

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	
Visiting Graduate Student Researcher, MIT	
Advisors: Jean-Jacques E. Slotine (MIT) and Chris H. Rycroft (Harvard)	
Thesis title: <i>Methods for scientific simulation, machine learning, and nonlinear control</i>	
Northwestern University	2010–2014
B.A. Mathematics, Physics, and Integrated Science <i>with honors</i>	
Advisor: Tamar Seideman	
Thesis title: <i>High harmonic generation from simple aromatic molecules</i>	

PROFESSIONAL POSITIONS

Carnegie Mellon University	2024–present
<i>Assistant Professor, Machine Learning Department (50%)</i>	
<i>Assistant Professor, Department of Mathematical Sciences (50%)</i>	
Courant Institute of Mathematical Sciences	2021–2024
<i>Courant Instructor & Assistant Professor</i>	
Google Brain	2020–2021
<i>Student Researcher</i>	
Advisor: Vikas Sindhwani	
Lawrence Berkeley National Lab	2016
<i>Computational Science Graduate Fellowship Practicum</i>	
Advisor: Adam Arkin	
Tel Aviv University	2014–2015
<i>Fulbright Research Scholar</i>	
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
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

GRANTS

Carl & Amy Jones Endowed Interdisciplinary Seed Fund	2025
Breaking the curse of dimensionality in scientific computing with reinforcement learning-based stochastic control	
<i>Principal Investigator</i> (with M. Simchowitz)	
<i>Awarded</i>	

PREPRINTS

* indicates co-first authorship; † indicates co-advising.

Student co-authors are color-coded by role: **PhD**, **masters**, **undergraduate**, **affiliated**.

Ayhan Suleymanzade, Chanhuk Lee, Floor Eijkelboom, **Nicholas M. Boffi**, İsmail İlkan Ceylan†, and Jinwoo Kim†. Thinking with looped flows. *arXiv:2609.11801*, 2026.

Jaehoon Yoo*, Wonjung Kim*, Floor Eijkelboom, Chanhuk Lee, **Nicholas M. Boffi**, Seunghoon Hong†, and Jinwoo Kim†. Self-conditioned flow map language models via fixed-point flows. *arXiv:2607.00714*, 2026.

RuiKang OuYang, Hanlin Yu*, Xinyue Ai*, Yutong He, **Nicholas M. Boffi**, Pradeep Ravikumar, José Miguel Hernández-Lobato, Max Simchowitz, Benjamin Kurt Miller, and Omar Chehab. Few-step Boltzmann generators via scalable likelihood flow maps. *arXiv:2606.29110*, 2026.

Manan Agarwal*, **Sheel Shah***, Chanhuk Lee, Jaehoon Yoo, **Jerry Huang**, Seunghoon Hong, Aditi Raghunathan, Jinwoo Kim†, and **Nicholas M. Boffi†**. Posterior refinement: Fast language generation via any-order flow maps. *arXiv:2606.24773*, 2026.

Gianluca Scarpellini, Ron Shprints, Peter Holderrieth, Juno Nam, Pranav Murugan, Rafael Gómez-Bombarelli, Tommi Jaakkola, Maruan Al-Shedivat, **Nicholas M. Boffi**, and Avishek Joey Bose. Few-step cofolding with all-atom flow maps. *arXiv:2606.08375*, 2026.

Rishal Aggarwal, David Ryan Koes, **Nicholas M. Boffi**, and Eric Vanden-Eijnden. Reactive flux matching: Mechanism discovery and adaptive sampling of rare events. *arXiv:2606.06295*, 2026.

Sanjit Dandapanthula and **Nicholas M. Boffi**. Are we really tilting? The mechanics of reward guidance in flow and diffusion models. *arXiv:2606.02884*, 2026.

Chanhuk 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.

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

* indicates co-first authorship; † indicates co-advising.

Student co-authors are color-coded by role: [PhD](#), [masters](#), [undergraduate](#), [affiliated](#).

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.

[Jerry Y. Huang](#), [Justin Lin](#), [Sheel Shah](#), [Kartik Nair](#), and [Nicholas M. Boffi](#). How to guide your flow: Few-step alignment via flow map reward guidance. In *International Conference on Machine Learning*, 2026.

Peter Holderrieth, Douglas Chen, Luca Eyring, [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. In *International Conference on Machine Learning*, 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 <code>flow-maps</code> , official code for flow map self-distillation (NeurIPS 2025) https://github.com/nmboffi/flow-maps	2025
Author of <code>structured_diffusion_public</code> , code for structured diffusion learning (ICLR 2025) https://github.com/nmboffi/structured_diffusion_public	2025
Author of <code>jax-interpolants</code> , a clean implementation of the stochastic interpolant framework https://github.com/nmboffi/jax-interpolants	2025
Author of <code>flow_map_matching_public</code> , an implementation of the flow map matching method https://github.com/nmboffi/flow_map_matching_public	2025
Author of <code>jax-edm2</code> , JAX implementation of NVIDIA’s EDM2 network https://github.com/nmboffi/jax-edm2	2025
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

SEMINARS AND SUMMER SCHOOLS

Foundations of flow and diffusion-based generative models, <i>Co-instructor</i>	Summer 2025
Summer seminar and reading group, Carnegie Mellon University	
<i>Co-taught with Max Simchowitz</i>	
Summer School on Scientific Machine Learning, <i>Invited lecturer</i>	Summer 2025
Brin Mathematics Research Center, Department of Mathematics, University of Maryland	
Delivered several lectures on flows and diffusions	
Neural networks for partial differential equations, <i>Co-organizer</i>	Fall 2024
Center for Nonlinear Analysis reading group, Carnegie Mellon University	
<i>Co-organized with Dejan Slepčev, Ameet Talwalkar, and Andrej Risteski</i>	
~40 participants across mathematics, computer science, machine learning, statistics, and engineering	

MENTORING

PhD Students	
Jerry Huang	2024–Present
Computer Science Department, Carnegie Mellon University	
Stephen Huan	2024–Present
Computer Science Department, Carnegie Mellon University	
<i>Co-advised with Andrej Risteski</i>	
Sanjit Dandapanthula	2025–Present
Department of Statistics and Machine Learning Department, Carnegie Mellon University	
<i>Co-advised with Aaditya Ramdas</i>	
Abbas Mammadov	2026–Present
Department of Statistics, University of Oxford	
<i>Co-advised with Yee Whye Teh</i>	
Masters Students	
Justin Lin	2025–Present
Machine Learning Department, Carnegie Mellon University	

Kartik Nair	2025–Present
Machine Learning Department, Carnegie Mellon University	
Manan Agarwal	Summer 2026
Machine Learning Department, Carnegie Mellon University	
Partha Kaushik	Summer 2026
Robotics Institute, Carnegie Mellon University	
Sheel Shah	2025–Present
Machine Learning Department, Carnegie Mellon University	
Undergraduate Students	
Ishin Shah	2025–2026
Master’s thesis, Department of Mathematical Sciences, Carnegie Mellon University	
Thesis title: <i>Efficient reward alignment in generative models via diamond flow maps and stochastic optimal control</i>	
<i>Co-advised with Max Simchowitz</i>	
Jimmy Almgren-Bell	2017–2019
Senior thesis, Harvard University	
Thesis title: <i>An agent-based numerical approach to Lenski’s long-term evolution experiment</i>	
Affiliated Students	
Anna Wei	
Machine Learning Department, Carnegie Mellon University	
<i>Advised by Ameet Talwalkar and Andrej Risteski</i>	
Rishal Aggarwal	
Computational Biology Department, Carnegie Mellon University	
<i>Advised by David Ryan Koes</i>	
Summer Programs	
Applied Mathematics Summer Undergraduate Research Experience (AM-SURE)	2022
Program coordinator, Courant Institute of Mathematical Sciences	
<i>Mentored ten undergraduate students through summer research projects</i>	

PROFESSIONAL ACTIVITIES

Workshop organizer

Theory of AI for Scientific Computing (TASC)	June 30, 2025
COLT 2025, Mérieux Amphitheater, ENS Lyon, France	
<i>Jointly organized with Misha Khodak (Princeton), Jianfeng Lu (Duke), Tanya Marwah (Polymathic AI, Flatiron Institute), and Andrej Risteski (CMU)</i>	
Measure Transport, Diffusion Processes, and Sampling	Dec. 4–6, 2023
Flatiron Institute, New York City	
<i>Jointly organized with Michael Albergo (NYU), Bob Carpenter (Flatiron Institute), Neha Wadia (Flatiron Institute), and Joan Bruna (Courant, Flatiron Institute)</i>	

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

<i>Flow map language models</i> Machine Learning Seminar, Flatiron Institute	2026
<i>Flow map language models</i> Machine Learning Reading Group, Google DeepMind, London	2026
<i>Flow map language models</i> Workshop on Geometry-grounded Representation Learning and Generative Modeling International Conference on Learning Representations (ICLR)	2026
<i>Flow map language models</i> Mathematical Foundations of Artificial Intelligence (MFAI) Workshop Instituto de Investigaciones en Matemáticas Aplicadas y en Sistemas (IIMAS) Universidad Nacional Autónoma de México	2026
<i>Flow map language models</i> Machine Learning Reading Group, Cerebras	2026
<i>Flow map language models</i> Starkly Speaking Reading Group	2026
<i>Flow map language models</i> Frontiers of Flows for Generative AI Workshop, Carnegie Mellon University	2026
<i>Flow maps: Generative models with lightning-fast inference</i> PSU-Purdue-UMD Joint Seminar on Mathematical Data Science	2026
<i>Flow maps: Generative models with lightning-fast inference</i> CMX Lunch Seminar, California Institute of Technology	2026
<i>Lectures on flows and diffusions for scientific machine learning</i> Summer School on Scientific Machine Learning Brin Mathematics Research Center, Department of Mathematics, University of Maryland <i>Invited lecturer; delivered several lectures</i>	2025
<i>Flow maps: Generative models with lightning-fast inference</i> Workshop on Mathematical Analysis in Machine Learning, University of Pittsburgh	2025
<i>Flow maps: Generative models with lightning-fast inference</i> Starkly Speaking Reading Group	2025
<i>Flow maps: Generative models with lightning-fast inference</i> Statistics & Data Science Seminar, Carnegie Mellon University	2025
<i>Flow maps: Generative models with lightning-fast inference</i> Kinetic Theory: Novel Statistical, Stochastic and Analytical Methods SLMath (Simons Laufer Mathematical Sciences Institute)	2025
<i>How to build a consistency model (with stochastic interpolants)</i> Workshop on Probabilistic Learning Methods for Inverse Problems Applied Inverse Problems Conference, Rio de Janeiro, Brazil	2025
<i>Stochastic interpolants: from generative modeling to generative science and engineering</i> Efficient and Reliable Deep Learning Methods and their Scientific Applications Banff International Research Station	2025
<i>Stochastic interpolants: from generative modeling to generative science and engineering</i> Statistical and Computational Challenges in Probabilistic Scientific Machine Learning Institute for Mathematical and Statistical Innovation, University of Chicago	2025

<i>Stochastic interpolants: from generative modeling to generative science and engineering</i> Frontiers in Applied Analysis Workshop, Carnegie Mellon University	2025
<i>Stochastic interpolants: from generative modeling to generative science and engineering</i> Machine Learning Department Seminar, Carnegie Mellon University	2025
<i>Stochastic interpolants: from generative modeling to generative science and engineering</i> Applied Mathematics and Statistics Seminar, Johns Hopkins University	2025
<i>Stochastic interpolants: from generative modeling to generative science and engineering</i> Applied and Computational Mathematics Seminar University of Wisconsin-Madison	2025
<i>Flow maps: Generative models with lightning-fast inference</i> Advances in Generative AI Conference, Duke University	2025
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Oden Institute & Department of Mathematics Seminar University of Texas at Austin	2025
<i>Stochastic interpolants: A unifying framework for flows and diffusions</i> CMU-Pitt Joint Computational Biology Seminar Carnegie Mellon University & University of Pittsburgh	2025
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Continuum Mechanics Seminar, University of Nebraska-Lincoln	2025
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Center for Computational Mathematics Seminar Flatiron Institute	2024
<i>Stochastic interpolants: A unifying framework for flows and diffusions</i> Genesis Therapeutics	2024
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Machine Learning / Duolingo Seminar, Carnegie Mellon University	2024
<i>Deep learning the entropy production rate in active matter physics</i> CECAM Multiscale Simulations of Soft Matter IV, Technical University of Darmstadt	2024
<i>Stochastic interpolants: A unifying framework for flows and diffusions</i> Allerton Control Conference, University of Illinois at Urbana-Champaign	2024
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Youth in High Dimensions Workshop, International Center for Theoretical Physics	2024
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> CRUNCH Seminar, Brown University	2024
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Applied Mathematics Seminar, University of Washington	2024
<i>On flows and diffusions: from many-body Fokker-Planck to stochastic interpolants</i> Applied and Interdisciplinary Mathematics Seminar, University of Michigan	2024
<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> Learning for Dynamics and Control Course, Google Brain (Virtual)	2021
<i>Learning stability certificates from data</i> Anandkumar Group, California Institute of Technology	2021
<i>Learning stability certificates from data</i> Neurophysics Group, Harvard University	2020
<i>A continuous-time analysis of distributed stochastic gradient</i> Google Brain Robotics, New York, New York	2020
<i>Adaptive control and statistical learning</i> Google Brain Robotics, New York, New York	2020

<i>Learning dynamical systems with deep feedforward and balanced recurrent networks</i> Neurophysics Group, Harvard University	2020
<i>Parallel three-dimensional simulations of quasi-static elastoplastic solids</i> Computational Science Graduate Fellowship Program Review, Arlington, Virginia	2019
<i>Continuum-level simulation of shear banding in metallic glasses on transforming grids</i> American Physical Society March Meeting, Boston, Massachusetts	2019
<i>Three-dimensional continuum-level simulation of shear banding in metallic glasses</i> American Physical Society March Meeting, Los Angeles, California	2018
<i>A quasi-static projection method for three-dimensional hypo-elastoplasticity</i> SIAM Conference on Computational Science and Engineering, Atlanta, Georgia	2017
<i>Amorphous plasticity and the shear transformation zone theory</i> Kavli Seminar, Harvard University	2016