

At the end of last lecture, I showed how to represent a heap using an array. Let's revisit the `add` and `removeMin` methods and express them using this representation.

`add(element)`

Suppose we have a heap with `k` elements which is represented using an array, and now we want to add a `k+1`-th element. Last lecture we sketched an algorithm doing so. Here I'll re-write that algorithm using the simple array indexing scheme. Let `size` be the number of elements in the heap. These elements are stored in array slots 1 to `size`, i.e. recall that slot 0 is unused so that we can use the simple relationship between a child and parent index.

```
add(element ){
    size = size + 1          // number of elements in heap
    heap[ size ] = element   // assuming array has room for another element
    i = size

    // the following is sometimes called "upHeap"

    while (i > 1 and heap[i] < heap[ i/2 ]){
        swapElements( i, i/2 )
        i = i/2
    }
}
```

Here is an example. Let's add `c` to a heap with 8 elements.

1	2	3	4	5	6	7	8	9	

a	e	b	f	l	u	k	m	c	
a	e	b	c	l	u	k	m	f	<---- c swapped with f (slots 9 & 4)
a	c	b	e	l	u	k	m	f	<---- c swapped with e (slots 4 & 2)

Building a heap

We can use the `add` method to build a heap as follows. Suppose we have a list of `size` elements and we want to build a heap.

```
buildHeap(list){
    create new heap array          // size == 0, length > list.size
    for (k = 0; k < list.size; k++){
        add( list[k] )             // add to heap[ ]
    }
}
```

Time complexity

How long does it take to build a heap in the best and worst case? Before answering this, let's introduce some notation. We have seen the "floor" operation a few times. Recall that it rounds down to the nearest integer. If the argument is already an integer then it does nothing. We also can define the ceiling, which rounds up. It is common to use the following notation:

- $\lfloor x \rfloor$ is the largest integer that is less than or equal to x . $\lfloor \cdot \rfloor$ is called the *floor* operator.
- $\lceil x \rceil$ is the smallest integer that is greater than or equal to x . $\lceil \cdot \rceil$ is called the *ceiling* operator.

For any positive integer $i \geq 1$, there is a unique integer – we call it *level* – such that

$$2^{\text{level}} \leq i < 2^{\text{level}+1}$$

or

$$\text{level} \leq \log_2 i < \text{level} + 1,$$

and so

$$\text{level} = \lfloor \log_2 i \rfloor.$$

In particular, if i is the index in the array representation of elements/nodes in a heap, then i is found at level *level* in the corresponding binary tree representation.

We can use this to examine the best and worst cases for building a heap. In the best case, the node i that we add to the heap satisfies the heap property immediately, and no swapping with parents is necessary. In this case, building a heap takes time proportional to the number of nodes n . So, best case is $\Omega(n)$.

What about the worst case? Since $\text{level} = \lfloor \log_2 i \rfloor$, when we add element i to the heap, in the *worst case* we need to do $\lfloor \log_2 i \rfloor$ swaps up the tree to bring element i to a position where it is less than its parent, namely we may need to swap it all the way up to the root. If we are adding n nodes in total, the worst case number of swaps is:

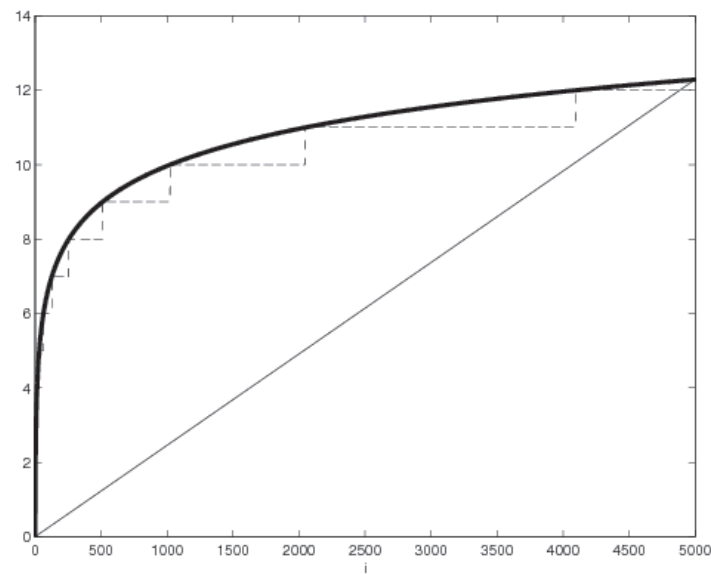
$$t(n) = \sum_{i=1}^n \lfloor \log_2 i \rfloor$$

To visualize this sum, consider the plot below which show the functions $\log_2 i$ (thick) and $\lfloor \log_2 i \rfloor$ (dashed) curves up to $i = 5000$. In this figure, $n = 5000$.

The area under the dashed curve is the above summation. It should be visually obvious from the figures that

$$\frac{1}{2}n \log_2 n < t(n) < n \log_2 n$$

where the left side of the inequality is the area under the diagonal line from (0,0) to $(n, \log_2 n)$ and the right side $(n \log_2 n)$ is the area under the rectangle of height $\log_2 n$. From the above inequalities, we conclude that in the worst of building a heap is $O(n \log_2 n)$. Indeed this is a upper tight bound and so we could say that this worst case $t(n)$ is $\Theta(n \log n)$. (It is a bit confusing to hear a "worst case" stated as $\Theta(n \log n)$ so in the lecture slides I only illustrated the $O(n \log n)$ bound, i.e. upper asymptotic bound.



removeMin

Next, recall the `removeMin()` algorithm from last lecture. We can write this algorithm using the array representation of heaps as follows.

```
removeMin(){
    element = heap[1]           // heap[0] not used.
    heap[1] = heap[size]
    heap[size] = null
    size = size - 1
    downHeap(1, size)          // see next page
    return element
}
```

This algorithm saves the root element to be returned later, and then moves the element at position `size` to the root. The situation now is that the two children of the root (node 2 and node 3) and their respective descendents each define a heap. But the tree itself typically won't satisfy the heap property: the new root will be greater than one of its children. In this typical case, the root needs to move down in the heap.

The `downHeap` helper method moves an element from a starting position in the array down to some maximum position in the heap. I will use this helper method in a few ways in this lecture.

```

downHeap(start,maxIndex){    // move element from starting position
                             // down to at most position maxIndex
    i = start
    while (2*i <= maxIndex){    // if there is a left child
        child = 2*i
        if (child < maxIndex) {    // if there is a right sibling
            if (heap[child + 1] < heap[child])
                // if rightchild < leftchild ?
            child = child + 1
        }
        if (heap[child] < heap[ i ]){ // swap with child?
            swapElements(i , child)
            i = child
        }
        else break // exit while loop
    }
}

```

This is essentially the same algorithm we saw last lecture. What is new here is that (1) I am expressing it in terms of the array indices, and (2) there are parameters that allow the downHeap to start and stop at particular indices.

Heapsort

A heap can be used to sort a set of elements. First, we take the elements and build a heap. At this point we are sure that the minimum element is the root. We then remove the minimum element from the heap, replace it by the last element and then downheap that element from the root. The size of the heap is reduced by 1 each time, and we use the freed slot in the array to store the removed element.

```

heapsort(list){
    buildheap(list)
    for i = 1 to size{
        swapElements( heap[1], heap[size + 1 - i])
        downHeap( 1, size - i)
    }
    return reverse(heap)
}

```

Note that after i times through the loop, the remaining heap has $\text{size} - i$ elements, and the last i elements in the array hold the smallest i elements in the original list. So, we only downheap to index $\text{size} - i$.

The end result is an array that is sorted from largest to smallest. If we want to order from smallest to largest, we simply reverse the order of the elements. This can be done in $\Theta(n)$ time, by swapping i and $n + 1 - i$ for $i = 1$ to $\frac{n}{2}$.

Example

The example below shows the state of the array after each pass through the for loop. The vertical line marks the boundary between the remaining heap (on the left) and the sorted elements (on the right).

1	2	3	4	5	6	7	8	9	

a	d	b	e	l	u	k	f	w	
b	d	k	e	l	u	w	f	a	(removed a, put w at root, ...)
d	e	k	f	l	u	w	b	a	(removed b, put f at root, ...)
e	f	k	w	l	u	d	b	a	(removed d, put w at root, ...)
f	l	k	w	u	e	d	b	a	(removed e, put u at root, ...)
k	l	u	w	f	e	d	b	a	(removed f, put u at root, ...)
l	w	u	k	f	e	d	b	a	(removed k, put w at root, ...)
u	w	l	k	f	e	d	b	a	(removed l, put u at root, ...)
w	u	l	k	f	e	d	b	a	(removed u, put w at root, ...)
w	u	l	k	f	e	d	b	a	(removed w, and done)

Note that the last pass through the loop doesn't do anything since the heap has only one element left (w in this example), which is the largest element. We could have made the loop go from $i = 1$ to $\text{size} - 1$.