Week 09: Search Tree Algorithms 1

Searching 1/74

An extremely common application in computing

- given a (large) collection of *items* and a *key* value
- find the item(s) in the collection containing that key
 - item = (key, val₁, val₂, ...) (i.e. a structured data type)
 - key = value used to distinguish items (e.g. student ID)

Applications: Google, databases,

... Searching

Since searching is a very important/frequent operation, many approaches have been developed to do it

Linear structures: arrays, linked lists, files

Arrays = random access. Lists, files = sequential access.

Cost of searching:

	Array	List	File
Unsorted	O(n) (linear scan)	O(n) (linear scan)	O(n) (linear scan)
Sorted	O(log n) (binary search)	O(n) (linear scan)	O(log n) (seek, seek>,)

- O(n) ... linear scan (search technique of last resort)
- $O(\log n)$... binary search, search trees (trees also have other uses)

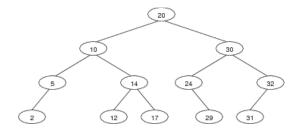
Also (cf. COMP9021): hash tables (O(1), but only under optimal conditions)

... Searching

Maintaining the order in sorted arrays and files is a costly operation.

Search trees are as efficient to search but more efficient to maintain.

Example: the following tree corresponds to the sorted array [2,5,10,12,14,17,20,24,29,30,31,32]:

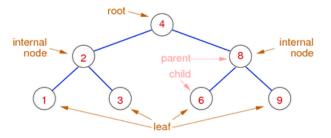


Tree Data Structures

Trees 5/74

Trees are connected graphs

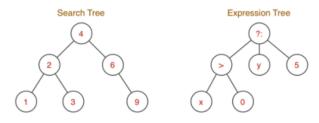
- consisting of nodes and edges (called *links*), with no cycles (no "up-links")
- each node contains a data value (or key+data)
- each node has links to $\leq k$ other child nodes (k=2 below)



... Trees

Trees are used in many contexts, e.g.

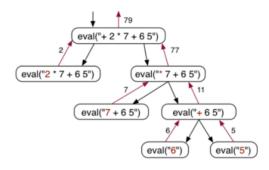
- representing hierarchical data structures (e.g. expressions)
- efficient searching (e.g. sets, symbol tables, ...)



... Trees

Trees can be used as a data structure, but also for illustration.

E.g. showing evaluation of a prefix arithmetic expression

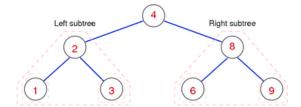


... Trees

Binary trees (k=2 children per node) can be defined recursively, as follows:

A binary tree is either

- empty (contains no nodes)
- consists of a *node*, with two *subtrees*
 - node contains a value
 - left and right subtrees are binary trees



... Trees 9/74

Other special kinds of tree

• *m-ary tree*: each internal node has exactly *m* children

• Ordered tree: all left values < root, all right values > root

• Balanced tree: has ≅minimal height for a given number of nodes

• Degenerate tree: has \(\sigma \) maximal height for a given number of nodes

Search Trees

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Binary Search Trees

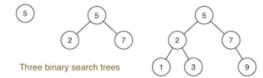
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Binary search trees (or BSTs) have the characteristic properties

- each node is the root of 0, 1 or 2 subtrees
- all values in any left subtree are less than root
- all values in any right subtree are greater than root
- these properties applies over all nodes in the tree

(perfectly) balanced trees have the properties

- #nodes in left subtree = #nodes in right subtree
- this property applies over all nodes in the tree



... Binary Search Trees

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Operations on BSTs:

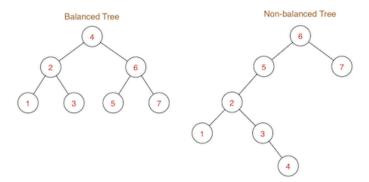
- insert(Tree,Item) ... add new item to tree via key
- delete(Tree,Key) ... remove item with specified key from tree
- search(Tree,Key) ... find item containing key in tree
- plus, "bookkeeping" ... new(), free(), show(), ...

Notes:

- nodes contain Items; we just show Item.key
- keys are unique (not technically necessary)

... Binary Search Trees

Examples of binary search trees:

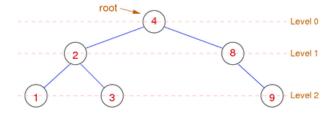


Shape of tree is determined by order of insertion.

... Binary Search Trees

Level of node = path length from root to node

Height (or: *depth*) of tree = max path length from root to leaf



Height-balanced tree: ∀ nodes: height(left subtree) = height(right subtree)

Time complexity of tree algorithms is typically *O*(*height*)

Exercise #1: Insertion into BSTs

For each of the sequences below

- start from an initially empty binary search tree
- show tree resulting from inserting values in order given
- (a) 4 2 6 5 1 7 3
- (b) 6 5 2 3 4 7 1
- (c) 1 2 3 4 5 6 7

Assume new values are always inserted as new leaf nodes

- (a) the balanced tree from 3 slides ago (height = 2)
- (b) the non-balanced tree from 3 slides ago (height = 4)
- (c) a fully degenerate tree of height 6

Representing BSTs

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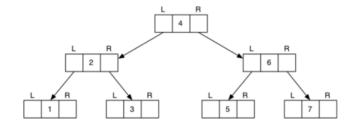
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Binary trees are typically represented by node structures

• containing a value, and pointers to child nodes

Most tree algorithms move *down* the tree.

If upward movement needed, add a pointer to parent.



... Representing BSTs

Typical data structures for trees ...

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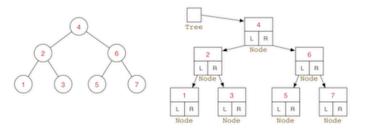
```
// a Tree is represented by a pointer to its root node
```

```
typedef struct Node *Tree;
// a Node contains its data, plus left and right subtrees
typedef struct Node {
  int data;
  Tree left, right;
} Node;
// some macros that we will use frequently
#define data(tree) ((tree)->data)
#define left(tree) ((tree)->left)
#define right(tree) ((tree)->right)
```

We ignore items \Rightarrow data in Node is just a key

... Representing BSTs

Abstract data vs concrete data ...



Tree Algorithms

Searching in BSTs

Most tree algorithms are best described recursively:

```
TreeSearch(tree,item):
   Input tree, item
   Output true if item found in tree, false otherwise
   if tree is empty then
      return false
   else if item < data(tree) then</pre>
      return TreeSearch(left(tree),item)
   else if item > data(tree) then
      return TreeSearch(right(tree),item)
                // found
   else
     return true
   end if
```

Insertion into BSTs

Insert an item into appropriate subtree:

```
insertAtLeaf(tree,item):
   Input tree, item
   Output tree with item inserted
   if tree is empty then
      return new node containing item
   else if item < data(tree) then</pre>
      return insertAtLeaf(left(tree),item)
   else if item > data(tree) then
      return insertAtLeaf(right(tree),item)
```

```
else
                  // avoid duplicates
   return tree
end if
```

23/74 **Tree Traversal**

Iteration (traversal) on ...

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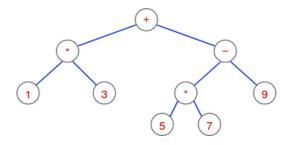
- Lists ... visit each value, from first to last
- Graphs ... visit each vertex, order determined by DFS/BFS/...

For binary Trees, several well-defined visiting orders exist:

- preorder (NLR) ... visit root, then left subtree, then right subtree
- inorder (LNR) ... visit left subtree, then root, then right subtree
- postorder (LRN) ... visit left subtree, then right subtree, then root
- level-order ... visit root, then all its children, then all their children

24/74 ... Tree Traversal

Consider "visiting" an expression tree like:



NLR: + * 1 3 - * 5 7 9 (prefix-order: useful for building tree)

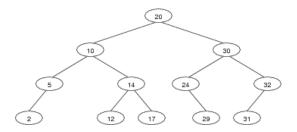
LNR: 1*3+5*7-9 (infix-order: "natural" order)

LRN: 13 * 57 * 9 - + (postfix-order: useful for evaluation)

Level: +* - 1 3 * 9 5 7 (level-order: useful for printing tree)

Exercise #2: Tree Traversal

Show NLR, LNR, LRN traversals for the following tree:



```
NLR (preorder): 20 10 5 2 14 12 17 30 24 29 32 31 LNR (inorder): 2 5 10 12 14 17 20 24 29 30 31 32 LRN (postorder): 2 5 12 17 14 10 29 24 31 32 30 20
```

Exercise #3: Non-recursive traversals

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Write a non-recursive *preorder* traversal algorithm.

Assume that you have a stack ADT available.

Joining Two Trees

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An auxiliary tree operation ...

Tree operations so far have involved just one tree.

An operation on two trees: $t = joinTrees(t_1, t_2)$

- Pre-conditions:
 - takes two BSTs; returns a single BST
 - max(key(t₁)) < min(key(t₂))
- Post-conditions:
 - o result is a BST (i.e. fully ordered)
 - o containing all items from t₁ and t₂

... Joining Two Trees

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Method for performing tree-join:

- find the min node in the right subtree (t₂)
- replace min node by its right subtree
- elevate min node to be new root of both trees

Advantage: doesn't increase height of tree significantly

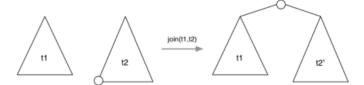
```
x \le height(t) \le x+1, where x = max(height(t_1), height(t_2))
```

Variation: choose deeper subtree; take root from there.

... Joining Two Trees

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Joining two trees:



Note: t2' may be less deep than t2

... Joining Two Trees

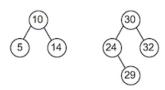
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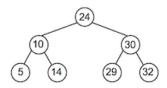
Implementation of tree-join:

```
joinTrees(t_1, t_2):
   Input trees t<sub>1</sub>,t<sub>2</sub>
   Output t<sub>1</sub> and t<sub>2</sub> joined together
   if t<sub>1</sub> is empty then return t<sub>2</sub>
   else if t_2 is empty then return t_1
       curr=t2, parent=NULL
      while left(curr) is not empty do
                                                    // find min element in t2
          parent=curr
          curr=left(curr)
       end while
       if parent≠NULL then
          left(parent)=right(curr) // unlink min element from parent
          right(curr)=t<sub>2</sub>
       end if
       left(curr)=t1
       return curr
                                          // curr is new root
   end if
```

Exercise #4: Joining Two Trees

Join the trees





Deletion from BSTs

Insertion into a binary search tree is easy.

Deletion from a binary search tree is harder.

Four cases to consider ...

- empty tree ... new tree is also empty
- zero subtrees ... unlink node from parent
- one subtree ... replace by child
- two subtrees ... replace by successor, join two subtrees

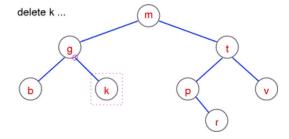
... Deletion from BSTs

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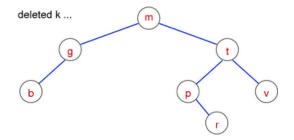
Case 2: item to be deleted is a leaf (zero subtrees)



Just delete the item

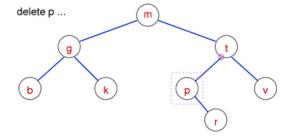
... Deletion from BSTs 37/74

Case 2: item to be deleted is a leaf (zero subtrees)



... Deletion from BSTs 38/74

Case 3: item to be deleted has one subtree

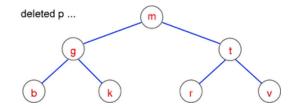


Replace the item by its only subtree

... Deletion from BSTs

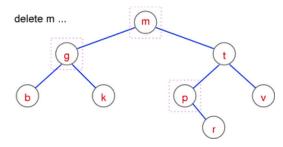
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Case 3: item to be deleted has one subtree



... Deletion from BSTs 40/74

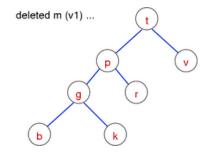
Case 4: item to be deleted has two subtrees



Version 1: right child becomes new root, attach left subtree to min element of right subtree

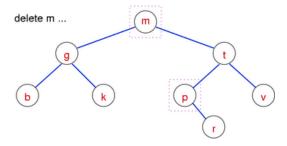
... Deletion from BSTs 41/74

Case 4: item to be deleted has two subtrees



... Deletion from BSTs 42/74

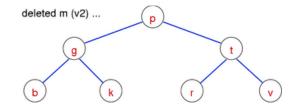
Case 4: item to be deleted has two subtrees.



Version 2: join left and right subtree

... Deletion from BSTs 43/74

Case 4: item to be deleted has two subtrees



... Deletion from BSTs 44/74

Pseudocode (version 2):

```
TreeDelete(t,item):
  Input tree t, item
  Output t with item deleted
  if t is not empty then
                                   // nothing to do if tree is empty
      if item < data(t) then</pre>
                                   // delete item in left subtree
         left(t)=TreeDelete(left(t),item)
      else if item > data(t) then // delete item in left subtree
         right(t)=TreeDelete(right(t),item)
                                   // node 't' must be deleted
      else
        if left(t) and right(t) are empty then
                                             // 0 children
            new=empty tree
        else if left(t) is empty then
           new=right(t)
                                             // 1 child
        else if right(t) is empty then
                                             // 1 child
            new=left(t)
        else
            new=joinTrees(left(t),right(t)) // 2 children
        end if
        free memory allocated for t
        t=new
     end if
  end if
  return t
```

Balanced BSTs

Balanced Binary Search Trees

Goal: build binary search trees which have

• minimum height ⇒ minimum worst case search cost

Perfectly balanced tree with N nodes has

- abs(#nodes(LeftSubtree) #nodes(RightSubtree)) < 2, for every node
- height of $log_2N \Rightarrow$ worst case search O(log N)

Three *strategies* to improving worst case search in BSTs:

- randomise reduce chance of worst-case scenario occuring
- amortise do more work at insertion to make search faster
- optimise implement all operations with performance bounds

Operations for Rebalancing

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To assist with rebalancing, we consider new operations:

Left rotation

• move right child to root; rearrange links to retain order

Right rotation

• move left child to root; rearrange links to retain order

Insertion at root

• each new item is added as the new root node

Tree Rotation

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In tree below: $t_1 < n_2 < t_2 < n_1 < t_3$



... Tree Rotation 49/74

Method for rotating tree T right:

- N₁ is current root; N₂ is root of N₁'s left subtree
- N₁ gets new left subtree, which is N₂'s right subtree
- N₁ becomes root of N₂'s new right subtree
- N₂ becomes new root

Left rotation: swap left/right in the above.

Cost of tree rotation: O(1)

... Tree Rotation 50/74

Algorithm for right rotation:

```
rotateRight(n<sub>1</sub>):
    Input tree n<sub>1</sub>
    Output n<sub>1</sub> rotated to the right

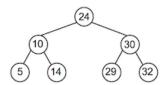
if n<sub>1</sub> is empty or left(n<sub>1</sub>) is empty then
    return n<sub>1</sub>
end if
    n<sub>2</sub>=left(n<sub>1</sub>)
    left(n<sub>1</sub>)=right(n<sub>2</sub>)
    right(n<sub>2</sub>)=n<sub>1</sub>
    return n<sub>2</sub>
```



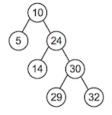
Exercise #5: Tree Rotation

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Consider the tree t:



Show the result of rotateRight(t)



Exercise #6: Tree Rotation

Write the algorithm for left rotation



```
rotateLeft(n<sub>2</sub>):
    Input tree n<sub>2</sub>
   {f Output} {\bf n_2} rotated to the left
    if n_2 is empty or right(n_2) is empty then
        return n<sub>2</sub>
    end if
    n_1 = right(n_2)
   right(n_2) = left(n_1)
    left(n_1)=n_2
    return n<sub>1</sub>
```

55/74 **Insertion at Root**

Previous description of BSTs inserted at leaves.

Different approach: insert new item at root.

Potential disadvantages:

• large-scale rearrangement of tree for each insert

Potential advantages:

- recently-inserted items are close to root
- low cost if recent items more likely to be searched

... Insertion at Root

Method for inserting at root:

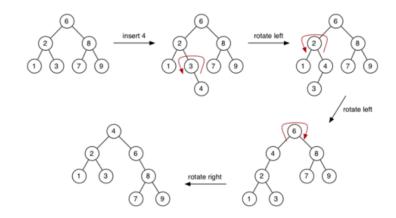
- base case:
 - o tree is empty; make new node and make it root
- recursive case:

- insert new node as root of appropriate subtree
- lift new node to root by rotation

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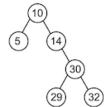
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... Insertion at Root



Exercise #7: Insertion at Root

Consider the tree t:



Show the result of insertAtRoot(t,24)



... Insertion at Root

Analysis of insertion-at-root:

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- same complexity as for insertion-at-leaf: *O(height)*
- tendency to be balanced, but no balance guarantee
- benefit comes in searching
 - o for some applications, search favours recently-added items
 - o insertion-at-root ensures these are close to root
- could even consider "move to root when found"
 - effectively provides "self-tuning" search tree

Rebalancing Trees

Reparationing frees

An approach to balanced trees:

- insert into leaves as for simple BST
- periodically, rebalance the tree

Question: how frequently/when/how to rebalance?

```
NewTreeInsert(tree,item):
    Input    tree, item
    Output    tree with item randomly inserted
    t=insertAtLeaf(tree,item)
    if #nodes(t) mod k = 0 then
         t=rebalance(t)
    end if
    return t
```

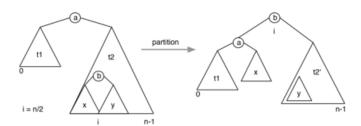
E.g. rebalance after every 20 insertions \Rightarrow choose k=20

Note: To do this efficiently we would need to change tree data structure and basic operations:

```
typedef struct Node {
  int data;
  int nnodes;  // #nodes in my tree
  Tree left, right; // subtrees
} Node;
```

... Rebalancing Trees

How to rebalance a BST? Move median item to root.



... Rebalancing Trees 63/74

Implementation of rebalance:

... Rebalancing Trees

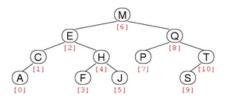
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New operation on trees:

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• partition(tree, i): re-arrange tree so that element with index i becomes root

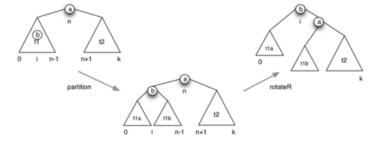


For tree with N nodes, indices are 0 ... N-1

... Rebalancing Trees

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Partition: moves i th node to root



... Rebalancing Trees 66/74

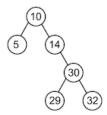
Implementation of partition operation:

```
partition(tree,i):
    Input tree with n nodes, index i
    Output tree with i<sup>th</sup> item moved to the root
    m=#nodes(left(tree))
    if i < m then
        left(tree)=partition(left(tree),i)
        tree=rotateRight(tree)
    else if i > m then
        right(tree)=partition(right(tree),i-m-1)
        tree=rotateLeft(tree)
    end if
    return tree
```

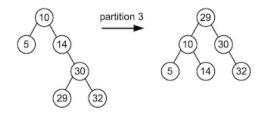
Note: size(tree) = n, size(left(tree)) = m, size(right(tree)) = n-m-1 (why -1?)

Exercise #8: Partition

Consider the tree ±:



Show the result of partition (t,3)



... Rebalancing Trees

Analysis of rebalancing: visits every node $\Rightarrow O(N)$

Cost means not feasible to rebalance after each insertion.

When to rebalance? ... Some possibilities:

- after every *k* insertions
- whenever "imbalance" exceeds threshold

Either way, we tolerate worse search performance for periods of time.

Does it solve the problem? ... Not completely ⇒ Solution: real balanced trees (next week)

Application of BSTs: Sets

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Trees provide efficient search.

Sets require efficient search

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- to find where to insert/delete
- to test for set membership

Logical to implement a set ADT via BSTree

... Application of BSTs: Sets

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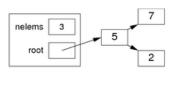
Assuming we have Tree implementation

- which precludes duplicate key values
- which implements insertion, search, deletion

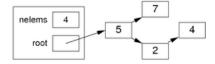
then Set implementation is

- SetInsert(Set, Item) = TreeInsert(Tree, Item)
- SetDelete(Set, Item) = TreeDelete(Tree, Item.Key)
- SetMember(Set, Item) = TreeSearch(Tree, Item.Key)

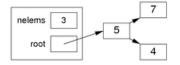
... Application of BSTs: Sets



After SetInsert(s,4):



After SetDelete(s,2):



... Application of BSTs: Sets

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Concrete representation:

```
#include <BSTree.h>

typedef struct SetRep {
   int nelems;
   Tree root;
} SetRep;

Set newSet() {
   Set S = malloc(sizeof(SetRep));
   assert(S != NULL);
   S->nelems = 0;
   S->root = newTree();
   return S;
}
```

Summary

- Binary search tree (BST) data structure
- BST insertion and deletion
- Other tree operations
 - tree rotation
 - tree partition
 - joining trees

- Suggested reading:
 - o Sedgewick, Ch.12.5-12.6,12.8-12.9

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