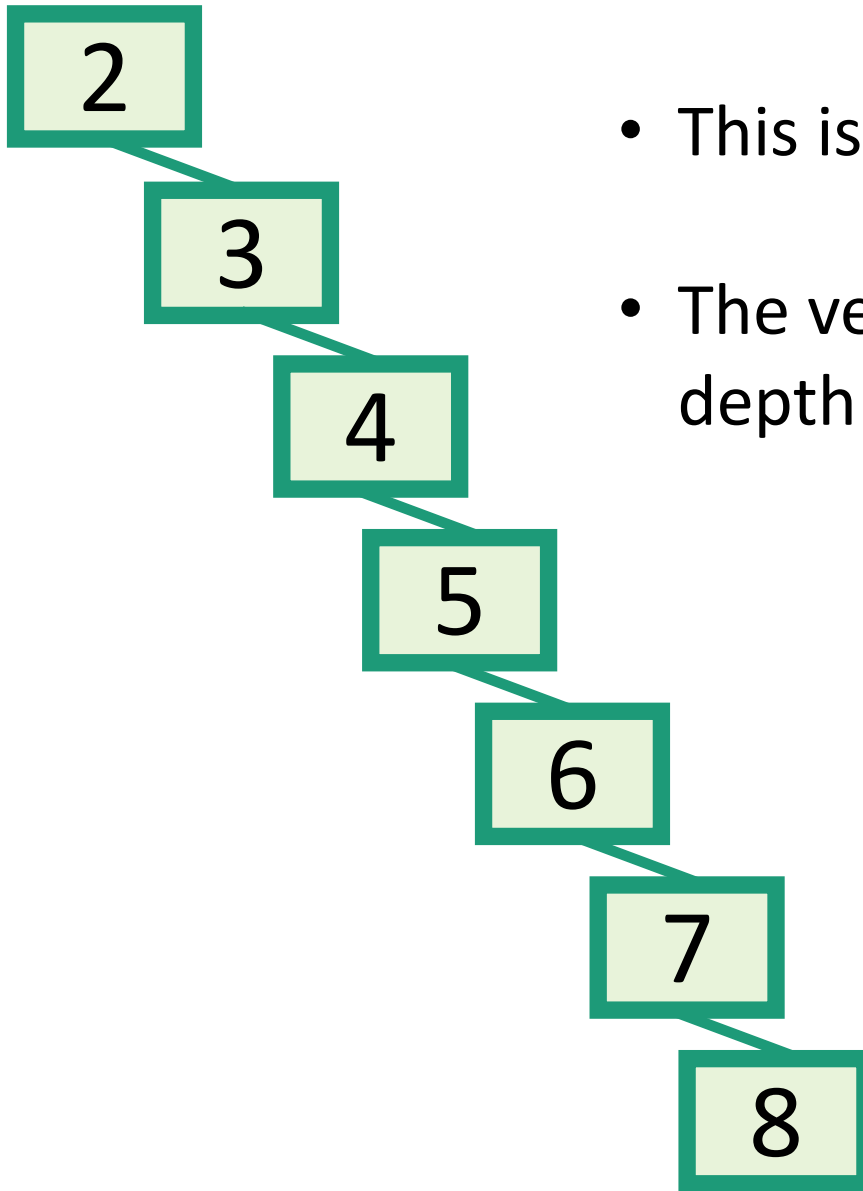
Plucky the  
Pedantic Penguin

## How long does search take?

1 minute think; 1 minute pair+share



# Search might take time $O(n)$ .



- This is a valid binary search tree.
- The version with  $n$  nodes has depth  $n$ , **not**  $O(\log(n))$ .

# What to do?

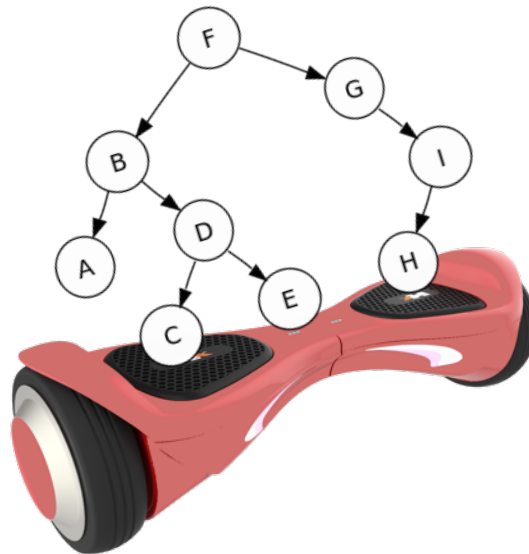
How often is “every so often” in the worst case?  
It’s actually pretty often!



Ollie the over-achieving ostrich

- Goal: Fast **SEARCH/INSERT/DELETE**
- All these things take time  $O(\text{height})$
- And the height might be big!!! ☹️
- Idea 0:
  - Keep track of how deep the tree is getting.
  - If it gets too tall, re-do everything from scratch.
    - At least  $\Omega(n)$  every so often....
- Turns out that’s not a great idea. Instead we turn to...

# Self-Balancing Binary Search Trees

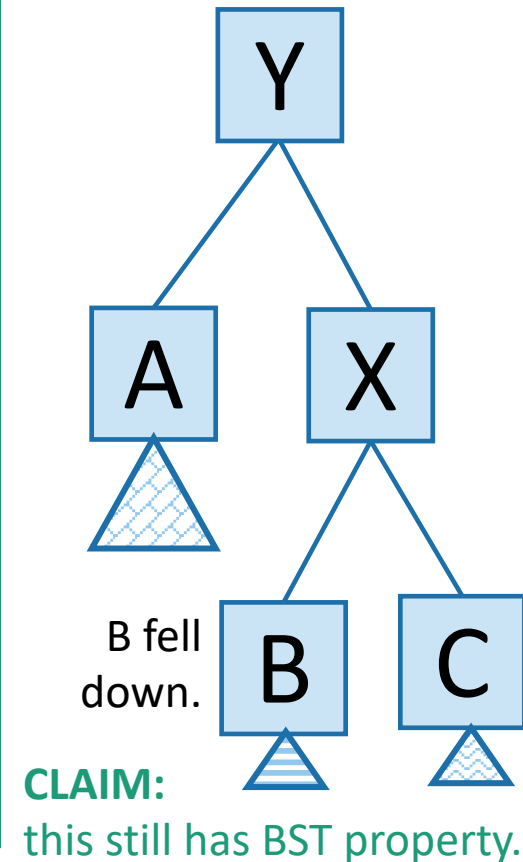
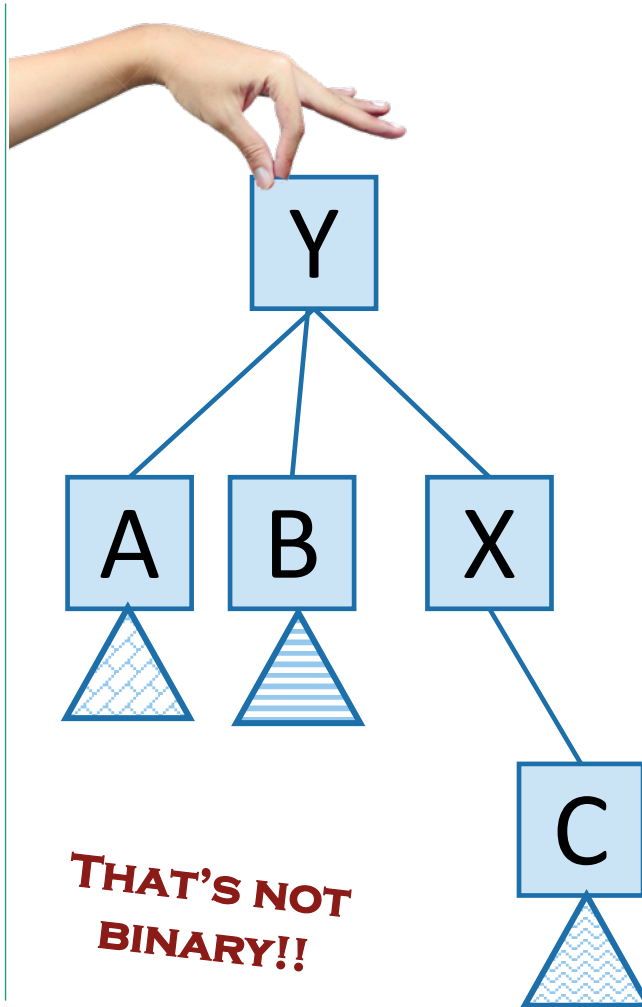
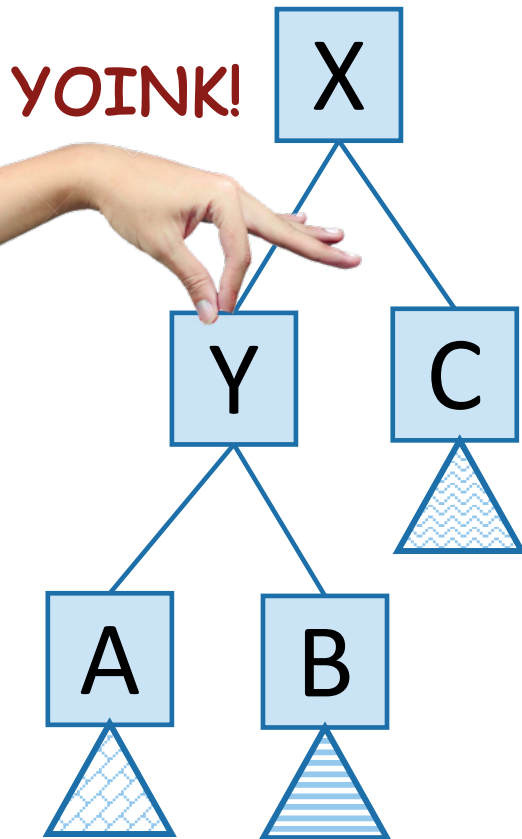


# Idea 1: Rotations

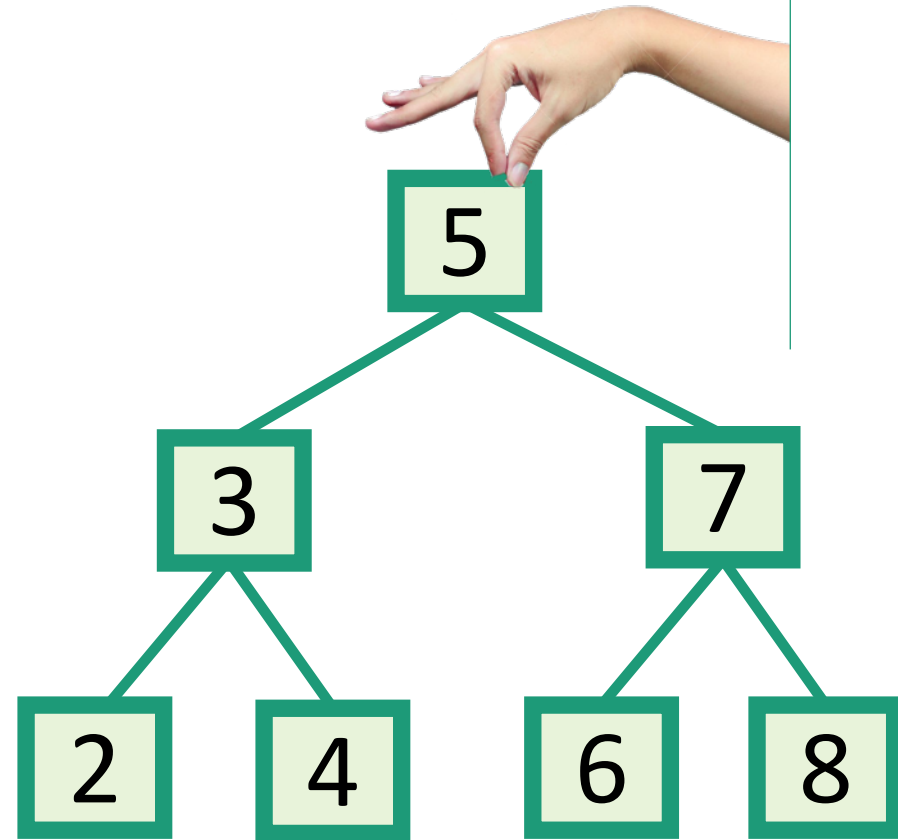
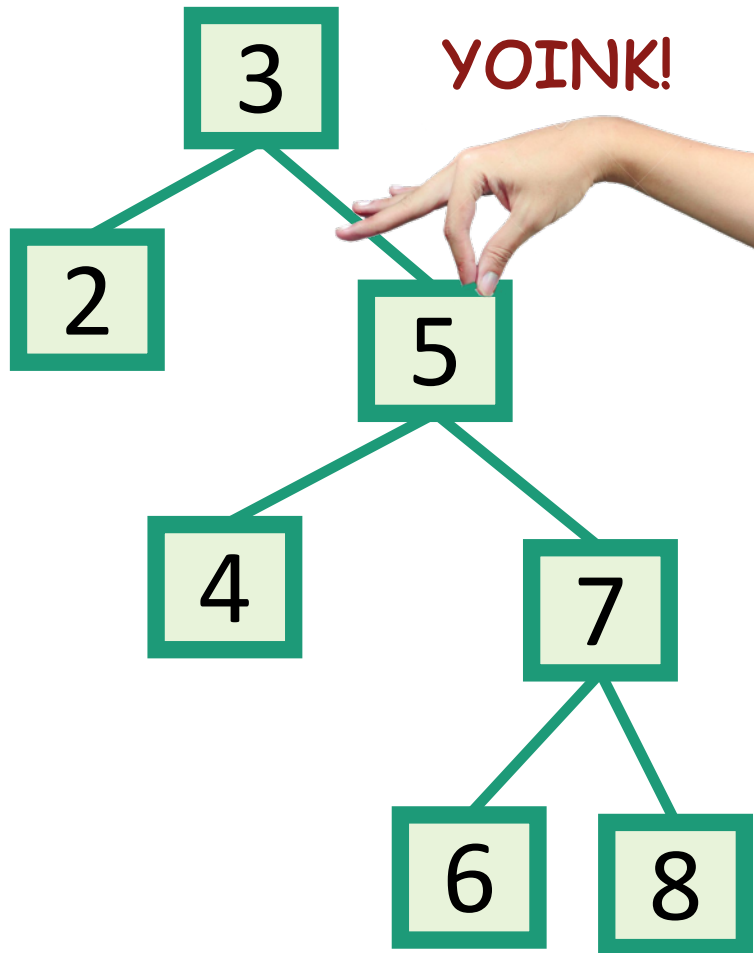
No matter what lives underneath A,B,C,  
this takes time  $O(1)$ . (Why?)

- Maintain Binary Search Tree (BST) property, while moving stuff around.

Note: A, B, C, X, Y are  
variable names, not the  
contents of the nodes.



This seems helpful





# Strategy?

- Whenever something seems unbalanced, do rotations until it's okay again.

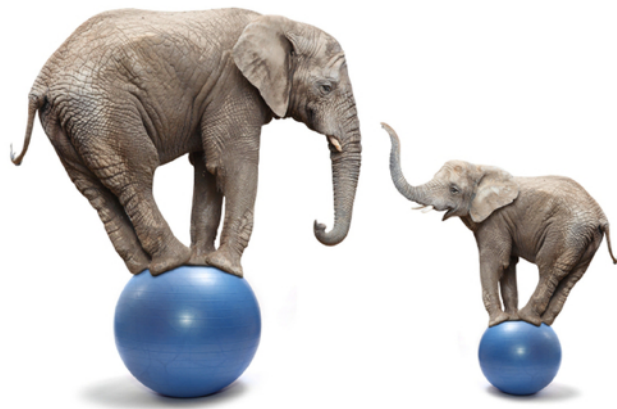


Lucky the Lackadaisical Lemur

Even for Lucky this is pretty vague.  
What do we mean by “seems unbalanced”? What’s “okay”?

# Idea 2: have some proxy for balance

- Maintaining **perfect balance** is too hard.
- Instead, come up with some **proxy for balance**:
  - If the tree satisfies **[SOME PROPERTY]**, then it's pretty balanced.
  - We can maintain **[SOME PROPERTY]** using rotations.



There are actually several ways to do this, but today we'll see...

# Red-Black Trees

- A Binary Search Tree that balances itself!
- No more time-consuming by-hand balancing!
- Be the envy of your friends and neighbors with the time-saving...

*Red-Black tree!*

Maintain balance by stipulating that **black nodes** are balanced, and that there aren't too many **red nodes**.

*It's just good sense!*





Leo Guibas

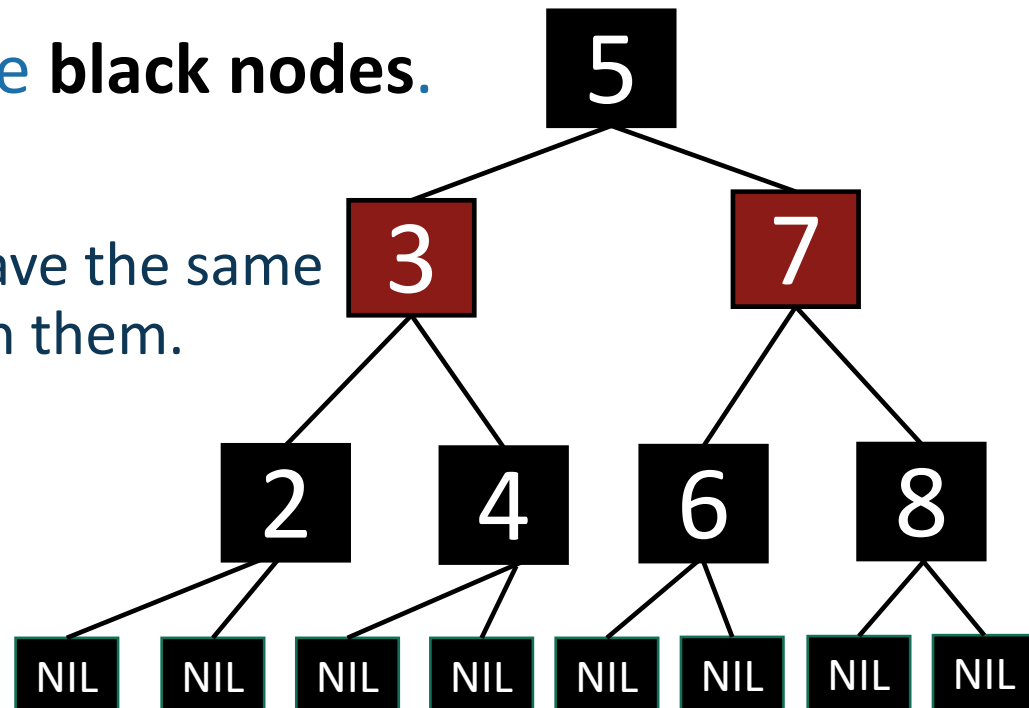


Bob Sedgewick

# Red-Black Trees

obey the following rules (which are a proxy for balance)

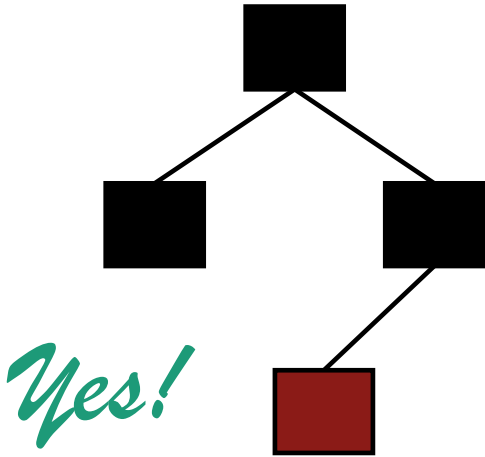
- Every node is colored **red** or **black**.
- The root node is a **black node**.
- NIL children count as **black nodes**.
- Children of a **red node** are **black nodes**.
- For all nodes  $x$ :
  - all paths from  $x$  to NIL's have the same number of **black nodes** on them.



I'm not going to draw the NIL children in the future, but they are treated as black nodes.

# Examples(?)

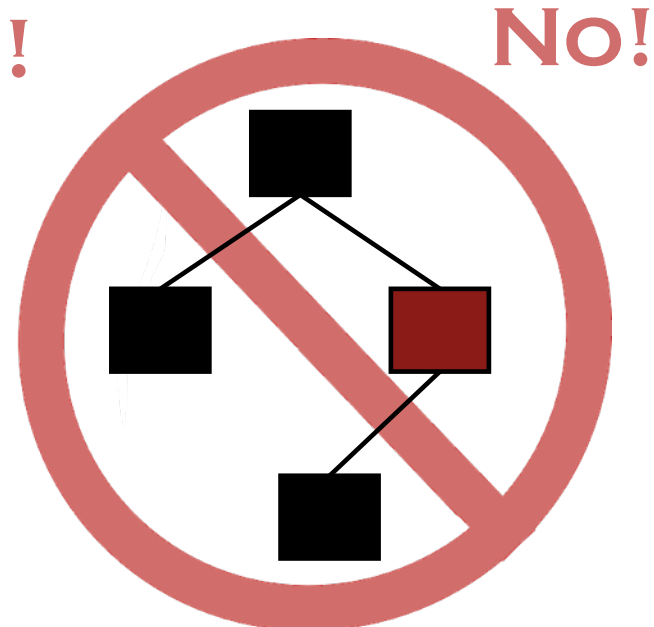
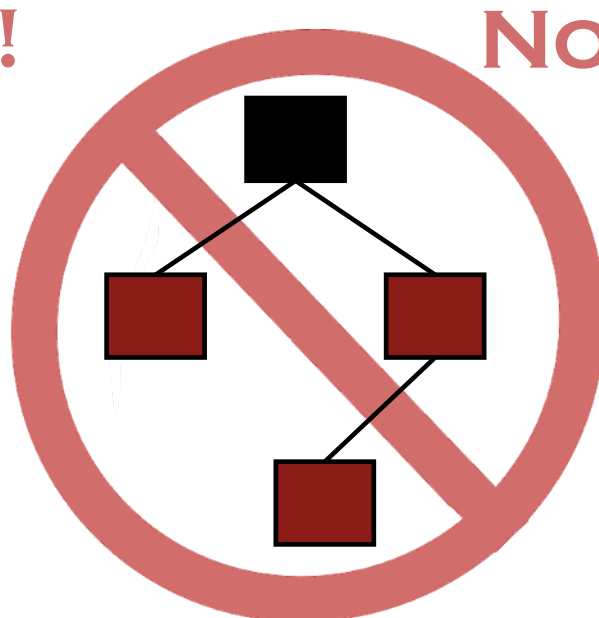
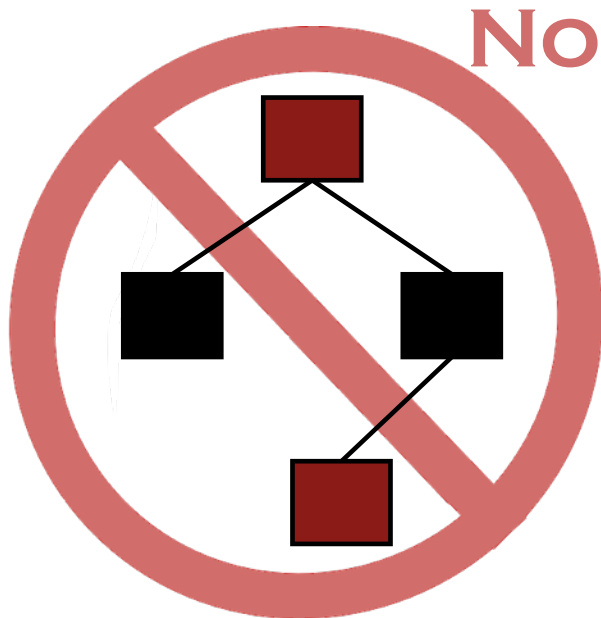
- Every node is colored **red** or **black**.
- The root node is a **black node**.
- NIL children count as **black nodes**.
- Children of a **red node** are **black nodes**.
- For all nodes x:
  - all paths from x to NIL's have the same number of **black nodes** on them.



Which of these  
are red-black trees?  
(NIL nodes not drawn)



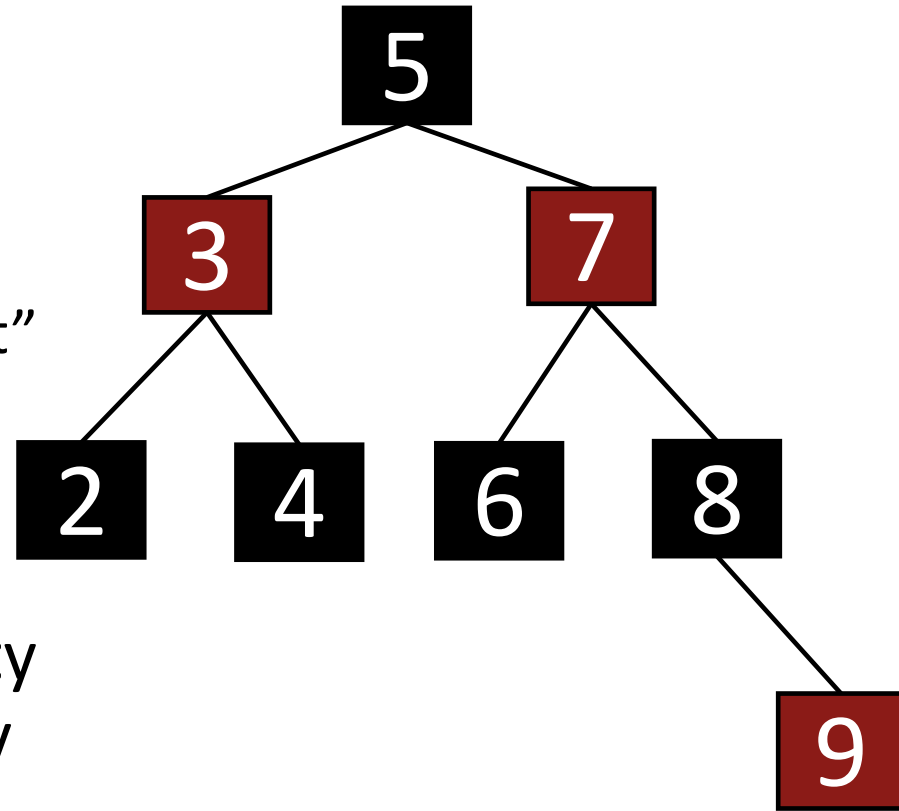
1 minute think  
1 minute share



Break

# Why these rules??????

- This is pretty balanced.
  - The **black nodes** are balanced
  - The **red nodes** are “spread out” so they don’t mess things up too much.
- We can maintain this property as we insert/delete nodes, by using rotations.



This is the really clever idea!

This **Red-Black** structure is a **proxy for balance**.

It’s just a smidge weaker than perfect balance, but we can actually maintain it!



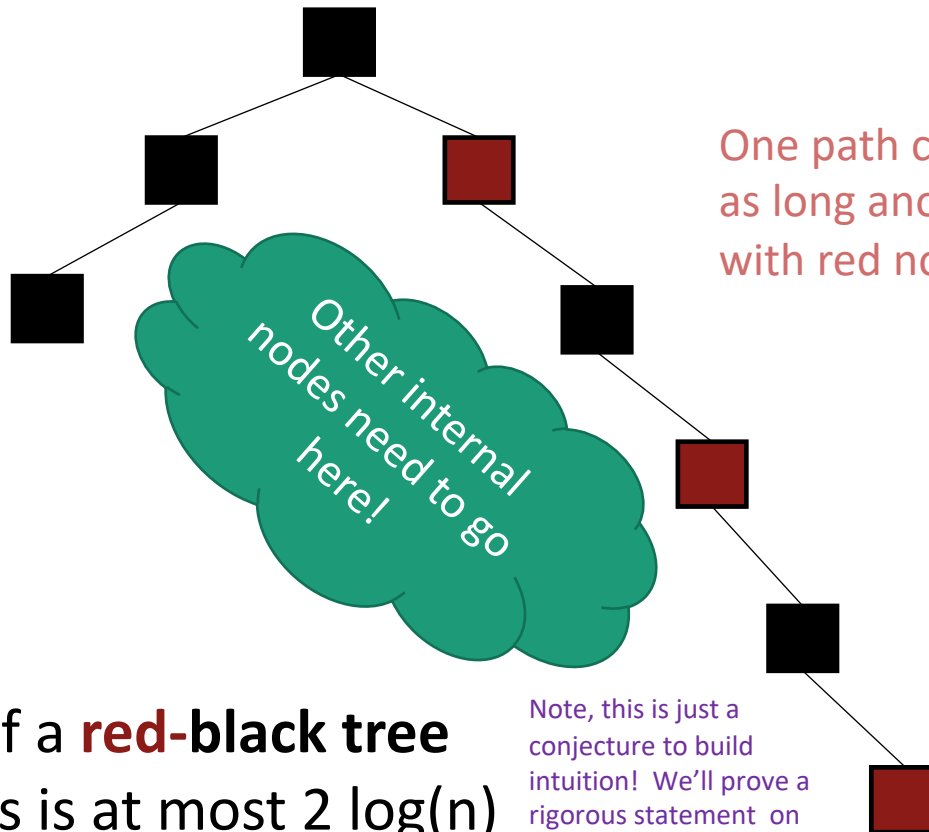
Let's build some intuition!



Lucky the  
lackadaisical  
lemur

# This is “pretty balanced”

- To see why, intuitively, let's try to build a Red-Black Tree that's unbalanced.



One path can be at most twice  
as long another if we pad it  
with red nodes.

**Conjecture:**  
the height of a **red-black tree**  
with  $n$  nodes is at most  $2 \log(n)$

Note, this is just a  
conjecture to build  
intuition! We'll prove a  
rigorous statement on  
the next slide.



- 
- The diagram shows a tree structure within a light green rounded rectangle. At the top is a red square node labeled 'x'. It has two children: a red square node labeled 'y' on the left and a red square node labeled 'z' on the right. Node 'y' has three children, all represented by black squares, arranged vertically. A thick purple line highlights a path starting from node 'x', going to its left child 'y', then to the top black square child of 'y', then to the middle black square child, and finally to the bottom black square child, which is labeled 'NIL' in a black box. A small penguin is in the top right corner, and a green arrow points upwards from the bottom center towards the tree.

Then:

$$\geq 2^{\text{height}/2} - 1 \quad \text{b(root) } \geq \text{height}/2 \text{ because of RBTree rules.}$$

$$n + 1 \geq 2^{height/2} \Rightarrow height \leq 2\log(n + 1)$$

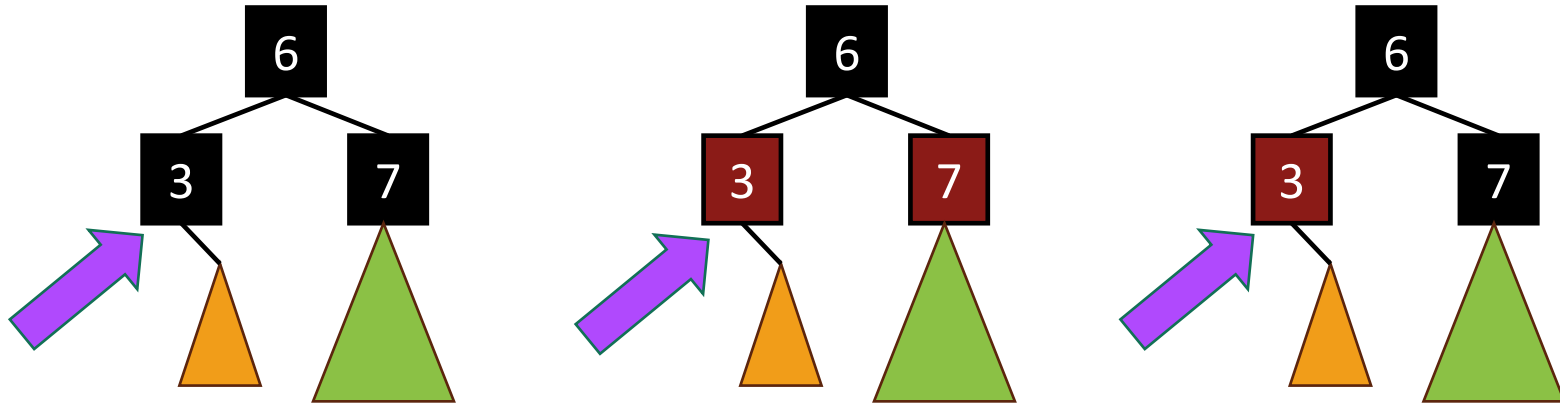
# This is great!

- SEARCH in an RBTree is immediately  $O(\log(n))$ , since the depth of an RBTree is  $O(\log(n))$ .
- What about INSERT/DELETE?
  - Turns out, you can INSERT and DELETE items from an RBTree in time  $O(\log(n))$ , while *maintaining* the RBTree property.
  - That's why this is a good property!

# INSERT/DELETE

- I expect we are out of time...
  - There are some slides which you can check out to see how to do INSERT/DELETE in RBTrees if you are curious.
  - See CLRS Ch 13. for even more details.
- You are **not responsible** for the details of INSERT/DELETE for RBTrees for this class.
  - You should know what the “proxy for balance” property is and why it ensures approximate balance.
  - You should know **that** this property can be efficiently maintained, but you do not need to know the details of how.

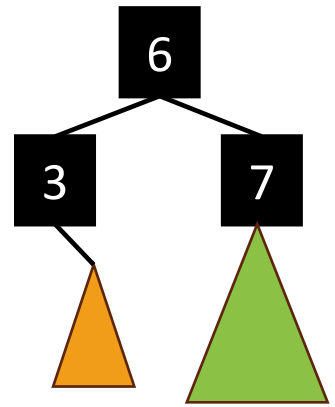
# INSERT: Many cases



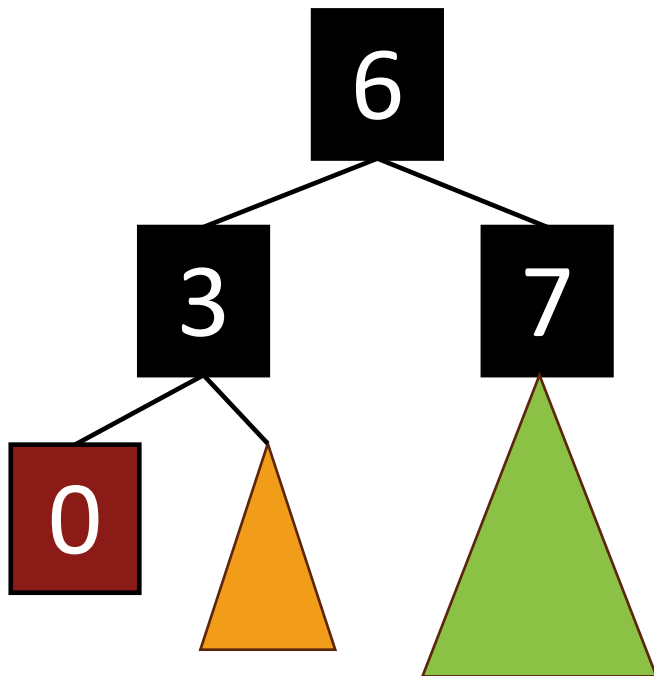
- Suppose we want to insert 0 **here**.
- There are 3 “important” cases for different colorings of the existing tree, and there are 9 more cases for all of the various symmetries of these 3 cases.

# INSERT: Case 1

- Make a new **red node**.
- Insert it as you would normally.



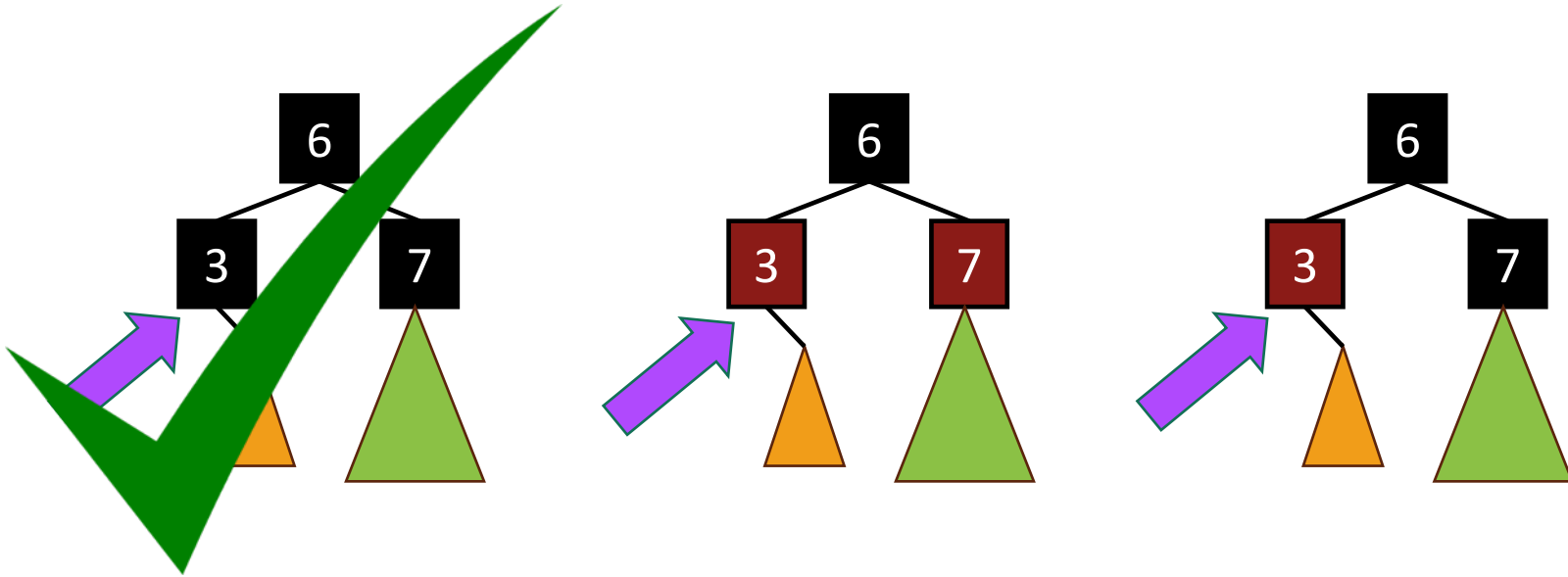
What if it looks like this?



Example: insert 0



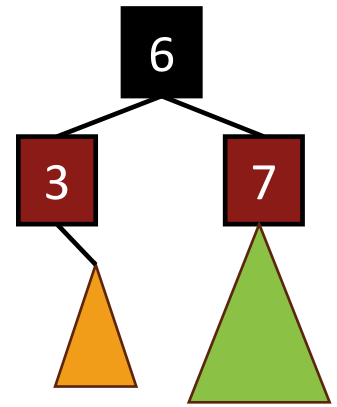
# INSERT: Many cases



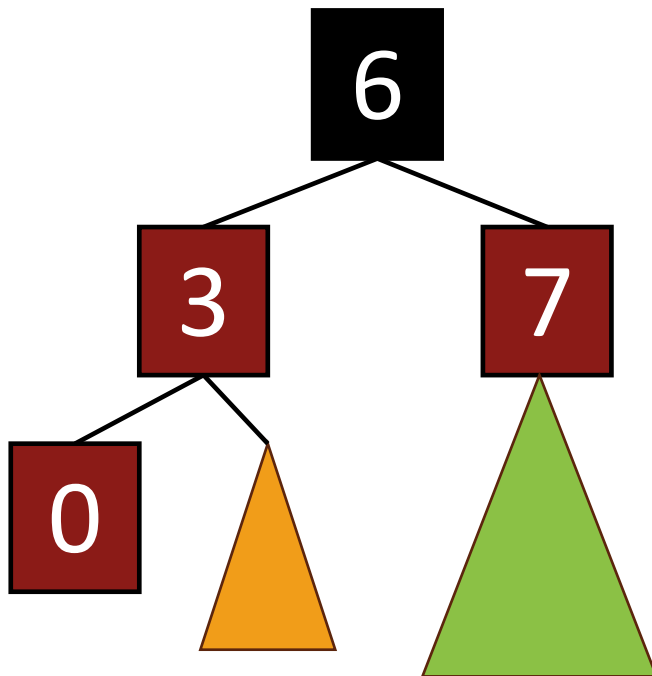
- Suppose we want to insert 0 **here**.
- There are 3 “important” cases for different colorings of the existing tree, and there are 9 more cases for all of the various symmetries of these 3 cases.

# INSERT: Case 2

- Make a new **red node**.
- Insert it as you would normally.
- Fix things up if needed.



What if it looks like this?



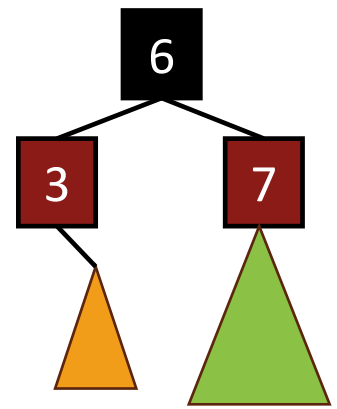
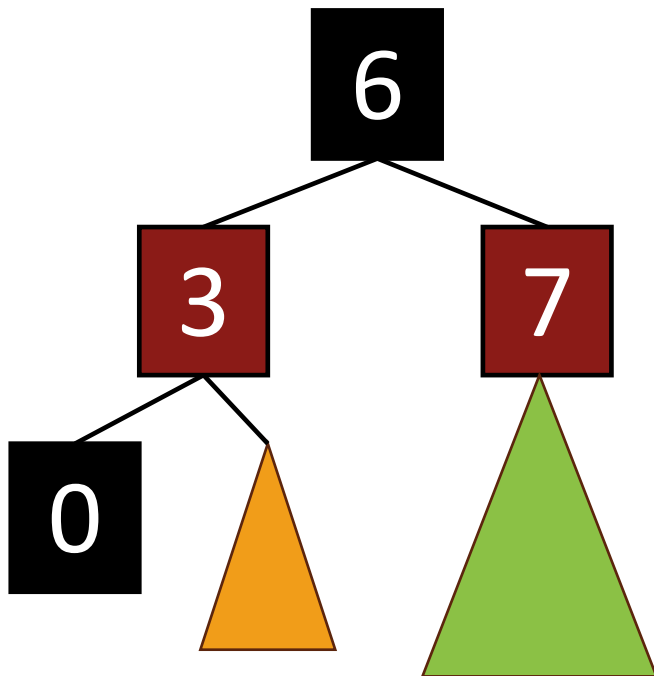
Example: insert 0





# INSERT: Case 2

- Make a new **red node**.
- Insert it as you would normally.
- Fix things up if needed.



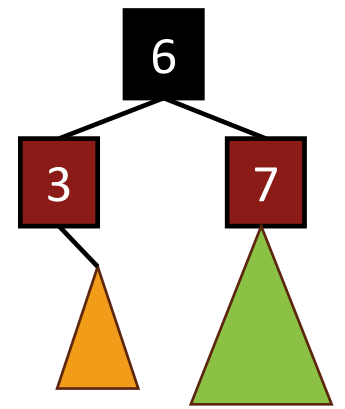
What if it looks like this?

Example: insert 0

Can't we just insert 0 as a **black node**?

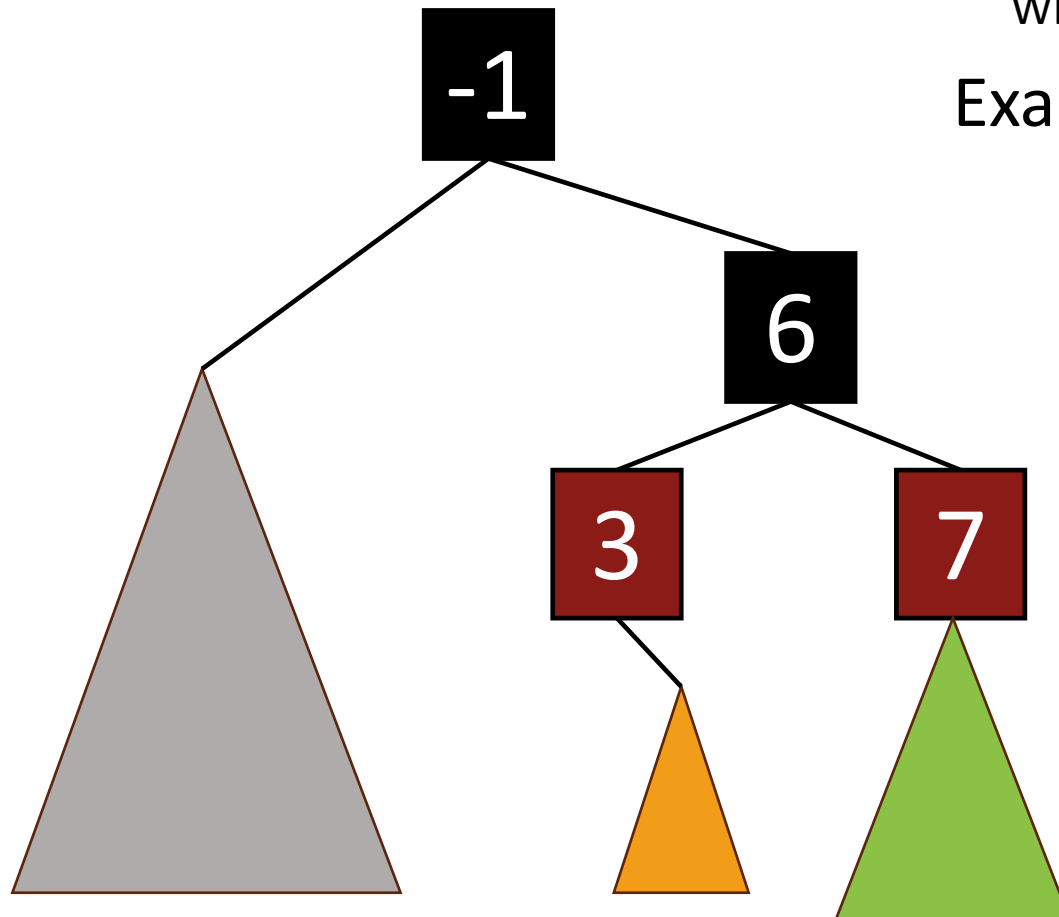


# We need a bit more context



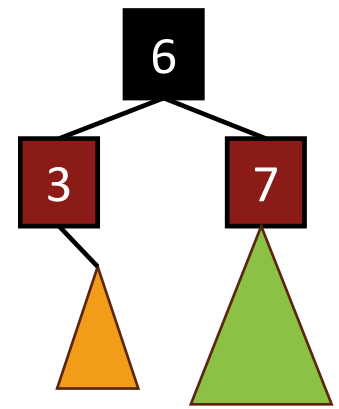
What if it looks like this?

Example: insert 0



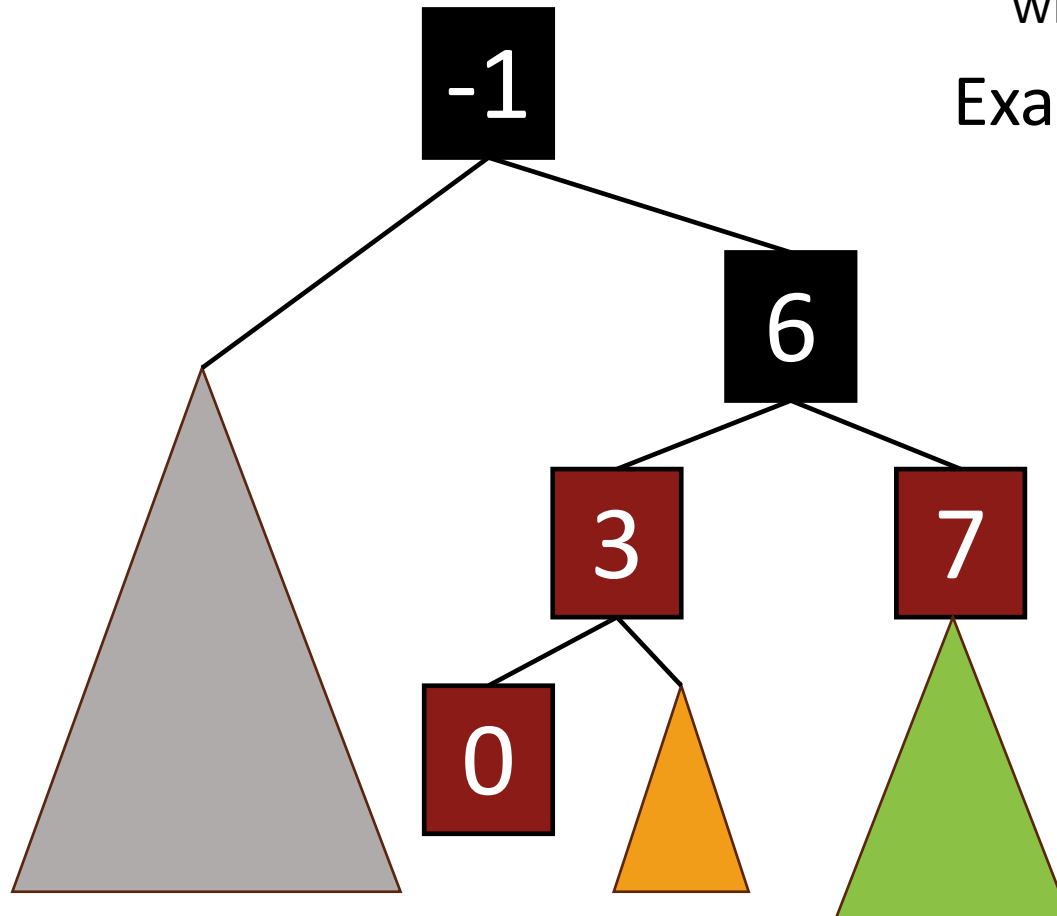
# We need a bit more context

- Add 0 as a red node.



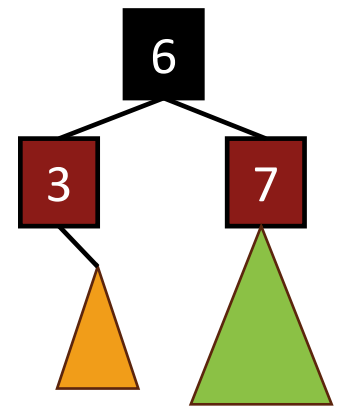
What if it looks like this?

Example: insert 0



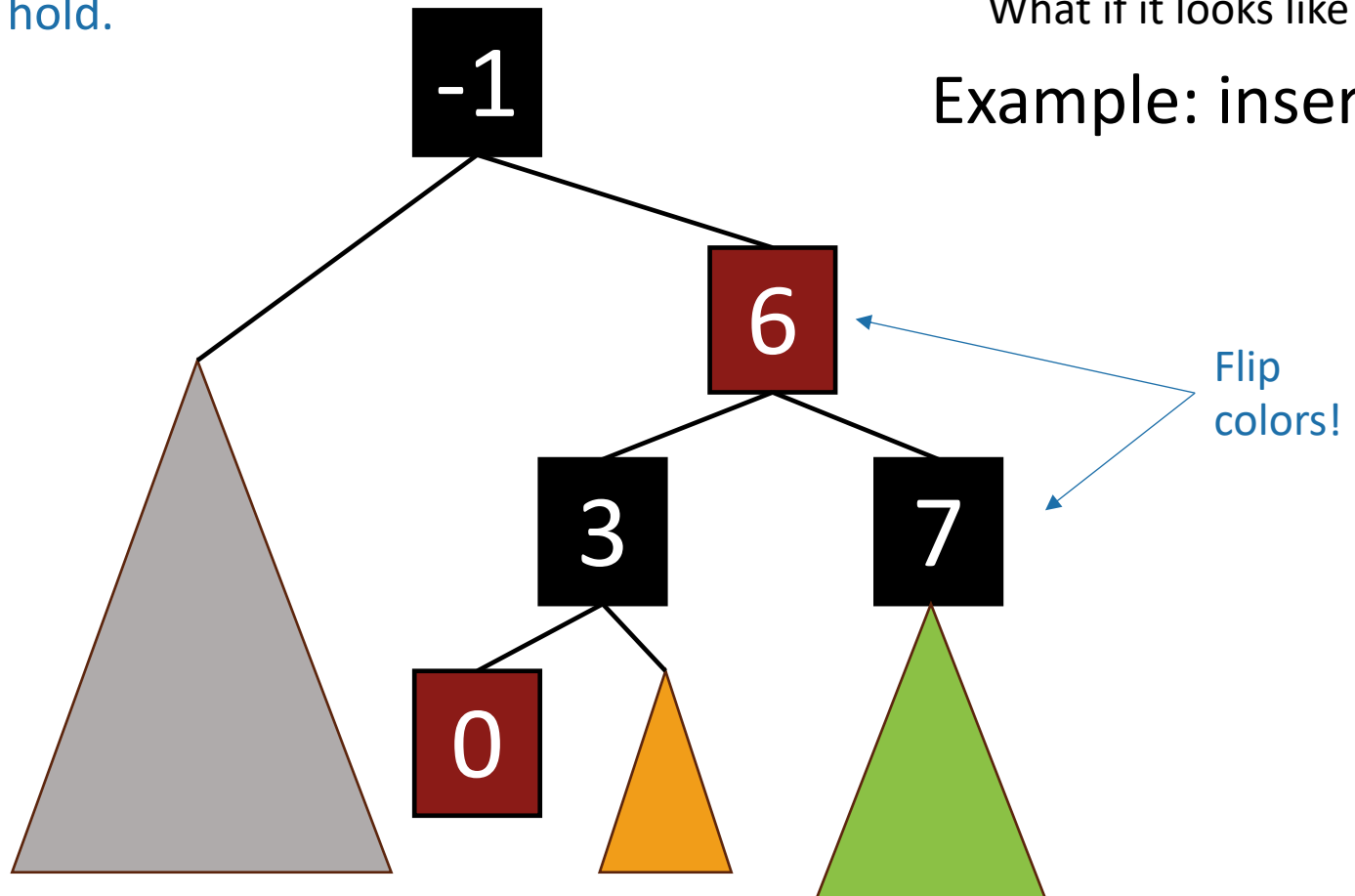
# We need a bit more context

- Add 0 as a red node.
- **Claim:** RB-Tree properties still hold.

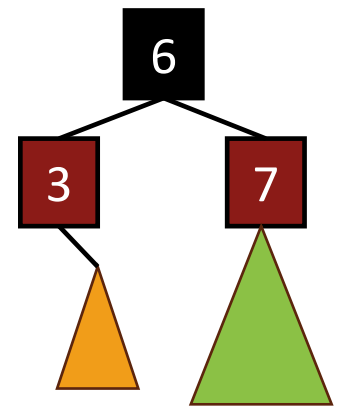
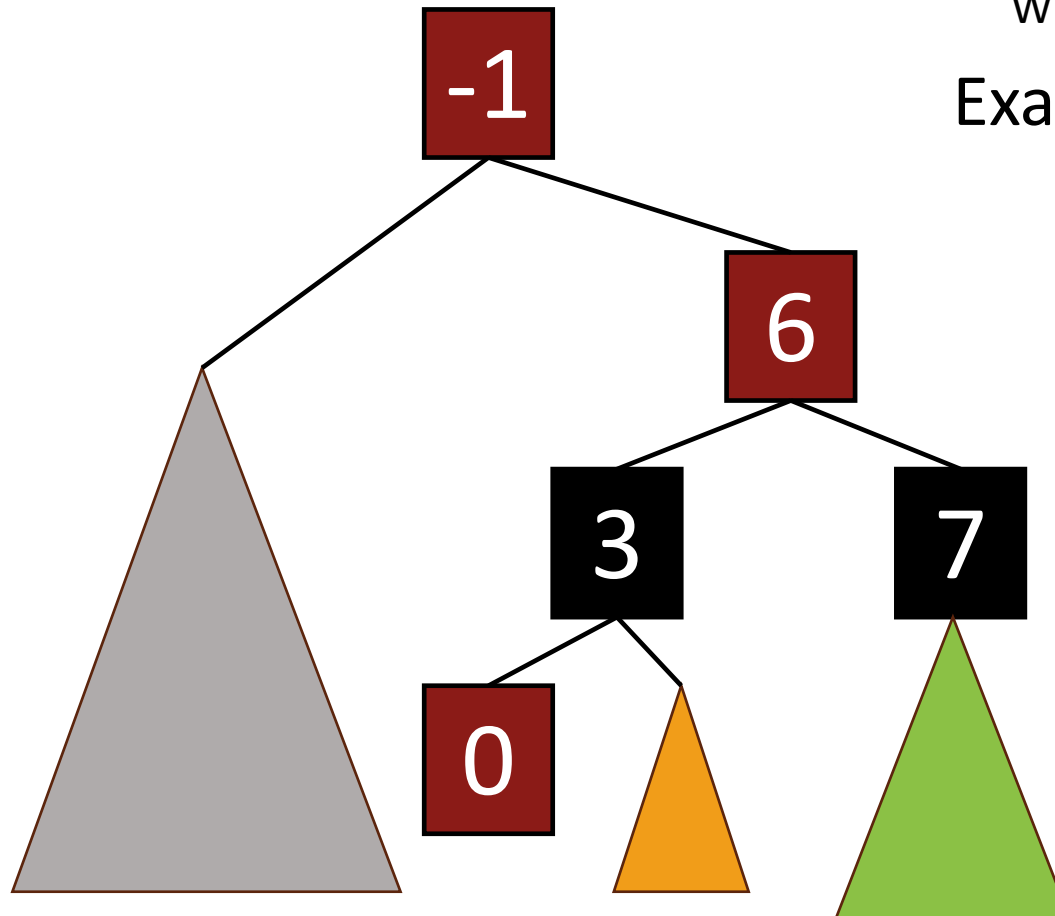
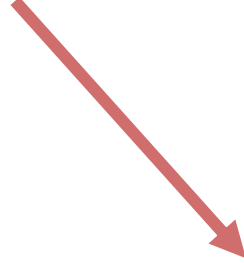


What if it looks like this?

Example: insert 0



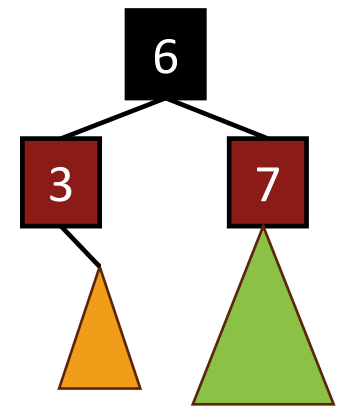
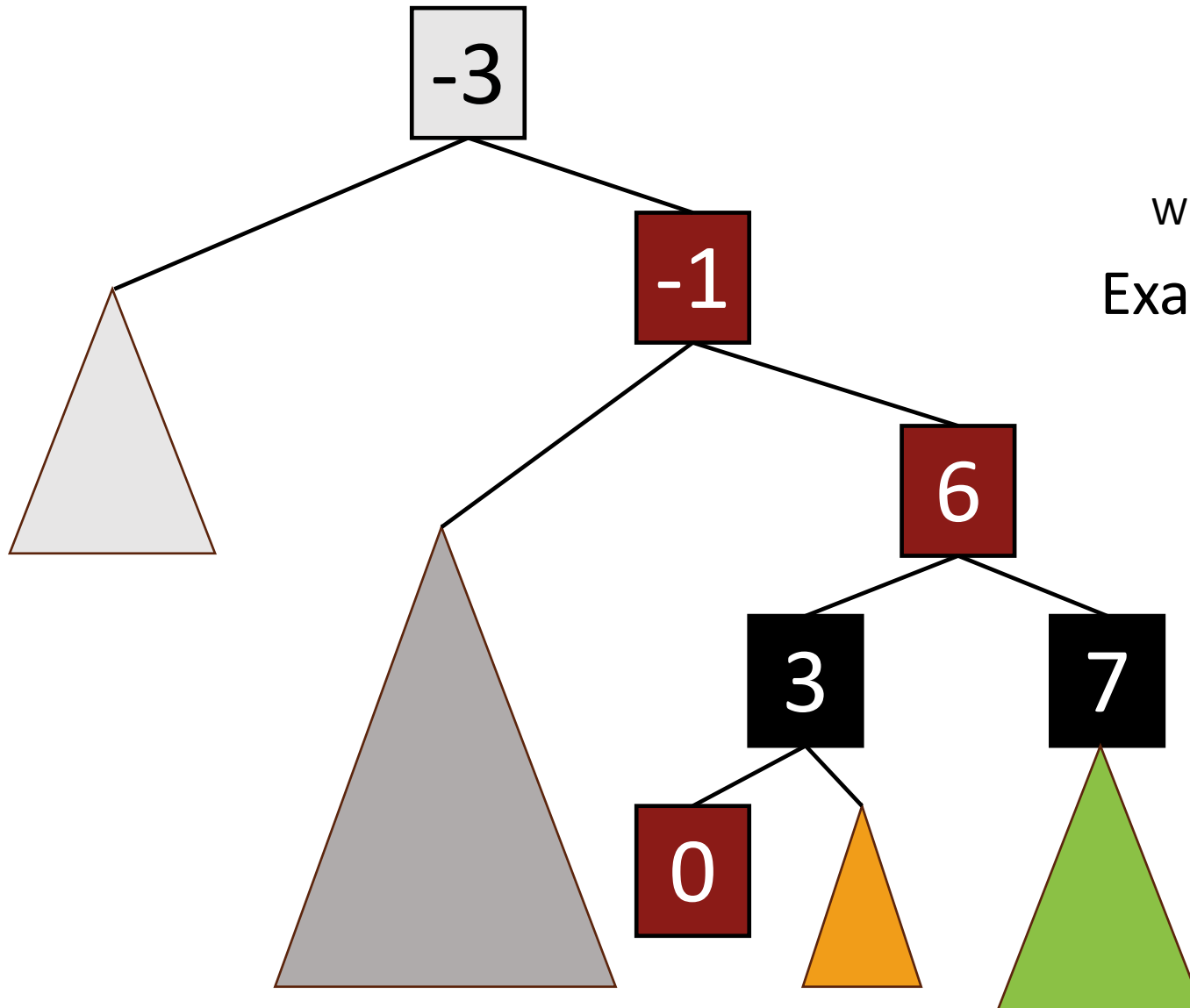
But what if **that** was red?



What if it looks like this?

Example: insert 0

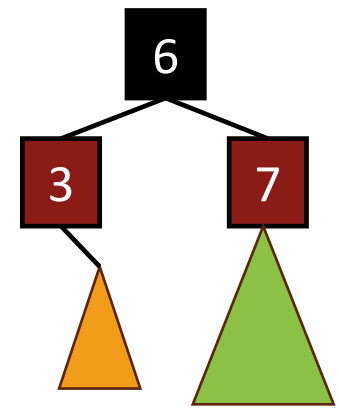
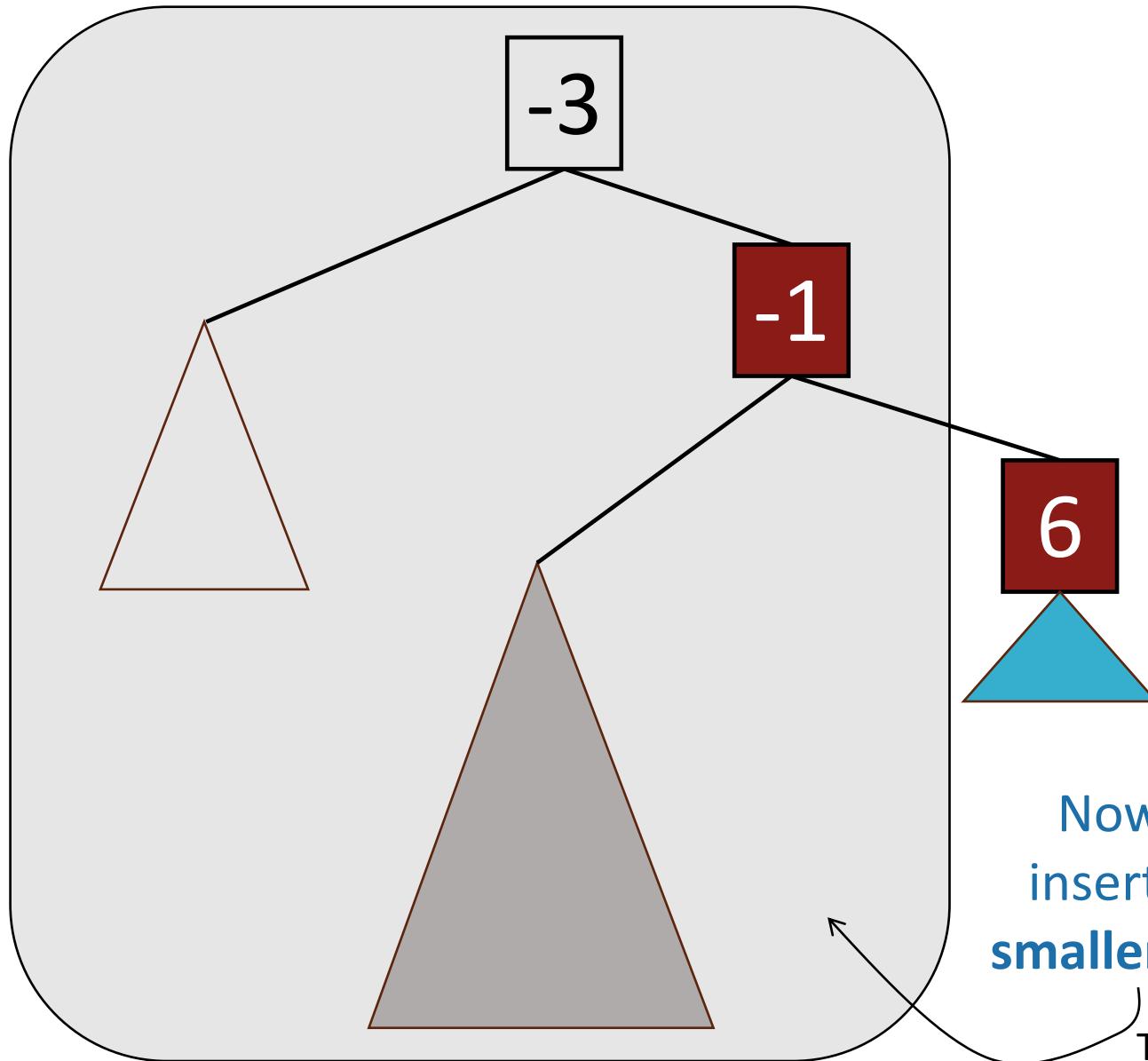
# More context...



What if it looks like this?

Example: insert 0

# More context...



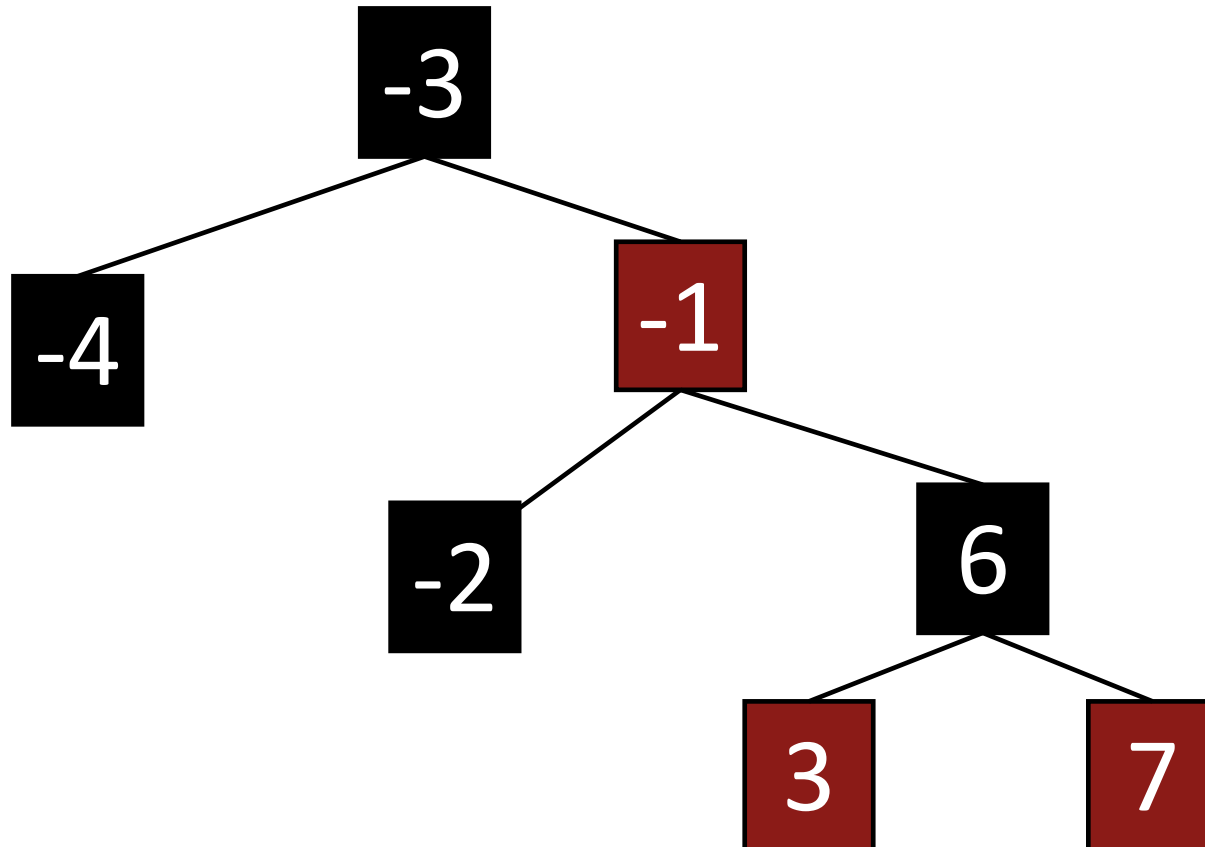
What if it looks like this?

Example: insert 0

Now we're basically  
inserting 6 into some  
**smaller tree**. Recurse!

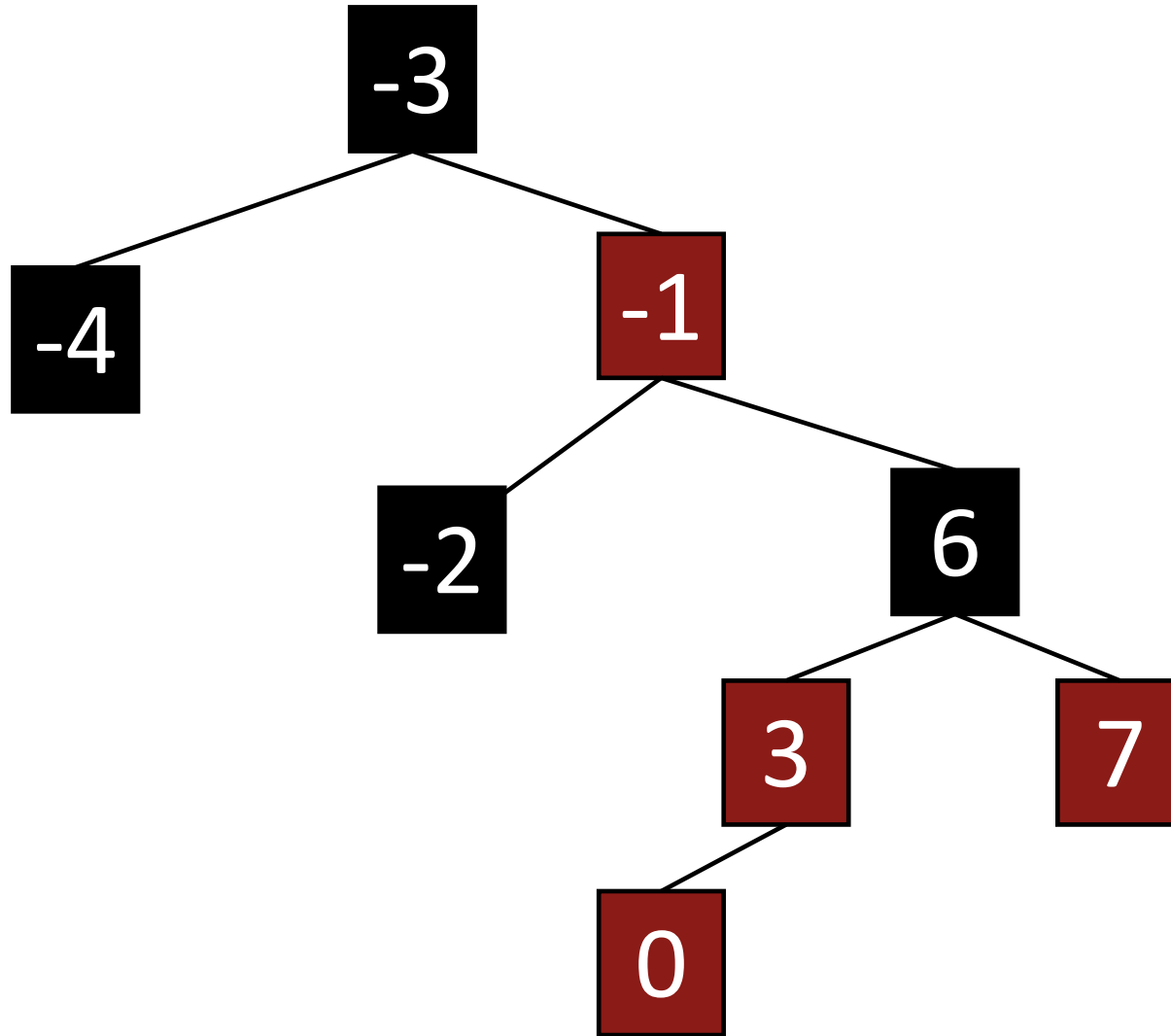
This one!

# Example, part I

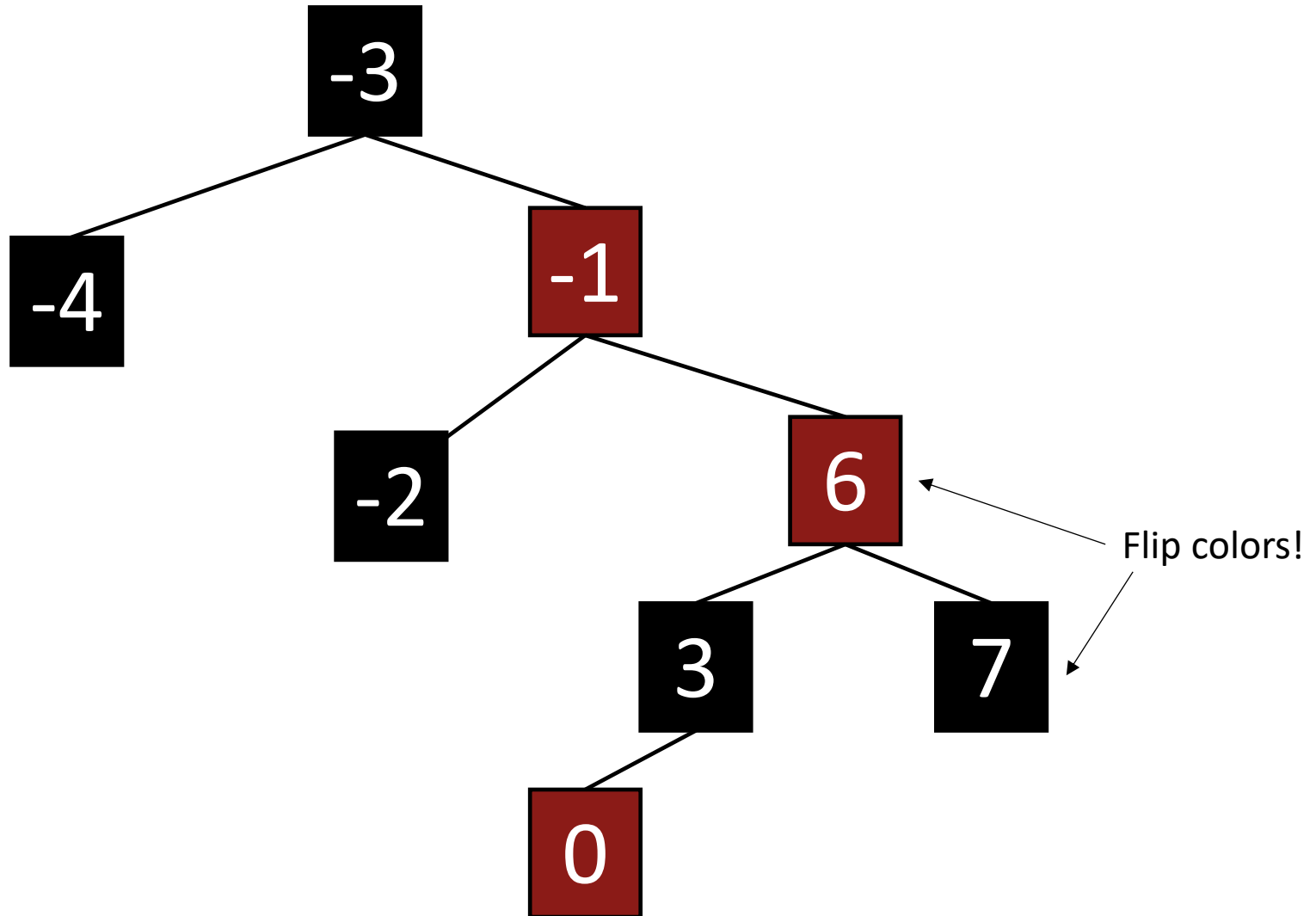




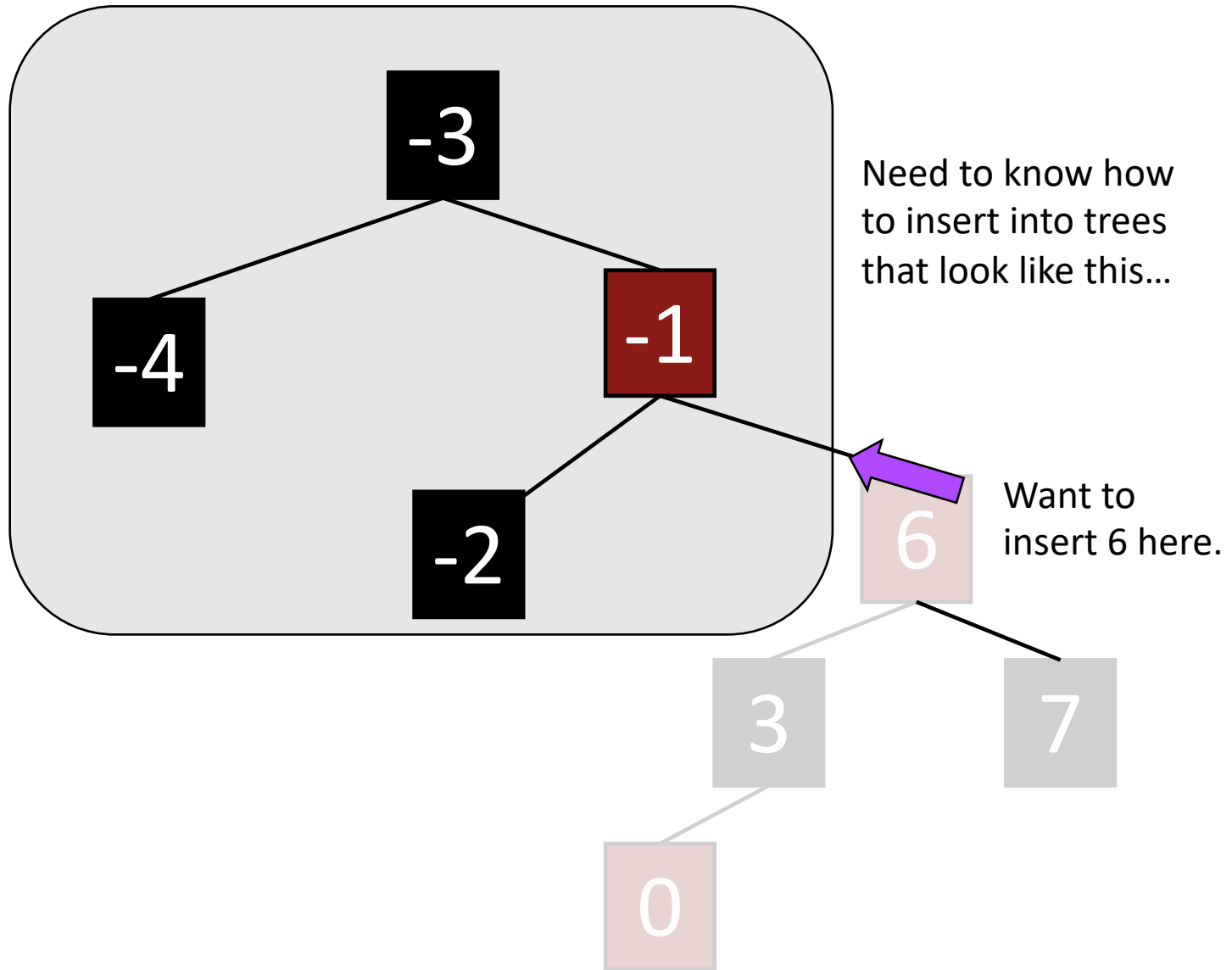
# Example, part I



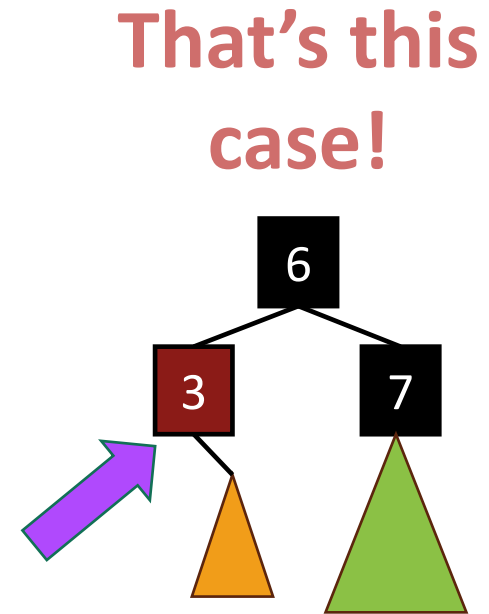
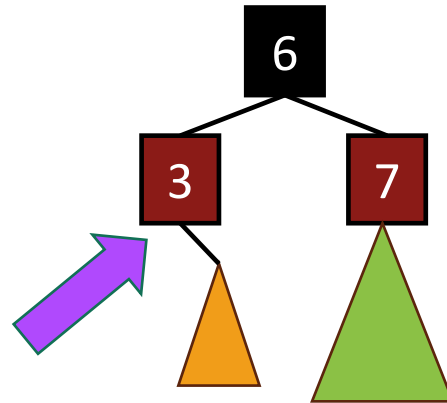
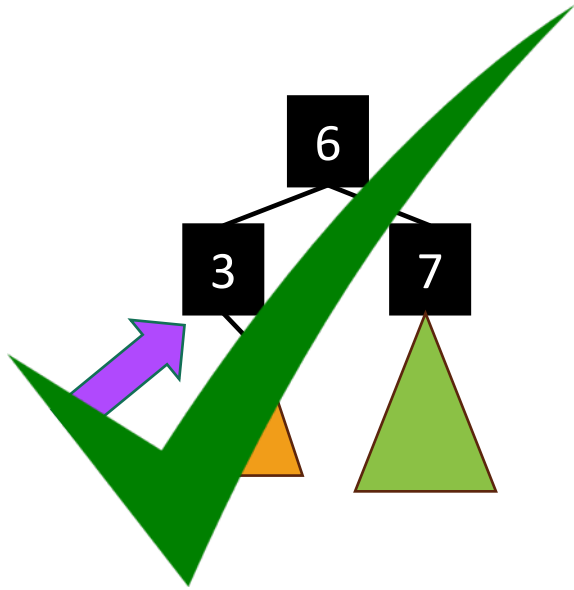
# Example, part I



# Example, part I



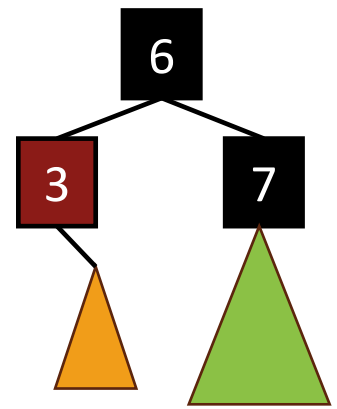
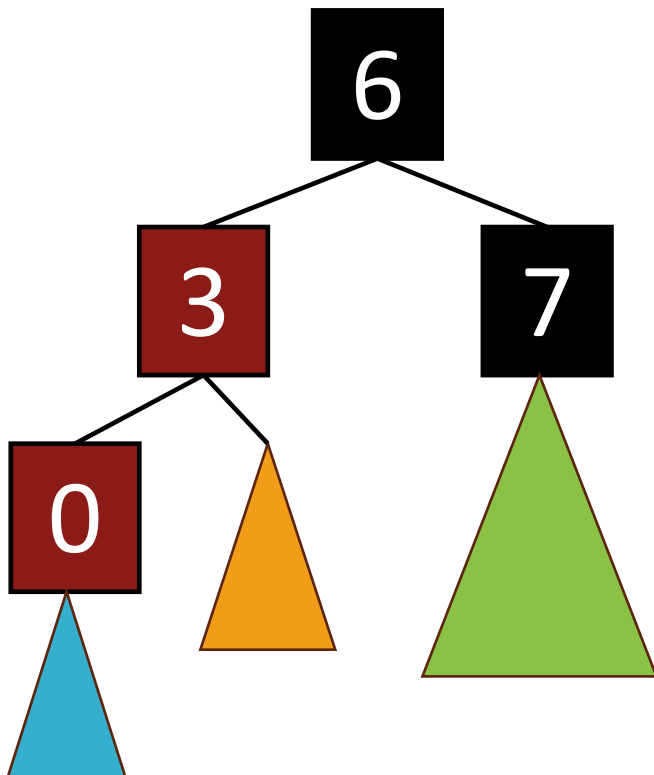
# INSERT: Many cases



- Suppose we want to insert 0 **here**.
- There are 3 “important” cases for different colorings of the existing tree, and there are 9 more cases for all of the various symmetries of these 3 cases.

# INSERT: Case 3

- Make a new **red node**.
- Insert it as you would normally.
- Fix things up if needed.



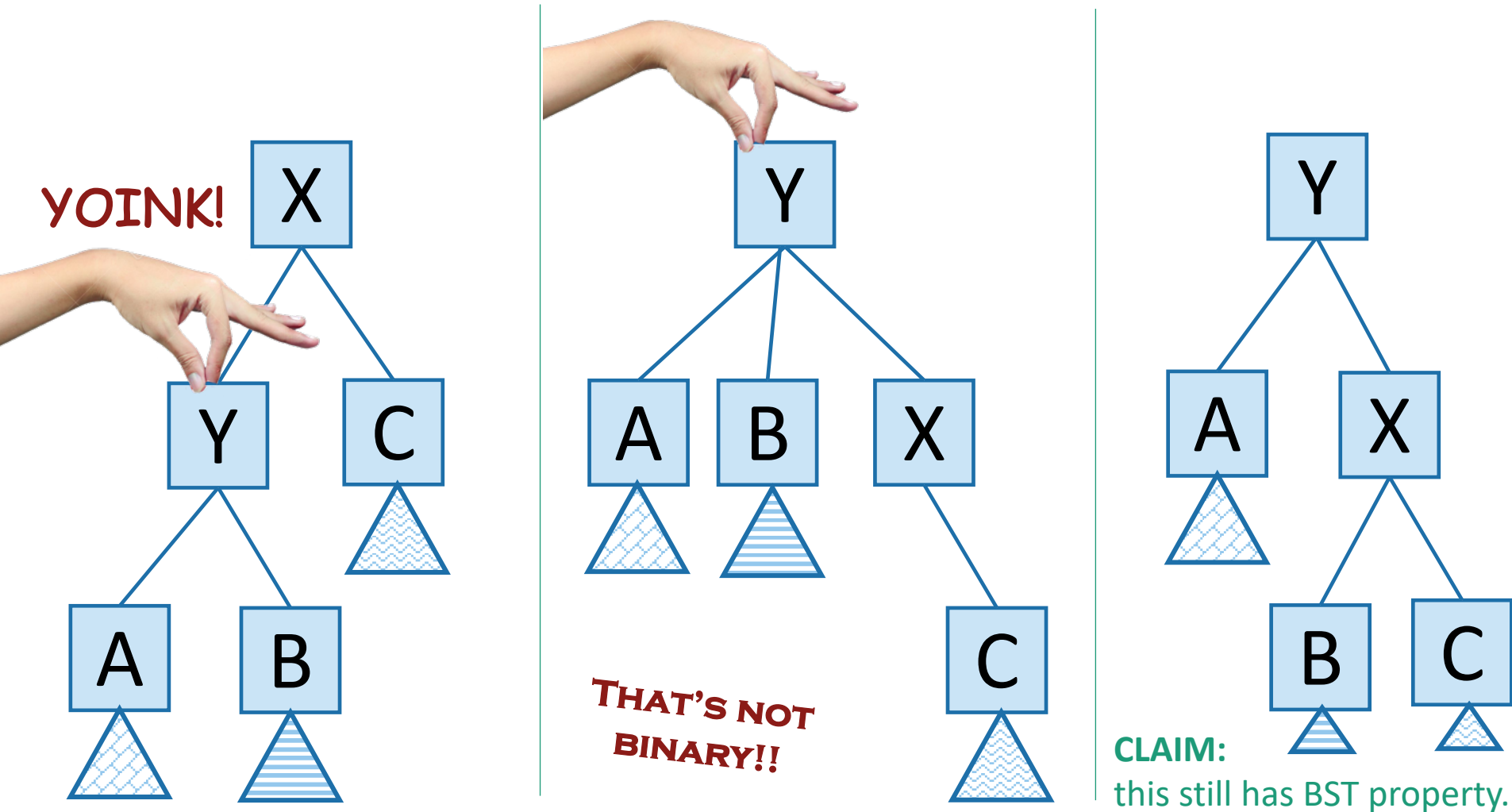
What if it looks like this?

Example: Insert 0.

- Maybe with a subtree below it.

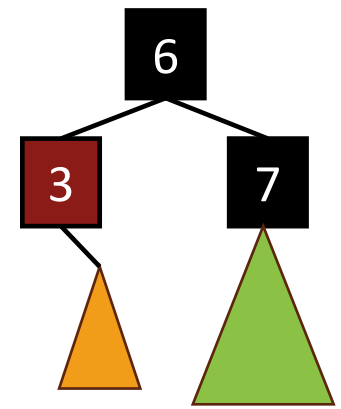
# Recall Rotations

- Maintain Binary Search Tree (BST) property, while moving stuff around.



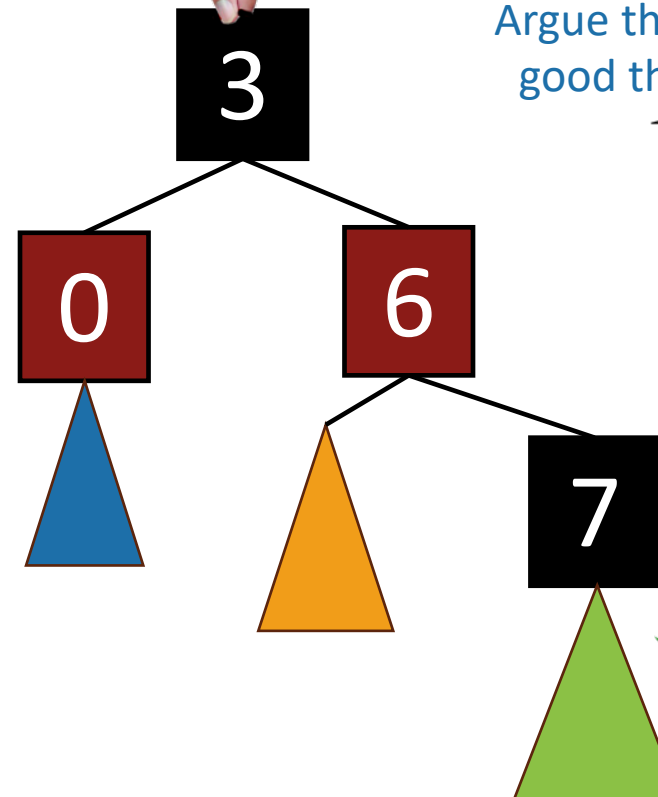
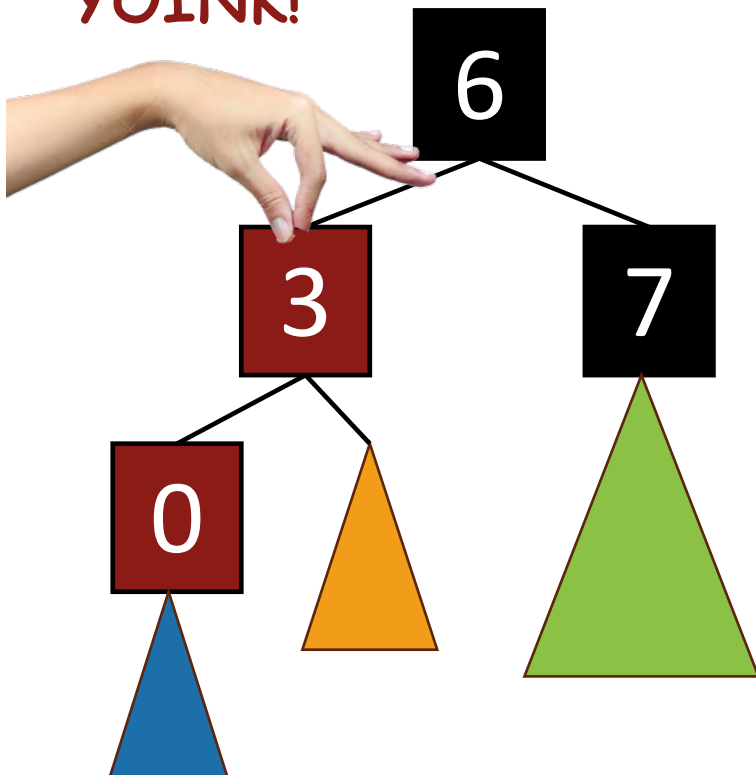
# Inserting into a Red-Black Tree

- Make a new **red node**.
- Insert it as you would normally.
- **Fix things up if needed.**



What if it looks like this?

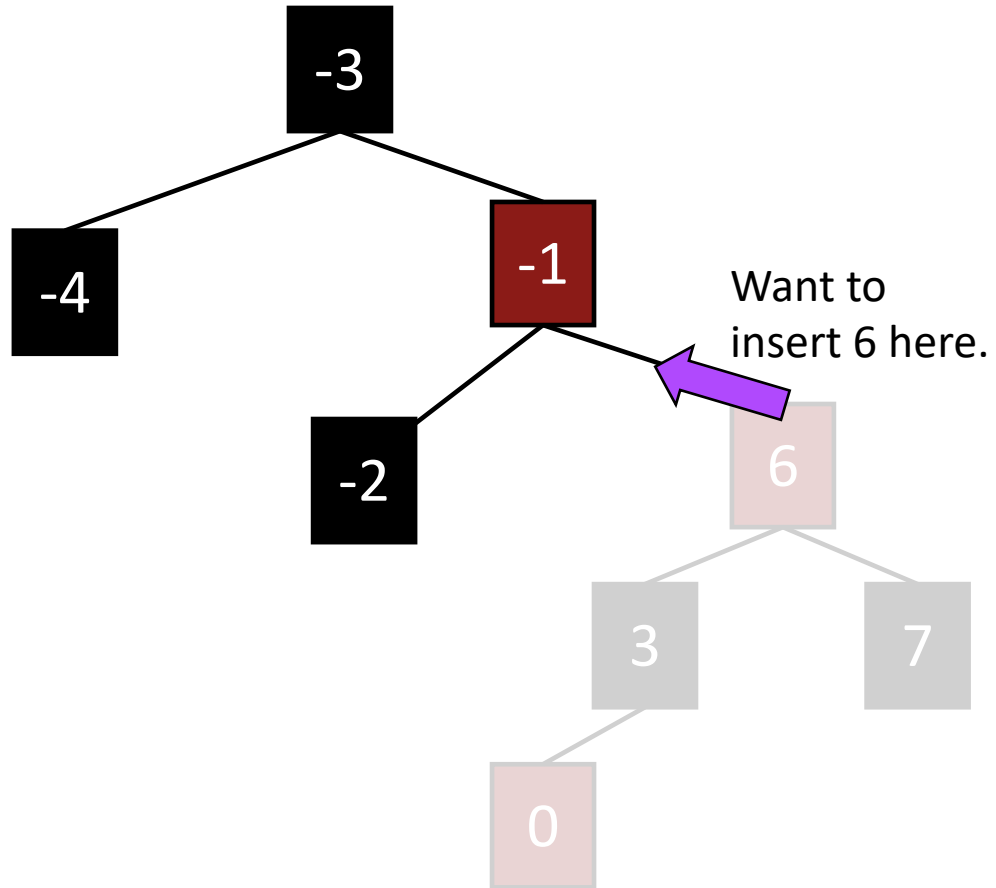
**YOINK!**



Argue that this is a good thing to do!

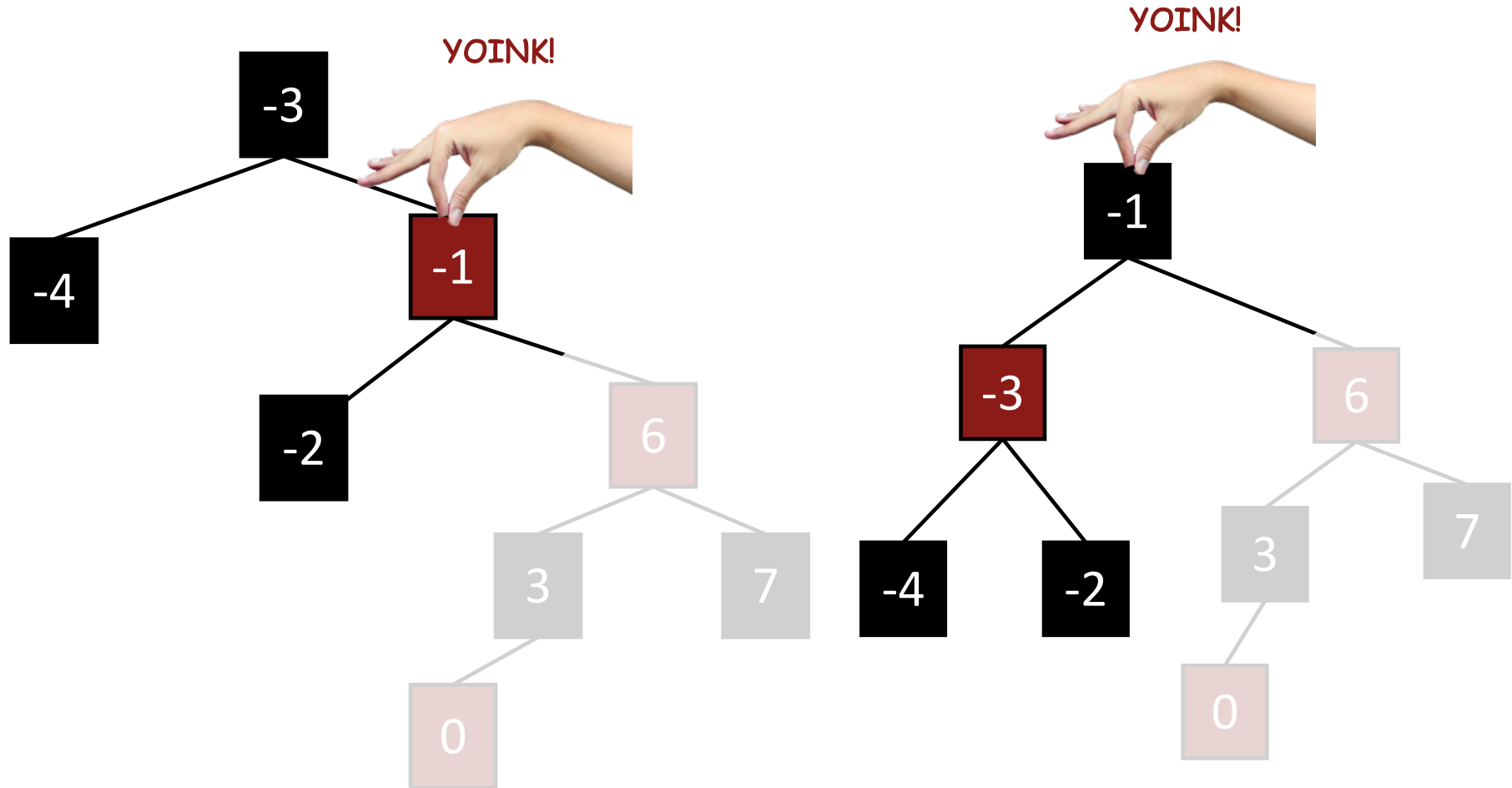


# Example, part 2

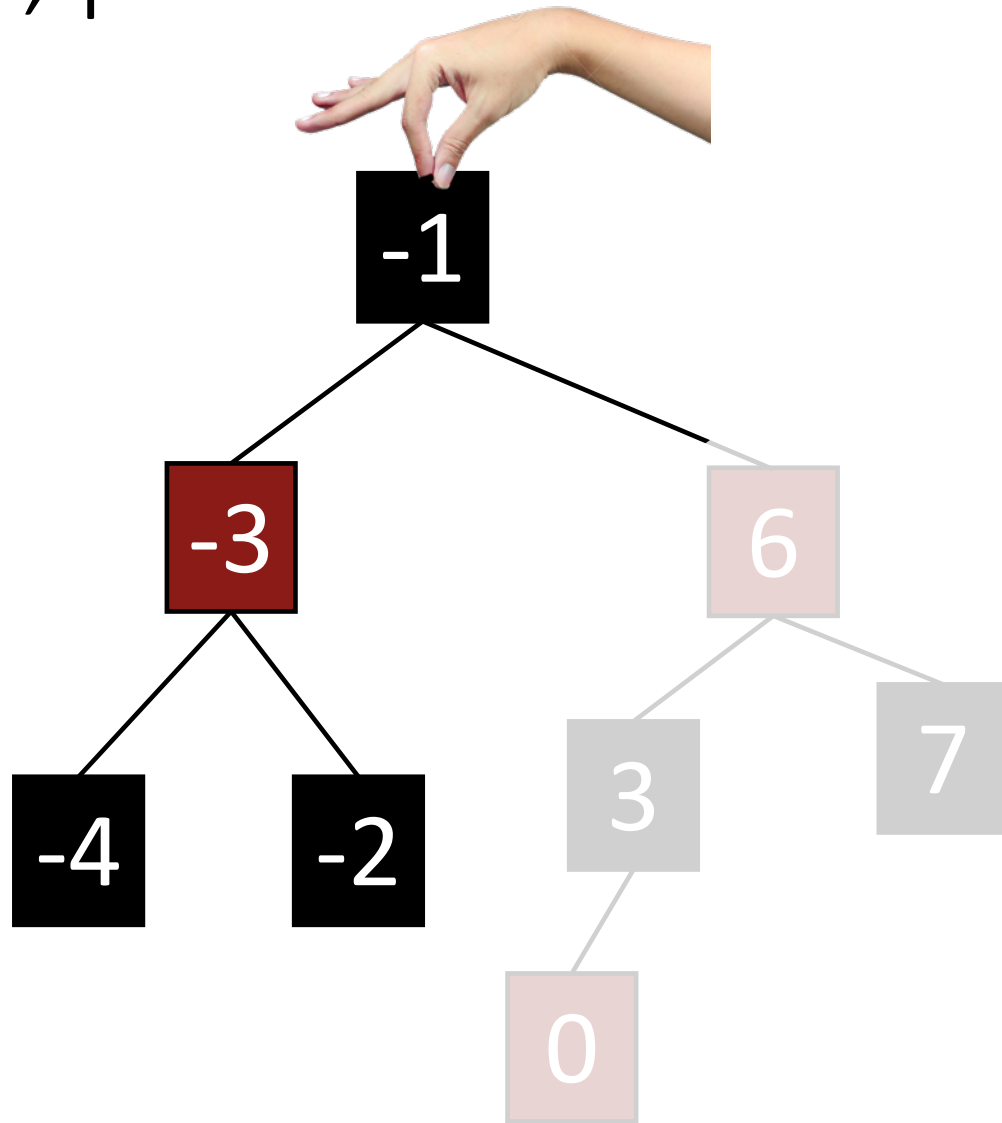




# Example, part 2

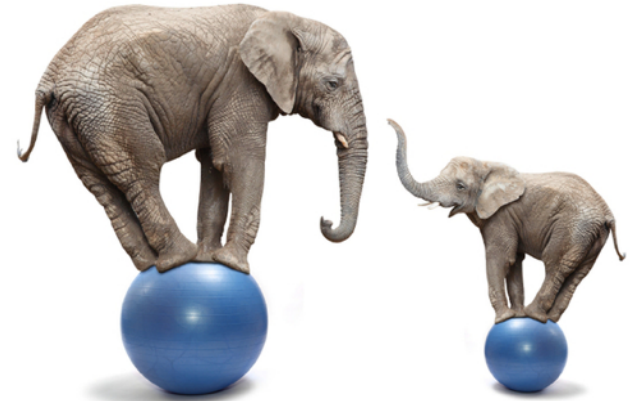
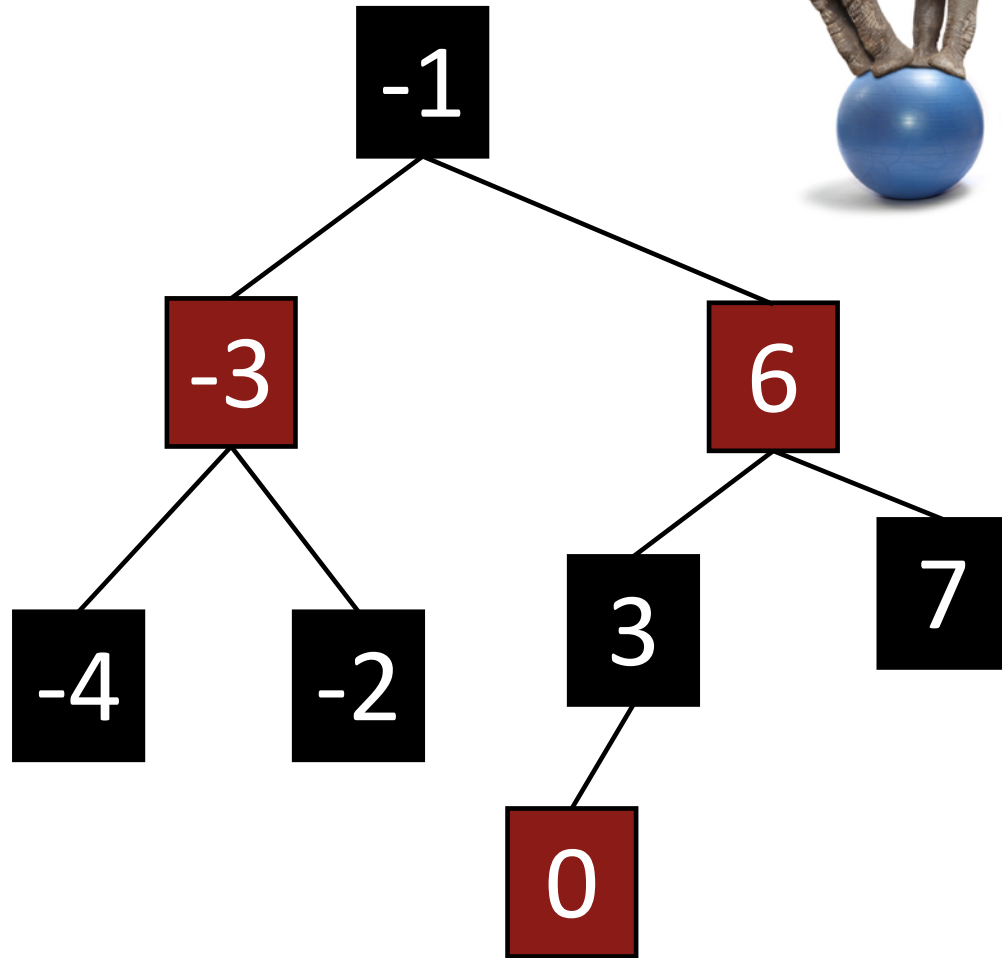


# Example, part 2 **YOINK!**

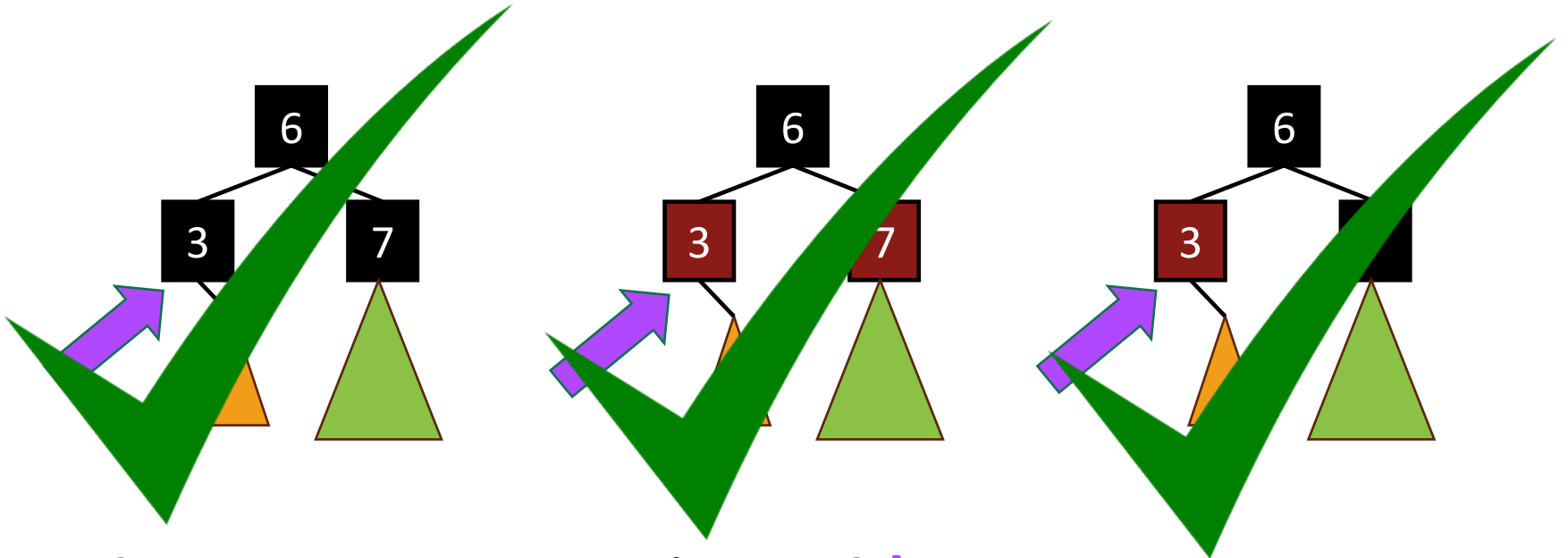


# Example, part 2

***TA-DA!***



# Many cases



- Suppose we want to insert 0 **here**.
- There are 3 “important” cases for different colorings of the existing tree, and there are 9 more cases for all of the various symmetries of these 3 cases.

# Deleting from a Red-Black tree

Fun exercise!



Ollie the over-achieving ostrich

# That's a lot of cases!

- You are **not responsible** for the nitty-gritty details of Red-Black Trees. (For this class)
  - Though implementing them is a great exercise!
- You should know:
  - What are the properties of an RB tree?
  - And (more important) why does that guarantee that they are balanced?

# What have we learned?

- Red-Black Trees always have height at most  $2\log(n+1)$ .
- As with general Binary Search Trees, all operations are  $O(\text{height})$
- So all operations with RBTrees are  $O(\log(n))$ .

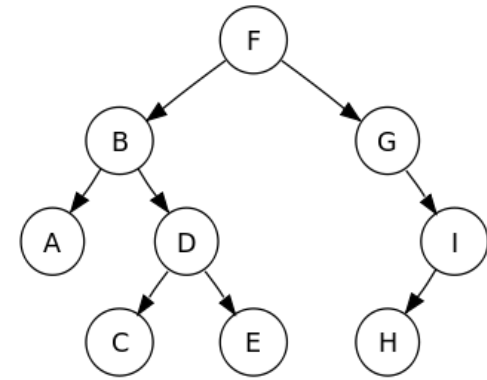
# Conclusion: The best of both worlds

	Sorted Arrays	Linked Lists	Binary Search Trees*
Search	$O(\log(n))$ 😊	$O(n)$ 😞	$O(\log(n))$ 😊
Delete	$O(n)$ 😞	$O(n)$ 😞	$O(\log(n))$ 😊
Insert	$O(n)$ 😞	$O(1)$ 😊	$O(\log(n))$ 😊



# Today

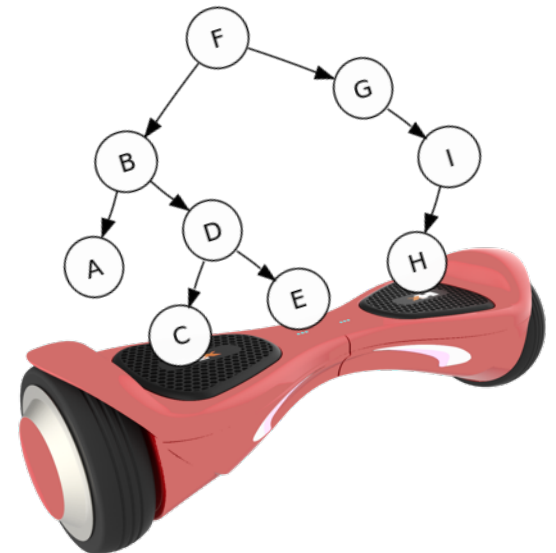
- Begin a brief foray into data structures!
  - See CS 166 for more!
- Binary search trees
  - You may remember these from CS 106B
  - They are better when they're balanced.



this will lead us to...

- Self-Balancing Binary Search Trees
  - **Red-Black** trees.

Recap



# Recap

- Balanced binary trees are the best of both worlds!
- But we need to keep them balanced.
- **Red-Black Trees** do that for us.
  - We get  $O(\log(n))$ -time INSERT/DELETE/SEARCH
  - Clever idea: have a proxy for balance



# Next time

- **Hashing!**

## Before next time

- Pre-lecture exercise for Lecture 8
- More probability yay!