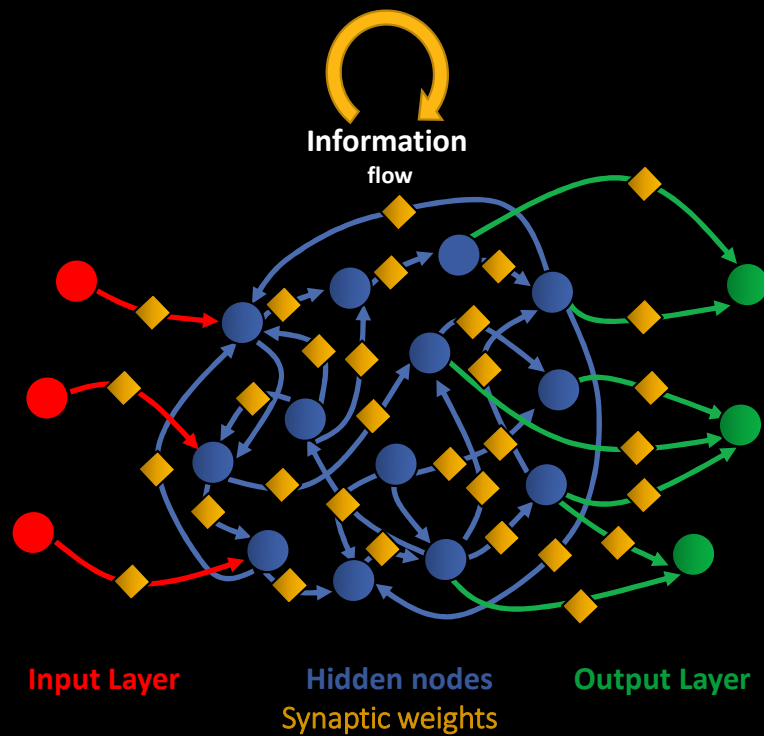


Cascaded Mach-Zehnder Interferometers for Efficient AI Acceleration

by Felix Hermann, Pascal Stark, Mustafa Yildirim, Folkert Horst, Jonas Weiss, Bert Jan Offrein

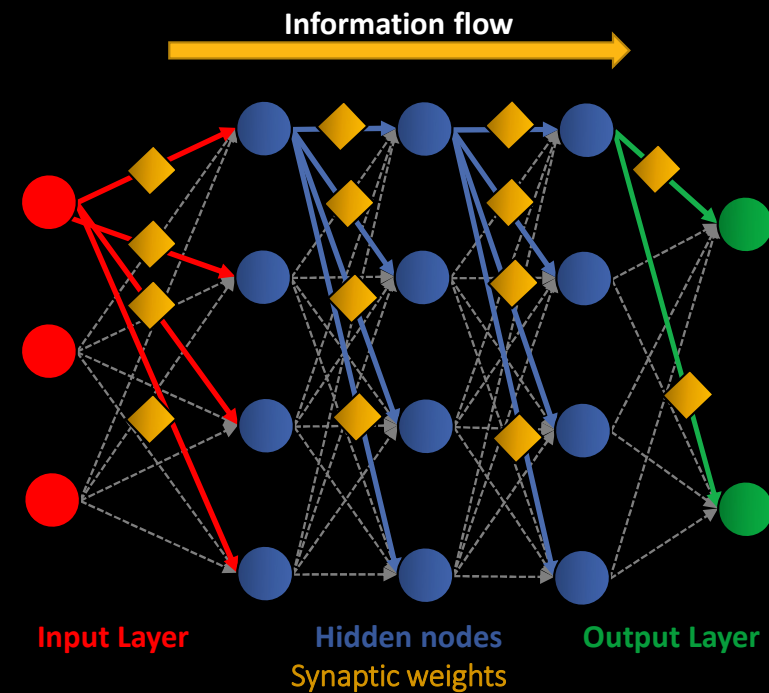
Neuromorphic Systems and Devices





human brains

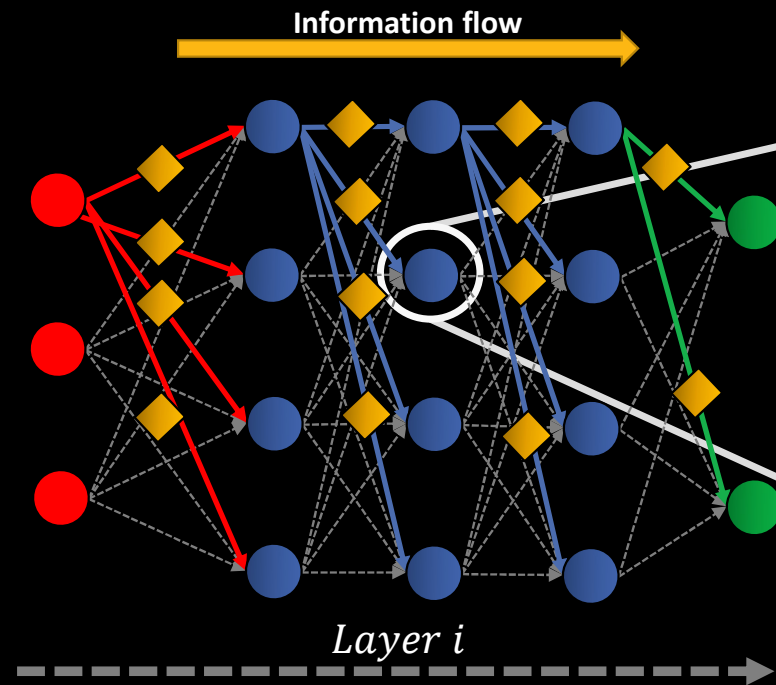
- recurrent connections
- information is stored in network
- training by creating new and tuning existing synapses



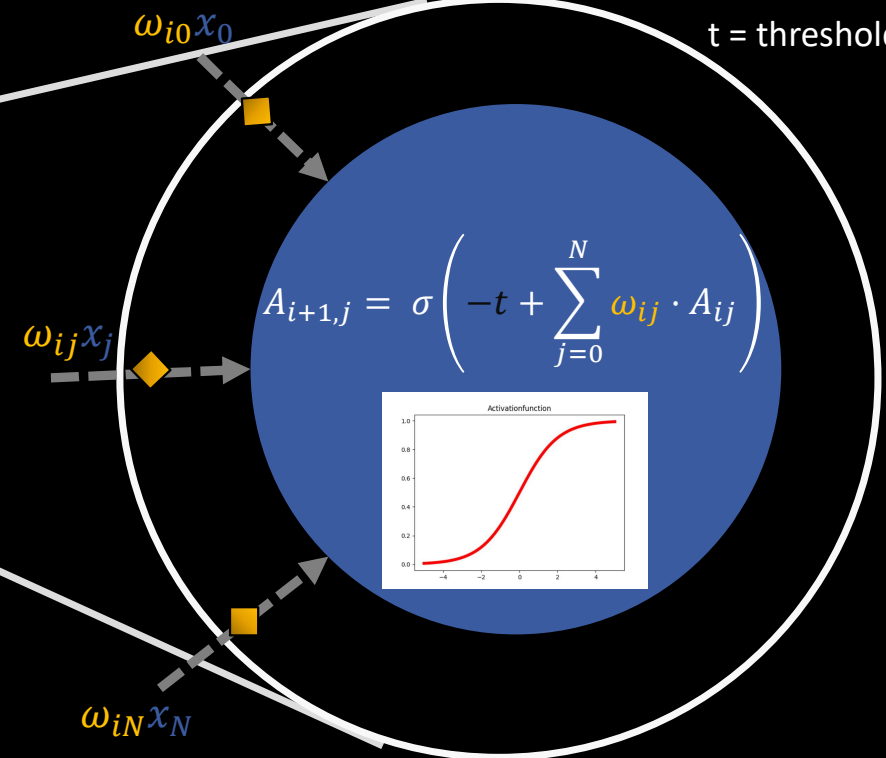
neural network

- layer-by-layer signal processing
- training by tuning synapses

-----> Synapse



Neuron j



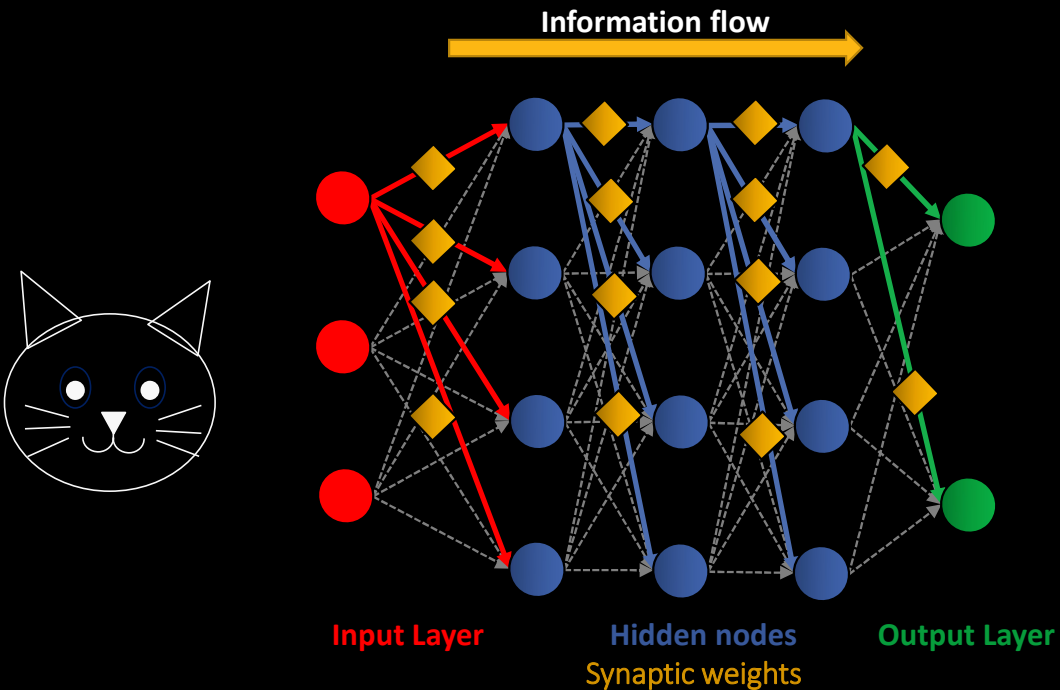
Neural network

- layer-by-layer signal processing
- training by tuning synapses

Neurons

- connected by weighted connections → Synapses
- activation is derived from evaluating activation function for a weighted sum of inputs

Tasks for Neural Networks



recognize:

- images from a camera,
- language from microphone or
- patterns in data set

Dog

Cat

- detect features by e.g., convolution
→ requires massive computing power

What is a Convolution in Image Processing?

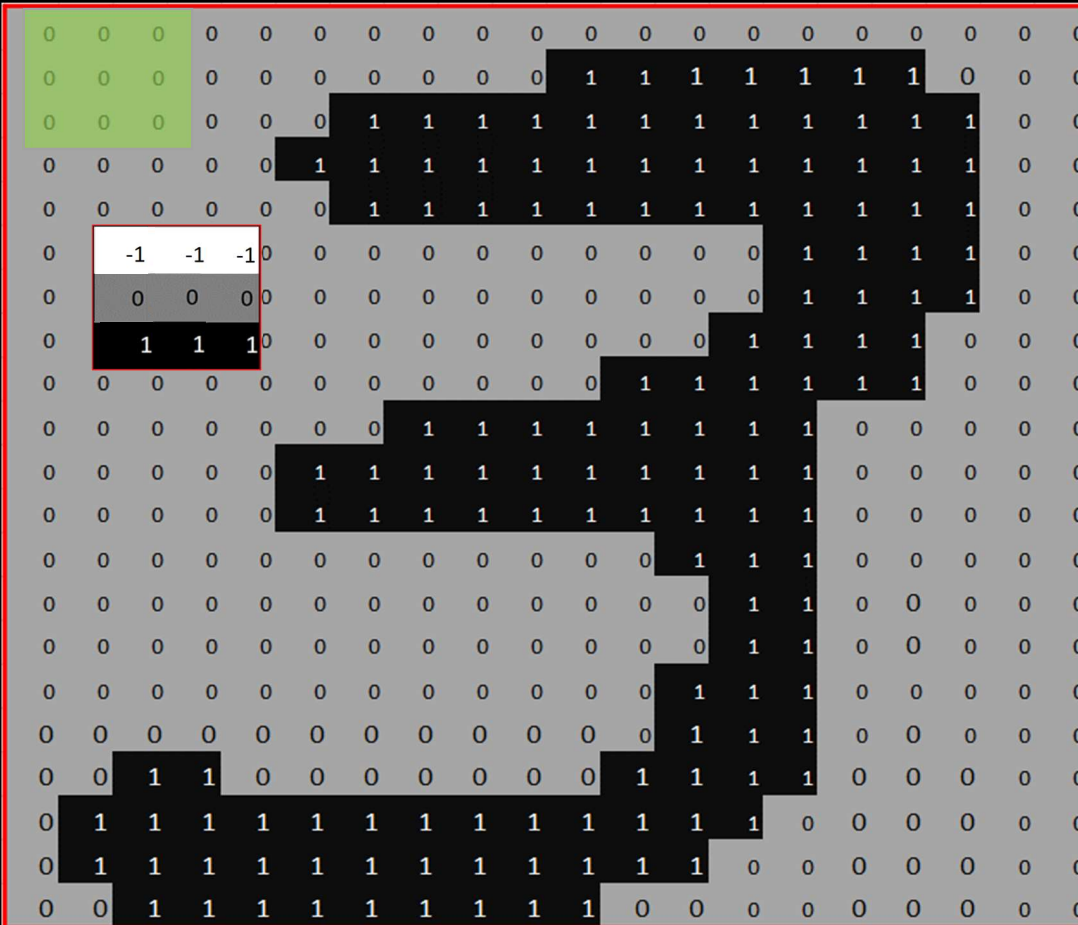
defined filter-kernel

-1	-1	-1
0	0	0
1	1	1

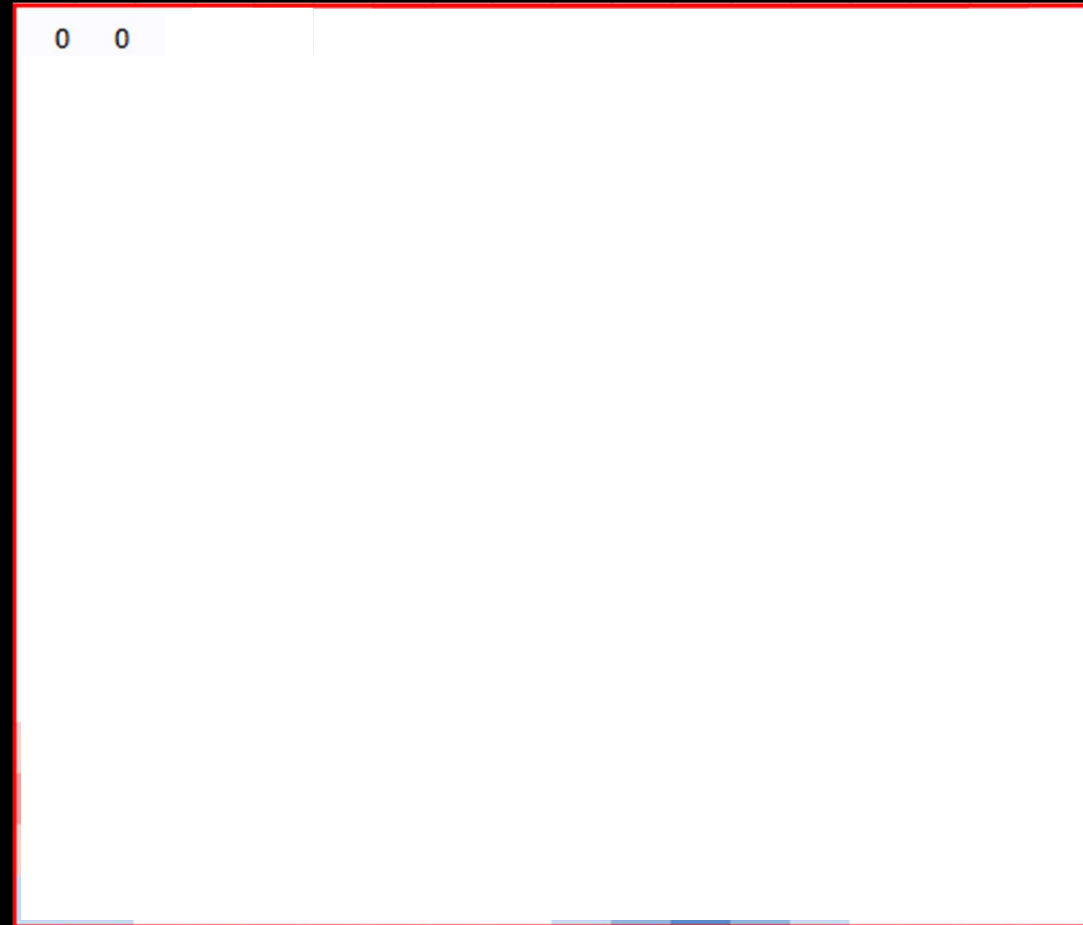
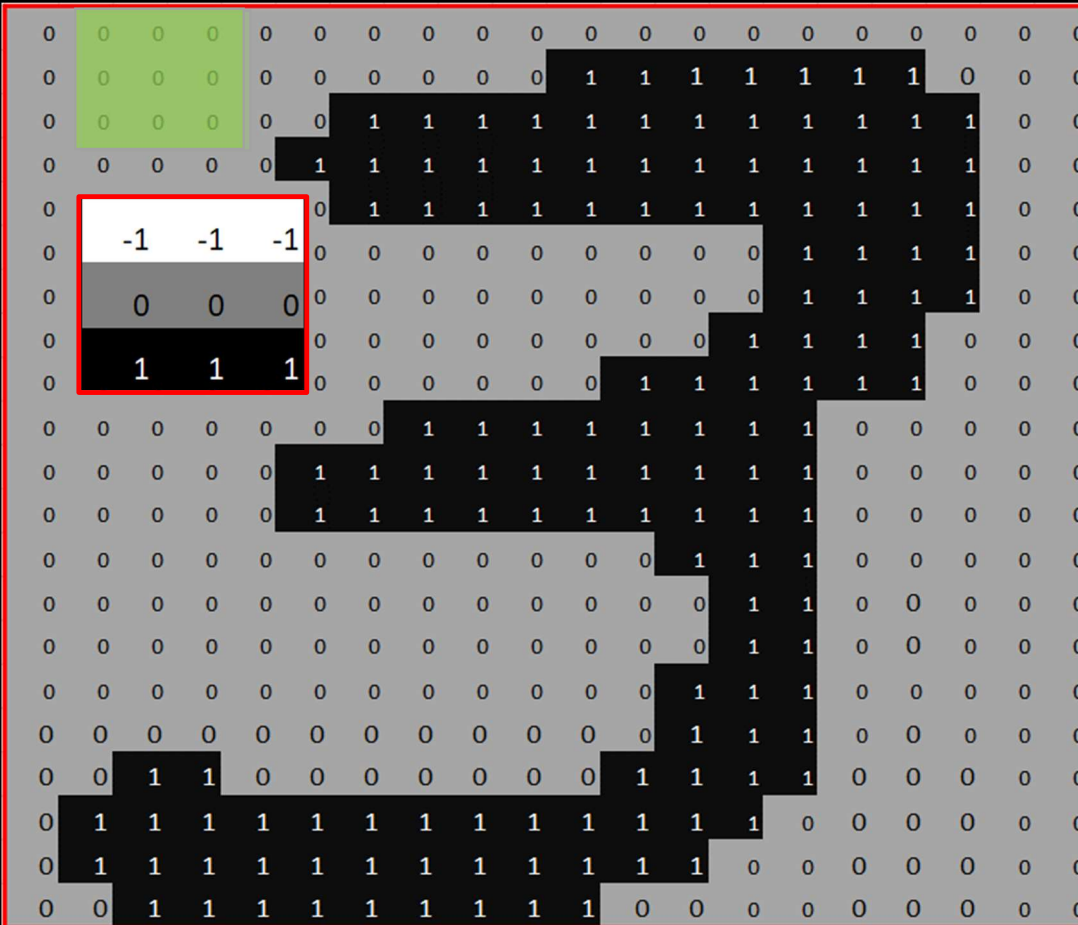
unknown image

[illegible]

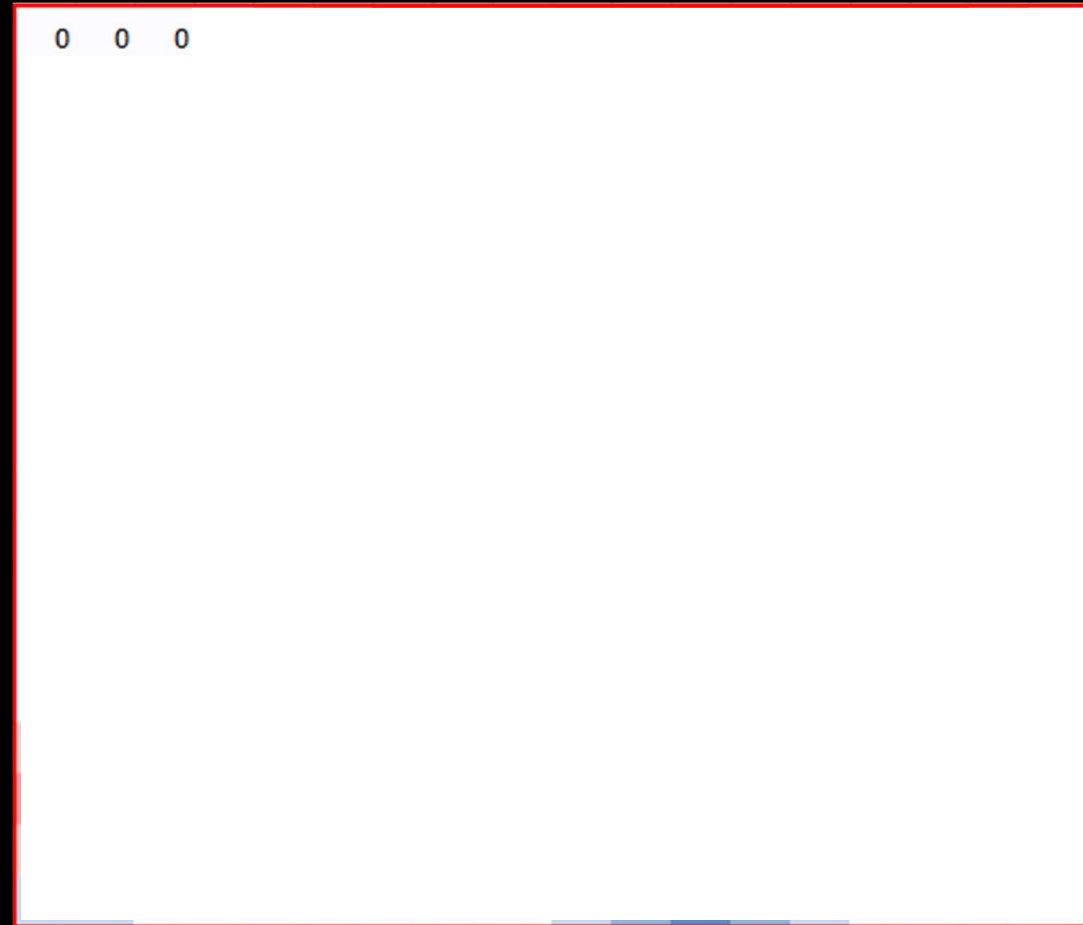
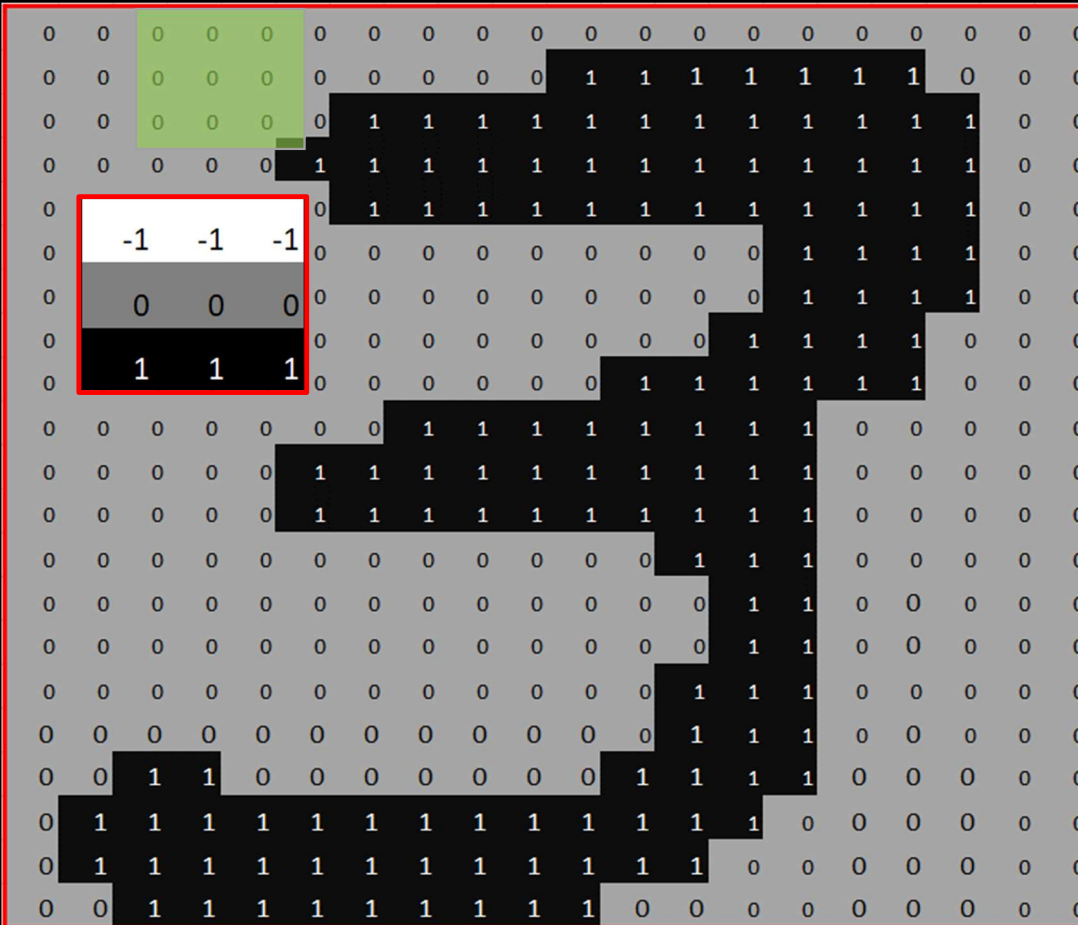
What is a Convolution in Image Processing?



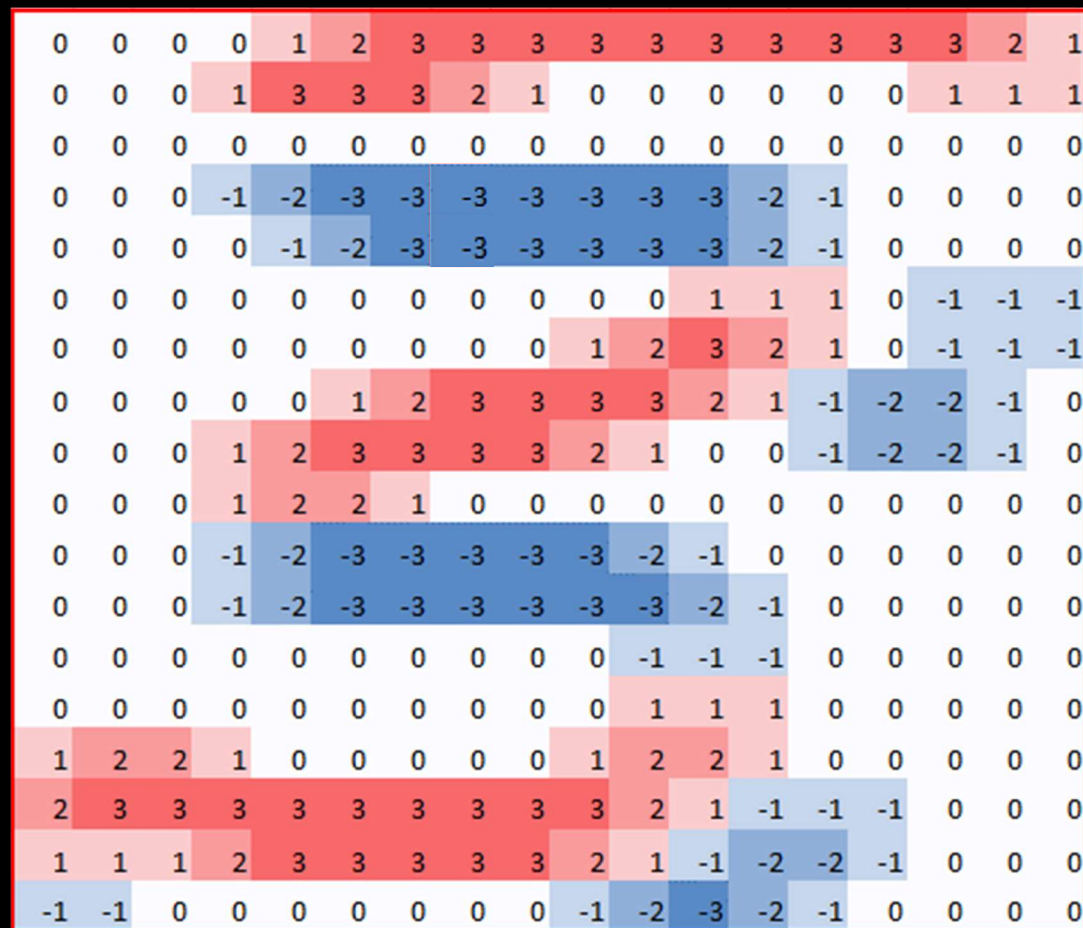
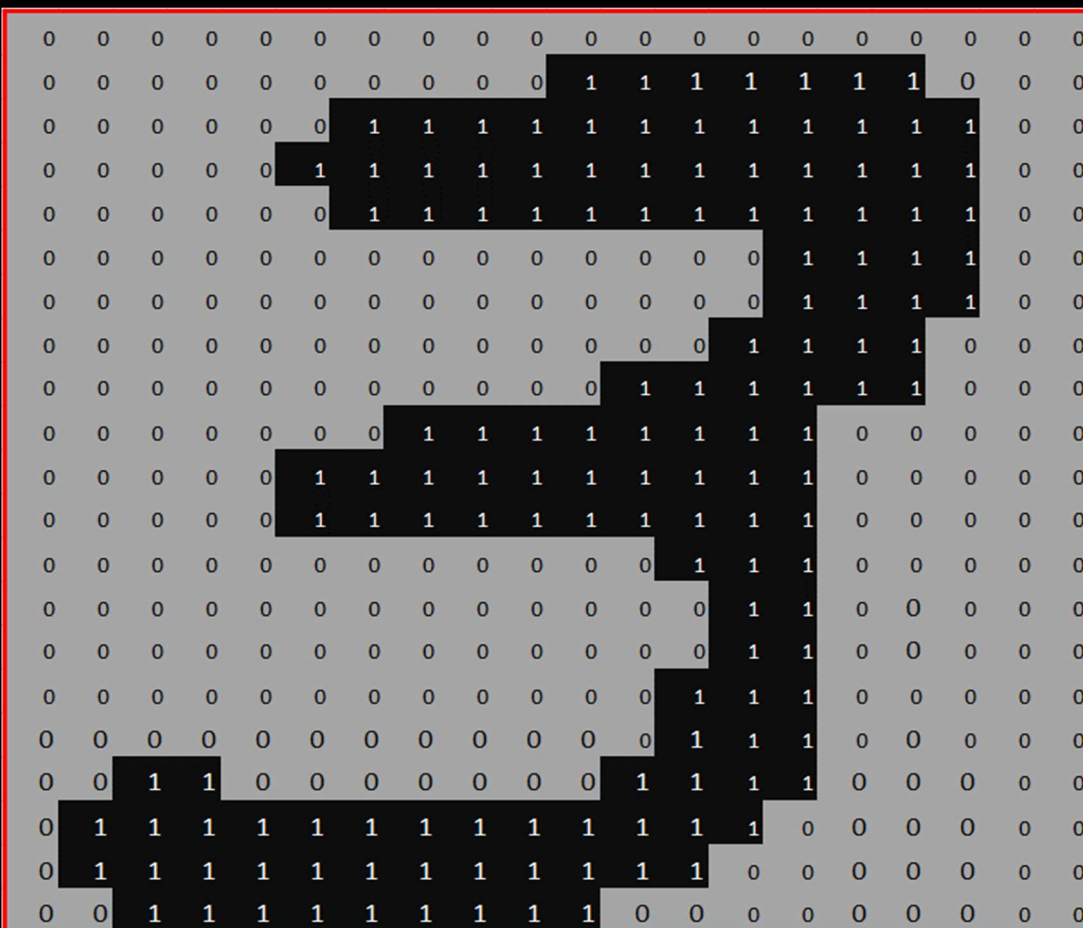
What is a Convolution in Image Processing?



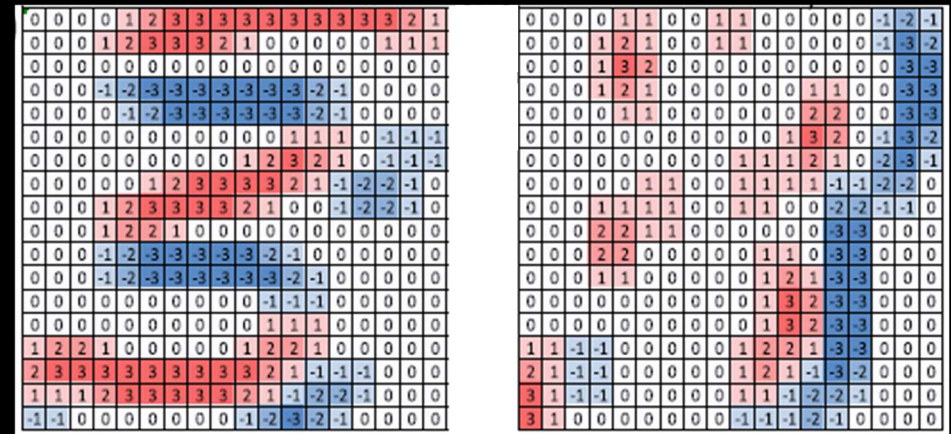
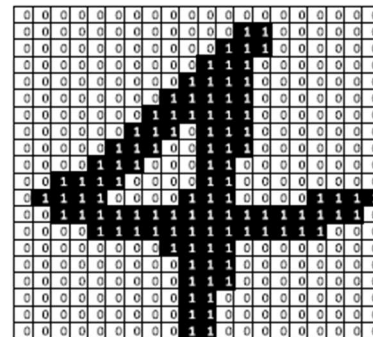
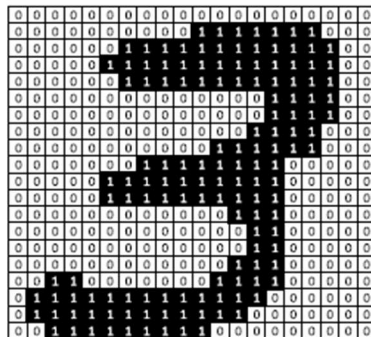
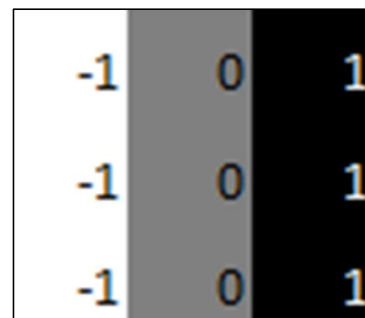
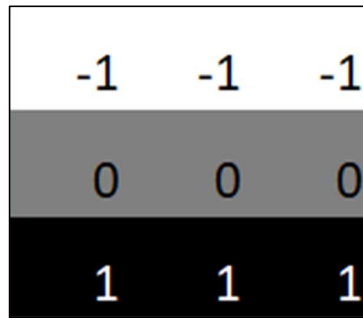
What is a Convolution in Image Processing?



What is a Convolution in Image Processing?

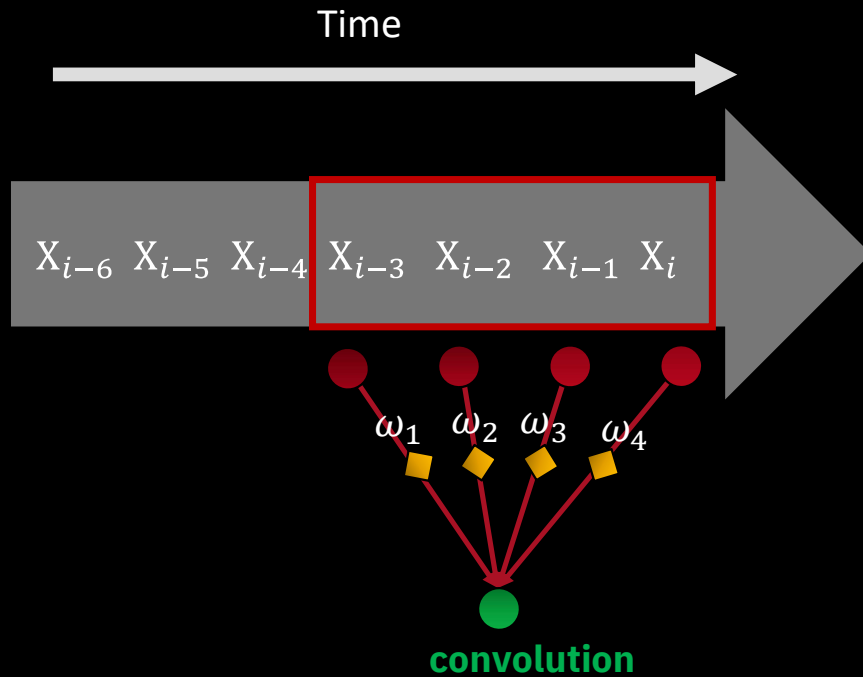


What is a Convolution in Image Processing?



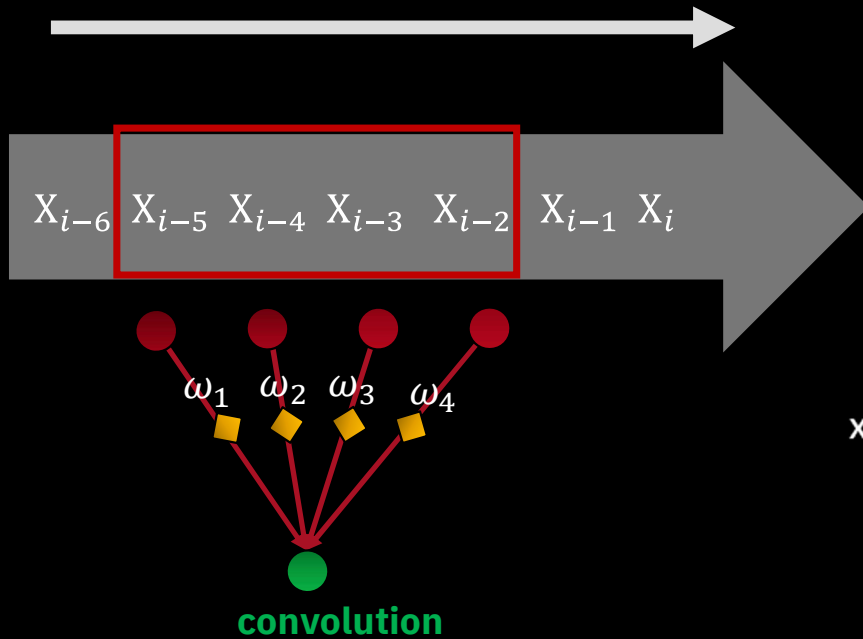
- convolution detects features, which allows for interpretation further processing

Convolution in Neural Networks

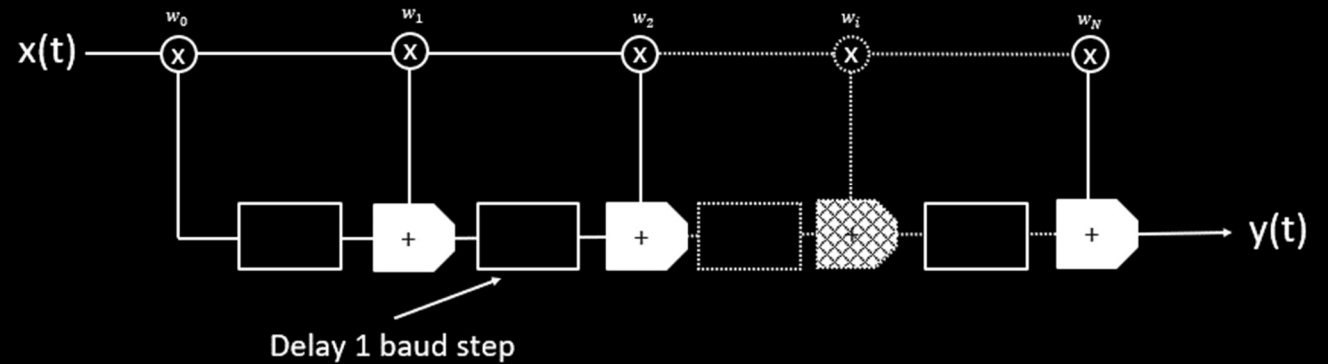


Convolution in Neural Networks

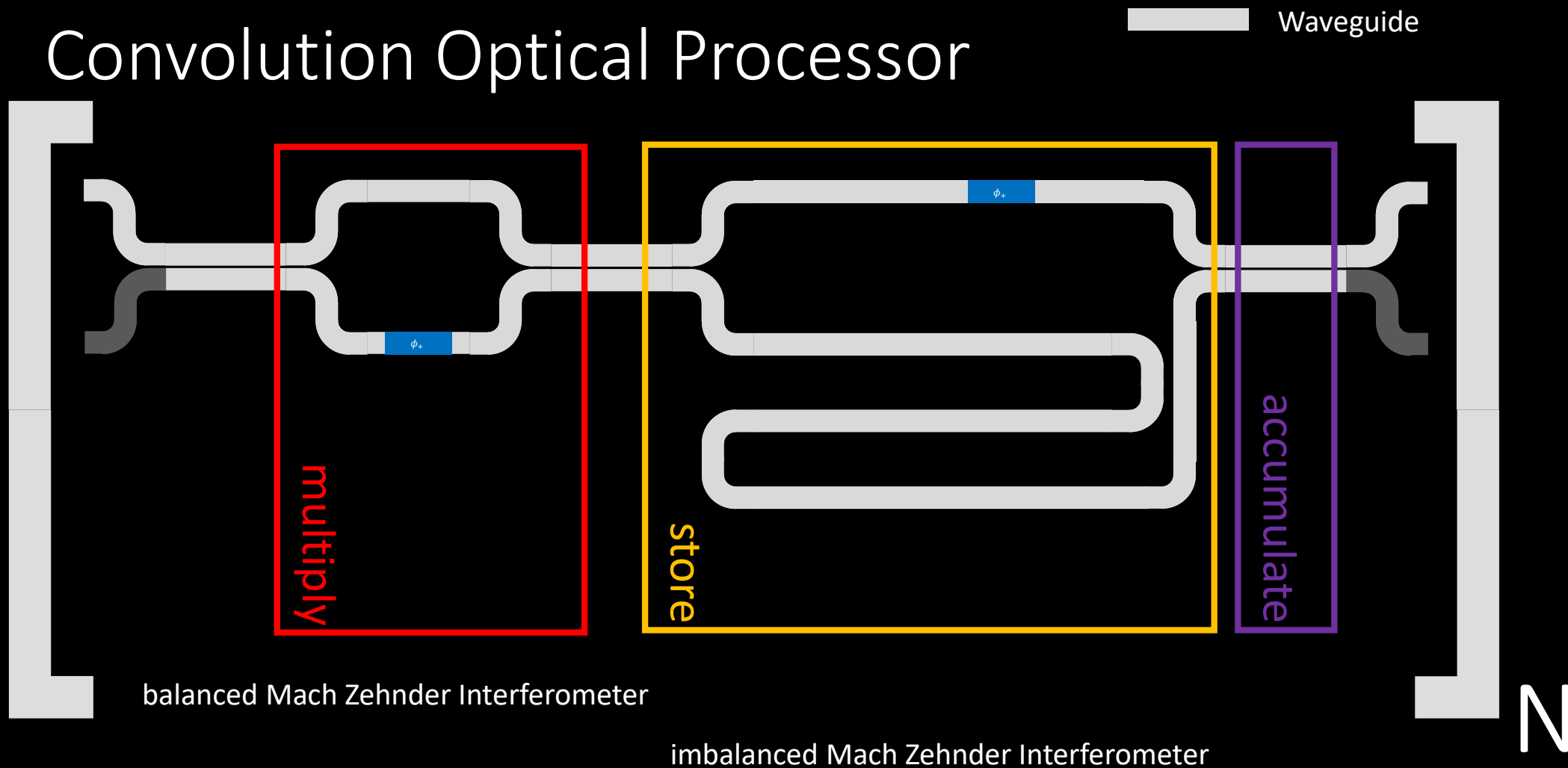
Time



$$y[n] = \sum_{i=0}^N \omega_i \cdot x[n - i]$$



Convolution Optical Processor



Layout and Setup

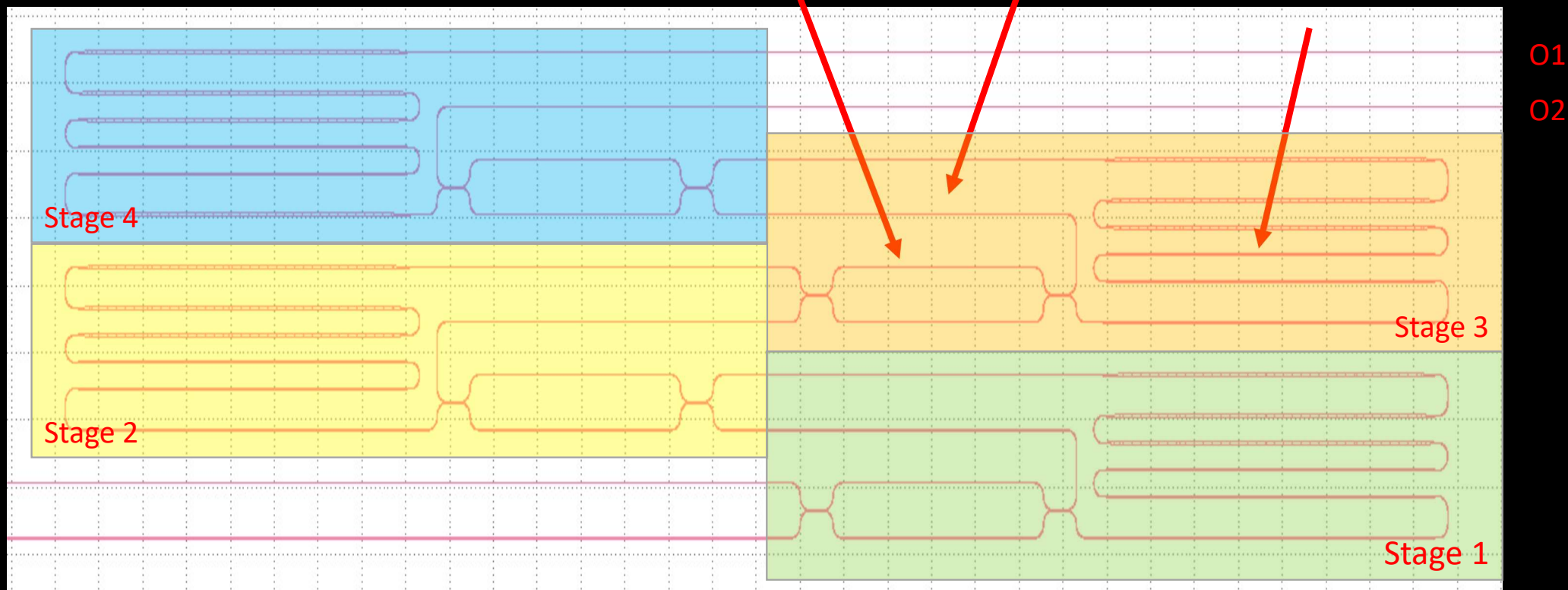


waveguide

Multiplier

short line

delay line



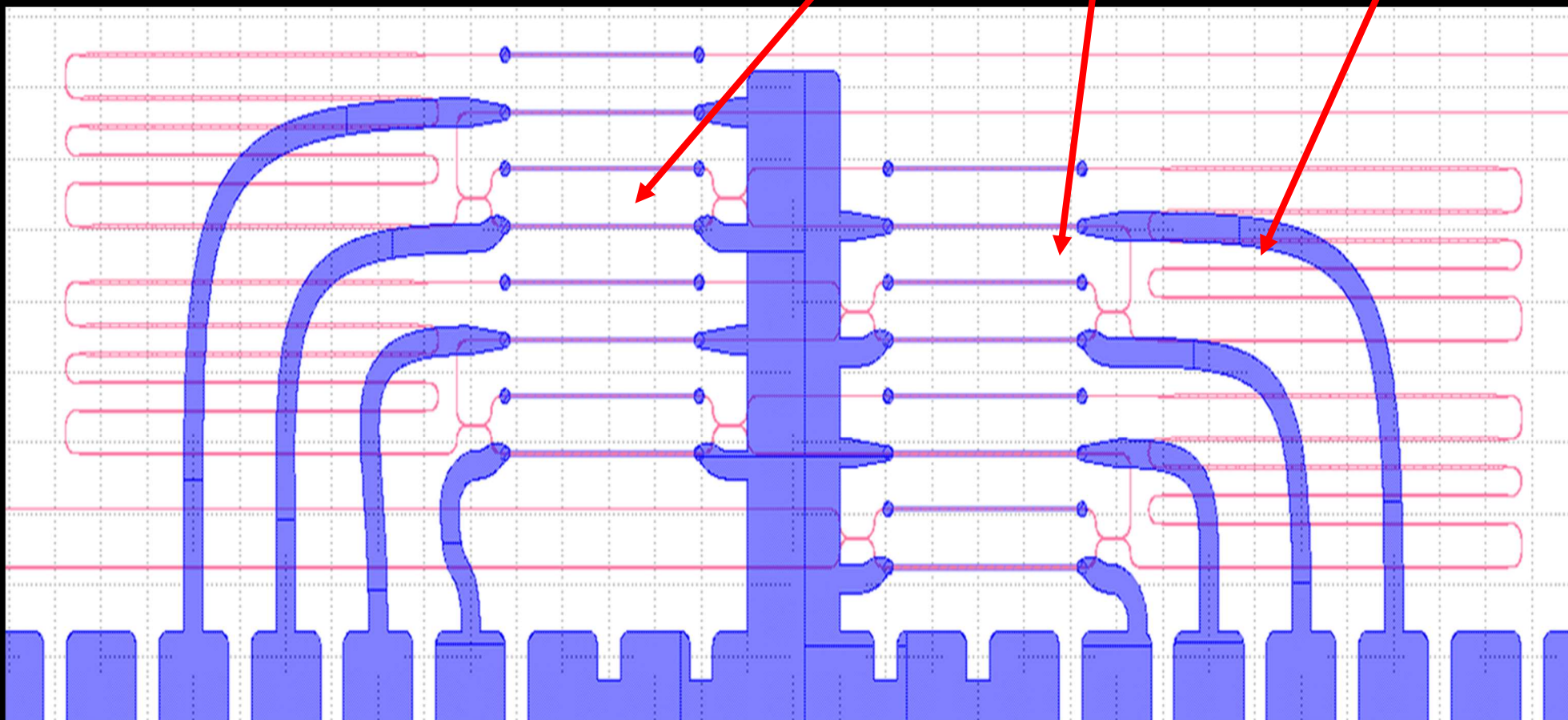
Layout and setup

■ metal ■ waveguide

Tunable coupler

short path

delay line



I1

I2

O1

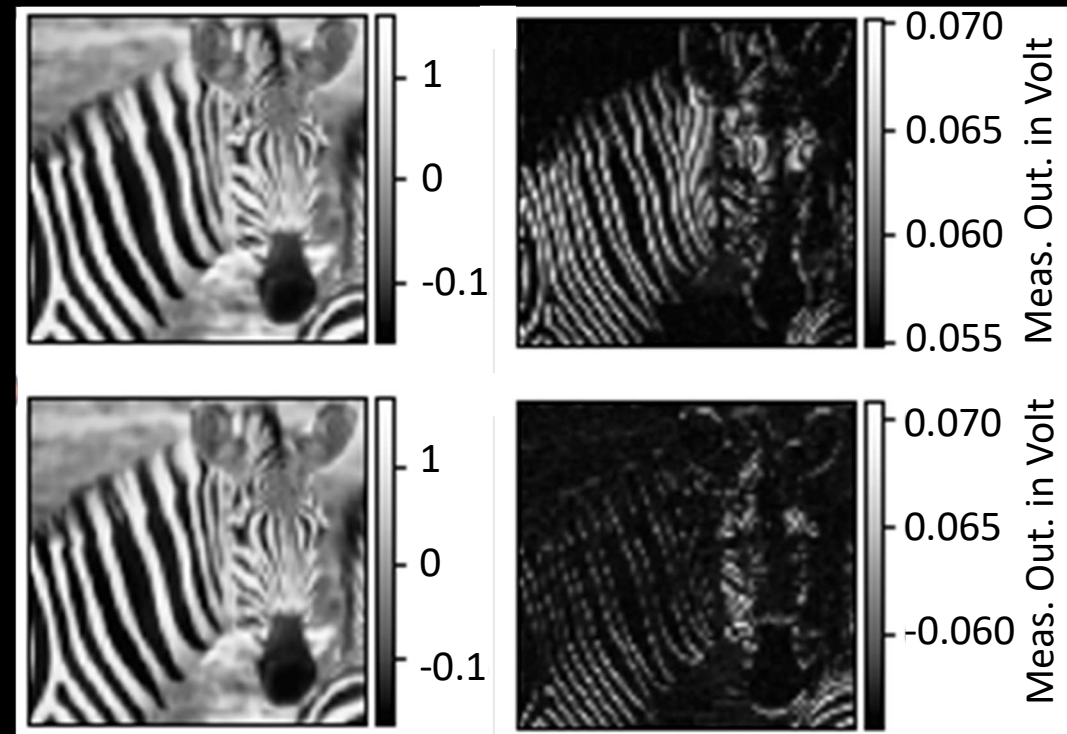
O2

proof of principle results: Sobel filter

- 2-element kernel:

$$\begin{pmatrix} 1 \\ -1 \end{pmatrix} \text{vertical}$$

$$\begin{pmatrix} 1 & -1 \end{pmatrix} \text{horizontal}$$

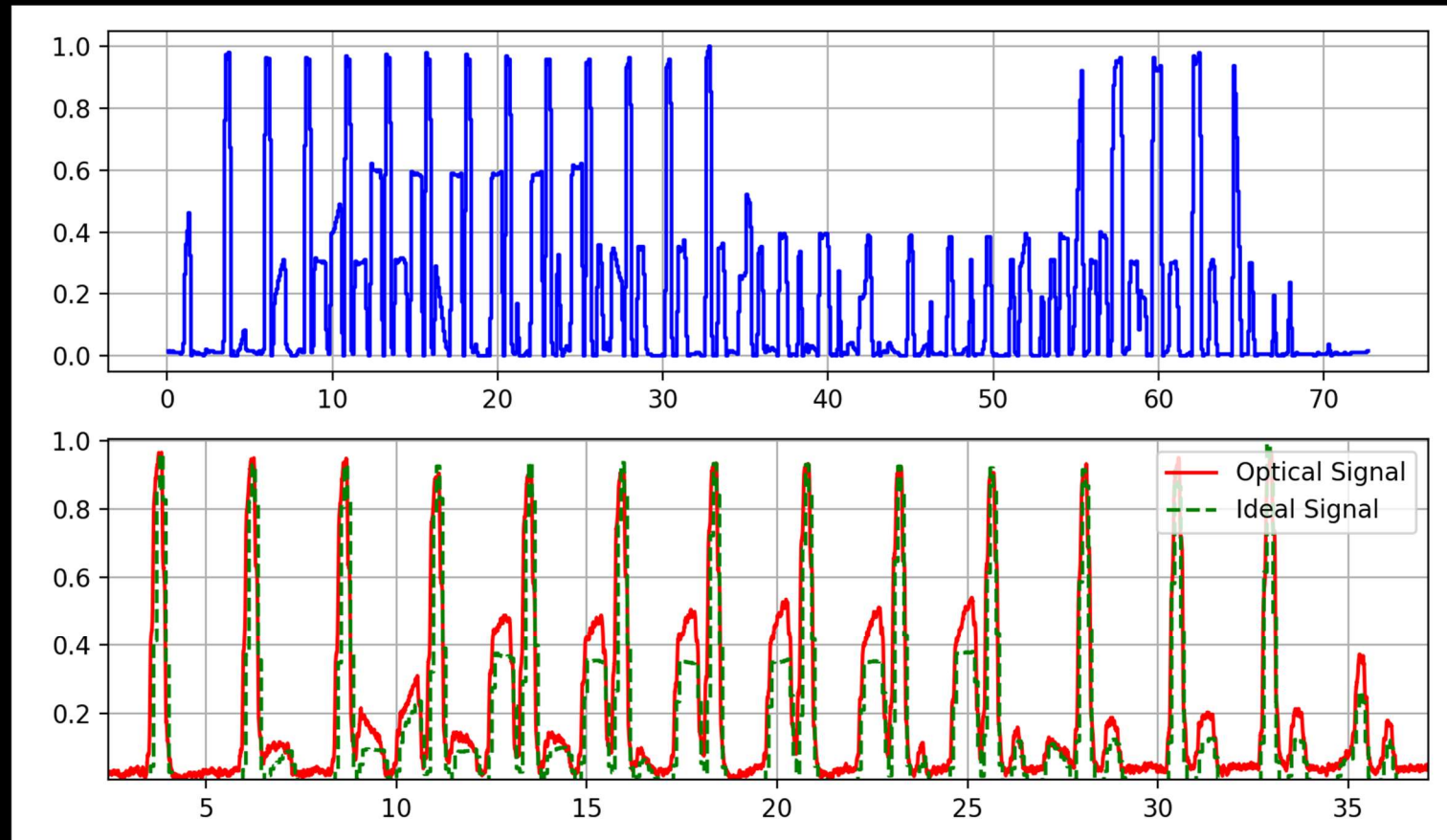


reprinted from Doctoral Thesis by Pascal Stark

proof of principle results: Sobel filter

- 3-element kernel:

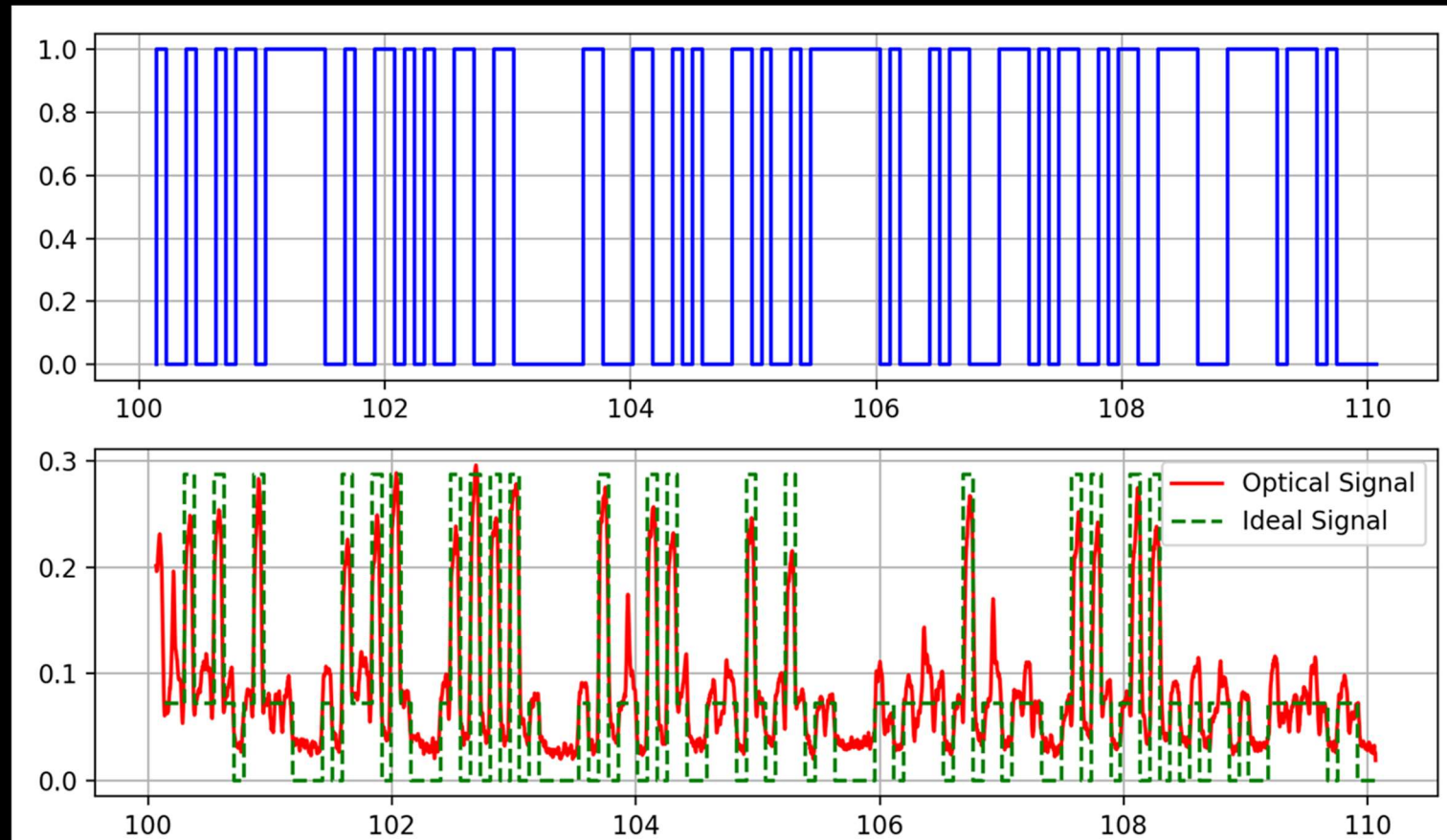
$$\begin{pmatrix} 1 & 2 & 1 \end{pmatrix}$$



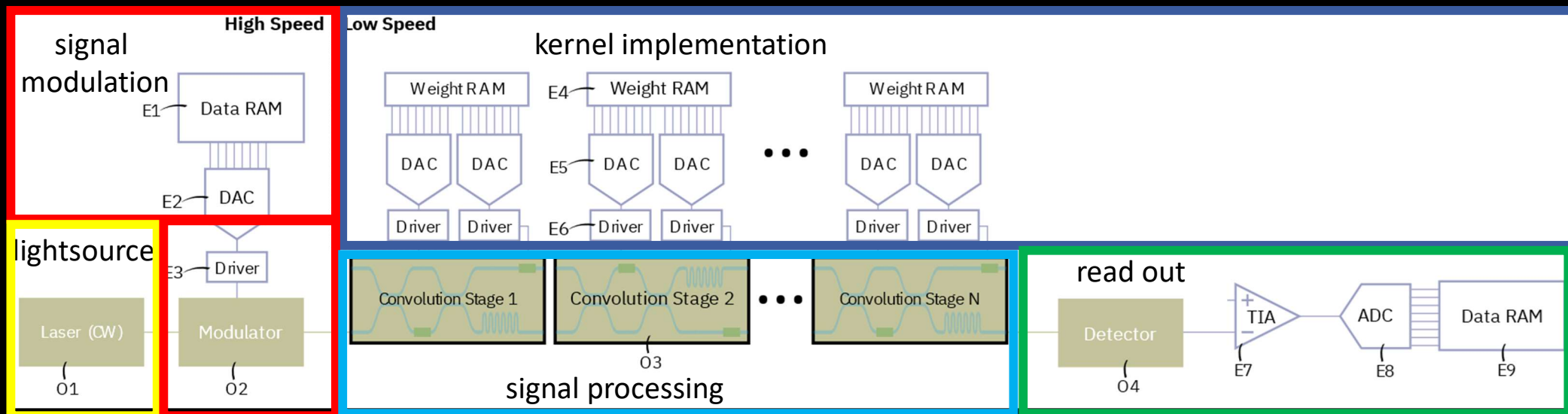
proof of principle results: Sobel filter

- 4-element kernel:

$$\begin{pmatrix} 1 & -1 & -1 & 1 \end{pmatrix}$$

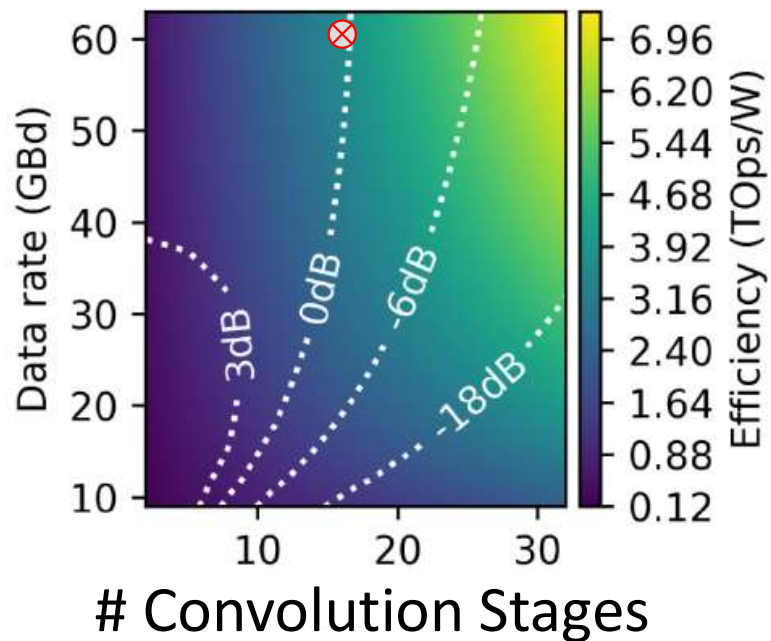


Benchmarking



Benchmarking

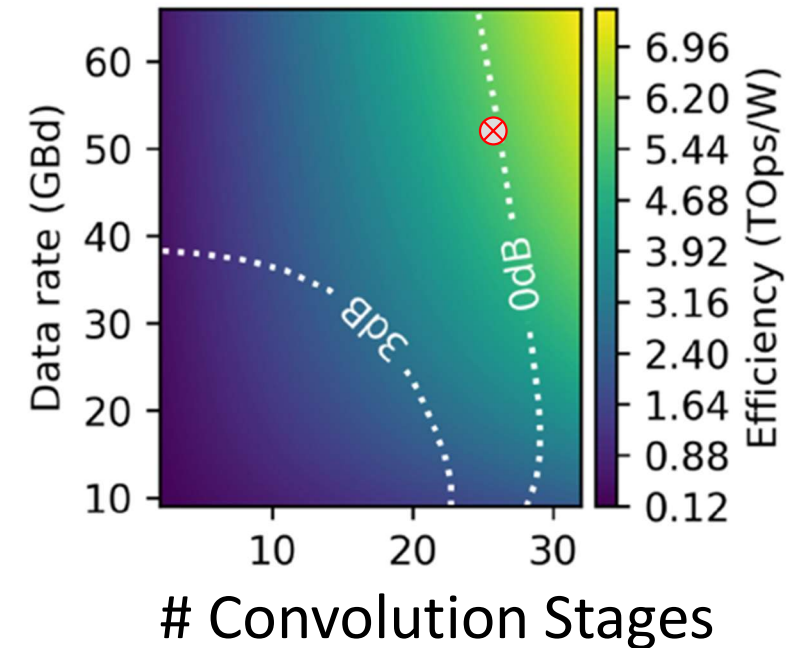
in-house Si platform + BTO



⊗ = most efficient device

0dB = power margin at receiver

commercial Si platform + BTO



low latency, high speed, fast kernel implementation

25 stage - devices with up to 6TOps/W are possible with today's available technology!

Acknowledgment



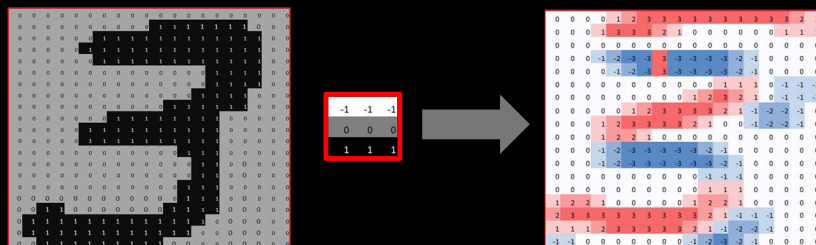
- my team at IBM
- Swiss National Science Foundation grant no. 175801 (Napreco)



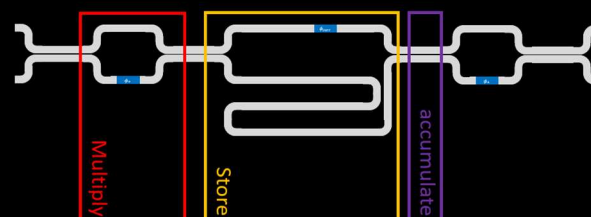
FONDS NATIONAL SUISSE
SCHWEIZERISCHER NATIONALFONDS
FONDO NAZIONALE SVIZZERO
SWISS NATIONAL SCIENCE FOUNDATION

Thank you and let's discuss!

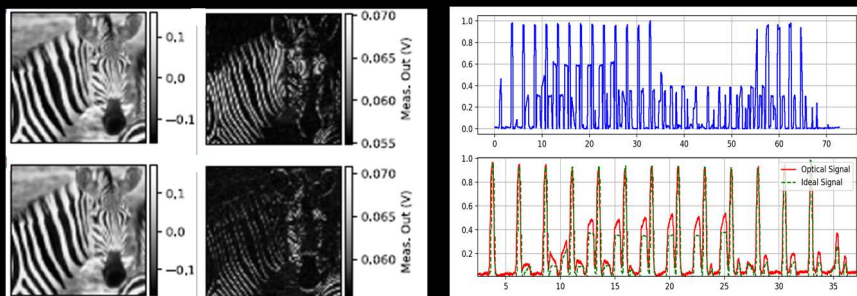
- convolutions for NNs



- device concept



- proof of principle



- benchmark

6TOps/s

