

# Bruteforce

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# Brute Force Algorithms

- String Matching
- Closest Pair Problem
- Exhaustive search: TSP, Knapsack Problem

# Sequential Search

**ALGORITHM** *SequentialSearch2*( $A[0..n]$ ,  $K$ )

//Implements sequential search with a search key as a sentinel

//Input: An array  $A$  of  $n$  elements and a search key  $K$

//Output: The index of the first element in  $A[0..n - 1]$  whose value is  
// equal to  $K$  or  $-1$  if no such element is found

$A[n] \leftarrow K$

$i \leftarrow 0$

**while**  $A[i] \neq K$  **do**

$i \leftarrow i + 1$

**if**  $i < n$  **return**  $i$

**else return**  $-1$



# Sequential Search

- Sequential search provides an excellent illustration of the brute-force approach, with its characteristic strength (simplicity) and weakness (inferior efficiency).
- If elements are sorted, then Binary search can be employed



# String Matching

- Pattern: a string of  $m$  characters to search for
- Text: a (longer) string of  $n$  characters to search in
- problem: find a substring in the text that matches the pattern



# Brute-force algorithm

- Step 1: Align pattern at beginning of text
- Step 2: Moving from left to right, compare each character of pattern to the corresponding character in text until all characters are found to match (successful search); or a mismatch is detected
- Step 3 While pattern is not found and the text is not yet exhausted, realign pattern one position to the right and repeat Step 2



**ALGORITHM** *BruteForceStringMatch*( $T[0..n - 1]$ ,  $P[0..m - 1]$ )

//Implements brute-force string matching

//Input: An array  $T[0..n - 1]$  of  $n$  characters representing a text and

// an array  $P[0..m - 1]$  of  $m$  characters representing a pattern

//Output: The index of the first character in the text that starts a

// matching substring or  $-1$  if the search is unsuccessful

**for**  $i \leftarrow 0$  **to**  $n - m$  **do**

$j \leftarrow 0$

**while**  $j < m$  **and**  $P[j] = T[i + j]$  **do**

$j \leftarrow j + 1$

**if**  $j = m$  **return**  $i$

**return**  $-1$



# Match

|       |     |              |     |              |     |              |     |           |             |
|-------|-----|--------------|-----|--------------|-----|--------------|-----|-----------|-------------|
| $t_0$ | ... | $t_i$        | ... | $t_{i+j}$    | ... | $t_{i+m-1}$  | ... | $t_{n-1}$ | text $T$    |
|       |     | $\Downarrow$ |     | $\Downarrow$ |     | $\Downarrow$ |     |           |             |
|       |     | $p_0$        | ... | $p_j$        | ... | $p_{m-1}$    |     |           | pattern $P$ |



# Complexity

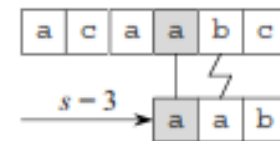
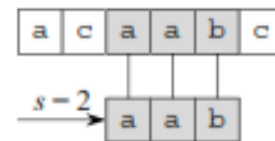
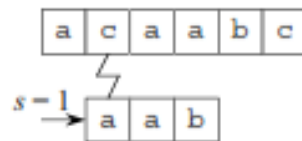
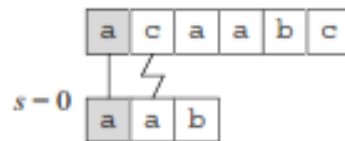
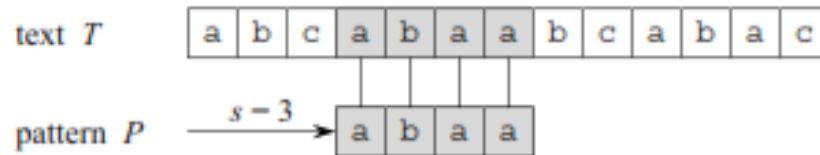
- Note that for this example, the algorithm shifts the pattern almost always after a single character comparison.
- The worst case is much worse: the algorithm may have to make all  $m$  comparisons before shifting the pattern, and this can happen for each of the  $n - m + 1$  tries.
- In the worst case, the algorithm makes  $m(n - m + 1)$  character comparisons, which puts it in the  $O(nm)$  class



|          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |
|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|----------|
| N        | O        | B        | O        | D        | Y        | _        | N        | O        | T        | I        | C        | E        | D        | _        | H        | I        | M        |
| <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |          |          |          |          |          |          |
|          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |          |          |          |          |          |
|          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |          |          |          |          |
|          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |          |          |          |
|          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |          |          |
|          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |          |
|          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |          |
|          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |          |
|          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |          |
|          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |          |
|          |          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |          |
|          |          |          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |          |
|          |          |          |          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |          |
|          |          |          |          |          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |          |
|          |          |          |          |          |          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |          |
|          |          |          |          |          |          |          |          |          |          |          |          |          |          |          | <b>N</b> | <b>O</b> | <b>T</b> |

**FIGURE 3.3** Example of brute-force string matching. (The pattern's characters that are compared with their text counterparts are in bold type.)

# Example



Determine the number of character comparisons made by the brute-force algorithm in searching for the pattern SEARCH in the text

`SORTING_ALGORITHM_CAN_USE_BRUTE_FORCE_METHOD`

Assume that the length of the text—it is 43 characters long—is known before the search starts.



# Example

How many comparisons (both successful and unsuccessful) will be made by the brute-force algorithm in searching for each of the following patterns in the binary text of one million zeros?

**a.** 01001    **b.** 00010    **c.** 01011



# Matching

one position to the right:

0 0 0 0 0 0 0 0

0 1 0 0 1

0 1 0 0 1

etc.

0 0 0 0 0 0 0 0

0 1 0 0 1

The total number of character comparisons will be  $c = 2 * 9,99,996$ .

# Example

Give an example of a text of length  $n$  and a pattern of length  $m$  that constitutes a worst-case input for the brute-force string-matching algorithm. Exactly how many character comparisons will be made for such input?



# Comparisons

The text composed of  $n$  zeros and the pattern  $\underbrace{0 \dots 0}_{m-1}1$  is an example of the worst-case input. The algorithm will make  $m(n - m + 1)$  character comparisons on such input.





# Boyer Moore Horspool Algorithm

- Example:

Pattern 'tooth'  
Text 'trusthardtoothbrushes'

# Match Table Pre Processing

- Construct Bad Match Table  
Value = length – index – 1 (Every other letter = length)

T O O T H      Length = 5  
0 1 2 3 4

| Letter | T | O | H | * |
|--------|---|---|---|---|
| Value  |   |   |   |   |

# Pre-Processing

$T = 5 - 0 - 1$

$O = 5 - 1 - 1$

| Letter | T | O | H | * |
|--------|---|---|---|---|
| Value  | 1 | 2 | 5 | 5 |

$O = 5 - 2 - 1$

$T = 5 - 3 - 1$

$H = 5$       Last letter = length if not already defined, else, leave

# Matching

T R U S T H A R D T O O T H B R U S H E S

T O O T H

T R U S T H A R D T O O T H B R U S H E S

T O O T H



# Contd...

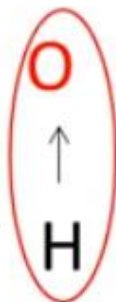
T R U S T H A R D T O O T H B R U S H E S  
↑  
T O O T H

T R U S T H A R D T O O T H B R U S H E S  
↑ ↑ ↑  
T O O T H

# Contd...

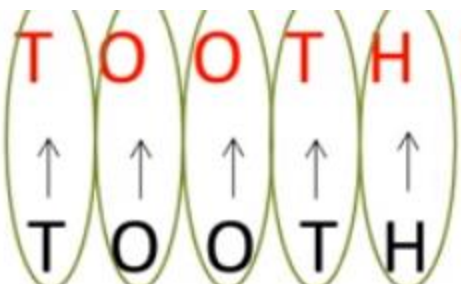
T R U S T H A R D T O O T H B R U S H E S

T O O T H



T R U S T H A R D T O O T H B R U S H E S

T O O T H



# Complexity

Worst case same as naïve example:

- $1^n$  input text (length  $n$ )
- $0111\dots 1$  pattern (length  $m$ )

Worst Case  $O(nm)$

Best case

- $1^n$  input text (length  $n$ )
- $0^m$  pattern (length  $m$ )

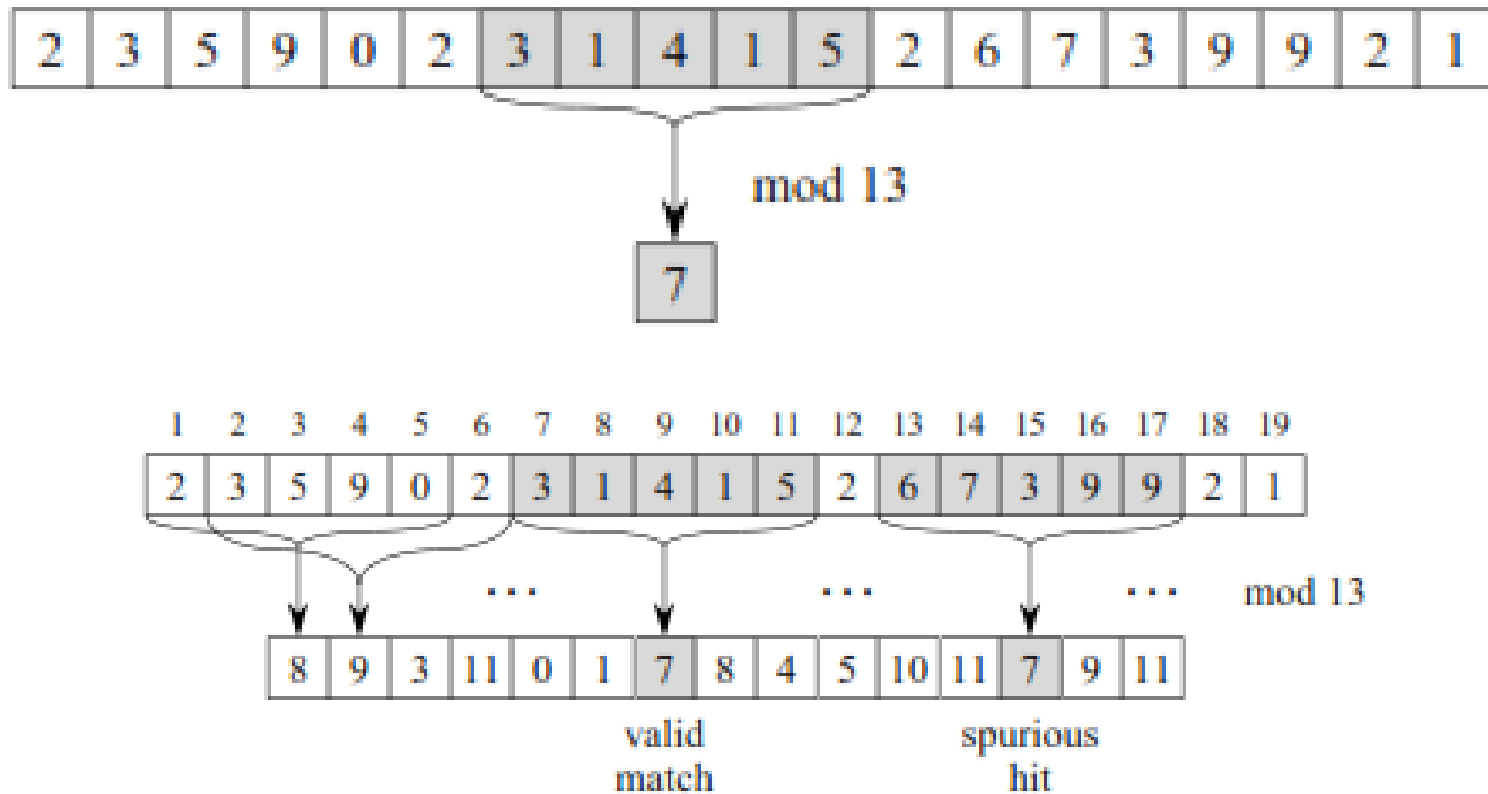
# Rabin Karp Algorithm

| Algorithm  | Preprocessing time | Matching time     |
|------------|--------------------|-------------------|
| Naive      | 0                  | $O((n - m + 1)m)$ |
| Rabin-Karp | $\Theta(m)$        | $O((n - m + 1)m)$ |

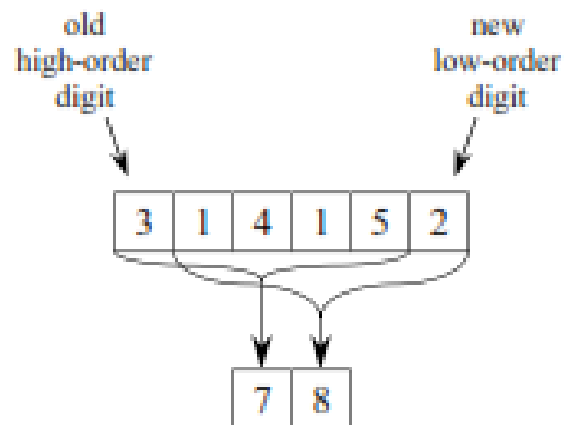
Given a pattern  $P[1..m]$ , let  $p$  denote its corresponding decimal value. In a similar manner, given a text  $T[1..n]$ , let  $t_s$  denote the decimal value of the length- $m$  substring  $T[s + 1..s + m]$ , for  $s = 0, 1, \dots, n - m$ . Certainly,  $t_s = p$  if and only if  $T[s + 1..s + m] = P[1..m]$ ; thus,  $s$  is a valid shift if and only if  $t_s = p$ . If we could compute  $p$  in time  $\Theta(m)$  and all the  $t_s$  values in a total of  $\Theta(n - m + 1)$  time,<sup>1</sup>



# Preprocessing



# Preprocessing



$$\begin{aligned}
 14152 &\equiv (31415 - 3 \cdot 10000) \cdot 10 + 2 \pmod{13} \\
 &\equiv (7 - 3 \cdot 3) \cdot 10 + 2 \pmod{13} \\
 &\equiv 8 \pmod{13}
 \end{aligned}$$

Diagram illustrating the preprocessing of a number 14152. The number is shown in a sequence of boxes: 1, 4, 1, 5, 2. An arrow labeled "old high-order digit" points to the first box (1). An arrow labeled "shift" points to the second box (4). An arrow labeled "new low-order digit" points to the last box (2).

$$\begin{aligned}
 t_{s+1} &= 10(31415 - 10000 \cdot 3) + 2 \\
 &= 14152 .
 \end{aligned}$$

# Example

Working modulo  $q = 11$ , how many spurious hits does the Rabin-Karp matcher encounter in the text  $T = 3141592653589793$  when looking for the pattern  $P = 26$ ?



# Solution

- Three spurious hits,  $15 \equiv 59 \equiv 92 \equiv 26 \equiv 4 \pmod{11}$

# Closest Pair Problem

- Find the two closest points in a set of  $n$  points (in the two-dimensional Cartesian plane).
- Brute-force algorithm
- Compute the distance between every pair of distinct points and return the indexes of the points for which the distance is the smallest.

# Closest Pair Problem

**ALGORITHM** *BruteForceClosestPoints(P)*

//Finds two closest points in the plane by brute force

//Input: A list  $P$  of  $n$  ( $n \geq 2$ ) points  $P_1 = (x_1, y_1), \dots, P_n = (x_n, y_n)$

//Output: Indices  $index1$  and  $index2$  of the closest pair of points

$dmin \leftarrow \infty$

**for**  $i \leftarrow 1$  **to**  $n - 1$  **do**

**for**  $j \leftarrow i + 1$  **to**  $n$  **do**

$d \leftarrow \text{sqrt}((x_i - x_j)^2 + (y_i - y_j)^2)$  //sqrt is the square root function

**if**  $d < dmin$

$dmin \leftarrow d; index1 \leftarrow i; index2 \leftarrow j$

**return**  $index1, index2$



# Strictly increasing function

$$\begin{aligned} C(n) &= \sum_{i=1}^{n-1} \sum_{j=i+1}^n 2 = 2 \sum_{i=1}^{n-1} (n-i) \\ &= 2[(n-1) + (n-2) + \dots + 1] = (n-1)n \in \Theta(n^2). \end{aligned}$$

# Bruteforce algorithms

- Strengths
  - wide applicability
  - simplicity
  - yields reasonable algorithms for some important problems  
(e.g., matrix multiplication, sorting, searching, string matching)
- Weaknesses
  - rarely yields efficient algorithms
  - some brute-force algorithms are unacceptably slow
  - not as constructive as some other design techniques





# Exhaustive Search

- A brute force solution to a problem involving search for an element with a special property, usually among combinatorial objects such as permutations, combinations, or subsets of a set.

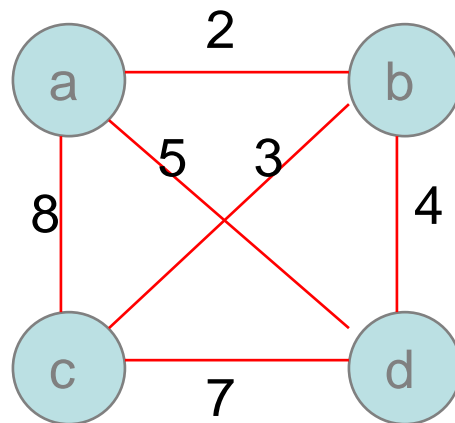
# Method

## Method:

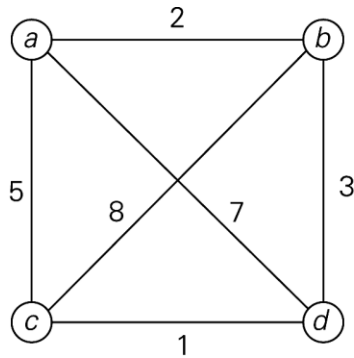
- generate a list of all potential solutions to the problem in a systematic
- evaluate potential solutions one by one, disqualifying infeasible ones and, for an optimization problem, keeping track of the best one found so far
- when search ends, announce the solution(s) found

# TSP

- Given  $n$  cities with known distances between each pair, find the shortest tour that passes through all the cities exactly once before returning to the starting city
- Alternatively: Find shortest *Hamiltonian circuit* in a weighted connected graph



# TSP



| <u>Tour</u>   | <u>Length</u>            |         |
|---|--------------------------|---------|
| $a \rightarrow b \rightarrow c \rightarrow d \rightarrow a$ | $l = 2 + 8 + 1 + 7 = 18$ |         |
| $a \rightarrow b \rightarrow d \rightarrow c \rightarrow a$ | $l = 2 + 3 + 1 + 5 = 11$ | optimal |
| $a \rightarrow c \rightarrow b \rightarrow d \rightarrow a$ | $l = 5 + 8 + 3 + 7 = 23$ |         |
| $a \rightarrow c \rightarrow d \rightarrow b \rightarrow a$ | $l = 5 + 1 + 3 + 2 = 11$ | optimal |
| $a \rightarrow d \rightarrow b \rightarrow c \rightarrow a$ | $l = 7 + 3 + 8 + 5 = 23$ |         |
| $a \rightarrow d \rightarrow c \rightarrow b \rightarrow a$ | $l = 7 + 1 + 8 + 2 = 18$ |         |

**FIGURE 3.7** Solution to a small instance of the traveling salesman problem by exhaustive search

# Knapsack

Given  $n$  items:

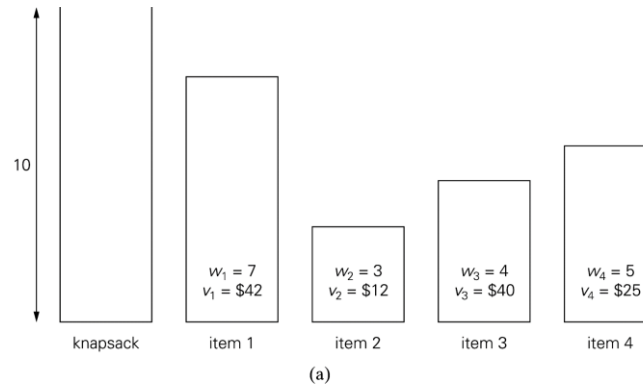
- weights:  $w_1 \ w_2 \ \dots \ w_n$
- values:  $v_1 \ v_2 \ \dots \ v_n$
- a knapsack of capacity  $W$

Find most valuable subset of the items that fit into the knapsack

Example: Knapsack capacity  $W=16$

| <u>item</u> | <u>weight</u> | <u>value</u> |
|-------------|---------------|--------------|
| 1           | 2             | \$20         |
| 2           | 5             | \$30         |
| 3           | 10            | \$50         |
| 4           | 5             | \$10         |

# Knapsack



| Subset        | Total weight | Total value  |
|---------------|--------------|--------------|
| $\emptyset$   | 0            | \$ 0         |
| {1}           | 7            | \$42         |
| {2}           | 3            | \$12         |
| {3}           | 4            | \$40         |
| {4}           | 5            | \$25         |
| {1, 2}        | 10           | \$36         |
| {1, 3}        | 11           | not feasible |
| {1, 4}        | 12           | not feasible |
| {2, 3}        | 7            | \$52         |
| {2, 4}        | 8            | \$37         |
| <b>{3, 4}</b> | <b>9</b>     | <b>\$65</b>  |
| {1, 2, 3}     | 14           | not feasible |
| {1, 2, 4}     | 15           | not feasible |
| {1, 3, 4}     | 16           | not feasible |
| {2, 3, 4}     | 12           | not feasible |
| {1, 2, 3, 4}  | 19           | not feasible |

(b)

**FIGURE 3.8** (a) Instance of the knapsack problem. (b) Its solution by exhaustive search. (The information about the optimal selection is in bold.)

# Assignment Problem

|  |                              |                             |      |
|--|------------------------------|-----------------------------|------|
| $C = \begin{bmatrix} 9 & 2 & 7 & 8 \\ 6 & 4 & 3 & 7 \\ 5 & 8 & 1 & 8 \\ 7 & 6 & 9 & 4 \end{bmatrix}$ | $\langle 1, 2, 3, 4 \rangle$ | cost = $9 + 4 + 1 + 4 = 18$ | etc. |
|  | $\langle 1, 2, 4, 3 \rangle$ | cost = $9 + 4 + 8 + 9 = 30$ |      |
|  | $\langle 1, 3, 2, 4 \rangle$ | cost = $9 + 3 + 8 + 4 = 24$ |      |
|  | $\langle 1, 3, 4, 2 \rangle$ | cost = $9 + 3 + 8 + 6 = 26$ |      |
|  | $\langle 1, 4, 2, 3 \rangle$ | cost = $9 + 7 + 8 + 9 = 33$ |      |
|  | $\langle 1, 4, 3, 2 \rangle$ | cost = $9 + 7 + 1 + 6 = 23$ |      |

**FIGURE 3.9** First few iterations of solving a small instance of the assignment problem by exhaustive search



# Assignment problem

There are  $n$  people who need to be assigned to  $n$  jobs, one person per job. The cost of assigning person  $i$  to job  $j$  is  $C[i,j]$ . Find an assignment that minimizes the total cost.

# Exhaustive Search

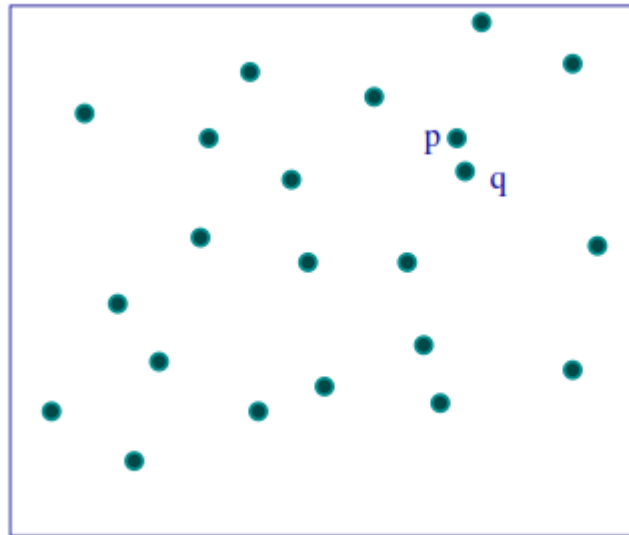
- Exhaustive-search algorithms run in a realistic amount of time only on very small instances
- In some cases, there are much better alternatives!
  - Euler circuits
  - shortest paths
  - minimum spanning tree
  - assignment problem

# Divide and Conquer / Dynamic Programming

Presentation by  
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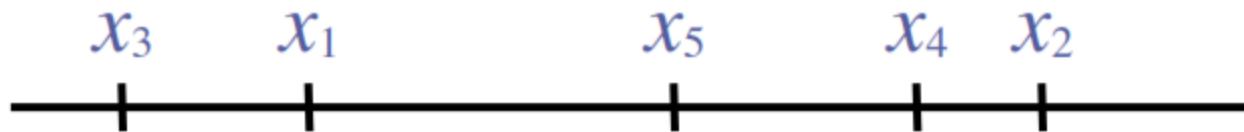
# Closest Pair Problem



- Complexity =  $n^2$
- For example, an air-traffic controller might be interested in two closest planes as the most probable collision candidates.
- A regional postal service manager - closest pair problem to find candidate post-office locations to be closed.



# 1D solution



# Algorithm

- If  $n > 3$ , we can divide the points into two subsets  $P_l$  and  $P_r$  of  $n/2$  and  $n/2$  points, respectively, by drawing a vertical line through the median  $m$  of their  $x$  coordinates so that  $n/2$  points lie to the left of or on the line itself, and  $n/2$  points lie to the right of or on the line. Then we can solve the closest-pair problem



# Algorithm

The algorithm:

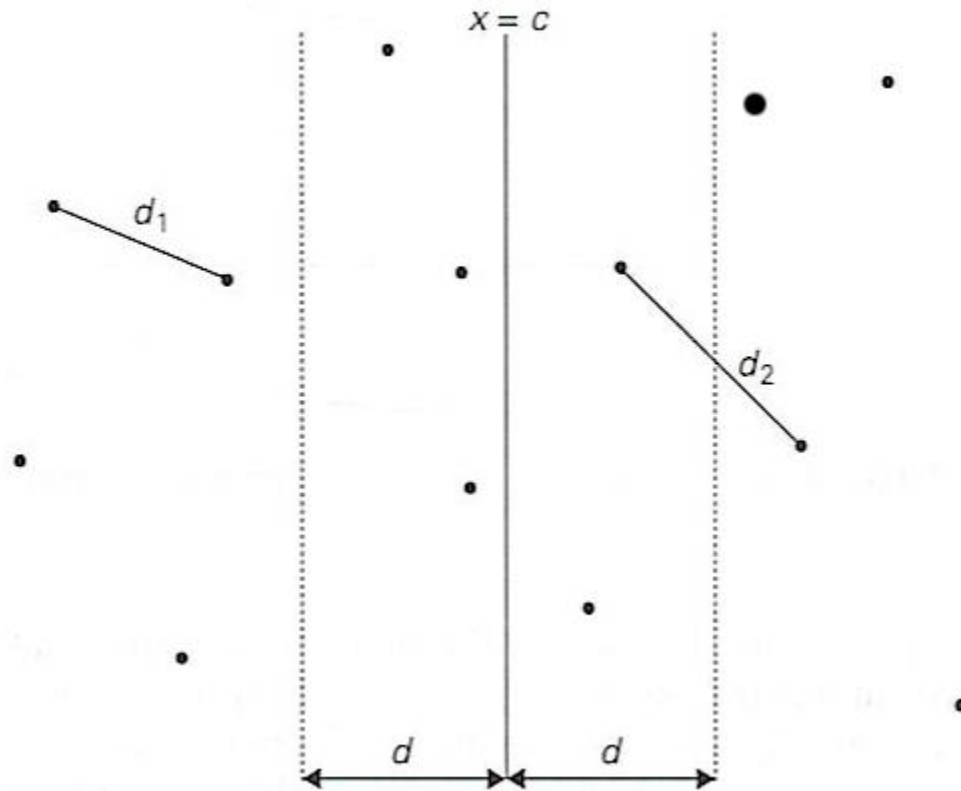
- Input: A set  $S$  of  $n$  planar points.
- Output: The distance between two closest points.



# Steps

Step 1 Divide the points given into two subsets  $S_1$  and  $S_2$  by a vertical line  $x = c$  so that half the points lie to the left or on the line and half the points lie to the right or on the line.

# Contd...



# Step 2

Step 2: Find recursively the closest pairs for the left and right subsets.

# Step 3

- Step 3: Set  $d = \min\{d_1, d_2\}$

We can limit our attention to the points in the symmetric vertical strip of width  $2d$  as possible closest pair. Let  $C_1$  and  $C_2$  be the subsets of points in the left subset  $S_1$  and of the right subset  $S_2$ , respectively, that lie in this vertical strip. The points in  $C_1$  and  $C_2$  are stored in increasing order of their  $y$  coordinates, which is maintained by merging during the execution of the next step.

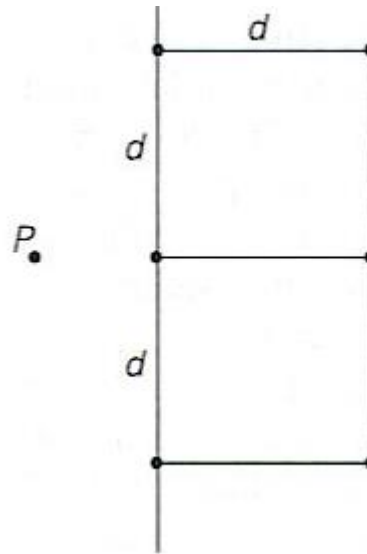


# Step 4

- Step 4:

For every point  $P(x,y)$  in  $C_1$ , we inspect points in  $C_2$  that may be closer to  $P$  than  $d$ . There can be no more than 6 such points (because  $d \leq d_2$ )!

# Worst Case Scenario



## Contd...

Step 1: Sort points in  $S$  according to their  $y$ -values.

Step 2: If  $S$  contains only one point, return infinity as its distance.

Step 3: Find a median line  $L$  perpendicular to the  $X$ -axis to divide  $S$  into  $S_L$  and  $S_R$ , with equal sizes.

## Contd...

- Step 4: Recursively apply Steps 2 and 3 to solve the closest pair problems of  $S_L$  and  $S_R$ . Let  $d_L(d_R)$  denote the distance between the closest pair in  $S_L$  ( $S_R$ ). Let  $d = \min(d_L, d_R)$ .



## Contd...

- Step 5: For a point  $P$  in the half-slab bounded by  $L-d$  and  $L$ , let its  $y$ -value be denoted as  $y_P$ . For each such  $P$ , find all points in the half-slab bounded by  $L$  and  $L+d$  whose  $y$ -value fall within  $y_P+d$  and  $y_P-d$ . If the distance  $d'$  between  $P$  and a point in the other half-slab is less than  $d$ , let  $d=d'$ . The final value of  $d$  is the answer.

**ALGORITHM** *EfficientClosestPair*( $P, Q$ )

//Solves the closest-pair problem by divide-and-conquer

//Input: An array  $P$  of  $n \geq 2$  points in the Cartesian plane sorted in

//       nondecreasing order of their  $x$  coordinates and an array  $Q$  of the

//       same points sorted in nondecreasing order of the  $y$  coordinates

//Output: Euclidean distance between the closest pair of points

**if**  $n \leq 3$

    return the minimal distance found by the brute-force algorithm

**else**

    copy the first  $\lceil n/2 \rceil$  points of  $P$  to array  $P_l$

    copy the same  $\lceil n/2 \rceil$  points from  $Q$  to array  $Q_l$

    copy the remaining  $\lfloor n/2 \rfloor$  points of  $P$  to array  $P_r$

    copy the same  $\lfloor n/2 \rfloor$  points from  $Q$  to array  $Q_r$

$d_l \leftarrow \text{EfficientClosestPair}(P_l, Q_l)$

$d_r \leftarrow \text{EfficientClosestPair}(P_r, Q_r)$

$d \leftarrow \min\{d_l, d_r\}$

$m \leftarrow P[\lceil n/2 \rceil - 1].x$

    copy all the points of  $Q$  for which  $|x - m| < d$  into array  $S[0..num - 1]$

$dminsq \leftarrow d^2$

**for**  $i \leftarrow 0$  **to**  $num - 2$  **do**

$k \leftarrow i + 1$

**while**  $k \leq num - 1$  **and**  $(S[k].y - S[i].y)^2 < dminsq$

$dminsq \leftarrow \min((S[k].x - S[i].x)^2 + (S[k].y - S[i].y)^2, dminsq)$

$k \leftarrow k + 1$

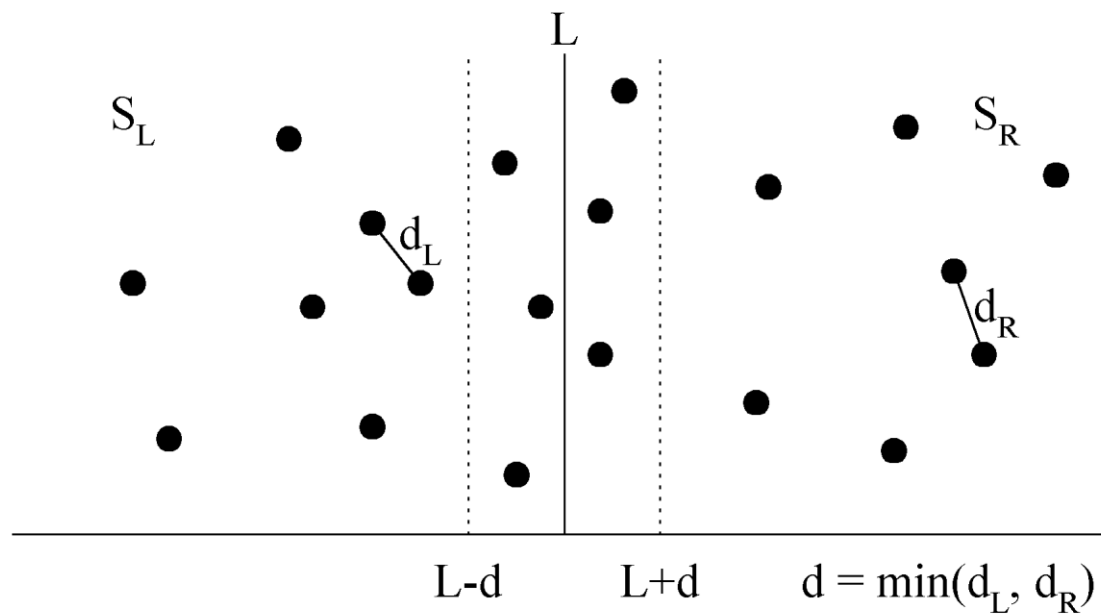
**return**  $\text{sqrt}(dminsq)$

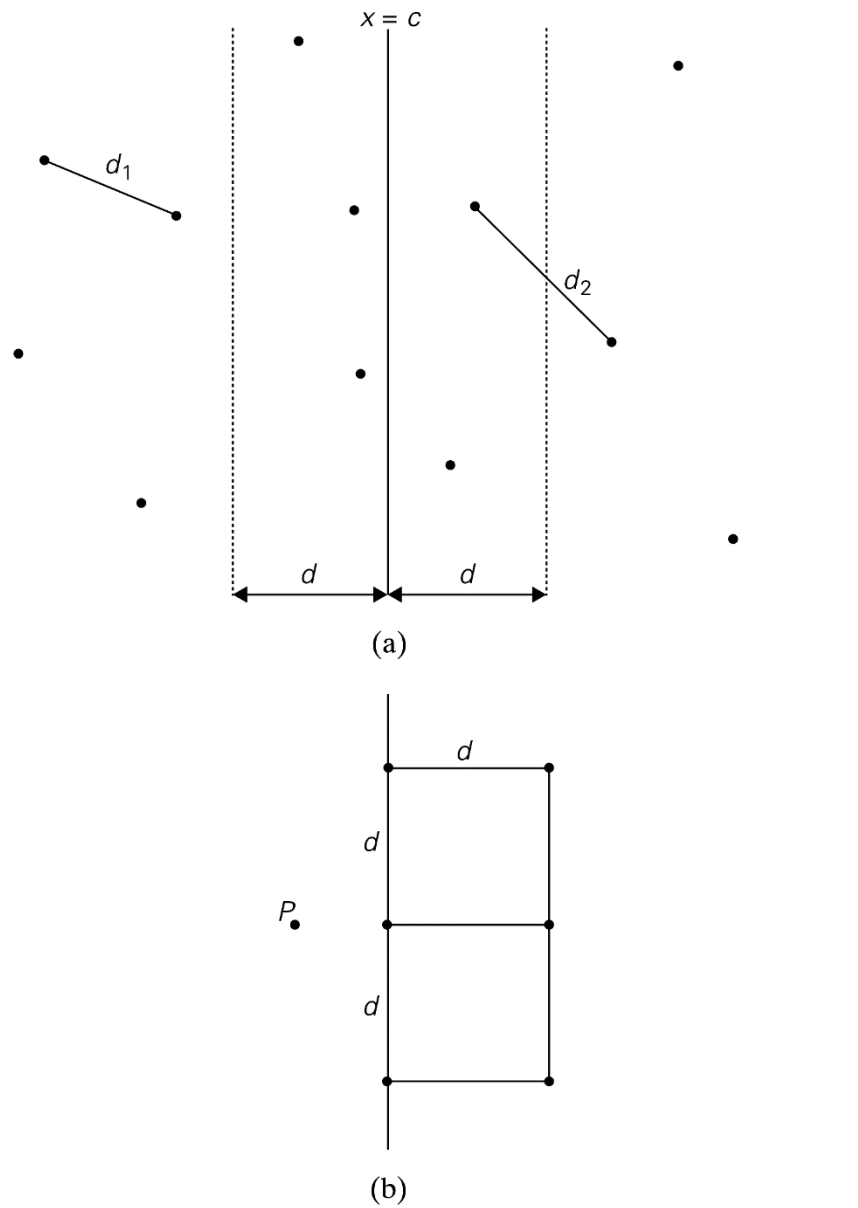


# Contd...

$$T(n) = \begin{cases} 2T(n/2) + O(n) + O(n) & , n > 1 \\ 1 & , n = 1 \end{cases}$$







**FIGURE 4.7** (a) Idea of the divide-and-conquer algorithm for the closest-pair problem.  
 (b) The six points that may need to be examined for point  $P$ .

# Analysis

Running time of the algorithm is described by

$$T(n) = 2T(n/2) + M(n), \text{ where}$$

$$M(n) \in O(n)$$

By the Master Theorem (with  $a = 2$ ,  
 $b = 2$ ,  $d = 1$ )

$$T(n) \in O(n \log n)$$



# Exercise

- 1. a.** For the one-dimensional version of the closest-pair problem, i.e., for the problem of finding two closest numbers among a given set of  $n$  real numbers, design an algorithm that is directly based on the divide-and-conquer technique and determine its efficiency class.
- b.** Is it a good algorithm for this problem?

# Solution

- $O(n \log n)$
- Without employing Divide and Conquer, we can sort it and do it.  
 $O(n \log n)$



# Exercise

2. Prove that the divide-and-conquer algorithm for the closest-pair problem examines, for every point  $p$  in the vertical strip (see Figures 5.7a and 5.7b), no more than seven other points that can be closer to  $p$  than  $d_{\min}$ , the minimum distance between two points encountered by the algorithm up to that point.