

## Recurrence Relation

1. What is the time & space complexity of this algorithm?

```
# Q1
def mystery(N):
    if N == 1:
        return 1
    else:
        return mystery(N-1)
```

①

$$T(N) = \begin{cases} T(N-1) + a & ; N > 1 \\ b & ; N = 1 \end{cases}$$

|  
constant

\* identify which var is decreasing in each func call.  
N becoming smaller / smaller till we reach base case.

*Telescoping*

$$\begin{aligned} T(N) &= T(N-1) + a \\ &= (T(N-2) + a) + a = T(N-2) + 2a \\ &= (T(N-3) + a) + 2a = T(N-3) + 3a \\ &= \underline{T(N-k) + ka} \quad \text{general form} \end{aligned}$$

↑  
eliminate function call (n=1)  
we want to reach base case.

*Base when  $T(1)$*

$$\begin{aligned} \therefore N-k &= 1 \\ k &= N-1 \end{aligned}$$

when  $k = N-1$

$$\begin{aligned} T(N) &= T(N-(N-1)) + (N-1)b \\ &= T(N-N+1) + bN - b \\ &= T(1) + bN - b \\ &= a + bN - b \\ &= O(N) \end{aligned}$$

Time complexity:  $O(N)$

Space complexity:  $O(N)$  = Input  $O(1)$  + Aux  $O(N)$  (Recursion height)

2. What is the time & space complexity of this algorithm?

```
# Q2
def mystery(N):
    if N == 1:
        return 1
    else:
        return mystery(N/2) + N
```

②

$$T(N) = \begin{cases} T(N/2) + a & ; N > 1 \\ b & ; N = 1 \end{cases}$$

\*  $\text{mys}(N/2) + \textcircled{N}$  — is a constant

Telescoping

$$\begin{aligned} T(N) &= T(N/2) + a \\ &= (T(N/4) + a) + a = T(N/2^2) + 2a \\ &= (T(N/8) + a) + 2a = T(N/2^3) + 3a \end{aligned}$$

$$\boxed{= T(N/2^k) + ka} \quad \text{general form}$$

Base when  $T(1)$

$$\therefore N/2^k = 1$$

$$N = 2^k$$

$$k = \log_2 N$$

$$\begin{aligned} T(N) &= T(N/2^{\log_2 N}) + (\log_2 N) b \\ &= T(1) + \log_2 N b \\ &= a + b \log_2 N \\ &= O(\log N) \quad \# \end{aligned}$$

Time complexity:  $O(\log N)$

Space complexity:  $O(\log N)$  = Input  $O(1)$  + Aux  $O(\log N)$  (Recursion height)

3. What is the time & space complexity of this algorithm?

```
# Q3
def mystery(N):
    if N == 0:
        return 0
    else:
        total = 0
        for _ in range(3):
            total += mystery(N-1)
        return total
```

$$\textcircled{3} \quad T(N) = \begin{cases} 3T(N-1) + a & ; N > 0 \\ b & ; N = 0 \end{cases}$$

*Telescoping*

$$\begin{aligned} T(N) &= 3T(N-1) + a \\ &= 3(3T(N-2) + a) + a = 3^2 T(N-2) + 3a + a \\ &= 3^2(3T(N-3) + a) + 3a + a = 3^3 T(N-3) + 3^2 a + 3a + a \\ &= 3^k T(N-k) + a(3^0 + 3^1 + 3^2 + \dots + 3^{k-1}) \end{aligned}$$

$$\frac{r^{n+1} - 1}{r - 1} \quad r=3 \quad n=k-1 = \frac{3^{k-1+1} - 1}{3 - 1} = \frac{3^k - 1}{2}$$

$$\boxed{= 3^k T(N-k) + a \left( \frac{3^k - 1}{2} \right)} \quad \text{general form}$$

*Base when  $T(0)$*

$$\therefore N - k = 0$$

$$k = N$$

*when  $k = N$*

$$T(N) = 3^N T(0) + a \left( \frac{3^N - 1}{2} \right)$$

$$= 3^N b + \frac{1}{2} a (3^N - 1)$$

$$= O(3^N) \quad \#$$

**Time Complexity:  $O(3^N)$**

**Space Complexity:  $O(N)$  = Input  $O(1)$  + Aux  $O(N)$  (Why? Because it will allocate and deallocate memory as it recurse and returns.)**

#### 4. What is the time & space complexity of this algorithm?

```
# Q4
def mystery(N):
    if N == 0:
        return 0
    else:
        value = 0
        for i in range(N):
            value += i
        return value + mystery(N-1)
```

④  $T(N) = \begin{cases} NT(N-1) + Nb + c & \text{OR } T(N-1) + Nb + c; N > 0 \\ a & ; N = 0 \end{cases}$

*be careful! function calling not in for loop,*

*Telescoping*

$$T(N) = T(N-1) + Nb + c$$

$$= (T(N-2) + (N-1)b + c) + Nb + c = T(N-2) + (N-1)b + Nb + 2c$$

$$= (T(N-3) + (N-2)b + c) + (N-1)b + Nb + 2c = T(N-3) + Nb + (N-1)b + (N-2)b + 3c$$

$$= T(N-k) + \left( \sum_{i=N-k+1}^N i \right) b + kc$$

$$\left( N + (N-1) + (N-2) + \dots + (N-k+1) \right) b$$

$$T(N-k) + \left( \sum_{i=n-k+1}^n i \right) b + kc$$

Base when  $T(0)$

$$\therefore N-k = 0$$

$$k = N$$

$$\begin{aligned} \text{When } k = N \\ T(N) &= T(N-N) + \left( \sum_{i=\underline{n-n+1}}^n i \right) b + Nc \\ &= T(0) + (1+2+\dots+N)b + Nc \\ &= a + \left( \frac{n(n+1)}{2} \right) b + Nc \\ &= O(N^2) \end{aligned}$$

**Time Complexity:  $O(N^2)$**

**Space Complexity:  $O(N)$  = Input  $O(1)$  + Aux  $O(N)$  (Why? Because the loop doesn't do anything for the space complexity).**

## 5. Identify the space & auxiliary complexity of this algorithm

```
# Q5
def binarySearch(array, left, right, key):
    if right >= left:
        mid = left + (right - left) // 2

        if array[mid] == key:
            return mid
        elif array[mid] > key:
            return binarySearch(array, left, mid-1, key)
        else:
            return binarySearch(array, mid+1, right, key)

    return -1
```

**Auxiliary Space :** The extra or temporary space used by an algorithm. (In other words, space that doesn't account for a function's/algorithm's input size).

The above is recursive implementation of binary search:

Auxiliary space:  $O(\log N)$  (recursive height)

Space complexity:  $O(N)$

If it's iterative implementation of binary search (Ian's lecture video):

Auxiliary Space:  $O(1)$  (in-place)

Space complexity:  $O(N)$

As you can see here, comparing the space used by both types of function (recurrence and iterative), although both of their space complexity is  $O(N)$ . But with auxiliary, we can identify with ease the amount of additional space used by the functions.