## Introduction to Theoretical Computer Science

#### **Lecture 18: Denotational Semantics**

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## Why learn this?

We can't prove anything about a computer program without first giving it a semantics.

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#### In this lecture

We focus mostly on denotational semantics as MCS's treatment is very informal and no other course touches it.

## **Denotational Semantics**

At its heart, it's quite simple:

 $[\![\cdot]\!]: \mathsf{Program} \to \mathsf{Semantics}$ 

More specifically, we define a function  $[\cdot]$  which maps *syntax* into (mathematical) *models*.

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#### Desideratum

We want this semantic function to be *compositional*: The semantics of a compound expression should be made from the semantics of its components.

Domain theory

## Robot Example

### Example (A Toy Language)

A robot moves along a grid according to a sequence of commands move (forward 1 unit) and turn (90 degrees counter-clockwise), separated by semicolons, with the command sequence terminated by the keyword stop:

 $\mathcal{R} ::= move; \mathcal{R} \mid turn; \mathcal{R} \mid stop$ 

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$$\begin{split} \llbracket \cdot \rrbracket^{\mathcal{R}} & : & \mathcal{R} \to \mathbb{Z}^{2} \\ \llbracket \operatorname{turn}; r \rrbracket^{\mathcal{R}} & = & \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \llbracket r \rrbracket^{\mathcal{R}} \\ \llbracket \operatorname{move}; r \rrbracket^{\mathcal{R}} & = & \llbracket r \rrbracket^{\mathcal{R}} + \begin{pmatrix} 1 \\ 0 \end{pmatrix} \\ \llbracket \operatorname{stop} \rrbracket^{\mathcal{R}} & = & \begin{pmatrix} 0 \\ 0 \end{pmatrix} \end{split}$$

Domain theory

# Arithmetic Expressions

$$\llbracket \cdot \rrbracket^{\mathcal{E}} : \mathcal{E} \rightarrow$$

$$\llbracket n \rrbracket^{\mathcal{E}} =$$

$$\llbracket x \rrbracket^{\mathcal{E}} =$$

$$\llbracket e_1 + e_2 \rrbracket^{\mathcal{E}} =$$

$$\llbracket e_1 * e_2 \rrbracket^{\mathcal{E}} =$$

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Domain theory

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## Arithmetic Expressions

$$\llbracket \cdot \rrbracket^{\mathcal{E}} : \mathcal{E} \to \Sigma \to \mathbb{Z}$$

Our denotation for arithmetic expressions is functions from *states* (mapping from variables to their values) to values.

$$\begin{bmatrix} n \end{bmatrix}^{\mathcal{E}} = \lambda \sigma. n \\ \llbracket x \end{bmatrix}^{\mathcal{E}} = \lambda \sigma. \sigma(x) \\ \llbracket e_1 + e_2 \rrbracket^{\mathcal{E}} = \lambda \sigma. \\ \llbracket e_1 * e_2 \rrbracket^{\mathcal{E}} = \lambda \sigma. \\ \llbracket \text{let } x = e_1 \text{ in } e_2 \rrbracket^{\mathcal{E}} = \lambda \sigma.$$

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Where  $\sigma[x := n]$  is a new state just like  $\sigma$  except the variable x now maps to n.

Note: From this point onwards I'll assume all standard arithmetic expressions are in  ${\cal E}$ 

Domain theory

# Boolean Expressions

$$[\![\cdot]\!]^{\mathcal{B}}:\mathcal{B}\rightarrow$$

$$\begin{bmatrix} e_1 == e_2 \end{bmatrix}^{\mathcal{B}} = \\ \begin{bmatrix} e_1 <= e_2 \end{bmatrix}^{\mathcal{B}} = \\ \begin{bmatrix} e_1 \&\& e_2 \end{bmatrix}^{\mathcal{B}} = \\ \begin{bmatrix} e_1 &\& e_2 \end{bmatrix}^{\mathcal{B}} = \\ \begin{bmatrix} e_1 & || & e_2 \end{bmatrix}^{\mathcal{B}} = \\ \begin{bmatrix} ! & e_1 \end{bmatrix}^{\mathcal{B}} =$$

Domain theory

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**Note**: C notation is used here to distinguish syntax from semantics, but from this point onwards I'll assume all standard boolean expressions are in  $\mathcal{B}$ 

Domain theory

## Imperative Programs

We are going to give semantics to non-deterministic imperative programs. Because of non-determinism, our models are relations not functions:

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#### Assignment statement

An *assignment* x := e simply assigns the value of the expression e to the variable x:

$$\llbracket x := e \rrbracket = \left\{ (\sigma_i, \sigma_f) \mid \sigma_f = \sigma_i \left[ x \mapsto \llbracket e \rrbracket^{\mathcal{E}}(\sigma_i) \right] \right\}$$

## More Statements

#### Sequencing

The semicolon, or *sequential composition* operator, is the operator that lets us first run *P*, and then run *Q*.

 $\llbracket P; Q \rrbracket = \llbracket P \rrbracket \operatorname{\mathfrak{g}} \llbracket Q \rrbracket$ 

where ; is forward-composition of relations:

 $X \notin Y = \left\{ (\sigma_i, \sigma_f) \mid \exists \sigma_m. \ (\sigma_i, \sigma_m) \in X \land (\sigma_m, \sigma_f) \in Y \right\}$ 

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#### Example (Swap)

$$(\{a \mapsto 4, b \mapsto 8, \dots\}, \{a \mapsto 8, b \mapsto 4, \dots\}) \\ \in [\![x := a; a := b; b := x]\!]$$

## More Statements

#### Choice and Guards

An *a nondeterministic choice* P + Q means that all observations of P and all observations of Q are possible:

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A boolean expression guard  $\phi$  (in B) doesn't change the state, but only those observations that satisfy  $\phi$  succeed:

 $[\![\boldsymbol{\varphi}]\!] = \left\{ (\boldsymbol{\sigma}, \boldsymbol{\sigma}) \mid \boldsymbol{\sigma} \in [\![\boldsymbol{\varphi}]\!]^{\mathcal{B}} \right\}$ 

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Using these ingredients, we can recover if-statements:

if  $\varphi$  then *P* else *Q* fi  $\simeq (\varphi; P) + (\neg \varphi; Q)$ 

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We will show that this is the same as:

$$\llbracket P^{\star} \rrbracket = \bigcup_{i \in \mathbb{N}_0} \llbracket P \rrbracket^i$$

Where superscripting is self-composition:  $\begin{array}{ll} R^0 &= I \\ R^{n+1} &= R \wr R^n \end{array}$ 

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 $n+1 = R \circ R$ 

R'

We can recover while loops: while g do P od  $\simeq (g; P)^*; \neg g$ 

Domain theory

## Great Scott!

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- **2**  $\omega$ -*chain-complete*: For every countable ascending sequence  $f_0 \sqsubseteq f_1 \sqsubseteq f_2 \ldots$  we have a least upper bound, written sup f or  $\bigsqcup_{n \in \mathbb{N}} f_n$ .

Domain theory



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■ The least upper bound of a chain  $f_0 \subseteq f_1 \subseteq f_2 \dots$  is just  $\bigcup_{i \in \mathbb{N}} f_i$ 

# **Climbing Chains**

Recalling our semantics for the star operator, we want to show that the least fixed point of a function f on our cpo is the least upper bound of the *ascending Kleene chain*:

 $\bot \sqsubseteq f(\bot) \sqsubseteq f(f(\bot)) \sqsubseteq f^3(\bot) \sqsubseteq f^4(\bot) \sqsubseteq \cdots$ 

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$$\bot \sqsubseteq f(\bot) \sqsubseteq f(f(\bot)) \sqsubseteq f^{3}(\bot) \sqsubseteq f^{4}(\bot) \sqsubseteq \cdots$$

#### But!

This chain doesn't exist for some f! Consider this f on the flat domain  $(\mathbb{N}_{\perp}, \sqsubseteq)$ :

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

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Requiring that f is *monotone* fixes this problem, i.e.  $a \le b \implies f(a) \le f(b)$ . Why?

## Monotone isn't enough

Consider this function *f* defined over a cpo  $(\mathbb{R} \cup \{-\infty, \infty\}, \leq)$ :

$$f(x) = \begin{cases} \tan^{-1} x & \text{if } x < 0\\ 1 & \text{otherwise} \end{cases}$$

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### Oh no

It has a fixed point of 1, but the chain approaches 0:

$$f(-\infty) = -\frac{\pi}{2}$$
  
 $f(-\frac{\pi}{2}) = -1$   
 $f(-1) \approx -0.78$ 

But f(0) = 1 — the least upper bound of the ascending Kleene chain is **not** the same as the least fixed point!

### Continuity

### Definition

In a cpo  $(S, \sqsubseteq)$ , a function  $f : S \to S$  is (Scott)-continuous if, for every chain  $x_0 \sqsubseteq x_1 \sqsubseteq x_2 \sqsubseteq \dots$ , f preserves the least upper bound operator:

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Requiring Scott-continuity instead of just monotonicity gives us the Kleene fixed point theorem...

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### Proof it is a fixed point:

 $f(\bigsqcup_{n\in\mathbb{N}}f^n(\bot)) = \bigsqcup_{n\in\mathbb{N}}f(f^n(\bot))$ 

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$$\begin{split} f(\bigsqcup_{n\in\mathbb{N}}f^{n}(\bot)) &= \bigsqcup_{n\in\mathbb{N}}f(f^{n}(\bot)) & \text{(continuity)} \\ &= \bigsqcup_{n\in\mathbb{N}}f^{n+1}(\bot) \\ &= \bigsqcup_{n=1,2\dots}f^{n}(\bot) & \text{(reindexing)} \end{split}$$

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Domain theory

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### Bringing it back to semantics

For our programming language, our cpo is  $(\mathcal{P}(\Sigma \times \Sigma), \subseteq)$ :

- $\blacksquare$  The least element  $\bot = \emptyset$
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All of our composite operators are Scott-continuous:

 $\llbracket P + Q \rrbracket = \llbracket P \rrbracket \cup \llbracket Q \rrbracket \qquad \llbracket P; Q \rrbracket = \llbracket P \rrbracket \operatorname{\mathfrak{z}} \llbracket Q \rrbracket$ 

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Thus, we know from the fixed point theorem that least solutions to our recursive equations always exist and they can be found by iteratively applying the function until we find a fixed point.

Domain theory

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#### Key idea

Add a special value, confusingly also written  $\bot$ , which represents non-terminating computations. Our models would now be  $\mathcal{P}(\Sigma \times \Sigma_{\bot})$  where  $\Sigma_{\bot}$  is either a state or the special "loop forever" value.

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It's quite tricky to define this ordering such that it is a cpo and such that our language operations are still continuous.

#### Further reading

Plotkin resolved this with his Powerdomain construction, which gives a general treatment of non-determinism such that any cpo can be lifted to a non-deterministic context. A programming language

Domain theory

# **Common Theorems**

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Let  $(\sigma, P) \Downarrow \sigma'$  be an operational semantics for our language. It says that, starting in state  $\sigma$ , evaluating the program P on a machine results in  $\sigma'$ .

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- $\blacksquare \textit{ Adequacy} \quad \text{ If } (\sigma, \sigma') \in \llbracket P \rrbracket \text{ then } (\sigma, P) \Downarrow \sigma'$
- *Full Abstraction*  $\llbracket P \rrbracket = \llbracket Q \rrbracket$  iff for all contexts *C* and states  $\sigma$  and  $\sigma'$ ,  $(\sigma, C[P]) \Downarrow \sigma' \Leftrightarrow (\sigma, C[Q]) \Downarrow \sigma'$

The first two are common. The last one is hard.

# More on denotations

This is just the tip of the iceberg in Denotational Semantics.

- Effectful programs use Kleisli categories (monads) for their domain
- Categorical semantics which use structures from category theory for denotations.
- Game semantics which use games as denotations.
- Probabilistic powerdomains and quasi-Borel spaces for probablistic programs.
- Concurrency semantics using traces, transition systems, event structures, Petri nets and so on. MCS



# Best of luck with your exams and the rest of your life! Please feel free to reach out if you're interested in learning more theory.