

Binary Search Trees

Reflecting on Dictionaries

Cost

- Complexity of various implementations of dictionaries
 - assuming it contains n entries

	<i>Unsorted array</i>	<i>Array sorted by key</i>	<i>Linked list</i>	<i>Hash Table</i>
lookup	$O(n)$	$O(\log n)$	$O(n)$	$O(1)$ <i>average</i>
insert	$O(1)$ <i>amortized</i>	$O(n)$	$O(1)$	$O(1)$ <i>average and amortized</i>

- Hash dictionaries are clearly the best implementation
 - $O(1)$ lookup and insertion are hard to beat!

Cost

- *Hash dictionaries are clearly the best implementation*

- *$O(1)$ lookup and insertion are hard to beat!*

or are they?

- It's $O(1)$ **average**

- we could be (very) unlucky and incur an $O(n)$ cost

- e.g., if we use a poor hash function

Always read
the fine prints!

- It's $O(1)$ **amortized**

- from time to time, we need to resize the table

- then the operation costs $O(n)$

- Operations like finding the entry with the smallest key cost $O(n)$

- we have to check every entry

Using hash dictionaries is too risky
or not good enough
for applications that require a
guaranteed (short) response time

But they are great for
applications that don't
have such constraints

Goal

- Develop a data structure that has **guaranteed** $O(\log n)$ worst-case complexity for **lookup**, **insert** and **find_min**

➤ **always!**

- $O(1)$ would be great but we can't get that

Returns the entry with the smallest key

	<i>Unsorted array</i>	<i>Array sorted by key</i>	<i>Linked list</i>	<i>Hash Table</i>	
lookup	$O(n)$	$O(\log n)$	$O(n)$	$O(1)$ <i>average</i>	$O(\log n)$
insert	$O(1)$ <i>amortized</i>	$O(n)$	$O(1)$	$O(1)$ <i>average and amortized</i>	$O(\log n)$
find_min	$O(n)$	$O(1)$	$O(n)$	$O(n)$	$O(\log n)$

Exercise

Getting Started

- The only $O(\log n)$ so far is **lookup** in sorted arrays

	<i>Unsorted array</i>	<i>Array sorted by key</i>	<i>Linked list</i>	<i>Hash Table</i>	
lookup	$O(n)$	$O(\log n)$	$O(n)$	$O(1)$ <i>average</i>	$O(\log n)$
insert	$O(1)$ <i>amortized</i>	$O(n)$	$O(1)$	$O(1)$ <i>average and amortized</i>	$O(\log n)$
find_min	$O(n)$	$O(1)$	$O(n)$	$O(n)$	$O(\log n)$

- That's binary search
 - Let's start there

Searching Sorted Data

Searching for a Number

- Consider the following sorted array

0	1	2	3	4	5	6	7	8	9
-2	0	4	7	12	19	22	42	65	

- When searching for a number x using binary search, we **always** start by looking at the midpoint, index 4

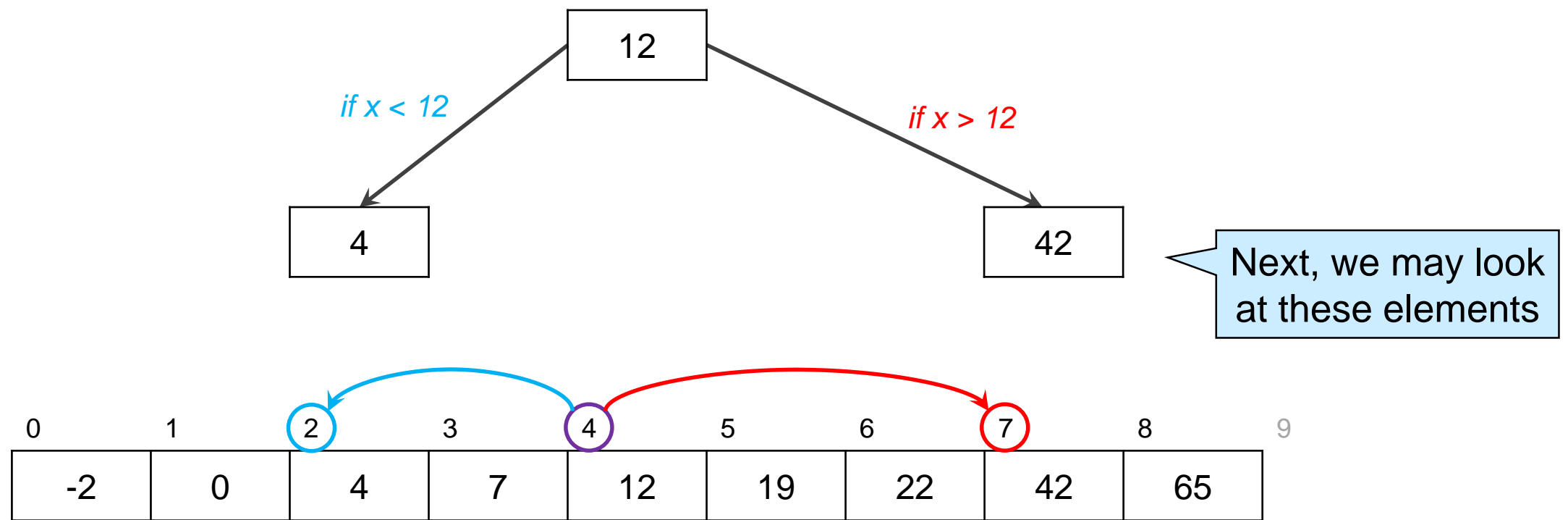
12

We always look at this element

- Then, 3 things can happen
 - $x = 12$ (and we are done)
 - $x < 12$
 - $x > 12$

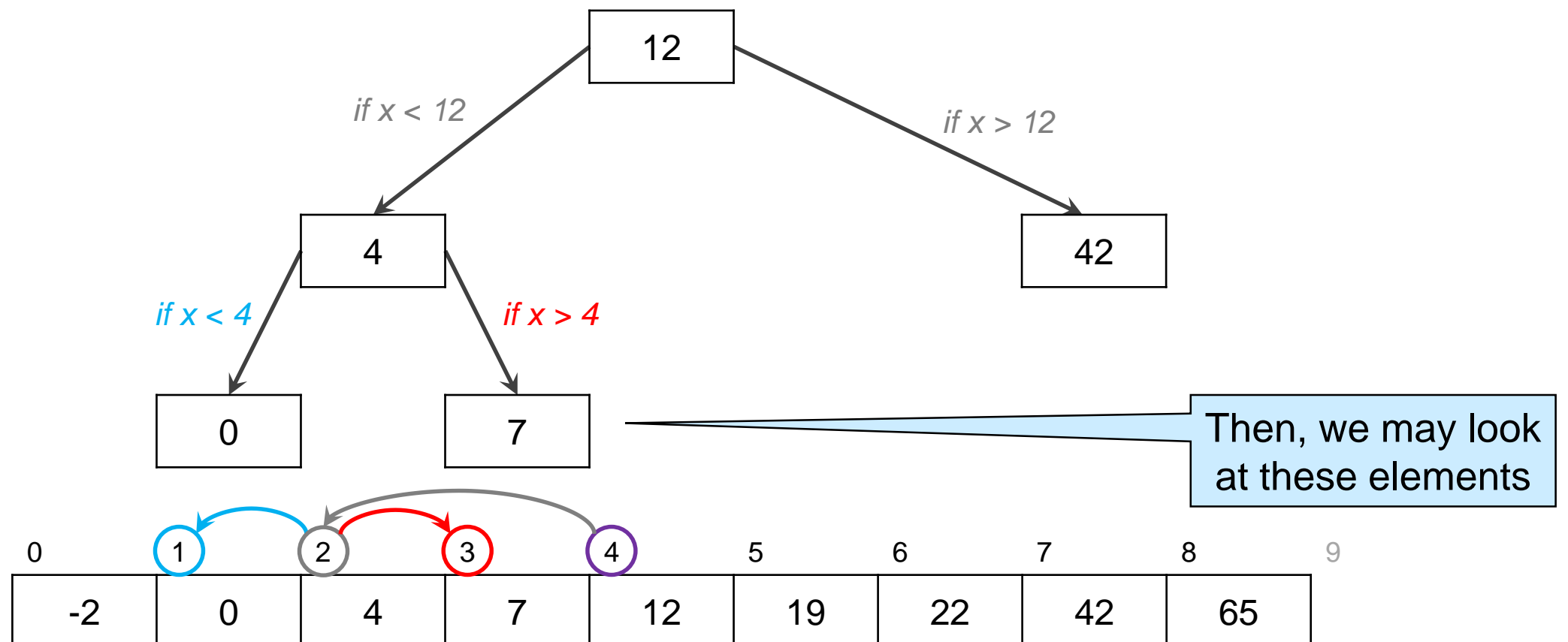
Searching for a Number

- If $x < 12$, the next index we look at is **necessarily 2**
- If $x > 12$, the next index we look at is **necessarily 7**



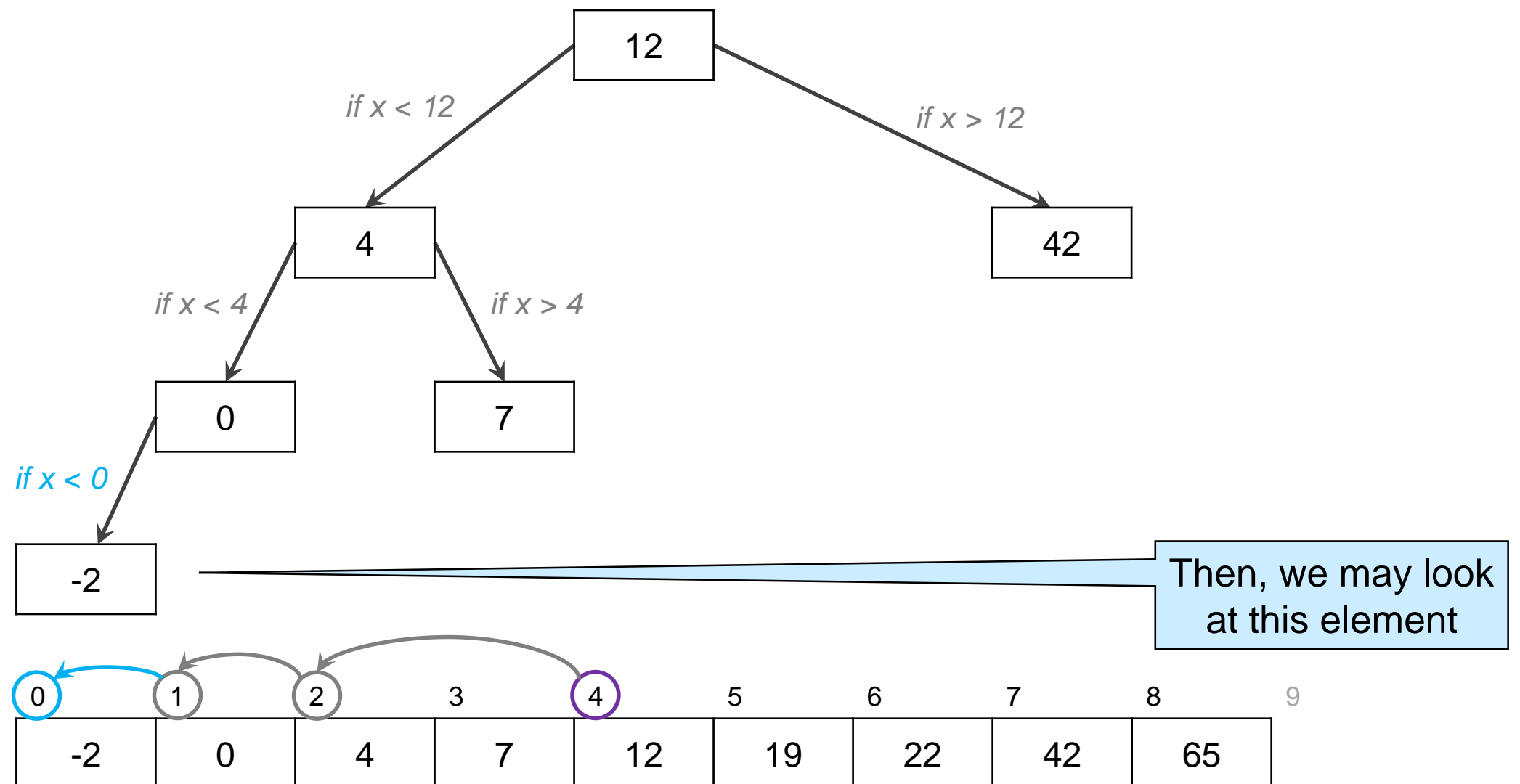
Searching for a Number

- Assume $x < 12$, so we look at 4
 - if $x = 4$, we are done
 - if $x < 4$, we **necessarily** look at 0
 - if $x > 4$, we **necessarily** look at 7



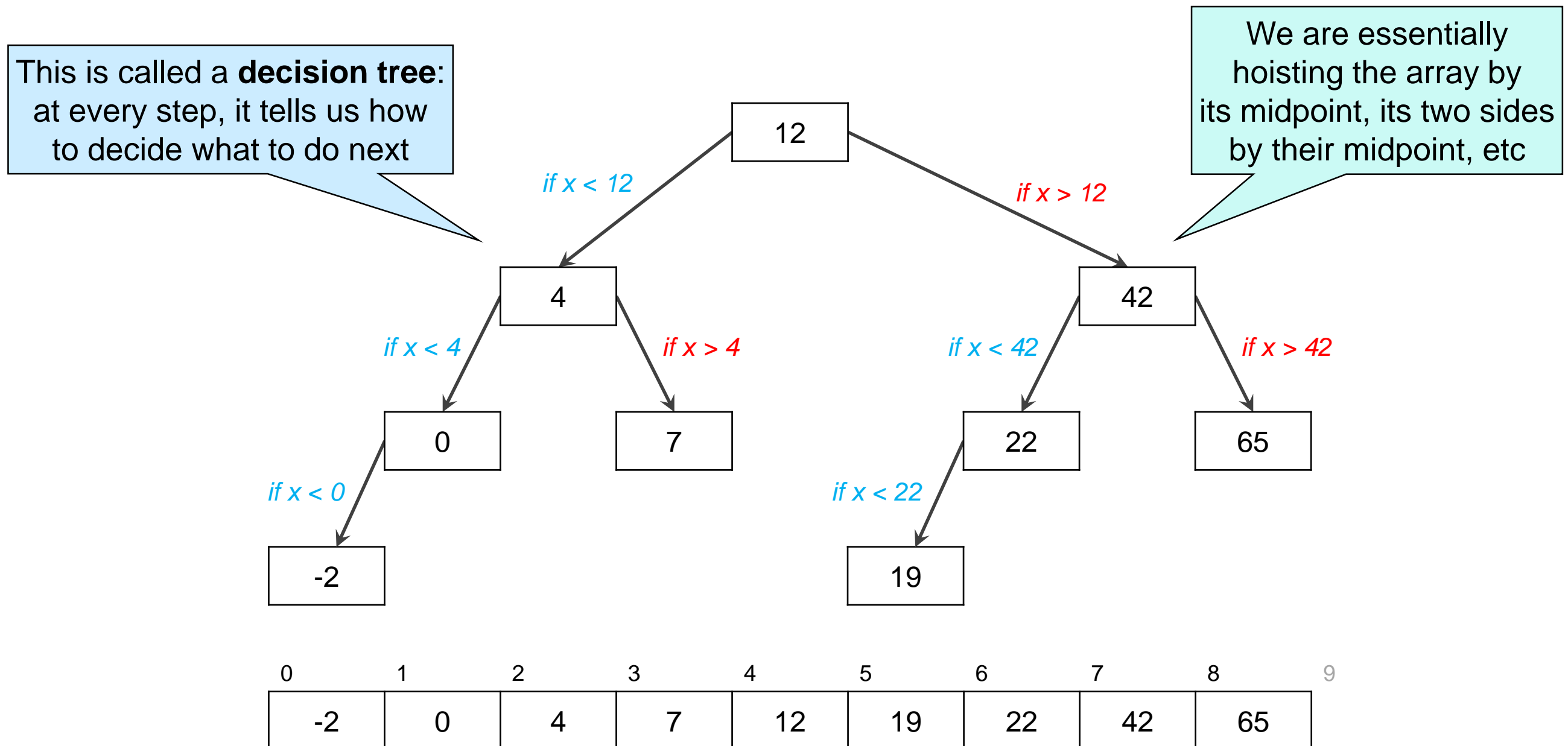
Searching for a Number

- Assume $x < 4$, so we look at 0
 - if $x = 0$, we are done
 - if $x < 0$, we **necessarily** look at -2



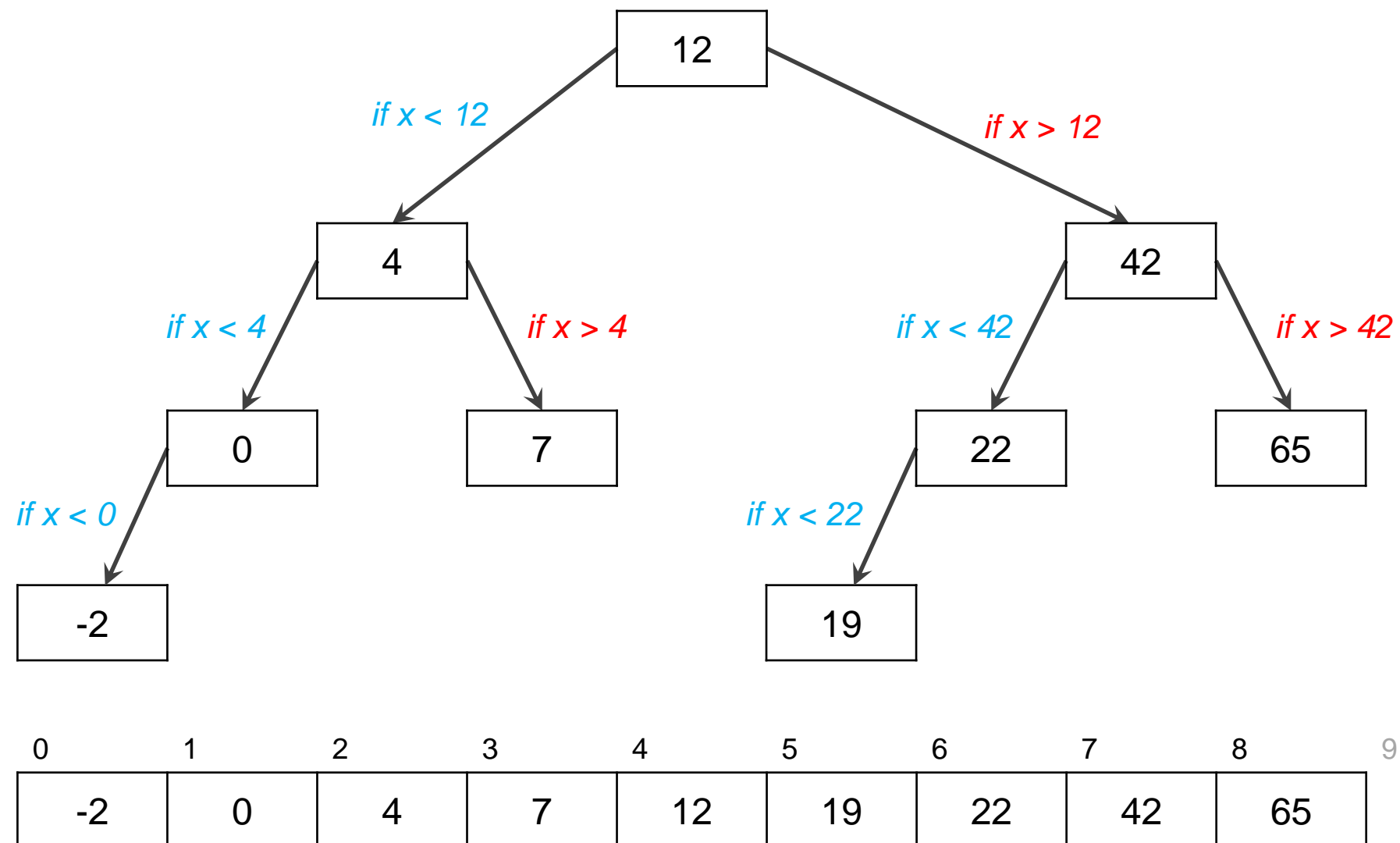
Searching for a Number

- We can map out all possible sequences of elements binary search may examine, for any x



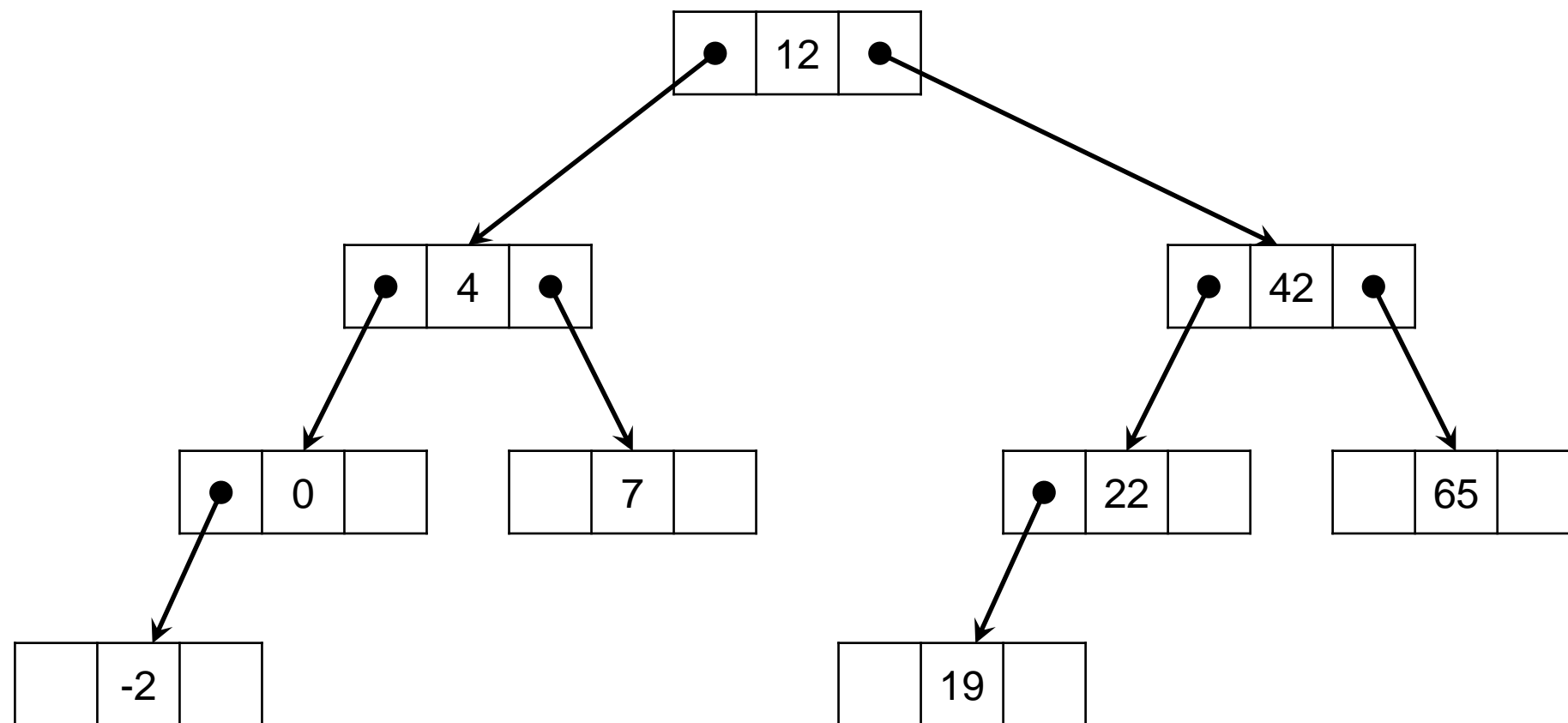
Searching for a Number

- An array provides direct access to all elements
 - This is overkill for binary search
 - At any point, it needs direct access to **at most two** elements



Searching for a Number

- We can achieve the same access pattern by pairing up each element with **two pointers**
 - one to each of the two elements that may be examined next



- We are losing direct access to arbitrary elements,
 - but it retains access to the elements that matter to binary search

Arrays gave us
more power
than needed

Towards an Implementation

- We can capture this idea with this **type declaration**:

```
typedef struct tree_node tree;  
struct tree_node {  
    tree* left;  
    int data;  
    tree* right;  
};
```

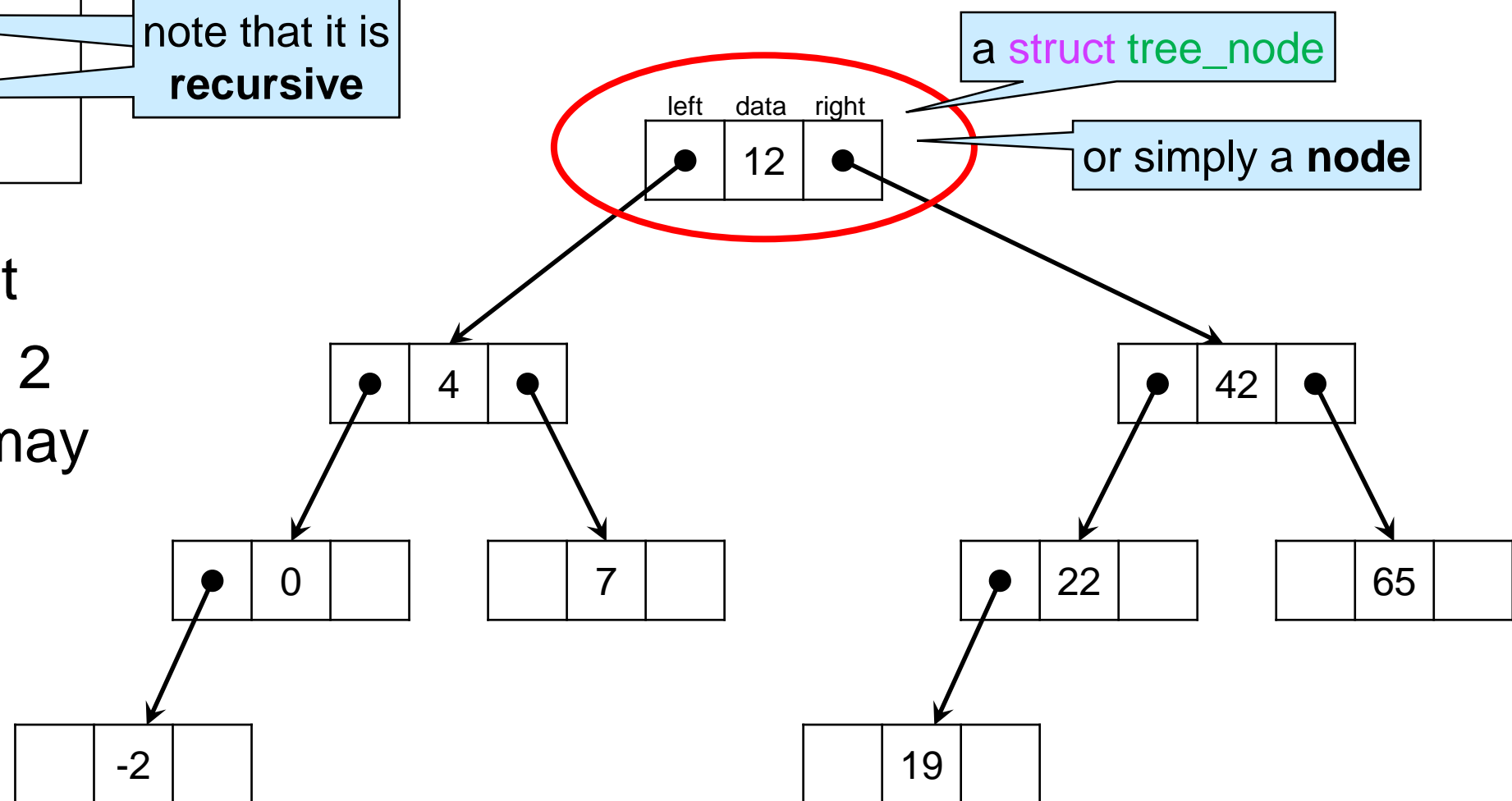
note that it is
recursive

a **struct tree_node**

or simply a **node**

- a data element
- pointers to the 2 elements we may look at next

- This is called a **node**



- This arrangement of data in memory is called a **tree**

Constructing this Tree

```
typedef struct tree_node tree;  
struct tree_node {  
    tree* left;  
    int data;  
    tree* right;  
};
```

- Let's build the first few nodes of this example

```
tree* T = alloc(tree);
```

```
T->data = 12;
```

```
T->left = alloc(tree);
```

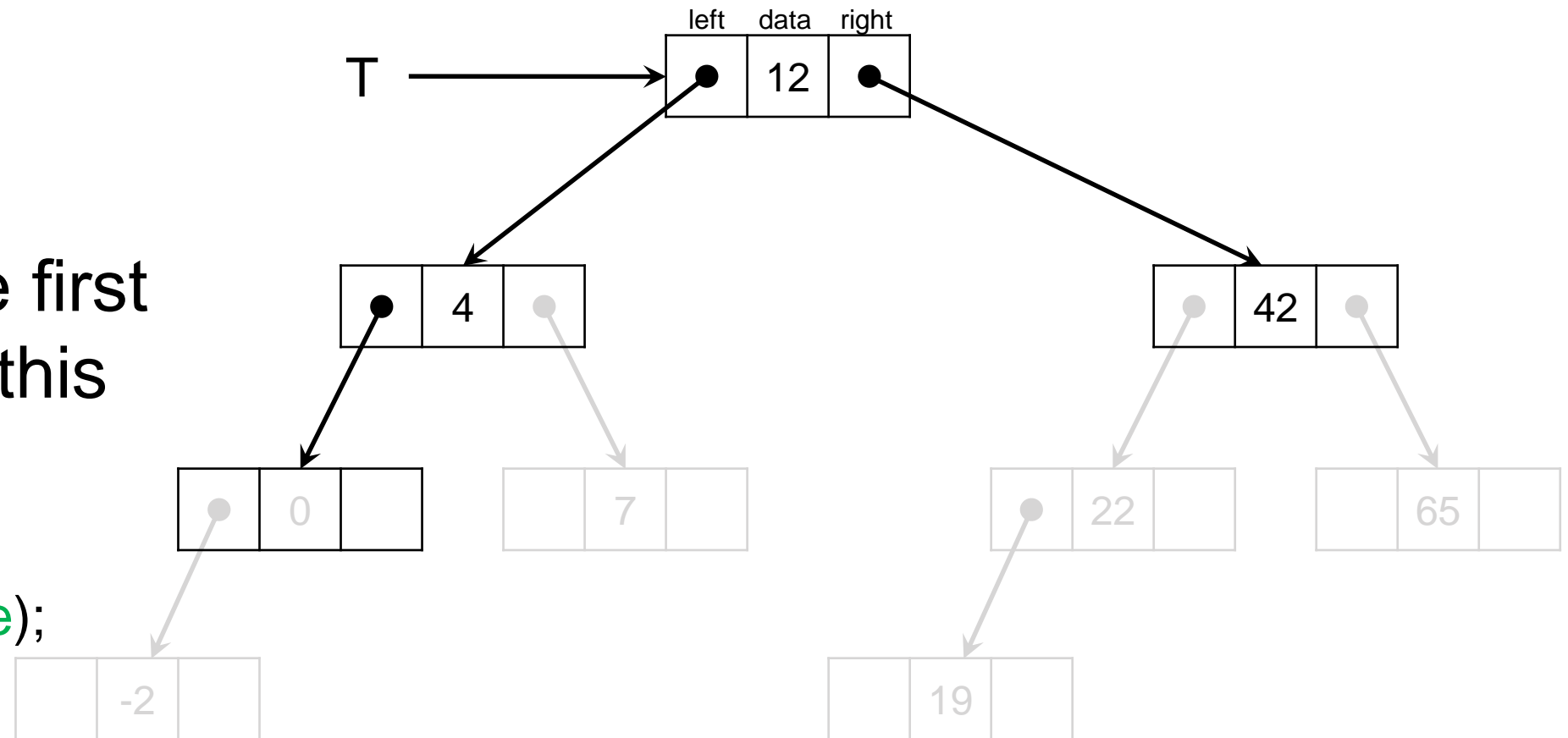
```
T->left->data = 4;
```

```
T->right = alloc(tree);
```

```
T->right->data = 42;
```

```
T->left->left = alloc(tree);
```

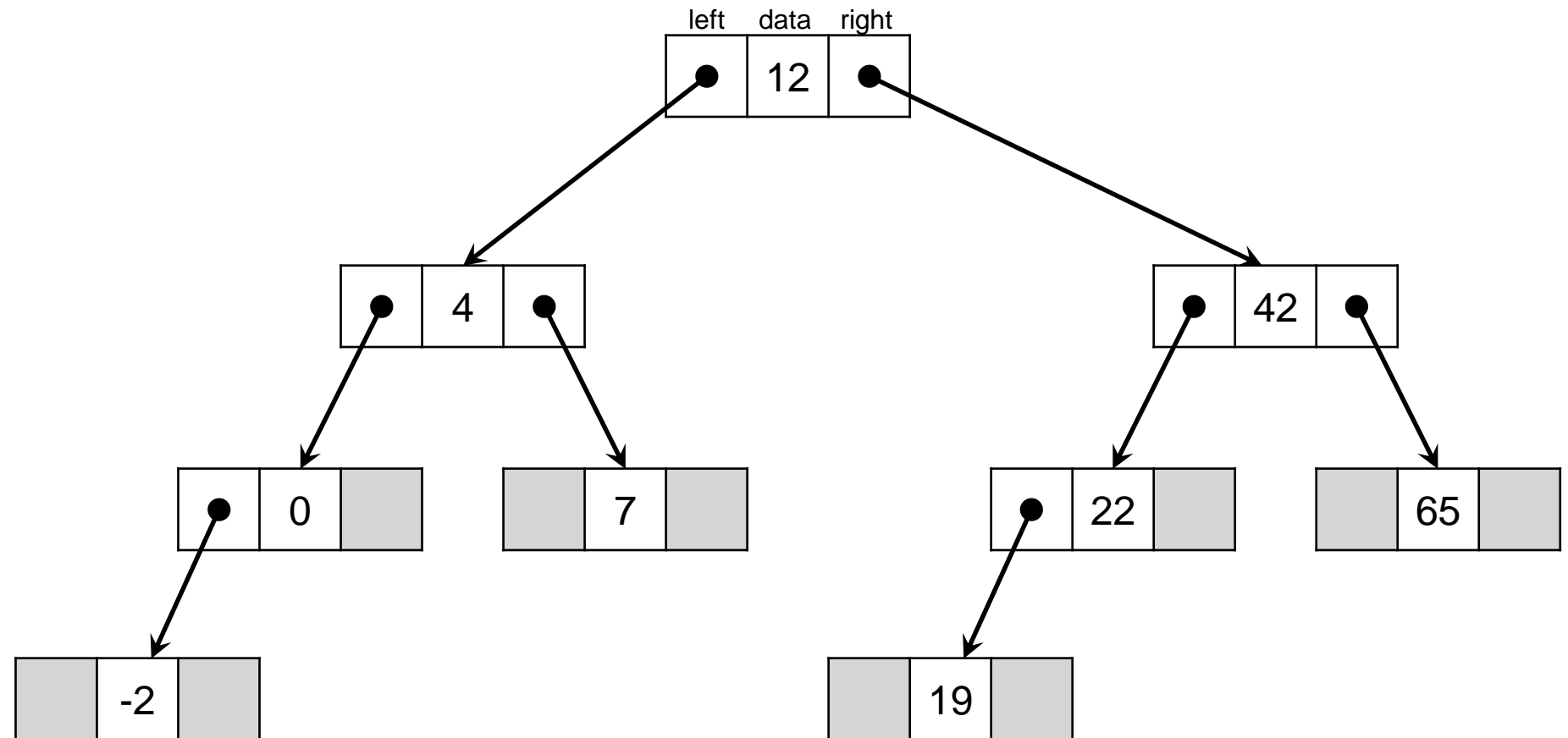
...



The End of the Line

```
typedef struct tree_node tree;  
struct tree_node {  
    tree* left;  
    int data;  
    tree* right;  
};
```

- What should the blank left/right fields point to?



- NULL



➤ each sequence of left/right pointers works like a NULL-terminated list

- a dummy node



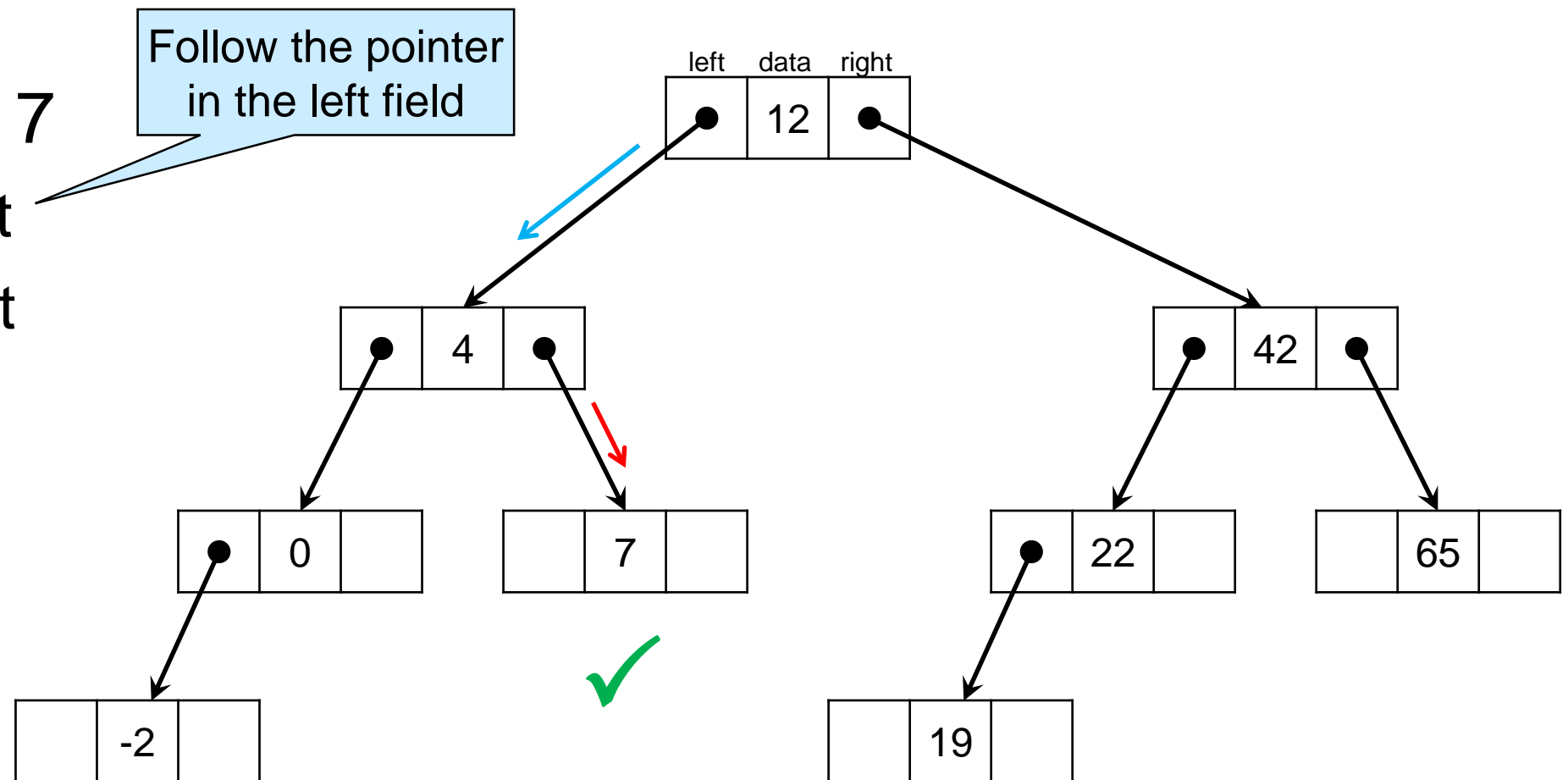
➤ not very useful here

We used dummy nodes to get direct access to the end of a list

Searching

- Searching for 7

- $7 < 12$: go left
- $7 > 4$: go right
- $7 = 7$: **found**



- We are doing the **same steps** as binary search
- Starting from an n -element array, the cost is $O(\log n)$

If the tree is obtained as in this example

Searching

- Searching for 5

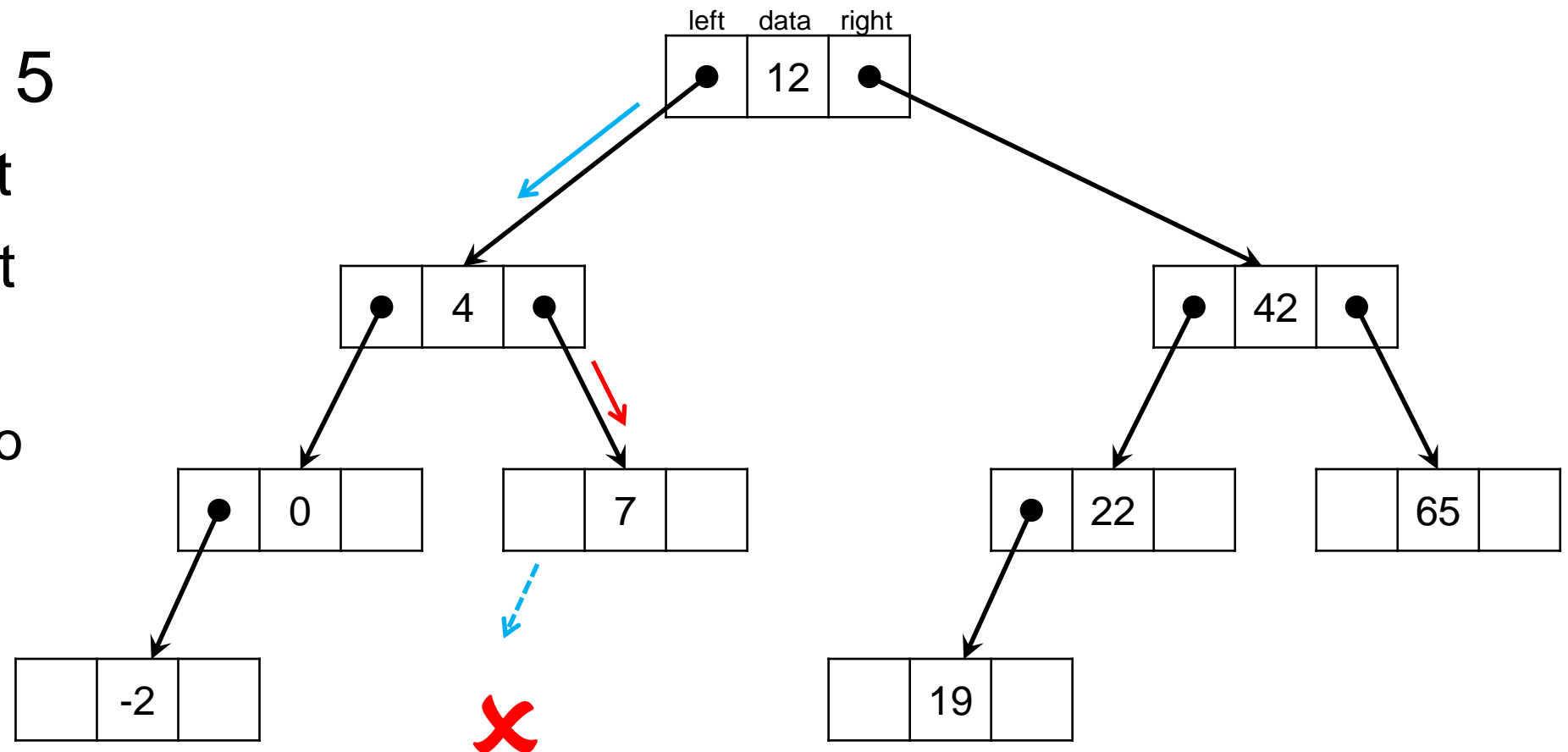
- $5 < 12$: go left

- $5 > 4$: go right

- $5 < 7$: go left

- nowhere to go

- **not there**



- We are doing the **same steps** as binary search

- Starting from an n-element array, the cost is $O(\log n)$

If the tree is obtained
as in this example

Recall our Goal

- Develop a data structure that has **guaranteed** $O(\log n)$ worst-case complexity for **lookup**, **insert** and **find_min**
 - **always!**

- **lookup** has cost $O(\log n)$

in *this* setup

	<i>Target data structure</i>
lookup	$O(\log n)$
insert	$O(\log n)$
find_min	$O(\log n)$

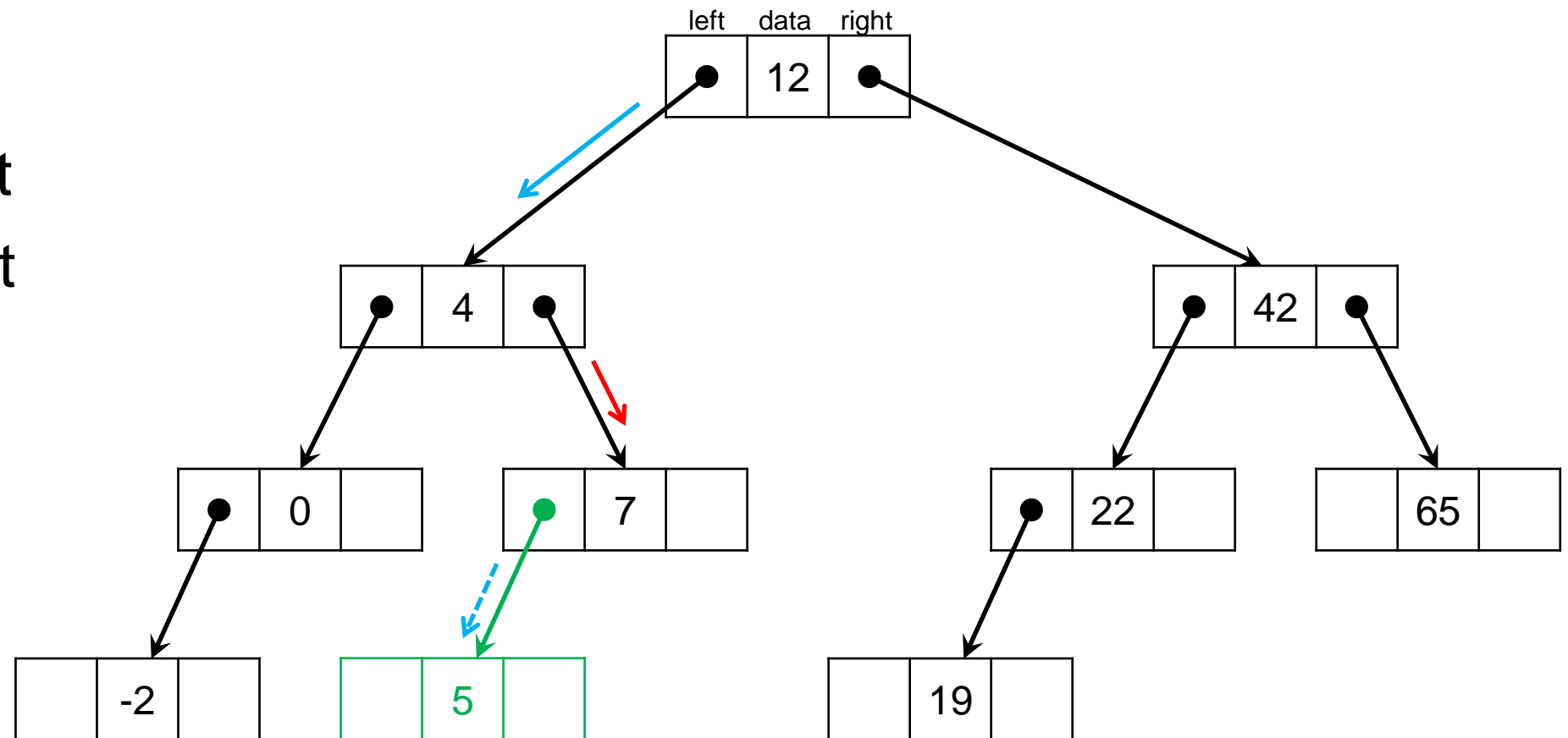


- What about **insert** and **find_min**?

Insertion

- Inserting 5

- $5 < 12$: go left
- $5 > 4$: go right
- $5 < 7$: go left
- put it there



- We are doing the **same steps** we would do to search for it, and then put it where it should have been
 - so that we find it when searching for it next time
- For an n-element array, this costs $O(\log n)$

We couldn't get this with sorted arrays

If the tree is obtained as in this example

Finding the Smallest Key

- Keep going left

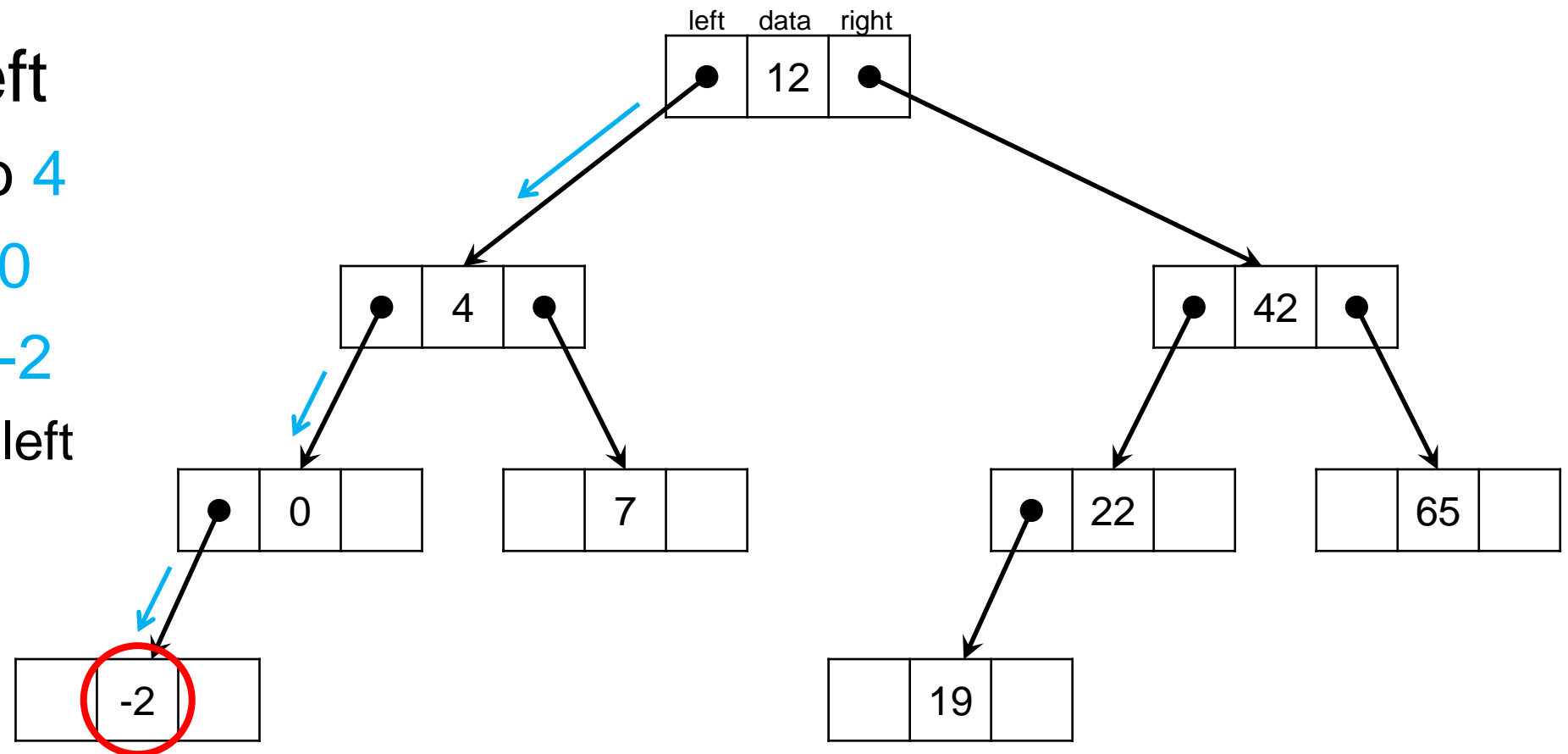
- left from 12 to 4

- left from 4 to 0

- left from 0 to -2

- nothing to its left

- **the smallest key is -2**



- Starting from an n-element array, we can go left at most $O(\log n)$ times

- The cost is $O(\log n)$

If the tree is obtained as in this example

Recall our Goal

- Develop a data structure that has **guaranteed** $O(\log n)$ worst-case complexity for *lookup*, *insert* and *find_min*
 - *always!*

- *lookup*, *insert* and *find_min* all have cost $O(\log n)$

in *this* setup

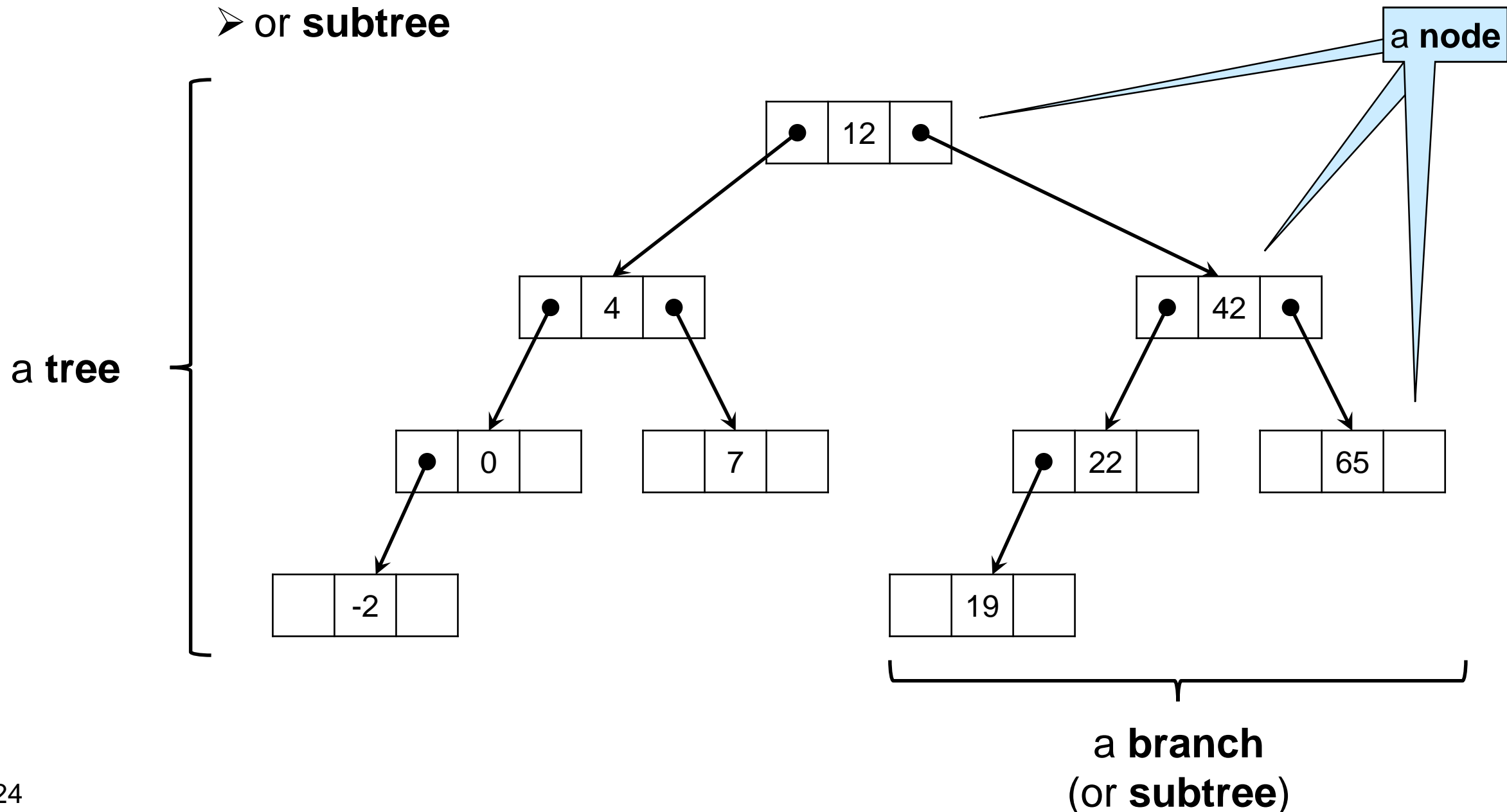
	Target data structure
<i>lookup</i>	$O(\log n)$
<i>insert</i>	$O(\log n)$
<i>find_min</i>	$O(\log n)$



Trees

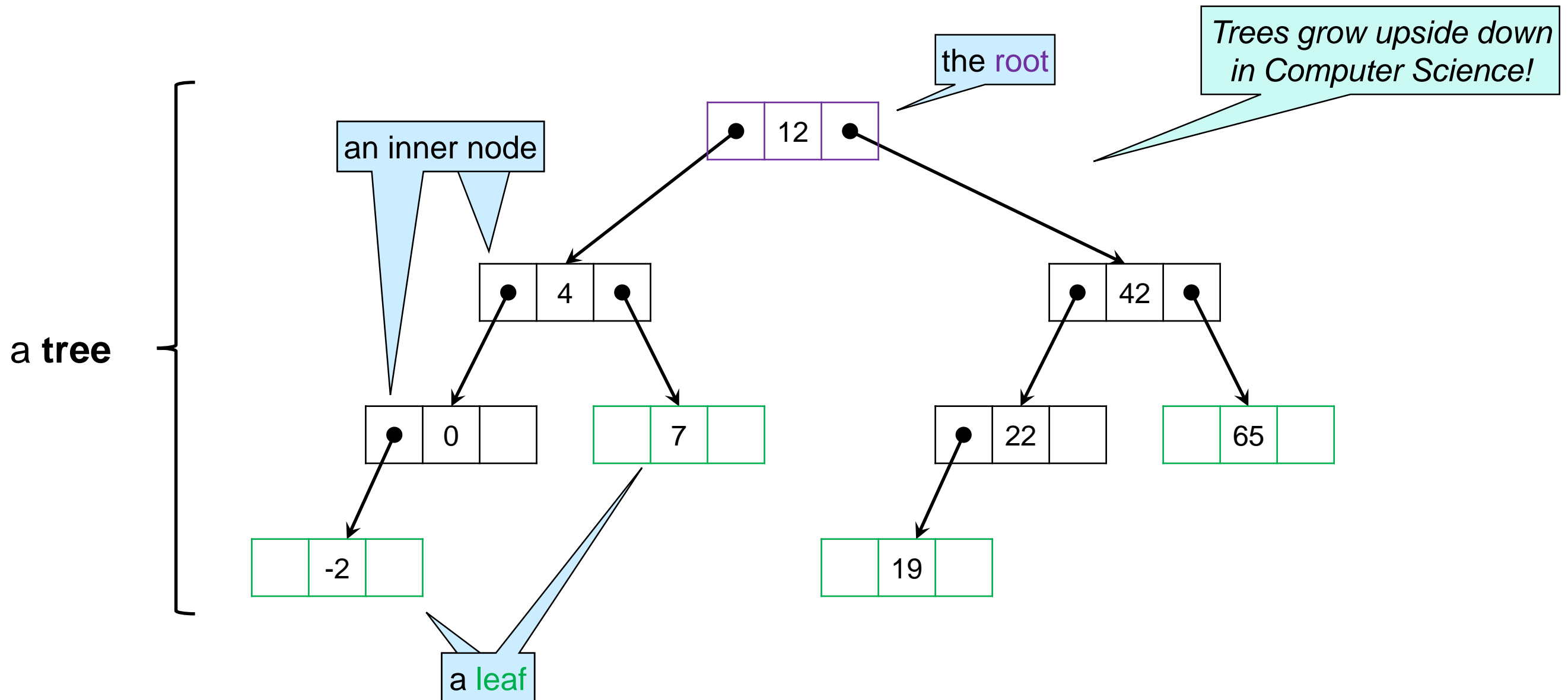
Terminology

- This arrangement of data is called a (binary) **tree**
 - each item in it is called a **node**
 - the part of a tree hanging from a node is called a **branch**
 - or **subtree**



Terminology

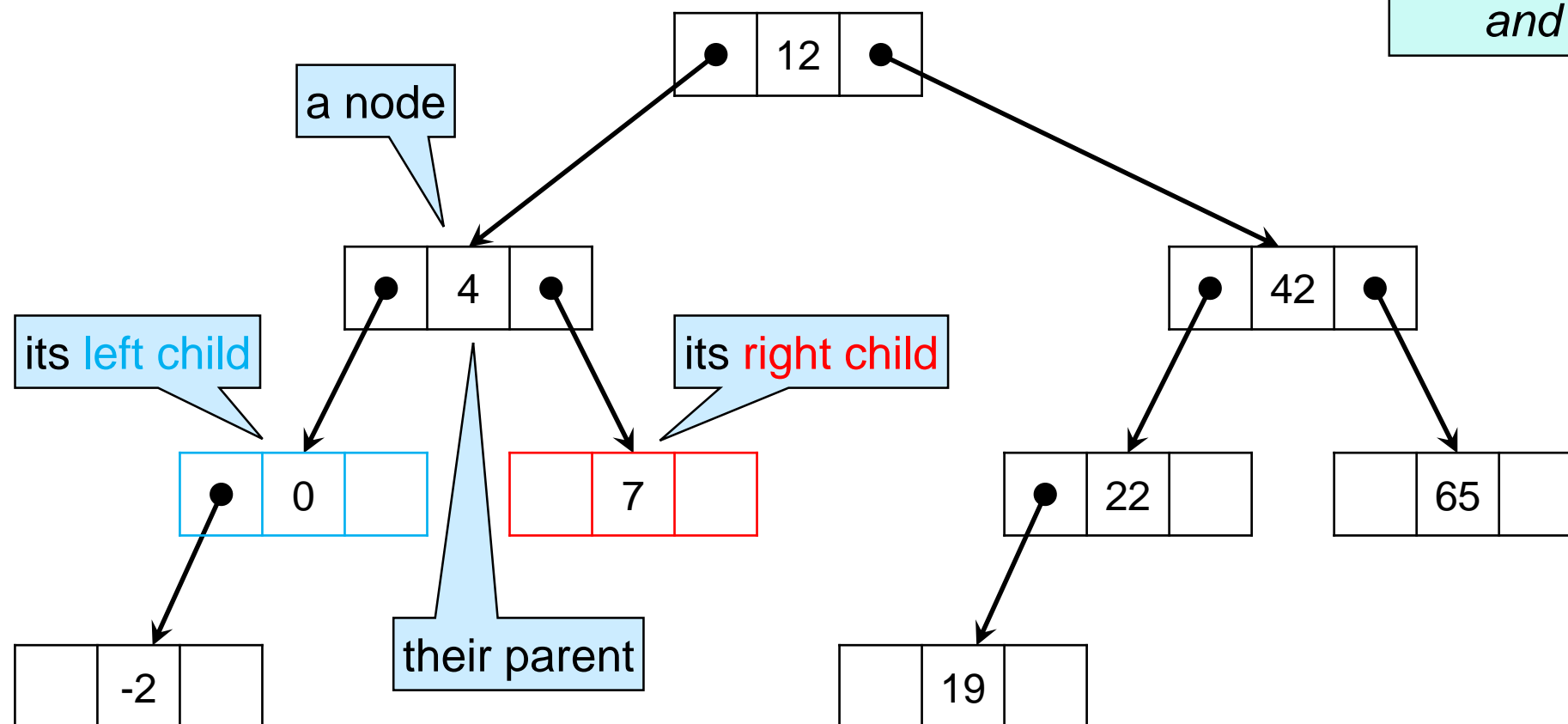
- The node at the top is called the **root** of the tree
 - the nodes at the bottom are the **leaves** of the tree
 - the other nodes are called **inner nodes**



Terminology

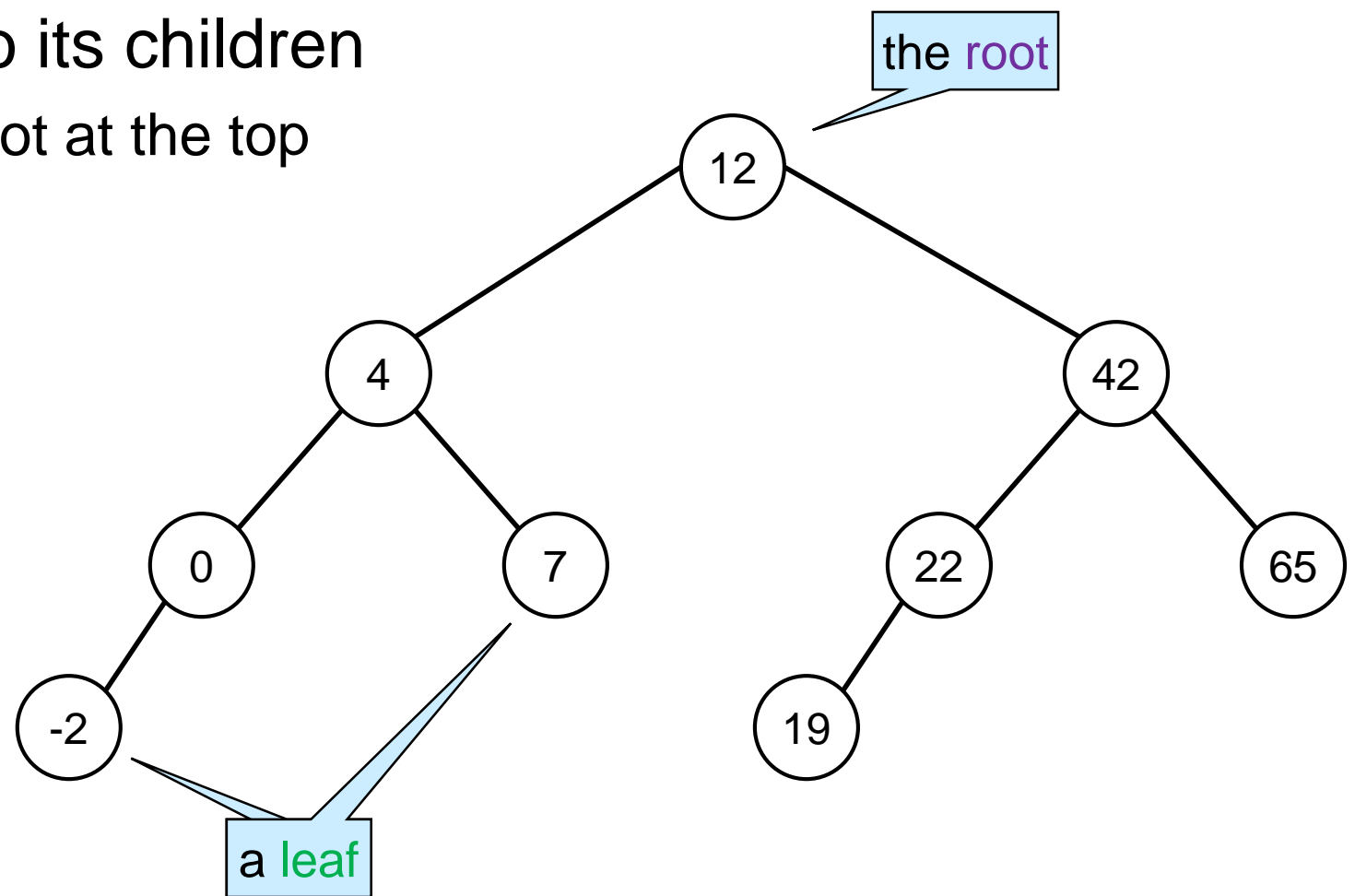
- Given any node
 - the node to its left is its **left child**
 - the node to its right is its **right child**
 - the node above it is its **parent**

.. and Computer Science mixes botanical trees and family trees!



Concrete Tree Diagrams

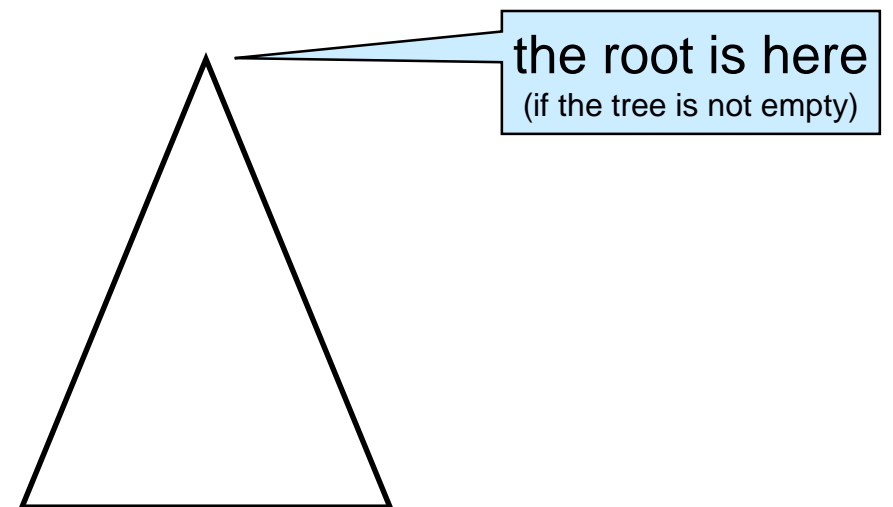
- When drawing trees, we generally omit the details of the memory diagrams
 - draw just the data in a node
 - not the pointer fields
 - and the connection to its children
 - we always draw the root at the top



Pictorial Abstraction

- We will often reason about trees that are arbitrary
 - their actual content is unimportant, so we abstract it away

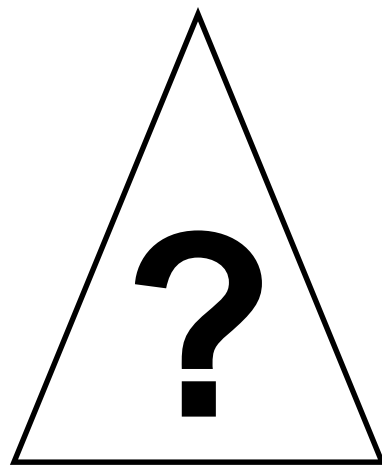
- We draw a generic tree as a triangle



- We represent the empty tree by simply writing "Empty"

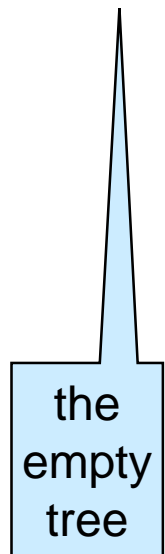
Empty

What do Trees Look Like?

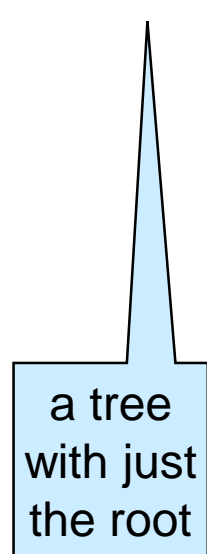


- Abstract trees come in many shapes

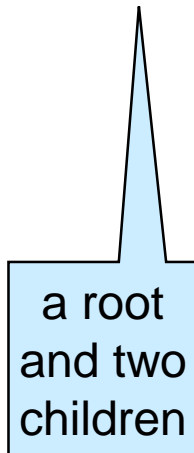
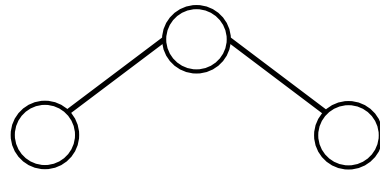
EMPTY



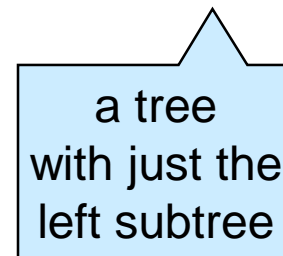
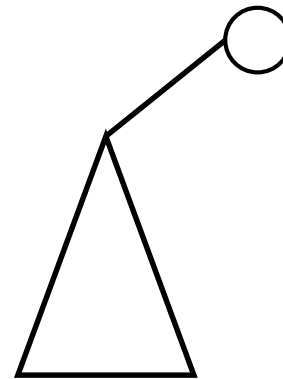
the
empty
tree



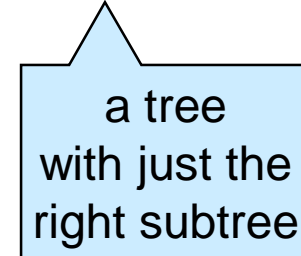
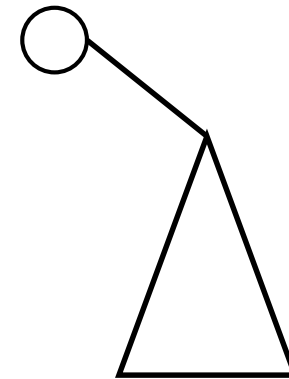
a tree
with just
the root



a root
and two
children

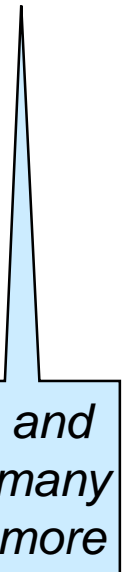


a tree
with just the
left subtree



a tree
with just the
right subtree

...

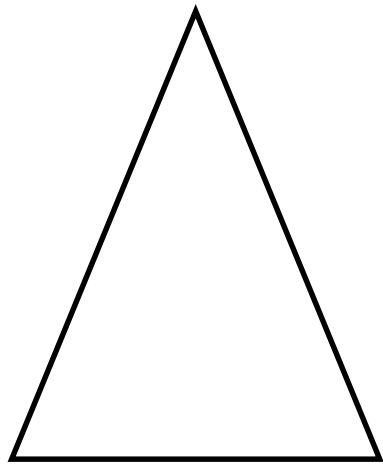


and
many
more

- When working with trees, we need to account for all these possibilities
 - we will forget some
- *Is there a simpler description?*

What Trees Look Like

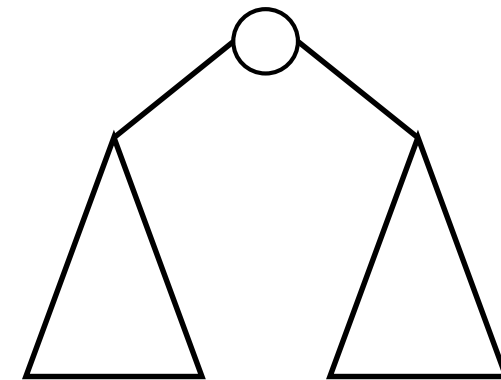
- A tree can be



- either empty

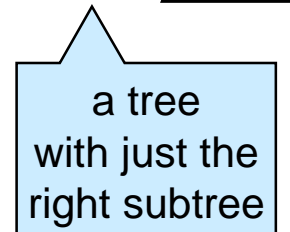
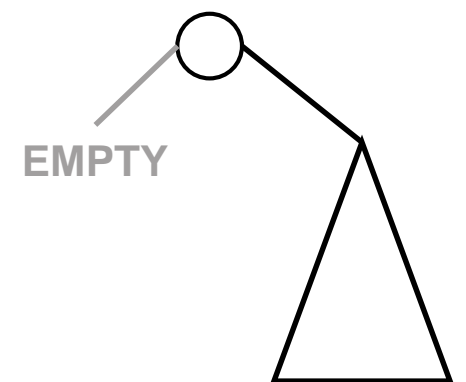
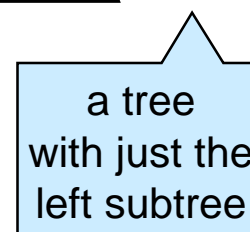
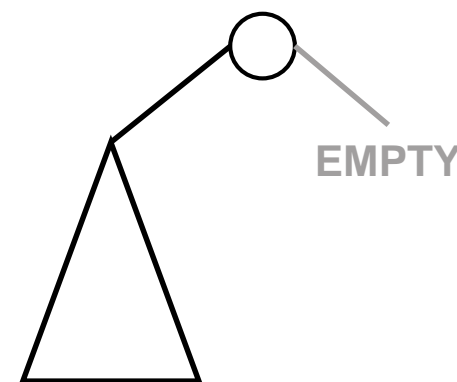
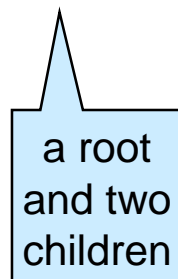
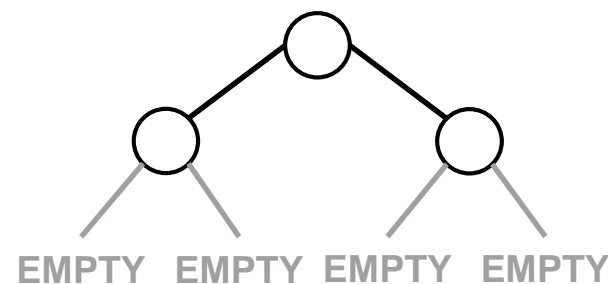
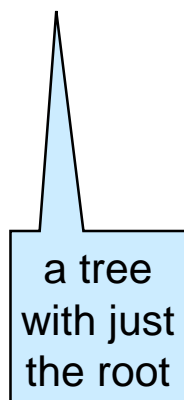
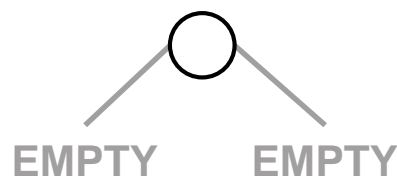
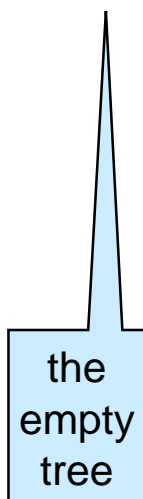
- or a root with a tree on its left and a tree on its right

EMPTY



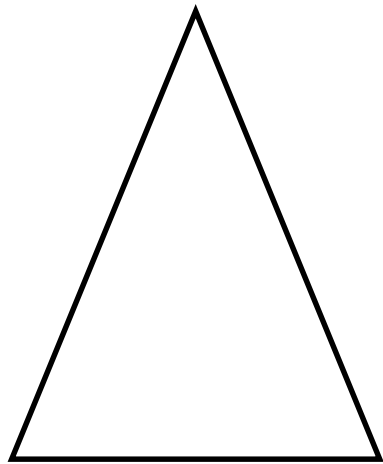
- **Every tree** reduces to these two cases

EMPTY



What Trees Look Like

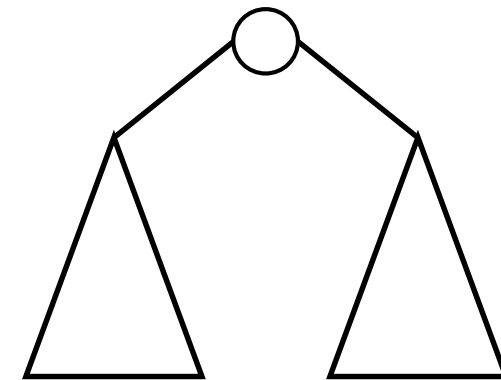
- *A tree can be*



- *either empty*

- *or a root with
a tree on its left and
a tree on its right*

EMPTY



- We only need to consider these two cases when

- writing code about trees

- reasoning about trees

A Minimal Tree Invariant

- *We only need to consider these two cases when writing code about trees*

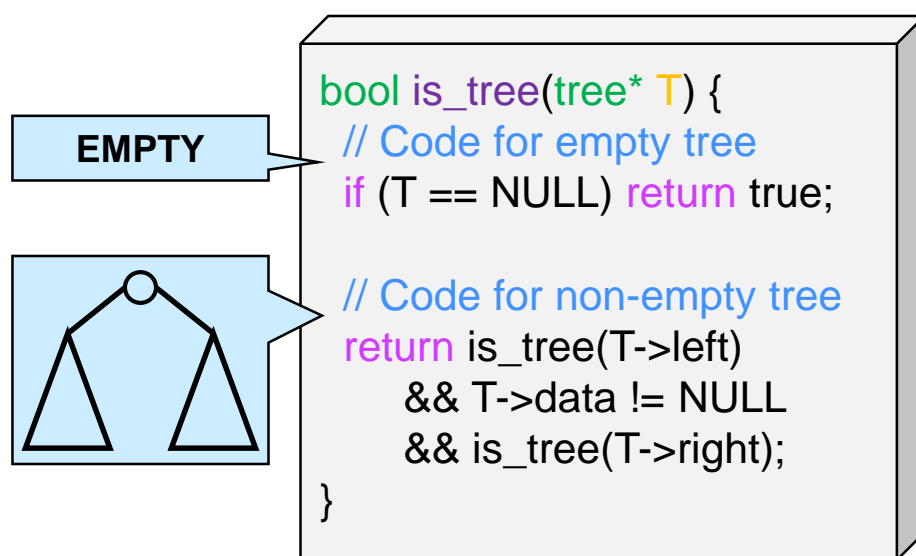
- Let's apply this to write a basic invariant about trees of *entries*

- Just check that the data field is never NULL

```
typedef struct tree_node tree;  
struct tree_node {  
    tree* left;  
    entry data; // != NULL  
    tree* right;  
};
```

Recall we are using trees to implement dictionaries:

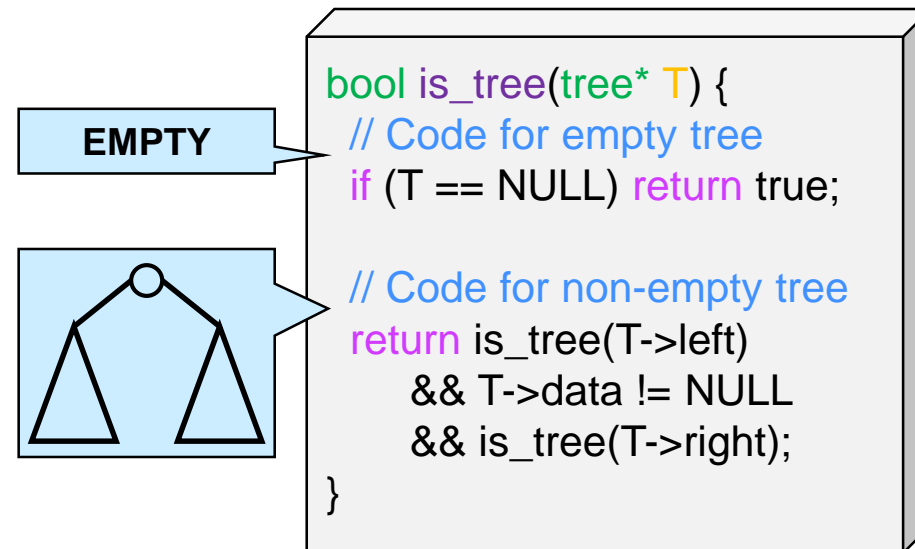
- we store entries in nodes
- valid entries are non-NULL



- This is a **recursive** function
 - the **base case** is about the empty tree
 - the **recursive case** is about every tree that is not empty
 - with a root
 - and two subtrees

A Minimal Tree Invariant

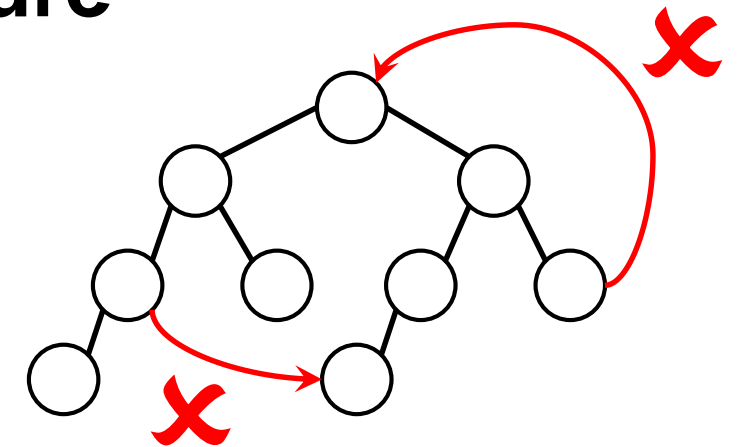
- *We just check that the data field is never NULL*



- But trees have constraints on their **structure**

- a node does not point to an ancestor
- a node has at most one parent

How to check them
is left as an exercise

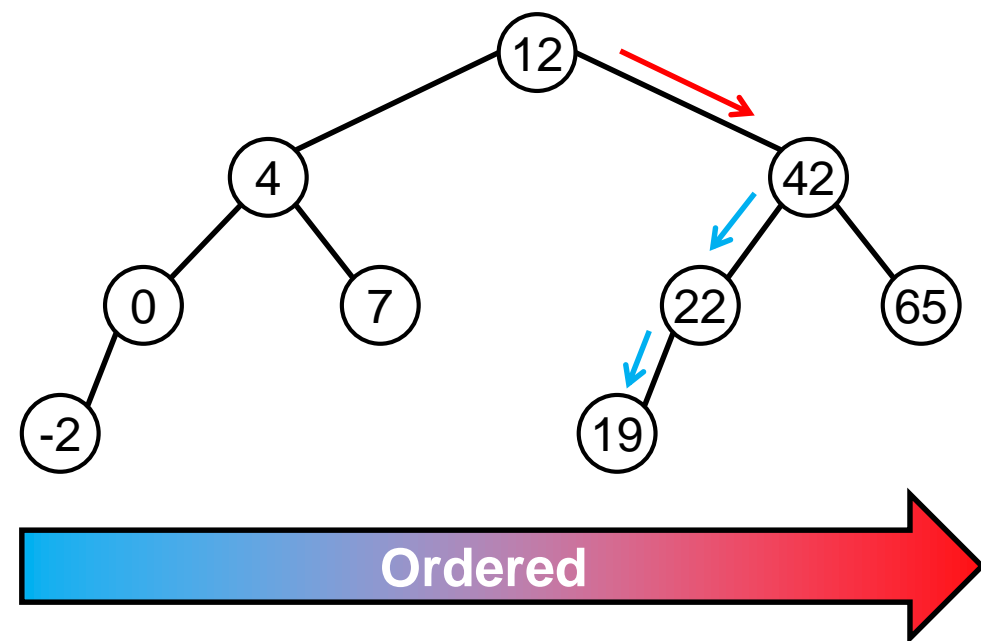


- *What additional constraints on contents do we need to use trees to implement **dictionaries**?*

Binary Search Trees

Binary Search Trees

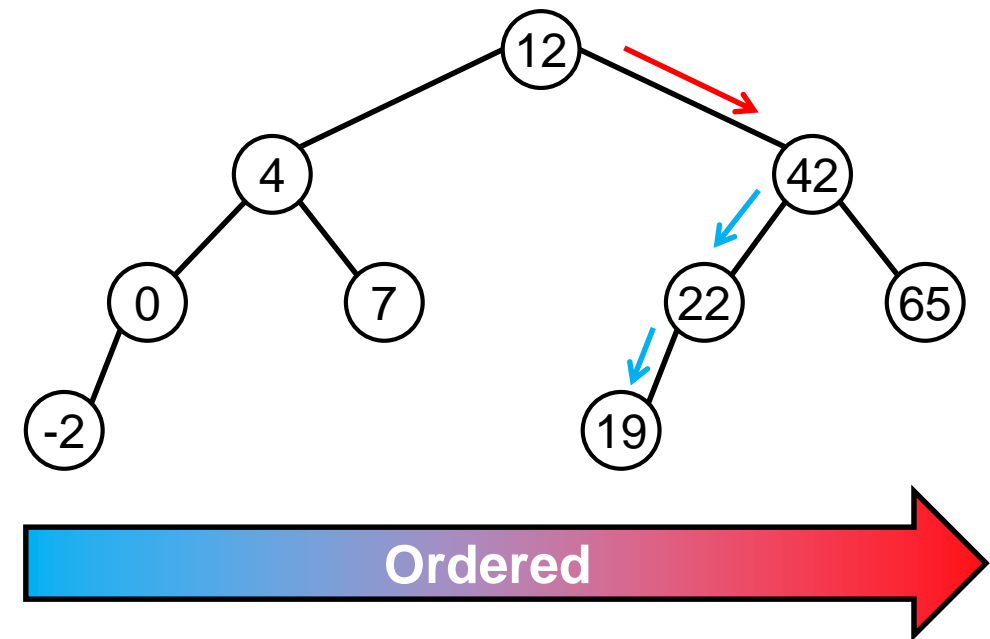
- *What additional constraints on the contents do we need to use trees to implement **dictionaries**?*
- Because lookup emulates binary search, the data in the tree need to be **ordered**
 - smaller values on the left
 - bigger values on the right



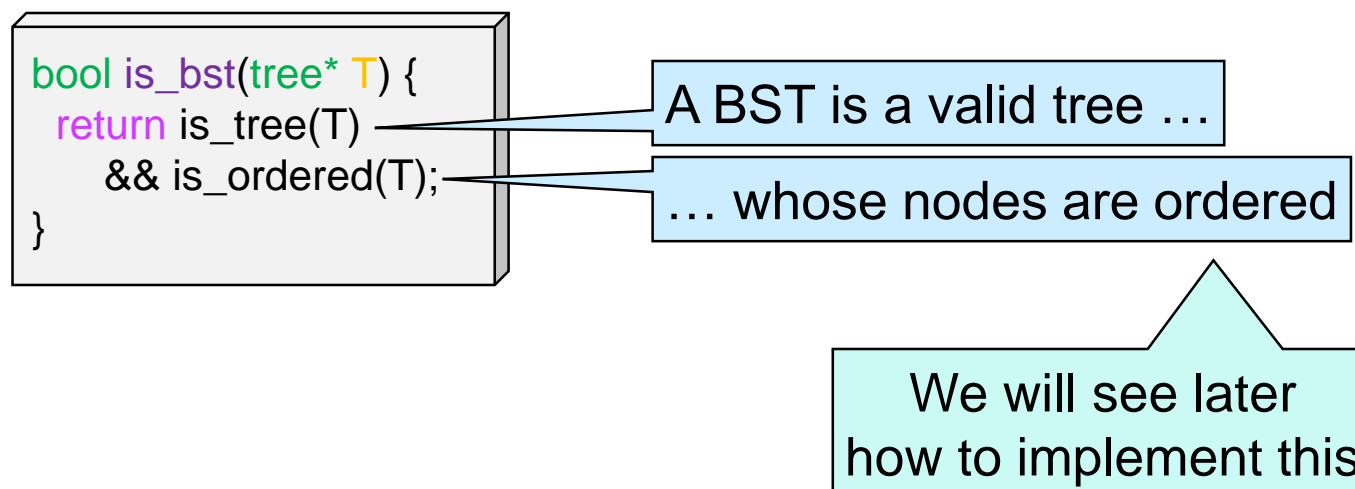
- A tree whose nodes are ordered is called a **binary search tree**

The BST Invariant

- *A tree whose nodes are ordered is called a **binary search tree***



- We can write a specification function that check BSTs



Looking Up Keys

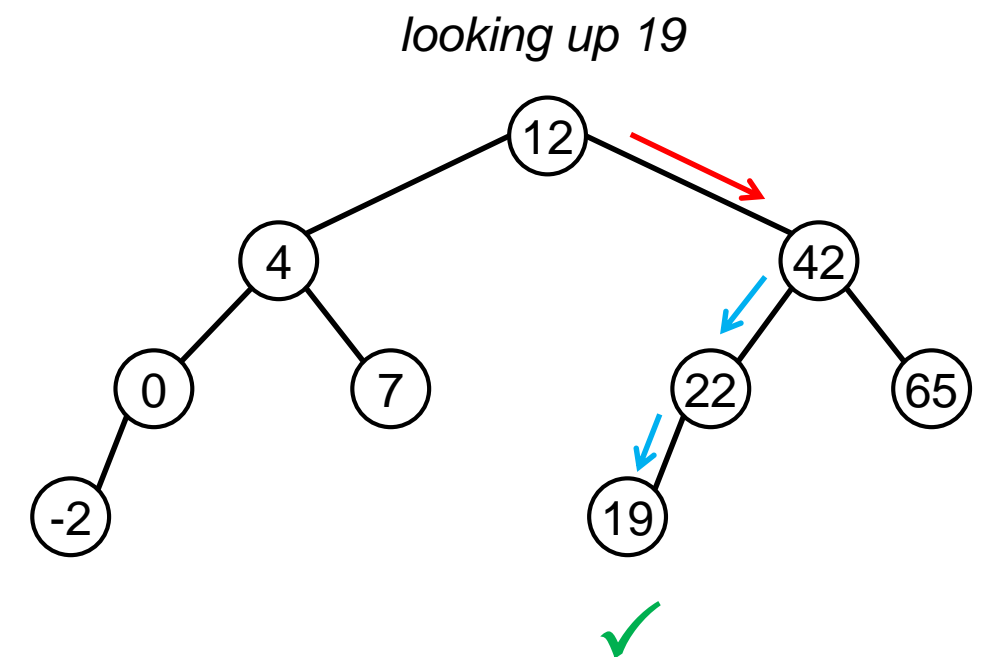
Implementing lookup

- Leverage the structure of the tree!

- empty: the key is not found

- non-empty:

- if root contains the key, **found**
 - if key is smaller than the root's **go left**
 - if key is bigger than the root's **go right**



- In code:

```
entry bst_lookup(tree* T, key k)
//@requires is_bst(T);
{
    // Code for empty tree
    if (T == NULL) return NULL;

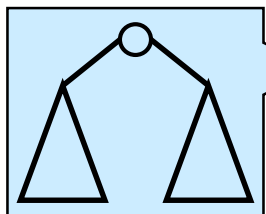
    // Code for non-empty tree
    if (k == T->data) return T->data;
    if (k < T->data) return bst_lookup(T->left, k);
    //@assert k > T->data;
    return bst_lookup(T->right, k);
}
```

lookup works only on BST's

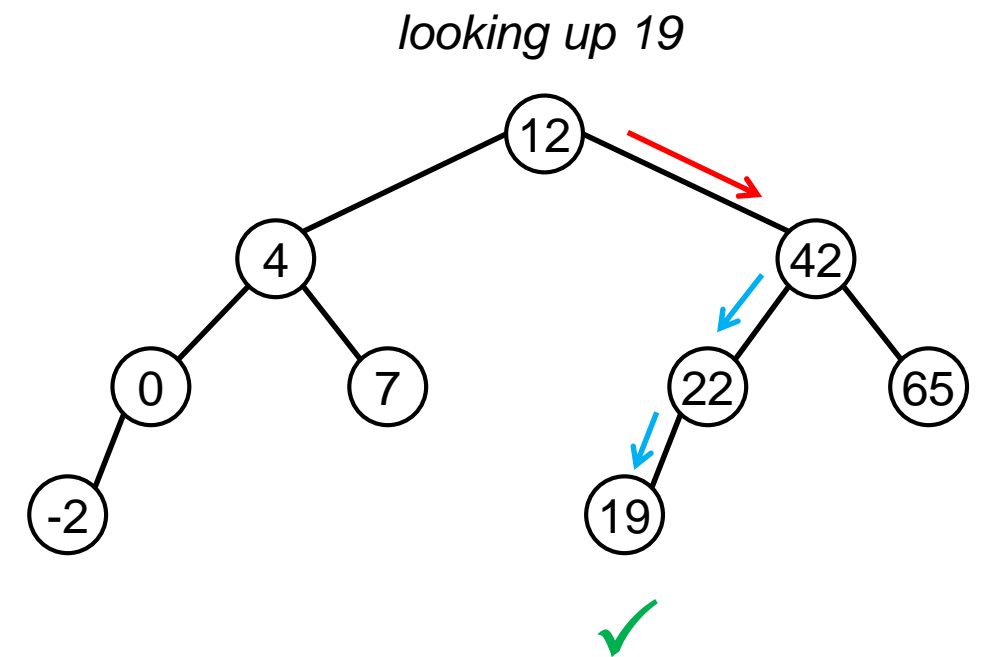
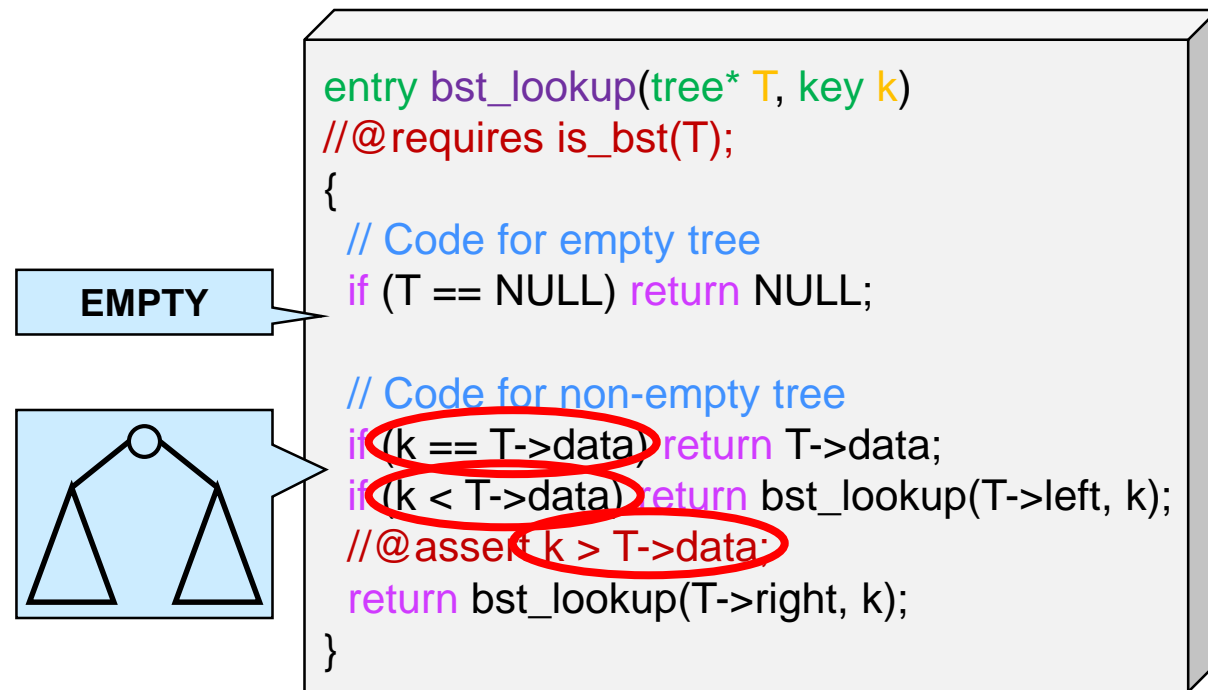
we **go left** by recursing on the left subtree

we **go right** by recursing on the right subtree

EMPTY



Implementing lookup



- *But $<$ and $>$ work only for integers!*
 - *also, keys and entries are not the same thing in general*
- We want a dictionary that uses trees
 - to store **entries of any type**
 - and look them up using **keys of any type**

Implementing lookup

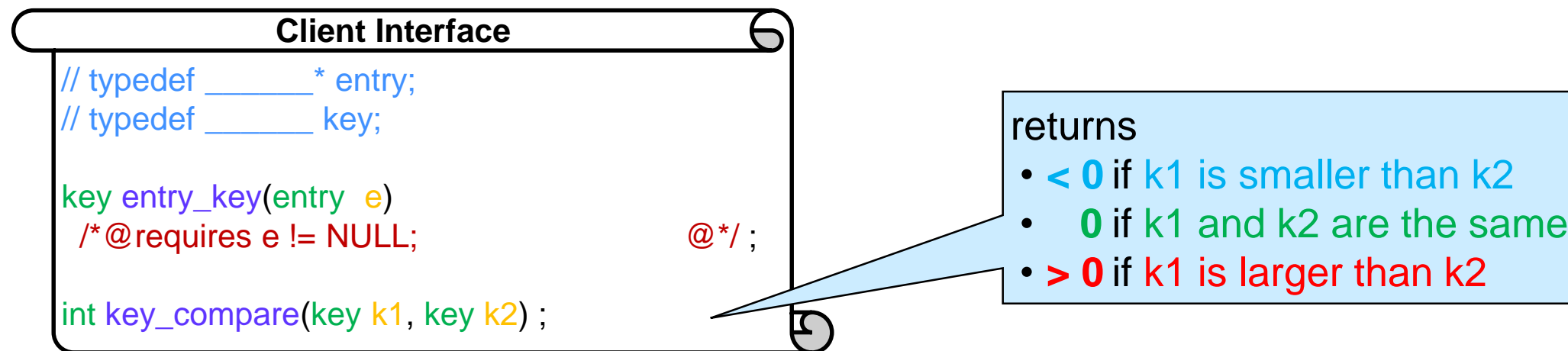
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- *We want a dictionary that uses trees*
 - *to store **entries of any type***
 - *and look them up using **keys of any type***
- We need functions that
 - extract the key from an entry: *entry_key*
 - compare two keys: *key_compare*
- It is for the client to decide on the type of keys and entries
 - So the client shall provide these functions

entry
key

just like for hash dictionaries

A Client Interface

- The BST dictionary needs a **client interface** that
 - requests the client to provide types **entry** and **key**
 - declares a function to extract the key of an entry
 - declares a function to compare two keys



- *We could make it fully generic*
 - *but let's keep things simple*

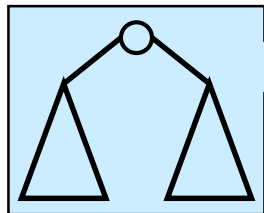
Implementing lookup

- With it, we can write a general implementation

```
entry bst_lookup(tree* T, key k)
//@requires is_bst(T);
//@ensures \result == NULL
        || key_compare(entry_key(\result), k) == 0;
{
    // Code for empty tree
    if (T == NULL) return NULL;

    // Code for non-empty tree
    int cmp = key_compare(k, entry_key(T->data));
    if (cmp == 0) return T->data;
    if (cmp < 0) return bst_lookup(T->left, k);
    //@assert cmp > 0;
    return bst_lookup(T->right, k);
}
```

EMPTY



We save the outcome of the comparison in the variable **cmp**

found!

go left

go right

- **< 0** if **k1** is smaller than **k2**
- **0** if **k1** and **k2** are the same
- **> 0** if **k1** is larger than **k2**

- We can now even provide a useful postcondition

➤ either lookup returns NULL

□ no entry in T has key k

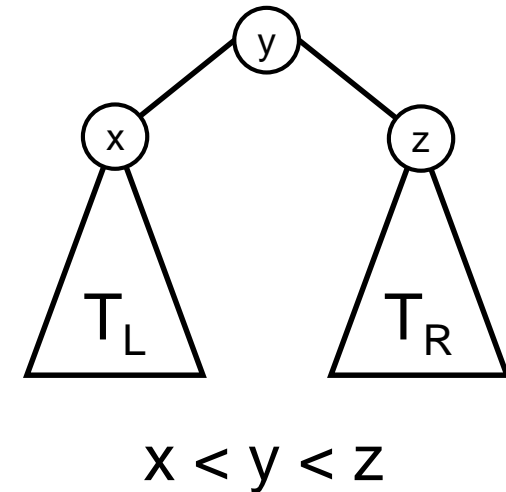
➤ or the key of the returned entry is the same as k

just like for
hash dictionaries

Checking Ordering

Ordered Trees – I

- The data in every node must be
 - bigger than its left child's
 - smaller than its right child's



- In code:

```
bool is_ordered(tree* T)
//@requires is_tree(T);
{
    // Code for empty tree
    if (T == NULL) return true;

    // Code for non-empty tree
    return (T->left == NULL || T->left->data < T->data)
        && (T->right == NULL || T->data < T->right->data)
        && is_ordered(T->left)
        && is_ordered(T->right);
}
```

EMPTY

The empty tree is ordered

If T has a left child, it must be smaller

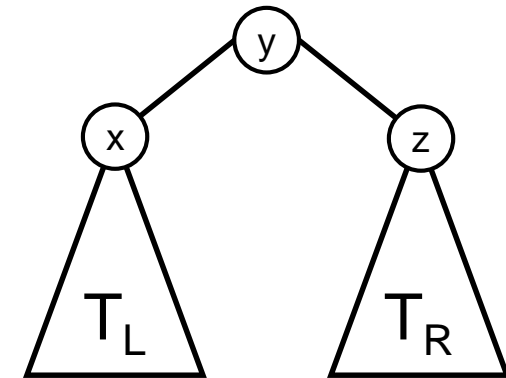
The left subtree must be ordered

and similarly
on the right

For simplicity,
assume `int` data

Ordered Trees – I

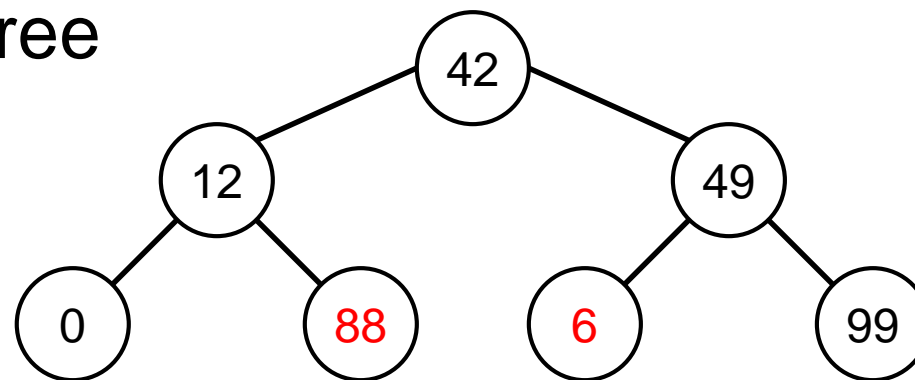
- *The data in every node must be*
 - *bigger than its left child's*
 - *smaller than its right child's*



$$x < y < z$$

- *Is this enough?*

- This is true of this tree



lookup cannot find 88 and 6

- but it is **not** ordered

X

- To be ordered, we want $T_L < y < T_R$

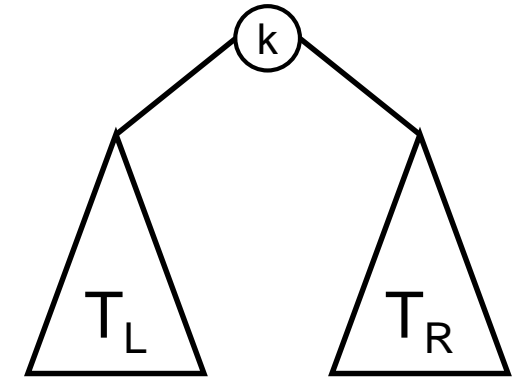
- not $x < y < z$

This is a **global** constraint:
we need to check the whole subtrees

This is a **local** constraint:
it only checks the children of each node

Ordered Trees – II

- The data in every node must be
 - bigger than **everything** in its left subtree
 - smaller than **everything** in its right subtree
- We need two helper functions
 - `gt_tree` that checks $k > T_L$ (i.e., $T_L < k$)
 - `lt_tree` that checks $k < T_R$



$$T_L < k < T_R$$

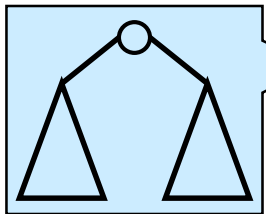


```
bool gt_tree(key k, tree* T) // checks that T < k
//@requires is_tree(T);
{
    // Code for empty tree
    if (T == NULL) return true;

    // Code for non-empty tree
    return key_compare(k, entry_key(T->data)) > 0
        && gt_tree(k, T->left)
        && gt_tree(k, T->right);
}
```

- `gt_tree` has cost $O(n)$
 - if T contains n nodes
 - because it compares k with every node in T

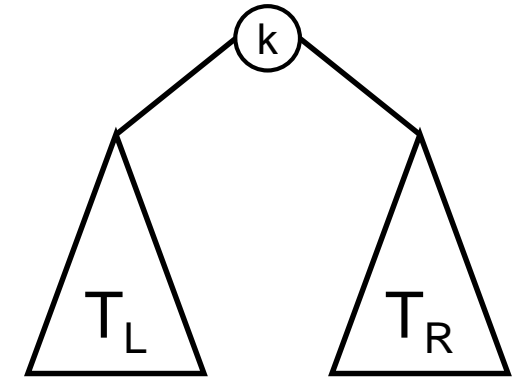
EMPTY



`lt_tree` is similar

Ordered Trees – II

- *The data in every node must be*
 - *bigger than everything in its left subtree*
 - *smaller than everything in its right subtree*



$$T_L < k < T_R$$



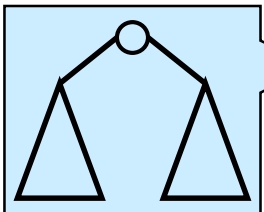
- In code:

```
bool gt_tree(key k, tree* T) {...} // O(n)
bool lt_tree(key k, tree* T) {...} // O(n)

bool is_ordered(tree* T)
//@requires is_tree(T);
{
    // Code for empty tree
    if (T == NULL) return true;

    // Code for non-empty tree
    key k = entry_key(T->data);
    return is_ordered(T->left) && gt_tree(k, T->left)
        && is_ordered(T->right) && lt_tree(k, T->right);
}
```

EMPTY



- **is_ordered** costs $O(n^2)$
 - if T contains n nodes
 - because it calls **gt_tree** and **lt_tree** on each node

Ordered Trees – III

- *Can we do better than $O(n^2)$?*

Even though we typically don't care about the cost of specification functions

- As we examine each key k , keep track of its **allowed range**

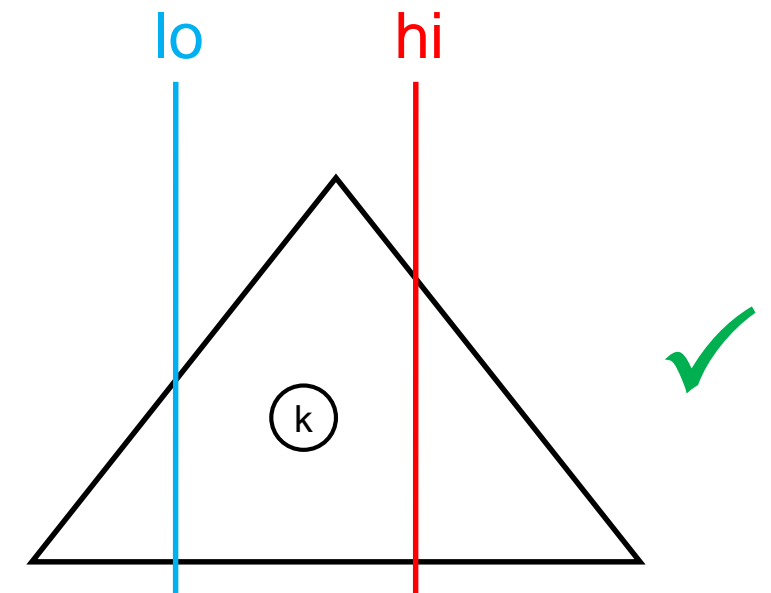
- if $lo < k < hi$, then

- $lo < k_L < k$ for the key k_L of its left child (if any)

- $k < k_R < hi$ for the key k_R of its right child (if any)

- if k is the root, then $-\infty < k < \infty$

This assumes integer keys



- For arbitrary keys,

- use *entries* as the bounds and *entry_key* to extract their key

- use *key_compare* to compare k with another key

- use NULL as $-\infty$ and ∞

NULL is a value of type *entry* that is not a valid entry

Ordered Trees – III

- *As we examine each key k , keep track of its allowed range*

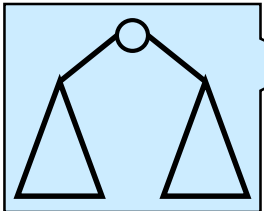
- In code:

We carry around the range (lo , hi) as additional parameters

```
bool is_ordered(tree* T, entry lo, entry hi)
//@requires is_tree(T);
{
    // Code for empty tree
    if (T == NULL) return true;

    // Code for non-empty tree
    key k = entry_key(T->data);
    return (lo == NULL || key_compare(entry_key(lo), k) < 0)
        && (hi == NULL || key_compare(k, entry_key(hi)) < 0)
        && is_ordered(T->left, lo, T->data)
        && is_ordered(T->right, T->data, hi);
}
```

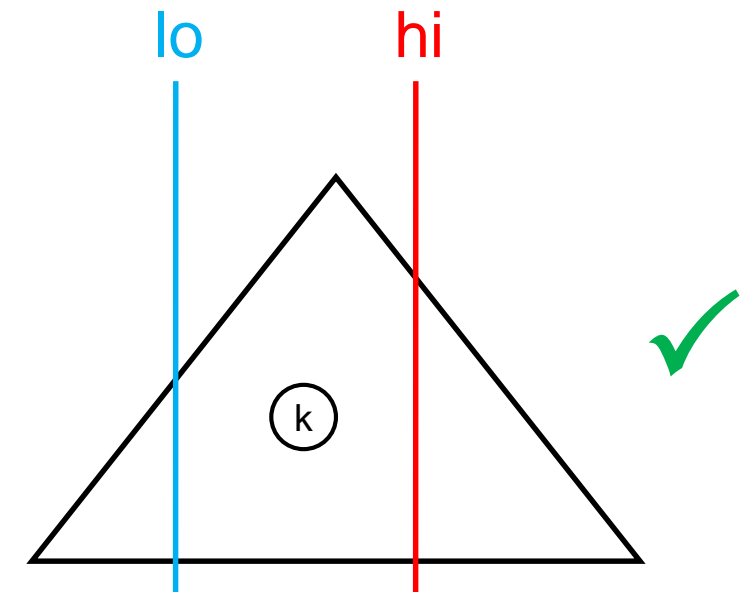
EMPTY



Check that $lo < k < hi$

Check that $lo < T_L < k$

Check that $k < T_R < hi$



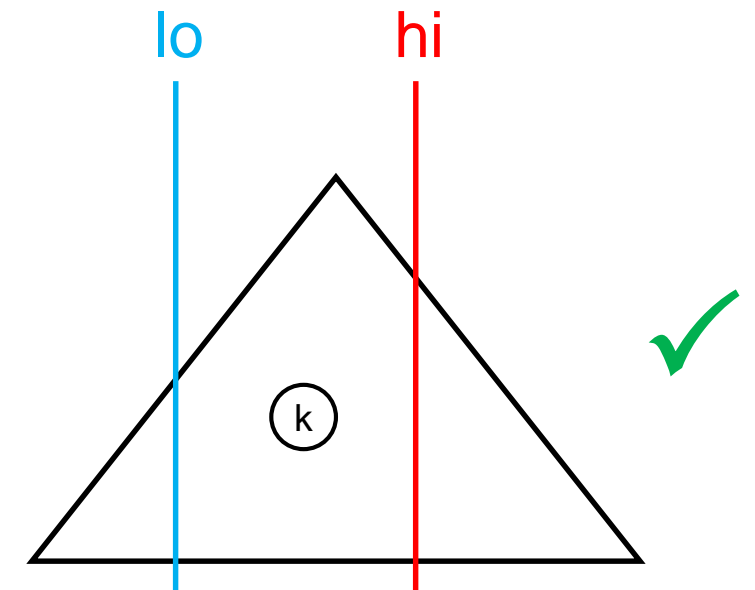
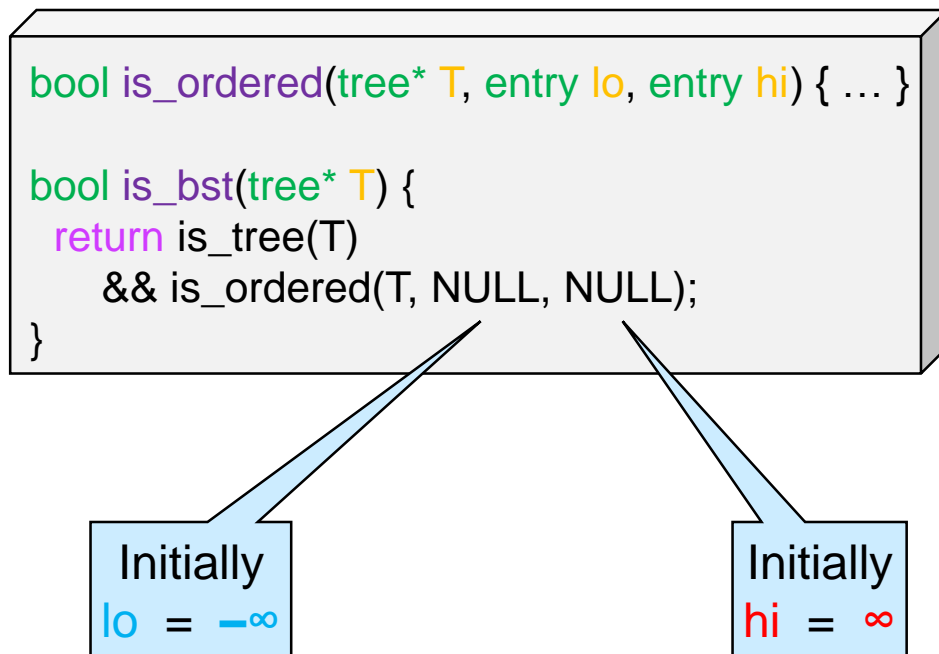
- Complexity: $O(n)$

➤ if T contains n nodes

○ we test every node in the tree

Ordered Trees – III

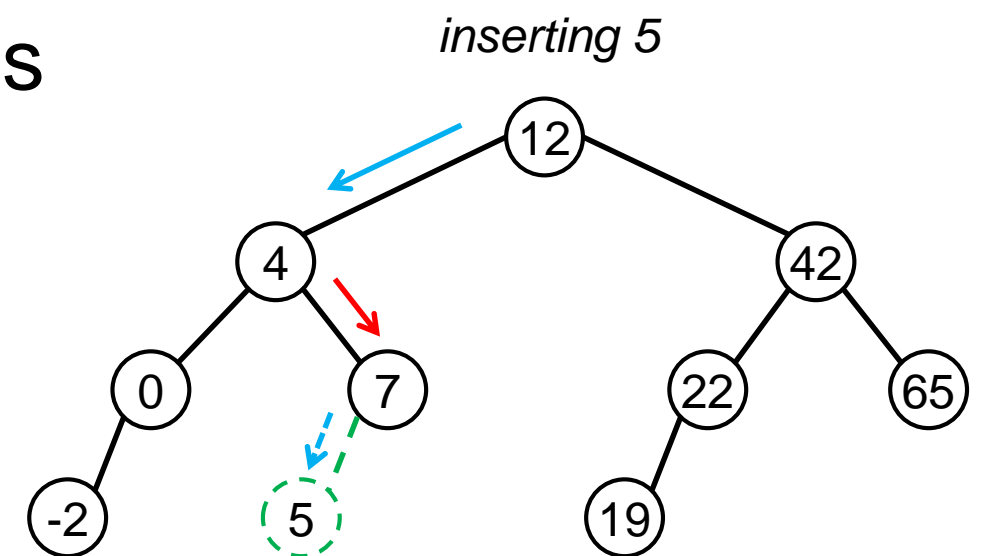
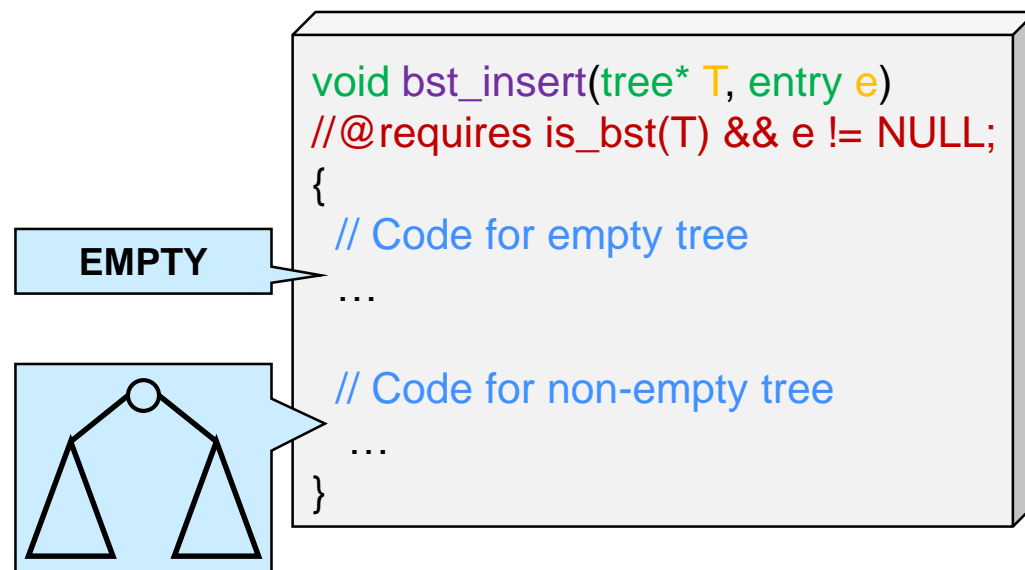
- We need to update `is_bst` slightly



Inserting Entries

Inserting into a BST

- *Do the same steps we would do to search for this entry, and then put it where it should have been*
- The code follows the possible shapes of the tree

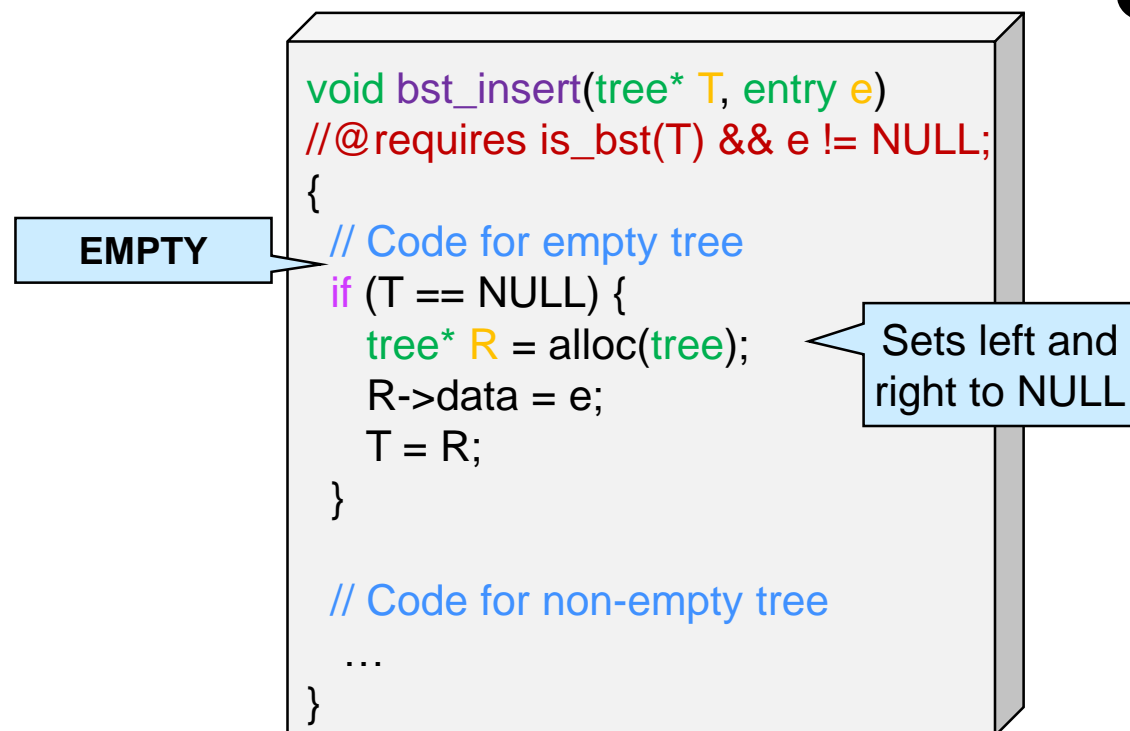


Inserting into an Empty BST

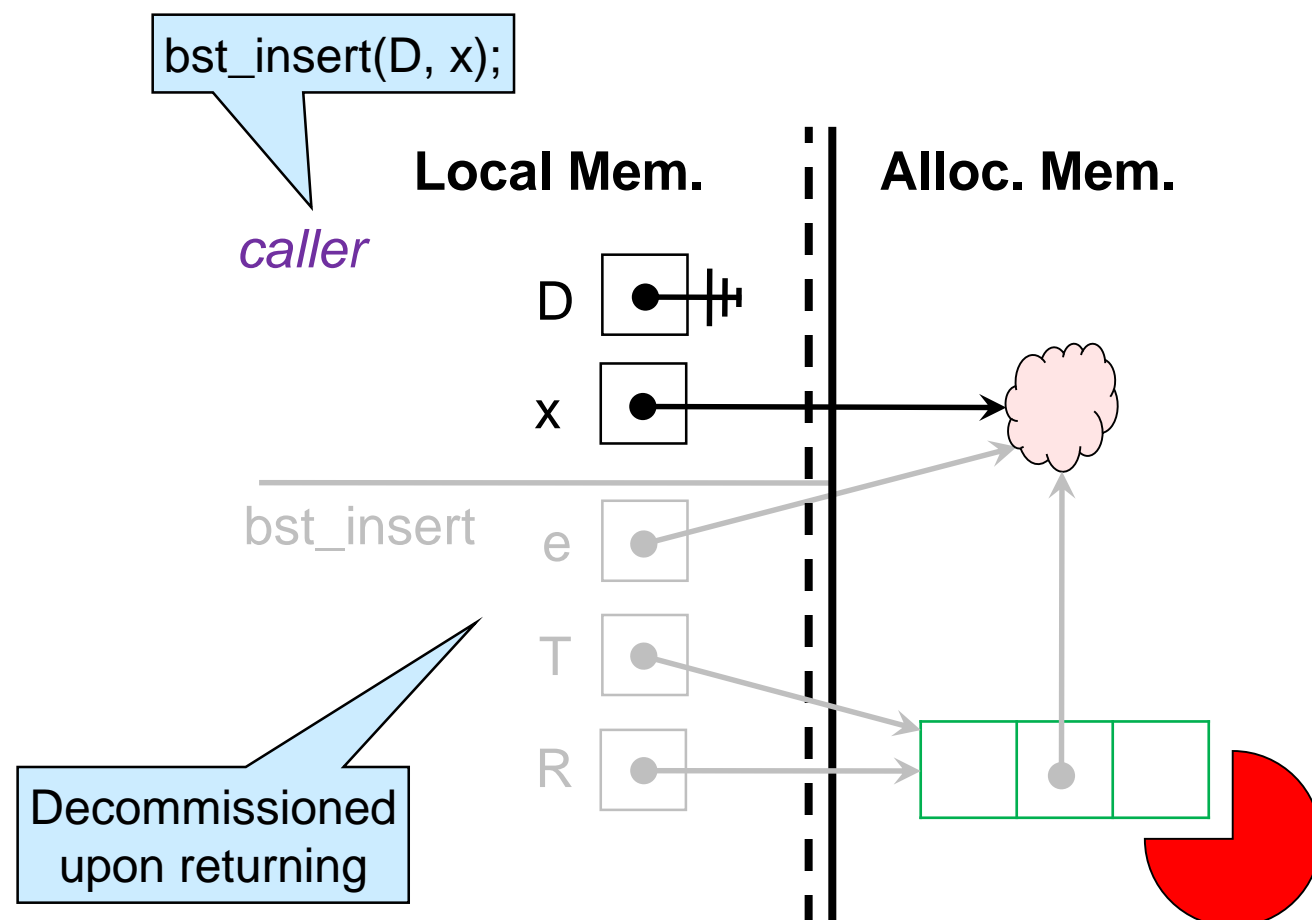
- We simply create a node for the new entry

inserting 5

5



- Does this achieve what we want?
 - No: T is a **copy** of the caller's tree
 - changing T does not change the original



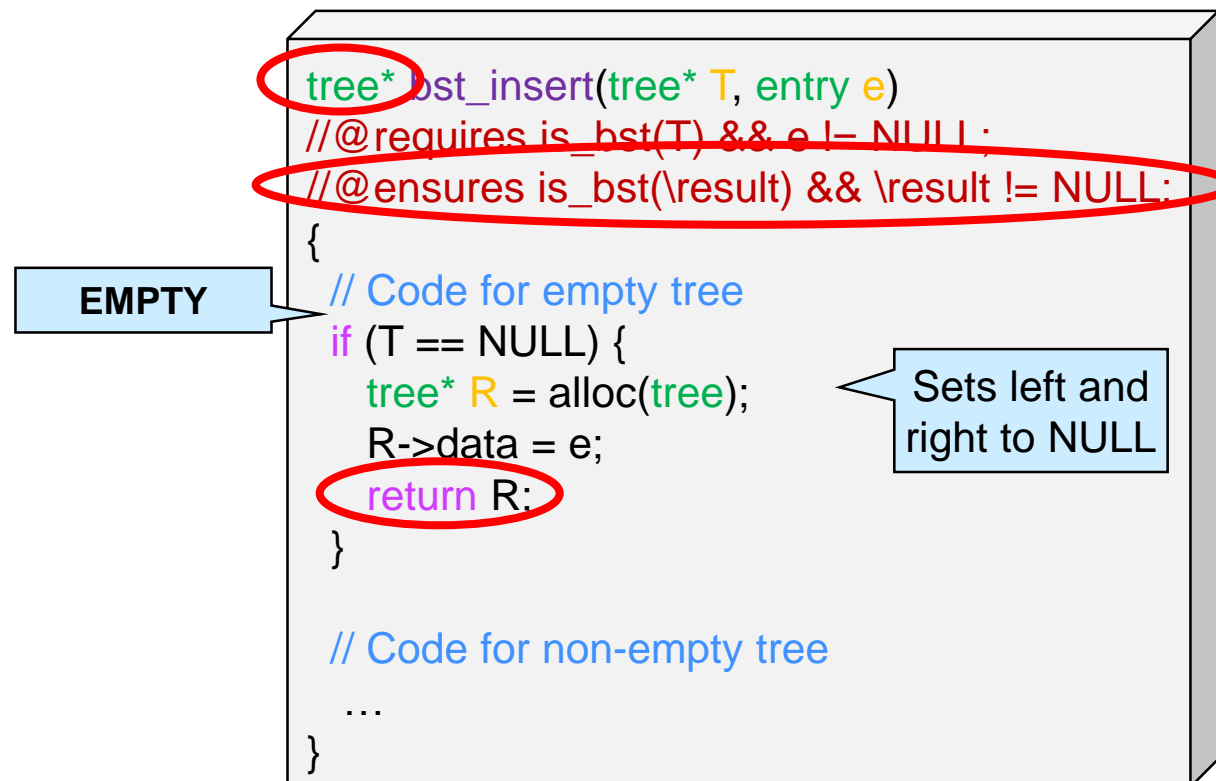
- We need to **return** the new node to the caller
 - `bst_insert` must return a **tree**

Inserting into an Empty BST

- We simply create a node for the new entry **and return it**

inserting 5

5

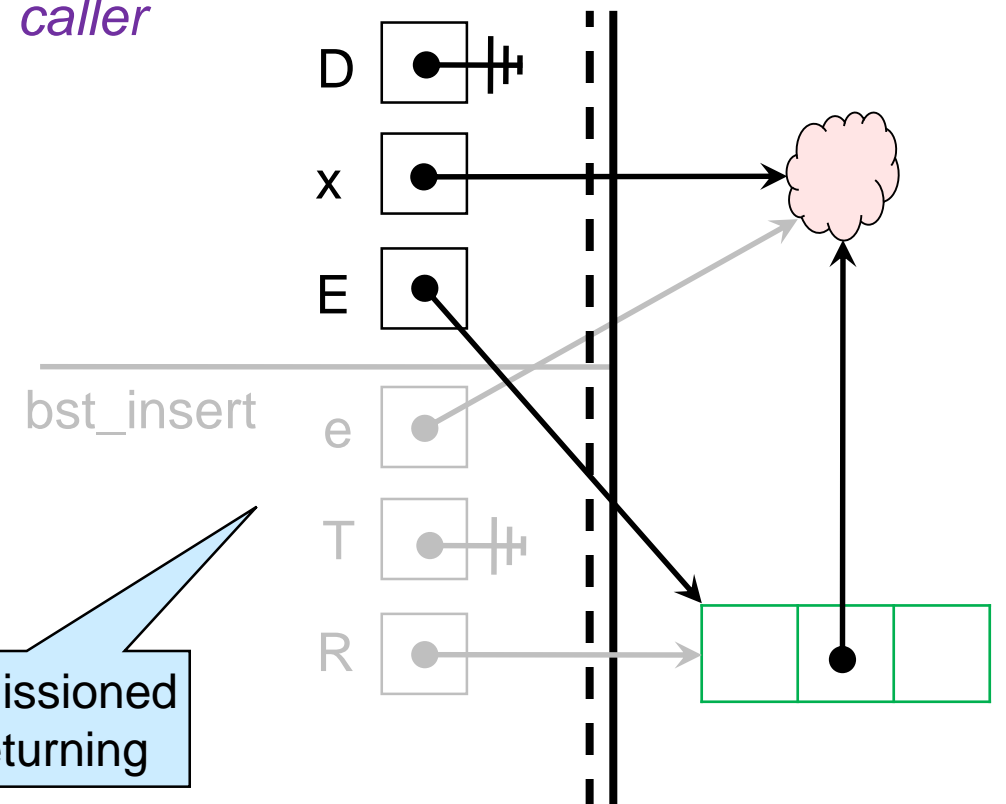


tree* E = bst_insert(D, x);

caller

Local Mem.

Alloc. Mem.



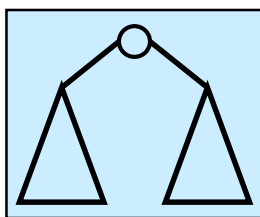
Decommissioned upon returning

- The returned tree must be a valid BST

Inserting in a Non-empty BST

- If an entry with the same key is present, we overwrite it

EMPTY



```
tree* bst_insert(tree* T, entry e)
//@requires is_bst(T) && e != NULL;
//@ensures is_bst(\result) && \result != NULL;
//@ensures bst_lookup(\result, entry_key(e)) == e;
{
    // Code for empty tree
    if (T == NULL) {
        tree* R = alloc(tree);
        R->data = e;
        return R;
    }

    // Code for non-empty tree
    int cmp = key_compare(entry_key(e), entry_key(T->data));
    if (cmp == 0) T->data = e;
    else if (cmp < 0) T->left = bst_insert(T->left, e);
    else { //@assert cmp > 0;
        T->right = bst_insert(T->right, e);
    }
    return T;
}
```

Additional postcondition

- < 0 if k1 is smaller than k2
- 0 if k1 and k2 are the same
- > 0 if k1 is larger than k2

We save the outcome of the comparison in the variable `cmp`

- When inserting in the left subtree, we **reattach** the tree returned by the recursive call
 - the pointer is the same except if it was NULL
- and similarly on the right

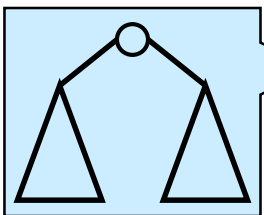
Inserting into a BST

```
tree* leaf(entry e)
//@requires e != NULL;
//@ensures is_bst(\result) && \result != NULL;
{
    tree* T = alloc(tree);
    T->data = e;
    T->left = NULL;    // not necessary
    T->right = NULL;   // not necessary
    return T;
}

tree* bst_insert(tree* T, entry e)
//@requires is_bst(T) && e != NULL;
//@ensures is_bst(\result) && \result != NULL;
//@ensures bst_lookup(\result, entry_key(e)) == e;
{
    // Code for empty tree
    if (T == NULL) return leaf(e);

    // Code for non-empty tree
    int cmp = key_compare(entry_key(e), entry_key(T->data));
    if (cmp == 0) T->data = e;
    else if (cmp < 0) T->left = bst_insert(T->left, e);
    else { // @assert cmp > 0;
        T->right = bst_insert(T->right, e);
    }
    return T;
}
```

EMPTY



- We make `bst_insert` more readable by
 - moving the code that creates a new leaf into a helper function
 - explicitly setting its children to NULL

Refactoring code to make it more readable is important for **maintainability**

BST Dictionaries

Are we There Yet?

- Our target dictionary interface is

Library Interface

```
// typedef _____* dict_t;

dict_t dict_new()
/*@ensures \result != NULL;          @*/;

entry dict_lookup(dict_t D, key k)
/*@requires D != NULL;              @*/
/*@ensures \result == NULL
           || key_compare(entry_key(\result), k) == 0; @*/;

void dict_insert(dict_t D, entry e)
/*@requires D != NULL && e != NULL;  @*/
/*@ensures dict_lookup(D, entry_key(e)) == e; @*/;

entry dict_min(dict_t D)
/*@requires D != NULL;              @*/;
```

Like hash
dictionaries

... plus
find_min

with this client
interface

Client Interface

```
// typedef _____* entry;
// typedef _____ key;

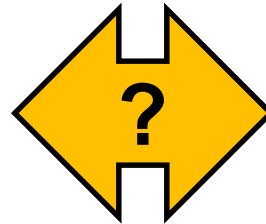
key entry_key(entry e)
/*@requires e != NULL;          @*/;

int key_compare(key k1, key k2);
```

- So far, we have implemented lookup and insertion

Are we There Yet?

```
entry bst_lookup(tree* T, key k);  
tree* bst_insert(tree* T, entry e);
```



Library Interface

```
// typedef _____* dict_t;  
  
dict_t dict_new()  
/*@ensures \result != NULL; @*/  
  
entry dict_lookup(dict_t D, key k)  
/*@requires D != NULL; @*/  
/*@ensures \result == NULL  
|| key_compare(entry_key(\result), k) == 0; @*/  
  
void dict_insert(dict_t D, entry e)  
/*@requires D != NULL && e != NULL; @*/  
/*@ensures ndict_lookup(D, entry_key(e)) == e; @*/  
  
entry dict_min(dict_t D)  
/*@requires D != NULL; @*/
```

- They do not match!

- `bst_insert` returns a `tree*` but `dict_insert` does not return anything
- `NULL` is a valid BST but not a valid dictionary

Implementing BST Dictionaries

- We can define a **header** that contains a pointer to a tree
 - and possibly other data

```
struct dict_header {  
    tree* root;  
    int size; // example of other data  
};  
typedef struct dict_header dict;
```

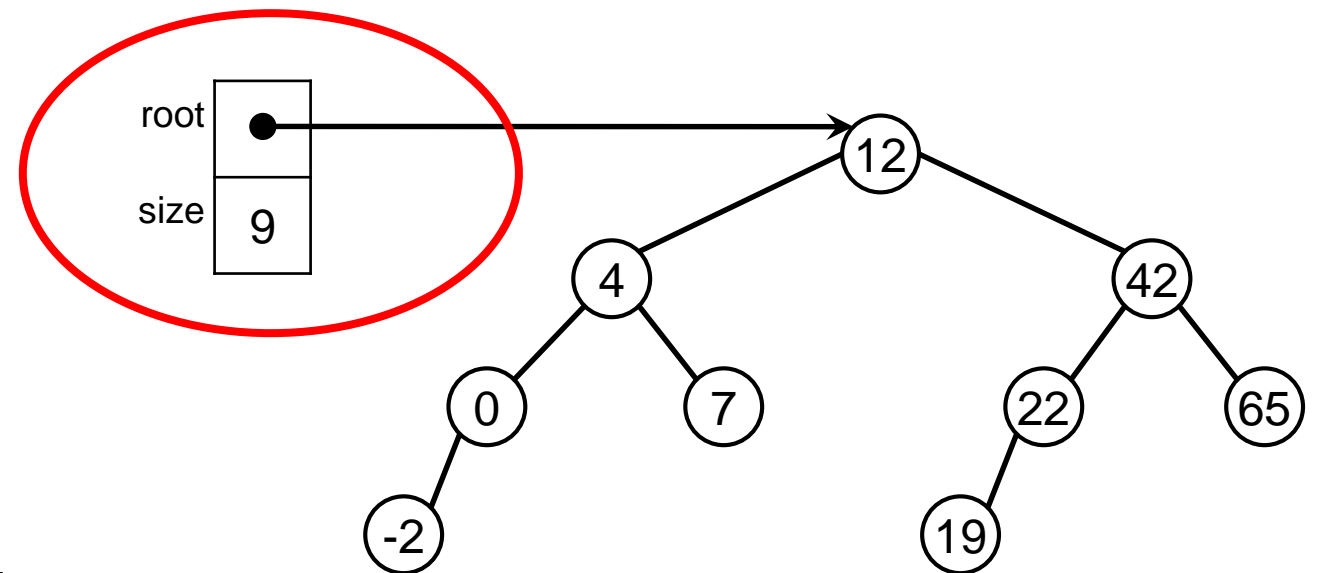
- and **wrappers** around the BST functions

- they mediate between **trees** and **dicts**

- Here's the specification function for BST dictionaries

```
bool is_dict(dict* D) {  
    return D != NULL  
        && is_bst(D->root);  
}
```

ignoring
other data



- the dictionary itself can't be NULL
 - this satisfies the dictionary interface
 - but the underlying BST can
 - that's how we represent the empty dictionary

Implementing BST Dictionaries

```
struct dict_header {  
    tree* root;  
    int size; // example of other data  
};  
typedef struct dict_header dict;
```

- We define **wrappers** around the BST functions

- they mediate between the **trees** and **dicts**

Lookup

```
entry dict_lookup(dict* D, key k)  
//@requires is_dict(D);  
//@ensures \result == NULL  
           || key_compare(entry_key(\result), k) == 0;  
{  
    return bst_lookup(D->root, k);  
}
```

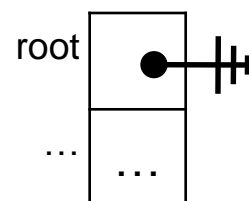
Insertion

```
void dict_insert(dict* D, entry e)  
//@requires is_dict(D) && e != NULL;  
//@ensures dict_lookup(D, entry_key(e)) == e;  
//@ensures is_dict(D);  
{  
    D->root = bst_insert(D->root, e);  
}
```

- Creating a dictionary

- allocates a header and
 - sets the root to the empty BST

```
dict* dict_new()  
//@ensures is_dict(\result);  
{  
    dict* D = alloc(dict);  
    D->root = NULL;  
    return D;  
}
```



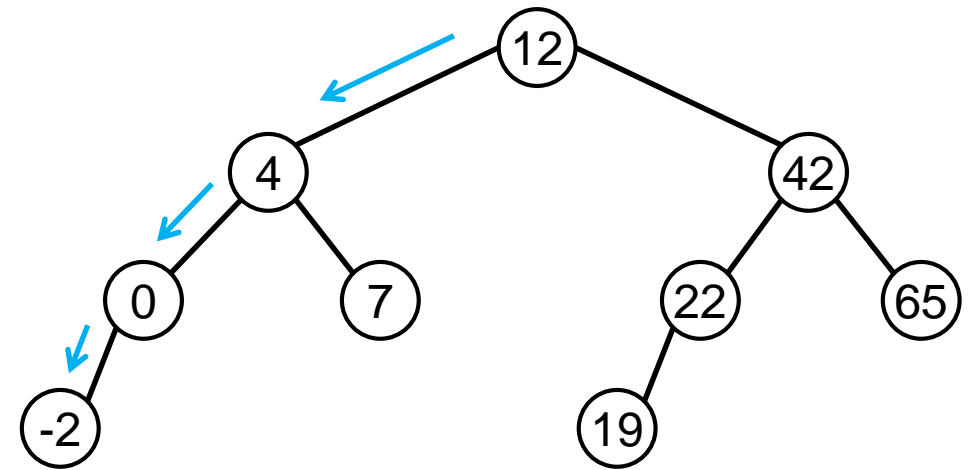
dict_new creates the empty dictionary

Implementing BST Dictionaries

```
struct dict_header {  
    tree* root;  
    int size; // example of other data  
};  
typedef struct dict_header dict;
```

- We are only left with implementing `find_min`

```
entry dict_min(dict* D)  
//@requires is_dict(D);  
{  
    if (D->root == NULL) return NULL;  
    tree* T = D->root;  
    while (T->left != NULL)  
        T = T->left;  
    return T->data;  
}
```



- The abstract client `dict_t` is just `dict*`

```
typedef dict* dict_t;
```

- That's it!



The BST Dictionary Library

```
// BSTs and auxiliary functions
typedef struct tree_node tree;
struct tree_node {
    entry data;      // data != NULL
    tree* left;
    tree* right;
};

// Representation invariant
bool is_bst (tree* T) { ... }

// BST auxiliary functions
entry bst_lookup(tree* T, key k)
/*@requires is_bst(T);
  @ensures \result == NULL
           || key_compare(entry_key(\result), k) == 0;
  { ... }

tree* bst_insert(tree* T, entry e)
/*@requires is_bst(T) && e != NULL;
  @ensures is_bst(\result) && \result != NULL;
  @ensures bst_lookup(\result, entry_key(e)) == e;
  { ... }

// Implementing the dictionary
// Concrete type
struct dict_header {
    tree* root;
};
typedef struct dict_header dict;

// Representation invariant
bool is_dict (dict* D) {
    return D != NULL && is_bst(D->root);
}
```

```
// Implementation of interface functions
dict* dict_new()
/*@ensures is_dict(\result);
{
    dict* D = alloc(dict);
    D->root = NULL;
    return D;
}

entry dict_lookup(dict* D, key k)
/*@requires is_dict(D);
  @ensures \result == NULL
           || key_compare(entry_key(\result), k) == 0;
{
    return bst_lookup(D->root, k);
}

void dict_insert(dict* D, entry e)
/*@requires is_dict(D) && e != NULL;
  @ensures dict_lookup(D, entry_key(e)) == e;
  @ensures is_dict(D);
{
    D->root = bst_insert(D->root, e);
}

entry dict_min(dict* D)
/*@requires is_dict(D);
{
    if (D->root == NULL) return NULL;
    tree* T = D->root;
    while (T->left != NULL)
        T = T->left;
    return T->data;
}

// Client type
typedef dict* dict_t;
```

Implementation

Client Interface

```
// typedef _____* entry;
// typedef _____ key;

key entry_key(entry e)
/*@requires e != NULL;
  @*/;

int key_compare(key k1, key k2);
```

Library Interface

```
// typedef _____* dict_t;

dict_t dict_new()
/*@ensures \result != NULL;
  @*/;

entry dict_lookup(dict_t D, key k)
/*@requires D != NULL;
  @ensures \result == NULL
           || key_compare(entry_key(\result), k) == 0;
  @*/;

void dict_insert(dict_t D, entry e)
/*@requires D != NULL && e != NULL;
  @ensures dict_lookup(D, entry_key(e)) == e;
  @*/;

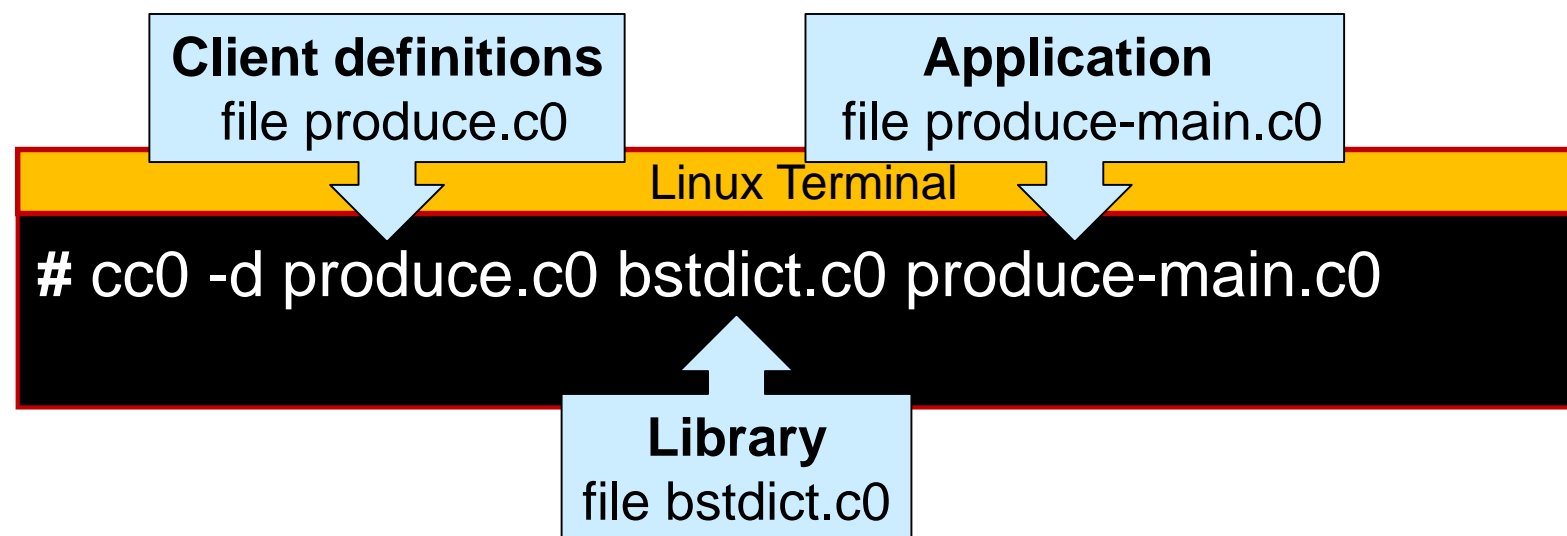
entry dict_min(dict_t D)
/*@requires D != NULL;
  @*/;
```

How

What

Using BST Dictionaries

- We can now use this new implementation of dictionaries for our application
 - once we write an appropriate client definition file



- We could easily make this library fully generic

Recall our Goal

- Develop a data structure that has **guaranteed** $O(\log n)$ worst-case complexity for **lookup**, **insert** and **find_min**
 - always!

- We have succeeded

	<i>Target data structure</i>	
lookup	$O(\log n)$	✓
insert	$O(\log n)$	✓
find_min	$O(\log n)$	✓

○ *or have we ...*