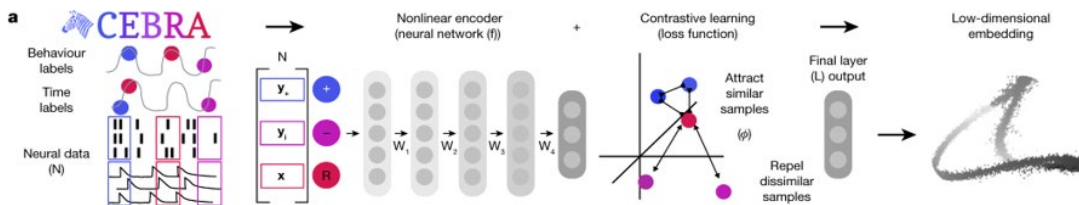


Dimensionality Reduction of Time-Series Data using Contrastive Learning (CEBRA)



CEBRA allows for self-supervised, supervised and hybrid approaches for both hypothesis- and discovery-driven analysis. Overview of pipeline: collect data (for example, pairs of behaviour (or time) and neural data (x, y)), determine positive and negative pairs, train CEBRA and produce embeddings. W_1, \dots, W_4 represent the neural network weights.

Ref. Nr

6.2273

Keywords

CEBRA, brain-machine interface, BCI, neural decoding, embedding, AI

Intellectual Property

US 12,499,131B2

Publications

Nature volume 617, pages 360-368 (2023)

Date

13/04/2026

Description

Current methods do not allow to extract meaningful patterns from high-dimensional time-series data (e.g., neural signals, sensor streams, behavioral recordings) as the number of variables (dimensions) increases, and the relationships are non-linear.

This technology relates to a self-supervised encoding method for dimensionality reduction of time-series data, leveraging contrastive learning techniques. It enables efficient transformation of multidimensional temporal datasets into a latent embedding preserving essential patterns and relationships.

Advantages

- Enables latent embeddings that are robust to noise, temporal distortions, and sampling variability;
- Reduced dependence on labeled datasets; Superior performance over linear and reconstruction-based methods;
- Robustness to domain shifts.

Applications

- brain-machine interface and neural decoding
- Mapping/modelling behavioural action to neural activity
- kinematics