Algorithms



ROBERT SEDGEWICK | KEVIN WAYNE

1.4 ANALYSIS OF ALGORITHMS

running time (experimental analysis)

running time (mathematical models)

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Algorithms

Robert Sedgewick | Kevin Wayne

https://algs4.cs.princeton.edu

1.4 ANALYSIS OF ALGORITHMS

introduction
 running time (experimental analysis)
 running time (mathematical models)

memory usage



Cast of characters



programmer needs to develop a working solution



client wants to solve problem efficiently





theoretician seeks to understand



student (you) might play all of these roles someday

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?

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)

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

By Matt Kennedy July 25, 2018



							Data.								1	Working Variables.	-			Result V	ariables.	
Number of Operation	Nature of Operation	Variables acted upon.	Variables receiving results.	Indication of change in the value on any Variable.	Statement of Results.	$1V_1 \\ 0 \\ 0 \\ 0 \\ 1 \\ 1$	¹ V ₂ O 0 2 2	¹ V ₃ O 0 4 <i>n</i>	⁰ V ₄ O 0 0 0	⁰ V ₅ O 0 0 0	⁰ Ve ○ 0 0 0	°V7 0000	⁰Vs ○ 0 0 0 □	°V ₉ 00 00 0	^o V ₁₀ O 0 0 0	⁰ V ₁₁ O 0 0 0	⁶ V ₁₂ O 0 0 0	0 0 0 0	[⊥] [⊥] [⊥] [⊥] ¹	[] B ₃ in a decimal O ₁₃ ₹ fraction.	[™] _a B _s in a decimalO ¹ ₁₅	⁰ V ₅₁ O 0 0 0 B ₇
1 2 3 4 5 6 7	× 1. + + + 1 1	$\frac{1}{1^{1}V_{2} \times {}^{1}V_{3}}$ $\frac{1}{1^{1}V_{4}} - \frac{1}{1^{1}V_{1}}$ $\frac{1}{1^{1}V_{5}} + \frac{1}{1^{1}V_{1}}$ $\frac{1}{1^{1}V_{11}} + \frac{1}{1^{1}V_{2}}$ $\frac{1}{1^{1}V_{13}} - \frac{1}{1^{1}V_{1}}$ $\frac{1}{1^{1}V_{3}} - \frac{1}{1^{1}V_{1}}$	¹ V ₄ , ¹ V ₅ , ¹ V ₆ ² V ₄ ² V ₅ ¹ V ₁₁ ¹ V ₁₃ ¹ V ₁₀	$ \begin{cases} V_2 = V_2 \\ IV_3 = IV_3 \\ IV_4 = 2V_4 \\ IV_1 = IV_1 \\ IV_5 = 2V_5 \\ IV_1 = V_1 \\ SV_5 = 0V_5 \\ 2V_4 = 0V_4 \\ IV_1 = 2V_1 \\ IV_2 = IV_2 \\ SV_{110} = V_{111} \\ 0V_{110} = IV_{111} \\ V_2 = IV_3 \\ IV_1 = IV_1 \\ IV_2 = IV_1 \\ IV_1 = IV_1 \\ SV_1 = IV_1 \\ IV_2 = IV_1 \\ IV_1 = IV_1 \\ $	$= 2n$ $= 2n-1$ $= 2n-1$ $= 2n-1$ $= 2n-1$ $= \frac{2n-1}{2n+1}$ $= \frac{1}{2} \cdot \frac{2n-1}{2n+1}$ $= -\frac{1}{2} \cdot \frac{2n-1}{2n+1} = \Lambda_0$ $= n-1 (=3)$	 1 1	2	n n	2 n 2 n - 1 0 	2 n 2 n+1 0 	2 n				 n - 1	$\frac{\frac{2n-1}{2n+1}}{\frac{1}{2},\frac{2n-1}{2n+1}}$		$-\frac{1}{2}\cdot\frac{2n-1}{2n+1}=\Lambda_0$				
	+ + × + -	${}^{1}V_{2} + {}^{6}V_{7}$ ${}^{1}V_{6} + {}^{1}V_{7}$ ${}^{1}V_{21} \times {}^{3}V_{11}$ ${}^{1}V_{12} + {}^{1}V_{13}$ ${}^{1}V_{10} - {}^{1}V_{1}$	¹ V ₇ ³ V ₁₁ ¹ V ₁₂ ² V ₁₃ ² V ₁₀	$\begin{cases} {}^{1}V_2 = {}^{1}V_2 \\ {}^{0}V_7 = {}^{1}V_7 \\ {}^{1}V_6 = {}^{1}V_6 \\ {}^{0}V_{11} = {}^{3}V_{11} \\ {}^{1}V_{21} = {}^{1}V_{21} \\ {}^{3}V_{11} = {}^{3}V_{11} \\ {}^{1}V_{12} = {}^{0}V_{12} \\ {}^{1}V_{12} = {}^{2}V_{12} \\ {}^{1}V_{12} = {}^{2}V_{12} \\ {}^{1}V_{12} = {}^{2}V_{13} \\ {}^{1}V_{11} = {}^{2}V_{11} \\ {}^{1}V_{11} = {}^{1}V_1 \\ {}^{1}V_1 \\ {}^{1}V_1 = {}^{1}V_1 \\ {}^{1}V_1 \\$	$\begin{split} &= 2 + 0 = 2 \dots \\ &= \frac{2n}{2} = \lambda_1 \dots \\ &= B_1 \cdot \frac{2n}{2} = B_1 \lambda_1 \dots \\ &= -\frac{1}{2} \cdot \frac{2n-1}{2n+1} + B_1 \cdot \frac{2n}{2} \dots \\ &= n-2 (=2) \dots \end{split}$	 1	2				 2 n 	99 99 :: ::			 n - 2	$\frac{\frac{2}{2}n}{\frac{2}{2}} = \Lambda_1$ $\frac{\frac{2}{2}n}{\frac{2}{2}} = \Lambda_1$	$B_1 \cdot \frac{2 n}{2} = B_1 A_2$	$\left\{-\frac{1}{2},\frac{2n-1}{2n+1}+B_{1},\frac{2n}{2}\right\}$	B1	1011 A. 10 - 11		
3 4 5 6 7 8 9	-++× × ++	${}^{1}V_{6} - {}^{1}V_{1}$ ${}^{1}V_{1} + {}^{1}V_{7}$ ${}^{2}V_{6} + {}^{2}V_{7}$ ${}^{1}V_{8} \times {}^{3}V_{11}$ ${}^{2}V_{6} - {}^{1}V_{1}$ ${}^{1}V_{1} + {}^{2}V_{7}$ ${}^{3}V_{6} + {}^{3}V_{7}$	² V ₆ ² V ₇ ¹ V ₈ ⁴ V ₁₁ ³ V ₆ ³ V ₇	$ \begin{bmatrix} V_{6} = 2V_{6} \\ V_{1} = V_{1} \\ V_{1} = V_{1} \\ V_{2} = 2V_{2} \\ V_{3} = 2V_{5} \\ 2V_{7} = 2V_{7} \\ 2V_{7} = 2V_{7} \\ V_{7} = 2V_{7} \\ V_{7} = 2V_{7} \\ V_{1} = 0V_{3} \\ V_{1} = 0V_{3} \\ V_{1} = 1V_{1} \\ V_{6} = 2V_{6} \\ V_{1} = 1V_{1} \\ 2V_{7} = 2V_{7} \\ V_{1} = 1V_{1} \\ 2V_{6} = 3V_{6} \\ V_{3} = 3V_{7} \\ 2V_{6} = 3V_{7} \\ V_{5} = 3V_{7} \\ V_{7} = 2V_{7} $	= 2n - 1 = 2 + 1 = 3 = $\frac{2n - 1}{3}$ = $\frac{2n - 2}{2}$ = $2n - 2$ = $3 + 1 = 4$ = $\frac{2n - 2}{4}$	1 1 1 1					2 n - 1 2 n - 1 2 n - 2 2 n - 2 2 n - 2	3 3 4 4	$\frac{2n-1}{3}$ 0	 2n - 2 4		$\left\{\frac{2n}{2} \cdot \frac{2n-1}{3} \\ \left\{\frac{2n}{2} \cdot \frac{2n-1}{3} \cdot \frac{2n-2}{3}\right\}\right\}$				12)		
	× + -	${}^{1}V_{9} \times {}^{4}V_{11}$ ${}^{1}V_{22} \times {}^{5}V_{12}$ ${}^{2}V_{12} + {}^{2}V_{12}$ ${}^{2}V_{12} - {}^{1}V_{1}$	^a V ₁₁ ^a V ₁₂ ^a V ₁₃ ^a V ₁₆	$\left\{\begin{array}{l} 1V_{9}=0V_{9}\\ 4V_{11}=5V_{11}\\ 0V_{12}=1V_{22}\\ 0V_{12}=2V_{12}\\ 2V_{12}=0V_{12}\\ 2V_{13}=3V_{13}\\ 2V_{13}=3V_{13}\\ 1V_{1}=1V_{1} \end{array}\right\}$	$= \frac{2n}{2} \cdot \frac{2n-1}{3} \cdot \frac{2n-2}{4} = \Lambda_3$ = $B_3 \cdot \frac{2n}{2} \cdot \frac{2n-3}{3} \cdot \frac{2n-2}{4} = B_3 \Lambda_3$ = $\Lambda_6 + B_1 \Lambda_1 + B_3 \Lambda_2$	····								4 0 	 n - 3	0 	B ₂ A ₃ 0	$\left\{A_3 + B_1 \Lambda_1 + B_2 \Lambda_3^{'}\right\}$		Ba		N THE CO
	++	*V ₁₃ +°V ₂	1V24	$ \left\{ \begin{array}{l} {}^{4}V_{13} = {}^{0}V_{13} \\ {}^{0}V_{24} = {}^{1}V_{24} \\ {}^{1}V_{1} = {}^{1}V_{1} \\ {}^{1}V_{3} = {}^{1}V_{3} \\ {}^{5}V = {}^{0}V \end{array} \right. $	$= B_{7}$ $= n + 1 = 4 + 1 = 5$ by a Variable-card	· ··· 1		 n+1			0	0			···.							

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



N-body simulation.

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





Andrew Appel PU '81







The challenge



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. Related to problems in computational geometry.





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





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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Measuring the running time

- **Q.** How to time a program?
- A. Manual.





% java ThreeSum 1Kints.txt

tick tick tick

% java ThreeSum 2Kints.txt

tick tick

% java ThreeSum 4Kints.txt

tick tick

Measuring the running time

- **Q.** How to time a program?
- A. Automatic.

```
import edu.princeton.cs.algs4.StdOut;
import edu.princeton.cs.algs4.Stopwatch;
public static void main(String[] args)
{
    In in = new In(args[0]);
    int[] a = in.readAllInts();
    Stopwatch stopwatch = new Stopwatch();
    StdOut.println(ThreeSum.count(a));
    double time = (stopwatch.elapsedTime();
    StdOut.println("elapsed time = " + time);
}
```



Empirical analysis

Run the program for various input sizes and measure running time.



Run the program for various input sizes and measure running time.

n	time (seconds) †			
250	0			
500	0			
1,000	0.1			
2,000	0.8			
4,000	6.4			
8,000	51.1			
16,000	?			

† on a 2.8GHz Intel PU-226 with 64GB DDR E3 memory and 32MB L3 cache; running Oracle Java 1.7.0_45-b18 on Springdale Linux v. 6.5

Data analysis

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



Hypothesis (power law). $T(n) = a n^{b}$. Questions. How to validate hypothesis? How to estimate *a* and *b*?

Data analysis

Log–log plot. Plot running time T(n) vs. input size *n* using log–log scale.



Regression. Fit straight line through data points. Hypothesis. The running time is about $1.006 \times 10^{-10} \times n^{2.999}$ seconds.

slope
(2)) =
$$2.999 \log_2 n + (-33.21)$$

$$n) = 2^{-33.21} \times n^{2.999}$$
$$= 1.006 \times 10^{-10} \times n^{2.999}$$

Hypothesis. The running time is about $1.006 \times 10^{-10} \times n^{2.999}$ seconds.

Predictions.

- 51.0 seconds for *n* = 8,000.
- 408.1 seconds for *n* = 16,000.

Observations.

n	time (seconds) †
8,000	51.1
8,000	51.0
8,000	51.1
16,000	410.8

validates hypothesis!

"order of growth" of running time is about *n*³ [stay tuned]

Doubling hypothesis. Quick way to estimate b in a power-law relationship.

Run program, doubling the size of the input.

n	time (seconds) †	ratio	log ₂ ratio
250	0		
500	0	4.8	2.3
1,000	0.1	6.9	2.8
2,000	0.8	7.7	2.9
4,000	6.4	8	3.0 🔸
8,000	51.1	8	3.0

Hypothesis. Running time is about $a n^b$ with $b = \log_2$ ratio. **Caveat.** Cannot identify logarithmic factors with doubling hypothesis.



seems to converge to a constant $b \approx 3$

Doubling hypothesis. Quick way to estimate *b* in a power-law relationship.

- **Q.** How to estimate *a* (assuming we know *b*)?
- A. Run the program (for a sufficient large value of *n*) and solve for *a*.

n	time (seconds) †
8,000	51.1
8,000	51.0
8,000	51.1

Hypothesis. Running time is about $0.998 \times 10^{-10} \times n^3$ seconds. almost identical hypothesis to one obtained via regression (but less work)

 $= a \times 8000^3$

 $= 0.998 \times 10^{-10}$

Analysis of algorithms: quiz 1

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

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







Analysis of algorithms: quiz 1

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

Α.	39 seconds	n	time (seconds)	ratio	log2 (ratio)
Β.	52 seconds	1,000	0.02	_	_
C.	117 seconds	2,000	0.05	2.5	1.3
D .	350 seconds	4,000	0.20	4.0	2.0
		8,000	0.81	4.1	2.0
		16,000	3.25	4.0	2.0
		32,000	13.01	4.0	2.0

$T(n) = a n^2$ seconds	T(3n)
T(32,000) = 13 seconds	
$\Rightarrow a = 1.2695 \times 10^{-8}$	

 \Rightarrow T(96,000) = 117 seconds

- $) = a (3n)^2$
- $= 9 a n^2$
- = 9 T(n) seconds

 \Rightarrow T(3 × 32,000) = 9 × 13 = 117 seconds



System independent effects.

- Algorithm.
- Input data.

determines exponent *b* in power law *a* 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)





Biology (1 experiment)

Computer Science (1 million experiments)

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 cost × frequency for all operations.

- Need to analyze program to determine set of operations.
- Cost depends on machine, compiler.
- Frequency depends on algorithm, input data.



Warning. No general-purpose method (e.g., halting problem).

perations. rations.

Example: 1-SUM

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	n
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)



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.** No idea.





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. $2n^2$

n-1n

> $2 \times (1 + 2 + 3 + \dots +$ \implies (1+2+3+ ... +



$$(n-1) = n(n-1)$$

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



Example: 2-SUM

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



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

 $0 + 1 + 2 + \ldots + (n - 1) = \frac{1}{2}n(n - 1)$ $=\binom{n}{2}$ + 2) $1/4 n^2 + 13/20 n + 13/10 ns$ to 1) $3/10 n^2 + 3/5 n + 13/10$ ns (tedious to count exactly) to n^2

Simplification 1: cost model

Cost model. Use some elementary operation as a proxy for running time.

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





Simplification 2: asymptotic notations

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

function	tilde	big Theta
$4 n^5 + 20 n + 16$	~ 4 n^5	$\Theta(n^5)$
$7 n^2 + 100 n^{4/3} + 56$	$\sim 7 n^2$	$\Theta(n^2)$
$\frac{1}{6}n^3 - \frac{1}{2}n^2 + \frac{1}{3}n$	~ $\frac{1}{6} n^3$	$\Theta(n^3)$

discard lower-order terms (e.g., n = 1,000: 166.67 million vs. 166.17 million)

Rationale.

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





Common order-of-growth classifications

order of growth	name	typical code framework
$\Theta(1)$	constant	a = b + c;
$\Theta(\log n)$	logarithmic	<pre>while (n > 1) { n = n/2; }</pre>
$\Theta(n)$	linear	for (int i = 0; i < n; i++) { }
$\Theta(n \log n)$	linearithmic	see mergesort lecture
$\Theta(n^2)$	quadratic	for (int i = 0; i < n; i++) for (int j = 0; j < n; j++) { }
$\Theta(n^3)$	cubic	<pre>for (int i = 0; i < n; i++) for (int j = 0; j < n; j++) for (int k = 0; k < n; k++) { }</pre>
$\Theta(2^n)$	exponential	see combinatorial search lecture

description	example	T(2n) / T(n)
statement	add two numbers	1
di∨ide in half	binary search	~ 1
single loop	find the maximum	2
divide and conquer	mergesort	~ 2
double loop	check all pairs	4
triple loop	check all triples	8
exhaustive search	check all subsets	2 <i>ⁿ</i>

Example: 2-SUM

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



A. ~ n^2 array accesses.

——— "inner loop"

$$0 + 1 + 2 + \ldots + (n - 1) = \frac{1}{2}n(n - 1)$$

= $\binom{n}{2}$

Example: 3-SUM

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



A. ~ $\frac{1}{2} n^3$ array accesses.

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

Estimating a discrete sum

- Q. How to estimate a discrete sum?
- A1. Take a discrete mathematics course (COS 340).



Estimating a discrete sum

Q. How to estimate a discrete sum?

A2. Replace the sum with an integral; use calculus!

Ex 1.
$$1 + 2 + ... + n$$
.
Ex 1. $1 + 2 + ... + n$.
Ex 2. $1 + \frac{1}{2} + \frac{1}{3} + ... + \frac{1}{n}$.
Ex 3. 3-sum triple loop.

$$\sum_{i=1}^{n} \sum_{j=i}^{n} \sum_{k=j}^{n} 1 \sim \int_{x=1}^{n} \int_{y=x}^{n} \int_{z=y}^{n} dz \, dy \, dx \sim \frac{1}{6}n^{3}$$
Ex 4. $1 + \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + ...$

$$\int_{x=0}^{\infty} \left(\frac{1}{2}\right)^{x} dx = \frac{1}{\ln 2} \approx 1.4427$$

$$\sum_{i=0}^{\infty} \left(\frac{1}{2}\right)^{i} = 2$$
integral trid doesn't always of





Estimating a discrete sum

- Q. How to estimate a discrete sum?
- A3. Use Maple or Wolfram Alpha.



https://www.wolframalpha.com

Analysis of algorithms: quiz 3

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 order of growth of 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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introduction



Basics

Bit. 0 or 1. NIST most computer scientists Byte. 8 bits. Megabyte (MB). 1 million or 2²⁰ bytes. Gigabyte (GB). 1 billion or 2³⁰ bytes.

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





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

Typical memory usage for primitive types and arrays

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



two-dimensional arrays (n-by-n)

Typical memory usage for objects in Java

Object overhead. 16 bytes.Reference. 8 bytes.Padding. Memory of each object rounded up to use a multiple of 8 bytes.

Ex 1. A Date object uses 32 bytes of memory.

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

16 bytes (object overhead)

4 bytes (int)

4 bytes (int)

4 bytes (int)

4 bytes (padding)

32 bytes

Total memory usage for a data type value in Java:

- Primitive type: 4 bytes for int, 8 bytes for double, ...
- Object reference: 8 bytes.
- Array: 24 bytes + memory for each array entry.
- Object: 16 bytes + memory for each instance variable.
- Padding: round up to multiple of 8 bytes.

Note. Depending on application, we often count the memory for any referenced objects (recursively).

"deep memory"

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

Α.	$\sim 4 n$ bytes
B.	$\sim 8 n$ bytes
С.	~ 4 n^2 bytes
D.	~ 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 use as a function of *n*?

16 bytes (object overhead)	
8 + (4 <i>n</i> + 24) bytes each (reference + int[] array)	\longrightarrow
4 bytes (int) 4 bytes (padding)	\longrightarrow

 $8n + 88 \sim 8n$ 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; . . . }





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 tilde and big-Theta notations to simplify analysis.
- Model enables us to explain behavior.

This course. Learn to use both.



ns. ysis.

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



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