

# Computability

- 2 approaches : operational  $\times$  functional
- Church thesis

## Foundation of Recursive Function Theory

We will try to identify such functions which are "computable" in common sense (without respect to a given computer or computational system)

To decrease an extreme amount of such functions we abstract domains and ranges to integers and we will consider functions of the form:

$$f: \mathbb{N}^m \rightarrow \mathbb{N}^n$$

where  $\mathbb{N} = \{0, 1, 2, \dots\}$ ,  $m, n \in \mathbb{N}$

We will call this functions partial functions with classification:

Partial functions

Total functions

Strictly partial functions

Example:

Total function

$$\text{plus}: \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$$

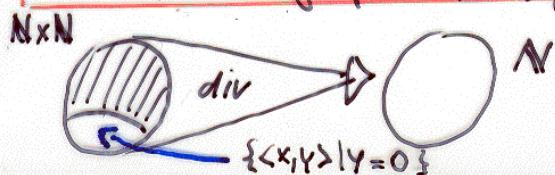
$$\text{plus}(x, y) = x + y$$



Strictly part. function

$$\text{div}: \mathbb{N} \times \mathbb{N} \rightarrow \mathbb{N}$$

$$\text{div}(x, y) = \text{integer part of } x/y, \text{ if } y \neq 0$$



### Convention:

a tuple  $(x_1, x_2, \dots, x_n) \in \mathbb{N}^n$  will be denoted as  $\bar{x}$

### Initial functions

The hierarchy of computable function is based on suitably chosen so called initial functions representing elementary functions (building stones for more complex functions). There are the following initial functions:

#### 1. Zero function

$$f() = 0$$

The function maps "empty tuple" to zero

#### 2. Successor function

$$\sigma: \mathbb{N} \rightarrow \mathbb{N}$$

$$\sigma(x) = x + 1$$

#### 3. Projections

$$\pi_k^n: \mathbb{N}^n \rightarrow \mathbb{N}$$

The function extracts from n-tuple the k-th component

Ex.:  $\pi_2^3(4, 6, 4) = 6$  or  $\pi_1^2(5, 17) = 5$

Special case:  $\pi_0^n: \mathbb{N}^n \rightarrow \mathbb{N}^0$

$$\pi_0^3(1, 2, 3) = ()$$

## Primitive Recursive Functions

These functions are defined by 3 principles of joining functions:

### 1. Combination

The combination of two functions  $f: \mathbb{N}^k \rightarrow \mathbb{N}^m$  and  $g: \mathbb{N}^n \rightarrow \mathbb{N}^l$  is a function:

such that

$$f \times g: \mathbb{N}^k \rightarrow \mathbb{N}^{m+n}$$

$$f \times g(\bar{x}) = (f(\bar{x}), g(\bar{x}))$$

Ex.:

$$\Pi_1^3 \times \Pi_3^3 (4, 12, 8) = (4, 8)$$

### 2. Composition

The composition of two functions  $f: \mathbb{N}^k \rightarrow \mathbb{N}^m$  and  $g: \mathbb{N}^m \rightarrow \mathbb{N}^n$  is a function

$$g \circ f: \mathbb{N}^k \rightarrow \mathbb{N}^n$$

such that  $g \circ f(\bar{x}) = g(f(\bar{x}))$

Ex.:

$$G \circ f() = 1 \quad G \circ G \circ f() = 2$$

$$G \circ \Pi_1^2(3, 1) = 4$$

### 3. Primitive recursion

#### Example

Let us suppose we wanted to define a function  $f: \mathbb{N}^2 \rightarrow \mathbb{N}$  such that  $f(x, y)$  would be the number of nodes in a full, balanced tree in which every nonleaf node has  $x$  children ( $x$ -ary tree) with depth of  $y$ .



We can see:

$$(1) f(x, 0) = 1$$

(2) The tree of depth  $n$  has  $x^n$  leafs. If we increase the depth to  $n+1$  we must add  $x^{n+1}$  nodes

$$x^{n+1} = x^n \cdot x$$

Now we are prepared to define the function  $f$  using following principle:

$$f(x, 0) = x^0$$

$$f(x, y+1) = f(x, y) + x^y \cdot x$$

For example:  $f(3, 2) = f(3, 1) + 3^1 \cdot 3 = 4 + 9 = 13$

$$f(3, 1) = f(3, 0) + 3^0 \cdot 3 = 1 + 3 = 4$$

3. Primitive recursion is a technique which defines a function  $f$

$$f: \mathbb{N}^{k+1} \rightarrow \mathbb{N}^m$$

from two functions  $g$  and  $h$

$$g: \mathbb{N}^k \rightarrow \mathbb{N}^m$$

$$h: \mathbb{N}^{k+m+1} \rightarrow \mathbb{N}^m$$

by the following prescription:

$$f(\bar{x}, 0) = g(\bar{x})$$

$$f(\bar{x}, y+1) = h(\bar{x}, y, f(\bar{x}, y)) \quad \bar{x} \in \mathbb{N}^k$$

A schema of computation (illustration for  $y=3$ )

$$f(\bar{x}, 3) = h(\bar{x}, 2, f(\bar{x}, 2))$$

$$f(\bar{x}, 2) = h(\bar{x}, 1, f(\bar{x}, 1))$$

$$f(\bar{x}, 1) = h(\bar{x}, 0, f(\bar{x}, 0))$$

$$f(\bar{x}, 0) = g(\bar{x})$$

e.g. two "directions"

- forward (until  $f(\bar{x}, 0)$  is computed)
- backward (until computation of  $f(\bar{x}, y)$  is completed)

### Example

Let us consider the function

$$\text{plus} : \mathbb{N}^2 \rightarrow \mathbb{N}$$

$$\text{plus}(x, y) = x + y$$

Using primitive recursion we can define the function plus in this way:

$$\text{plus}(x, 0) = \Pi_1^1(x)$$

$$\text{plus}(x, y+1) = \sigma \circ \Pi_3^3(x, y, \text{plus}(x, y))$$

e.g.

$$x + 0 = 0$$

$x + (y+1)$  is obtained (recursively) by finding the successor of  $x+y$

### Definition

The class of primitive recursion functions is a class of all function constructed from initial function using combination, composition, and primitive recursion, including initial functions.

### Theorem

Any primitive recursive function is a total function.

### Proof.

(a) initial function are total functions

(b) by application of combination, composition or primitive recursion to total functions results to a total function

## Typical primitive recursive functions (PRF)

PRF's are used in typical applications of computers.

Convention: Instead functional notation we will use common notation frequently:

$$\text{Instead } h \equiv \text{plus}_0(\Pi_1^3 \times \Pi_3^3)$$

$$\text{we use } h(x, y, z) = \text{plus}(x, z) \text{ or}$$

$$h(x, y, z) = x + z$$

### Constant functions

We introduce function  $K_m^n$  which to any tuple  $\bar{x} \in \mathbb{N}^n$  assigns the constant value  $m \in \mathbb{N}$ :

$$K_m^0 \equiv \underbrace{G_0 G_0 \dots \circ G}_m \{$$

Also for  $n > 0$  function  $K_m^n$  is PRF:

$$K_m^n(\bar{x}, 0) = K_m^{n-1}(\bar{x})$$

$$K_m^n(\bar{x}, y+1) = \Pi_{n+1}^{n+1}(\bar{x}, y, K_m^n(\bar{x}, y))$$

$$\begin{aligned} \text{For example: } K_3^2(1, 1) &= \Pi_3^3(1, 0, K_3^2(1, 0)) = K_3^1(1) = K_3^1(0) = \\ &= K_3^0() = 3 \end{aligned}$$

Constants from  $\mathbb{N}^n$  can be constructed by combination of  $K_m^n$ . For example  $K_2^3 \times K_5^3(x, y, z) = (2, 5)$ .

## Multiplication

$$\text{mult}(x, 0) = k_0^1(x)$$

$$\text{mult}(x, y+1) = \text{plus}(x, \text{mult}(x, y))$$

## Exponentiation

$$\text{exp}(x, 0) = k_1^1(x)$$

$$\text{exp}(x, y+1) = \text{mult}(x, \text{exp}(x, y))$$

## Predecessor function

$$\text{pred}(0) = f()$$

$$\text{pred}(y+1) = \Pi_1^2(y, \text{pred}(y))$$

## Function monus (modification of minus)

$$\text{monus}(x, 0) = x$$

$$\text{monus}(x, y+1) = \text{pred}(\text{monus}(x, y))$$

$$\text{e.g. } \text{monus}(x, y) = \begin{cases} x-y & \text{if } x \geq y \\ 0 & \text{otherwise} \end{cases}$$

$$\text{notation : monus}(x, y) \equiv x - y$$

## Function equal

$$\text{eq}(x, y) = \begin{cases} 1 & \text{if } x = y \\ 0 & \text{if } x \neq y \end{cases}$$

$$\text{eq}(x, y) = 1 - ((y - x) + (x - y))$$

$$\text{eq}(5, 3) = 1 - ((3 - 5) + (5 - 3)) = 1 - (0 + 2) = 1 - 2 = 0$$

We can define also

$$\begin{aligned}\top \varphi &= \text{monus}_0(K_1^2 \times \varphi), \\ &\equiv 1 \div \varphi\end{aligned}$$

### Tabular function

Let us consider functions of the type

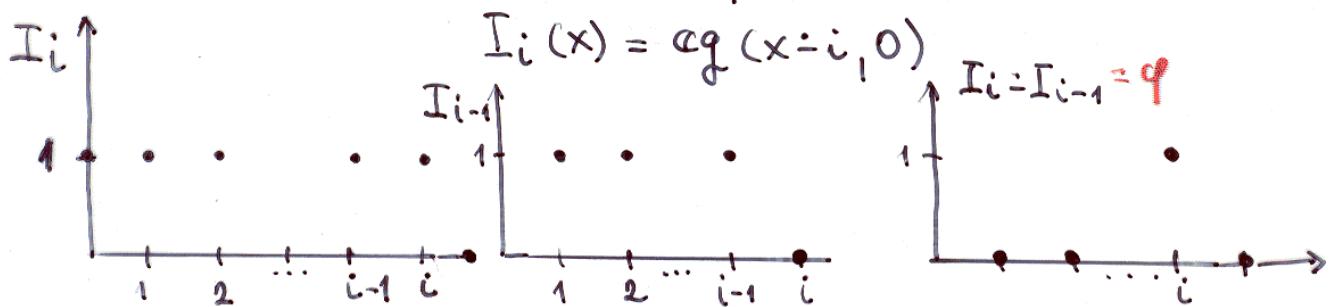
$$f(x) = \begin{cases} 3 & \text{if } x=0 \\ 5 & \text{if } x=4 \\ 2 & \text{otherwise} \end{cases}$$

which are usually defined by a table. These functions can be described using characteristic functions

$$\varphi_i(x) = \begin{cases} 1 & \text{if } x=i \\ 0 & \text{otherwise} \end{cases}$$

which can be expressed as

monus( $I_i, I_{i-1}$ ) where



Tabular functions can now be created by finite sum of multiplications of constants with  $\varphi_i$  or  $\top \varphi_i$ .

Example (function f above)

$$f \equiv \text{mult}(3, \varphi_0) + \text{mult}(5, \varphi_4) + \text{mult}(2, \text{mult}(\top \varphi_0, \top \varphi_4))$$

## Function quo (quotient)

$$\text{quo}(x, y) = \begin{cases} \text{integer part of } x/y & \text{if } y \neq 0 \\ 0 & \text{if } y = 0 \end{cases}$$

Using primitive recursion:

$$\text{quo}(0, y) = 0$$

$$\text{quo}(x+1, y) = \text{quo}(x, y) + \text{eq}(x+1, \text{mult}(\text{quo}(x, y), y) + y)$$

## Beyond Primitive Recursive Function

There are functions which are computable, but which are not PRFs. Such functions are all strictly partial functions (as div) but even some total functions. An example of such total function is so called Ackermann's function (1928), which is defined by equations:

$$A(0, y) = y + 1$$

$$A(x+1, 0) = A(x, 1)$$

$$A(x+1, y+1) = A(x, A(x+1, y))$$

e.g.  $A : \mathbb{N}^2 \rightarrow \mathbb{N}$

## Theorem

There is a total function from  $\mathbb{N}$  to  $\mathbb{N}$  which is computable but not PRF.

Proof : Considering PRFs as strings, we can order<sup>\*</sup> them :  $f_1, f_2, \dots, f_n$ . Let us define the function  $f : \mathbb{N} \rightarrow \mathbb{N}$  now such that

$$f(n) = f_n(n) + 1$$

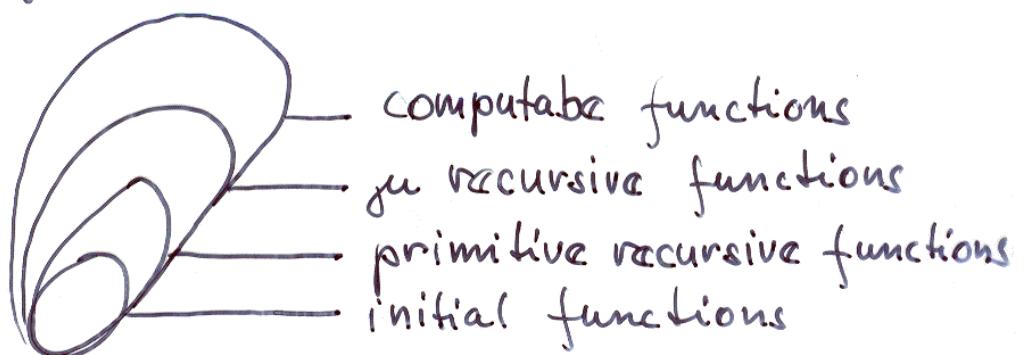
for all  $n \in \mathbb{N} \setminus \{0\}$ .

Clearly,  $f$  is total and computable function. But  $f$  is not PRF. Indeed if  $f$  would be a PRF then  $f \equiv f_m$  for some  $m \in \mathbb{N}$ .

But in this case  $f(m) = f_m(m)$  but not  $f_m(m) + 1$  according to definition of.

## Definition

The class of total computable function is called recursiv functions.



\*) in lexicographical order for example

## Partial recursive functions

To extend the class of computable functions beyond total computable functions we introduce a technique called minimisation:

This technique enables to construct function

$$f: \mathbb{N}^n \rightarrow \mathbb{N}$$

from function  $g: \mathbb{N}^{n+1} \rightarrow \mathbb{N}$

in such way that  $f(\bar{x})$  is the smallest  $y$  for which

$$(1) \quad g(\bar{x}, y) = 0$$

(2)  $g(\bar{x}, z)$  is defined for all  $z < y, z \in \mathbb{N}$

For this construction we will used notation

$$f(\bar{x}) = \mu y [g(\bar{x}, y) = 0]$$

### Examples

$$1. \quad f(x) = \mu y [\text{plus}(x, y) = 0], \text{ e.g. } f(x) = \begin{cases} 0 & \text{if } x=0 \\ \text{undef} & \text{otherwise} \end{cases}$$

$$2. \quad \text{div}(x, y) = \mu t [((x+1) \div (\text{mult}(t, y) + y)) = 0]$$

$$3. \quad i(x) = \mu y [\text{monus}(x, y) = 0] - \text{identical function}$$

Indeed, a function defined by minimisation is computable. The computation of  $f(\bar{x})$  includes computations of

$$g(\bar{x}, 0)$$

$$g(\bar{x}, 1)$$

$$g(\bar{x}, 2)$$

:

until

(a)  $g(\bar{x}, y) = 0$  (then  $f(\bar{x}) = y$ )

or (b)  $g(\bar{x}, z)$  is not defined ( $f(\bar{x})$  is not defined)

### Definition

The class of partial recursive functions is the class of partial functions which can be constructed from initial functions using

- (a) combination
- (b) composition
- (c) primitive recursion
- (d) minimisation

## Turing-computable functions

### Definicion

A Turing machine  $M = (Q, \Sigma, \Gamma, \delta, q_0, q_F)$  computes a partial function

$$f: \Sigma^{*m} \rightarrow \Sigma^{*n} \quad (\Sigma \subseteq \Gamma, \Delta \notin \Sigma)$$

iff for any  $(w_1, w_2, \dots, w_m) \in \Sigma^{*m}$  and corresponding initial configuration

$$\Delta w_1 \Delta w_2 \Delta \dots \Delta w_m \Delta \Delta \Delta \dots$$

the machine M

(1) stops in the case when  $f(w_1, w_2, \dots, w_m)$  is defined with content of the tape

$$\Delta v_1 \Delta v_2 \Delta \dots \Delta v_n \Delta \Delta \Delta$$

$$\text{and } (v_1, v_2, \dots, v_n) = f(w_1, w_2, \dots, w_m)$$

(2) cycles or stops abnormally when  $f(w_1, w_2, \dots, w_m)$  is not defined

The partial function which can be computed by some Turing machine is said the Turing-computable function.

### Examples

1.



The machine computes the function  $f(w_1, w_2, w_3) = (w_1, w_3)$

## 2. The function

$$g(w) = \begin{cases} 1 & \text{iff } w = \sigma(M) \text{ for the selfterminating} \\ & \text{machine } M \\ 0 & \text{otherwise} \end{cases}$$

is not Turing computable function.

## Turing Computability of Partial Recursive Functions

Let us consider Turing machines with alphabet  $\Sigma = \{0, 1\}$  and partial functions  $f: \mathbb{N}^m \rightarrow \mathbb{N}^n$

$m$ -tuples ( $n$ -tuples) will be coded as tuples of binary numbers:

$\Delta 11 \Delta 10 \Delta 100 \Delta \Delta$  represents 3-tuples (3, 2, 4)

### Theorem

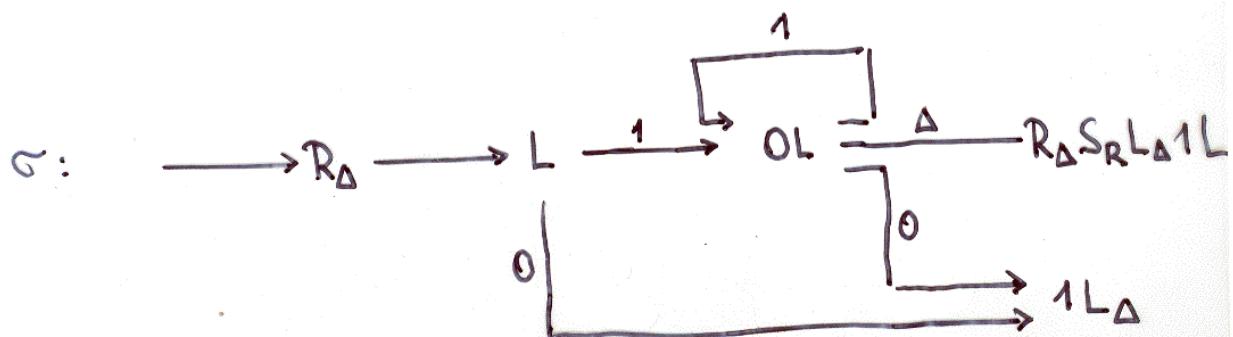
Every partial recursive function is Turing-computable.

Proof. 1. At first we have to find Turing machines which compute the initial functions

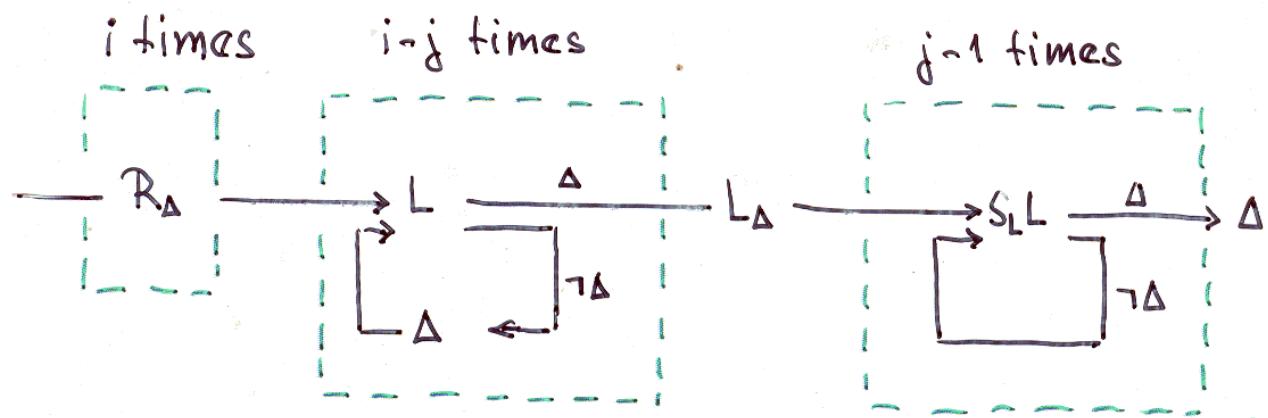
$\xi, \sigma, \pi$ :

(a)  $\xi: \Sigma^* \rightarrow ROL$

(b) a machine to compute  $\sigma$



(c) a machine to compute  $\Pi_j^i$



2. Now we construct Turing machines for applications of (a) combination, (b) composition, (c) primitive recursion and (d) minimization:

(a) combination

Let T. machine  $M_1$ , resp.  $M_2$  computes partial functions  $g_1$ , resp.  $g_2$ . We construct the T. machine M that computes the function  $g_1 \times g_2$ :

M is 3-tape T. machine with following behaviour:

- (i) M starts with copying 1st tape (input) to 2nd and 3rd tape
- (ii) then M simulates the machine  $M_1$  using 2nd tape and the machine  $M_2$  using 3rd tape
- (iii) If  $M_1$  and  $M_2$  terminate normally then copies the content of 2nd and 3rd tape (separated by  $\Delta$ ) to 1st tape and stops.

(b) The composition  $g_1 \circ g_2$  is done by the machine

$$\rightarrow M_1 M_2$$

## (c) primitive recursion

17

Consider the schema of primitive recursion

$$f(\bar{x}, 0) = g(\bar{x})$$

$$f(\bar{x}, y+1) = h(\bar{x}, y, f(\bar{x}, y))$$

where p. functions  $g$ , resp.  $h$  are computed by machines  $M_1$ , resp.  $M_2$ . The function  $f$  can be computed by Turing machine  $M$  which has to do following actions:

- (I) If the last tape symbol is 0, then  $M$  deletes this symbol, goes back to origin of the tape and simulates the machine  $M_1$
- (II) If the last symbol is not 0, then the tape contains sequence:  
 $\Delta \bar{x} \Delta y \Delta \dots \Delta \Delta \Delta$  for some  $y+1 > 0$

Then

- (i) using machines for copying and decrementation (viz exercise)  $M$  transforms the tape to :

$$\Delta \bar{x} \Delta y \Delta \bar{x} \Delta y-1 \Delta \bar{x} \Delta y-2 \Delta \dots \Delta \bar{x} \Delta 0 \Delta \bar{x} \Delta \Delta$$

Then  $M$  shifts the head just beyond 0 and simulates  $M_1$

- (ii) Now the tape will be

$$\Delta \bar{x} \Delta y \Delta x \Delta y-1 \Delta \dots \Delta \bar{x} \Delta 0 \Delta g(\bar{x}) \Delta \Delta \quad \text{e.g. i.e.}$$

$\sim \quad \sim \quad \bar{x} \Delta 0 \Delta f(\bar{x}, 0) \Delta \Delta$

Then  $M$  shifts the head before the last  $\bar{x}$  and simulates  $M_2$

$$\text{The result is } \Delta \bar{x} \Delta y \Delta x \Delta y-1 \Delta \dots \Delta x \Delta 1 \Delta h(\bar{x}, 0, f(\bar{x}, 0)) \Delta \Delta$$

$$\text{i.e. } \sim \quad \sim \quad +(\bar{x}, 1) \Delta \Delta$$

- (iii) Continue with applications of  $M_2$  to the rest of the tape. The last application of  $M_2$  to  $\Delta \bar{x} \Delta y f(\bar{x}, y)$  gives  $\Delta h(\bar{x}, y, f(\bar{x}, y)) \Delta \Delta$  which is the required output  $\Delta f(\bar{x}, y+1) \Delta \Delta \dots$

## (d) minimization

Let us consider a partial function

$$\mu y [g(x, y) = 0]$$

where the p. function  $g$  is computed by a machine  $M_1$ .

Construct 3-tape machine  $M$  with following actions:

1. Write 0 to the tape 2
2. Copy 1st tape and 2nd tape (in sequence) to the 3rd tape
3. Simulate machine  $M_1$  with the 3rd tape
4. If tape 3 is 0, delete the content of tape 1, copy the tape 2 to tape 1 and halt. Otherwise increment the tape 2, erase tape 3 and continue with step 2.

## The partial recursive nature of Turing machines

### Theorem

Any computational process performed by a Turing machine is actually the process of computing a partial recursive function.

### Proof:

Let  $M = (Q, \Sigma, \Gamma, \delta, q_0, q_F)$  be a Turing machine and let  $b = |\Gamma|$ . We will interpret the content of the tape as positive integer with  $b$ -base representation stored in reverse order. For example, if  $\Gamma = \{x, y, \Delta\}$  with coding  $x \approx 1, y \approx 2, \Delta \approx 0$  (always) then the content of the tape  $\Delta y x \Delta \Delta y \Delta \Delta$  is equivalent  $02100200\dots$

and after reversion  $\dots 00200120$  is the number 501. ( $b = 3$ )

With this interpretation we can consider a run of T-machine M as a computation of partial function

$$f: \mathbb{N} \rightarrow \mathbb{N}$$

which maps the numbers corresponding to initial content of the tape to the number corresponding to the content of the tape after M halts.

Let  $M'$  be coded machine M, where the codes of states are:  $0 \approx q_0, 1 \approx q_F$  and other states are coded by integers  $2, 3, \dots, k-1$ , if  $|Q| = k$ .

Because states and symbols are coded by integers, we can represent transition function  $\delta$  by integer functions:

$$\text{mov}(p_i x) = \begin{cases} 2 & \text{if } \delta(p_i x) = (*, R) \\ 1 & \text{if } \delta(p_i x) = (*, L) \\ 0 & \text{otherwise} \end{cases}$$

$$\text{sym}(p_i x) = \begin{cases} Y & \text{if } \delta(p_i x) = (*, Y) \\ X & \text{otherwise} \end{cases}$$

$$\text{state}(p_i x) = \begin{cases} g & \text{if } \delta(p_i x) = (g, *) \\ k & \text{if } p=0 \text{ or } \delta(p_i x) \text{ is not defined} \end{cases}$$

The functions mov, sym and state are tabular functions and hence partial recursive functions.

Now consider a configuration of T-machine as a triple

$$(w, p, n)$$

where  $w$  is content of the tape,  $p$  is the current state, and  $n, n \geq 1$ , is the position of the head.

A symbol under the head can be computed from the configuration by p. recursive function cursym:

$$\text{cursym}(w, p, n) = \text{quo}(w, b^{n-1}) \div \text{mult}(b, \text{quo}(w, b^n))$$

Let it illustrate for  $w = 1120121$ ,  $n = 5$ ,  $b = 3$

$$w = 1120121$$

$$\begin{array}{ccc} & \downarrow \text{divide } b^{n-1} & \\ 112 & \xrightarrow{\div} & 2 \\ 110 & & \\ \uparrow \text{multiple } b & & \\ 11 & & \\ \uparrow \text{divide } b^n & & \\ w = 1120121 & & \end{array}$$

We define another functions:

$$\text{nexthead}(w, p, n) = n \div \text{eq}(\text{mov}(p, \text{cursym}(w, p, n)), 1) + \text{eq}(\text{mov}(p, \text{cursym}(w, p, n)), 2)$$

specifies next position of the head (0 value means erroneous shift)

$$\text{nextstate}(w, p, n) = \text{state}(p, \text{cursym}(w, p, n)) + \text{mult}(k, \neg \text{nexthead}(w, p, n))$$

(If  $\text{nexthead} = 0$  (erroneous shift) then then  $\text{nextstate}$  evaluates to value greater than  $k-1$ , i.e. illegal state.)

and finally

21

$$\text{nexttape}(w, p, n) = (w - \text{mult}(b^n, \text{cursym}(w, p, n))) + \text{mult}(b^n, \text{sym}(p, \text{cursym}(w, p, n)))$$

which evaluates to integer representing the new content of the tape

By combination of previous functions we construct a function step which models one step of Turing machine:

$$\text{step} = \text{nexttape} \times \text{nextstate} \times \text{nexthead}$$

Now we define a function run:  $N^4 \rightarrow N^3$ ;  $\text{run}(w, p, n, t)$  performs  $t$  transitions (steps) from configuration  $(w, p, n)$ :

$$\begin{aligned}\text{run}(w, p, n, 0) &= (w, p, n) \\ \text{run}(w, p, n, t+1) &= \text{step}(\text{run}(w, p, n, t))\end{aligned}$$

A function computed by the machine  $M'$  (for input  $w$ ) is "value" of the tape after the final state (state 0) is reached. The number of required steps is given by a function stoptime:

$$\text{stoptime}(w) = \text{fut} [\Pi_2^3 (\text{run}(w, 1, 1, t) = 0)]$$

We can conclude:

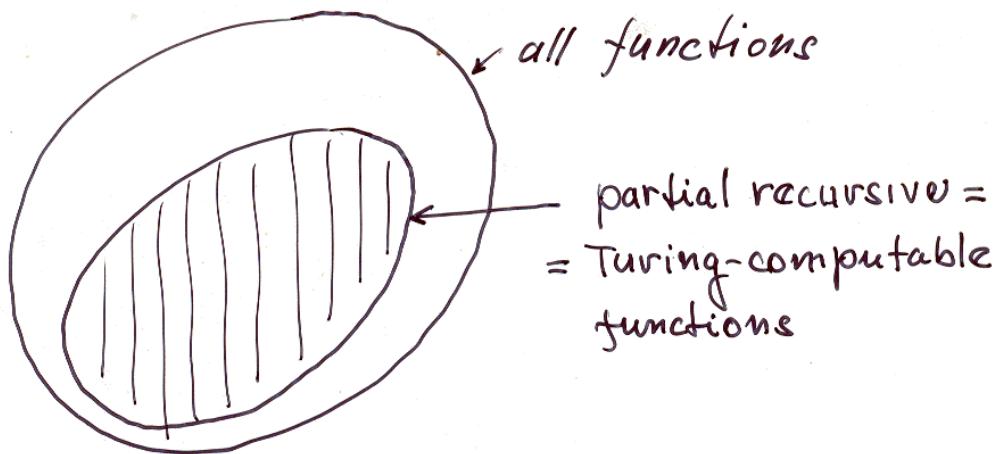
Partial function which represents the machine  $M'$  is the function  $f: N \rightarrow N$

$$f(w) = \Pi_1^3 (\text{run}(w, 1, 1, \text{stoptime}(w)))$$

It is clear from the construction that  $f$  is partial recursive function.

Q.E.D.

To conclude:



## The power of programming languages

### A Bare-Bones Programming Language

We introduce a "minimal" programming language PLBB with the following expression facilities:

Data structures: simple integer variables represented by identifiers (non-negative values)

#### Statements:

- (a) incr name
- (b) decr name
- (c) while name  $\neq$  0 do  
  :  
end

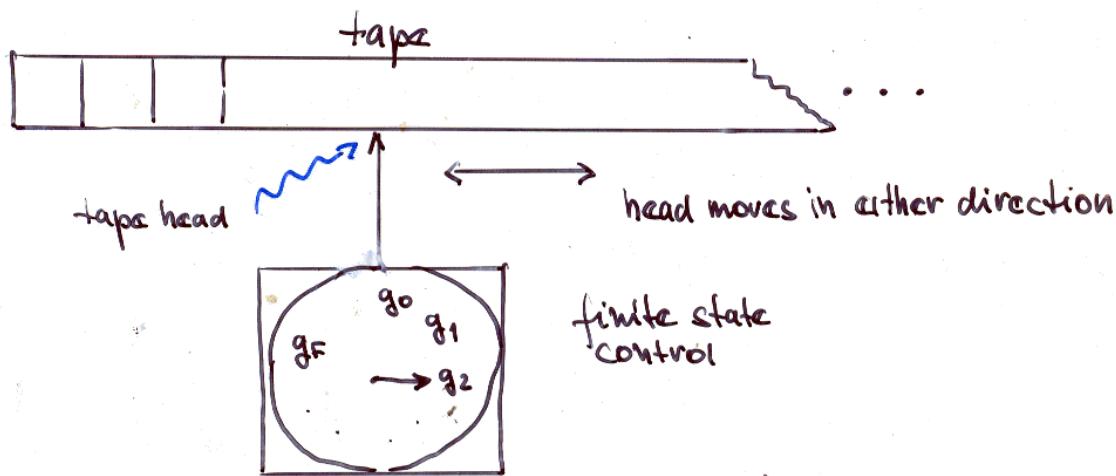
Theorem: Any partial recursive function can be programmed in PLBB and any PLBB program describes some partial recursive functions.

(Proof: homework)

# 7. TURING MACHINES

Alan M. Turing, Emil E. Post - 1936

## 7.1. Basic definition of Turing machine



Turing thesis:

The computational power of Turing machines is as great as any possible computational system.

### Definition 7.1.

Turing machine is the 6-tuple

$$M = (Q, \Sigma, \Gamma, \delta, q_0, q_F), \text{ where}$$

- (1)  $Q$  is a finite set of states
- (2)  $\Sigma$  is a finite set of nonblank symbols, called machine (input) alphabet
- (3)  $\Gamma$  is a finite set of symbols called tape alphabet,  $\Sigma \subseteq \Gamma$
- (4)  $\delta$  is mapping

$$\delta: (Q \setminus \{q_F\}) \times \Gamma \rightarrow Q \times (\Gamma \cup \{L, R\}), \quad L, R \in \Gamma$$

- (5)  $q_0$  is initial state

- (6)  $q_F$  is final state ( $q_0, q_F \in Q$ )

## Comments :

2

$\Delta \in \Gamma$ ,  $\Delta$  denotes the blank symbol

An example of notation for tape contents :  $\Delta x y z \underline{\underline{z}} \Delta x \Delta \Delta \dots$

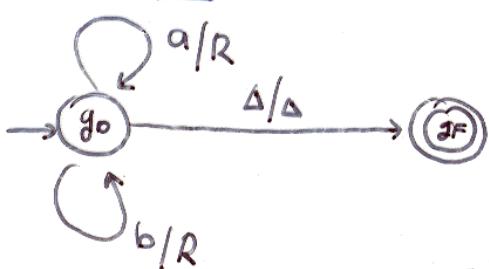
## The semantics of transition function $\delta$ :

	Meaning	Graphical representation
1. $\delta(p_i x) = (q_j y)$ $p_i \in Q, x, y \in \Gamma$	write operation (replace $x$ by $y$ )	
2. $\delta(p_i x) = (q_j L)$	move tape head one cell to the left	
3. $\delta(p_i x) = (q_j R)$	move right	
	initial state	
	final state	

## Termination of computation

- (a) normal termination - moving to the final state
- (b) abnormal termination ↘ for given  $(p_i x)$  is  $\delta(p_i x)$  undefined  
↙ "fall off" the left end of the tape

## Example



T. machine that shifts  
the head to the first  
symbol  $\Delta$  right from the  
current position

For example:

$\Delta a a b \Delta b a a \dots \rightarrow \Delta a a b \underline{\Delta} b a a \dots$

## 4.2 Modular construction of Turing machines

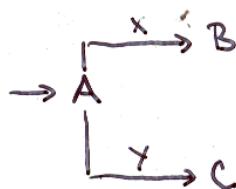
### Combining T. machines

An approach is similar to the case of combining finite automata (your project).

#### Example 7.1.

See Fig. 3.3.

The composite machine shown in Fig. 3.3 could be summarized by the composite diagram:



where

A represents machine that moves its tape head one cell to the right

B - " - that searches for an x

C - " - a y

### Conventions for drawing composite diagrams:

1. The sequence of machines  $\rightarrow A \rightarrow B \rightarrow C$

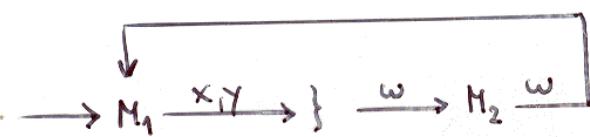
could be represented as  $\rightarrow ABC$

2. Parametrization

$$\underline{x, y, z} \rightarrow \{ \xrightarrow{\omega}$$

$\omega$  has value of that  $x, y, z$  which is current symbol on the tape

E.g.



### 3. Branching

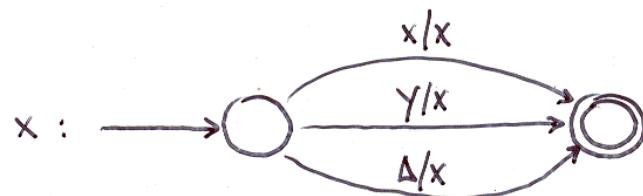
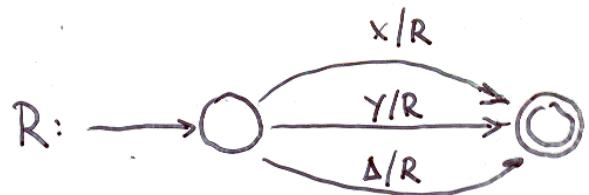
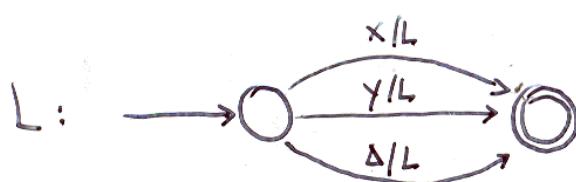


(The difference between  $M_1 \rightarrow M_2$  and  $M_1 \xrightarrow{x} M_2$ )

### Basic buildings blocks

$$\text{Let } \Gamma = \{x, y, \Delta\}$$

#### Machines $L, R, x$

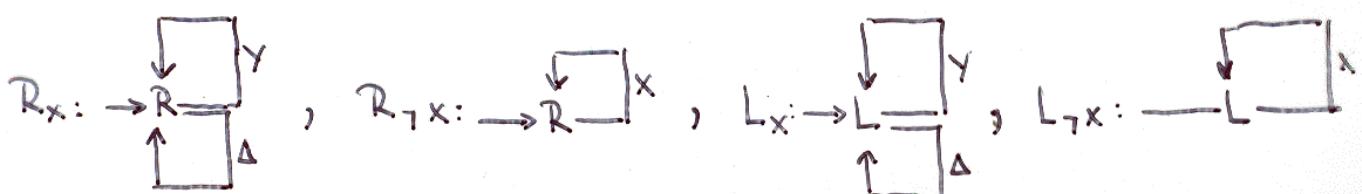


#### An example

$\rightarrow R \rightarrow y \rightarrow L$  or shortly  $\rightarrow RyL$   
transforms the tape

$\Delta x y x y x \Delta \Delta \dots$  to  $\Delta x y x y x y \Delta \dots$

#### Machines $L_x, L_{\bar{x}}, R_x, R_{\bar{x}}$



## Turing machines $S_R$ and $S_L$ for shifts of tape contents

The machine  $S_R$  shifts the string of non blank symbols found at the left of the current cell one cell to the right.

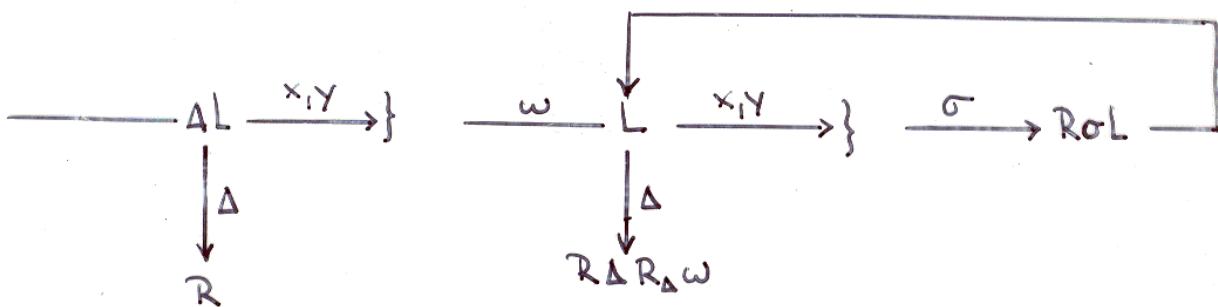
### Examples of activations of $S_R$

$$\begin{array}{lll} \Delta xyy\Delta x\Delta\Delta\dots & \rightarrow & \Delta\Delta xy\Delta x\Delta\Delta\dots \\ \Delta yxy\Delta\Delta xx\Delta\Delta\dots & \rightarrow & \Delta\Delta y\Delta x\Delta xx\Delta\Delta\dots \\ xy\Delta\Delta x\Delta\Delta\dots & \rightarrow & xy\Delta\Delta x\Delta\Delta\dots \text{ - special case} \end{array}$$

The machine  $S_L$  works symmetrically:

$$\Delta\Delta yxx\Delta\Delta\dots \rightarrow \Delta\Delta yxx\Delta\Delta\dots$$

### Realization of $S_R$ :



### Realization of $S_L$ :

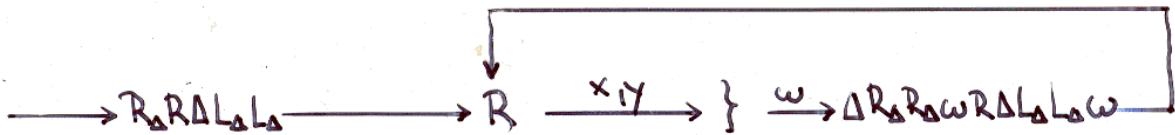
Exercise

## Examples of application

### (a) A copying machine

The machine transforms the tape

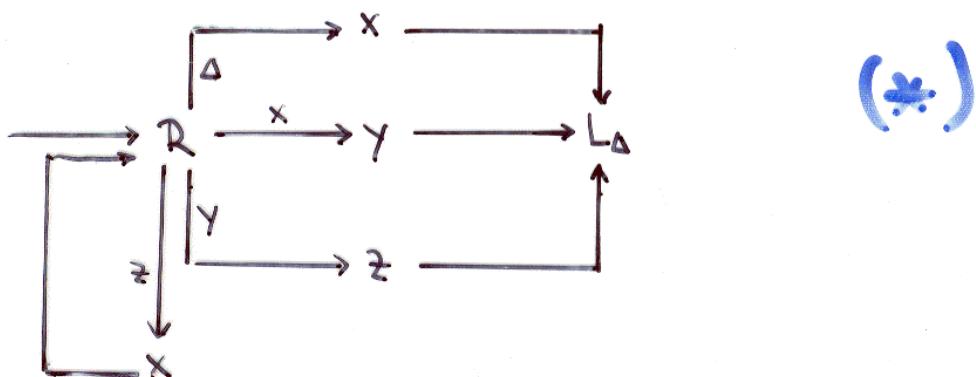
$$\underline{\Delta w \Delta} \text{ to } \underline{\Delta w} \underline{\Delta w \Delta} \quad w \in \{x, y\}^*$$



Example

- |  |   |
|--|---|
| 1. $\underline{\Delta x y \Delta}$ ( $R_R\Delta L\Delta L\Delta$ ) | 5. $\underline{\Delta \Delta y \Delta \Delta}$ ( $\omega R\Delta$ )           |
| 2. $\underline{\Delta x y \Delta \Delta}$ ( $R$ )                  | 6. $\underline{\Delta \Delta y \Delta x \Delta}$ ( $L\Delta L\Delta \omega$ ) |
| 3. $\underline{\Delta x y \Delta \Delta}$ ( $\omega = x, \Delta$ ) | 7. $\underline{\Delta x y \Delta x \Delta}$<br>⋮                              |
| 4. $\underline{\Delta \Delta y \Delta \Delta}$ ( $R_R\Delta$ )     |   |

### (b) Strings generation



The machine produces the next strings in the sequence :

$x, x_1y_1z_1, x_2x_1y_2x_1z_2x_1, x_3y_3, y_4y_3z_4y_2x_2z_2, x_5z_5, y_6z_6, z_7z_6, x_8x_7y_8x_6z_7x_5, y_9x_8x_7y_9, \dots$

## 4

### 4.3. Turing machines as languages acceptors

Definition 7.2.

A string  $w \in \Sigma^*$  is said to be accepted by T. machine

$M = (Q, \Sigma, \Gamma, \delta, q_0, q_f)$  iff  $M$  starting from initial contents of the tape  $\Delta w \Delta \Delta \dots$  reads  $w$  and terminates in  $q_f$ .

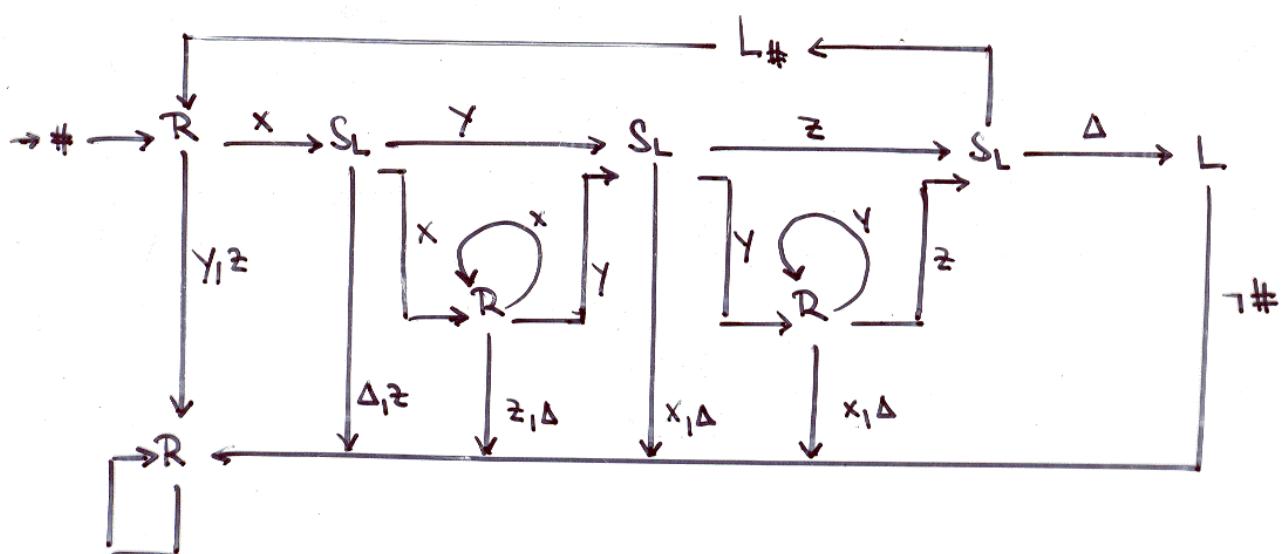
The set

$$L(M) = \{w \mid w \text{ is accepted by } M\}$$

is said to be the language accepted by Turing machine M

Example 4.3.

Let  $T$  be T. machine



Then  $L(T) = \{x^n y^n z^n \mid n \geq 1\}$

$T$  repeats reductions

$$x^n y^n z^n \rightarrow x^{n-1} y^{n-1} z^n \rightarrow x^{n-1} y^{n-1} z^{n-1}$$

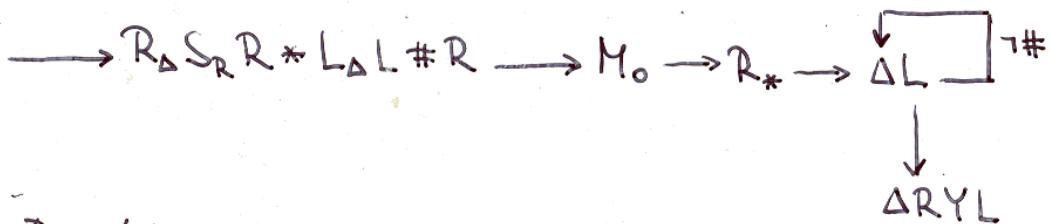
until  $\epsilon$  (empty string) is reached

$\# \in \Gamma$  denotes the left end of the tape

In alternative way of the string acceptation by T. machine:

$$\Delta w \Delta \Delta \rightarrow \Delta Y \Delta \Delta \quad Y \in \Pi \text{ (Yes)}$$

Realization



$M_0$  is T. machine accepting  $L$  by going to the final state

#### 4.4. Variants of Turing machines

##### 4.4.1 Multiple-tape Turing machine

We will consider T. machine with  $k$  tapes and  $k$  corresponding heads. The transition function  $\delta$  can be defined as follows:

$$\delta: (Q \setminus \{q_F\}) \times (\Gamma_1 \times \Gamma_2 \times \dots \times \Gamma_k) \rightarrow Q \times (\Gamma_i \cup \{L, R\}) \times \{1, 2, \dots, k\}$$

##### Theorem 4.1

For each  $k$ -tape Turing machine  $M$  there is a one tape Turing machine  $M'$  such that  $L(M) = L(M')$ .

Proof :

## 7.4.2. Nondeterministic Turing machines

Definition 7.3.

A Turing machine  $M = (Q, \Sigma, \Gamma, \delta, q_0, q_F)$  is said to be a nondeterministic Turing machine if  $\delta$ , the transition function, has the form

$$\delta: (Q \setminus \{q_F\}) \times \Gamma \rightarrow 2^{Q \times (\Gamma \cup \{L, R\})}$$

Theorem 7.2.

For each nondeterministic Turing machine  $M$  there is a deterministic Turing machine  $M'$  such that  $L(M) = L(M')$ .

Proof NTM  $M$  will be simulated by 3-tape deterministic Turing machine  $M'$

Purposes of the tapes:

- Tape 1. input tape - it contains an input string  
 2. working tape  
 3. stores coded sequence of transitions

The activities of  $M'$  proceed as follows:

1. Copy the input string from tape 1 to tape 2
3. Generate the next transition sequence on tape 3
3. Simulate this sequence using tape 2
4. If this simulation leads to a final state in  $M$ , halt.  
 Otherwise erase tape 2 and return to step 1

## 4.5 Turing machines languages and languages of type 0

(also: phrase structured languages or recursively enumerable languages)

We can define a configuration of T. machine. It is given

- (1) by a state of finite control
- (2) by contents of the tape
- (3) by position of the head

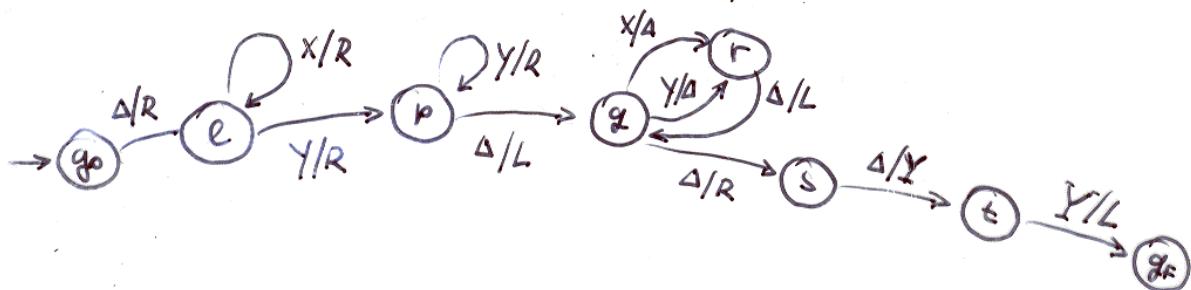
### Convention :

If current state is  $p, p \in Q$  and contents of tape is

$\Delta x y z \Delta$  we will represent the configuration in the form  
 $[\Delta x p y z \Delta]$  or  $[\Delta x p y z \Delta \Delta]$  or ...

### Example #4.

Let us consider the skeleton of Turing machine which accepts the language  $\{x^m y^n \mid m \geq 0, n > 0\}$



The sequence of configurations for accepting  $xx$ :

- |                                       |                              |                                  |
|---------------------------------------|------------------------------|----------------------------------|
| 1. $[\Delta q_0 \Delta x x y \Delta]$ | 6. $[\Delta x x q y \Delta]$ | 10. $[\Delta r \Delta]$          |
| 2. $[\Delta \ell x x y \Delta]$       | 7. $[\Delta x x r \Delta]$   | 11. $[\Delta q \Delta]$          |
| 3. $[\Delta x \ell x y \Delta]$       | 8. $[\Delta x y x \Delta]$   | 12. $[\Delta s \Delta]$          |
| 4. $[\Delta x x \ell y \Delta]$       | 9. $[\Delta x r \Delta]$     | 13. $[\Delta t Y \Delta]$        |
| 5. $[\Delta x x y p \Delta]$          | 10. $[\Delta q x \Delta]$    | 14. $[\Delta f \Delta Y \Delta]$ |

### Theorem 4.3.

Each language accepted by Turing machine is the language of type 0. <sup>M</sup>

Proof. We will construct a grammar  $G = (N, \Sigma, P, S)$  in which there is a derivation for each string accepted by Turing machine such that it corresponds to the reversal of the sequence of configurations accepting the string:

$$(1) \quad N = \{S\} \cup \{\sqbrack, ]\} \cup Q \cup (\Gamma \setminus \Sigma)$$

(2)  $P$  contains productions of the form:

$$(a) \quad S \rightarrow [q_F \Delta Y \Delta]$$

(b)  $\Delta] \rightarrow \Delta \Delta$  (using this production we can extend the string  $[q_F \Delta Y \Delta]$  to required length  $[q_F \Delta Y \Delta \Delta \dots \Delta]$ )

$$(c) \quad qy \rightarrow px \quad \text{if } \delta(p_i x) = (q, y)$$

$$(d) \quad xq \rightarrow px \quad \text{if } \delta(p_i x) = (q, R)$$

$$(e) \quad qyx \rightarrow ypx \quad \text{for each } y \text{ if } \delta(p_i x) = (q, L)$$

$$(f) \quad \left. \begin{array}{l} [q_0 \Delta \rightarrow \epsilon \\ \Delta \Delta] \rightarrow \Delta \\ \Delta] \rightarrow \epsilon \end{array} \right\} \quad \text{eliminate symbols } q_0, \sqbrack, ], \Delta$$

It can be shown that  $L(G) = L(M)$  i.e.

$w \in L(M) \Leftrightarrow$  there is the derivation

$$S \Rightarrow [q_F \Delta Y \Delta] \Rightarrow \dots \Rightarrow [q_0 \Delta w \Delta] \Rightarrow w \Delta \Rightarrow w$$

Theorem 4.4.

Each language of type 0 is the language accepted by Turing machine.

Proof.

Let  $L = L(G)$  be the language generated by the G of type 0. We will construct nondeterministic 2-tape Turing machine N such that  $L(G) = L(N)$ .

Basic action: simulation of application of a production  $\alpha \rightarrow \beta$   
Contents of tapes:

1. tape - input string
2. tape - working tape for construction  
a derivation using substitution  
according to productions  $\alpha \rightarrow \beta$

4.6. The scope of the class of 0 type languagesTheorem 7.5.

For each alphabet  $\Sigma$  there is a language L over  $\Sigma$  such that L is not the language of type 0.

Proof.

1. At first we have to understand that any 0 type language over alphabet  $\Sigma$  can be accepted by T. machine with tape alphabet  $\Sigma \cup \{\Delta\}$ .

Really, if T. machine M works with more symbols than we can code other tape symbols using symbol from  $\Sigma$ .

2. With this assumption we can now list all T. machines 13

with tape symbols in  $\Gamma = (\Sigma \cup \{\Delta\})$ . We start with all 2-state machines, then 3-state, 4-state and so on.

This list can be indexed and hence the set of all such T. machines, i.e. the set of all 0-type languages over  $\Sigma$  is countable.

3. On the other hand we now show, that the set  $2^{\Sigma^*}$  which represents the set of all languages over  $\Sigma$  is uncountable. We use the schema of proof called diagonalisation. (It was used by Cantor to prove that the set  $\mathbb{R}$  is uncountable):

Let us suppose that the set of all languages over  $\Sigma$  is countable. Then, by definition, there is a bijection  $f$

$$f: \mathbb{N} \leftrightarrow 2^{\Sigma^*}$$

Let's try express this bijection in the form of infinite matrix (for  $\Sigma = \{x, y, z\}$ )

	x	y	z	xx	xy	xz	...	xxx	xxz	...
f(1)	a <sub>11</sub>	a <sub>12</sub>	a <sub>13</sub>	...				a <sub>11</sub>	a <sub>11+1</sub>	...
f(2)	a <sub>21</sub>	a <sub>22</sub>	a <sub>23</sub>	...				a <sub>21</sub>	a <sub>21+1</sub>	...
f(3)	a <sub>31</sub>	a <sub>32</sub>	a <sub>33</sub>	...						
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
f(k)	a <sub>k1</sub>	a <sub>k2</sub>			a <sub>kk</sub>					
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮

The language  $L_n = f^{-1}(n)$  is coded by (infinite) binary vector:

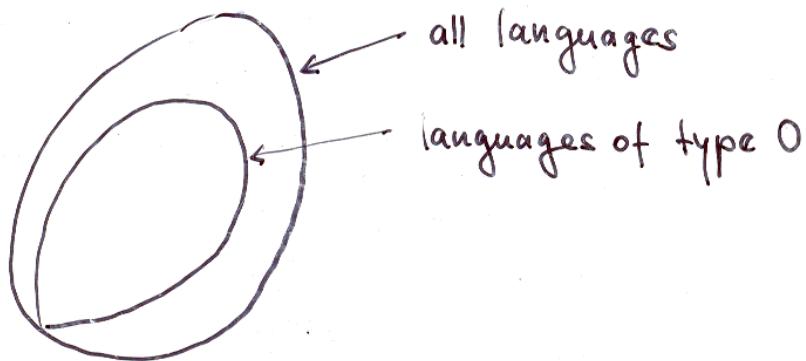
$(a_{n1}, a_{n2}, a_{n3}, \dots)$ , where  $a_{nj} = \begin{cases} 0 & \text{if the } j\text{-th string is not in } L_n \\ 1 & \text{if the } j\text{-th string is in } L_n \end{cases}$

But now consider the language corresponding to the "negation" of diagonal elements:

$$\tilde{L} = (\tilde{a}_1, \tilde{a}_2, \tilde{a}_3, \dots) \text{ where } \tilde{a}_i = \begin{cases} 0 & \text{if } a_{ii} = 1 \\ 1 & \text{if } a_{ii} = 0 \end{cases}$$

The language  $\tilde{L}$  differs from any languages  $f(n)$ ,  $n \in \mathbb{N}$ , but  $\tilde{L} \in 2^{\Sigma^*}$ . It means that  $f$  is not onto mapping (surjective map.) and hence there is not bijection like  $f$  and the cardinality of  $2^{\Sigma^*}$  is greater than cardinality of  $\mathbb{N}$  - the set of all languages over  $\Sigma$  is uncountable.

### Conclusions



### Consequences of Turing's thesis :

1. The languages without grammatical foundation can not be recognized by any algorithmic process
2. If natural intelligence is based on the level more powerful than algorithm execution, then Turing's thesis suggests that the dream of developing truly intelligent computing machines is destined to remain a dream forever.

## 4.7. Beyond phrase-structured languages

### 4.7.1 A coding system for Turing machine

The system includes a coding of

- (1) states
- (2) symbols in  $\Gamma$  and symbols L and R
- (3) transition function  $S$

The result will be the binary string which represents completely given Turing machine.

#### state encoding:

Let  $Q$  is ordered :  $Q = \{q_0, q_F, q_1, p_1, r_1, \dots\}$

the code for the  $j$ -th state is  $0j$

e.g.  $q_0 \approx 0, q_F \approx 00, 5^{\text{th}} \text{ state} \approx 00000$

#### symbol encoding (we suppose the tape symbol in $\Sigma \cup \{\Delta\}$ )

Again we order the alphabet  $\Sigma = \{a_1, a_2, \dots, a_n\}$  and choose the following encoding:

$\Delta \approx \epsilon$  (empty string)

$L \approx 0$

$R \approx 00$

$a_i \approx 0^{i+2} \quad i = 1, 2, \dots, n$

## transition encoding

the transition  $\delta(p, x) = (q_f, y)$ ,  $y = \begin{cases} x & \in \Pi \\ L & \\ R & \end{cases}$

is encoded as 4-tuple

$(p, x, q_f, y)$  by concatenating the codes for  $p, x, q_f, y$  and using 1 as a separator:

"code p" 1 "code x" 1 "code q" 1 "code y"

### Example 4.5

The transition  $\delta(q_0, x) = (q_F, R)$ , the code for  $x$  is 000, will be coded as 01000100100

$$\delta(q_0, \Delta) = (q_F, \Delta) \approx 011001$$

## machine encoding

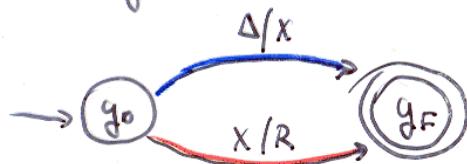
concatenation of all transitions (in systematic way) and separators 1 :

1 "code for 1st transition" 1 "code for 2nd transition" 1 ..

... 1 "code for m-th transition" 1

### Example 4.6.

#### Turing Machine



#### It's code

10110010001010001001001

## An example of non phrase structural languages

The coding system enables mapping of each T. machine to the binary number (not in opposite direction).

Now let us consider the class  $\mathcal{T}$  of all Turing machines over an input alphabet  $\Sigma$  and tape alphabet  $\Sigma \cup \{\Delta\}$  and the mapping

$$\gamma : \mathbb{N} \rightarrow \mathcal{T}$$

such that

$$\gamma(n) = \begin{cases} T & \text{iff } n \text{ is the code of } T \\ \bar{T} & \text{otherwise, } \bar{T} \text{ is the trivial } T \text{ machine.} \end{cases}$$



$\gamma$  is clearly onto mapping.

Now construct another function

$$g : \Sigma^* \rightarrow \mathcal{T} \text{ such that}$$

$$\text{for any } w \in \Sigma^* \quad g(w) = \gamma(\text{len}(w)) \equiv M_w$$

$g$  is again onto mapping.

Because  $w \in \Sigma^*$  and  $\Sigma$  is the tape alphabet of  $M_w$ , there is possible to apply  $M_w$  on string  $w$ . Let's define language  $L_0$ :

$$L_0 = \{w \mid M_w \text{ does not accept } w\}$$

We will show that  $L_0$  can not be accepted by any T. machine.

We will suppose an opposite, i.e. there is  $T_0 \in \mathcal{T}$  such that  $L_0 = L(T_0)$ .

Then because exists  $w_0 \in \Sigma^*$  such that  $L_0 = L(M_{w_0})$ . ( $T_0 \equiv M_{w_0}$ )

Now we are asking if  $w_0 \in L(M_{w_0})$ . Either  $w_0 \in L(M_{w_0})$  or

$w_0 \notin L(M_{w_0})$ . (Tercia non datur). From definition of  $L_0$ :

$$w_0 \in L(M_{w_0}) \Rightarrow w_0 \in L_0 \text{ and } w_0 \notin L(M_{w_0}) \Rightarrow w_0 \in L_0, \text{ since } L_0 = L(M_{w_0})$$

the both implications are contradictory and hence  $L_0 \notin \mathcal{L}_0$ .

## 7.2 Universal Turing machines

18

It is a concept of "programmable" machine enabling in the form of input string to specify concrete Turing machine as well as data for it.

We use the binary coding for the specification of a T. machine. In the same manner we will code also input string representing "data":

10000 1 000 1 000000 1  
  \u2191   \u2191   \u2191  
  \u2191   \u2191   \u2191  
symbol codes

The whole "problem" can be now given by concatenation of machine code and data code.

For example:

10110010001 010001001 1000100010001  
  \u2191   \u2191   \u2191  
  \u2191   \u2191   \u2191  
the code of machine                                  the code of input string

The universal Turing machine which can process this specification can be 3-tape T. machine:

1st tape - programm + data (+ output data)

2nd tape - working tape

3rd tape holds a representation of current state of the machine being simulated

As we know we can construct an equivalent 1 tape machine to this one. Let us denote that machine by  $T_U$ .

## 4.8. Acceptable versus Decidable languages

The ability to accept a language is not symmetric to the ability to accept its complement for T. machines, i.e. Turing languages are not closed under complement. We show it using universal T. machine  $T_U$  and the language  $L_0$ .

We construct T. machine  $M'_w$  which for any string  $w \in \Sigma^*$

1. computes binary number  $\varphi^{-1}(\rho(w))$ , i.e. the code of  $M_w$
2. concatenates the code of  $M_w$  with the code of  $w$

Then we do composition of T. machines  $\rightarrow M'_w \rightarrow T_U$

The resulting machine accepts languages  $L_1$

$$L_1 = \{ w \mid M_w \text{ accepts the string } w \}$$

which is clearly the complement of  $L_0$ .

We see that generally is not possible to construct T. machine  $M$  which writes „Y“ if input string  $w \in L(M)$  and „N“ if  $w \notin L(M)$ .

The languages for which it is not possible are called the Turing acceptable languages (or partially decidable languages).

The languages for which such T. machine exists are called Turing decidable languages.

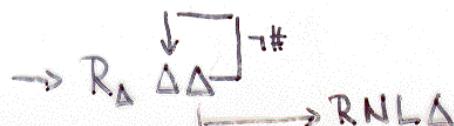
### Examples 4.7.

1.  $L_1$  is acceptable but not decidable
2. CFL's are decidable languages
3.  $L = \{a^n b^n c^n \mid n \in \mathbb{N}\}$  is decidable

In given T. machine for  $L$  the case



can be replaced by



## Terminological note

20

The class of languages accepted by Turing machine is called  
recursively enumerable languages

$\sim \in \sim$  decidable by Turing machines is called  
recursive languages

### 4.9. The halting problem

Now we are going to prove the classical and representative undecidable problem for Turing machine which is called the halting problem.

Let us suppose the binary alphabets of a Turing machines i.e  $I = \{0, 1\}$ ,  $\Sigma = \{0, 1, \Delta\}$  and let us denote the encoded version of M as  $\Gamma(M)$ .  $\Gamma(M)$  is binary string which can be putted on input of M. We are interesting if the machine M for  $\Gamma(M)$  will be halted (M is selfterminating) or not (M is nonselfterminating).

We define the language  $L_F$  over alphabet  $I = \{0, 1\}$  as

$$L_F = \{\Gamma(M) \mid M \text{ is selfterminating}\}$$

The halting problem is stated as a problem of decidability of the language  $L_F$ . This problem is, unfortunately! Undecidable. We can prove it, analogically to decidability of  $L_0$ , by contradiction.

So, let us suppose that there is a Turing machine, denote's it  $M_F$ , which decides the language  $L_F$ . Now let us modify the machine  $M_F$  in such way, that modified machine  $M'_F$  outputs 0, resp. 1 iff  $M_F$  gives a message N, resp. Y. The machine  $M'_F$  has hence tape alphabet  $\{0, 1, \Delta\}$  and can be used to construction of T. machine  $M_0$ .

$$M_0 : \xrightarrow{\quad} M'_F \xrightarrow{\quad} R \xrightarrow{1} \boxed{R}$$

We will examine if the machine  $M_0$  is selfterminating. If  $L_F$  is decidable, then the answer is either "yes" or "no".

- (a) Let us suppose that  $M_0$  is selfterminating. Then for  $\sigma(M_0)$  the machine  $M'_F$  halts with output 1 and  $M_0$  goes along the arc  $R \xrightarrow{1} R$  and cycles forever.

Hence

$M_0$  is "selfterminating"  $\Rightarrow M_0$  is "nonselfterminating"

- (b) Let us suppose that  $M_0$  is nonselfterminating. Then for input  $\sigma(M_0)$  the machine  $M'_F$  halts with output 0 and the machine  $M_0$  halts. Hence

$M_0$  is nonselfterminating  $\Rightarrow M_0$  is selfterminating

From inconsistence of the both implications we conclude that the assumption of decidability of  $L_F$  was false. So the halting problem is undecidable.

## 7.10. Post's correspondence problem

### Definition 7.

The set  $S$  of ordered pairs of strings over alphabet  $\Sigma$

$$S = \{(\alpha_1, \beta_1), (\alpha_2, \beta_2), \dots, (\alpha_k, \beta_k)\}, k \geq 1, \alpha_i, \beta_i \in \Sigma^+$$

is called Post system over alphabet  $\Sigma$ .

The solution of Post's system is each nonempty sequence of natural numbers  $i_1, i_2, \dots, i_m$ ,  $i_j \leq k$ , such that

$$\alpha_{i_1} \alpha_{i_2} \dots \alpha_{i_m} = \beta_{i_1} \beta_{i_2} \dots \beta_{i_m}$$

### The Post's correspondence problem :

Is there a solution for given (any) Post's system?

### Theorem 7.

Post's correspondence problem is undecidable.

Proof: It can be shown that Post's correspondence problem is equivalent to the halting problem.

### Example 7.7.

- Let  $S = \{(b, bbb), (babbb, ba), (ba, a)\}$  be the Post system over  $\{a, b\}$ .

The system has a solution  $i_1=2, i_2=1, i_3=1, i_4=3$   
(so  $m=4$ )

$$\begin{array}{ccccccccc} \underline{babbb}, & \underline{b}, & \underline{b}, & \underline{ba}, & = & \underline{ba}, & \underline{bbb}, & \underline{bbb}, & \underline{a}, \\ \alpha_2, & \alpha_1, & \alpha_1, & \alpha_3, & & \beta_2, & \beta_1, & \beta_1, & \beta_3 \end{array}$$

2. On the other hand the Post system

23

$$S = \{(ab, abb), (a, ba), (b, bb)\}$$

has no solution, because  $|x_i| < |y_i|$  for  $i = 1, 2, 3$ .

The Post correspondence problem is often used for a proof of undecidability of some important problems:

1. The problem of equivalence of context-free grammars
2. The problem of non-emptiness of a language generated by context-sensitive grammar
3. The problem of membership of  $w$  in  $L(G)$  where  $G$  is grammar of type 0

## 8. COMPLEXITY

The study of computational processes has led to the classification of problems into two broad categories:

- the solvable problems
- the unsolvable problems

Now we will deal with the solvable problems. We concentrate our attention to questions "how problems are difficult to solve with respect of required time or space". This aspect is called complexity of problems.

Two basic types of complexity are distinguished:

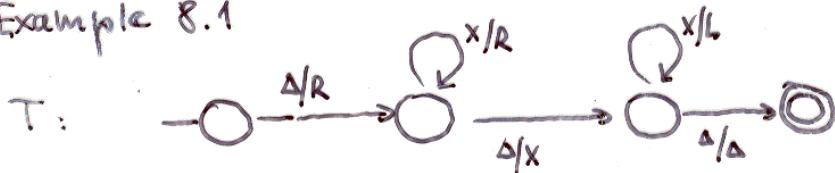
- time complexity
- space complexity

### 8.1. Complexity of Turing Machine Computations

Time complexity - the number of steps executed during computation

Space complexity - the number of tape cells required by the computation

Example 8.1



for input  $\Delta \text{xxx} \Delta \Delta$ : time complexity of computation is 9  
space - " - is 5

Simple implication:

If the time complexity of computation of  $T$  is  $n$ , then space complexity of computation of  $T$  is not greater than  $n+1$ .

## 8.2. Complexity of algorithms and problems

The theoretical approach is based on Turing thesis :

Each algorithm is implementable by some Turing machine

Then the complexity of algorithm is done by analysis of the corresponding Turing machine. The aim is to express (quantified) required resources (time and space) as a function of the length of input strings (input data)

Usually we distinguish

- (1) worst-case analysis
- (2) best-case analysis
- (3) average-case analysis

### Complexity of problems:

The complexity of problem is the complexity of the best (known or proved) algorithm of its solution

#### Example:

Find the complexity of string comparison problem

#### Alternative approach:

Rates (orders) of growth :

$O(f(n))$	- asymptotical	upper bound
$\Theta(f(n))$	- asymptotical	tight bound
$\Omega(f(n))$	- asymptotical	lower bound

## 8.3. Time complexity of language recognition problems

3

### Definition 8.1.

Let  $M$  be (single or multitape) Turing machine which computes the partial function  $f: \Sigma_1^* \rightarrow \Sigma_2^*$ . We say that  $M$  computes the function in polynomial time if there is a polynomial  $p(x)$  such that for each  $w \in \Sigma_1^*$ , for which  $f(w)$  is defined,  $M$  computes  $f(w)$  in no more than  $p(|w|)$  steps. The function  $f$  is called the function computable in polynomial time.

### Theorem 8.1.

Let  $f_1$  and  $f_2$  are the functions computable in polynomial time. Then composition  $f_2 \circ f_1$  is also computable in polynomial time

### Proof:

There is a Turing machine  $M_1$  that computes  $f_1(w)$  in no more than  $p_1(|w|)$  steps and a Turing machine  $M_2$  that computes  $f_2(w)$  in no more than  $p_2(|w|)$  steps.  $p_1(x)$  a  $p_2(x)$  are polynomials. The Turing machine  $\rightarrow M_1 M_2$  which computes  $f_2 \circ f_1(w)$  has for any  $w$  the following complexity:

$$p_1(|w|) + p_2(p_1(|w|)+1), \text{ where } .$$

$p_1(|w|)$  is the complexity of computation  $f_1(w)$

$p_1(|w|)+1$  is maximal length of output of  $M_2$

$p_2(p_1(|w|)+1)$  is the complexity of computation of  $f_2$  applied on  $f_1(w)$

Resulting complexity is obviously polynomial.

### Theorem 8.2.

The class of functions which are computable by multi-tape Turing machines in polynomial time is the same as the class of functions computable by single-tape T. machines in polynomial time.

Proof:

### Definition 8.2.

We say that Turing machine  $M$  accepts the language  $L$  in polynomial time if  $L = L(M)$  and there is a polynomial  $p(n)$  such that the number of steps required to accept any  $w \in L(M)$  is no greater than  $p(|w|)$ . We define P to be the class of languages that can be accepted by Turing machines in polynomial time.

### Theorem 8.3.

If a T. machine  $M$  accepts the language  $L$  in polynomial time, then there is another T. machine  $M'$  that also accepts  $L$  in polynomial time but indicates acceptance by halting with tape configuration  $\sqcup Y \Delta \Delta \dots$

Proof:

### Theorem 8.4.

Let  $L$  be a language accepted by multi-tape Turing machine in polynomial time. Then  $L$  belongs to class P.

Proof: We define the function  $f$

$$f(w) = \begin{cases} Y & \text{if } w \in L \\ \text{is not defined} & \text{if } w \notin L \end{cases}$$

and apply Theorem 8.2.

Conclusion: Robustness of the class P

## 8.4. Polynomial-time decidable languages

Definition of polynomial-time decidable languages is obvious.

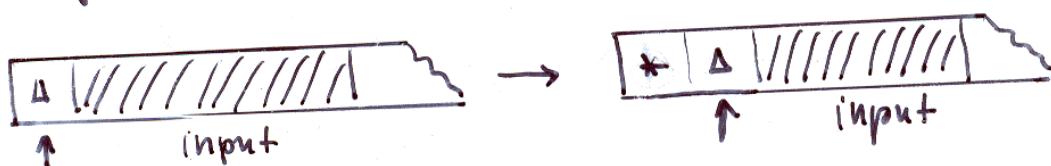
Q: What is relation between polyn.-time acceptable and poly-n.-time decidable languages?

### Theorem 8.5.

If  $L$  is polynomial time acceptable language then  $L$  is also polynomial time decidable language.

Proof. Let  $M$  be T-machine that accepts  $L$  without abnormal halting.

If it is a danger of abnormal halting, we transform input tape in this way



According assumption there is a polynomial  $p(n)$ , which bounds the number of steps of  $M$  accepting  $w \in L(M)$  by value  $p(|w|)$ .

We will construct 2-tape T-machine  $M'$  which decides  $L$  in polynomial time.  $M'$  is a composition of machines  $M_1$  and  $M_2$ .

1.  $M_1$  makes copy of input  $w$  and its substitution of string  $p(|w|)$  on 2nd tape



It can be shown that  $M_1$  runs in polyn. time.

2.  $M_2$  simulates  $M$  with some modifications:

(a) each step of  $M$  implies shift on 2nd tape 1 cell right

(b) reaching final state of  $M$  implies output  $\Delta Y 00$

(b) -  $w$  - blank on 2nd tape -  $w$  -  $\Delta N 00$

## 8.5. Time complexity of nondeterministic machines

Definition 8.3.

We say that a nondeterministic T. machine accepts the language L in polynomial time provided  $L = L(M)$  and there is a polynomial  $p(x)$  such that for any  $w \in L$ , M can accept w by a computation involving no more than  $p(|w|)$  steps. Furthermore, we define NP to be the class of languages that can <sup>be</sup> accepted by nondeterministic T. machines in polynomial time.

We can immediately claim that

$$P \subseteq NP$$

The question

$$P \stackrel{?}{=} NP$$

is one of the most known open problems. It is believed that  $P \subset NP$ .

Summary:

Polynomial time deterministic computations:

$$\begin{array}{c} \text{solvble decision} \\ \text{problems} \end{array} = \begin{array}{c} \text{decidable} \\ \text{languages} \end{array} = \begin{array}{c} \text{acceptable} \\ \text{languages} \end{array}$$

Polynomial time nondeterministic computations:

$$\begin{array}{c} \text{solvble decision} \\ \text{problems} \end{array} = \begin{array}{c} \text{decidable} \\ \text{languages} \end{array} \stackrel{?}{=} \begin{array}{c} \text{acceptable} \\ \text{languages} \end{array}$$

## 8.6. Polynomial reduction (also polynomial transformation)

### Definition 8.4

A polynomial reduction from a language  $L_1$  over  $\Sigma_1$ , to another language  $L_2$  over  $\Sigma_2$  is a function  $f$  such that

$$(1) \forall w \in \Sigma_1^*: w \in L_1 \Leftrightarrow f(w) \in L_2$$

(2)  $f$  can be computed by T.machine in polynomial time

If there is a polynomial reduction from  $L_1$  to  $L_2$ , we say that  $L_1$  reduces to  $L_2$ , and write  $L_1 \leq L_2$ .

### Theorem 8.6.

If  $L_1 \leq L_2$  and  $L_2$  is in P, then  $L_1$  is in P.

#### Proof

Let  $M_f$  be T.machine which computes  $f$  and let  $p(x)$  is the time complexity of  $M_f$ . For any  $w \in L_1$ , the computation  $f(w)$  requires no more than  $p(|w|)$  steps and gives the output of the length less or equal  $p(|w|) + |w|$ .

Let  $M_2$  accepts  $L_2$  with complexity  $q(x)$ . Now consider the composition  $\rightarrow M_f M_2$ . This machine accepts  $L_1$ ; for any  $w \in L_1$ , the machine  $\rightarrow M_f M_2$  makes maximally  $p(|w|) + q(p(|w|) + |w|)$  steps. Hence  $\rightarrow M_f M_2$  has polynomial complexity and  $L_1 \in P$ .

### Example 8.

The function  $f: \{x, y\}^* \rightarrow \{x, y, z\}^*$  defined by  $f(v) = v z v$  is a polynomial reduction from

$$L_1 = \{w \mid w \text{ is palindrome in } \{x, y\}^*\}$$

to

$$L_2 = \{w z w^R \mid w \in \{x, y\}^*\}$$

Theorem 8.6. can be used practically to show that given language  $L$  is in  $P$  (if we know that another language  $L' \in P$ ). Moreover, by "inverse" of Theorem 8.6:

If  $L_1 \in L_2$  and  $L_1 \notin P$  then  $L_2 \notin P$

We can show non-belonging to  $P$ .

### 8.4. Cook's theorem

Cook's theorem identifies a language (a decision problem) in the class NP to which any other language in NP can be reduced by some polynomial reduction:

SAT problem - the problem of satisfiability

Let  $V = \{v_1, v_2, \dots, v_n\}$  be a finite set of Boolean variables (propositions). We suppose truth assignment:  $V \rightarrow \{\text{true}, \text{false}\}$ . Each  $v_i$  or  $\bar{v}_i$  (negation of  $v_i$ ) is called literal.

We define a clause over  $V$  to be formula containing literals connected by  $\vee$  (or).

Examples of clauses

$$v_1 \vee \bar{v}_2$$

$$v_2 \vee v_3$$

$$\bar{v}_1 \vee \bar{v}_3 \vee v_2$$

SAT problems is formulated as follows:

Given a finite set  $V$  of variables and a collection of clauses over  $V$ , is there a truth assignment that satisfies the clauses?

A given SAT problem can be coded by a single string : 9

Let  $V = \{v_1, v_2, \dots, v_n\}$ . Each literal  $v_i$  is coded by the string of 0's, but at  $i$ th position is p if coded literal is  $v_i$  or n if coded literal is  $\bar{v}_i$ . Each clause is coded as a list of literals separated by /. The SAT-problem is a list of clauses in parenthesis :

### Example

The SAT problem over  $V = \{v_1, v_2, v_3\}$  with clauses

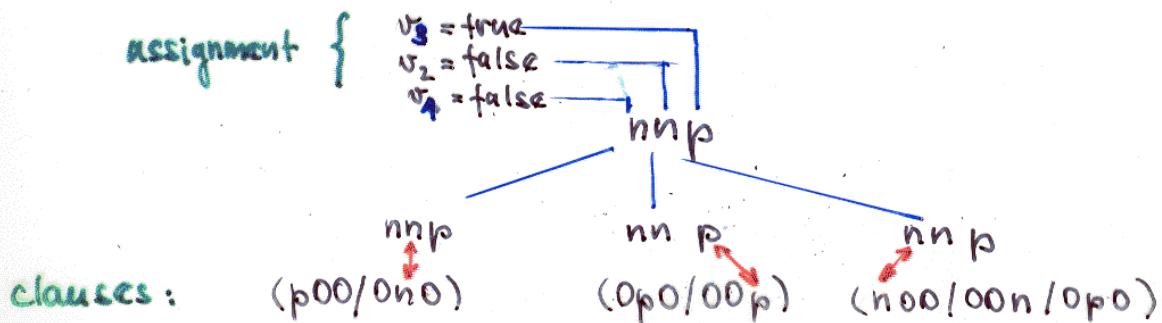
$$v_1 \vee \bar{v}_2 ; v_2 \vee v_3 ; \bar{v}_1 \vee \bar{v}_3 \vee v_2$$

is represented by the string

$$(p00/0n0)(0p0/00p)(n00/00n/0p0)$$

Let  $L_{SAT}$  be a language of such strings which represent satisfiable sets of clauses. For example, the above string is in  $L_{SAT}$  ( $v_1 = F, v_2 = F, v_3 = T$ ). On the other hand  $(p00/0p0)(n00/0p0)(p00/0n0)(n00/0n0)$  is not in  $L_{SAT}$ .

Let us now represent truth assignments by strings over  $\{p, n\}$  where p in  $i$ th position indicates that  $v_i$  is true and n in  $j$ -th position indicates that  $v_j$  is false. Then test, if for given truth assignment is satisfiable for given set of clauses is simple :



Following this principle of testing (and cooling) truth assignments we can construct 2-tape T. machine which is nondeterministic and accepts LSAT in polynomial time. The machine will

1. start computation by checking that its input is a string representing a collection of clauses
2. write, in nondeterministic manner, a string of p's and n's of length equal to the number of variables in the clauses on its second tape
3. move the tape head on the first tape across the input string while testing to see that the truth assignment on its 2nd tape satisfies the clauses being scanned

This process can easily be implemented with polynomial complexity. Hence  $LSAT \in NP$ .

### Theorem 8.7. (Cook's theorem)

If  $L$  is a language in NP, then  $L \leq L_{SAT}$ .

#### Proof:

Because  $L \in NP$ , there is nondeterministic T. machine  $M$  and polynomial  $p(x)$  such that for any  $w \in L$  machine  $M$  accepts  $w$  in no more than  $p(|w|)$  steps. The heart of the proof is a polynomial reduction  $f$  from  $L$  to  $L_{SAT}$ :

for any  $w \in L$ , the value  $f(w)$  will be the set of clauses which are satisfiable if and only if  $M$  accepts  $w$

## NP-complete languages

Since the discovery of Cook's theorem, many other languages in NP have been found to possess the same pivotal properties as L-SAT.  
 (To be polynomial reductions of any other languages in NP)  
 These languages are said to be NP-complete (~~see appendix~~<sup>NPC</sup>)

If any one of these languages can be accepted by deterministic T-machine in polynomial time, then NP must be equal P.

Conversely, if any language in NP can be shown to lie outside of, then all the NP-complete languages must also lie outside of P and search for efficient solution to these problems is useless.

## A Sampling of NP Problems

In fact, all the listed problems here are NPC problems.

## Formal language and automata theory

1. Two nondeterministic finite automaton over the same alphabet accept different languages.
2. Two regular expressions over the same alphabet represent different languages
3. Given a finite collection of DFA, respond Y if and only if there is a string that is accepted by each of the automaton in the collection.

## Number theory

1. Given three positive integers  $a, b$ , and  $c$ , respond Y if and only if the equation

$$ax^2 + by = c$$

has a solution that consist of positive integers

2. Given three positive integer  $a, b$ , and  $c$ , respond Y if and only if there is positive integer  $x$  that is less than  $c$  and  $x^2 \equiv a \pmod{b}$
3. Given a finite collection  $K$  of positive integer and another posive integer  $n$ , respond Y iff there is a subset of  $K$  whose sum is  $n$ .

## Logic

1. Given a collection of clauses in which each clause contains exactly three literals, respond Y iff there is a true assignment that satisfied the entire collection. This problem is known as 3SAT.
2. Given an expression consist of variables connected by symbols  $\neg, \vee, \wedge, \rightarrow$ , respond Y iff there is a model of the expression. (An assignment that evaluates the expression to true value).

## Scheduling problems

- Given a set  $T$  of tasks, the amount of time required to complete each individual task, a positive integer  $n$ , and a deadline, respond  $Y$  iff there is a way of assigning the tasks to  $n$  processors so that entire set of tasks will be completed by the deadline.

## Graph theory

Definition

Let  $\vec{G} = (E, V, \sigma)$  be an oriented graph.

A feedback vertex set is a subset  $S$ ,  $S \subseteq V$  such that any cycle of  $\vec{G}$  consist of some vertex from  $S$ .

A feedback edge set is a subset  $F$ ,  $F \subseteq E$  such that any cycle of  $\vec{G}$  consist of some edge from  $F$

Following are the NPC problems :

1. Is there a clique with  $k$  vertexes ( $k$ -clique) of given (nonoriented) graph  $G$ ?

2. (Vertex Cover) Is there a dominating set of  $k$  vertexes?

3. (Hamilton circuit) Is there a Hamilton circuit in graph  $G$ ?

4. (Colorability) Is a graph  $G$   $k$ -chromatic?

5. (Feedback vertex set) Is there a feedback vertex set with  $k$  vertices in  $\vec{G}$ ?

6. (Feedback edge set) Is there a feedback edge set with  $k$  edges in graph  $\vec{G}$ ?

7. (Directed Hamilton circuit) Is there a Hamilton cycle in oriented graph  $\vec{G}$ ?

## Set theory

### 1. (Set cover)

Let  $S_1, S_2, \dots, S_n$  is a collection of sets. Is there a subset with  $k$  sets  $S_{i_1}, S_{i_2}, \dots, S_{i_k}$  of the collection such that

$$\bigcup_{j=1}^k S_{i_j} = \bigcup_{j=1}^n S_j$$

### 2. (Exact cover)

Let  $S_1, S_2, \dots, S_n$  is a collection of sets. Is there set cover which consist of subset of pair-wise disjoint sets of the collection?

## Another approaches to the complexity analysis

### Rates (Orders) of Growth

#### Definition

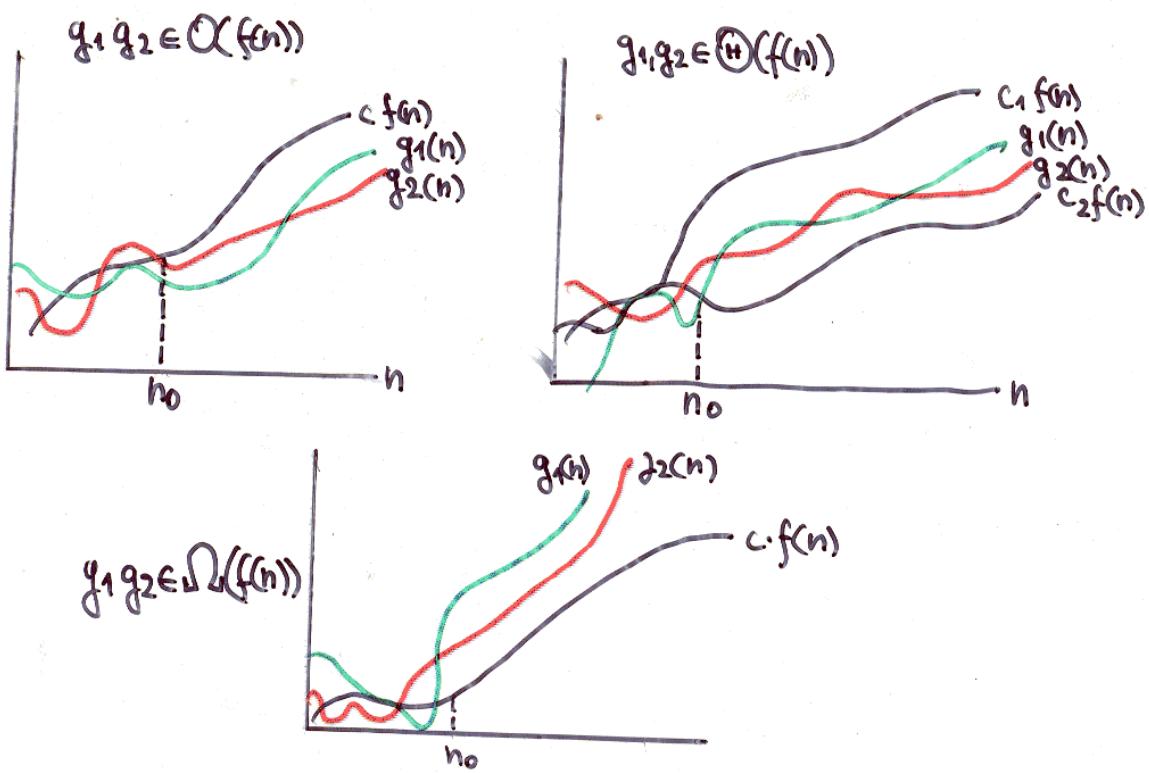
Let  $\mathcal{T}$  be a set of functions from  $\mathbb{N}$  to  $\mathbb{N}$ . For given function  $f$  from  $\mathcal{T}$  we define set of functions  $O(f)$ ,  $\Theta(f)$ ,  $\Omega(f)$  called asymptotically upper bound, asymptotically tight bound and asymptotically lower bound respectively in the following way :

$$O(f(n)) = \{g(n) \mid \text{there are positive constants } c \text{ and } n_0 \text{ such that } 0 \leq g(n) \leq c \cdot f(n) \text{ for all } n \geq n_0\}$$

$$\Theta(f(n)) = \{g(n) \mid g(n) \in \mathcal{T} \text{ and there are positive constants } c_1, c_2, \text{ and } n_0 \text{ such that } 0 \leq c_1 f(n) \leq g(n) \leq c_2 f(n) \text{ for all } n \geq n_0\}$$

$$\Omega(f(n)) = \{g(n) \mid g(n) \in \mathcal{T} \text{ and there are positive constants } c \text{ and } n_0 \text{ such that } 0 \leq c \cdot f(n) \leq g(n) \text{ for all } n \geq n_0\}$$

The following figures illustrate the definitions.



### Theorem

For any two functions  $f(n)$  and  $g(n)$  from  $\mathbb{N}$  the function  $f(n)$  is in  $\Theta(g(n))$  if  $f(n) \in O(g(n))$  and  $f(n) \in \Omega(g(n))$ .

Proof : Follows from definition.

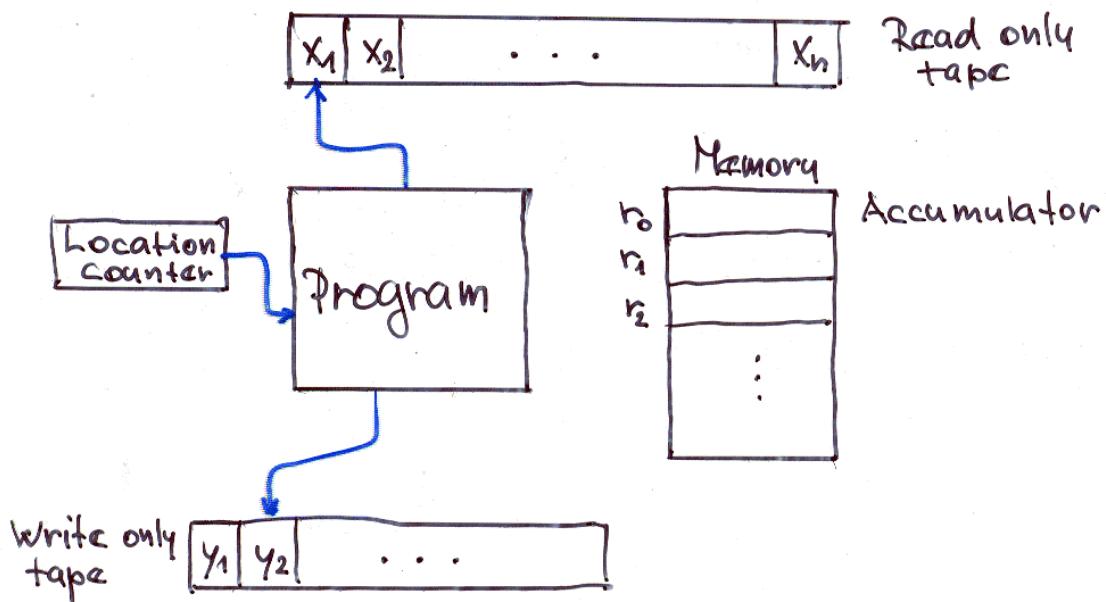
### Theorem :

Let  $p(n)$  is a polynomial with degree  $d$ . Then  
 $|p(n)| \in \Theta(n^d)$

Because  $\Theta(n^2) \subseteq \Theta(n^3)$  for example, we will write  $\Theta(n^2) \leq \Theta(n^3)$  and we are considering a problem with complexity  $\Theta(n^2)$  as less complex than problem with complexity (rate of growth)  $\Theta(n^3)$

## RAM and RASP models

### 1. Random access machine - RAM model



#### Instructions

LOAD	READ
STORE	WRITE
ADD	JUMP
SUB	JGTZ
MULT	JZERO
DIV	HALT

#### Operands

integers (same as  $x_i, y_i$ )

#### Addressing modes

- (a) literal (constant) = 5
- (b) direct address 5
- (c) indirect address \* 5

### 2. Random access stored program machine - RASP model

The same as RAM model but the program is stored in Memory, hence there is no addressing of type (c) - the program can be modified itself.

Theorem: For any RAM program with time complexity  $T(n)$  there is a constant  $k$  such that time complexity of equivalent RASP program is  $k \cdot T(n)$ .