

Using Machine Learning to classify Timbre based on EEG Data

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Objective



Explore how different timbral characteristics are encoded in EEG data



Current goal: Determine if a machine can discriminate between EEG data recorded when a participant is listening to different musical instruments



Purpose: For use in development of auditory prosthesis such as BCI-based hearing aids

Current music listening experience for hearing aid users less than ideal

Methods

10 participants



Age (SD)	Gender
27.4 (11.05)	7 Female 3 Male

Experiment

0.5-second A3 (F0: 220 Hz) tones of 4 different instrument timbres: Piano, Trombone, String and Clarinet, and Pure tone



Each instrument presented a total of 200 times

Data recording and pre-processing

EEG recorded from 70 electrodes

5 identical tones were presented in sets, and participants reported the instrument they heard after each set

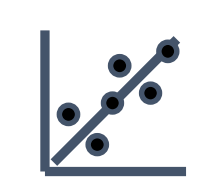


High pass filter of 0.1Hz

Artifact removal using ICA

Epochs segmented from -1s to 2s

Classification

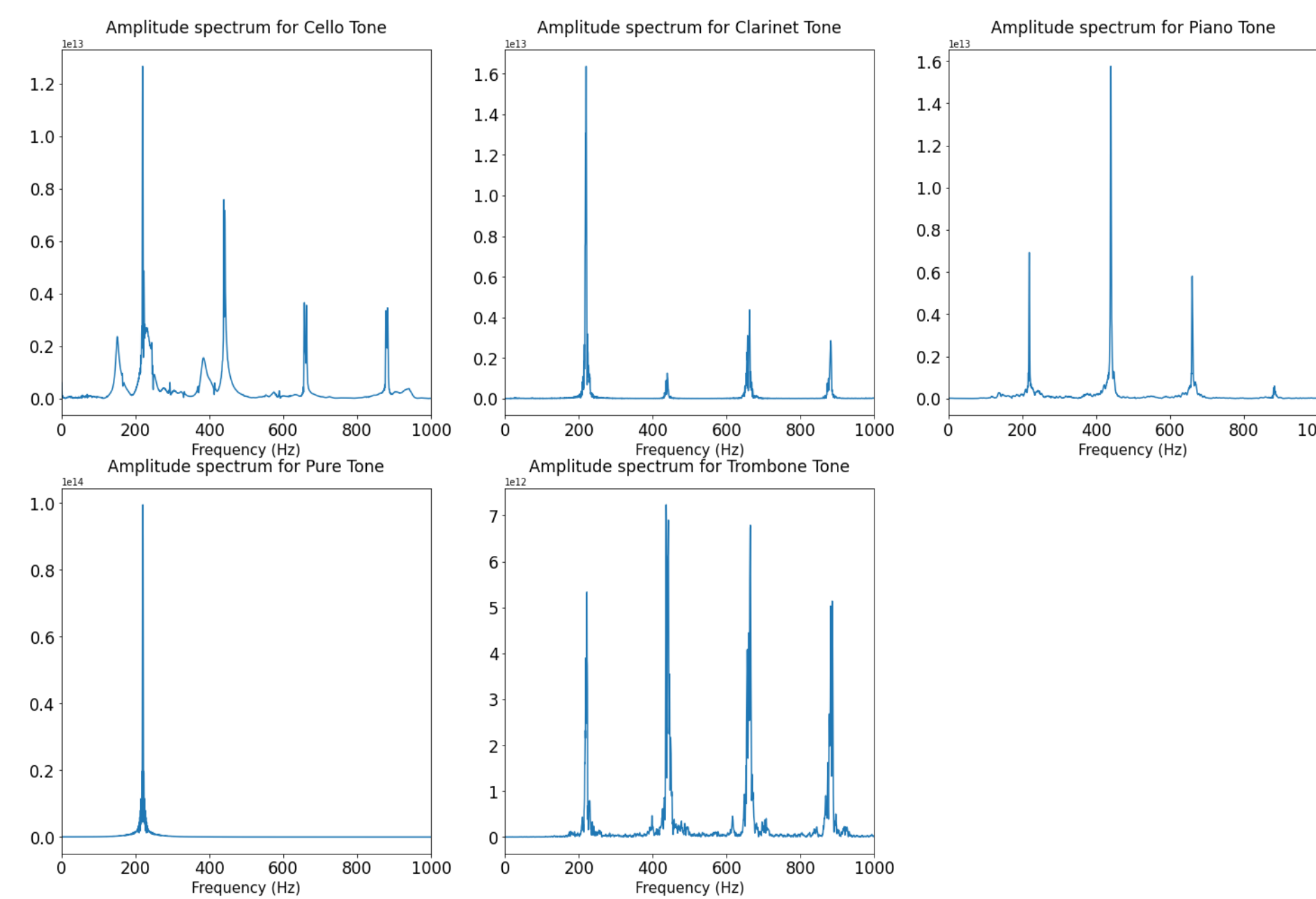


Tested three classifiers: LDA, GB, k-NN

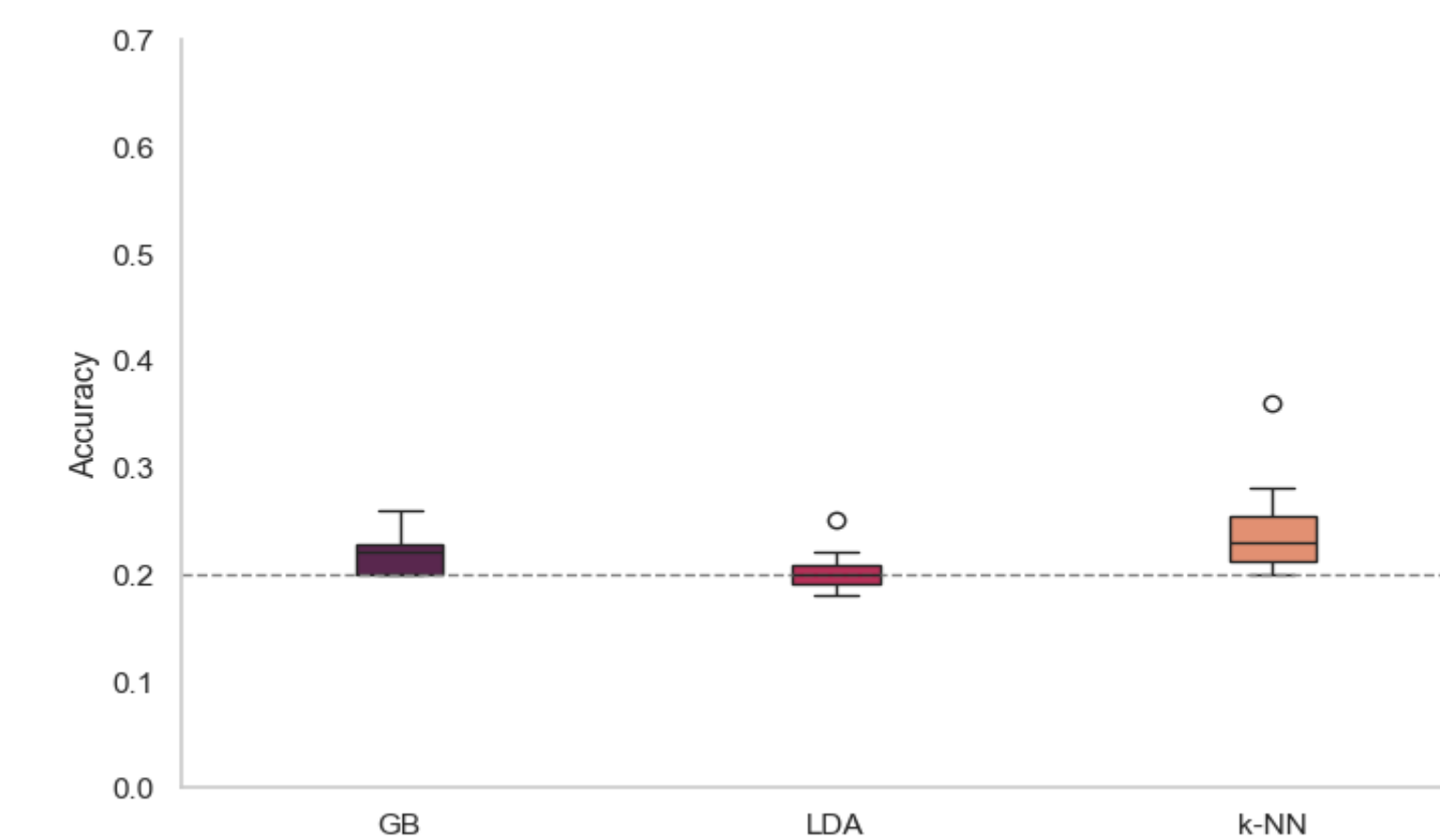
Funding



Results: Harmonics

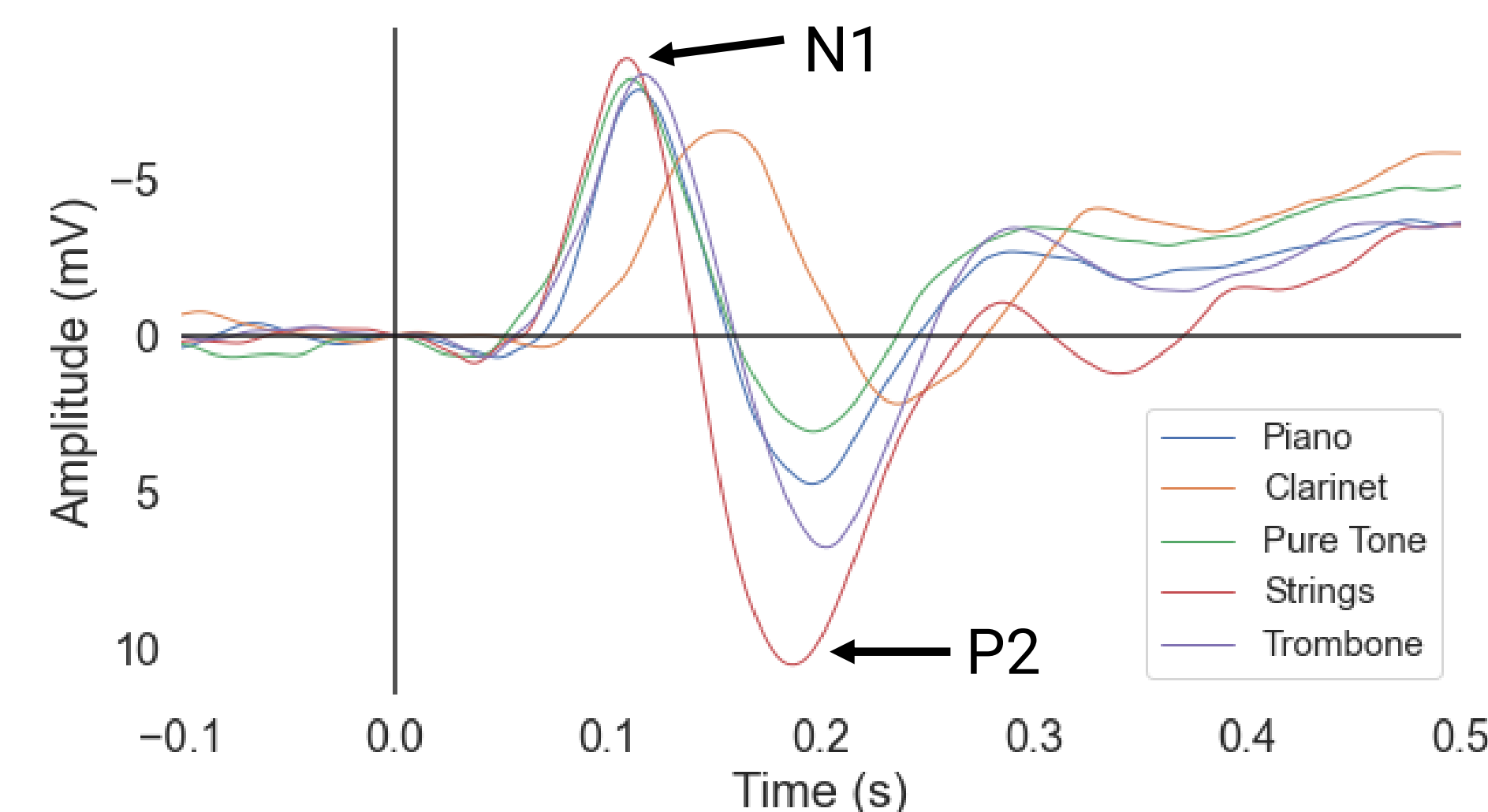


Within-Subject Accuracy using Harmonic Frequency-based Features

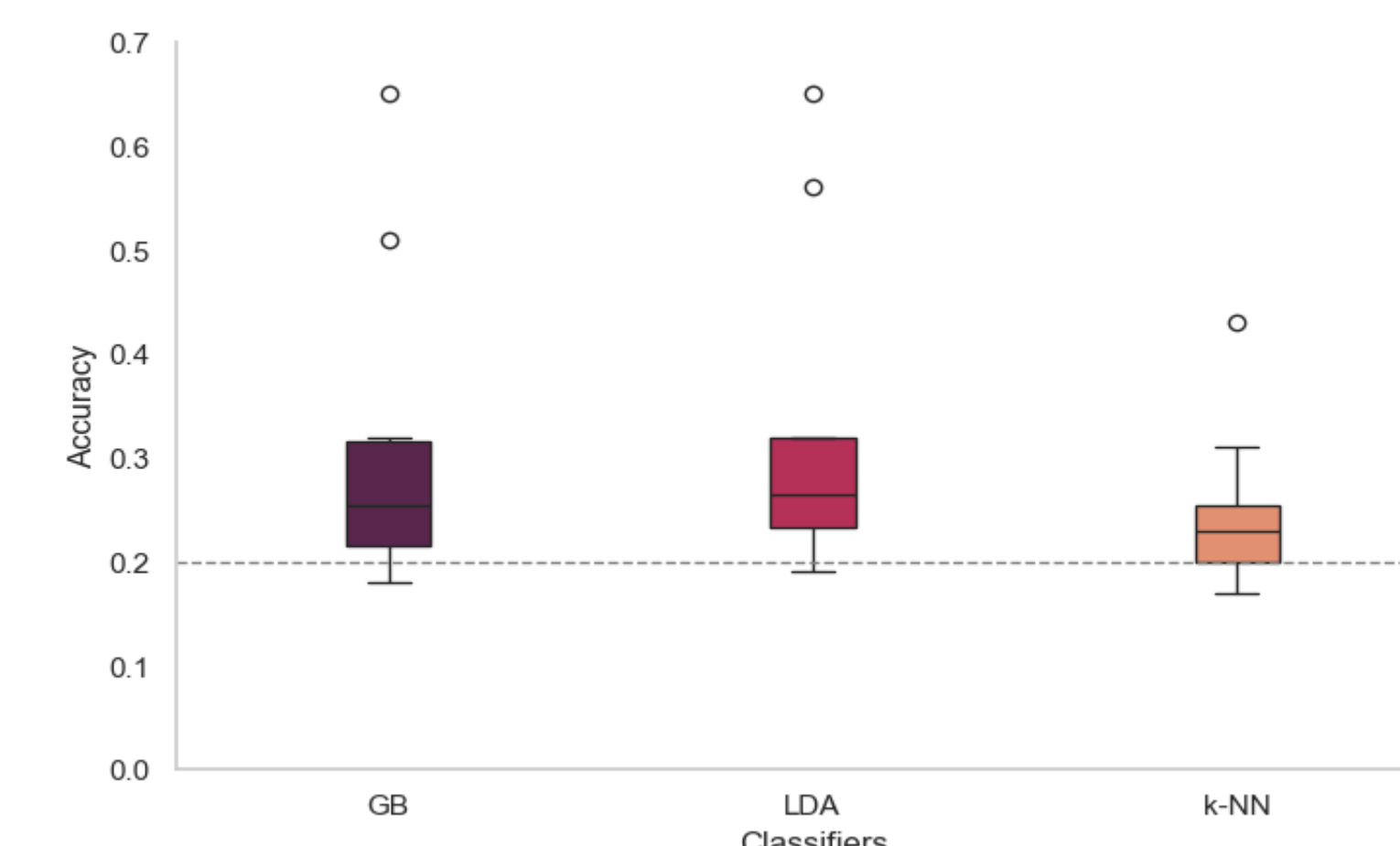


Results: ERP-N1/P2

Grand Average ERPs across Participants

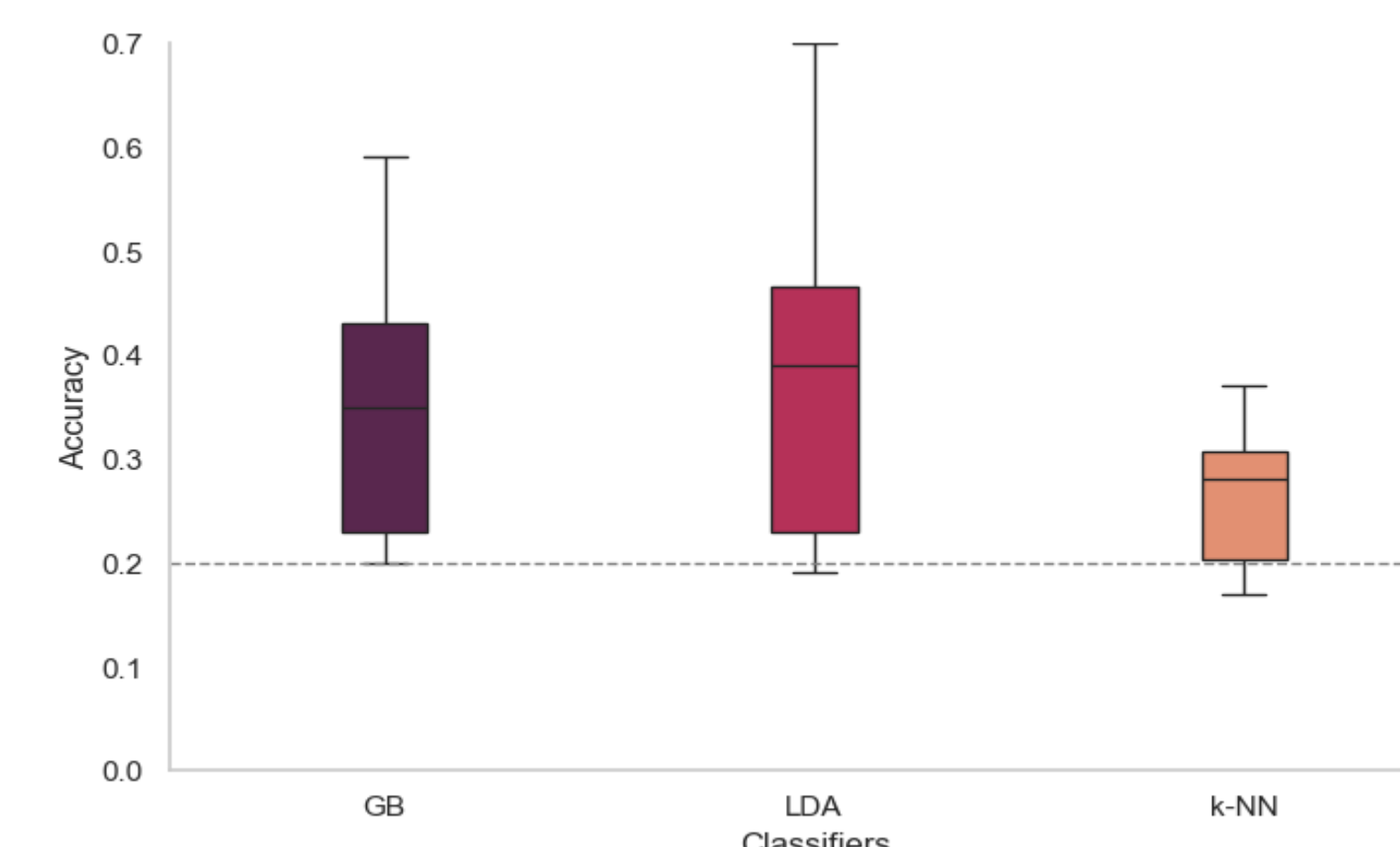


Within-Subject Accuracy using ERP-based Features



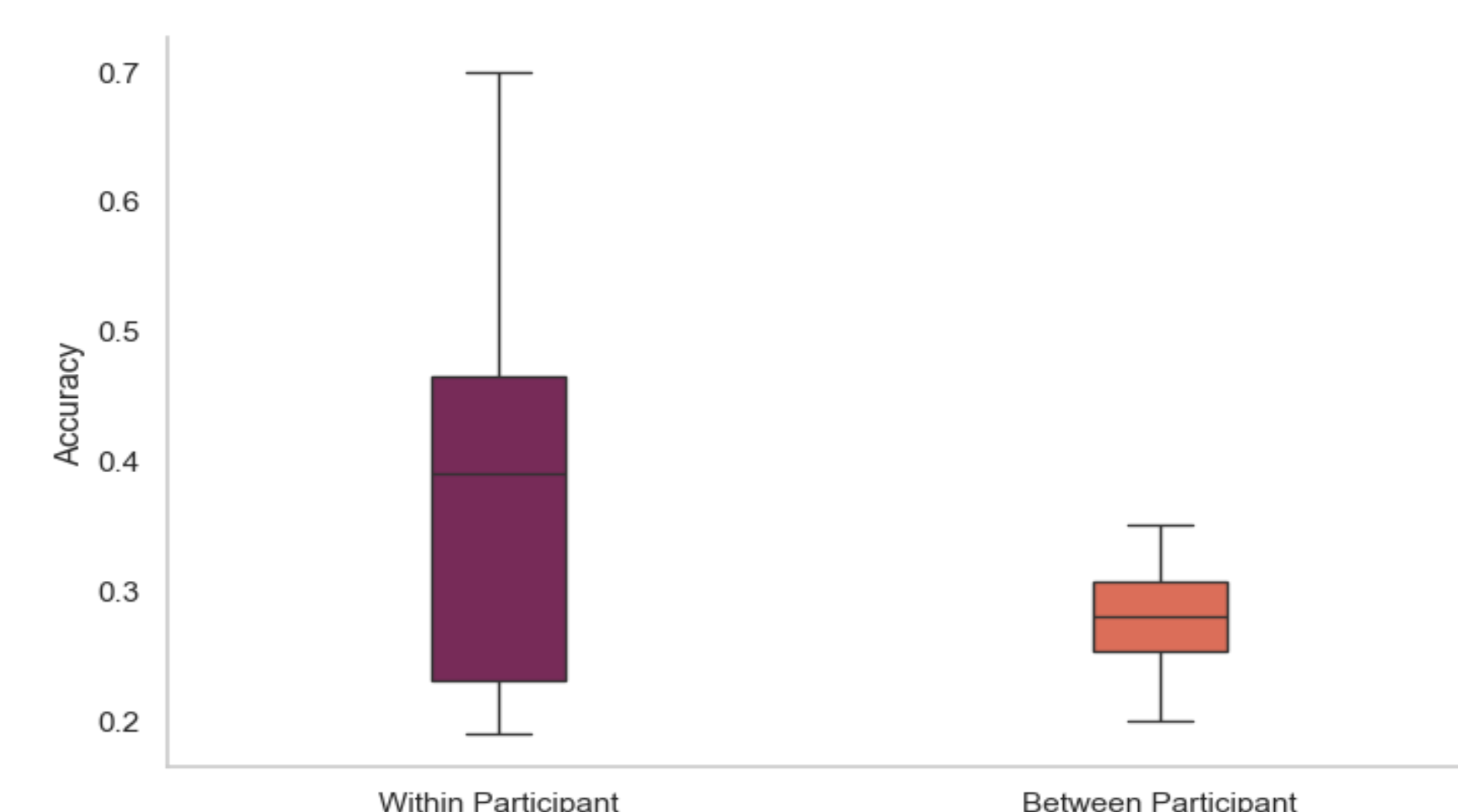
Results: Raw EEG

Within-Subject Accuracy using Raw EEG

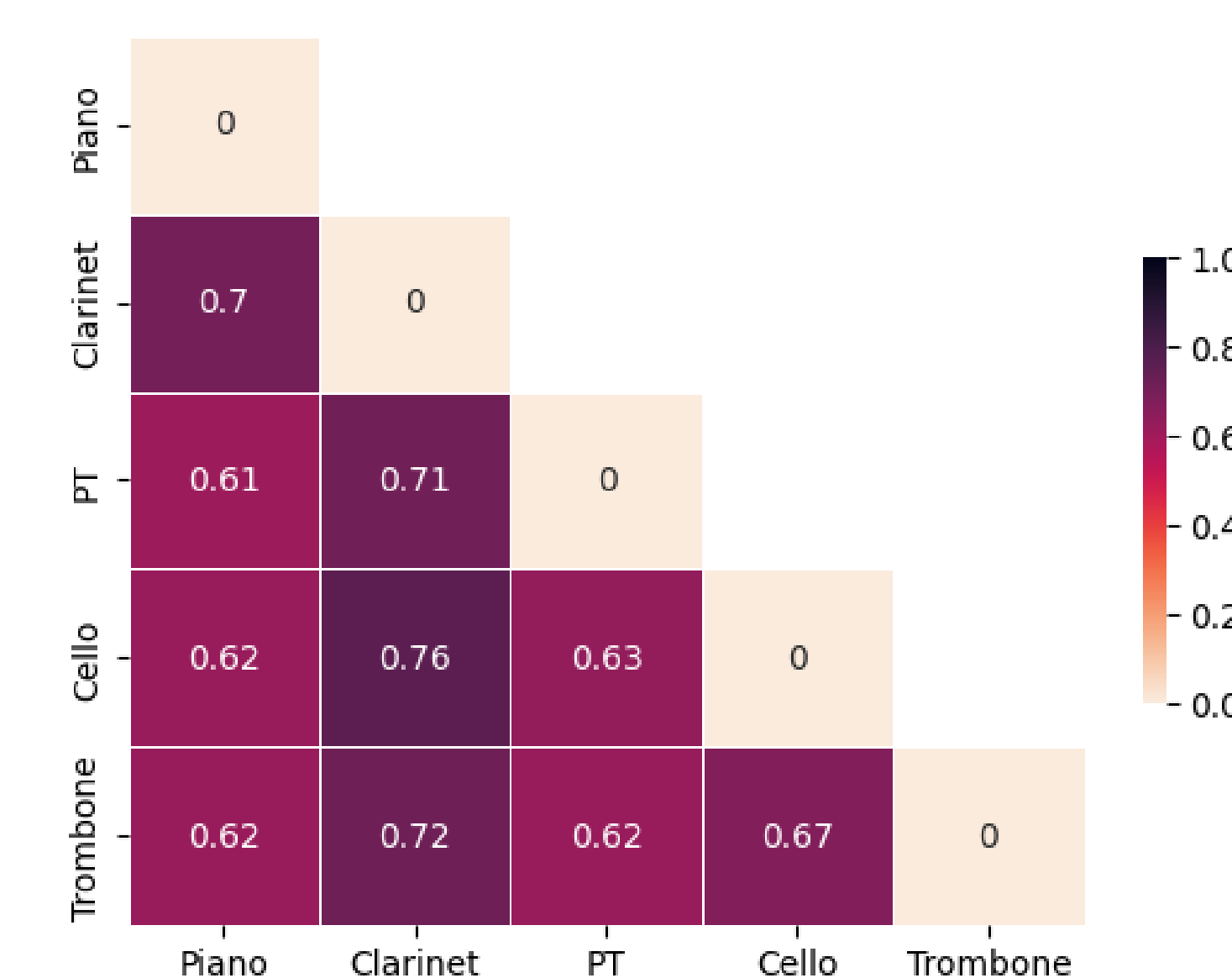


Results: Within-vs Between Participant

Within vs Between-Participant Accuracy using Raw EEG



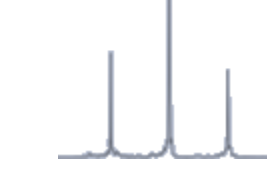
Results: Pairwise Classification



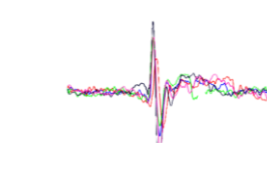
Methods

Within-Subject models trained on following features individually:

Spectral information around **harmonic frequencies** of audio tone:



Performed FFT on EEG signal
Extracted power around frequencies **27.5Hz, 55Hz, 110Hz, 220Hz, 440Hz, 880Hz**



ERP – N1/P2:

Used mean and peak amplitude and latency of N1 and P2



Raw EEG

Raw EEG input without any feature extraction

PCA used for feature reduction

Between-Participant classification performed with **Raw EEG**

Discussion

Classifier performed **above chance**

Features related to **harmonic frequencies** did not seem to contribute to classifier performance as much compared to **raw EEG** and **ERP-Based features**

Performance of **within-participant model** > **between-participant model**

More complex models could possibly lead to higher accuracy rates

Pairwise classification indicates **Clarinet** and **Cello** are the most easily discriminated

