



Algorithms

COMP 53 - DATA STRUCTURE

UNIVERSITY OF THE PACIFIC

Recursive Definition

- ▶ A recursive algorithm is an algorithm that breaks the problem into **smaller sub-problems** and applies the algorithm itself to solve the smaller sub-problems.
- ▶ **base case:** A case where a recursive algorithm completes without applying itself to a smaller sub-problem.
- ▶ **A recursive function** is a function that **calls itself**.

Factorial(N):

```
if (N == 1)  
    return 1
```

Base case

```
else  
    return N * Factorial(N - 1)
```

Non-base case

Recursive logic

Cumulative Summation

- ▶ Summation of all the numbers from 0 to N

```
CumulativeSum(N)  {
    if (N == 0)
        return 0
    else
        return N + CumulativeSum(N - 1)
}
```

Reverse a List

```
ReverseList(list, startIndex, endIndex) {  
    if (startIndex >= endIndex)  
        return  
    else {  
        Swap elements at startIndex and endIndex  
        ReverseList(list, startIndex + 1, endIndex - 1)  
    }  
}
```

Fibonacci Sequence

- ▶ a numerical sequence where each term is the sum of the previous 2 terms in the sequence
- ▶ except the first 2 terms, which are 0 and 1

```
FibonacciNumber(termIndex) {  
    if (termIndex == 0)  
        return 0  
    else if (termIndex == 1)  
        return 1  
    else  
        return FibonacciNumber(termIndex - 1) + FibonacciNumber(termIndex - 2)  
}
```

Runtime Complexity of Recursive functions

- ▶ The runtime complexity $T(N)$ of a recursive function will have function T on both sides of the equation
- ▶ The runtime complexity $T(N)$ will be equal to
 - ▶ the runtime of all of the statements in the function
 - ▶ plus the runtime of the function called inside with the new input size
- ▶ Such a function is known as a **recurrence relation**
 - ▶ A function $T(N)$ that is defined in terms of T on a value $< N$.

Runtime Complexity of Recursive functions

- ▶ Using O-notation to express runtime complexity of a recursive function requires solving the recurrence relation.
- ▶ For simpler recursive functions, it can be determined by expressing the **number of function calls**
- ▶ Other useful tool for solving recurrences is a **recursion tree**, a visual diagram of operations done by a recursive function

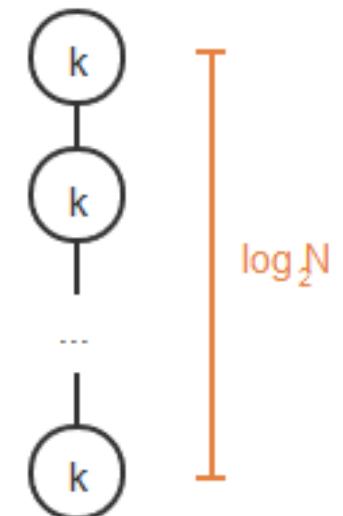
Runtime Complexity of Recursive functions

```
int logCalculator(int n) {  
    if (n == 1) {  
        return 0;  
    }  
    return 1 + logCalculator(n / 2);  
}
```

Input Size	Number of function calls
1	1
2	2
4	3
8	4
16	5
32	6
...	...
n	$\log n + 1$

If: $T(N) = O(1) + T(N / 2)$
Then: $T(N) = O(\log N)$

$$T(N) = k + T(N / 2)$$



$$T(N) = k \times \log N$$

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

Runtime Complexity of Recursive functions

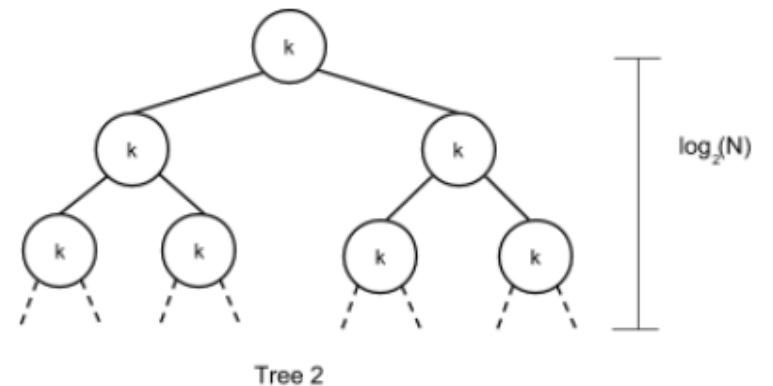
```

int cumulativeSum(int a[], int start, int end, int size) {
    if (size == 1) {
        return a[start];
    }
    return cumulativeSum(a, start, end/2, size/2 ) + cumulativeSum(a, end/2 +1, end, size - size/2);
}

```

Input Size	Number of function calls
1	1
2	3
4	7
8	15
16	31
32	63
...	...
n	$2n + 1$

If: $T(N) = O(1) + T(N / 2) + T(N / 2)$
 Then: $T(N) = O(N)$



$$\begin{aligned}
 T(N) &= k \times 1 + k \times 2 + k \times 4 + \dots + k \times N \\
 T(N) &= k \times (1 + 2 + 4 + \dots + N) \\
 T(N) &= k \times (2^0 + 2^1 + 2^2 + \dots + 2^{\log_2 N}) \\
 T(N) &= k \times (2^{\log_2 N + 1}) \\
 T(N) &= k \times (N + 1) \\
 T(N) &= O(N)
 \end{aligned}$$