

References

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6 Proof of Lemma 1

Proof. Let us prove first that inequalities (5) are sufficient conditions for inclusion of X into Y .

Based on Lemma 1 of [2], let us define $\gamma_{lin}(Y_+)$ as

$$\gamma_{lin}(Y_+) = \bigcap_{1 \leq i \leq k} \left\{ x \in \mathbb{R}^p \mid |u_i^T x| \leq \|Y_+ u_i\|_1 \right\}$$

where each u_i is normal to the faces of $\gamma(Y)$ (or equivalently of $\gamma_{lin}(Y_+)$). Let x be any point such that $x \in \gamma(X)$. Let $x = x' + c_x$ such that $x' \in \gamma_{lin}(X_+)$ and x'' be any point such that $x'' = x' + (c_x - c_y)$. Let us assume that

$$\left| \langle u_i, c_x - c_y \rangle \right| \leq \|Y_+ u_i\|_1 - \|X_+ u_i\|_1, \forall i = 1, \dots, k$$

Under this assumption and also by Lemma 2 of [2] i.e., $\sup_{x' \in \gamma_{lin}(X_+)} \langle u, x' \rangle = \|X_+ u\|_1$, we can say that

$$\left| \langle u_i, c_x - c_y \rangle \right| + \langle u_i, x' \rangle \leq \|Y_+ u_i\|_1$$

This implies $\langle u_i, x'' \rangle \leq \|Y_+ u_i\|_1$ which means $x'' \in \gamma_{lin}(Y_+)$. Thus, $x \in \gamma(Y)$ where the difference between $\gamma_{lin}(Y_+)$ and $\gamma(Y)$ is the translation c_y .

Let us prove now that inequalities (5) are necessary conditions for inclusion of X into Y .

By Lemma 4 of [2], we know that if $\gamma(X) \subseteq \gamma(Y)$ then $\forall u$, $\left| \langle u_i, c_x - c_y \rangle \right| \leq \|Y_+ u\|_1 - \|X_+ u\|_1$. Thus, we can say that if $\gamma(X) \subseteq \gamma(Y)$ then $\forall i = 1, \dots, k$ $\left| \langle u_i, c_x - c_y \rangle \right| \leq \|Y_+ u_i\|_1 - \|X_+ u_i\|_1$

□

7 Example for computing u

Example 2. Consider a zonotope defined by $n=4$ generators in 3-dimension ($p=3$): $((2, -4, 2), (-1, 2, -4), (0, 0, 1), (0, 1, 0))$. We want to compute the normals u_i such that each face (dimension $p-1$) of the zonotope has a vector in u that is normal to it (Lemma 1). Consider one of the faces characterized by the set $((2, -4, 2), (-1, 2, -4))$. Now, we compute the singular value decomposition of the matrix formed from these vectors and we obtain $U\Sigma V^T =$

$$\begin{pmatrix} -0.3374 & 0.2935 & -0.8944 \\ 0.6748 & -0.5871 & -0.4472 \\ -0.6564 & -0.7545 & 0.0000 \end{pmatrix} \begin{pmatrix} 6.3689 & 0 \\ 0 & 2.1066 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} -0.7359 & 0.6771 \\ 0.6771 & 0.7359 \end{pmatrix} \quad (12)$$

The vector normal to the face in question is $(-0.8944, -0.4472, 0)$. In this manner, we compute the remaining normal vectors.

8 Example with the proposed meet operation

Example 3. We set \mathfrak{Z}_1 to be the initial box $(x, y) \in S_0 = [-2, 2]^2$. It can be abstracted using zonotopes as $\mathfrak{Z}_1 = (2\varepsilon_1 \ 2\varepsilon_2)^T$. Consider now \mathfrak{Z}_2 to be the effect of the body of a loop on \mathfrak{Z}_1 :

$$F^\sharp(S_0) = \begin{pmatrix} 1.4\varepsilon_1 & 1.4\varepsilon_2 \\ 1.4\varepsilon_1 & -1.4\varepsilon_2 \end{pmatrix}.$$

The intersection $S_0 \cap F^\sharp(S_0)$ is over-approximated by

$$\alpha \begin{pmatrix} 2\varepsilon_1 \\ 2\varepsilon_2 \end{pmatrix} + (1 - \alpha) \begin{pmatrix} 1.4 \times (\varepsilon_3 - \varepsilon_4) \\ 1.4 \times (\varepsilon_3 + \varepsilon_4) \end{pmatrix}.$$

We then choose α so as to minimize the distance $\|A_+ u_i\|_1$ (this can be solved by linear programming), for u_i the normals to the faces of zonotopes \mathfrak{Z}_1 and \mathfrak{Z}_2 . Here, we obtain α equal to 0.5.

9 Inductive invariants found on Sine programs

Fig. 7(a), 7(b), 7(c) and 7(d) compare the result of the algorithm on the *Sine* program using intervals, octagons, polyhedras and zonotopes.

10 Programs

The programs are illustrated as shown in the Fig. 8, a structure compliant with respect to the analyzer. In Figure 8, *init* stands for the initial state of the program, which the final inductive invariant must contain. If any abstract element does not intersect with the *init*, then it is not a *necessary* abstract element. *Body* denotes the loop of the program and the goal is the candidate invariant.

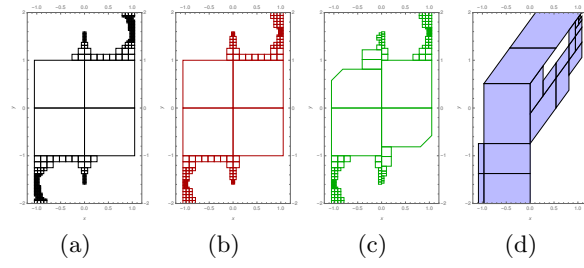


Fig. 7. Inductive invariant for *Sine* example (a) 240 boxes 1448 iterations, 0.4395 s; (b) 154 octagons, 348 iterations, 0.1102 s; (c) 136 polyhedras, 286 iterations, 1.1145 s; (d) 21 zonotopes, 33 iterations, 0.0547 s.

```

init
{

}
body
{

}
goal
{

}

```

Fig. 8. Structure of a program for the analyzer

Octagon

```

init {
  x=[-1,1];
  y=[-1,1];
}

body {
  t=0.7*(x+y);
  y=0.7*(x-y);
  x=t;
}

goal {
  x=[-2,2];
  y=[-2,2];
}

```

Filter

```
init
{
  x=[-0.1,0.1];
  y=[-0.1,0.1];
}
body
{
  r = 1.5*x - 0.7*y + [-0.1,0.1];
  y = x;
  x = r;
}
goal
{
  x=[-4,4];
  y=[-4,4];
}
```

Filter2

```

init {
  x=[0,1];
  y=[0,1];
}

body {
  x = (3.0/4.0) * x - (1.0/8.0)* y;
  y = x;
}

goal {
  // This goal was not taken from the paper itself,
  // but was chosen as a reasonable property to prove.
  x=[-0.2,1];
  y=[-0.2,1];
}
/*
Example file: from LMCS '12: ACCURATE NUMERICAL INVARIANTS
              by Adje, Gaubert, and Goubault
              Figure 11: The program Filter
*/

/*
// Code from the paper:
x = [0,1];
y = [0,1];
while ( true ) {
  x = (3/4)* x-(1/8)*y;
  y = x;
}
*/

```

Arrow-Hurwicz

```

/* Example file: from ESOP'10: Coupling Policy Iteration with
   Semi-definite Relaxation to Compute
   Accurate Numerical Invariants in Static
   Analysis by Adje, Gaubert, and Goubault
   ****
NOTE: This is a *modified* version of the Arrow-Hurwicz example
   because it does not represent u and v as separate
   variables, and it has no loop condition.
   ****
*/
/*
// Original code from
// http://www.lix.polytechnique.fr/~adje/uploads/Codes.pdf
a = 1;b = -1;c = -1;r = 1/2;
u = [0, 1];v = [0, 1];x = [0, 3/2];y = [3/8, 11/8];
while (max(|x-u|, |y-v|) > 1e-9) {
    u = x;
    v = y;
    x = u - r * (a * u + b * v);
    y = v + (r / 2) * (b * u - c);
    if (y <= 0) {
        y = 0;
    }
}
*/
init {
    x = [0,1.5];
    y = [0.375,1.375];
}
body {
    // Note: We have removed the loop condition,
    // so u and v are temporary variables,
    // not carried from one iteration to the next.
    // We have also substituted in constants a,b,c,r
    u = x;
    v = y;
    x = u - 0.5 * (1 * u + (-1) * v);
    y = v + (0.5 / 2) * ((-1) * u - (-1));
    if (y <= 0) {
        y = 0;
    }
}
goal {
    // Note: these bounds are not from the paper,
    // they are simply a reasonable guess.
    x = [-1.73,1.73];
    y = [-1.73,1.73];
}

```

Harm

```

init {
  x0=[0,1];
  x1=[0,1];
}

body {
  x0p = x0; x1p = x1;
  x0 = 0.95*x0p + 0.09975*x1p;
  x1 = -0.1 *x0p + 0.95 *x1p;
}

goal {
  // Bounds given in the Roux paper (Table 3)
  x0=[-1.27,1.27];
  x1=[-1.27,1.27];
  // Goal from ESOP'10 paper (policy iteration goal)
  //x=[-1.42,1.42];
  //v=[-1.42,1.42];
}

/*
Example file:
  Practical policy iterations
  Roux and Garoche
  Form Methods Syst Des 2015
  DOI 10.1007/s10703-015-0230-7
  Also known as Ex.8 (Harmonic oscillator)
  ultimately from LMCS '12: ACCURATE NUMERICAL INVARIANTS
  by Adje, Gaubert, and Goubault
  Figure 12: An implementation of the Symplectic method
*/
/*
// Code from benchmarks tarball
node top(ix0, ix1 : real) returns (x0, x1 : real);
let
  assert(ix0 > 0. and ix0 < 1.);
  assert(ix1 > 0. and ix1 < 1.);
  x0 = ix0 -> 0.95 * pre x0 + 0.09975 * pre x1;
  x1 = ix1 -> -0.1 * pre x0 + 0.95 * pre x1;
tel
*/
/*
// Original code from the Goubault paper
tau = 0.1;
x = [0,1];
v = [0,1];
while (true) {
  x = (1-( tau / 2 ) ) * x + (tau -(( tau^3 ) / 4 ) ) * v ;
  v = -tau *x+(1-( tau / 2 ) ) * v ;
}
*/

```

```

init {
  x0=[0,1];
  x1=[0,1];
}
body {
  x0p = x0; x1p = x1;
  x0 = 0.95*x0p + 0.09975*x1p;
  x1 = -0.1 *x0p + 0.95 *x1p;
  if ([0,1] > 0.5) {
    x0 = 1;
    x1 = 1;
  }
}
goal {
  // Bounds given in the Roux paper (Table 3) (for reset Ex8)
  //x0=[-1.00,1.00];
  //x1=[-1.01,1.01];
  // Bounds given in the Roux paper (Table 3) (for original Ex8)
  x0=[-1.27,1.27];
  x1=[-1.27,1.27];
  // Goal from ESOP'10 paper (policy iteration goal)
  //x=[-1.42,1.42];
  //v=[-1.42,1.42];
}
/* Example file:
   Practical policy iterations
   Roux and Garoche
   Form Methods Syst Des 2015
   DOI 10.1007/s10703-015-0230-7
   Also known as Ex.8 (Harmonic oscillator reset)
   ultimately from LMCS '12: ACCURATE NUMERICAL INVARIANTS
   by Adje, Gaubert, and Goubault
   Figure 12: An implementation of the Symplectic method */
/*
  // Code from benchmarks tarball
  node top(r : bool; ix0, ix1 : real) returns (x0, x1 : real);
  let
    assert(ix0 > 0. and ix0 < 1.);
    assert(ix1 > 0. and ix1 < 1.);
    x0 = ix0 -> if r then 1. else 0.95*pre x0 + 0.09975*pre x1;
    x1 = ix1 -> if r then 1. else -0.1*pre x0 + 0.95*pre x1;
  tel
*/
/* Original code from the Goubault paper
  tau = 0.1;
  x = [0,1];
  v = [0,1];
  while (true) {
    x = (1-( tau / 2 ) ) * x + (tau -(( tau^ 3 ) / 4 ) ) * v ;
    v = -tau *x+(1-( tau / 2 ) ) * v ;
  }
*/

```

Harm-saturated

```

init {
  x0=[0,1];
  x1=[0,1];
}
body {
  x0p = x0; x1p = x1;
  x0 = 0.95*x0p + 0.09975*x1p;
  x1 = -0.1 *x0p + 0.95*x1p;

  if (x0 > 0.5) { x0 = 0.5; }
  if (x0 < -0.5) { x0 = -0.5; }
}
goal {
  // Bounds given in the Roux paper (Table 3)
  x0=[-1.27,1.27];
  x1=[-1.27,1.27];
  // Goal from ESOP'10 paper (policy iteration goal)
  //x=[-1.42,1.42];
  //v=[-1.42,1.42];
}
/* Example file:
   Practical policy iterations
   Roux and Garoche
   Form Methods Syst Des 2015
   DOI 10.1007/s10703-015-0230-7
   Also known as Ex.8 (Harmonic oscillator saturate)
   ultimately from LMCS '12: ACCURATE NUMERICAL INVARIANTS
   by Adje, Gaubert, and Goubault
   Figure 12: An implementation of the Symplectic method
*/
/* Code from benchmarks tarball
node top(ix0, ix1 : real) returns (sx0, x0, x1 : real);
let
  assert(ix0 > 0. and ix0 < 1.);
  assert(ix1 > 0. and ix1 < 1.);
  x0=ix0 -> 0.95 * pre sx0 + 0.09975 * pre x1;
  x1=ix1 -> -0.1 * pre sx0 + 0.95 * pre x1;
  sx0=if x0 > 0.5 then 0.5 else if x0<-0.5 then -0.5 else x0;
tel
*/
/*
// Original code from the Goubault paper
tau = 0.1;
x = [0,1];
v = [0,1];
while (true) {
  x = (1-( tau / 2 ) ) * x + (tau -(( tau^3 ) / 4 ) ) * v ;
  v = -tau *x+(1-( tau / 2 ) ) * v ;
}
*/

```

```
init
{
  x = [-1.57079632679,1.57079632679];
  y=[0,0];
}
body
{
  y=x - x^3/6 + x^5/120 - x^7/5040;
}
goal
{
  x=[-2,2];
  y=[-1.05,1.05];
}
/*
  Example file from Leopold Haller's benchmark
  Original code from
  http://www.cprover.org/cdfpl/
  Simple Taylor expansion of sine

  =>
  can prove a bound of 1.05 for the output
  We wanted to illustrate that it works for tighter bounds

  the vertical bar for y=0 is expected
  (the inductive invariant must
  include both the init state,
  where y=0, and the end state, where
  y is the approximate sine)
*/
```

Newton

```
init {
  x=[-1,1];
  out=[0,0];
}

body {
  y = x - x*x*x/6 + x*x*x*x*x/120 + x*x*x*x*x*x*x/5040;
  z = 1 - x*x/2 + x*x*x*x/24 + x*x*x*x*x*x/720;
  x = x - y / z;
  out = x;
}

goal {
  x=[-1,1];
  out=[-0.56,0.56];
}
/*
  Example file from Leopold Haller's benchmark
  Original code from
  http://www.cprover.org/cdfpl/
  Newton iterations: one step
  =>
  can prove an output bound of 0.56
*/
```

Newton2

```

init {
  x=[-1,1];
  out=[0,0];
}

body {
  y = x - x*x*x/6 + x*x*x*x*x/120 + x*x*x*x*x*x*x/5040;
  z = 1 - x*x/2 + x*x*x*x/24 + x*x*x*x*x*x/720;
  x = x - y / z;

  y = x - x*x*x/6 + x*x*x*x*x/120 + x*x*x*x*x*x*x/5040;
  z = 1 - x*x/2 + x*x*x*x/24 + x*x*x*x*x*x/720;
  x = x - y / z;

  out = x;
}

goal {
  x=[-1,1];
  out=[-0.1,0.1];
}
/*
Example file from Leopold Haller's benchmark
Original code from
http://www.cprover.org/cdfpl/
Newton iterations: two steps
=>
works for bound of 0.1
*/

```

Square root

```

init {
  x=[0,1];
  out=[0,0];
}

body {
  out = 1 + 0.5*x - 0.125*x*x + 0.0625*x*x*x - 0.0390625*x*x*x*x;
}

goal {
  x=[0,1];
  out=[0,1.5];
}
/*
  Example file from Leopold Haller's benchmark
  Original code from
  http://www.cprover.org/cdfpl/
  Simple polynomial interpolation function for square root.
  =>
  we can prove that the output is <= 1.5
  we can prove tighter bound (such as 1.4 or 1.39844)
  using zonotopes but not using boxes and octagons
*/

```

Corner

```

init {
  x=[0.9,1.1];
  y=[0.9,1.1];
}

body {
  d = (0.2 + x*x + y*y + 1.53*x*x*y*y)/2;
  x = x / d;
  y = y / d;
}

goal {
  x=[-2.1,2.1];
  y=[-2.1,2.1];
}

```
