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

July 2011 - May 2015

B.Tech. in Computer Science and Engineering

Major GPA: 9.55 / 10.0

APPOINTMENTS

University of Pennsylvania, Philadelphia, PA

January 2023 (expected)

Magerman Term Assistant Professor, Computer and Information Science

Microsoft Research, New York, NY

July 2020 - November 2022 (expected)

Postdoctoral Researcher, Machine Learning Group

Manager: Sham M. Kakade

Institute for Advanced Study, 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 with an emphasis on 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 Kay	Dissertation	Award for	best	dissertation	in	CS	at	UT	Austin
------	----------	--------------	-----------	------	--------------	----	----	---------------------	----	--------

- 2019 Rising Stars in ML by University of Maryland
- 2019 Rising Stars in EECS by UIUC
- 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

* indicates α - β (alphabetical) ordering.

THESES

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

Benjamin L. Edelman*, **Surbhi Goel***, Sham M. Kakade*, Cyril Zhang * Inductive Biases and Variable Creation in Self-Attention Mechanisms International Conference on Machine Learning (ICML) 2022

Nikunj Saunshi, Jordan T. Ash, **Surbhi Goel**, Dipendra Misra, Cyril Zhang, Sanjeev Arora, Sham M. Kakade, Akshay Krishnamurthy

Understanding Contrastive Learning Requires Incorporating Inductive Biases International Conference on Machine Learning (ICML) 2022

Jordan T. Ash, Cyril Zhang, **Surbhi Goel**, Akshay Krishnamurthy, Sham M. Kakade **Anti-Concentrated Confidence Bonuses For Scalable Exploration** International Conference on Learning Representations (ICLR) 2022

Jordan T. Ash*, **Surbhi Goel***, Akshay Krishnamurthy*, Dipendra Misra* Investigating the Role of Negatives in Contrastive Representation Learning International Conference on Artificial Intelligence and Statistics (AISTATS) 2022

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

Anthimos-Vardis Kandiros, Yuval Dagan, Nishanth Dikkala, **Surbhi Goel**, Constantinos Daskalakis **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

WORKING PAPERS

Surbhi Goel*, Sham M. Kakade*, Adam T. Kalai*, Cyril Zhang * RCNNs Learn Succinct Learning Algorithms in Polynomial Time In submission, 2022

Boaz Barak*, Benjamin L. Edelman*, **Surbhi Goel***, Sham M. Kakade*, Eran Malach*, Cyril Zhang* **Hidden Progress in Deep Learning: SGD Learns Parities Near the Computational Limit** In submission, 2022

UNPUBLISHED PAPERS

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

The Hidden Progress Behind Loss Curves Workshop on Learning: Optimization and Stochastics at EPFL	July 2022
Demystifying Attention-based Architectures in Deep Learning	
Workshop on Algorithms for Learning and Economics in Naxos, Greece	June~2022
Pricipled Algorithm Design in the Era of Deep Learning	
CS/CSE Colloquium at NYU Courant/Tandon	$April\ 2022$
CS Colloquium at UW-Madison	March~2022
CS Colloquium at Halicioglu Data Science Institute, UCSD	March~2022
$CS\ Colloquium\ at\ UMD$	February 2022
SCS Talk at CMU	February 2022
CS Colloquium at Duke	February 2022
CIS Colloquium at UPenn	February 2022
CS Colloquium at Cornell University	February 2022
Talks at TTIC	February 2022

What Functions do Self-attention Blocks Prefer to Represent? ML Symposium at USC	December 2021
ELLIS Talk Series at IST Austria	December 2021
Learning Theory Workshop at Google	October 2021
Learning Theory Workshop at Google	0 000001 2021
Computational Barriers For Learning Some Generalized Linear Models Information-Computation Trade-offs Workshop at Simons Institute [video][slides]	September 2021
Computational Complexity of ReLU Regression The Multifaceted Complexity of Machine Learning Workshop at IMSI [video]	$April\ 2021$
Computational Complexity of Learning Neural Networks over Gaussian	_
MIC Seminar at NYU	May 2020
Algorithms Seminar at Duke University	October 2020
ML Theory Seminar at Harvard University [video]	October 2020
ARC Colloquium at Georgia Tech	November 2020
IDEAL Seminar at TTIC	November 2020
TOC Colloquium at MIT	December 2020
SILO Seminar at UW-Madison	January 2020
Statistics Seminar at Stanford University	July 2021
Learning Ising and Potts Models with Latent Variables International Conference on Artificial Intelligence and Statistics (AISTATS)	August 2020
Thiermational Conference on Artificial Intelligence and Statistics (AISTATS)	August 2020
Approximation Schemes for ReLU Regression	
Conference on Learning Theory (COLT) [video]	July 2020
Deep Learning Program Reunion at Simons Institute	August~2020
Provably Efficient Algorithms for Learning Neural Networks	
Microsoft Research New York	February 2020
Microsoft Research New England	February 2020
Microsoft Research Redmond	February 2020
Time/Accuracy Tradeoffs for Learning a ReLU wrt Gaussian Marginals	-
Spotlight Talk at Neural Information Processing Systems (NeurIPS)	December 2019
Exploring Surrogate Losses for Learning Neural Networks TTIC Young Researcher Seminar Series	December 2019
Efficiently Learning Simple Neural Networks	
Rising Star in ML Talk at University of Maryland	$September\ 2019$
Taranian Trian Madala with Indones dant Pailance	
Learning Ising Models with Independent Failures	I.J. 0010
Conference on Learning Theory (COLT) Research Fellows Talk at Simons Institute	July 2019 July 2019
	<i>y</i> •
Learning Neural Networks with Two Nonlinear Layers in Polynomial Tir Conference on Learning Theory (COLT)	me July 2019
Efficiently Learning Simple Convolutional Naturalis	
Efficiently Learning Simple Convolutional Networks China Theory Week at Tringhya University	Santambar 2010
China Theory Week at Tsinghua University	September 2019

Learning One Convolutional Layer with Overlapping Patches

Google Research Theory Reading Group June 2018 July 2018

International Conference on Machine Learning (ICML)

Reliably Learning the ReLU in Polynomial Time

Conference on Learning Theory (COLT) July 2017

December 2016

OPT-ML Workshop at Neural Information Processing Systems (NeurIPS)

WORK EXPERIENCE

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 May - August 2014

Software Engineering Intern Supervisor: Neha Jha

University of Michigan, Ann Arbor MI May - July 2013

 $Supervisor:\ Atul\ Prakash$ Research Scholar

TEACHING EXPERIENCE

University of Texas at Austin Spring 2018

Course: Data Mining (Hons.) Guest Lecture

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

OUTREACH

Co-founder 2020-Present

Learning Theory Alliance (LeT-All)

Co-organized the COLT 2022 Mentoring Panel

Co-organized the ALT 2022 Mentoring Workshop

Co-organized 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

Pa	ne	lis	st
\mathbf{r}	ше	:11:	วเ

New Horizons in Theoretical Computer ScienceJune 2022VMware Nirman for Women in TechJanuary 2021

Mentor 2018-19

UT Austin's Women in CS (GWC-WiCS) Mentoring Program

SERVICE ROLES

Virtual Experience Chair

Conference on Learning Theory (COLT)

Co-organized the virtual part of the hybrid conference, including the 2-day virtual-only program

Co-organizer 2020-2021

2021

One World Machine Learning Seminar Series

Treasurer 2016-17

Graduate Representative Association of Computer Sciences (GRACS)

Program Committee

International Conference on Algorithmic Learning Theory (ALT)	2021/22
Conference on Learning Theory (COLT)	2021/22

Conference Reviewing

Symposium on Theory of Computing (STOC)	2019/20/21
Neural Information Processing Systems (NeurIPS)	$2018\ (top\ 30\%)/20/21$
Conference on Learning Theory (COLT)	2018/19/20
International Conference on Learning Representations (ICLR)	2019/20
Symposium on Discrete Algorithms (SODA)	2020
Foundations of Computer Science (FOCS)	2020/22
International Conference on Machine Learning (ICML)	2019 (top 5%)

Journal Reviewing

Journal of Machine Learning Research	2021
IEEE Transactions on Information Theory	2020