

# Algorithm Efficiency and Sorting

# Measuring the Efficiency of Algorithms

- Analysis of algorithms
  - Provides tools for contrasting the efficiency of different methods of solution
- A comparison of algorithms
  - Should focus of significant differences in efficiency
  - Should not consider reductions in computing costs due to clever coding tricks

# Measuring the Efficiency of Algorithms

- Three difficulties with comparing programs instead of algorithms
  - How are the algorithms coded?
  - What computer should you use?
  - What data should the programs use?
- Algorithm analysis should be independent of
  - Specific implementations
  - Computers
  - Data

# The Execution Time of Algorithms

- Counting an algorithm's operations is a way to access its efficiency
  - An algorithm's execution time is related to the number of operations it requires
  - Examples
    - Traversal of a linked list
    - The Towers of Hanoi
    - Nested Loops

# Algorithm Growth Rates

- An algorithm's time requirements can be measured as a function of the problem size
- An algorithm's growth rate
  - Enables the comparison of one algorithm with another
  - Examples
    - Algorithm A requires time proportional to  $n^2$
    - Algorithm B requires time proportional to  $n$
- Algorithm efficiency is typically a concern for large problems only

# Algorithm Growth Rates

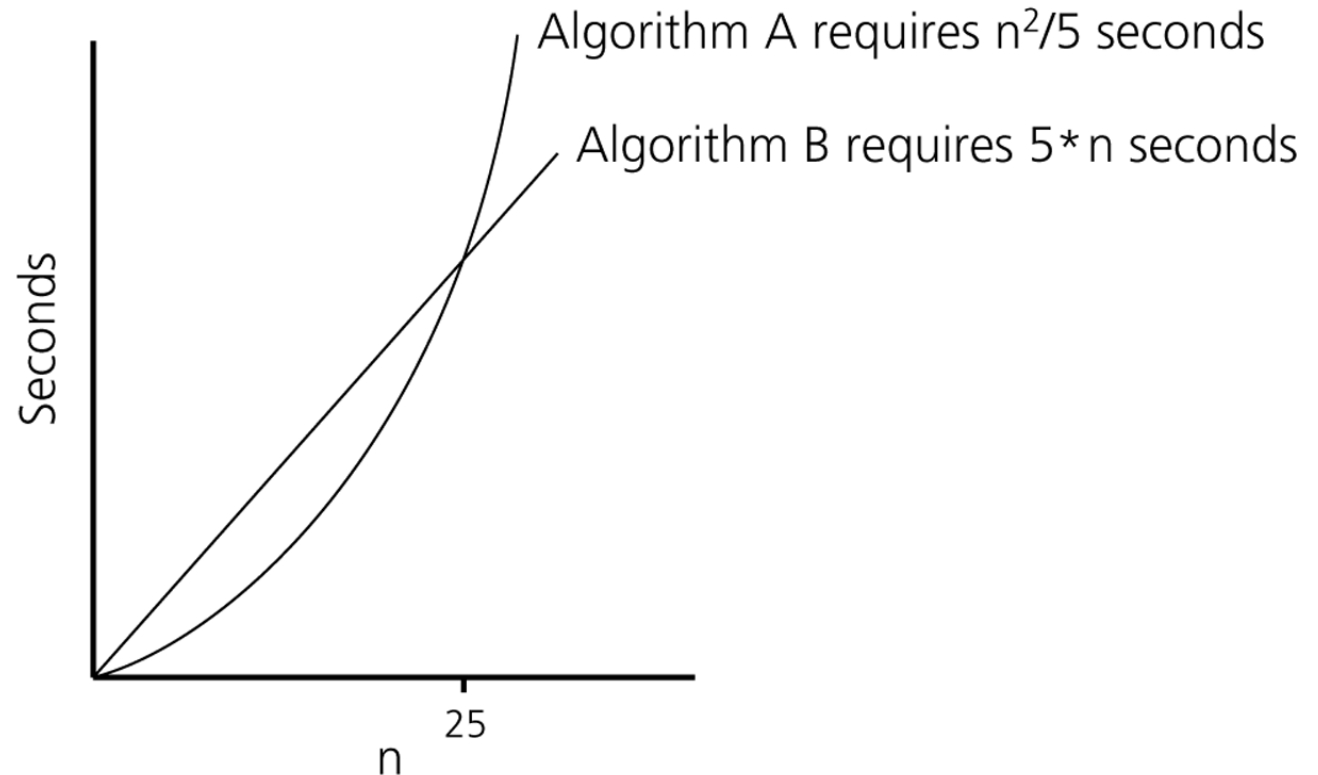


Figure 10-1

Time requirements as a function of the problem size  $n$

# Order-of-Magnitude Analysis and Big O Notation

- Definition of the order of an algorithm
  - Algorithm A is order  $f(n)$  – denoted  $O(f(n))$  – if constants  $k$  and  $n_0$  exist such that A requires no more than  $k * f(n)$  time units to solve a problem of size  $n \geq n_0$
- Growth-rate function
  - A mathematical function used to specify an algorithm's order in terms of the size of the problem
- Big O notation
  - A notation that uses the capital letter O to specify an algorithm's order
  - Example:  $O(f(n))$

# Order-of-Magnitude Analysis and Big O Notation

(a)

Function	n					
	10	100	1,000	10,000	100,000	1,000,000
1	1	1	1	1	1	1
$\log_2 n$	3	6	9	13	16	19
$n$	$10$	$10^2$	$10^3$	$10^4$	$10^5$	$10^6$
$n * \log_2 n$	30	664	9,965	$10^5$	$10^6$	$10^7$
$n^2$	$10^2$	$10^4$	$10^6$	$10^8$	$10^{10}$	$10^{12}$
$n^3$	$10^3$	$10^6$	$10^9$	$10^{12}$	$10^{15}$	$10^{18}$
$2^n$	$10^3$	$10^{30}$	$10^{301}$	$10^{3,010}$	$10^{30,103}$	$10^{301,030}$

Figure 10-3a

A comparison of growth-rate functions: a) in tabular form



# Order-of-Magnitude Analysis and Big O Notation

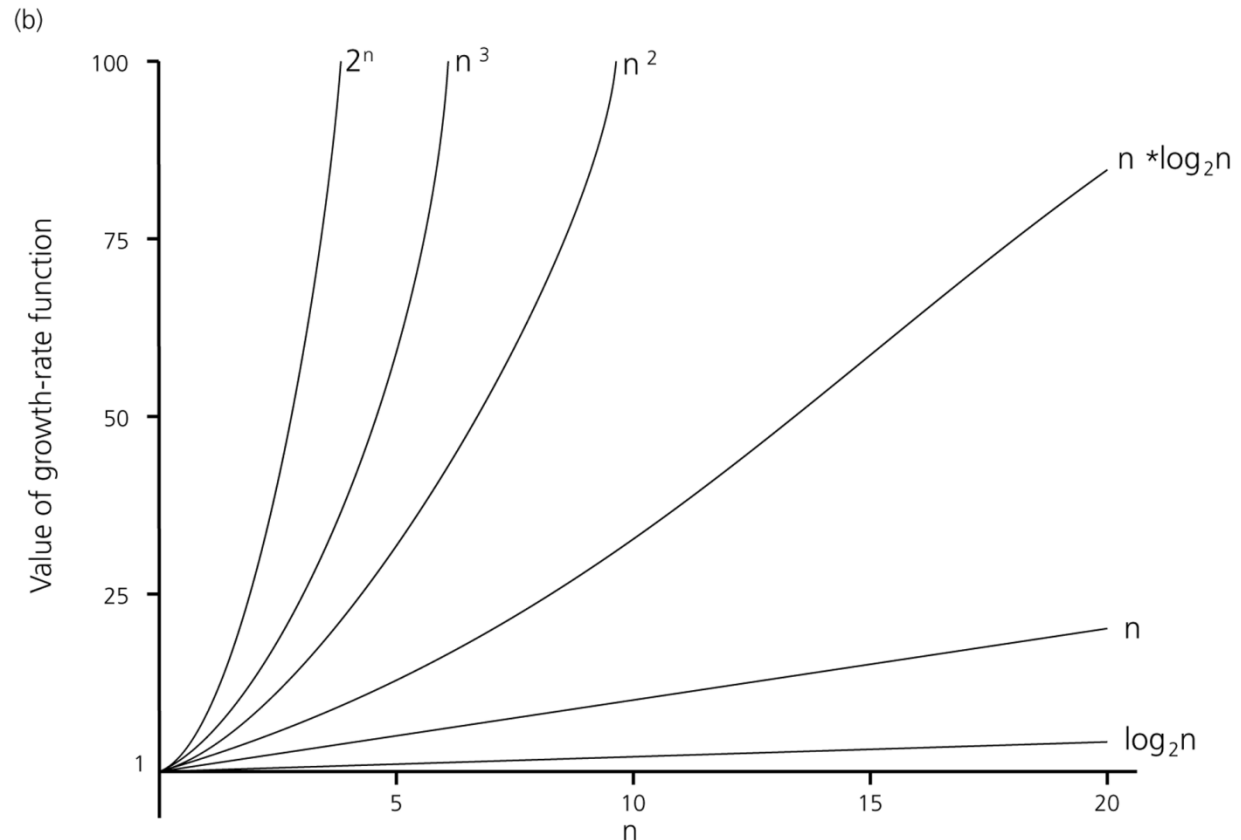


Figure 10-3b

A comparison of growth-rate functions: b) in graphical form

# Order-of-Magnitude Analysis and Big O Notation

- Order of growth of some common functions  
 $O(1) < O(\log_2 n) < O(n) < O(n * \log_2 n) < O(n^2) < O(n^3) < O(2^n)$
- Properties of growth-rate functions
  - You can ignore low-order terms
  - You can ignore a multiplicative constant in the high-order term
  - $O(f(n)) + O(g(n)) = O(f(n) + g(n))$

# Order-of-Magnitude Analysis and Big O Notation

- Worst-case and average-case analyses
  - An algorithm can require different times to solve different problems of the same size
    - Worst-case analysis
      - A determination of the maximum amount of time that an algorithm requires to solve problems of size  $n$
    - Average-case analysis
      - A determination of the average amount of time that an algorithm requires to solve problems of size  $n$

# Keeping Your Perspective

- Throughout the course of an analysis, keep in mind that you are interested only in significant differences in efficiency
- When choosing an ADT's implementation, consider how frequently particular ADT operations occur in a given application
- Some seldom-used but critical operations must be efficient

# Keeping Your Perspective

- If the problem size is always small, you can probably ignore an algorithm's efficiency
- Weigh the trade-offs between an algorithm's time requirements and its memory requirements
- Compare algorithms for both style and efficiency
- Order-of-magnitude analysis focuses on large problems

# The Efficiency of Searching Algorithms

- Sequential search
  - Strategy
    - Look at each item in the data collection in turn, beginning with the first one
    - Stop when
      - You find the desired item
      - You reach the end of the data collection

# The Efficiency of Searching Algorithms

- Sequential search
  - Efficiency
    - Worst case:  $O(n)$
    - Average case:  $O(n)$
    - Best case:  $O(1)$

# The Efficiency of Searching Algorithms

- Binary search
  - Strategy
    - To search a sorted array for a particular item
      - Repeatedly divide the array in half
      - Determine which half the item must be in, if it is indeed present, and discard the other half
  - Efficiency
    - Worst case:  $O(\log_2 n)$
- For large arrays, the binary search has an enormous advantage over a sequential search



# Sorting Algorithms and Their Efficiency

- Sorting
  - A process that organizes a collection of data into either ascending or descending order
- Categories of sorting algorithms
  - An internal sort
    - Requires that the collection of data fit entirely in the computer's main memory
  - An external sort
    - The collection of data will not fit in the computer's main memory all at once but must reside in secondary storage

# Sorting Algorithms and Their Efficiency

- Data items to be sorted can be
  - Integers
  - Character strings
  - Objects
- Sort key
  - The part of a record that determines the sorted order of the entire record within a collection of records

# Selection Sort

- Selection sort
  - Strategy
    - Select the largest item and put it in its correct place
    - Select the next largest item and put it in its correct place, etc.

Shaded elements are selected;  
boldface elements are in order.

Initial array:	29	10	14	37	13
After 1 <sup>st</sup> swap:	29	10	14	13	<b>37</b>
After 2 <sup>nd</sup> swap:	13	10	14	<b>29</b>	<b>37</b>
After 3 <sup>rd</sup> swap:	13	10	<b>14</b>	<b>29</b>	<b>37</b>
After 4 <sup>th</sup> swap:	<b>10</b>	<b>13</b>	<b>14</b>	<b>29</b>	<b>37</b>

Figure 10-4

A selection sort of an array of  
five integers

# Selection Sort

- Analysis
  - Selection sort is  $O(n^2)$
- Advantage of selection sort
  - It does not depend on the initial arrangement of the data
- Disadvantage of selection sort
  - It is only appropriate for small  $n$

## Selection Code

// This code will compile with warnings about unchecked exceptions

```
public class SortsClass {

    public static void selectionSort(Comparable[] theArray,
    int n) {
// -----
// Sorts the items in an array into ascending order.
// Precondition: theArray is an array of n items.
// Postcondition: theArray is sorted into
// ascending order.
// Calls: indexOfLargest.
// -----
// last = index of the last item in the subarray of
// items yet to be sorted
// largest = index of the largest item found
```

## Selection Code

```
for (int last = n-1; last >= 1; last--) {  
    // Invariant: theArray[last+1..n-1] is sorted  
    // and > theArray[0..last]  
    // select largest item in theArray[0..last]  
    int largest = indexOfLargest(theArray, last+1);  
    // swap largest item theArray[largest] with  
    // theArray[last]  
    Comparable temp = theArray[largest];  
    theArray[largest] = theArray[last];  
    theArray[last] = temp;  
} // end for  
} // end selectionSort
```

## Selection Code

```
private static int indexOfLargest(Comparable[] theArray,
    int size) {
    // -----
    // Finds the largest item in an array.
    // Precondition: theArray is an array of size items;
    // size >= 1.
    // Postcondition: Returns the index of the largest
    // item in the array.
    // -----
    int indexSoFar = 0; // index of largest item found so far
    // Invariant: theArray[indexSoFar]>=theArray[0..currIndex-1]
    for (int currIndex = 1; currIndex < size; ++currIndex) {
        if (theArray[currIndex].compareTo(theArray[indexSoFar])>0) {
            indexSoFar = currIndex;
        } // end if
    } // end for
    return indexSoFar; // index of largest item
} // end indexOfLargest
```

# Performance Analysis



# Bubble Sort

- Bubble sort
  - Strategy
    - Compare adjacent elements and exchange them if they are out of order
      - Comparing the first two elements, the second and third elements, and so on, will move the largest (or smallest) elements to the end of the array
      - Repeating this process will eventually sort the array into ascending (or descending) order

# Bubble Sort

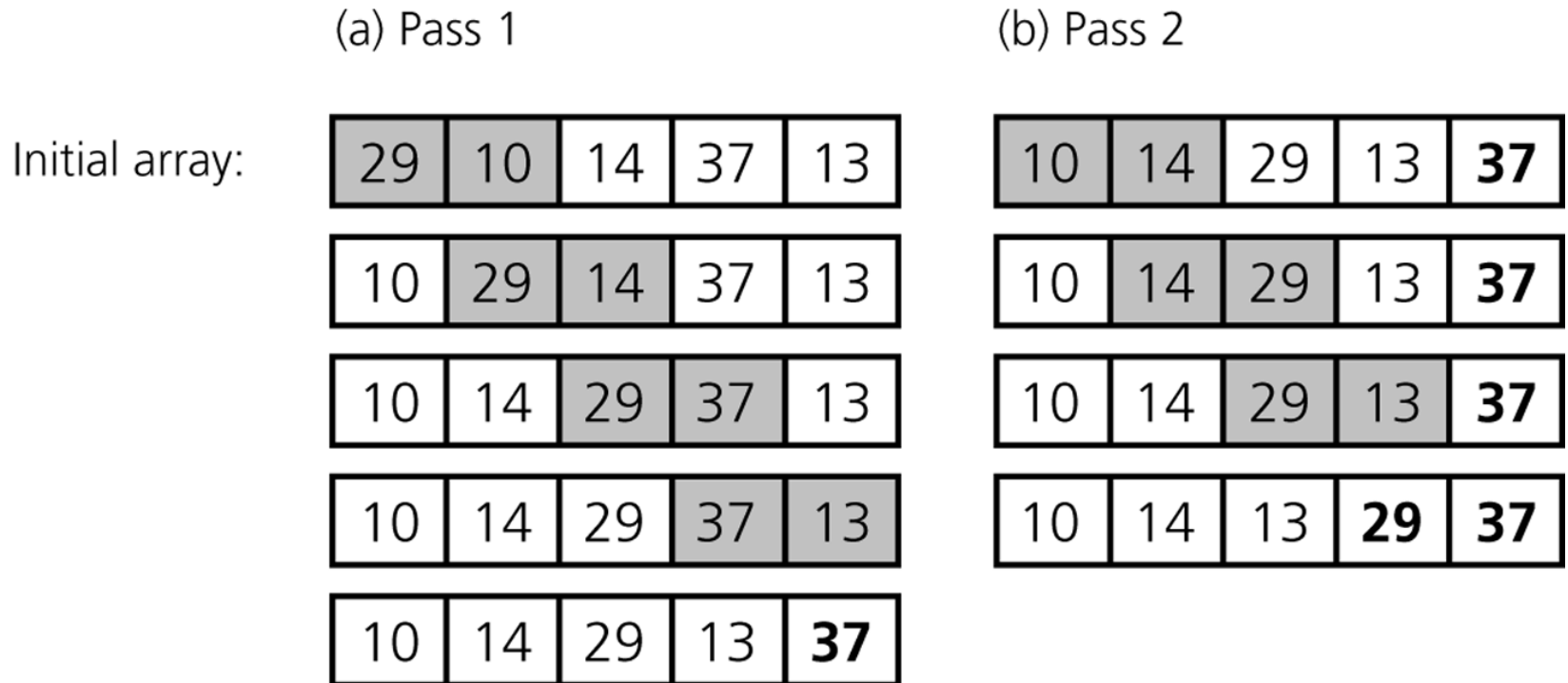


Figure 10-5

The first two passes of a bubble sort of an array of five integers: a) pass 1;

b) pass 2

## Bubble Sort Code

```
public static void bubbleSort(Comparable[] theArray, int n) {  
  
    // -----  
    // Sorts the items in an array into ascending order.  
    // Precondition: theArray is an array of n items.  
    // Postcondition: theArray is sorted into ascending  
    // order.  
    // -----  
    boolean sorted = false; // false when swaps occur  
    for (int pass = 1; (pass < n) && !sorted; ++pass) {  
        // Invariant: theArray[n+1-pass..n-1] is sorted  
        // and > theArray[0..n-pass]  
        sorted = true; // assume sorted  
        for (int index = 0; index < n-pass; ++index) {  
            // Invariant: theArray[0..index-1] <= theArray[index]  
            int nextIndex = index + 1;  
            if (theArray[index].compareTo(theArray[nextIndex]) > 0) {  
                // exchange items  
                Comparable temp = theArray[index];  
                theArray[index] = theArray[nextIndex];  
                theArray[nextIndex] = temp;  
                sorted = false; // signal exchange  
            } // end if  
        } // end for  
        // Assertion: theArray[0..n-pass-1] < theArray[n-pass]  
    } // end for  
} // end bubbleSort
```

# Bubble Sort Analysis

- Analysis
  - Worst case:  $O(n^2)$
  - Best case:  $O(n)$

# Insertion Sort

- Insertion sort
  - Strategy
    - Partition the array into two regions: sorted and unsorted
    - Take each item from the unsorted region and insert it into its correct order in the sorted region

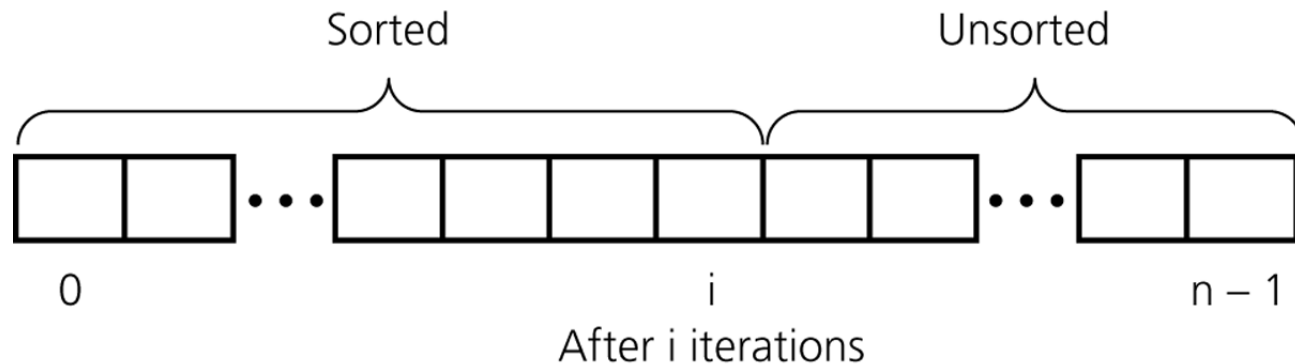
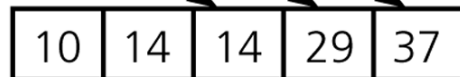
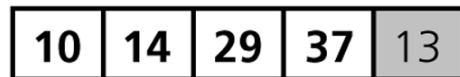
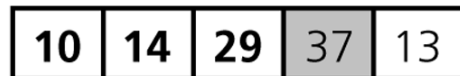
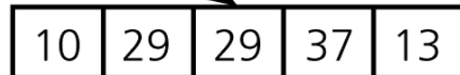
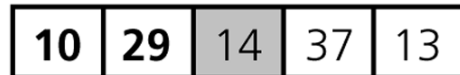
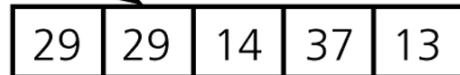
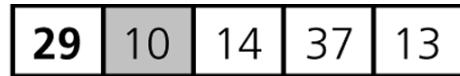


Figure 10-6

An insertion sort partitions the array into two regions

# Insertion Sort

Initial array:



Sorted array:



Copy 10

Shift 29

Insert 10; copy 14

Shift 29

Insert 14; copy 37, insert 37 on top of itself

Copy 13

Shift 37, 29, 14

Insert 13

**Figure 10-7**

An insertion sort of an array of five integers.

# Insertion Sort Code

```
public static void insertionSort(Comparable[] theArray,
    int n) {
    // -----
    // Sorts the items in an array into ascending order.
    // Precondition: theArray is an array of n items.
    // Postcondition: theArray is sorted into ascending
    // order.
    // -----
    // unsorted = first index of the unsorted region,
    // loc = index of insertion in the sorted region,
    // nextItem = next item in the unsorted region
    // initially, sorted region is theArray[0],
    // unsorted region is theArray[1..n-1];
    for (int unsorted = 1; unsorted < n; ++unsorted) {
        // Invariant: theArray[0..unsorted-1] is sorted
        // find the right position (loc) in
        // theArray[0..unsorted] for theArray[unsorted],
        // which is the first item in the unsorted
        // region; shift, if necessary, to make room
```

# Insertion Sort Code

```
Comparable nextItem = theArray[unsorted];
    int loc = unsorted;
    while ((loc > 0) &&
        (theArray[loc-1].compareTo(nextItem) > 0)) {
        // shift theArray[loc-1] to the right
        theArray[loc] = theArray[loc-1];
        loc--;
    } // end while
    // insert nextItem into sorted region
    theArray[loc] = nextItem;
} // end for
} // end insertionSort
```



# Insertion Sort

- Analysis
  - Worst case:  $O(n^2)$
  - For small arrays
    - Insertion sort is appropriate due to its simplicity
  - For large arrays
    - Insertion sort is prohibitively inefficient

# Mergesort

- Important divide-and-conquer sorting algorithms
  - Mergesort
  - Quicksort
- Mergesort
  - A recursive sorting algorithm
  - Gives the same performance, regardless of the initial order of the array items
  - Strategy
    - Divide an array into halves
    - Sort each half
    - Merge the sorted halves into one sorted array

# Mergesort

theArray: 

8	1	4	3	2
---	---	---	---	---

Divide the array in half

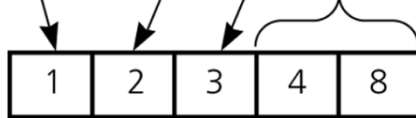


Sort the halves

Merge the halves:

- a.  $1 < 2$ , so move 1 from left half to `tempArray`
- b.  $4 > 2$ , so move 2 from right half to `tempArray`
- c.  $4 > 3$ , so move 3 from right half to `tempArray`
- d. Right half is finished, so move rest of left half to `tempArray`

Temporary array  
tempArray:



Copy temporary array back into original array

theArray:

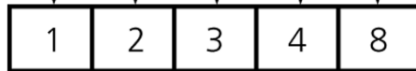


Figure 10-8

A mergesort with an auxiliary temporary array

# Mergesort

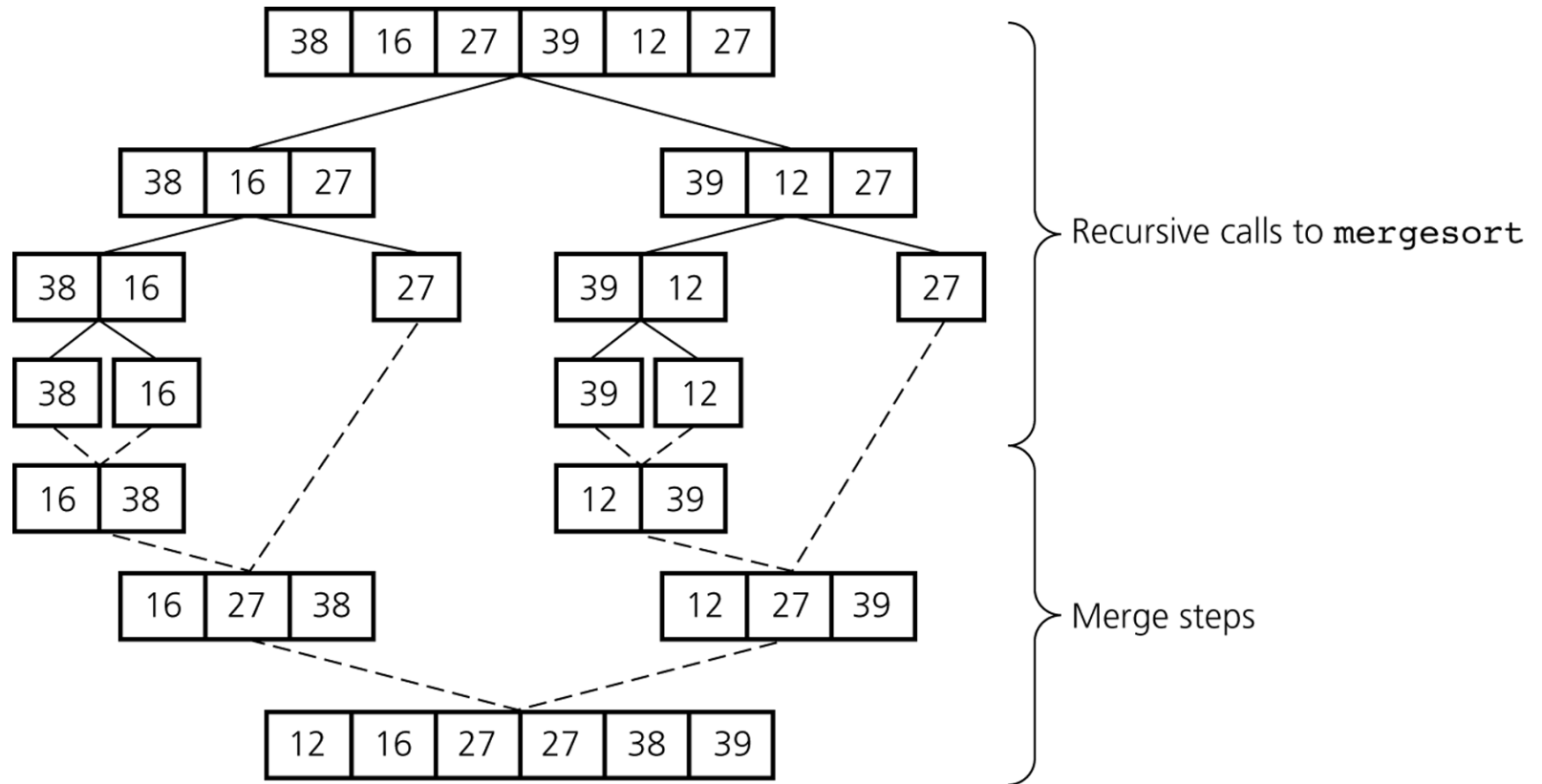


Figure 10-9

A mergesort of an array of six integers

# Mergesort

- Analysis
  - Worst case:  $O(n * \log_2 n)$
  - Average case:  $O(n * \log_2 n)$
  - Advantage
    - It is an extremely efficient algorithm with respect to time
  - Drawback
    - It requires a second array as large as the original array

# Mergesort

Click [here](#) to open the mergesort program

# Quicksort

- Quicksort
  - A divide-and-conquer algorithm
  - Strategy
    - Partition an array into items that are less than the pivot and those that are greater than or equal to the pivot
    - Sort the left section
    - Sort the right section

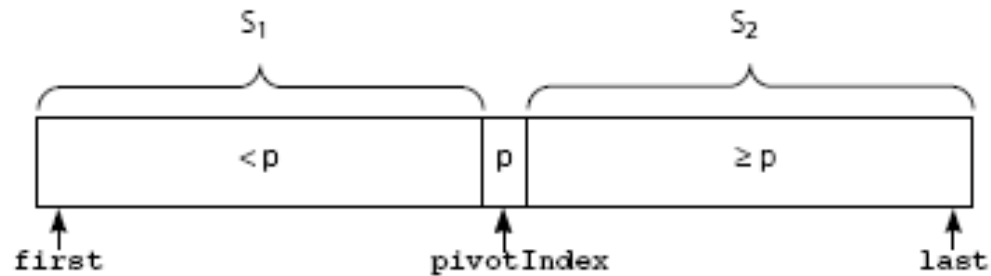


Figure 10-12

A partition about a pivot

# Quicksort

- Using an invariant to develop a partition algorithm

- Invariant for the partition algorithm

The items in region  $S_1$  are all less than the pivot, and those in  $S_2$  are all greater than or equal to the pivot

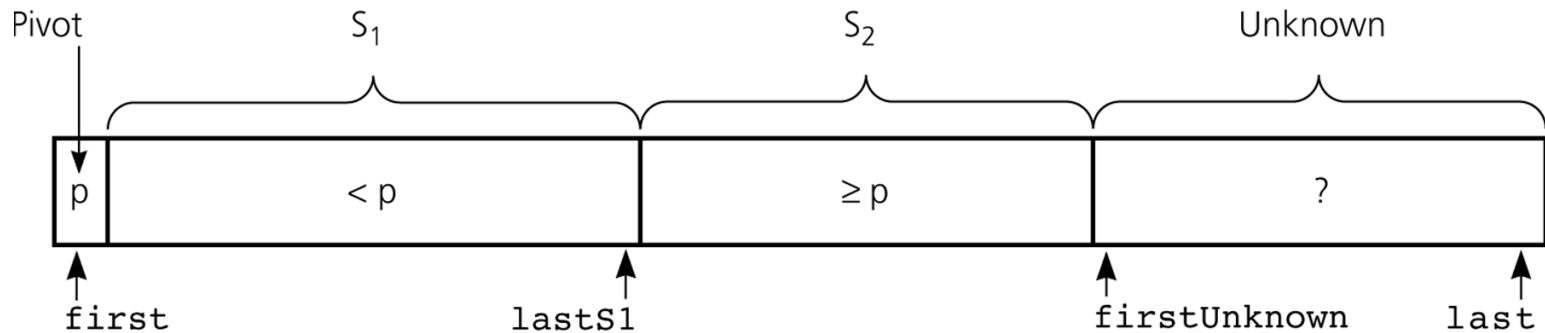


Figure 10-14

Invariant for the partition algorithm



# Quicksort

- Analysis

- Worst case

- quicksort is  $O(n^2)$  when the array is already sorted and the smallest item is chosen as the pivot

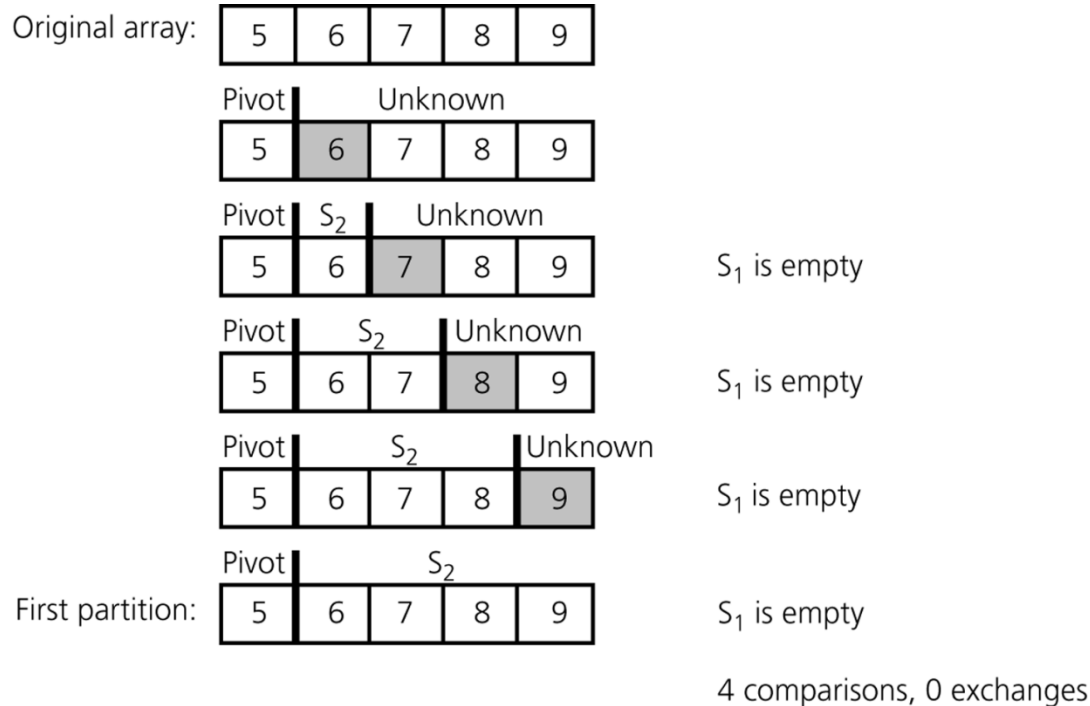


Figure 10-19  
A worst-case partitioning  
with *quicksort*

# Quicksort

- Analysis

- Average case

- quicksort is  $O(n * \log_2 n)$  when  $S_1$  and  $S_2$  contain the same – or nearly the same – number of items arranged at random

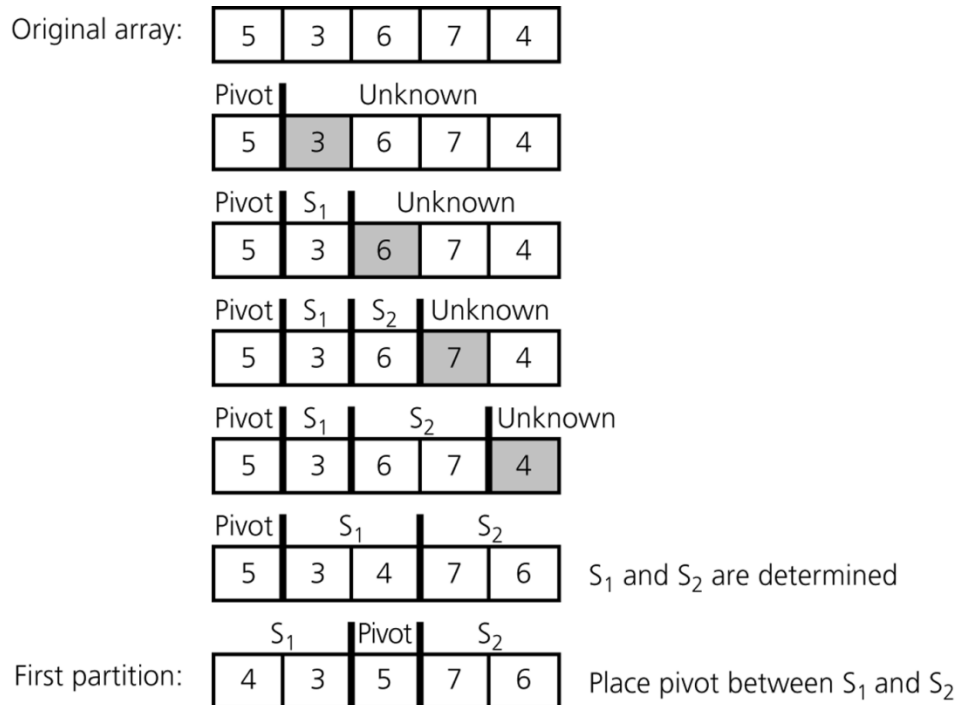


Figure 10-20

A average-case partitioning with *quicksort*

# Quicksort

- Analysis
  - `quicksort` is usually extremely fast in practice
  - Even if the worst case occurs, `quicksort`'s performance is acceptable for moderately large arrays
  - Click [here](#) to open the quicksort program

# Radix Sort

- Radix sort
  - Treats each data element as a character string
  - Strategy
    - Repeatedly organize the data into groups according to the  $i^{\text{th}}$  character in each element
- Analysis
  - Radix sort is  $O(n)$

# Radix Sort

0123, 2154, 0222, 0004, 0283, 1560, 1061, 2150	Original integers
(156 <b>0</b> , 215 <b>0</b> ) (106 <b>1</b> ) (022 <b>2</b> ) (012 <b>3</b> , 028 <b>3</b> ) (215 <b>4</b> , 000 <b>4</b> )	Grouped by fourth digit
1560, 2150, 1061, 0222, 0123, 0283, 2154, 0004	Combined
(000 <b>4</b> ) (022 <b>2</b> , 012 <b>3</b> ) (215 <b>0</b> , 215 <b>4</b> ) (156 <b>0</b> , 106 <b>1</b> ) (028 <b>3</b> )	Grouped by third digit
0004, 0222, 0123, 2150, 2154, 1560, 1061, 0283	Combined
(000 <b>4</b> , 106 <b>1</b> ) (012 <b>3</b> , 215 <b>0</b> , 215 <b>4</b> ) (022 <b>2</b> , 028 <b>3</b> ) (156 <b>0</b> )	Grouped by second digit
0004, 1061, 0123, 2150, 2154, 0222, 0283, 1560	Combined
(000 <b>4</b> , 012 <b>3</b> , 022 <b>2</b> , 028 <b>3</b> ) (106 <b>1</b> , 156 <b>0</b> ) (215 <b>0</b> , 215 <b>4</b> )	Grouped by first digit
0004, 0123, 0222, 0283, 1061, 1560, 2150, 2154	Combined (sorted)

**Figure 10-21**

A radix sort of eight integers

# A Comparison of Sorting Algorithms

	<u>Worst case</u>	<u>Average case</u>
Selection sort	$n^2$	$n^2$
Bubble sort	$n^2$	$n^2$
Insertion sort	$n^2$	$n^2$
Mergesort	$n * \log n$	$n * \log n$
Quicksort	$n^2$	$n * \log n$
Radix sort	$n$	$n$
Treesort	$n^2$	$n * \log n$
Heapsort	$n * \log n$	$n * \log n$

Figure 10-22

Approximate growth rates of time required for eight sorting algorithms

# Summary

- Order-of-magnitude analysis and Big O notation measure an algorithm's time requirement as a function of the problem size by using a growth-rate function
- To compare the inherent efficiency of algorithms
  - Examine their growth-rate functions when the problems are large
  - Consider only significant differences in growth-rate functions

# Summary

- Worst-case and average-case analyses
  - Worst-case analysis considers the maximum amount of work an algorithm requires on a problem of a given size
  - Average-case analysis considers the expected amount of work an algorithm requires on a problem of a given size
- Order-of-magnitude analysis can be used to choose an implementation for an abstract data type
- Selection sort, bubble sort, and insertion sort are all  $O(n^2)$  algorithms
- Quicksort and mergesort are two very efficient sorting algorithms



# Acknowledgement

All of the material for the slides were adapted from

Data Abstraction and Problem Solving with Java, 2/E

**Frank Carrano**, *University of Rhode Island*

**Janet Prichard**, *Bryant College*