

NICHOLAS M. BOFFI: CURRICULUM VITAE

Department of Mathematical Sciences & Machine Learning Department
Carnegie Mellon University
Pittsburgh, PA 15213

nboffi@andrew.cmu.edu
<https://nmboffi.github.io>

EDUCATION

Harvard University 2015–2021
Ph.D. Applied Mathematics
Advisors: Jean-Jacques E. Slotine (MIT) and Chris H. Rycroft (Harvard)
Thesis title: *Methods for scientific simulation, machine learning, and nonlinear control*

Northwestern University 2010–2014
B.A. Mathematics, Physics, and Integrated Science *with honors*
Advisor: Tamar Seideman
Thesis title: *High harmonic generation from simple aromatic molecules*

PROFESSIONAL POSITIONS

Carnegie Mellon University 2025–present
Assistant Professor, Machine Learning Department (50%)
Assistant Professor, Department of Mathematical Sciences (50%)

Carnegie Mellon University 2024–2025
Assistant Professor, Department of Mathematical Sciences
Affiliated Faculty, Machine Learning Department

Courant Institute of Mathematical Sciences 2021–2024
Courant Instructor / Assistant Professor

Google Brain 2020–2021
Research Intern / Student Researcher
Advisor: Vikas Sindhwani

Massachusetts Institute of Technology 2016–2021
Visiting Graduate Student Researcher
Advisor: Jean-Jacques Slotine

Lawrence Berkeley National Lab 2016
Computational Science Graduate Fellowship Practicum
Advisor: Adam Arkin

Tel Aviv University 2014–2015
Fulbright Research Scholar
Advisor: Amir Natan

RESEARCH INTERESTS

Broadly: machine learning, computational mathematics

Specifically: generative modeling, dynamical transport of measure, stochastic thermodynamics, active matter, partial differential equations, adaptive control and learning, optimal control, dynamical systems, deep learning, optimization, numerical analysis, elasticity, continuum mechanics, electronic structure

HONORS AND AWARDS

| | |
|---|-----------|
| NSF Postdoctoral Fellowship in the Mathematical Sciences, <i>declined</i> | 2021 |
| Harvard University Certificate of Distinction in Teaching | 2016 |
| Department of Energy Computational Science Graduate Research Fellowship | 2015–2019 |
| Fulbright Research Fellowship | 2014 |
| Honorary Cambridge Trust Fellowship, <i>declined</i> | 2014 |
| University of Chicago McCormick Fellowship, <i>declined</i> | 2014 |
| Cornell Graduate Fellowship, <i>declined</i> | 2014 |
| Phi Beta Kappa, Northwestern Chapter | 2014 |
| Hypercube Award for excellence in theoretical chemistry research, Northwestern University | 2013 |
| Fletcher Undergraduate Research Award, Northwestern University, <i>Finalist</i> | 2012 |

PREPRINTS

Chanhyuk Lee, Jaehoon Yoo, Manan Agarwal, Sheel Shah, Jerry Huang, Aditi Raghunathan, Seunghoon Hong, [Nicholas M. Boffi[†]](#), and Jinwoo Kim[†]. One-step language modeling via continuous denoising. *arXiv:2602.16813*, 2026.

Peter Holderrieth, Douglas Chen, Luca Eyting, Ishin Shah, Giri Anantharaman, Yutong He, Zeynep Akata, Tommi Jaakkola, [Nicholas M. Boffi[†]](#), and Max Simchowitz[†]. Diamond maps: Efficient reward alignment via stochastic flow maps. *arXiv:2602.05993*, 2026.

Naman Choudhary, Vedant Singh, Ameet Talwalkar, [Nicholas M. Boffi](#), Mikhail Khodak, and Tanya Marwah. Pre-generating multi-difficulty PDE data for few-shot neural PDE solvers. *arXiv:2512.00564*, 2025.

Amirmojtaba Sabour, Michael S. Albergo, Carles Domingo-Enrich, [Nicholas M. Boffi](#), Sanja Fidler, Karsten Kreis, and Eric Vanden-Eijnden. Test-time scaling of diffusions with flow maps. *arXiv:2511.22688*, 2025.

Matthew S. Zhang, Stephen Huan, Jerry Huang, [Nicholas M. Boffi](#), Sitan Chen, and Sinho Chewi. Sublinear iterations can suffice even for DDPMs. *arXiv:2511.04844*, 2025.

Burak Varıcı, Che-Ping Tsai, Ritabrata Ray, [Nicholas M. Boffi](#), and Pradeep Ravikumar. Eigenfunction extraction for ordered representation learning. *arXiv:2510.24672*, 2025.

PUBLICATIONS

Xinyue Ai, Yutong He, Albert Gu, Ruslan Salakhutdinov, J. Zico Kolter, [Nicholas M. Boffi](#), and Max Simchowitz. Joint distillation for fast likelihood evaluation and sampling in flow-based models. In *International Conference on Learning Representations*, 2026.

Chaoyi Pan, Giri Anantharaman, Nai-Chieh Huang, Claire Jin, Daniel Pfrommer, Chenyang Yuan, Frank Permenter, Guannan Qu[†], [Nicholas M. Boffi[†]](#), Guanya Shi[†], and Max Simchowitz[†]. Much ado about noising: Dispelling the myths of generative robotic control. In *International Conference on Learning Representations*, 2026.

Oscar Davis, Michael S. Albergo, [Nicholas M. Boffi](#), Michael M. Bronstein, and Avishek Joey Bose. Generalised flow maps for few-step generative modelling on Riemannian manifolds. In *International Conference on Learning Representations*, 2026.

Rishal Aggarwal, Jacky Chen, [Nicholas M. Boffi](#), and David Ryan Koes. BoltzNCE: Learning likelihoods for Boltzmann generation with stochastic interpolants and noise contrastive estimation. In *Advances in Neural Information Processing Systems*, 2025.

Nicholas M. Boffi, Michael S. Albergo, and Eric Vanden-Eijnden. How to build a consistency model: Learning flow maps via self-distillation. In *Advances in Neural Information Processing Systems*, 2025.

Nicholas M. Boffi and Eric Vanden-Eijnden. Model-free learning of probability flows: Elucidating the nonequilibrium dynamics of flocking. *Physical Review Letters*, 135(23):238301, 2025.

Nicholas M. Boffi, Arthur Jacot, Stephen Tu, and Ingvar Ziemann. Shallow diffusion networks provably learn hidden low-dimensional structure. In *International Conference on Learning Representations*, 2025.

Michael S. Albergo*, Nicholas M. Boffi*, and Eric Vanden-Eijnden. Stochastic Interpolants: A Unifying Framework for Flows and Diffusions. *Journal of Machine Learning Research*, 26(209):1–80, 2025.

Nicholas M. Boffi*, Michael S. Albergo*, and Eric Vanden-Eijnden. Flow map matching with stochastic interpolants: A mathematical framework for consistency models. *Transactions on Machine Learning Research*, 2025.

Nanye Ma, Mark Goldstein, Michael S. Albergo, Nicholas M. Boffi, Eric Vanden-Eijnden, and Saining Xie. Sit: Exploring flow and diffusion-based generative models with scalable interpolant transformers. In *Computer Vision – ECCV 2024*, pages 23–40. Springer Nature Switzerland, 2024.

Yifan Chen, Mark Goldstein, Mengjian Hua, Michael S. Albergo, Nicholas M. Boffi, and Eric Vanden-Eijnden. Probabilistic forecasting with stochastic interpolants and Föllmer processes. In *International Conference on Machine Learning*, 2024.

Michael S. Albergo, Mark Goldstein, Nicholas M. Boffi, Rajesh Ranganath, and Eric Vanden-Eijnden. Stochastic interpolants with data-dependent couplings. In *International Conference on Machine Learning*, 2024.

Nicholas M. Boffi and Eric Vanden-Eijnden. Deep learning probability flows and entropy production rates in active matter. *Proceedings of the National Academy of Sciences*, 121(25):e2318106121, June 2024.

Michael Samuel Albergo, Nicholas M. Boffi, Michael Lindsey, and Eric Vanden-Eijnden. Multimarginal generative modeling with stochastic interpolants. In *International Conference on Learning Representations*, 2024.

Nicholas M. Boffi, Yipei Guo, Chris H. Rycroft, and Ariel Amir. How microscopic epistasis and clonal interference shape the fitness trajectory in a spin glass model of microbial long-term evolution. *eLife*, 12, 2023.

Nicholas M. Boffi and Eric Vanden-Eijnden. Probability flow solution of the Fokker–Planck equation. *Machine Learning: Science and Technology*, 4(3):035012, 2023.

Saminda Abeyruwan, Alex Bewley, Nicholas M. Boffi, Krzysztof Marcin Choromanski, David B D’Ambrosio, Deepali Jain, Pannag R Sanketi, Anish Shankar, Vikas Sindhwani, Sumeet Singh, Jean-Jacques Slotine, and Stephen Tu. Agile catching with whole-body mpc and blackbox policy learning. In *Proceedings of The 5th Annual Learning for Dynamics and Control Conference*, volume 211 of *Proceedings of Machine Learning Research*, pages 851–863, 2023.

Nicholas M. Boffi*, Stephen Tu*, and Jean-Jacques E. Slotine. Nonparametric adaptive control and prediction: theory and randomized algorithms. *Journal of Machine Learning Research*, 23(281):1–46, 2022.

Thomas Zhang, Stephen Tu, Nicholas M. Boffi, Jean-Jacques Slotine, and Nikolai Matni. Adversarially robust stability certificates can be sample-efficient. In *Proceedings of The 4th Annual Learning for Dynamics and Control Conference*, volume 168 of *Proceedings of Machine Learning Research*, pages 532–545, 2022.

Nicholas M. Boffi*, Stephen Tu*, and Jean-Jacques Slotine. The role of optimization geometry in single neuron learning. In *Proceedings of The 25th International Conference on Artificial Intelligence and Statistics*, volume 151 of *Proceedings of Machine Learning Research*, pages 11528–11549, 2022.

Nicholas M. Boffi*, Stephen Tu*, Nikolai Matni, Jean-Jacques Slotine, and Vikas Sindhwani. Learning stability certificates from data. In *Proceedings of the 2020 Conference on Robot Learning*, volume 155 of *Proceedings of Machine Learning Research*, pages 1341–1350, 2021.

Katiana Kontolati, Darius Alix-Williams, Nicholas M. Boffi, Michael L. Falk, Chris H. Rycroft, and Michael D. Shields. Manifold learning for coarse-graining atomistic simulations: Application to amorphous solids. *Acta Materialia*, 215:117008, 2021.

Nicholas M. Boffi*, Stephen Tu*, and Jean-Jacques Slotine. Nonparametric adaptive control and prediction: Theory and randomized algorithms. In *Proceedings of the 60th IEEE Conference on Decision and Control (CDC)*, pages 2935–2942, 2021.

Nicholas M. Boffi*, Stephen Tu*, and Jean-Jacques E. Slotine. Regret bounds for adaptive nonlinear control (**selected for oral presentation**). In *Proceedings of the 3rd Conference on Learning for Dynamics and Control*, volume 144 of *Proceedings of Machine Learning Research*, pages 471–483, 2021.

Nicholas M. Boffi and Jean-Jacques E. Slotine. Implicit regularization and momentum algorithms in nonlinearly parameterized adaptive control and prediction (**featured on the cover**). *Neural Computation*, 33(3):590–673, 2021.

Nicholas M. Boffi and Chris H. Rycroft. Coordinate transformation methodology for simulating quasistatic elastoplastic solids. *Physical Review E*, 101:053304, 2020.

Nicholas M. Boffi and Chris H. Rycroft. Parallel three-dimensional simulations of quasi-static elastoplastic solids. *Computer Physics Communications*, 257:107254, 2020.

Nicholas M. Boffi and Jean-Jacques E. Slotine. A continuous-time analysis of distributed stochastic gradient. *Neural Computation*, 32(1):36–96, 2020.

Nicholas M. Boffi, Manish Jain, and Amir Natan. Efficient computation of the Hartree–Fock exchange in real-space with projection operators. *Journal of Chemical Theory and Computation*, 12(8):3614–3622, 2016.

Nicholas M. Boffi, Manish Jain, and Amir Natan. Asymptotic behavior and interpretation of virtual states: The effects of confinement and of basis sets. *The Journal of Chemical Physics*, 144(8):084104, 2016.

Nicholas M. Boffi, Judith C. Hill, and Matthew G. Reuter. Characterizing the inverses of block tridiagonal, block Toeplitz matrices. *Computational Science & Discovery*, 8(1):015001, 2014.

Matthew G. Reuter, Nicholas M. Boffi, Mark A. Ratner, and Tamar Seideman. The role of dimensionality in the decay of surface effects. *The Journal of Chemical Physics*, 138(8):084707, 2013.

SOFTWARE

Author of `flow-maps`, official code for flow map self-distillation (NeurIPS 2025) 2025
<https://github.com/nmboffi/flow-maps>

Author of `structured_diffusion_public`, code for structured diffusion learning (ICLR 2025) 2025
https://github.com/nmboffi/structured_diffusion_public

Author of `jax-interpolants`, a clean implementation of the stochastic interpolant framework 2025
<https://github.com/nmboffi/jax-interpolants>

Author of `flow_map_matching_public`, an implementation of the flow map matching method 2025
https://github.com/nmboffi/flow_map_matching_public

Author of `jax-edm2`, JAX implementation of NVIDIA’s EDM2 network 2025
<https://github.com/nmboffi/jax-edm2>

| | |
|---|------|
| Author of <code>vicsek_public</code> , code for entropy production from trajectories in collective motion https://github.com/nmboffi/vicsek_public | 2024 |
| Author of <code>active_probability_flows</code> , a method for learning physical probability flows https://github.com/nmboffi/active_pflows | 2024 |
| Co-author of <code>stochastic-interpolants</code> , an implementation of the stochastic interpolant method https://github.com/malbergo/stochastic-interpolants | 2023 |
| Author of <code>sbtm</code> , an implementation of the score-based transport modeling algorithm https://github.com/nmboffi/sbtm | 2023 |
| Author of <code>spin_glass_evodyn</code> , a simulation of evolutionary dynamics via spin glass physics https://github.com/nmboffi/spin_glass_evodyn | 2023 |
| Author of <code>stzpp</code> , a simulation of the shear transformation zone theory of amorphous plasticity https://github.com/nmboffi/stzpp | 2021 |
| Contributor to PARSEC, a real-space electronic structure code http://real-space.org/ | 2015 |

TEACHING

Carnegie Mellon University

| | |
|---|-------------|
| Methods of optimization, <i>Instructor</i> | Spring 2025 |
| Introduction to partial differential equations: A computational approach, <i>Instructor</i> | Fall 2024 |

Courant Institute of Mathematical Sciences, New York University

| | |
|--|-------------|
| Honors numerical analysis, <i>Instructor</i> | Spring 2024 |
| Linear and nonlinear optimization, <i>Instructor</i> | Fall 2023 |
| Linear and nonlinear optimization, <i>Instructor</i> | Spring 2023 |
| Numerical analysis, <i>Instructor</i> | Fall 2022 |
| Linear and nonlinear optimization, <i>Instructor</i> | Spring 2022 |
| Numerical analysis, <i>Instructor</i> | Fall 2021 |

Harvard University

| | |
|--|-------------|
| Advanced scientific computing: Numerical methods II, <i>Teaching Fellow</i> | Spring 2021 |
| Learning, estimation, and control of dynamical systems, <i>Teaching Fellow</i> | Spring 2020 |
| Advanced scientific computing: Numerical methods, <i>Teaching Fellow</i> | Fall 2019 |
| Advanced scientific computing: Numerical methods, <i>Teaching Fellow</i> | Fall 2016 |

Northwestern University

| | |
|---|-------------------------|
| Integrated Science Program 101, <i>Instructor</i> | Academic Year 2013–2014 |
| Integrated Science Program 101, <i>Teaching Assistant</i> | Academic Year 2012–2013 |
| Integrated Science Program 101, <i>Teaching Assistant</i> | Spring 2012 |

MENTORING

PhD Students

| | |
|---|--------------|
| Jerry Huang Computer Science Department, Carnegie Mellon University | 2024–Present |
| Stephen Huan Computer Science Department, Carnegie Mellon University <i>Co-advised with Andrej Risteski</i> | 2024–Present |

Undergraduate Students

Ishin Shah 2025–Present
Senior thesis, Carnegie Mellon University
Thesis title: *Generative modeling with transition map matching*
Co-advised with Max Simchowitz

Jimmy Almgren-Bell 2017–2019
Senior thesis, Harvard University
Thesis title: *An agent-based numerical approach to Lenski’s long-term evolution experiment*

Summer Programs

Applied Mathematics Summer Undergraduate Research Experience (AM-SURE) 2022
Program coordinator, Courant Institute of Mathematical Sciences
Mentored ten undergraduate students through summer research projects

PROFESSIONAL ACTIVITIES

Workshop organizer

Theory of AI for Scientific Computing (TASC) June 30, 2025
COLT 2025, Mérieux Amphitheater, ENS Lyon, France
*Jointly organized with Misha Khodak (Princeton), Jianfeng Lu (Duke),
Tanya Marwah (Polymathic AI, Flatiron Institute), and Andrej Risteski (CMU)*

Measure Transport, Diffusion Processes, and Sampling Dec. 4–6, 2023
Flatiron Institute, New York City
*Jointly organized with Michael Albergo (NYU), Bob Carpenter (Flatiron Institute),
Neha Wadia (Flatiron Institute), and Joan Bruna (Courant, Flatiron Institute)*

Reviewer

Annals of Applied Probability, Banff International Research Center, Nature Physics, Journal of Statistical Mechanics: Theory and Experiment, Journal of Computational Physics, SIAM Journal on Scientific Computing, SIAM Journal on Applied Mathematics, Annals of Statistics, International Journal of Robotics Research, Applied and Computational Harmonic Analysis, Journal of Machine Learning Research, Proceedings of the National Academy of Sciences, Physica D: Nonlinear Phenomena, IEEE Transactions on Automatic Control, IEEE Systems & Control Letters, SIAM Journal on Mathematics of Data Science, Neural Information Processing Systems, International Conference on Learning Representations, International Conference on Machine Learning, AISTATS, Learning for Dynamics and Control

Area Chair

Learning for Dynamics and Control 2026

SELECTED TALKS

Flow maps: Generative models with lightning-fast inference 2026
PSU-Purdue-UMD Joint Seminar on Mathematical Data Science

Flow maps: Generative models with lightning-fast inference 2026
CMX Lunch Seminar, California Institute of Technology

Flow maps: Generative models with lightning-fast inference 2025
Workshop on Mathematical Analysis in Machine Learning, University of Pittsburgh

Flow maps: Generative models with lightning-fast inference 2025
Statistics & Data Science Seminar, Carnegie Mellon University

Flow maps: Generative models with lightning-fast inference 2025
Kinetic Theory: Novel Statistical, Stochastic and Analytical Methods
SLMath (Simons Laufer Mathematical Sciences Institute)

| | |
|---|------|
| <i>How to build a consistency model (with stochastic interpolants)</i> | 2025 |
| Workshop on Probabilistic Learning Methods for Inverse Problems Applied Inverse Problems Conference, Rio de Janeiro, Brazil | |
| <i>Stochastic interpolants: from generative modeling to generative science and engineering</i> | 2025 |
| Efficient and Reliable Deep Learning Methods and their Scientific Applications Banff International Research Station | |
| <i>Stochastic interpolants: from generative modeling to generative science and engineering</i> | 2025 |
| Statistical and Computational Challenges in Probabilistic Scientific Machine Learning Institute for Mathematical and Statistical Innovation, University of Chicago | |
| <i>Stochastic interpolants: from generative modeling to generative science and engineering</i> | 2025 |
| Frontiers in Applied Analysis Workshop, Carnegie Mellon University | |
| <i>Stochastic interpolants: from generative modeling to generative science and engineering</i> | 2025 |
| Machine Learning Department Seminar, Carnegie Mellon University | |
| <i>Stochastic interpolants: from generative modeling to generative science and engineering</i> | 2025 |
| Applied Mathematics and Statistics Seminar, Johns Hopkins University | |
| <i>Stochastic interpolants: from generative modeling to generative science and engineering</i> | 2025 |
| Applied and Computational Mathematics Seminar University of Wisconsin-Madison | |
| <i>Flow maps: Generative models with lightning-fast inference</i> | 2025 |
| Advances in Generative AI Conference, Duke University | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2025 |
| Oden Institute & Department of Mathematics Seminar University of Texas at Austin | |
| <i>Stochastic interpolants: A unifying framework for flows and diffusions</i> | 2025 |
| CMU-Pitt Joint Computational Biology Seminar Carnegie Mellon University & University of Pittsburgh | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2025 |
| Continuum Mechanics Seminar, University of Nebraska-Lincoln | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2024 |
| Center for Computational Mathematics Seminar Flatiron Institute | |
| <i>Stochastic interpolants: A unifying framework for flows and diffusions</i> | 2024 |
| Genesis Therapeutics | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2024 |
| Machine Learning / Duolingo Seminar, Carnegie Mellon University | |
| <i>Deep learning the entropy production rate in active matter physics</i> | 2024 |
| CECAM Multiscale Simulations of Soft Matter IV, Technical University of Darmstadt | |
| <i>Stochastic interpolants: A unifying framework for flows and diffusions</i> | 2024 |
| Allerton Control Conference, University of Illinois at Urbana-Champaign | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2024 |
| Youth in High Dimensions Workshop, International Center for Theoretical Physics | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2024 |
| CRUNCH Seminar, Brown University | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2024 |
| Applied Mathematics Seminar, University of Washington | |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> | 2024 |
| Applied and Interdisciplinary Mathematics Seminar, University of Michigan | |

| | |
|---|------|
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Computational and Applied Mathematics Seminar, University of Chicago | 2024 |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Applied Mathematics Seminar, Massachusetts Institute of Technology | 2024 |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Mathematics Seminar, University of North Carolina | 2024 |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Center for Nonlinear Analysis Seminar, Carnegie Mellon University | 2023 |
| <i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Modeling and Simulation Seminar, Courant Institute of Mathematical Sciences | 2023 |
| <i>Deep learning probability flows and entropy production rates in active matter</i> Generative Modeling Foundations, Courant Institute of Mathematical Sciences | 2023 |
| <i>Deep learning probability flows and entropy production rates in active matter</i> Scientific Machine Learning Seminar, Courant Institute of Mathematical Sciences | 2023 |
| <i>Neural networks for computational mathematicians</i> Modeling and Simulation Seminar, Courant Institute of Mathematical Sciences | 2023 |
| <i>A spin glass model of microbial long-term evolution</i> Mostly Biomathematics Seminar, Courant Institute of Mathematical Sciences | 2023 |
| <i>On stochastic and deterministic generative models</i> Generative Modeling Foundations, Courant Institute of Mathematical Sciences | 2023 |
| <i>Representation and optimization in adaptive control</i> Azizan Group, Massachusetts Institute of Technology | 2023 |
| <i>Probability flow solution of the Fokker-Planck equation</i> Google Brain Robotics, New York, New York | 2022 |
| <i>Probability flow solution of the Fokker-Planck equation</i> Sampling, Transport, and Diffusions Workshop, Flatiron Institute | 2022 |
| <i>Probability flow solution of the Fokker-Planck equation</i> Computational Mathematics Seminar, Courant Institute of Mathematical Sciences | 2022 |
| <i>Probability flow solution of the Fokker-Planck equation</i> Generative Modeling Foundations, Courant Institute of Mathematical Sciences | 2022 |
| <i>Probability flow solution of the Fokker-Planck equation</i> Bruna Group, Courant Institute of Mathematical Sciences | 2022 |
| <i>Nonparametric adaptive control: theory and randomized algorithms</i> Courant Instructor Day, Courant Institute of Mathematical Sciences | 2022 |
| <i>A spin glass model of microbial long-term evolution</i> Modeling and Simulation Seminar, Courant Institute of Mathematical Sciences | 2021 |
| <i>Nonlinear adaptive control theory: a view from optimization and machine learning</i> Bruna Group, Courant Institute of Mathematical Sciences | 2021 |
| <i>Nonparametric adaptive control: theory and randomized algorithms</i> CRAN, Université de Lorraine (Virtual) | 2021 |
| <i>Regret bounds for adaptive nonlinear control</i> 3rd Annual Conference on Learning for Dynamics and Control (Virtual) | 2021 |
| <i>Projection methods for quasi-static hypo-elastoplasticity</i> Numerical Methods for PDEs Seminar, Massachusetts Institute of Technology | 2021 |
| <i>Projection methods for quasi-static hypo-elastoplasticity</i> Modeling and Simulation Seminar, Courant Institute of Mathematical Sciences | 2021 |

| | |
|--|------|
| <i>Adaptive control theory</i> | 2021 |
| Learning for Dynamics and Control Course, Google Brain (Virtual) | |
| <i>Learning stability certificates from data</i> | 2021 |
| Anandkumar Group, California Institute of Technology | |
| <i>Learning stability certificates from data</i> | 2020 |
| Neurophysics Group, Harvard University | |
| <i>A continuous-time analysis of distributed stochastic gradient</i> | 2020 |
| Google Brain Robotics, New York, New York | |
| <i>Adaptive control and statistical learning</i> | 2020 |
| Google Brain Robotics, New York, New York | |
| <i>Learning dynamical systems with deep feedforward and balanced recurrent networks</i> | 2020 |
| Neurophysics Group, Harvard University | |
| <i>Parallel three-dimensional simulations of quasi-static elastoplastic solids</i> | 2019 |
| Computational Science Graduate Fellowship Program Review, Arlington, Virginia | |
| <i>Continuum-level simulation of shear banding in metallic glasses on transforming grids</i> | 2019 |
| American Physical Society March Meeting, Boston, Massachusetts | |
| <i>Three-dimensional continuum-level simulation of shear banding in metallic glasses</i> | 2018 |
| American Physical Society March Meeting, Los Angeles, California | |
| <i>A quasi-static projection method for three-dimensional hypo-elastoplasticity</i> | 2017 |
| SIAM Conference on Computational Science and Engineering, Atlanta, Georgia | |
| <i>Amorphous plasticity and the shear transformation zone theory</i> | 2016 |
| Kavli Seminar, Harvard University | |