

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

Research Practice

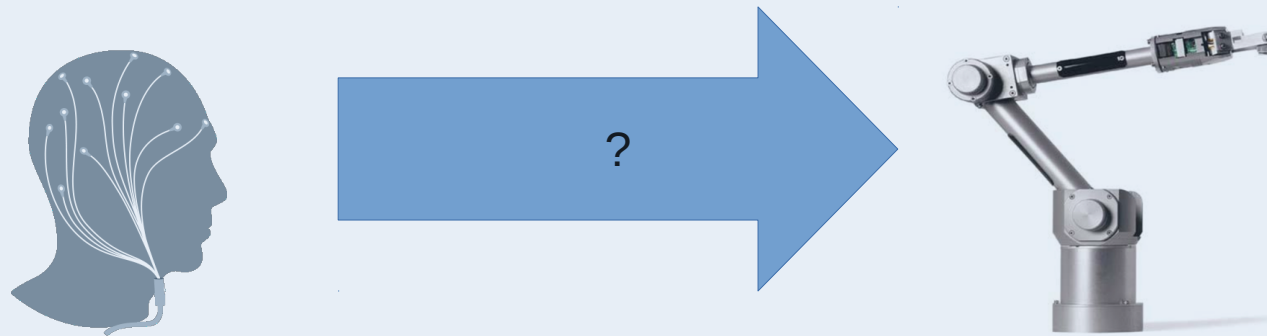
Juri Fedjaev

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Overview

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

Introduction

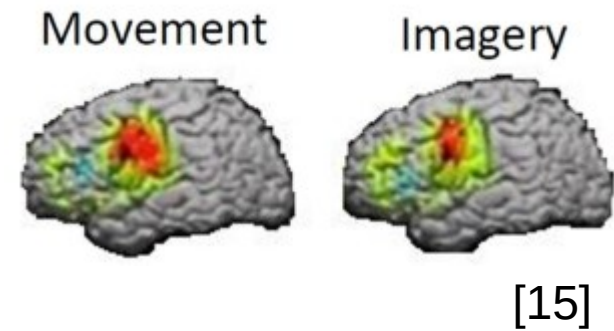


How to control a robot arm using nothing but our mind?

Motor Imagery

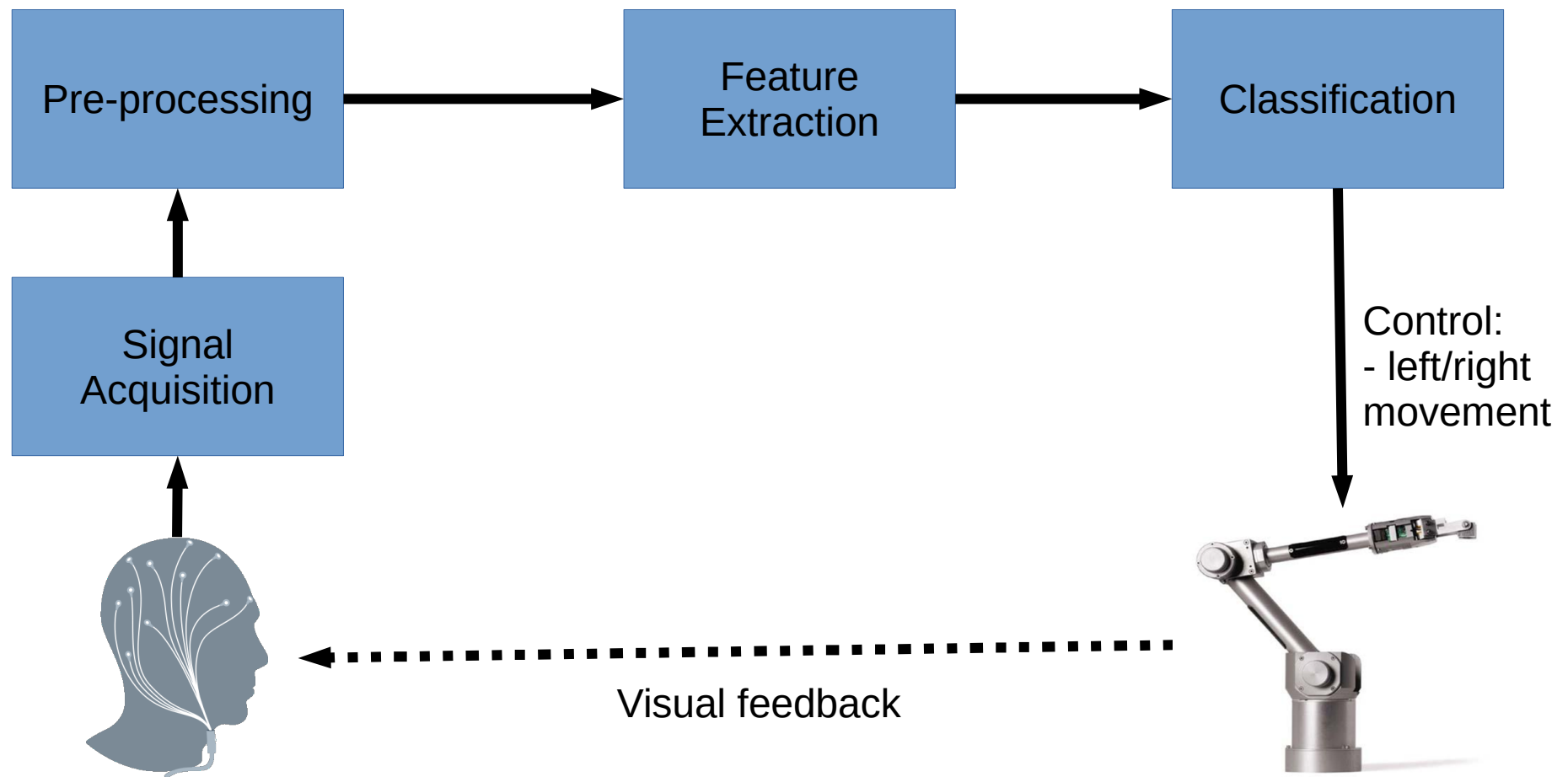
Introduction

- 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)



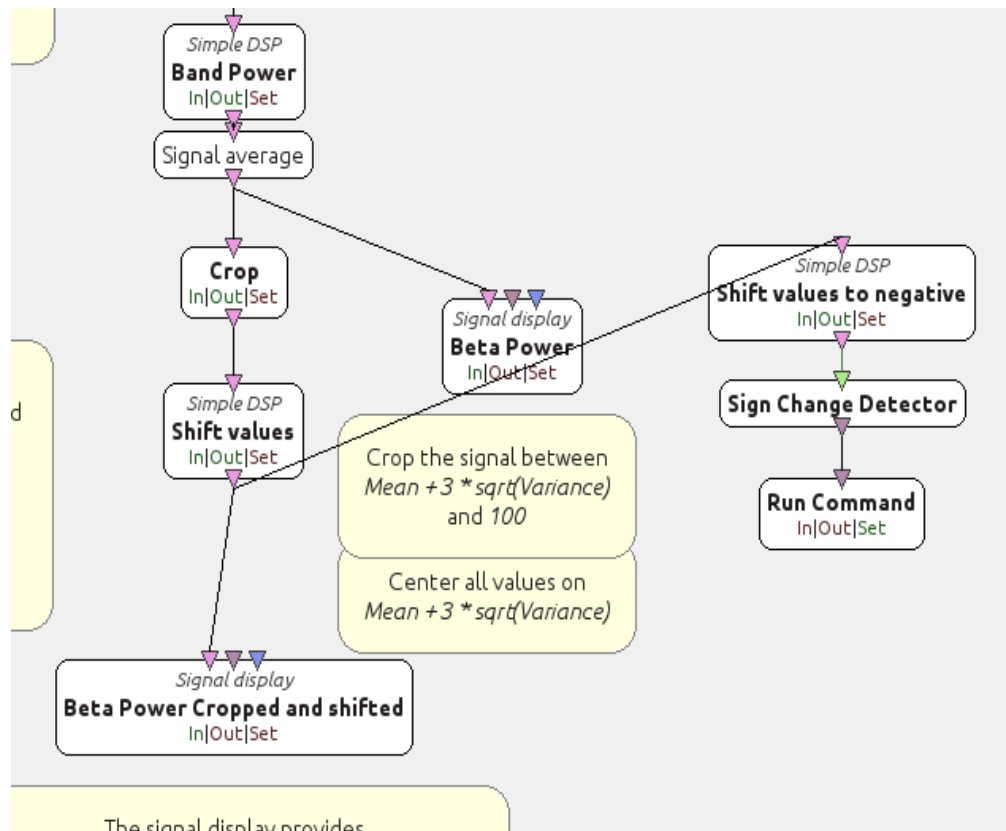
Full system overview

Introduction



First approach – OpenVibe & Emotiv

Experimental Design

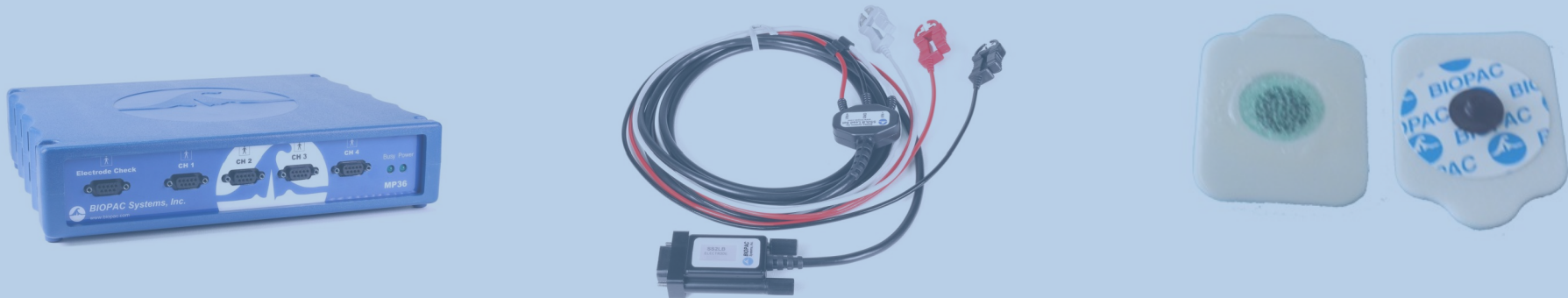


Emotiv EPOC+ headset

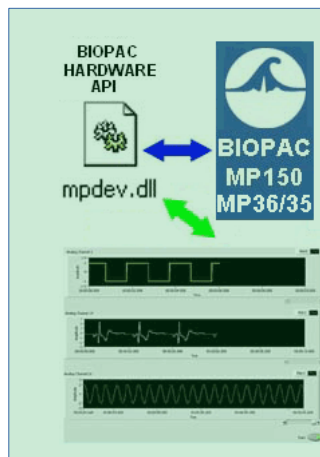
OpenVibe Workflow

Second Approach – Biopac System

Experimental Design



Biopac MP36 Signal Acquisition Device

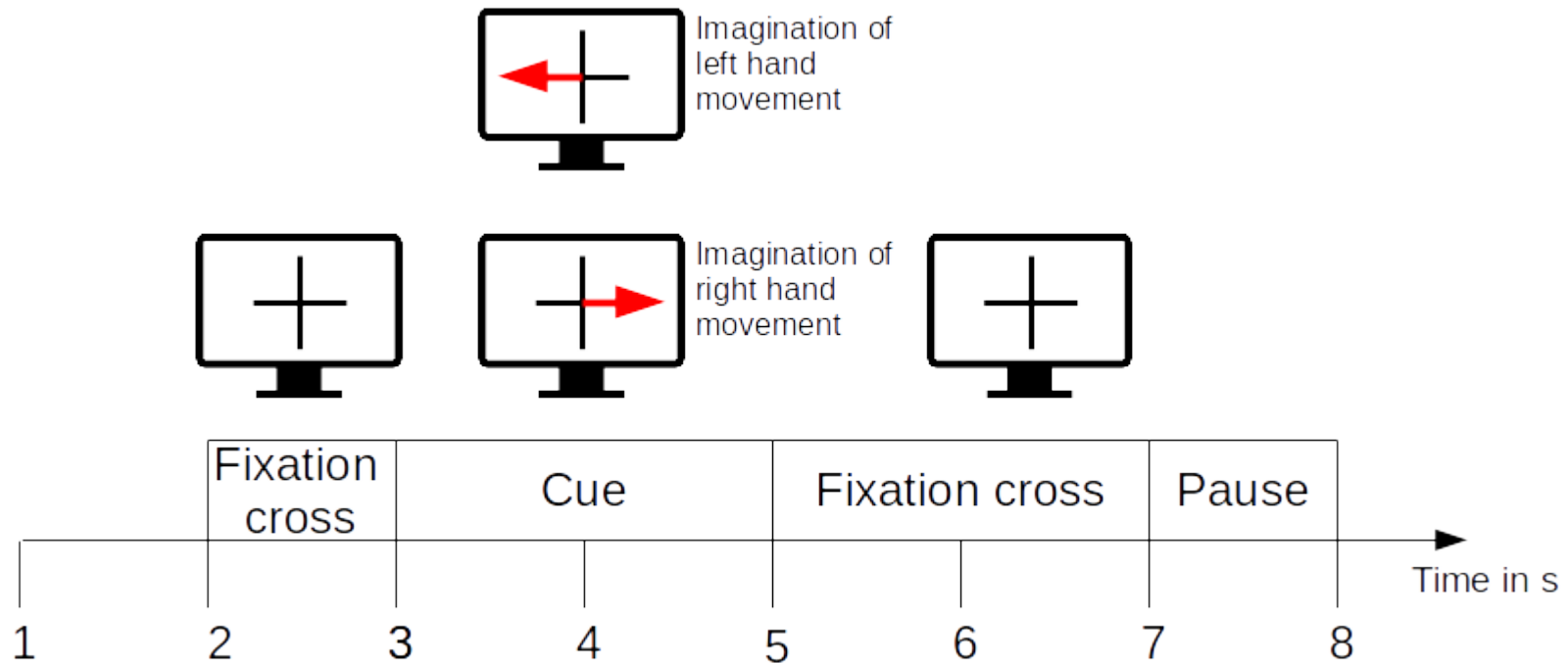


Biopac Hardware API



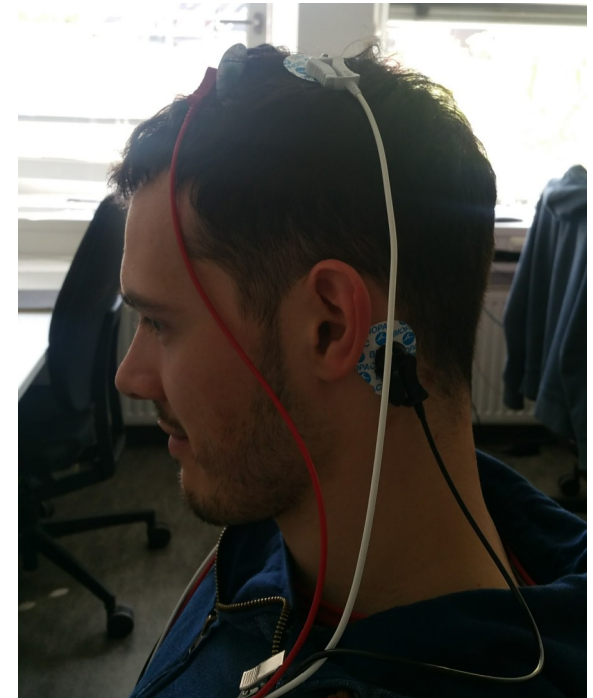
Recording Protocol

Experimental Design



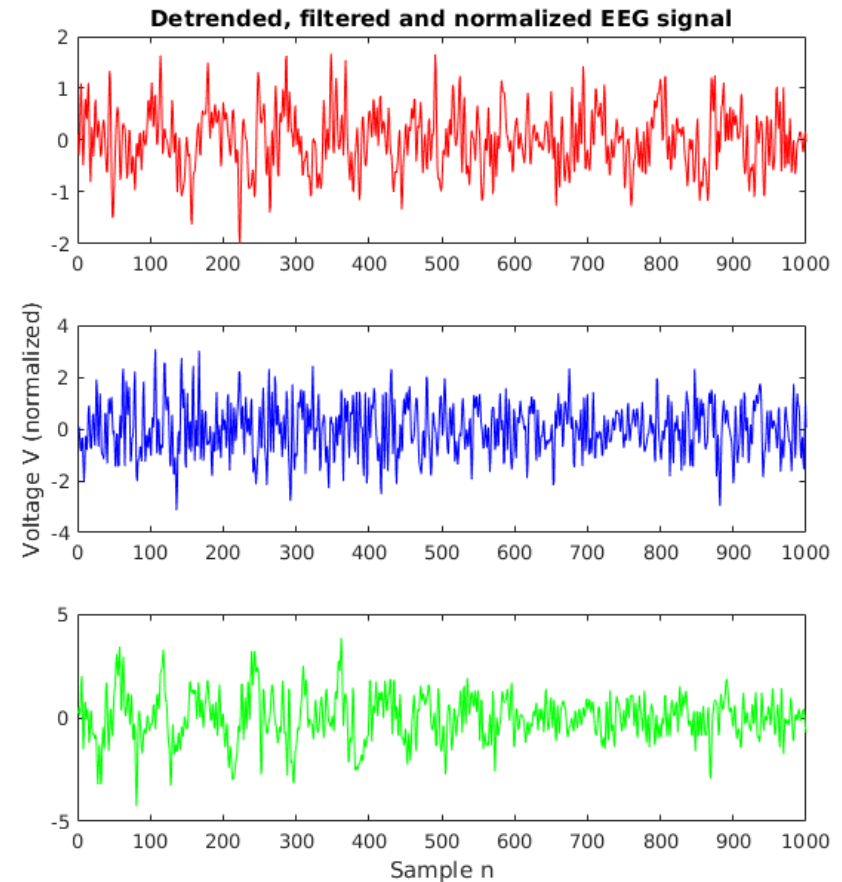
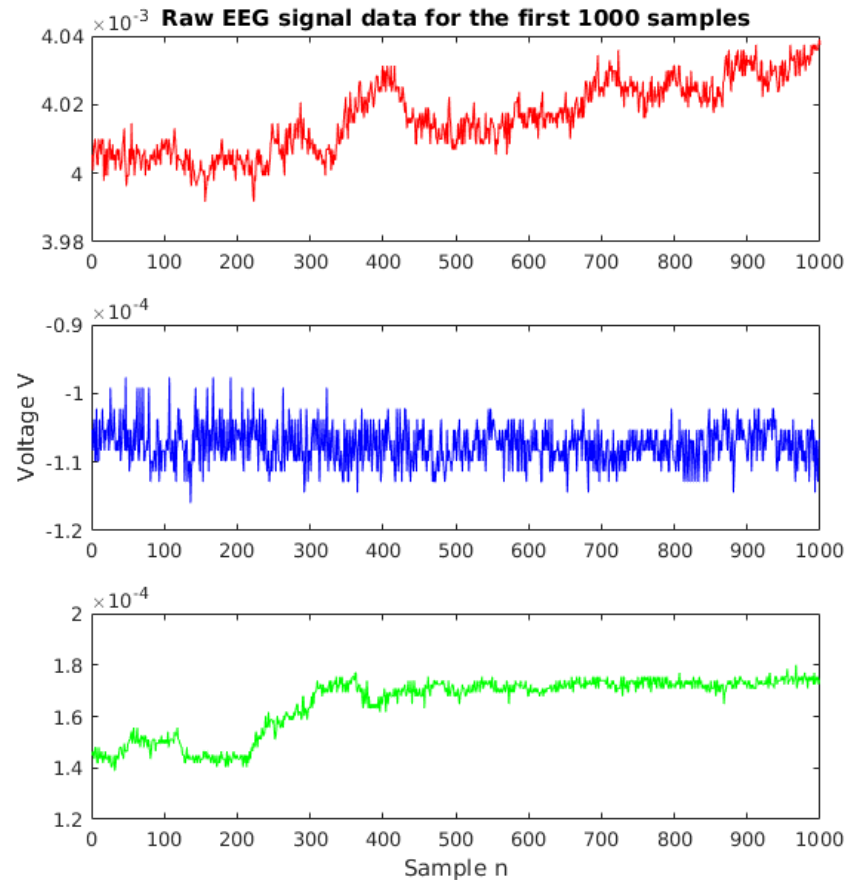
Data Acquisition

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	77
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

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Questions?