

# Single-trial classification of evoked responses to auditory tones using OPM- and SQUID-MEG

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## Introduction

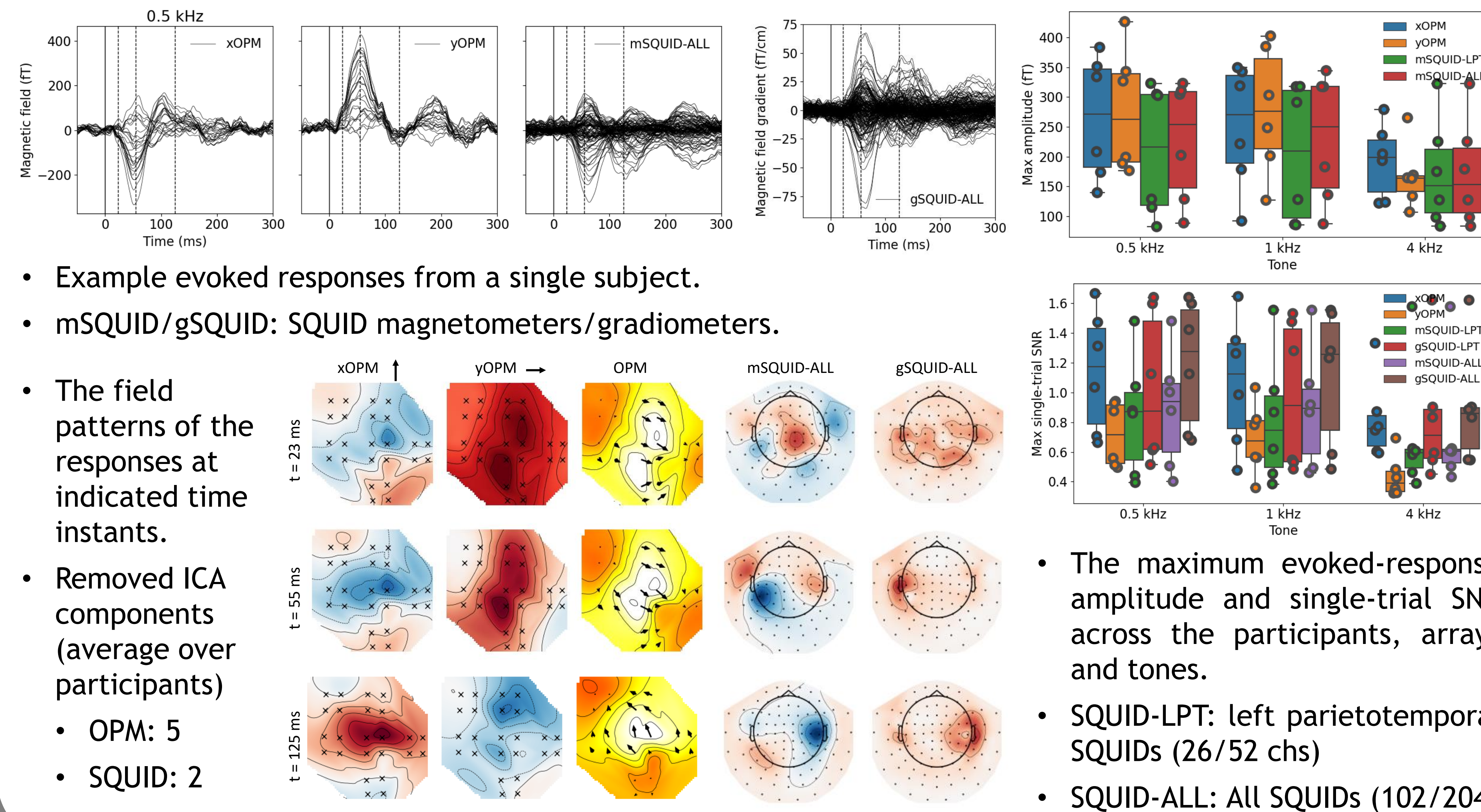
Optically pumped magnetometers (OPMs) are a near-room-temperature alternative to superconducting quantum interference devices (SQUIDs) for magnetoencephalography (MEG). In contrast to SQUIDs, OPMs can be placed in a close proximity to subject's scalp potentially increasing the signal-to-noise ratio and spatial resolution of MEG [1]. However, experimental demonstrations of these benefits are still scarce.

To compare our 24-channel OPM-MEG system [2] to a commercial whole-head SQUID system (MEGIN Oy) in a data-driven way, we quantified their performance in classifying single-trial evoked responses to auditory tones in six participants.

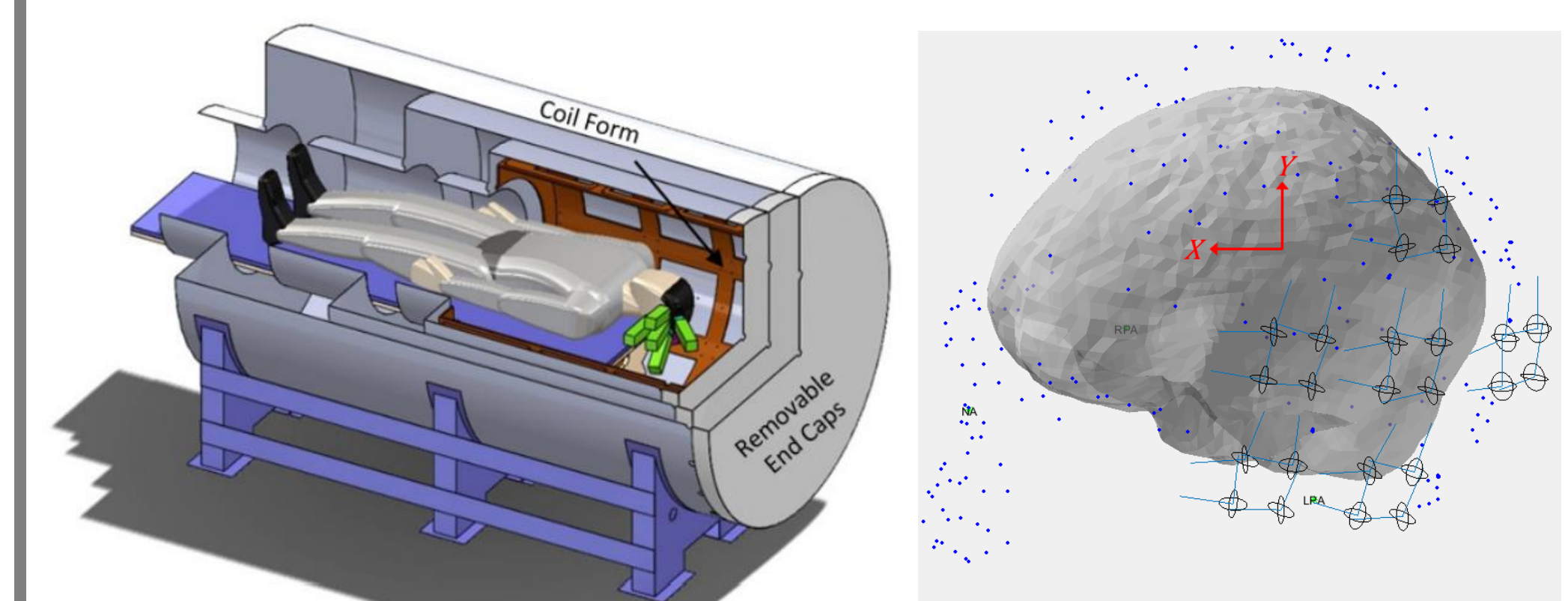
We performed

- Pairwise temporal classification of the responses
  - Linear discriminant analysis (LDA)
- Multiclass classification
  - EEGNet convolutional neural network [3]
  - xDAWN decoding + Riemannian geometry + Logistic regression [4,5]

## Evoked responses

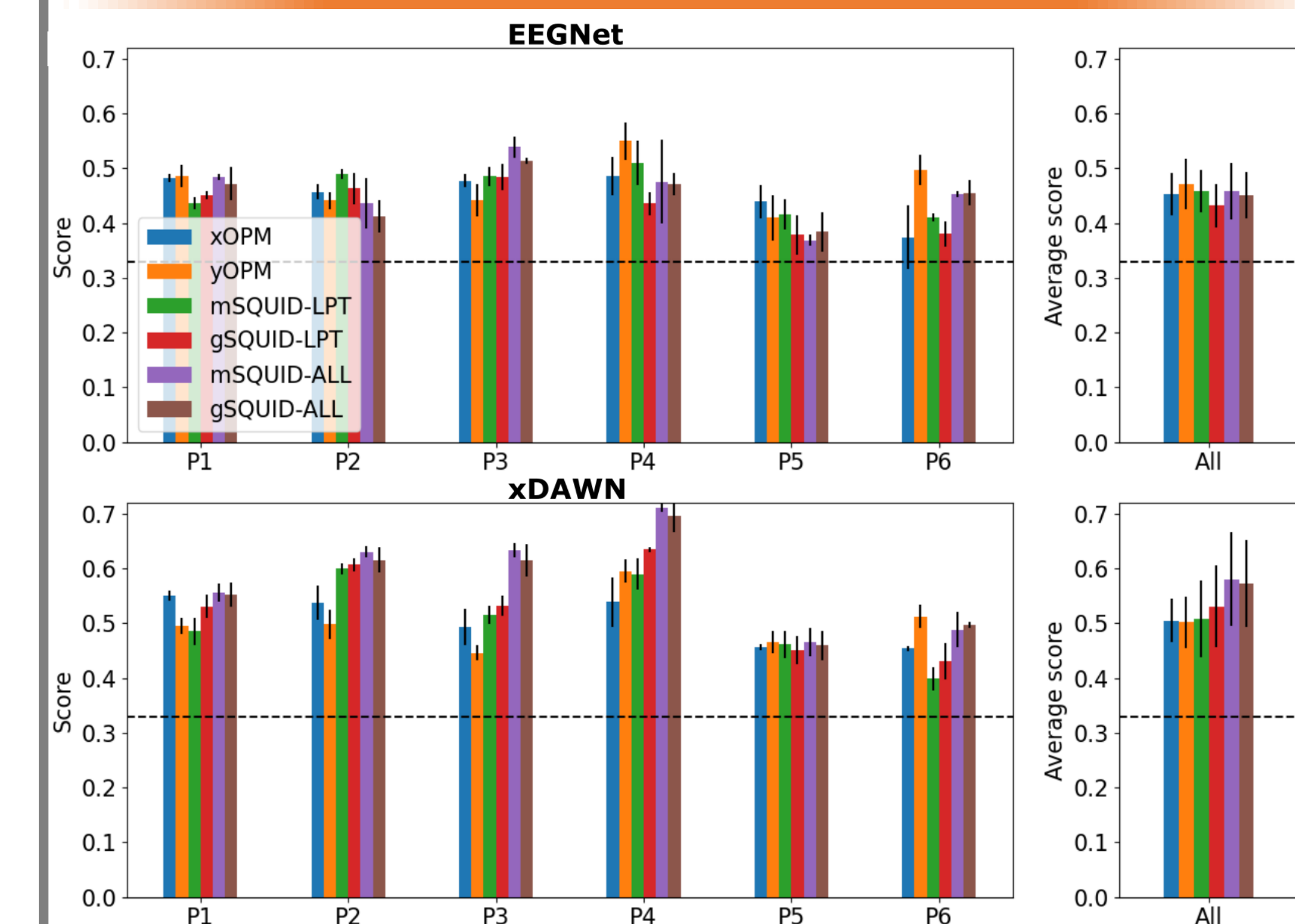


## OPM-MEG system



- 24-channel OPM-MEG system inside a person-sized magnetic shield [2].
- The locations and orientations of xOPM and yOPM with respect to a subject's brain surface.
- xOPM and yOPM are measured sequentially.

## EEGNet and xDAWN



- The classification accuracies of EEGNet and xDAWN spatial filtering across the participants

## Methods

### Stimuli

- Binaural sinusoidal tone pulses (0.5, 1 and 4 kHz)
- Pulse duration: 50 ms, ISI: 650 ms, jitter: 100 ms

### Preprocessing

- SQUID: MaxFilter
- LDA: bandpass filter at 1–90 Hz
- EEGNet: 1–43 Hz; downsample to 128 Hz
- Independent component analysis (ICA) data cleaning
- Classification: Equalize epoch counts for OPM and SQUID

### LDA

- Principal component analysis (PCA) to 99% variance
- 10-fold cross validation (CV), 9:1 training-to-test ratio

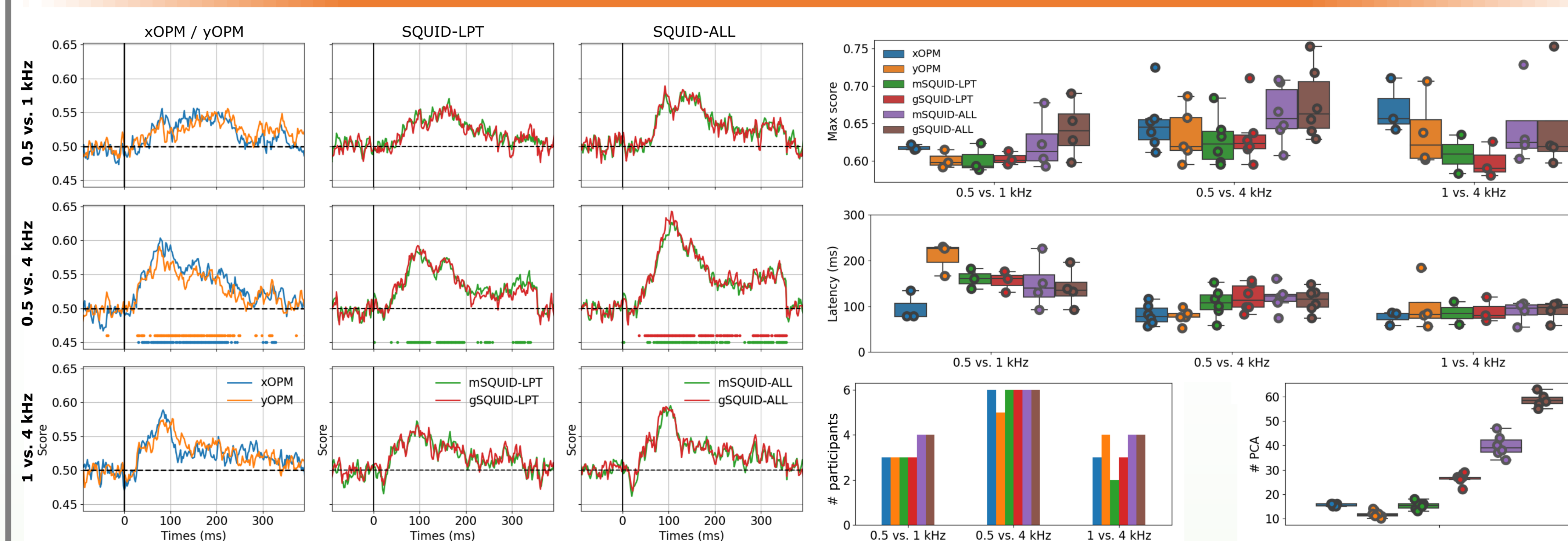
### EEGNet

- $F_1 = 8$ ,  $D = 2$ ,  $F_2 = 16$ ; kernel length: 64 samples; dropout rate: 0.5
- 4-fold CV: training:validation:test split ratio 2:1:1

### xDAWN + Riemannian Geometry + Logistic regression

- 4-fold CV: training:test split ratio 3:1

## LDA classification



- The average classification accuracy of LDA across the participants as a function of time.
- The points underneath the plots show time instances when the accuracy is significantly higher than the chance level ( $p < 0.05$ ).
- Summary of the classifier performance across the participants. The maximum classification accuracy and its latency. Dot = a participant.
- The number of participants that showed a time instant with significant classification accuracy.
- The number of PCA components needed to explain 99% of the data variance.

## Conclusions

- OPMs provided higher classification accuracies than SQUIDs having a similar coverage of the left hemisphere of the participant.
- SQUID sensors covering the whole helmet had classification scores larger than those of OPMs demonstrating the benefits of a whole-head measurement.
- Simultaneous measurement of xOPM and yOPM may yield even higher accuracies.