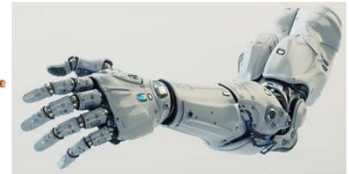
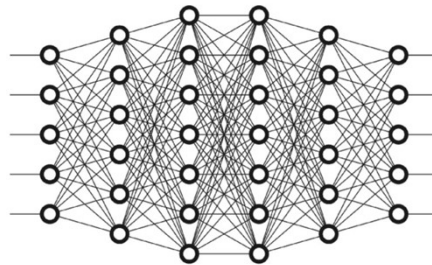
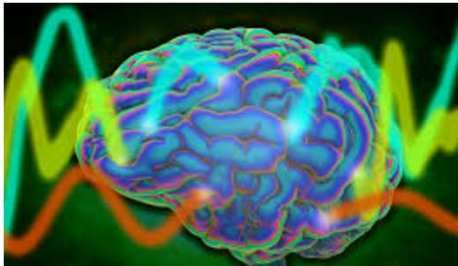


Hypernetwork for self-calibrating brain computer interface

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Department of Computer Science, University of Chicago
August 16, 2024

Introduction: Brain Computer Interface



Ongoing brain
Activity

Real-time data
streaming



Brain
computer
interface

Processing



Human interpretable
concepts

Visual Stimuli



Our Reconstructed Video



Deo et.al 2023 Translating deep learning to neuroprosthetic control
Willett et.al 2023 A high-performance speech neuroprosthesis

Introduction: BCI calibration for new user



Differences of brain activity and hardware between individuals

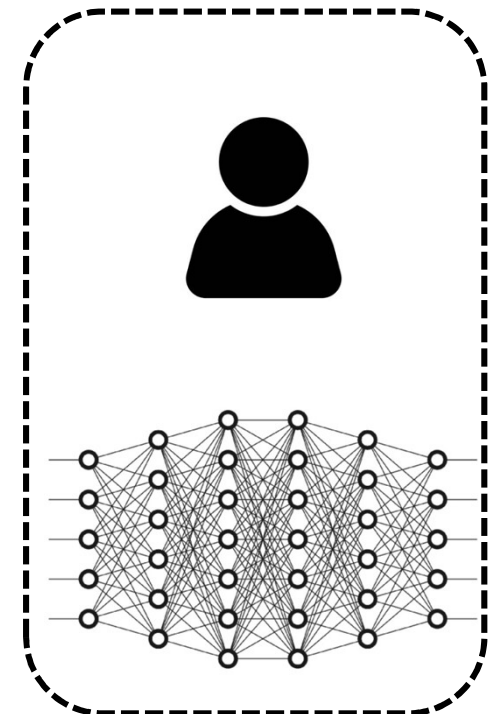
Existing users



Model calibration



New user



Challenge in *supervised* calibration

Supervised calibration requires **labeled** data. Accessing data labels from new user is impractical for real-time BCI applications.

Unlabeled brain data



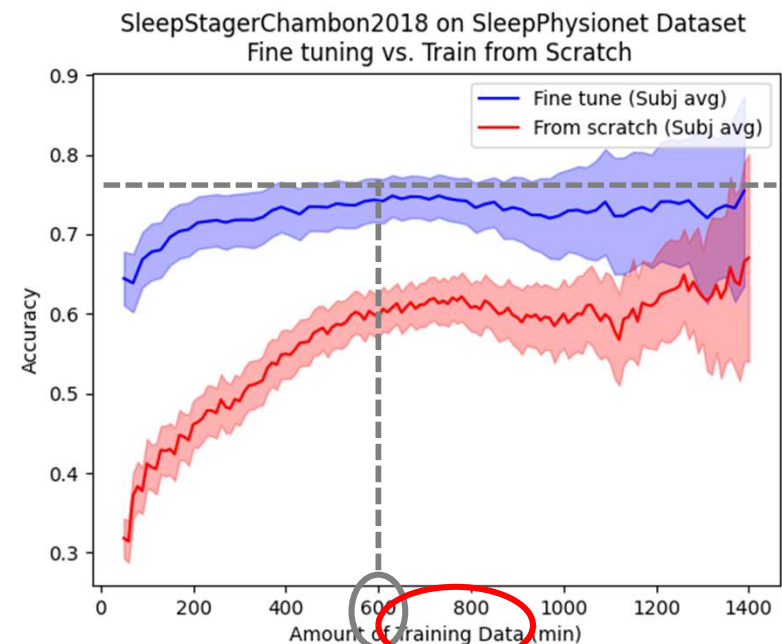
Human labeling



Labeled data



Sleep stage classification

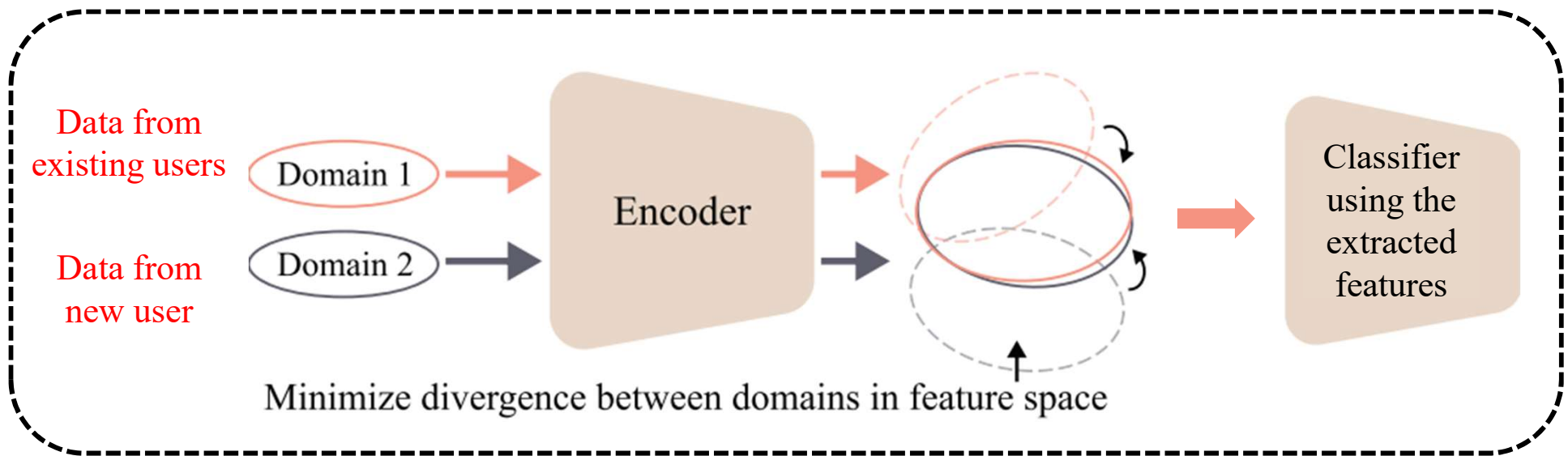


labeled

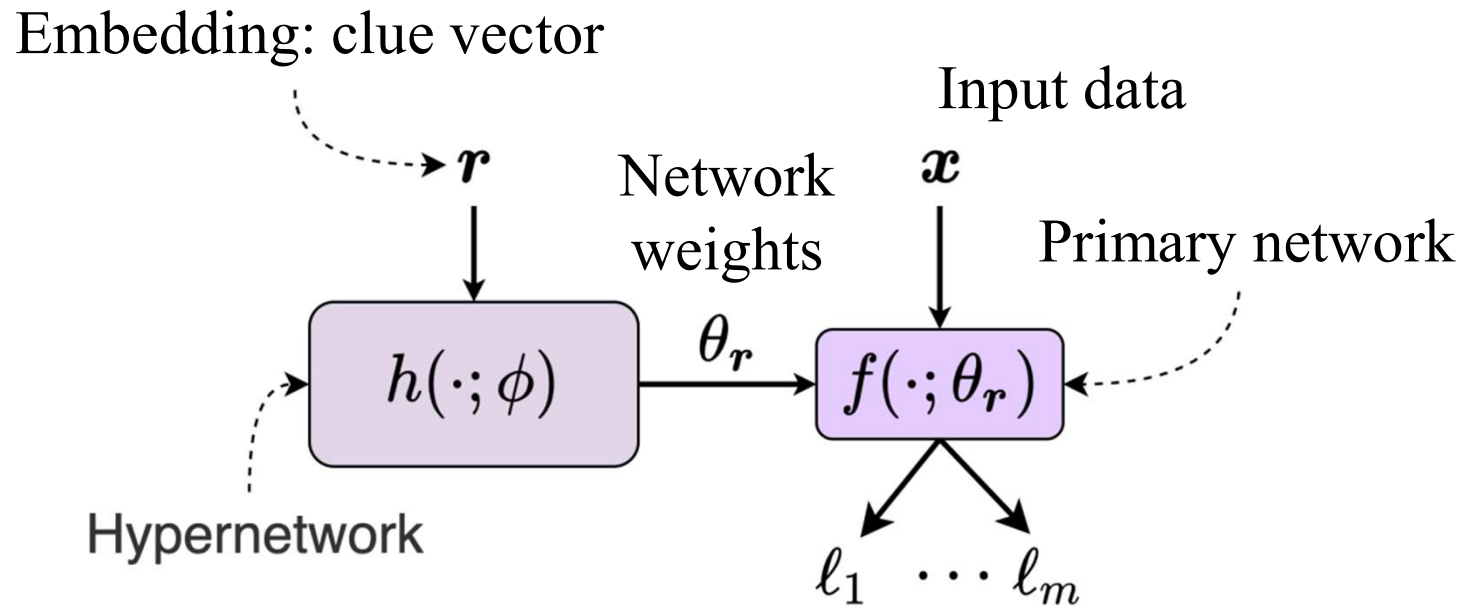
Challenge in *unsupervised* calibration

Unsupervised calibration typically involves feature space alignment, which usually requires **brain data from existing users**.

May be challenging to access due to privacy or sheer quantity.

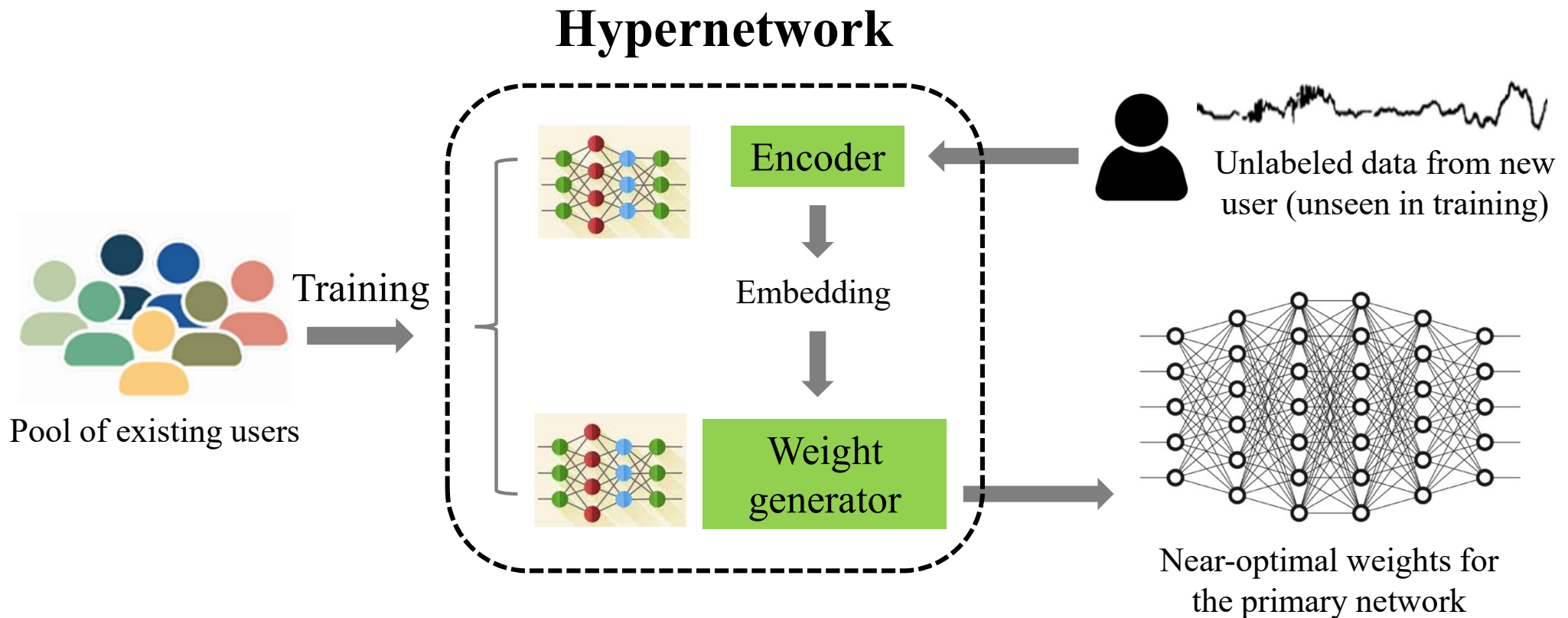


Method: Hypernetwork

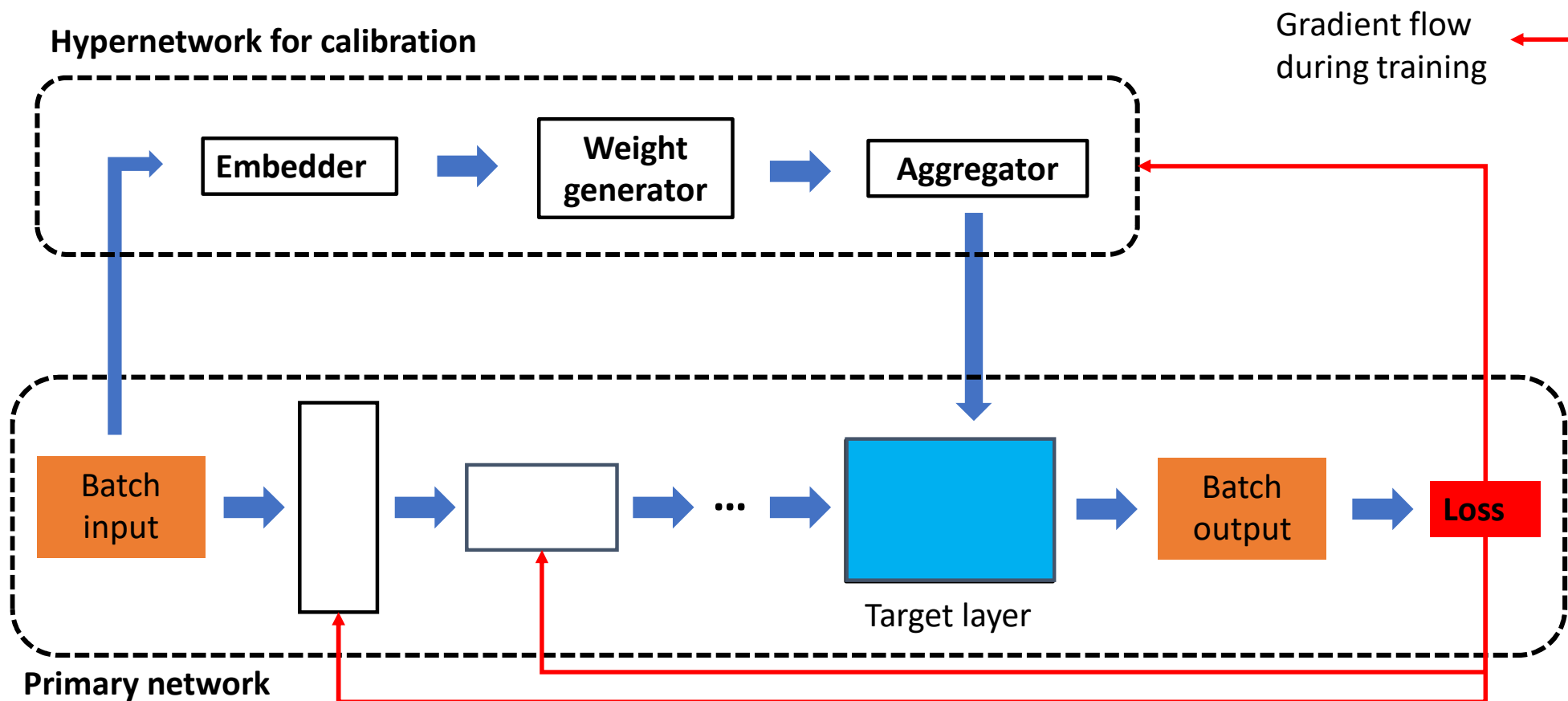


If given proper clue, hypernetworks can **generate near-optimal** weights, which are usually acquired through iterative, time-consuming optimization process.

Method: Hypernetwork for unsupervised, real-time calibration

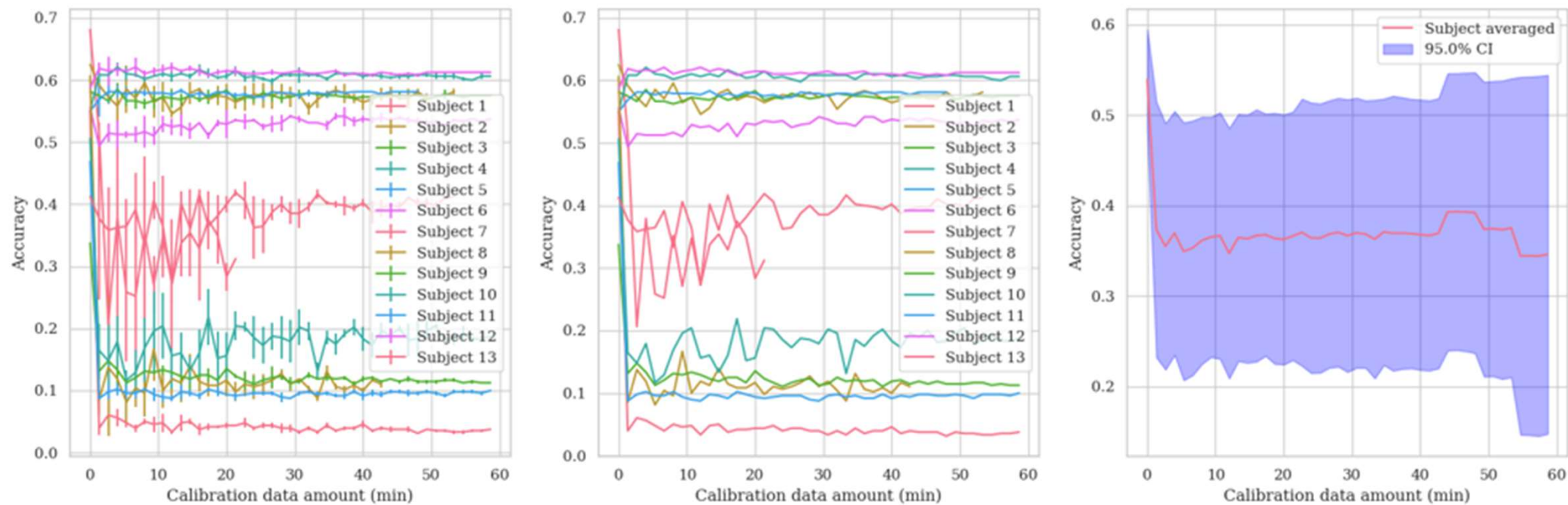


Method: HypernetBCI architecture



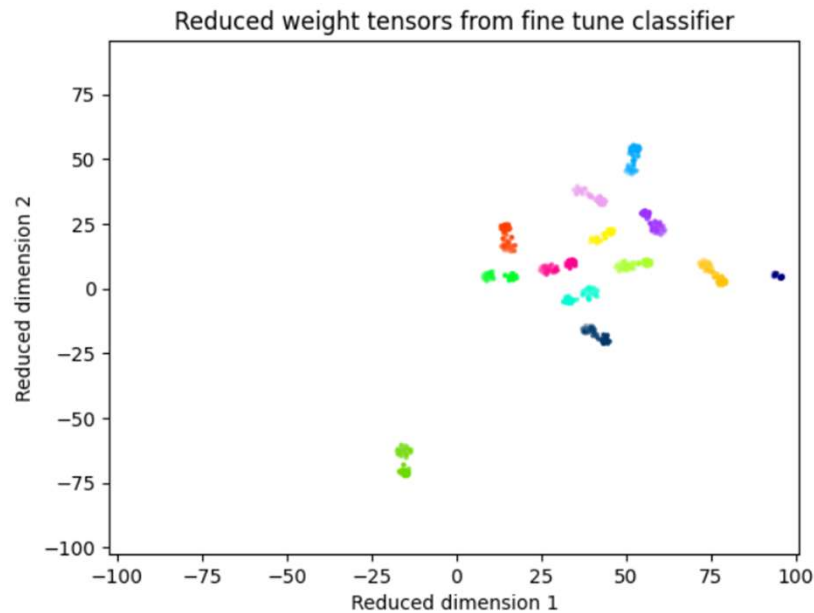
Result: test accuracy vs. amount of calibration data

HYPERShallowFBCSPNet on Schirrmeister2017 Dataset
, Calibrate model for each subject (cross subject calibration), 3 reps each point

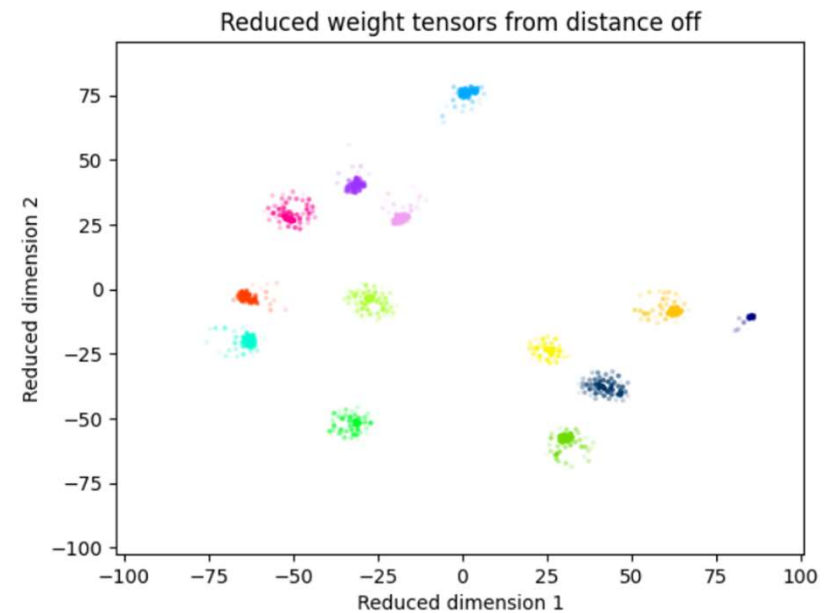


Result: visualizing weights to understand calibration efficacy

Weight tensors produced by fine tuning



Weight tensors produced by hypernetBCI



Contributions



First to apply hypernetwork for **unsupervised domain adaptation**



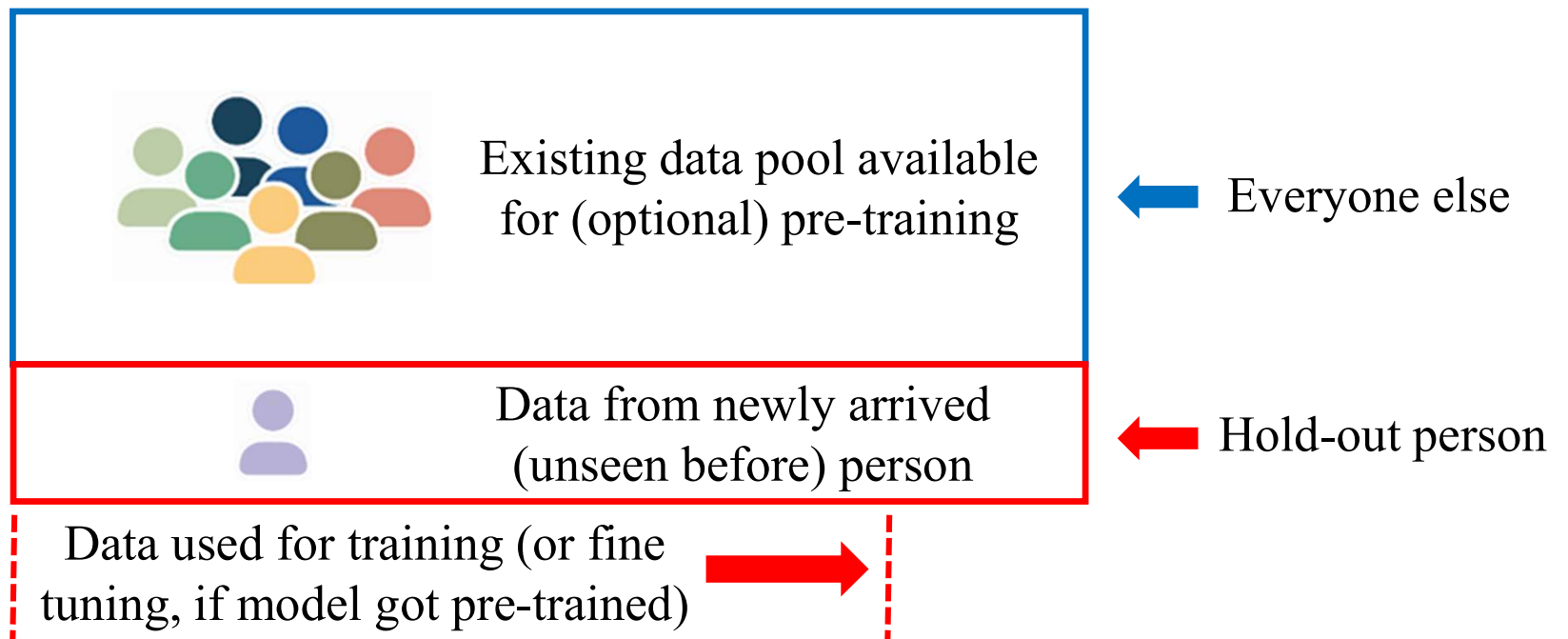
Calibration takes one forward pass, repeatable and good for **real-time** adjustment



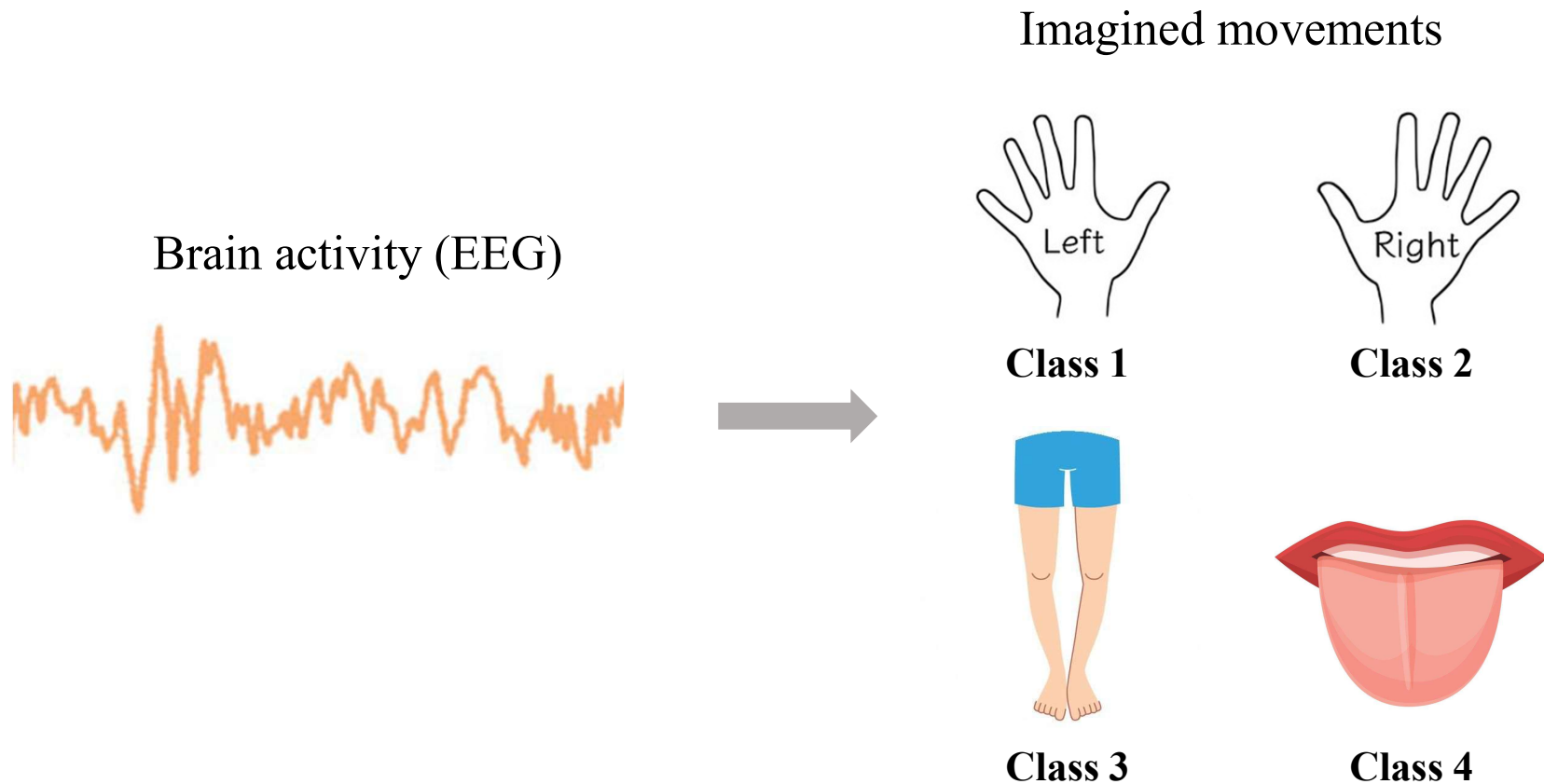
Data efficient, good for edge deployment (brain implants)

Appendix I: Experiment setup

Dataset

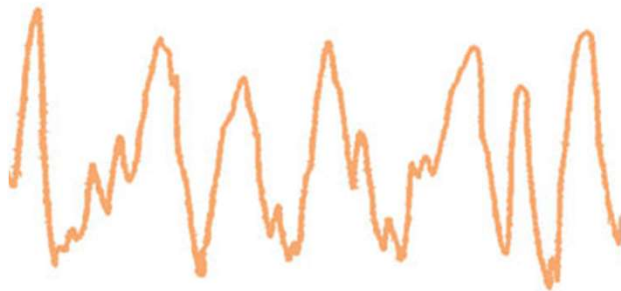


Appendix II: motor imagery classification dataset

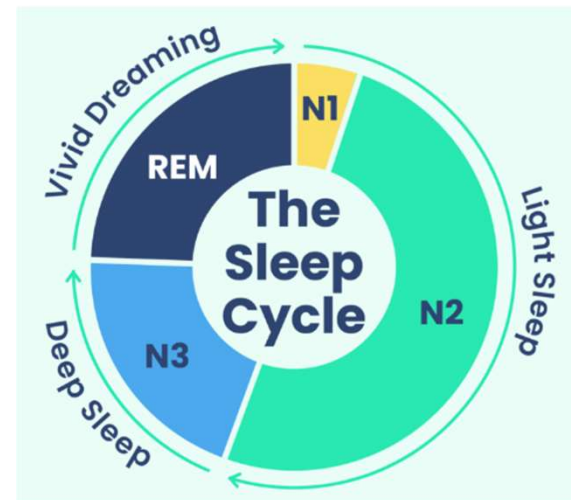


Appendix II: sleep stage classification dataset

Brain activity (EEG)



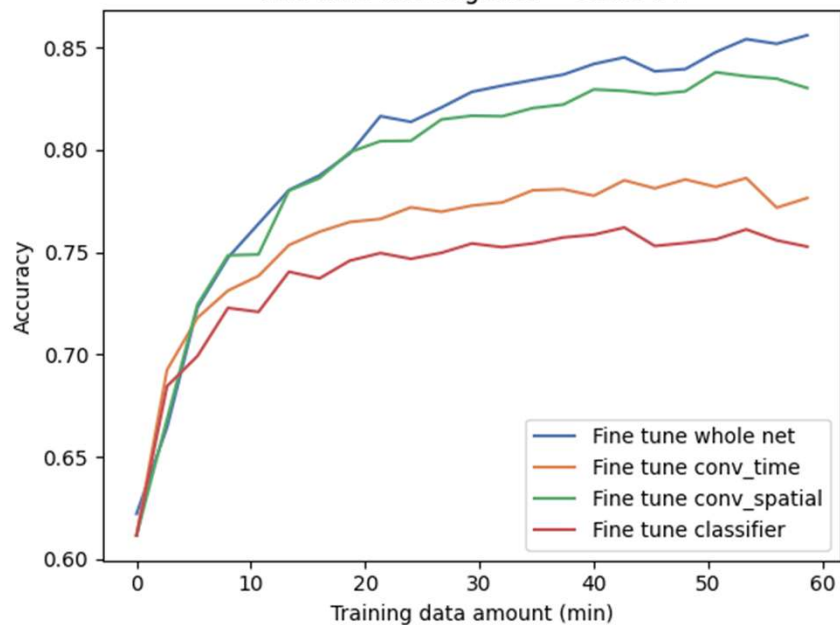
Sleep stages (or awake)



Appendix III: fine tuning part(s) of the model while freezing others

Motor Imagery Classification

Fine tune part(s) of ShallowFBCSPNet for Schirrmeister2017 dataset
fine tune learning rate = 6.50×10^{-4}



Sleep Stage Classification

Fine tune part(s) of SleepStagerChambon2018 for SleepPhysionet dataset
fine tune learning rate = 1.00×10^{-3}

