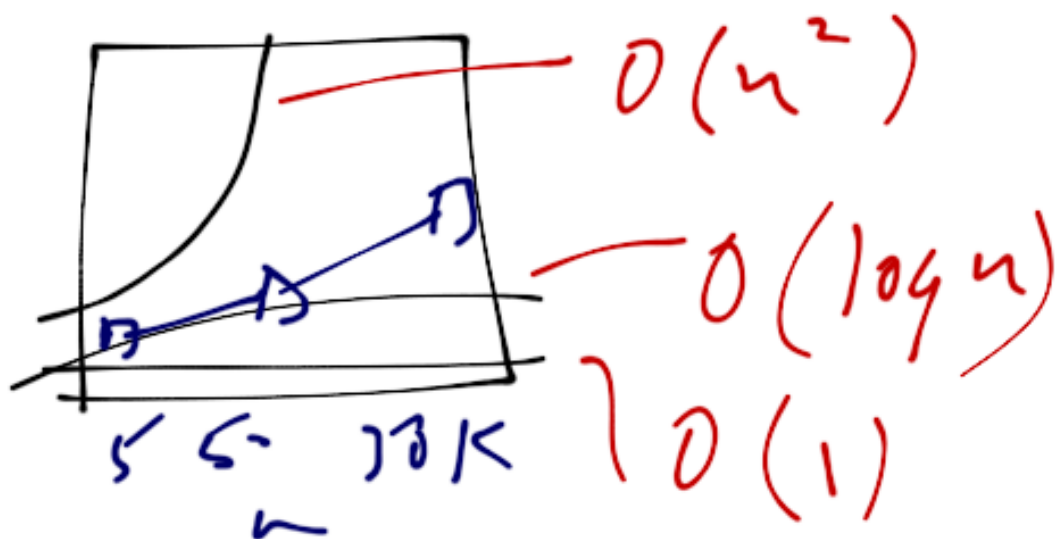


Alg. Analysis (Chp. 2)

- goal: compare & predict alg. performance (run time)
- empirical approach:
 - use timer_t class



- graph & report
(e.g. in tables)

(still relevant today
15. graphics papers)

- problems (w/empirical
approach):

- machine-specific

- unreliable for

few runs - need to
run several times

- any given run-on
Unix may take a
diff. amount of
time, Unix is

MULTI-tasking OS:

many processes

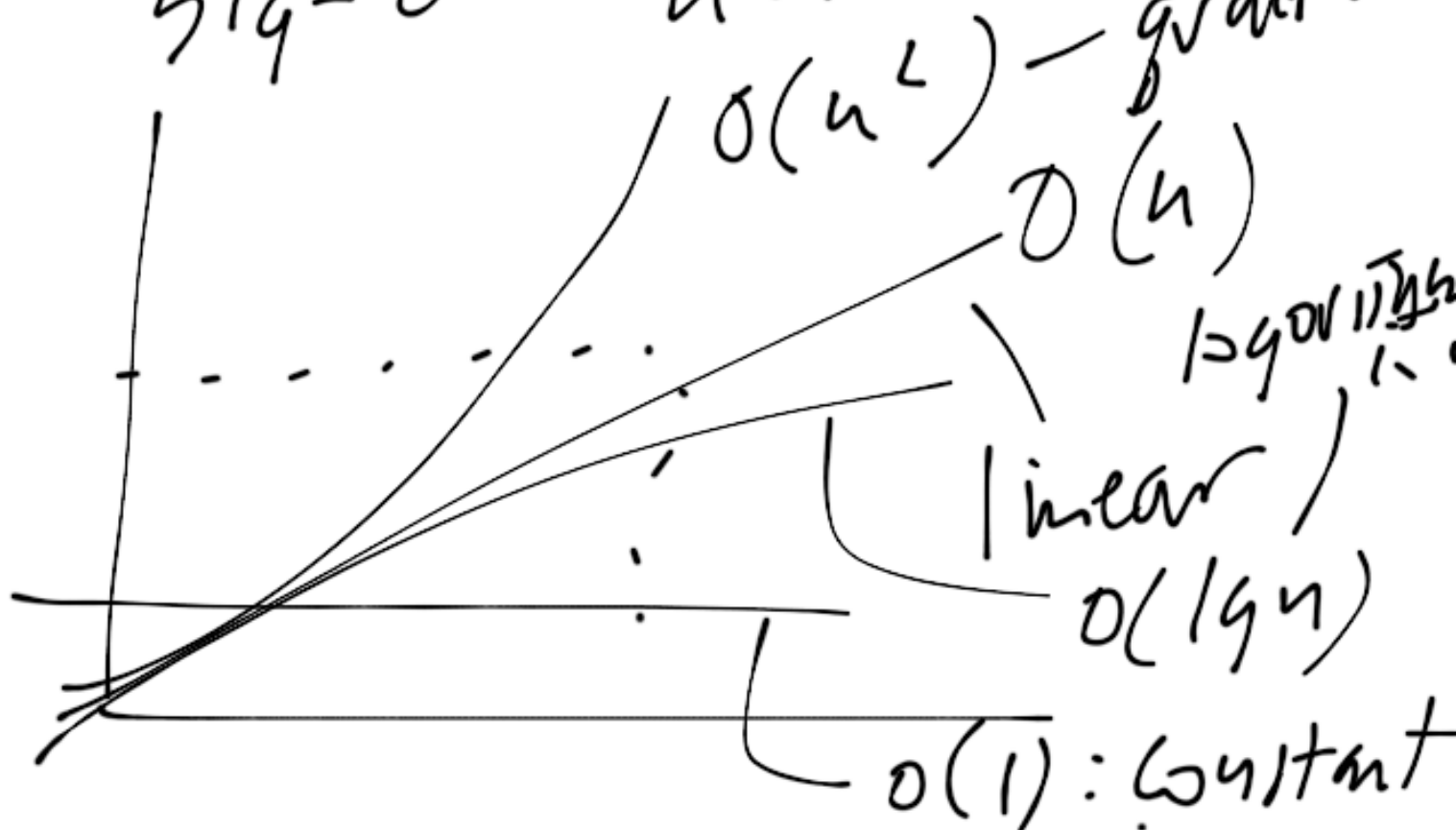
running "simultaneously"

but only one CPU

⇒ context switching

- analytical approach:

big-oh notation



n : input size

(e.g. n : size of array

5, 50, 500, 50,000)

Comparisons:

$$O(1) < O(\lg n) < O(\lg^2 n)$$

constant logarithmic iterated

$$< O(n) \text{ linear} \quad \text{logarithm}$$

$$< O(n \lg n) < O(n^2)$$

log linear

quadratic

$$< O(n^3)$$

cubic

$$< O(2^n)$$

exponential

- What to count?

operations
(cost)

```
int sum(int n)  
{
```

0 ————— int sum;

1 ————— sum = 0;

2n+2 ————— for (i=1 to n)

2n ————— sum += i*i*i;

1 ————— return sum;

————— }
6n+4

- $f(n) + 4$:

- reduce all constants
to 1

$\Rightarrow |n+1| \Rightarrow n+1$

- apply $O(f(n))$ definition

$T(n) = O(f(n))$ if

$\exists c, n_0 \mid T(n) \leq c f(n)$

when $n \geq n_0$.

- $n+1 \in O(n)$ ✓

$T(n)$ $O(f(n))$

$f(n)$ is n

$$n+1 \leq cn$$

$$\left(\uparrow T(n) \leq cf(n) \right)$$

such that

when $n \geq n_0$

e.g. let $n_0 = 100$, $c = 2$

$$100+1 \leq 2 \times 100 \quad ?$$

- in general, drop lower-order terms

$$\begin{aligned}T(n) &= O(2n^2) \\ &= O(n^2 + n) \\ &= O(n^2)\end{aligned}$$

- big-oh ($O(f(n))$)
specifies upper bound:

alg can't be any
worse than $O(f(n))$

- $O(f(n))$: worst-case
running time

- $\Omega(g(n))$:

$$T(n) = \Omega(g(n)) \Leftrightarrow$$

$$\exists c, n_0 \mid$$

$$T(n) \geq c g(n)$$

- $\Omega(g(n))$ means
that $g(n)$ must be
~~the~~ alg's lower bound

alg. is at least as

fast as this ($\Omega(g(n))$)
(slow)

(best-case performance)

- $T(n) = \Theta(h(n))$ | iff

$$T(n) = O(h(n)) \ \&$$

$$T(n) = \Omega(h(n))$$

$$\Omega(n) = T(n) = O(n)$$

\Rightarrow upper & lower bounds

the same, so $\Theta()$

denotes average running
time.

- $T(n) = o(p(n))$ if

little
oh

$\forall c \exists n_0$ |

$T(n) < c p(n)$ when
($n > n_0$)

strict inequality

(cf. $O(f(n))$ allows \leq)

little oh does not)

(no one was little-oh :)

- program analysis
(rules of thumb)

1. for loops

- no. statements \times
no. iterations

2. nested loops

- count inside out

$O(n)$ — for($i=0; i < n; i++$)

$O(n)$ — for($j=0; j < n; j++$)

$k++;$

$O(n^2)$

3. consecutive statements,
just add up

(Σ)

4. conditionals :

use larger of the
branches

(overestimate rather
than underestimate)

5. Recursion: ~~the~~
tricky one

— need to solve

recurrence relations

eg- fib(int n)

{ if (n <= 1) return 1;

else

return fib(n-1) +

fib(n-2)

T(0) + T(1) = 1

$$T(n) = T(n-1) + T(n-2) + \dots$$

book says for $n > 4$,

$$T(n) \geq \left(\frac{3}{2}\right)^n$$

inductive proof:

$$T(n) = T(n-1) + T(n-2) \geq \left(\frac{3}{2}\right)^n$$

assume this holds for $n > 4$

show it holds for $n+1$

$$\text{show } T(n+1) = T(n) + T(n-1)$$

so if n is $n+1$,

$$(n+1)-1 = n$$

$$(n+1)-2 = n-1$$

$$\begin{aligned} T(n+1) &= T(n) + T(n-1) \\ &= \left(\frac{3}{2}\right)^n + \left(\frac{3}{2}\right)^{n-1} \end{aligned}$$

(from top $T(n) \geq \left(\frac{3}{2}\right)^n$,

$$T(n-1) \geq \left(\frac{3}{2}\right)^{n-1}$$

$$T(n) \geq \left(\frac{3}{2}\right)^n$$

$$\binom{3}{2}^u + \binom{3}{2}^{u-1}$$

want this to be

$$\geq \binom{3}{2}^{u+1}$$

$$\binom{3}{2}^{-1} \binom{3}{2}^{u+1}$$

$$\binom{3}{2}^{-1} \binom{3}{2}^{-1} \binom{3}{2}^{u+1}$$

$$\left(\frac{3}{2}\right)^{-1} \left(\frac{3}{2}\right)^{n+1} + \left(\frac{3}{2}\right)^{-1} \left(\frac{3}{2}\right)^{-1} \left(\frac{3}{2}\right)^{n+1}$$

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

$$\left(\frac{2}{3} + \frac{4}{9}\right) \left(\frac{3}{2}\right)^{n+1}$$

$$= \left(\frac{10}{9}\right) \left(\frac{3}{2}\right)^{n+1}$$

which is larger than $\left(\frac{3}{2}\right)^{n+1}$



- generally we deal with recursive calls that split lists up in two, then recurse on each half

$$T(n) = 2T\left(\frac{n}{2}\right) + n$$

What
 $O(f(n))$
does this
mean?

time to process
half the list,
twice
 $+ n$