CMPSC 465: LECTURE IV

Solving Recurrences

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September 05, 2025

Solving recurrences

Recall that the time complexity T(n) of MergeSort satisfies:

$$T(n) = T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + \Theta(n).$$

Simplification: assume n is a power of 2 so we can ignore floors and ceilings.

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Method 1: Solve by substitution

- ▶ Make a guess, e.g., $T(n) = O(n \log n)$.
- Try to prove the guess by induction.

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In order to prove our guess, we need to show that $T(n) \le c \cdot n \log n$ for some constant c.

Assume this is true for all m < n, in particular, for m = n/2. Substituting into the recurrence gives

$$\begin{split} T(n) &= 2T(n/2) + \Theta(n) \\ &\leq 2c \cdot (n/2) \log(n/2) + O(n) \\ &= 2c \cdot (n/2) \log n - 2c \cdot n/2 + O(n) \\ &\leq c \cdot n \log n - c \cdot n + c' \cdot n \\ &= c \cdot n \log n - (c - c') \cdot n \\ &\leq c \cdot n \log n. \end{split}$$

The last step holds as long as we choose $c \geq c'$.

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

$$\leq 2c \cdot n/2 + O(n)$$

$$\leq 2c \cdot n/2 + c' \cdot n$$

$$= (c + c')n$$

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$$= (c + c')n > c \cdot n$$

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Note: When the induction proof fails, it does not necessarily mean the initial guess was wrong.

$$T(n) = 2T(n/2) + \Theta(n).$$

We can also guess a lower bound $T(n) = \Omega(n \log n)$.

Need to show that $T(n) \ge c \cdot n \log n$ for some constant c.

Assume this is true for all m < n, in particular, for m = n/2. Substituting into the recurrence gives

$$\begin{split} T(n) &= 2T(n/2) + \Theta(n) \\ &\geq 2c \cdot (n/2) \log(n/2) + \Omega(n) \\ &= 2c \cdot (n/2) \log n - 2c \cdot n/2 + \Omega(n) \\ &\geq c \cdot n \log n - c \cdot n + c' \cdot n \\ &= c \cdot n \log n - (c - c') \cdot n \\ &\geq c \cdot n \log n. \end{split}$$

The last step holds as long as we choose $c \le c'$.

$$T(n) = 2T(n/2) + \Theta(n).$$

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$$\le 2(2T(n/4) + c \cdot (n/2)) + c \cdot n$$

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$$= 4T(n/4) + 2c \cdot n$$

$$T(n) = 2T(n/2) + \Theta(n).$$

$$\begin{split} T(n) & \leq 2T(n/2) + c \cdot n \\ & \leq 2\left(2T(n/4) + c \cdot (n/2)\right) + c \cdot n \\ & = 4T(n/4) + 2c \cdot n \\ & \leq 4\left(2T(n/8) + c \cdot (n/4)\right) + 2c \cdot n \end{split}$$

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$$T(n) = 2T(n/2) + \Theta(n).$$

To get an upper bound, we can try unrolling the recurrence :

$$\begin{split} T(n) & \leq 2T(n/2) + c \cdot n \\ & \leq 2\left(2T(n/4) + c \cdot (n/2)\right) + c \cdot n \\ & = 4T(n/4) + 2c \cdot n \\ & \leq 4\left(2T(n/8) + c \cdot (n/4)\right) + 2c \cdot n \\ & = 8T(n/8) + 3c \cdot n \\ & \vdots \\ & \leq 2^k T(n/2^k) + k \cdot c \cdot n \end{split}$$

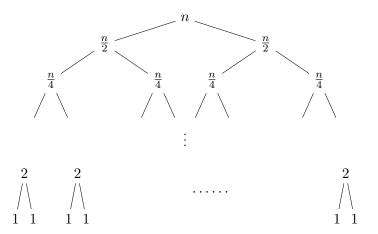
When $n/2^k=1$, or $k=\log n$, we reach the base case T(1)=O(1). So the final result is $T(n)\leq 2^kO(1)+k\cdot c\cdot n$

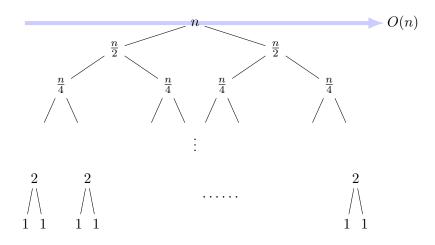
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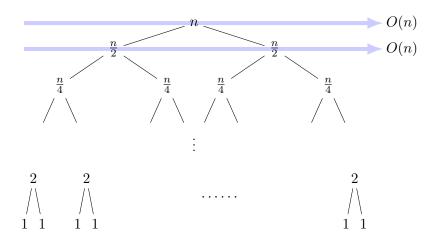
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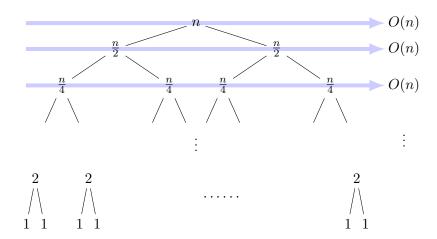
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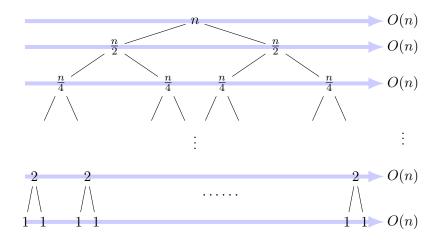
When $n/2^k=1$, or $k=\log n$, we reach the base case T(1)=O(1). So the final result is $T(n)\leq 2^kO(1)+k\cdot c\cdot n=O(n)+c\cdot n\log n=O(n\log n)$.

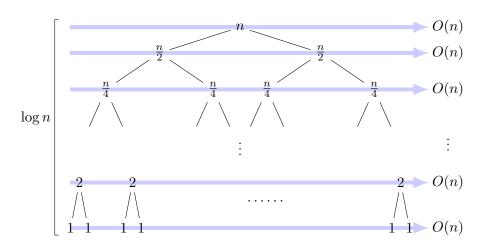


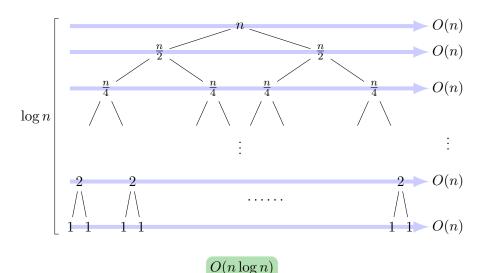




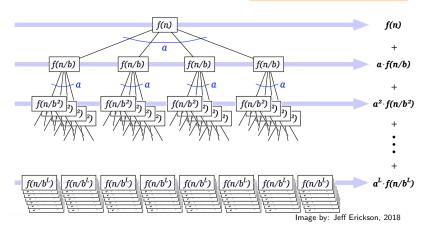




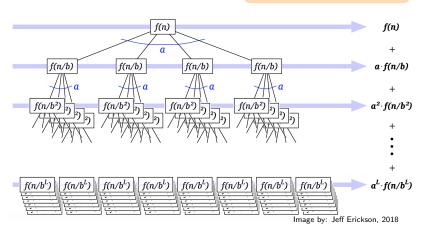




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So
$$T(n) = \sum\limits_{i=0}^{L} a^i \cdot f(n/b^i)$$
 where $L = \log_b n$.

$$T(n) = a \cdot T(n/b) + f(n)$$
 solves to $T(n) = \sum_{i=0}^{\log_b n} a^i \cdot f(n/b^i)$.

If f(n) is bounded above by $f(n) = O(n^d)$, we have

$$T(n) = \sum_{i=0}^{\log_b n} a^i \cdot O\left(\left(\frac{n}{b^i}\right)^d\right)$$

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$$= O(n^d) \cdot \sum_{i=0}^{\log_b n} \left(\frac{a}{b^d}\right)^i$$

$$T(n) = a \cdot T(n/b) + O(n^d)$$
 solves to $T(n) = O(n^d) \cdot \sum_{i=0}^{\log_b n} \left(\frac{a}{b^d}\right)^i$.

If $a/b^d = 1$, $T(n) = O(n^d \log n)$.

Fact: The sum of any geometric series is a constant times its largest term:

- ▶ If $a/b^d < 1$, largest term is the first term, $T(n) = O(n^d)$.
- ▶ If $a/b^d > 1$, largest term is the last term

$$(a/b^d)^{\log_b n} = a^{\log_b n}/(b^{\log_b n})^d = n^{\log_b a}/n^d,$$

so
$$T(n) = O(n^{\log_b a})$$
 .

Master theorem If $T(n) = aT(\lceil n/b \rceil) + O(n^d)$ for some constants a > 0, b > 1, and $d \ge 0$, then

$$T(n) = \begin{cases} O(n^d) & \text{if } a/b^d < 1\\ O(n^d \log n) & \text{if } a/b^d = 1\\ O(n^{\log_b a}) & \text{if } a/b^d > 1 \end{cases}.$$

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