

# Binary Search Trees

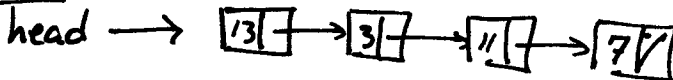
L8.1

Today continue discussion of Data Structures

Operations: Insert, Delete, Search

Introduce: linked List (§ 10.2 of text CLRS)

example



<u>Data Structures</u>	<u>Insert</u>	<u>Delete</u>	<u>Search</u>	<u>Model Elements</u>
Linked List	$O(1)$	$O(1)$	$O(n)$	exact match
Sorted Array	$O(n)$	$O(n)$	$O(\lg n)$	comparison ( $<$ , $>$ , $=$ )
Hash Table	$O(1)$	$O(1)$	$O(1)$	hash function

Insert into head: assumes have pointer & don't search  
 Delete: move data  
 Search: binary search: Divide & conquer with middle element

Since we have achieved constant time with hashing, why would we need any other type of data structure?

It turns out that there are hundreds of data structures and this is a very rich research field. Reason is that different types of structures better suited to different tasks. The 3 operations (insert, delete, search) are not the only ones we may be interested in.

- Simple: minimum (easy for sorted, hard for linked list or hash table)
- Useful: nearest match (hash)

Examples: (Need fast search, and fast min (sorted array))  $O(1)$  in best

- imagining running both structures in parallel
- once built, do search in hash ( $O(1)$ ) - min in sorted array ( $O(1)$ )
- problem: to maintain each data structure's integrity, need to insert/delete into each ( $O(n)$  for sorted array)
- does it help to maintain pointers between corresponding elements of each data structure?

② Inexact Searches

- "Introduction to Algorithms"

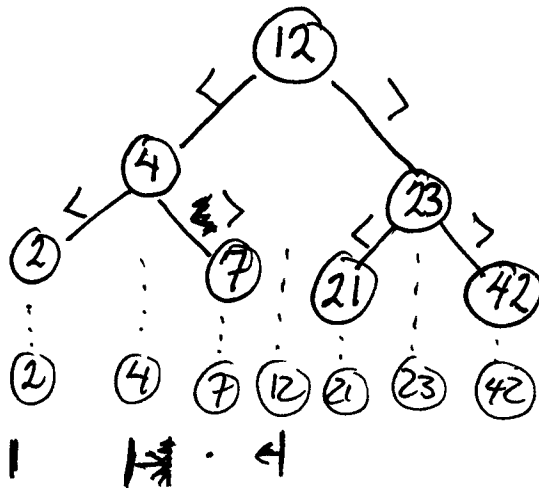
Introd. to Algor., 2<sup>nd</sup> ed  
 Introd. to Algor.  
~~Introduction to Algorithms~~  
 Introduction to Algorithms

- nearest (lexical) match

- hashing doesn't do this; in fact, hash function often chosen so similar keys do not end up in nearby locations  
~~hashing~~ (Uniformity)

- sorted array with binary search will find where element would be in array  $O(\log n)$ , but require  $O(n)$  to build/maintain.

TODAY WE INTRODUCE BINARY SEARCH TREES (BST)



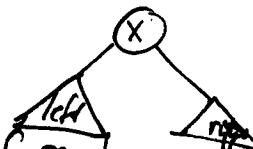
looks like sorted array built into a tree.

Let's imagine searching for the key 7 in this tree and in the sorted array.

Binary Search

BST Property

- ① all nodes in left subtree of  $x$  have  $key \leq key[x]$
- ② all nodes in right subtree of  $x$  have  $key \geq key[x]$



BST traits

- binary tree (at most 2 branches per node, can be nil), left/right, rooted
- each node contains:
  - ①  $key[x]$
  - ②  $left[x]$
  - ③  $right[x]$
  - ④  $parent[x] = p[x]$  ← can be nil
  - ⑤  $Satellite\_data[x]$

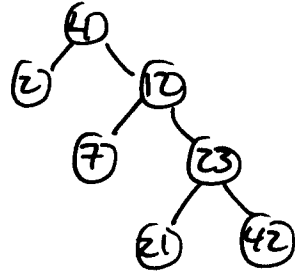
To ~~show~~ see how tree works,  
 Let's ~~the~~ look at how would  
 find sorted array from BST

Inorder [x]

Inorder (left [x]) unless left [x] = nil

Print key [x]

Inorder (right [x]) unless right [x] = nil



look at our example

Recursively find leftmost element and prints. Then prints parent and right, etc.

2 4 7 17 21 23 42

That is, Inorder tree walk visits each node in sorted key order. It really just visits each node once, and through recursive calls keeps "placeholders" throughout the tree.

Time:  $O(n)$  where  $n = \#$  nodes in tree

Now, Imagine using the binary search tree as a data structure for information retrieval

Search (T, k)

```

x ← root(T)
while x ≠ nil and key[x] ≠ k
  do if k < key[x]
     then x ← left[x]
     else x ← right[x]
  ← return x
    
```

Start at root node.  
 If find key, done.  
 Otherwise, use BST property to decide which subsearch to follow. If reach nil, then key does not exist.

Like binary search, but are not guaranteed of subdividing in half each subsearch. If tree is reasonably well balanced (proportional) will still be efficient.

Do example  
 - successful  
 - unsuccessful

This is iterative version

## Analysis of Search

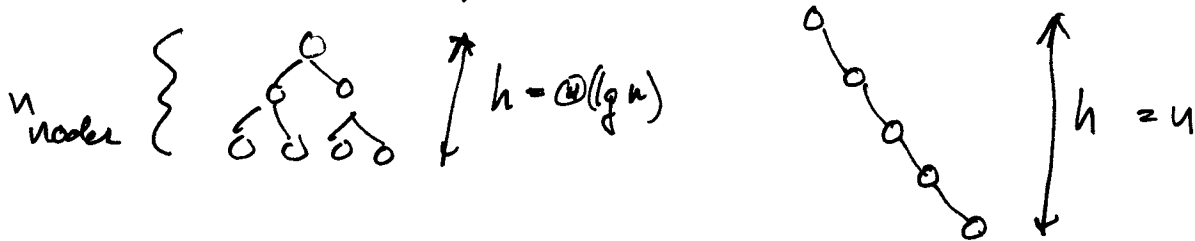
① Fast form of comparison model (like binary search)

② Correctness (idea)

if  $k < \text{key}[x]$ , then by BST property,  $k < \text{all keys}$  in right subtree (and can safely ignore)

③ Running Time

- To successfully find node  $x \rightarrow \Theta(\text{depth}(x))$   
 where  $\text{depth} = \text{distance from root down to } x$
- Worst case  $\rightarrow \Theta(\text{height}(T)) \equiv \Theta(h)$
- Worst case for any tree of  $n$  nodes =  $\Theta(n)$



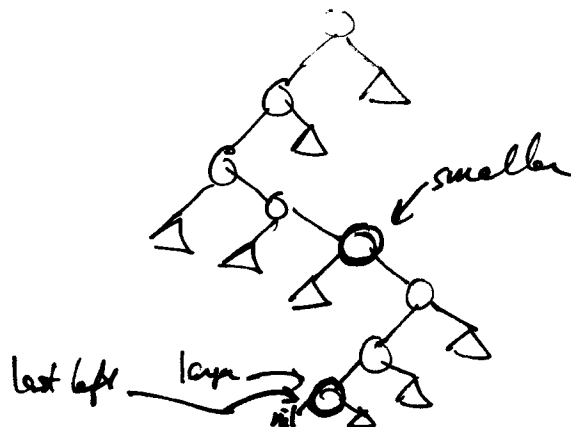
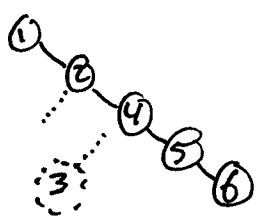
## What about nearest matches?

• Unsuccessful searches identify location where key would fit.

i.e. if best match was left

- next larger element is last node visited
- next smaller element is last true inner node right

(& symmetric statements if best match was right)



## Dynamic BST

Insert (T, k)

- run search until reach nil (pretend  $kft[x] = nil$ )
- $kft[x] \leftarrow$  new node with
  - key = k
  - left = nil
  - right = nil
  - parent = x

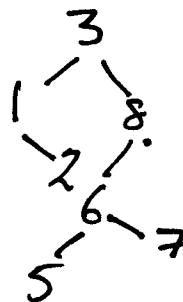
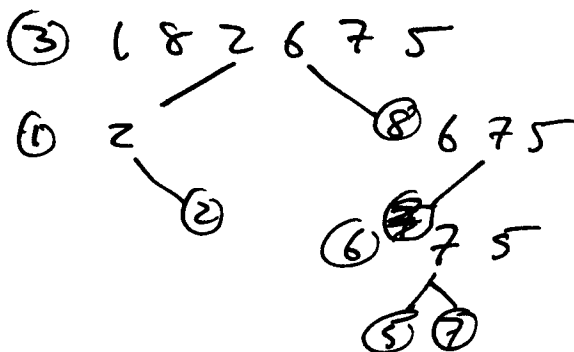
(Example)

Running time  $\Theta(h)$

## BST Sort (A)

for  $i \leftarrow 1$  to  $n$   
 do Insert (A[i])  
 Inorder (root)

Analysis: same comparisons as quicksort, but in different order



Sean  $\Theta(n \lg n)$  time in average case (uniformly random permutations)  
 $\rightarrow$  avg depth of node =  $\Theta(\lg n)$   
 $=$  avg time for insert/search  
 In fact, max depth of node  $\rightarrow \Theta(\lg n)$  in avg case  
~~recitation~~

More BST operations

Minimum (x)

- minimum key in subtree rooted at x

```

while left [x] != nil
  do x ← left [x]
return x
    
```

Time =  $O(h)$

Maximum is simple change

Successor (x)

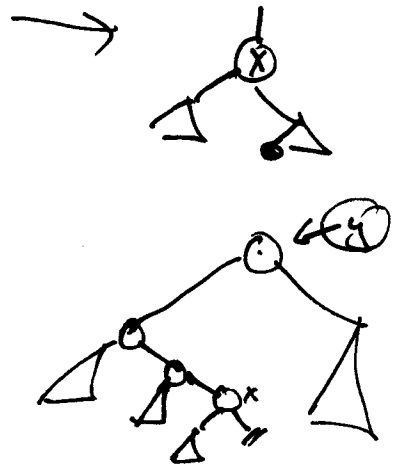
- next higher element in inorder seq

```

if right [x] != nil
  then return minimum (right [x])
else y ← parent [x]
  while y != nil and x = right [y]
    do x ← y
    y ← parent [y]
return y
    
```

Time =  $O(h)$

predecessor is simple change



Delete (x)

- given pointer to a node

① If x has no children (leaf of tree) then remove x

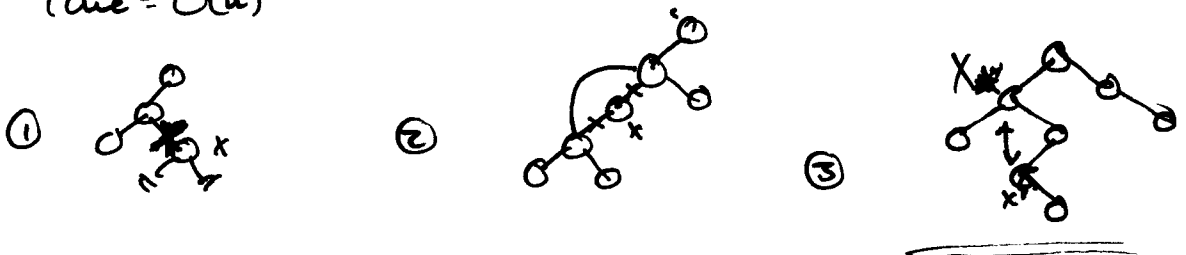
② If x has exactly one child, then splice it out

③ If x has ≥ children, then swap x with successor(x)

- do ① or ② with

(guaranteed to have no children)

Time =  $O(h)$



Problem: Worst Case BST height is  $\Theta(n)$   
 $\Rightarrow$  degenerates to linked list

Solution: Force BST (or related) to stay "balanced"  
 $\Rightarrow$  have  $\Theta(\log n)$  height

Will get all ops in  $\Theta(\log n)$  time

## BST ops

Search  
 Insert  
 Delete  
 Minimum  
 Maximum  
 Successor  
 Predecessor

}  $\Theta(h)$

Started discussion with  $\Theta(n)$  data problems. Now, with balanced tree, are all  $\Theta(\log n)$