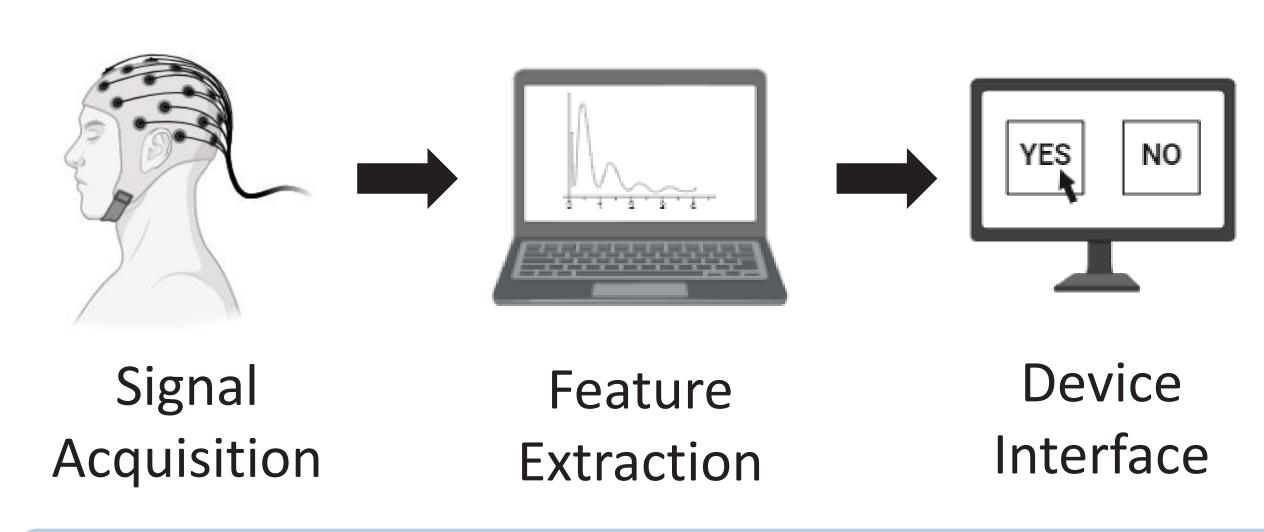
**Project**: Deep Learning (DL) Enabled EEG Artifact Removal Algorithms for Real-World BCI Applications

Wan. J, Floreani. E, and Chau. T

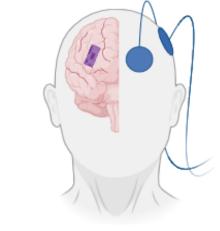
### Background

Typical BCI workflow



Accuracy of the BCI task is affected by the amount of noise in the EEG signal

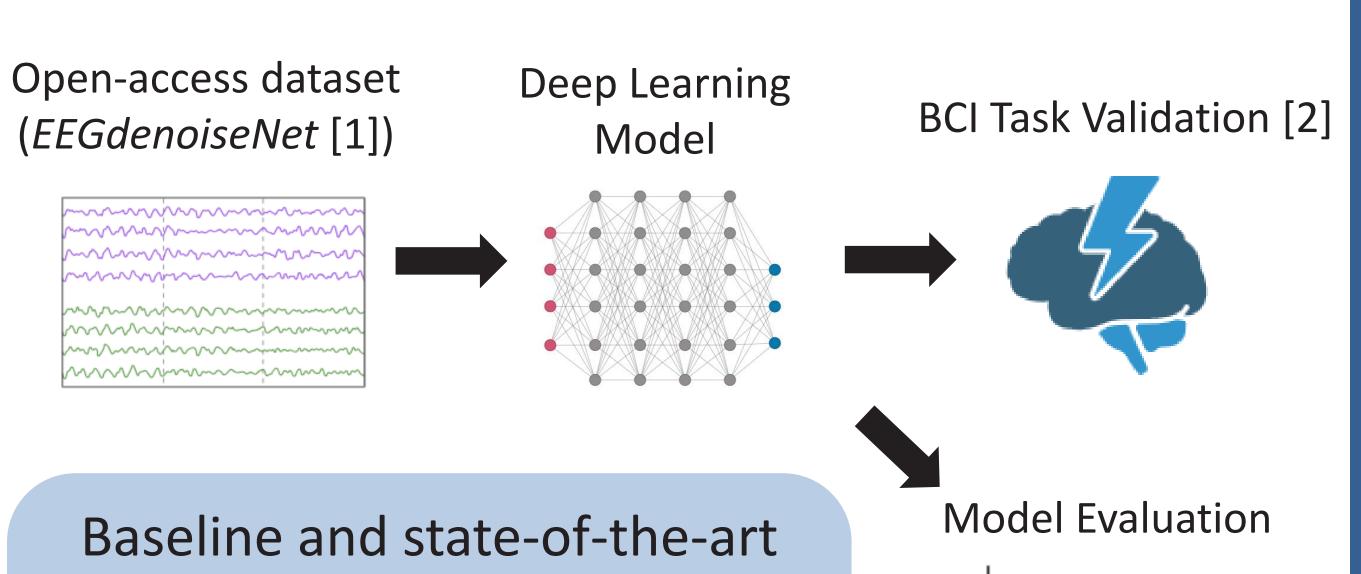
# Objective



Does EEG noise removal using DL techniques improve performance of real world BCI tasks?

#### Methods

# DL Pipeline for BCI Validation



Baseline and state-of-the-art models from literature, as well as a novel Short-time Fourier Transform based CNN model were compared.

Removing Noise From Brain Signals with AI to Improve Brain-Computer Interface (BCI) Tasks

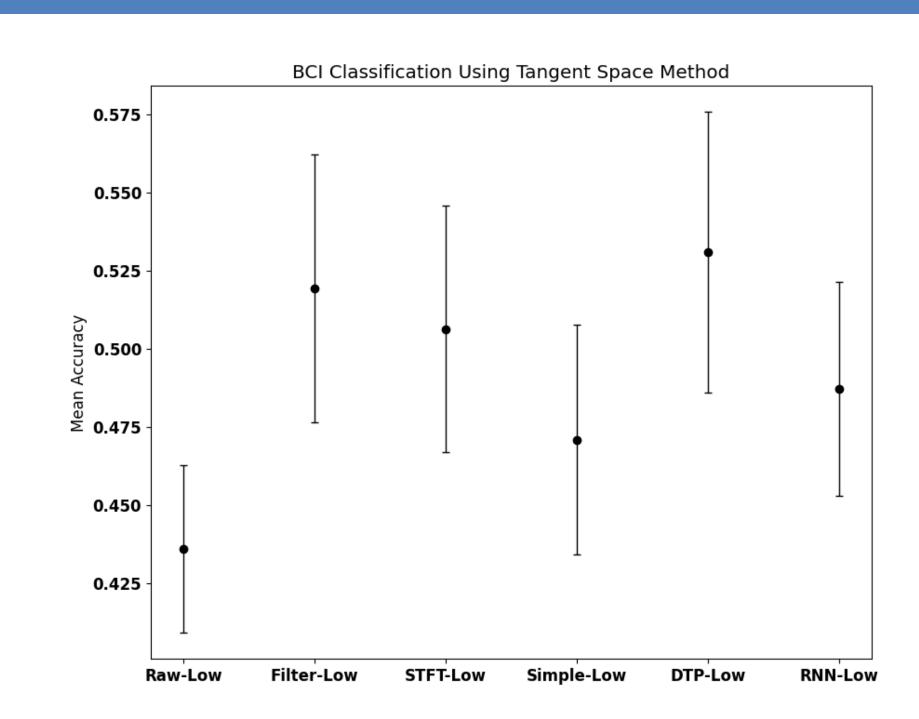


Holland Bloorview

Kids Rehabilitation Hospital



### Results



Model	Accuracy	Model Size
SimpleCNN	$0.47 \pm 0.04$	10.5M
RNN	$0.49 \pm 0.03$	788k
DTPNet	0.53 ± 0.04	40M
Our model (STFT)	0.51 ± 0.04	16M

### Conclusion

- Overall, DL models improved the performance of the classification task.
- Results from DL models were comparable with bandpass filtering, a traditional noise removal technique.
- Our novel CNN is the second-best performing model amongst all the DL algorithms.

# **Next Steps**

Evaluate level
 of noise in
 validation dataset

2. Further analyze properties of BCI data

3. Experiment with other BCI datasets

#### Relevance

Development of novel EEG artifact removal techniques will enhance BCI usability in real-world environments, extending benefits children with severe motor disabilities.

**References:** [1] Zhang, H., Zhao, M., Wei, C., Mantini, D., Li, Z., and Liu, Q. (2021b). EEGdenoiseNet: a benchmark dataset for deep learning solutions of EEG denoising. *J. Neural Eng.* 18:056057. doi: 10.1088/1741-2552/ac2bf8

[2] Goldberger, A., Amaral, L., Glass, L., Hausdorff, J., Ivanov, P. C., Mark, R., ... & Stanley, H. E. (2000). PhysioBank, PhysioToolkit, and PhysioNet: Components of a new research resource for complex physiologic signals. Circulation [Online]. 101 (23), pp. e215–e220.