

Ans: to the Que! No! 2

Implementation! 1

base case!

if $n \leq 0$
and $n \leq 2$

$$T(1) = 1$$

$$T(2) = 2$$

$$T(0) = \text{invalid input}$$

$$T(n) = T(n-1) + T(n-2) + O(1) \quad \left[\begin{array}{l} \text{two base case} = O(2) \\ \text{one addition} = O(1) \\ \text{two subtraction} = O(2) \end{array} \right]$$

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

$$[T(n-1)]$$

$$= \text{approximate } T(n-2)$$

$$\Rightarrow O(1)$$

$$T(n-1) \Rightarrow$$

$$\Rightarrow 2(2T(n-2) + 1) + 1$$

$$T(n-2) \Rightarrow 2(2(2T(n-3) + 1) + 1) + 1$$

$$\Rightarrow 2^n + (n \times 1)$$

$$\Rightarrow 2^n$$

above,
 we are getting 1 whenever we call the function.
 thus we will get n times of 1 [From $(n-1)$ to $(n-n)$]
 So the time complexity will be $O(n)$.

again
 the recursive function is giving us a 2 as its
 multiplication. Thus, the time complexity will be,

$$T(5) = 2 \cdot 2 \cdot 2 \cdot 2 \cdot 2 + 1$$

$$= 2^5 + 1$$

$$T(6) = 2^6 + 1$$

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

$$\therefore T(n) = 2^n + n$$

$$= 2^n$$

$$= O(2^n)$$

implementation-2 $(c-m) = m$ for $m \leq T$ and $n > T$
 big Ω \Rightarrow $\Omega(n)$ \Rightarrow $\Omega(n)$ \Rightarrow $\Omega(n)$
~~base case~~

$$n < 0 \Rightarrow O(1)$$

$$n \leq 2 \Rightarrow O(1)$$

big O:

$$O(n) + O(1) + O(1)$$

$$\Rightarrow O(n) + O(1)$$

$$\Rightarrow O(n)$$

For loop is iterating n times $\Rightarrow O(n)$

For example: $n=5$

thus iteration:

| | | |
|------------------|-------|---|
| (0-0) | | |
| (2-1) | (2-2) | 1 |
| (3-1) | (3-2) | 2 |
| (4-1) | (4-2) | 3 |

iteration: $n=3$ $T = (n-2)$
 $n=5$ $T = (n-2) = 3$

Thus we can say $n \approx (n-2)$ as a result don't the time complexity will be $O(n)$, / Ann.

$$(1)0 \in \dots$$

$$(1)0 \in \dots$$

$$(1)0 + (1)0 + \dots (n)0$$

$$(1)0 + (n)0 \in$$

$$(n)0$$

Problem 4)

initializing = $O(1)$ iteration = $O(1)$

First for loop = $O(n)$

Second " " = $O(n)$

Third " " = $O(n)$

∴ Thus, all are nested,

$$\begin{aligned} O(T) &= O(n) \times O(n) \times \cancel{O(n)} \times O(n) \times 1 \\ &= O(n^3) \times 1 \\ &= O(n^3) \end{aligned}$$

| beac First For loop | 2nd For loop | 3rd For loop | total |
|-----------------------------------|--------------|--------------|-------------------|
| 1 | 1 | 1 | 1 |
| 2 | 4 | 8 | 8 |
| 3 | 9 | 27 | 27 |
| n | n^2 | n^3 | $\Rightarrow n^3$ |

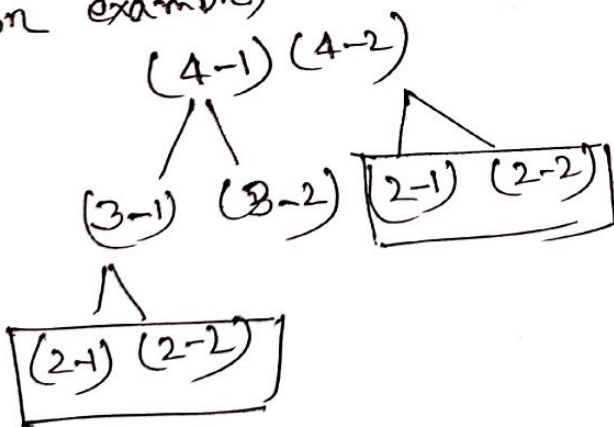
From above we can see every loop is in multiplication
∴ thus final time complexity $n \times n \times n = O(n^3)$

Ans to the ques no :3

The difference between fibonacci 1 and 2 algorithm is clearly showed after generating the time complexity graphs. we can see that till $n=15$ or 16 both algorithm worked in constant time ($O(1)$).

After $n \geq 16$ there is a difference. The red line (dibo-1) has increased suddenly. Thus we can say till $(14-16)$ both algorithm work same time or gone through approx same amount of iteration. But after that the recursive algo showed some dissimilarity well because recursive algo gone throw same number twice sometimes.

For example,



$$\Rightarrow 2^n$$

But in iterative algo the loop gone through n times
only. Thus the ~~at~~ ~~of~~ big O is not more than
 n . As input increases we can see red line is
increasing where blue stay as its previous position.