Lecture 6: Balanced Search Trees

Michael Dinitz

September 12, 2024 601.433/633 Introduction to Algorithms

- HW2 due now, HW3 released
- Regrade policy: 72 hours from when grades released
 - Don't abuse this!
 - If too many of your regrade requests do not result in positive changes, will ban you from regrade requests
 - Grading can go down!

Introduction

Today, and next few weeks: data structures.

 Since "Data Structures" a prereq, focus on advanced structures and on interesting analysis

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Today and later: data structures for dictionaries

Definition

A dictionary data structure is a data structure supporting the following operations:

- insert(key,object): insert the (key, object) pair.
- lookup(key): return the associated object
- delete(key): remove the key and its object from the data structure. We may or may not care about this operation.

Reminder: all running times for worst case

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Approach 1: Sorted array

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Goal: $O(\log n)$ for both.

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- Lookup: O(log n)
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Approach 2: Unsorted (linked) list

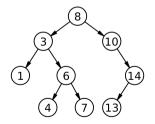
- ▶ Insert: *O*(1)
- Lookup: Ω(n)

Goal: **O(log n)** for both. Approach today: search trees

Binary Search Tree Review

Binary search tree:

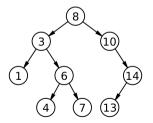
- All nodes have at most **2** children
- Each node stores (key, object) pair
- All descendants to left have smaller keys
- All descendants to the right have larger keys



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Lookup: follow path from root!

Dictionary Operations in Simple Binary Search Tree insert(x):

- If tree empty, put x at root
- Else if x < root.key recursively insert into left child</p>
- Else (if x > root.key) recursively insert into right child

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Example: H O P K I N S

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Want to make tree *balanced*.

Rest of today:

- B-trees: perfect balance, not binary
- Red-black trees: approximate balance, binary
- Turn out to be related!

B-Trees

B-tree Definition

Parameter $t \ge 2$.

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Definition (B-tree with parameter t)

- 1. Each node has between t 1 and 2t 1 keys in it (except the root has between 1 and 2t 1 keys). Keys in a node are stored in a sorted array.
- Each non-leaf has degree (number of children) equal to the number of keys in it plus 1. If v is a node with keys [a₁, a₂,..., a_k] and the children are [v₁, v₂,..., v_{k+1}], then the tree rooted at v_i contains only keys that are at least a_{i-1} and at most a_i (except the the edge cases: the tree rooted at v₁ has keys less than a₁, and the tree rooted at v_{k+1} has keys at least a_k).
- 3. All leaves are at the same depth.

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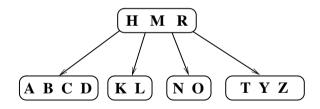
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When t = 2 known as a 2-3-4 tree, since # children either 2, 3, or 4

B-tree: Example

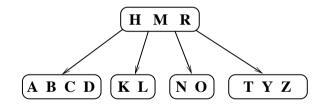
t = 3:

- \blacktriangleright Root has between 1 and 5 keys, non-roots have between 2 and 5 keys
- Non-leaves have between **3** and **6** children (root can have fewer).

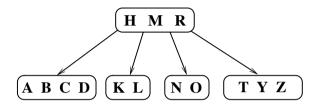


Lookups

Binary search in array at root. Finished if find item, else get pointer to appropriate child, recurse.



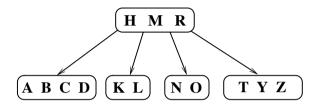
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Obvious approach: do a lookup, put x in leaf where it should be.

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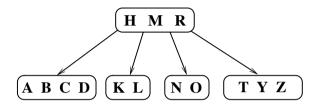


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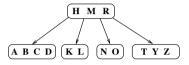
Problem: What if leaf is full (already has 2t - 1 keys)?

Split:

- Only used on *full* nodes (nodes with 2t 1 keys) whose parents are *not* full.
- Pull median of its keys up to its parent
- Split remaining 2t 2 keys into two nodes of t 1 keys each. Reconnect appropriately.

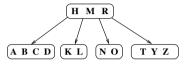
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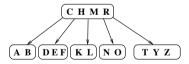


Insert **E**, **F** into example.

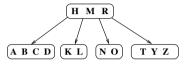
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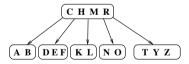
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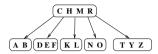
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Note: since split on the way down, when a node is split, its parent is not full!

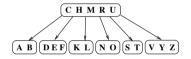




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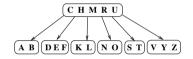


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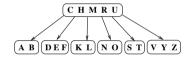
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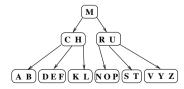
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Second property (correct degrees, subtrees have keys in correct ranges): Hooked nodes up correctly after split. \checkmark

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Used a lot in databases

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Large t: shallow trees. Fits well with memory hierarchy

t = 2:

- 2-3-4 tree
- Can be implemented as binary tree using red-black trees

Red-Black Trees

Red-Black Trees: Intro

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Most famous: *red-black trees*

- Default in Linux kernel, used to optimize Java HashMap,
- ► Today: Quick overview, connection to 2-3-4 trees.
- Not traditional or practical point of view on red-black trees. See book!

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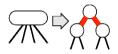


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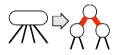
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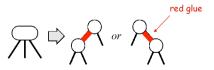
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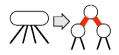
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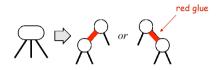
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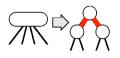


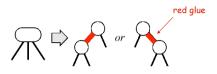
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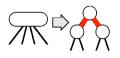


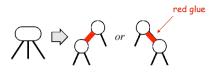




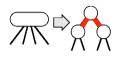


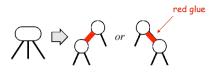
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 - ▶ Red edge is "internal", never have more than one "internal" edge in a row.



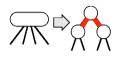


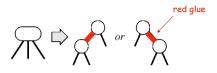
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 - Each black edge is a 2-3-4 tree edge
 - All leaves in 2-3-4 tree at same distance from root





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Want to insert while preserving two properties.

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 \mathcal{F}

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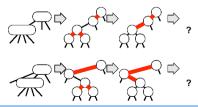


Want to insert while preserving two properties. 2-3-4 trees: split full nodes on way down.

Easy cases:



Harder cases:



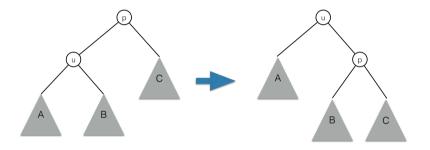
Lecture 6: Balanced Search Trees

Tree Rotations

Used in many different tree constructions.

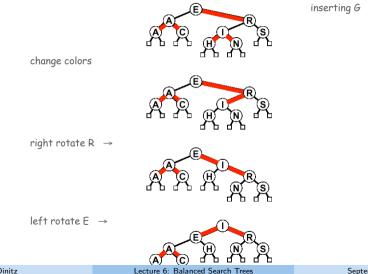
Tree Rotations

Used in many different tree constructions.



Using Rotations

Can use rotations to "fix" hard cases. Example:



A few more complications to deal with – see lecture notes, textbook.

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Main points:

- Red-Black trees can be thought of as a binary implementation of 2-3-4 trees
- Approximately balanced, so O(log n) lookup time
- ▶ Insert time (basically) same as 2-3-4 tree, so also *O*(log *n*).
- See book for direct approach (not through 2-3-4 trees).