

# Fixed-Point implementation of Lattice Wave Digital Filter: comparison and error analysis

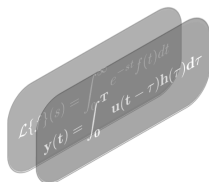
Anastasia Volkova, Thibault Hilaire

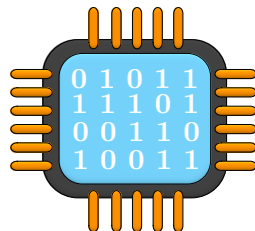
Sorbonne Universités, University Pierre and Marie Curie, LIP6,  
Paris, France

EUSIPCO 15  
September 2, 2015



# Motivation


$$\mathcal{L}\{y(t)\}(s) = \int_0^\infty e^{-st} \int_0^t u(t-\tau) h(\tau) d\tau dt$$
$$y(t) = \int_0^t u(t-\tau) h(\tau) d\tau$$



## Need to deal with

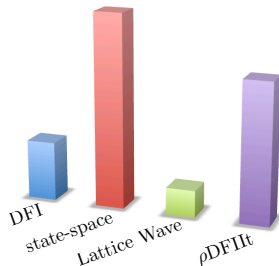
- Discretize functions and coefficients
  - parametric errors
  - computational errors
- Implementation under constraints
  - software implementation
  - hardware implementation

# Motivation

Different filter structures:

- Direct Form I, Direct Form II
- State-space
- Wave, Lattice Wave, ...
- $\rho$ -operator:  $\rho$ DFIIt,  $\rho$ State-space...
- LGS, LCW, etc.

Number of coefficients



## Problem:

They are equivalent in *infinite* precision but no more in *finite* precision. The finite precision degradation depends on the realization.

# Motivation

Given transfer function and a target, we want:

- Represent various realizations (in an easy way)
- Evaluate finite precision degradation (a priori/a posteriori)
- Find an optimal realization (need to compare realizations)

Tradeoff:

- |                     |   |                     |
|---------------------|---|---------------------|
| • Error             | } | w.r.t. exact filter |
| • Quality           |   |                     |
| • Power consumption | } | resources           |
| • Area              |   |                     |
| • Speed             |   |                     |

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Specialized Implicit Framework (SIF)

# Outline

- 1 Motivation
- 2 Specialized Implicit Framework
- 3 Lattice Wave Digital Filters
- 4 LWDF-to-SIF conversion
- 5 Example and comparison
- 6 Summary

# SIF: Specialized Implicit Framework

SIF is:

- Macroscopic description
- Based on state-space
- Explicit all the computations and their order
- Any DFG can be transformed to this form
- Analytical derivation of measures

$$\mathcal{H} \begin{cases} \mathbf{J}\mathbf{t}(k+1) = \mathbf{M}\mathbf{x}(k) + \mathbf{N}u(k) \\ \mathbf{x}(k+1) = \mathbf{K}\mathbf{t}(k+1) + \mathbf{P}\mathbf{x}(k) + \mathbf{Q}u(k) \\ \mathbf{y}(k) = \mathbf{L}\mathbf{t}(k+1) + \mathbf{R}\mathbf{x}(k) + \mathbf{S}u(k) \end{cases}$$

Denote  $\mathbf{Z}$  the matrix containing  
all the coefficients

$$\mathbf{Z} \triangleq \begin{pmatrix} -\mathbf{J} & \mathbf{M} & \mathbf{N} \\ \mathbf{K} & \mathbf{P} & \mathbf{Q} \\ \mathbf{L} & \mathbf{R} & \mathbf{S} \end{pmatrix}$$

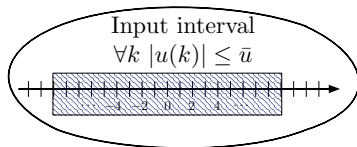
# SIF: measures

## Measures

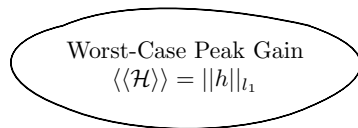
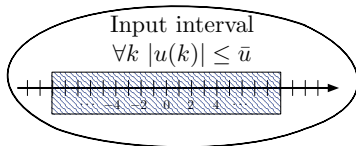
- *a priori* measures
  - transfer function sensitivity (based on  $\frac{\partial H}{\partial \mathbf{Z}}$ )
    - stochastic measure, takes into account coefficient wordlengths
  - poles or zeros sensitivity (e.g based on  $\frac{\partial |\lambda_i|}{\partial \mathbf{Z}}$  for a pole  $\lambda_i$ )
    - stochastic measure, takes into account coefficient wordlengths
  - RNG, ...
- *a posteriori* measures
  - Signal to Quantization Noise Ratio
  - **output error**



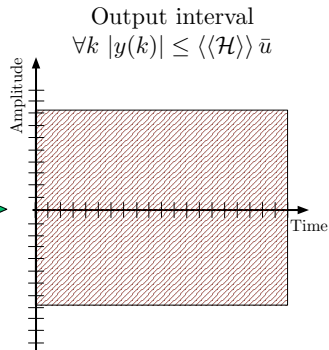
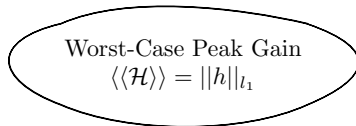
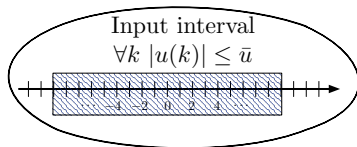
# SIF: Worst-Case Peak Gain theorem



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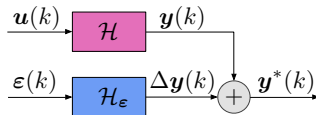
# SIF: Worst-Case Peak Gain theorem



# SIF: Worst-Case Peak Gain theorem

WCPG theorem permits to determine:

- the output error interval



- the Most Significant Bit, therefore Fixed-Point Formats

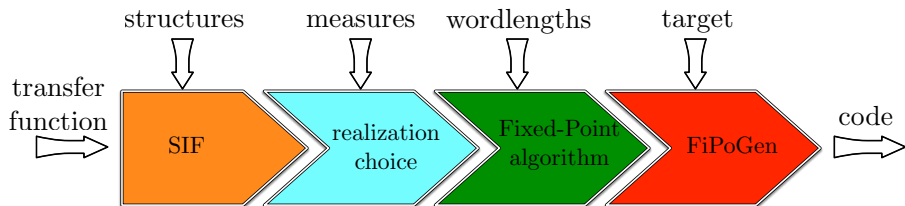
$$m_y = \lfloor \log_2 (\langle\langle \mathcal{H} \rangle\rangle \bar{u}) \rfloor + 1$$

**Equivalent technique:** WCPG-scaling, it guarantees that no overflows occur.

## Fixed Point Code Generator (FiPoGen)

- Generates bit-accurate fixed-point algorithms
- Optimizes the wordlength under certain criteria (e.g. area)

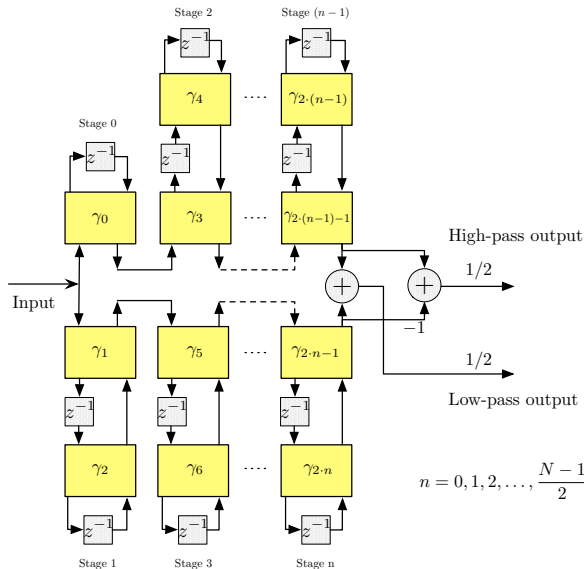
# SIF: from transfer function to Fixed-Point code



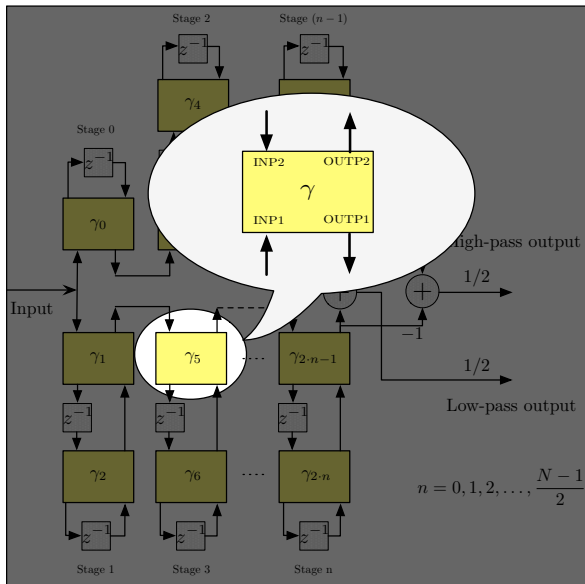
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# Lattice Wave Digital Filters



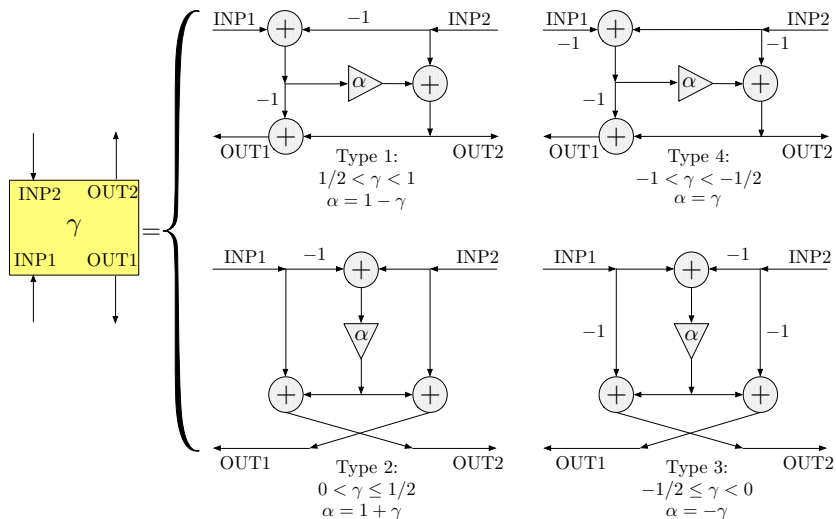
# Lattice Wave Digital Filters





# Lattice Wave Digital Filters

Two-port adaptor: Richard's structures



# Lattice Wave Digital Filters

## Positive sides

- parallelizable
- modular, convenient for VLSI
- often referred to as *stable*

## Drawbacks

- Studies of Fixed-Point implementation include complicated infinite-precision optimization
- Comparison is difficult

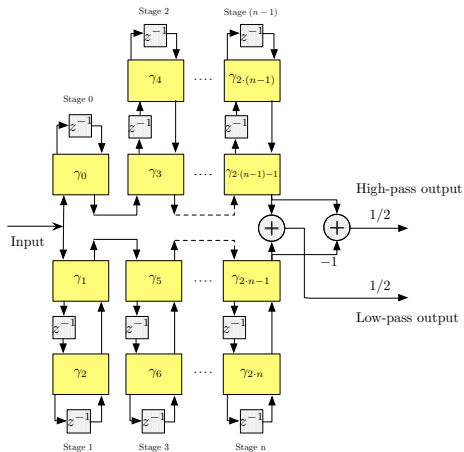
## Objectives

- Represent LWDF in terms of SIF
- Perform *rigorous* error analysis
- Instantly compare with other structures

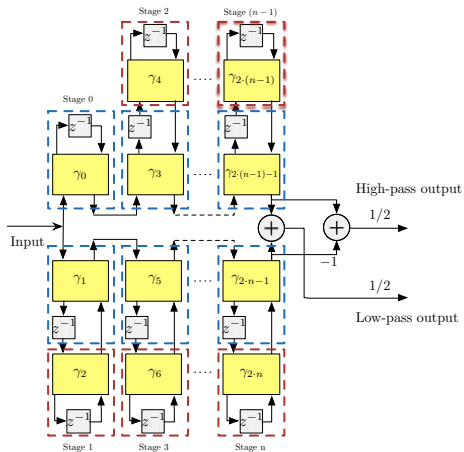
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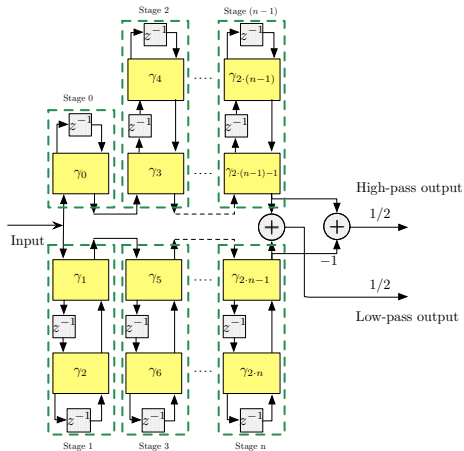
# LWDF-to-SIF conversion



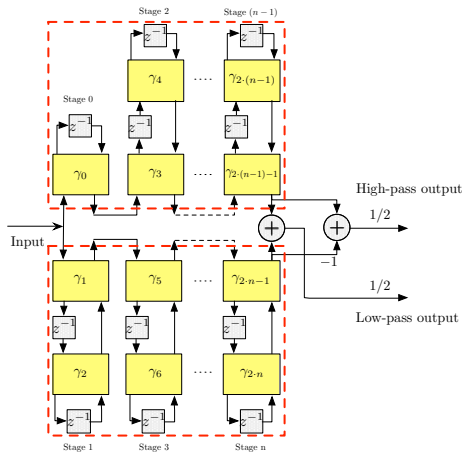
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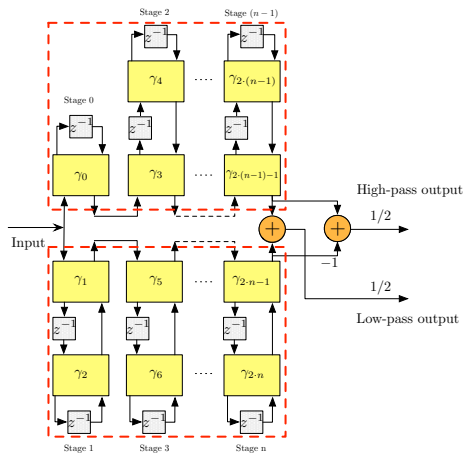
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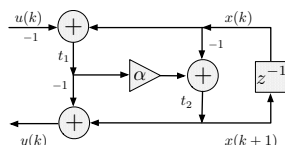
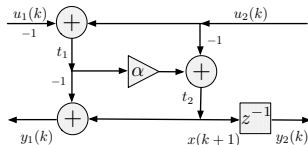
# LWDF-to-SIF conversion





# LWDF-to-SIF conversion: example

Convert DFGs of two adaptors into SIFs:

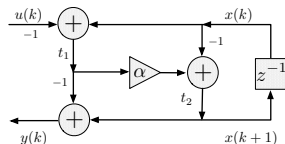
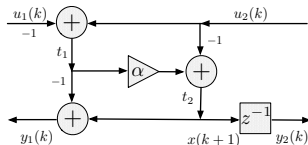


$$\left( \begin{array}{cc|cc} 1 & 0 & 0 & 0 \\ -\alpha & 1 & 0 & 0 \\ \hline 0 & 1 & 1 & 0 \\ 1 & -1 & 0 & 1 \\ \hline 0 & 0 & 0 & 1 \end{array} \right) \begin{pmatrix} \mathbf{t}(k+1) \\ \mathbf{x}(k+1) \\ \mathbf{y}(k) \end{pmatrix} = \left( \begin{array}{cc|cc} 0 & 0 & 0 & -1 \\ 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 1 & 0 \end{array} \right) \begin{pmatrix} \mathbf{t}(k) \\ \mathbf{x}(k) \\ \mathbf{u}(k) \end{pmatrix}$$

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# LWDF-to-SIF conversion: example

Convert DFGs of two adaptors into SIFs:



$$\mathbf{Z}_A \triangleq \begin{pmatrix} -\mathbf{J}_A & \mathbf{M}_A & \mathbf{N}_A \\ \mathbf{K}_A & \mathbf{P}_A & \mathbf{Q}_A \\ \mathbf{L}_A & \mathbf{R}_A & \mathbf{S}_A \end{pmatrix} = \begin{pmatrix} -1 & 0 & 0 & -1 & 1 \\ \alpha & -1 & 0 & 0 & -1 \\ 0 & -1 & 0 & 0 & 0 \\ -1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \end{pmatrix}$$

$$\mathbf{Z}_B \triangleq \begin{pmatrix} -\mathbf{J}_B & \mathbf{M}_B & \mathbf{N}_B \\ \mathbf{K}_B & \mathbf{P}_B & \mathbf{Q}_B \\ \mathbf{L}_B & \mathbf{R}_B & \mathbf{S}_B \end{pmatrix} = \begin{pmatrix} -1 & 0 & 1 & -1 \\ \alpha & -1 & 1 & 0 \\ 0 & -1 & 0 & 0 \\ -1 & 1 & 0 & 0 \end{pmatrix}$$

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## Example and comparison

Reference filter: low-pass 5<sup>th</sup> order Butterworth filter with cutoff frequency 0.1.

Structures for the comparison:

- LWDF
- state-space
- $\rho$ -Direct Form II transposed
- Direct Form I

Normalized (*i.e.* all coefficients have the same wordlength) measures:

- transfer function error:  $\bar{\sigma}_{\Delta H}^2$
- pole error:  $\bar{\sigma}_{\Delta|\lambda|}^2$
- output error:  $\overline{\Delta_y}$

# Example and comparison

<span style="color: red;">□</span> 1	<span style="color: magenta;">●</span> coefficient
<span style="color: blue;">□</span> -1	+ power of 2

$$\mathbf{Z} = \begin{pmatrix} \text{matrix of red and blue squares and magenta dots} \\ + \end{pmatrix}$$

LWDF,  $\mathbf{Z}$  is  $22 \times 22$

$$\mathbf{Z} = \begin{pmatrix} \text{matrix of red and blue squares and magenta dots} \end{pmatrix}$$

DFI,  $\mathbf{Z}$  is  $12 \times 12$

$$\mathbf{Z} = \begin{pmatrix} \text{matrix of magenta dots} \end{pmatrix}$$

State-Space,  $\mathbf{Z}$  is  $12 \times 12$

$$\mathbf{Z} = \begin{pmatrix} \text{matrix of red and blue squares, magenta dots, and plus signs} \end{pmatrix}$$

$\rho$ DFIIt,  $\mathbf{Z}$  is  $12 \times 12$

# Example and comparison

Realization	size(Z)	coeff.	$\bar{\sigma}_{\Delta H}^2$	$\bar{\sigma}_{\Delta \lambda}^2$	$\overline{\Delta_y}$
LWDF	$22 \times 22$	5	0.3151	0.56	122.9
state-space	$6 \times 6$	36	1.15	5.75	23.33
$\rho$ DFIIt	$11 \times 11$	11	0.09	0.45	94.3
DFI	$12 \times 12$	11	1.42e+6	-	7.961

# Conclusion and perspectives

## Conclusion:

- LWDF converted to SIF
- Normalized sensitivity and output error measures applied
- Comparison with several popular structures presented

## Perspectives:

- Use VHDL code generator (FloPoCo) to compare hardware implementations
- Apply  $\rho$ -operator to LWDF

Thank you!  
Questions?



## SIF: the rigorous filter error bound

Exact filter:

$$\mathcal{H} \left\{ \begin{array}{l} \mathbf{J} \mathbf{t} (k+1) = \mathbf{M} \mathbf{x} (k) + \mathbf{N} u(k) \\ \mathbf{x} (k+1) = \mathbf{K} \mathbf{t} (k+1) + \mathbf{P} \mathbf{x} (k) + \mathbf{Q} u(k) \\ \mathbf{y} (k) = \mathbf{L} \mathbf{t} (k+1) + \mathbf{R} \mathbf{x} (k) + \mathbf{S} u(k) \end{array} \right.$$

## SIF: the rigorous filter error bound

Implemented filter:

$$\mathcal{H}^* \begin{cases} \mathbf{J}\mathbf{t}^*(k+1) = \mathbf{M}\mathbf{x}^*(k) + \mathbf{N}u(k) + \boldsymbol{\varepsilon}_t(k) \\ \mathbf{x}^*(k+1) = \mathbf{K}\mathbf{t}^*(k+1) + \mathbf{P}\mathbf{x}^*(k) + \mathbf{Q}u(k) + \boldsymbol{\varepsilon}_x(k) \\ \mathbf{y}^*(k) = \mathbf{L}\mathbf{t}^*(k+1) + \mathbf{R}\mathbf{x}^*(k) + \mathbf{S}u(k) + \boldsymbol{\varepsilon}_y(k) \end{cases}$$

where  $\boldsymbol{\varepsilon}_t(k)$ ,  $\boldsymbol{\varepsilon}_x(k)$  and  $\boldsymbol{\varepsilon}_y(k)$  are the computational errors.

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The output error

$$\Delta\mathbf{y}(k) \triangleq \mathbf{y}^*(k) - \mathbf{y}(k)$$

can be seen as the output of a MIMO filter  $\mathcal{H}_\varepsilon$ .

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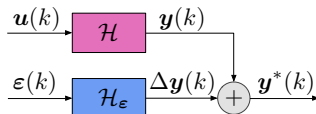
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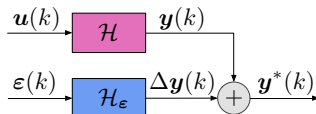
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WCPG theorem on  $\mathcal{H}_\varepsilon$  gives the output error interval.