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 BTech CSF 5th Sem
 Section C
 1961152

Assignment 1 (DAA).

i) Asymptotic means towards infinity. These are used to tell the complexity of an algorithm when the input is very large (10^8 , 10^9 , etc).

Types of Asymptotic Notations:

i) Big-O $\Rightarrow f(n) = O(g(n))$

where $g(n)$ is tight upper bound of $f(n)$. $f(n)$ can never go beyond $g(n)$.

$$\text{Eg} \Rightarrow f(n) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$$

$$f(n) = O(x^n)$$

ii) Big Omega (Ω) $\Rightarrow f(n) = \Omega(g(n))$

where $g(n)$ is tight upper bound of $f(n)$

& $f(n)$ is always never perform better than $g(n)$.

Eg \Rightarrow For the sorting of the array if array is already sorted so insertion sort will take place.

$\Omega(n)$ is the Best time complexity

$$\text{iii) Master - O} \Rightarrow f(n) = O(g(n)) \text{ if } c_1 g(n) \leq f(n) \leq c_2 g(n)$$

It gives the time complexity of algorithm at average cases.

Eg \Rightarrow When array is unsorted so all complexity will be $O(n)$.

2>

Sol > for ($i=1$ to n) $\{ i = i * 2 \}$

The series formed is: 1, 2, 4, 8, 16, ..., K^{th} term upto n terms

So by using G.P: $a_n = a \gamma^{k-1}$

$$a_n = n, \quad a = 1, \quad \gamma = 2,$$

$$n = 1 \cdot 2^{k-1}$$

$$n = 2^k \Rightarrow 2n = 2^k$$

taking \log_2 on both side.

$$\log_2 2n = \log_2 2^k$$

$$\log_2 2 + \log_2 n = K \log_2 2$$

$$1 + \log_2 n = K$$

$K = \text{Time complexity} = O(\log_2 n) =$

$$K = \log n$$

3) $T(n) = \begin{cases} 3T(n-1) & \text{if } n > 0, \text{ otherwise } 1 \\ T(n) = 3T(n-1) & T(0) = 1 \end{cases}$

Sol. Using substitution method

$$T(n) = 3T(n-1) \rightarrow T(n-1) = 3T(n-2)$$

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

$$T(n) = 3(3(3T(n-3)))$$

$$T(n) = 3^3 T(n-3)$$

For K terms

$$T(n) = 3^K T(n-K)$$

$$\text{let } T(n-K) = T(0)$$

$$n = K$$

Using $n = K$

$$T(n) = 3^n T(n-n)$$

$$T(n) = 3^n \cdot 1$$

$$T(n) = 3^n$$

4) $T(n) = 2T(n-1) - 1 \quad T(0) = 1$

Sol. $T(n) = 2T(n-1) - 1 \rightarrow T(n-1) = 2T(n-2) - 1$

$$T(n) = 2[2T(n-2) - 1] - 1 \rightarrow T(n-2) = 2T(n-3) - 1$$

$$T(n) = 2[2\{2T(n-3) - 1\} - 1] - 1$$

$$T(n) = 2[4T(n-3) - 2 - 1] - 1$$

$$\Rightarrow 8T(n-3) - 7$$

For k^{th} term

$$T(n) = 2^K (T(n-K)) - (2^K - 1)$$

$$\text{let } T(n-K) = T(0)$$

$$n = K$$

$$T(n) = 2^n T(0) - (2^n - 1)$$

$$T(n) = 2^n - 2^n + 1 = 1$$

5) `int i=1, s=1;`

`while (s <= n) {`

`i++;`

`s = s + i;`

`printf ("%#");`

`}`

Sol. Here no. of steps.

1

1

8

2

2

1

3

3

6

4

4

10

5

5

15

Order of growth is not constant so general term is $\frac{k(k+1)}{2}$

$$\text{so } n = \frac{k(k+1)}{2}$$

$$n = \frac{k^2 + k}{2} \Rightarrow 2n = k^2 + k$$

ignoring lower order terms

$$n = k^2$$

$$K = \sqrt{n}$$

6) `for (i=1 ; i * i <=n ; i++) count++;`

sol: so i is moving from 1 to \sqrt{n}
with linear growth
 $T(n) = \sqrt{n}$.

```
7) for (i = n/2 ; i <= n ; i++)
    for (j = 1 ; j <= n ; j = j * 2)
        for (k = 1 ; k <= n ; k = k * 2)
            count ++;
```

Since the j & k loops are running from 1 to n with exponential growth is Time complexity for these 2 loops will be:

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

8 i loop is moving from $\frac{1}{2}$ to n with
constant growth.

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

so overall complexity w.r.t the f^n will be
 $T(n) = O(n \log n \cdot \log n)$

$$8) T(n) = T(n-3) + n^2 \quad \& \quad T(1) = 1$$

$$T(n) = T(n-3) + n^2 \rightarrow T(n-3) = T(n-6) + (n-3)^3$$

$$T(n) = T(n-6) + (n-3)^2 + n^2 \rightarrow T(n-6) = T(n-9) + (n-6)^2$$

$$T(n) = T(n-9) + (n-6)^2 + (n-3)^2 + n^2 \rightarrow T(n-9) = T(n-12) + (n-9)^2$$

$$T(n) = T(n-12) + (n-9)^2 + (n-6)^2 + (n-3)^2 + n^2$$

General term is

$$T(n) = T(n-k) + (n-(k-3))^2 + (n-(k-6))^2 + (n-(k-9))^2 + (n-(k-12))^2 + \dots + (n-(k-k))^2$$

$$\text{Since } T(n-k) = T(1)$$

$$n = 1+k \quad \& \quad k = n-1$$

$$= T(n) = T(n-(n-1)) + [n-(n-1-3)]^2 + [n-(n-1-6)]^2 + [n-(n-1-9)]^2 + \dots + n^2$$

$$T(n) = T(1) + 4^2 + 7^2 + 10^2 + \dots + n^2$$

$$= 1 + 4^2 + 7^2 + 10^2 + \dots + n^2$$

General formula is

$$n = (3k-2)^2$$

$$n = 9k^2 + 4 - 12k$$

By ignoring lower terms

$$n = k^2$$

$$k = \sqrt{n}$$

$$T(n) = O(\sqrt{n})$$

a) for ($i=1$ to n)

 for ($j=1$ to n) { $j=j+i$ }

 print ("*")

sol: steps \Rightarrow

i	1	2	3	4	5	n
j	n	$\frac{n}{2}$	$\frac{n}{3}$	$\frac{n}{4}$	$\frac{n}{5}$	$\frac{n}{n} = 1$

$$TC = n + \frac{n}{2} + \frac{n}{3} + \frac{n}{4} + \frac{n}{5} + \dots + \frac{n}{n}$$

$$TC = n \left[1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \frac{1}{5} + \dots + \frac{1}{n} \right]$$

$$TC = n \int_1^n \frac{1}{x} dx$$

$$TC = n [\log x]_1^n = n [\log n - \log 1]$$

$$TC = n \log n$$

10) if $c > 1$, then the exponential cn will grow very fast, so that ans is n^k is $O(c^n)$

11) init j=1, i=0;
while (i < n) {

$$i = i + j;$$

$$j++;$$

3

$$\text{sol. } i = 0 \quad 1 \quad 3 \quad 6 \quad 10 \quad 15 \quad \dots$$

$$j = 1 \quad 2 \quad 3 \quad 4 \quad 5 \quad 6 \quad \dots$$

Here 'i' will go till n & general formula for k^{th} term is $n = \frac{k(k+1)}{2}$

$$\text{so } TC = O(\sqrt{n})$$

12) Use relation for the fibonacci series is
~~so~~

$$T(n) = T(n-1) + T(n-2) + C$$

$$\text{let } T(n-1) \cong T(n-2)$$

$$= T(n) = 2T(n-1) + C \rightarrow T(n-1) = 2T(n-2) + C$$

$$T(n) = 2[2T(n-2) + C] + C \rightarrow T(n-2) = 2T(n-3) + C$$

$$T(n) = 4T(n-2) + 3C$$

$$T(n) = 4[2T(n-3) + C] + 3C$$

$$T(n) = 8T(n-3) + 7C$$

~~so~~ for the K^{th} term until we

$$T(n) = 2^K T(n-K) + (2^K - 1)C$$

$$= T(n-K) \text{ for } T(0) = 1$$

$$= n-K = 0 \Rightarrow n = K$$

$$T(n) = 2^n \cdot 1 + (2^n - 1)C$$

$$T(n) = 2^n + 2^n C - C$$

$$= 2^n [1 + C] - C$$

neglect the C

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

And max depth for recursive function is
 based on n so space complexity is $O(n)$

13) $n \log n$

Eg \Rightarrow for (int i=0; i<n; i++)

 for (int j=0; j<n; j *= 2)

 cout << "

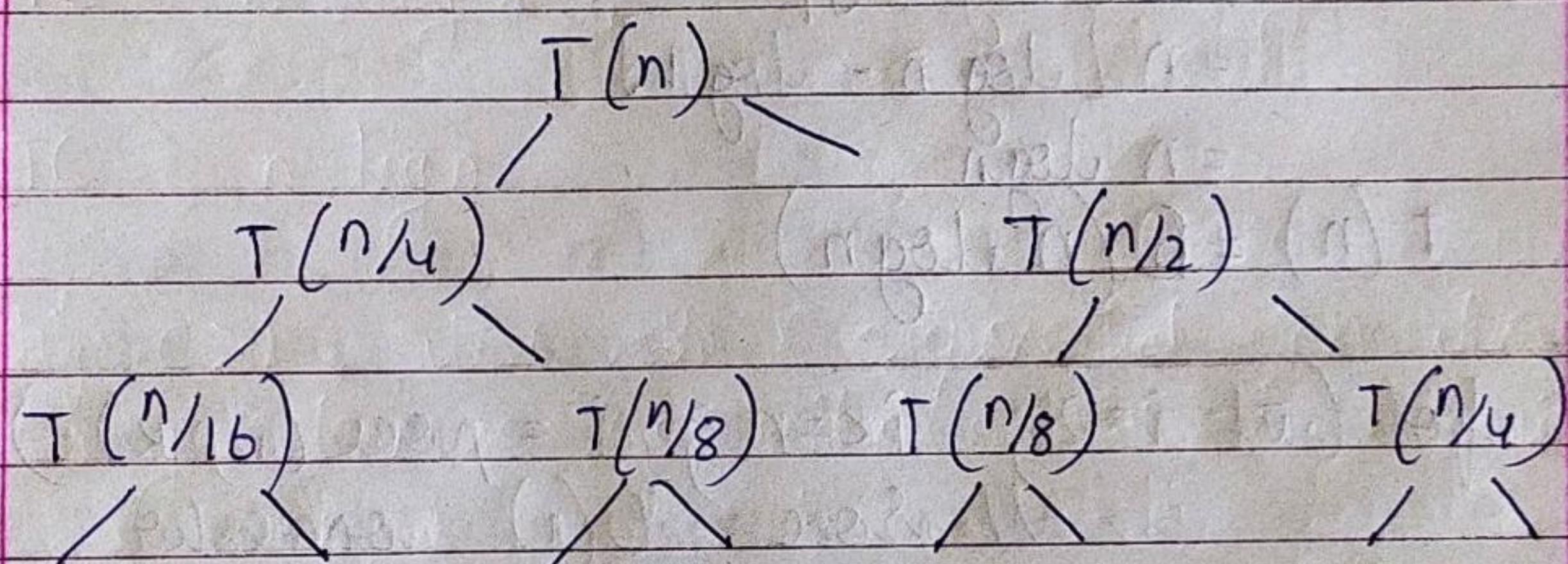
ii) n^3

eg \Rightarrow for (int $i=0$; $i < n$; $i++$)
 for (int $j=0$; $j < n$; $j++$)
 for (int $k=0$; $k < n$; $k++$)
 printf ("%#");

iii) $\log(\log n)$

eg \Rightarrow for (int $i=0$; $i < n$; $i = i * i$)
 printf ("Hello");

$$14) T(n) = T(n/4) + T(n/2) + cn^2$$



$$T(n) = C \left[n^2 + \frac{5n^2}{16} + \frac{25n^2}{256} + \dots \right]$$

so here

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

15) for (int $i=1$; $i < n$; $i++$)
 for (int $j=1$; $j < n$; $j += i$)
 ignore $O(1)$ operation

Sol.

$$\begin{array}{ccccccccc} i & = & 1 & 2 & 3 & 4 & 5 & \dots & n \\ j & = & n & \frac{n}{2} & \frac{n}{3} & \frac{n}{4} & \frac{n}{5} & & \frac{n}{n} \end{array}$$

so, here;

$$\begin{aligned} T(n) &= n + \frac{n}{2} + \frac{n}{3} + \frac{n}{4} + \dots + \frac{n}{n} \\ &= n \left[1 + \frac{1}{2} + \frac{1}{3} + \frac{1}{4} + \dots + \frac{1}{n} \right] \\ &= n \int_1^n \frac{1}{x} dx \\ &= n [\log x]_1^n \\ &= n [\log n - \log 1] \\ &= n \log n \\ T(n) &= O(n \log n) \end{aligned}$$

16) `for (int i=2; i<=n; i = pow(i, k))`
 // some $O(1)$ operation

Sol for any function increasing with exponential rate the $T(n)$ becomes

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

18) a) $n, n!, \log n, \log \log n, \sqrt{n}, \log(n!),$
 $n \cdot \log n, 2^n, 2^{2^n}, 4^n, n^2, 100.$

Sol $100 < \log(\log(n)) < \log n < \text{root}(n) < n < n \log n$
 $< n^2 < 2^n < 2^{2^n} < 4^n < \log(n!) < n!$

b) $2(2^n), 4n, 2n, 1, \log(n), \log(\log n),$

$\sqrt{\log(n)}, \log 2n, 2\log n, n, \log n!, n!, n^2, n \log n$.

Sol: $1 < \log(\log n) < \sqrt{\log(n)} \leq \log n \leq \log 2n \leq \log(n^2) \leq 2 \log(n) < 2n \leq 2^n \leq 4n \leq n \log(n) \leq n^2 \leq 2(2^n) \leq n!$

c) $8^{2n}, \log_2(n), n \log_6(n), n \log_2(n), \log(n!), n!, \log_8(n^8), 76, 8n^2, 7n^3, 5n$.

Sol: $76 < \log_8 n < \log_2 n < n \log_6 n < n \log_2 n < 5n < 8n^2 < 7n^3 < 8^{2n} < \log(n!) < n!$

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Sol: Search(A, n)

low = 0 high = n - 1, K

while (low ≤ high)

mid = $\frac{(low + high)}{2}$

if A[mid] = Key

return mid

else if (A[mid] < key)

low = mid + 1

else

high = mid - 1

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Sol: Iterative method →

insertion sort (A)

for j = 2 to A.size

Key = A[j];

$$i = j - 1$$

while $i > 0$ AND $A[i] > \text{key}$

$$A[i+1] = A[i]$$

$$A[i+1] = \text{key}$$

Recursive Method \Rightarrow

Insertion sort (A, n)

if ($n \leq 1$)

return

Insertion sort ($A, n-1$)

$$\text{key} = A[n-1]$$

$$j = n-2$$

while $j \geq 0$ AND $A[j] > \text{key}$

$$A[j+1] = A[j]$$

$$A[j+1] = \text{key}$$

This sort considers one i/p element per iteration and produces a partial solution without considering future elements. That's why it is called online sorting.

Other sorting algo's i.e.

Bubble sort $O(n^2)$

Insertion sort $O(n^2)$

Heap sort $O(n \cdot \log n)$

selection sort $O(n^2)$

merge sort $O(n \cdot \log n)$

Quick sort $O(n \cdot \log n)$

d21-

sol.

Algos.	Best case	Average case	Worst case.
Bubble sort	$\Omega(n)$	$\Theta(n^2)$	$O(n^2)$
Selection sort	$\Omega(n^2)$	$\Theta(n^2)$	$O(n^2)$
Insertion sort	$\Omega(n)$	$\Theta(n^2)$	$O(n^2)$
Merge sort	$\Omega(n \log n)$	$\Theta(n \log n)$	$O(n \log n)$
Heap sort	$\Omega(n \log n)$	$\Theta(n \log n)$	$O(n \log n)$
Quick sort	$\Omega(n \log n)$	$\Theta(n \log n)$	$O(n^2)$
Counting sort	$\Omega(n+K)$	$\Theta(n+K)$	$O(n+K)$

d22

sol. Algos	Inplace	stable	Online
Bubble sort	Yes	Yes	Yes
Insertion sort	Yes	Yes	Yes
selection sort	Yes	No	Yes
Merge sort	No	Yes	Yes
Quick sort	Yes	No	Yes
Heap sort	Yes	No	Yes
Count sort.	Yes	Yes	Yes.

d23

sol. linear iterative method

Binary Search (A, n, Key)

i = 0, j = n - 1

while (i <= j)

mid = (i + j) / 2

if A[mid] == key

return mid

else if A[mid] < key

i = mid + 1

else

$j = \text{mid} - 1$
 return -1

Time complexity = $O(\log n)$
 Space complexity = $O(1)$

Recursive method \Rightarrow

Binary-search(A, i, j, key)

if ($i > j$)

return -1

mid = $(i + j) / 2$

if $A[\text{mid}] == \text{key}$

return mid

else if $A[\text{mid}] < \text{key}$

return binary-search(A, mid + 1, j, key)

else

return binary-search(A, i, mid - 1, key)

Time complexity $\Rightarrow O(\log n)$

Space complexity $\Rightarrow O(1)$

Linear Search \Rightarrow Time Complexity $\Rightarrow O(n)$

Space Complexity $\Rightarrow O(1)$

24)

Sol Recursion relation for binary search is

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

