

Hash tables:

Some common hash functions;  
 Table size =  $m$  ( $[0..(m-1)]$ )  
 Keys are integers:  
 $h(k) = k \bmod m$   
 Keys are strings:  
 int  $h(\text{char } *s)$  //  $s$  is null-terminated string  
 $\{$   
 int sum;  
 sum = 0;  
 while (\*s) {  
 sum =  $\frac{37}{NUM} * \text{sum} + *s$ ;  
 sum % = m;  
 s++;  
 }  
 Best if NUM is coprime with m  
 (no factors in common)  
 NUM  $\neq 1$   
 NUM  $\neq m-1$   
 Best if m is prime

Hash table with  $m$  entries  
 & storing  $n$  items:  
Load  $\alpha = \frac{n}{m}$   
 = ave length of a linked list if chaining is used  
 (aka, utility factor, traffic intensity).  
 Good to have, say,  $\frac{1}{2} \leq \alpha \leq 2$

Probing (open addressing)  
 All items stored in the array itself.  
 Dealing with collisions is fraught.

- one approach: linear search  
 if new item collides with existing item, search sequentially (cyclically) for free entry to put the new item.

disadvantages:  
 - deletion is messy  
 - clustering:  


General disadvantage with open addr:  $\alpha \leq 1$   
 ( $\alpha = 1$  makes poor performance).  
 Other probing strategies for collisions: use a hash function (or run a B+ tree) but not for sequential search.

Binary Search Trees

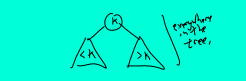
Collection supporting  
 Insert ( $x$ )  
 Search ( $x$ )  
 Delete ( $x$ )

Advantages of B+ trees vs a BST  
 - (non-ordered) insertion  
 - selection of the  $k$ th smallest item.

Binary tree, each node stores one data item (with key), no duplicates.

Disadvantages over a hash  
 - items must (easily) comparable  
 - worst case behavior is bad for  $n$  items  
 (but there are ways to improve on this)  
 actual worst-case behavior is  $\Theta(d)$  where  $d$  is the tree depth.  
 hence depth balancing to keep within range of big- $O$  run

- AVL trees  
 - red-black trees (related to 2-3 trees)  
 Does randomness help? Yes  
 worst case behavior is  $\Theta(\log n)$



If items inserted in random order (all  $n!$  insertion orderings equally likely), then

expected depth is  $O(\lg n)$

Analysis is similar to that of randomized quicksort.

$n$  items, laid out in increasing order



$E(n)$  = expected time for  $n$  insertions in random order satisfies

$$E(n) = \Theta(n) + \frac{1}{n} \sum_{j=0}^{n-1} (E(j) + E(n-j-1))$$

same recurrence as for randomized quicksort, so  $E(n) = n \lg n$

Similar analysis gives expected depth =  $O(\lg n)$ .

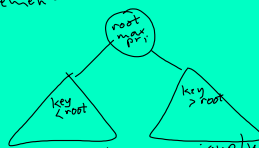
Works if all  $n$  items are available beforehand, but this is usually not the case.

Treap data structure

Binary search tree where each item has both a key (for the BST) and a priority (numerical) such that the keys are arranged in BST order and the priorities are in max-heap order (child priority is  $\leq$  parent priority)

Theorem: Given (key, priority) pairs (no duplicate keys, no duplicate priorities), the shape of the treap is uniquely determined by these pairs, so shape is independent of the insertion order.

Proof: root is max priority element



contents of subtrees uniquely determined by the root.

Applying this same idea recursively for all nodes, their shapes are uniquely determined by the key:priority pairs in their respective subtrees.  $\square$

Example

key	A	B	C	D	E	F	G	H
priority	1	7.5	7	3.7	8.7	3.5	2	9.6

