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Problem 0

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| Points: |
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Acknowledgements

- (a) I did not work in a group.
- (b) I did not consult without anyone my group members.
- (c) I did not consult any non-class materials.

Problem 1**Points:****Solving recurrences**

$$T(n) = aT(n/b) + \theta(n^d)$$

- (a) Here, $d = 1.3$ and $\log_b a = \log_5 11 = 1.48$

So, $\log_b a > d$, we will use case-3 of Master's Theorem:

$$T(n) = \underline{\theta(n^{\log_5 11})} = \underline{\theta(n^{1.48})}$$

- (b) Here, $d = 2.8$ and $\log_b a = \log_2 6 = 2.58$

So, $d > \log_b a$, we will use case-1 of Master's Theorem:

$$T(n) = \underline{\theta(n^{2.8})}$$

- (c) Here, $d = 0$ and $\log_b a = \log_3 5 = 1.46$

So, $\log_b a > d$, we will use case-3 of Master's Theorem:

$$T(n) = \underline{\theta(n^{\log_3 5})} = \underline{\theta(n^{1.46})}$$

- (d) $T(n) = T(n-2) + \log(n)$

$$= [T(n-4) + \log(n-2)] + \log(n)$$

$$= [T(n-6) + \log(n-4)] + \log(n-2) + \log(n)$$

$$\vdots$$

$$= [T(n-2k) + \log(n-(2k-2))] + \dots + \log(n-2) + \log(n)$$

Let $2k = n$,

$$= T(0) + \log(2) + \log(4) + \log(6) + \dots + \log(n)$$

$$= 1 + \log(2 \cdot 4 \cdot 6 \dots n)$$

$$= 1 + \log(2^{n/2} \cdot (1 \cdot 2 \cdot 3 \dots n/2))$$

$$= 1 + (n/2)(\log(2)) + \log((n/2)!)$$

$$\Rightarrow \log(1) + \dots + \log(1) < \log(1) + \log(2) + \dots + \log(n/2) < \log(n) + \dots + \log(n)$$

$$\Rightarrow 0 < \log((n/2)!) < \log((n/2)^{n/2})$$

$$\text{Also, } \log((n/2)^{n/2}) = (n/2) \cdot \log(n/2) = \frac{n \cdot \log_c(n)}{2 \cdot \log_c(2)} = O(n \log(n))$$

$$= \underline{O(n \log(n))}$$

Problem 2

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| Points: |
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Sorted Array

```
def Search(low, high, A):  
    if (low == high):  
        if (A[low] == 1):  
            return low  
        else:  
            return False  
    else:  
        mid = (low+high)//2  
        if (A[mid] == mid):  
            return mid  
        elif (A[mid] > mid):  
            return Search(low, mid-1, A)  
        else:  
            return Search(mid+1, high, A)
```

Run-time Analysis: $O(\log(n))$

Problem 3

Points:

Linear Time Sorting

```

def linearTimeSorting(unsortedList, k):
    sortedList = [0 for _ in range(len(unsortedList))]
    tempList = [0 for _ in range(k+1)]

    for j in range(0, len(unsortedList)):
        tempList[unsortedList[j]] += 1

    for i in range(1, k+1):
        tempList[i] = tempList[i] + tempList[i-1]

    for j in range(len(unsortedList)-1, -1, -1):
        sortedList[tempList[unsortedList[j]]-1] = unsortedList[j]
        tempList[unsortedList[j]] -= 1

    return sortedList

```

Initially, the *tempList* is initialized and it takes $O(k)$ time where k is $\max_i(x_i)$.

The first ‘for loop’ is making *tempList*[*i*] to hold the number of input elements equal to *i* for each integer $i = 1, 2, \dots, k \Rightarrow$ frequency of elements. It takes $O(n)$ time.

The second ‘for loop’ is keeping running sum of array *tempList* to get how many input elements $\leq i$. It takes $O(k)$ time.

The third ‘for loop’ placing each element *unsortedList*[*j*] in its correct sorted position in the *sortedList*; also we decrement *tempList*[*unsortedList*[*j*]] each time we place a value *unsortedList*[*j*] into the *sortedList*. It takes $O(n)$ time.

Therefore, the time is $O(k+n)$. Note that, there is a significance of $\min_i(x_i)$ that the *tempList* starts changing from the index $= \min_i(x_i)$ and not from index 0 everytime. This is because the given list may or may not have the minimum integer as 0. Depending on what minimum number the list contains, the *tempList* is starting to change from that index instead of 0.

Therefore, the **Overall Time Complexity is:** $O(n + \max_i(x_i) - \min_i(x_i))$

$$\Rightarrow \underline{O(n+M)} \text{ (where } M = \max_i(x_i) - \min_i(x_i))$$

The $\Omega(n \cdot \log(n))$ does not apply in this case because it is not a comparison sort. There is no comparison among input elements in this algorithm.