# CMSC 420 - 0201 - Fall 2019 Lecture 12

Extended and Scapegoat Trees



- In today's lecture, we will discuss two unrelated topics that arise in the programming assignment:
  - Extended Binary Search Trees
  - Scapegoat Trees
- We will also discuss the SG Tree, which is featured in the Programming Project, Part 1

- Extended Binary Tree (from Lecture 3)
  - Internal nodes: Have exactly 2 children
  - External nodes: Have 0 children
- Basic properties
  - Any extended binary tree with n internal nodes has n + 1 leaves



- Extended Binary Search Trees
  - Each external node contains an entry, a key-value pair, (x, v)

S

- Each internal node contains a splitter, s
  - $If x \le s \rightarrow Left subtree$
  - $\text{ If } x > s \rightarrow \text{ Right subtree}$
- Note that a key can be both a splitter and part of a key-value pair

< s



Contents: {2, 6, 7, 9, 11, 14, 17}

Why?

- Memory locality: We saw with B+ trees, we can store many splitters in a single node, increasing fan-out, thus decreasing tree height
- Heterogenous data: In some applications the data and splitters are different
  - Example: Binary-space partition tree
    - Data are points
    - Splitters are lines



Differences with standard (unbalanced) binary search trees

- find(x):
  - Descend to the external node, as directed by internal nodes
  - If key matches then found, else not
  - Warning: Matching a splitter means nothing!
- Example:
  - find(7) yes
  - find(15) no
  - find(10) no! (even though root matches)



Differences with standard (unbalanced) binary search trees

- insert(x,v):
  - Descend to the external node. Let y be its key. If x = y duplicate-key error
  - Create a new external node for x and internal node to split between x and y
  - Splitter *s* satisfies:  $min(x, y) \le s < max(x, y)$



CMSC 420 – Dave Mount

Differences with standard (unbalanced) binary search trees

- delete(x):
  - Descend to the external node. Let y be its key. If  $x \neq y$  key-not-found error
  - Replace this node and its parent with its sibling



Another Amortized Dictionary Data Structure

- Amortized cost -
  - The total cost divided by the number of operations
  - Splay trees Amortized cost  $O(\log n)$  for dictionary operations, even though any single operation may take O(n) time
- Are there other efficient dictionaries in the amortized sense? Scapegoat trees!
- Origins:
  - Original idea by Arne Andersson (of AA-Tree fame), 1989
  - Rediscovered by Galperin and Rivest, 1993 (gave the name "Scapegoat Tree")
- Resources:
  - <u>http://opendatastructures.org/versions/edition-0.1g/ods-python/8\_Scapegoat\_Trees.html</u>
  - <u>http://opendatastructures.org/newhtml/ods/latex/scapegoat.html</u>

Another Amortized Dictionary Data Structure

- Why should we care?
  - Amortized structures are often simpler than worst-case efficient structures
  - The update rules for scapegoat trees can be adapted to many other search trees where rotations cannot be applied (e.g., spatial decomposition trees)
  - The SG Tree in the programming assignment is a variant of the scapegoat tree



Overview - Balance through Rebuilding

#### Insertion:

- Insert just as in a standard (unbalanced) binary tree
- Monitor the depth of the inserted node after each insertion
- If it is too high, then there must be at least one node on the search path that has poor weight balance (left and right children have very different sizes)
- Find such a node it's the scapegoat! (It is given the blame for the high depth)
- Rebuild the subtree rooted at this node so that it is perfectly balanced
- Deletion:
  - Delete as in a standard (unbalanced) binary tree
  - Once the number of deletions is sufficiently large relative to the entire tree size, rebuild the entire tree so it is perfectly balanced

Overview - Balance through Rebuilding

- How to rebuild a subtree?
  - Perform an inorder traversal of the subtree, and copy the n elements to a (sorted) array  $A[0 \dots n-1]$
  - Take the median of the array as the root, and recursively build left and right subtrees from the two halves of the array
- buildSubtree(A,i,k): Build subtree for k-element subarray A[i ... i + k 1]
  - If k = 0, return null
  - Otherwise, let  $m = \left|\frac{k}{2}\right|$ . Create new node p with median key, A[i + m]
  - p.left = buildSubtree(A,i,m)
  - p.right = buildSubtree(A,i+m+1,k-m-1)
- A subtree with n nodes can be rebuilt in O(n) time

**Overview** - **Details** 

- A scapegoat tree stores no balance or height information with the nodes
- In addition to the tree we maintain:
  - n the current number of nodes in the tree
  - -m an upper bound on the tree size (we maintain:  $n \le m \le 2n$ )
- Height condition: never exceeds  $\log_{3/2} m$  ( $\Rightarrow$ Tree height is  $O(\log n)$ )
- Size condition:
  - Initially: n = m = 0
  - After insertion: n++, m++
  - After deletion: n - (but do not change m)
  - If 2n < m, rebuild the entire tree, and set m = n

Overview - More Details

- find(x):
  - Identical to any binary search time (time:  $O(\log n)$ )
- delete(x):
  - Identical to delete for an unbalanced binary tree
  - Decrement n (but do not change m)
  - If 2n < m, rebuild the entire tree, and set m = n

Overview - More Details

- insert(x):
  - Same as standard binary search tree insertion, keep track of inserted node's depth (number of edges from the root)
  - Increment both n and m
  - If inserted depth exceeds  $\log_{3/2} m$ :
    - Walk back up the search path until we find the first node u such that

$$\frac{\text{size}(u, \text{child})}{\text{size}(u)} > \frac{2}{3}$$

-Here size(u) is the number of nodes in u's subtree and u.child is u's child on search path

- A node on the insertion path satisfying this is called a candidate scapegoat
- Rebuild the subtree rooted at u

Overview - More Details

• insert(5):



Overview - More Details

- Will we always find a scapegoat node? Yes!
- Is it unique? No! (9, 12, and 13 are all candidate scapegoats)
- Lemma: If there exists a node p such that  $depth(p) > \log_{3/2} m$ , then p has an ancestor u that is a candidate scapegoat, that is,

$$\frac{\text{size}(u.\,\text{child})}{\text{size}(u)} > \frac{2}{3}$$

- Proof: By contradiction.
  - Suppose that for every node u along the path to p, size(u.child)  $\leq (2/3)$ size(u)
  - Letting  $k = \operatorname{depth}(p)$ , by induction have  $\operatorname{size}(p) \le (2/3)^k n$
  - Since size $(p) \ge 1$ , this implies  $(3/_2)^k \le n$ , implies  $k \le \log_{3/_2} n \le \log_{3/_2} m$ , contradiction

Overview - More Details

- How do we compute size(u) for each node u?
- Two methods:
  - 1. Maintain a separate field, u.size, for each node storing the size of u's subtree (and update as needed)
  - 2. Compute it on the fly, after each insertion that requires rebalancing:
    - Walk up the search path toward the root
    - Let u be any ancestor of the inserted node. Assume we know size(u).
    - We want to compute size(*u*. parent):
      - Let u' be u's sibling. Traverse the subtree rooted at u' and count the number of nodes.
      - Set size(u. parent) = 1 + size(u) + size(u')
      - This may seem costly, but it can all be done within the amortized time bound!

Amortized Complexity

- Theorem: Starting with an empty tree, any sequence of k dictionary operations costs a total of O(k log k)
- Proof: (Sketch)
  - Find: Cost is  $O(\log n)$  always (by height bound)
  - Delete: In order to rebuild a tree due to deletions, at least half the entries have been deleted. A token-based analyses (recall stacks and rehashing from earlier lectures) can be applied here.
  - Insert: This is analyzed by a potential argument. Intuitively, after any subtree of size k is rebuilt it takes O(k) insertions to force this subtree to be rebuilt again. Charge the rebuilding time against these "cheap" insertions.
- Corollary: The amortized complexity of the scapegoat tree with at most n nodes is O(log n)

A data structure invented just for the programming assignment

- Overview An SG Tree is:
  - An extended binary search tree that is rebalanced like a scapegoat tree
  - Updated concepts:
    - The size of an internal node is the number of external nodes in its subtree
    - The height of a node is the maximum number of edges to any external node
  - Similarities with the scapegoat tree:
    - Maintain total size n and upper bound m, where  $n \leq m \leq 2n$
    - Height condition: Rebuild if tree height exceeds  $\log_{3/2} m$  ( $\Rightarrow$ Tree height is  $O(\log n)$ )
    - -Candidate scapegoat: Any node on search path such that size(u.child)/size(u) >  $\frac{2}{3}$
    - Deletion condition: If 2n < m, rebuild the entire tree, and set m=n

#### Differences from the scapegoat tree:

- Nodes: Two types of nodes:
  - External store data only (a city for the programming assignment)
  - Internal store splitter, left, right, subtree height, and subtree size
- Scapegoat Node:
  - When insertion causes the tree's height to exceed  $\log_{3/2} m$ , if multiple nodes satisfy the scapegoat condition, we chose the one closest to the root
  - Why? By rebuilding the largest subtree, we make the overall tree more balanced

#### Conventions:

To avoid floating-point round-off errors, use integer arithmetic to test the scapegoat condition:

 $2 \cdot \text{size}(u) < 3 \cdot \text{size}(u. \text{child}) \implies u \text{ is candidate scapegoat}$ 

- When inserting a new external node, the parent's splitter is taken from its left child

![](_page_21_Figure_5.jpeg)

#### More conventions:

- When rebuilding a subtree with k external nodes:
  - If k is even, split the internal nodes evenly among the left and right subtrees
  - If k is odd, the left subtree gets  $\lfloor k/2 \rfloor$  external nodes and the right subtree gets  $\lfloor k/2 \rfloor$
  - -More formally: When splitting the k-element subarray A[i ... i + k 1]:
    - Set m = [k/2]
    - Build left subtree with m keys: A[i ... i + m 1]
    - The splitter is A[i + m 1]
    - Build right subtree with k m keys:  $A[i + m \dots i + k 1]$
  - This convention results in the most even split and most balanced splitter value

#### Implementation hints

- Abstract class Node and two derived classes
  - ExternalNode stores just a city object
  - InternalNode stores splitter (a city), left, right, size, and height
- Take advantage of virtual functions when defining node operations
  - Don't do this:

```
Node insert(Node p) {
    if (p.isExternal) {
        ExternalNode pe = (ExternalNode) p;
        /* external node processing */
    }
    else {
        InternalNode pi = (InternalNode) p;
        /* internal node processing */
    }
}
```

![](_page_23_Figure_8.jpeg)

Implementation hints

Instead, do this:

```
abstract class Node {
    // ...
    abstract Node insert(Key x);
}
class InternalNode extends Node {
    // ...
    Node insert(Key x) { ... } // insertion at internal node
}
class ExternalNode extends Node {
    // ...
    Node insert(Key x) { ... } // insertion at external node
}
```

Implementation hints

- Your SGTree class:
  - Generic? It's up to you.
    - We don't maintain key-value pairs. We store city objects.
    - The print command assumes that the data object has a name and (x,y) coordinates
    - -We made ours generic, but the data type must support getName(), getX(), and getY()
  - Use inner classes for nodes:
    - -Node, InternalNode, ExternalNode

– Private data:

```
Node root;
Comparator comparator; (Optional. Given with the constructor)
Document resultsDoc; (Needed by print command)
int n, m; (Used by the scapegoat functions)
```

Implementation hints

- insert(x):
  - Insert the key using the standard recursive insertion algorithm
    - Some modifications needed because we have an extended tree
  - While backing out from recursion, update the size and height values for each node
  - Increment both n and m
  - If the new tree height exceeds  $\log_{3/2} m$ :
    - Traverse the search path from root down until finding the first candidate scapegoat

 $2 \cdot \text{size}(u) < 3 \cdot \text{size}(u. \text{child})$ 

- Rebuild this subtree (Note: *u* must be an internal node)
- (Be sure that your rebuilding function updates heights and sizes for all nodes)

Implementation hints

- delete(x):
  - Delete the key using the standard recursive deletion algorithm
    - Some modifications needed because we have an extended tree
  - While backing out from recursion, update the size and height values for each node
  - Decrement n but not m
  - If 2*n* < *m*:
    - Rebuild the entire tree
    - -Set m = n

Implementation hints

- Write utilities for handling size and height:
  - getSize(Node p): return (p.isExternal ? 1 : p.size)
  - getHeight(Node p): return (p.isExternal ? 0 : p.height)
  - InternalNode.update():

```
size = getSize(left) + getSize(right);
```

height = 1 + max(getHeight(left), getHeight(right));

- Write a debugging utility for "pretty printing" your tree
  - Call this function after each major operation (insert, delete, subtree rebuilding)
- Insert a boolean flag (e.g., DEBUG), which you can turn on and off for debugging

Implementation hints

- Problem:
  - My SG Tree is ordered by (x, y)-coordinates. How do I delete a city given just its name?
- Answer:
  - This is why we have the binary-search tree (which is ordered by name)
  - Create a "bogus city" with just a name (no coordinates)
  - Find this city in your binary-search tree and save this "complete city"
  - Delete this complete city from both data structures

# Supplemental

Example of a rebuild operation

![](_page_30_Figure_2.jpeg)

# Supplemental

Example of SG-Tree operations

![](_page_31_Figure_2.jpeg)

# Supplemental

Another example of SG-Tree operations

![](_page_32_Figure_2.jpeg)

# Summary

#### Extended Binary Search Trees

- Data stored only at the leaves (external nodes)
- Internal nodes are used only for locating the data
- Scapegoat Trees
  - Another amortized binary search tree data structure
  - Rebalancing through rebuilding subtrees
  - Unlike splay trees, height is guaranteed to be  $O(\log n)$

SG Tree

- An extended-tree variant of the scapegoat tree

![](_page_33_Picture_10.jpeg)