**Problem 1** Let A and B be arrays of n integers each (do not assume they are sorted). Given an integer x, describe a  $O(n \log n)$ -time algorithm for determining if there is an integer a in A and an integer b in B such that x = a + b.

Present pseudocode and analyze the running time.

## Solution

```
11 11 11
```

If Integer in A and B present such that a + b = x then we will return a, b else will return None, None.

- 1. Here sorting of a list and b list will take  $O(n \log n)$  time both
- 2. While loop will take O(n) time to traverse from start to end where start and end will both move towards one another
  - 3. Else will be constant time complexity c1

```
So total time complexity of the algorithm will be O(2 \ n \ \log n) + O(n) + O(c1) only bottleneck is O(n \ \log n) so total time complexity will be O(n \ \log n)
```

## def is\_sum\_present\_in\_2\_lists(a: [], b: [], x: int):

```
\# Sorting here will take O(n log n) time where n = number of elements in both list
```

2 a.sort()

# Sorting here will take O(n log n) time where n = number of elements in both list

# O(n) time complexity for the while loop till be go in between where start  $\geq$  end

```
6
        while start < len(a) and end > -1:
7
             if a[start] + b[end] == x:
8
                 return a[start], b[end]
9
             elif a[start] + b[end] < x:</pre>
10
                 start += 1
11
             else:
12
                 end -= 1
13
        return None, None
```

**Problem 2** Characterize each of the following recurrence equations using the master method (assuming that T(n) = c for n < d, for constants  $d \ge 1$ ).

#### Solution

According to master's Theorem

The Master Theorem: Let  $a \ge 1$  and b > 1 be constants, let f(n) be a function, and let T(n) be defined on the nonnegative integers by the recurrence

$$T(n) = aT(n/b) + f(n)$$

where we interpret n/b as  $\lfloor n/b \rfloor$  or  $\lceil n/b \rceil$ . Then T(n) can be bounded asymptotically as follows:

- 1. If  $f(n) = O(n^{\log_b a \epsilon})$  for some constant  $\epsilon > 0$ , then  $T(n) = \Theta(n^{\log_b a})$ .
- 2. If  $f(n) = \Theta(n^{\log_b a})$ , then  $T(n) = \Theta(n^{\log_b a} \log n)$ .
- 3. If  $f(n) = \Omega(n^{\log_b a + \epsilon})$  for some constant  $\epsilon > 0$ , and if  $af(n/b) \le cf(n)$  for some constant c < 1, and all sufficiently large n, then  $T(n) = \Theta(f(n))$ .

**a.** 
$$T(n) = 2T(n/2) + (n \log n)^4$$

### Solution

$$f(n) = (n \log n)^4$$
  
 $a = 2, b = 2$   
 $\log_b a = \log_2 2 = 1$   
 $n^{(\log_b a + e)} = n < n^4$  and  $e = 3 > 0$  here  
 $T(n) = theta(f(n)) = theta((n logn)^4)$ 

**b.** 
$$T(n) = 2T(n/2) + \log^2 n$$

### Solution

$$f(n) = log^{2}n$$

$$a = 2, b = 2$$

$$log_{b}a = log_{2}2 = 1$$

$$n^{(log_{b}a) > n^{0}$$

$$n^{(log_{b}a - e)} = n^{0} \text{ and } e = 1 > 0 \text{ here}$$

$$T(n) = theta(n^{log_{b}a}) = theta(n)$$

**c.** 
$$T(n) = 9T(n/3) + n^2$$

## **Solution**

$$f(n) = n^2$$
  
 $a = 9$ ,  $b = 3$   
 $log_b a = log_3 9 = 2$   
 $n^*(log_b a) = n^2 = f(n)$   
 $T(n) = theta(n^*(log_b a) logn) = theta(n^2 logn)$ 

**d.** 
$$T(n) = 9T(n/3) + n^3$$

## **Solution**

$$f(n) = n^3$$
  
 $a = 9, b = 3$   
 $log_b a = log_3 9 = 2$   
 $n^{(log_b a)} = n^2 < n^3 = n^{(log_b a + e)}$ , where  $e = 1$   
 $T(n) = theta(f(n)) = theta(n^3)$ 

**e.** 
$$T(n) = 7T(n/2) + n^2$$

# Solution

$$f(n) = n^2$$
  
 $a = 7$ ,  $b = 2$   
 $n^{1}\log_{b}a = n^{1}\log_{2}7 > n^2 = n^{1}(\log_{b}a - e)$ , where  $e > 0$   
 $n^{1}\log_{b}a > n^3$   
 $T(n) = theta(n^{1}\log_{b}a) = theta(n^{1}\log_{2}7)$ 

**Problem 3** Show that the running time of QUICKSORT is  $\Theta(n^2)$  when array A contains distinct elements and is sorted in decreasing order.

#### Solution

```
Let's say we have array of following numbers x_1 x_2 x_3 x_4 x_5 x_6 x_7 x_8 x_9 x_{10}
here numbers are in decreasing order so
X_1 > X_2 > X_3 > X_4 > X_5 > X_6 > X_7 > X_8 > X_9 > X_{10}
```

```
Quick Sort Algorithm will be like the following:
class QuickSort:
   def init (self, elements):
       self.elements = elements
   def quick sort(self):
       self. quick sort(0, len(self.elements) - 1)
   def quick sort(self, left: int, right: int):
       if left < right:
           pivot = self. partition(left, right)
            self. quick sort(left, pivot - 1)
           self. quick sort(pivot + 1, right)
   def partition(self, left: int, right: int):
       # we are taking all small elements to pivot to the left
       # of it and all big element to pivot to the right of it
       pivot = right
       divider = left
       for iterator in range(left, right):
            if self.elements[iterator] <= self.elements[pivot]:</pre>
                self.elements[divider], self.elements[iterator] =
self.elements[iterator], self.elements[divider]
                divider += 1
       self.elements[divider], self.elements[pivot] =
self.elements[pivot], self.elements[divider]
       print(self.elements[left:right])
       pivot = divider
       return pivot
Here in partition, the left-hand side of the pivot will be less
than the pivot and the right-hand side of the pivot will be
greater than the pivot.
First Pass
X_1 \quad X_2 \quad X_3 \quad X_4 \quad X_5 \quad X_6 \quad X_7 \quad X_8 \quad X_9 \quad X_{10}
pivot = x_{10}
```

```
partition = \mathbf{x}_{10} (pivot) \mathbf{x}_2 \mathbf{x}_3 \mathbf{x}_4 \mathbf{x}_5 \mathbf{x}_6 \mathbf{x}_7 \mathbf{x}_8 \mathbf{x}_9 \mathbf{x}_1
```

Here we can see that because the pivot is the smallest element in the list after partition it does not contain left elements which are lesser than it. So there will be no elements to sort on the left side of the pivot.

```
2nd Pass left no element right  x_2 \ x_3 \ x_4 \ x_5 \ x_6 \ x_7 \ x_8 \ x_9 \ x_1  pivot = x_1 partition = x_2 \ x_3 \ x_4 \ x_5 \ x_6 \ x_7 \ x_8 \ x_9 \ \textbf{x_1(Pivot)}
```

Here we can see that because the pivot is the largest element in the list after partition it does not contain the right elements which are greater than it. So there will be no elements to sort on the right side of the pivot.

```
3rd Pass
left
x_2 x_3 x_4 x_5 x_6 x_7 x_8 x_9
pivot = x_9
partition = x_9 (pivot) x_3 x_4 x_5 x_6 x_7 x_8 x_2
right no element
4th Pass
left no element
right
X<sub>3</sub> X<sub>4</sub> X<sub>5</sub> X<sub>6</sub> X<sub>7</sub> X<sub>8</sub> X<sub>2</sub>
pivot = x_2
partition = x_3 x_4 x_5 x_6 x_7 x_8 x_2 (pivot)
5th Pass
left
X_3 X_4 X_5 X_6 X_7 X_8
pivot = x_8
partition = x_8 (pivot) x_4 x_5 x_6 x_7 x_3
right no element
6th Pass
left no element
```

```
right
X<sub>4</sub> X<sub>5</sub> X<sub>6</sub> X<sub>7</sub> X<sub>3</sub>
pivot = x_3
partition = x_4 x_5 x_6 x_7 x_3 (pivot)
7th Pass
left
X<sub>4</sub> X<sub>5</sub> X<sub>6</sub> X<sub>7</sub>
pivot = x7
partition = x_7 (pivot) x_5 x_6 x_4
right no element
8th Pass
left no element
right
X<sub>5</sub> X<sub>6</sub> X<sub>4</sub>
pivot = x_4
partition = x_5 x_6 x_4 (pivot)
9th Pass
left
X<sub>5</sub> X<sub>6</sub>
pivot = x_6
partition = x_6 (pivot) x_5
right no element
10th Pass
left no element
right
X_5
pivot = x_5
partition = x_5 (pivot)
       Here time complexity of every pass of the partition will be
as following:
T(n) = T(n-1) + O(n)
      Because the pivot will get sorted to its position and
remaining n - 1 element will be on the right side of the pivot or
left side of the partition according to our partition algorithm.
Here O(n) time complexity will be of the partitioning algorithm.
    T(n) = theta(n + n-1 + n-2 + .... + 1)
      T(n) = theta(n^2)
```

**Problem 4** Describe an O(n)-time algorithm that, given a set S of n distinct numbers and a positive integer  $k \le n$ , determines the k numbers in S that are closest to the median of S.

Assume n is odd and the set S is given as an **unsorted** array of size n. You cannot assume the input array is sorted.

Example: if  $S = \{1, 3, 5, 9, 13, 21, 101\}$  and k = 4, the solution is  $\{3, 5, 9, 13\}$ . That is, the median itself is included. The answer  $\{5, 9, 13, 21\}$  is not correct since 3 is closer to the median (which is 9) than 21. The algorithm should write the output in a separate array, and the numbers **do not** have to be sorted.

You can use the selection algorithm as a subroutine. Precisely, assume that the following procedure is given: SELECT(A,p,q,i) returns (finds) the index j such that A[j] is the  $i^{th}$  smallest 1 number among A[p], A[p+1], . . . , A[q]. SELECT correctly runs in time O(q-p) even if the elements of A are not distinct. SELECT here is an extension of Quick-select(A,1,n,i) as in the notes, and only requires two extra tricks for implementation.

Partial credit will be given to correct algorithms, but with larger running time.

#### Solution

Here I have used Selection in the worst linear time algorithm found in the book chapter 9 Medians and Order Statistics and implemented **median\_kth\_smallest\_element**, which gives k\_th smallest element of the unsorted array in linear time O(n).

Main Algorithm is below:

11 11 11

first we will find median\_value which will take O(n) time To find distances to the median will take O(n) time kth\_distance from distances will take O(n) time k nearest to the k th distance will take O(n) time

So total time complexity will O(4\*n) = O(n)

```
1   def find_k_closet_to_median(elements: [], k: int):
2     mid = len(elements) // 2

3     median_index, median_value =
median kth smallest element(elements, 0, len(elements) - 1, mid)
```

```
distances to median = []
```

```
5
        for index in range(0, len(elements)):
            distances to median.append(abs(elements[index] -
median value))
        kth distance index, kth distance value =
median kth smallest element (distances to median, 0,
len(distances to median) - 1, k - 1)
8
        k nearest = []
9
        for index in range(0, len(elements)):
            distances to median = abs(elements[index] -
median value)
            if distances to median <= kth distance value:
12
                k nearest.append(elements[index])
13
        return k nearest
Supporting Algorithms is as follows:
11 11 11
    Here sorting will take O(5 log 5) time so time complexity is
O(1) and for loop will take O(n) time bcz
   all all inside it is of constant time so total time complexity
of this algo will be O(n)
def find medians(elements: [], left: int, right: int):
  medians = []
   for index in range(left, right, 5):
       if index + 5 < right:
           lis = elements[index: index + 5]
       else:
           lis = elements[index: right]
       lis.sort()
       mid = len(lis) // 2
       medians.append(lis[mid])
   return medians
11 11 11
  Algo will be O(n) time complexity by definition in the book
```

```
def median kth smallest element(elements: [], left: int, right:
int, k: int):
   if left <= right:</pre>
       medians = find medians(elements, left, right)
       number of median = len(medians)
       if number of median == 1:
           median of median = medians[0]
       else:
           median of median = median kth smallest element (medians,
left, number of median - 1,
number of median // 2)
       divider = element partition (elements, left, right,
median of median)
       if divider == k:
           return divider, elements[divider]
       elif k < divider:</pre>
           return median kth smallest element (elements, left,
divider - 1, k)
       else:
           return median kth smallest element (elements, divider +
1, right, k)
11 11 11
    Here first for loop will take O(n) time + last for loop will
take O(n) time rest will take constant time
  complexity so total time complexity will be O(n)
def element partition(elements: [], left: int, right:
                                                                int,
element: int):
   # we are taking all small elements to pivot to the left
   # of it and all big element to pivot to right of it
   index position = left
   for index in range(left, right):
       if elements[index] == element:
           index position = index
           break
     elements[index position], elements[right] = elements[right],
elements[index position]
   pivot = right
```

#### Problem 5

1. Draw the 11-item hash table resulting from hashing the keys 12, 44, 13, 88, 23, 94, 11, 39, 20, 16, and 5, using the hash function  $h(i) = (2i + 5) \mod 13$  and assuming collisions are handled by chaining. Here the hash table has 13 slots.

### Solution

```
h(k) = (2k + 5) \mod m

k = key to be inserted

m = number of slots = 13
```

```
Hashing
slot: 0 --> None
slot: 1 --> 11
slot: 2 --> 5 <--> 44
slot: 3 --> 12
slot: 4 --> None
slot: 5 --> 39 <--> 13
slot: 6 --> 20
slot: 7 --> None
slot: 8 --> None
slot: 9 --> None
slot: 10 --> None
slot: 10 --> None
```

2. What is the result of the previous exercise, assuming collisions are handled by linear probing? In the notation from the textbook, use h as defined above for the auxiliary function h'.

#### Solution

Here collisions are handled by hashfunction  $h\left(k,\ i\right)$  which is as following:-

```
h(k, i) = (h'(k) + i) \mod m

h'(k) = (2k + 5)

k = \text{key to be inserted}

m = \text{number of slots} = 13
```

```
Linear Probing
slot: 0 key 23
slot: 1 key 11
slot: 2 key 44
slot: 3 key 12
slot: 4 key 16
slot: 5 key 13
slot: 6 key 39
slot: 7 key 20
slot: 8 key 5
slot: 9 key None
slot: 10 key None
slot: 11 key 94
slot: 12 key 88
```

3. Show the result of Exercise 1 above in this problem, assuming collisions are handled by quadratic probing, up to the point where the method fails because no empty slot is found. In the notation from the textbook, use h as defined above for the auxiliary function h' and c1 = 0, c2 = 1. Also, the key will not be inserted if no empty slot is discovered with at most 10 probes.

#### Solution

Here collisions are handled by hashfunction h(k, i) which is as following:-

```
h(k , i) = (h'(k) + c1 * i + c2 * i^2) \mod m

h'(k) = (2k + 5)

k = \text{key to be inserted}

m = \text{number of slots} = 13
```

```
Quadratic Probing
slot: 0 key 23
slot: 1 key 11
slot: 2 key 44
slot: 3 key 12
slot: 4 key None
slot: 5 key 13
slot: 6 key 39
slot: 7 key 20
slot: 8 key None
slot: 9 key None
slot: 10 key 16
slot: 11 key 94
slot: 12 key 88
```

4. What is the result of Exercise 1 above in this problem, assuming collisions are handled by double hashing using a secondary hash function  $h2(k) = 7- (k \mod 7)$ ? In the notation from the textbook, use h1 = h.

#### Solution

Here collisions are handled by hashfunction  $h\left(k\text{, i}\right)$  which is as following:-

```
h(k, i) = (h1(k) + i h2(k)) mod m
h1(k) = (2k + 5)
h2(k) = 7- (k mod 7)
m = number of slots = 13
k = key to be inserted
```

```
Double Hashing
slot: 0 key 16
slot: 1 key 11
slot: 2 key 44
slot: 3 key 12
slot: 4 key 23
slot: 5 key 13
slot: 6 key 20
slot: 7 key None
slot: 8 key 39
slot: 9 key None
slot: 10 key 5
slot: 11 key 94
slot: 12 key 88
```