

**Homework 4**

**Instructions:** Please submit your solutions via Gradescope by **Friday, 18 February 2022, 10:00am**. Make sure your name, your class group number, and the name of your class teacher is put on **every page** of your submission. Your submission should, ideally, be a **PDF** file.

**Exercise 4.1: Master Theorem****3 pts**

Use the Master Theorem to find solutions to the following recurrences. Make sure that you state and justify which case of the Master Theorem you use.

(a)  $T(n) = 3 \cdot T(n/4) + n \log_2(n)$

(b)  $T(n) = 3 \cdot T(n/2) + n$

(c)  $T(n) = 3 \cdot T(n/3) + n/2$

**Exercise 4.2: HeapSort****1 pts**

In the lectures last week, we saw the HeapSort algorithm, which sorts a list in place and in worst-case time  $O(n \cdot \log_2(n))$ . Using the figures from the lecture as a model, illustrate the operation of BuildMaxHeap on the list

$$A = [5, 3, 17, 10, 84, 19, 6, 22, 9].$$

**Exercise 4.3: Randomised algorithms****3 pts**

Suppose you are given a list  $A$  of even length,  $\text{len}(A) = n$ , and you know that  $n/2$  of the entries are equal to  $x$ , and the other  $n/2$  entries are equal to  $y \neq x$ . You want to find  $x$  and  $y$ .

A simple deterministic algorithm (not relying on randomisation) for this problem that is guaranteed to find  $x$  and  $y$  proceeds as follows. First it sets  $x = A[0]$ . It then goes through the elements at positions  $i = 2, \dots, n-1$ . As soon as  $A[i] \neq x$ , the algorithm sets  $y = A[i]$  and outputs  $x$  and  $y$ .

- (a) Implement this algorithm in Python, and analyse its running time using big  $\Theta$ -notation.

Now consider the following randomised algorithm for this problem. Repeatedly draw a random entry (with replacement, so you may draw the same entry more than once) until you have seen two distinct numbers for the first time.

- (b) Implement this algorithm in Python.
- (c) What is the expected number of random draws performed by this algorithm?
- (d) Use your answer to (c) to analyse the expected running time of this algorithm using big  $O$ -notation.
- (e) Is this a Monte Carlo or a Las Vegas algorithm?

**Hint:** For part (c) it may be useful to consider the random variable  $Y$  which counts the number of random draws, and compute the probability that  $\Pr[Y > k]$ .

**Exercise 4.4: Binary Search****3 pts**

Given an arbitrary list, an algorithm searching for a specific target value in the list cannot do much better than scanning the list elements one by one, which requires  $\Omega(n)$  time in the worst case. On the other hand, if the algorithm is guaranteed to be given a sorted list (say, in non-decreasing order), it is possible to do drastically better. An algorithm, with a single comparison, can eliminate half of the entries of the list as possible locations. This is the idea underlying the “Binary Search” algorithm.

- (a) Implement a recursive Python function that searches for value  $x$  in the list  $A$ , which is sorted in non-decreasing order. Your function should return an index  $i$ , where  $A[i]=x$  (or the special Python value `None` if no such  $i$  exists) and have time complexity  $O(\log n)$ , where  $n$  is the length of the input list. (Do not use slicing, as it is a linear-time operation.)
- (b) Implement an iterative version of the binary search in Python, `BinarySearch(A, x)`, where  $A$ ,  $x$ , and the time complexity are the same as above.
- (c) Give a loop invariant for your algorithm in part (b). Prove the correctness of that algorithm using your loop invariant.

### Ex 4.1

$$(a) T(n) = 3 \cdot T(n/4) + n \log_2(n)$$

$$\log_b a = \log_4 3 < 1$$

Case 3:

$$\text{As } n \log_2 n = \Omega(n^{\log_4 3 + \epsilon}) \Rightarrow T(n) = \Theta(n \log_2 n)$$

$$(b) T(n) = 3 \cdot T(n/2) + n$$

$$\log_b a = \log_2 3 > 1 \quad \underline{\text{Case 1:}}$$

$$\text{So } n = O(n^{\log_2 3 - \epsilon}) \Rightarrow T(n) = \Theta(n^{\log_2 3})$$

$$(c) T(n) = 3 \cdot T(n/3) + n/2$$

$$\log_b a = \log_3 3 = 1 \quad \underline{\text{Case 2:}}$$

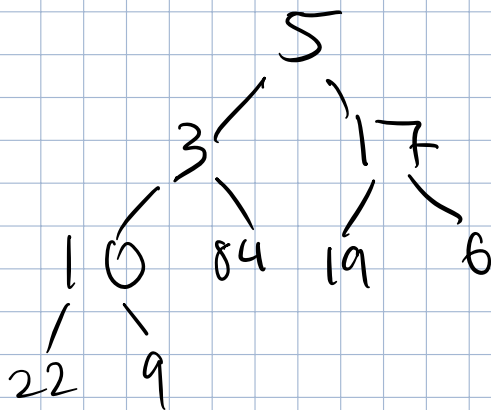
$$f(n) = n = \Theta(n^1) \Rightarrow T(n) = (n \lg n)$$

## Exercise 4.2: HeapSort

1 pts

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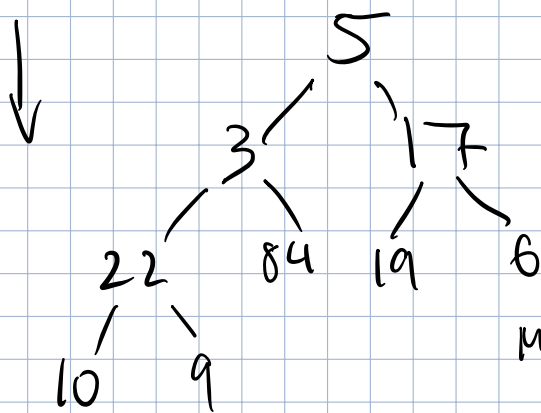
$A = [5, 3, 17, 10, 84, 19, 6, 22, 9]$ .



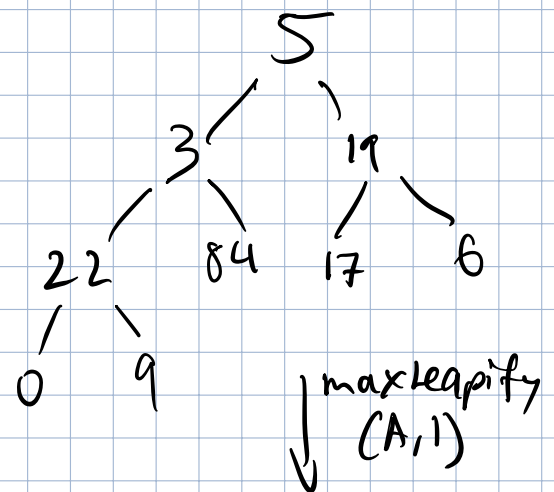
As  $\text{len}(A) = 9$

We only call max heapify from  $A[\text{len}(A)/2 - 1] \dots A[0]$

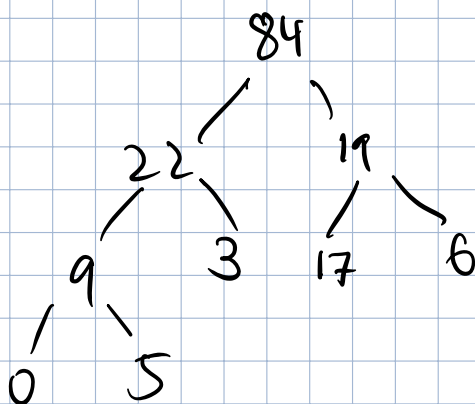
So start with  $\text{maxheapify}(A, 3)$



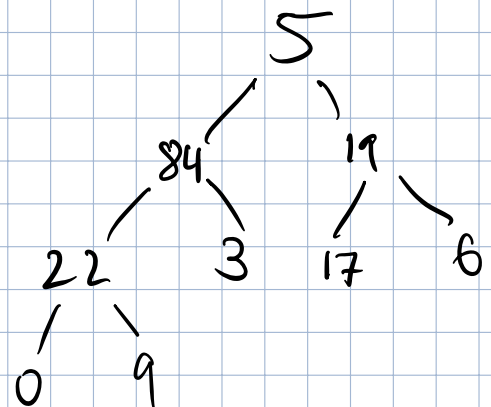
$\text{Maxheapify}(A, 2)$



$\text{maxheapify}(A, 1)$



$\text{Maxheapify}(A, 0)$



### Exercise 4.3: Randomised algorithms

3 pts

Suppose you are given a list  $A$  of even length,  $\text{len}(A) = n$ , and you know that  $n/2$  of the entries are equal to  $x$ , and the other  $n/2$  entries are equal to  $y \neq x$ . You want to find  $x$  and  $y$ .

A simple deterministic algorithm (not relying on randomisation) for this problem that is guaranteed to find  $x$  and  $y$  proceeds as follows. First it sets  $x = A[0]$ . It then goes through the elements at positions  $i = 2, \dots, n-1$ . As soon as  $A[i] \neq x$ , the algorithm sets  $y = A[i]$  and outputs  $x$  and  $y$ .

- (a) Implement this algorithm in Python, and analyse its running time using big  $\Theta$ -notation.

```
def findXAndY(A):  
    x = A[0]  
    for i in A:  
        if i != x:  
            return x, i
```

$\Theta(1)$   
 $\Theta(n/2)$   
 $\Theta(1)$   
 $\Theta(1)$

In the worst case half are on each side. Then

the algorithm will run  $\Theta(\frac{n}{2}) = \Theta(n)$  through the for loop.

Hence  $\Theta(n)$ .

Now consider the following randomised algorithm for this problem. Repeatedly draw a random entry (with replacement, so you may draw the same entry more than once) until you have seen ~~two~~ distinct numbers for the first time.

(b) Implement this algorithm in Python.

b)

```
def findXAndYRandom(A):  
    x = A[randint(0, len(A) - 1)]  
    new = x  
    while(new == x):  
        new = A[randint(0, len(A) - 1)]  
    return x, new
```

(c) What is the expected number of random draws performed by this algorithm?

Let  $X$  be the number of random draws.

$$\begin{aligned} E(X) &= \sum_{i=0}^{\infty} P(X=i) i \\ &= \sum_{i=0}^{\infty} \left(\frac{1}{2}\right)^i i = 1 \frac{1}{2} + 2 \frac{1}{2^2} + 3 \frac{1}{2^3} + 4 \frac{1}{2^4} \\ &= \frac{\frac{1}{2}}{\left(1 - \frac{1}{2}\right)^2} = \frac{\frac{1}{2}}{\frac{1}{4}} = 2 \end{aligned}$$

(d) Use your answer to (c) to analyse the expected running time of this algorithm using big O-notation.

As  $E(X) = 2$  this is not dependent on  $n$ ,  
so we have an algorithm of  $O(1)$ .

(e) Is this a Monte Carlo or a Las Vegas algorithm?

Las Vegas Algorithm as it will keep trying till we get the right answer.

#### Exercise 4.4: Binary Search

3 pts

Given an arbitrary list, an algorithm searching for a specific target value in the list cannot do much better than scanning the list elements one by one, which requires  $\Omega(n)$  time in the worst case. On the other hand, if the algorithm is guaranteed to be given a sorted list (say, in non-decreasing order), it is possible to do drastically better. An algorithm, with a single comparison, can eliminate half of the entries of the list as possible locations. This is the idea underlying the “Binary Search” algorithm.

- (a) Implement a recursive Python function that searches for value  $x$  in the list  $A$ , which is sorted in non-decreasing order. Your function should return an index  $i$ , where  $A[i]=x$  (or the special Python value `None` if no such  $i$  exists) and have time complexity  $O(\log n)$ , where  $n$  is the length of the input list. (Do not use slicing, as it is a linear-time operation.)

```
def recurseBinarySearch(A, start, end, x):
    if start > end:
        return None

    mid = (start + end) // 2
    if A[mid] == x:
        return mid
    elif A[mid] > x:
        return recurseBinarySearch(A, start, mid - 1, x)
    elif A[mid] < x:
        return recurseBinarySearch(A, mid + 1, end, x)
```

- (b) Implement an iterative version of the binary search in Python, `BinarySearch(A, x)`, where  $A$ ,  $x$ , and the time complexity are the same as above.

```
def iterativeBinarySearch(A, x):
    i = 0
    j = len(A) - 1
    while i <= j:
        mid = (i + j) // 2
        if A[mid] == x:
            return mid
        elif A[mid] > x:
            j = mid - 1
        elif A[mid] < x:
            i = mid + 1
    return None
```

(c) Give a loop invariant for your algorithm in part (b). Prove the correctness of that algorithm using your loop invariant.

(c) Loop invariant if  $x$  is in  $A$  and  $A[k] = x$   
then  $i \leq k \leq j$ .

Initialisation Clearly if  $x$  is in  $A$  then it must  
be in the range  $A[0] \dots A[\text{len}(A)-1]$

Maintenance: Suppose it is true at the start of the iteration.  
Choose the middle element.

▷ If  $\text{mid} = x$  we're done

▷ If  $\text{mid}$  is greater, then as the list is sorted  
it must be that  $i \leq k < \text{mid}$  so  $j = \text{mid} - 1$   
achieves this

▷ If  $\text{mid}$  is less, then as sorted, it must be  
that  $\text{mid} < k \leq j$  so  $i = \text{mid} + 1$  achieves  
this

Termination ▷ If we terminate early, we're done.

▷ Otherwise if we terminate, then  $i > j$

so then  $k$  must be in the range  $j \dots i$

but that is the empty set. So  $\text{None}$  is correctly  
returned.