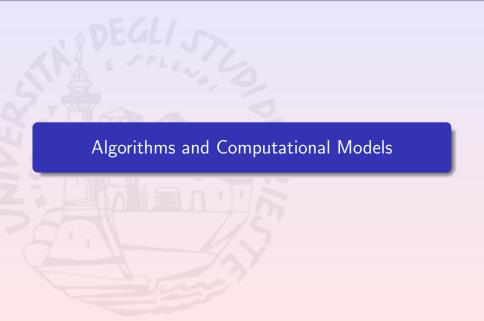
# Fundations Advanced Programming and Algorithmic Design

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## What is an Algorithm?

#### Definition (Algorithm)

Is a sequence of well-defined steps that transforms a set of inputs into a set of outputs in a finite amount of time

A function described by an algorithm is calculable.

A function implementable in a computational model is computable.

# Functions, Computability and Calcolability

Are all the functions computable in any specific model?

If this is not the case

- are there calculable functions that are not computable?
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# Why Is This Relevant For Us?

What if calculability will not be related to computability?

# Why Is This Relevant For Us?

What if calculability will not be related to computability?

Algorithms would not guarantee implementability!

# Halting Problem

Let h be the function that establish whether any program p eventually ends its execution  $(\downarrow)$  on an input i or runs forever $(\uparrow)$ 

$$h(p,i) \stackrel{\text{def}}{=} \left\{ \begin{array}{l} 0 & \text{if } p(i) \text{ never ends} \\ 1 & \text{otherwise} \end{array} \right.$$

#### Definition (Halting problem)

Can we implement h?

For any computable function f(a, b), define

$$g_f(i) \stackrel{\text{def}}{=} \left\{ \begin{array}{l} 0 & \text{if } f(i,i) = 0 \\ \uparrow & \text{otherwise} \end{array} \right.$$

Since f is computable, so it is  $g_f$ . Let  $G_f$  implement it.

Can h be one of the f's?

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Since f is computable, so it is  $g_f$ . Let  $G_-f$  implement it.

Can h be one of the f's?

• If 
$$f(G_-f, G_-f) = 0 \implies g_f(G_-f) = 0$$
 and  $h(G_-f, G_-f) = 1$ 

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Thus,  $h \neq f$  for all computable f's and h is not computable.

# Church-Turing Thesis

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Every effectively calculable function is a computable function.

 $calculability \Longrightarrow computability$ 

If we have an algorithm for f, then f can be formally computed

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#### calculability ⇒ computability

If we have an algorithm for f, then f can be formally computed

#### It also means that:

- all the "reasonable" computational models are equivalent
- we can avoid "hard-to-be-programmed" models (e.g., Turing machine)

# Random-Access Machine (RAM)

- variables to store data (no types)
- arrays
- integer and floating point constants
- algebraic functions: +, -, /, \*,  $\lfloor \cdot \rfloor$ ,  $\lceil \cdot \rceil$
- assignments
- pointers (no pointer arithmetic)
- conditional and loop statements
- procedure definitions and recursion
- simple "reasonable" functions, e.g., the length of an array

Algorithms are defined as programs on RAM.

# A Simple Algorithm

```
Input: An array A of numbers \langle a_1, \ldots, a_n \rangle.
Output: The maximum among a_1, \ldots, a_n.
def find_max(A):
     max_value \leftarrow A[1]
     for i \leftarrow 2..|A|:
           if A[i] > max_value:
                 max_value \leftarrow A[i]
           endif
     endfor
      return max_value
enddef
```

#### RAM is not Real Hardware!!!

RAM models real hardware, but it lacks

- real HW limitations s.a. finiteness
- memory hierarchy
- instruction execution time



# How to Measure Algorithm Efficiency?

What about execution time?



# How to Measure Algorithm Efficiency?

What about execution time? (for what input?)

Algorithms are not programs

Assuming 1 time unit per instruction are not realistic because execution time depends on:

- CPU instruction sets
- CPU/Memory/Bus Clock
- language and compiler
- OS memory handling
- . . .

# How to Measure Algorithm Efficiency?

What about execution time?

Any other ideas?

Time Complexity

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What about execution time?

Any other ideas?

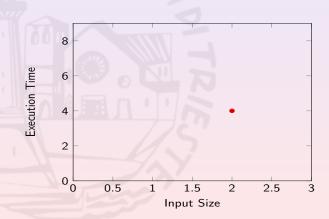
What about scalability?

#### Definition (Scalability)

Capacity for a system to handle input growth.

## **Growth Complexity**

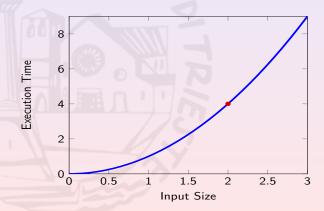
We do not measure the execution time for a given input



## **Growth Complexity**

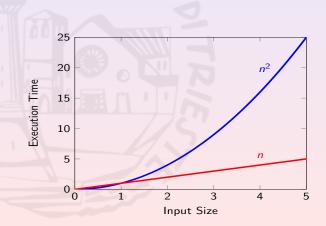
We do **not** measure the execution time for a given input

We estimate the relation between input size and execution time



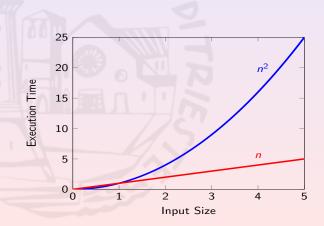
Which growth is preferable between:

•  $n^2$  and n?

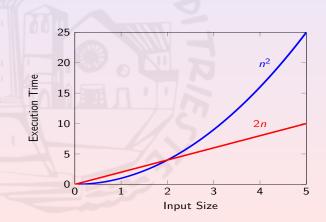


Which growth is preferable between:

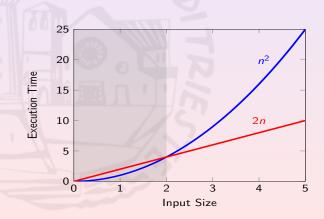
•  $n^2$  and  $\underline{n}$ ?



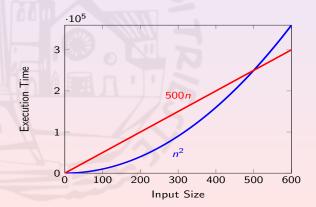
- $n^2$  and  $\underline{n}$ ?
- $n^2$  and 2 \* n?



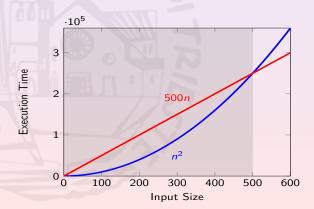
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- $n^2$  and  $\underline{n}$ ?
- $n^2$  and 2 \* n?
- $n^2$  and 500 \* n?



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# Asymptotic Time Complexity

Constants are not useful. We are looking at asymptotic behaviour.

We can abstract the single instruction execution time !!!

This intuition is also supported by linear time speedup theorem.

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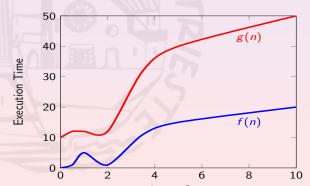
This intuition is also supported by linear time speedup theorem.

How to group all the functions that asymptotically are the same?

## big O notation

$$O(f(n)) \stackrel{def}{=} \{g(n)| \exists c > 0 \exists n_0 > 0 \ m \ge n_0 \Longrightarrow g(m) \le c * f(m)\}$$

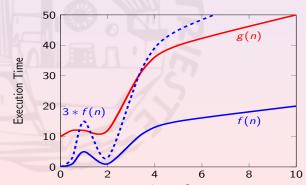
So,  $g(n) \in O(f(n))$  iff



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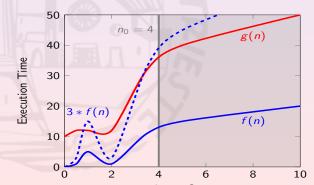
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# Some Useful Properties

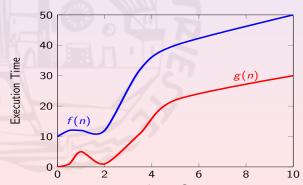
For any  $c_1, c_2 \in \mathbb{N}$  and for any  $k \in \mathbb{Z}$ 

- $f(n) \in O(f(n))$
- $O(f(n)) = O(c_1 * f(n) + k)$
- if  $c_1 \ge c_2$ , then  $O(f(n)^{c_1} + k * f(n)^{c_2}) = O(f(n)^{c_1})$
- $O(f(n)^{c_1}) \subseteq O(f(n)^{c_1+c_2})$  es.  $n \in O(n^2)$
- if  $h(n) \in O(f(n))$  and  $h'(n) \in O(g(n))$ , then
  - $h(n) + h'(n) \in O(g(n) + f(n))$
  - $h(n) * h'(n) \in O(g(n) * f(n))$

## big $\Omega$ notation

$$\Omega(f(n)) \stackrel{def}{=} \{g(n) | \exists c > 0 \exists n_0 > 0 \ m \ge n_0 \Longrightarrow c * f(m) \le g(m) \}$$

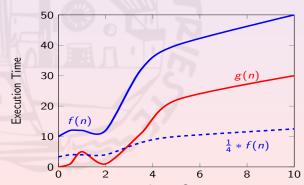
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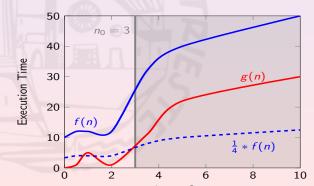
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So,  $g(n) \in \Omega(f(n))$  iff



# big $\Theta$ notation

$$\Theta(f(n)) \stackrel{\text{def}}{=} \{g(n) | \exists c_1, c_2 > 0 \exists n_0 > 0$$

$$m \ge n_0 \Longrightarrow c_1 * f(m) \le g(m) \le c_2 * f(m) \}$$

#### **Theorem**

$$f(n) \in \Theta(g(n)) \iff f(n) \in O(g(n)) \cap \Omega(g(n))$$

```
Input: An array A of numbers < a_1, \ldots, a_n >. Output: The maximum among a_1, \ldots, a_n.
```

```
def
         find_max(A):
                                                 • 2 costs O(1)
         max_value \leftarrow A[1]
         for i \leftarrow 2 upto len(A):
               if A[i] > max_value:
                    max_value \leftarrow A[i]
6
               endif
         endfor
8
                                               O(1
9
         return max_value
    enddef
10
```

```
Input: An array A of numbers < a_1, \ldots, a_n >. Output: The maximum among a_1, \ldots, a_n.
```

- 2 costs *O*(1)
- 4-6 cost O(1)
- 4-6 repeatedO(n) times
- 9 costs O(1)

$$O(1+1) =$$

```
Input: An array A of numbers < a_1, ..., a_n >. Output: The maximum among a_1, ..., a_n.
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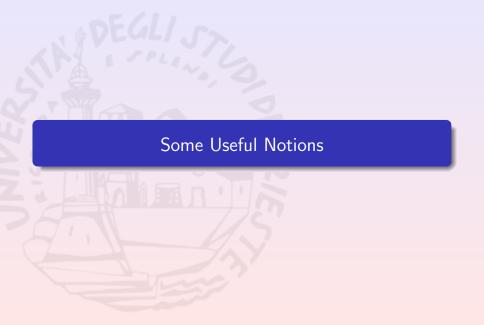
$$O(1+1*n) =$$

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Input: An array A of numbers < a_1, \ldots, a_n >. Output: The maximum among a_1, \ldots, a_n.
```

```
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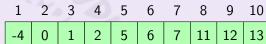
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$$O(1+1*n+1)=O(n)$$



# Arrays and Lists (Abstract Data Types)

Arrays Are indexed collections of values fixed in length.





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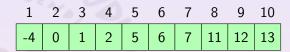
Arrays Are indexed collections of values fixed in length.

Single-Linked Lists Are sequences of values supporting head and next operations

Head 
$$\xrightarrow{\text{next}}$$
  $\xrightarrow{-4}$   $\xrightarrow{\text{next}}$   $\xrightarrow{0}$   $\xrightarrow{\text{next}}$   $\xrightarrow{1}$   $\xrightarrow{\text{next}}$ 

# Arrays and Lists (Abstract Data Types)

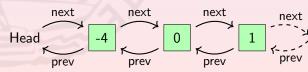
Arrays Are indexed collections of values fixed in length.



Single-Linked Lists Are sequences of values supporting head and next operations

Head 
$$\xrightarrow{\text{next}}$$
  $\xrightarrow{-4}$   $\xrightarrow{\text{next}}$   $\xrightarrow{0}$   $\xrightarrow{\text{next}}$   $\xrightarrow{1}$   $\xrightarrow{\text{next}}$ 

Double-Linked Lists Are sequences of values supporting head, next and previous operations



# Queue and Stacks (Abstract Data Types)

- Queues Are collections of values ruled according the FIFO policy. They support head, is\_empty, insert\_back, extract\_head operations
- Stacks Are collections of values ruled according the LIFO policy. They support top, is\_empty, insert\_top, extract\_top operations

# Graphs (Graph Theory)

Are pairs (V, E) where:



Are pairs (V, E) where:

V is a set of nodes



b



e



(g)

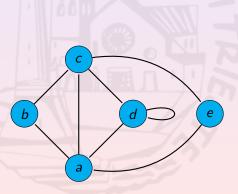
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# Graphs (Graph Theory)

Are pairs (V, E) where:

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*E* is a set of edges





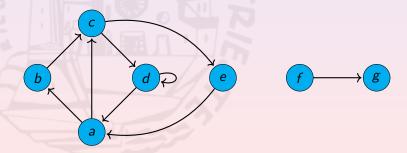
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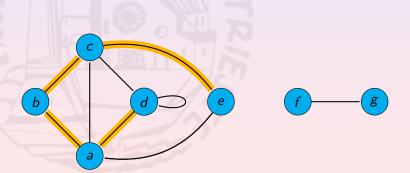
If the edges are (un)directed, the graph is (un)directed



# Paths and Cycles

A path of length n between  $a, b \in V$  is a sequence  $e_1, \ldots, e_n$  s.t.

- e<sub>1</sub> involves a
- e<sub>n</sub> involves b
- $e_i$  and  $e_{i+1}$  involve a common node  $n_i$

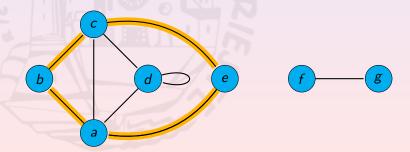


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- e<sub>n</sub> involves b
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A cycle is a path whose initial and final node coincide.



# Connected and Acyclic Graphs (Graph Theory)

A graph is connected if there is a path between every pairs of nodes

A graph is acyclic if it does not contains cycles

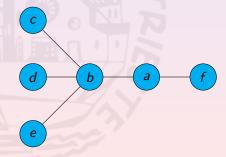


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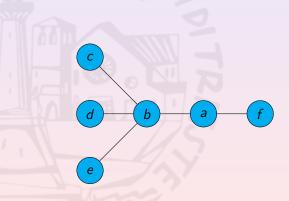
A graph is acyclic if it does not contains cycles

A tree is an connected and acyclic undirected graphs



# Trees (Abstract Data Types)

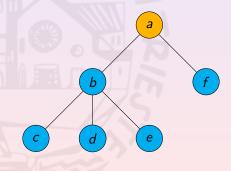
Organize data in a hierarchical finite tree (graph theory)



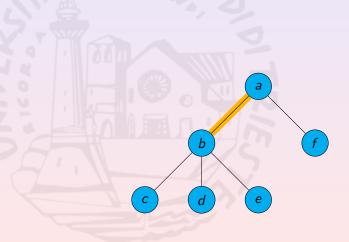
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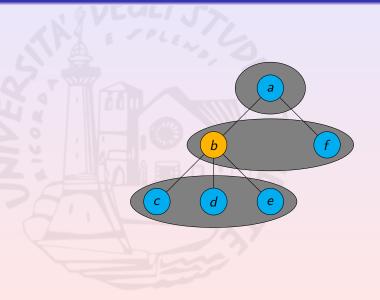
Organize data in a hierarchical finite tree (graph theory)

One of the nodes is the root of the graph

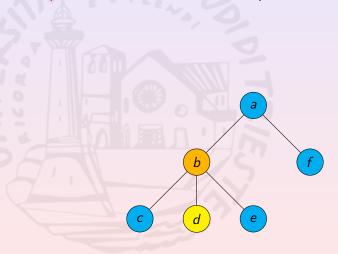


The depth of a node is its distance from the root



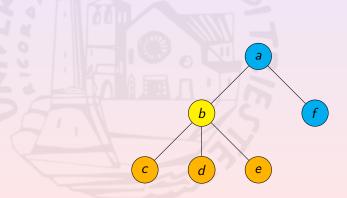


The parent of a node is a node one step closer to the root



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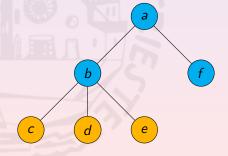
The children of a node have it as parent



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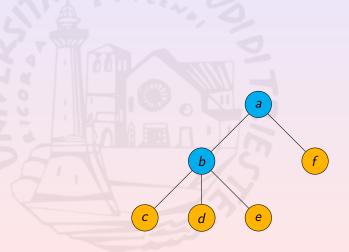
The children of a node have it as parent

Two nodes are siblings if they have the same parent



# Tree Leaves and Height

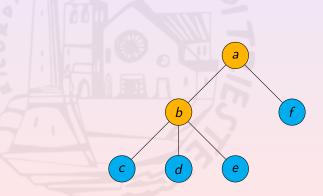
The leaves are nodes without children



# Tree Leaves and Height

The leaves are nodes without children

The internal nodes have children

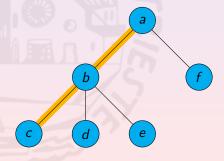


### Tree Leaves and Height

The leaves are nodes without children

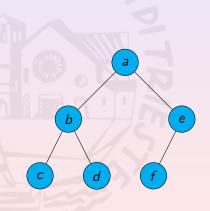
The internal nodes have children

The height of a tree is the max depth among those of its leaves



### *n*-ary Tree and Completeness

Every node of a n-ary tree can have up to n children



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Every node of a n-ary tree can have up to n children

A n-ary tree is complete if the nodes in all the levels but the last one have n children

