Wannan (Winnie) Yang

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EDUCATION

Ph.D. candidate in Computational Neuroscience. New York University, Buzsáki Lab 🗹	Graduating in 2025.9
Visiting Student. MIT, Tye Lab	2018.6 - 2019.6
B.S. in Computational Neuroscience. University of Edinburgh. GPA: 4.0 (USA equivalent)	2014.9 - 2018.5

RESEARCH EXPERIENCE

LLM Interpretability and Alignment (ICLR 🗹) 🖸

2024.3 - present

Collaborator: Chen Sun, Google DeepMind

- Designed and conducted experiments to study two safety-related problems in large language models (LLMs): deception and jailbreaks.
- Implemented a battery of interpretability tools including contrastive activation steering, activation patching and sparse auto-encoders (SAEs) to understand and control LLMs.
- Built a pipeline (github 🗹) to evaluate, analyze and steer 25+ large language models from different model families (Gemma, Llama, Pythia, Qwen and Yi) of different sizes (form 1.5 billion to 70 billion parameters).
- Published a series of technical blog posts to share the research findings.

Memory Representation and Consolidation (Science ☑, Nature ☑, NeurIPS ☑) 2020.9 – present Mentor: György Buzsáki, NYU

- Led a project to study a key mechanism for selective memory consolidation in the brain. This novel discovery has led to a publication in *Science* (leading author).
- Developed a novel latent-space based decoding method and applied various ML tools (including Bayesian decoding) to decode the content of memory reactivations ('replays') from neural population activity during learning and sleep.
- Created a pipeline for decoding large-scale (50TB) electrophysiology data. The pipeline can be applied to a diverse range of tasks. Implemented variants of the method to different datasets and projects, which enabled further key publications, including a collaboration project recently accepted at *Nature* (in press) and a first author paper at *NeurIPS* Symmetry and Geometry in Neural Representations Workshop.

Brain-inspired Deep Reinforcement Learning (NeurIPS ☑)

2021.3 - 2023.9

Collaborator: Chen Sun, Google DeepMind

- Co-developed a brain-inspired (memory consolidation and reflection) framework to build a novel deep RL algorithm.
- The resulting simple and scalable algorithm greatly improved long-term credit assignments in a diverse set of RL tasks (including grid-world, Montezuma's Revenge and other Atari games).

PUBLICATIONS

- W. Yang, C. Sun, G. Buzsáki. (2024). Interpretability for Safe AI: Jailbreak as a case study. In preparation.
- W. Yang, G. Buzsáki. (2024). Interpretability of LLMs Deception: Universal Motif. ICLR 🗹 (under review).
- W. Yang, C. Sun, R. Huszár, T. Hainmueller, K. Kiselev, G. Buzsáki. (2024). Selection of experience for memory by hippocampal sharp wave ripple. Science 383, 1478-1483.
- I. Zutshi, A. Apostolelli, W. Yang, Z. Zheng, T. Dohi, E. Balzani, A. H. Williams, C. Savin, G. Buzsáki. (2024). Hippocampal neuronal activity is aligned with action plans. *Nature* (in press) .
- C. Sun, W. Yang, T. Jiralerspong, D. Malenfant, B. Alsbury- Nealy, Y. Bengio, B. Richards. (2023). Contrastive Retrospection: honing in on critical steps for rapid learning and generalization in RL. NeurIPS.
- W. Yang, C. Sun, R. Huszár, G. Buzsáki. (2023). Changes in the geometry of hippocampal representations across brain states. Symmetry and Geometry in Neural Representations Workshop NeurIPS.
- E. Y. Kimchi, A. Burgos-Robles, G. A. Matthews, T. Chakoma, M. Patarino, J. Weddington, C. A. Siciliano, <u>W. Yang</u>, S. Foutch, R. Simons, M. Fong, M. Jing, Y. Li, D. B. Polley, Kay M. Tye. (2023). **Reward contingency gates** selective cholinergic suppression of amygdala neurons. eLife ∠
- S. Tennant, I. Hawes, H. Clark, W. Tam, J. Hua, W. Yang, K. Gerlei, E. Wood, M. Nolan. (2022). Analogue representation of a spatial memory by ramp-like neural activity in retrohippocampal cortex. Current Biology
- C. Sun, W. Yang, J. Martin, S. Tonegawa. (2020). Hippocampal neurons represent events as transferable units of experience. Nature Neuroscience ...

SKILLS

ML: Pytorch, scikit-learn

LLM Agent: LangChain, AutoGen

 $\textbf{LLM Interpretability:} \ \text{transformer-lens, Hugging Face Transformers, Contrastive Activation Steering, Activation Patching,} \\$

SAE Steering

Programming: Python, MATLAB, HTML, LaTeX

 $\textbf{Computational Neuroscience:} \ \, \text{Large-scale High-dimensional Data Analysis, Linear and Nonlinear Dimensionality Reductional Neuroscience:} \ \, \text{Large-scale High-dimensional Data Analysis, Linear and Nonlinear Dimensionality Reductional Neuroscience:} \ \, \text{Large-scale High-dimensional Data Analysis, Linear and Nonlinear Dimensionality Reductional Neuroscience:} \ \, \text{Large-scale High-dimensional Data Analysis, Linear and Nonlinear Dimensionality Reductional Neuroscience:} \ \, \text{Large-scale High-dimensional Data Analysis, Linear and Nonlinear Dimensionality Reductional Neuroscience:} \ \, \text{Large-scale High-dimensional Data Analysis, Linear and Nonlinear Dimensionality Reductional Data Analysis,} \ \, \text{Large-scale High-dimensional Data Analysis,} \ \, \text{Large-scale High-dime$

tion, Time Series Data Analysis, Neural Data Decoding

COURSES

Large Language Model Agents
Ongoing

Instructer: Dawn Song 🗹

Deep Learning

NYU. Grade: A

Instructer: Yann LeCun 🗹

Computational Cognitive Modeling

NYU. Grade: A

Instructer: Brenden Lake

Reinforcement Learning UCL.

Instructor: David Silver

Neural Circuits and Computational Modeling

NYU. Grade: A

Instructor: Xiaojing Wang

Neural Networks and Deep Learning deeplearning.ai

Instructor: Andrew Ng 🗹

Applied Machine Learning

UoE. Grade: A

Instructor: Oisin Mac Aodha 🗹