CS 341: Algorithms Lecture 2: Solving Recurrences

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Based on lecture notes by Éric Schost and many previous CS 341 instructors

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Design Idea for MergeSort

Input: Array A of n integers

- Step 1: We split A into two subarrays: A_L consists of the first $\lceil \frac{n}{2} \rceil$ elements in A and A_R consists of the last $\lfloor \frac{n}{2} \rfloor$ elements in A.
- **Step 2:** Recursively run MergeSort on A_L and A_R .
- **Step 3:** After A_L and A_R have been sorted, use a function *Merge* to merge them into a single sorted array.

Recurrence Relations

The mergesort recurrence is

$$T(n) = \begin{cases} T\left(\left\lceil \frac{n}{2} \right\rceil\right) + T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + \Theta(n) & \text{if } n > 1\\ \Theta(1) & \text{if } n = 1. \end{cases}$$

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It is simpler to consider the following *exact* recurrence, with constant factors c and d replacing Θ 's:

$$T(n) = \begin{cases} T\left(\left\lceil \frac{n}{2} \right\rceil\right) + T\left(\left\lfloor \frac{n}{2} \right\rfloor\right) + cn & \text{if } n > 1 \\ d & \text{if } n = 1. \end{cases}$$

Recurrence Relations (cont.)

The following is the corresponding *sloppy* recurrence (it has floors and ceilings removed):

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

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We solve the sloppy recurrence when $n = 2^{j}$ using the *recursion tree method*.

We can construct a recursion tree for the sloppy recurrence, assuming $n = 2^{j}$, as follows.

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- Sum the values on each level of the tree, and then compute the sum of all these sums; the result is T(n).

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- If this solution is expressed as a function of *n* using
 Θ-notation, then we obtain the complexity of the solution of the exact recurrence for *all n*.
- This is not a proof, however. If a real mathematical proof is required, then it is necessary to use induction.

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Theorem (Master theorem)

Suppose that $a \ge 1$ and b > 1. Consider the recurrence

$$T(n) = a T\left(\frac{n}{b}\right) + \Theta(n^{y})$$

in sloppy or exact form. Denote $x = \log_b a$. Then

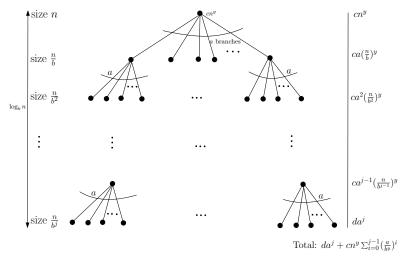
$$T(n) \in \begin{cases} \Theta(n^{x}) & \text{if } y < x \\ \Theta(n^{y} \log n) & \text{if } y = x \\ \Theta(n^{y}) & \text{if } y > x. \end{cases}$$

Suppose that $n = b^j$, $a \ge 1$, $b \ge 2$ are integers and

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Let $n = b^j$.

size of subproblem	# nodes	cost/node	total cost
$n = b^j$	1	c n ^y	c n ^y
$n/b = b^{j-1}$	а	$c (n/b)^{y}$	c a (n/ b) ^y
$n/b^2 = b^{j-2}$	a ²	$c(n/b^2)^y$	$c a^2 (n/b^2)^y$
÷	÷	:	÷
$n/b^{j-1} = b$	a^{j-1}	$c(n/b^{j-1})^y$	$c a^{j-1} (n/b^{j-1})^y$
$n/b^j=1$	a ^j	d	d a ^j

Summing the costs of all levels of the recursion tree, we have that

$$T(n) = d a^j + c n^y \sum_{i=0}^{j-1} \left(rac{a}{b^y}
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There are three cases, depending on whether r > 1, r = 1 or r < 1.

Complexity of T(n)

case	r	y, x	complexity of $T(n)$
heavy leaves	r > 1	y < x	$T(n) \in \Theta(n^{x})$
balanced	r = 1	y = x	$T(n)\in \Theta(n^y\log n)$
heavy top	r < 1	y > x	$T(n)\in \Theta(n^{y})$

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"heavy top" means that cost of the recursion tree is dominated by the cost of the root node.

The substitution method

To solve a recurrence do the following:

- Guess the solution (or the form of the solution)
- Use induction to prove it (and if needed, find a constant)

Example: $T(n) = 2T(\frac{n}{2}) + n$