

\* Unbounded minimisation

Given a function (possibly partial)

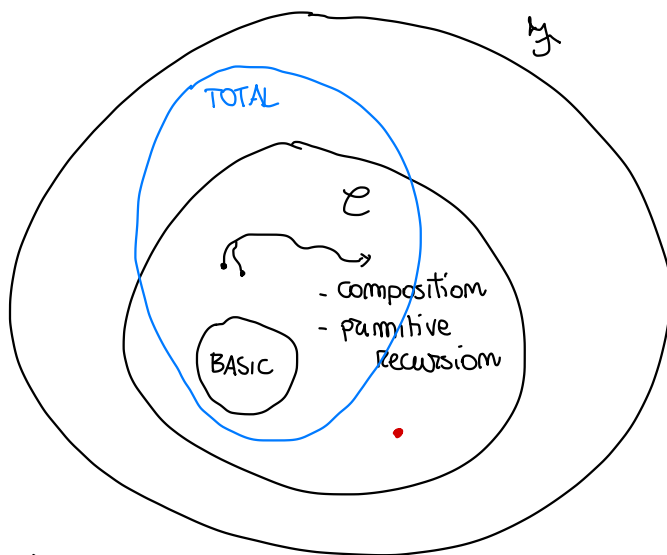
$$f: \mathbb{N}^{k+1} \rightarrow \mathbb{N}$$

$$f(\vec{x}, z)$$

define  $h: \mathbb{N}^k \rightarrow \mathbb{N}$

$$h(\vec{x}) = \text{least } y \text{ s.t. } f(\vec{x}, y) = 0$$

$$= \mu y. f(\vec{x}, y)$$

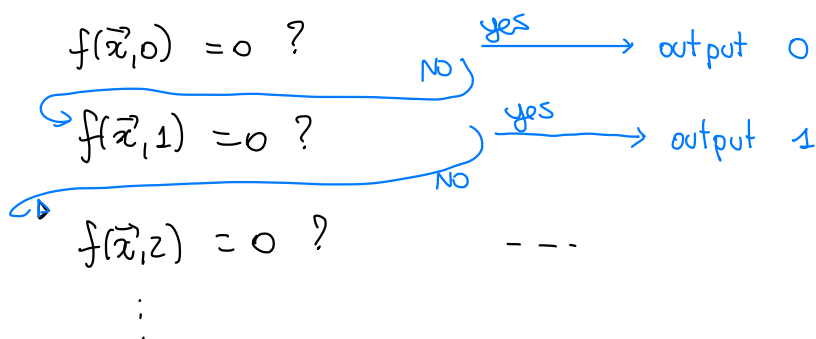


two issues:

- there could be no  $y$  s.t.  $f(\vec{x}, y) = 0$
  - $f(\vec{x}, z)$  might be undefined "before" finding  $y$  s.t.  $f(\vec{x}, y) = 0$
- ↳ minimisation is  $\uparrow$

$$h(\vec{x}) = \mu y. f(\vec{x}, y) = \begin{cases} y & \text{if there is } y \text{ s.t. } f(\vec{x}, y) = 0 \\ & \text{and } \forall z < y \quad f(\vec{x}, z) \downarrow \text{ and } f(\vec{x}, z) \neq 0 \\ \uparrow & \text{if there is no such } y \end{cases}$$

In order to compute  $\mu y. f(\vec{x}, y)$



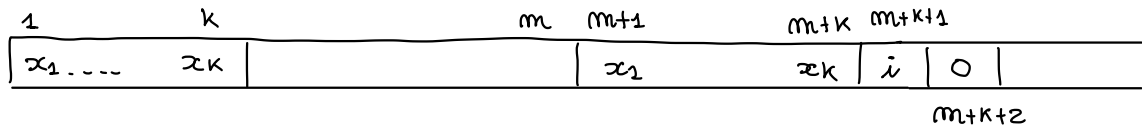
Proposition: Class  $C$  is closed under (unbounded) minimisation

proof

let  $f: \mathbb{N}^{k+1} \rightarrow \mathbb{N}$  be a computable function

and let  $P$  be a (standard form) program for  $f$

We show that  $h(\vec{x}) = \mu y. f(\vec{x}, y)$  is computable by showing a program for it



$$m = \max \{ p(P), k+1 \}$$

$$f(\vec{x}, i) \quad \begin{matrix} i=0 \\ =1 \\ =2 \\ \vdots \end{matrix}$$

program for  $h$  is

$T(1, m+1)$  // save  $\vec{x}$  to a safe place  
 $\vdots$   
 $T(k, m+k)$

LOOP:  $P [m+1, \dots, m+k, m+k+1 \rightarrow 1]$  //  $f(\vec{x}, i)$  in  $R_1$

$J(1, m+k+2, \text{END})$  //  $f(\vec{x}, i) = 0$  ?

$S(m+k+1)$  //  $i++$

$J(1, 1, \text{LOOP})$

END:  $T(m+k+1, 1)$  // output  $i$

□

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

$$f(x) = \begin{cases} x/2 & \text{if } x \text{ even} \\ \uparrow & \text{otherwise} \\ \mu y. |2 \times y - x| & \\ \mu y. |y + y - x| & \end{cases}$$

computable  
by minimization

EXAMPLE  $g: \mathbb{N}^2 \rightarrow \mathbb{N}$

$$g(x, y) = \begin{cases} x/y & \text{if } y \neq 0 \text{ and } y \text{ is a divisor of } x \\ \uparrow & \text{otherwise} \\ \neq \mu z. |z \times y - x| \end{cases}$$

$\uparrow$   
 $y=0 \ \& \ x=0$

we get  $0$   
we want  $\uparrow$

$$= \mu z. (|z \times y - x| + \overline{\text{sg}}(y))$$

$$\begin{array}{l} \uparrow \\ 1 \text{ if } y=0 \\ 0 \text{ if } y \neq 0 \end{array}$$

OBSERVATION : Every finite (domain) function is computable

proof

Let  $\vartheta : \mathbb{N} \rightarrow \mathbb{N}$  be a finite function

$$\vartheta(x) = \begin{cases} y_1 & \text{if } x = x_1 \\ y_2 & \text{if } x = x_2 \\ \vdots & \vdots \\ y_m & \text{if } x = x_m \\ \uparrow & \text{otherwise} \end{cases} \quad \text{dom}(\vartheta) = \{x_1, \dots, x_m\}$$

$$= \{(x_1, y_1), \dots, (x_m, y_m)\}$$

$\vartheta$  is computable

$$\vartheta(x) = \sum_{i=1}^m y_i \cdot \underbrace{\overline{\text{sg}}(|x - x_i|)}_{\substack{0 \text{ if } x \neq x_i \\ 1 \text{ if } x = x_i}} + \mu z. \underbrace{\frac{y(xz)}{\prod_{i=1}^m |x - x_i|}}_{\substack{0 \text{ if } x \in \text{dom}(\vartheta) \\ \neq 0 \text{ otherwise}}}$$

$$\underbrace{\substack{y_i \text{ if } x = x_i \\ 0 \text{ if } x \neq x_i}}_{\substack{0 \text{ if } x \in \text{dom}(\vartheta) \\ \uparrow \text{ otherwise}}}$$

□

Example :

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

$$f(x) = \begin{cases} 1 & \text{if } x = 0 \text{ and } P = NP \\ 0 & \text{if } x = 0 \text{ and } P \neq NP \\ \uparrow & \text{if } x \neq 0 \end{cases}$$

computable

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

fix a program  $P$

$$g(x) = \begin{cases} 0 & \text{if } x=0 \text{ and } P(0) \uparrow \\ 1 & \text{if } x=0 \text{ and } P(0) \downarrow \\ \uparrow & \text{otherwise} \end{cases}$$

computable

OBSERVATION : let  $f: \mathbb{N} \rightarrow \mathbb{N}$  TOTAL computable and injective

Then

$$f^{-1}(y) = \begin{cases} x & \text{if there is } x \text{ st. } f(x)=y \\ \uparrow & \text{if there is no such } x \end{cases}$$

is computable.

proof

$$f^{-1}(y) = \mu x. |f(x) - y|$$

□

Not working for partial functions

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

$$f(x) = \begin{cases} x+1 & \text{if } x > 0 \\ \uparrow & \text{otherwise} \end{cases}$$

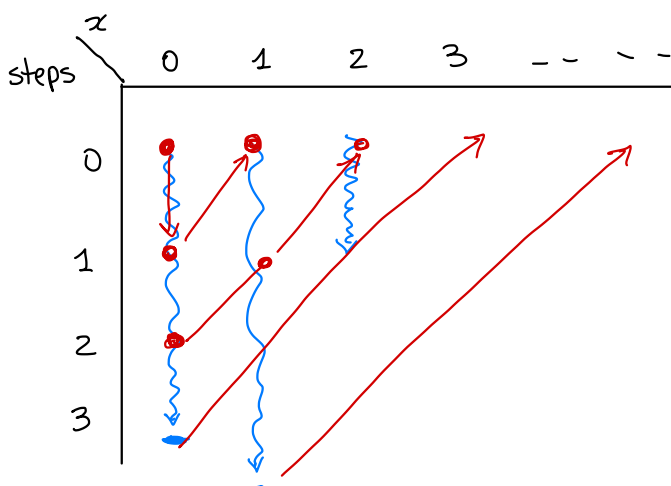
$$g(x,z) = \bar{s}g(x)$$

$$= (x+1) + \boxed{\mu z. \bar{s}g(x)} \quad \text{computable}$$

$$x=0 \quad (f(0)-y) \neq 0 ?$$

$$f^{-1}(y) = y+1 \neq \mu x. |f(x) - y| \quad \forall y$$

\* What if  $f$  is partial? Does the result hold? YES



$y$  find  $x$  s.t.  
 $f(x)=y$   
 $f$  computed by  $P$

you check  $P(x)$   
for any  $x$   
and any number of steps

# Partial Recursive Functions

computational models: TM,  $\lambda$ -calculus, Post systems, -----, URM-machines

Church Turing thesis: A function is computable by an effective procedure  
iff  
it is URM-computable

TO DO LIST:

→ define the class  $\mathcal{R}$  of partial recursive functions

→  $\mathcal{E} = \mathcal{R}$

Def. The class of partial recursive functions  $\mathcal{R}$  is the least class  
of functions which  
w.r.t.  $\subseteq$

→ contains

(a) zero

(b) successor

(c) projections

→ closed under

(1) composition

(2) primitive recursion

(3) minimisation

→ define rich class of functions  $\mathcal{A}$  as a class of functions which

\* contains (a), (b), (c)

\* is closed w.r.t. (1), (2), (3)

→  $\mathcal{R}$  is the least rich class, i.e. for all rich class  $\mathcal{A}$   $\mathcal{R} \subseteq \mathcal{A}$

→ OBSERVE: • given  $\mathcal{A}_i$   $i \in I$  rich classes then  $\bigcap_{i \in I} \mathcal{A}_i$  is rich

• the class of all functions is rich

define  $\mathcal{R} = \bigcap_{\mathcal{A} \text{ rich class}} \mathcal{A}$

Equivalently:  $\mathcal{R}$  is the class of functions that one obtains from the  
BASIC FUNCTIONS using operations (1), (2), (3) a finite number of  
times.

Theorem :  $\mathcal{C} = \mathcal{R}$

proof

$(\mathcal{R} \subseteq \mathcal{C})$   $\mathcal{C}$  is rich,  $\mathcal{R}$  is the least rich class

$(\mathcal{C} \subseteq \mathcal{R})$  let  $f : \mathbb{N}^k \rightarrow \mathbb{N}$  in  $\mathcal{C}$  desize  $\rightsquigarrow$   $f \in \mathcal{R}$

let  $P$  be a program for  $f$



$\begin{cases} C_P^1 : \mathbb{N}^{k+1} \rightarrow \mathbb{N} \\ C_P^1(\vec{x}, t) = \text{content of } R_1 \text{ after } t \text{ steps of computation of } P(\vec{x}) \end{cases}$

$\begin{cases} J_P : \mathbb{N}^{k+1} \rightarrow \mathbb{N} \\ J_P(\vec{x}, t) = \begin{cases} \text{instruction to be executed after } t \text{ steps of } P(\vec{x}) \\ 0 \text{ if } P(\vec{x}) \text{ in } t \text{ steps or fewer} \end{cases} \end{cases}$

let  $\vec{x} \in \mathbb{N}^k$ . We distinguish two cases

$\rightarrow f(\vec{x}) \downarrow$  then  $P(\vec{x}) \downarrow$  in a number steps

$$t_0 = \mu t. J_P(\vec{x}, t)$$

and

$$f(\vec{x}) = C_P^1(\vec{x}, t_0) = C_P^1(\vec{x}, \mu t. J_P(\vec{x}, t))$$

$\rightarrow f(\vec{x}) \uparrow$  then  $P(\vec{x}) \uparrow$  and thus  $\mu t. J_P(\vec{x}, t) \uparrow$

$$f(\vec{x}) = C_P^1(\vec{x}, \underbrace{\mu t. J_P(\vec{x}, t)}_{\uparrow \text{ undefined}}) \uparrow$$

for all  $\vec{x}$

$$f(\vec{x}) = C_p^1(\vec{x}, \mu t. J_p(\vec{x}, t))$$

If we are able to show  $C_p^1, J_p \in \mathcal{R}$  then  $f \in \mathcal{R}$

NEXT LESSON