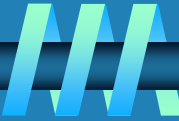


# Topic 2

## Brute Force

# Brute Force



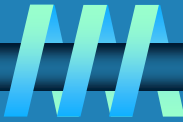
**A straightforward approach, usually based directly on the problem's statement and definitions of the concepts involved**

## **Examples:**

- 1. Computing  $a^n$  ( $a > 0$ ,  $n$  a nonnegative integer)**
- 2. Computing  $n!$**
- 3. Multiplying two matrices**
- 4. Searching for a key of a given value in a list**




# 1. Brute-Force Sorting Algorithm



## Selection Sort

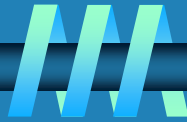
- Scan the array to find its smallest element and swap it with the first element.
- Then, starting with the second element, scan the elements to the right of it to find the smallest among them and swap it with the second element.
- Do this for all elements (until the second last).

$$0 \leq i \leq n-2$$

$$A[0] \leq \dots \leq A[i-1] \mid A[i], \dots, A[\min], \dots, A[n-1]$$


Example: 7 3 2 5

# Analysis of Selection Sort



**ALGORITHM** *SelectionSort*( $A[0..n - 1]$ )

//Sorts a given array by selection sort

//Input: An array  $A[0..n - 1]$  of orderable elements

//Output: Array  $A[0..n - 1]$  sorted in nondecreasing order

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

$min \leftarrow i$

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

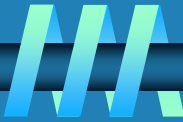
**if**  $A[j] < A[min]$   $min \leftarrow j$

    swap  $A[i]$  and  $A[min]$

As an example, the action of the algorithm on the list 89, 45, 68, 90, 29, 34, 17

	89	45	68	90	29	34	<b>17</b>
17	45	68	90	<b>29</b>	34	89	
17 29	68	90	45	<b>34</b>	89		
17 29 34	90	<b>45</b>	68	89			
17 29 34 45	90	<b>68</b>	89				
17 29 34 45 68	90	<b>89</b>					
17 29 34 45 68 89	90						

# Analysis of Selection Sort



The analysis of selection sort is straightforward. The input size is given by the number of elements  $n$ ; the basic operation is the key comparison  $A[j] < A[\min]$ . The number of times it is executed depends only on the array size and is given by the following sum:

$$\begin{aligned} C(n) &= \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1] = \sum_{i=0}^{n-2} (n-1-i). \\ &= \frac{(n-1)n}{2}. \end{aligned}$$

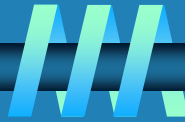
Thus, selection sort is a  $\Theta(n^2)$  algorithm on all inputs.

**Time efficiency:**

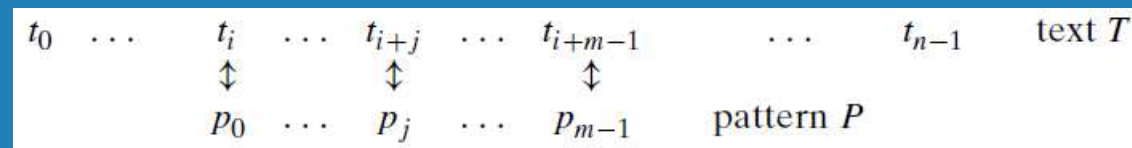
**$\Theta(n^2)$**



## 2. Brute-Force 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



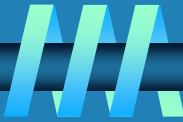
# Examples of Brute-Force String Matching

1. **Pattern:** 001011  
**Text:** 10010101101001100101111010

2. **Pattern:** happy  
**Text:** It is never too late to have a happy life.



# Pseudocode and Efficiency



```
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  $-1$ 
```

**Time efficiency:**  $\Theta(mn)$  comparisons (in the worst case)

Why?

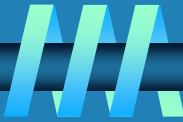
Ans: For each  $(n-m+1)$  tries, we compare  $m$  times at maximum.

→  $\Theta(mn)$  comparisons (in the worst case)





# 3. Brute-Force Polynomial Evaluation



Problem: Find the value of polynomial

$$p(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x^1 + a_0$$

at a point  $x = x_0$

Brute-force algorithm

$p \leftarrow 0.0$

**for**  $i \leftarrow n$  **downto** 0 **do**

$power \leftarrow 1$

**for**  $j \leftarrow 1$  **to**  $i$  **do**      //compute  $x^i$

$power \leftarrow power * x$

$p \leftarrow p + a[i] * power$

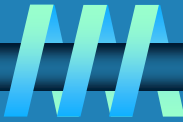
**return**  $p$

Efficiency:

$\sum_{0 \leq i \leq n} (i+1) = \Theta(n^2)$  multiplications



# Polynomial Evaluation: Improvement



We can do better by evaluating from right to left:

## Better brute-force algorithm

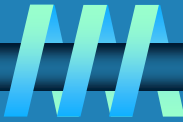
```
p ← a[0]
power ← 1
for i ← 1 to n do
    power ← power * x
    p ← p + a[i] * power
return p
```

Efficiency:  **$\Theta(n)$  multiplications**

Horner's Rule is another linear time method.



# 4. 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 Brute-Force Algorithm (cont.)

## ALGORITHM *BruteForceClosestPair(P)*

//Finds distance between 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: The distance between the closest pair of points

$d \leftarrow \infty$

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

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

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

**return**  $d$

**Basic Operation: Square root (a complicated operation) and multiplication**

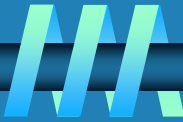
Can you avoid the square root in the loop?

**Efficiency:**

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

Divide-and-conquer approach can make it faster!

# 5. Brute-Force Strengths and Weaknesses



## ⌚ 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



# 6. 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:

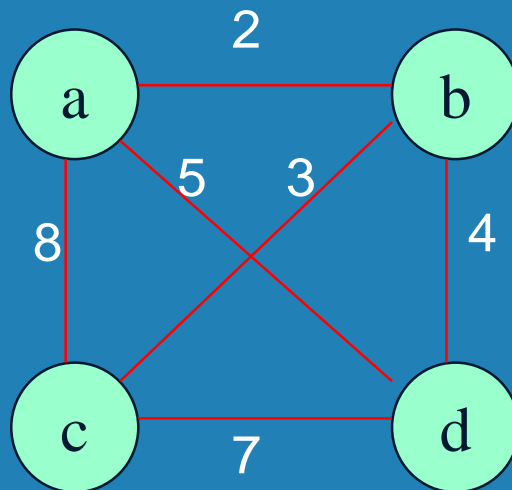
- generate a list of all potential solutions to the problem in a systematic manner (see algorithms in Sec. 5.4)
- evaluate potential solutions one by one, disqualifying infeasible ones and, for an optimization problem, keeping track of the best one found so far
- when search ends, announce the solution(s) found



# Example 1: Traveling Salesman Problem

- 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

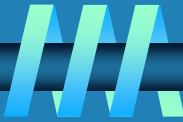
Example:



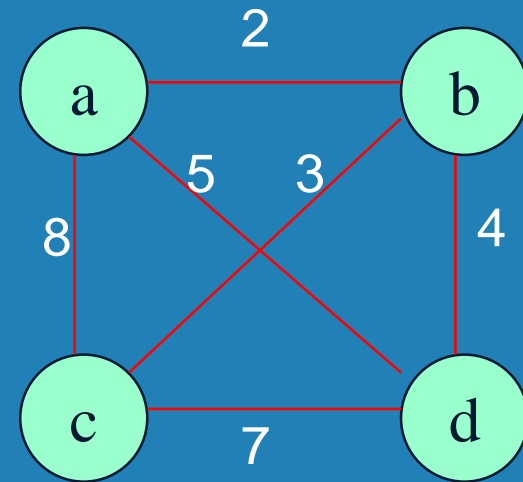
A Hamiltonian circuit is defined as a cycle that passes through all the vertices of the graph exactly once.



# TSP by Exhaustive Search



Tour	Cost
$a \rightarrow b \rightarrow c \rightarrow d \rightarrow a$	$2+3+7+5 = 17$
$a \rightarrow b \rightarrow d \rightarrow c \rightarrow a$	$2+4+7+8 = 21$
$a \rightarrow c \rightarrow b \rightarrow d \rightarrow a$	$8+3+4+5 = 20$
$a \rightarrow c \rightarrow d \rightarrow b \rightarrow a$	$8+7+4+2 = 21$
$a \rightarrow d \rightarrow b \rightarrow c \rightarrow a$	$5+4+3+8 = 20$
$a \rightarrow d \rightarrow c \rightarrow b \rightarrow a$	$5+7+3+2 = 17$



**Efficiency:**  $\Theta((n-1)!)$

Chapter 5 discusses how to generate permutations fast.



# Example 2: Knapsack Problem



Given  $n$  items:

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

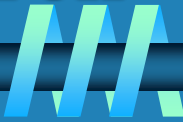
Find most valuable subset of the items that fit into the knapsack

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 Problem by Exhaustive Search



<u>Subset</u>	<u>Total weight</u>	<u>Total value</u>
{1}	2	\$20
{2}	5	\$30
{3}	10	\$50
{4}	5	\$10
{1,2}	7	\$50
{1,3}	12	\$70
{1,4}	7	\$30
<b>{2,3}</b>	<b>15</b>	<b>\$80</b>
{2,4}	10	\$40
{3,4}	15	\$60
{1,2,3}	17	not feasible
{1,2,4}	12	\$60
{1,3,4}	17	not feasible
{2,3,4}	20	not feasible
{1,2,3,4}	22	not feasible

**Efficiency:  $\Theta(2^n)$**

Each subset can be represented by a binary string (bit vector, Ch 5).

# Example 3: The Assignment Problem

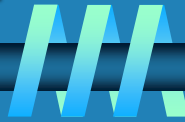
There are  $n$  people who need to be assigned to  $n$  jobs, one person per job. The cost of assigning person  $i$  to job  $j$  is  $C[i,j]$ . Find an assignment that minimizes the total cost.

	Job 0	Job 1	Job 2	Job 3
Person 0	9	2	7	8
Person 1	6	4	3	7
Person 2	5	8	1	8
Person 3	7	6	9	4

Algorithmic Plan: Generate all legitimate assignments, compute their costs, and select the cheapest one.

How many assignments are there?  $n!$

# Assignment Problem by Exhaustive Search



Assignment (col.#s)	Total Cost
1,2,3,4	$9+4+1+4 = 18$
1,2,4,3	$9+4+8+9 = 30$
1,3,2,4	$9+3+8+4=24$
1,3,4,2	$9+3+8+6=26$
1,4,2,3	$9+7+8+9=33$
1,4,3,2	$9+7+1+6=23$
2,1,3,4	$6+2+1+4 = 13$
2,1,4,3	...
2,3,1,4	
2,3,4,1	
2,4,1,3	
2,4,3,1	
...	

	Job 0	Job 1	Job 2	Job 3
Person 0	9	2	7	8
Person 1	6	4	3	7
Person 2	5	8	1	8
Person 3	7	6	9	4

- One can pick the best person for the job by looking at the minimum in each column.
- This can reduce the search space.

Number of possible  
assignment =  $24 = 4!$

# Final Comments on 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
- ⌚ In many cases, exhaustive search or its variation is the only known way to get exact solution

The Hungarian method  
runs in  $O(n^3)$  time.