

Lecture 5 - January 21

Asymptotic Analysis of Algorithms

From Absolute RT to Relative RT

Approximating RT Functions

Asymptotic Upper Bound (Big-O): Def.

Announcements/Reminders

- **Assignment 1** released
- ***splitArrayHarder***: an extended version released
- Office Hours: 3pm to 4pm, Mon/Tue/Wed/Thu
- Contact Information of TAs on common eClass site

Example 2: Counting Number of Primitive Operations

(Exercise)

```
1  boolean foundEmptyString = false;
2  int i = 0;
3  while (!foundEmptyString && i < names.length) {
4      if (names[i].length() = 0) {
5          /* set flag for early exit */
6          foundEmptyString = true;
7      }
8      i = i + 1;
9  }
```

Q. # of times **Line 3** is executed?

Q. # of times **loop body (Lines 4 to 8)** is executed?

Q. # of POs in the **loop body (Lines 4 to 8)**?

Comparing Algorithms: From Absolute RT to Relative RT

t
 exact/absolute
 time taken to
 execute a P.O.
 in some exper. environment

e.g. Mac M1 $t = 2$ (ms)

e.g. Mac'14 $t = 4$ (ms)

absolute
 RTs of
 two alg.
 (solving
 the same
 problem)

abs. RT	Mac M1	Mac'14
$7n - 2$	$(100 \cdot 7 - 2) \cdot 2$	$(100 \cdot 7 - 2) \cdot 4$
$10n + 3$	$(100 \cdot 10 + 3) \cdot 2$	

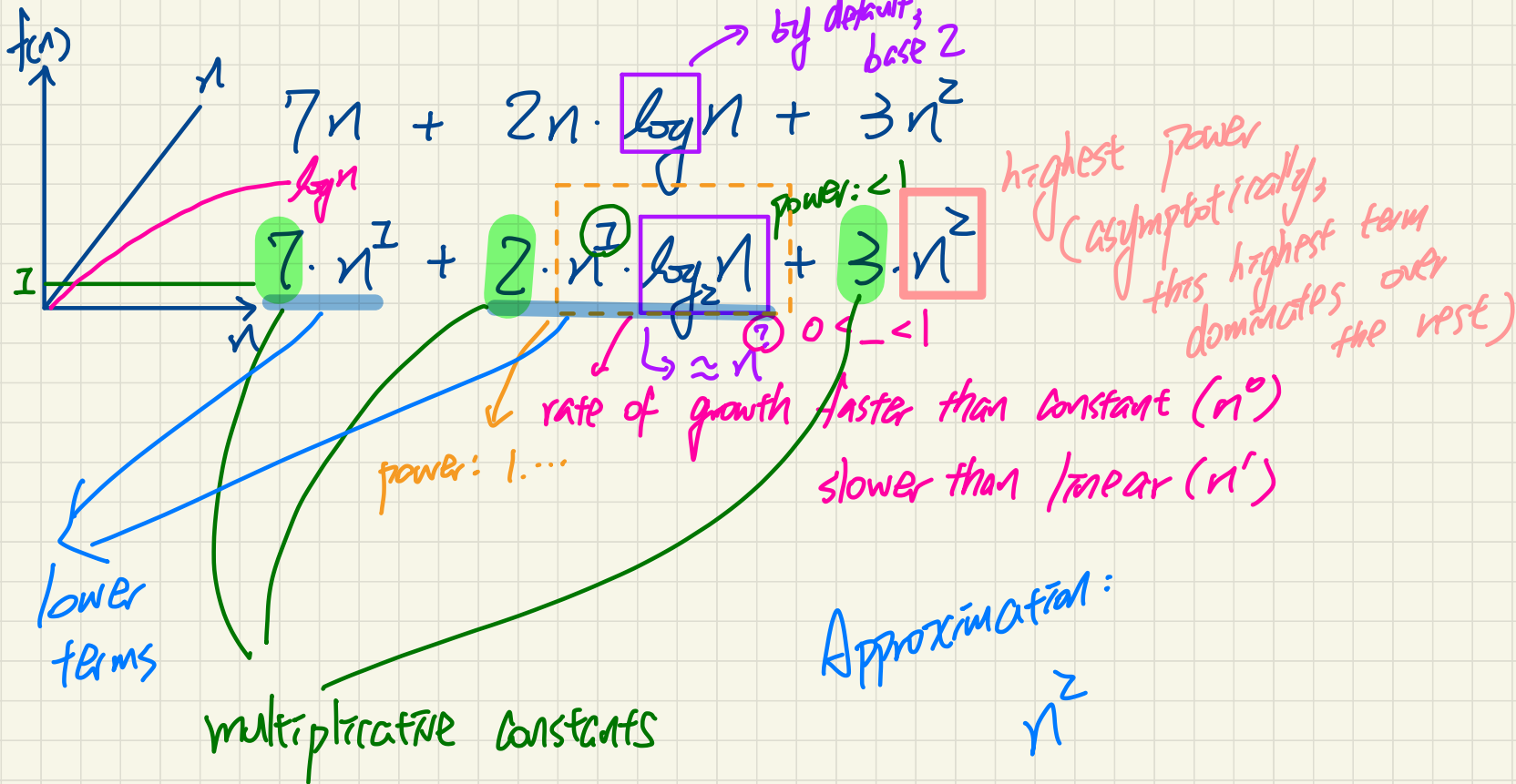
input
 size
 (assume $n=100$)

when comparing
 two diff. alg.,
 not necessary to consider
 the extra common difference
 factor. $\rightarrow t$ disre.

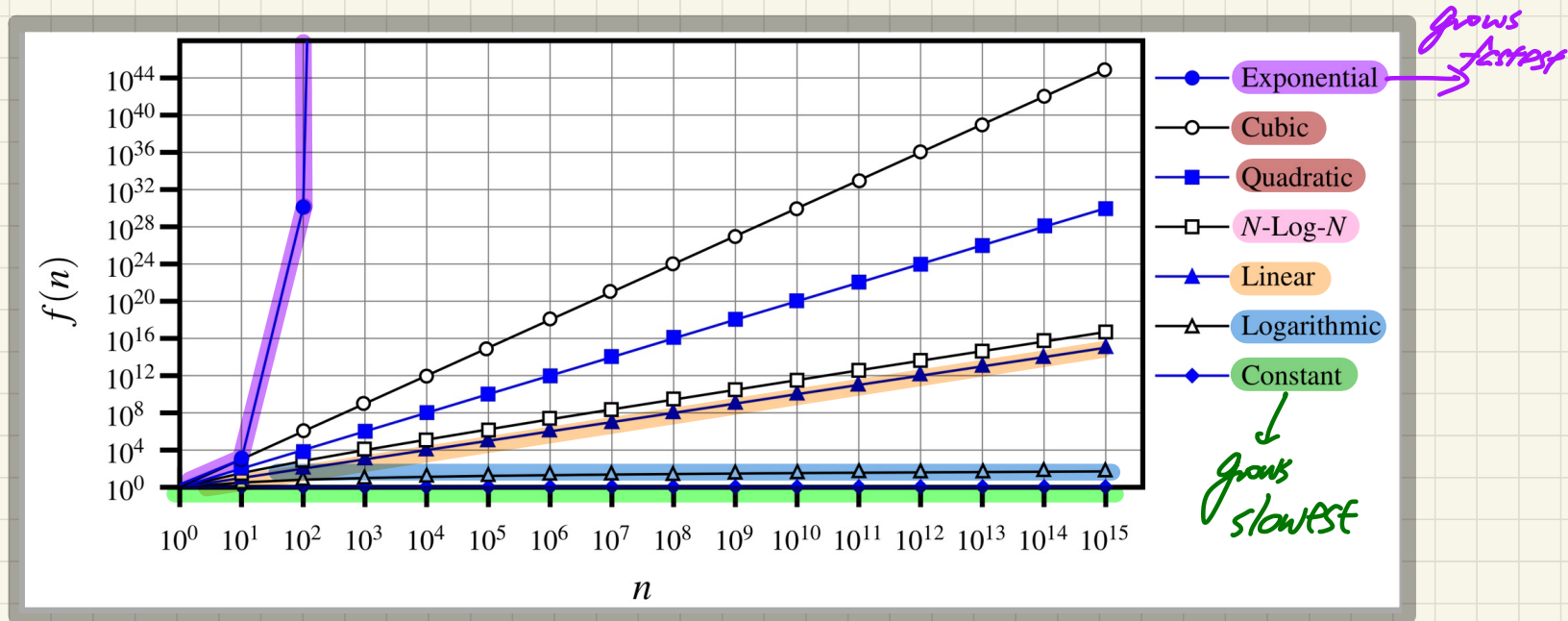
when consid.
 the same alg.,
 it's not
 necessary to
 know the
 precise RT

n^0 n^1 $\log n$

Exercise: Approximating $f(n) = 7n + 2n \cdot \log n + 3n^2$



RT Functions: Rates of Growth (w.r.t. Input Sizes)



RT

$$\begin{aligned} f(n) &= 1000 = 1000 \cdot n^0 & f(1000) &= 1000 \\ f(1) &= 1000 & f(1M) &= 1000 \end{aligned}$$

Comparing **Relative**, **Asymptotic** RTs of Algorithms

Q1. Compare:

$$\text{RT}_1(n) = \cancel{3}n^2 + \cancel{7}n + \cancel{18} \approx n^2$$

$$\text{RT}_2(n) = \cancel{100}n^2 + \cancel{3}n - \cancel{100} \approx n^2$$

↳ equally efficient, asymptotically.

Q2: Compare:

$$\text{RT}_1(n) = n^3 + \cancel{7}n + \cancel{18} \approx n^3$$

$$\text{RT}_2(n) = \cancel{100}n^2 + \cancel{100}n + \cancel{2000} \approx n^2$$

↳ RT_2 more efficient (taking less time), asymptotically.

$f(n)$: RT function

↳ input size \leadsto relative RT

$f(n)$

$\rightarrow O(g(n))$ e.g. $f(n) = 7n - 2$

$g(n)$: reference function

(further manipulation on $g(n)$ expected)

$g(n) \leadsto c \cdot g(n)$

$f(n)$ being a member in the family means that it can be upper-bounded by $g(n)$

Goal: Prove $f(n)$ is $O(g(n))$

u.b.e. : upper-bound effect

Asymptotic Upper Bound: **Big-O**

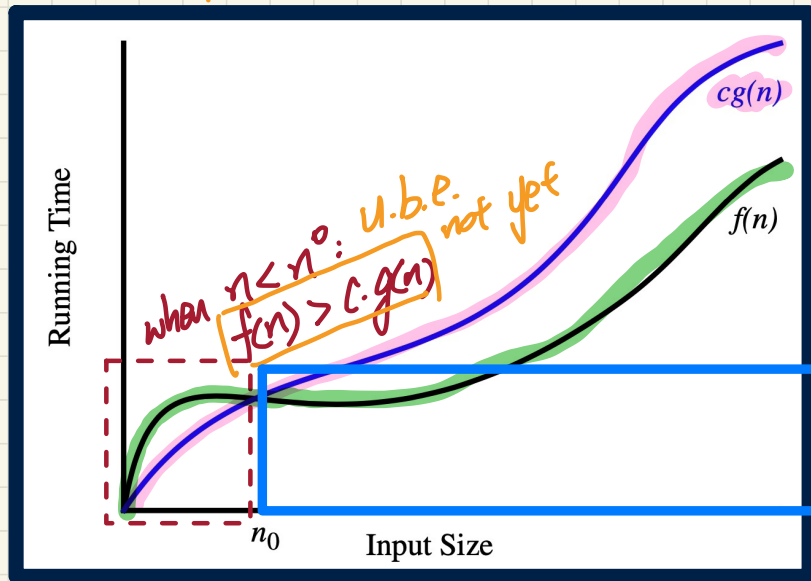
$f(n) \in O(g(n))$ if there are:

- A real constant $c > 0$
- An integer constant $n_0 \geq 1$

such that:

$$f(n) \leq c \cdot g(n) \text{ for } n \geq n_0$$

starting point of the u.b.e.
multiplicative constant applied to $g(n)$ to change its slope



Example:

$$f(n) = 8n + 5$$

$$g(n) = n$$

Can n_0 be 1?

$$f(1) \stackrel{?}{\leq} 9 \cdot 1$$

$$8 + 5 = 13$$

False

↳ u.b.e. not there.

Prove:

$$f(n) \text{ is } O(g(n))$$

Choose $c = 9$

What about n_0 ?

starting from $n = n_0$
 $f(n) \leq c \cdot g(n)$
u.b.e. is there!

Proving $f(n)$ is $O(g(n))$

If $f(n)$ is a polynomial of degree d , i.e.,

$$f(n) = a_0 \cdot n^0 + a_1 \cdot n^1 + \dots + a_d \cdot n^d$$

and a_0, a_1, \dots, a_d are integers (i.e., negative, zero, or positive),

then $f(n)$ is $O(n^d)$. $\rightarrow g(n)$

We prove by choosing

$$\begin{aligned} c &= |a_0| + |a_1| + \dots + |a_d| \\ n_0 &= 1 \end{aligned}$$

$$\begin{aligned} (1) f(1) &\leq c \cdot 1^d \\ (2) f(n) &\leq c \cdot n^d \quad (n > 1) \end{aligned}$$

Upper-bound effect: $n_0 = 1$?

$$[f(1) \leq (|a_0| + |a_1| + \dots + |a_d|) \cdot 1^d]$$

Upper-bound effect holds?

$$[f(n) \leq (|a_0| + |a_1| + \dots + |a_d|) \cdot n^d]$$