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*Enabling Ultra-low Power Machine Learning at the Edge*

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# An embedded EOG-based BCI system for robotic control

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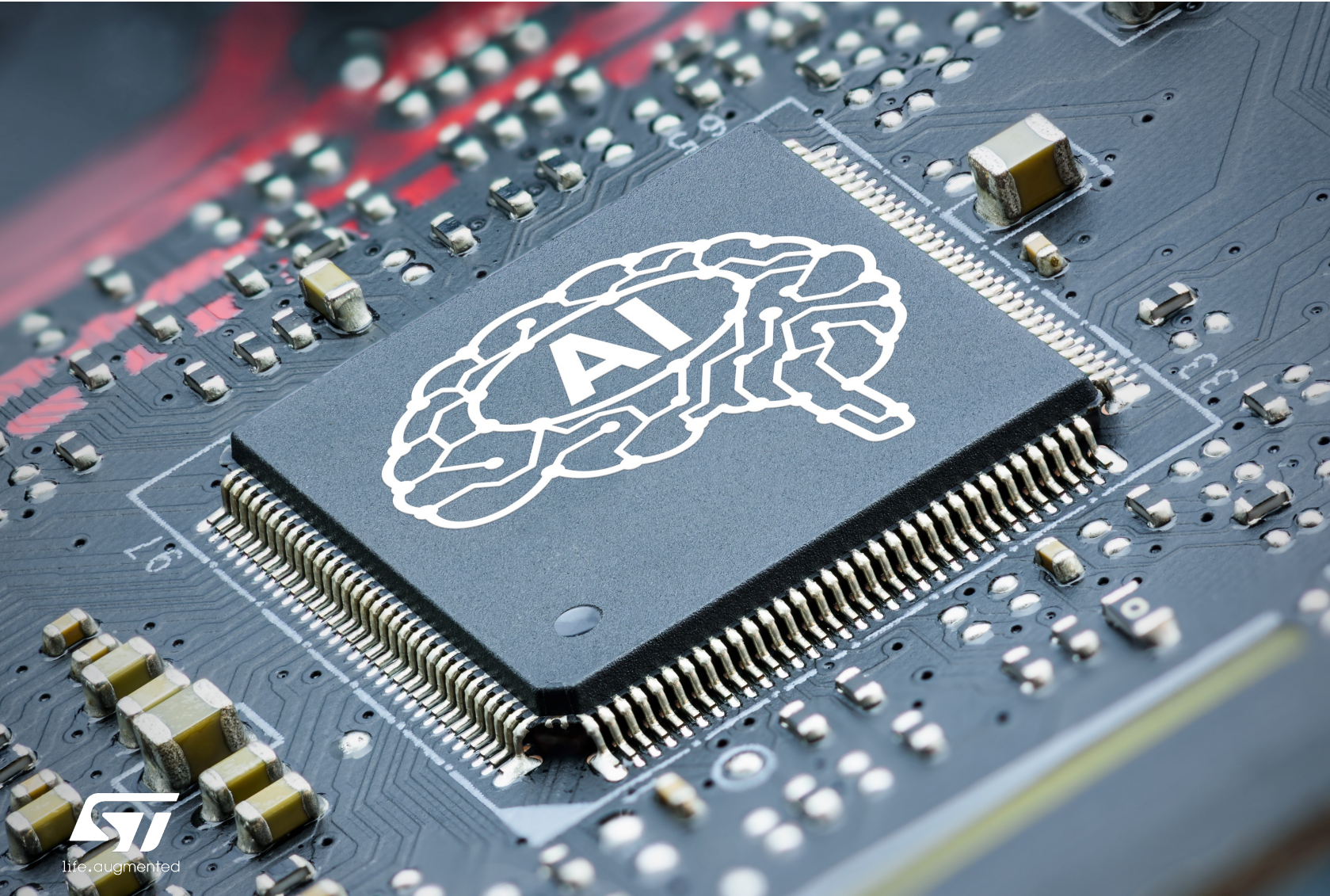




**The brain is wider than the sky**  
*Emily Dickinson*

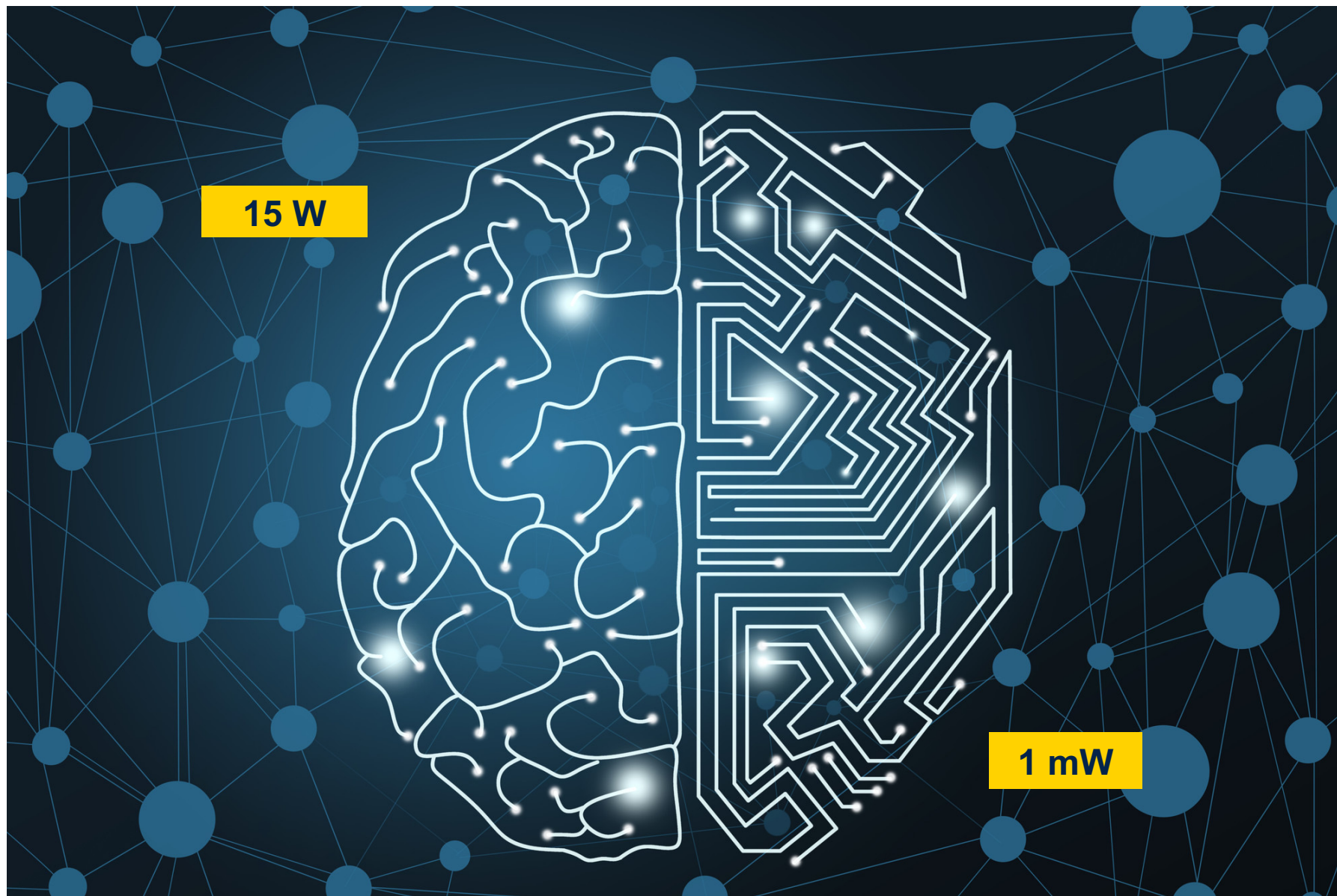


# The future of machine learning is tiny





# Power Consumption





# Brain Computer Interface (BCI)



Allows to establish a direct communication link between the human brain and an external device

It was mainly conceived to assist people with severe motor disabilities

It is based on grabbing and decoding brain waves

It is considered a new frontier of Human Machine Interaction (HMI)



# State of the art

Li et al. [1] proposed a single blink EOG-based switch for the “start/stop” control of a wheelchair

Djeha et al. [2] investigated a combination of EEG and EOG eye movements to allow the control of a wheelchair in a virtual environment. NN where used, with 93% accuracy

O’Brien et al. [3] took a multimodal approach using EOG and EMG data from the eyes to move Forward, Left, Right and Stop a wheelchair. Multiple Classification Ripple Down Rules model used, with 97.7% accuracy

Sharma et al. [4] detected eye closing/opening from EOG to build an interface for robotic arm control for a pick and place task with the average detection accuracy was 96.9%

[1] Y. Li, S. He, Q. Huang, Z. Gu, and Zhu Liang Yu. A eog-based switch and its application for “start/stop” control of a wheelchair. *Neurocomputing*, 275:1350–1357, 2018.

[2] M. Djeha, F. Sbagoud, M. Guiatni, K. Fellah, and N.Ababou. A combined eeg and eog signals based wheelchair control in virtual environment. In 2017 5th International Conference on Electrical Engineering-Boumerdes (ICEE-B), pages 1–6. IEEE, 2017.

[3] S. O’Brien and G. Alici. Control of an electric wheelchair using multimodal biosignals and machine learning. In 2021 IEEE/ASME International Conference on Advanced Intelligent Mechatronics (AIM), pages 994–999. IEEE, 2021.

[4] K. Sharma, N. Jain, and P. K Pal. Detection of eye closing/opening from eog and its application in robotic arm control. *Biocybernetics and Biomedical Engineering*, 40(1):173–186, 2020.



# Challenges of BCI systems

A photograph of a person rappelling down a steep, rocky cliff face. The person is shirtless and wearing green pants, suspended by ropes. The background shows a vast blue ocean and a coastline with buildings under a clear sky.

Few control dimensions

Low classification accuracy

Need to execute commands synchronously with an external stimulus

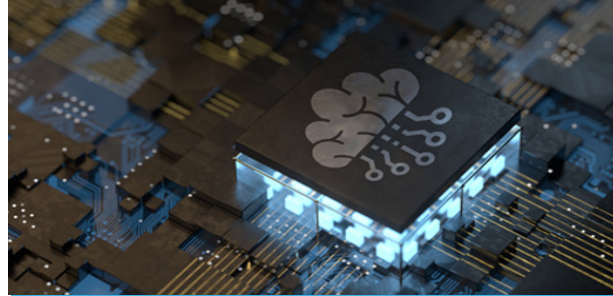
Subjects need an extensive training to be able to control the system



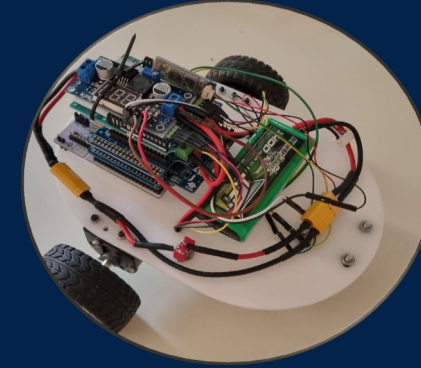
# This work: brain computer interface for robotic control



READ SIGNALS FROM  
THE BRAIN

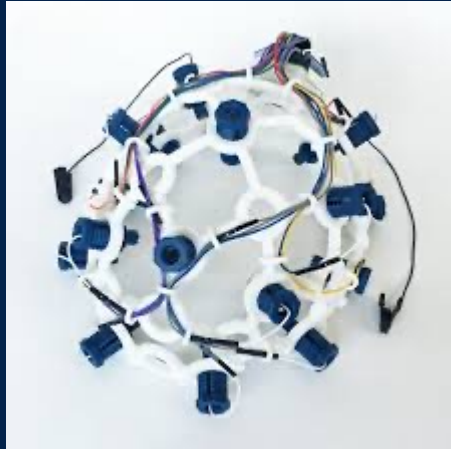


SIGNAL PROCESSING  
AND MACHINE  
LEARNING

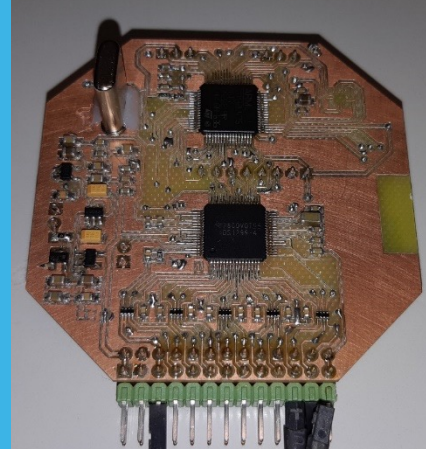


SEND COMMANDS TO A  
ROBOT

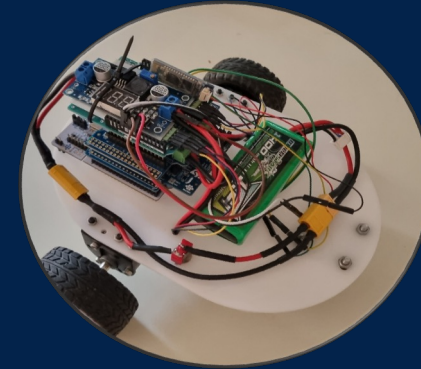
# The hardware



HELMET FROM THE  
OPEN BCI PROJECT



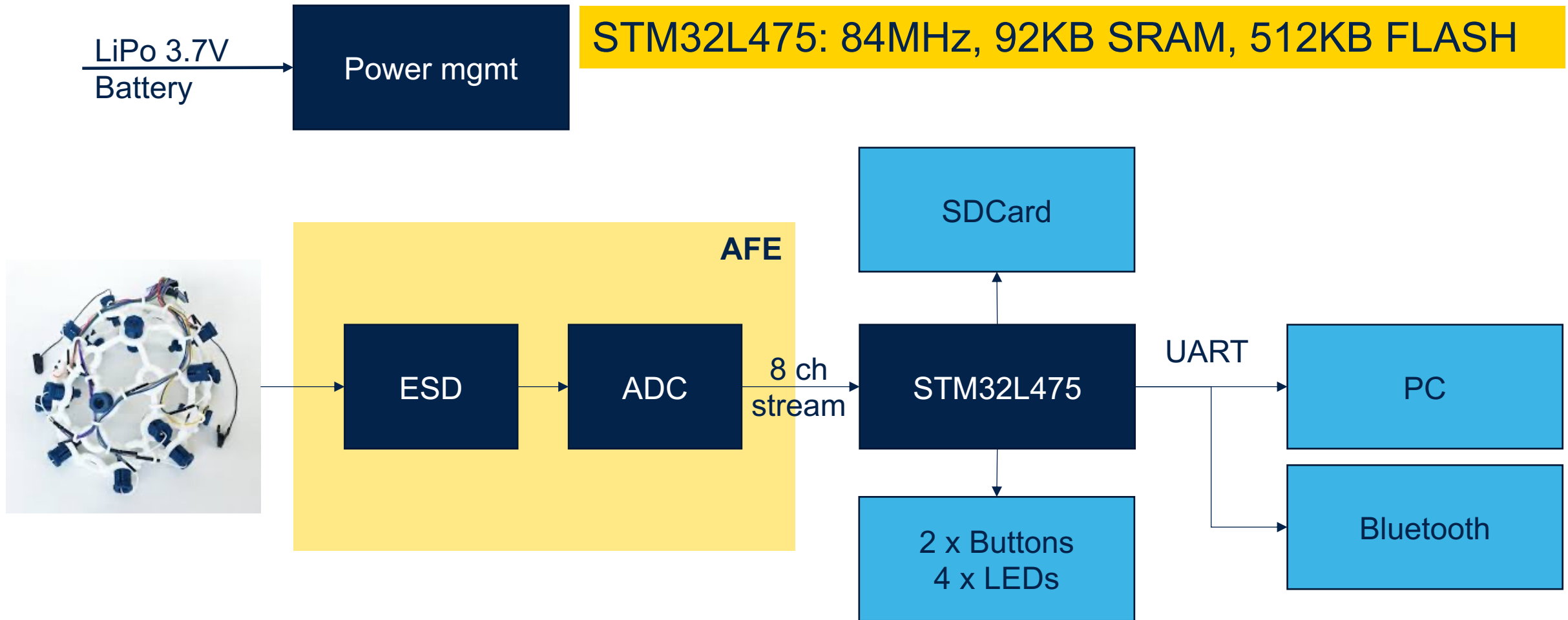
CUSTOM BOARD WITH  
AN STM32  
MICROCONTROLLER



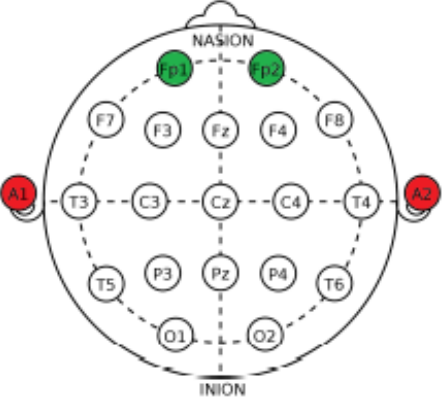
WILL-E: 3 WHEELED  
ROBOTIC PLATFORM



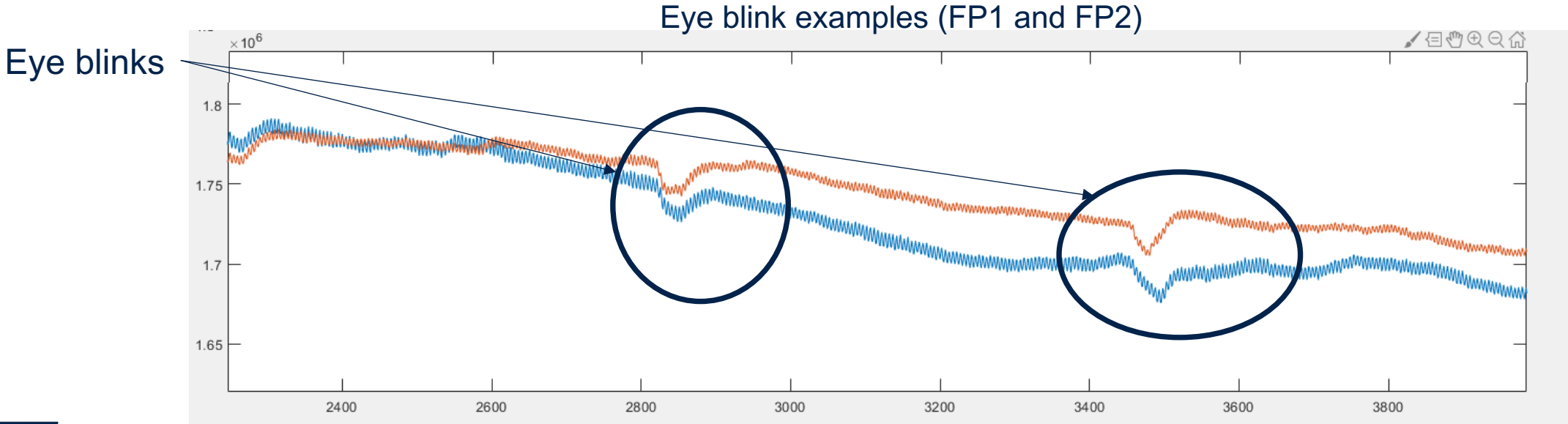
# Custom board with STM32 microcontroller



# Machine learning on Electro-Oculography (EOG) signals



Goal: recognize voluntary/involuntary blinks and right/left winks from fronto polar sensors (Fp1 and Fp2)





# Data collection

8 healthy adults without neurological disorders and without previous experience with BCI



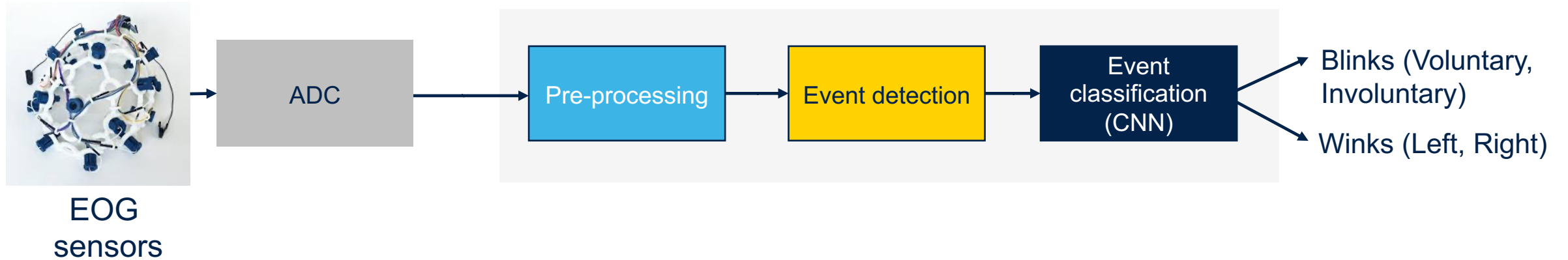
Visual cue was presented on the screen indicating:

- LW (left wink)
- RW (right wink)
- VB (voluntary blink)

Note: IB (Involuntary blink) was spontaneous

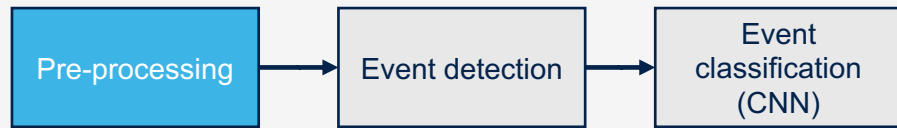
Balanced dataset with 2600 samples has been collected and labelled

# Firmware pipeline



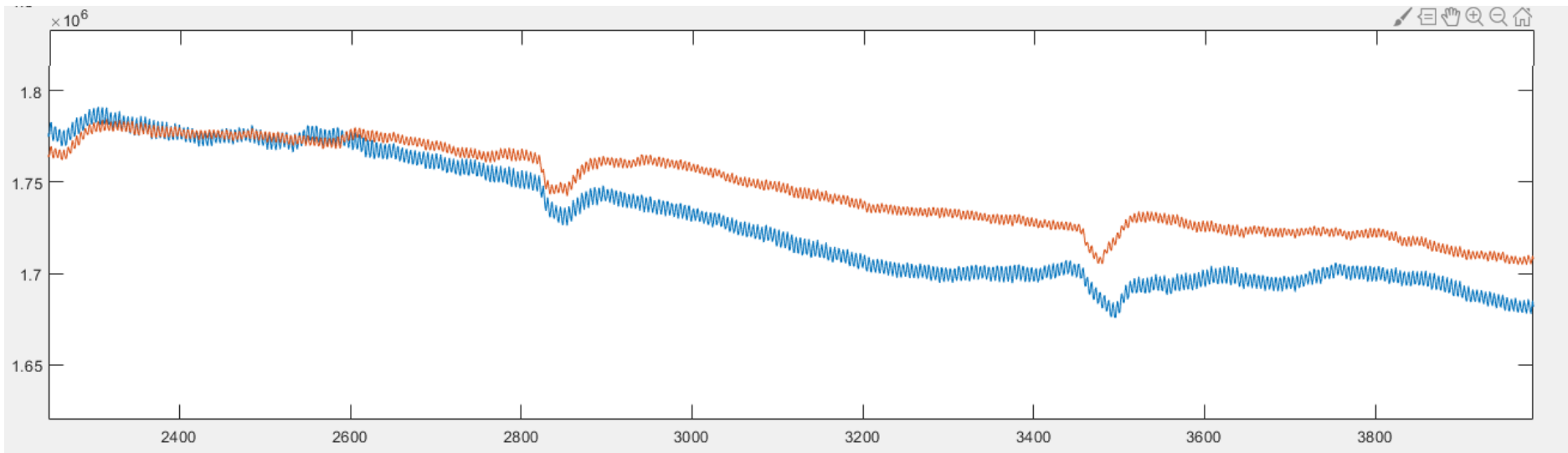


# Pre-processing

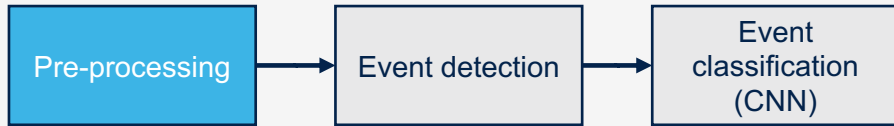


Aims at:

- Eliminating the DC component
- Reducing noise (50Hz/60Hz)

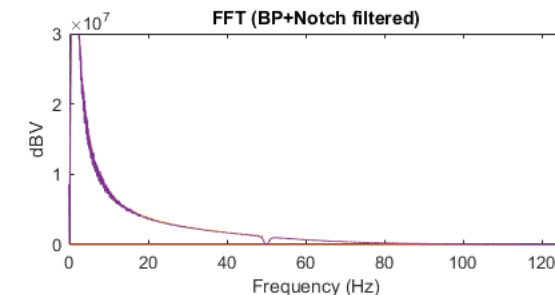
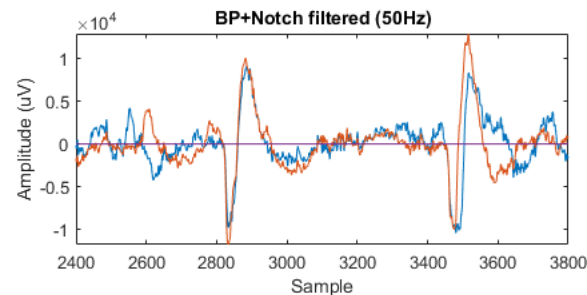
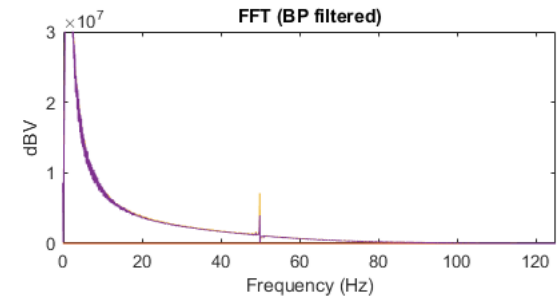
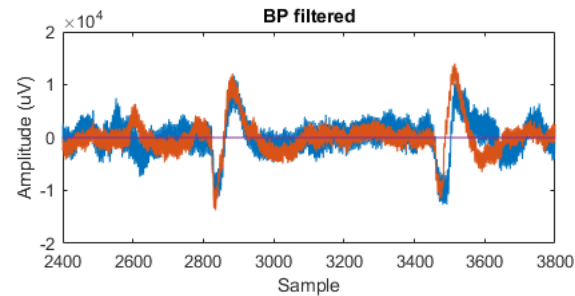
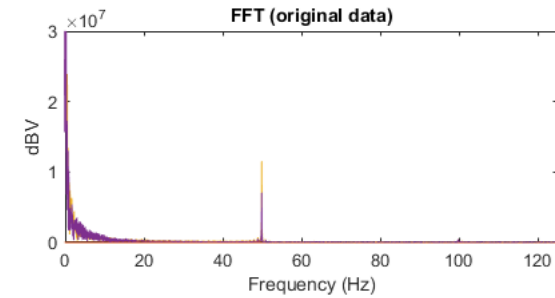
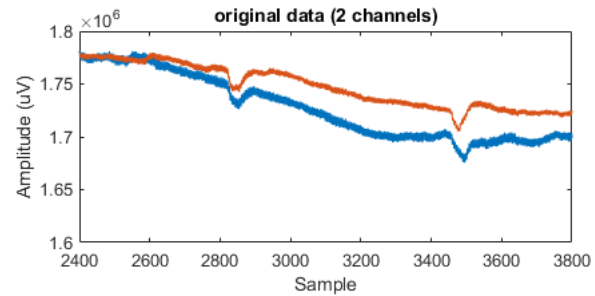


# Pre-processing example

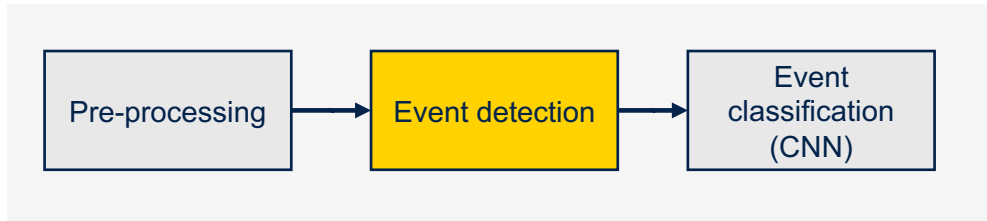


Real-time IIR band-pass filter  
Cut-off frequencies:  
 $f_1 = 0.1\text{ Hz}$   
 $f_2 = 50\text{ Hz}$

Real time IIR notch filter  
 $f = 50\text{ Hz}$





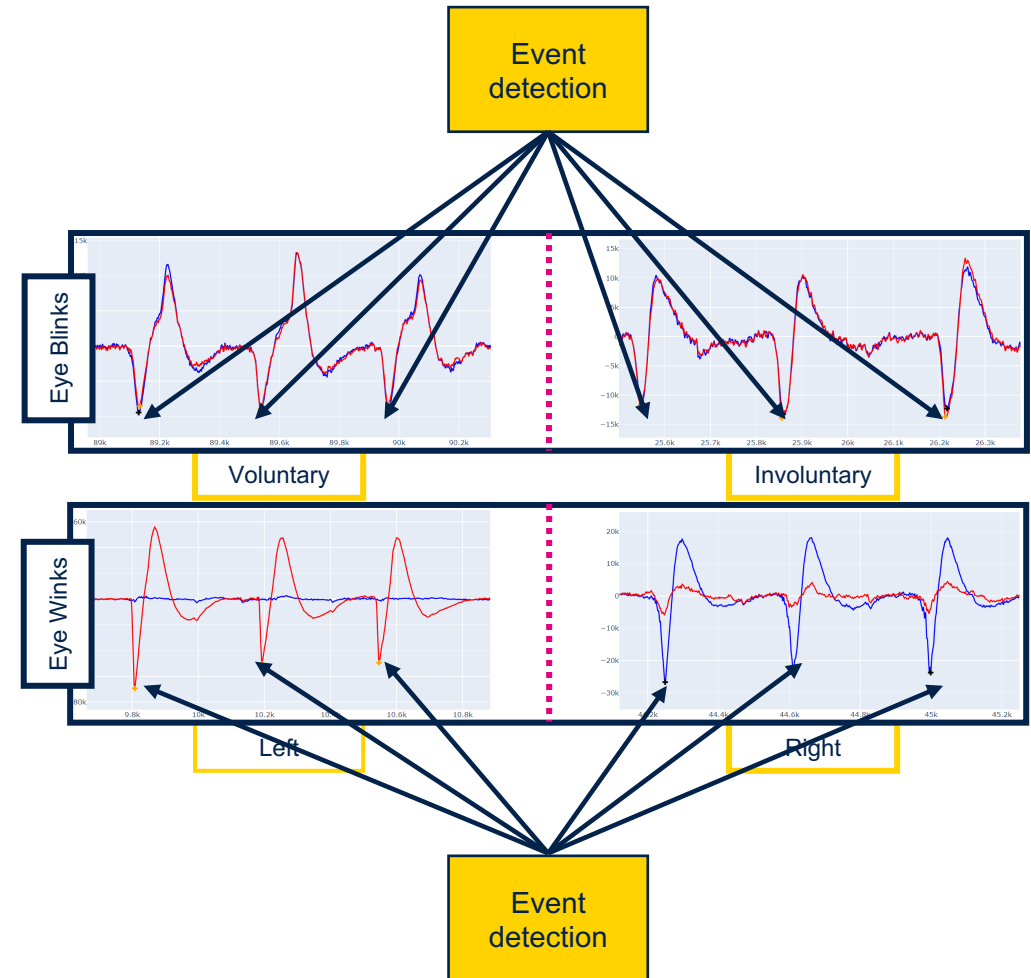


# Event detection

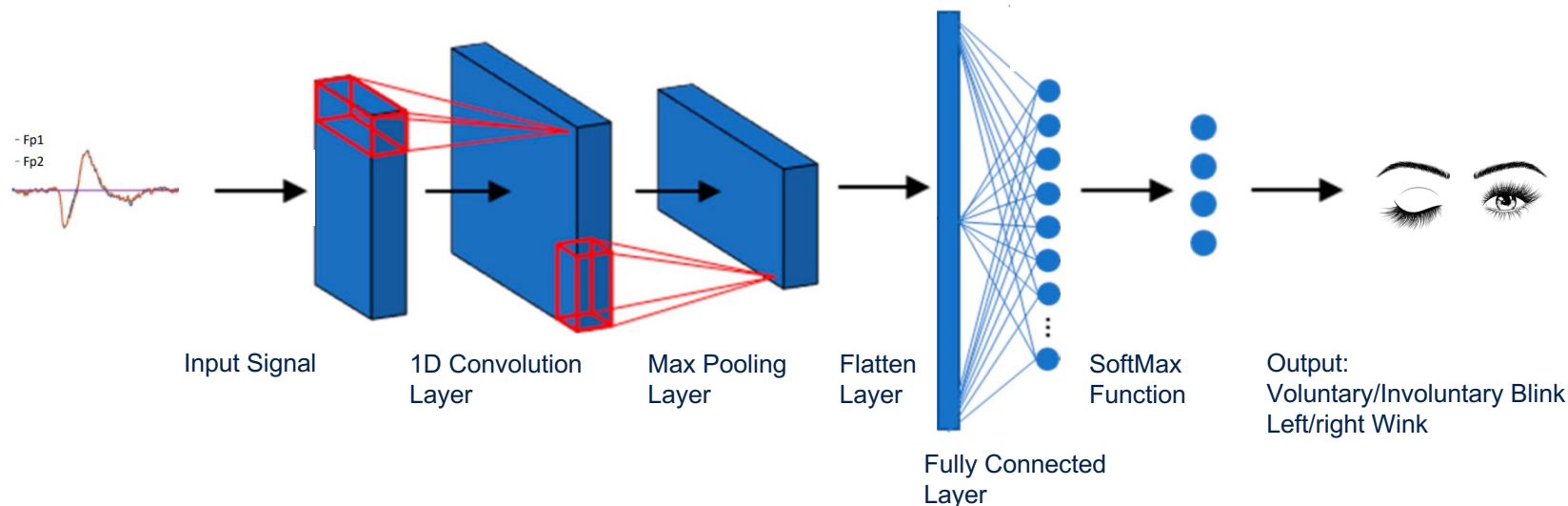
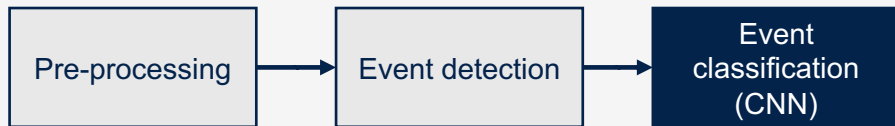
Aims to detect negative peaks

Algorithm is based on z-score

It triggers the collection of 200 samples



# Event classification (CNN)



Model: "sequential"

Layer (type)	Output Shape	Param #
conv1d (Conv1D)	(None, 191, 60)	1260
max_pooling1d (MaxPooling1D)	(None, 9, 60)	0
flatten (Flatten)	(None, 540)	0
dense (Dense)	(None, 20)	10820
dense_1 (Dense)	(None, 4)	84

Total params: 12,164  
 Trainable params: 12,164  
 Non-trainable params: 0

From keras to C



Memory usage on MCU

Complexity: 262504 MACC  
 Used Flash: 65.64 KiB (65.64 KiB over 1024.00 KiB Internal)  
 Used Ram: 10.48 KiB (10.48 KiB over 128.00 KiB Internal)  
 Achieved compression: -  
 Analysis status: done



# Results

Accuracy: 99%

Footprint on STM32L475 MCU:

- Flash: 66 KB
- RAM: 11 KB

Complexity

- 270K MACC

$$\frac{\# \text{ training samples}}{\# \text{ parameters}} = \frac{2000}{12000} \cong 17\%$$

Confusion matrix:

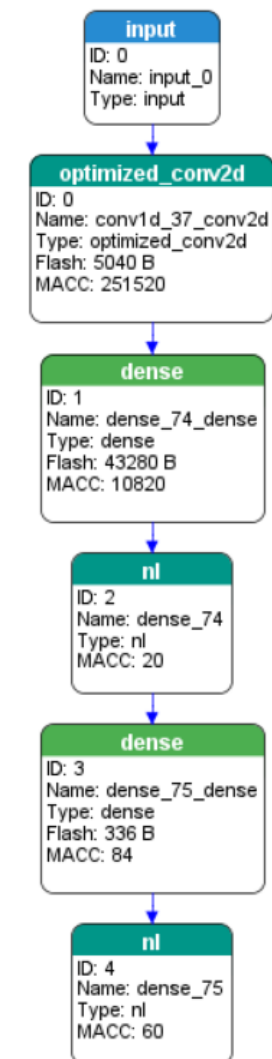
	VB	IB	LW	RW
VB	98.6%	1%	0.3%	0.1%
IB	0.7%	99,3%	0	0
LW	0	0.2%	99.8%	0
RW	0.3%	0	0	99.7%

**VB:** Voluntary (or forced) Blinks

**IB:** Involuntary Blinks

**LW:** Left Winks

**RW:** Right Winks



# Execution on STM32L475 MCU

Inference time: 0.02 s

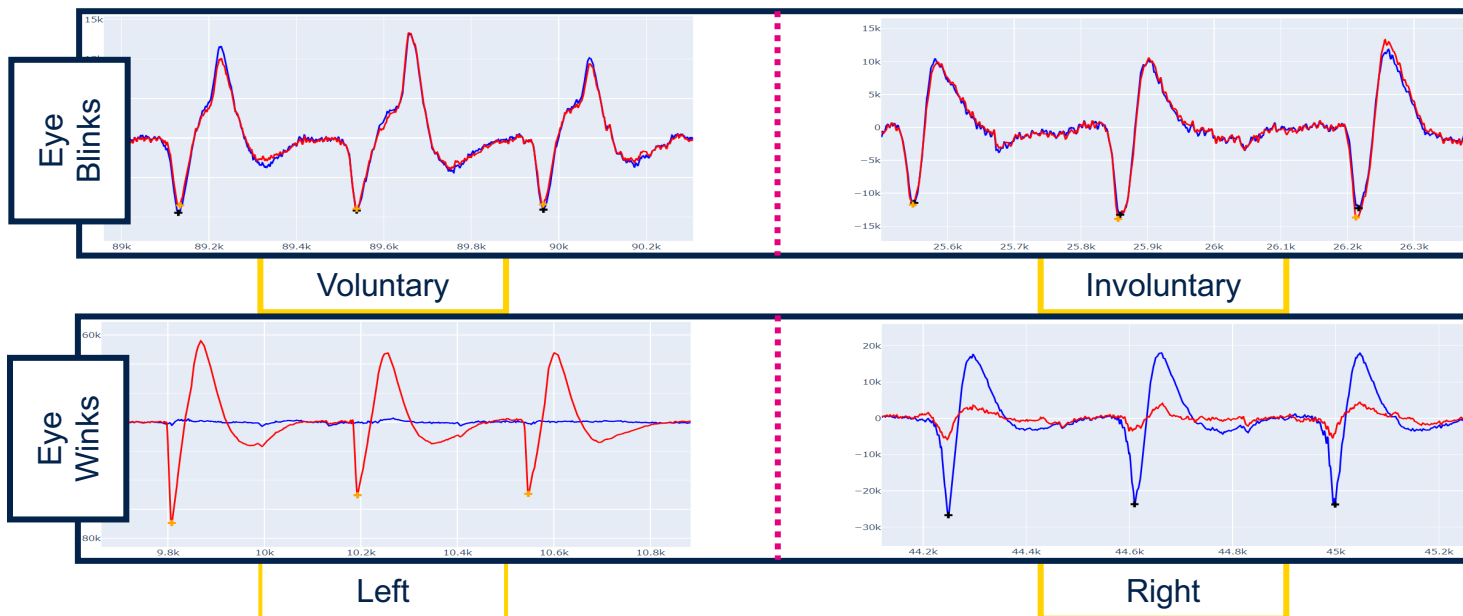
Acquisition time:  $200/250 = 0.8$  s (fs = 250 hz)

Total time: 0.82 s

Power consumption: 0.67 mW

GO AHEAD 20cm

NOTHING TO DO



Commands sent to the Robot

**VB:** Go ahead 20 cm

**IB:** Nothing to do

**LW:** Turn 45° left

**RW:** Turn 45° right

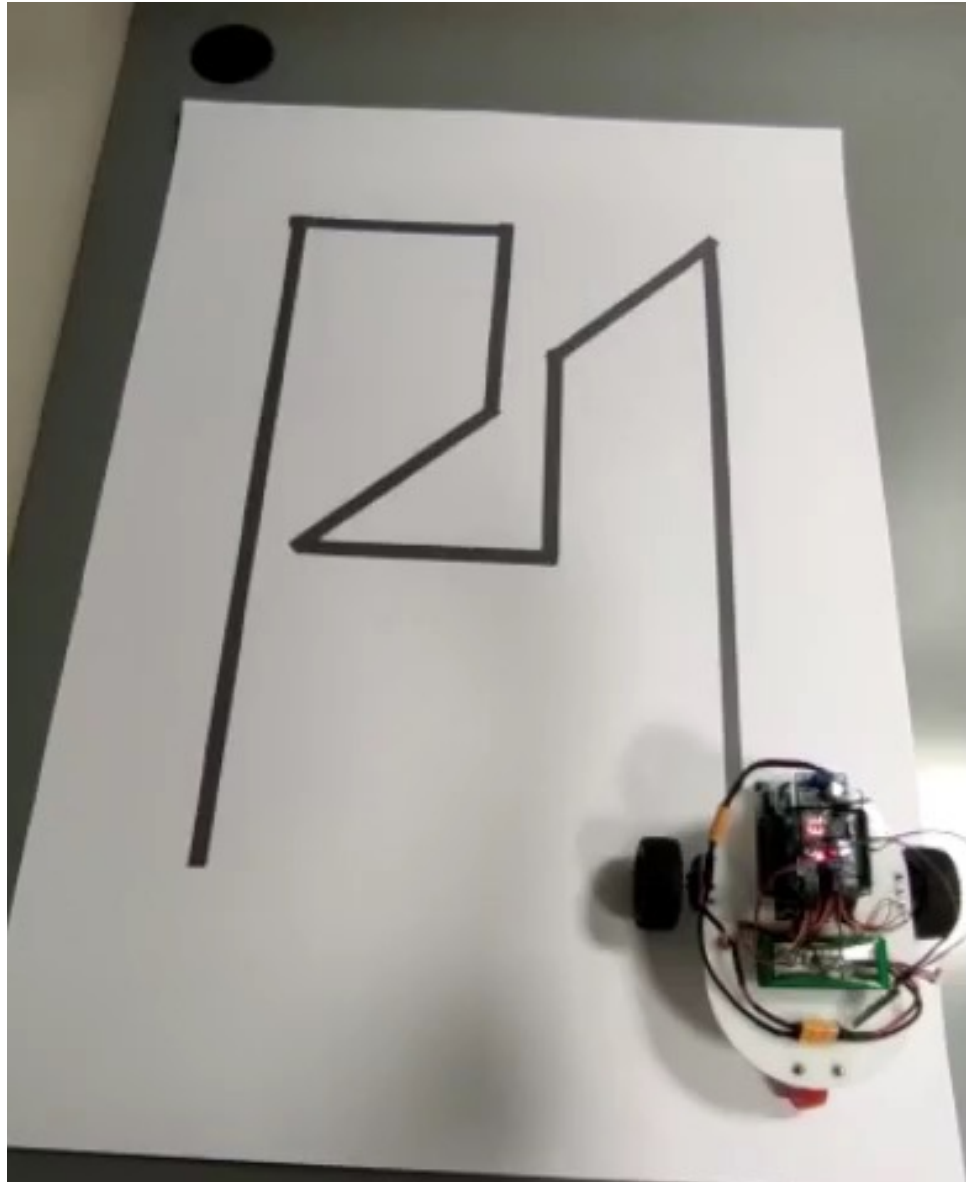
TURN LEFT 45°

TURN RIGHT 45°





# Complete algorithm running on the microcontroller



# Conclusions

First BCI system running on a tiny embedded device

The CNN achieves 99% of accuracy

It works in real time (0.02 s)

## Future works

Increase robustness to noise artifacts and body movements

Personalization

Gaze tracking from EOG signals





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# Questions?

[1] Bruna A, Tomaselli V, Chepyk O, Mammone N, Morabito F C, Ruggeri G, Campolo M, "An Embedded EOG-based Brain Computer Interface System for Robotic Control", International Conference on smart and sustainable technologies (SpliTech), Split 20-23 June, 2023

[2] Chepyk O, Bruna A, Tomaselli V, Mammone N, Campolo M, Ruggeri G, Ieracitano C, Morabito F C, "An Embedded Deep Neural Model for EOG-controlled Brain Computer Interfaces", WIRN 2023, 7-9 June 2023, Vietri sul mare (Italy)

[3] Lo Giudice M, Mammone N, Ieracitano C, Campolo M, Bruna A R, Tomaselli V, Morabito F C, "Visual Explanations of Deep Convolutional Neural Network for eye blinks detection in EEG-based BCI applications", IEEE World Congress On Computational Intelligence, 18-23 July 2022, Padua (Italy)

[4] Ieracitano C, Mammone N, Lo Giudice M, Tomaselli V, Bruna A R, Morabito F C, "EEG recordings and Eye Tracking for Brain-Computer Interfaces and Robotics", Automatica 2021, Catania 8-9 September 2021

[5] Lo Giudice M, Varone G, Ieracitano C, Mammone N, Bruna A R, Tomaselli V, Morabito F C, "1D Convolutional Neural Network approach to classify voluntary eye blinks in EEG signals for BCI applications", IEEE World Congress On Computational Intelligence, 2020



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