



i) Big-oh (Θ):

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

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

$$\therefore f(n) \leq c \cdot g(n)$$

$\forall n > n_0$, some constant $c > 0$

Asymptotic Notations:

Asymptotic \rightarrow towards infinity

- While defining complexity of our algo, we will use asymptotic notations, assuming our input size is very large.

* Time Complexity

No. of instructions to carry out an algorithm

Space Complexity

Extra space reqd. by an algorithm except input

* While calculating complexities:

- Constraints are ignored

- Lower order terms are ignored in add "anubhava"
Take the highest order term.

ii) Big Omega (Ω):

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

$g(n)$ is "tight" lower bound of $f(n)$

$$\therefore f(n) \geq c \cdot g(n)$$

$\forall n > n_0 \& c > 0$

iii) Theta (Θ): gives "tight" upper & lower bound both

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

$$\therefore C_1 \cdot g(n) \leq f(n) \leq C_2 \cdot g(n)$$

$\forall n > \max(n_1, n_2)$

for some constant $C_1, C_2 > 0$

$$\Rightarrow f(n) = O(g(n)) \& f(n) = \Omega(g(n))$$

iv) Small-oh (o):

$$f(n) = o(g(n))$$

$$\therefore f(n) < c \cdot g(n) \& \forall n > n_0 \& c > 0$$



5) Small omega (ω):

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

$$f(n) > c g(n)$$

$$\forall n > n_0 \text{ & } c > 0$$

Note:- $\omega(f(n)) = O(g(n)) \rightarrow g(n) = \Omega(f(n))$

* $f(n) = \Omega(g(n)) \rightarrow g(n) = \omega(f(n))$

* $f(n) = \Theta(g(n)) \rightarrow f(n) = O(g(n)) \& f(n) = \Omega(g(n))$
Reflexive Symmetric Transitive

O	✓	✗	✓
Ω	✓	✗	✓
O	✗	✗	✓
ω	✗	✗	✓

A-2) for $c_i = 1; i \leq n; i = i+2$ {
}

Values of $i: 1, 2, 4, 8, \dots, 2^k$

which forms a G.P.

$$a = 1, r = 2$$

$$\therefore t_k = a \cdot r^{k-1}$$

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

$$2^n = 2^k$$

$$\therefore k \cdot \log_2 2 = \log_2(2^n)$$

$$k \cdot \log_2 2 = \log_2(2) + \log_2(n)$$

$$k = 1 + \log_2(n)$$

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

A-3) $T(n) = 3T(n-1)$ if $n > 0$, otherwise 1?

$$T(0) = 1$$

Using forward substitution:

$$T(1) = 3T(0) = 3$$

$$T(2) = 3 \quad T(1) = 3 \times 3 = 9 \quad - \text{for } n=2 \in \{n\}$$

$$T(3) = 3T(2) = 3 \times 9 = 27 \quad - \text{for } n=3 \in \{n\}$$

$$\therefore T(n) = 3^n, n \in \{n\}$$

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

$$T(0) = 1$$

Using backward substitution,

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

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

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

$$= 4T(n-2) + 2 + 1$$

$$= 8T(n-3) + 4 + 2 + 1$$

$$= 2^k \cdot T(n-k) + \sum_{i=0}^{k-1} (2^i)$$

Put $k=n$

$$\therefore T(n) = 2^n \cdot T(0) + \sum_{i=0}^{n-1} (2^i)$$

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

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

$$\therefore T(n) = O(1)$$

$$(A-5) \quad i=1, s=1;$$

while ($s \leq n$) {

$i++$; $s += i$;

 print ("#");

}

Values of i & s will change as -

i	1	2	3	4	5	...
s	1	3	6	10	15	...

$$s_i = s_{i-1} + i, \text{ where } i = 1, 2, 3, \dots, k$$

$$\text{Sum of A.P. is } \frac{n(n+1)}{2} = \frac{k(k+1)}{2} \therefore T(n) = O(\sqrt{n})$$

A-6) void function (int n) {
 int i, count = 0;
 for (i = 1; i * i <= n; i++)
 count++;
}

Values of i will be: 1, 2, ..., k

where $k = \sqrt{n}$

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

A-7) void function (int n) {
 int i, j, k, count = 0;
 for (i = n/2; i <= n; i++) {
 for (j = 1; j <= n; j = j * 2) {
 for (k = 1; k <= n; k = k * 2) {
 count++;
 }
 }
 }
}

Values of i will change as: $n/2, (\frac{n}{2})+1, (\frac{n}{2})+2, \dots, n$
 $\Rightarrow (\frac{n}{2})+1 \text{ times} = O(n)$

Values of j & k change as: 1, 2, 4, 8, 16, ...

The complexity of these loops will be $O(\log n)$

$$\begin{aligned} T(n) &= \Theta(\log n) \cdot \Theta(\log n) \cdot \Theta(n) \\ &= \Theta(n \cdot (\log n)^2) \end{aligned}$$

A-8) function (int n) {

```

    if (n == 1) return;
    for (i = 1; i < n; i++) {
        for (j = 1; j < n; j++)
            cout << "*";
    }
    function(n - 3);
}

```

Values of n will be $n, n-3, n-6, \dots, 1 \Rightarrow O(n)$

Complexity of i loop = $O(n)$

Complexity of j loop = $O(n)$

$$\therefore T(n) \leq O(n^2)$$

A-9) void function (int n) {

```

    for (i = 1; i < n; i++) {
        for (j = 1; j <= n; j += i)
            cout << "*";
    }
}

```

Complexity of i loop is $O(n)$

Complexity of j loop = $O(\sqrt{n})$

$$\therefore T(n) \leq O(n \cdot \sqrt{n})$$

A-10) void function (int n) {

int j = 1, i = 0;

while (i < n) {

$i += j; \quad j++; \quad \{ \quad \}$

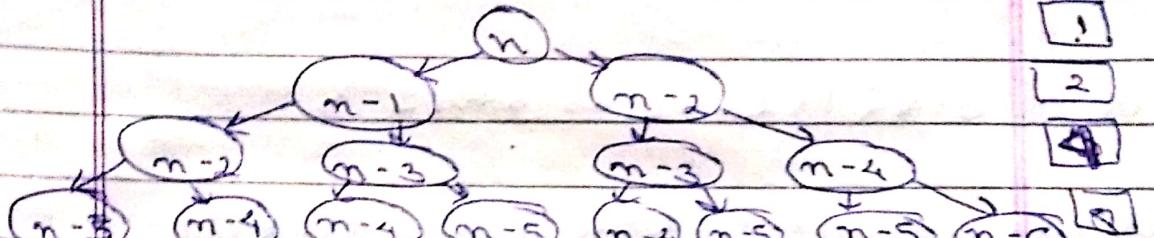
Complexity = $O(\sqrt{n})$

A-12) $\text{int fib (int } n\text{) } \{$

if ($n \leq 1$) return n ;

return fib ($n-1$) + fib ($n-2$); }

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



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

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

$$\Rightarrow O(2^n)$$

A-13) i) $n(\log n)$

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

`for (int j = 1; j <= n; j = j * 2) {`

`cout++; } }`

ii) n^3

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

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

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

`cout++; } }`

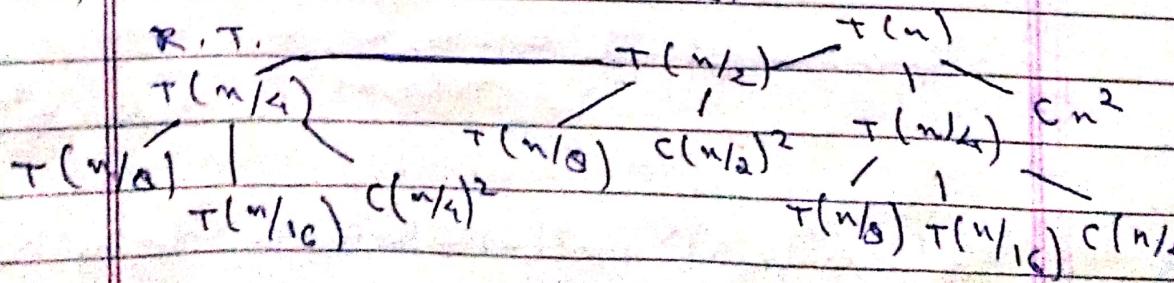
iii) $\log(\log n)$

`for (int i = n; i >= 1; i = sqrt(i)) {`

`cout++; }`

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

R.T.



No. of nodes per level will be 1, 3, 6, 12, ...

∴ Total no. of nodes = $1 + 3 + (2^{n/2} - 1)$

$$(2-1)$$

$$= 1 + 3 \cdot (2^{n/2} - 1) = 2^n = O(2^n)$$

A-15) $O(\sqrt{n})$, same as ans 5 & 11

A-16) for (int i=2; i<=n; i = pow(i, k)) {
 count++; }

Values of i will be as follows:

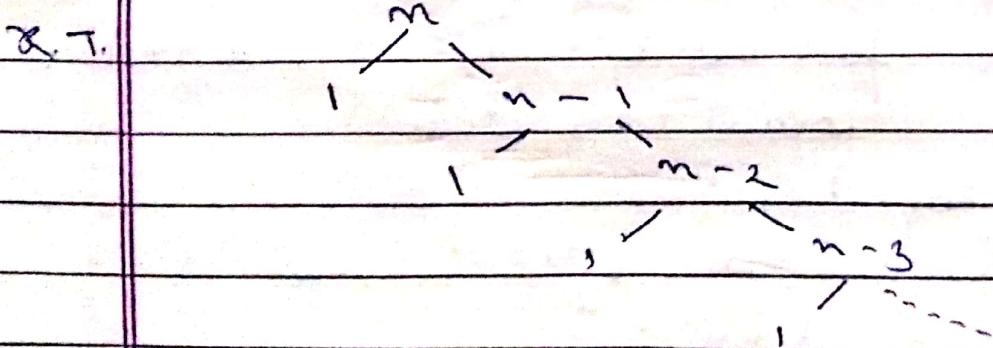
$$2, 2^k, 2^{k^2}, 2^{k^3}, \dots, 2^{k^c}$$

$$\text{As } 2^{k^c} \leq n \Rightarrow k^c \cdot \log_2 2 = \log_2 n$$

$$\Rightarrow k^c = \log_2 n \Rightarrow c \cdot \log_2 k = \log_2 (\log_2 n)$$

$$\Rightarrow c = \log_2 (\log_2 n) \Rightarrow T(n) = O(\log n (\log \log n))$$

A-17) Quick sort will repeatedly divide array into 2 parts of $\lceil \frac{n}{2} \rceil$ & $\lfloor \frac{n}{2} \rfloor$, i.e. when pivot chosen by partition is always either the smallest or largest element in array.
 $\therefore T(n) = T(n-1) + 1 \Rightarrow T(1) = 1$



Using forward substitution,

$$T(1) = 1, T(2) = 2, T(3) = 3, \dots, T(n) = n$$

$$\therefore \text{Time complexity} = 1 + 2 + 3 + \dots + n$$

$$= n(n+1)/2 = O(n^2)$$

A-18) $100 < \log(\log n) < \sqrt{n} \leq \log(n) \leq \log(n) \leq \log(\log n!) \leq n \log(\log n!) < n!$

$$2^n = \underline{n^2} < 2^{2n} = 4^n$$

b) $1 < \sqrt{\log n} \leq \log(\log n) < \log(n) < 2\log(n) < \log(2n)$
 $\leq n < \underline{2^n} < 4^n < \log(n!) < n \log n < n^2 < 2(2^n) \leq n^3$

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

Q-19)

index = 0

while (index < n)

if (arr[index] == key)

break

index += 4

if (index == n) // not found

else // found

A-2a) Insertion Sort
with insertion sort (at i from n-1)
for last $i = 0 \rightarrow i < n-1$
and $\text{start} = \text{arr}[i+1]$,
 $\text{last} = \text{arr}[n]$,

while ($j >= 0 \wedge \text{arr}[j] > \text{arr}[j+1]$)
 $\text{arr}[j+1] = \text{arr}[j]$,

$j = j - 1$

$\text{arr}[j+1] = \text{last}$;

Previous insertion sort

with insertion sort (at arr[i] until $i \neq$
 $\text{arr}[i+1]$)

elsewise

insertion sort (arr, n-1),

int last = arr[n-1],

int j = n-2,

while ($j >= 0 \wedge \text{arr}[j] > \text{arr}[j+1]$)

$\text{arr}[j+1] = \text{arr}[j]$,

$j = j - 1$

$\text{arr}[j+1] = \text{last}$;

Insertion sort is called a in-place sort because we can add new elements to array's end while sorting by being inserted. At any given iteration, say iteration i , only first k elements of array participate in sorting. Therefore we can add new elements to array during sort.

A-21) Complexity of all algorithms discussed in class -

	Best	Average	Worst
• Bubble	$O(n^2)$	$O(n^2)$	$O(n^2)$
• Selection	$O(n^2)$	$O(n^2)$	$O(n^2)$
• Insertion	$O(n)$	$O(n^2)$	$O(n^2)$
• Merge	$O(n \cdot \log n)$	$O(n \cdot \log n)$	$O(n \cdot \log n)$

A-22) • Bubble sort : in-place, stable, offline

- Selection sort : in-place, unstable, offline
- Insertion sort : in-place, stable, online
- Merge sort : not in-place, stable, offline

A-23) 24) Recursive Binary Search -

```
int binary-search (int arr, int l, int r){\n    if (l <= r) {\n        m = l + (r - l) / 2;\n        if (arr[m] == key)\n            return m;\n        else if (arr[m] < key)\n            return binary-search (arr, m+1, r);\n        else\n            return binary-search (arr, l, m-1);\n    }\n}
```

• Recurrence relation for recursive binary search :-

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

• Time complexity of recursive binary search :

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

• Time complexity of iterative binary search : $T(n) = O(\log n)$

• Time complexity of linear search $\rightarrow T(n) = O(n)$

• Space complexity in all cases = $O(1)$