

DAA

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ASSIGNMENT-1

Ans 1 - Asymptotic notations are methods/languages using which we can define the running time of the algorithm based on input size.

There are mainly 3 asymptotic notations

(1) Big-O Notation: Big-O notation represents the upper bound of the running time of an algorithm. Thus, it gives the worst-case complexity of an algorithm.

eg:

$O(\log n)$ = Binary search

$O(n)$ = Simple search

(2) Omega Notation: It represents the lower bound of the running time of an algorithm. Thus, it provides the best case complexity of an algorithm.

$$f(n) = \Omega(g(n))$$

EXAMPLE: for a given function $g(n)$, we denote by $\Omega(g(n))$ the set of functions.

$\Omega(g(n)) = \{f(n) : \text{there exist positive constants } c \text{ and } n_0 \text{ such that } c * g(n) \leq f(n) \text{ for all } n \geq n_0\}$

(a) Theta Notation (Θ): Theta notation encloses the function from above & below since, it represents the upper and the lower bound of the running time of an algorithm, it is used for analyzing the average-case complexity of an algorithm.

for a function $g(n)$, $\Theta(g(n))$:-

$\Theta(g(n)) = \{ f(n) : \text{there exist positive constants } c_1, c_2 \text{ and } n_0 \text{ such that}$
 $0 \leq c_1 g(n) \leq f(n) \leq c_2 g(n) \text{ for all } n \geq n_0 \}$

Ans 2 : for $i = 1$ to n
 $\quad \{ i \times 2$
 $\quad \}$

Time complexity : $O(\log n)$

As per q. stn. it will run $2^k = n$
 which leads to $k = \log n$

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

$$\begin{aligned} T(n) &= 3T(n-1) \quad (\text{solve using backward substitution}) \\ &= 3(3T(n-2)) \\ &= 3^2 T(n-2) \\ &= 3^3 T(n-3) \\ &= \vdots \\ &= 3^n T(n-n) \\ &= 3^n T(0) \\ &= 3^n \end{aligned}$$

This clearly shows that the complexity of this function is $O(3^n)$

Ans 4 3

$$T(n) = \begin{cases} 2T(n-1) - 1, & \text{if } n > 0 \\ 1, & \text{otherwise} \end{cases}$$

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

Ans 7: Outer loop = $\frac{n}{2} = n$
 Inner 1st loop = $\log n$
 Inner 2nd loop = $\log n$

Time Complexity : $O(n^2 \log n)$

Ans 8: function (int n)
 { if (n == 1)
 return 1;
 for (i = 1 to n)
 for (j = 1 to n)
 printf(" * ");
 function (n-3);
 }

$$T(n) = n^2 + T(n-3)$$

$$T(1) = O(1)$$

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

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

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

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

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

$$T(n) = n^2 + (n-3)^2 + (n-6)^2 + \dots + 4^2 + T(1)$$

$$= n^2 + (n-3)^2 + (n-6)^2 + \dots + 1^2 + 4^2 + 1$$

$$= \sum_{k=1}^{(n+2)/3} (3k-2)^2 = \sum (9k^2 + 4 - 12k)$$

$$T.C \approx \Theta(n^3)$$

Ans 9 \Rightarrow void function (withn)

for (i=1 to n)

for (j=1, j<=n; j=j+i)

printf("%*");

}

i = 1, 2, 3, ..., n

j = $\frac{n}{1}, \frac{n}{2}, \frac{n}{3}, \dots, \frac{n}{n}$

$T(n) = \frac{n}{1} + \frac{n}{2} + \frac{n}{3} + \dots + \frac{n}{n}$
 $= n \left[1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \right]$
 $T.O.C = n \log n$

Ans 11: The loop variable 'i' is incremented by 1, 2, 3, 4, ... until it becomes greater than or equal to n.

The value of i is $x(x+1)/2$ after x iterations.

So if loop runs x times, then $x(x+1)/2 < n$. Therefore

TIME COMPLEXITY can be written as (\sqrt{n}) .

Ans 12: fibonacci series:-

0 1 1 2 3 5 8 13 21 34 55 ...

$$\text{fib}(n) = \begin{cases} 1, 0 & (n=1 \text{ or } n=0 \text{ resp.}) \\ \text{fib}(n-2) + \text{fib}(n-1), & n > 1 \end{cases}$$

The recursive equation for fibonacci series is:-

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

First of all, assume $T(n-2) = T(n-1)$

Solving it using Backward Substitution.

$$T(n) = 2T(n-1) + 1 \quad // \text{now solve}$$

$$T(n) = T(n-1)$$

$$T(n) = 2[2T(n-2) + 1] + 1$$

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

Next, we can substitute in $T(n-2) = 2T(n-3) + 1$

$$T(n) = 2[2[2T(n-3) + 1] + 1] = 8T(n-3) + 7$$

$$\vdots$$

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

Now, we can find k & thereby solve $T(n)$, by substituting in $T(0)=1$

For $T(0)$, we can see that $n-k=0$.

Rearranging, we get $k=n$. Now substituting in our values for $T(0)$ & k , we get

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

$$\text{Time Complexity} = O(2^n)$$

Space Complexity would be: $O(1)$. Since, the program does not use extra spaces.

Ques, 13 $T(n) = O(n \log n)$

```
for (int i=0; i<n; i++)
```

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

```
cout << "x";
```

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

```
for (int i=0; i<n; i++)
```

```
for (int j=0; j<n; j++)
```

```
for (int k=0; k<n; k++)
```

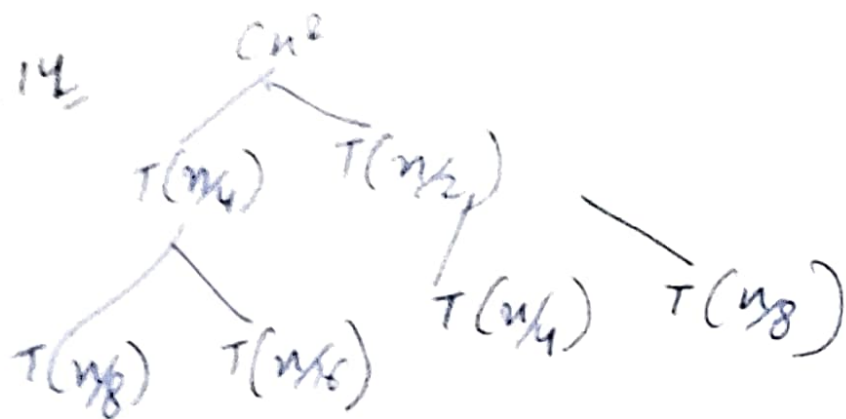
```
cout << i << " " << k << endl;
```

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

```
for (int i=0; i<log(n); i*=2)
```

```
cout << i << " ";
```

Ques 14



$$T(n) = cn^2 + \frac{5n^2}{16} + \frac{25n^2}{256} + \dots$$

This GP with ratio $\frac{5}{16}$

$$\therefore T(n) = \frac{n^2}{1 - \frac{5}{16}} \Rightarrow T.C. = O(n^2)$$

Q 15 \Rightarrow int fun (int n)

{ for (int i=1; i<=n; i++)

for (int j=1; j<=n; j+=i)

o(1)

i = 1, 2, 3, 4, ..., n

j = n, n/2, n/3, n/4, ..., n/n

$\therefore T.C. = n, \frac{n}{2}, \frac{n}{3} + \dots + \frac{n}{n}$

$$n \left[1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \right]$$

$$= n \log n$$

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

Q16 for (int i=0; i<=n; i+=pow(2,k))
 { O(1)

$2^0 = 2, 2^k, 2^{k^2}, 2^{k^3}, \dots, 2^{k^x}$ i.e. (x+1) terms

$$2^{k^x} = n$$

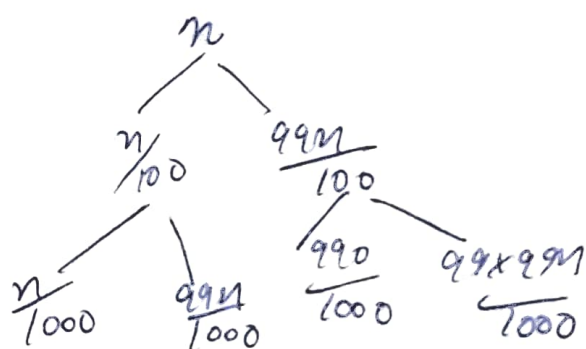
$$k^x = \log_2 n$$

$$x = \log_k (\log_2 n)$$

$$\therefore \text{T.C.} = O(\log_k \log_2 n)$$

Q17 $\Rightarrow T(n) = T\left(\frac{99n}{100}\right) + T\left(\frac{n}{100}\right) + c^n$

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



Q18 \Rightarrow int Linear Search (int arr[], int key)

```

{ int i;
for (i=0; i<n; i++)
    if (arr[i] == key)
        return i;
else if (arr[i] > key)
    return -1;
}
    
```

$$\boxed{\begin{array}{l} T.C. = O(n) \\ S.C. = O(1) \end{array}}$$

Q30 \Rightarrow Iterative Insertion Sort

```
void insertionSort (int *a, int n)
```

```
{ int i, temp;
```

```
  for (i  $\leftarrow$  1 to n)
```

```
    temp  $\leftarrow$  a[1]
```

```
    j  $\leftarrow$  i - 1
```

```
    while (j  $\geq$  0 && a[j]  $>$  temp)
```

```
      a[j+1] = a[j]
```

```
      j  $\leftarrow$  j - 1
```

```
      a[j+1]  $\leftarrow$  temp
```

```
}
```

Recursive Insertion sort.

```
void insertionSort (int *a, int n)
```

```
{ if (n  $<$  2)
```

```
  return;
```

```
  insertionSort (a, n-1)
```

```
  int last  $\leftarrow$  a[n-1]
```

```
  int j  $\leftarrow$  n-2
```

```
  while (j  $\geq$  0 && a[j]  $>$  last)
```

```
    a[j+1]  $\leftarrow$  a[j]
```

```
    j = j - 1
```

```
    a[j+1]  $\leftarrow$  last
```

```
}
```

Online Sorting is one that will work if elements to be sorted are provided 1 at a time with understanding that algo must keep sequence sorted as more & more elements are added. Insertion Sort is online.

Ques, 21, 22

Algo	Best	avg	worst	worst space	Inplace	Stable	On
Bubble	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	✓	✓	X
Selection	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	✓	X	✓
Insertion	$O(n)$	$O(n^2)$	$O(n^2)$	$O(1)$	✓	X	✓
Merge	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	$O(n)$	X	✓	X
Quick	$O(n \log n)$	$O(n \log n)$	$O(n^2)$	$O(n)$	X	X	X
Heap	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	$O(1)$	✓	X	X

Q 23 ⇒ Iterative Binary Search

```
int binarySearch(int *a, int l, int r, int key)
```

```
{ while (l <= r)
```

```
    m = (l+r)/2
```

```
    if (a[m] == key)
```

```
        return m
```

```
    if (a[m] < key)
```

```
        l = m+1
```

```
    else
```

```
        r = m-1
```

return -1

}

Q29 \Rightarrow Binary Recursive Search

$$T(n) = T(n/2) + 1$$
