

SMT and Z3

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February 5, 2014

Plan

Mon An invitation to SMT with Z3

Tue Equalities and Theory Combination

Wed Theories: Arithmetic, Arrays, Data types

Thu Quantifiers and Theories

Fri Programming Z3: Interfacing and Solving

Quiz

Show: A *difference logic graph* without negative cycles has a model. Give a procedure for extracting a model.

True or false: A formula over difference logic has a model over reals **iff** it has a model over integers?

Give an *efficient* algorithm to extract models for UTVPI over integers.

Encode lambda Calculus into *map, K, read* (without *I*).

Plan

- Arithmetic
- Arrays and friends
- Data types [Introduction]

What Theories?

Overall aim:

*Rich Theories (and logics) with
Efficient Decision Procedures*

Auth

MSOL

Sequences

XDucers

Queues

ASP

DL

homomor
phisms

Optimiz
ation

Orders

Objects

HOL

MultiSets

BAPA

Strings

Reg. Exprs.

NRA

NIA

Floats

f*

*

SAT

EUF

LRA

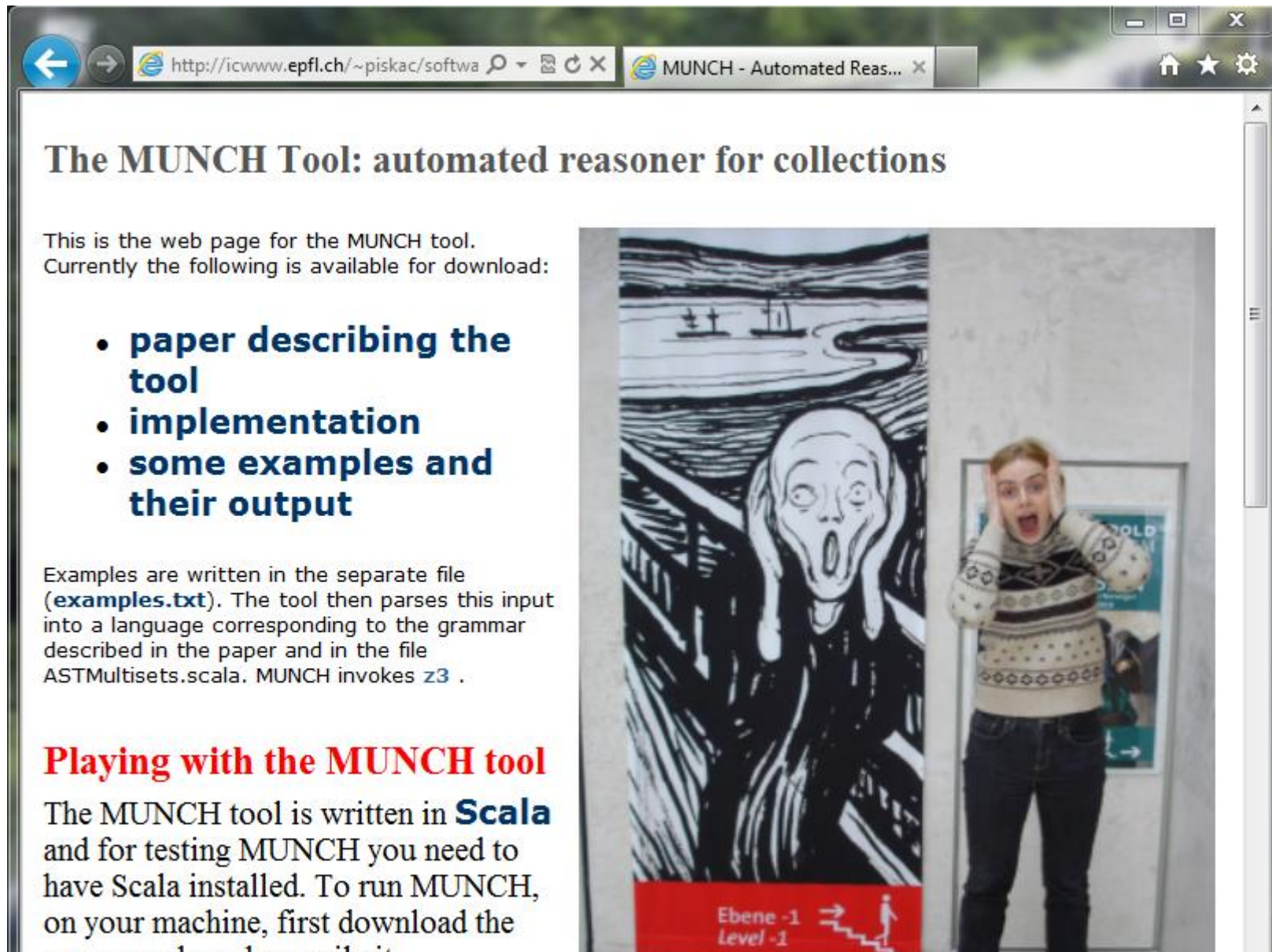
LIA

Arrays

Bit-Vectors

Alg. DT

Be afraid!

A screenshot of a web browser window. The address bar shows the URL 'http://icwww.epfl.ch/~piskac/softwa'. The browser title is 'MUNCH - Automated Reas...'. The page content includes a title 'The MUNCH Tool: automated reasoner for collections', a paragraph about the tool's availability, a bulleted list of resources, a paragraph about examples, and a section titled 'Playing with the MUNCH tool' with text about the tool's implementation in Scala. On the right side of the page, there is a vertical image strip showing a person mimicking the expression of a figure in 'The Scream' painting.

The MUNCH Tool: automated reasoner for collections

This is the web page for the MUNCH tool.
Currently the following is available for download:

- **paper describing the tool**
- **implementation**
- **some examples and their output**

Examples are written in the separate file (**examples.txt**). The tool then parses this input into a language corresponding to the grammar described in the paper and in the file `ASTMultisets.scala`. MUNCH invokes `z3`.

Playing with the MUNCH tool

The MUNCH tool is written in **Scala** and for testing MUNCH you need to have Scala installed. To run MUNCH, on your machine, first download the



Linear Real Arithmetic

- Many approaches
 - Graph-based for difference logic: $a - b \leq 3$
 - Fourier-Motzkin elimination:
$$t_1 \leq ax, bx \leq t_2 \Rightarrow bt_1 \leq at_2$$
 - Standard Simplex
 - General Form Simplex
 - GDPLL [McMillan],
Unate Resolution [Coton],
Conflict Resolution [Korovin et.al.]

Difference Logic: $a - b \leq 5$

Very useful in practice!

Most arithmetical constraints in software verification/analysis are in this fragment.

$$x := x + 1$$



$$x_1 = x_0 + 1$$



$$x_1 - x_0 \leq 1, x_0 - x_1 \leq -1$$

Job shop scheduling

$d_{i,j}$	Machine 1	Machine 2
Job 1	2	1
Job 2	3	1
Job 3	2	3

$max = 8$

Solution

$t_{1,1} = 5, t_{1,2} = 7, t_{2,1} = 2,$
 $t_{2,2} = 6, t_{3,1} = 0, t_{3,2} = 3$

Encoding

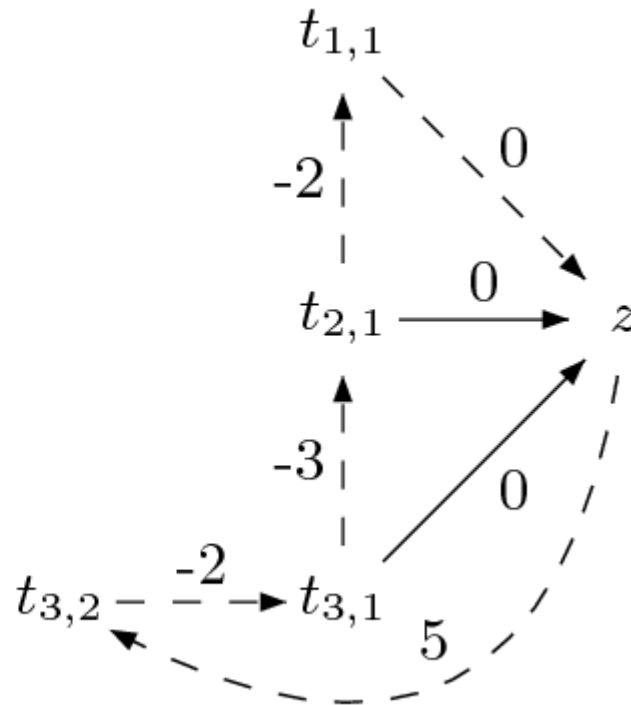
$(t_{1,1} \geq 0) \wedge (t_{1,2} \geq t_{1,1} + 2) \wedge (t_{1,2} + 1 \leq 8) \wedge$
 $(t_{2,1} \geq 0) \wedge (t_{2,2} \geq t_{2,1} + 3) \wedge (t_{2,2} + 1 \leq 8) \wedge$
 $(t_{3,1} \geq 0) \wedge (t_{3,2} \geq t_{3,1} + 2) \wedge (t_{3,2} + 3 \leq 8) \wedge$
 $((t_{1,1} \geq t_{2,1} + 3) \vee (t_{2,1} \geq t_{1,1} + 2)) \wedge$
 $((t_{1,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{1,1} + 2)) \wedge$
 $((t_{2,1} \geq t_{3,1} + 2) \vee (t_{3,1} \geq t_{2,1} + 3)) \wedge$
 $((t_{1,2} \geq t_{2,2} + 1) \vee (t_{2,2} \geq t_{1,2} + 1)) \wedge$
 $((t_{1,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{1,2} + 1)) \wedge$
 $((t_{2,2} \geq t_{3,2} + 3) \vee (t_{3,2} \geq t_{2,2} + 1))$

Difference Logic

Chasing negative cycles!

Algorithms based on Bellman-Ford ($O(mn)$).

$$\begin{array}{rcll} z & - & t_{1,1} & \leq 0 \\ z & - & t_{2,1} & \leq 0 \\ z & - & t_{3,1} & \leq 0 \\ t_{3,2} & - & z & \leq 5 \\ t_{3,1} & - & t_{3,2} & \leq -2 \\ t_{2,1} & - & t_{3,1} & \leq -3 \\ t_{1,1} & - & t_{2,1} & \leq -2 \end{array}$$



Unit Two Variables Per Inequality

$$x + y \leq 5 \wedge -x + y \leq -4 \wedge y + y \geq 1$$

Unit Two Variables Per Inequality

$$x + y \leq 5 \wedge -x + y \leq -4 \wedge 2y \geq 1$$

$$2y \leq 1 \wedge 2y \geq 1$$

Unit Two Variables Per Inequality

$$x + y \leq 5 \wedge -x + y \leq -4 \wedge 2y \geq 1$$

$$2y \leq 1 \wedge 2y \geq 1$$

$$y \leq 0 \wedge y \geq 1$$

Unit Two Variables Per Inequality: UTVPI

Reduce to Difference Logic:

- For every variable x introduce fresh variables x^+, x^-
- Meaning: $2x := x^+ - x^-$
- Rewrite constraints as follows:

- $x - y \leq k \quad \Rightarrow \quad \begin{cases} x^+ - y^+ \leq k \\ y^- - x^- \leq k \end{cases}$

UTVPI

- $x - y \leq k \Rightarrow \begin{cases} x^+ - y^+ \leq k \\ y^- - x^- \leq k \end{cases}$

- $x \leq k \Rightarrow x^+ - x^- \leq 2k$

- $x + y \leq k \Rightarrow \begin{cases} x^+ - y^- \leq k \\ y^+ - x^- \leq k \end{cases}$

- $x + y \leq k \Rightarrow$ chalkboard

UTVPI

$$x + y \leq 5 \wedge -x + y \leq -4 \wedge 2y \geq 1$$

$$x^+ - y^- \leq 5 \wedge y^+ - x^- \leq 5 \wedge$$

$$-x^+ + y^+ \leq -4 \wedge x^- - y^- \leq -4 \wedge$$

$$y^- - y^+ \leq 1$$

UTVPI

- Solve for x^+ and x^-
- $M(x) := (M(x^+) - M(x^-))/2$
- Nothing can go wrong...
$$2y \leq 1 \wedge 2y \geq 1$$

UTVPI

- $M(x) := (M(x^+) - M(x^-))/2$
- Nothing can go wrong... as if
- What if:
 - x is an integer
 - $M(x^+)$ is **odd** and
 - $M(x^-)$ is **even**
- **Thm:** Parity can be fixed **iff** there is no tight loop forcing the wrong parity

UTVPI

$$x^- - y^+ \leq 5$$

$$y^+ - z^- \leq -6$$

$$z^- - x^+ \leq -2$$

$$x^+ - v^+ \leq 3$$

$$v^+ - x^- \leq 0$$

$$\Rightarrow \begin{aligned} x^- - x^+ &\leq -3 \\ x^+ - x^- &\leq 3 \end{aligned}$$

General Form

General Form: $Ax = 0$ and $l_j \leq x_j \leq u_j$

Example:

$$x \geq 0, (x + y \leq 2 \vee x + 2y \geq 6), (x + y = 2 \vee x + 2y > 4)$$

\rightsquigarrow

$$s_1 \equiv x + y, s_2 \equiv x + 2y,$$

$$x \geq 0, (s_1 \leq 2 \vee s_2 \geq 6), (s_1 = 2 \vee s_2 > 4)$$

Only **bounds** (e.g., $s_1 \leq 2$) are asserted during the search.

Unconstrained variables can be **eliminated** before the beginning of the search.

From Definitions to a Tableau

$$s_1 \equiv x + y, \quad s_2 \equiv x + 2y$$

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$$s_1 = x + y,$$

$$s_2 = x + 2y$$

From Definitions to a Tableau

$$s_1 \equiv x + y, \quad s_2 \equiv x + 2y$$



$$s_1 = x + y,$$
$$s_2 = x + 2y$$



$$s_1 - x - y = 0$$
$$s_2 - x - 2y = 0$$

From Definitions to a Tableau

$$s_1 \equiv x + y, \quad s_2 \equiv x + 2y$$



$$s_1 = x + y,$$

$$s_2 = x + 2y$$



$$s_1 - x - y = 0$$

$$s_2 - x - 2y = 0$$

s_1, s_2 are basic (dependent)

x, y are non-basic

Pivoting

A way to swap a basic with a non-basic variable!

It is just equational reasoning.

Key invariant: a basic variable occurs in only one equation.

Example: swap s_1 and y

$$s_1 - x - y = 0$$

$$s_2 - x - 2y = 0$$

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$$-s_1 + x + y = 0$$

$$s_2 - x - 2y = 0$$

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$$-s_1 + x + y = 0$$

$$s_2 - 2s_1 + x = 0$$

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It is just substituting equals by equals.

Definition:

An assignment (model) is a mapping from variables to values

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It is just equational reasoning.

Key invariant: a basic variable occurs in only one equation.

Example: swap s_1 and y

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$$s_2 - x - 2y = 0$$



$$-s_1 + x + y = 0$$

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It is just substituting equals by equals.

Key Property:

If an assignment satisfies the equations before a pivoting step, then it will also satisfy them after!

Definition:

An assignment (model) is a mapping from variables to values

A way to swap a basic with a non-basic variable!

It is just equational reasoning.

Key invariant: a basic variable occurs in only one equation.

Example: swap s_2 and y

$$s_1 - x - y = 0$$

$$s_2 - x - 2y = 0$$



$$-s_1 + x + y = 0$$

$$s_2 - x - 2y = 0$$



$$-s_1 + x + y = 0$$

$$s_2 - 2s_1 + x = 0$$

Example:

$$M(x) = 1$$

$$M(y) = 1$$

$$M(s_1) = 2$$

$$M(s_2) = 3$$

It is just substituting equals by equals.

Key Property:

If an assignment satisfies the equations before a pivoting step, then it will also satisfy them after!

Equations + Bounds + Assignment

An **assignment** (model) is a mapping from variables to values.

We maintain an **assignment** that satisfies all **equations** and **bounds**.

The assignment of non dependent variables implies the assignment of dependent variables.

Equations + Bounds can be used to derive **new bounds**.

Example: $x = y - z, y \leq 2, z \geq 3 \rightsquigarrow x \leq -1$.

The **new bound** may be inconsistent with the already known bounds.

Example: $x \leq -1, x \geq 0$.

“Repairing Models”

If the assignment of a non-basic variable does not satisfy a bound, then fix it and propagate the change to all dependent variables.

$$a = c - d$$

$$b = c + d$$

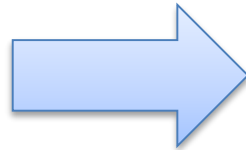
$$M(a) = 0$$

$$M(b) = 0$$

$$M(c) = 0$$

$$M(d) = 0$$

$$1 \leq c$$



$$a = c - d$$

$$b = c + d$$

$$M(a) = 1$$

$$M(b) = 1$$

$$M(c) = 1$$

$$M(d) = 0$$

$$1 \leq c$$

“Repairing Models”

If the assignment of a non-basic variable does not satisfy a bound, then fix it and propagate the change to all dependent variables. **Of course, we may introduce new “problems”.**

$$a = c - d$$

$$b = c + d$$

$$M(a) = 0$$

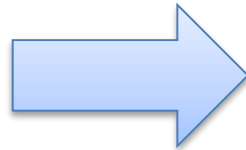
$$M(b) = 0$$

$$M(c) = 0$$

$$M(d) = 0$$

$$1 \leq c$$

$$a \leq 0$$



$$a = c - d$$

$$b = c + d$$

$$M(a) = 1$$

$$M(b) = 1$$

$$M(c) = 1$$

$$M(d) = 0$$

$$1 \leq c$$

$$a \leq 0$$

“Repairing Models”

If the assignment of a basic variable does not satisfy a bound, then pivot it, fix it, and propagate the change to its new dependent variables.

$a = c - d$	$c = a + d$	$c = a + d$
$b = c + d$	$b = a + 2d$	$b = a + 2d$
$M(a) = 0$	$M(a) = 0$	$M(a) = 1$
$M(b) = 0$	$M(b) = 0$	$M(b) = 1$
$M(c) = 0$	$M(c) = 0$	$M(c) = 1$
$M(d) = 0$	$M(d) = 0$	$M(d) = 0$
$1 \leq a$	$1 \leq a$	$1 \leq a$

“Repairing Models”

Sometimes, a model cannot be repaired. It is pointless to pivot.

$$a = b - c$$

$$a \leq 0, 1 \leq b, c \leq 0$$

$$M(a) = 1$$

$$M(b) = 1$$

$$M(c) = 0$$

The value of $M(a)$ is too big. We can reduce it by:

- reducing $M(b)$

not possible b is at lower bound

- increasing $M(c)$

not possible c is at upper bound

“Repairing Models”

Extracting proof from failed repair attempts is easy.

$$s_1 \equiv a + d, s_2 \equiv c + d$$

$$a = s_1 - s_2 + c$$

$$a \leq 0, 1 \leq s_1, s_2 \leq 0, 0 \leq c$$

$$M(a) = 1$$

$$M(s_1) = 1$$

$$M(s_2) = 0$$

$$M(c) = 0$$

“Repairing Models”

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$$a \leq 0, 1 \leq s_1, s_2 \leq 0, 0 \leq c$$

$$M(a) = 1$$

$$M(s_1) = 1$$

$$M(s_2) = 0$$

$$M(c) = 0$$

$\{ a \leq 0, 1 \leq s_1, s_2 \leq 0, 0 \leq c \}$ is inconsistent

“Repairing Models”

Extracting proof from failed repair attempts is easy.

$$s_1 \equiv a + d, s_2 \equiv c + d$$

$$a = s_1 - s_2 + c$$

$$a \leq 0, 1 \leq s_1, s_2 \leq 0, 0 \leq c$$

$$M(a) = 1$$

$$M(s_1) = 1$$

$$M(s_2) = 0$$

$$M(c) = 0$$

$\{ a \leq 0, 1 \leq s_1, s_2 \leq 0, 0 \leq c \}$ is inconsistent

$\{ a \leq 0, 1 \leq a + d, c + d \leq 0, 0 \leq c \}$ is inconsistent

Arrays and Combinatory Array Logic

What are arrays?

- Applicative stores:

$$\mathit{write}(a, i, v)[i] = v$$

$$i \neq j \Rightarrow \mathit{write}(a, i, v)[j] = a[j]$$

- Or, special combinator:

$$\mathit{write}(a, i, v) = \lambda j. \mathit{ite}(i = j, v, a[j])$$

What are arrays?

- Special combinator:

$$\mathit{write}(a, i, v) = \lambda j. \mathit{ite}(i = j, v, a[j])$$

- Existential fragment is decidable by reduction to congruence closure using finite set of instances.
- Models for arrays are finite maps with default values.

What else are arrays?

- Special combinators:

$$\mathit{write}(a, i, v) = \lambda j. \mathit{ite}(i = j, v, a[j])$$

$$K(v) = \lambda j. v$$

$$\mathit{map}_f(a, b) = \lambda j. f(a[j], b[j])$$

- **Result:** Existential fragment is decidable and in NP by reduction to congruence closure using finite set of instances.

What else are arrays++?

- Extra special combinators:

$$\mathit{write}(a, i, v) = \lambda j. \mathit{ite}(i = j, v, a[j])$$

$$K(v) = \lambda j. v$$

$$\mathit{map}_f(a, b) = \lambda j. f(a[j], b[j])$$

$$I = \lambda j. j$$

- Easy to encode lambda calculus

What else are arrays++?

- Encoding lambda terms into CAL+:

$$\begin{array}{l|l} [[\lambda x.M]] = tr(x, [[M]]) & tr(x, x) = I \\ [[x]] = x & tr(x, y) = K(y) \\ [[(MN)]] = map_{read} ([[M]], [[N]]) & tr(x, f(M, N)) = map_f(tr(x, M), tr(x, N)) \end{array}$$

- Where

$$M, N ::= x \mid \lambda x.M \mid (MN)$$

Exercise: encode lambda calculus without /

NB. Our procedure is going to assume that function passed to map is not from *read*.

Example translation

$$\begin{aligned} & [[\lambda x.((\lambda y.(yx))x)]] \\ & = tr(x, [[((\lambda y.(yx))x)]])) \\ & = tr(x, map_{read} ([[\lambda y.(yx)]], [[x]])) \\ & = tr(x, map_{read} ([[\lambda y.(yx)]], x)) \\ & = tr(x, map_{read} (tr(y, [[(yx)]]), x)) \\ & = tr(x, map_{read} (tr(y, map_{read} (y, x)), x)) \\ & = tr(x, map_{read} (map_{map_{read}} (tr(y, y), tr(y, x))), x)) \end{aligned}$$

$$\begin{aligned} & = tr(x, map_{read} (map_{map_{read}} (I, K(x)), x)) \\ & = map_{map_{read}} (tr(x, map_{map_{read}} (I, K(x))), tr(x, x)) \\ & = map_{map_{read}} (map_{map_{map_{read}}} (tr(x, I), tr(x, K(x))), I) \\ & = map_{map_{read}} (map_{map_{map_{read}}} (K(I), tr(x, K(x))), I) \\ & = map_{map_{read}} (map_{map_{map_{read}}} (K(I), map_K (tr(x, x))), I) \\ & = map_{map_{read}} (map_{map_{map_{read}}} (K(I), map_K (I)), I) \end{aligned}$$

... But there are arrays#:

- Restricted theory using I .

$$K(v) = \lambda j.v$$

$$\text{map}_{ite}(a, b, c) = \lambda j.ite(a[j], b[j], c[j])$$

$$\text{map}_=(a, b) = \lambda j.(a[j] = b[j])$$

$$I = \lambda j.j$$

- Then: $\text{write}(a, i, v) = \text{map}_{ite}(\text{map}_=(K(i), I), K(v), a)$
- Theory of arrays# is decidable.

Last combinator for the road...

- Can I access a *default* array value?

$\delta(a) - \text{default}$

$\delta(K(v)) = v$

$\delta(\text{map}_f(a, b)) = f(\delta(a), \delta(b))$

$\delta(\text{write}(a, i, v)) = \delta(a)$

Only sound for infinite domains

Let's use CAL:

- Simple set and bag operations:

\emptyset	□	$K(\text{false})$	\emptyset_{Bag}	□	$K(0)$
$\{a\}$	□	$\text{write}(\emptyset, a, \text{true})$	$\{a\}$	□	$\text{write}(\emptyset, a, 1)$
$a \in A$	□	$A[a]$	$\text{mult}(a, A)$	□	$A[a]$
$A \cup B$	□	$\text{map}_{\vee}(A, B)$	$A \oplus B$	□	$\text{map}_{+}(A, B)$
$A \cap B$	□	$\text{map}_{\wedge}(A, B)$	$A \Pi B$	□	$\text{map}_{\min}(A, B)$
$\text{finite}(A)$	□	$(\delta(A) = \text{false})$	$\text{finite}_{\text{Bag}}(A)$	□	$(\delta(A) = 0)$

- But not cardinality $|A|$, power-set 2^A , ...

CAL: Arrays as Combinators

- McCarthy Arrays:
store/select

$$\begin{aligned} & \text{select}(\text{store}(a, i, v), i) = v \\ i \neq j & \Rightarrow \text{select}(\text{store}(a, i, v), j) = \text{select}(a, j) \end{aligned}$$

- Array
combinators:

$$\begin{aligned} \text{store}(a, i, v) & := \lambda j. \mathbf{if} \ i = j \ \mathbf{then} \ v \ \mathbf{else} \ \text{select}(a, j) \\ \text{const}(v) & := \lambda i. v \\ \text{map}_f(a, b) & := \lambda i. f(\text{select}(a, i), \text{select}(b, i)) \end{aligned}$$

- Takeaway: A common procedure for Array
Combinators

A reduction-based approach

$$\text{Sat}(T_{\text{Array}} \wedge \varphi)?$$



Use saturation rules to reduce arrays to the theory of un-interpreted functions

$$\text{Sat}(T_{\text{Equality}} \wedge \text{Closure}_{\text{Array}}(\varphi) \wedge \varphi)?$$



Extract models for arrays as finite graphs

```
sat
partitions:
*0 -> true
*1 -> false
*2 {a2} -> { *4 -> *5; *7 -> *12; else -> *13 }
*3 {a1} -> { *7 -> *12; else -> *13 }
*4 {i1} -> 1
*5 {v1} -> 2
*6 {a3} -> { *4 -> *5; *7 -> *8; else -> *13 }
*7 {i2 j} -> 3
*8 {v2} -> 7
*9 {a4} -> { *4 -> *5; *7 -> *8; *10 -> *11; else -> *13 }
*10 {i3} -> 4
```

Deciding *store*

For every sub-term $store(a, i, v)$, every index j in φ , add equation to φ :

$$select(store(a, i, v), j) = \mathbf{if} \ i = j \ \mathbf{then} \ v \ \mathbf{else} \ select(a, j)$$

EUF model of $\varphi \Rightarrow$ Array Model:

For each array a define

$$M_{array}(a) := \{ M(i) \rightarrow M(select(a, i)), else \rightarrow \blacklozenge_{Ma} \}$$

where $select(a, i)$ occurs in φ .

Deciding *store*

For each array a in φ define

$$M_{array}(a) := \{ M(i) \rightarrow M(select(a, i)), else \rightarrow \blacklozenge_{Ma} \}$$

Does M satisfy axioms for *store*?

$$M(store(a, i, v)) = \lambda j. \mathbf{if} M(i) = j \mathbf{then} M(v) \mathbf{else} M(select(a, j))$$

Recall, we added

$$select(store(a, i, v), j) = \mathbf{if} i = j \mathbf{then} v \mathbf{else} select(a, j)$$

Thus, $M(select(store(a, i, v), j))$

$$= M(\mathbf{if} i = j \mathbf{then} v \mathbf{else} select(a, j))$$

$$= \mathbf{if} M(i) = M(j) \mathbf{then} M(v) \mathbf{else} M(select(a, j))$$

Extensionality

$$\forall a, b \left(\left(\forall i . \text{select}(a, i) = \text{select}(b, i) \right) \Rightarrow a = b \right)$$

Not automatically satisfied by basic decision procedure.

Skolemized:

$$\forall a, b \left(\left(\text{select}(a, \delta(a, b)) = \text{select}(b, \delta(a, b)) \right) \Rightarrow a = b \right)$$

Add instance for every pair a, b .

More Efficiently Deciding *store*

- $a \sim b$ – a and b are equal in current context
- $a \equiv t$ – a is a name for the term t

$$\begin{array}{c}
 \text{idx} \frac{a \equiv \text{store}(b, i, v)}{a[i] \simeq v} \\
 \Downarrow \frac{a \equiv \text{store}(b, i, v), \quad w \equiv a'[j], \quad a \sim a'}{i \simeq j \vee a[j] \simeq b[j]} \\
 \Uparrow \frac{a \equiv \text{store}(b, i, v), \quad w \equiv b'[j], \quad b \sim b'}{i \simeq j \vee a[j] \simeq b[j]} \\
 \text{ext} \frac{a: (\sigma \Rightarrow \tau), \quad b: (\sigma \Rightarrow \tau)}{a \simeq b \vee a[k_{a,b}] \not\simeq b[k_{a,b}]}
 \end{array}$$

What makes it more *Efficient*?

- Axioms for *store* are only added by the model induced by EUF

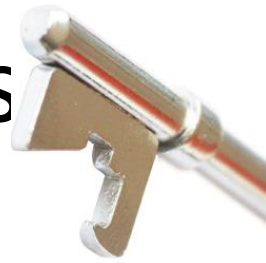
Bottlenecks

$$\text{ext} \frac{a: (\sigma \Rightarrow \tau), \quad b: (\sigma \Rightarrow \tau)}{a \simeq b \vee a[k_{a,b}] \not\simeq b[k_{a,b}]}$$

$$\uparrow \frac{a \equiv \text{store}(b, i, v), \quad w \equiv b'[j], \quad b \sim b'}{i \simeq j \vee a[j] \simeq b[j]}$$

- Extensionality axiom is instantiated on every pair of array variables.
- Upwards propagation distributes index over all modifications of same array.

Bottlenecks



$$\text{ext} \frac{a: (\sigma \Rightarrow \tau), \quad b: (\sigma \Rightarrow \tau)}{a \simeq b \vee a[k_{a,b}] \not\simeq b[k_{a,b}]}$$

Bottleneck:

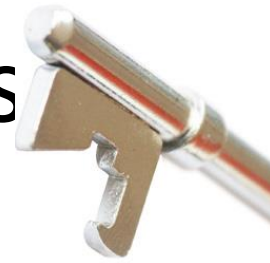
Extensionality axiom is instantiated on every pair of array variables.

$$\text{ext}_{\neq} \frac{p \equiv a \simeq b, \quad \Gamma(p) = \text{false}}{a \simeq b \vee a[k_{a,b}] \not\simeq b[k_{a,b}]}$$

$$\text{ext}_r \frac{a: (\sigma \Rightarrow \tau), \quad b: (\sigma \Rightarrow \tau), \quad \{a, b\} \subseteq \text{foreign}}{a \simeq b \vee a[k_{a,b}] \not\simeq b[k_{a,b}]}$$

Optimization: Restrict to variables asserted different, or shared.

Bottlenecks



$$\uparrow \frac{a \equiv \text{store}(b, i, v), \quad w \equiv b'[j], \quad b \sim b'}{i \simeq j \vee a[j] \simeq b[j]}$$

- **Bottleneck:** Upwards propagation distributes index over all modifications of same array.

$$\uparrow_r \frac{a \equiv \text{store}(b, i, v), \quad w \equiv b'[j], \quad b \sim b', \quad b \in \text{non-linear}}{i \simeq j \vee a[j] \simeq b[j]}$$

- **Optimization:** Only use \uparrow for updates where ancestor has multiple children. *Formulas from programs are well-behaved.*

Saturating K, map, δ

$$\text{K}\Downarrow \frac{a \equiv K(v), \quad w \equiv a'[j], \quad a \sim a'}{a[j] \simeq v}$$

$$\text{map}\Downarrow \frac{a \equiv \text{map}_f(b_1, \dots, b_n), \quad w \equiv a'[j], \quad a \sim a'}{a[j] \simeq f(b_1[j], \dots, b_n[j])}$$

$$\text{map}\Uparrow \frac{\begin{array}{l} a \equiv \text{map}_f(b_1, \dots, b_n), \quad w \equiv b'_k[j], \\ b_k \sim b'_k, \text{ for some } k \in \{1, \dots, n\} \end{array}}{a[j] \simeq f(b_1[j], \dots, b_n[j])}$$

$$\epsilon_{\neq} \frac{v \equiv a[i], \quad i:\sigma, \quad i \text{ is not } \epsilon_{\sigma}}{\epsilon_{\sigma} \neq i} \quad \epsilon_{\delta} \frac{a: (\sigma \Rightarrow \tau)}{a[\epsilon_{\sigma}] \simeq \delta_a}$$

Algebraic Data types

Scalars, Tuples and Composites

Fruit = Apple | Orange | Banana

Person = { name : String, age : Int, sex : M | F }

IntOption = Some of { ofSome : Int } | None

Recursive and Mutual Recursive types

List = Nil | Cons of { head : Int, tail : List }

Ping = DropP | WinP | Pi of { pong : Pong }

Pong = WinP | DropP | Po of { ping : Pong }

ADTs: Algebraic Data-types

- Constructors are injective:
 - $\text{head}(\text{cons}(x, xs)) = x$
 - $\text{tail}(\text{cons}(x, xs)) = xs$
- Terms are well-founded:
 - $xs \neq \text{cons}(x, xs)$
 - $xs \neq \text{cons}(x, \text{cons}(y, xs))$
 - $xs \neq \text{cons}(x, \text{cons}(y, \text{cons}(z, xs)))$
 - $xs \neq \text{cons}(x, \text{cons}(y, \text{cons}(z, \text{cons}(u, xs))))$

ADTs

- Outline of a decision Procedure:
 - Force injectivity:
 - For $\text{cons}(t1,t2)$ add lemmas:
 - $\text{head}(\text{cons}(t1,t2)) = t1$
 - $\text{tail}(\text{cons}(t1,t2)) = t2$
 - Build pre-model for constants of data-type sort.
 - $x = \text{Nil}$ $y = \text{Nil}$ $z = \text{Nil}$
 - Perform occurs check in each equivalence class.
 - Q: can there be two constructors in an equivalence class?