Week 23

Tongue EMG XR Project

Getting Started

Anatomy, EMG theory and Related Work

Testing OpenBCI Cyton board

First Recording (02/06/2025)

First Data Analysis

More recordings and their data analysis

Real time & Unity

Next Week

Getting Started

- I got started with the project, read the project overview, prepared a folder and GitHub.
- HyperDecahedron/tongue_emg_xr_project: Project to identify tongue's position as an XR input.

Anatomy, EMG theory and Related Work

- Researched the mouth, tongue and neck anatomy to familiarise with the muscles.
 Noted down some muscles of interest:
 - Submental triangle
 - o Sternocleidomastoideus
 - o Masseter: right/left?
 - o Genioglossus
 - Styloglossus
 - Pulls the tongue back and up.
 - Harder to access.
 - Hyoglossus
 - Under chin, slightly lateral.
 - Suprahyoid muscles (mylohyoid, geniohyoid)
 - Under chin, near genioglossus.
 - o Forehead for teeth clench
- I read SEMGIntro.fm to learn about EMG basics.

- Read the work by Zhang 2014 (TongueSee), and noted down important information for the online classification
 - o 2048 Hz
 - Window of 128 samples
 - RMS of the window.
 - o RMS of signed difference between signal value and mean value.
 - Majority voting

Testing OpenBCI Cyton board

- Muscles work up to 500 Hz, but the board samples at 250 Hz. Zhang2014 employs 2048 Hz. We might have to increase the sampling rate.
- Installed drivers and GUI
- Tested the signal in several muscles and selected those that worked better.
 - All of them: very sensitive when I move my head!
 - o Forehead for teeth clench VS Masseter
 - **■** Forehead much better than masseter
 - Can detect different right/left.
 - Submental triangle left/right
 - o Sternocleidomastoideus
 - Clear signal, no difference left/right
 - Genioglossus & Styloglossus
 - Genioglossus: same as submental
 - Styloglossus: very light signal when going left/right
 - Simple tongue touch
 - Useless
 - Hyoglossus
 - Good if touching the last molars
- Tested board settings
 - o Set SRB2 off
 - Creep-: default value is the best.

First Recording (02/06/2025)

- Created a key logger file (IDE Visual Studio) that runs via an Anaconda Prompt.
- 3 classes (left, right, front)
- First I press the class on the keyboard, then I move the tongue.

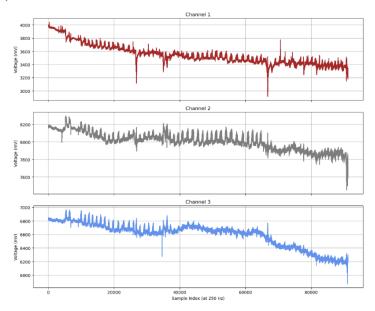
Channels:

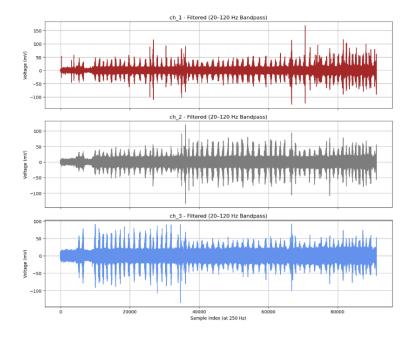
- o 3x channels in submental triangle.
- Channel 1: Submental center
- o Channel 2: Submental right
- Channel 3: Submental left

First Data Analysis

Filtering

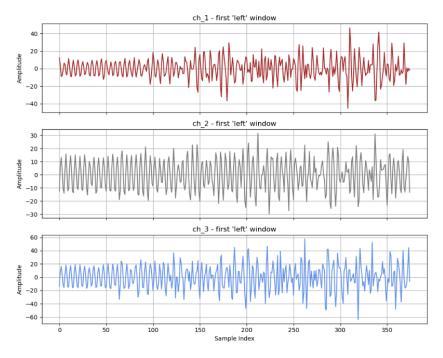
 Bandpass filter with 20-100 Hz. Can't raise the upper bound because of Nyquist.





Windows

- I used the timestamps from key_logger to detect the windows.
- Window size: 1.5s



• Feature extraction

o Added RMS, ZC and WL.

Classification

- o SVM
 - Best results with linear kernel
 - Accuracy: 0.94 (left, right, front)
- Random Forest Classifier
 - Changing the number of classificators doesn't affect the result
 - Accuracy: 0.89

Fine tuning

- o I created a list of variables to check for the next recording.
 - Recording
 - Hz
 - Creep-
 - Position of the sensors
 - Filtering
 - Highcut
 - Type of filtering
 - Windowing
 - Size of the window in seconds
 - Feature extraction
 - RMS, ZC, WL
 - Anything else?
 - Classification

- SVM
 - Kernel
- Random Forest Tree

More recordings and their data analysis

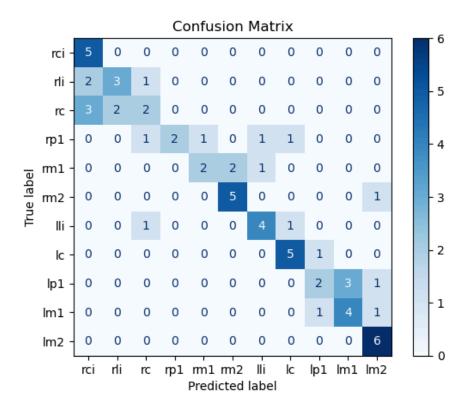
- **Higher sample rate?** Not directly possible, we would need an SD card and the process is not straightforward. To not waste time, I will skip this.
- Creep- and low limit: not much change, will use default values.
- Second Recording
 - This recording was a repetition of the first one but with four classes instead of three (left, right, front, none).
 - Data Analysis:

| Offline accuracy | Classifier | Window size | Observations |
|------------------|------------|----------------|---|
| 0.58 | SVM | 1.5 | |
| 0.96 | RF | 1.5 | |
| 0.62 | SVM | 1 | |
| 0.75 | RF | 1 | |
| 0.62 | SVM | 2 | |
| 0.92 | RF | 2 | |
| 0.96 | RF | 1.5 | Z-Score normalization. |
| 0.88 | SVM | 1.5 | Z-Score, rectify, smooth |
| 0.79 | SVM | 1 | Z-Score, rectify, smooth |
| 0.92 | RF | 1 | Z-Score, rectify, smooth |
| 0.88 | SVM | 2 | Z-Score, rectify, smooth |
| 0.92 | RF | 2 | Z-Score, rectify, smooth |
| | | | Tested both of them with more features, but got less accuracy. tested features are: 'RMS', 'ZC', 'WL', 'MAV', 'SSC', 'WAMP', 'IEMG' |

• Third recording

- o To test tooth-by-tooth precision
- Updated the key_logger to include a class for each tooth.

- I recorded tooth by tooth, 20 times each, split in two recordings of the left and the right side.
- The accuracy was awful...
 - but I saw that the tooth that was more difficult to detect was the second premolar in both left and right sides...
 - ... so I repeated the analysis with both sides together and without the second premolar.
 - The final offline accuracy was 0.62, which is low, but the confusion matrix is very diagonal, so it always detects the correct tooth +- 1 tooth of error.



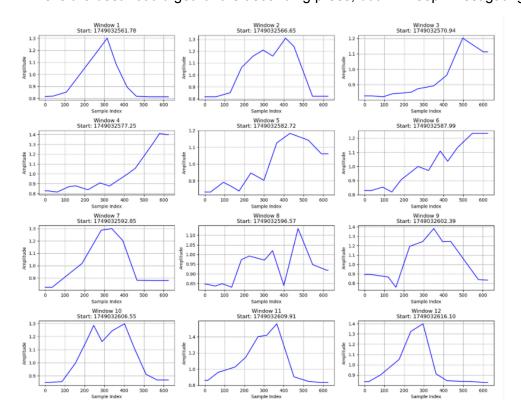
| TOOTH CLASSES | | | | | |
|---------------|-----|----------------------|-----|-----|-----------------------|
| key | id | description | key | id | description |
| 1 | lci | left central incisor | а | rci | right central incisor |
| 2 | Ili | left lateral incisor | s | rli | right lateral incisor |
| 3 | lc | left canine | d | rc | right canine |
| 4 | lp1 | left premolar 1 | f | rp1 | right premolar 1 |
| 5 | lp2 | left premolar 2 | g | rp2 | right premolar 2 |

| 6 | lm1 | left molar 1 | h | rm1 | right molar 1 |
|---|-----|--------------|---|-----|---------------|
| 7 | lm2 | left molar 2 | j | rm2 | right molar 2 |

| RIGHT + LEFT SIDES | | | | |
|--------------------|------------|----------------|--|--|
| Accuracy | Classifier | Window size | Observations | |
| 0.52 | SVM | 1 | all features with wamp threshold 1.1 | |
| 0.51 | RF | 1 | | |
| | | | The worst tooth to detect were lp2 and rp2 | |
| 0.62 | SVM | 1 | removing rp2 and lp2 | |
| 0.58 | RF | 1 | | |

Fourth recording

- Recorded tongue pressure from light to hard and logged the timestamp with key_logger.
- Logged in class:
 - I ascending
 - r light press
 - f hard press
- I tried to get a smooth envelop for the ascending class without using a classifier
- o This is the best I could get for the ascending press, but will keep investigating



Real time & Unity

- **Implemented the code** to execute the filtering, windowing, feature extraction and classification **in real time.**
 - o Decided to use a window size of 125 samples, but I will update it in the future.
- Implemented UDP receiver in Unity
 - o Moves XR origin depending on the class it receives.
- Tested with the board, but without the sensors.
 - I used the classifier that I had previously trained offline in the second recording.

Next Week

| ☐ Prepare presentation Kick off |
|---|
| ☐ 4 classes classifier |
| ☐ Try real-time classification of 4 classes with sensors + in Unity. |
| ☐ Online training? Get 4-5 samples to fine tune before starting the testing, is it |
| necessary? |
| ☐ Make windows overlap? |
| \square Implement code to save the predicted class and timestamp, while doing a |
| keylogger, then evaluate online accuracy. |
| ☐ 6 classes classifier |
| ☐ Record more data with 6 classes. Maybe add more channels. |
| ☐ Train classifier and expand to left, right, front, back, up, down. |
| ☐ Try real-time accuracy |
| ☐ If the accuracy is bad: |
| ☐ Reconsider positioning of senors. |
| ☐ Try to increase sampling rate. |
| ☐ Validate classifier generalizability (train/test split, cross-validation) |
| ☐ Think how to use those classes for navigation in XR. |
| ☐ Investigate regression for analog/force estimation (e.g., pressure via RMS amplitude) |