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Digital Oscillatory Neural Networks for AI Edge Applications

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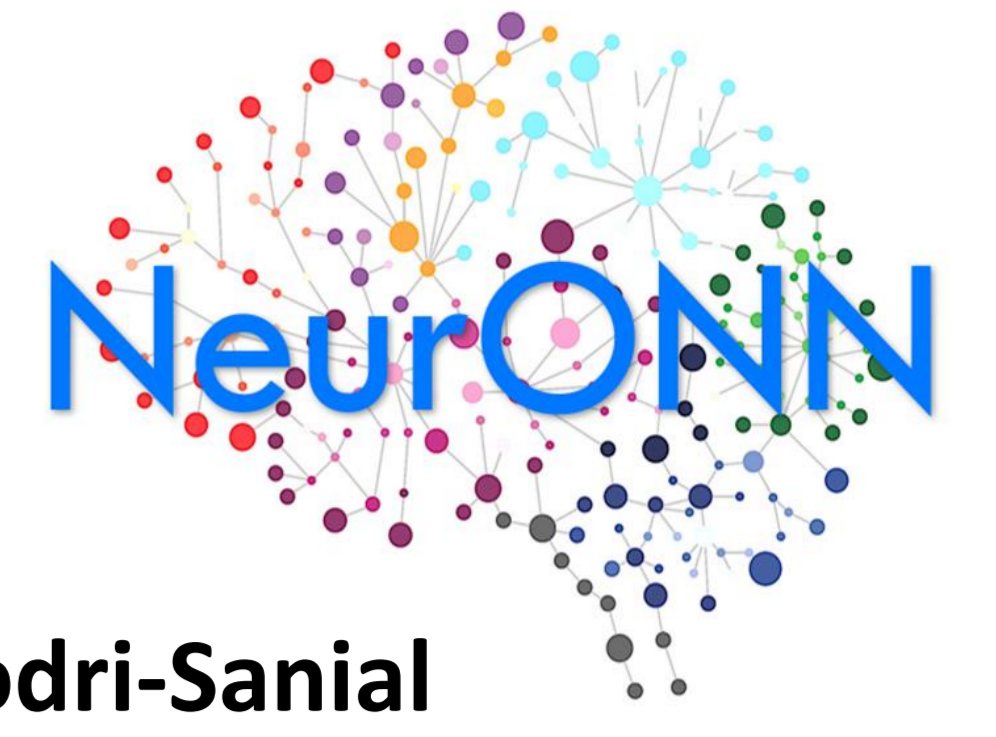
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Digital Oscillatory Neural Networks for AI Edge Applications

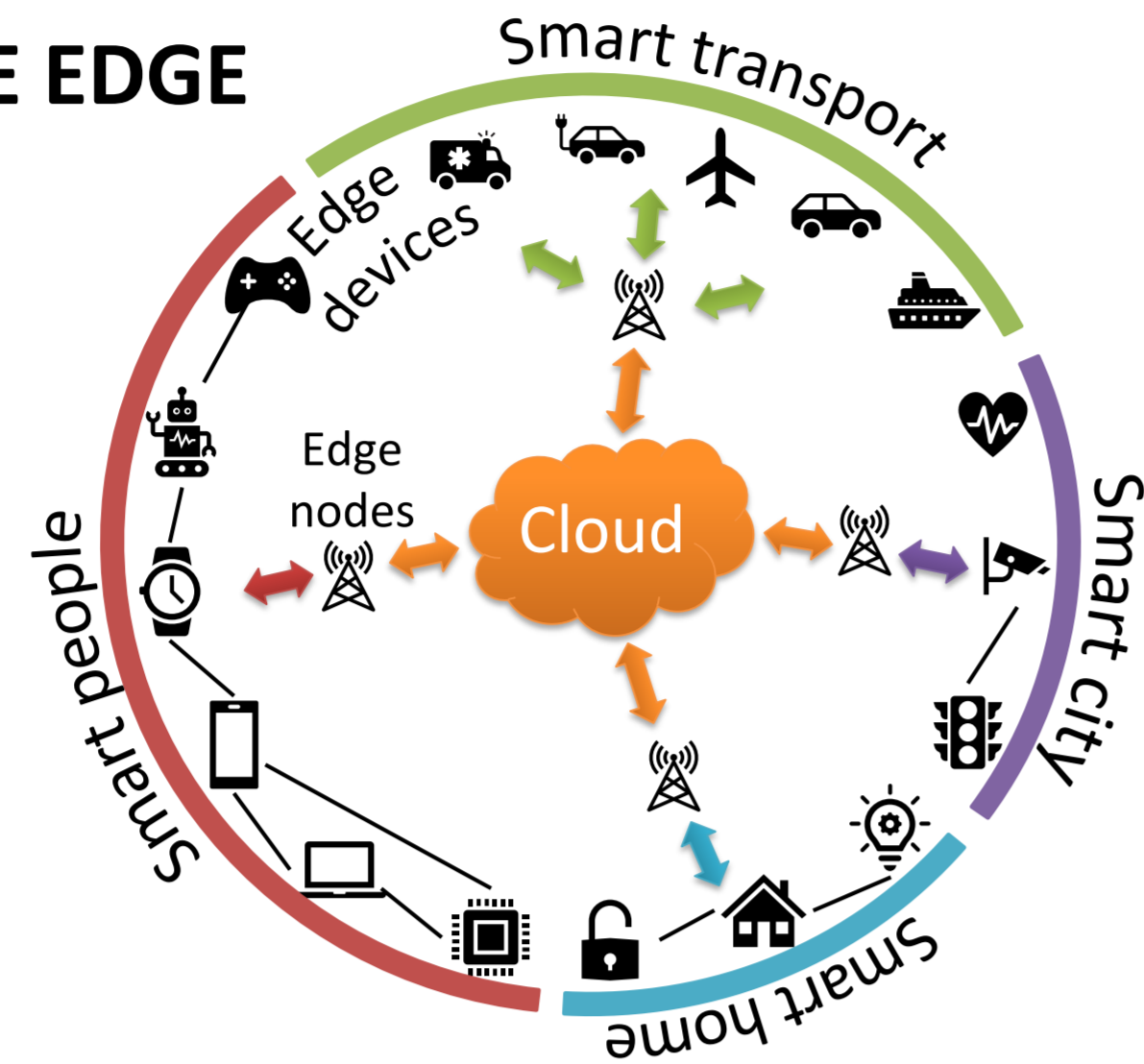


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MOTIVATION AND GOALS

ARTIFICIAL INTELLIGENCE AT THE EDGE

- Constraints:
 - Bandwidth (inference/second)
 - Latency (frames/second)
 - Privacy concerns
 - Power consumption



NEUROMORPHIC COMPUTING

- Support online learning
- Fast and efficient inference
- Low power consumption
- Scalability
- Low cost

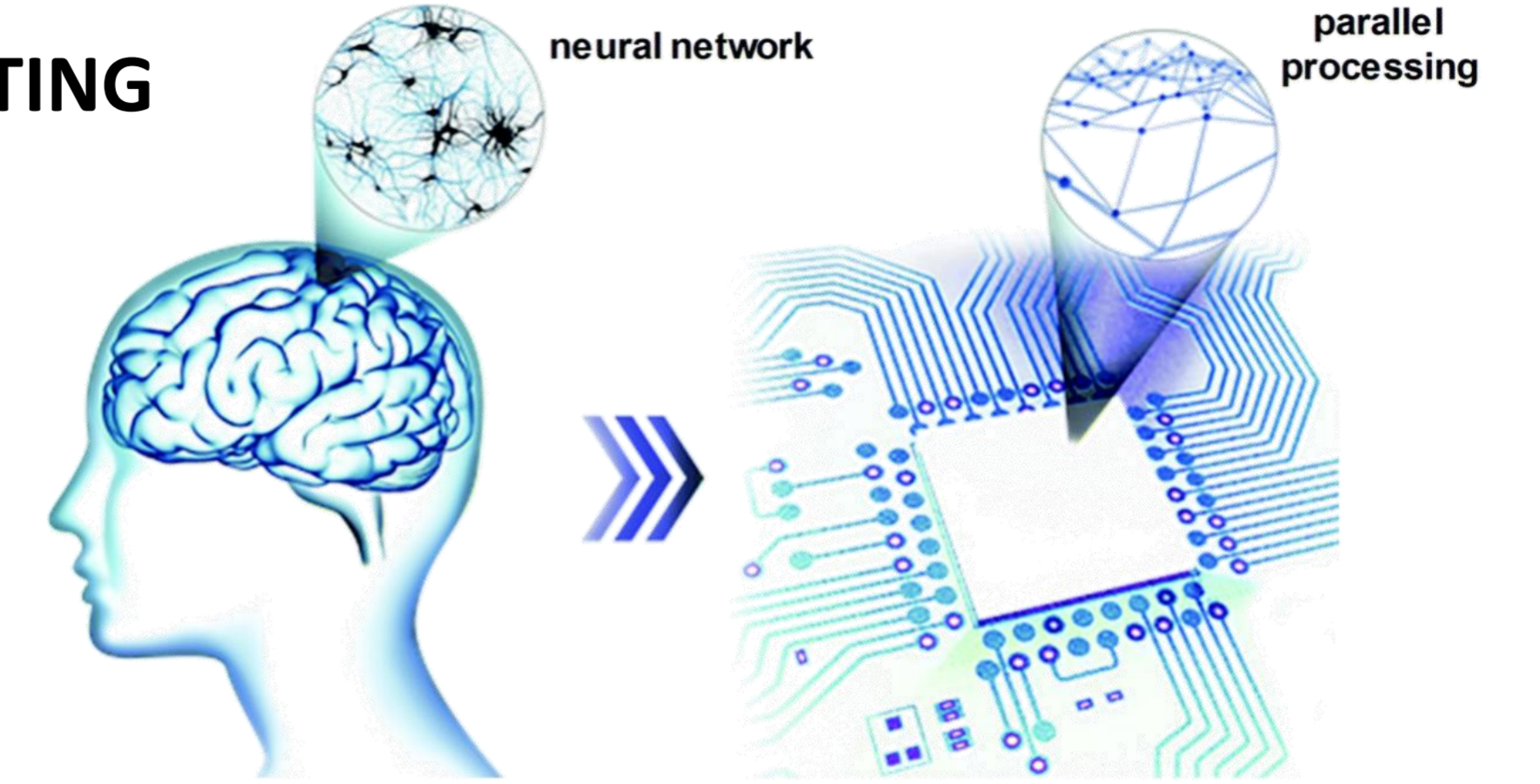
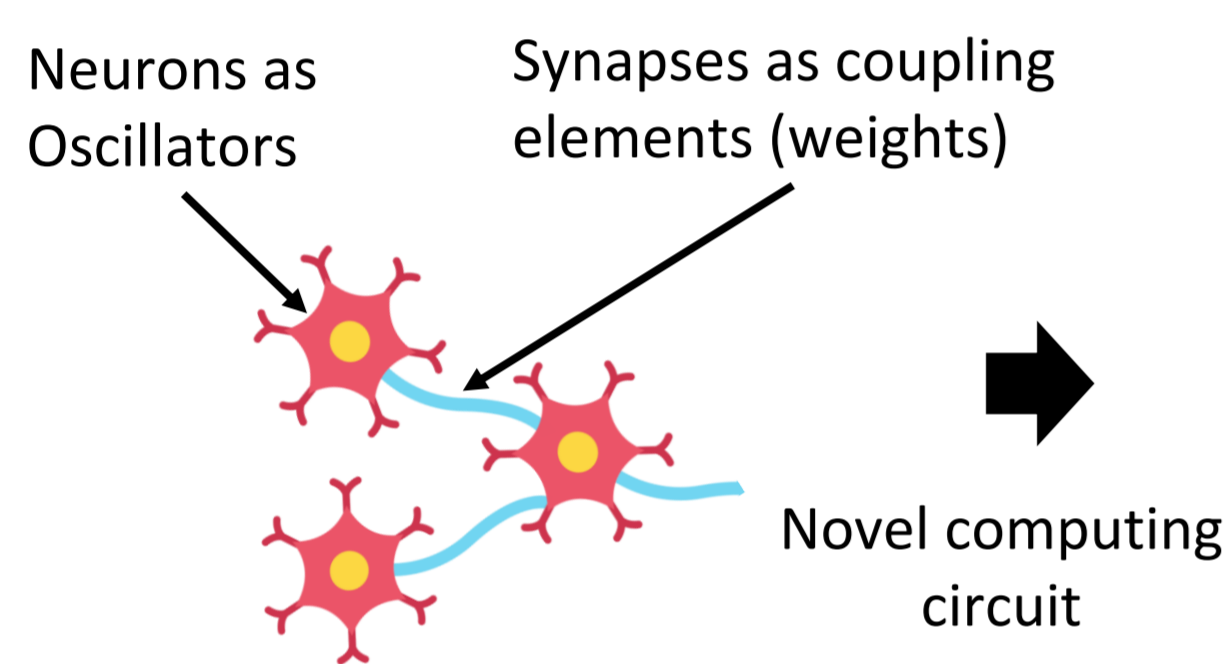


Image: Kim S., Lee Y., Kim HD., Choi SJ. Parallel weight update protocol for a carbon nanotube synaptic transistor array for accelerating neuromorphic computing, Nanoscale, 2020

OSCILLATORY NEURAL NETWORKS

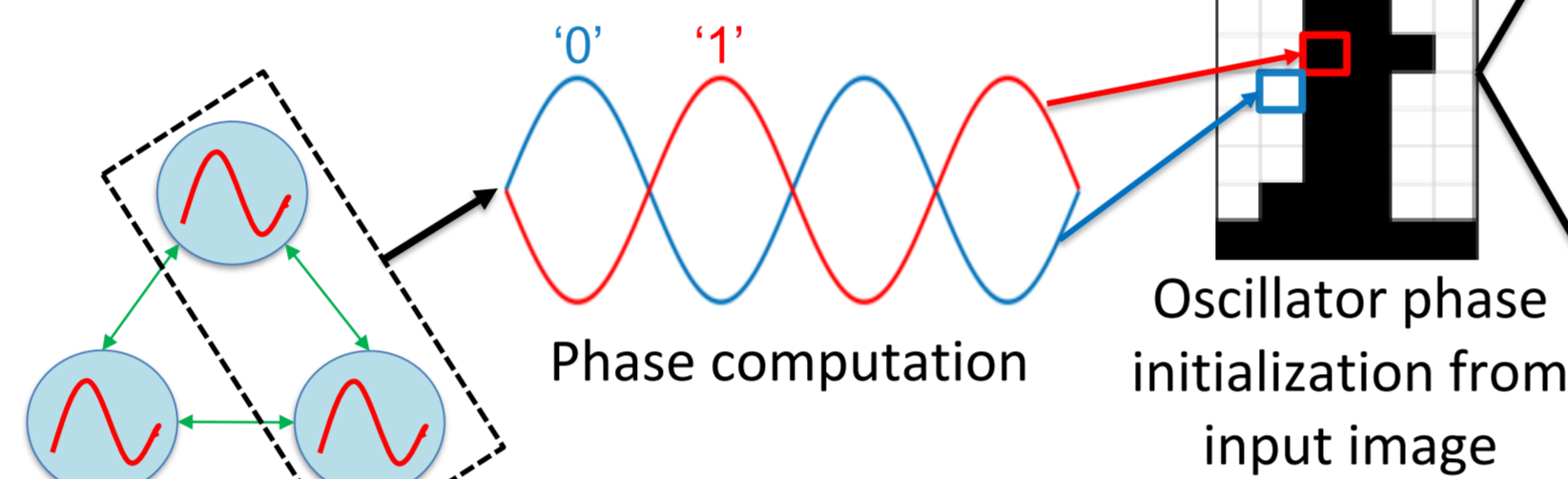
PHASE COMPUTING PARADIGM [1]

- Brain-inspired computing paradigm
- Neurons are oscillators
- Synapses are coupling elements between oscillators
- Information encoded in oscillators' phases

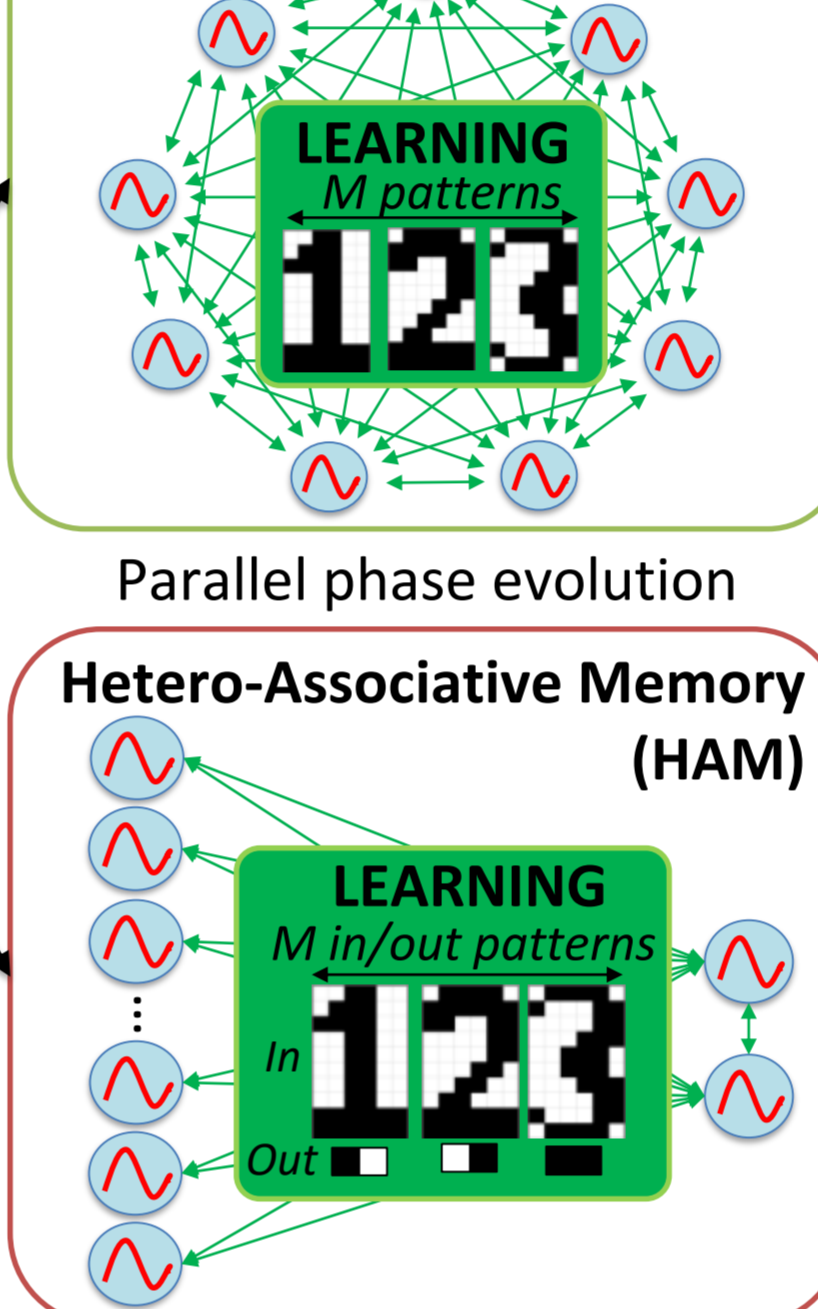


ASSOCIATIVE MEMORY

- Learn patterns
- Associate corrupted input with correct output

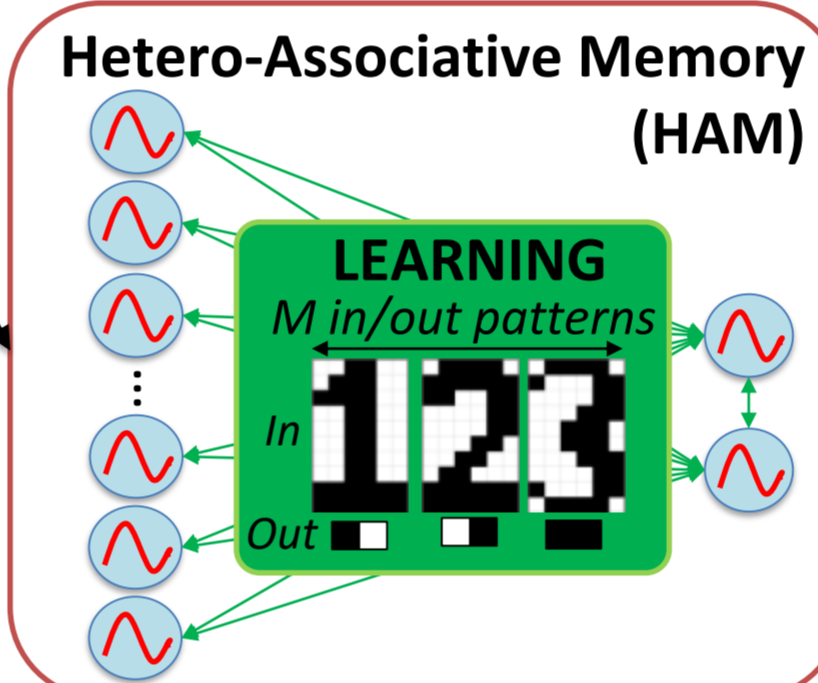


Auto-Associative Memory (AAM)



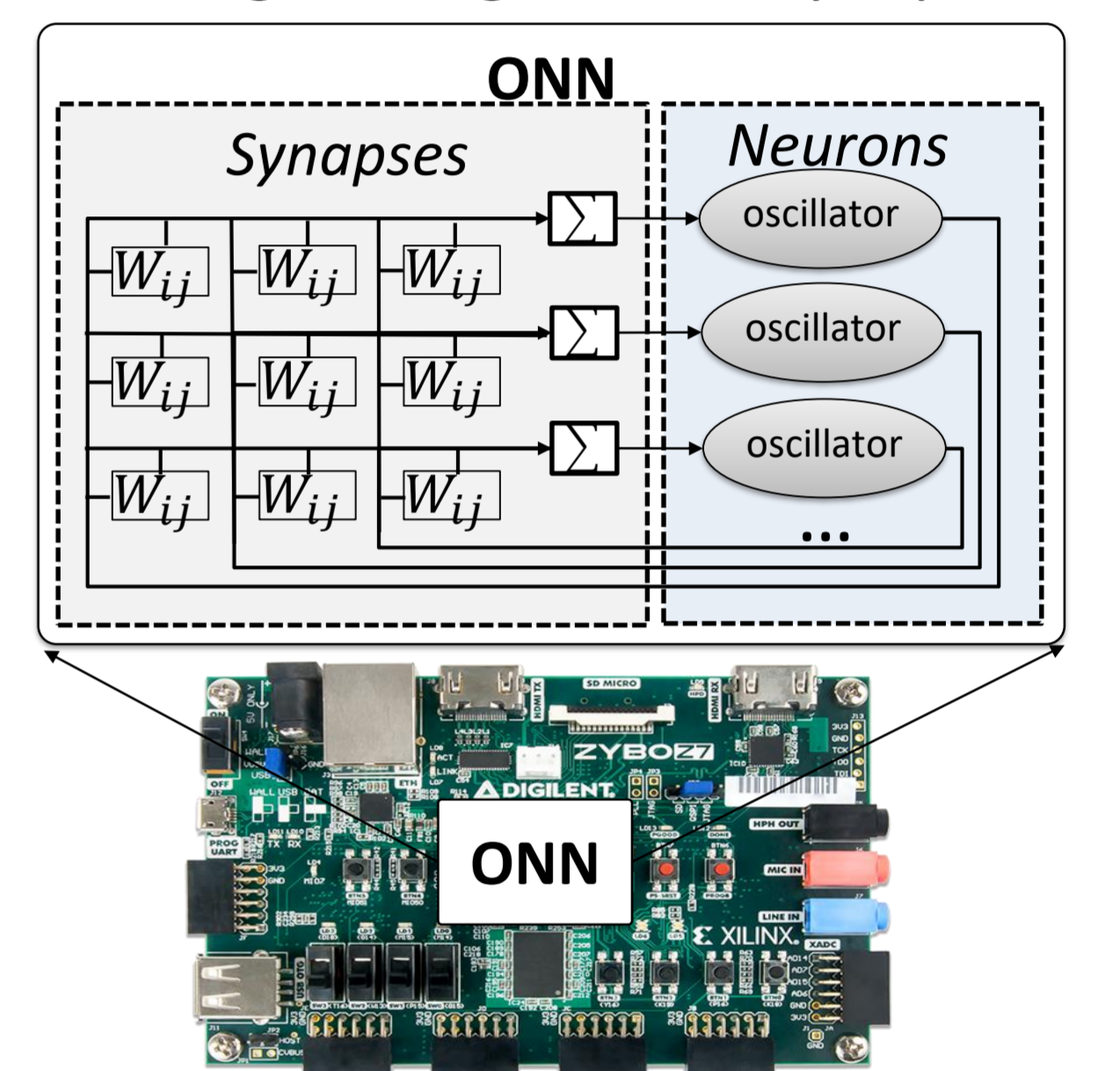
Parallel phase evolution

Phase measurement



FPGA IMPLEMENTATION [2]

- Digital oscillators
- 5-bits signed registers as synapses



APPLICATIONS AND USE CASES

Digits recognition (AAM) [2]

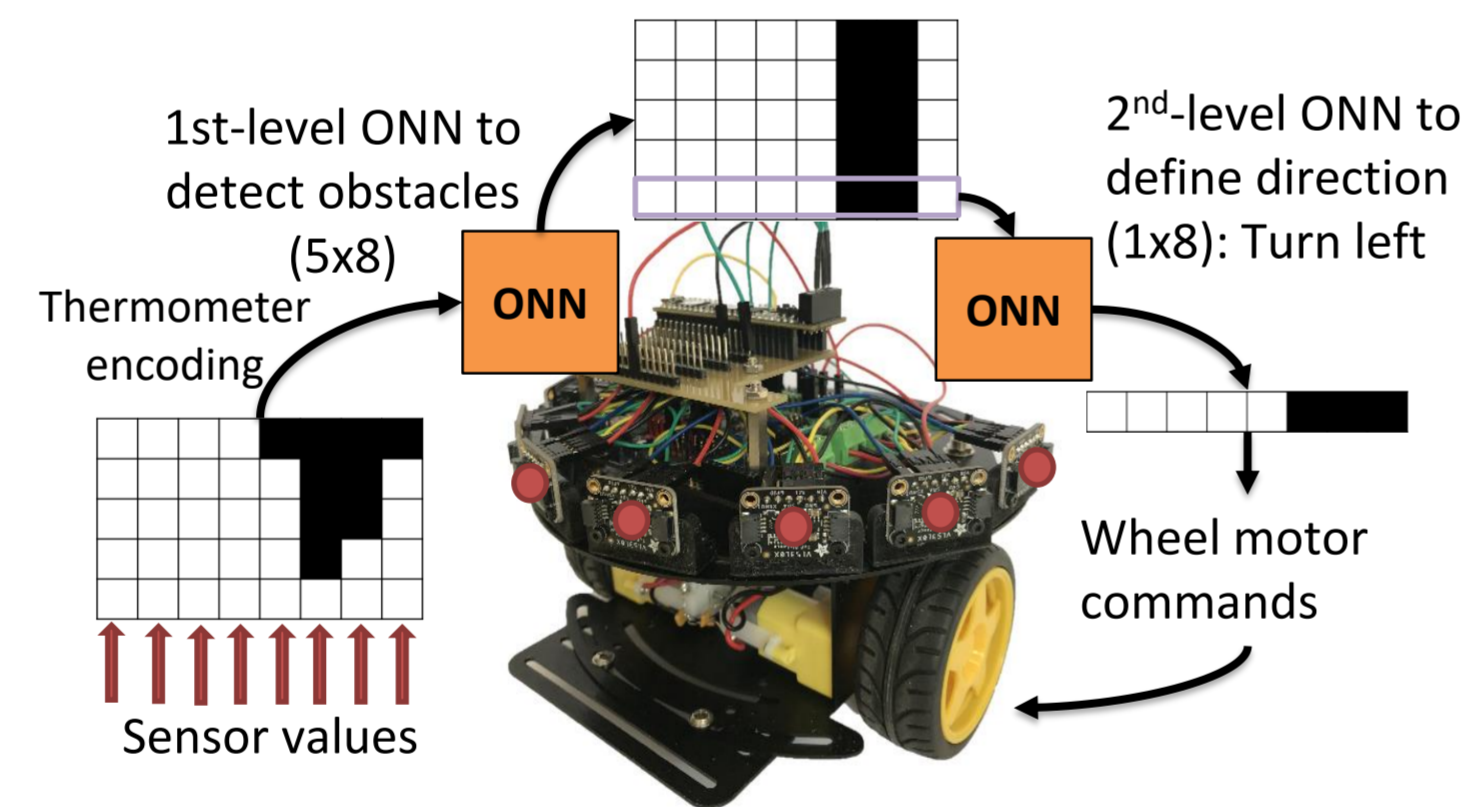
Camera stream to HDMI screen



ONN	10x6
LUTs	12%
Flip-Flops	2,6%
ONN Frequency	488 kHz
Init & comp time	13,2 us
FPS	75000
Accuracy (25 test images)	80 %

Obstacle avoidance (AAM) [3]

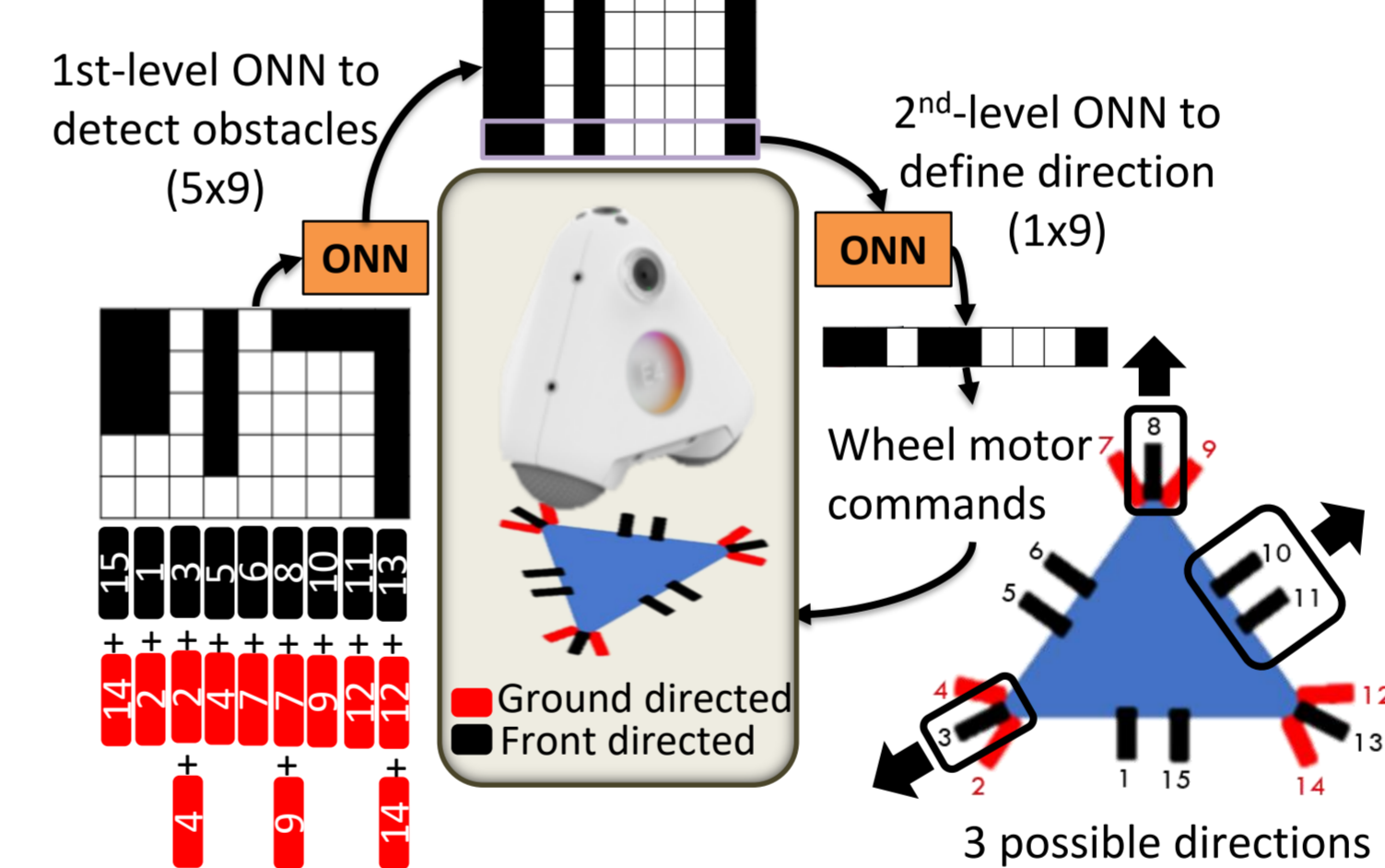
with an Arduino robot



ONN Performances	5x8	1x8	Full system performances (FPGA frequency: 12 MHz)	
#Training Patterns	256	16	8-sensor measurement	18 ms
LUTs (33 280)		11,5%		
Flip-Flops (41 600)		5,4%	FPS	40
ONN freq. (KHz)	187,5	187,5	Battery	6V/2850mAh
Init & Comp time	24 us	17 us	Current cons.	700 mA
Accuracy	100 %	74 %	Robot life time	4h

Obstacle avoidance (AAM) [4]

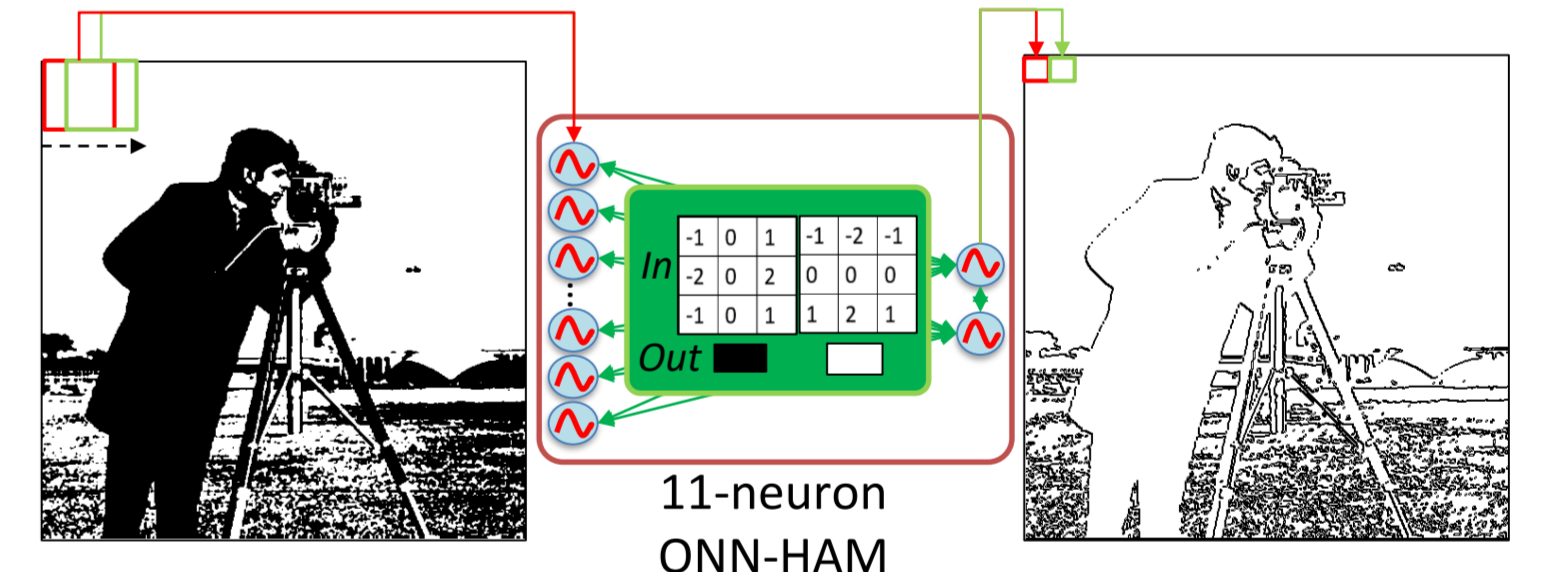
with the industrial robot E4 from A.I.Mergence



ONN Performances	5x9	1x9	Full system performances (FPGA frequency: 12 MHz)	
#Training Patterns	512	64	15-sensor measurement	27 ms
LUTs (33 280)		20,07 %		
Flip-Flops (41 600)		7,74 %	FPS	30
ONN freq. (KHz)	187,5	187,5	Robot life time estimation	2h/3h
Init & Comp time	27 us	17 us		
Accuracy	100 %	100 %		

Image edge detection (HAM) [5]

Camera stream to HDMI screen (10x6 ONN)



Single ONN characteristics	
Input - output size	9 - 2 neurons
ONN Frequency	2,7 MHz
Init - Comp time	240 ns - 1 or 2 us
Resources (LUTs)	402 (0,76 %)
Resources (Flip-Flops)	443 (0,42 %)
Estimation of full-image sequential processing with single ONN (computation time: 2 us)	
Image size	Processing time
28x28	1,5 ms
120x120	31,9 ms
512x512	582,6 ms

CONCLUSION

- Development of a proof of concept of the ONN computing paradigm with a digitally implemented ONN on FPGA
- Development of various demonstrators using the digital ONN on FPGA
 - Digits recognition from a camera stream
 - Obstacle avoidance on mobile robots from sensory data measurements
 - Image edge detection using ONN as HAM

REFERENCES

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- M. Abernot, T. Gil, A. Todri-Saniai. Oscillatory Neural Network as Hetero-Associative Memory for Image Edge Detection. *NICE workshop*, 2022.

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