

Constrained Horn Clauses as a Basis of Automatic Program Verification: The Higher-order Case

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“Constrained Horn clauses are a suitable basis for automatic program verification, i.e., *symbolic model checking*.” [Bjørner et al. 2012]

Constrained means truth of formula is relative to a *decidable 1st-order background theory* (e.g. ZLA).

Example: safety verification

Recursive predicate $\left\{ \begin{array}{l} \forall \bar{x} . (Initial(\bar{x}) \Rightarrow Reach(\bar{x})) \\ \forall \bar{x} . (Reach(\bar{x}) \wedge Trans(\bar{x}, \bar{x}') \Rightarrow Reach(\bar{x}')) \end{array} \right.$

Query: $\forall \bar{x} . (Reach(\bar{x}) \Rightarrow Safe(\bar{x}))$

Solve for (unknown) predicate *Reach*, which defines an *inductive invariant*.

Many *algorithmic solutions*. Examples: CLP (Jaffar et al.); IC3 algorithms (Bradley); lazy annotation (Jaffar, McMillan, etc.).

Desirable features of Horn clauses

Horn clauses originated from theorem proving in 1st-order logic.

- 1 **Syntactic simplicity** eases presentation of proof procedure.
E.g. **1st-order resolution**: resolvent of two Horn clauses is a Horn clause.

- 2 **Solving satisfiability of Horn clause fragments is simpler**

Logic	Horn	General
Propositional	P	NP
Bernays-Schönfinkel ($\exists^* \forall^*$)	DEXPTIME	NEXPTIME

- 3 Horn clauses enjoy **least model property**:
 - ▶ useful for **model building**: as symbolic representation of partial models (even for non-Horn theories).

Why constrained Horn clauses (rather than model checking)?

- 1 **Expressivity**: Horn constraints can express standard verification proof rules, and encode safety, liveness, CTL+FO, and game solving. [Rybalchenko et al. PLDI12, POPL14]
- 2 **Adoption of standards** (i.e. SMT formats *and* Horn constraints) promotes
 - ▶ **exchange** of software model checking benchmarks
 - ▶ **separation of concerns**: let verification-condition generators worry about specificities of programming languages, whilst “model checking” is kept purely logical, and hence generic.
- 3 **Extensibility** and **retargetability** of verification tool (chain).

Why *higher-order* constrained Horn clauses?

The reasons above are just as applicable to higher-order computation! ... More on this anon.

- 1 Higher-order constrained Horn clauses (HoCHC): satisfiability and safety problems
- 2 Standard semantics of higher-order logic
- 3 Monotone semantics satisfies least model property
- 4 Algorithmic solutions of HoCHC safety problem: 1. via refinement types
- 5 Automation via prototype tool Horus

Outline

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Higher-order constrained Horn clauses arise *naturally* as **definitions of inductive invariants of higher-order programs**.

Example: safety verification

```
let add x y = x + y
letrec iter f s n = if n ≤ 0 then s else f n (iter f s (n - 1))
  in λn. assert (n ≤ (iter add 0 n))
```

- $(iter\ f\ s\ n)$ computes $f\ n\ (f\ (n - 1)\ (f\ (n - 2)\ (\dots\ (f\ 1\ s)\ \dots)))$.
- Thus $(iter\ add\ 0\ n) = n + (n - 1) + \dots + 1 + 0$.

Say the program is **safe** if assertion is never violated.

Example: safety verification

```
let add x y = x + y
letrec iter f s n = if n ≤ 0 then s else f n (iter f s (n - 1))
  in λn. assert (n ≤ (iter add 0 n))
```

An **inductive invariant** of a defined function is a relation overapproximating its input-output graph.

The system below describes the class of all **invariants sufficiently strong to guarantee the assertion**:

$$\begin{aligned} &\forall x y z. (z = x + y \Rightarrow \text{Add } x y z) \\ &\forall f s n m. (n \leq 0 \wedge m = s \Rightarrow \text{Iter } f s n m) \\ &\forall f s n m. \\ &\quad (n > 0 \wedge (\exists p. \text{Iter } f s (n - 1) p \wedge f n p m) \Rightarrow \text{Iter } f s n m) \\ &\forall n m. (\text{Iter } \text{Add } 0 n m \Rightarrow n \leq m) \end{aligned}$$

Some features of HoCHC

$$\begin{aligned} &\forall x y z . (z = x + y \Rightarrow \textit{Add } x y z) \\ &\forall f s n m . (n \leq 0 \wedge m = s \Rightarrow \textit{Iter } f s n m) \\ &\forall f s n m . \\ &\quad (n > 0 \wedge (\exists p . \textit{Iter } f s (n - 1) p \wedge f n p m) \Rightarrow \textit{Iter } f s n m) \\ &\forall n m . (\textit{Iter } \textit{Add } 0 n m \Rightarrow n \leq m) \end{aligned}$$

- Higher-order “unknown” relation:

$$\textit{Iter} : (\text{int} \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{bool}) \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{bool}$$

- Quantification at higher sort: $\text{int} \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{bool}$

- Literals headed by variables: $f n p m$

Every model of the system *is* an invariant witnessing safety of the program.

Higher-order constrained Horn clauses (HoCHC): definitions

Relational sorts: $\sigma ::= \text{int} \rightarrow \text{bool} \mid \text{int} \rightarrow \sigma \mid \sigma \rightarrow \sigma'$

Fix a sorting Δ of higher-order relational variables (“unknowns”)

goal $G ::= A \mid \varphi \mid G \wedge G \mid G \vee G \mid \exists x:\sigma. G$
definite $D ::= \text{true} \mid \forall x:\sigma. D \mid D \wedge D \mid G \Rightarrow X \ x_1 \dots x_n$

- A ranges over atoms e.g. $\text{Iter } f \ m \ (n - 1) \ p, \ f \ n \ p \ r$
- φ ranges over constraints e.g. $x > 3$
- X ranges over Δ e.g. Iter

- **Satisfiability Problem:** $\langle \Delta, D \rangle$ is solvable if for all models \mathcal{A} of background theory Th , there is valuation α of Δ s.t. $\mathcal{A}, \alpha \models D$.
- **Safety Problem:** $\langle \Delta, D, G \rangle$ is solvable if for all models \mathcal{A} of Th , there is valuation α of Δ s.t. $\mathcal{A}, \alpha \models D$, yet $\mathcal{A}, \alpha \not\models G$.

Example: an instance of HoCHC safety problem $\langle \Delta, D, G \rangle$

- (1) $\forall x y z . (z = x + y \Rightarrow \text{Add } x y z)$
- (2) $\forall f s n m . (n \leq 0 \wedge m = s \Rightarrow \text{Iter } f s n m)$
- (3) $\forall f s n m .$
 $(n > 0 \wedge (\exists p . \text{Iter } f s (n - 1) p \wedge f n p m) \Rightarrow \text{Iter } f s n m)$
- (4) $\forall n m . (\text{Iter } \text{Add } 0 n m \Rightarrow n \leq m)$

- Sorting Δ of relational variables:

$$\begin{cases} \text{Add} & : \text{int} \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{bool} \\ \text{Iter} & : (\text{int} \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{bool}) \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{int} \rightarrow \text{bool} \end{cases}$$

- Definite formula $D = (1) \wedge (2) \wedge (3)$.

- Goal formula $G = \neg(4) = \exists n m . ((\text{Iter } \text{Add } 0 n m) \wedge m < n)$.

Safety problem $\langle \Delta, D, G \rangle$ is solvable. I.e. w.r.t. the unique model of ZLA (\because complete theory), there is a valuation satisfying D but refuting G .

Systems of definite clauses can be presented (equivalently) in **program form**.

$$Add = \lambda x y z. (z = x + y)$$

$$Iter = \lambda f s n m. \left(\begin{array}{l} (n \leq 0 \wedge m = s) \\ \vee \exists p. 0 < n \wedge Iter f s (n - 1) p \wedge f n p m \end{array} \right)$$

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Standard semantics of higher-order logic

Sorts: $\sigma ::= \text{one} \mid \text{bool} \mid \text{int} \mid \sigma_1 \rightarrow \sigma_2$

$$\mathcal{S}[\text{one}] := \{\star\}$$

$$\mathcal{S}[\text{bool}] := \{0, 1\}$$

$$\mathcal{S}[\text{int}] := \mathbb{Z}$$

$$\mathcal{S}[\sigma_1 \rightarrow \sigma_2] := \mathcal{S}[\sigma_1] \Rightarrow \mathcal{S}[\sigma_2] \quad (\text{all functions})$$

Syntax: Standard presentation as a **simply-typed λ -calculus** with **logical constants**: $\neg, \wedge, \vee, \forall_\sigma, \exists_\sigma$, etc.

$$\neg : \text{bool} \rightarrow \text{bool} \quad \forall_\sigma, \exists_\sigma : (\sigma \rightarrow \text{bool}) \rightarrow \text{bool}$$

We write $\exists_\sigma(\lambda x:\sigma. M)$ as $\exists x:\sigma. M : \text{bool}$.

Semantics: completely standard.

Example: $\mathcal{A} \models_{\mathcal{S}} \exists x : (\text{int} \rightarrow \text{bool}) \rightarrow \text{bool} . G$

“There is some predicate x on sets of integers that makes G true in \mathcal{A} .”

Failure of least model property in standard semantics!

Counterexample:

$$\begin{cases} P : ((\text{one} \rightarrow \text{bool}) \rightarrow \text{bool}) \rightarrow \text{bool} \\ Q : \text{one} \rightarrow \text{bool} \end{cases}$$

$$\forall x : (\text{one} \rightarrow \text{bool}) \rightarrow \text{bool} . (x Q \Rightarrow P x)$$

Theorem

Satisfiable systems of higher-order constrained Horn clauses do not necessarily possess (unique) least models.

(Least with respect to inclusion of relations.)

$\forall x. (x Q \Rightarrow P x)$ has two minimal models (=valuations) α & β

$P : ((\text{one} \rightarrow \text{bool}) \rightarrow \text{bool}) \rightarrow \text{bool}$ $Q : \text{one} \rightarrow \text{bool}$

$\mathcal{S}[\text{one}] := \{\star\}$

$\mathcal{S}[\text{one} \rightarrow \text{bool}] := \left\{ \underbrace{\{\star \mapsto 0\}}_{-}, \underbrace{\{\star \mapsto 1\}}_{+} \right\}$

$\mathcal{S}[(\text{one} \rightarrow \text{bool}) \rightarrow \text{bool}] :=$

$\left\{ \underbrace{\left\{ \begin{array}{l} - \mapsto 0 \\ + \mapsto 1 \end{array} \right\}}_{\text{id}}, \underbrace{\left\{ \begin{array}{l} - \mapsto 0 \\ + \mapsto 0 \end{array} \right\}}_{\text{cst0}}, \underbrace{\left\{ \begin{array}{l} - \mapsto 1 \\ + \mapsto 1 \end{array} \right\}}_{\text{cst1}}, \underbrace{\left\{ \begin{array}{l} - \mapsto 1 \\ + \mapsto 0 \end{array} \right\}}_{\text{neg}} \right\}$

$\alpha(Q)$	$=$	$-$	$\beta(Q)$	$=$	$+$
$\alpha(P)(\text{id})$	$=$	0	$\beta(P)(\text{id})$	$=$	1
$\alpha(P)(\text{cst0})$	$=$	0	$\beta(P)(\text{cst0})$	$=$	0
$\alpha(P)(\text{cst1})$	$=$	1	$\beta(P)(\text{cst1})$	$=$	1
$\alpha(P)(\text{neg})$	$=$	1	$\beta(P)(\text{neg})$	$=$	0

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Monotone semantics of higher-order logic

Interpret \rightarrow as the **monotone function space**.

$$\mathcal{M}[\text{int}] := \mathbb{Z} \quad (\text{ordered discretely})$$

$$\mathcal{M}[\text{bool}] := \text{lattice } \{0, 1\} \text{ (or } \{f, t\}) \text{ with } 0 \sqsubseteq 1$$

$$\mathcal{M}[\sigma_1 \rightarrow \sigma_2] := \mathcal{M}[\sigma_1] \Rightarrow_m \mathcal{M}[\sigma_2] \quad (\text{monotone fns})$$

Example: $\mathcal{A} \models_{\mathcal{M}} \exists x : (\text{int} \rightarrow \text{bool}) \rightarrow \text{bool} . G$

“There is some **monotone** predicate x on sets of integers that makes G true in \mathcal{A} .”

In monotone semantics, satisfiable Horn clauses have least models (because “**immediate consequence operator**” is monotone) and constructible by Knaster-Tarski.

$\mathcal{M}[\text{int} \rightarrow \text{bool}]$ All sets of integers

$\mathcal{M}[(\text{int} \rightarrow \text{bool}) \rightarrow \text{bool}]$ All upward-closed (w.r.t. \subseteq) sets of sets of integers

$\mathcal{M}[(\text{int} \rightarrow \text{bool}) \rightarrow \text{bool}) \rightarrow \text{bool}]$ All upward-closed sets of upward-closed sets of sets of integers

Counter-intuitive (?) Take $x : (\text{int} \rightarrow \text{bool}) \rightarrow \text{bool}$.

$$x \mapsto \{\{1\}\} \not\models \exists y : (\text{int} \rightarrow \text{bool}). \exists z : \text{int}. (x y \wedge y z)$$

(\therefore valuation is invalid: $\{\{1\}\} \notin \mathcal{M}[(\text{int} \rightarrow \text{bool}) \rightarrow \text{bool}]$)

Standard Semantics

😊 Completely standard satisfiability problem (modulo background theory) in higher-order logic.

😞 No least model.

Monotone Semantics

😞 Bespoke satisfiability problem with a restricted class of models.

😊 Least model arising in the usual way.

Can we have the best of both worlds?

I.e. can we **specify** problems in standard semantics, but solve / **compute** in monotone semantics?

Standard and monotone semantics are equivalent for the HoCHC Satisfiability Problem

We *can* have the best of both worlds!

Theorem (Model correspondence)

Given a clause (set) H , H is satisfiable in the standard semantics iff H is satisfiable in the monotone semantics.

Proof idea

For each sort of relations ρ , **monotone** and **standard** semantics are locked in **two-sided Galois connections**:

$$\begin{array}{ccccc} \mathcal{S}[\rho] & \begin{array}{c} \xleftarrow{I_\rho} \\ \xrightarrow{L_\rho} \end{array} & \mathcal{M}[\rho] & \begin{array}{c} \xleftarrow{U_\rho} \\ \xrightarrow{J_\rho} \end{array} & \mathcal{S}[\rho] \\ \text{Standard} & & \text{Monotone} & & \text{Standard} \end{array}$$

Define, by recursion over sorts:

$$\begin{array}{ll} I_{\text{bool}}(b) & := b & J_{\text{bool}}(b) & := b \\ I_{\text{int} \rightarrow \rho}(r) & := I_\rho \circ r & J_{\text{int} \rightarrow \rho}(r) & := J_\rho \circ r \\ I_{\rho_1 \rightarrow \rho_2}(r) & := I_{\rho_2} \circ r \circ L_{\rho_1} & J_{\rho_1 \rightarrow \rho_2}(r) & := J_{\rho_2} \circ r \circ U_{\rho_1} \end{array}$$

where

- U_ρ is the **right adjoint** of J_ρ , i.e., uniquely determined by: for all a, b

$$J_\rho a \subseteq b \Leftrightarrow a \subseteq U_\rho b$$

- L_ρ is the **left adjoint** of I_ρ

Standard and monotone semantics are equivalent also for HoCHC Safety Problem

Theorem (Equivalence / Inter-reducibility)

For all Δ, D and G , T.F.A.E.

- (i) HoCHC Safety Problem $\langle \Delta, D, G \rangle$ in *standard semantics* is solvable
- (ii) HoCHC Safety Problem $\langle \Delta, D, G \rangle$ in *monotone semantics* is solvable
- (iii) In all models of the background theory, the *least valuation* $\mathcal{M}[[D]]$ invalidates G (i.e. $\mathcal{M}[[G]](\mathcal{M}[[D]]) = 0$).

Thus: we can specify problems using the standard semantics, and then solve in the monotone semantics.

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Refinement types as higher-order invariants

Solving HoCHC problems is about finding (higher-order) **symbolic models**. Models are **valuations**.

$$R : ((\text{int} \rightarrow \text{bool}) \rightarrow \text{bool}) \rightarrow \text{bool}$$

Symbolic model:

$$R \mapsto \lambda f. (\forall g. f\ g \Rightarrow (\forall x. g\ x \Rightarrow \phi) \Rightarrow \psi) \Rightarrow \chi$$

Dependent refinement type:

$$R : f : (g : (x : \text{int} \rightarrow \text{bool}\langle\phi\rangle) \rightarrow \text{bool}\langle\psi\rangle) \rightarrow \text{bool}\langle\chi\rangle$$

Dependency – “ $f : x : T_1 \rightarrow T_2$ ” means: for each $a : T_1$, the value of $f\ a$ has type $T_2[a/x]$.

Refinement – “ $b : \text{bool}\langle\varphi\rangle$ ” means: $b \Rightarrow \varphi$

Dependent refinement types: syntax and semantics

$$T \quad := \quad \text{bool}\langle\varphi\rangle \mid x:\text{int} \rightarrow T \mid T_1 \rightarrow T_2$$

Refinement at **bool**: φ is a 1st-order formula of constraint language

Dependence at **int**: x can occur freely in T

Order-ideal semantics: Given a valuation α of int-sorted vars:

$$\llbracket \text{int} \rrbracket(\alpha) := \mathbb{Z}$$

$$\llbracket \text{bool}\langle\varphi\rangle \rrbracket(\alpha) := \{\text{f}, \llbracket \varphi \rrbracket(\alpha)\}$$

$$\llbracket x:\text{int} \rightarrow T \rrbracket(\alpha) := \prod_{d \in \mathbb{Z}} \llbracket T \rrbracket(\alpha[x \mapsto d])$$

Examples of refinement type

Idea: $\llbracket \text{bool}\langle\varphi\rangle \rrbracket(\alpha)$ is downward closure of value of φ .

$$1. \llbracket \text{bool}\langle x \leq y \rangle \rrbracket(\{x \mapsto 1, y \mapsto 2\}) = \{\text{f}, \text{t}\}$$

Fact: $b \in \llbracket \text{bool}\langle\varphi\rangle \rrbracket(\alpha) \Leftrightarrow \alpha \vDash b \Rightarrow \varphi$

2. A function type.

$$\begin{aligned} & \llbracket x:\text{int} \rightarrow \text{bool}\langle\varphi\rangle \rrbracket \\ = & \prod n \in \mathbb{Z}. \llbracket \text{bool}\langle\varphi[n/x]\rangle \rrbracket \\ = & \{f \mid \forall n \in \mathbb{Z}. f\ n \in \llbracket \text{bool}\langle\varphi[n/x]\rangle \rrbracket\} \\ = & \{f \mid \forall n \in \mathbb{Z}. (f\ n \Rightarrow \varphi[n/x])\} \\ = & \{f \mid \forall x : \text{int}. (f\ x \Rightarrow \varphi)\} \end{aligned}$$

Decidable judgement: $\Gamma \vdash G : T$

- Type environment Γ : finite map from variables to refinement types
- Goal term G : subterm of body of Horn clause
- T : refinement type

Intuition:

$$\Gamma \vdash G : \text{bool}\langle\varphi\rangle$$

In symbolic model Γ (i.e. models satisfying Γ), truth of G is bounded above by constraint φ (or “ G implies φ ”).

Thus φ is an over-approximation of G , which may have higher-order subterms.

Some proof rules of typing judgements

$$\text{(TConstraint)} \quad \frac{}{\Gamma \vdash \varphi : \text{bool}\langle\varphi\rangle} \varphi \in Fm$$

$$\text{(TExists)} \quad \frac{\Gamma, x : \iota \vdash G : \text{bool}\langle\varphi\rangle}{\Gamma \vdash \exists x:\iota. G : \text{bool}\langle\psi\rangle} Th \models \varphi \Rightarrow \psi$$

$$\text{(TAbsl)} \quad \frac{\Gamma, x : \text{int} \vdash G : T}{\Gamma \vdash \lambda x:\text{int}. G : x:\text{int} \rightarrow T}$$

$$\text{(TAppl)} \quad \frac{\Gamma \vdash G : x:\text{int} \rightarrow T \quad \Gamma \vdash N : \text{int}}{\Gamma \vdash G N : T[N/x]}$$

$$\text{(TSub)} \quad \frac{\Gamma \vdash G : T_1 \quad \vdash T_1 \sqsubseteq T_2}{\Gamma \vdash G : T_2}$$

(Nothing surprising here.)

Subtyping judgement: $\vdash T_1 \sqsubseteq T_2$

Subtyping, \sqsubseteq , captures **implication** in the background theory Th .

$$\frac{}{\vdash \text{bool}\langle\varphi\rangle \sqsubseteq \text{bool}\langle\psi\rangle} (Th \models \varphi \Rightarrow \psi)$$

$$\frac{\vdash T_1 \sqsubseteq T_2}{\vdash x:\text{int} \rightarrow T_1 \sqsubseteq x:\text{int} \rightarrow T_2}$$

$$\frac{\vdash T'_1 \sqsubseteq T_1 \quad \vdash T_2 \sqsubseteq T'_2}{\vdash T_1 \rightarrow T_2 \sqsubseteq T'_1 \rightarrow T'_2}$$

Theorem (Soundness)

If $\Gamma \vdash G : T$ then $\Gamma \models G : T$.

A sound approach to solving HoCHC

Given HoCHC safety problem $\langle \Delta, D, G \rangle$:

- If there is a type environment Γ (that refines Δ) such that $\vdash D : \Gamma$ and $\Gamma \vdash G : \text{bool}\langle \text{false} \rangle$, then for each model \mathcal{A} of background theory, $\mathcal{M}[\llbracket \Gamma \rrbracket]$ is a valuation that satisfies D but refutes G .
- Typability of clauses, $\Gamma \vdash G : T$, is reducible to 1st-order constrained Horn clause solving. This is more or less standard.

The method is incomplete.

Working example revisited

From safety verification problem:

```
let add x y = x + y
letrec iter f s n = if n ≤ 0 then s else f n (iter f s (n - 1))
  in λn. assert (n ≤ (iter add 0 n))
```

obtain HoCHC safety problem:

$$\begin{aligned} &\forall x y z. (z = x + y \Rightarrow \text{Add } x y z) \\ &\forall f s n m. (n \leq 0 \wedge m = s \Rightarrow \text{Iter } f s n m) \\ &\forall f s n m. \\ &\quad (n > 0 \wedge (\exists p. \text{Iter } f s (n - 1) p \wedge f n p m) \Rightarrow \text{Iter } f s n m) \\ &\forall n m. (\text{Iter } \text{Add } 0 n m \Rightarrow n \leq m) \end{aligned}$$

Goal clause: $G = \exists m n. (\text{Iter } \text{Add } 0 n m) \wedge n > m$

Task (type checking): Find Γ s.t. $\vdash D : \Gamma$ and $\Gamma \vdash G : \text{bool}\langle \text{false} \rangle$.

Working example problem: a solution

Model = valuation, here expressed as refinement type assignment

$$\left\{ \begin{array}{l} \textit{Add} \mapsto x:\textit{int} \rightarrow y:\textit{int} \rightarrow z:\textit{int} \rightarrow \textit{bool}\langle z = x + y \rangle \\ \textit{Iter} \mapsto (x:\textit{int} \rightarrow y:\textit{int} \rightarrow z:\textit{int} \rightarrow \textit{bool}\langle 0 < x \Rightarrow y < z \rangle) \rightarrow \\ \quad s:\textit{int} \rightarrow n:\textit{int} \rightarrow m:\textit{int} \rightarrow \textit{bool}\langle 0 \leq s \Rightarrow n \leq m \rangle \end{array} \right.$$

For example...

$r : \text{int}$

$n : \text{int}$

$\text{Add} : (x : \text{int} \rightarrow \dots \rightarrow \text{bool}\langle z = x + y \rangle)$

$\text{Iter} : (x : \text{int} \rightarrow \dots \rightarrow \text{bool}\langle 0 < x \Rightarrow y < z \rangle) \rightarrow \dots \rightarrow \text{bool}\langle 0 \leq m \Rightarrow n \leq r \rangle$



Γ

$$\overline{\Gamma \vdash > : (x:\text{int} \rightarrow y:\text{int} \rightarrow \text{bool}\langle x > y \rangle)}$$
$$\overline{\Gamma \vdash n : \text{int}}$$
$$\Gamma \vdash > n : (y:\text{int} \rightarrow \text{bool}\langle n > y \rangle)$$
$$\Gamma \vdash r : \text{int}$$
$$\Gamma \vdash n > r : \text{bool}\langle n > r \rangle$$

$$T_1 = (x: \text{int} \rightarrow \dots \rightarrow \text{bool}\langle 0 < x \Rightarrow y < z \rangle) \rightarrow m: \text{int} \rightarrow \dots \rightarrow \text{bool}\langle 0 \leq m \Rightarrow n \leq r \rangle$$

$$\Gamma \vdash \text{Add} : (x: \text{int} \rightarrow \dots \rightarrow \text{bool}\langle z = x + y \rangle)$$

$$\Gamma \vdash \text{Iter} : T_1$$

$$\Gamma \vdash \text{Add} : (x: \text{int} \rightarrow \dots \rightarrow \text{bool}\langle 0 < x \Rightarrow y < z \rangle)$$

$$\Gamma \vdash \text{Iter Add} : x: \text{int} \rightarrow \dots \rightarrow \text{bool}\langle 0 \leq x \Rightarrow y \leq z \rangle$$

●
●
●

$$\Gamma \vdash \text{Iter Add } 0 \ n \ r : \text{bool}\langle n \leq r \rangle$$

$$\Lambda : \forall XY. \text{bool}\langle X \rangle \rightarrow \text{bool}\langle Y \rangle \rightarrow \text{bool}\langle X \wedge Y \rangle$$

$$\Gamma \vdash \Lambda : \text{bool}\langle n \leq r \rangle \rightarrow \text{bool}\langle n > r \rangle \rightarrow \text{bool}\langle n \leq r \wedge n > r \rangle$$

•
•
•

$$\Gamma \vdash (\textit{Iter Add 0 n r}) \wedge (n > r) : \text{bool}\langle n \leq r \wedge n > r \rangle$$

$$\Gamma \vdash (\textit{Iter Add 0 n r}) \wedge (n > r) : \text{bool}\langle \text{false} \rangle$$

Type inference: how to grow a symbolic model

1. From higher-order relational vars: $\begin{cases} R : (\text{int} \rightarrow \text{bool}) \rightarrow \text{bool} \\ S : \text{int} \rightarrow (\text{int} \rightarrow \text{bool}) \end{cases}$
2. Create refinement template:

$$\Gamma = \begin{cases} R : (x:\text{int} \rightarrow \text{bool}\langle Z_1 x \rangle) \rightarrow \text{bool}\langle Z_2 \rangle \\ S : y:\text{int} \rightarrow (x:\text{int} \rightarrow \text{bool}\langle Z_3 x y \rangle) \end{cases}$$

3. Check that type environment Γ is a model.
4. Except, whenever forced to check the validity of an implication:

$$\frac{Th \models Z_3 n z \Rightarrow Z_1 z}{\text{bool}\langle Z_3 n z \rangle \sqsubseteq \text{bool}\langle Z_1 z \rangle} \text{ (Sub-Bool)}$$

add clause ' $Z_3 n z \Rightarrow Z_1 z$ ' to the (1st-order) Horn constraint system.

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Web interface to **Horus**: <http://mjolnir.cs.ox.ac.uk/horus>

Tests

Verification problems taken from **MoCHi** test suite (Kobayashi et al. PLDI'11) but reexpressed as HoCHC safety problems.

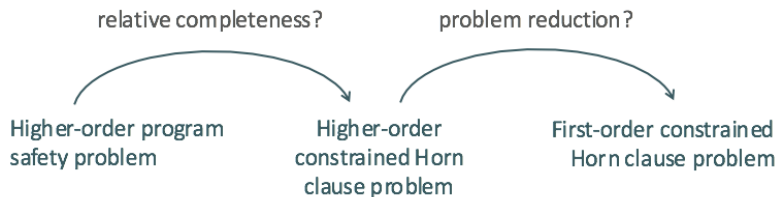
In all the examples (without local assertions), except neg:

- Horus takes around 0.01s to transform the system of clauses and
- Z3 takes around 0.02s to solve the transformed 1st-order system.

Example. In Problem mc91, we verify: $M(n) = 91$ for all $n \leq 101$.

<https://github.com/penteract/HigherOrderHornRefinement>

Related work: see paper on arXiv.



- 1 Other approaches to reduce HoCHC problems to 1st-order problems (e.g. via Reynolds' [defunctionalisation](#))
- 2 Adequacy of HoCHC for safety verification of higher-order programs in general (*cf.* Blass & Gurevich)