

Sam Foreman

PHYSICS · ML / DATA SCIENCE

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Education

PhD. Physics
B.S. Engineering Physics
B.S. Applied Mathematics

University of Iowa
University of Illinois at Urbana-Champaign
University of Illinois at Urbana-Champaign

Dissertation

Learning Better Physics: A Machine Learning Approach to Lattice Gauge Theory

University of Iowa

SUPERVISOR: YANNICK MEURICE

Spring 2016 – Summer 2019

- Developed new technique for applying renormalization group transformations to arbitrary 2D images.
- Designed and implemented a new algorithmic technique for Markov Chain Monte Carlo simulations that is able to generate gauge configurations.
- Carried out interdisciplinary research focused on applying ideas from machine learning and data science to simulations in high-energy physics.
- Discovered a new method for describing the phase transition in the 2-dimensional Ising model by applying unsupervised learning techniques (PCA, k-means clustering) to Monte Carlo simulation images.
- Helped to create a new technique for implementing renormalization group transformations on arbitrary image sets, and explored potential applications in dynamic image analysis and action recognition.
- Worked with Tensorflow/Keras to construct convolutional neural networks capable of classifying configurations of the Ising model by temperature.
- Current work focuses on improving the efficiency of the Hybrid Monte Carlo algorithm by using neural networks to improve the quality of the sampler. These improvements have wide applications across a variety of industries.

Experience

Argonne National Laboratory

Computational Sciences Division

GRADUATE RESEARCH FELLOW

Summer 2018 — Summer 2019

- Software development focused on applying machine learning models to help improve the efficiency of Hybrid Monte Carlo simulations and their use in Lattice QCD.
- Built and deployed learning models using Tensorflow/Keras on some of the world's fastest supercomputers using state-of-the-art high-performance computing techniques.
- Developed a method for training Markov Chain Monte Carlo kernels parameterized with deep neural networks that shows promise in outperforming traditional methods on a variety of different models.

University of Iowa

Department of Physics & Astronomy

RESEARCH ASSISTANT

Spring 2016 — Fall 2016

- Software and hardware development for HaloSat, a nanosatellite built with the goal of better understanding the missing baryon problem.
- Implemented a variety of in-flight optimization algorithms aimed at maximizing the incoming X-ray signals (by minimizing background noise) while in operation.

University of Illinois

Center for Complex Systems Research

RESEARCH ASSISTANT

Spring 2011 — Spring 2015

- Actively maintained the legacy code base (C++ / MATLAB) for our research group and was in charge of quality analysis of new contributions.
- Constructed a model capable of describing the energy density and self-discharge time of nanoscale capacitors.
- This work was submitted as a patent (pending) titled "Energy Storage in Quantum Resonators", on which I was designated a co-inventor together with my advisor Alfred Hübler.

Publications & Talks

- S. Foreman, X.J. Jin, and J. Osborn, "**Machine Learning and Neural Networks for Field Theory**"
- S. Foreman, Y. Meurice, J. Giedt and J. Unmuth-Yockey, "**Examples of renormalization group transformations for image sets**" *Physical Review E*.
- S. Foreman, "**Machine learning inspired analysis of the Ising model transition**" The 36th Annual International Symposium on Lattice Field Theory
- S. Foreman, J. Giedt, Y. Meurice and J. Unmuth-Yockey, "**RG inspired Machine Learning for lattice field theory.**" *arXiv:1710.02079*
- S. Foreman, "**Machine Learning Analysis of Ising Worms.**" Brookhaven National Laboratory, Dec. 2017 (invited speaker)
- A. Hubler, S. Foreman, J. Liu, and L. Wortsman, "**Large Energy Density in Three-Plate Nanocapacitors due to Coulomb Blockade**" *Journal of Applied Physics*