

## 4.3 Binary Search Trees

Binary search trees  
Randomized BSTs

Reference: Chapter 12, Algorithms in Java, 3<sup>rd</sup> Edition, Robert Sedgewick.

Robert Sedgewick and Kevin Wayne · Copyright © 2005 · <http://www.Princeton.EDU/~cos226>

Symbol table. Key-value pair abstraction.

- **Insert** a value with specified key.
- **Search** for value given key.
- **Delete** value with given key.

Challenge 1. Guarantee symbol table performance.

hashing analysis depends on input distribution

Challenge 2. Expand API when keys are ordered.

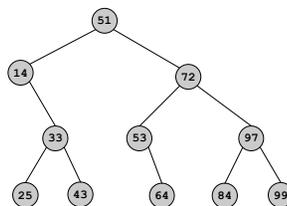
find the kth largest

### Binary Search Trees

Def. A **binary search tree** is a binary tree in symmetric order.

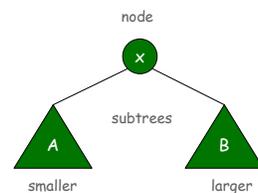
Binary tree is either:

- Empty.
- A key-value pair and two binary trees.



Symmetric order:

- Keys in nodes.
- No smaller than left subtree.
- No larger than right subtree.



### Binary Search Trees in Java

A **BST** is a reference to a node.

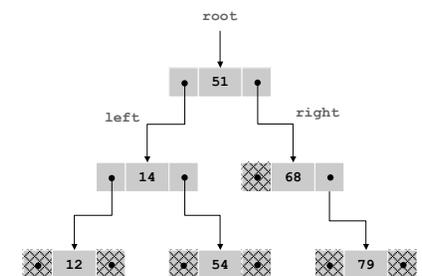
A **Node** is comprised of four fields:

- A key and a value.
- A reference to the left and right subtree.

smaller larger

```
private class Node {
    Key key;
    Val val;
    Node l, r;
}
```

Key and Val are generic types;  
Key is Comparable



```
public class BST<Key extends Comparable, Val> {
    private Node root;

    private class Node {
        private Key key;
        private Val val;
        private Node l, r;

        private Node(Key key, Val val) {
            this.key = key;
            this.val = val;
        }
    }

    private boolean less(Key k1, Key k2) { ... }
    private boolean eq (Key k1, Key k2) { ... }

    public void put(Key key, Val val) { ... }
    public Val get(Key key) { ... }
}
```

Get. Return val corresponding to given key, or null if no such key.

```
public Val get(Key key) {
    Node x = root;
    while (x != null) {
        if (eq(key, x.key)) return x.val;
        else if (less(key, x.key)) x = x.l;
        else x = x.r;
    }
    return null;
}
```

BST: Insert

Put. Associate val with key.

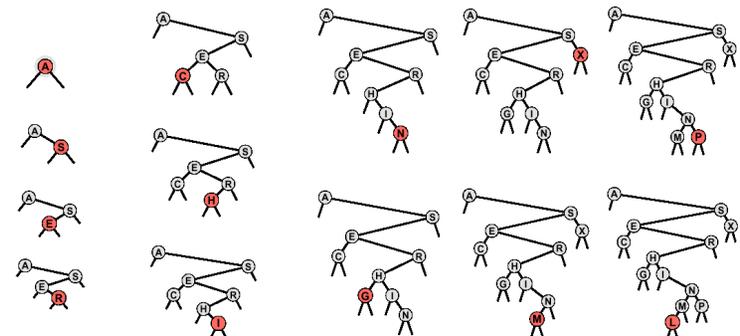
- Search, then insert.
- Concise (but tricky) recursive code.

```
public void put(Key key, Val val) {
    root = insert(root, key, val);
}

private Node insert(Node x, Key key, Val val) {
    if (x == null) return new Node(key, val);
    else if (eq(key, x.key)) x.val = val;
    else if (less(key, x.key)) x.l = insert(x.l, key, val);
    else x.r = insert(x.r, key, val);
    return x;
}
```

BST: Construction

Insert the following keys into BST. A S E R C H I N G X M P L

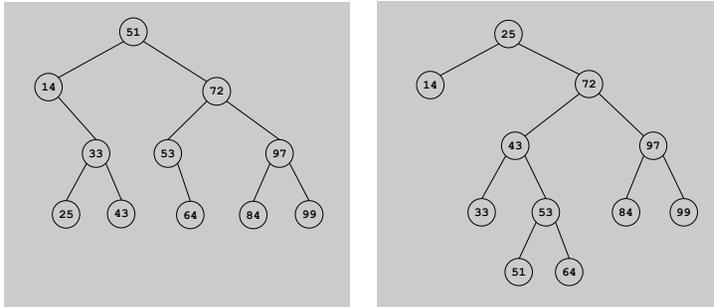


## Tree Shape

### Tree shape.

- Many BSTs correspond to same input data.
- Cost of search/insert proportional to depth of node.
- 1-1 correspondence between BST and quicksort partitioning.

depth of node corresponds to depth of function call stack when node is partitioned



## BST: Analysis

**Theorem.** If keys are inserted in random order, height of tree is  $\Theta(\log N)$ , except with exponentially small probability.

mean =  $4.311 \ln N$ , variance =  $O(1)$

**Property.** If keys are inserted in random order, expected number of comparisons for a search/insert is about  $2 \ln N$ .

**But...** Worst-case for search/insert/height is  $N$ .

e.g., keys inserted in ascending order

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## Symbol Table: Implementations Cost Summary

Implementation	Worst Case			Average Case		
	Get	Put	Remove	Get	Put	Remove
Sorted array	$\log N$	$N$	$N$	$\log N$	$N/2$	$N/2$
Unsorted list	$N$	$N$	$N$	$N/2$	$N$	$N$
Hashing	$N$	1	$N$	1*	1*	1*
BST	$N$	$N$	$N$	$\log N$	$\log N$	???

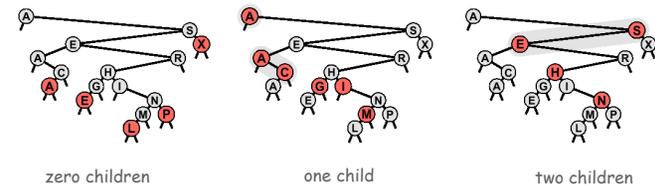
\* assumes hash function is random

**BST.**  $O(\log N)$  insert and search if keys arrive in **random** order.

## BST: Eager Delete

**Delete a node in a BST.** [Hibbard]

- Zero children: just remove it.
- One child: pass the child up.
- Two children: find the next largest node using right-left\* or left-right\*, swap with next largest, remove as above.



**Problem.** Eager deletion strategy clumsy, not symmetric.

**Consequence.** Trees not random (!)  $\Rightarrow \sqrt{N}$  per op.

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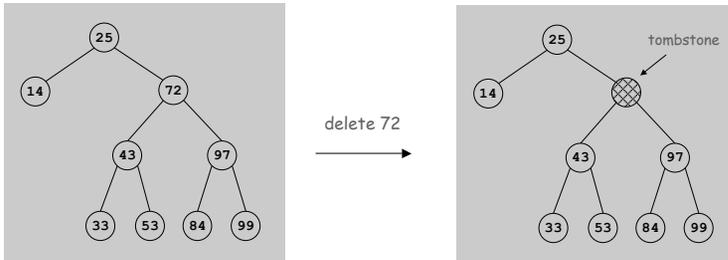
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## BST: Lazy Delete

**Lazy delete.** To delete node with a given key, set its value to `null`.

**Cost.**  $O(\log N')$  per insert, search, and delete, where  $N'$  is the number of elements ever inserted in the BST.

under random input assumption



## Symbol Table: Implementations Cost Summary

Implementation	Worst Case			Average Case		
	Get	Put	Remove	Get	Put	Remove
Sorted array	log N	N	N	log N	N/2	N/2
Unsorted list	N	N	N	N/2	N	N
Hashing	N	1	N	1*	1*	1*
BST	N	N	N	log N †	log N †	log N †

\* assumes hash function is random

† assumes N is number of keys ever inserted

**BST.**  $O(\log N)$  insert and search if keys arrive in **random** order.

**Q.** Can we achieve  $O(\log N)$  independent of input distribution?

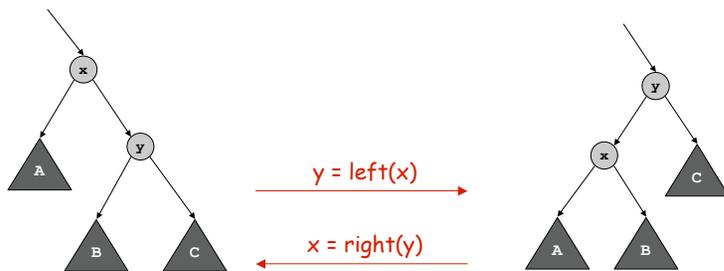
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## Right Rotate, Left Rotate

Two fundamental ops to rearrange nodes in a tree.

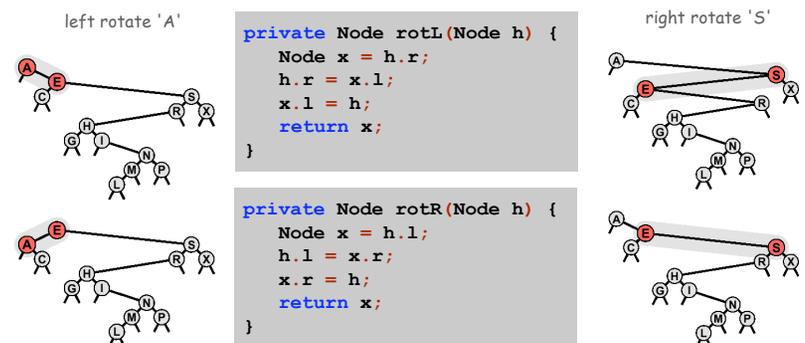
- Maintains symmetric order.
- Local transformations, change just 3 pointers.



## Right Rotate, Left Rotate

**Rotation.** Fundamental operation to rearrange nodes in a tree.

- Easier done than said.



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## Recursive BST Root Insertion

**Root insertion:** insert a node and make it the new root.

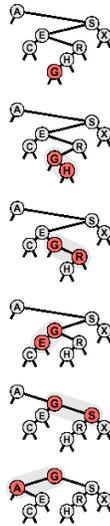
- Insert using standard BST.
- Rotate it up to the root.

**Why bother?**

- Faster if searches are for recently inserted keys.
- Basis for advanced algorithms.

```
private Node rootInsert(Node h, Key key, Val val) {
    if (h == null) return new Node(key, val);
    if (less(key, h.key)) {
        h.l = rootInsert(h.l, key, val);
        h = rotR(h);
    }
    else {
        h.r = rootInsert(h.r, key, val);
        h = rotL(h);
    }
    return h;
}
```

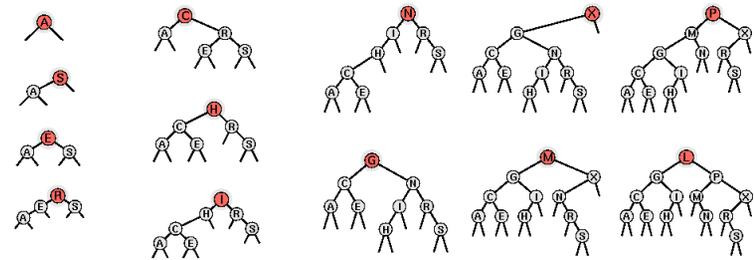
insert G



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## BST Construction: Root Insertion

**Ex. A S E R C H I N G X M P L**



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## Randomized BST

**Intuition.** If keys are inserted in random order, height is logarithmic.

**Idea.** When inserting a new node, make it the root (via root insertion) with probability  $1/(N+1)$ , and do so recursively.

```
private Node insert(Node h, Key key, Val val) {
    if (h == null) return new Node(key, val);
    if (Math.random() * (h.N + 1) < 1)
        return rootInsert(h, key, val);
    else if (less(key, h.key)) h.l = insert(h.l, key, val);
    else h.r = insert(h.r, key, val);
    h.N++;
    return h;
}
```

maintain size of subtree rooted at h

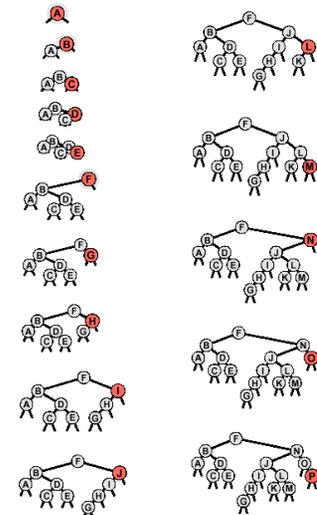
**Fact.** Tree shape distribution is identical to tree shape of inserting keys in random order.

but now, no assumption made on the input distribution

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## Randomized BST Example

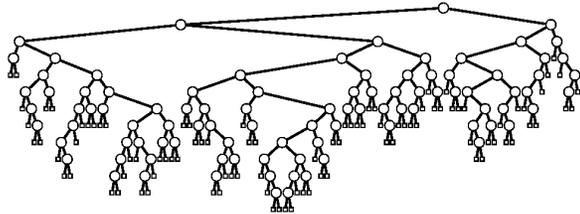
**Ex:** Insert keys in ascending order.



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### Randomized BST

**Property.** Always "looks like" random binary tree.

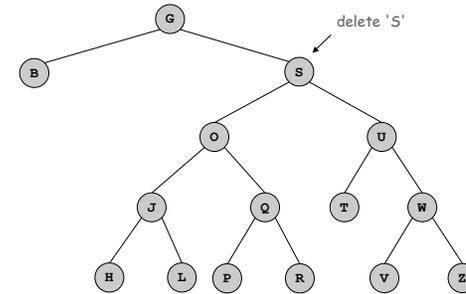


- As before, expected height is  $\Theta(\log N)$ .
- Exponentially small chance of bad balance.

**Implementation.** Need to maintain subtree size in each node.

### Randomized BST: Delete

**Delete.** Delete node containing given key; **join** two broken subtrees.

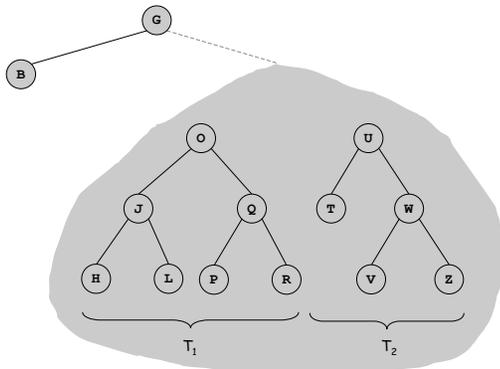


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### Randomized BST: Delete

**Delete.** Delete node containing given key; **join** two broken subtrees.

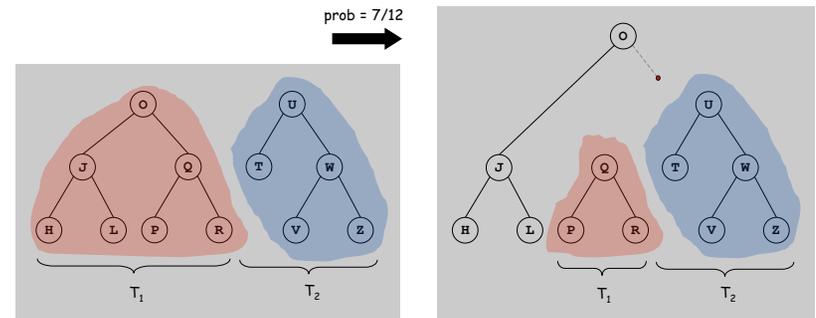


**Goal.** Join  $T_1$  and  $T_2$ , where all keys in  $T_1$  are less than all keys in  $T_2$ .

### Randomized BST: Join

**Join.** Merge  $T_1$  (of size  $N_1$ ) and  $T_2$  (of size  $N_2$ ) assuming all keys in  $T_1$  are less than all keys in  $T_2$ .

- Use root of  $T_1$  as root with probability  $N_1 / (N_1 + N_2)$ , and recursively join right subtree of  $T_1$  with  $T_2$ .
- Use root of  $T_2$  as root with probability  $N_2 / (N_1 + N_2)$ , and recursively join left subtree of  $T_2$  with  $T_1$ .



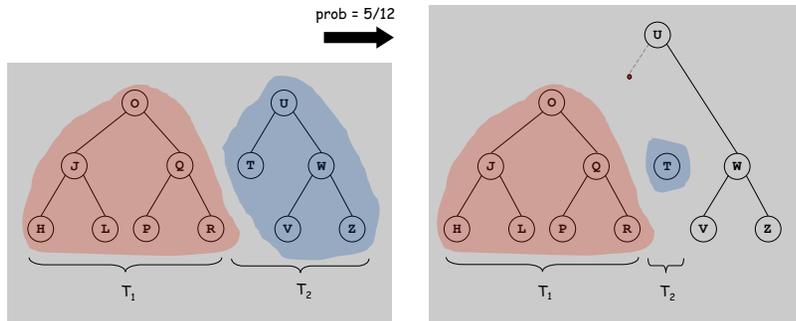
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## Randomized BST: Join

**Join.** Merge  $T_1$  (of size  $N_1$ ) and  $T_2$  (of size  $N_2$ ) assuming all keys in  $T_1$  are less than all keys in  $T_2$ .

- Use root of  $T_1$  as root with probability  $N_1 / (N_1 + N_2)$ , and recursively join right subtree of  $T_1$  with  $T_2$ .
- Use root of  $T_2$  as root with probability  $N_2 / (N_1 + N_2)$ , and recursively join left subtree of  $T_2$  with  $T_1$ .



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## Randomized BST: Delete

**Join.** Merge  $T_1$  (of size  $N_1$ ) and  $T_2$  (of size  $N_2$ ) assuming all keys in  $T_1$  are less than all keys in  $T_2$ .

**Delete.** Delete node containing given key; join two broken subtrees.

**Analysis.** Running time bounded by height of tree.

**Theorem.** Tree still random after delete.

**Corollary.** Expected number of comparisons for a search/insert/delete is  $\Theta(\log N)$ .

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## Symbol Table: Implementations Cost Summary

Implementation	Worst Case			Average Case		
	Search	Insert	Delete	Search	Insert	Delete
Sorted array	$\log N$	$N$	$N$	$\log N$	$N/2$	$N/2$
Unsorted list	$N$	$N$	$N$	$N/2$	$N$	$N$
Hashing	$N$	1	$N$	1*	1*	1*
BST	$N$	$N$	$N$	$\log N^\dagger$	$\log N^\dagger$	$\log N^\dagger$
Randomized BST	$\log N^\ddagger$	$\log N^\ddagger$	$\log N^\ddagger$	$\log N$	$\log N$	$\log N$

\* assumes our hash function can generate random values for all keys  
 † assumes  $N$  is the number of keys ever inserted  
 ‡ assumes system can generate random numbers, randomized guarantee

**Randomized BST.** Guaranteed  $\log N$  performance!  
**Next lecture.** Can we achieve deterministic guarantee?

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## BST: Advanced Operations

**Sort.** Iterate over keys in ascending order.

- Inorder traversal.
- Same comparisons as quicksort, but pay space for extra links.

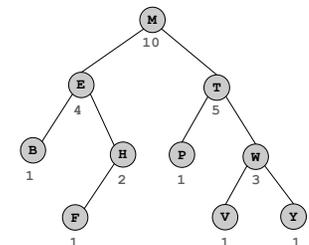
**Range search.** Find all items whose keys are between  $k_1$  and  $k_2$ .

**Find  $k^{\text{th}}$  largest/smallest.** Generalizes PQ.

- Special case: find min, find max.
- Add subtree size to each node.
- Takes time proportional to height of tree.

```
private class Node {
    Key key;
    Val val;
    Node l, r;
    int N;
}
```

↖ subtree size



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## BST: Bin Packing Application

**Ceiling.** Given key  $k$ , return smallest element that is  $\geq k$ .

**Best-fit bin packing heuristic.** Insert the item in the bin with the least remaining space among those that can store the item.

**Theorem.** [D. Johnson] Best-fit decreasing is guaranteed use at most  $11B/9 + 1$  bins, where  $B$  is the best possible.

- Within 22% of best possible.
- Original proof of this result was over 70 pages of analysis!

## Symbol Table: Implementations Cost Summary

Asymptotic Cost

Implementation	Search	Insert	Delete	Find $k^{\text{th}}$	Sort	Join	Ceil
Sorted array	$\log N$	$N$	$N$	$\log N$	$N$	$N$	$\log N$
Unsorted list	$N$	$N$	$N$	$N$	$N \log N$	$N$	$N$
Hashing	$1^*$	$1^*$	$1^*$	$N$	$N \log N$	$N$	$N$
BST	$N$	$N$	$N$	$N$	$N$	$N$	$N$
Randomized BST	$\log N^\ddagger$	$\log N^\ddagger$	$\log N^\ddagger$	$\log N^\ddagger$	$N$	$\log N^\ddagger$	$\log N^\ddagger$

\* assumes our hash function can generate random values for all keys  
‡ assumes system can generate random numbers, randomized guarantee