# Binary Search Trees CMSC 420: Lecture 6



How much space is used?

# **Threaded Trees**

- Traversals:
  - Require extra memory, and
  - Must be started from the root
- Use NULL pointers to store in-order predecessors and successors.
- Extra bit associated with each pointer marks whether it is a thread.



## **Threaded Trees**

```
void inorder_succ(BinNode *T)
{
    BinNode * next = T->right();
    if(!next) return NULL;
    if(!is_thread(T->right))
    {
        while(next->left() &&
            is_thread(next->left)
            ) next = next->left();
    }
```





return next;

}

```
In general, in order
successor = leftmost
item in right subtree
```

# **Using Threads for Preorder**

preorder\_succ(H) = right child of the lowest ancestor of H that both has H in its left subtree and has a right child.



```
void preorder_succ(BinNode *T)
{
    if(T->left() &&
        !is_thread(T->left) return T->left();
    for(BinNode* next = T->right();
        is_thread(next->right); Walk up right
        next = next->right()) {} threads
```

```
return RC(P); Return right child
}
```

# **Serializing Trees**

- Often want to write trees out to disk in a space efficient way.
- Preorder traversal will let you store the nodes.
  - What's the preorder traversal of this tree?
- Need to encode the structure somehow.



# **Serializing Trees**

 In preorder traversal, output a mark when you finish processing a node's children.



• Leaves = empty children lists:  $\emptyset$ 

 $\begin{array}{cccc} A & B & E & \varnothing \end{array} \\ & & e & b \end{array} \begin{array}{c} F & \varnothing \end{array} \begin{array}{c} G & \emptyset \end{array} \begin{array}{c} G & \emptyset \end{array} \begin{array}{c} H & \emptyset \end{array} \begin{array}{c} O & 0 \\ h & d \end{array} \begin{array}{c} C & \emptyset \end{array} \begin{array}{c} O \\ c & a \end{array}$ 

- Ø symbols are redundant:
  - ABE))C)D)F)G)H))
- ")" means "go up one level".

# **Binary Search Trees**

- BST Property: If a node has key k then keys in the left subtree are < k and keys in the right subtree are > k.
- We disallow duplicate keys.
- Generalization of the binary search process we saw before:
  - ordering
  - partitioning
  - linking
- Good for implementing the *dictionary ADT* we've already seen: insert, delete, find.



# **Sorted Set Problem**

- If keys are *totally ordered*, dictionary ADT is sometimes extended to a "sorted set ADT."
  - Totally ordered means: for every pair (a,b) either a < b or b < a).</li>
  - Operations:
    - s = make\_sorted\_set
    - find(s, k)
    - insert(s, k)
- *Dictionary operations*

- delete(s, k)
- *join*(s<sub>1</sub>, k, s<sub>2</sub>): make a new sorted set from s<sub>1</sub>, {k}, s<sub>2</sub>; destroy s<sub>1</sub> and s<sub>2</sub>. Assumes every item in s<sub>1</sub> < k and k < every time in s<sub>1</sub>.
- *split*(*s*, *k*): return 3 new sorted sets: *s*<sub>1</sub> with items < *k*, {*k*}, and *s*<sub>2</sub> with items > *k*.

#### **BST Find**

Find k = 6: Is k < 5? No, go right Is k < 8? Yes, go left



#### **BST Find**

Find k = 9:5Is k < 5?No, go rightIs k < 8?No, go rightIs k < 11?Yes, go left246

#### **BST Find**



#### **BST Insert**

```
insert(T, K):
                                         Same idea as BST Find
   q = NULL
   p = T
   while p != NULL and p.key != K:
      q = p
      if p.key < K:</pre>
       p = p.left
      else if p.key > K:
                                                             q
        p = p.right
                                        3
                                                      8
   if p != NULL: error DUPLICATE
   N = new Node(K)
   if q.data > K:
                                    2
                                          4
     q.left = N
   else:
     q.right = N
                                                               NULL
                                                        9
```

#### **BST Insert with Extended Binary Trees**



#### **BST FindMin**



#### Walk left until you can't go left any more

Can you express inorder\_successor using find\_min?





**Python Implementation of BST** 

# *How would you implement join and split?*



• What's the worst possible insertion order?

• What's the best possible insertion order?

- Suppose the keys  $k_1$ ,  $k_2$ ,  $k_3$ , ...,  $k_n$  are inserted in a random order (every permutation equally likely).
- What is the expected path length to a node in the BST built by inserting these keys?
- **Idea**: consider the leftmost path as a representative path.
- New key  $k_i$  added to left-most path when it is the smallest encountered so far ( $k_i < k_j$  for j < i).
- In a random permutation, how often does the minimum change?
   [This is the length of the leftmost path]

 $k_1, k_2, k_3, k_4, k_5, k_6, k_7, k_8$ 

- What's the probability that  $k_i$  is the smallest so far?
- $\Pr[k_i \text{ is smallest among } k_1, \dots, k_i] = 1/i$
- Why?
- In a random permutation of  $k_1, \dots, k_i$ , the minimum is equally likely to be in any one of the *i* positions.
- Probability it is in the last position = 1/i.

keys =  $k_1$ ,  $k_2$ ,  $k_3$ ,  $k_4$ ,  $k_5$ ,  $k_6$ ,  $k_7$ ,  $k_8$ random variables =  $x_1$ ,  $x_2$ ,  $x_3$ ,  $x_4$ ,  $x_5$ ,  $x_6$ ,  $x_7$ ,  $x_8$ 

> $x_i = 1$  if  $k_i$  is smallest among  $k_1, \dots, k_i$ 0 otherwise

• sum of  $x_i$  = length of leftmost path.

• Expected length = 
$$E[\sum x_i] = \sum E[x_i]$$
  
=  $\sum [(1/i)1 + 0(1-1/i)]$   
=  $\sum (1/i) = H_n = O(\log n)$ 



Insertion order: 9 3 10 6 1 8 5



Left chain has 3 nodes

Minimum changes 3 times

Insertion order: 8 9 5 10 3 6 1





Left chain has 4 nodes

Minimum changes 4 times

#### **Optimal Static BSTs – Cost of trees**

 $k_1$ ,  $k_2$ ,  $k_3$ ,  $k_4$ ,  $k_5$ ,  $k_6$ ,  $k_7$ ,  $k_8$  $p_1$ ,  $p_2$ ,  $p_3$ ,  $p_4$ ,  $p_5$ ,  $p_6$ ,  $p_7$ ,  $p_8$ 

• Define the cost of a tree built on keys  $k_j, ..., k_m$ :

$$C(T) = \sum_{i=j}^{m} p_i(Depth(T, k_i) + 1)$$
Why is it
Depth + 1?

- T is **optimal** if C(T) is smallest among any possible T containing the same keys.
- C(T) = expected cost of searching for a key in T.

#### Subtrees of optimal tree are optimal trees

- **Goal**: find tree that minimizes C(T).
- **Claim**: every subtree of optimal tree is optimal.
- **Proof:** Let T be an optimal tree on  $k_j, ..., k_m$ , with root =  $k_r$  (j ≤ r ≤ m)

$$C(T) = p_{r} + \sum_{i=j}^{r-1} p_{i}(\text{Depth}(T, k_{i}) + 1) + \sum_{i=r+1}^{m} p_{i}(\text{Depth}(T, k_{i}) + 1)$$

$$\sum_{i=j}^{r-1} p_{i} + \sum_{i=j}^{r-1} p_{i}\text{Depth}(T, k_{i}) = \sum_{i=j}^{r-1} p_{i} + \sum_{i=j}^{r-1} p_{i}\text{Depth}(T, k_{i})$$

$$C(T_{left}) = C(T_{right})$$

$$C(T) = \sum_{i=j}^{m} p_i + C(T_{left}) + C(T_{right})$$

So,

$$C(T) = \sum_{i=j}^{m} p_i + C(T_{left}) + C(T_{right})$$

- If there were a lower cost T<sub>left</sub> or T<sub>right</sub> we could reduce the total cost of T, contradicting that T is optimal.
- Hence, T<sub>left</sub> and T<sub>right</sub> must be optimal.

## **Dynamic Programming to Find OPT Tree**

$$C[j, m] = \begin{cases} 0 & \text{if } m < j \text{ (tree is empty)} \\ p_j & \text{if } j = m \text{ (tree is single node)} \\ \sum_{i=j}^{m} p_i + \min_{r} \{ C[j, r-1] + C[r+1, m] \} & \text{if } j < m \end{cases}$$

$$k_{2}, k_{3}, k_{4}, k_{5}, k_{6}, k_{7}, k_{8}$$
  
 $\uparrow \uparrow \qquad m$ 

So: if you fill in the C[j, m] table from in order of increasing *m*-*j*, you'll always have the value of C[j,m] computed when you need it.

Dynamic Programming The chosen values of *r* partition the nodes and give you the optimal tree structure.

