SURBHI GOEL

https://www.surbhigoel.com goel.surbhi@microsoft.com

EDUCATION

The University of Texas at Austin

August 2015 - June 2020

M.S. and Ph.D. in Computer Science

GPA: 4.0/4.0

Advisor: Adam R. Klivans

Dissertation: Towards Provably Efficient Algorithms for Learning Neural Networks

Committee: Alex Dimakis, Raghu Meka, Eric Price

Indian Institute of Technology, Delhi

B.Tech. in Computer Science and Engineering

July 2011 - May 2015

Major GPA: 9.55 / 10.0

APPOINTMENTS

Microsoft Research, New York, NY

July 2020 - Present

Postdoctoral Researcher, Machine Learning Group

Manager: Sham M. Kakade

Institute for Advanced Studies, Princeton, NJ

January - May 2020

Visiting Graduate Student, Theoretical Machine Learning Program

Simons Institute for Theory of Computing, Berkeley, CA

May - August 2019

Research Fellow, Foundations of Deep Learning Program

RESEARCH INTERESTS

My research is on the theoretical aspects of the modern practice of machine learning, where my goal is to develop the next generation of principled machine learning methods. In the pursuit of this goal, my work focuses on advancing the theoretical foundations of modern machine learning, particularly quantifying the computational and statistical aspects of deep learning methods, and expanding the toolbox of current algorithms using new theoretically grounded insights.

AWARDS AND FELLOWSHIPS

2020	Bert Kav	Dissertation .	Award fo	or PhD	dissertation

- 2019 Rising Stars in ML
- 2019 Rising Stars in EECS
- 2019 The University of Texas at Austin Graduate Dean's Prestigious Fellowship Supplement
- 2019 J.P. Morgan AI PhD Fellowship
- 2019 Simons-Berkeley Research Fellowship for Foundations of Deep Learning program
- 2018 The University of Texas at Austin Graduate Continuing Bruton Fellowship
- 2017 The University of Texas at Austin Graduate School Summer Fellowship
- 2015 ICIM Stay Ahead Award for Undergraduate Thesis
- 2015 Suresh Chandra Memorial Trust Award for Undergraduate Thesis
- 2011 Aditya Birla Scholarship awarded to 12 students from all over India
- 2011 OPJEM Scholarship awarded to 1 out of 850 students in the batch at IIT Delhi
- 2011 All India Rank 37 (Rank 2 in girls) in IITJEE among 450,000 students
- 2010 National Mathematics Olympiad finalist (1 out of 30 from all over India)

PUBLICATIONS

Authors are ordered alphabetically unless specified with *.

THESIS

Surbhi Goel

Towards Provably Efficient Algorithms for Learning Neural Networks The University of Texas at Austin, 2020 Received the Bert Kay dissertation award

CONFERENCE PAPERS

Jordan T. Ash*, **Surbhi Goel**, Akshay Krishnamurthy, Sham M. Kakade **Gone Fishing:** Neural Active Learning with Fisher Embeddings Neural Information Processing Systems (NeurIPS) 2021

Naman Agarwal, **Surbhi Goel**, Cyril Zhang **Acceleration via Fractal Learning Rate Schedules** International Conference on Machine Learning (ICML) 2021

Yuval Dagan, Constantinos Daskalakis, Nishanth Dikkala, **Surbhi Goel**, Anthimos Vardis Kandiros Statistical Estimation from Dependent Data
International Conference on Machine Learning (ICML) 2021

Surbhi Goel, Adam R. Klivans, Pasin Manurangsi, Daniel Reichman Tight Hardness Results for Learning One-Layer ReLU Networks Innovations in Theoretical Computer Science (ITCS) 2021

Surbhi Goel, Adam R. Klivans, Frederic Koehler From Boltzmann Machines to Neural Networks and Back Again Neural Information Processing Systems (NeurIPS) 2020

Surbhi Goel, Aravind Gollakota, Adam R., Klivans Statistical-Query Lower Bounds via Functional Gradients Neural Information Processing Systems (NeurIPS) 2020

Surbhi Goel, Aravind Gollakota, Zhihan Jin, Sushrut Karmalkar, Adam R. Klivans Superpolynomial Lower Bounds for Learning One-Layer Neural Networks using Gradient Descent

International Conference on Machine Learning (ICML) 2020

Omar Montasser*, **Surbhi Goel**, Ilias Diakonikolas, Nathan Srebro Efficiently Learning Adversarially Robust Halfspaces with Noise International Conference on Machine Learning (ICML) 2020

Jessica Hoffmann*, Soumya Basu, **Surbhi Goel**, Constantine Caramanis Learning Mixtures of Graphs from Epidemic Cascades International Conference on Machine Learning (ICML) 2020

Ilias Diakonikolas, **Surbhi Goel**, Sushrut Karmalkar, Adam R. Klivans, Mahdi Soltanolkotabi **Approximation Schemes for ReLU Regression**

Conference on Learning Theory (COLT) 2020

Surbhi Goel

Learning Ising and Potts Models with Latent Variables International Conference on Artificial Intelligence and Statistics (AISTATS) 2020

Surbhi Goel, Sushrut Karmalkar, Adam R. Klivans

Time/Accuracy Trade-offs for Learning a ReLU with respect to Gaussian Marginals Neural Information Processing Systems (NeurIPS) 2019 Selected for a spotlight presentation

Surbhi Goel, Daniel Kane, Adam R. Klivans Learning Ising Models with Independent Failures Conference on Learning Theory (COLT) 2019

Surbhi Goel, Adam R. Klivans

Learning Neural Networks with Two Nonlinear Layers in Polynomial Time Conference on Learning Theory (COLT) 2019

Surbhi Goel, Adam R. Klivans, Raghu Meka Learning One Convolutional Layer with Overlapping Patches International Conference on Machine Learning (ICML) 2018 Selected for a full oral presentation

Surbhi Goel, Adam R. Klivans

Eigenvalue Decay Implies Polynomial-Time Learnability for Neural Networks Neural Information Processing Systems (NeurIPS) 2017

Surbhi Goel, Varun Kanade, Adam R. Klivans, Justin Thaler Reliably Learning ReLU in Polynomial Time Conference on Learning Theory (COLT) 2017

WORKSHOP PAPERS

Jessica Hoffmann*, Soumya Basu, **Surbhi Goel**, Constantine Caramanis Disentangling Mixtures of Epidemics on Graphs Graph Representation Learning, Neural Information Processing Systems (NeurIPS) 2019

Surbhi Goel, Adam R. Klivans

Learning Depth-Three Neural Networks in Polynomial Time
Deep Learning: Bridging Theory and Practice, Neural Information Processing Systems (NeurIPS) 2017

Surbhi Goel, Varun Kanade, Adam R. Klivans, Justin Thaler Reliably Learning ReLU in Polynomial Time Optimization for Machine Learning (OPT), Neural Information Processing Systems (NeurIPS) 2016 Selected for an oral presentation

PREPRINTS

Benjamin L. Edelman, **Surbhi Goel**, Sham M. Kakade, Cyril Zhang Inductive Biases and Variable Creation in Self-Attention Mechanisms In submission, 2021

Jordan T. Ash, **Surbhi Goel**, Akshay Krishnamurthy, Dipendra Misra Investigating the Role of Negatives in Contrastive Representation Learning In submission, 2021.

Jordan T. Ash*, Cyril Zhang, **Surbhi Goel**, Akshay Krishnamurthy, Sham M. Kakade **Anti-Concentrated Confidence Bonuses For Scalable Exploration**In submission, 2021

Surbhi Goel, Rina Panigrahy

Learning Two layer Networks with Multinomial Activation and High Thresholds Manuscript, 2019

Matthew Jordan*, Naren Manoj, **Surbhi Goel**, Alexandros Dimakis Quantifying Perceptual Distortion of Adversarial Examples Manuscript, 2019

Simon Du, Surbhi Goel

Improved Learning of One-hidden-layer Convolutional Neural Networks with Overlaps Manuscript, 2018.

INVITED TALKS

What Functions do Self-attention Blocks Prefer to Represent?

2021

Learning Theory Workshop at Google [virtual]

Computational Barriers For Learning Some Generalized Linear Models

2021

Rigorous Evidence for Information-Computation Trade-offs Workshop at Simons Institute [virtual]

Computational Complexity of ReLU Regression

2021

The Multifaceted Complexity of Machine Learning Workshop at IMSI [virtual]

Computational Complexity of Learning Neural Networks over Gaussian Marginals 2020-21

MIC Seminar at NYU [virtual]

Algorithms Seminar at Duke University [virtual]

ML Theory Seminar at Harvard University [virtual]

ARC Colloquium at Georgia Tech [virtual]

IDEAL Seminar at TTIC [virtual]

TOC Colloquium at MIT [virtual]

SILO Seminar at UW-Madison [virtual]

Statistics Seminar at Stanford University [virtual]

Learning Ising and Potts Models with Latent Variables

2020

International Conference on Artificial Intelligence and Statistics (AISTATS) [virtual]

Approximation Schemes for ReLU Regression

2020

Conference on Learning Theory (COLT) [virtual]

Deep Learning Program Reunion at Simons Institute [virtual]

Provably Efficient Algorithms for Learning Neural Networks

2020

Microsoft Research New York

Microsoft Research New England

Microsoft Research Redmond

Time/Accuracy Tradeoffs for Learning a ReLU wrt Ga Spotlight Talk at Neural Information Processing Systems (Neur	_
Exploring Surrogate Losses for Learning Neural Network TTIC Young Researcher Seminar Series	orks 2019
Efficiently Learning Simple Neural Networks Rising Star in ML Talk at University of Maryland Institute for	2019 r Advanced Computer Studies
Learning Ising Models with Independent Failures Conference on Learning Theory (COLT) Research Fellows Talk at Simons Institute	2019
Learning Neural Networks with Two Nonlinear Layers Conference on Learning Theory (COLT)	s in Polynomial Time 2019
Efficiently Learning Simple Convolutional Networks China Theory Week	2018
Learning One Convolutional Layer with Overlapping P Google Research Theory Reading Group International Conference on Machine Learning (ICML)	Patches 2018
	2016-17
Reliably Learning the ReLU in Polynomial Time Oral at OPT-ML Workshop at Neural Information Processing INTERNSHIPS	
Oral at OPT-ML Workshop at Neural Information Processing INTERNSHIPS Google, Mountain View CA	Systems (NeurIPS) May - August 2018
Oral at OPT-ML Workshop at Neural Information Processing INTERNSHIPS Google, Mountain View CA Research Intern Dell, Round Rock TX	Systems (NeurIPS) May - August 2018 Supervisor: Rina Panigrahy
Oral at OPT-ML Workshop at Neural Information Processing INTERNSHIPS Google, Mountain View CA Research Intern Dell, Round Rock TX Research Intern Google, New York, NY	May - August 2018 Supervisor: Rina Panigrahy June - August 2017 May - August 2016
Oral at OPT-ML Workshop at Neural Information Processing INTERNSHIPS Google, Mountain View CA Research Intern Dell, Round Rock TX Research Intern Google, New York, NY Research Intern Google, Mountain View CA	May - August 2018 Supervisor: Rina Panigrahy June - August 2017 May - August 2016 Supervisor: Natalia Ponomareva May - August 2014
INTERNSHIPS Google, Mountain View CA Research Intern Dell, Round Rock TX Research Intern Google, New York, NY Research Intern Google, Mountain View CA Software Engineering Intern University of Michigan, Ann Arbor MI	May - August 2018 Supervisor: Rina Panigrahy June - August 2017 May - August 2016 Supervisor: Natalia Ponomareva May - August 2014 Supervisor: Neha Jha May - July 2013
INTERNSHIPS Google, Mountain View CA Research Intern Dell, Round Rock TX Research Intern Google, New York, NY Research Intern Google, Mountain View CA Software Engineering Intern University of Michigan, Ann Arbor MI Research Scholar	May - August 2018 Supervisor: Rina Panigrahy June - August 2017 May - August 2016 Supervisor: Natalia Ponomareva May - August 2014 Supervisor: Neha Jha May - July 2013

Indian Institute of Technology Delhi

Course: Data Structures

 $\begin{array}{c} \text{Spring } 2015 \\ \textit{Teaching Assistant} \end{array}$

OUTREACH

Co-founder 2020-Present Learning Theory Alliance (LeT-All) Co-organizing the Graduate Applications Support Program Co-organized the COLT 2021 Mentoring Workshop Co-organized the ALT 2021 Mentoring Workshop Mentor 2021-Present Women in Machine Learning Theory (WiML-T) Mentoring Program **Panelist** 2021 VMware Nirman for Women in Tech Mentor 2018-19 UT Austin's Women in CS (GWC-WiCS) Mentoring Program SERVICE ROLES Virtual Experience Chair 2021 Conference on Learning Theory (COLT) Co-organized the virtual part of the hybrid conference, including the 2-day virtual-only program Co-organizer 2020-Present One World Machine Learning Seminar Series Treasurer 2016-17Graduate Representative Association of Computer Sciences (GRACS) **Program Committee** International Conference on Algorithmic Learning Theory (ALT) 2021/22 Conference on Learning Theory (COLT) 2021 Reviewing Journal of Machine Learning Research 2021 Symposium on Theory of Computing (STOC) 2019/20/21 Neural Information Processing Systems (NeurIPS) (top 30%) 2018/20/21 Conference on Learning Theory (COLT) 2018/19/20

International Conference on Learning Representations (ICLR) 2019/20

IEEE Transactions on Information Theory 2020

Symposium on Discrete Algorithms (SODA) 2020

Foundations of Computer Science (FOCS) 2020

International Conference on Machine Learning (ICML) 2019 (top 5%)