

Cynthia Rudin

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My work focuses on **interpretable machine learning**, which is crucial for responsible and trustworthy AI. This includes the design of algorithms for interpretable modeling, interpretable policy design, variable importance measures, causal inference methods, and methods that can incorporate domain knowledge-based constraints into machine learning. These techniques are applied to critical societal problems in healthcare, criminal justice, energy grid reliability, and other areas. The interpretable machine learning algorithms heavily rely on efficient discrete optimization techniques.

In order to encourage the use of interpretable models, I aim to: (1) design interpretable machine learning methods that are as accurate as the black box methods (e.g., sparse decision trees, scoring systems and interpretable neural networks), (2) provide a theoretical foundation for why interpretable-yet-accurate models exist for most real data problems in supervised learning, (3) warn the greater ML world of the danger of using black box models in high stakes decisions, and provide educational tools on interpretable machine learning (via online course materials).

I am the winner of the 2022 *Squirrel AI Award for Artificial Intelligence for the Benefit of Humanity* from the Association for the Advancement of Artificial Intelligence (AAAI). I am also a 2022 Guggenheim fellow, and a three-time winner of the INFORMS Innovative Applications in Analytics award based on my applied projects with domain experts. My team won second place in the prestigious 2023 Bell Labs Prize. Some accomplishments:

- The Rashomon Set Paradigm: My collaborators and I introduced a new paradigm for machine learning – instead of the algorithm providing one model (as usual), we return *all* good models, allowing the user to choose between them. This eliminates the *interaction bottleneck* to users. It allows users to easily obtain models with fairness constraints and monotonicity constraints – by taking a simple loop over the good models. We developed this for sparse decision trees and sparse generalized additive models (GAMs). This relies on a decade of work with Margo Seltzer and our students on algorithms for optimal sparse decision trees and generalized additive models.
- Why Interpretable Models Are Accurate: Ron Parr, Lesia Semenova and I established a framework for why interpretable models are as accurate as deep learning or other black box models for most tabular data sets.
- Risk Scores: Our algorithms for sparse risk scores were applied to important healthcare and criminal justice applications. My team’s work on seizure prediction in ICU patients allowed for better allocation of patient monitoring resources. This work won the 2019 INFORMS Innovative Applications in Analytics Award. It helps to prevent severe brain damage in critically ill patients, and yielded the only AI model widely used in critical care brain monitoring.
- Energy Reliability: I led the first major effort to maintain an underground electrical distribution network using ML, in work with Con Edison in NYC. This won the 2013 INFORMS Innovative Applications in Analytics Award.
- Crime Series Analysis: My collaborators and I developed code for detecting crime series in cities. This methodology (specifically, the Series Finder algorithm) was adapted by the NYPD and their application (Patternizr) has been running live in NYC since 2016 for determining whether each new crime is related to past crimes.
- Margins for AdaBoost: I solved a well-known (previously open) theoretical problem in machine learning as a PhD student, which is whether AdaBoost maximizes the ℓ_1 -margin. Subsequent work solved a COLT open problem.
- I enjoy competing in data science competitions and coaching student teams. We have won awards including the FICO Recognition Award for the Explainable Machine Learning Challenge (2018), NTIRE Superresolution Competition (2018), PoeTix Literary Turing Competition (2018), and American Statistical Association Data Challenge Expo Student Competition (2022 and 2023).
- I have given invited/keynote/plenary talks at INFORMS, KDD (2014 and 2019), AISTATS, ECML-PKDD, ML in Healthcare, FAT-ML (Fairness, Accountability, and Transparency), SIAM International Conference on Data Mining (SDM), and the Nobel Conference. My work appears in the media, including the NY Times, Washington Post, Boston Globe, Wall Street Journal, MSNBC, and National Public Radio.
- I am currently a member of the National AI Advisory Committee’s Subcommittee on Law Enforcement. I am a member of the National Academies Committee on Facial Recognition Technology that just released a report.
- I am writing an e-book on introductory machine learning called “Intuition for the Algorithms of Machine Learning,” which teaches interpretable machine learning as part of the basic ML curriculum, linked on my website.

Education

- **Ph.D.: Princeton University.** Program in Applied and Computational Mathematics. Title: **Boosting, Margins, and Dynamics**. Advisors: Ingrid Daubechies and Robert Schapire.
- **BS/BA: University at Buffalo (SUNY), Honors Program.** Outstanding Senior Award in the Arts and Sciences (one awarded per year university-wide), separate outstanding senior awards from the Physics Department, Mathematics Department, and Music Department (one awarded per department per year), BS Mathematical Physics, BA Music Theory, Minor in Computer Science, Summa Cum Laude, 1999

Employment History

- **Duke University**, Computer Science Department (50%), Electrical and Computer Engineering Department (50%), Secondary appointments in Statistical Science, Biostatistics & Bioinformatics, and Mathematics. Associate Professor 2016-2019, Professor 2019-present.
- **Massachusetts Institute of Technology**, MIT Computer Science and Artificial Intelligence Laboratory and Sloan School of Management, Associate Professor of Statistics 2013-2016, Assistant Professor of Statistics 2009-2013.
- **Columbia University**, Center for Computational Learning Systems, Associate Research Scientist, 2007-2009.
- **NSF Postdoctoral Research Fellow, New York University**, 2004-2007.

Peer-Reviewed Publications

Papers associated with major awards (winner and finalist)

1. Yiyang Sun, Zhi Chen, Vittorio Orlandi, Tong Wang and Cynthia Rudin. **Sparse and Faithful Explanations without Sparse Models.** *Winner of Data Mining Best Paper Award, INFORMS 2023*, AISTATS, 2024.
2. Edwin Agnew, Michelle Qiu, Lily Zhu, Sam Wiseman and Cynthia Rudin. **The Mechanical Bard: An Interpretable Machine Learning Approach to Shakespearean Sonnet Generation**, Meeting of the Association for Computational Linguistics (ACL), 2023. *Outstanding Paper Award.*
3. Gah-Yi Ban and Cynthia Rudin. *MSOM Best OM paper in OR Award, 2021, INFORMS.* (Awarded to the best paper in Operations Research within the last 3 years, awarded by Manufacturing and Service Operations Management Society of INFORMS.) **The Big Data Newsvendor: Practical Insights from Machine Learning**, Operations Research, Vol. 67, No. 1, pages 90-108, 2019.
4. Aaron F. Struck, Berk Ustun, Andres Rodriguez Ruiz, Jong Woo Lee, Suzette LaRoche, Lawrence J. Hirsch, Emily J Gilmore, Jan Vlachy, Hiba Arif Haider, Cynthia Rudin, M Brandon Westover. *2019 INFORMS Innovative Applications in Analytics Award* (shared with paper below). **Association of an Electroencephalography-Based Risk Score With Seizure Probability in Hospitalized Patients**, JAMA Neurology, 74 (12), 1419-1424, 2017.
5. Berk Ustun and Cynthia Rudin. *2019 INFORMS Innovative Applications in Analytics Award*, also *2017 INFORMS Computing Society Student Paper Prize.* **Learning Optimized Risk Scores.** Journal of Machine Learning Research, 2019. Shorter version **Learning Optimized Risk Scores from Large-Scale Datasets.** Knowledge Discovery in Databases (KDD), 2017.
6. Chaofan Chen, Kangcheng Lin, Cynthia Rudin, Yaron Shaposhnik, Sijia Wang, Tong Wang. **A Holistic Approach to Interpretability in Financial Lending: Models, Visualizations, and Summary-Explanations.** Decision Support Systems, 2021.
Preliminary work:
 - Chaofan Chen, Kangcheng Lin, Cynthia Rudin, Yaron Shaposhnik, Sijia Wang, Tong Wang. *Winner of the FICO Recognition Award for the Explainable Machine Learning Challenge, 2018.* **An Interpretable Model with Globally Consistent Explanations for Credit Risk.** NIPS 2018 Workshop on Challenges and Opportunities for AI in Financial Services: the Impact of Fairness, Explainability, Accuracy, and Privacy, 2018.

7. Cynthia Rudin and Berk Ustun. *Finalist for 2017 Daniel H. Wagner Prize for Excellence in Operations Research, Institute for Operations Research and Management Science (INFORMS). Optimized Scoring Systems: Towards Trust in Machine Learning for Healthcare and Criminal Justice*. INFORMS Journal on Applied Analytics, Special Issue: 2017 Daniel H. Wagner Prize for Excellence in Operations Research Practice, 48(5), pages 399–486, September–October, 2018.
8. William Souillard-Mandar, Randall Davis, Cynthia Rudin, Rhoda Au, David J. Libon, Rodney Swenson, Catherine C. Price, Melissa Lamar, Dana L. Penney. *2016 INFORMS Innovative Applications in Analytics Award* (shared with paper below). **Learning Classification Models of Cognitive Conditions from Subtle Behaviors in the Digital Clock Drawing Test**. Machine Learning, volume 102, number 3, 2016.
9. Berk Ustun and Cynthia Rudin. *2016 INFORMS Innovative Applications in Analytics Award* and also *Runner up, Invenia Labs SEE Award 2018 - Supporting Machine Learning Research with a Positive Impact on Social, Economic, or Environmental (SEE) Challenges*. **Supersparse Linear Integer Models for Optimized Medical Scoring Systems**. Machine Learning, volume 102, number 3, 2016.
10. Cynthia Rudin, Şeyda Ertekin, Rebecca Passonneau, Axinia Radeva, Ashish Tomar, Boyi Xie, Stanley Lewis, Mark Riddle, Debbie Pangsrivini, Tyler McCormick. *2013 INFORMS Innovative Applications in Analytics Award* **Analytics for Power Grid Distribution Reliability in New York City**. INFORMS Journal on Applied Analytics, volume 44, issue 4, pages 364–383, 2014.
11. Indraneel Mukherjee, Cynthia Rudin, and Robert E. Schapire. *This paper answered an open question published in COLT 2010*. **The Rate of Convergence of AdaBoost**, Journal of Machine Learning Research, volume 14, pages 2315–2347, August 2013.
Preliminary work:
 - Indraneel Mukherjee, Cynthia Rudin, and Robert E. Schapire. **The Rate of Convergence of AdaBoost**, Proceedings of the 24th Annual Conference on Learning Theory (COLT), 2011.

Publications and other best paper awards (not including those above)

2024

12. Jon Donnelly, Jon, Luke Moffett, Alina Barnett, Hari Trivedi, Fides Regina Schwartz, Joseph Lo, Cynthia Rudin. **Asym-Mirai: Interpretable Mammography-Based Deep Learning Model for 1- to 5-year Breast Cancer Risk Prediction**, Radiology, in press, 2024
13. Travis Seale-Carlisle, Saksham Jain, Courtney Lee, Caroline Levenson, Swathi Ramprasad, Brandon Garrett, Sudeepa Roy, Cynthia Rudin, Alexander Volfovsky **Evaluating Pre-trial Programs Using Machine Learning Matching Algorithms**, AAAI (Oral), 2024.
14. Siong Thye Goh, Lesia Semenova, and Cynthia Rudin. **Sparse Density Trees and Lists: An Interpretable Alternative to High-Dimensional Histograms**, INFORMS Journal on Data Science, 2024.
15. Srikanth Katta, Harsh Parikh, Cynthia Rudin, Alexander Volfovsky. **Interpretable Causal Inference for Analyzing Wearable, Sensor, and Distributional Data**. *2024 Joint Statistical Meeting Paper Award, American Statistical Association, Biometrics Section*, AISTATS, 2024.
16. Rui Zhang, Rui Xin, Margo Seltzer, and Cynthia Rudin. **Optimal Sparse Survival Trees**, AISTATS, 2024.
17. Harsh Parikh, Quinn Lanners, Zade Akkas, Sahar Zafar, M Brandon Westover, Cynthia Rudin, Alexander Volfovsky, **Safe and Interpretable Estimation of Optimal Treatment Regimes**, AISTATS, 2024.
18. Cheng Ding, Zhicheng Guo, Cynthia Rudin, Ran Xiao, Amit Shah, Duc H. Do, Randall J Lee, Gari Clifford, Fadi B Nahab, Xiao Hu **Learning From Alarms: A Robust Learning Approach for Accurate Photoplethysmography-Based Atrial Fibrillation Detection using Eight Million Samples Labeled with Imprecise Arrhythmia Alarms**, IEEE Journal of Biomedical and Health Informatics (JBHI), accepted, 2024.
19. Ashokkumar Manickam, Jackson J Peterson, Yuriko Harigaya, David M Murdoch, David M Margolis, Alex Oesterling, Zhicheng Guo, Cynthia D Rudin, Yuchao Jiang, and Edward P Browne. **Integrated single-cell multiomic analysis of HIV latency reversal reveals novel regulators of viral reactivation**, Genomics, Proteomics, and Bioinformatics, 2024.

2023

20. Brandon L Garrett and Cynthia Rudin. **Interpretable Algorithmic Forensics**, The Proceedings of the National Academy of Sciences (PNAS), 2023.
21. Lesia Semenova, Harry Chen, Ronald Parr, Cynthia Rudin. **A Path to Simpler Models Starts With Noise**, NeurIPS, 2023.
22. Zhi Chen, Chudi Zhong, Margo Seltzer, Cynthia Rudin. **Exploring and Interacting with the Set of Good Sparse Generalized Additive Models**, NeurIPS, 2023.
23. Chiyu Ma, Brandon Zhao, Chaofan Chen, Cynthia Rudin. **This Looks Like Those: Illuminating Prototypical Concepts Using Multiple Visualizations**, NeurIPS, 2023.
24. Jiachang Liu, Sam Rosen, Chudi Zhong, Cynthia Rudin. **OKRidge: Scalable Optimal k-Sparse Ridge Regression for Learning Dynamical Systems**, NeurIPS (Spotlight), 2023.
25. Jon Donnelly, Srikar Katta, Cynthia Rudin, Edward P Browne. **The Rashomon Importance Distribution: Getting RID of Unstable, Single Model-based Variable Importance**, NeurIPS (Spotlight), 2023.
26. Stephen Hahn, Rico Zhu, Yue Jiang, Simon Mak, and Cynthia Rudin. **New Orleans: An Adventure in Music**, NeurIPS Creative AI track. (demo) 2023.
27. Stephen Hahn, Rico Zhu, Simon Mak, Cynthia Rudin, Yue Jiang. **An Interpretable, Flexible, and Interactive Probabilistic Framework for Melody Generation**, KDD 2023.
28. Dennis Tang, Frank Willard, Ronan Tegerdine, Luke Triplett, Jon Donnelly, Luke Moffett, Lesia Semenova, Alina Jade Barnett, Jin Jing, Cynthia Rudin, and M. Brandon Westover. **ProtoEEGNet: An Interpretable Approach for Detecting Interictal Epileptiform Discharges**, Medical Imaging meets NeurIPS Workshop (MedNeurIPS), (Oral), 2023.
29. Samantha M. McDonald, Emily K. Augustine, Quinn Lanners, Cynthia Rudin, L. Catherine Brinson, and Matthew L. Becker. **Applied Machine Learning as a Driver for Polymeric Biomaterials Design**, Nature Communications, 2023.
30. Jacob Peloquin, Alina Kirillova, Cynthia Rudin, L.C. Brinson, Ken Gall. **Prediction of tensile performance for 3D printed photopolymer gyroid lattices using structural porosity, base material properties, and machine learning**, Materials & Design, 2023.
31. Jacob Peloquin, Alina Kirillova, Elizabeth Mathey, Cynthia Rudin, L. Catherine Brinson, Ken Gall. **Tensile performance data of 3D printed photopolymer gyroid lattices**, Data in Brief, 2023.
32. Quinn Lanners, Harsh Parikh, Alexander Volfovsky, Cynthia Rudin, David Page. **From Feature Importance to Distance Metric: An Almost Exact Matching Approach for Causal Inference**, UAI, 2023.
A shorter version appears at this workshop:
 - Quinn Lanners, Harsh Parikh, Alexander Volfovsky, Cynthia Rudin, and David Page. **Matching Using Feature Importance: An Auditable Approach to Causal Inference**, 9th International Conference on Computational Social Science IC²S², 2023
33. Harsh Parikh, Kentaro Hoffman, Haoqi Sun, Wendong Ge, Jin Jing, Rajesh Amerineni, Lin Liu, Jimeng Sun, Sahar Zafar, Aaron Struck, Alexander Volfovsky, Cynthia Rudin, M. Brandon Westover. **Effects of Epileptiform Activity on Discharge Outcome in Critically Ill Patients: A Retrospective Cross-Sectional Study**, The Lancet Digital Health, 2023.
34. Rui Zhang, Rui Xin, Margo Seltzer, and Cynthia Rudin. **Optimal Sparse Regression Trees**, AAAI, 2023.
35. Cynthia Rudin and Yaron Shaposhnik. **Globally-Consistent Rule-Based Summary-Explanations for Machine Learning Models: Application to Credit-Risk Evaluation**, Journal of Machine Learning Research, 2023.
 - An earlier version appeared at the Conference on Information Systems and Technology (CIST), 2019.
36. Fernanda Bravo, Cynthia Rudin, Yuting Yuan and Yaron Shaposhnik. **Interpretable Prediction Rules for Congestion Risk in ICUs**. Stochastic Systems, 2023.

- Conference version: Fernanda Bravo, Cynthia Rudin, Yuting Yuan and Yaron Shaposhnik. **Simple Rules for Predicting Congestion Risk in Queueing Systems: Application to ICUs**, 2019 INFORMS Workshop on Data Science (DS 2019), (Oral).

37. Shane D Falcinelli, Alicia Volkheimer, Lesia Semenova, Ethan Wu, Alexander Richardson, Manickam Ashokkumar, David M Margolis, Nancie M Archin, Cynthia D Rudin, David Murdoch, Edward P Browne. **Impact of cannabis use on immune cell populations and the viral reservoir in people with HIV on suppressive antiretroviral therapy**, The Journal of Infectious Disease (JID), 2023.
38. Yanchen Jessie Ou, Alina Jade Barnett, Anika Mitra, Fides Regina Schwartz, Chaofan Chen, Lars Grimm, Joseph Y. Lo, Cynthia Rudin. **A user interface to communicate interpretable AI decisions to radiologists**. Proceedings of SPIE Medical Imaging, 2023.
39. Edward W. Felten, Jennifer Mnookin, Thomas D. Albright, Ricardo Baeza-Yates, Bob Blakey, Patrick Grother, Marvin B. Haiman, Aziz Z. Huq, Anil K. Jain, Elizabeth E. Joh, Michael C. King, Nicol Turner Lee, Ira S. Reese, and Cynthia Rudin. **Facial Recognition Technology: Current Capabilities, Future Prospects, and Governance**, National Academies Press, 2024. (Consensus study report of the National Academies of Sciences Engineering, and Medicine)

2022

40. Rui Xin, Chudi Zhong, Zhi Chen, Takuya Takagi, Margo Seltzer, Cynthia Rudin. **Exploring the Whole Rashomon Set of Sparse Decision Trees**, *Finalist for INFORMS 2022 Data Mining Best Paper Competition Award, Student Track*, NeurIPS (Oral), 2022.
41. Jiachang Liu, Chudi Zhong, Boxuan Li, Margo Seltzer, Cynthia Rudin. **FasterRisk: Fast and Accurate Interpretable Risk Scores**, NeurIPS, 2022.
42. Zijie Wang, Chudi Zhong, Rui Xin, Takuya Takagi, Zhi Chen, Duen Horng Chau, Cynthia Rudin and Margo Seltzer. **TimberTrek: Exploring and Curating Trustworthy Decision Trees with Interactive Visualization**, IEEE VIS: Visualization & Visual Analytics, 2022.
43. Harsh Parikh, Cynthia Rudin, Alexander Volfovsky. **MALTS: Matching After Learning to Stretch**. Journal of Machine Learning Research, 2022
44. Haiyang Huang, Yingfan Wang, Cynthia Rudin, and Edward Browne. **Towards a Comprehensive Evaluation of Dimension Reduction Methods for Transcriptomic Data Visualization**. Communications Biology (Nature), 2022.
45. Vaishali Jain, Ted Enamorado, and Cynthia Rudin. **The Importance of being Ernest, Ekundayo, or Eswari: An Interpretable Machine Learning Approach to Name-based Ethnicity Classification**, Harvard Data Science Review, 2022.
46. Elita Lobo, Harvineet Singh, Marek Petrik, Cynthia Rudin, Himabindu Lakkaraju. **Data Poisoning Attacks on Off-Policy Policy Evaluation Methods**, (Oral - top 5%), UAI, 2022.
47. Lesia Semenova, Cynthia Rudin, and Ron Parr. **On the Existence of Simpler Machine Learning Models**. ACM Conference on Fairness, Accountability, and Transparency (ACM FAccT), 2022.
48. Cynthia Rudin. **Why Black Box Machine Learning Should be Avoided for High Stakes Decisions, in Brief**, Nature Reviews Methods Primers, 2022.
49. Jiachang Liu, Chudi Zhong, Margo Seltzer, and Cynthia Rudin. **Fast Sparse Classification for Generalized Linear and Additive Models**, AISTATS, 2022.
50. Cynthia Rudin, Chaofan Chen, Zhi Chen, Haiyang Huang, Lesia Semenova, Chudi Zhong. **Interpretable Machine Learning: Fundamental Principles and 10 Grand Challenges**, Statistics Surveys, 2022.
51. Hayden McTavish, Chudi Zhong, Reto Achermann, Ilias Karimalis, Jacques Chen, Cynthia Rudin, Margo Seltzer. **How Smart Guessing Strategies Can Yield Massive Scalability Improvements for Sparse Decision Tree Optimization**, AAAI, 2022.
52. Zhi Chen, Alexander Ogren, Chiara Daraio, L. Catherine Brinson, Cynthia Rudin. *Winner of the 2022 Physical and Engineering Sciences (SPES) and the Quality and Productivity (Q&P) Student Paper Competition of the American Statistical Association*. **How to See Hidden Patterns in Metamaterials with Interpretable Machine Learning**. Extreme Mechanics Letters, 2022.

53. Tong Wang and Cynthia Rudin. **Subgroup Identification for Enhanced Treatment Effect with Decision Rules**, INFORMS Journal on Computing, 2022.
54. Caroline Wang, Bin Han, Bhrij Patel, Feroze Mohideen, Cynthia Rudin. **In Pursuit of Interpretable, Fair and Accurate Machine Learning for Criminal Recidivism Prediction**. Journal of Quantitative Criminology, 2022.
55. Masoud Afnan, Michael Anis Mihdi Afnan, Yanhe Liu, Julian Savulescu, Abhishek Mishra, Vincent Conitzer, Cynthia Rudin. **Data solidarity for machine learning for embryo selection; a call for the creation of an open access repository of embryo data**. Reproductive BioMedicine Online, 2022.
56. Chunxiao Li, Cynthia Rudin, and Tyler McCormick. **Rethinking Nonlinear Instrumental Variable Models through Prediction Validity**, Journal of Machine Learning Research, 2022.
57. Marco Morucci, Md. Noor-E-Alam, and Cynthia Rudin. **A Robust Approach to Quantifying Uncertainty in Matching Problems of Causal Inference**, INFORMS Journal on Data Science, 2022.
58. Edwin Agnew, Lily Zhu, Sam Wiseman, and Cynthia Rudin. **Can a Computer Really Write Poetry?** Harvard Data Science Review, 2022.
59. Ali Behrouz, Mathias Lécuyer, Cynthia Rudin and Margo Seltzer. **Fast Optimization of Weighted Sparse Decision Trees for use in Optimal Treatment Regimes and Optimal Policy Design**, Workshop on Advances in Interpretable Machine Learning and Artificial Intelligence (AIMLAI) at CIKM, 2022.
60. Gaurav Rajesh Parikh, Jenny Huang, Albert Sun, Lesia Semenova, Cynthia Rudin. *Winner of the American Statistical Association Data Challenge Expo (Student Winners), 2022*, **Making the World More Equal, One Ride at a Time: Studying Public Transportation Initiatives Using Interpretable Causal Inference**. NeurIPS 2022 Workshop on Causality for Real-world Impact (CML4Impact), 2022.

2021

61. Cynthia Rudin and Shawn Bushway. **A Truth Serum for your Personal Perspective on Facial Recognition Software in Law Enforcement**, Translational Criminology, Fall 2021
62. Tianyu Wang and Cynthia Rudin. *Best Paper Second Prize for INFORMS Data Mining Best Paper Competition - General Track 2021*, **Bandit Learning for Proportionally Fair Allocations**, 2021.
63. Yingfan Wang, Haiyang Huang, Cynthia Rudin, and Yaron Shaposhnik. **Understanding How Dimension Reduction Tools Work: An Empirical Approach to Deciphering t-SNE, UMAP, TriMAP, and PaCMAP for Data Visualization**, Journal of Machine Learning Research, 2021.
 - Our PaCMAP software is the *Winner of the 2023 John M. Chambers Statistical Software Award from the American Statistical Association*.
64. Tianyu Wang, Marco Morucci, M. Usaid Awan, Yameng Liu, Sudeepa Roy, Cynthia Rudin, Alexander Volfovsky. **FLAME: A Fast Large-scale Almost Matching Exactly Approach to Causal Inference**. Journal of Machine Learning Research, 2021.
65. Neha R. Gupta, Vittorio Orlandi, Chia-Rui Chang, Tianyu Wang, Marco Morucci, Pritam Dey, Thomas J. Howell, Xian Sun, Angikar Ghosal, Sudeepa Roy, Cynthia Rudin, Alexander Volfovsky. **dame-flame: A Python Library Providing Fast Interpretable Matching for Causal Inference**. 2021.
 - Software for FLAME by our students Vittorio Orlandi and Neha Gupta was the *Honorable Mention for the 2022 John M. Chambers Statistical Software Award from the American Statistical Association*.
66. Stefano Tracà, Cynthia Rudin, Weiyu Yan. *Best paper award, INFORMS 2016 Data Mining & Decision Analytics Workshop*, also *Finalist for 2015 IBM Service Science Best Paper Award*. **Regulating Greed over Time in Multi-Armed Bandits**. Journal of Machine Learning Research, 2021.
67. Jianyou Wang, Xiaoxuan Zhang, Yuren Zhou, Christopher Suh, Cynthia Rudin. **There Once Was a Really Bad Poet, It Was Automated But You Didn't Know It**. Transactions of the Association for Computational Linguistics (TACL), 2021. (Presented at ACL, 2021)

68. Divya Koyyalagunta, Anna Sun, Rachel Lea Draelos, Cynthia Rudin. **Playing Codenames with Language Graphs and Word Embeddings**. Journal of Artificial Intelligence Research (JAIR), 2021.
69. Michael Anis Mihdi Afnan, Cynthia Rudin, Vincent Conitzer, Julian Savulescu, Abhishek Mishra, Yanhe Liu and Masoud Afnan. **Ethical Implementation of Artificial Intelligence to Select Embryos in In Vitro Fertilization**. Fourth AAAI/ACM Conference on AI, Ethics, and Society (AIES), 2021. (Our related work “Embryo selection using Artificial Intelligence (AI): Epistemic and ethical considerations” accepted for oral presentation at European Society of Human Reproduction and Embryology, ESHRE, 2021, and our related work “Embryo selection by “black-box” artificial intelligence: The ethical and epistemic considerations” accepted for oral presentation at Fertility Society of Australia and New Zealand Annual Conference, 2021)
70. Michael Anis Mihdi Afnan, Yanhe Liu, Vincent Conitzer, Cynthia Rudin, Abhishek Mishra, Julian Savulescu, Masoud Afnan. **Interpretable, Not Black-Box, Artificial Intelligence Should be Used for Embryo Selection**, Human Reproduction Open, 2021.
71. Alina Jade Barnett, Fides Regina Schwartz, Chaofan Tao, Chaofan Chen, Yinhao Ren, Joseph Y. Lo, Cynthia Rudin. **IAIA-BL: A Case-based Interpretable Deep Learning Model for Classification of Mass Lesions in Digital Mammography**, Nature Machine Intelligence, 2021. Conference papers related to this project:
 - Alina Jade Barnett, Fides Regina Schwartz, Chaofan Tao, Chaofan Chen, Yinhao Ren, Joseph Y. Lo, and Cynthia Rudin. **Interpretable Mammographic Image Classification using Cased-Based Reasoning and Deep Learning**, Oral Presentation, IJCAI-21 Workshop on Deep Learning, Case-Based Reasoning, and AutoML: Present and Future Synergies, 2021.
 - Alina Jade Barnett, Vaibhav Sharma, Neel Gajjar, Jerry Fang, Fides Schwartz M.D., Chaofan Chen, Joseph Y. Lo, Cynthia Rudin. **A user interface to communicate interpretable AI decisions to radiologists**. SPIE Medical Imaging, 2022.
72. Zhicheng Guo, Cheng Ding, Xiao Hu, and Cynthia Rudin. **A Supervised Machine Learning Semantic Segmentation Approach for Detecting Artifacts in Plethysmography Signals from Wearables**. Physiological Measurement, 2021.
73. Alex Oesterling, Angikar Ghosal, Haoyang Yu, Rui Xin, Yasa Baig, Lesia Semenova, and Cynthia Rudin. **Duke Data Science 3C Citation Classification Task B**, (Outlines our strategy for our *third-place competition entry in the 2021 3C Shared Task Citation Context Classification based on Purpose competition*), Oral Presentation, Second Workshop on Scholarly Document Processing (SDP) Workshop, NAACL, 2021

2020

74. Zhi Chen, Yijie Bei and Cynthia Rudin. **Concept Whitening for Interpretable Image Recognition**. Nature Machine Intelligence, 2020.
75. Jiayun Dong and Cynthia Rudin. **Variable Importance Clouds: A Way to Explore Variable Importance for the Set of Good Models**, Nature Machine Intelligence, 2020
76. Beau Coker, Cynthia Rudin and Gary King. **A Theory of Statistical Inference for Ensuring the Robustness of Scientific Results**. Management Science, 2020.
77. Jimmy Lin, Chudi Zhong, Diane Hu, Cynthia Rudin, Margo Seltzer. **Generalized and Scalable Optimal Sparse Decision Trees**. ICML, 2020.
78. Tianyu Wang and Cynthia Rudin. **Bandits for BMO Functions**. ICML, 2020.
79. Marco Morucci, Vittorio Orlandi, Sudeepa Roy, Cynthia Rudin, Alexander Volfovsky. **Adaptive Hyper-box Matching for Interpretable Individualized Treatment Effect Estimation**. UAI, 2020.
80. Cynthia Rudin, Caroline Wang and Beau Coker. **The Age of Secrecy and Unfairness in Recidivism Prediction**. Harvard Data Science Review, 2020.
 - Broader Issues Surrounding Model Transparency in Criminal Justice Risk Scoring. (Rejoinder) Harvard Data Science Review, 2020.

81. Sachit Menon, Alexandru Damian, Nikhil Ravi, Shijia Hu, Cynthia Rudin. **PULSE: Self-Supervised Photo Upsampling via Latent Space Exploration of Generative Models**. CVPR, 2020.
82. Marco Morucci, Vittorio Orlandi, Sudeepa Roy, Alexander Volfovsky. **Almost-Matching-Exactly for Treatment Effect Estimation under Network Interference**. AISTATS, 2020.
83. Tianyu Wang, Weicheng Ye, Dawei Geng and Cynthia Rudin. **Towards Practical Lipschitz Bandits**. Foundations of Data Science (FODS), 2020.
84. Jerry Liu, Nathan O’Hara, Alexander Rubin, Rachel Draelos, Cynthia Rudin. **Metaphor Detection Using Contextual Word Embeddings from Transformers**. Proceedings of the Second Workshop on Figurative Language Processing, pages 250-255, 2020.
85. Hunter Gregory, Steven Li, Pouya Mohammadi, Natalie Tarn, Rachel Draelos, Cynthia Rudin. **A Transformer Approach to Contextual Sarcasm Detection in Twitter**. Proceedings of the Second Workshop on Figurative Language Processing, pages 270-275, 2020.
86. Qinwen Huang, Ye Zhou, Xiaochen Du, Reed Chen, Jianyou Wang, Cynthia Rudin, Alberto Bartesaghi. **Cryo-ZSSR: Multiple-Image Super-Resolution Based on Deep Internal Learning**. Machine Learning for Structural Biology Workshop at the 34th Conference on Neural Information Processing Systems, 2020.

2019

87. Cynthia Rudin and Joanna Radin. **Why Are We Using Black Box Models in AI When We Don’t Need To? A Lesson From an Explainable AI Competition**. Harvard Data Science Review, 2019.
88. Aaron Fisher, Cynthia Rudin, Francesca Dominici. **All Models are Wrong, but Many are Useful: Learning a Variable’s Importance by Studying an Entire Class of Prediction Models Simultaneously**. Journal of Machine Learning Research, 2019.
89. Harsh Parikh, Cynthia Rudin, and Alexander Volfovsky. **An Application of Matching After Learning To Stretch (MALTS) to the ACIC 2018 Causal Inference Challenge Data**. Observational Studies, Issue 5, pages 118-130, 2019.
90. Chaofan Chen, Oscar Li, Chaofan Tao, Alina Jade Barnett, Jonathan Su, Cynthia Rudin. **This Looks Like That: Deep Learning for Interpretable Image Recognition**. NeurIPS (Spotlight), 2019.
91. Xiyang Hu, Cynthia Rudin, and Margo Seltzer. **Optimal Sparse Decision Trees**. NeurIPS (Spotlight), 2019.
92. Peter Hase, Chaofan Chen, Oscar Li, Cynthia Rudin. **Interpretable Image Recognition with Hierarchical Prototypes**. AAAI Human Computation (AAAI-HCOMP), 2019.
93. Cynthia Rudin. **Stop Explaining Black Box Machine Learning Models for High Stakes Decisions and Use Interpretable Models Instead**. Nature Machine Intelligence, 2019.
 - Shorter version called **Please Stop Explaining Black Box Machine Learning Models for High Stakes Decisions** appeared at NIPS 2018 Workshop on Critiquing and Correcting Trends in Machine Learning, 2018.
94. Stefano Tracà, Weiyu Yan, and Cynthia Rudin. **Reducing Exploration of Dying Arms in Mortal Bandits**. UAI, 2019
95. Usaid Awan, Yameng Liu, Marco Morucci, Sudeepa Roy, Cynthia Rudin, and Alexander Volfovsky. **Interpretable Almost-Exact Matching With Instrumental Variables**. UAI, 2019
96. Awa Dieng, Yameng Liu, Sudeepa Roy, Cynthia Rudin, and Alexander Volfovsky. **Interpretable Almost-Exact Matching For Causal Inference**. AISTATS, 2019.

2018

97. Fulton Wang, Tyler McCormick, Cynthia Rudin, and John Gore. *Best Poster Award, Conference of the ASA Section on Statistical Learning and Data Mining, 2014*. **Modeling Recovery Curves With Application to Prostatectomy**. Biostatistics, 2018.

98. Cynthia Rudin and Yining Wang. *Finalist for 2017 QSR (Quality, Reliability and Statistics) best refereed paper competition, INFORMS 2017. On Direct Learning to Rank and Rerank*. Artificial Intelligence and Statistics (AISTATS), 2018.
99. John Benhardt, Tianlin Duan, Peter Hase, Liuyi Zhu, Cynthia Rudin. *Winner of the 2018 PoetiX Literary Turing Test Award for computer-generated poetry. Shall I Compare Thee to a Machine-Written Sonnet? An Approach to Algorithmic Sonnet Generation*, 2018.
100. Yijie Bei, Alex Damian, Shijia Hu, Sachit Menon, Nikhil Ravi, and Cynthia Rudin. *NTIRE-CVPR 2018 Image Super-Resolution Challenge: winner for Track 1 (classic bicubic), honorable mention for Track 2 (realistic mild adverse conditions). New Techniques for Preserving Global Structure and Denoising with Low Information Loss in Single-Image Super-Resolution*, New Trends in Image Restoration and Enhancement Workshop and Challenges on Super-Resolution, Dehazing, and Spectral Reconstruction, NTIRE-CVPR, 2018.
101. Cynthia Rudin and Şeyda Ertekin. **Learning Customized and Optimized Lists of Rules with Mathematical Programming**. Mathematical Programming C (Computation), Mathematical Programming Computation, Volume 10, Number 4, pages 659-702, 2018.
102. Elaine Angelino, Nicholas Larus-Stone, Daniel Alabi, Margo Seltzer, and Cynthia Rudin. **Learning Certifiably Optimal Rule Lists for Categorical Data**, Journal of Machine Learning Research, volume 18, no. 234, pages 1-78, 2018.
103. Nicholas Larus-Stone, Elaine Angelino, Daniel Alabi, Margo Seltzer, Vassilios Kaxiras, Aditya Saligrama and Cynthia Rudin. **Systems Optimizations for Learning Certifiably Optimal Rule Lists**. Conference on Machine Learning and Systems (MLSys), 2018.
104. Oscar Li, Hao Liu, Chaofan Chen, and Cynthia Rudin. **Deep Learning for Case-based Reasoning through Prototypes: A Neural Network that Explains its Predictions**. Association for the Advancement of Artificial Intelligence (AAAI), 2018.
105. Chaofan Chen and Cynthia Rudin. **An Optimization Approach to Learning Falling Rule Lists**. Artificial Intelligence and Statistics (AISTATS), 2018.

2017

106. Elaine Angelino, Nicholas Larus-Stone, Daniel Alabi, Margo Seltzer, and Cynthia Rudin. **Certifiably Optimal Rule Lists for Categorical Data**, Knowledge Discovery in Databases (KDD - oral presentation), 2017.
107. Berk Ustun, Lenard A. Adler, Cynthia Rudin, Stephen V. Faraone, Thomas J. Spencer, Patricia Berglund, Michael J. Gruber, Ronald C. Kessler. **The World Health Organization Adult Attention-Deficit/Hyperactivity Disorder Self-Report Screening Scale for DSM-5**. JAMA Psychiatry, April 2017.
108. Tong Wang, Cynthia Rudin, Finale Doshi, Yimin Liu, Erica Klampfl, and Perry MacNeille. **Bayesian Rule Sets for Interpretable Classification, with Application to Context-Aware Recommender Systems**. Journal of Machine Learning Research (JMLR), volume 18, number 70, pages 1-37, 2017.
109. Fulton Wang and Cynthia Rudin. **Causal Falling Rule Lists**, Fairness, Accountability, and Transparency (FATML), 2017 (longer version on ArXiv).
110. Hongyu Yang, Cynthia Rudin, and Margo Seltzer. *Winner of Student Paper Competition sponsored by the Statistical Learning and Data Mining section (SLDM) of the American Statistical Association, 2016. Scalable Bayesian Rule Lists*. International Conference on Machine Learning (ICML), 2017.
111. Himabindu Lakkaraju and Cynthia Rudin. *Finalist for 2017 INFORMS Data Mining Best Paper Competition. Learning Cost-Effective and Interpretable Treatment Regimes*. Artificial Intelligence and Statistics (AISTATS), 2017.

- Shorter versions accepted to workshops: NIPS Machine Learning for Healthcare (ML4HC), NIPS Workshop on Interpretable Machine Learning in Complex Systems, and NIPS Workshop on ML and the Law, NIPS 2016.

2016

112. Jiaming Zeng, Berk Ustun, and Cynthia Rudin. *Winner of 2015 Undergraduate Statistics Research Project Competition (USRESP) sponsored by the American Statistical Association (ASA) and the Consortium for Advancement of Undergraduate Statistics Education (CAUSE). Interpretable Classification Models for Recidivism Prediction.* Journal of the Royal Statistical Society Series A, volume 180, issue 3, 2016.
113. Tong Wang, Cynthia Rudin, Finale Doshi, Yimin Liu, Erica Klampfl, and Perry MacNeille. **Bayesian Rule Sets for Interpretable Classification.** IEEE International Conference on Data Mining (ICDM), 2016.
114. Ramin Moghaddass, Cynthia Rudin, and David Madigan. **The Factorized Self-Controlled Case Series Method: An Approach for Estimating the Effects of Many Drugs on Many Outcomes.** Journal of Machine Learning Research, 17(185):1-24, 2016.
115. Berk Ustun, M. Brandon Westover, Cynthia Rudin, and Matt T. Bianchi. **Clinical Prediction Models for Sleep Apnea: The Importance of Medical History over Symptoms.** Journal of Clinical Sleep Medicine, volume 12, number 2, 2016.
116. Vikas Garg, Cynthia Rudin, and Tommi Jaakola. **CRAFT: ClusteR-specific Assorted Feature selecTion,** Artificial Intelligence and Statistics (AISTATS), 2016.
117. Benjamin Letham, Portia A. Letham, Cynthia Rudin, Edward P. Browne. **Prediction Uncertainty and Optimal Experimental Design for Learning Dynamical Systems.** Chaos, volume 26, number 6, 2016.
118. Benjamin Letham, Lydia M. Letham and Cynthia Rudin. **Bayesian Inference of Arrival Rate and Substitution Behavior from Sales Transaction Data with Stockouts.** Knowledge Discovery in Databases (KDD), 2016
119. Edward P. Browne, Benjamin Letham, and Cynthia Rudin. **A Computational Model of Inhibition of HIV-1 by Interferon-Alpha.** PLoS ONE, vol 11, no 3, pages 1–16, March, 2016.
120. Şeyda Ertekin and Cynthia Rudin. **A Bayesian Approach to Learning Scoring Systems.** Big Data, volume 3, number 4, 2016.
121. John Guckenheimer, Thomas Overbye (Co-chairs), and committee: Daniel Bienstock, Anjan Bose, Terry Boston, Jeffery Dagle, Marija D. Ilic, Christopher K. Jones, Frank P. Kelly, Yannis G. Kevrekidis, Ralph D. Masiello, Juan C. Meza, Cynthia Rudin, Robert J. Thomas, and Margaret H. Wright. **Analytic Research Foundations for the Next-Generation Electric Grid.** The National Academies Press, 2016. (Consensus study report of the National Academies of Sciences Engineering, and Medicine)

2015

122. Fulton Wang and Cynthia Rudin. *Winner of Best Student Paper Competition, Statistical Learning and Data Mining section (SLDM) of the American Statistical Association, 2015, also Finalist for INFORMS Data Mining Section Best Student Paper Award, 2015. Falling Rule Lists.* Proceedings of the 18th International Conference on Artificial Intelligence and Statistics (AISTATS), 2015.
123. Benjamin Letham, Cynthia Rudin, Tyler H. McCormick, and David Madigan. *Winner of Data Mining Best Student Paper Competition, INFORMS 2013, also Winner of Student Paper Competition sponsored by the Statistical Learning and Data Mining section (SLDM) of the American Statistical Association, 2014. Building Interpretable Classifiers with Rules using Bayesian Analysis: Building a Better Stroke Prediction Model.* Annals of Applied Statistics, volume 9, number 3, pages 1350-1371, 2015.
Shorter version:
 - Benjamin Letham, Cynthia Rudin, Tyler McCormick, and David Madigan. **An Interpretable Model for Stroke Prediction Using Rules and Bayesian Analysis.** Proceedings of 2014 KDD Workshop on Data Science for Social Good, 2014
124. Tong Wang, Cynthia Rudin, Daniel Wagner and Rich Sevieri. *Second place in INFORMS 2015 Doing Good with Good OR Paper Competition. Finding Patterns with a Rotten Core: Data Mining for Crime Series with Core Sets.* Big Data. volume 3, issue 1, 2015. (Special issue on Data for Social Good)
125. Ramin Moghaddass and Cynthia Rudin. **The Latent State Hazard Model, with Application to Wind Turbine Reliability.** Annals of Applied Statistics, volume 9, number 4, pages 1823–1863, 2015.

126. Şeyda Ertekin, Cynthia Rudin, and Tyler McCormick. **Reactive Point Processes: A New Approach to Predicting Power Failures in Underground Electrical Systems**. Annals of Applied Statistics, volume 9, No 1, pages 122–144, 2015.

2014

127. Theja Tulabandhula and Cynthia Rudin. **Tire Changes, Fresh Air and Yellow Flags: Challenges in Predictive Analytics for Professional Racing**. Big Data, vol 2 issue 2, pages 97-112, June 20, 2014.
128. Şeyda Ertekin, Cynthia Rudin and Haym Hirsh. **Approximating the Crowd**. Data Mining and Knowledge Discovery, volume 28 issue 5-6, pages 1189-1221, September 2014.
- Shorter versions appeared at NIPS Workshop on Computational Social Science and the Wisdom of Crowds, 2011, Proceedings of Collective Intelligence (CI) 2012, and Proceedings of the 2012 AAAI Fall Symposium on Machine Aggregation of Human Judgment, MAGG-2012.
129. Been Kim, Cynthia Rudin, and Julie Shah. **The Bayesian Case Model: A Generative Approach for Case-Based Reasoning and Prototype Classification**. Neural Information Processing Systems (NIPS), 2014.
130. Theja Tulabandhula, Cynthia Rudin. **On Combining Machine Learning with Decision Making**. Machine Learning (ECML-PKDD journal track), volume 93, Pages 33-64, 2014
131. Siong Thye Goh and Cynthia Rudin. **Box Drawings for Learning with Imbalanced Data**. Proceedings of 20th ACM SIGKDD Conference on Knowledge Discovery and Data Mining (KDD), 2014.
132. Theja Tulabandhula and Cynthia Rudin. **Generalization Bounds for Learning with Linear, Polygonal, Quadratic and Conic Side Knowledge**. Machine Learning (ECML-PKDD journal track), December, 2014, pages 1-34.
- Shorter version: Theja Tulabandhula, Cynthia Rudin. **Generalization Bounds for Learning with Linear and Quadratic Side Knowledge**. Proceedings of ISAIM 2014.
133. Jonathan Huggins and Cynthia Rudin. **Towards a Theory of Pattern Discovery** Proceedings of SIAM Conference on Data Mining (SDM) 2014.
134. Been Kim and Cynthia Rudin. **Learning About Meetings**, Data Mining and Knowledge Discovery, (ECML-PKDD Journal track), volume 28 issue 5-6, pages 1134-1157, September 2014.

2013

135. Benjamin Letham, Cynthia Rudin and Katherine Heller. **Growing a List**. Data Mining and Knowledge Discovery (DAMI), ECML-PKDD journal track. volume 27, pages 372-395, 2013.
136. Tong Wang, Cynthia Rudin, Daniel Wagner, Richard Sevieri. **Learning to Detect Patterns of Crime**, Proceedings of European Conference on Machine Learning and Principles and Practice of Knowledge Discovery in Databases (ECML-PKDD), 2013.
- Ideas from this work were implemented by the NYPD by Alex Chohlas-Wood and E.S. Levine in their algorithm Patternizr, which operates live in New York City.
137. Cynthia Rudin, Benjamin Letham, and David Madigan. **Learning Theory Analysis for Association Rules and Sequential Event Prediction**. Journal of Machine Learning Research (JMLR), volume 14, pages 3385-3436, 2013.
- Shorter version: Cynthia Rudin, Ben Letham, Ansaf Salieb-Aouissi, Eugene Kogan, and David Madigan. **Sequential Event Prediction with Association Rules**, Proceedings of the 24th Annual Conference on Learning Theory (COLT), 2011.
138. Benjamin Letham, Cynthia Rudin and David Madigan. **Sequential Event Prediction**. Machine Learning, volume 93, pages 357-380, 2013
139. Theja Tulabandhula and Cynthia Rudin. *Finalist, Data Mining Best Student Paper Competition, INFORMS 2012*. **Machine Learning with Operational Costs**. Journal of Machine Learning Research (JMLR), volume 14, pages 1989-2028, July 2013. Preliminary work is in the following conference paper.

- Theja Tulabandhula and Cynthia Rudin. **The Influence of Operational Costs on Estimation**, Proceedings of the International Symposium on Artificial Intelligence and Mathematics (ISAIM), 2012.

2012

140. Tyler McCormick, Cynthia Rudin, and David Madigan. **Hierarchical Models for Association Rule Mining: A New Approach for Adverse Event Prediction in Clinical Trials**, Annals of Applied Statistics, volume 6, No. 2, pages 652–668, 2012.
141. Cynthia Rudin, David Waltz, Roger N. Anderson, Albert Boulanger, Ansaf Salieb-Aouissi, Maggie Chow, Haimonti Dutta, Philip Gross, Bert Huang, Steve Ierome, Delfine Isaac, Arthur Kressner, Rebecca J. Passonneau, Axinia Radeva, Leon Wu. **Machine Learning for the New York City Power Grid**, IEEE Transactions on Pattern Analysis and Machine Intelligence, Vol 34, No 2, February 2012. *Spotlight Paper for the February 2012 Issue*.
142. Allison Chang, Cynthia Rudin, Mike Cavaretta, Robert Thomas and Gloria Chou. **Reverse-Engineering Quality Ratings**, Machine Learning: volume 88, issue 3, pages 369-398, 2012.

2011

143. Cynthia Rudin, Rebecca J. Passonneau, Axinia Radeva, Steve Ierome, and Delfina Isaac. **21st-Century Data Miners Meet 19th-Century Electrical Cables**, IEEE Computer, volume 44 no. 6, pages 103-105, June 2011.
(One of three articles featured on the cover of the magazine.)
144. Şeyda Ertekin and Cynthia Rudin. **On Equivalence Relationships Between Classification and Ranking Algorithms**, Journal of Machine Learning Research, volume 12, pages 2905–2929, 2011.

2010

145. Cynthia Rudin, Rebecca J. Passonneau, Axinia Radeva, Haimonti Dutta, Steve Ierome, and Delfina Isaac. **A Process for Predicting Manhole Events in Manhattan**. Machine Learning, volume 80, pages 1–31, 2010.

- Also oral presentation at ICML 2012

The following conference papers are also related to my projects on grid reliability.

- Rebecca J. Passonneau, Cynthia Rudin, Axinia Radeva, Ashish Tomar, Boyi Xie. **Treatment Effect of Repairs to an Electrical Grid: Leveraging a Machine Learned Model of Structure Vulnerability**, Proceedings of the KDD Applications in Sustainability (SustKDD) Workshop on Data Mining, 17th Annual ACM SIGKDD Conference on Knowledge Discovery and Data Mining, 2011.
- Dingquan Wang, Rebecca Passonneau, Michael Collins and Cynthia Rudin. **Modeling Weather Impact on a Secondary Electrical Grid**, 4th International Conference on Sustainable Energy Information Technology (SEIT-2014), 2014.
- Leon Wu, Timothy Teräväinen, Gail Kaiser, Roger Anderson, Albert Boulanger, and Cynthia Rudin. **Estimation of System Reliability Using a Semiparametric Model**, Proceedings of IEEE EnergyTech, 2011.
- Leon Wu, Gail Kaiser, Cynthia Rudin, and Roger Anderson. **Data Quality Assurance and Performance Measurement of Data Mining for Preventive Maintenance of Power Grid**, Proceedings of the KDD Workshop on Data Mining for Service and Maintenance (KDD4Service), 17th Annual ACM SIGKDD Conference on Knowledge Discovery and Data Mining, 2011.
- Leon Wu, Gail Kaiser, Cynthia Rudin, David Waltz, Roger Anderson, Albert Boulanger, Ansaf Salieb-Aouissi, Haimonti Dutta, and Manoj Poolery. **Evaluating Machine Learning for Improving Power Grid Reliability**, Proceedings of the ICML 2011 workshop on Machine Learning for Global Challenges, International Conference on Machine Learning, 2011.
- Axinia Radeva, Cynthia Rudin, Rebecca Passonneau and Delfina Isaac. **Report Cards for Manholes**, Proceedings of the International Conference on Machine Learning and Applications (ICMLA), 2009. *Best Poster Award*.

- Rebecca Passonneau, Cynthia Rudin, Axinia Radeva and Zhi An Liu. **Reducing Noise in Labels and Features for a Real World Dataset: Application of NLP Corpus Annotation Methods**, Proceedings of the 10th International Conference on Computational Linguistics and Intelligent Text Processing (CICLing), 2009.
- Haimonti Dutta, Cynthia Rudin, Becky Passonneau, Fred Seibel, Nandini Bhardwaj, Axinia Radeva, Zhi An Liu, Steve Ierome, Delfina Isaac. **Visualization of Manhole and Precursor-Type Events for the Manhattan Electrical Distribution System**, Workshop on GeoVisualization of Dynamics, Movement and Change, 11th AGILE International Conference on Geographic Information Science, 2008.
- Boyi Xie, Rebecca J. Passonneau, Haimonti Dutta, Jing-Yeu Miaw, Axinia Radeva, Ashish Tomar, Cynthia Rudin. **Progressive Clustering with Learned Seeds: An Event Categorization System for Power Grid**. 24th International Conference on Software Engineering and Knowledge Engineering (SEKE 2012). Redwood City, CA. July 1-3, 2012.

2009

146. Cynthia Rudin. **The P-Norm Push: A Simple Convex Ranking Algorithm that Concentrates at the Top of the List**, Journal of Machine Learning Research, volume 10, pages 2233–2271, 2009.

- Shorter version: Cynthia Rudin. **Ranking with a P-Norm Push**. Proceedings of the Nineteenth Annual Conference on Learning Theory (COLT), pages 589 - 604, 2006.

An application of the P-Norm Push is described in this conference paper:

- Heng Ji, Cynthia Rudin, and Ralph Grishman. **Re-ranking Algorithms for Name Tagging**. In Proc. Human Language Technology conference - North American chapter of the Association for Computational Linguistics annual meeting (HLT-NAACL) Workshop on Computationally Hard Problems and Joint Inference in Speech and Language Processing, 2006.

147. Cynthia Rudin and Robert E. Schapire. **Margin-Based Ranking and an Equivalence Between AdaBoost and Rank-Boost**. Journal of Machine Learning Research, volume 10, pages 2193–2232, 2009.

- Preliminary version: Cynthia Rudin, Corinna Cortes, Mehryar Mohri, and Robert E. Schapire. **Margin Based Ranking Meets Boosting in the Middle**. Proceedings of the Eighteenth Annual Conference on Learning Theory (COLT), pages 63 - 78, 2005.

2008 and before

148. Cynthia Rudin, Robert E. Schapire and Ingrid Daubechies. **Analysis of Boosting Algorithms Using the Smooth Margin Function**. Annals of Statistics, volume 35, number 6, pages 2723-2768, 2007.

Preliminary material:

- Cynthia Rudin, Robert E. Schapire, and Ingrid Daubechies. (2007) **Precise Statements of Convergence for AdaBoost and arc-gv**. In Proc. AMS-IMS-SIAM Joint Summer Research Conference: Machine Learning, Statistics, and Discovery, pages 131-145, 2007.
- Cynthia Rudin, Robert E. Schapire, and Ingrid Daubechies. **Boosting Based on a Smooth Margin**. Proceedings of the Seventeenth Annual Conference on Computational Learning Theory, (COLT), pages 502-517, 2004.
- Cynthia Rudin, Ingrid Daubechies, and Robert E. Schapire. **On the Dynamics of Boosting**. Advances in Neural Information Processing Systems (NIPS) 16, 2003.

149. Cynthia Rudin, Ingrid Daubechies, and Robert E. Schapire. **The Dynamics of AdaBoost: Cyclic Behavior and Convergence of Margins**. Journal of Machine Learning Research, 5 (Dec): 1557–1595, 2004.

Preliminary material for this work appears partly within the NIPS paper below, and the open problem in COLT is from the JMLR paper:

- Cynthia Rudin, Ingrid Daubechies, and Robert E. Schapire. **On the Dynamics of Boosting**. Advances in Neural Information Processing Systems (NIPS) 16, 2003.
 - Cynthia Rudin, Robert E. Schapire and Ingrid Daubechies. **Does AdaBoost Always Cycle?** JMLR: Workshop and Conference Proceedings, Published as a COLT Open problem, 2012.
150. Ryan Roth, Owen Rambow, Nizar Habash, Mona Diab, and Cynthia Rudin. **Arabic Morphological Tagging, Diacritization, and Lemmatization Using Lexeme Models and Feature Ranking**, The 46th Annual Meeting of the Association for Computational Linguistics: Human Language Technologies (ACL/HLT), 2008.
 151. Cynthia Rudin and Brian Spencer. **Equilibrium Ridge Arrays in Strained Solid Films**. Journal of Applied Physics, vol 86, pp 5530-5536, 1999.

Non-Peer-Reviewed Publications

152. Ezra Miller, Cynthia Rudin and Ingrid Daubechies. **Response to Letter to AMS Notices: Boycott collaboration with police**, September issue, AMS Notices, 2020. *We urge the mathematical community not to boycott interactions with police, but instead to work together to improve society.*
153. Matthew J Salganik, Lauren Maffeo and Cynthia Rudin. **Prediction, Machine Learning, and Individual Lives: an Interview with Matthew Salganik**. Harvard Data Science Review, 2020
154. Sarah Desmarais, Brandon Garrett, and Cynthia Rudin. **Risk Assessment Tools Are Not A Failed ‘Minority Report’**. Perspectives, Law360, July 19, 2019, 5:50 PM EDT
155. Cynthia Rudin and David Carlson. **The Secrets of Machine Learning: Ten Things You Wish You Had Known Earlier to be More Effective at Data Analysis**. INFORMS TutORial, 2019.
156. Cynthia Rudin (with credit to Robin Smith). **Algorithms and Justice: Scrapping the Black Box**. The Crime Report, 2018.
157. Cynthia Rudin, David Dunson, Rafael Irizarry, Hongkai Ji, Eric Laber, Jeffrey Leek, Tyler McCormick, Sherri Rose, Chad Schafer, Mark van der Laan, Larry Wasserman, Lingzhou Xue. **Discovery with Data: Leveraging Statistics with Computer Science to Transform Science and Society**. American Statistical Association whitepaper, <http://www.amstat.org/policy/pdfs/BigDataStatisticsJune2014.pdf>, 2014.
158. Cynthia Rudin and Kiri Wagstaff. **Machine Learning for Science and Society**, Machine Learning, (Introduction to the Special Issue on Machine Learning for Science and Society), volume 95, issue 1, April 2014, pp 1-9.
159. Cynthia Rudin. **Teaching “Prediction: Machine Learning and Statistics”**, Proceedings of the ICML Workshop on Teaching ML, 2012.
160. Peter Qian, Yilu Zhou, and Cynthia Rudin, **Proceedings of the 6th INFORMS Workshop on Data Mining and Health Informatics (DM-HI)**, editors, 2011.
161. Cynthia Rudin and Miroslav Dudík, **Lecture Notes for the AMS Short Course on Statistical Learning**, editors, includes contributions by Robert E Schapire, Lawrence Saul, Lisa Hellerstein, Adam Tauman-Kalai, and John Lafferty, 2007.

Book

Intuition for the Algorithms of Machine Learning, Multimedia textbook, including videos, in progress, 2022. Free and publicly available. Introduces interpretable ML as part of introductory graduate ML.
<https://users.cs.duke.edu/~cynthia/teaching.html>

Grants

Comparative effectiveness of EEG guided anti-seizure treatment in acute brain injury, NIH, PI Sahar Zafar, MGH, 12/1/2023-11/30/2028.

DMREF/Collaborative Research: Accelerated Discovery of Sustainable Bioplastics: Automated, Tunable, Integrated Design, Processing and Modeling, NSF, co-PI. 10/01/2023-09/30/2027. Amount of Award: \$790,000 (with Cate Brinson, Eleftheria Roumeli, Linda Schadler, and Kayla Sprenger).

Novel Algorithm and Data Strategies to Detect and Predict Atrial Fibrillation for Poststroke Patients (NADSP), NIH (DHHS). co-PI, with Pi Xiao Hu (Emory). 3/10/2023-2/29/2024. Amount of Award: \$551,830

Exploring the Whole Set of Sparse Explanations, DOE. PI. 09/01/2022-08/31/2024. Amount of Award: \$400,000

FW-HTF-R: Interpretable Machine Learning for Human-Machine Collaboration in High Stakes Decisions in Mammography, NSF. PI, with co-PI Joseph Lo. 09/01/2022-08/31/2026. Amount of Award: \$1,800,000

Patient-Focused Collaborative Hospital Repository Uniting Standards (CHoRUS) for Equitable AI, National Institutes of Health, co-PI, with PI Eric Rosenthal and several other co-PIs. 09/01/2022 - 08/31/2026. Amount of Award: \$5,880,300

FAI: An Interpretable AI Framework for Care of Critically Ill Patients Involving Matching and Decision Trees, NSF & Amazon. PI, with co-PIs Alex Volfovsky and Sudeepa Roy. 07/01/2022-06/30/2025. Amount of Award: \$1,000,000

Center for Virtual Imaging Trials. NIH/NIBIB. Co-lead for Project 3, with PIs Ehsan Samei and Joseph Lo. 4/01/21-12/31/25. Technology Research and Development (TR&D) Project 3. Amount of Award: \$6,371,435 (total) \$2,620,393 (Project 3).

MedX: Interpretable Deep Learning Models for Better Clinician-AI Communication in Clinical Mammography, 11/1/2021-10/31/2022. Internal Duke Award, with Joseph Lo, Amount of Award: \$50,000

A Machine Learning Framework for Understanding Impacts on the HIV Latent Reservoir Size, Including Drugs of Abuse, National Institute on Drug Abuse (NIDA), PI, with co-PI's David Murdoch, Nilu Goonetilleke and Nancie Archin. 9/30/2021-9/30/2026. Amount of Award: \$2,284,375

Research on Explainable AI on Multiple Models, Fujitsu, PI, 07/16/2021-7/16/2024. Amount of Award: \$272,307

EAGER: Creating an Unsupervised Interpretable Representation of the World Through Concept Disentanglement, NSF, PI, 11/01/2021-10/31/2024. Amount of Award: \$169,345

NSF Workshop on Seamless/Seamful Human-Technology Interaction, NSF, PI, 09/01/2021-08/31/2022. Amount of Award: \$49,777

FAIR Data and Interpretable AI Framework for Architected Metamaterials, DOE, co-PI, with PI Cate Brinson, and co-PI Chiara Daraio, 9/1/2020-8/31/2023. Amount of Award: \$870,580

NRT-HDR Harnessing AI for Understanding & Designing Materials (aiM), DGE-2022040, co-PI, with PI Cate Brinson, co-PIs David Banks, Stafano Curtarolo, Johann Guilleminot, 09/01/2020-08/31/2025. Amount of Award: \$2,252,971

NSF-HDR TRIPODS: Innovations in Data Science, CCF-1934964 co-PI, with PI Sayan Mukherjee, co-PI's Robert Calderbank, Rong Ge, and Jianfeng Liu, 10/1/2019-9/30/23. Amount of Award: \$1.5 million

Utilizing Key Past Experiences from Large Datasets to Make Better Prediction in Multi-Class Settings, co-PI with Ramin Moghaddass, Amazon AWS Machine Learning Research Awards program, 2018. Amount of Award: \$20,000

An Integrated Nonparametric Bayesian and Deep Neural Network Framework for Biologically-Inspired Lifelong Learning (co-PI, with PI Katherine Heller, and other co-PI's Lawrence Carin, David Dunson, Nicolas Brunel, Tamara Broderick, Joshua Tenenbaum, Michael Jordan, Thomas Griffiths), DARPA, 2017-2020. Amount of Award: \$4,388,119

Collaborative Research Framework: Data: HDR: Nanocomposites to Metamaterials: A Knowledge Graph Framework, OAC-1835782 (co-PI, with PI Cate Brinson, and other co-PI's Chiara Daraio, Linda Schadler, Deborah McGuinness, and Wei Chen), NSF, 11/1/2018-10/31/2023. Amount of Award: \$2,590,810

Lord Foundation “Duke’s Super Superresolution Team (Continued)” (Single PI) 2019-2020. Amount of Award: \$24,000

Lord Foundation “Duke’s Super Superresolution Team” (Single PI) 2018-2019. Amount of Award: \$32,000

Duke Energy Initiative “Enabling Better Energy Decisions Through Better Interpretable Causal Inference Methods for Personalized Treatment Effects” Duke University Energy Initiative Energy Research Seed Fund (ERSF). (PI, with co-PI’s Alex Volfovsky and Sudeepa Roy), 2018-2021. Amount of Award: \$40,500 Stage I, \$40,500 Stage II

QuBBD: Collaborative Research: Matching Methods for Causal Inference: Big Data and Networks, 1R01EB025021-01, sponsored by DHHS, PHS, NIH, NIBIB&B (co-Investigator, with PI Alexander Volfovsky, co-Investigators Sudeepa Roy and Allison Aiello). 2017-2020. Amount of Award: \$849K (\$514K to Duke)

Alfred P. Sloan Foundation “Interdisciplinary Energy Data Analytics Ph.D. Fellows Program,” (co-PI, with Brian Murray, William Pizer, and Kyle Bradbury). 2018-2020. Amount of Award: \$225,000

Laura and John Arnold Foundation “Interpretable Machine Learning for Pre-Trial Risk Analysis,” 2017-2020. (Single PI) Amount of Award: \$87,507

Duke Institute for Health Innovation “Palliatyctics: Using analytics to inform a palliative care population health management intervention” (co-PI with Jonathan Fischer, Leslie Alabi, Eugenie Komives from Duke Medical). 2017-2018. Amount of award: \$55,500

2016 Adobe Digital Marketing Research Award, “Compute-intensive causal machine learning models for finding customer segments” (Single PI). 2016. Amount of Award: \$50K

MIT-Lincoln Labs “Adaptable, Interpretable, Machine Learning” (co-PI with Jonathan Su from MIT-LL), 2016-2019. Amount of Subaward: \$542K

Xerox Research “Causal Inference-related Research Threads in the Prediction Analysis Lab, continued” (co-PI with Theja Tulabandhula), 2016. Amount of Award: \$30K

DARPA “Foundations of Sequential Learning” (co-PI with Ron Parr and Kamesh Munagala), April 2016 - Jan 2017. Amount of Award: \$242K

Philips “Algorithms for Interpretable Risk-Scoring” (Single PI), July 2017-December 2017. Amount of Award: \$122K

Xerox Research “Causal Inference-related Research Threads in the Prediction Analysis Lab” (co-PI with Theja Tulabandhula), 2015. Amount of Award: \$30K

Philips “Self-Learning Systems and User-Behavior Modeling” (Single PI), June 2015-May 2016. Amount of Award: \$240K

Army Research Office “Uncertainty Quantification for Unobserved Variables in Dynamical Systems and Optimal Experimental Design” (Single PI), Spring 2015. Amount of Award: \$50K

Big Data Seed Grant, MIT Big Data Initiative “Interpretable, Scalable and Causal Models from Machine Learning” (Single PI), Spring 2015. Amount of award: 1 semester RA

Accenture and MIT Alliance in Business Analytics “Creating a Model of the Usual State of a Machine” (Single PI), March 20 2015 - March 20 2016. Amount of award: \$100,000

Wistron Corporation “Interpretable Predictive Models from Machine Learning” (Single PI), September 4, 2014 - August 31, 2016. Amount of award \$ 300,000

Accenture and MIT Alliance in Business Analytics “Big Data Analysis for Plant and Commercial Optimization,” June 15 2013 - June 14 2014. Amount of award: \$ 100,000

Ford Racing “Predictive Analytics for Racing” (Single PI),
September 1 2012 - August 31 2013. Amount of award: \$75,000
September 1 2013 - December 31 2014. Amount of award: \$75,000
December 31 2014 - December 31 2015. Amount of award: \$157,000

Siemens Energy, CKI University Innovation Initiative. “CKI Proposal: Augmented Data-Driven Diagnosis using Physical Models” (Single MIT PI with co-PIs at Siemens), June 1, 2013 - May 31, 2016
Amount of award: \$350,000

Siemens Energy, CKI University Innovation Initiative. “CKI Proposal: Incorporating Prediction Analysis into XHQ” (Single MIT PI with co-PIs at Siemens), June 1 2013 - May 31 2015.
Amount of award: 150,000 euro \approx \$200,000

Ford - MIT Alliance. “Develop advanced in-vehicle SYNC advertisement features to target drivers based on their context information, system interaction and past choices” (Single PI), July 1 2012 - August 31 2014.
Amount of award: \$298,000

MIT Sloan Research Fund. “Predictive Models for Highly Imbalanced data,” (Single PI), Amount of award: \$20,000

Solomon Buchsbaum Research Fund. “A New Foundation for Statistical Decision-Making,” June 2 2011-present. (Single PI), Amount of award: \$50,000

NSF-CAREER IIS-1053407. “New Approaches for Ranking in Machine Learning,” September 1 2011 – August 31 2018.
Amount of award: \$480,000

MIT Lincoln Labs. “A Knowledge Discovery Framework for Threat Identification (Phase III)” (co-PI with Shirish Ranjit from MIT-LL) 7/1/2013-6/30/2014.
Amount of award: \$100,000

MIT Lincoln Labs. “A Knowledge Discovery Framework for Threat Identification (Phase II)” (co-PI with Shirish Ranjit from MIT-LL) 7/1/2012-6/30/2013.
Amount of award: \$100,000

MIT Lincoln Labs. “A Knowledge Discovery Framework for Threat Identification” (co-PI with Shirish Ranjit from MIT-LL and Regina Barzilay from MIT), 6/1/2011- 5/31/2012.
Amount of award: \$83,000

MIT Intelligence Initiative. “Combining human and machine predictions using “boosting” algorithms” (co-PI with Thomas Malone and Patrick Winston) November 1 2010 - October 1 2011.
Amount of award: 0.5 postdoctoral research fellowship

MIT Energy Initiative (MITEI) Seed Fund Program. “A Novel Framework for Electrical Grid Maintenance” (Single PI) May 17 2010 - May 16 2012.
Amount of award: \$150,000

Ford - MIT Alliance. “Achieving Top Quality Ratings with Minimal Cost” (Single PI) July 1 2010 - June 20 2011.
Amount of award: \$106,013

Con Edison Company of New York. “Manhole Events and Secondary System - Machine Learning Project”
Secondary System Project, Manhattan Backbone, December 1, 2011-December 31, 2013, Amount of subaward to MIT (Single PI): \$300,129

Secondary System Project, Manhattan Corollary, May 1, 2011 - May 31, 2012, Amount of subaward to MIT (Single PI): \$93,056

Update Manhattan Consolidated Database and Ranking Model through 2009, and Analyze 2004-2005 Secondary Inspections Data, July 1 2010 - December 31 2010, Amount of subaward to MIT (Single PI): \$81,661

Phase 2, Application to B, Q, and X, December 1 2009 - May 31 2010, Amount of subaward to MIT (Single PI): \$81,899

Phase 1, Application to B, Q, and X, January 2009 - June 2009, Co-PI , Amount: \$464,127

Phase 4, July, 2008 - December, 2008, Co-PI , Amount: \$413,947

Phase 3, January, 2008 - June, 2008, Co-PI , Amount: \$347,104

Phase 2, August, 2007 - December, 2007, Co-PI, Amount: \$486,296

Phase 1, March, 2007 - July, 2007, Co-PI , Amount: \$339,251

National Science Foundation. Postdoctoral Research Fellowship in Biological Informatics, Grant DBI-0434636, March 2005 - February 2007.

Amount of award: \$120,000

Honors (not including awards listed with papers above)

Bell Labs Prize, Second Place, 2023. Recognizes game-changing innovations in science, technology, engineering and mathematics. 107 teams competing, 10 semifinalists, 5 finalists.

David and Janet Vaughan Brooks Teaching Award, Duke University, 2023. Awarded by the Trinity College Arts and Sciences Council based on equitable and inclusive teaching, sparking excitement about learning, encouraging a deep dive into disciplinary ways of thinking, making connections beyond the courses they teach, and teaching innovations. Highest teaching award at Duke, 4 teaching awards per year across all departments.

Grand Prize Winner, A- μ -Sing Competition, Consortium for the Advancement of Undergraduate Statistics Education (CAUSE), 2023. This award was based on three bluegrass songs that I wrote about machine learning for my class.

Named by AIo as one of the top 10 Women in the World of AI in 2023

7th International Workshop on Health Intelligence (W3PHIAI-23, AAAI-23 workshop), Winner of Hackathon (\$1K awarded to student Srikar Katta), 2023. Objective: This hackathon challenged the AI community to design an optimal age predictor. There was no restriction on the type of data used for this challenge, and we used sleep EEG to predict biological "brain" age.

Bass Chair (Earl D. McLean, Jr. Professor), Duke University, 2022-2027

Chron15 Pioneer 2021-2022, Duke Chronicle, 2022

Bass Society of Fellows, Duke University, Inducted 2022

Terng Lecturer, Women in Mathematics Program, Institute for Advanced Study, Princeton, 2022

2022 Squirrel AI Award for Artificial Intelligence for the Benefit of Humanity from the Association for the Advancement of Artificial Intelligence (AAAI). This is the most prestigious award in artificial intelligence. This award, similar only to world-renowned recognitions, such as the Nobel Prize and the Turing Award, carries a monetary reward at the million-dollar level.

Guggenheim Fellow, 2022

Fellow of the Association for the Advancement of Artificial Intelligence (AAAI), 2022-present

Thomas Langford Lecture Award, Duke University, 2019-2020

Fellow of the American Statistical Association, 2019-present

Fellow of the Institute of Mathematical Statistics, 2019-present

Duke AI for Art competition (with students Alina Barnett, James Hctor, Chaofan Chen, and Oscar Li), second place, 2019

Faculty associate, Berkman Klein Center for Internet and Society at Harvard University, 2015-2019.

2016 Adobe Digital Marketing Research Award, 2016

Named by Businessinsider.com as one of the 12 most impressive professors at MIT in 2015

National Science Foundation CAREER Award, 2011

"Top 40 Under 40" business school professors of *Poets & Quants*, 2015. (Published in Forbes magazine.)

Nominated for Outstanding UROP Mentor Award, UROP (undergraduate research opportunities) Program, MIT, 2012

Nominated for 2012 Sloan Excellence in Teaching Awards, MIT, 2012

Second Place for Phase 1 in the ICML Exploration and Exploitation 3 Challenge, 2012. Goal is to design a recommender system with high click through rates for Yahoo! Front Page News Article Recommendations
teammates: Virot Ta Chiraphadhanakul and Edward Su

National Science Foundation Postdoctoral Research Fellowship in Biological Informatics, 2005-2007

University at Buffalo College of Arts and Sciences Outstanding Senior Award in Sciences and Mathematics 1999, one per year at the university (also Department of Physics Outstanding Senior 1998, Department of Mathematics Outstanding Senior 1999, Department of Music Outstanding Senior 1999)

Barry J. Goldwater Scholarship, 1997-1998

State University of New York Chancellors Award for Student Excellence 1997, 1999

Dr. Stanley T. Sekula Memorial Scholarship, University at Buffalo Physics Department, 1996, 1997

Hildegard F. Shinnars Prize, 1999, Phi Beta Kappa award to recognize mathematics thesis and music thesis

Phi Beta Kappa, inducted 1997

UB Music Department Scholarships, 1994, 1995, 1996

Intellectual Breath and Liberal Knowledge Award, UB Honors Program, 1999

Second Place-National Winner of the 1993 Young Inventors' and Creators' Competition in the Copyright Category of Popular Music Composition, sponsored by the Foundation for a Creative America. (This came with a congratulatory letter signed by Vice President Al Gore!)

Media (Selected)

Squirrel AI Award:

"Duke Computer Scientist Wins \$1 Million Artificial Intelligence Prize, A 'New Nobel' " AAAI press release, October 12, 2021

"Duke Professor Recognized for Clarifying AI Decision Making" by John McCormick, Wall Street Journal, Lead article in AI section, October 14, 2021

"She won a \$1 million prize for predicting which manholes would explode" by James Barron, New York Times, New York Today section, October 12, 2021 (This title is misleading.)

AI Regulation:

"Biden's Executive Order on AI Is a Good Start, Experts Say, but Not Enough" by Lauren Leffer, Scientific American, October 31, 2023

"What You Need to Know About Biden's Sweeping AI Order: The executive order covers AI safety, algorithmic bias, and privacy" by Eliza Strickland, IEEE Spectrum, October 10, 2023

NYT quote in "How Easy Is It to Fool A.I.-Detection Tools?" by Tiffany Hsu, New York Times, June 28, 2023

CNBC International (newsroom in Singapore), interview on live television, May 29 2023

"Senators reaffirm worries about AI during hearing with ChatGPT founder," Ahtra Elnashar, The National Desk, WJLA (ABC News 7 - based in Virginia), May 16th, 4:00pm EST, 2023

"Duke professor - a critic of 'runaway train' AI development - is named one of the field's top women", WRAL TechWire, April 6, 2023

WRAL TechWire interview: "Artificial intelligence industry is out of control, requires regulation, Duke researcher warns," March 6, 2023

CNN quote in "Experts are warning A.I. could lead to human extinction. Are we taking it seriously enough?" May 30, 2023

Machine Learning Interpretability Interviews:

LinkedIn Live Conversation with Scott Zoldi from FICO, March 8, 2023

"What Do Conspiracy Theories And AI Explainability Have In Common?" by Kareem Saleh, Forbes, May 4, 2021

"Meet Cynthia Rudin—A Champion of Interpretable Machine Learning," by Sam Behseta & Michelle Dunn, Chance Magazine, American Statistical Association, April 22, 2020

"Rise of Robot Radiologists," by Sara Reardon, Nature, Innovations In, December 18, 2019

BBC radio, Digital Planet, December 10, 2019 (discussing "This Looks Like That")

“Machine vision that sees things more the way we do is easier for us to understand” MIT Technology review, Artificial Intelligence, Dec 6 2019

Computer Vision:

“Accurate Neural Network Computer Vision Without the ‘Black Box’: Duke team disentangles neural networks to understand how they see the world,” December 15, 2020

“Artificial Intelligence Makes Blurry Faces Look More Than 60 Times Sharper: This AI turns even the blurriest photo into realistic computer-generated faces in HD”: by Robin A. Smith, Duke Today, June 12, 2020 - published on DukeToday, covered by Newsweek.com, Independent.co.uk and other media outlets.

“Making Blurry Faces Photorealistic Goes Only So Far,” by Mark Anderson. IEEE Spectrum, June 23 2020

Crime Data Mining:

“Possible to predict recidivism? Here’s how...,” The Docket, MSNBC (Live TV), Tuesday May 19, 2015.

“Crime-Fighting Computer Code from Cambridge Police and MIT,” *WBUR Radio Boston* (National Public Radio), Tuesday August 13 2013

“Cambridge police look at math to solve crimes,” *Boston Globe*, Metro Section, front page, Sunday August 4 2013

“Statistician enlisted to fight crime by numbers,” *The Times of London*, US & Americas section, Tuesday August 6 2013

“Predictive Policing: Using Machine Learning to Detect Patterns of Crime,” *WIRED Innovation Insights*, August 22, 2013

Meetings Analysis:

“At Work: Just Say ‘Yeah’,” *Wall Street Journal*, Business section, on page B8 in the U.S. edition, June 19, 2013

“How to be effective at meetings? Say ‘yeah’,” *Toronto Star*, Business section, June 28, 2013

“Researchers discover the key to persuasion,” *ABC News*, consumer report blog / business, June 24, 2013

Energy Grid:

My work discussed in book The Alignment Problem: Machine Learning and Human Values 1st Edition by Brian Christian, W. W. Norton and Company, 2020.

“Why Manhole Explosions Happen in the Summer,” *NBC News*, Business/Energy section, August 19, 2015

“New York’s Exploding Manhole Covers Pose Unexpected Winter Hazard,” *Reuters*, appeared in *New York Times*, and *MSN.com*, February 28, 2015

My work discussed in book Big Data: A Revolution that Will Transform how we Live, Work, and Think, by Victor Mayer-Schönberger and Kenneth Cukier, Houghton Mifflin Harcourt Publishing Company, 2013

Analytics Magazine, INFORMS. Headlines: Innovative Applications in Analytics Award, April 18, 2013

“Machine Versus Manhole,” *ScienceNews*, *U.S. News and World Report*, *WIRED Science*, *Slashdot*, *Discovery News / Discovery Channel*, and others, July 8-9 2010

Information Retrieval / Building New Search Engines:

Radio segment about my work on Growing a List. “A New Way to Google,” *Boston Public Radio*, show on innovation at 12:45pm-1pm, hosted by Kara Miller, October 9, 2012

Health and Interpretable Predictive Models:

“How Can Doctors Be Sure A Self-Taught Computer Is Making The Right Diagnosis?,” by Richard Harris, All Things Considered, NPR, April 1, 2019

“Do You Zone Out? Procrastinate? Might Be Adult ADHD,” by Rebecca Hersher, NPR, April 5 2017

“Algorithms Learn From Us, and We’ve Been Bad Parents,” by Bahar Gholipour, Mach Technology, NBC News, March 10 2017, 2:17 PM ET

“New Computer Tool Can Predict Dementia From Your Simple Drawings” Popular Science, August 13, 2015

“Digital Pen is Better Dementia-Prediction Tool than a Doctor” WIRED Magazine, August 17, 2015

“Computers that teach by example: New system enables pattern-recognition systems to convey what they learn to humans.” MIT News (also front page of MIT main website), December 5-10, 2014

“New Statistical Model Lets Patient’s Past Forecast Future Ailments,” Science News section, *Science Daily*, June 9, 2012

Other topics:

Quoted in “AI Is Getting Powerful. But Can Researchers Make It Principled?” by Mordechai Rorvig, *Scientific American*, April 4, 2023.

Scientific Sonnets: Duke team wins competition for poetry-generating algorithm, *Duke Chronicle*, Dec 27, 2018

Article discussing the whitepaper effort I led: *AmstatNews: The Membership Magazine of the American Statistical Association News*. Cover Story, *Statistical Scientists Advance Federal Research Initiatives*, July 1, 2014.

Article about my work: “How to Improve Product Rankings,” *Businessweek*, B-School Research Briefs section, March 9, 2012

Interview: “Should There Be Enforceable Ethics Regulations on Generative AI?” *InformationWeek*, article by Joao-Pierre S. Ruth, March 14, 2023

Op-Eds:

Brandon Garrett and Cynthia Rudin. **Seeking To Regulate AI Is a Good Start. Next, Tackle the Secretive Way Government Uses It.** *The Messenger*, November 2, 2023.

Cynthia Rudin, Zhicheng Guo, Cheng Ding and Xiao Hu. **How good are AI health technologies? We have no idea**, *STAT*, First Opinion, Oct 11, 2023

Brandon Garrett and Cynthia Rudin. **What the Supreme Court Doesn’t Understand About AI**, *The Messenger*, June 3, 2023

Cynthia Rudin and Lance Browne. **A reasonable right to biometric privacy.** (They retitled it to this title: “That selfie you posted on Instagram? Companies are using it in unethical ways.”) Published in: *Charlotte Observer*, *Raleigh News & Observer*, *Durham Herald-Sun*, and *Las Vegas Sun*, March 27, 2023

Cynthia Rudin. **How dangerous is AI? Regulate it before it’s too late**, *The Hill*, February 8, 2023

Cynthia Rudin. **‘The Marriage Pact’ and the risks we take with data**, *The News&Observer*, February 28, 2021

Cynthia Rudin. **No more excuses. Make data more accessible.** *Washington Post*, Opinions, part of collection “We need smart solutions to mitigate the coronavirus’s impact. Here are 46.” June 18, 2020

Professional Societies and Government Committees

National Artificial Intelligence Advisory Committee Law Enforcement Subcommittee (NAIAC-LE), 2023-2026

Committee on Facial Recognition Technology: Current Capabilities, Future Prospects, and Governance, National Academies of Science Engineering and Medicine, 2022-present

Executive Committee Member, ACM SIGKDD, 2021-present

US Department of Labor Technical Working Group on Automated Approaches for Data Catalogs, May 17, 2023

Associate Director, Statistical and Applied Mathematical Sciences Institute (SAMSI), 2018-2021

Chair of Committee to Choose the First Editor-in-Chief of the new INFORMS Journal on Data Science, INFORMS, 2019-2020

Judge for Edelman Award, INFORMS 2019-2020, 2020-2021

Technology Strategy Committee, INFORMS, 2019-present

Member of Committee for Computing Community Consortium (CCC) for Interaction for AI / Roadmap for Future AI, January 2019

Chair, Section on Statistical Learning and Data Mining, American Statistical Association, 2017-2018.

Member of Committee on Applied and Theoretical Statistics (CATS), National Academies of Sciences, Engineering, and Medicine, 2016-2022

Member of Committee on Law and Justice (CLAJ), National Academies of Sciences, Engineering, and Medicine, 2017-present

Councilor of the AAAI, 2017-2020

Chair, INFORMS Data Mining Section, 2015-2016 (Vice Chair for 2014-2015, Council Member 2017-2018, Council Member, 2011- 2013).

Member of Committee on Analytical Research Foundations for the Next-Generation Electric Grid, and author of consensus report: “Analytical Foundations for the Next Generation Electric Grid,” National Academies of Sciences, Engineering and Medicine, 2014-2016.

Bureau of Justice Assistance Criminal Justice Technology Forecasting Group (BJA CJTFG), United States Department of Justice, 2014-2016.

DARPA Information Science and Technology (ISAT) study group (faculty advisory board of DARPA), 2014-2018.

American Statistical Association Committee on Funded Research, 2015-2018.

MIT Energy Initiative, Energy Education Task Force, 2013-2014.

Activities

Events Organized

Seamless/Seamful Human Computer Interaction, NSF workshop, co-organizer, May 17 and 20, 2021.

FAIF: Fair AI in Finance, NeurIPS workshop, co-organizer, 2020.

Self-Supervised Learning – Theory and Practice, NeurIPS workshop, co-organizer, 2020.

Triangle Machine Learning Day, lead organizer, co-organized with David Banks, Jeremy Freeman, and Ted Enamorado, September 20, 2019.

SAMSI Deep Learning Program, co-organizer and leader of working group, 2019

Conference co-Chair, Conference on Statistical Learning and Data Science / Nonparametric Statistics, co-Chair with Annie Qu, American Statistical Association, June 4-6, NYC, 2018

Triangle Machine Learning Day, lead organizer, co-organized with Jade Vinson, Richard Lucic, and Kirsten Shaw, April 3, 2018.

Member of planning committee, SAMSI Program on Statistical, Mathematical, and Computational Methods for Precision Medicine 2016-2018.

TAMALE: Toolkit of Algorithms for Machine Learning, DARPA ISAT workshop, co-organized with Margo Seltzer, March 2018.

Judge for INFORMS Data Mining Best Paper competitions, INFORMS, 2018.

Member of organizing committee, Conference on Statistical Learning and Data Science, UNC Chapel Hill. 2016.

What if: Machine Learning Models for Causal Inference, DARPA ISAT workshop, co-organized with Dustin Tingley, February 2016.

The Cassandra Problem: Building Trust in Predictive Models, DARPA ISAT workshop, co-organized with Carla Brodley and Stephen Boyd, April 2015.

Invited Session: The Fifth V in “Big Data” is *Variables*, co-organizer with Tyler McCormick, Joint Statistical Meetings, 2015.

Topic Contributed Session: Predictive Policing, organizer, Joint Statistical Meetings, 2014.

Judge for Statistical Learning and Data Mining Best Student Paper competition, American Statistical Association, 2015.

Judge for 2015 INFORMS Innovative Applications in Analytics Award, 2014-2015.

Workshop on Data Analytics: Challenges in Big Data for Data Mining, Machine Learning and Statistics organizer, MIT CSAIL Big Data, March 26, 2014.

The ISBIS (International Society of Business and Industrial Statistics) 2014 and SLDM (Statistical Learning and Data Mining section of the American Statistical Association) Meeting on Data Mining in Business and Industry, Program Committee, June 9-11, 2014.

Statistical Analysis and Data Mining (a journal of the American Statistical Association), committee to choose the next editor-in-chief, 2014-2015.

ECML/PKDD 2013 Workshop on “DARE: Data Analytics for Renewable Energy Integration”, Technical program committee member, 2013.

Workshop on Hospital Readmission Prediction and Clinical Risk Management (HRPCRM) at IEEE International Conference on Healthcare Informatics (ICHI) 2013, program committee member, organizers are John Cromwell and Si-Chi Chin

Session on Smart Grid Data Analytics (SGDA) at International Conference on Smart Grid and Clean Energy Technologies (ICSGCE) 2013, co-chair with Zeyar Aung

Dagstuhl Seminar on Preference Learning, co-organizer, 5 day seminar, 45 participants, Germany, March 3th to March 7th, 2014.

IMS/ASA Spring Research Conference, organizer of Machine Learning session, Harvard University, June 14th, 2012.

New England Machine Learning Day, Co-organizer, Microsoft Research New England, May 16th, 2012.
New England Statistics Symposium (NESS), organizer of Machine Learning session, Boston University, April 21, 2012.
Collective Intelligence 2012 (CI 2012), Local Arrangements Chair, 2012
INFORMS Data Mining and Health Informatics Workshop (DH-MI), co-organizer, 2011
MIT Energy Initiative, Organizing Committee for the MITEI Seminar Series, member, Fall 2010- Spring 2012
AMS Short Course on Aspects of Statistical Learning, organizer, 2007 AMS joint math meetings, New Orleans, January 3-4, 2007
AMS/AWM/MAA Special Session on Mathematical Results and Challenges in Learning Theory, session organizer, AMS joint math meetings, San Antonio Texas, January 12-15, 2006
Program for Women in Mathematics, Institute for Advanced Study, Program Committee Member, 2003-2006. Women in Science Seminar organizer, 2004, 2005, 2006. TA for the Wavelets course in 2002. Discussion group organizer, 2001. Research seminar speaker 2003. Poster session organizer 2003, Panel discussion organizer 2003, Panelist 2007
PACM Conference Princeton University, organizer, 2002-2004, speaker 2005

Editorial Responsibilities

Associate Editor for Harvard Data Science Review, 2019 - present
 Associate Editor for Management Science, in the Big Data Analytics department, 2018 - 2024
 Associate Editor for INFORMS Journal on Data Science, 2020 - 2023
 Associate Editor for Journal of Quantitative Criminology, 2021 - 2022
 Action editor: 2013 - 2017, (Editorial board member: June 30 2010 - June 30 2013), Machine Learning Journal
 Action Editor: Statistical Analysis and Data Mining (SAM, a journal of the American Statistical Association), 2012-2017
 Editor for Special Issue on Sports Analytics for Statistical Analysis and Data Mining (SAM, a journal of the American Statistical Association), co-editor with Theja Tulabandhula, 2015
 Editorial Board, Journal of Artificial Intelligence Research (JAIR), July 2014 - June 2017
 Member of Guest Editorial Board for ECMLPKDD 2014 journal track
 Editor for Special Issue on Machine Learning for Science and Society, co-editor with Kiri Wagstaff, 2012-2013
 Editorial board member: Journal of Machine Learning Research, 2012 - present

Program Committee Memberships

NIPS Workshop on Machine Learning for Healthcare ML4HC (2016, 2017, 2018), ECML-PKDD Workshop on Social Good (2016), ICML Workshop on Human Interpretability for Machine Learning (2016), ECML-PKDD Workshop on Data Science for Social Good (2016), SDM senior pc (2016), ICDM area chair (2015), AAAI senior pc (2016), NIPS area chair (2015), Visual Data Science (2015), IJCAI senior pc (2015), ICML area chair (2015) ICDM area chair (2014), IEEE Big Data (2013), NIPS area chair (2013), AAAI (2013), ICML area chair (2013), NIPS area chair (2012), ECML-PKDD (2012), ICML (2012), COLT (2011), IJCAI (2011), ECML-PKDD (2010), ICML (2009), ICML (2006), AAAI (2006), KDD (2005)

Referee Assignments

Journal Reviewing:

Biometrics, Harvard Data Science Review, Journal of Artificial Intelligence Research (JAIR), Journal of Quantitative Criminology (JOQC), Data Mining and Knowledge Discovery (DAMI), Journal of Machine Learning Research (JMLR), Machine Learning Journal, IEEE Transactions on Pattern Analysis and Machine Intelligence (TPAMI), Mathematical Programming, Journal of Computational Statistics and Data Analysis, Communications on Pure and Applied Mathematics (CPAM), Management Science, International Journal of Renewable Energy Research (IJRER), Annals of Operations Research, Operations Research, Recent Patents on Computer Science, The Lancet, Scientific Reports - Nature, Science Advances, Decision Support Systems, AI Magazine

Conference and Other Reviewing: NeurIPS/NIPS (Neural Information Processing Systems) 2020, 2019, 2018, 2017, 2011, 2010, 2009, 2008, 2007. IEEE Computer 2019, ICDM (International Conference on Data Mining) 2011, Grant Proposal Review for Natural Sciences and Engineering Research Council of Canada 2016, Book Proposal for Springer 2013, COLT (Conference on Learning Theory) 2012, COLT 2010, COLT 2005, AISTATS (Conference on Artificial Intelligence and Statistics) 2012, AISTATS 2010, ICMLA (International Conference on Machine Learning and Applications) 2011, Book Proposal for Manning Publications 2010, ALT (Algorithmic Learning Theory) 2010, ICML

(International Conference on Machine Learning) 2010, NIPS Ranking Workshop 2009, ACM SIGKDD (Conference on Knowledge Discovery and Data Mining) 2009, Applied and Computational Harmonic Analysis Journal, 2006, Conference on Machine and Statistical Learning: Prediction and Discovery 2006, Reviewer for MRC Biostatistics Unit at the University of Cambridge 2023

Presentations

Keynotes/Invited/Plenary for Notable Conferences

Conference on Statistical Practice, American Statistical Association, Keynote, February 28, 2024

SPIE Medical Imaging, Opening Plenary, February 18, 2023

International Conference on Data Mining (ICDM), Keynote, November 30, 2022

INFORMS (Institute for Operations Research and the Management Sciences), Plenary, October 17, 2022

SDM (SIAM International Conference on Data Mining), Keynote, April 29, 2022

AAAI (Association for the Advancement of Artificial Intelligence), Plenary talk for Squirrel AI Award, February 24, 2022

Nobel Conference, Gustavus Adolphus College, October 4-6, 2021

CODE (Conference on Digital Experimentation), MIT, November 2, 2019

DSAA (IEEE International Conference on Data Science and Advanced Analytics), Washington DC, October 7th, 2019

KDD (25th ACM SIGKDD Conference on Knowledge Discovery and Data Mining), Anchorage, Alaska, August 8th, 2019

Women in Data Science Cambridge Conference (organized by Harvard, MIT, Microsoft Research New England, Stanford) March 4, 2019

ECML-PKDD (The European Conference on Machine Learning & Principles and Practice of Knowledge Discovery in Databases), Dublin, September 11, 2018

Females Excelling More in Math, Engineering, and Science (FEMMES), keynote for capstone event, 200 girls grades 4-6, Duke University, February 17, 2018

ISAIM (International Symposium of Artificial Intelligence and Mathematics), January 3, 2018

MLHC (Machine Learning for Healthcare), August 18, 2017

AISTATS (Artificial Intelligence and Statistics), April 22, 2017

The Frontiers of Machine Learning, Forum on Machine Learning, National Academy of Sciences and The Royal Society, Jan 31-Feb 1, 2017

FAT-ML (Fairness, Accountability and Transparency in Machine Learning), November 18, 2016

KDD (20th ACM SIGKDD Conference on Knowledge Discovery and Data Mining), August 24-27, 2014

Other Conference Invited Presentations

(This list does not include regular or invited talks at annual meetings such as JSM, INFORMS, SLDM/SLDS, American Society for Criminology, Atlantic Causal Inference Conference, which I participate in most years.)

Winter School for Ethics of Neuroscience and AI, invited talk, March 1, 2024

XAI4Sci, Explainable AI for Sciences Workshop, AAAI, invited speaker, February 26, 2024

CanQueue (Conference on Queueing Theory), keynote, August 25, 2023

Disney Data & Analytics Conference (DDAC), featured speaker, August 15, 2023

Introductory Overview Lecture (IOL) in Interpretable Machine Learning at the Joint Statistical Meeting (JSM), August 8, 2023

5th International Conference on Statistics: Theory and Applications (ICSTA 2023), keynote, August 4, 2023

International Conference on AI and the Digital Economy, keynote, June 26, 2023

Government Advances in Statistical Programming (GASP), keynote, June 14, 2023

INFORMS Conference on Quality, Statistics, and Reliability, North Carolina State University, plenary, June 7, 2023

AI for Scientific Data Analysis: Mini-conference, Chalmers University of Technology, Gothenburg, Sweden, June 1, 2023

Forum for Information Retrieval Evaluation (FIRE), plenary, December 13, 2022
NeurIPS Human in the Loop Learning (HiLL) Workshop, invited talk, December 2, 2022
IEEE/WIC/ACM International Conference on Web Intelligence (WI'22), Keynote, November 18, 2022
Explainable AI in Finance Workshop at the 3rd ACM International Conference on AI in Finance (XAI-FIN@ICAIF2022),
 Invited Talk, November 2, 2022
Information Technology Laboratory (ITL) Science Day, National Institute for Standards and Technology (NIST), Keynote,
 October 24, 2022
ECCV Workshop on Safe Artificial Intelligence for Automated Driving (SAIAD), Invited Speaker, October 24, 2022
ECCV Workshop on Responsible Computer Vision, Panel Member, October 23, 2022
Making Sense of Interpretable Machine Learning Workshop, Keynote, Lorentz Center, Leiden, Netherlands, (virtual),
 October 17, 2022
INFORMS 17th Workshop on Data Mining & Decision Analytics, and **INFORMS 6th Workshop on Data Science**,
 Keynote, INFORMS, Indianapolis, October 16, 2022
AI-ML Systems Second International Conference on AI-ML Systems, Keynote, October 14, 2022
ICCBR International Conference on Case-Based Reasoning, Keynote, September 14, 2022
ICML workshop on Machine Learning in Healthcare, Keynote, July 23, 2022
ICML workshop on Responsible Decision-Making in Dynamic Environments, Keynote, July 23, 2022
IEEE World AI IOT Congress, Keynote, June 8, 2022
Interpretability in Artificial Intelligence Workshop, Banff International Research Station, Keynote, May 3, 2022
Discover Experiential AI, The Institute for Experiential AI, Northeastern University, Panelist, April 6, 2022
Walmart AI Summit, Plenary speaker, April 6, 2022
INFORMS Optimization Society Conference, Plenary speaker, Greenville, SC, March 14, 2022
5th Annual Swedish Symposium on Deep Learning / 39th Annual Swedish Symposium on Image Analysis, March 14,
 2022
Stu Hunter Research Conference on Statistics and Statistical Engineering, 2022, keynote, Duke University, March 8,
 2022
AAAI Workshop on Interactive Machine Learning, keynote, February 28, 2022
AAAI Explainable Agency in Artificial Intelligence Workshop, Keynote, February 28, 2022
Government of Canada Data Conference 2022: Driving Data Value and Insights for All Canadians, Panelist, February
 23, 2022
Fidelity AI Day, Keynote, February 16, 2022
The Where and Why of Explainable AI, Science and Technology Expert Partnership Advances in Explainable AI Work-
 shop, MITRE Labs, panelist, January 11, 2022
5th Joint International Conference on Data Science & Management of Data (CODS-COMAD 2022), keynote, India
 (virtual), January 7, 2022
**HICSS-55: SWT on Future of Human Work: Harnessing the Power of Augmented Intelligence and Augmented
 Cognition**, panelist, January 3, 2022
Interpretability, Safety, and Security in AI, The Alan Turing Institute, December 14, 2021
Human-Centered AI (HCAI) Workshop at NeurIPS, Keynote, December 13, 2021
Machine Learning & Supply Chain Management Workshop, TRIPODS-X, December 13, 2021
Forward Summit, Puerto Rico Science, Technology & Research Trust, December 10, 2021
2021 IEEE Symposium Series on Computational Intelligence (SSCI), Deep Learning Track, Keynote, December 5,
 2021
Gillmore Symposium: Explainable, Interpretable AI: the Future of Investment Management, Warwick University,
 November 19, 2021
Conference on Non-traditional Data, Machine Learning and Natural Language Processing in Macroeconomics, Bank
 of Canada, panelist, November 18, 2021
DARPA DSO Futures meeting, plenary speaker, November 18, 2021
Duke in DC discussion on “The Equitable, the Ethical and the Technical: Artificial Intelligence’s Role in the U.S. Criminal
 Justice System,” November 15, 2021

McCombs School of Business' Center for Analytics and Transformative Technologies 2021 Global Analytics Conference, keynote, November 12, 2021

Artificial Intelligence in Consumer Finance: Defining and Insuring Fairness, Federal Reserve Bank of Philadelphia and Federal Reserve Bank of Cleveland, November 9, 2021

AAAI Fall Symposium, keynote, November 5, 2021

Advanced Analytics: New Methods and Applications for Macroeconomic Policy, organized by the Bank of England, the European Central Bank, King's College London and King's Business School, keynote, November 4, 2021

Methodological Approaches for Whole Person Research Workshop, NIH, September 29-30, 2021

CNRS School on Explainability, September 30, 2021

iMIMIC Workshop on Interpretability of Machine Intelligence in Medical Image Computing at MICCAI 2021, September 27 2021

Explainable AI Virtual Workshop, Caltech, September 23, 2021

AI Meets Regulators Symposia, AI for Health at Imperial College, September 21, 2021

ICCAI'21 International Conference on Complex Acute Illness, keynote, September 10, 2021

FICO Mastermind Event, Aug 30 – Sept 19, 2021

ARES / CD-MAKE 2021 Conference, keynote, August 20, 2021

ELLIS workshop on Causethical ML, Invited talk, July 26, 2021

ICML Workshop on Theoretic Foundation, Criticism, and Application Trend of Explainable AI, Invited talk, July 23, 2021

CVPR 2021 Tutorial on Interpretable Machine Learning

Forecasting the future for sustainable development, Centre for Excellence and Transdisciplinary Studies, hosted by OECD, keynote, June 17, 2021

Analytics Summit 2021, University of Cincinnati, June 8, 2021

Responsible Machine Learning, keynote, North Carolina State University, June 4, 2021

ICLR Workshop on Responsible AI, May 7, 2021

Bringing Artificial Intelligence to the Bedside, Purdue University Workshop, April 23, 2021

Trustworthy Automated Decision Making (ETAPS 2021 Workshop), keynote, March 28, 2021

Explainable AI Mini-Summit, Re-Work, February 17, 2021

Panel: Bias in AI: How scientists are trying to fix it?, Intuit, February 15, 2021

Florida Women in Mathematics Day, Florida Atlantic University, keynote, February 13, 2021

IEEE EMBS Forum on Data Science and Engineering in Medical Imaging, Symposium #1: Grand Challenges in Data Science and Engineering in Healthcare: Medical Imaging, February 10, 2021

The WACV 2021 Workshop on Explainable & Interpretable Artificial Intelligence for Biometrics (xAI4Biometrics Workshop 2021), keynote, January 5, 2021

Machine Learning in Science & Engineering (MLSE), Columbia University, December 14, 2020

NeurIPS Workshop on Broader Impact of AI, December 12, 2020

Toronto Machine Learning Summit, November 19, 2020

Advancing Analytics 2020, National Conference, Institute of Analytics Professionals of Australia, November 17, 2020

The Machine Learning Conference (MLconf), November 6, 2020

Explainable AI Planning (XAIP) @ ICAPS, October 21, 2020

Symposium on Artificial Intelligence for Learning Health Systems (SAIL), Presymposium, October 21, 2020

Advances in Interpretable Machine Learning and Artificial Intelligence (AIMLAI), keynote, workshop at CIKM, October 20, 2020

Workshop on Credit Card Lending and Payments, Federal Reserve Bank of Philadelphia, September 17, 2020

National Health Symposium, Johns Hopkins Applied Physics Laboratory, September 14, 2020

Data Science, Statistics & Visualization (DSSV), SAMSI, July 29, 2020

Workshop on New Directions in Optimization, Statistics and Machine Learning, Institute for Advanced Study, April 16, 2020

TutORial, INFORMS, October 21, 2019

Advances in Decision Analysis, INFORMS Decision Analysis Society, (keynote), Bocconi University, Milan, June 19, 2019

Debugging Machine Learning Models: ICLR 2019 workshop, May 6, 2019

Safe Machine Learning: Specification, Robustness and Assurance: ICLR 2019 Workshop, May 6th, 2019

FEAP-AI4Fin 2018 : NeurIPS 2018 Workshop on Challenges and Opportunities for AI in Financial Services: the Impact of Fairness, Explainability, Accuracy, and Privacy, invited talk, December 7, 2018

NeurIPS 2018 Workshop on Critiquing and Correcting Trends in Machine Learning, (Spotlight talk), December 7, 2018

Nicholas Institute, Duke University, invited commenter on deep learning session, October 5, 2018

3rd International Workshop on Biomedical Informatics with Optimization and Machine Learning (BOOM), invited talk, IJCAI workshop, June 13, 2018

Workshop on Human Interpretability in Machine Learning (WHI), invited talk, ICML workshop, June 17, 2018

CVPR NTIRE 2018 New Trends in Image Restoration and Enhancement workshop and challenges on super-resolution, dehazing, and spectral reconstruction, invited talk, representing our competition entry on superresolution, June 18, 2018

Triangle Machine Learning Day, invited talk, April 3, 2018

4th Annual Morgan Stanley Quantitative Equity Research Conference, invited talk, November 17, 2016

Predictive Applications and APIs (PAPIs), October 2016

Japan-America Frontiers of Engineering Symposium, National Academy of Engineering, invited talk, Beckman Center, Irvine CA, June 16, 2016

Criminal Justice in the Age of Big Data, panelist. Harvard Kennedy School. November 13, 2015

Discovery Science, Banff, Canada. October 5, 2015

Duke University Workshop on Sensing and Analysis of High-Dimensional Data, Duke University, July 27-29, 2015

Operations, Technology and Information Management Research Camp, Johnson School of Management, Cornell University. June 23, 2015

MIT Big Data Workshop – “From Data to Insights”. May 21, 2015

MIT Media Lab – DARPA ISAT Workshop on Computational Health. April 20, 2015

Workshop on Big Data and Statistical Machine Learning. (Part of the Thematic Program on Statistical Inference, Learning, and Models for Big Data.) Fields Institute, Toronto, January 29, 2015

Bringing Social Science Back In: The ‘Big Data’ Revolution and Urban Theory, Radcliffe Institute, Harvard University, December 15-16, 2014

Understanding and Improving Cities: Policy/Research Partnerships in the Digital Age, Invited Speaker/Panelist, District Hall, Boston, December 12, 2014

Weathering the Data Storm: The Promise and Challenges of Data Science, Third Annual Symposium on the Future of Computation in Science and Engineering, Harvard University, Jan 26, 2014.

New England Machine Learning Day, May 1, 2013

Exploration and Exploitation 3, Presentation for Second Place team in Data Mining Contest, ICML Workshop, June 2012

IMA Workshop on User-Centered Modeling, May 2012

IMA Workshop on Machine Learning, March 2012

ICML Workshop on Global Challenges, International Conference on Machine Learning, July 2, 2011

American Institute of Mathematics (AIM) workshop on the Mathematics of Ranking, August 16-20, 2010

DIMACS/CCICADA Workshop on Statistical Issues in Analyzing Information from Diverse Sources, Rutgers University, May 6-7, 2010

New England Statistics Symposium (NESS), Harvard University, April 17, 2010

International Utility Working Group: Workshop on Computer-Aided Lean Management (CALM), Columbia University, April 16, 2008

Conference on Machine and Statistical Learning: Prediction and Discovery, AMS-IMS-SIAM Summer Research Conferences in the Mathematical Sciences, invited by organizers Joe Verducci, Xiaotong Shen, and John Lafferty, Snowbird, Utah, June 25-29th, 2006

FOCM 2005 Foundations of Computational Mathematics, Workshop 4 on Learning Theory, Santander Spain. Invited by organizer Steve Smale, 2005

Machine Learning Summer School, Workshop on the Dynamics of Learning, TTI-C, Chicago, invited by organizer Steve Smale, May 16-26, 2005

Notable Panel Discussions

U.S. Senate AI Insight Forum: Transparency, Explainability, Intellectual Property, & Copyright. Invited by Senator Chuck Schumer, facilitated by Leader Schumer, Senator Rounds, Senator Heinrich, and Senator Young, Kennedy Caucus Room at the U.S. Senate, Washington DC, November 30, 2023.

AI Roundtable with Congresswoman Deborah Ross and Congresswoman Valerie Foushee, SAS Campus, November 27, 2023 and February 20, 2024

Invited Presentations at Universities, Societies, and Research Labs

Princeton Quantitative Social Science Colloquium, Princeton University, December 8, 2023

AI for Good Seminar, Organized by International Telecommunication Union (ITU) and United Nations partners, November 15, 2023

Statistics Student Seminar, University of Michigan, November 9, 2023

RAND Statistics Seminar, November 8, 2023

NPL Artificial Intelligence Interest Group, National Physical Laboratory, United Kingdom, May 18, 2023

WNAR Webinar, Western North American Region of The International Biometric Society, May 12, 2023

Harvard Statistics Department Seminar, February 6, 2023

University of Michigan, AI Seminar, January 24, 2023

Women in AI Speaker and Mentorship Series, Deloitte and the Schwartz Reisman Institute for Technology and Society (SRI) at the University of Toronto, January 18, 2023 and January 24, 2023

Tutorial on Machine Learning, Merck, December 12, 2022

Quantitative Biology Seminar Series, Cancer Research UK Cambridge Institute, December 12, 2022

IAA Summit Keynote, Institute for Assured Autonomy, Johns Hopkins University, November 1, 2022

SILO Seminar, University of Wisconsin, October 26, 2022

ATHENA Seminar, Duke University, October 25, 2022

Data, Environments, and Learners: Theory and Algorithms, UCL Statistical Science, October 21, 2022

National Science Foundation (NSF) CISE Distinguished Lecture Series, October 20, 2022

Computational Biomedicine Grand Rounds, Cedars Sinai Medical Center, October 12, 2022

Imageomics External Speaker Series, Imageomics Institute, Ohio State University, October 3, 2022

OIM department seminar at the Wisconsin School of Business, September 30, 2022

NISS/Merck Meetup: Interpretable/Explainable Machine Learning, September 28, 2022

Colloquium, Department of Statistics and Actuarial Science, University of Iowa, September 22, 2022

Keynote for Annual Meeting, Tel Aviv University Center for AI and Data Science (TAD), June 8, 2022

Institute for Advanced Study, Terng Lecturer, Women and Mathematics Program, May 23-27, 2022

Fields Machine Learning Seminar, Fields Institute for Research in Mathematical Sciences University of Toronto, April 25, 2022

Data Valorization Institute (IVADO) in Montreal, Zooming in on Multidisciplinary AI, April 21, 2022

Michigan Institute for Data Science Seminar Series, April 14, 2022

International Conference on Foundations and Applications of AI, Peking University, April 8, 2022

UCSB Center for Responsible Machine Learning Distinguished Lecture Series, UC Santa Barbara, April 8, 2022

“I Can’t Believe It’s Not Better” Seminar Series, April 7, 2022

Elon University Analytics Day, March 24, 2022

Cambridge Centre for AI in Medicine Seminar, Cambridge UK, March 16, 2022

Janelia Research Campus Computation & Theory Seminar, Janelia Farms, March 10, 2022

The Alan Turing Institute, Human-Machine Teams Seminar, February 25, 2022

Institute for Experiential AI, Northeastern University, Bay Area, Distinguished Speaker Series February 23, 2022

University of Virginia, AI and Machine Learning Seminar, February 4, 2022

Institute for Assured Autonomy Speaker Series, Johns Hopkins University, February 3, 2022

London School of Economics & Political Science, Data Science Seminar, January 31, 2022

Roche Advanced Analytics Network, Roche/Genentech, January 26, 2022

Image Guided Cancer Therapy Seminar, MD Anderson Cancer Center, January 19, 2022

Perspectives in Mathematical Sciences, Dr. F.C. Kohli Centre of Excellence, Chennai Mathematical Institute, January 19, 2022

Trustworthy ML, Fireside Chat, December 16, 2021

The Myth of “Explainable” AI and why “Interpretable” AI is the Answer, Collective AI Podcast, December 15, 2021

AI for Good, Accelerating the United Nations Sustainable Development Goals, Seminar Series, International Telecommunication Union, December 9, 2021

Montreal Speaker Series in the Ethics of AI, December 9, 2021

Chalmers AI Talks, Chalmers University in Sweden, December 8, 2021

Teaching Youth to Build and Deploy Responsible AI for Justice Workshop, Technovation, December 7, 2021

ECE Distinguished Speaker Series, Rice University, December 6, 2021

Oxford Women in Computer Science Distinguished Speaker Series, November 9, 2021

Smarsh Advanced AI panel on Driving Adoption of AI, November 1, 2021

Responsible Modelling in Uncertain Times — CEST-UCL Seminar series, Keynote, November 3, 2021

RBCDSAI LatentView Colloquium, Robert Bosch Centre for Data Science and AI at IIT-Madras, October 29, 2021

TOM Seminar, Harvard Business School, October 23, 2021

Optimization and Data Science Community Seminar, Exxon, October 7, 2021

Neyman Seminar, Berkeley Statistics Department, September 15, 2021

Trustworthy ML Initiative Seminar, Harvard, July 8, 2021

Fidelity AI Seminar, June 29, 2021

University of Basel, Computational Seminar Series, Computational Biology core program of the Biozentrum, Basel, Switzerland, May 31, 2021

University of Oxford, Oxford Computational Statistics and Machine Learning (OxCSML) Seminar, May 14, 2021

FDIC Center for Financial Research Seminar, May 11, 2021

EPFL Statistics Seminar, May 7 2021

Distinguished Seminars on Explainable AI, connected to ERC project “XAI Science and technology for the eXplanation of AI decision making,” April 20, 2021

Frontiers of Big Data, AI, and Analytics, Virtual Seminar Series (centered in Australia), April 14, 2021

Iowa State University, Theoretical and Applied Data Science Seminar, April 8, 2021

National Institute of Cancer, Biostatistics Branch, Division of Cancer Epidemiology and Genetics, Seminar, April 7, 2021

New Jersey Institute of Technology Data Science Seminar, March 10, 2021

Metron/George Mason University AI/ML Seminar Series, March 10, 2021

NeEDS Mathematical Optimization Seminar Series, Network of European Data Scientists, February 8, 2021

Brown University, Division of Applied Mathematics Colloquium, February 4, 2021

Texas A&M Institute of Data Science, TAMIDS Seminar, January 30, 2021

Bias² Seminar, Harvard Data Science, January 28, 2021

University of Oxford, Computer Science, OATML group, January 26, 2021

University of California, Santa Cruz, Winter Seminar Series, Department of Statistics, January 25, 2021

Verizon Media Research Day, keynote, December 17, 2020

University of Iowa, Department of Business Analytics Seminar Tippie College of Business, December 11, 2020

Duke Computing Roundtable, December 10, 2020

Energy Data Analytics Symposium: Transforming Energy Systems with Data Science Techniques, Duke University, December 9, 2020

Alberta Machine Intelligence Institute, University of Alberta, December 4, 2020

North Carolina Agricultural and Technical State University (NCAT), Seminar, organized by Student Leadership Council and Faculty of the ACIT Institute and TECHLAV Center, November 20, 2020

University of Pittsburgh, Statistics Department Seminar, October 27, 2020

University of California, Davis, Mathematics of Data and Decision in Davis Seminar, October 20, 2020

University of Toronto, MIE Distinguished Seminar Series, Department of Mechanical and Industrial Engineering, October 2, 2020

University of Colorado, Denver, Biostatistics Seminar, September 16, 2020

Online Causal Inference Seminar, August 25, 2020

Tutorial on Interpretable Machine Learning, Joint Statistical Meetings, August 3, 2020

TRIPODS Seminar Series, TRIPODS DATA-INSPIRE Institute, DIMACS & Rutgers CS/Math/Statistics, July 17, 2020

Melbourne Centre for Data Science Seminar Series, June 25, 2020

Decision Making in an Uncertain World, Seminar Series, INFORMS Stochastic Programming Society, June 12, 2020

Center for Human-Compatible AI (CHAI) Seminar, Berkeley, June 10, 2020
 Boeing Research Seminar, June 5, 2020
 Research Triangle Institute (TRI) Seminar, April 21, 2020
 Carnegie Mellon Artificial Intelligence Seminar, April 7, 2020
 GSS/SSS webinar program (offered by the American Statistical Association's Government Statistics Section and Social Science Section, February 13, 2020)
 Ingrid Daubechies Lecture in Computer Science, Duke University, January 21, 2020
 North Carolina State University RED Talk, NCSU Data Science Initiative, November 6, 2019
 North Carolina Chapter of the American Statistical Association, Webinar, November 1, 2019
 Duke University, Langford Lecture, October 31, 2019
 North Carolina State University Statistics Department Seminar, October 4, 2019
 North Carolina State University ECE Interdisciplinary Distinguished Seminar Series, September 27, 2019
 New York University, PRIISM Seminar, September 25, 2019
 University of North Carolina at Chapel Hill, Biostatistics Seminar, August 29, 2019
 SAMSI Seminar, August 28, 2019
 Federal Judicial Center and Duke Law's Bolch Judicial Institute, Duke University, Workshop on Law and Technology for Judges, May 30, 2019
 Yale University, MacMillan-CSAP Workshop on Quantitative Research Methods, April 25, 2019
 University of Pennsylvania, Mahoney Institute for Neurosciences Seminar, April 8, 2019
 Microsoft Research New England Colloquium Series, December 12, 2018
 Boston University, Distinguished Speaker Series, Artificial Intelligence Research Initiative, December 10, 2018
 Princeton University, Quantitative Social Science Seminar, November 16, 2018
 University of North Carolina Statistics and Operations Research Department Colloquium, October 15, 2018
 Research Triangle Analysts, lecture October 16, 2018
 NSF Webinar: Statistics at a Crossroads: Challenges and Opportunities in the Data Science Era, October 2, 2018
 University of North Carolina Health Care and NC Women in Machine Learning and Data Science MeetUp, June 2, 2018
 University of North Carolina, Chapel Hill, Causal Inference Research Group Seminar, April 6, 2018
 North Carolina State University, Bioinformatics Seminar, March 29, 2018
 University of Maryland, Distinguished Speaker Series, Computer Science Department, December 1, 2017
 Temple University, Fox School of Business, Statistics Department Colloquium, October 6, 2017
 Duke University Algorithms Seminar, September 28, 2017
 University of Toronto Law School, Law and Economics Colloquium, September 26, 2017
 Microsoft Research, New York City, Sept 19, 2017
 University at Buffalo, CSE, Distinguished Speaker Series, May 11, 2017
 Columbia University, IEOR-DRO Seminar, May 9, 2017
 Statistics C. V. Starr Lectureship Series, Biostatistics Department, Brown University, April 24 2017
 Computational Social Science and Public Policy Colloquium, Harris School of Public Policy, University of Chicago, April 14, 2017
 Statistics Department Seminar, University of Chicago, April 11, 2017
 Duke Network Analysis Center Seminar, Duke University, February 6, 2017
 Applied Mathematics Seminar, Duke University, October 3, 2016
 Statistics and Complex Systems Seminar, University of Michigan, March 5, 2016
 One Day University, New York, April 29, 2016
 Urban Social Processes Workshop and Quantitative Methods Workshop, Harvard University, March 10, 2016
 Center for Statistics and Machine Learning Seminar Series, Princeton University, October 20, 2015
 Applied Statistics Workshop, Harvard University, October 14, 2015.
 Brainstorming Session on the Next Generation of Search Engines, Berkman Center for Internet and Society, Harvard University, September 22, 2015
 Machine Learning Seminar, Gatsby Unit, University College London. September 16, 2015
 MIT Conversation Series, Accenture. July 10, 2015
 ENAR webinar, (Eastern North American Region, International Biometric Society), May 8, 2015
 Oracle Labs, Research Seminar, April 29, 2015
 NYU-Poly Center for Urban Science and Progress, Research Seminar, March 12, 2015
 American Express (NYC), Decision Science Seminar, March 11, 2015
 Columbia University, IEOR-DRO Distinguished Seminar Series, March 10, 2015
 University of Washington, Statistics Seminar, March 2, 2015

Duke University, Machine Learning Seminar, February 23, 2015
Harvard Business School, Technology and Operations Management Seminar, December 19, 2014
NYU Stern School, Department of Information, Operations & Management Sciences, IOMS Colloquium Series, December 3, 2014
UC Berkeley, Seminar in Computer Science Department, November 12, 2014
MIT Lincoln Labs, Seminar, November 4, 2014.
Brown University, Pattern Theory Group Seminar, October 22, 2014
University of Washington, Data Science Seminar, October 15, 2014
University of Alberta, Operations and Information Systems Seminar, October 3, 2014.
Carnegie Mellon University, ECE Seminar / Machine Learning Special Seminar, September 18, 2014
Harvard University, Computer Science Seminar, September 3, 2014
IBM TJ Watson Research Center, KDD Speaker Day, invited talk, August 28, 2014
MIT CSAIL CAP Meeting, May 30, 2014
MIT Lincoln Labs, Seminar, May 27, 2014
Stanford University, Operations Management Seminar, May 6, 2014
Stanford University, Institute for Research in the Social Sciences, Data Science and Inference Seminar, May 5, 2014
UMass Amherst, Machine Learning and Friends Lunch Seminar, April 29, 2014
MIT International Liaison Program Conference, Plenary Speaker, April 23, 2014
Schlumberger-Doll Research Center Seminar, April 14, 2014
Cornell University Operations Research and Industrial Engineering Seminar, April 8, 2014
Liberty Mutual Research Center Seminar, March 14, 2014
University of Pennsylvania Criminology Seminar, March 7, 2014
MIT Theory of Computation Seminar, March 4, 2014
Harvard Applied Statistics Workshop, October 9, 2013
Harvard/MIT Econometrics Workshop, September 12, 2013
MIT Lincoln Laboratory, Seminar, June 4, 2013
Massachusetts General Hospital (MGH), Quantitative Medicine Seminar, April 29, 2013
UT Austin McCombs School of Business, Research Seminar, April 5, 2013
Laboratory for Information Decision Systems (LIDS) lunchtime seminar, February 26, 2013
North Carolina State University, Statistics Department Seminar, February 28, 2013
Yale Statistics Department Seminar, January 14, 2013
Harvard High Dimensional and Correlated Data Seminar, December 17, 2012
MIT Operations Research Center Seminar, December 11, 2012
Yale Computer Science Department Seminar, December 6, 2012
Columbia University, Statistics Department Seminar, October 22, 2012
Robert H. Smith School of Business, University of Maryland, DO&IT Seminar Series, October 12, 2012
MIT Center for Collective Intelligence, July 17, 2012
Notre Dame Computer Science Department Seminar, November 3, 2011
Rutgers Statistics Department Seminar, October 26, 2011
Wharton Statistics Department Seminar, University of Pennsylvania, September 21, 2011
MIT Energy Initiative External Board Meeting Speaker, October 15, 2010
MIT Energy Initiative Seminar Series, October 12, 2010
Boston University, Probability and Statistics Seminar, October 7, 2010
Microsoft Research New England, Machine Learning Seminar, October 4, 2010
Tufts, Computer Science Seminar, September 30, 2010
ABB (Asea Brown Boveri Ltd) Corporate Research Center - United States, Lunch time Seminar, September 1, 2010.
Harvard Statistics Colloquium, April 5, 2010
MIT Operations Research Center Seminar, February 11, 2010
MIT Imaging Seminar, October 22, 2009
University of Chicago, Statistics Department Seminar, February 23, 2009
Ohio State University, Computer Science Department Seminar, February 19, 2009
Brown University, Applied Mathematics Seminar, February 17, 2009
Indiana University, Computer Science Department Seminar, March 5, 2009
University of Houston, Mathematics Department Seminar, Spring 2009
Polytechnic University, Brooklyn, Computer Science Colloquium, April 30, 2007
Columbia University, Applied Math Seminar, April 3, 2007

New York University, Theory Seminar, Computer Science Department, November 9, 2005
 IBM Yorktown Heights, April 5, 2005
 Rensselaer Polytechnic Institute, March 8, 2005
 CCR (IDA Center for Communications Research), Princeton NJ, March 2, 2005
 Institute for Advanced Study, Computer Science/Discrete Math Seminar, Princeton, February 14, 2005
 Columbia University CCLS Center, February 4, 2005
 Google Labs Inc., Research Seminar, October 29, 2004
 New York University, Harmonic Analysis and Signal Processing Seminar, October 20, 2004
 SUNY at Buffalo, Math Department Seminar, May 2004
 NYU, Workshop on Computational and Biological Learning, January 16, 2004

Teaching and Mentoring

- **CS 671D / STA 671D / ECE 687D Machine Learning, Duke**, graduate and undergraduate machine learning, 2016 (Fall), 2018 (Spring), 2019 (Spring), 2019 (Fall), 2020 (Fall), 2021 (Fall), 2022 (Fall), 2023 (Fall)
- **CS 290 / CS 474, Data Science Competition, Duke**, undergraduate course, 2018 (Spring), 2020 (Spring), 2021 (Spring), 2022 (Spring), 2023 (Spring)
- **ME 555 Applications in Data and Materials Science**, (co-taught) 2021 (Spring)
- **Machine Learning Summer School, Duke**, 2018
- **Microsoft-DAT203x, Data Science and Machine Learning Essentials**, co-taught with Stephen Elston, free online course, edX, 2015. Over 17,500 students registered.
- **Microsoft-DAT203.2x Principals of Machine Learning**, co-taught with Stephen Elston, free online course, edX, 2016. Over 14,500 students registered.
- **Microsoft-DAT203.3x, Applied Machine Learning**, co-taught with Stephen Elston, free online course, edX, 2016. Over 8,000 students registered.
- **15.060 Data Models and Decisions, MIT**, MBA course (core course), Fall 2014, Instructor
- **15.075 Statistical Thinking and Data Analysis, MIT**, undergraduate course, Fall 2009, Fall 2010, Fall 2011, Spring 2013, Instructor.
- **15.097 Prediction: Machine Learning and Statistics, MIT**, graduate course, Spring 2012, Instructor. Course materials available on MIT Open Courseware.
- **15.060 Data Models and Decisions, MIT** MBA course (core course), Fall 2012, Instructor
- **15.064 Probability and Statistics, MIT**, Summer 2010, Summer 2011. masters student course (Leaders for Global Operations Program), Co-Instructor, 2010 and 2011
- **COMS 4771 Machine Learning, Computer Science Department, Columbia University**, Spring 2008, lectures on regression, boosting, logistic regression, and ranking.
- **Math 103 Calculus, Princeton**, Fall 2002, Fall 2001, Instructor
 - My lectures were videotaped and placed online. I was the first instructor at Princeton in the sciences to have their lectures videotaped. Class average was over 10 percentage points higher than the average of the other sections on a shared final exam that was worth 50% of their grade; this class was the top scoring class, and it scored 5 percentage points above the second highest class.
- **Math 199 Math Alive, Princeton**, Fall 2003, Teaching Assistant, responsible for the cryptography section, taught by Dr. Ingrid Daubechies
- **Wavelets Course, Program for Women in Mathematics, Institute for Advanced Study**, Summer 2002, Teaching Assistant, taught by Dr. Ingrid Daubechies

- **Physics Classes, Buffalo Seminary Women's High School**, Substitute Teacher, part-time during winter and spring, 1999, taught physics classes daily to freshmen (conceptual physics) and seniors (physics and advanced physics).

Service to Duke

Provost's Forum Planning Committee, 2022-2023
 Founding faculty member, AI/Materials (aiM) graduate program at Duke 2020-present, graduate admissions committee, 2021-2022
 Duke CS Department Graduate Affairs Committee, 2020-2023
 Faculty search committees (CS, ECE, and Biostatistics), 2019-2022
 Tenure committee for Prof. David Page, Duke Biostatistics and Bioinformatics, 2020
 CS department strategic planning committee, 2019
 Tenure Committee for Prof. Kirsten Wickelgren, Duke Math, 2019
 Graduate admissions committee, Duke CS, 2019, 2020, 2021
 Working group member for white paper "Current State and Near-Term Priorities for AI-Enabled Diagnostic Support Software in Health Care," Duke-Margolis Center for Health Policy
 Lead organizer for the Machine Learning seminar, 2016-present
 Lead organizer of Triangle Machine Learning Day, 2018, 2019
 Chair of faculty search committee, Duke CS/ECE 2017-2018
 Bass connections reviewing, 2016
 Grad student admissions reviewing 2015-present
 Committee for a Prof. Katherine Heller's reappointment 2017-2018
 Committee of Guillermo Sapiro, Vince Conitzer, and I, appointed by Provost Kornbluth to write "Computing For Humanity," 2017
 Committee of 7 CS/ECE faculty members led by Carlo Tomasi to write a document similar to the above on AI, 2018
 Review committee for reappointment of Valerie Ashby, Dean of Trinity College of Arts and Sciences, Duke University, 2018
 CS graduate awards committee, 2018
 Reviewer for Data+ proposals, 2016-present
 Member of numerous RIP, prelim, PhD, MS thesis, and undergraduate honors thesis committees, 2017-2018
 Outreach such as hosting Duke Conversations, giving keynote for FEMMES at Duke, meeting with Visiting Committee, etc., 2018-present

Supervision

Postdocs

Dr. Aaron Fisher, co-advised with Francesca Dominici, Harvard, 2016-2019.
 Dr. Keivan Sadeghzadeh, MIT Sloan, 2016.
 Dr. Berk Ustun, Harvard CS, 2017-2020. (Now assistant professor at UCSD)
 Dr. Noor-E-Alam, MIT Sloan, 2014-2015. (Now assistant professor at Northeastern University)
 Dr. Ramin Moghaddass, MIT Sloan, 2013-2015 (Now tenured professor at University of Miami).
 Dr. Şeyda Ertekin, MIT Sloan, 2010-2014. (Now assistant professor at Middle East Technical University).

Graduate Students

MS student Saksham Jain, Duke ECE MS student, 2021-2022.
 PhD student Cristina Molero Del Rio, Visiting PhD student, Universidad de Sevilla, 2022-2023.
 PhD student Srikar Katta, Duke CS PhD student, 2022-present.
 PhD student Jon Donnelly, Duke CS PhD student, 2022-present.
 PhD student Sam Rosen, Duke statistics PhD student, 2022-present.
 PhD student Quinn Lanners, Duke Biostatistics PhD student, 2021-present.
 MS student Henry Ma, Duke statistics student, 2022-present.
 MD student Sully Chen, Duke Medical Center MD student, 2022-present.
 PhD student Kentaro Hoffman, UNC PhD student, 2019-2022.

MS student Pranay Jain, Duke CS MS student, 2021-2023.

PhD student Rui Zhang, Duke CS PhD student, 2021-present.

PhD student Stephen Hahn, Duke ECE PhD student, 2021-present.

PhD student Jiachang Liu, Duke ECE PhD student, 2021-present.

PhD student Zhicheng (Stark) Guo, Duke PhD ECE student, 2021-present.

MS student Xian (Jesse) Sun, Duke MS ECE student, 2020-2021.

MS student Vaishali Jain, Duke MS CS student, 2020-2021.

MS student Bin Han, Duke MS Statistics student, 2019-2020 (now PhD student at University of Washington)

MS student Neha Gupta, Duke MS Economics Computation student and then Duke Economics PhD student, 2019-present

PhD student Yingfan Wang (advised as UG, and then Duke PhD student), 2019-present

PhD Student Vittorio Orlandi, Duke Stats student, 2019-2023.

PhD Student Haiyang Huang, Duke CS student, 2019-present.

MS student Henry Yuren Zhang, Duke MS Statistics student 2019-2020.

MS/PhD student Chudi Zhong, Duke MS Statistics student and then Duke CS PhD student, 2019-present

MS student Jiali Xing, Duke Economics and CS student, 2019-2020.

MS student Matias Benitez Sr., Duke Economics and CS student, 2018-2019.

MS student Chunxiao Li, Duke MS Statistics student, 2018-2019.

MS student Weiyu Yan, Duke ECE student, 2018-2019.

PhD student Usaid Awan, Duke Economics PhD student, 2018-2020.

PhD student Zhi Chen, Duke CS PhD student, 2018-2023.

PhD student Jiayun Dong, Duke Economics PhD student, 2018-2019.

MS student Kangcheng Lin, Duke MS Statistics student, 2018-2019 (now at UIUC PhD program)

MS student Yang Bao, Duke Statistics student, 2018-2019.

MS student Sijia Wang, Duke ECE student, 2018-2019.

MS student Lei Chen, Duke ECE student, 2018-2020.

MS student Xiyang Hu, Duke Statistics student, 2018-2019 (now at CMU PhD program)

PhD Student Alina Barnett, Duke CS student, 2017-present.

PhD Student Harsh Parikh, Duke CS student, 2018-2023.

MS Student Yameng Liu, Duke Computer Science student, 2017-2019.

PhD Student Lesia Semenova, Duke CS student, 2016-present.

PhD Student Chaofan Chen, Duke CS student, 2016-2020 (now faculty at University of Maine)

PhD Student Tianyu Wang, Duke CS student, 2016-2021 (now faculty at Fudan University)

MS Student Beau Coker, Duke Statistics student, 2017-2018 (now at Harvard PhD program)

PhD Student Marco Morucci, Duke Political Science student, 2017-2021 (now postdoc at NYU)

PhD Student Hongyu Yang, MIT EECS student, 2014-2019.

MS Student Peter Alexander Lee, MIT ORC student, 2015-2016.

MS Student Prashan Wanigasekara, MIT EECS student, 2014-2016.

MS Student Christopher Choo, Engineering and Management, 2014-2015 (now at SUTD and Singapore Grand Prix)

PhD Student Vikas Garg, MIT EECS student, 2014-2016 (co-advised with Tommi Jaakola)

PhD Student Fulton Wang, MIT EECS student, 2013-2018. (Now at Sandia National Labs)

PhD Student Berk Ustun, MIT EECS student, 2012-2017 (Now faculty at UCSD)

PhD Student Stefano Tracà, MIT ORC student, 2012-2018 (now working at Disney Research)

PhD Student Siong Thye Goh, MIT ORC student, 2012-2018.

PhD Student Tong Wang, MIT EECS student, 2012-2016. (Now faculty at University of Iowa)

Project Student Ashia Wilson, MIT Sloan, 2012. (5 months before starting a PhD program at Berkeley)

PhD Student Theja Tulabandhula, MIT EECS student, 2010-2014. (Now senior lecturer at University of Sydney Business School)

PhD Student Ben Letham, MIT ORC student, 2010-2015. (Now at Facebook)

PhD Student Allison Chang, MIT ORC student, co-supervised with Dimitris Bertsimas, 2009-2012 (now at MIT Lincoln Labs).

Masters Student William Harris, MIT ORC Student, co-advised with Michael Ricard, 2014-2015 (now in the US military)

MS Student Oscar Moll, MIT CSAIL student, 2010-2011.

MS Graduate Research Assistant, Nandini Bhardwaj, Columbia & Con Edison Secondary Events Project, 2008.

Masters Project Course, Jawwad Sultan, Columbia & Con Edison Secondary Events Project, Fall 2007.

Summer Students, Supervision of 2 masters students and 1 undergraduate. Columbia & Con Edison Secondary Events Project, Summer 2007.

Undergraduate Students

Duke undergraduate, Chloe Zhu, 2023.

Duke undergraduate, Muhang (Tony) Tian, 2023.

Duke undergraduate, Danny Luo, 2022-2023.

Duke undergraduate, Boxuan Li, 2022-2023.

Duke undergraduate, Harry Chen, 2022-2023.

Duke undergraduate, Flora Shi, 2022-2023.

Duke undergraduate, Jenny Huang, 2022.

Duke undergraduate, Gaurav Parikh, 2022-2023.

Duke undergraduate, Michelle Qiu, 2022-2023.

Duke undergraduate, Anika Mitra, 2022-2023.

Duke undergraduate, Jessie (Yanchen) Ou, 2021-2022.

Duke undergraduate, Jerry Fang, 2021.

Duke undergraduate, Vaibhav Sharma, 2021-2022.

Duke undergraduate, Rui Xin, 2021-2023.

Duke undergraduate, Harsha Srijay, 2021.

Duke undergraduate, Vijit Singh, 2021.

Duke undergraduate, Caleb Kornfeld, 2021-2022.

Duke undergraduate, George Wang, 2021.

Duke undergraduate, Alexander Oesterling, 2021-2022.

Duke undergraduate, Haoning Jiang, 2021-2022.

Duke undergraduate, Lily Zhu, 2021-2023.

Duke undergraduate, Yunyao Zhu, 2021.

Duke undergraduate, Jerry Liu, 2020-2021.

Duke undergraduate, Nathan O'Hara, 2020.

Duke undergraduate, Krystal Hu, 2020.

Duke undergraduate, Angikar Ghosal, 2020-2021.

Duke undergraduate, Thomas Howell, 2020-2021.

Duke undergraduate, Edwin Agnew, 2020-2021.

Duke undergraduate, Benjamin Burnette, 2020-2021.

Duke undergraduate, Jordan Diamond, 2020-2021.

Duke undergraduate, Reed Chen, 2020.

Duke undergraduate, Kari Larson, 2020-2021.

Duke undergraduate, Brandon Zhao, 2019-2021.

Duke undergraduate, Alexander Rubin, 2019.

Duke undergraduate, Feroze Mohideen, 2019.

Duke undergraduate, Diane Hu, 2019-2020.

Duke undergraduate, Isaac Zhang, 2019-2021.

Duke undergraduate, Andre Wang, 2019-2021.

Duke undergraduate, Bhrij Patel, 2019-2021.
 Duke undergraduate, Jake Shulman, 2019.
 Duke undergraduate, Kenny Green, 2019.
 Duke undergraduate, Jerry Pan, 2018-2019.
 Duke undergraduate, Alexandru Damian, 2018-2020.
 Duke undergraduate, Nikhil Ravi, 2018-2020.
 Duke undergraduate, Sachit Menon, 2018-2020.
 Duke undergraduate, Chris Suh, 2018-2019.
 Duke undergraduate, McCourt Hu, 2018-2019.
 Duke undergraduate, Webster Bei, 2018-2020.
 Duke undergraduate, Jerry Chia Rui Chang, 2018.
 Duke undergraduate, Wilson Zhang, 2018-2019.
 Duke undergraduate, Divya Koyyalagunta, 2018-2019.
 Duke undergraduate, Anna Sun, 2018-present.
 Duke undergraduate, Peter Hase, 2018-2019.
 Duke undergraduate, Daniel Tau, 2017-2019.
 Duke undergraduate, Caroline Wang, 2017-2020.
 Duke undergraduate, Jerry Chia-Rui Chang, 2018.
 Duke undergraduate, Hao Liu, summer 2017.
 Duke undergraduate, Oscar Li, 2017-2019.
 MIT undergraduate, Chelsea Ge, summer 2014.
 MIT undergraduate, Jeffrey Chan, spring-fall 2014.
 MIT undergraduate, Jiaming Zeng, 2014-2015.
 MIT undergraduate, Shawn Qian, summer-fall 2012.
 Undergraduate exchange student, Yining Wang, spring 2013.
 PhD student at MIT and previously undergraduate from Arizona State University, Lydia Letham, summer 2012, summer 2014.
 MIT undergraduate project courses, three students (Kang Zhang, Arash Delijani, Kevin Pang) 2011-2012.
 Undergraduate Visiting Student from Ecole Centrale Paris (through MISTI), Fabrice Vegetti, 2012.
 Undergraduate Visiting Student from Ecole Centrale Paris (through MISTI), Adel Basli, 2011.
 Undergraduate project course on Collaborative Filtering, Association Rules and Information Retrieval, Eugene Kogan, Columbia University, co-supervision with Dr. Ansaf Salleb-Aouissi, Spring 2008.
 Undergraduate thesis advisement at Princeton, Krysta Svore, entitled "Multiscale Image Processing Using Single and Double Gaussian Techniques, and Hidden Markov Models," 2001-2002.

Thesis/Prelim Committees (not including my students)

PhD thesis proposal committee for Emory/Georgia Tech PhD student, Cheng Ding, 2023
 Prelim committee for Duke ECE PhD student, Fakrul Islam Tushar, 2023
 Qual committee for Duke MEMS PhD student, Jake Peloquin, 2022
 Qual committee for Duke ECE PhD student, Lin Duan, 2021
 Prelim and PhD defense committee for Duke CS student Qinwen Huang, 2021, 2023
 PhD defense committee for Duke ECE student Haibei Zhu, 2021
 Prelim and PhD defense committees for Duke ECE student Mojtaba Zarei, 2021, 2023
 Undergraduate honors thesis committee for Yunyao Zhu, 2021
 Prelim committee for Duke MEMS student Peiyi Chen, 2021
 Prelim and PhD Thesis committee for Duke Biomedical Engineering and Radiology PhD student Yinhao Ren, 2021, 2023
 Prelim & PhD thesis committee for Duke ECE PhD student Bohao Huang, 2019, 2020
 Prelim committee for Duke MEMS PhD student Bingyin Hu, 2020
 Prelim committee for Duke ECE PhD student Jiachang Liu, 2019

Qual committee for Duke ECE PhD student Kavinayan Sivakumar, 2019
 Prelim committee for Duke Statistics PhD student Filipe Ossa, 2019
 Prelim & PhD Thesis committee for Duke Economics Student Usaid Awan, 2019, 2022
 RIP committee for Duke ECE PhD student Jerry Wang, 2019
 Prelim committee for Duke Fuqua PhD student Shuyu Chen, 2019
 RIP & Prelim committee for Duke CS PhD student Shuai Yuan, 2019, 2020
 Qual, Prelim, & PhD committee for Duke ECE PhD student Yuting Ng, 2019, 2021, 2023
 Qual committee for Duke ECE PhD student Qian Huang, 2019
 Qual, Prelim, & PhD committees for Duke ECE PhD student Ghassen Jerfel, 2019, 2020, 2021
 Prelim & PhD thesis committee for Duke ECE PhD student Wanyi Fu, 2019, 2021
 Qualifying committee for Duke ECE PhD student Claire Lin, 2019
 Prelim committee for Duke CS PhD student Shuzhi Yu, 2018
 Prelim and PhD thesis committee for Duke CS PhD student Swarna Ravindran, 2018, 2023
 Prelim committee for PhD student Paidamoyo Chapfuwa, ECE PhD student, 2018
 RIP committee and prelim committee for Duke CS PhD student Andrew Lee, 2017, 2018
 Graduation with honors committee for Duke undergraduate Peter Hase, 2018
 Graduation with honors committee for Duke undergraduate Wuming Zhang, 2018
 Graduation with honors committee for Duke undergraduate Tianlin Duan, 2018
 RIP for Duke PhD student Xiaonan Hu, 2017, 2018
 Prelim committee for Duke PhD student Greg Spell, 2018
 RIP, Prelim and Thesis committees for Duke PhD student Rachel Draelos, 2017, 2019, 2021
 Prelim and thesis committees for Duke PhD student Stavros Sintos, 2017, 2020
 RIP and Prelim committee for Duke CS PhD student Zilong Tan, PhD in fall 2018
 Thesis committee for Stanford CS student Himabindu Lakkaraju, PhD in spring 2018
 RIP committee for Duke CS PhD student Xiaonan Hu, 2017
 RIP committee for Duke CS PhD student Shuzhi Yu, 2017
 RIP, prelim, PhD committees for Duke CS PhD student Abe Frandsen, 2018, 2019, 2022
 Thesis committee for Duke CS undergraduate with distinction Aditya Mukund, 2017
 Thesis committee for Duke CS masters student Guan-Wun Hao, 2017
 Thesis committee for Duke CS masters student Mona Prakash, 2016
 Thesis committee for Duke statistics masters student Emily Shao, 2017
 Thesis committee for Duke statistics masters student Sanjay Harihanan, 2017
 Thesis committee for Duke ECE PhD student Jordan Hashemi, 2017
 Thesis committee for Duke ECE PhD student Zhuoqing Chang, 2017, 2018, 2020
 Thesis reader for Harvard CS undergraduate Nicholas Larus-Stone, 2017
 Thesis committee for Duke PhD student Shan Shan, 2017
 Thesis committee for Duke PhD student Narayanan Rengaswamy, 2017.
 Thesis committee for MIT PhD student Yingxiang Yang, 2015.
 Thesis committee for MIT PhD student Been Kim, PhD in spring 2015.
 Thesis committee for MIT PhD student Anima Singh, PhD in spring 2015.
 Thesis committee for Pannaga Shivaswamy at Columbia University CS, PhD in spring 2009.

Society Memberships

- INFORMS
- International Machine Learning Society
- American Statistical Association (ASA)
- Institute of Mathematical Statistics (IMS)
- Association for the Advancement of Artificial Intelligence (AAAI)
- Association for Computing Machinery (ACM)
- American Association for the Advancement of Science (AAAS)
- Society for Causal Inference (SCI)