

**CSC236H**

**Introduction to the Theory of Computation**

## Example – When we cannot use the Master Theorem

**Theorem** (Master Theorem). Let  $T : \mathbb{N} \rightarrow \mathbb{R}^+$  be a recursively defined function with recurrence relation

$$T(n) = c T\left(\frac{n}{d}\right) + f(n)$$

for some constants  $c, d \in \mathbb{Z}^+$ ,  $d > 1$ , and  $f : \mathbb{N} \rightarrow \mathbb{R}^+$ .

Furthermore, suppose  $f(n) \in \Theta(n^k)$  for some  $k \in \mathbb{R}, k \geq 0$ . Then:

1. if  $k = \log_d c$ , then  $T(n) \in \mathcal{O}(n^k \log n)$ ;
2. if  $k < \log_d c$ , then  $T(n) \in \mathcal{O}(n^{\log_d c})$ ;
3. if  $k > \log_d c$ , then  $T(n) \in \mathcal{O}(n^k)$ .

- **Example:**

$$T(n) = \begin{cases} a, & n = 1 \\ T(\frac{n}{2}) + 2^n, & n \geq 2 \end{cases}$$

## Example – When we cannot use the Master Theorem

**Theorem** (Master Theorem). Let  $T : \mathbb{N} \rightarrow \mathbb{R}^+$  be a recursively defined function with recurrence relation

$$T(n) = c T\left(\frac{n}{d}\right) + f(n)$$

for some constants  $c, d \in \mathbb{Z}^+$ ,  $d > 1$ , and  $f : \mathbb{N} \rightarrow \mathbb{R}^+$ .

Furthermore, suppose  $f(n) \in \Theta(n^k)$  for some  $k \in \mathbb{R}, k \geq 0$ . Then:

1. if  $k = \log_d c$ , then  $T(n) \in \mathcal{O}(n^k \log n)$ ;
2. if  $k < \log_d c$ , then  $T(n) \in \mathcal{O}(n^{\log_d c})$ ;
3. if  $k > \log_d c$ , then  $T(n) \in \mathcal{O}(n^k)$ .

- **Example:**

$$T_1(n) = \begin{cases} a, & n = 1 \\ 2T\left(\frac{n}{2}\right) + \log_2 n, & n \geq 2 \end{cases}$$

# Divide-and-Conquer Algorithms

```
1. divide-and-conquer(P):  
2.   if P has "small enough" size:  
3.     solve P directly  
4.   else:  
5.     # Solve c sub-problems of the same size, recursively  
6.     for i from 1 to c:  
7.       # Solve the sub-problem P_i recursively  
8.         s_i = divide_and_conquer(P_i)  
9.     combine(s_1, ..., s_c)
```

# Divide-and-Conquer Algorithms

```
""" Given a non-empty list A, sort the list in non-decreasing order. """
def MergeSort(A):
    2.     if len(A) == 1:
    3.         return A
    4.     else:
    5.         m = len(A) // 2 # integer division
    6.         L1 = MergeSort(A[0..m-1])
    7.         L2 = MergeSort(A[m..len(A)-1])
    8.         return merge(L1, L2)
def merge(A, B):
    9.     i = 0
    10.    j = 0
    11.    C = []
    12.    while (i < len(A)) and (j < len(B)):
    13.        if (A[i] <= B[j]):
    14.            C.append(A[i]) # Add A[i] to the end of C
    15.            i += 1
    16.        else:
    17.            C.append(B[j])
    18.            j += 1
    19.    return C + A[i..len(A)-1] + B[j..len(B)-1] # List concatenation
```

# The Master Theorem and Designing Efficient Recursive Algorithms

**Theorem** (Master Theorem). Let  $T : \mathbb{N} \rightarrow \mathbb{R}^+$  be a recursively defined function with recurrence relation

$$T(n) = c T\left(\frac{n}{d}\right) + f(n)$$

for some constants  $c, d \in \mathbb{Z}^+$ ,  $d > 1$ , and  $f : \mathbb{N} \rightarrow \mathbb{R}^+$ .

Furthermore, suppose  $f(n) \in \Theta(n^k)$  for some  $k \in \mathbb{R}$ ,  $k \geq 0$ . Then:

1. if  $k = \log_d c$ , then  $T(n) \in \mathcal{O}(n^k \log n)$ ;
2. if  $k < \log_d c$ , then  $T(n) \in \mathcal{O}(n^{\log_d c})$ ;
3. if  $k > \log_d c$ , then  $T(n) \in \mathcal{O}(n^k)$ .

- **Example:** Suppose  $rec\_func\_1(n)$  and  $rec\_func\_2(n)$  are two different algorithms for solving the same problem. Suppose  $T_1$  represents the running-time complexity of  $rec\_func\_1(n)$ , and  $T_2$  represents the running-time complexity of  $rec\_func\_2(n)$ , where

$$T_1(n) = \begin{cases} a_1, & n = 1 \\ 4T_1\left(\frac{n}{2}\right) + b_1, & n \geq 2 \end{cases} \quad T_2(n) = \begin{cases} a_2, & n = 1 \\ 3T_2\left(\frac{n}{2}\right) + n, & n \geq 2 \end{cases}$$

Which algorithm should we choose?

1. Analyze the running time of each of the following recursive algorithms.

- a) 1. `def sum(A):`  
2.     `if len(A) == 0:`  
3.         `return 1`  
4.     `else:`  
5.         `return A[0] + sum(A[1..len(A)-1])`
- b) 1. `def fun(A):`  
2.     `if len(A) < 2:`  
3.         `return len(A) == 0`  
4.     `else:`  
5.         `return fun(A[2..len(A)-1])`
- c) 1. `def double_fun(A):`  
2.     `n = len(A)`  
3.     `if n < 2:`  
4.         `return n`  
5.     `else:`  
6.         `return double_fun(A[0..n-2]) + double_fun(A[1..n-1])`

2. Find a closed-form expression for each of the following functions. Also use the Master Theorem to find an asymptotic upper-bound for each of them.

a) In your calculations and proof, you may assume that  $n$  is a power of 3.

$$T_1(n) = \begin{cases} a, & n = 1 \\ T_1(\frac{n}{3}) + 100, & n \geq 2 \end{cases}$$

b) In your calculations and proof, you may assume that  $n$  is a power of 2.

$$T_2(n) = \begin{cases} a, & n = 1 \\ 4T_2(\frac{n}{2}) + n^{1.5} - 1, & n \geq 2 \end{cases}$$

3. Find an asymptotic upper-bound for the following function

$$T(n) = \begin{cases} a, & n = 1 \\ 2T(\frac{n}{2}) + \log_{10} n, & n \geq 2 \end{cases}$$