

Homework Sheet 6

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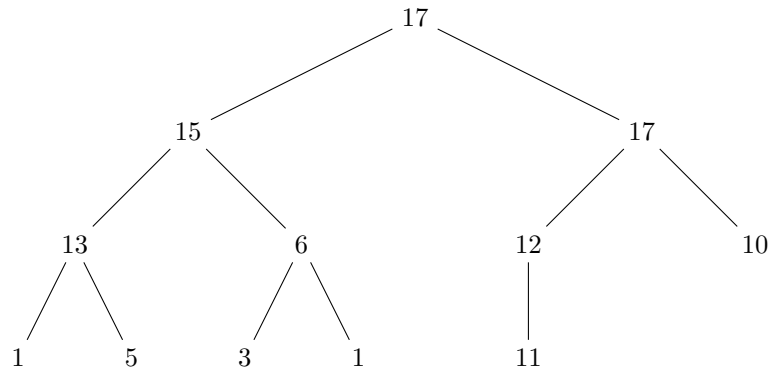
Exercise 1

(a)

We are given the array

$$A = [17, 15, 17, 13, 6, 12, 10, 1, 5, 3, 1, 11].$$

The heap tree would be



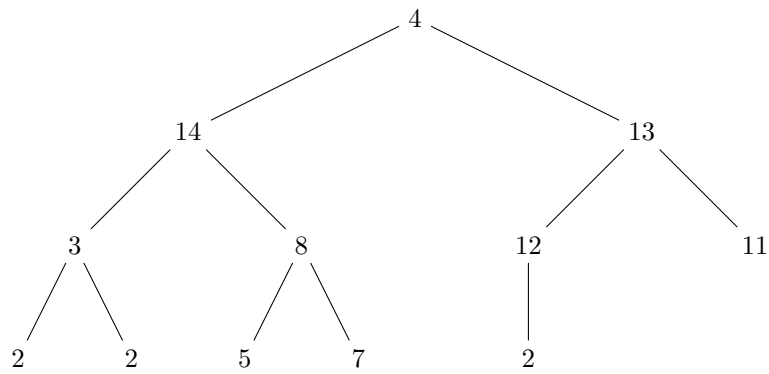
This is a heap since all nodes satisfy the max heap property.

(b)

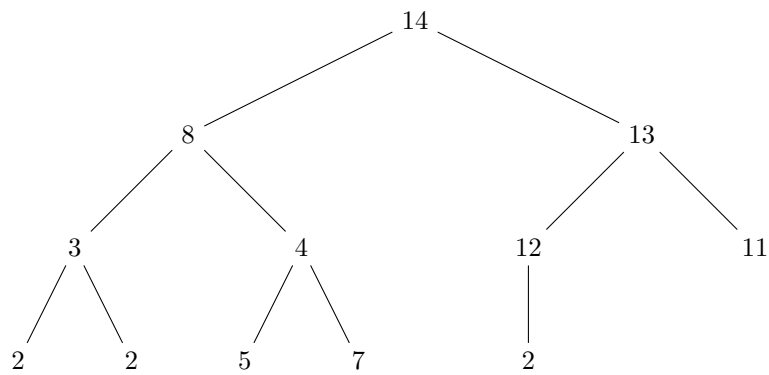
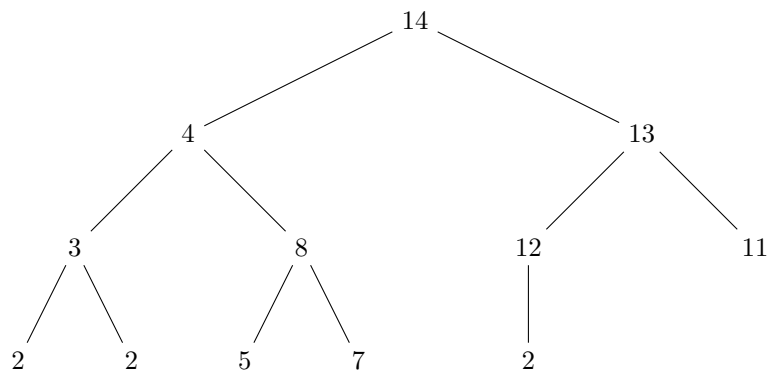
We are given the array

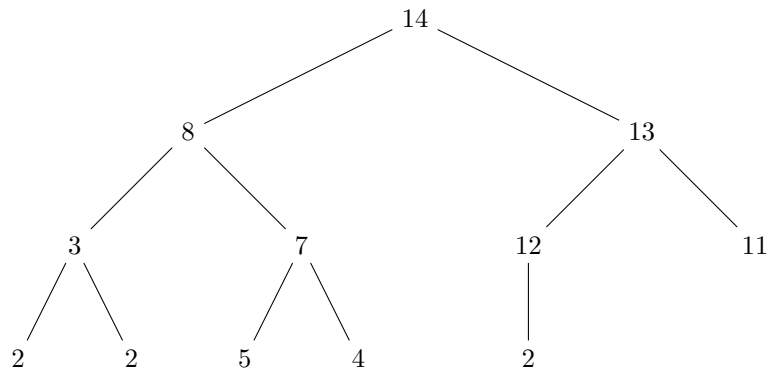
$$A = [4, 14, 13, 3, 8, 12, 11, 2, 2, 5, 7, 2].$$

The heap tree would be



This is a near heap because only the root node (4) violates the max heap property. Every other node satisfies it.
Steps of heapify operation:



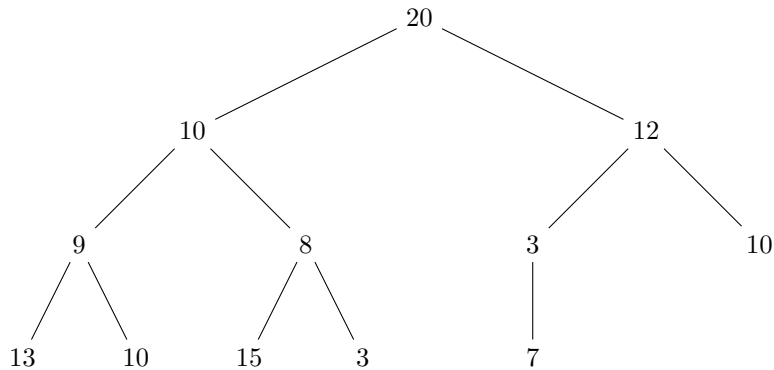


(c)

We are given the array

$$A = [20, 10, 12, 9, 8, 3, 10, 13, 10, 15, 3, 7].$$

The heap tree would be



This is not a heap also not a near heap because multiple nodes violate the max heap property. For example the node with the value 9 and the node with the value 8 both violates the max heap property and they are not descendants of each other.

Exercise 3

We are gonna use the same approach that we used in the lecture. We will build a max heap from the given array and then we will call `DeleteMax()` k many times.

```

int KthLargestElement(A[1..n], k)
    BuildMaxHeap(A)

```

```

    for i = 1 to k do
        result = DeleteMax(A)
    return result

void BuildMaxHeap(A[1..n]) // this is called makeheap() in the lecture slides
    for i = n/2 down to 1 do
        Heapify(A, i) // heapify function from the lecture

```

Running Time Analysis:

Building the max heap takes $O(n)$ time as we saw in the lecture. Each call to `DeleteMax()` takes $O(\log n)$ time. Since we are calling `DeleteMax()` k many times, this part takes $O(k \log n)$ time. Therefore the total running time of the algorithm is

$$O(n) + O(k \log n) = O(n + k \log n).$$

Correctness Proof:

The `BuildMaxHeap()` function builds a valid max heap from the given array A . In a max heap the maximum element is always at the root node. The `DeleteMax()` function removes and returns the maximum element from the heap and then reestablishes the max heap property by calling `Heapify()`. So at the time we call `DeleteMax()` for the k th time we get the k th largest element from the original array A and then we return it.

Exercise 4

We will maintain two heaps: a max heap H_{low} to store the lower half of the elements and a min heap H_{high} to store the upper half of the elements. The max heap H_{low} will allow us to efficiently retrieve the maximum element of the lower half. The min heap H_{high} will allow us to retrieve the minimum element of the upper half.

```

int[] RunningMedian(A[1..n])
    Hlow = new MaxHeap()
    Hhigh = new MinHeap() // the difference is the comparison function
    R = new int[1..n]
    for i = 1..n do
        if Hlow.isEmpty() or A[i] <= Hlow.Max() then
            Hlow.Insert(A[i])
        else
            Hhigh.Insert(A[i])

    // Balance the heaps
    if Hlow.size() > Hhigh.size() + 1 then
        Hhigh.Insert(Hlow.DeleteMax())
    else if Hhigh.size() > Hlow.size() then

```

```

        Hlow.Insert(Hhigh.DeleteMin())

    // Calculate median
    // either they are equal size or Hlow has one more element
    if Hlow.size() == Hhigh.size() then
        R[i] = Hhigh.Min()
    else
        R[i] = Hlow.Max()
    return R

```

Running Time Analysis:

Each insertion into a heap takes $O(\log m)$ time where m is the number of elements in the heap. Since we are inserting n elements the total time for insertions is $O(n \log n)$. Balancing the heaps involves at most one deletion and one insertion, which also takes $O(\log n)$ time. Since we do this for each of the n elements, the total time for balancing is also $O(n \log n)$. Calculating the median takes $O(1)$ time for each element, resulting in $O(n)$ time for all n elements. Therefore, the overall time complexity of the algorithm is $O(n \log n)$.

Correctness Proof:

The algorithm maintains two heaps to keep track of the lower and upper halves of the elements seen so far. By ensuring that the sizes of the heaps differ by at most one, we can determine the median after each insertion. If the heaps are of equal size, the median is the minimum of the upper half (Hhigh). If Hlow has one more element than Hhigh, the median is the maximum of the lower half (Hlow). This approach guarantees that we always have access to the median in $O(1)$ time after each insertion, and thus correctly computes the running median for each prefix of the array A .

Exercise 5

Assume train struct looks like this:

```

struct Train {
    int arrivalTime;
    int departureTime;
};

```

The idea is to sort the arrival and departure times of the trains separately and then use two pointers to traverse these sorted lists.

```

int MinPlatforms(Train L[1..n])
    arrivalTimes = new int[1..n]
    departureTimes = new int[1..n]

    for i = 1 to n do
        arrivalTimes[i] = L[i].arrivalTime

```

```

    departureTimes[i] = L[i].departureTime

    sort(arrivalTimes) // O(n log n)
    sort(departureTimes) // O(n log n)

    platformNeeded = 0
    maxPlatforms = 0
    i = 1 // pointer for arrivalTimes
    j = 1 // pointer for departureTimes

    while i <= n and j <= n do
        if arrivalTimes[i] < departureTimes[j] then
            platformNeeded++
            maxPlatforms = max(maxPlatforms, platformNeeded)
            i++
        else
            platformNeeded--
            j++

    return maxPlatforms

```

Running Time Analysis:

Sorting the arrival and departure times takes $O(n \log n)$ time each (for example assume MergeSort() we have seen in the lecture). The while loop runs in $O(n)$ time since each pointer (i and j) traverses the list of trains once. Therefore the overall time complexity of the algorithm is $O(n \log n)$.

Correctness Proof:

The algorithm sorts the arrival and departure times of the trains, allowing us to process events in chronological order. By using two pointers, we can track the number of platforms needed at any given time. When a train arrives before the next train departs, we need an additional platform. Conversely when a train departs before the next train arrives, we can free up a platform. The maximum number of platforms needed at any point during this process gives us the minimum number of platforms required for the train station.