Algorithms and Data Structures

A11. Runtime Analysis: Solving Recurrences

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Algorithms and Data Structures

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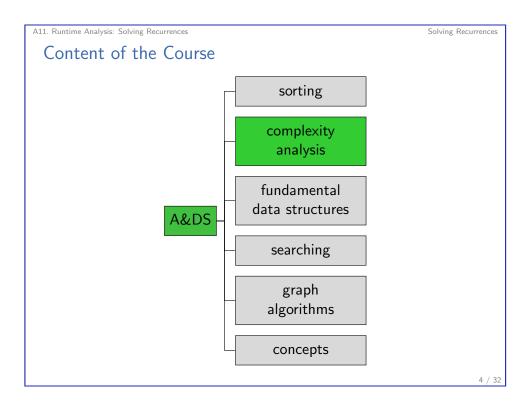
A11.6 Summary

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A11. Runtime Analysis: Solving Recurrences

Solving Recurrences

A11.1 Solving Recurrences



A11. Runtime Analysis: Solving Recurrences

Introduction

In Ch. A10, we derived (algorithmic) recurrences from divide-and-conquer algorithms:

- ► $T(m) = T(m/2) + \Theta(m)$ for merge sort.
- ► $T(n) = 8T(n/2) + \Theta(1)$ for simple recursive matrix multiplication.
- ► $T(n) = 7T(n/2) + \Theta(n^2)$ for Strassen's algorithm for matrix multiplication.

For the asymptotic running time, we want an expression that only depends on the input size (and not recursively on T)!

How can we solve such recurrences?

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Substitution Method

A11.2 Substitution Method

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Approaches

- substitution method
- recursion-tree method
- master theorem
- ► Akra-Bazzi method
 - ► Generalization of the master theorem.
 - Not covered in this course.

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A11. Runtime Analysis: Solving Recurrences

Substitution Method

Solving Recurrences

Substitution Method

- Guess the form of the solution using symbolic constants.
- 2 Use mathematical induction to show that the solution works.

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Example: Top-down Merge Sort I

Try to Guess the Form of the Solution

Consider $T(m) = 2T(m/2) + \Theta(m)$

Consider $m = 2^k$ with $k \in \mathbb{N}_{>0}$

$$T(m) = 2T(m/2) + c'm$$

$$= 2(2T(m/4) + c'(m/2)) + c'm$$

$$= 2c'm + 4T(m/4)$$

$$= 2c'm + 4(2T(m/8) + c'(m/4))$$

$$= 3c'm + 8T(m/8)$$

$$= ...$$

$$= kc'm + 2^k c_0 \qquad \text{(use } c_0 \text{ for } T(1)\text{)}$$

$$= c'm \log_2 m + mc_0 \qquad (k = \log_2 m, 2^k = m)$$

$$\leq (c_0 + c')m \log_2 m \qquad (\log_2 m = k \geq 1)$$

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Example: Top-down Merge Sort II

Guess Solution and Formulate Hypothesis

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Guess: $T(m) \in O(m \log_2 m)$

Hypothesis: $T(m) \le cm \log_2 m$ for all $m \ge m_0$ for some constants $c, m_0 > 0$ (taken care of later).

Let's try the inductive step with this hypothesis.

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A11. Runtime Analysis: Solving Recurrences

Substitution Method

Example: Top-down Merge Sort II

Verify Guessed Solution with Induction (Inductive Case)

Assume by induction that

 $T(m') \le cm' \log_2(m')$ for all m' with $m_0 \le m' < m$.

Inductive step: $m-1 \rightarrow m$, where $m > 2m_0$:

$$T(m) = 2T(m/2) + c'm$$

$$\leq 2c(m/2)\log_2(m/2) + c'm \qquad \text{(induction hypothesis)}$$

$$= cm\log_2(m) - cm\log_2 2 + c'm$$

$$= cm\log_2(m) - cm + c'm$$

$$< cm\log_2(m) \qquad \text{if } c > c'$$

Inductive steps works if we constrain c to be sufficiently large such that cm > c'm for all $m > 2m_0$ (with c' hidden constant from O(m)).

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Substitution Method

Example: Top-down Merge Sort II

Verify Guessed Solution with Induction (Base Case)

Show that $T(m) \le cm \log_2(m)$ for all m with $m_0 \le m < 2m_0$. Consider $m_0 = 2$.

- ▶ Let $d = \max\{T(2), T(3)\}$
 - ► $T(2) \le d \le d2 \log_2 2$
 - $T(3) < d < d3 \log_2 3$

With $c = \max\{c', d\}$ it holds for all $m \ge 2$ that

$$T(m) \leq cm \log_2(m)$$
.

We have shown that $T(m) \in O(m \log_2 m)$.

Method A11. Runtime Analysis: Solving Recurrences

Recursion-tree Method

Recursion-tree Method

In a recursion tree, each node represents the cost of a single subproblem somewhere in the set of recursive function invocations.

Analyse the cost on each level of the tree and the depth of the tree to get an idea of the overall running time.

Suitable for making a good guess (to be verified by induction).

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A11.3 Recursion-tree Method

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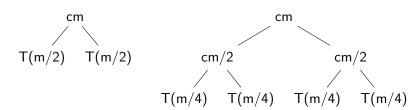
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Recursion-tree Method

Example: Top-down Merge Sort I

Consider again $m = 2^k$ with $k \in \mathbb{N}_{>0}$

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



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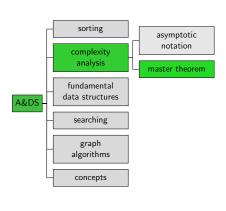
Master Theorem

A11.4 Master Theorem

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A11. Runtime Analysis: Solving Recurrences

Content of the Course



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A11. Runtime Analysis: Solving Recurrences

Master Theorem

Master Recurrences

A common instantiation of the divide-and-conquer algorithm scheme works as follows:

- ▶ For inputs of small size n < C, solve the problem directly.
- ► Otherwise:
 - ① Construct A smaller inputs of size n/B.
 - 2 Recursively solve these inputs using the same algorithm.
 - Ompute the result from the recursively computed results.

If 1.+3. take time f(n), the overall run-time for n > Ccan be expressed as $T(n) = A \cdot T(n/B) + f(n)$.

- ► We call this a master recurrence.
- ightharpoonup f(n) is called the driving function.
- \blacktriangleright We do not care about run-time for n < Cbecause it does not affect asymptotic analysis.

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Master Theorem

Master Recurrences – Examples

Reminder:

- ① Construct A smaller inputs of size n/B.
- 2 Recursively solve these inputs using the same algorithm.
- 3 Compute the result from the recursively computed results.

master recurrence: $T(n) = A \cdot T(n/B) + f(n)$

Examples:

- ► Merge sort: A = 2, B = 2, $f(n) = \Theta(n)$
- ► Strassen's algorithm: A = 7, B = 2, $f(n) = \Theta(n^2)$

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Master Theorem: Intuition (Case 1)

Master Theorem

The theorem compares the asymptotic growth of the driving function to the one of the watershed function $n^{\log_B A}$:

Theorem

Let $A \ge 1$, B > 1 be constants and f(n) be a driving function that is defined and nonnegative on all sufficiently large reals. Let T satisfy the master recurrence $T(n) = A \cdot T(n/B) + f(n)$. Then:

- ▶ If $f(n) = O(n^{\log_B A \varepsilon})$ for some $\varepsilon > 0$, then $T(n) = \Theta(n^{\log_B A})$.
- If $f(n) = \Theta(n^{\log_B A} \log_2^k n)$ for some $k \ge 0$, then $T(n) = \Theta(n^{\log_B A} \log_2^{k+1} n)$.
- ▶ If $f(n) = \Omega(n^{\log_B A + \varepsilon})$ for some $\varepsilon > 0$ and if $Af(n/B) \le cf(n)$ for some c < 1 and all sufficiently large n, then $T(n) = \Theta(f(n))$.

 $f(n) = O(n^{\log_B A - \varepsilon})$

- ► Watershed function grows polynomially faster than the driving function.
- ► Cost per level in recursion tree grows at least geometrically from root to leaves.
- Cost of leaves dominates total cost of inner nodes.

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Asster Theorem

Master Theorem: Intuition (Case 2)

$$f(n) = \Theta(n^{\log_B A} \log_2^k n)$$

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- ▶ Both functions grow at nearly the same asymptotic rates.
- ▶ Precisely: driving function only grows faster than the watershed function by a factor of $\log_2^k n$.
- ► Each level of the tree costs approximately the same.
- ▶ With k = 0, the second case covers case $f(n) = \Theta(n^{\log_B A})$.

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Master Theorer

Master Theorem: Intuition (Case 3)

$$f(n) = \Omega(n^{\log_B A + \varepsilon})$$

- ▶ Driving function grows polynomially faster than the watershed function.
- ▶ Regularity condition $Af(n/B) \le cf(n)$ is typically satisfied (no big growth differences of driving function in different areas of the recursion tree).
- ► Cost per level in recursion tree drops at least geometrically from root to leaves.
- Cost of root dominates cost of other nodes in the recursion tree.

Master Theorem

Application: Merge Sort

Reminder: $T(n) = A \cdot T(n/B) + f(n)$

- $f(n) = O(n^{\log_B A \varepsilon}) \rightsquigarrow T(n) = \Theta(n^{\log_B A})$
- $f(n) = \Theta(n^{\log_B A} \log_2^k n) \rightsquigarrow T(n) = \Theta(n^{\log_B A} \log_2^{k+1} n)$
- $f(n) = \Omega(n^{\log_B A + \varepsilon}) \rightsquigarrow T(n) = \Theta(f(n))$

Merge sort: A = 2, B = 2, $f(n) = \Theta(n)$ $\rightarrow \log_B A = \log_2 2 = 1$

- ► Case 1 $f(n) = O(n^{1-\varepsilon})$ for some $\varepsilon > 0$? \rightsquigarrow No.
- ► Case 2 $f(n) = \Theta(n^1 \log_2^k n)$ for some k? \rightsquigarrow Yes with k = 0! $\rightsquigarrow T(n) = \Theta(n^1 \log_2 n)$

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Floors and Ceilings?

A11.5 Floors and Ceilings?

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Master Theorem

Application: Strassen's Algorithm

Reminder: $T(n) = A \cdot T(n/B) + f(n)$

- $f(n) = O(n^{\log_B A \varepsilon}) \rightsquigarrow T(n) = \Theta(n^{\log_B A})$
- $f(n) = \Theta(n^{\log_B A} \log_2^k n) \rightsquigarrow T(n) = \Theta(n^{\log_B A} \log_2^{k+1} n)$
- $f(n) = \Omega(n^{\log_B A + \varepsilon}) \rightsquigarrow T(n) = \Theta(f(n))$

Strassen's algorithm: A = 7, B = 2, $f(n) = \Theta(n^2)$ $\Rightarrow \log_B A = \log_2 7 \approx 2.807...$

► Case 1 $f(n) = O(n^{\log_2 7 - \varepsilon})$ for some $\varepsilon > 0$? \rightsquigarrow Yes, for instance with $\varepsilon = 0.8! \rightsquigarrow T(n) = \Theta(n^{\log_2 7})$

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Floors and Ceilings?

Merge Sort Revisited

We used $T(m) = 2T(m/2) + \Theta(m)$ as recurrence for merge sort.

If m = 5, we have one recursive call for 2 and one for 3 elements.

The precise recurrence for the running time is

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

Does this make a difference for the asymptotic growth?

Good News

Ignoring floors and ceilings does not generally affect the order of growth of the solution of a divide-and-conquer recurrence.

The master theorem also holds for recurrences

$$T(n) = A'T(\lceil n/B \rceil) + A''T(\lceil n/B \rceil) + f(n)$$

for some constant $A', A'' \ge 0$ (set A := A' + A'').

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Summar

A11.6 Summary

Example: Merge Sort (Optional Material)

Inductive Case Revisited

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Assume by induction that

 $T(m') \le cm' \log_2(m')$ for all m' with $m_0 \le m' < m$.

Inductive step: $m-1 \rightarrow m$, where $m \ge 2m_0$:

$$T(m) = T(\lfloor m/2 \rfloor) + T(\lceil m/2 \rceil) + c'm$$

$$\leq c \lfloor m/2 \rfloor \log_2(\lfloor m/2 \rfloor) + c \lceil m/2 \rceil \log_2(\lceil m/2 \rceil) + c'm$$

$$\leq c \lfloor m/2 \rfloor \log_2(\lceil m/2 \rceil) + c \lceil m/2 \rceil \log_2(\lceil m/2 \rceil) + c'm$$

$$\leq c m \log_2(\lceil m/2 \rceil) + c'm \qquad (\lfloor m/2 \rfloor + \lceil m/2 \rceil = m)$$

$$\leq c m \log_2((m+1)/2) + c'm \qquad (\lceil m/2 \rceil \leq (m+1)/2)$$

$$= c m \log_2(m+1) - cm + c'm$$

$$\leq c m \log_2(m+1) - cm + c'm \qquad (\log_2(m+1) \leq \log_2(m) + \frac{1}{m})$$

$$= c m \log_2 m + c - cm + c'm$$

$$\leq c m \log_2 m \qquad \text{if } c > 2c' \text{ and } m \geq 2$$

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A11. Runtime Analysis: Solving Recurrences

Summan

Summary

- ► The substitution method is the most general one:
 - ► Guess the running time (typically by substituting the recursive term a few times).
 - Prove by mathematical induction that the guess is correct.
- ► The recursion-tree method is good for quickly getting an impression of a running time.
- ► The master theorem is not always applicable. If it is, it is the quickest way to determine the running time.
- ► Top-down merge sort has linearithmic running time $\Theta(m \log_2 m)$.