

NSAD'05

**Numerical Abstract Domains  
for Digital Filters**

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# Overview

1. Introduction
2. Case study
3. Generic framework
4. Simplified filters
5. Expanded filters
6. Conclusion

# Context

- proving the absence of run time error in critical embedded software.

Filter behavior 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 seek precise bounds on the output, using information inferred about the input. (Linear invariants do not yield accurate bounds).
- **Floating-point arithmetics:**  
(in the concrete semantics and when implementing the abstract domain).

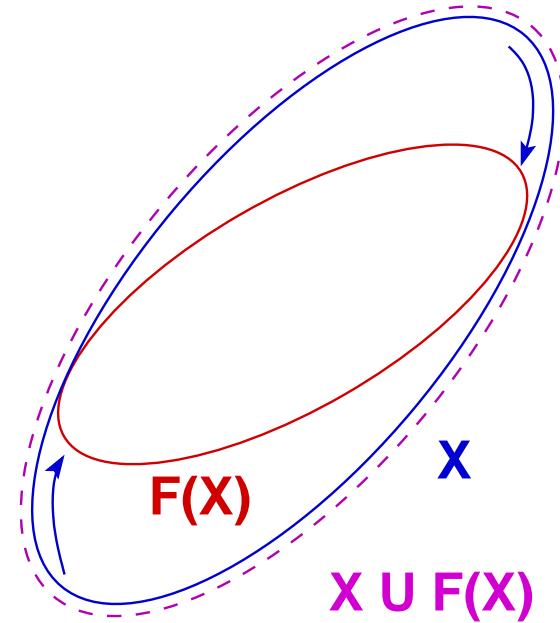
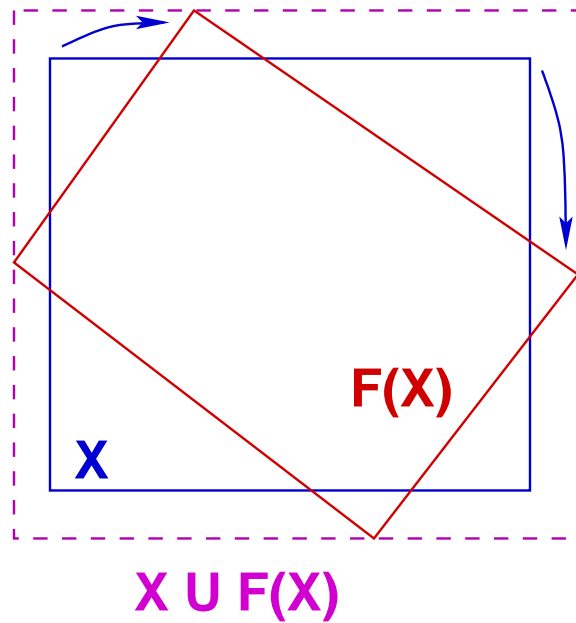
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# The second order filter (simplified)

```
V ∈ ℝ;  
E := 0; S0 := 0; S1 := 0; S2 := 0;  
while (V ≥ 0) {  
    V ∈ ℝ; T ∈ ℝ;  
    E ∈ [-1; 1];  
    if (T ≥ 0) {S0 := E; S1 := E;}  
    else {S0 := 1.5 × S1 - 0.7 × S2 + E;}  
    S2 := S1; S1 := S0;  
}
```

# Linear versus quadratic invariants



# Ellipsoidal constraints

## Theorem 1 (second order filter (simplified))

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.  $Z^2 - aZX - bX^2 \leq (\sqrt{-bK} + m)^2;$

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



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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 a second order filter using the last three inputs:

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

# 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  $\wp(\mathcal{V} \rightarrow \mathbb{R})$ , by a concretization function:

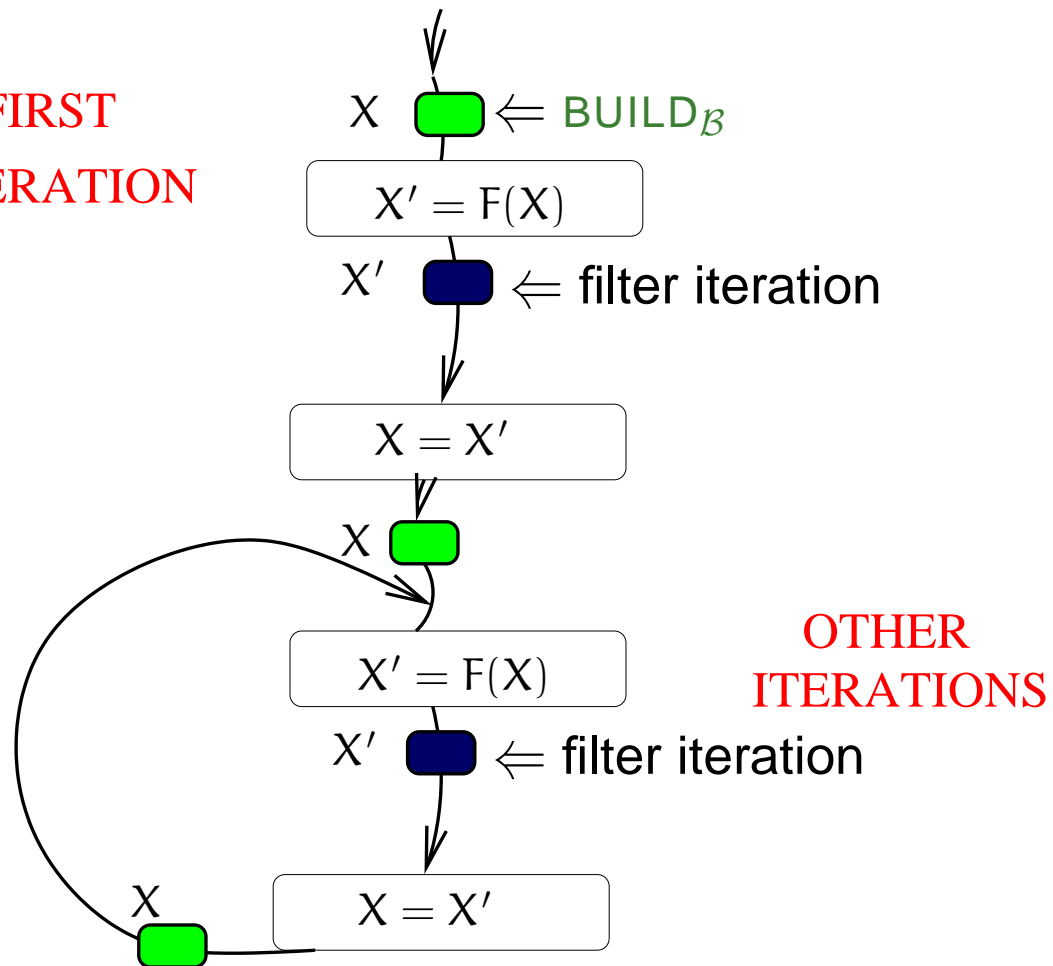
$$\gamma_{\mathcal{B}} : \mathcal{T}_{\mathcal{B}} \times \mathcal{B} \rightarrow \wp(\mathcal{V} \rightarrow \mathbb{R}).$$

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.

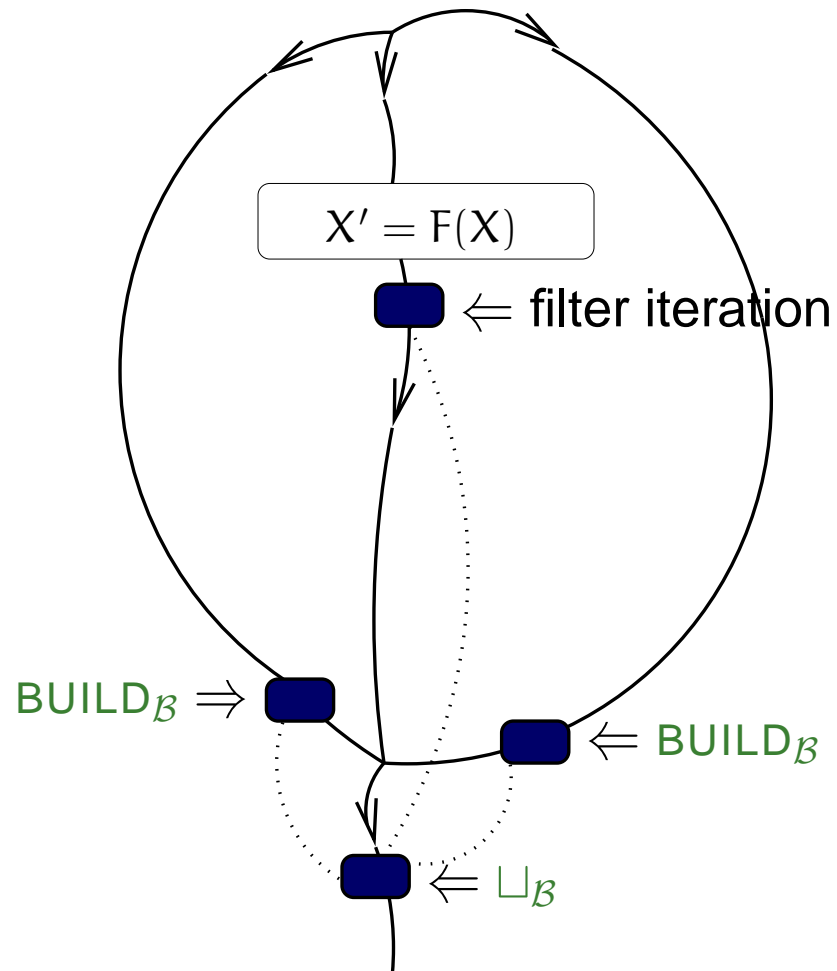
# Assignment

FIRST  
ITERATION



OTHER  
ITERATIONS

# Merging computation paths



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# Simplified first order filter

## Theorem 2 (Including rounding errors)

Let  $\alpha, \varepsilon_\alpha \geq 0, D \geq 0, m \geq 0, X$  and  $Z$  be real numbers such that:

1.  $|X| \leq D$ ;
2.  $\alpha X - (m + \varepsilon_\alpha |X|) \leq Z \leq \alpha X + (m + \varepsilon_\alpha |X|)$ .

We have:

- $|Z| \leq (|\alpha| + \varepsilon_\alpha)D + m$ ;

- $\left[ |\alpha| + \varepsilon_\alpha < 1 \text{ and } D \geq \frac{m}{1 - (|\alpha| + \varepsilon_\alpha)} \right] \implies |Z| \leq D.$

□

# Simplified second order filter

## Theorem 3 (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

$$1. Z^2 - aZX - bX^2 \leq ((\sqrt{-b} + \delta)\sqrt{K} + m)^2;$$

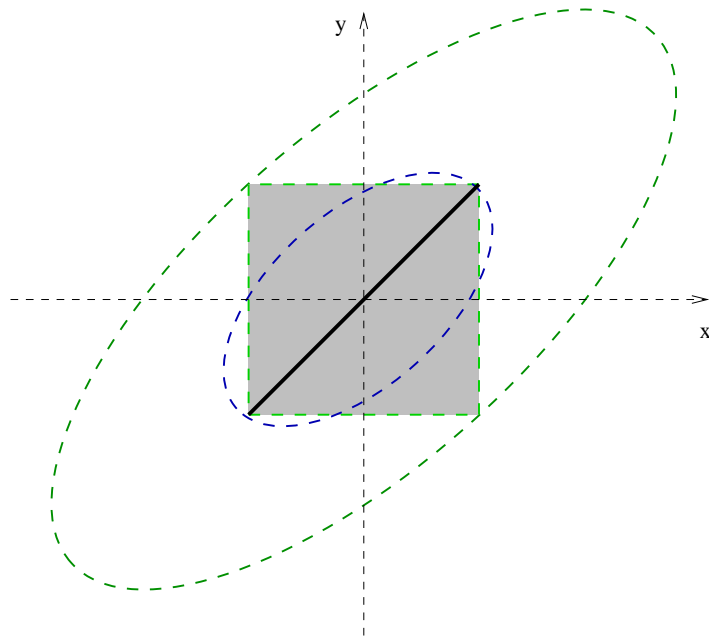
$$2. \begin{cases} \sqrt{-b} + \delta < 1 \\ K \geq \left( \frac{m}{1 - \sqrt{-b} - \delta} \right)^2 \end{cases} \implies Z^2 - aZX - bX^2 \leq K,$$

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

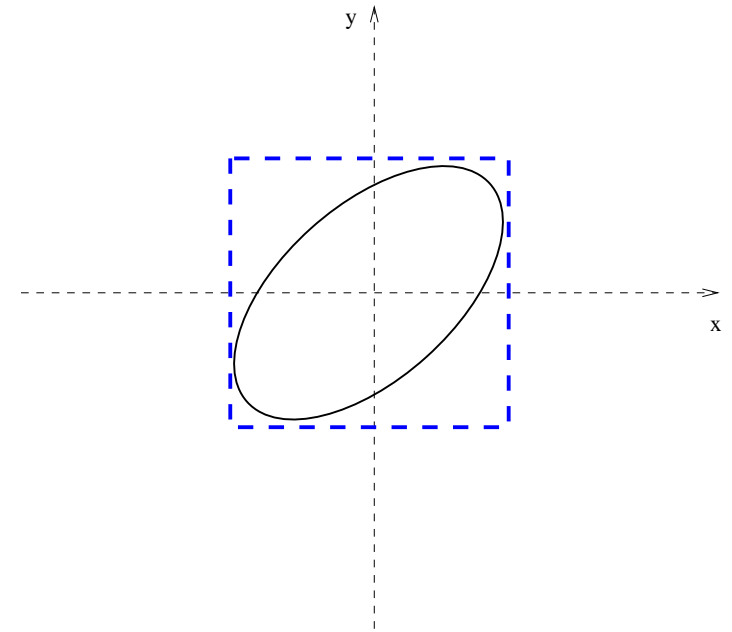
□



# Reduced product



Initial conditions



Output refinement

# Higher order simplified filters

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

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

where the polynomial  $P = X^p - a_p.X^{p-1} - \dots - a_1.X^0$  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:

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

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  $\tau$ ,

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  $\tau'$  such that the characteristic polynomial of  $\tau'$  is  $P'$ ,
5. we bound the difference between  $\tau$  and  $\tau'$  (by using symbolic computation),
6. we integrate this bound into the input stream.

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# Other filters

We have:

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

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  $(\varepsilon_n)$ ;

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

we can **use the simplified filter domain** to limit  $(\varepsilon_n)$ .

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

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

# Limiting $R_n$

To refine the output, we need to limit the sequence  $R_n$ :

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

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

2. Since the filter is linear, we have, for  $n > N + p$ :

- $\mathit{last}_n^N = \mathit{last}_{N+p}^N$ ;
- $\mathit{res}_n^N$  can be limited by using the corresponding simplified filter domain.

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

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 [MinéPhD];

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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