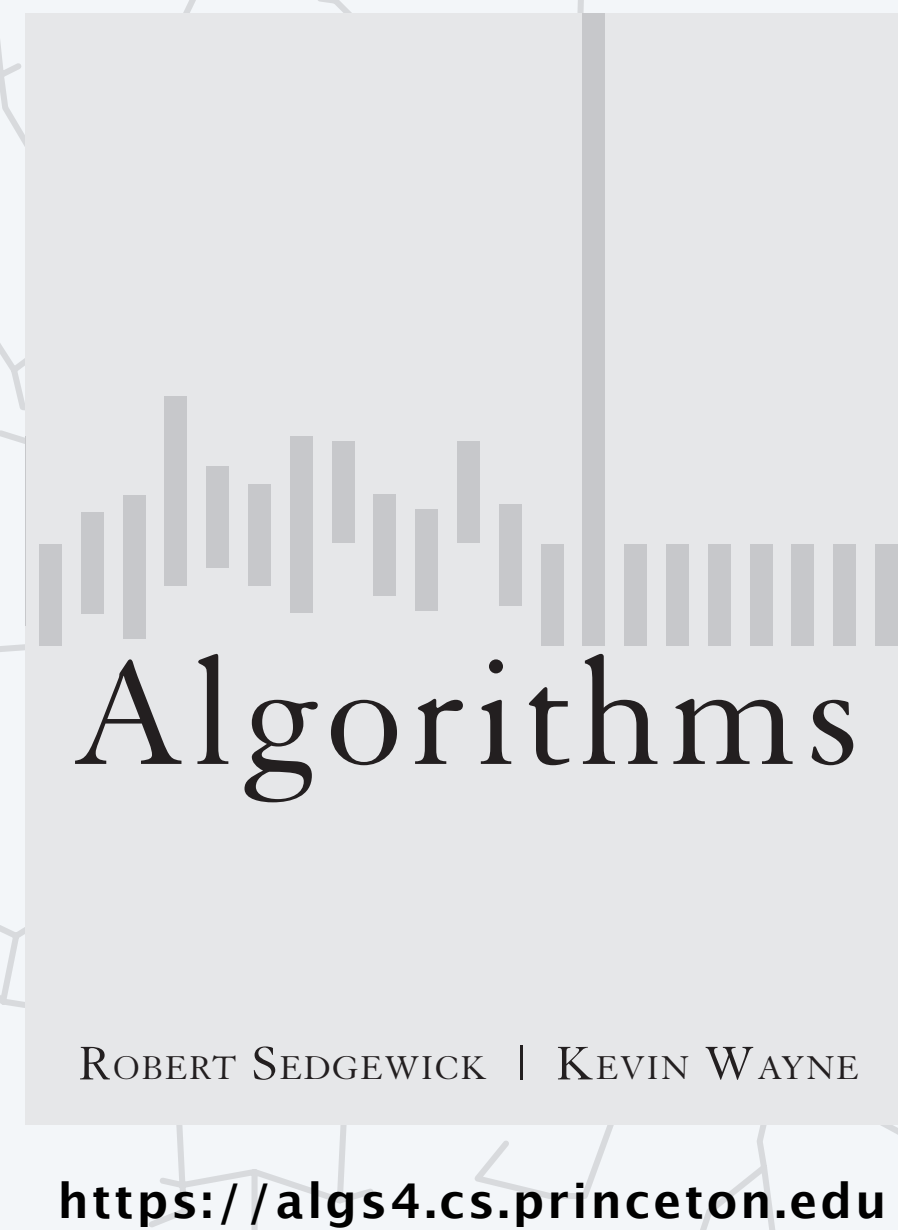




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1.4 ANALYSIS OF ALGORITHMS

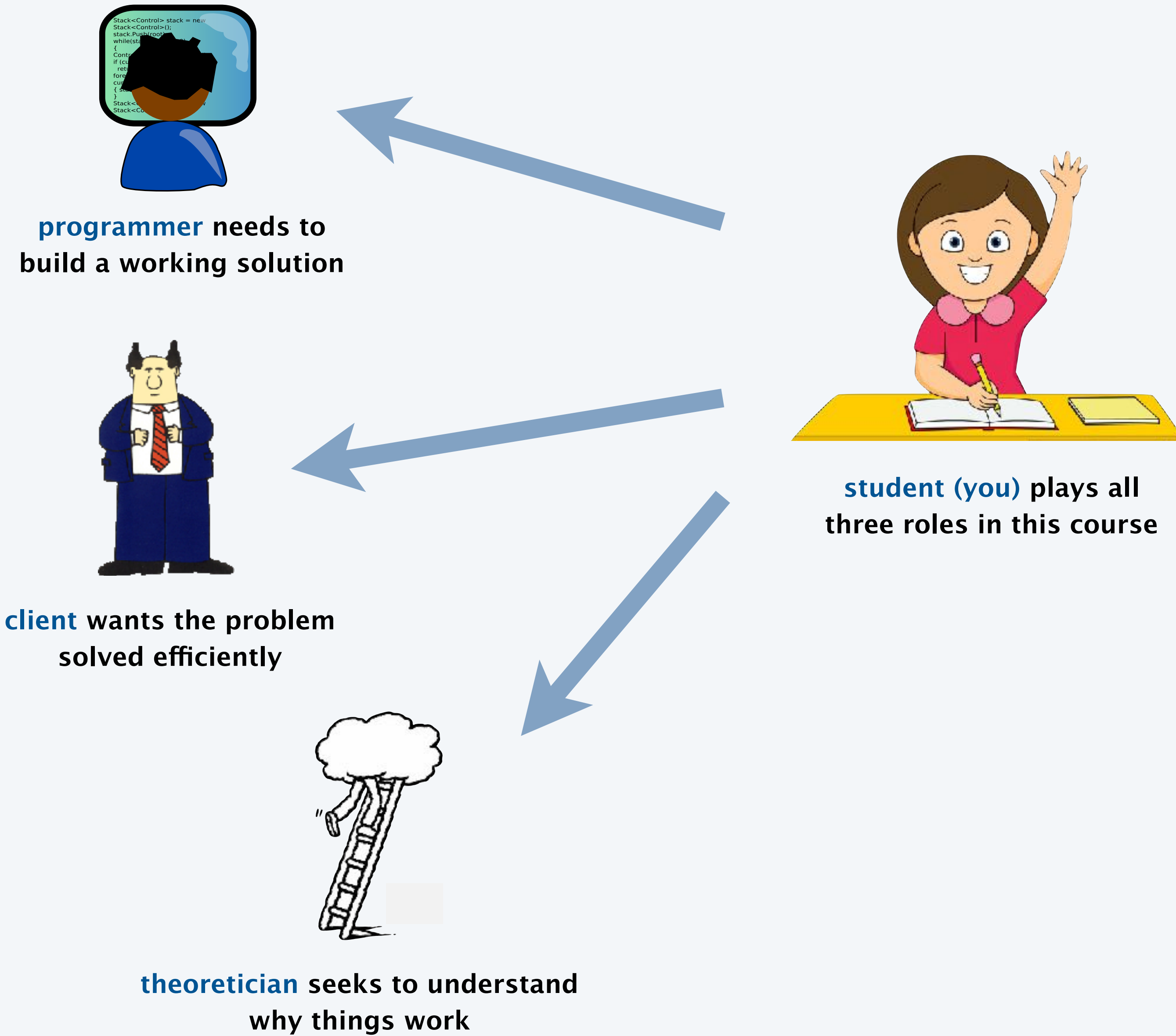
- *introduction*
- *running time (experimental analysis)*
- *running time (mathematical models)*
- *memory usage*



1.4 ANALYSIS OF ALGORITHMS

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

Different viewpoints

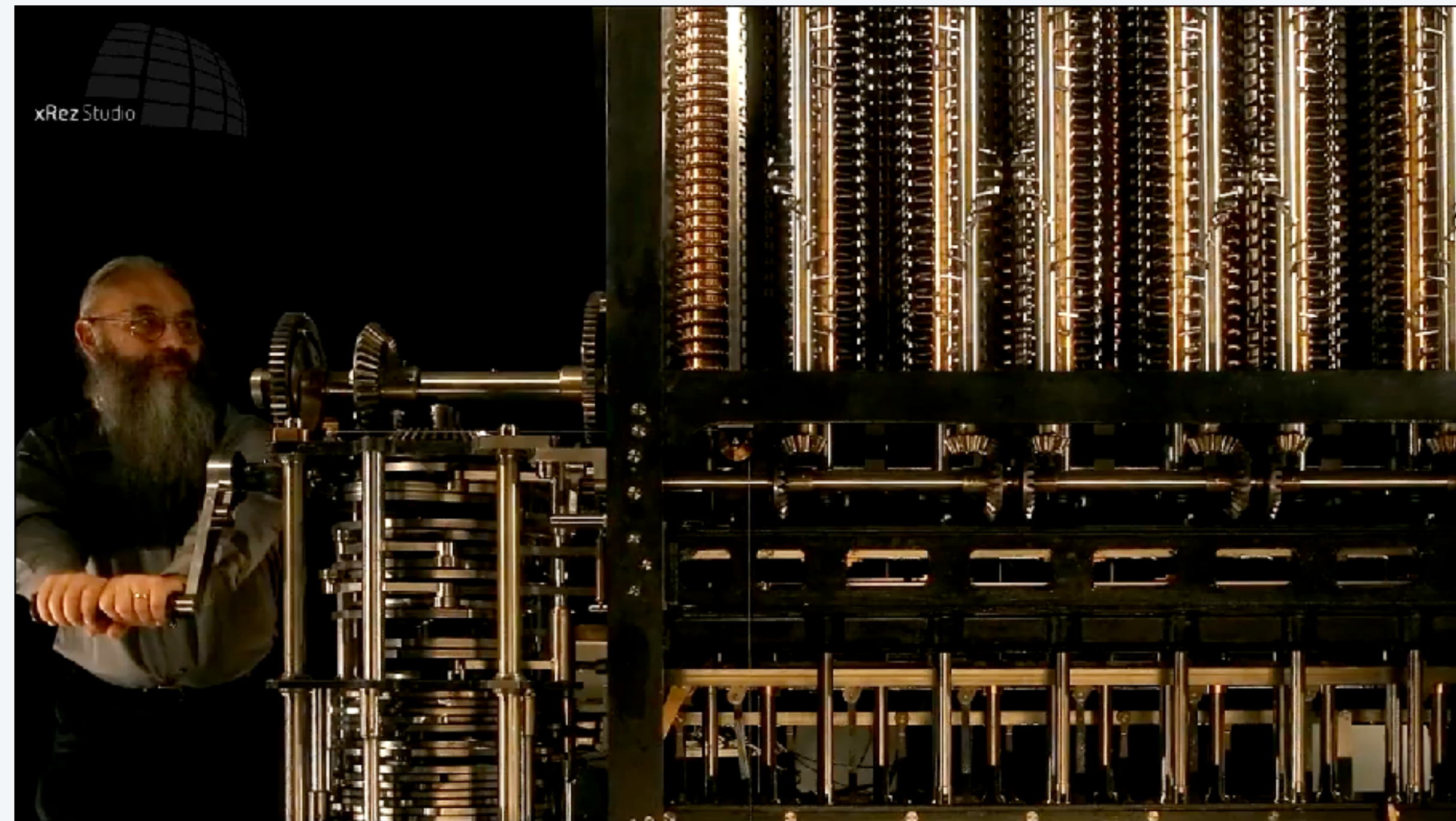


Running time

*“ 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

*“ 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)*



Diagram for the comparison by the Rule of the Numbers of Demolish. See Note G, page 755 of text.

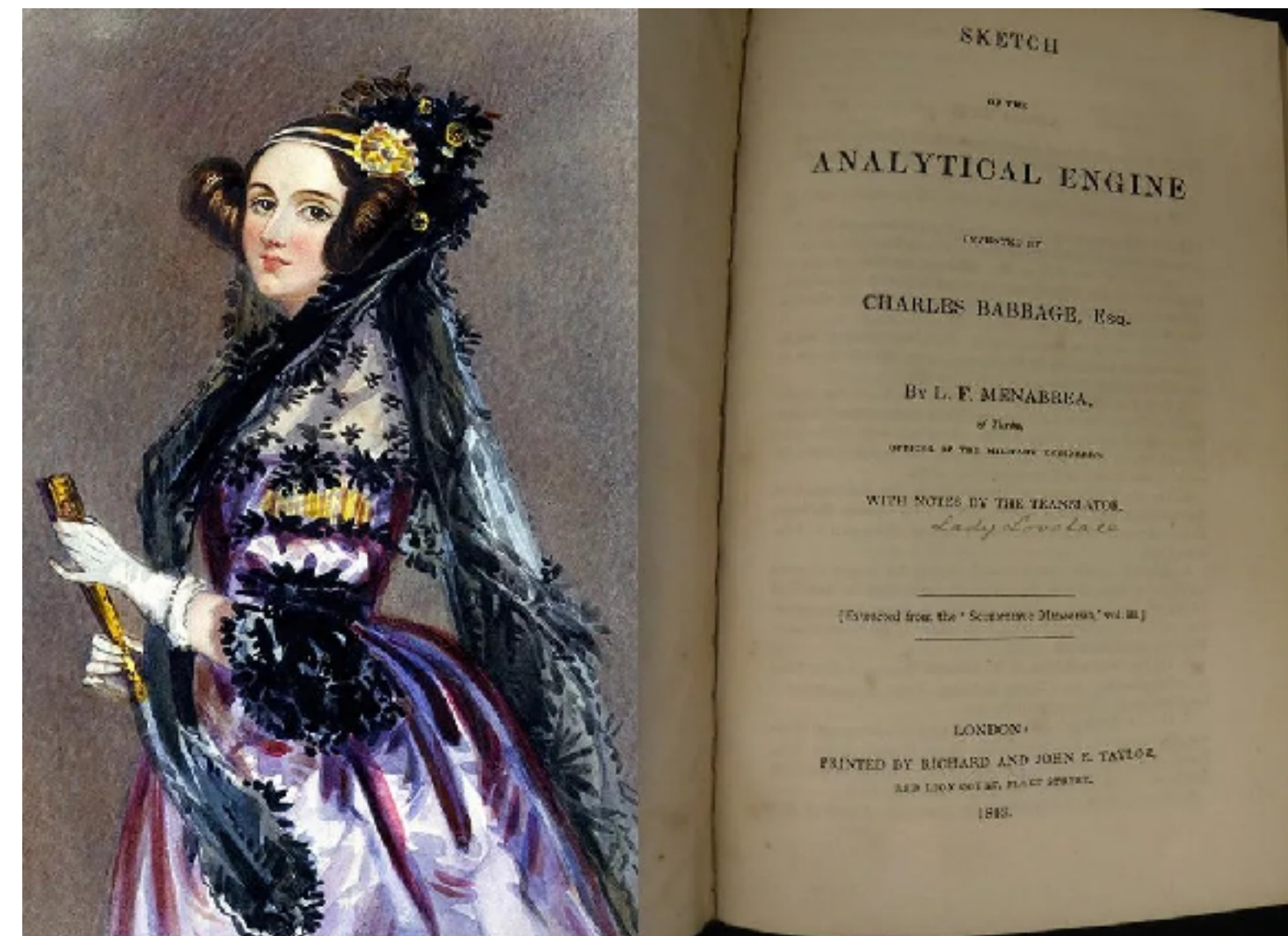
Number of Operations	Nature of Operations	Variables actually involved	Variables actually involved	Set of changes in the value of each Variable	Statement of Result	Working Variables																Result Variables							
						U_1	U_2	U_3	U_4	U_5	U_6	U_7	U_8	U_9	U_{10}	U_{11}	U_{12}	U_{13}	U_{14}	U_{15}	U_{16}	U_{17}	U_{18}	U_{19}	U_{20}				
1	$U_1 = U_2$	U_1	U_2	$U_1 = U_2$	$U_1 = U_2$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
2	$U_1 = U_3$	U_1	U_3	$U_1 = U_3$	$U_1 = U_3$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
3	$U_1 = U_4$	U_1	U_4	$U_1 = U_4$	$U_1 = U_4$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
4	$U_1 = U_5$	U_1	U_5	$U_1 = U_5$	$U_1 = U_5$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
5	$U_1 = U_6$	U_1	U_6	$U_1 = U_6$	$U_1 = U_6$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
6	$U_1 = U_7$	U_1	U_7	$U_1 = U_7$	$U_1 = U_7$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
7	$U_1 = U_8$	U_1	U_8	$U_1 = U_8$	$U_1 = U_8$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
8	$U_1 = U_9$	U_1	U_9	$U_1 = U_9$	$U_1 = U_9$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
9	$U_1 = U_{10}$	U_1	U_{10}	$U_1 = U_{10}$	$U_1 = U_{10}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
10	$U_1 = U_{11}$	U_1	U_{11}	$U_1 = U_{11}$	$U_1 = U_{11}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
11	$U_1 = U_{12}$	U_1	U_{12}	$U_1 = U_{12}$	$U_1 = U_{12}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
12	$U_1 = U_{13}$	U_1	U_{13}	$U_1 = U_{13}$	$U_1 = U_{13}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
13	$U_1 = U_{14}$	U_1	U_{14}	$U_1 = U_{14}$	$U_1 = U_{14}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
14	$U_1 = U_{15}$	U_1	U_{15}	$U_1 = U_{15}$	$U_1 = U_{15}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
15	$U_1 = U_{16}$	U_1	U_{16}	$U_1 = U_{16}$	$U_1 = U_{16}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
16	$U_1 = U_{17}$	U_1	U_{17}	$U_1 = U_{17}$	$U_1 = U_{17}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
17	$U_1 = U_{18}$	U_1	U_{18}	$U_1 = U_{18}$	$U_1 = U_{18}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
18	$U_1 = U_{19}$	U_1	U_{19}	$U_1 = U_{19}$	$U_1 = U_{19}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
19	$U_1 = U_{20}$	U_1	U_{20}	$U_1 = U_{20}$	$U_1 = U_{20}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				
20	$U_1 = U_{21}$	U_1	U_{21}	$U_1 = U_{21}$	$U_1 = U_{21}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20				

Here follows a summary of Operations taken to be typical.

21	$U_1 = U_{22}$	U_1	U_{22}	$U_1 = U_{22}$	$U_1 = U_{22}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
22	$U_1 = U_{23}$	U_1	U_{23}	$U_1 = U_{23}$	$U_1 = U_{23}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
23	$U_1 = U_{24}$	U_1	U_{24}	$U_1 = U_{24}$	$U_1 = U_{24}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
24	$U_1 = U_{25}$	U_1	U_{25}	$U_1 = U_{25}$	$U_1 = U_{25}$	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20

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

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



An algorithmic success story

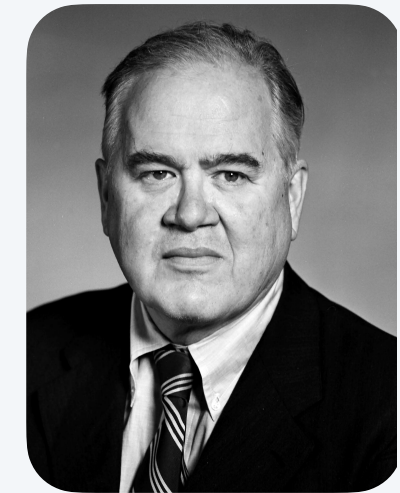
Goal. Multiply two polynomials of degree n .

$$(x^3 + x^2 - 2x + 1) \cdot (3x^3 - x^2 + 2x + 1) = 3x^6 + 2x^5 - 5x^4 + 8x^3 - 4x^2 + 1$$

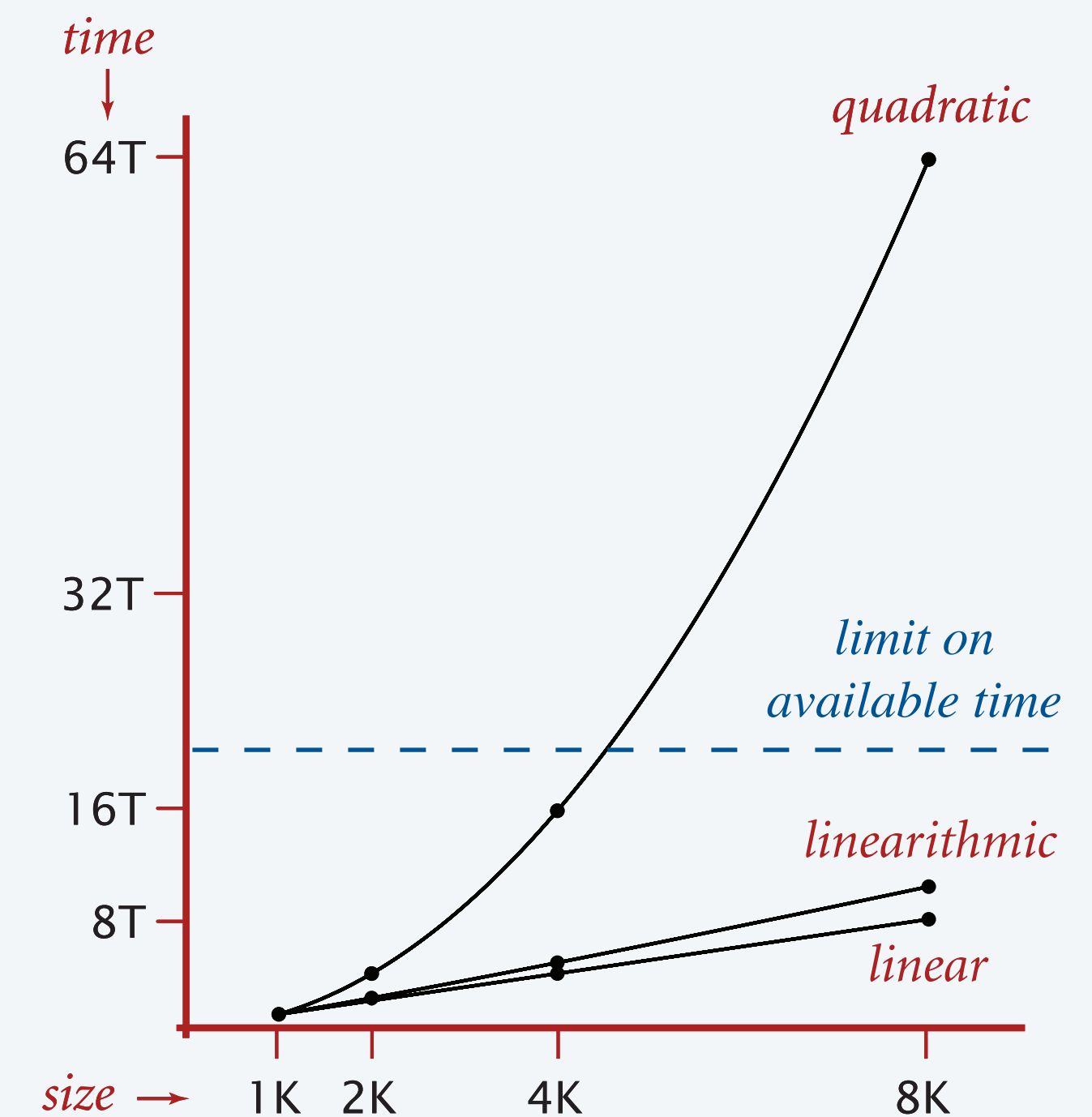
- Applications: JPEG compression, MRI, astrophysics, and more.
- Grade-school algorithm: $\Theta(n^2)$ operations.
- FFT algorithm: $\Theta(n \log n)$ operations, **enabling modern technology**.



James
Cooley



John
Tukey



Another algorithmic success story?

CNBC

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TECH

POLITICS & POLICY

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INVESTING CLUB

PRO

LIVESTREAM

MAKE IT

TECH

Chinese AI models are gaining ground with U.S. companies as OpenAI, Anthropic costs surge

PUBLISHED TUE, JUL 7 2026 1:00 AM EDT | UPDATED TUE, JUL 7 2026 5:30 AM EDT



Kai Nicol-Schwarz

@IN/KAINS

SHARE    

KEY POINTS

- U.S. companies are increasingly using Chinese-built models as they build out tools and products using AI.
- Recent model releases from Chinese companies are seen by many as highly competitive compared to leading U.S. frontier systems.
- Companies are hunting for cheaper model alternatives as they grapple with unexpectedly high costs related to AI.

Chinese-built AI models are gaining traction among U.S. companies as they narrow the performance gap with leading American rivals while remaining significantly cheaper to use.

Recent model releases from Chinese companies, including DeepSeek and Z.ai,

WATCH LIVESTREAM

Prefer to Listen?

NOW

UP NEXT

Squawk on the S

Halftime R

TRENDING NOW



1

Apple event live up
First foldable iPhor
new watch series
expected as Ternus
makes debut



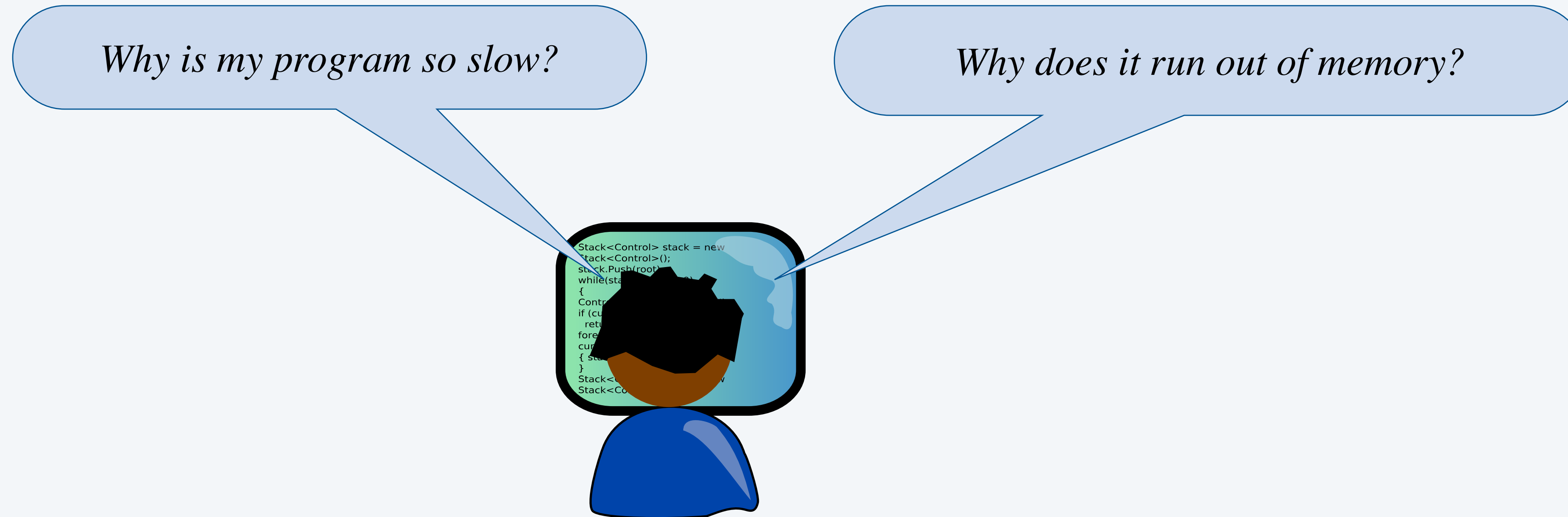
2

Treasury Departme
buy back up to \$6 l
in longer-term debt
the normal level

7

The core challenge

- Q1. Will my program handle on large, real-world inputs?
- Q2. If not, how can I analyze and improve its performance?



Our approach: a combination of **experiments** and **mathematical modeling**.

Example: 3-SUM problem



Goal. Given an array of n distinct integers, count triples $i < j < k$ such that $a[i] + a[j] + a[k] = 0$.

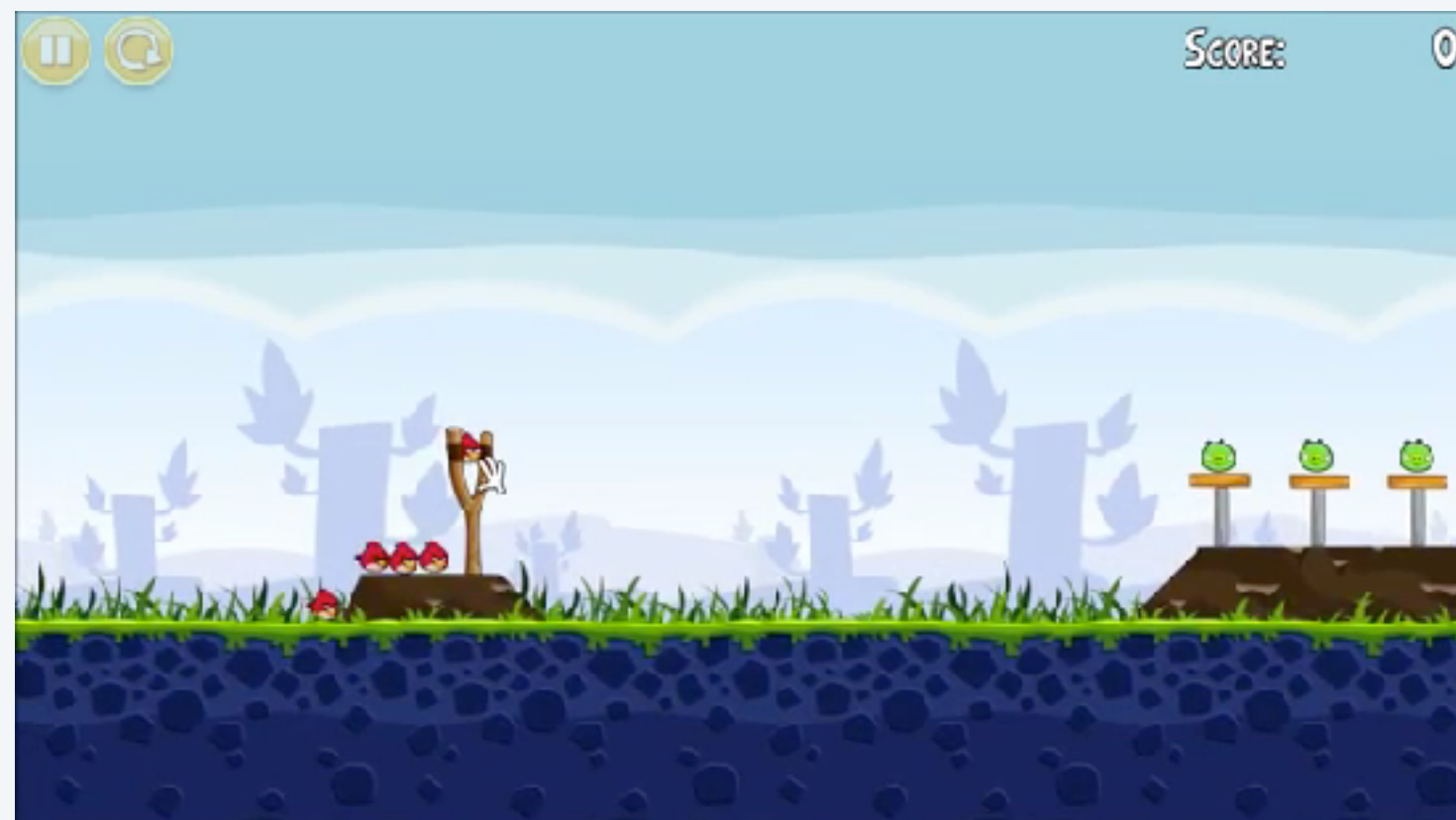
```
~/cos226/analysis> more 8ints.txt
8
30 -40 -20 -10 40 0 10 5

~/cos226/analysis> java ThreeSum 8ints.txt
4
```

	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	✓

Context. Arises in computational geometry (and even in computer games!)

Open problem. What is the optimal running time for solving 3-SUM ?



3-SUM problem: 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));
```

```
    }
```

```
}
```

← *count distinct triples that sum to 0*

← *assume no integer overflow*



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1.4 ANALYSIS OF ALGORITHMS

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

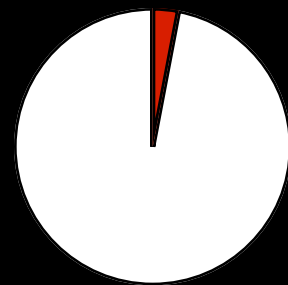


Measuring running time

Experiment. Measure the program's **running time** on inputs of different sizes.

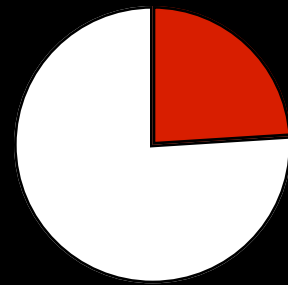
Observation. The running time $T(n)$ increases with the input size n .

```
~/cos226/analysis> java ThreeSum 1Kints.txt  
70
```



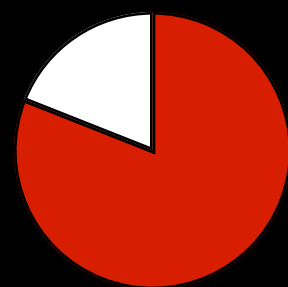
tick tick

```
~/cos226/analysis> java ThreeSum 2Kints.txt  
528
```



*tick tick tick tick tick tick tick tick
tick tick tick tick tick tick tick tick*

```
~/cos226/analysis> java ThreeSum 3Kints.txt  
1670
```



*tick tick tick tick tick tick tick tick tick tick tick tick tick tick
tick tick tick tick tick tick tick tick tick tick tick tick tick tick tick
tick tick tick tick tick tick tick tick tick tick tick tick tick tick tick
tick tick tick tick tick tick tick tick tick tick tick tick tick tick tick
tick tick tick tick tick tick*



Measuring running time

Experiment. Measure the program's **running time** on inputs of different sizes.

Observation. The running time $T(n)$ increases with the input size n .

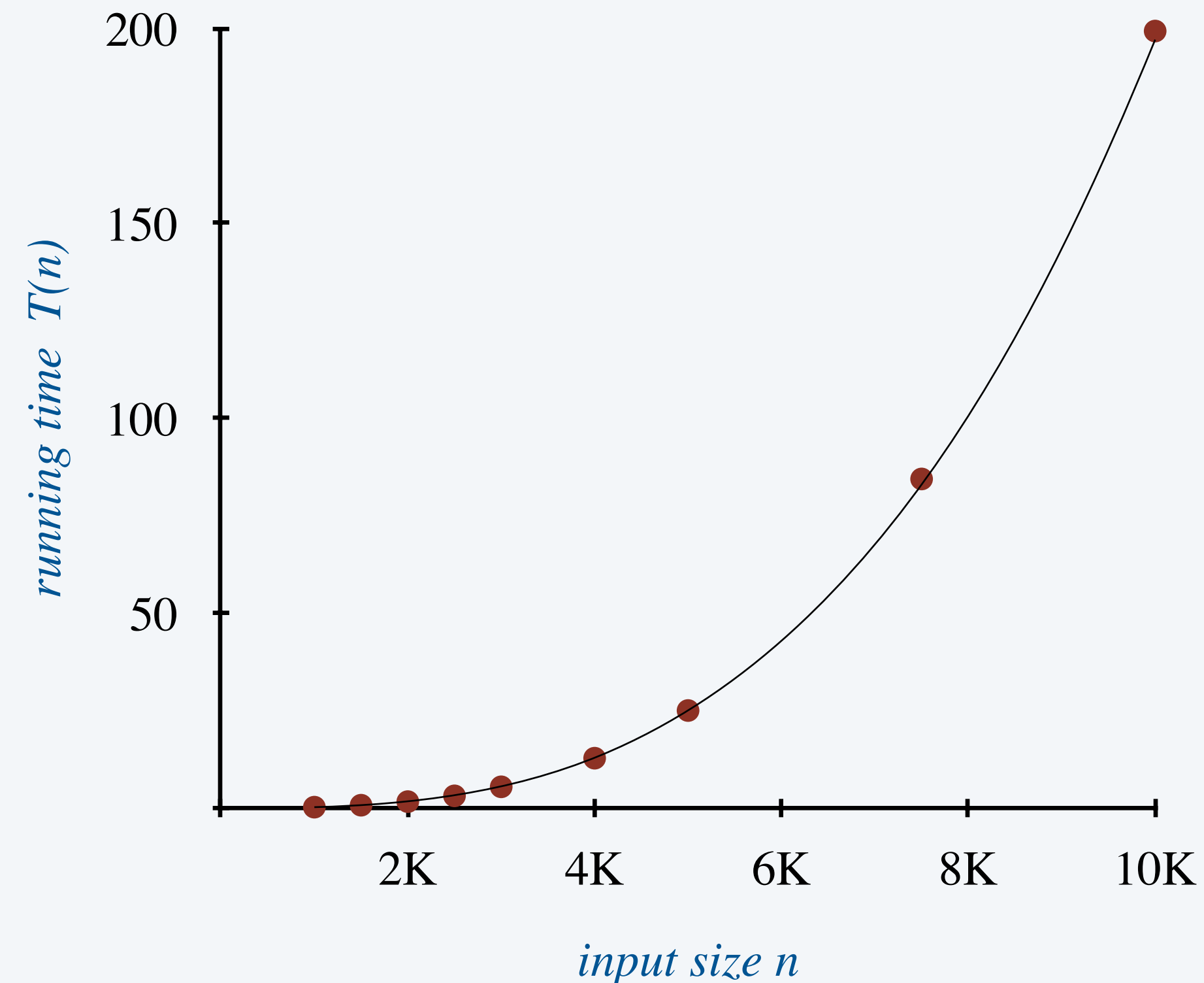
n	time (seconds) †
1,000	0.21
1,500	0.71
2,000	1.63
2,500	3.11
3,000	5.43
4,000	12.8
5,000	25.0
7,500	84.4
10,000	199.3



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

Data analysis: running time vs. input size

Visualization. Plot the running time $T(n)$ versus the input size n .



Hypothesis. The running time follows a **power law**: $T(n) = a \times n^b$ seconds.

Questions. How can we test this hypothesis? How can we estimate a and b ?

Doubling test: estimating the exponent b

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

- Assume running time satisfies $T(n) = a \times n^b$.
- Estimate b as $\log_2(\text{ratio})$.

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 $b \approx 3.0$*

$$\frac{T(n)}{T(n/2)} = \frac{an^b}{a(n/2)^b} = 2^b$$
$$\implies b = \log_2 \frac{T(n)}{T(n/2)}$$

why the log₂(ratio) works

Doubling test: estimating the leading constant a

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

- Assume running time satisfies $T(n) = a \times n^b$.
- Estimate b as $\log_2(\text{ratio})$.
- Estimate a from $T(n) = a \times n^b$ using 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	$819.0 = a \times 16,000^3 \implies a = 2.00 \times 10^{-10}$

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



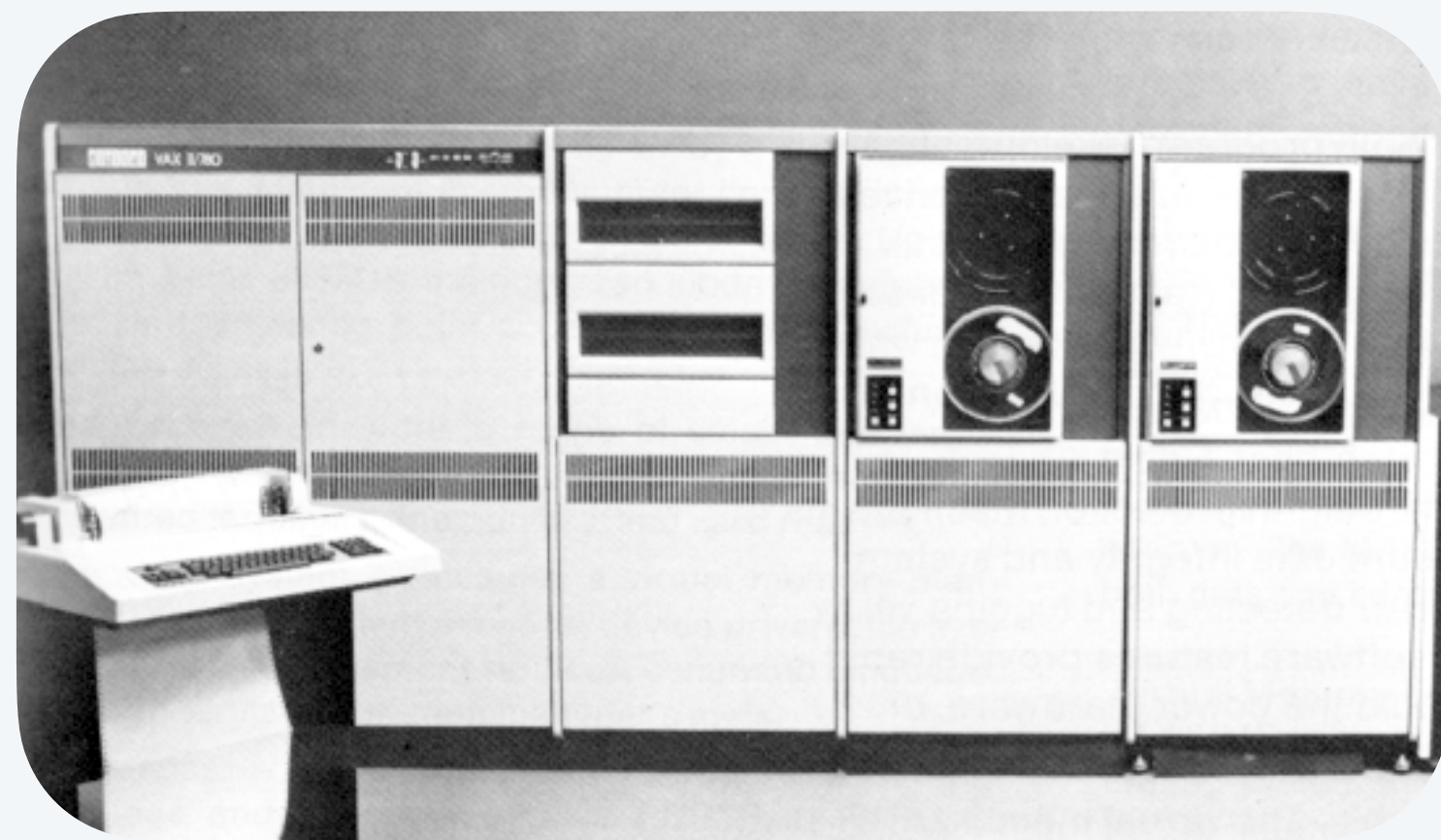
Estimate the running time when $n = 96,000$ (assuming power law hypothesis).

	n	time (seconds)
A.	39 seconds	
	1,000	0.02
B.	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

Machine invariance

Hypothesis. For a fixed algorithm, the running times on different computers are the same up to a constant factor.

Note. That constant factor can be large, sometimes several orders of magnitude.



1970s
(VAX-11/780)



2020s
(Macbook Pro M2)

What affects the running time?

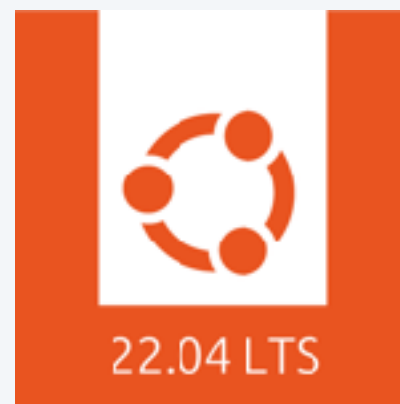
System independent effects.

- 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, ...

← *determines leading coefficient a
in power law $T(n) = a \times n^b$*



Bad news. Getting accurate timing measurements can be difficult. ← *system-dependent effects
can introduce noise*

Context: the scientific method



Experimental algorithmics follows the **scientific method**: observe, hypothesize, experiment, repeat.



Chemistry
(an expensive experiment)



Biology
(a slow experiment)



Computer Science
(millions of cheap, fast, and safe experiments)



Physics
(a dangerous experiment)

Good news. Computational experiments are cheap, fast, and safe.



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1.4 ANALYSIS OF ALGORITHMS

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Mathematical model of running time

Model. The running time = $\sum$ (frequency of operation) $\times$ (cost of operation).

- **Frequency of operation:** depends on the algorithm and the specific input.
- **Cost of operation:** depends on hardware, software, system, and low-level implementation details.



Warning. For arbitrary programs, frequencies may be impossible to determine. $\longleftarrow$ *halting problem*

Example: one-sum problem

Q. How many operations does this code perform as a function of the input size n ?

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

operation	cost (ns) †	frequency
variable declaration	2 / 5	2
assignment statement	1 / 5	2
less than compare	1 / 5	$n + 1$
equality test	1 / 10	n
array read	1 / 10	n
increment	1 / 10	n to $2n$

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

painful to count exactly

† representative estimates (with a bit of poetic license)

Simplification 1: cost model

Cost model. Pick one elementary operation as a **proxy** for running time. ← *array accesses, compares, API calls, floating-point operations, ...*

```
int count = 0;
for (int i = 0; i < n; i++)
    if (a[i] == 0) ← “inner loop”
        count++;
```

operation	cost (ns) †	frequency
<i>variable declaration</i>	2 / 5	2
<i>assignment statement</i>	1 / 5	2
<i>less than compare</i>	1 / 5	$n + 1$
<i>equality test</i>	1 / 10	n
<i>array read</i>	1 / 10	n ← <i>cost model = array accesses</i>
<i>increment</i>	1 / 10	n to $2 n$

Simplification 2: asymptotic notation

Tilde notation. Ignore lower-order terms.

Big Theta notation. Ignore both lower-order terms and the leading constant.

← rigorous definitions involve limits

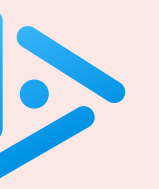
function	tilde notation	big Theta ← “order of growth”
$4 n^5 + 20 n^3 + 16$	$\sim 4 n^5$	$\Theta(n^5)$
$0.01 n^2 + 100 n^{4/3} - 56$	$\sim 0.01 n^2$	$\Theta(n^2)$
$2^n + n^5$	$\sim 2^n$	$\Theta(2^n)$
$\underbrace{\frac{1}{6} n^3 - \frac{1}{2} n^2 + \frac{1}{3} n}_{\text{discard lower-order terms}}$	$\sim \frac{1}{6} n^3$	$\Theta(n^3)$

discard lower-order terms

(e.g., $n = 1,000$: 166.667 million vs. 166.167 million)

Rationale.

- For large n , lower-order terms have negligible effect.
- For small n , the value is so small that we don’t care.



Which of the following correctly describes the function $f(n) = n \log_2 n + 3n^2 + 10n$?

- A. $\sim 10n$
- B. $\sim n \log_2 n$
- C. $\sim n^2$
- D. $\Theta(n \log n)$
- E. $\Theta(n^2)$

Example: 2-SUM analysis

Q. Approximately how many operations 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++)
        if (a[i] + a[j] == 0)
            count++;
```

← “inner loop”

$$\begin{array}{ccccccccc} i=0 & & i=1 & & i=2 & & & & i=n-2 & & i=n-1 \\ j=1, \dots, n-1 & & j=2, \dots, n-1 & & j=3, \dots, n-1 & & & & j=n-1 & & \text{no } j \\ \downarrow & & \downarrow & & \downarrow & & & & \downarrow & & \downarrow \\ (n-1) & + & (n-2) & + & (n-3) & + & \dots & + & 1 & + & 0 \\ \underbrace{\hspace{10em}} & & & & & & & & & & \\ & & & & & & = & \frac{n(n-1)}{2} \end{array}$$

Step 1. Pick a cost model: array accesses.

Step 2. Count array accesses: $2 \times \frac{n(n-1)}{2} \sim 1n^2$.

↑
*body inner loop makes
2 array accesses*

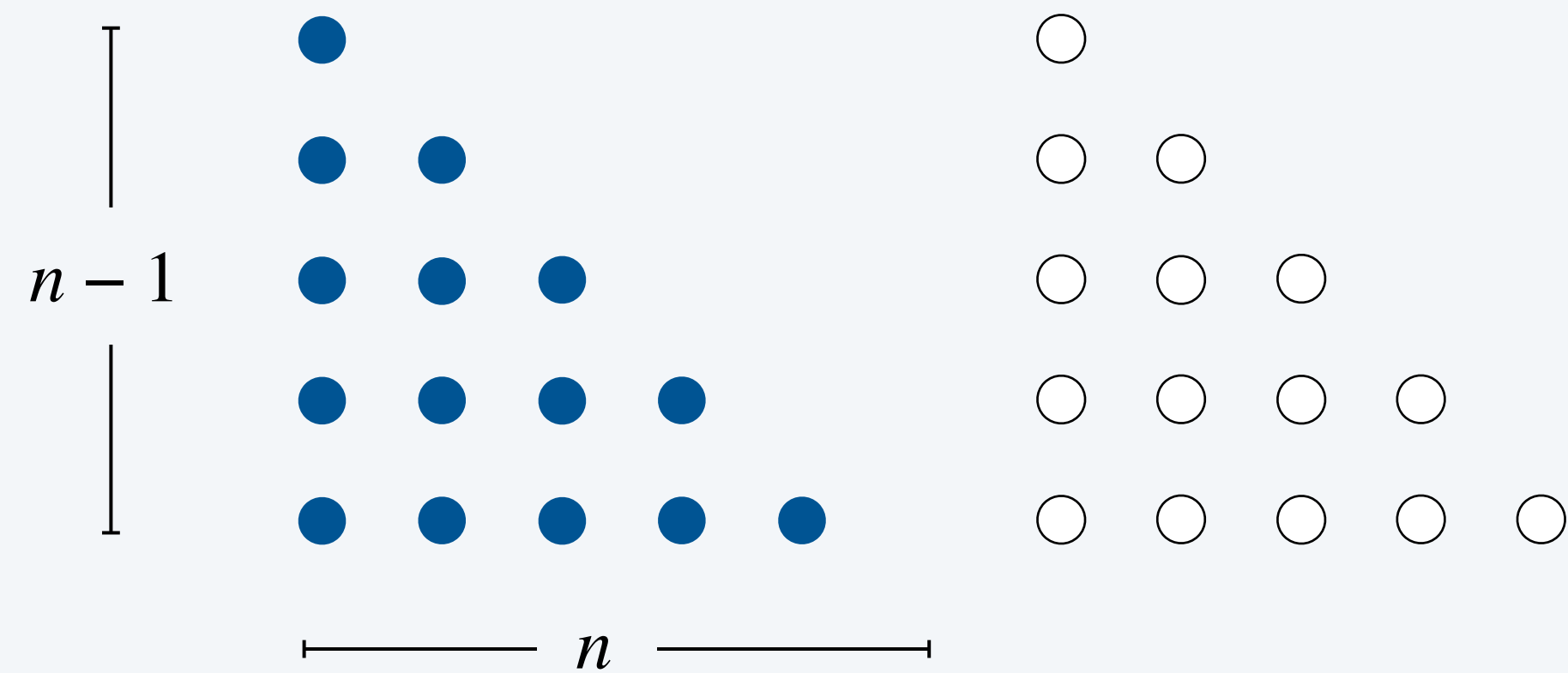
Nested loops.

- Independent loops: analyze separately and multiply.
- Dependent loops: write a sum (and simplify).

Triangular sum

Claim. $0 + 1 + \dots + (n-2) + (n-1) = \frac{1}{2} n(n-1).$

Proof. [by picture]



$$2 \times (1 + 2 + 3 + \dots + n-1) = n(n-1)$$

$$\implies (1 + 2 + 3 + \dots + n-1) = \frac{1}{2} n(n-1)$$

Example: 3-SUM analysis

Q. Approximately many operations 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++;
```

← “inner loop” $\binom{n}{3} = \frac{n(n-1)(n-2)}{3!} \sim \frac{1}{6}n^3$

see COS 240

Step 1. Pick a cost model: array accesses.

Step 2. Count the number of array accesses: $\Theta(n^3)$.

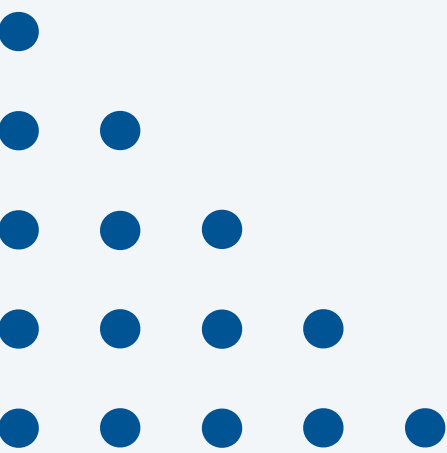
Bottom line. Using a **cost model** and **asymptotic notation** makes analysis manageable.

Common orders of growth

order of growth	emoji	name	typical code pattern	description	example	$T(2n) / T(n)$
$\Theta(1)$	💕	constant	<code>a = b + c;</code>	statement	<i>add two numbers</i>	1
$\Theta(\log n)$	😎	logarithmic	<code>for (int i = n; i > 0; i /= 2) { ... }</code>	repeatedly divide in half	<i>binary search</i>	~ 1
$\Theta(n)$	😄	linear	<code>for (int i = 0; i < n; i++) { ... }</code>	single loop	<i>find the maximum</i>	2
$\Theta(n \log n)$	😁	linearithmic	<i>mergesort</i>	divide-and-conquer	<i>mergesort</i>	~ 2
$\Theta(n^2)$	😞	quadratic	<code>for (int i = 0; i < n; i++) for (int j = 0; j < n; j++) { ... }</code>	double loop	<i>check all pairs</i>	4
$\Theta(n^3)$	😡	cubic	<code>for (int i = 0; i < n; i++) for (int j = 0; j < n; j++) for (int k = 0; k < n; k++) { ... }</code>	triple loop	<i>check all triples</i>	8
$\Theta(2^n)$	😈	exponential	<i>towers of Hanoi</i>	brute-force search	<i>check all subsets</i>	2^n

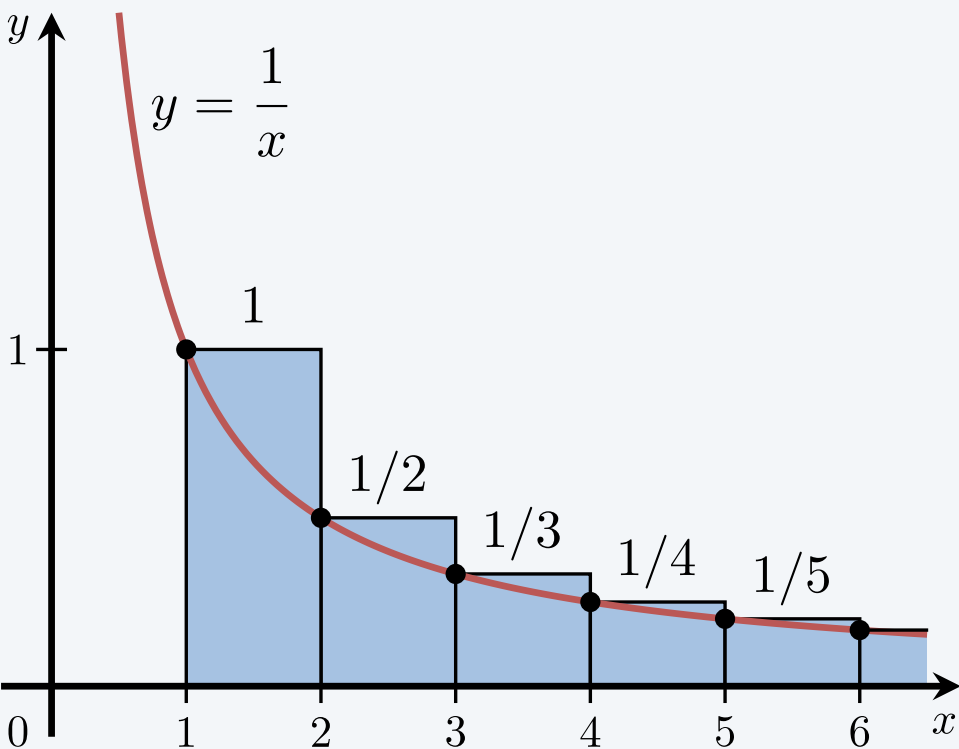
Useful discrete sums and approximations

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



Logarithmic sum. $\log_2 1 + \log_2 2 + \log_2 3 + \dots + \log_2 n \sim \int_{x=1}^n \log_2 x \, dx = n \log_2 n$

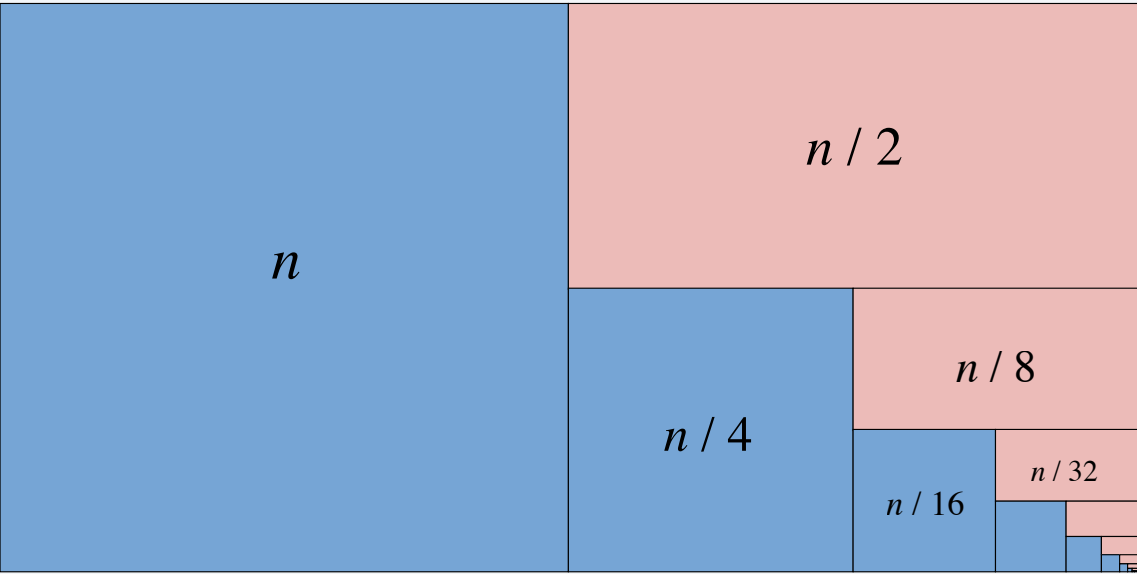
Harmonic sum. $\frac{1}{1} + \frac{1}{2} + \frac{1}{3} + \dots + \frac{1}{n} \sim \int_{x=1}^n \frac{1}{x} \, dx = \ln n$



Geometric sum. $1 + 2 + 4 + 8 + \dots + n = 2n - 1$

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

n is a power of 2



Geometric sum meme



<https://marekbennett.com/2014/03/06/recursive-load>



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

```
int count = 0;
for (int i = 0; i < n; i++)
    for (int j = i+1; j < n; j++)
        for (int k = 1; k <= n; k = k*2)
            if (a[i] + a[j] >= a[k])
                count++;
```

- A. $\sim n^2 \log_2 n$
- B. $\sim \frac{3}{2} n^2 \log_2 n$
- C. $\sim \frac{1}{2} n^3$
- D. $\sim \frac{3}{2} n^3$



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

```
int count = 0;
for (int i = n; i >= 1; i = i/2)
    for (int j = 1; j <= i; j++)
        count++;
```

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



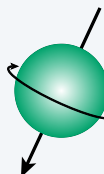
<https://algs4.cs.princeton.edu>

1.4 ANALYSIS OF ALGORITHMS

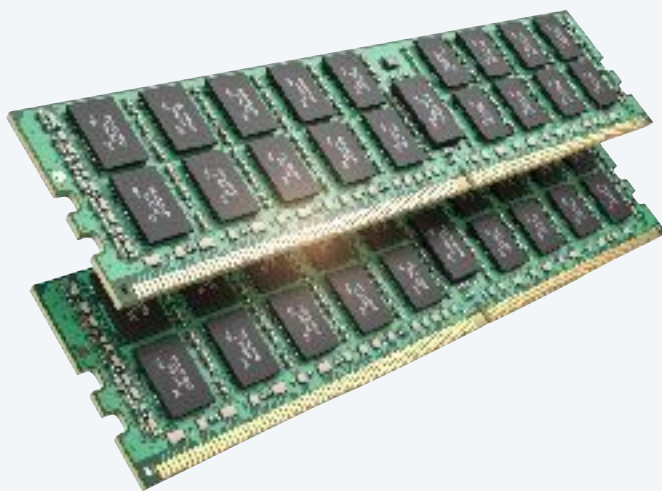
- *introduction*
- *running time (experimental analysis)*
- *running time (mathematical models)*
- *memory usage*

Memory basics: bits, bytes, and pointers

Bit. A single binary digits (0 or 1).

0				
1				

term	symbol	size
byte	B	8 bits
kilobyte	KB	10^3 bytes
megabyte	MB	10^6 bytes
gigabyte	GB	10^9 bytes
terabyte	TB	10^{12} bytes



↑
*some systems use powers of 2
(e.g., 1 MB = 2^{20} bytes)*

Assumption. Running on a 64-bit machine with 8-byte pointers.



↑
*some JVMs “compress” pointers
to 4 bytes to avoid this cost*

Typical memory usage of primitive types and arrays in Java

type	bytes
boolean	1
byte	1
char	2
int	4
float	4
long	8
double	8
primitive types	

type	bytes
boolean[]	$1n + 24$ ← <i>wasteful</i> <i>(but ~ 36n bytes in Python 3)</i>
int[]	$4n + 24$
double[]	$8n + 24$ ← <i>array overhead = 24 bytes</i>
one-dimensional arrays (length n)	

type	bytes
boolean[][]	$\sim 1 n^2$
int[][]	$\sim 4 n^2$
double[][]	$\sim 8 n^2$
two-dimensional arrays (n-by-n array of arrays)	

type	bytes
object reference	8
	↑ <i>64-bit machine</i>

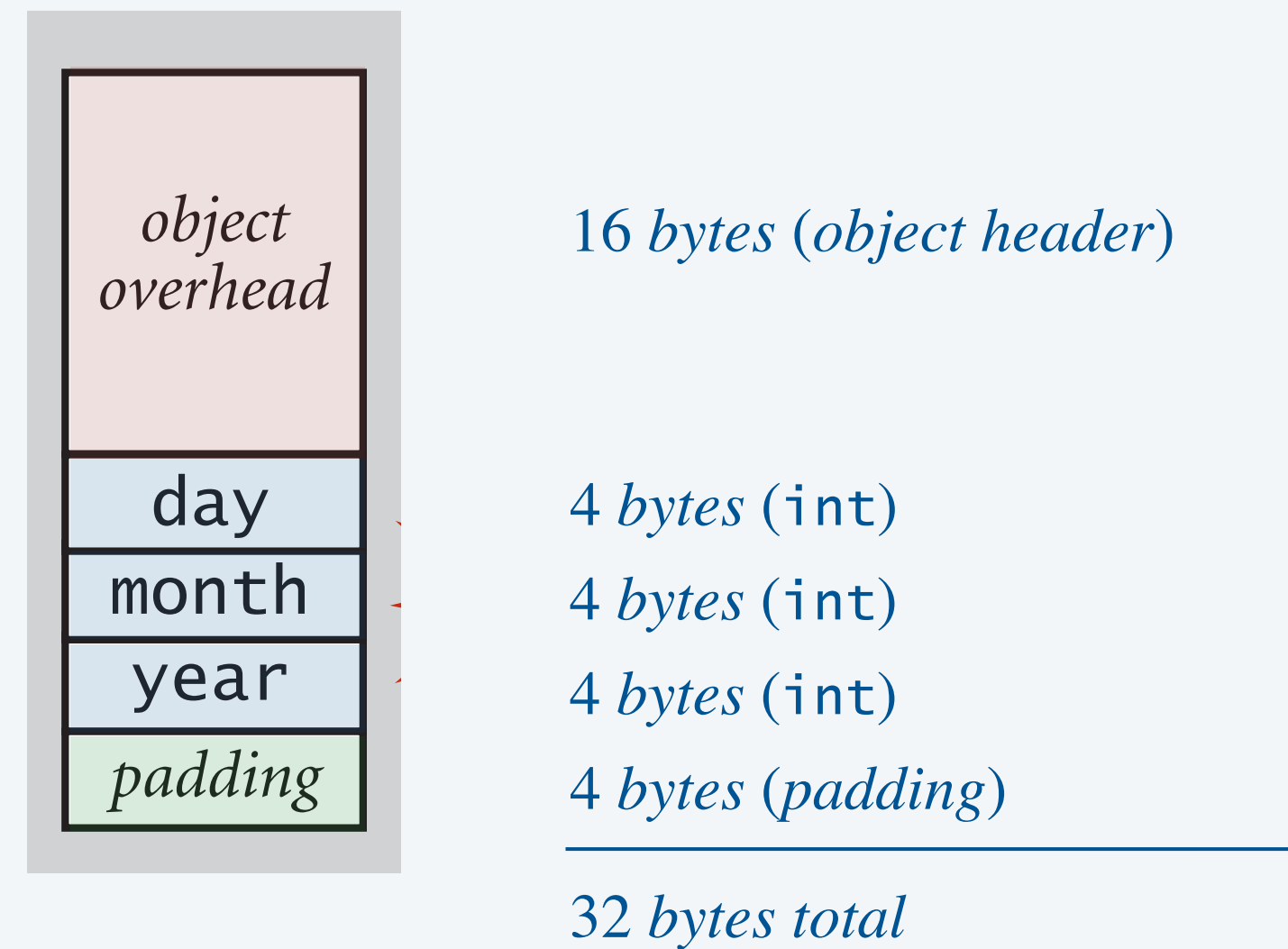
Typical memory usage of objects in Java

Object header. Every Java object begins with a fixed 16-byte header. ← *class pointer, garbage collector bits, lock state, hash code, ...*

Padding. Object size is rounded up to a multiple of 8 bytes.

Example. A *Date* object uses 32 bytes of memory.

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





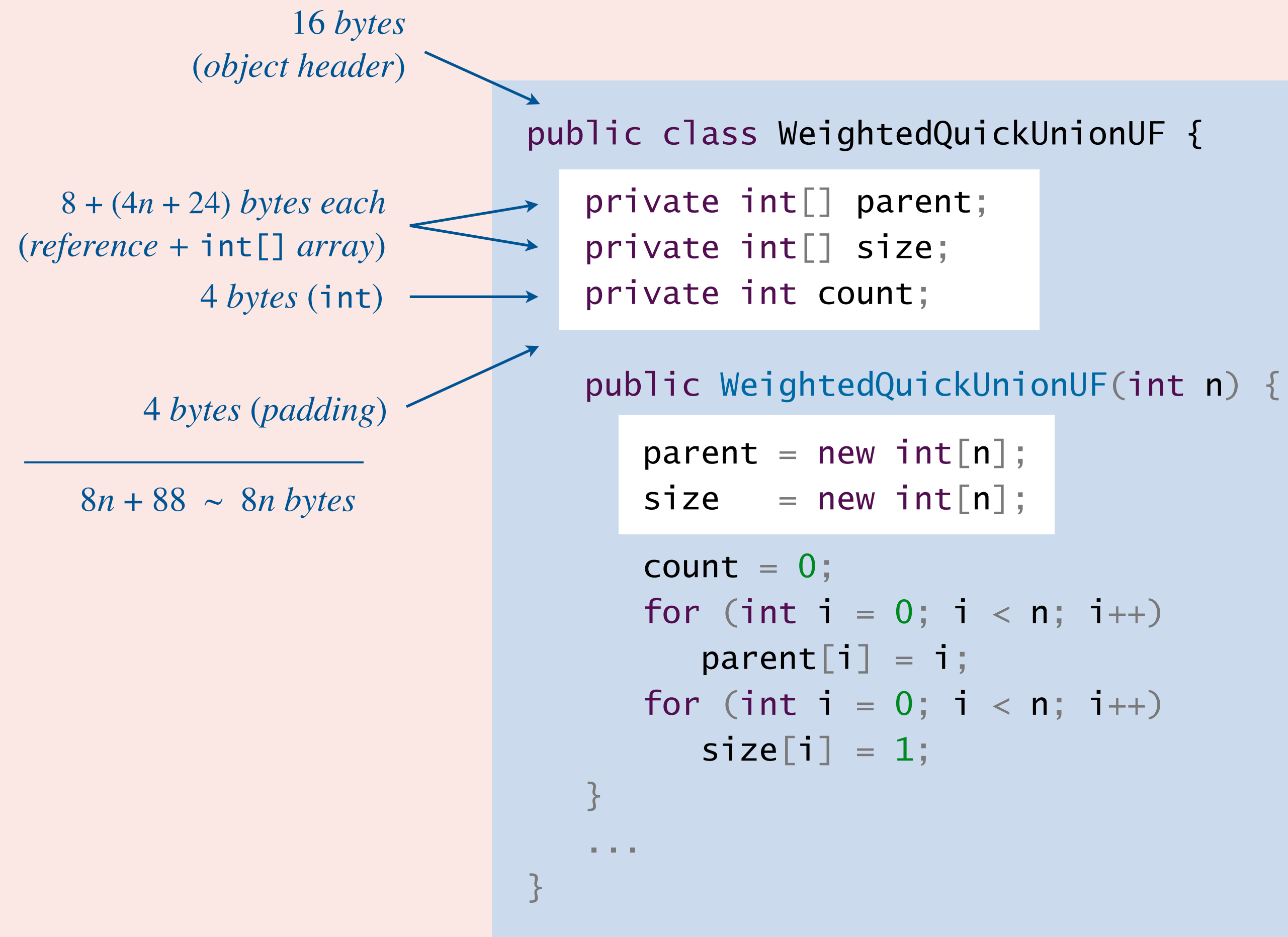
How much memory does a `WeightedQuickUnionUF` object use as a function of n ?

- A. $\sim 4n$ bytes
- B. $\sim 8n$ bytes
- C. $\sim 4n^2$ bytes
- D. $\sim 8n^2$ 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;  
    }  
    ...  
}
```



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



Turning the crank: summary

Goal. Analyze running time (or memory) as a function of input size.

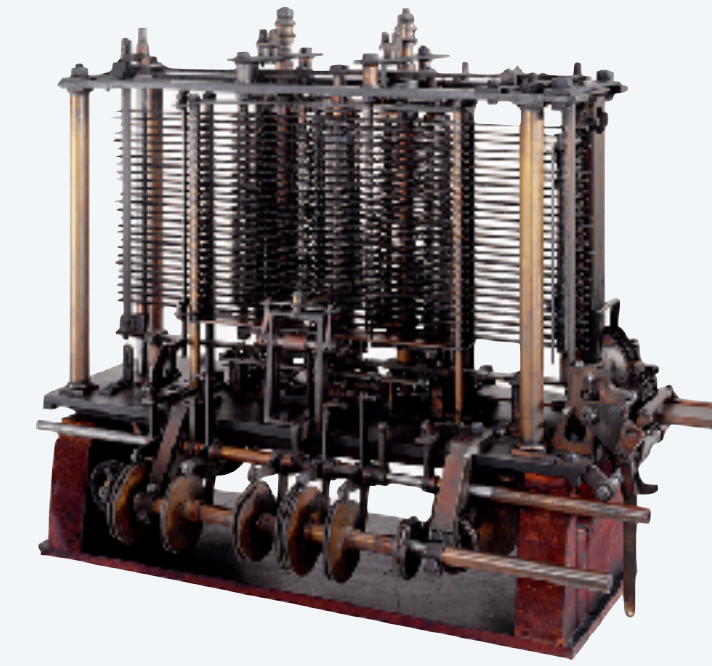
Empirical analysis.

- Run the program and measure how long it takes.
- Formulate a hypothesis for the running time.
- Use the model to **predict** behavior.

Mathematical analysis.

- Analyze the algorithm on an abstract machine.
- Count the frequency of its dominant operations.
- Apply a cost model and asymptotic notation to simplify analysis.
- Use the model to **explain** behavior.

This course. You will learn to use both.



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

Credits

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A final thought

“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)

