

Name: Yash Santosh Sainane
Roll No: 21102A0070
Sub: AOA
Div: CMPNA

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Homework Assignment 5

Q.1

$$T(n) = T(n/2) + n$$

a.

$$= T(n/2) + \frac{n}{2} + n$$

$$= T\left(\frac{n}{2^3}\right) + \frac{n}{2^2} + \frac{n}{2} + \frac{n}{2^0}$$

⋮

⋮

⋮

$$= T\left(\frac{n}{2^k}\right) + \left[\frac{n}{2^{k-1}} + \dots + \frac{n}{2} + \frac{n}{2^0} \right]$$

$$= T\left(\frac{n}{2^k}\right) + n \left(\frac{1 - (1/2)^k}{1 - 1/2} \right)$$

$$= T\left(\frac{n}{2^k}\right) + 2n \left(1 - \left(\frac{1}{2}\right)^k \right)$$

Now, recursion bottoms out when $\frac{n}{2^k} = 1$

$$\therefore k = \log\left(\frac{n}{2}\right)$$

When this happens we get,

$$T(n) = T(2) + 2n \left(1 - \frac{1}{2^{\log n/2}} \right)$$

$$= T(2) + 2n \left(1 - \frac{2}{n} \right)$$

$$= T(2) - 4 + 2n$$

$T(2)$ will take $O(1)$ time i.e. constant say 1.

$$\therefore T(n) = 2n - 3$$

For upper bound:

we need to show that there exist c, n_0
such that $0 \leq T(n) \leq c \cdot g(n) \quad \forall n \geq n_0$.

\therefore Here $g(n) = n$

$$\therefore 2n - 3 \leq cn$$

setting $c=3$ and $n_0 = 1$ proves the upper bound.

$$\therefore \boxed{T(n) = O(n)}$$

For lower bound:

we need to show that there exist c, n_0
such that $0 \leq c \cdot g(n) \leq T(n) \quad \forall n \geq n_0$.

$$\therefore 2n - 3 \geq cn$$

This is same as showing
 $(2-c)n \geq 3$

setting $c=1$ and $n_0 = 4$ proves the lower bound.

$$\therefore \boxed{T(n) = \Omega(n)}$$

$$\text{Hence } \boxed{T(n) = \Theta(n)}$$

$$T(n) = 2T\left(\frac{n}{2}\right) + 2n$$

$$= 2 \left[2T\left(\frac{n}{2^2}\right) + n \right] + 2n$$

$$= 2^2 T\left(\frac{n}{2^2}\right) + 2(2n)$$

$$= 2^2 \left[2T\left(\frac{n}{2^3}\right) + \frac{n}{2} \right] + 2(2n)$$

$$= 2^3 T\left(\frac{n}{2^3}\right) + 3(2n)$$

⋮

⋮

$$= 2^k T\left(\frac{n}{2^k}\right) + k(2n)$$

Recursion bottoms out when $\frac{n}{2^k} = 2$.

$$\text{i.e. } k = \log(n/2)$$

when this happens, we get

$$T(n) = 2^{\log(n/2)} T\left(\frac{n}{2^{\log(n/2)}}\right) + \log\left(\frac{n}{2}\right) \times (2n)$$

$$= \frac{n}{2} T(2) + 2n \log\left(\frac{n}{2}\right)$$

$T(2)$ will take $O(1)$ time i.e constant say 1.

$$T(n) = 2n \log\left(\frac{n}{2}\right) + \frac{n}{2} - 2n$$

$$T(n) = 2n \log\left(\frac{n}{2}\right) - \frac{3n}{2}$$

For upper bound:

$$T(n) \leq c \cdot g(n) \quad \forall n \geq n_0 \text{ for some constant } c$$

here $g(n) = n \log(n)$

$$\therefore n \log(n) - \frac{3n}{2} \leq c \cdot n \log(n)$$

Setting $c = 3$ and $n_0 = 2$ proves the upper bound.

$$\boxed{T(n) = O(n)} \quad \boxed{T_n = O(n \log(n))}$$

For lower bound:

$$T(n) \geq c \cdot g(n) \quad \forall n \geq n_0 \text{ for some constant } c$$

here $g(n) = n \log(n)$

$$n \log(n) - \frac{3n}{2} \geq c \cdot n \log(n)$$

This will be seen as showing,

$$n \log(n) [2 - c] \geq \frac{3n}{2}$$

Setting $c = \frac{3}{2}$ and $n_0 = 2$ proves the lower bound.

$$\boxed{T(n) = \Omega(n \log(n))}$$

Hence, $T(n) = \Theta(n \log(n))$

c. $T(n) = T(\sqrt{n}) + 1$

$$\begin{aligned} \rightarrow \quad \therefore T(n) &= T(n^{1/2}) + 1 \\ &= T(n^{1/2^2}) + 2 \\ &= T(n^{1/2^3}) + 3 \\ &\vdots \\ &\vdots \\ T(n) &= T(n^{1/2^k}) + k \end{aligned}$$

recursion bottoms out when, $n^{1/2^k} = 2$.

$$\therefore k = \log_2(\log_2(n))$$

when this happens, we get.

$$T(n) = T(2) + \log_2(\log_2(n))$$

~~and n₀~~
T(2) will take ~~O(1)~~ O(1) time i.e constant

say 1

$$\therefore T(n) = 1 + \log_2(\log_2(n))$$

Now,

for upper bound:

$$T(n) \leq c \cdot g(n) \quad \forall n \geq n_0 \text{ for some } c \text{ and } n_0$$

here

$$g(n) = \log_2(\log_2(n))$$

$$\therefore 1 + \log_2(\log_2(n)) \leq c \cdot \log_2(\log_2(n))$$

setting $c = 4$ and $n_0 = 4$ proves the upper bound.

$$\therefore T(n) = O(\log(\log(n)))$$

For lower bound:

$$T(n) \geq c \cdot g(n) \quad \forall n \geq n_0 \text{ for some } c \text{ and } n_0$$

here $g(n) = \log_2 \log_2(n)$

$$1 + \log_2(\log_2(n)) \geq c \log_2 \log_2(n)$$

$$(1-c) \log_2(\log_2(n)) \geq -1$$

setting $c = \frac{1}{2}$ and $n_0 = \frac{1}{2}$ proves the lower bound

$$\therefore T(n) = \Omega(\log_2(\log_2(n)))$$

$$\text{Hence, } T(n) = \Theta(\log_2(\log_2(n)))$$

d.

$$\begin{aligned}
 T(n) &= 2T(n-1) + 1 \\
 &= 2[2T(n-2) + 1] + 1 \\
 &= 2^2 T(n-2) + 2 + 2^0 \\
 &= 2^2 [2T(n-3) + 1] + 2^1 + 2^0 \\
 &= 2^3 T(n-3) + 2^2 + 2^1 + 2^0 \\
 &\quad \vdots \\
 &= 2^k T(n-k) + [2^{k-1} + \dots + 2^1 + 2^0] \\
 T(n) &= 2^k T(n-k) + (2^k - 1)
 \end{aligned}$$

Recursion bottoms out when $n-k=2$

$$\therefore n = k+2 \quad \therefore k = n-2$$

when this happens, we get

$$\begin{aligned}
 T(n) &= 2^{n-2} T(2) + 2^{n-2} - 1 \\
 &= \frac{2^n T(2)}{4} + \frac{2^n}{4} - 1
 \end{aligned}$$

$T(2)$ will take $O(1)$ time i.e constant say 1

$$\therefore T(n) = \frac{2^n}{4} + \frac{2^n}{4} - 1$$

$$T(n) = \frac{2^n}{2} - 1$$

For upper bound:

$T(n) \leq c.g(n) \quad \forall n \geq n_0$ for some $c \& n_0$.

$$\therefore \frac{2^n}{2} - 1 \leq c \cdot 2^n$$

setting $c=2$, $n=2$ proves the upper bound.

$$T(n) = O(2^n)$$

For lower bound:

$$T(n) \geq c \cdot 8(n)$$

$$\frac{2^n}{2} - 1 \geq c \cdot 2^n$$

This will be same as showing

$$2^n \left(\frac{1}{2} - c\right) \geq 1$$

Setting $c = \frac{1}{4}$ and $n_0 = 4$ proves the lower bound.

$$\therefore T(n) = \Omega(2^n)$$

$$\text{Hence } T(n) = \Theta(2^n)$$

Q.2

a) Algorithm:

1. int s[n], i, k;
2. mergesort(s[n])
3. for(i=0 ; i < n ; i++) → $n \log(n)$
4. k = BinarySearch(s, z - s[i])
5. if (k ≠ -1 & k ≠ i)
6. return true.
7. return false

In above algorithm mergesort will take $\Theta(n \log n)$
 and for loop will run for max n times.
 and Binary search will take $O(\log n)$ time
 \therefore sum of all above we will get

$$T = \Theta(n \log n)$$

b) If s ← sorted - (Given)

Algorithm:

1. int *a = 0, int *e = n - 1, int sum
2. while (*a < e)
3. { sum = (g[*a] + *s[*e])
4. if (sum == z)
- return true;
5. if (sum < z)
- a++ // increase start pointer
6. if (sum > z)
- e-- // decrease end pointer
7. if (e > s)
- return false; #

above algorithm can take maximum n iterations
 \therefore it will take $O(n)$ time.

Q.3)

Algorithm - 1:

Given $\rightarrow A[1...n]$ is sorted array.1. int $s = A[A.length - 1];$

2. int check[s] = {0};

3. for (int $i = 0; i < n; i++$)

{

check[A[i]]++;

4. Merge sort (check, s)

5. if (Binary Search(c, s, $n/2$) == 1)

{

// we will make Binary search return

"1" if there is element greater than

 $n/2$ in c.. otherwise "0".

printf("Majority element is present")

}

c. else

printf("Majority element is not present")

In above algorithm for loop will take max $O(n)$ iterations, Merge sort will take $O(n \log(n))$ time, and binary search will take $O(\log(n))$ time and rest will take constant time.

$$\therefore T(n) = O(n) + O(n \log n) + O(\log n)$$

$$T(n) = O(n \log n)$$

Better Algorithm.

Algorithm-2:

```

1. int count = 0, majorityelement = 0; int s = n/2;
2. for (int i=0; i<n; i++)
{
    if (count == 0)
        majorityelement = A[i];
    if (A[i] == majorityelement)
        count = count + 1;
    else
        count = count - 1;
}
3. count = 0; // assign count = 0;
4. for (int i=0; i<n; i++)
{
    if (majorityelement == A[i])
        count = count + 1;
}
if (count > s)
    return true;
else
    return false.
5.

```

In above algorithm, 1st for loop will run for maximum n time and so does the 2nd.
 $\therefore T(n) = O(n) + O(n)$ and it will take $O(n)$ time.

$$\therefore T(n) = O(n)$$