

Next-generation IoT insights

VEDLIoT project overview and results

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Big Picture

Requirements

Smart Home

Industrial IoT

Automotive AI

Security & Safety

Applications



Modelling & Verification

Middleware

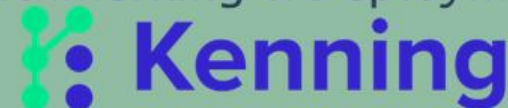
Toolchain



Emulation



Benchmarking & Deployment



Trusted Execution & Communication

Microserver & Accelerators



Xilinx Kria

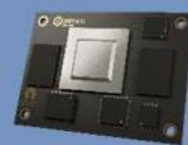


Coral SoM

COM-HPC
Xilinx Zynq
UltraScale+



Jetson AGX
NVIDIA Xavier



RPI CM4
ARVSOM

SMARC
Xilinx Zynq
UltraScale+



Monitoring

Hardware Platforms

Embedded/
Far Edge



u.RECS

Near Edge

t.RECS



Cloud

RECS|Box



RISC-V extensions

Safety & Robustness

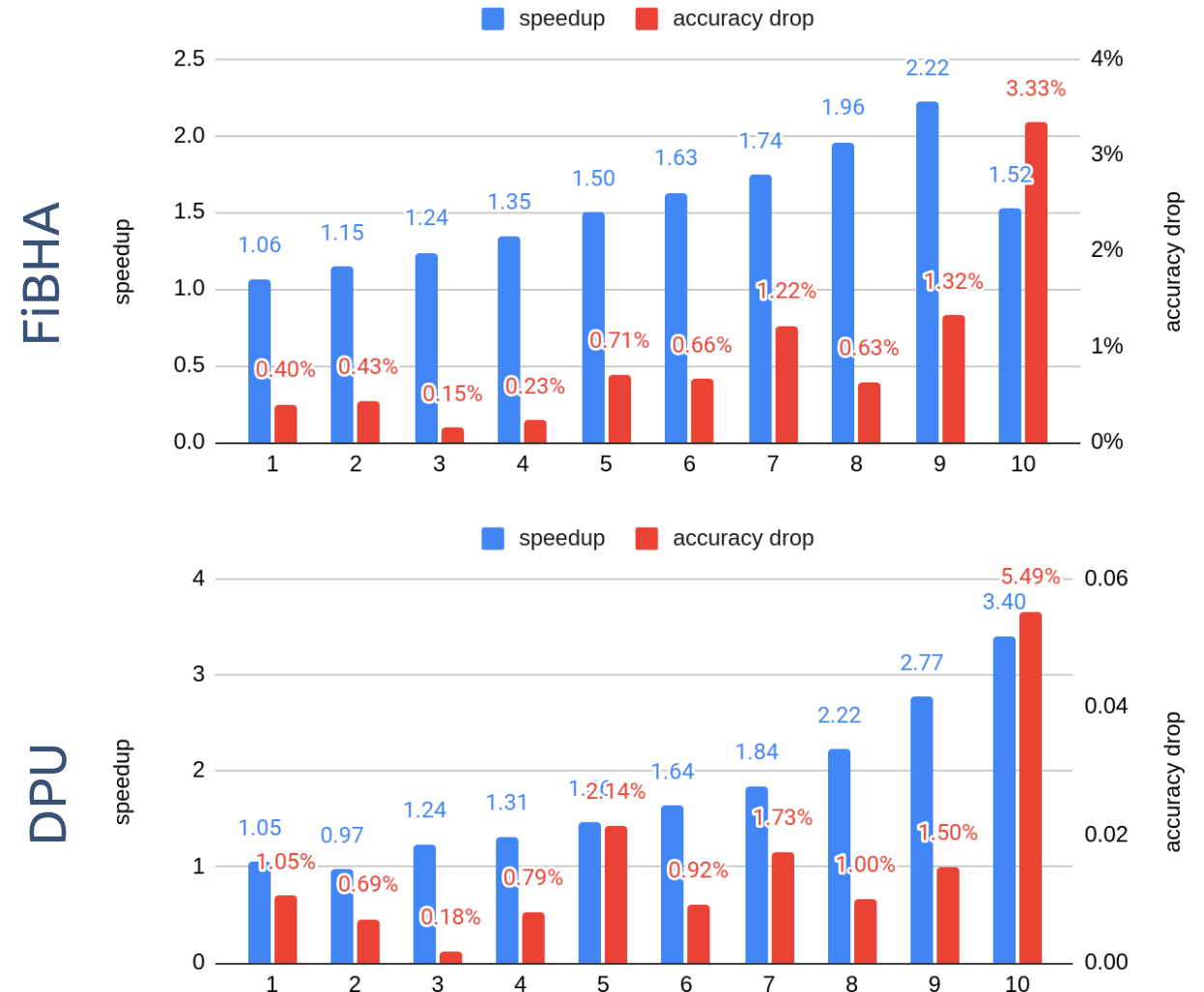
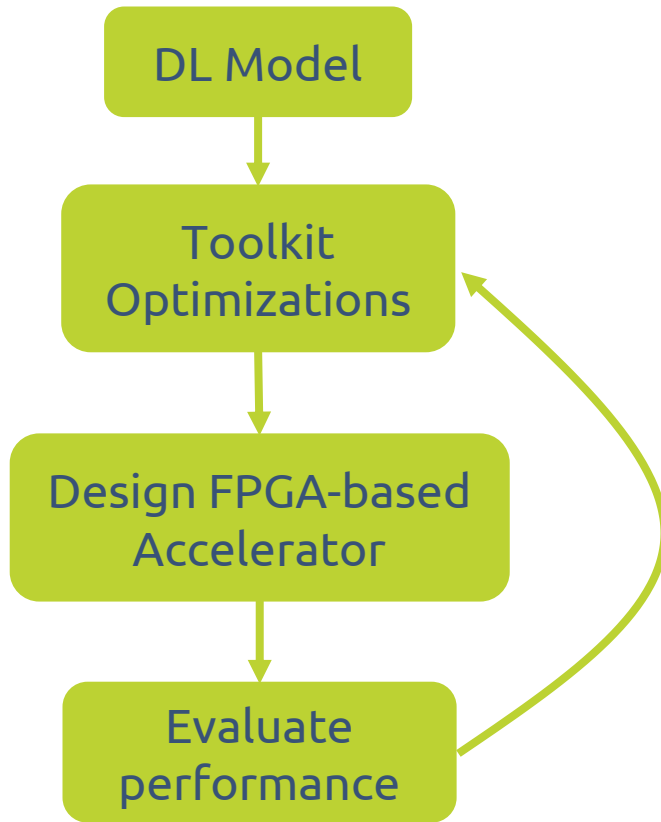
VEDLiOT Hardware Platform

CPU	SMARC 2.1 Intel Atom	COM-HPC Server Size D Intel Xeon D	COM Express ARM v8 Server SoC Hi1616	COM Express AMD EPYC 3451
GPU SoC	Jetson nano NVIDIA Orin NX	COM-HPC Size B NVIDIA Orin AGX	Jetson TX2 NVIDIA Tegra X2	
FPGA SoC	SMARC Xilinx Zynq UltraScale+	COM-HPC client Xilinx Zynq UltraScale+	COM Express Intel Stratix 10	COM Express Xilinx Zynq 7045
ML SoC	M.2 PCIe Hailo-8	M.2 PCIe Hailo-8	Grayskull Tenstorrent	
Far Edge Computing		Near Edge Computing		Cloud Computing



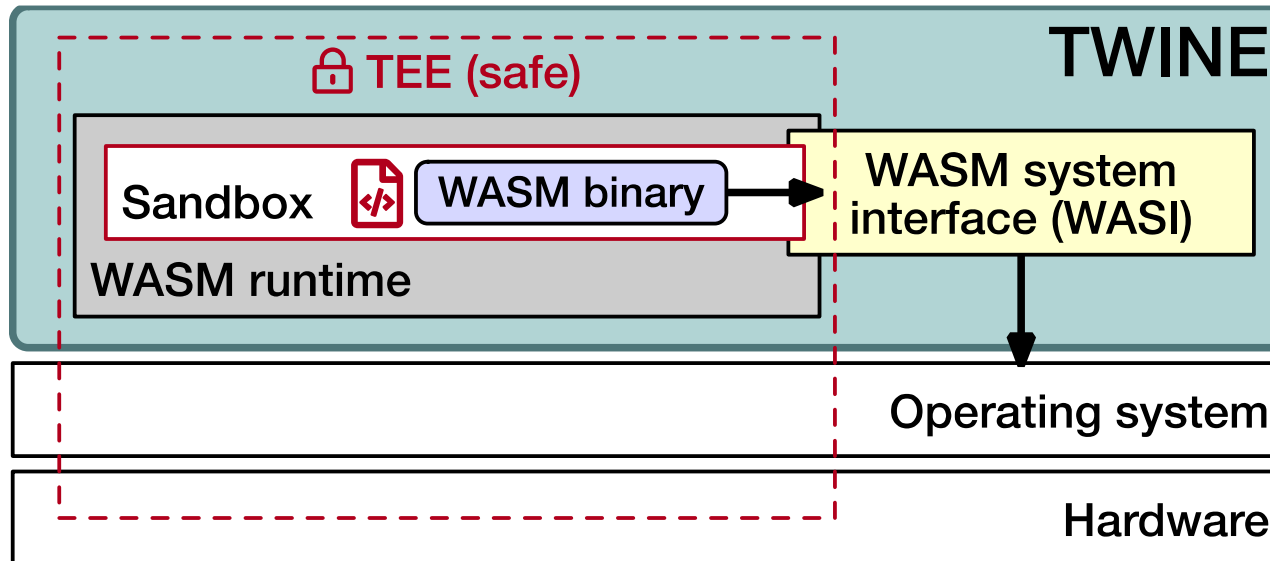
# Sites	>100K	>10K	100-10K	<100
Footprint	Custom	Compact (1RU)	Medium (2RU)	Large (3RU)
Power Budget	<30 W	< 500 W	500 W – 2 KW	> 2 KW
# Microserver	max 2	up to 3	up to 48	up to 144

Full co-design: DL accelerator and model optimizations



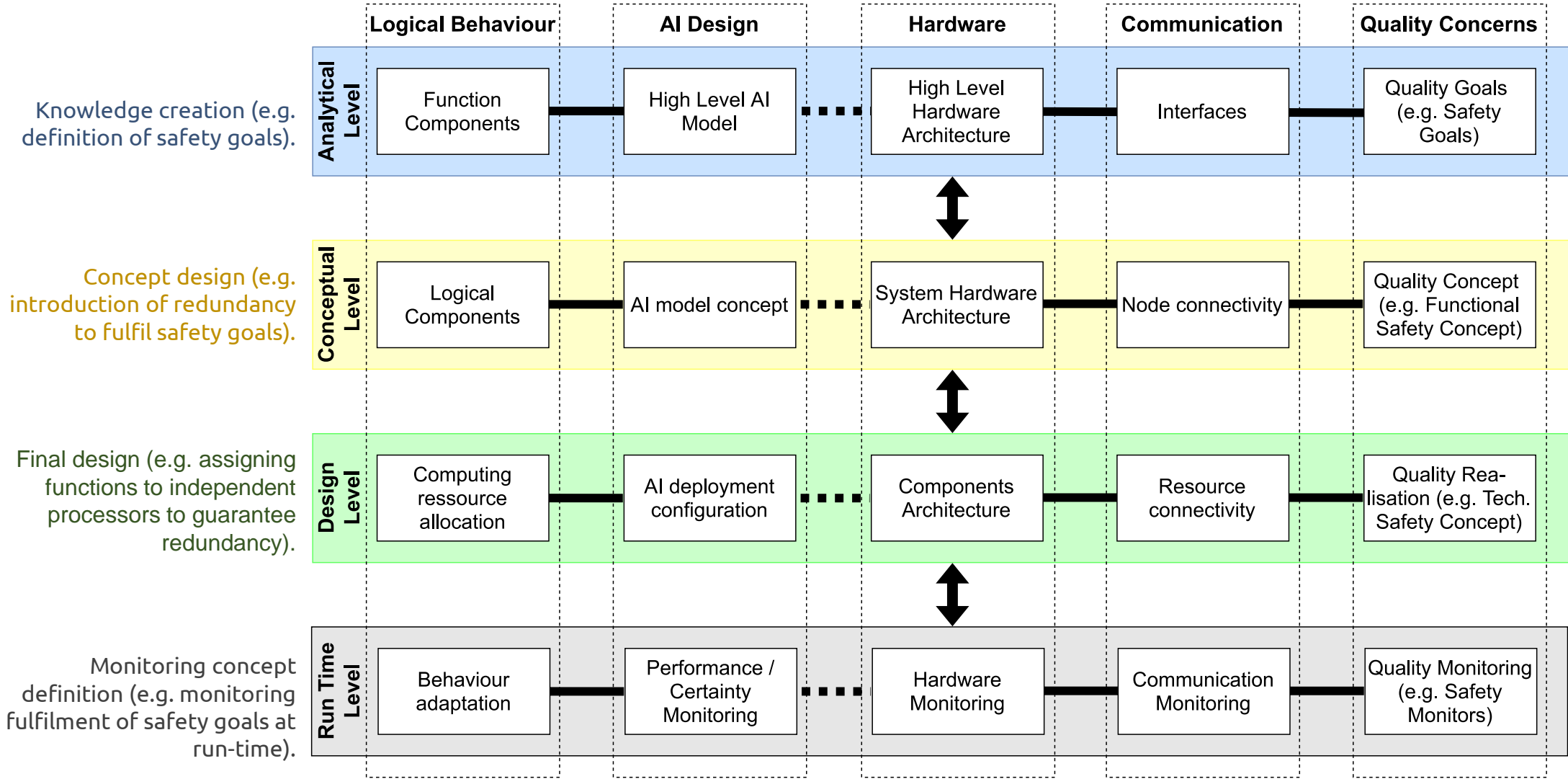
Security

- Common environment for running distributed applications
 - WebAssembly runtime + Trusted Execution Environment
 - Security for edge (and cloud) devices
- Advances on attestation
 - Better support for edge devices
 - Distributed (Byzantine fault-tolerant) attestation and configuration service

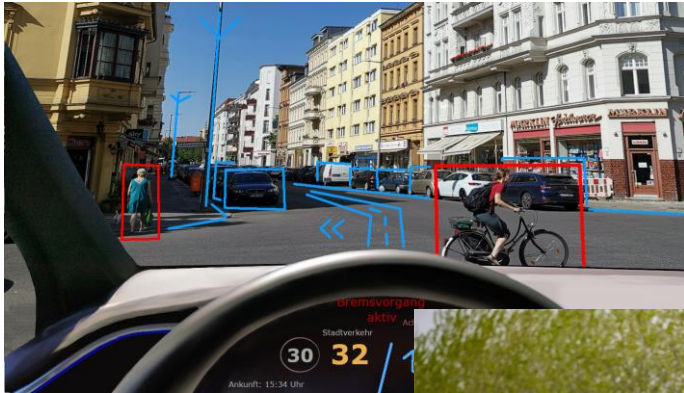


A compositional architecture framework for AIoT

Problem Space
 Solution Space



Use case: Automotive



**Challenge:
Distribution
of work**

- Focus on collision detection/avoidance scenario
- Improve performance/cost ratio – AI processing hardware distributed over the entire chain

Demonstrated distribution across embedded, edge, cloud with at least 2x eff. improvement

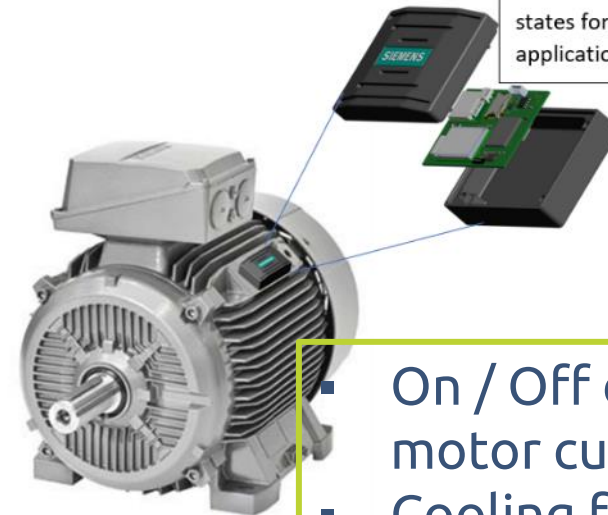
Use case: Industrial IoT – drive condition classification

VEDLiOT
Very Efficient Deep Learning in IoT

- Control applications need DL-based condition classification
 - On the edge device for low power consum
 - Suggestions for control and maintenance
- DL methods on all communication layers
 - DL in a distributed architecture
 - Dynamically configured systems
- Sensored testbench with 2 motors
 - Acceleration, Magnetic field, Temperature, IR-Cam (temperature), Current-Sensors, Torque

Challenge:
Low-power /
Efficiency

Edge devices with AI for sensing
communication and detection of complex
states for local safety and control
applications



- On / Off detection without motor current or voltage
- Cooling fault detection
- Bearing fault detection

Demonstrated > 2 years of battery
lifetime (> 7x eff. improvement)

Use case: Industrial IoT – Arc detection

- AI based pattern recognition for different local sensor data
 - current, magnetic field, vibration, temperature, low resolution infrared picture
- Safety critical nature
 - response time should be <10ms
 - AI based or AI supported decision made by the sensor node itself or by a local part of the sensor network



Combining the information from the IR-Camera and the magnetic field sensor to localize electric faults in power cabinets by deep learning methods



- 5G, Wi-Fi
- Magnetic Field sensor
- Vibration, Temperature
- IR-Camera

Specifications:

- Industrial temperature range (-20°C ... +85°C)
- Industrial batteries (rechargeable for ID-Tag)
- IP65 protection
- RoHS and IEC 61850-3 complaint
- Pre-certified wireless transceivers
- Target price: 100€ (ID-Tag)
- SIM on Chip*

Demonstrated > 99 % accuracy (> 20 x eff. improvement)

Challenge:
Accuracy

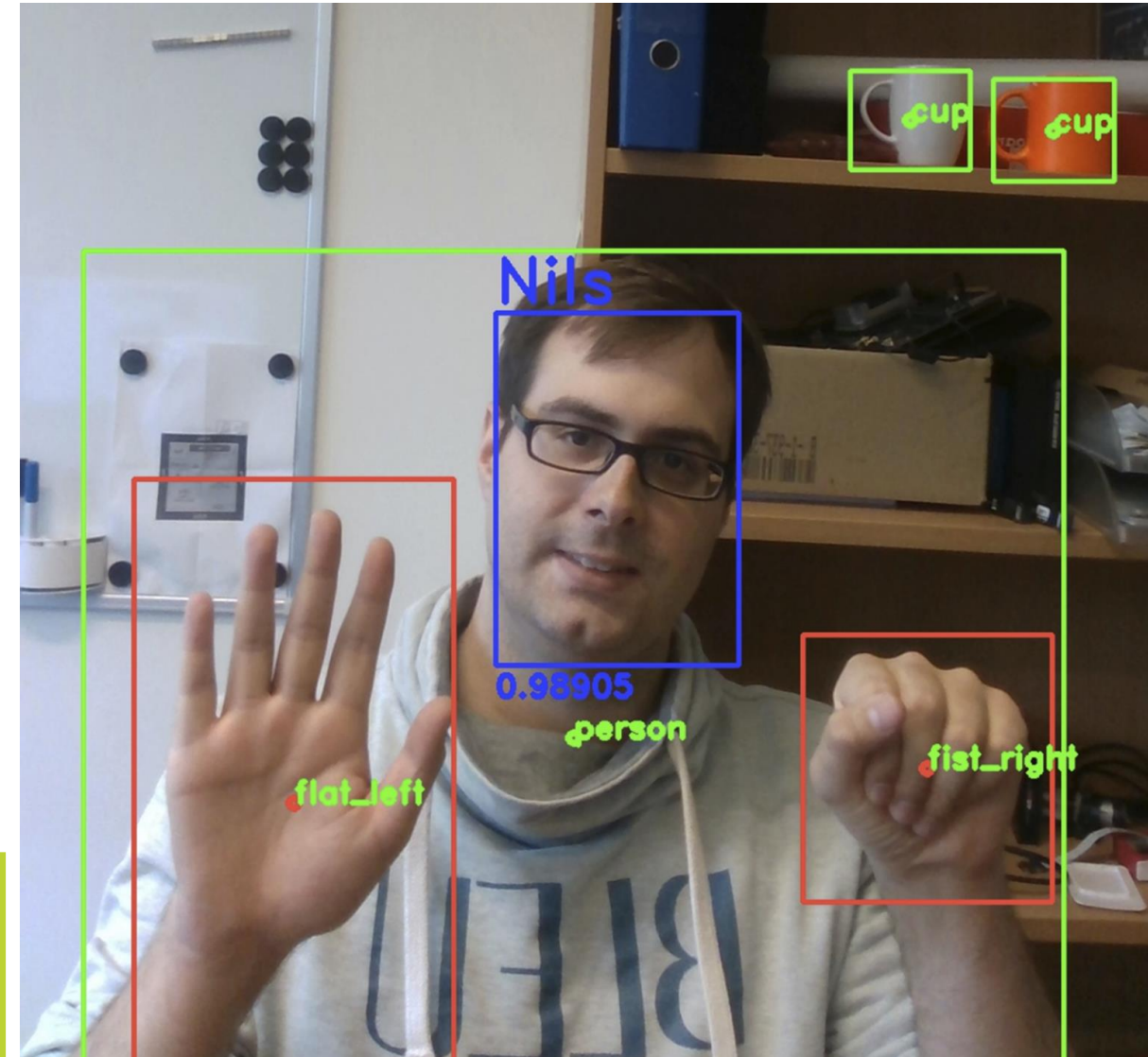
Use case: Smart Mirror – Neural Networks

- Face recognition
 - Mobilenet SSD trained on WIDERFACE dataset
- Object detection
 - YoloV3, Efficient-Net, yoloV4-tiny
- Gesture detection
 - YoloV4-tiny with 3 Yolo layers (usually: 2 layers)
- Speech recognition
 - Mozilla DeepSpeech

- AI Art: Style-Gan trained on works of arts
- Collect usage data in situation memory

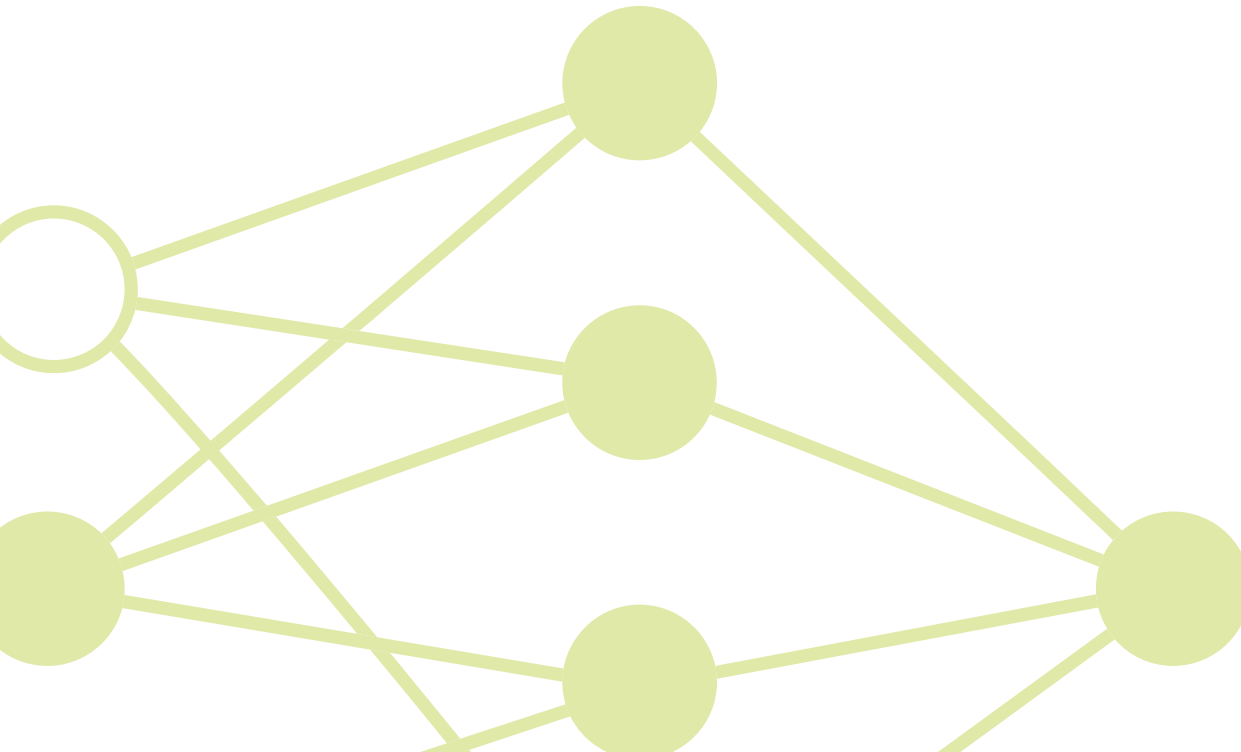
Challenge:
Data privacy,
Efficiency

Demonstrated local
processing (> 9 x eff.
improvement)



Summary

- VEDLIoT key results
 - Scalable, heterogenous, cognitive IoT hardware platform (Embedded – Edge – Cloud)
 - Full co-design approach, model optimization and extensive benchmarking
 - Secure environment for distributed applications, WebAssembly runtime + Trusted Execution Environment + Distributed (Byzantine fault-tolerant) attestation
 - Demonstrated efficiency improvement of at least on order of magnitude for wide range of use cases
- VEDLIoT lessons learned
 - Heterogeneity and reconfigurability is key for next generation AIoT platforms
 - Toolchain coverage for mapping applications to this heterogeneous hardware is vital
 - Security and robustness are mission critical for a broad spectrum of AIoT applications
 - Systematic requirements engineering for AIoT is vital to meet all requirements complying with regulatory constraints, such as the AI Act.



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Thank you for your attention.



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