

CSE 750
10/3/2023

Hash tables (cont.)

①

Binary Search Trees

Hash table: array $H[0..(m-1)]$ with m entries
& easy to compute

"Random-looking" but deterministic hash function

$h(k) \in [0..(m-1)]$ for any key k

Collision: $k_1 \neq k_2$ but $h(k_1) = h(k_2)$

Handling collisions:

- Chaining — $h[j]$ is a linked list of all keys k such that $h(k) = j$.

- Open addressing — all items stored in the array H itself. If inserting k results in a collision: ($h(k)$ is occupied) then look for another location.

Different strategies:

— try $h(k)+1, h(k)+2, \dots$ until empty space found. [sequential search]
Naive & poorly performing

— better: try a different $h_2(k)$.
If this is occupied, then sequential search

Analysis (with chaining)

Uniform simple hashing assumption:

$$\forall j \quad \Pr_k [h(k) = j] = \frac{1}{m}$$

$0 \leq j \leq m-1$

Reasonable assumption if you have a sufficiently "crazy" hash function.

Worst Case: All keys hash to the same index;

	Insert	$O(1)$] no better than a linked list
unsuccessful	Search	$\Theta(n)$	
(unsuccessful)	Delete		

Expected time for an unsuccessful search given the uniform simple hashing assumption is

$$A(n) = \underbrace{\Theta(1)}_{\substack{\text{constant} \\ \text{time to} \\ \text{compute} \\ h(k)}} + \underbrace{\frac{1}{m}}_{\substack{\text{prob} \\ \Pr[h(k)=j]}} \left(\sum_{j=0}^{m-1} \frac{1}{m} \cdot \text{Length}(H[j]) \right)$$

$$= \Theta(1) + \left(\frac{1}{m} \sum_{j=0}^{m-1} \text{Length}(H[j]) \right)$$

$n = \# \text{ items in the hash table}$
 $m = \# \text{ array entries}$

$$= \Theta\left(1 + \frac{n}{m}\right)$$

If $n = O(m)$, then this is
constant time.

③

Binary Search Trees (BSTs)

Collection of items with comparable keys.

Supporting

Insert

Search

Delete

⋮

*Rooted
ordered* Binary tree of keys (assume no duplicate keys)

$k = \text{root key} \Rightarrow$ left subtree is a BST
containing only keys $< k$ &
right subtree is a BST
containing only keys $> k$.

Search, Insert, Delete all take ^{worst-case} $\Theta(h)$ time
where h is the height of the tree.

* Avoid BSTs where h is big.

Generally want $h = O(\lg n)$ ($n = \# \text{items}$)

Some standard techniques to ~~keep~~ keep a BST balanced so that $h = O(\lg n)$ always: (4)

- AVL trees

- Red-Black trees (2-3 trees)

Randomization! Does that help?

~~If items~~

With n items, there are $n!$ many permutations.

If one of these perms is chosen uniformly at random for insertion order into the BST, then the expected ^{time for each op} height is $O(\lg n)$.

[Analysis is the same (essentially) as randomized quicksort]

Can't count on a random insertion order but can add randomness to the algo to mimic a random insertion order. Treap (tree/heap hybrid).

Def: A treap T is a binary tree of items that each have a comparable key and a numerical priority. Such that

T is a BST of the keys (ignoring priorities)

2 T is a max heap of the priorities (ignoring keys) (5)
each nonroot
priority is $<$ parent priority

Fact: [assume no duplicate priorities]

The shape of T depends only on the set of (key, priority) pairs.

Proof: by induction on the size of T
[base case: T is empty]

T nonempty: root is item with highest priority
[regardless of anything else]

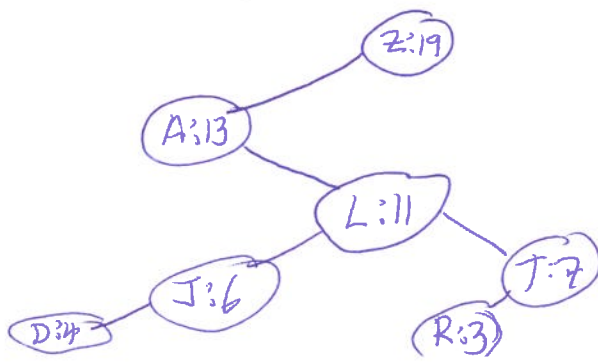
Left subtree: determined by the root key ~~and~~
among set of all keys.

Inductive hypothesis: root, left shape is uniquely determined

Same for the right subtree.

Ex: key:priority items

J:6, A:13, R:3, T:7, Z:19, L:11, D:4

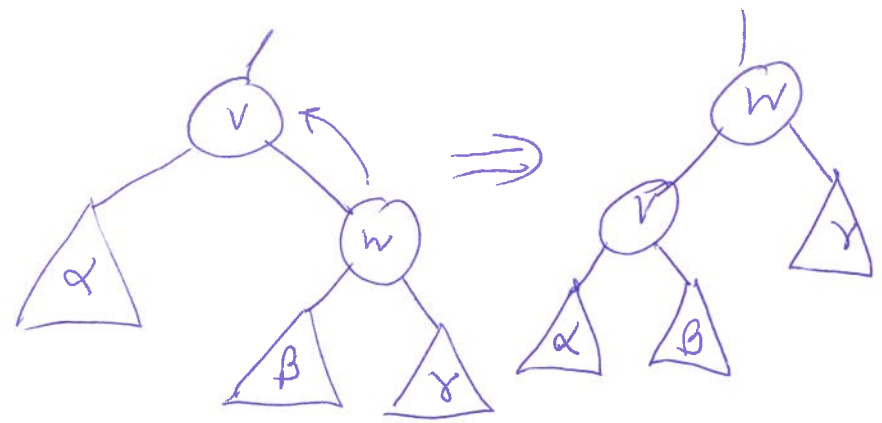


Rotations 2 types:

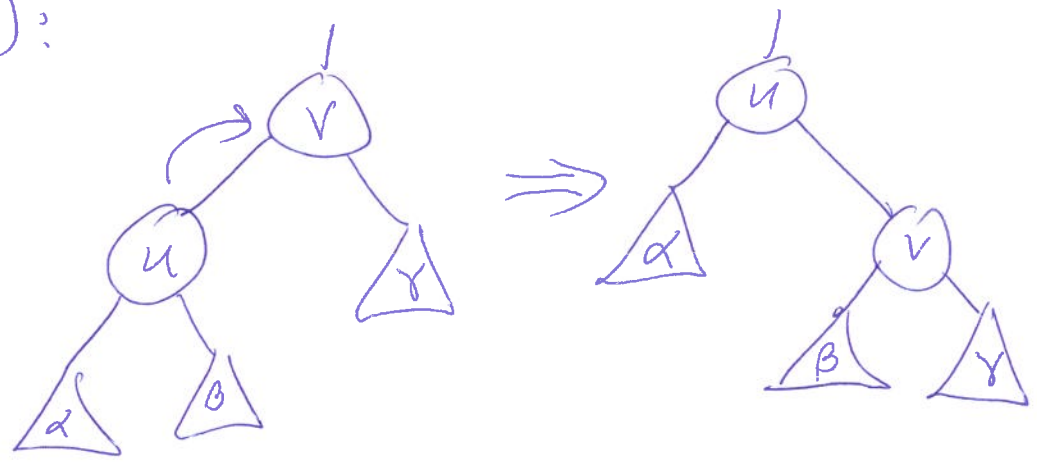
6

$O(1)$
time

LeftRotate(v):



RightRotate(v):



~~Implementing a BST using a randomized treap~~
~~treap~~

Inserting into a treap:

- 1) Insert as usual with a BST (new item at a leaf)
- 2) Go up the search path, ~~giving~~ making rotations until the max-heap order is restored.