

Madeleine Udell

Assistant Professor, Management Science and Engineering
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Academic Employment **Stanford University** Stanford, CA
Assistant Professor July 2022 –
Management Science and Engineering and, by courtesy, Electrical Engineering
Gabilan Fellow
Affiliated with the Institute of Computational and Mathematical Engineering,
Stanford Data Science, Human Centered AI, and BioX

Cornell University Ithaca, NY
Associate Professor with tenure (on leave) July 2022 – June 2024
Assistant Professor July 2016 – June 2022
Department of Operations Research and Information Engineering
Richard and Sybil Smith Sesquicentennial Fellow
Graduate field member in Operations Research, Computer Science, Applied Mathematics,
Data Science, Statistics, Electrical and Computer Engineering, and Systems Engineering

California Institute of Technology Pasadena, CA
Postdoctoral Fellow, Center for the Mathematics of Information June 2015 – June 2016

Education **Stanford University** Stanford, CA
Ph.D. in Computational and Mathematical Engineering June 2015
GPA: 4.0.
Thesis: *Generalized Low Rank Models*. Advisor: Stephen P. Boyd.

Yale University New Haven, CT
B.S. in Mathematics and Physics June 2009
Summa cum laude, with honors in mathematics and honors in physics.
GPA: 3.95.
Thesis: *Local Parametrizations via Laplacian Eigenfunctions*. Advisor: Peter W. Jones.

Papers In the pipeline

12. C. Lawless, Y. Li, A. Wikum, M. Udell, and E. Vitercik. LLMs for cold-start cutting plane separator configuration. 2025, 2412.12038
11. C. Lawless, J. Schöffer, and M. Udell. “It was a magical box”: Understanding practitioner workflows and needs in optimization. 2025, 2509.16402
10. W. Gao, Y.-C. Chu, Y. Ye, and M. Udell. Gradient methods with online scaling part 2. theoretical foundations. 2025, 2509.11007
9. W. Gao, Y.-C. Chu, Y. Ye, and M. Udell. Gradient methods with online scaling part i. theoretical foundations. 2025, 2505.23081
8. M. V. Ness and M. Udell. dnamite: A python package for neural additive models. *Submitted*, 2025, 2503.07642

7. J. Sun, Z. Frangella, and M. Udell. Sapphire: Preconditioned stochastic variance reduction for faster large-scale statistical learning. *Major revision at SIAM Journal on Mathematics of Data Science (SIMODS)*, 2025, 2501.15941
6. L. Lessard and M. Udell. Algebraic characterization of equivalence between optimization algorithms. *Minor revision at Math Programming*, 2025, 2501.04972
5. W. Gao, H. Liu, Y. Ye, and M. Udell. When does primal interior point method beat primal-dual in linear optimization? *Submitted*, 2024, 2411.16015
4. P. Rathore, Z. Frangella, J. Yang, M. Dereziński, and M. Udell. Have ASkotch: A neat solution for large-scale kernel ridge regression. *Submitted*, 2025, 2407.10070
3. A. AhmadiTeshnizi, W. Gao, H. Brunborg, S. Talaei, and M. Udell. OptiMUS-0.3: Using large language models to model and solve optimization problems at scale. *Submitted*, 2024, 2407.19633
2. W. Gao, Z. Qu, M. Udell, and Y. Ye. Scalable approximate optimal diagonal preconditioning. *Submitted*, 2023, 2312.15594
1. Z. Frangella, S. Zhao, T. Diamandis, B. Stellato, and M. Udell. On the (linear) convergence of generalized newton inexact ADMM. *Submitted*, 2023, 2302.03863

Refereed Journal Articles

25. T. Diamandis, Z. Frangella, S. Zhao, B. Stellato, and M. Udell. GeNIOS: an (almost) second-order operator-splitting solver for large-scale convex optimization. *Accepted at Math Programming Computation*, 2023, 2310.08333
24. S. Fazliani, Z. Frangella, and M. Udell. Enhancing physics-informed neural networks through feature engineering. *Transactions on Machine Learning Research (TMLR)*, 2025, 2502.07209
23. Y.-C. Chu, L.-R. Santos, and M. Udell. Randomized nyström preconditioned interior point-proximal method of multipliers. *Accepted to SIAM Journal on Scientific Computing (SISC)*, 2024, 2404.14524
22. P. Rathore, Z. Frangella, and M. Udell. PROMISE: Preconditioned stochastic optimization methods by incorporating scalable curvature estimates. *JMLR*, 2024, 2309.02014
21. Z. Frangella, P. Rathore, S. Zhao, and M. Udell. SketchySGD: Reliable stochastic optimization via robust curvature estimates. *SIAM Journal on Mathematics of Data Science (SIMODS)*, 2024, 2211.08597
20. Y. Zhao and M. Udell. gcompute: A package for missing data imputation. *Journal of Statistical Software*, 2024, 2203.05089
19. D. Kouri, Z. Hua, and M. Udell. A greedy galerkin method to efficiently select sensors for linear dynamical systems. *Linear Algebra and its Applications*, 2023

18. J. Fan, L. Ding, C. Yang, and M. Udell. Low-rank tensor recovery with Euclidean-norm-induced Schatten-p quasi-norm regularization. *Transactions on Machine Learning Research*, 2023, 2012.03436
17. Z. Frangella, J. A. Tropp, and M. Udell. Randomized Nyström preconditioning. *SIAM Journal on Matrix Analysis and Applications*, 2022, 2110.02820
16. L. Ding and M. Udell. A strict complementarity approach to error bound and sensitivity of solution of conic programs. *Optimization Letters*, 2022, 2012.00183
15. N. Sengupta, M. Udell, N. Srebro, and J. Evans. Sparse data reconstruction, missing value and multiple imputation through matrix factorization. *Sociological Methodology*, 2022
14. L. Ding and M. Udell. On the simplicity and conditioning of low rank semidefinite programs. *SIAM Journal on Optimization (SIOPT)*, 2021, 2002.10673
13. L. Ding, A. Yurtsever, V. Cevher, J. A. Tropp, and M. Udell. An optimal-storage approach to semidefinite programming using approximate complementarity. *SIAM Journal on Optimization (SIOPT)*, 2021, 1902.03373
12. J. Fan, C. Yang, and M. Udell. Robust non-linear matrix factorization for dictionary learning, denoising, and clustering. *IEEE Trans. Signal Processing (TSP)*, 69:1755–1770, 2021, 2005.01317
11. R. Muthukumar, D. P. Kouri, and M. Udell. Randomized sketching algorithms for low-memory dynamic optimization. *SIAM Journal on Optimization (SIOPT)*, 31(2):1242–1275, 2021
10. A. Yurtsever, J. A. Tropp, O. Fercoq, M. Udell, and V. Cevher. Scalable semidefinite programming. *SIAM Journal on Mathematics of Data Science (SIMODS)*, 3(1):171–200, 2021, 1912.02949
9. Y. Sun, Y. Guo, C. Luo, J. A. Tropp, and M. Udell. Low-rank Tucker approximation of a tensor from streaming data. *SIAM Journal on Mathematics of Data Science (SIMODS)*, 2(4):1123–1150, 2020, 1904.10951
8. J. A. Tropp, A. Yurtsever, M. Udell, and V. Cevher. Streaming low-rank matrix approximation with an application to scientific simulation. *SIAM Scientific Computing (SISC)*, 41(4):A2430–A2463, 2019, 1902.08651
7. M. Udell and O. Toole. Optimal design of efficient rooftop photovoltaic arrays. *INFORMS Journal on Applied Analytics (Interfaces)*, 49(4):281–294, 2019
6. M. Udell and A. Townsend. Why are big data matrices approximately low rank? *SIAM Journal on Mathematics of Data Science (SIMODS)*, 1(1):144–160, 2019, 1705.07474
5. N. Kallus and M. Udell. Dynamic assortment personalization in high dimensions. *Operations Research*, 2019, 1610.05604
4. J. A. Tropp, A. Yurtsever, M. Udell, and V. Cevher. Practical sketching algorithms for low-rank matrix approximation. *SIAM Journal of Matrix Analysis and Applications (SIMAX)*, 38(4):1454–1485, 2017, 1609.00048

3. M. Udell, C. Horn, R. Zadeh, and S. Boyd. Generalized low rank models. *Foundations and Trends in Machine Learning*, 9(1), 2016, 1410.0342
2. M. Udell and S. Boyd. Bounding duality gap for separable problems with linear constraints. *Computational Optimization and Applications*, 64(2):355–378, 2016, 1410.4158
1. E. Birch, M. Udell, and M. Covert. Incorporation of flexible objectives and time-linked simulation with flux balance analysis. *Journal of Theoretical Biology*, 345:12–21, 2014

Refereed Conference Proceedings

43. C. Lawless, T.-W. Weng, B. Ustun, and M. Udell. Understanding fixed predictions via confined regions. In *International Conference on Machine Learning (ICML)*, 2025, 2502.16380
42. P. Rathore, Z. Frangella, S. Garg, S. Fazliani, M. Dereziński, and M. Udell. Turbocharging gaussian process inference with approximate sketch-and-project. In *NeurIPS*, 2025, 2505.13723
41. Y.-C. Chu, W. Gao, Y. Ye, and M. Udell. Provable and practical online learning rate adaptation with hypergradient descent. In *International Conference on Machine Learning (ICML)*, 2025, 2502.11229
40. W. Gao, Y.-C. Chu, Y. Ye, and M. Udell. Gradient methods with online scaling. In *Conference on Learning Theory (COLT)*, 2025, 2411.01803
39. C. Lawless, Y. Li, A. Wikum, M. Udell, and E. Vitercik. LLMs for cold-start cutting plane separator configuration. In *International Conference on the Integration of Constraint Programming, Artificial Intelligence, and Operations Research (CPAIOR)*, 2025, 2412.12038
38. M. V. Ness and M. Udell. Interpretable prediction and feature selection for survival analysis. In *ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2025, 2404.14689
37. M. Van Ness, B. Block, and M. Udell. DNAMite: interpretable calibrated survival analysis with discretized additive models. In *Machine Learning for Health (ML4H)*, 2024, 2411.05923
36. A. AhmadiTeshnizi, W. Gao, and M. Udell. OptiMUS: Scalable optimization modeling using MIP solvers and large language models. In *International Conference on Machine Learning (ICML)*, 2024, 2402.10172
35. P. Rathore, W. Lei, Z. Frangella, L. Lu, and M. Udell. Challenges in training PINNs: A loss landscape perspective. In *International Conference on Machine Learning (ICML)*, 2024, 2402.01868
34. M. Van Ness, T. Bosschieter, N. Din, A. Ambrosy, A. S. Singh, and M. Udell. Interpretable survival analysis for heart failure risk prediction. In *Machine Learning for Health (ML4H)*, 2023, 2310.15472

33. M. Van Ness, T. Bosschieter, R. Halpin-Gregorio, and M. Udell. The missing indicator method: From low to high dimensions. In *29th SIGKDD Conference on Knowledge Discovery and Data Mining - Applied Data Science Track*, 2023, 2211.09259
32. C.-H. Chang, J. Yoon, S. Arik, M. Udell, and T. Pfister. Data-efficient and interpretable tabular anomaly detection. In *SIGKDD Conference on Knowledge Discovery and Data Mining - Applied Data Science Track*, 2023
31. I. Drori, S. J. Zhang, R. S. Shuttleworth, S. Zhang, K. Tyser, Z. Chin, P. Langtigua, S. Surbehera, G. Hunter, D. Austin, L. Tang, Y. Hicke, S. Simhon, S. Karnik, D. Granberry, and M. Udell. From human days to machine seconds: Automatically answering and generating machine learning final exams. In *SIGKDD Conference on Knowledge Discovery and Data Mining - Applied Data Science Track*, 2023
30. Y. Zhao, A. Townsend, and M. Udell. Probabilistic missing value imputation for mixed categorical and ordered data. In *NeurIPS*, 2022, 2210.06673
29. C. Yang, G. Bender, H. Liu, P.-J. Kindermans, M. Udell, Y. Lu, Q. Le, and D. Huang. Resource-constrained neural architecture search on tabular datasets. In *NeurIPS*, 2022, 2204.07615
28. S. Zhao, Z. Frangella, and M. Udell. NysADMM: faster composite convex optimization via low-rank approximation. In *International Conference on Machine Learning (ICML)*, 2022, 2202.11599
27. C. Yang, Z. Wu, J. Chee, C. D. Sa, and M. Udell. How low can we go: Trading memory for error in low-precision training. In *International Conference on Learning Representations (ICLR)*, 2022, 2106.09686
26. Y. Zhao, E. Landgrebe, E. Shekhtman, and M. Udell. Online missing value imputation and correlation change detection for mixed-type data via gaussian copula. In *AAAI*, 2021, 2009.12326
25. W. T. Stephenson, Z. Frangella, M. Udell, and T. Broderick. Can we globally optimize cross-validation loss? Quasiconvexity in ridge regression. In *Advances in Neural Information Processing Systems (NeurIPS)*, 2021, 2107.09194
24. B. Liu, M. Xie, and M. Udell. ControlBurn: Feature selection by sparse forests. In *ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2021, 2107.00219
23. C. Yang, L. Ding, Z. Wu, and M. Udell. TenIPS: Inverse propensity sampling for tensor completion. In *International Conference on Artificial Intelligence and Statistics (AISTATS)*, 2021, 2101.00323
22. I. Drori, B. Kates, W. Sickinger, A. Kharkar, B. Dietrich, A. Shporer, and M. Udell. Galaxy TSP: A new billion-node benchmark for TSP. In *NeurIPS Workshop on Learning Meets Combinatorial Algorithms*, 2020
21. I. Drori, A. Kharkar, W. R. Sickinger, B. Kates, Q. Ma, S. Ge, E. Dolev, B. Dietrich, D. P. Williamson, and M. Udell. Learning to solve combinatorial optimization problems on real-world graphs in linear time. In *IEEE International Conference on Machine Learning and Applications (IEEE ICMLA)*, 2020, 2006.03750

20. W. Stephenson, M. Udell, and T. Broderick. Approximate cross-validation with low-rank data in high dimensions. In *Advances in Neural Information Processing Systems (NeurIPS)*, 2020, 2008.10547
19. Y. Zhao and M. Udell. Matrix completion with quantified uncertainty through low rank gaussian copula. In *Advances in Neural Information Processing Systems (NeurIPS)*, 2020, 2006.10829
18. C. Yang, J. Fan, Z. Wu, and M. Udell. AutoML pipeline selection: Efficiently navigating the combinatorial space. In *ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2020, 2006.04216
17. Y. Zhao and M. Udell. Missing value imputation for mixed data through gaussian copula. In *ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, 2020, 1910.12845
16. J. Fan, Y. Zhang, and M. Udell. Polynomial matrix completion for missing data imputation and transductive learning. In *Thirty-Fourth AAAI Conference on Artificial Intelligence*, pages 3842–3849, 2020, 1912.06989
15. J. Fan, L. Ding, Y. Chen, and M. Udell. Factor group-sparse regularization for efficient low-rank matrix recovery. In *Advances in Neural Information Processing Systems (NeurIPS)*, volume 32, pages 5105–5115, 2019, 1911.05774
14. C. Yang, Y. Akimoto, D. W. Kim, and M. Udell. OBOE: Collaborative filtering for AutoML model selection. In *ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD)*, volume 25, pages 1173–1183. ACM, 2019, 1808.03233
13. J. Fan and M. Udell. Online high-rank matrix completion. In *Computer Vision and Pattern Recognition (CVPR)*, pages 8690–8698, 2019
12. J. Chen, N. Kallus, X. Mao, G. Svacha, and M. Udell. Fairness under unawareness: Assessing disparity when protected class is unobserved. In *FAT*: Conference on Fairness, Accountability, and Transparency*, pages 339–348, 2019, 1811.11154
11. S. Zhou, S. Gupta, and M. Udell. Limited memory Kelley’s method converges for composite convex and submodular objectives. In *Advances in Neural Information Processing Systems*, 2018, 1807.07531
10. N. Kallus, X. Mao, and M. Udell. Causal inference with noisy and missing covariates via matrix factorization. In *Advances in Neural Information Processing Systems*, 2018, 1806.00811
9. J. A. Tropp, A. Yurtsever, M. Udell, and V. Cevher. Fixed-rank approximation of a positive-semidefinite matrix from streaming data. In *Advances in Neural Information Processing Systems*, 2017, 1706.05736
8. A. Yurtsever, M. Udell, J. A. Tropp, and V. Cevher. Sketchy decisions: Convex low-rank matrix optimization with optimal storage. In *International Conference on Artificial Intelligence and Statistics (AISTATS)*, pages 1188–1196, 2017, 1702.06838

7. X. Shen, S. Diamond, M. Udell, Y. Gu, and S. Boyd. Disciplined multi-convex programming. In *Chinese Control and Decision Conference (CCDC)*, 2017, 1609.03285
6. D. Davis, B. Edmunds, and M. Udell. The sound of APALM clapping: Faster nonsmooth nonconvex optimization with stochastic asynchronous PALM. In *Advances in Neural Information Processing Systems*, 2016, 1606.02338
5. A. Schuler, V. Liu, J. Wan, A. Callahan, M. Udell, D. Stark, and N. Shah. Discovering patient phenotypes using generalized low rank models. In *Pacific Symposium on Biocomputing (PSB)*, 2016
4. N. Kallus and M. Udell. Revealed preference at scale: Learning personalized preferences from assortment choices. In *The 2016 ACM Conference on Economics and Computation*, New York, NY, USA, 2016. ACM
3. H. Mehmood, M. Udell, and J. Cioffi. Revenue maximization for broadband service providers using revenue capacity. In *IEEE Global Communications Conference*, 2015
2. E. Lee, M. Udell, and S. Wong. Factorization for analog-to-digital matrix multiplication. In *International Conference on Acoustics, Speech, and Signal Processing (ICASSP)*, 2015
1. P. LePendou, Y. Liu, S. Iyer, M. Udell, and N. Shah. Analyzing patterns of drug use in clinical notes for patient safety. *Proceedings of the AMIA Summits on Translational Science*, 2012:63, 2012

Refereed Workshops

15. S. Fazliani, Z. Frangella, and M. Udell. Enhancing physics-informed neural networks through feature engineering. In *Theory of AI for Scientific Computing Workshop at COLT*, 2025, 2502.07209
14. V. S. Lokhande, K. Sohn, J. Yoon, M. Udell, C. Lee, and T. Pfister. Towards group robustness in the presence of partial group labels. In *ICML 2022: Workshop on Spurious Correlations, Invariance and Stability*, volume abs/2201.03668, 2022, 2201.03668
13. M. Van Ness and M. Udell. CDF normalization for controlling the distribution of hidden layer activations. In *I (Still) Can't Believe It's Not Better! NeurIPS 2021 Workshop*, 2021
12. I. Drori, L. Liu, Q. Ma, B. Kates, and M. Udell. Zero-shot AutoML. In New York Academy of Sciences, editor, *Annual Machine Learning Symposium*, 2020
11. C. Yang, L. Ding, Z. Wu, and M. Udell. TenIPS: Inverse propensity sampling for tensor completion (workshop). In *OPT2020: 12th Annual Workshop on Optimization for Machine Learning*, 2020
10. E. Landgrebe, Y. Zhao, and M. Udell. Online mixed missing value imputation using gaussian copula. In *ICML Workshop on the Art of Learning with Missing Values (Artemiss)*, 2020

9. I. Drori, L. Liu, S. Koorathota, N. Yi, J. Li, A. Moretti, J. Freire, and M. Udell. AutoML using metadata language embeddings. In *NeurIPS Workshop on Meta-Learning*, 2019, 1910.03698
8. Y. Zhang, K. Song, Y. Sun, S. Tan, and M. Udell. “Why should you trust my explanation?” understanding uncertainty in LIME explanations. In *ICML Workshop AI for Social Good*, 2019, 1904.12991
7. Y. Sun, Y. Guo, J. A. Tropp, and M. Udell. Tensor random projection for low memory dimension reduction. In *NeurIPS Workshop on Relational Representation Learning*, 2018, 2012.03436
6. C. Yang, Y. Akimoto, D. W. Kim, and M. Udell. OBOE: Collaborative filtering for AutoML initialization (workshop version). *NeurIPS Workshop on Automated Machine Learning*, 2018, 1808.03233
5. M. Paradkar and M. Udell. Graph-regularized generalized low rank models. In *CVPR Workshop on Tensor Methods in Computer Vision*, 2017
4. N. Kallus and M. Udell. Learning preferences from assortment choices in a heterogeneous population. In *ICML Workshop on Computational Frameworks for Personalization*, 2016, 1509.05113
3. M. Udell, K. Mohan, D. Zeng, J. Hong, S. Diamond, and S. Boyd. Convex optimization in Julia. In *SC14 Workshop on High Performance Technical Computing in Dynamic Languages*, 2014, 1410.4821
2. M. Udell, C. Horn, R. Zadeh, and S. Boyd. Generalized low rank models. *NeurIPS Workshop on Distributed Machine Learning and Matrix Computations*, 2014
1. M. Udell and R. Takapoui. Linear bandits, matrix completion, and recommendation systems. *NeurIPS Workshop on Large Scale Matrix Analysis and Inference*, 2013

Miscellaneous: theses, technical reports, book chapters, newsletters, etc.

21. I. Drori, G. Longhitano, M. Mao, S. Hyun, Y. Zhang, S. Park, Z. Meeks, X.-Y. Zhang, B. Segev, H. Yong, N. Verma, A. Shporer, A. Amit, and M. Udell. Diverse inference and verification for advanced reasoning. 2025, 2502.09955
20. K. Tyser, B. Segev, G. Longhitano, X.-Y. Zhang, Z. Meeks, J. Lee, U. Garg, N. Belsten, A. Shporer, M. Udell, D. Te’eni, and I. Drori. AI-driven review systems: Evaluating LLMs in scalable and bias-aware academic reviews. 2024, 2408.10365
19. K. Tyser, J. Lee, A. Shporer, M. Udell, D. Te’eni, and I. Drori. OpenReviewer: Mitigating challenges in LLM reviewing. 2023
18. A. AhmadiTeshnizi, W. Gao, and M. Udell. OptiMUS: Optimization modeling using MIP solvers and large language models. 2023, 2310.06116
17. M. Udell and Z. Frangella. Randomized numerical linear algebra for optimization. *SIAG/OPT Views and News*, 2022

16. B. Liu, M. Xie, H. Yang, and M. Udell. ControlBurn: Nonlinear feature selection with sparse tree ensembles, 2022
15. S. Zhao, L. Lessard, and M. Udell. An automatic system to detect equivalence between iterative algorithms. 2021, 2105.04684
14. N. Singh, B. Kates, J. Mentch, A. Kharkar, M. Udell, and I. Drori. Privileged zero-shot AutoML, 2021, 2106.13743
13. C. Yang, J. Fan, Z. Wu, and M. Udell. Efficient AutoML pipeline search with matrix and tensor factorization, 2020, 2006.04216
12. I. Drori, L. Liu, Q. Ma, J. Deykin, B. Kates, and M. Udell. Real-time AutoML, 2020
11. B. Liu and M. Udell. Impact of accuracy on model interpretations. 2020, 2011.09903
10. L. Ding, J. Fan, and M. Udell. *kFW*: A Frank-Wolfe style algorithm with stronger subproblem oracles. *In revision at Computational Optimization and Applications*, 2020, 2006.16142
9. E. A. Ricci, M. Udell, and R. A. Knepper. An information-theoretic approach to persistent environment monitoring through low rank model based planning and prediction. 2020, 2009.01168
8. M. Udell. Big data is low rank. *SIAG/OPT Views and News*, 2019
7. A. Ratner, D. Alistarh, G. Alonso, D. G. Andersen, P. Bailis, S. Bird, N. Carlini, B. Catanzaro, E. Chung, B. Dally, J. Dean, I. S. Dhillon, A. G. Dimakis, P. Dubey, C. Elkan, G. Fursin, G. R. Ganger, L. Getoor, P. B. Gibbons, G. A. Gibson, J. E. Gonzalez, J. Gottschlich, S. Han, K. M. Hazelwood, F. Huang, M. Jaggi, K. G. Jamieson, M. I. Jordan, G. Joshi, R. Khalaf, J. Knight, J. Konecný, T. Kraska, A. Kumar, A. Kyrillidis, J. Li, S. Madden, H. B. McMahan, E. Meijer, I. Mitliagkas, R. Monga, D. G. Murray, D. S. Papailiopoulos, G. Pekhimenko, T. Rekatsinas, A. Rostamizadeh, C. Ré, C. D. Sa, H. Sedghi, S. Sen, V. Smith, A. Smola, D. Song, E. R. Sparks, I. Stoica, V. Sze, M. Udell, J. Vanschoren, S. Venkataraman, R. Vinayak, M. Weimer, A. G. Wilson, E. P. Xing, M. Zaharia, C. Zhang, and A. Talwalkar. SysML: The new frontier of machine learning systems. *CoRR*, abs/1904.03257, 2019, 1904.03257
6. J. A. Tropp, A. Yurtsever, M. Udell, and V. Cevher. More practical sketching algorithms for low-rank matrix approximation. Technical Report 2018-01, California Institute of Technology, Pasadena, California, 2018
5. L. Ding and M. Udell. Frank-Wolfe style algorithms for large scale optimization. In *Large-Scale and Distributed Optimization*. Springer, 2018
4. M. Udell. *Generalized Low Rank Models*. PhD thesis, Stanford University, 2015
3. M. Udell and S. Boyd. PCA on a data frame. 2015
2. M. Udell and S. Boyd. Beyond principal component analysis (PCA). *Biomedical Computation Review*, 2014

1. M. Udell and S. Boyd. Maximizing a sum of sigmoids. 2013

Patents

Madeleine Udell and Oliver Toole. Optimal Design of Residential Photovoltaic Arrays.

Application No. 62/400,542, filed on 9/27/2017, published 3/29/2018.

Jinsung Yoon, Sercan Omer Arik, Madeleine Richards Udell, Chun-Hao Chang. Interpretable Anomaly Detection By Generalized Additive Models With Neural Decision Trees.

Application No. 18/113,267, filed on 2/23/2023, published 8/31/2023.

Da Huang, Chengrun Yang, Pieter-Jan Kindermans, Hanxiao Liu, Quoc V Le, Madeleine Richards Udell, Yifeng Lu, Gabriel Mintzer Bender. Neural Architecture Search with Improved Computational Efficiency.

Application No. 18/839,379, filed on 12/27/2025, published 5/15/2025.

Grants

Current

AFOSR (PI): Online Scaled Gradient Method for Stochastic and Distributed Optimization, \$519,711, December 2025 to November 2028.

Stanford Human-Centered AI – IBM Research award (PI): Democratizing Decision Optimization for Organizational Decision Makers via LLMs, \$85,000, September 2024.

Stanford Woods Institute for the Environmental Venture Projects (co-I): Sustainable and resilient performance-based design solutions using an inverse analysis approach employing multi-objective optimization, \$250,000, June 2024 to November 2026.

ONR (PI): Randomized Numerical Linear Algebra for Optimization, \$523,238, May 2024 to November 2027.

Past

ONR (PI): Fast Re-routing using Machine Learning, \$350,000, July 2022 to June 2025.

Alfred P. Sloan Foundation Fellowship, \$75,000, September 2021 to August 2023.

NSF IIS-1943131 (PI): CAREER: Accelerating Machine Learning with Low Dimensional Structure, \$550,000, October 2020 to September 2025.

ONR (PI): Young Investigator Award: DREAMI: Dimension Reduction for Efficient Automated Machine Intelligence, \$528,890, June 2020 to August 2024.

Canadian Institutes of Health Research (Senior Personnel): Using data to guide population health management: A comprehensive evaluation of analytic approaches for population segmentation. PI: Laura Rosella. \$291,082, March 2020 to March 2023. (Provides some student funding.)

NSF CCF-1740822 (Senior Personnel): TRIPODS: Data Science for Improved Decision-Making: Learning in the Context of Uncertainty, Causality, Privacy, and Network Structures, \$1,496,655, October 1, 2017 to September 30, 2020. (.09 calendar [ie, nominal level] effort.) This grant establishes a major data science center at Cornell.

Cornell: Digital Agriculture (co-PI): Development of a High-Resolution Weather Forecast Database for Digital Agricultural Research and Outreach Applications. Co-PI: Art Degaetano. \$214,000, 2017–2020.

DARPA FA8750-17-2-0101 (PI): Composable Robust Structured Data Inference, \$1,411,602, March 27, 2017 to February 24, 2020. (2 summer months, 1 academic month effort.) The goal of this project is to develop robust tools for imputing missing data in large scale, heterogeneous data tables by using side information about which rows or columns are similar.

Capital One (co-PI). Co-PI: Nathan Kallus. \$60,000, August 1 2018.

Cornell Tech Faculty Exchange Grant (co-PI). Co-PI: Nathan Kallus. \$3,000, 2016–2017.

Awards	US National Academy of Sciences Kavli Fellow,	2023
	Microsoft Research Faculty Fellowship Finalist,	2021
	Alfred P. Sloan Foundation Research Fellowship,	2021
	Cornell Engineering Research Excellence Award, Cornell University	2020
	INFORMS Optimization Society Student Research Prize, INFORMS	2019
	<i>First place for student Lijun Ding’s paper on Storage-Optimal Semidefinite Programming</i>	
	INFORMS Undergraduate Operations Research Prize, INFORMS	2018
	<i>Honorable mention for student Song Zhou’s paper on Limited Memory Kelley’s Method</i>	
	Douglas Whitney ’61 Engineering Teaching Excellence Award, Cornell University	2018
	Doing Good with Good OR Student Paper Competition, INFORMS	2017
	<i>Second place, for Optimal design of efficient rooftop photovoltaic arrays.</i>	
	Center for the Mathematics of Information Postdoctoral Fellowship	2015
	California Institute of Technology	
	Gerald J. Lieberman Fellowship, Stanford University	2014
	<i>Awarded to doctoral students demonstrating the potential to become academic leaders. (12 Lieberman Fellows are selected among all doctoral candidates at Stanford each year.)</i>	
	Best Force Multiplier, DARPA PlanX	2013
	Graduate Research Fellowship, National Science Foundation	2010
	Gabilan Graduate Fellowship, Stanford University	2009
	Phi Beta Kappa, Yale University	2009
	Henry Edwards Ellsworth Prize, Yale University	2009
	<i>Awarded for the best senior thesis research paper in the sciences.</i>	
	US Physics Olympics Team Member	2005

Advising Below I list my Stanford PhD advisees. See my webpage for details.

8. **Wanyu Zhang, MS&E** *September 2025 – June 2030 (expected)*
7. **Yinjun Wang, MS&E** *September 2024 – June 2029 (expected)*
6. **Wenzhi Gao, ICME** *September 2023 – June 2028 (expected)*
5. **Shaghayegh Fazliani, Math** *April 2024 – June 2027 (expected)*
4. **Ya-Chi Chu, Math** *Jan 2023 – June 2026 (expected)*
3. **Pratik Rathore, EE** *July 2022 – June 2026 (expected)*
2. **Mike Van Ness, MS&E** *Sept 2020 – June 2025*
Interpretable machine learning with applications in healthcare
1. **Zachary Frangella, MS&E** *Jan 2020 – June 2025*
Randomized numerical linear algebra for large-scale optimization.

PhD committee member

16. Max Schaller (EE)	<i>expected graduation June 2026</i>
15. Akshay Rao (CEE)	<i>expected graduation June 2026</i>
14. Sam Liu (MS&E)	<i>expected graduation June 2026</i>
13. Sergio Gomez (GSB OIT)	<i>June 2025</i>
12. Tianyi Zhang (CS)	<i>June 2025</i>
11. Chunlin Sun (ICME)	<i>June 2025</i>
10. Zhuoyang Liu (GSB OIT)	<i>December 2025</i>
9. Izabel Aguilar (MS&E)	<i>June 2024</i>
8. Robert Dyro (AA)	<i>June 2024</i>
7. Jiahong Ouyang (EE)	<i>June 2024</i>
6. Robin Alexandra Brown (ICME)	<i>June 2024</i>
5. Ethan Steinberg (BDS)	<i>September 2023</i>
4. Yujia Jin (MS&E)	<i>September 2023</i>
3. Yiping Lu (MS&E)	<i>September 2023</i>
2. Aldo Carranzo (ICME)	<i>June 2023</i>
1. Gradey Wang (Mechanical Engineering)	<i>March 2023</i>

Postdoctoral advising

1. Connor Lawless *July 2025 – June 2027 (expected)*

Cornell advising At Cornell, I advised 3 postdoctoral researchers; 6 PhDs (5 graduated as of June 2022); 16 undergraduates; and 26 Masters students. I served on several other PhD committees and was a member of seven graduate fields: ORIE, CS, CAM, Statistics, ECE, Systems, and Data Science.

Below I list my Cornell PhD advisees. See my webpage for details.

6. Shipu Zhao, **Systems Engineering** *May 2023*
New perspectives in continuous optimization: theory and methodology
5. Yuxuan Zhao, **Statistics** *May 2022*
Gaussian copula for mixed data with missing values: model estimation and imputation
4. Chengrun Yang, **Electrical and Computer Engineering** *May 2022*
Automated machine learning under resource constraints
3. Lijun Ding, **ORIE (co-advised by Yudong Chen)** *August 2021*
Large scale semidefinite programming: simplicity, conditioning, and an efficient algorithm.
2. Xiaojie Mao, **Statistics (co-advised by Nathan Kallus)** *May 2021*
Machine Learning Methods for Data-driven Decision Making: Contextual Optimization, Causal Inference, and Algorithmic Fairness.
1. Yiming Sun, **Statistics (co-advised by Sumanta Basu)** *October 2019*
High Dimensional Data Analysis with Dependency and Under Limited Memory.

Cornell PhD committee member

3. Tianyi Shi (Applied Math) *graduated May 2022*

2. **Zhengze Zhou (Statistics)** *graduated May 2021*

1. **Yingjie Bi (Electrical and Computer Engineering)** *graduated January 2020*

Teaching	CME 307 / MS&E 311: Optimization	Stanford University
	<i>Instructor</i>	<i>Spring 2023, Winter 2024, Fall 2024, Fall 2025</i>
	PhD level required course; 46 students (2023), 28 students (W 2024), 44 students (F 2024).	
	MS&E 478: ASES Breakthrough	Stanford University
	<i>Instructor</i>	<i>Spring 2024</i>
	Student-led course on venture finance	
	CME 218 / MS&E 218: Applied Data Science	Stanford University
	<i>Instructor</i>	<i>Spring 2023</i>
	masters level project-based elective; 24 students.	
	MS&E 125: Intro to Applied Statistics	Stanford University
	<i>Instructor</i>	<i>Spring 2023</i>
	sophomore level required course; 71 students.	
	ORIE 7391: Faster: Algorithmic Ideas for Speeding Up Optimization	Cornell University
<i>Instructor</i>	<i>Spring 2022</i>	
PhD level elective course; 13 students.		
ORIE 4741: Learning with Big Messy Data	Cornell University	
<i>Instructor</i>	<i>Fall 2016, 2017, 2019, 2020, 2021</i>	
Undergraduate level elective course in data analysis; 2016: 85 students. 2017: 117 students. 2019: 147 students. 2020: 117 students. 2021: 155 students.		
ORIE 3120: Practical Tools for Operations Research, Data Science, and Machine Learning	Cornell University	
<i>Instructor</i>	<i>Spring 2020</i>	
Undergraduate level required course; 222 students		
ORIE 7191: Optimization for Machine Learning	Cornell University	
<i>Instructor</i>	<i>Spring 2019</i>	
PhD level elective course; 15 students.		
CS+ORIE+STSCI 1380: Data Science for All	Cornell University	
<i>Co-instructor</i>	<i>Spring 2018</i>	
Undergraduate level elective course in data analysis; 30 students.		
ORIE 6326: Convex Optimization	Cornell University	
<i>Instructor</i>	<i>Spring 2017</i>	
PhD level elective course; 47 students.		
EE 364b: Convex Optimization II	Stanford University	
<i>Teaching assistant</i>	<i>Spring 2014</i>	
CVX 101: Convex Optimization	EdX Stanford	
<i>Head teaching assistant</i>	<i>Winter 2014</i>	
Taught 10,000 students worldwide.		

EE 364a: Convex Optimization I <i>Instructor</i>	Stanford University <i>Summer 2013</i>
EE 364a: Convex Optimization I <i>Teaching Assistant</i>	Stanford University <i>Winter 2012</i>
CME Refresher Course: Discrete Math and Algorithms <i>Instructor</i>	Stanford University <i>September 2011, September 2012</i>
CME 305: Discrete Mathematics and Algorithms <i>Teaching Assistant</i>	Stanford University <i>Winter 2011</i>

**Academic
Service**

INFORMS Annual Meeting Organizing Committee <i>Invited Sessions Committee Member</i>	INFORMS <i>2026</i>
Conference on AI for Operations and Operations for AI: Charting the Future Stanford University <i>Co-organizer (with Mohsen Bayati, Ramesh Johari, Arvind Karunakaran, Vasilis Syrgkanis, Ben Van Roy, Gabriel Weintraub)</i>	<i>2023</i>
Generative AI Public Working Group <i>Committee Member</i>	NIST <i>2023–2024</i>
SciML Scientific Machine Learning Advisory Committee <i>Committee Member</i>	NumFocus <i>2020–</i>
JuliaOpt <i>Co-owner</i> The JuliaOpt organization curates high quality optimization software in the Julia language.	Github <i>Fall 2014 –</i>
INFORMS AI Strategy Advisory Committee <i>Committee Member</i> <i>Advise the INFORMS board to develop synergies and opportunities in artificial intelligence</i>	INFORMS <i>2019</i>
INFORMS Session: Large Scale Semidefinite Programming <i>Co-organizer (with Lijun Ding)</i>	Seattle <i>2019</i>
BoydFest: Stephen Boyd’s 60th Birthday Conference <i>Co-organizer (with Maryam Fazel and Mung Chiang)</i>	Stanford University <i>2018</i>
ICDM Workshop: Data Driven Discovery of Models <i>Co-organizer (with Christophe Giraud-Carrier and Ishanu Chattopadhyay)</i>	New Orleans <i>2017</i>
SIAM Annual: mini-symposium on Robust Low-Rank Models and Applications New Orleans <i>Co-organizer (with Tamara Kolda)</i>	<i>2016</i>
MOPTA: session on Large-scale Distributed Convex Optimization <i>Organizer</i>	Lehigh <i>2015</i>

**University
Service**

ICME Xpo <i>Co-organized workshop for affiliates</i>	Stanford University <i>2022–2023</i>
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	MS&E Teaching Committee <i>Committee Member</i>	Stanford University 2022–2023
	ORIE Curriculum Committee <i>Committee Member</i>	Cornell University 2020–2021
	CAM Colloquium Committee <i>Committee Member</i>	Cornell University 2018–2019
	CAM PhD Admissions Committee <i>Committee Member</i>	Cornell University 2018, 2021
	ORIE Curriculum Review Committee <i>Committee Member</i>	Cornell University 2017–2018
	ORIE PhD Admissions Committee <i>Committee Member</i>	Cornell University 2016, 2017
	Committee on the Future of the School of Engineering <i>Committee Member</i>	Stanford University Fall 2014 – Spring 2015
	Represented all engineering doctoral students on faculty committee. Collaborated on proposal addressing faculty hiring and development, research themes and centers, space and facilities, education and outreach, and interdisciplinary research.	
	C² Computational Consulting <i>Consultant</i>	Stanford University Fall 2011 – Spring 2015
	Helped researchers across the university (in physics, computer science, neuroscience, law, immunology, ...) formulate and solve numerical problems.	
	EE Faculty Search Committee <i>Committee Member</i>	Stanford University Fall 2014 – Spring 2015
	Student member on Electrical Engineering broad area search committee.	
	Information Systems Laboratory Colloquium <i>Coordinator</i>	Stanford University Winter 2012 – Spring 2013
	Invited and hosted academic speakers for weekly seminar series.	
	Committee on Graduate Studies <i>Committee Member</i>	Stanford University Fall 2011 – Spring 2013
	Debated and decided policies for all graduate students at Stanford. Approved and reauthorized interdisciplinary graduate programs.	
	Graduate Student Housing <i>Community Associate</i>	Stanford University Winter 2011 – Spring 2013
	Planned and led events for 800 graduate students.	
	Judicial Affairs <i>Juror</i>	Stanford University Fall 2009 – Spring 2010
Editorial	Associate Editor	Foundations and Trends in Machine Learning 2024–
	Associate Editor	Operations Research (OR) 2024–
	Associate Editor	SIAM Journal on Mathematics of Data Science (SIMODS)

2023–

Technical Editor

Mathematical Programming Computation (MPC)

2021–

2025: NeurIPS Area Chair; NeurIPS Datasets and Benchmarks Senior Area Chair; ICML Senior Area Chair; ...

2024: NeurIPS Area Chair; ICML Senior Area Chair; AFRL; Journal of Machine Learning Research; Math Programming Computation (2); SIAM Journal on Matrix Analysis; Statistics and Computation ...

2023: NeurIPS Area Chair; NeurIPS Datasets and Benchmarks Senior Area Chair; Journal of the Royal Statistical Society; Journal of Machine Learning Research; Math Programming Computation (2); Math of OR; DeepMath; International Journal of Computer Vision ...

2022: Math of OR (2); Operations Research; IEEE transactions on signal processing; IEEE transactions on automatic control; NeurIPS Area Chair; technical editor for three papers at MPC.

2021: Linear Algebra and Applications (LAA); Proceedings of the National Academy of Sciences (PNAS); Mathematics of Operations Research (MOR); SIAM Journal on the Mathematics of Data Science (SIMODS) (2); NeurIPS area chair; Statistics and Computing; SIAM Journal on Matrix Analysis (SIMAX).

2020: SIAM Journal on Optimization (SIOPT); Journal of the American Statistical Society (JASA); Annals of Applied Statistics (AOAS); Linear Algebra and Applications (LAA); SIAM Journal on the Mathematics of Data Science (SIMODS); Science Advances; TPAMI Special Issue on AutoML (2); Proceedings of the National Academy of Sciences (PNAS). Conferences: Learning for Dynamics and Control (6); NeurIPS (6). Grants: NSF RI, ONR.

2019: Journal of the American Statistical Society (JASA); Annals of Applied Statistics (AOAS); SIAM Journal on the Mathematics of Data Science (SIMODS) (2); Journal of Statistical Software (JSS); Optimization Letters (OPTL); SIAM Journal on Matrix Analysis and Applications (SIMAX) (2); SIAM Journal on Optimization (SIOPT). Conferences: Learning for Dynamics and Control (5); ICML (3); NeurIPS (6); AAAI (3). Grants: NSF RI.

2018: SIAM Review; SIAM Journal on Scientific Computing; Linear Algebra and Applications; Stochastic Systems; NeurIPS (6); AAAI meta-reviewer (23); SysML (now MLSys).

2017: Mathematical Programming; NIPS (6); SIAM Journal on Matrix Analysis; AAAI metareviewer (26); ICDM Workshop on Data Driven Discovery of Models (2); Journal of Statistical Software; Automatica; Springer book chapters (2);

Industry Experience

Technical Advisor

Oakland, CA

Orchestrated Intelligence

2022 –

Advised on data-driven and robust solutions for supply chain modeling and optimization.

Technical Consultant

Stanford, CA

Two Sigma

2021 –

Developed and advised on algorithms for risk management and portfolio optimization.

Visiting Researcher

Ithaca, NY and Mountain View, CA

Google Cloud AI Research

2021

Research ideas to accelerate and improve deep learning for tabular and time-series data to solve high-impact business problems in finance, manufacturing, retail, and beyond.

Technical Advisor

Santa Monica, CA

Retina AI 2017 – 2022
Advised on technical solutions for problems in e-commerce, including rapid assessment of long term value, retention analysis, and customer segmentation.

Technical Advisor Palo Alto, CA
Aurora Solar 2014 – 2019
Designed optimization algorithms tailored for problems in the solar industry, including design of efficient rooftop photovoltaic array configurations. Compared to designs produced by solar installation experts, the resulting optimized designs deliver the same energy output at lower cost for more than 70% of homes.

Senior Research Scientist San Francisco, CA
Qadium (renamed Expanse, acquired by Palo Alto Networks) 2012 – 2015
Won grants exceeding \$6.5M from DARPA for research in data analytics and cybersecurity.

Lead Data Scientist Arlington, VA
DARPA (via Data Tactics) 2012 – 2013
Wrote 3 white papers to define mission for \$100M DARPA cybersecurity program.

Data Scientist Chicago, IL
Obama for America Fall 2011
Analyzed graph of 70M Facebook users to identify potential donors and target voter registration campaign.

Research Scientist San Mateo, CA
Apixio (acquired by Centene) Summer 2011
Developed a tool to extract structured information about diseases from the unstructured text of doctors' notes.

Sales and Trading Strategist New York, NY
Goldman Sachs Summer 2009
Corrected model of commodities derivatives risk using multiple parameter estimation.

Market Risk Management Analyst New York, NY
Goldman Sachs Summer 2008
Designed and automated a system to evaluate and graph mutual fund risk.

Presentations (Plenary) **International Conference on Preconditioning Techniques for Scientific and Industrial Applications**, Edinburgh 2026
Plenary: Learning a Preconditioner

Mathematics of Operations Research, Soesterland, Netherlands 2025
Keynote: Big Data is Low Rank
Keynote: Hunting the Hessian
Plenary: AI and the Future of Optimization

TILOS workshop on optimization for deep learning, UCSD 2025
Hunting the Hessian

AlgoPerf Workshop, Meta, Menlo Park 2025
Hunting the Hessian

MOPTA, Lehigh 2024
Plenary: Automated Optimization Modeling

AutoML Conference, Berlin 2023
Plenary: Recommending Learners

East Coast Optimization Meeting , (online) <i>Keynote: Scalable Semidefinite Programming</i>	2021
JuliaCon , Baltimore <i>Keynote: Big Data is Low Rank using LowRankModels</i>	2019
Women in Data Science , Stanford <i>Plenary: Big Data is Low Rank</i> 100,000 conference attendees worldwide!	2019
JuliaCon , Berkeley <i>Plenary: Julia: the Type of Language for Mathematical Programming</i>	2017
MIIS (Tutorial and Invited Talk) , Chinese University of Hong Kong, Shenzhen	2016

**Presentations
(Invited)**

INFORMS , Atlanta <i>Interpretable Prediction and Feature Selection for Survival Analysis</i>	2025
KDD , Toronto <i>Interpretable Prediction and Feature Selection for Survival Analysis</i>	2025
International Conference on Continuous Optimization (ICCOPT) , USC <i>Online Scaled Gradient Method</i>	2025
Strategies for Handling Applications with Nonconvexity , Banff <i>Online Scaled Gradient Method</i>	2025
INFORMS , Seattle <i>AI and the Future of Optimization Modeling</i>	2024
Discrete Optimization Talks (DOTs) , https://talks.discreteopt.com/ <i>Automated Optimization Modeling</i>	2024
ORAI , China (online) <i>Automated Optimization Modeling</i>	2024
AAAI Workshop on AI in OR , Vancouver <i>AI and the Future of Optimization Modeling</i>	2024
Information Theory and Applications (ITA) , San Diego <i>AI and the Future of Optimization Modeling</i>	2024
Knowledge Discovery and Data Mining (KDD) , Long Beach <i>Panel on AI in Education</i>	2023
International Conference on Stochastic Programming , UC Davis <i>Low rank approximation for faster convex optimization</i>	2023
Kavli frontiers of science symposium , Irvine <i>Low rank approximation for faster convex optimization</i>	2023
One World MINDS seminar , online <i>Low rank approximation for faster convex optimization</i>	2023
SIAM Math of Data Science , San Diego (online) <i>Low rank approximation for faster convex optimization</i>	2022
Aspects of Logic and Machine Learning , ASL, Cornell University, <i>Automating Machine Learning</i>	2022
ICME XPO , Stanford University <i>Automating Machine Learning</i>	2022
Workshop on Automated Data Science , ECML (online),	2021

Structured Models for Automated Machine Learning

Workshop on Low Rank models and Applications , Fields Institute (online) <i>Imputing Missing Data with the Gaussian Copula</i>	2021
Complexity of Matrix Computation Seminar , (online) <i>Panelist: What does it mean to compute a low rank approximation of a matrix?</i>	2021
E-NLA Numerical Linear Algebra Seminar , (online) <i>Scalable Semidefinite Programming</i>	2021
Mathematical Foundations and Algorithms for Tensor Computations IPAM, UCLA (online) <i>Low Rank Tucker Approximation of a Tensor from Streaming Data</i>	2021
Women in Data Science Workshop , WiDS Global Conference (online) <i>Automating Machine Learning</i>	2021
OPTML++ Seminar , MIT (online) <i>Big Data is Low Rank</i>	2020
QMnet Seminar , Melbourne (online) <i>Missing Data Imputation with Low Rank Models</i>	2020
The Art of Learning with Missing Values (ARTEMISS) Workshop , ICML (online) <i>Imputing Missing Data with the Gaussian Copula</i>	2020
SIAM Mathematics of Data Science , (online) <i>Imputing Missing Data with the Gaussian Copula</i>	2020
Science on Tap , Ithaca, NY <i>Filling in Missing Data: Elections, _____, Healthcare.</i>	2020
Low-rank models winter school , Villars-Sur-Ollon, Switzerland <i>Low Rank Models for Missing Data and Optimization</i>	2020
Statistics and Computation , Alan Turing Institute, London <i>Big Data is Low Rank</i>	2020
Reunion Conference on Foundations of Data Science , Simons Institute <i>Missing Value Imputation for Mixed Data Through Gaussian Copula</i>	2019
INFORMS , Seattle <i>Low Rank Tucker Approximation of a Tensor from Streaming Data</i>	2019
Knowledge Discovery and Data Mining (KDD) , Anchorage <i>Oboe: Collaborative Filtering for AutoML Initialization</i>	2019
Machine Learning for Health (ML4H) , Vector Institute, Toronto <i>Representation Learning, Patient Similarity, and Subtyping</i>	2019
Low Rank Optimization Workshop , Leipzig MPI for Mathematics in the Sciences <i>Low Rank Tucker Approximation of a Tensor from Streaming Data</i>	2019
Optimization and Statistical Learning , Les Houches <i>Optimal-Storage Semidefinite Programming using Approximate Complementarity</i>	2019
Women and Mathematics (WAM) Ambassador Program , Cornell University <i>Filling in Missing Data: Elections, _____, Healthcare.</i>	2019
NeurIPS workshop on AI in financial services , Montreal <i>Moderated Industry Panel</i>	2018
NeurIPS workshop on AI in financial services , Montreal	2018

Fairness under Unawareness

NeurIPS spotlight talk , Montreal <i>Limited Memory Kelley's Method Converges for Composite Convex and Submodular Optimization</i>	2018
ISMP , Bordeaux <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2018
DARPA D3M Workshop , Arlington <i>Composable Robust Structured Data Inference: AutoML, Causal Inference, Big Data is Low Rank</i>	2018
AI in advancement , Cornell <i>Panel Discussion</i>	2018
INFORMS , Houston <i>Optimal Design of Rooftop Photovoltaic Arrays</i>	2017
SIMONS Institute , Berkeley <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
Schonfeld Quantitative Conference , New York <i>Convex Optimization Modeling</i>	2017
STRATA , New York <i>Generalized Low Rank Models</i>	2017
LCCC workshop on Distributed Optimization (Invited) , Lund <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
SIOPT , Vancouver <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
DARPA D3M Kickoff , Arlington <i>Composable Robust Structured Data Inference</i>	2017
Optimization Under Uncertainty Workshop , Duke <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
Yale Alumni in Science and Engineering Talk , New York <i>Filling in Missing Data: Elections, ———, Healthcare.</i>	2017
INFORMS , Nashville <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2016
ICCOPT , Tokyo <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2016
SIAM Annual Meeting , Boston <i>Generalized Low Rank Models</i>	2016
DARPA ISAT Workshop on the Future of Storage , New York <i>Low Rank Models for Compression</i>	2016
TDA 2016 , Leuven <i>Generalized Low Rank Models</i>	2016
DARPA SIMPLEX program meeting , Stanford University <i>Generalized Low Rank Models</i>	2015
H2O World , Santa Clara <i>Generalized Low Rank Models</i>	2015
INFORMS , Philadelphia	2015

Bounding Duality Gap for Separable Problems with Linear Constraints

ISMP , Pittsburgh	2015
<i>Convex optimization in Julia</i>	
Optimization in Julia , JuliaCon, Cambridge	2015
<i>Convex optimization in Julia</i>	
Bay Area Julia Users Meetup , San Francisco	2014
<i>Convex optimization in Julia</i>	
HPTCDL Workshop , SC14, New Orleans	2014
<i>Convex optimization in Julia</i>	
INFORMS , San Francisco	2014
<i>Sigmoidal Programming</i>	
IPAM Workshop on Optimization , UCLA	2013
<i>Sigmoidal Programming</i>	

Presentations (Contributed)	SIAM optimization conference , Seattle	2023
	<i>Detecting equivalence between iterative algorithms for optimization</i>	

Department Seminars	IEOR-DRO Seminar , Columbia	2025
	<i>Hunting the Hessian</i>	
	Quantitative Sciences Unit Seminar , Stanford University	2025
	<i>Interpretable Survival Analysis</i>	
	Wharton Optimization Seminar , Philadelphia	2024
	<i>AI and the Future of Optimization Modeling</i>	
	ISE Seminar , USC	2024
	<i>Low rank approximation for faster convex optimization</i>	
	Keller Distinguished Lecture , Caltech	2024
	<i>Low rank approximation for faster convex optimization</i>	
	EE faculty meeting , Stanford University	2024
	<i>AI and the Future of Optimization Modeling</i>	
	BCG Tech Talk , online	2024
	<i>AI and the Future of Optimization Modeling</i>	
	MS&E Alumni Weekend , Stanford University(online)	2024
	<i>AI and the Future of Optimization Modeling</i>	
	Biostatistics Seminar , University of Tennessee Health Sciences (online)	2024
	<i>Big Data is Low Rank</i>	
	Applied Math Seminar , UC Santa Cruz	2023
	<i>Low rank approximation for faster convex optimization</i>	
	Applied Math Seminar , UC Boulder	2023
	<i>Detecting equivalence between iterative algorithms for optimization</i>	
	SILO Seminar , UW Madison	2023
	<i>Detecting equivalence between iterative algorithms for optimization</i>	
	ISL seminar , Stanford University	2023
	<i>Low rank approximation for faster convex optimization</i>	
	Two Sigma Academic Seminar (online)	2022
	<i>Low rank approximation for faster convex optimization</i>	

Statistics Seminar , University of Chicago (online) <i>Low rank approximation for faster convex optimization</i>	2022
ORC Seminar , MIT Sloan (online) <i>Big Data is Low Rank</i>	2022
ORFE Seminar , Princeton (online) <i>Big Data is Low Rank</i>	2022
MS&E Seminar , Stanford (online) <i>Big Data is Low Rank</i>	2022
Ocurate / PredictWise Seminar , online <i>Imputing Missing Data with the Gaussian Copula</i>	2021
Intel Site Visit , Cornell University <i>Pareto optimization to pick the perfect precision</i>	2021
Data Science Seminar , Johns Hopkins (online) <i>Detecting equivalence between iterative algorithms for optimization</i>	2021
IEOR-DRO Seminar , Columbia (online) <i>Detecting equivalence between iterative algorithms for optimization</i>	2021
IOE Seminar , Michigan (online) <i>Detecting equivalence between iterative algorithms for optimization</i>	2021
AI seminar , Microsoft Research (online) <i>Structured Models for Automated Machine Learning</i>	2021
Industrial Engineering Seminar , Sharif University of Technology (online) <i>Big Data is Low Rank</i>	2021
Optimization and Statistical Learning Seminar , Northwestern (online) <i>Detecting equivalence between iterative algorithms for optimization</i>	2021
Statistics Seminar , University of California at Santa Barbara (online) <i>Big Data is Low Rank</i>	2021
AI seminar , Cornell University(online) <i>Automating Machine Learning</i>	2021
Computational Science and Engineering Seminar , Georgia Tech (online) <i>Big Data is Low Rank</i>	2021
Computational Mathematics and Applications Seminar , Oxford (online) <i>Big Data is Low Rank</i>	2021
Ezra Systems Seminar , Cornell (online) <i>Imputing Missing Data with the Gaussian Copula</i>	2020
Scientific Computing Seminar , Emory (online) <i>Imputing Missing Data with the Gaussian Copula</i>	2020
Mathematics of Data Science Math Seminar , Tufts (online) <i>Big Data is Low Rank</i>	2020
Applied Math Seminar , Princeton <i>Scalable Semidefinite Programming</i>	2020
Applied Math Seminar , UC Boulder <i>Optimal-Storage Semidefinite Programming using Approximate Complementarity</i>	2019
CME 300 , Stanford <i>Big Data is Low Rank</i>	2019

Johns Hopkins AMS seminar , Baltimore <i>Big Data is Low Rank</i>	2019
CAM Colloquium , Cornell University <i>Low Memory Convex Optimization</i>	2019
Rutgers Optimization Seminar , New Brunswick <i>Low Memory Convex Optimization</i>	2018
Princeton Optimization Seminar , Princeton <i>Low Memory Convex Optimization</i>	2018
UC Davis Mathematics of Data and Decisions Seminar , Davis <i>Big Data is Low Rank</i>	2018
Georgia Tech OR Colloquium , Atlanta <i>Big Data is Low Rank</i>	2018
Stanford Linear Algebra and Optimization Seminar , Stanford <i>Low Memory Convex Optimization</i>	2018
Ecole Polytechnique: Statistics Special Seminar , Paris <i>Big Data is Low Rank</i>	2018
Penn State OR Colloquium , State College, PA <i>Big Data is Low Rank</i>	2018
Cornell Engineering College Council , New York, <i>The New Educational Paradigm: Data Science</i>	2017
MIT ORC Seminar , Cambridge, MA <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
Capital One Tech Talk , New York <i>Low Rank Models for Automatic Machine Learning and Interpretability</i>	2017
Two Sigma Tech Talk , New York <i>Generalized Low Rank Models</i>	2017
CATALYST Academy Field Session: Operations Research , Cornell <i>Outreach session to introduce URM high school students to the discipline of OR</i>	2017
CURIE Academy Field Session: Operations Research , Cornell <i>Outreach session to introduce female high school students to the discipline of OR</i>	2017
UW Optimization Seminar , Seattle <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
NYU Numerical Analysis Seminar , New York <i>Sketchy Decisions: Convex Optimization with Optimal Storage</i>	2017
Goldman Sachs Tech Talk , New York	2017
CS Brown-Bag Colloquium , Cornell	2017
SCAN Seminar , Cornell	2016
CAM Colloquium , Cornell	2016
JPL Seminar , Pasadena	2016
Kaiser Permanente , Oakland	2016
CMI Seminar (I) , California Institute of Technology	2015
CMI Seminar (II) , California Institute of Technology	2015
Uber Tech Talk , San Francisco	2015

Applied Math Seminar , UCLA	2015
Sandia National Lab Seminar , Livermore	2015
Google Tech Talk , Mountain View	2015
Biomedical Informatics Seminar , Stanford University	2015
Palantir Tech Talk , Palo Alto	2015
Twitter Tech Talk , San Francisco	2015
ICME PhD Oral Examination , Stanford University	2015
H2O Tech Talk , Santa Clara	2015
Civis Analytics Tech Talk , Chicago	2015
TTIC Seminar , Toyota Technical Institute of Chicago	2015
IBM T. J. Watson Research Seminar , Yorktown Heights	2015
Hutchin Hill Capital Seminar , New York	2015
ORIE Seminar , Cornell University	2015
IEOR Seminar , UC Berkeley	2015
CMS Seminar , California Institute of Technology	2015
Heinz College Seminar , Carnegie Mellon University	2015
Mobilize Seminar , Stanford University	2014
ICME Seminar , Stanford University	2014
BlackRock SAE Tech Talk , Stanford University	2014
JuliaCon , Chicago	2014
Verizon Labs Tech Talk , Palo Alto	2014
ICME Seminar , Stanford University	2013
Marin Software Tech Talk , San Francisco	2013
Political Psychology Research Seminar , Stanford University	2013
ICME Student Seminar , Stanford University	2010

Numbers **Erdős number: 3**

Bacon number: 3

Erdős-Bacon number: 6

Hobbies Harp, backpacking, running, foraging, ergonomics, carbon sequestration.