Algorithms



Two classic sorting algorithms: mergesort and quicksort

Critical components in our computational infrastructure.

Mergesort. [this lecture]

















Quicksort. [next lecture]



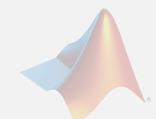




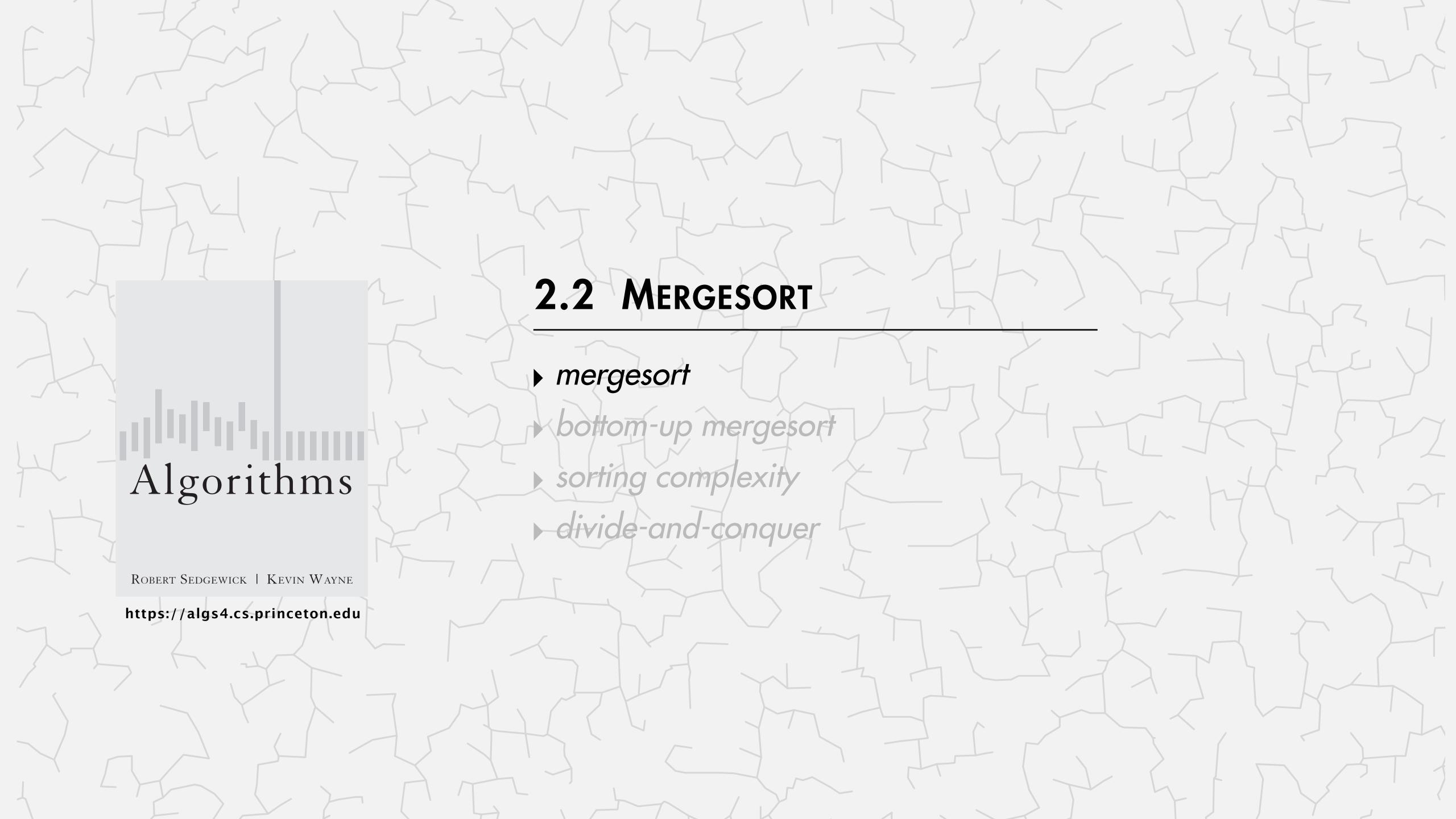








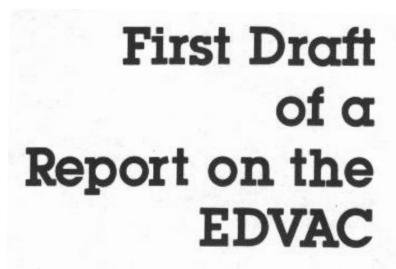




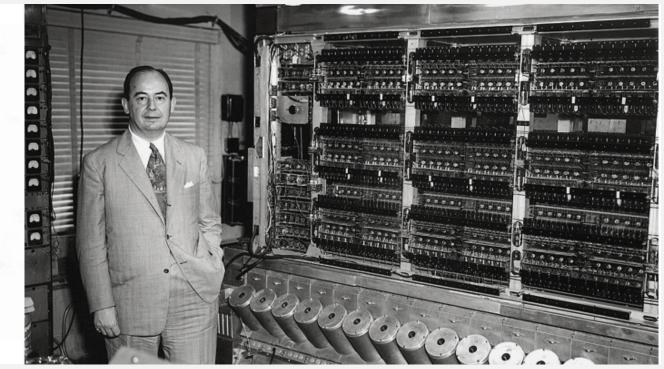
Mergesort

Basic plan.

- Divide array into two halves.
- Recursively left half.
- Recursively sort right half.
- Merge two sorted halves.



John von Neumann

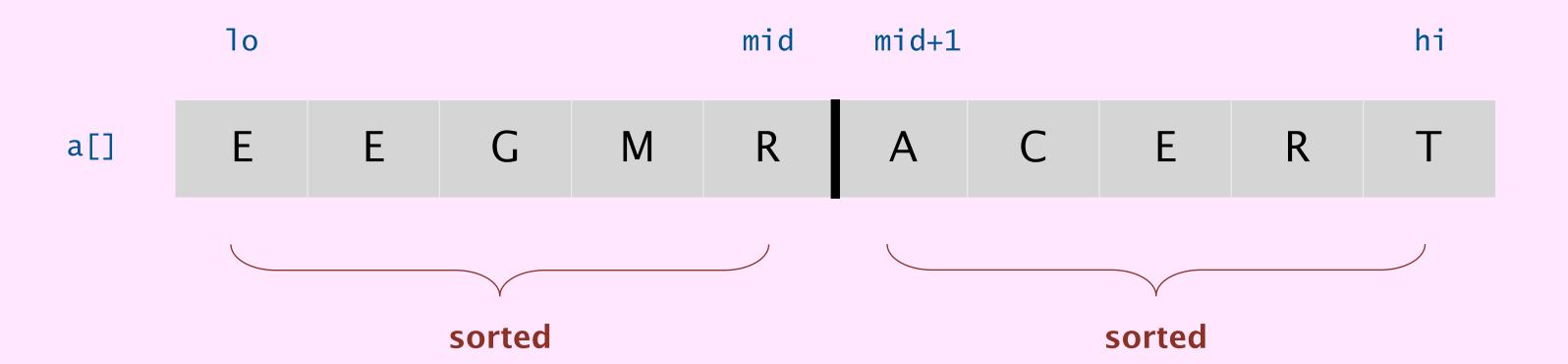


input	M	Ε	R	G	Ε	S	O	RT	Ε	X	A	M	P	L	Ε
sort left half	Ε	Ε	G	M	0	R	R	S T	Ε	X	A	M	Р	L	Ε
sort right half	Ε	Е	G	M	0	R	R	S A	Ε	Ε	L	M	P	Т	X
merge results	A	Ε	Ε	Ε	Ε	G	L	M M	0	Р	R	R	S	Т	X

Abstract in-place merge demo

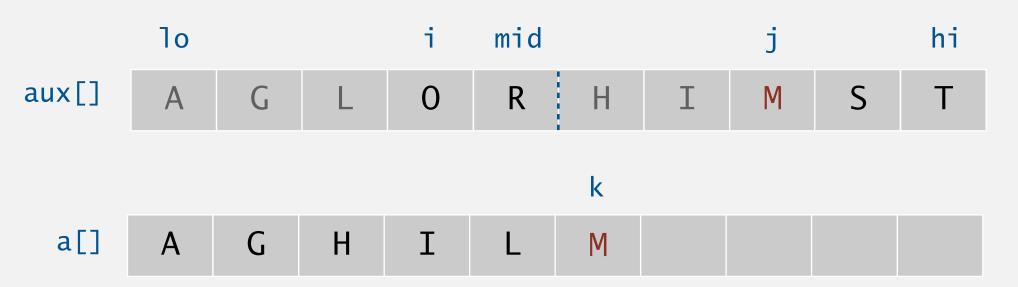


Goal. Given two sorted subarrays a[lo] to a[mid] and a[mid+1] to a[hi], replace with sorted subarray a[lo] to a[hi].



Merging: Java implementation

```
private static void merge(Comparable[] a, Comparable[] aux, int lo, int mid, int hi)
  for (int k = lo; k \ll hi; k++)
                                   copy
     aux[k] = a[k];
  int i = lo, j = mid+1;
                                                merge
  for (int k = 10; k \le hi; k++)
        (i > mid) a[k] = aux[j++];
     if
     else if (j > hi) a[k] = aux[i++];
     else if (less(aux[j], aux[i])) a[k] = aux[j++];
                                a[k] = aux[i++];
     else
```



Mergesort quiz 1



How many calls does merge() make to less() in order to merge two sorted subarrays, each of length n/2, into a sorted array of length n?

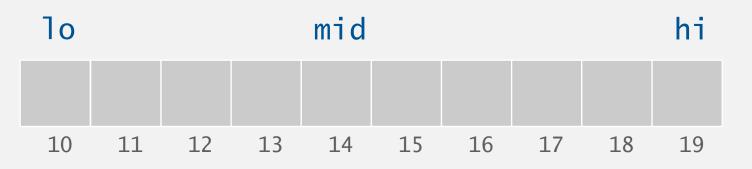
- **A.** $\sim \frac{1}{4} n$ to $\sim \frac{1}{2} n$
- **B.** $\sim \frac{1}{2} n$
- C. $\sim \frac{1}{2} n$ to $\sim n$
- $\sim n$

merging two sorted arrays, each of length n/2

 $a_0 \quad a_1 \quad a_2 \quad a_3 \quad b_0 \quad b_1 \quad b_2 \quad b_3$

Mergesort: Java implementation

```
public class Merge
   private static void merge(...)
   { /* as before */ }
   private static void sort(Comparable[] a, Comparable[] aux, int lo, int hi)
      if (hi <= lo) return;</pre>
      int mid = (lo + hi) >>> 1;
      sort(a, aux, lo, mid);
      sort(a, aux, mid+1, hi);
      merge(a, aux, lo, mid, hi);
   public static void sort(Comparable[] a)
                                                               avoid allocating arrays
      Comparable[] aux = new Comparable[a.length]; <----</pre>
                                                            within recursive function calls
      sort(a, aux, 0, a.length - 1);
```



Mergesort: trace

```
——— result after recursive call
     merge(a, aux, 0, 0,
     merge(a, aux, 2, 2, 3)
   merge(a, aux, 0, 1, 3)
     merge(a, aux, 4, 4, 5)
     merge(a, aux, 6, 6, 7)
   merge(a, aux, 4, 5, 7)
 merge(a, aux, 0, 3, 7)
     merge(a, aux, 8, 8, 9)
     merge(a, aux, 10, 10, 11)
   merge(a, aux, 8, 9, 11)
     merge(a, aux, 12, 12, 13)
     merge(a, aux, 14, 14, 15)
   merge(a, aux, 12, 13, 15)
 merge(a, aux, 8, 11, 15)
merge(a, aux, 0, 7, 15) A E E E G L M M O P R R S T X
```

Mergesort quiz 2

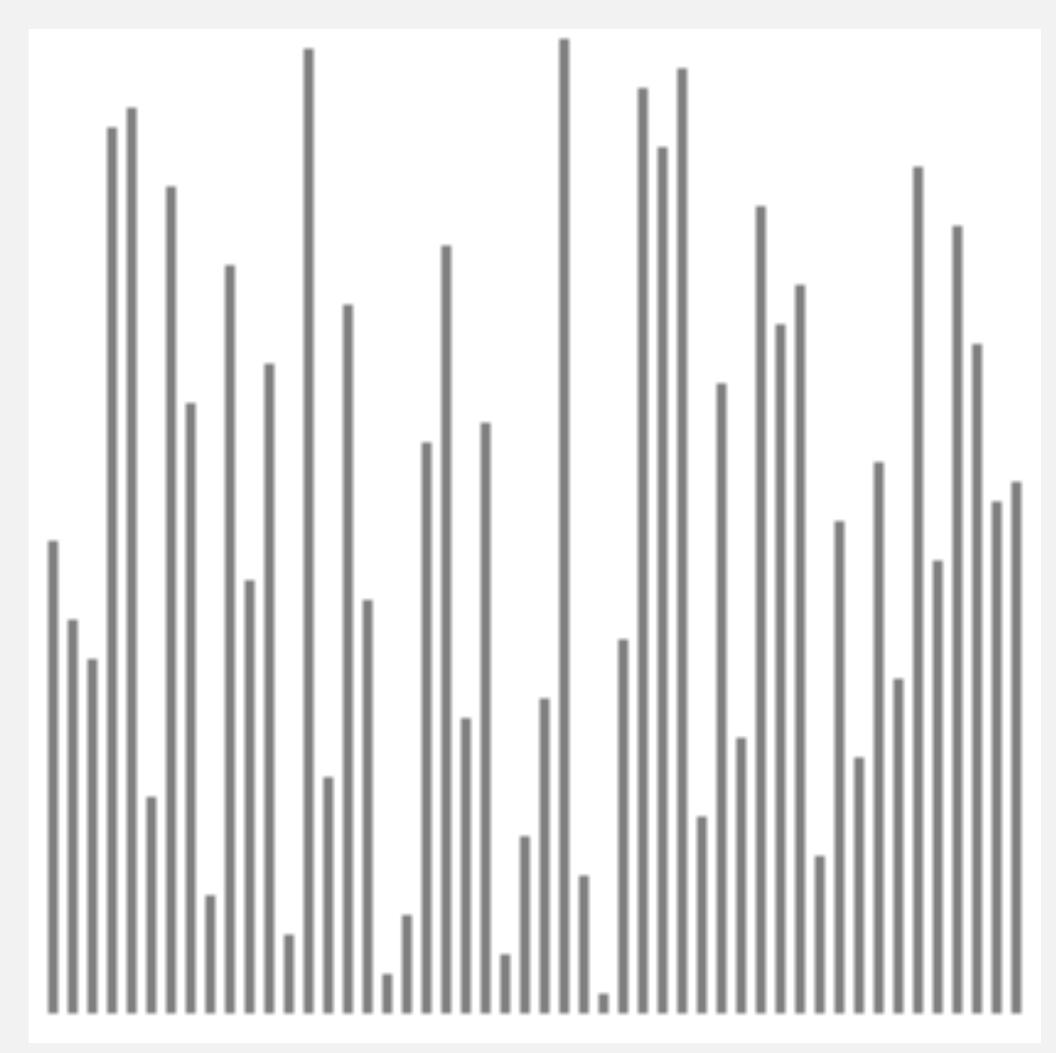


Which subarray lengths will arise when mergesorting an array of length 12?

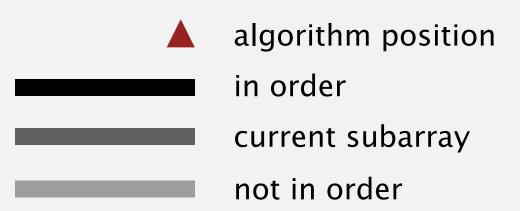
- **A.** { 1, 2, 3, 4, 6, 8, 12 }
- **B.** { 1, 2, 3, 6, 12 }
- **C.** { 1, 2, 4, 8, 12 }
- **D.** { 1, 3, 6, 9, 12 }

Mergesort: animation

50 random items

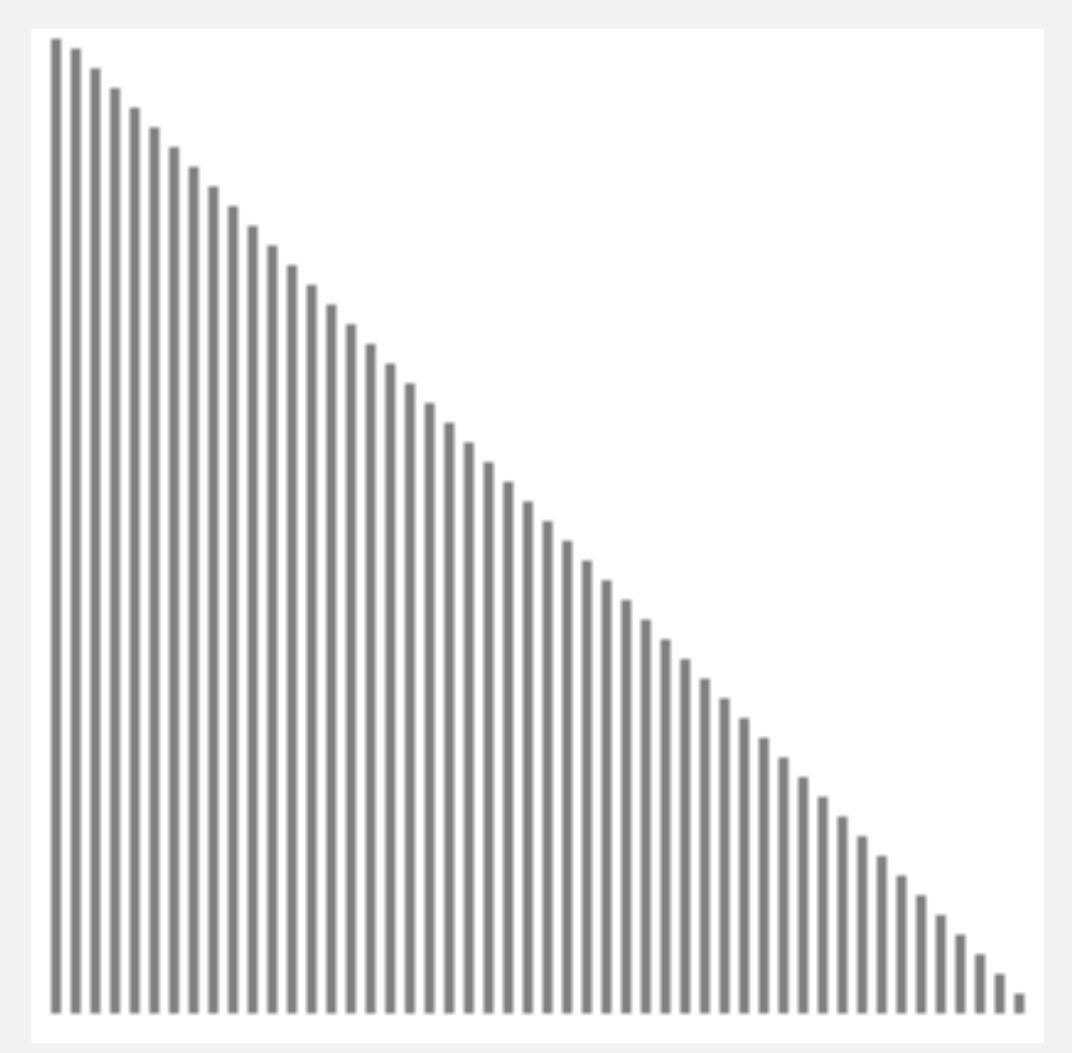


http://www.sorting-algorithms.com/merge-sort

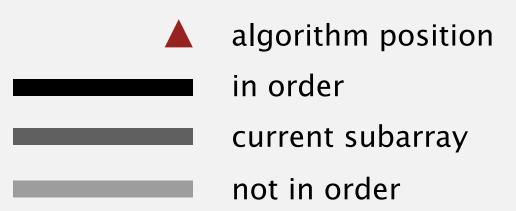


Mergesort: animation

50 reverse-sorted items



http://www.sorting-algorithms.com/merge-sort



Mergesort: empirical analysis

Running time estimates:

- Laptop executes 10⁸ compares/second.
- Supercomputer executes 10¹² compares/second.

	ins	ertion sort (n ²)	mergesort (n log n)				
computer	thousand	million	billion	thousand	million	billion		
home	instant	2.8 hours	317 years	instant	1 second	18 min		
super	instant	1 second	1 week	instant	instant	instant		

Bottom line. Good algorithms are better than supercomputers.

Mergesort analysis: number of compares

Proposition. Mergesort uses $\leq n \log_2 n$ compares to sort any array of length n.

Pf sketch. The number of compares C(n) to mergesort any array of length n satisfies the recurrence:

$$C(n) \le C(\lceil n/2 \rceil) + C(\lfloor n/2 \rfloor) + n-1$$
 for $n > 1$, with $C(1) = 0$.

sort
sort
left half

For simplicity: Assume *n* is a power of 2 and solve this recurrence:

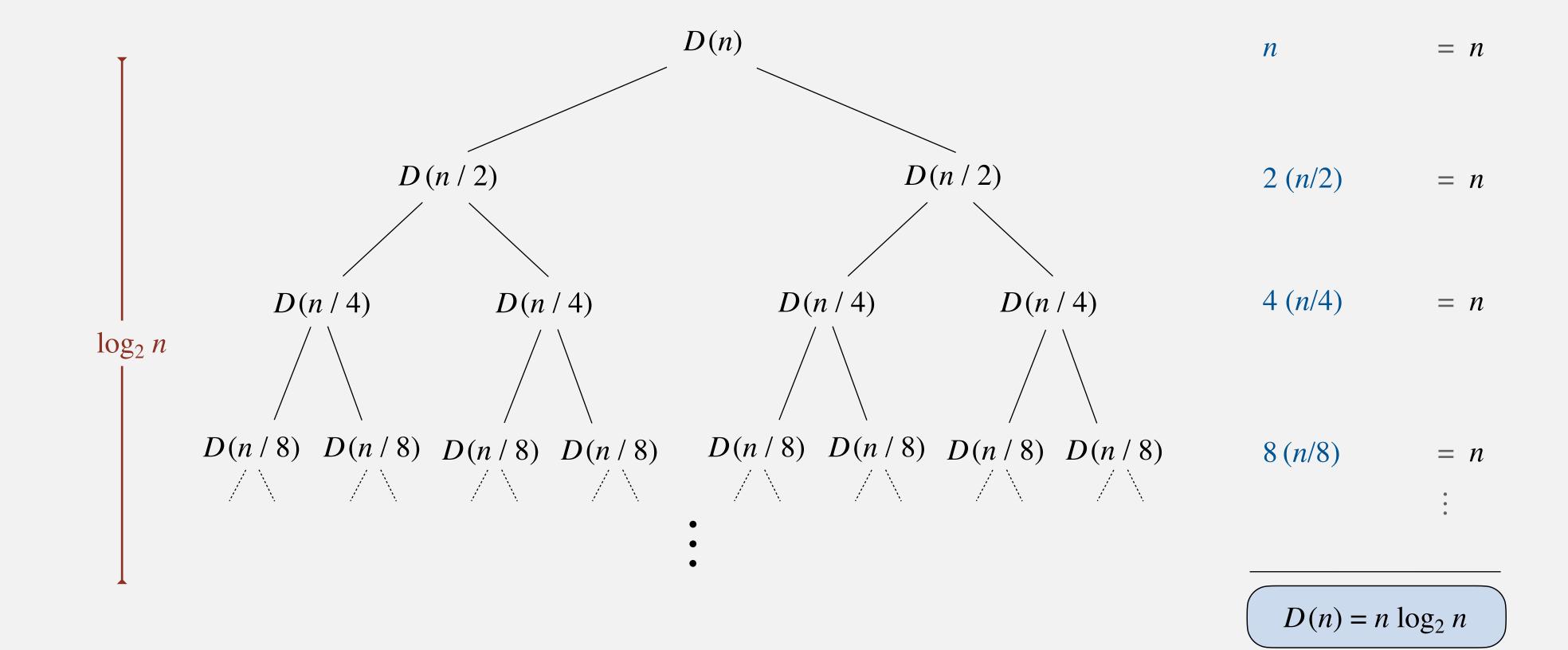
proposition holds even when *n* is not a power of 2 (but analysis cleaner in this case)

$$D(n) = 2 D(n/2) + n$$
, for $n > 1$, with $D(1) = 0$.

Divide-and-conquer recurrence

Proposition. If D(n) satisfies D(n) = 2D(n/2) + n for n > 1, with D(1) = 0, then $D(n) = n \log_2 n$.

Pf by picture. [assuming *n* is a power of 2]



15

Mergesort analysis: number of array accesses

Proposition. Mergesort makes $\Theta(n \log n)$ array accesses.

Pf sketch. The number of array accesses A(n) satisfies the recurrence:

$$A(n) = A([n/2]) + A([n/2]) + \Theta(n)$$
 for $n > 1$, with $A(1) = 0$.

Key point. Any algorithm with the following structure takes $\Theta(n \log n)$ time:

Famous examples. FFT and convolution, hidden-line removal, Kendall-tau distance, ...

Mergesort analysis: memory

Proposition. Mergesort uses $\Theta(n)$ extra space.

Pf. The length of the aux[] array is n, to handle the last merge.

two sorted subarrays



merged result



essentially negligible

Def. A sorting algorithm is in-place if it uses $\Theta(\log n)$ extra space (or less).

Ex. Insertion sort and selection sort.

Challenge 1 (not hard). Get by with an aux[] array of length $\sim \frac{1}{2} n$ (instead of n). Challenge 2 (very hard). In-place merge. [Kronrod 1969]

Mergesort quiz 3



Consider the following modified version of mergesort.

How much total memory is allocated over all recursive calls?

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

```
private static void sort(Comparable[] a, int lo, int hi)
{
   if (hi <= lo) return;
   int mid = (lo + hi) >>> 1;
   int n = hi - lo + 1;
   Comparable[] aux = new Comparable[n];
   sort(a, lo, mid);
   sort(a, mid+1, hi);
   merge(a, aux, lo, mid, hi);
}
```

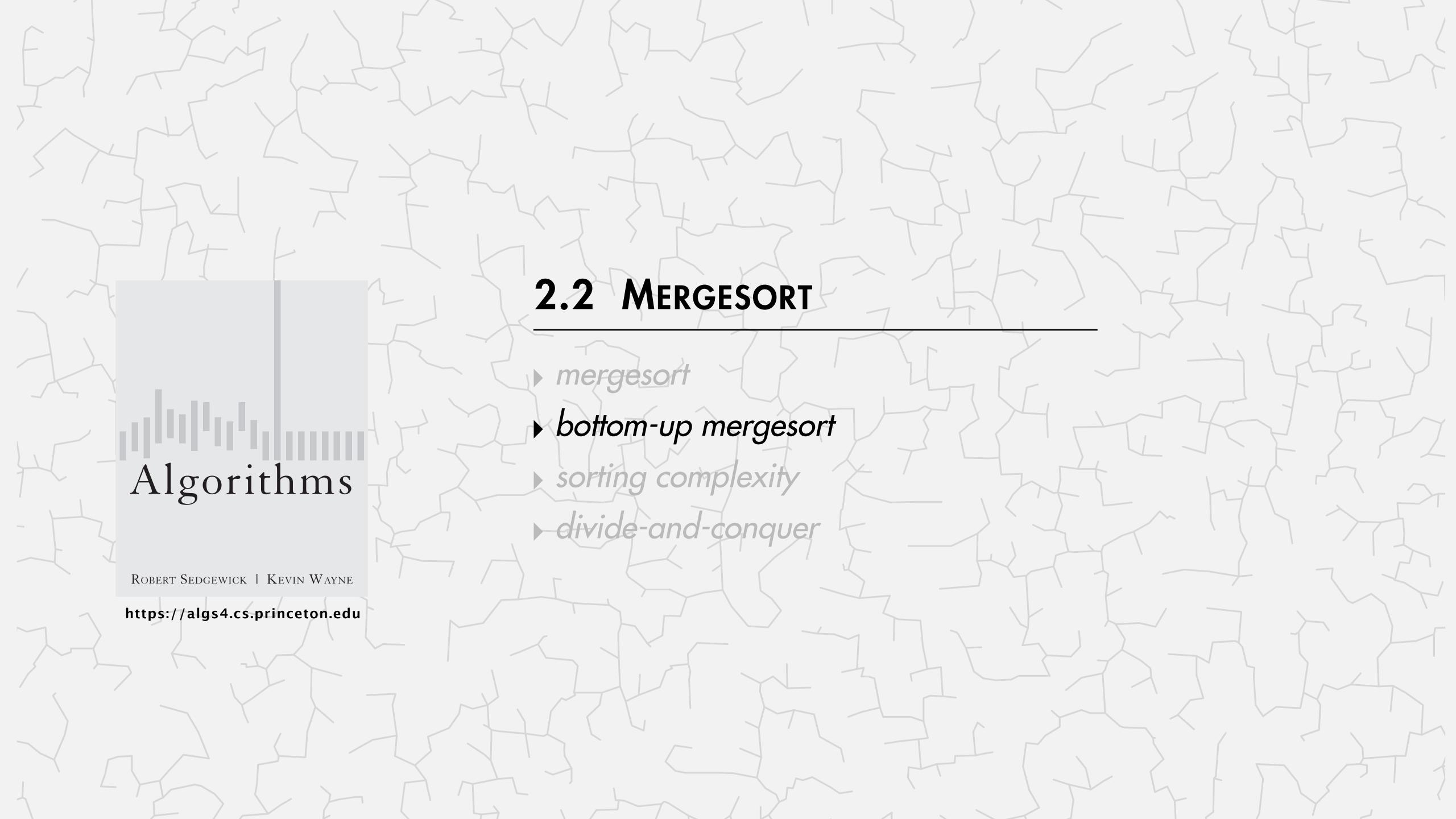
Mergesort: practical improvement

Use insertion sort for small subarrays.

- · Mergesort has too much overhead for tiny subarrays.
- Cutoff to insertion sort for ≈ 10 items.

```
private static void sort(...)
{
    if (hi <= lo + CUTOFF - 1)
    {
        Insertion.sort(a, lo, hi);
        return;
    }

    int mid = (lo + hi) >>> 1;
    sort (a, aux, lo, mid);
    sort (a, aux, mid+1, hi);
    merge(a, aux, lo, mid, hi);
}
```



Bottom-up mergesort

Basic plan.

- Pass through array, merging subarrays of length 1.
- Repeat for subarrays of length 2, 4, 8,

```
a[i]
     sz = 1
     merge(a, aux, 0, 0, 1)
     merge(a, aux, 2, 2, 3)
     merge(a, aux, 4, 4, 5)
     merge(a, aux, 6, 6,
     merge(a, aux, 8, 8, 9)
     merge(a, aux, 10, 10, 11)
     merge(a, aux, 12, 12, 13)
     merge(a, aux, 14, 14, 15)
   sz = 2
   merge(a, aux, 0, 1, 3)
   merge(a, aux, 4, 5, 7)
   merge(a, aux, 8, 9, 11)
   merge(a, aux, 12, 13, 15)
 sz = 4
 merge(a, aux, 0, 3, 7)
                              E E G M O R R S A E E L M P T X
 merge(a, aux, 8, 11, 15)
sz = 8
merge(a, aux, 0, 7, 15) A E E E G L M M O P R R S T X
```

Bottom-up mergesort: Java implementation

```
public class MergeBU
   private static void merge(...)
   { /* as before */ }
  public static void sort(Comparable[] a)
     int n = a.length;
     Comparable[] aux = new Comparable[n];
     for (int sz = 1; sz < n; sz = sz+sz)
        for (int lo = 0; lo < n-sz; lo += sz+sz)
           merge(a, aux, lo, lo+sz-1, Math.min(lo+sz+sz-1, n-1));
                                             hi
                                 mid
```

Proposition. At most $n \log_2 n$ compares; $\Theta(n)$ extra space.

Bottom line. Simple and non-recursive version of mergesort.

Mergesort quiz 4

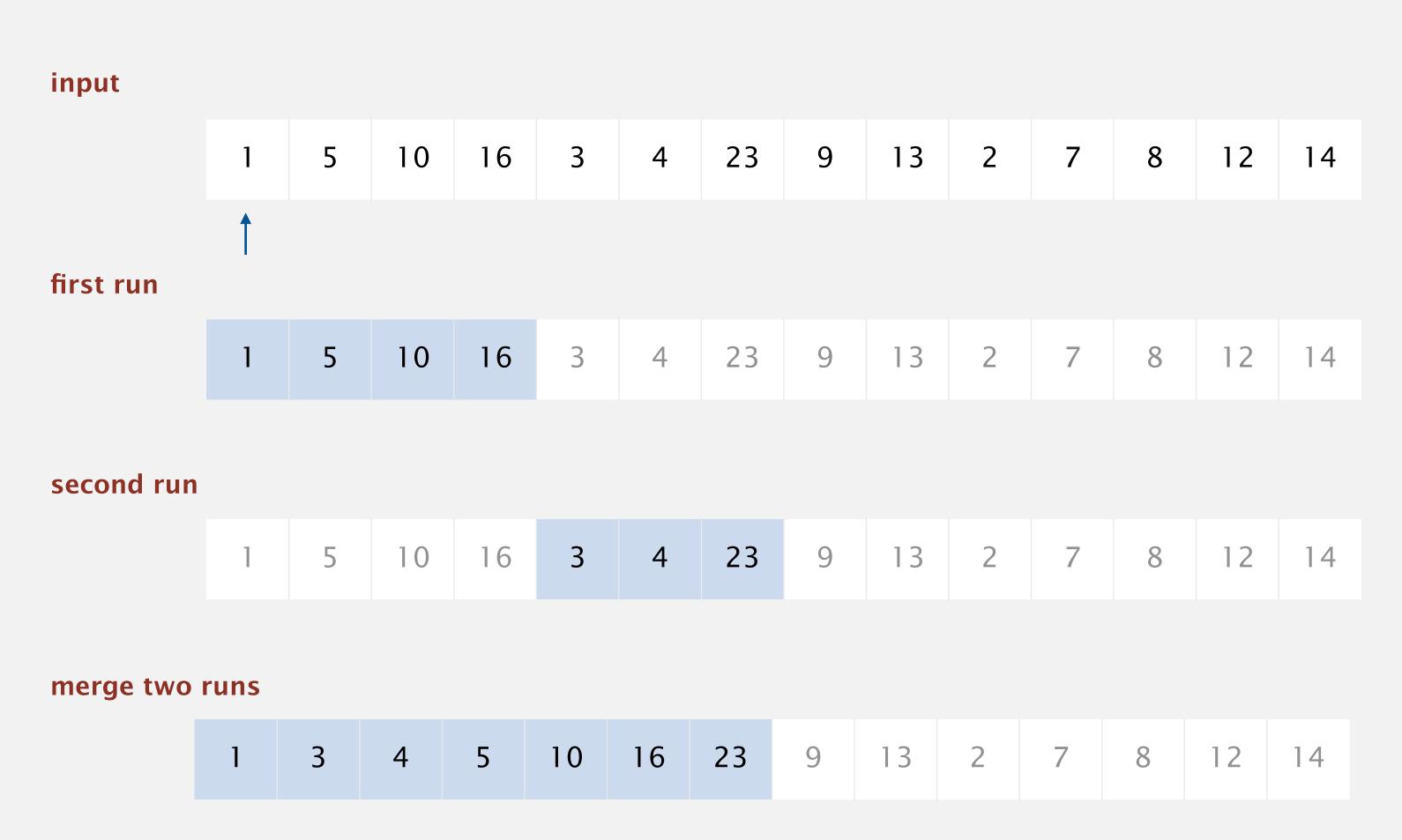


Which is faster in practice for $n=2^{20}$, top-down mergesort or bottom-up mergesort?

- A. Top-down (recursive) mergesort.
- B. Bottom-up (non-recursive) mergesort.
- C. No difference.
- **D.** I don't know.

Natural mergesort

Idea. Exploit pre-existing order by identifying naturally occurring runs.



Tradeoff. Fewer passes vs. extra compares per pass to identify runs.

Timsort (2002)

- Natural mergesort.
- Use binary insertion sort to make initial runs (if needed).
- A few more clever optimizations.

Intro

This describes an adaptive, stable, natural mergesort, modestly called timsort (hey, I earned it <wink>). It has supernatural performance on many kinds of partially ordered arrays (less than lg(n!) comparisons needed, and as few as n-1), yet as fast as Python's previous highly tuned samplesort hybrid on random arrays.

In a nutshell, the main routine marches over the array once, left to right, alternately identifying the next run, then merging it into the previous runs "intelligently". Everything else is complication for speed, and some hard-won measure of memory efficiency.

. . .



Tim Peters

Consequence. Only $\Theta(n)$ compares on many arrays with pre-existing order.

Widely used. Python, Java, GNU Octave, Android,

Timsort bug (February 2015)



Proving that Android's, Java's and Python's sorting algorithm is broken (and showing how to fix it)

Tim Peters developed the Timsort hybrid sorting algorithm in 2002. It is a clever combination of ideas from merge sort and insertion sort, and designed to perform well on real world data. TimSort was first developed for Python, but later ported to Java (where it appears as java.util.Collections.sort and java.util.Arrays.sort) by Joshua Bloch (the designer of Java Collections who also pointed out that most binary search algorithms were broken). TimSort is today used as the default sorting algorithm for Android SDK, Sun's JDK and OpenJDK. Given the popularity of these platforms this means that the number of computers, cloud services and mobile phones that use TimSort for sorting is well into the billions.

Timsort bug (May 2018)



JDK / JDK-8203864

Execution error in Java's Timsort

Details

Status: RESOLVED

Priority: 3 P3

Resolution: Fixed

Affects Version/s: None

Fix Version/s: 11

Component/s: core-libs

Labels: None

Subcomponent: java.util:collections

Introduced In Version: 6

Resolved In Build: b20

Description

Carine Pivoteau wrote:

While working on a proper complexity analysis of the algorithm, we realised that there was an error in the last paper reporting such a bug (http://envisage-project.eu/wp-content/uploads/2015/02/sorting.pdf).

This implies that the correction implemented in the Java source code (changing Timsort stack size) is wrong and that it is still possible to make it break. This is explained in full details in our analysis: https://arxiv.org/pdf/1805.08612.pdf.

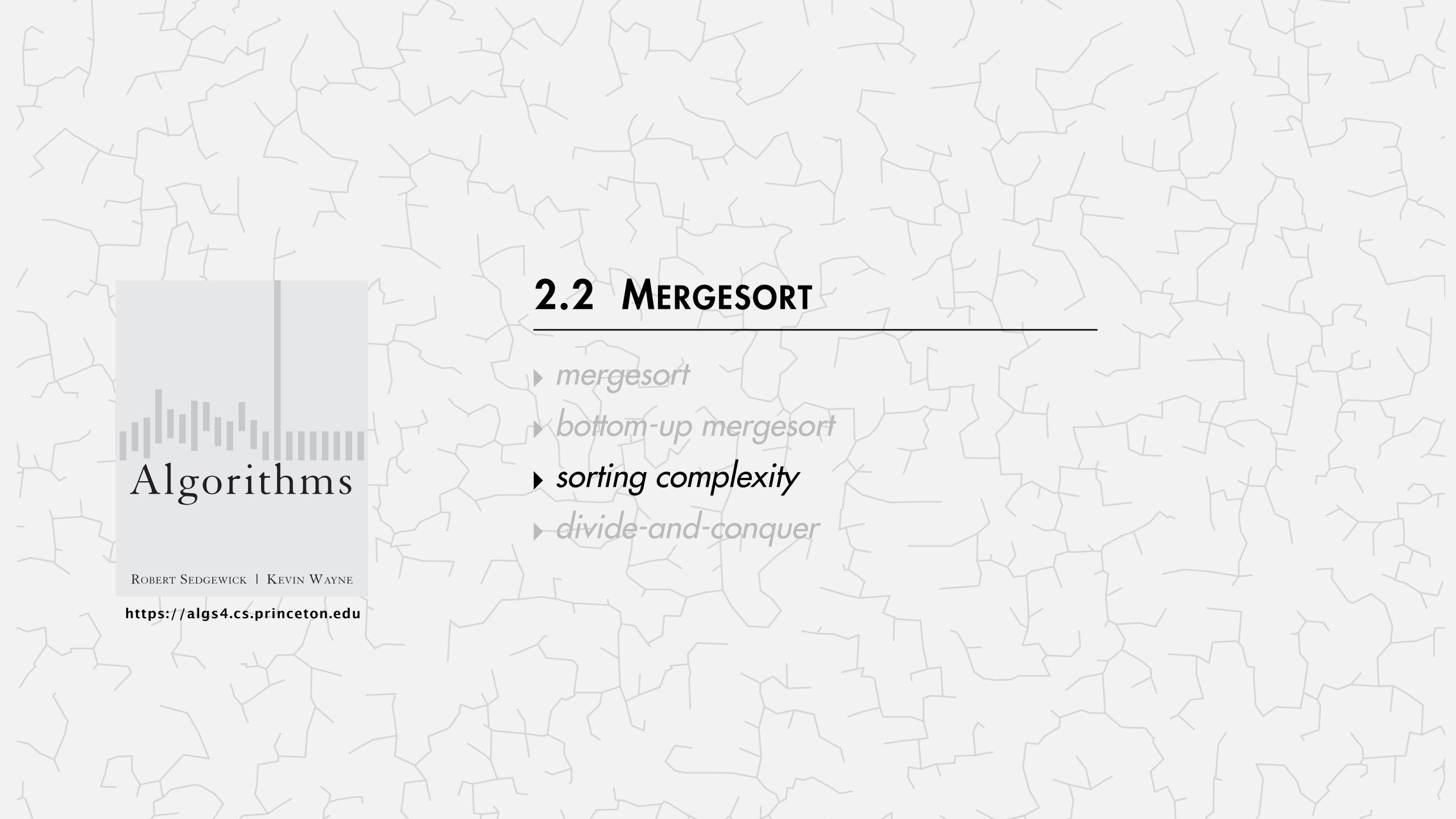
We understand that coming upon data that actually causes this error is very unlikely, but we thought you'd still like to know and do something about it. As the authors of the previous article advocated for, we strongly believe that you should consider modifying the algorithm as explained in their article (and as was done in Python) rather than trying to fix the stack size.

Sorting summary

	in-place?	stable?	best	average	worst	remarks
selection	✓		½ n ²	$\frac{1}{2}$ n^2	½ n ²	n exchanges
insertion	✓	•	n	½ n ²	½ n ²	use for small <i>n</i> or partially sorted
merge		✓	$\frac{1}{2} n \log_2 n$	$n \log_2 n$	$n \log_2 n$	$\Theta(n \log n)$ guarantee; stable
timsort		•	n	$n \log_2 n$	$n \log_2 n$	improves mergesort when pre-existing order
?	✓	✓	n	$n \log_2 n$	$n \log_2 n$	holy sorting grail

number of compares to sort an array of n elements

asserted but not proved until a decade after widespread adoption (!)



Computational complexity

A framework to study efficiency of algorithms for solving a particular problem X.

Model of computation. Allowable operations.

Cost model. Operation counts.

Upper bound. Cost guarantee provided by some algorithm for *X*.

Lower bound. Proven limit on cost guarantee for all algorithms for *X*.

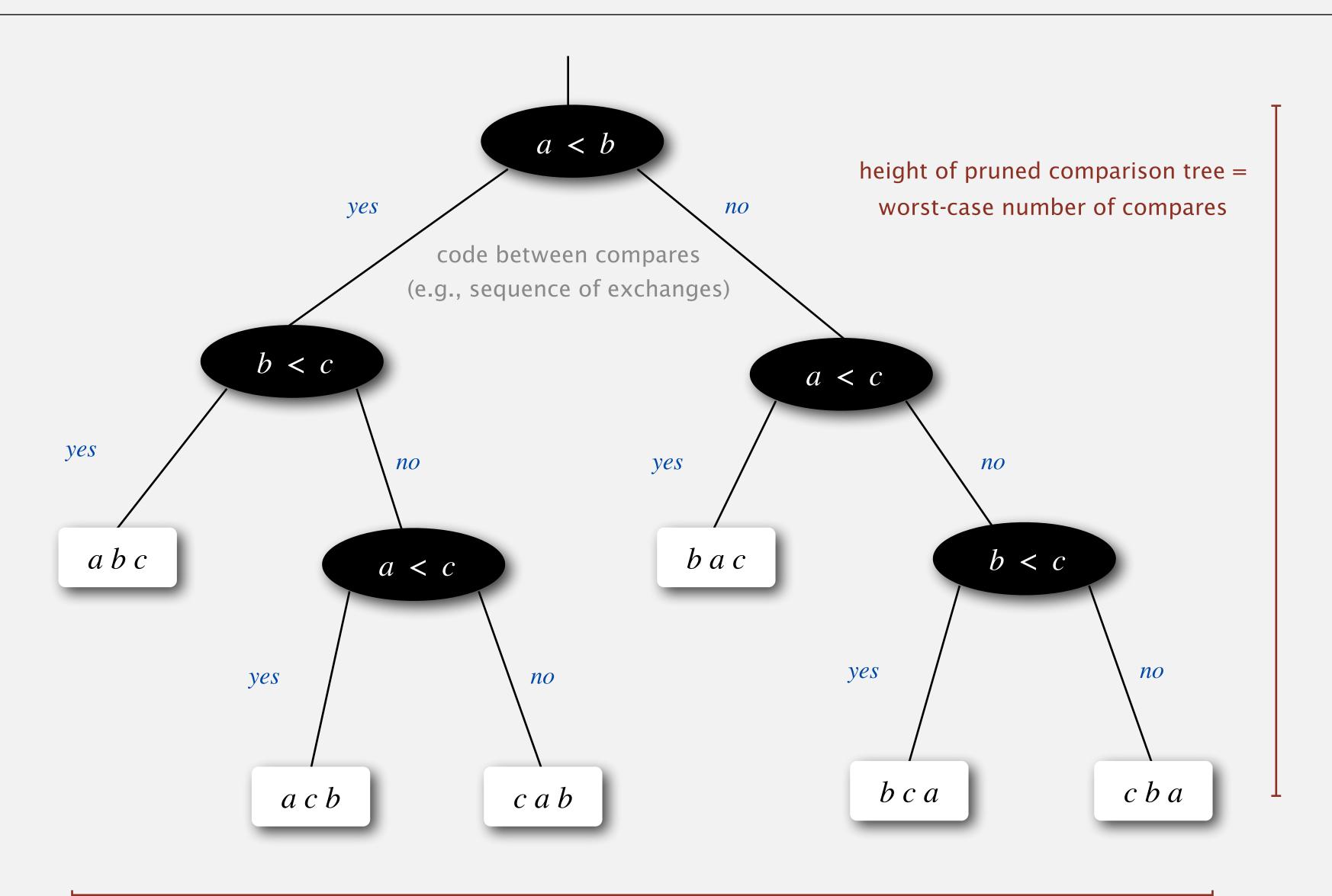
Optimal algorithm. Algorithm with best possible cost guarantee for *X*.



model of computation	comparison tree	~	can gain knowledge about input only through compares				
cost model	# compares		(e.g., Java Comparable framework)				
upper bound	$\sim n \log_2 n$	—	from mergesort				
lower bound	?						
optimal algorithm	?						

computational complexity of sorting

Comparison tree (for 3 distinct keys a, b, and c)

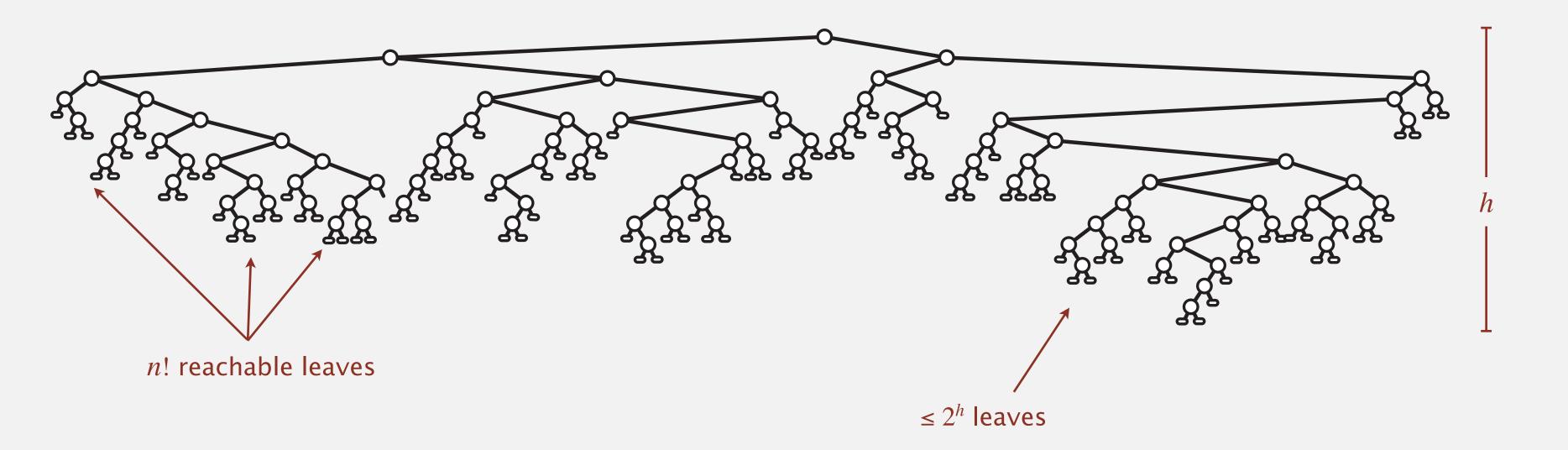


Compare-based lower bound for sorting

Proposition. In the worst case, any compare-based sorting algorithm must make at least $\log_2(n!) \sim n \log_2 n$ compares.

Pf.

- Assume array consists of n distinct values a_1 through a_n .
- n! different orderings $\Rightarrow n!$ reachable leaves.
- Worst-case number of compares = height h of pruned comparison tree.
- Binary tree of height h has $\leq 2^h$ leaves.

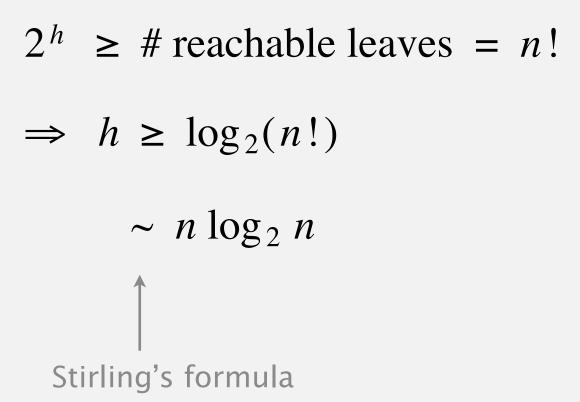


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Complexity of sorting

Model of computation. Allowable operations.

Cost model. Operation count(s).

Upper bound. Cost guarantee provided by some algorithm for *X*.

Lower bound. Proven limit on cost guarantee of all algorithms for *X*.

Optimal algorithm. Algorithm with best possible cost guarantee for *X*.

model of computation	comparison tree		
cost model	# compares		
upper bound	$\sim n \log_2 n$		 from mergesort
lower bound	$(\sim n \log_2 n)$	~	 comparison tree
optimal algorithm	mergesort		

complexity of sorting

First goal of algorithm design: optimal algorithms.

Complexity results in context

Compares? Mergesort is optimal with respect to number compares.

Space? Mergesort is not optimal with respect to space usage.

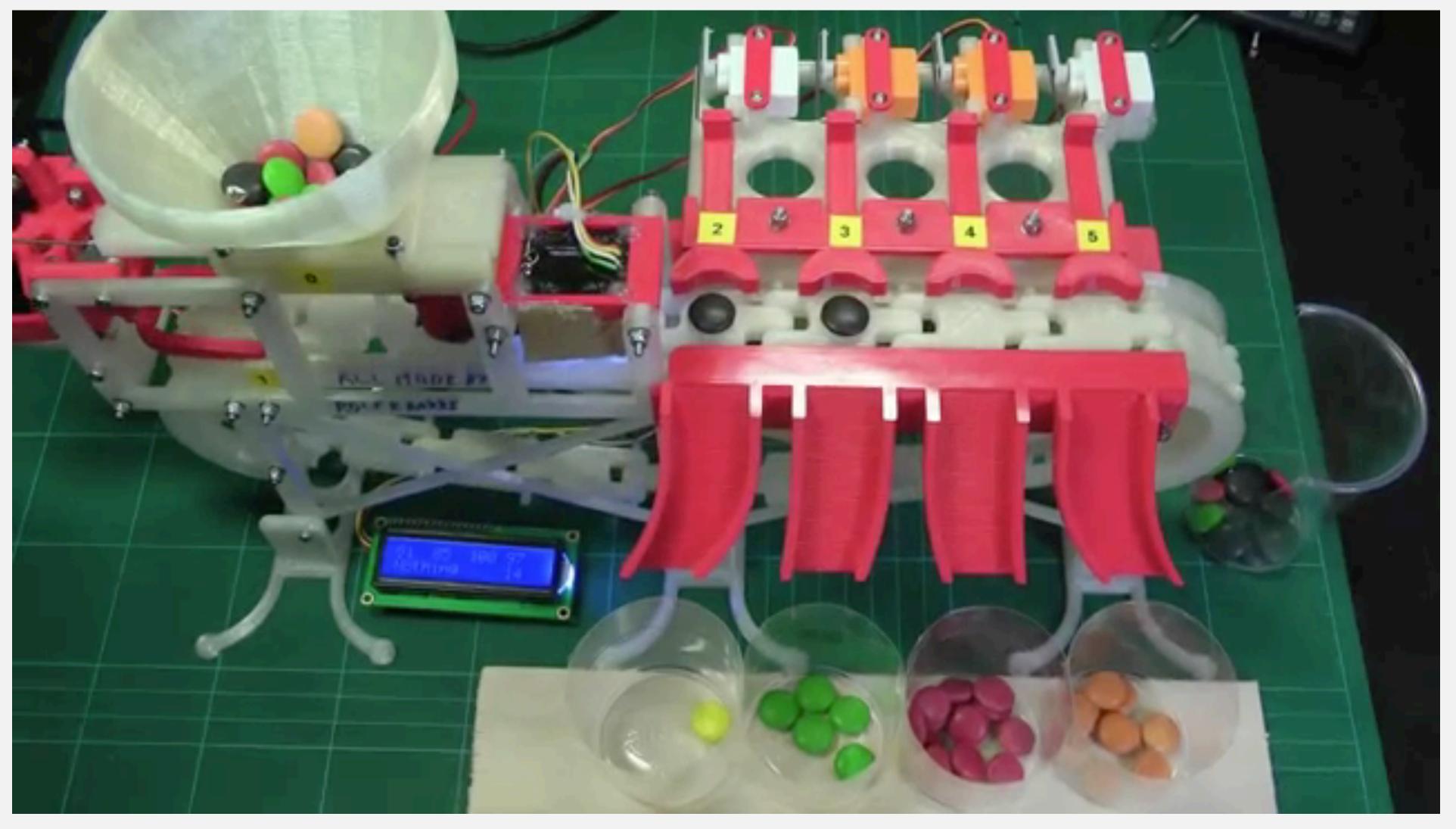


Lesson. Use theory as a guide.

- Ex. Design sorting algorithm that makes $\sim \frac{1}{2} n \log_2 n$ compares in worst case?
- Ex. Design sorting algorithm that makes $\Theta(n \log n)$ compares and uses $\Theta(1)$ extra space.



Q. Why doesn't this Skittles sorter violate the sorting lower bound?



Complexity results in context (continued)

Lower bound may not hold if the algorithm can take advantage of:

The initial order of the input array.

Ex: insertion sort makes only $\Theta(n)$ compares on partially sorted arrays.

• The distribution of key values.

Ex: 3-way quicksort makes only $\Theta(n)$ compares on arrays with a small number of distinct keys. [stay tuned]

• The representation of the keys.

Ex: radix sorts do not make any key compares; they access the data via individual characters/digits. [stay tuned]

Asymptotic notations

notation	provides	example	shorthand for	Warning
tilde (~)	leading term	$\sim \frac{1}{2} n^2$	$\frac{1}{2} n^2$ $\frac{1}{2} n^2 + 3n + 22$ $\frac{1}{2} n^2 + n \log_2 n$	ignore lower-order terms
big Theta (Θ)	order of growth	$\Theta(n^2)$	$\frac{1/2}{7} n^2 + n^{1/2}$ $5 n^2 - 3 n$	also ignore leading coefficient
big O (O)	upper bound	$O(n^2)$	$10 n^2$ $22 n$ $\log_2 n$	$\Theta(n^2)$ or smaller
big Omega (Ω)	lower bound	$\Omega(n^2)$	$n^{3} + 3n$ 2^{n}	$\Theta(n^2)$ or larger

Warning: many programmers
wisuse O to mean Θ .

Mergesort quiz 5



Which of the following correctly describes the function $f(n) = 3n^2 + 30n$?

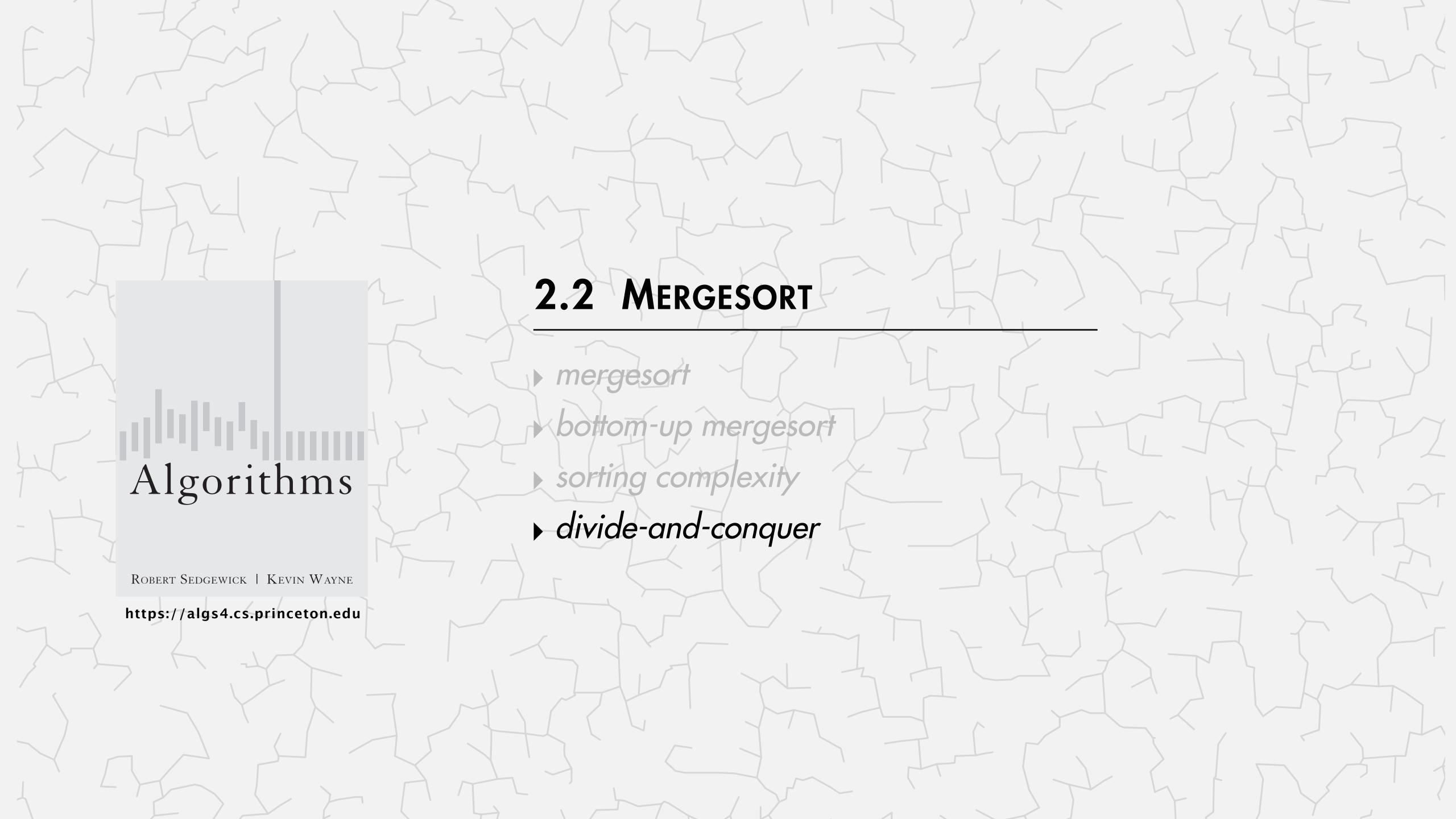
- $\sim n^2$
- **B.** $\Theta(n)$
- C. $O(n^3)$
- **D.** All of the above.
- E. None of the above.

SORTING LOWER BOUND



Interviewer. Give a formal description of the sorting lower bound for sorting arrays of n elements.





SORTING A LINKED LIST



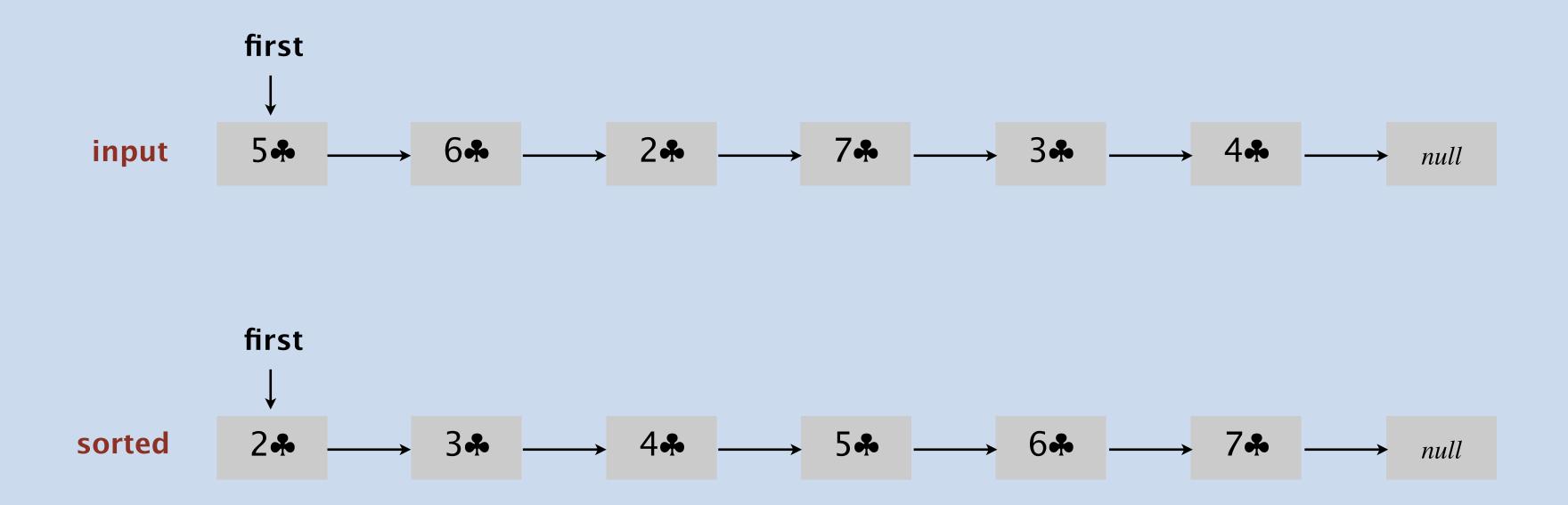
Problem. Given a singly linked list, rearrange its nodes in sorter order.

Application. Sort list of inodes to garbage collect in Linux kernel.

Version 0. $\Theta(n \log n)$ time, $\Theta(n)$ extra space.

Version 1. $\Theta(n \log n)$ time, $\Theta(\log n)$ extra space.

Version 2. $\Theta(n \log n)$ time, $\Theta(1)$ extra space.



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