#### Lecture 7: Balanced Search Trees

Jessica Sorrell

September 16, 2025 601.433/633 Introduction to Algorithms Slides by Michael Dinitz

#### **Announcements**

- ▶ HW1 due now, HW2 released
- Regrade policy: 120 hours (five days) 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

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

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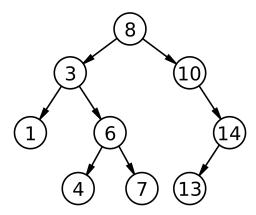
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Approach today: search trees

## Binary Search Tree Review

#### Binary search tree:

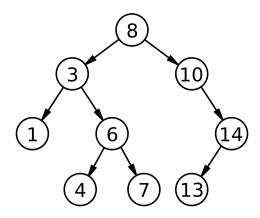
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- Each node stores (key, object) pair
- All descendants to left have smaller keys
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## Binary Search Tree Review

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Lookup: follow path from root!

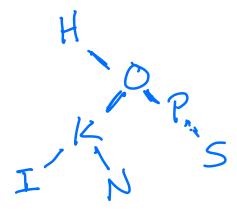
# Dictionary Operations in Simple Binary Search Tree insert(x):

- ▶ If tree empty, put **x** at root
- Else if x < root.key recursively insert into left child</p>
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Example: H O P K I N S



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a, b, c, d, e, 5

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#### 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.
- 2. 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  $[a_1, a_2, \ldots, a_k]$  and the children are  $[v_1, v_2, \ldots, v_{k+1}]$ , then the tree rooted at  $v_i$  contains only keys that are at least  $a_{i-1}$  and at most  $a_i$  (except the edge cases: the tree rooted at  $v_1$  has keys less than  $a_1$ , and the tree rooted at  $v_{k+1}$  has keys at least  $a_k$ ).
- 3. All leaves are at the same depth.

#### **B-tree Definition**

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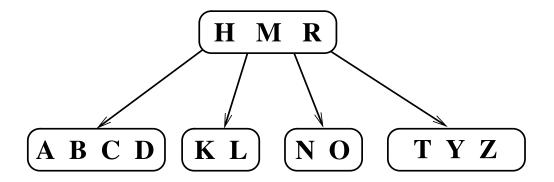
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When t = 2 known as a 2-3-4 tree, since # children either 2, 3, or 4

### B-tree: Example

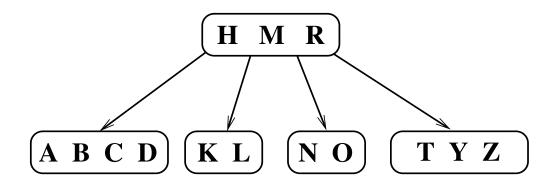
$$t=3$$
:

- ▶ 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).

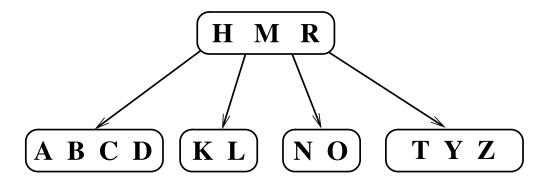


## Lookups

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



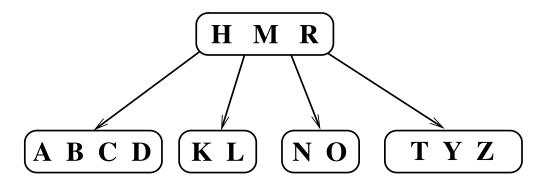
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Obvious approach: do a lookup, put x in leaf where it should be.

Example: insert **E** 

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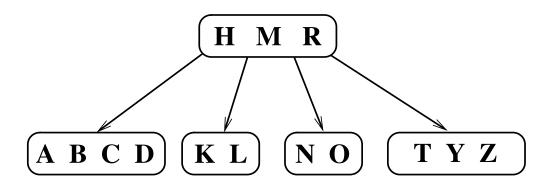


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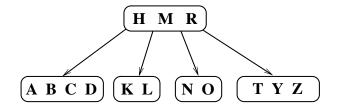
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## Split:

- ▶ 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.

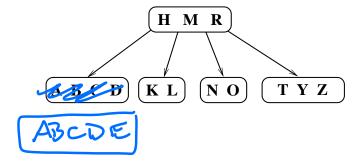
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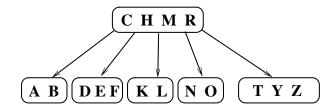


Insert *E*, *F* into example.

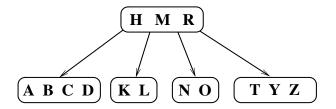
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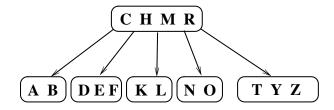
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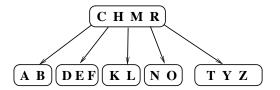
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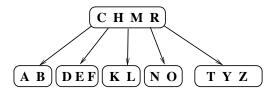


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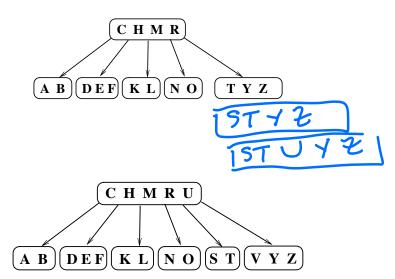
Note: since split on the way down, when a node is split, its parent is not full!

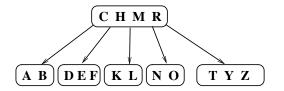




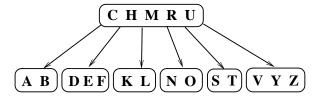
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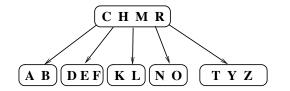




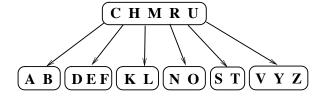
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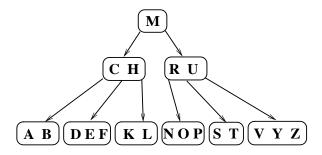
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t=2:

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

Red-Black Trees

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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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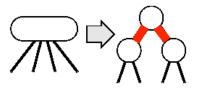
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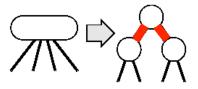


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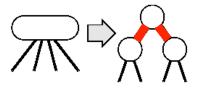
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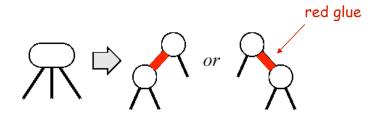
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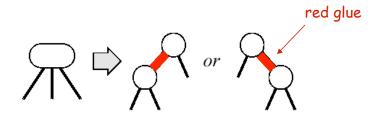
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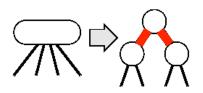


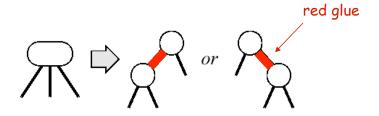
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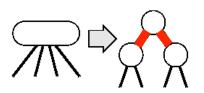


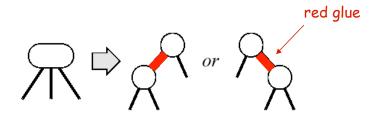




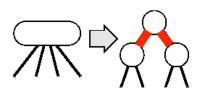


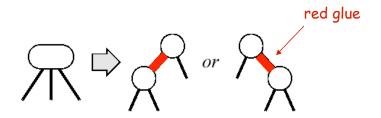
- 1. Never have two red edges in a row.
  - ▶ Red edge is "internal", never have more than one "internal" edge in a row.





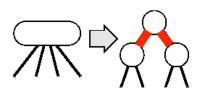
- 1. Never have two red edges in a row.
  - ▶ Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of black edges on path to root (black-depth)
  - ► Each black edge is a 2-3-4 tree edge
  - ▶ All leaves in 2-3-4 tree at same distance from root

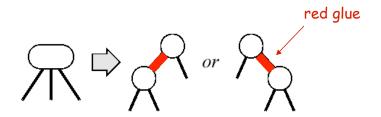




- 1. Never have two red edges in a row.
  - ▶ Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of *black* edges on path to root (*black-depth*)
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 $\implies$  path from root to deepest leaf  $\leq 2 \times$  path to shallowest leaf





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- $\implies$  depth  $\leq O(\log n)$

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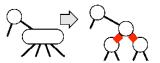
2-3-4 trees: split full nodes on way down.

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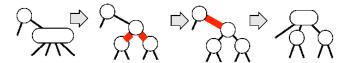
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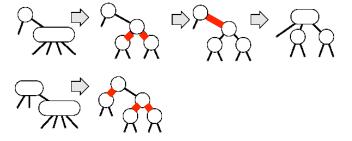
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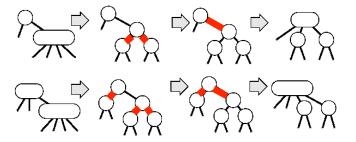
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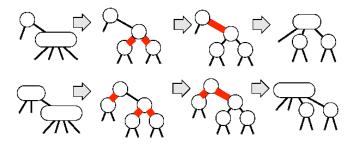
2-3-4 trees: split full nodes on way down.



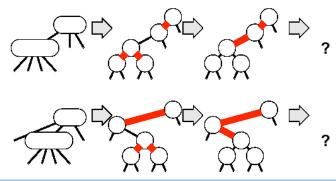
Want to insert while preserving two properties.

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#### Easy cases:



Harder cases:

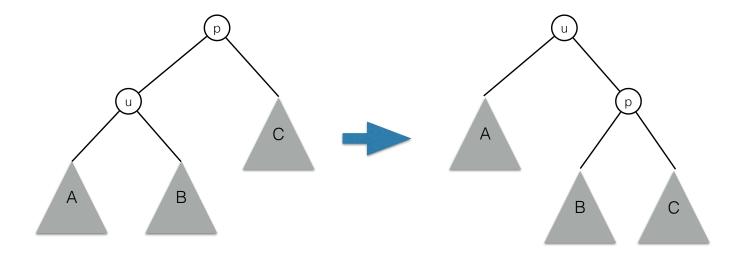


# Tree Rotations

Used in many different tree constructions.

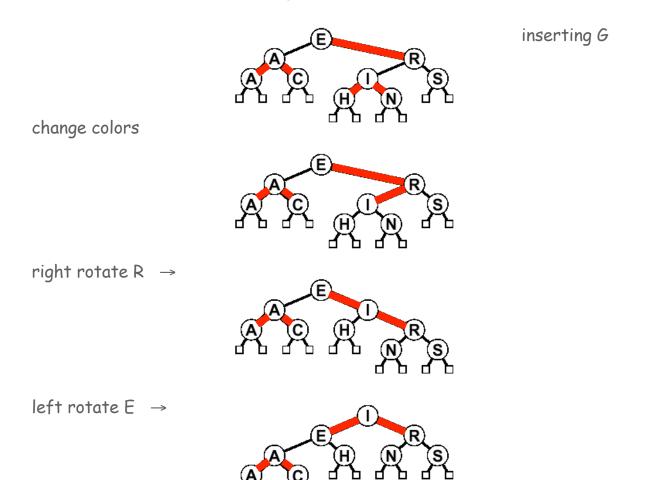
# Tree Rotations

Used in many different tree constructions.



# Using Rotations

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



Jessica Sorrell

Lecture 7: Balanced Search Trees

September 16, 2025

## End

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

#### End

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

#### Main points:

- ▶ Red-Black trees can be thought of as a binary implementation of 2-3-4 trees
- ightharpoonup 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).