

# E2011: Theoretical fundamentals of computer science

## Introduction to algorithms – part II

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RECETOX

# Outline

- 1 Review of pseudocode constructs
- 2 Basic data structures
- 3 Exercises

# Pseudocode

- *variables* - store some values (e.g.  $x, y$ ); may refer to simple (e.g. scalar) values, or more complicated *data structures* (vectors, matrices, lists, etc.)
- *input* to specify the required values for the algorithm to compute the *output*
- variables are assigned values:  $x \leftarrow 50$  or  $x \leftarrow y$ , but values are never assigned variables or other values:  $50 \leftarrow x$  is a nonsense
- mathematical operators can be used as usual

## Branching - conditional execution

```
if <condition> then
    code for <condition> is True
[
else
    code for <condition> is False
]
end if
```

- "else" branch is optional
- one can use "**continue**" to force quitting a loop or "**next**" to force jumping to the next iteration within a loop

# Loops

Repeat as long as the condition is true:

```
while <condition> do  
    instruction  
    ...  
end while
```

Repeat as long as the condition is false:

```
repeat  
    instruction  
    ...  
until <condition>
```

Repeat for all values in a series:

```
for <iterator> do  
    instructions  
end for  
for all <iterator> do  
    instructions  
end for
```

# Subroutines

## Procedures

```
procedure  $\langle name \rangle(\langle params \rangle)$   
    block  
end procedure
```

- may change the values of the parameters
- use **return** to cause immediate exit from the procedure

## Functions

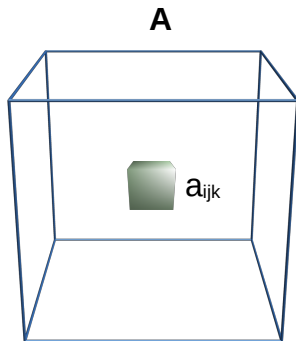
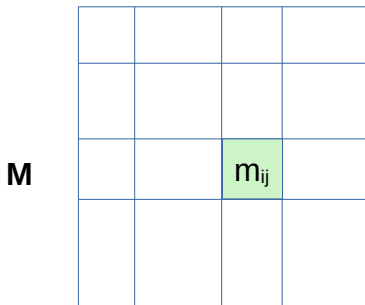
```
function  $\langle name \rangle(\langle params \rangle)$   
    body  
    return value  
end function
```

- does not change the values of the parameters
- returns a computed value

# Vectors and arrays

- may contain  $\geq 0$  elements
- all elements have the same type (e.g.  $\mathbb{N}, \mathbb{R}$ )
- each element is addressable by an index - e.g.  $i \in \mathbb{N}^*$  or  $i, j \in \mathbb{N}^*$
- the indexing induces an order among the elements
- for the purpose of this introductory course, we stick to single element addressing

$$\mathbf{x} \in \mathbb{R}^n : [x_j]; \quad \mathbf{M} \in \mathcal{M}_{m \times n}(\mathbb{R}) : [m_{ij}]; \quad \mathbf{A} \in \mathbb{R}^n \times \mathbb{R}^m \times \mathbb{R}^p$$





## Arrays - access patterns - examples

Access all  $[x_i]$   
sequentially:

```
for  $i = 1, 2, \dots, n$  do  
    work with  $x_i$   
    ...  
end for
```

Access all elements in odd-numbered columns:

```
for  $j = 2k + 1, k = 1, \dots, \lfloor (n - 1)/2 \rfloor$  do  
    for  $i = 1, \dots, m$  do  
        work with  $m_{ij}$   
        ...  
    end for  
end for
```

# Sets

- bag of elements, usually of the same type
- no inherent ordering
- needs an element selection strategy to be specified
- you can use sets operations to make things clear(er)
- e.g.  $A \subset \mathbb{Z}$ ,  $|A| = n$
- if asked to detail some operation (e.g. union), then you need to describe the algorithm, not only say " $A \cup B$ "

## Problem 1

Find the minimum and maximum of a (a) vector, (b) matrix, and (c) set of real numbers.

## Solution 1(a)

**Input:**  $x = [x_i] \in \mathbb{R}^n$

**Output:**  $x_{min} = \min_i(x); x_{max} = \max_i(x)$

$x_{min} \leftarrow x_1; x_{max} \leftarrow x_1$

▷ initial values

**for**  $i = 2, \dots, n$  **do**

**if**  $x_i > x_{max}$  **then**

$x_{max} \leftarrow x_i$

▷ a larger value was found

**end if**

**if**  $x_i < x_{min}$  **then**

$x_{min} \leftarrow x_i$

▷ a smaller value was found

**end if**

**end for**

## Solution 1(b)

**Input:**  $X = [x_{ij}] \in \mathcal{M}_{m,n}(\mathbb{R})$

**Output:**  $x_{min} = \min_{ij}(X)$ ;  $x_{max} = \max_{ij}(X)$

$x_{min} \leftarrow x_{11}$ ;  $x_{max} \leftarrow x_{11}$

▷ initial values

**for**  $i = 1, \dots, m$  **do**

**for**  $j = 1, \dots, n$  **do**

**if**  $x_{ij} > x_{max}$  **then**

$x_{max} \leftarrow x_{ij}$

▷ a larger value was found

**end if**

**if**  $x_{ij} < x_{min}$  **then**

$x_{min} \leftarrow x_{ij}$

▷ a smaller value was found

**end if**

**end for**

**end for**

## Solution 1(c)

**Input:**  $A \subset \mathbb{R}, A \neq \emptyset$

**Output:**  $a_{min} = \min(A); a_{max} = \max(A)$

$a_{min} \leftarrow \infty; a_{max} \leftarrow -\infty$

▷ initial values

**while**  $A \neq \emptyset$  **do**

$a \leftarrow$  pick random element from  $A$

**if**  $a > a_{max}$  **then**

$a_{max} \leftarrow a$

▷ a larger value was found

**end if**

**if**  $a < a_{min}$  **then**

$a_{min} \leftarrow a$

▷ a smaller value was found

**end if**

$A \leftarrow A \setminus \{a\}$

▷ remove the element from  $A$

**end while**

## Problem 2

Given a vector of real numbers, sort its elements in increasing order.

## Solution 2 (Bubble sort)

**Input:**  $x = [x_i] \in \mathbb{R}^n$

**Output:** sorted  $x$

**repeat**

swapped  $\leftarrow False$

**for**  $i = 1, \dots, n - 1$  **do**

**if**  $x_i > x_{i+1}$  **then**

$t \leftarrow x_i$

$x_i \leftarrow x_{i+1}$

$x_{i+1} \leftarrow t$

swapped  $\leftarrow True$

**end if**

**end for**

**until** not swapped

▷ need to swap  $x_i$  and  $x_{i+1}$



### Problem 3

Given a real-valued vector  $x$ , compute the (sample-based) estimates of the mean and standard deviation.

## Solution 3 - first version

Use  $\hat{\mu} = \frac{1}{n} \sum_i x_i$ ;  $\hat{\sigma}^2 = \frac{1}{n-1} \sum_i (x_i - \hat{\mu})^2$

**Input:**  $x = [x_i] \in \mathbb{R}^n$

**Output:**  $m$  - the mean;  $\sigma$  - the standard deviation

$m \leftarrow 0$

**for**  $i = 1, \dots, n$  **do**

$m \leftarrow m + x_i$

**end for**

$m \leftarrow \frac{m}{n}$

$s \leftarrow 0$

**for**  $i = 1, \dots, n$  **do**

$s \leftarrow s + (x_i - m)^2$

**end for**

$\sigma \leftarrow \sqrt{\frac{s}{n-1}}$

# Discussion

- why not use updates like  $m \leftarrow m + \frac{x_i}{n}$ ?
- what happens if  $n = 1$ ?
- what about underflow and precision of the result? how can we improve the robustness?
- can we make it faster - e.g. by passing only once through data?

## Solution 3 - second version

It can be show (prove it!) that

$$\hat{\sigma}^2 = \frac{1}{n-1} \sum_i (x_i - \mu)^2 = \frac{1}{n-1} \sum_i x_i^2 + \frac{1}{n(n-1)} \left( \sum_i x_i \right)^2$$

**Input:**  $x = [x_i] \in \mathbb{R}^n$

**Output:**  $m$  - the mean;  $\sigma$  - the standard deviation

$s_1 \leftarrow 0; s_2 \leftarrow 0$

**for**  $i = 1, \dots, n$  **do**

$s_1 \leftarrow s_1 + x_i$

$s_2 \leftarrow s_2 + x_i^2$

**end for**

$m \leftarrow \frac{s_1}{n}$

$\sigma \leftarrow \sqrt{\frac{s_2}{n-1} + \frac{s_1^2}{n(n-1)}}$

# Questions?