

Divide and Conquer Algorithms

CPSC 320 2023W2

Definitions

A divide and conquer algorithm proceeds by...

- Dividing the input into two or more **smaller instances** of the same problems
 - We call these **subproblems**
- Solving the subproblems recursively
- Combining the subproblem solutions to obtain a solution to the original problem

Examples

Some divide and conquer algorithms you are already familiar with:

- QuickSort
- MergeSort

Recurrence relations

- The running time **$T(n)$** of a recursive function can be described using a **recurrence relation**:
 - **$T(n)$** is defined in terms of one or more terms of the form **$T(\text{something smaller than } n)$**
 - Example:

One recursive call on $n/2$ items.

Two recursive calls on $n/4$ items.

n^2 work not done inside a recursive call

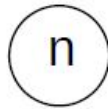
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Recursion Trees

- One way to solve a recurrence relation is to draw a **recursion tree**
 - Represent the recursion with a tree where each node represents a recursive subproblem
 - Inside each node, write the size of the subproblem this call to the function solves
 - Next to each node, write the amount of work done by the call to the function, **not including** any time spent in recursive calls
 - Compute the total amount of non-recursive work done on each row
 - Then add up the work done over all rows

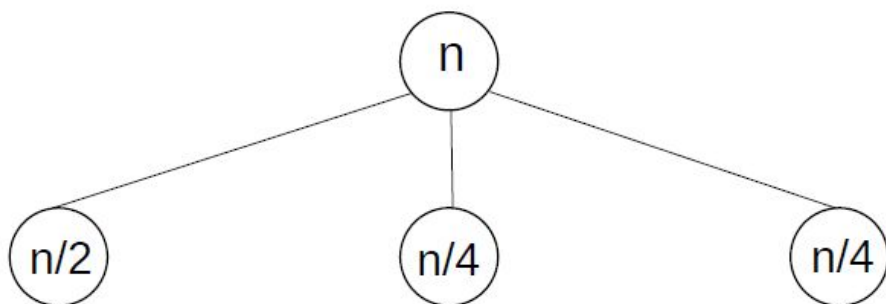
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- Example: drawing the tree



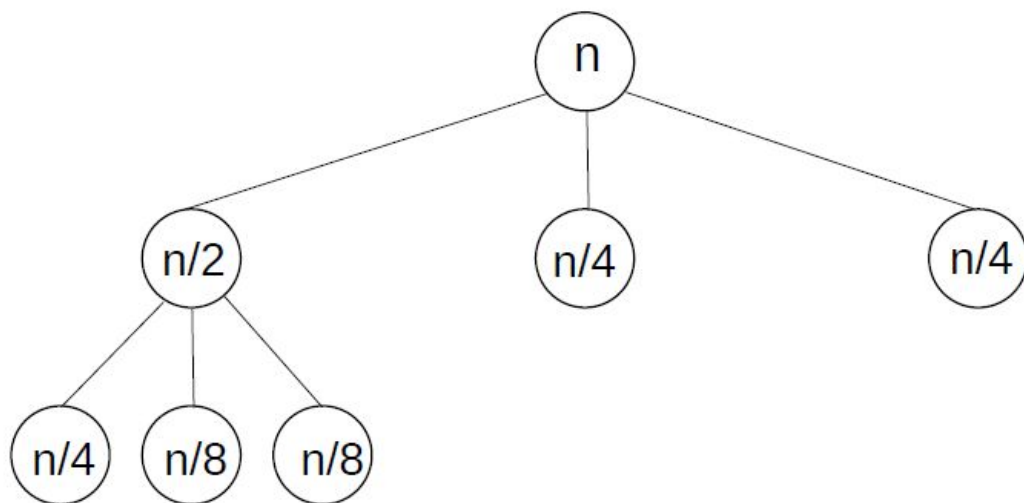
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- Example: drawing the tree (continued)



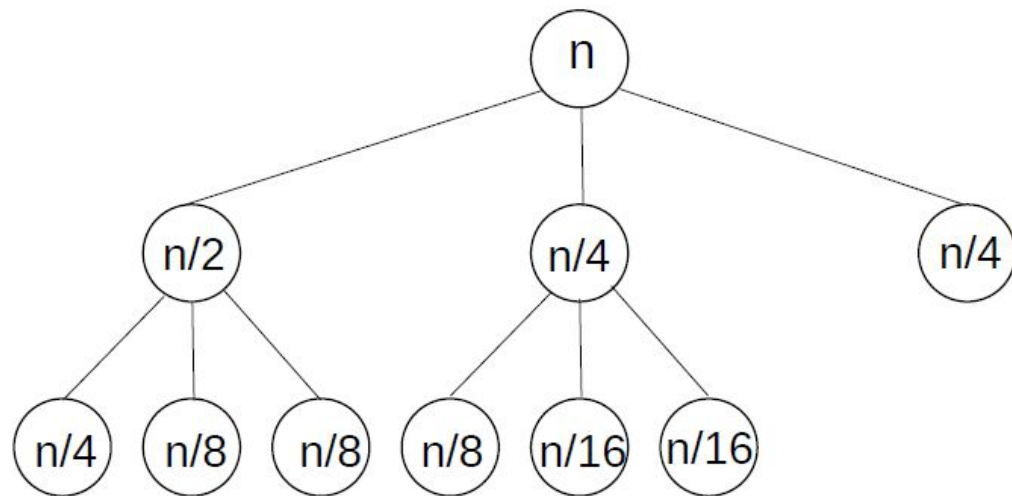
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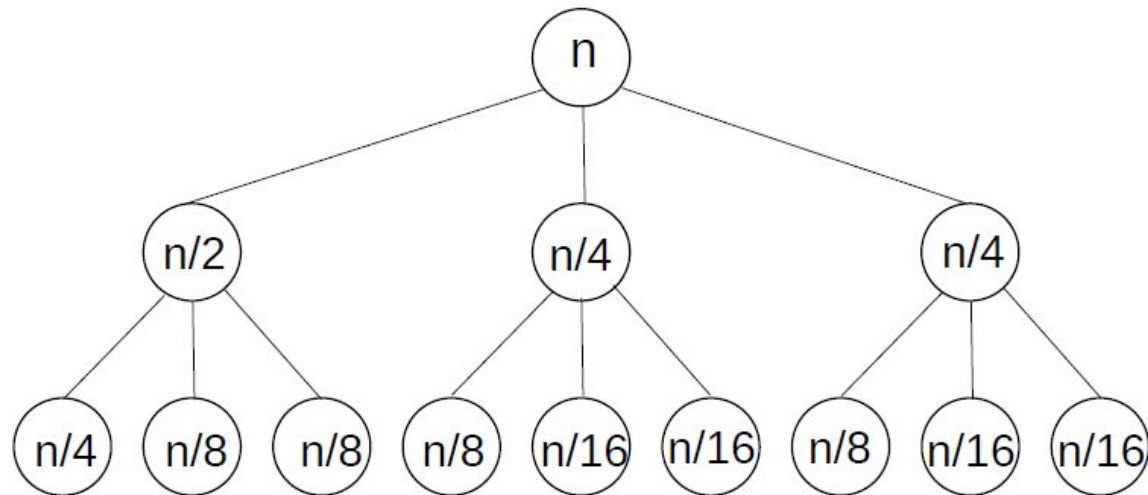
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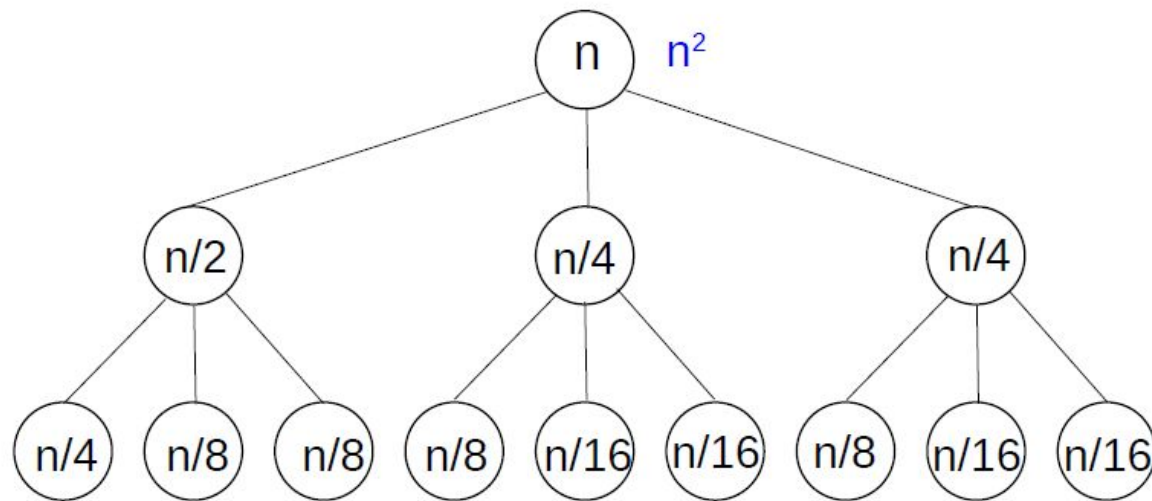
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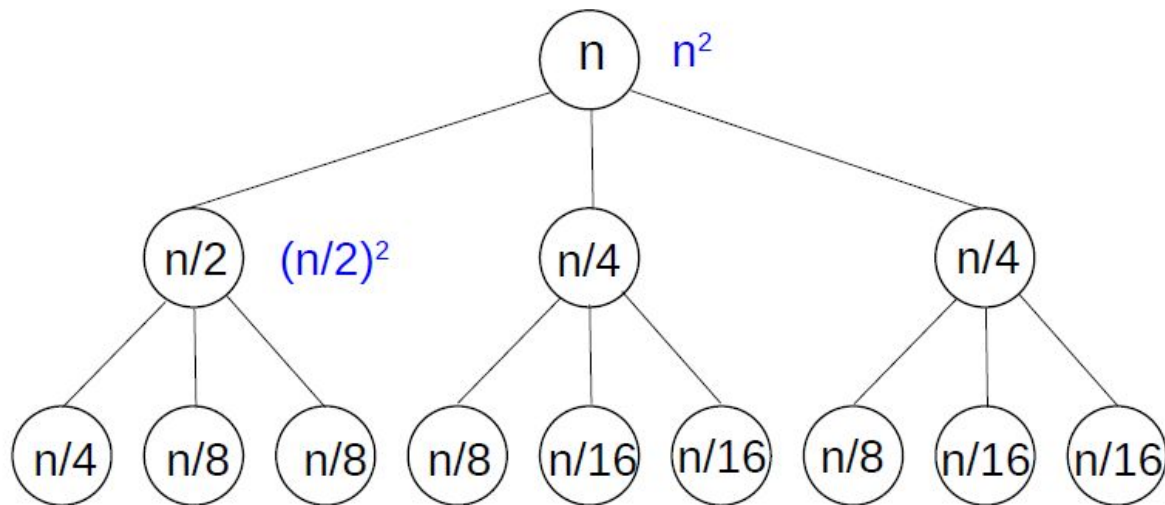
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Example: work done at each node



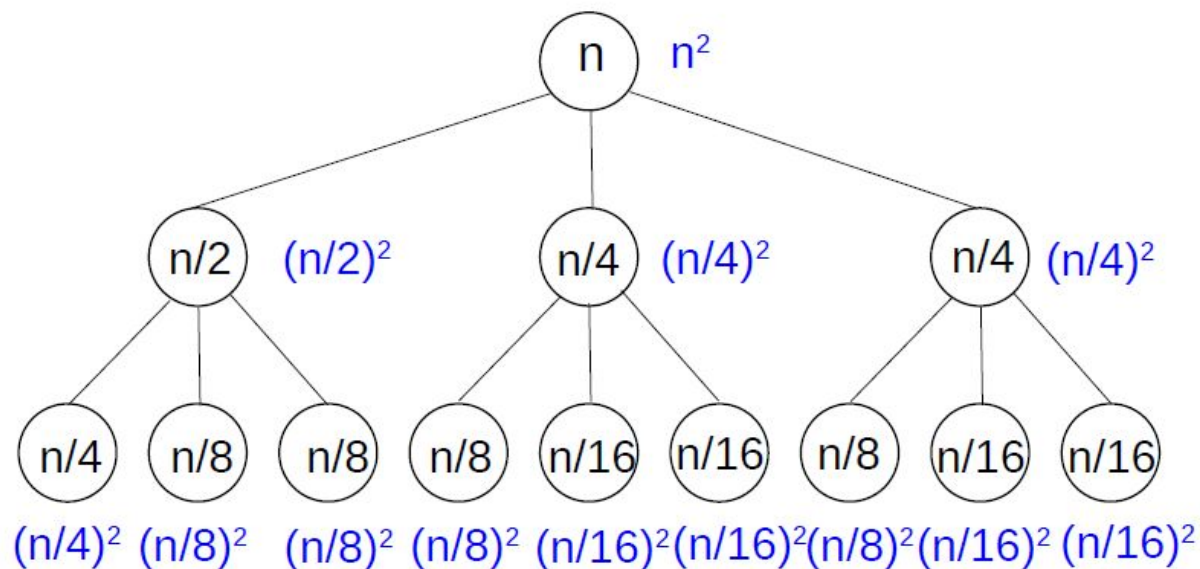
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Example: work done at each node (continued)



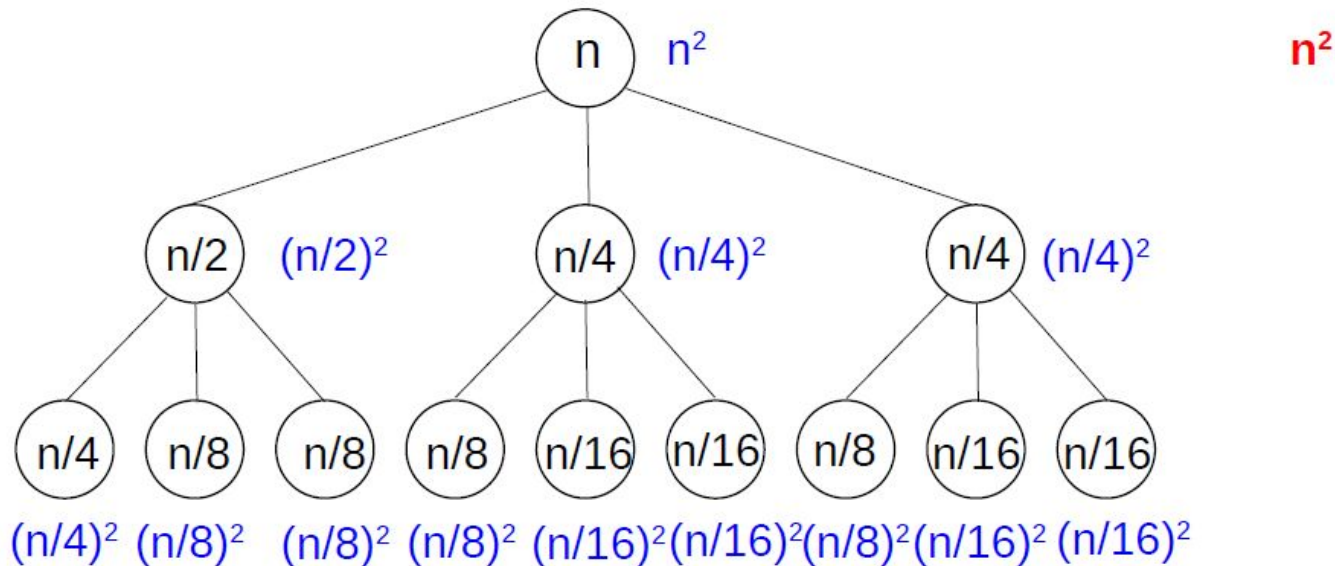
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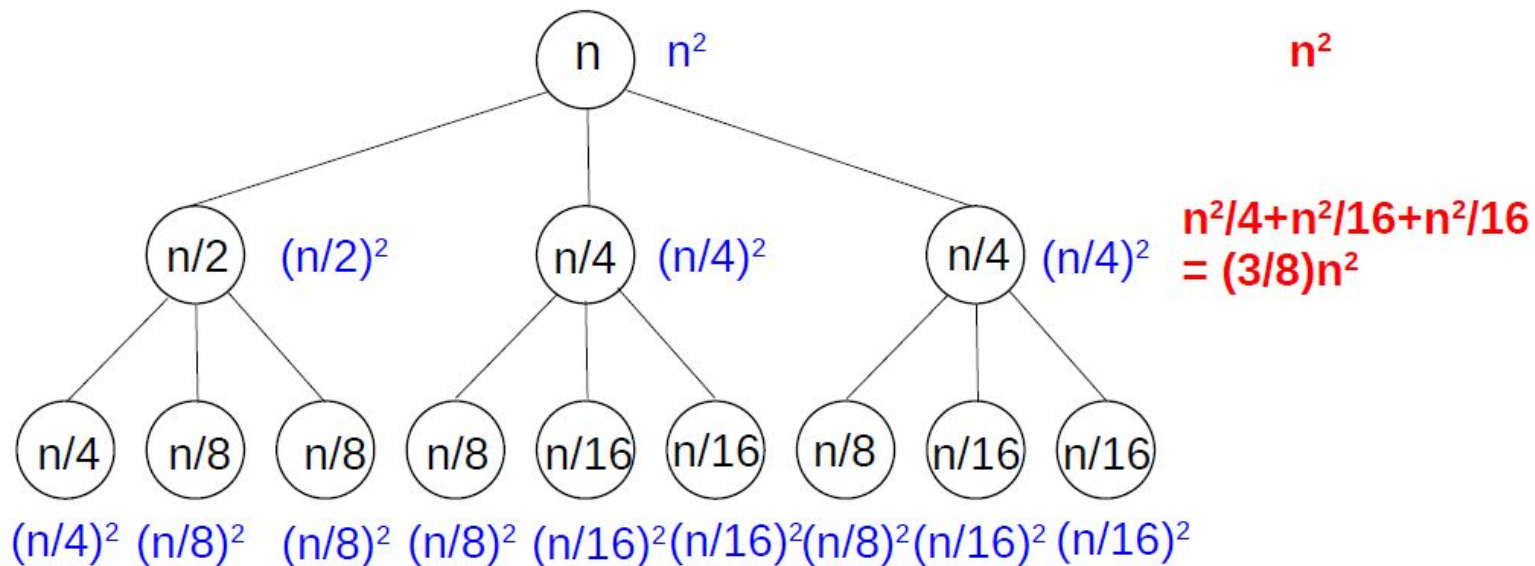
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- Example: work done on each row



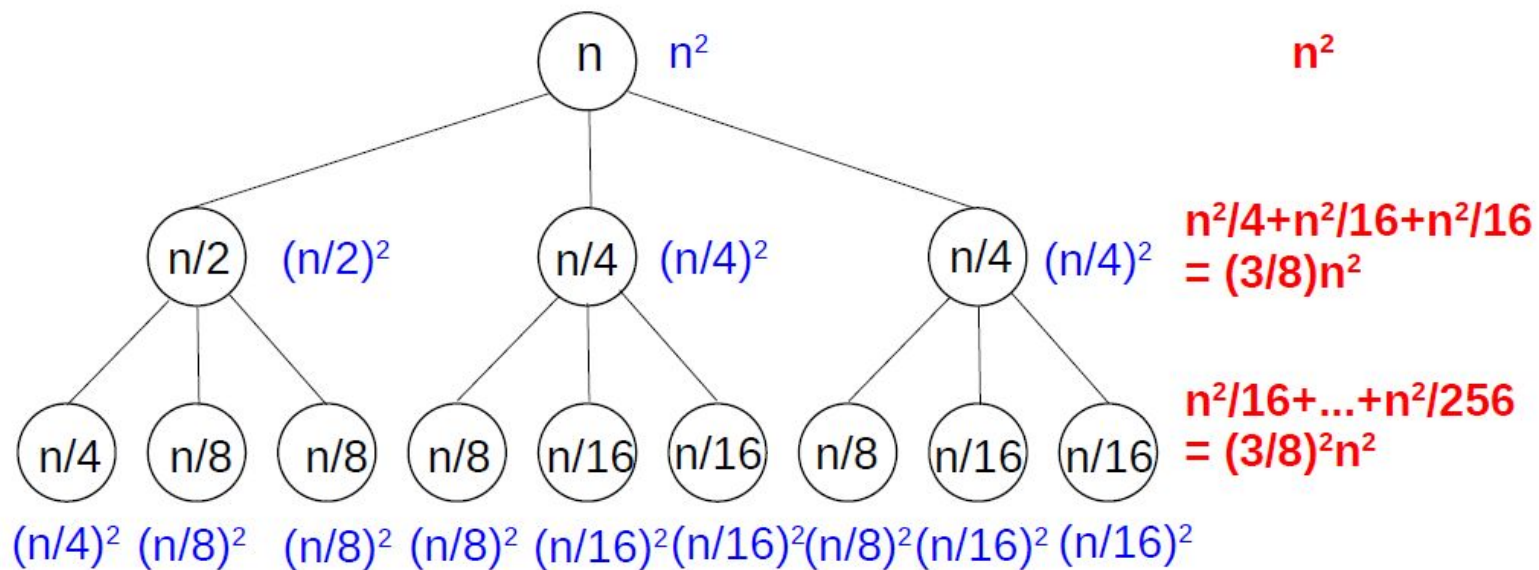
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- Example: summing up the work on all the rows
 - The total work is $n^2 + (3/8)n^2 + (3/8)^2n^2 + \dots$
 - This is a geometric series, and $3/8 < 1$.
 - So the sum converges to $\frac{1}{1-3/8}n^2$
 - Hence $T(n) \in \Theta(n^2)$

The Master Theorem

- Most divide and conquer algorithms split the input into equal-size subproblems
- Most recursion trees fall into one of three categories:
 - The work per level increases geometrically
 - The work per level is constant (e.g., MergeSort)
 - The work per level decreases geometrically (e.g., the previous example)

The Master Theorem [Bentley, Haken, Saxe]

- Theorem: Let $a \geq 1$, $b > 1$ be real constants, let $f: \mathbb{N} \rightarrow \mathbb{R}^+$, and let $T(n)$ be defined by:

$$T(n) = \begin{cases} aT(n/b) + f(n) & \text{if } n \geq n_0 \\ \Theta(1) & \text{if } n < n_0 \end{cases}$$

where n/b might be either $\lfloor n/b \rfloor$ or $\lceil n/b \rceil$. Then

1. If $f(n) \in O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$ then $T(n) \in \Theta(n^{\log_b a})$.
2. If $f(n) \in \Theta(n^{\log_b a} \log^k n)$ for some $k \geq 0$ then $T(n) \in \Theta(n^{\log_b a} \log^{k+1} n)$.
3. If $f(n) \in \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$ and $af(n/b) < \delta f(n)$ for some $0 < \delta < 1$ and all n large enough, then $T(n) \in \Theta(f(n))$.

↑
regularity condition

The Master Theorem [Bentley, Haken, and Saxe]

How to apply the theorem:

- Compute $\log_b a$
- Compare it to the exponent of n in $f(n)$
 - If $\log_b a$ is larger: case 1.
 - If they are equal: maybe case 2.
 - If $\log_b a$ is smaller: check regularity condition, and if it holds then case 3.