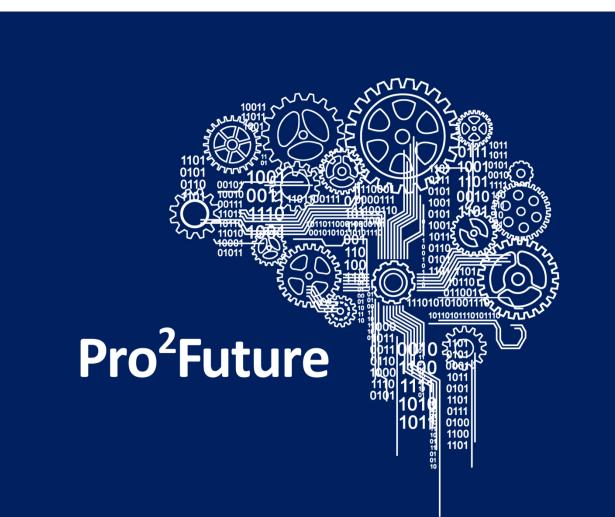
# StreamingAl

Federated Embedded Al for the digital transformation of Austrian Industries.



Alois Ferscha<sup>1,2</sup>, Bernhard Anzengruber-Tanase<sup>1</sup>, Michael Haslgrübler<sup>1</sup>, Ekaterina Sysoykova<sup>1</sup>, Georgios Sopidis<sup>1</sup>, Behrooz Azadi<sup>1</sup>, Michael Siegl<sup>1</sup>, Miguel Vazquez<sup>2</sup>, Patrick Denzler<sup>2</sup>, Sepp Hochreiter<sup>3</sup>

Pro2Future GmbH<sup>1</sup>, JKU-IPC (Institute of Pervasive Computing)<sup>2</sup>, JKU-IML (Institute of Machine Learning)<sup>3</sup>

- <sup>1</sup> Science Park 4, Altenberger Strasse 69, 4040 Linz
- <sup>2</sup> Science Park 3, Altenberger Strasse 69, 4040 Linz
- <sup>3</sup> Science Park 3, Altenberger Strasse 69, 4040 Linz



# **MOTIVATION & GOALS**

Streaming Al aims to drive low TRL, foundational research to develop Al for industrial applications. In contrast to conventional pre-trained, holistic, and resource-intensive Al,

streaming machine learning methods

**FRAMEWORK** 

ii. on-device machine learning methods are to be introduced,

# **Project FactBox**

18 Months

**Project Name** StreamingAl **Project ID** 

Area 1

**Area Perception** 

**Project Lead** 

**Duration** 

Dr. Bernhard Anzengruber-

Tanase

thereby reducing dependence on mass training data and supporting ecological sustainability.

# STATIC FEDERATION and FEW-SHOT LEARNING

Developing privacy-preserving Al for time-series classification using Federated Few-Shot Learning (FFSL), enabling cross-institutional model training without data sharing and rapid local adaptation to individuals.

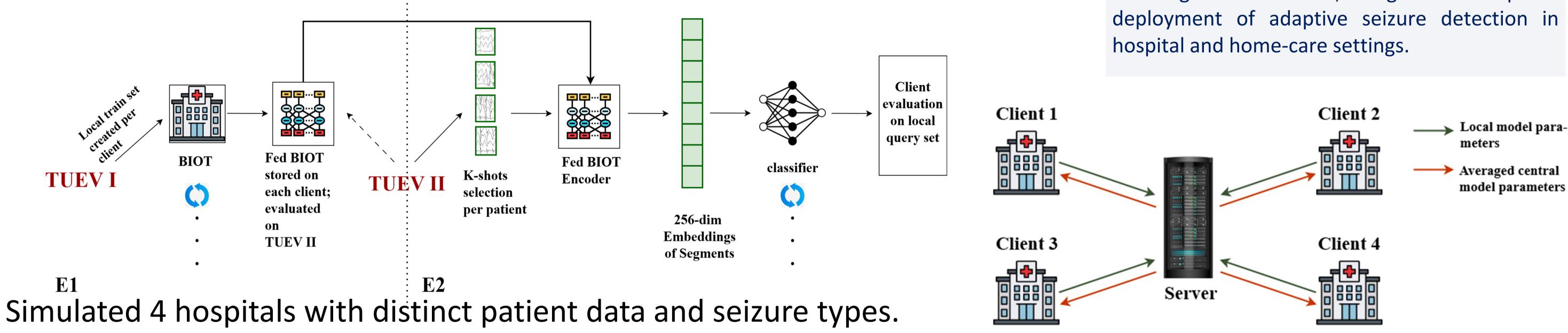
# CONTRIBUTION

### **Scientific contribution**

Demonstrating the integration of few-shot and federated learning for privacy-preserving, personalized time series classification.

### **Economic contribution**

Enabling cost-efficient, regulation-compliant deployment of adaptive seizure detection in



**Federated Averaging** 

Two-stage training:

E1: Federated pre-training of a transformer-based seizure classification model on distributed EEG data.

E2: Federated few-shot learning for patient-specific seizure detection.

# Client Patient # Bal. Accuracy 0.838 0.912 0.533 6 0.797

Metric	BIOT	Fed BIOT
Bal. Accuracy	0.5207	0.4328

# **RESULTS**

- Federated model shows slightly lower performance than centralized but gains real-world applicability through privacy compliance.
- FFSL achieves up to 91% balanced accuracy with just 5 labeled samples per patient, demonstrating feasible privacy-preserving, personalized seizure detection.

# **IMPACT**

- Predictive maintenance: cross-site training, local adaptation with few samples
- Wearable health monitoring: shared model, personalized to each user.

Contact: Dr. Bernhard Anzengruber-Tanase, Pro2Future GmbH, bernhard.anzengruber@pro2future.at, +43 732 2468 - 9474 **Acknowledgement**: This work was supported by Pro<sup>2</sup>Future II (FFG, 911655) and the Province of Upper Austria (Land OÖ).





















