

Program Verification using Constraint Handling Rules and Array Constraint Generalizations

Emanuele De Angelis^{1,3}, Fabio Fioravanti¹,
Alberto Pettorossi², and Maurizio Proietti³

¹University of Chieti-Pescara 'G. d'Annunzio', Italy

²University of Rome 'Tor Vergata', Italy

³CNR - Istituto di Analisi dei Sistemi ed Informatica, Rome, Italy

2nd International Workshop on
Verification and Program Transformation, VPT 2014
Vienna, July 17–18, 2014

Proof of Partial Correctness: An Example

Consider a **program** and a **partial correctness triple** for the program:

```
prog: while(x < n) {  
    x = x + 1;  
    y = y + 2;  
}
```

$$\{ x = 0 \wedge y = 0 \wedge n \geq 1 \} \text{ prog } \{ y > x \}$$

(A) Generate the **Verification Conditions** (VC's)

- | | | |
|----|---|----------------|
| 1. | $x = 0 \wedge y = 0 \wedge n \geq 1 \rightarrow P(x, y, n)$ | Initialization |
| 2. | $P(x, y, n) \wedge x < n \rightarrow P(x + 1, y + 2, n)$ | Loop |
| 3. | $P(x, y, n) \wedge x \geq n \rightarrow y > x$ | Exit |

(B) Then, prove **satisfiability** of the VC's.

If the VC's are **satisfiable**, then the **partial correctness triple** holds.

Proof of Satisfiability of VC's

VC's are **satisfiable** if there is an **interpretation** that makes them true.

In our case,

$$P(x,y,n) \equiv (x=0 \wedge y=0 \wedge n \geq 1) \vee y > x$$

makes the VC's true. Indeed,

$$1'. \quad x=0 \wedge y=0 \wedge n \geq 1 \rightarrow (x=0 \wedge y=0 \wedge n \geq 1) \vee y > x$$

$$2'. \quad ((x=0 \wedge y=0 \wedge n \geq 1) \vee y > x) \wedge x < n$$

$$\rightarrow (x+1=0 \wedge y+2=0 \wedge n \geq 1) \vee y+2 > x+1$$

$$3'. \quad ((x=0 \wedge y=0 \wedge n \geq 1) \vee y > x) \wedge x \geq n \rightarrow y > x$$

Thus, $\{x=0 \wedge y=0 \wedge n \geq 1\} \text{ prog } \{y > x\}$ holds.

■ How to **automatically** prove **satisfiability** of the VC's?

Automatic Proofs of Satisfiability of VC's

Various methods:

- CounterExample Guided Abstraction Refinement (CEGAR), Interpolation, Satisfiability Modulo Theories [Rybalchenko et al., McMillan, Alberti et al.]
- Symbolic execution of Constraint Logic Programs [Jaffar et al.]
- Static Analysis and Transformation of Constraint Logic Programs [Gallagher et al., Albert et al.]

Our CLP Transformation Method

(A) Generate the VC's as a **constraint logic program (a CLP program)**:

- V : 1*. $p(X,Y,N) :- X=0, Y=0, N \geq 1.$ (a constrained fact)
2*. $p(X1,Y1,N) :- X < N, X1 = X + 1, Y1 = Y + 2, p(X, Y, N).$
3*. **incorrect** $:- X \geq N, Y \leq X, p(X, Y, N).$

THM: The VC's are **satisfiable** iff **incorrect** \notin **the least model** $M(V).$

(B) Apply transformation rules that **preserve the least model** $M(V).$

- V' : 4. $q(X1, Y1, N) :- X < N, X > Y, Y \geq 0, X1 = X + 1, Y1 = Y + 2, q(X, Y, N).$
5. **incorrect** $:- X \geq N, Y \leq X, Y \geq 0, N \geq 1, q(X, Y, N).$

least model preserved: **incorrect** $\notin M(V)$ iff **incorrect** $\notin M(V')$

no constrained facts for q : **incorrect** $\notin M(V')$

Thus, $\{x=0 \wedge y=0 \wedge n \geq 1\} \text{ prog } \{y > x\}$ holds.

- (A) How to generate the VC's, i.e., V ?
- (B) How to prove satisfiability of the VC's, i.e., transform V into V' ?

Basic ideas from [PEPM-13].

Rules and strategies for programs on integers and integer arrays.

- CLP program transformation:
 - **Unfold/fold rules**: preserving the least model
 - (A) **Strategies for VC's generation**:
specialization of the interpreter (getting V)
 - (B) **Strategies for VC's satisfiability proof**:
propagation of constraints (getting V')
- **Running example**: Ascending Array Initialization, e.g., [3, 4, 5, 6]
- **Experimental evaluation**.

Rules for Transforming CLP Programs

R1. **Definition.** Introducing a new predicate (e.g., a loop invariant)

$$\text{newp}(X) \text{ :- } c, A.$$

R2. **Unfolding.** A symbolic evaluation step (e.g., a resolution step)

given $H \text{ :- } c, \underline{A}, G.$

$$\underline{A} \text{ :- } d_1, G_1, \dots, \underline{A} \text{ :- } d_m, G_m.$$

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

R3. **Folding.** Using a predicate definition

given $H \text{ :- } d, \underline{A}, G.$

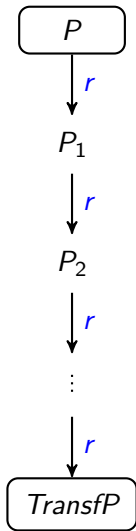
$$\text{newp}(X) \text{ :- } c, \underline{A}. \quad \text{and} \quad d \rightarrow c$$

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

R4. **Clause Removal.** Delete clauses with

(i) unsatisfiable constraint or (ii) subsumed by other clauses

'Rule + Strategies' Program Transformation



- The transformation **rules**:
 $r \in \{ \text{Definition, Unfolding, Folding, Clause Removal} \}$

- The rules **preserve** the least model:

Theorem (Least model preservation)

incorrect $\in M(P)$ iff **incorrect** $\in M(\text{Transf}P)$

- The rules must be guided by **strategies**.

[Burstall-Darlington 77, Tamaki-Sato 84, Etalle-Gabbrielli 96]

Encoding Partial Correctness into CLP

Consider the triple $\{\varphi_{init}\} prog \{\neg\varphi_{error}\}$.

A program $prog$ is **incorrect** w.r.t. φ_{init} and φ_{error}

if a final configuration satisfying φ_{error}

is reachable from an initial configuration satisfying φ_{init} .

Definition (the interpreter Int with the transition predicate $tr(X,Y)$)

$reach(X) :- \text{initConf}(X).$

$reach(Y) :- tr(X,Y), reach(X).$

incorrect $:- \text{errorConf}(X), reach(X).$

+ clauses for tr (i.e., the operat. semantics of the programming language)

Theorem

$prog$ is **incorrect** iff **incorrect** $\in M(Int)$

A program $prog$ is **correct** iff it is not **incorrect**.

tr(X, Y): the operational semantics

<code>L: Id = Expr</code>	<code>tr(cf(cmd(L, asgn(Id, Expr)), S), cf(cmd(L1, C1), S1)) :- aeval(Expr, S, V), <i>evaluate expression</i> update(Id, V, S, S1), <i>update store</i> nextlabel(L, L1), <i>next label</i> at(L1, C1). <i>next command</i></code>
<code>L: if(Expr) { L1: ... } else L2: ... }</code>	<code>tr(cf(cmd(L, ite(Expr, L1, L2)), S), cf(C, S)) :- beval(Expr, S), <i>expression is true</i> at(L1, C). <i>next command</i> <code>tr(cf(cmd(L, ite(Expr, L1, L2)), S), cf(C, S)) :- beval(not(Expr), S), <i>expression is false</i> at(L2, C). <i>next command</i></code></code>
<code>L: goto L1</code>	<code>tr(cf(cmd(L, goto(L1)), S), cf(C, S)) :- at(L1, C). <i>next command</i></code>

tr(X,Y): the operational semantics for array assignment

array assignment: $L : a[ie] = e$

tr(cf(cmd(L,asgn(elem(A,IE),E)),S),	<i>source configuration</i> cf
cf(cmd(L1,C),S1)) :-	<i>target configuration</i> cf
eval(IE,S,I),	<i>evaluate index expr</i> IE
eval(E,S,V),	<i>evaluate expression</i> E
lookup(S,array(A),FA),	<i>get array FA from store</i>
write(FA,I,V,FA1),	<i>update array FA, getting FA1</i>
update(S,array(A),FA1,S1),	<i>update store S, getting S1</i>
nextlab(L,L1),	<i>next label</i> L1
at(L1,C).	<i>command C at next label</i>

Running Example: Ascending Array Initialization

Given the **program** *SeqInit* and the **partial correctness triple**

```
i=1;
while(i < n) {
  a[i] = a[i-1]+1;
  i = i+1;
}
```

$$\{i \geq 0 \wedge n \geq 1 \wedge n = \dim(a)\}$$

SeqInit

$$\{\forall j (0 \leq j \wedge j + 1 < n \rightarrow a[j] < a[j+1])\}$$

CLP encoding of program *SeqInit*

- A set of **at**(label, command) facts.
- while = ite + goto.
- **elem**(a, i) stands for a[i].

at(l_0 , asgn(i, 1)).

at(l_1 , ite(less(i, n), l_2 , l_h)).

at(l_2 , asgn(elem(a, i),
plus(elem(a, minus(i, 1)), 1))).

at(l_3 , asgn(i, plus(i, 1))).

at(l_4 , goto(l_1)).

at(l_h , halt).

CLP encoding of φ_{init} and φ_{error}

initConf(l_0, I, N, A) :-

$I \geq 0, N \geq 1$.

errorConf(l_h, N, A) :-

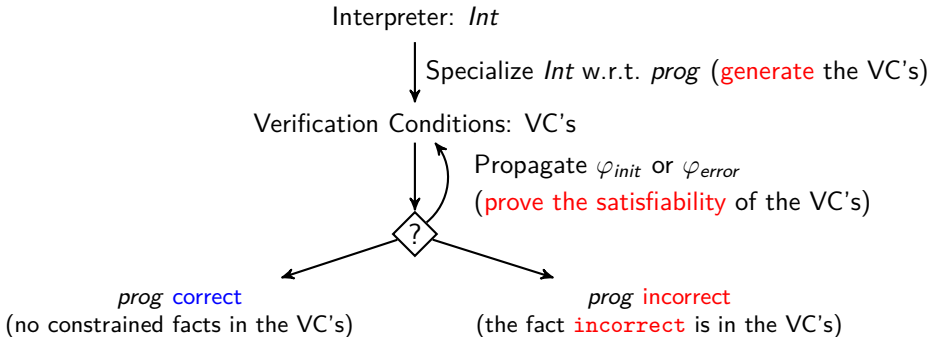
$W \geq 0, W + 1 < N, Z = W + 1, U \geq V,$

read(A, W, U), read(A, Z, V).

The Transformation-based Verification Method

Program Transformation of CLP is used to

- (A) generate the VC's
- (B) prove the satisfiability of the VC's



The Strategy for Generation

Transform(P)

```
TransfP =  $\emptyset$ ;  
Defs = { incorrect :- errorConf( $X$ ), reach( $X$ ) };  
while  $\exists q \in$  Defs do  
    %execute a symbolic evaluation step (i.e., resolution)  
    Cls = Unfolding( $q$ );  
    %remove unsatisfiable and subsumed clauses  
    Cls = ClauseRemoval(Cls);  
    %introduce new predicates (i.e., a loop invariant)  
    Defs = (Defs - { $q$ })  $\cup$  Definition(Cls);  
    %match a predicate definition  
    TransfP = TransfP  $\cup$  Folding(Cls, Defs);  
od
```

Verification Conditions Generation

The specialization of *Int* w.r.t. *prog* removes all references to:

- *tr* and
- *at*

VC's: the Specialized Interpreter for *SeqInt*

```
incorrect :- Z=W+1, W $\geq$ 0, W+1<N, U $\geq$ V, N $\leq$ I,  
            read(A,W,U), read(A,Z,V), new1(I,N,A).  
new1(I1,N,B) :- 1 $\leq$ I, I<N, D=I-1, I1=I+1, V=U+1,  
              read(A,D,U), write(A,I,V,B), new1(I,N,A).  
new1(I,N,A) :- I=1, N $\geq$ 1.
```

- A constrained fact is present:
we cannot conclude that the program is *correct*.
- The fact *incorrect* is not present:
we cannot conclude that the program is *incorrect* either.

The Strategy for Satisfiability

Transform(P)

```
TransfP =  $\emptyset$ ;  
Defs = { incorrect :- errorConf( $X$ ), reach( $X$ ) };  
while  $\exists q \in$  Defs do  
  Cls = Unfolding( $q$ );  
  Cls = ConstraintReplacement(Cls);  
  Cls = ClauseRemoval(Cls);  
  Defs = (Defs - { $q$ })  $\cup$  Definitionarray(Cls);  
  TransfP = TransfP  $\cup$  Folding(Cls, Defs);  
od
```


Constraint Replacement Rule

If $\mathcal{A} \models \forall (c_0 \leftrightarrow (c_1 \vee \dots \vee c_n))$, where \mathcal{A} is the Theory of Arrays

Then replace $H :- c_0, d, G$

by $H :- c_1, d, G, \dots, H :- c_n, d, G$

Constraint Handling Rules for Constraint Replacement:

AC1. Array-Congruence-1: **if $i=j$ then $a[i]=a[j]$**

$\text{read}(A, I, X) \setminus \text{read}(A1, J, Y) \Leftrightarrow A == A1, I = J \mid X = Y.$

AC2. Array-Congruence-2: **if $a[i] \neq a[j]$ then $i \neq j$**

$\text{read}(A, I, X), \text{read}(A1, J, Y) \Rightarrow A == A1, X \langle \rangle Y \mid I \langle \rangle J.$

ROW. Read-Over-Write: **$\{a[i]=x; y=a[j]\}$ if $i=j$ then $x=y$**

$\text{write}(A, I, X, A1) \setminus \text{read}(A2, J, Y) \Leftrightarrow A1 == A2 \mid$

$(I = J, X = Y) ; (I \langle \rangle J, \text{read}(A, J, Y)).$

Ascending Array Initialization

```
new3(A,B,C) :- A=2+H, B-H ≤ 3, E-H ≤ 1, E ≥ 1, B-H ≥ 2, ...,  
  read(N,H,M), read(C,D,F), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW rule:

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ...,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ...,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW, AC1, and AC2 rules:

```
new3(A,B,C) :- A=1+H, E=1+D, J=-1+H, K=1+L, D-H ≤ -2, H < B, ...  
  read(N,E,G), read(N,D,F), read(N,J,L), write(N,H,K,C),  
  reach(J,B,M).
```

Ascending Array Initialization

```
new3(A,B,C) :- A=2+H, B-H ≤ 3, E-H ≤ 1, E ≥ 1, B-H ≥ 2, ...,  
  read(N,H,M), read(C,D,F), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW rule:

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ..., J=E, K=G,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ...,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW, AC1, and AC2 rules:

```
new3(A,B,C) :- A=1+H, E=1+D, J=-1+H, K=1+L, D-H ≤ -2, H < B, ...  
  read(N,E,G), read(N,D,F), read(N,J,L), write(N,H,K,C),  
  reach(J,B,M).
```

Ascending Array Initialization

```
new3(A,B,C) :- A=2+H, B-H ≤ 3, E-H ≤ 1, E ≥ 1, B-H ≥ 2, ...,  
  read(N,H,M), read(C,D,F), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW rule:

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ..., J=E, K=G,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ..., J <> E,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW, AC1, and AC2 rules:

```
new3(A,B,C) :- A=1+H, E=1+D, J=-1+H, K=1+L, D-H ≤ -2, H < B, ...  
  read(N,E,G), read(N,D,F), read(N,J,L), write(N,H,K,C),  
  reach(J,B,M).
```

Ascending Array Initialization

```
new3(A,B,C) :- A=2+H, B-H ≤ 3, E-H ≤ 1, E ≥ 1, B-H ≥ 2, ...,  
  read(N,H,M), read(C,D,F), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW rule:

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ..., J=E, K=G,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

```
new3(A,B,C) :- J=1+D, A=2+D, K=1+I, I < F, ..., J <> E,  
  read(C,D,F), read(N,D,I), write(N,J,K,C), read(C,E,G),  
  reach(J,B,N).
```

- by applying the ROW, AC1, and AC2 rules:

```
new3(A,B,C) :- A=1+H, E=1+D, J=-1+H, K=1+L, D-H ≤ -2, H < B, ...  
  read(N,E,G), read(N,D,F), read(N,J,L), write(N,H,K,C),  
  reach(J,B,M).
```

Definition Introduction

Introduction of suitable new predicate **definitions** (they correspond to **program invariants**).

Difficulty: Introduction of an unbounded number of new predicate definitions.

Solution: Use of **generalization** operators:

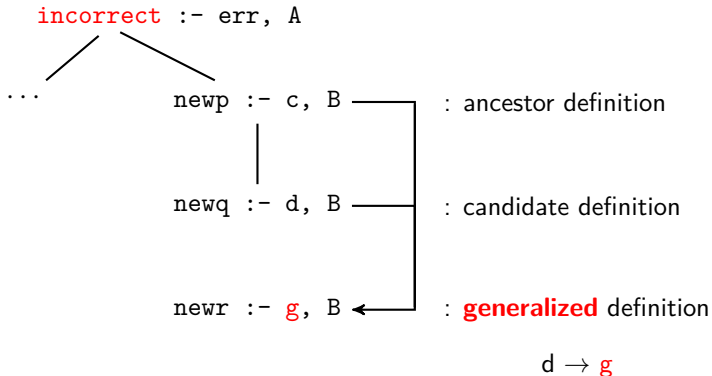
- to ensure the **termination** of the transformation,
- to generate program **invariants**.

Note. They are two somewhat conflicting requirements:

- (**efficiency**) introduction of as few definitions as possible, and
- (**precision**) proof of as many satisfiability properties as possible.

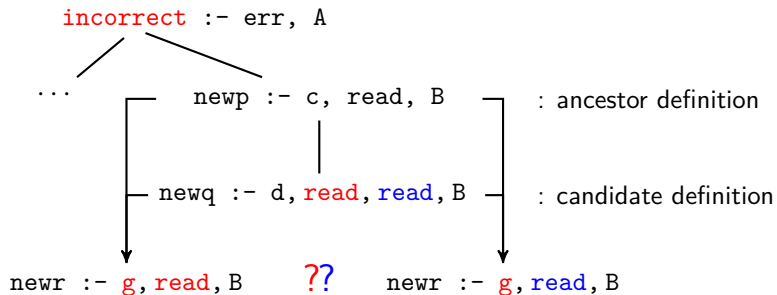
Constraint Generalizations

Definitions are arranged as a tree:



Generalization operators based on **widening** and **convex-hull**.

Array Constraint Generalizations



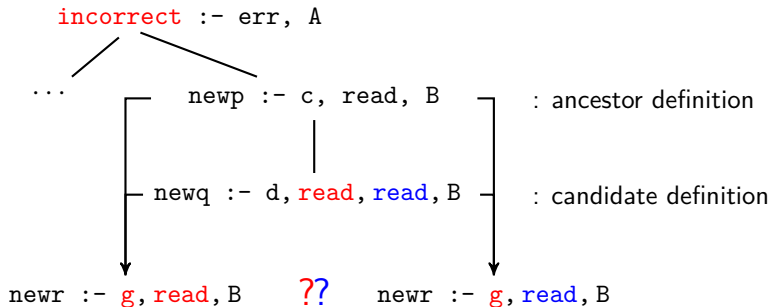
- We decorate CLP variables with the **variable identifiers** of the imperative program.

VC's: the Specialized Interpreter for *SeqInit*

```

incorrect :- Z=W+1, W≥0, W+1<N, U≥V, N≤I,
           read(A,Wj,Ua[j]), read(A,Zj1,Va[j1]), new1(I,N,A).
new1(I1,N,B) :- 1≤I, I<N, D=I-1, I1=I+1, V=U+1,
               read(A,Di,Ua[i]), write(A,I,V,B), new1(I,N,A).
new1(I,N,A) :- I=1, N≥1.
    
```


Array Constraint Generalizations



- We decorate CLP variables with the **variable identifiers** of the imperative program.

VC's: the Specialized Interpreter for *SeqInit*

```
incorrect :- Z=W+1, W ≥ 0, W+1 < N, U ≥ V, N ≤ I,  
           read(A, Wj, Ua[j]), read(A, Zj1, Va[j1]), new1(I, N, A).  
new1(I1, N, B) :- 1 ≤ I, I < N, D=I-1, I1=I+1, V=U+1,  
                 read(A, Di, Ua[i]), write(A, I, V, B), new1(I, N, A).  
new1(I, N, A) :- I=1, N ≥ 1.
```

Ascending Array Initialization

: ancestor definition

```
new3(I,N,A) :- E+1=F, E≥0, I>F, G≥H, N>F, N≤I+1,  
read(A,Ej,Ga[j]), read(A,Fj1,Ha[j1]), reach(I,N,A).
```

: candidate definition

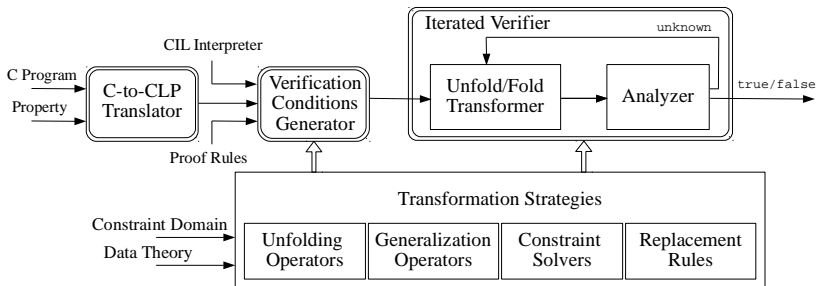
```
new4(I,N,A) :- E+1=F, E≥0, I>F, G≥H, I=1+I1, I1+2≤C, N≤I1+3,  
read(A,Ej,Ga[j]), read(A,Fj1,Ha[j1]), read(A,Pi,Qa[i]),  
reach(I,N,A).
```

: **generalized** definition

```
new5(I,N,A) :- E+1=F, E≥0, I>F, G≥H, N>F,  
read(A,Ej,Ga[j]), read(A,Fj1,Ha[j1]), reach(I,N,A).
```

In the paper: a variable of the form G^v is encoded by `val(v,G)`.

- The VeriMAP tool <http://map.uniroma2.it/VeriMAP>



Experimental evaluation

Program	$Gen_{W,I,\mathbb{M}}$	$Gen_{H,V,\subseteq}$	$Gen_{H,V,\mathbb{M}}$	$Gen_{H,I,\subseteq}$	$Gen_{H,I,\mathbb{M}}$
bubblesort-inner	0.9	<i>unknown</i>	<i>unknown</i>	<i>unknown</i>	1.52
copy-partial	<i>unknown</i>	<i>unknown</i>	3.52	3.51	3.54
copy-reverse	<i>unknown</i>	<i>unknown</i>	5.25	<i>unknown</i>	5.23
copy	<i>unknown</i>	<i>unknown</i>	5.00	4.88	4.90
find-first-non-null	0.14	0.66	0.64	0.28	0.27
find	1.04	6.53	2.35	2.33	2.29
first-not-null	0.11	0.22	0.22	0.22	0.22
init-backward	<i>unknown</i>	1.04	1.04	1.03	1.04
init-non-constant	<i>unknown</i>	2.51	2.51	2.47	2.47
init-partial	<i>unknown</i>	0.9	0.89	0.9	0.89
init-sequence	<i>unknown</i>	4.38	4.33	4.41	4.29
init	<i>unknown</i>	1.00	0.97	0.98	0.98
insertionsort-inner	0.58	2.41	2.4	2.38	2.37
max	<i>unknown</i>	<i>unknown</i>	0.8	0.81	0.82
partition	0.84	1.77	1.78	1.76	1.76
rearrange-in-situ	<i>unknown</i>	<i>unknown</i>	3.06	3.01	3.03
selectionsort-inner	<i>unknown</i>	<i>time-out</i>	<i>unknown</i>	2.84	2.83
precision	6	10	15	15	17
total time	3.61	21.42	34.76	31.81	38.45
average time	0.60	2.14	2.31	2.12	2.26

Conclusions and Future Work

- Parametric verification framework (semantics, logics, constraint domains)
 - CLP as a metalanguage
 - agile way of synthesizing software verifiers [Rybalchenko et al.]
- Semantics preserving transformations
 - iterative verification
 - use Horn clauses for passing information between verifiers [McMillan]
- **Future work**
 - more experiments (including programs with nested loops)
 - more theories (lists, heaps, etc.)
 - Other programming languages and properties.