

PROBLEM SET 3

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Problem 1

- Find the median of an array:
 - Divide n elements into groups of 5
 - Find the median of each group
 - Recursively, find the median x of the $\lfloor n/5 \rfloor$ medians
 - Partition the n elements around x and drop most of the smallest and largest.
 - The time complexity of finding the median is:
 - $T(n) = T(\lfloor n/5 \rfloor) + T(7n/10 + 6) + O(n)$, which is guaranteed $T(n) \in O(n)$
- Since all we want is to use as less bags as possible, as well more chocolates. In other word, we want those heaviest chocolates. Therefore, we have this divide procedure:

Python

```
def buy(choco, target):
    if sum(choco) < target:
        return False # it's impossible to achieve the goal
    plan = []
    def divide(sub, target):
        median = get_median(sub) # This function is illustrated above
        heavy, light = [], []
        for w in sub:
            if w >= median:
                heavy.append(x)
            else:
                light.append(x)
        sum_heavy = sum(heavy)
        if 0 <= sum_heavy - target < median: # Reach the best deal
            return heavy
        elif sum_heavy - target < 0:
            target -= sum_heavy
            plan += heavy
            return divide(light, target)
        else:
            return divide(heavy, target)
    return plan += divide(choco, target)
```

The time complexity is $T(n) = T(n/2) + O(n) \in O(n)$

Problem 2: Binary Search Tree Practice

1. Firstly, let's specify some functions:

- `find_min`: go down to the leftmost node of the BST, this node has the smallest value in the BST or sub-BST.
- `find_next`: to find the successor that is only larger than the node in the BST or sub-BST:
 - If the node has a right child, then the leftmost node of the right sub-BST is the successor.
 - If the node has no right child, then:
 - If the node is its parent's left child, then the successor is its parent.
 - If the node is its parent's right child, then we keep following the pointers to parents until an ancestor has a larger value. Once we reached the root node, terminate the traversal.

Thus, from this naive procedure above, we shall see that we'll visit any node in the BST at most 3 times during the in-order traversal. Therefore, the time complexity $T(n) = n * \Theta(1) = O(n)$

2.

1. Python

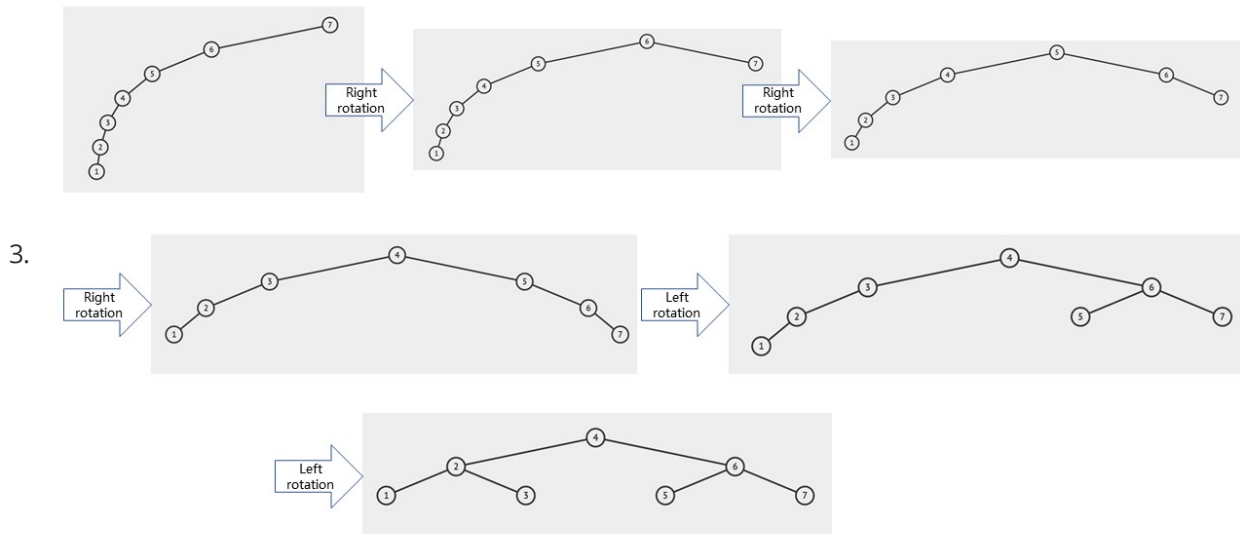
```
def isBST(self, root):  
    """  
    :type root: TreeNode  
    :rtype: bool  
    """  
    self.inorder = []  
    self.traverse(root)  
    return all(self.inorder[i] <= self.inorder[i+1] for i in  
               range(len(self.inorder)-1))  
  
def traverse(self, root):  
    if not root:  
        return  
    else:  
        self.traverse(root.left)  
        self.inorder.append(root.val)  
        self.traverse(root.right)
```

The time complexity $T(n) = O(n) + O(n) \in O(n)$

2. Python

```
def isBalanced(self, root):  
  
    def check(root):  
        if root is None:  
            return 0  
        left = check(root.left)  
        right = check(root.right)  
        if left == -1 or right == -1 or abs(left - right) > 1:  
            return -1  
        return 1 + max(left, right)  
  
    return check(root) != -1
```

The time complexity $T(n) = n * O(1) \in O(n)$



Problem 3: Strongly 2-Universal Hashing

1. If H is strongly 2-universal, then for every pair of keys (x, y) where $x \neq y$ and for every $i \in \{0, 1, \dots, m-1\}$, we have:

- $\Pr[(h(x), h(y)) = (i, i)] = \frac{1}{m^2}$, for every $h \in H$

Therefore, here are exactly m possible ways to make two keys x, y collide:

- $h(x) = h(y) = i$, for all $i \in \{0, 1, \dots, m-1\}$ and all $h \in H$
- Thus, $\Pr[h(x) = h(y)] = \sum_{i=0}^{m-1} \Pr[(h(x), h(y)) = (i, i)] = \frac{m}{m^2} = \frac{1}{m}$, for every $h \in H$

Thus, H is universal by definition.

2. A simple example with $m = |H| = |U| = 2$:

	\mathbf{x}	\mathbf{y}
h_1	0	0
h_2	1	0

It's easy to see the probability that two keys \mathbf{x}, \mathbf{y} collide is $1/2$ since only h_1 can cause the collision.

However, if \mathbf{H} is a strongly 2-universal hash family, for a randomly chosen hash function h , all the possible pairs of $(h(\mathbf{x}), h(\mathbf{y}))$ must be equally likely.

- In this case, they're $(0, 0), (0, 1), (1, 0), (1, 1)$.

But $(h(\mathbf{x}), h(\mathbf{y}))$ never equals to $(0, 1), (1, 1)$ in our \mathbf{H} .

Therefore, \mathbf{H} is not strongly 2-universal.

3. Firstly, I assume \mathbf{H} is a universal hash family. Thus we have $\Pr[h(\mathbf{x}) = h(\mathbf{y})] = \frac{1}{m}$, for every $h \in \mathbf{H}$

- I denote the index of $h \in \mathbf{H}$ by n , where $n \geq 0$, therefore $\mathbf{H} = \{h_0, h_1, \dots, h_n\}$
- Now we set a special \mathbf{x}_s as a probe:
 - For every $h_i \in \mathbf{H}$, we get $o = h_i(\mathbf{x}_s)$. If $o \neq i$, we set $h_i(\mathbf{x}_s) = i \bmod m$, and then we traverse the \mathbf{U} to find a key \mathbf{y} which satisfies that $h_i(\mathbf{y}) = i$. Once we get it, we set $h_i(\mathbf{y}) = o$, otherwise, just over.
- Now we have a new hash family \mathbf{H} , where we can index the target hash functions by \mathbf{x}_s :
 - $h_i(\mathbf{x}_s) = i \bmod m$, thus only $\lceil n/m \rceil$ hash functions are candidates. We can guarantee that the probability to make a collision to be greater than $\frac{1}{m}$ since we have the information of the target functions.
 - For all $h \in \mathbf{H}$, $\Pr[h(\mathbf{x}) = h(\mathbf{y})] = \frac{1}{m}$ for any pair of $(\mathbf{x}, \mathbf{y}), (\mathbf{x} \in \mathbf{U}, \mathbf{y} \in \mathbf{U}, \mathbf{x} \neq \mathbf{y})$ is still satisfied. Since I don't change the number of pairs of collision in the output domain of any hash function $h_i \in \mathbf{H}$.

Therefore, the new \mathbf{H} maintains its universality via the transformation and we guarantee the collide probability to be greater than $\frac{1}{m}$ as well.

4. Knowing $h(\mathbf{x})$ gives the adversary no information about $h(\mathbf{y})$ if h is randomly chosen from a strongly 2-universal hash family \mathbf{H} . We can format this problem like this:

- $\mathbf{X} = h(\mathbf{x}_s)$, where the \mathbf{X} is our information towards the hash function h .
- $\Pr[h(\mathbf{x}) = h(\mathbf{y}) | h(\mathbf{x}) = \mathbf{X}] = \frac{\Pr[h(\mathbf{x})=h(\mathbf{y}), h(\mathbf{x})=\mathbf{X}]}{\Pr[h(\mathbf{x})=\mathbf{X}]} = \frac{1/m^2}{1/m} = \frac{1}{m}$

Therefore, no matter which \mathbf{x}_s the adversary choose and which $\mathbf{X} = h(\mathbf{x}_s)$ the adversary knows, the probability of any particular \mathbf{y} colliding with \mathbf{x} is still $1/m$.