Mat 2345 — Discrete Math

Week 12

Fall 2013

Student Responsibilities — Week 12

- ▶ Reading: Textbook, Section 4.4 & 4.5
- ► Assignments:

```
Sec 4.4 8, 10, 24, 28
Sec 4.5 2, 4, 7, 12
```

► Attendance: Frostily Encouraged

Week 12 Overview

- ► Sec 4.4. Recursive Algorithms
- ► Sec 4.5. Program Correctness

Section 4.4 Recursive Algorithms

A recursive procedure to find the max in a non-empty list.

We will assume we have built-in functions:

- ▶ Length() which returns the number of elements in the list
- ► Max() which returns the larger of two values
- ▶ Listhead() which returns the first element in a list

Note: Max() requires one comparison

What happens with the list $\{29\}$? With the list $\{3, 8, 5\}$?

How Many Comparisons?

The recurrence equation for the number of comparisons required for a list of length n, C(n) is:

```
C(1)=0 the initial condition C(n)=1+C(n-1) the recurrence equation So, C(n)\in O(n) as we would expect
```

A Variant of Maxlist()

Assuming the list length is a power of 2, here is a variant of Maxlist() using a Divide—and—Conquer approach.

- ▶ Divide the list in half, and find the maximum of each half
- ▶ Find the Max() of the maximum of the two halves
- ▶ Apply these steps to each list half recursively.
- ► What could the base case(s) be?

Maxlist2() Algorithm

```
procedure Maxlist2(...list...){
// PRE: list is not empty
// POST: returns the largest element in list
// Divide list into two lists, take the max of
// the two halves (recursively)

if Length(list) is 1 then
    return Listhead(list)

else
    a = Maxlist2(first half of list)
    b = Maxlist2(second half of list)
    return Max(a, b)
}
What happens with the list {29,7}?
```

How Many Comparisons in Maxlist2()?

- ▶ There are two calls to Maxlist2(), each of which requires $C(\frac{n}{2})$ operations to find maximum.
- ► One comparison is required by the Max() function

The **recurrence equation** for the number of comparisons required for a list of length n, C(n), is:

$$C(1) = 0$$
 the initial condition $c(n) = 2C(\frac{n}{2}) + 1$ the recurrence equation

Consider A Sampling

With the list $\{3, 8, 5, 7\}$?

n	$C(n)=2C(\frac{n}{2})+1$	
$2^0 = 1$	1	$= 2^1 - 1$
$2^1 = 2$	3	$= 2^2 - 1$
$2^2 = 4$	7	$= 2^3 - 1$
$2^3 = 8$	15	$= 2^4 - 1$
$2^4 = 16$	31	$= 2^5 - 1$
÷		:
$2^{\log n} = n$	$2^{\log(n)+1}-1$	$= 2n-1 \in O(n)$

Thus,
$$C(n) = 2^{\log(n)+1} - 1 \in O(n)$$

Practice I: **Prove** $3n^2 + 5n + 4 \in O(n^2)$

Definition of Big–Oh: $f(n) \in O(g(n))$ if there exists positive constants c and N_0 such that $\forall n \geq N_0$ we have $f(n) \leq cg(n)$

We need to find c > 0 and $N_0 > 0$ such that:

$$3n^2 + 5n + 4 \leq cn^2 \quad \forall n \geq N_0$$

We note that

Practice III.

$$3n^2 + 5n + 4 \le 3n^2 + 5n^2 + 4n^2$$
, when $n > 0$
 $\le 12n^2$

and we can choose c = 12

Practice I, Cont.

To find
$$N_0$$
: $3n^2 + 5n + 4 = 12n^2$
 $0 = 9n^2 - 5n - 4$

when
$$n = 1$$
, $9(1)^2 - 5(1) - 4 = 9 - 5 - 4 = 0$

Thus,
$$3n^2 + 5n + 4 \le 12n^2 \quad \forall n \ge 1$$
, and therefore, $3n^2 + 5n + 4 \in O(n^2)$

Practice II. Given
$$T(n) = 2n - 1$$
, prove that $T(n) \in O(n)$

Prove that
$$T(n)=3n+2$$
 if
$$T(n)=\left\{\begin{array}{ll}2&n=0\\3+T(n-1)&n>0\end{array}\right.$$

MergeSort Algorithm

```
list MergeSort(list[1..n]){
// PRE: none
// POST: returns list[1..n] in sorted order
// Functional dependency: Merge()

if n is 0
    return an empty list
else if n is 1
    return list[1]
else {
    list A = MergeSort(list[1..n/2])
    list B = MergeSort(list[n/2 + 1..n])
    list C = Merge(A, B)
    return C
}
```

Time Complexity of MergeSort()

Prove by induction that the time complexity of MergeSort(), $T(n) \in O(n \log n)$

What we need to do:

- ightharpoonup Establish a Base Case for some small n
- ▶ Prove $T(k) \le cf(k) \to T(2k) \le cf(2k)$

In particular, we need to prove $\forall k \geq N_0$ that:

$$T(k) \le ck \log k$$
 \rightarrow $T(2k)$ $\le c2k \log(2k)$
= $c2k(\log 2 + \log k)$
= $c2k \log k + c2k$

where $k = 2^m$ for some $m \ge 0$, $wlog^*$

*without loss of generality

Base Case

```
Let n=1. n\log n=(1)\log(1)=1(0)=0 But, T(n) is always positive, so this is not a good base case. Try a larger number. Let n=2. T(2)= Time to divide + time to MergeSort halves + time to Merge =1+1+1+2=5 while n\log n=2\log 2=2(1)=2 Can we find a constant c>0 such that 5\leq 2c? \frac{5}{2}\leq c, so \frac{5}{2} is a lower bound on c
```

Inductive Hypothesis

Assume for some arbitrary $k \geq 2$ that $T(k) \leq ck \log k$

Inductive Step — Show $T(2k) \le 2ck \log k + 2ck$

$$T(2k) \leq 1 + T(\lceil \frac{2k}{2} \rceil) + T(\lfloor \frac{2k}{2} \rfloor) + 2k$$

$$\leq T(k) + T(k) + 2k + 1$$

$$\leq 2T(k) + 2k + 1$$

$$\leq 2(ck \log k) + 2k + 1$$

$$\leq 2ck \log k + 2k + 1$$

Inductive Step, Cont.

Now, can we find a c such that

$$2ck \log k + 2k + 1 \leq 2ck \log k + 2ck$$
$$2k + 1 \leq 2ck$$
$$1 \leq 2ck - 2k$$
$$1 \leq 2k(c - 1)$$

Since $k \ge 2$ from base case, $(c-1) \ge \frac{1}{4}$ or $c \ge \frac{5}{4}$ We had a lower bound of $\frac{5}{2}$, so we can choose c = 3.

Thus, $T(n) \in O(n \log n) \quad \forall n \geq 2$.

More on Complexity

- If an algorithm is composed of several parts, then its time complexity is the sum of the complexities of its parts.
- ▶ We must be able to evaluate these summations.
- Things become even more complicated when the algorithm contains loops, each iteration of which is a different complexity.

An Example: Suppose $S_n = \sum_{i=1}^n i^2$

- ▶ We saw $\sum_{i=1}^{n} i = \frac{n(n+1)}{2} = \frac{(n^2+n)}{2} \leq n^2$, and is, in fact, $\Theta(n^2)$
- ▶ So, we guess $\sum_{i=1}^{n} i^2 \leq \sum_{i=1}^{n} n^2 = n^3$. Maybe $S_n \in \Theta(n^3)$.
- ▶ We can prove our guess correct, and find the minimum constant of difference between S_n and n^3 by induction:
- Guess: $\sum_{i=1}^{n} i^2 = an^3 + bn^2 + cn + d = P(n)$
- ▶ Notice that $\sum_{i=1}^{n+1} i^2 \sum_{i=1}^{n} i^2 = (n+1)^2$

So,
$$P(n+1) = P(n) + (n+1)^2$$

Thus,

$$a(n+1)^3 + b(n+1)^2 + c(n+1) + d =$$

 $an^3 + bn^2 + cn + d + (n+1)^2$

$$a(n^3 + 3n^2 + 3n + 1) + b(n^2 + 2n + 1) + cn + c + d =$$

$$an^3 + bn^2 + cn + d + n^2 + 2n + 1$$

$$an^3 + 3an^2 + 3an + a + bn^2 + 2bn + b + cn + c + d =$$

 $an^3 + bn^2 + cn + d + n^2 + 2n + 1$

Hence:

$$3an^2 + 3an + a + 2bn + b + c = n^2 + 2n + 1$$
, or $3an^2 + (3a + 2b)n + (a + b + c) = n^2 + 2n + 1$

Since coefficients of the same power of n must be equal:

And we can choose d = 0

Hanca

$$P(n) = \frac{1}{3}(n^3) + \frac{1}{2}(n^2) + \frac{1}{6}(n)$$

$$= \frac{2}{6}(n^3) + \frac{3}{6}(n^2) + \frac{1}{6}(n)$$

$$= \frac{(2n^3 + 3n^2 + n)}{6}$$

$$= \frac{n(2n^2 + 3n + 1)}{6}$$

$$= \frac{n(n+1)(2n+1)}{6}$$

Now, we wish to prove $S_n \in \Theta(n^3)$, or, that S_n is a third degree polynomial, by induction.

Base Case

Let
$$n = 1$$

Ihs:
$$\sum_{i=1}^{1} i^2 = 1^2 = 1$$

rhs: $P(1) = \frac{1(1+1)(2(1)+1)}{6} = \frac{1(2)(3)}{6} = \frac{6}{6} = 1 \quad \checkmark$

Inductive Hypothesis

Assume for some arbitrary $k \ge 1$, that $S_k = P(k)$

That is,
$$\sum_{i=1}^{k} i^2 = \frac{k(k+1)(2k+1)}{6}$$

Inductive Step

Show
$$S_{k+1} = P(k+1) = \frac{(k+1)(k+2)(2k+3)}{6}$$

Ihs $= S_{k+1} = S_k + (k+1)^2$ defn of \sum $= \frac{k(k+1)(2k+1)}{6} + (k+1)^2$ IH & subst. $= \frac{k(k+1)(2k+1) + 6(k+1)^2}{6}$ Alg. Man. $= \frac{(k+1)[k(2k+1) + 6(k+1)]}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man. $= \frac{(k+1)(2k^2 + k + 6k + 6)}{6}$ Alg. Man.

Thus, $S_n = P(n) \ \forall \ n \geq 1$ Hence, $S_n \in \Theta(n^3)$

Section 4.5 — Program Correctness

- A brief introduction to the area of program verification, tying together the rules of logic, proof techniques, and the concept of an algorithm.
- ▶ **Program verification** means to prove the correctness of the program.
- ▶ Why is this important? Why can't we merely run testcases?
- ► A program is said to be **correct** if it produces the correct output for every possible input.

Correctness Proof

A correctness proof for a program consists of two parts:

- Establish the partial correctness of the program.
 If the program terminates, then it halts with the correct answer.
- 2. Show that the program always terminates.

Proving Output Correct

We need two propositions to determine what is meant by $\boldsymbol{produce}$ \boldsymbol{the} $\boldsymbol{correct}$ $\boldsymbol{output}.$

- 1. Initial Assertion: the properties the input values must have. ($\mbox{\bf p}$)
- 2. Final Assertion: the properties the output of the program should have if the program did what was intended. (q)

A program segment S is said to be partially correct with respect to p and q, $[p \{S\} q]$, if — whenever p is TRUE for the input values of S and S terminates, — then q is TRUE for the output values of S.

Example

Is
$$[p \{S\} q]$$
 TRUE?

Composition Rule: $[p\ \{S_1\}\ q]$ and $[q\ \{S_2\}\ r] \to [p\ \{S_1;S_2\}\ r]$

Rules of Inference: Conditional Statements

IF condition THEN block

 ${\tt BLOCK}$ is executed when condition is ${\tt TRUE},$ and it is not executed when condition is ${\tt FALSE}.$

To verify correctness with respect to p and q, we must show:

- 1. When p is TRUE and condition is also TRUE, then q is TRUE after BLOCK terminates.
- 2. When p is TRUE and condition is FALSE, q is TRUE (since BLOCK does not execute.

This leads to the following rule of inference:

Example

Is $[p \{S\} q]$ TRUE?

IF...THEN...ELSE Statements

IF condition THEN block1 ELSE block2

If condition is TRUE, then block1 executes; if condition is FALSE, then block2 executes.

To verify correctness with respect to p and q, we must show:

- 1. When p is TRUE and condition is also TRUE, then q is TRUE after BLOCK1 terminates.
- 2. When p is TRUE and condition is FALSE, q is TRUE after BLOCK2 terminates.

This leads to the following rule of inference:

Example

I.e., is the segment correct?

Loop Invariants — While Loops

WHILE condition block

Where $\ensuremath{\mathtt{BLOCK}}$ is repeatedly executed until condition becomes $\ensuremath{\mathtt{FALSE}}.$

Loop Invariant: an assertion that remains \mathtt{TRUE} each time \mathtt{BLOCK} is executed.

I.e., p is a loop invariant if $(p \land condition)\{block\}p$ is TRUE

Let p be a loop invariant.

If p is TRUE before Segment S is executed, then p and \neg condition are TRUE after the loop terminates (if it does).

Hence: $(p \land condition)\{S\}p$

 \therefore $p\{while\ condition\ S\}(\neg condition\ \land\ p)$

Example

We wish to verify the following code segment terminates with factorial = n! when n is a positive integer.

Our loop invariant p is: factorial = i! and $i \le n$

```
i = 1
factorial = 1
while i < n {
   i = i + 1
   factorial = factorial * i
}</pre>
```

[Base Case] p is TRUE before we enter the loop since factorial = 1 = 1!, and $1 \le n$.

[Inductive Hypothesis] Assume for some arbitrary $k \ge 1$ that p is TRUE. Thus i < k (so we enter the loop again), and factorial= (i-1)!.

[Inductive Step] Show p is still TRUE after execution of the loop. Thus $i \leq k$ and factorial = i!.

First, i is incremented by 1

Thus $i \leq k$ since we assumed i < k, and i and $k \geq 1$.

Also, factorial, which was (i-1)! by IH, is set to (i-1)!*i=i!

Hence, p remains true.

Since p remains $\mbox{TRUE}, \ p$ is a loop invariant and thus the assertion:

$$[p \land (i < n)]{S}p$$
 is TRUE

It follows that the assertion:

 $p\{\textit{while } i < n \ S\}[(i \ge n) \land p] \text{ is also true.}$

Furthermore, the loop terminates after n-1 iterations with i=n, since:

- $1. \ i$ is assigned the value 1 at the beginning of the program,
- 2. 1 is added to i during each iteration of the loop, and
- 3. the loop terminates when $i \ge n$

Thus, at termination, factorial = n!.

We split larger segments of code into component parts, and use the rule of composition to build the correctness proof.

$$(p = p_1)\{S_1\}q_1, q_1\{S_2\}q_2, \dots, q_{n-1}\{S_n\}(q_n = q) \rightarrow p\{S_1; S_2; \dots; S_n\}q$$