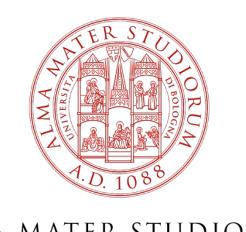
# ETHZURICH



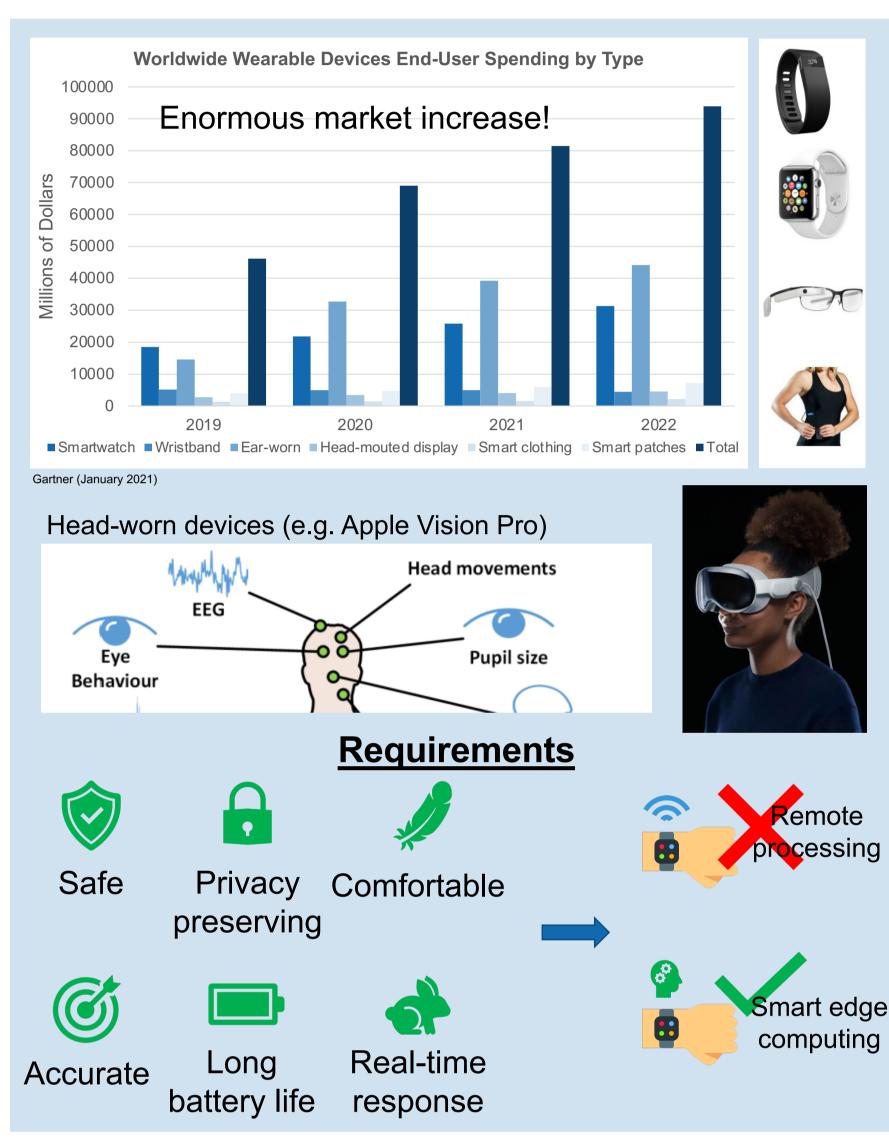
Near-Sensor Analytics and Machine Learning for Long-Term Wearable Biomedical Systems

X. Wang<sup>1</sup>, T. M. Ingolfsson<sup>1</sup>, L. Mei<sup>1</sup>, C. Cioflan<sup>1</sup>, S. Frey<sup>1</sup>, V. Kartsch<sup>2</sup>, S. Benatti<sup>2</sup>, A. Cossettini<sup>1</sup>, L. Benini<sup>1,2</sup> <sup>1</sup> Integrated Systems Laboratory, ETH Zurich, Switzerland <sup>2</sup> DEI, University of Bologna, Bologna, Italy

**BioWolf and BioGAP ExG Acquisition** 

# **Smart Wearable Devices**

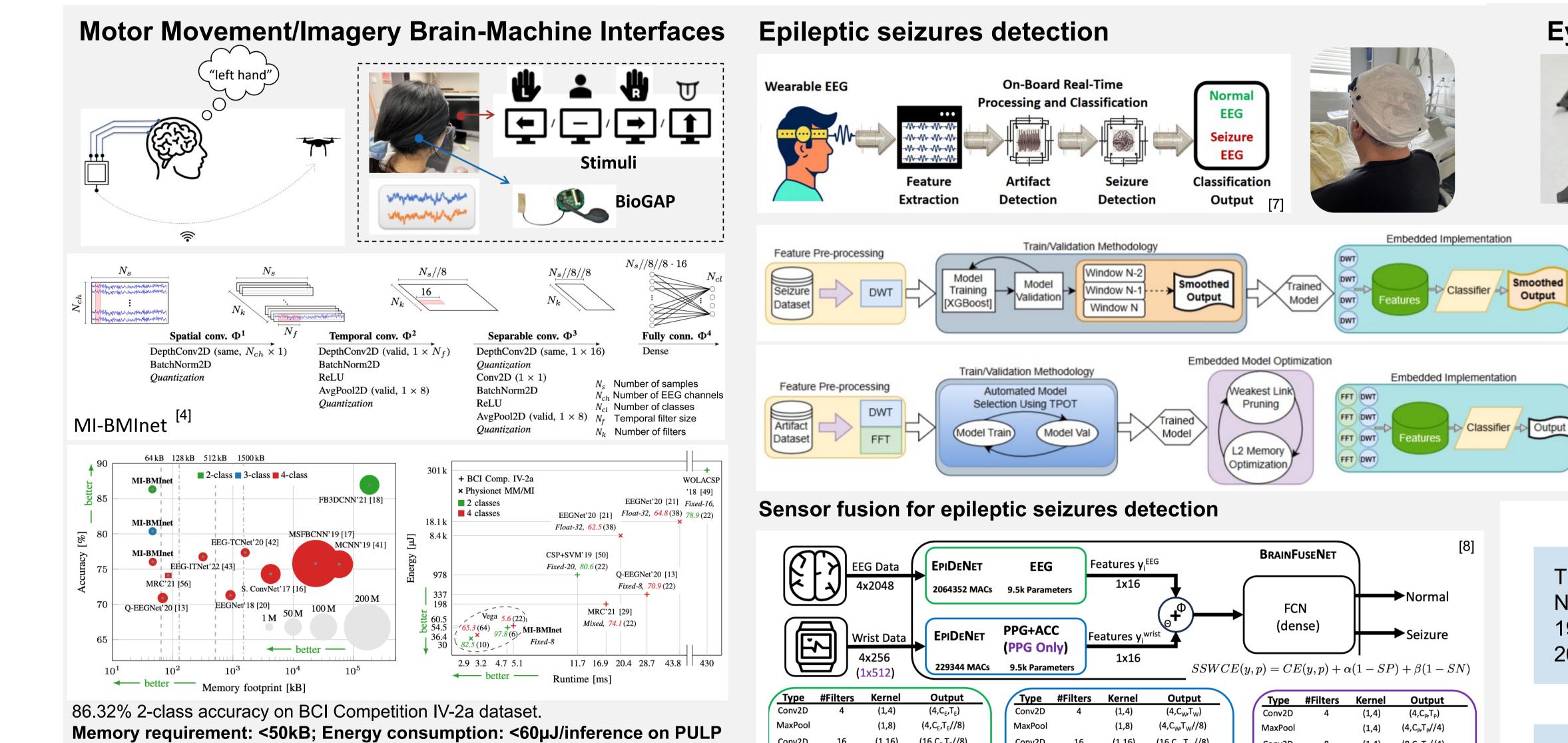
# **Proposed Systems**



### **PCB** implementation EEG PCB **Baseboard ADS1298** 12C Up to 8 ExG channels ADS1298 • 24-bit of resolution LSM6DSO EEG AFE GAP9 0.5 to 32 – Ksps. SoC IMU (-----i) SPI • PGA up to 12x. 12C • Dry active/Gel electrodes Electr nRF52811 MAX20303 compatible. **Usage example (Patch-like encase)** BLE SoC PMIC MAX86150 PPG AFE nRF52832 ARM Cortex-M4. **PPG PCB** CR1254 64 MHz. Host PC OFAAR Bluetooth 5 capable. Battery nRF52840 NFC-A. Dongle • 512/64 KB Flash/RAM. [3] In-ear PULP SoC \_\_\_\_\_ DC/DC L2 BANK O L2 BANK 1 e.g. Mr. Wolf: LDO Flex-rigid PCB to • 8 RISC-V cluster processors. L2 BANK 2 L2 BANK Custom **EOG** interface Voltage range (0.8V-1.1V). + FLLs SoC Domain and EEG electrodes • 40nm LP CMOS. electrodes L2 BANK 4 64kB/512kB L1/L2 memory. **BioGAP** for acquisition and L2 BANK 6 L2 BANK Peak Perf → 1GFLOP/s process En.Eff. → 15MMAC/s/mW and 9MFMAC/s/mW **Cluster Domain** Headband **ExG-glasses** Miniaturized and non-**Developed Applications** stigmatizing!

Smoothed

Output



Conv2D

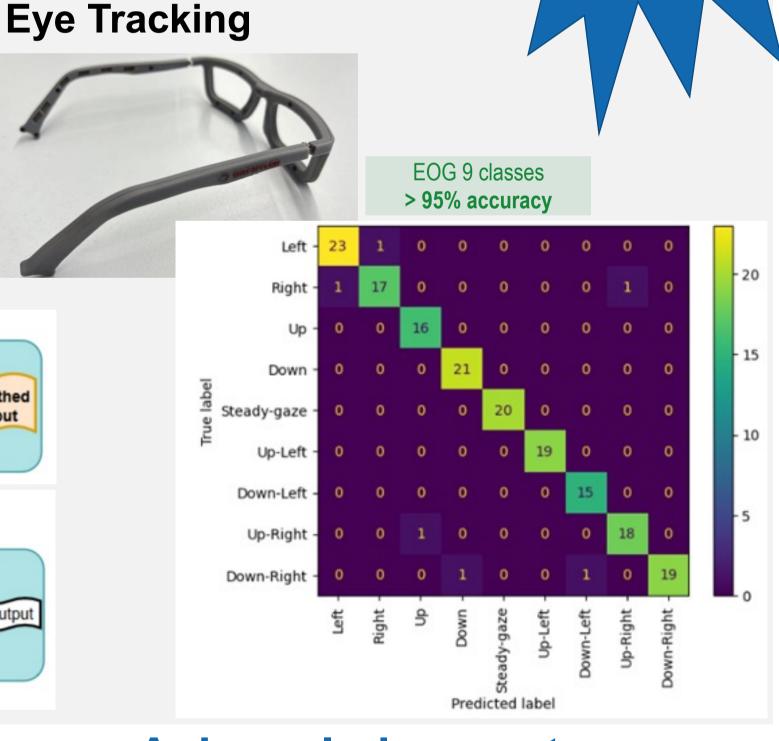
MaxPoo

Conv2D

MaxPoo

Conv2D

MaxPoo

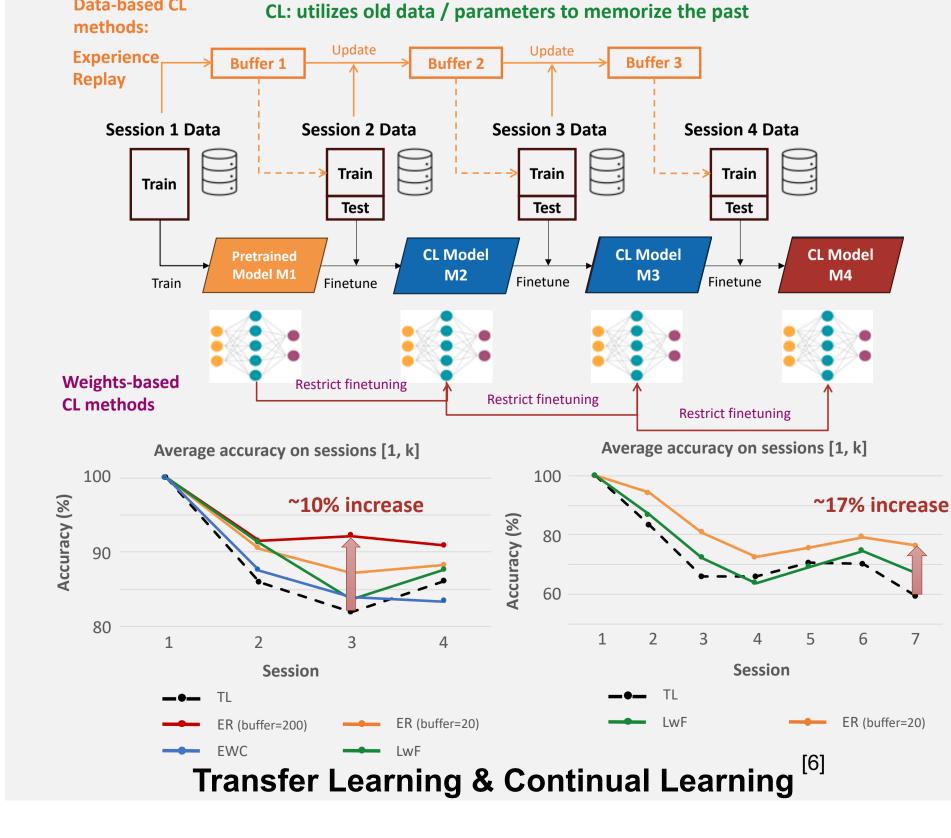


# **Acknowledgements**

The authors acknowledge support from the Swiss National Science Foundation under the grant n. 193813 (PEDESITE project) and the grant n. 207913 (TinyTrainer project).

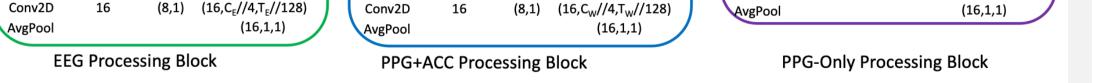
# References

- 1. Kartsch V, et al., "BioWolf: A Sub-10-mW 8-Channel Advanced Brain-Computer Interface Platform With a Nine-Core Processor and BLE Connectivity," IEEE Trans Biomed Circuits Syst., 2019.
- 2. X. Wang, et al., "Enhancing Performance, Calibration Time and



Vega; Runtime: <5ms/inference.

Data-based CL



(1,4)

(1,8)

(1,4)

(16,1)

(4,1)

(16,C<sub>w</sub>,T<sub>w</sub>//8)

(16,C<sub>w</sub>,T<sub>w</sub>//32)

(16,C<sub>w</sub>,T<sub>w</sub>//32)

(16,C<sub>w</sub>,T<sub>w</sub>//128)

(16,C<sub>w</sub>,T<sub>w</sub>//128)

(16,C<sub>w</sub>//4,T<sub>w</sub>//128)

Conv2D

MaxPool

Conv2D

MaxPoo

Conv2D

Conv2D

(8,C<sub>P</sub>,T<sub>P</sub>//4)

 $(16, C_P, T_P//16)$ 

 $(16, C_{p}, T_{p}//16)$ 

(16,C<sub>P</sub>,T<sub>P</sub>//64)

(16,C<sub>P</sub>,T<sub>P</sub>//64)

(16,C<sub>P</sub>,T<sub>P</sub>//64)

(1,4)

(1,4)

(1,8)

(1,16)

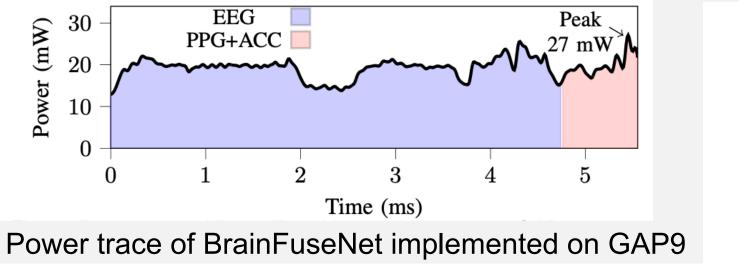
**Demo videos** 

https://efcl.ethz.ch/Media/p

roject-videos---demos.html

## Seizure detection results on the PEDESITE dataset

	EEG CE			EEG (smooth) SSWCE			EEG+PPG (smooth) SSWCE			EEG+PPG+ACC (smooth) SSWCE		
Patient	<b>Sens</b> [%]	<b>Spec</b> [%]	FP/h	<b>Sens</b> [%]	<b>Spec</b> [%]	FP/h	<b>Sens</b> [%]	<b>Spec</b> [%]	FP/h	<b>Sens</b> [%]	<b>Spec</b> [%]	FP/h
P1	9.70	99.52	2.16	25.20	99.62	1.71	22.90	99.76	1.08	-	-	-
P2	57.00	100	0	59.44	100	0	59.10	100	0	46.31	99.98	0.11
P3	68.20	99.48	2.34	65.45	99.67	1.48	69.95	99.56	1.98	60.07	99.95	0.24
P4	73.40	99.72	1.26	77.86	99.78	0.99	82.29	99.82	0.81	77.95	99.85	0.68
P5	91.10	99.82	0.81	92.50	100	0	91.24	100	0	91.06	100	0
P6	48.70	96.90	14.0	43.49	99.36	2.88	41.81	99.50	2.15	46.03	100	0
Average	58.02	99.24	3.43	60.66	99.74	1.18	61.22	99.77	1.00	64.28	99.96	0.21



 $(16, C_{E}, T_{E}//8)$ 

 $(16, C_E, T_E//32)$ 

 $(16, C_E, T_E//32)$ 

 $(16, C_E, T_E / / 128)$ 

 $(16, C_E, T_E / / 128)$ 

(16,C<sub>E</sub>//4,T<sub>E</sub>//128)

(4,1)

Conv2D

MaxPool

Conv2D

MaxPool

Conv2D

MaxPool

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