Data Structures and Algorithms Dr. L. Rajya Lakshmi

Analysis of insertion sort

```
for j \leftarrow 1 to n-1 do
                                                                  n
                                                                  n-1
        key \leftarrow A[j]
         {insert A[j] into the sorted
                  sequence A[0. .j-1]}
                                                                  n-1
        i \leftarrow j-1
        while i \ge 0 and A[i] > \text{key do}
                 A[i+1] \leftarrow A[i]
        A[i+1] \leftarrow key
                                                                  n-1
```

Analysis of insertion sort

$$T(n) = n + (n-1) + (n-1) + \sum_{j=1}^{n-1} t_j + \sum_{j=1}^{n-1} (tj - 1) + \sum_{j=1}^{n-1} (tj - 1) + (n-1)$$

$$= n + 3(n-1) + (n-1)n/2 + (n-1)(n-2)/2$$

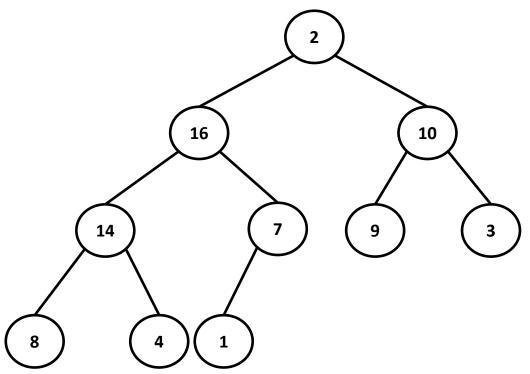
$$= an^2 + bn + c$$

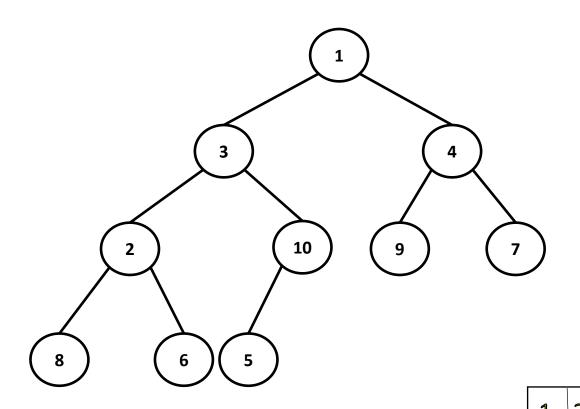
- O-notation describes an upper bound; when we use it to bound the worst-case running time of an algorithm, we have a bound on the running time of that algorithm on every input
- Does $\Theta(n^2)$ bound on the worst-case running time of insertion sort imply $\Theta(n^2)$ bound on every input?

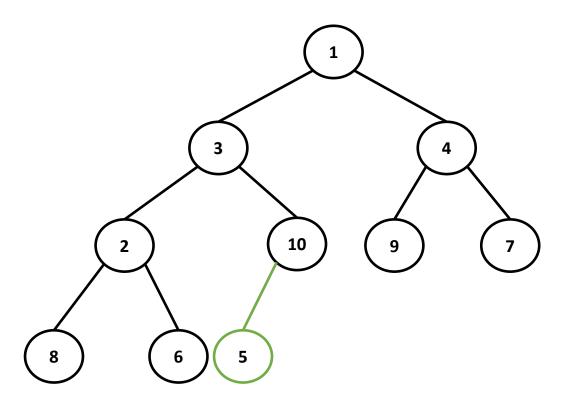
Selection sort

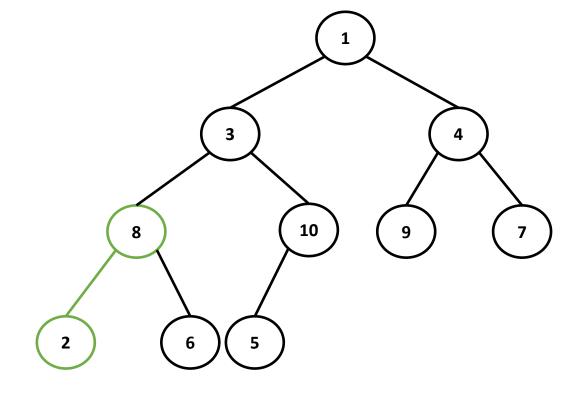
```
Algorithm Selection Sort(A[0..n-1], n)
         for i \leftarrow 0 to n-2 do
                  m \leftarrow i
                  for j \leftarrow i+1 to n-1 do
                           if A[i] < A[m]
                                    m \leftarrow i
                  swap A[i] and A[m]
• T(n) = \sum_{i=0}^{n-2} \sum_{i=i+1}^{n-1} c
        = \sum_{i=0}^{n-2} c(n-i-1) = c \sum_{i=0}^{n-2} (n-1) - c \sum_{i=0}^{n-2} i
         = c(n-1)(n-1) - c(n-2)(n-1)/2 = cn(n-1)/2
```

```
Algorithm Max_Heapify(A, i)
        I \leftarrow Left(i)
        r \leftarrow Right(i)
        if I \le A.heap_size and A[I] > A[i]
                largest ← I
        else
                largest ← i
        if r \le A.heap\_size and A[r] > A[largest]
                largest ← r
        if largest ≠ i
                swap A[i], A[largest]
                Max_Heapify(A, largest)
```



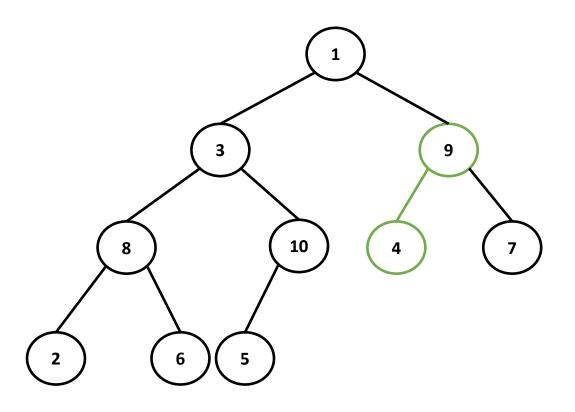


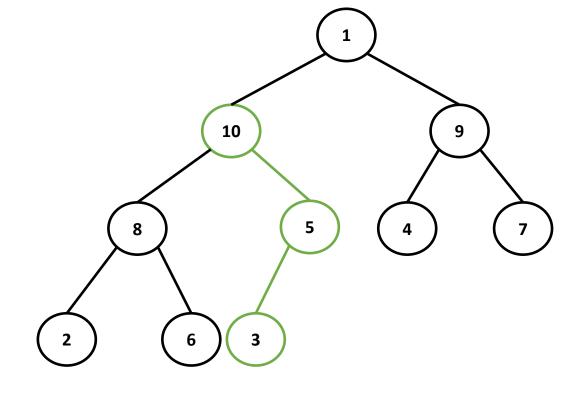




1 3 4 2 10 9 7 8 6 5

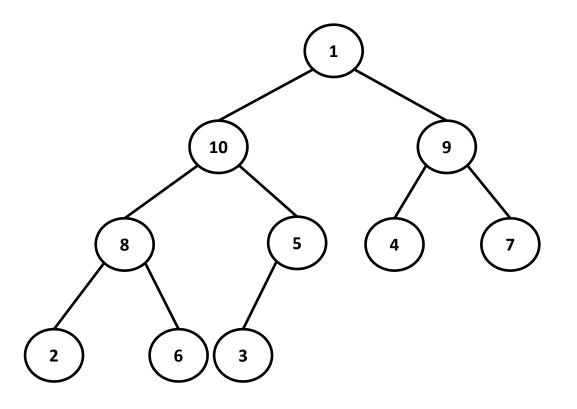
1 3 4 8 10 9 7 2 6 5	1	സ	4	8	10	9	7	2	6	5
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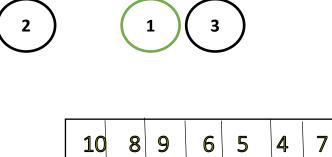




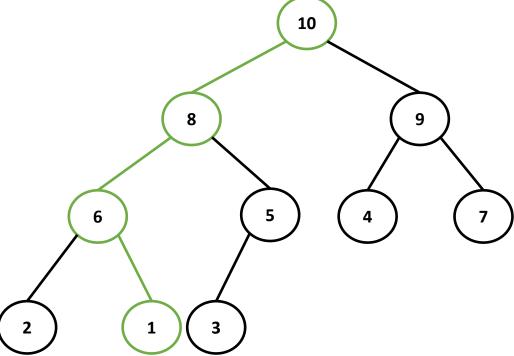
1	3	9	8	10	4	7	2	6	5
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1 10 9 8 5 4 7 2 6 3





1	10	9	8	5	4	7	2	6	3	
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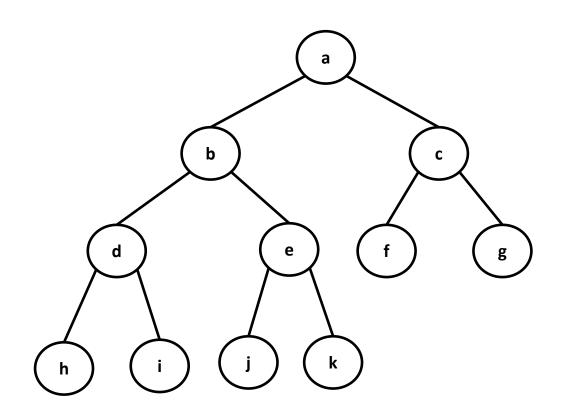


```
Algorithm Build_Max_Heap(A)

A.heap_size \leftarrow A.length

for i \leftarrow [A.lenght/2] downto 1

Max_Heapify(A, i)
```



- Time to run Max_Heapify at a node varies with the height of the node
- Time required to run Max_Heapify at a node of height h is O(h)
- An n element heap has height $\lfloor \log(n) \rfloor$
- At height "h", there would be at most [n/2^{h+1}] nodes
- The total cost of Build_Max_Heap:

$$\sum_{h=1}^{\lfloor \log(n) \rfloor} [n/2^{h+1}] O(h)$$
 which is $O(n \sum_{h=1}^{\lfloor \log(n) \rfloor} [h/2^h]$

$$\sum_{i=0}^{\infty} x^i = \frac{1}{(1-x)} \text{ if } |\mathbf{x}| < 1 \text{ {differentiate both sides}}$$

$$\sum_{i=1}^{\infty} i x^{i-1} = \frac{1}{(1-x)^2} \text{ {multiply both sides by x}}$$

$$\sum_{i=1}^{\infty} i x^i = \frac{x}{(1-x)^2}$$
Taking $\mathbf{x} = \frac{x}{2}$ yields,
$$\sum_{i=1}^{\infty} i x^i = \frac{1/2}{(1-1/2)^2} = 2$$

$$O(n \sum_{h=1}^{\lfloor \log(n) \rfloor} \lceil h/2^h \rceil \text{ is } O(n \sum_{h=1}^{\infty} \lceil h/2^h \rceil \text{ which can be written}$$
as $O(n \frac{1/2}{(1-\frac{1}{2})^2})$ which is $O(n)$

- Consider an array A[1 . . n], n is A.length
- Run Build_Max_Heap on A
- The largest element is sitting in A[1]
- Swap A[1] and A[n], decrement A.heap_size by 1, and run Max_Heapify on the new root
- Repeat the above step until A.heap size becomes 1

```
Algorithm Heapsort(A)

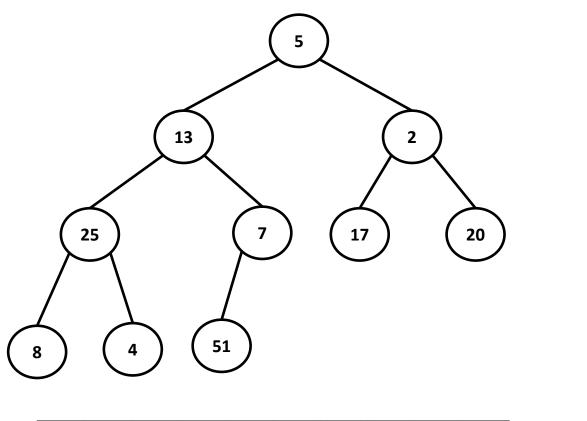
Build_Max_Heap(A)

for i ← A.length downto 2

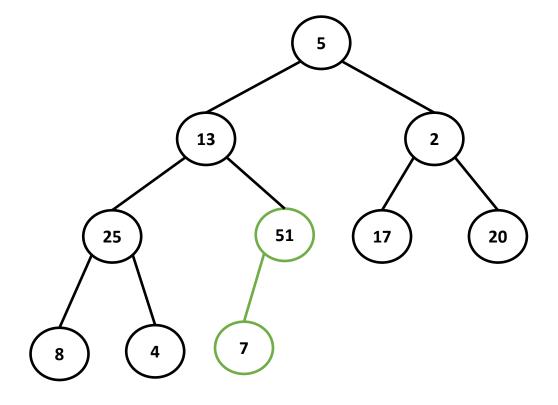
swap A[1] and A[i]

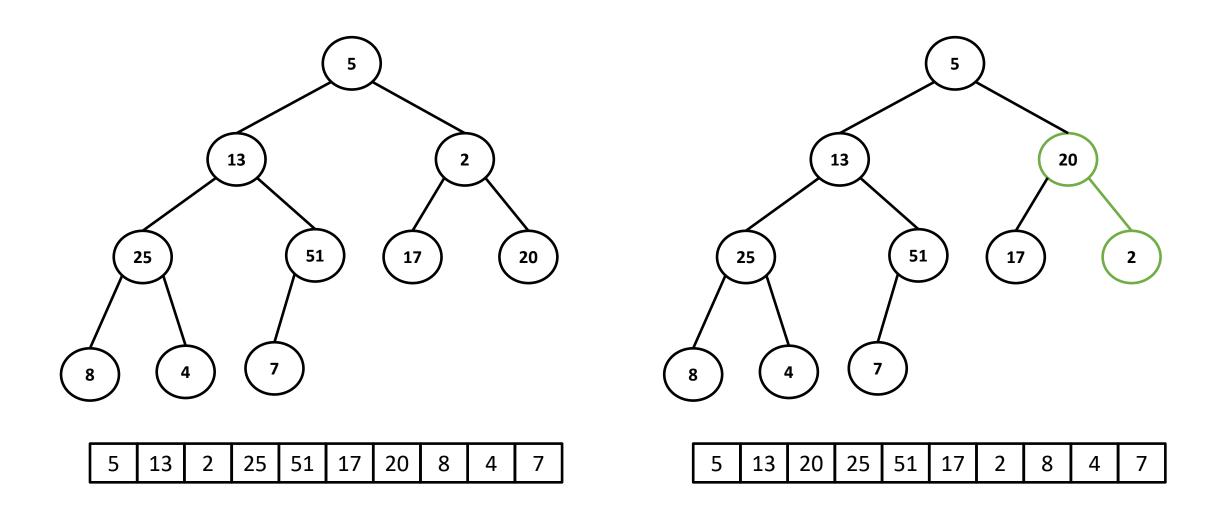
Max_Hepify(A, 1)
```

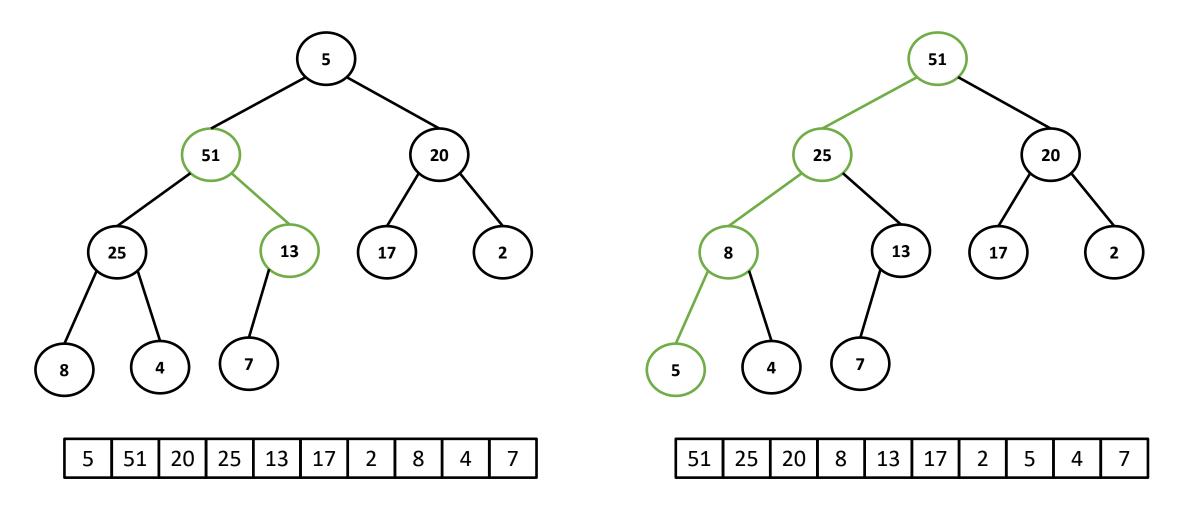
- Build_Max_Heap() takes O(n) time
- Each of n-1 calls of Max_Heapify() takes O(log n) time
- Running time is O(n log n)

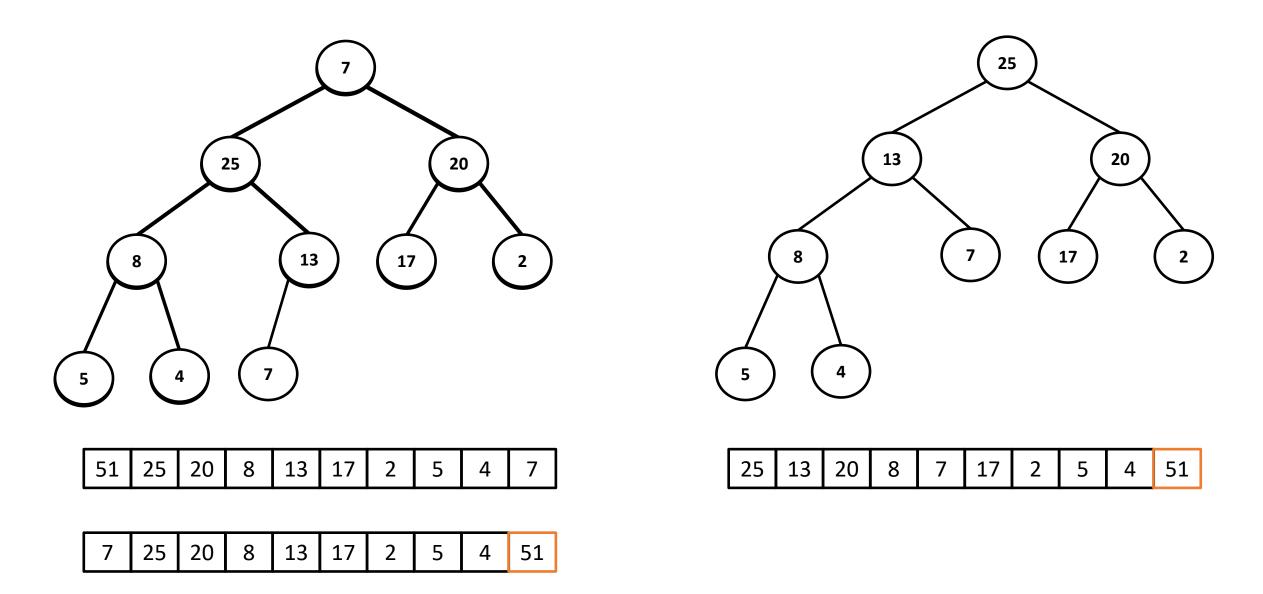


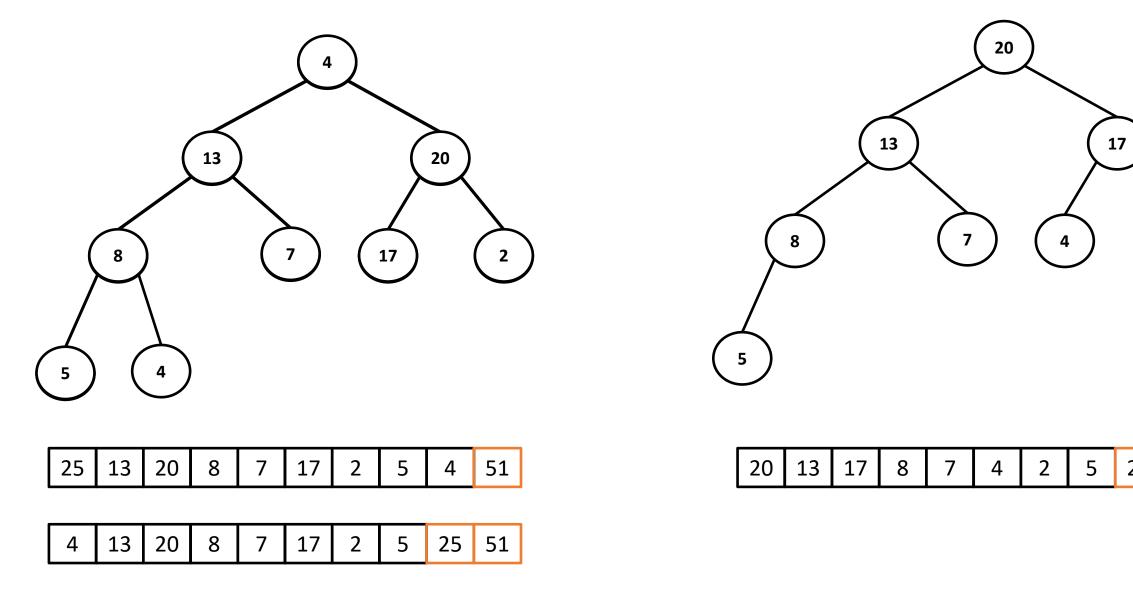


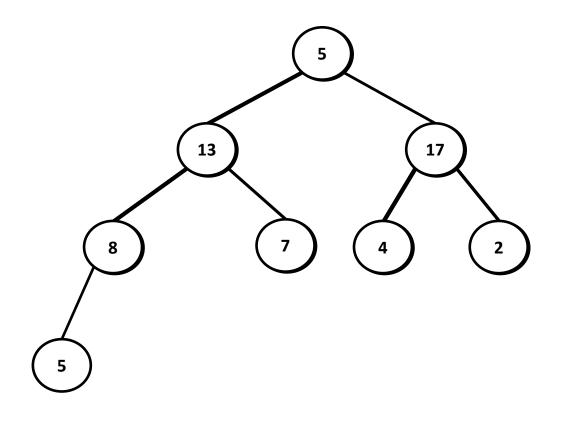


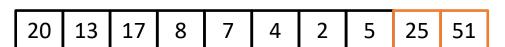




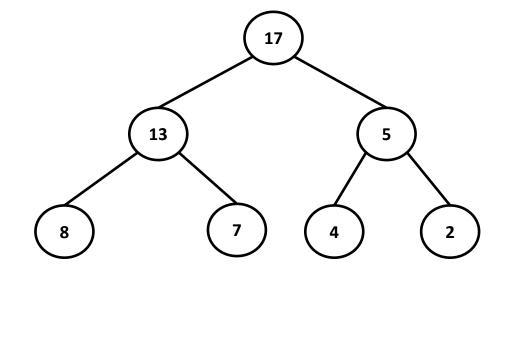


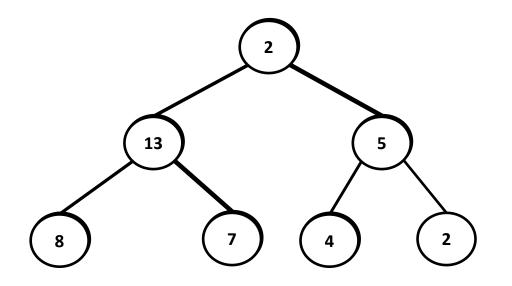


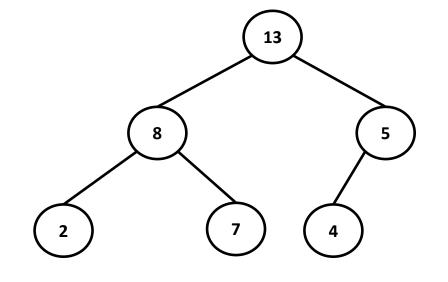


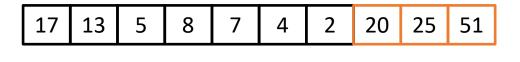




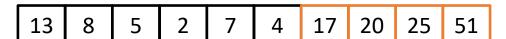


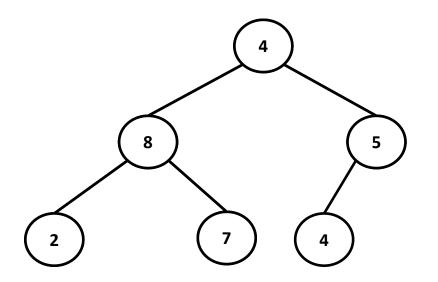


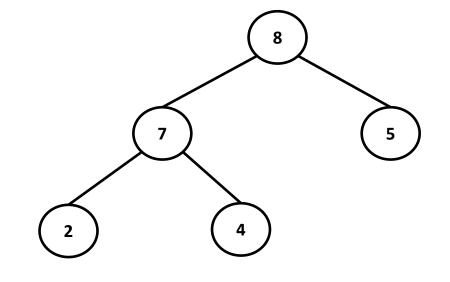


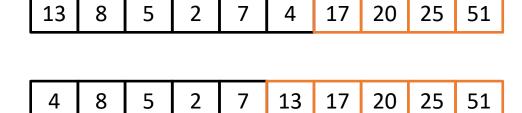




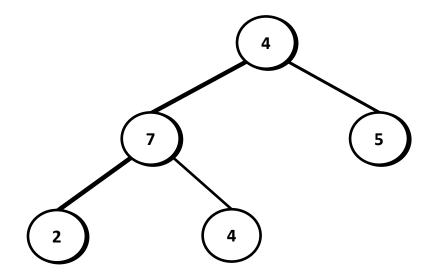


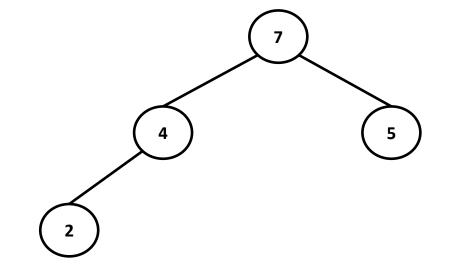


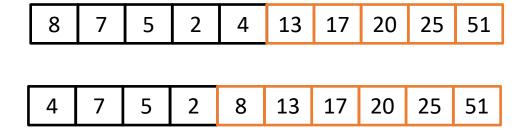


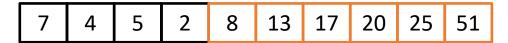


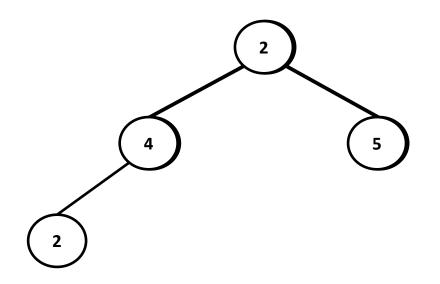


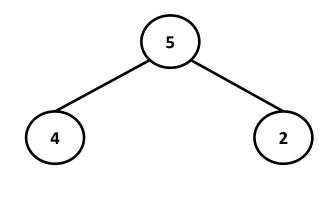








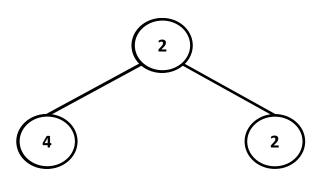


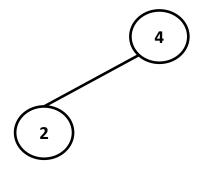


7 4 5 2	8	13	17	20	25	51
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4	5	7	8	13	17	20	25	51







5	4	2	7	8	13	17	20	25	51
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2 4 5 7 8 13 17 20 25 51

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4 2 5 7 8 13 17 20 25 51
```

