Recurrences

Recurrences and Running Time

- Recurrences arise when an algorithm contains recursive calls to itself
- An equation or inequality that describes a function in terms of its value on smaller inputs.

$$T(n) = T(n-1) + n$$

- · What is the actual running time of the algorithm?
- · Need to solve the recurrence
 - Find an explicit formula of the expression
 - Bound the recurrence by an expression that involves n

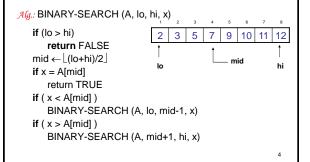
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Example Recurrences

- T(n) = T(n-1) + n $\Theta(n^2)$
 - Recursive algorithm that loops through the input to eliminate one item
- T(n) = T(n/2) + c $\Theta(Iqn)$
 - Recursive algorithm that halves the input in one step
- T(n) = T(n/2) + n $\Theta(n)$
 - Recursive algorithm that halves the input but must examine every item in the input
- T(n) = 2T(n/2) + 1 $\Theta(n)$
 - Recursive algorithm that splits the input into 2 halves and does a constant amount of other work

Recurrent Algorithms BINARY-SEARCH

• for an ordered array A, finds if x is in the array A[lo...hi]



Methods for Solving Recurrences

- · Iteration method
- · Substitution method
- · Master method
- · Recursion tree method

The Iteration Method

- Convert the recurrence into a summation and try to bound it using known series
 - Iterate the recurrence until the initial condition is reached.
 - Use back-substitution to express the recurrence in terms of n and the initial (boundary) condition.

The Iteration Method

Iteration Method - Example

```
T(n) = n + 2T(n/2) \qquad \text{Assume: } n = 2^k
T(n) = n + 2T(n/2) \qquad T(n/2) = n/2 + 2T(n/4)
= n + 2(n/2 + 2T(n/4))
= n + n + 4T(n/4)
= n + n + 4(n/4 + 2T(n/8))
= n + n + n + 8T(n/8)
... = in + 2^iT(n/2^i)
= kn + 2^kT(1)
= n|gn + nT(1) = \Theta(n|gn)
```

Methods for Solving Recurrences

- · Iteration method
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- · Recursion tree method
- · Master method

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The substitution method

- 1. Guess a solution
- 2. Use induction to prove that the solution works

Substitution method

- · Guess a solution
 - T(n) = O(g(n))
 - Induction goal: apply the definition of the asymptotic notation
 - $T(n) \le cg(n)$, for some c > 0 and $n \ge n_0$
 - Induction hypothesis: T(k) ≤ cg(k) for all k < n
- Prove the induction goal
 - Use the induction hypothesis to find some values of the constants c and n₀ for which the induction goal holds

Example: Binary Search

$$T(n) = c + T(n/2)$$

- Guess: T(n) = O(lgn)
 - Induction goal: $T(n) \le d \lg n$, for some d and $n \ge n_0$
 - Induction hypothesis: $T(n/2) \le d \lg(n/2)$
- Proof of induction goal:

$$T(n) = T(n/2) + c \le d \lg(n/2) + c$$
$$\le d \lg n - d + c$$

 $T(n) \le d \lg n$

 $(d \lg n - d + c \le d \lg n)$ if: $-d + c \le 0, d \ge c$

Example 2

T(n) = T(n-1) + n

- Guess: $T(n) = O(n^2)$
 - Induction goal: $\mathbf{T}(\mathbf{n}) \leq \mathbf{c} \ \mathbf{n}^2$, for some c and $\mathbf{n} \geq \mathbf{n}_0$
 - Induction hypothesis: $T(n-1) \le c(n-1)^2$
- Proof of induction goal:

$$\begin{split} T(n) &= T(n\text{-}1) + n \le c \; (n\text{-}1)^2 + n \\ &\le \; cn^2 - (2cn - c - n) \\ &\quad T(n) \le cn^2 \; \text{ (i.e. } cn^2 - (2cn - c - n) \le cn^2 \text{)} \end{split}$$

if: $2cn - c - n \ge 0 \implies c \ge n/(2n-1) \implies c \ge 1/(2-1/n)$

- For $n \ge 1 \implies 2 - 1/n \ge 1 \implies$ any $c \ge 1$ will work

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Methods for Solving Recurrences

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Master's method

• Solving recurrences of the form:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

where, $a \ge 1$, $b \ge 1$, and $f(n) \ge 0$

Idea: Compare f(n) with nlog a

- f(n) is asymptotically smaller or larger than $n^{log}_b{}^a$ by a polynomial factor n^g
- f(n) is asymptotically equal with $n^{\log_b a}$

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Master's method

· Solving recurrences of the form:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

where, $a \ge 1$, b > 1, and f(n) > 0

Case 1: if $f(n) = O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$, then: $T(n) = \Theta(n^{\log_b a})$

Case 2: if $f(n) = \Theta(n^{\log_b a})$, then: $T(n) = \Theta(n^{\log_b a} \lg n)$

Case 3: if $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$, and if

 $af(n/b) \le cf(n)$ for some c < 1 and all sufficiently large n, then:

 $T(n) = \Theta(f(n))$ regularity condition

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Examples

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

$$a = 2$$
, $b = 2$, $log_2 2 = 1$

Compare $n^{\log_2 2} (n^{\log_b a})$ with f(n) = n

- \rightarrow f(n) = Θ (n) (Case 2)
- \rightarrow T(n) = Θ (nlgn)

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Examples (cont.)

$$\mathbf{T}(\mathbf{n}) = 2\mathbf{T}(\mathbf{n}/2) + \sqrt{n}$$

$$a = 2$$
, $b = 2$, $log_2 2 = 1$

Compare **n** with $f(n) = n^{1/2}$

$$\Rightarrow$$
 f(n) = O(n^{1-\varepsilon}) (Case 1)

 \Rightarrow T(n) = Θ (n)

Examples

$$T(n) = 2T(n/2) + n^2$$
 a = 2, b = 2, $\log_2 2 = 1$

Compare \mathbf{n} with $\mathbf{f}(\mathbf{n}) = \mathbf{n}^2$

$$\rightarrow$$
 $f(n) = \Omega(n^{1+\epsilon})$ (Case 3)

→ verify regularity condition (a $f(n/b) \le c f(n)$)

$$2 n^2/4 \le c n^2 \implies c = \frac{1}{2}$$
 is a solution (c<1)

$$\Rightarrow$$
 T(n) = Θ (n²)

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$$T(n) = 3T(n/4) + nlgn$$

$$a = 3$$
, $b = 4$, $\log_4 3 = 0.793$

Compare $n^{0.793}$ with f(n) = nlgn

$$f(n) = \Omega(n^{\log_4^{3+\epsilon}})$$
 (Case 3)

Check regularity condition:

$$3*(n/4)lg(n/4)$$

$$\leq$$
 (3/4)nlgn = c *f(n), c=3/4

$$\Rightarrow$$
T(n) = Θ (nlgn)

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Examples

$$T(n) = 2T(n/2) + nlgn$$

$$a = 2$$
, $b = 2$, $log_2 2 = 1$

- Compare n with f(n) = nlgn
 - seems like case 3 should apply
- f(n) must be polynomially larger by a factor of n^ϵ
- In this case it is only larger by a factor of Ign

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Master's method

· Solving recurrences of the form:

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

where, $a \ge 1$, b > 1, and f(n) > 0

Case 1: if $f(n) = O(n^{\log_b a - \epsilon})$ for some $\epsilon > 0$, then: $T(n) = \Theta(n^{\log_b a})$

Case 2: if $f(n) = \Theta(n^{\log_b a})$, then: $T(n) = \Theta(n^{\log_b a} \log n)$

Case 3: if $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some $\epsilon > 0$, and if

 $af(n/b) \le cf(n)$ for some c < 1 and all sufficiently large n, then:

 $T(n) = \Theta(f(n))$ regularity condition

$$T(n) = aT\left(\frac{n}{b}\right) + f(n)$$

$$T(n) = aT\left(\frac{n}{b}\right)$$

$$A^{2}T\left(\frac{n}{b^{2}}\right)$$

$$A^{3}T\left(\frac{n}{b^{2}}\right)$$

$$\vdots$$

$$T(n) = a^{i}T\left(\frac{n}{b^{i}}\right)$$

$$\vdots$$

$$T(n) = a^{i}T\left(\frac{n}{b^{i}}\right) \quad \forall i$$

$$\cdot \text{ Case 3:}$$

$$- \text{ If } f(n) \text{ is dominated by } n^{\log_{b}a^{i}}:$$

$$\cdot T(n) = \Theta(n^{\log_{b}a})$$

$$\cdot T(n) = \Theta(f^{\log_{b}a})$$

$$\cdot T(n) = \Theta(f(n))$$

$$\cdot T(n) = \Theta(f(n))$$

$$\cdot T(n) = \Theta(f(n))$$

$$\cdot T(n) = \Theta(f^{\log_{b}a}):$$

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The recursion-tree method

Convert the recurrence into a tree:

- Each node represents the cost incurred at various levels of recursion

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- Sum up the costs of all levels
- · Used to "guess" a solution for the recurrence

Common Summations

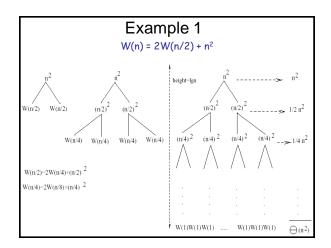
 $\sum_{k=1}^{n} k = 1 + 2 + \dots + n = \frac{n(n+1)}{2}$ · Arithmetic series:

 $\sum_{k=0}^{n} x^{k} = 1 + x + x^{2} + \dots + x^{n} = \frac{x^{n+1} - 1}{x - 1} (x \neq 1)$ · Geometric series:

Special case: |x| < 1:

 $\sum_{k=1}^{n} \frac{1}{k} = 1 + \frac{1}{2} + \dots + \frac{1}{n} \approx \ln n$ · Harmonic series:

• Other important formulas: $\sum_{i=1}^n \lg k \approx n \lg n$ $\sum_{i=1}^n k^p = 1^p + 2^p + ... + n^p \approx \frac{1}{p+1} n^{p+1}$



Subproblem size at level i is:
$$n/2^{i}$$

Subproblem size hits 1

when $1 = n/2^{i} \Rightarrow i = \lg n$

Cost of the problem at level $i = (n/2^{i})^{2}$

No. of nodes at level $i = 2^{i}$

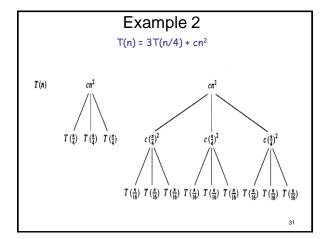
Total cost:

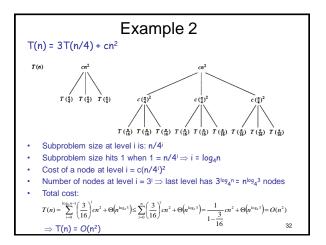
$$W(n) = \sum_{i=0}^{\lg n-1} \frac{n^{2}}{2^{i}} + 2^{i} * W(1) = \sum_{i=0}^{\lg n-1} \frac{n^{2}}{2^{i}} + 2^{\lg n} * W(1)$$

$$= n^{2} \sum_{i=0}^{\lg n-1} \left(\frac{1}{2}\right)^{i} + n \le n^{2} \sum_{i=0}^{\infty} \left(\frac{1}{2}\right)^{i} + O(n)$$

$$= n^{2} \frac{1}{1 - \frac{1}{2}} + O(n) = 2n^{2}$$

$$\Rightarrow W(n) = O(n^{2})$$





Changing variables

$$T(n) = 2T(n/2) + lgn$$

- Rename:
$$m = lgn \Rightarrow n = 2^m$$

$$T(2^m) = 2T(2^{m/2}) + m$$

- Rename:
$$S(m) = T(2^m)$$

$$S(m) = 2S(m/2) + m \Rightarrow S(m) = O(mlgm)$$

(demonstrated before)

$$T(n) = T(2^m) = S(m) = O(mlgm) = O(lgnlglgn)$$

Idea: transform the recurrence to one that you have seen before

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Substitution method more examples

Example 3

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

- Guess: T(n) = O(nlgn)
 - Induction goal: $T(n) \le cn \ lgn$, for some c and $n \ge n_0$
 - Induction hypothesis: $T(n/2) \le cn/2 \lg(n/2)$
- Proof of induction goal:

$$T(n) = 2T(n/2) + n \le 2c (n/2)lg(n/2) + n$$

= $cn lgn - cn + n$

$$T(n) \le cn lgn (i.e. cn lgn - cn + n \le cn lgn)$$

if:
$$-cn + n \le 0 \Rightarrow c \ge 1$$

Example 4

$$T(n) = 3T(n/4) + cn^2$$

- Guess: $T(n) = O(n^2)$
 - Induction goal: $T(n) \le dn^2$, for some d and $n \ge n_0$

if: $d \ge (16/13)c$

- Induction hypothesis: $T(n/4) \le d(n/4)^2$
- Proof of induction goal:

$$T(n) = 3T(n/4) + cn^2$$

$$\leq 3d (n/4)^2 + cn^2$$

$$= (3/16) d n^2 + cn^2$$

• Therefore: $T(n) = O(n^2)$

 $\leq d n^2$