

CSE 373: Data Structures and Algorithms

# Sorting and recurrence analysis techniques

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# Sorting problem statement

Given  $n$  comparable elements, rearrange them in an increasing order.

## Input

- An array  $A$  that contains  $n$  elements
- Each element has a key  $k$  and an associated data
- Keys support a comparison function (e.g., keys implement a Comparable interface)

## Expected output

- An output array  $A$  such that for any  $i$  and  $j$ ,
- $A[i] \leq A[j]$  if  $i < j$  (increasing order)
- Array  $A$  can also have elements in reverse order (decreasing order)

# Desired properties in a sorting algorithm

→  $(1, \underline{a})$   $(3, b)$   $(1, \underline{d})$   $(4, x)$

## Stable

- In the output, equal elements (i.e., elements with equal keys) appear in their original order

## In-place

- Algorithm uses a constant additional space,  $O(1)$  extra space

## Adaptive

- Performs better when input is almost sorted or nearly sorted
- (Likely different big-O for best-case and worst-case)

## Fast. $O(n \log n)$

$(1, a)$   $(1, \underline{d})$   $(3, b)$   $(4, x)$

$(1, \underline{d})$   $(1, \underline{a})$   $(3, b)$   $(4, x)$

No algorithm has all of these properties. So choice of algorithm depends on the situation.

# Sorting algorithms – High-level view

- $O(n^2)$ 
  - Insertion sort
  - Selection sort
  - Quick sort (worst)
- $O(n \log n)$ 
  - Merge sort
  - Heap sort
  - Quick sort (avg)
- $\Omega(n \log n)$  -- lower bound on comparison sorts
- $O(n)$  – non-comparison sorts
  - Bucket sort (avg)

# A framework to think about sorting algos

Some questions to consider when analyzing a sorting algorithm.

Loop/step invariant: \_\_\_\_\_ What is the state of the data during each step while sorting

Runtime: Worst \_\_\_\_\_ Average \_\_\_\_\_ Best \_\_\_\_\_

Input: Worst \_\_\_\_\_ Best \_\_\_\_\_

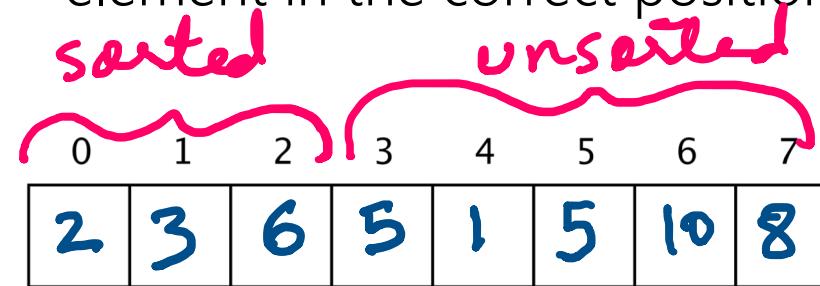
Stable Yes/No/Can-be In-place Yes/No/Can-be Adaptive Yes/No

Operations: Comparisons > or < or approx. equal Moves

Data structure \_\_\_\_\_ Which data structure is better suited for this algo

# Insertion sort

Idea: At step  $i$ , insert the  $i^{th}$  element in the correct position among the first  $i$  elements.



Loop/step invariant: Subarray  $0 \dots i-1$  is sorted

Runtime: Worst  $O(n^2)$  Average  $O(n^2)$  Best  $O(n)$

Input: Worst Reverse sorted Best Sorted ↗

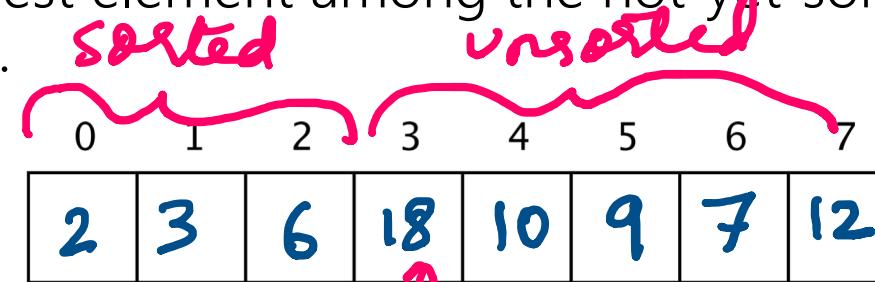
Stable Yes (if careful) In-place Yes Adaptive Yes

Operations: Comparisons depends on case Moves

Data structure linked list (ideal to insert elements in a list)

# Selection sort

Idea: At step  $i$ , find the smallest element among the not-yet-sorted elements ( $i \dots n$ ) and swap it with the element at  $i$ .



*Subarray  $0 \dots i-1$  is sorted and contains  $i$  smallest element in Array*

Loop/step invariant:  $0 \dots i-1$  is sorted and contains  $i$  smallest element in Array

Runtime: Worst  $O(n^2)$

Average  $O(n^2)$

Best  $O(n^2)$

Input: Worst \_\_\_\_\_ Best \_\_\_\_\_

Stable Yes (if careful) In-place Yes Adaptive No

Operations: Comparisons      $\Rightarrow$      Moves see

Data structure Array. To get min, could we keep  $\rightarrow$  heap sort

# Heap sort

Idea: *buildHeap* with all  $n$  elements

```
for i = 0 to n do  
    A[i] = removeMin()  
end for
```

Loop/step invariant: \_\_\_\_\_

Runtime: Worst  $O(n \log n)$  Average \_\_\_\_\_ Best \_\_\_\_\_

Input: Worst \_\_\_\_\_ Best \_\_\_\_\_

Stable No In-place No (But can be) Adaptive \_\_\_\_\_

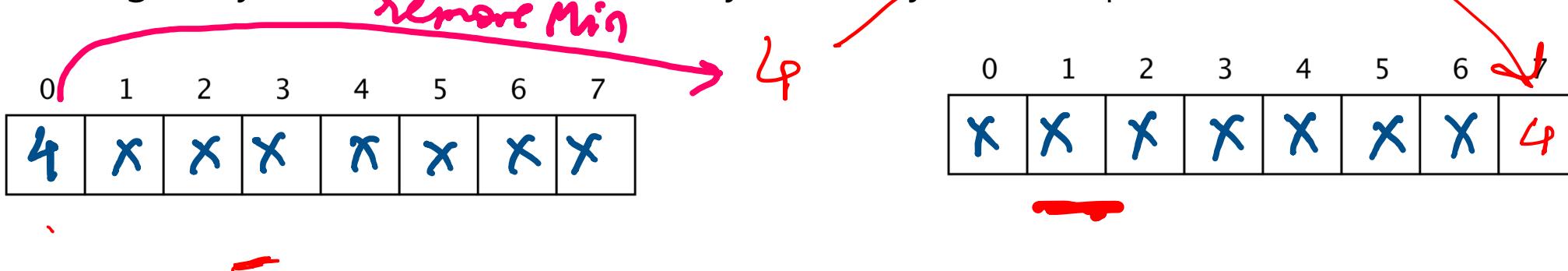
Operations: Comparisons \_\_\_\_\_ Moves \_\_\_\_\_

Data structure Heap

# In-place heap sort

Idea:

1. Treat initial array as a heap
2. When you call `removeMin()`, that frees up a slot towards the end in the array. Put the ~~extract min~~ element there.
3. More specifically, when you remove the  $i^{th}$  element, put it at  $A[n - i]$
4. This gives you a reverse sorted array. But easy to fix in-place.



# Design technique: Divide-and-conquer

Very important technique in algorithm to attack problems

Three steps:

1. Divide: Split the original problem into smaller parts
2. Conquer: Solve individual parts independently (think recursion)
3. Combine: Put together individual solved parts to produce an overall solution

Merge sort and Quick sort are classic examples of sorting algorithms that use this technique

# Merge sort

To sort a given array,

Divide: Split the input array into two halves

Conquer: Sort each half independently

Combine: Merge the two sorted halves into one sorted whole (HW3 Problem 6!)

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```
function mergeSort(A)
    if A.length == 1 then
        return A;
    else
        mid = A.length / 2
        firstHalf = mergeSort(new [0, ... mid])
        secondHalf = mergeSort(new [mid+1, ... ])
        return merge(firstHalf, secondHalf)
    end if
end function
```

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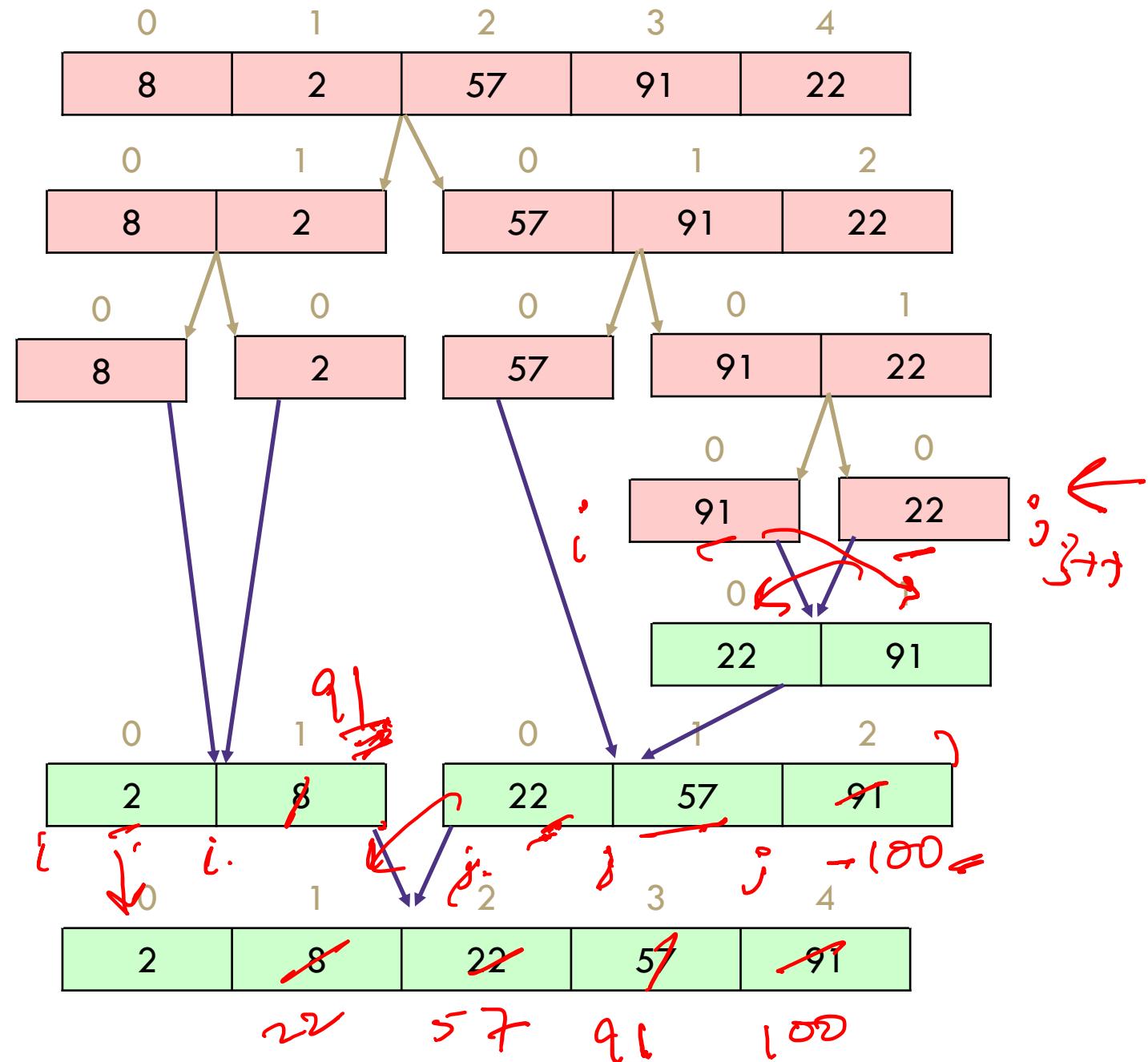
# Merge sort

Split array in the middle

Sort the two halves

Merge them together

$$T(n) = \begin{cases} c_1 & \text{if } n \leq 1 \\ 2T\left(\frac{n}{2}\right) + c_2 n & \text{otherwise} \end{cases}$$



# Review: Unfolding (technique 1)

$$\begin{aligned} T(n) &= 2\underline{T(n/2)} + c_2 n \\ &= 2 \left( 2T\left(\frac{n}{2 \cdot 2}\right) + c_2 \frac{n}{2} \right) + c_2 n \\ &= 2^2 T\left(\frac{n}{2 \cdot 2}\right) + c_2 n + c_2 n \\ &= 2^2 \left( 2T\left(\frac{n}{2^3}\right) + c_2 \frac{n}{2^2} \right) + c_2 n + c_2 n \\ &= 2^3 T\left(\frac{n}{2^3}\right) + c_2 n + c_2 n + c_2 n \quad i \\ &= \dots \\ &= 2^{\log n} T(1) + \underbrace{c_2 n + c_2 n + \dots + c_2 n}_{\text{about } \log(n) \text{ times}} \\ &= c_1 n + c_2 n \log n \end{aligned}$$

$$T(n) = \begin{cases} c_1 & \text{if } n \leq 1 \\ 2T\left(\frac{n}{2}\right) + c_2 n & \text{otherwise} \end{cases}$$

# Technique 2: Tree method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$$

Level	Number of Nodes at level	Work per Node	Work per Level
0			
1			
2			
$i$			
base			

Last recursive level:

# Technique 2: Tree method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$$

$n$

Level	Number of Nodes at level	Work per Node	Work per Level
0			
1			
2			
$i$			
base			

Last recursive level:

# Technique 2: Tree method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$$

$T(n)$  is circled in red.

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$n$  is at the root node.

$\frac{n}{2}$  is at the left child node.

$\frac{n}{2}$  is at the right child node.

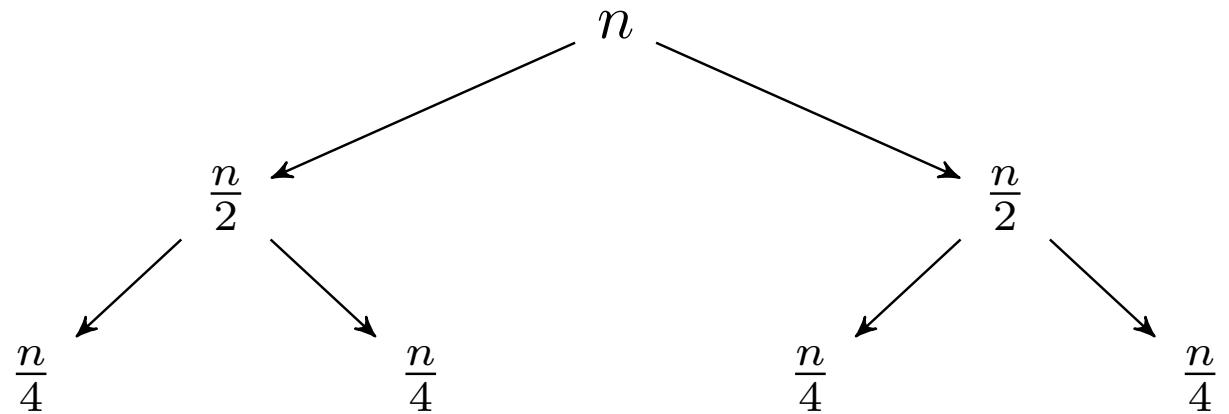
$T(n/2) + T(n/2)$  is written above the tree structure.

Level	Number of Nodes at level	Work per Node	Work per Level
0			
1			
2			
$i$			
base			

Last recursive level:

# Technique 2: Tree method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$$

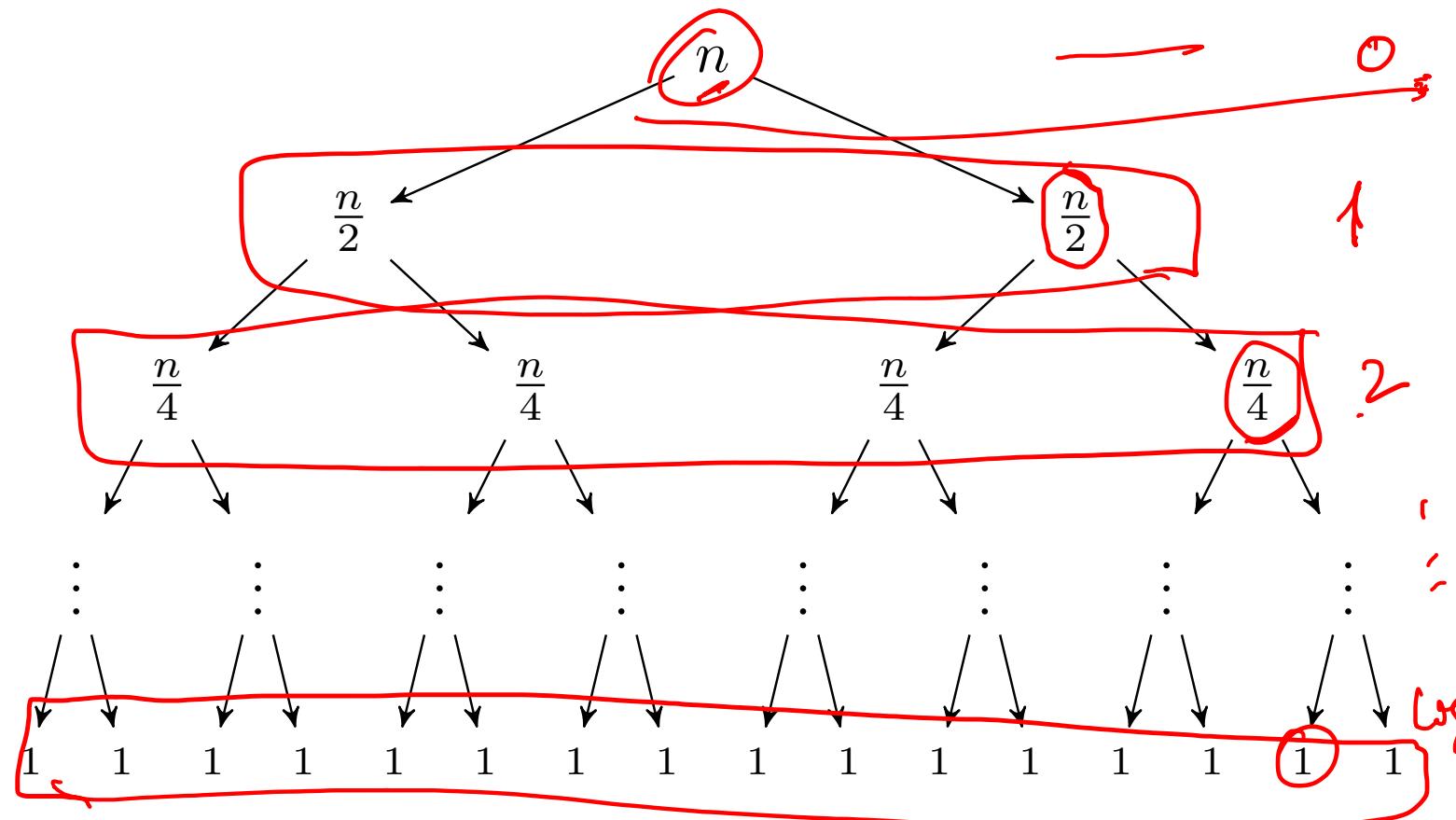


Level	Number of Nodes at level	Work per Node	Work per Level
0			
1			
2			
$i$			
base			

Last recursive level:

# Technique 2: Tree method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$$



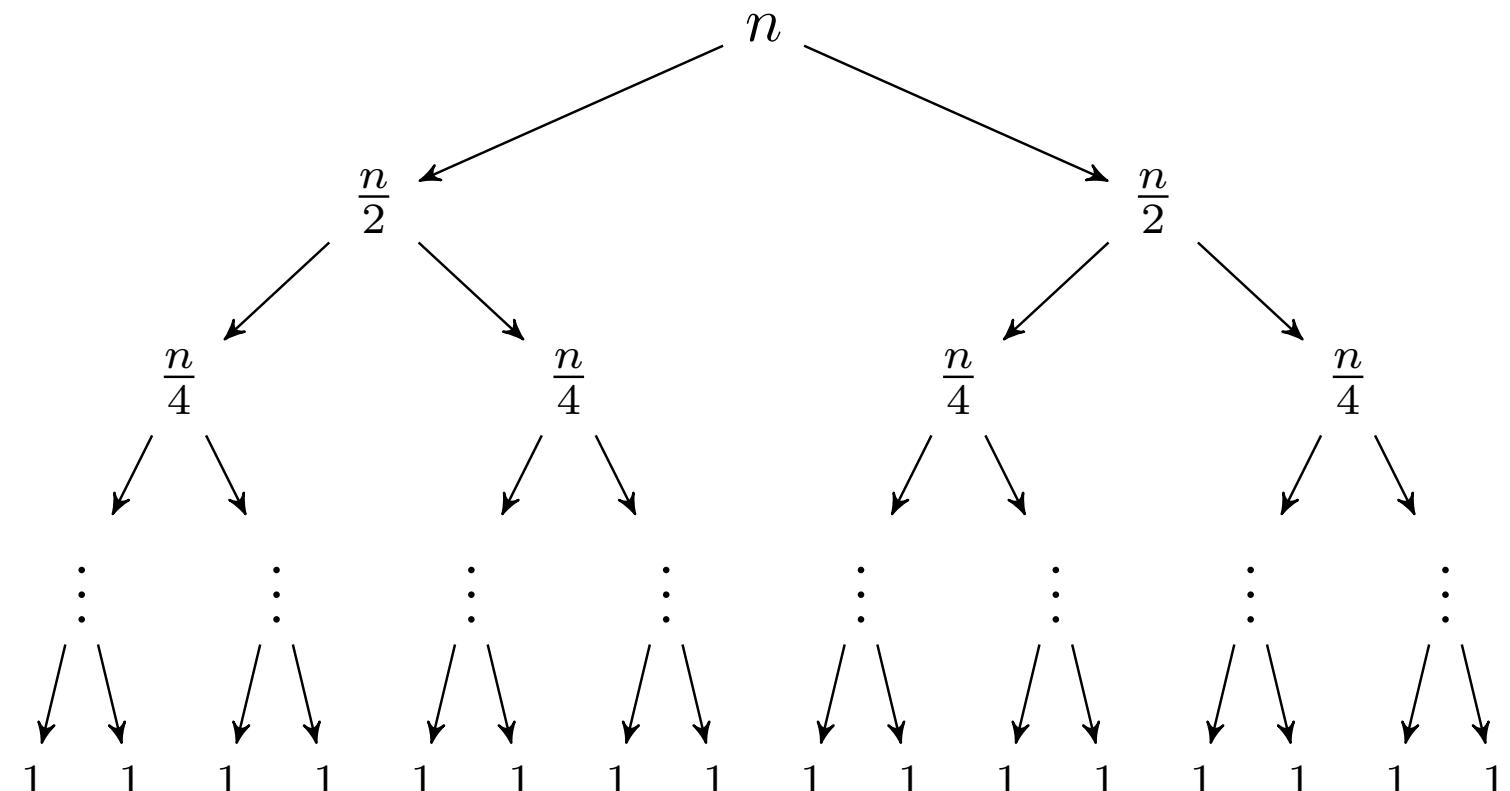
Level	Number of Nodes at level	Work per Node	Work per Level
0	$\frac{n}{2^0}$	$\frac{n}{2^0}$	$n$
1	$\frac{n}{2^1}$	$\frac{n}{2^1}$	$n$
2	$\frac{n}{2^2}$	$\frac{n}{2^2}$	$n$
$i$	$\frac{n}{2^i}$	$\frac{n}{2^i}$	$n$
base	$n$	1	$n$

Last recursive level:  $\frac{\log n - 1}{\log n - 1}$

$$\begin{aligned} T(n) &= \text{Base case} + \sum_{i=0}^{\log n - 1} n \\ \log n &= n + \sum_{i=0}^{\log n - 1} n \\ &= n + n \log n \end{aligned}$$

# Technique 2: Tree method

$$T(n) = \begin{cases} 1 & \text{if } n = 1 \\ 2T(n/2) + n & \text{otherwise} \end{cases}$$



Level	Number of Nodes at level	Work per Node	Work per Level
0	1	$n$	$n$
1	2	$\frac{n}{2}$	$n$
2	4	$\frac{n}{4}$	$n$
$i$	$2^i$	$\frac{n}{2^i}$	$n$
base	$2^{\log_2 n}$	1	$n$

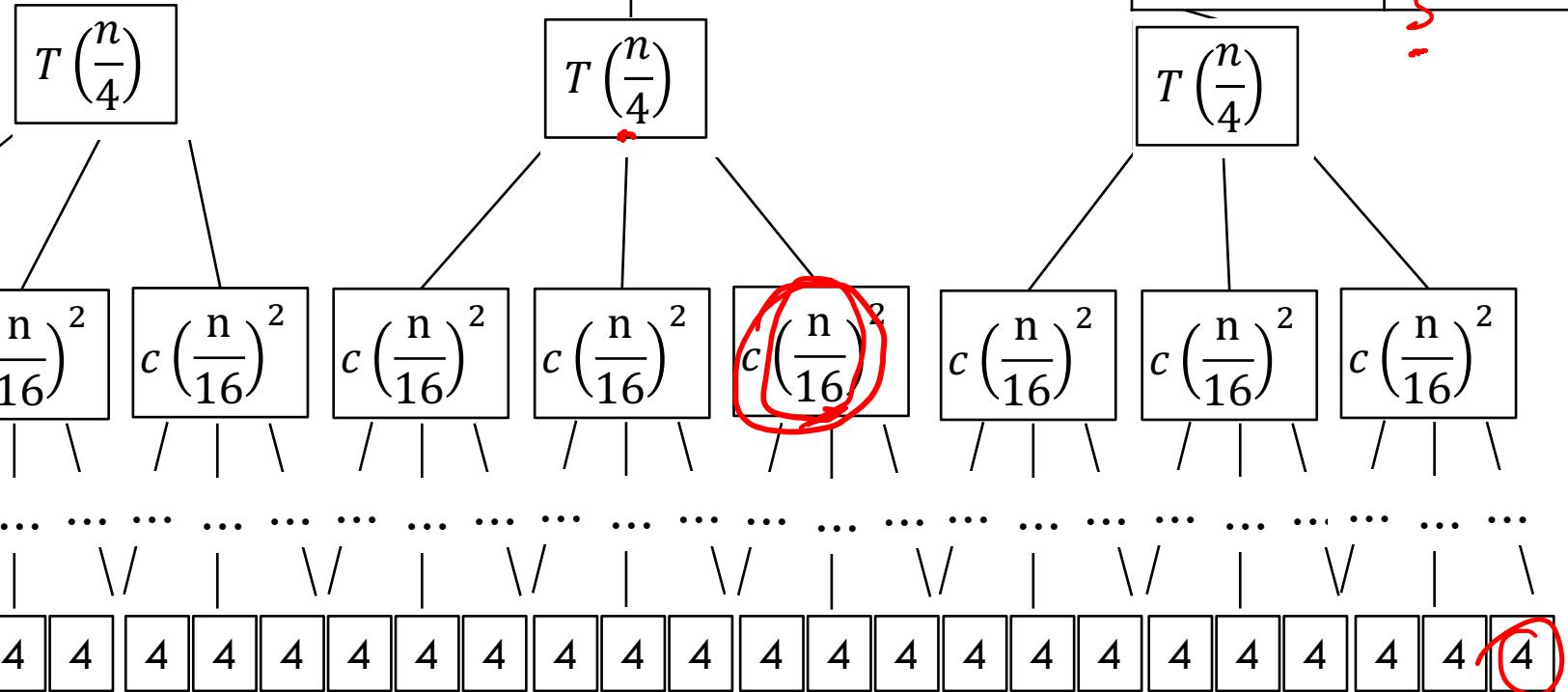
Last recursive level:  $\log n - 1$

Combining it all together...

$$T(n) = n + \sum_{i=0}^{\log_2 n - 1} n$$

# Tree Method Practice

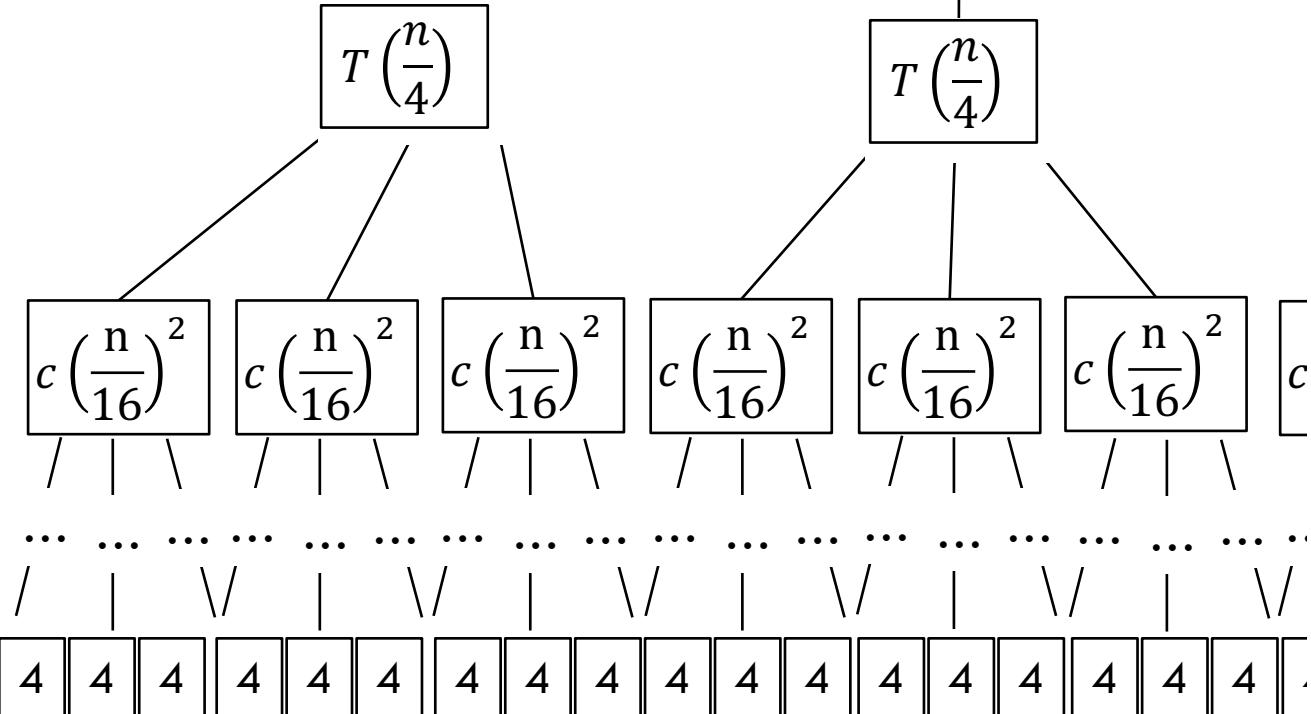
$$T\left(\frac{n}{4}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{4}\right) + cn^2$$



Level (i)	Number of Nodes	Work per Node	Work per Level
0	1	$\cancel{X} cn^2$	$cn^2$
1	3	$\cancel{X} c(n/4)^2$	
2	9	$\cancel{X} c(n/16)^2$	
i	$3^i$	$\cancel{X} c(n/4^i)^2$	$c(3n/16)^i$
base	$3^{log_4 n}$	$\cancel{X} 4$	$4 \cdot 3^{log_4 n}$

# Tree Method Practice

$$T\left(\frac{n}{4}\right) + T\left(\frac{n}{4}\right) + T\left(\frac{n}{4}\right) + cn^2$$



Level (i)	Number of Nodes	Work per Node	Work per Level
0	1	$cn^2$	$cn^2$
1	3	$c\left(\frac{n}{4}\right)^2$	$\frac{3}{16}cn^2$
2	9	$c\left(\frac{n}{16}\right)^2$	$\frac{9}{256}cn^2$
$i$	$3^i$	$c\left(\frac{n}{4^i}\right)^2$	$\left(\frac{3}{16}\right)^i cn^2$
base	$3^{\log_4 n}$	4	$4 \cdot 3^{\log_4 n}$

Last recursive level:  $\log_4 n - 1$

Combining it all together...

$$T(n) = 4n^{\log_4 3} + \sum_{i=0}^{\log_4 n - 1} \left(\frac{3}{16}\right)^i cn^2$$

# Technique 3: Master Theorem

Given a recurrence of the following form:

$$T(n) = \begin{cases} d & \text{when } n = 1 \\ aT\left(\frac{n}{b}\right) + n^c & \text{otherwise} \end{cases}$$

Where  $a$ ,  $b$ , and  $c$  are constants, then  $T(n)$  has the following asymptotic bounds

If  $\underline{\log_b a < c}$  then  $\underline{T(n) \in \Theta(n^c)}$

If  $\underline{\log_b a = c}$  then  $\underline{T(n) \in \Theta(n^c \log_2 n)}$

If  $\underline{\log_b a > c}$  then  $\underline{T(n) \in \Theta(n^{\log_b a})}$

# Apply Master Theorem

Given a recurrence of the form:

$$T(n) = \begin{cases} d & \text{when } n = 1 \\ aT\left(\frac{n}{b}\right) + n^c & \text{otherwise} \end{cases}$$

If  $\log_b a < c$  then  $T(n) \in \Theta(n^c)$

If  $\log_b a = c$  then  $T(n) \in \Theta(n^c \log_2 n)$

If  $\log_b a > c$  then  $T(n) \in \Theta(n^{\log_b a})$

$$T(n) = \begin{cases} 1 & \text{when } n \leq 1 \\ 2T\left(\frac{n}{2}\right) + n^c & \text{otherwise} \end{cases}$$

$$\begin{aligned} a &= 2 \\ b &= 2 \\ c &= 1 \\ d &= 1 \end{aligned}$$

$$\log_b a = c \Rightarrow \log_2 2 = 1$$

$$T(n) \in \Theta(n^c \log_2 n) \Rightarrow \Theta(n^1 \log_2 n)$$

# Reflecting on Master Theorem

Given a recurrence of the form:

$$T(n) = \begin{cases} d & \text{when } n = 1 \\ aT\left(\frac{n}{b}\right) + n^c & \text{otherwise} \end{cases}$$

If  $\log_b a < c$  then  $T(n) \in \Theta(n^c)$

If  $\log_b a = c$  then  $T(n) \in \Theta(n^c \log_2 n)$

If  $\log_b a > c$  then  $T(n) \in \Theta(n^{\log_b a})$

$height \approx \log_b a$

$branchWork \approx n^c \log_b a$

$leafWork \approx d(n^{\log_b a})$

The  $\log_b a < c$  case

- Recursive case conquers work more quickly than it divides work
- Most work happens near “top” of tree
- Non recursive work in recursive case dominates growth,  $n^c$  term

The  $\log_b a = c$  case

- Work is equally distributed across call stack (throughout the “tree”)
- Overall work is approximately work at top level  $\times$  height

The  $\log_b a > c$  case

- Recursive case divides work faster than it conquers work
- Most work happens near “bottom” of tree
- Leaf work dominates branch work

# Recurrence analysis techniques

## 1. Unfolding method

- more of a brute force method
- Tedious but works

## 2. Tree methods

- more scratch work but less error prone

## 3. Master theorem

- quick, but applicable only to certain type of recurrences
- does not give a closed form (gives big-Theta)