

# Analyzing Merge Sort + Recurrence

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CSC 212

# Announcements

- Assignment 1 OUT
  - Due in three weeks (Oct 24, 11:59pm)
  - Find the link on the Schedule on the course webpage
- Quiz 2 on Tuesday
  - Will cover all the materials covered until today

# Recap: Merge Sort

**MergeSort** (A)

if A's size > 1

**Divide** array A in halves

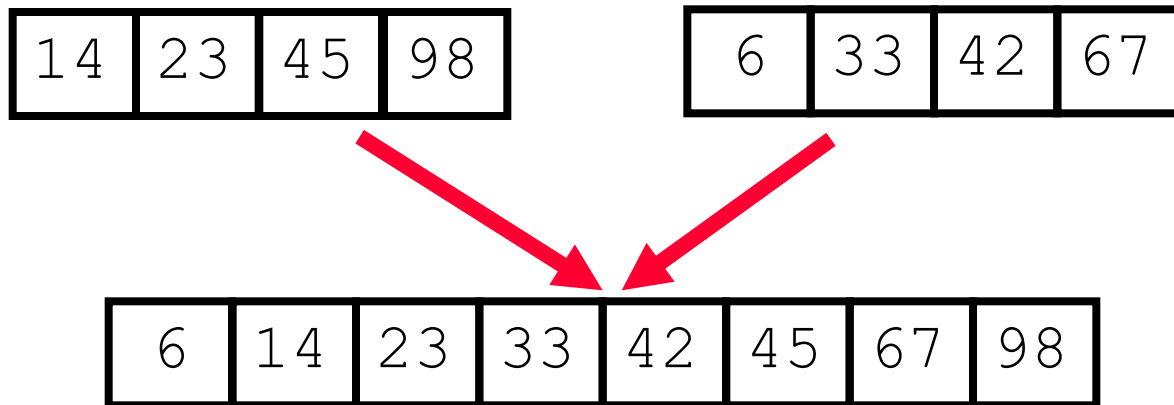
Call **MergeSort** on first half.

Call **MergeSort** on second half.

**Merge** two results (**combine**).

# Recap: How to Merge?

- **Merge the sub-problem solutions together:**
  - Compare the sub-array's first elements
  - Remove the smallest element and put it into the result array
  - Continue the process until all elements have been put into the result array



# Recurrence Relations

- When computing the complexity of Divide and Conquer algorithms, we have to consider:

- The **sub-problems created**
- The **size of the sub-programs**
- **Effort needed to create** the sub-problems
- **Effort needed to combine** the sub-problems

- Formally

- $T(n) = aT(n/b) + D(n) + C(n)$

Note that there a  $T(..)$  in the equation of  $T(n)$ ?

- Where

- $a$  is the number of sub-problems that  $n$  is divided into
- $b$  is the size of the sub-problem
- $D(n)$  is the complexity it takes to divide the problem
- $C(n)$  is the complexity of combining the solutions

# Merge Sort Recurrence (1)

- Recurrence relation for MergeSort(A)
  - $T(n) = ???$
- Every time MergeSort(A) is called:
  - How many sub-problems are created?
    - $a = ?$
  - What is the size of each sub-problem?
    - $b = ?$
  - What is the complexity of creating the sub-problems?
    - $D(n) = O(?)$
  - What is the complexity of merging them?
    - $C(n) = O(?)$

```
MergeSort(A)
    if A's size > 1
        Divide array A in halves
        Call MergeSort on first half.
        Call MergeSort on second half.
        Merge two results (combine).
```

# Merge Sort Recurrence (1)

- Recurrence relation for MergeSort(A)
  - $T(n) = 2T(n/2) + O(1) + O(n)$
- OR
  - $T(n) = 2T(n/2) + c*n$
- How do we solve  $T(n)$  and find out the overall complexity?
  - Unrolling the recursion (iterative method)
  - **Substitution method** (guess the answer and prove by induction)
  - **Master method** (memorize a few rules and apply them)

# Unrolling the Recursion

- $$\begin{aligned} T(n) &= 2 T(n/2) + cn \\ &= 2 (2 T(n/4) + n/2) + cn = 4 T(n/4) + 2cn \\ &= 4(2T(n/8) + n/4) + 2cn = 8 T(n/8) + 3cn \\ &\dots\dots \\ &= 2^k T(n/2^k) + k*cn \end{aligned}$$

- If  $\frac{n}{2^k} = 1 \Rightarrow k = \log_2 n$ . Then:

$$\begin{aligned} &= 2^{\lg n} T(1) + c*n \lg n \\ &= n + c*n \lg n \end{aligned}$$

OR

- $T(n) = O(n \lg n)$

Why do we do this?



# Example: Unrolling Recursion

- $$\begin{aligned} T(n) &= T(n-1) + c \\ &= T(n-2) + c + c \\ &= T(n-3) + c + c + c \\ &\dots \\ &= T(n-k) + k*c \end{aligned}$$

*If  $n-k = 1$ , then  $k = n-1$ . Therefore*

$$\begin{aligned} &= T(n-(n-1)) + (n-1)*c = T(1) + nc - c \\ &= nc - c \end{aligned}$$

$$T(n) = O(n)$$

# One more:

- $$\begin{aligned} T(n) &= 3T(n/2) + n^2 \\ &= 3(3T(n/4) + (n/2)^2) + n^2 = 3^2T(n/4) + 3/4 n^2 + n^2 \\ &= 3^2(3T(n/8) + (n/4)^2) + 3/4 n^2 + n^2 = 3^3T(n/8) + 9/16 n^2 + 3/4 n^2 + n^2 \\ &\dots \\ &= 3^k T(n/2^k) + n^2 (1 + 3/4 + 9/16 + 27/64 + \dots) \end{aligned}$$

If  $\frac{n}{2^k} = 1 \Rightarrow k = \log_2 n$ . Then:

$$3^{\lg n} T(1) + n^2 (1 + (3/4) + (3/4)^2 + \dots + (3/4)^{k-1})$$

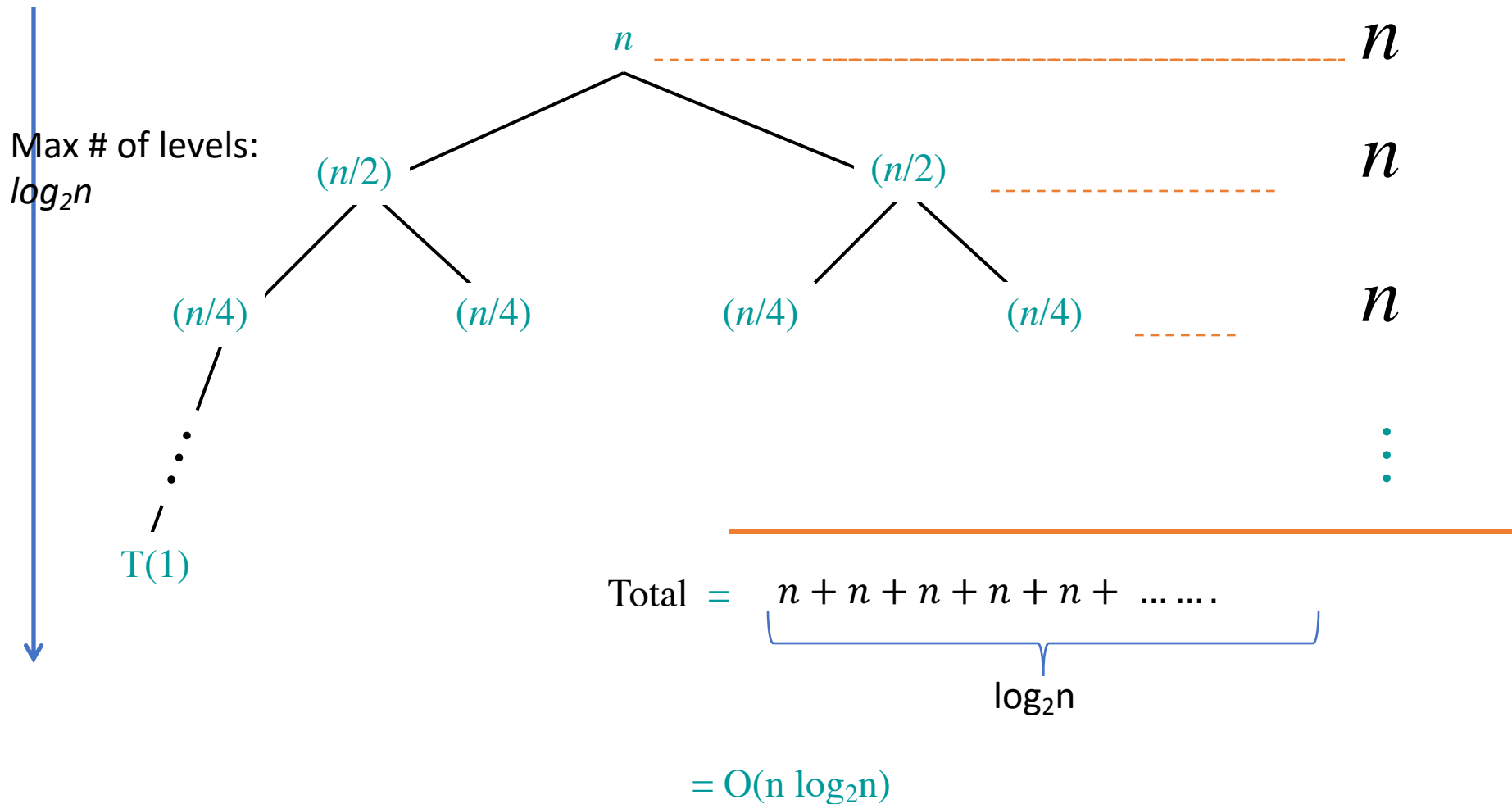
$$\sim 3^{\lg n} + n^2$$

$$= O(n^2)$$

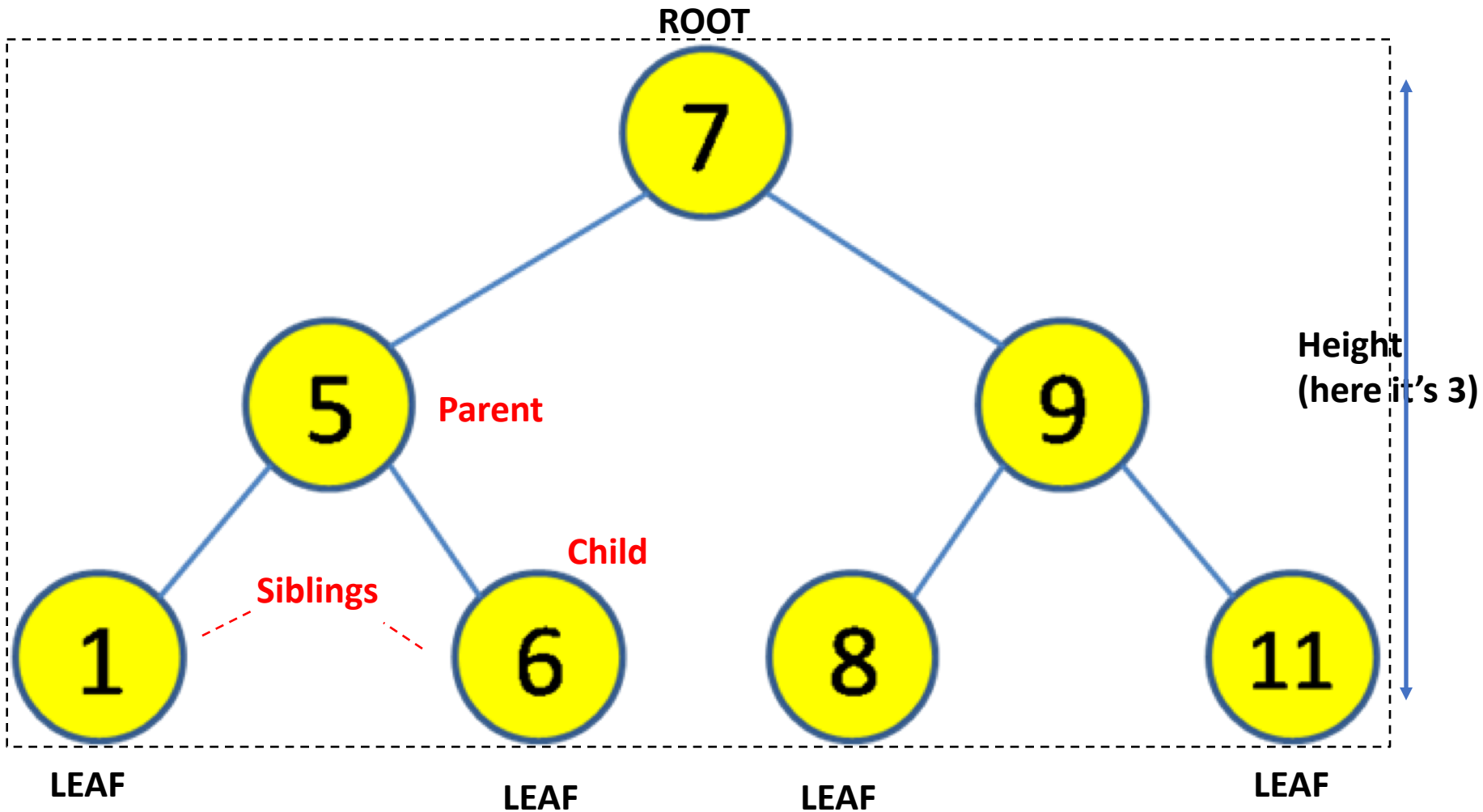
Infinitely decreasing geometric series

# Unspooling Visually

Solve  $T(n) = 2T(n/2) + n$ :



# Quick Note: A TREE Structure



(All leaf nodes need not be at the same level)

# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :

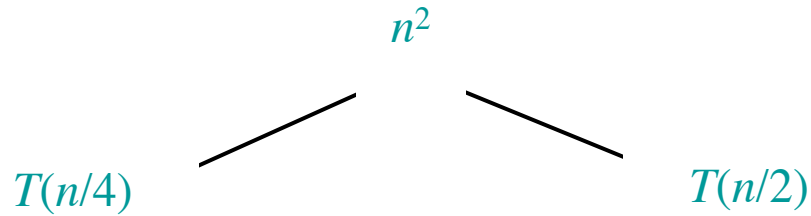
# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :

$$n^2$$

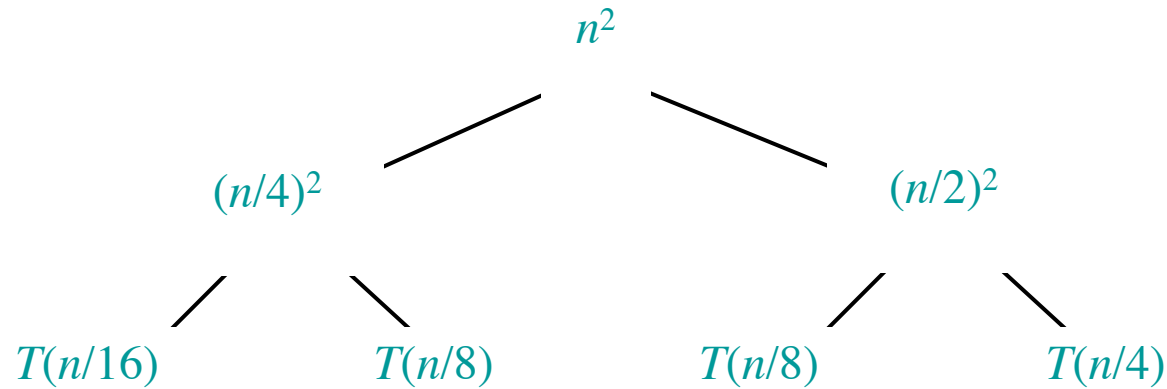
# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :



# Another Example of Recursion Tree

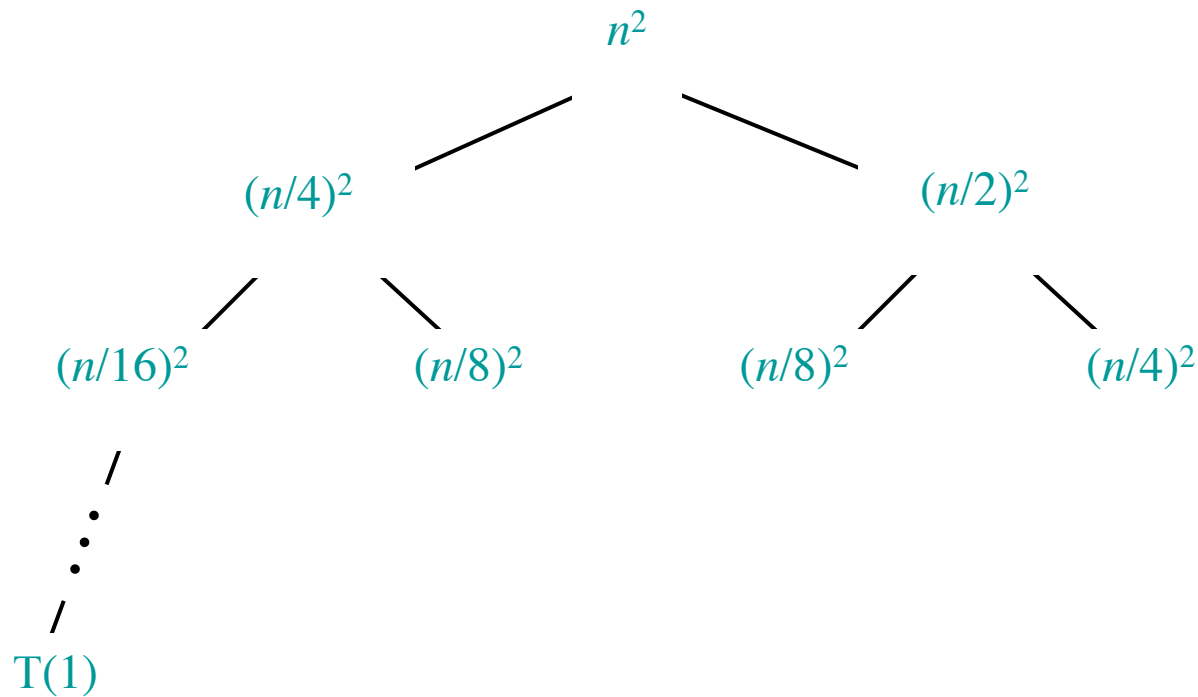
Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :





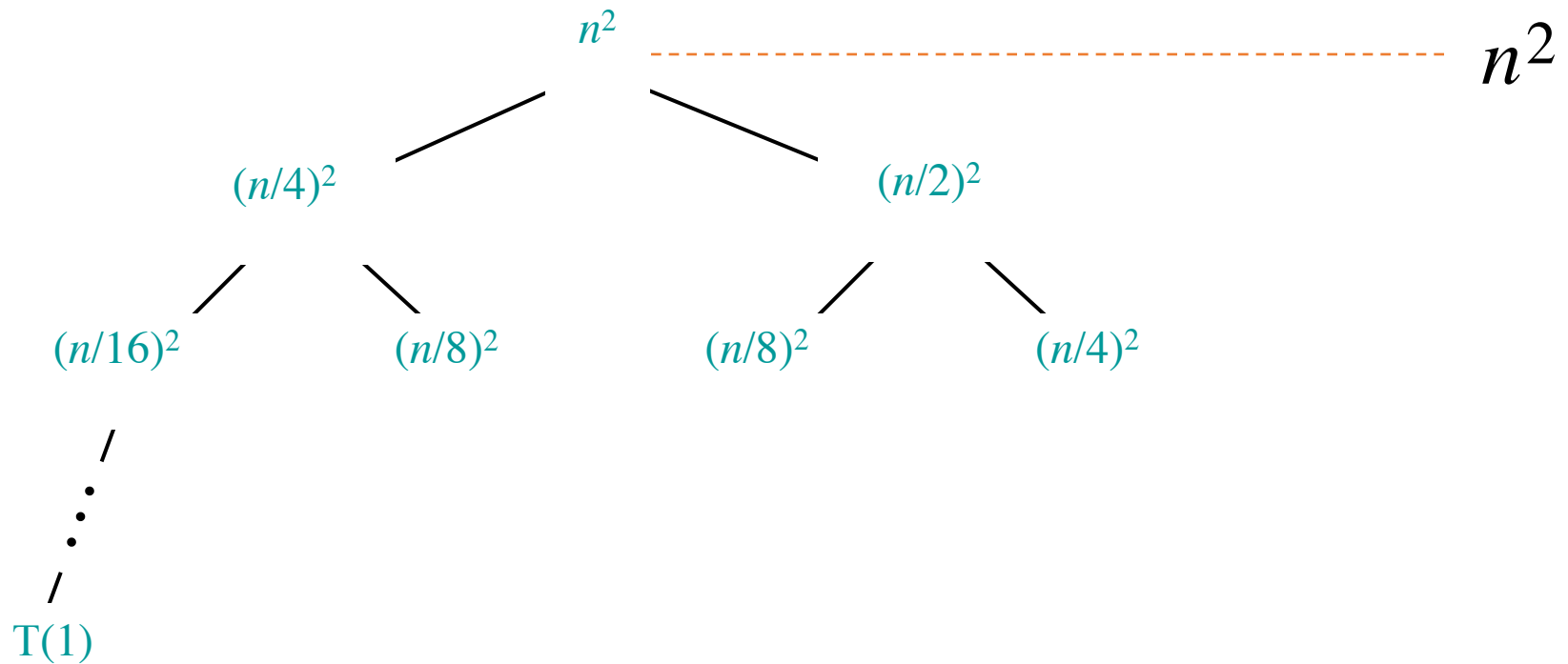
# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :



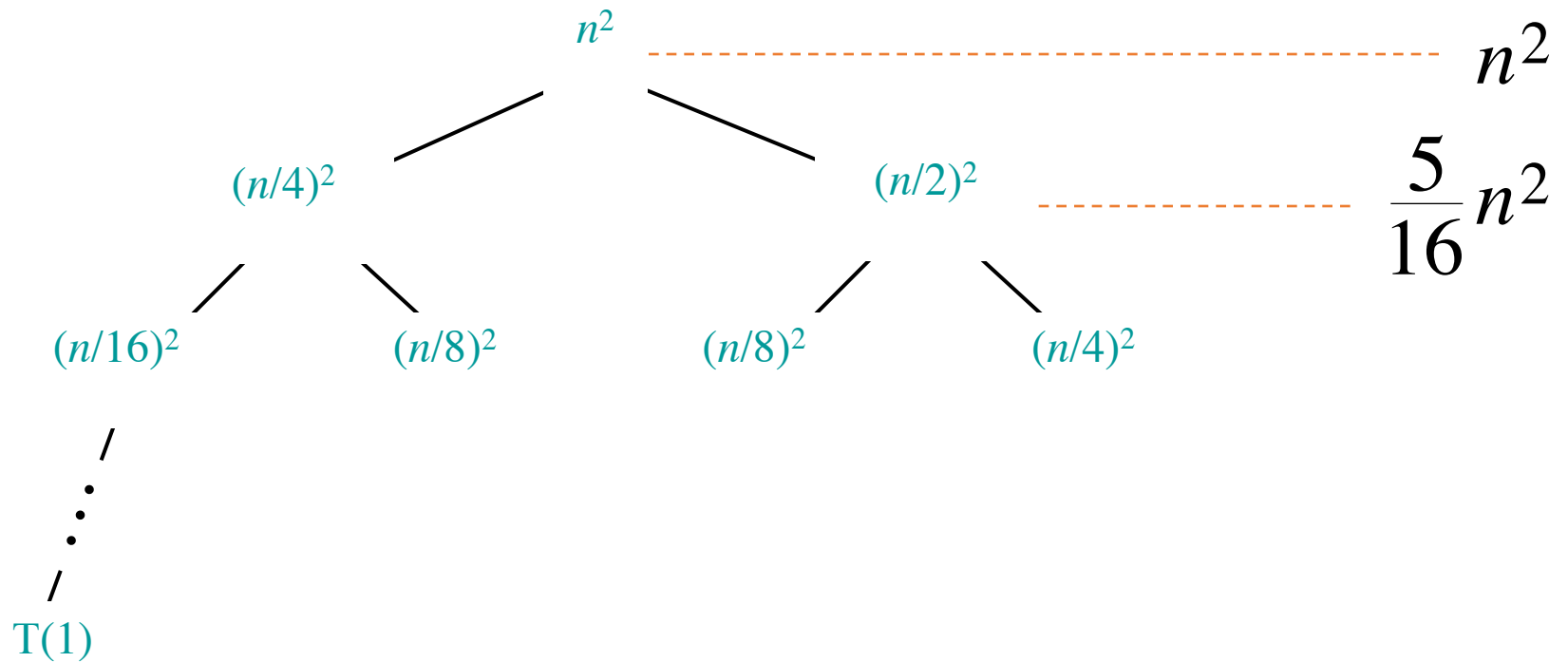
# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :



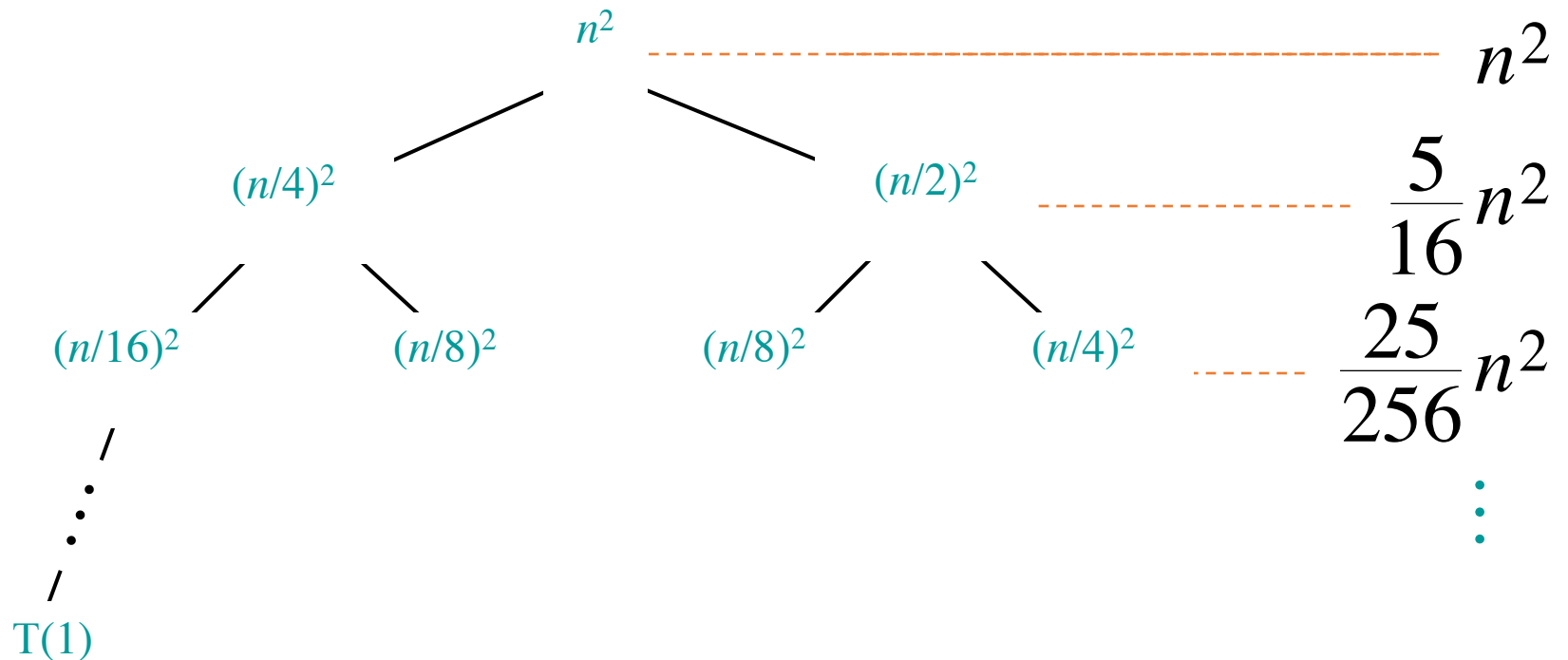
# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :



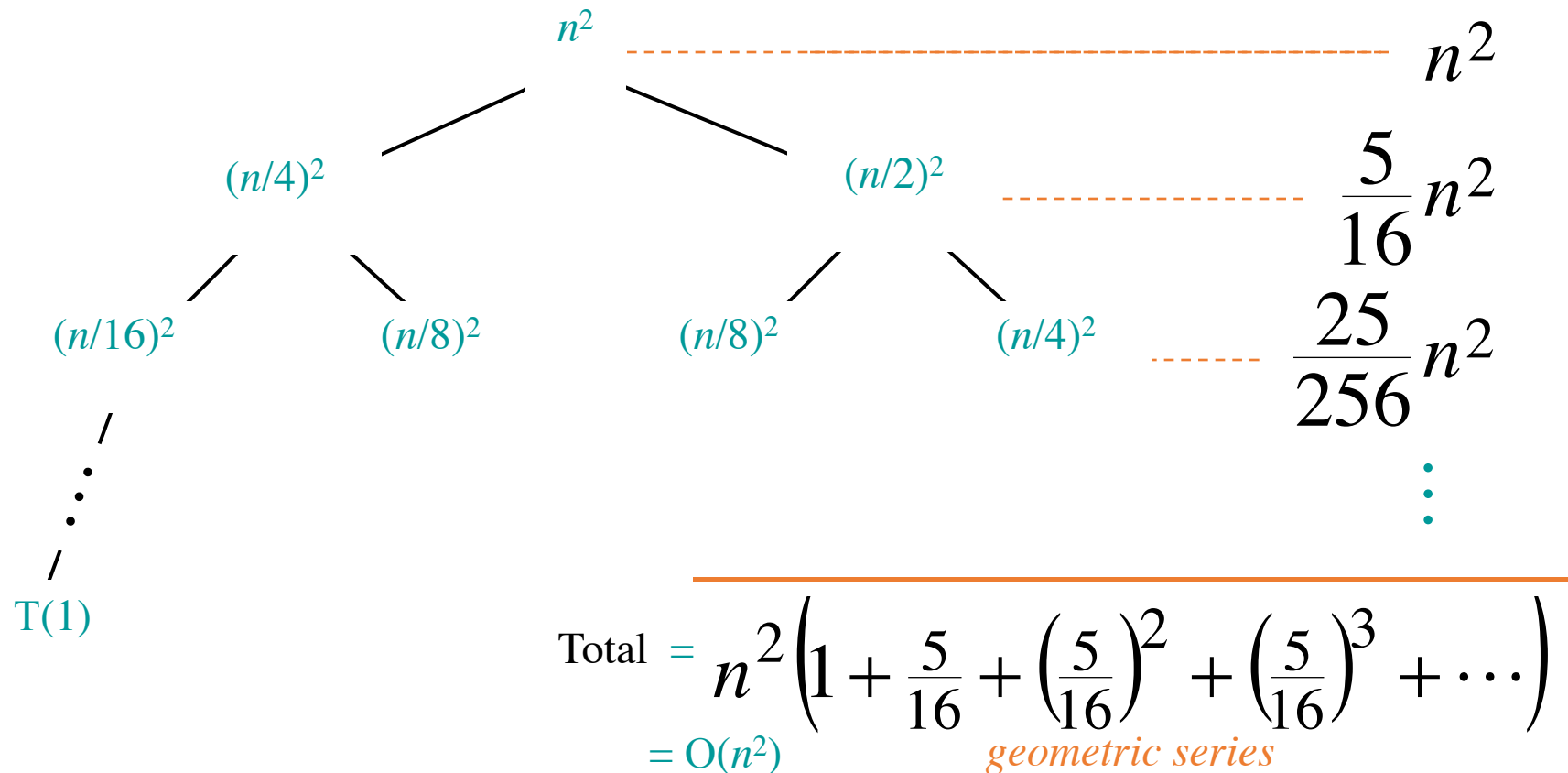
# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :



# Another Example of Recursion Tree

Solve  $T(n) = T(n/4) + T(n/2) + n^2$ :



# Imagine Another Merge Sort

**ImaginaryMergeSort** (A)

if A's size > 1

Divide array A into 1/3s and 2/3s

Call **ImaginaryMergeSort** on first 1/3.

Call **ImaginaryMergeSort** on second 2/3.

Merge two results (combine).

- Recurrence relation for **ImaginaryMergeSort**(A)
  - $T(n) = ??$

# Another MergeSort

MergeSort (  $A, r, s$  )

if (  $r \geq s$  ) return;

$m = r + (s - r) / 3$ ;

$A1 = \text{MergeSort} ( A, r, m )$ ;

$A2 = \text{MergeSort} ( A, m + 1, s )$ ;

Merge (  $A1, A2$  );

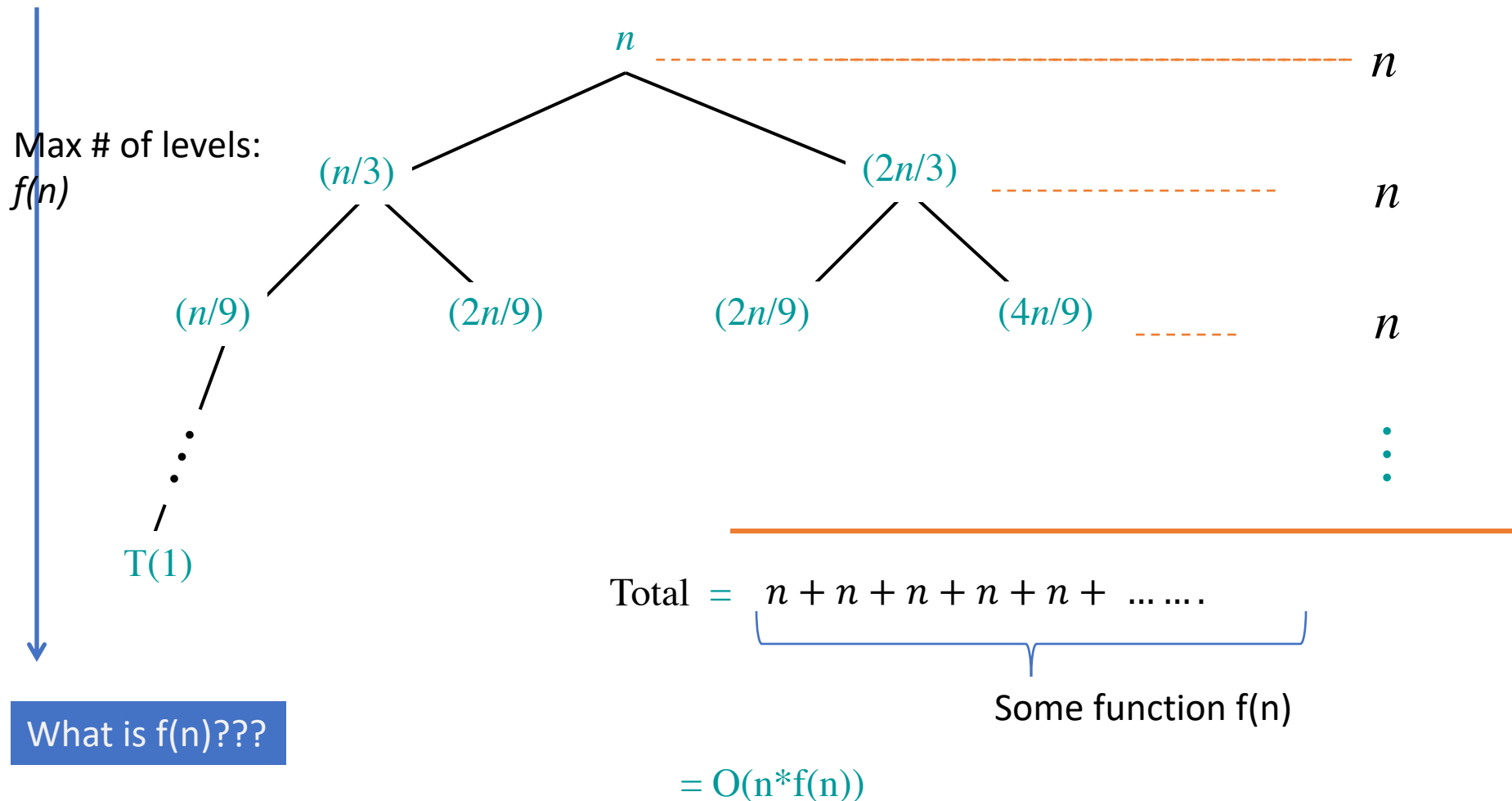
- Recurrence relation for MergeSort(A,1,n)

- $T(n) = T(n/3) + T(2n/3) + cn$

We will ignore the 'c' going forward. It's sloppy but makes things easier to see

# Another Example of Recursion Tree

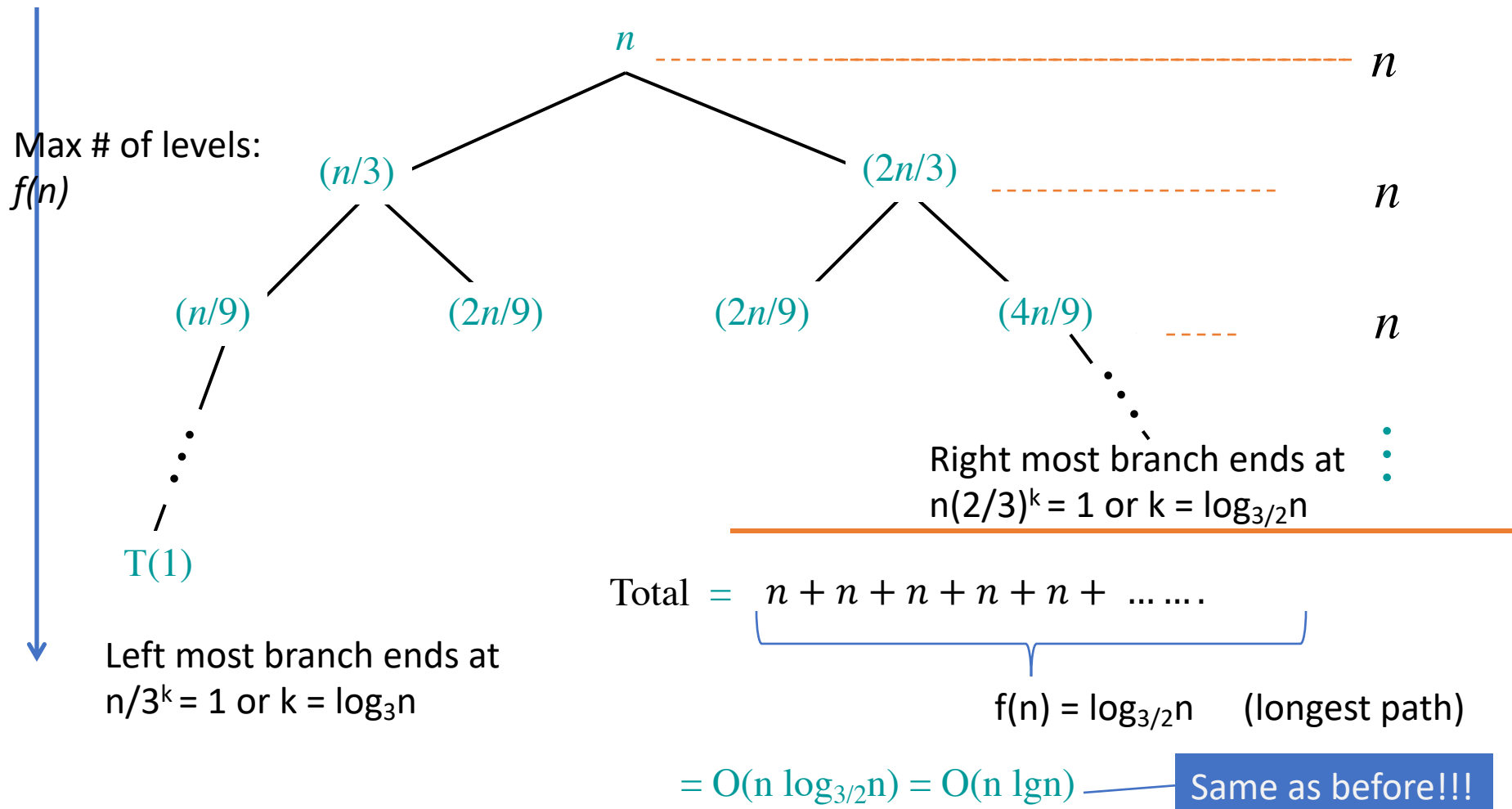
Solve  $T(n) = T(n/3) + T(2n/3) + n$ :





# Another Example of Recursion Tree

Solve  $T(n) = T(n/3) + T(2n/3) + n$ :



# Practice for home

- $T(n) = T(n - 1) + cn$ 
  - $T(n) = O(n^2)$
- $T(n) = T(n - 1) + cn^2$ 
  - $T(n) = O(n^3)$
- $T(n) = T\left(\frac{n}{2}\right) + c$ 
  - $T(n) = O(\lg n)$
- $T(n) = T\left(\frac{n}{3}\right) + cn$ 
  - $T(n) = O(n)$
- $T(n) = 2T\left(\frac{n}{2}\right) + cn^2$ 
  - $T(n) = O(n^2)$
- $T(n) = 4T\left(\frac{n}{2}\right) + cn$ 
  - $T(n) = O(n^2)$
- $T(n) = 4T\left(\frac{n}{2}\right) + cn^2$ 
  - $T(n) = O(n^2 \lg n)$
- **Practice as home through unspooling + drawing recursion trees**

