

Week 24

Tongue EMG XR Project

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Digital Future Events

- Presented the project overview in the Kick-off Introduction Event.
- Visited Astra Zeneca in Södertäljeon on Friday.

XR inputs

- Researched common uses of EMG in XR and proposed a list of actions/effects. To be further discussed when the online classifier is ready.

CNN

- Tested a simple CNN for 4-classes classification. The achieved accuracy was less than SVM (**CNN only reaches 53% in offline**), so this option was dismissed for the early state of the project, as I believe many more than 20 samples per class would be needed. However, I don't have experience in CNNs, so this remains to be further discussed as a potential classifier in later states.

7-Classes Online Classification

- I am stuck in this section.
- I started recording 30 samples of each class: left, left-front, front, right-front, right, swallow, clench, none (equivalent to no movement). Then, I repeated the offline classification, obtaining **94.9% of offline accuracy**.
- While recording, I found the "clench" movement to be painful, so I removed one of the classes. This way, the **offline accuracy increased up to 97.7%**.

- The new trained Random Forest classifier was tested online, achieving around **15% of online accuracy**. Currently, I am trying to find the problem and increase the accuracy.
- **Training pipeline:**
 - Data is recorded with 3 channels in the submental area. **The movement is continuous, not discrete.**
 - Tested two approaches: (A) filtering the windows or (B) filtering the whole signal. Both reach similar offline accuracy. I selected the (A) approach because I have implemented the online classification with window segmentation.
 - For each annotation timestamp, create 5 windows with time offset $[0, -2*\text{offset}, -\text{offset}, \text{offset}, 2*\text{offset}]$. This way there are more windows to train.
 - Filter: bandpass. Removed Z-score.
 - Get features. Added 4 more features.
 - Predict with RF using feature scaling.
- **Testing pipeline (real and fake):**
 - **To avoid having to repeat the same movements, I faked the online accuracy test with one recording I didn't use for training.**
 - The pipeline is exactly the same as the training, using the saved classifier (RF) and the save feature scaler.
 - The windows are buffers of 250 samples, as in the training.
 - The next buffer is 150 samples overlapped with the previous buffer.
- **Things that I tried and didn't improve the online accuracy**
 - Recording exactly in the same tooth position and on the same day, one for training and one for testing.
 - I repeated all recordings with discrete movements (not continuous). I got high offline accuracy, but still bad online accuracy.
 - Adding even more features, removing feature scaling and adding Z-score.
 - Tried SVM and KNN.
 - Training with and without offset windows.
 - Changing the bandpass filter.
 - Filtering techniques from this paper: [Surface Electromyography Signal Processing and Classification Techniques - PMC](#)
- **Why is the online classifier not working properly? Possible reasons:**
 - The training data is actually clustering in different positions for the same label. According to the PCA analysis, it seems that the classifier is overfitting (see image below).

- The amplitude of the signals varies between recordings, even if done on the same day. I should find a way to normalise it.



Next Week

- ☐ Online classifier
 - ☐ Discover why the online classification is currently so bad.
 - ☐ Validate classifier generalizability (train/test split, cross-validation)
 - ☐ Test with other users.
- ☐ Unity XR setup
 - ☐ Discuss VR setup, inputs and actions. Should the user move around the VR world, control a UI or select/interact with an object?