

CS311 Data Structures

Lecture 02 — BigO

Jyh-Ming Lien

June 4, 2018

Logistics

At Home

- ▶ Read Weiss Ch 1-4: Java Review
- ▶ Read Weiss Ch 5: Big-O
- ▶ Get your java environment set up

Goals

- ▶ Max Subarray problem (Code provided on Course webpage.)
- ▶ Review Big O and other asymptotic notations

How Fast/Big?

Review

Algorithmic time/space complexity depend on **problem size**

- ▶ Problem size: Often have some input parameter like n or N or (M, N)
- ▶ Describe both time and space complexity as *functions* of those parameters
- ▶ **Example:** For an input array of size N , the maximum element can be found in $5 * N + 3$ operations while the array can be sorted in $2N^2 + 11N + 7$ operations.

Big O

Big-O notation: upper bounding how fast functions grow based on input $T(n)$ is $O(F(n))$ if there are positive constants c and n_0 such that

- ▶ When $n \geq n_0$
- ▶ $T(n) \leq cF(n)$

Bottom line:

- ▶ If $T(n)$ is $O(F(n))$
- ▶ Then $F(n)$ grows as fast or faster than $T(n)$

Show It

Show

$$f(n) = 2n^2 + 3n + 2 \text{ is } O(n^3)$$

- ▶ Pick $c = 0.5$ and $n_0 = 6$

n	$f(n)$	$0.5n^3$
0	2	0
1	7	0.5
2		
3		
4		
5		
6		
7		

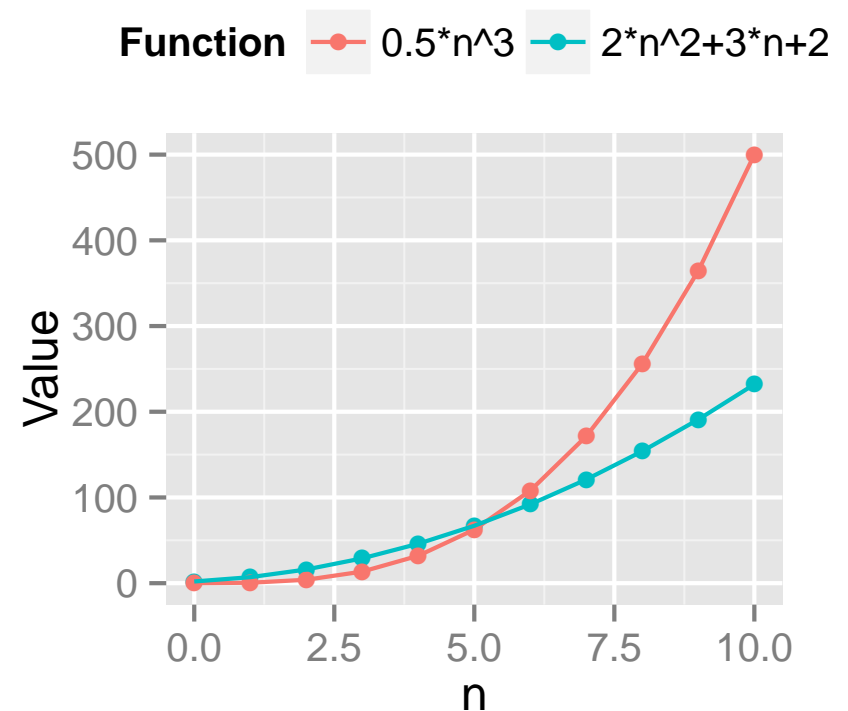
Show It

Show

$f(n) = 2n^2 + 3n + 2$ is $O(n^3)$

► Pick $c = 0.5$ and $n_0 = 6$

n	$f(n)$	$0.5n^3$
0	2	0
1	7	0.5
2	16	4
3	29	13
4	46	32
5	67	62
6	92	108
7	121	171



How about the opposite? Show

$g(n) = n^3$ is $\Omega(2n^2 + 3n + 2)$

Basic Rules

- ▶ Constant additions disappear
 - ▶ $N + 5$ is $O(N)$
- ▶ Constant multiples disappear
 - ▶ $0.5N + 2N + 7$ is $O(N)$
- ▶ Non-constant multiples multiply:
 - ▶ Doing a constant operation $2N$ times is $O(N)$
 - ▶ Doing a $O(N)$ operation $N/2$ times is $O(N^2)$
 - ▶ Need space for half an array with N elements is $O(N)$ space overhead
- ▶ Function calls **are not free** (including library calls)
 - ▶ Call a function which performs 10 operations is $O(1)$
 - ▶ Call a function which performs $N/3$ operations is $O(N)$
 - ▶ Call a function which copies object of size N takes $O(N)$ time and uses $O(N)$ space

Growth Ordering of Some Functions

Name	Leading Term	Big-Oh	Example
Constant	$1, 5, c$	$O(1)$	$2.5, 85, 2c$
Log-Log	$\log(\log(n))$	$O(\log \log n)$	$10 + (\log \log n + 5)$
Log	$\log(n)$	$O(\log(n))$	$5 \log n + 2$ $\log(n^2)$
Linear	n	$O(n)$	$2.4n + 10$ $10n + \log(n)$
N-log-N	$n \log n$	$O(n \log n)$	$3.5n \log n + 10n + 8$
Super-linear	$n^{1.x}$	$O(n^{1.x})$	$2n^{1.2} + 3n \log n - n + 2$
Quadratic	n^2	$O(n^2)$	$0.5n^2 + 7n + 4$ $n^2 + n \log n$
Cubic	n^3	$O(n^3)$	$0.1n^3 + 8n^{1.5} + \log(n)$
Exponential	a^n	$O(2^n)$ $O(10^n)$	$8(2^n) - n + 2$ $100n^{500} + 2 + 10^n$
Factorial	$n!$	$O(n!)$	$0.25n! + 10n^{100} + 2n^2$

Common Patterns

- ▶ Adjacent Loops Additive: $2 \times n$ is $O(n)$

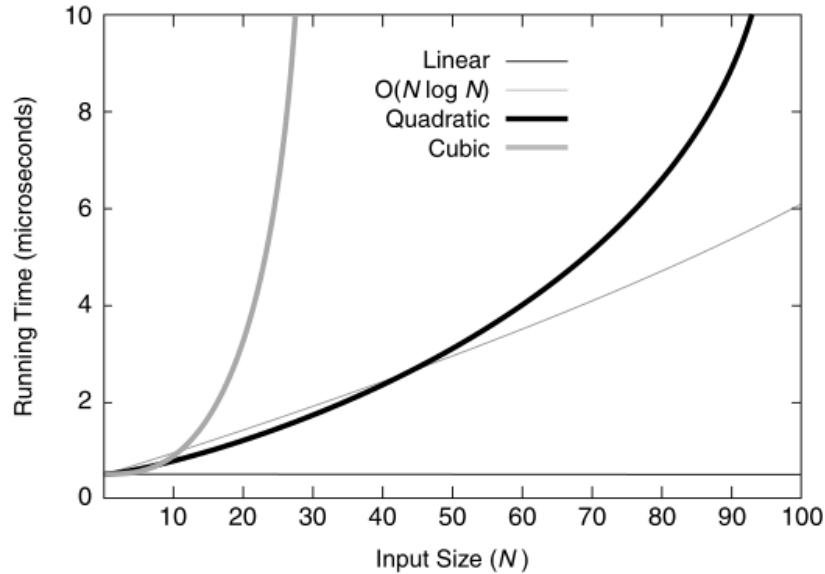
```
for(int i=0; i<N; i++){
    blah blah blah; //constant time operation
}
for(int j=0; j<N; j++){
    yakkety yack; //another constant time operation
}
```

- ▶ Nested Loops Multiplicative usually polynomial
 - ▶ 1 loop, $O(n)$
 - ▶ 2 loops, $O(n^2)$
 - ▶ 3 loops, $O(n^3)$
- ▶ Repeated halving usually involves a logarithm
 - ▶ Binary search is $O(\log n)$
 - ▶ Fastest sorting algorithms are $O(n \log n)$
 - ▶ Proofs are harder, require solving recurrence relations

Lots of special cases so be careful

FIGS/Idealized Functions

Smallish Inputs



Larger Inputs

