

SPENCER FREI

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RESEARCH INTERESTS

Machine learning, with a particular interest in understanding and improving deep learning.

EMPLOYMENT

Tenure-track Assistant Professor, Department of Statistics, University of California, Davis 2023–present
Postdoctoral Fellow, Simons Institute for the Theory of Computing, UC Berkeley 2021–2023
— Mentors: Peter Bartlett and Bin Yu.
— Part of the [NSF/Simons Collaboration on the Theoretical Foundations of Deep Learning](#).
— Visitor at École polytechnique fédérale de Lausanne (EPFL), Sep. & Oct. 2022, hosted by Emmanuel Abbé.

EDUCATION

Ph.D, Statistics, UCLA 2015–2021
— Co-advisors: Quanquan Gu and Ying Nian Wu; committee members: Arash Amini and Qing Zhou.
MSc., Mathematics, University of British Columbia, Vancouver 2013–2015
— Advisor: Edwin A. Perkins.
BSc., Mathematics, McGill University, Montréal 2009–2013
— First class honours.

PUBLICATIONS AND PREPRINTS

1. Zhiwei Xu, Yutong Wang, Spencer Frei, Gal Vardi, and Wei Hu. Benign overfitting and grokking in ReLU networks for XOR cluster data. *Preprint*, arXiv:2310.02541.
2. Nikhil Ghosh, **Spencer Frei**, Wooseok Ha, and Bin Yu. The effect of SGD batch size on autoencoder learning: Sparsity, sharpness and feature learning. *Preprint*, arXiv:2308.03215.
3. Ruiqi Zhang, **Spencer Frei**, and Peter L. Bartlett. Trained transformers learn linear models in-context. *Preprint*, arXiv:2306.09927.
4. **Spencer Frei***, Gal Vardi*, Peter L. Bartlett, and Nathan Srebro. The double-edged sword of implicit bias: Generalization vs. robustness in ReLU networks. *Advances in Neural Information Processing Systems (NeurIPS)*, 2023.
5. **Spencer Frei***, Gal Vardi*, Peter L. Bartlett, and Nathan Srebro. Benign overfitting in linear classifiers and leaky ReLU networks from KKT conditions for margin maximization. *Conference on Learning Theory (COLT)*, 2023.
6. **Spencer Frei***, Gal Vardi*, Peter L. Bartlett, Nathan Srebro, and Wei Hu. Implicit bias in leaky ReLU networks trained on high-dimensional data. *International Conference on Learning Representations (ICLR)*, 2023. (**Spotlight**)
7. **Spencer Frei**, Niladri Chatterji, and Peter L. Bartlett. Random feature amplification: Feature learning and generalization in neural networks. *Journal of Machine Learning Research*, to appear.
8. **Spencer Frei**, Niladri Chatterji, and Peter L. Bartlett. Benign overfitting without linearity: Neural network classifiers trained by gradient descent for noisy linear data. *Conference on Learning Theory (COLT)*, 2022.

9. **Spencer Frei***, Difan Zou*, Zixiang Chen*, and Quanquan Gu. Self-training converts weak learners to strong learners in mixture models. *International Conference on Artificial Intelligence and Statistics (AISTATS)*, 2022.
10. **Spencer Frei** and Quanquan Gu. Proxy convexity: A unified framework for the analysis of neural networks trained by gradient descent. *Advances in Neural Information Processing Systems (NeurIPS)*, 2021.
11. Difan Zou*, **Spencer Frei***, and Quanquan Gu. Provable robustness of adversarial training for learning halfspaces with noise. *International Conference on Machine Learning (ICML)*, 2021.
12. **Spencer Frei**, Yuan Cao, and Quanquan Gu. Provable generalization of SGD-trained neural networks of any width in the presence of adversarial label noise. *International Conference on Machine Learning (ICML)*, 2021.
13. **Spencer Frei**, Yuan Cao, and Quanquan Gu. Agnostic learning of halfspaces with gradient descent via soft margins. *International Conference on Machine Learning (ICML)*, 2021. **(Long Talk)**
14. **Spencer Frei**, Yuan Cao, and Quanquan Gu. Agnostic learning of a single neuron with gradient descent. *Advances in Neural Information Processing Systems (NeurIPS)*, 2020.
15. Ariana E. Anderson, Mirella Diaz-Santos, **Spencer Frei et al.** Hemodynamic latency is associated with reduced intelligence across the lifespan: an fMRI DCM study of aging, cerebrovascular integrity, and cognitive ability. *Brain Structure and Function*, 2020.
16. **Spencer Frei**, Yuan Cao, and Quanquan Gu. Algorithm-dependent generalization bounds for overparameterized deep residual networks. *Advances in Neural Information Processing Systems (NeurIPS)*, 2019.
17. **Spencer Frei** and Edwin Perkins. A lower bound for p_c in range- R bond percolation in two and three dimensions. *Electronic Journal of Probability* 21(56), 2016.
18. **Spencer Frei**, Kathryn Lockwood, Greg Stewart, Justin Boyer, and Burt S. Tilley, On thermal resistance in concentric residential geothermal heat exchangers. *Journal of Engineering Mathematics* 86(1), 2014.

* indicates equal contribution.

INDUSTRY EXPERIENCE

Applied Scientist Intern, Amazon Alexa AI, Cambridge, MA	Summer 2020
— Worked on natural language understanding using Transformer-based multilingual language models.	
Student Researcher, Chatterbaby/UCLA School of Medicine, Los Angeles, CA	2018–2020
— Developed deep learning models for audio analysis of infant cries that were deployed in Chatterbaby app (100k+ downloads on Android). Employed 20 hr/week when not teaching.	
Biostatistical Consultant, Ritter Pharmaceuticals, Los Angeles, CA	2017–2019
— Consulted on the analysis of clinical trial data using linear regression and mixed effects models.	
Student Researcher, Blackthorn Therapeutics/UCLA School of Medicine, Los Angeles, CA	2016–2018
— Analyzed MRI and neuropsychiatric data using generalized linear models for a biotech company-sponsored project. Employed 20 hr/week when not teaching.	

HONORS

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- Rising Star in Machine Learning, University of Maryland, 2022.
 - Best Reviewer Award: ICML 2020, ICML 2021, NeurIPS 2021, ICLR 2022.
 - Dissertation Year Fellowship, UCLA, 2020–2021.
 - Most Promising Computational Statistician, UCLA, 2016.
 - Research Fellowship, Montréal Institut des sciences mathématiques, 2012.

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- “Reconsidering Overfitting in the Age of Overparameterized Models” (three hour tutorial)
— NeurIPS 2023 Tutorial, presented with Vidya Muthukumar and Fanny Yang *December 2023*

 - “Learning Linear Models In-Context with Transformers”
— Apple Machine Learning Research, Cupertino *November 2023*
— University of Basel, Aurelien Lucchi group meeting *November 2023*
— University of Oxford, Computational Statistics and Machine Learning Seminar *October 2023*
— University of Cambridge, Machine Learning Group *October 2023*
— Google DeepMind, London *October 2023*
— Imperial College London, Imperial + AI Seminar *October 2023*
— The Alan Turing Institute, London, Rough Paths Interest Group^v *October 2023*
— Stanford University, Tengyu Ma group meeting *September 2023*
— Google Research, In-Context Learning Reading Group^v *August 2023*

 - “Benign Overfitting from KKT Conditions for Margin Maximization”
— Conference on Learning Theory, Bangalore *July 2023*

 - “Implicit Regularization and Benign Overfitting for Neural Networks in High Dimensions”
— Youth in High Dimensions Workshop, International Center for Theoretical Physics, Trieste *May 2023*
— UC Berkeley, Department of Biostatistics Seminar *April 2023*
— University of British Columbia, Mathematics of Information, Learning, and Data Seminar *January 2023*
— University of Alberta, Dept. of Mathematical and Statistical Sciences, Statistics Seminar *October 2022*
— EPFL, Fundamentals of Learning and Artificial Intelligence Seminar *September 2022*

 - “Statistical and Computational Phenomena in Deep Learning”
— University of Wisconsin–Madison, Department of Statistics Seminar^v *February 2023*
— University of British Columbia, Department of Statistics Seminar *February 2023*
— University of Illinois, Urbana-Champaign, Department of Statistics Seminar *February 2023*
— University of Michigan, Department of Statistics Seminar *January 2023*
— University of British Columbia, Department of Mathematics Colloquium *January 2023*
— Northwestern University, Department of Statistics and Data Science Seminar^v *January 2023*
— University of California, Davis, Department of Statistics Seminar *January 2023*
— University of California, Davis, Department of Mathematics Colloquium *January 2023*
— Boston University, Department of Mathematics and Statistics Seminar *January 2023*
— University of California, San Diego, Department of Mathematics Colloquium *December 2022*
— University of California, Irvine, Department of Statistics Seminar *December 2022*

 - “Statistical Learning Theory, Optimization, and Neural Networks” (two hour tutorial)
— Simons Institute, Deep Learning Theory Workshop & Summer School *August 2022*

 - “Benign Overfitting without Linearity”
— University of Southern California, Symposium on Frontiers of Machine Learning and AI *November 2022*
— Joint Statistical Meetings, Washington, D.C. *August 2022*
— Conference on Learning Theory, London *July 2022*
— ETH Zürich, Data, Algorithms, Combinatorics, & Optimization Seminar *June 2022*
— Harvard University, Probabilistic Seminar^v *May 2022*
— University of Toronto, Statistics Research Day^v *May 2022*
— University of British Columbia, Christos Thrampoulidis group meeting^v *April 2022*
— Theory of Overparameterized Machine Learning Workshop^v *April 2022*
— Google Research, Algorithms Seminar^v *March 2022*
— Oxford University, Yee Whye Teh group meeting^v *March 2022*
— NSF/Simons Mathematics of Deep Learning Seminar^v *March 2022*

- “Random Feature Amplification: Feature Learning and Generalization in Neural Networks”
— EPFL, Nicolas Flammarion group meeting October 2022
- Microsoft Research, Machine Learning Foundations Seminar[✓] April 2022
- Columbia University, Daniel Hsu group meeting[✓] April 2022
- Theory of Overparameterized Machine Learning Workshop[✓] April 2022
- “Self-Training Converts Weak Learners to Strong Learners in Mixture Models”
— Simons Institute for the Theory of Computing, Deep Learning Theory Symposium December 2021
- “Proxy Convexity: A Unified Optimization Framework For Neural Networks Trained by Gradient Descent”
— Simons Institute for the Theory of Computing, Meet the Fellows Welcome Event September 2021
- “Generalization of SGD-Trained Neural Networks in the Presence of Adversarial Label Noise”
— ETH Zürich, Young Data Science Researchers Seminar[✓] April 2021
- Johns Hopkins University, Machine Learning Seminar[✓] April 2021
- Max-Planck-Institute MiS, Machine Learning Seminar[✓] March 2021
- NSF/Simons Mathematics of Deep Learning Seminar[✓] February 2021

PROFESSIONAL SERVICE

- Workshop co-organizer: Deep Learning Theory Workshop & Summer School, Simons Institute for the Theory of Computing, Berkeley, 2022.
- Area Chair/Senior PC for conferences: NeurIPS 2023, ALT 2024.
- Reviewer for journals: *Annals of Statistics*, *Journal of Machine Learning Research*, *SIAM Journal on Mathematics of Data Science*, *Neural Computation*, *Mathematics of Operations Research*, *Transactions on Machine Learning Research*.
- Reviewer for conferences: ICML 2020, NeurIPS 2020, AISTATS 2021, ICML 2021, NeurIPS 2021, ICLR 2022, AISTATS 2022, ICML 2022, ICLR 2023, COLT 2023.
- Reviewer for workshops: Theory of Overparameterized Machine Learning (TOPML) 2021, ICML Workshop on Overparameterization: Pitfalls & Opportunities (ICMLOPPO) 2021, TOPML 2022, NeurIPS Workshop on Mathematics of Modern Machine Learning (M3L) 2023, NeurIPS Workshop on Robustness of Few-Shot Learning in Foundation Models (R0-FoMo) 2023.
- Reviewer, ENVISION Research Competition for Women in STEM, 2022.
- Volunteer for Queer in AI.

CONFERENCE, WORKSHOP, AND PROGRAM PARTICIPATION

- Neural Information Processing Systems. New Orleans, 2023.
- Conference on Learning Theory. Bangalore, 2023.
- Youth in High Dimensions. Trieste, 2023.
- Joint Statistical Meetings. Washington, DC, 2022.
- Deep Learning Theory Summer Cluster. Simons Institute for the Theory of Computing, Berkeley, 2022.
- Conference on Learning Theory. London, 2022.
- Deep Learning Theory Symposium. Simons Institute for the Theory of Computing, Berkeley, 2021.
- Mathematical/Scientific Foundations of Deep Learning Annual Meeting. Simons Foundation, New York, 2021.
- ICML Workshop on Overparameterization: Pitfalls and Opportunities. Online, 2021.
- Theory of Overparameterized Machine Learning (TOPML) Workshop. Rice University (online), 2021; 2022.
- Theory of Deep Learning Special Quarter. TTIC/Northwestern Institute for Data, Econometrics, Algorithms, and Learning (online) 2020.
- Neural Information Processing Systems. Vancouver, BC, 2019.
- Emerging Challenges in Deep Learning Workshop. Simons Institute, Berkeley, 2019.
- Summer School in Probability. Pacific Institute for the Mathematical Sciences, Vancouver, BC, 2014.

TEACHING EXPERIENCE

- UC Davis, Department of Statistics
 - Instructor, Stats 35B: Statistical Data Science II, Winter 2024.
 - Instructor, Stats 250: Deep Learning Theory, Winter 2024.
 - Organizer, Stats 290: Seminar in Statistics. Winter 2024.
- UCLA, Department of Statistics
 - TA, Stats 100C: Linear Models, Spring 2020.
 - TA, Stats 102C: Monte Carlo Methods, Fall 2019.
 - TA, Stats 100B: Mathematical Statistics, Winter 2016.
 - TA, Stats 100A: Probability Theory, Fall 2016.
 - TA, Stats 10: Intro to Statistics, Summer 2016.

PROGRAMMING LANGUAGES AND SKILLS

Python (including PyTorch & TensorFlow), R, Matlab.