



PULP PLATFORM

Open Source Hardware, the way it should be!

Working with RISC-V

Part 5 of 5 : PULP based chips

Davide Rossi

<davide.rossi@unibo.it>

Luca Benini

<luca.benini@unibo.it>

ETH zürich



<http://pulp-platform.org>



[@pulp_platform](https://twitter.com/pulp_platform)



https://www.youtube.com/pulp_platform



Summary

- Part 1 – Introduction to RISC-V ISA
- Part 2 – Advanced RISC-V Architectures
- Part 3 – PULP concepts
- Part 4 – PULP Extensions and Accelerators
- Part 5 – PULP based chips
 - From concept to reality
 - Single core microcontrollers PULPino to PULPissimo
 - Many core systems OpenPULP
 - Advanced systems with accelerators
 - Lessons learned, the good, the bad and the ugly.





We will discuss chips we have made with PULP

■ Why make chips at all?

- MPW: Only limited samples
- Use cases

■ Single core PULP chips

- PULPino (Imperio)
- PULPissimo (Arnold)

■ Many core PULP chips

- Cluster only (Honey Bunny, Dustin)
- PULPopen (Mr. Wolf, Vega)

■ Advanced PULP chips

- Kosmodrom: 2x 64b Ariane cores + ML accelerators
- Making use of technology: Body biasing

■ Lessons learned

- There are many pitfalls
- We had great success, but..
- Sometimes you have embarrassing failures. Part of the process



Multi Project Wafer, chips for prototyping

■ Cost sharing method for ICs

- Multiple ICs are manufactured together. They share the mask costs
 - 1.5M cost / 10 projects = 150k per project
 - But you only get 1 / 10 of the area
- Dedicated MPW services available
 - Europractice-IC for SMEs and academia

■ You only get few chips

- Usually 50 to 200
- Per chip costs very high (few kUSD)

■ All our chips through MPWs

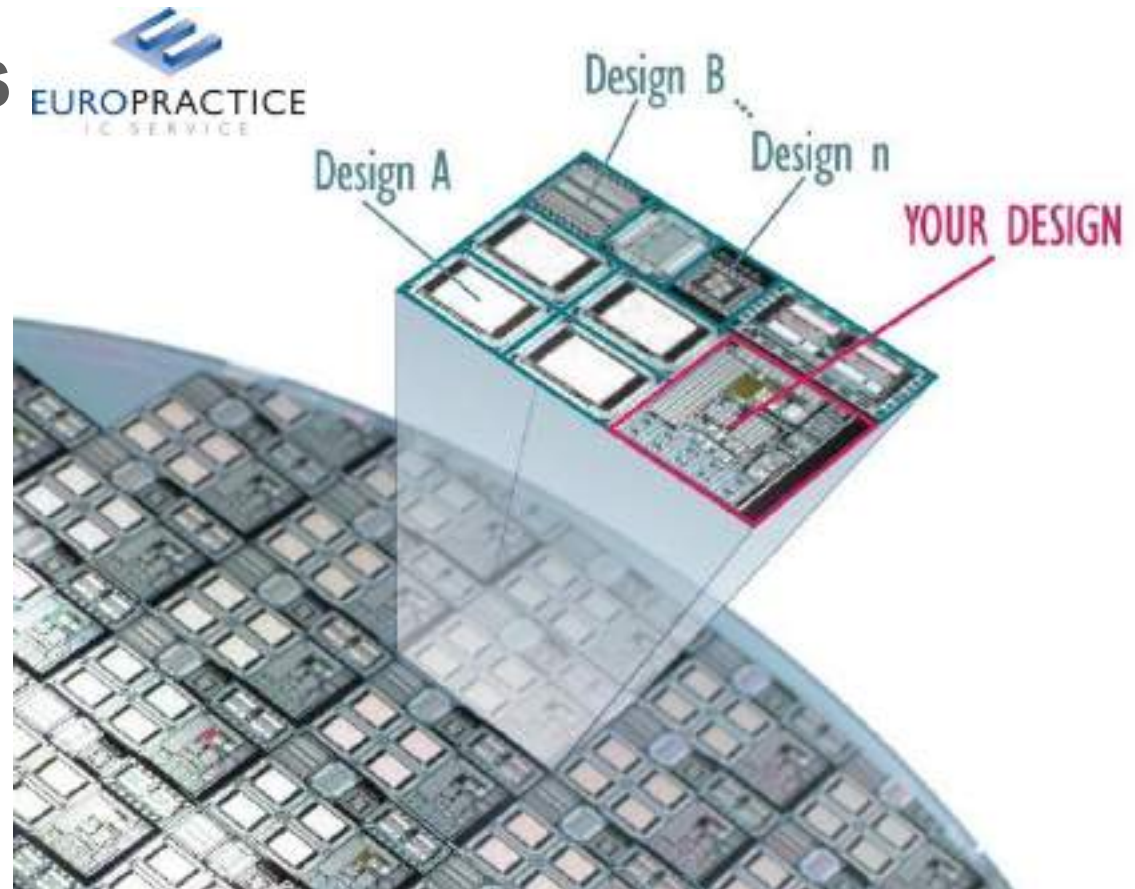
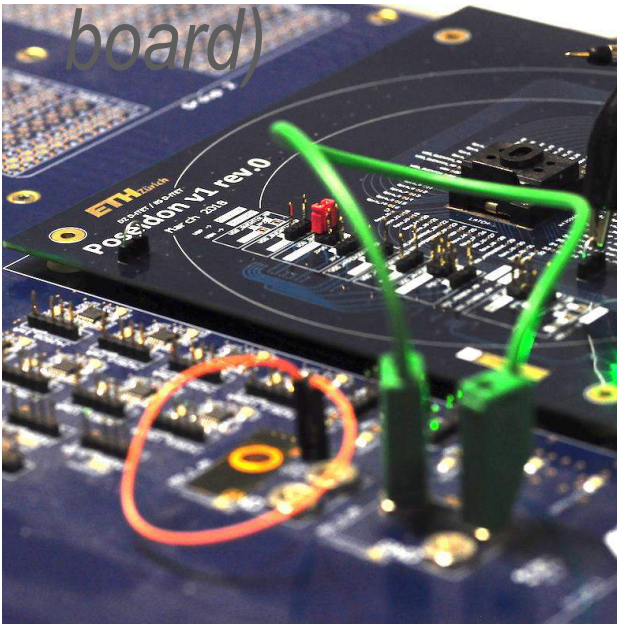


Image taken from <https://europractice-ic.com/mpw-prototyping/general/mpw-minisic/>




Our ASICs have different use cases

- Chips characterized on an IC tester (*Poseidon 22nm*)
- Research demonstrators (*Nano drone with Mr. Wolf/GAP8*)
- Industrial uses of our cores/peripherals (*open-isa.org Vega*)



ETH zürich


- 

- 

pulp-platform

[Repositories 100](#)
[Packages](#)
[People 30](#)
[Teams](#)
[Projects](#)
[Settings](#)


Pinned repositories



pulpissimo

This is the top-level project for the PULPissimo Platform. It instantiates a PULPissimo open-source system with a PULP SoC domain, but no cluster.


SystemVerilog ★ 85 ▼ 32



ariano

Ariano is a 6-stage RISC-V CPU capable of loading Linux

SystemVerilog ★ 308 ▼ 118




pulp

This is the top-level project for the PULP Platform. It instantiates a PULP open-source system with a PULP SoC (microcontroller) domain accelerated by a PULP cluster with 8 cores.


SystemVerilog ★ 75 ▼ 28

Customize pinned repositories




riscv

RISC-V is an open 6-stage RISC-V RV32M1Cpulp CPU



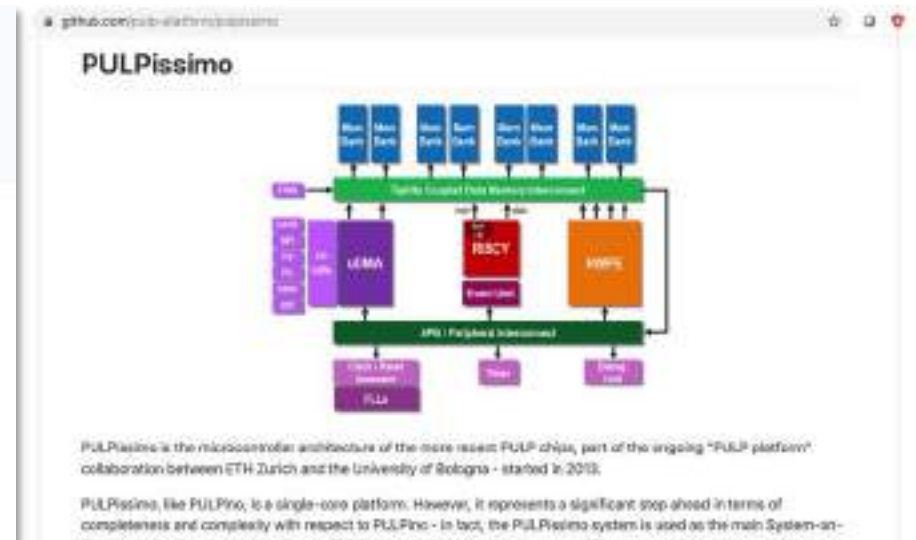
legosrv

RISC-V microprocessor evaluation for RISC-V legPULP hardware platform



pulp-droplet

A deep learning generalist neural navigation engine to enable autonomous navigation of pocket-size quadrotor - running on PULP



PULP has released a large number of IPs

RISC-V Cores

RI5CY	Ibex	Snitch	Ariane + Ara
32b	32b	32b	64b

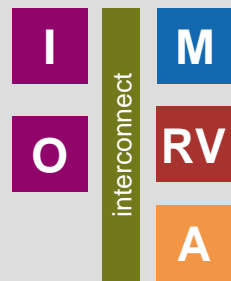
Peripherals

JTAG	SPI
UART	I2S
DMA	GPIO

Interconnect

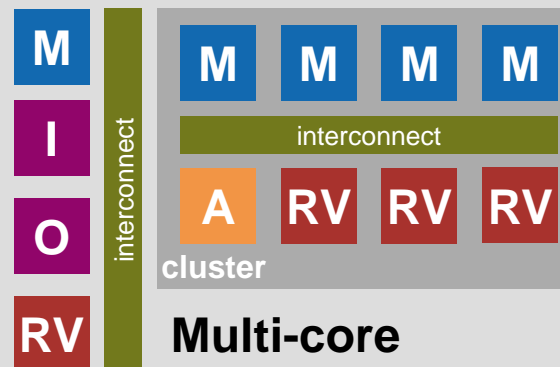
Logarithmic interconnect
APB – Peripheral Bus
AXI4 – Interconnect

Platforms



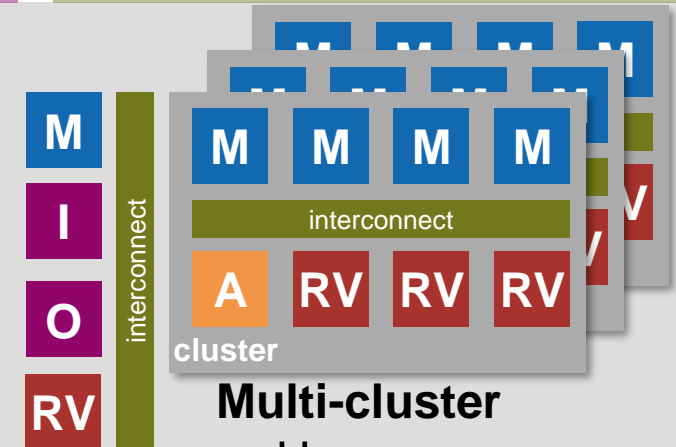
Single Core

- PULPino
- PULPissimo



Multi-core

- Fulmine
- Mr. Wolf



Multi-cluster

- Hero
- Open Piton

IOT

HPC

Accelerators

HWCE
(convolution)

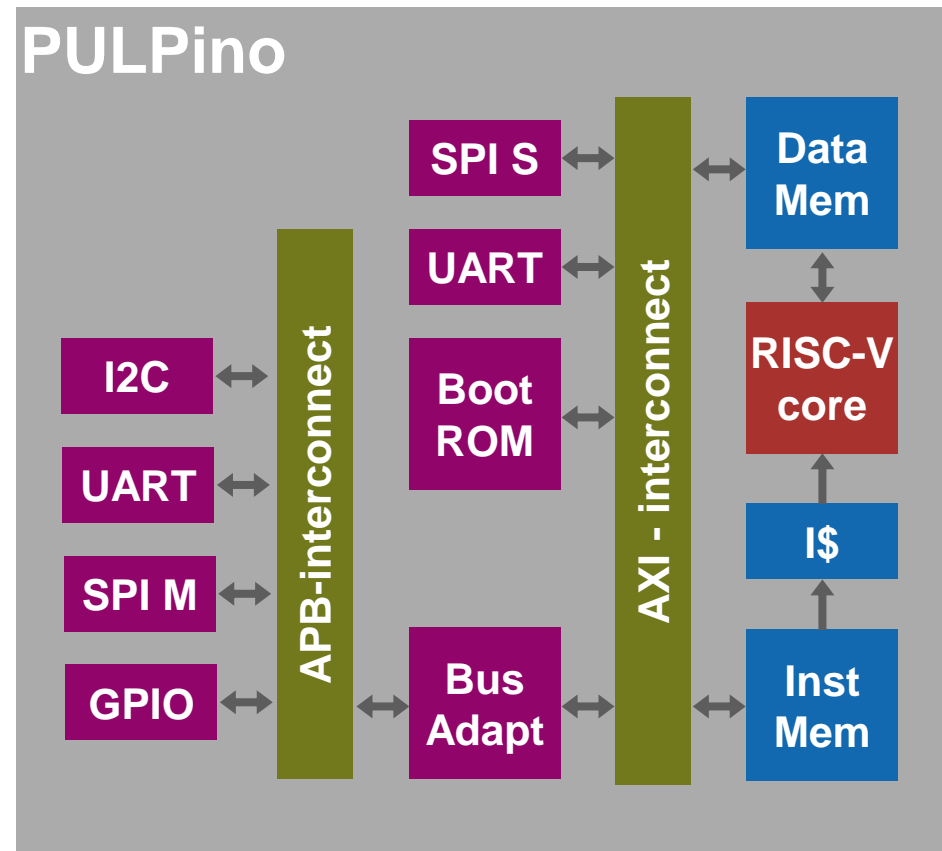
Neurostream
(ML)

HWCrypt
(crypto)

PULPO
(1st ord. opt)

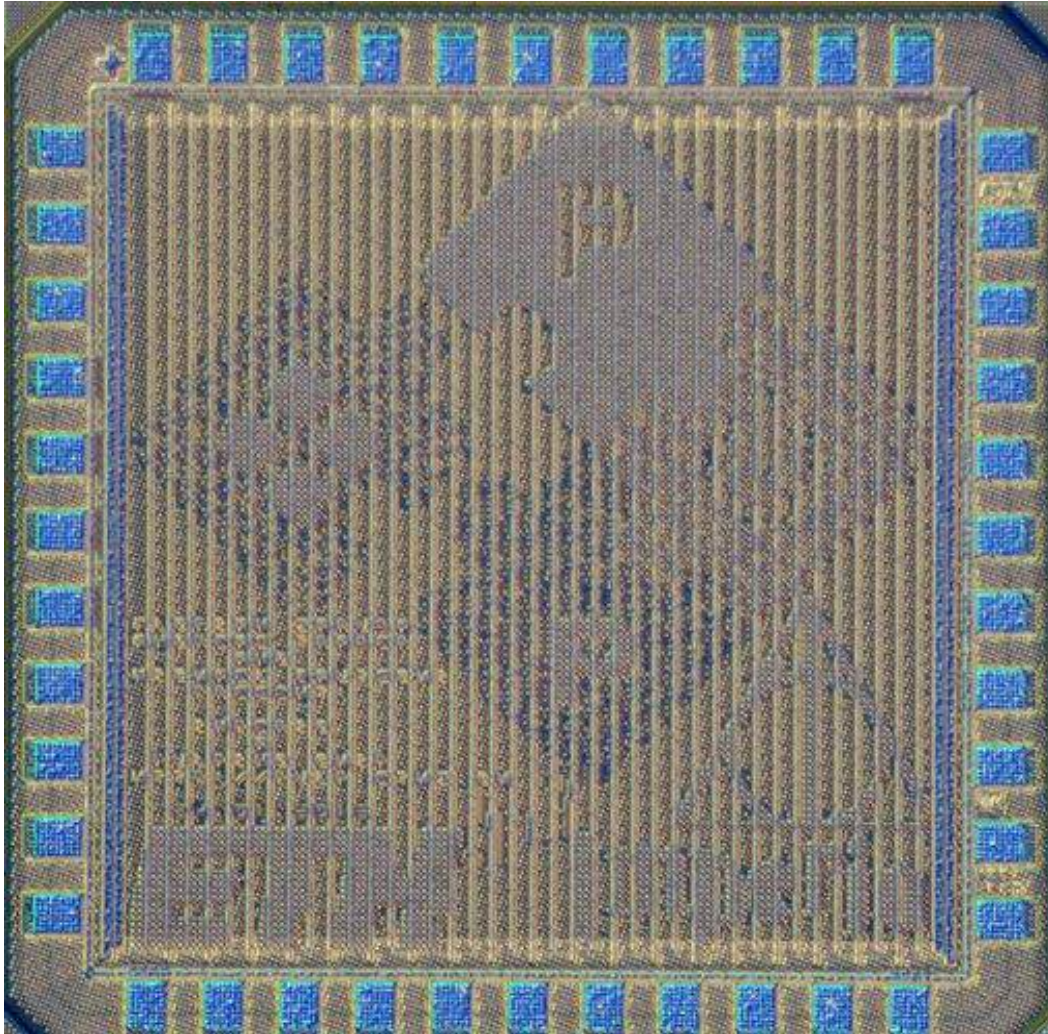
PULPino: Our first open source release

- **Simple design**
 - Meant as a quick release
- **Separate data and inst. mem**
 - Makes it easy in HW
 - Not meant as a Harvard arch.
- **Can use all our 32bit cores**
 - RI5CY, Zero/Micro-Riscy (Ibex)
- **Peripherals from other projects**
 - Any AXI and APB peripherals could be used





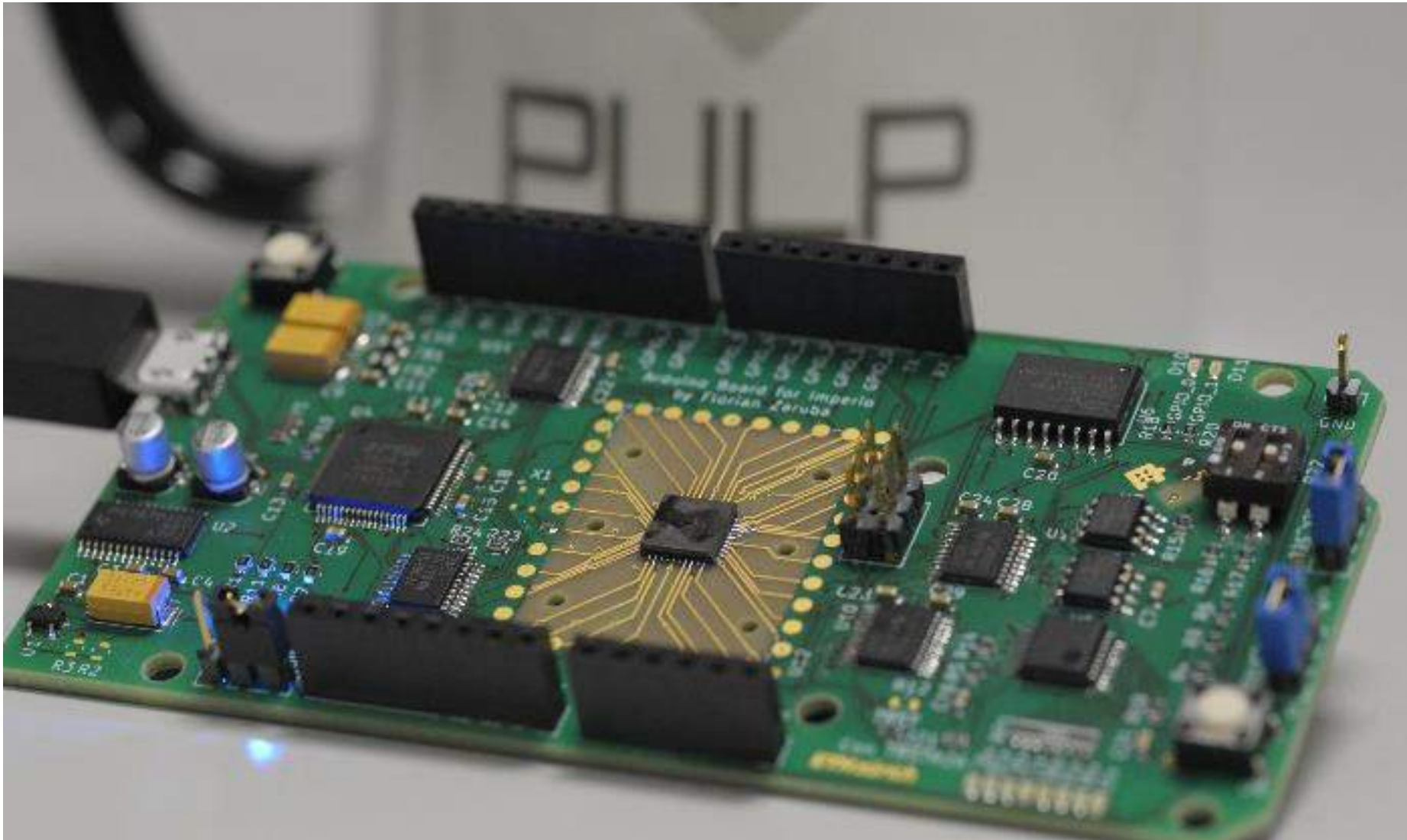
Imperio – 65nm RISC-V core



- **Chip implemented in 65nm**
 - Using RI5CY (RV32IMC) core
 - 64 kBytes of memory
 - Basic peripherals (GPIO, SPI, I2C)
 - Working debug interface
- **Functional up to 500 MHz**
 - Main challenge was to find fast memory cuts to work at that speed.
 - Memory made of multiple smaller cuts to maximize the operating speed.



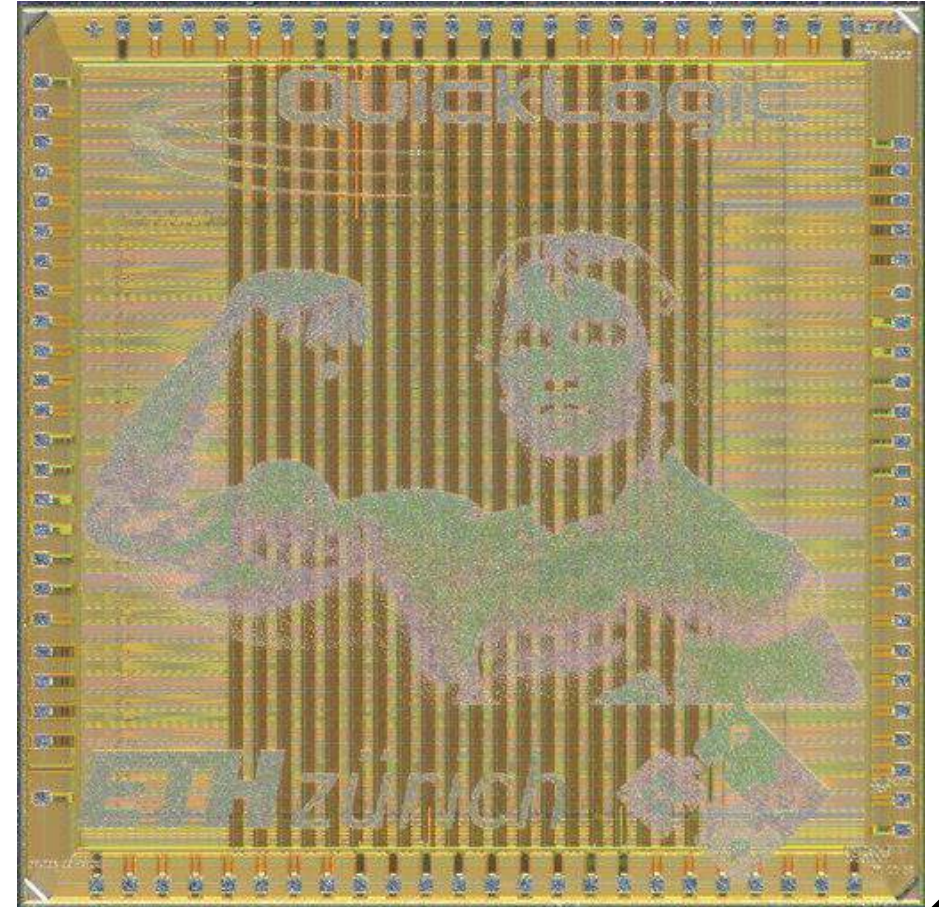
Working chip on an Arduino compatible board





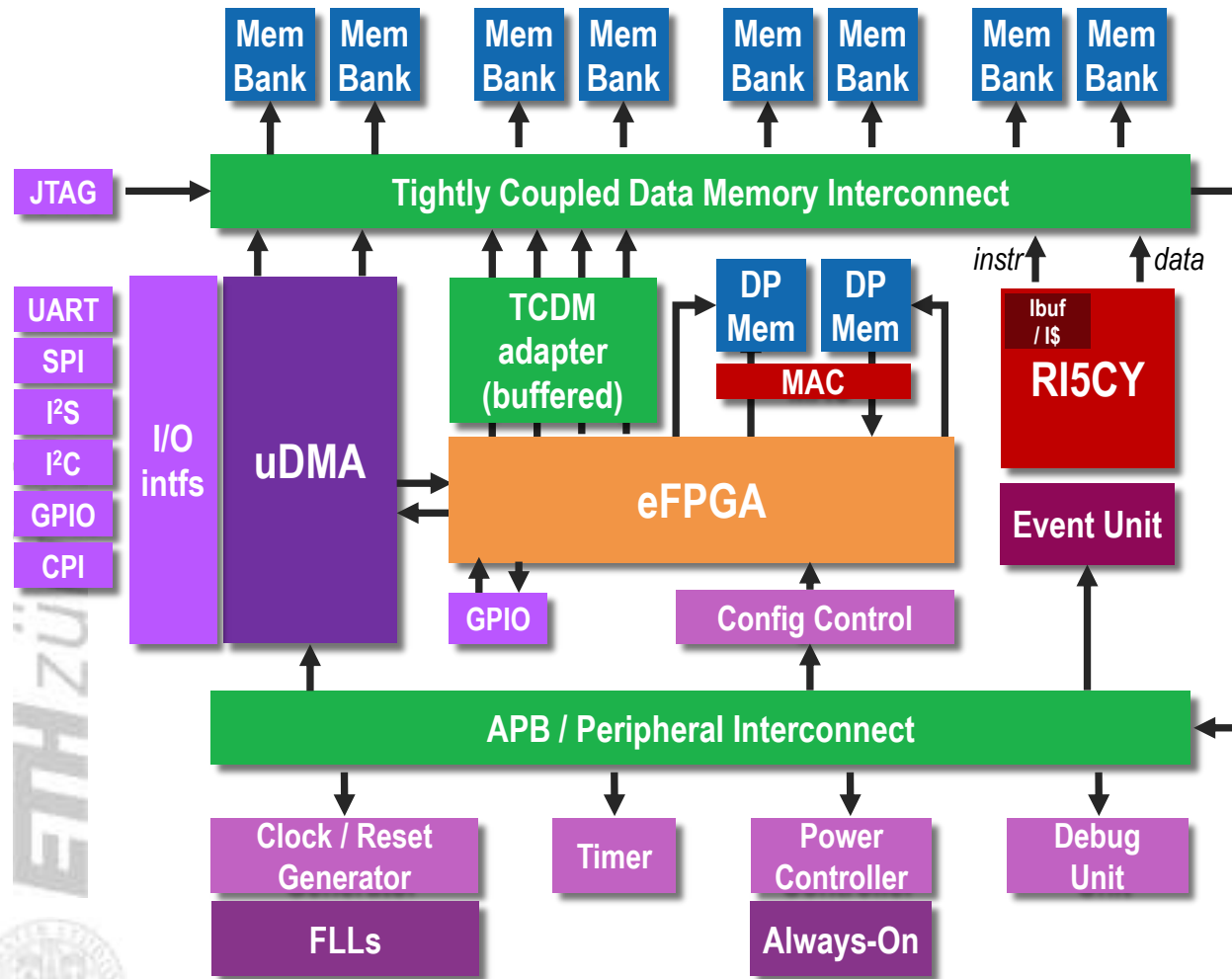
Arnold (2018) – Fastest collaboration

- **GF22nm**
 - RISC-V microcontroller with eFPGA
 - Based around PULPissimo
- **Collaboration with Quicklogic**
 - Met at GTC 2017 by coincidence
 - In one year chip was taped out
 - Only possible because of open source nature
- **Quicklogic is going open source**
 - They announced June 2020 the Quicklogic Open Reconfigurable Computing
<https://www.quicklogic.com/QORC/>



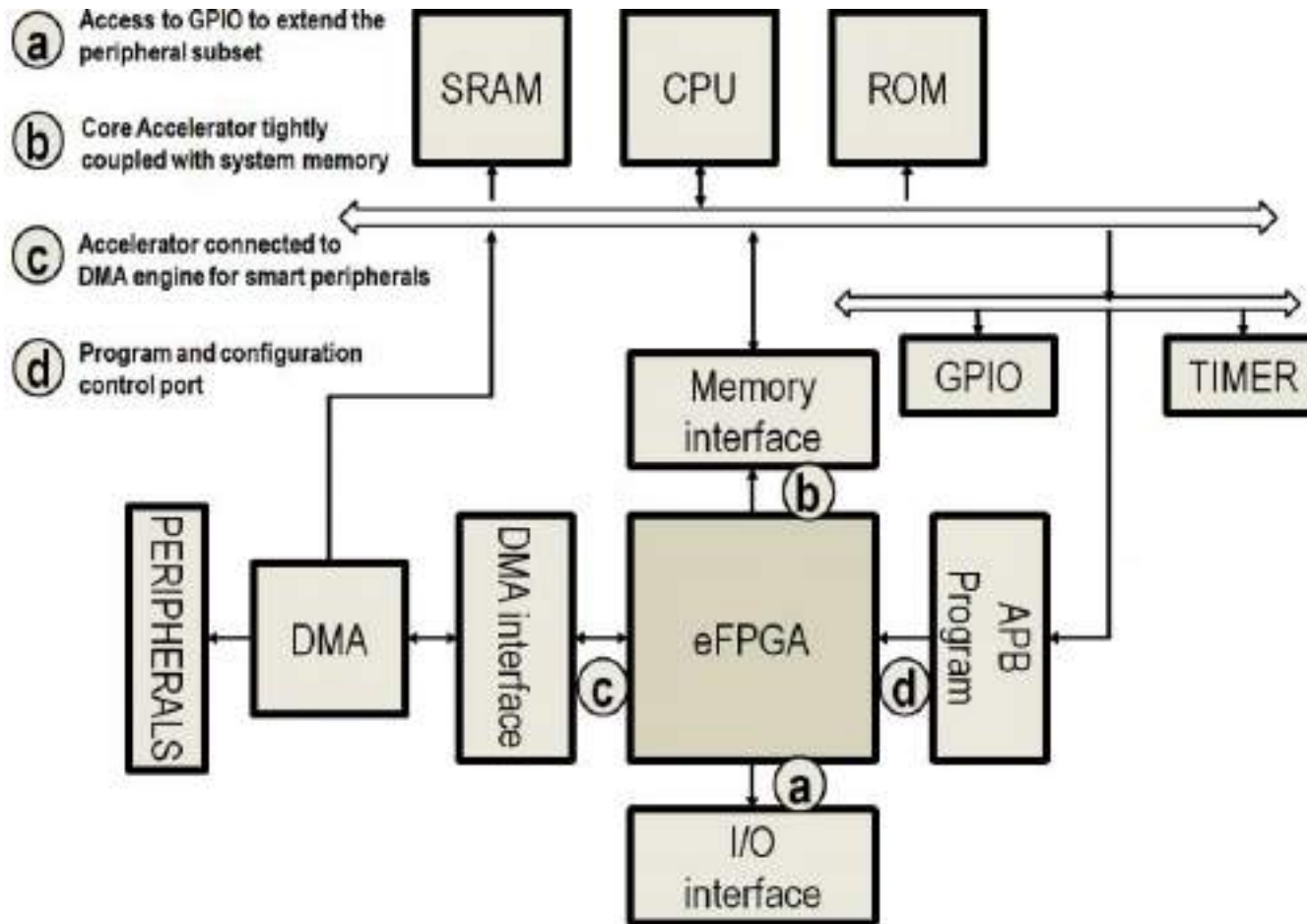
P. D. Schiavone et al., "Arnold: An eFPGA-Augmented RISC-V SoC for Flexible and Low-Power IoT End Nodes," TVLSI, vol. 29, no. 4, pp. 677-690, April 2021.

PULPissimo: very good platform for extensions



- **eFPGA added as accel.**
 - Easy plug and play
 - Configuration over APB
 - Additional ALU and memory
 - Uses the same memory
- **Multiple operation modes**
 - Configurable peripheral
 - Accelerator for core
 - Accelerator for independent I/O

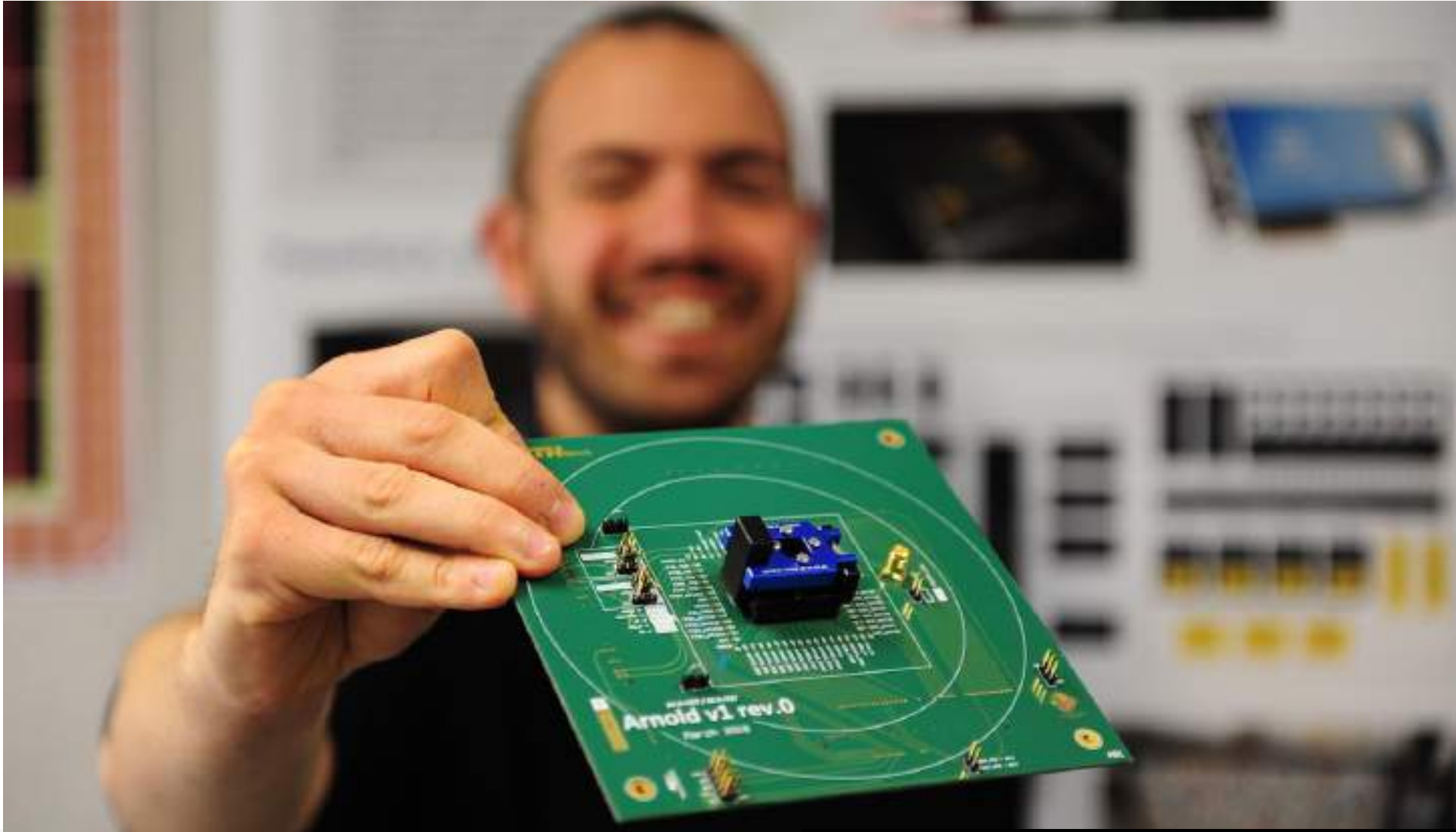
Experimental platform with many configurations



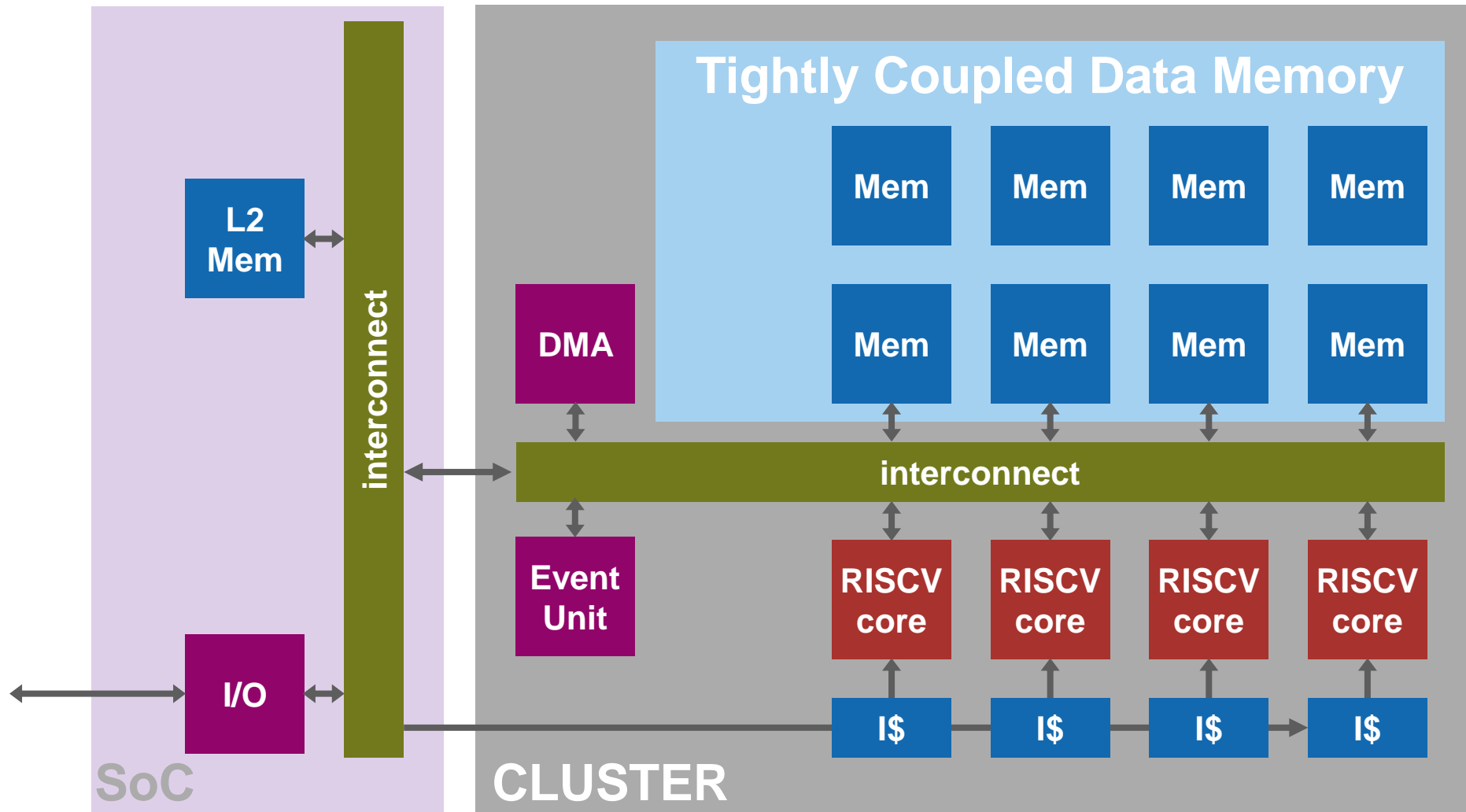
- I/O subsystem accel
 - 6.0mW, 2.5x
- Custom I/O interface
 - BNN interface 12.5mW 2.2x
- CPU accelerator
 - CRC 7.5mW 42x
- Many more ideas
 - Dynamic reconfiguration



Arnold test board with D. Schiavone



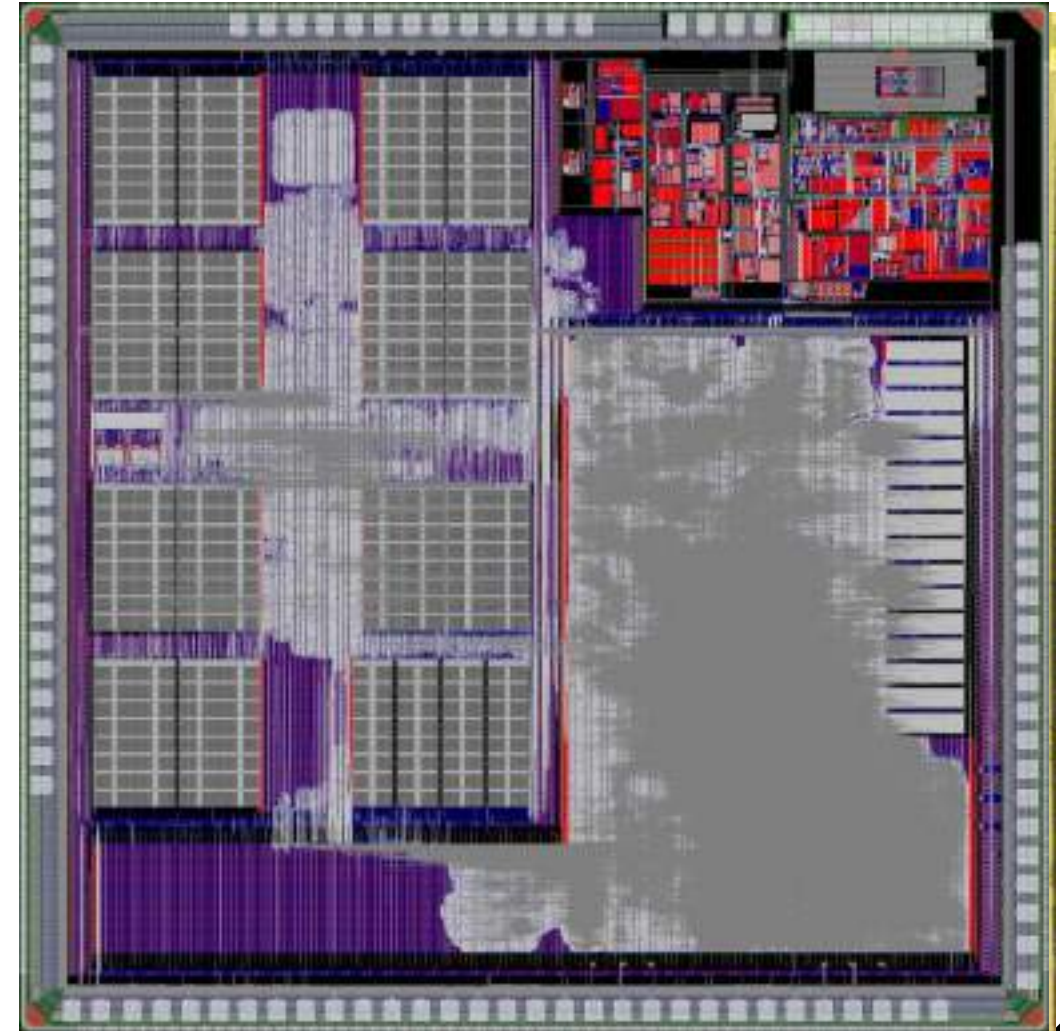
Full Multi-Cluster SoCs





Mr. Wolf (TSMC 40): 8+1 core IoT Processor

- **One cluster with**
 - 8 RISC-V cores
 - 2x shared FPU units
 - 64 kByte of TCDM
- **One controller with**
 - 512 kByte L2 RAM
 - Peripherals
- **On chip voltage regulators**
 - By Dolphin Integration



A. Pullini, D. Rossi, I. Loi, G. Tagliavini and L. Benini, "Mr.Wolf: An Energy-Precision Scalable Parallel Ultra Low Power SoC for IoT Edge Processing," in IEEE Journal of Solid-State Circuits, vol. 54, no. 7, pp. 1970-1981, July 2019.





On-chip regulators allow different power modes

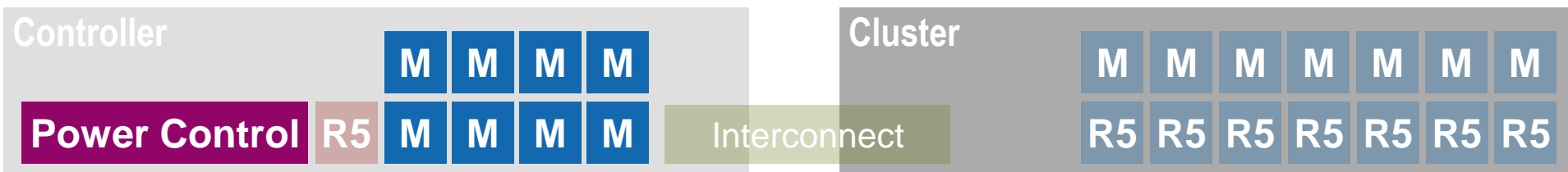
Power Mode	VDD	Frequency Range	Power
Deep Sleep	0.8 V	n.A.	72 μ W

ETH zürich



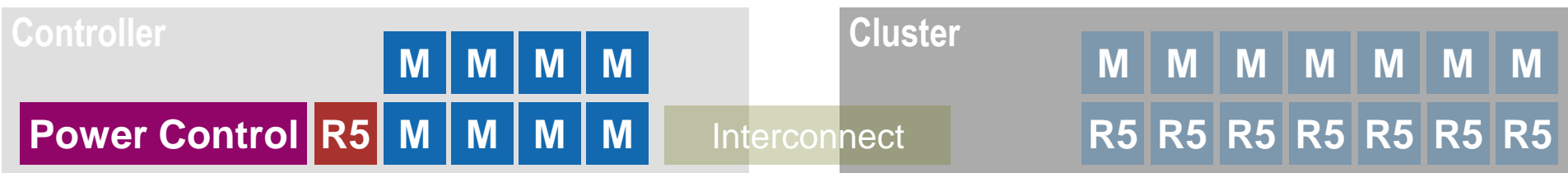
It is possible to keep memory state intact

Power Mode	VDD	Frequency Range	Power
Deep Sleep	0.8 V	n.A.	72 μ W
State Retentive Deep Sleep	0.8 V	n.A.	77 – 108 μ W



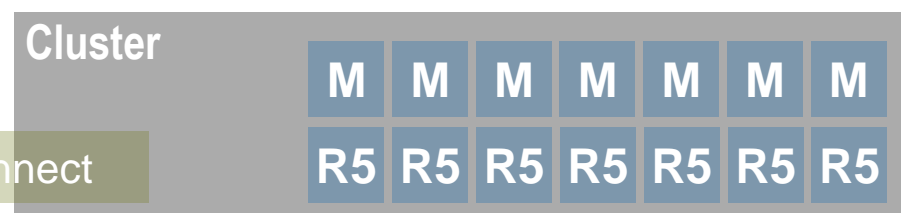
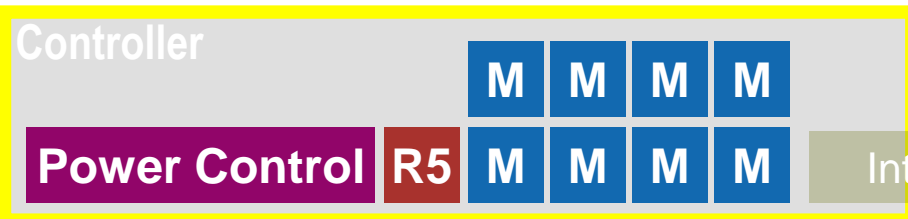
SoC is awake but is clock gated

Power Mode	VDD	Frequency Range	Power
Deep Sleep	0.8 V	n.A.	72 μ W
State Retentive Deep Sleep	0.8 V	n.A.	77 – 108 μ W
SoC Idle	0.8 – 1.1V	SoC clock gated	0.55 – 1.96 mW



Only SoC with a single RISC-V core running

Power Mode	VDD	Frequency Range	Power
Deep Sleep	0.8 V	n.A.	72 μ W
State Retentive Deep Sleep	0.8 V	n.A.	77 – 108 μ W
SoC Idle	0.8 – 1.1V	SoC clock gated	0.55 – 1.96 mW
SoC active	0.8 – 1.1V	32 kHz – 450 MHz	0.97 – 38 mW



Interconnect

Cluster is active, but clock gated

Power Mode	VDD	Frequency Range	Power
Deep Sleep	0.8 V	n.A.	72 μ W
State Retentive Deep Sleep	0.8 V	n.A.	77 – 108 μ W
SoC Idle	0.8 – 1.1V	SoC clock gated	0.55 – 1.96 mW
SoC active	0.8 – 1.1V	32 kHz – 450 MHz	0.97 – 38 mW
Cluster Idle	0.8 – 1.1V	Cluster clock gated	1.2 – 4.6 mW



Cluster with 8 RISC-V cores is active

Power Mode	VDD	Frequency Range	Power
Deep Sleep	0.8 V	n.A.	72 μ W
State Retentive Deep Sleep	0.8 V	n.A.	77 – 108 μ W
SoC Idle	0.8 – 1.1V	SoC clock gated	0.55 – 1.96 mW
SoC active	0.8 – 1.1V	32 kHz – 450 MHz	0.97 – 38 mW
Cluster Idle	0.8 – 1.1V	Cluster clock gated	1.2 – 4.6 mW
Cluster Active	0.8 – 1.1V	32 kHz – 350 MHz	1.6 – 153 mW

Controller



Interconnect

Cluster





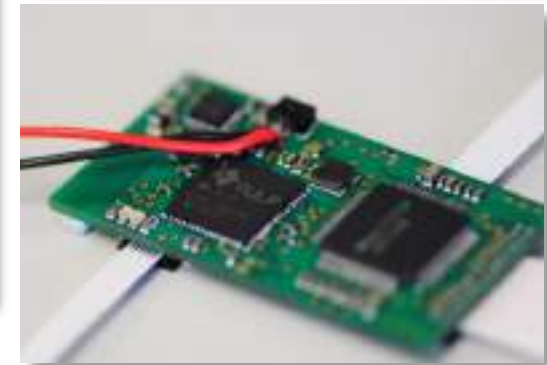
Our OpenPULP release is Mr. Wolf

- **With Mr. Wolf, most of what we have is open sourced**
 - This is a **complex IoT processor**, not like the much simpler PULPino
 - 8 + 1 cores, FPU, shared accelerators, multiple power down modes.
- **Still many parts can still not be open source**
 - Technology specific information, P&R scripts
 - Memory macros, selected cuts, their performance
 - I/O cells
 - FLL, analog macros, I/O cells, memory cuts (affects performance), P&R scripts
- **Interesting industry collaboration**
 - Greenwaves, BitCraze, Dolphin



Mr. Wolf has been used in multiple systems

- Designed as an application processor
 - We still build boards with it
 - Despite only 200 manufactured
- **Widespread industrial use:**
 - Dolphin IP was validated on this chip
 - Greenwaves GAP8 is based on the open source release OpenPULP
 - BitCraze AI Deck is related

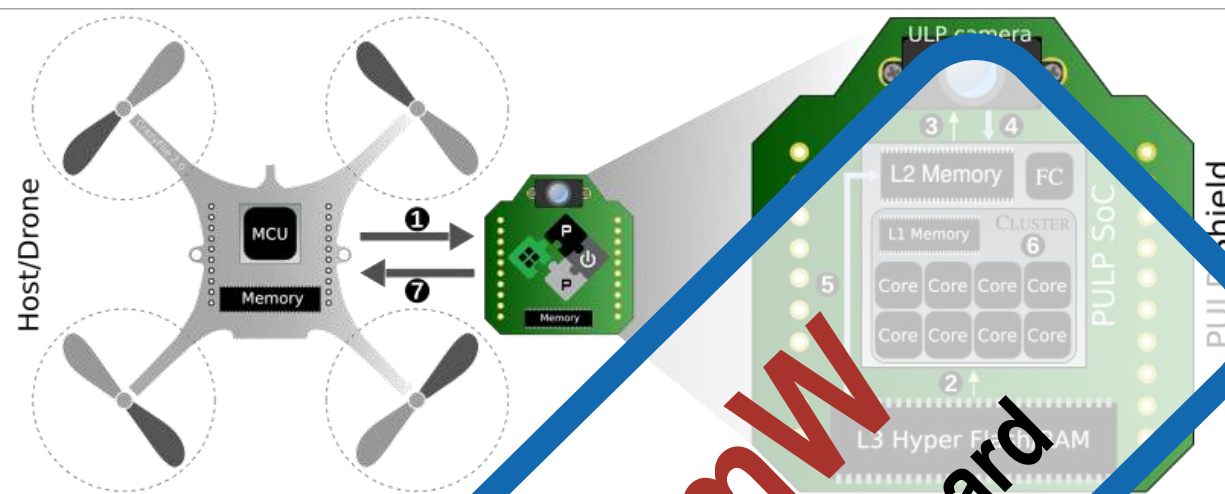


GREENWAVES
TECHNOLOGIES

bitcraze

Complete Application: DroNET on NanoDrone

- 1 Init interrupt (GPIO)
- 2 Load binary (HyperBus)
- 3 Configure camera (I2C)
- 4 Grab frames (µDMA)
- 5 Load weights (HyperBus)
- 6 PULP computation
- 7 Write-back results (SPI)



Pluggable PCB:
PULP-Shield

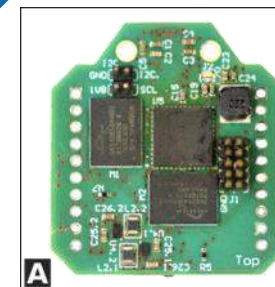
- ~5g, 30x28mm
- GAP8 SoC
- 8 MB HDRAM
- 16 MB HFlash
- QVGA ULP
- HiMax camera
- Crazyflie 2.0 nano-drone (27g)



Copyright 2019 © ETH zürich



Credit: F. K. Gürkaynak & Daniele Palossi



A

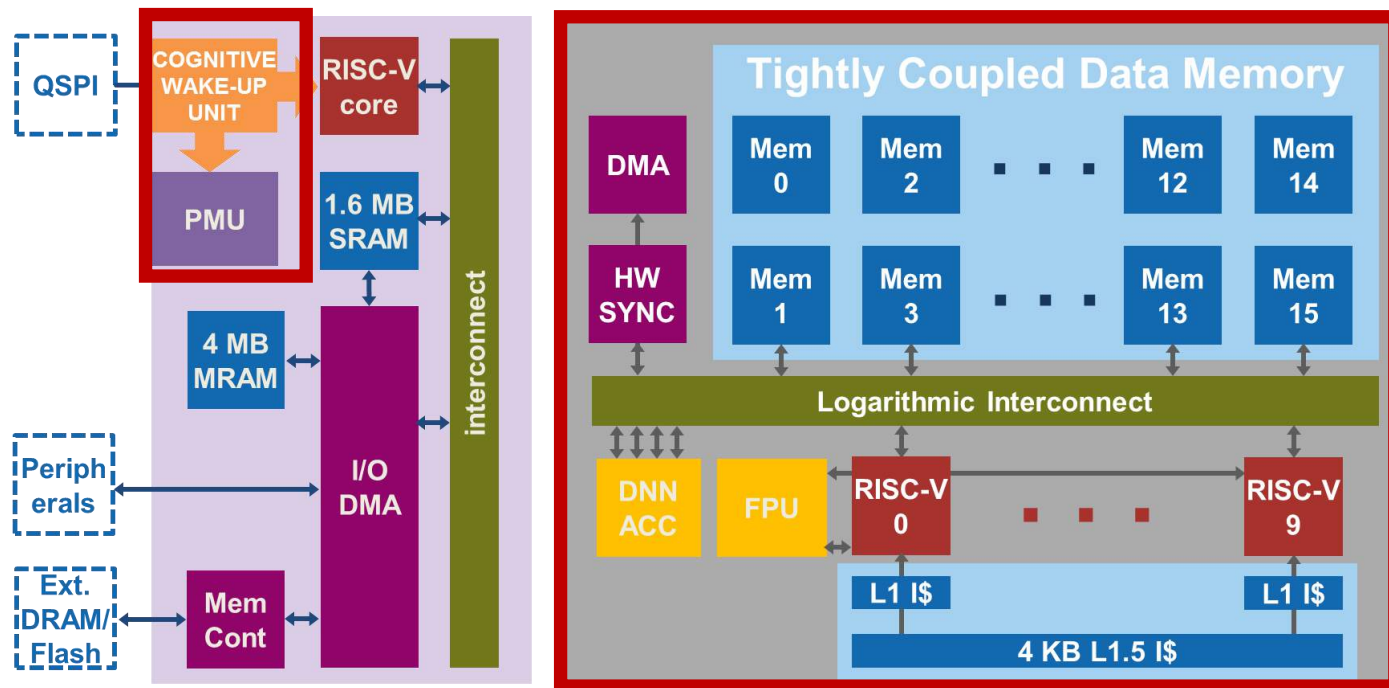


B

Only onboard computation for autonomous flight + obstacle avoidance
no human operator, no ad-hoc external signals, and no remote base-station!

VEGA: Extreme Edge IoT Processor

- Fully programmable RISC-V based cluster targeting highly dynamic Near-Sensor Analytic Applications (NSAA)
- 1.7 μW cognitive unit for autonomous wake-up from retentive sleep mode



ETH zürich

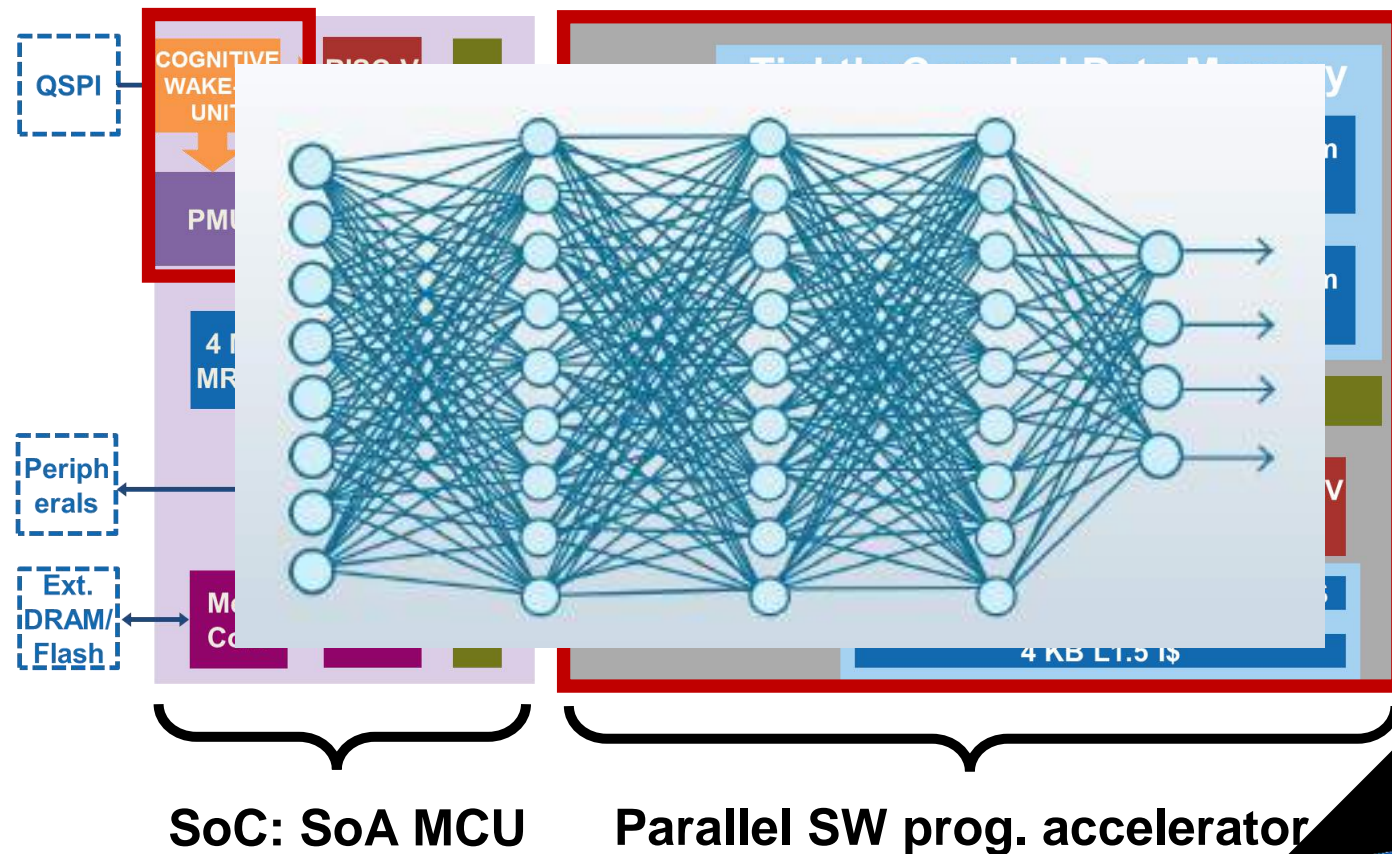


GREENWAVES
TECHNOLOGIES

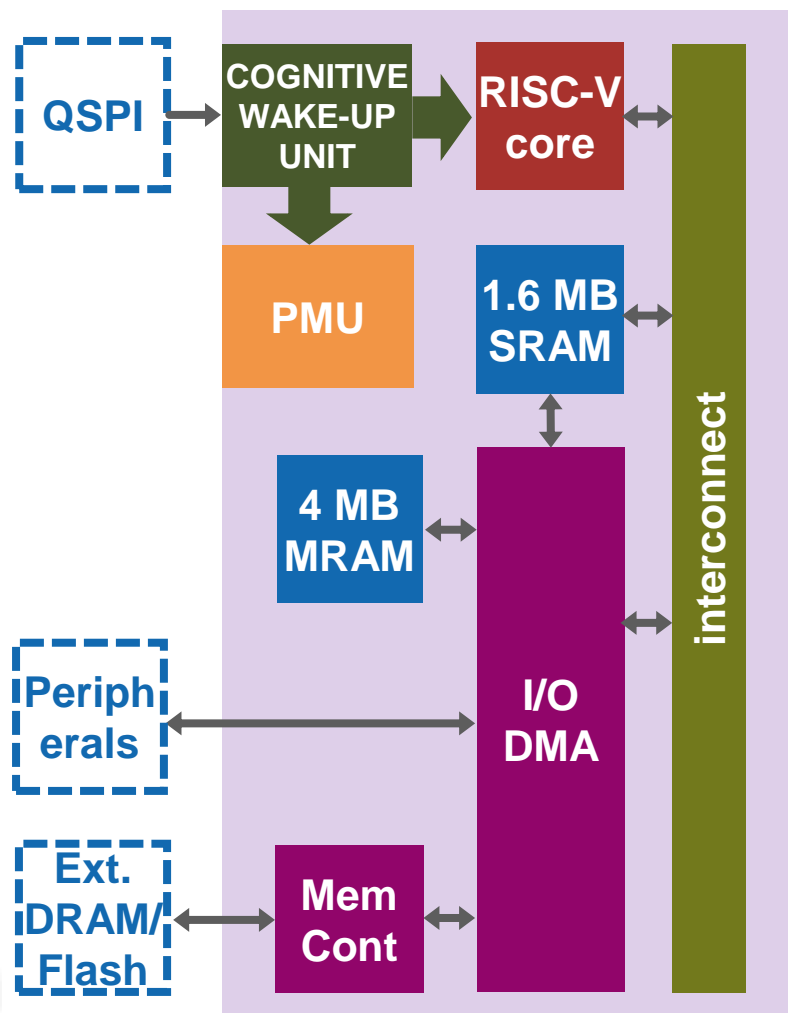


VEGA: Extreme Edge IoT Processor

- Fully programmable RISC-V based cluster targeting highly dynamic Near-Sensor Analytic Applications (NSAA)
- 1.7 μ W cognitive unit for autonomous wake-up from retentive sleep mode
- Fully integrated execution of real-life DNN from 4 MB of non-volatile MRAM (first time for an IoT end-node)



SoC Overview



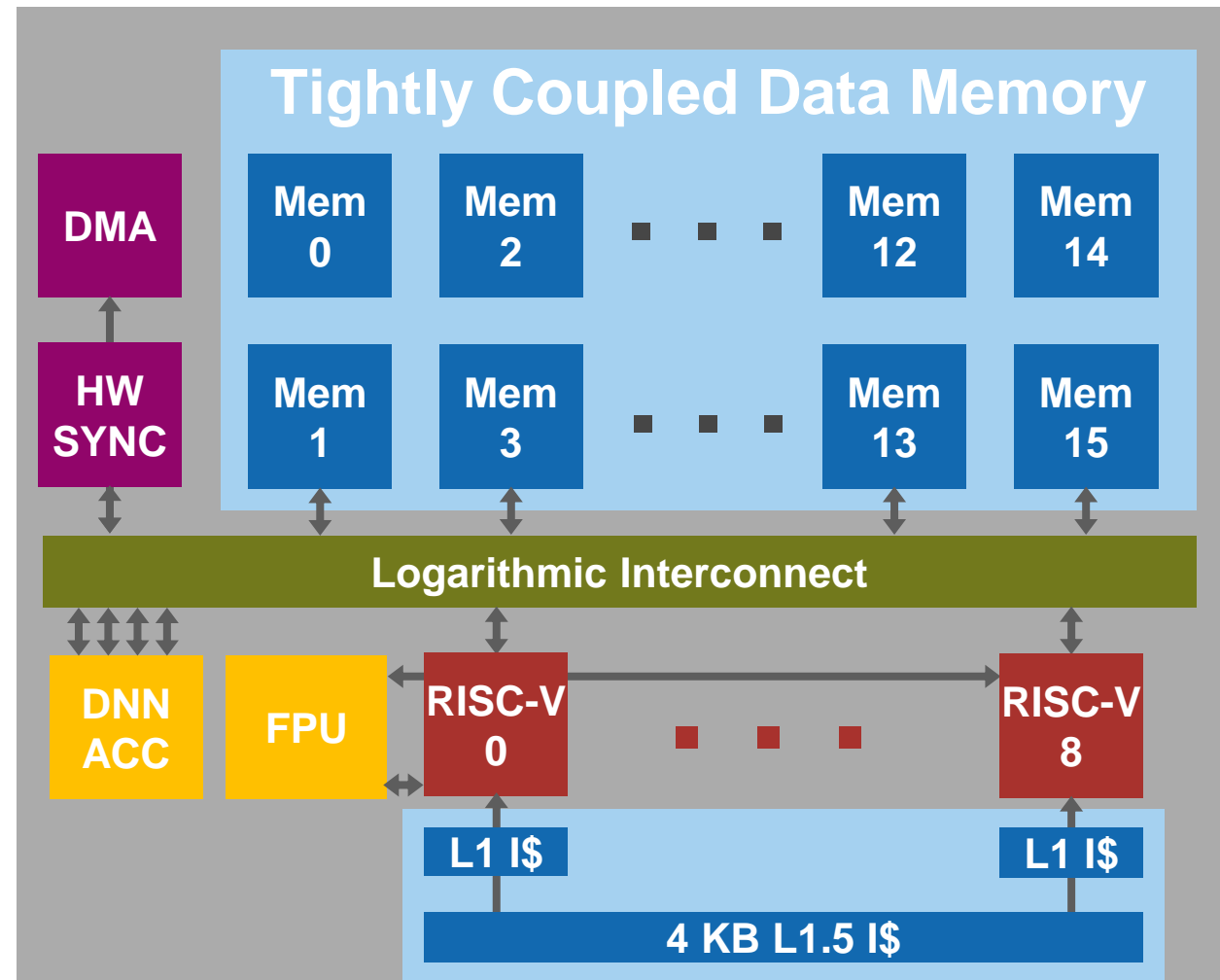
- 32-bit RISC-V core (Fabric Controller)
- 1.6 MB L2 SRAM
- 4 MB non-volatile MRAM
- Standard set of peripherals (SPI, I2C, UART, CSI2...)
- Off-chip memory (*HyperRAM™ DRAM / Flash)
- Autonomous I/O DMA
- Cognitive smart wake-up
- 3 Frequency-Locked Loops (FLL)
- 2 Voltage regulators (HP/LP) + 1 LDO (COTS) + PMU

*<https://www.cypress.com/products/hyperram-memory>



Software-Programmable Accelerator

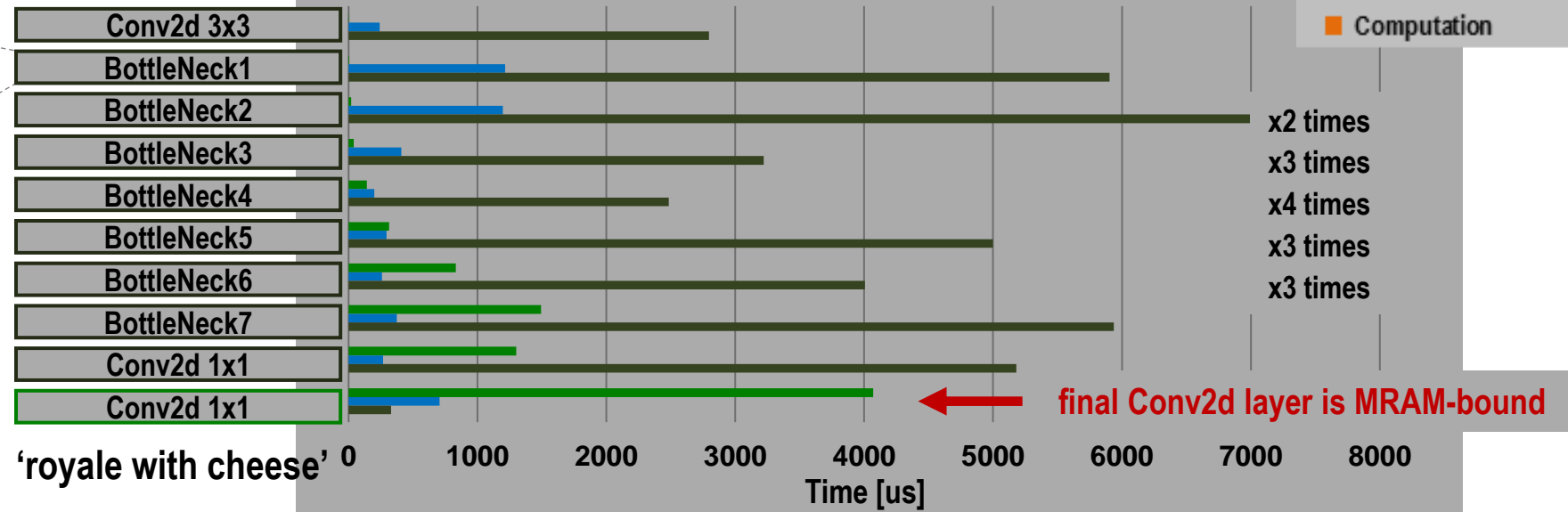
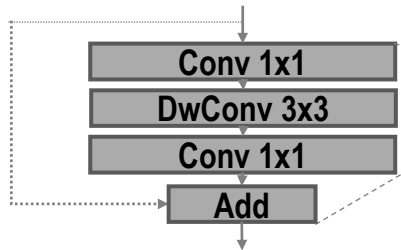
- 9 RISC-V DSP cores
- 128KB 16-Banks TCDM (scratchpad, no cache)
- Single-cycle latency, word-level interleaved Interconnect
- DMA for explicit memory mgmt.
- I\$: 9x 0.5kB L1 I\$ + 4KB L1.5 I\$
- Hardware Synchronizer (HW SYNC)
- Shared SIMD Floating-Point Unit (FPU)
- DNN Accelerator



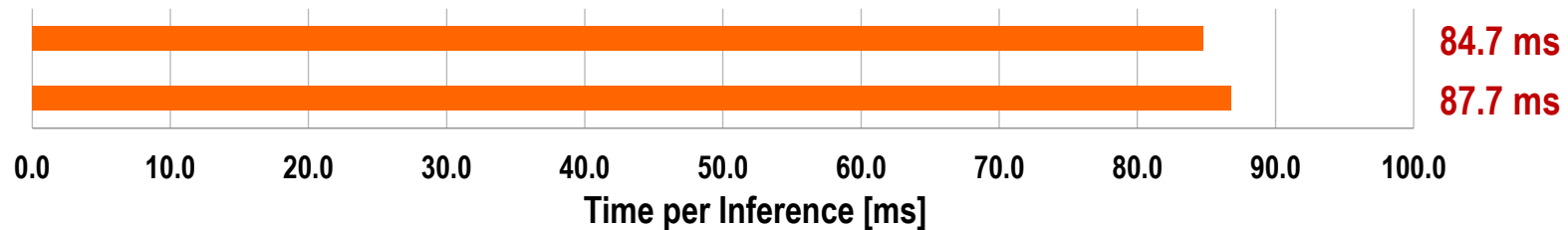
Full DNN Performance (MobileNetV2)



@ Vdd_SOC=0.8V, f_SOC=250 MHz, f_CL=250 MHz



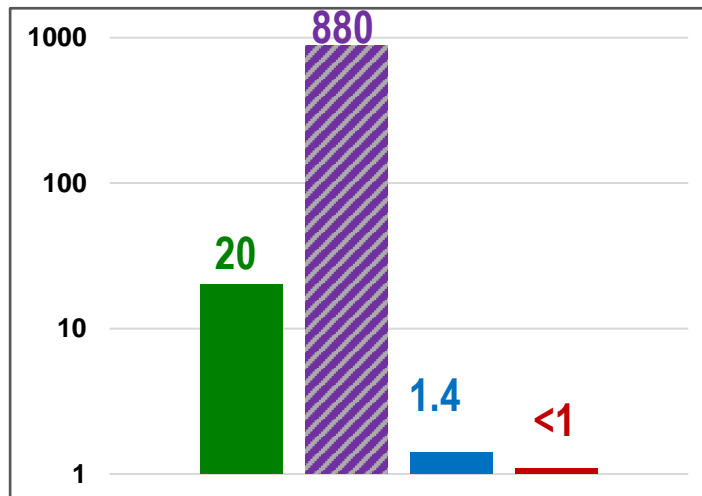
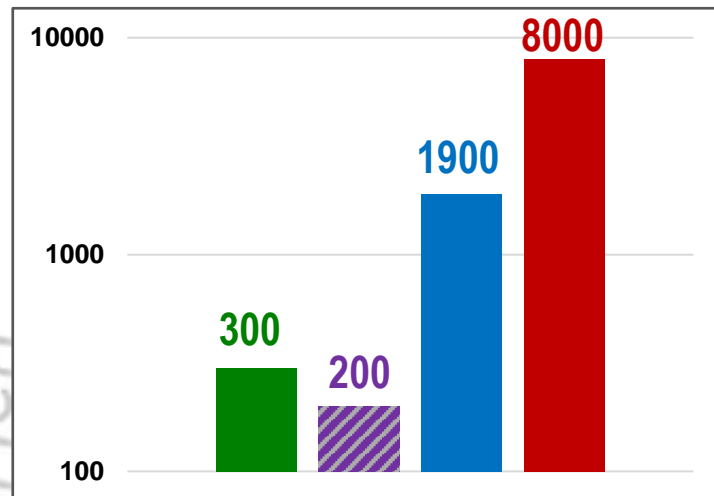
weights on MRAM
weights on HyperRAM



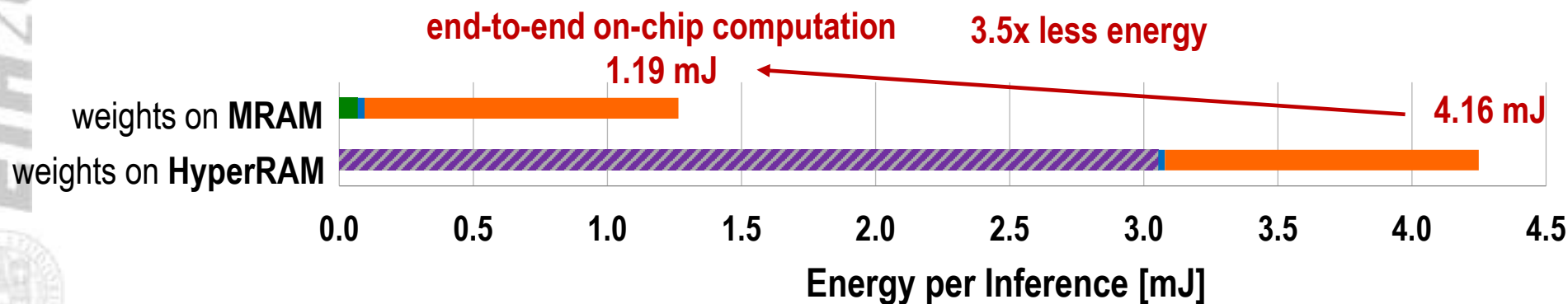
Full DNN Energy (MobileNetV2)

Bandwidth [MB/s]

Energy per byte [pJ/B]



- HyperRAM (ext)↔L2 w/ I/O DMA
- MRAM↔L2 w/ I/O DMA
- L2↔L1 w/ Cluster DMA
- L1 access



World-Record Efficiency Among IoT Processors

	SleepRunner [2]	Mr.Wolf [3]	Samurai [4]	Vega (work)
Embedded NVM	-	-	-	4 MB MRAM
Wake-up Sources	WiC	GPIO, RTC	WuR, RTC, In GPIO	GPIO, RTC, Cognitive
Best Int Perf.	31 MOPS (32b)	12.1 GOPS	1.5 GOPS	15.6 GOPS
Best.Int Eff.	97 MOPS/mW (32b)	190 GOPS/W	230 GOPS/W	14 GOPS/W
@ Perf.	@ 18.6 MOPS (32b)	@ 3.8 GOPS	@ 3.8 MOPS	@ 7.6 GOPS
Best FP Perf.	-	1 GFLOPS	-	2 GFLOPS
Best FP Eff.	-	18 GFLOPS/W	-	79 GFLOPS/W
@Perf	-	@ 350 MFLOPS	-	@ 1 GFLOPS
Best ML Perf.	-	-	36 GOPS	32.2 GOPS
Best ML Eff.	-	-	1.3 TOPS/W	1.3 TOPS/W
@Perf	-	-	@ 2.8 GOPS	@ 15.6 GOPS

Presented at ISSCC 2021

3.2x Efficiency
@ 2x Performance

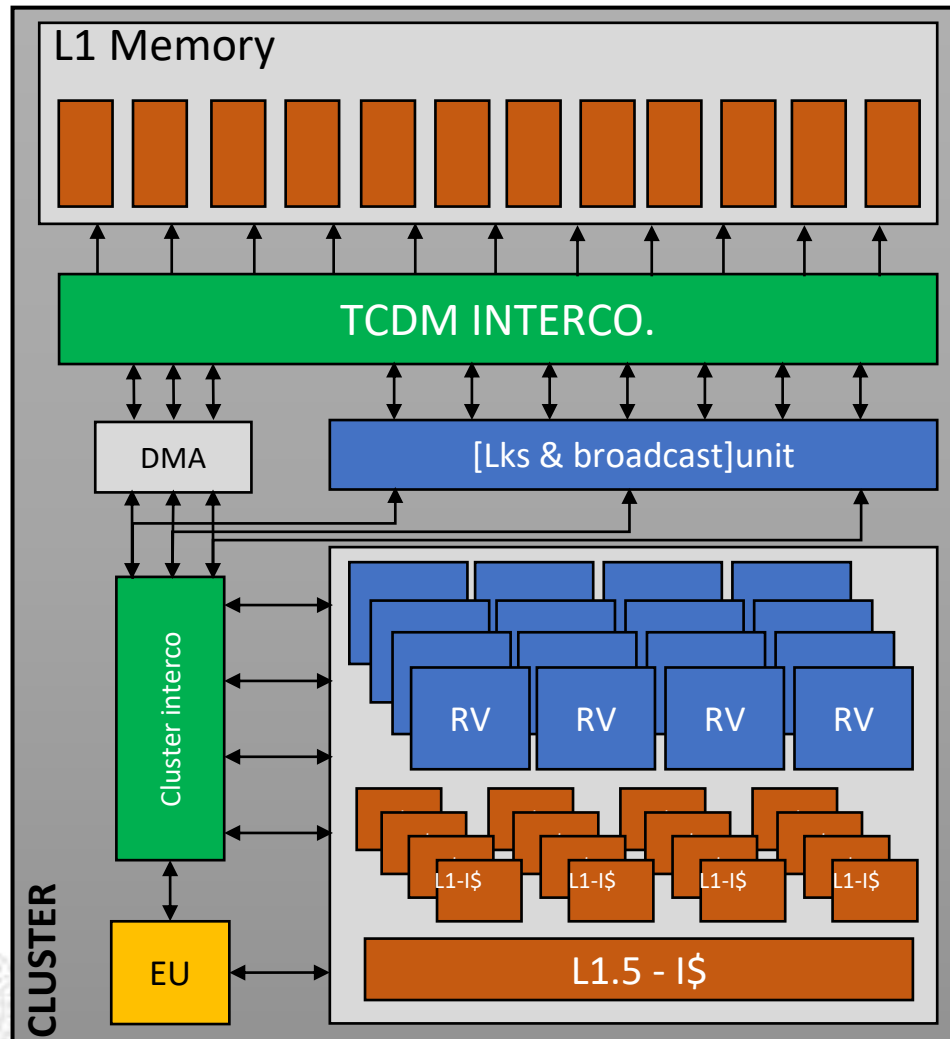
4.3x Efficiency
@ 2.8x Performance

Similar Efficiency
@ 5.5x Performance

D. Rossi *et al.*, "4.4 A 1.3TOPS/W @ 32GOPS Fully Integrated 10-Core SoC for IoT End-Nodes with 1.7 μ W Cognitive Wake-Up From MRAM-Based State-Retentive Sleep Mode," 2021 IEEE International Solid-State Circuits Conference (ISSCC), 2021, pp. 60-62



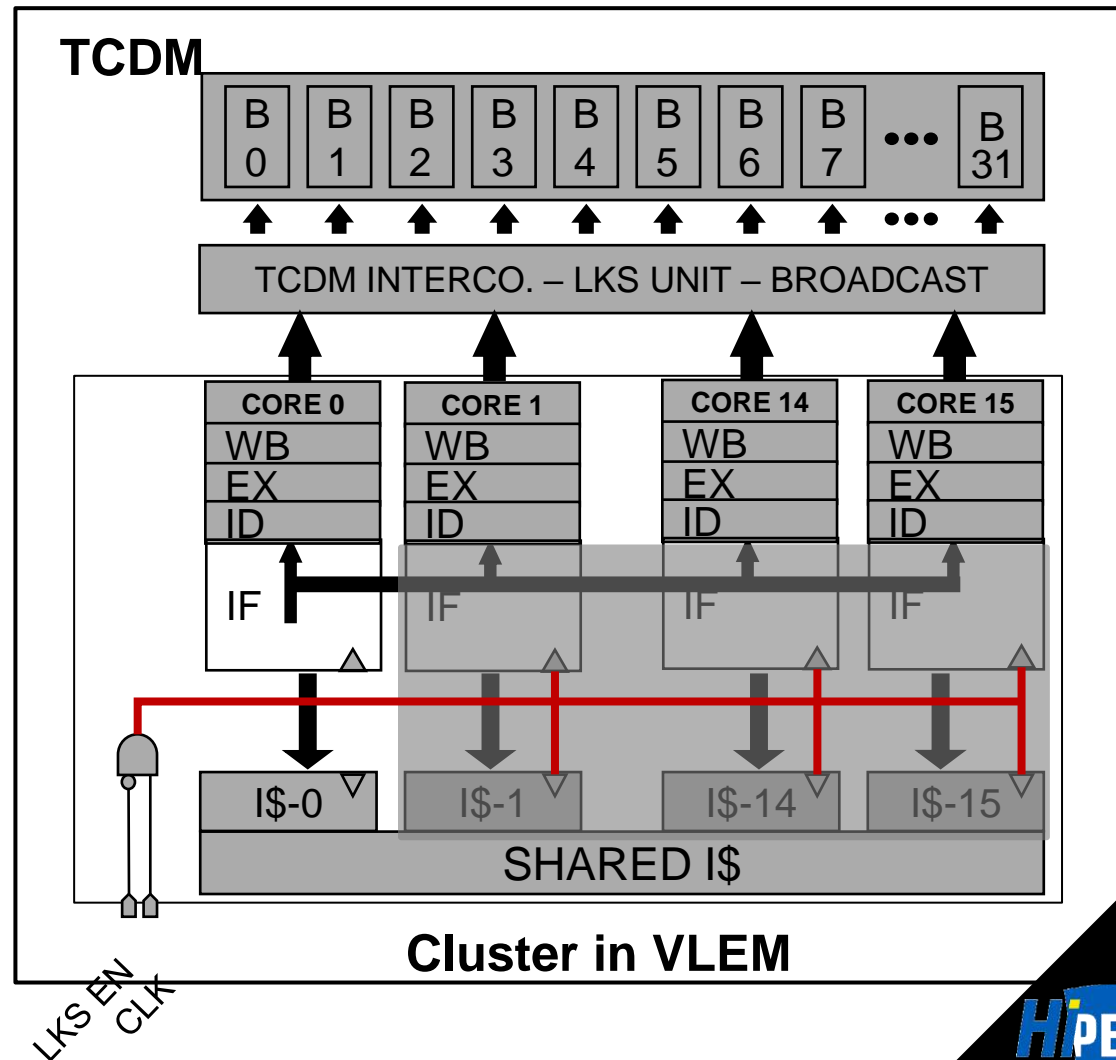
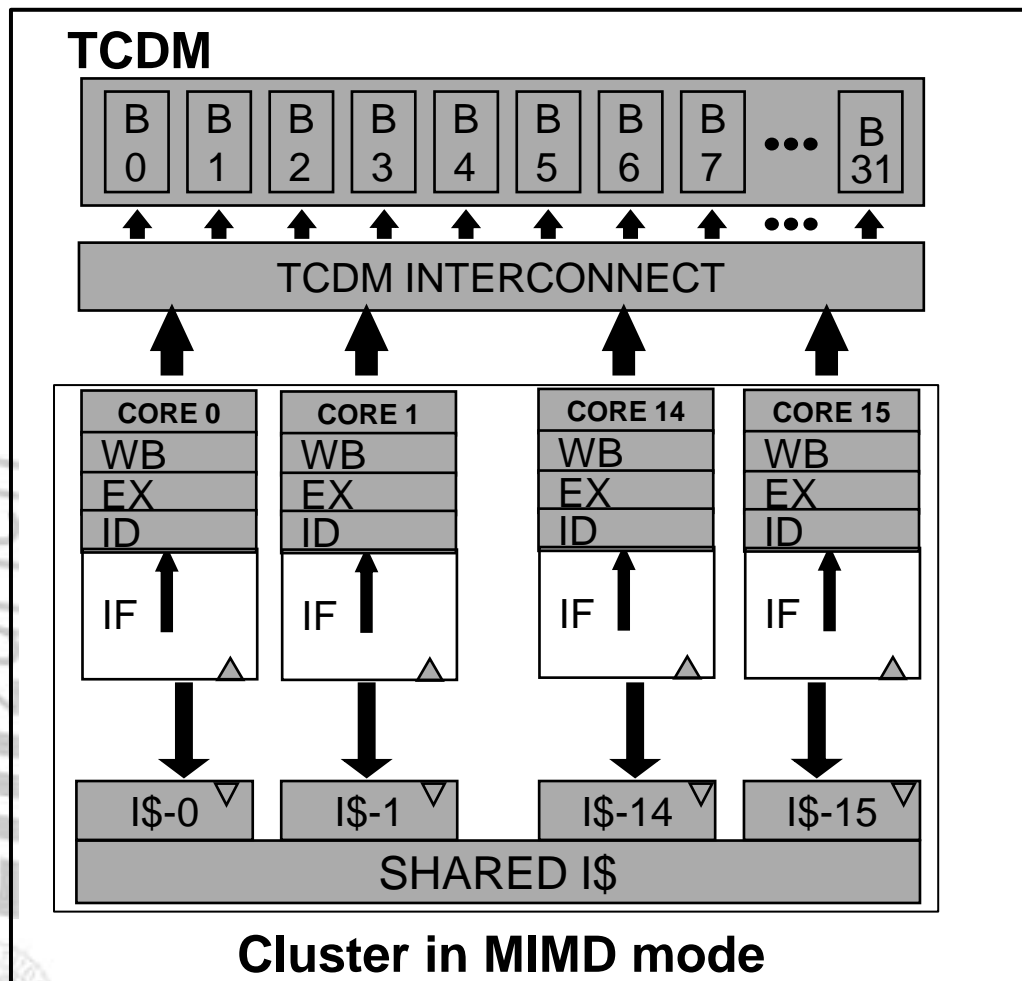
DUSTIN: Mixed-Precision Cluster



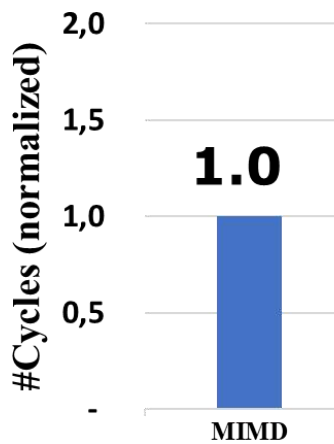
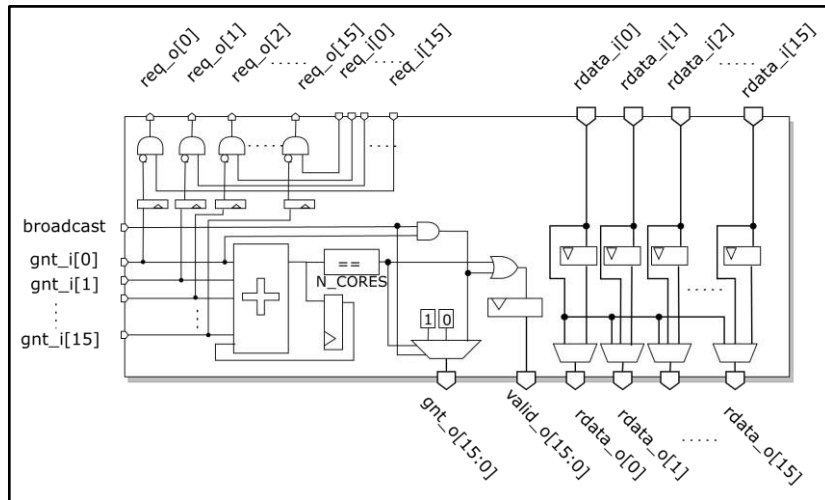
Accelerator Cluster

- ❑ 16 RI5CY (*) cores augmented with 2b-to-32b SIMD instructions;
- ❑ Software Configurable *Vector Lockstep Execution Mode (VLEM)*;
- ❑ Single-cycle latency TCDM interco. leveraging a req/gnt protocol, word-level interleaved scheme.
- ❑ 128 kB of Shared Tightly-Coupled L1 Data Memory;
- ❑ Hierarchical Instruction Cache;
- ❑ High performance DMA (L2 <-> L1);
- ❑ Event Unit supporting efficient synchronization among the cores;

Vector Lockstep Exec. Mode (VLEM)

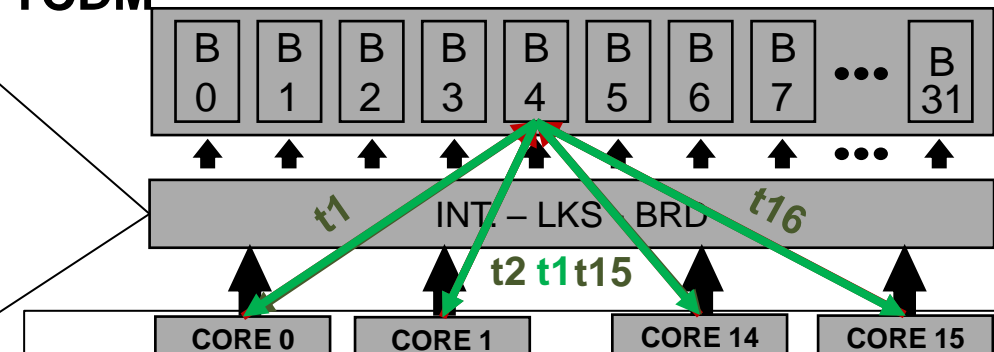


VLEM: Broadcast Unit



MatMul exec. kernels

TCDM



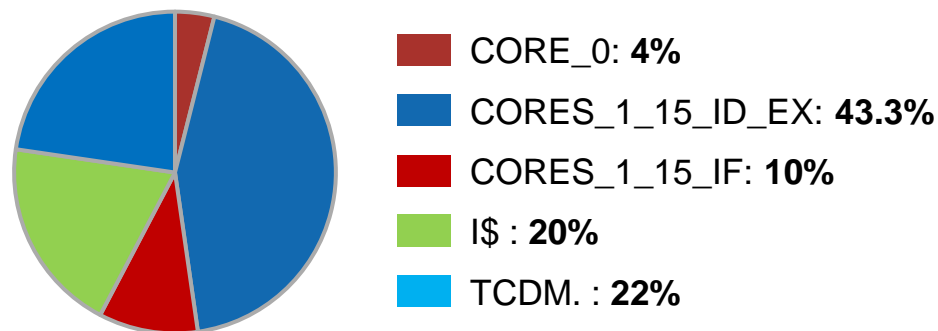
- **Overhead:** at least 16 clk cycles to unlock the execution in case of concurrent accesses;
- **Solution:**
 - eliminate the overhead in case of access to the same mem address → **BROADCAST UNIT**.
 - Misalign static data and stacks to avoid accesses to the same mem bank;

Cluster in VLEM

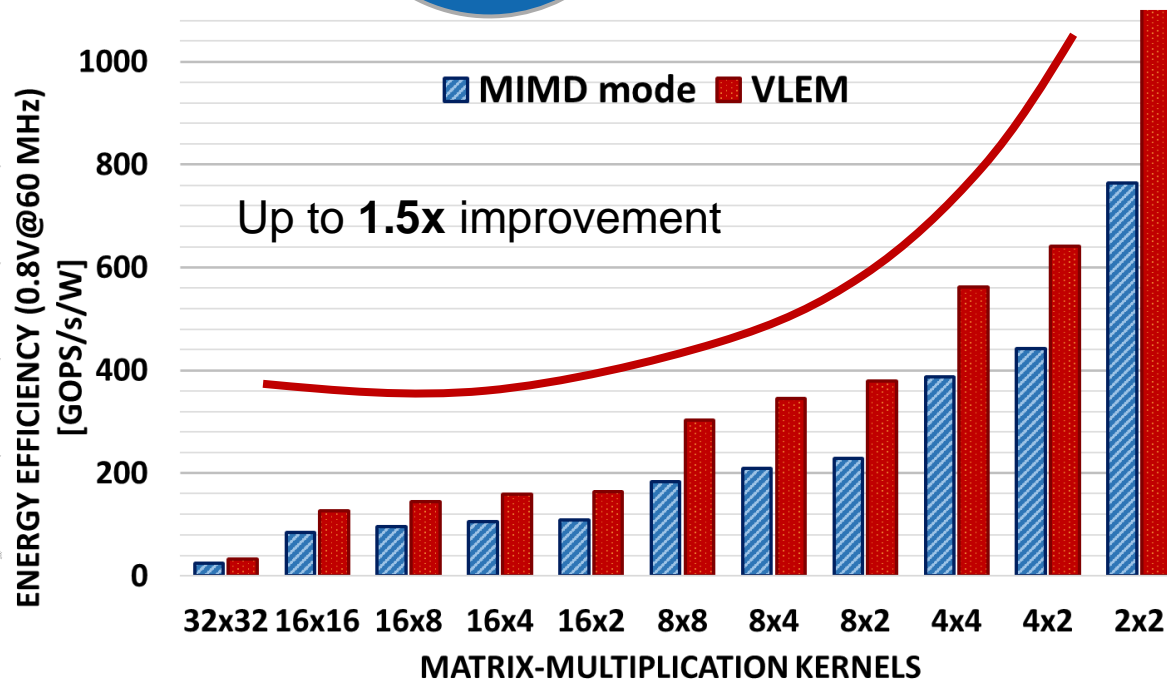
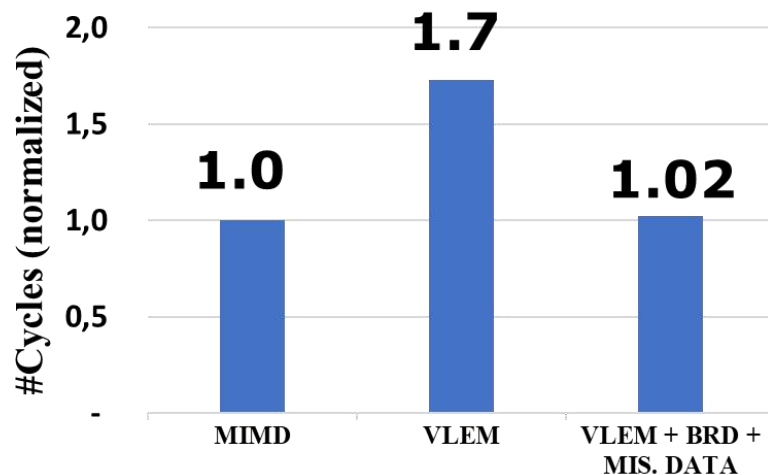
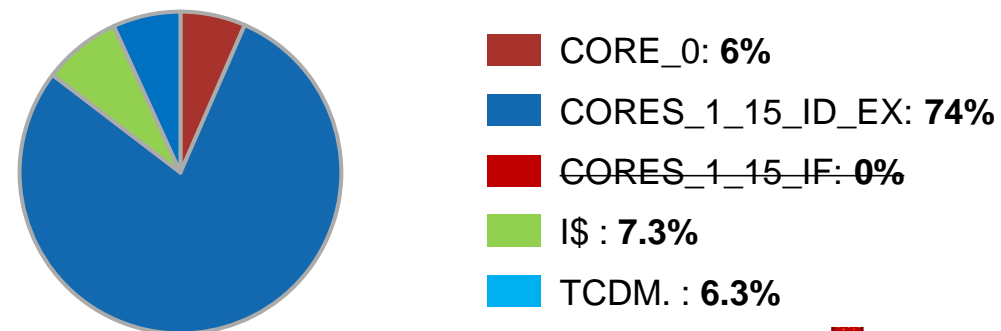
LKS EN
CLK

Energy Efficiency on MatMul kernels

8x4 MatMul Power in MIMD: **29.2 mW**



Power in VLEM: **16.1 mW** (~45% reduction wrt MIMD)



Comparison with the SoA

	SleepRunner [6]	SamuraiAI [7]	Mr. Wolf [8]	Vega [9]	Dustin (this work)
Technology	CMOS 28nm FDSOI	CMOS 28nm FDSOI	CMOS 40nm LP	CMOS 22nm FDSOI	CMOS 65nm
Die Area	0.68 mm ²	4.5 mm ²	10 mm ²	12 mm ²	1 mm ²
Applications	IoT GP	IoT GP + DNN	IoT GP + DNN	IoT GP + DNN + QNNs	IoT GP + DNN + QNNs
CPU/ISA	CM0DS Thumb-2 subset	1x RI5CY RVC32IMFXpulp	9x RI5CY RVC32IMFXpulp	16x RI5CY RVC32IMFXpulp + SF	16x RISC-V CORES (RISC-V)
Int Precision (bits)	32	8, 16, 32	8, 16, 32	8, 16, 32	2, 4, 8, 16, 32 (plus Mixed-Precision)
Supply Voltage	0.4 - 0.8 V	0.45 - 0.9 V	0.8 - 1.2 V	0.5 - 0.8 V	0.8 - 1.2 V
Max Frequency	80 MHz	350 MHz	450 MHz	450 MHz	205 MHz
Power Envelope	320 μ W	96 mW	153 mW	4.4 mW	156 mW
¹ Best Integer Performance	31 MOPS (32b)	1.5 GOPS (8b) ²	12.1 GOPS (8b)	15.6 GOPS (8b)	15 GOPS (8b) 30 GOPS (4b) 58 GOPS (2b)
¹ Best Integer Efficiency	97 MOPS/mW @ 18.6 MOPS (32b)	230 GOPS/W @ 110 MOPS (8b)	190 GOPS/W @ 3.8 GOPS (8b)	614 GOPS/W @ 7.6 GOPS	303 GOPS/W @ 4.4 GOPS (8b) 570 GOPS/W @ 8.8 GOPS (4b) 1152 GOPS/W @ 17.3 GOPS (2b)

Presented at
ESSCIRC 2021

Dustin supports Mixed-Precision computation in HW

Better efficiency wrt solutions in 28nm and 40nm tech node

Comparable efficiency wrt Vega (22 nm)

A.Garofalo et. al. "A 1.15 TOPS/W, 16-Cores Parallel Ultra-Low Power Cluster with 2b-to-32b Fully Flexible Bit-Precision and Vector Lockstep Execution Mode", ESSCIRC 2021





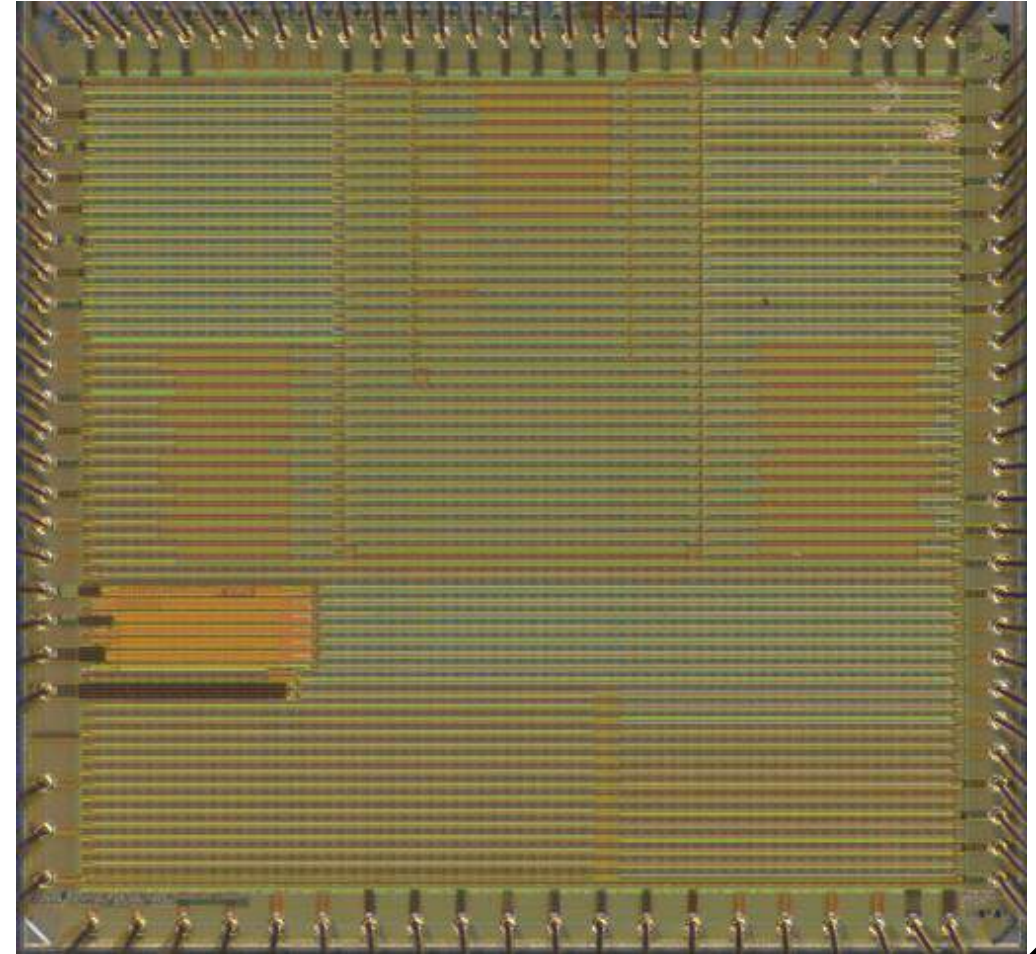
Moving to HPC: Kosmodrom

■ Globalfoundries 22FDX

- In 2018, most advanced node for us
- Minimum size 3mm x 3mm
 - That fits about **100 million transistors**
- Allows body biasing

■ With great power comes...

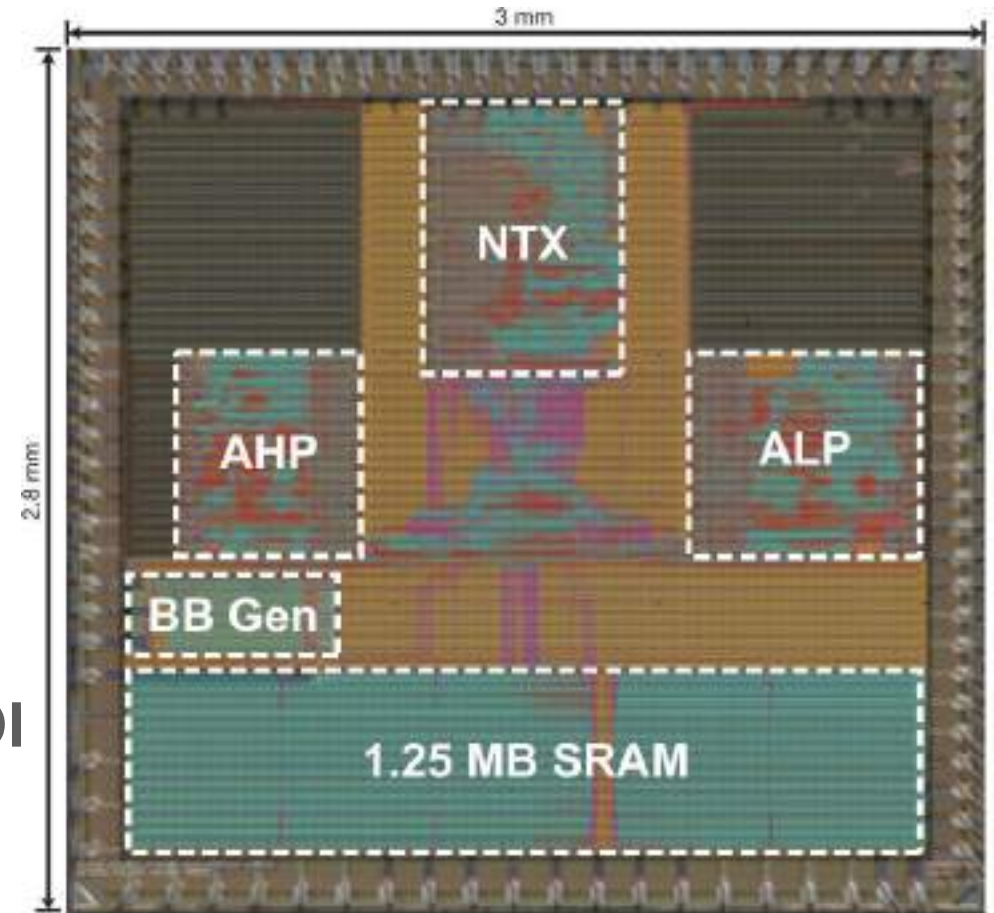
- Designs in 22FDX are more involved
- More blocks, more functionality
 - More things that can go wrong
- Challenging design
- Collaboration with Globalfoundries



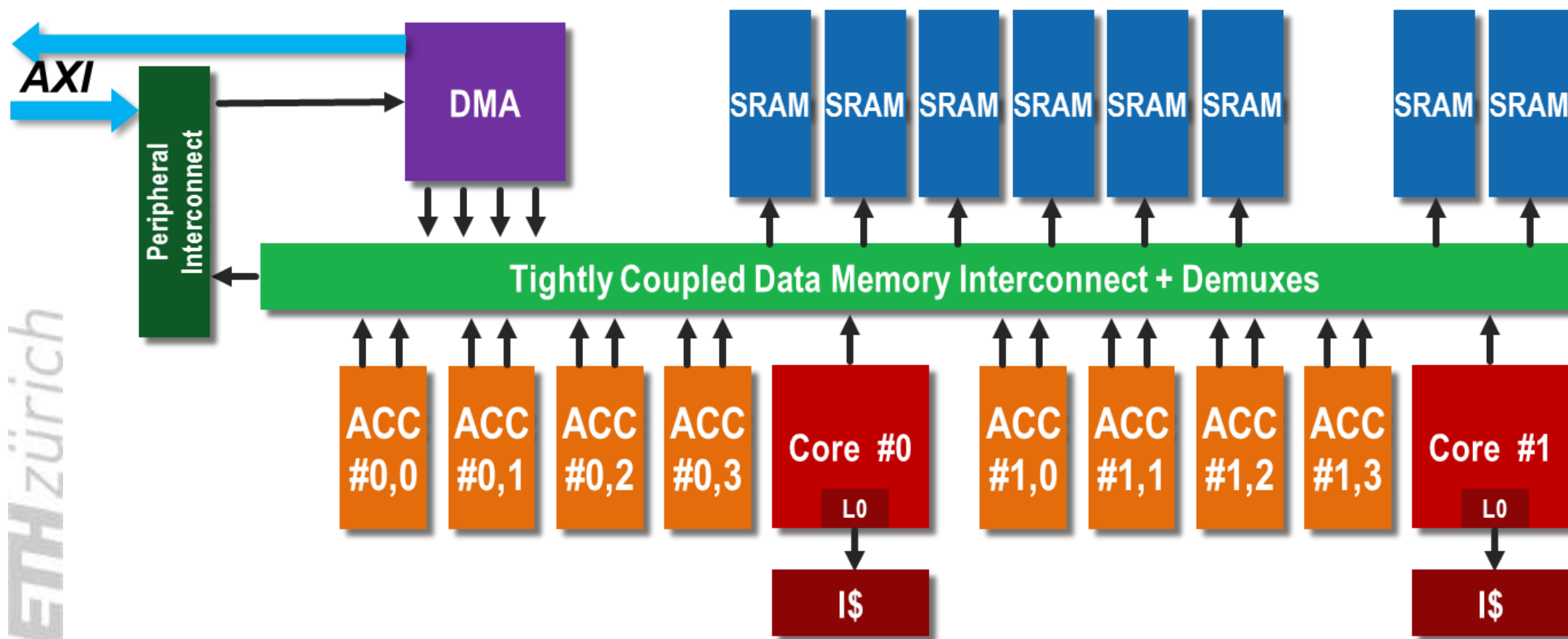


Kosmodrom: Main components

- **2x Ariane 64b RISC-V cores**
 - AHP optimized for high speed
 - ALP optimized for low power
- **Automatic Body Bias Gen.**
 - IP by INVECAS
 - Allows body bias to be tuned
- **NTX: Neural Training Accelerator**
 - 260 Gflops/Watt efficiency
- **Common infrastructure**
 - SRAM, Debug, I/Os



Fine-Grained Shared-Memory Accelerators



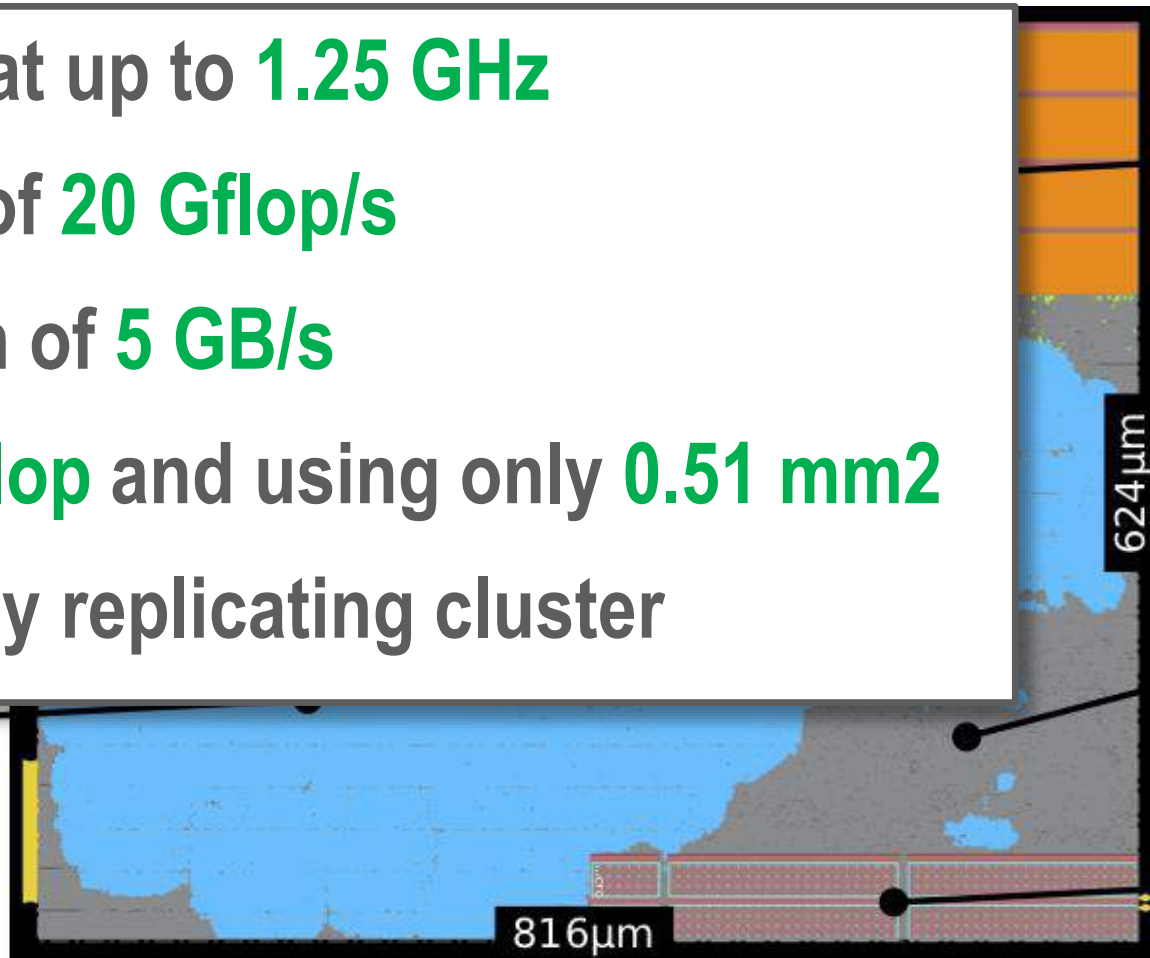
Similar concept as **OpenPULP**, but **fewer RISC-V cores** and **more accelerators**



NTX uses 1 RISC-V core to control 8 units

- NTX runs at up to **1.25 GHz**
- Compute of **20 Gflop/s**
- Bandwidth of **5 GB/s**
- At **9.3 pJ/flop** and using only **0.51 mm²**
- Scale up by replicating cluster

coprocessors



64 kB TCDM
in 32 banks

1x RISC-V
processor and
peripherals

2 kB ICACHE

816 μm

624 μm



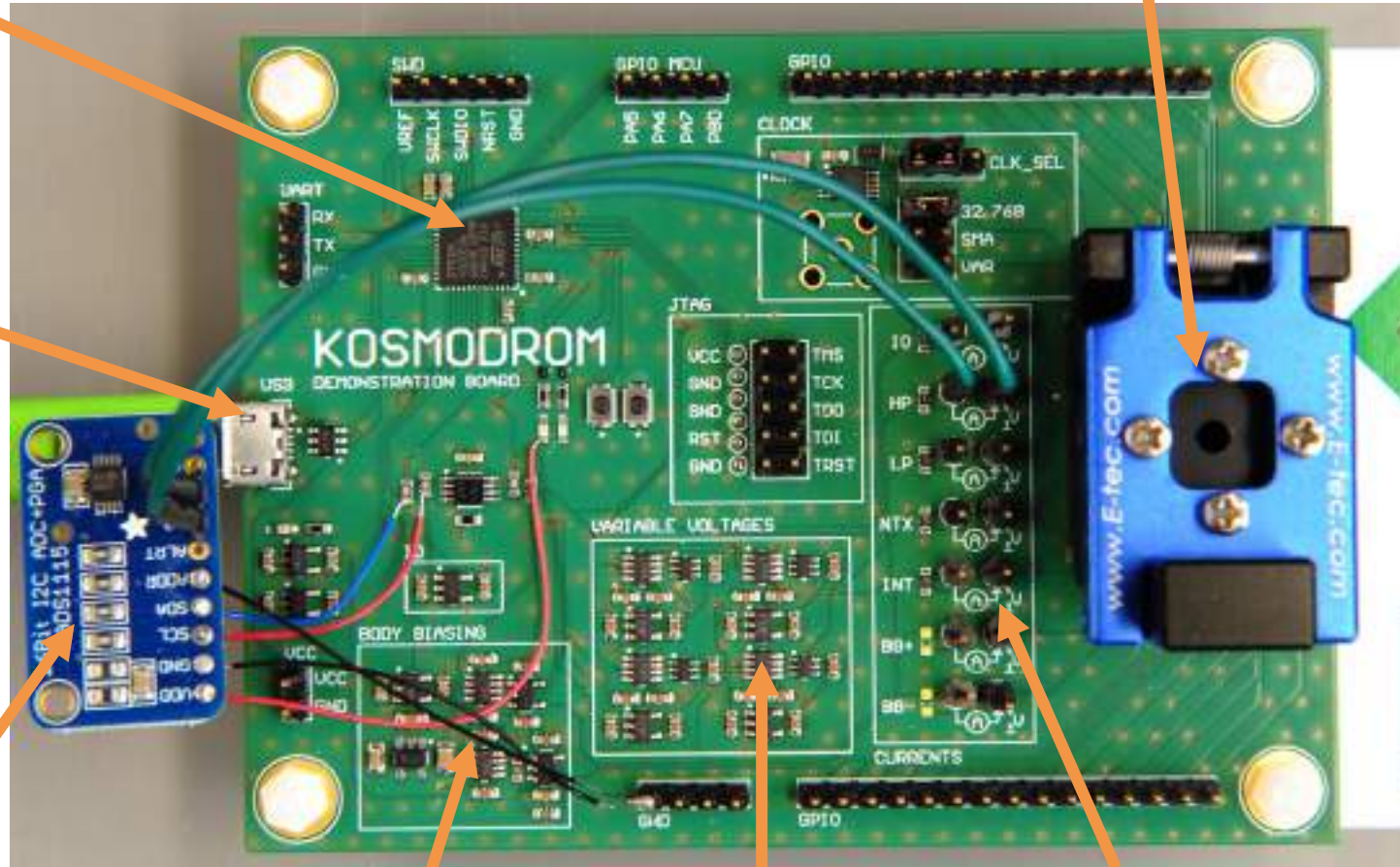
F. Schuiki, M. Schaffner, F. K. Gürkaynak and L. Benini, "A Scalable Near-Memory Architecture for Training Deep Neural Networks on Large In-Memory Datasets," in IEEE Transactions on Computers, vol. 68, no. 4, pp. 484-497, 1 April 2019, doi: 10.1109/TC.2018.2876312.

Kosmodrom Demonstration Board

STM microcontroller for control

Test socket for Kosmodrom chip

USB connection to computer



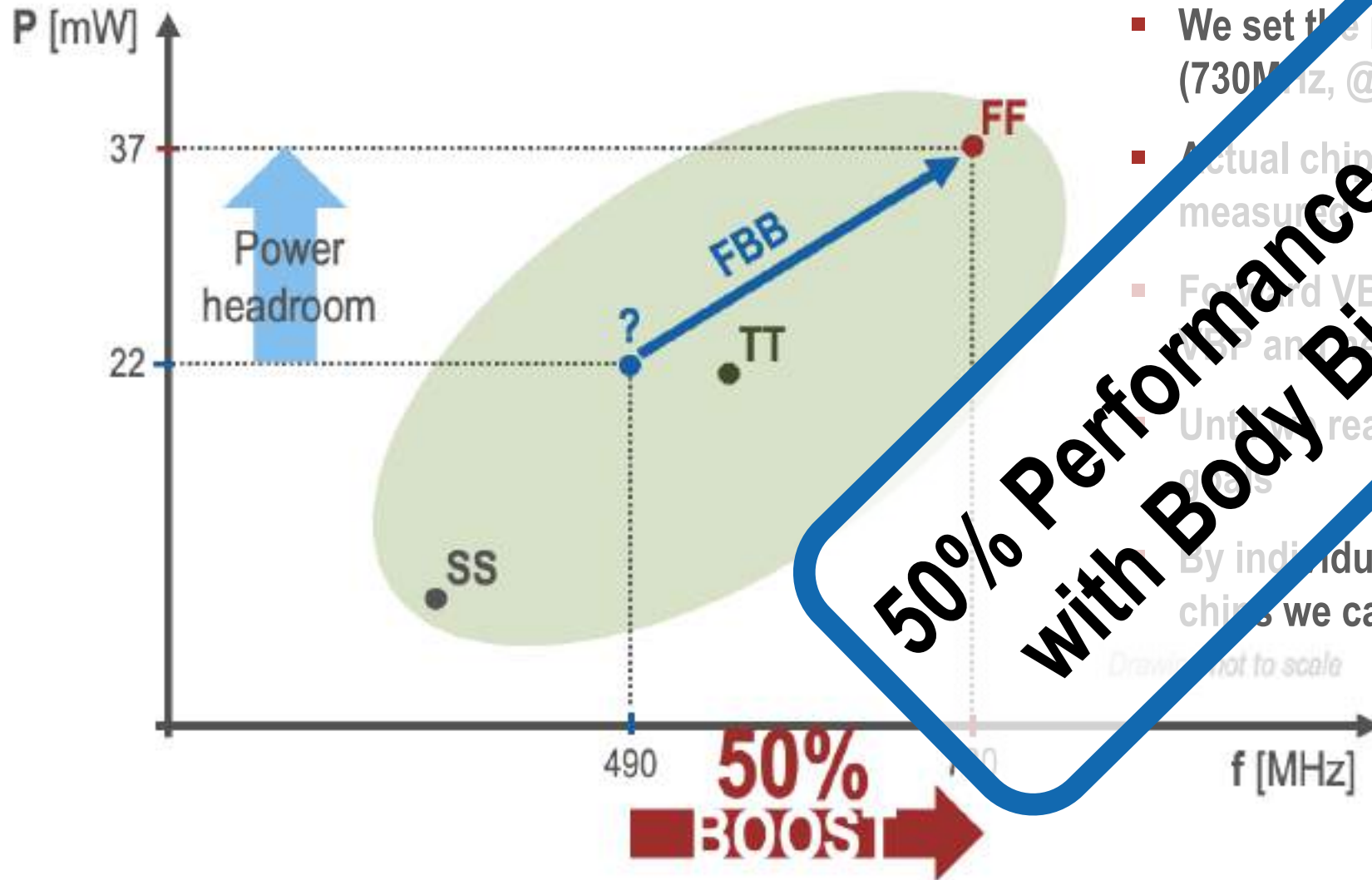
Analog to Digital
Converter module

Body bias voltage
generation

Supply voltage
generation

Measurement points for
all supplies

Boosting performance with Body Biasing



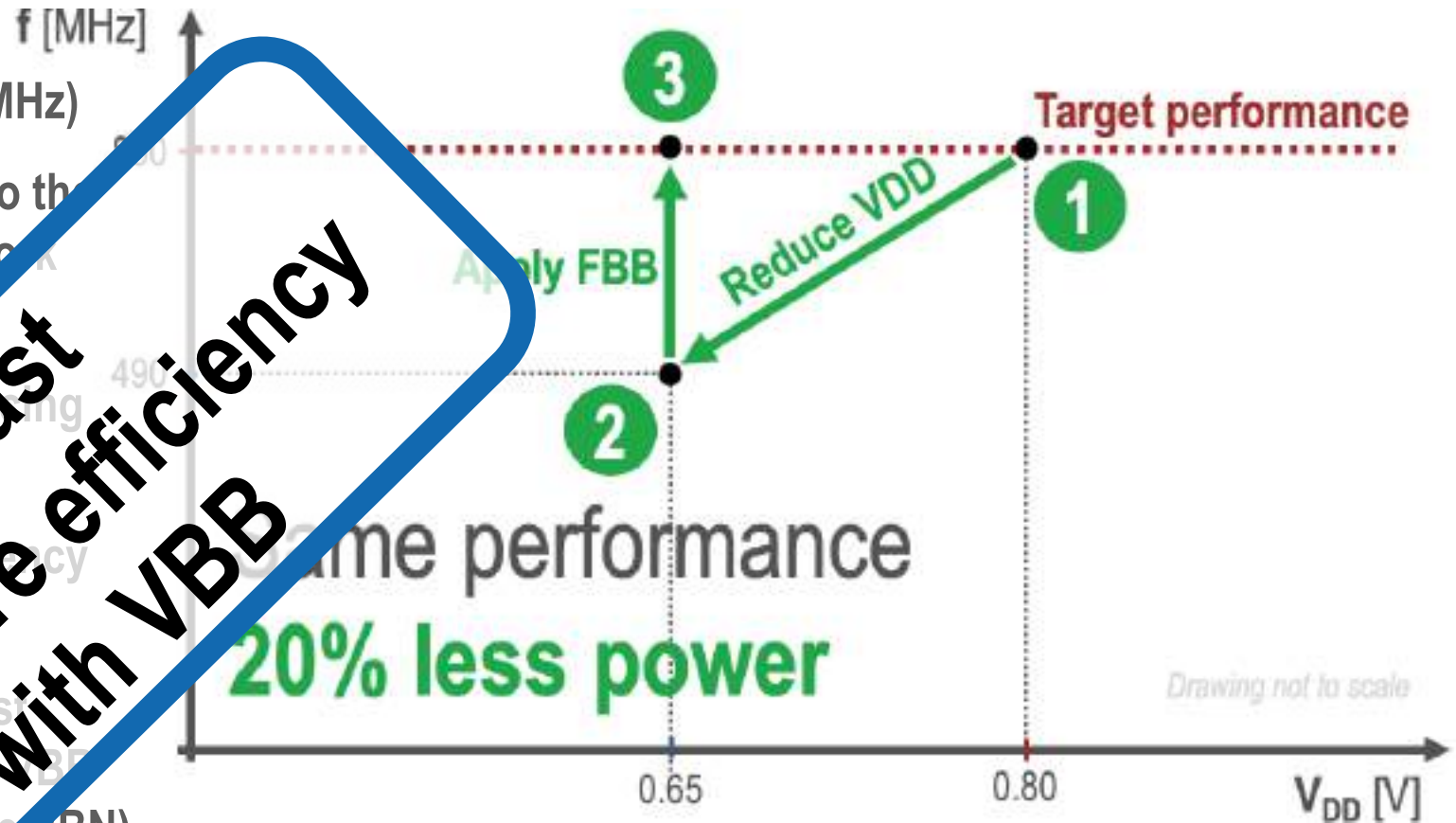
- We set the performance target (730MHz, @0.65V, 40mW)
- Actual chip performance is measured
- Forward VBB is applied (positive V_{FB} and negative V_{BN})
- Unable to reach the performance goals
- By individually applying VBB to chips we can improve yield

50% Performance gain with Body Biasing

Drawn not to scale

Gaining Energy Efficiency with Body Biasing

- We set the desired operating frequency (800MHz)
- We decrease the voltage to the minimum level chip will work (0.8V)
- At this point we start reducing voltage further (0.65V)
- Maximum operating frequency will also drop (~500MHz)
- We compensate for the lost performance with forward BSB (positive V_{BB} and negative V_{BN})
- Until we reach the desired operating frequency.

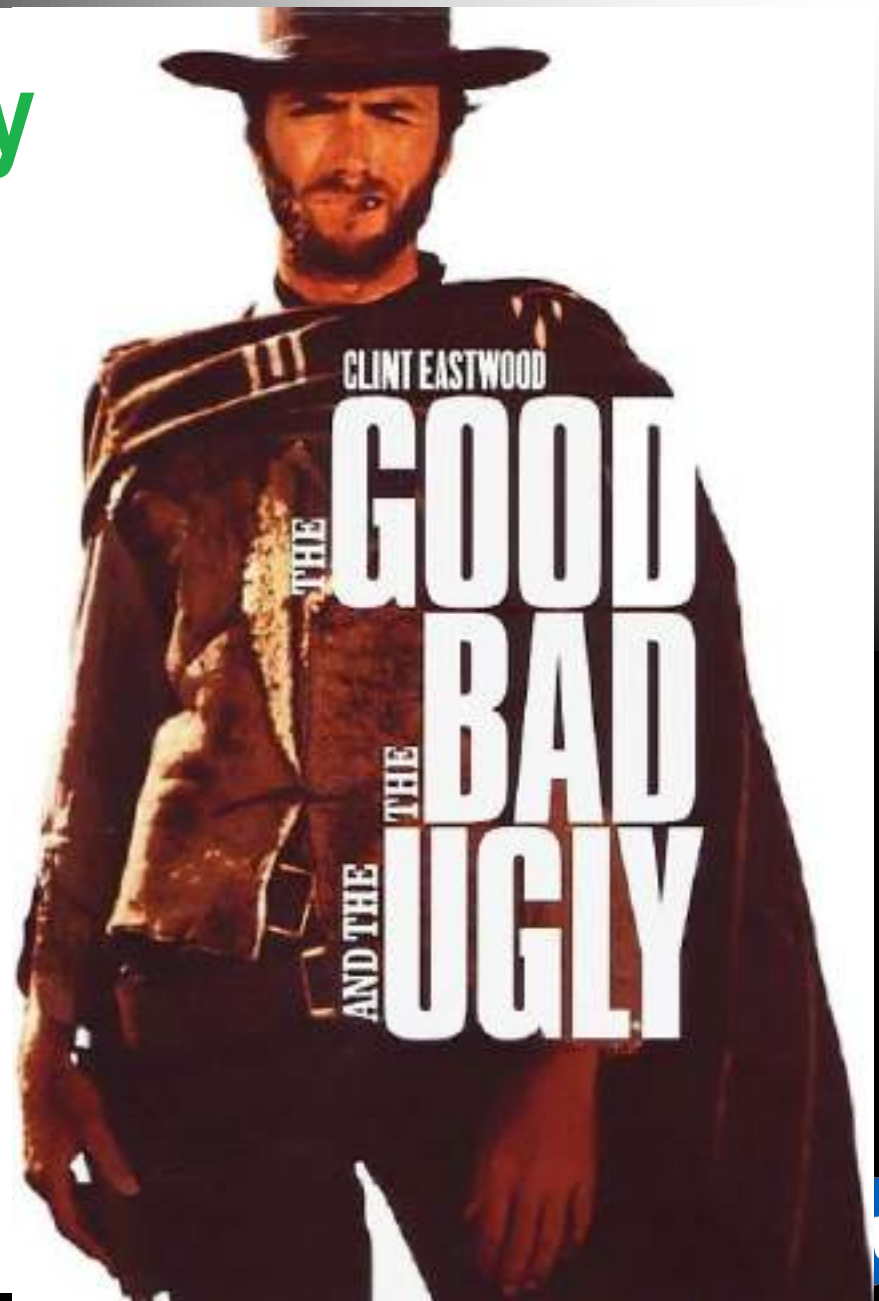


At least 20% more efficiency with VBB




The good the bad and the ugly

- We designed and tested 43 chips as part of PULP project
- Most worked great
- But there were also mistakes
- Here is a look at some highs and some lows





Good: Fulmine the award winning one



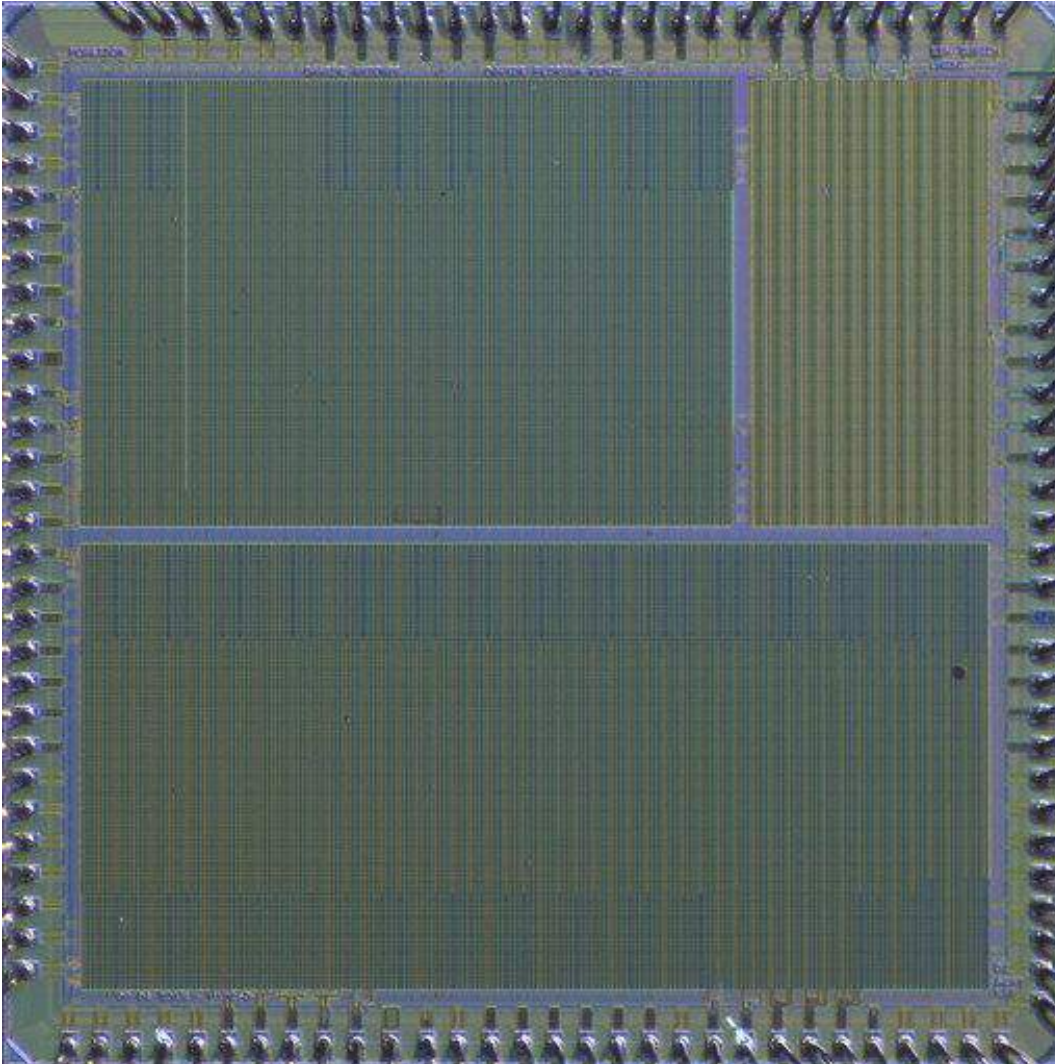
IEEE Circuits and Systems,
Darlington Award for Best
Paper in 2020

- UMC65
- Earlier chip (2015)
 - 4x OpenRISC cores (not yet RISC-V)
 - 192 kBytes L2 + 64 kBytes TCDM
 - 2x HW accelerators
 - HW – Crypt (together with TU-Graz)
 - HW – Convolution Engine
- Publication from this chip

F. Conti et. Al., "An IoT Endpoint System-on-Chip for Secure and Energy-Efficient Near-Sensor Analytics", IEEE Transactions on Circuits and Systems I: Regular Papers, Vol: 64, Issue: 9, Sept. 2017, pp 2481 – 2494"

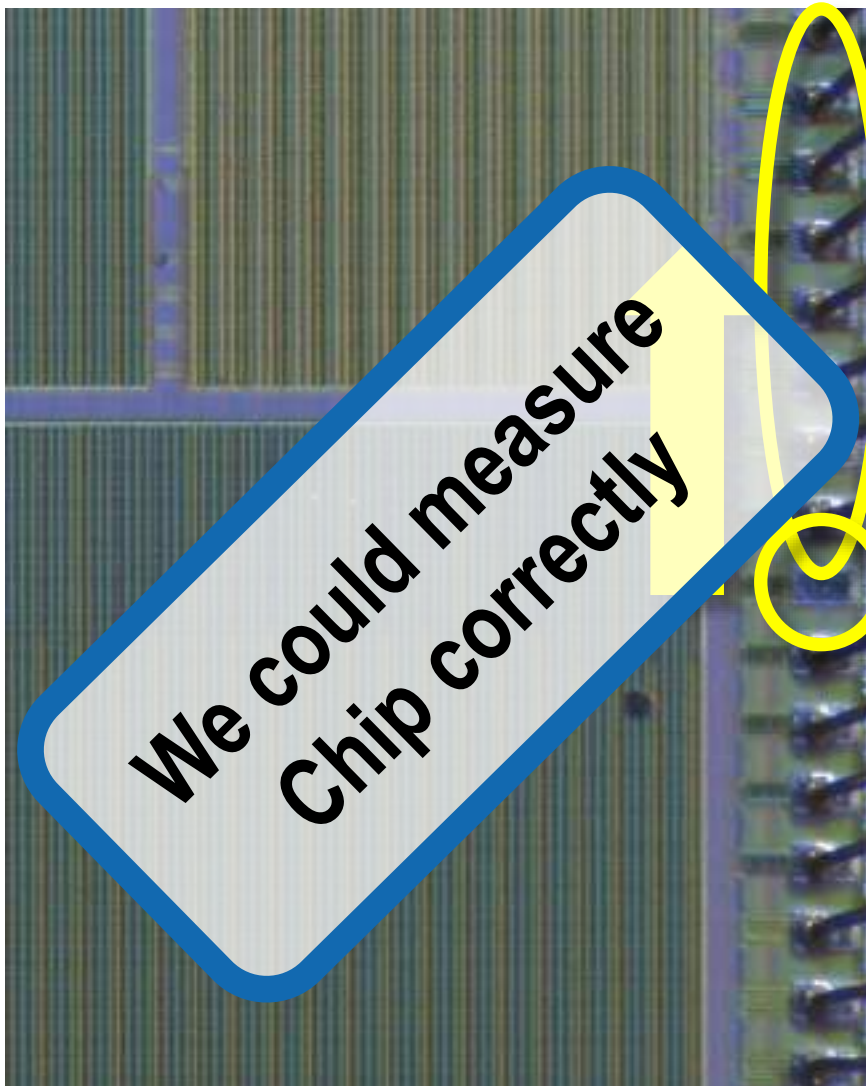


Bad: Bonding issues on Poseidon



- **First GF22nm chip**
 - Used Europractice IC service
 - Cost 150k CHF for 50 samples
- **Has three parts (trident..)**
 - PULPissimo system
 - Ariane core
 - Independent ML accelerator
- **30 of 50 chips were packaged**
 - We provide a bonding diagram
 - Mostly simple manual work

Bad: Bonding issues on Poseidon

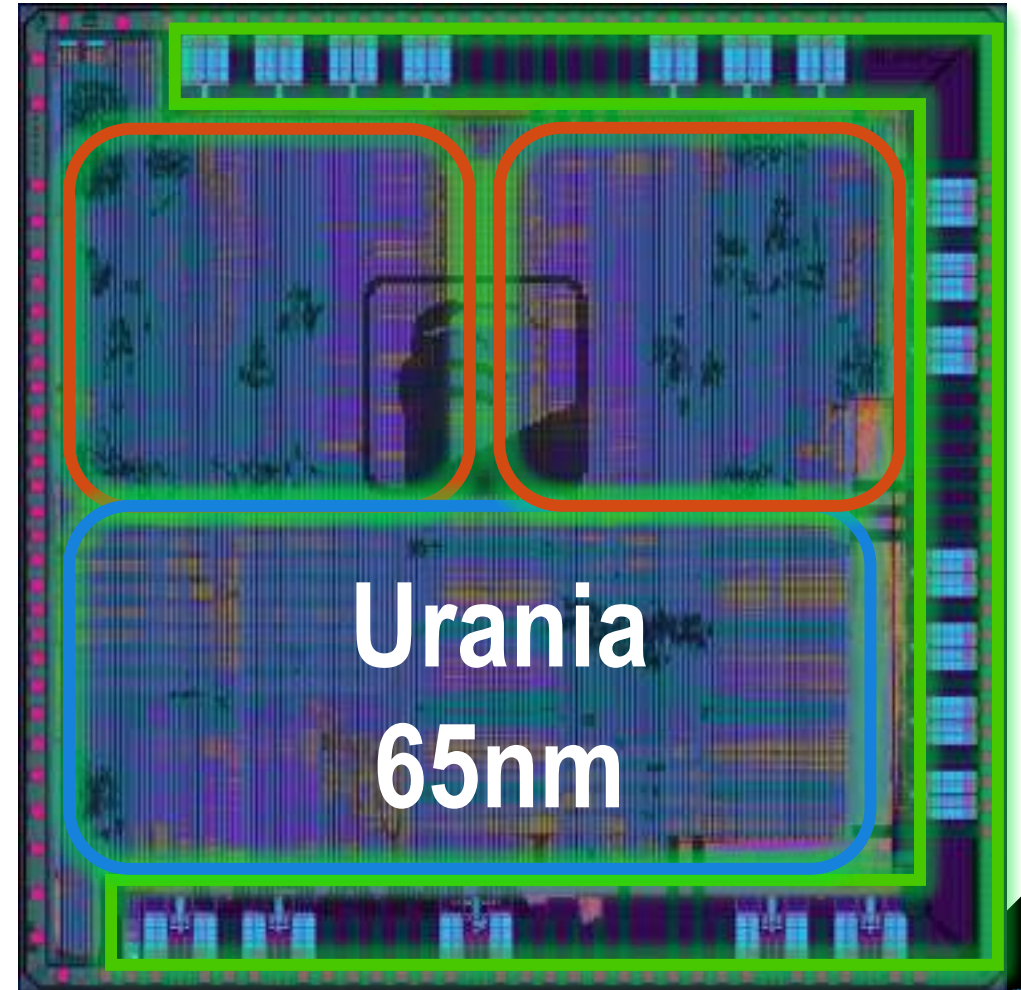


- **Look closer on the right side**
 - There is a pad that is not bonded
- **We skipped one pad**
 - All connections are shifted by one
- **VDD and GND are one after other**
 - Bonding causes shorts between VDD and GND
 - Pretty much catastrophic.
- **Fortunately: unpackaged dies**
 - There were 20 unpackaged dies
 - We could bond those correctly

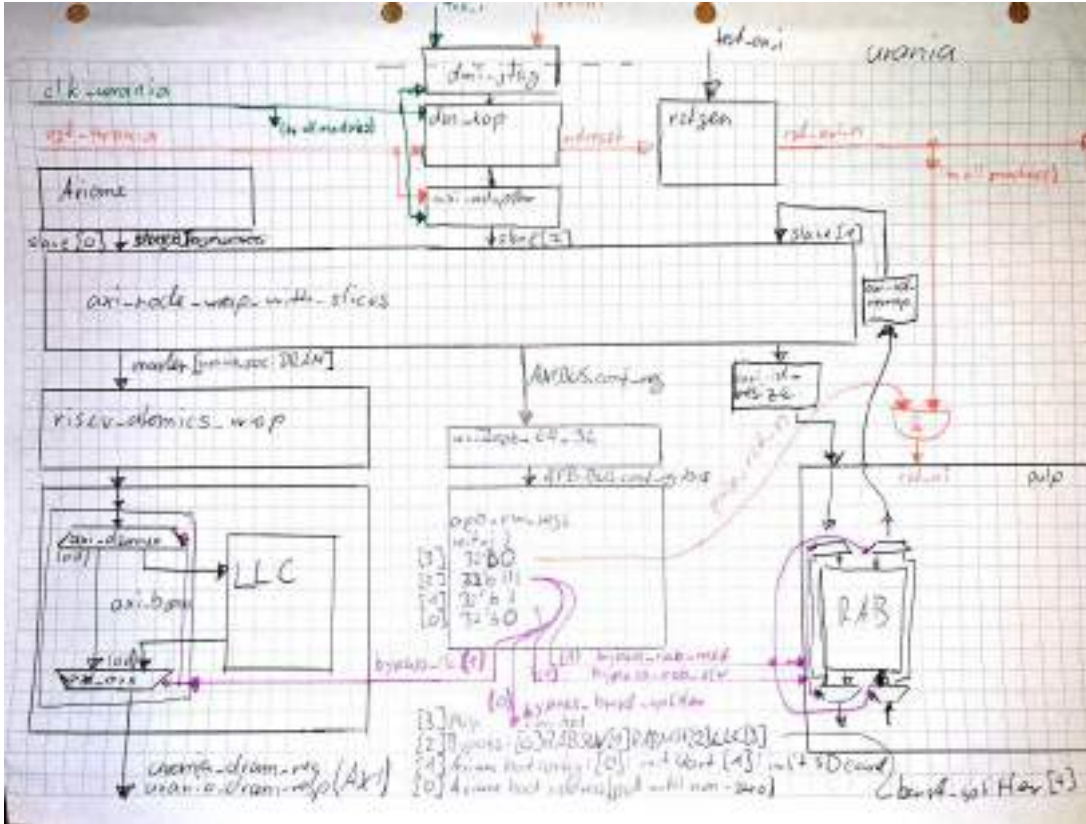


Downright Ugly, reset problem of Urania

- **2 PULP clusters, each with**
 - 4x RV32 RI5CY cores
 - 4x transprecision FPU's
 - 1x PULPO accelerator
 - 64 kB TCDM in 8 banks
- **Ariane RV64 host processor**
 - 128 KiB Shared LLC
 - software-managed IOMMU
- **DDR3 DRAM Controller + PHY by TUKL**



ETH zürich



- 






IC Design is tricky and demands attention

- **Even the simplest things can derail a complex chip**
 - A copy paste error in a bonding diagram, a mistake in reset
- **Academic research chips are not industrial products**
 - Designed to test and verify ideas, not mass production
 - Much more effort needed in DfT and verification to make a successful product
- **Experience is key in IC Design**
 - All the mistakes we make, add to our future success
 - Some lessons you learn the hard way
 - But these stay with you and help you for your future designs



We hope this was helpful/fun for you

- **Covered the basics of RISC-V**
 - Explained the ISA
 - Examples of Implementations
 - Advanced cores and Concepts
- **Talked about building open source systems around RISC-V**
 - Showed the main concepts and talked about our ICs
- **You can find PULP related information**
 - GitHub:  http://github.com/pulp_platform
 - PULP Webpage:  <http://pulp-platform.org>
 - Follow us on Twitter:  [@pulp_platform](https://twitter.com/pulp_platform)



PULP

Parallel Ultra Low Power

Luca Benini, Davide Rossi, Andrea Borghesi, Michele Magno, Simone Benatti, Francesco Conti, Francesco Beneventi, Daniele Palossi, Giuseppe Tagliavini, Antonio Pullini, Germain Haugou, Manuele Rusci, Florian Glaser, Fabio Montagna, Bjoern Forsberg, Pasquale Davide Schiavone, Alfio Di Mauro, Victor Javier Kartsch Morinigo, Tommaso Polonelli, Fabian Schuiki, Stefan Mach, Andreas Kurth, Florian Zaruba, Manuel Eggimann, Philipp Mayer, Marco Guermandi, Xiaying Wang, Michael Hersche, Robert Balas, Antonio Mastrandrea, Matheus Cavalcante, Angelo Garofalo, Alessio Burrello, Gianna Paulin, Georg Rutishauser, Andrea Cossettini, Luca Bertaccini, Maxim Mattheeuws, Samuel Riedel, Sergei Vostrikov, Vlad Niculescu, Hanna Mueller, Matteo Perotti, Nils Wistoff, Luca Bertaccini, Thorir Ingulfsson, Thomas Benz, Paul Scheffler, Alessio Burello, Moritz Scherer, Matteo Spallanzani, Andrea Bartolini, Frank K. Gurkaynak, and many more that we forgot to mention



<http://pulp-platform.org>



@pulp_platform