

Design and Analysis of Algorithms

Merge Sort & Recurrence Relation

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Merge sort

- The **merge sort** algorithm closely follows the **divide-and-conquer paradigm**. Intuitively,
- it operates as follows.
- **Divide:** Divide the n -element sequence to be sorted into two subsequences of $n/2$ elements each.
- **Conquer:** Sort the two subsequences recursively using merge sort.
- **Combine:** Merge the two sorted subsequences to produce the sorted answer.

Merge Sort (Divide and Conquer)

MERGE-SORT $A[1 \dots n]$

1. If $n = 1$, done.
2. Recursively sort $A[1 \dots \lceil n/2 \rceil]$
and $A[\lceil n/2 \rceil + 1 \dots n]$.
3. “*Merge*” the 2 sorted lists.

Key subroutine: **MERGE**

Key Operation

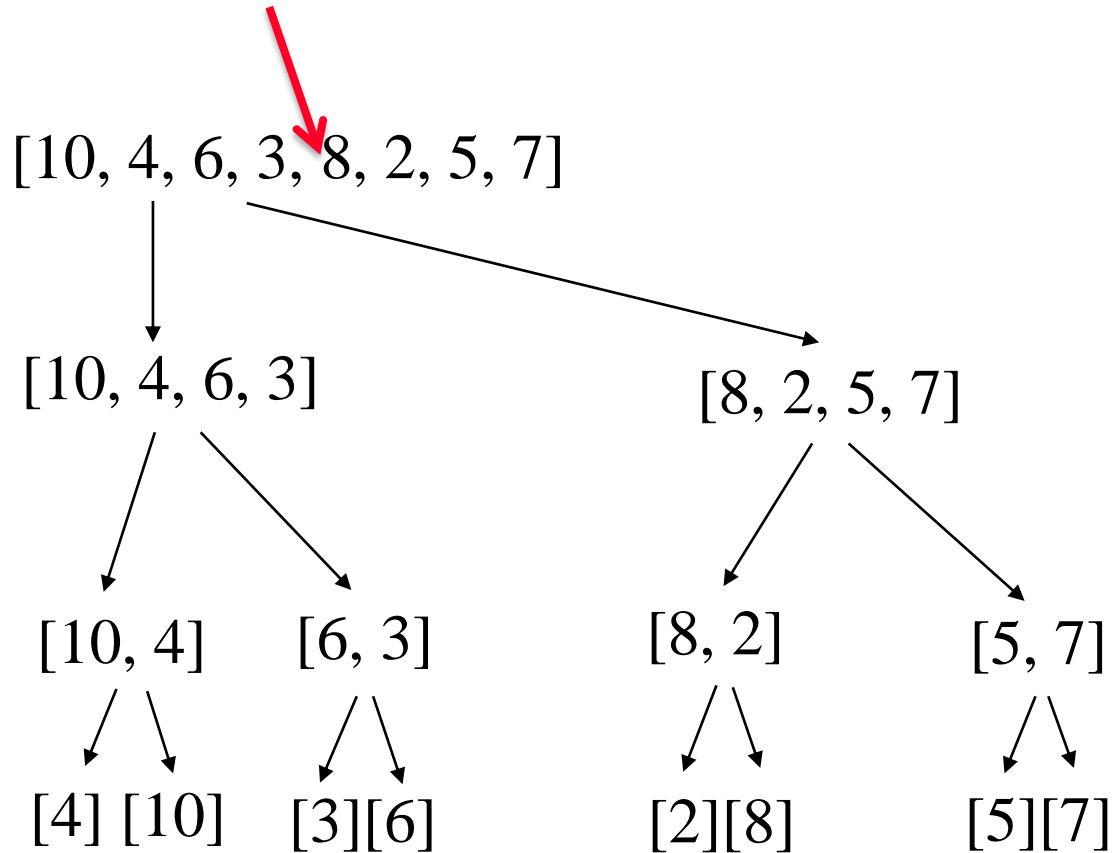
- The key operation of the merge sort algorithm is the merging of two sorted sequences in the “combine” step.
- We merge by calling an auxiliary procedure
- **MERGE** (A, p, q, r), where A is an array and p, q , and r are indices into the array such that $p \leq q < r$.
- The procedure assumes that the subarrays $A[p..q]$ and $A[q+1..r]$ are in sorted order.
- It **merges** them to form a single sorted subarray that replaces the current subarray $A[p..r]$

Visualization MergeSort

- <https://opensa-server.cs.vt.edu/embed/mergesortAV>

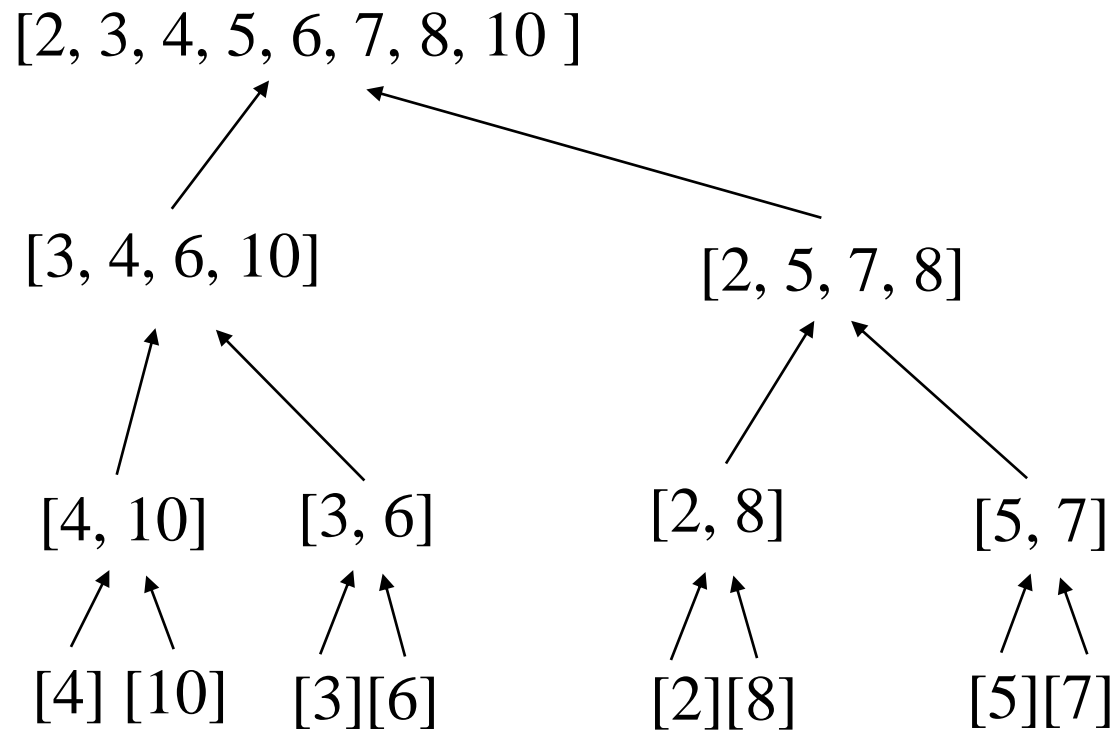
Example

- Partition into lists of size $n/2$



Example Cont'd

- **Merge**



MERGE(A, p, q, r)

```
1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
12 for  $k = p$  to  $r$ 
13     if  $L[i] \leq R[j]$ 
14          $A[k] = L[i]$ 
15          $i = i + 1$ 
16     else  $A[k] = R[j]$ 
17          $j = j + 1$ 
```

Reference: Pg#31
Introduction to algorithms

Merge-Sort

MERGE-SORT(A, p, r)

```
1  if  $p < r$   
2       $q = \lfloor (p + r) / 2 \rfloor$   
3      MERGE-SORT( $A, p, q$ )  
4      MERGE-SORT( $A, q + 1, r$ )  
5      MERGE( $A, p, q, r$ )
```

Analysis of Merge Sort

$$T(n) = \begin{cases} \Theta(1) & \text{if } n \leq c , \\ aT(n/b) + D(n) + C(n) & \text{otherwise .} \end{cases}$$

Divide: The divide step just computes the middle of the subarray, which takes constant time. Thus, $D(n) = \Theta(1)$.

Conquer: We recursively solve two subproblems, each of size $n/2$, which contributes $2T(n/2)$ to the running time.

Combine: We have already noted that the MERGE procedure on an n -element subarray takes time $\Theta(n)$, and so $C(n) = \Theta(n)$.

Worst-case running time $T(n)$ of Merge sort:

$$T(n) = \begin{cases} \Theta(1) & \text{if } n = 1, \\ 2T(n/2) + \Theta(n) & \text{if } n > 1. \end{cases}$$

- Merge sort, with its $\theta(n \lg n)$ running time, outperforms insertion sort, whose running time is $\theta(n^2)$, in the worst case.

Analysis of Merge Sort



Recurrences

- The expression:

$$T(n) = \begin{cases} c & n = 1 \\ 2T\left(\frac{n}{2}\right) + cn & n > 1 \end{cases}$$

is a *recurrence*.

- **Recurrence:** an equation that describes a function in terms of its value on smaller functions

Recurrence Examples

$$s(n) = \begin{cases} 0 & n = 0 \\ c + s(n-1) & n > 0 \end{cases}$$

$$s(n) = \begin{cases} 0 & n = 0 \\ n + s(n-1) & n > 0 \end{cases}$$

$$T(n) = \begin{cases} c & n = 1 \\ 2T\left(\frac{n}{2}\right) + c & n > 1 \end{cases}$$

$$T(n) = \begin{cases} c & n = 1 \\ aT\left(\frac{n}{b}\right) + cn & n > 1 \end{cases}$$

Example

Function(int number)

if number <= 1

then return;

else

Function(number/2)

Function(number/2)

for(i=1 to number)

Display i;

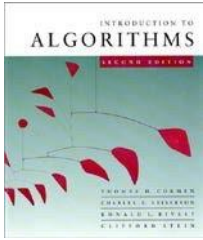
It follows that

$$T(n) = 2 T(n/2) + \textcircled{cn}$$

number of sub-problems

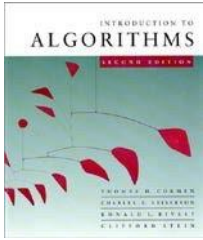
Sub-problem size

work done dividing and combining



Recursion tree

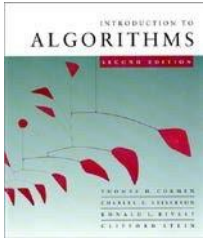
Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.



Recursion tree

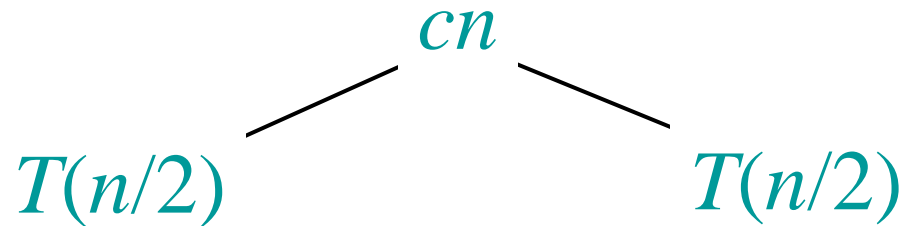
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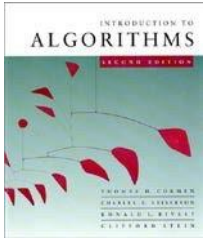
$$T(n)$$



Recursion tree

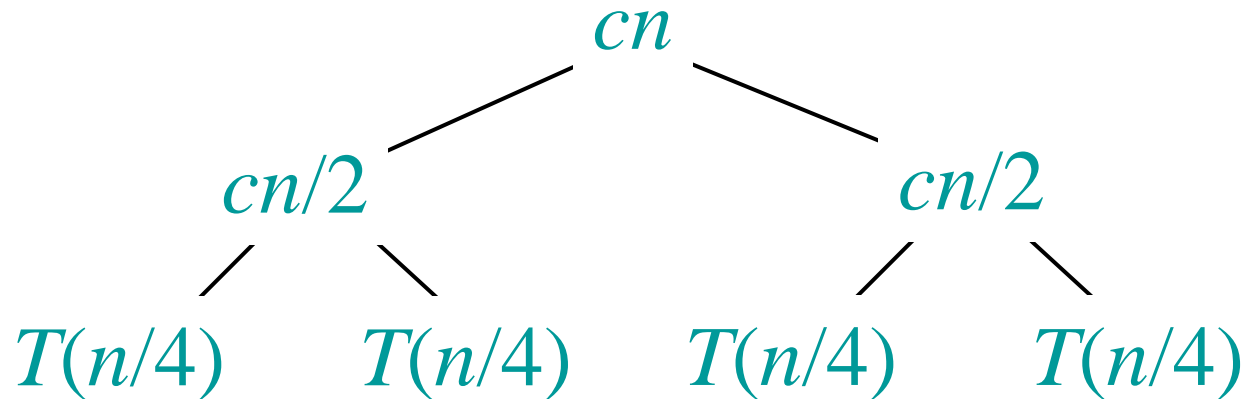
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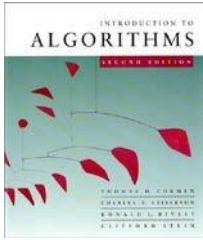




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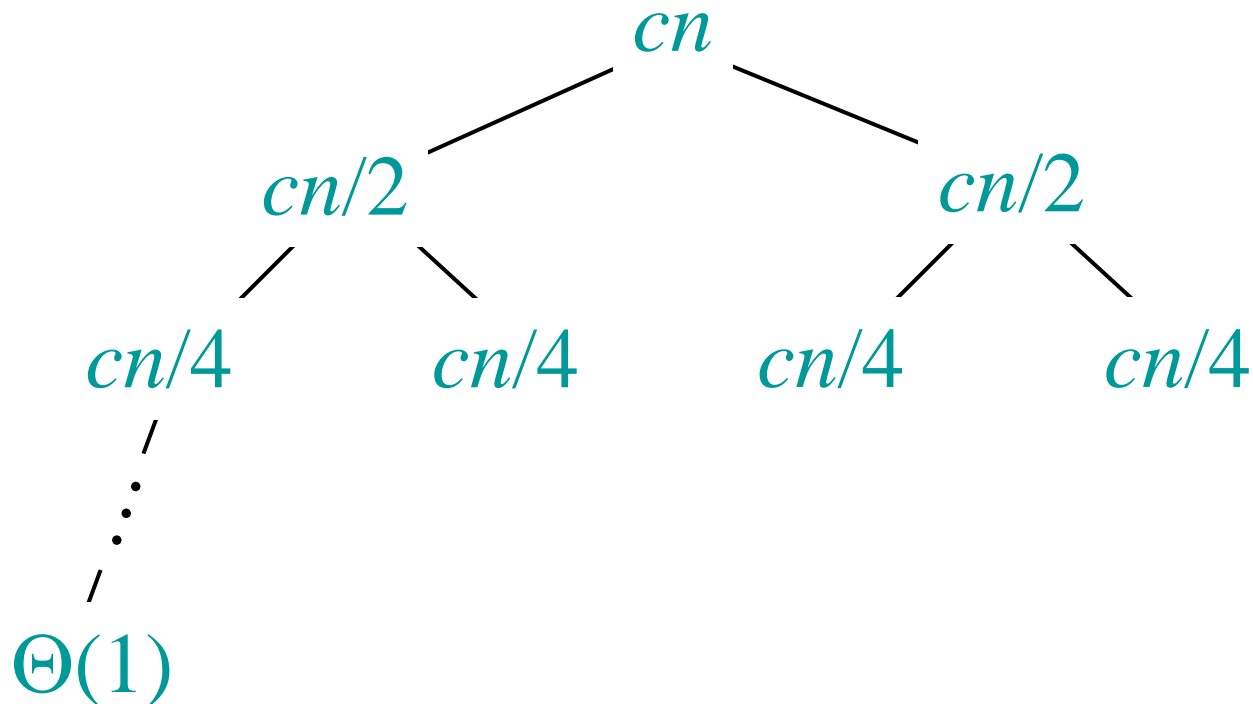
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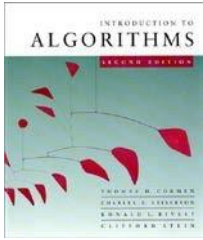




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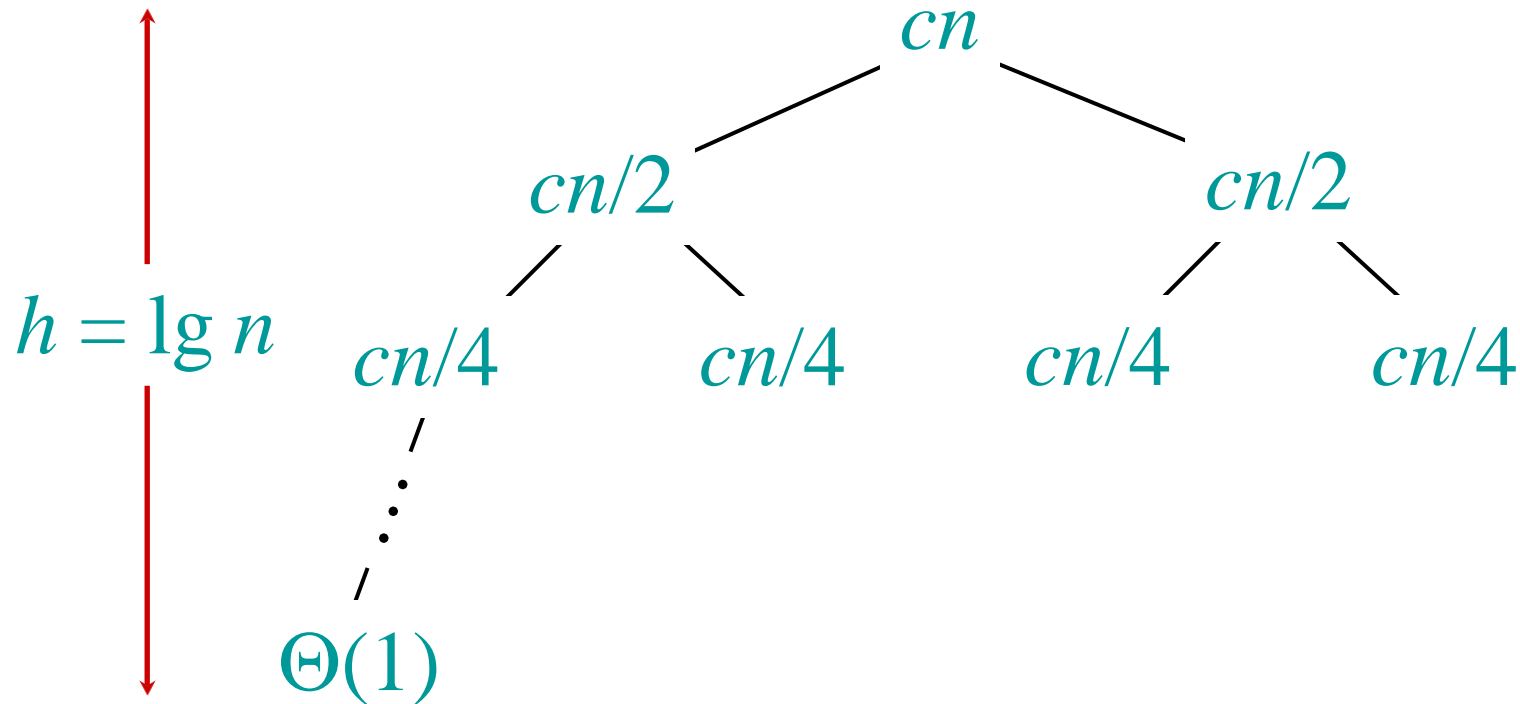
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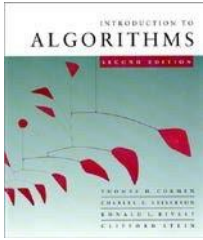




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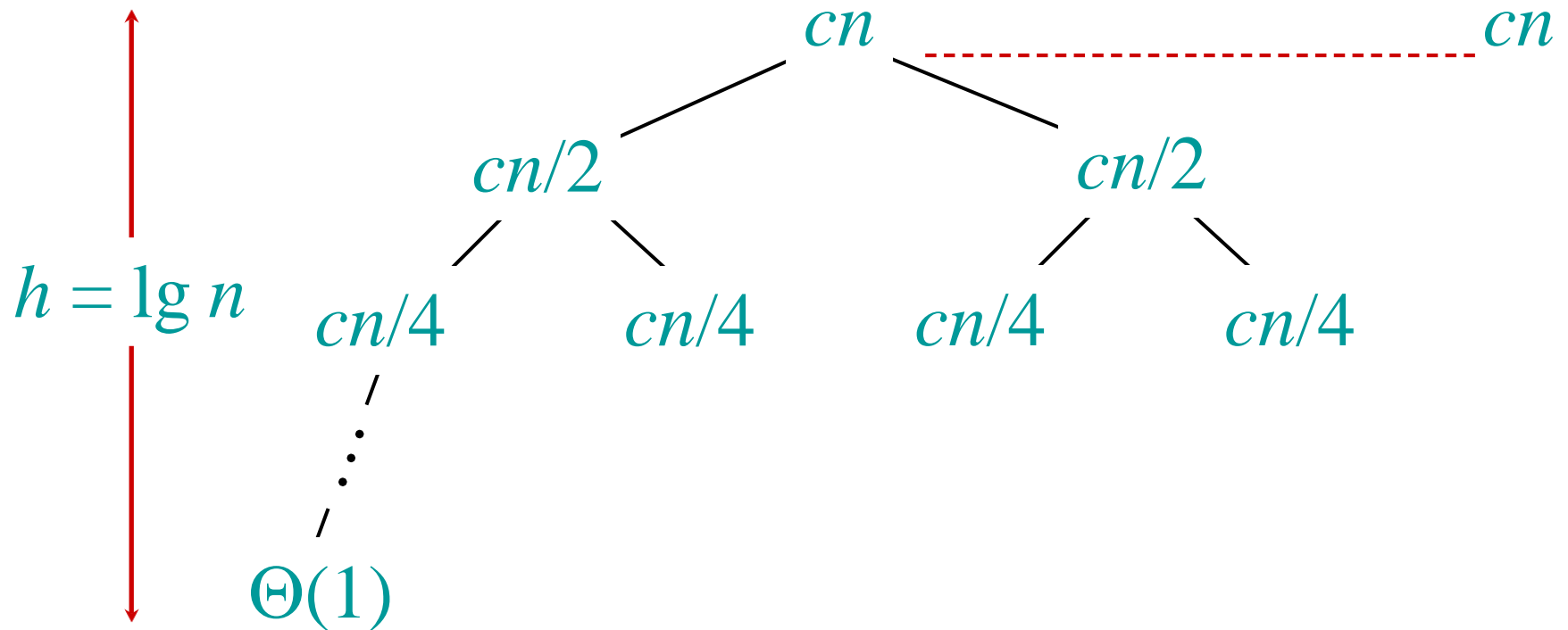
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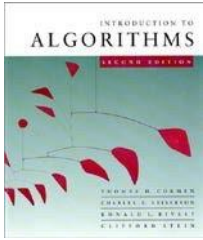




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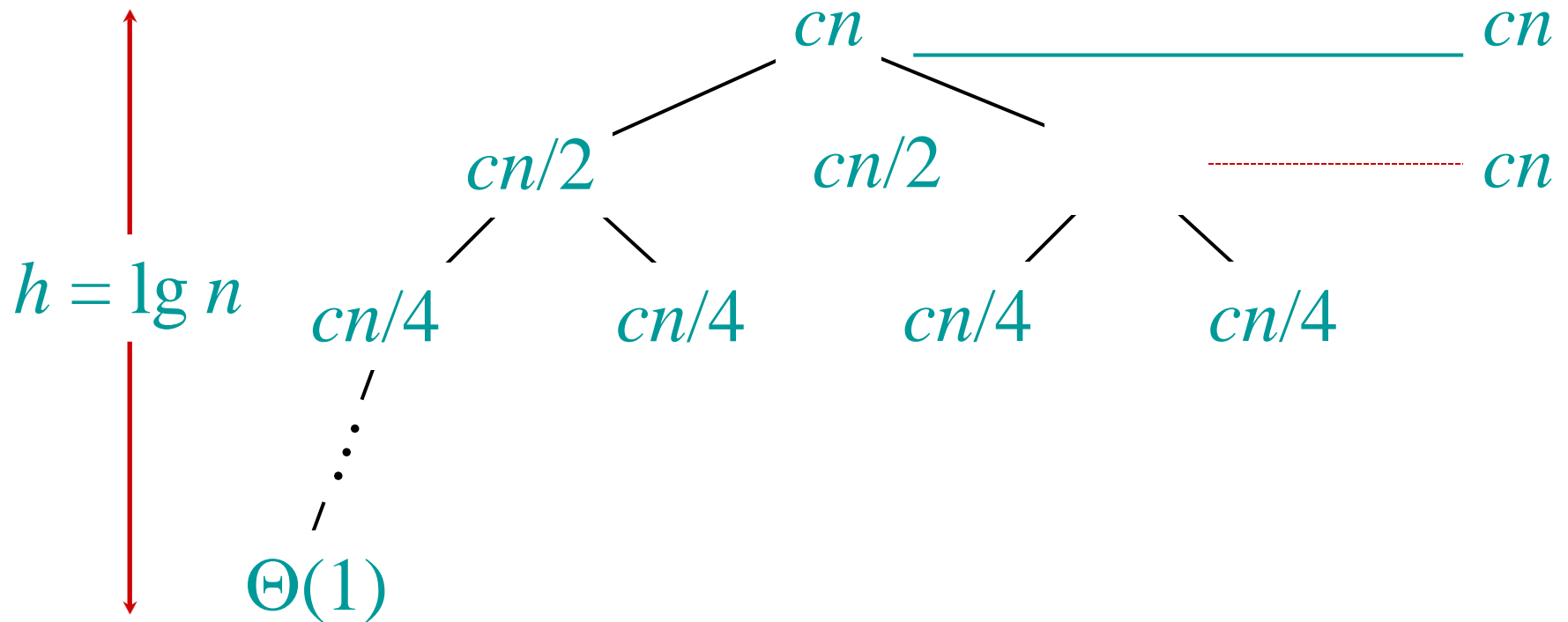
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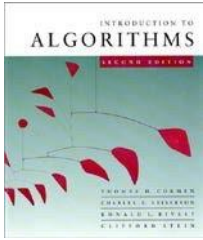




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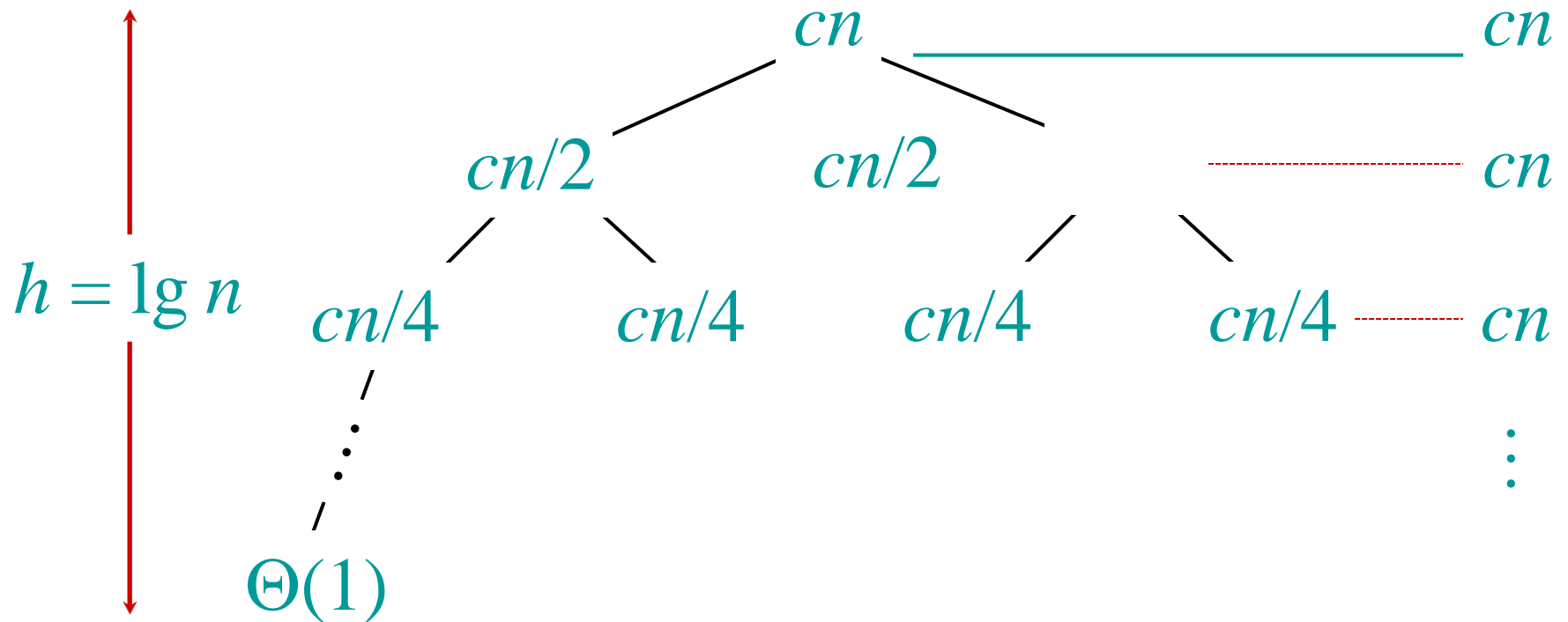
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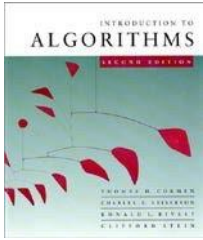




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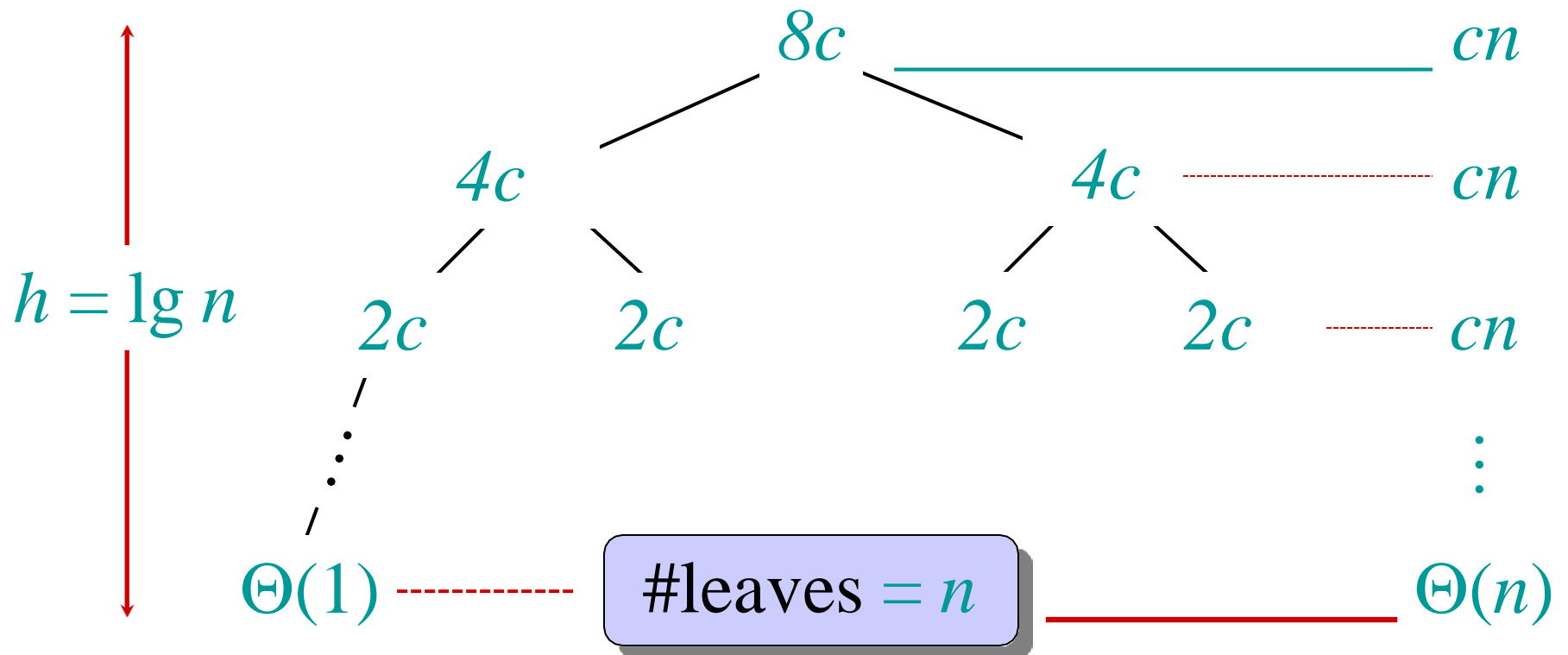


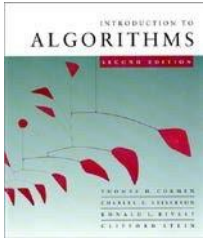


$$n=8$$

Recursion tree

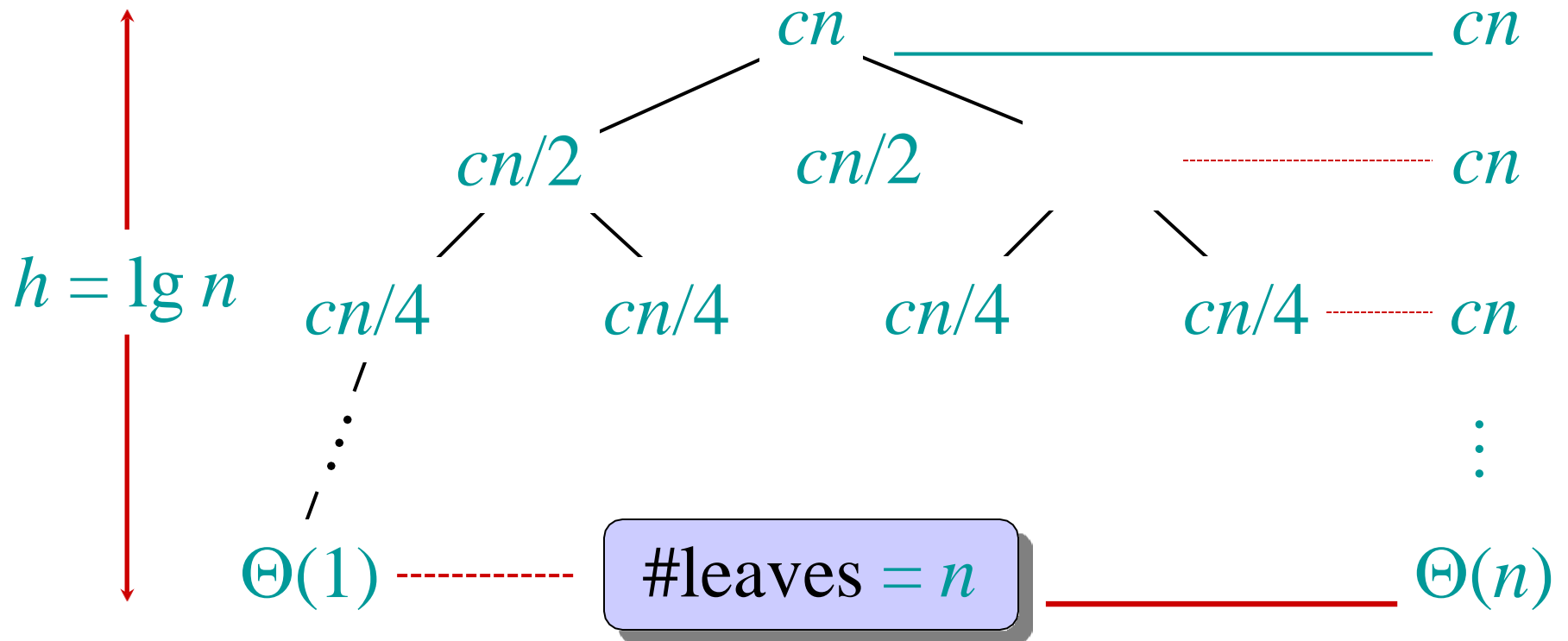
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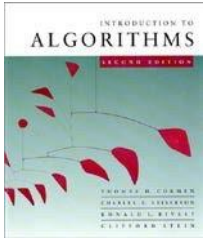




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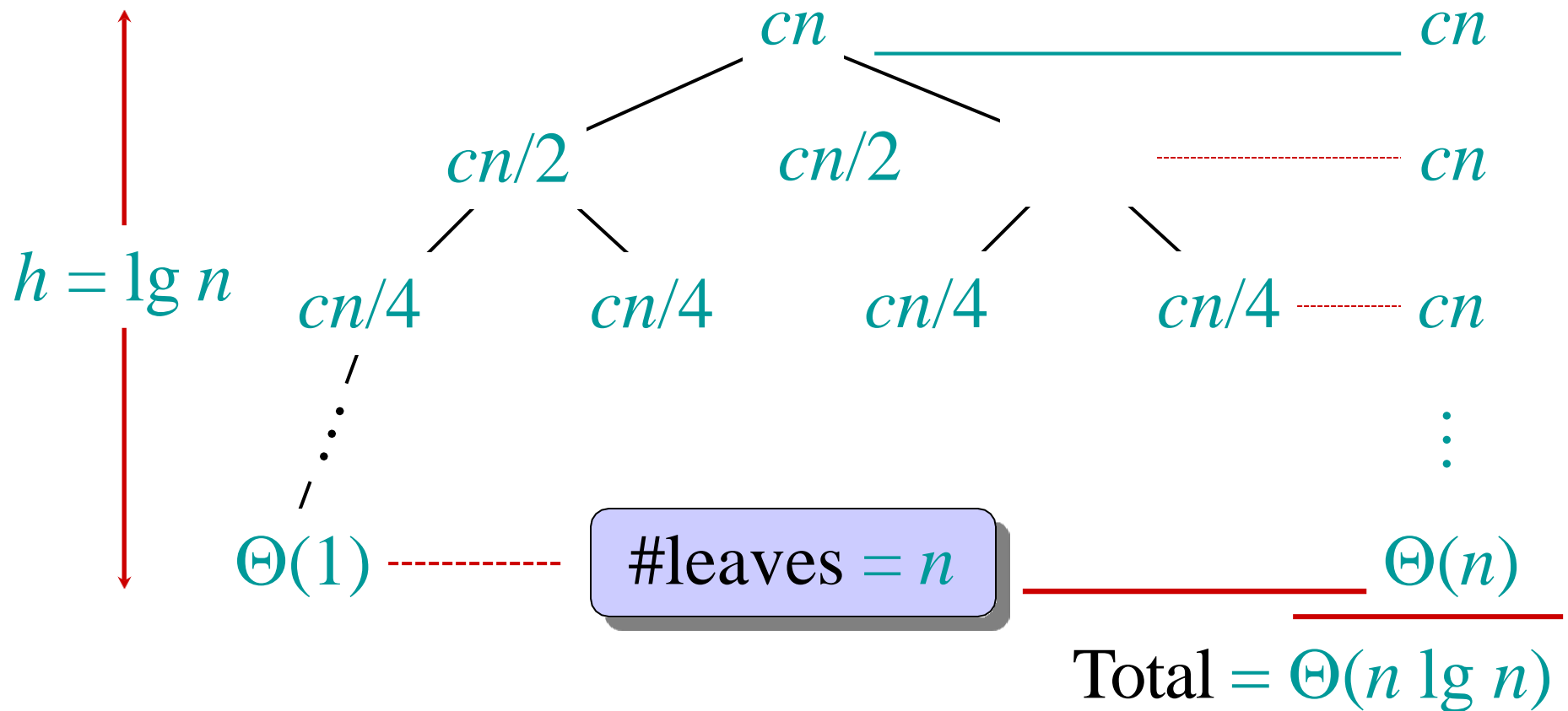
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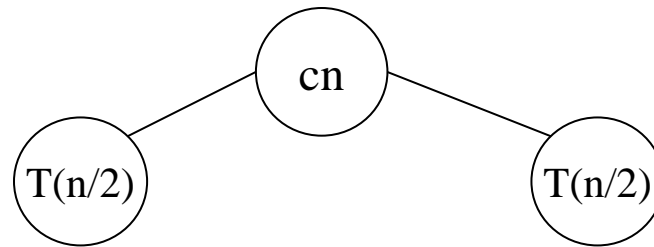


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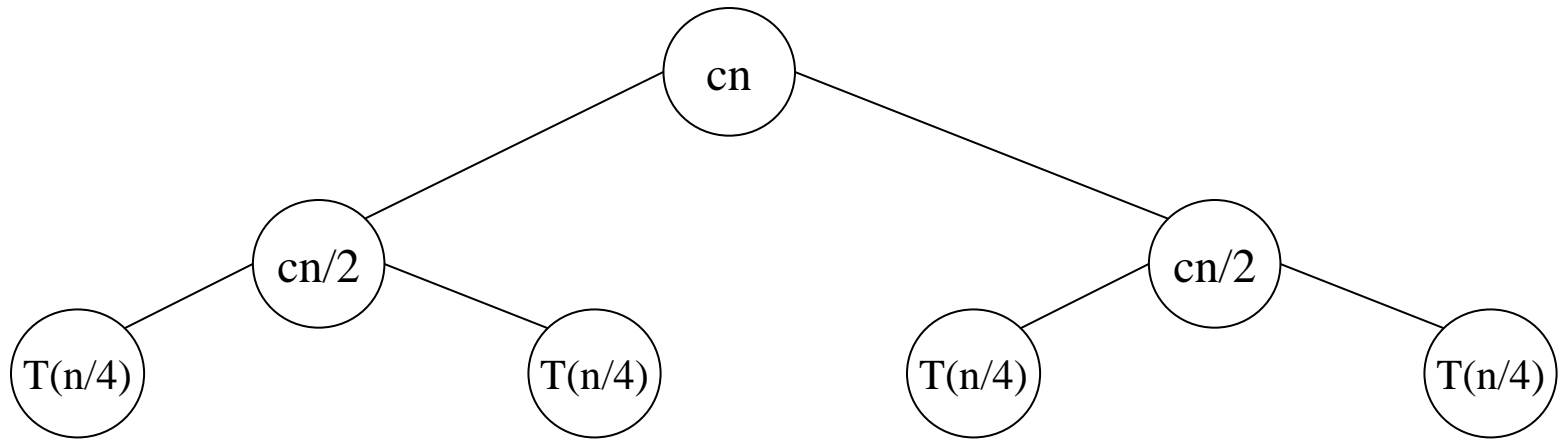
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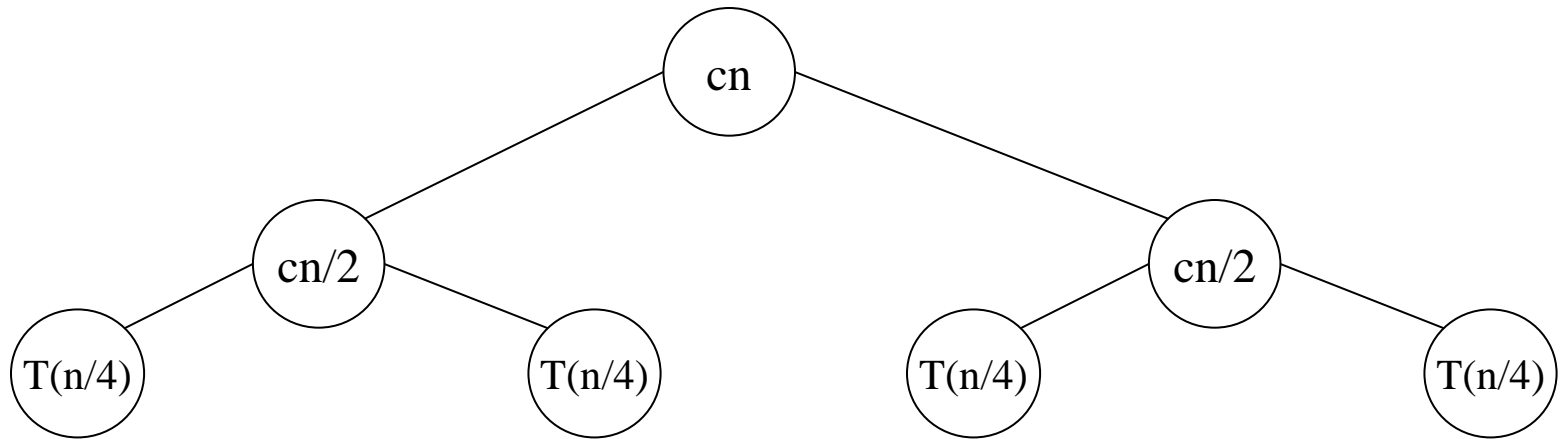
Recursion Tree for Algorithm



Recursion Tree for Algorithm

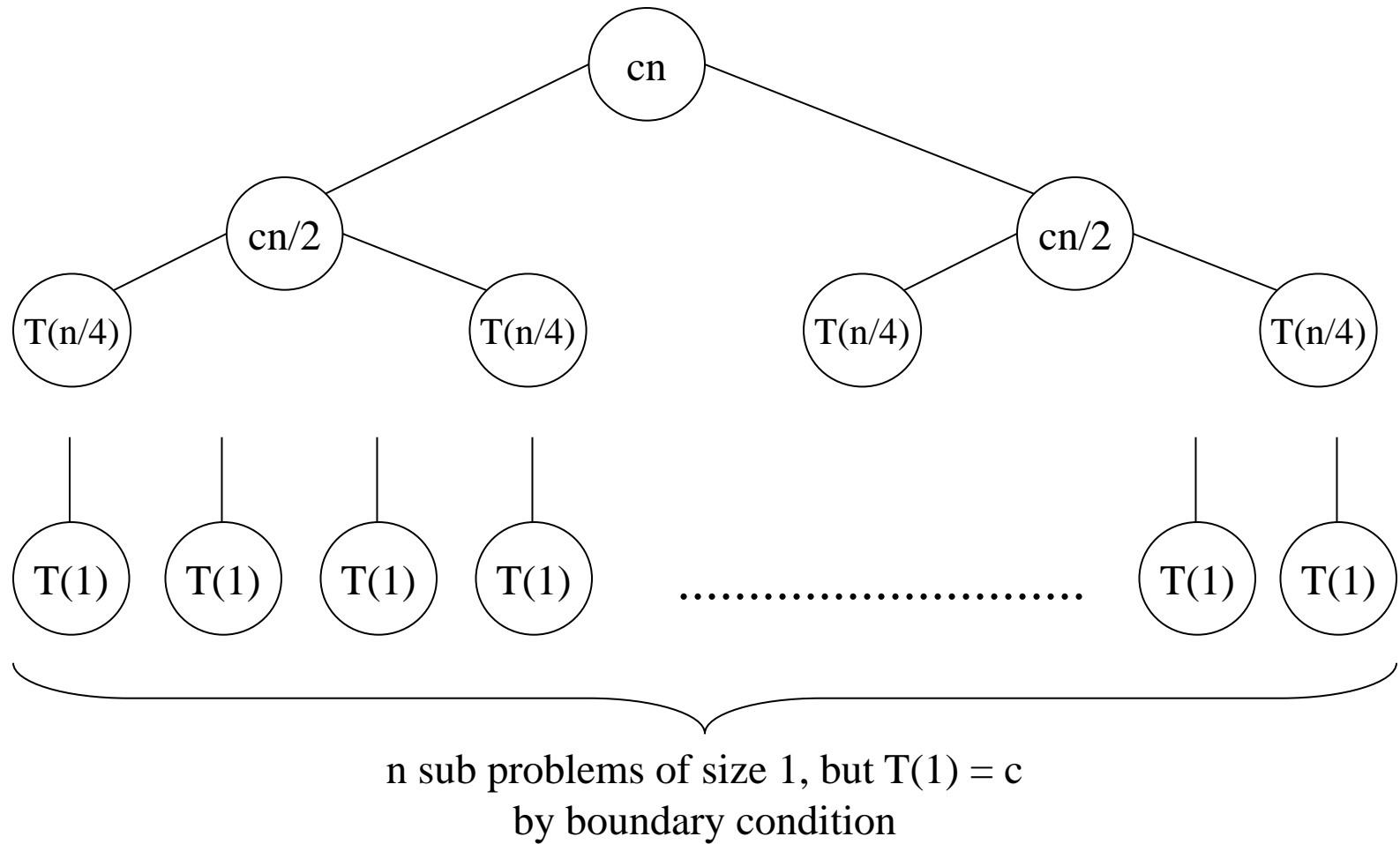


Recursion Tree for Algorithm

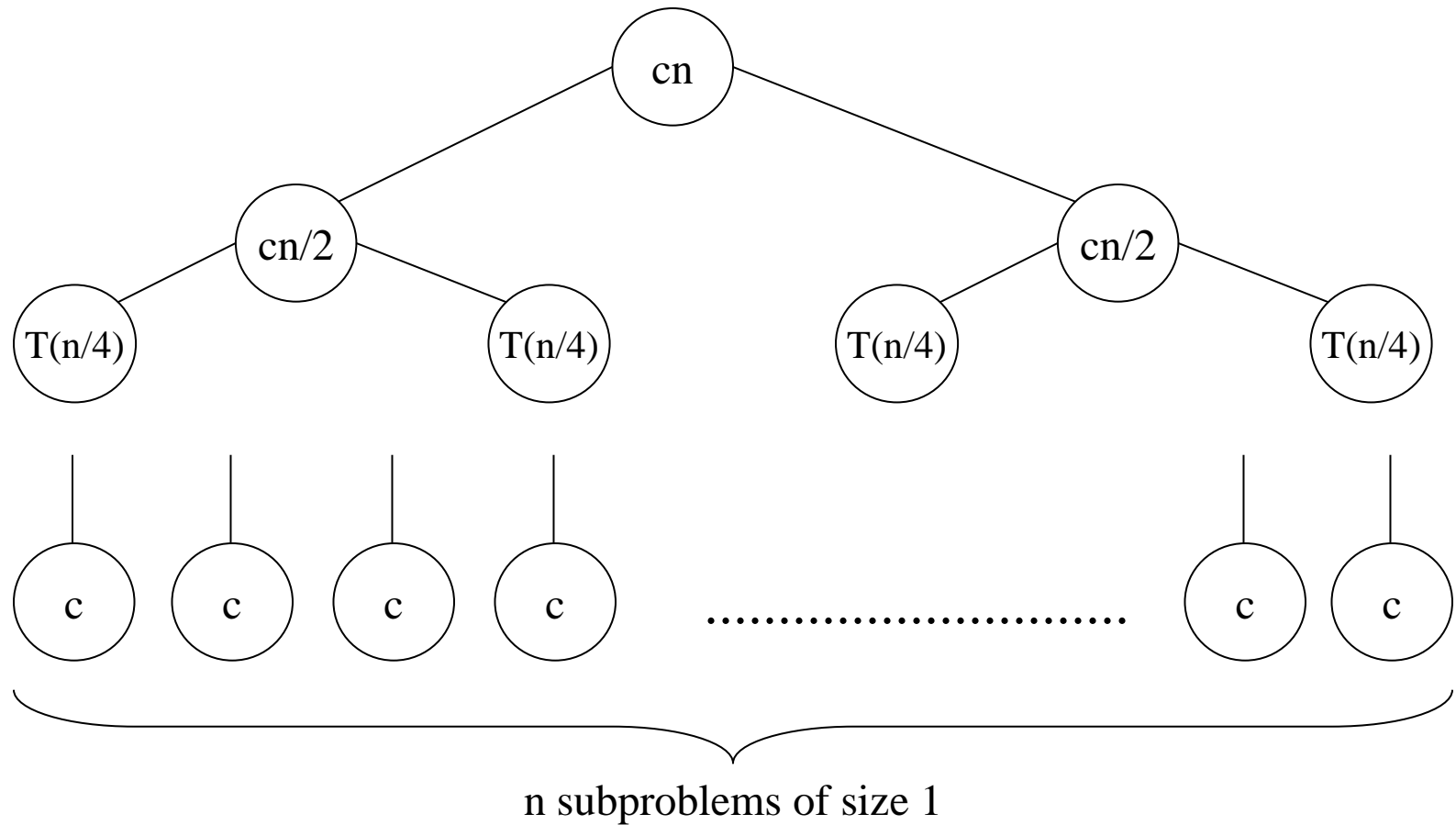


Eventually, the input size (the argument of T) goes to 1, so...

Recursion Tree for Algorithm



Recursion Tree for Algorithm



Recursion Tree for Algorithm

level	nodes/ level	cost/ level
0	$2^0 = 1$	cn
1	$2^1 = 2$	cn
2	$2^2 = 4$	cn

.

.

.

N-1 $2^{N-1} = n$

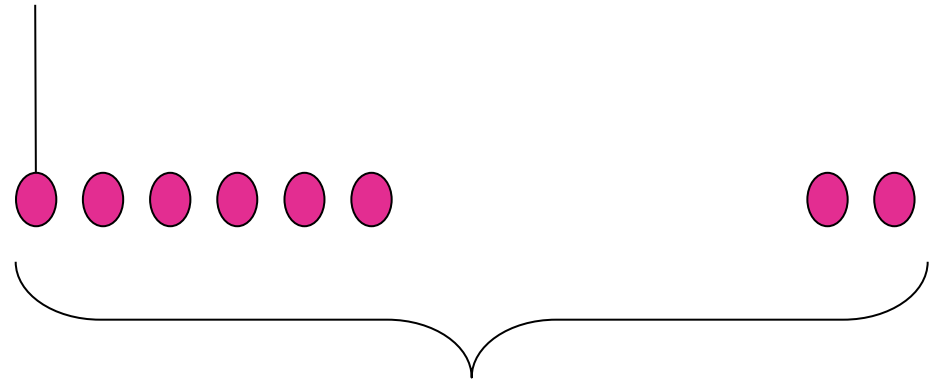
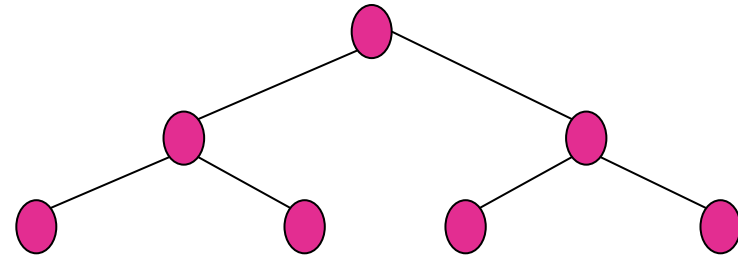
Since $2^{N-1} = n$,

$N-1 = \lg(n)$

levels = N = $1 + \lg(n)$

$T(n) = \text{total cost} = (\text{levels})(\text{cost/level})$

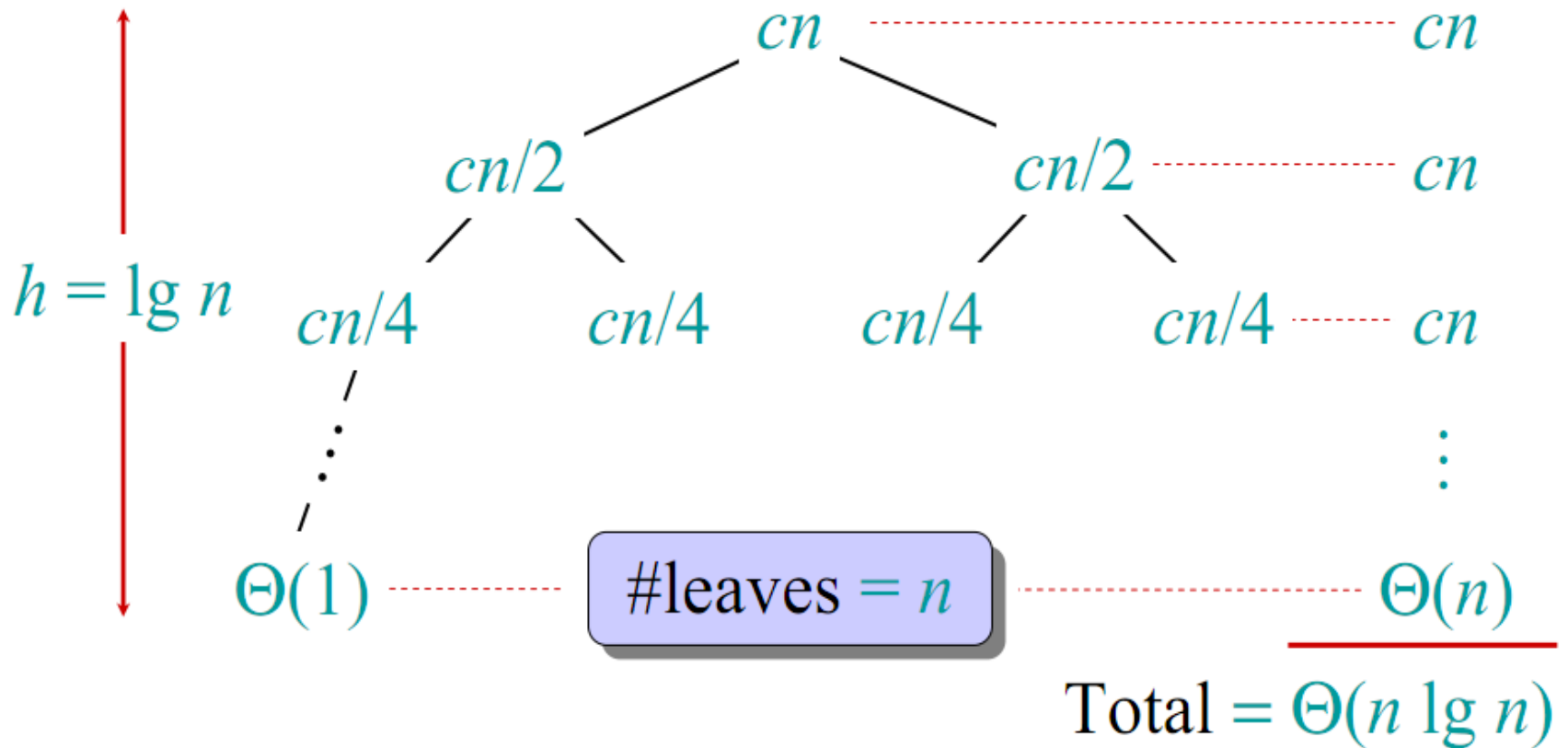
$T(n) = cn [1 + \lg(n)] = O(n \lg(n))$



n nodes at level N-1

Visual Representation of the Recurrence for Merge Sort

Solve $T(n) = 2T(n/2) + cn$, where $c > 0$ is constant.



Time Complexity (Using Master Theorem)

- **Recurrence Relation**

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

Using Master Theorem applying case 2:

$$\Theta\left(n^{\log_b a} \log n\right)$$

So time complexity is $O(n \log n)$

- $\Theta(n \lg n)$ grows more slowly than $\Theta(n^2)$
- Therefore, **merge sort asymptotically beats insertion sort in the worst case.**
- In practice, merge sort beats insertion sort for $n \geq 3$

Sorting algorithms

- **Selection and bubble sort have quadratic best/average/worst-case performance**
- **Insertion sort has quadratic average-case and worst-case performance**
- **The faster comparison based algorithm ?
 $O(n \log n)$**
- **Mergesort and Quicksort**

Solving Recurrences

- **The substitution method**
 - A.k.a. “making a good guess method”
 - Guess the form of the answer, then use mathematical induction to find the constants and show that the solution works
 - Example:
 - $T(n) = 2T(n/2) + n \quad \square \quad T(n) = O(n \lg n)$

- **<https://opensa-server.cs.vt.edu/embed/mergesortAV>**
- **https://www.youtube.com/watch?v=4V30R3l1vLI&ab_channel=AbdulBari**