

What is tinyML and what is it used for?

On-device machine learning applications in the single mW and below



Vibration and motion Any 'signal'

Predictive maintenance, sensor fusion, accelerometer, pressure, lidar/radar, speed, shock, vibration, pollution, density, viscosity, etc.



Voice and sound

Recognition and creation

Keyword spotting, speech recognition, natural language processing, speech synthesis, sound recognition, etc.



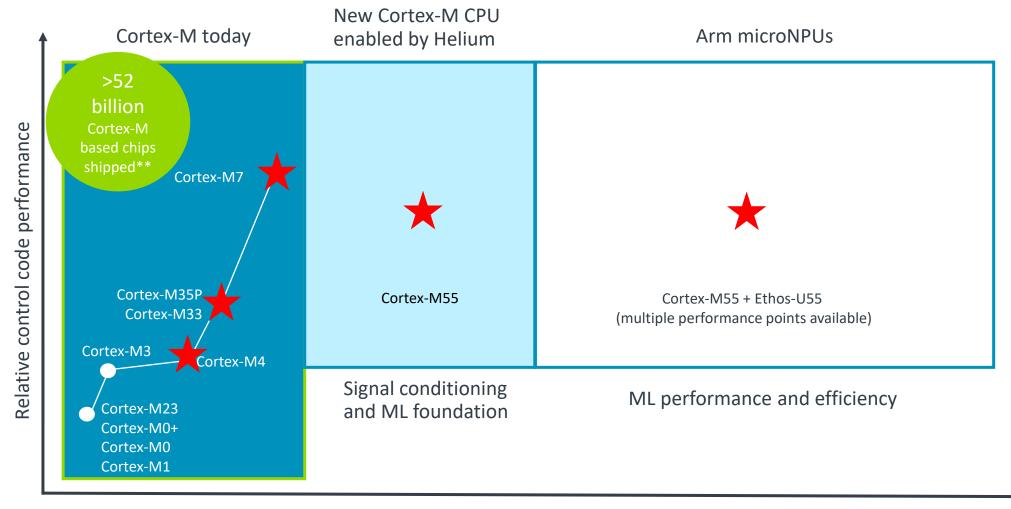
Vision

Images and video

Object detection, face unlock, object classification etc.



Pushing the Boundaries for Real-time On-device Processing



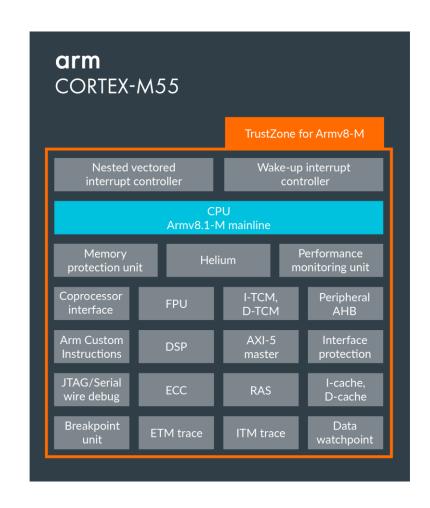
Relative ML and signal processing performance

**Based on Arm data



Cortex-M55: The Most Al-capable Cortex-M Processor

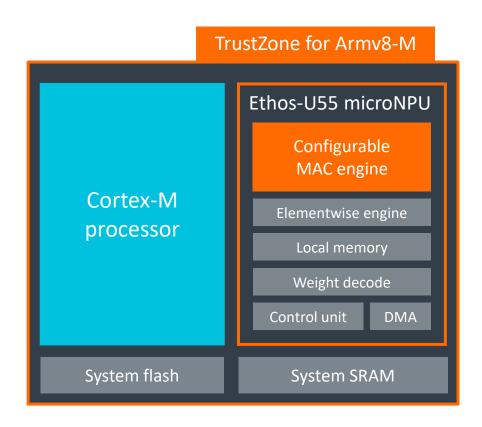
- ✓ First CPU based on Arm Helium technology
 - Energy-efficient and configurable with vector processing capabilities
 - Delivers up to 5x DSP performance and up to 15x ML performance*
 - Versatile capability for both classical ML and NN inference
- ✓ Advanced memory interfaces for fast access to ML data and weights
- ✓ TrustZone support
- Extensive configurability





Ethos-U55: The first microNPU for Cortex-M

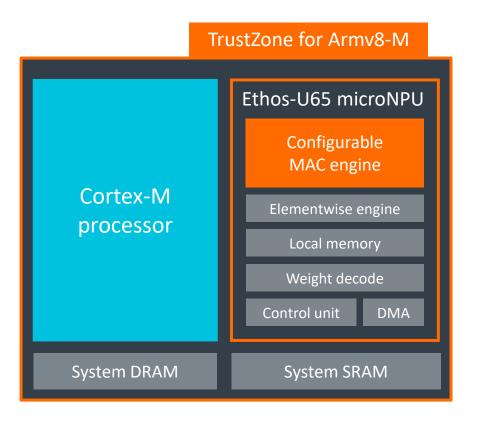
- ✓ Highest efficiency and small memory footprint
- ✓ 32, 64, 128, or 256 unit multiply-accumulate (MAC) engine
- ✓ Weight decoder and DMA for on-the-fly weight decompression
- ✓ Tooling available for offline optimization.
- ✓ Works with a range of Cortex-M processors:
 - Cortex-M55
- Cortex-M7
- Cortex-M33
- Cortex-M4





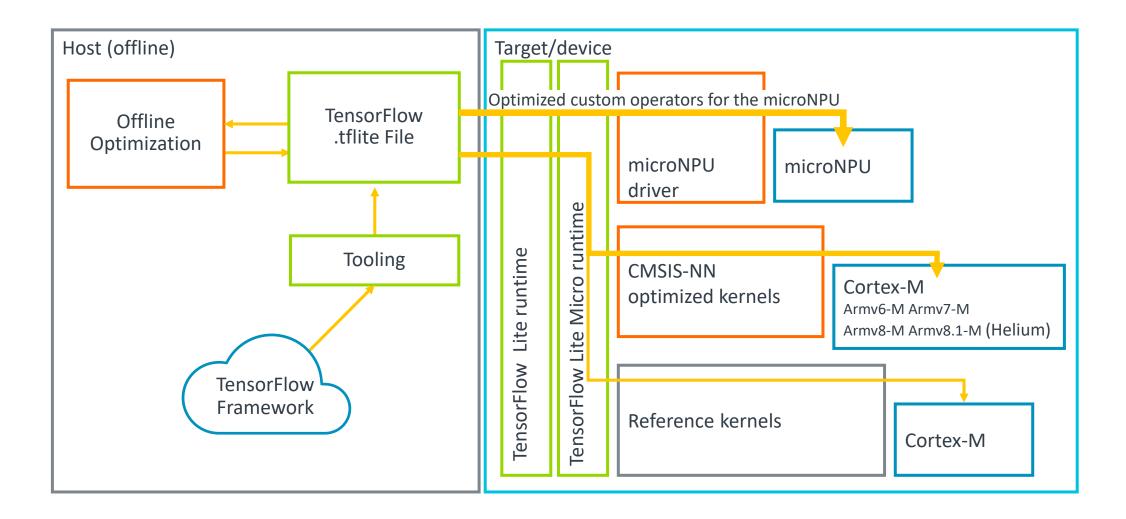
Ethos-U65: The second generation microNPU

- ✓ 256 or 512 unit multiply-accumulate (MAC) engine
- ✓ DMA update for DRAM as well as flash support
- ✓ Can be an M-class subsystem inside an A-class system





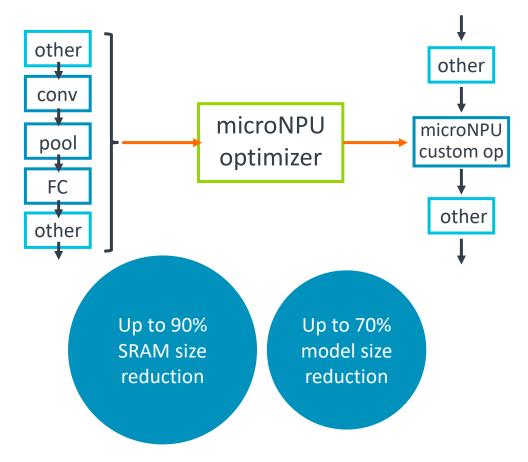
Mapping of NNs to Ethos-U using TensorFlow Lite





The Vela Optimizer

- Open source compiler
- Reads a tflite file and identifies subgraphs
- Optimizes scheduling of subgraphs
- Loss-less compression of weights
- Generates commands for microNPU
- Writes out a modified tflite file

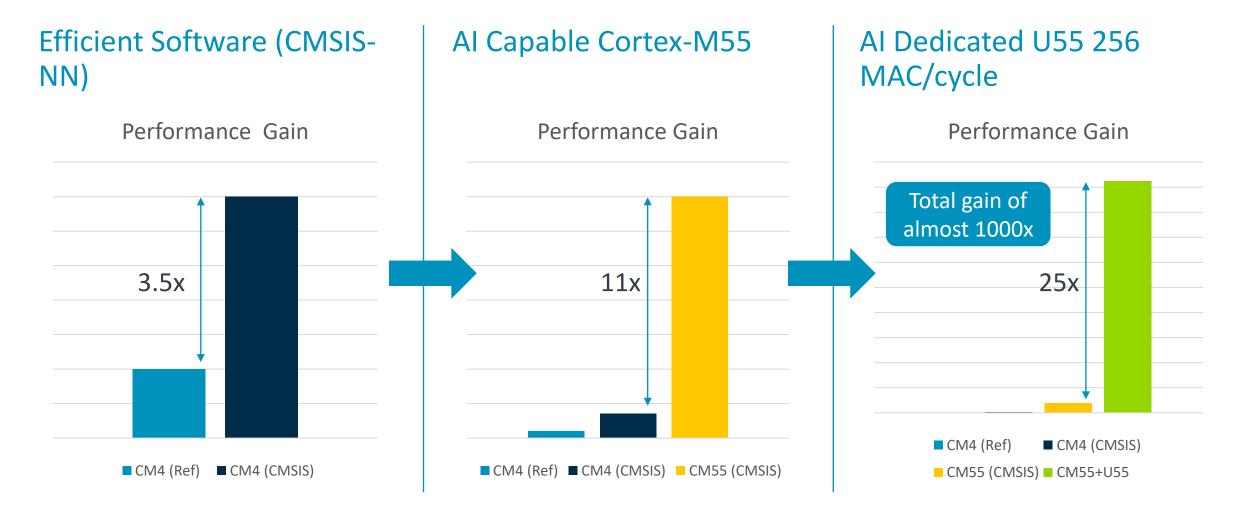


Enabling networks not before feasible in embedded systems



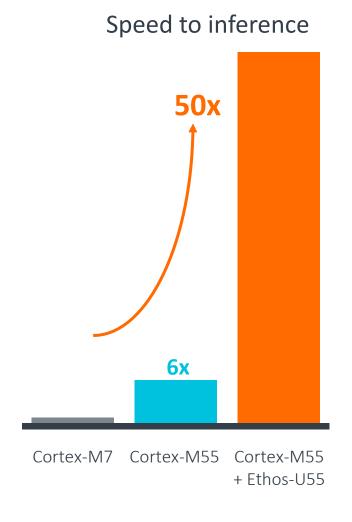
Neural Network performance Across ARM IPs

Wav2Letter

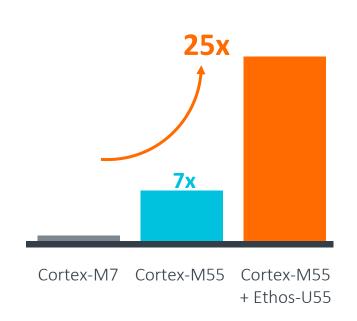




Full example: Typical ML Workload for a Voice Assistant





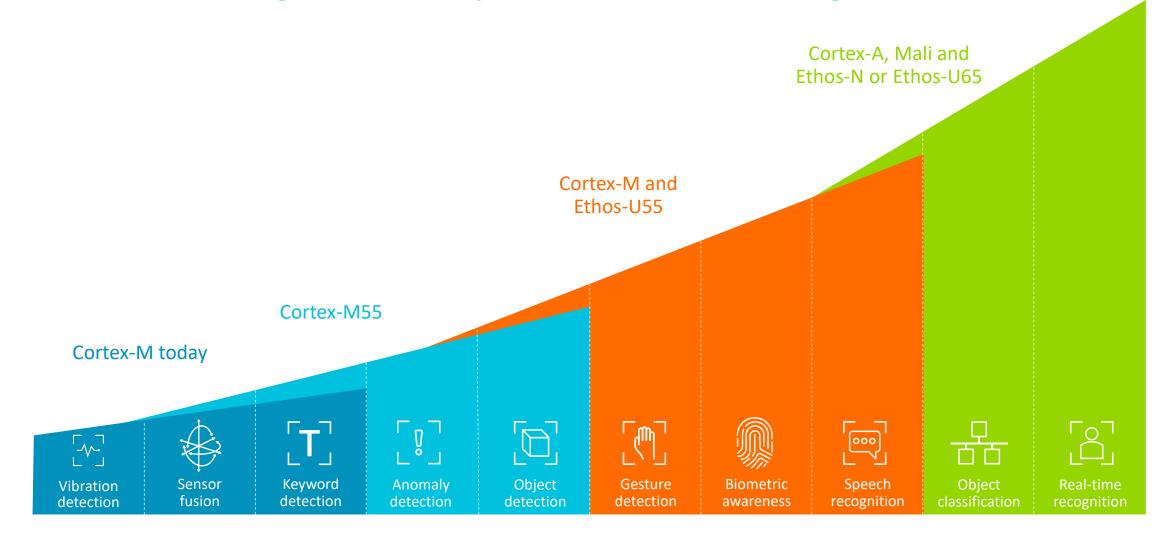


- ✓ Faster responses
- ✓ Smaller form-factors
- ✓ Improved accuracy

Latency and energy spent for all tasks listed combined: voice activity detection, noise cancellation, two-mic beamforming, echo cancellation, equalizing, mixing, keyword spotting, OPUS decode, and automatic speech recognition.



Broadest Range of ML-optimized Processing Solutions







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