SURBHI GOEL

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EDUCATION

University of Texas at Austin

August 2015 - June 2020

M.S. and PhD in Computer Science

Advisor: Adam Klivans

Thesis: Towards Provably Efficient Algorithms for Learning Neural Networks

Indian Institute of Technology, Delhi

July 2011 - May 2015

Bachelor of Technology

Department of Computer Science and Engineering

WORK EXPERIENCE

Microsoft Research, New York NY

July 2020 - Present

 $\cdot \ Postdoctoral \ Researcher$

IAS, Princeton NJ

January - May 2020

· Visiting Graduate Student

Simons Institute for Theory of Computing, Berkeley CA

May - August 2019

· Research Fellow

Google, Mountain View CA

May - August 2018

· Research Intern

Supervisor: Rina Panigrahy

Dell, Round Rock TX

June - August 2017

· Research Intern

Google, New York, NY

May - August 2016

· Research Intern

Supervisor: Natalia Ponomareva

Google, Mountain View CA
Software Engineering Intern

May - August 2014 Supervisor: Neha Jha

University of Michigan, Ann Arbor MI

May - July 2013

· Research Scholar

Supervisor: Atul Prakash

RESEARCH INTERESTS

Theory, Machine Learning

PUBLICATIONS

All papers are in alphabetical ordering unless specified with *.

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

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

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

Surbhi Goel, Aravind Gollakota, Zhihan Jin, Sushrut Karmalkar, Adam 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. *Disentangling Mixtures of Epidemics on Graphs*. Short Version: Graph Representation Learning Workshop at Neural Information Processing Systems (NeurIPS) 2019. Full version: International Conference on Machine Learning (ICML) 2020.

Ilias Diakonikolas, **Surbhi Goel**, Sushrut Karmalkar, Adam 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 Klivans. *Time/Accuracy Trade-offs for Learning a ReLU with respect to Gaussian Marginals*. Neural Information Processing Systems (NeurIPS) 2019 [Spotlight Presentation].

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

Surbhi Goel, Adam Klivans. Learning Neural Networks with Two Nonlinear Layers in Polynomial Time. Short version: Deep Learning Bridging Theory and Practice Workshop, Neural Information Processing Systems (NeurIPS) 2017. Full version: Conference on Learning Theory (COLT) 2019.

Surbhi Goel, Adam Klivans, Raghu Meka. Learning One Convolutional Layer with Overlapping Patches. International Conference on Machine Learning (ICML) 2018 [Full Oral].

Surbhi Goel, Adam Klivans. Eigenvalue Decay Implies Polynomial-Time Learnability for Neural Networks. Neural Information Processing Systems (NeurIPS) 2017.

Surbhi Goel, Varun Kanade, Adam Klivans, Justin Thaler. Reliably Learning ReLU in Polynomial Time. Short Version: Workshop on Optimization for Machine Learning, Neural Information Processing Systems (NeurIPS) 2016 [Oral Presentation]. Full Version: Conference on Learning Theory (COLT) 2017.

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

Jordan T. Ash*, **Surbhi Goel**, Akshay Krishnamurthy, Sham Kakade. *Gone Fishing: Neural Active Learning with Fisher Embeddings*.

OTHER MANUSCRIPTS

Surbhi Goel, Rina Panigrahy. Learning Two layer Networks with Multinomial Activation and High Thresholds.

Simon Du, Surbhi Goel. Improved Learning of One-hidden-layer Convolutional Neural Networks with Overlaps.

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

TEACHING EXPERIENCE

University of Texas at Austin

Spring 2016

· Course: Distributed Computing (Hons.)

Teaching Assistant

University of Texas at Austin

Fall 2015

· Course: Data Structures

Teaching Assistant

Indian Institute of Technology Delhi

Spring 2015

· Course: Data Structures

Teaching Assistant

TALKS

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 Gaussian Marginals · Neural Information Processing Systems (NeurIPS) 2019	2019
Exploring Surrogate Losses for Learning Neural Networks · TTIC Young Researcher Seminar Series	2019
Efficiently Learning Simple Neural Networks · Rising Star in ML Talk at University of Maryland Institute for Advanced Computer Studies	2019
Learning Neural Networks with Two Nonlinear Layers in Polynomial Time \cdot Conference on Learning Theory (COLT)	2019
Learning Ising Models with Independent Failures Conference on Learning Theory (COLT) Research Fellows Talk at Simons Institute	
Efficiently Learning Simple Convolutional Networks · China Theory Week	2018
Learning One Convolutional Layer with Overlapping Patches Google Research Theory Reading Group International Conference on Machine Learning (ICML)	2018
Conference on Learning Theory (COLT) · OPT-ML Workshop at Neural Information Processing Systems (NeurIPS) PROGRAM COMMITTEE 2021/22 Algorithmic Learning Theory (ALT)	
REVIEWING	
2021 Journal of Machine Learning Research 2020 IEEE Transactions on Information Theory 2020 Symposium on Discrete Algorithms (SODA) 2020 Foundations of Computer Science (FOCS) 2019 International Conference on Machine Learning (ICML) (top 5%) 2019/20 International Conference on Learning Representations (ICLR) 2019/20 Symposium on Theory of Computing (STOC) 2018/19/20/21 Conference on Learning Theory (COLT) 2018/20/21 Neural Information Processing Systems (NeurIPS) (top 30%) SERVICE	
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2021 Conference on Learning Theory (COLT) 2021 Virtual Experience Cha 2020-Present WiML-T Mentoring Program Mentor 2021 COLT 2021 Mentorship Workshop Co-organizer 2020-Present One World Machine Learning Seminar Series Co-organizer 2021-Present Learning Theory Alliance (LeT-All) Co-founder 2021 ALT 2021 Learning Theory Mentorship Workshop Co-organizer	41

ACHIEVEMENTS

2020	Bert Kay Dissertation Award
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
2016-19	Professional Development Award for travel to conferences
2015	ICIM Stay Ahead Award for Undergraduate Thesis
2015	Suresh Chandra Memorial Trust Award for Undergraduate Thesis
2011-15	Aditya Birla Scholarship awarded to 12 students from all over India
2011	OPJEM Scholarship awarded to 1 out of 850 students in the batch
2011	All India Rank 37 (second in girls) in IITJEE among 450,000 students
2010-11	National Mathematics Olympiad finalist, IMO training camp attendee
2010-11	KVPY Fellowship awarded to 250 from all over India pursuing science