Analysis of Algorithms

Recurrences



Recurrences and Running Time

 An equation or inequality that describes a function in terms of its value on smaller inputs.

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

- Recurrences arise when an algorithm contains recursive calls to itself
- 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

Example Recurrences

•
$$T(n) = T(n-1) + n$$

$$\Theta(n^2)$$

 Recursive algorithm that loops through the input to eliminate one item

•
$$T(n) = T(n/2) + c$$

$$\Theta(Ign)$$

- 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

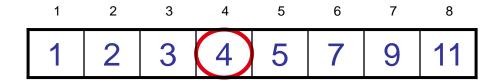
Recurrent Algorithms BINARY-SEARCH

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

```
Alg.: BINARY-SEARCH (A, lo, hi, x)
                                          2
                                              3
                                                                7
                                                       5
                                                                     8
                                                            6
    if (lo > hi)
                                         3
                                              5
        return FALSE
    mid \leftarrow \lfloor (lo+hi)/2 \rfloor
                                                        mid
                                     lo
                                                                    hi
    if x = A[mid]
        return TRUE
    if (x < A[mid])
        BINARY-SEARCH (A, Io, mid-1, x)
    if (x > A[mid])
        BINARY-SEARCH (A, mid+1, hi, x)
```

Example

• $A[8] = \{1, 2, 3, 4, 5, 7, 9, 11\}$ - Io = 1 Io = 1 Io = 8 Io = 7



mid = 4, lo = 5, hi = 8

mid = 6, A[mid] = xFound!

Another Example

• $A[8] = \{1, 2, 3, 4, 5, 7, 9, 11\}$ - lo = 1 hi = 8 x = 65 mid = 4, lo = 5, hi = 8† high † low 3 mid = 6, A[6] = 7, Io = 5, Hi = 5† high low 3 mid = 5, A[5] = 5, Io = 6, Ho = 5**NOT FOUND!** 3 11 5 9 4 high

Analysis of BINARY-SEARCH

```
Alg.: BINARY-SEARCH (A, Io, hi, x)
     if (lo > hi)
                                                       constant time: c<sub>1</sub>
        return FALSE
     mid \leftarrow \lfloor (lo+hi)/2 \rfloor
                                                       constant time: c<sub>2</sub>
     if x = A[mid]
                                                       constant time: c<sub>3</sub>
         return TRUE
     if (x < A[mid])
         BINARY-SEARCH (A, Io, mid-1, x) \leftarrow same problem of size n/2
     if (x > A[mid])
        BINARY-SEARCH (A, mid+1, hi, x) ← same problem of size n/2
```

•
$$T(n) = c + T(n/2)$$

T(n) – running time for an array of size n

Methods for Solving Recurrences

Iteration method

Substitution method

Recursion tree method

Master 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

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

 $T(n) = c + T(n/2)$
 $T(n/2) = c + T(n/4)$
 $T(n/2) = c + T(n/4)$
 $T(n/4) = c + T(n/8)$
 $T(n/4) = c + T(n/8)$
Assume $n = 2^k$
 $T(n) = c + c + ... + c + T(1)$
 $t + times$
 $t = clgn + T(1)$
 $t = \Theta(lgn)$

Iteration Method – Example

```
T(n) = n + 2T(n/2) Assume: n = 2^k
T(n) = n + 2T(n/2)
                              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^{i}T(n/2^{i})
     = kn + 2^kT(1)
     = nlgn + nT(1) = \Theta(nlgn)
```

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) ≤ d g(n), for some d > 0 and n ≥ n₀
 - Induction hypothesis: $T(k) \le d g(k)$ for all k < n (strong induction)
- Prove the induction goal
 - Use the induction hypothesis to find some values of the constants d 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) ≤ d lgn, for some d and n ≥ n₀
 - Induction hypothesis: T(n/2) ≤ d Ig(n/2)
- Proof of induction goal:

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

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

Base case?

Example 2

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

- Guess: $T(n) = O(n^2)$
 - Induction goal: T(n) ≤ c n², for some c and n ≥ n₀
 - Induction hypothesis: T(n-1) ≤ c(n-1)² for all k < n</p>
- Proof of induction goal:

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

Example 3

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

- Guess: T(n) = O(nlgn)
 - Induction goal: T(n) ≤ cn Ign, for some c and n ≥ n₀
 - Induction hypothesis: T(n/2) ≤ cn/2 lg(n/2)
- Proof of induction goal:

T(n) = 2T(n/2) + n
$$\leq$$
 2c (n/2)lg(n/2) + n
= cn lgn - cn + n \leq cn lgn
if: - cn + n \leq 0 \Rightarrow c \geq 1

Base case?

Changing variables

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

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

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

- 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

The recursion-tree method

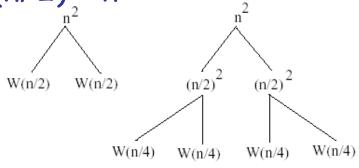
Convert the recurrence into a tree:

- Each node represents the cost incurred at various levels of recursion
- Sum up the costs of all levels

Used to "guess" a solution for the recurrence

Example 1





 $W(n/2)=2W(n/4)+(n/2)^{-2}$

 $W(n/4)=2W(n/8)+(n/4)^{-2}$

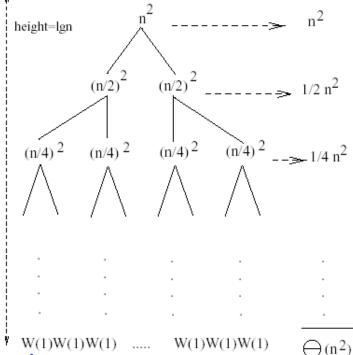


- Subproblem size hits 1 when $1 = n/2^i \Rightarrow i = lgn$
- Cost of the problem at level $i = (n/2^i)^2$ No. of nodes at level $i = 2^i$

• Total cost:

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

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



Example 2

E.g.:
$$T(n) = 3T(n/4) + cn^2$$

$$T(\frac{n}{4}) T(\frac{n}{4}) T(\frac{n}{4}) T(\frac{n}{4}) C(\frac{n}{4})^2 C(\frac{n}{4})^2 C(\frac{n}{4})^2$$

$$T(\frac{n}{16}) T(\frac{n}{16}) T(\frac{n}{16}) T(\frac{n}{16}) T(\frac{n}{16}) T(\frac{n}{16}) T(\frac{n}{16}) T(\frac{n}{16})$$

- Subproblem size at level i is: n/4ⁱ
- Subproblem size hits 1 when $1 = n/4^i \Rightarrow i = log_4 n$
- Cost of a node at level i = c(n/4ⁱ)²
- Number of nodes at level $i = 3^i \Rightarrow$ last level has $3^{\log_4 n} = n^{\log_4 3}$ nodes
- Total cost:

$$T(n) = \sum_{i=0}^{\log_4 n - 1} \left(\frac{3}{16}\right)^i cn^2 + \Theta\left(n^{\log_4 3}\right) \le \sum_{i=0}^{\infty} \left(\frac{3}{16}\right)^i cn^2 + \Theta\left(n^{\log_4 3}\right) = \frac{1}{1 - \frac{3}{16}} cn^2 + \Theta\left(n^{\log_4 3}\right) = O(n^2)$$

$$\Rightarrow T(n) = O(n^2)$$
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Example 2 - Substitution

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

- Guess: $T(n) = O(n^2)$
 - Induction goal: T(n) ≤ dn², for some d and n ≥ n₀
 - Induction hypothesis: T(n/4) ≤ d (n/4)²
- Proof of induction goal:

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

 $\le 3d (n/4)^2 + cn^2$
= $(3/16) d n^2 + cn^2$
 $\le d n^2$ if: $d \ge (16/13)c$

• Therefore: $T(n) = O(n^2)$

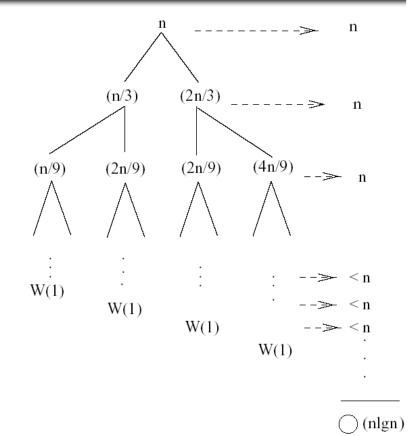
Example 3 (simpler proof)

$$W(n) = W(n/3) + W(2n/3) + n$$

 The longest path from the root to a leaf is:

$$n \rightarrow (2/3)n \rightarrow (2/3)^2 \ n \rightarrow ... \rightarrow 1$$

- Subproblem size hits 1 when
 1 = (2/3)ⁱn ⇔ i=log_{3/2}n
- Cost of the problem at level i = n
- Total cost:



$$W(n) < n + n + \dots = n(\log_{3/2} n) = n \frac{\lg n}{\lg \frac{3}{2}} = O(n \lg n)$$

$$\Rightarrow$$
 W(n) = O(nlgn)

Example 3

W(n) = W(n/3) + W(2n/3) + n

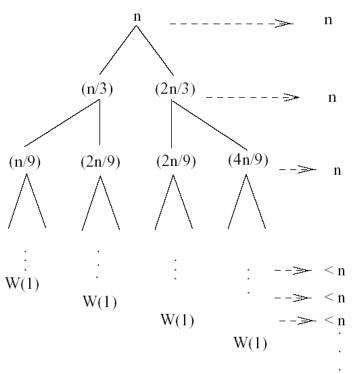
The longest path from the root to a leaf is:

$$n \rightarrow (2/3)n \rightarrow (2/3)^2 \ n \rightarrow ... \rightarrow 1$$

- Subproblem size hits 1 when $1 = (2/3)^{i} n \Leftrightarrow i = \log_{3/2} n$
- Cost of the problem at level i = n
- Total cost:

Cost:
$$W(n) < n + n + ... = \sum_{i=0}^{(\log_{3/2} n) - 1} n + 2^{(\log_{3/2} n)} W(1) < < n \sum_{i=0}^{\log_{3/2} n} 1 + n^{\log_{3/2} 2} = n \log_{3/2} n + O(n) = n \frac{\lg n}{\lg 3/2} + O(n) = \frac{1}{\lg 3/2} n \lg n + O(n)$$

$$\Rightarrow W(n) = O(n \lg n)$$
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Example 3 - Substitution

$$W(n) = W(n/3) + W(2n/3) + O(n)$$

- Guess: W(n) = O(nlgn)
 - Induction goal: W(n) ≤ dnlgn, for some d and n ≥ n₀
 - Induction hypothesis: W(k) ≤ d klgk for any K < n (n/3, 2n/3)
- Proof of induction goal:

Try it out as an exercise!!

• T(n) = O(nlgn)

Master's method

"Cookbook" for 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

Idea: compare f(n) with nlogba

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

Master's method

"Cookbook" for 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
```

Examples

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

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

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

$$\Rightarrow$$
 f(n) = Θ (n) \Rightarrow Case 2

$$\Rightarrow$$
 T(n) = Θ (nlgn)

Examples

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

$$\alpha = 2, b = 2, log_2 2 = 1$$
Compare n with $f(n) = n^2$

$$\Rightarrow f(n) = \Omega(n^{1+\epsilon}) \text{ Case } 3 \Rightarrow \text{verify regularity cond.}$$

$$\alpha f(n/b) \le c f(n)$$

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

$$\Rightarrow T(n) = \Theta(n^2)$$

Examples (cont.)

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

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

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

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

$$\Rightarrow$$
 T(n) = Θ (n)

Examples

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

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

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

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

Check regularity condition:

$$3*(n/4)Ig(n/4) \le (3/4)nIgn = c *f(n), c=3/4$$

$$\Rightarrow$$
T(n) = Θ (nlgn)

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 lgn

Readings