

Towards reliable implementation of Digital Filters

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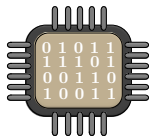
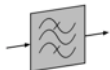
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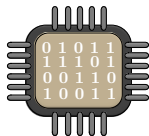
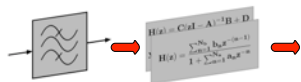
April 5, 2016



Motivation



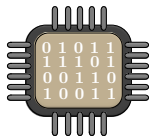
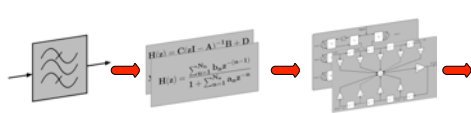
Motivation



Filter implementation flow:

- Transfer function generation

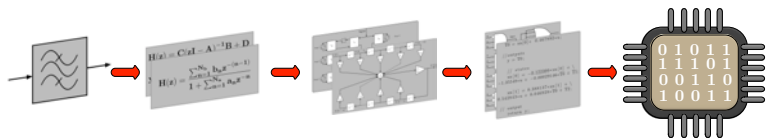
Motivation



Filter implementation flow:

- Transfer function generation
- State-space, DFI, DFII, ...

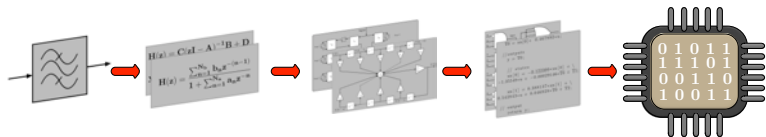
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Filter implementation flow:

- Transfer function generation
- State-space, DFI, DFII, ...
- Software or Hardware implementation

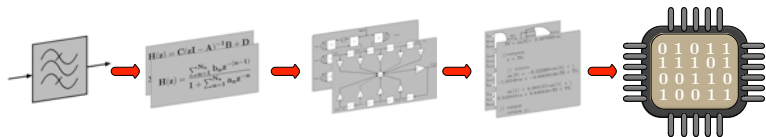
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Filter implementation flow:

- Transfer function generation
 - ! Coefficient quantization
- State-space, DFI, DFII, ...
- Software or Hardware implementation

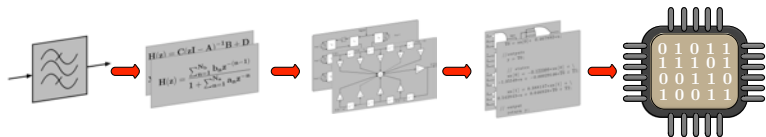
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Filter implementation flow:

- Transfer function generation
 - ! Coefficient quantization
- State-space, DFI, DFII, ...
 - ! Large variety of structures with no common quality criteria
- Software or Hardware implementation

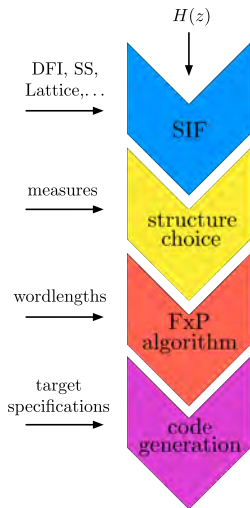
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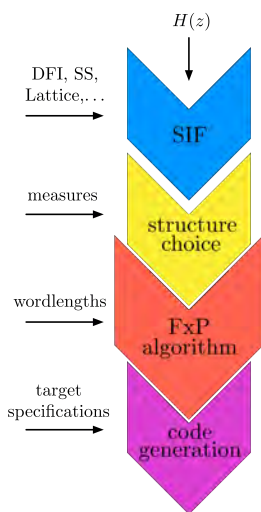
Filter implementation flow:

- Transfer function generation
 - ! Coefficient quantization
- State-space, DFI, DFII, ...
 - ! Large variety of structures with no common quality criteria
- Software or Hardware implementation
 - ! Constraints: power consumption, area, error, speed, etc.

Motivation: Automated filter implementation flow



Motivation: Automated filter implementation flow



Focus on Fixed-Point realization.

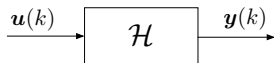
Optimization of wordlengths:

- Take more - pay more
- Take less - overflow risk

What we want in the end:

- Rigorous algorithm for Fixed-Point Formats (FxPF) determination
- No simulations but mathematical proofs
- Integration into automated code generator for filters

LTI filters



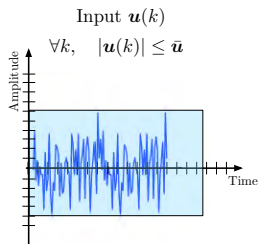
$\mathcal{H} := (\mathbf{A}, \mathbf{B}, \mathbf{C}, \mathbf{D})$ is a Bounded Input Bounded Output LTI filter in state-space representation:

$$\mathcal{H} \begin{cases} \mathbf{x}(k+1) & = \mathbf{A}\mathbf{x}(k) + \mathbf{B}\mathbf{u}(k) \\ \mathbf{y}(k) & = \mathbf{C}\mathbf{x}(k) + \mathbf{D}\mathbf{u}(k) \end{cases}$$

with q inputs, n states and p outputs and state matrices

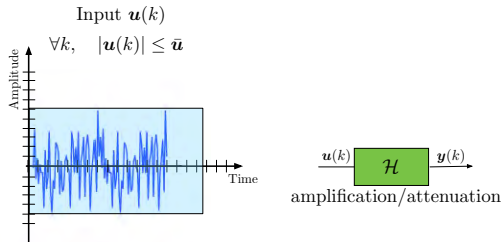
Basic brick: interval through filter

The Worst-Case Peak Gain theorem



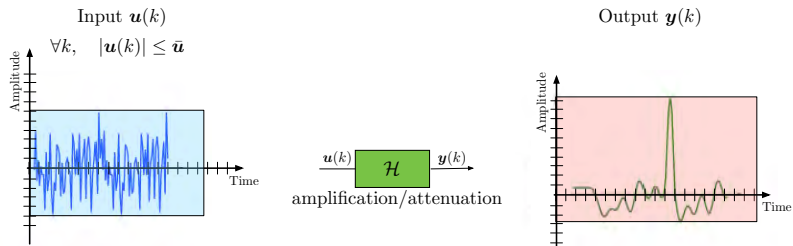
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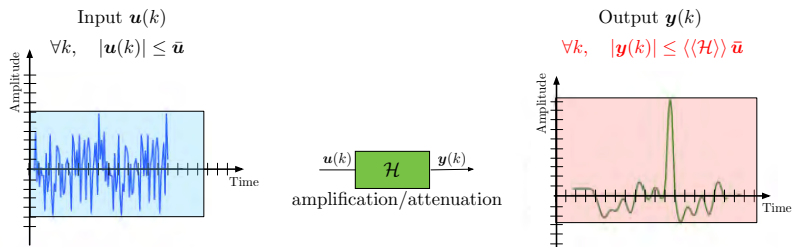
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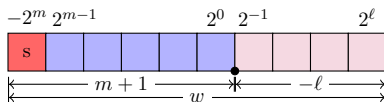
The Worst-Case Peak Gain theorem



Worst-Case Peak Gain

$$\langle\langle \mathcal{H} \rangle\rangle = |\mathbf{D}| + \sum_{k=0}^{\infty} |\mathbf{C}\mathbf{A}^k\mathbf{B}|$$

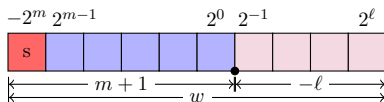
Two's complement Fixed-Point arithmetic



$$t = -2^m t_m + \sum_{i=\ell}^{m-1} 2^i t_i$$

- Wordlength: w
- Most Significant Bit position: m
- Least Significant Bit position: $\ell := m - w + 1$

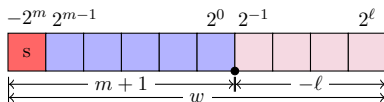
Two's complement Fixed-Point arithmetic



$$t = -2^m t_m + \sum_{i=\ell}^{m-1} 2^i t_i$$

- quantization step: 2^ℓ
- t represented by integer $T = t \cdot 2^{-\ell}$
- $T \in [-2^m; 2^m - 2^\ell] \cap \mathbb{Z}$

Two's complement Fixed-Point arithmetic



$$t = -2^m t_m + \sum_{i=\ell}^{m-1} 2^i t_i$$

- $y(k) \in \mathbb{R}$
- wordlength w bits
- minimal Fixed-Point Format (FPF) is the least m :

$$\forall k, \quad y(k) \in [-2^m; 2^m - 2^{m-w+1}]$$

Taking the quantization errors into account

The exact filter \mathcal{H} is:

$$\mathcal{H}^\diamond \begin{cases} \mathbf{x}(k+1) = \mathbf{A}\mathbf{x}(k) + \mathbf{B}\mathbf{u}(k) \\ \mathbf{y}(k) = \mathbf{C}\mathbf{x}(k) + \mathbf{D}\mathbf{u}(k) \end{cases}$$

Taking the quantization errors into account

The actually implemented filter \mathcal{H}^\diamond is:

$$\mathcal{H}^\diamond \begin{cases} \mathbf{x}^\diamond(k+1) &= \mathbf{A}\mathbf{x}^\diamond(k) + \mathbf{B}\mathbf{u}(k) + \boldsymbol{\varepsilon}_x(k) \\ \mathbf{y}^\diamond(k) &= \mathbf{C}\mathbf{x}^\diamond(k) + \mathbf{D}\mathbf{u}(k) + \boldsymbol{\varepsilon}_y(k) \end{cases}$$

with

$$|\boldsymbol{\varepsilon}_x(k)| \leq 2^{m_x - w_x + 1} \quad \text{and} \quad |\boldsymbol{\varepsilon}_y(k)| \leq 2^{m_y - w_y + 1}.$$

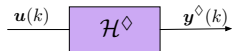
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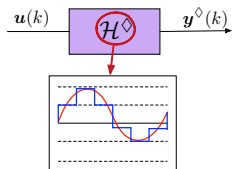
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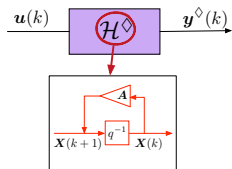
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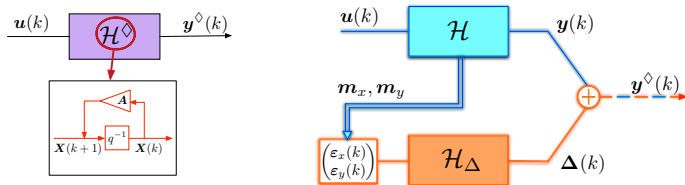
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Computing the MSB using the WCPG theorem

Applying the WCPG theorem on filter \mathcal{H} gives

$$\forall k, \quad |\mathbf{y}_i(k)| \leq (\langle\langle \mathcal{H} \rangle\rangle \bar{\mathbf{u}})_i.$$

We can compute the MSB positions with

$$\mathbf{m}_{y_i} = \lceil \log_2 (\langle\langle \mathcal{H} \rangle\rangle \bar{\mathbf{u}})_i - \log_2 (1 - 2^{1-w_{y_i}}) \rceil.$$

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We can compute the MSB positions with

$$\begin{aligned} \widehat{\mathbf{m}}_{y_i} &= \left[\log_2 (\langle\langle \mathcal{H} \rangle\rangle \bar{\mathbf{u}})_i - \log_2 (1 - 2^{1-w_{y_i}}) + \log_2 \left(1 + \frac{(\varepsilon_{WCPG \mathbf{u}})_i}{(\langle\langle \mathcal{H} \rangle\rangle \bar{\mathbf{u}})_i} \right) \right] \\ &= \mathbf{m}_{y_i} + [\dots] \end{aligned}$$

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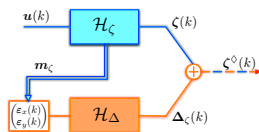
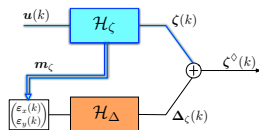
Adjust error term ε_{WCPG} in order to be at most **off by one**.

Algorithm

Step 1 Determine the MSBs m_y for the exact filter \mathcal{H} , applying the WCPG theorem;

Step 2 Compute the error-filter \mathcal{H}_Δ , induced by the format m_y and deduce the MSB m_y^\diamond of the implemented filter;

Step 3 If $m_{y_i}^\diamond == m_{y_i}$ then return $m_{y_i}^\diamond$ otherwise $m_{y_i} \leftarrow m_{y_i} + 1$ and go to Step 2.



Numerical examples

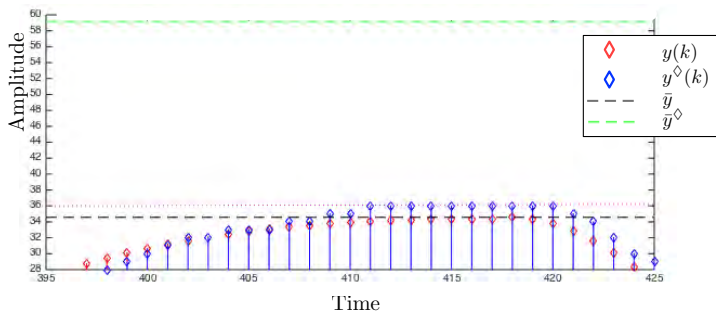
Example:

- Random filter with 3 states, 1 input, 1 output
- $\bar{u} = 5.125$, wordlengths set to 7 bits

	states			output
	$\mathbf{x}_1(k)$	$\mathbf{x}_2(k)$	$\mathbf{x}_3(k)$	$\mathbf{y}(k)$
Step 1	6	7	5	6
Step 2	6	7	6	6
Step 3	6	7	6	6

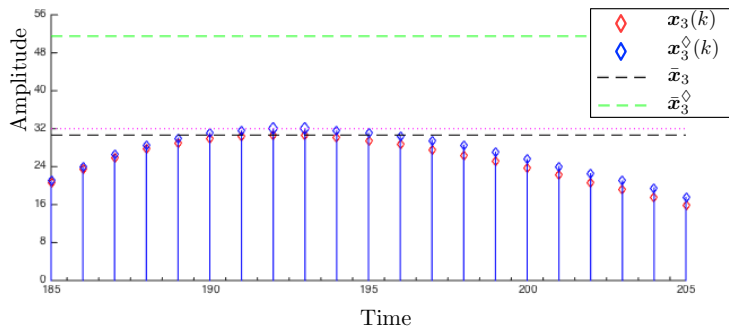
Table Evolution of the MSB positions

Numerical examples



The exact and quantized outputs of the example. Quantized output does not pass over to the next binade.

Numerical examples



The exact and quantized third state of the example. Quantized output passes over to the next binade.

Conclusion and Perspectives

Conclusion

- Computing WCPG in arbitrary precision
- Rigorous procedure for Fixed-Point Formats determination
- Filter computation errors are taken into account, ensuring that no overflow occurs
- Do not use any (non-exhaustive) simulations but proofs.

Perspectives

- Integrate into optimization procedures in automatic workflow
- Solve off-by-one problem for the MSB computation (e.g. using linear integer programming, lattices)
- Compute WCPG for interval matrices (requires interval solver for Lyapunov equations)