

### **6.1.1 What are the minimum and maximum number of elements in a heap of height $h$ ?**

We assume the question is asking about a binary heap. Since the height is  $h$  there is a number  $k$  of elements such that if there is one more element the height is  $h + 1$  and there is a number  $j$  such that if we take one element of this heap our height will be  $h - 1$ . We know that each node has at most 2 children and this level indicates the height a heap has the following property, at height  $= 1$ , we consider the node at height  $= 1$ , it has 1 element. height 2 has at most 3 elements and at least 2 elements height 3 has at most 9 elements and minimum of 6 elements.

So it has at most  $2^h - 1$  elements and at minimum  $2^{h-1}$

---

### **6.1.2 Show that an $n$ - element heap has height $\lceil \lg n \rceil$**

We showed that in a heap of height  $h$  there are at least  $2^{h-1}$  elements and  $2^h - 1$  elements at most. So:

$$2^{h-1} \leq n \leq 2^h - 1 \rightarrow h - 1 \leq \log(n) \leq \log(2^h - 1) \leq h \rightarrow$$

$h - 1 \leq \log(n) \leq h \rightarrow$  adjust for convention in which we begin from height 0

$h \leq \lg(n) < h + 1 \rightarrow h$  is an integer so it can only be  $\lceil \lg n \rceil$

---

### **6.1.3 Show that in any subtree of a max - heap, the root of the subtree contains the largest value occurring anywhere in that subtree**

This is clearly what guarantees the max - heap property since the parents are greater than the children, by the transitivity property of  $>$  we know this is true.

Suppose that it's not, so there exists an element in that subtree, let denote it by  $S[j]$  such that  $S[j] > S[1]$ ,  $S[j]$  has a parent of at least 1 order, and the index of its parent is lower than its index, and we have  $A[\text{parentNode}(S[j])] < S[j]$  but this is a max heap and we need to have  $A[\text{parentNode}(S[j])] \geq S[j]$

---

### **6.1.4 Where in a max heap the smallest element resides, assuming that all elements are distinct?**

The minimum element must reside in a leaf, so without children, if it would have children

This would imply that their values are  $\leq$  this min, since they are different, it would imply that the children have a lower value than the min, what is a contradiction.

We cannot, although, get any more information based on that, it can be in any leaf since we don't have any relation between brothers

---

### **6.1.5 At which levels in a max - heap the $k$ - th largest element resides, for $2 \leq k \leq \left\lfloor \frac{n}{2} \right\rfloor$ , assuming that all elements are distinct?**

The largest element is clearly in 0 level, the second is clearly in the first level, the third can be at the first level or second. The deepest we can go is to achieve the  $k - 1$  level.

Furthermore it needs to be in a level that has at least  $k$  elements

$2^h \leq k$ . if  $k = 2$  we can go from at least level 1 to level 1.

if  $k = \left\lfloor \frac{n}{2} \right\rfloor \rightarrow 2^h \leq \left\lfloor \frac{n}{2} \right\rfloor \rightarrow h \leq \log \left( \left\lfloor \frac{n}{2} \right\rfloor \right) \rightarrow$  since it must be an integer

and it could go to  $\left\lfloor \frac{n}{2} \right\rfloor - 1$  of height. since there are  $n$  elements in our tree

The height of our tree is  $\lceil \lg n \rceil$  which is  $< \left\lfloor \frac{n}{2} \right\rfloor - 1$  so it can go everywhere except the root

---

**6.1.6 Is an array that is in sorted order a min heap?**

Yes, it clearly respect the condition to be a min heap

---

**6.1.7 Is the array with values (33, 19, 20, 15, 13, 10, 2, 13, 16, 12) a max heap?**

We can verify just by drawing, get the index beginning from 0

To be a max heap we need that  $\text{array}[i] \geq \text{array}[2i + 1], \text{array}[2i + 2]$

Element 15,  $i = 3$ , has element  $i = 7$  as child, 13, and element  $i = 8$  as child 16, which is larger than 15, so its not a max heap

---

**6.1.8 Show that, with the array representation for storing an  $n$  – element heap, the leaves are the nodes indexed by  $\left\lfloor \frac{n}{2} \right\rfloor + 1, \left\lfloor \frac{n}{2} \right\rfloor + 2 \dots n$**

Clearly after we find the exact separation, point where  $k$  is a leaf and  $k - 1$  its not the nexts elements are all leaves.

In this situation lets show that  $\left\lfloor \frac{n}{2} \right\rfloor$  is a node. since it's a node it must have a left child all nodes have a left node whose index is equal the index of the node times 2

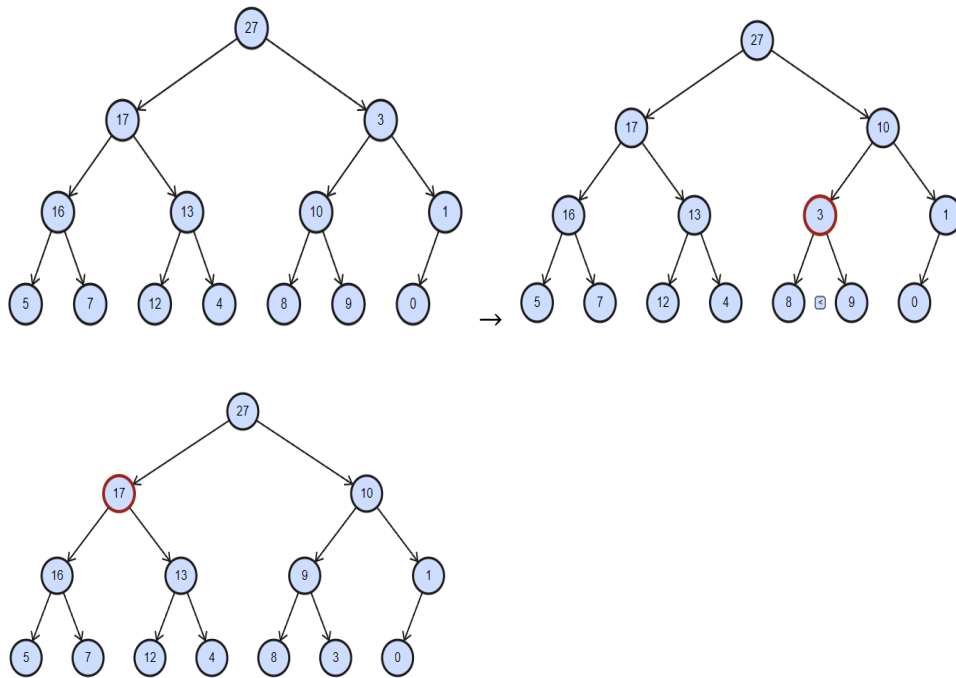
since the node is  $\left\lfloor \frac{n}{2} \right\rfloor < \frac{n}{2} \rightarrow 2 \left\lfloor \frac{n}{2} \right\rfloor \leq n$ , it's a node. Now let's show that

$\left\lfloor \frac{n}{2} \right\rfloor + 1$  has not a left child  $\rightarrow 2 \left\lfloor \frac{n}{2} \right\rfloor + 2 = 2 * \left( \left\lfloor \frac{n}{2} \right\rfloor + 1 \right) > 2 \left( \frac{n}{2} \right) = n$

So the index of the left child exceeds the actual number of elements

---

**6.2.1 Using figure 6.2 as a model, illustrate the operation of MAX – HEAPIFY (A, 3) on the array**  
**A = 27; 17; 3; 16; 13; 10; 1; 5; 7; 12; 4; 8; 9; 0**



**6.2.2 Show that each child of the root of an  $n$  – node heap is the root of a subtree containing at most  $\frac{2n}{3}$  nodes. What is the smallest constant  $\alpha$  such that each subtree has at most  $\alpha n$  nodes? How does that affect the recurrence 6.1 and its solution?**

We always create the children of the left node first, so the left nodes are the only ones who can get the maximum number. this always happen when we achieve a new height and we add only the necessary child to complete the first subtree of the left from the left child of the root. Lets assume we have a height  $H$ . the number of maximum elements are  $2^{h+1} - 1$ , and we have the maximum leaves as  $2^h, 2^{h-1}$  for the left subtree and  $2^{h-1}$  for the right subtree, above of that, if we take out all the leaves of this level our height go to  $h - 1$  and the number of total elements is  $2^h - 1$ .  $\rightarrow 2^h - 2$  the

subtree. So we have the following:  $\frac{\left(\frac{2^h - 2}{2} + 2^{h-1}\right)}{2^h - 1 + 2^{h-1}}$  ratio of left subtree

elements for height  $h$ . This function monotonic increasing, and the limit goes to  $\frac{2}{3}$

So the ratio is always  $< \frac{2}{3}$ .

In the situation above we forced the left subtree have more weight. Now we force them to be equal so in a height  $h$ , each one have  $(2^{h+1} - 2)/2$  in a total of  $2^{h+1} - 1 \rightarrow$

$\frac{2^{h+1} - 2}{2 \cdot (2^{h+1} - 1)}$ . The smallest constant is for  $n = 3$   $\alpha = \frac{1}{3}$ .

It doesn't affect it all since for  $\alpha < 1$  the solution is still  $O(\lg(n))$

**6.2.3 Starting with the procedure Max – Heapify, write pseudocode for the procedure MIN – HEAPIFY(A, i) which performs the corresponding manipulation on a min – heap. How does the running time of MIN – Heapify compare with that of MAX – heapify?**

*Its the same thing, just swap > and < the runing time is the same*

---

**6.2.4 What is the effect of calling MAX – HEAPIFY(A, i) when the element is larger than its children?**

*Nothing*

---

**6.2.5 What is the effect of calling MAX – HEAPIFY(A, i) for  $i > A.\text{heap-size}/2$**

*We are at the leaves, so nothing*

---

**6.2.6 The code for MAX – HEAPIFY is quite efficient in terms of constant factors, except possibly for the recursive call in line 10, for which some compilers might produce inefficient code. Write an efficient MAX – HEAPIFY that used an iterative control construct instead of a recursion:**

```
MAX – HEAPIFY(A, i):  
while  $i < \lfloor \frac{n}{2} \rfloor$ :  
     $l = \text{LEFT}(i)$   
     $r = \text{RIGHT}(i)$   
    if ( $l \leq A.\text{heap-size}$  and  $A[l] > A[i]$ )  
         $\text{largest} = l$   
    else  $\text{largest} = i$   
    if ( $r \leq A.\text{heap-size}$  and  $A[r] > A[\text{largest}]$ ):  
         $\text{largest} = r$   
    if  $\text{largest} \neq i$   
        swap  $A[i]$  with  $A[\text{largest}]$   
         $i = \text{largest}$   
    else:  
        break
```

---

**6.2.7 Show that the worst case running time of MAX – HEAPIFY on a heap of size  $n$  is  $\Omega(\lg(n))$ . (Hint: For a heap with  $n$  nodes, give node values that cause MAX – HEAPIFY to be called recursively at every node on a simple path from the root down to a leaf**

*Well. the hint gives everything, in the worst case you call it from the root and consecutively in all levels, until the highest height possible given  $n$  elements*

*With  $n$  elements the height is  $\lfloor \lg n \rfloor \rightarrow$  we have then  $\lfloor \lg n \rfloor + 1$  calls on heapify  
In this case MAX – HEAPIFY run exactly in  $\lfloor \lg n \rfloor + 1 \rightarrow \Theta(\lg(n)) \rightarrow \Omega(\lg(n))$*

---

**6.3.1 Using figure 6.3 as a model, illustrate the operation BUILDMAX – HEAP on the array  $A = (5; 3; 17; 10; 84; 19; 6; 22; 9)$**

Use the following Site with the array above as input

<http://btv.melezinek.cz/binary-heap.html>

---

**6.3.2 Show that  $\left\lceil \frac{n}{2^{h+1}} \right\rceil \geq \frac{1}{2}$  for  $0 \leq h \leq \lfloor \lg(n) \rfloor$**

For a complete binary tree we have that

$$2^h \leq n \leq 2^{h+1} - 1 \rightarrow 2^{h+1} \leq 2n \rightarrow 2^{h+1} > 2^{h+1} \rightarrow$$

$$2^h \leq n \leq 2^{h+1} - 1 < 2^{h+1} \leq 2n \rightarrow \frac{2^h}{2^{h+1}} \leq \frac{n}{2^{h+1}} \rightarrow \frac{1}{2} \leq \frac{n}{2^{h+1}} \leq \left\lceil \frac{n}{2^{h+1}} \right\rceil$$

---

**6.3.3 Why does the loop index  $i$  in line 2 of BUILD – MAX – HEAP decrease from  $\left\lceil \frac{n}{2} \right\rceil$**

We want to make a bottom – up approach so we want to go from the leaves to the root

---

**6.3.4 Show that there are at most  $\left\lceil \frac{n}{2^{h+1}} \right\rceil$  nodes of height  $h$  in any  $n$  –element heap**

we can show it by induction, we know that a heap with only a root has 1 element

A heap with two levels has at most 3 elements and at least 2 elements

A heap with three levels has at most 7 elements and at least 4 elements

in these cases we verify that we have 4, 2 and 1 as maximum number of leaves

Suppose for a  $n$  – heap, the number of leaves is at most  $2^{\lfloor \lg n \rfloor} = h \rightarrow$

We showed its valid for two, suppose its valid for  $h$ , now in the next level

So there are at most  $2^{\lfloor \lg n \rfloor}$ , if all these leaves have children

then we have  $2 \cdot 2^{\lfloor \lg n \rfloor}$  new leaves and height + 1. so  $2^{\lfloor \lg n \rfloor + 1}$  that is  $2^{h+1}$

Now, since there are at most  $2^{\lfloor \lg(n) \rfloor}$  leaves, this occurs at the last level.

Then the previous level,  $h - 1$ , there must be a maximum of  $2^{\lfloor \lg n \rfloor - 1}$  nodes

This is equivalent to the statement above, but we chose  $h$  as 0 from the root till  $h$  in the leaves

---

**6.4.1 Using figure 6.4 as a model, illustrate the operation Heapsort on the array  $A = 5; 13; 2; 25; 7; 17; 20; 8; 4$**

Use the following Site with the array above as input

<http://btv.melezinek.cz/binary-heap.html>

---

**6.4.2 Argue the correctness of HeapSort using the following loop invariant**  
**At the start of each iteration of the for loop of lines 2 – 5 the subarray**  
 **$A[1:i]$  is a max – heap containing the  $i$  smallest elements of  $A[1:n]$ , and the**  
**subarray  $A[i+1:n]$  contains the  $n-i$  largest elements of  $A[1:n]$ , sorted**

Lets verify the base case, where we are just before the first iteration,  $i = n$   
 $A[1:n]$  is a max – heap since we just called Build – Max – Heap on it.  $A[1:n]$  obviously  
 contain all elements and therefore it contains the  $n$  smallest elements of  $A[1:n]$  and  
 the subarray  $A[n+1:n]$  is empty and contains the 0 largest element of  $A[1:n]$

Now suppose it works for  $i$

Then  $A[1:i]$  is a max – heap containing the  $i$  smallest elements of  $A[1:n]$

$A[i]$  is  $\geq$  all other elements from  $A[1:i]$  so it makes him the  $n-i+1$  largest element

We then make a swap, what guarantess that  $A[i:n]$  is properly sorted, Now if it was a Max  
 heap then  $A[2]$  and  $A[3]$  were Max – Heaps two, because we just removed a child  
 from one of them, what maintains the property so if we apply Max – Heapify  
 in the first element transform  $A[1:i-1]$  in a heap, so it guarantees

For the termination when  $i$  would be 1 so  $A[1:1]$  is clearly a max – heap which is the  
 smallest element of the array by the previous iteration the following properties also holds

**6.4.3 What is the running time of HEAPSORT on an array  $A$  of length  $n$  that is already  
 sorted in increasing order? How about if the array is already sorted in decreasing order?**

Even though its in an increasing order we will re  
 necessity so the algorithm still is  $O(n \lg(n))$

**6.4.4 Show that the worst case running time of HEAPSORT is  $\Omega(n \lg n)$**

The build part is  $O(n)$  so we won't bother with it. At the worst case for each iteration of the loop  
 We will call MAX – HEAPIFY  $h$  times so we have:

$$\sum_{i=1}^{n-1} \lg(n-i) - 1 \leq \sum_{i=1}^{n-1} \lfloor \lg(n-i) \rfloor \leq \sum_{i=1}^{n-1} \lg(n-i)$$

$$\Theta(n \lg n) = \log(n-1!) - n + 1 \leq \sum_{i=1}^{n-1} \lfloor \lg(n-i) \rfloor \leq \log(n-1!) = \Theta(n \lg n)$$

So in the worst case it's still  $\Omega(n \lg n)$

**6.4.5 Show that when all elements of  $A$  distinct, the best – case running time of  
 HeapSort is  $\Omega(n \lg(n))$**

This showed to be a very tricky question and will be done later

**6.5.1 Suppose that the objects in a max – priority queue are just keys. Illustrate the operation of MAX – HEAP – EXTRACT – MAX on the heap 15; 13; 9; 5; 12; 8; 7; 4; 0; 6; 2; 1**

*We basically copy the  $A[0]$ , swap it with  $A[\text{size}]$  and apply a max – heap on it since we removed the last element, it swapped to be the first, we can still apply max – heapify and it will position in the right position again*

---

**6.5.2 Suppose that the objects in a max – priority queue are just keys. Illustrate the operation of MAX – HEAP – INSERT( $A, 10$ ) on the heap  $A = \{15, 13, 9, 5, 12, 8, 7, 4, 0, 6\}$**

*We basically insert 10 as the last element, all the other subtrees that aren't a direct parental lineage of this element is still a max – heap, so we can use a bottom up approach to rearrange the elements as needed*

---

**6.5.4 Write pseudocode for the procedure MAX – HEAP – DECREASEKEY( $A, x, k$ ) in a max – heap. What is the running time of your procedure?**

*Since we are decreasing the value of the key we just need to apply a max – heapify on the element. Its still  $\log(n)$*

---

**6.5.5 Why does MAX – HEAP – INSERT bother setting the key of the inserted object to  $-\infty$  in line 5 given that line 8 will set the object's key to the desired value**

*This is just a work round to assert that the increase method verify if its in fact increasing the value. Since the Insertion uses it as subroutine and doesn't change the value we needed to do this .*

---

**6.5.6 Professor Uriah suggests replacing the while loop of lines 5 – 7 in MAX – HEAP – INCREASE – KEY by a call to MAX – HEAPIFY. Explain the flaw in the professor idea**

*Here we are doing a bottom up approach, max – heapify uses a top – down approach Since the value were increased the object maintain itself being a max – heap, so the heapify method won't do anything*

---

**6.5.7 Argue the correctness of MAX – HEAP – INCREASE – KEY using the following loop invariant:**

**At the start of each iteration of the while loop of lines 5 – 7:**

**a. If both nodes  $\text{PARENT}(i)$  and  $\text{LEFT}(i)$  exist, then  $A[\text{PARENT}].\text{key} \geq A[\text{LEFT}(i)].\text{key}$ .**

***b. If both nodes  $PARENT(i)$  and  $RIGHT(i)$  exist, then  $A[PARENT(i)].key \geq A[RIGHT(i)].key$ .***

***c. The subarray  $A[1 : A.heap - size]$  satisfies the max – heap property, except that there may be one violation, which is that  $A[i].key$  may be greater than  $A[Parent(i)].key$***

***You may assume that the subarray  $A[1 : A.heap - size]$  satisfies the max – heap property at the time  $MAX - HEAP - INCREASE - KEY$  is called.***

*Initialization: since it passed from the catch error  $k$  is greater than the actual key  
Since it was max – heap just before the update the parent of  $i$  were greater than him that were greater than the left child. Since they didn't change values it still holds that parent  $>$  left. The same thing for the right child*

*Since we increased the value of  $i$  it still is bigger than left and right, if they exist.  
In this situation we could have increased the value of its parent, so there might be one violation that this cell is now greater than the parent*

*Maintenance, Since we do the swap and previously we already showed that the parent, that now is in  $i$  position, were greater than Left and Right, this still holds  
Since the swap occurred this violation was solved, but now there is possibly a new violation between the new  $i$  and its new parent, this is guaranteed by the  $i = Parent(i)$  assignment*

*Termination: if the node is the root it has no parent or we found a parent( $i$ ) greater than the actual  $i$ , if this case never occurs we will stop at the root because we have a clearly convergent sequence*

---

***6.5.8 Each exchange operation on line 6 of  $MAX - HEAP - INCREASE - KEY$  typically requires three assignments, not counting the updating of the mapping from the objects to array indices. Show how to use the idea of the inner loop of  $INSERTION - SORT$  to reduce the three assignments to just one***

*The idea is reuse  $x.key$  already saved in a variable, so during the while loop we just use the assignment  $A[i] = A[parent(i)]$  and after the loop  $A[i] = x$*

---

***6.5.9 Show how to implement a first – in, first – out queue with a priority queue. Show how to implement a stack with a priority queue***

*The idea is that always that you insert a new element it must have a high priority for the stack case, in this sense the priority can have a counter indicating how many objects there are and this will be associated with the priority  
So, just call insert with the actual value of this counter and increment it  
For the queue the idea is the opposite, just decrement it, or use a min – priority queue with the same strategy as said above*

---



**6.5.10 The Operation MAX – HEAP – DELETE( $A, x$ ) deletes the object  $x$  from max – heap  $A$ . Give an implementation of MAX – HEAP – DELETE for an  $n$  – element max – heap that runs in  $O(\lg(n))$  time plus the overhead for mapping priority queue objects to array indices**

We can swap the desired element with the last one, based on their values we decide if we will do a max – heapify or a bottom – up approach to fix the heap  
 So we make  $A[i] = A[A.size]$ . if  $A[i]$  were lower we do the bottom up approach  
 So we swap the  $A[i]$  with its parent if necessary until the heap property be okay  
 In the case  $A[i]$  were larger we just call max – heapify. We just do swaps equal the height

**6.5.11 Give an  $O(n \lg(k))$  – time algorithm to merge  $k$  sorted lists into one sorted list, where  $n$  is the total number of elements in all input lists. Hint (use a min – heap for  $k$  – way merging)**

We can maintain a min – heap of size  $k$ , we begin from the smallest elements of each list. We then proceed to extract this element what takes  $\lg(k)$   
 We insert another from the same list the one was extracted, what takes still  $\lg(k)$   
 We do this for each element, what leads us to deduce, without a proof although that the algorithm is  $n \lg(k)$

## **6.1 Building a heap using insertion**

One way to build a heap is by repeatedly calling MAX – HEAP – INSERT to insert the elements into the heap. Consider the procedure BUILD – MAX – HEAP' on the facing page. It assumes that the objects inserted are just the heap elements

```
BUILD-MAX-HEAP'(A, n)
1  A.heap-size = 1
2  for i = 2 to n
3      MAX-HEAP-INSERT(A, A[i], n)
```

**a. Do the procedures BUILD – MAX – HEAP and BUILD – MAX – HEAP' always create the same heap when run on the same input array? Prove that they do or provide a counter example**

**b. Show that in the worst case BUILD – MAX – HEAP' requires  $\Theta(n \lg n)$  time to build an  $n$  – element heap**

No, both procedures may fail to produce the same structure, but they do produce heaps  
 a counter example would be an array 1,2,3 → In this case 2 and 3 are leaves so  
 build max heap will call max – heapify on 1 and switch with 3 resulting in 321  
 Now we will insert 2 and 3 in the heap. two will begin as the root so 213 then we insert 3  
 Who will take the place from 2, so we finish with 312

We will realize  $n$  inserts. when we do an insert we add at the end of the heap, at this time we can make at maximum  $h$  swaps. so we have

$$\begin{aligned}\Theta(n \lg n) &= \sum_{i=2}^n c_i \lg(i) - c_i \leq \sum_{i=2}^n c_i \lfloor \lg(i) \rfloor \leq \sum_{i=2}^n c_i \lg(i) \leq \sum_{i=2}^n c_{\max} \lg(i) = c_{\max} \lg(n!) \\ &= \Theta(n \log n)\end{aligned}$$

---

**6.2 Analysis of a d ary heap** a d – ary heap is like a binary heap, but (with one possible exception) nonleaf nodes have d children instead of two children. In all parts of this problem, assume that the time to maintain the mapping between objects and heap elements is  $O(1)$  per operation

- Describe how to represent a d – ary heap in an array
- Using  $\Theta$  notation, express the height of a d ary heap of n elements in terms of n and d
- Given an efficient implementation of EXTRACT – MAX in a d ary max heap Analyze its running time in terms of d and n
- Given an efficient implementation of INCREASE – KEY in a d – ary max heap. Analyze its running time in terms of d and n
- Given an efficient implementation of INSERT in a d – ary max heap. Analyze its running time in terms of d and n

a) Let an array of Index 1,2,3,4 ... n. We would have the root as index 1 the k children from 1 + 1, until 1 + k. Then the k children of the first would have 2 + 1 ... 2 + k. So we have their children going from (k + 2)  $A[k^2 + k + 1]$

We could represent it like the D – ary parent of (i):

$$\left\lfloor \frac{i-1}{d} \right\rfloor + 1$$

And the j element of the element i as:

$$d(i-1) + j + 1$$

b) The first level have 1 element, the second have k, the third have  $k^2$ , the fourth have  $k^3$  at maximum, the height then is obviously  $\lceil \lg_k n \rceil$

c) So extract MAX will do the same thing as in the max – heap, we will swap it with the last element and perform a max – heapify. In this case we perform d comparisons at each level and we have  $a \lfloor d \log_d n \rfloor$  for the height  $\Theta(d \log_d n)$

d) So we will perform the same, given that we want to increase the value of the lowest to the larger, then we realize just a comparison between the new value and the value of the parent. if your new is bigger than swap, repeat until you are the root or the parent be bigger than you. In the worst case we have a  $\Theta(\log_d n)$

e) MAX – HEAP insert works the same and have the  $\Theta(\log_d n)$

---

### 6.3 Young tableaux

An  $m \times n$  Young tableau is an  $m \times n$  matrix such that the entries of each row are in sorted order from left to right and the entries of each column are in sorted order from top to bottom. Some of the entries of a Young tableau can be used to hold  $r \leq mn$  finite numbers

a) Draw a  $4 \times 4$  Young tableau containing the elements 9; 16; 3; 2; 4; 8; 5; 14; 12

b) Argue that an  $m \times n$  Young tableau  $Y$  is empty if  $Y[1:1] = \infty$ . Argue that  $Y$  is full (contains  $mn$  elements) if  $Y[m,n] < \infty$

c) Give an algorithm to implement EXTRACT – MIN on a nonempty  $m \times n$  Young tableau that runs in  $O(m + n)$  time. Your algorithm should use a recursive subroutine that solves  $m \times n$  problem by recursively solving either an  $m - 1 \times n$  or an  $m \times n - 1$  subproblem. (Hint: think about MAX – HEAPIFY). Explain why your implementation of EXTRACT – MIN runs in  $O(m + n)$

d) Show how to insert a new element into a nonfull  $m \times n$  Young tableau in  $O(m + n)$

e) Using no other sorting method as a subroutine, show how to use an  $n \times n$  Young tableau to sort  $n^2$  numbers in  $O(n^3)$  time

f) Given an  $O(m + n)$  – time algorithm to determine whether a given number is stored in a given  $m \times n$  Young tableau

a) Many ways, one of them:

2	3	9	12
4	8	14	$\infty$
5	$\infty$	$\infty$	$\infty$
16	$\infty$	$\infty$	$\infty$

b) Well we have that  $A_{11} \leq A_{1n}$  for all  $n$  and  $A_{11} \leq A_{m1}$  for all  $m$  and we know that  $A_{mn} \leq A_{jk}$  for all  $m \leq j$  given  $n = k$  or  $n \leq k$  for  $m = j$ . The argument, that could be proved by induction is that the rest of line 1 is  $\infty$

So all other elements below these must be  $\infty$ . if it less than  $\infty$  its finite, so its a number

c) Ok the minimum element is obviously the first, while the largest is the last  $A_{mn}$ . Since we will do a extract movement we need to change the value of the cell for  $\infty$ , but  $\infty$  is not suitable in that position. In this situation we will begin to make a similar MAX – HEAPIFY STRATEGY, we will compare the  $\infty$  position with its neighbours (the right and below one) and we will choose the lowest of them since the neighbours (the right and below one) must be greater than the actual position. So when we take the lowest one we guarantee that the position obeys the rules of the Young tableau. If we do  $n$  swaps in the columns and  $m$  swaps in the rows we put the  $\infty$  in the last position and all cells are obeying the rule

This takes obviously  $O(m + n)$ ,  $m$  swaps in the rows,  $n$  swaps in the column

d) If it's a non full tableau then the last element is for sure the  $\infty$  we then place this new element in this position we now realize the comparison with its neighbours, the left and top one and we choose with the highest, we proceed it until all cells obey the rule  
This is  $O(m + n)$

e) To insert a  $n$  element in a  $n^2$  Tableau we use  $O(2n)$  time, we insert  $n^2$  so  $n^2 O(n) = O(n^3)$ . We then use EXTRACT min for each  $n$ , so  $n^2 O(n) = O(n^3)$

f) Since the tableau is sorted by row and column we can use this information in our favor. We can begin from the top right position or from the bottom left. For the top right: In each iteration, if the element is lower than the one in the cell we reduce by 1 the column. If it is higher we sum 1 in the row so the Complexity is  $O(n + m)$ . Why this is true? If it is lower than the cell no other element in that column will be useful since they are greater. If it is greater than, no other element in that row will be useful, since they are all lower. We verify that we can eliminate a row or column at a time, so we reduce the problem to a  $m - 1 \times n$  or  $m \times n - 1$  subproblem that is guaranteed to have the value, if initially the tableau already had the value.