

# Time Complexity in DSA

```
1. for i in range(n): # c time
    i = i*2 # n*c

# Time Complexity(TC) = n*c = O(n)
```

So, the time complexity of the above code is  $O(n)$ .

```
2. def fact_iter(n): # c1 time
    "assumes n an int >= 0"
    answer = 1 # c2 time
    while n > 1: # c3 time
        answer *= n # n*c4 time
        n -= 1 # n*c5 time
    return answer # c6 times

#Time Complexity(TC) = c1 + c2 + c3 + n*c4 + n*c5 + c6 = O(n)
```

So, the worst time complexity of the above code is  $O(n)$ .

```
3. A = [1, 2, 3, 4] # c1
   B = [2, 3, 4, 5, 6] # c2
   for i in A: # c3
       for j in B: # c4
           if i<j:
               print('{} {}'.format(i, j)) # n*m*c5

#Time Complexity (TC) = c1 + c2 + c3 + c4 + n*m*c5 = O(n*m)
```

So, the time complexity of the above code is the  $O(n*m)$

```
4. L = [1, 2, 3, 4, 5, 6, 7, 8] # c1
   for i in range(len(L)//2): # c2
       other = len(L) - i - 1
       temp = L[i]
       L[i] = L[other]
       L[other] = temp # len(n)/2*c3
```

Here, the code is for the swapping, so the above code will run for  $n/2$  times the length of the list.

Time Complexity =  $O(n)$

```
5. def fib(n):  
    if n==1 or n==0:  
        return 1  
    else:  
        return fib(n-1) + fib(n-2)
```

This is the recursive function to calculate the Fibonacci of the given number  $n$ . So, the time complexity

Time complexity (TC) =  $2^n$ .