

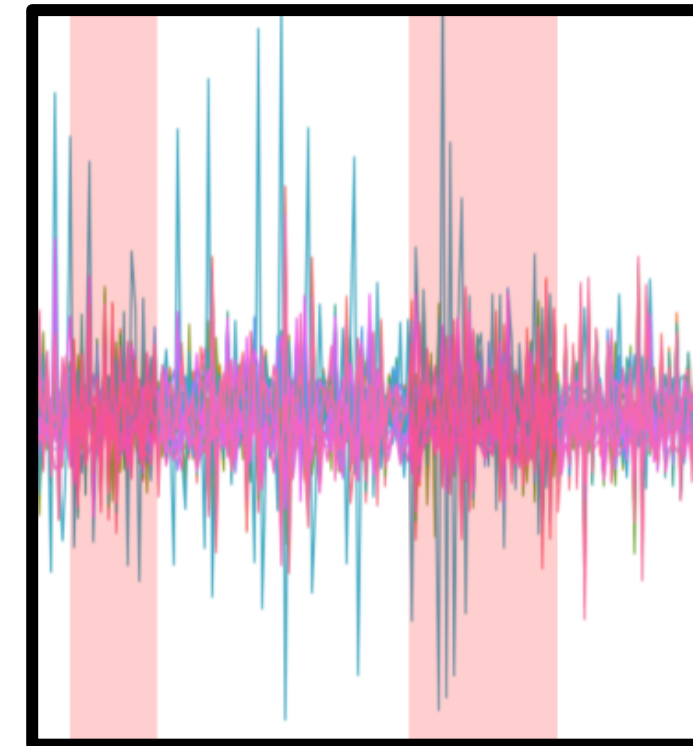
# GAN-Based Data Augmentation for Epileptic Seizure Annotation

## 1 Introduction

- Epilepsy affects **>50M individuals** worldwide
- Manual analysis takes **hours** per patient
- Automating it means **more time for patients**

**Dataset:** Annotated EEGs of seizure patients [1]

**Challenge:** Annotate types of seizures per timestamp

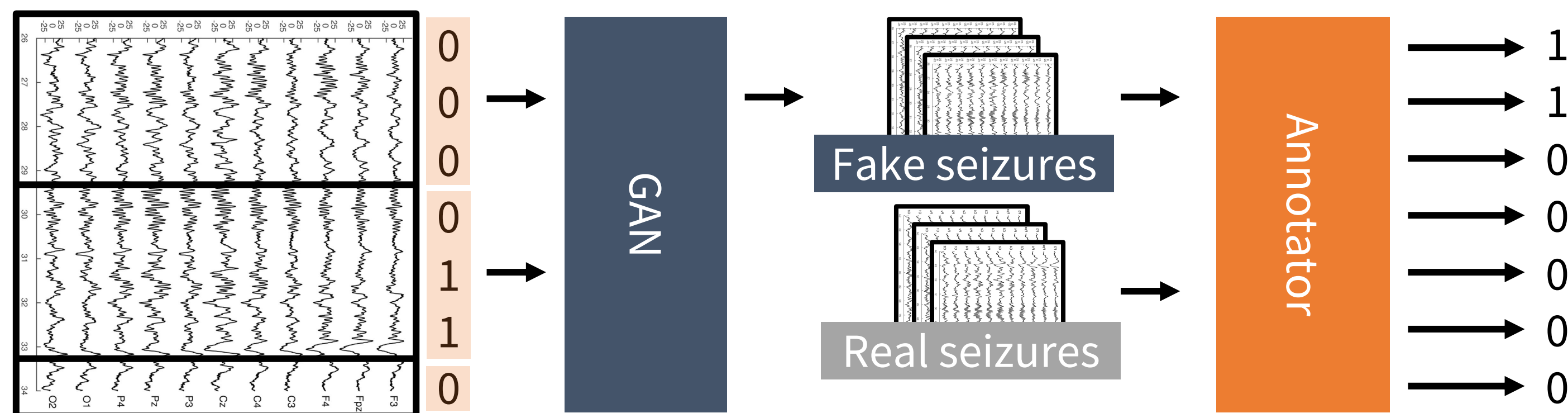


## 2 Approach

**Problem:** Seizure events relatively rare in recordings

**Our approach:**

- Augment dataset with GAN-generated fake seizures (based on [2])
- Implement and train LSTM and CNN models with Keras
- Use baseline comparison model to test GAN

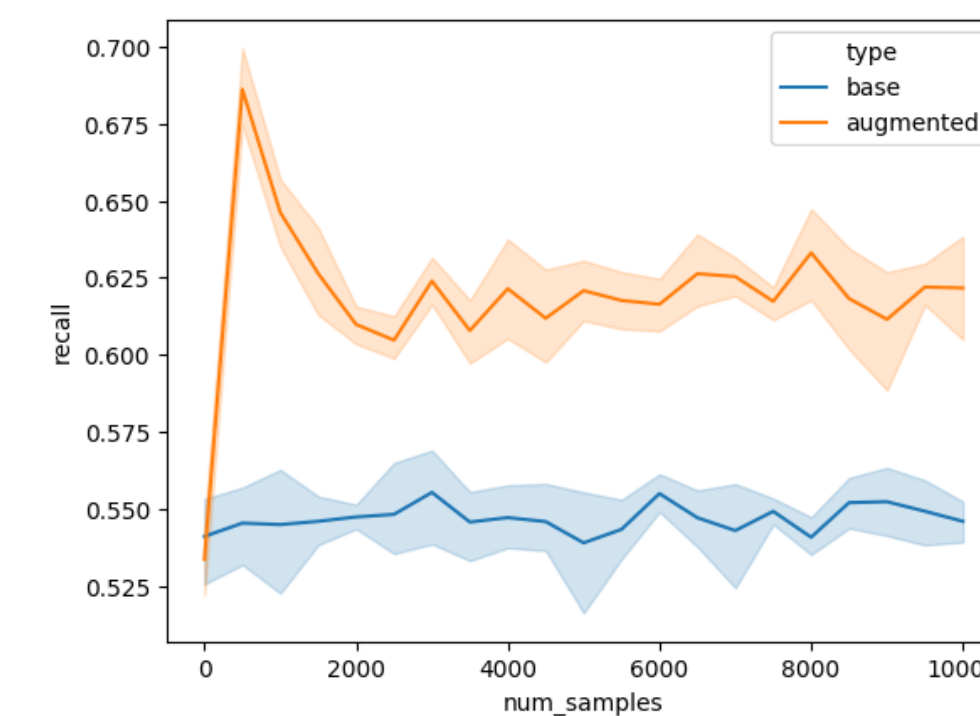


## 3 Research Questions

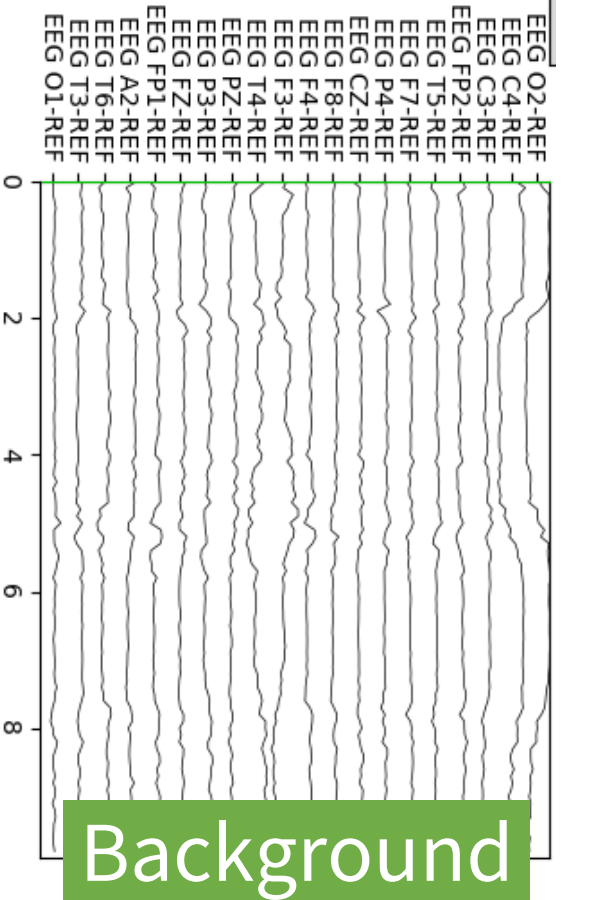
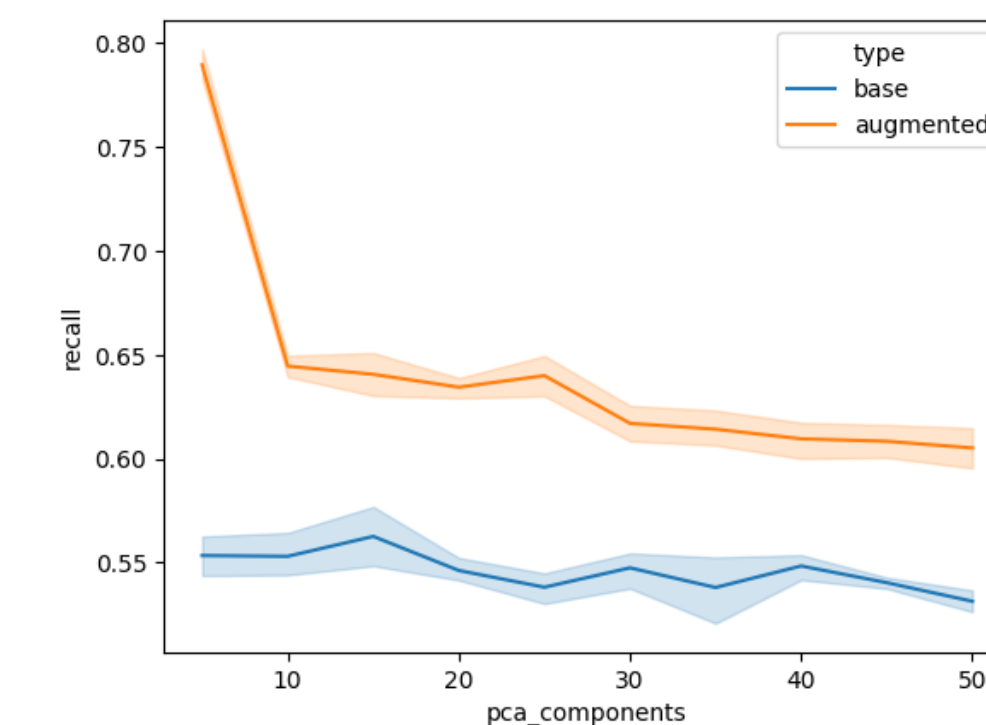
1. How do existing methods compare against each other, in terms of **precision** and **recall**?
2. Can GAN-based **data augmentation** improve performance?

## 4 Experiments

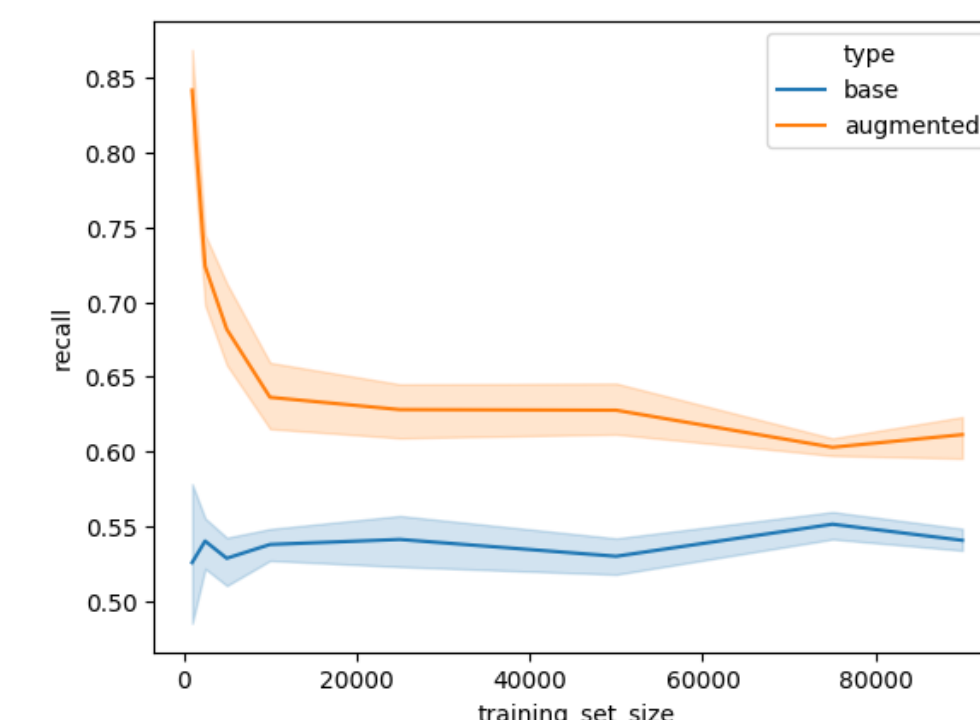
1. Num. Fake Seizures



2. Num. PCA Components

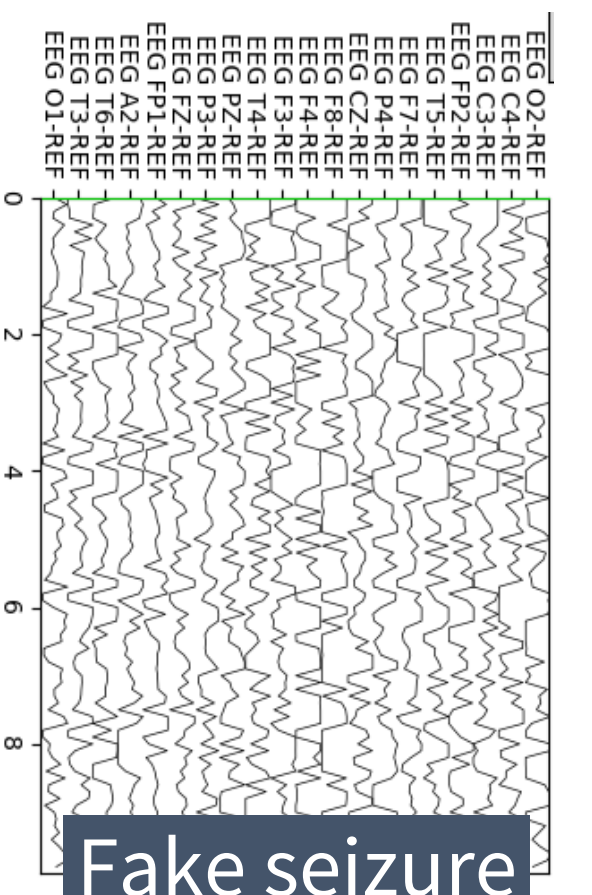


3. Num. Training Samples



### Key insights

- Adding generated seizures improves performance
- Data augmentation especially strong with small training set



## 5 Discussion

- Risk of overfitting
- Single window size vs. variable-length seizures
- Generate different types of seizures (instead of binary)

## 6 Conclusion

- Annotating with raw data is challenging (2k+ features in window)
- Rarity of seizure events can be compensated with GANs
- Data augmentation improves performance significantly

[1] Harati et al. "The TUH EEG CORPUS: A big data resource for automated EEG interpretation." IEEE, 2014.

[2] Radford et al. "Unsupervised representation learning with deep convolutional generative adversarial networks." Preprint on arXiv:1511.06434, 2015.