CS062 DATA STRUCTURES AND ADVANCED PROGRAMMING

5: Analysis of Algorithms



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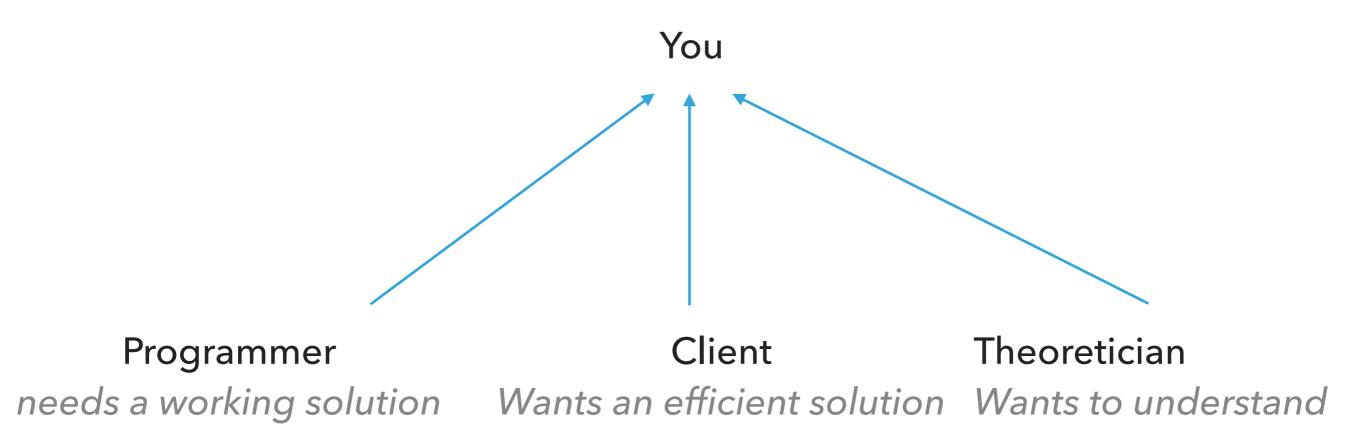


Tom Yeh he/him/his

Introduction

- Experimental Analysis of Running Time
- Mathematical Models of Running Time
- Order of Growth Classification
- Analysis of Memory Consumption

Different Roles



Why analyze algorithmic efficiency?

- Predict performance.
- Compare algorithms that solve the same problem.
- Provide guarantees.
- Understand theoretical basis.
- Avoid performance bugs.

Why is my program so slow? Why does it run out of memory?

We can use a combination of experiments and mathematical modeling.

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▶ 3-SUM: Given *n* distinct numbers, how many unordered triplets sum to 0?

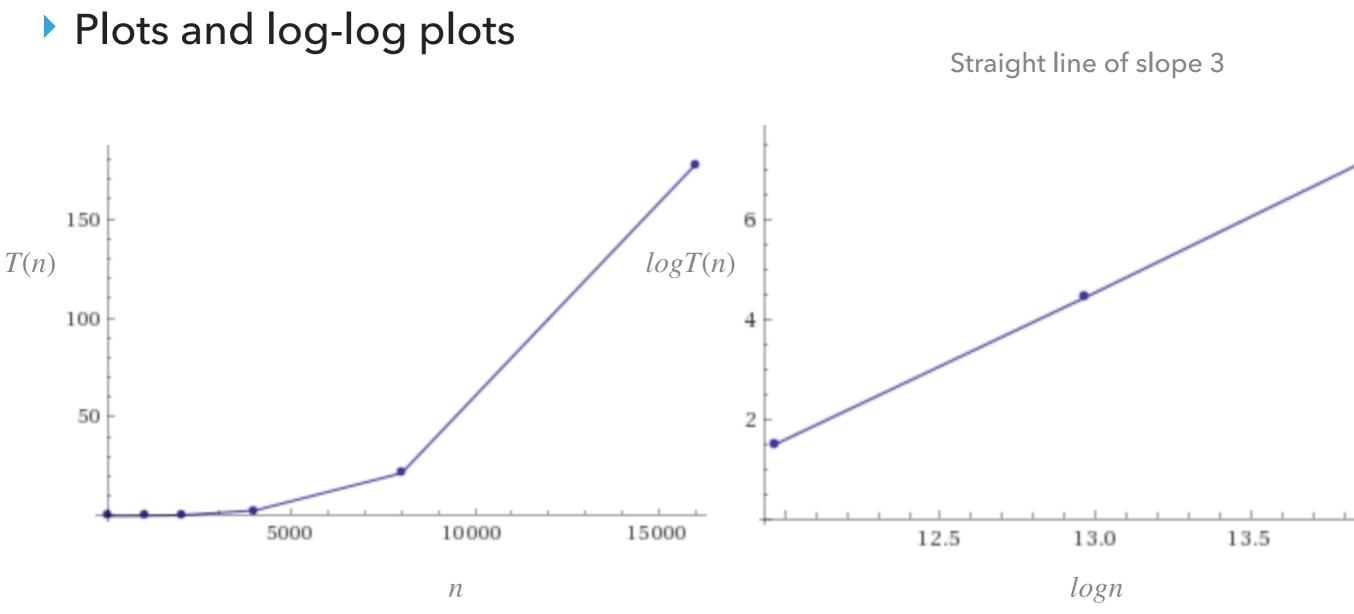
- Input: 30 -40 -20 -10 40 0 10 5
- Output: 4
 - ▶ 30 -40 10
 - ▶ 30 -20 -10
 - ► -40 40 0
 - ► -10 0 10

3-SUM: brute-force algorithm

```
public class ThreeSum {
public static int count(int[] a) {
         int n = a.length;
         int count = 0;
         for (int i = 0; i < n; i++) {</pre>
             for (int j = i+1; j < n; j++) {
                  for (int k = j+1; k < n; k++) {
                       if (a[i] + a[j] + a[k] == 0) {
                            count++;
                       }
                  }
              }
                                         public static void main(String[] args) {
         }
                                                   String filename = args[0];
         return count;
                                                   int fileSize = Integer.parseInt(args[1]);
    }
                                                   try {
                                                        Scanner scanner = new Scanner(new File(filename));
                                                       int intList[] = new int[fileSize];
                                                       int i=0;
                                                       while(scanner.hasNextInt()){
                                                            intList[i++]=scanner.nextInt();
                                                        }
                                                       Stopwatch timer = new Stopwatch();
                                                       int count = count(intList);
                                                       System.out.println("elapsed time = " + timer.elapsedTime());
                                                       System.out.println(count);
                                                   }
                                                   catch (IOException ioe) {
                                                       throw new IllegalArgumentException("Could not open " + filename, ioe);
                                                   }
                                              }
```

Empirical Analysis

 Input: 8ints.txt Output: 4 and 0 	Input size	Time
 Input: 1Kints.txt Output: 70 and 0.081 	8	0
 Input: 2Kints.txt Output: 528 and 0.38 	1000	0.081
Input: 2Kints.txt	2000	0.38
 Output: 528 and 0.371 Input: 4Kints.txt 	2000	0.371
Output: 4039 and 2.792	4000	2.792
 Input: 8Kints.txt Output: 32074 and 21.623 	8000	21.623
 Input: 16Kints.txt Output: 255181 and 177.344 	16000	177.344



• Regression: $T(n) = an^b$ (power-law).

- ▶ $\log T(n) = b \log n + \log a$, where *b* is slope.
- Experimentally: $\sim 0.42 \times 10^{-10} n^3$, in our example for ThreeSum.

	Input size	Time
EXPERIMENTAL ANALYSIS OF RUNNING TIME	8	0
	1000	0.081
	2000	0.38
	4000	2.792
Doubling hypothesis	8000	21.623
	16000	177.344

- Doubling input size increases running time by a factor of $\frac{T(n)}{T(n/2)}$
- Run program doubling the size of input. Estimate factor of growth: $\frac{T(n)}{T(n/2)} = \frac{an^b}{a(\frac{n}{2})^b} = 2^b.$
- E.g., in our example, for pair of input sizes 8000 and 16000 the ratio is 8.2, therefore b is approximately 3.
- Assuming we know *b*, we can figure out *a*.
 - E.g., in our example, $T(16000) = 177.34 = a \times 16000^3$.
 - Solving for a we get $a = 0.42 \times 10^{-10}$.

Practice Time

Suppose you time your code and you make the following observations. Which function is the closest model of *T*(*n*)?
 A. *n*²

Β.	$6 \times 10^{-4} n$
C.	$5 \times 10^{-9} n^2$
D.	$7 \times 10^{-9} n^2$

Input size	Time
1000	0
2000	0.0
4000	0.1
8000	0.3
16000	1.3
32000	5.1

- Answer
- C. $5 \times 10^{-9} n^2$
- Ratio is approximately 4, therefore b = 2.
- ► $T(32000) = 5.1 = a \times 32000^2$.
- Solving for $a = 4.98 \times 10^{-9}$.s

Input size	Time
1000	0
2000	0.0
4000	0.1
8000	0.3
16000	1.3
32000	5.1

- Effects on performance
- System independent effects: Algorithm + input data
 - Determine *b* in power law relationships.
- System dependent effects: Hardware (e.g., CPU, memory, cache) + Software (e.g., compiler, garbage collector) + System (E.g., operating system, network, etc).
- Dependent and independent effects determine a in power law relationships.
- Although it is hard to get precise measurements, experiments in Computer Science are cheap to run.

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Total Running Time

- Popularized by Donald Knuth in the 60s in the four volumes of "The Art of Computer Programming".
 - Knuth won the Turing Award (The "Nobel" in CS) in 1974.
- In principle, accurate mathematical models for performance of algorithms are available.
- Total running time = sum of cost x frequency for all operations.
- Need to analyze program to determine set of operations.
- Exact cost depends on machine, compiler.
- Frequency depends on algorithm and input data.

- Cost of basic operations
- Add < integer multiply < integer divide < floating-point add < floating-point multiply < floating-point divide.</p>

Operation	Example	Nanoseconds
Variable declaration	int a	<i>c</i> ₁
Assignment statement	a = b	<i>c</i> ₂
Integer comparison	a < b	<i>C</i> ₃
Array element access	a[i]	<i>C</i> ₄
Array length	a.length	<i>C</i> ₅
1D array allocation	new int[n]	<i>c</i> ₆ <i>n</i>
2D array allocation	new int[n][n]	$c_7 n^2$
string concatenation	s+t	<i>c</i> ₈ <i>n</i>

```
Example: 1-SUM
```

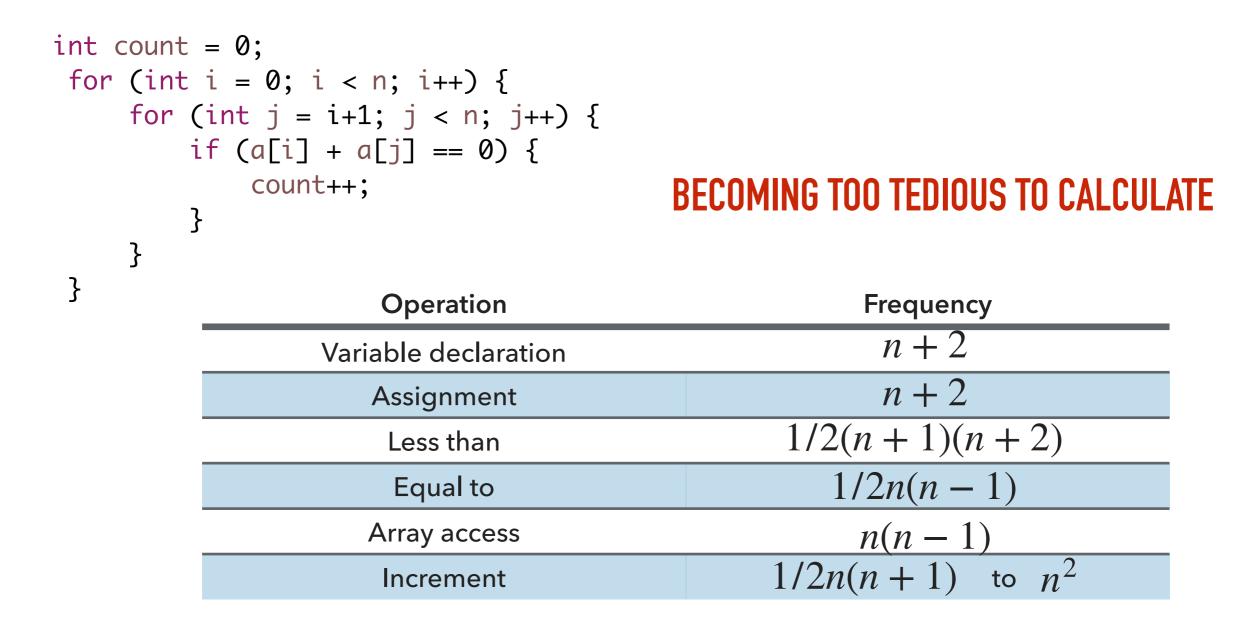
How many operations as a function of n?

```
int count = 0;
for (int i = 0; i < n; i++) {
    if (a[i] == 0) {
        count++;
    }
}
```

Operation	Frequency
Variable declaration	2
Assignment	2
Less than	<i>n</i> + 1
Equal to	n
Array access	п
Increment	<i>n</i> to 2 <i>n</i>

```
Example: 2-SUM
```

How many operations as a function of n?



Tilde notation

Estimate running time (or memory) as a function of input size *n*.

Ignore lower order terms.

When *n* is large, lower order terms become negligible.

• Example 1:
$$\frac{1}{6}n^3 + 10n + 100 \sim n^3$$

• Example 2: $\frac{1}{6}n^3 + 100n^2 + 47 \sim n^3$
• Example 3: $\frac{1}{6}n^3 + 100n^{\frac{2}{3}} + \frac{1/2}{n} \sim n^3$
• Technically $f(n) \sim g(n)$ means that $\lim_{n \to \infty} \frac{f(n)}{g(n)} = 1$

Simplification

- Cost model: Use some basic operation as proxy for running time.
 - E.g., array accesses
- Combine it with tilde notation.

Operation	Frequency	Tilde notation
Variable declaration	<i>n</i> + 2	~ <i>N</i>
Assignment	<i>n</i> + 2	~ 11
Less than	1/2(n+1)(n+2)	~ n ²
Equal to	1/2n(n-1)	$\sim n^2$
Array access	n(n-1)	~ n ²
Increment	$1/2n(n+1)$ to n^2	~ n ²

 $\sim n^2$ array accesses for the 2-SUM problem

- Back to the 3-SUM problem
- Approximately how many array accesses as a function of input size n?

```
int count = 0;
for (int i = 0; i < n; i++) {
    for (int j = i+1; j < n; j++) {
        for (int k = j+1; k < n; k++) {
            if (a[i] + a[j] + a[k] == 0) {
                count++;
            }
        }
    }
}

n<sup>3</sup> array accesses.
```

- Useful approximations for the analysis of algorithms
- Harmonic sum: $H_n = 1 + 1/2 + 1/3 + \ldots + 1/n \sim \ln n$
- Triangular sum: $1 + 2 + 3 + ... + n \sim n^2$
- Geometric sum: $1 + 2 + 4 + 8 + ... + n = 2n 1 \sim n$, when *n* power of 2.
- Binomial coefficients: $\binom{n}{k} \sim \frac{n^k}{k!}$ when k is a small constant.
- Use a tool like Wolfram alpha.

Practice Time

How many array accesses does the following code make?

```
int count = 0;
        for (int i = 0; i < n; i++) {</pre>
             for (int j = i+1; j < n; j++) {</pre>
                 for (int k = 1; k < n; k=k*2) {
                     if (a[i] + a[j] >= a[k]) {
                     count++;
                 }
             }
        }
A. n^2
B. n^2 \log n
C. n^{3}
D. n^3 \log n
```

Answer

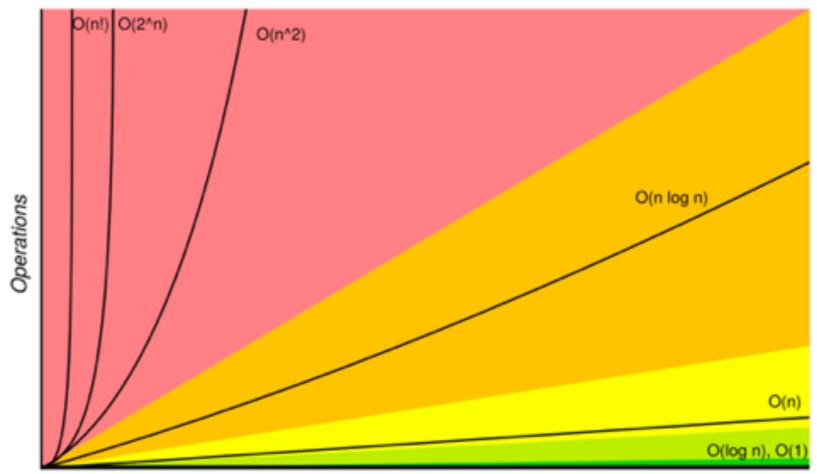
$$hn^2\log n$$

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Order-of-growth

- Definition: If f(n)~cg(n) for some constant c > 0, then the order of growth of f(n) is g(n).
 - Ignore leading coefficients.
 - Ignore lower-order terms.
- We will use this definition in the mathematical analysis of the running time of our programs as the coefficients depend on the system.
- E.g., the order of growth of the running time of the ThreeSum program is n^3 .

- Common order-of-growth classifications
- Good news: only a small number of function suffice to describe the order-of-growth of typical algorithms.
- 1: constant
- log *n*: logarithmic
- n : linear
- n log n : linearithmic
- n^2 : quadratic
- n^3 : cubic
- ▶ 2ⁿ: exponential
- n!: factorial



Elements

bigocheatsheet.com

Common order-of-growth classifications

Order-of-growth	Name	Typical code	T(n)/T(n/2)
1	Constant	a=b+c	1
log n	Logarithmic	while(n>1){n=n/2;}	~ 1
n	Linear	for(int i =0; i <n;i++{ }</n;i++{ 	2
$n\log n$	Linearithmic	mergesort	~ 2
<i>n</i> ²	Quadratic	for(int i =0;i <n;i++) for(int="" j="0;" j<n;j++){}}<="" td="" {=""><td>4</td></n;i++)>	4
<i>n</i> ³	Cubic	for(int i =0;i <n;i++) for(int="" j="0;" j<n;j++){="" k="0;" k++){}}}<="" k<n;="" td="" {=""><td>8</td></n;i++)>	8

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Basics

- **Bit:** 0 or 1.
- Byte: 8 bits.
- ▶ Megabyte (MB): 2²⁰ bytes.
- Gigabyte: 2³⁰ bytes.

- Typical memory usage for primitives and arrays
- boolean: 1 byte
- byte: 1 byte
- char: 2 bytes
- int: 4 bytes
- float: 4 bytes
- long: 8 bytes
- double: 8 bytes
- Array overhead: 24 bytes
- char[]:2n+24 bytes
- int[]:4n+24 bytes
- b double[]:8n+24 bytes

- Typical memory usage for objects
- Object overhead: 16 bytes
- Reference: 8 bytes
- Padding: padded to be a multiple of 8 bytes
- Example:

```
> public class Date {
    private int day;
    private int month;
    private int year;
}
```

16 bytes overhead + 3x4 bytes for ints + 4 bytes padding =
 32 bytes

Practice Time

How much memory does WeightedQuickUnionUF use as a function of n?

```
public class WeightedQuickUnionUF{
    private int[] parent;
    private int[] size;
    private int count;
    public WeightedQuickUnionUF(int n) {
        parent = new int[n];
        size = new int[n];
        count = 0;
...
}
A. \sim 4n bytes
B. ~8n bytes
C. ~4n^2 bytes
D. ~8n^2 bytes
```

Answer

- B. ~8n bytes
- 16 bytes for object overhead
- Each array: 8 bytes for reference + 24 overhead + 4n for integers
- 4 bytes for int
- 4 bytes for padding
- Total $88 + 8n \sim 8n$

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Readings:

- Textbook:
 - Chapter 1.4 (pages 172-196, 200-205)
- Website:
 - Analysis of Algorithms: https://algs4.cs.princeton.edu/14analysis/

Practice Problems:

▶ 1.4.1-1.4.9