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## Digital Oscillatory Neural Networks for AI Edge Applications

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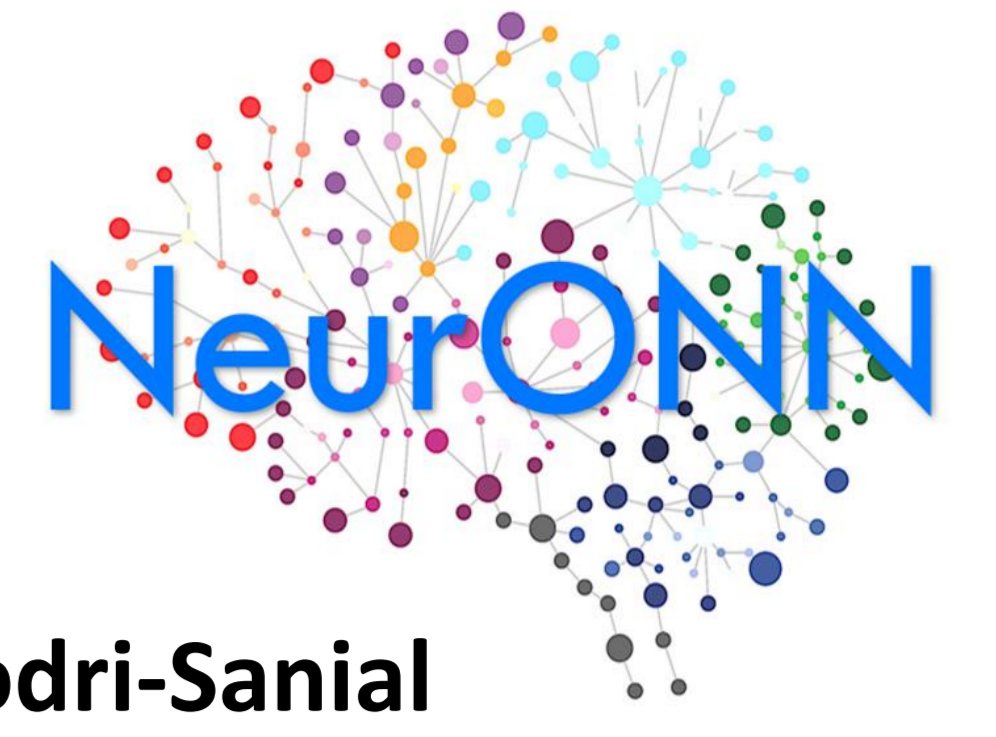
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# Digital Oscillatory Neural Networks for AI Edge Applications

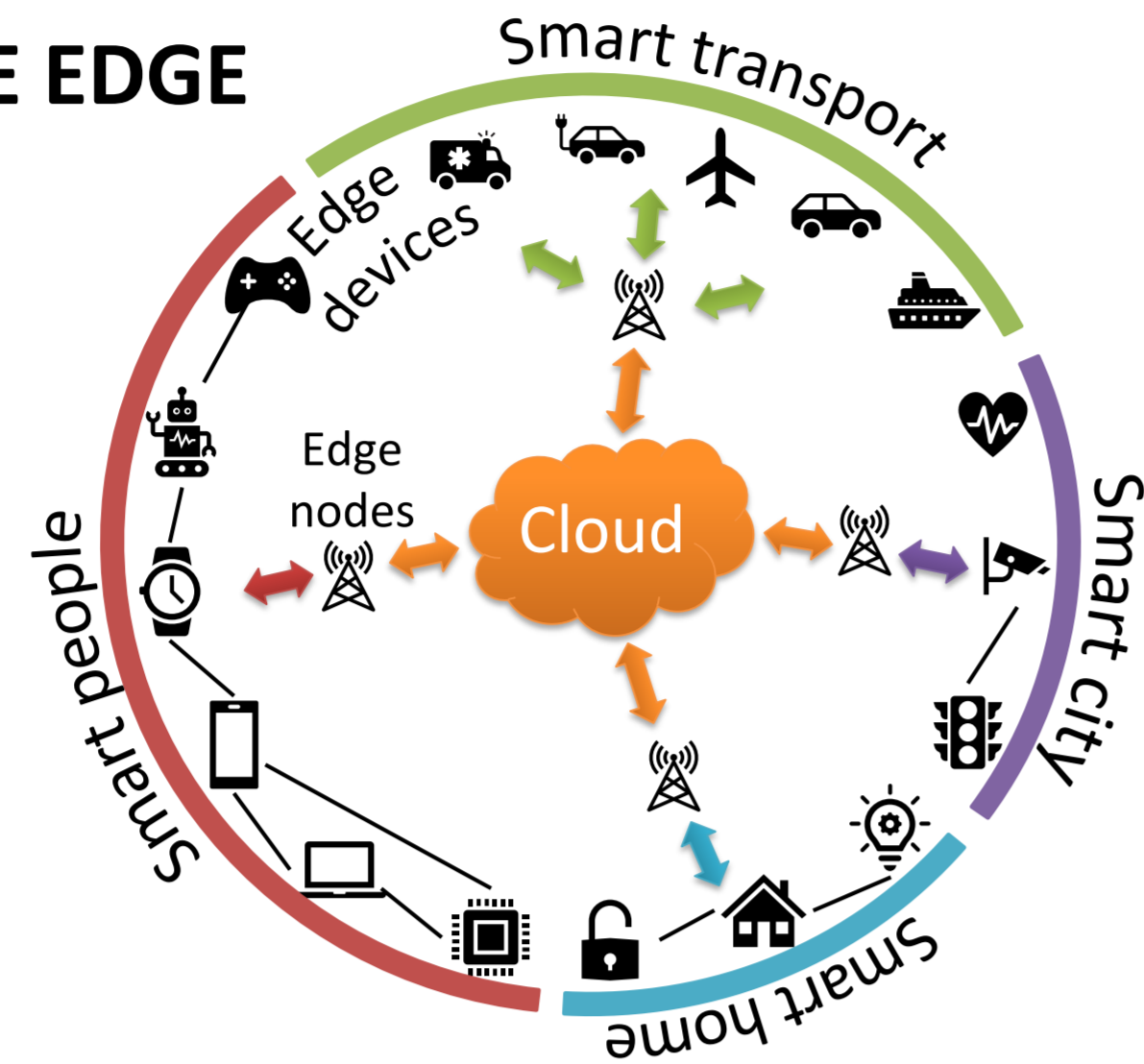


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## MOTIVATION AND GOALS

### 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

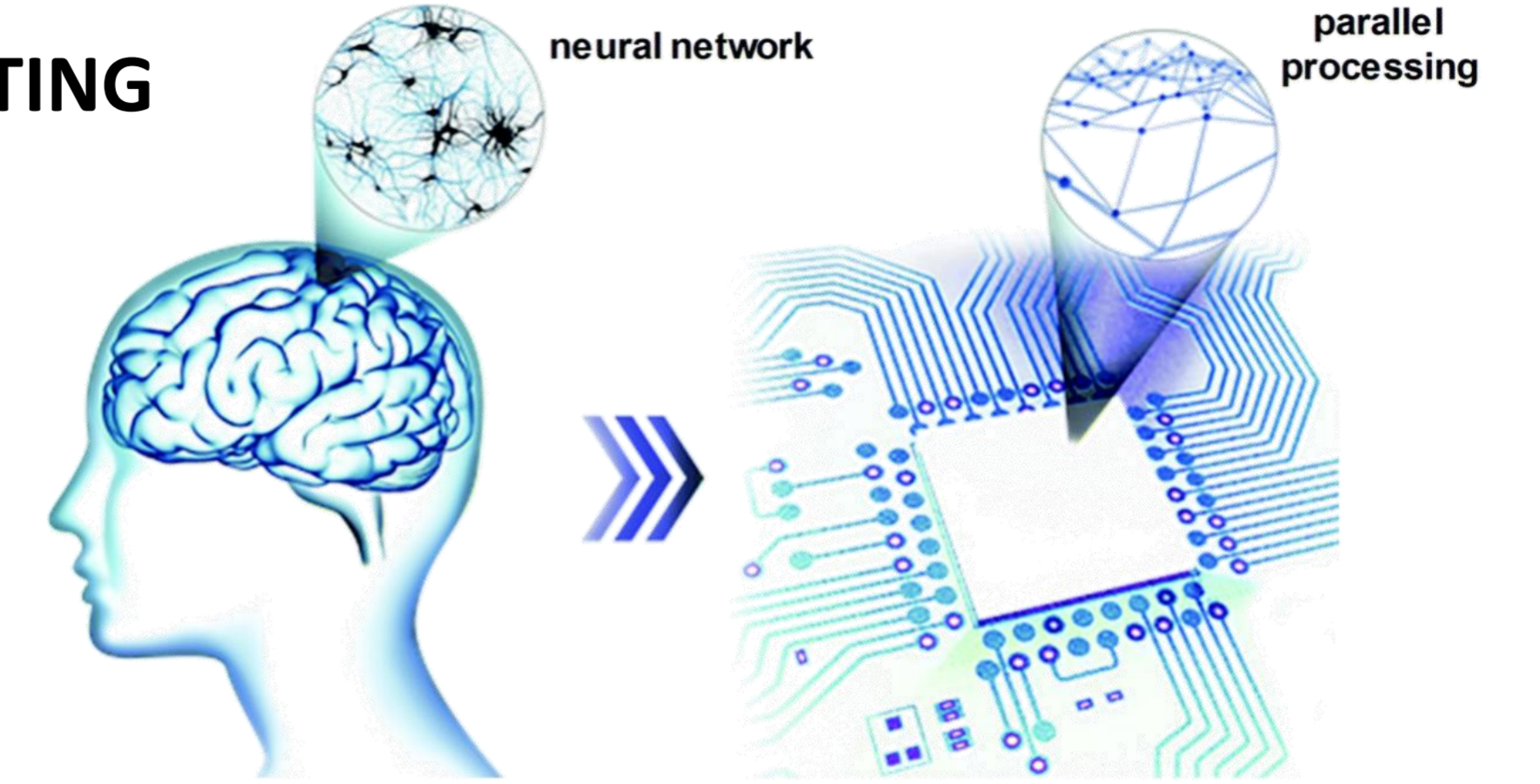
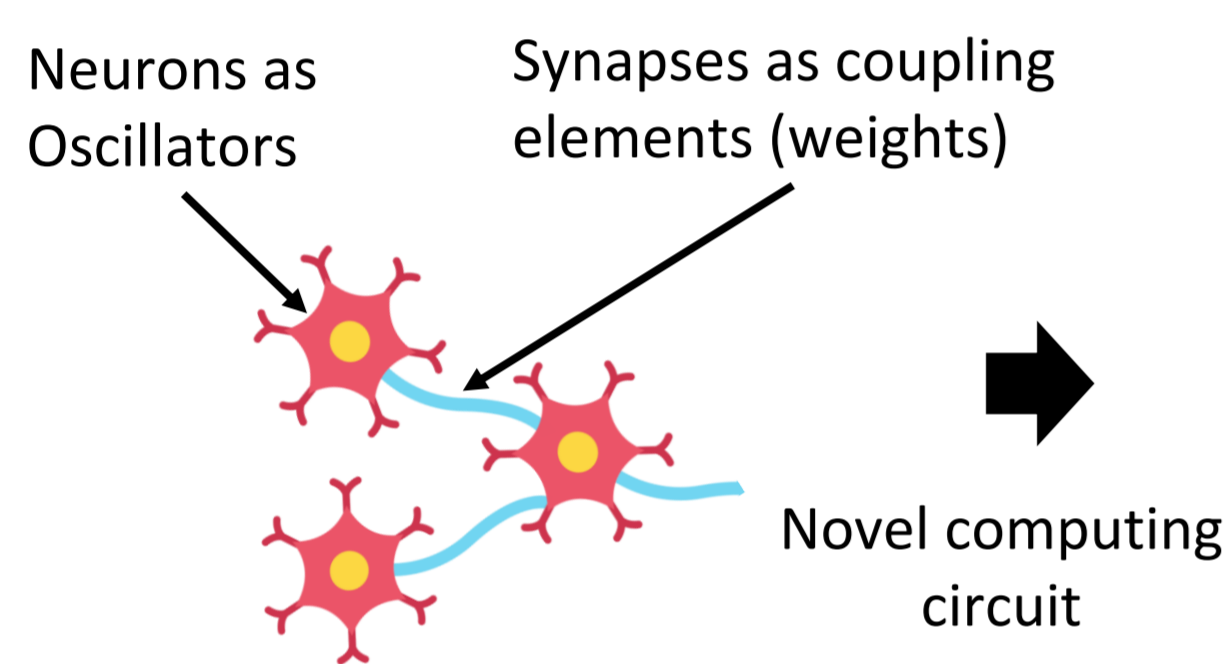


Image: Kim S., Lee Y., Kim HD., Choi SJ. Parallel weight update protocol for a carbon nanotube synaptic transistor array for accelerating neuromorphic computing, Nanoscale, 2020

## OSCILLATORY NEURAL NETWORKS

### PHASE COMPUTING PARADIGM [1]

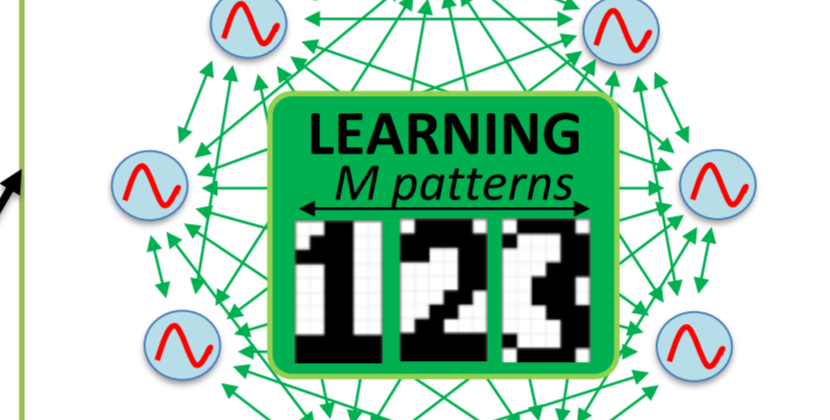
- Brain-inspired computing paradigm
- Neurons are oscillators
- Synapses are coupling elements between oscillators
- Information encoded in oscillators' phases



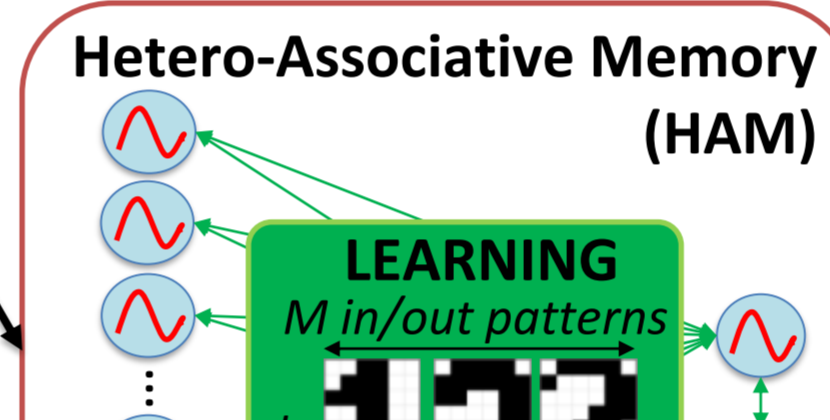
### ASSOCIATIVE MEMORY

- Learn patterns
- Associate corrupted input with correct output

#### Auto-Associative Memory (AAM)

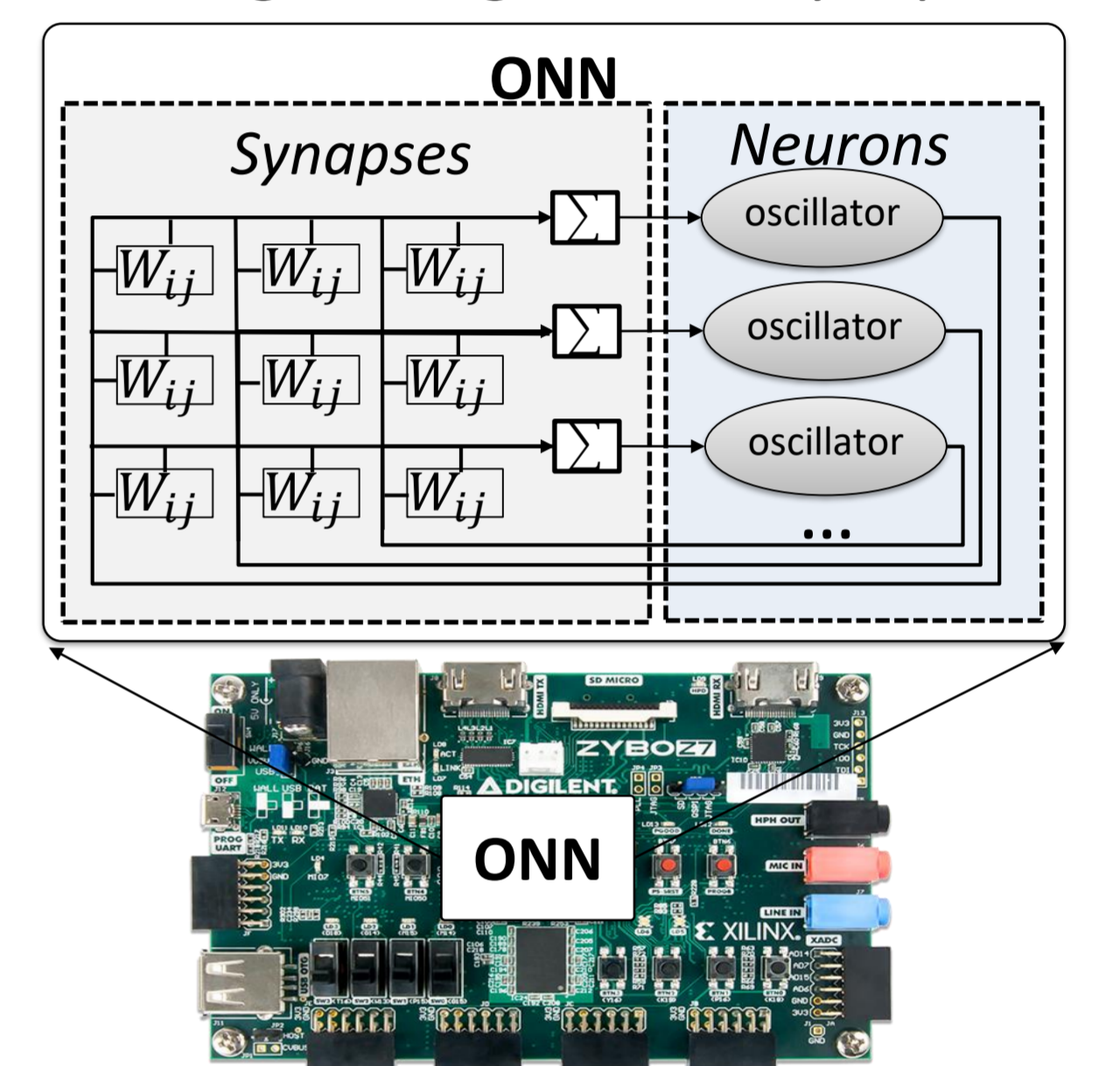


#### Hetero-Associative Memory (HAM)



### FPGA IMPLEMENTATION [2]

- Digital oscillators
- 5-bits signed registers as synapses



## APPLICATIONS AND USE CASES

### Digits recognition (AAM) [2]

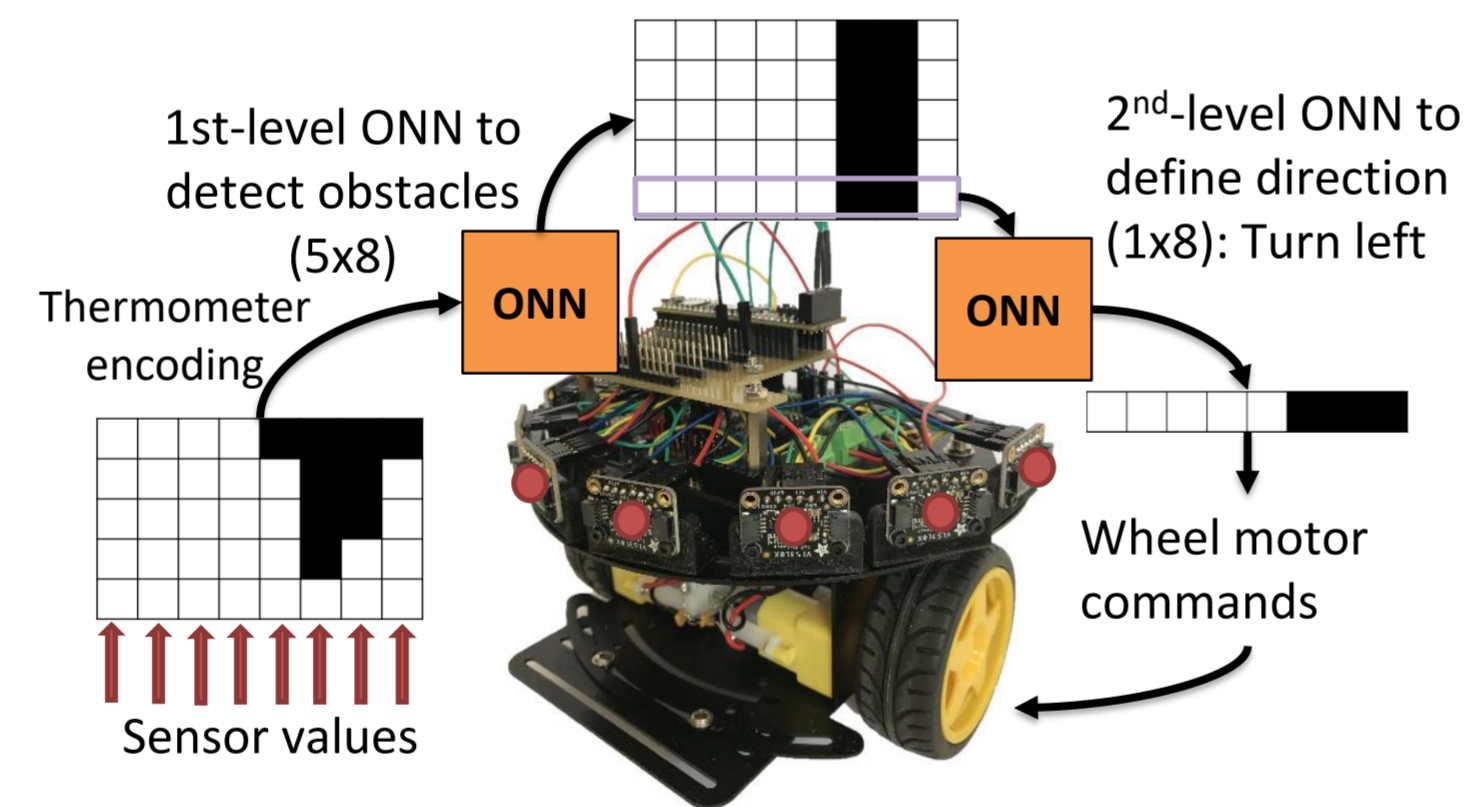
Camera stream to HDMI screen



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]

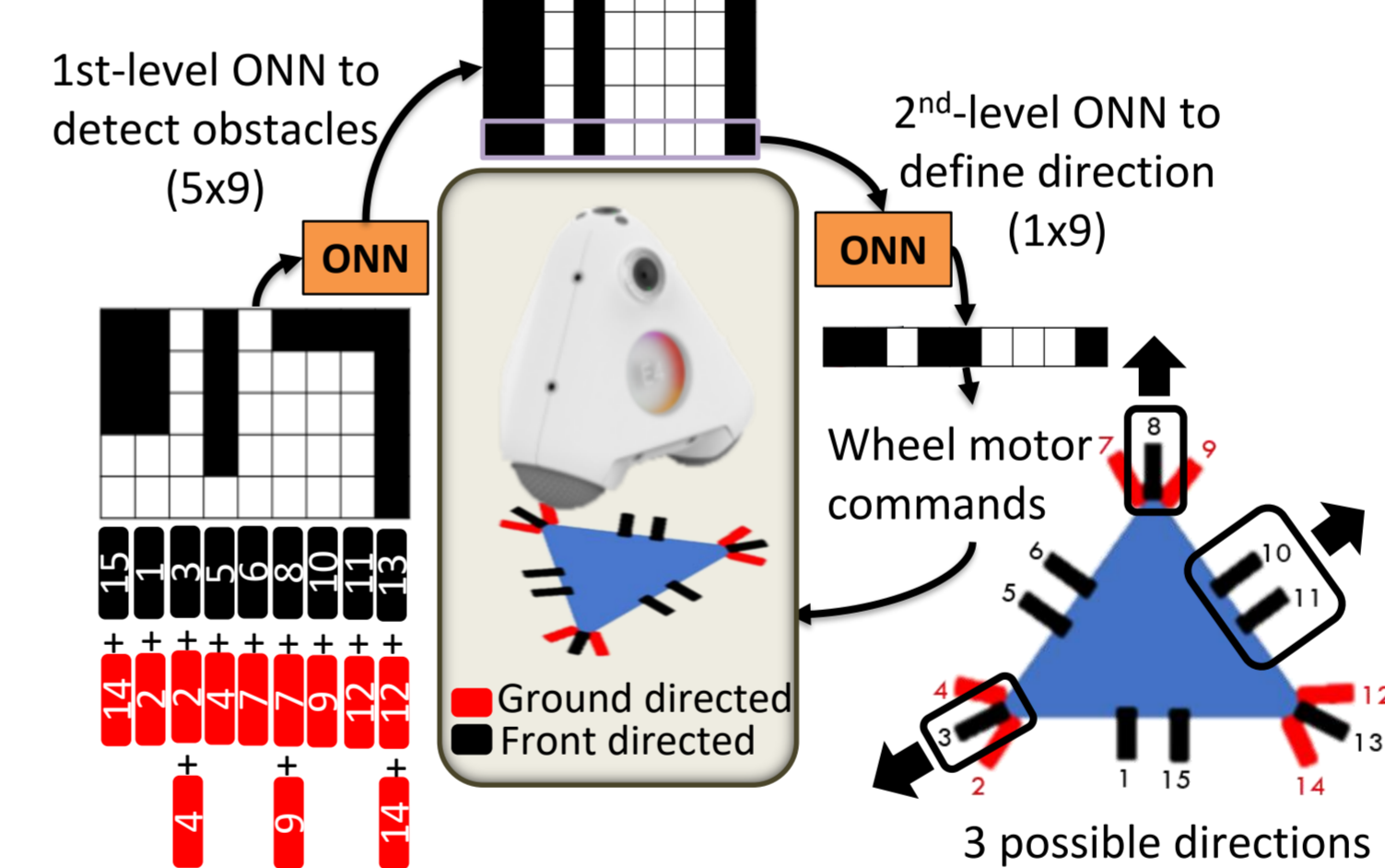
with an Arduino robot



ONN Performances	5x8	1x8	Full system performances (FPGA frequency: 12 MHz)
#Training Patterns	256	16	
LUTs (33 280)		11,5%	8-sensor measurement 18 ms
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]

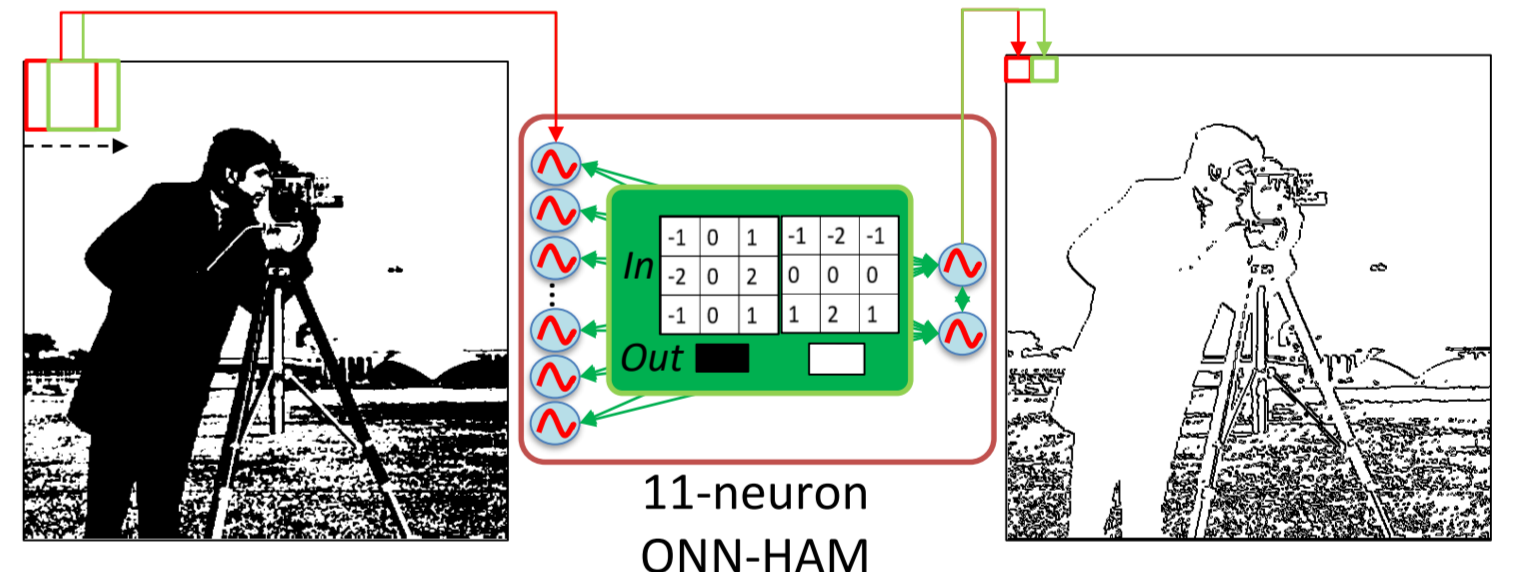
with the industrial robot E4 from A.I.Mergence



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

- A. Todri-Saniai, et al. EU H2020 NEURONN: Two-Dimensional Oscillatory Neural Networks for Energy Efficient Neuromorphic Computing. *EF ECS*, 2020.
- M. Abernot, et al. Digital Implementation of Oscillatory Neural Networks for Image Recognition Applications, *Front. In Neuroscience*, 2021.
- M. Abernot, et al.. Mobile Robot Obstacle Avoidance with Oscillatory Neural Networks on FPGA. *IBM-IEEE AI Compute Symposium*, 2021.
- M. Abernot, et al. Oscillatory Neural Networks for Obstacle Avoidance on Mobile Surveillance Robot E4. *IJCNN*, 2022.
- M. Abernot, T. Gil, A. Todri-Saniai. Oscillatory Neural Network as Hetero-Associative Memory for Image Edge Detection. *NICE workshop*, 2022.

## ACKNOWLEDGEMENTS AND FUNDING



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