GUIDELINES FOR ASYMPTOTIC ANALYSIS-PART 1A

[BASED ON SOME SUB-CODES OF PROGRAMS]

TILL NOW WE HAVE LEARNT THAT:

Big - Oh or Worst Case Complexity:

 $f(n) \le cg(n)$, where c and n_0 are constants and $n \ge n_0$, then f(n) = O(g(n))

<u>Big - Omega or Best Case Complexity:</u>

 $f(n) \ge cg(n)$, where c and n_0 are constants and $n \ge n_0$, then $f(n) = \Omega(g(n))$

$\underline{\textit{Big-Theta or Average Case Complexity}}:$

 $c_1g(n)\leq f(n)\leq c_2g(n), where~c~, n_1~and~n_2~are~constants~and~n\geq \{n_1,n_2\},$ $then~f(n)=\Theta\big(g(n)\big)$

And from an algorithm the priority is to find Worst Case Time Complexity.

1. LOOPS: The running time of a loop is, at most, the running time of the statements inside the loop (including tests) multiplied by the number of iterations.

```
//executes\ n\ times
for(i = 1; i \le n; i + +)\{
m = m + 2; //constant\ time, c
\}
```

Total time = contant time $c \times n = O(n)$.

Evaluation:

We can say outer loop executes 1 to n times = O(n) for outer loop.

Again, we can say inner loop runs or iterates n times 1 to n =

That is:

```
i=1, i terates 1 time, i=2, i terates 1 time i=3, i terates 1 time, i=4, i terates 1 time Similarly, up to n time i=n, i terates 1 time.
```

And we know by counting technique that : When 1 counted to n times gives $1 \times n = n$.

Hence O(n).

2. NESTED LOOPS: Analyse from the inside out. Total running time is the product of the sizes of all the loops.

```
\begin{tabular}{ll} //outer loop executed $n$ times \\ for (i=1;i\leq n;i++)\{ \\ //inner loop executes $n$ times \\ for (j=1;j\leq n;j++)\{ \\ k=k+1;// \ constant \ time. \\ \end{tabular}
```

So, 1^{st} For loop will be executed *n times*.

Or, 2nd For Loop will be executed as depending upon the first for loop as shown below:

i=1	i=2	i = 3	i = 4	 i = n
j=1 to n	j= 1 to n	j=1 to n	j=1 to n	 j=1 to n

Hence inner loop also executes n times:

Which gives us the Multiplication rule:

Total Time =
$$c \times n \times n = cn^2 = O(n^2)$$

Evaluation:

We can say outer loop executes 1 to n times = O(n) for outer loop.

Again, we can say inner loop runs or iterates n times 1 to n =

$$i = 1$$
, $k = \{1, 2, 3...n\}$ i. e k iterates 1 time $i = 2$, $k = \{\{1, 2, 3...n\}, \{1, 2, 3, ..., n\}\}$ i. e k iterates 2 times.

.....

i = n,

 $k = \{\{1, 2, 3...n\}, \{1, 2, 3, ..., n\}, \{1, 2, 3, ..., n\}\}$

i.e k iterates n time.

$$1 + 2 + 3 + 4 + \cdots + n - 1 + n$$
 times

By arithmetic series(Arithmetic Progression to find general term):

$$\Rightarrow S(n) = \frac{n}{2} ((2 \times a) + ((n-1) \times (d)))$$

Where, a = First Term.

 $d = (T_n - T_{n-1})$ or it can be 2^{nd} term – $(minus)1^{st}$ term.

i.e. the common difference.

 $T_{n-1} = Second\ Last\ term \implies n-1.$ $T_n = Last\ Term \implies n.$ $n-1 = Second\ last\ term\ i.\ e.\ T_{n-1}.$

$$\Rightarrow S(n) = \frac{n}{2} ((2 \times 1) + ((n-1) \times (1)))$$

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

$$\Rightarrow S(n) = \frac{n}{2}(1+n)$$

$$= \frac{n(n+1)}{2} = \frac{n^2 + n}{2} = 0\left(\frac{1}{2} \times n^2 + \frac{1}{2} \times n\right)$$

By Constant Rule: $O(k \times n) = O(n)$, where k is constant.

$$0\left(\frac{1}{2} \times n^2 + \frac{1}{2} \times n\right) = 0(n^2 + n) = 0(n^2)$$

Hence inner loop iterates dominates over outer loop i.e.

By addition rule (Outer Loop + Inner Loop):

$$\mathbf{O}(n) + \mathbf{O}(n^2) = \mathbf{O}\{max(n^2 + n)\} = \mathbf{O}(n^2)$$

Also, how much inner loop will be iterated , it is depended over outer loop.