

# A low-cost EMG sensor for real-time data acquisition

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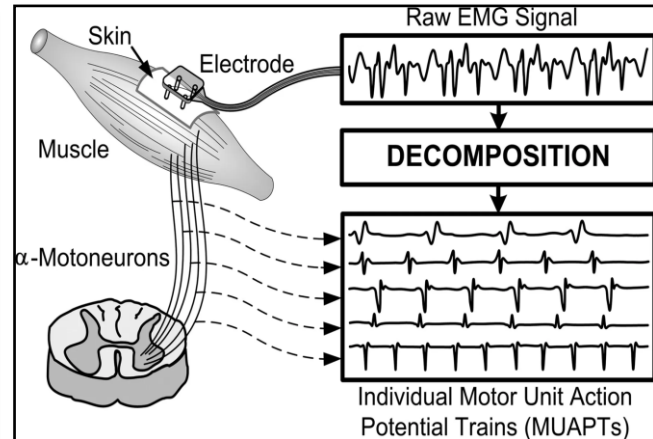
- Project background
- Aims & objectives
- Processing and classification stages
- EMG sensor
- Conclusion and future steps
- References

- **What is a myoelectric prosthesis?**

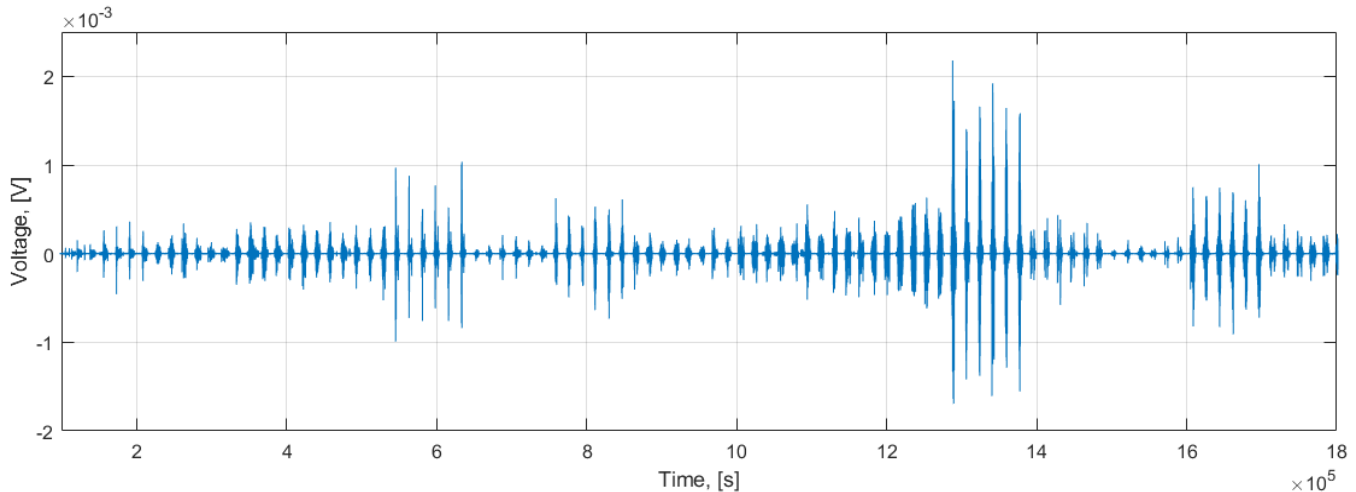
- An *artificial limb* actuated through **surface electromyography signals** (sEMG) and aimed to restore the necessary functions of an intact hand

- **What is an sEMG signal?**

- A sum of *individual electrical pulses* in a muscle recorded by electrodes during muscle contraction



- sEMG is a **non-stationary** bio-signal
  - sEMG amplitude: **0.5 - 2 mV**
  - sEMG useful frequency band: **20 – 300 Hz**
  - Total gain of the circuit:  **$\sim 1000$**



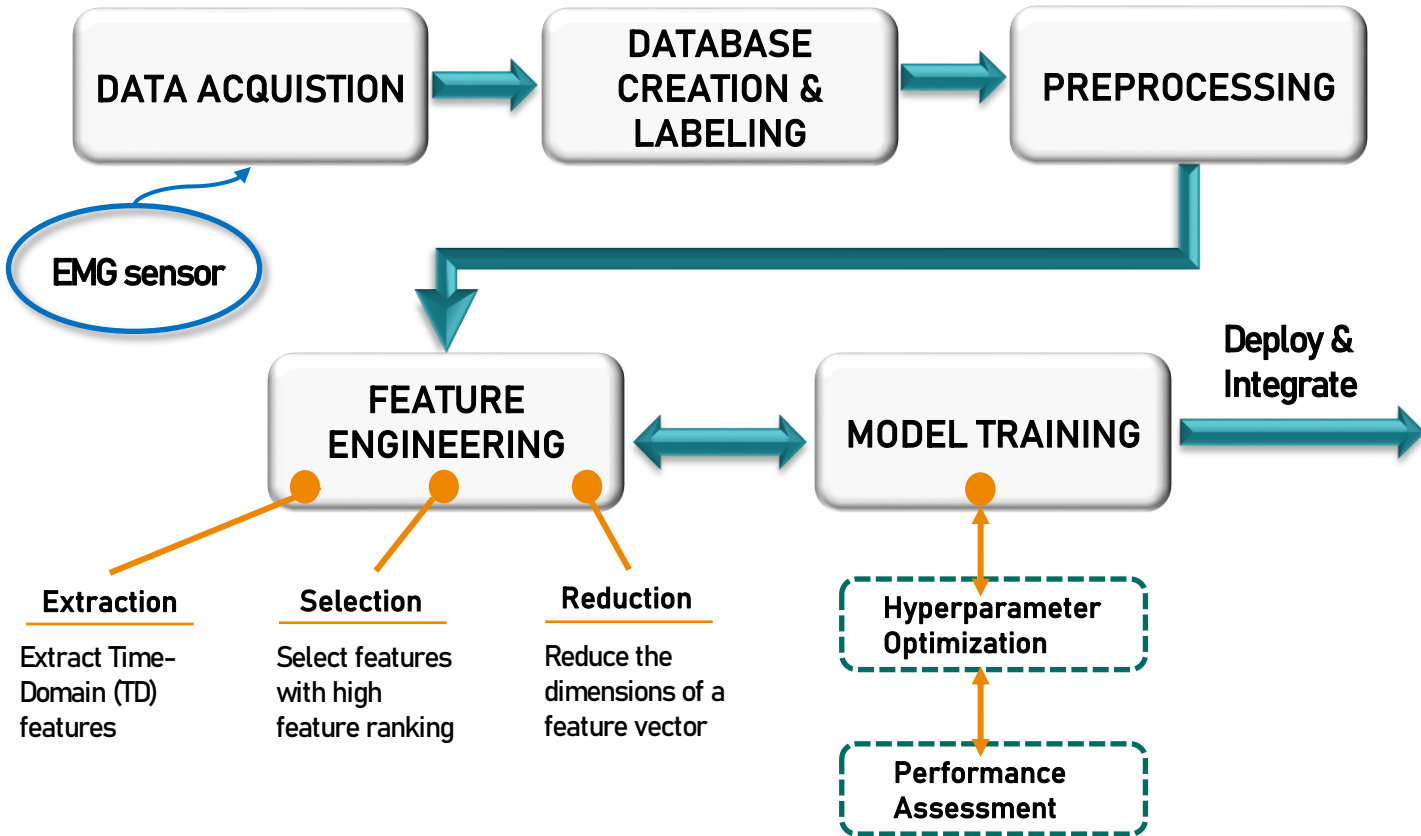
**Figure 1.** The raw EMG signal

## Aims

- Develop a *real-time* sEMG signal **processing & classification** model for a myoelectric prosthesis design
- Produce an efficient and cost-effective myoelectric hand prototype

## Objectives

- Make a **comprehensive** study of the most relevant sEMG PR models
- Apply *filtering*, *feature extraction* and *classification* algorithms resulting into highest classification metrics and real-time performance
- Implement the model on a microcontroller

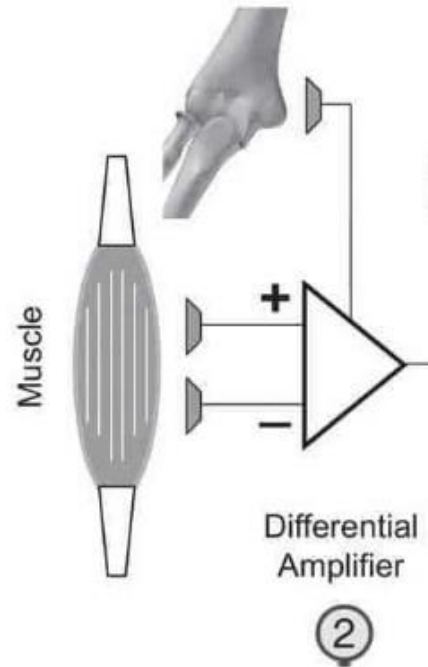


## Differential/Instrumentation

### Saturation:

- Common-mode signals from adjacent equipment (up to 1.5 V)
- DC potentials at different parts of the body of  $\pm 500\text{ mV}$
- Raw AC EMG signals are of **0.5 -2 mV**

The unwanted **DC signals** could be x300-400 times of **EMG signals**, also amplified by the Differential Amplifier, causing it to saturate.



**Figure 2.** EMG signal acquisition

## Solutions:

- 1) Limit the input gain of the Diff. Amp/INA to not to amplify DC
- 2) Implement the integrator/LP filtering to remove the differential DC
- 3) Apply high gain in the output stage of the Diff. Amp/INA to boost the AC (EMG) signal of interest

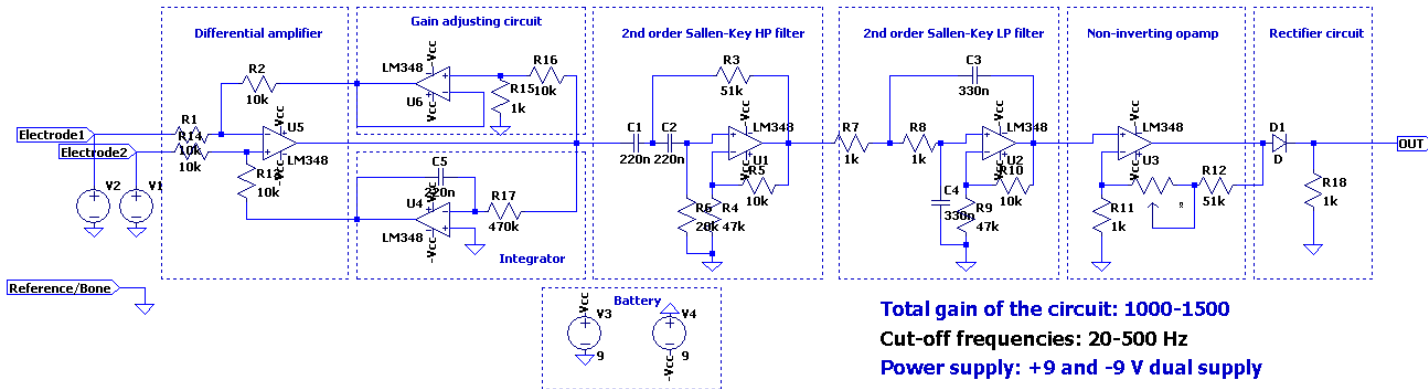


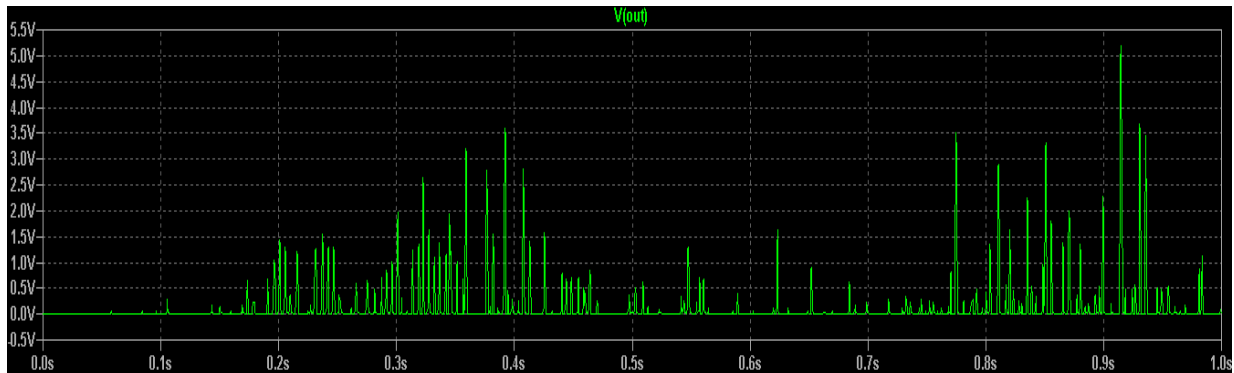
Figure 3. EMG sensor



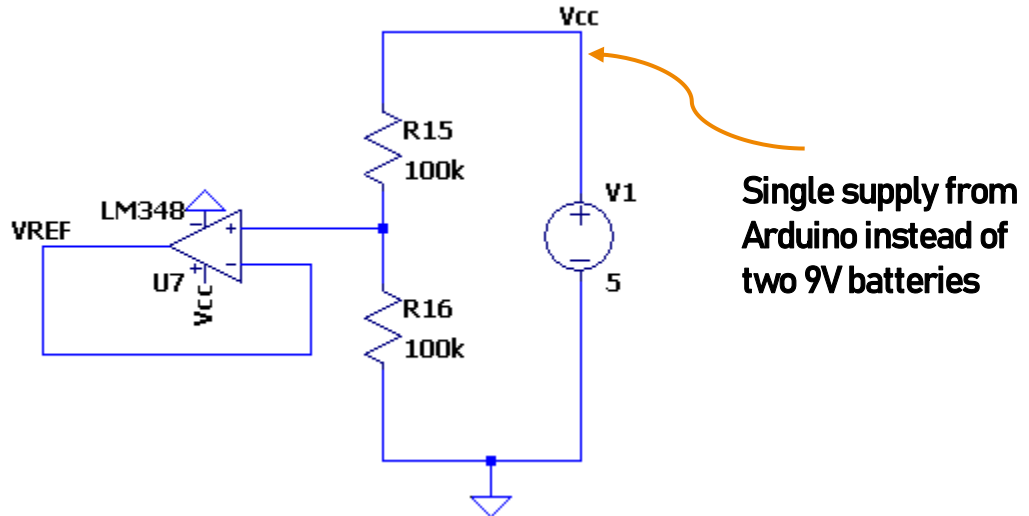
**Different raw EMG input:**



**Output signal:**



If we do not want a rectified version of the processed EMG, we can shift our reference value (offset) from 0 V to 2.5 V (**Arduino** accepts **0-5V**) using the schematic below.



**Figure 4. Voltage divider and voltage buffer to produce a new reference voltage of 2.5 V**

- The reasons of the OpAmp saturation in real-time experiments were discussed
- The solutions to this problem, as well as the models of the modified EMG sensor were introduced

## **By the end of Capstone I, I plan to achieve:**

- ✓ The dataset of EMG signals acquired from a DIY sensor
- ✓ Trained and optimized Pattern-Recognition (PR) model based on an ML algorithm

- [1] M. Atzori and H. Muller, "The Ninapro database: a Resource for sEMG Naturally Controlled Robotic Hand Prosthetics", in *37th Annual International Conference of the IEEE Engineering in Medicine and Biology Society (EMBC)*, Milan, 2016.
- [2] C. Spiewak, "A Comprehensive Study on EMG Feature Extraction and Classifiers", *Open Access Journal of Biomedical Engineering and Biosciences*, vol. 1, no. 1, 2018. Available: [10.32474/oajbeeb.2018.01.000104](https://doi.org/10.32474/oajbeeb.2018.01.000104).
- [3] M. Atzori et al., "Electromyography data for non-invasive naturally-controlled robotic hand prostheses", *Scientific Data*, vol. 1, no. 1, 2014. Available: [10.1038/sdata.2014.53](https://doi.org/10.1038/sdata.2014.53).
- [4] M. Ortiz-Catalan, R. Brånemark and B. Håkansson, "BioPatRec: A modular research platform for the control of artificial limbs based on pattern recognition algorithms", *Source Code for Biology and Medicine*, vol. 8, no. 1, 2013.

# Thank you!

