

# Can we define what an algorithm is?

#### **Simone Martini**

Dipartimento di Informatica – Scienza e Ingegneria

### A pre-mathematical concept

A sequence of rules, where
each rule
the way they are composed
are "effective"

The intuitive notion is the "touchstone" of any formal definition

Its formalisation will always loose (or add) some *nuance* in relation to the intuitive sense



### **Turing, 1936**

It does *not* deal with the notion of algorithm

It defines the notion of *effective*, i.e. *mechanically computable* 

effective = there is an algorithm that can compute it

But for a single effective function, there are many distinct algorithms

Many other systems for computable functions:

- Post systems, lambda-calculus, Gödel's general recursive functions
- All programming languages: e.g. C, Java, Python

They are all equivalent to each other: the Church-Turing thesis



### **Equivalence between formalisms for computability**

The equivalence holds "modulo coding" (à codage près)

Let S be a system for computability (lambda-calculus, Post systems, Python, etd.)

- 1. In S we may "code" the integers: [n]
- 2. For every (Turing-) computable functions over the integers f, there is a term F in S which "computes" f

$$F[n] \rightarrow [f(n)]$$

No guarantees on the preservation of algorithms!



# The "parallel or"

$$por(x,y) = \begin{cases} 0 & \text{if } x \text{ terminates or } y \text{ terminates} \\ it \ does \ not \ terminate & \text{otherwise} \end{cases}$$

Computable: evaluate "in parallel" x and y



# The "parallel or" in the lambda-calculus

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Computable: evaluate "in parallel" x and y

G. Berry: there is no lambda-term P which, for any pair of lambda terms M,N:

P N M has a normal form if and only if

(at least) one of the two terms M and N has a normal form

And yet, the lambda calculus is Turing-complete!



# The "parallel or" in the lambda-calculus

$$por(x,y) = \begin{cases} 0 & \text{if } x \text{ terminates or } y \text{ terminates} \\ it \ does \ not \ terminate & \text{otherwise} \end{cases}$$

Equivalence is "modulo coding":

- Each lambda term M is coded by another lambda term [M]
- We have a term Eval which encode the procedure of reduction
- There is a term[OR]

such that

Eval [OR] [M] [N]

has a normal form if and only if one of M and N has a normal form



## **Algorithms and abstraction levels**

An algorithm is only well defined "modulo one level of abstraction" [Gurevich]

Palindromicity of a sequence: rêver, radar, kayak



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**KAYAK** 

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Elementary operation: "select a generic element from a sequence"

Complexity: length('KAYAK')/2



## **Algorithms and abstraction levels**

Palindromicity of a sequence: rêver, radar, kayak

Elementary operation: "select a generic element from a sequence"

Complexity: length('KAYAK')/2

A Turing machine cannot ""select a generic element from a sequence"!

Back and forth every time!

RESSASSER

Complexity: length('RESSASSER')<sup>2</sup>



### **Algorithms and programs**

At the limit, one risks identifying algorithms and programs

For computer science, they are different concepts:

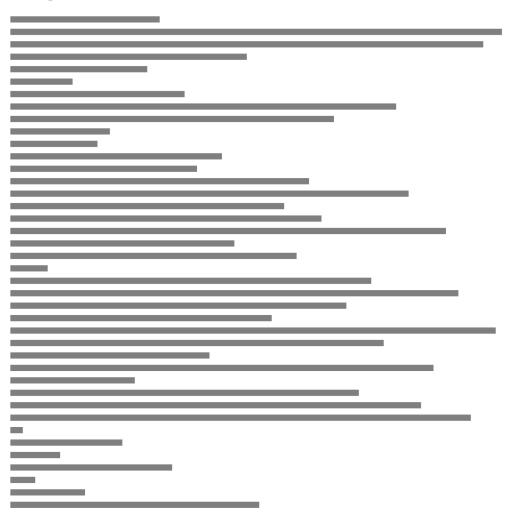
- the generic description of a process;
- its "translation", its "coding", in a programming language

And yet,

two languages never have the same set of elementary operations



Quicksort [Hoare, 1961]





#### Python:



JAVA:

```
public static void quickSort(int[] arr, int start, int end){
        int partition = partition(arr, start, end);
        if(partition-1>start) {
            quickSort(arr, start, partition - 1);
        }
        if(partition+1<end) {</pre>
            quickSort(arr, partition + 1, end);
        }
public static int partition(int[] arr, int start, int end){
        int pivot = arr[end];
        for(int i=start; i<end; i++){</pre>
            if(arr[i]<pivot){</pre>
                 int temp= arr[start];
                 arr[start]=arr[i];
                 arr[i]=temp;
                                                  Python:
                 start++;
                                                  def QuickSort(L):
                                                    if L==[]: return L
                                                    pivot = L[0]
                                                    return QuickSort([x for x in L[1:] if x < pivot])</pre>
        int temp = arr[start];
                                                           + [pivot] +
        arr[start] = pivot;
                                                           QuickSort([x for x in L[1:] if x >= pivot])
        arr[end] = temp;
        return start;
```

return start;

#### JAVA: public static void quickSort(int[] arr, int start, int end){ int partition = partition(arr, start, end); if(partition-1>start) { quickSort(arr, start, partition - 1); } if(partition+1<end) {</pre> quickSort(arr, partition + 1, end); } public static int partition(int[] arr, int start, int end){ int pivot = arr[end]; for(int i=start; i<end; i++){</pre> if(arr[i]<pivot){</pre> int temp= arr[start]; arr[start]=arr[i]; arr[i]=temp; Python: start++; def QuickSort(L): if L==[]: return L pivot = L[0]return QuickSort([x for x in L[1:] if x < pivot])</pre> int temp = arr[start]; + [pivot] + arr[start] = pivot; QuickSort([x for x in L[1:] if x >= pivot]) arr[end] = temp;

Are they really two different encodings of the same algorithm?

#### Two actors:

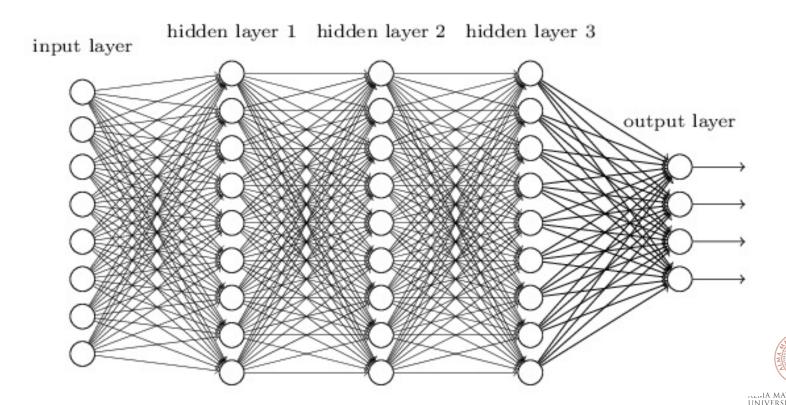
- the learning algorithm (generic)
- the learning outcome
   (the behaviour of the network, after learning)

Learning specialises (through training) a generic network into a specific function



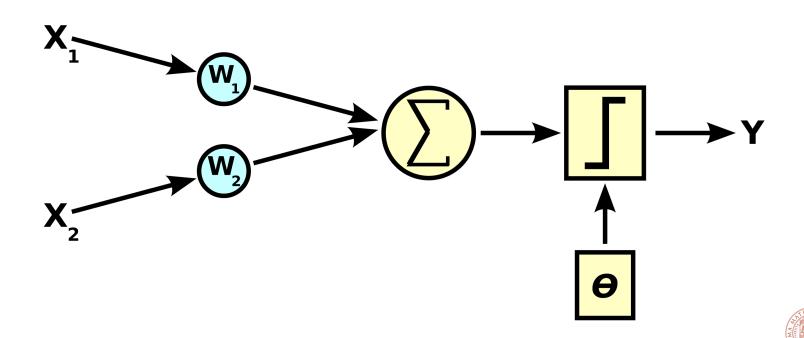
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Learning specialises (through training) a generic network into a specific function

The whole learning outcome is contained in the network weights

"classic" algorithm: the choices made by the algorithm are explicit

"neural" algorithm: everything is "opaque", hidden in the weights

An issue of *accountability...* 



# algorithms that do not terminate

Classical algorithms: termination, "the outcome", the "result"

An operating system: an infinite loop which does things through interaction with the actors of computation (resources, processes, environment, people, ecc. )

Not transformation of input data into a result, but interaction, which is a function of data, time, human actors...



### the message of this lecture

An *intuitive* concept: normative, touchstone a sequence of effective steps

There is not *a single* formalisation of the algorithm concept of the same algorithm

There are *various* formalisations

at different abstraction levels

each one accounts for some aspects, and loses others



## the message of this lecture

the plurality of formal approaches

even, cum grano salis, their disagreeement

is good

how is good the pluralism of ideas in a healthy democratic society



## algorithms in our daily life

knowledge: the algorithm must be public!

responsability: who is responsible for decisions?

- understanding: what skills are needed?

E.g., Maël Pégny, Issam Ibnouhsein.

Quelle transparence pour les algorithmes d'apprentissage machine? *Rev. d'Intelligence Artif. 32(4): 447-478 (2018).* 





#### **Simone Martini**

Dipartimento di Informatica – Scienza e Ingegneria

simone.martini@unibo.it

www.unibo.it