#### Lecture 7: Balanced Search Trees

Michael Dinitz

September 16, 2025 601.433/633 Introduction to Algorithms

#### **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

Today, and next few weeks: data structures.

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

#### Introduction

Today, and next few weeks: data structures.

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

Today and later: data structures for *dictionaries* 

#### Introduction

Today, and next few weeks: data structures.

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

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

Reminder: all running times for worst case

Approach 1: Sorted array

Reminder: all running times for worst case

Approach 1: Sorted array

► Lookup:

Reminder: all running times for worst case

Approach 1: Sorted array

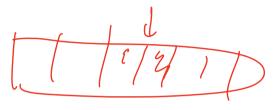
▶ Lookup:  $O(\log n)$ 

Reminder: all running times for worst case

Approach 1: Sorted array

▶ Lookup:  $O(\log n)$ 

Insert:



Reminder: all running times for worst case

Approach 1: Sorted array

▶ Lookup:  $O(\log n)$ 

• Insert:  $\Omega(n)$ 

Reminder: all running times for worst case

Approach 1: Sorted array

• Lookup:  $O(\log n)$ 

• Insert:  $\Omega(n)$ 

Approach 2: Unsorted (linked) list

Reminder: all running times for worst case

Approach 1: Sorted array

▶ Lookup:  $O(\log n)$ 

• Insert:  $\Omega(n)$ 

Approach 2: Unsorted (linked) list

► Insert:

Reminder: all running times for worst case

Approach 1: Sorted array

▶ Lookup:  $O(\log n)$ 

• Insert:  $\Omega(n)$ 

Approach 2: Unsorted (linked) list

▶ Insert: *O*(1)

Reminder: all running times for worst case

Approach 1: Sorted array

• Lookup:  $O(\log n)$ 

• Insert:  $\Omega(n)$ 

Approach 2: Unsorted (linked) list

- ▶ Insert: *O*(1)
- ► Lookup:

Reminder: all running times for worst case

Approach 1: Sorted array

• Lookup:  $O(\log n)$ 

• Insert:  $\Omega(n)$ 

Approach 2: Unsorted (linked) list

- ▶ Insert: *O*(1)
- Lookup:  $\Omega(n)$

Reminder: all running times for worst case

Approach 1: Sorted array

- ▶ Lookup:  $O(\log n)$
- ▶ Insert:  $\Omega(n)$

Approach 2: Unsorted (linked) list

- ▶ Insert: *O*(1)
- Lookup:  $\Omega(n)$

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

Reminder: all running times for worst case

Approach 1: Sorted array

- ▶ Lookup:  $O(\log n)$
- Insert:  $\Omega(n)$

Approach 2: Unsorted (linked) list

- ▶ Insert: *O*(1)
- Lookup:  $\Omega(n)$

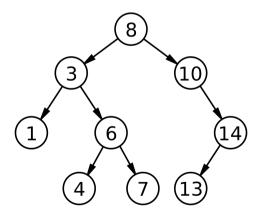
Goal:  $O(\log n)$  for both.

Approach today: search trees

## Binary Search Tree Review

#### Binary search tree:

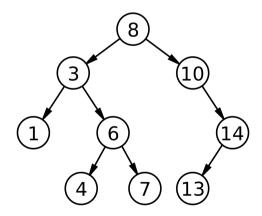
- ▶ All nodes have at most 2 children
- Each node stores (key, object) pair
- All descendants to left have smaller keys
- All descendants to the right have larger keys



## 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



Lookup: follow path from root!

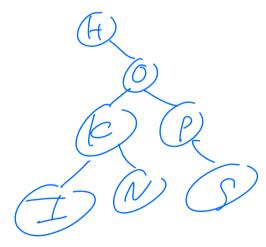
# Dictionary Operations in Simple Binary Search Tree insert(x):

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

# 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
- ► Else (if *x* > *root.key*) recursively insert into right child

Example: H O P K I N S



Pluses: easy to implement

Pluses: easy to implement

(Worst-case) Running time:

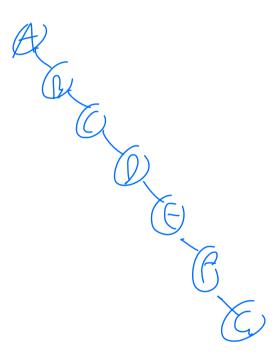
Pluses: easy to implement

(Worst-case) Running time: if depth d, then  $\Theta(d)$ 

Pluses: easy to implement

(Worst-case) Running time: if depth d, then  $\Theta(d)$ 

• If very unbalanced d could be  $\Omega(n)$ !



Pluses: easy to implement

(Worst-case) Running time: if depth d, then  $\Theta(d)$ 

• If very unbalanced d could be  $\Omega(n)$ !

Want to make tree balanced.

Pluses: easy to implement

(Worst-case) Running time: if depth d, then  $\Theta(d)$ 

• If very unbalanced d could be  $\Omega(n)$ !

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  $t \ge 2$ .

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





Parameter  $t \geq 2$ .

## 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 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

Parameter  $t \geq 2$ .

### Definition (B-tree with parameter *t*)

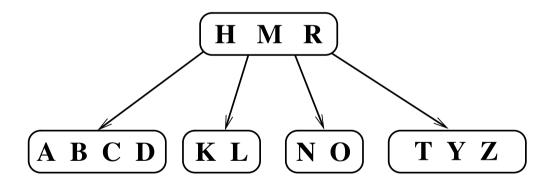
- 1. Each node has between t-1 and 2t-1 keys in it (except the root has between 1 and 2t-1 keys). Keys in a node are stored in a sorted array.
- 2. Each non-leaf has degree (number of children) equal to the number of keys in it plus 1. If v is a node with keys  $[a_1, a_2, \ldots, a_k]$  and the children 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 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.

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

### B-tree: Example

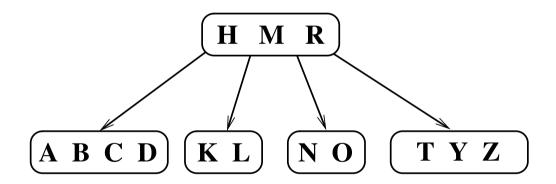
$$t = 3$$
:

- ▶ 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).

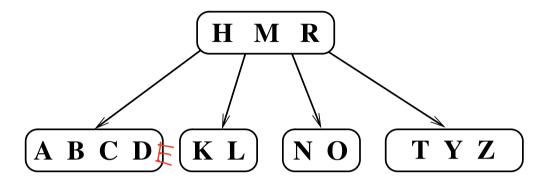


## Lookups

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



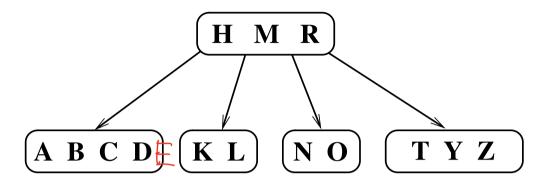
## Insert(x)



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

Example: insert **E** 

## Insert(x)

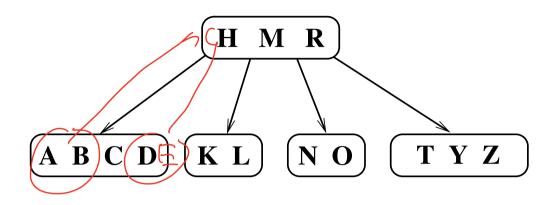


Obvious approach: do a lookup, put x in leaf where it should be.

Example: insert *E* 

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

# Insert(x)



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

Example: insert *E* 

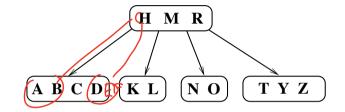
Problem: What if leaf is full (already has 2t - 1 keys)?

### 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.

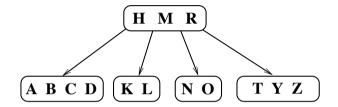
Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.

Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.

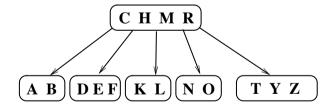


Insert *E*, *F* into example.

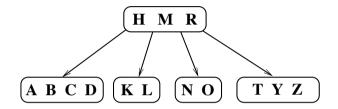
Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.



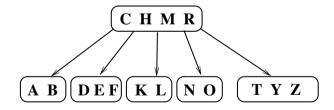
Insert **E**, **F** into example.



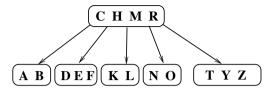
Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.

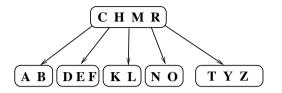


Insert **E**, **F** into example.

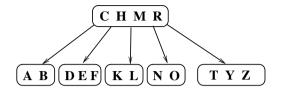


**Note:** since split on the way down, when a node is split, its parent is not full!

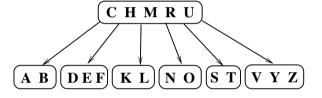


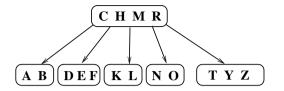


Insert *S*, *U*, *V*:

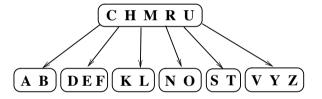


Insert *S*, *U*, *V*:

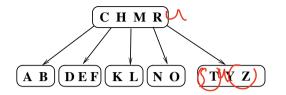




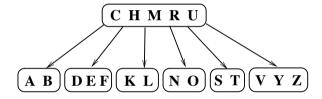
Insert *S*, *U*, *V*:



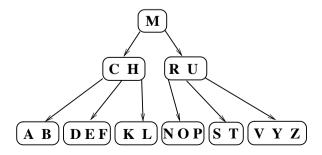
Insert **P**:



Insert *S*, *U*, *V*:



Insert **P**:



Induction. Start with a valid B-tree, insert x.

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth):

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t - 1 and 2t - 1 keys):

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t - 1 and 2t - 1 keys):

► No split:

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t-1 and 2t-1 keys):

▶ No split: only leaf changes, was not full (or would have split)

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t-1 and 2t-1 keys):

- ▶ No split: only leaf changes, was not full (or would have split)
- Split:

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t-1 and 2t-1 keys):

- No split: only leaf changes, was not full (or would have split)
- ▶ Split: Parent was not full. New nodes have exactly t 1 keys.

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t-1 and 2t-1 keys):

- No split: only leaf changes, was not full (or would have split)
- ▶ Split: Parent was not full. New nodes have exactly t 1 keys.

Second property (correct degrees, subtrees have keys in correct ranges):

Induction. Start with a valid B-tree, insert x.

Third property (all leaves at same depth): Tree grows up. ✓

First property (all non-leaves other than root have between t - 1 and 2t - 1 keys):

- No split: only leaf changes, was not full (or would have split)
- ▶ Split: Parent was not full. New nodes have exactly t 1 keys.

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

Suppose n keys, depth d

Suppose n keys, depth  $d \leq O(\log_t n)$ 

Suppose n keys, depth  $d \leq O(\log_t n)$ 

### Lookup:

▶ Binary search on array in each node we pass through

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

#### Insert:

Same as insert, but need to split on the way down & insert into leaf

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- Same as insert, but need to split on the way down & insert into leaf
- ▶ Total: lookup time + splitting time + time to insert into leaf

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- ▶ Same as insert, but need to split on the way down & insert into leaf
- ▶ Total: lookup time + splitting time + time to insert into leaf
  - Insert into leaf:

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- Same as insert, but need to split on the way down & insert into leaf
- Total: lookup time + splitting time + time to insert into leaf
  - ▶ Insert into leaf: O(t)

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- ▶ Same as insert, but need to split on the way down & insert into leaf
- ▶ Total: lookup time + splitting time + time to insert into leaf
  - ▶ Insert into leaf: O(t)
  - Splitting time:

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- Same as insert, but need to split on the way down & insert into leaf
- ▶ Total: lookup time + splitting time + time to insert into leaf
  - ► Insert into leaf: O(t)
  - Splitting time: O(t) per split

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- ▶ Same as insert, but need to split on the way down & insert into leaf
- ▶ Total: lookup time + splitting time + time to insert into leaf
  - ▶ Insert into leaf: O(t)
  - ▶ Splitting time: O(t) per split  $\implies O(td) = O(t \log_t n)$  total

Suppose n keys, depth  $d \leq O(\log_t n)$ 

#### Lookup:

- ▶ Binary search on array in each node we pass through  $\implies O(\log t)$  time per node.
- ► Total time  $O(d \times \log t) = O(\log_t n \times \log t) = O(\frac{\log n}{\log t} \times \log t) = O(\log n)$

- Same as insert, but need to split on the way down & insert into leaf
- Total: lookup time + splitting time + time to insert into leaf
  - ▶ Insert into leaf: O(t)
  - ▶ Splitting time: O(t) per split  $\implies O(td) = O(t \log_t n)$  total
- $O(t \log_t n) = O(\frac{t}{\log t} \log n) \text{ total}$

### B-tree notes

Used a lot in databases

► Large t: shallow trees. Fits well with memory hierarchy

### B-tree notes

Used a lot in databases

▶ 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

B-Trees great, but binary is nice: lookups very simple! Want *binary* balanced tree.

#### Red-Black Trees: Intro

B-Trees great, but binary is nice: lookups very simple! Want *binary* balanced tree.

- Classical and super important data structure question
- Many solutions!

#### Red-Black Trees: Intro

B-Trees great, but binary is nice: lookups very simple! Want *binary* balanced tree.

- Classical and super important data structure question
- Many solutions!

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!

Can we turn a 2-3-4 tree into a binary tree with all the same properties?

Can we turn a 2-3-4 tree into a binary tree with all the same properties?

► *No*: can't have perfect balance!

Can we turn a 2-3-4 tree into a binary tree with all the same properties?

- ► No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

Can we turn a 2-3-4 tree into a binary tree with all the same properties?

- ► No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

Nodes in 2-3-4 tree have degree 2, 3, or 4

Can we turn a 2-3-4 tree into a binary tree with all the same properties?

- No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

Nodes in 2-3-4 tree have degree 2, 3, or 4

Degree 2: good!

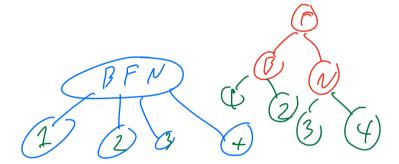


Can we turn a 2-3-4 tree into a binary tree with all the same properties?

- ► No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

Nodes in 2-3-4 tree have degree 2, 3, or 4

- ▶ Degree 2: good!
- Degree 4:

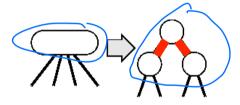


Can we turn a 2-3-4 tree into a binary tree with all the same properties?

- No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

Nodes in 2-3-4 tree have degree 2, 3, or 4

- Degree 2: good!
- Degree 4:



Can we turn a 2-3-4 tree into a binary tree with all the same properties?

- No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

Nodes in 2-3-4 tree have degree 2, 3, or 4

- Degree 2: good!
- Degree 4:



Degree 3:

Can we turn a 2-3-4 tree into a binary tree with all the same properties?

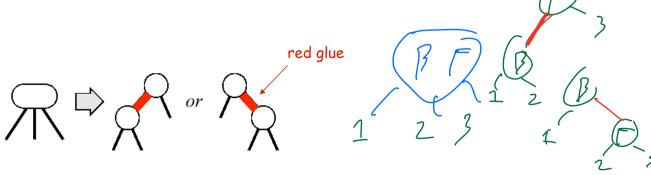
- No: can't have perfect balance!
- ▶ Just need depth  $O(\log n)$

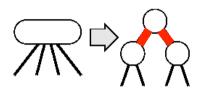
Nodes in 2-3-4 tree have degree 2, 3, or 4

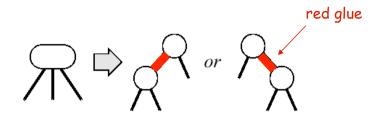
- Degree 2: good!
- Degree 4:

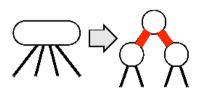


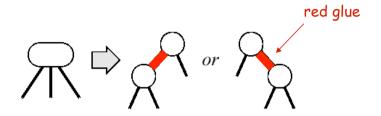
Degree 3:



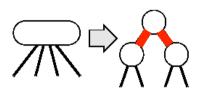


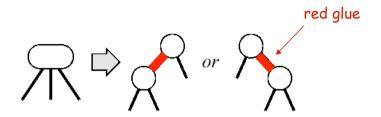




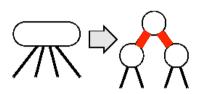


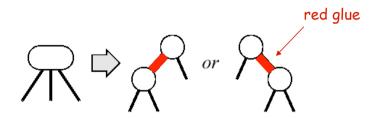
- 1. Never have two red edges in a row.
  - ▶ Red edge is "internal", never have more than one "internal" edge in a row.





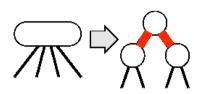
- 1. Never have two red edges in a row.
  - ▶ Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of black edges on path to root (black-depth)
  - ► Each black edge is a 2-3-4 tree edge
  - ▶ All leaves in 2-3-4 tree at same distance from root

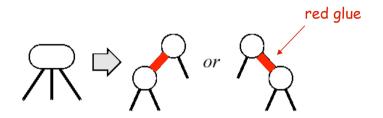




- 1. Never have two red edges in a row.
  - ▶ Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of black edges on path to root (black-depth)
  - ► Each black edge is a 2-3-4 tree edge
  - ► All leaves in 2-3-4 tree at same distance from root

 $\implies$  path from root to deepest leaf  $\leq 2 \times$  path to shallowest leaf





- 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
- $\implies$  path from root to deepest leaf  $\leq 2 \times$  path to shallowest leaf
- $\implies$  depth  $\leq O(\log n)$

Want to insert while preserving two properties.

Want to insert while preserving two properties.

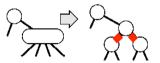
2-3-4 trees: split full nodes on way down.

Want to insert while preserving two properties.

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

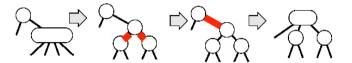
Want to insert while preserving two properties.

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



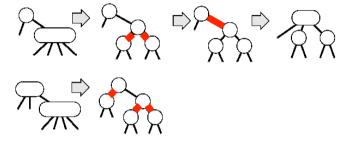
Want to insert while preserving two properties.

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



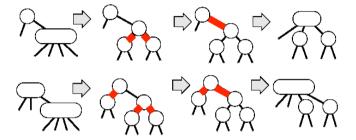
Want to insert while preserving two properties.

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



Want to insert while preserving two properties.

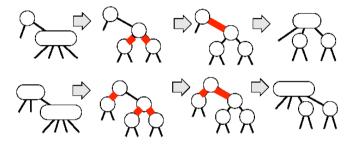
2-3-4 trees: split full nodes on way down.



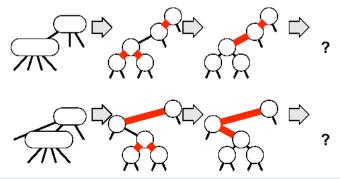
Want to insert while preserving two properties.

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

#### Easy cases:



#### Harder cases:

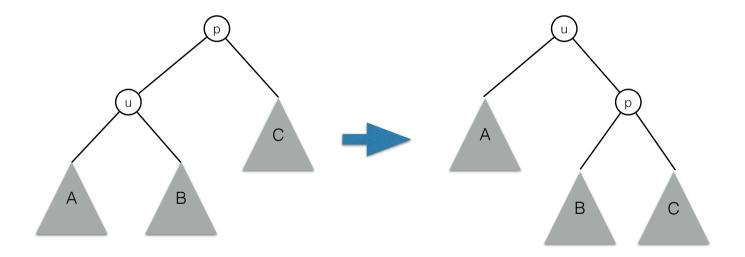


### Tree Rotations

Used in many different tree constructions.

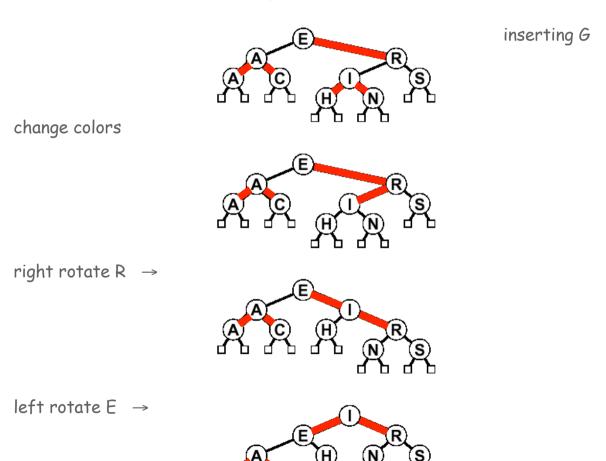
### Tree Rotations

Used in many different tree constructions.



# Using Rotations

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



Michael Dinitz

### 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).