

EMBEDDED AI LAB – EEMCS – UNIVERSITY OF TWENTE

UNIVERSITY OF TWENTE.

EMBEDDED AI

DSI SEMINAR BY SEBASTIAN BUNDA MSc

18 SEPTEMBER 2024

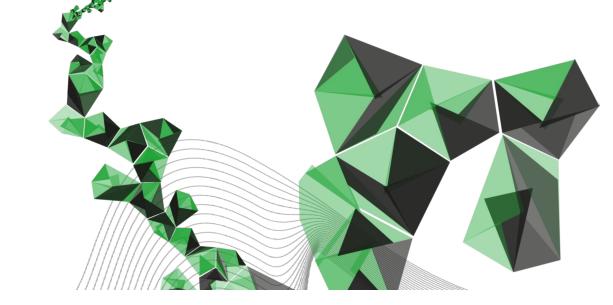
DSI EMBEDDED AI SEMINAR SERIES



WHO AM I?

- PhD Student @ DMB and CAES
 - Subject: Embedded Al
- Background in Electrical Engineering: 2015-2022
 - Specialisation Computer Vision & Biometrics @Twente





EMBEDDED AI



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ARTIFICIAL INTELLIGENCE DEFINITION

Artificial Intelligence is an umbrella term for various computational strategies able to display human-like capabilities such as reasoning and learning

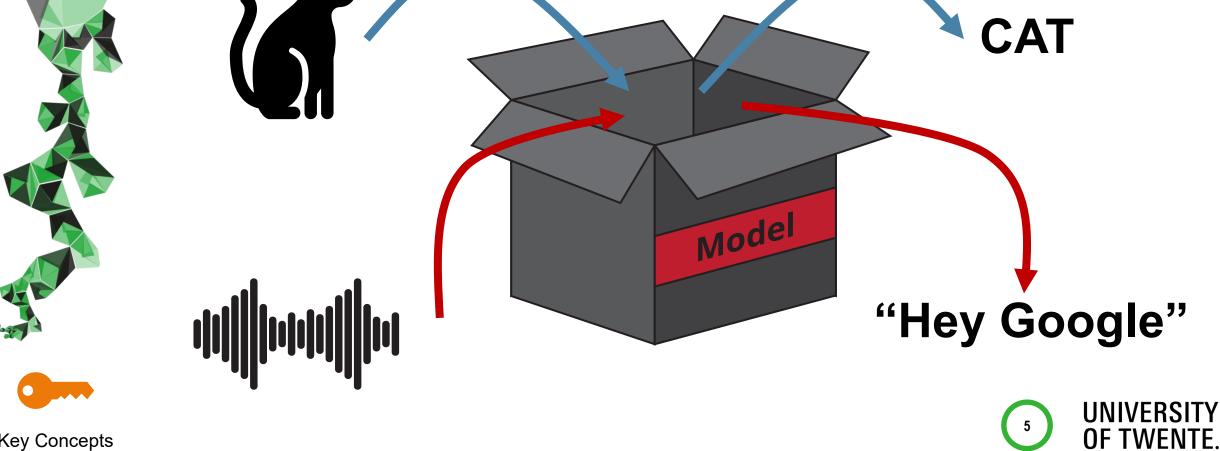
Examples: machine learning, robotics and natural language processing





Key Concepts

MACHINE LEARNING DEEP LEARNING MODEL



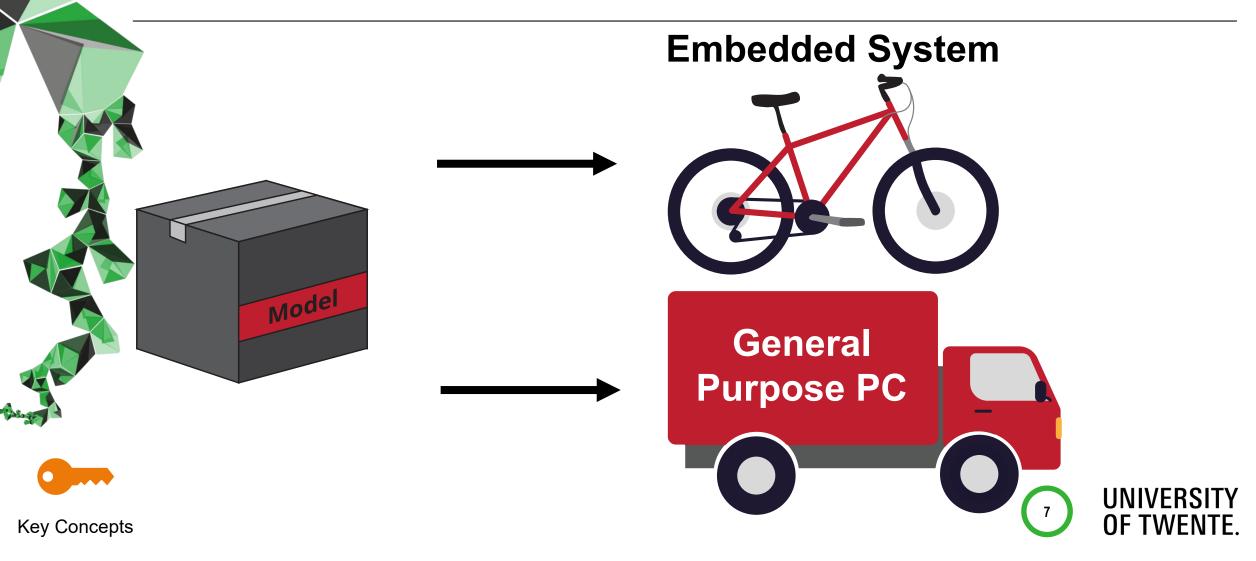
Key Concepts

TYPES DEEP NEURAL NETWORKS Transformer e.g. Natural Language Processing **Convolutional Neural Network** e.g. Image Classification Model Generative **Adversarial Networks Recurrent** e.g. Synthetic Face Generation **Neural Network** e.g. Speech Processing UNIVERSITY OF TWENTE.

Key Concepts

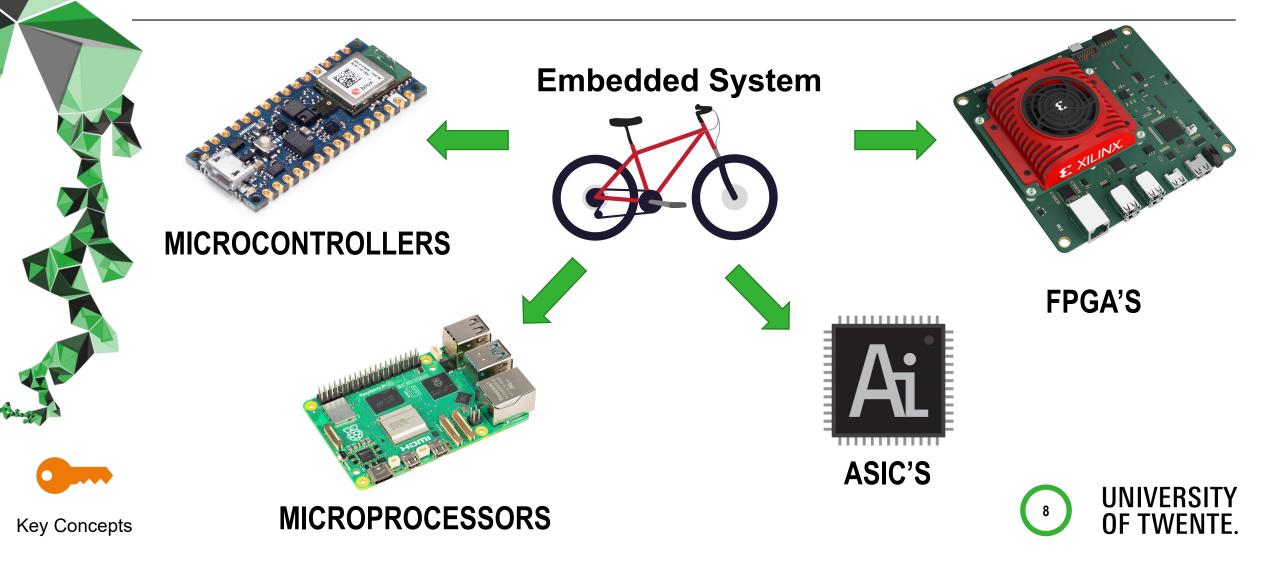
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IMPLEMENTING AI ON EMBEDDED SYSTEMS CHALLENGES



EMBEDDED SYSTEMS

A DEVICE DESIGNED FOR ONE SPECIFIC TASK WITHIN A SYSTEM



IMPLEMENTING AI ON EMBEDDED SYSTEMS COMPARE EMBEDDED SYSTEM WITH PC



Embedded System

- Special Purpose Hardware
- Pre-programmed firmware
- Real-time response and efficiency is key
- Little power consumption
- Cheap
- Local data processing



- Generic Hardware
- Programmable by user
- Performance and capacity is key
- High power consumption
- Expensive
- Cloud-based data processing





Key Concepts

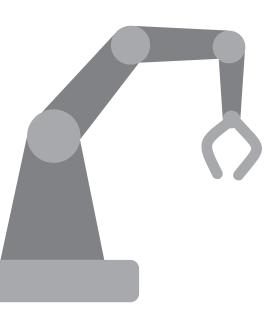


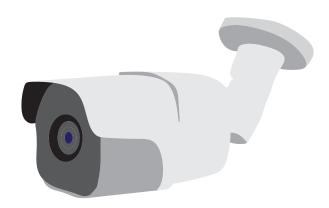
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APPLICATIONS

Personal Healthcare

Predictive Maintenance

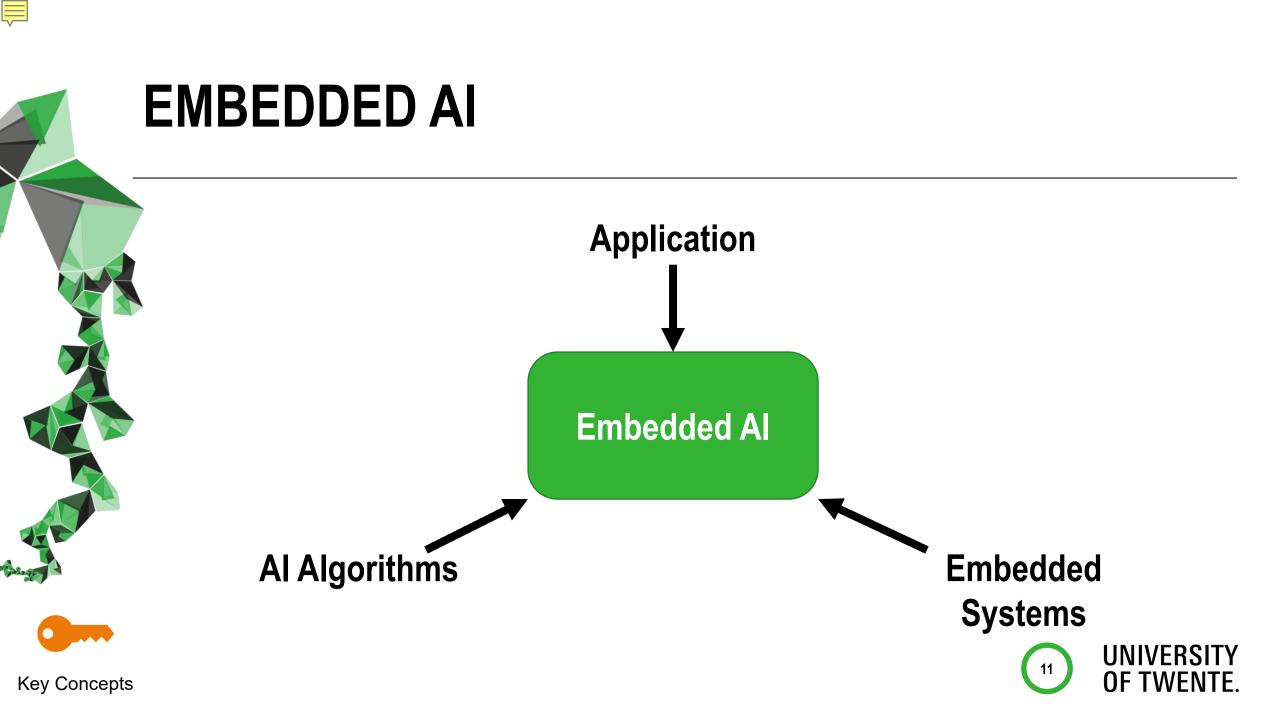




Smart Camera's











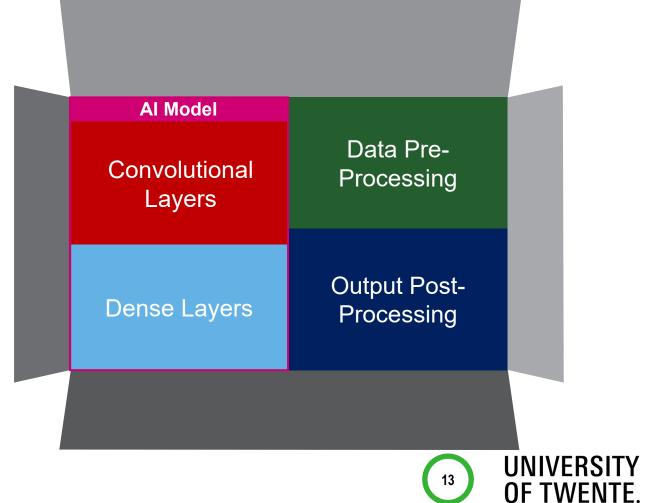
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Data Management & Biometrics Computer Architectures & Embedded Systems

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MODEL DESIGN & PERFORMANCE

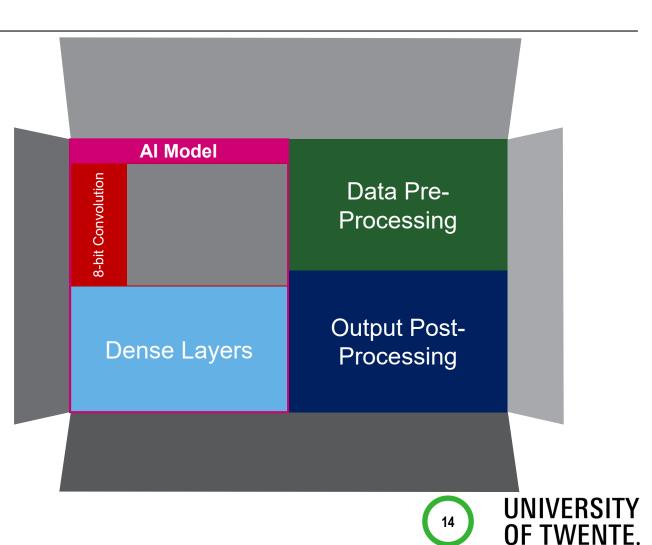
 AI model contains several different layers



Techniques Efficient Computation

MODEL DESIGN & PERFORMANCE

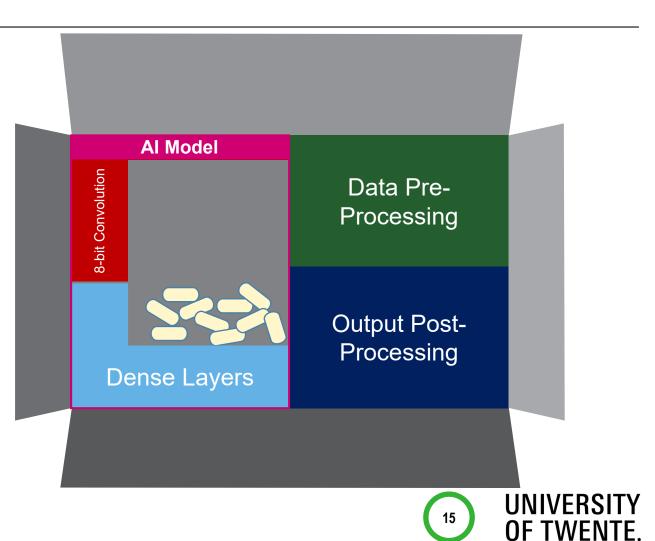
- Quantization
 - Approximating a 32-bit number with an 8-bit number
 - Can reduce footprint by a factor 4



Techniques Efficient Computation

MODEL DESIGN & PERFORMANCE

- Pruning
 - Removing connections in neural network that do not contribute
 - Like packaging peanuts that fill up space



Techniques Efficient Computation

EFFECT OF QUANTIZATION ON FACE RECOGNITION

Face Recognition Dataset	32-bit	8-bit	4-bit	2-bit
Post Training Quantization	98.85%	94.65%	63.15%	51.55%
Quantize Aware Training	98.85%	98.68 ± 1.15%	98.63 ± 0.18%	93.45 ± 0.66%



Techniques

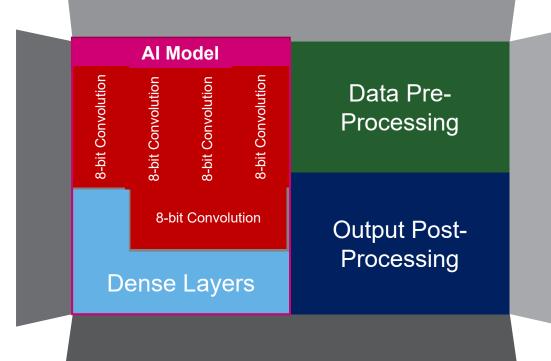
Efficient Computation

S. Bunda, L. Spreeuwers and C. Zeinstra, "Sub-byte quantization of Mobile Face Recognition Convolutional Neural Networks," 2022 International Conference of the Biometrics Special Interest Group (BIOSIG), Darmstadt, Germany, 2022, pp. 1-5, doi: 10.1109/BIOSIG55365.2022.9897025.





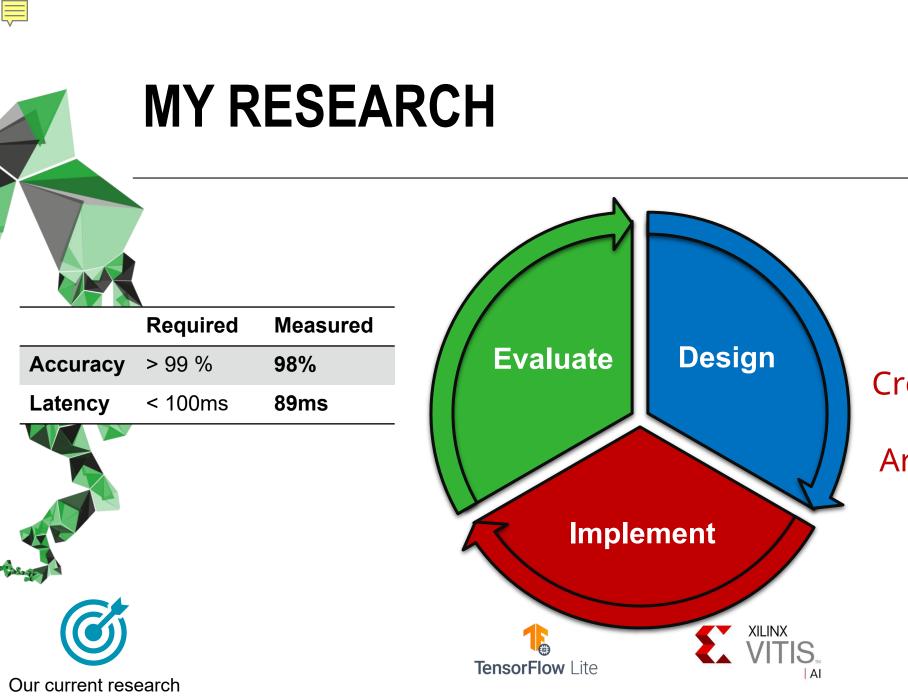
OPTIMIZING HARDWARE USAGE



Beware power consumption and latency!



Efficient Computation



and goals

Create/Adapt

Model Architecture



HARDWARE CO-DESIGN NEURAL ARCHITECTURE SEARCH

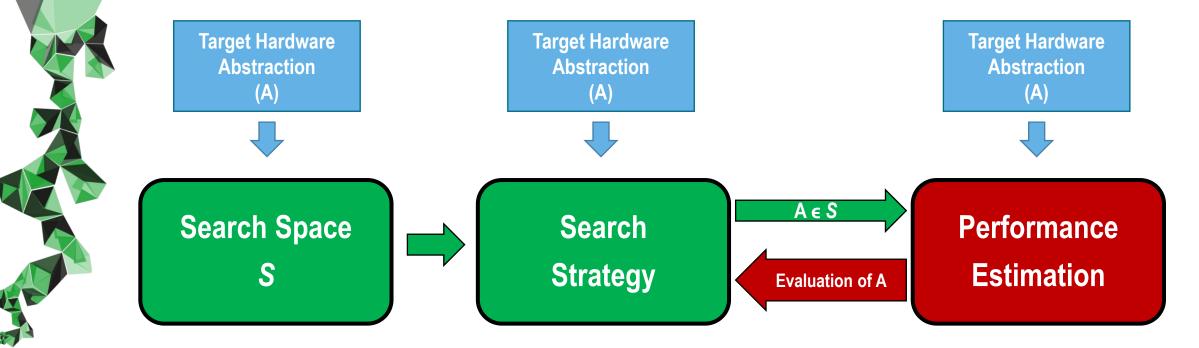
Optimize design by:

- Proposing models that fit within the target hardware memory
- Search for optimal implementation
- Evaluate based on e.g. accuracy, latency and energy





HARDWARE CO-DESIGN NEURAL ARCHITECTURE SEARCH





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Our current research and goals



ACTIVE PROJECTS



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SOME ACTIVE PROJECTS

- Image classification optimization using FINN-aware neural architecture search
- Vision-based object distribution detection using YOLO and Raspberry Pi
- Efficient Transformer Networks by researching 8-bit Arithmetics for Transformers
- Hardware acceleration for Genetic Selective Sweep Detection using modern technologies



Our current research and goals



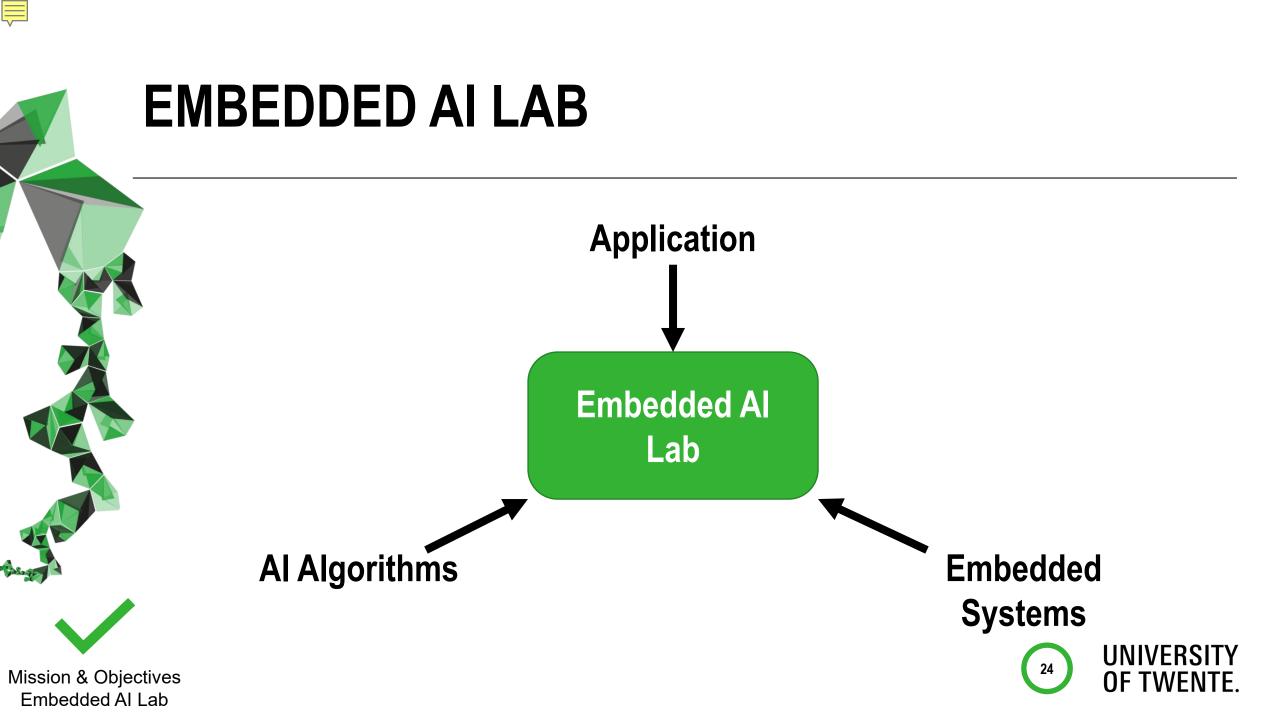
OUTLOOK OF EMBEDDED AI

- Expect new efficient AI-acceleration hardware
 - AMD Ryzen AI NPU
 - Neuromorphic Processors
- Leverage the knowledge of the domain of AI and Embedded Systems for Specific Applications



Our current research and goals





OBJECTIVES EMBEDDED AI LAB

- **Objective 1**: Create an Embedded AI Community at UT
 - Creating new contacts and collaborations within the University of Twente
- Objective 2: Inspire using DSI Embedded AI Seminars
 - Everyone is welcome to present their work and start discussions
- Objective 3: Stimulate student projects through collaboration and teaching
 - Co-supervise students on Embedded AI topics
 - Teaching the basics through Master course



Mission & Objectives Embedded Al Lab





Embedded Al Lab

PLANNING



Mario Ruiz from AMD on their new Neural Processing Units (NPUs)

December

Talk on Neuromorphic Computing

Qing Wang on his research on Embedded AI in TU Delft





EMBEDDED AI LAB

- Simulate collaboration on the topic of Embedded AI within University of Twente and with industry
- Develop AI methods to achieve State-of-the-Art performance with limited resources
- Build **Demonstrators** to show applications



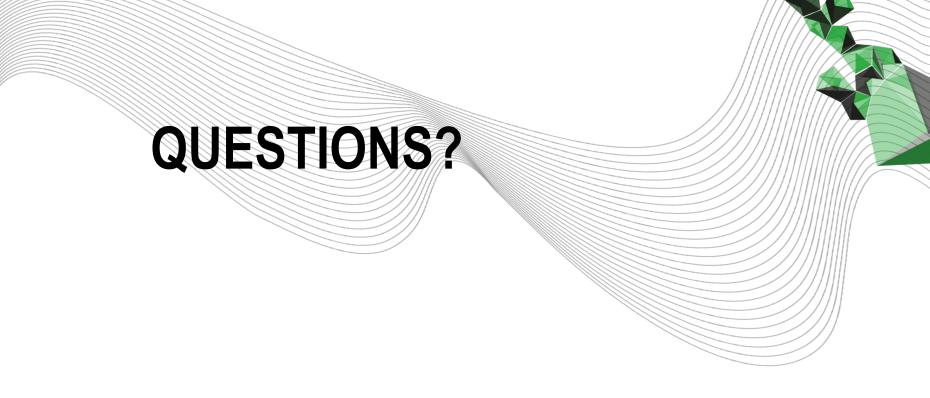


Mission & Objectives Embedded AI Lab

SIGN UP FOR NEXT SEMINAR 9TH OF OCTOBER







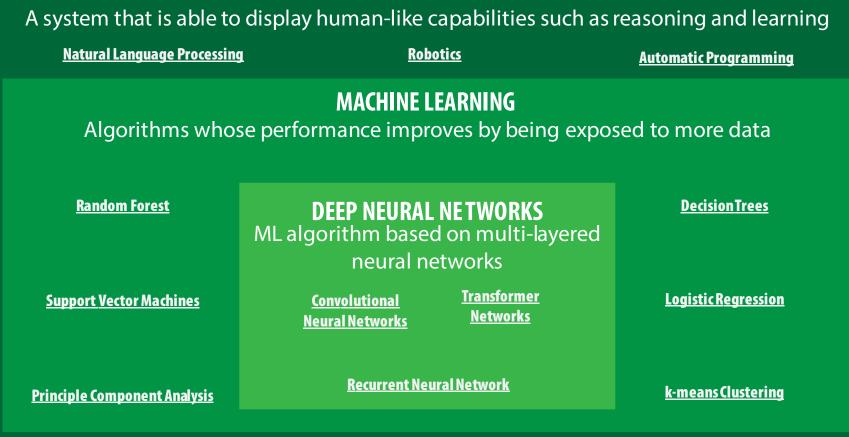






ARTIFICIAL INTELLIGENCE

ARTIFICIAL INTELLIGENCE





TENSORFLOW LITE & TINYML COMPILATION TOOLING

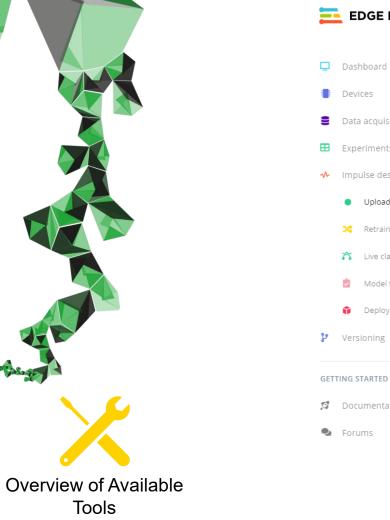
- tensorflow.org/lite/microcontrollers
- tinyurl.com/tinyml-book





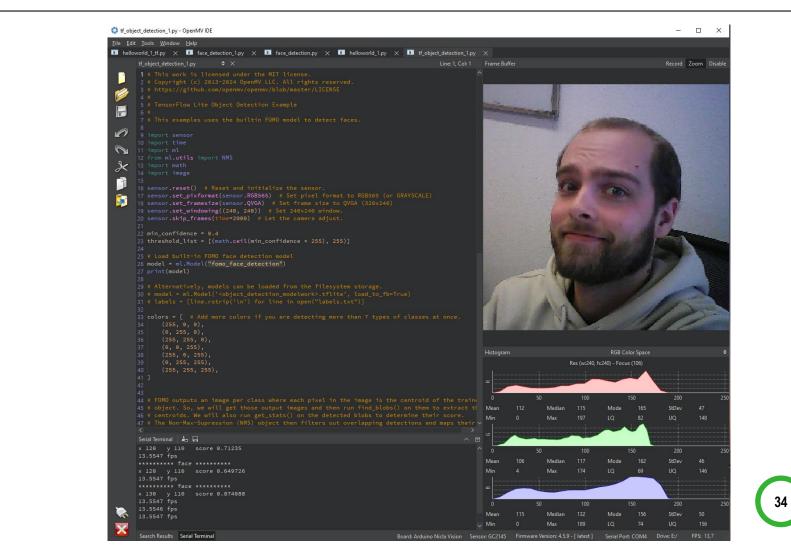
Overview of Available Tools

EDGE IMPULSE DEPLOYMENT TOOLING



EDGE IMPULSE	Sebastian / m7715157-project-1 PERSONAL		Target: Arduino Nicla Vis S	50 CO
	Impulse #1			
Dashboard				
Devices	 Step 2: Process "mobilenet-v1-tflite-0-25-128-qu 		On-device performance Arduino Nicla Vision (
Data acquisition				
Experiments	Configure model settings for optimal processing.		PROCESSING TIME FLASH USAGE 35 ms. 485,6K	
Impulse design 🔹	Model input Input shape: (128, 128, 3)	Image (RGB)	This model won't run on MCUs. Calculated arena size is >6MB	
 Upload model 	How is your input	Pixels ranging 0255 (not no 💙		
🗙 Retrain model	scaled?	Input should be in RGB format (one value per pixel). If your model uses a	Check model behavior	
Tive classification		different channel order, or is scaled differently, then select "Other".		
Model testing	Model output	Classification 🗸	Upload test data to ensure correct model settings and proper model processing. (Optional)	
🗘 Deployment	Output shape: (1001)		Upload an image	
Versioning	Output labels (1001) Enter labels for your model separated by ','.	class 1, class 2, class 3, class 4, c	Upload an image to try out your model. The image will be automatically resized to 128x128 (RGB).	
NG STARTED			Bestand kiezen Geen bestand gekozen Test sample	
Documentation	Save model			
Forums				
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				OF TWENTE.

MICROPYTHON IN OPENMV DEPLOYMENT TOOLING





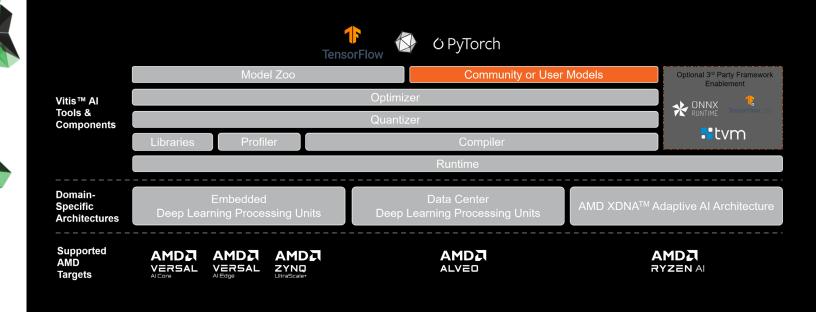
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Overview of Available Tools

VITIS AI & FINN DEPLOYMENT TOOLING

AMD Vitis[™] AI Integrated Development Environment





https://xilinx.github.io/Vitis-AI/3.5/html/index.html https://xilinx.github.io/finn/

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Overview of Available Tools



STM32CUBE.AI

- Optimize and deploy Deep Neural Networks on STM32
 microcontrollers:
 - <u>https://stm32ai.st.com/stm32-cube-ai/</u>

