# Binary Search Trees

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COMPCSI220: WEEK 9





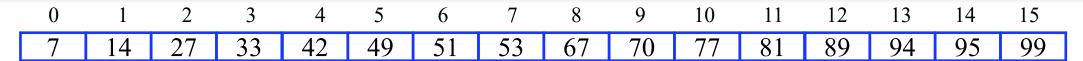
#### OUTLINE

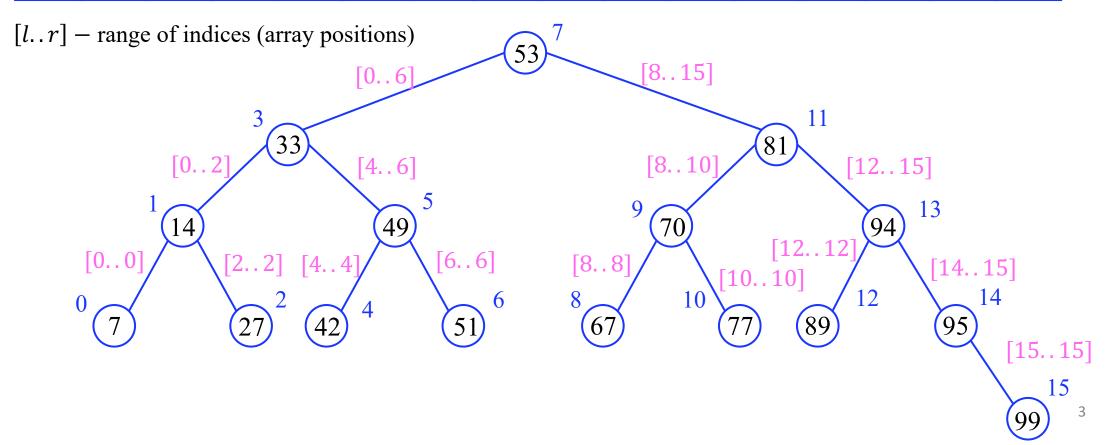
Tree Data Structure

- Binary Search Tree Operations
- Time Complexity Analysis



#### Tree Structure of Binary Search

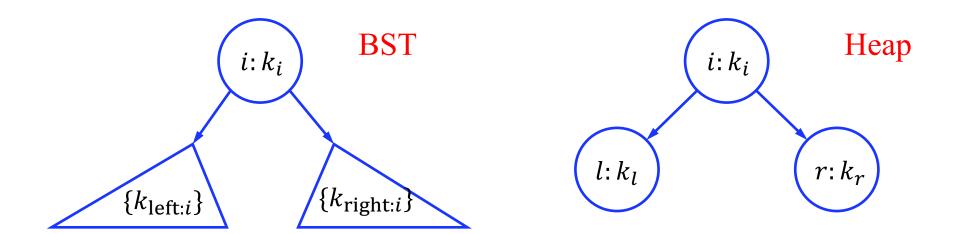




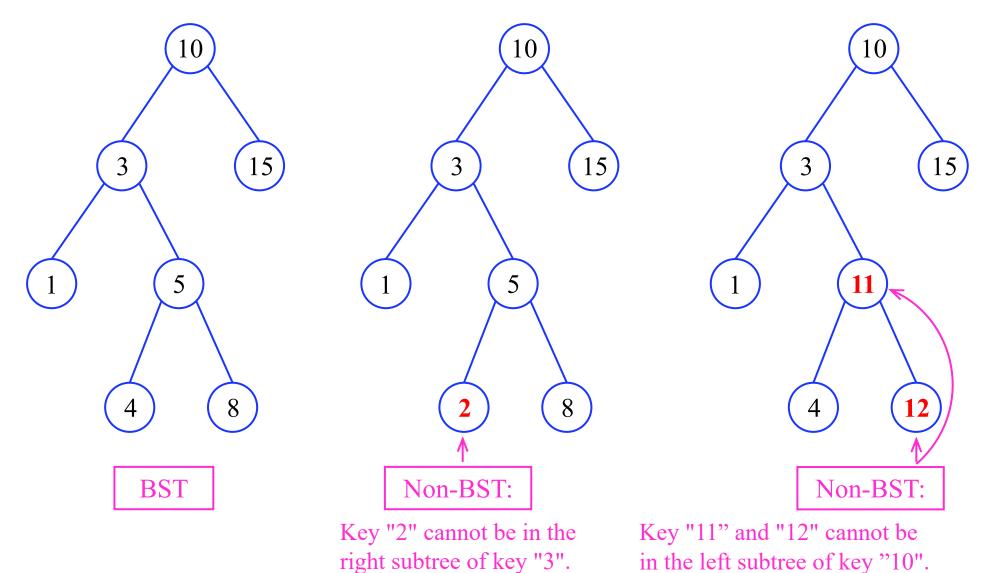


#### Binary Search Tree: Left-Right Ordering of Keys

- Left-to-right numerical ordering in a BST: for every node i,
  - the values of all the keys  $k_{\text{left};i}$  in the left subtree are smaller than the key  $k_i$  in i and
  - the values of all the keys  $k_{\mathrm{right}:i}$  in the right subtree are larger than the key  $k_i$  in i:  $\{k_{\mathrm{left}:i}\} \ni l < k_i < r \in \{k_{\mathrm{right}:i}\}$



# Binary Search Tree: Left-Right Ordering of Keys





#### Binary Search Tree Operations

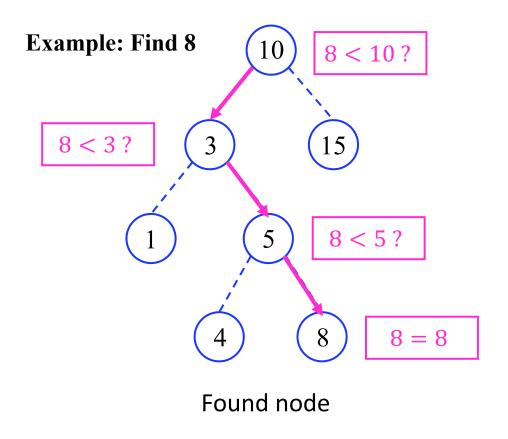
- BST is an explicit data structure implementing the table ADT.
  - BST are more complex than heaps: any node may be removed, not only a root or leaves.
  - The only practical constraint: no duplicate keys (attach them all to a single node).
- Basic operations
  - Find a given search key or detect that it is absent in the BST.
  - Insert a node with a given key to the BST if it is not found.
  - FindMin: find the minimum key.
  - findMax: find the maximum key.
  - Remove a node with a given key and restore the BST if necessary.

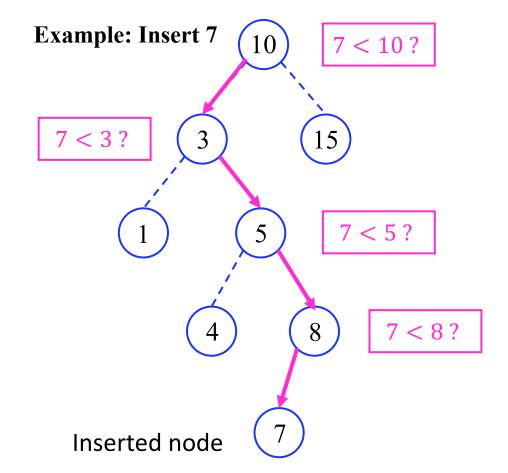


## BST Operations: Find / Insert a Node

find: a successful binary search

**insert**: creating a new node at the point where an unsuccessful search stops.







## BST Operations: FindMin / FindMax

- Extremely simple: starting at the root, branch repeatedly left (findMin) or right (findMax) as long as a corresponding child exists.
- The root of the tree plays a role of the pivot in quicksort and quickselect.
- As in quicksort, the in-order traversal of the tree can sort the items:
  - First visit the left subtree;
  - Then visit the root, and
  - Then visit the right subtree.
- $O(\log n)$  average-case and O(n) worst-case running time for find, insert, findMin, and findMax operations, as well as for selecting a single item



#### BST Operations: Remove a Node

- The most complex because the tree may be disconnected. Need to reconnect some nodes!
  - Reconnection must retain the ordering condition.
  - Reconnection should not needlessly increase the tree height.

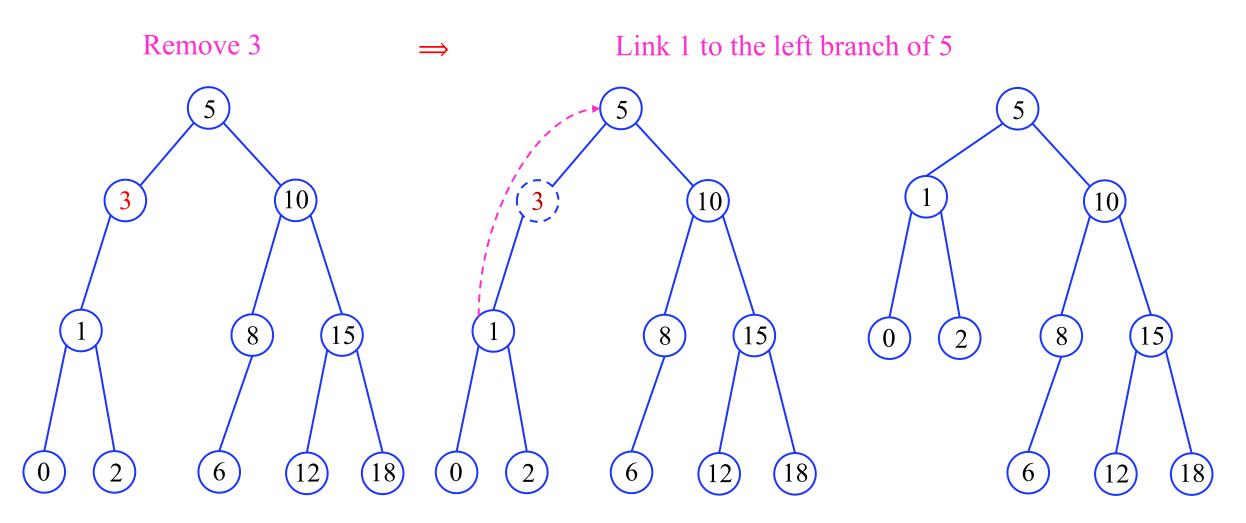


## BST Operations: Remove a Node

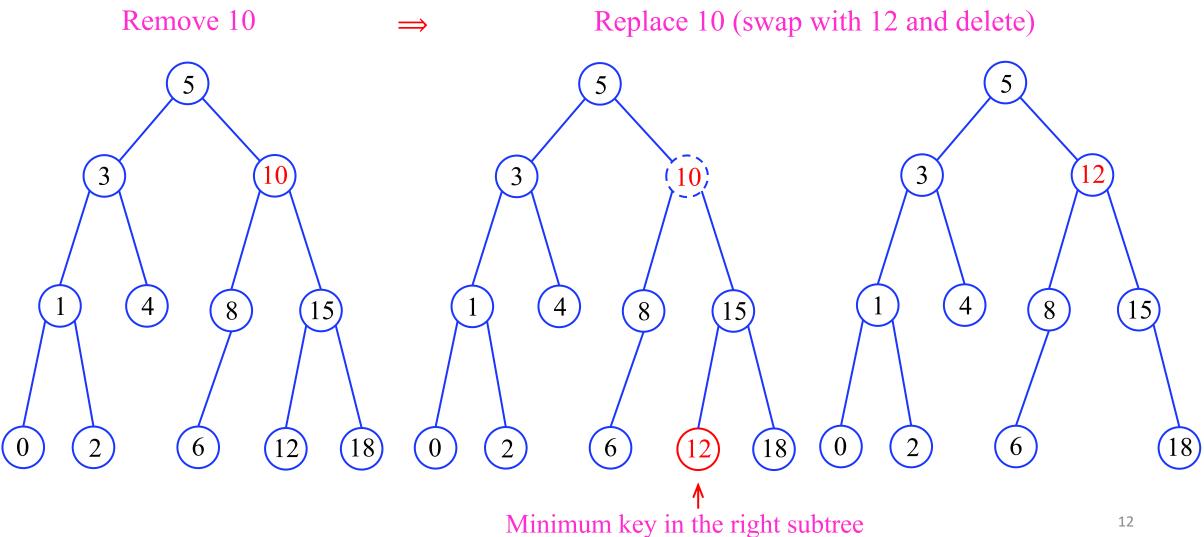
• Standard method of removing a node i with c children:

С	ACTION
0	Simply remove the leaf <i>i</i> .
1	Remove the node <i>i</i> after linking its child to its parent node.
2	Swap the node $i$ with the node $j$ having the smallest key $k_j$ in the right subtree of the node $i$ .
	After swapping, remove the node <i>i</i> (as now it has at most one right child).

# BST Operation: Remove a Node



## BST Operation: Remove a Node



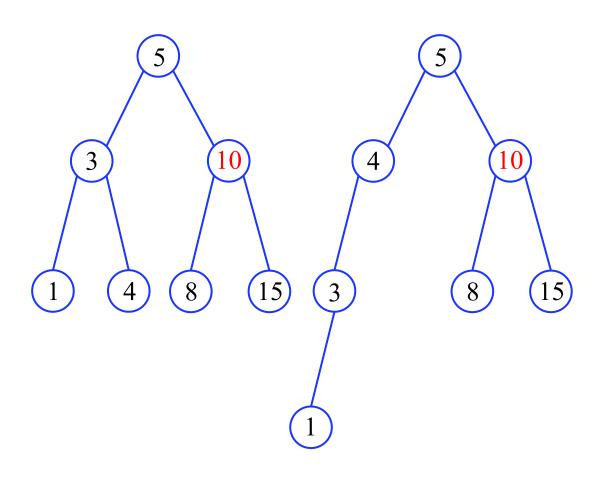


## The Worst-Case Time Complexity

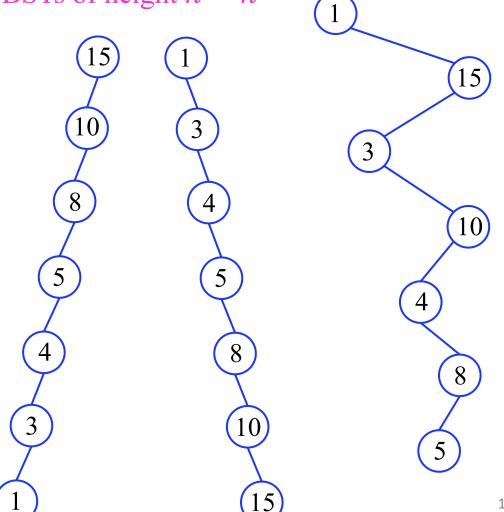
- The find, insert, and remove operations in a BST all take time in O(h) in the worst case, where h is the height of the tree.
- **Proof**: The running time T(n) of these operations is proportional to the number of nodes visited.
  - Find / insert: 1+*h*.
  - Remove: "1 + the depth of the node + the height of its highest subtree"  $\rightarrow$ 1+h.
  - In each case  $T(n) = \Theta(h)$ .
  - For a well-balanced BST,  $T(n) \in O(\log n)$  (logarithmic time).
  - In the worst case  $T(n) \in \Theta(n)$  (linear time) because insertions and deletions may heavily destroy the balance.

# The Worse-Case Time Complexity

BSTs of height  $h \approx \log n$ 



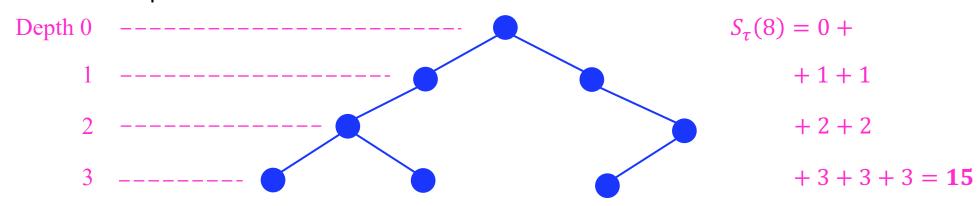
BSTs of height  $h \approx n$ 





## The Average-Case Time Complexity

- More balanced trees are more frequent than unbalanced ones.
- **Definition** (Internal Path Length): The total internal path length,  $S_{\tau}(n)$ , of a binary tree  $\tau$  is the sum of the depths of all its nodes.

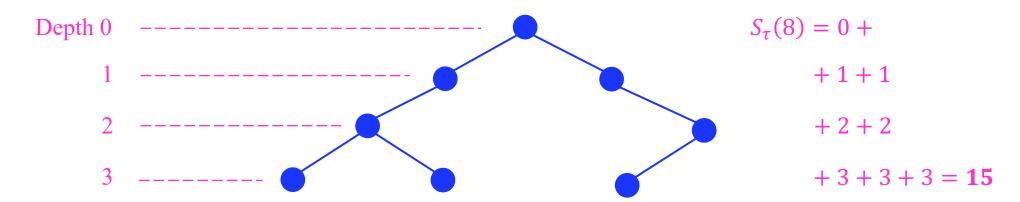


• Average complexity of a successful search in  $\tau$ : the average node depth,  $1/n S_{\tau}(n)$ , e.g. 1/8  $S_{\tau}(8)$  =15/8=1.875 in this example.



## The Average-Case Time Complexity (Contd.)

- Average-case complexity of searching:
  - Averaging  $S_{\tau}(n)$  for all the trees of size n, i.e. for all possible n! Insertion orders, occurring with equal probability  $\frac{1}{n!}$





## The $\Theta(\log n)$ Average-case BST Operations

- Let S(n) be the average of the total internal path length,  $S_{\tau}(n)$ , over all BST  $\tau$  created from an empty tree by sequences of n random insertions, each sequence considered as equally possible.
- The expected time for successful and unsuccessful search (insertion and deletion) in such BST is  $\Theta(\log n)$
- **Proof**: It should be proven that  $S(n) \in \Theta(n \log n)$ 
  - Obviously, S(1) = 0.
  - Any n-node tree, n>1, contains a left subtree with i nodes, a root at level 0, and a right subtree with n-i-1 nodes;  $0 \le i \le n-1$ .
  - For a fixed i, S(n) = (n-1) + S(i) + S(n-i-1), as the root adds 1 to the path length of each other node.



#### The $\Theta(\log n)$ Average-case BST Operations (Contd.)

• After summing these recurrences for  $0 \le i \le n-1$  and averaging, just the same recurrence as for the average-case quicksort analysis is obtained:

$$S(n) = (n-1) + \frac{2}{n} \sum_{i=0}^{n-1} S(i)$$

- $S(n)=(n-1)+\frac{2}{n}\sum_{i=0}^{n-1}S(i)$  Therefore,  $S(n)\in\Theta(n\log n)$ , and the expected depth of a node is  $\frac{1}{n}S(n)\in\Theta(\log n)$ .
- Thus, the average-case search and insertion time is in  $\Theta(\log n)$ .
- It is possible to prove (but in a more complicate way) that the average-case deletion time is also in  $\Theta(\log n)$ .
- The BST allow for a special balancing, which prevents the tree height from growing too much, i.e. avoids the worst cases with linear time complexity  $\Theta(n)$ .



#### **SUMMARY**

- Tree Data Structure
- Binary Search Tree Operations
  - find, insert, and remove operations
- Time Complexity Analysis