

Chapter 2.1

EDGE COMPUTING AND EMBEDDED ARTIFICIAL INTELLIGENCE

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Scope

This chapter focuses on **computing components**, and more specifically on **embedded architectures, edge computing devices and systems using artificial intelligence at the edge**, including, but not exclusively:

- Processors (CPU, MPU) with high energy efficiency,
- Accelerators (for AI and for other tasks, such as security):
- GPU (and their generic usage),
- NPU (Neural processing unit)
- DPU (Data processing Unit, e.g. logging and collecting information for automotive and other systems) and processing data early (decreasing the load on processors/accelerators),
- Other accelerators xPU (FPU, IPU, TPU, XPU, ...)
- **Memories** and associated controllers, specialized for low power and/or for processing data locally (e.g. using non-volatile memories such as PCRAM, CBRAM, MRAM, and ***In/Near Memory Computing***), etc.
- **Power management systems** and techniques.

Key trends

More and more convergence between edge computing and embedded (generative) AI, but ***still a lot of edge will be without AI.***

Emergence of ***Gen-AI at the edge*** (Copilot+PC, Apple Intelligence), for automatizing tasks locally, processing local data, for natural user interface, but also for ***image interpretation and robotics.***

New Recommendations:

- Managing diversity and the ***dynamic range of computing*** -> A system becomes an ***orchestration of federated services, distributed or centralized*** (Software Defined X).

Impact on architecture and of the programmability and interoperability (cloud techniques used: containers (silo), orchestration, WASM, interoperability, “Web standards”) therefore their hardware support, including for networking.



- Disaggregation of complex SoC into chiplet + interposers, for example potentially for the automotive market, but still no ecosystem of interoperable chiplets and overall architecture.

Key trends

- Generative AI is not only for the cloud, Intelligent Agents (federations of Small Action Models) or **Agentic AI** will emerge at the edge, and not only in smartphones.
- **Fine tuning** of models should be possible at the edge only with local data
- **Memory cost is crucial for generative AI at the edge.** New innovations required to avoid to waste RAM
- Emergence of (very) cheap Chinese Risc-V microcontrollers
- Further **reducing standby power** and fast on operation (stop and go for chips?) and **proportionality of power/load** for reducing overall power consumption
- Still research required for **new computing paradigms** (neuromorphic, **using physics to make computation** – analog computing -, etc) and their **validation** in product ready solutions.

Major challenges

The Major Challenges (still) are :

- 1. **Energy Efficiency**: Developing innovative hardware architectures and minimizing data movement are critical for energy-efficient computing systems. **Memory is becoming an important challenge** as we are moving from a computing centric paradigm to a data centric (driven by AI). **Zero standby energy and energy proportionality to load** is essential for edge devices. 
- 2. **System Complexity Management**: Addressing the complexity of embedded systems through interoperability, modularity, and **dynamic resource allocation** in a (**distributed,**) safe and secure way. **Web technologies cascade to edge** (containerization, WASM, protocols, ...) forming a **continuum of computing resources** 
- 3. **Lifespan of Devices**: Enhancing hardware support for software upgradability, interoperability, and second-life applications.
- 4. **Sustainability**: Ensuring European sustainability by developing solutions aligned with ethical principles (for embedded AI) and transforming innovations into commercial successes (for example, based on open standards, such as Risc-V, and for innovative solutions such as neuromorphic computing)

R&I focus areas

- Processing data locally and reducing data movements (towards the computing continuum)
- Co-design of algorithms, hardware and software
- Efficient management of storage resources
 - Unified memory
 - Innovations in memory technology
- Energy proportionality
- Ultra-low standby current
- Leveraging physical phenomena for computation
- Complexity management utilizing AI
- Decomposition of complex SoCs into chiplets and interposers
- Ensuring programmability and interoperability
- Combining processing devices to work together
- Modeling interactions among system components
- Ensuring long-term functionality and up to date operation
- Designing with modularity and extensibility in mind
- Reuse of components or systems in a downgraded or requalified use case
- Leveraging open-source hardware and software for innovation and cost reduction
- Developing and federating smaller specialized AI models
- Training models with European data and ethical compliance
- Deploying efficient and sustainable embedded AI-oriented ECS
- Accelerating development of robust, cost-effective solutions

