

What Are the Questions?

Principal concern throughout engineering:

"There is someone who can do for a dime what any fool can do for a dollar."

Can

Actual cost (for programs, time to run, space requirements).

Development costs: How much engineering time? When delivered?

Upgrade costs: Upgrades, bug fixes.

Failure: How robust? How safe?

Can it be **fast enough**? Depends on:

Purpose;

Input data.

Space (memory, disk space)?

Depends on what input data.

Scale, as input gets big?

7:15 2019

CS61B: Lecture #16 2

Cost Measures (Time)

Real execution time

Do this at home:

`java FindPrimes 1000`

Pros: easy to measure, meaning is obvious.

Cons: time where time is critical (real-time systems, e.g.).

Usage: applies only to specific data set, compiler, machine.

Statement counts of # of times statements are executed:

Pros: more general (not sensitive to speed of machine).

Usage: doesn't tell you actual time, still applies only to data sets.

Execution times:

Formulas for execution times as functions of input size.

Pros: applies to all inputs, makes scaling clear.

Usage: practical formula must be approximate, may tell about actual time.

7:15 2019

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Handy Tool: Order Notation

Try to produce specific functions that specify size, but **ignore** *features of functions with similarly behaved magnitudes*.

Something like " f is bounded by g if it is in g 's family."

For a function $g(x)$, the functions $2g(x)$, $0.5g(x)$, or for any $K > 0$, have the same "shape". So put all of them into g 's family.

Find a $h(x)$ such that $h(x) = K \cdot g(x)$ for $x > M$ (for some M). h has g 's shape "except for small values." So put all of them into g 's family.

For limits, throw in all functions whose absolute value is eventually less than some member of g 's family. Call this set $O(g)$ or $O(g(n))$.

For limits, throw in all functions whose absolute values is eventually greater than some member of g 's family. Call this set $\Omega(g)$.

The $\Theta(g) = O(g) \cap \Omega(g)$ —the set of functions **bracketed** by two members of g 's family.

7:15 2019

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CS61B Lecture #16: Complexity

Enlightening Example

Search a text corpus (say 10^8 bytes or so), and find and print frequently used words, together with counts of how often

Usage: Heavy-Duty data structures

Cons: implementation, randomized placement, pointers galore, pages long.

BDoug McIlroy): UNIX shell script:

```
'[:alpha:]' ' [\n*]' < FILE | \
\
tr -k 1,1 | \
```

After?

Can be faster,

took 5 minutes to write and processes 100MB in ≈ 50 sec.

In many cases, almost anything will do: **Keep It Simple**.

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Asymptotic Cost

Execution time lets us see **shape** of the cost function.

Even approximating anyway, pointless to be precise about constants:

Focus on small inputs:

Always pre-calculate some results.

For small inputs not usually important.

More interested in **asymptotic behavior** as input size is very large.

Factors (as in "off by factor of 2"):

Changing machines causes constant-factor change.

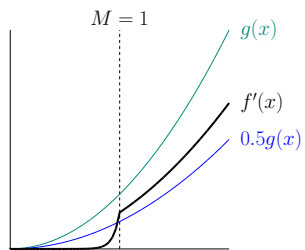
Don't get distracted away from (i.e., ignore) these things?

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Big Omega

by bounding from below:



$\geq \frac{1}{2}g(x)$ as long as $x > 1$,

g 's "bounded-below family," written

$$f'(x) \in \Omega(g(x)),$$

though $f(x) < g(x)$ everywhere.

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de: Various Mathematical Pedantry

if I am going to talk about $O(\cdot)$, $\Omega(\cdot)$ and $\Theta(\cdot)$ as sets of functions, I really should write, for example,

$$f \in O(g) \quad \text{instead of} \quad f(x) \in O(g(x))$$

$f(x) \in O(g(x))$ is short for $\lambda x. f(x) \in O(\lambda x. g(x))$.

And notation outside this course, in fact, is $f(x) = O(g(x))$, which, I think that's a serious abuse of notation.

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Why It Matters

Scientists often talk as if constant factors didn't matter, but the difference of $\Theta(N)$ vs. $\Theta(N^2)$.

They do matter, but at some point, constants always get

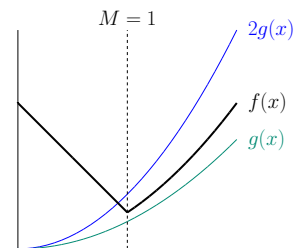
\sqrt{n}	n	$n \lg n$	n^2	n^3	2^n
1.4	2	2	4	8	4
2	4	8	16	64	16
2.8	8	24	64	512	256
4	16	64	256	4,096	65,536
5.7	32	160	1024	32,768	4.2×10^9
8	64	384	4,096	262,144	1.8×10^{19}
11	128	896	16,384	2.1×10^9	3.4×10^{38}
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
32	1,024	10,240	1.0×10^6	1.1×10^9	1.8×10^{308}
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
1024	1.0×10^6	2.1×10^7	1.1×10^{12}	1.2×10^{18}	$6.7 \times 10^{315,652}$

7:15 2019

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Big Oh

by bounding from above.



$\leq 2g(x)$ as long as $x > 1$,

g 's "bounded-above family," written

$$f(x) \in O(g(x)),$$

though (in this case) $f(x) > g(x)$ everywhere.

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Big Theta

From the previous slides, we not only have $f(x) \in O(g(x))$ and $f(x) \in \Omega(g(x))$,...

$f(x) \in \Omega(g(x))$ and $f'(x) \in O(g(x))$.

Summarize this all by saying $f(x) \in \Theta(g(x))$ and $f'(x) \in \Theta(g(x))$.

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How We Use Order Notation

In mathematics, you'll see $O(\dots)$, etc., used generally to denote bounds on functions.

$$\pi(N) = \Theta\left(\frac{N}{\ln N}\right)$$

and we prefer to write

$$\pi(N) \in \Theta\left(\frac{N}{\ln N}\right)$$

where $\pi(N)$ is the number of primes less than or equal to N .)

We see things like

$$= x^3 + x^2 + O(x) \quad (\text{or } f(x) \in x^4 + x^2 + O(x)),$$

$$f(x) = x^3 + x^2 + g(x) \text{ where } g(x) \in O(x).$$

In computer science, the functions we will be bounding will be *cost functions* that measure the amount of execution time or the space required by a program or algorithm.

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Using the Notation

order notation for any kind of real-valued function.
them to describe cost functions. Example:
position of X in list L, or -1 if not found. */
List L, Object X) {
= 0; L != null; L = L.next, c += 1)
(X.equals(L.head)) return c;
-1;
representative operation: number of .equals tests.
th of L, then loop does at most N tests: worst-case
sts.
al # of instructions executed is roughly proportional
worst case, so can also say worst-case time is O(N),
f units used to measure.
provision (in defn. of O(·)) to ignore empty list.

Intuition on Meaning of Growth

problem can you solve in a given time?
giving table, left column shows time in microseconds to
problem as a function of problem size N.
y the size of problem that can be solved in a second,
(31 days), and century, for various relationships be-
quired and problem size.

c) for ze N	1 second	Max N Possible in 1 hour	1 month	1 century
	10 ³⁰⁰⁰⁰⁰	10 ¹⁰⁰⁰⁰⁰⁰⁰⁰⁰	10 ^{8·10¹¹}	10 ^{10¹⁴}
	10 ⁶	3.6 · 10 ⁹	2.7 · 10 ¹²	3.2 · 10 ¹⁵
	63000	1.3 · 10 ⁸	7.4 · 10 ¹⁰	6.9 · 10 ¹³
	1000	60000	1.6 · 10 ⁶	5.6 · 10 ⁷
	100	1500	14000	150000
	20	32	41	51

Effect of Nested Loops

s often lead to polynomial bounds:
i = 0; i < A.length; i += 1)
nt j = 0; j < A.length; j += 1)
(i != j && A[i] == A[j])
return true;
else;
is O(N²), where N = A.length. Worst-case time is
icient though:
i = 0; i < A.length; i += 1)
nt j = i+1; j < A.length; j += 1)
(A[i] == A[j]) return true;
else;
ase time is proportional to
- 1 + N - 2 + ... + 1 = N(N - 1)/2 ∈ Θ(N²)
ic time unchanged by the constant factor).

Be Careful

e that the worst-case time is O(N²), since N ∈ O(N²)
bounds are loose.
ase time is Ω(N), since N ∈ Ω(N), but that does not
le loop always takes time N, or even K · N for some K.
are just saying something about the function that maps
argest possible time required to process any array of
ch as possible about our worst-case time, we should try
ound: in this case, we can: Θ(N).
hat still tells us nothing about best-case time, which
n we find X at the beginning of the loop. Best-case time

Binary Search: Slow Growth

is an element of S[L .. U]. Assumes
ding order, 0 <= L <= U-1 < S.length. */
String X, String[] S, int L, int U) {
return false;
)/2;
+ X.compareTo(S[M]);
< 0) return isIn(X, S, L, M-1);
rect > 0) return isIn(X, S, M+1, U);
true;
case time, C(D), (as measured by # of calls to .compareTo),
ize D = U - L + 1.
e S[M] from consideration each time and look at half the
e D = 2^k - 1 for simplicity, so:
$$C(D) = \begin{cases} 0, & \text{if } D \leq 0, \\ 1 + C((D-1)/2), & \text{if } D > 0. \end{cases}$$
$$= \underbrace{1 + 1 + \dots + 1}_k + 0$$
$$= k = \lg(D+1) \in \Theta(\lg D)$$

rsion and Recurrences: Fast Growth

e of recursion. In the worst case, both recursive calls
ff X is a substring of S */
curs(String S, String X) {
uals(X)) return true;
ngth() <= X.length()) return false;

(S.substring(1), X) ||
(S.substring(0, S.length()-1), X);
to be the worst-case cost of occurs(S,X) for S of
f fixed size N₀, measured in # of calls to occurs. Then
$$C(N) = \begin{cases} 1, & \text{if } N \leq N_0, \\ 2C(N-1) + 1 & \text{if } N > N_0 \end{cases}$$

ws exponentially:
N - 1) + 1 = 2(2C(N - 2) + 1) + 1 = ... = 2(⋯2·1 + 1) + ... + 1
N-N₀
N₀ + 2^{N-N₀-1} + 2^{N-N₀-2} + ... + 1 = 2^{N-N₀+1} - 1 ∈ Θ(2^N)

Other Typical Pattern: Merge Sort

```
sort(L) {  
  if (|L| < 2) return L;  
  L0 and L1 of about equal size;  
  sort(L0); L1 = sort(L1);  
  merge of L0 and L1  
}
```

Merge ("combine into a single ordered list") takes time proportional to size of its result.

At size of L is $N = 2^k$, worst-case cost function, $C(N)$,
merge time (which is proportional to # items merged):

$$\begin{aligned} C(N) &= \begin{cases} 0, & \text{if } N < 2; \\ 2C(N/2) + N, & \text{if } N \geq 2. \end{cases} \\ &= 2(2C(N/4) + N/2) + N \\ &= 4C(N/4) + N + N \\ &= 8C(N/8) + N + N + N \\ &= N \cdot 0 + \underbrace{N + N + \dots + N}_{k=\lg N} \\ &= N \lg N \end{aligned}$$

Can say it's $\Theta(N \lg N)$ for arbitrary N (not just 2^k).