# Analysis and Design of Algorithms

## **Chapter 3: Brute Force**



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## **Brute Force**

### **Brute Force**

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

Just do it

### • Example:

- ▲ Computing  $a^n$  (a > 0, n a nonnegative integer)
- **▲** Computing n!
- ▲ *Multiplying two matrices*
- ▲ Searching for a key of a given value in a list
- ▲ Consecutive Integer Algorithm for gcd (m,n)

## Brute-Force Sorting Alg. — Selection Sort

### Idea of Selection Sort

#### → Problem

Given an array of *n* orderable items (e.g. numbers, characters from some alphabet, character strings), rearrange them in non-decreasing order

### → Idea

- Scan the entire array to find its smallest element and swap it with the first element. — put the smallest element in its final position in the sorted array
- starting with the second element, to find the smallest among the next n-1 elements and swap it with the second element. — put the second smallest element in its final position in the sorted array
- Generally, on pass i ( $0 \le i \le n-2$ ), find the smallest element in A[i..n-1] and swap it with A[i]:
- After n-1 passes, the array is sorted

 $A[0] \leq \ldots \leq A[i-1] \mid A[i], \ldots, A[min], \ldots, A[n-1]$ in their final positions

```
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 ascending 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] \quad min \leftarrow j

swap A[i] and A[min]
```

### • Example:

Selection Sort on the list {89, 45, 68, 90, 29, 34, 17}

**FIGURE 3.1** Example of sorting with selection sort. Each line corresponds to one iteration of the algorithm, i.e., a pass through the list tail to the right of the vertical bar; an element in bold indicates the smallest element found. Elements to the left of the vertical bar are in their final positions and are not considered in this and subsequent iterations.

### **Analysis of Selection Sort**

- Basic operation: key comparison A[j] < A[min]
- Input size: number of elements, n
- Time efficiency  $\Theta(n^2)$

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=i+1}^{n-1} 1 = \sum_{i=0}^{n-2} [(n-1) - (i+1) + 1] = \sum_{i=0}^{n-2} (n-i-1)$$

$$= \sum_{i=0}^{n-2} (n-1) - \sum_{i=0}^{n-2} i = (n-1) \sum_{i=0}^{n-2} 1 - \sum_{i=0}^{n-2} i = (n-1)^2 - \frac{(n-2)(n-1)}{2}$$

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

• number of key swaps:  $\Theta(n)$ 

## Brute-Force Sorting Alg. — Bubble Sort

### **Idea of Bubble Sort**

### → Idea

- Compare adjacent elements of the list and exchange them if they are out of order
- By doing it repeatedly, we end up "bubbling" the largest element to the last position on the list
- The next past bubbles up the second largest element, and so on until, after n-1 passes, the list is sorted
- Pass i

$$A_0 \dots A_j \stackrel{?}{\longleftrightarrow} A_{j+1} \dots A_{n-i-1} \mid A_{n-i} \leq \dots \leq A_{n-1}$$

in their final positons

## **Bubble Sort**

```
ALGORITHM BubbleSort (A [0...n-1]) {

// Sorts a given array by bubble sort;

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

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

For i \leftarrow 0 to n-2 do

For j \leftarrow 0 to n-2-i do

if A[j+1] < A[j] swap A[j] and A[j+1]
```

## **Bubble Sort**

### • Example:

Bubble Sort on the list {89, 45, 68, 90, 29, 34, 17}

89
 
$$\stackrel{?}{\longrightarrow}$$
 45
 68
 90
 29
 34
 17

 45
 89
  $\stackrel{?}{\longrightarrow}$ 
 68
 90
 29
 34
 17

 45
 68
 89
  $\stackrel{?}{\longrightarrow}$ 
 90
  $\stackrel{?}{\longrightarrow}$ 
 34
 17

 45
 68
 89
 29
 90
  $\stackrel{?}{\longrightarrow}$ 
 34
 17

 45
 68
 89
 29
 34
 17
 90

 45
  $\stackrel{?}{\longrightarrow}$ 
 68
  $\stackrel{?}{\longrightarrow}$ 
 29
 34
 17
 90

 45
 68
 29
 89
  $\stackrel{?}{\longrightarrow}$ 
 34
 17
 90

 45
 68
 29
 34
 17
 90

 45
 68
 29
 34
 17
 90

 45
 68
 29
 34
 17
 90

## **Bubble Sort**

### Analysis of Bubble Sort

- Basic operation: key comparison
- Input size: number of elements, n
- Time efficiency  $\Theta(n^2)$

$$C(n) = \sum_{i=0}^{n-2} \sum_{j=0}^{n-2-i} 1 = \sum_{i=0}^{n-2} [(n-2-i) - 0 + 1] = \sum_{i=0}^{n-2} (n-i-1)$$
$$= \frac{n(n-1)}{2} \in \Theta(n^2)$$

number of key swaps: depends on the input

$$S_{worst}(n) = C(n) = \frac{n(n-1)}{2} \in \Theta(n^2)$$

**Thinking:** if a pass through the list makes no exchanges, the list has been sorted and we can stop the algorithm

### Idea of Brute-Force String Matching

text: a (longer) string of n characters to search in pattern: a string of m characters to search for (m <= n)

#### → Problem

find a substring in the text that matches the pattern,

precisely, find i — the index of the leftmost character of the first matching substring in the text — such that

$$t_i = p_0 \dots t_{i+j} = p_i \dots t_{i+m-1} = p_{m-1}$$

### → Idea

- S1: Align pattern against the first *m* characters of the text
- S2: compare corresponding pairs of characters from left to right, starting with the first character of the pattern and its counter part in the text, until
  - \_ Case1: all m pairs are found to match (successful search); or
  - \_ Case2: a mismatching pair is detected
- S3: In Case2, the text is not yet exhausted, realign pattern one position to the right and repeat S2, starting again with the first left pair.

Note: the last position in the text which can still be a beginning of a matching substring is *n-m* 

```
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
```

### • Example:

```
N O B O D Y _ N O T I C E D _ H I M
N O T
N O T
N O T
N O T
N O T
N O T
N O T
N O T
N O T
N O T
N O T
```

### **#** Analysis of Brute-Force String Matching

- Basic operation: key comparison
- Input size: n, m
- Time efficiency
  - ▲ worst case: it has to make all m comparisons before shifting the pattern, and this can happen for each of the n-m+1 tries.

$$C_{worst} = \Theta(nm)$$

▲ average case: for a typical word search, we can expect most shifts would happen after very few comparisons

$$C_{avg} = \Theta(n+m) = \Theta(n)$$

### Idea of Closest-Pair Problem

#### Problem

Find the two closest points in a set of n points (in the twodimensional Cartesian plane).

### → Idea

Compute the Euclidean distance between every pair of distinct points;

and return the indexes of the points for which the distance is the smallest.

$$d(P_i, P_j) = \sqrt{(x_i - x_j)^2 + (y_i - y_j)^2}$$

```
ALGORITHM BruteForceClosestPoints(P)

//Input: A list P of n (n \ge 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 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
```

### Idea of Closest-Pair Problem

### How to make it faster?

The basic operation of the algorithm is computing the Euclidean distance between two points.

The square root is a complex operation who's result is often irrational, therefore the results can be found only approximately. Computing such operations are not trivial.

 $-\rightarrow$  One can *avoid* computing square roots by comparing distance squares instead.

### Analysis of Closest-Pair Problem

- Basic operation: squaring a number
- Input size: number of points, n
- Time efficiency  $\Theta(n^2)$

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

## Brute-Force Polynomial Evaluation

### Idea of 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  $x = x_0$ 

Idea

## Brute-Force Polynomial Evaluation

```
\begin{aligned} p &\leftarrow 0.0 \\ \textbf{for } i \leftarrow n \ \textbf{down to} \ 0 \ \textbf{do} \\ power &\leftarrow 1 \\ \textbf{for } j \leftarrow 1 \ \textbf{to} \ i \ \textbf{do} \quad // \textbf{compute} \ x^i \\ power &\leftarrow power * x \\ p &\leftarrow p + a[i] * power \\ \text{return } p \end{aligned}
```

# Brute-Force Polynomial Evaluation

### **Better Polynomial Evaluation**

evaluating from right to left:

```
p \leftarrow a[0]
power \leftarrow 1
\mathbf{for}\ i \leftarrow 1\ \mathbf{to}\ n\ \mathbf{do}
power \leftarrow power * x
p \leftarrow p + a[i] * power
\mathbf{return}\ p
```

## Exhaustive Search

### **Problem**

searching for an element with a special property, in a domain that grows exponentially (or faster) with an instance size,

usually involve combinatorial objects such as permutations, combinations, or subsets of a set.

Many such problems are optimization problems, to find an element that maximizes or minimizes some desired characteristic

such as a path's length or an assignment's cost

## Exhaustive Search

### **Exhaustive Search— Brute-Force for combinatorial**

- generate a list of all potential solutions to the problem in a systematic manner
- selecting those of them that satisfy all the constraints
- evaluate potential solutions one by one, disqualifying infeasible ones and, for an optimization problem, keeping track of the best one found so far
- then search ends, announce the desired solution(s) found (e.g. the one that optimizes some objective function)
- typically requires for generating certain combinatorial objects

## Exhaustive Search: Traveling Salesman Problem

### **III** Idea

### → Problem

Given *n* cities with known distances between each pair, find the shortest tour that passes through <u>all</u> the cities <u>exactly once</u> before returning to the starting city

#### → Idea

weighted graph:

vertices: cities

edge weights: distances

 Alternatively: To find shortest Hamiltonian circuit in a weighted connected graph

Hamiltonian circuit: a cycle that passes through all the vertices of the graph exactly once

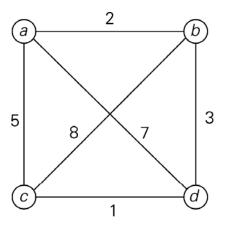
## Exhaustive Search: Traveling Salesman Problem

### → Idea

- Hamiltonian circuit can be defined as a sequence of n+1 adjacent vertices  $v_{i0}, v_{i1}, v_{i2}, \ldots, v_{in-1}, v_{i0}$
- generating all the permutations of n-1 intermediate cities
- computing the tour lengths
- find the shortest among them

## Traveling Salesman Problem

### Example:



Tour Length

$$a \longrightarrow b \longrightarrow c \longrightarrow d \longrightarrow a$$
  $l = 2 + 8 + 1 + 7 = 18$ 

$$I = 2 + 8 + 1 + 7 = 18$$

$$a \longrightarrow b \longrightarrow d \longrightarrow c \longrightarrow a$$
  $l = 2 + 3 + 1 + 5 = 11$ 

$$l = 2 + 3 + 1 + 5 = 11$$
 optimal

$$a \longrightarrow c \longrightarrow b \longrightarrow d \longrightarrow a$$
  $l = 5 + 8 + 3 + 7 = 23$ 

$$I = 5 + 8 + 3 + 7 = 23$$

$$a \longrightarrow c \longrightarrow d \longrightarrow b \longrightarrow a$$
  $l = 5 + 1 + 3 + 2 = 11$ 

$$I = 5 + 1 + 3 + 2 = 11$$
 optimal

$$a \longrightarrow d \longrightarrow b \longrightarrow c \longrightarrow a$$
  $l = 7 + 3 + 8 + 5 = 23$ 

$$I = 7 + 3 + 8 + 5 = 23$$

$$a \longrightarrow d \longrightarrow c \longrightarrow b \longrightarrow a$$
  $l = 7 + 1 + 8 + 2 = 18$ 

$$I = 7 + 1 + 8 + 2 = 18$$

## Traveling Salesman Problem

- **#** Analysis of Exhaustive Search for TSP
  - number of permutations (n-1)!

## Exhaustive Search: Knapsack Problem

### **III** Idea

#### → Problem

#### Given

```
weights: w_1 w_2 ... w_n values: v_1 v_2 ... v_n a knapsack of capacity W
```

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

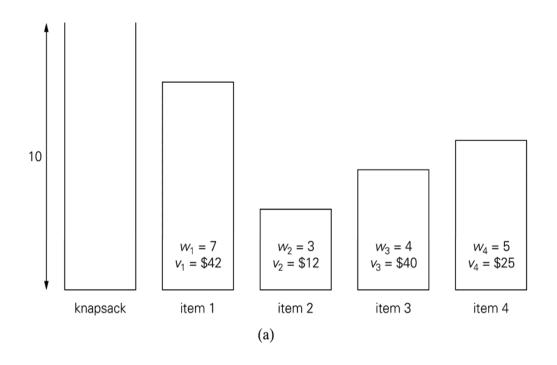
## Exhaustive Search: Knapsack Problem

### → Idea

- generating all subsets of the set of n items given
- computing the total weight of each feasible subset (i.e. the ones with the total weight not exceeding the knapsack's capacity)
- finding a subset of the largest value among them

## Knapsack Problem

### • Example:



Subset	Total weight	Total value	
Ø	0	\$ 0	
{1}	7	\$42	
{2}	3	\$12	
{3}	4	\$40	
<b>{4</b> }	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>	9	<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)

## Knapsack Problem

### **Analysis of Exhaustive Search for Knapsack**

• number of subsets for an n-element set 2<sup>n</sup>

For Exhaustive Search for Knapsack Problem and TSP problem,

- examples of so-called NP-hard problem
- no polynomial-time algorithm is known for NP-hard problem

## Exhaustive Search: Assignment Problem

### **III** Idea

#### Problem

There are n people who need to be assigned to n jobs, one person per job.

each person is assigned to exactly one job, and each job is assigned to exactly one person

The cost of assigning person i to job j is C[i, j]

Find an assignment that minimizes the total cost.

## Exhaustive Search: Assignment Problem

#### → Idea

describe the feasible solutions to the Assignment Problem as n-tuples  $\langle j_1, ..., j_n \rangle$  in which the i-th component indicates the column of the element selected in the i-th row (i.e. job number assigned to the i-th person)

- generating all legitimate assignments,
- compute their costs
- select the cheapest one

## Assignment Problem

### • Example:

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

Pose the problem as the one about a cost matrix:

$$C = \begin{bmatrix} 9 & 2 & 7 & 8 \\ 6 & 4 & 3 & 7 \\ 5 & 8 & 1 & 8 \\ 7 & 6 & 9 & 4 \end{bmatrix}$$

$$<1, 2, 3, 4> cost = 9 + 4 + 1 + 4 = 18$$

$$<1, 2, 4, 3> cost = 9 + 4 + 8 + 9 = 30$$

$$<1, 3, 2, 4> cost = 9 + 3 + 8 + 4 = 24$$

$$<1, 3, 4, 2> cost = 9 + 3 + 8 + 6 = 26$$

$$<1, 4, 2, 3> cost = 9 + 7 + 8 + 9 = 33$$

$$<1, 4, 3, 2> cost = 9 + 7 + 1 + 6 = 23$$

## Assignment Problem

### **#** Analysis of Exhaustive Search for Assignment

number of permutations n!

 no known polynomial-time algorithms for problems whose domain grows exponentially with instance size

## Final Comments

- → 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

## Final Comments

### Comments on Exhaustive Search

- Brute-force is a straightforward approach to solving a problem, directly based on the definitions or statement of a problem
- 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