# Non-Invasive Brain-Computer Interfaces KU (709.028) | Summer term 2019

Jonas Ditz | Institute of Neurotechnology | TU-Graz

Unit #4 (09.05.2019)

# Steady state visual evoked potential (SSVEP) classification

#### **Exercise 4.1: Generate an artificial SSVEP signals**

The goal is to construct artificial SSVEP signals by superposition of sine waves and noise. Make use of the following specifications:

- Length of each signal: 100 seconds
- Sampling Rate of each signal: 256 Hz
- Class 1 signal:
  - o Frequency/Amplitude of sine: 8 Hz / 1
  - o Frequency/Amplitude of sine first harmonic: 16 Hz / 0.5
- Class 2 signal:
  - o Frequency/Amplitude of sine 2: 13 Hz / 1
  - o Frequency/Amplitude of sine 2 first harmonic: 26 Hz / 0.5
- Noise: (Matlab randn)
  - o Noise STD: 3

#### Task:

- Calculate 2 signals (class1 and class2) by summing sine signals and the noise.
- Make a plot of the signals. For the plot use a one second window of each of the two signals.
  What is the SNR of our signals for the main component?

## **Exercise 4.2: Extract feature from artificial SSVEP signals**

The goal for this exercise is to extract features and create training and test data sets for classification. Generate the signal using the following procedure:

- Divide your artificial SSVEP signals into non-overlapping segments with a length of one second. Note: You should get 100 signal segments per class.
- Apply a DFT to each of the windows (use your own implementation from unit 3. Alternatively use the Matlab function fft).
- Extract the 4 relevant frequency amplitudes (8, 13, 16, 26 Hz) from each Fourier transformed segments. These 4 amplitudes are the features for classification.
- Use half of class1 and half of class2 data to form the training set (features X\_train, class labels Y\_train). Use the remaining halves of these data for your test set (features X\_test, classlabels Y\_test).

#### Task:

Create the matrices X\_train (size 100x4; 2 classes 50 trials each), Y\_train (size 100x1), X\_test (size 100x4; 2 classes 50 trials each) and Y\_test (size 100x1)

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#### **Exercise 3: Artificial SSVEP classification with LDA at different SNRs**

In this exercise we want to classify the test data ( $X_test$ ) with an LDA classifier trained on the training data ( $X_test$ ). You can either use your LDA implementation (unit 2) or the code provided.

#### Task:

- Which accuracy do you achieve?
- Vary the SNR (change STD for the randn function). What are your accuracies achieved for STD = [3, 4, 5, 6, 7, 8, 9]? Make a plot with the accuracy and SNR over the STD.

## **Exercise 4: Classification of patient SSVEP data**

The goal for this exercise is to classify SSVEP data recorded from a patient (files  $X \cdot m$  and  $Y \cdot m$ ). The data was prepared as follows:

- Application on an fft to 1-second EEG segments (from second 4 to 5 of each trial).
- Selection of the most important frequencies.

The prepared data are built up as described below:

- X: (160 trials x 6 frequencies)
  - o 6 frequency bins (8,13,16,24,26,39 Hz)
  - O Stimulation for class 1: 8 Hz (harmonics: 16 and 24 Hz)
  - O Stimulation for class 2: 13 Hz (harmonics: 26 and 39 Hz)
- Y: (160 trials x 1 class label)

#### Task:

• Calculate a 10-times 10-fold cross-validation to estimate the classification accuracy. Again, you can either use your own LDA implementation or that I provided to you.

Pre-submission via TeachCenter at the end of the unit; final submission 19.05.2019