Algorithms



ROBERT SEDGEWICK | KEVIN WAYNE

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ROBERT SEDGEWICK | KEVIN WAYNE

1.4 ANALYSIS OF ALGORITHMS

introduction

running time (experimental analysis)

running time (mathematical models)

memory usage

Last updated on 2/1/24 10:26AM





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introduction

Algorithms

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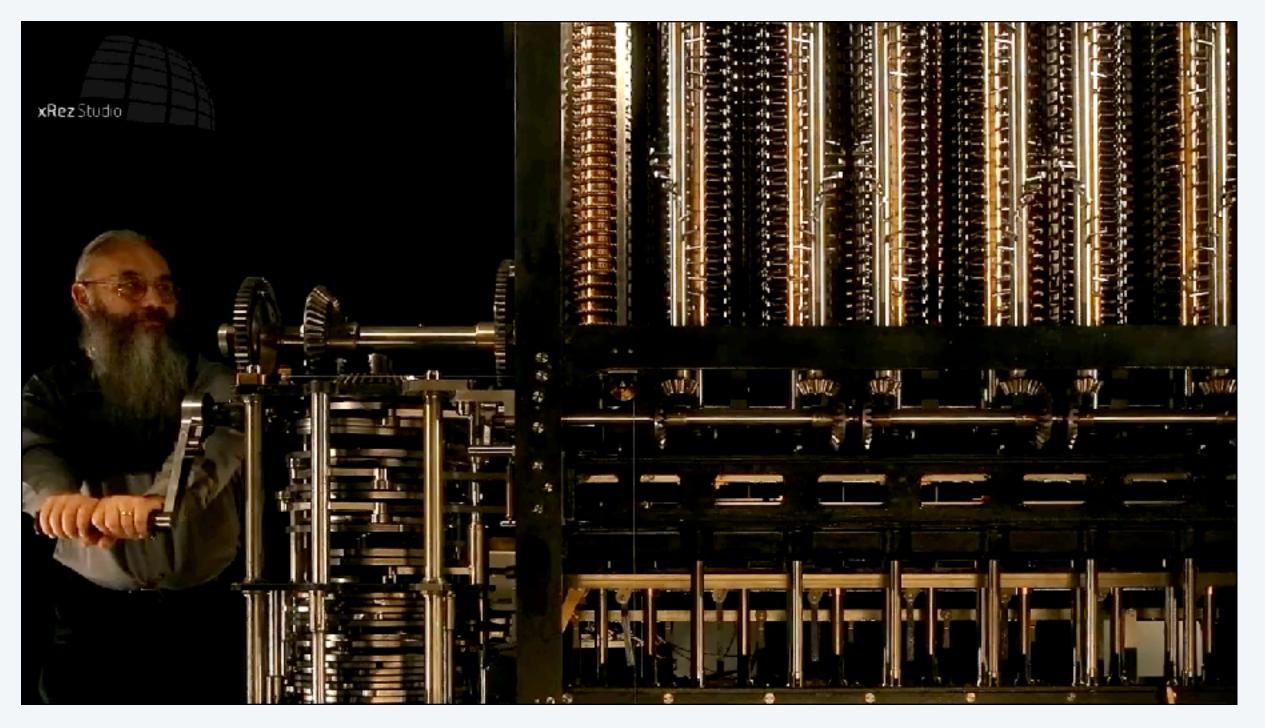
- running time (experimental analysis)
- running time (mathematical models)
- memory usage

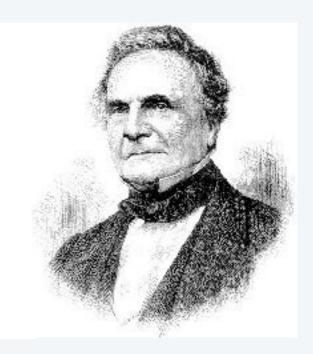


Running time

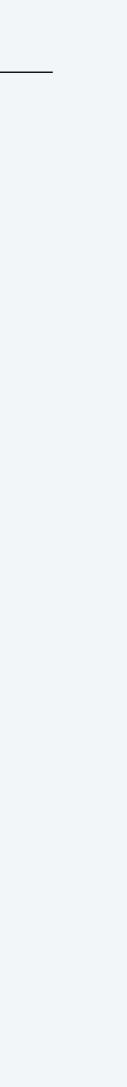
"As soon as an Analytical Engine exists, it will necessarily guide the future course of the science. Whenever any result is sought by its aid, the question will then arise—By what course of calculation can these results be arrived at by the machine in the shortest time?" — Charles Babbage (1864)

how many times do you have to turn the crank?





https://vimeo.com/49080293



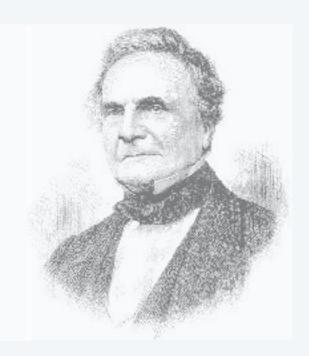


Running time

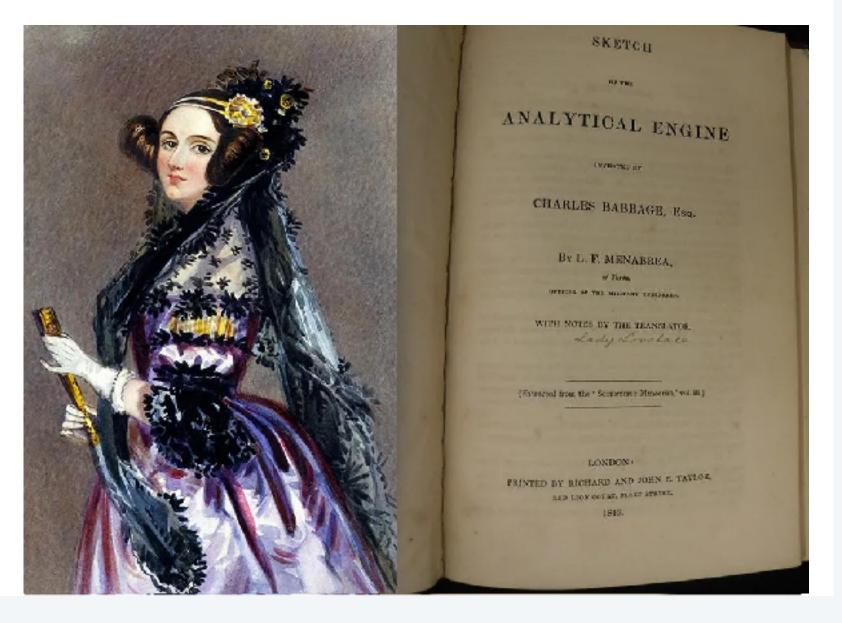
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Ada Lovelace's algorithm to compute Bernoulli numbers on Analytic Engine (1843)

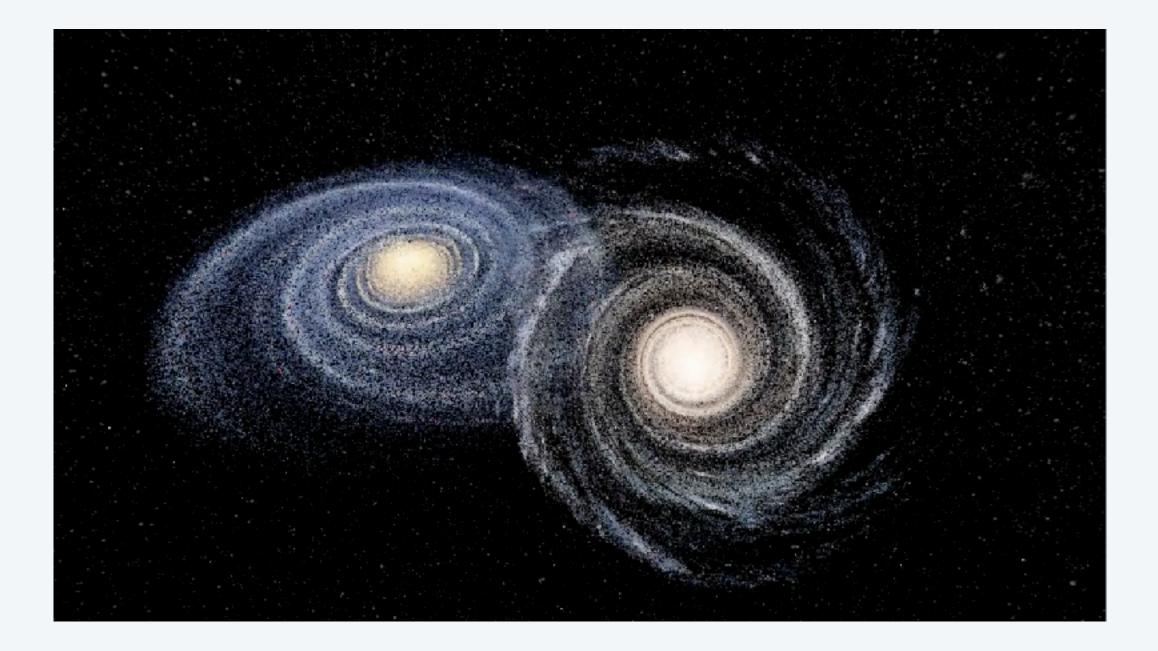


Rare book containing the world's first computer algorithm earns \$125,000 at auction



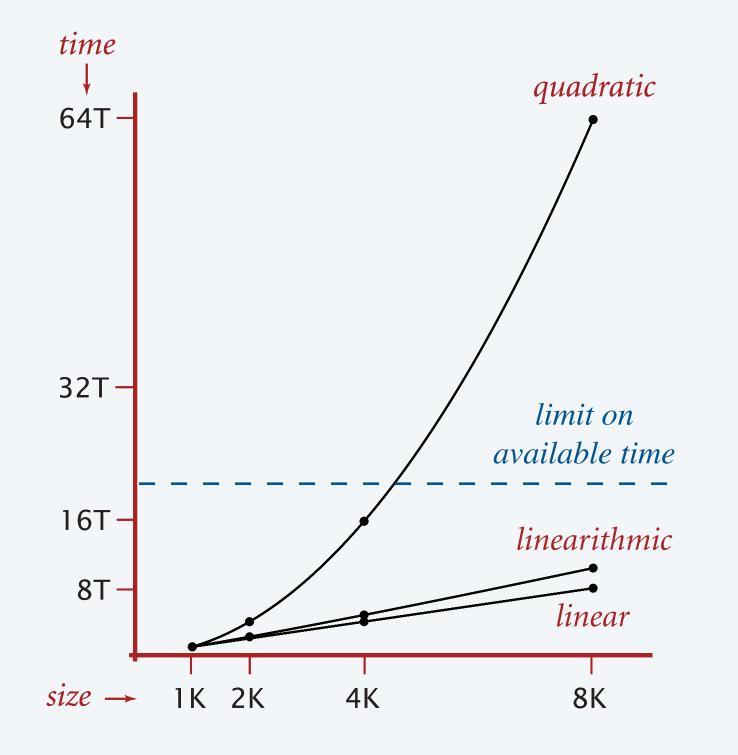
N-body simulation.

- Simulate gravitational interactions among *n* bodies.
- Applications: cosmology, fluid dynamics, semiconductors, ...
- Brute force: $\Theta(n^2)$ steps.
- Barnes-Hut algorithm: $\Theta(n \log n)$ steps, enables new research.





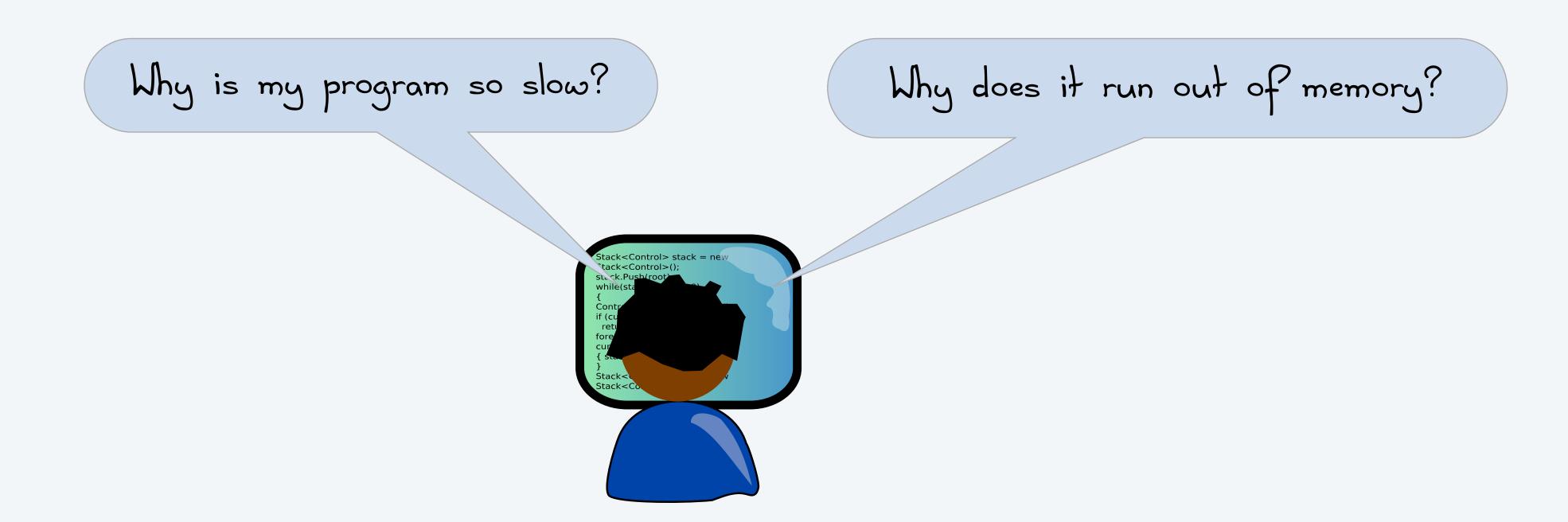
Andrew Appel PU '81



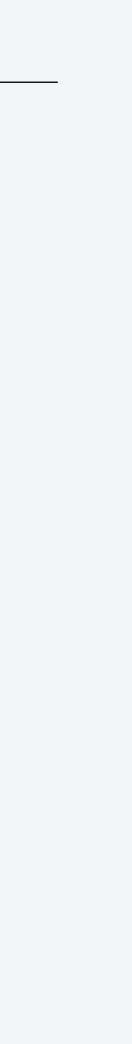
The challenge

Q1. Will my program be able to solve a large practical input?

Q2. If not, how might I understand its performance characteristics so as to improve it?



Our approach. Combination of experiments and mathematical modeling.



Example: 3-SUM

3-SUM. Given *n* distinct integers, how many triples sum to exactly zero?

Context. Connected with problems in computational geometry.



	a[i]	a[j]	a[k]	sum	
1	30	-40	10	0	~
2	30	-20	-10	0	~
3	-40	40	0	0	~
4	-10	0	10	0	~





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++)
         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++;
      return count;
  }
  public static void main(String[] args) {
      In in = new In(args[0]);
      int[] a = in.readAllInts();
      StdOut.println(count(a));
```

check distinct triples

for simplicity, ignore integer overflow



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memory usage

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running time (experimental analysis)

running time (mathematical models)



Measuring the running time

Running time. Run the program for inputs of varying size; measure running time.

Observation. The running time T(n) grows as a function of the input size n.







Measuring the running time

Running time. Run the program for inputs of varying size; measure running time.

n	time (seconds) †
1,000	0.21
1,500	0.71
2,000	1.63
2,500	3.11
3,000	5.43
4,000	12.8
5,000	25.0
7,500	84.4
10,000	199.3



† Apple M2 Pro with 32 GB memory running OpenJDK 11 on MacOS Ventura

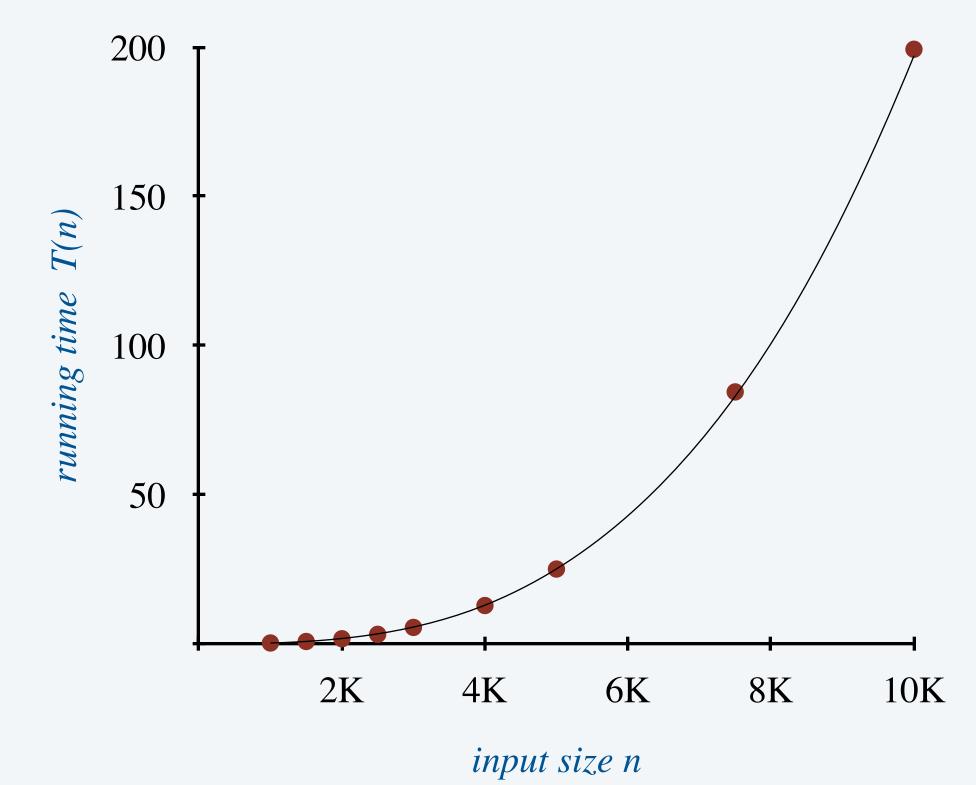


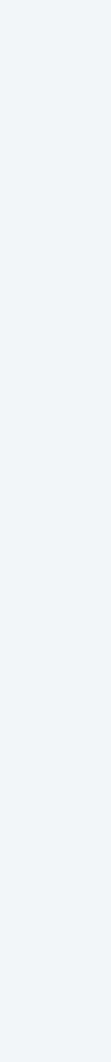
Data analysis: standard plot

Standard plot. Plot running time T(n) vs. input size n.

n	time (seconds) †
1,000	0.21
1,500	0.71
2,000	1.63
2,500	3.11
3,000	5.43
4,000	12.8
5,000	25.0
7,500	84.4
10,000	199.3

Hypothesis. The running time obeys a power law: $T(n) = a \times n^b$ seconds. How to validate hypothesis? How to estimate constants *a* and *b*? Questions.



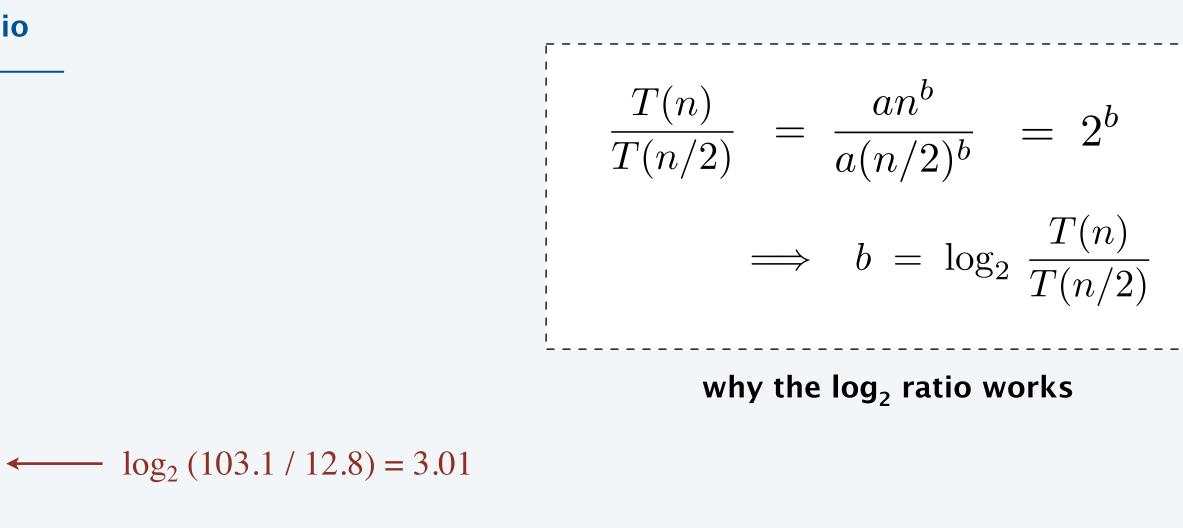


Doubling test: estimating the exponent b

Doubling test. Run program, **doubling** the size of the input.

- Assume running time satisfies the "power law" $T(n) = a \times n^b$.
- Estimate $b = \log_2$ ratio.

n	time (seconds)	ratio	log ₂ ratio
500	0.05		
1,000	0.21	4.20	2.07
2,000	1.63	7.76	2.96
4,000	12.8	7.85	2.97
8,000	103.1	8.05	3.01
16,000	819.0	7.94	2.99



seems to converge to a constant $b \approx 3.0$

Doubling test: estimating the leading coefficient a

Doubling test. Run program, **doubling** the size of the input.

- Assume running time satisfies $T(n) = a \times n^b$.
- Estimate $b = \log_2$ ratio.
- Estimate *a* by solving $T(n) = a \times n^b$ for a sufficiently large value of *n*.

n	time (seconds)	ratio	log ₂ ratio
500	0.05		
1,000	0.21	4.20	2.07
2,000	1.63	7.76	2.96
4,000	12.8	7.85	2.97
8,000	103.1	8.05	3.01
16,000	819.0	7.94	2.99

Hypothesis. Running time is about $2.00 \times 10^{-10} \times n^3$ seconds.

$$819.0 = a \times 16,000^3 \implies a = 2.00 \times 10^{-10}$$



Analysis of algorithms: quiz 1

Estimate the running time to solve a problem of size n = 96,000.

Α.	39 seconds	n	time (seconds)
D	52 cocodo	1,000	0.02
Β.	52 seconds	2,000	0.05
C.	117 seconds	4,000	0.20
D.	350 seconds	8,000	0.81
		16,000	3.25
		32,000	13.01



s)



Order of growth

Hypothesis. Running times on different computers differ by a constant factor.

Note. That factor can be several orders of magnitude.



1970s (VAX-11/780)







System independent effects.

- Algorithm.
- Input data.

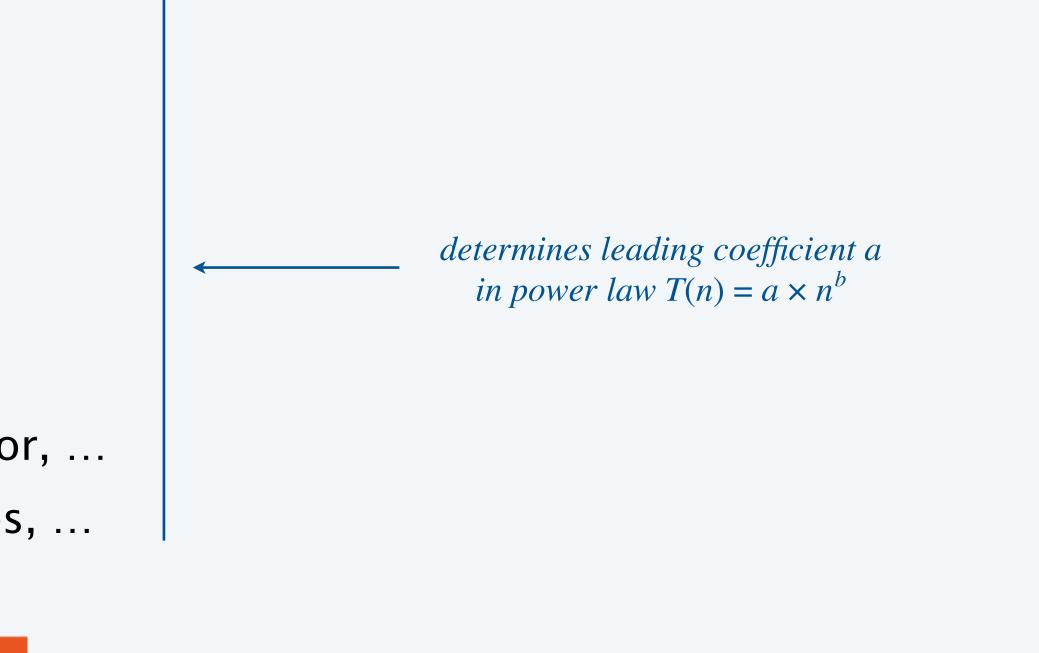
determines exponent b in power law $T(n) = a \times n^b$

System dependent effects.

- Hardware: CPU, memory, cache, ...
- Software: compiler, interpreter, garbage collector, ...
- System: operating system, network, other apps, ...



Bad news. Sometimes difficult to get accurate measurements.





Context: the scientific method

Experimental algorithmics is an example of the scientific method.



Chemistry (1 experiment)





Computer Science (1 million experiments)

Biology (1 experiment)

Good news. Experiments are easier and cheaper than other sciences.



Physics (1 experiment)





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Mathematical models for running time

Total running time: sum of frequency × cost for all operations.



Warning. No general-purpose method (e.g.,



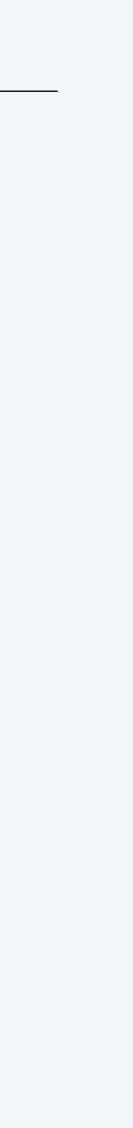
Q. How many operations as a function of input size *n*?

operation	cost (ns) †	frequency
variable declaration	2/5	2
assignment statement	1/5	2
less than compare	1/5	<i>n</i> + 1
equal to compare	1/10	n
array access	1/10	п
increment	1/10	<i>n</i> to 2 <i>n</i>

† representative estimates (with some poetic license)

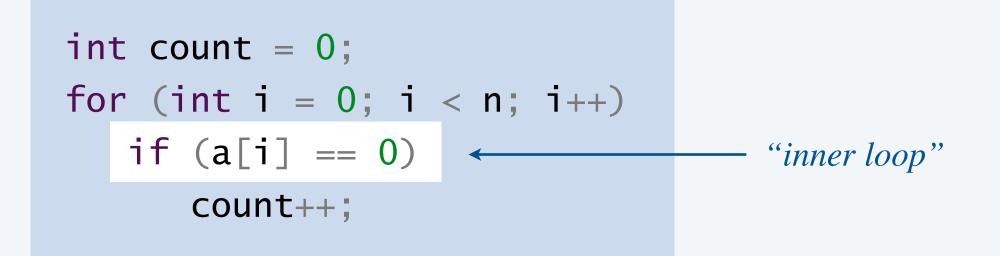
in practice, depends on caching, bounds checking, ... (see COS 217)

tedious to count exactly



Simplification 1: cost model

Cost model. Use some elementary operation as a proxy for running time. *API calls*, *API calls*,



operation	cost (ns) †	frequency
variable declaration	2/5	2
assignment statement	1/5	2
less than compare	1/5	<i>n</i> + 1
equal to compare	1/10	n
array access	1/10	n 🔸
increment	1/10	<i>n</i> to 2 <i>n</i>

floating-point operations, ...

- *cost model* = *array accesses*

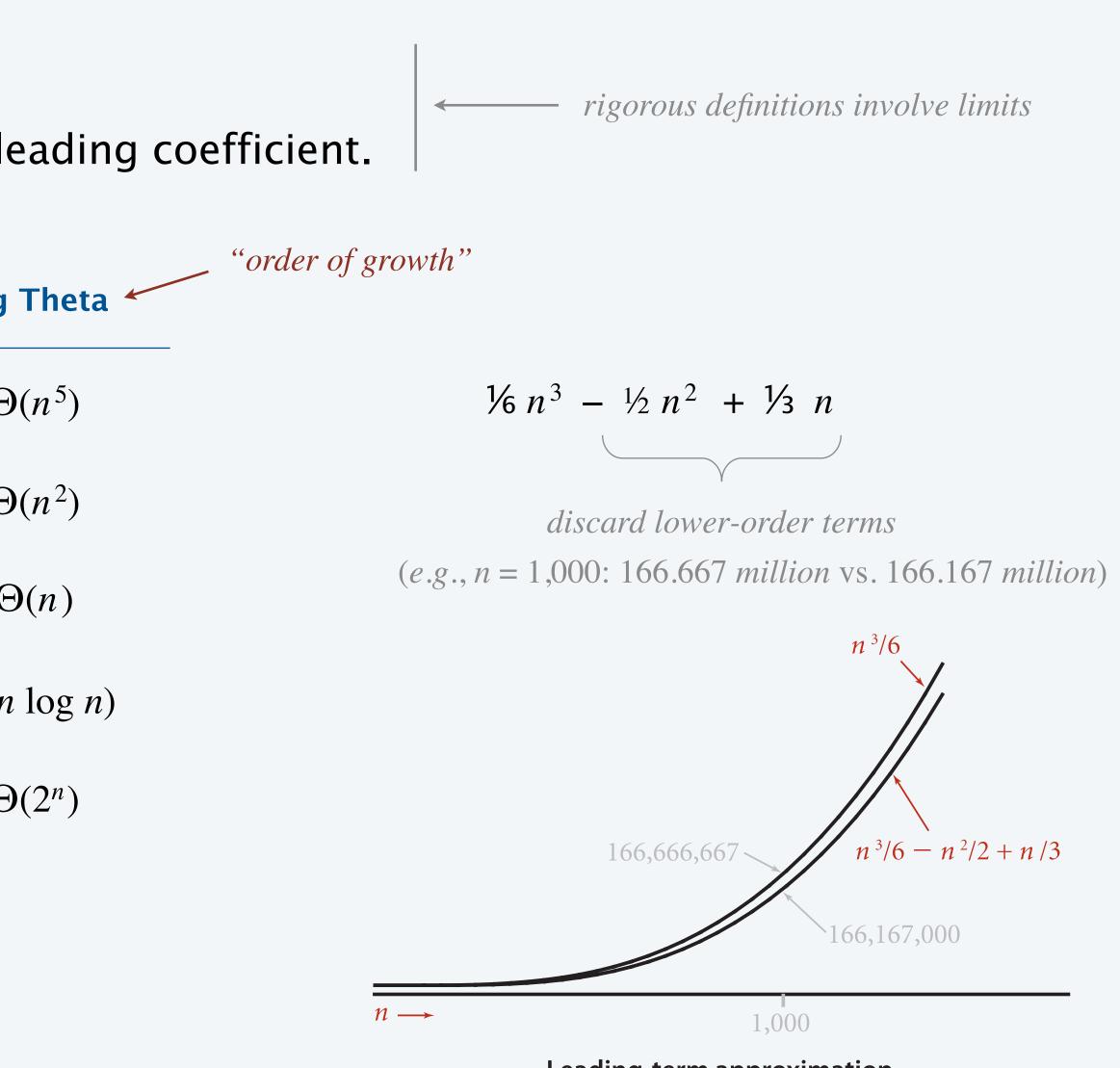
Simplification 2: asymptotic notations

Tilde notation.Discard lower-order terms.Big Theta notation.Discard lower-order terms and leading coefficient.

function	tilde notation	big
$4 n^5 + 20 n^3 + 16$	$\sim 4 n^5$	Θ
$0.01 n^2 + 100 n^{4/3} - 56$	$\sim 0.01 \ n^2$	Θ
$8 \log^2(n) + 7 n$	~ 7 n	$oldsymbol{arepsilon}$
$10 n + 3 n \log n$	~ 3 <i>n</i> log <i>n</i>	$\Theta(n$
$2^{n} + n^{5}$	$\sim 2^n$	Θ

Rationale.

- When *n* is large, lower-order terms are negligible.
- When *n* is small, we don't care.



Leading-term approximation



Analysis of algorithms: quiz 2

How many array accesses as a function of n?

A.
$$\frac{1}{2}n(n-1)$$

B.
$$n(n-1)$$

- **C.** $2 n^2$
- **D.** 2n(n-1)

"inner loop"



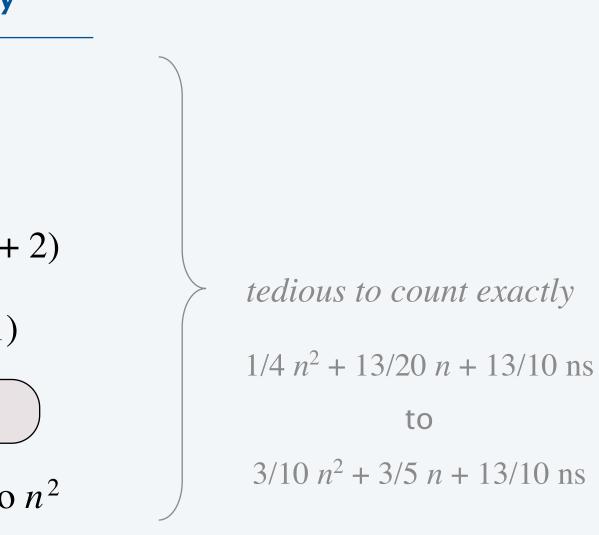


Q. How many operations as a function of input size *n*?

operation	cost (ns) †	frequency
variable declaration	2/5	<i>n</i> + 2
assignment statement	1/5	<i>n</i> + 2
less than compare	1/5	$\frac{1}{2}(n+1)(n+1)$
equal to compare	1/10	$\frac{1}{2} n (n-1)$
array access	1/10	(<i>n</i> (<i>n</i> – 1)
increment	1/10	$\frac{1}{2} n (n + 1)$ to

$$(n-1) + (n-2) + (n-3) \dots + 2 + 1 + 0$$

$$\frac{1}{2} n (n-1)$$



Example: 2-SUM

Q. Approximately how many operations as a function of input size *n*?

operation	cost (ns) †	frequency
variable declaration	2/5	$\Theta(n)$
assignment statement	1/5	$\Theta(n)$
less than compare	1/5	$\Theta(n^2)$
equal to compare	1/10	$\Theta(n^2)$
array access	1/10	$\Theta(n^2)$
increment	1/10	$\Theta(n^2)$

$$\underbrace{(n-1) + (n-2) + (n-3) \dots + 2 + 1 + 0}_{\frac{1}{2} n (n-1)}$$



Example: 3-SUM

Q. Approximately how many array accesses as a function of input size *n*?

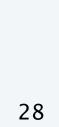
A1.
$$\sim \frac{1}{2}n^3$$
 array accesses.
A2. $\Theta(n^3)$ array accesses.

Bottom line. Use cost model and asymptotic notation to simplify analysis.

$$-\binom{n}{3} = \frac{n(n-1)(n-2)}{3!} \sim \frac{1}{6}n^3$$
see cos 240

Common order-of-growth classifications

order of growth	emoji	name	typical code framework	description	example
$\Theta(1)$		constant	a = b + c;	statement	add two numbers
$\Theta(\log n)$		logarithmic	for (int i = n; i > 0; i /= 2) { }	divide in half	binary search
$\Theta(n)$		linear	for (int i = 0; i < n; i++) { }	single loop	find the maximum
$\Theta(n \log n)$		linearithmic	mergesort	divide and conquer	mergesort
$\Theta(n^2)$		quadratic	for (int i = 0; i < n; i++) for (int j = 0; j < n; j++) { }	double loop	check all pairs
$\Theta(n^3)$		cubic	<pre>for (int i = 0; i < n; i++) for (int j = 0; j < n; j++) for (int k = 0; k < n; k++) { }</pre>	triple loop	check all triples
$\Theta(2^n)$		exponential	towers of Hanoi	exhaustive search	check all subsets



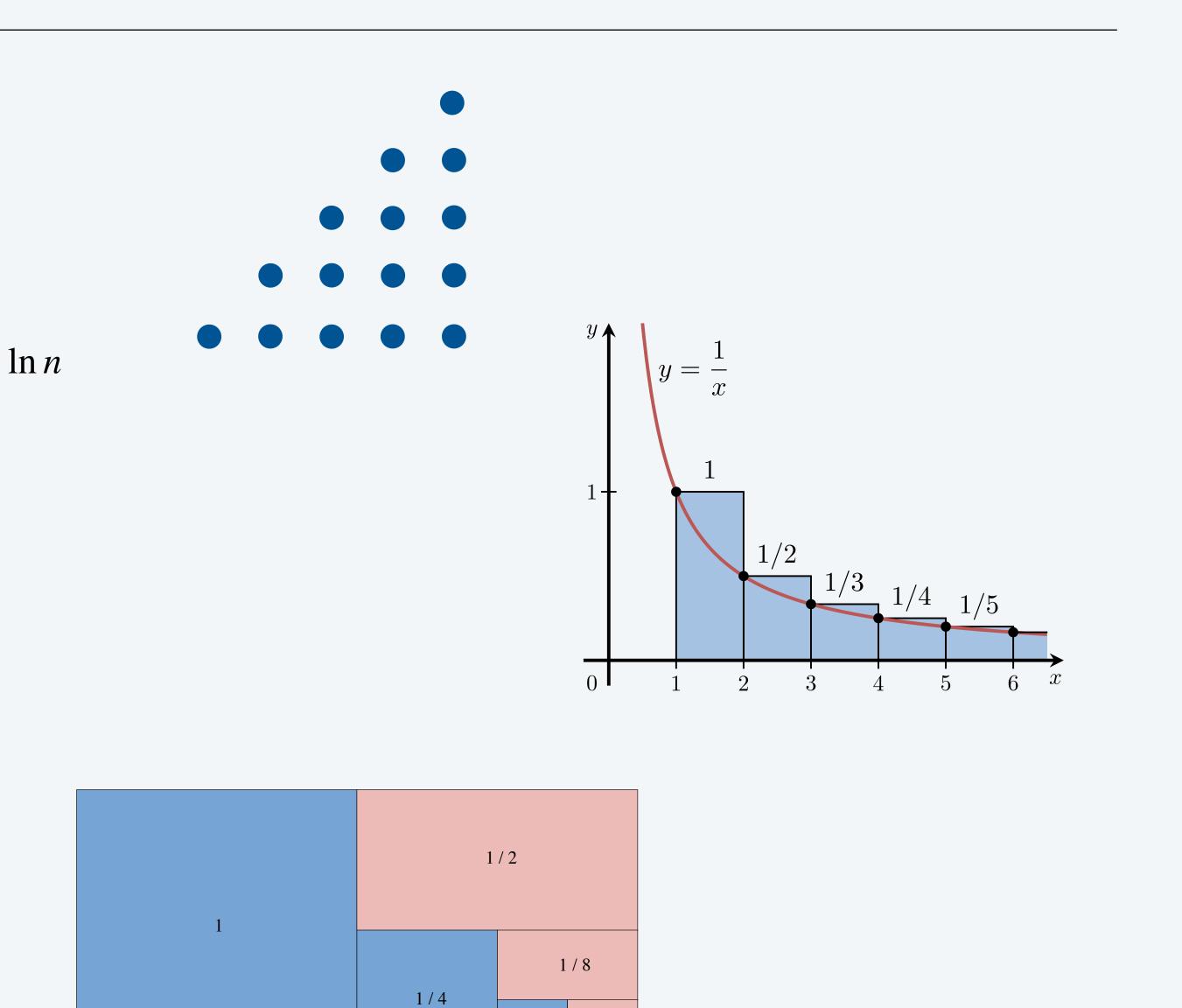
Some useful discrete sums and approximations

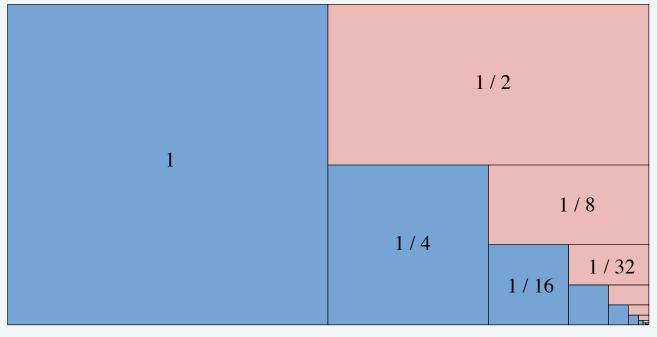
Triangular sum.
$$1 + 2 + 3 + ... + n \sim \frac{1}{2}n^2$$

Harmonic sum.
$$1 + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \sim \int_{x=1}^{n} \frac{1}{x} dx =$$

Geometric sum. 1 + 2 + 4 + 8 + ... + n = 2n - 1n a power of 2

Geometric sum'. $n + \frac{n}{2} + \frac{n}{4} + \dots + 1 = 2n - 1$





Analysis of algorithms: quiz 3

Approximately how many array accesses as a function of *n*?

A. ~
$$n^2 \log_2 n$$

B. ~
$$3/2 n^2 \log_2 n$$

C. ~
$$1/2 n^3$$

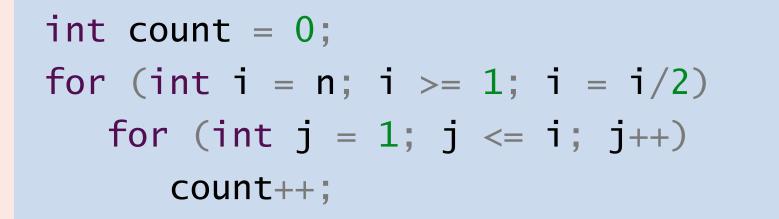
D. ~ $3/2 n^3$





Analysis of algorithms: quiz 4

What is the order of growth of the running time as a function of *n*?



- A. $\Theta(n)$
- **B.** $\Theta(n \log n)$
- **C.** $\Theta(n^2)$
- **D.** $\Theta(2^n)$







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running time (mathematical models)

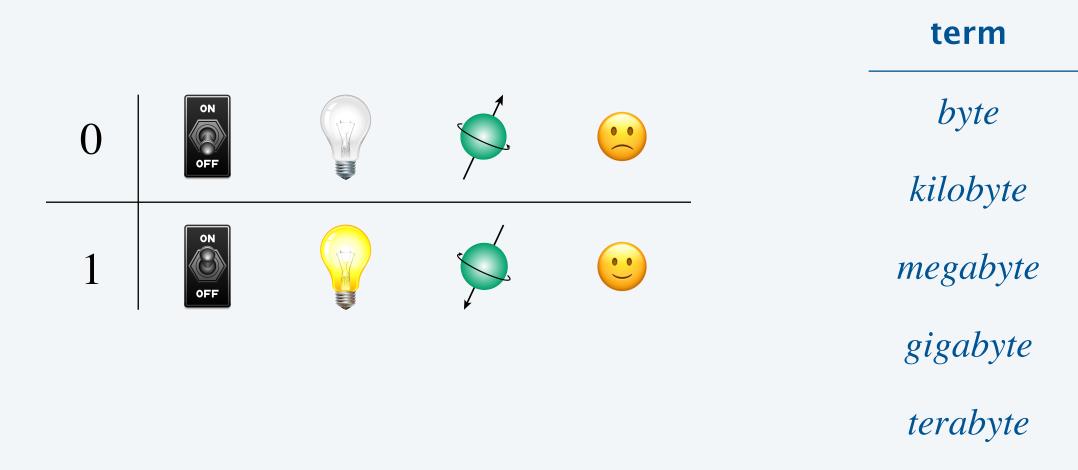
memory usage

introduction



Memory basics

Bit. 0 or 1.

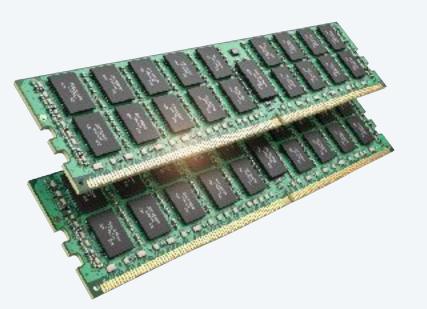


64-bit machine. We assume a 64-bit machine with 8-byte pointers.



some JVMs "compress" pointers to 4 bytes to avoid this cost

symbol	quantity
В	8 bits
KB	1000 bytes
MB	1000^2 bytes
GB	1000^3 bytes
TB	1000^4 bytes



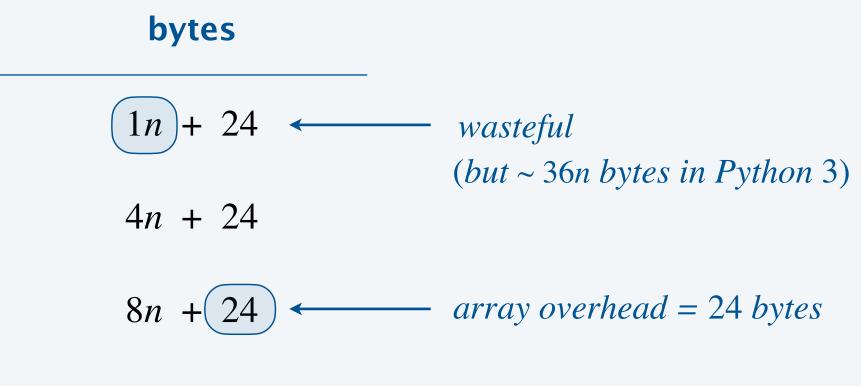
some define using powers of 2 (MB = 2^{10} bytes)

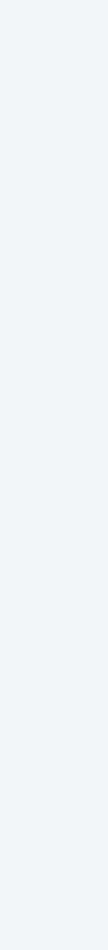
3–byte pointers. ↑

Typical memory usage for primitive types and arrays

type	bytes	type	bytes
boolean	1	boolean[]	<u>1</u> <i>n</i> + 24 ←
byte	1	int[]	4 <i>n</i> + 24
char	2	double[]	8 <i>n</i> + 24 ←
int	4	one-dimensiona	al arrays (length n)
float	4		
long	8		
double	Q	 type	bytes
double 8 primitive types		boolean[][]	$\sim 1 n^2$
		int[][]	$\sim 4 n^2$
		double[][]	$\sim 8 n^2$

two-dimensional arrays (n-by-n)







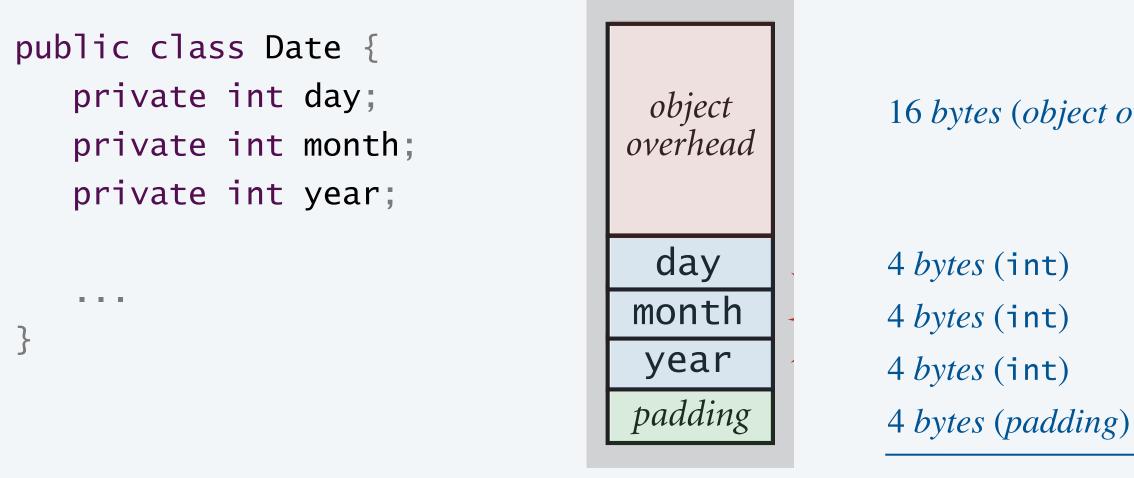
Typical memory usage for objects in Java

Reference. 8 bytes.

Object overhead. 16 bytes.

Padding. Round up memory of each object to be a multiple of 8 bytes.

Ex. Each *Date* object uses 32 bytes of memory.



16 bytes (object overhead)

32 bytes

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

Α.	~4 <i>n</i>	bytes
Β.	~ 8 n	bytes
С.	$\sim 4 n^2$	bytes
D.	$\sim 8 n^2$	bytes

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;
for (int i = 0; i < n; i++)
    parent[i] = i;
for (int i = 0; i < n; i++)
    size[i] = 1;
}
```

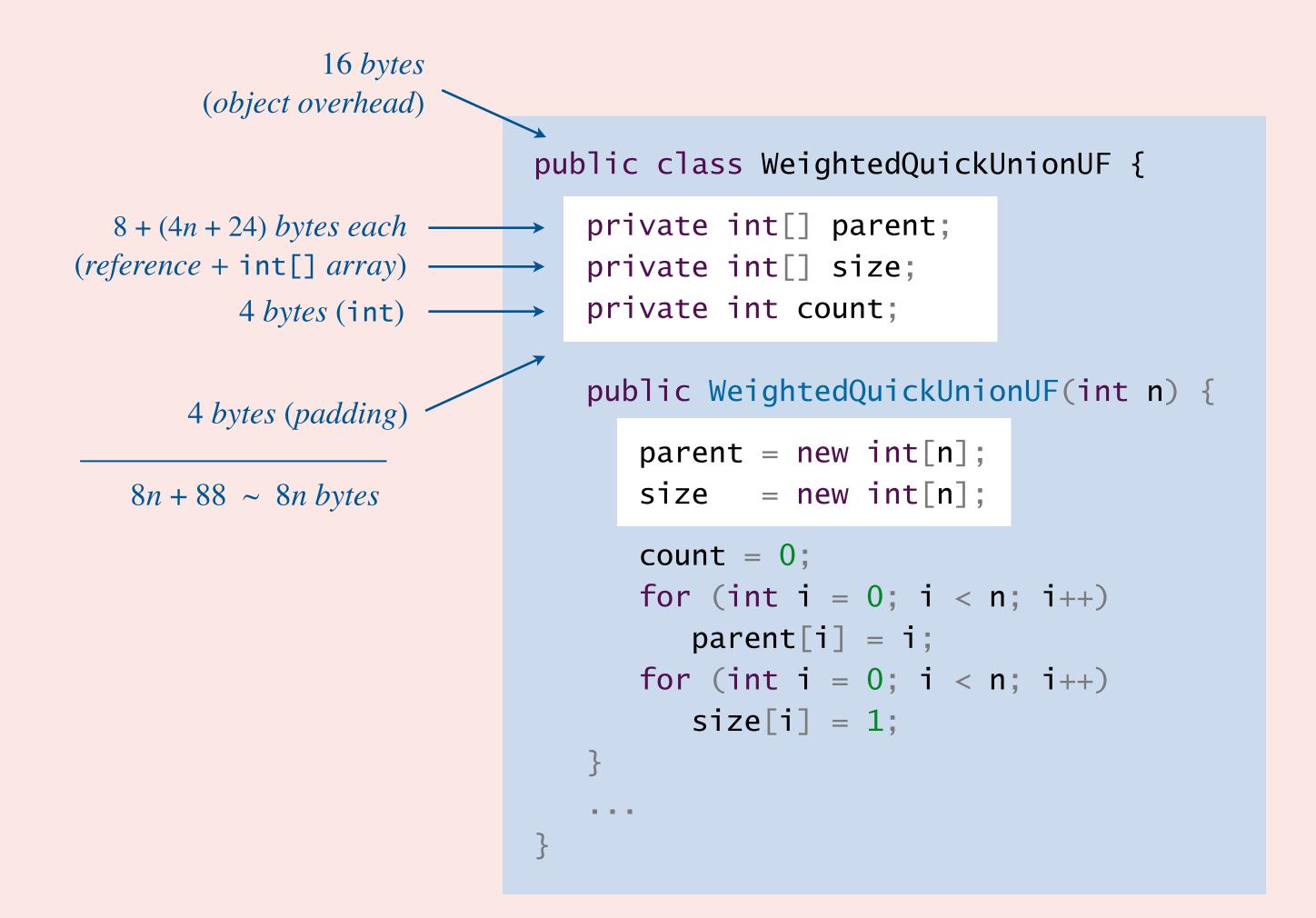






Analysis of algorithms: quiz 5

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







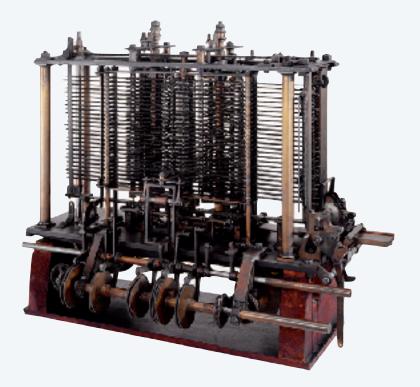
Empirical analysis.

- Execute program to perform experiments.
- Assume power law.
- Formulate a hypothesis for running time.
- Model enables us to make predictions.

Mathematical analysis.

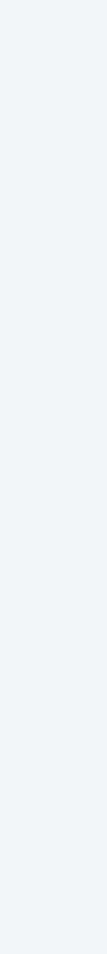
- Analyze algorithm to count frequency of operations.
- Use cost model and asymptotic notations to simplify analysis.
- Model enables us to explain behavior.

This course. Learn to use both.



ns. Iify analysis

$$\sum_{h=0}^{\lfloor \lg n \rfloor} \left\lceil n/2^{h+1} \right\rceil h \sim n$$



Credits

image

Charles Babbage

Babbage Enginine in Operation Algorithm for the Analytic Engine Ada Lovelace and Book Galaxies Colliding Andrew Appel Programmer Icon Head in the Clouds Student Raising Hand Running Time Analog Stopwatch

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"It is convenient to have a measure of the amount of work involved in a computing process, even though it be a very crude one. We may count up the number of times that various elementary operations are applied in the whole process and then give them various weights." — Alan Turing (1947)

