

# Computer Science 331

## Binary Search Trees

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Lectures #18–19

## Outline

- 1 The Dictionary ADT
- 2 Binary Trees
  - Definitions
  - Relationship Between Size and Height
- 3 Binary Search Trees
  - Definition
  - Searching
  - Finding an Element with Minimal Key
  - BST Insertion
  - BST Deletion
  - Complexity Discussion
- 4 References

## The Dictionary ADT

A *dictionary* is a finite set (no duplicates) of elements.

Each element is assumed to include

- A **key**, used for searches.
  - Keys are required to belong to some ordered set.
  - The keys of the elements of a dictionary are required to be distinct.
- Additional **data**, used for other processing.

Permits the following operations:

- search by key
- insert (key/data pair)
- delete an element with specified key

Similar to Java's Map (unordered) and SortedMap (ordered) interfaces.

## Binary Tree

A **binary tree**  $T$  is a hierarchical, recursively defined data structure, consisting of a set of **vertices** or **nodes**.

A binary tree  $T$  is **either**

- an “empty tree,”

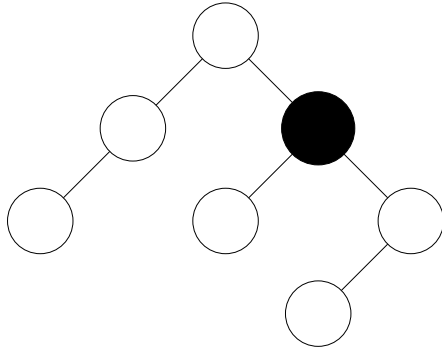
**or**

- a structure that includes
  - the **root** of  $T$  (the node at the top)
  - the **left subtree**  $T_L$  of  $T$  ...
  - the **right subtree**  $T_R$  of  $T$  ...

... where both  $T_L$  and  $T_R$  are also binary trees.

## Example and Implementation Details

## Example:



Each node has a:

- **parent**: unique node above a given node
- **left child**: node in left subtree directly below a given node (root of left subtree)
- **right child**: node in right subtree directly below a given node (root of right subtree)

Each of these may be null

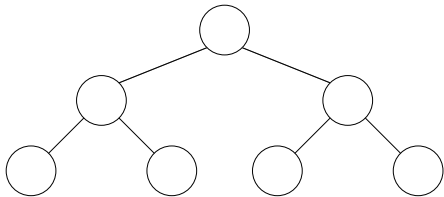
## Additional Terminology

Additional terms related to binary trees:

- **siblings**: two nodes with the same parent
- **descendant (of N)**: any node occurring in the tree with root N
- **ancestor (of N)**: root of any tree containing node N
- **leaf**: node with no children
- **size**: number of nodes in the tree
- **depth (of N)**: length (# of edges) of path from the root to N
- **height**: length of longest path from root to a leaf ( $height(emptytree) = -1$ )

**Note:** depth and height are sometimes (as in the text) defined in terms of number of nodes as opposed to number of edges.

## Size vs. Height: One Extreme



- Size: 7
- Height: 2
- Relationship:

$$n = 1 + 2 + 4 = \sum_{i=0}^h 2^i$$

$$= 2^{h+1} - 1,$$

and

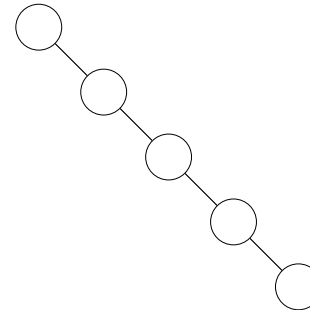
$$h = \log_2(n + 1) - 1$$

This binary tree is said to be *full*:

- all leaves have the same depth
- all non-leaf nodes have exactly two children

**Upper bound:** a binary tree of height  $h$  has size *at most*  $2^{h+1} - 1$ .

## Size vs. Height: Another Extreme



- Size: 5
- Height: 4
- Relationship:  $n = h + 1$

Essentially a linked list!

**Lower bound:** a binary tree with height  $h$  has size *at least*  $h + 1$ .

# Binary Search Tree

A **binary search tree**  $T$  is a data structure that can be used to store and manipulate a finite ordered set or mapping.

- $T$  is a binary tree
- Each element of the dictionary is stored at a node of  $T$ , so

set size = size of  $T$

- In order to support efficient searching, elements are arranged to satisfy the **Binary Search Tree Property** ...

# Binary Search Tree Property

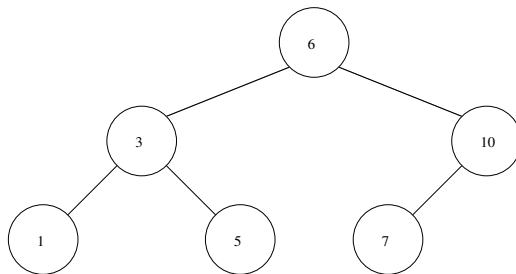
**Binary Search Tree Property:** If  $T$  is nonempty, then

- The left subtree  $T_L$  is a binary search tree including all dictionary elements whose keys are *less than* the key of the element at the root
- The right subtree  $T_R$  is a binary search tree including all dictionary elements whose keys are *greater than* the key of the element at the root

# Example

One binary search tree for a dictionary including elements with keys

$\{1, 3, 5, 6, 7, 10\}$



# Binary Search Tree Data Structure

```

public class BST<E extends Comparable<E>,V> {
    protected bstNode<E,V> root;
    ...

    protected class bstNode<E,V> {
        E key;
        V value;
        bstNode<E,V> left;
        bstNode<E,V> right;
        ...
    }
}
  
```

`bstNode` can also include a reference to its parent

## Specification of “Search” Problem:

### Precondition 1:

- a) T is a BST storing values of some type V along with keys of type E
- b) key is an element of type E stored with a value of type V in T

### Postcondition 1:

- a) Value returned is (a reference to) the value in T with key key
- b) T and key are not changed

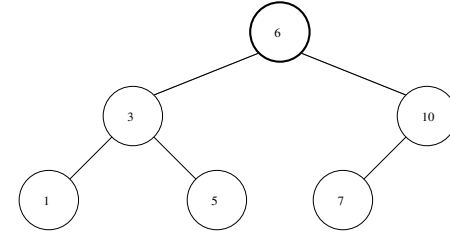
### Precondition 2: same, but key is not in T

### Postcondition 2:

- a) A `NotFoundException` is thrown
- b) T and key are not changed

## Searching: An Example

Searching for 5:



Nodes Visited:

- Start at 6 : since  $5 < 6$ , search in left subtree
- Next node 3 : since  $5 > 3$ , search in right subtree
- Next node 5 : equal to key, so we're finished

## A Recursive Search Algorithm

```

public V search(bstNode<E,V> T, E key)
    throws NotFoundException {
    if (T == null)
        throw new NotFoundException();
    else if (key.compareTo(T.key) == 0)
        return T.value;
    else if (key.compareTo(T.key) < 0)
        return search(T.left, key);
    else
        return search(T.right, key);
}
  
```

## Partial Correctness

Proved by induction on the height of T:

- 1 Base case is correct (empty tree, height  $-1$ )
- 2 Assume that the algorithm is partially correct for all trees of height  $\leq h - 1$ . By the BST property:
  - if `key == root.key`, correctness of output is clear by inspection of the code
  - otherwise, by the BST property:
    - if `key < root.key`, it is in the left subtree (or not in the tree)
    - otherwise `key > root.key` and it must be in the right subtree (or not in the tree)

In either case, algorithm is called recursively on a subtree of height at most  $h - 1$  and outputs correct result by assumption.  $\square$

## Termination and Running Time

Let  $\text{Steps}(T)$  be the number of steps used to search in a BST  $T$  in the worst case. Then there are positive constants  $c_1$ ,  $c_2$  and  $c_3$  such that

$$\text{Steps}(T) \leq \begin{cases} c_1 & \text{if } \text{height}(T) = -1, \\ c_2 & \text{if } \text{height}(T) = 0, \\ c_3 + \max(\text{Steps}(T.\text{left}), \text{Steps}(T.\text{right})) & \text{if } \text{height}(T) > 0. \end{cases}$$

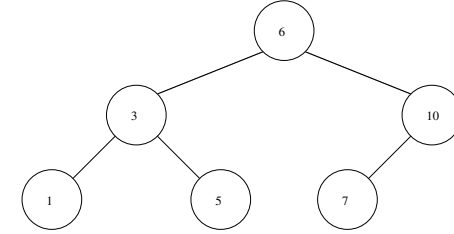
**Exercise:** Use this to prove that

$$\text{Steps}(T) \leq c_3 \times \text{height}(T) + \max(c_1, c_2)$$

**Exercise:** Prove that  $\text{Steps}(T) \geq \text{height}(T)$  as well.

$\implies$  The worst-case cost to search in  $T$  is in  $\Theta(\text{height}(T))$ .

## Minimum Finding: The Idea



Idea: value in a node is the minimum if the node has no left child

- recursively (or iteratively) visit left children
- first node with no left child encountered contains the minimum key

Example: minimum is 1

## A Recursive Minimum-Finding Algorithm

```
// Precondition: T is non-null
// Postcondition: returns node with minimal key,
// null if T is empty
```

```
public bstNode<E,V> findMin(bstNode<E,V> T) {
    if (T == null)
        return null;
    else if (T.left == null)
        return T;
    else
        return findMin(T.left);
}
```

## Analysis: Correctness and Running Time

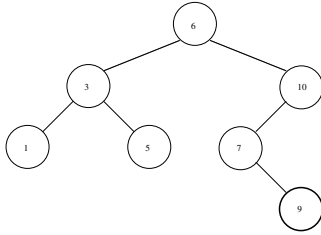
Partial Correctness (tree of height  $h$ ):

- Exercise (similar to proof for Search)

Termination and Bound on Running Time (tree of height  $h$ ):

- after each recursive call, the height is reduced by at least 1
- worst case running time is  $\Theta(h)$  (and hence  $\Theta(n)$ )

## Insertion: An Example



Idea: use search to find empty subtree where node should be

Nodes Visited (inserting 9):

- Start at 6 : since  $9 > 6$ , new node belongs in right subtree
- Next node 10 : since  $9 < 10$ , new node belongs in left subtree
- Next node 7 : since  $9 > 7$ , new node belongs in right subtree
- Next node null: insert new node at this point

## A Recursive Insertion Algorithm

```

// Non-recursive public function calls recursive worker function
public void insert(E key, V value)
    { root = insert(root, key, Value); }

protected
bstNode<E,V> insert(bstNode<E,V> T, E newKey, V newValue) {
    if (T == null)
        T = new bstNode<E,V>(newKey,newValue,null,null);
    else if (newKey.compareTo(T.key) < 0)
        T.left = insert(T.left, newKey, newValue);
    else if (newKey.compareTo(T.key) > 0)
        T.right = insert(T.right, newKey, newValue);
    else
        throw new FoundException();
    return T;
}
  
```

## Analysis: Correctness and Running Time

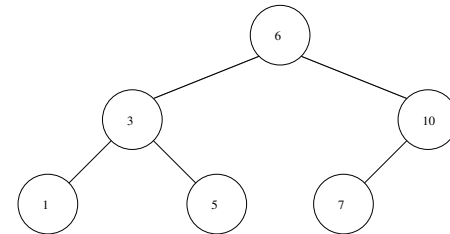
Partial Correctness (tree of height  $h$ ):

- Exercise (similar to proof for Search)

Termination and Bound on Running Time (tree of height  $h$ ):

- worst case running time is  $\Theta(h)$  (and hence  $\Theta(n)$ )
- Proof: exercise

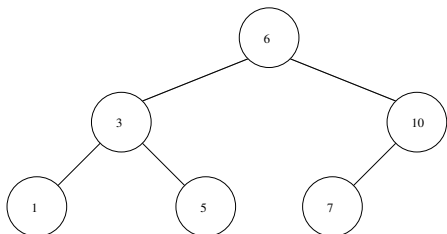
## Deletion: Four Important Cases



Key is/has ...

- 1 Not Found (Eg: Delete 8)
- 2 At a Leaf (Eg: Delete 7)
- 3 One Child (Eg: Delete 10)
- 4 Two Children (Eg: Delete 6)

## First Case: Key Not Found



Idea: search for key 8, throw `notFoundException` when not found

Nodes Visited (delete 8):

- Start at 6 : since  $8 > 6$ , delete 8 from right subtree
- Next node 10 : since  $8 < 10$ , delete 8 from left subtree
- Next node 7 : since  $8 > 7$ , delete 8 from right subtree
- Next node `null`: conclude that 8 is not in the tree

## Algorithm and Analysis

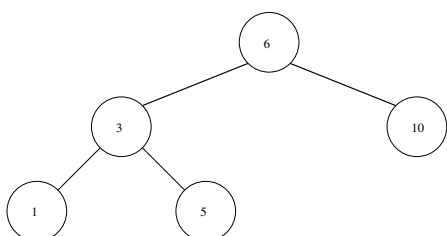
```

protected bstNode<E,V> delete(bstNode<E,V> T, E key) {
    if (T != null) {
        if (key.compareTo(T.key) < 0)
            T.left = delete(T.left, key);
        else if (key.compareTo(T.key) > 0)
            T.right = delete(T.right, key);
        else if ...
            // found node with given key
    }
    else
        throw new notFoundException();
    return T;
}
  
```

Correctness and Efficiency For This Case:

- tree is not modified if key is not found (base case will be reached)
- worst-case cost  $\Theta(h)$  (same as search)

## Second Case: Key is at a Leaf



Idea: set appropriate reference in parent to `null`

Nodes Visited (delete 7):

- Start at 6 : since  $7 > 6$ , delete 7 from right subtree
- Next node 10 : since  $7 < 10$ , delete 7 from left subtree
- Next node 7 : set reference to left child of parent to `null`

## Algorithm and Analysis

Extension of Algorithm:

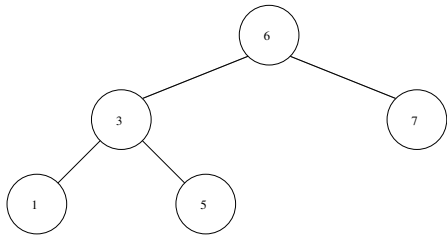
```

else if (T.left == null && T.right == null)
    T = null;
  
```

Correctness and Efficiency For This Case:

- test detects whether the node is a leaf
- replacing T with `null` deletes the leaf at T
- removing a leaf does not affect BST property
- worst-case cost is  $\Theta(h)$  for this case ( $\Theta(h)$  to locate leaf,  $\Theta(1)$  to remove it)

## Third Case: Key is at a Node with One Child



Idea: remove node, put the one subtree in its place

Nodes Visited (delete 10):

- Start at 6 : since  $10 > 6$ , delete 10 from right subtree
- Next node 10 : set reference to right child of parent to child of 10

## Algorithm and Analysis

Extension of Algorithm:

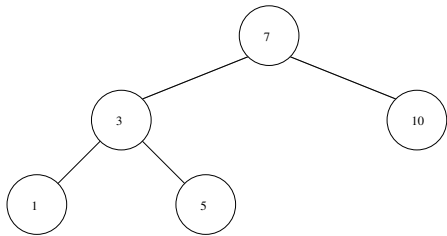
```

else if (T.left == null)
    T = T.right;
else if (T.right == null)
    T = T.left;
  
```

Correctness and Efficiency For This Case:

- T is replaced with its one non-empty subtree
  - node originally at T is deleted
  - BST property still holds (new subtree at T still contains keys that were in the old subtree)
- worst case cost is  $\Theta(h)$  ( $\Theta(h)$  to locate node,  $\Theta(1)$  to remove it)

## Fourth Case: Key is at a Node with Two Children



Idea: replace node with its successor (minimum in the right subtree)

Nodes Visited (delete 6):

- Start at 6 : found node to delete
- replace data at node with data from the node of minimum key in the right subtree
- delete node with minimal key from the right subtree

## Algorithm and Analysis

Extension of Algorithm:

```

else {
    bstNode<E,V> min = findMin(T.right);
    T.key = min.key; T.value = min.value;
    T.right = delete(T.right, T.key);
}
  
```

Correctness and Efficiency For This Case:

- BST property holds: all entries in the new right subtree have keys  $>$  the smallest key from the original right subtree
- worst case cost is  $\Theta(h)$  :
  - findMin costs  $\Theta(h)$  (from last lecture)
  - recursive call deletes a node with at most one child from a tree of height  $< h$  (cost is  $\Theta(h)$ )



## More on Worst Case

All primitive operations (search, insert, delete) have worst-case complexity  $\Theta(n)$

- all nodes have exactly one child (i.e., tree only has one leaf)
- Eg. will occur if elements are inserted into the tree in ascending (or descending) order

On average, the complexity is  $\Theta(\log n)$

- Eg. if the tree is full, the height of the tree is  $h = \log_2(n + 1) - 1$
- the height of a randomly constructed tree (inserting  $n$  elements uniformly randomly) is  $3 \log_2 n$  for sufficiently large  $n$  (see lecture supplement)

Need techniques to ensure that all trees are close to full

- want  $h \in \Theta(\log n)$  in the worst case
- one possibility: red-black trees (next three lectures)

## References

**Introduction to Algorithms**, Chapter 12

and,

**Data Structures: Abstraction and Design Using Java**, Chapter 6.1-6.4