

~ ASSIGNMENT - 1 ~

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1) Asymptotic Notations :- are mathematical tools to represent the time complexity of algorithms for asymptotic analysis.

The main idea of asymptotic analysis is to have a measure of the efficiency of algorithms that depend on machines.

Most used asymptotic notations are -

(i) Θ Notation :- The theta notation bounds a func. from above & below, so it defines exact asymptotic behaviour.

(ii) Big O Notation :- It defines an upper bound of an algorithm, it bounds a func. only from above.

(iii) Ω Notation :- it provides an asymptotic lower bound.

For e.g.,

consider Insertion Sort.

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

We can say, Insertion Sort have:-

$O(n^2)$ / $\Theta(n^2)$ for worst case / $\Theta(n)$ for best case / $\Omega(n)$

$$2) O(\log n)$$

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

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

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

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

$$= 2^2(2T(n-3) - 1) - 2 - 1$$

$$= 2^3 T(n-3) - 2^3 - 2^1 - 2^0$$

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

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

$$= 2^n - 2^n + 1 = 1$$

$$\underline{T(n) = 1}$$

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

$$T(n) = 3T(n-1)$$

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

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

⋮

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

$$= \underline{\underline{3^n}}$$

$$5) S_i = S_{i-1} + i$$

If k is total no. of iterations taken by program, then while loop terminates.

$$1 + 2 + 3 + \dots + k = \left[\frac{k(k+1)}{2} \right] > n$$

$$\underline{k = O(\sqrt{n})}$$

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

7) j loop executing $\log n$ times
 k " " " $\log n$ times
 i " " " $n/2$ times $n/2 \approx n$
 Time complexity = $O(n \log^2 n)$

8) $O(n^3)$

9/15) Inner loop will execute $(n + \frac{n}{2} + \frac{n}{3} + \dots + \frac{n}{n})$
 $= n \left(1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \right)$

\Rightarrow It's equal to $O(n \log n)$

10) n^k vs a^n
 $k > 1$ $a > 1$
 Taking $k = a = 2$
 n^2 2^n

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

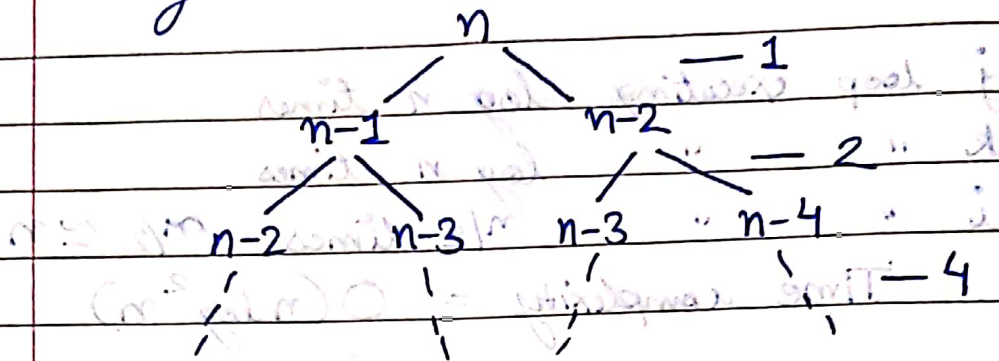
$n^k = O(a^n)$

11) $O(\sqrt{n})$: (NOTE: Logic same as given in 5)

12) Recurrence Relation

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

Making recurrence tree,



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

$$a=1, r=2$$

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

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

$$\text{Space complexity} = O(n)$$

This is because maximum stack frame is equal to n only as func. called like this:-

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

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

∴ It is equal to $O(n)$.

13) $n \log n$

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

```
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 func(int n)
{
    if (n <= 2)
        return 1;
    else
        return (func(floor(sqrt(n))) + n);
}

```

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

assuming,

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

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

Applying masters method,

$$a = 2, b = 2$$

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

$$n^k = n$$

$$f(n) = n^2$$

It is $\Theta(n^2)$

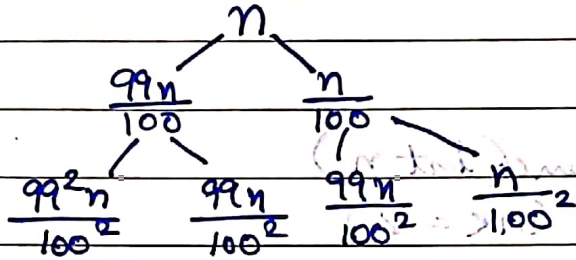
But as $T(n) \leq \Theta(n^2)$

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

16) If k is a constant > 1

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

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



If we take longer branch, i.e., $99n/100$

$\therefore T.C = \log_{\text{base}}(100/99) n \approx \log n$

We can say, the base of log doesn't matter as it's only a matter of constant.

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

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

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

19) ~~Linear~~ $T(n) = O(n)$

19) linear search (array, key)

for i in array

if value == key

return i

20) Iterative Insertion Sort

insertion_sort(arr, n)

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

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

Recursive Insertion Sort

insertion_sort(arr, n)

{ if $n \leq 1$

return

recursively sort $n-1$ elements

insertion_sort(arr, n-1)

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

}

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

It's called online sorting algorithm.

20/21/22 > Considering the 3 sorting algorithms taught,

Algorithm	Best case	Avg Case	Worst Case	SC	Stable	Inplace	On
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	$O(n)$	$O(n^2)$	$O(n^2)$	$O(1)$	✓	✓	✓

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

23) Binary Search

$A \leftarrow$ Sorted array

$n \leftarrow$ Size of array

$X \leftarrow$ value to be searched

while X not found:

if upperbound \leq lowerbound

EXIT: X does not exist

(1, 2, 3, 4, 5, 6, 7, 8, 9, 10)

set $midpt = \frac{lower\ bound + (upper\ b. - lower\ b.)}{2}$

if $A[midpt] < X$:

lowerbound = midpoint + 1

if $A[midpt] > X$:

upperbound = midpoint - 1

if $A[midpoint] = X$:

EXIT: X found at midpoint

	Time complexity	Space Complexity
Linear Search	$O(n)$	$O(1)$
Binary Search (recursive)	$O(\log n)$	$O(\log n)$
Bin. Search (Iterative)	$O(\log n)$	$O(1)$

X — X — X