



CSE332: Data Abstractions

Lecture 4: Priority Queues

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Spring 2012

A new ADT: Priority Queue

- Textbook Chapter 6
 - Will go back to binary search trees and hash tables
 - Nice to see a new and surprising data structure first
- A **priority queue** holds *compare-able data*
 - Unlike stacks and queues need to *compare items*
 - Given x and y , is x less than, equal to, or greater than y
 - Meaning of the ordering can depend on your data
 - Many data structures require this: dictionaries, sorting
 - Integers are comparable, so will use them in examples
 - But the priority queue ADT is much more general
 - Typically two fields, the *priority* and the *data*

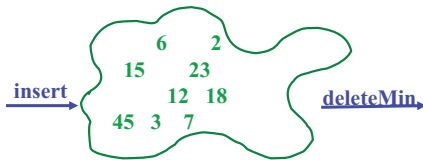
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Priorities

- Each item has a “priority”
 - The *lesser* item is the one with the *greater* priority
 - So “priority 1” is more important than “priority 4”
 - (Just a convention)
- Operations:
 - `insert`
 - `deleteMin`
 - `is_empty`
- Key property: `deleteMin` *returns* and *deletes* the item with greatest priority (lowest priority value)
 - Can resolve ties arbitrarily



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Example

```
insert x1 with priority 5
insert x2 with priority 3
insert x3 with priority 4
a = deleteMin // x2
b = deleteMin // x3
insert x4 with priority 2
insert x5 with priority 6
c = deleteMin // x4
d = deleteMin // x1
```

- Analogy: `insert` is like `enqueue`, `deleteMin` is like `dequeue`
 - But the whole point is to use priorities instead of FIFO

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Applications

- Like all good ADTs, the priority queue arises often
- Sometimes blatant, sometimes less obvious
 - Run multiple programs in the operating system
 - “critical” before “interactive” before “compute-intensive”
 - Maybe let users set priority level
 - Treat hospital patients in order of severity (or triage)
 - Select print jobs in order of decreasing length?
 - Forward network packets in order of urgency
 - Select most frequent symbols for data compression (cf. CSE143)
 - Sort (first `insert` all, then repeatedly `deleteMin`)
 - Much like Project 1 uses a stack to implement reverse

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More applications

- “Greedy” algorithms
 - Will see an example when we study graphs in a few weeks
- Discrete event simulation (system simulation, virtual worlds, ...)
 - Each event e happens at some time t , updating system state and generating new events e_1, \dots, e_n at times $t+t_1, \dots, t+t_n$
 - Naïve approach: advance “clock” by 1 unit at a time and process any events that happen then
 - Better:
 - *Pending events* in a priority queue (priority = event time)
 - Repeatedly: `deleteMin` and then `insert` new events
 - Effectively “set clock ahead to next event”

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Finding a good data structure

- Will show an efficient, non-obvious data structure
 - But first let's analyze some "obvious" ideas for n data items
 - All times worst-case; assume arrays "have room"

data	insert algorithm / time	deleteMin algorithm / time
unsorted array		
unsorted linked list		
sorted circular array		
sorted linked list		
binary search tree		

Need a good data structure!

- Will show an efficient, non-obvious data structure for this ADT
 - But first let's analyze some "obvious" ideas for n data items
 - All times worst-case; assume arrays "have room"

data	insert algorithm / time	deleteMin algorithm / time
unsorted array	add at end $O(1)$	search $O(n)$
unsorted linked list	add at front $O(1)$	search $O(n)$
sorted circular array	search / shift $O(n)$	move front $O(1)$
sorted linked list	put in right place $O(n)$	remove at front $O(1)$
binary search tree	put in right place $O(n)$	leftmost $O(n)$

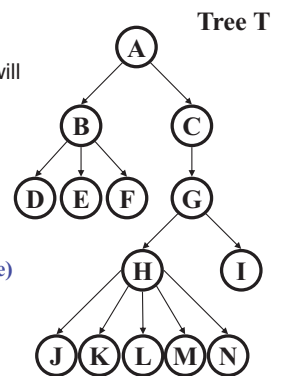
More on possibilities

- If priorities are random, binary search tree will likely do better
 - $O(\log n)$ insert and $O(\log n)$ deleteMin on average
- One more idea: if priorities are $0, 1, \dots, k$ can use array of lists
 - insert: add to front of list at `arr[priority]`, $O(1)$
 - deleteMin: remove from lowest non-empty list $O(k)$
- We are about to see a data structure called a "binary heap"
 - $O(\log n)$ insert and $O(\log n)$ deleteMin worst-case
 - Possible because we don't support unneeded operations; no need to maintain a full sort
 - Very good constant factors
 - If items arrive in random order, then insert is $O(1)$ on average

Tree terms (review?)

The binary heap data structure implementing the priority queue ADT will be a tree, so worth establishing some terminology

- root(tree)*
- leaves(tree)*
- children(node)*
- parent(node)*
- siblings(node)*
- ancestors(node)*
- descendants(node)*
- subtree(node)*
- depth(node)*
- height(tree)*
- degree(node)*
- branching factor(tree)*



Kinds of trees

Certain terms define trees with specific structure

- **Binary tree**: Each node has at most 2 children (branching factor 2)
- **n -ary tree**: Each node has at most n children (branching factor n)
- **Perfect tree**: Each row completely full
- **Complete tree**: Each row completely full except maybe the bottom row, which is filled from left to right

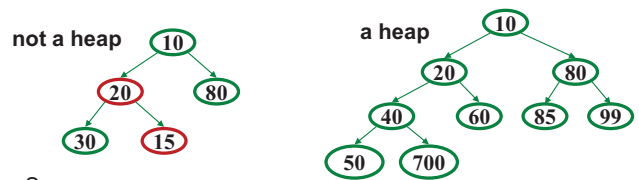


What is the height of a perfect tree with n nodes? A complete tree?

Our data structure

Finally, then, a *binary min-heap* (or just *binary heap* or just *heap*) is:

- **Structure property**: A complete binary tree
- **Heap property**: The priority of every (non-root) node is greater than the priority of its parent
 - Not a binary search tree

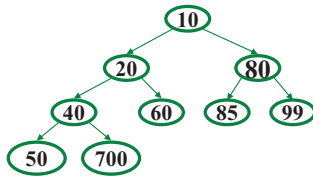


So:

- Where is the highest-priority item?
- What is the height of a heap with n items?

Operations: basic idea

- **findMin**: return `root.data`
- **deleteMin**:
 1. `answer = root.data`
 2. Move right-most node in last row to root to restore structure property
 3. "Percolate down" to restore heap property
- **insert**:
 1. Put new node in next position on bottom row to restore structure property
 2. "Percolate up" to restore heap property

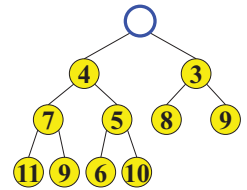


Overall strategy:

- Preserve structure property
- Break and restore heap property

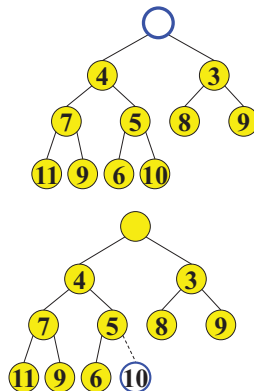
DeleteMin

1. Delete (and later return) value at root node

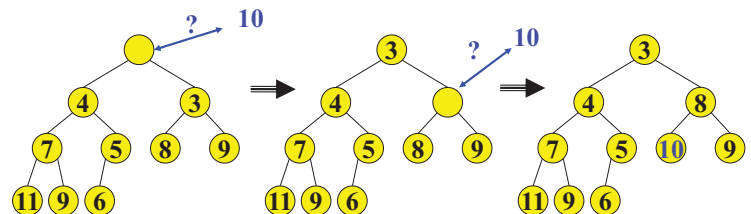


2. Restore the Structure Property

- We now have a "hole" at the root
 - Need to fill the hole with another value
- When we are done, the tree will have one less node and must still be complete



3. Restore the Heap Property



Percolate down:

- Keep comparing with both children
- Swap with lesser child and go down one level
- Done if both children are \geq item or reached a leaf node

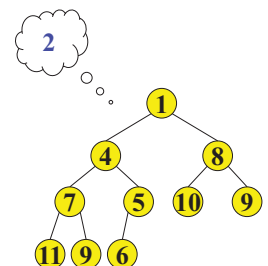
Why is this correct? What is the run time?

DeleteMin: Run Time Analysis

- Run time is $O(\text{height of heap})$
- A heap is a complete binary tree
- Height of a complete binary tree of n nodes?
 - height = $\lfloor \log_2(n) \rfloor$
- Run time of `deleteMin` is $O(\log n)$

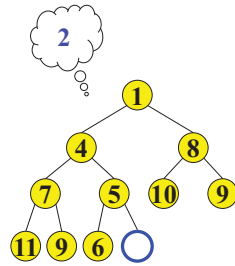
Insert

- Add a value to the tree
- Afterwards, structure and heap properties must still be correct

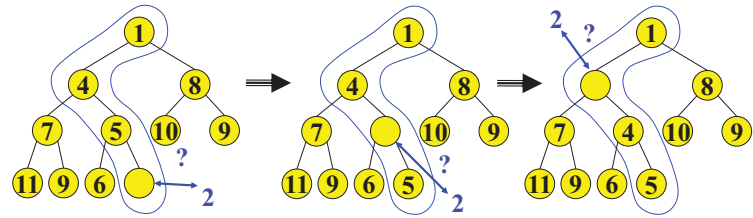


Insert: Maintain the Structure Property

- There is only one valid tree shape after we add one more node
- So put our new data there and then focus on restoring the heap property



Maintain the heap property



Percolate up:

- Put new data in new location
- If parent larger, swap with parent, and continue
- Done if parent \leq item or reached root

Why is this correct? What is the run time?

Insert: Run Time Analysis

- Like `deleteMin`, worst-case time proportional to tree height
 - $O(\log n)$
- But... `deleteMin` needs the “last used” complete-tree position and `insert` needs the “next to use” complete-tree position
 - If “keep a reference to there” then `insert` and `deleteMin` have to adjust that reference: $O(\log n)$ in worst case
 - Could calculate how to find it in $O(\log n)$ from the root given the size of the heap
 - But it's not easy
 - And then `insert` is always $O(\log n)$, promised $O(1)$ on average (assuming random arrival of items)
- There's a “trick”: don't represent complete trees with explicit edges!