

CIS 500

Software Foundations

Fall 2003

15 September

Administrivia

- ◆ If you still want to join a study group, send mail to cis500@seas. We currently have one pending request.
- ◆ Sign-up sheets for recitations are being passed around the room. Please write down your name and recitation preferences. We will announce assignments as soon as possible.
- ◆ HW3 is available now; due next Monday.
 - ◆ Can be handed in either physically or electronically
 - ◆ If you foresee a lot of technical writing in your future, you should consider learning how to use Latex now.
 - ◆ Please look at (at least) the first two problems from HW3 **before** coming to this week's recitation.

Where we're going

Going Meta...

The functional programming style used in OCaml is based on treating **programs as data** — i.e., on writing functions that manipulate other functions as their inputs and outputs.

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Everything in this course is based on treating **programs as mathematical objects** — i.e., we will be building mathematical theories whose basic objects of study are programs (and whole programming languages).

Jargon: We will be studying the **metatheory** of programming languages.

Warning!

The material in the next couple of lectures is more slippery than it may first appear.

“I believe it when I hear it” is not a sufficient test of understanding.

A much better test is “I can explain it so that someone else believes it.”

Basics of Induction (Review)

Induction

Principle of **ordinary induction** on natural numbers

Suppose that P is a predicate on the natural numbers. Then:

If $P(0)$

and, for all i , $P(i)$ implies $P(i + 1)$,

then $P(n)$ holds for all n .

Example

Theorem: $2^0 + 2^1 + \dots + 2^n = 2^{n+1} - 1$, for every n .

Proof:

◆ Let $P(i)$ be “ $2^0 + 2^1 + \dots + 2^i = 2^{i+1} - 1$.”

◆ Show $P(0)$:

$$2^0 = 1 = 2^1 - 1$$

◆ Show that $P(i)$ implies $P(i + 1)$:

$$\begin{aligned} 2^0 + 2^1 + \dots + 2^{i+1} &= (2^0 + 2^1 + \dots + 2^i) + 2^{i+1} \\ &= (2^{i+1} - 1) + 2^{i+1} && \text{by IH} \\ &= 2 \cdot (2^{i+1}) - 1 \\ &= 2^{i+2} - 1 \end{aligned}$$

◆ The result ($P(n)$ for all n) follows by the principle of induction.

Shorthand form

Theorem: $2^0 + 2^1 + \dots + 2^n = 2^{n+1} - 1$, for every n .

Proof: By induction on n .

◆ Base case ($n = 0$):

$$2^0 = 1 = 2^1 - 1$$

◆ Inductive case ($n = i + 1$):

$$\begin{aligned} 2^0 + 2^1 + \dots + 2^{i+1} &= (2^0 + 2^1 + \dots + 2^i) + 2^{i+1} \\ &= (2^{i+1} - 1) + 2^{i+1} && IH \\ &= 2 \cdot (2^{i+1}) - 1 \\ &= 2^{i+2} - 1 \end{aligned}$$

Complete Induction

Principle of **complete induction** on natural numbers

Suppose that **P** is a predicate on the natural numbers. Then:

if, for each natural number **n**,

given **P(i)** for all $i < n$

we can show **P(n)**,

then **P(n)** holds for all **n**.

Ordinary and complete induction are **interderivable** — assuming one, we can prove the other.

Thus, the choice of which to use for a particular proof is purely a question of style.

We'll see some other (equivalent) styles as we go along.

Syntax

Simple Arithmetic Expressions

Here is a BNF grammar for a very simple language of arithmetic expressions:

<code>t ::=</code>	
<code> true</code>	constant true
<code> false</code>	constant false
<code> if t then t else t</code>	conditional
<code> 0</code>	constant zero
<code> succ t</code>	successor
<code> pred t</code>	predecessor
<code> iszero t</code>	zero test

Terminology:

- ◆ `t` here is a **metavariable**

Abstract vs. concrete syntax

Q1: Does this grammar define a set of **character strings**, a set of **token lists**, or a set of **abstract syntax trees**?

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Q1: Does this grammar define a set of **character strings**, a set of **token lists**, or a set of **abstract syntax trees**?

A: In a sense, all three. But we are primarily interested, here, in abstract syntax trees.

For this reason, grammars like the one on the previous slide are sometimes called **abstract grammars**. An abstract grammar **defines** a set of abstract syntax trees and **suggests** a mapping from character strings to trees.

We then **write** terms as linear character strings rather than trees simply for convenience. If there is any potential confusion about what tree is intended, we use parentheses to disambiguate.

Q: So, are

`succ 0`

`succ (0)`

`((succ (((((0)))))))`

“the same term”?

What about

`succ 0`

`pred (succ (succ 0))`

?

A more explicit form of the definition

The set \mathcal{T} of **terms** is the smallest set such that

1. $\{\text{true}, \text{false}, 0\} \subseteq \mathcal{T}$;
2. if $t_1 \in \mathcal{T}$, then $\{\text{succ } t_1, \text{pred } t_1, \text{iszero } t_1\} \subseteq \mathcal{T}$;
3. if $t_1 \in \mathcal{T}$, $t_2 \in \mathcal{T}$, and $t_3 \in \mathcal{T}$, then $\text{if } t_1 \text{ then } t_2 \text{ else } t_3 \in \mathcal{T}$.

Inference rules

An alternate notation for the same definition:

$$\begin{array}{c} \text{true} \in \mathcal{T} \\ \frac{t_1 \in \mathcal{T}}{\text{succ } t_1 \in \mathcal{T}} \end{array} \qquad \begin{array}{c} \text{false} \in \mathcal{T} \\ \frac{t_1 \in \mathcal{T}}{\text{pred } t_1 \in \mathcal{T}} \end{array} \qquad \begin{array}{c} 0 \in \mathcal{T} \\ \frac{t_1 \in \mathcal{T}}{\text{iszero } t_1 \in \mathcal{T}} \end{array}$$
$$\frac{t_1 \in \mathcal{T} \quad t_2 \in \mathcal{T} \quad t_3 \in \mathcal{T}}{\text{if } t_1 \text{ then } t_2 \text{ else } t_3 \in \mathcal{T}}$$

Note that “the smallest set closed under...” is implied (but often not stated explicitly).

Terminology:

- ◆ axiom vs. rule
- ◆ concrete rule vs. rule scheme

Terms, concretely

Define an infinite sequence of sets, $\mathcal{S}_0, \mathcal{S}_1, \mathcal{S}_2, \dots$, as follows:

$$\mathcal{S}_i = \emptyset$$

$$\mathcal{S}_{i+1} = \{\text{true, false, 0}\}$$

$$\cup \{\text{succ } t_1, \text{pred } t_1, \text{iszero } t_1 \mid t_1 \in \mathcal{S}_i\}$$

$$\cup \{\text{if } t_1 \text{ then } t_2 \text{ else } t_3 \mid t_1, t_2, t_3 \in \mathcal{S}_i\}$$

Now let

$$\mathcal{S} = \bigcup_i \mathcal{S}_i$$

Comparing the definitions

We have seen two different presentations of terms:

1. as the **smallest** set that is **closed** under certain rules (\mathcal{T})
 - ◆ explicit inductive definition
 - ◆ BNF shorthand
 - ◆ inference rule shorthand
2. as the **limit** (\mathcal{S}) of a series of sets (of larger and larger terms)

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 - ◆ BNF shorthand
 - ◆ inference rule shorthand
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What does it mean to assert that “these presentations are equivalent”?

Induction on Syntax

Why two definitions?

The two ways of defining the set of terms are both useful:

1. the definition of terms as the smallest set with a certain closure property is compact and easy to read
2. the definition of the set of terms as the limit of a sequence gives us an **induction principle** for proving things about terms...

Induction on Terms

Definition: The **depth** of a term t is the smallest i such that $t \in \mathcal{S}_i$.

From the definition of \mathcal{S} , it is clear that, if a term t is in \mathcal{S}_i , then all of its immediate subterms must be in \mathcal{S}_{i-1} , i.e., they must have strictly smaller depths.

This observation justifies the **principle of induction on terms**.

Let P be a predicate on terms.

If, for each term s ,
given $P(r)$ for all immediate subterms r of s
we can show $P(s)$,
then $P(t)$ holds for all t .

Inductive Function Definitions

The set of constants appearing in a term t , written $\text{Consts}(t)$, is defined as follows:

$\text{Consts}(\text{true})$	$=$	$\{\text{true}\}$
$\text{Consts}(\text{false})$	$=$	$\{\text{false}\}$
$\text{Consts}(0)$	$=$	$\{0\}$
$\text{Consts}(\text{succ } t_1)$	$=$	$\text{Consts}(t_1)$
$\text{Consts}(\text{pred } t_1)$	$=$	$\text{Consts}(t_1)$
$\text{Consts}(\text{iszero } t_1)$	$=$	$\text{Consts}(t_1)$
$\text{Consts}(\text{if } t_1 \text{ then } t_2 \text{ else } t_3)$	$=$	$\text{Consts}(t_1) \cup \text{Consts}(t_2) \cup \text{Consts}(t_3)$

Simple, right?

First question:

Normally, a “definition” just assigns a convenient name to a previously-known thing. But here, the “thing” on the right-hand side involves the very name that we are “defining”!

So in what sense is this a definition??

Second question: Suppose we had written this instead...

The set of constants appearing in a term t , written $\text{BadConsts}(t)$, is defined as follows:

$$\text{BadConsts}(\text{true}) = \{\text{true}\}$$

$$\text{BadConsts}(\text{false}) = \{\text{false}\}$$

$$\text{BadConsts}(0) = \{0\}$$

$$\text{BadConsts}(0) = \{\}$$

$$\text{BadConsts}(\text{succ } t_1) = \text{BadConsts}(t_1)$$

$$\text{BadConsts}(\text{pred } t_1) = \text{BadConsts}(t_1)$$

$$\text{BadConsts}(\text{iszero } t_1) = \text{BadConsts}(\text{iszero } (\text{iszero } t_1))$$

What is the essential difference between these two definitions? How do we tell the difference between well-formed inductive definitions and ill-formed ones?

What, exactly, does a well-formed inductive definition mean?

First, recall that a **function** can be viewed as a two-place **relation** (called the “graph” of the function) with certain properties:

- ◆ It is **total**: every element of its domain occurs at least once in its graph
- ◆ It is **deterministic**: every element of its domain occurs at most once in its graph.

We have seen how to define relations inductively. E.g....

Let **Consts** be the smallest two-place relation closed under the following rules:

$$(\text{true}, \{\text{true}\}) \in \text{Consts}$$

$$(\text{false}, \{\text{false}\}) \in \text{Consts}$$

$$(0, \{0\}) \in \text{Consts}$$

$$\frac{(t_1, C) \in \text{Consts}}{(\text{succ } t_1, C) \in \text{Consts}}$$

$$\frac{(t_1, C) \in \text{Consts}}{(\text{pred } t_1, C) \in \text{Consts}}$$

$$\frac{(t_1, C) \in \text{Consts}}{(\text{iszero } t_1, C) \in \text{Consts}}$$

$$\frac{(t_1, C_1) \in \text{Consts} \quad (t_2, C_2) \in \text{Consts} \quad (t_3, C_3) \in \text{Consts}}{(\text{if } t_1 \text{ then } t_2 \text{ else } t_3, (\text{Consts}(t_1) \cup \text{Consts}(t_2) \cup \text{Consts}(t_3))) \in \text{Consts}}$$

This definition certainly defines a **relation** (i.e., the smallest one with a certain closure property).

Q: How can we be sure that this relation is a **function**?

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Q: How can we be sure that this relation is a **function**?

A: **Prove it!**

Theorem: The relation **Consts** defined by the inference rules a couple of slides ago is total and deterministic.

I.e., for each term **t** there is exactly one set of terms **C** such that $(t, C) \in \text{Consts}$.

Proof:

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Proof: By induction on **t**.

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Proof: By induction on t .

To apply the induction principle for terms, we must show, for an arbitrary term t , that if

for each immediate subterm s of t , there is exactly one set of terms C_s such that $(s, C_s) \in \text{Consts}$

then

there is exactly one set of terms C such that $(t, C) \in \text{Consts}$.

Proceed by cases on the form of t .

- ◆ If t is 0 , $true$, or $false$, then we can immediately see from the definition of $Consts$ that there is exactly one set of terms C (namely $\{t\}$) such that $(t, C) \in Consts$.

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- ◆ If t is $succ\ t_1$, then the induction hypothesis tells us that there is exactly one set of terms C_1 such that $(t_1, C_1) \in Consts$. But then it is clear from the definition of $Consts$ that there is exactly one set C (namely C_1) such that $(t, C) \in Consts$.

Proceed by cases on the form of t .

- ◆ If t is `0`, `true`, or `false`, then we can immediately see from the definition of `Consts` that there is exactly one set of terms C (namely $\{t\}$) such that $(t, C) \in \text{Consts}$.
- ◆ If t is `succ` t_1 , then the induction hypothesis tells us that there is exactly one set of terms C_1 such that $(t_1, C_1) \in \text{Consts}$. But then it is clear from the definition of `Consts` that there is exactly one set C (namely C_1) such that $(t, C) \in \text{Consts}$.

Similarly when t is `pred` t_1 or `iszero` t_1 .

- ◆ If t is `if s_1 then s_2 else s_3` , then the induction hypothesis tells us
 - ◆ there is exactly one set of terms C_1 such that $(t_1, C_1) \in \text{Consts}$
 - ◆ there is exactly one set of terms C_2 such that $(t_2, C_2) \in \text{Consts}$
 - ◆ there is exactly one set of terms C_3 such that $(t_3, C_3) \in \text{Consts}$

But then it is clear from the definition of Consts that there is exactly one set C (namely $C_1 \cup C_2 \cup C_3$) such that $(t, C) \in \text{Consts}$.

How about the bad definition?

$$(\text{true}, \{\text{true}\}) \in \text{BadConsts}$$
$$(\text{false}, \{\text{false}\}) \in \text{BadConsts}$$
$$(0, \{0\}) \in \text{BadConsts}$$
$$(0, \{\}) \in \text{BadConsts}$$
$$\frac{(t_1, C) \in \text{BadConsts}}{(\text{succ } t_1, C) \in \text{BadConsts}}$$
$$\frac{(t_1, C) \in \text{BadConsts}}{(\text{pred } t_1, C) \in \text{BadConsts}}$$
$$\frac{(\text{iszero } (\text{iszero } t_1), C) \in \text{BadConsts}}{(\text{iszero } t_1, C) \in \text{BadConsts}}$$

This set of rules defines a perfectly good **relation** — it's just that this relation does not happen to be a function!

Just for fun, let's calculate some cases of this relation...

- ◆ For what values of **C** do we have $(\text{false}, \mathbf{C}) \in \text{Consts}$?
- ◆ For what values of **C** do we have $(\text{succ } 0, \mathbf{C}) \in \text{Consts}$?
- ◆ For what values of **C** do we have $(\text{if false then } 0 \text{ else } 0, \mathbf{C}) \in \text{Consts}$?
- ◆ For what values of **C** do we have $(\text{iszero } 0, \mathbf{C}) \in \text{Consts}$?

Another Inductive Definition

<code>size(true)</code>	<code>= 1</code>
<code>size(false)</code>	<code>= 1</code>
<code>size(0)</code>	<code>= 1</code>
<code>size(succ t₁)</code>	<code>= size(t₁) + 1</code>
<code>size(pred t₁)</code>	<code>= size(t₁) + 1</code>
<code>size(iszero t₁)</code>	<code>= size(t₁) + 1</code>
<code>size(if t₁ then t₂ else t₃)</code>	<code>= size(t₁) + size(t₂) + size(t₃) + 1</code>

Another proof by induction

Theorem: The number of distinct constants in a term is at most the size of the term. I.e., $|\text{Consts}(t)| \leq \text{size}(t)$.

Proof:

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There are three cases to consider:

Case: t is a constant

Immediate: $|\text{Consts}(t)| = |\{t\}| = 1 = \text{size}(t)$.

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There are three cases to consider:

Case: t is a constant

Immediate: $|\text{Consts}(t)| = |\{t\}| = 1 = \text{size}(t)$.

Case: $t = \text{succ } t_1, \text{pred } t_1, \text{ or iszero } t_1$

By the induction hypothesis, $|\text{Consts}(t_1)| \leq \text{size}(t_1)$. We now calculate as follows: $|\text{Consts}(t)| = |\text{Consts}(t_1)| \leq \text{size}(t_1) < \text{size}(t)$.

Case: $t = \text{if } t_1 \text{ then } t_2 \text{ else } t_3$

By the induction hypothesis, $|\text{Consts}(t_1)| \leq \text{size}(t_1)$,
 $|\text{Consts}(t_2)| \leq \text{size}(t_2)$, and $|\text{Consts}(t_3)| \leq \text{size}(t_3)$. We now calculate as follows:

$$\begin{aligned} |\text{Consts}(t)| &= |\text{Consts}(t_1) \cup \text{Consts}(t_2) \cup \text{Consts}(t_3)| \\ &\leq |\text{Consts}(t_1)| + |\text{Consts}(t_2)| + |\text{Consts}(t_3)| \\ &\leq \text{size}(t_1) + \text{size}(t_2) + \text{size}(t_3) \\ &< \text{size}(t). \end{aligned}$$

Operational Semantics

Abstract Machines

An **abstract machine** consists of:

- ◆ a set of **states**
- ◆ a **transition relation** on states, written \longrightarrow

A state records **all** the information in the machine at a given moment. For example, an abstract-machine-style description of a conventional microprocessor would include the program counter, the contents of the registers, the contents of main memory, and the machine code program being executed.

For the very simple languages we are considering at the moment, however, the term being evaluated is the whole state of the abstract machine.

Nb. Often, the transition relation is actually a partial function: i.e., from a given state, there is at most one possible next state. But in general there may be many.

Operational semantics for Booleans

Syntax of terms and values

$t ::=$

true

false

if t then t else t

terms

constant true

constant false

conditional

$v ::=$

true

false

values

true value

false value

The evaluation relation $t \longrightarrow t'$ is the smallest relation closed under the following rules:

$\text{if true then } t_2 \text{ else } t_3 \longrightarrow t_2$ (E-IFTRUE)

$\text{if false then } t_2 \text{ else } t_3 \longrightarrow t_3$ (E-IFFALSE)

$$\frac{t_1 \longrightarrow t'_1}{\text{if } t_1 \text{ then } t_2 \text{ else } t_3 \longrightarrow \text{if } t'_1 \text{ then } t_2 \text{ else } t_3}$$
 (E-IF)

Terminology

Computation rules:

if true then t_2 else $t_3 \longrightarrow t_2$ (E-IFTRUE)

if false then t_2 else $t_3 \longrightarrow t_3$ (E-IFFALSE)

Congruence rule:

$$\frac{t_1 \longrightarrow t'_1}{\text{if } t_1 \text{ then } t_2 \text{ else } t_3 \longrightarrow \text{if } t'_1 \text{ then } t_2 \text{ else } t_3} \quad (\text{E-IF})$$

Computation rules perform “real” computation steps.

Congruence rules determine **where** computation rules can be applied next.

Evaluation, more explicitly

\longrightarrow is the smallest two-place relation closed under the following rules:

$$((\text{if true then } t_2 \text{ else } t_3), t_2) \in \longrightarrow$$

$$((\text{if false then } t_2 \text{ else } t_3), t_3) \in \longrightarrow$$

$$(t_1, t'_1) \in \longrightarrow$$

$$((\text{if } t_1 \text{ then } t_2 \text{ else } t_3), (\text{if } t'_1 \text{ then } t_2 \text{ else } t_3)) \in \longrightarrow$$