

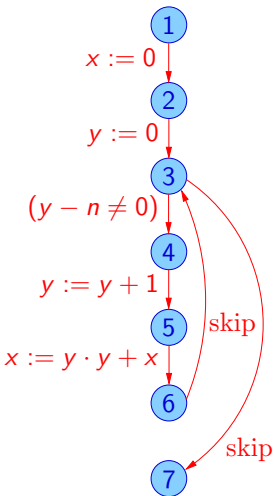
Inferring polynomial invariants with Polyinvar

Helmut Seidl and Michael Petter

TU-München

Chair Workshop, 2005

Problem:



Question

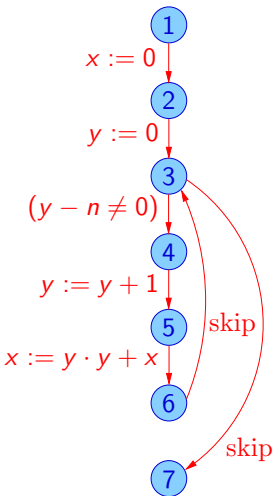
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What relation holds at program point 7?

\Rightarrow *Polynomial invariants*

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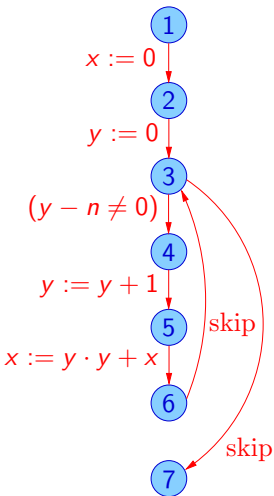
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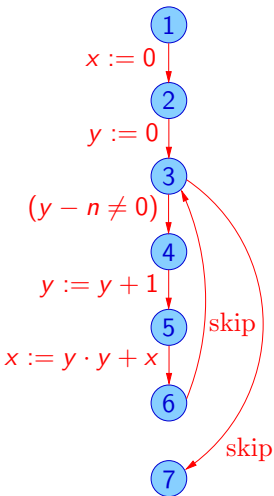
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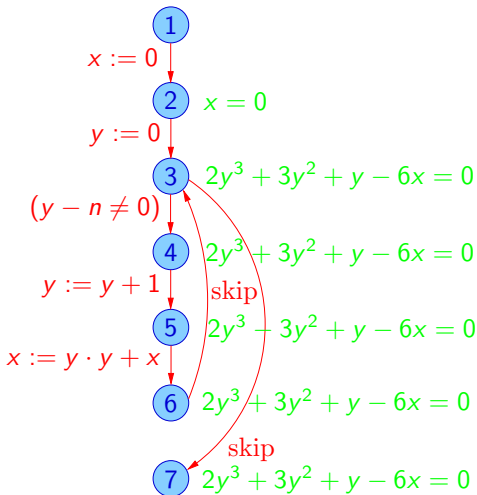
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Valid invariants:



Power sum

The example program calculates the square power sum $x = \sum_{y=0}^n y^2$, therefore

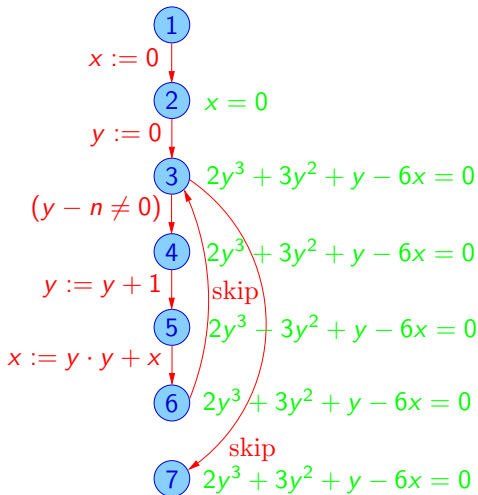
$$x = \frac{2y^3 + 3y^2 + y}{6}$$

holds at program point 7

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\Rightarrow but how to automate this cognition?

Valid invariants:



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Related work

Approaches with ideals

E.RC, D.K. Program Verification Using Automatic Generation Of Invariants 2004

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Approach with modules

M.P. Berechnung von polynomiellen Invarianten 2004

Initial point

Interpret program states as Ideals of polynomials;
Store generators of the ideal as representation

→ M.MO., H.S.

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Abstract Model

Polynomial programs...

- modelling control flow with (possibly annotated) edges
- assignments of multivariate polynomial expressions (without division) $x := y \cdot y + x$
- method calls $x := f(y, z)$
- unknown assignments $x := ?$

... with guards

- negative polynomial equality guards $(y - n) \neq 0$
- positive polynomial equality guards $(y - n) = 0$
- non deterministic choice for the rest *skip*

→ **Goal:** inferring all valid polynomial relations

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Intraprocedural example

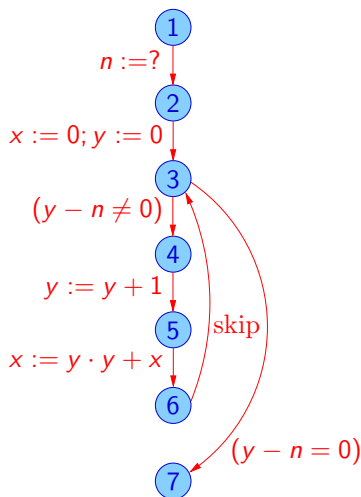
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State abstraction

Still, we have to find an abstraction for program states that serves our analysis...



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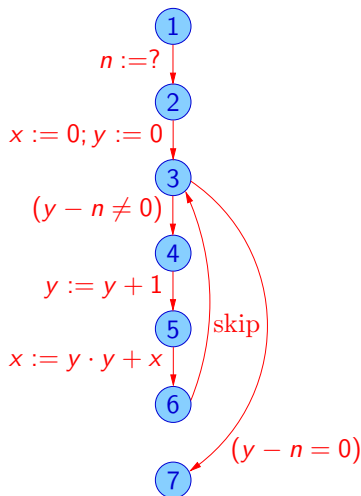
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State abstraction

Polynomials

Polynomials are expressed by equations from the set $\mathbb{R}[\mathbf{X}]$, polynomials over \mathbb{R} and the variables from \mathbf{X} , for example $x - y^2 + 25 = 0$.

Polynomial relations

$$\textcircled{1} \quad \forall PR[s] \subseteq \mathbb{R}[\mathbf{X}] \quad \exists p \in PR[s] \quad \forall c \in \mathbb{R} \cup \mathbf{X} \Rightarrow c \cdot p \in PR[s]$$

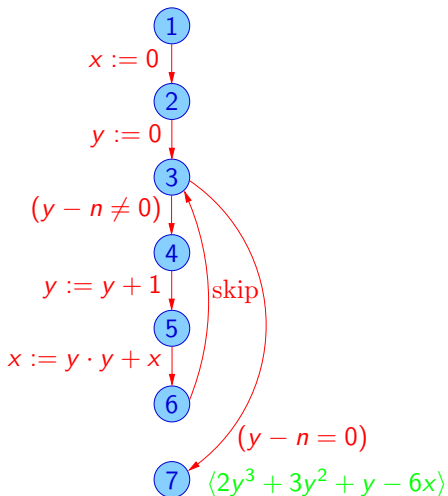
$$\textcircled{2} \quad \forall PR[s] \subseteq \mathbb{R}[\mathbf{X}] \quad \exists p \in PR[s] \quad \forall q \in PR[s], \circ \in \{+, -, \cdot\} \Rightarrow q \circ p \in PR[s]$$

Polynomial ideals – finitely generated

Polynomial ideals are infinite sets of polynomials, with the upper properties. All ideals can be represented by a minimal number of generating polynomials.

For example $\langle \{x - y^2 + 25, x^2 - z\} \rangle$

Verifying polynomial relations



Fixpoint analysis

Associating program states with polynomial ideals.

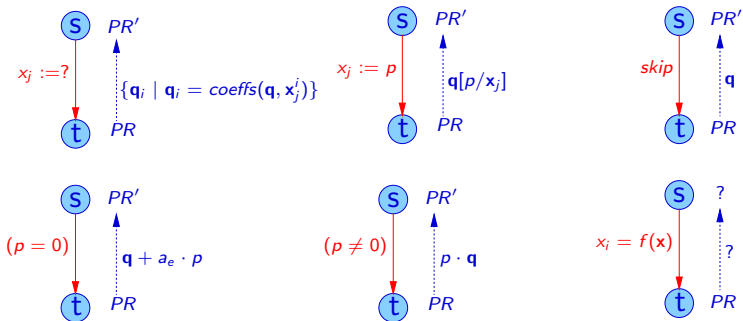
Verifying polynomials

Computing the weakest precondition for a polynomial invariant ideal

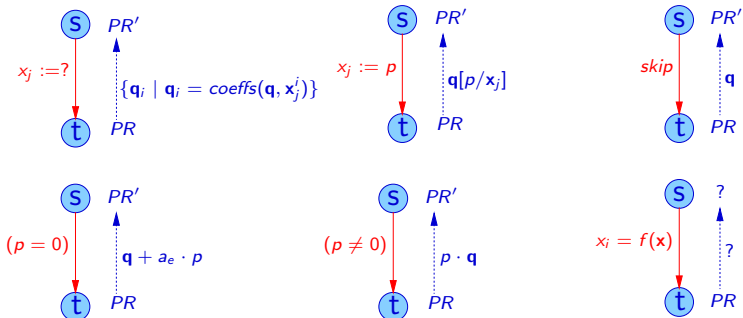
Weakest precondition

The only valid precondition can only be the relation $0 = 0$.

Incremental fixpoint iteration: semantics



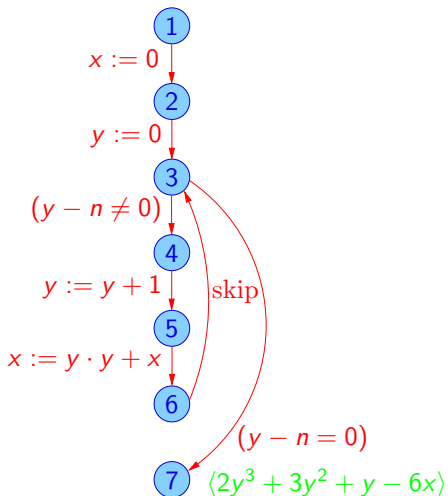
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Propagate only new generators – incremental iteration

Recalculation of ideals at each iteration step is expensive
 \Rightarrow Only new generators q have to be propagated via edges.

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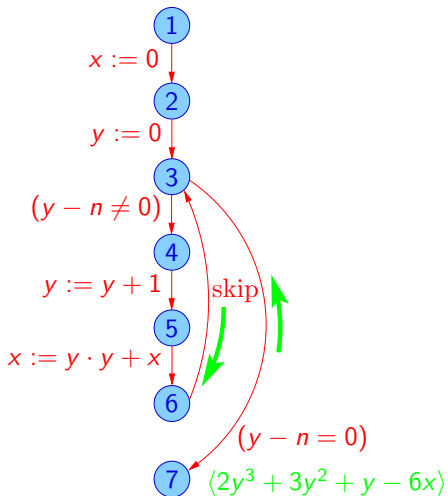
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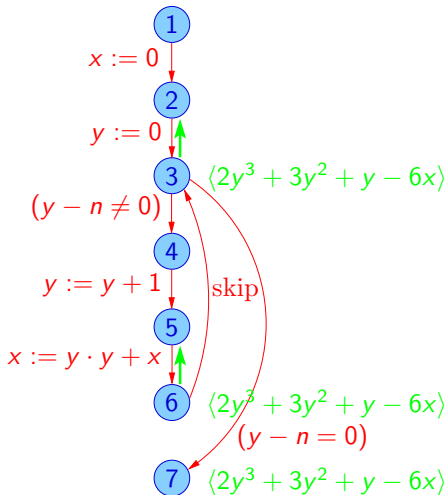
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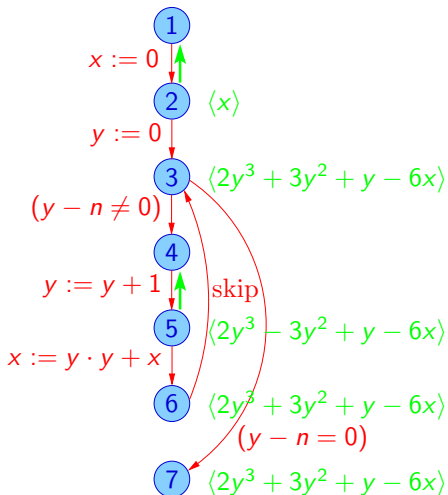
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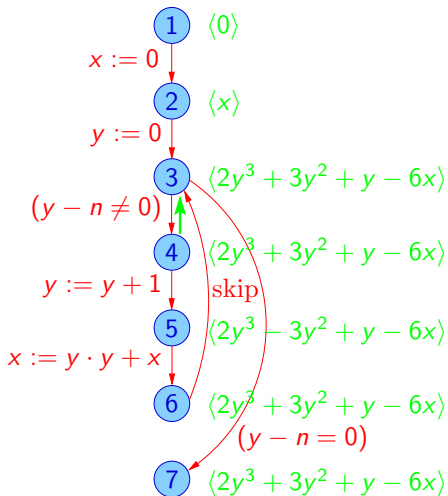
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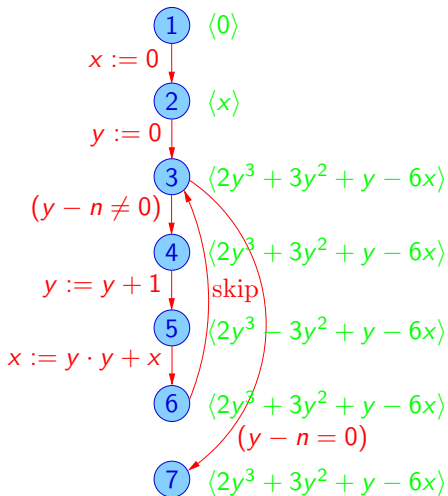
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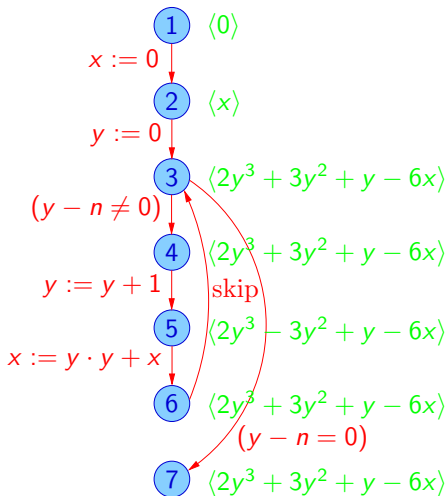
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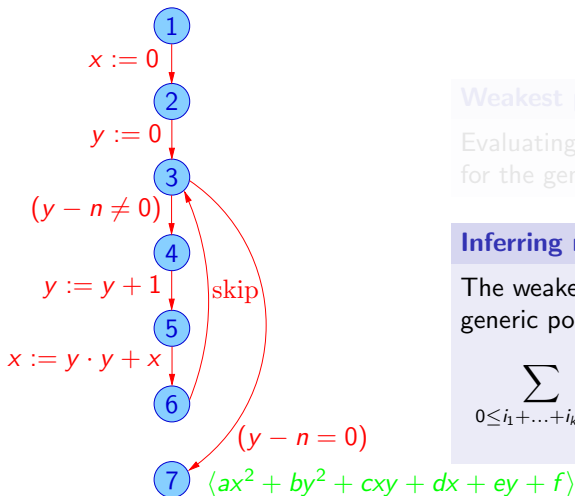
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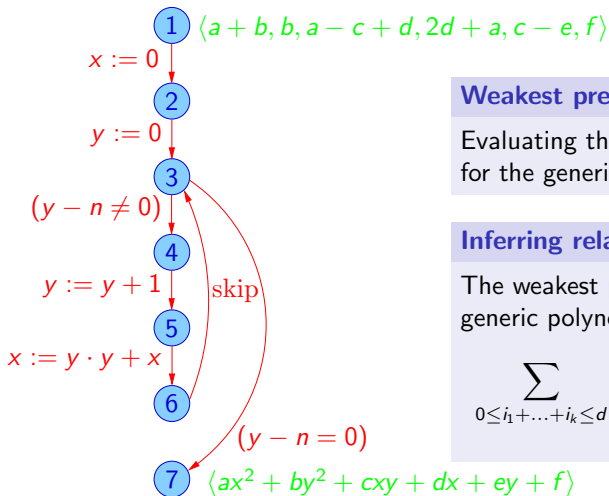
Evaluating the WP provides values for the generic parameters

Inferring relations

The weakest precondition for a generic polynomial of degree n . E.g:

$$\sum_{0 \leq i_1 + \dots + i_k \leq d} \mathbf{a}_{i_1, \dots, i_k} \cdot \mathbf{x}_1^{i_1} \cdot \dots \cdot \mathbf{x}_k^{i_k}$$

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Performance issues

Bad news

- Reductions on polynomial ideals are perform doubly exponentially on the number of participating variables.
- Ideal membership is in general EXPSPACE-hard
- Ideal membership is NP-hard for fixed number of variables

▽ Problem

Using generic polynomials with many variables turns polynomial reductions infeasible.

⇒ **Observation:** Generic variables don't occur in programs, merely model the structure of invariants; they also contribute linear to the polynomials

▽ Idea

Mark generic variables for special treatment in the reduction algorithm.

⇒ Model of vectors and Modules

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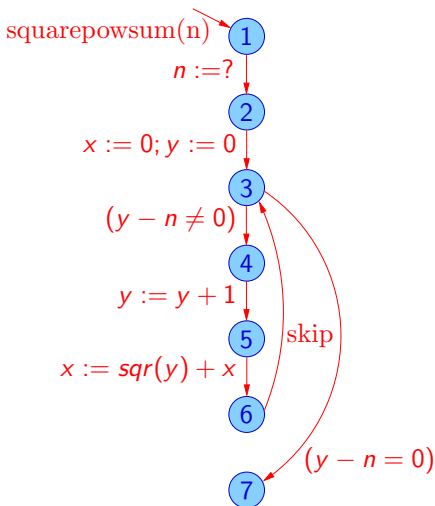
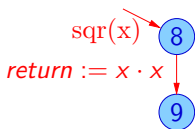
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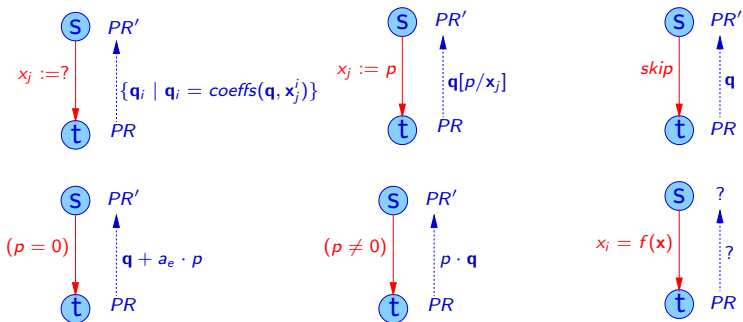
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     $x \leftarrow \text{sqr}(y) + x$ ;
  }
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sqr ( $x \in \mathbb{N}$ )  $\in \mathbb{N}$  {
  return  $x \cdot x$ ;
}

```



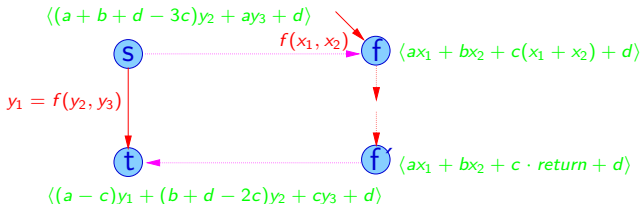
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Method call details

Idea

Use precomputed templates to carry the effect of each method call.

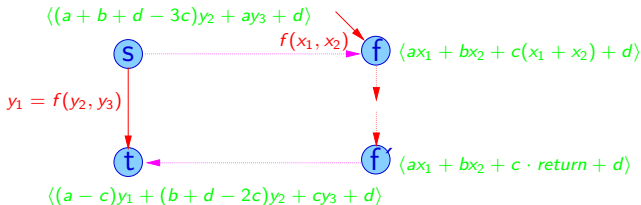


→ **But:** Has yet to be implemented

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Complete analysis

```

Set fixpointiteration (Node  $u_t$ , Vector  $v_t$ , Set Vars, Set Edges, Set Nodes) {
  Set []  $G \leftarrow$  new Set[|Nodes|];
  forall ( $u \in$  Nodes)  $G[u] \leftarrow \emptyset$ ;
  Set  $W \leftarrow \{(v_t, u_t)\}$ ;
  while ( $W \neq \emptyset$ ) {
    ( $v, t$ )  $\leftarrow$  extract( $W$ );
     $v \leftarrow$  reduce( $v, G[t]$ );
    if ( $v \neq 0$ ) {
       $G[t] \leftarrow G[t] \cup \{v\}$ ;
      forall ( $(s, \text{"skip"}, t) \in$  Edges)
         $W \leftarrow W \cup \{(v, s)\}$ ;
      forall ( $(s, \text{"x}_j := p''$ ,  $t) \in$  Edges)
         $W \leftarrow W \cup \{(v[p/x_j], s)\}$ ;
      forall ( $(s, \text{"(p} \neq 0)$ ,  $t) \in$  Edges)
         $W \leftarrow W \cup \{(p \cdot v, s)\}$ ;
      forall ( $(s, \text{"x}_j := ?"$ ,  $t) \in$  Edges)
        let  $l = \max(\{i \mid ax^i \in \text{monoms}(v)\})$ 
        in let  $v \Rightarrow (p_{0_0} x_{j_0}^0 + \dots + p_{0_j} x_{j_j}^l, \dots, p_{k_0} x_{j_0}^0 + \dots + p_{k_j} x_{j_j}^l)$ 
        in let  $v_i \leftarrow (p_{0_i}, p_{1_i}, \dots, p_{k_i})$ 
        in  $W \leftarrow W \cup \{(v_0, u), \dots, (v_l, u)\}$ ;
    }
  }
  return  $\langle G[u_{start}] \rangle$ ;
}

```

Benchmarks

Name	Calculation		ass-deg	Invariant	Time
geoSeries1	$x = (z - 1) \cdot \sum_{k=0}^K z^k$	$y = z^K$	≤ 2	$x = y - 1$	0, 356s
geoSeries2	$x = \sum_{k=0}^K z^k$	$y = z^{K-1}$	≤ 2	$x \cdot (z - 1) = yz - 1$	0, 569s
geoSeries3	$x = \sum_{k=0}^K a \cdot z^k$	$y = z^{K-1}$	≤ 2	$x \cdot (z - 1) = azy - a$	1, 47s

Name	Calculation		ass-deg	Invariant	Time
powSum1	$x = \sum_{k=0}^K 1$	$y = \sum_{k=0}^K 1$	≤ 1	$x = y$	0, 331s
powSum2	$x = \sum_{k=0}^K k$	$y = \sum_{k=0}^K 1$	≤ 1	$2x = y^2 + y$	0, 776s
powSum3	$x = \sum_{k=0}^K k^2$	$y = \sum_{k=0}^K 1$	≤ 2	$6x = 2y^3 + 3y^2 + y$	1, 47s
powSum4	$x = \sum_{k=0}^K k^3$	$y = \sum_{k=0}^K 1$	≤ 3	$4x = y^4 + 2y^3 + y^2$	2, 71s
powSum5	$x = \sum_{k=0}^K k^4$	$y = \sum_{k=0}^K 1$	≤ 4	$30x = 6y^5 + 15y^4 + 10y^3 - y$	10, 3s
powSum6	$x = \sum_{k=0}^K k^5$	$y = \sum_{k=0}^K 1$	≤ 5	$12x = 2y^6 + 6y^5 + 5y^4 - y^2$	787, 2s

Strategy	gs3/5	gs3/6	ps3/5	ps4/5	ps4/6	ps5/5	ps5/6	ps6/6
Original vector	8,4s	29,4s	3,83s	14,7s	...			
Reduced vector	7,3s	26,6s	2,9s	3,5s	8,1s	10,9s	30,0s	787s

Future Work

Implementation

- Treatment of procedure calls
- Scope on relevant variables
- Face large/real examples

Theory

Find a better upper complexity bound

⇒ <http://www2.cs.tum.edu/~petter/polyinvar>



Thank You for Your attention!