MPRI

Static Analysis of Digital Filters

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Overview

Context

We want to prove run time error absence, in critical embedded software.
Filter behaviour is implemented at the software level, using hardware floating point numbers.

Full certification requires special care about these filters.

Issues

- Control flow detection: to locate filter resets and filter iterations.

- Invariant inference: we are not interested in functional properties.

  We seek precise bounds on the output, using information inferred about the input.

  (Linear invariants do not yield accurate bounds).

- To take into account floating-point rounding:
  - in the semantics,
  - when implementing the abstract domain.

Overview

The high bandpass filter

We consider the following example:

$$
\begin{array}{l}
V \in \mathbb{R};\ \ E_1 := 0; S := 0;\\
\text{while } (V \geq 0)\ \{\\
\quad V \in \mathbb{R};\ \ T \in \mathbb{R};\\
\quad E_0 \in [-1;1];\\
\quad \text{if } (T \geq 0)\ \{S := 0\}\\
\quad \text{else } \{S := 0.999 \times S + E_0 - E_1\}\\
\quad E_1 := E_0;\\
\}
\end{array}
$$

Interval approximation (simplified)

With a view to simplifying, we ignore rounding errors !!!

The analyzer infers the following sound counterpart $\mathbb{F}^\sharp$:

$$\mathbb{F}^\sharp(X) = \{0.999 * s + e_0 + e_1 \mid s \in X,\ e_0, e_1 \in [-1; 1]\}$$

to the loop body.

Abstract iteration

1. The analyzer starts iterating $\mathbb{F}^\sharp$:
   $\mathbb{F}^\sharp(\{0\}) = [-2;2]$,
   $\mathbb{F}^\sharp([-2;2]) = [-3.998;3.998]$,
   $\ldots$;
2. then it widens the iterates:
   $\mathbb{F}^\sharp([-10;10]) \not\subseteq [-10;10]$,
   $\mathbb{F}^\sharp([-100;100]) \not\subseteq [-100;100]$,
   $\ldots$;
3. until it discovers a stable threshold:
   $\mathbb{F}^\sharp([-10000;10000]) = [-9992;9992]$;
4. finally, it keeps iterating to refine the solution:
   $\mathbb{F}^\sharp([-9992;9992]) = [-9984.008;9984.008]$.

Driving the analysis

Better results could have been obtained by driving the analysis:

Theorem 1 (High bandpass filter (history-insensitive))
Let $D \geq 0$, $m \geq 0$, a, X *and* Z *be real numbers such that:*

1. $|X| \leq D$;
2. $aX - m \leq Z \leq aX + m$;

then we have:

1. $|Z| \leq |a|D + m$;
2. $\left[|a| < 1 \text{ and } D \geq \frac{m}{1-|a|}\right] \implies |Z| \leq D$. $\square$

Theorem 1 implies that 2000 can be used as a threshold.

History sensitive approximation

Theorem 2 (High bandpass filter (history-sensitive version))
Let $\alpha \in [\frac{1}{2};1[$, i *and* $m > 0$ *be real numbers.*
Let E_n *be a real number sequence, such that* $\forall k \in \mathbb{N},\ E_k \in [-m;m]$.
Let S_n *be the following sequence:*

$$\begin{cases} S_0 = i \\ S_{n+1} = \alpha.S_n + E_{n+1} - E_n. \end{cases}$$

We have:

1. $S_n = \alpha^n.i + E_n - \alpha^n E_0 + \sum_{l=1}^{n-1}(\alpha - 1)\alpha^{l-1}E_{n-l}$
2. $|S_n| \leq |\alpha|^n|i| + (1 + |\alpha|^n + |1 - \alpha^{n-1}|)m$;
3. $|S_n| \leq 2.m + |i|$. □

Theorem 2 implies that 2 is a sound bound on $|S|$.

The second order filter

$V \in \mathbb{R};$

$E_1 := 0;\ \ E_2 := 0;\ \ S_0 := 0;\ \ S_1 := 0;\ \ S_2 := 0;$

while $(V \geq 0)$ $\{$

$V \in \mathbb{R};\ T \in \mathbb{R};$

$E_0 \in [-1;1];$

if $(T \geq 0)$ $\{S_0 := E_0; S_1 := E_0; E_1 := E_0\}$

else $\{S_0 := 1.5 \times S_1 - 0.7 \times S_2$

$+ 0.5 \times E_0 - 0.7 \times E_1 + 0.4 \times E_2\};$

$E_2 := E_1;\ E_1 := E_0;$

$S_2 := S_1;\ S_1 := S_0$

$\}$

Ellipsoidal constraints

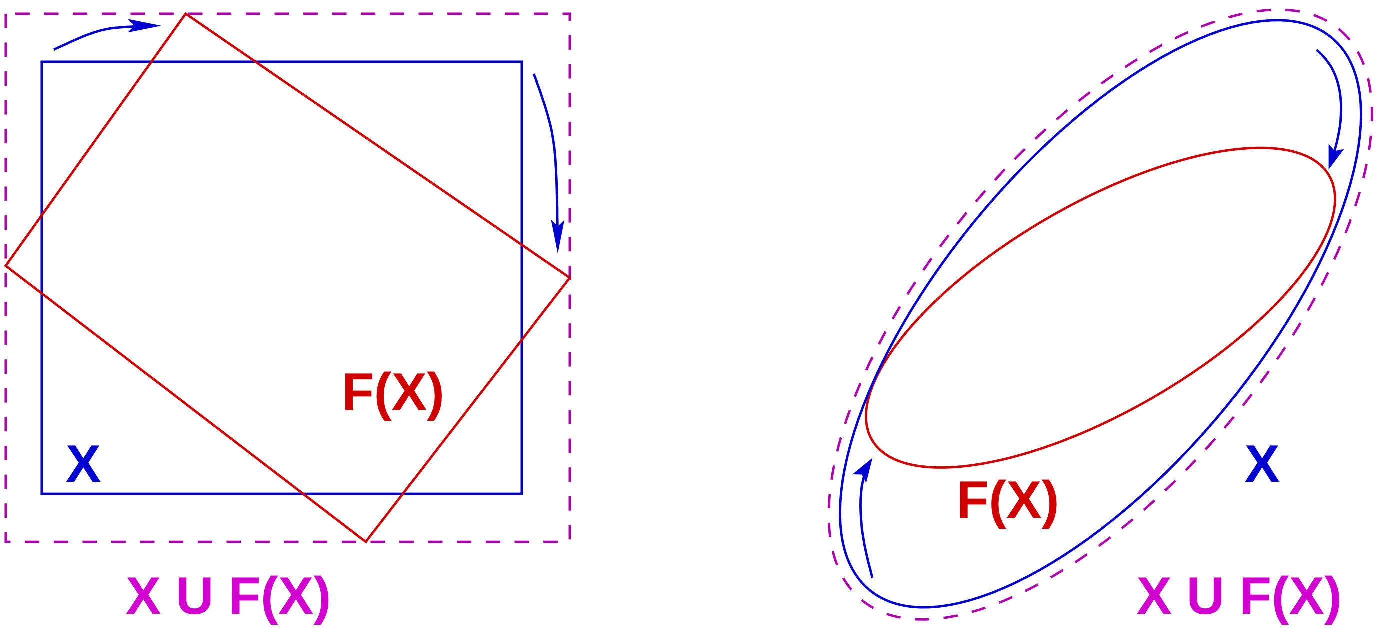
Theorem 3 (second order filter (history insensitive))
Let a, b, $K \geq 0$, $m \geq 0$, X, Y, Z *be real numbers such that:*

1. $a^2 + 4b < 0$,
2. $X^2 - aXY - bY^2 \leq K$,
3. $aX + bY - m \leq Z \leq aX + bY + m$.

We have:

1. $Z^2 - aZX - bX^2 \leq \left(\sqrt{-bK} + m\right)^2$;
2. $\begin{cases} \sqrt{-b} < 1 \\ K \geq \left(\frac{m}{1-\sqrt{-b}}\right)^2 \end{cases} \implies Z^2 - aZX - bX^2 \leq K$.

Linear versus quadratic invariants



Second order filter approximation

1. without relational domain,
   we cannot limit $|S_2|$;
2. with ellipsoidal constraints (history insensitive abstraction),
   we can infer that $|S_2| < 22.111$;
3. by formally expanding the output as a sum of all previous inputs,
   we can prove that $|S_2| < 1.41824$;

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Syntax

Let $\mathcal{V}$ be a finite set of variables.
Let $\mathcal{I}$ be the set of real intervals (including $\mathbb{R}$).
Expressions $\mathcal{E}$ are affine forms of variables $\mathcal{V}$ with real interval coefficients:

$$E ::= I + \sum_{j \in J} I_j . V_j$$

Programs are given by the following grammar:

$$\begin{array}{rcl} P & ::= & \text{skip} \\ & | & P;P \\ & | & V := E \\ & | & \text{if } (V \geq 0) \; \{P\} \text{ else } \{P\} \\ & | & \text{while } (V \geq 0) \; \{P\} \end{array}$$

Semantics

We define the semantics of a program P:

$$[\![P]\!] : (\mathcal{V} \rightarrow \mathbb{R}) \rightarrow \wp(\mathcal{V} \rightarrow \mathbb{R})$$

by induction over the syntax of P:

$[\![\text{skip}]\!](\rho) = \{\rho\}$,

$[\![P_1;P_2]\!](\rho) = \{\rho'' \mid \exists \rho' \in [\![P_1]\!](\rho),\ \rho'' \in [\![P_2]\!](\rho')\}$,

$[\![V := I + \sum_{j \in J} I_j.V_j]\!](\rho) = \left\{\rho\left[V \mapsto i + \sum_{j \in J} i_j.\rho(V_j)\right] \;\middle|\; i \in I,\ \forall j \in J, i_j \in I_j\right\}$,

$[\![\text{if } (V \geq 0)\ \{P_1\} \text{ else } \{P_2\}]\!](\rho) = \begin{cases} [\![P_1]\!](\rho) & \text{if } \rho(V) \geq 0 \\ [\![P_2]\!](\rho) & \text{otherwise,} \end{cases}$

$[\![\text{while } (V \geq 0)\ \{P\}]\!](\rho) = \{\rho' \in \mathit{Inv} \mid \rho'(V) < 0\}$
where $\mathit{Inv} = \text{lfp}\,(X \mapsto \{\rho\} \cup \{\rho'' \mid \exists \rho' \in X,\ \rho'(V) \geq 0 \text{ and } \rho'' \in [\![P]\!](\rho')\})$.

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Abstract domain

An abstract domain $\text{ENV}^\sharp$ is a set of environment properties.

A concretization map γ relates each property to the set of its solutions:

$$\gamma : \text{ENV}^\sharp \to \wp(\mathcal{V} \to \mathbb{R}).$$

Some primitives simulate concrete computation steps in the abstract:

- an abstract control path merge $\sqcup$;
- an abstract guard GUARD and an abstract assignment ASSIGN;
- an abstract least fixpoint $\text{lfp}^\sharp$ operator, which maps sound counterpart $f^\sharp$ to monotonic function f, to an abstraction of the least fixpoint of f.
  $\text{lfp}^\sharp$ is defined using extrapolation operators $(\bot, \nabla, \Delta)$.
  Soundness follows from the monotony of the concrete semantics.

Abstract semantics

$$[\![\text{skip}]\!]^\sharp(a) = a$$

$$[\![P_1;P_2]\!]^\sharp(\rho^\sharp) = [\![P_2]\!]^\sharp([\![P_1]\!]^\sharp(\rho^\sharp))$$

$$[\![V := E]\!]^\sharp(a) = \text{ASSIGN}(V, E, a)$$

$$[\![\text{if } (V \geq 0)\ \{P_1\} \text{ else } \{P_2\}]\!]^\sharp(a) = a_1 \sqcup a_2,$$

$$\text{with } \begin{cases} a_1 = [\![P_1]\!]^\sharp(\text{GUARD}(V, [0;+\infty[, a)) \\ a_2 = [\![P_2]\!]^\sharp(\text{GUARD}(V,]-\infty;0[, a)) \end{cases}$$

$$[\![\text{while } (V \geq 0)\ \{P\}]\!]^\sharp(a) = \text{GUARD}(V,]-\infty;0[, \mathit{Inv}^\sharp)$$

$$\text{where } \mathit{Inv}^\sharp = \text{lfp}^\sharp\left(X \mapsto a \sqcup [\![P]\!]^\sharp(\text{GUARD}(V, [0;+\infty[, X))\right)$$

Soundness

We prove by induction over the syntax:

Theorem 4 (Soundness) *For any program P, environment ρ, abstract element a, we have:*

$$\rho \in \gamma(a) \implies [\![P]\!](\rho) \subseteq \gamma\left([\![P]\!]^\sharp(a)\right).$$

$\square$

Extrapolation operators

- iteration basis: $\bot \in \text{ENV}^\sharp$
- a widening operator ∇ such that:
  1. $\nabla \in (\text{ENV}^\sharp \times \text{ENV}^\sharp) \rightarrow \text{ENV}^\sharp$,
  2. $\forall a, b \in \text{ENV}^\sharp,\ \gamma(a) \cup \gamma(b) \subseteq \gamma(a \nabla b)$,
  3. $\forall\, (a_i) \in (\text{ENV}^\sharp)^{\mathbb{N}}$, the sequence (a_i^{∇}) defined by:
  $$a_0^{\nabla} = a_0 \text{ and } a_{n+1}^{\nabla} = a_n^{\nabla} \nabla a_{n+1}$$
  is ultimately stationary;
- a narrowing operator Δ such that:
  1. $\Delta \in (\text{ENV}^\sharp \times \text{ENV}^\sharp) \rightarrow \text{ENV}^\sharp$,
  2. $\forall a, b \in \text{ENV}^\sharp,\ \gamma(a) \cap \gamma(b) \subseteq \gamma(a \Delta b)$,
  3. $(a_i) \in (\text{ENV}^\sharp)^{\mathbb{N}}$, the sequence (a_i^{Δ}) defined by:
  $$a_0^{\Delta} = a_0 \text{ and } a_{n+1}^{\Delta} = a_n^{\Delta} \Delta a_{n+1}$$
  is ultimately stationary;

Abstract iterations

Let $f^\sharp$ be a map in $\mathrm{ENV}^\sharp \rightarrow \mathrm{ENV}^\sharp$.

Abstract upward-iterates:

$$\begin{cases} C_0^\triangledown = \bot, \\ C_{n+1}^\triangledown = C_n^\triangledown \triangledown f^\sharp(C_n^\triangledown), \end{cases}$$

is eventually stationary: We denote by $C_\omega^\triangledown$ its limit.

Abstract downward-iterates:

$$\begin{cases} D_0^\vartriangle = C_\omega^\triangledown, \\ D_{n+1}^\vartriangle = D_n^\vartriangle \vartriangle f^\sharp(D_n^\vartriangle), \end{cases}$$

is eventually stationary: We define $\mathrm{lfp}^\sharp(f^\sharp)$ as this limit.

Soundness

Let f be a $\cup$-complete morphism such that:

$$\forall a \in \text{ENV}^\sharp,\ f(\gamma(a)) \subseteq \gamma(f^\sharp(a)).$$

We want to prove that $\text{lfp}(f) \subseteq \gamma(\text{lfp}^\sharp(f^\sharp))$.

Since $\text{lfp}(a) = \bigcap\{a \mid f(a) \subseteq a\}$ (Tarski), we only have to prove that:

$$\exists a \in \wp(\mathcal{V} \to \mathbb{R}),\ f(a) \subseteq a \text{ and } a \subseteq \gamma(\text{lfp}^\sharp(f^\sharp)).$$

Soundness proof (continued)

1. $f(\gamma(C_\omega^\nabla)) \subseteq \gamma(C_\omega^\nabla)$ since:
   $f(\gamma(C_\omega^\nabla)) \subseteq \gamma(f^\sharp(C_\omega^\nabla))$, (soundness of $f^\sharp$)
   $\gamma(f^\sharp(C_\omega^\nabla)) \subseteq \gamma(C_\omega^\nabla \nabla f^\sharp(C_\omega^\nabla))$, (soundness of ∇)
   $C_\omega^\nabla \nabla f^\sharp(C_\omega^\nabla) = C_\omega^\nabla$, ($C_\omega^\nabla$ is a limit)
2. $\forall n \in \mathbb{N},\ \exists a \in \wp(\mathcal{V} \to \mathbb{R})$ such that $f(a) \subseteq a$ and $a \subseteq \gamma(D_n^\Delta)$:
   (a) $\gamma(D_0^\Delta) = \gamma(C_\omega^\nabla)$ and $f(\gamma(C_\omega^\nabla)) \subseteq \gamma(C_\omega^\nabla)$;
   (b) let $a \in \wp(\mathcal{V} \to \mathbb{R})$ such that $f(a) \subseteq a$ and $a \subseteq \gamma(D_n^\Delta)$,
   then
   - $f(f(a)) \subseteq f(a)$ (f is monotonic),
     $f(a) \subseteq f(\gamma(D_n^\Delta)) \subseteq \gamma(f^\sharp(D_n^\Delta))$,
   - $f(a \cap f(a)) \subseteq f(a) \cap f(f(a)) \subseteq a \cap f(a)$,
     $a \cap f(a) \subseteq \gamma(D_n^\Delta) \cap \gamma(f^\sharp(D_n^\Delta)) \subseteq \gamma(D_n^\Delta \Delta f^\sharp(D_n^\Delta)) \subseteq \gamma(D_{n+1}^\Delta)$

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Filter family

A filter class is given by:

- the number p of outputs and the number q of inputs involved in the computation of the next output;
- a (generic/symbolic) description of F with parameters;
- some conditions over these parameters.

In the case of the second order filter:

- $p = 2, \ \ q = 3$;
- $F(S_{n-1}, S_{n-2}, E_{n+2}, E_{n+1}, E_n) = a.S_{n-1} + b.S_{n-2} + c.E_{n+2} + d.E_{n+1} + e.E_n$;
- $a^2 + 4b < 0$.

Filter domain

A filter constraint is a couple in $\mathcal{T}_{\mathcal{B}} \times \mathcal{B}$ where:

- $\mathcal{T}_{\mathcal{B}} \in \wp_{\text{finite}}(\mathcal{V}^m \times \mathbb{R}^n)$ with:
  - m, the number of variables that are involved in the computation of the next output. m depends on the abstraction;
  - n, the number of filter parameters;
- $\mathcal{B}$ is an abstract domain encoding some "ranges".

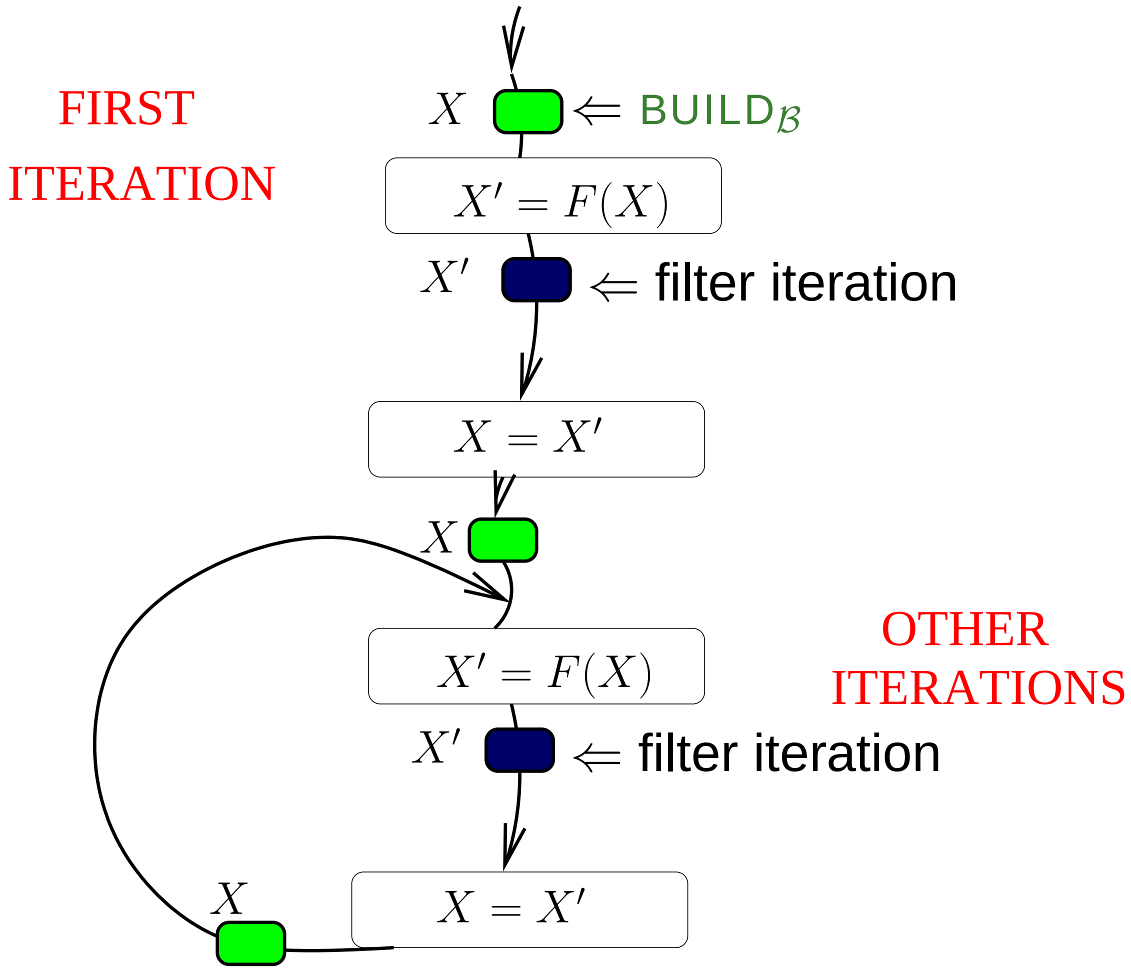
A constraint (t, d) is related to $\wp(\mathcal{V} \to \mathbb{R})$, by a concretization function:

$$\gamma_{\mathcal{B}} : \ \mathcal{T}_{\mathcal{B}} \times \mathcal{B} \to \wp(\mathcal{V} \to \mathbb{R}).$$

An approximation of second order filter may consist in relating:

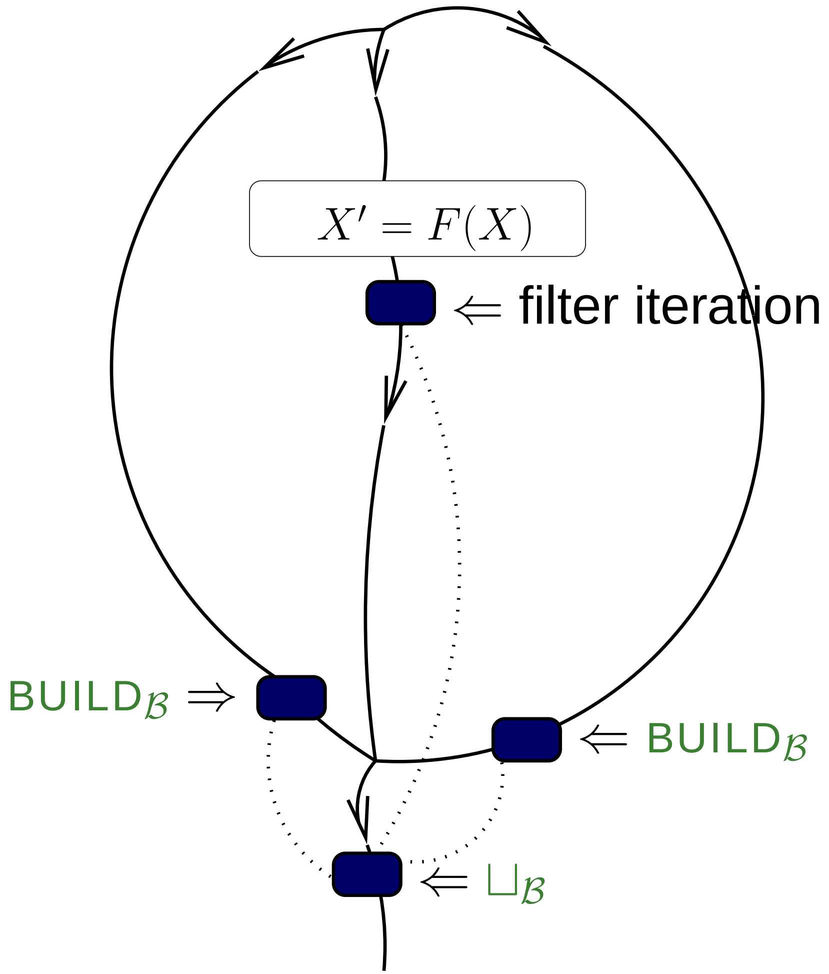
- the last two outputs and the first two coefficients of the filter (a and b)
- to the 'ratio' of an ellipsoid.

Assignment



FIRST
ITERATION
X
$\Leftarrow$ BUILD$_{\mathcal{B}}$
$X' = F(X)$
X'
$\Leftarrow$ filter iteration
$X = X'$
X
OTHER
ITERATIONS
$X' = F(X)$
X'
$\Leftarrow$ filter iteration
X
$X = X'$

Merging computation paths



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Floating point domain

Let:

- $\mathbb{F}$ be a finite subset of $\mathbb{R}$ closed upon opposite,
- L is a finite subset of $\mathbb{F}$;
- q, r two natural parameters for setting extrapolation strategy.

We define $\mathcal{F}_{q,r}$ as follows:

- $\mathcal{F}_{q,r} = \overline{\mathbb{F}} = \mathbb{F} \cup \{-\infty; +\infty\}$;
- $\gamma_{\overline{\mathbb{F}}} : \begin{cases} \overline{\mathbb{F}} & \mapsto \wp(\mathbb{R}) \\ a & \to \begin{cases} [-a; a] & \text{if } a \in \mathbb{F} \\ \mathbb{R} & \text{otherwise}; \end{cases} \end{cases}$
- $\lceil \_ \rceil : \begin{cases} \mathbb{R} \to \overline{\mathbb{F}} \\ r \to \mathit{min}(\{f \in \overline{\mathbb{F}} \mid f \geq r\}); \end{cases}$
- $a \nabla_{\overline{\mathbb{F}}} b = \mathit{max}(\{a, \mathit{min}(\{l \in L \cup \{a; +\infty\} \mid l \geq b\})\})$.

Extrapolation strategy

- Delayed widening:

$$(a_1, k_1) \nabla_{\mathcal{F}_{q,r}} (a_2, k_2) = \begin{cases} (a_1, k_1) & \text{if } a_1 \geq a_2 \\ (a_2, k_1 + 1) & \text{if } a_2 > a_1 \text{ and } k_1 < q \\ (a_1 \nabla_{\overline{\mathbb{F}}} a_2, 0) & \text{otherwise}; \end{cases}$$

  Constraints are only widened when they have been unstable (not necessarily successively) q times, since their last widening.

- Bounded narrowing:

$$(a_1, k_1) \Delta_{\mathcal{F}_{q,r}} (a_2, k_2) = \begin{cases} (a_1, k_1) & \text{if } a_1 \leq a_2 \text{ or } k_1 \leq (-r) \\ (a_2, \mathit{min}(k_1, 0) - 1) & \text{if } a_2 < a_1 \text{ and } k_1 > (-r); \end{cases}$$

  Constraints are only narrowed r times.

Approximating contracting functions

When analyzing filter, we iterate functions f such that:

- $f : I \times \mathbb{F} \to \mathbb{F}$
- $\forall i \in I$, the map $[x \to f(i,x)]$ is contracting;
- we can compute $f_l : I \to \mathbb{F}$ such that $\forall i \in I,\ f(i, f_l(i)) \leq f_l(i)$;

where I is a set of inputs.

Since $[x \to f(i,x)]$ is contracting, we have:

- $\forall i \in I,\ \forall x \geq f_l(i),\ f(i,x) \leq x$.

Our goal

We want to find a iterating strategy which ensures:

- soundness (even if f_l is unsound)
- accuracy (if f_l is sound):
  - do not jump directly at the limit f_l: (to analyze not iterated filter, loop unrolling...)
  - do not jump higher than the limit when the input is constant;
  - do not jump higher than the limit in most cases.
- termination (even if the input depend on the output).

Reduced product

We use an approximation of the reduced product of two domains:
Let q,r be two natural parameters.

1. the first domain iterates f in $\mathcal{F}_{0,r}$

   $\Longrightarrow$ widened at each step;

2. the second domain iterates $[(i,x) \rightarrow \mathit{max}(f(i,x), f_l(i))]$ in $\mathcal{F}_{q,0}$

   $\Longrightarrow$ soundness does not depend on f_l

   $\Longrightarrow$ not widened at each step to wait until input are stables.

We use the reduction:

$$\rho : \begin{cases} \mathcal{F}_{0,r} \times \mathcal{F}_{q,0} & \mapsto \mathcal{F}_{0,r} \times \mathcal{F}_{q,0} \\ (x_0, m_0), (x_1, m_1) & \rightarrow (\mathit{min}(x_0, x_1), m_0), (x_1, m_1) \end{cases}$$

after each computation step.

$\Longrightarrow$ The second domain is used to reduce the first one, when it is not accurate.

Unstable filters

In case the iterated function is not contracting, filters are very likely to diverge.
In case of linear filters, the iterated function is linear.
We may use the arithmetic-geometric progression domain [VMCAI'2005].
We require an external clock to relate the divergence to the value of the clock.

Overview

Simplified second order filter

Theorem 5 (Including rounding errors)
Let a, b, $\varepsilon_a \geq 0$, $\varepsilon_b \geq 0$, $K \geq 0$, $m \geq 0$, X, Y, Z *be real numbers, such that:*

1. $a^2 + 4b < 0$,
2. $X^2 - aXY - bY^2 \leq K$,
3. $aX + bY - (m + \varepsilon_a|X| + \varepsilon_b|Y|) \leq Z \leq aX + bY + (m + \varepsilon_a|X| + \varepsilon_b|Y|)$.

We have

1. $Z^2 - aZX - bX^2 \leq \left((\sqrt{-b} + \delta)\sqrt{K} + m\right)^2$;
2. $\begin{cases} \sqrt{-b} + \delta < 1 \\ K \geq \left(\frac{m}{1-\sqrt{-b}-\delta}\right)^2 \end{cases} \implies Z^2 - aZX - bX^2 \leq K$,

where $\delta = 2\frac{\varepsilon_b + \varepsilon_a\sqrt{-b}}{\sqrt{-(a^2+4b)}}$.

□

Domain

- The domain relates the variables describing the last two outputs and the four filter parameters to the square of the ellipsoid ratio:

  $\gamma_{\mathcal{B}_1}((X, Y, a, \varepsilon_a, b, \varepsilon_b), K)$ is given by the set of environments ρ that satisfy:

  $$(\rho(X))^2 - a\rho(X)\rho(Y) - b(\rho(Y))^2 \leq K;$$

- in order to interpret assignment $Z = E$ under range constraints $\rho^\sharp$, we test whether E matches:

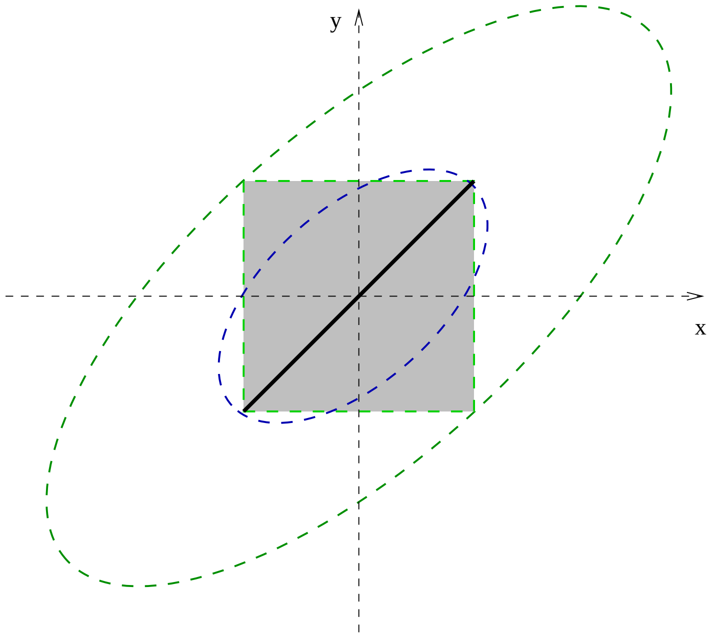
  $$[a - \varepsilon_a; a + \varepsilon_a] \times X + [b - \varepsilon_b; b + \varepsilon_b] \times Y + E'$$

  with $a^2 + 4b < 0$,

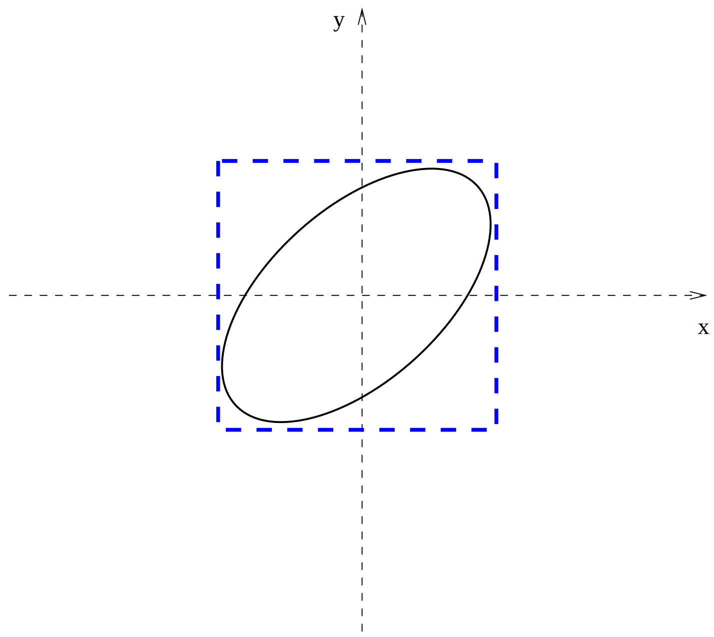
  and capture:

  - filter parameters: $(a, \varepsilon_a, b, \varepsilon_b)$;
  - variables tied before (X, Y) and after the iteration (Z, X),
  - an approximation of the current input: $\text{EVAL}^\sharp(E', \rho^\sharp)$.

Approximated reduced product



Initial conditions



Output refinement

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Higher order simplified filters

A simplified filter of class (k, l) is defined as a sequence:

$$S_{n+p} = a_1.S_n + \ldots + a_p.S_{n+p-1} + E_{n+p},$$

where the polynomial $P = X^p - a_p.X^{p-1} - \ldots - a_1.X^0$ has no multiple roots (in $\mathbb{C}$) and can be factored into the product of k second order irreducible polynomials $X^2 - \alpha_i.X - \beta_i$ and l first order polynomials $X - \delta_j$.
Then, there exists sequences $(x_n^i)_{n\in\mathbb{N}}$ and $(y_n^j)_{n\in\mathbb{N}}$ such that:

$$\begin{cases} S_n = (\sum\limits_{1\leq i\leq k} x_n^i) + (\sum\limits_{1\leq j\leq l} y_n^j) \\ x_{n+2}^i = \alpha_i.x_{n+1}^i + \beta_i.x_n^i + F^i(E_{n+2}, E_{n+1}) \\ y_{n+1}^j = \delta_j.y_n^j + G^j(E_{n+1}). \end{cases}$$

The initial outputs (x_0^i, x_1^i, y_0^j) and filter inputs F^i, G^j are given by solving symbolic linear systems, they only depend on the roots of P.

Higher order simplified filters

Soundness of the factoring algorithm into irreducible polynomials is not required.

Whenever we meet a higher order filter assignment τ,

1. we compute the characteristic polynomial P,
2. we compute a potentially unsound factoring P' of P,
3. we expand P',
4. we consider the filter assignment τ' such that the characteristic polynomial of τ' is P',
5. we bound the difference between τ and τ'
   (by using symbolic computation),
6. we integrate this bound into the input stream.

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Other filters

We have:

$$\begin{cases} S_k = i_k, 0 \leq k < p \\ S_{n+p} = \overline{F}(S_n, \ldots, S_{n+p-1}) \mp \overline{G}(E_{n+p+1-q}, \ldots, E_{n+p}) \end{cases}$$

Having bounds:

- on the input sequence (E_n),
- and on the initial outputs $(i_k)_{0 \leq k < p}$;

we want to infer a bound on the output sequence (S_n).

Splitting S_n

We split the output sequence $S_n = R_n + \varepsilon_n$ into

- the contribution of the errors (ε_n);

$$\begin{cases} \varepsilon_k = 0, 0 \leq k < p; \\ \varepsilon_{n+p} = F(\varepsilon_n, \ldots, \varepsilon_{n+p-1}) + \textit{err}_{n+p} \end{cases}$$

  we can use the simplified filter domain to limit (ε_n).

- the ideal sequence (R_n) (in the real field);

$$\begin{cases} R_k = i_k,\ 0 \leq k < p \\ R_{n+p} = F(R_n, \ldots, R_{n+p-1}) + G(E_{n+p+1-q}, \ldots, E_{n+p}) \end{cases}$$

Bounding R_n

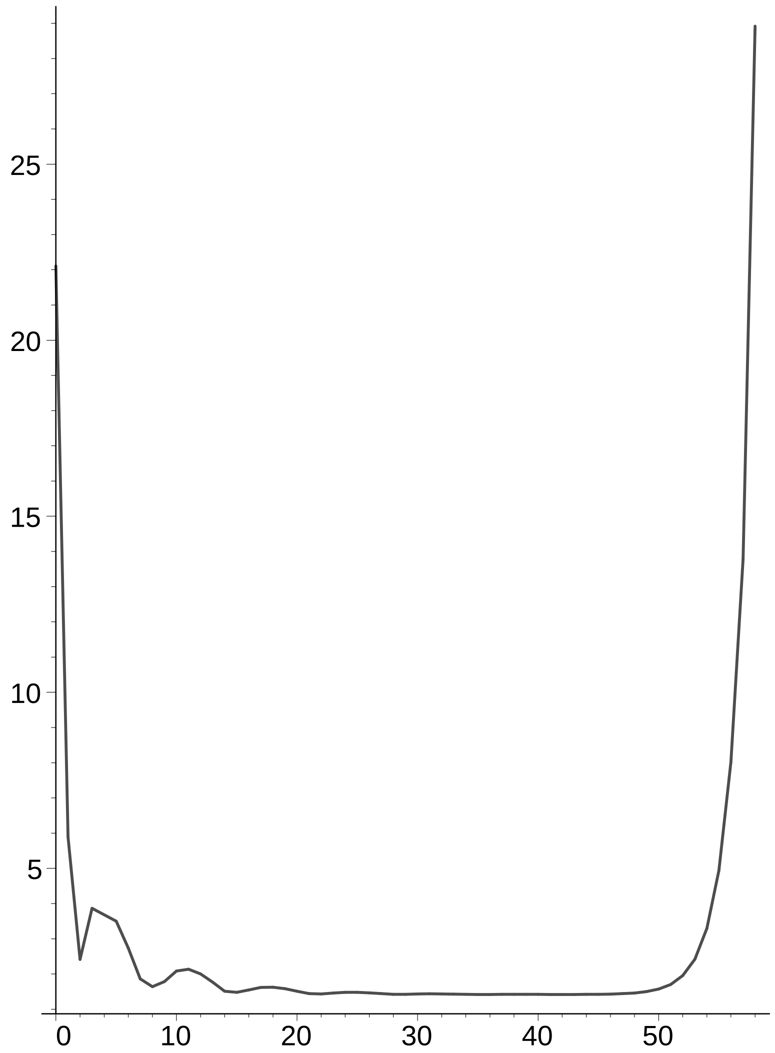
To refine the output, we need to bound the sequence R_n:

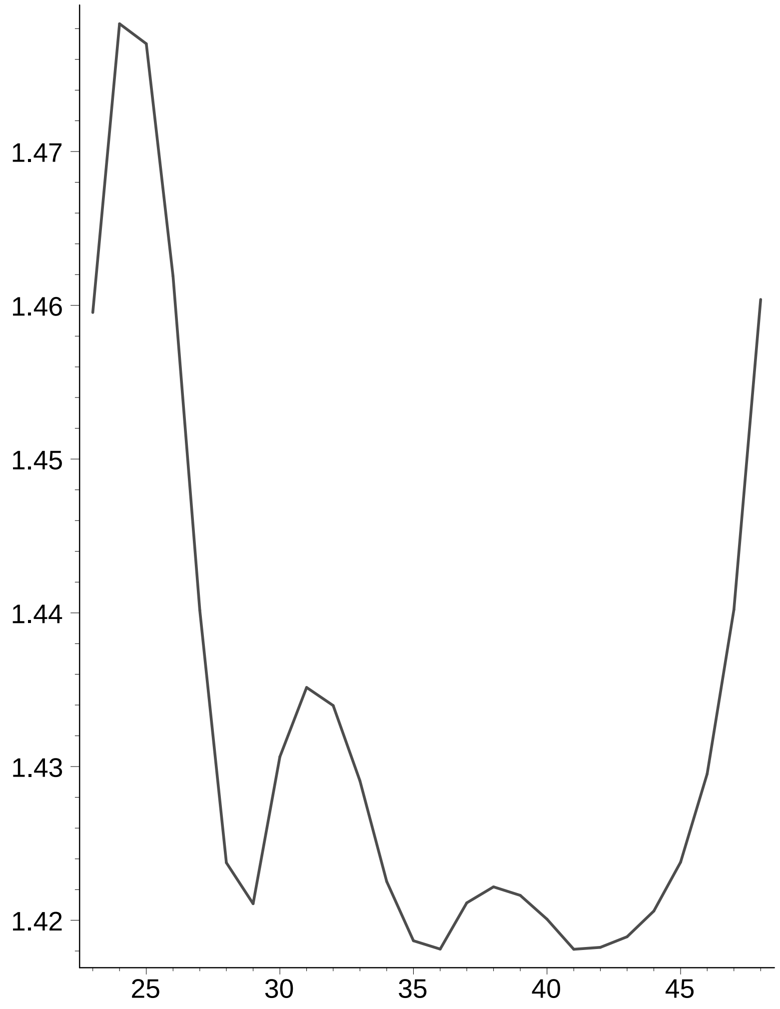
1. We isolate the contribution of the N last inputs:

$$R_n = \textit{last}_n^N(E_n, \ldots, E_{n+1-N}) + \textit{res}_n^N.$$

2. Since the filter is linear, we have, for $n > N + p$:
   - $\textit{last}_n^N = \textit{last}_{N+p}^N$;
   - $\textit{res}_n^N$ can be limited by using the corresponding simplified filter domain.

Abstract gain with respect to N





Overview

Benchmarks

We analyze three programs in the same family on a AMD Opteron 248, 8 Gb of RAM (analyses use only 2 Gb of RAM).

lines of C	70,000			216,000			379,000		
global variables	13,400			7,500			9,000		
iterations	72	41	**37**	161	75	**53**	151	187	**74**
time/iteration	52s	1mn18s	**1mn16s**	3mn07s	5mn08s	**4mn40s**	4mn35s	9mn25s	**8mn17s**
analysis time	1h02mn	53mn	**47mn**	8h23mn	6h25mn	**4h08mn**	11h34mn	30h26mn	**10h14mn**
false alarms	574	3	**0**	207	0	**0**	790	0	**0**

1. without filter domains;
2. with simplified filter domains;
3. with expanded filter domains.

Conclusion

- a highly generic framework to analyze programs with digital filtering:

  a technical knowledge of used filters allows the design of the adequate abstract domain;

- the case of linear filters is fully handled:

  We need to solve a symbolic linear system for each filter family. We need an unsound polynomial reduction algorithm for each filter instance.

- filter detection is left as a parameter:
  - term rebuilding can be used [Miné:VMCAI 2006];

This framework has been used and was necessary in the full certification of the absence of runtime error in industrial critical embedded software.

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