

CS 361, Lecture 4

Jared Saia
University of New Mexico

- Recurrence Relations

In-Class Soln 1

Let $f(n)$ be an always positive function and let $g(n) = f(n) \log n$. Show that $f(n) = o(g(n))$

- For any positive constant c , we want to show there is a $n_0 > 0$ such that $0 \leq f(n) < cg(n)$ for all $n \geq n_0$.
- In other words, we want to show that there is $n_0 > 0$ such that

$$0 \leq f(n) < cf(n) \log n$$

Dividing by $f(n) \log n$, we get:

$$0 \leq \frac{f(n)}{f(n) \log n} < c$$

- We know that $\lim_{n \rightarrow \infty} \frac{f(n)}{f(n) \log n} = \lim_{n \rightarrow \infty} \frac{1}{\log n} = 0$. Thus, for any constant c , there must be a n_0 such that the above inequality is satisfied for all $n \geq n_0$.

Today's Outline

In-Class Soln 2

Let $f(n)$ be an always positive function and let $g(n) = f(n) \log n$. Show that $f(n) = o(g(n))$

- For any positive constant c , we want to show there is a $n_0 > 0$ such that $0 \leq f(n) < cg(n)$ for all $n \geq n_0$.
- In other words, we want to show that there is $n_0 > 0$ such that

$$0 \leq f(n) < cf(n) \log n$$

Dividing by $f(n) \log n$, we get:

$$\begin{aligned} \frac{1}{\log n} &< c \\ (1/c) &< \log n \\ 2^{1/c} &< n \end{aligned}$$

- Thus, for any c , if we choose $n_0 > 2^{1/c}$, for all $n \geq n_0$, $f(n) < cg(n)$

Asymptotic Analysis - Take Away

- In studying behavior of algorithms, we'll be more concerned with *rate* of growth than with constants
- $O, \Theta, \Omega, o, \omega$ give us a way to talk about rates of growth
- Asymptotic analysis is an extremely useful way to compare run times of algorithms
- However, empirical analysis is also important (you'll be studying this in your project)

4

Alg: Binary Search

```
bool BinarySearch (int arr[], int s, int e, int key){
    if (e-s<=0) return false;
    int mid = (e-s)/2;
    if (arr[key]==arr[mid]){
        return true;
    }else if (key < arr[mid]){
        return BinarySearch (arr,s,mid,key);}
    else{
        return BinarySearch (arr,mid,e,key)}
}
```

6

Recurrence Relations

- Getting the run times of recursive algorithms can be challenging
- Recall our algorithm for binary search
- Let $T(n)$ be the run time of this algorithm on an array of size n
- Then we can write $T(1) = 1, T(n) = T(n/2) + 1$

5

What?

- $T(n)$ is a function giving the run time of Binary Search on an array of size n
- $T(n) = T(n/2) + 1$ is an example of a *recurrence relation*
- A *Recurrence Relation* is any equation for a function $T(n)$, where $T(n)$ appears on both the left and right sides of the equation.
- We always want to “solve” these recurrence relation by getting an equation for $T(n)$, where T appears on just the left side of the equation

7

Use of Recurrences

- We can use recurrence relations to analyze many properties of recursive algorithms e.g. run time, value returned, etc.
- To do this we need to: 1) write down the correct recurrence relation 2) solve the recurrence relation
- Step 1 is usually easier than step 2

8

A Side Note

- The running time of an algorithm on a constant size input is always $\Theta(1)$
- Thus for convenience, we frequently omit statements of the boundary conditions and just assume $T(n)$ is constant when n is a constant.
- Example: Instead of saying "If $n = 1$, $T(n) = \theta(1)$, and if $T(n) = 2 * T(n/2) + \Theta(n)$ ", we just say " $T(n) = 2 * T(n/2) + \Theta(n)$ "

9

Alg1

```
Alg1 (int n){
  if (n<=1) return 1;
  else
    return Alg1(n/2) + Alg1(n/2) + n;
}
```

10

Example1

- Let $T(n)$ be the run time of Alg1 on input n
- Then we can write $T(n) = 2T(n/2) + 1$
- Let $f(n)$ be the value returned by Alg1 on input n
- Then we can write $f(n) = 2f(n/2) + n$ and $f(1) = 1$

11

What now?

- To get the “real” run time or value returned, we need to solve the recurrence relation
- This means that no function appear on the right hand side
- We will review several techniques for solving recurrences including: the substitution method, recursion trees, the Master method, and annihilators

12

Substitution Method

- One way to solve recurrences is the substitution method aka “guess and check”
- What we do is make a good guess for the solution to $T(n)$, and then try to prove this is the solution by induction

13

Example

- Let’s guess that the solution to $T(n) = 2 * T(n/2) + n$ is $T(n) = O(n \log n)$
- We want to show that $T(n) \leq cn \log n$ for appropriate choice of constant c
- We can prove this by induction.

14

Proof

- Base Case: $T(2) \leq c * 2$
- Inductive Hypothesis: For all $j < n$, $T(j) \leq cj \log j$
- Inductive Step:

$$\begin{aligned} T(n) &= 2T(n/2) + n && (1) \\ &\leq 2(cn/2 \log(n/2)) + n && (2) \\ &\leq cn \log(n/2) + n && (3) \\ &= cn(\log n - \log 2) + n && (4) \\ &= cn \log n - cn + n && (5) \\ &\leq cn \log n && (6) \end{aligned}$$

The last step holds if $c \geq 1$

15

Recurrences and Induction

Recurrences and Induction are closely related:

- To *find* some solution to $f(n)$, solve a recurrence
- To *prove* that a solution for $f(n)$ is correct, use induction

For both recurrences and induction, we always solve a big problem by reducing it to smaller problems!

16

Some Examples

- The next several problems can be attacked by induction/recurrences
- For each problem, we'll need to reduce it to smaller problems
- Question: How can we reduce each problem to a smaller subproblem?

17

Sum Problem

- $f(n)$ is the sum of the integers $1, \dots, n$

18

Tree Problem

- $f(n)$ is the maximum number of leaf nodes in a binary tree of height n

Recall:

- In a binary tree, each node has at most two children
- A *leaf* node is a node with no children
- The height of a tree is the length of the longest path from the root to a leaf node.

19

Binary Search Problem

- $f(n)$ is the maximum number of queries that need to be made for binary search on a sorted array of size n .

20

Dominoes Problem

- $f(n)$ is the number of ways to tile a 2 by n rectangle with dominoes (a domino is a 2 by 1 rectangle)

21

Simpler Subproblems

- Sum Problem: What is the sum of all numbers between 1 and $n - 1$ (i.e. $f(n - 1)$)?
- Tree Problem: What is the maximum number of leaf nodes in a binary tree of height $n - 1$? (i.e. $f(n - 1)$)
- Binary Search Problem: What is the maximum number of queries that need to be made for binary search on a sorted array of size $n/2$? (i.e. $f(n/2)$)
- Dominoes problem: What is the number of ways to tile a 2 by $n - 1$ rectangle with dominoes? What is the number of ways to tile a 2 by $n - 2$ rectangle with dominoes? (i.e. $f(n - 1)$, $f(n - 2)$)

22

Recurrences

- Sum Problem: $f(n) = f(n - 1) + n$, $f(1) = 1$
- Tree Problem: $f(n) = 2 * f(n - 1)$, $f(0) = 1$
- Binary Search Problem: $f(n) = f(n/2) + 1$, $f(2) = 1$
- Dominoes problem: $f(n) = f(n - 1) + f(n - 2)$, $f(1) = 1$, $f(2) = 1$

23

Guesses

- Sum Problem: $f(n) = (n + 1)n/2$
- Tree Problem: $f(n) = 2^n$
- Binary Search Problem: $f(n) = \log n$
- Dominoes problem: $f(n) = \frac{1}{\sqrt{5}} \left(\frac{1+\sqrt{5}}{2} \right)^n - \frac{1}{\sqrt{5}} \left(\frac{1-\sqrt{5}}{2} \right)^n$

24

Inductive Proofs

“Trying is the first step to failure” - Homer Simpson

- Now that we’ve made these guesses, we can try using induction to prove they’re correct (the substitution method)
- We’ll give inductive proofs that these guesses are correct for the first three problems

25

Sum Problem

- Want to show that $f(n) = (n + 1)n/2$.
- Prove by induction on n
- Base case: $f(1) = 2 * 1/2 = 1$
- Inductive hypothesis: for all $j < n$, $f(j) = (j + 1)j/2$
- Inductive step:

$$f(n) = f(n - 1) + n \quad (7)$$

$$= n(n - 1)/2 + n \quad (8)$$

$$= (n + 1)n/2 \quad (9)$$

26

Tree Problem

- Want to show that $f(n) = 2^n$.
- Prove by induction on n
- Base case: $f(0) = 2^0 = 1$
- Inductive hypothesis: for all $j < n$, $f(j) = 2^j$
- Inductive step:

$$f(n) = 2 * f(n - 1) \quad (10)$$

$$= 2 * (2^{n-1}) \quad (11)$$

$$= 2^n \quad (12)$$

27

Binary Search Problem

- Want to show that $f(n) = \log n$. (assume n is a power of 2)
- Prove by induction on n
- Base case: $f(2) = \log 2 = 1$
- Inductive hypothesis: for all $j < n$, $f(j) = \log j$
- Inductive step:

$$f(n) = f(n/2) + 1 \quad (13)$$

$$= \log n/2 + 1 \quad (14)$$

$$= \log n - \log 2 + 1 \quad (15)$$

$$= \log n \quad (16)$$

28

In Class Exercise

- Consider the recurrence $f(n) = 2f(n/2) + 1$, $f(1) = 1$
- Guess that $f(n) \leq cn - 1$:
- Q1: Show the base case - for what values of c does it hold?
- Q2: What is the inductive hypothesis?
- Q3: Show the inductive step.

29

Todo

- Read Chapter 4 (Recurrences) in text

30