Lecture 5: Three More Models

Models of Computation

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Course overview

Monday, July 7 10.30 – 12.30	Tuesday, July 8 10.30 – 12.30	Wednesday, July 9 10.30 – 12.30	Thursday, July 10 10.30 – 12.30	Friday, July 11
intro	classic models			additional models
Introduction to Computability	Machine Models	Recursive Functions	Lambda Calculus	
computation and decision problems, from logic to computability, overview of models of computation relevance of MoCs	Post Machines, typical features, Turing's analysis of human computers, Turing machines, basic recursion theory	primitive recursive functions, Gödel-Herbrand recursive functions, partial recursive funct's, partial recursive = = Turing-computable, Church's Thesis	$\begin{array}{ll} \lambda\text{-terms, }\beta\text{-reduction,}\\ \lambda\text{-definable functions,}\\ \text{partial recursive}\\ =\lambda\text{-definable}\\ =\text{Turing computable} \end{array}$	
	imperative programming	algebraic programming	functional programming	
				14.30 – 16.30
				Three more Models of Computation
				Post's Correspondence Problem, Interaction-Nets, Fractran comparing computational power

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Some Models of Computation

machine model	mathematical model	sort
Turing machine Post machine register machine	Combinatory Logic \(\lambda\text{-calculus}\) Herbrand-Gödel recursive functions partial-recursive/\(\mu\text{-recursive functions}\) Post canonical system (tag system) Post's Correspondence Problem Markov algorithms Lindenmayer systems	classical
	Fractran	less well known
cellular automata neural networks		
hypercomputation		speculative
quantum computing bio-computing reversible computing		physics-/biology- inspired

Overview

▶ Post's Correspondence Problem (by Emil Post, 1946, [5])

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- Fractran (by John Horton Conway, 1987, [2])

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Post's Correspondence Problem

Emil Leon Post:

- article:
 - "A Variant of a Recursively Unsolvable Problem"
 Bulletin of the American Mathematical Society, 1946.

▶ Simulation of models of computation $\mathcal{M}_1 = \langle D_1, \mathcal{F}_1 \rangle$, $\mathcal{M}_2 = \langle D_2, \mathcal{F}_2 \rangle$:

Simulation of functions: function f₂ simulates function f₁ via encoding ρ if:

 $egin{array}{cccc} D_1 & & & & D_2 \ f_1 & & & & & & & & \mathcal{M} \ D_1 & & & & & & \mathcal{M} \ \end{array}$

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$$D_1 \xrightarrow{\rho} D_2$$
 $\mathcal{M}_1 \qquad \forall f_1 \in \mathcal{F}_1 \qquad f_1 \downarrow \qquad \downarrow f_2 \qquad \mathcal{M}_2$
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$$\forall f_1 \in \mathcal{F}_1 \ \exists f_2 \in \mathcal{F}_2 \ (f_2 \ \text{simulates} \ f_1 \ \text{via} \ \rho)$$

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Weak requirements on encodings (Boker/Dershowitz)

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- (ii) bijective functions

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- (i) *injective* functions
- (ii) bijective functions

Definition (power subsumption pre-order [Boker/Dershowitz 2006 [1]])

- (i) $\mathcal{M}_1 \lesssim \mathcal{M}_2$ if: there is an injective ρ such that $\mathcal{M}_1 \lesssim_{\rho} \mathcal{M}_2$
- (ii) $\mathcal{M}_1 \lesssim_{\text{bijective}} \mathcal{M}_2$ if: there is a bijective ρ such that $\mathcal{M}_1 \lesssim_{\rho} \mathcal{M}_2$

However, we found anomalies of these definitions.

$$\mathcal{M} = \langle D, \mathcal{F} \rangle$$
 is a decision model if $\{0,1\} \subseteq D$, $\forall f \in \mathcal{F} (f[D] \subseteq \{0,1\})$.

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Theorem (Endrullis/G/Hendriks, [3])

Let Σ and Γ with $\{0,1\} \subseteq \Sigma$, Γ be alphabets.

Then for every countable decision model $\mathcal{M} = \langle \Sigma^*, \mathcal{F} \rangle$, it holds:

$$\mathcal{M} \lesssim \mathsf{DFA}(\Gamma)$$
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 $\mathsf{TMD}(\Sigma)$: class of Turing machine deciders with input alphabet Σ

Anomaly (example)

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- depend on uncomputable encodings
- can be extended to some moc's with unbounded output domain
- but do not extend to all moc's

models $M_1 \in \mathcal{M}_1$ and $M_2 \in \mathcal{M}_2$ simulate each other with respect to computable coding $\cdot \cdot : I_{\mathcal{M}_1} \to I_{\mathcal{M}_2}$ and decoding $\cdot \cdot : O_{\mathcal{M}_2} \to O_{\mathcal{M}_1}$ if:

$$x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I_{\mathcal{M}_1} \qquad \qquad x_2 \in I_{\mathcal{M}_2} \qquad \qquad x_1 \in I$$

(defines a Galois connection)

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Models of computation, viewed abstractly

A(n abstractly viewed) model of computation (MoC) is a class \mathcal{M} of machines/systems/... such that every $M \in \mathcal{M}$ it holds:

 \triangleright *M* has a countable set $I_{\mathcal{M}}$ of input objects, and a countable set $O_{\mathcal{M}}$ of output objects that are specific to the MoC \mathcal{M} ;

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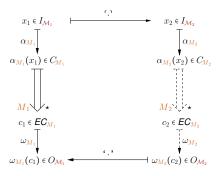
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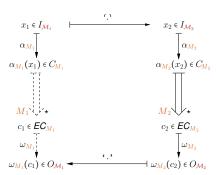
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- ightharpoonup M has a partial output function $\omega_M : EC_M
 ightharpoonup O_M$, which maps some end-configurations of M to output objects of M; ω_M is computable, and membership of end-configurations in $dom(\omega_M)$ is decidable.

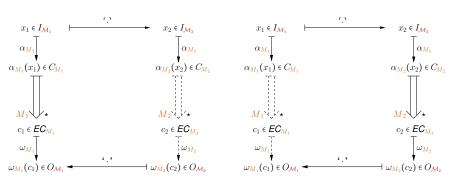
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Comparing Computational Power of MoC's

Definition

Let \mathcal{M}_1 and \mathcal{M}_2 be MoC's.

1 The computational power of \mathcal{M}_1 is subsumed by that of \mathcal{M}_2 , denoted symbolically by $\mathcal{M}_1 \leq \mathcal{M}_2$, if:

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1 The computational power of \mathcal{M}_1 is subsumed by that of \mathcal{M}_2 , denoted symbolically by $\mathcal{M}_1 \leq \mathcal{M}_2$, if:

2 The computational power of \mathcal{M}_1 is equivalent to that of \mathcal{M}_2 , denoted by $\mathcal{M}_1 \sim \mathcal{M}_2$, if both $\mathcal{M}_1 \leq \mathcal{M}_2$ and $\mathcal{M}_2 \leq \mathcal{M}_1$ hold.

Comparing Computational Power of MoC's

Theorem

For all models \mathcal{M}_1 and \mathcal{M}_2 , and encoding and decoding functions $: I_{\mathcal{M}_1} \to I_{\mathcal{M}_2}$ and $: : O_{\mathcal{M}_2} \to O_{\mathcal{M}_1}$ it holds:

$$\mathcal{M}_1 \leq_{(`\cdot,\cdot,\cdot')} \mathcal{M}_2 \implies \mathcal{F}(\mathcal{M}_1) \subseteq \{`\cdot,'\circ f \circ \cdot'\cdot' \mid f \in \mathcal{F}(\mathcal{M}_2)\}.$$

Turing completeness and equivalence

By $\mathcal{TM}(\Sigma)$ we mean the model of Turing machines over input alphabet Σ .

Definition

Let \mathcal{M} a model of computation.

 \mathcal{M} is Turing-complete if $\mathcal{TM}(\Sigma) \leq \mathcal{M}$ for some alphabet Σ with $\Sigma \neq \emptyset$.

 \mathcal{M} is Turing-equivalent if $\mathcal{M} \sim \mathcal{TM}(\Sigma)$ for some alphabet $\Sigma \neq \emptyset$.

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Interaction Nets

Yves Lafont (1990) [4] (link pdf) proposed:

a programming language with a simple graph rewriting semantics

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An interaction net is specified by:

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Analogy with:

- electric circuits:
 - ▶ agents [≙] gates,
 - ▶ edges ^ˆ wires

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Analogy with:

- electric circuits:
 - ▶ agents [≙] gates,
 - ▶ edges [≙] wires
- agents as computation entities:
 - interaction rules specify behavior

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Fractran

John Horton Conway:

- article:
 - FRACTRAN:
 A Simple Universal Programming Language for Arithmetic
- talk video:
 - "Fractran: A Ridiculous Logical Language"

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Summary

- ▶ Post's Correspondence Problem (by Emil Post, 1946, [5])
- Compare computational power of models of computation
- Interaction Nets (by Yves Lafont, 1990, [4])
- Fractran (by John Horton Conway, 1987, [2])

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cellular automata neural networks		
hypercomputation		speculative
quantum computing bio-computing reversible computing		physics-/biology- inspired

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References I



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