

# Feature Reconstruction Based Channel Selection For Emotion Recognition Using EEG

---

James Ronald Msonda  
Zhimin He, Ph. D.  
Chuan Lu, Ph. D.

# Introduction

---

- Emotions are natural responses to circumstances (internal and/or external events).
- Human emotions have been predicted by observing parameters like facial expression.
- There is interest in the use of Electroencephalogram (EEG)



# The Problem

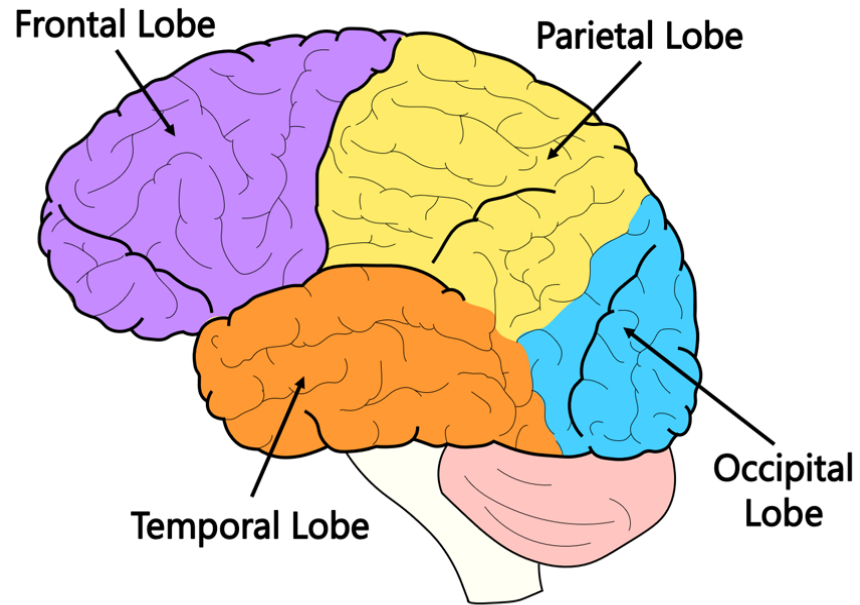
---

- Established EEG systems have up to 256 electrodes but are very expensive.
- Wearable EEG devices come in different configurations
- The question is: **What is the minimum number of electrodes and where do we place them on the scalp for emotion recognition?**

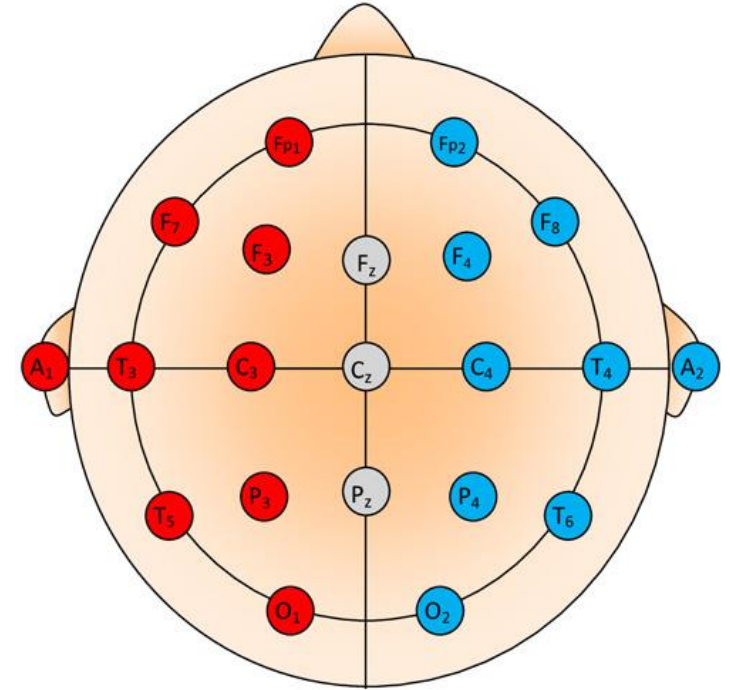


# International 10/20 System

- Standardised method to describe EEG electrode location.
- Based on percentage of left to right or front to back distance

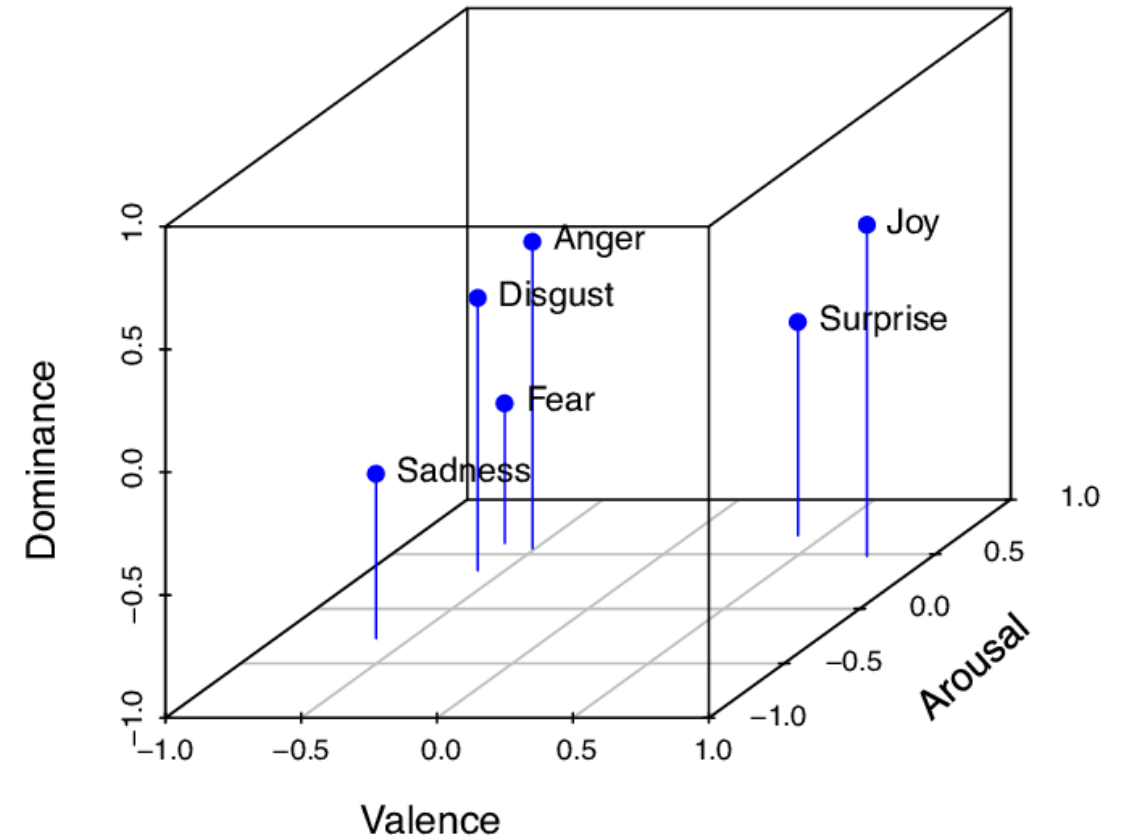


- F - Frontal lobe
- T - Temporal lobe
- C - Central region
- P - Parietal lobe
- O - Occipital lobe



# Data Description

- **DEAP**: 32 EEG channels
- **DREAMER**: 14 channels
- **AMIGOS**: 14 channels
- Self-Assessment Manikin (SAM) for self rating of emotions.



3D Representation of the SAM Scale

Image Credit: <https://www.mdpi.com/2073-8994/12/1/21>

# Our Approach

---

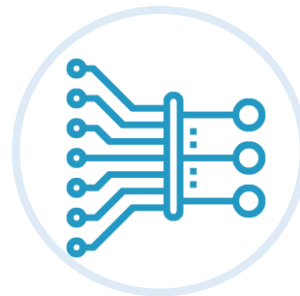
- Goal is to identify suitable features for emotion recognition and thereafter, **reconstruct the selected features** using fewer channels/electrodes.
- It is a five-stage pipeline.



SIGNAL  
PREPROCESSING



FEATURE  
GENERATION



FEATURE  
SELECTION



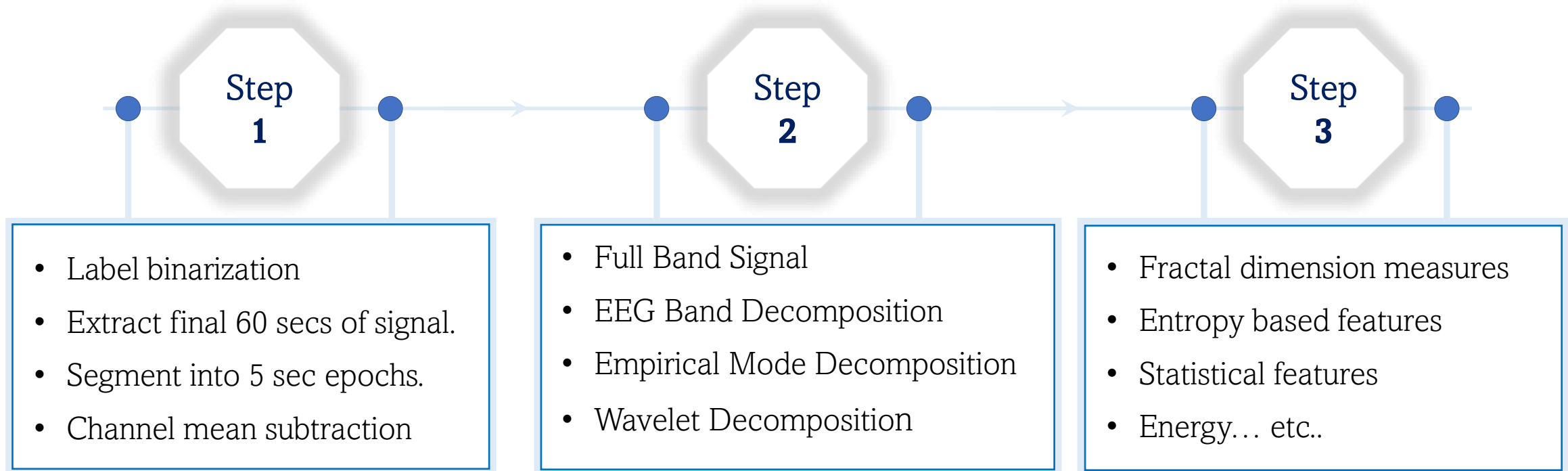
CHANNEL  
SELECTION



EMOTION  
CLASSIFICATION

# Preliminary Steps

---



- Tested feature selection methods like ANOVA, BorutaShap, Tree Ensemble feature importance, recursive feature elimination etc.

# Channel Selection Algorithm Summary

---

- Given output of feature selection as set of  $B$  positional features.
- To get best  $k$  out of  $N$  channels:
  1. Collect all possible channel combinations,  $C_k^N$ ;
  2. Calculate minimum sum of square error between each combination and corresponding features in  $B$ ;
  3. REPEAT (2) for all combinations; and
  4. RETURN the combination with the least reconstruction error



# Channel Selection Algorithm

[1]

- Assuming we had :
  - five electrode positions - F3, T8, T7, P7, AF4
  - four features extracted from each position: power (POW), entropy (ENT), fractal dimension (FD) and zero crossing rate (ZR).
  - 2 class labels – 0, 1
- We will obtain 20 positional features and a label vector as below :

F3_POW	F3_ENT	F3_FD	F3_ZR	T8_POW	T8_ENT	T8_FD	T8_ZR	T7_POW	T7_ENT	T7_FD	T7_ZR	P7_POW	P7_ENT	P7_FD	P7_ZR	AF4_POW	AF4_ENT	AF4_FD	AF4_ZR	LABEL
0.676	0.285	0.836	0.398	0.529	0.986	0.684	0.096	0.989	0.373	0.412	0.465	0.478	0.669	0.445	0.732	0.303	0.241	0.709	0.518	0
0.385	0.467	0.833	0.858	0.287	0.157	0.146	0.351	0.737	0.986	0.898	0.491	0.156	0.552	0.459	0.589	0.818	0.685	0.563	0.153	1
0.086	0.736	0.093	0.602	0.918	0.895	0.049	0.463	0.133	0.935	0.469	0.854	0.207	0.377	0.664	0.471	0.781	0.988	0.538	0.988	1
0.759	0.732	0.478	0.053	0.937	0.547	0.851	0.332	0.541	0.131	0.555	0.037	0.763	0.908	0.601	0.347	0.179	0.085	0.271	0.549	0
0.869	0.281	0.634	0.772	0.385	0.504	0.201	0.293	0.361	0.856	0.528	0.244	0.205	0.565	0.221	0.793	0.589	0.922	0.468	0.148	0
0.893	0.253	0.403	0.727	0.698	0.006	0.408	0.501	0.301	0.009	0.779	0.979	0.349	0.482	0.573	0.801	0.929	0.908	0.606	0.168	1
0.692	0.012	0.924	0.371	0.163	0.883	0.524	0.735	0.244	0.319	0.225	0.693	0.387	0.282	0.192	0.893	0.253	0.119	0.205	0.342	0

AF4_FD	AF4_ZR	LABEL
0.709	0.518	0
0.562	0.153	1
0.538	0.988	1
0.271	0.549	0

# Channel Selection Algorithm

[2]

- Again, assume feature selection algorithm output as F3\_POW, T8\_ENT, AF4\_ZR, P7\_FD.
- To select the **best two** channels:

**STEP 1:** Combinations of two from five electrodes (F3, T8, T7, P7, AF4). Thus,

$$C_2^5 = \frac{5!}{2!(5-2)!} = 10$$

[F3, T8],  
[T8,P7],

[F3, T7],  
[T8, AF4],

[F3, P7],  
[T7,P7],

[F3, AF4],  
[T7, AF4],

[T8,T7],  
[P7, AF4]

# Channel Selection Algorithm

[3]

**STEP 2:** For each of the combinations in [step 1](#), we find the best substitute for each of the target features obtained after feature selection.

E.g., for candidate combination [\[T7, AF4\]](#) and target feature [F3\\_POW](#), we compute:

$$A = \|T7\_POW - F3\_POW\|_2$$

$$B = \|AF4\_POW - F3\_POW\|_2$$

The one with a lower error is a better substitute for [F3\\_POW](#)

**STEP 3:** **REPEAT** [step 2](#) to get substitutes for [T8\\_ENT](#), [AF4\\_ZR](#) and [P7\\_FD](#)

# Channel Selection Algorithm

[4]

STEP 4: REPEAT steps 2 and 3 for all combinations to get substitutes such as:

COMBINATION	F3_POW	T8_ENT	AF4_ZR	P7_FD
[F3, T8]	F3_POW	T8_ENT	T8_ZR	F3_FD
[F3, T7]	-	-	-	-
[F3, P7]	-	-	-	-
[F3, AF4]	F3_POW	AF4_ENT	AF4_ZR	AF4_FD
[T8, T7]	-	-	-	-
[T8, P7]	-	-	-	-
[T8, AF4]	-	-	-	-
[T7, P7]	-	-	-	-
[T7, AF4]	T7_POW	AF4_ENT	T7_ZR	AF4_FD
[P7, AF4]	AF4_POW	P7_ENT	AF4_ZR	AF4_FD

# Channel Selection Algorithm

[5]

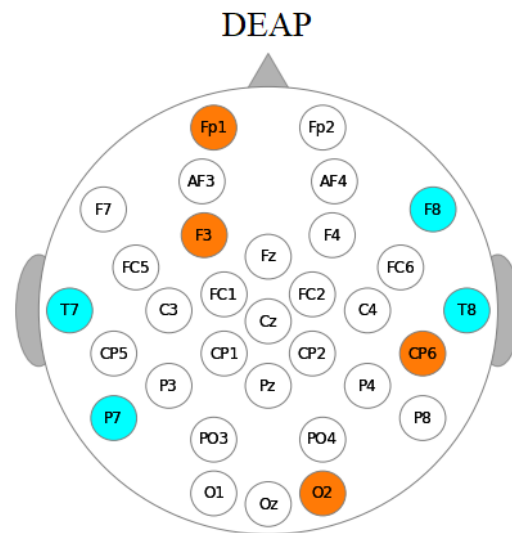
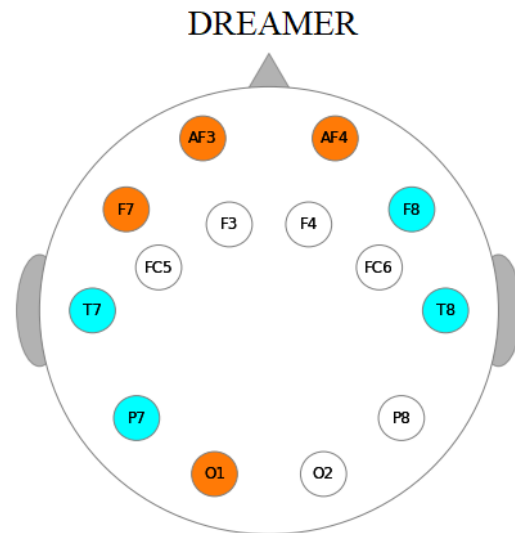
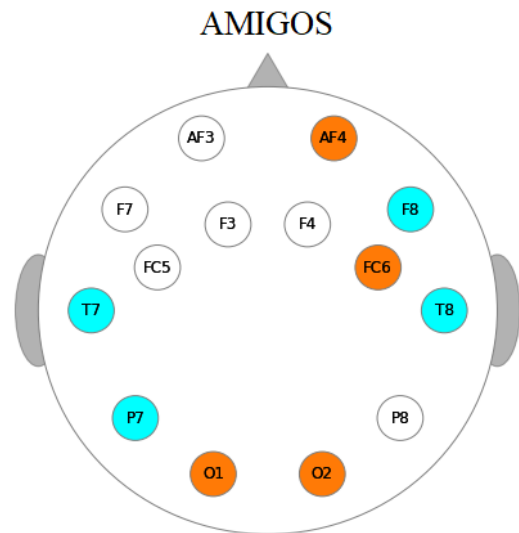
**STEP 5:** Calculate reconstruction error between channel combinations and the target features as the total mean squared errors from all substitute features to the targets.

COMBINATION	F3_POW	T8_ENT	AF4_ZR	P7_FD	SUM
[F3, T8]	0.000	0.009	0.002	0.009	0.0246
[F3, T7]	0.000	0.011	0.008	0.006	0.0251
[F3, P7]	0.000	0.002	0.003	0.007	0.0132
<b>[F3, AF4]</b>	<b>0.002</b>	<b>0.001</b>	<b>0.005</b>	<b>0.002</b>	<b>0.0099</b>
[T8,T7]	0.006	0.004	0.001	0.002	0.0127
[T8,P7]	0.001	0.009	0.003	0.002	0.0116
[T8, AF4]	0.011	0.009	0.008	0.005	0.0334
[T7,P7]	0.006	0.001	0.003	0.004	0.0138
[T7, AF4]	0.002	0.009	0.007	0.004	0.0217
[P7, AF4]	0.008	0.004	0.003	0.006	0.0216

Best substitute features **F3\_POW**, **AF4\_ENT**, **F3\_ZR** and **AF4\_FD**

# Results [1]

NUMBER OF CHANNELS	AMIGOS	DREAMER	DEAP
2	['FC6', 'O1']	['AF4', 'P7']	['CP6', 'T8']
3	['FC6', 'O1', 'P7']	['AF4', 'P7', 'T8']	['CP6', 'F3', 'T8']
4	['FC6', 'O1', 'P7', 'T8']	['AF4', 'F8', 'P7', 'T8']	['F3', 'Fp1', 'P7', 'T8']
5	['AF4', 'FC6', 'O1', 'P7', 'T8']	['AF4', 'F7', 'F8', 'P7', 'T8']	['F3', 'Fp1', 'O2', 'P7', 'T8']
6	['AF4', 'FC6', 'O1', 'P7', 'T7', 'T8']	['AF3', 'AF4', 'F7', 'F8', 'P7', 'T8']	['F3', 'Fp1', 'O2', 'P7', 'T7', 'T8']
7	['AF4', 'F8', 'FC6', 'O1', 'P7', 'T7', 'T8']	['AF4', 'F7', 'F8', 'FC6', 'O1', 'P7', 'T8']	['F3', 'F8', 'Fp1', 'O2', 'P7', 'T7', 'T8']
8	['AF4', 'F8', 'FC6', 'O1', 'O2', 'P7', 'T7', 'T8']	['AF3', 'AF4', 'F7', 'F8', 'O1', 'P7', 'T7', 'T8']	['CP6', 'F3', 'F8', 'Fp1', 'O2', 'P7', 'T7', 'T8']



## KEY



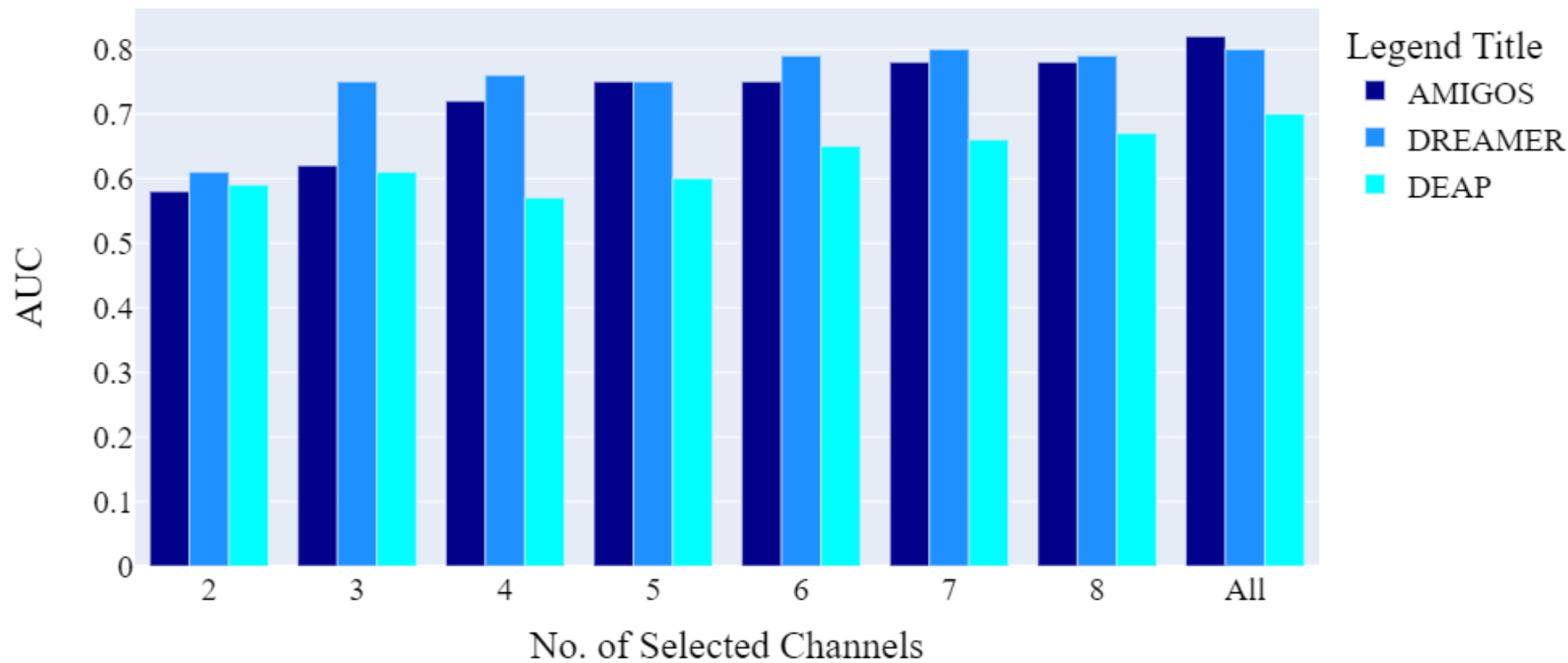
Active electrode



Active electrode in all three datasets

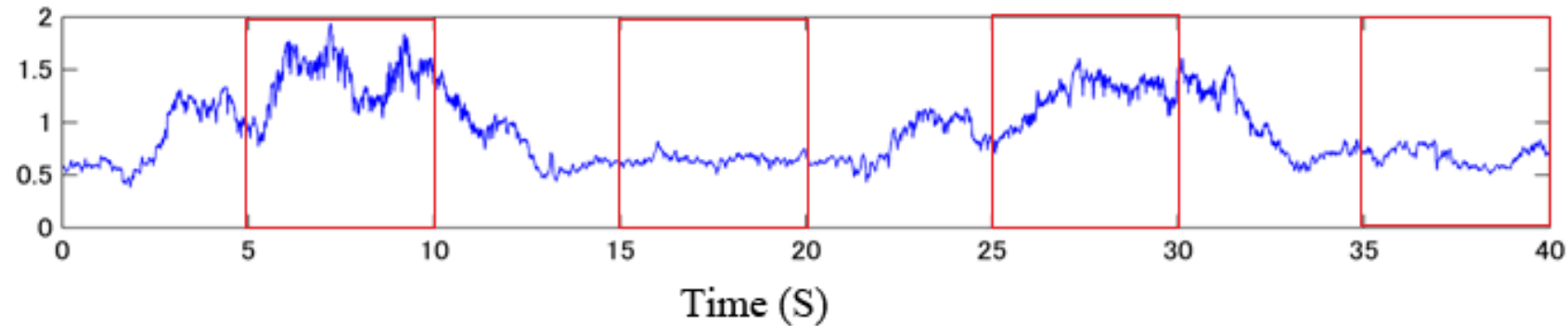
# Results [2]

---



# Weaknesses and Future Work

---



- Time segmentation of the signal may not be ideal
- L2 norm tends to penalize outliers
- Next step : Graph Convolution Network models



# Thank You!

Contact: [jam139@aber.ac.uk](mailto:jam139@aber.ac.uk) / [jmsonda@mubas.ac.mw](mailto:jmsonda@mubas.ac.mw)