Lecture Notes on

Computation Theory

for the

Computer Science Tripos, Part IB

Andrew M. Pitts University of Cambridge Computer Laboratory First edition 1995. Second edition 2009; revised 2011.

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Learning Guide

These notes are designed to accompany 12 lectures on computation theory for Part IB of the Computer Science Tripos. The aim of this course is to introduce several apparently different formalisations of the informal notion of algorithm; to show that they are equivalent; and to use them to demonstrate that there are uncomputable functions and algorithmically undecidable problems. At the end of the course you should:

- be familiar with the register machine, Turing machine and λ -calculus models of computability;
- understand the notion of coding programs as data, and of a universal machine;
- be able to use diagonalisation to prove the undecidability of the Halting Problem;
- understand the mathematical notion of partial recursive function and its relationship to computability.

The prerequisites for taking this course are the Part IA courses *Discrete Mathematics* and *Regular Languages and Finite Automata*.

This incaration of the *Computation Theory* course builds on previous lecture notes by Ken Moody, Glynn Winskel, Larry Paulson and myself. It contains some material that **everyone who calls themselves a computer scientist should know**. It is also a prerequisite for the Part IB course on *Complexity Theory*.

Recommended books

- Hopcroft, J.E., Motwani, R. & Ullman, J.D. (2001). *Introduction to Automata Theory, Languages and Computation, Second Edition*. Addison-Wesley.
- Hindley, J.R. & Seldin, J.P. (2008). *Lambda-Calculus and Combinators, an Introduction*. Cambridge University Press (2nd ed.).
- Cutland, N.J. (1980) Computability. An introduction to recursive function theory. Cambridge University Press.
- Davis, M.D., Sigal, R. & Wyuker E.J. (1994). *Computability, Complexity and Languages*, 2nd edition. Academic Press.
- Sudkamp, T.A. (1995). Languages and Machines, 2nd edition. Addison-Wesley.

Exercises and Tripos Questions

A course on *Computation Theory* has been offered for many years. Since 2009 the course has incorporated some material from a Part IB course on *Foundations of Functional Programming* that is no longer offered. A guide to which Tripos questions from the last five years are relevant to the current course can be found on the course web page (follow links from www.cl.cam.ac.uk/teaching/). Here are suggestions for which of the older ones to try, together with some other exercises.

- 1. Exercises in register machine programming:
 - (a) Produce register machine programs for the functions mentioned on slides 36 and 37.
 - (b) Try Tripos question 1999.3.9.
- 2. Undecidability of the halting problem:
 - (a) Try Tripos question 1995.3.9.
 - (b) Try Tripos question 2000.3.9.
 - (c) Learn by heart the poem about the undecidability of the halting problem to be found at the course web page and recite it to your non-compsci friends.
- 3. Let ϕ_e denote the unary partial function from numbers to numbers (i.e. an element of $\mathbb{N} \rightarrow \mathbb{N}$ —cf. slide 30) computed by the register machine with code e (cf. slide 63). Show that for any given register machine computable unary partial function f, there are infinitely many numbers e such that $\phi_e = f$. (Equality of partial functions means that they are equal as sets of ordered pairs; which is equivalent to saying that for all numbers x, $\phi_e(x)$ is defined if and only if f(x) is, and in that case they are equal numbers.)
- 4. Suppose S_1 and S_2 are subsets of the set $\mathbb{N} = \{0,1,2,3,\ldots\}$ of natural numbers. Suppose $f \in \mathbb{N} \to \mathbb{N}$ is register machine computable and satisfies: for all x in \mathbb{N} , x is an element of S_1 if and only if f(x) is an element of S_2 . Show that S_1 is register machine decidable (cf. slide 66) if S_2 is.
- 5. Show that the set of codes $\langle e, e' \rangle$ of pairs of numbers e and e' satisfying $\phi_e = \phi_{e'}$ is undecidable.
- 6. For the example Turing machine given on slide 75, give the register machine program implementing

$$(S,T,D) := \delta(S,T)$$

as described on slide 83. [Tedious!—maybe just do a bit.]

- 7. Try Tripos question 2001.3.9. [This is the Turing machine version of 2000.3.9.]
- 8. Try Tripos question 1996.3.9.
- 9. Show that the following functions are all primitive recursive.
 - (a) Exponentiation, $exp(x,y) \triangleq x^y$.

- (b) Truncated subtraction, $minus(x,y) \triangleq \begin{cases} x-y & \text{if } x \geq y \\ 0 & \text{if } x < y \end{cases}$
- (c) Conditional branch on zero, if $z = (x, y, z) \triangleq \begin{cases} y & \text{if } x = 0 \\ z & \text{if } x > 0 \end{cases}$
- (d) *Bounded summation*: if $f \in \mathbb{N}^{n+1} \to \mathbb{N}$ is primitive recursive, then so is $g \in \mathbb{N}^{n+1} \to \mathbb{N}$ where

$$g(\vec{x}, x) \triangleq \begin{cases} 0 & \text{if } x = 0\\ f(\vec{x}, 0) & \text{if } x = 1\\ f(\vec{x}, 0) + \dots + f(\vec{x}, x - 1) & \text{if } x > 1. \end{cases}$$

- 10. Recall the definition of Ackermann's function ack from slide 122. Sketch how to build a register machine M that computes $ack(x_1, x_2)$ in R0 when started with x_1 in R1 and x_2 in R2 and all other registers zero. [Hint: here's one way; the next question steers you another way to the computability of ack. Call a finite list $L = [(x_1, y_1, z_1), (x_2, y_2, z_2), \ldots]$ of triples of numbers suitable if it satisfies
 - (i) if $(0, y, z) \in L$, then z = y + 1
 - (ii) if $(x + 1, 0, z) \in L$, then $(x, 1, z) \in L$
 - (iii) if $(x+1,y+1,z) \in L$, then there is some u with $(x+1,y,u) \in L$ and $(x,u,z) \in L$.

The idea is that if $(x,y,z) \in L$ and L is suitable then z = ack(x,y) and L contains all the triples (x',y',ack(x,y')) needed to calculate ack(x,y). Show how to code lists of triples of numbers as numbers in such a way that we can (in principle, no need to do it explicitly!) build a register machine that recognizes whether or not a number is the code for a *suitable* list of triples. Show how to use that machine to build a machine computing ack(x,y) by searching for the code of a suitable list containing a triple with x and y in it's first two components.]

- 11. If you are not already fed up with Ackermann's function, try Tripos question 2001.4.8.
- 12. If you are *still* not fed up with Ackermann's function $ack \in \mathbb{N}^2 \to \mathbb{N}$, show that the λ -term ack $\triangleq \lambda x. x (\lambda f y. y f (f \underline{1}))$ Succ represents ack (where Succ is as on slide 152).
- 13. Let I be the λ -term λx . x. Show that $\underline{n}I =_{\beta} I$ holds for every Church numeral \underline{n} . Now consider

$$\mathsf{B} \triangleq \lambda f \, g \, x. \, g \, x \, \mathsf{I} \, (f \, (g \, x))$$

Assuming the fact about normal order reduction mentioned on slide 145, show that if partial functions $f,g \in \mathbb{N} \to \mathbb{N}$ are represented by closed λ -terms F and G respectively, then their composition $(f \circ g)(x) \equiv f(g(x))$ is represented by B F G. Now try Tripos question 2005.5.12.

Introduction

Computation Theory , L 1 2/17

Algorithmically undecidable problems

Computers cannot solve all mathematical problems, even if they are given unlimited time and working space.

Three famous examples of computationally unsolvable problems are sketched in this lecture.

- ▶ Hilbert's *Entscheidungsproblem*
- ► The Halting Problem
- ► Hilbert's 10th Problem.

Computation Theory , L 1 3/17

Hilbert's Entscheidungsproblem

Is there an algorithm which when fed any statement in the formal language of first-order arithmetic, determines in a finite number of steps whether or not the statement is provable from Peano's axioms for arithmetic, using the usual rules of first-order logic?

Such an algorithm would be useful! For example, by running it on

$$\forall k > 1 \,\exists p, q \, (2k = p + q \land prime(p) \land prime(q))$$

(where prime(p) is a suitable arithmetic statement that p is a prime number) we could solve Goldbach's Conjecture ("every strictly positive even number is the sum of two primes"), a famous open problem in number theory.

Computation Theory , L 1

Hilbert's Entscheidungsproblem

Is there an algorithm which when fed any statement in the formal language of first-order arithmetic, determines in a finite number of steps whether or not the statement is provable from Peano's axioms for arithmetic, using the usual rules of first-order logic?

Posed by Hilbert at the 1928 International Congress of Mathematicians. The problem was actually stated in a more ambitious form, with a more powerful formal system in place of first-order logic.

In 1928, Hilbert believed that such an algorithm could be found. A few years later he was proved wrong by the work of Church and Turing in 1935/36, as we will see.

Decision problems

Entscheidungsproblem means "decision problem". Given

a set S whose elements are finite data structures of some kind

(e.g. formulas of first-order arithmetic)

a property P of elements of S
 (e.g. property of a formula that it has a proof)

the associated decision problem is:

find an algorithm which terminates with result $\mathbf{0}$ or $\mathbf{1}$ when fed an element $s \in S$ and yields result $\mathbf{1}$ when fed s if and only if s has property P.

Computation Theory , L 1 6/171

Algorithms, informally

No precise definition of "algorithm" at the time Hilbert posed the *Entscheidungsproblem*, just examples, such as:

- Procedure for multiplying numbers in decimal place notation.
- Procedure for extracting square roots to any desired accuracy.
- Euclid's algorithm for finding highest common factors.

Computation Theory , L 1 7/17

Algorithms, informally

No precise definition of "algorithm" at the time Hilbert posed the *Entscheidungsproblem*, just examples.

Common features of the examples:

- finite description of the procedure in terms of elementary operations
- deterministic (next step uniquely determined if there is one)
- procedure may not terminate on some input data, but we can recognize when it does terminate and what the result is.

Computation Theory, L 1

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Algorithms, informally

No precise definition of "algorithm" at the time Hilbert posed the *Entscheidungsproblem*, just examples.

In 1935/36 Turing in Cambridge and Church in Princeton independently gave negative solutions to Hilbert's *Entscheidungsproblem*.

 First step: give a precise, mathematical definition of "algorithm".

(Turing: Turing Machines; Church: lambda-calculus.)

► Then one can regard algorithms as data on which algorithms can act and reduce the problem to...

Computation Theory, L 1

The Halting Problem

is the decision problem with

- ▶ set S consists of all pairs (A, D), where A is an algorithm and D is a datum on which it is designed to operate;
- ▶ property P holds for (A, D) if algorithm A when applied to datum D eventually produces a result (that is, eventually halts—we write $A(D) \downarrow$ to indicate this).

Turing and Church's work shows that the Halting Problem is undecidable, that is, there is no algorithm H such that for all $(A,D)\in S$

$$H(A,D) = egin{cases} 1 & ext{if } A(D) \downarrow \ 0 & ext{otherwise.} \end{cases}$$

Computation Theory , L $\mathbf{1}$

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There's no
$$H$$
 such that $H(A,D) = \begin{cases} 1 & \text{if } A(D) \downarrow \\ 0 & \text{otherwise.} \end{cases}$ for all (A,D) .

Informal proof, by contradiction. If there were such an H, let C be the algorithm:

"input A; compute H(A,A); if H(A,A)=0 then return 1, else loop forever."

So
$$\forall A \ (C(A) \downarrow \leftrightarrow H(A,A) = 0)$$
 (since H is total) and $\forall A \ (H(A,A) = 0 \leftrightarrow \neg A(A) \downarrow)$ (definition of H). So $\forall A \ (C(A) \downarrow \leftrightarrow \neg A(A) \downarrow)$.

Taking A to be C, we get $C(C) \downarrow \leftrightarrow \neg C(C) \downarrow$, contradiction!

From HP to Entscheidungsproblem

Final step in Turing/Church proof of undecidability of the *Entscheidungsproblem*: they constructed an algorithm encoding instances (A, D) of the Halting Problem as arithmetic statements $\Phi_{A,D}$ with the property

$$\Phi_{A,D}$$
 is provable $\leftrightarrow A(D) \downarrow$

Thus any algorithm deciding provability of arithmetic statements could be used to decide the Halting Problem—so no such exists.

Computation Theory , L 1

Hilbert's Entscheidungsproblem

Is there an algorithm which when fed any statement in the formal language of first-order arithmetic, determines in a finite number of steps whether or not the statement is provable from Peano's axioms for arithmetic, using the usual rules of first-order logic?

With hindsight, a positive solution to the *Entscheidungsproblem* would be too good to be true. However, the algorithmic unsolvability of some decision problems is much more surprising. A famous example of this is. . .

Hilbert's 10th Problem

Give an algorithm which, when started with any Diophantine equation, determines in a finite number of operations whether or not there are natural numbers satisfying the equation.

One of a number of important open problems listed by Hilbert at the International Congress of Mathematicians in 1900.

Computation Theory, L 1

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Diophantine equations

$$p(x_1,\ldots,x_n)=q(x_1,\ldots,x_n)$$

where p and q are polynomials in unknowns x_1, \ldots, x_n with coefficients from $\mathbb{N} = \{0, 1, 2, \ldots\}$.

Named after Diophantus of Alexandria (c. 250AD).

Example: "find three whole numbers x_1 , x_2 and x_3 such that the product of any two added to the third is a square" [Diophantus' Arithmetica, Book III, Problem 7].

In modern notation: find $x_1, x_2, x_2 \in \mathbb{N}$ for which there exists $x, y, z \in \mathbb{N}$ with

$$x_1^2x_2^2 + x_2^2x_3^2 + x_3^2x_1^2 + \dots = x^2x_1x_2 + y^2x_2x_3 + z^2x_3x_1 + \dots$$

[One solution:
$$(x_1, x_2, x_3) = (1, 4, 12)$$
, with $(x, y, z) = (4, 7, 4)$.]

Computation Theory, L 1

Hilbert's 10th Problem

Give an algorithm which, when started with any Diophantine equation, determines in a finite number of operations whether or not there are natural numbers satisfying the equation.

- ▶ Posed in 1900, but only solved in 1990: Y Matijasevič, J Robinson, M Davis and H Putnam show it undecidable by reduction to the Halting Problem.
- ▶ Original proof used Turing machines. Later, simpler proof [JP Jones & Y Matijasevič, J. Symb. Logic 49(1984)] used Minsky and Lambek's register machines—we will use them in this course to begin with and return to Turing and Church's formulations of the notion of "algorithm" later.

Computation Theory , L 1 16/171

Register Machines

Computation Theory , L 2 17/17

Algorithms, informally

No precise definition of "algorithm" at the time Hilbert posed the *Entscheidungsproblem*, just examples.

Common features of the examples:

- finite description of the procedure in terms of elementary operations
- deterministic (next step uniquely determined if there is one)
- procedure may not terminate on some input data, but we can recognize when it does terminate and what the result is.

Register Machines, informally

They operate on natural numbers $\mathbb{N} = \{0, 1, 2, ...\}$ stored in (idealized) registers using the following "elementary operations":

- add 1 to the contents of a register
- ▶ test whether the contents of a register is 0
- subtract 1 from the contents of a register if it is non-zero
- jumps ("goto")
- conditionals ("if_then_else_")

Computation Theory, L 2

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Definition. A register machine is specified by:

- ▶ finitely many registers R_0 , R_1 , ..., R_n (each capable of storing a natural number);
- a program consisting of a finite list of instructions of the form label:body, where for i=0,1,2,..., the $(i+1)^{\text{th}}$ instruction has label L_i .

Instruction body takes one of three forms:

| $R^+ ightarrow L'$ | add ${f 1}$ to contents of register ${m R}$ and jump to instruction labelled ${m L'}$ |
|-----------------------------|---|
| D- 1/1// | if contents of R is > 0 , then subtract |
| $R^- ightarrow L'$, L'' | ${f 1}$ from it and jump to L' , else jump to L'' |
| HALT | stop executing instructions |

Computation Theory , L 2

Example

registers:

 $R_0 R_1 R_2$ program:

$$\texttt{L}_0: \texttt{R}_1^- \to \texttt{L}_1, \texttt{L}_2$$

$$L_1: R_0^{\overline{+}} \rightarrow L_0$$

$$L_{2}: \mathbb{R}_{2}^{-} \to L_{3}, L_{4}$$
 $L_{3}: \mathbb{R}_{0}^{+} \to L_{2}$

$$L_3: R_0^+ \rightarrow L_2$$

$$L_4$$
: HALT

example computation:

| L_i | R_0 | R_1 | R_2 |
|-------|-------|-------|-------|
| 0 | 0 | 1 | 2 |
| 1 | 0 | 0 | 2 |
| 0 | 1 | 0 | 2 |
| 2 | 1 | 0 | 2 |
| 3 | 1 | 0 | 1 |
| 2 | 2 | 0 | 1 |
| 3 | 2 | 0 | 0 |
| 2 | 3 | 0 | 0 |
| 4 | 3 | 0 | 0 |

Computation Theory , L 2

Register machine computation

Register machine configuration:

$$c = (\ell, r_0, \ldots, r_n)$$

where $\ell =$ current label and $r_i =$ current contents of R_i .

Notation: " $R_i = x$ [in configuration c]" means $c=(\ell,r_0,\ldots,r_n)$ with $r_i=x$.

Initial configurations:

$$c_0 = (0, r_0, \ldots, r_n)$$

where r_i = initial contents of register R_i .

Register machine computation

A computation of a RM is a (finite or infinite) sequence of configurations

$$c_0, c_1, c_2, \ldots$$

where

- $ightharpoonup c_0 = (0, r_0, \dots, r_n)$ is an initial configuration
- each $c = (\ell, r_0, \dots, r_n)$ in the sequence determines the next configuration in the sequence (if any) by carrying out the program instruction labelled L_ℓ with registers containing r_0, \dots, r_n .

Computation Theory , L 2

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Halting

For a finite computation c_0, c_1, \ldots, c_m , the last configuration $c_m = (\ell, r, \ldots)$ is a halting configuration, i.e. instruction labelled L_ℓ is

E.g.
$$\begin{bmatrix} L_0 : R_0^+ \to L_2 \\ L_1 : HALT \end{bmatrix}$$
 halts erroneously.

Halting

For a finite computation c_0, c_1, \ldots, c_m , the last configuration $c_m = (\ell, r, \ldots)$ is a halting configuration.

Note that computations may never halt. For example,

$$L_0: \mathbb{R}_0^+ \to L_0$$
 $L_1: \text{HALT}$

only has infinite computation sequences

$$(0,r), (0,r+1), (0,r+2), \dots$$

Computation Theory , L 2 25/171

Graphical representation

- one node in the graph for each instruction
- arcs represent jumps between instructions
- ▶ lose sequential ordering of instructions—so need to indicate initial instruction with START.

| instruction | representation |
|--------------------|-------------------------------|
| $R^+ ightarrow L$ | $R^+ {\longrightarrow} [L]$ |
| $R^-	o L$, L' | $R^ [L']$ |
| HALT | HALT |
| L_0 | $START \longrightarrow [L_0]$ |

Example

registers:

 $\mathsf{R}_0 \ \mathsf{R}_1 \ \mathsf{R}_2$

program:

$$\texttt{L}_0: \texttt{R}_1^- \to \texttt{L}_1, \texttt{L}_2$$

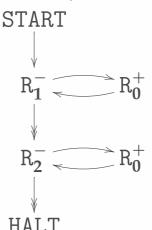
$$L_1: \mathbb{R}_0^{\widehat{+}} o L_0$$

$$L_2: R_2^{\stackrel{\circ}{-}} \rightarrow L_3$$
, L_4

$$L_3: \mathbb{R}_0^{\overline{+}} \to L_2$$

 L_4 : HALT

graphical representation:



Claim: starting from initial configuration (0, 0, x, y), this machine's computation halts with configuration (4, x + y, 0, 0).

Computation Theory, L 2

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Partial functions

Register machine computation is deterministic: in any non-halting configuration, the next configuration is uniquely determined by the program.

So the relation between initial and final register contents defined by a register machine program is a partial function...

Definition. A partial function from a set X to a set Y is specified by any subset $f \subseteq X \times Y$ satisfying

$$(x,y) \in f \land (x,y') \in f \rightarrow y = y'$$

for all $x \in X$ and $y, y' \in Y$.

Partial functions

ordered pairs $\{(x,y) \mid x \in X \land y \in Y\}$

i.e. for all $x \in X$ there is at most one $y \in Y$ with $(x,y) \in f$

Definition. A partial function from a set X to a set Y is specified by any subset $f \subseteq X \times Y$ satisfying

$$\rightarrow$$
 $(x,y) \in f \land (x,y') \in f \rightarrow y = y'$

for all $x \in X$ and $y, y' \in Y$.

Computation Theory, L 2

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Partial functions

Notation:

- "f(x) = y" means $(x, y) \in f$
- " $f(x)\downarrow$ " means $\exists y \in Y (f(x) = y)$
- "f(x)\"" means $\neg \exists y \in Y (f(x) = y)$
- ▶ $X \rightarrow Y$ = set of all partial functions from X to Y $X \rightarrow Y$ = set of all (total) functions from X to Y

Definition. A partial function from a set X to a set Y is total if it satisfies

$$f(x) \downarrow$$

for all $x \in X$.

Computable functions

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is (register machine) computable if there is a register machine $oldsymbol{M}$ with at least n+1 registers R_0 , R_1 , ..., R_n (and maybe more) such that for all $(x_1, \ldots, x_n) \in \mathbb{N}^n$ and all $y \in \mathbb{N}$, the computation of M starting with $\mathbb{R}_0 = \mathbf{0}$, $R_1 = x_1, \ldots, R_n = x_n$ and all other registers set to 0, halts with $R_0 = y$ if and only if $f(x_1, \ldots, x_n) = y$.

Note the [somewhat arbitrary] I/O convention: in the initial configuration registers R_1, \ldots, R_n store the function's arguments (with all others zeroed); and in the halting configuration register R_0 stores it's value (if any).

Computation Theory , L 2

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Example

registers: R_0 R_1 R_2

program:

$$L_0: \mathbb{R}_1^- \to L_1, L_2$$

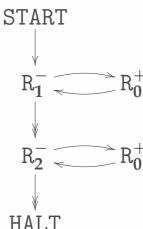
$$L_1: \mathbb{R}_0^+ \to L_0$$

$$\begin{array}{c} \mathtt{L_2:R_2^-} \rightarrow \mathtt{L_3,L_4} \\ \mathtt{L_3:R_0^+} \rightarrow \mathtt{L_2} \end{array}$$

$$L_3: \mathbb{R}_0^+ \to L_2$$

L₄: HALT

graphical representation:



Claim: starting from initial configuration (0,0,x,y), this machine's computation halts with configuration (4, x + y, 0, 0). So $f(x, y) \triangleq x + y$ is computable.

Computable functions

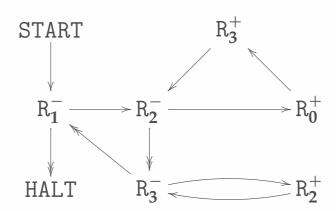
Recall:

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is (register machine) computable if there is a register machine M with at least n+1 registers R_0, R_1, \ldots, R_n (and maybe more) such that for all $(x_1, \ldots, x_n) \in \mathbb{N}^n$ and all $y \in \mathbb{N}$, the computation of M starting with $R_0 = 0$, $R_1 = x_1, \ldots, R_n = x_n$ and all other registers set to 0, halts with $R_0 = y$

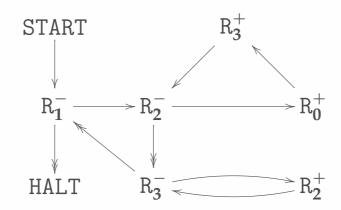
if and only if $f(x_1, \ldots, x_n) = y$.

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Multiplication $f(x,y) \triangleq xy$ is computable



Multiplication $f(x,y) \triangleq xy$ is computable



If the machine is started with $(R_0, R_1, R_2, R_3) = (0, x, y, 0)$, it halts with $(R_0, R_1, R_2, R_3) = (xy, 0, y, 0)$.

Computation Theory, L 3

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Further examples

The following arithmetic functions are all computable. (Proof—left as an exercise!)

projection: $p(x,y) \triangleq x$

constant: $c(x) \stackrel{\triangle}{=} n$

truncated subtraction: $x - y \triangleq \begin{cases} x - y & \text{if } y \leq x \\ 0 & \text{if } y > x \end{cases}$

Further examples

The following arithmetic functions are all computable. (Proof—left as an exercise!)

integer division:

$$x div y \triangleq \begin{cases} integer \ part \ of \ x/y & \text{if } y > 0 \\ 0 & \text{if } y = 0 \end{cases}$$

integer remainder: $x \mod y \triangleq x - y(x \operatorname{div} y)$

exponentiation base 2: $e(x) \triangleq 2^x$

logarithm base 2:

$$\log_2(x) \triangleq \begin{cases} greatest \ y \ such \ that \ 2^y \le x & \text{if } x > 0 \\ 0 & \text{if } x = 0 \end{cases}$$

Computation Theory, L 3

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Coding Programs as Numbers

Computation Theory , L 3

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Turing/Church solution of the Etscheidungsproblem uses the idea that (formal descriptions of) algorithms can be the data on which algorithms act.

To realize this idea with Register Machines we have to be able to code RM programs as numbers. (In general, such codings are often called Gödel numberings.)

To do that, first we have to code pairs of numbers and lists of numbers as numbers. There are many ways to do that. We fix upon one...

Computation Theory , L 3

Numerical coding of pairs

For
$$x, y \in \mathbb{N}$$
, define $\left\{ \begin{array}{l} \langle \langle x, y \rangle \rangle & \triangleq & 2^x(2y+1) \\ \langle x, y \rangle & \triangleq & 2^x(2y+1)-1 \end{array} \right.$

So

$$\begin{array}{c|c}
0b\langle\langle x,y\rangle\rangle &=& 0by & 1 & 0 \cdots 0 \\
\hline
0b\langle x,y\rangle &=& 0by & 0 & 1 \cdots 1
\end{array}$$

(Notation: $0bx \triangleq x \text{ in binary.}$)

E.g.
$$27 = 0$$
b $11011 = \langle 0, 13 \rangle = \langle 2, 3 \rangle$

Computation Theory, L 3

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Numerical coding of pairs

For
$$x,y \in \mathbb{N}$$
, define $\left\{ \begin{array}{ll} \langle\!\langle x,y \rangle\!\rangle & \triangleq & 2^x(2y+1) \\ \langle x,y \rangle & \triangleq & 2^x(2y+1)-1 \end{array} \right.$

So

 $\langle -, - \rangle$ gives a bijection (one-one correspondence) between $\mathbb{N} \times \mathbb{N}$ and \mathbb{N} .

 $\langle -, - \rangle$ gives a bijection between $\mathbb{N} \times \mathbb{N}$ and $\{n \in \mathbb{N} \mid n \neq 0\}$.

Numerical coding of lists

 $list \mathbb{N} \triangleq set$ of all finite lists of natural numbers, using ML notation for lists:

- ► empty list: []
- ▶ list-cons: $x :: \ell \in list \mathbb{N}$ (given $x \in \mathbb{N}$ and $\ell \in list \mathbb{N}$)
- $| [x_1, x_2, \ldots, x_n] \triangleq x_1 :: (x_2 :: (\cdots x_n :: [] \cdots))$

Computation Theory , L 3

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Numerical coding of lists

 $list \mathbb{N} \triangleq set of all finite lists of natural numbers, using ML notation for lists.$

For $\ell \in \mathit{list}\,\mathbb{N}$, define $\lceil \ell \rceil \in \mathbb{N}$ by induction on the length of the list ℓ :

$$\left\{ \begin{array}{c} \lceil \rceil \rceil \triangleq 0 \\ \lceil x :: \ell \rceil \triangleq \langle \langle x, \lceil \ell \rceil \rangle \rangle = 2^x (2 \cdot \lceil \ell \rceil + 1) \end{array} \right.$$

Thus
$$\lceil [x_1, x_2, \ldots, x_n] \rceil = \langle \langle x_1, \langle \langle x_2, \cdots \langle \langle x_n, 0 \rangle \rangle \cdots \rangle \rangle \rangle$$

Computation Theory , L 3

Numerical coding of lists

 $list \mathbb{N} \triangleq set$ of all finite lists of natural numbers, using ML notation for lists.

For $\ell \in list \mathbb{N}$, define $\lceil \ell \rceil \in \mathbb{N}$ by induction on the length of the list ℓ :

$$\left\{ \begin{array}{ccc} & \lceil [\rceil \rceil & \triangleq & \mathbf{0} \\ & \lceil x :: \ell \rceil & \triangleq & \langle \langle x, \lceil \ell \rceil \rangle \rangle = 2^x (2 \cdot \lceil \ell \rceil + 1) \end{array} \right.$$

$$\boxed{0} \text{b} \lceil [x_1, x_2, \dots, x_n] \rceil = \boxed{1} \boxed{0 \cdots 0} \boxed{1} \boxed{0 \cdots 0} \cdots \boxed{1} \boxed{0 \cdots 0}$$

Hence $\ell \mapsto \lceil \ell \rceil$ gives a bijection from $list \mathbb{N}$ to \mathbb{N} .

Computation Theory , L 3 44/171

Numerical coding of lists

 $list \mathbb{N} \triangleq set$ of all finite lists of natural numbers, using ML notation for lists.

For $\ell \in list \mathbb{N}$, define $\lceil \ell \rceil \in \mathbb{N}$ by induction on the length of the list ℓ :

$$\left\{ \begin{array}{c} \lceil \rceil \rceil \triangleq 0 \\ \lceil x :: \ell \rceil \triangleq \langle \langle x, \lceil \ell \rceil \rangle \rangle = 2^x (2 \cdot \lceil \ell \rceil + 1) \end{array} \right.$$

For example:

$$\lceil [3] \rceil = \lceil 3 :: [] \rceil = \langle (3,0) \rangle = 2^{3}(2 \cdot 0 + 1) = 8 = 0 \cdot 1000$$

$$\lceil [1,3] \rceil = \langle (1,\lceil [3] \rceil) \rangle = \langle (1,8) \rangle = 34 = 0 \cdot 100010$$

$$\lceil [2,1,3] \rceil = \langle (2,\lceil [1,3] \rceil) \rangle = \langle (2,34) \rangle = 276 = 0 \cdot 100010100$$

Computation Theory , L 3

Numerical coding of programs

then its numerical code is

$$\lceil P \rceil \triangleq \lceil \lceil body_0 \rceil, \ldots, \lceil body_n \rceil \rceil$$

where the numerical code $\lceil body \rceil$ of an instruction body

Computation Theory, L 3

Any $x \in \mathbb{N}$ decodes to a unique instruction body(x):

```
if x = 0 then body(x) is HALT,
else (x > 0 \text{ and}) let x = \langle \langle y, z \rangle \rangle in
   if y = 2i is even, then
      \overline{body}(x) is \mathbb{R}_i^+ \to \mathbb{L}_z,
   else y = 2i + 1 is odd, let z = \langle j, k \rangle in
       body(x) is R_i^- \to L_i, L_k
```

So any $e \in \mathbb{N}$ decodes to a unique program prog(e), called the register machine program with index e:

$$prog(e) riangleq egin{bmatrix} \mathbb{L}_0 : body(x_0) \ dots \ \mathbb{L}_n : body(x_n) \end{bmatrix}$$
 where $e = \lceil [x_0, \ldots, x_n]
ceil$

Example of prog(e)

▶
$$786432 = 2^{19} + 2^{18} = 0$$
b $110...0$ = $\lceil [18, 0] \rceil$

▶
$$18 = 0$$
b $10010 = \langle \langle 1, 4 \rangle \rangle = \langle \langle 1, \langle 0, 2 \rangle \rangle \rangle = \lceil \mathsf{R}_0^- \to \mathsf{L}_0, \mathsf{L}_2 \rceil$

▶
$$0 = \lceil \text{HALT} \rceil$$

So
$$prog(786432) = \begin{bmatrix} \texttt{L}_0: \texttt{R}_0^- o \texttt{L}_0, \texttt{L}_2 \\ \texttt{L}_1: \texttt{HALT} \end{bmatrix}$$

N.B. In case e=0 we have $0=\lceil [\rceil \rceil$, so prog(0) is the program with an empty list of instructions, which by convention we regard as a RM that does nothing (i.e. that halts immediately).

Computation Theory , L 3

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Universal Register Machine, *U*

Computation Theory , L 4

High-level specification

Universal RM U carries out the following computation, starting with $R_0=0$, $R_1=e$ (code of a program), $R_2=a$ (code of a list of arguments) and all other registers zeroed:

- ightharpoonup decode e as a RM program P
- ightharpoonup decode a as a list of register values a_1, \ldots, a_n
- riangleright carry out the computation of the RM program P starting with $R_0 = 0$, $R_1 = a_1, \ldots, R_n = a_n$ (and any other registers occurring in P set to 0).

Mnemonics for the registers of \boldsymbol{U} and the role they play in its program:

 $\mathbb{R}_1 \equiv \mathbb{P}$ code of the RM to be simulated

 ${\tt R}_2 \equiv {\tt A}$ code of current register contents of simulated RM

 $R_3 \equiv PC$ program counter—number of the current instruction (counting from 0)

 $\mathbb{R}_4 \equiv \mathbb{N}$ code of the current instruction body

 $R_5 \equiv C$ type of the current instruction body

 $R_6 \equiv R$ current value of the register to be incremented or decremented by current instruction (if not HALT)

 $R_7 \equiv S$, $R_8 \equiv T$ and $R_9 \equiv Z$ are auxiliary registers.

 R_0 result of the simulated RM computation (if any).

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Overall structure of U's program

- $\boxed{1}$ copy PCth item of list in P to N (halting if PC > length of list); goto $\boxed{2}$
- 2 if $\mathbb{N} = 0$ then halt, else decode \mathbb{N} as $\langle y, z \rangle$; $\mathbb{C} := y$; $\mathbb{N} := z$; goto 3

{at this point either C=2i is even and current instruction is $R_i^+ \to L_z$, or C=2i+1 is odd and current instruction is $R_i^- \to L_j$, L_k where $z=\langle j,k\rangle$ }

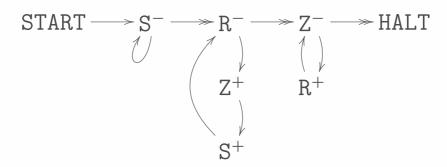
- 3 copy *i*th item of list in A to R; goto $\boxed{4}$
- 4 execute current instruction on \mathbb{R} ; update PC to next label; restore register values to \mathbb{A} ; goto $\boxed{1}$

To implement this, we need RMs for manipulating (codes of) lists of numbers. . .

Computation Theory, L 4

The program START \rightarrow S ::= R \rightarrow HALT

to copy the contents of R to S can be implemented by



precondition:

$$R = x$$

$$S = y$$

$$Z = 0$$

postcondition:

$$R = x$$

$$S = x$$

$$Z = 0$$

Computation Theory, L 4

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The program START $\rightarrow \begin{vmatrix} push & X \\ to & L \end{vmatrix} \rightarrow HALT$

to carry out the assignment (X, L) := (0, X :: L) can be implemented by

$$START \longrightarrow Z^{+} \xrightarrow{\qquad \qquad } L^{-} \xrightarrow{\gg} Z^{-} \xrightarrow{\gg} X^{-} \xrightarrow{\gg} HALT$$

$$\downarrow \qquad \qquad \downarrow \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \downarrow \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad$$

precondition:

$$X = x$$

$$L = \ell$$

$$Z = 0$$

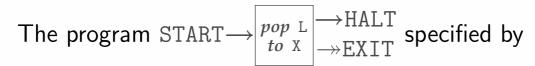
postcondition:

$$X = 0$$

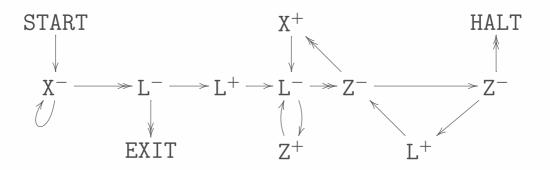
$$L = \langle \langle x, \ell \rangle \rangle = 2^{x} (2\ell + 1)$$

$$Z = 0$$

$$z = 0$$



"if L = 0 then (X := 0; goto EXIT) else let $L = \langle \langle x, \ell \rangle \rangle$ in $(X := x; L := \ell; goto HALT)$ " can be implemented by



Computation Theory , L 4

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Overall structure of U's program

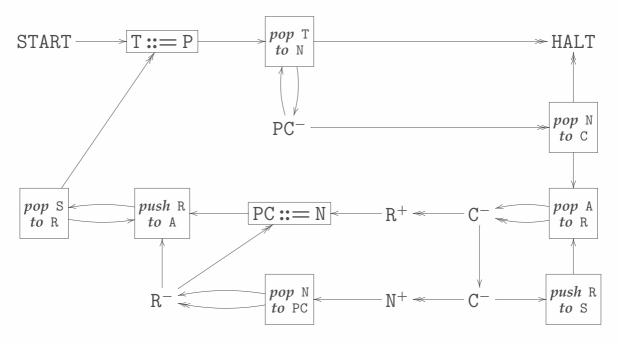
- $\boxed{1}$ copy PCth item of list in P to N (halting if PC > length of list); goto $\boxed{2}$
- 2 if $\mathbb{N} = 0$ then halt, else decode \mathbb{N} as $\langle y, z \rangle$; $\mathbb{C} := y$; $\mathbb{N} := z$; goto 3

{at this point either C=2i is even and current instruction is $R_i^+ \to L_z$, or C=2i+1 is odd and current instruction is $R_i^- \to L_j$, L_k where $z=\langle j,k\rangle$ }

- 3 copy *i*th item of list in A to R; goto $\boxed{4}$
- 4 execute current instruction on R; update PC to next label; restore register values to A; goto $\boxed{1}$

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The program for \boldsymbol{U}



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The Halting Problem

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Definition. A register machine H decides the Halting Problem if for all $e, a_1, \ldots, a_n \in \mathbb{N}$, starting H with

$$R_0 = 0$$
 $R_1 = e$ $R_2 = \lceil [a_1, \dots, a_n] \rceil$

and all other registers zeroed, the computation of H always halts with R_0 containing $\mathbf{0}$ or $\mathbf{1}$; moreover when the computation halts, $R_0 = \mathbf{1}$ if and only if

the register machine program with index e eventually halts when started with $R_0 = 0$, $R_1 = a_1, \ldots, R_n = a_n$ and all other registers zeroed.

Theorem. No such register machine H can exist.

Proof of the theorem

Assume we have a RM \boldsymbol{H} that decides the Halting Problem and derive a contradiction, as follows:

▶ Let H' be obtained from H by replacing START → by START → $Z := R_1$ → $\begin{bmatrix} push & Z \\ to & R_2 \end{bmatrix}$ → (where Z is a register not mentioned in H's program).

- Let C be obtained from H' by replacing each HALT (& each erroneous halt) by $\longrightarrow \mathbb{R}_0^- \longrightarrow \mathbb{R}_0^+$.
- ▶ Let $c \in \mathbb{N}$ be the index of C's program.

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Proof of the theorem

Assume we have a RM \boldsymbol{H} that decides the Halting Problem and derive a contradiction, as follows:

C started with $R_1=c$ eventually halts if & only if H' started with $R_1=c$ halts with $R_0=0$ if & only if H started with $R_1=c$, $R_2=\lceil [c] \rceil$ halts with $R_0=0$ if & only if $R_1=c$ does not halt if & only if $R_1=c$ does not halt $R_1=c$ does not halt

Computable functions

Recall:

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is (register machine) computable if there is a register machine M with at least n+1 registers R_0, R_1, \ldots, R_n (and maybe more) such that for all $(x_1, \ldots, x_n) \in \mathbb{N}^n$ and all $y \in \mathbb{N}$, the computation of M starting with $R_0 = 0$,

the computation of M starting with $R_0 = 0$, $R_1 = x_1, \ldots, R_n = x_n$ and all other registers set to 0, halts with $R_0 = y$

if and only if $f(x_1,\ldots,x_n)=y$.

Note that the same RM M could be used to compute a unary function (n=1), or a binary function (n=2), etc. From now on we will concentrate on the unary case. . .

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Enumerating computable functions

For each $e \in \mathbb{N}$, let $\varphi_e \in \mathbb{N} \rightarrow \mathbb{N}$ be the unary partial function computed by the RM with program prog(e). So for all $x, y \in \mathbb{N}$:

 $\varphi_e(x) = y$ holds iff the computation of prog(e) started with $R_0 = 0$, $R_1 = x$ and all other registers zeroed eventually halts with $R_0 = y$.

Thus

$$e \mapsto \varphi_e$$

defines an <u>onto</u> function from $\mathbb N$ to the collection of all computable partial functions from $\mathbb N$ to $\mathbb N$.

An uncomputable function

Let $f \in \mathbb{N} \to \mathbb{N}$ be the partial function with graph $\{(x,0) \mid \varphi_x(x) \uparrow \}$.

Thus
$$f(x) = \begin{cases} 0 & \varphi_x(x) \uparrow \\ undefined & \varphi_x(x) \downarrow \end{cases}$$

f is not computable, because if it were, then $f=\varphi_e$ for some $e\in\mathbb{N}$ and hence

- ▶ if $\varphi_e(e)\uparrow$, then f(e)=0 (by def. of f); so $\varphi_e(e)=0$ (by def. of e), i.e. $\varphi_e(e)\downarrow$
- ▶ if $\varphi_e(e)\downarrow$, then $f(e)\uparrow$ (by def. of e); so $\varphi_e(e)\uparrow$ (by def. of f)

—contradiction! So f cannot be computable.

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(Un)decidable sets of numbers

Given a subset $S \subseteq \mathbb{N}$, its characteristic function

$$\chi_S \in \mathbb{N} o \mathbb{N}$$
 is given by: $\chi_S(x) ext{ } ex$

(Un)decidable sets of numbers

Definition. $S \subseteq \mathbb{N}$ is called (register machine) decidable if its characteristic function $\chi_S \in \mathbb{N} \to \mathbb{N}$ is a register machine computable function. Otherwise it is called undecidable.

So S is decidable iff there is a RM M with the property: for all $x \in \mathbb{N}$, M started with $R_0 = 0$, $R_1 = x$ and all other registers zeroed eventually halts with R_0 containing $\mathbf{1}$ or $\mathbf{0}$; and $R_0 = \mathbf{1}$ on halting iff $x \in S$.

Basic strategy: to prove $S \subseteq \mathbb{N}$ undecidable, try to show that decidability of S would imply decidability of the Halting Problem.

For example...

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Claim: $S_0 \triangleq \{e \mid \varphi_e(0) \downarrow \}$ is undecidable.

Proof (sketch): Suppose M_0 is a RM computing χ_{S_0} . From M_0 's program (using the same techniques as for constructing a universal RM) we can construct a RM H to carry out:

let
$$e = R_1$$
 and $\lceil [a_1, \ldots, a_n] \rceil = R_2$ in $R_1 := \lceil (R_1 := a_1); \cdots; (R_n := a_n); prog(e) \rceil;$ $R_2 := 0;$ run M_0

Then by assumption on M_0 , H decides the Halting Problem—contradiction. So no such M_0 exists, i.e. χ_{S_0} is uncomputable, i.e. S_0 is undecidable.

Claim: $S_1 \triangleq \{e \mid \varphi_e \text{ a total function}\}$ is undecidable.

Proof (sketch): Suppose M_1 is a RM computing χ_{S_1} . From M_1 's program we can construct a RM M_0 to carry out:

let
$$e = R_1$$
 in $R_1 := \lceil R_1 := 0$; $prog(e) \rceil$; run M_1

Then by assumption on M_1 , M_0 decides membership of S_0 from previous example (i.e. computes χ_{S_0})—contradiction. So no such M_1 exists, i.e. χ_{S_1} is uncomputable, i.e. S_1 is undecidable.

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Turing Machines

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Algorithms, informally

No precise definition of "algorithm" at the time Hilbert posed the *Entscheidungsproblem*, just examples.

Common features of the examples:

- ► finite description of the procedure in terms of elementary operations ←
- deterministic (next step uniquely determined if there is one)
- procedure may not terminate on some input data, but we can recognize when it does terminate and what the result is.

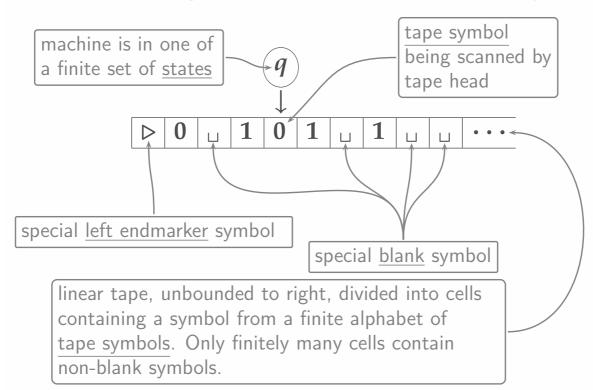
e.g. multiply two decimal digits by looking up their product in a table

Register Machine computation abstracts away from any particular, concrete representation of numbers (e.g. as bit strings) and the associated elementary operations of increment/decrement/zero-test.

Turing's original model of computation (now called a Turing machine) is more concrete: even numbers have to be represented in terms of a fixed <u>finite</u> alphabet of symbols and increment/decrement/zero-test programmed in terms of more elementary symbol-manipulating operations.

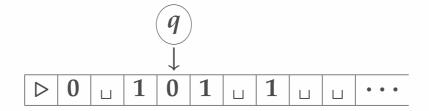
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Turing machines, informally



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Turing machines, informally



- Machine starts with tape head pointing to the special left endmarker ▷.
- Machine computes in discrete steps, each of which depends only on current state (q) and symbol being scanned by tape head (0).
- Action at each step is to overwrite the current tape cell with a symbol, move left or right one cell, or stay stationary, and change state.

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Turing Machines

are specified by:

- Q, finite set of machine states
- ▶ Σ , finite set of tape symbols (disjoint from Q) containing distinguished symbols \triangleright (left endmarker) and \sqcup (blank)
- $ightharpoonup s \in Q$, an initial state
- ▶ $\delta \in (Q \times \Sigma) \rightarrow (Q \cup \{\text{acc,rej}\}) \times \Sigma \times \{L,R,S\}$, a transition function, satisfying:

for all
$$q \in Q$$
, there exists $q' \in Q \cup \{acc, rej\}$ with $\delta(q, \triangleright) = (q', \triangleright, R)$

(i.e. left endmarker is never overwritten and machine always moves to the right when scanning it)

Example Turing Machine

$$M=(Q,\Sigma,s,\delta)$$
 where states $Q=\{s,q,q'\}$ $(s \ ext{initial})$ symbols $\Sigma=\{ riangle, \sqcup, 0, 1\}$

transition function

$$\delta \in (Q \times \Sigma) \rightarrow (Q \cup \{\text{acc,rej}\}) \times \Sigma \times \{L,R,S\}$$
:

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Turing machine computation

Turing machine configuration: (q, w, u) where

- ▶ $q \in Q \cup \{acc, rej\} = current state$
- w = non-empty string (w = va) of tape symbols under and to the left of tape head, whose last element (a) is contents of cell under tape head
- u = (possibly empty) string of tape symbols to the right of tape head (up to some point beyond which all symbols are \Box)

Initial configurations: (s, \triangleright, u)

Turing machine computation

Given a TM $M = (Q, \Sigma, s, \delta)$, we write

$$(q, w, u) \rightarrow_M (q', w', u')$$

to mean $q \neq acc, rej, w = va$ (for some v, a) and

either
$$\delta(q,a)=(q',a',L)$$
, $w'=v$, and $u'=a'u$ or $\delta(q,a)=(q',a',S)$, $w'=va'$ and $u'=u$ or $\delta(q,a)=(q',a',R)$, $u=a''u''$ is non-empty, $w'=va'a''$ and $u'=u''$

or
$$\delta(q,a)=(q',a',R)$$
, $u=\varepsilon$ is empty, $w'=va'_{\sqcup}$ and $u'=\varepsilon$.

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Turing machine computation

A computation of a TM M is a (finite or infinite) sequence of configurations c_0, c_1, c_2, \ldots

where

- $c_0 = (s, \triangleright, u)$ is an initial configuration
- $ightharpoonup c_i
 ightharpoonup_M c_{i+1}$ holds for each $i=0,1,\ldots$

The computation

- does not halt if the sequence is infinite
- ▶ halts if the sequence is finite and its last element is of the form (acc, w, u) or (rej, w, u).

Example Turing Machine

$$M=(Q,\Sigma,s,\delta)$$
 where states $Q=\{s,q,q'\}$ $(s \ \text{initial})$ symbols $\Sigma=\{\rhd,\sqcup,0,1\}$

transition function

$$\delta \in (Q \times \Sigma) \rightarrow (Q \cup \{\text{acc,rej}\}) \times \Sigma \times \{L,R,S\}$$
:

Claim: the computation of M starting from configuration $(s, \triangleright, \bot 1^n 0)$ halts in configuration $(acc, \triangleright, \bot 1^{n+1} 0)$.

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The computation of M starting from configuration $(s, \triangleright, \sqcup 1^n 0)$:

$$(s, \triangleright, \sqcup 1^{n}0) \rightarrow_{M} (s, \triangleright, \sqcup, 1^{n}0)$$
 $\rightarrow_{M} (q, \triangleright, \sqcup, 1^{n-1}0)$
 \vdots
 $\rightarrow_{M} (q, \triangleright, \sqcup 1^{n}, 0)$
 $\rightarrow_{M} (q, \triangleright, \sqcup 1^{n}0, \varepsilon)$
 $\rightarrow_{M} (q, \triangleright, \sqcup 1^{n+1}, \varepsilon)$
 $\rightarrow_{M} (q', \triangleright, \sqcup 1^{n+1}, 0)$
 \vdots
 $\rightarrow_{M} (q', \triangleright, \sqcup, 1^{n+1}0)$
 $\rightarrow_{M} (\operatorname{acc}, \triangleright, \sqcup, 1^{n+1}0)$

Theorem. The computation of a Turing machine M can be implemented by a register machine.

Proof (sketch).

- Step 1: fix a numerical encoding of M's states, tape symbols, tape contents and configurations.
- Step 2: implement M's transition function (finite table) using RM instructions on codes.
- Step 3: implement a RM program to repeatedly carry out \longrightarrow_M .

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Step 1

Identify states and tape symbols with particular numbers:

▶ Code configurations c = (q, w, u) by:

$$\lceil c \rceil = \lceil [q, \lceil [a_n, \ldots, a_1] \rceil, \lceil [b_1, \ldots, b_m] \rceil] \rceil$$

where $w=a_1\cdots a_n\ (n>0)$ and $u=b_1\cdots b_m\ (m\geq 0)$ say.

Step 2

Using registers

Q = current state

A = current tape symbol

D = current direction of tape head (with L=0, R=1 and S=2, say)

one can turn the finite table of (argument, result)-pairs specifying δ into a RM program \rightarrow $(Q, A, D) := \delta(Q, A) \rightarrow$ so that starting the program with Q = q, A = a, D = d (and all other registers zeroed), it halts with Q = q', A = a', D = d', where $(q', a', d') = \delta(q, a)$.

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Step 3

The next slide specifies a RM to carry out M's computation. It uses registers

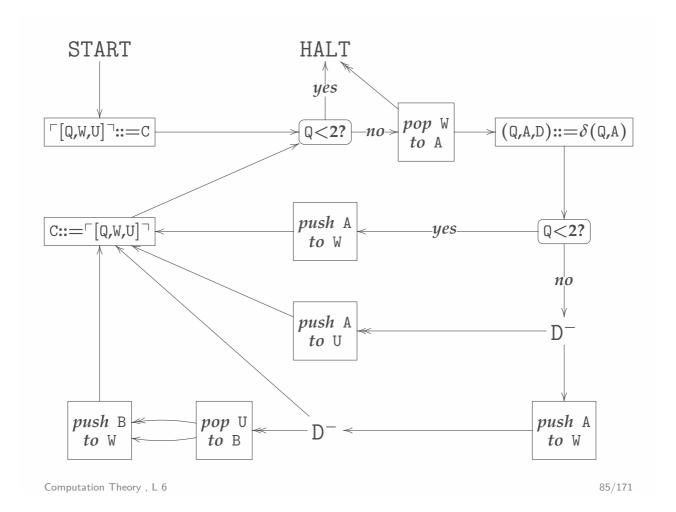
 $C = \mathsf{code}\ \mathsf{of}\ \mathsf{current}\ \mathsf{configuration}$

 $\mathbb{W}=\operatorname{code}$ of tape symbols at and left of tape head (reading right-to-left)

U = code of tape symbols right of tape head (reading left-to-right)

Starting with $\mathbb C$ containing the code of an initial configuration (and all other registers zeroed), the RM program halts if and only if M halts; and in that case $\mathbb C$ holds the code of the final configuration.

Computation Theory, L 6



Computable functions

Recall:

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is (register machine) computable if there is a register machine M with at least n+1 registers R_0, R_1, \ldots, R_n (and maybe more) such that for all $(x_1, \ldots, x_n) \in \mathbb{N}^n$ and all $y \in \mathbb{N}$, the computation of M starting with $R_0 = 0$, $R_1 = x_1, \ldots, R_n = x_n$ and all other registers set to 0, halts with $R_0 = y$ if and only if $f(x_1, \ldots, x_n) = y$.

We've seen that a Turing machine's computation can be implemented by a register machine.

The converse holds: the computation of a register machine can be implemented by a Turing machine.

To make sense of this, we first have to fix a tape representation of RM configurations and hence of numbers and lists of numbers...

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Tape encoding of lists of numbers

Definition. A tape over $\Sigma = \{\triangleright, \sqcup, 0, 1\}$ codes a list of numbers if precisely two cells contain 0 and the only cells containing 1 occur between these.

Such tapes look like:

$$\triangleright_{\square} \cdots {\scriptstyle \square} 0 \underbrace{1 \cdots 1}_{n_1} {\scriptstyle \square} \underbrace{1 \cdots 1}_{n_2} {\scriptstyle \square} \cdots {\scriptstyle \square} \underbrace{1 \cdots 1}_{n_k} 0 \underbrace{\scriptstyle \square} \cdots \underbrace{\scriptstyle \text{all}}_{\square' s}$$

which corresponds to the list $[n_1, n_2, \ldots, n_k]$.

Computation Theory, L 7

Turing computable function

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is Turing computable if and only if there is a Turing machine M with the following property:

Starting M from its initial state with tape head on the left endmarker of a tape coding $[0, x_1, \ldots, x_n]$, M halts if and only if $f(x_1, \ldots, x_n) \downarrow$, and in that case the final tape codes a list (of length ≥ 1) whose first element is y where $f(x_1, \ldots, x_n) = y$.

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Theorem. A partial function is Turing computable if and only if it is register machine computable.

Proof (sketch). We've seen how to implement any TM by a RM. Hence

f TM computable implies f RM computable.

For the converse, one has to implement the computation of a RM in terms of a TM operating on a tape coding RM configurations. To do this, one has to show how to carry out the action of each type of RM instruction on the tape. It should be reasonably clear that this is possible in principle, even if the details (omitted) are tedious.

Notions of computability

- ▶ Church (1936): λ -calculus [see later]
- ► Turing (1936): Turing machines.

Turing showed that the two very different approaches determine the same class of computable functions. Hence:

Church-Turing Thesis. Every algorithm [in intuitive sense of Lect. 1] can be realized as a Turing machine.

Computation Theory , L 7

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Notions of computability

Church-Turing Thesis. Every algorithm [in intuitive sense of Lect. 1] can be realized as a Turing machine.

Further evidence for the thesis:

- ► Gödel and Kleene (1936): partial recursive functions
- ▶ Post (1943) and Markov (1951): canonical systems for generating the theorems of a formal system
- ▶ Lambek (1961) and Minsky (1961): register machines
- ► Variations on all of the above (e.g. multiple tapes, non-determinism, parallel execution...)

All have turned out to determine the same collection of computable functions.

Computation Theory , L 7

Notions of computability

Church-Turing Thesis. Every algorithm [in intuitive sense of Lect. 1] can be realized as a Turing machine.

In rest of the course we'll look at

- ▶ Gödel and Kleene (1936): partial recursive functions (~> branch of mathematics called recursion theory)
- Church (1936): λ-calculus
 (→ branch of CS called functional programming)

Computation Theory , L 7

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Aim

A more abstract, machine-independent description of the collection of computable partial functions than provided by register/Turing machines:

they form the smallest collection of partial functions containing some basic functions and closed under some fundamental operations for forming new functions from old—composition, primitive recursion and minimization.

The characterization is due to Kleene (1936), building on work of Gödel and Herbrand.

Basic functions

▶ Projection functions, $proj_i^n \in \mathbb{N}^n \rightarrow \mathbb{N}$:

$$\operatorname{proj}_{i}^{n}(x_{1},\ldots,x_{n})\triangleq x_{i}$$

ightharpoonup Constant functions with value 0, zeroⁿ $\in \mathbb{N}^n \rightarrow \mathbb{N}$:

$$\mathsf{zero}^n(x_1,\ldots,x_n) \triangleq 0$$

▶ Successor function, $succ \in \mathbb{N} \rightarrow \mathbb{N}$:

$$succ(x) \triangleq x + 1$$

Computation Theory , L 7

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Basic functions

are all RM computable:

ightharpoonup Projection projⁿ is computed by

$$\mathtt{START} {\longrightarrow} \boxed{\mathtt{R}_0 ::= \mathtt{R}_i} {\longrightarrow} \mathtt{HALT}$$

ightharpoonup Constant zeroⁿ is computed by

$$START \longrightarrow HALT$$

Successor succ is computed by

$$\mathtt{START} {\longrightarrow} \mathtt{R}_1^+ {\longrightarrow} \boxed{\mathtt{R}_0 ::= \mathtt{R}_1} {\longrightarrow} \mathtt{HALT}$$

Composition

Composition of $f \in \mathbb{N}^n \to \mathbb{N}$ with $g_1, \dots, g_n \in \mathbb{N}^m \to \mathbb{N}$ is the partial function $f \circ [g_1, \dots, g_n] \in \mathbb{N}^m \to \mathbb{N}$ satisfying

$$f \circ [g_1, \ldots, g_n](x_1, \ldots, x_m) \equiv f(g_1(x_1, \ldots, x_m), \ldots, g_n(x_1, \ldots, x_m))$$

where \equiv is "Kleene equivalence" of possibly-undefined expressions: LHS \equiv RHS means "either both LHS and RHS are undefined, or they are both defined and are equal."

Computation Theory, L 7

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Composition

Composition of $f \in \mathbb{N}^n \to \mathbb{N}$ with $g_1, \dots, g_n \in \mathbb{N}^m \to \mathbb{N}$ is the partial function $f \circ [g_1, \dots, g_n] \in \mathbb{N}^m \to \mathbb{N}$ satisfying

$$f \circ [g_1, \ldots, g_n](x_1, \ldots, x_m) \equiv f(g_1(x_1, \ldots, x_m), \ldots, g_n(x_1, \ldots, x_m))$$

So $f \circ [g_1, \ldots, g_n](x_1, \ldots, x_m) = z$ iff there exist y_1, \ldots, y_m with $g_i(x_1, \ldots, x_m) = y_i$ (for i = 1...m) and $f(y_1, \ldots, y_m) = z$.

N.B. in case n=1, we write $f \circ g_1$ for $f \circ [g_1]$.

Composition

$f \circ [g_1, \ldots, g_n]$ is computable if f and g_1, \ldots, g_n are.

Proof. Given RM programs $\begin{cases} F \\ G_i \end{cases}$ computing $\begin{cases} f(y_1,\ldots,y_n) \\ g_i(x_1,\ldots,x_m) \end{cases}$ in \mathbb{R}_0 starting with $\begin{cases} \mathbb{R}_1,\ldots,\mathbb{R}_n \\ \mathbb{R}_1,\ldots,\mathbb{R}_m \end{cases}$ set to $\begin{cases} y_1,\ldots,y_n \\ x_1,\ldots,x_m \end{cases}$, then the next slide specifies a RM program computing $f\circ [g_1,\ldots,g_n](x_1,\ldots,x_m)$ in \mathbb{R}_0 starting with $\mathbb{R}_1,\ldots,\mathbb{R}_m$ set to x_1,\ldots,x_m .

(**Hygiene** [caused by the lack of *local names* for registers in the RM model of computation]: we assume the programs F, G_1, \ldots, G_n only mention registers up to R_N (where $N \ge \max\{n, m\}$) and that $X_1, \ldots, X_m, Y_1, \ldots, Y_n$ are some registers R_i with i > N.)

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$$(X_{1},...,X_{m})::=(R_{1},...,R_{m}) \rightarrow G_{1} \rightarrow Y_{1}::=R_{0} \rightarrow (R_{1},...,R_{N})::=(0,...,0)$$

$$(R_{1},...,R_{m})::=(X_{1},...,X_{m}) \rightarrow G_{2} \rightarrow Y_{2}::=R_{0} \rightarrow (R_{1},...,R_{N})::=(0,...,0)$$

$$(R_{1},...,R_{m})::=(X_{1},...,X_{m}) \rightarrow G_{n} \rightarrow Y_{n}::=R_{0} \rightarrow (R_{1},...,R_{N})::=(0,...,0)$$

$$(R_{1},...,R_{m})::=(Y_{1},...,Y_{n}) \rightarrow F$$

Partial Recursive Functions

Computation Theory , L 8

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Aim

A more abstract, machine-independent description of the collection of computable partial functions than provided by register/Turing machines:

they form the smallest collection of partial functions containing some basic functions and closed under some fundamental operations for forming new functions from old—composition, primitive recursion and minimization.

The characterization is due to Kleene (1936), building on work of Gödel and Herbrand.

Computation Theory , L $8\,$

Examples of recursive definitions

$$\begin{cases} f_1(0) & \equiv 0 \\ f_1(x+1) & \equiv f_0(x) + (x+1) \end{cases}$$

$$\begin{cases} f_2(0) & \equiv 0 \\ f_2(1) & \equiv 1 \\ f_2(x+2) & \equiv f_2(x) + f_2(x+1) \end{cases}$$

$$\begin{cases} f_3(0) & \equiv 0 \\ f_3(x+1) & \equiv f_3(x+2) + 1 \end{cases}$$

$$f_4(x) = if \ x > 100 \ then \ x - 10 \ else \ f_4(f_4(x+11))$$

$$\begin{cases} f_4(x) = \sup_{x \in A} f_1(x) = \sup_{x \in A} f_2(x) = \sup_{x \in A} f_2(x) = \inf_{x \in A} f_3(x) = \inf_{x \in A} f_4(x) = \inf$$

Primitive recursion

Theorem. Given $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ and $g \in \mathbb{N}^{n+2} \rightarrow \mathbb{N}$, there is a unique $h \in \mathbb{N}^{n+1} \rightarrow \mathbb{N}$ satisfying

$$\begin{cases} h(\vec{x},0) & \equiv f(\vec{x}) \\ h(\vec{x},x+1) & \equiv g(\vec{x},x,h(\vec{x},x)) \end{cases}$$

for all $\vec{x} \in \mathbb{N}^n$ and $x \in \mathbb{N}$.

We write $\rho^n(f,g)$ for h and call it the partial function defined by primitive recursion from f and g.

Primitive recursion

Theorem. Given $f \in \mathbb{N}^n \longrightarrow \mathbb{N}$ and $g \in \mathbb{N}^{n+2} \longrightarrow \mathbb{N}$, there is a unique $h \in \mathbb{N}^{n+1} \longrightarrow \mathbb{N}$ satisfying

$$(*) \begin{cases} h(\vec{x},0) & \equiv f(\vec{x}) \\ h(\vec{x},x+1) & \equiv g(\vec{x},x,h(\vec{x},x)) \end{cases}$$

for all $\vec{x} \in \mathbb{N}^n$ and $x \in \mathbb{N}$.

Proof (sketch). *Existence*: the set

$$h \triangleq \{(\vec{x}, x, y) \in \mathbb{N}^{n+2} \mid \exists y_0, y_1, \dots, y_x \\ f(\vec{x}) = y_0 \land (\bigwedge_{i=0}^{x-1} g(\vec{x}, i, y_i) = y_{i+1}) \land y_x = y\}$$
 defines a partial function satisfying (*).

Uniqueness: if h and h' both satisfy (*), then one can prove by induction on x that $\forall \vec{x} \ (h(\vec{x}, x) = h'(\vec{x}, x))$.

Computation Theory , L 8

Example: addition

Addition $add \in \mathbb{N}^2 \rightarrow \mathbb{N}$ satisfies:

$$\begin{cases} add(x_1,0) & \equiv x_1 \\ add(x_1,x+1) & \equiv add(x_1,x)+1 \end{cases}$$

So
$$add =
ho^1(f,g)$$
 where $egin{cases} f(x_1) & riangleq x_1 \ g(x_1,x_2,x_3) & riangleq x_3+1 \end{cases}$

Note that $f = \text{proj}_1^1$ and $g = \text{succ} \circ \text{proj}_3^3$; so add can be built up from basic functions using composition and primitive recursion: $add = \rho^1(\text{proj}_1^1, \text{succ} \circ \text{proj}_3^3)$.

Computation Theory, L 8

Example: predecessor

Predecessor $pred \in \mathbb{N} \rightarrow \mathbb{N}$ satisfies:

$$\begin{cases} pred(0) & \equiv 0 \\ pred(x+1) & \equiv x \end{cases}$$

So
$$pred =
ho^0(f,g)$$
 where $egin{cases} f() & riangleq 0 \ g(x_1,x_2) & riangleq x_1 \end{cases}$

Thus pred can be built up from basic functions using primitive recursion: $pred = \rho^0(zero^0, proj_1^2)$.

Computation Theory , L 8

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Example: multiplication

Multiplication $mult \in \mathbb{N}^2 \rightarrow \mathbb{N}$ satisfies:

$$\begin{cases} mult(x_1,0) & \equiv 0 \\ mult(x_1,x+1) & \equiv mult(x_1,x) + x_1 \end{cases}$$

and thus $add = \rho^1(\mathsf{zero}^1, add \circ (\mathsf{proj}_3^3, \mathsf{proj}_1^3))$.

So *mult* can be built up from basic functions using composition and primitive recursion (since *add* can be).

Definition. A [partial] function f is primitive recursive $(f \in \mathbf{PRIM})$ if it can be built up in finitely many steps from the basic functions by use of the operations of composition and primitive recursion.

In other words, the set **PRIM** of primitive recursive functions is the <u>smallest</u> set (with respect to subset inclusion) of partial functions containing the basic functions and closed under the operations of composition and primitive recursion.

Computation Theory , L $8\,$

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Definition. A [partial] function f is primitive recursive $(f \in \mathbf{PRIM})$ if it can be built up in finitely many steps from the basic functions by use of the operations of composition and primitive recursion.

Every $f \in \mathbf{PRIM}$ is a total function, because:

- all the basic functions are total
- if f, g_1, \ldots, g_n are total, then so is $f \circ (g_1, \ldots, g_n)$ [why?]
- if f and g are total, then so is $\rho^n(f,g)$ [why?]

Computation Theory, L 8

Definition. A [partial] function f is primitive recursive $(f \in PRIM)$ if it can be built up in finitely many steps from the basic functions by use of the operations of composition and primitive recursion.

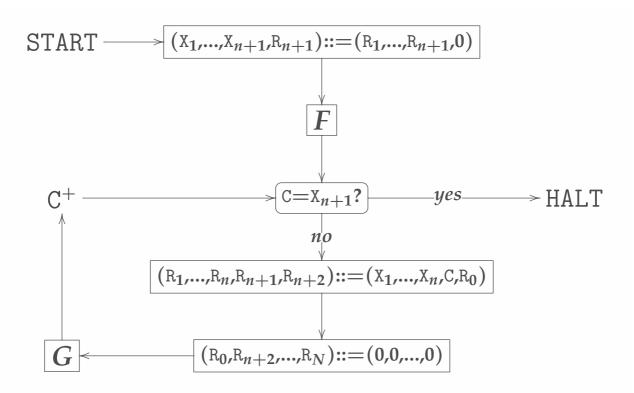
Theorem. Every $f \in \mathbf{PRIM}$ is computable.

Proof. Already proved: basic functions are computable; composition preserves computability. So just have to show:

$$ho^n(f,g)\in \mathbb{N}^{n+1} o \mathbb{N}$$
 computable if $f\in \mathbb{N}^n o \mathbb{N}$ and $g\in \mathbb{N}^{n+1} o \mathbb{N}$ are.

Suppose f and g are computed by RM programs F and G (with our usual I/O conventions). Then the RM specified on the next slide computes $\rho^n(f,g)$. (We assume X_1,\ldots,X_{n+1} , C are some registers not mentioned in F and G; and that the latter only use registers R_0,\ldots,R_N , where $N\geq n+2$.)

Computation Theory , L 8



Computation Theory , L 8

Aim

A more abstract, machine-independent description of the collection of computable partial functions than provided by register/Turing machines:

they form the smallest collection of partial functions containing some basic functions and closed under some fundamental operations for forming new functions from old—composition, primitive recursion and minimization.

The characterization is due to Kleene (1936), building on work of Gödel and Herbrand.

Computation Theory, L 9

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Minimization

```
Given a partial function f \in \mathbb{N}^{n+1} \rightarrow \mathbb{N}, define \mu^n f \in \mathbb{N}^n \rightarrow \mathbb{N} by \mu^n f(\vec{x}) \triangleq \text{least } x \text{ such that } f(\vec{x}, x) = \mathbf{0} and for each i = 0, \dots, x - 1, f(\vec{x}, i) is defined and > 0 (undefined if there is no such x)
```

In other words

$$\mu^{n} f = \{ (\vec{x}, x) \in \mathbb{N}^{n+1} \mid \exists y_{0}, \dots, y_{x}$$

$$(\bigwedge_{i=0}^{x} f(\vec{x}, i) = y_{i}) \wedge (\bigwedge_{i=0}^{x-1} y_{i} > 0) \wedge y_{x} = 0 \}$$

Computation Theory, L 9

Example of minimization

integer part of
$$x_1/x_2 \equiv \text{least } x_3 \text{ such that}$$
 (undefined if $x_2{=}0$) $\qquad \qquad x_1 < x_2(x_3+1)$ $\qquad \equiv \qquad \mu^2 f$

where $f \in \mathbb{N}^3 {
ightarrow} \mathbb{N}$ is

$$f(x_1, x_2, x_3) \triangleq \begin{cases} 1 & \text{if } x_1 \ge x_2(x_3 + 1) \\ 0 & \text{if } x_1 < x_2(x_3 + 1) \end{cases}$$

Computation Theory , L 9 115/17

Definition. A partial function f is partial recursive $(f \in \mathbf{PR})$ if it can be built up in finitely many steps from the basic functions by use of the operations of composition, primitive recursion and minimization.

In other words, the set **PR** of partial recursive functions is the <u>smallest</u> set (with respect to subset inclusion) of partial functions containing the basic functions and closed under the operations of composition, primitive recursion and minimization.

Definition. A partial function f is partial recursive $(f \in \mathbf{PR})$ if it can be built up in finitely many steps from the basic functions by use of the operations of composition, primitive recursion and minimization.

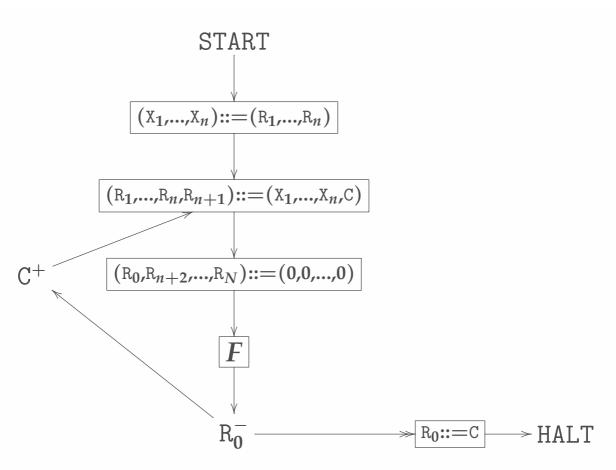
Theorem. Every $f \in PR$ is computable.

Proof. Just have to show:

$$\mu^n f \in \mathbb{N}^n {
ightarrow} \mathbb{N}$$
 is computable if $f \in \mathbb{N}^{n+1} {
ightarrow} \mathbb{N}$ is.

Suppose f is computed by RM program F (with our usual I/O conventions). Then the RM specified on the next slide computes $\mu^n f$. (We assume X_1, \ldots, X_n , C are some registers not mentioned in F; and that the latter only uses registers $\mathbb{R}_0, \ldots, \mathbb{R}_N$, where N > n+1.)

Computation Theory , L 9 117/171



Computation Theory , L 9

Computable = partial recursive

Theorem. Not only is every $f \in \mathbf{PR}$ computable, but conversely, every computable partial function is partial recursive.

Proof (sketch). Let f be computed by RM M. Recall how we coded instantaneous configurations $c = (\ell, r_0, \ldots, r_n)$ of M as numbers $\lceil [\ell, r_0, \ldots, r_n] \rceil$. It is possible to construct primitive recursive functions $lab, val_0, next_M \in \mathbb{N} \to \mathbb{N}$ satisfying

$$lab(\lceil [\ell, r_0, \dots, r_n] \rceil) = \ell$$
 $val_0(\lceil [\ell, r_0, \dots, r_n] \rceil) = r_0$ $next_M(\lceil [\ell, r_0, \dots, r_n] \rceil) = code of M's next configuration$

(Showing that $next_M \in \mathbf{PRIM}$ is tricky—proof omitted.)

Computation Theory, L 9

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Proof sketch, cont.

Let $config_M(\vec{x}, t)$ be the code of M's configuration after t steps, starting with initial register values \vec{x} . It's in **PRIM** because:

$$\begin{cases} config_M(\vec{x}, 0) &= \lceil [0, \vec{x}] \rceil \\ config_M(\vec{x}, t+1) &= next_M(config_M(\vec{x}, t)) \end{cases}$$

Can assume M has a single HALT as last instruction, Ith say (and no erroneous halts). Let $halt_M(\vec{x})$ be the number of steps M takes to halt when started with initial register values \vec{x} (undefined if M does not halt). It satisfies

$$halt_M(\vec{x}) \equiv \text{least } t \text{ such that } I - lab(config_M(\vec{x},t)) = 0$$
 and hence is in PR (because $lab, config_M, I - (\) \in \text{PRIM}$). So $f \in \text{PR}$, because $f(\vec{x}) \equiv val_0(config_M(\vec{x}, halt_M(\vec{x})))$.

Computation Theory, L 9

Definition. A partial function f is partial recursive $(f \in \mathbf{PR})$ if it can be built up in finitely many steps from the basic functions by use of the operations of composition, primitive recursion and minimization.

The members of **PR** that are total are called recursive functions.

Fact: there are recursive functions that are not primitive recursive. For example. . .

Computation Theory, L 9

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Ackermann's function

There is a (unique) function $ack \in \mathbb{N}^2 { o} \mathbb{N}$ satisfying

$$ack(0, x_2) = x_2 + 1$$

 $ack(x_1 + 1, 0) = ack(x_1, 1)$
 $ack(x_1 + 1, x_2 + 1) = ack(x_1, ack(x_1 + 1, x_2))$

- ▶ ack is computable, hence recursive [proof: exercise].
- ▶ **Fact:** *ack* grows faster than any primitive recursive function $f \in \mathbb{N}^2 \rightarrow \mathbb{N}$:

$$\exists N_f \ \forall x_1, x_2 > N_f \ (f(x_1, x_2) < ack(x_1, x_2)).$$
 Hence ack is not primitive recursive.

Computation Theory , L 9

Lambda-Calculus

Computation Theory , L 10

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Notions of computability

- ▶ Church (1936): λ -calculus
- ► Turing (1936): Turing machines.

Turing showed that the two very different approaches determine the same class of computable functions. Hence:

Church-Turing Thesis. Every algorithm [in intuitive sense of Lect. 1] can be realized as a Turing machine.

λ -Terms, M

are built up from a given, countable collection of

 \triangleright variables x, y, z, \dots

by two operations for forming λ -terms:

- ▶ λ -abstraction: $(\lambda x.M)$ (where x is a variable and M is a λ -term)
- ▶ application: (M M') (where M and M' are λ -terms).

Some random examples of λ -terms:

$$x (\lambda x.x) ((\lambda y.(xy))x) (\lambda y.((\lambda y.(xy))x))$$

Computation Theory, L 10

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λ -Terms, M

Notational conventions:

- $(\lambda x_1 x_2 \dots x_n M)$ means $(\lambda x_1 (\lambda x_2 \dots (\lambda x_n M) \dots))$
- $(M_1 M_2 ... M_n)$ means $(... (M_1 M_2) ... M_n)$ (i.e. application is left-associative)
- drop outermost parentheses and those enclosing the body of a λ -abstraction. E.g. write $(\lambda x.(x(\lambda y.(yx))))$ as $\lambda x.x(\lambda y.yx)$.
- x # M means that the variable x does not occur anywhere in the λ -term M.

Computation Theory , L 10

Free and bound variables

In $\lambda x.M$, we call x the bound variable and M the body of the λ -abstraction.

An occurrence of x in a λ -term M is called

- ▶ binding if in between λ and . (e.g. $(\lambda x.y x) x$)
- ▶ bound if in the body of a binding occurrence of x (e.g. $(\lambda x.y.x)x$)
- free if neither binding nor bound (e.g. $(\lambda x.yx)x$).

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Free and bound variables

Sets of free and bound variables:

$$FV(x) = \{x\}$$

 $FV(\lambda x.M) = FV(M) - \{x\}$
 $FV(MN) = FV(M) \cup FV(N)$
 $BV(x) = \emptyset$
 $BV(\lambda x.M) = FV(M) \cup \{x\}$
 $BV(MN) = BV(M) \cup BV(N)$

If $FV(M) = \emptyset$, M is called a closed term, or combinator.

$$\alpha$$
-Equivalence $M =_{\alpha} M'$

 $\lambda x.M$ is intended to represent the function f such that

$$f(x) = M$$
 for all x .

So the name of the bound variable is immaterial: if $M' = M\{x'/x\}$ is the result of taking M and changing all occurrences of x to some variable x' # M, then $\lambda x.M$ and $\lambda x'.M'$ both represent the same function.

For example, $\lambda x.x$ and $\lambda y.y$ represent the same function (the identity function).

Computation Theory, L 10

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$$\alpha$$
-Equivalence $M =_{\alpha} M'$

is the binary relation inductively generated by the rules:

$$\frac{z \# (MN) \qquad M\{z/x\} =_{\alpha} N\{z/y\}}{\lambda x. M =_{\alpha} \lambda y. N}$$

$$\frac{M =_{\alpha} M' \qquad N =_{\alpha} N'}{MN =_{\alpha} M'N'}$$

where $M\{z/x\}$ is M with all occurrences of x replaced by z.

α -Equivalence $M =_{\alpha} M'$

For example:

because
$$\lambda x.(\lambda xx'.x) \ x' =_{\alpha} \lambda y.(\lambda x \ x'.x) x'$$
 because
$$(\lambda z \ x'.z) x' =_{\alpha} (\lambda x \ x'.x) x'$$
 because
$$(\lambda x'.u) x' =_{\alpha} (\lambda x'.u) x'$$
 because
$$\lambda x'.u =_{\alpha} \lambda x'.u and x' =_{\alpha} x'$$
 because
$$u =_{\alpha} u and x' =_{\alpha} x'.$$

Computation Theory , L 10

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α -Equivalence $M =_{\alpha} M'$

Fact: $=_{\alpha}$ is an equivalence relation (reflexive, symmetric and transitive).

We do not care about the particular names of bound variables, just about the distinctions between them. So α -equivalence classes of λ -terms are more important than λ -terms themselves.

- Textbooks (and these lectures) suppress any notation for α -equivalence classes and refer to an equivalence class via a representative λ -term (look for phrases like "we identify terms up to α -equivalence" or "we work up to α -equivalence").
- For implementations and computer-assisted reasoning, there are various devices for picking canonical representatives of α -equivalence classes (e.g. de Bruijn indexes, graphical representations, . . .).

Substitution N[M/x]

$$x[M/x] = M$$
 $y[M/x] = y$ if $y \neq x$
 $(\lambda y.N)[M/x] = \lambda y.N[M/x]$ if $y \# (M x)$
 $(N_1 N_2)[M/x] = N_1[M/x]N_2[M/x]$

Side-condition y # (M x) (y does not occur in M and $y \neq x$) makes substitution "capture-avoiding".

E.g. if
$$x \neq y$$

$$(\lambda y.x)[y/x] \neq \lambda y.y$$

Computation Theory, L 10

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Substitution N[M/x]

```
x[M/x] = M
y[M/x] = y if y \neq x
(\lambda y.N)[M/x] = \lambda y.N[M/x] if y \# (M x)
(N_1 N_2)[M/x] = N_1[M/x]N_2[M/x]
```

Side-condition y # (M x) (y does not occur in M and $y \neq x$) makes substitution "capture-avoiding".

E.g. if
$$x \neq y \neq z \neq x$$

$$(\lambda y.x)[y/x] =_{\alpha} (\lambda z.x)[y/x] = \lambda z.y$$

 $N \mapsto N[M/x]$ induces a <u>total</u> operation on α -equivalence classes.

β -Reduction

Recall that $\lambda x.M$ is intended to represent the function f such that f(x) = M for all x. We can regard $\lambda x.M$ as a function on λ -terms via substitution: map each N to M[N/x].

So the natural notion of computation for λ -terms is given by stepping from a

$$\beta$$
-redex $(\lambda x.M)N$

to the corresponding

$$\beta$$
-reduct $M[N/x]$

Computation Theory , L 10

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β -Reduction

One-step β -reduction, $M \to M'$:

$$\frac{M \to M'}{\lambda x.M)N \to M[N/x]} \qquad \frac{M \to M'}{\lambda x.M \to \lambda x.M'}$$

$$\frac{M \to M'}{M N \to M' N} \qquad \frac{M \to M'}{N M \to N M'}$$

$$\frac{N =_{\alpha} M \qquad M \to M' \qquad M' =_{\alpha} N'}{N \to N'}$$

β -Reduction

E.g.

$$(\lambda x.xy)((\lambda y.\lambda z.z)u) \xrightarrow{((\lambda y.\lambda z.z)u)y} (\lambda z.z)y \longrightarrow y$$

$$(\lambda x.xy)(\lambda z.z)$$

E.g. of "up to α -equivalence" aspect of reduction:

$$(\lambda x.\lambda y.x)y =_{\alpha} (\lambda x.\lambda z.x)y \to \lambda z.y$$

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Many-step β -reduction, $M \rightarrow M'$:

$$rac{M =_{lpha} M'}{M o M'}$$
 $rac{M o M'}{M o M'}$ $rac{M o M'}{M o M'}$ $rac{M o M'}{M o M''}$ (no steps) (1 step) (1 more step)

E.g.

$$(\lambda x.xy)((\lambda yz.z)u) \rightarrow y$$

 $(\lambda x.\lambda y.x)y \rightarrow \lambda z.y$

Lambda-Definable Functions

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β -Conversion $M =_{\beta} N$

Informally: $M =_{\beta} N$ holds if N can be obtained from M by performing zero or more steps of α -equivalence, β -reduction, or β -expansion (= inverse of a reduction).

E.g.
$$u\left((\lambda x\,y.\,v\,x)y\right) =_{\beta} (\lambda x.\,u\,x)(\lambda x.\,v\,y)$$

because $(\lambda x.\,u\,x)(\lambda x.\,v\,y) \to u(\lambda x.\,v\,y)$
and so we have
$$u\left((\lambda x\,y.\,v\,x)y\right) =_{\alpha} u\left((\lambda x\,y'.\,v\,x)y\right) \\ \quad \to u(\lambda y'.\,v\,y) \qquad \text{reduction} \\ \quad =_{\alpha} u(\lambda x.\,v\,y) \\ \quad \leftarrow (\lambda x.\,u\,x)(\lambda x.\,v\,y) \quad \text{expansion}$$

β -Conversion $M =_{\beta} N$

is the binary relation inductively generated by the rules:

$$\frac{M =_{\alpha} M'}{M =_{\beta} M'} \qquad \frac{M \to M'}{M =_{\beta} M'} \qquad \frac{M =_{\beta} M'}{M' =_{\beta} M}$$

$$\frac{M =_{\beta} M'}{M =_{\beta} M''} \qquad \frac{M =_{\beta} M'}{\lambda x. M =_{\beta} \lambda x. M'}$$

$$\frac{M =_{\beta} M'}{M =_{\beta} M'} \qquad N =_{\beta} N'$$

$$M =_{\beta} M' N'$$

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Church-Rosser Theorem

Theorem. wo is confluent, that is, if $M_1 \twoheadleftarrow M wo M_2$, then there exists M' such that $M_1 wo M' \twoheadleftarrow M_2$.

Corollary. $M_1 =_{\beta} M_2$ iff $\exists M (M_1 \twoheadrightarrow M \twoheadleftarrow M_2)$.

Proof. $=_{\beta}$ satisfies the rules generating \twoheadrightarrow ; so $M \twoheadrightarrow M'$ implies $M =_{\beta} M'$. Thus if $M_1 \twoheadrightarrow M \twoheadleftarrow M_2$, then $M_1 =_{\beta} M =_{\beta} M_2$ and so $M_1 =_{\beta} M_2$.

Conversely, the relation $\{(M_1,M_2)\mid \exists M\ (M_1\twoheadrightarrow M\twoheadleftarrow M_2)\}$ satisfies the rules generating $=_{\beta}$: the only difficult case is closure of the relation under transitivity and for this we use the Church-Rosser theorem. Hence $M_1=_{\beta}M_2$ implies $\exists M\ (M_1\twoheadrightarrow M'\twoheadleftarrow M_2)$.

β -Normal Forms

Definition. A λ -term N is in β -normal form (nf) if it contains no β -redexes (no sub-terms of the form $(\lambda x.M)M'$). M has β -nf N if $M=_{\beta}N$ with N a β -nf.

Note that if N is a β -nf and $N \to N'$, then it must be that $N =_{\alpha} N'$ (why?).

Hence if $N_1=_{\beta}N_2$ with N_1 and N_2 both β -nfs, then $N_1=_{\alpha}N_2$. (For if $N_1=_{\beta}N_2$, then by Church-Rosser $N_1 \twoheadrightarrow M' \twoheadleftarrow N_2$ for some M', so $N_1=_{\alpha}M'=_{\alpha}N_2$.)

So the β -nf of M is unique up to α -equivalence if it exists.

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Non-termination

Some λ terms have no β -nf.

E.g. $\Omega \triangleq (\lambda x.xx)(\lambda x.xx)$ satisfies

So there is no β -nf N such that $\Omega =_{\beta} N$.

A term can possess both a β -nf and infinite chains of reduction from it.

E.g.
$$(\lambda x.y)\Omega \to y$$
, but also $(\lambda x.y)\Omega \to (\lambda x.y)\Omega \to \cdots$.

Non-termination

Normal-order reduction is a deterministic strategy for reducing λ -terms: reduce the "left-most, outer-most" redex first.

- \blacktriangleright left-most: reduce M before N in M N, and then
- outer-most: reduce $(\lambda x.M)N$ rather than either of M or N.

(cf. call-by-name evaluation).

Fact: normal-order reduction of M always reaches the β -nf of M if it possesses one.

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Encoding data in λ -calculus

Computation in λ -calculus is given by β -reduction. To relate this to register/Turing-machine computation, or to partial recursive functions, we first have to see how to encode numbers, pairs, lists, ... as λ -terms.

We will use the original encoding of numbers due to Church...

Church's numerals

Notation:
$$egin{cases} M^0 N & riangleq N \ M^1 & riangleq M N \ M^{n+1} N & riangleq M(M^n N) \end{cases}$$

so we can write \underline{n} as $\lambda f x \cdot f^n x$ and we have $\underline{n} M N =_{\beta} M^n N$.

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λ -Definable functions

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is λ -definable if there is a closed λ -term F that represents it: for all

$$(x_1,\ldots,x_n)\in\mathbb{N}^n$$
 and $y\in\mathbb{N}$

- \vdash if $f(x_1,\ldots,x_n)=y$, then $F\underline{x_1}\cdots\underline{x_n}=_\beta y$
- ightharpoonup if $f(x_1,\ldots,x_n)\uparrow$, then $Fx_1\cdots x_n$ has no β -nf.

For example, addition is λ -definable because it is represented by $P \triangleq \lambda x_1 x_2 . \lambda f x . x_1 f(x_2 f x)$:

$$P \underline{m} \underline{n} =_{\beta} \lambda f x. \underline{m} f(\underline{n} f x)$$

$$=_{\beta} \lambda f x. \underline{m} f(f^{n} x)$$

$$=_{\beta} \lambda f x. f^{m}(f^{n} x)$$

$$= \lambda f x. f^{m+n} x$$

$$= m + n$$

Computable = λ -definable

Theorem. A partial function is computable if and only if it is λ -definable.

We already know that

Register Machine computable

- = Turing computable
- = partial recursive.

Using this, we break the theorem into two parts:

- ightharpoonup every partial recursive function is λ -definable
- \triangleright λ -definable functions are RM computable

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λ -Definable functions

Definition. $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ is λ -definable if there is a closed λ -term F that represents it: for all

$$(x_1,\ldots,x_n)\in\mathbb{N}^n$$
 and $y\in\mathbb{N}$

- \vdash if $f(x_1,\ldots,x_n)=y$, then $F\underline{x_1}\cdots\underline{x_n}=_{\beta}y$
- ightharpoonup if $f(x_1,\ldots,x_n)\uparrow$, then $F\underline{x_1}\cdots\underline{x_n}$ has no β -nf.

This condition can make it quite tricky to find a λ -term representing a non-total function.

For now, we concentrate on total functions. First, let us see why the elements of **PRIM** (primitive recursive functions) are λ -definable.

Basic functions

▶ Projection functions, $proj_i^n \in \mathbb{N}^n \rightarrow \mathbb{N}$:

$$\operatorname{proj}_{i}^{n}(x_{1},\ldots,x_{n})\triangleq x_{i}$$

ightharpoonup Constant functions with value 0, zeroⁿ $\in \mathbb{N}^n \rightarrow \mathbb{N}$:

$${\sf zero}^n(x_1,\ldots,x_n) \stackrel{\triangle}{=} 0$$

▶ Successor function, $succ \in \mathbb{N} \rightarrow \mathbb{N}$:

$$\operatorname{succ}(x) \triangleq x + 1$$

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Basic functions are representable

- lacktriangledown proj $_i^n \in \mathbb{N}^n { o} \mathbb{N}$ is represented by $\lambda x_1 \ldots x_n.x_i$
- lacktriangledown zero $^n\in\mathbb{N}^n{ o}\mathbb{N}$ is represented by $\lambda x_1\ldots x_n.\underline{0}$
- lacktriangleright succ $\in \mathbb{N} {
 ightarrow} \mathbb{N}$ is represented by

$$\mathsf{Succ} \triangleq \lambda x_1 f x_1 f (x_1 f x)$$

since

Succ
$$\underline{n} =_{\beta} \lambda f x. f(\underline{n} f x)$$

 $=_{\beta} \lambda f x. f(f^{n} x)$
 $= \lambda f x. f^{n+1} x$
 $= n+1$

Representing composition

If total function $f \in \mathbb{N}^n \to \mathbb{N}$ is represented by F and total functions $g_1, \ldots, g_n \in \mathbb{N}^m \to \mathbb{N}$ are represented by G_1, \ldots, G_n , then their composition $f \circ (g_1, \ldots, g_n) \in \mathbb{N}^m \to \mathbb{N}$ is represented simply by

$$\lambda x_1 \ldots x_m \cdot F(G_1 x_1 \ldots x_m) \ldots (G_n x_1 \ldots x_m)$$

because
$$F\left(G_{1} \underbrace{a_{1} \ldots a_{m}}\right) \ldots \left(G_{n} \underbrace{a_{1} \ldots a_{m}}\right) \\ =_{\beta} F\underbrace{g_{1}(a_{1}, \ldots, a_{m}) \ldots g_{n}(a_{1}, \ldots, a_{m})}_{=\beta} \\ =_{\beta} \underbrace{f\left(g_{1}(a_{1}, \ldots, a_{m}), \ldots, g_{n}(a_{1}, \ldots, a_{m})\right)}_{f \circ \left(g_{1}, \ldots, g_{n}\right)\left(a_{1}, \ldots, a_{m}\right)}$$

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Representing composition

If total function $f \in \mathbb{N}^n \to \mathbb{N}$ is represented by F and total functions $g_1, \ldots, g_n \in \mathbb{N}^m \to \mathbb{N}$ are represented by G_1, \ldots, G_n , then their composition $f \circ (g_1, \ldots, g_n) \in \mathbb{N}^m \to \mathbb{N}$ is represented simply by

$$\lambda x_1 \dots x_m \cdot F(G_1 x_1 \dots x_m) \dots (G_n x_1 \dots x_m)$$

This does not necessarily work for <u>partial</u> functions. E.g. totally undefined function $u \in \mathbb{N} \to \mathbb{N}$ is represented by $U \triangleq \lambda x.\Omega$ (why?) and $\mathsf{zero}^1 \in \mathbb{N} \to \mathbb{N}$ is represented by $Z \triangleq \lambda x.\underline{0}$; but $\mathsf{zero}^1 \circ u$ is not represented by $\lambda x_1.Z(U\,x)$, because $(\mathsf{zero}^1 \circ u)(n)\uparrow$ whereas $(\lambda x_1.Z(U\,x))\,\underline{n} =_{\beta} Z\,\Omega =_{\beta} \underline{0}$. (What is $\mathsf{zero}^1 \circ u$ represented by?)

Primitive recursion

Theorem. Given $f \in \mathbb{N}^n \rightarrow \mathbb{N}$ and $g \in \mathbb{N}^{n+2} \rightarrow \mathbb{N}$, there is a unique $h \in \mathbb{N}^{n+1} \rightarrow \mathbb{N}$ satisfying

$$\begin{cases} h(\vec{x},0) & \equiv f(\vec{x}) \\ h(\vec{x},x+1) & \equiv g(\vec{x},x,h(\vec{x},x)) \end{cases}$$

for all $\vec{x} \in \mathbb{N}^n$ and $x \in \mathbb{N}$.

We write $\rho^n(f,g)$ for h and call it the partial function defined by primitive recursion from f and g.

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Representing primitive recursion

If $f \in \mathbb{N}^n \to \mathbb{N}$ is represented by a λ -term F and $g \in \mathbb{N}^{n+2} \to \mathbb{N}$ is represented by a λ -term G, we want to show λ -definability of the unique $h \in \mathbb{N}^{n+1} \to \mathbb{N}$ satisfying

$$\begin{cases} h(\vec{a},0) &= f(\vec{a}) \\ h(\vec{a},a+1) &= g(\vec{a},a,h(\vec{a},a)) \end{cases}$$

Representing primitive recursion

If $f \in \mathbb{N}^n \to \mathbb{N}$ is represented by a λ -term F and $g \in \mathbb{N}^{n+2} \to \mathbb{N}$ is represented by a λ -term G, we want to show λ -definability of the unique $h \in \mathbb{N}^{n+1} \to \mathbb{N}$ satisfying $h = \Phi_{f,g}(h)$ where $\Phi_{f,g} \in (\mathbb{N}^{n+1} \to \mathbb{N}) \to (\mathbb{N}^{n+1} \to \mathbb{N})$ is given by

$$\Phi_{f,g}(h)(\vec{a},a) \stackrel{\triangle}{=} if \ a = 0 \ then \ f(\vec{a})$$

$$else \ g(\vec{a},a-1,h(\vec{a},a-1))$$

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Representing primitive recursion

If $f \in \mathbb{N}^n \to \mathbb{N}$ is represented by a λ -term F and $g \in \mathbb{N}^{n+2} \to \mathbb{N}$ is represented by a λ -term G, we want to show λ -definability of the unique $h \in \mathbb{N}^{n+1} \to \mathbb{N}$ satisfying $h = \Phi_{f,g}(h)$ where $\Phi_{f,g} \in (\mathbb{N}^{n+1} \to \mathbb{N}) \to (\mathbb{N}^{n+1} \to \mathbb{N})$ is given by... Strategy:

- show that $\Phi_{f,g}$ is λ -definable;
- lacktriangleright show that we can solve fixed point equations $X=M\,X$ up to eta-conversion in the λ -calculus.

Representing booleans

True
$$\triangleq \lambda x y. x$$

False $\triangleq \lambda x y. y$
If $\triangleq \lambda f x y. f x y$

satisfy

- ▶ If True $MN =_{\beta} \text{True } MN =_{\beta} M$
- If False $MN =_{\beta} False MN =_{\beta} N$

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Representing test-for-zero

$\mathsf{Eq}_0 \triangleq \lambda x. x(\lambda y. \mathsf{False}) \mathsf{True}$

satisfies

- ► Eq₀ $\underline{0} =_{\beta} \underline{0} (\lambda y. \text{False})$ True $=_{\beta}$ True

Representing ordered pairs

Pair
$$\triangleq \lambda x y f \cdot f x y$$

Fst $\triangleq \lambda f \cdot f$ True
Snd $\triangleq \lambda f \cdot f$ False

satisfy

Fst(Pair
$$MN$$
) $=_{\beta}$ Fst($\lambda f.fMN$) $=_{\beta}$ ($\lambda f.fMN$) True $=_{\beta}$ True MN $=_{\beta}$ M

▶ Snd(Pair MN) $=_{\beta} \cdots =_{\beta} N$

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Representing predecessor

Want λ -term **Pred** satisfying

$$\begin{array}{ccc} \operatorname{Pred} \underline{n+1} & =_{\beta} & \underline{n} \\ \operatorname{Pred} \underline{0} & =_{\beta} & \underline{0} \end{array}$$

Have to show how to reduce the "n+1-iterator" $\underline{n+1}$ to the "n-iterator" n.

Idea: given f, iterating the function $g_f:(x,y)\mapsto (f(x),x)$ n+1 times starting from (x,x) gives the pair $(f^{n+1}(x),f^n(x))$. So we can get $f^{n+1}(x)$ from $f^n(x)$ parametrically in f and x, by building g_f from f, iterating n+1 times from (x,x) and then taking the second component.

Hence...

Representing predecessor

Want λ -term **Pred** satisfying

$$\begin{array}{ccc} \operatorname{Pred} \underline{n+1} & =_{\beta} & \underline{n} \\ \operatorname{Pred} \underline{0} & =_{\beta} & \underline{0} \end{array}$$

$$\mathsf{Pred} \triangleq \lambda y \, f \, x. \, \mathsf{Snd}(y \, (G \, f)(\mathsf{Pair} \, x \, x))$$
 where $G \triangleq \lambda f \, p. \, \mathsf{Pair}(f(\mathsf{Fst} \, p))(\mathsf{Fst} \, p)$

has the required $oldsymbol{eta}$ -reduction properties. [Exercise]

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Curry's fixed point combinator Y

$$\mathbf{Y} \triangleq \lambda f. (\lambda x. f(x x)) (\lambda x. f(x x))$$

satisfies
$$\mathbf{Y}M \to (\lambda x. M(x\,x))(\lambda x. M(x\,x))$$

 $\to M((\lambda x. M(x\,x))(\lambda x. M(x\,x)))$

hence $YM \rightarrow M((\lambda x. M(xx))(\lambda x. M(xx))) \leftarrow M(YM)$.

So for all λ -terms M we have

$$\mathbf{Y}M =_{\beta} M(\mathbf{Y}M)$$

Representing primitive recursion

If $f\in\mathbb{N}^n\to\mathbb{N}$ is represented by a λ -term F and $g\in\mathbb{N}^{n+2}\to\mathbb{N}$ is represented by a λ -term G, we want to show λ -definability of the unique $h\in\mathbb{N}^{n+1}\to\mathbb{N}$ satisfying $h=\Phi_{f,g}(h)$ where $\Phi_{f,g}\in(\mathbb{N}^{n+1}\to\mathbb{N})\to(\mathbb{N}^{n+1}\to\mathbb{N})$ is given by

$$\Phi_{f,g}(h)(\vec{a},a) \stackrel{\triangle}{=} if \ a = 0 \ then \ f(\vec{a})$$

$$else \ g(\vec{a},a-1,h(\vec{a},a-1))$$

We now know that h can be represented by $Y(\lambda \vec{x}x. \operatorname{If}(\operatorname{Eq}_0 x)(F\vec{x})(G\vec{x}(\operatorname{Pred} x)(z\vec{x}(\operatorname{Pred} x)))).$

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Representing primitive recursion

Recall that the class **PRIM** of primitive recursive functions is the smallest collection of (total) functions containing the basic functions and closed under the operations of composition and primitive recursion.

Combining the results about λ -definability so far, we have: **every** $f \in \mathbf{PRIM}$ **is** λ -**definable**.

So for λ -definability of all recursive functions, we just have to consider how to represent minimization. Recall...

Minimization

```
Given a partial function f \in \mathbb{N}^{n+1} \rightarrow \mathbb{N}, define \mu^n f \in \mathbb{N}^n \rightarrow \mathbb{N} by \mu^n f(\vec{x}) \triangleq \text{least } x \text{ such that } f(\vec{x}, x) = 0 and for each i = 0, \dots, x - 1, f(\vec{x}, i) is defined and > 0 (undefined if there is no such x)
```

Can express $\mu^n f$ in terms of a fixed point equation: $\mu^n f(\vec{x}) \equiv g(\vec{x}, \mathbf{0})$ where g satisfies $g = \Psi_f(g)$ with $\Psi_f \in (\mathbb{N}^{n+1} \rightarrow \mathbb{N}) \rightarrow (\mathbb{N}^{n+1} \rightarrow \mathbb{N})$ defined by $\Psi_f(g)(\vec{x}, x) \equiv if \ f(\vec{x}, x) = 0 \ then \ x \ else \ g(\vec{x}, x+1)$

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Representing minimization

Suppose $f \in \mathbb{N}^{n+1} \to \mathbb{N}$ (totally defined function) satisfies $\forall \vec{a} \exists a \ (f(\vec{a}, a) = 0)$, so that $\mu^n f \in \mathbb{N}^n \to \mathbb{N}$ is totally defined.

Thus for all $\vec{a} \in \mathbb{N}^n$, $\mu^n f(\vec{a}) = g(\vec{a}, 0)$ with $g = \Psi_f(g)$ and $\Psi_f(g)(\vec{a}, a)$ given by if $(f(\vec{a}, a) = 0)$ then a else $g(\vec{a}, a + 1)$.

So if f is represented by a λ -term F, then $\mu^n f$ is represented by

$$\lambda \vec{x}.Y(\lambda z \vec{x} x. \text{If}(\text{Eq}_0(F \vec{x} x)) x (z \vec{x} (\text{Succ } x))) \vec{x} \underline{0}$$

Recursive implies λ -definable

Fact: every partial recursive $f \in \mathbb{N}^n \to \mathbb{N}$ can be expressed in a standard form as $f = g \circ (\mu^n h)$ for some $g, h \in \mathbf{PRIM}$. (Follows from the proof that computable = partial-recursive.)

Hence every (total) recursive function is λ -definable.

More generally, every partial recursive function is λ -definable, but matching up \uparrow with $\not\supseteq \beta$ -nf makes the representations more complicated that for total functions: see [Hindley, J.R. & Seldin, J.P. (CUP, 2008), chapter 4.]

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Computable = λ -definable

Theorem. A partial function is computable if and only if it is λ -definable.

We already know that computable = partial recursive \Rightarrow λ -definable. So it just remains to see that λ -definable functions are RM computable. To show this one can

- ightharpoonup code λ -terms as numbers (ensuring that operations for constructing and deconstructing terms are given by RM computable functions on codes)
- write a RM interpreter for (normal order) β -reduction.

The details are straightforward, if tedious.