

# Introduction to Computability

13/12/2004  
7/12/2005  
15/10/2007  
2.1

Question: Using a counting argument, we have seen that there are functions that cannot be computed (or, in other words, problems that cannot be solved by any algorithm).

How can we exhibit a specific problem of this form?

Solution: we need a formal definition of algorithm

Let us start with something we know: Java

Let's show that there is no Java program that solves a specific problem?

Hello - World problem:

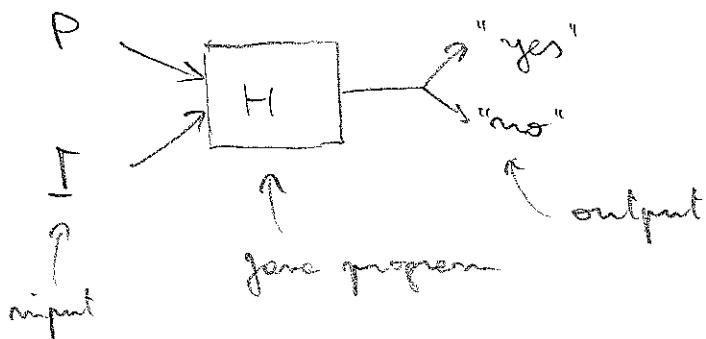
Your first Java program HW:

```
public class HW {  
    public static void main (String [] args) {  
        System.out.println ("Hello, world");  
    }  
}
```

The first 12 characters output by HW are "Hello, world".

Hello-world problem (HWP): given an arbitrary Java program P and an input I for P, does P(I) print "Hello, world" as its first 12 characters?

Consider a solution to HWP:



Does such a program H exist?

- we could see P for println statements
- but, how do we know whether they are executed?

To give you an idea how difficult this can become, consider Fermat's last theorem:

The equation  $x^n + y^n = z^n$  has no integer solution for  $n \geq 3$ .

For  $n=2$ : a solution is  $x=3, y=4, z=5$

For  $n \geq 3$ : mathematicians have believed that the theorem is true, but no proof was found until recently (proof given by Wiles is very complex, and still under verification)

Consider a simple Java program  $P_1$ , that:

- 1) needs input n
- 2) for all possible  $x, y, z$  do  
if  $(x^n + y^n = z^n)$   
println ("Hello, world");

Consider input  $n=3$ :  $P_1$  prints "Hello, world" only if F.L.T. is false, otherwise  $P_1$  loops forever.

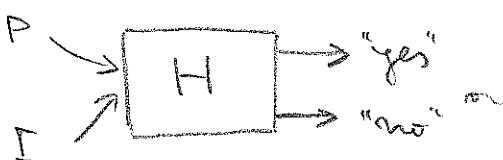
⇒ If we could solve HWP, we would also have proved or disproved F.L.T.

This would be too nice !! Where is the problem ?

Theorem: There is no Java program  $H$  that decides HWP.

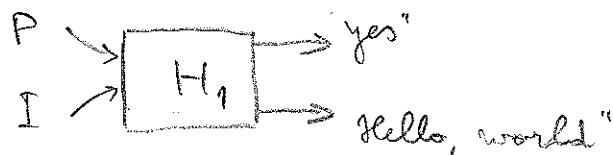
Proof: assume  $H$  exists and derive a contradiction.

Consider  $H$ :



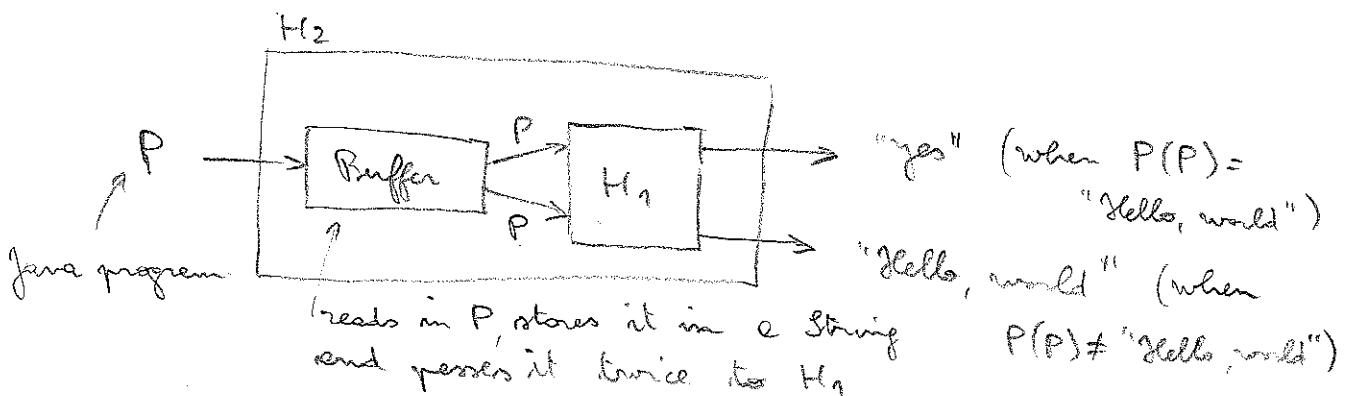
We modify  $H$  to  $H_1$ , s.t.  $H_1$  prints "Hello, world"

instead of "no"



(Note: we have to modify the println statements in  $H$ )

We modify  $H_1$  to  $H_2$ , which takes only input  $P$  and feeds it to  $H_1$  as both  $P$  and  $I$ :



Let us consider  $H_2(P)$  when  $P = H_2$ .

suppose  $H_2(H_2) = \text{"yes"} \Rightarrow P(P) = \text{"Hello, world"}$

suppose  $H_2(H_2) = \text{"Hello, world"} \Rightarrow P(P) \neq \text{"Hello, world"}$

But  $P = H_2 \Rightarrow$  contradiction  $\Rightarrow H_1, H_2, H_2$  cannot exist!

Q.e.d.

We have shown HWP to be undecidable,

27/11/2006, 2.4

i.e., there cannot be an algorithm (or a program) that solves it.

We can show that other problems are undecidable by "reducing" HWP to them

### Reductions

foo-problem: given a program  $R$  and its input  $z$ , does  $R$  even call a function named `foo` while executing on input  $z$ .

Idea: we reduce the HWP to the foo-problem, i.e. we show that if it's possible to solve the foo-problem on  $(R, z)$ , then we can solve HWP on  $(Q, y)$ , for any program  $Q$  with input  $y$ .

Since HWP is undecidable, so is the foo-problem.

Suppose there is a program  $F$  that takes as input  $(R, z)$  and decides the foo-problem for  $(R, z)$ .

We show how  $F$  can be used to construct  $H$  that decides HWP on input  $(Q, y)$

Idea: apply modifications to  $Q$

2.5

- 1) remove function  $\text{foo}$  in  $Q$  (if present) to  $Q_1$   
 $\Rightarrow Q_1$
- 2) add a dummy function ' $\text{foo}$ ' to  $Q_1 \Rightarrow Q_2$
- 3) modify  $Q_2$  to store all its output in some array  $A$   
 $\Rightarrow Q_3$
- 4) modify  $Q_3$  so that after every println statement  
it checks array  $A$  to see if "Hello, world" has been  
printed. If yes, then call function  $\text{foo} \Rightarrow Q_4$

Note: We can write a Java program that takes as input a  
Java sourcefile and modifies it as specified above.

Let  $R = Q_4$  and  $z = y$

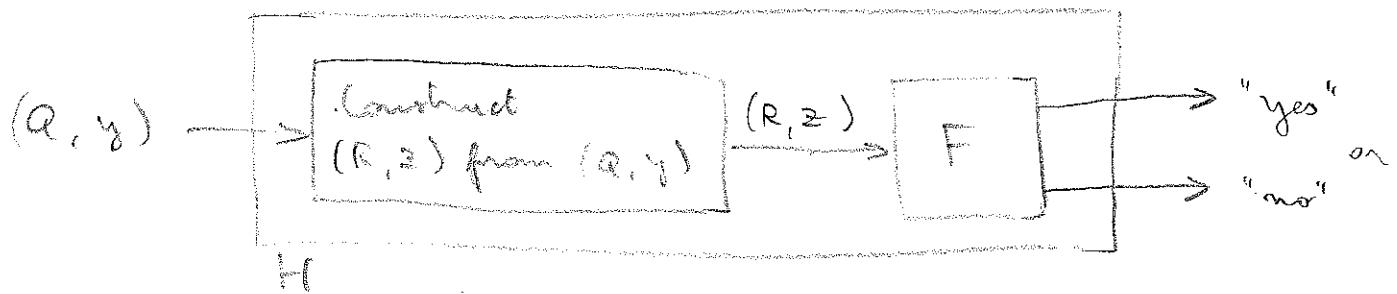
We have by construction:

$Q(y)$  prints "Hello, world"  $\Leftarrow$

$R(z)$  calls function  $\text{foo}$ .

Hence, we can use  $F$  that solves  $\text{foo}$ -problem on  $R(z)$   
to construct  $H$  that solves HWP on  $Q(y)$ .

Schematically:



But since  $H$  does not exist, also  $F$  cannot exist.

Q.e.d

## Showing undecidability by reduction from undecidable problem

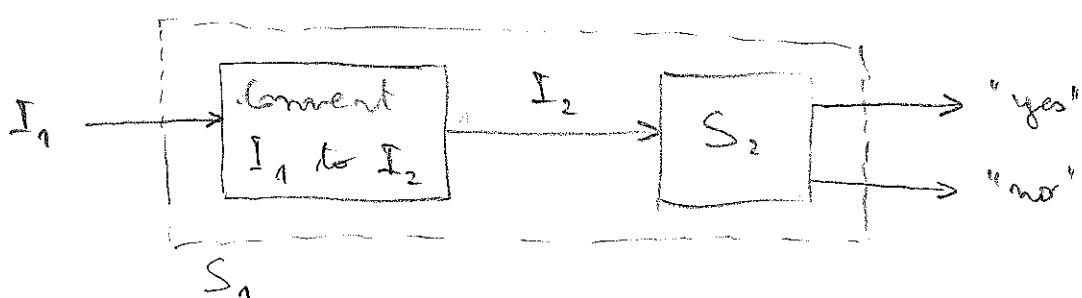
Problem  $P_1$  taking input  $I_1$ , known to be undecidable  
 $\dashv \dashv P_2 \dashv \dashv I_2$  to show undecidable.

Reduction: convert  $I_1$  to  $I_2$  such that

$$P_1(I_1) = \text{"yes"} \text{ iff } P_2(I_2) = \text{"yes"}$$

Given solution program  $S_2$  for  $P_2$ , we could obtain

$$\dashv \dashv S_1 \text{ for } P_1$$



Since  $S_1$  does exist, we obtain that  $S_2$  cannot exist  
 $\Rightarrow P_2$  is undecidable.

## Existence of undecidable problems:

While it was tricky to show that a specific problem is undecidable, it is rather easy to show that there are infinitely many undecidable problems.

We use a counting argument:

- a problem  $P$  is a language over  $\Sigma$  (for some finite  $\Sigma$ )  
 (the strings in the language represent those instances of  $P$  for which the answer is "yes")  
 $\Rightarrow$  there are uncountably many problems
- an algorithm is a string over  $\Sigma'$  (for some finite  $\Sigma'$ )  
 $\Rightarrow$  there are countably many algorithms  
 $\Rightarrow$  There must be (uncountably many) problems for which there is no algorithm.

17/10/2007

Turing Machines

Java (or C, Pascal, ...) programs are not well-suited to develop a theory of computation:

- run-time environment and run-time errors
- complex language constructs
- finite memory
- "state" of the computation is complicated to represent
- would need to show that the results for a specific programming language are in fact general

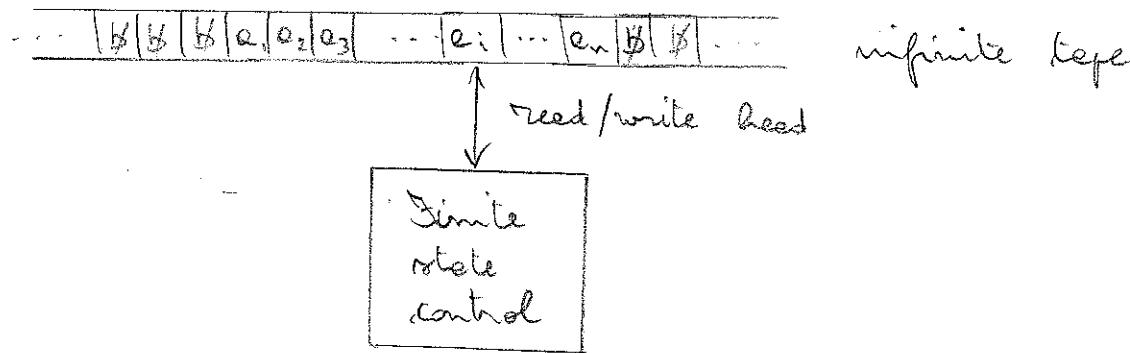
⇒ We resort to an abstract computing device, the Turing Machine (TM)

- simple and universal programming language
- state of computation is easy to describe
- unbounded memory
- can simulate any known computing device

Church - Turing hypothesis:

All reasonably powerful computation models are equivalent to TMs (but not more powerful).

⇒ TMs model anything we can compute.



Programmed by specifying transitions

- move depends on
  - current state (finitely many)
  - symbol under the tape head

- effects of a move:
  - new state
  - write new symbol on tape cell under the head
  - move head left/right/stay

Observations:

relationship to real computers: CPU  $\leftrightarrow$  finite state control  
memory  $\leftrightarrow$  tape

"differences" (features lost in the abstraction)

- no random access memory
- limited instruction set

However: a TM can simulate a computer (with a cubic increase in running time - see book 8.6)

Definition A TM  $M = (Q, \Sigma, \Gamma, \delta, q_0, \$, F)$

$Q$  ... set of states (finite)

$q_0 \in Q$  ... initial state

$\Sigma$  ... input alphabet (finite)

$\Gamma$  ... tape alphabet (finite)

$F \subseteq Q$  ... final states

$\$ \in \Gamma$  ... blank symbol

- Conditions:  $\Sigma \subseteq \Gamma$ , since input is written initially on tape
- $\$ \in \Gamma - \Sigma$ , since the rest of the tape is blank

Initially:

- state  $q_0$

- tape contains  $w$  surrounded by  $\$$

- tape head is at the leftmost cell of the input

Transitions:  $\delta: Q \times \Gamma \rightarrow Q \times \Gamma \times \{L, R, S\}$

$\delta(q, x) = (p, y, d)$  means that

if  $M$  is in state  $q$  and tape head is over symbol  $x$ ,  
then  $M$  changes state to  $p$ ,

- replaces  $x$  by  $y$  on the tape

- moves tape head by one cell in direction  $d$   
(left for L, right for R; S for stay in place)

The TM is deterministic:

for each  $\delta(q, x)$  we have at most one move  
( $\delta(q, x)$  could also be undefined)

Acceptance:  $w$  is accepted by TM  $M$  if  $M$ , when started with  $w$  on the tape, eventually enters a final state

We can assume that all final states are halting, i.e. no transition is defined for them

Rejection:

- halts in non final state (i.e., no transition defined)
- never halts (infinite loop)

Difference between FA / PDA and TM:

FA / PDA scans over  $w$  and accepts/rejects when it has reached its end

TM can move back and forth over  $w$  and accepts/rejects when it halts or rejects if it loops forever

Example:  $L = \{ w\#^t w^t \mid w \in \{0,1\}^+, t \in \{0,1,\#\}^* \}$

initially

$\$ | \# | \$ \dots w \dots \# | \dots | \# | \dots w \dots | \dots t \dots | \$ | \$ \dots$

TM idea: remember (in the state) leftmost symbol, and erase it

- move to leftmost symbol after #'s
- if the two don't match, then reject
- otherwise replace the symbol by #, move left and start again

$$M = (Q, \Sigma, \Gamma, \delta, q_0, \$, F)$$

$$Q = \{q_0, q_1, \dots, q_7\} \quad F = \{q_7\}$$

$$\Sigma = \{0, 1, \#\} \quad \Gamma = \{0, 1, \#, \$\}$$

$$\begin{array}{l} \delta(q_0, 0) = (q_1, \$, R) \\ \delta(q_0, 1) = (q_2, \$, R) \end{array} \left. \begin{array}{l} \text{Erase } 0 \text{ and look for matching } 0 \\ \dots - 1 \dots \end{array} \right. \quad \begin{array}{c} \\ \\ 1 \end{array}$$

$$\begin{array}{l} \delta(q_1, 0) = (q_1, 0, R) \\ \delta(q_1, 1) = (q_1, 1, R) \\ \delta(q_1, \#) = (q_3, \#, R) \end{array} \left. \begin{array}{l} \text{Skip over } 0's \text{ and } 1's, \\ \text{till } \# \text{ is found (remembering } 0) \end{array} \right.$$

$$\begin{array}{l} \delta(q_2, 0) = (q_2, 0, R) \\ \delta(q_2, 1) = (q_2, 1, R) \\ \delta(q_2, \#) = (q_4, \#, R) \end{array} \left. \begin{array}{l} \dots \\ \dots \\ \text{(remembering } 1) \end{array} \right.$$

2.11

$$\begin{cases} \delta(q_3, \#) = (q_3, \#, R) \\ \delta(q_3, 0) = (q_5, \#, L) \end{cases} \quad \left. \begin{array}{l} \text{Skip over \#}'s, look for 0,} \\ \text{and replace it by \#}. \end{array} \right\}$$

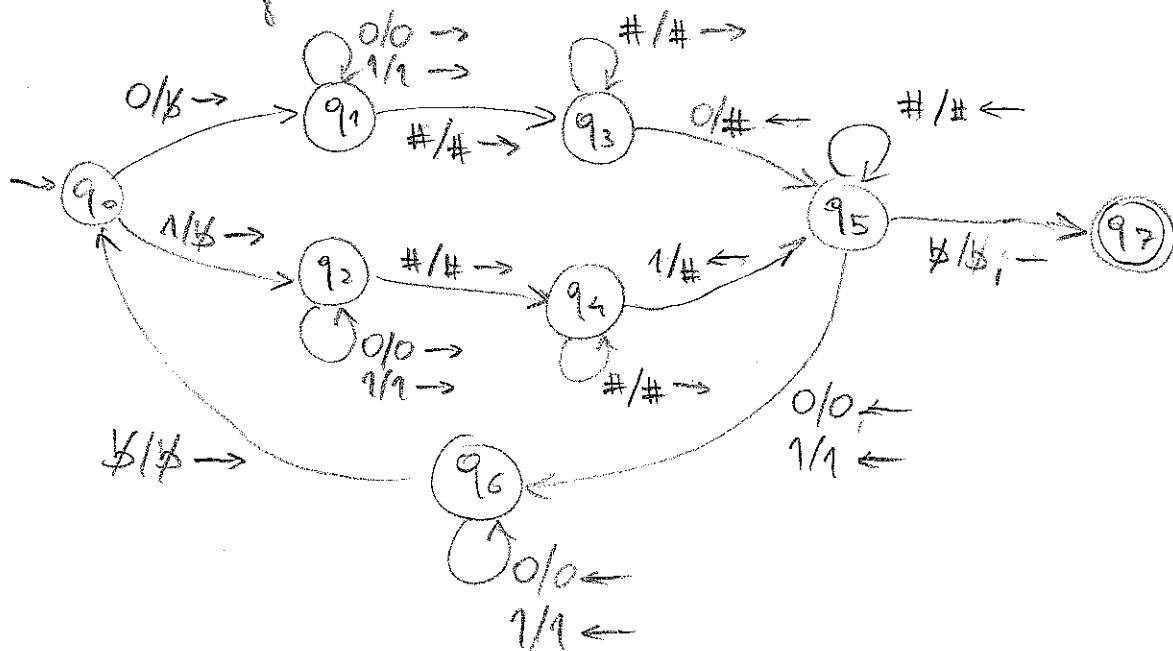
Note: if after \#'s a 1 or a \$ is found, M halts and rejects.

$$\begin{cases} \delta(q_4, \#) = (q_4, \#, R) \\ \delta(q_4, 1) = (q_5, \#, L) \end{cases} \quad \left. \begin{array}{l} \text{as previous ones, replacing 0/1} \\ \text{with 1/0.} \end{array} \right\}$$

$$\begin{cases} \delta(q_5, \#) = (q_5, \#, L) \\ \delta(q_5, 0) = (q_6, 0, L) \\ \delta(q_5, 1) = (q_6, 1, L) \\ \delta(q_5, \$) = (q_7, \$, S) \end{cases} \quad \left. \begin{array}{l} \text{Move left skipping \#}'s.} \\ \text{If to the left of the \#}'s a 0 or 1} \\ \text{is found, move to } q_6 \text{ to skip them} \\ \text{also. If \$ is found, accept.} \end{array} \right\}$$

$$\begin{cases} \delta(q_6, 0) = (q_6, 0, L) \\ \delta(q_6, 1) = (q_6, 1, L) \\ \delta(q_6, \$) = (q_0, \$, R) \end{cases} \quad \left. \begin{array}{l} \text{Move left, skipping 0's and 1's,} \\ \text{and restart again.} \end{array} \right\}$$

Transition diagram



$\xrightarrow{q \xrightarrow{x/y d} r}$  represents  $\delta(q, x) = (q, y, d)$

## 2.12

Instantaneous description (I.D.) or configuration of a TM

describes the current situation of TM and tape.

$$I.D. = \alpha_1 q \alpha_2 \quad \text{with } q \in Q \\ \alpha_1, \alpha_2 \in \Gamma^*$$

- means:
- non-blank portion of tape contains  $\alpha_1 \alpha_2$
  - head is on leftmost symbol of  $\alpha_2$
  - machine is in state  $q$

Corresponds to

BLANKS	$\alpha_1$	$\alpha_2$	BLANKS
$\downarrow$ state $q$			

Let  $ID = \Gamma^* \times Q \times \Gamma^*$  be the set of instantaneous descriptions. We use a relation  $\vdash \subseteq ID \times ID$  to describe the transitions of the TM.

Example:

$$\begin{aligned} q_0 01\#01 &\vdash q_1 1\#01 \vdash q_1 \#01 \vdash \\ &\vdash 1\#q_3 01 \vdash q_5 \#\#1 \vdash \\ &\vdash q_5 1\#\#1 \vdash q_6 \$ 1\#\#1 \vdash \\ &\vdash q_6 1\#\#1 \vdash \dots \vdash \\ &\vdash q_5 \$ \#\# \vdash q_7 \#\# \leftarrow \text{accepts} \end{aligned}$$

Note: we can define  $\vdash$  formally, making use of  $S$ . [Exercise]

Making use of the closure  $\vdash^*$  of  $\vdash$  we can define the language accepted by a TM

Definition: Let  $M = (Q, \Sigma, \Gamma, S, q_0, \$, F)$  be a TM.

Then the language  $L(M)$  accepted by  $M$  is

$$L(M) = \{ w \in \Sigma^* \mid q_0 w \vdash^* \alpha_1 q \alpha_2 \text{ with } q \in F \text{ and} \\ \alpha_1, \alpha_2 \in \Gamma^* \}$$

Notes:

1) We have used TMs for language recognition, which in turn corresponds to solving decision problems

- We can, however, consider also TMs as computing functions:
  - the output (result of the function) is left on the tape

2) The class of languages accepted by TMs are called recursively enumerable

- for a string  $w$  in the language
  - the TM halts on input  $w$  in a final state
- for a string  $w$  not in the language
  - the TM may halt in a non-final state, or
  - it may loop forever

Those languages for which the TM always halts (regardless of whether it accepts or not) are called recursive:

- these languages correspond to recursive functions
- TMs that always halt are a good model of algorithms and they correspond to decidable problems

We present some notational conveniences that make it easier to write TM programs

Idee: use structured states and tape symbols

1) Storage in the state: ("CPU register")

Idee: state names are a tuple of the form

$$[q_1, D_1, \dots, D_k]$$

$D_i$  .. etc as stored symbol

$q$  ... control portion of the state

Example: TM  $M = (Q, \Sigma, \Gamma, \delta, q_0, B, F)$  for  $L = 01^* + 10^*$

Idee: M remembers the first symbol and checks that it does not reappear

$$\begin{aligned} Q &= \{ [q_0, e] \mid e \in \{0, 1\}, e \notin \{0, 1, -\} \} = \\ &\quad \{ [q_0, -], [q_0, 0], [q_0, 1], [q_0, -], [q_1, 0], [q_1, 1] \} \\ \Sigma &= \{0, 1\} \quad \Gamma = \{0, 1, B\} \\ q_0 &= [q_0, B] \quad F = \{[q_1, -]\} \end{aligned}$$

Meaning of  $[q_i, e]$

- control portion  $q_i$ :

$q_0$  ... M has not yet read its first symbol

$q_1$  ... M has read its first symbol

- tape portion  $e$ :  $e$  is the first symbol read

transitions:

$$\delta([q_0, -], e) = ([q_1, e], e, R), \text{ for } e \in \{0, 1\}$$

... M remembers in  $[q_1, e]$  that it has read e

$$\begin{aligned} \delta([q_1, 0], 1) &= ([q_1, 0], 1, R) \\ \delta([q_1, 1], 0) &= ([q_1, 1], 0, R) \end{aligned} \quad \left. \begin{array}{l} M \text{ moves right as} \\ \text{long as it does not} \\ \text{see the first symbol} \end{array} \right\}$$

$$\delta([q_1, e], \$) = ([q_1, -], \$, R), \text{ for } e \in \{0, 1\}$$

... M accepts when it reaches the first \$

## 2) Multiple tracks:

Idee: view tape as having multiple tracks, i.e. if each symbol in  $\Gamma$  has multiple components

	0	*	\$	
..	1	0	0	..
a	a	a	c	

the symbols on the tape are  $[0]_a, [*]_a, [\$]_c$

Example:  $L = \{ww \mid w \in \{0, 1\}^+\}$

We first need to find midpoint, and then we can match corresponding symbols.

To find midpoint: we view tape as 2 tracks

			*				
0	1	1	0	1	1		

← used to put markers on symbols

Hence:  $\Gamma = \{[*], [\$], [\$_1], [\$_0], [*], [\$_1]\}$

(note: we need no \* over \$)

We put markers on two outermost symbols and move them inwards: 2.16

$$\begin{array}{l} \delta(q_0, [\overset{*}{\underset{i}{\alpha}}]) = (q_1, [\overset{*}{\underset{i}{\alpha}}], R) \\ \delta(q_1, [\overset{*}{\underset{i}{\alpha}}]) = (q_1, [\overset{*}{\underset{i}{\alpha}}], R) \\ \delta(q_1, [\overset{*}{\underset{i}{\alpha}}]) = (q_2, [\overset{*}{\underset{i}{\alpha}}], L) \\ \delta(q_1, [\overset{*}{\underset{i}{\alpha}}]) = (q_2, [\overset{*}{\underset{i}{\alpha}}], L) \\ \delta(q_2, [\overset{*}{\underset{i}{\alpha}}]) = (q_3, [\overset{*}{\underset{i}{\alpha}}], L) \\ \delta(q_3, [\overset{*}{\underset{i}{\alpha}}]) = (q_3, [\overset{*}{\underset{i}{\alpha}}], L) \\ \delta(q_3, [\overset{*}{\underset{i}{\alpha}}]) = (q_0, [\overset{*}{\underset{i}{\alpha}}], R) \end{array} \quad \left. \begin{array}{l} \text{move right till end} \\ \text{or first marked symbol} \\ \text{move rightmost mark} \\ \text{one symbol to the left} \\ \text{move left till end} \\ \text{or first marked symbol} \end{array} \right.$$

Note: we have each of the above for  $i \in \{0, 1\}$

At the end: head is over first symbol of second w,  
with a \* above it, in state  $q_0$ .

### 3) Subroutines / procedure calls

6/12/2006

Example: shifting over

$$\begin{array}{ll} \text{given: } ID_1 = \alpha q_i \times \beta & \text{for } \alpha \in \Gamma \\ \text{want: } ID_2 = \alpha \square q_i \times \beta & \alpha, \beta \in \Gamma^* \\ & \square \in \Gamma \end{array}$$

Subroutine for shifting over can be used repeatedly to  
create space in the middle of the tape

E.g. to implement a counter

$$\begin{aligned} \$0\$ &\rightarrow \$1\$ \rightarrow \$\square 1\$ \rightarrow \$01\$ \rightarrow \$10\$ \rightarrow \\ &\rightarrow \$11\$ \rightarrow \$\square 11\$ \rightarrow \$011\$ \rightarrow \dots \end{aligned}$$

Procedure cell:  $\delta(q_i, x) = ([q_i, x], [\square], R)$ ,  $\forall x \in \Gamma$

- remember return state  $q_i$ , end erased symbol  $x$
- state  $p$  calls procedure

Procedure  $\varphi$  for shifting

1) shift 1 cell to the right

$$\delta([p, x], y) = ([p, y], x, R) \quad \forall x, y \in \Gamma \text{ with } y \neq \square$$

2) till we have reached end of  $B$

$$\delta([p, y], \square) = (r, y, L) \quad \forall y \in \Gamma$$

3) return to calling point by moving left

$$\delta(r, y) = (r, y, L) \quad \forall y \neq [\square]$$

4) exit and return to state  $q_i$

$$\delta(r, [\square]) = (q_i, \square, R)$$

In fact, we can implement arbitrary complex procedures, with any kind of parameter passing

**Exercise:** redesign the TMs you have seen so far to take advantage of storage in the state, multiple tracks, and subroutines

## Extensions to the basic TM

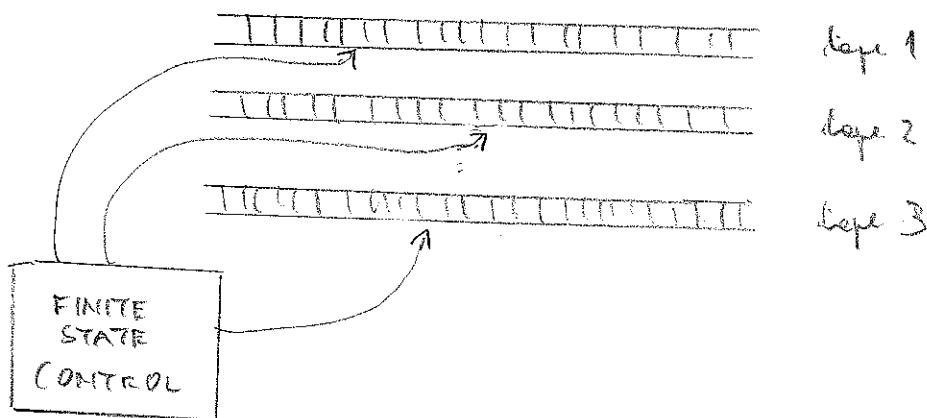
Note: if the TM seen so far can compute all that can be computed, then it should not become more expressive by extending it.

We consider two extensions:

- multiple tapes
- non-determinism

and show that both can be captured by the basic T.M.

### 1) Multi-tape T.M.



Initially: input  $w$  is on tape 1 with tape-head on the leftmost symbol. Other tapes are all blank.

Transitions: specify behaviour of each head independently

$$\delta(q, x_1, \dots, x_h) = (q, (y_1, d_1), \dots, (y_h, d_h))$$

$x_i$  ... symbol under head  $i$

$y_i$  ... new symbol written to head  $i$

$d_i$  ... direction in which head  $i$  moves

To simulate  $k$ -tape TM  $M_k$  with a 1-tape TM  $M_1$ ,  
 we use  $2k$  tracks in  $M_1$ : for each tape of  $M_k$

- one track of  $M_1$  to store tape content
- one track of  $M_1$  to mark head position with \*

A	B	A	C	B	A		Type 1
		*					head 1
0	0	1	1	1	0		Type 2
		*					head 2
b	b	s	b	s	b		Type 3
						*	head 3

Each transition of  $M_k$  is simulated by a series of transitions of  $M_1$ :  $\delta(q, x_1, \dots, x_k) = (q, (y_1, d_1), \dots, (y_k, d_k))$

- start at leftmost head position marker
- sweep right and remember in appropriate "CPU registers" the symbols  $x_i$  under each head (note: there are exactly  $k$ , and hence finitely many)
- knowing all  $x_i$ 's, sweep left, change each  $x_i$  to  $y_i$ , and move the marker for tape  $i$  according to  $d_i$

Note:  $M_1$  needs to remember always how many of the  $k$  heads are to its left (uses an additional (PV-)register)

The final states of  $M_1$  are those that have in the state-component a final state of  $M_k$ .

We can verify that we can construct  $M_1$  so that  
 $L(M_1) = L(M_k)$

(details are straightforward, but cumbersome)

Simulation speed:

Note :- enhancements do not affect the expressive power of e TM

- they do effect its efficiency

Definition: e TM is said to have running time  $T(n)$  if it halts within  $T(n)$  steps on all inputs of length  $n$ .

Note :  $T(n)$  could be infinite

Theorem: If  $M_h$  has running time  $T(n)$ , then  $M_s$  will simulate it with running time  $O(T(n)^2)$ .

Proof: Consider input  $w$  of length  $n$ .

- $M_h$  runs at most  $T(n)$  time on it.
- At each step, leftmost and rightmost heads can drift apart by at most 2 additional cells.
- It follows that after  $T(n)$  steps the 2 heads cannot be more than  $2 \cdot T(n)$  apart, and  $M_h$  uses  $\leq 2 \cdot T(n)$  tape cells

Consider  $M_s$ :

- makes two sweeps for each transition of  $M_h$
- each sweep takes at most  $O(T(n))$
- number of transitions of  $M_h$  is  $\leq T(n)$

It follows that the total running time is  $O(T(n)^2)$ .

## 2) Non-deterministic TMs (NTM)

In a (deterministic) TM,  $\delta(q, x)$  is unique or undefined.

In a NTM,  $\delta(q, x)$  is a finite set of triples

$$\delta(q, x) = \{(q_1, y_1, d_1), \dots, (q_k, y_k, d_k)\}$$

At each step, the NTM can non-deterministically choose which transition to make.

As for other ND devices: a string  $w$  is accepted if the NTM has at least one execution leading to a final state.

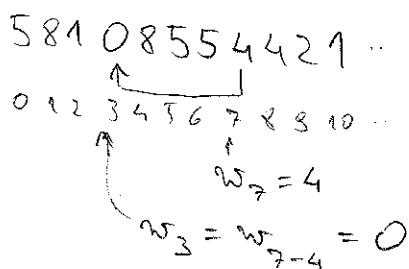
Example:  $\Sigma = \{0, 1, \dots, 9\}$

$L = \{w \in \Sigma^* \mid \text{a } 0 \text{ appears in positions to the left of some } i \text{ in } w, \text{ with } 0 < i \leq 9\} =$

$$= \{w \in \Sigma^* \mid \exists j > 0 \text{ s.t. } w_{j-1} = 0\}$$

( $w_i$  indicates the  $i$ -th character of  $w$ )

Ex. 02146  $\in L$



NTM  $N$  s.t.  $I(N) = L$

$$Q = \{q_0, f, [q, 0], [q, 1], \dots, [q, 9]\}$$

$$F = \{f\}$$

$$\Gamma = \{0, 1, \dots, 9, \$\}$$

Idee für N: scan w from left to right.

2.22

- guess at some  $w_j = i$ ,
- store i in CPU register, and
- move i steps left to find 0

Transitions:

- $\delta(q_0, 0) = \{(q_0, 0, R)\}$  (since  $w_j > 1$ )
- $\forall i > 0 : \delta(q_0, i) = \{(q_0, i, R), ([\uparrow, i], i, L)\}$   
↑  
guess
- $\forall i \geq 2, \forall x \in \Gamma : \delta([\uparrow, i], x) = \{[\uparrow, i-1], x, L\}$
- accepting:  $\delta([\uparrow, 1], 0) = \{(\downarrow, 0, R)\}$

Execution traces on input  $w = 103332$

$q_0 103332 \xrightarrow{} q_0 03332 \xrightarrow{} 10 q_0 3332 \xrightarrow{} 103 q_0 332 \xrightarrow{} \dots$   
 $\xrightarrow{} 10 [\uparrow, 3] 3332 \xrightarrow{} 1 [\uparrow, 2] 03332 \xrightarrow{} [\uparrow, 1] 103332 \xrightarrow{} \dots$   
 $\Rightarrow \text{reject}$

$q_0 103332 \xrightarrow{*} 1033 q_0 32 \xrightarrow{} 103 [\uparrow, 3] 332 \xrightarrow{} \dots$   
 $\xrightarrow{} 10 [\uparrow, 2] 3332 \xrightarrow{} 1 [\uparrow, 1] 03332 \xrightarrow{} 10 f 3332 \xrightarrow{} \dots$   
 $\Rightarrow \text{accept}$

11/12/2006

Theorem: Let N be a NTM. Then there exists a DTM D s.t.

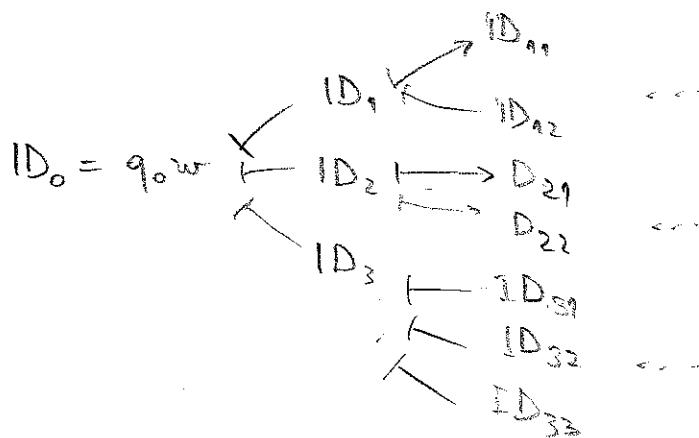
$$L(D) = L(N)$$

Proof: Given N and w, we show how a multi-tape DTM can simulate the execution of N on input w.

We can then convert the multi-tape DTM to a single-tape DTM

Idea for the simulation:

Consider the execution tree of  $N$  on  $w$



DTM  $D$  will perform a breadth-first search of the execution tree, systematically enumerating the IDs, until it finds an accepting one.

We use two tapes:

Type 2: is for working

Type 1: contains a sequence of ID's of  $N$  in BFS order

- \* used to separate two ID's
- ^ marks next ID to be explored
  - ID's to the left of ^ have been explored
  - ID's to the right of ^ are still to be explored
- initially, only  $ID_0 = q_0 w$  is on the tape
- we can use multiple blocks for convenience

Algorithm: repeat the following steps

Step 0: examine current  $ID_c$  (the one after  $\hat{*}$ ) and read  $q, e$  from it

if  $q \in F$ , then accept and halt

Step 1: let  $\delta(q, e)$  have  $k$  possible transitions

- copy  $ID_c$  onto tape 2

- make  $k$  new copies of  $ID_c$  and place them at the end of tape 1

Step 2: modify the  $k$  copies of  $ID_c$  on tape 1 to become the  $k$  possible outcomes of  $\delta(q, e)$  on  $ID_c$

Step 3: move  $\hat{*}$  right past  $ID_c$ .

clean up tape 2

return to step 0

It is possible to verify:

- the above steps can all be implemented in a DTM.
- the construction is correct, i.e.  $w \in L(D)$  iff  $w \in L(N)$

Evolution of tape 1:

1)  $\hat{*} ID_0 *$

2)  $\hat{*} ID_0 * ID_0 * ID_0 * ID_0 *$

3)  $\hat{*} ID_0 * ID_1 * ID_2 * ID_3 *$

4)  $\hat{*} ID_0 * ID_1 * ID_2 * ID_3 *$

5)  $\hat{*} ID_0 * ID_1 * ID_2 * ID_3 * ID_4 * ID_5 *$

6)  $* \dots \dots \hat{*} ID_{n1} * ID_{n2} *$

7)  $\hat{*} ID_0 * ID_1 * ID_2 * ID_3 * ID_{n1} * ID_{n2} *$

## Simulation time:

(2.25)

Let NTM N have running time  $T(n)$ .

What is the running time of D?

Let  $m$  be the maximum number of non-det. choices for each transition (i.e., the maximum size of  $\delta(q, x)$ )

Consider execution tree of N on w.

let  $t = T(|w|) \Rightarrow$  exec. tree has at most  $t$  levels

$$\text{size of the tree is } \leq 1 + m + m^2 + \dots + m^t =$$

$$\leq \frac{m^{t+1} - 1}{m - 1} = O(m^t)$$

Thus D has at most  $O(m^t)$  iterations of steps 0-3.

Each iteration requires at most  $O(m^t)$  steps

$\Rightarrow$  Total running time is  $m^{O(t)}$ , i.e. exponential