The State of NeuroBench

NICE 2025



What is NeuroBench?

Benchmark

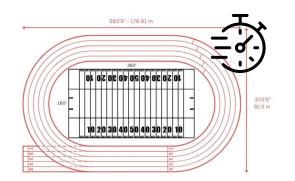
A standard for measuring performance, including task, metrics, methodology.

Benchmarking

Running benchmarks to profile and compare different solutions.

Benchmark Framework

Systematic benchmarking and implementation of novel benchmarks.



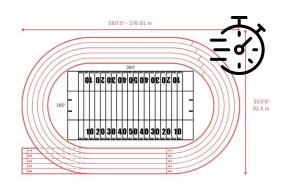






Benchmark

A standard for measuring performance.



Benchmarking under a **common framework** <u>aligns research</u>, <u>identifies best practices</u>, and <u>drives technological progress</u>.

Benchmark Framework

Systematic benchmarking and implementation of novel benchmarks.







A <u>Benchmark Framework</u> for Neuromorphic Computing Research









NeuroBench Goals

Inclusive

Community driven, minimal assumptions of the solutions

Actionable

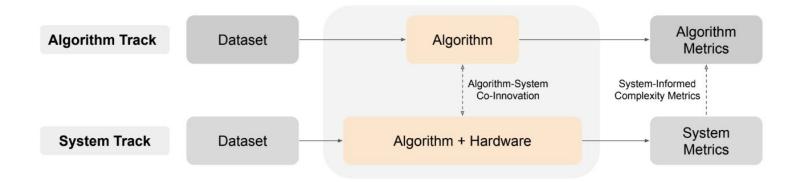
Common tooling, portable, standardized

Iterative

Provide starting point, evolve over time



Benchmark Tracks: Algorithms and Systems



Algorithm track: profile algorithm cost in hardware-agnostic setting

System track: measuring deployed performance on neuromorphic hardware



Tasks

Algorithm Track



Keyword Few-shot, Continual Learning



Event Camera
Object Detection



Motor Neural Decoding



Chaotic Sequence Prediction

System Track



Acoustic Scene Classification



QUBO

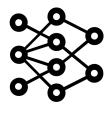


Metrics

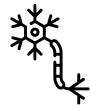
Algorithm Track







Connection Sparsity



Activation Sparsity



Synaptic Operations

System Track







Efficiency



Algorithm Track Harness

https://github.com/NeuroBench/neurobench

pip install neurobench



Automatic benchmarking for neural network models



Datasets,
Data Processing,
Metrics



Natively compatible with **torch**-based libraries



Actively maintained and built for extensibility



The State of NeuroBench



NeuroBench Community Timeline

April @ NICE Conf, Whitepaper presentation 2023



April @ NICE Conf, Keynote presentation 2024

July @ Telluride Workshop, Tutorial

Sept @ NCN Summit, Poster

Oct @ BioCAS Conf, Grand Challenge on Neural Decoding









NeuroBench Community Timeline

2023 2024 <u>Jason Yik</u> [™], <u>Korneel Van den Berghe, Douwe den Blanken, Younes Bouhadjar, Maxime Fabre, Paul</u> 2025 Feb in Nat Mar @ NI

Perspective Open access | Published: 11 February 2025

The neurobench framework for benchmarking neuromorphic computing algorithms and systems

Hueber, Weijie Ke, Mina A. Khoei, Denis Kleyko, Noah Pacik-Nelson, Alessandro Pierro, Philipp Stratmann, Pao-Sheng Vincent Sun, Guangzhi Tang, Shengi Wang, Biyan Zhou, Soikat Hasan Ahmed, Corradi, Guido de Croon, Andreea Danielescu, Anurag Daram, Mike Davies, Yigit Demirag, Jason Haoyuan Ma, Rajit Manohar, Josep Maria Margarit-Taulé, Christian Mayr, Konstantinos Michmizos, Dylan Sandamirskaya, Clemens J. S. Schaefer, André van Schaik, Johannes Schemmel, Samuel Schmidgall, Catherine Schuman, Jae-sun Seo, Sadique Sheik, Sumit Bam Shrestha, Manolis Sifalakis, Amos Sironi,

Urgese, Marian Verhelst, Craig M. Vinevard, Bernhard Vogginger, Amirreza Yousefzadeh, Fatima Tuz

ecoding accepted



June @

Nature Communications 16, Article number: 1545 (2025) | Cite this article

Zohora, Charlotte Frenkel & Vijay Janapa Reddi

Benchmarking of hardware-efficient real-time neural decoding in brain-computer interfaces

Paul Hueber, Guangzhi Tang, Manolis Sifalakis, Hua-Peng Liaw, Aurora Micheli, Nergis Tomen and

Yao-Hong Liu

Decoding finger velocity from cortical spike trains with recurrent spiking neural networks

Tengjun Liu, Julia Gygax, Julian Rossbroich, Yansong Chua, Shaomin Zhang, Friedemann Zenke

Leveraging Recurrent Neural Networks for Predicting Motor Movements from Primate Motor Cortex Neural Recordings

ces (BMIs) can significantly improve the life onetheless, externally mounted pedestals pose

otential to dramatica

their ability to perfor

Yuanxi Wang, Zuowen Wang, Shih-Chii Liu

movements from neural recordings

This paper presents an efficient dee Hybrid Spiking Neural Networks for Low-Power Intra-Cortical Brain-Machine Interfaces

Alexandru Vasilache, Jann Krausse, Klaus Knobloch, Juergen Becker

Neurobench: DCASE 2020 Acoustic Scene Classification benchmark on XyloAudio 2

Weijie Ke, Mina Khoei, Dylan Muir

XyloAudio is a line of ultra. Zero-Shot Temporal Resolution Domain Adaptation in unique dynamics. While backpropagation through time (BPTT) with near-microphone analysis for Spiking Neural Networks

Sanja Karilanova, Maxime Fabre, Emre Neftci, Ayça Özçelikkale

Spiking Neural Networks (SNNs) are biologically-inspired deep neural networks that efficiently extract temporal information while offering promising gains in terms of

Natively neuromorphic LMU architecture for encoding-free SNN-based HAR on commercial edge devices

Vittorio Fra, Benedetto Leto, Andrea Pignata, Enrico Macii, Gianvito Urgese

Neuromorphic models take inspiration from the human brain by adopting bioplausible neuron models to build alternatives to traditional Machine Learning (ML)

> Reinforcement Learning for Spiking Neural Networks

Recurrent Reinforcement Learning with Surrogate Gradients

Korneel Van den Berghe

Slax: A Composable JAX Library for Rapid and Flexible Prototyping of Spiking Neural Networks

Thomas M. Summe, Siddharth Joshi

Inces to algorithms for training spiking neural networks (SNNs) often



Grand Challenge on Neural Decoding (Oct 2024)

Benchmarking under a **common framework** <u>aligns research</u>, <u>identifies best practices</u>, and <u>drives technological progress</u>.

	Model	R^2	Footprint (bytes)	SynOps		
				Dense	Eff_MACs	Eff_ACs
	ANN Baseline	0.5755	27160	6236	4970	0
	SNN Baseline	0.5805	29248	7300	0	413
	AEGRU	0.6982	45520	54283	25316	0
	RSNN-L	0.6978	4833360	1206272	0	42003
	RSNN-S	0.6604	27144	13440	0	304
•	ConvGRU	0.6209	26568	4947	627	247





Open-source Harness Tooling







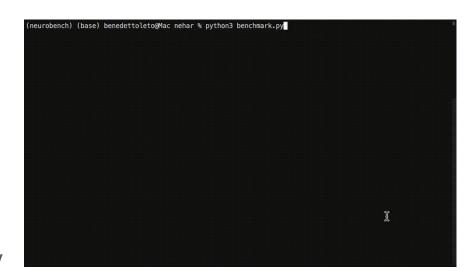


Latest Release: 2.0.0 (Jan 2025)

- Full re-organization of library
- New interface for custom metrics
- Connection to NIR

WIP updates

- Progress bar / results UI
- New metrics (activation sparsity by layer)













What's Next for NeuroBench?



What are effective methods for better benchmark performance?

Are there tasks which are a better fit for the technology?

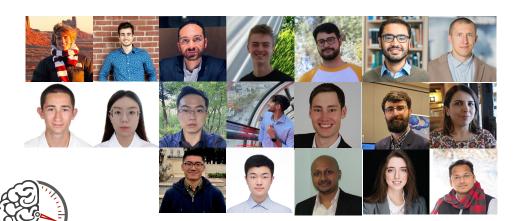
Are other metrics more representative of better performance?

Are there further harness features to better facilitate research?



NeuroBench Tutorial Tomorrow

- How to use the harness.
- Discussing next directions for integrating algorithm and system benchmarking
- Invitation to open discussion



Perspective Open access | Published: 11 February 2025

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Jason Yik 🗹, Korneel Van den Berghe, Douwe den Blanken, Younes Bouhadiar, Maxime Fabre, Paul Hueber, Weijie Ke, Mina A. Khoei, Denis Kleyko, Noah Pacik-Nelson, Alessandro Pierro, Philipp Stratmann, Pao-Sheng Vincent Sun, Guangzhi Tang, Shengi Wang, Biyan Zhou, Soikat Hasan Ahmed, George Vathakkattil Joseph, Benedetto Leto, Aurora Micheli, Anurag Kumar Mishra, Gregor Lenz, Tao Sun, Zergham Ahmed, Mahmoud Akl, Brian Anderson, Andreas G. Andreou, Chiara Bartolozzi, Arindam Basu, Petrut Bogdan, Sander Bohte, Sonia Buckley, Gert Cauwenberghs, Elisabetta Chicca, Federico Corradi, Guido de Croon, Andreea Danielescu, Anurag Daram, Mike Davies, Yigit Demirag, Jason Eshraghian, Tobias Fischer, Jeremy Forest, Vittorio Fra, Steve Furber, P. Michael Furlong, William Gilbin Aditya Gilra, Hector A. Gonzalez, Giacomo Indiveri, Siddharth Joshi, Vedant Karia, Lyes Khacef, James C. Knight, Laura Kriener, Raikumar Kubendran, Dhireesha Kudithipudi, Shih-Chii Liu, Yao-Hong Liu, Haoyuan Ma, Rajit Manohar, Josep Maria Margarit-Taulé, Christian Mayr, Konstantinos Michmizos, Dylan R. Muir, Emre Neftci, Thomas Nowotny, Fabrizio Ottati, Ayca Ozcelikkale, Privadarshini Panda, Jongkil Park, Melika Payvand, Christian Pehle, Mihai A. Petrovici, Christoph Posch, Alpha Renner, Yulia Sandamirskaya, Clemens J. S. Schaefer, André van Schaik, Johannes Schemmel. Samuel Schmidgall. Catherine Schuman, Jae-sun Seo, Sadigue Sheik, Sumit Bam Shrestha, Manolis Sifalakis, Amos Sironi Kenneth Stewart, Matthew Stewart, Terrence C. Stewart, Jonathan Timcheck, Nergis Tömen, Gianvito Urgese, Marian Verhelst, Craig M. Vineyard, Bernhard Vogginger, Amirreza Yousefzadeh, Fatima Tuz Zohora, Charlotte Frenkel & Vijay Janapa Reddi Show fewer authors

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