

Jonathan W. Siegel

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EDUCATION

University of California at Los Angeles 2013-2018
Ph.D. in Mathematics Los Angeles, CA
Advisor: Russel E. Caflish
Thesis: “Accelerated First-Order Optimization with Orthogonality Constraints”

University of California at Santa Cruz 2009-2013
B.Sc. (Honors) in Mathematics Santa Cruz, CA

ACADEMIC APPOINTMENTS

Assistant Professor 2022-present
Texas A&M University College Station, TX

Assistant Research Professor 2021-2022
Pennsylvania State University University Park, PA

Postdoctoral Scholar 2018-2021
Pennsylvania State University University Park, PA

RESEARCH PAPERS

Published

- The Extended Regularized Dual Averaging Method for Composite Optimization. *to appear in Journal of Computational Mathematics*. With Jinchao Xu.
Preprint available at: <https://arxiv.org/abs/1904.02316>
- Sharp Bounds on the Approximation Rates, Metric Entropy, and n -widths of Shallow Neural Networks. *Foundations of Computational Mathematics* 1-57, 2022. With Jinchao Xu.
Preprint available at: <https://arxiv.org/abs/2101.12365>
- Characterization of the Variation Spaces Corresponding to Shallow Neural Networks. *to appear in Constructive Approximation*. With Jinchao Xu.
Preprint available at: <https://arxiv.org/abs/2106.15002>
- Uniform Approximation Rates and Metric Entropy of Shallow Neural Networks. *Research in the*

Mathematical Sciences 9.3:1-21, 2022. With Limin Ma and Jinchao Xu.

- Optimal Convergence Rates for the Orthogonal Greedy Algorithm. *IEEE Transactions on Information Theory* 68.5:3354-3361, 2022. With Jinchao Xu.
- Extensible Structure-Informed Prediction of Formation Energy with Improved Accuracy and Usability employing Neural Networks. *Computational Materials Science* 208:111254, 2021. With Adam Krajewski, Zi-Kui Liu, and Jinchao Xu.
- High-Order Approximation Rates for Shallow Neural Networks with Cosine and ReLU^k Activation Functions. *Computational and Applied Harmonic Analysis* 58:1-26, 2022. With Jinchao Xu.
- Approximation rates for neural networks with general activation functions. *Neural Networks* 128:313-321, 2020. With Jinchao Xu.
- Accuracy, Efficiency and Optimization of Signal Fragmentation. *Multiscale Modeling and Simulation* 18(2):737-757, 2020. With Russel Caflisch and Hung Hsu Chou
- Accelerated Optimization with Orthogonality Constraints. *Journal of Computational Mathematics* 39(2):207-226, 2020.
- Compact Support of L^1 Penalized Variational Problems. *Communications in Mathematical Sciences* 15(6):1771-1790, 2017. With Omer Tekin.

After First Revision

- Greedy Training Algorithms for Neural Networks and Applications to PDEs. *First Revision submitted to Journal of Computational Physics*, 2022. With Qingguo Hong, Xianlin Jin, Wenrui Hao, and Jinchao Xu.
Preprint available at: <https://arxiv.org/abs/2107.04466>

Under Review

- Sharp Lower Bounds on Interpolation by Deep ReLU Neural Networks at Irregularly Spaced Data. *Submitted to Machine Learning*, 2023.
Preprint available at: <https://arxiv.org/abs/2302.00834>
- Optimal Approximation Rates for Deep ReLU Neural Networks on Sobolev Spaces. *Submitted to Journal of Machine Learning Research*, 2023.
Preprint available at: <https://arxiv.org/abs/2211.14400>
- Training Sparse Neural Networks using Compressed Sensing. *Submitted to IEEE Transactions on Neural Networks and Learning Systems*, 2021. With Jianhong Chen, Pengchuan Zhang and Jinchao Xu.
Preprint available at: <https://arxiv.org/abs/2008.09661>
- Accelerated First-Order Methods: Differential Equations and Lyapunov Functions. *Submitted to Journal of Computational Mathematics*, 2021.
Preprint available at: <https://arxiv.org/abs/1903.05671>

- On the Activation Function Dependence of the Spectral Bias of Neural Networks. *Submitted to ICLR*, 2023. With Qingguo Hong, Qingyang Tan and Jinchao Xu

GRANTS

Penn State Institute for CyberScience Seed Grant (co-PI) 2018-2019
 “Deep Learning for CALPHAD Database Development and Uncertainty Quantification” \$35,000

NSF DMS-2111387 (co-PI) 2021-2024
 “Comparative Study of Finite Element and Neural Networks Discretizations for Partial Differential Equations” \$550,000

NSF DMS-2216799 (PI) 2022
 “US Participation at the Twenty-sixth International Domain Decomposition Conference” \$15,000

NSF CCF-2205004 (co-PI) 2022-2025
 “CIF: Small: Interpretable Machine Learning based on Deep Neural Networks: A Source Coding Perspective” \$600,000

TEACHING EXPERIENCE

Texas A&M University Fall 2022
Instructor College Station, TX
 Math 308H - Honors Differential Equations

Pennsylvania State University Fall 2021
Instructor University Park, PA
 Math 141 - Integral Calculus

Pennsylvania State University Spring 2021
Co-Instructor University Park, PA
 Math 555 - Optimization Theory (Graduate Course)

Pennsylvania State University Fall 2020
Instructor University Park, PA
 Math 140 - Differential Calculus

Pennsylvania State University Spring 2020
Instructor University Park, PA
 Math 251 - Differential Equations
 Math 251H - Honors Differential Equations

Pennsylvania State University Fall 2019
Instructor University Park, PA
 Math 141 - Integral Calculus

Pennsylvania State University/Peking University
Co-Instructor
Math 497 - Introduction to Deep Learning

Summer 2019
Beijing, China

Pennsylvania State University
Instructor
Math 141 - Integral Calculus

Fall 2018
University Park, PA

University of California, Los Angeles
Teaching Assistant
Math 32B - Integral Vector Calculus
Math 32A - Differential Vector Calculus
Math 110B - Finite Group Theory

2014-2017
Los Angeles, CA

INVITED SEMINAR AND CONFERENCE TALKS

SUNY Albany
Data Science Seminar

January 23, 2023

Brown University
Crunch Seminar on Scientific Computing

December 23, 2022

Texas A&M University
TAMIDS Seminar

December 8, 2022

University of Oslo
Scientific and Machine Learning Seminar

December 1, 2022

King Abdullah University of Science and Technology (KAUST)
Mathematics and Computational Science Seminar

November 22, 2022

King Abdullah University of Science and Technology (KAUST)
Conference on Scientific Computing and Machine Learning

November 15, 2022

Texas State University
SyDATA Symposium

September 30, 2022

Texas A&M University
CAMDA Seminar

August 31, 2022

Czech Technical University in Prague
27th International Conference on Domain Decomposition Methods, Invited Plenary Talk

July 26, 2022

Princeton University
Wilks Seminar

May 16, 2022

Georgia Institute of Technology <i>Applied and Computational Mathematics (ACM) Seminar</i>	April 4, 2022
Illinois Institute of Technology <i>Mathematics Department Colloquium</i>	January 21, 2022
Texas A&M University <i>Mathematics Department Colloquium</i>	December 6, 2021
University of South Carolina <i>Mathematics Department Colloquium</i>	November 29, 2021
Rensselaer Polytechnic Institute <i>Mathematics in Imaging, Data and Optimization Seminar</i>	October 6, 2021
RWTH Aachen <i>Applied Mathematics Group Lunch Seminar</i>	October 4, 2021
ETH Zurich <i>FoMICS Seminar Talk and Lecture</i>	June 2, 2021
University of Texas, Austin <i>Applied Mathematics Seminar</i>	May 21, 2021
University of California, San Diego <i>CCoM Seminar</i>	May 11, 2021
Purdue University <i>Mathematical Data Science Webinar</i>	May 10, 2021
University of Notre Dame <i>ACMS Applied Mathematics Seminar</i>	April 15, 2021
University of California, Irvine <i>Computational Mathematics Seminar</i>	March 15, 2021
California Institute of Technology <i>CMX (Computational Mathematics) Seminar</i>	February 17, 2021
Pennsylvania State University <i>CCMA Workshop on Mathematical Machine Learning and Applications</i>	December 15, 2020

INVITED MINI-SYMPOSIUM TALKS

SIAM Conference on the Mathematics of Data Science

September 27, 2022

Recent Advances in Machine Learning and Optimization

SIAM Conference on Uncertainty Quantification

April 12, 2022

Recent Advances in Machine Learning and Data-Driven Methods for Physical Sciences and Engineering

AMS Fall Western Sectional Meeting

October 23, 2021

Special Session on Theoretical and Applied perspectives in Machine Learning

SIAM Conference on Analysis of PDEs

March 16, 2021

Mathematics of Machine Learning Methods for PDEs

Kunming, China

August 15, 2019

International Multigrid Conference (IMG)

SERVICE

Students Co-Advised

- Jianhong Chen, Penn State Graduate Student, 2019-2020
- Xianlin Jin, Peking University Graduate student, 2021-present

Conferences Co-Organized

- CCMA Workshop on Mathematical Machine Learning and Applications, December 14-16, 2020.

Seminars Co-Organized

- Computational and Applied Mathematics (CAM) Colloquium at Penn State, Fall 2020-Fall 2021

Ad-Hoc Reviewer for

- Mathematical Programming, Neural Networks, AISTATS, Numerical Algorithms, Calculus of Variations and Partial Differential Equations, Expert Systems with Applications, IEEE Transactions on Neural Networks and Learning Systems, SIAM Journal on Optimization, Journal of Machine Learning Research, SIAM Journal on Mathematics of Data Science, SIAM Journal on Numerical Analysis

AWARDS AND SCHOLARSHIPS

University of California, Los Angeles

2018

- Pacific Journal of Mathematics Dissertation Award

University of California, Los Angeles

2013-2014

- University of California Regents Fellow

University of California, Santa Cruz 2012
• Stephen M. Palais Award

University of California, Santa Cruz 2011
• Putnam Mathematical Competition Honorable Mention

AFFILIATIONS

American Mathematical Association (AMS) 2021-present
Society of Industrial and Applied Mathematicians (SIAM) 2021-present

INDUSTRY EXPERIENCE

Google Intern June 6, 2016-August 26, 2016
Mountainview, CA

I worked with the Network Architecture team on improving the efficiency of a Monte Carlo network reliability simulation. Specifically, I implemented importance sampling, which reduced the number of samples required by a factor of 3.

Google Intern June 12, 2017-September 1, 2017
Los Angeles, CA

I worked with the Budgetplanner Team (a division working on advertisement). I built a data processing pipeline that collected and processed data which was scattered across multiple relational databases to create training data for a machine learning model. Then I used TensorFlow to design and test multiple machine learning models on the resulting dataset.

TECHNICAL STRENGTHS

Programming Skills: C/C++, Java, Matlab, Latex, Python