

# CSE 326 Data Structures

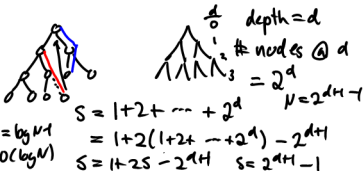
CSE 326 : Dave Bacon

Priority Queues : Floyd's Algorithm,  
D heaps, Leftist heaps, ...

Homework 2 due Friday

# Binary Min Heaps (summary)

- **insert**: percolate up.  $O(\log N)$  time.
- **deleteMin**: percolate down.  $O(\log N)$  time.



# Other Priority Queue Operations

- **decreaseKey**

- given a pointer to an object in the queue, reduce its priority value



Solution: change priority and

percolate up

- **increaseKey**

- given a pointer to an object in the queue, increase its priority value

15 → 29

$O(\log N)$

Why do we need a *pointer*? Why not simply data value?

Solution: change priority and

percolate down

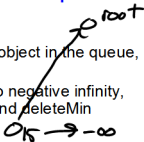
Hard to Find

# More Priority Queue Operations

- **Remove(objPtr)**

– given a pointer to an object in the queue, remove it

**Solution:** set priority to negative infinity, percolate up to root and deleteMin



Worst case Running time for all of these:

FindMax?



$O(N)$



ExpandHeap – when heap fills, copy into new space.  $O(N)$

# More Priority Queue Operations

5, 15, 25, 16, 39  $N$

- **buildHeap**

Naïve solution:

But in 1 by 1



Running time:

insert  $O(\log N)$

→  $O(N \log N)$

Can we do better?



# Buildheap pseudocode

```
private void buildHeap() {  
    for ( int i = currentSize/2; i > 0; i-- )  
        percolateDown( i );  
}
```

$O(\log N)$



$N$



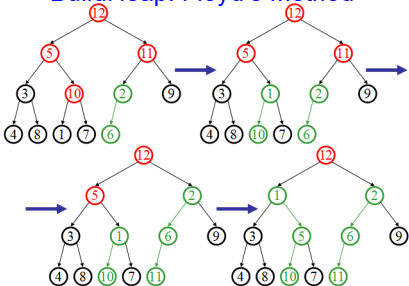
↓

runtime:

1st guess

$O(N \log N)$  ??

# BuildHeap: Floyd's Method





Finally... Sum of the heights of the nodes.

$$\sum_{i=0}^h \sum_{j=0}^i 2^{h-j}$$

height #  
 $0 \quad 2^h$   
 $1 \quad 2^{h-1}$   
 $\dots$   
 $h \quad 1$   
 $\sum_{i=0}^h i \cdot 2^{h-i}$   
 height # at height



height = h  
 # nodes  $2^{h+1} - 1$

$$\underline{2^{h+1} - 2 - h}$$

$O(N)$  vs

~~$O(N \log N)$~~   
 $2^i (h-1)$

runtime:

$$S = 2^h \cdot 0 + 2^{h-1} \cdot 1 + 2^{h-2} \cdot 2 + 2^{h-3} \cdot 3 + \dots + 1 \cdot h$$

$$\underline{-2S} = 2^h \cdot 1 + 2^{h-1} \cdot 2 + 2^{h-2} \cdot 3 + \dots + 2h$$

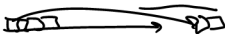
$$S = \frac{2^h + 2^{h-1} + 2^{h-2} + \dots + 2 - h}{2^{h+1} - 1 - 1 + h} = \underline{2^{h+1} - 2 - h}$$

# Facts about Heaps

Observations:

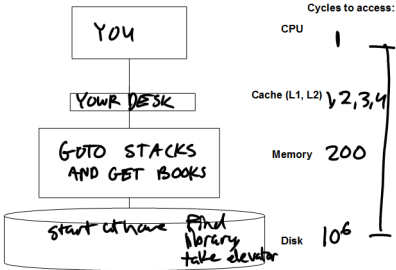
- finding a child/parent index is a multiply/divide by two
- operations jump widely through the heap
- each percolate step looks at only two new nodes
- inserts are at least as common as deleteMins

fast  
↓



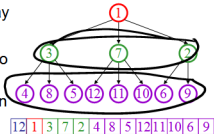
Realities:

- division/multiplication by powers of two are equally fast
- looking at only two new pieces of data: bad for cache!
- with huge data sets, disk accesses dominate



# A Solution: $d$ -Heaps

- Each node has  $d$  children
- Still representable by array
- Good choices for  $d$ :
  - (choose a power of two for efficiency)
  - fit one set of children in a cache line
  - fit one set of children on a memory page/disk block



$$\log_d N \quad I \quad \Delta$$

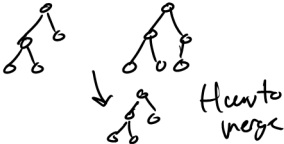
## Operations on $d$ -Heap

- Insert : runtime =  $O(\log_d N)$
- deleteMin: runtime =  $O(d \log_d N)$   
↑

Does this help insert or deleteMin more?

# One More Operation

- Merge two heaps. Ideas?



Leftist Heaps → Friday

# Leftist Heaps

Idea:

Focus all heap maintenance work in one small part of the heap

Leftist heaps:

1. Most nodes are on the left
2. All the merging work is done on the right

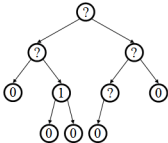
# Definition: Null Path Length

*null path length (npl)* of a node  $x$  = the number of nodes between  $x$  and a null in its subtree

OR

$npl(x)$  = min distance to a descendant with 0 or 1 children

- $npl(\text{null}) = -1$
- $npl(\text{leaf}) = 0$
- $npl(\text{single-child node}) = 0$



Equivalent definitions:

1.  $npl(x)$  is the height of largest complete subtree rooted at  $x$
2.  $npl(x) = 1 + \min \{npl(\text{left}(x)), npl(\text{right}(x))\}$



## Leftist Heap Size

- A leftist tree with  $r$  nodes on the right path must have at least  $2^r - 1$  nodes
- Induction
- $r=1$
  
- Assume true for  $1, \dots, r-1$ . Then leftist heap size  $r$ :

# Leftist Heap Properties

- Heap-order property
  - parent's priority value is  $\leq$  to childrens' priority values
  - result: minimum element is at the root
- Leftist property
  - For every node  $x$ ,  $npl(\text{left}(x)) \geq npl(\text{right}(x))$
  - result: tree is at least as "heavy" on the left as the right

Are leftist trees...

complete?

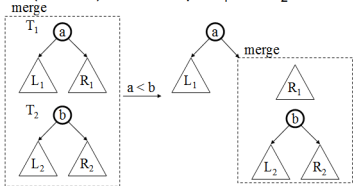
balanced?

## Merge two leftist heaps (basic idea)

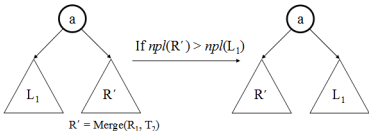
- Put the smaller root as the new root,
- Hang its left subtree on the left.
- Recursively merge its right subtree and the other tree.

# Merging Two Leftist Heaps

- $\text{merge}(T_1, T_2)$  returns one leftist heap containing all elements of the two (distinct) leftist heaps  $T_1$  and  $T_2$

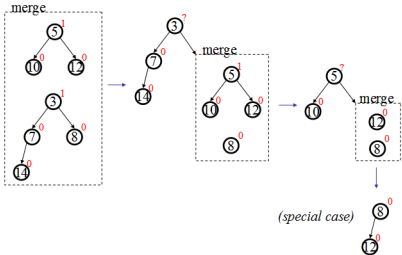


# Leftist Merge Continued

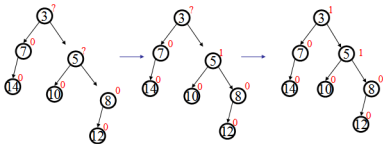


runtime:

# Leftist Merge Example

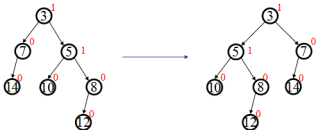


# Sewing Up the Leftist Example



Done?

## Finally...(Leftist)



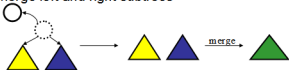


# Operations on Leftist Heaps

- merge with two trees of total size  $n$ :  $O(\log n)$
- insert with heap size  $n$ :  $O(\log n)$ 
  - pretend node is a size 1 leftist heap
  - insert by merging original heap with one node heap



- deleteMin with heap size  $n$ :  $O(\log n)$ 
  - remove and return root
  - merge left and right subtrees



# Random Definition: Amortized Time

am·or·tized time:

**Running time limit resulting from “writing off” expensive runs of an algorithm over multiple cheap runs of the algorithm, usually resulting in a lower overall running time than indicated by the worst possible case.**

If  $M$  operations take total  $O(M \log N)$  time,  
*amortized* time per operation is  $O(\log N)$

Difference from **average time**: