An Abstract Domain to Infer Linear Absolute Value Equalities

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Overview

- Motivation
- An abstract domain of linear absolute value equalities
- Implementation and Experiments
- Conclusion

MotivationAn abstract domain of linear absolute value equalities
Implementation and Experiments

Motivation

Goal: numerical static analysis

discover numerical properties of a program statically and automatically

Applications:

- check for runtime errors (e.g., arithmetic overflows, division by zero, array out-of-bounds, etc.)
- optimize programs
- . .

<u>Theoretical framework</u>: abstract interpretation

to design static analyses that are

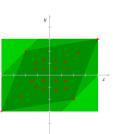
- sound by construction (no behavior is omitted)
- approximate (trade-off between precision and efficiency)

Abstract domain: key ingredient of abstract interpretation

- a specific kind of computer-representable properties
 - e.g., a family of constraints
- sound (but maybe incomplete) algorithms for semantic actions
 - e.g., join, meet, widening,...

Numerical abstract domains

- infer relationships among numerical variables
- examples
 - non-relational: **intervals** $(a \le x \le b)$
 - weakly relational: **octagons** $(\pm x \pm y \leq c)$
 - strongly relational: **polyhedra** $(\sum_{k} a_k x_k \le b)$
 - . . .



Convexity limitations: a motivating example

```
float x, y;

if (x \ge 0 \ /* \ |x| == x \ */\ ) \ \{ \ y := x; \ \}

else \ /* \ |x| == -x \ */\ \{ \ y := -x; \ \}

① if (x \ge 0 \ /* \ |x| == x \ */\ ) \ \{ \ assert(y == x); \ \}

else \ /* \ |x| == -x \ */\ \{ \ assert(y == -x); \ \}

}
```

Loc	PolkaEq	AVI	AVE
1	Т	$y == x \wedge$	$y == x \wedge$
		y == y	y == y



Absolute Value (AV): y = |x|

piecewise linear expressiveness

$$|x| = \begin{cases} x & \text{if } x \ge 0 \\ -x & \text{if } x < 0 \end{cases}$$



Possible applications

• to encode disjunctions of linear constraints in the program

•
$$(x = -1 \lor x = 1) \iff |x| = 1$$

- AV functions in C: abs(), fabs(), ...
- Min/Max functions in C: fmax(), fmin(), ...

• e.g.,
$$\max(x, y) = \frac{1}{2}(|x - y| + x + y)$$

- Rel U function in neural network
 - $ReLU(x,0) = \frac{1}{2}(|x| + x)$

The domain of linear AV inequalities: $(\Sigma_k a_k x_k + \Sigma_k b_k | x_k | \le b)$ [Chen et al. ESOP'11]

- idea: extending polyhedra domain $(\sum_k a_k x_k \leq b)$ with absolute value
- pros: piecewise linear expressiveness
- cons: exponential complexity

New idea: A domain of linear AV equalities: $(\sum_k a_k x_k + \sum_k b_k |x_k| = b)$

- goal: less costly but with non-convex expressiveness
- idea: extending the affine (linear) equality domain with absolute value
 - affine (linear) equality domain ($\sum_k a_k x_k = b$): scalable, widely used in practice

An abstract domain of linear absolute value equalities

The AVE abstract domain

An abstract domain of linear absolute value equalities (AVE)

 goal: to infer linear equality relations among values and absolute values of program variables

$$\sum_{k} a_k x_k + \sum_{k} b_k |x_k| = c$$

Domain representation for domain element P

- AVE representation: a linear AVE system Ax + B|x| = c
- semantics: $\gamma(\mathbf{P}) = \{x \in \mathbb{R}^n : Ax + B|x| = c\}$

Topological properties: can be non-convex, even unconnected

- a (possibly empty) affine space (within the orthant boundary)
 in each orthant
- e.g., y = |x|

Expressiveness limitation: $\sum_k a_k x_k + \sum_k b_k |x_k| = c$

• | · | applies to only (single) variables rather than expressions

An example: max(x, y) = z, i.e.,

$$z = \left\{ \begin{array}{ll} x & \text{if } y \le x \\ y & \text{if } x < y \end{array} \right.$$

Expressiveness lifting

- introduce new auxiliary variables to denote expressions inside the AV function
- e.g., w = x y \wedge $\frac{1}{2}(|w| + x + y) = z$

Horizontal Linear Complementary Problem (HLCP)

• given $M, N \in \mathbb{Q}^{m \times n}$ and $q \in \mathbb{Q}^m$, find $x^+, x^- \in \mathbb{Q}^n$ so that

$$Mx^+ + Nx^- = q \tag{1}$$

$$x^+, x^- \ge 0 \tag{2}$$

$$(x^+)^T x^- = 0. (3)$$

Complementarity condition: condition (3), which implies

$$x_i^+ x_i^- = 0$$
 for $i = 1, ..., n$

Equivalence of AVEs and HLCPs

Let
$$x^+ = (\max(x_i, 0))_{i=1}^n$$
 and $x^- = (\max(-x_i, 0))_{i=1}^n$, so that $x^+ \ge 0, x^- \ge 0, (x^+)^T x^- = 0$

and

$$x = x^{+} - x^{-}$$
 $|x| = x^{+} + x^{-}$
 $x^{+} = \frac{1}{2}(x + |x|)$ $x^{-} = \frac{1}{2}(|x| - x).$

Then, AVE

$$Ax + B|x| = c$$

can be reformulated as the following HLCP:

$$(A + B)x^{+} + (B - A)x^{-} = c$$

 $x^{+}, x^{-} \ge 0 \wedge (x^{+})^{T}x^{-} = 0$

Domain representation (HLCP constraints):

$$Ax^{\pm} = b, \quad x^{\pm} \ge 0, (x^{+})^{T}x^{-} = 0 \qquad (x^{\pm} \in \{x^{+}, x^{-}\})$$

• linear system part: $Ax^{\pm} = b$ in reduced row echelon form

Definition (Reduced row echelon form)

Ax = b where A is of size $m \times n$, is in reduced row echelon form if

- 1) Every row i_0 of A has at least one non-zero entry
- 2) Let $x_{j_0}^{\pm}$ be the leading variable of row i_0 of A. Then
 - $\begin{array}{lll} \bullet \ A_{i_0j_0} = 1 \\ \bullet \ \ \text{for all} \ \ i > i_0, j \leq j_0, A_{ij} = 0 \\ \bullet \ \ \text{for all} \ \ i < i_0, A_{ij_0} = 0. \end{array} \qquad \left(\begin{array}{ccccc} 1 & 0 & 0 & 1 & 2 \\ 0 & 1 & 0 & 2 & 4 \\ 0 & 0 & 1 & 0 & 7 \end{array} \right)$
- complementary condition part: standard, with no need of being stored explicitly

Double Description Method for Polyhedra

Theorem (Minkowski-Weyl Theorem)

The set $P \subseteq \mathbb{R}^n$ is a polyhedron, iff it is finitely generated, i.e., there exist finite sets $V, R \in \mathbb{R}^n$ such that P can be generated by (V, R):

$$P = \left\{ \sum_{i=1}^{|V|} \lambda_i V_i + \sum_{j=1}^{|R|} \mu_j R_j \ \middle| \ \forall i, \lambda_i \ge 0, \forall j, \mu_j \ge 0, \sum_{i=1}^{|V|} \lambda_i = 1 \right\}$$

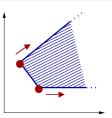
Dual representations

• constraint representation: $Ax \le b$

• e.g.,
$$\{-y \le -1, x-y \le 1, -x-y \le -3\}$$

- generator representation: G = (V, R)
 - e.g., $(\{(2,1),(1,2)\}, \{(0,1),(1,1)\})$

<u>Dual conversion:</u> Chernikova's algorithm



Computing complementary generators for HLCP:

$$Mx^+ + Nx^- = c \wedge x^+, x^- \ge 0 \wedge (x^+)^T x^- = 0$$

Step1:
$$G \leftarrow \text{Polyhedra.Cons2Gens} (Mx^+ + Nx^- = c \land x^+, x^- \ge 0)$$

Step2: $G^c \leftarrow \{g \in G \mid g \text{ satisfies } (x_{\sigma}^+)^T x_{\sigma}^- = 0\}$

Dual representations

• HLCP constraint representation:

$$Mx^+ + Nx^- = c \wedge x^+, x^- \ge 0 \wedge (x^+)^T x^- = 0$$

• complementary generator representation: $G^c = (V^c, R^c)$

How to implement AVE domain operations for static analysis

• maintain the map between abstract environments over x and abstract environments over x^+, x^- :

$$x = x^{+} - x^{-},$$
 $|x| = x^{+} + x^{-}$
 $x^{+} = \frac{1}{2}(x + |x|),$ $x^{-} = \frac{1}{2}(|x| - x)$

where
$$x^+, x^-$$
 satisfy $x^+ \ge 0, x^- \ge 0, (x^+)^T x^- = 0$

• compute $G^c = (V^c, R^c)$, the set of complementary generators of HLCP system (when needed):

$$Mx^{+} + Nx^{-} = b$$

 $x^{+} \ge 0, x^{-} \ge 0, (x^{+})^{T}x^{-} = 0$

Domain operations

- lattice operations
 - meet: P □ P' is an AVE domain element whose HLCP constraint representation is

$$Mx^{+} + Nx^{-} = b$$

$$M'x^{+} + N'x^{-} = b'$$

$$x^{+} \ge 0, x^{-} \ge 0, (x^{+})^{T}x^{-} = 0$$

• where $\{Mx^+ + Nx^- = b, M'x^+ + N'x^- = b'\}$ can be converted into reduced row echelon form via Gaussian elimination

Domain operations

- lattice operations
 - join: P ⊔ P' is the least AVE element containing P and P', whose set of complementary generators is the union of those of P and P': (V^c ∪ V'^c, R^c ∪ R'^c).
 - ① Compute the complementary generator representation (V^c, R^c) , (V'^c, R'^c) respectively for **P** and **P'**;
 - ② Compute $(V^c \cup V'^c, R^c \cup R'^c)$, and suppose $V^c \cup V'^c = \{v_1, \dots, v_a\}, R^c \cup R'^c = \{r_1, \dots, r_a\}$;
 - **3** Project out variables $\lambda_j(j=1,\ldots,p), \mu_k(k=1,\ldots,q)$ (via Gaussian elimination) from the following system:

$$\begin{cases} (x^{+} x^{-})^{T} = \sum_{j=1}^{p} (\lambda_{j} v_{j}) + \sum_{k=1}^{q} (\mu_{k} r_{k}) \\ \sum_{j=1}^{p} \lambda_{j} = 1 \end{cases}$$

Suppose we get $\hat{M}x^+ + \hat{N}x^- = \hat{b}$

4 Finally, the resulting HLCP representation of $P \sqcup P'$ is:

$$\hat{M}x^{+} + \hat{N}x^{-} = \hat{b}$$

$$x^{+} > 0, x^{-} > 0, (x^{+})^{T}x^{-} = 0$$

float
$$x, y$$
; if $(x \ge 0 \ /* \ |x| == x \ */\) \ \{ \ y := x; \ \textcircled{1} \ \}$ else $\ /* \ |x| == -x \ */\ \{ \ y := -x; \ \textcircled{2} \ \}$ $\textcircled{3} \dots$

Loc	AVE/HLCP constraints	Complementary generators
1	$ \begin{aligned} \mathbf{P} &= \{(x \ y)^T \mid x - y = 0, x = x\} \\ \{(x^+ \ x^- \ y^+ \ y^-)^T \mid x^+ - y^+ + y^- = 0, x^- = 0, \\ x^{\pm} &\geq 0, y^{\pm} \geq 0, x^+ x^- = 0, y^+ y^- = 0\} \end{aligned} $	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array} \right) \right\} \right)$
2	$P' = \{(x \ y)^T \mid -x - y = 0, x = -x\} = \{(x^+ \ x^- \ y^+ \ y^-)^T \mid x^ y^+ + y^- = 0, x^+ = 0, x^{\pm} \ge 0, y^{\pm} \ge 0, x^+ x^- = 0, y^+ y^- = 0\}$	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array} \right) \right\} \right)$
3	?	?

float
$$x, y$$
;
if $(x \ge 0 \ /* |x| == x */) \{ y := x; ① \}$
else $/* |x| == -x */ \{ y := -x; ② \}$
③ ...

Loc	AVE/HLCP constraints	Complementary generators
1	$P = \{(x y)^T \mid x - y = 0, x = x\} = \{(x^+ x^- y^+ y^-)^T \mid x^+ - y^+ + y^- = 0, x^- = 0, x^+ \ge 0, y^{\pm} \ge 0, x^{\pm} x^{-} = 0, y^{\pm} y^{-} = 0\}$	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array} \right) \right\} \right)$
2	$\begin{aligned} & \mathbf{P}' = \{(x \ y)^T \ \ -x - y = 0, x = -x\} = \\ & \{(x^+ \ x^- \ y^+ \ y^-)^T \ \ x^ y^+ + y^- = 0, x^+ = 0, \\ & x^{\pm} \ge 0, y^{\pm} \ge 0, x^{\pm} x^- = 0, y^{\pm} y^- = 0\} \end{aligned}$	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array} \right) \right\} \right)$
3	?	$ \left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array} \right), \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array} \right) \right\} \right) $

float
$$x, y$$
; if $(x \ge 0 \ /* \ |x| == x \ */\) \ \{ \ y := x; \ @ \ \}$ else $\ /* \ |x| == -x \ */\ \{ \ y := -x; \ @ \ \}$ $\ @ \dots$

Loc	AVE/HLCP constraints	Complementary generators			
1					
2					
3	?	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array} \right), \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array} \right) \right\} \right)$			

Projecting out λ_1, μ_1, μ_2 (wherein $\lambda_1 = 1$) from

$$\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array}\right) = \lambda_1 \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array}\right) + \mu_1 \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array}\right) + \mu_2 \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array}\right)$$

will result in
$$x^+ + x^- - y^+ = 0$$
, $y^- = 0$.

float
$$x, y$$
;
if $(x \ge 0 \ /* |x| == x */) \{ y := x; ① \}$
else $/* |x| == -x */ \{ y := -x; ② \}$
③ ...

Loc	AVE/HLCP constraints	Complementary generators
1	$P = \{(x y)^T \mid x - y = 0, x = x\} = \{(x^+ x^- y^+ y^-)^T \mid x^+ - y^+ + y^- = 0, x^- = 0, x^{\pm} \ge 0, y^{\pm} \ge 0, x^+ x^- = 0, y^+ y^- = 0\}$	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array} \right) \right\} \right)$
2	$\begin{aligned} & \mathbf{P}' = \{(x \ y)^T \ \ -x - y = 0, \ x = -x\} = \\ & \{(x^+ \ x^- \ y^+ \ y^-)^T \ \ x^- \ -y^+ + y^- = 0, \ x^+ = 0, \\ & x^{\pm} \ge 0, \ y^{\pm} \ge 0, \ x^+ x^- = 0, \ y^+ y^- = 0\} \end{aligned}$	$\left(\left(\begin{array}{c} x^+ \\ x^- \\ y^+ \\ y^- \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array} \right) \right\} \right)$
3	$\begin{aligned} \mathbf{P}' &= \{(x \ y)^T \ \ y = x , \ y = y\} = \\ \{(x^+ \ x^- \ y^+ \ y^-)^T \ \ x^+ + x^ y^+ = 0, y^- = 0, \\ x^{\pm} &\geq 0, y^{\pm} \geq 0, x^+ x^- = 0, y^+ y^- = 0\} \end{aligned}$	$ \left(\left(\begin{array}{c} x^{+} \\ x^{-} \\ y^{+} \\ y^{-} \end{array} \right) : \left\{ \left(\begin{array}{c} 0 \\ 0 \\ 0 \\ 0 \end{array} \right) \right\}, \left\{ \left(\begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \end{array} \right), \left(\begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \end{array} \right) \right\} \right) $

Domain operations

- transfer functions
 - test transfer function: $\tau[[cx + d|x] = e]]^{\sharp}(\mathbf{P})$, whose HLCP system is defined as

$$Mx^{+} + Nx^{-} = b$$

$$(c+d)x^{+} + (d-c)x^{-} = e$$

$$x^{+} \ge 0, x^{-} \ge 0, (x^{+})^{T}x^{-} = 0$$

- where $\{Mx^+ + Nx^- = b, (c+d)x^+ + (d-c)x^- = e\}$ can be converted into reduced row echelon form via Gaussian elimination
- projection: $\tau[x_j := random()]^{\sharp}(\mathbf{P})$, can be implemented by projecting out x_j^+, x_j^- via Gaussian elimination from

$$Mx^+ + Nx^- = b$$

• assignment transfer function: $\tau[[x_j := \sum_i a_i x_i + \sum_i b_i | x_i | + c]]^{\sharp}(\mathbf{P})$, can be implemented as:

$$\left(\tau\llbracket x_j := \mathit{random}()\rrbracket^\sharp \circ \tau\llbracket \Sigma_i a_i x_i + \Sigma_i b_i | x_i | + c - x_j' = 0 \rrbracket^\sharp(\mathbf{P})\right) [x_j'/x_j]$$

Domain operations

- Second Extrapolations (Widening):
 - the lattice of linear equalities (in a program) has finite height, and thus
 we do not need a widening operation for the domain of linear equalities.
 - the intersection of an AVE element with each orthant, results in an affine space, i.e., an element in the domain of linear equalities.
 - the number of the orthants are finite (for a given program)
 - → we also do not need a widening operation for the AVE domain. At each widening point, we use the join operator ⊔ instead of the widening.

Implementation and Experiments

Prototype

Prototype implementation rAVE using:

- GMP (the GNU Multiple Precision arithmetic library)
 - to guarantee the soundness of the implementation
- NewPolka: a rational implementation of the polyhedra domain
 - for Chernikova's algorithm

Interface:

- plugged into the APRON library [Jeannet Miné]
- programs analyzed with INTERPROC [Jeannet et al.]

Comparison with

- PolkaEq [Jeannet]: the linear equality domain in APRON
- rAVI: the domain of linear absolute value inequalities [Chen et al. ESOP11]

Example analyses

```
real x, y; assume x = 1 or x = -1; assume y = 1 or y = -1; while (true) {
① if (x \ge 0 \ /* \ |x| == \ x \ */\ ) \ \{ \ x := \ x + 1; \ \} else \ /* \ |x| == \ -x \ */\ \{ \ x := \ x - 1; \ \} else \ /* \ |y| == \ y \ */\ ) \ \{ \ y := \ y + 1; \ \} else \ /* \ |y| == \ -y \ */\  \ \{ \ y := \ y - 1; \ \} }
```

Loc	PolkaEq	rAVE	rAVI			
1	Т	x = y	$ x = y \land x \ge 1$			

Preliminary experimental results

	PolkaEq		rAVE		rAVI		Invariant	
Program							PolkaEq	rAVE
6	#iter.	t(ms)	#iter.	t(ms)	#iter.	t(ms)	vs. rAVE	vs. rAVI
MotivEx	1	3.3	1	4.0	1	5.1		=
AVtest1	3	7.1	4	11.2	3	12.3		
Complexity_cav08	3	4.4	4	14.4	4	20.7		
Synergy1	3	5.3	3	17.3	4	30.9		=
Reverse	3	4.0	3	5.6	4	8.7		=
Recwhile	3	3.6	7	24.5	7	31.2		
Speed_popl09	3	5.5	4	25.0	4	30.3		

- These programs involve non-convex behaviors (such as absolute value functions, max functions, disjunctions, etc.) that are out of the expressiveness of convex domains (including PolkaEq)
- ullet rAVE outputs $1{\sim}6$ linear AV equality invariants for each example at loop head
- rAVI also infers certain linear inequalities and linear AV inequalities, which are out of the expressiveness of rAVE

Conclusion

Summary:

a new abstract domain: linear absolute value equalities (AVE)

$$(\Sigma_k a_k x_k + \Sigma_k b_k | x_k | = c)$$

- idea: extend the affine equality domain with absolute value
 - can express non-convex (even unconnected) properties
- key:
 - making use of the equivalence between AVEs and HLCPs
 - maintaining the reduced row echelon form for the linear system part of HLCP representation
 - \rightarrow at most 2n linear AV equalities for a program involving n variables

Future Work

- for precision
 - introducing automatically auxiliary variables inside the AV function
 - combining the AVE abstract domain with the interval domain
- more experiments on large realistic programs