



Institut des
Nanotechnologies
de Lyon UMR 5270



INSA



Signed Convolution in Photonics with Phase-Change Materials using Mixed-Polarity Bitstreams

*Raphael Cardoso, C Zrounba, M Abdalla, P Jimenez,
MG de Queiroz, B Charbonnier, F Pavanello,
I O'Connor, S Le Beux*





Institut des
Nanotechnologies
de Lyon UMR 5270

Application and platform

Signed Convolution in Photonics



*Raphael Cardoso, C Zrounba, M Abdalla, P Jimenez,
MG de Queiroz, B Charbonnier, F Pavanello,
I O'Connor, S Le Beux*





Institut des
Nanotechnologies
de Lyon UMR 5270

Specific material

with Phase-Change Materials



*Raphael Cardoso, C Zrounba, M Abdalla, P Jimenez,
MG de Queiroz, B Charbonnier, F Pavanello,
I O'Connor, S Le Beux*





Institut des
Nanotechnologies
de Lyon UMR 5270

Mathematical technique

using Mixed-Polarity Bitstreams

*Raphael Cardoso, C Zrounba, M Abdalla, P Jimenez,
MG de Queiroz, B Charbonnier, F Pavanello,
I O'Connor, S Le Beux*

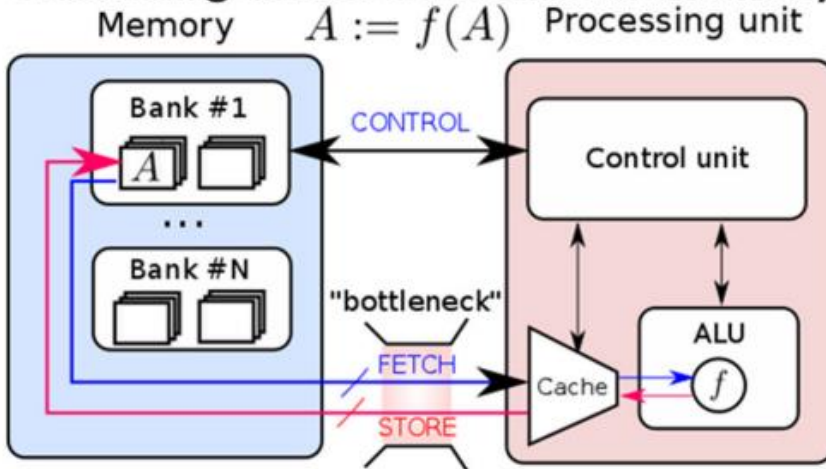


Summary

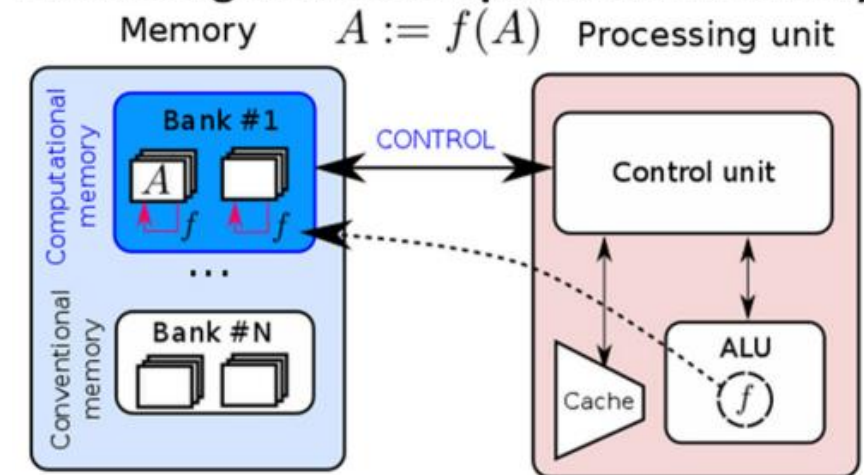
- *Motivation*
- *Photonic computing cells*
- *Proposed circuits*
- *Simulation results*
- *Conclusions and challenges*

An architectural paradigm shift

Processing unit & Conventional memory



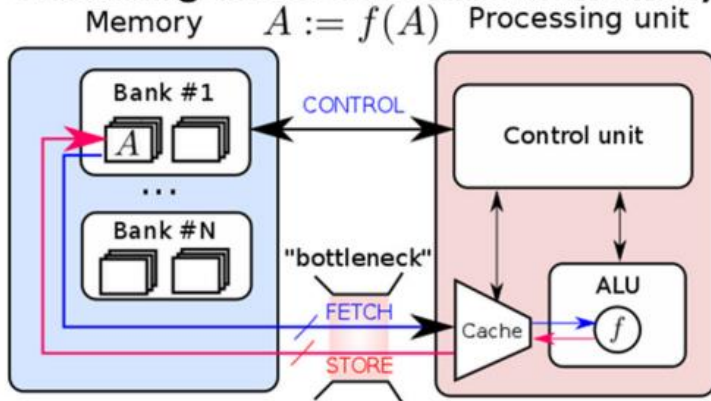
Processing unit & Computational memory



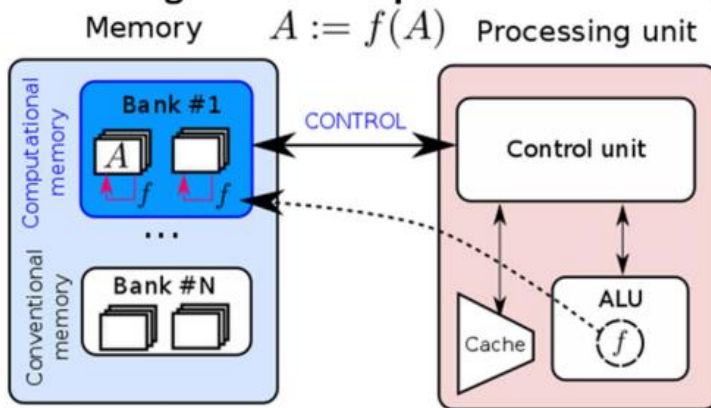
Sebastian et al., Tutorial: Brain-inspired computing using phase-change memory devices, 2018

Operations in neural networks

Processing unit & Conventional memory

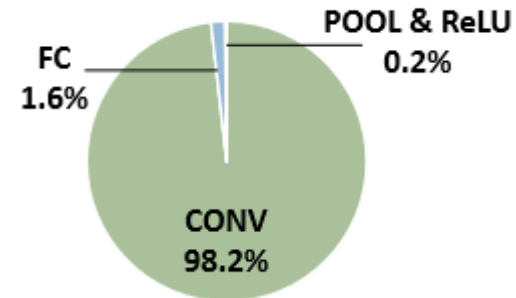


Processing unit & Computational memory



Sebastian A. et al., Tutorial: Brain-inspired computing using phase-change memory devices, 2018

operations of VGG-11



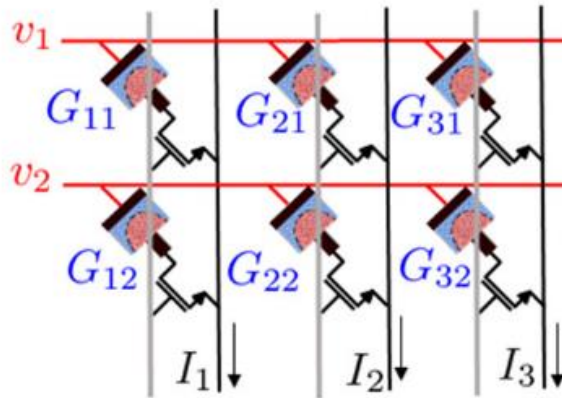
Guo et al., A Survey of FPGA-Based Neural Network Inference Accelerator, 2018

$$(f * g)(t) := \int_{-\infty}^{\infty} f(\tau)g(t - \tau) d\tau.$$

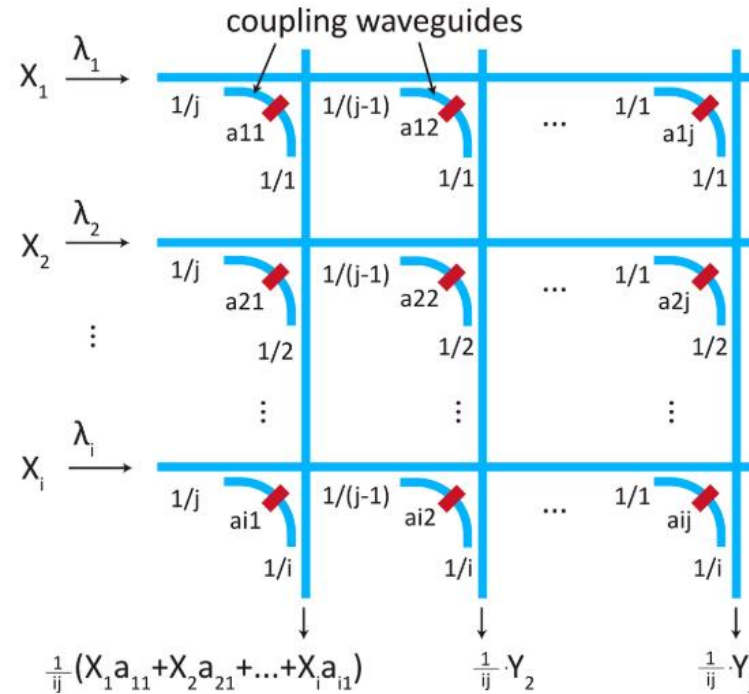
accumulate multiply
(MAC)

MAC crossbars in memory

$$\mathbf{a} \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \\ A_{31} & A_{32} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

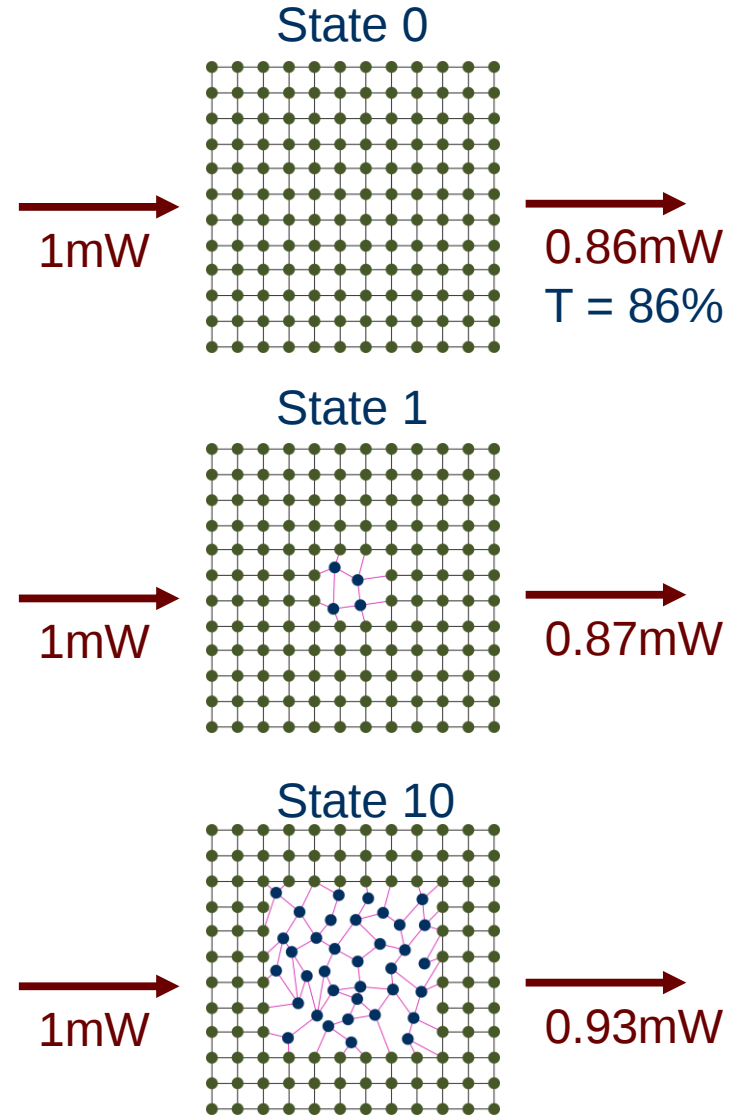
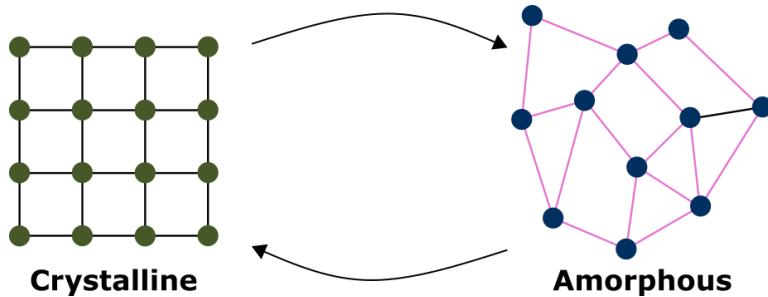
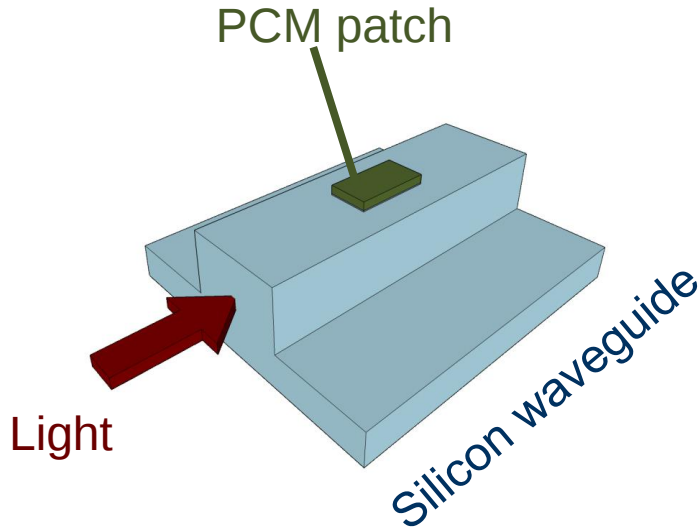


Sebastian et al., Computational phase-change memory: beyond von Neumann computing, 2019

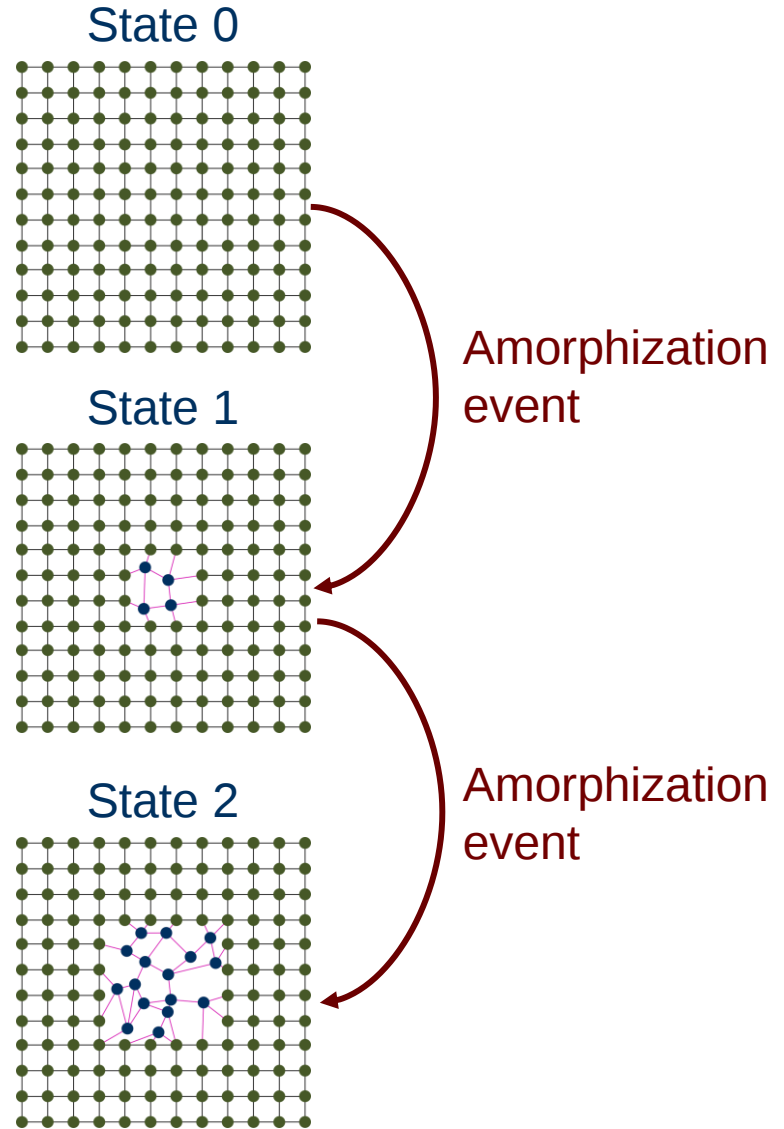
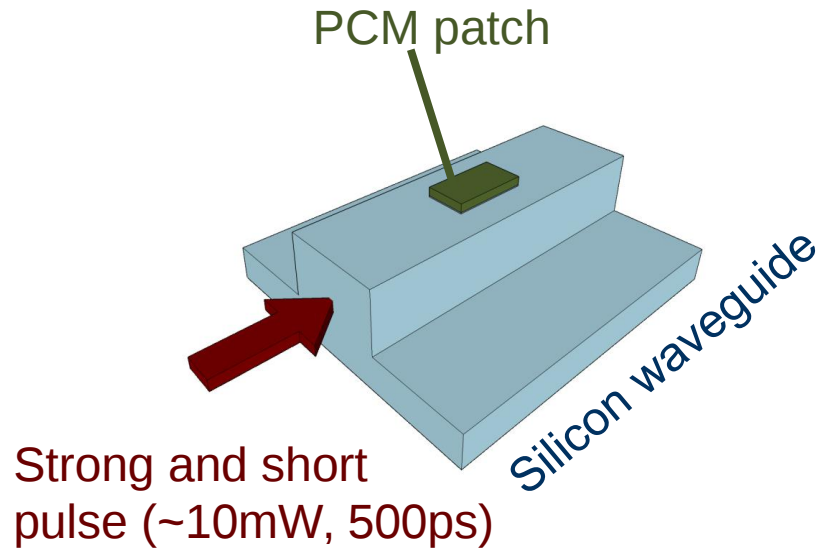


Brückerhoff-Plückelmann, et al., Chalcogenide phase-change devices for neuromorphic photonic computing, 2022

Phase-change materials in optics (oPCM)

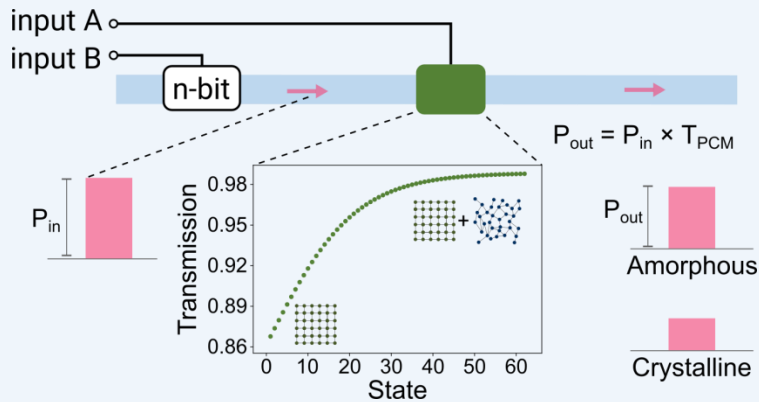


Changing between states



Photonic computing cells

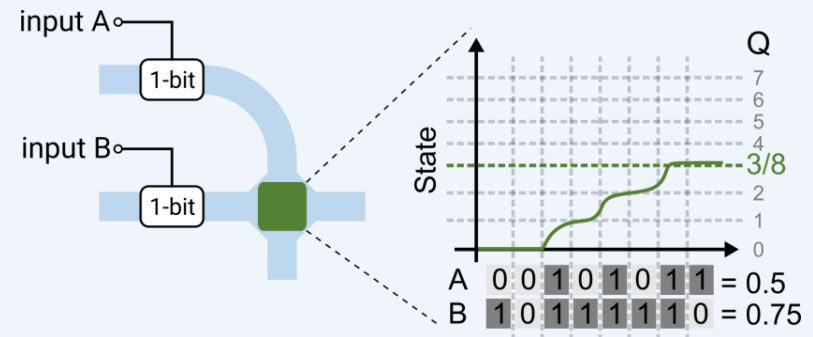
Baseline



- Performs a multiplication
- Constant PCM state (read)
- No native error

J. Feldmann, et al., Parallel convolutional processing using an integrated photonic tensor core, 2021

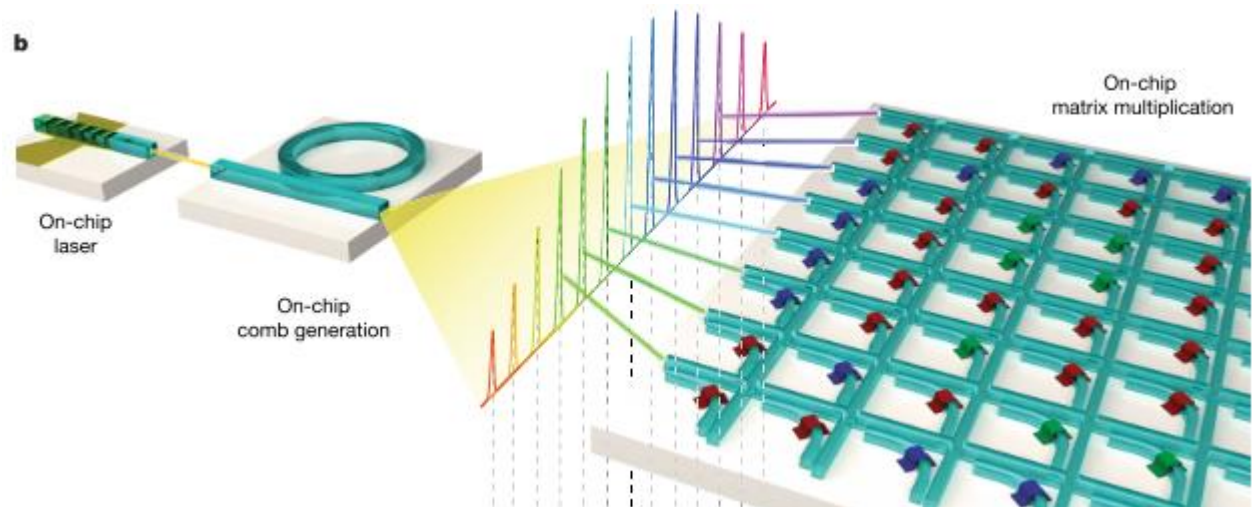
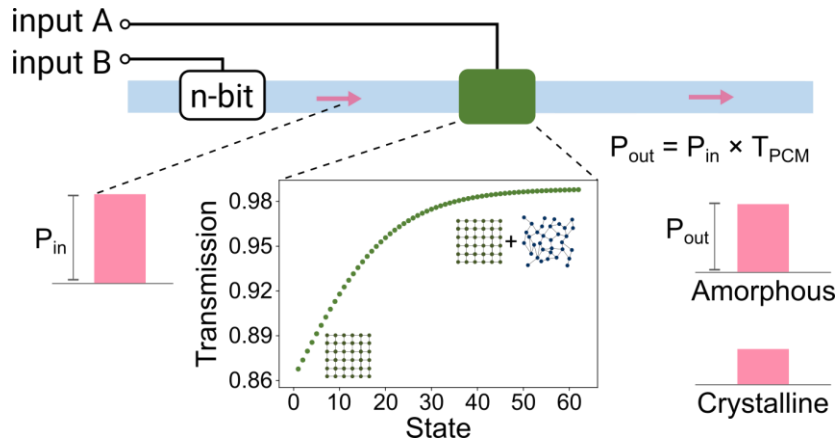
Stochastic-oPCM



- Performs sums of multiplications
- Changes PCM state (write)
- Has native error

R. Cardoso, et al., Towards a Robust Multiply-Accumulate Cell in Photonics using Phase-Change Materials, 2023

Baseline circuit



J. Feldmann, et al., Parallel convolutional processing using an integrated photonic tensor core, 2021

Stochastic-oPCM convolution circuit

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} * \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix}$$

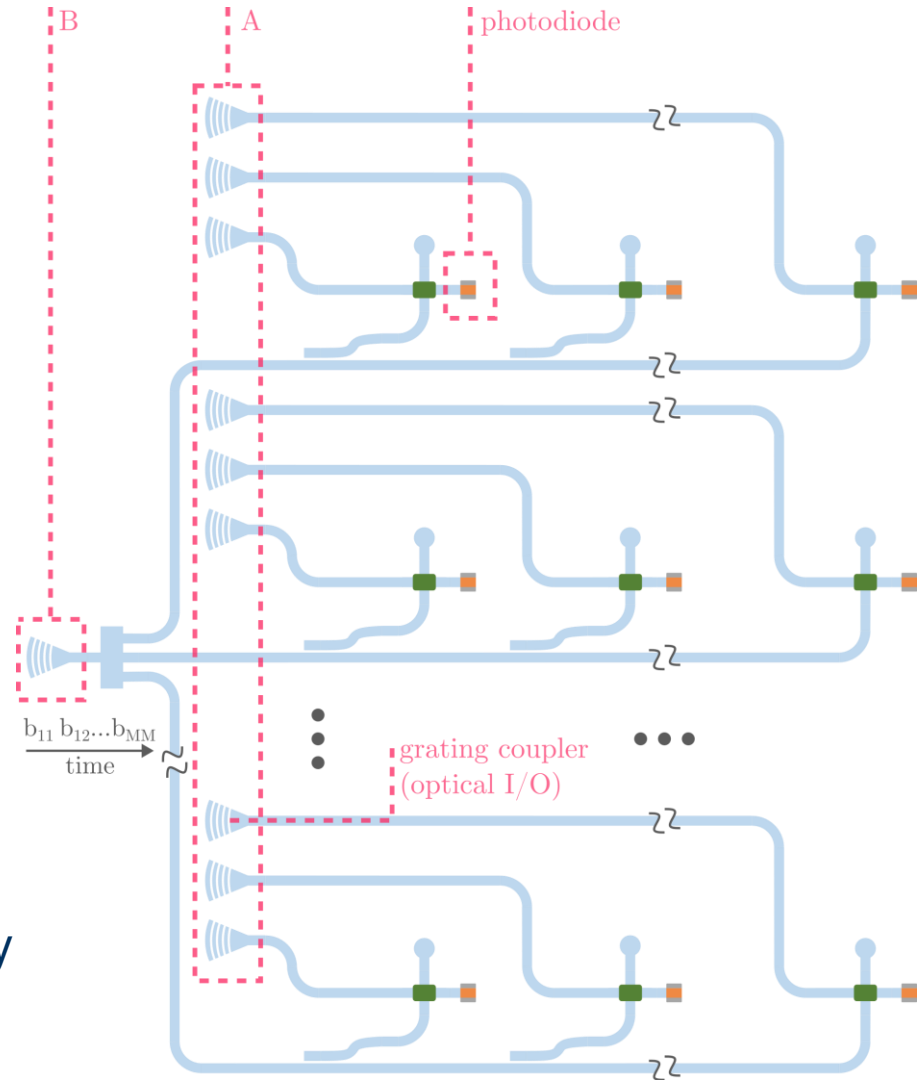
$$c_{11} = a_{11} b_{11} + a_{12} b_{12} + a_{21} b_{21} + a_{22} b_{22}$$

$$c_{12} = a_{12} b_{11} + a_{13} b_{12} + a_{22} b_{21} + a_{23} b_{22}$$

$$c_{21} = a_{21} b_{11} + a_{22} b_{12} + a_{31} b_{21} + a_{32} b_{22}$$

$$c_{22} = a_{22} b_{11} + a_{23} b_{12} + a_{32} b_{21} + a_{33} b_{22}$$

time →



- Convolution result **written** as PCM states → no need to spend energy and time sending to memory

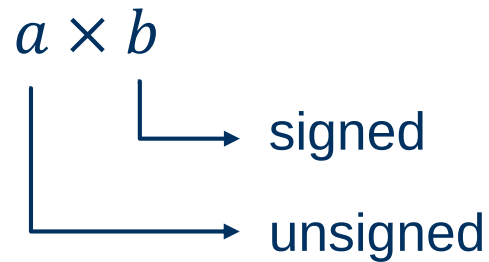
R. Cardoso, et al., Photonic Convolution Engine Based on Phase-Change Materials and Stochastic Computing, 2023

Only positive operands are supported

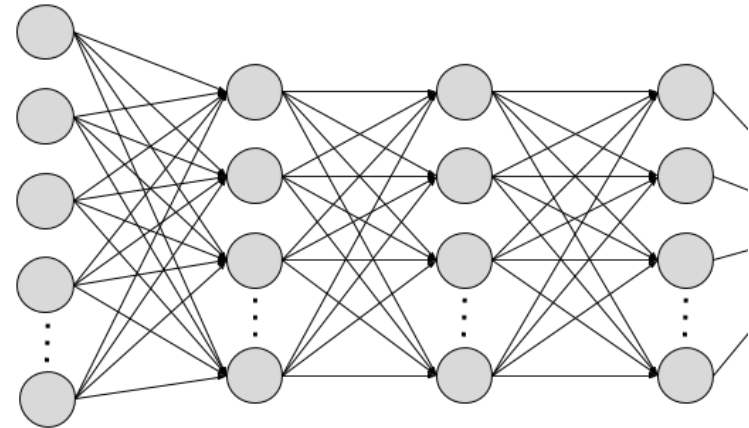


Applications of Mixed-Polarity operation

- Mixed polarity:

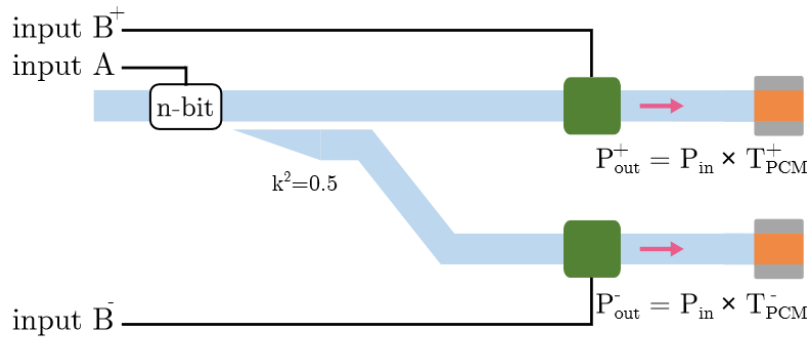


Filter-based image processing

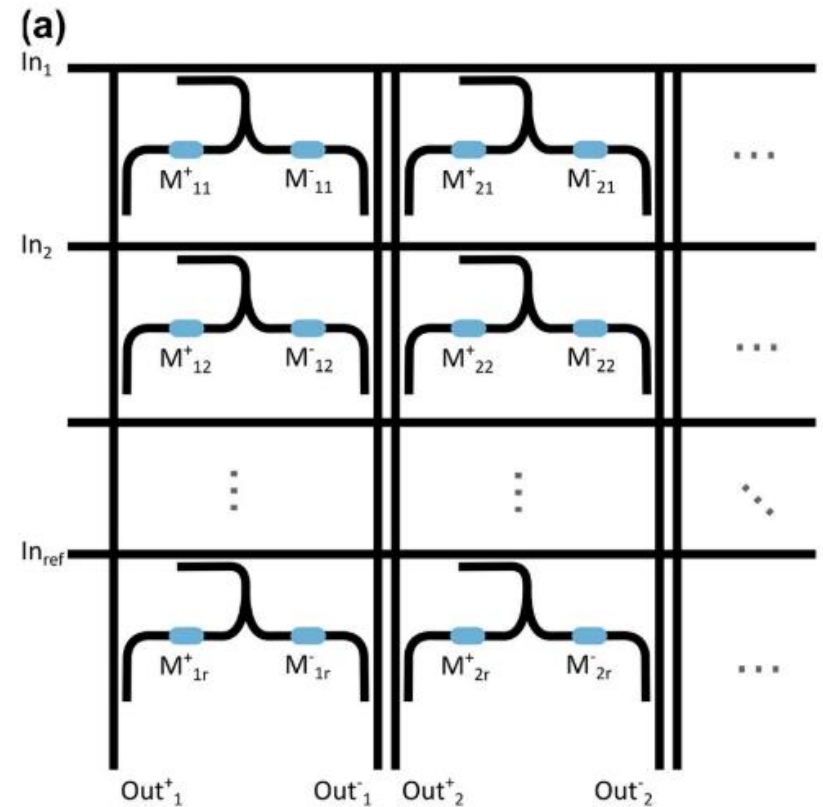


Deep learning *

Extension of baseline circuit



Needs two PCMs to deal with sign



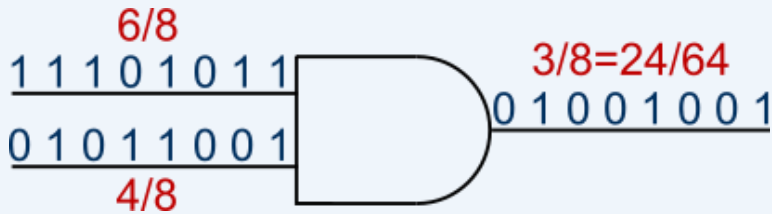
F. Brückerohoff-Plückelmann, et al., Broadband photonic tensor core with integrated ultra-low crosstalk wavelength multiplexers, 2022

Our contributions

1. Development of a **Stochastic Computing** formalism to Mixed-Polarity inputs
2. Development of a mixed-polarity photonic **cell** and **circuit** with PCM
3. **Conservative comparison** between methods with respect to noise

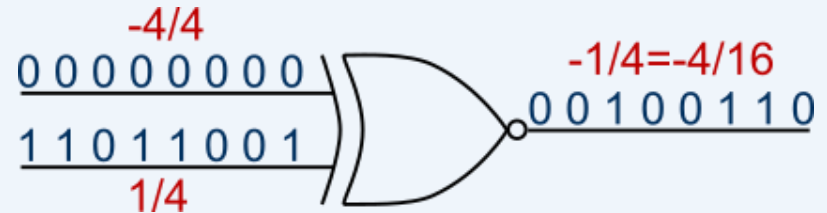
Stochastic computing

Unipolar



- Digital 0 encoded to +0, digital 1 encoded to +1
- Multiplication done by AND

Bipolar



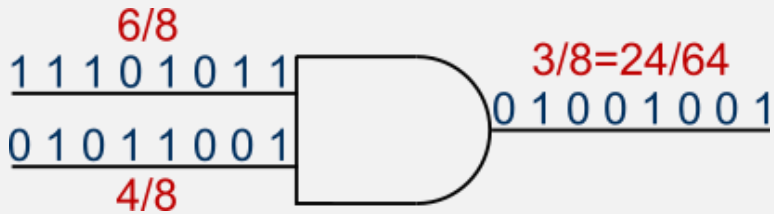
- Digital 0 encoded to -1, digital 1 encoded to +1
- Multiplication done by XNOR

A. Alaghi, et al., Survey of stochastic computing, 2013

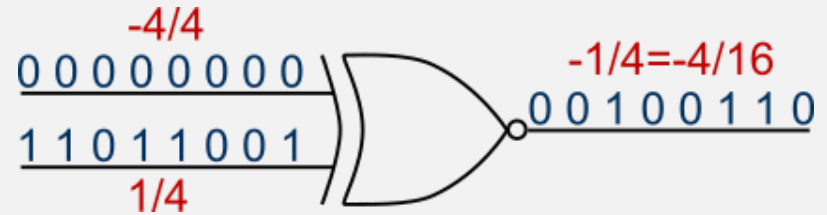
In both cases, the result is recovered by a counter at the output

Mixed-polarity in stochastic computing

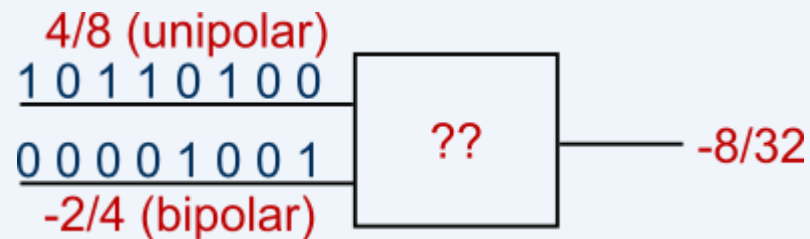
Unipolar



Bipolar

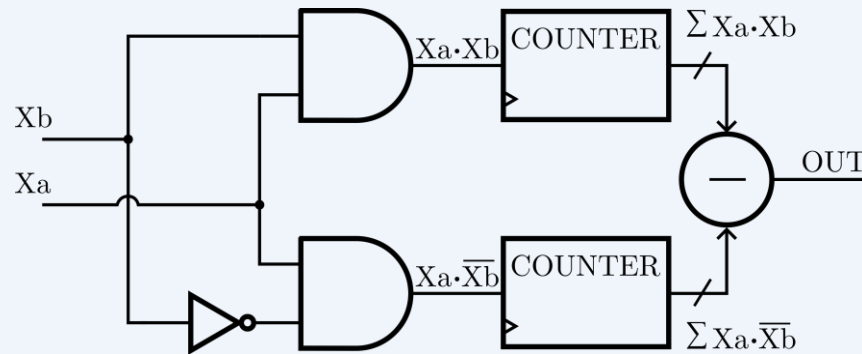


Mixed-polarity



Mixed-polarity in stochastic computing

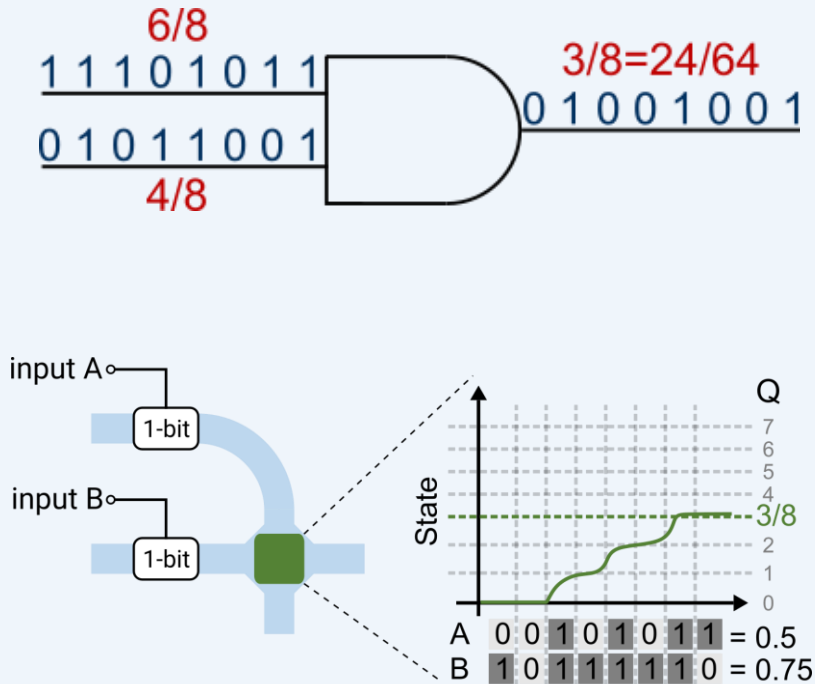
Mixed-polarity



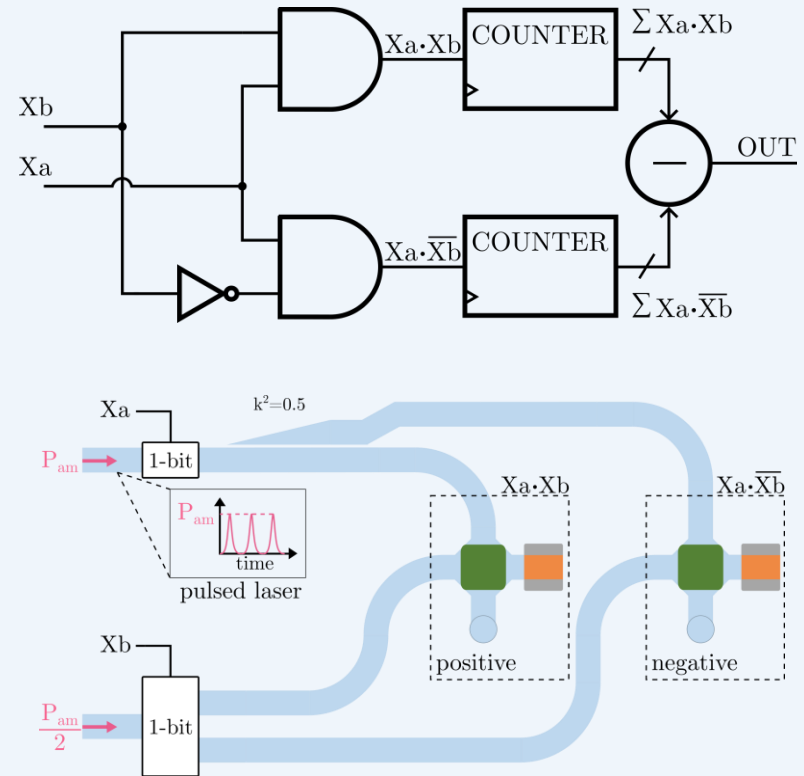
- Output is not stochastic
- Uses **only** AND and counters
- Proof of validity provided

Photonic mixed-polarity cell

Unipolar



Mixed-polarity



Photonic mixed-polarity circuit

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} * \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} c_{11} & c_{12} \\ c_{21} & c_{22} \end{bmatrix}$$

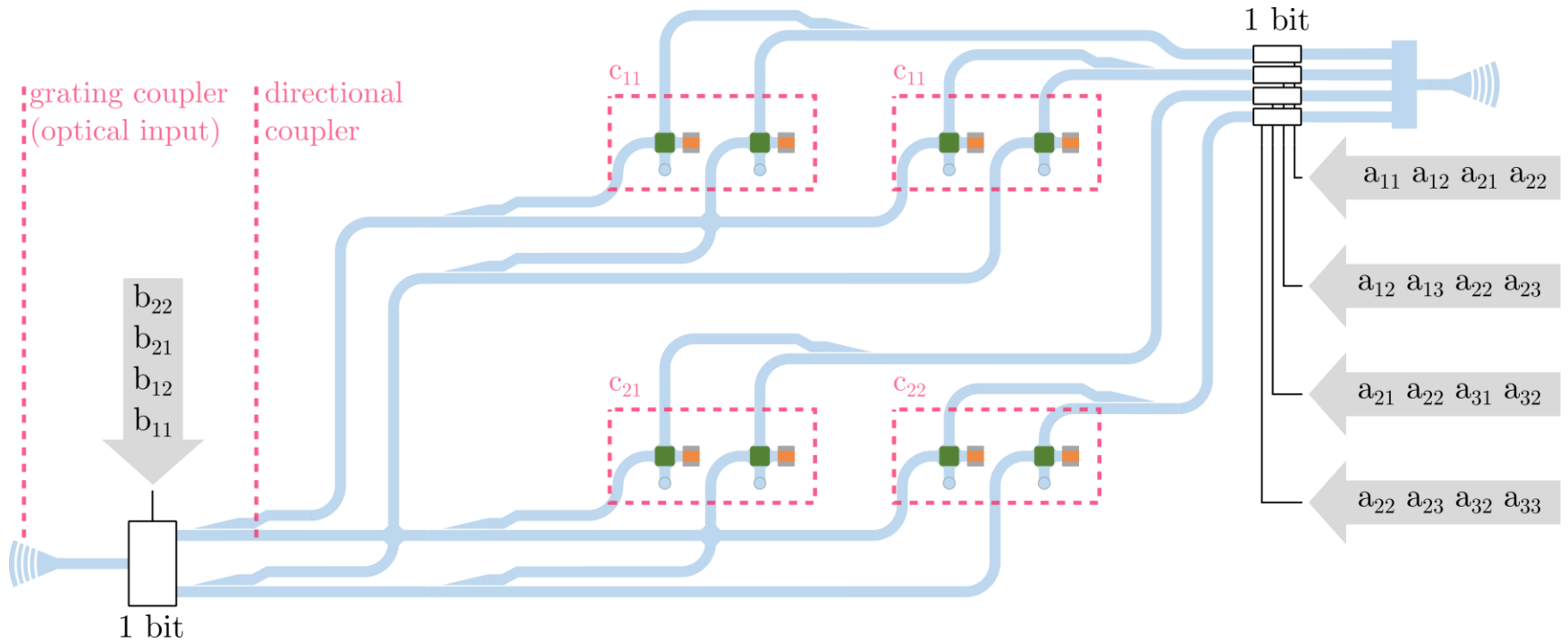
$$c_{11} = a_{11}b_{11} + a_{12}b_{12} + a_{21}b_{21} + a_{22}b_{22}$$

$$c_{12} = a_{12}b_{11} + a_{13}b_{12} + a_{22}b_{21} + a_{23}b_{22}$$

$$c_{21} = a_{21}b_{11} + a_{22}b_{12} + a_{31}b_{21} + a_{32}b_{22}$$

$$c_{22} = a_{22}b_{11} + a_{23}b_{12} + a_{32}b_{21} + a_{33}b_{22}$$

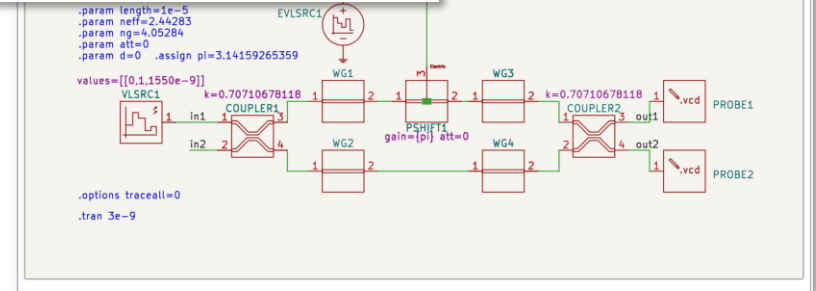
time



A visualizable convolution application

Introducing SPECS: Scalable Photonic Event-driven Circuit Simulator

Clément Zrounba⁽¹⁾, Raphael Cardoso⁽¹⁾, Maurício Gomes de Queiroz⁽¹⁾, Paul Jimenez⁽¹⁾,
Mohab Abdalla^(1,2), Alberto Bosio⁽¹⁾, Sébastien Le Beux⁽³⁾, Fabio Pavanello⁽¹⁾, Ian O'Connor⁽¹⁾

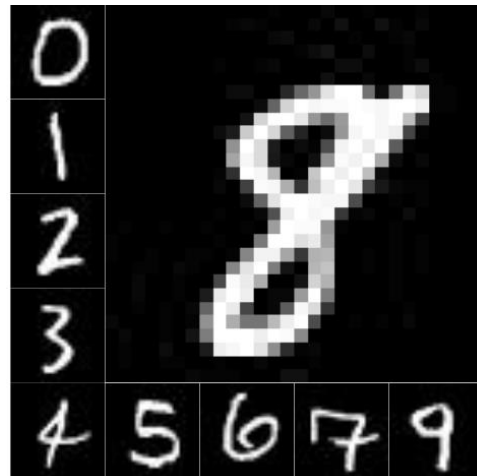


demo.zrounba.fr/specs/

A visualizable convolution application

Edge detection filter

- Input: 28x28 grayscale images
- Application: edge detection with Sobel filters
- Quantization: 6 bits
- Output bandwidth: 1 GHz



$$\begin{matrix} * \begin{pmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{pmatrix} \\ * \begin{pmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{pmatrix} \end{matrix} \left. \vphantom{\begin{matrix} * \begin{pmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{pmatrix} \\ * \begin{pmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{pmatrix} } \right\}$$

Ideal operation

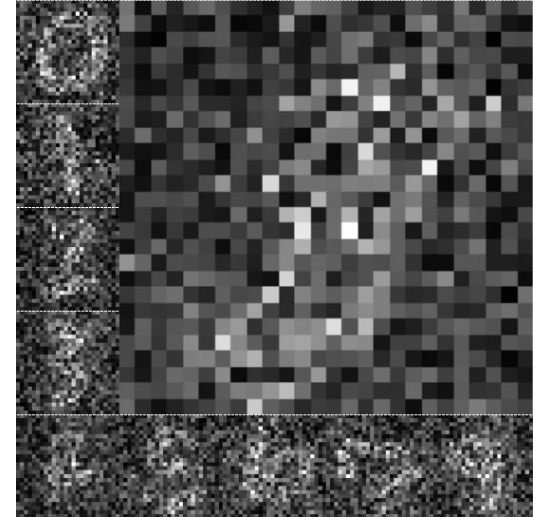


Filter in photonics

Ideal



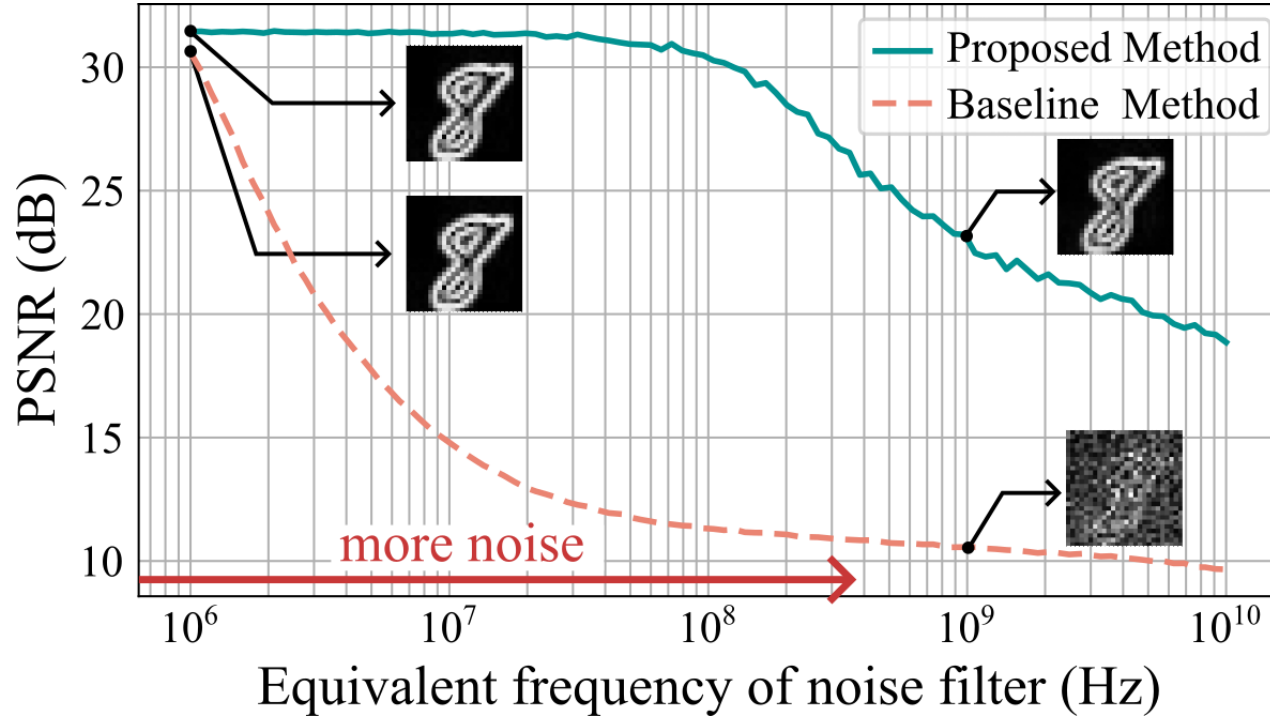
Baseline Method



Proposed Method



Noise dependency



- Noise is a function of the filter in the output (bandwidth)
- Speed is limited by noise in the baseline

Conclusions and perspectives

- Confirmed **noise robustness** of the stochastic computing cell in photonics
- **Speed** in our case limited by the time between amorphization events in a PCM ($\sim 1\text{ns/event}$)
- Challenges:
 - Limited **endurance** of the PCMs
 - Natural **randomness** in PCM phase dynamics
 - Testing a fabricated stochastic-oPCM cell (**taped out!**)

raphael.cardoso@ec-lyon.fr