

CS 5321: Advanced Algorithms – Analysis Using Recurrence

Dr. Ali Ebneenasir
Department of Computer Science
Michigan Technological University

Acknowledgement

- Eric Torng
- Moon Jung Chung
- Charles Ofria

Outline

- Recursive algorithms vs. recurrence relations
- Specifying recurrence relations
- Solution techniques
 - Substitution method
 - Recursion-tree
 - Master theorem
 - Characteristic equations

Example: Towers of Hanoi

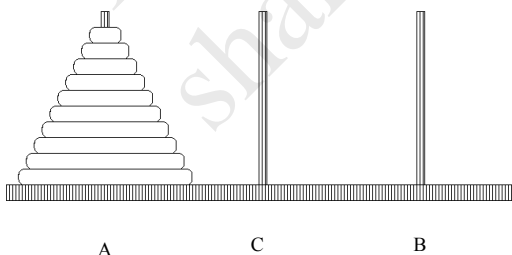
Problem Definition:

- Three pegs: A,B,C
- n disks
- Move n disks from A to B using C as a working peg

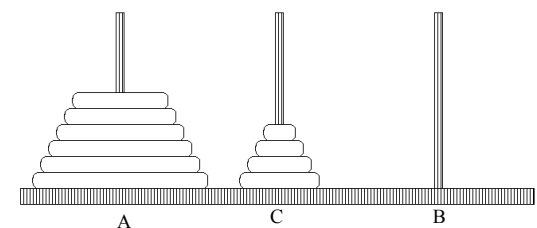
• Constraints:

- One disk at a time
- A larger disk cannot be on top of a smaller disk

The Towers of Hanoi

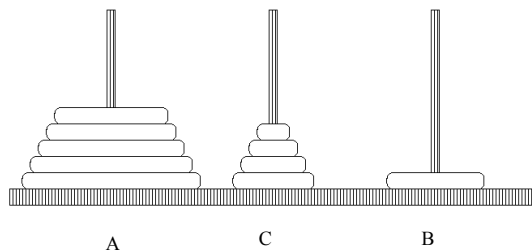


What if we knew we could solve
part of the problem?

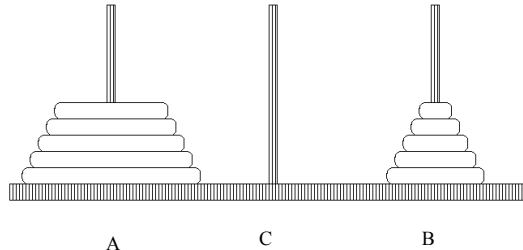


Assume we can move k (in this case, 4) different rings

Can we do one better?



Solved for one more!



Recurrence Relation

- $T(n)$: minimum number of moves needed for n disks
- $T(n-1)$: minimum number of moves needed for $n-1$ disks
- What is the relation between these two?
 - Try to calculate the number of moves from small values of n

Where do recurrence relations come from?

- Analysis of a divide and conquer algorithm
 - Towers of Hanoi, Merge Sort, Binary Search
- Analysis of a combinatorial object
 - up-down permutations
- **This is the key analysis step you should master**
- Use small cases to check correctness of your recurrence relation

Can recurrence relations be solved?

- No general procedure for solving recurrence relations is known, which is why it is an art.
- **However, linear, finite history, constant coefficient recurrences always can be solved**
- Example: $a_n = 2a_{n-1} + 2a_{n-2} + 1$; $a_1 = 1$; $a_2 = 1$
 - degree = 1
 - history = 2
 - coefficients = 2, 2, and 1

Recurrence Formulas

- $T(n) = a T(n/b) + f(n)$
- Divide the problem into a subproblems
 - The size of each subproblem is n/b
- The function $f(n)$ often represents the cost of dividing the problem to subproblems and then composing the solutions of subproblems; $f(n) = D(n) + C(n)$

Solution Techniques: Substitution Method

Substitution Method

1. Guess a solution
 - Try back-substituting until you know what is going on
 - Draw a recursion tree
2. Use induction to verify your guess

Recursion *and* Mathematical Induction

In both, we have general and boundary conditions:

The **general** conditions break the problem into smaller and smaller pieces.

The **initial** or **boundary** condition(s) terminate the recursion.

Both take a **Divide and Conquer** approach to solving mathematical problems.

First Step

- Using the base case and the recursive case, calculate small values
- Use these values to help guess a solution
- Use induction to verify the correctness of your solution

Guessing a Solution

We can use mathematical induction to prove that a general function solves for a recursive one.

$$T_n = 2T_{n-1} + 1 ; T_0 = 0$$

n	= 0	1	2	3	4	5	6	7	8
T _n	= 0	1	3	7	15	31	63	127	255

Guess what the solution is?

Prove by Induction

Prove: $T_n = 2^n - 1$ by induction:

1. Base Case: $n=0$: $T_0 = 2^0 - 1 = 0$
2. Inductive hypothesis: assume $T_n = 2^n - 1$ for $n \geq 0$
3. Inductive Step: Show $T_{n+1} = 2^{n+1} - 1$ for $n \geq 0$

$$\begin{aligned}
 T_{n+1} &= 2T_n + 1 \\
 &= 2(2^n - 1) + 1 \\
 &= 2^{n+1} - 1
 \end{aligned}$$

Substitution Method Drawbacks

- Sometimes it is difficult to guess a solution
- We need systematic approaches for coming up with a guess
 - Back-substitution
 - Recursion tree

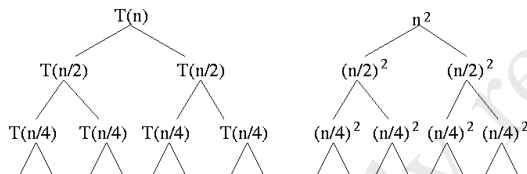
Back-Substitution

Example: $T(n) = 3T(\lfloor n/4 \rfloor) + n$, $T(1) = 1$

$$\begin{aligned}
 &= 3(3T(\lfloor n/16 \rfloor) + n/4) + n \\
 &= 9T(\lfloor n/16 \rfloor) + 3n/4 + n \\
 &= 9(3T(\lfloor n/64 \rfloor) + n/16) + 3n/4 + n \\
 &= 27T(\lfloor n/64 \rfloor) + 9n/16 + 3n/4 + n \\
 &= n \cdot \sum_{i=0}^{\log_4 n} \left(\frac{3}{4}\right)^i \leq \frac{1}{1 - 3/4} n = 4n
 \end{aligned}$$

Recursion Trees

$$T(n) = 2T(n/2) + n^2, \quad T(1) = 1$$



Example Problem

Use induction to prove that MergeSort is an $O(n \log n)$ algorithm.

```

MergeSort(array)
  n = size(array)
  if ( n == 1 ) return array
  array1 = MergeSort(array[1 .. n/2])
  array2 = MergeSort(array[n/2 .. n])
  return Merge(array1, array2)
    
```

Induction Proof

Example: Prove that $T(n) = 2T(\lfloor n/2 \rfloor) + n$, $T(1) = 1$ is $O(n \log n)$.

We need to prove that $T(n) \leq c n \log n$, for all n greater than some value.

1. Base cases: $T(2) = 4 \leq c \cdot 2$ and $T(3) = 5 \leq c \cdot 3 \log_2 3$
 $c \geq 2$ suffices
2. Inductive hypothesis: Assume $T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \log \lfloor n/2 \rfloor$

What is happening to the quantifications?

Induction Step

Given : $T(\lfloor n/2 \rfloor) \leq c \lfloor n/2 \rfloor \log \lfloor n/2 \rfloor$

$$\begin{aligned}
 T(n) &= 2T(\lfloor n/2 \rfloor) + n \\
 &\leq 2(c \lfloor n/2 \rfloor \log \lfloor n/2 \rfloor) + n \\
 &\leq 2(c(n/2) \log(n/2)) + n \quad (\text{dropping floors makes it bigger!}) \\
 &= c n \log(n/2) + n \\
 &= c n (\log(n) - \log(2)) + n \\
 &= c n \log(n) - c n + n \quad (\log_2 2 = 1) \\
 &= c n \log(n) - (c - 1) n \\
 &< c n \log(n) \quad (c > 1)
 \end{aligned}$$

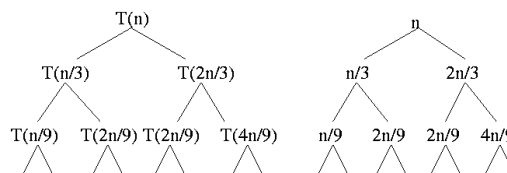
Example Problem 2

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

$$T(1) = 1$$

Show $T(n)$ is $\Omega(n \log n)$ by appealing to the recursion tree

Recursion Tree



What is the length of the longest path from root to a leaf?

Solution Techniques:
Master Theorem

Master Theorem

- $T(n) = a T(n/b) + f(n)$
 - Ignore floors and ceilings for n/b
 - constants $a \geq 1$ and $b > 1$ and $\epsilon > 0$
 - $f(n)$ an asymptotically positive function
- Case 1:
 - If $f(n) = O(n^{\log_b a - \epsilon})$ for constant $\epsilon > 0$, $T(n) = \Theta(n^{\log_b a})$
- Case 2:
 - If $f(n) = \Theta(n^{\log_b a})$, $T(n) = \Theta(n^{\log_b a} \lg n)$
- Case 3: If
 - $f(n) = \Omega(n^{\log_b a + \epsilon})$ for some constant $\epsilon > 0$, and
 - $a f(n/b) \leq c f(n)$ for some constant $c < 1$ and all sufficiently large n ,
 - Then, we have $T(n) = \Theta(f(n))$.

• Regularity condition: $a f(n/b) \leq c f(n)$ for some constant $c < 1$ and all sufficiently large n

Key idea: Compare $n^{\log_b a}$ with $f(n)$

Examples

- $T(n) = 9 T(n/3) + n$
 - $a=9, b=3, n^{\log_b a} = n^{\log_3 9} = \Theta(n^2)$
 - Thus, we have $f(n) = O(n^{\log_3 9 - \epsilon})$, case 1 applies
 - $T(n) = \Theta(n^2)$
- $T(n) = T(2n/3) + 1$
 - $a=1, b=3/2, n^{\log_b a} = n^0 = 1$
 - Thus, we have $f(n) = \Theta(n^{\log_b a})$; case 2 applies
 - $T(n) = \Theta(n^{\log_b a} \lg n) = \Theta(\lg n)$
- $T(n) = 3 T(n/4) + n \log n$
 - $a=3, b=4, n^{\log_b a} = O(n^{0.79})$
 - we have $f(n) = \Omega(n^{\log_b a + \epsilon})$ for $\epsilon \approx 0.2$, and
 - $3(n/4) \log(n/4) < (3/4) n \log n$, for $c = 3/4$; case 3 applies
 - $T(n) = \Theta(f(n)) = \Theta(n \log n)$

Examples

- $T(n) = 2 T(n/2) + n \log n$
 - $a=2, b=2, n^{\log_b a} = n^{\log_2 2} = \Theta(n)$
 - $f(n)$ is asymptotically larger than $(n^{1+\epsilon})$
 - Thus case 3 applies
 - Wrong! Why?
 - $f(n)$ is not polynomially larger; $f(n)/n = \log n$, which is asymptotically smaller than n^ϵ for any $\epsilon > 0$
 - Therefore, the regularity condition does not hold.

Solution Techniques: Characteristics Equations

Characteristic Equation Approach

- $t_n = 3t_{n-1} + 4t_{n-2}$ for $n > 1$
 - $t_0 = 0, t_1 = 5$
- Rewrite recurrence
 - $t_n - 3t_{n-1} - 4t_{n-2} = 0$
- Properties
 - **Homogeneous**: no terms not involving t_n
 - **Linear**: t_n terms have no squares or worse
 - **Constant coefficients**: 1, -3, -4

Characteristic Equation

- $t_n - 3t_{n-1} - 4t_{n-2} = 0$
- Rewrite assuming solution of the form $t_n = x^n$
- $x^n - 3x^{n-1} - 4x^{n-2} = 0$
- $x^{n-2}(x^2 - 3x - 4) = 0$
- Find roots of $(x^2 - 3x - 4)$
 - $(x+1)(x-4) \rightarrow$ roots are -1 and 4
- Solution is of form $c_1(-1)^n + c_2 4^n$

Solving for Constants

- $t_n = c_1(-1)^n + c_2 4^n$
- Use base cases to solve for constants
 - $t_0 = 0 = c_1(-1)^0 + c_2 4^0 = c_1 + c_2$
 - $t_1 = 5 = c_1(-1)^1 + c_2 4^1 = -c_1 + 4c_2$
 - $5c_2 = 5 \rightarrow c_2 = 1 \rightarrow c_1 = -1$
- $t_n = (-1)^{n+1} + 4^n$
- Always test solution on small values!

Repeated Roots for Characteristic Equations

- $t_n - 5t_{n-1} + 8t_{n-2} - 4t_{n-3} = 0$
 - boundary conditions: $t_n = n$ for $n = 0, 1, 2$
- $x^3 - 5x^2 + 8x - 4 = 0$
- $(x-1)(x-2)^2 \rightarrow$ roots are 1, 2, 2
- Solution is of form $c_1(1)^n + c_2 2^n + c_3 n 2^n$
 - If root is repeated third time, then $n^2 2^n$ term, and so on

Solving for Constants

- $t_n = c_1(1)^n + c_2 2^n + c_3 n 2^n$
- Use base cases to solve for constants
 - $t_0 = 0 = c_1(1)^0 + c_2 2^0 + c_3 0 2^0 = c_1 + c_2$
 - $t_1 = 1 = c_1(1)^1 + c_2 2^1 + c_3 1 2^1 = c_1 + 2c_2 + 2c_3$
 - $t_2 = 2 = c_1(1)^2 + c_2 2^2 + c_3 2 2^2 = c_1 + 4c_2 + 8c_3$
 - $c_1 = -2, c_2 = 2, c_3 = -1/2$
- $t_n = 2^{n+1} - n 2^{n-1} - 2$
- Test the solution on small values!

Inhomogeneous Equation

- $t_n - 2t_{n-1} = 3^n$
 - base case value for t_0 **only**
- $(x-2)(x-3) = 0$
 - $(x-2)$ term comes from homogeneous solution
 - If rhs is of form b^n poly(n) of degree d
 - In this case, $b = 3$, poly(n) = 1 is of degree 0
 - Plug $(x-b)^{d+1}$ into characteristic equation

Solving for constants

- $(x-2)(x-3) = 0$
- $t_n = c_1 2^n + c_2 3^n$
- Solve for c_1 and c_2 with only t_0 base case
- This is only 1 equation and 2 unknowns
- Use recurrence to generate extra equations
 - $t_n - 2t_{n-1} = 3^n \rightarrow t_1 = 2t_0 + 3$
- Now we have two equations
 - $t_0 = c_1 2^0 + c_2 3^0 = c_1 + c_2$
 - $t_1 = 2t_0 + 3 = 2c_1 + 2c_2 + 3 = 2c_1 + 3c_2$ results in $c_2 = 3$
 - $c_1 = t_0 - 3$ and $c_2 = 3$
- $t_n = (t_0 - 3)2^n + 3^{n+1}$

Changing variable

- $t_n - 3t_{n/2} = n$ if n is a power of 2
 - $t_1 = 1$
- Let $n = 2^i$ and $s_i = t_n$
- $s_i - 3s_{i-1} = 2^i$ for $i \geq 1$
 - $s_0 = 1$
- $(x-3)(x-2) = 0$
 - $x-3$ from characteristic equation
 - $x-2$ bⁿpoly(n) rhs

Solving for constants

- $(x-3)(x-2) = 0$
- $s_i = c_1 3^i + c_2 2^i$
- Generating two equations
 - $t_1 = s_0 = 1 = c_1 3^0 + c_2 2^0 = c_1 + c_2$
 - $t_2 = s_1 = 3t_1 + 2 = 5 = c_1 3^1 + c_2 2^1 = 3c_1 + 2c_2$
 - $c_1 = 3, c_2 = -2$
- $s_i = 3^{i+1} - 2^{i+1}$ for $i \geq 0$
- $t_n = 3n \lg 3 - 2n$ for n a power of 2 ≥ 1

Example: Towers of Hanoi

- Suppose we have two disks of each size.
- Let n be the number of sizes of disks
 - What is recurrence relation?
 - What is solution?

Example: Merge Sort

- Merge sort breaking array into 3 pieces
 - What is recurrence relation?
 - What is solution?
 - How does this compare to breaking into 2 pieces?