

# Real-time robot arm control using motor imaginary movements decoded from EEG signals

## Research Practice

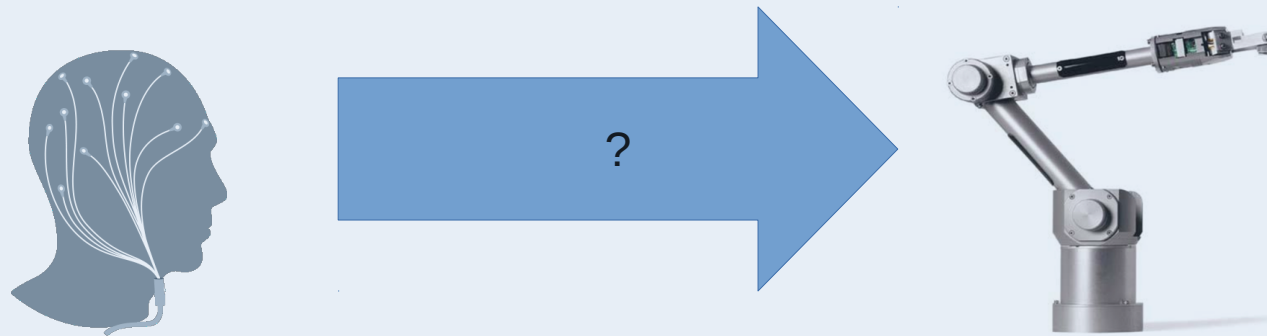
Juri Fedjaev

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# Overview

- 1) Introduction
- 2) Experimental Design
- 3) Results
- 4) Conclusion

# Introduction

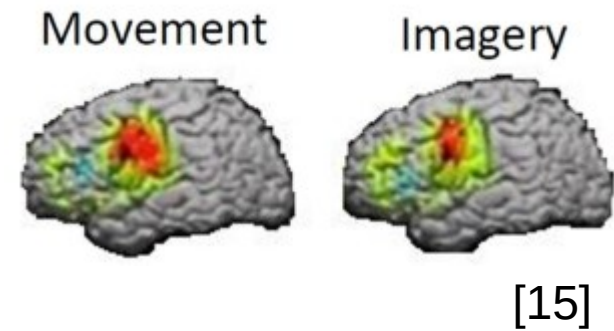


How to control a robot arm using nothing but our mind?

# Motor Imagery

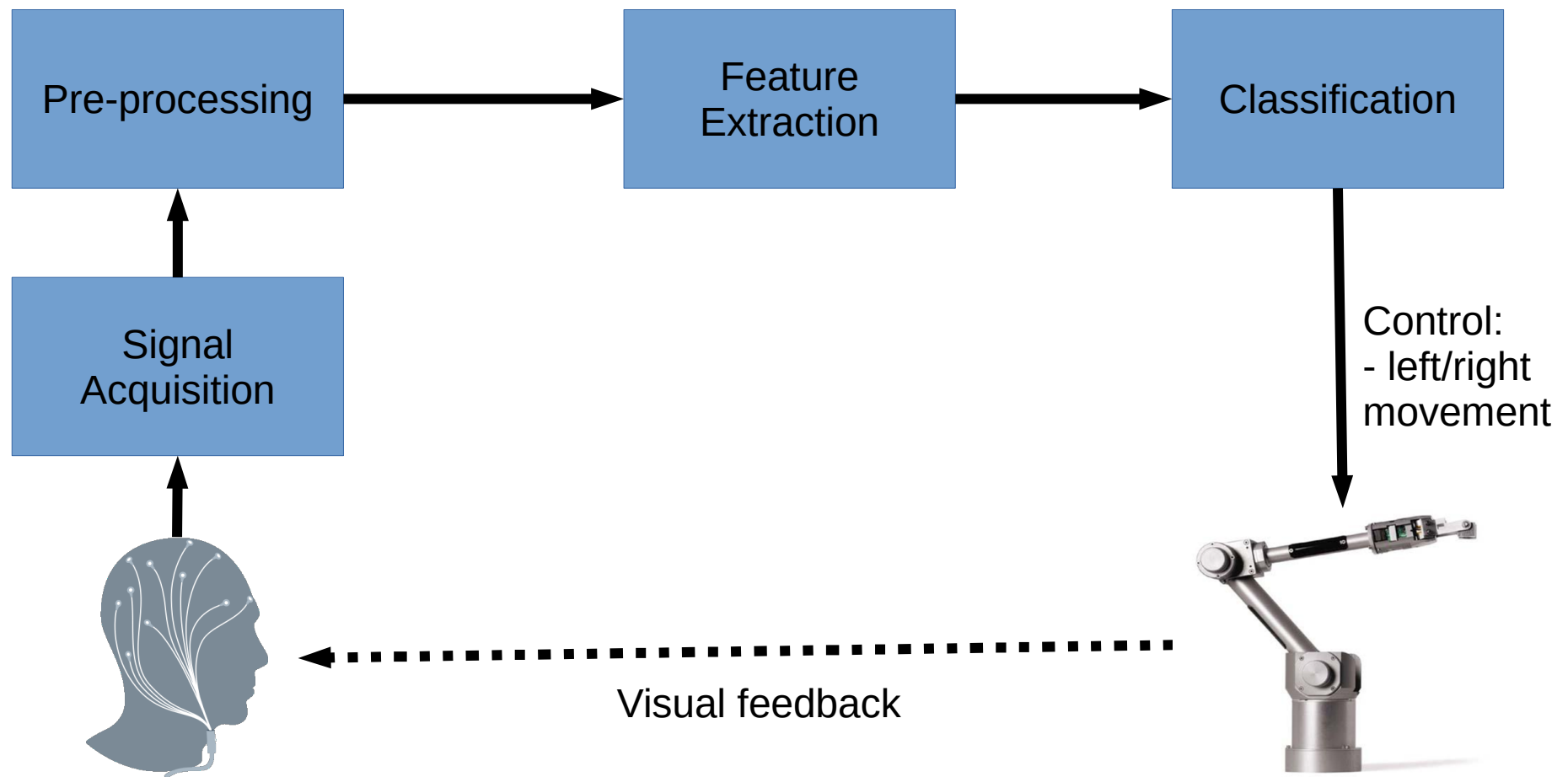
## Introduction

- Motor Imagery (MI) is the mental rehearsal of physical tasks
- Similar brain areas are activated as in physical execution
- Mainly present in mu rhythm (8-13 Hz) & beta band (14-30 Hz)



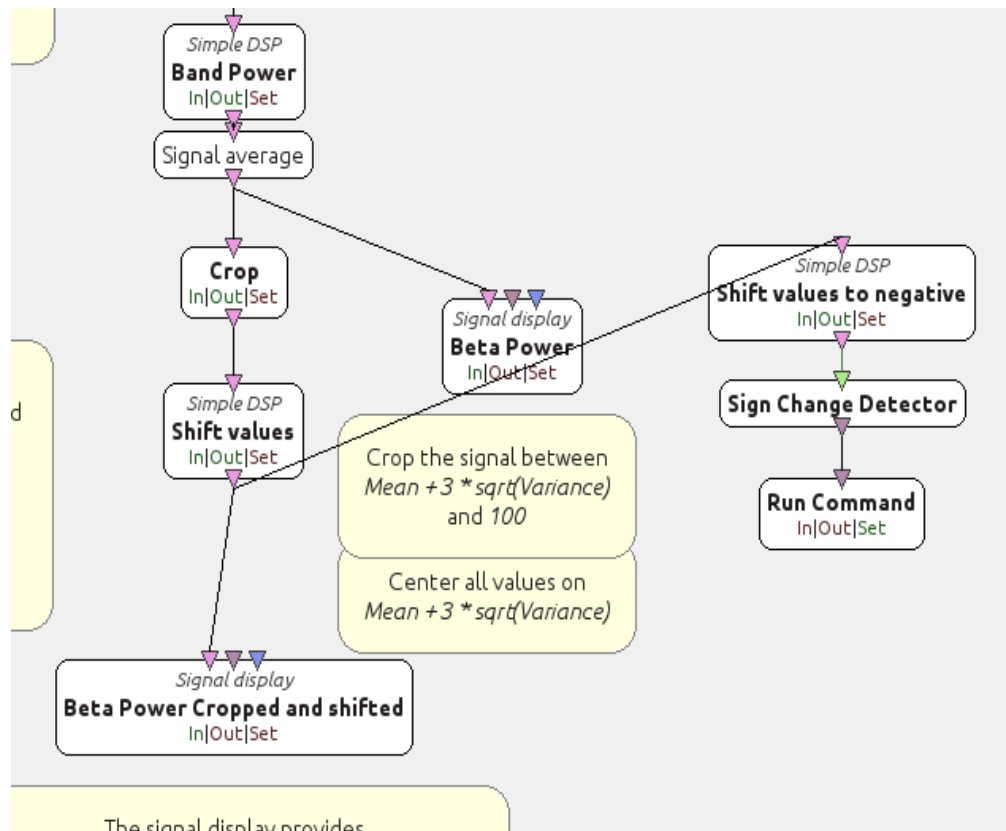
# Full system overview

## Introduction



# First approach – OpenVibe & Emotiv

## Experimental Design



The signal display provides

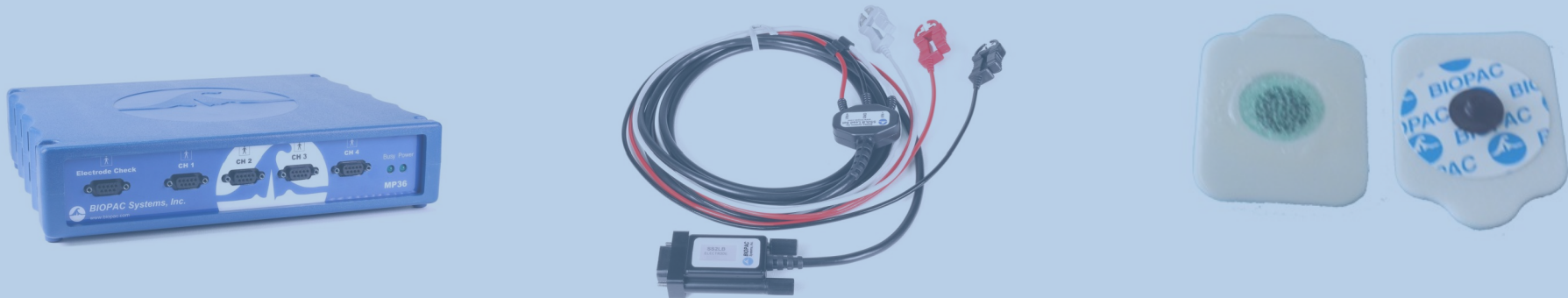


Emotiv EPOC+ headset

OpenVibe Workflow

# Second Approach – Biopac System

## Experimental Design

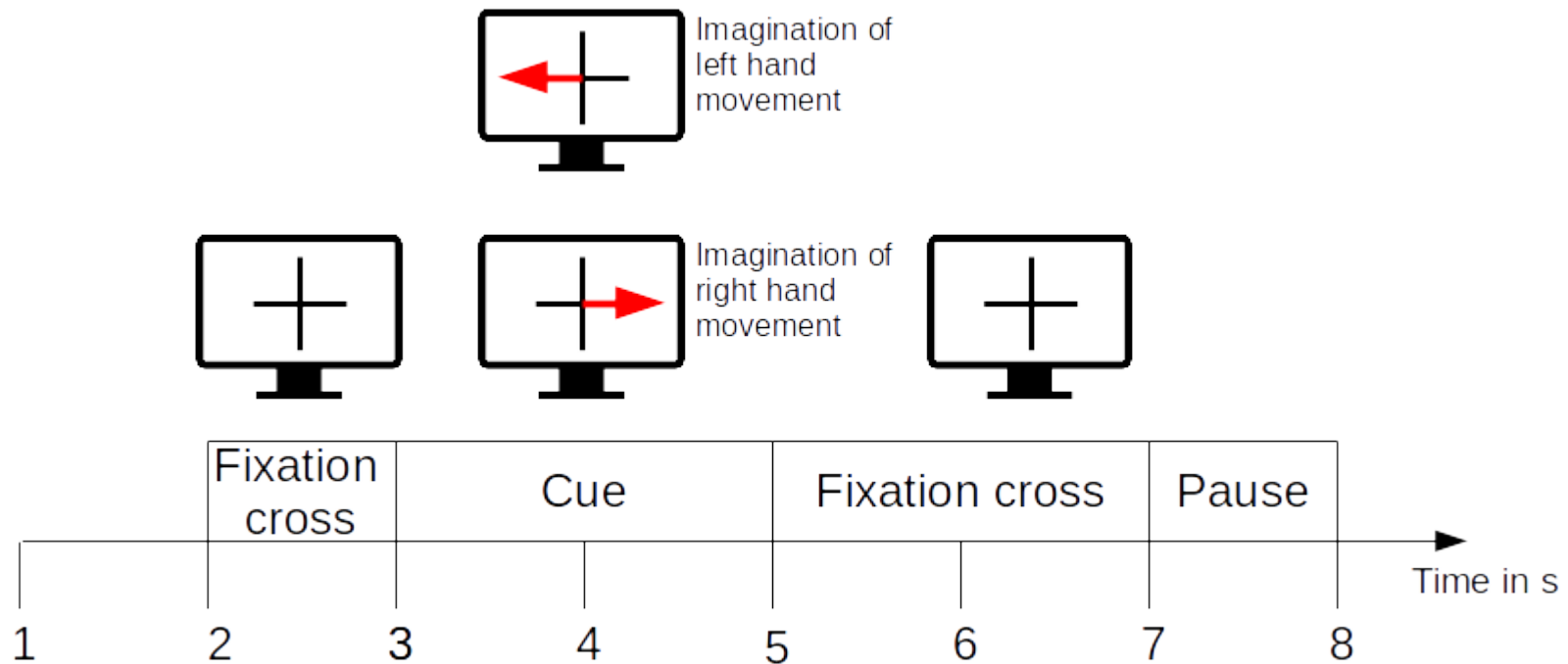


**Biopac MP36 Signal Acquisition Device**



# Recording Protocol

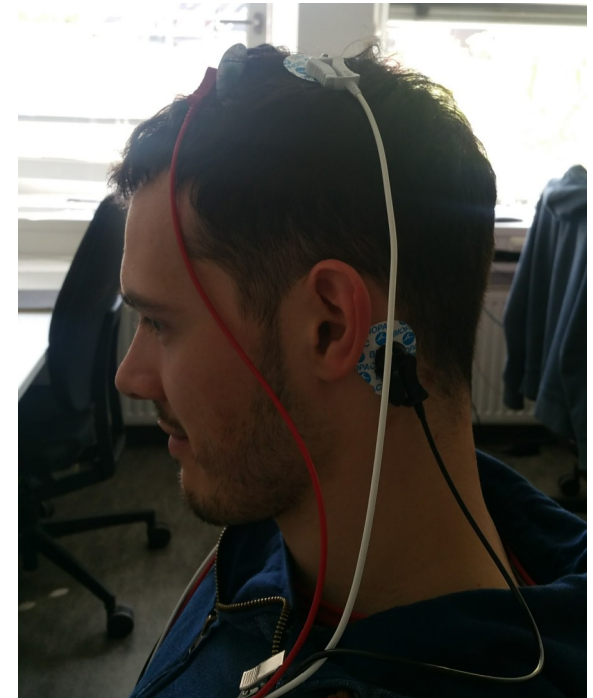
## Experimental Design





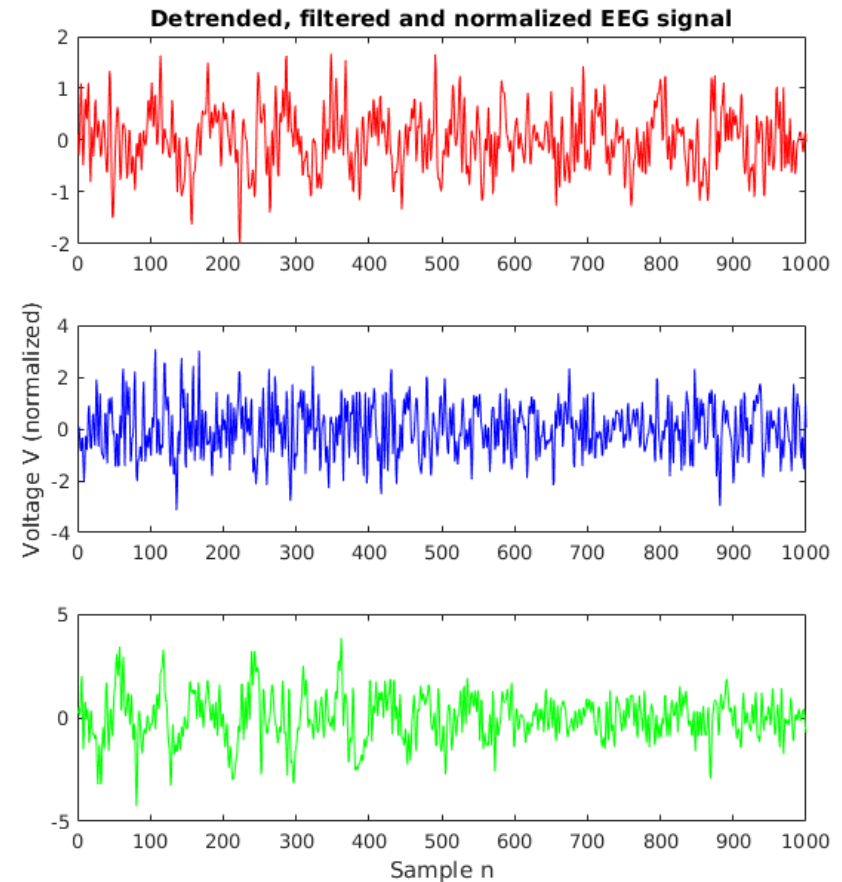
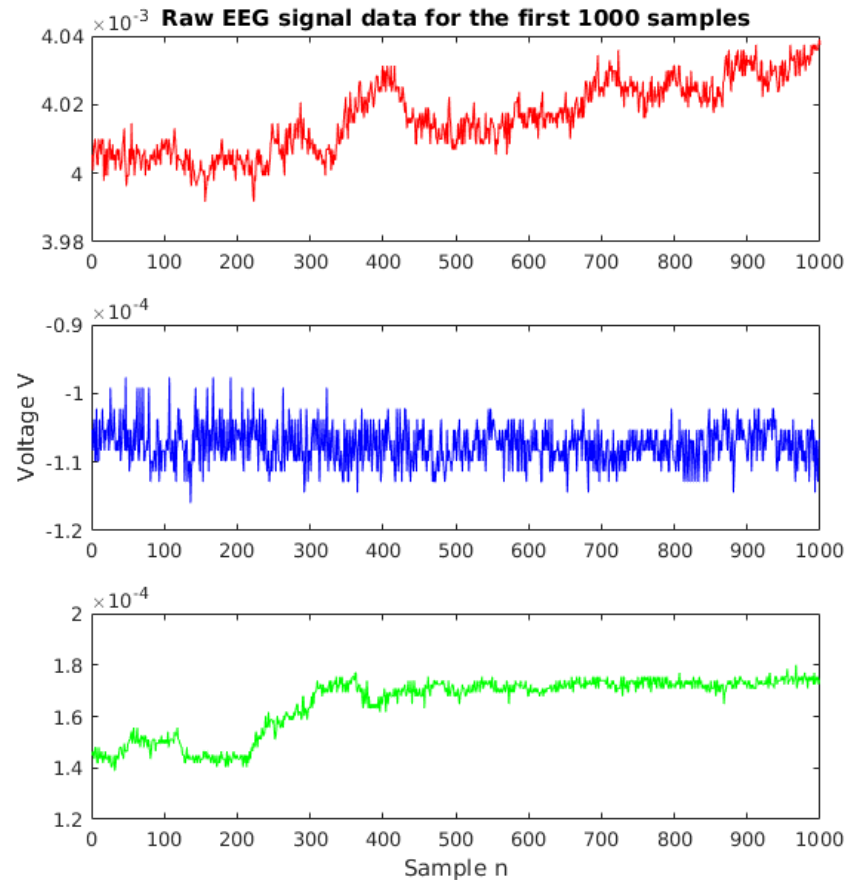
# Data Acquisition

## Experimental Design



# Signal Processing

## Experimental Design



# Feature Extraction

## Experimental Design

- 2000 ms epochs; 500 ms pre-stimulus baseline
- Band power (BP) features
  - 72 frequency bands using overlapping narrow bands in range between 8 and 30 Hz
- Total of 216 BP features
- Principal component analysis (PCA) with 12 coefficients → 12-D feature vector per trial

# Classification

## Experimental Design

- SVM model `fitcsvm()` from Matlab toolbox „Statistics and Machine Learning“
- Radial basis function (RBF) kernel
- 10-fold cross-validation with `crossval()`
- Misclassification rate with `kfoldLoss()`

# Robot Arm Control

## Experimental Design



- Python script for communication with SOAP server
  - `py.KatanaSoap.KatanaSoap()`
  - `katana.calibrate()`
  - `katana.moveMotAndWait(axis, val)`
  - `katana.closeGripper()`,  
`katan.openGripper()`
  - ...

# Off-line Classification

## Results

Recording ID	Number of trials	Accuracy (%)
AS01-11	40	63
AS02-11	40	67
JF02-10	40	63
JF03-10	40	74
JF04-10	40	74
AS02-12	40	<b>77</b>
JF02-15	50	66
JF02-16	50	60
JF01-15	50	58
JF01-16	50	60
AS01-16	50	46

# On-line Classification

## Results

[Video](#)

# Conclusion

- Simple 2-class BCI
  - Decent off-line accuracy of ~77%
- Unsuitable for on-line applications
  - Signal quality cannot be maintained over longer periods of time (→ electrode design)
  - Discomfort for user
- Possible improvements
  - Try different features: common spatial patterns (CSP), discrete wavelet transform (DWT), dynamic time warping (DWT), ...
  - Try other classifiers: Artificial or spiking neural networks (ANNs, SNNs), linear discriminant analysis (LDA), deep neural networks, ...



Thank you for your attention!

# Sources

- [1] N. Birbaumer, \Brain-computer-interface research: coming of age," 2006.
- [2] A. Sivakami and S. S. Devi, \Analysis of eeg for motor imagery based classification of hand activities,"
- [3] J. Meng, S. Zhang, A. Bekyo, J. Olsoe, B. Baxter, and B. He, \Noninvasive electroencephalogram based control of a robotic arm for reach and grasp tasks," *Scientific Reports*, vol. 6, 2016.
- [4] X. Yong and C. Menon, \Eeg classification of different imaginary movements within the same limb," *PloS one*, vol. 10, no. 4, p. e0121896, 2015.
- [5] J. Decety, \The neurophysiological basis of motor imagery," *Behavioural brain research*, vol. 77, no. 1, pp. 45-52, 1996.
- [6] M. Lotze, P. Montoya, M. Erb, E. Hülsmann, H. Flor, U. Klose, N. Birbaumer, and W. Grodd, \Activation of cortical and cerebellar motor areas during executed and imagined hand movements: an fmri study," *Journal of cognitive neuroscience*, vol. 11, no. 5, pp. 491-501, 1999.
- [7] R. W. Homan, J. Herman, and P. Purdy, \Cerebral location of international 10-20 system electrode placement," *Electroencephalography and clinical neurophysiology*, vol. 66, no. 4, pp. 376-382, 1987.
- [8] C. A. Kothe and S. Makeig, \Bcilab: a platform for brain-computer interface development," *Journal of neural engineering*, vol. 10, no. 5, p. 056014, 2013.
- [9] Y. Renard, F. Lotte, G. Gibert, M. Congedo, E. Maby, V. Delannoy, O. Bertrand, and A. Lécuyer, \Openvibe: An open-source software platform to design, test, and use brain-computer interfaces in real and virtual environments," *Presence*, vol. 19, no. 1, pp. 35-53, 2010.
- [10] G. Schalk and J. Mellinger, *A practical guide to brain-computer interfacing with BCI2000: General-purpose software for brain-computer interface research, data acquisition, stimulus presentation, and brain monitoring*. Springer Science & Business Media, 2010.
- [11] I. Biopac Systems, \Biopac product site," 2017.
- [12] S. Ge, R. Wang, and D. Yu, \Classification of four-class motor imagery employing single-channel electroencephalography," *PloS one*, vol. 9, no. 6, p. e98019, 2014.
- [13] R. Leeb, F. Lee, C. Keinrath, R. Scherer, H. Bischof, and G. Pfurtscheller, \Brain-computer communication: motivation, aim, and impact of exploring a virtual apartment," *IEEE Transactions on Neural Systems and Rehabilitation Engineering*, vol. 15, no. 4, pp. 473-482, 2007.
- [14] Z. Tayeb, E. Ercelik, and J. Conradt, \Decoding of motor imagery movements from eeg signals using spinnaker neuromorphic hardware," *arXiv*, 2017.
- [15] Miller, K. J., Schalk, G., Fetz, E. E., Nijss, M. den, Ojemann, J. G., & Rao, R. P. N. (2010). Cortical activity during motor execution, motor imagery, and imagery-based online feedback. *Proceedings of the National Academy of Sciences of the United States of America*, 107(9), 4430-5.

Questions?