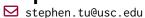
Stephen Tu



https://stephentu.github.io

Research Summary

I study problems at the intersection of machine learning, optimization, and control theory. My recent work focuses on developing a principled understanding of using machine learning models for data-driven control, with specific emphasis on applications in robotics.

Employment History

2024 – Now	Assistant Professor Ming Hsieh Department of Electrical and Computer Engineering Thomas Lord Department of Computer Science (by courtesy) University of Southern California
2019 – 2023	Research Scientist Google DeepMind Robotics New York, NY
2014 – 2019	Research Assistant Department of Electrical Engineering and Computer Sciences University of California, Berkeley
2011 – 2014	Research Assistant Department of Electrical Engineering and Computer Science Massachusetts Institute of Technology

Education

2014 – 2019	Ph.D., EECS , University of California, Berkeley Thesis title: Sample Complexity Bounds for the Linear Quadratic Regulator Advisor: Prof. Benjamin Recht
2011 – 2014	S.M., EECS , Massachusetts Institute of Technology Thesis title: <i>Fast Transactions for Multicore In-Memory Databases</i> Advisor: Prof. Samuel Madden
2006 – 2011	B.A., Computer Science, University of California, Berkeley
	B.S., Mechanical Engineering, University of California, Berkeley

Research Funding

9/25-9/26	2025 Okawa Foundation Research Grant
	Role: Solo-PI
	Amount: 10k
8/25-8/28	NSF CPS Medium: Latent Representation Learning for Verifiable Sensor Rich Systems
	Role: Lead-PI (Co-PIs: Lars Lindemann, Yue Wang)
	Amount: 400k of 1.2M
8/25-8/26	Generative AI for Financial Decision-Making in Noisy, Dynamic, and Agentic Settings
	Sponsor: USC-Capital One Center for Responsible AI Decision Making in Finance (CREDIF)
	Role: Co-PI (Lead-PI: Mahdi Soltanolkotabi)
	Amount: 20k

Research Publications

Preprints

- [1] A. Liang, P. Czempin, M. Hong, Y. Zhou, E. Bıyık, and **S. Tu**, CLAM: Continuous latent action models for robot learning from unlabeled demonstrations, Preprint, 2025. arXiv: arXiv: 2505.04999 [cs.R0].
- [2] E. Shekhtman, Y. Zhou, I. Ziemann, N. Matni, and **S. Tu**, *Nearly instance-optimal parameter recovery from many trajectories via hellinger localization*, Preprint, 2025. arXiv: arXiv: 2510.06434 [cs.LG].

Journal Articles

- [1] A. A. Ahmadi, A. Chaudhry, V. Sindhwani, and **S. Tu**, "Safely learning dynamical systems," *Foundations of Computational Mathematics*, 2025, ISSN: 1615-3383.
- [2] L. Lindemann, A. Robey, L. Jiang, S. Das, **S. Tu**, and N. Matni, "Learning robust output control barrier functions from safe expert demonstrations," *IEEE Open Journal of Control Systems*, vol. 3, pp. 158–172, 2024.
- [3] S. Singh, **S. Tu**, and V. Sindhwani, "Revisiting energy based models as policies: Ranking noise contrastive estimation and interpolating energy models," *Transactions on Machine Learning Research*, 2024, ISSN: 2835-8856.
- [4] **S. Tu**, R. Frostig, and M. Soltanolkotabi, "Learning from many trajectories," *Journal of Machine Learning Research*, vol. 25, no. 216, pp. 1–109, 2024.
- [5] N. M. Boffi, **S. Tu**, and J.-J. E. Slotine, "Nonparametric adaptive control and prediction: Theory and randomized algorithms," *Journal of Machine Learning Research*, vol. 23, no. 281, pp. 1–46, 2022.
- [6] S. Dean, H. Mania, N. Matni, B. Recht, and **S. Tu**, "On the sample complexity of the linear quadratic regulator," *Foundations of Computational Mathematics*, vol. 20, no. 4, pp. 633–679, 2020.

Conference Proceedings

- [1] N. M. Boffi, A. Jacot, **S. Tu**, and I. Ziemann, "Single hidden layer diffusion models provably learn simple low-dimensional structure," in *The Thirteenth International Conference on Learning Representations*, 2025.
- [2] E. Dietrich, R. Devonport, **S. Tu**, and M. Arcak, "Data-driven reachability with scenario optimization and the holdout method," in 2025 IEEE 64th Conference on Decision and Control (to appear), 2025.
- [3] P. Lutkus, D. Anantharaman, **S. Tu**, and L. Lindemann, "Incremental composition of learned control barrier functions in unknown environments," in 2025 IEEE 64th Conference on Decision and Control (to appear), 2025.
- [4] P. Lutkus, K. Wang, L. Lindemann, and **S. Tu**, "Latent representations for control design with provable stability and safety guarantees," in 2025 IEEE 64th Conference on Decision and Control (to appear), 2025.
- [5] S. Park and **S. Tu**, "Integration matters for learning PDEs with backwards SDEs," in *Advances in Neural Information Processing Systems (to appear)*, **(Spotlight)**, 2025.
- [6] I. Ziemann, **S. Tu**, G. J. Pappas, and N. Matni, "Sharp rates in dependent learning theory: Avoiding sample size deflation for the square loss," in *Proceedings of the 41st International Conference on Machine Learning*, (**Spotlight**), vol. 235, PMLR, 2024.
- [7] S. Abeyruwan, A. Bewley, N. M. Boffi, K. M. Choromanski, D. B. D'Ambrosio, D. Jain, P. R. Sanketi, A. Shankar, V. Sindhwani, S. Singh, J.-J. Slotine, and **S. Tu**, "Agile catching with whole-body mpc and blackbox policy learning," in *Proceedings of The 5th Annual Learning for Dynamics and Control Conference*, vol. 211, PMLR, 2023, pp. 851–863.
- [8] D. Brandfonbrener, **S. Tu**, A. Singh, S. Welker, C. Boodoo, N. Matni, and J. Varley, "Visual backtracking teleoperation: A data collection protocol for offline image-based reinforcement learning," in 2023 IEEE International Conference on Robotics and Automation (ICRA), 2023.

- [9] C. Le Lan, **S. Tu**, M. Rowland, A. Harutyunyan, R. Agarwal, M. G. Bellemare, and W. Dabney, "Bootstrapped representations in reinforcement learning," in *Proceedings of the 40th International Conference on Machine Learning*, vol. 202, PMLR, 2023.
- [10] D. Pfrommer, M. Simchowitz, T. Westenbroek, N. Matni, and **S. Tu**, "The power of learned locally linear models for nonlinear policy optimization," in *Proceedings of the 40th International Conference on Machine Learning*, vol. 202, PMLR, 2023.
- [11] A. Z. Ren, A. Dixit, A. Bodrova, S. Singh, **S. Tu**, N. Brown, P. Xu, L. Takayama, F. Xia, J. Varley, Z. Xu, D. Sadigh, A. Zeng, and A. Majumdar, "Robots that ask for help: Uncertainty alignment for large language model planners," in 7th Annual Conference on Robot Learning, (Best Student Paper), 2023.
- [12] T. T. Zhang, K. Kang, B. D. Lee, C. Tomlin, S. Levine, **S. Tu**, and N. Matni, "Multi-task imitation learning for linear dynamical systems," in *Proceedings of The 5th Annual Learning for Dynamics and Control Conference*, vol. 211, PMLR, 2023, pp. 586–599.
- [13] I. Ziemann, **S. Tu**, G. J. Pappas, and N. Matni, "The noise level in linear regression with dependent data," in *Advances in Neural Information Processing Systems*, vol. 36, 2023.
- [14] N. Boffi, **S. Tu**, and J.-J. Slotine, "The role of optimization geometry in single neuron learning," in *Proceedings of The 25th International Conference on Artificial Intelligence and Statistics*, vol. 151, PMLR, 2022.
- [15] C. Le Lan, **S. Tu**, A. Oberman, R. Agarwal, and M. G. Bellemare, "On the generalization of representations in reinforcement learning," in *Proceedings of The 25th International Conference on Artificial Intelligence and Statistics*, vol. 151, PMLR, 2022.
- [16] D. Pfrommer, T. T. Zhang, **S. Tu**, and N. Matni, "Tasil: Taylor series imitation learning," in *Advances in Neural Information Processing Systems*, vol. 35, 2022.
- [17] **S. Tu**, A. Robey, T. Zhang, and N. Matni, "On the sample complexity of stability constrained imitation learning," in *Proceedings of The 4th Annual Learning for Dynamics and Control Conference*, PMLR, 2022.
- [18] X. Xiao, T. Zhang, K. Choromanski, E. Lee, A. Francis, J. Varley, **S. Tu**, S. Singh, P. Xu, F. Xia, S. M. Persson, D. Kalashnikov, L. Takayama, R. Frostig, J. Tan, C. Parada, and V. Sindhwani, "Learning model predictive controllers with real-time attention for real-world navigation," in *Proceedings of the 2022 Conference on Robot Learning*, PMLR, 2022.
- [19] T. Zhang, **S. Tu**, N. Boffi, J.-J. Slotine, and N. Matni, "Adversarially robust stability certificates can be sample-efficient," in *Proceedings of The 4th Annual Learning for Dynamics and Control Conference*, PMLR, 2022.
- [20] I. Ziemann and **S. Tu**, "Learning with little mixing," in *Advances in Neural Information Processing Systems*, vol. 35, 2022.
- [21] A. A. Ahmadi, A. Chaudhry, V. Sindhwani, and **S. Tu**, "Safely learning dynamical systems from short trajectories," in *Proceedings of the 3rd Conference on Learning for Dynamics and Control*, vol. 144, PMLR, 2021.
- [22] N. M. Boffi, **S. Tu**, and J.-J. Slotine, "Nonparametric adaptive control and prediction: Theory and randomized algorithms," in 2021 60th IEEE Conference on Decision and Control (CDC), 2021, pp. 2935–2942.
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- [27] A. Robey, H. Hu, L. Lindemann, H. Zhang, D. V. Dimarogonas, **S. Tu**, and N. Matni, "Learning control barrier functions from expert demonstrations," in 2020 59th IEEE Conference on Decision and Control (CDC), 2020, pp. 3717–3724.
- [28] X. Song, Y. Jiang, **S. Tu**, Y. Du, and B. Neyshabur, "Observational overfitting in reinforcement learning," in *International Conference on Learning Representations*, 2020.
- [29] S. Dean, **S. Tu**, N. Matni, and B. Recht, "Safely learning to control the constrained linear quadratic regulator," in *2019 American Control Conference (ACC)*, 2019, pp. 5582–5588.
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Talks

10/25 Learning from many trajectories

IFML Seminar, UT Austin

Fa. 25 Challenges in Learning Behavior Certificates from Data

CCDC Seminar, UC Santa Barbara

Statistical Uncertainty Quantification in the Era of AI-Enabled Robots Workshop, RSS 2025

- 3/25 On the Effectiveness of Generative Modeling for Planning and Control Computer Science Colloquium, University of Southern California CRI Seminar, UC San Diego
- 2/25 **Shallow diffusion networks provably learn hidden low-dimensional structure** Information Theory and Applications, 2025
- 2/24 **Towards instance optimal rates in learning from dependent data** Information Theory and Applications, 2024
- 11/23 **Revisiting Energy Based Models as Policies**Stanford Vision and Learning Lab, Stanford University
- 6/23 **jax4dc Tutorial**Learning for Dynamics & Control Conference, 2023
- Sp. 23 The foundations of machine learning for feedback control

Computer Science Colloquium, UC Riverside

EECS Colloquium, UC Irvine

Computer Science Colloquium, University of Chicago

ECE Colloquium, University of Michigan

ECE Colloquium, University of Southern California

Talks (continued)

2022–23	Learning from many trajectories Information Theory and Applications, 2023 Learning Theory for Control Workshop, IEEE Conference on Decision and Control, 2022 Machine Learning in Feedback Systems (CS 6784) Guest Lecture, Cornell University
	Intelligent Robot Motion Lab, Princeton University
2022-23	Introduction to High Dimensional Probability (Part I and II) Google Brainiversity Lecture Series
11/22	Independent learning from sequential data Foundations of AI Seminar, University of Southern California
6/21	Learning and Control with Safety and Stability Guarantees for Nonlinear Systems Foundations and mathematical guarantees of data-driven control summer school, ETHZ-EPFL
Fa. 20	Learning Stability Certificates from Data CAN Lab, New York University Research Seminar, TU Darmstadt
12/19	Concentration Inequalities for System Identification Tutorial Session, IEEE Conference on Decision and Control, 2019 Seminar, Lund University
10/19	Finite-time Analysis of Approximate Policy Iteration for the Linear Quadratic Regulator INFORMS Conference, 2019
7/19	Minimax Lower Bounds for H_{∞} -Norm Estimation American Control Conference, 2019
Su. 19	The Gap Between Model-Based and Model-Free Methods on the Linear Quadratic Regulator: An Asymptotic Viewpoint International Conference on Continuous Optimization, 2019 Conference on Learning Theory, 2019
Sp. 19	Safe and Reliable Reinforcement Learning for Continuous Control Amazon Research, Palo Alto Facebook AI Research, Menlo Park Robot Locomotion Group, Massachusetts Institute of Technology Machine Learning Seminar, University of Washington Google DeepMind, London Google Brain Robotics, Mountain View Computer Science Colloquium, Princeton University Microsoft Research, New York City
6/18	On the Approximation of Toeplitz Operators for Nonparametric H_∞ -norm Estimation American Control Conference, 2018
5/18	Finite Sample Guarantees for Control of an Unknown Linear Dynamical System Stanford Information Systems Laboratory, Stanford University
Fa. 17	A Lyapunov Analysis for Accelerated Block Gauss-Seidel Simons Institute, UC Berkeley
8/17	Breaking Locality Accelerates Block Gauss-Seidel International Conference on Machine Learning, 2017
5/17	Convergence and Geometry of Non-convex Matrix Sensing SIAM Conference on Optimization, 2017
6/16	Low-rank Solutions of Linear Matrix Equations via Procrustes Flow International Conference on Machine Learning, 2016

Talks (continued)

Speedy Transactions in Multicore In-Memory Databases
Symposium on Operating Systems Principles, 2013
Microsoft Research, Redmond

8/13 **Processing Analytical Queries over Encrypted Data**International Conference on Very Large Data Bases, 2013

Awards and Honorable Mentions

2023	Best Student Paper Award, Conference on Robot Learning 2023
	(co-author on paper with A. Ren, lead student author)
2021	Outstanding Reviewer, NeurIPS 2021
2018	Google PhD Fellowship in Machine Learning
2011	CRA Undergraduate Research Award Honorable Mention

Teaching

EE660	Mathematical Foundations of Machine Learning, University of Southern California
	Semesters: Sp. 2024, Fa. 2024, Fa. 2025
CS189	Introduction to Machine Learning (GSI), University of California, Berkeley
	Semesters: Fa. 2016, Fa. 2018

Service

Ongoing	Reviewer, NeurIPS, ICML, ACC, CDC, AISTATS, COLT, CoRL, L4DC, and JMLR
	Program Co-Chair, Learning for Dynamics and Control Conference, 2026
	Guest Associate Editor, IEEE Open Journal of Control Systems (OJ-CSYS)
2025	Area Chair, NeurIPS, 2025
	Program Committee, Learning for Dynamics and Control Conference, 2025
2024	Co-Organizer, Southern California Control Workshop, Fall 2024
	Session Organizer , Information Theory and Applications, 2024 Session: <i>Statistical learning for analysis and design of control systems</i>
2023	Area Chair, NeurIPS, 2023
2022	Mentor, Google CS Research Mentorship Program
2018	Mentor , BAIR Undergraduate Mentoring Program

References

Available on Request