Lecture 6: Balanced Search Trees

Michael Dinitz

September 12, 2024 601.433/633 Introduction to Algorithms

Announcements

- ▶ 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

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: $\Omega(n)$

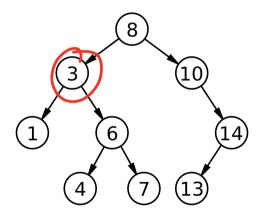
Goal: $O(\log n)$ for both.

Approach today: search trees

Binary Search Tree Review

Binary search tree:

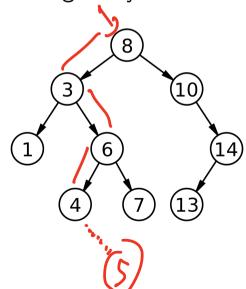
- ▶ All nodes have at most 2 children
- Each node stores (key, object) pair
- All descendants to left have smaller keys
- All descendants to the right have larger keys



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!

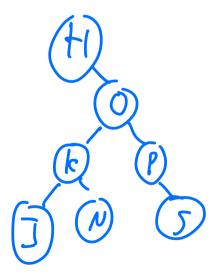
Dictionary Operations in Simple Binary Search Tree insert(x):

- ▶ 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:

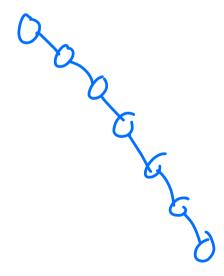
Pluses: easy to implement

(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 $\Omega(n)$!

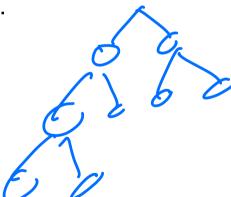


Pluses: easy to implement

(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:

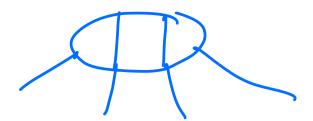
- ▶ B-trees: perfect balance, not binary
- Red-black trees: approximate balance, binary
- Turn out to be related!

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*)

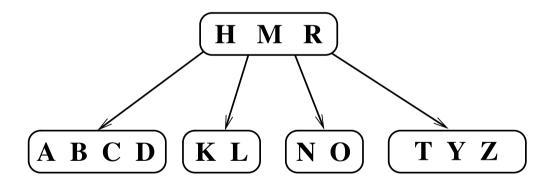
- 1. Each node has between t-1 and 2t-1 keys in it (except the root has between 1 and 2t-1 keys). Keys in a node are stored in a sorted array.
- 2. Each non-leaf has degree (number of children) equal to the number of keys in it plus 1. If v is a node with keys $[a_1, a_2, \ldots, a_k]$ and the children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

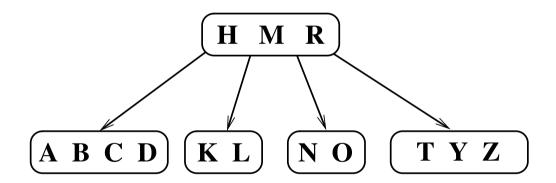
$$t = 3$$
:

- ▶ 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).

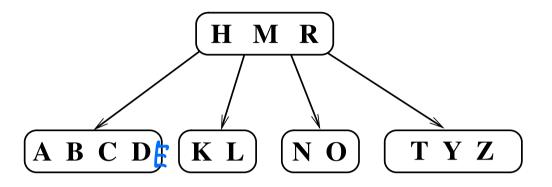


Lookups

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



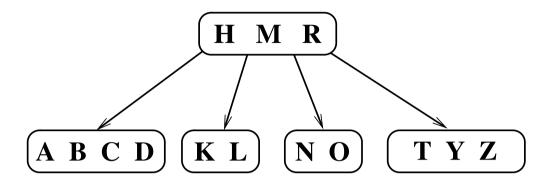
Insert(x)



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

Example: insert **E**

Insert(x)

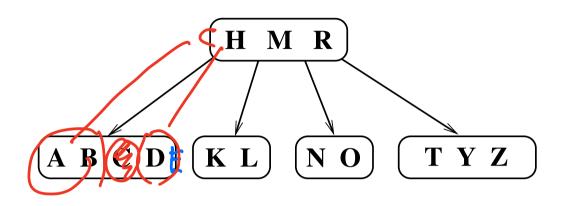


Obvious approach: do a lookup, put x in leaf where it should be.

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*

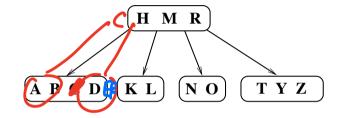
Problem: What if leaf is full (already has 2t - 1 keys)?

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.

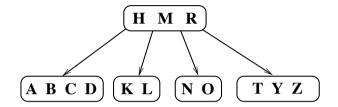
Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.

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

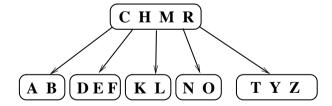


Insert *E*, *F* into example.

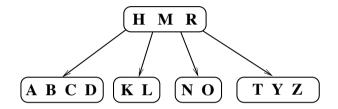
Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.



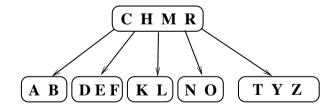
Insert **E**, **F** into example.



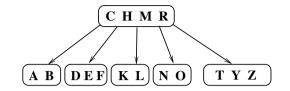
Insert: do a lookup and insert at leaf, but when we encounter a full node on way down, split it.

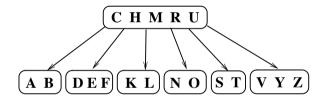


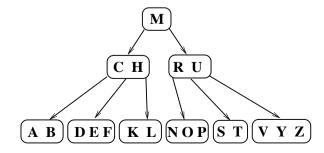
Insert **E**, **F** into example.



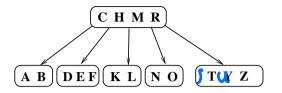
Note: since split on the way down, when a node is split, its parent is not full!



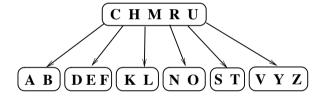


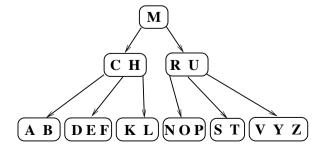


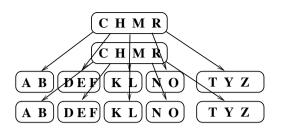
Michael Dinitz Lecture 6: Balanced Search Trees



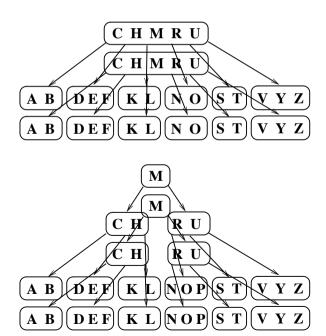
Insert *S*, *U*, *V*:

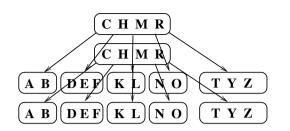




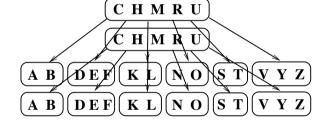


Insert *S*, *U*, *V*:

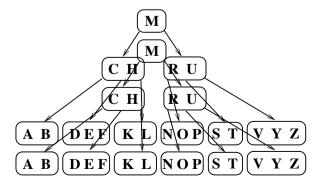


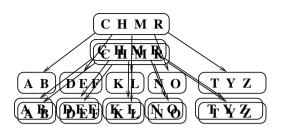


Insert *S*, *U*, *V*:



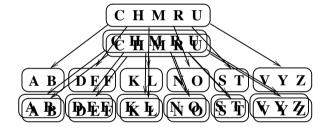
Insert **P**:

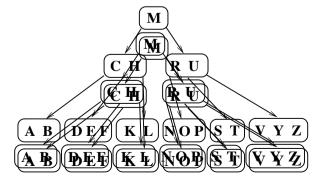




Insert *S*, *U*, *V*:

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

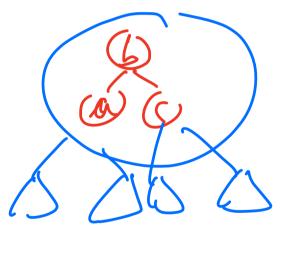
▶ Degree 2: good!

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:

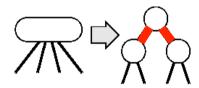


Can we turn a 2-3-4 tree into a binary tree with all the same properties?

▶ No: can't have perfect balance!

epth $O(\log n)$

ree have degree 2, 3, or 4 ood!



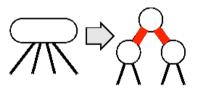
20 / 25

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

► No: can't have perfect balance!

epth $O(\log n)$

ree have degree 2, 3, or 4 ood!



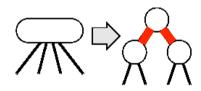
Degree 3:

20 / 25

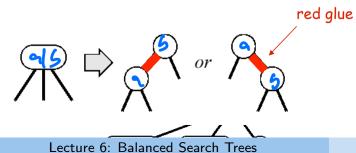
Can we turn a 2-3-4 tree into a binary tree with all the same properties?

► No: can't have perfect balance! •pth $O(\log n)$

ree have degree 2, 3, or 4 ood!

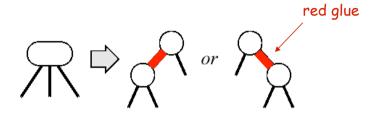


Degree 3:

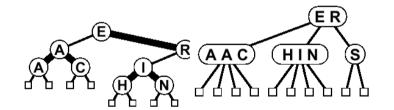


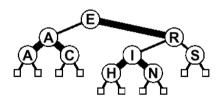
Michael Dinitz





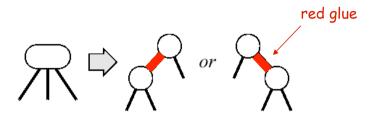






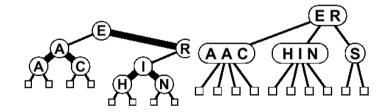
U/~cos226

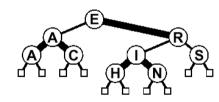




- 1. Never have two red edges in a row.
 - ▶ Red edge is "internal", never have more than one "internal" edge in a row.

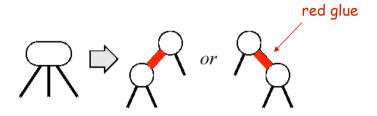






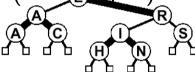
U/~cos226





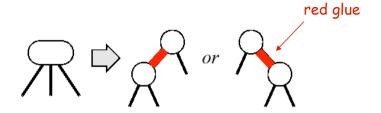
- 1. Never have two red edges in a row.
 - Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of black edges on Fath to root (black depth)
 - ► Each black edge is A2=3-4 tree edge AC HIN S

 All leaves in 2-3-4 tree at same distance from root



U/~cos226

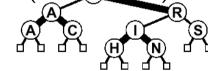




- 1. Never have two red edges in a row.
 - Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of black edges on Fath to root (black depth)

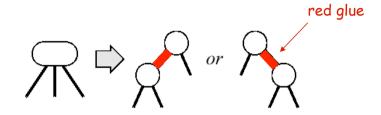
 - ► Each black edge is A2-3-4 tree edge A C HIN S

 All leaves in 2-3-4 tree at summer distance from root.



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

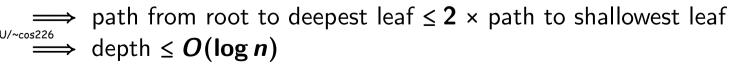




- 1. Never have two red edges in a row.
 - Red edge is "internal", never have more than one "internal" edge in a row.
- 2. Every leaf has same number of black edges on Fath to root (black depth)

 - ► Each black edge is A2-3-4 tree eliga A C HIN S

 ► All leaves in 2-3-4 tree at same distance from root



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.

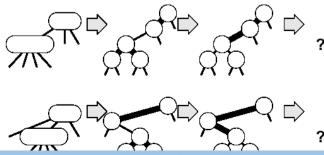
2-3-4 trees: split full nodes on way down.

Easy cases:

it to insert while preserving two properties.

4 trees: split full nodes on way down.

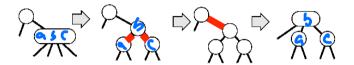
/ cases:

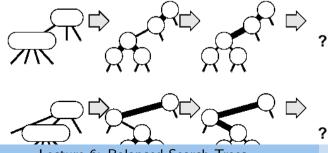


Lecture 6: Balanced Search Trees

- t to insert while preserving two properties.
- trees: split full nodes on way down.

cases:





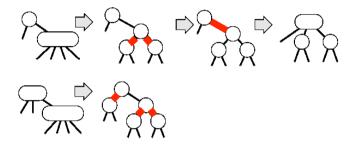
Michael Dinitz

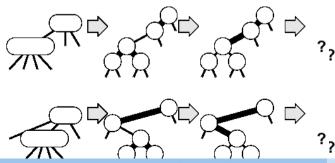
Lecture 6: Balanced Search Trees

Incort

- t to insert while preserving two properties.
- trees: split full nodes on way down.

cases:





Michael Dinitz

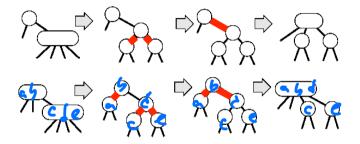
Lecture 6: Balanced Search Trees

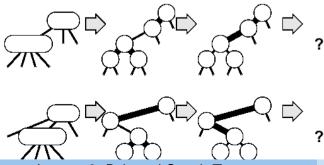
Incort

t to insert while preserving two properties.

trees: split full nodes on way down.

cases:



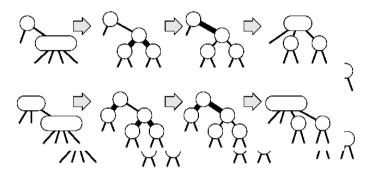


Lecture 6: Balanced Search Trees

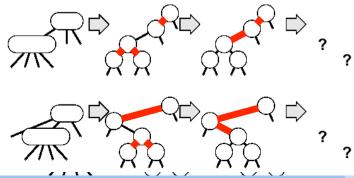
`^rt

- t to insert while preserving two properties.
- trees: split full nodes on way down.

cases:



irder cases:

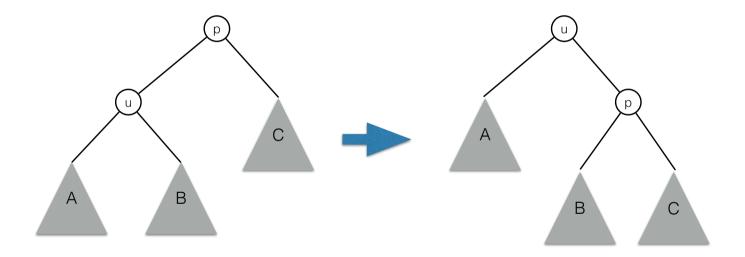


Tree Rotations

Used in many different tree constructions.

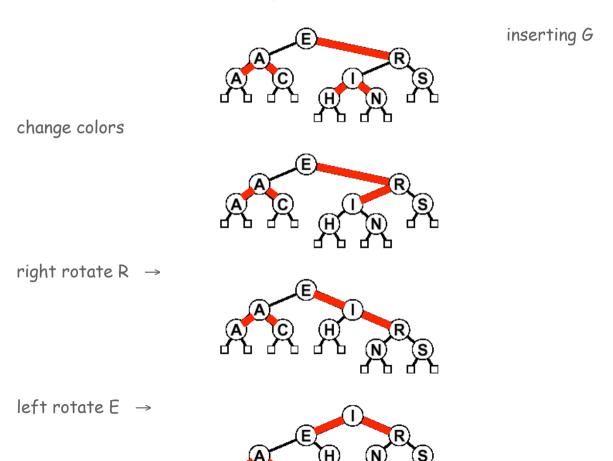
Tree Rotations

Used in many different tree constructions.



Using Rotations

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



Michael Dinitz

Lecture 6: Balanced Search Trees

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).