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Ans: 1 Asymptotic notation are mathematical tools to represent the time complexity of algorithm for asymptotic analysis.

The main idea of asymptotic analysis is to have a measure of the efficiency of algorithm that doesn't depends on machine specific constants and doesn't requires to be implemented and time taken by the program to be compared.

Following are the asymptotic notations that are mostly used.

1. Θ Notation :- The theta notation bounds a function from above and below, so it define exact asymptotic behaviour.
2. Big O Notation :- It define an upper bounds of an algorithm. it bound a function only from above.
3. Ω Notation :- Ω Notation provides an asymptotic lower bound.

For Example consider Insertion Sort.

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It takes linear time in best case and quadratic time in worst case.

We can say that Insertion sort have,

$$O(n^2)$$

$O(n^2)$ for worst case.

$O(n)$ for best case

$$O(n)$$

Ans 2 $O(\log n)$

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

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

Aakash

Ans: 4 $T(n) = \begin{cases} 2T(n-1) - 1, & \text{if } n > 0 \\ 1 & \text{otherwise} \end{cases}$

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

$T(n) = 1$

Ans: 5 $S_i = S_{i-1} + i$

If k is total number of iterations taken by the program then while loop terminates.

$$1+2+3+\dots+k = \left[\frac{k(k+1)}{2} \right] \geq n$$
$$\therefore k = O(\sqrt{n})$$

Ans: 6 $O(\sqrt{n})$

Akash Chaurasia

Ans: 7 j is loop executing $\log n$ times
 k is loop executing $\log n$ times
 i is loop executing $n/2$ times $n/2 \approx n$

Time complexity = $O(n \log^2 n)$

Ans: 8 $O(n^3)$

Ans: 9 Inner loop will execute $(n + \frac{n}{2} + \frac{n}{3} + \dots + 1)$
 $n(1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n})$
 It is equal to $O(n \log n)$

Ans: 10

n^k a^n

$k \geq 1$ $a \geq 1$

Taking $k = a = 2$

n^2 2^n

we can say $n^2 = O(2^k)$

$n^k = O(a^n)$

Ans: 11 $O(\sqrt{n})$ same logic given in ques 5

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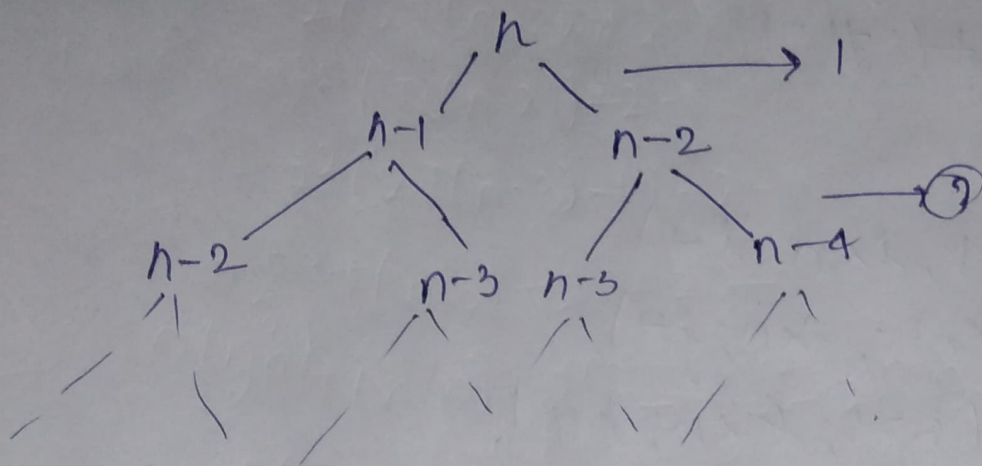
Ans-12

Recurrence Relation

5

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

Making Recurrence Tree.



$$T.C = 1 + 2 + 4 + \dots + 2^n$$

$a = 1$
 $\mu = 2$

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

$$O(2^{n+1}) = O(2 + 2^n) = O(2^n)$$

space complexity = $O(n)$

This is because maximum stack frame is equal to n only as function is called like this.

$$f(n-1) + f(n-2)$$

$f(n-2)$ is called when we get the return value from $f(n-1)$

\therefore It is equal to $O(n)$

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$n \log n$

```
for (i=1; i<n; i++)
```

```
for (j=1; j<n; j++)
```

```
printf("#");
```

 n^3

```
for (i=1; i<n; i++)
```

```
for (j=1; j<n; j++)
```

```
for (k=1; k<n; k++)
```

```
printf("#");
```

 $\log \log n$

```
int fun(int n)
```

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

```
    return 1;
```

```
else
```

```
    return (fun(float(sqrt(n))) + n);
```

```
}
```

Ans: 14

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

we can assume,

$$T(n/2) \geq T(n/4)$$

$$T(n) \geq 2T(n/2) + cn^2$$

Applying Masters Method,

$$a \geq 2, b \geq 2$$

$$k = \log_b a = \log_2 2 = 1$$

$$n^k \geq n$$

$$f(n) \sim n^2$$

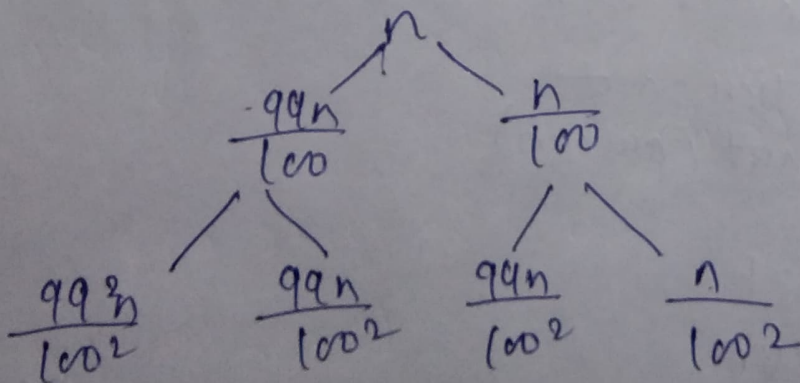
$$It \text{ is } O(n^2)$$

$$\text{But as } T(n) \propto O(n^2)$$

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

Ans: 16 If k is a constant greater than 1
Then $T.C = O(\log \log n)$

Ans: 17 $T(n) = T\left(\frac{99n}{100}\right) + T\left(\frac{n}{100}\right)$



$$T.C = \log_{100/99} n \approx \log n.$$

Akash Singh

we can say that the base of log does not matter as 6
it only a matter of constant.

Ans: 18

a) $60 \log \log n \ln n \log n! \quad n \log n \quad n^2 2^n \quad 2^m / 4^n \quad n!$

b) $1 \log \log n \sqrt{\log n} \log n \quad 2 \log n \log 2n \quad n^2 n^4 n \log n!$
 $n \log n \quad n^2 2(2^n) n!$

c) $96 \log_8^n 5n \log n! \quad n \log_6 n \quad n \log_2 n \quad 8n^2 7n^3$
 $8^{2n} n!$

Ans: 19

Linear Search (array, key)
for i in array
if value == key
return i ;

Ans: 20

Iterative Insertion sort
insertion sort (arr, n)

loop from $i=1$ to $i=n-1$

Pick element $arr[i]$ and insert
it into sorted sequence $arr[0 \dots i-1]$

Recursive Insertion sort

insertion sort (arr, n)

{

if $n \geq 1$

return

recursively sort $n-1$ element
insertion sort (arr, $n-1$)

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Pick last element $arr[i]$ and insert it into sorted sequence $arr[0 \dots i-1]$

1

5.

Insertion sort considers one input element per iteration and produces a partial solution without considering future elements.

It is called online sorting Algorithm.

Ans

considering only 3 sorting Algo. till now as we get the lecture get the lecture of these 3 only.

Algo	Best case	Avg. case	worst case	SC	stable	Inplace
Bubble sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	✓	✓
Selection sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	$O(1)$	✗	✓
Insertion sort	$O(n)$	$O(n^2)$	$O(n^2)$	$O(1)$	✓	✓

Ans: 23. Binary Search.

$A \leftarrow$ sorted Array.

$n \leftarrow$ size of Array

$x \leftarrow$ value to be sorted

while x not found.

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if upperbound & lower bound.

10

EXIT: X does not exist

Set mid point = $(\text{lower bound} + \text{Upper bound} - \text{Lower bound}) / 2$

if $A[\text{mid point}] < X$.

Lower bound = mid point + 1

if $A[\text{mid point}] > X$

upper bound = mid point - 1

if $A[\text{mid point}] = X$

Exit = X found at mid point.

Linear

Time complexity

$O(n)$

space complexity

$O(1)$

Binary Search
(Recursive)

$O(\log n)$

$O(\log n)$

Binary Search
(Iterative)

$O(\log n)$

$O(1)$

Ans 24 $T(n) = T(n/2) + C$

Akash Kumar