

Optimization & Algorithm Analysis (Ch 1.2, 10.4)

Concerned with:

- space utilization (eg memory)
- time efficiency

Usually optimize for just one of the two

- usually a tradeoff between space & time
- fast alg uses more mem for example

Not as critical for smaller programs, but many tasks where still important

- real-time devices
- games
- large-scale data crunching

Time efficiency

how long it takes to do a task

affected by several of the following:

- amount of data
- nature of data
- algorithm

use approximate measurement instead of seconds

seconds can be affected by hardware, OS, system load, etc

need a standardized measure that is machine independent

approximate count of instructions in the algorithm

can't use count of machine instructions (compiled code), not machine independent

Counting algorithm instructions

must look at each line of code and see how many steps it takes

basic code (eg assignment, math) is seen as taking 1 step

loops more complex

have to figure how many times the loop executes

Ex: `for(i=0; i<n; i++)`

goes from 0 to n-1, that's n times

plus 1 more step to check the stopping condition

then have to consider how often each stmt in the loop body executes

Ex: `for(i=0; i<n; i++) { sum += i; }`

from above there are n loops

so loop body executes n times

how long for each loop body?

look at each loop stmt

have one basic code stmt: add i to sum

since basic, takes 1 step

so 1 step for each loop body

now multiply steps in a single loop body by number of loops

Ex: 1 step for each loop body, n loops total

$$n * 1 = n$$

so n steps for loop body plus 1 step to check stopping condition

this loop has n+1 steps

Ex: Two methods to find sum in Ch1.2

Algorithm 1

1. Have user input value for n
2. Set sum = 0
3. For each i in the range 1 to n
 - a. add i to sum
4. Return sum

Counting steps

1. Basic task so 1 step
 2. Basic task, 1 step
 3. Loop, same as above example, n+1 steps
 4. Basic task, 1 step
- Total: n+4 steps

Algorithm 2

1. Have user input value for n
2. Return $(n * (n + 1)) / 2$

Counting steps

1. Basic task, 1 step
 2. Basic task, 1 step
- Total: 2 steps

recursive functions are counted similarly to loops

find out how many times it calls itself and how long

recursive case body takes

add in number of steps for stopping case

time can also be affected by structure of data

this is one way data structures differ

for example, fast to print a sorted list if data is stored

sorted

have to consider following for each alg:

best case

worst case

average case

often the worst case is what most look at

sets an upper bound on the performance

can't get any worse

Big O notation

just a formal way to express counted steps

takes the order of magnitude

ignores constants

$3n$ is order of magnitude n

3 is order of magnitude 1 (constant)

takes the largest factor

eg in n+1, n is greater factor than 1, so considered

order of magnitude n

Above examples

Alg1 is n +4 steps, so that's order of magnitude n

Alg2 is 2 steps, so that's order of magnitude 1

Alg1 is $O(n)$

Alg2 is $O(1)$

Alg2 is $O(1)$

Common notations

$O(1)$ is constant

$O(n)$ is linear

$O(n^2)$ is quadratic

$O(n^3)$ is cubic

$O(2^n)$ is exponential

Figure 10.6 on page 558 shows how these graph out for various values of n

Can be used to express space utilization too

Space utilization

how much memory data is stored in

again, must have a standardized method for comparison

can't just see how much mem the program used

also use a count instead of size (ie n units)

count says what it is a count of

Ex: you are storing student records

you'd say the storage is $O(n)$

where n is the number of student records

computing space utilization for an entire program can be tedious

course will focus on memory utilization for each data

structure

Know the Problem

no one best alg or data structure

have to analyze the problem

come up with the best solution for the problem at hand

don't forget about the context of the problem

what sort of system is it geared for?

a PDA will have different ideas of "optimal" than a gaming PC

what are the target users tolerances

some people want speed

some people want low "footprint"

remember that big-O is approximate

ignores constants & smaller factors that may be a big part of

performance

particularly w/ small data sets

many compilers can optimize code

differences between similar algs may be minimized