Graphing and exploring EMG data

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Overview

The objective of this data analysis practical is to graph and explore EMG recordings made from different skeletal muscles during various physical activities. The recordings to be analyzed can be found in our GitHub repository (https://github.com/emckiernan/electrophys (<a href="https://github.com/emckierna

Setting up the notebook

We begin by setting up the Jupyter notebook and importing the Python modules for plotting figures. We include commands to view plots in the notebook, and to create figures with good resolution and large labels.

```
In [62]: # command to view figures in Jupyter notebook
%matplotlib inline

# import plotting module
import matplotlib.pyplot as plt

# commands to create high-resolution figures with large labels
%config InlineBackend.figure_formats = {'png', 'retina'}
plt.rcParams['axes.labelsize'] = 18 # fontsize for figure labels
plt.rcParams['axes.titlesize'] = 20 # fontsize for figure titles
plt.rcParams['font.size'] = 16 # fontsize for figure numbers
plt.rcParams['lines.linewidth'] = 1.6 # line width for plotting
```

Next, we import various modules for extracting data and scientific computing.

```
In [63]: import numpy as np import scipy as sc import wave
```

Extracting and graphing the data

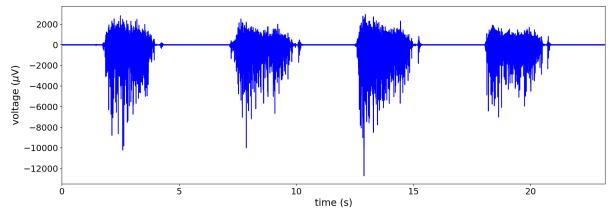
EMG recordings were obtained using the Backyard Brains EMG Spiker Box, and are saved as audio files in .wav format. So, we first have to open the .wav files and extract the data. We can also extract the number of recording channels and sampling rate.

```
\# open .wav file by specifying the path and filename
In [64]:
         record = wave.open('../data/S10 EMG calf intermittent.wav', 'r')
         # extract number of channels, sample rate, data
         numChannels = record.getnchannels() # number of channels
         N = record.getnframes() # humber of frames
         sampleRate = record.getframerate() # sampling rate
         dstr = record.readframes(N * numChannels)
         waveData = np.frombuffer(dstr, np.int16)
         # get the time window
         timeEMG=np.linspace(0, len(waveData)/sampleRate, num=len(waveData))
         # calculate frequency
         freq = 1/np.mean(np.diff(timeEMG))
         print('The recording has %d channel(s).' % (numChannels))
         print('The sampling rate of the recording is %d Hz.' % (sampleRate))
         The recording has 1 channel(s).
```

Now, let's plot the EMG. The following recording was made from the calf muscle during repeated calf raises.

The sampling rate of the recording is 44100 Hz.

```
In [60]: plt.figure(figsize=(18,6))
   plt.xlabel(r'time (s)')
   plt.ylabel(r'voltage ($\mu$V)')
   plt.plot(timeEMG,waveData, 'b')
   plt.xlim(0,max(timeEMG));
```



We can define a function to allow us to quickly extract and graph different EMG recordings. The input to the function is the name of the file we want to analyze.

```
In [57]: def EMG(file):
             # open .wav file by specifying the path and filename
             record = wave.open(file)
             # extract number of channels, sample rate, data
             numChannels = record.getnchannels() # number of channels
             N = record.getnframes() # number of frames
             sampleRate = record.getframerate() # sampling rate
             # extract data from the .wav file
             dstr = record.readframes(N * numChannels)
             waveData = np.frombuffer(dstr, np.int16)
             # print the number of channels and sample rate
             print('The recording has %d channel(s).' % (numChannels))
             print('The sampling rate of the recording is %d Hz.' % (sampleRate))
             # calculate time window
             timeEMG=np.linspace(0, len(waveData)/sampleRate, num=len(waveData))
             # calculate frequency
             freq = 1/np.mean(np.diff(timeEMG))
             # plot EMG
             plt.figure(figsize=(18,6))
             plt.xlabel(r'time (s)')
             plt.ylabel(r'voltage ($\mu$V)')
             plt.plot(timeEMG, waveData, 'b')
             plt.xlim(0,max(timeEMG));
             return
```

Exploring the activity of different muscles

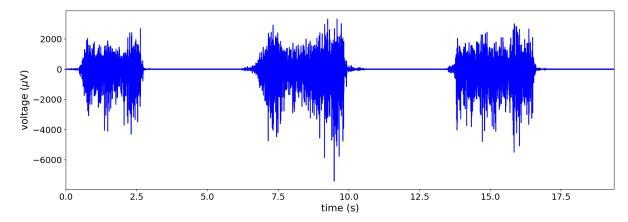
Using our function, we can now dive into our repository of recordings and look at the EMGs of different muscles recorded during various physical activities.

Bicep muscle

The following EMG was recorded from the bicep muscle during repeated (3) contractions with intermittent rest periods.

```
In [65]: EMG(file='../data/S10_EMG_bicep_intermittent.wav')
```

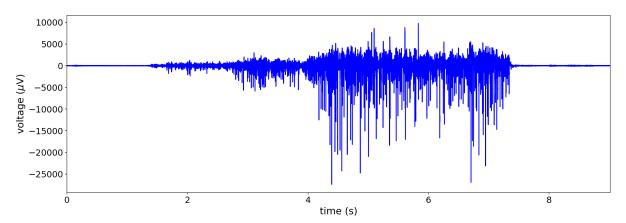
The recording has 1 channel(s). The sampling rate of the recording is 44100 Hz.



The following EMG was recorded from the bicep as the subject sustained a contraction and gradually increased the amount of force (i.e., recruitment).

In [28]: EMG(file='../data/S10_EMG_bicep_recruitment.wav')

The recording has 1 channel(s). The sampling rate of the recording is 44100 Hz.



Study questions and exercises:

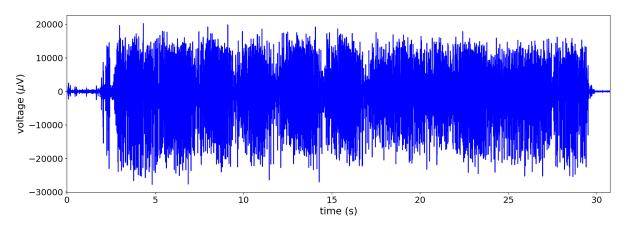
- How do the recordings differ when we look at intermittent versus sustained muscle contractions?
- Which features of the EMG recording change as the subject increases the force of contraction, and how? In other words, how might we quantify motor unit recruitment?
- What other exercises could you carry out while recording from the bicep muscle, and how would you expect the EMG recordings to look?
- If you have recording equipment available, record from your own bicep muscle and then extract and graph your data.

Tricep muscle

The following recording was made from the tricep muscle while the subject performed tricep dips.

```
In [36]: EMG(file='../data/S1_EMG_tricep_dips.wav')
```

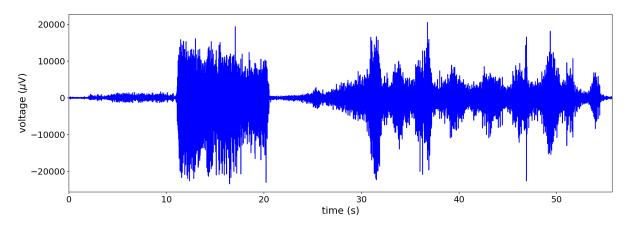
The recording has 1 channel(s). The sampling rate of the recording is 44100 Hz.



This next recording is more complex. It was made from the tricep muscle and included the following sequence of activities: (1) for the first 10 seconds, the subject was at rest, (2) then for the following 10 seconds, the subject turned their arm to activate the tricep isometrically, (3) then the subject changed position, and (4) finally, during the last 20 seconds, the subject repeatedly lifted and lowered a bottle with water inside (i.e., similar to an overhead tricep extension).

```
In [38]: EMG(file='../data/S1_EMG_tricep_twistWeight.wav')
```

The recording has 1 channel(s).
The sampling rate of the recording is 44100 Hz.



Study questions and exercises:

- Why do you think the tricep dip recording shows a high level of continuous activity despite the fact that the dips involve an up-and-down movement?
- How is the tricep dip activity different from the repeated tricep extension at the end of the second recording, and why?
- What other exercises could you carry out while recording from the tricep muscle, and how would you expect the EMG recordings to look?
- If you have recording equipment available, record from your own tricep muscle and then extract and graph your data.

Forearm muscles

The following recording was made from the forearm while the subject repeatedly pressed a hand gripper exercise device. It shows five contractions in total, with the last one sustained for a longer period of time than the previous four.

```
In [39]: EMG(file='../data/S2_EMG_forearm_grip.wav')

The recording has 1 channel(s).

The sampling rate of the recording is 44100 Hz.
```

The next recording was made from the forearm while the subject participated in an arm wrestling match. Around second 14, the subject lost the match.

10

time (s)

12

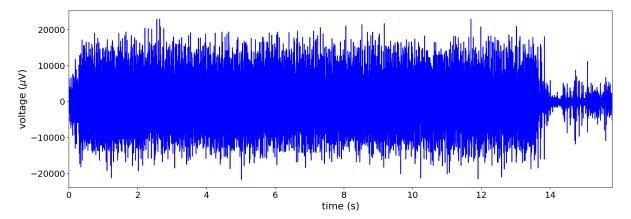
14

16

4

```
In [42]: EMG(file='../data/S2_EMG_forearm_wrestle.wav')
```

The recording has 1 channel(s). The sampling rate of the recording is 44100 Hz.



Study questions and exercises:

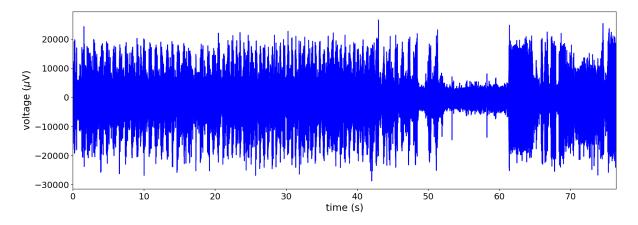
- What muscles are found in the forearm, and which could be contributing to the EMG signal?
- What other muscles, besides those in the forearm, might be activated during an arm wrestling match? Would you expect to see differences in their EMGs during this activity?
- What other exercises could you carry out while recording from the forearm muscles, and how would you expect the EMG recordings to look?
- If you have recording equipment available, record from your own forearm and then extract and graph your data.

Jaw muscles

The following EMG was recorded from the jaw muscles. For the first 50 seconds, the subject chewed a soft, gummy candy. Then there were 10 seconds of rest. And finally, for the last ten seconds the subject made a forced smile.

```
In [43]: EMG(file='../data/S3_EMG_jawMuscle_chewSmile.wav')
```

The recording has 1 channel(s). The sampling rate of the recording is 44100 Hz.

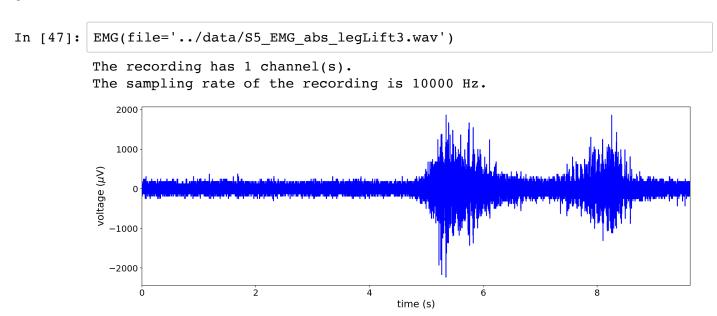


Study questions and exercises:

- · Which muscles control chewing and smiling, and which could be contributing to the EMG signal?
- What differences do you see in the EMGs recorded when the subject is chewing versus smiling, and what do you think causes these differences?
- What other activites could you carry out while recording from the jaw or other facial muscles, and how would you expect the EMG recordings to look?
- If you have recording equipment available, record from your own jaw muscles, try experimenting with different electrode placements on the face, and then extract and graph your data.

Abdominal muscles

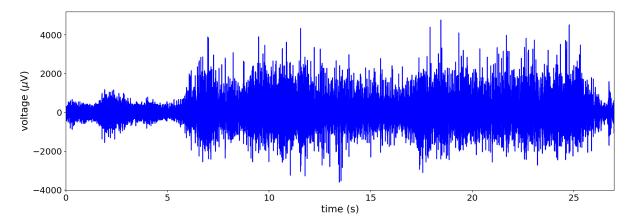
The following EMG recording was made from the abdominal muscles, specifically rectus abdominis, during lying leg raises.



During the next recording, the subject performed a sustained plank.

```
In [48]: EMG(file='../data/S5_EMG_abs_plank.wav')
```

The recording has 1 channel(s).
The sampling rate of the recording is 10000 Hz.



Study questions and exercises:

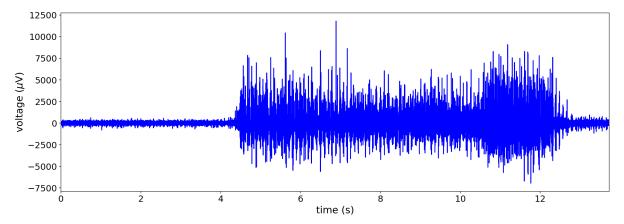
- What do you think would happen to the EMG recordings if the subject performed more lying leg raises, or tried to hold the plank for longer? In other words, how might fatigue show up on the two recordings?
- How could we quantify fatigue, and how might the quantification be different for the two types of exercise?
- What other exercises could you do to activate the abdominal muscles, or which other ab muscles could you record from, and how would you expect the EMG recordings to look?
- If you have recording equipment available, record from your own abdominal muscles, try experimenting with different electrode placements and exercises, and then extract and graph your data.

Quadricep muscles

The following EMG was recorded from the quadricep muscles while the subject performed a one-legged squat, also known as a Bulgarian squat. The subject also had a backpack filled with books on during the exercise for added weight.

```
In [52]: EMG(file='../data/S9_EMG_quadricep_oneLegSquat1.wav')
```

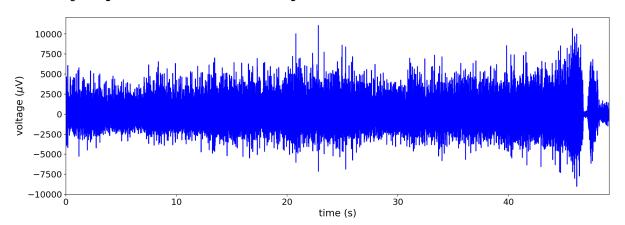
The recording has 1 channel(s). The sampling rate of the recording is 10000 Hz.



In the next recording, the subject sustained a wall squat, maintaining the backpack on top of the legs for added weight.

In [61]: EMG(file='../data/S9_EMG_quadricep_squatSustained.wav')

The recording has 1 channel(s).
The sampling rate of the recording is 10000 Hz.



Study questions and exercises:

- What muscles make up the quadricep group, and which could be contributing to the EMG signal?
- What factors could explain the fluctuations seen in the two recordings? What might have caused the increased amplitude at the end of the wall squat recording?
- What other exercises could you perform while recording from the quadricep muscles, and how would you expect the EMG recordings to look?
- If you have recording equipment available, record from your own quadriceps, try experimenting with different electrode placements, and then extract and graph your data.

You can explore more recordings from our repository. Or, record your own EMGs and experiment with electrode placement and different exercises to see how it affects the signal.

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