Priority Queue

Recall the definition of a queue. It is a collection where we remove the element that has been in the collection for the longest time. Alternatively stated, we remove the element that first entered the collection. A natural way to implement such a queue was using a linear data structure, such as a linked list or a (circular) array.

A priority queue is a different kind of queue, in which the next element to be removed is defined by (possibly) some other criterion. For example, in a hospital emergency room, patients are treated not in a first-come first-serve basis, but rather the order may also or instead be determined by the urgency of the case. To define the next element to be removed, it is necessary to have some way of comparing any two objects and deciding which has greater priority. Once a comparison method is chosen for determining priority, the next element to be removed is the one with greatest priority. Heads up: with priority queues, one typically assigns low numerical values to high priorities. Think "my number one priority", "my number 2 priority", etc.

One way to implement a priority queue of elements (often called keys) is to maintain a sorted list. This could be done with a linked list or array list. Each time a element is added, it would need to be inserted into the sorted list. If the number of elements were huge, however, then this would be an inefficient representation since in the worst case the adds and removes would be $\Theta(n)$. [ASIDE: in future, you will find people saying O(n) to express what I wrote, loosely identifying 'worst case' with O(n).]

A second way to implement a priority queue would be to use a binary search tree. The element that is removed next is found by the findMinimum() operation. This would be a better way to implement a priority queue than the linear list method, since add and remove tend to take $\log n$ steps rather than n steps, if the tree is balanced. (I mentioned balanced binary search trees briefly in the last lecture. This topic will be covered in more depth – pun intended – in COMP 251.) One problem with using a balanced binary search tree for a priority queue is that it is overkill. In the next few lectures, we will look at a simpler data structure, called a heap.

Heaps

To define a heap, we first need to define a complete binary tree. We say a binary tree of height h is complete if every level l less than h has the maximum number (2^l) of nodes, and in level h all nodes are as far to the left as possible. A heap is a complete binary tree, whose nodes are comparable and satisfy the property that each node is less than its children. (To be precise, when I say that the nodes are comparable, I mean that the elements are comparable. And by that I mean that there is an ordering on all the elements.¹) This is the default definition of a heap, and is sometimes called a min heap.

A max heap is defined similarly, except that the element stored at each node is greater than the elements stored at the children of that node. Unless otherwise specified, we will assume a min heap in the next few lectures. Note that it follows from the definition that the smallest element in a heap is stored at the root.

¹In Section 002, one student asked about the rock-paper-scissors problem where elements are pairwise ordered but there is no global ordering. That is NOT what is happening here. Here we have a global ordering. I will talk about what I mean by Comparable later in the course.

As with stacks and queues, the two main operations we perform on heaps are add and remove.

add

To add an element to a heap, we create a new node and insert it in the next available position of the complete tree. If level h is not full, then we insert it next to the rightmost leaf. If level h is full, then we start a new level at height h+1.

Once we have inserted the new node, we need to be sure that the heap property is satisfied. The problem could be that the parent of the node is greater than the node. This problem is easy to solve. We can just swap the element of the node and its parent. We then need to repeat the same test on the new parent node, etc, until we reach either the root, or until the parent node is less than the current node. This process of moving a node up the heap, is often called "upheap".

```
add(element){
   cur = new node at next leaf position
   cur.element = element
   while (cur != root) && (cur.element < cur.parent.element){
      swapElement(cur, parent)
      cur = cur.parent
   }
}</pre>
```

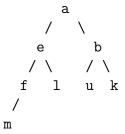
You might ask whether swapping the element at a node with its parent's element can cause a problem with the node's sibling (if it exists). It is easy to see that no problem exists though. Before the swap, the parent is less than the sibling.² So if the current node is less than its parent, then the current node must be less than the sibling. So, swapping the node's element with its parent's element preserves the heap property with respect to the node's current sibling.

For example, suppose we have a heap with two elements e and g. Then we add an element to the * position below and we find that * < e. So we swap them. But if * < e then * < g.

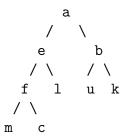
```
e
/ \
g *
```

²Here I say that one node is less than another, but what I really mean is that the element at one node is less than the element at the other node.

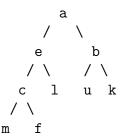
Here is a bigger example. Suppose we add element c to the following heap.



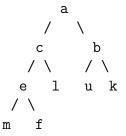
We add a node which is a sibling to m and assign c as the element of the new node.



Then we observe that c is less than the element f of its parent, so we swap c,f to get:



Now we continue up the tree. We compare c with the element in its new parent e, see that the elements need to be swapped, and swap them to get:



Again we compare c to its parent. Since c is greater than a, we stop and we're done.

removeMin

Next, let's look at how we remove elements from a heap. Since the heap is used to represent a priority queue, we remove the minimum element, which is the root.

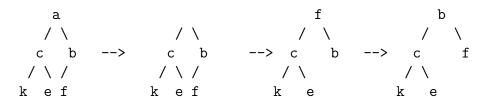
How do we fill the hole that is left by the element we removed? We first copy the last element in the heap (the rightmost element in level h) into the root, and delete the node containing this last element. We then need to manipulate the elements in the tree to preserve the heap property that each parent is less than its children.

We start at the root, which now contains an element that was previously the rightmost leaf in level h. We compare the root to its two children. If the root is greater than at least one of the children, we swap the root with the smaller child. Moving the smaller child to the root does not create a problem with the other child and with the heap property, since by definition the smaller child is greater than the larger child.

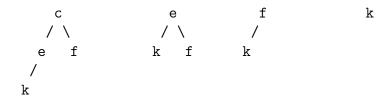
Here is a sketch of the algorithm:

The condition in the while loop is rather complicated, and you may have just skipped it. Don't. There are several possible events that can happen and you need to consider each of them. One is that the current node has no children. In that case, there is nothing to do. The second is that the current node has one child, in which case it is the left child. The potential problem here is that the left child might be smaller. The third is that the current node might have two children and one of these two children has to be smaller. (Are we guarenteed that one of the two children is smaller? See Exercises.)

Here is an example:

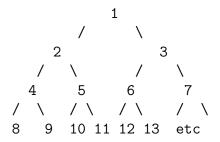


If we apply removeMin() again and again until all the elements are gone, we get the following sequence of heaps with elements removed in the following order: b, c, e, f, k.



Implementing a heap using an array

A heap is defined to be a complete binary tree. If we number the nodes of a heap by a level order traversal and start with index 1, rather than 0, then we get an indexing scheme as shown below.



These numbers are NOT the elements stored at the node, rather we are just numbering the nodes so we can index them. I will discuss the more next lecture.