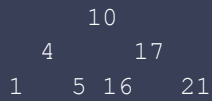
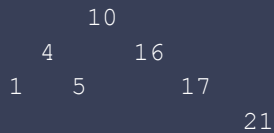


1. For the set of {1, 4, 5, 10, 16, 17, 21} of keys, draw binary search trees of height 2, 3, 4, 5, and 6.

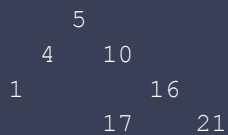
• Height 2



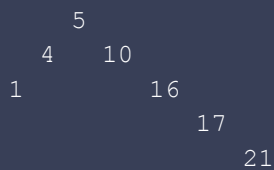
• Height 3



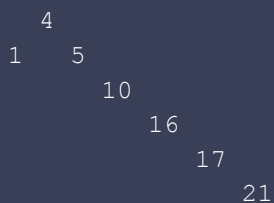
• Height 4



• Height 5

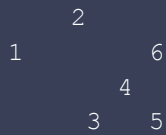


• Height 6



2. Suppose that the search for the key k in a binary search tree ends up in a leaf. Consider three sets: 1) the keys to the left of the search path; 2) the keys on the search path; 3) the keys to the right of the search path. One claims that any three keys a , b and c from these three sets, respectively (say a from 1, b from 2 and c from 3), it is always true that $a \leq b \leq c$. Is this claim true? Explain your answer.

No. Consider we search for 3 in the following tree. Then a in {1}, b in {2, 3, 4, 6}, c in {5}. It is clear we can pick $a = 1$, $b = 6$ and $c = 5$ which violate $a \leq b \leq c$.



3. We can sort a given set of n numbers by first building a binarysearchtree containing these numbers (using `Tree_insert` algorithm repeatedly to insert the numbers on by one) and then printing the number in an inorder tree walk. What are the worst-case and best-case running times for this sorting algorithm?

We know the running time for inserting is proportional to the hight of tree.

Therefore, in the best case, we have nearly balance tree and the hight of that tree is equal to $\log n$. Hence, the running time for best case is $\sum (\log i + d)$, which i from 1 to n , and d is const cost. Accordingly, the total cost is $O(n \log n)$.

For worst case, we have complete biased linear tree, When we insert the n th element to the tree, the hight of the tree is $n-1$. Hence, the cost is $\sum ((i-1) + d)$ which i from 1 to n , and d is const cost. Accordingly, the total cost is $O(n^2)$.

4. Suppose two teams A and B are playing a match to see who is the first to win n games (from given n). Assume that A and B are equally competent, so each has a 50% chance of winning any particular game. Suppose they have already played $i+j$ games, of which A won i and B has won j . Given an efficient algorithm to compute the probability that A will go on to win the match. For example, if $i=n-1$ and $j=n-3$ then the probability that A will win the match is $7/8$, since it must win any of the next three games.

Assume x, y be the remaining games team A, B need to win, then $x = n - i, y = n - j$.

$P(x, y)$ be the change the team A will win.

For $P(1, 1)$, we know team A has 0.5 to win the match, because A have 50% to win a game. For $P(1, 2)$, We know there is 0.5 that Team A will win next game and win the match, and 0.5 that Team A will lose the game and fallback to $P(1, 1)$ sceniario.

Accordingly, We can get formula that $P(1, 1) = 0.5 + 0.5 \cdot P(1, 2)$. We can use induction to get general relation $P(1, y) = 0.5 + 0.5 \cdot P(1, y+1)$. Due to the rule of symmetry, we also know that $P(2, 1)$ is also equal to $0.5 + 0.5 \cdot P(1, 1)$ as well.

In fact, we can get the overall recursion function,

1. $P(x, y) = 0.5 \cdot P(x-1, y) + 0.5 \cdot P(x, y-1)$
2. $P(1, 1) = 0.5$.

Or we can replace the parameter back to get

1. $P(n-i, n-j) = 0.5 \cdot P(n-i+1, n-j) + 0.5 \cdot P(n-i, n-j+1)$
2. $P(n-1, n-1) = 0.5$.

5. Give an $O(nt)$ algorithm for the following task. Input: a list of n positive integers a_1, a_2, \dots, a_n ; a positive integer t . Question: does some subset of the a_i 's add up to t ?

The recursion function is

1. $F(t, \text{Set}) = F(t-a_1, \text{Set}/a_1) \parallel F(t-a_2, \text{Set}/a_2) \parallel F(t-a_3, \text{Set}/a_3) \dots \parallel F(t-a_n, \text{Set}/a_n)$ for all $a_i \in \text{Set}$ and $a_i < t$.
2. $F(t, \emptyset) = \text{false}$.
3. $F(0, \text{Set}) = \text{true}$.

(Set/a_1 mean remove a_1 element from Set .)

$F(t-a_i, \text{Set})$ will be computed for all a_i in Set , and therefore is $O(n)$.

We repeat such recursion until we hit $F(0, \text{Set})$ or $F(t, \emptyset)$. And the worst case is repeat t times, so the overall time complexity is $O(n \cdot t)$.

6. Professor Midas drives an automobile from Newark to Reno along Interstate 80. His car's gas tank, when full, holds enough gas to travel n miles, and his map gives the distances between gas stations on his route. The professor wishes to make as few gas stops as possible along the way. Give an efficient method by which Professor Midas can determine at which gas stations he should stop, and prove that your strategy yields an optimal solution.

Assume the route start from x_0 (Newark), and end at x_n (Reno), and i gas stations are located in $x_1, x_2, x_3, \dots, x_i$, where x_i is between x_0 and x_n .

We can draw the picture like following.

```
x0 ---- x1 -- x2 ---- x3 --- x4 - ... --- xn-1 -- xn
```

Now we can define the dynamic recursion:

```
if D < n
    S(D) = min { S(x_i)+1, 0 } for all x_i < D, and (D - x_i) <= n
else
    S(D) = min { S(x_i)+1 } for all x_i < D, and (D - x_i) <= n
```

D is the target place, and S is the number of stops we get to place D . For $D < n$ we can choose not to stop at any gas stations, and arrival D .

For example:

```
x0 = 0, x1 = 4, x2 = 7, x3 = 9, x4 = 11, x5 = 14, x6 = 17, x7 = 19
n = 8, D is 20

S(20) = min { S(19)+1, S(17)+1, S(14)+1 }
S(19) = min { S(14)+1, S(11)+1 }
S(17) = min { S(14)+1, S(11)+1, S(9)+1 }
```

```
S(14) = min { S(11)+1, S(9)+1, S(7)+1 }
S(11) = min { S(9)+1, S(7)+1, S(4)+1 }
S(9) = min { S(7)+1, S(4)+1 }
S(7) = min { S(4)+1, 0 }
S(4) = min { 0 }
```

We can proof we get the optimal solution by induction. Let $OS(k)$ is the set of optimal solution for travel from x_0 to x_i for all $1 < i < k$, and $O(k)$ is the optimal solution for travel to x_k .

$OS(1) = \{O(1)\}$ and $OS(4) = \{O(1), O(2), O(3), O(4)\}$

Base case: if $k = 1$, $OS(1) = O(1)$ we know 0 is the solution for this problem.

Induction step: Assume we get the $OS(j)$ for $j < k$, we need to proof we can find the optimal solution for $OS(k)$ through this algorithm.

Because we check all the gas station that can be reached by x_k in distance n , say they are $x_i, x_{i+1}, \dots, x_{k-1}$, and we have the optimal solution for $O(i), O(i+1) \dots O(k-1)$, we know we can get all possible optimal solution for reaching x_k .

In algorithm, we choose the optimal solution among those subproblem, so we can guarantee we get the optimal solution for x_k . That is $O(k)$, and $OS(k) = OS(k-1) \cup \{O(k)\}$.