

Transforming Constrained Horn Clauses for Program Verification

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Overview

1. Rule-based transformations: Fold/Unfold transformations, CHC specialization, predicate tupling
2. Generating verification conditions via CHC specialization
3. CHC specialization as CHC solving
4. Increasing the power of CHC solving via predicate tupling

Constrained Horn Clauses (CHCs)

- First order formulas of the form:

$$A_1 \wedge \dots \wedge A_n \wedge c \rightarrow A_0$$

where A_0, A_1, \dots, A_n are atomic formulas and c is a formula in a theory Th of constraints (any first-order theory). All variables are assumed to be universally quantified in front.

- Prolog-like syntax:

$$A_0 \text{ :- } c, A_1, \dots, A_n.$$

CHCs for Program Verification

Program *sumupto*: summing the first n integers

```
x=0; y=0; while (x<n) { x=x+1; y=x+y}
```

Specification

```
{n≥1} SumUpto {y≥x}
```

Translation

Constrained Horn Clauses

```
false :- N≥1, X=0, Y=0, p(X, Y, N). %Init
```

```
p(X, Y, N) :- X<N, X1=X+1, Y1=X+Y, p(X1, Y1, N). %Loop
```

```
p(X, Y, N) :- X≥ N, Y<X. %Exit
```

- The program satisfies the specification iff the set of CHCs is **satisfiable**.
- Satisfiability of CHCs is **undecidable**: no ultimate verifier exists.
- Two ways for improving the effectiveness of CHC-based verification:
 1. Designing smart heuristics for satisfiability **solvers**;
 2. **Transforming** difficult CHC problems into equisatisfiable, easy ones.(1 and 2 not mutually exclusive)

1. Rule-based Transformations

Transformations of Functional and Logic Programs

Main idea of this talk: Transformation techniques introduced for improving functional and logic programs [Burstall-Darlington 1977, Tamaki-Sato 1984] can be adapted to ease satisfiability proofs for CHCs.

Initial program $P_0 \rightarrow P_1 \rightarrow \dots \rightarrow P_n$ Final program

where ' \rightarrow ' is an application of a transformation rule.

- Each rule application preserves the semantics:
 $M(P_0) = M(P_1) = \dots = M(P_n)$
- The application of the rules is guided by a strategy that guarantees that P_n is more efficient than P_0 .

Transformation Rules for CHCs

Initial clauses

$S_0 \rightarrow S_1 \rightarrow \dots \rightarrow S_n$

Final clauses

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introduce $C: \text{newp}(X) :- c, G$

$S_{i+1} = S_i \cup \{C\}$ $\text{Defs} := \text{Defs} \cup \{C\}$

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R2. **Unfolding.** Apply a Resolution step

given C: $H \text{ :- } c, A, G$ $A \text{ :- } d_1, G_1 \dots A \text{ :- } d_m, G_m$ in S_i

derive $S_i = \{ H \text{ :- } c, d_1, G_1, G \dots H \text{ :- } c, d_m, G_m, G \}$

$S_{i+1} = (S_i - \{C\}) \cup S$

... Transformation rules for CHCs

R3. **Folding.** Replace a conjunction by a new predicate

given C: $H :- d, B, G$ in S_i $\text{newp}(X) :- c, B.$ with $d \rightarrow c$ in Defs

derive D: $H :- d, \text{newp}(X), G.$

$$S_{i+1} = (S_i - \{C\}) \cup \{D\}$$

... Transformation rules for CHCs

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R4. **Constraint replacement.** Replace a constraint by an equivalent one

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derive D: $H :- d, B, G$

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... Transformation rules for CHCs

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... Transformation rules for CHCs

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$$S_{i+1} = (S_i - \{C\})$$

Theorem [Tamaki-Sato 84, Etalle-Gabbrielli 96]: If every new definition is unfolded at least once in $S_0 \rightarrow S_1 \rightarrow \dots \rightarrow S_n$ then **S_0 satisfiable iff S_n satisfiable.**

Transformation Strategies

- Transformation rules need to be guided by suitable **strategies**.
- Main idea: exploit some knowledge about the query to produce a customized, easier to verify set of clauses.
- **Specialization** [Gallagher,Leuschel,FPP,...]: Given a set of clauses S and a query **false :- c,A**, where **A** is atomic, transform S into a set of clauses S_{sp} such that
$$S \cup \{\text{false} \text{ :- } c,A\} \text{ satisfiable} \iff S_{sp} \cup \{\text{false} \text{ :- } c,A\} \text{ satisfiable.}$$
- **Predicate Tupling** (also known as **Conjunctive Partial Deduction**) [PP, Leuschel,...]: Given a set of clauses S and a query **false :- c,G**, where **G** is a (non-atomic) conjunction, introduce a new predicate **newp(X) :- G** and transform set of clauses S_T such that
$$S \cup \{\text{false} \text{ :- } c,G\} \text{ satisfiable} \iff S_T \cup \{\text{false} \text{ :- } c,\text{newp}(X)\} \text{ satisfiable.}$$

Specialization Strategy: An Example

```
false :- X<0, p(X,b).
```

```
p(X,C) :- X=Y+1, p(Y,C).
```

```
p(X,a).
```

```
p(X,b) :- X≥0, tm_halts(X).
```

$\% \forall X. p(X,b) \rightarrow X \geq 0$

$\% \text{ the } X\text{-th Turing machine halts on } X$

S_0

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%  $\forall X. p(X,b) \rightarrow X \geq 0$ 
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S_0

```
% the X-th Turing machine halts on X
```

Define: `q(X) :- X<0, p(X,b).`

```
% q(X) is a specialization of p(X,C)
```

S_1

```
% to a specific constraint on X and value of C
```


Specialization Strategy: An Example

```
false :- X<0, p(X,b).
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p(X,C) :- X=Y+1, p(Y,C).
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S_0

```
% the X-th Turing machine halts on X
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```
Define: q(X) :- X<0, p(X,b).
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S_1

```
% to a specific constraint on X and value of C
```

```
Unfold: q(X) :- X<0, X=Y+1, p(Y,b).
```

S_2

```
q(X) :- X<0, X≥0, tm_halts(X).
```

Specialization Strategy: An Example

```
false :- X<0, p(X,b).
```

```
p(X,C) :- X=Y+1, p(Y,C).
```

```
p(X,a).
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p(X,b) :- X≥0, tm_halts(X).
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```
%  $\forall X. p(X,b) \rightarrow X \geq 0$ 
```

S_0

```
% the X-th Turing machine halts on X
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Define: `q(X) :- X<0, p(X,b).`

```
% q(X) is a specialization of p(X,C)
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S_1

```
% to a specific constraint on X and value of C
```

Unfold: `q(X) :- X<0, X=Y+1, p(Y,b).`

S_2

```
q(X) :- X<0, X≥0, tm_halts(X).
```

```
% clause removal
```

Specialization Strategy: An Example

false :- X<0, p(X,b).

p(X,C) :- X=Y+1, p(Y,C).

p(X,a).

p(X,b) :- X≥0, tm_halts(X).

% $\forall X. p(X,b) \rightarrow X \geq 0$

S₀

% the X-th Turing machine halts on X

Define: q(X) :- X<0, p(X,b).

% q(X) is a specialization of p(X,C)

S₁

% to a specific constraint on X and value of C

Unfold: q(X) :- X<0, X=Y+1, p(Y,b).

S₂

~~q(X) :- X<0, X≥0, tm_halts(X).~~

% clause removal

Fold: false :- X<0, q(X).

q(X) :- X<0, X=Y+1, q(Y).

S₃

Satisfiability of S₃ is easy to check: q(X) \equiv false makes all clauses true (no facts for q)

Tupling Strategy: An Example

$\text{asum}(A,I,N,S)$: “the sum of the elements of array A from index I to N is S ”

$\text{amax}(A,I,N,M)$: “the largest element of array A from index I to N is M ”

$\text{false} :- S < M, 0 \leq I < N, \text{asum}(A,I,N,S), \text{amax}(A,I,N,M).$

$\text{asum}(A,I,N,S) :- I = N, S = 0.$

$\text{asum}(A,I,N,S) :- 0 \leq I < N, I1 = I + 1, \text{read}(A,I,X), X \geq 0, S = S1 + X, \text{asum}(A,I1,N,S1).$

$\text{amax}(A,I,N,M) :- I = N, M = 0.$

$\text{amax}(A,I,N,M) :- 0 \leq I < N, I1 = I + 1, \text{read}(A,I,X), X \geq 0,$
 $((X \geq M1, M = X) \vee (X < M1, M = M1)), \text{amax}(A,I1,N,M1).$

S_0

Proof of satisfiability:

$\text{asum}(A,I,N,S) \rightarrow \forall K,Y.(I \leq K < N, \text{read}(A,K,Y) \rightarrow S \geq Y)$
 $\text{amax}(A,I,N,M) \rightarrow \exists K.(I \leq K < N, \text{read}(A,K,M))$ } $\rightarrow S \geq M$

The proof uses **quantified array** constraints.

Eldarica (with the SimpleArray(1) theory) **does not solve** these clauses.

Tupling Strategy: An Example

false :- S < M, 0 ≤ I < N, asum(A, I, N, S), amax(A, I, N, M).

S_0

Tupling Strategy: An Example

false :- $S < M$, $0 \leq I < N$, $\text{asum}(A, I, N, S)$, $\text{amax}(A, I, N, M)$.

S_0

Define: $\text{asummax}(A, I, N, S, M)$:- $\text{asum}(A, I, N, S)$, $\text{amax}(A, I, N, M)$.

S_1

Tupling Strategy: An Example

false :- $S < M$, $0 \leq I < N$, $\text{asum}(A, I, N, S)$, $\text{amax}(A, I, N, M)$.

S_0

Define: $\text{asummax}(A, I, N, S, M)$:- $\text{asum}(A, I, N, S)$, $\text{amax}(A, I, N, M)$.

S_1

Unfold: $\text{asummax}(A, I, N, S, M)$:- $I = N - 1$, $S = 0$, $M = 0$.

S_2

& CR $\text{asummax}(A, I, N, S, M)$:- $0 \leq I$, $I < N - 1$, $X \geq 0$, $I_1 = I + 1$, $S = S_1 + X$, $\text{read}(A, I, X)$,
 $((X \geq M_1, M = X) \vee (X < M_1, M = M_1))$,
 $\text{asum}(A, I_1, N, S_1)$, $\text{amax}(A, I_1, N, M_1)$.

Tupling Strategy: An Example

false :- S < M, 0 ≤ I < N, **asum(A,I,N,S), amax(A,I,N,M)**. S₀

Define: **asummax(A,I,N,S,M)** :- **asum(A,I,N,S), amax(A,I,N,M)**. S₁

Unfold: **asummax(A,I,N,S,M)** :- I = N - 1, S = 0, M = 0. S₂

& CR **asummax(A,I,N,S,M)** :- 0 = < I, I < N - 1, X ≥ 0, I1 = I + 1, S = S1 + X, read(A,I,X),
((X ≥ M1, M = X) ∨ (X < M1, M = M1)),
asum(A,I1,N,S1), amax(A,I1,N,M1).

Fold: **asummax(A,I,N,S,M)** :- I = N - 1, S = 0, M = 0. S₃

asummax(A,I,N,S,M) :- 0 = < I, I < N - 1, X ≥ 0, I1 = I + 1, S = S1 + X, read(A,I,X),
((X ≥ M1, M = X) ∨ (X < M1, M = M1)), **asummax(A,I1,N,S1,M1)**.

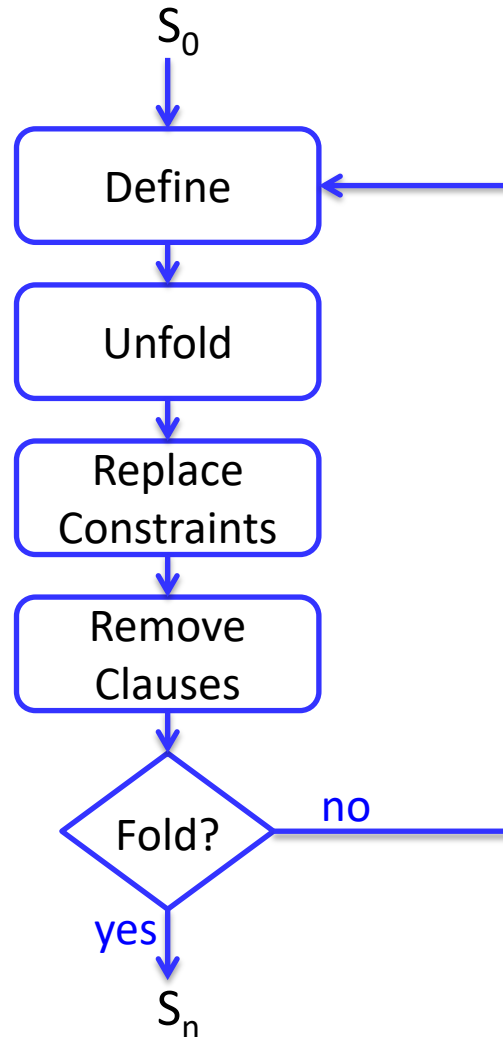
false :- S < M, 0 ≤ I < N, **asummax(A,I,N,S,M)**.

Proof of satisfiability: **asummax(A,I,N,S,M)** → S ≥ M

The proof only uses **linear arithmetic** constraints.

Eldarica (with the SimpleArray(1) theory) **solves** these clauses.

A Generic U/F Transformation Strategy

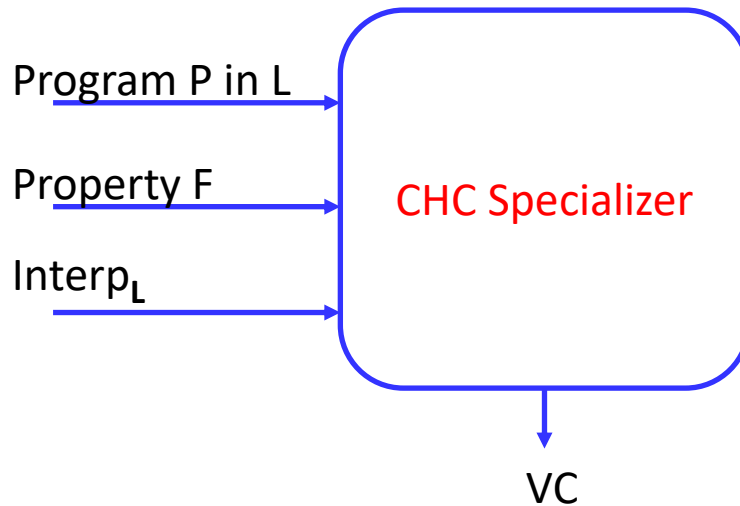


Some Issues About the U/F Strategy

- **Unfolding:** Which atoms should be unfolded? When to stop?
- **Constraint replacement:** A suitable constraint reasoner is needed
- **Definition:** Suitable new predicates need to be introduced to guarantee **termination** and **effectiveness** of strategy

2. Generating Verification Conditions via CHC Specialization

CHC Specialization as a Verification Condition Generator



L: Programming language

Interp_L: CHC interpreter for L

VC: Verification Conditions, i.e.,
a set of CHCs independent of L

F holds for P iff VC is satisfiable

The CHC specializer is **parametric with respect to the programming language L** and the class of properties.

Translating Imperative Programs into CHC

- C-like imperative language with assignments, conditionals, jumps. While-loops translated to conditionals and jumps.
- Commands encoded as atomic assertions: `at(Label, Cmd)`.

<code>x=0;</code>	0. <code>x=0;</code>	<code>at(0,asgn(int(x), int(0))).</code>
<code>y=0;</code>	1. <code>y=0;</code>	<code>at(1,asgn(int(y), int(0))).</code>
<code>while (x<n) {</code>	2. <code>if (x<n) 3 else 6;</code>	<code>at(2, ite(less(int(x), int(n)), 3, 6)).</code>
<code>x=x+1;</code>	3. <code>x=x+1;</code>	<code>at(3, asgn(int(x), plus(int(x), int(1)))).</code>
<code>y=x+y</code>	4. <code>y=x+y;</code>	<code>at(4, asgn(int(y), plus(int(x), int(y)))).</code>
<code>}</code>	5. <code>goto 2;</code>	<code>at(5, goto(2)).</code>
	6. <code>halt</code>	<code>at(6, halt).</code>

A Small-Step Operational Semantics

- The operational semantics is a **one-step transition relation** between **configurations**

$$\langle n1:cmd1, env1 \rangle \Rightarrow \langle n2:cmd2, env2 \rangle$$

where: $n:cmd$ is a labelled command and env is an environment mapping variable identifiers to values;

- **Assignment**

$$\langle n: x=e, env \rangle \Rightarrow \langle next(n), update(env, x, [e]env) \rangle$$

where: $next(n)$ is the next labelled command and $update(env, x, [e]env)$ updates the value of x in env to the value of expression e in env ;

- **Conditional**

$$\begin{array}{ll} \langle n: \text{if } (e) \text{ } n1 \text{ else } n2, env \rangle \Rightarrow \langle next(n1), env \rangle & \text{if } [e]\delta \neq 0 \\ \langle n: \text{if } (e) \text{ } n1 \text{ else } n2, env \rangle \Rightarrow \langle next(n2), env \rangle & \text{if } [e]\delta = 0 \end{array}$$

- **Jump**

$$\langle n: \text{goto } n1, env \rangle \Rightarrow \langle next(n1), env \rangle$$

A CHC Interpreter for the Small-Step Semantics

- **Configurations:** $cf(LC, Env)$
where:
 - LC is a labelled command represented by a term of the form $cmd(L,C)$,
L is a label, C is a command
 - Env is an environment represented as a list of (variable-id,value) pairs:
 $[(x,X),(y,Y),(z,Z)]$
- **One-step transition relation** between configurations:
 $tr(cf(LC1,Env1), cf(LC2,Env2))$

CHC Interpreter (Asgn)

```
assignment  x=e;
```

```
tr( source configuration      target configuration  
tr( cf(cmd(L, asgn(X,E)), Env1), cf(cmd(L1, C), Env2) ) :-  
  nextlab(L,L1),           % next label  
  at(L1,C),                % next command  
  eval(E,Env1,V),         % evaluate expression  
  update(Env1,X,V,Env2).  % update environment
```

More clauses for predicate tr to encode the semantics of the other commands.

Encoding Partial Correctness Properties

- Partial correctness specification (Hoare triple):

$$\{\varphi\} \text{ prog } \{\psi\}$$

If the initial values of the program variables satisfy the precondition φ and *prog* terminates, **then** the final values of the program variables satisfy the postcondition ψ .

- CHC encoding of partial correctness:

false :- initConf(Cf), reach(Cf).		<i>PC-prop</i>
reach(Cf1) :- tr(Cf1,Cf2), reach(Cf2).	PC property	
reach(Cf) :- errorConf(Cf).		
initConf(cf(C, Env)) :- at(0,C), $\varphi(\text{Env})$.	Initial configuration	
errorConf(cf(C, Env)) :- at(h,C), $-\psi(\text{Env})$.	Error configuration	
tr(cf1,cf2) :- ...	Interp _t	

- $\{\varphi\} \text{ prog } \{\psi\}$ is **valid** iff *PC-prop* is **satisfiable**.

VCGen: Generating Verification Conditions

- *VCGen* is a transformation strategy that **specializes** *PC-prop* to a given $\{\varphi\} prog \{\psi\}$, and removes explicit reference to the interpreter (function **cf**, predicates **at**, **tr**, etc.).
- All new definitions are of the form $newp(X) :- reach(cf)$, corresponding to a program point.
- Limited reasoning about constraints at specialization time (satisfiability only).
- *VCGen* is parametric wrt $Interp_L$ (to a large extent).
- If $PC-prop \xrightarrow{VCGen} VC$ then *PC-prop* is **satisfiable** iff *VC* is **satisfiable**

Generating Verification Conditions: An Example

PC property:

$\{n \geq 1\} \text{SumUpto } \{y > x\}$

CHC encoding:

```
false :- initConf(Cf), reach(Cf). PC-prop
reach(Cf1) :- tr(Cf1,Cf2), reach(Cf2).
reach(Cf) :- errorConf(Cf).
initConf(cf(C, [(x,X),(y,Y),(n,N)])) :- at(0,C),  $N \geq 1$ .
errorConf(cf(C, [(x,X),(y,Y),(n,N)])) :- at(6,C),  $Y \leq X$ .
tr(cf1,cf2) :- ...
...
at(0,asgn(int(x), int(0))).
...
```

VCGen



Verification
Conditions:

```
false :-  $N \geq 1$ , X=0, Y=0, p(X, Y, N). VC
p(X, Y, N) :- X < N, X1=X+1, Y1=Y+2, p(X1, Y1, N).
p(X, Y, N) :- X  $\geq$  N,  $Y < X$ .
```

Experimental evaluation

- Other semantics: multi-step for recursive functions, exceptions, etc.
- Checking the satisfiability of the VCs using QARMC, Z3 (PDR), MathSAT (IC3), Eldarica
- VCGen+QARMC compares favorably to HSF+QARMC

	Small-step (SS_p^s)				Multi-step (MS)				HSF(C)
	QARMC	Z3	MSAT	ELD	QARMC	Z3	MSAT	ELD	
Correct answers	217	208	205	217	210	196	177	182	189
safe problems	161	150	158	158	160	144	147	141	158
unsafe problems	56	58	47	59	50	52	30	41	31
Incorrect answers	5	0	3	2	3	0	1	0	12
false alarms	3	0	1	0	1	0	1	0	3
missed bugs	2	0	2	2	2	0	0	0	9
Timeouts	98	112	112	101	120	124	142	138	119
Total problems	320	320	320	320	320	320	320	320	320
VCG time	221.68	221.68	221.68	221.68	141.85	141.85	141.85	141.85	
Solving time	3656.24	4221.39	2988.86	8809.58	2674.00	2704.95	1896.96	2779.18	
Total time	3877.92	4443.07	3210.54	9031.26	2815.85	2846.80	2038.81	2921.03	
Average Time	17.87	21.36	15.66	41.62	13.41	14.52	11.52	16.05	

3. CHC Specialization as CHC Solving

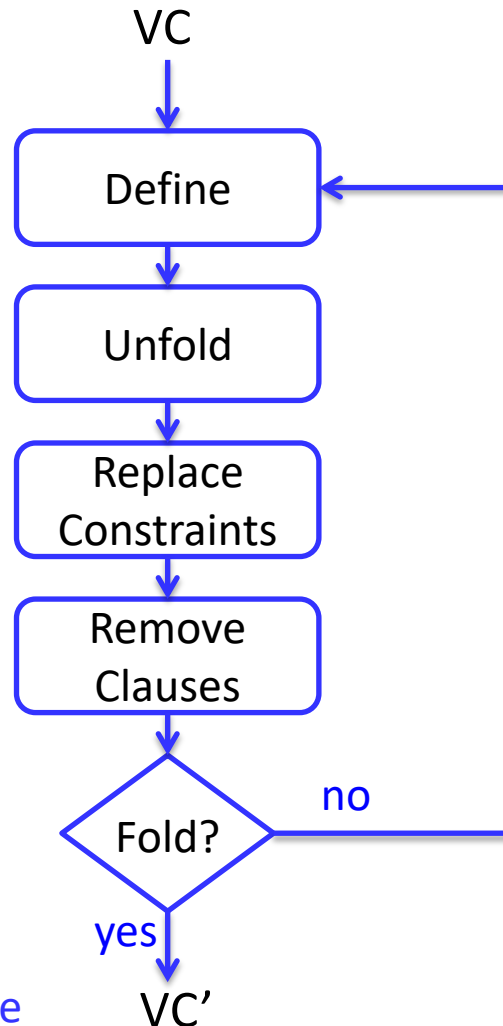
VCTransf: Specializing Verification Conditions

false :- c, p(X)

newp(X) :- c, p(X)

apply theory of constraints

VC is **satisfiable** iff VC' is **satisfiable**



Specializing verification conditions by **propagating constraints**.

Introduction of new predicates by **generalization** (e.g., **widening** and **convex hull** techniques)

VCTransf as CHC Solving

The effect of applying VCTransf can be:

1. A set VC' of verification conditions without constrained facts for the predicates on which the queries depend (i.e., no clauses of the form $p(X) :- c$).
 VC' is satisfiable.
2. A set VC' of verification conditions including $false :- true$.
 VC' is unsatisfiable.
3. Neither 1 nor 2 (constrained facts of the form $p(X) :- c$, but not $false :- true$).
Satisfiability is unknown.

propagation of constraint $X < 0$ and constant b

VC
 $false :- X < 0, p(X, b).$
 $p(X, C) :- X = Y + 1, p(Y, C).$
 $p(X, a).$
 $p(X, b) :- X \geq 0, tm_halts(X).$

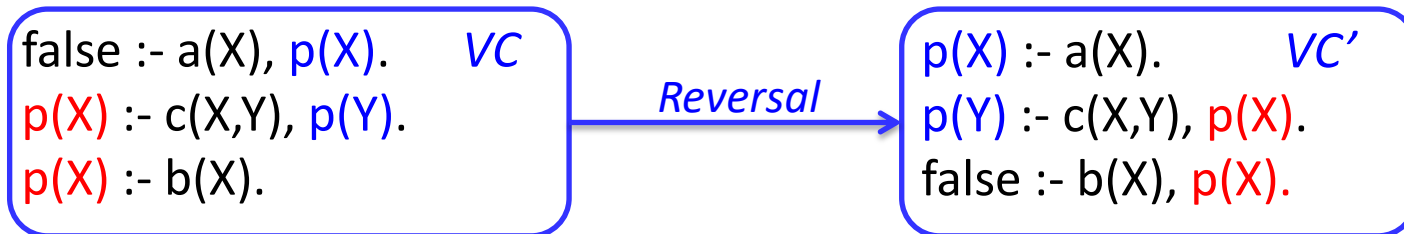
$VCTransf$

VC'
 $false :- X < 0, q(X).$
 $q(X) :- X < 0, X = Y + 1, q(Y).$

No constrained facts: VC' satisfiable

Iterated CHC Specialization

- If the satisfiability of VC' is unknown, $VCTransf$ can be iterated.
- Between two applications of $VCTransf$ we can apply the *Reversal* transformation (particular case of the *query-answer* transformation [KafleGallagher 15] for linear programs) that interchanges premises and conclusions of clauses (backward reasoning from queries simulates forward reasoning from facts).



VC is *satisfiable* iff VC' is *satisfiable*

$$VC_0 \xrightarrow{VCTransf} VC_1 \xrightarrow{Reversal} VC_2 \xrightarrow{VCTransf} VC_3 \dots \xrightarrow{VCTransf} VC_n$$

Iterated CHC Specialization: *SumUpto* Example

false :- $N \geq 1$, $X=0$, $Y=0$, $p(X, Y, N)$. VC_0
 $p(X, Y, N) :- X < N$, $X1=X+1$, $Y1=Y+2$, $p(X1, Y1, N)$.
 $p(X, Y, N) :- X \geq N$, $Y < X$.

Iterated CHC Specialization: *SumUpto* Example

false :- $N \geq 1$, $X=0$, $Y=0$, p(X, Y, N). VC_0
p(X, Y, N) :- $X < N$, $X1=X+1$, $Y1=Y+2$, p(X1, Y1, N).
p(X, Y, N) :- $X \geq N$, $Y < X$.

$VCTransf$

false :- $N \geq 1$, $X1=1$, $Y1=1$, new2(X1, Y1, N). VC_1
new2(X, Y, N) :- $X=1$, $Y=1$, $N > 1$, $X1=2$, $Y1=3$, new3(X1, Y1, N).
new3(X, Y, N) :- $X1 \geq 1$, $Y1 \geq X1$, $X < N$, $X1=X+1$, $Y1=X1+Y$, new3(X1, Y1, N).
new3(X, Y, N) :- $Y \geq 1$, $N \geq 1$, $X \geq N$, $Y < X$.

Iterated CHC Specialization: *SumUpto* Example

false :- N ≥ 1, X=0, Y=0, p(X, Y, N). VC_0
p(X, Y, N) :- X < N, X1=X+1, Y1=Y+2, p(X1, Y1, N).
p(X, Y, N) :- X ≥ N, Y < X.

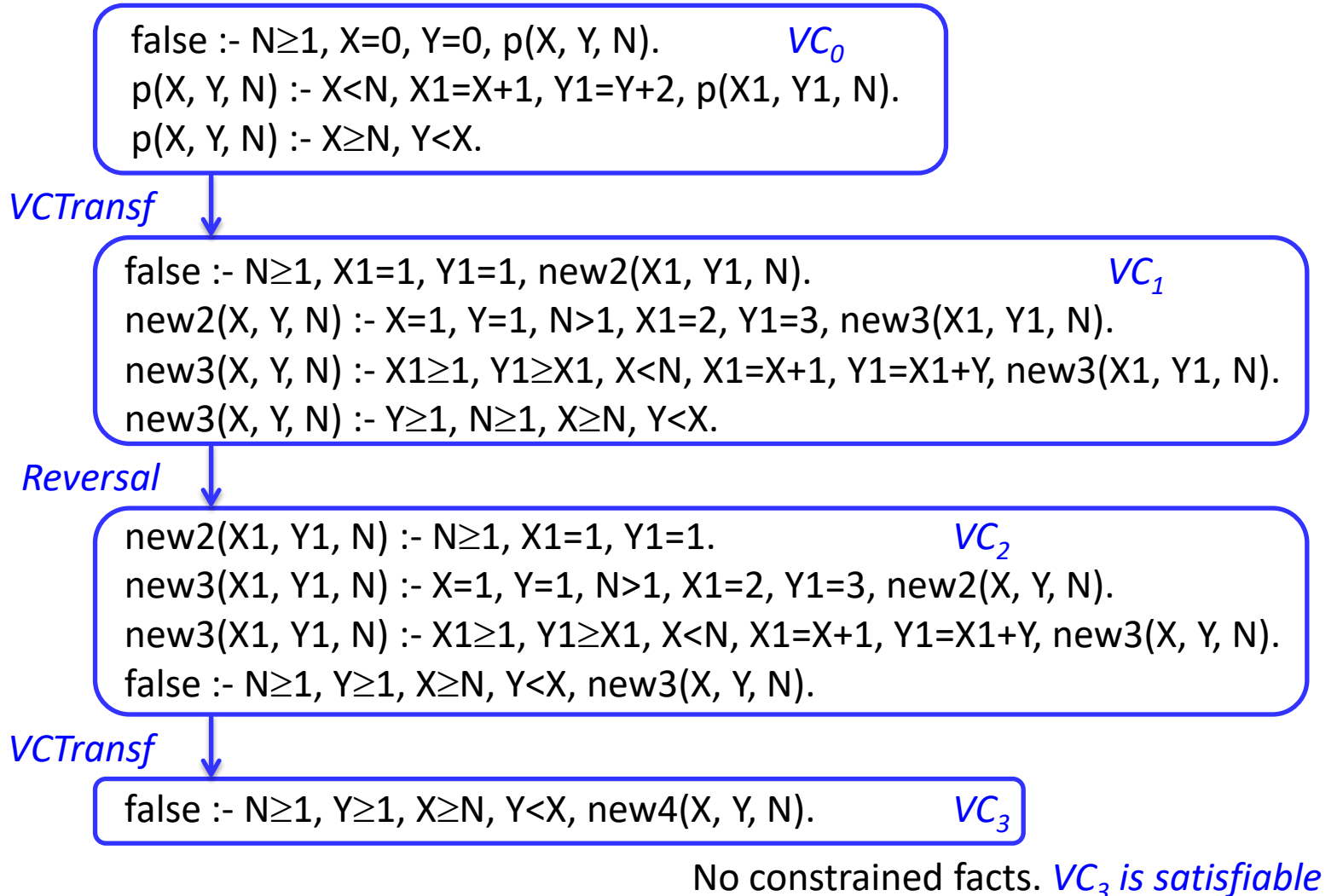
VCTransf

false :- N ≥ 1, X1=1, Y1=1, new2(X1, Y1, N). VC_1
new2(X, Y, N) :- X=1, Y=1, N > 1, X1=2, Y1=3, new3(X1, Y1, N).
new3(X, Y, N) :- X1 ≥ 1, Y1 ≥ X1, X < N, X1=X+1, Y1=X1+Y, new3(X1, Y1, N).
new3(X, Y, N) :- Y ≥ 1, N ≥ 1, X ≥ N, Y < X.

Reversal

new2(X1, Y1, N) :- N ≥ 1, X1=1, Y1=1. VC_2
new3(X1, Y1, N) :- X=1, Y=1, N > 1, X1=2, Y1=3, new2(X, Y, N).
new3(X1, Y1, N) :- X1 ≥ 1, Y1 ≥ X1, X < N, X1=X+1, Y1=X1+Y, new3(X, Y, N).
false :- N ≥ 1, Y ≥ 1, X ≥ N, Y < X, new3(X, Y, N).

Iterated CHC Specialization: *SumUpto* Example



Experiments with VeriMAP

216 examples taken from: DAGGER, TRACER, InvGen, and TACAS 2013 Software Verification Competition.

- ARMC [Podelski, Rybalchenko PADL 2007]
- HSF(C) [Grebenshchikov et al. TACAS 2012]
- TRACER [Jaffar, Murali, Navas, Santosa CAV 2012]

		VeriMAP (Gen_{PH})	ARMC	HSF(C)	TRACER	
					$SPost$	$WPre$
1	<i>correct answers</i>	185	138	159	91	103
2	<i>safe problems</i>	154	112	137	74	85
3	<i>unsafe problems</i>	31	26	22	17	18
4	<i>incorrect answers</i>	0	9	5	13	14
5	<i>false alarms</i>	0	8	3	13	14
6	<i>missed bugs</i>	0	1	2	0	0
7	<i>errors</i>	0	18	0	20	22
8	<i>timed-out problems</i>	31	51	52	92	77
9	<i>total score</i>	339 (0)	210 (-40)	268 (-28)	113 (-52)	132 (-56)
10	<i>total time</i>	10717.34	15788.21	15770.33	27757.46	23259.19
11	<i>average time</i>	57.93	114.41	99.18	305.03	225.82

Table 1: Verification results using VeriMAP, ARMC, HSF(C) and TRACER. For each column the sum of the values of lines 1, 4, 7, and 8 is 216, which is the total number of the verification problems we have considered. The timeout limit is five minutes. Times are in seconds.

4. Improving CHC Solving via Predicate Tupling

Limitations of Linear Arithmetic Specifications

- Not very expressive.
- Example: computing Fibonacci numbers

```
fibonacci: while (n>0) {  
    t=u;  
    u=u+v;  
    v=t;  
    n=n-1  
}
```

$\{n=N, N \geq 0, u=1, v=0, t=0\}$ *fibonacci* $\{fib(N,u)\}$

- The postcondition of *fibonacci* **cannot be specified** by using linear arithmetic constraints *only*.

Recursive Horn Clause Specifications

- Recursive Horn specifications:

$\{z_1=P_1, \dots, z_s=P_s, \mathbf{pre}(P_1, \dots, P_s)\} \text{ prog } \{\mathbf{post}(P_1, \dots, P_s, z)\}$

where :

- z_1, \dots, z_s are global variables of *prog*;
- P_1, \dots, P_s are parameters;
- z is a variable in $\{z_1, \dots, z_s\}$;
- **pre** and **post** are defined by a set of CHCs;
- **post** is a functional relation: $z=F(P_1, \dots, P_s)$ for some function F defined for all P_1, \dots, P_s that satisfy **pre**.

- All computable functions on integers can be specified by sets of CHCs.

Recursive Horn Specification for Fibonacci

Fibonacci specification:

$\{n=N, N \geq 0, u=1, v=0, t=0\}$ *fibonacci* $\{\text{fib}(N,u)\}$

where:

Fib:

$\text{fib}(0,F) :- F=1.$

$\text{fib}(1,F) :- F=1.$

$\text{fib}(N3,F3) :- N1 \geq 0, N2=N1+1, N3=N2+1, F3=F1+F2, \text{fib}(N1,F1), \text{fib}(N2,F2).$

Translating Partial Correctness into CHCs

A recursive Horn specification can be translated into CHCs in two steps:

Step1. Translate the **operational semantics** into CHCs;

Step 2. Generate **verification conditions** as a set of CHCs.

Translating the Operational Semantics

- Define a relation **fibonacci_prog**(N,U) such that, for all integers N, if the program variables satisfy the precondition

$$n=N, N \geq 0, u=1, v=0, t=0$$

then the final value of u computed by program *fibonacci* is U.

- fibonacci_prog** is defined by a set *OpSem* of clauses that encode the **operational semantics**:

fibonacci_prog(N,U) :- initConf(Cf0,N), reach(Cf0,Cfh), finalConf(Cfh,U).

initConf(cf(LC, [(n,N),(u,U),(v,V),(t,T)]), N) :- N ≥ 0, U=1, V=0, T=0, at(0,LC).

finalConf(cf(LC, [(n,N),(u,U),(v,V),(t,T)]), U) :- at(last,LC).

reach(Cf,Cf).

reach(Cf0,Cf2) :- tr(Cf0,Cf1), reach(Cf1,Cf2).

tr(Cf0,Cf1) is the **interpreter** of the imperative language.

Specializing the Operational Semantics

- Apply VCGen and **specialize** *OpSem* w.r.t. the program *fibonacci*.
- *OpSem*_{SP}:
fibonacci_prog(N,F) :- %Initialization
 N>=0, U=1, V=0, T=0,
 r(N,U,V,T, N1,F,V1,T1).
r(N,U,V,T, N2,U2,V2,T2) :- %Loop
 N>=1, N1=N-1, U1=U+V, V1=U, T1=U,
 r(N1,U1,V1,T1, N2,U2,V2,T2).
r(N,U,V,T, N,U,V,T) :- N=<0. %Loop exit
- For every query false :- G,
OpSem U {false :- G} is **satisfiable** iff *OpSem*_{SP} U {false :- G} is **satisfiable**.
- *OpSem*_{SP} is **linear recursive** (at most one predicate in the premise).

Generating Verification Conditions (Nonlinear recursive)

Q: false :- F1≠F2, fibonacci_prog(N,F1), fib(N,F2).

plus clauses for fibonacci_prog ($OpSem_{SP}$) and fib (Fib).

Program *fibonacci* is partially correct iff $OpSem_{SP} \cup Fib \cup \{Q\}$.

Q is **not linear recursive**; the clauses in *Fib* are **not linear recursive**.

Generating Verification Conditions (Almost linear recursive)

- Under suitable assumptions, **linear recursive clauses, except for queries**.
- Transform each clause $\mathbf{post}(P1, \dots, Ps, Z) :- B$ defining the postcondition, into a query for **prog**:
 - (1) **Replace post** by **prog** in the head and in the body
 $\mathbf{prog}(P1, \dots, Ps, Z) :- B'$
 - (2) **Move the conclusion to the premise** (exploiting **functionality** of **prog**):
 $\mathbf{false} :- Y \neq Z, \mathbf{prog}(P1, \dots, Ps, Y), B'$ where Y is a new variable
 - (3) **Case split**:
 $\mathbf{false} :- Y > Z, \mathbf{prog}(X1, \dots, Xs, Z), B'$
 $\mathbf{false} :- Y < Z, \mathbf{prog}(X1, \dots, Xs, Z), B'$
- *If* for all generated queries $\mathbf{false} :- G$, $OpSem_{sp} \cup \{\mathbf{false} :- G\}$ is **satisfiable**, *then* $\{z1=P1, \dots, zs=Ps, \mathbf{pre}(P1, \dots, Ps)\} \mathbf{prog} \{\mathbf{post}(P1, \dots, Ps, z)\}$ is **valid**.

Verification Conditions for Fibonacci

Generating the **verification conditions** for *fibonacci*

`fib(0,F) :- F=1.`

(1) **Replace fib** by **fibonacci_prog** in the head and in the body

`fibonacci_prog(0,F) :- F=1.`

(2) **Move the conclusion to the premise:**

`false :- F≠1, fibonacci_prog(0,F).`

(3) **Case split**

Q1: `false :- F>1, fibonacci_prog(0,F).`

Q2: `false :- F<1, fibonacci_prog(0,F).`

Verification Conditions for Fibonacci

- Verification conditions for *fibonacci*

Q1: false :- $F > 1$, fibonacci_prog(0,F).

Q2: false :- $F < 1$, fibonacci_prog(0,F).

Q3: false :- $F > 1$, fibonacci_prog(1,F).

Q4: false :- $F < 1$, fibonacci_prog(1,F).

Q5: false :- $N1 \geq 0$, $N2 = N1 + 1$, $N3 = N2 + 1$, $F3 > F1 + F2$,
fibonacci_prog(N1,F1),
fibonacci_prog(N2,F2),
fibonacci_prog(N3,F3).

Q6: false :- $N1 \geq 0$, $N2 = N1 + 1$, $N3 = N2 + 1$, $F3 < F1 + F2$,
fibonacci_prog(N1,F1),
fibonacci_prog(N2,F2),
fibonacci_prog(N3,F3).

- Program *fibonacci* is **partially correct** if, for $i=1, \dots, 6$, $OpSem_{SP} \cup \{Q_i\}$ is **satisfiable**.

Satisfiability via LA-Solvability

- Consider constraints \mathbf{C}_{LA} in **Linear (Integer) Arithmetics** (linear equalities and inequalities over the integers). An **LA-solution** of a set S of CHCs is a mapping

$$\Sigma: \text{Atom} \rightarrow \mathbf{C}_{\text{LA}}$$

such that, for every clause $A_0 :- c, A_1, \dots, A_n$ in S ,

$$\mathbf{LA} \models \forall (c, \Sigma(A_1) \wedge \dots \wedge \Sigma(A_n) \rightarrow \Sigma(A_0))$$

- A set of CHCs is **LA-solvable** if it has an LA-solution.
- LA-solvability implies satisfiability, but not vice versa.
- Satisfiability is undecidable and **not semidecidable**.
LA-solvability is **semidecidable**. (\mathbf{C}_{LA} is r.e. and entailment is decidable.)
Complete LA-solving methods exist (e.g., CEGAR).

Limitations of LA-solving

- Program *fibonacci* is **partially correct** and each $OpSem \cup \{Q_i\}$ is satisfiable.
- However, there is **no LA-solution** for $OpSem \cup \{Q_5\}$ (nor for $OpSem \cup \{Q_6\}$).

Proof (see details in ICLP-15 paper): there exists no LA constraint $c(N,F)$ which is an LA-solution of the clauses for **r_fibonacci** and:

$$\mathbf{LA} \models \forall (N_1 \geq 0, N_2 = N_1 + 1, N_3 = N_2 + 1, F_3 > F_1 + F_2, \\ c(N_1, F_1), c(N_2, F_2), c(N_3, F_3) \rightarrow \text{false})$$

- LA-solvers **cannot prove** the partial correctness of *fibonacci*.

Improving LA-solving by Transforming Verification Conditions

- Solution 1: **More powerful constraint theories**, but decidability of entailment is lost for non-linear polynomials [Matijasevic 70].
- Solution 2: **Transform** the verification conditions into **equisatisfiable** CHCs that are (sometimes) **LA-solvable**.
- Transformation: **Linearization via predicate tupling**.

Linearization

Q5: $\text{false} :- N1 \geq 0, N2 = N1 + 1, N3 = N2 + 1, F3 > F1 + F2,$
 $\text{fibonacci_prog}(N1, F1), \text{fibonacci_prog}(N2, F2), \text{fibonacci_prog}(N3, F3).$

- No LA-solution of **single fibonacci_prog** atoms is able to prove that the premise of Q5 is false.
- An “LA-solution” for the **conjunction** of the three **fibonacci_prog** atoms exists. The conjunction of the three atoms implies the LA-constraint:

$$N1 \geq 0, N2 = N1 + 1, N3 = N2 + 1, F3 = F1 + F2$$

which implies satisfiability of Q5.

- Apply **predicate tupling** and transform conjunctions of atoms into single atoms, i.e., transform $OpSem_{SP} \cup \{Q5\}$ into a set of **linear recursive** clauses.

The Linearization Transformation

Nonlinear queries



Nonlinear clauses

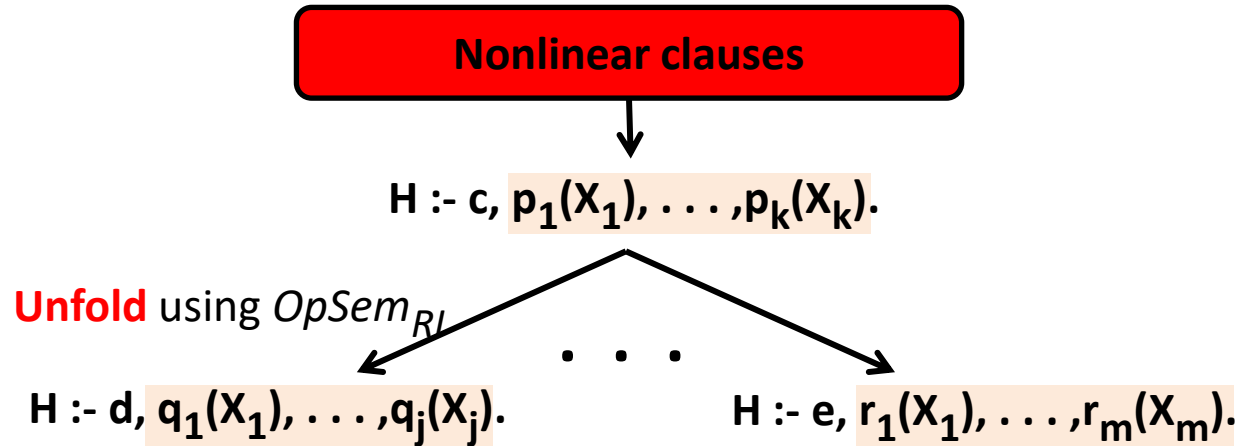
The Linearization Transformation

Nonlinear clauses



$H :- c, p_1(x_1), \dots, p_k(x_k).$

The Linearization Transformation



The Linearization Transformation

Nonlinear clauses

$H :- c, p_1(X_1), \dots, p_k(X_k).$

Unfold using $OpSem_{RI}$

$H :- d, q_1(X_1), \dots, q_j(X_j).$

...

$H :- e, r_1(X_1), \dots, r_m(X_m).$

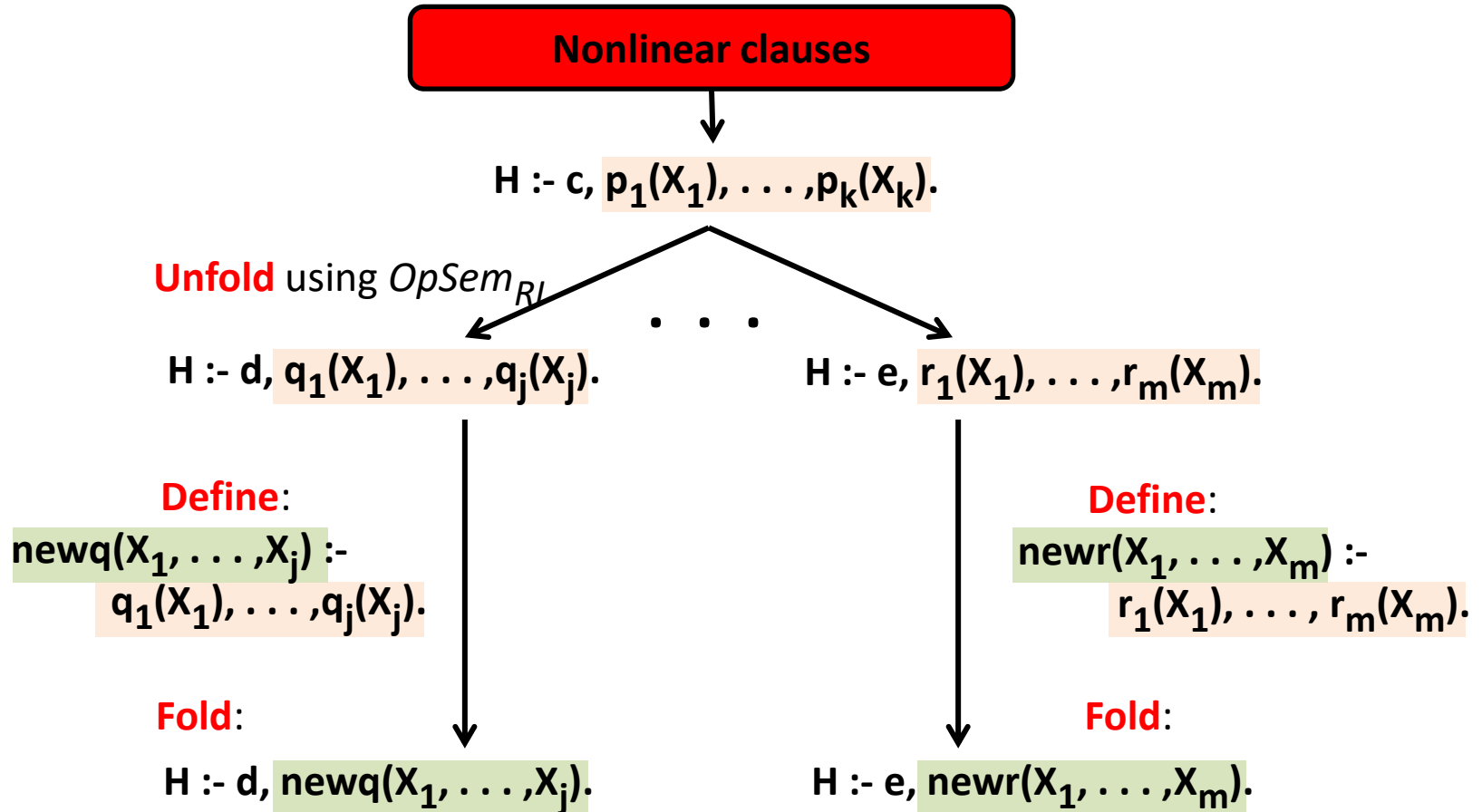
Define:

$newq(X_1, \dots, X_j) :-$
 $q_1(X_1), \dots, q_j(X_j).$

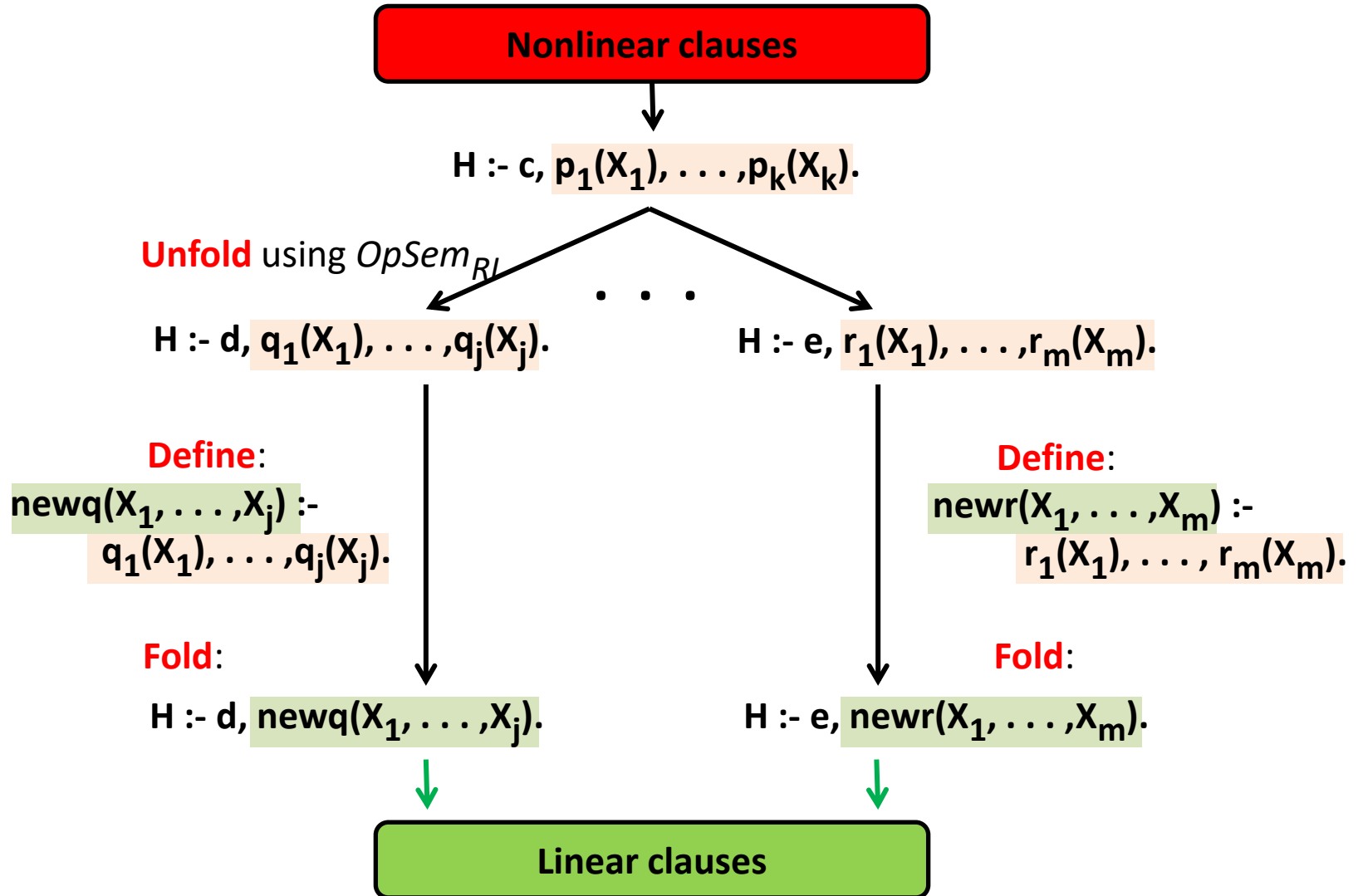
Define:

$newr(X_1, \dots, X_m) :-$
 $r_1(X_1), \dots, r_m(X_m).$

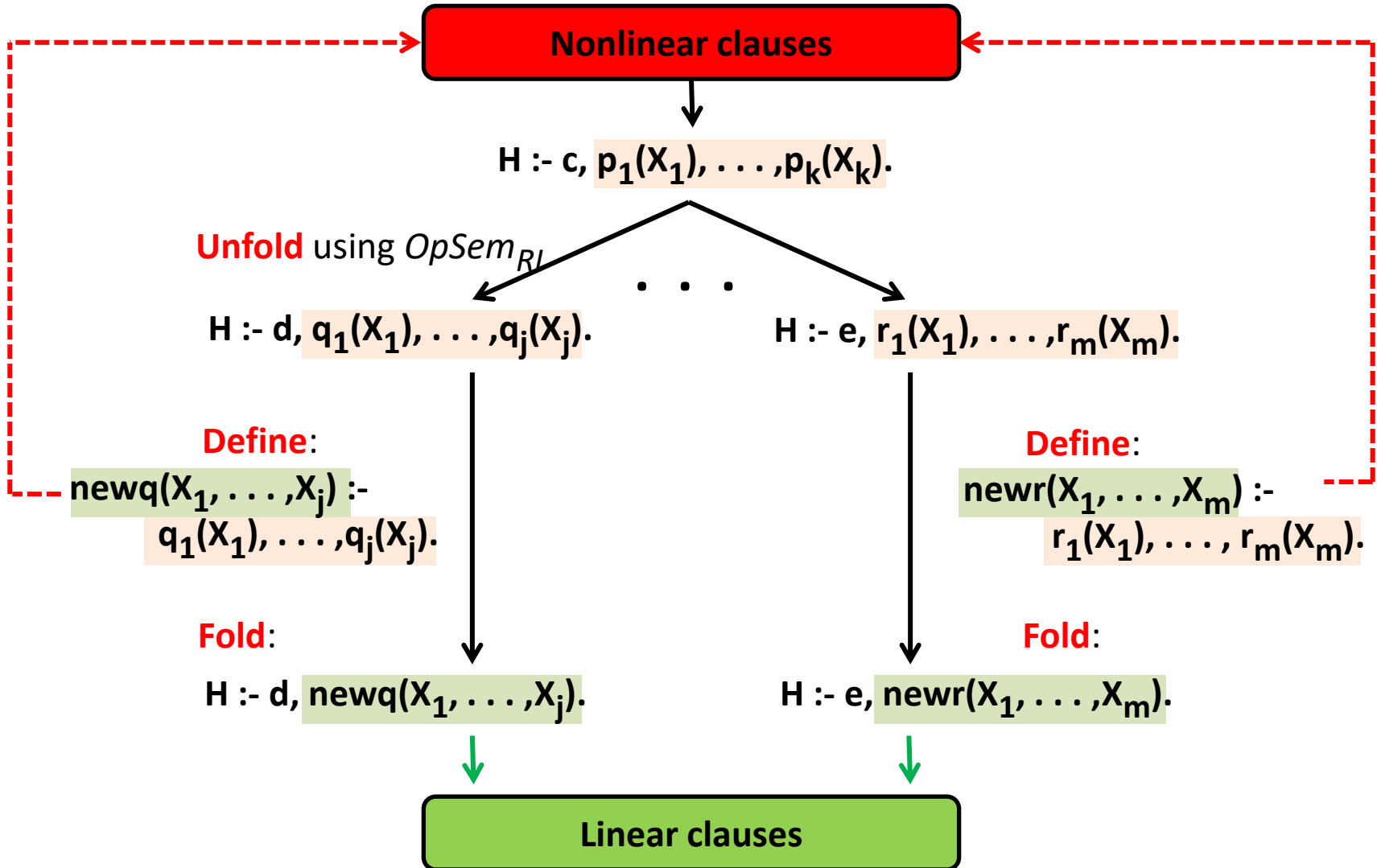
The Linearization Transformation



The Linearization Transformation



The Linearization Transformation



Linearized Fibonacci

false :- $N1 \geq 0, N2 = N1 + 1, N3 = N2 + 1, F3 > F1 + F2, U = 1, V = 0, \mathbf{new1}(N3, U, V, F3, N2, F2, N1, F1).$

new1($N1, U, V, U, N2, U, N3, U$) :- $N1 = < 0, N2 = < 0, N3 = < 0.$

new1($N1, U, V, U, N2, U, N3, F3$) :- $N1 = < 0, N2 = < 0, N4 = N3 - 1, W = U + V, N3 \geq 1, \mathbf{new2}(N4, W, U, F3).$

new1($N1, U, V, U, N2, F2, N3, U$) :- $N1 = < 0, N4 = N2 - 1, W = U + V, N2 \geq 1, N3 = < 0, \mathbf{new2}(N4, W, U, F2).$

new1($N1, U, V, U, N2, F2, N3, F3$) :- $N1 = < 0, N4 = N2 - 1, N2 \geq 1, N5 = N3 - 1, N3 \geq 1, \mathbf{new3}(N4, W, U, F2, N5, F3).$

new1($N1, U, V, F1, N2, U, N3, U$) :- $N4 = N1 - 1, W = U + V, N1 \geq 1, N2 = < 0, N3 = < 0, \mathbf{new2}(N4, W, U, F1).$

new1($N1, U, V, F1, N2, U, N3, F3$) :- $N4 = N1 - 1, N1 \geq 1, N2 = < 0, N5 = N3 - 1, W = U + V, N3 \geq 1, \mathbf{new3}(N4, W, U, F1, N5, F3).$

new1($N1, U, V, F1, N2, F2, N3, U$) :- $N4 = N1 - 1, N1 \geq 1, N5 = N2 - 1, W = U + V, N2 \geq 1, N3 = < 0, \mathbf{new3}(N4, W, U, F1, N5, F2).$

new1($N1, U, V, F1, N2, F2, N3, F3$) :- $N4 = N1 - 1, N1 \geq 1, N5 = N2 - 1, N2 \geq 1, N6 = N3 - 1, W = U + V, N3 \geq 1,$
new1($N4, W, U, F1, N5, F2, N6, F3$).

plus linear clauses for **new2** and **new3**.

new1, new2, new3 have been introduced by the following **definitions**:

new1($N1, U, V, F1, N2, F2, N3, F3$) :- $r(N1, U, V, V, X1, F1, Y1, Z1), r(N2, U, V, V, X2, F2, Y2, Z2), r(N3, U, V, V, X3, F3, Y3, Z3).$

new2(N, U, V, F) :- $r(N, U, V, V, X, F, Y, Z).$

new3($N2, U, V, F2, N1, F1$) :- $r(N1, U, V, V, X1, F1, Y1, Z1), r(N2, U, V, V, X2, F2, Y2, Z2).$

The linearized clauses for *fibonacci* are LA-solvable.

Properties of Linearization

- $OpSem_{SP}$ is a set of **linear recursive** clauses if no recursive functions in the imperative language.
- For every query **false :- G**, linearization **terminates** for the input
 $OpSem_{SP} \cup \{\mathbf{false} \text{ :- } \mathbf{G}\}$
- Let $TransfCls$ be the output of linearization.
 $OpSem_{SP} \cup \{\mathbf{false} \text{ :- } \mathbf{G}\}$ is **satisfiable** iff $TransfCls$ is **satisfiable**.
- If $OpSem_{SP} \cup \{\mathbf{false} \text{ :- } \mathbf{G}\}$ is **LA-solvable**, then $TransfCls$ is **LA-solvable**.
Not vice versa: **LA-solvability can be increased**.

Experiments

<i>Program</i>	VCG	<i>LA-solving-1</i>		LIN	<i>LA-solving-2: VeriMAP &</i>		
		Z3	Eldarica		Z3	MathSAT	Eldarica
1. <i>binary_division</i>	0.02	4.16	<i>TO</i>	0.04	17.36	17.87	20.98
2. <i>fast_multiplication_2</i>	0.02	<i>TO</i>	3.71	0.01	1.07	1.97	7.59
3. <i>fast_multiplication_3</i>	0.03	<i>TO</i>	4.56	0.02	2.59	2.54	9.31
4. <i>fibonacci</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	2.00	47.74	6.97
5. <i>Dijkstra_fusc</i>	0.01	1.02	3.80	0.05	2.14	2.80	10.26
6. <i>greatest_common_divisor</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	0.89	1.78	0.04
7. <i>integer_division</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	0.88	1.90	2.86
8. <i>91-function</i>	0.01	1.27	<i>TO</i>	0.06	117.97	14.24	<i>TO</i>
9. <i>integer_multiplication</i>	0.02	<i>TO</i>	<i>TO</i>	0.01	0.52	14.76	0.54
10. <i>remainder</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	0.87	1.70	3.16
11. <i>sum_first_integers</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	1.79	2.30	6.81
12. <i>lucas</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	2.04	8.39	9.46
13. <i>padovan</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	2.24	<i>TO</i>	11.62
14. <i>perrin</i>	0.01	<i>TO</i>	<i>TO</i>	0.02	2.23	<i>TO</i>	11.89
15. <i>hanoi</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	1.81	2.07	6.59
16. <i>digits10</i>	0.01	<i>TO</i>	<i>TO</i>	0.01	4.52	3.10	6.54
17. <i>digits10-itmd</i>	0.06	<i>TO</i>	<i>TO</i>	0.04	<i>TO</i>	10.26	12.38
18. <i>digits10-opt</i>	0.08	<i>TO</i>	<i>TO</i>	0.10	<i>TO</i>	<i>TO</i>	15.80
19. <i>digits10-opt100</i>	0.01	<i>TO</i>	<i>TO</i>	0.02	<i>TO</i>	58.99	8.98

Conclusions

- CHC transformations are useful for CHC satisfiability
 - For generating verification conditions from the program semantics
 - For proving the satisfiability of CHCs
 - For pre-processing the input of CHC solvers:
CHC solving < transformation + CHC solving
- Future work:
 - Characterization of the power of fold/unfold: How much improvement?
 - Other applications (more languages, properties, etc.)
 - Integration with CHC solvers