

# CS 180 - Lecture 7

Note Title

9/9/2009

## Announcements

- HW due tomorrow
- Lab 2 tomorrow
- programming project #1 - out tonight
  - probably due in  $1\frac{1}{2}$  - 2 weeks
  - pair project

Ch 3 - how to analyze the running time?  
So how?

time it

- disadvantage: computer matters

OS

language  
input set matters

## Counting primitive operations

Identik/ high-level primitive operations  
independent of language compiler, OS, or  
computers

Ex: variable assignment / update

comparisons

branching  
add

subtract

multiplication

⋮

Ex: (pseudocode to find max in an array)  
↳ write easily readable, language independent code

Algorithm arrayMax(A, n):

Input: An array A of  $n \geq 1$  numbers  
Output: The maximum element of A

currentMax  $\leftarrow$  A[0]

for  $i \leftarrow 1$  to  $n-1$

if currentMax < A[i] then  
currentMax  $\leftarrow$  A[i]

return currentMax

Advantage of pseudocode:

- independent of language
- easy to read + translate to any language

Ex: (in C++)

```
int arrayMax(int A[], int n) {  
    int currentMax = A[0];  
    for (int i = 1; i < n; i++)  
        if (currentMax < A[i])  
            currentMax = A[i];  
    return currentMax;  
}
```

## Counting operations:

Algorithm  $\text{arrayMax}(A, n)$ :

Input: An array  $A$  of  $n \geq 1$  numbers

Output: The maximum element of  $A$

```
1 currentMax ← A[0] ← 3
2 for i ← 1 to n-1
3   if currentMax < A[i] then
4     currentMax ← A[i] ← between 0 and (n-1) operations
5 return currentMax 1
```

$$\text{min: } 3 + 2 + n - 1 + n - 1 + 0 + 1 = 2n + 4$$

$$\text{max: } 3 + 2 + n - 1 + n - 1 + 2(n - 1) + 1 = 4n + 2$$

So how many operations in, best (or worst) case?

$2n+4$  to  $4n+2$   
(see previous)

Average case versus worst case

→ Look at all possible inputs + average

→ analyze time for worst possible input

average:  $3n+3$

worst case:  $4n+2$



# Asymptotic Notation

How important is exact number of computations?

In general, any primitive statement depends on a small number of low-level operations, independent of language or computer.

So we'll focus on big-picture, or how the running time grows in proportion to input size (usually  $n$ ).

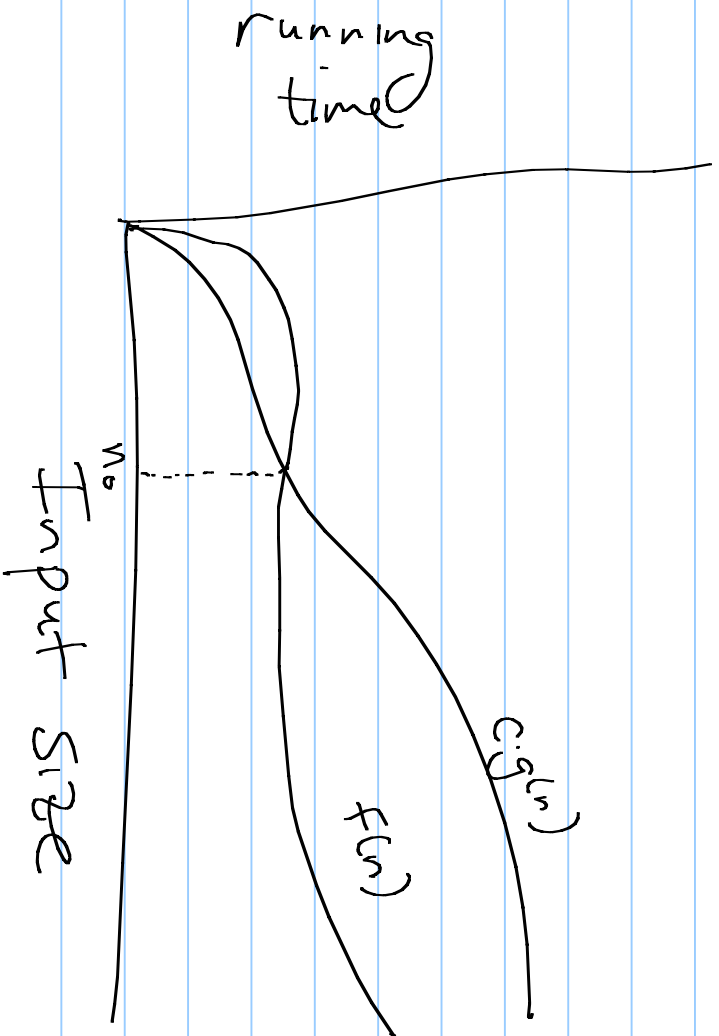
Formalize: Big-Oh notation

Let  $f(n)$  and  $g(n)$  be two functions from non-negative integers to reals.

We say  $f(n)$  is  $O(g(n))$  if there exists a constant  $c$  and integer  $n_0 > 0$  such that  $f(n) \leq c \cdot g(n)$  for all  $n \geq n_0$ .

$f(n)$  is big-Oh of  $g(n)$

Picture



Ex:  $\Theta(n+2)$  is  $O(n)$ .

Why? Find  $c$  at  $n_0$ :

Let  $c = 5$ . Want  $\Theta(n+2) < 5n$

Let  $n_0 = 2$

If  $n > n_0$ , then  $\Theta(n+2) < 5n$

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Let  $c = 100$  and  $n_0 = 100$

$\Theta(n+2) < 100n$  if  $n > 100$

Ex: running time of arrayMax is  $O(n)$ :

Algorithm arrayMax( $A, n$ ):

Input: An array  $A$  of  $n \geq 1$  numbers  
Output: The maximum element of  $A$

```
currentMax  $\leftarrow A[0]$ 
for  $i \leftarrow 1$  to  $n-1$ 
  if currentMax  $< A[i]$  then
    currentMax  $\leftarrow A[i]$ 
return currentMax
```

Why? Just showed worst case running time is  $O(n)$ .

$$\underline{\text{Ex:}} \quad \underline{20n^3 + 10n \log n + 5}$$

$$\underline{O(n^3)}$$

Why?

$$\underline{C = 35} \quad \forall n_0 = 2$$

$$20n^3 + 10n \log n + 5 < 35n^3$$

Any polynomial:  $a_k n^k + a_{k-1} n^{k-1} + \dots + a_0$

$$O(n^k)$$

$$25n^6 + \dots$$

$$O(n^6)$$

$$\underline{\text{Ex:}} \quad 2^{100} = O(1)$$

$$\text{Let } c = 2^{100} + 1 \quad n_0 = 1$$

$$2^{100} \leq 1 \cdot 2^{100}$$



P. 126 in book: Rules

Examples:

- If  $d(n)$  is  $O(f(n))$  and  $f(n)$  is  $O(g(n))$ ,  
then  $d(n)$  is  $O(g(n))$ .

-  $\log n^c$  is  $O(\log n)$  for any constant  $c > 0$ .

$c \cdot \log n$

etc.

Useful things to remember:

$$\bullet \sum_{i=a}^b f(i) = f(a) + f(a+1) + \dots + f(b)$$

(loops often produce these!)

• For any  $n \geq 1$  and  $0 < a \neq 1$ :

$$\sum_{i=0}^n a^i = 1 + a + \dots + a^n = \frac{1 - a^{n+1}}{1 - a}$$

and if  $a < 1$ , then

$$\sum_{i=0}^{\infty} a^i = \frac{1}{1-a}$$

Another useful thing:

for any  $n \geq 1$ ,

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

When might this come in handy?

2 nested loops

for  $i \leftarrow 1$  to  $n$

for  $j \leftarrow 1$  to  $n$

## Logarithms (see p. 115)

$$- \log_b(ac) = \log_b a + \log_b c$$

$$- \log_b(a/c) = \log_b a - \log_b c$$

$$- \log_b(a^c) = c \cdot \log_b(a)$$

$$- \log_b b^a = a$$

$$- b^{\log_c a} = a^{\log_c b}$$

$$-(b^a)^c = b^{ac}$$

$$- b^a b^c = b^{a+c}$$

etc ...