

## **SIGNAL SPACE GENERATIVE ADVERSARIAL NETWORKS FOR MODELLING EEG BRAIN ACTIVITY AND PREDICTING EMOTIONAL DECISIONS**

### **ABSTRACT:**

We recently proposed an architecture for **generating EEG signals** called a **Signal Space Generative Adversarial Network (SigS-GAN)**, that **learns a latent space representation of the signals** it was trained on. We impose a **regularization on these latent representations of signals**, which makes them **useful for understanding and predicting the processes** that were visible in the EEG activity.

The regularization (which is an **extension of Path-Length Regularization to the frequency domain**) encourages the **learning of a latent space** where a **distance between two points approximately corresponds to a measure of distance between the two signals that would be generated if we were to put these points into the generator**. This is useful as it (a) adds **smoothness to the representation**, such that **signals that are similar correspond to points that are near each other**, (b) **directions in latent space start to correspond to useful features of the signals**, which makes it easier to describe and perform classification, (c) you can use such a latent space in order to **perform a new kind of EEG analysis**, where you analyze, **in the latent space**, the **differences between point corresponding to signals** that, for example, **lead to two different decisions**.

The goal of this project is to **develop the architecture**, and **create the analysis methods and tool** needed to pursue this last opportunity (c) for a new way of EEG analysis. We will brainstorm what **modification to the present architecture would make a latent space analysis of EEG signals easier and more fruitful**. Implement them, and **train the networks** on several different datasets of ERP studies, where participants **made different types of decisions based on a processing of emotional words**. Next we will apply the developed methods in order to **explain what differences in electrical activity correlated with different decisions** and try to **predict them on data unseen by the model**. The techniques developed as a part of this project could lead to a scientific publication.

### **LIST OF MATERIALS:**

[1] Path-Length Regularization:

<https://paperswithcode.com/method/path-length-regularization>

[2] EEG and GANs:

<https://www.sciencedirect.com/science/article/abs/pii/S0208521621001273?via%3Dihub>

### **LIST OF REQUIREMENTS FOR TAKING PART IN THE PROJECT:**

- Either knowledge of python (we'll use PyTorch for the neural net) or mathematics (linear algebra, multi-variate calculus), as we'll spend some time working out Fourier-analysis-based regularization terms and statistical approaches to the analysis of a trained latent space.

- It is not required to be proficient in the topics discussed in the abstract (GANs, path-length regularization, latent space representations), as we will spend some time at the beginning of the project acquainting ourselves with them.

**MAXIMAL ALLOWED NUMBER OF TEAM MEMBERS: 10**