

Education \_\_\_\_\_

PhD. Physics University of Iowa

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

**B.S. Engineering Physics, B.S. Applied Mathematics** 

Univ. of Illinois at Urbana-Champaign

**Energy Storage in Quantum Resonators** 

2010—2015

2015-2019

# Experience\_

### **Argonne National Laboratory**

Leadership Computing Facility (ALCF)

2019 — Current

Postdoctoral Research AssociateData science group

- Responsible for building, testing, and documenting current ML frameworks and libraries on modern HPC infrastructures
- Presentation on **Hyperparameter Search Using DeepHyper on Theta** @ ALCF: Simulation, Data, and Learning Workshop
- Presentation on Scaling Deep Learning Applications (recording) @ 2021 Computational Performance Workshop
- Lead and contribute to discussions in our journal club, most recently presented on **Switch Transformers**
- Designed and implemented novel network architectures to improve current sampling techniques for Lattice QCD
- Extensive development (sole author) on <a href="Lighter: 124">124</a>mc-qcd, a python library for distributed, data-parallel training, tracking, and testing of different generative models for sampling in lattice gauge models
  - Shown to provide ~100 × speedup compared to current methods for simple models

# **Argonne National Laboratory**

**Computational Sciences Division** 

GRADUATE RESEARCH FELLOW

*2018 — 2019* 

- Recipient of the DOE Office of Science Graduate Student Research Program
- Software development focused on applying machine learning models to help improve the efficiency of Hybrid Monte Carlo simulations
- · Developed techniques for efficiently scaling ML models on next generation HPC infrastructures

## **University of Iowa**

Department of Physics & Astronomy

RESEARCH ASSISTANT

*2016 — 2017* 

- · Software and hardware development for HaloSat, a nanosatellite built to help the missing baryon problem
- · Implemented a variety of in-flight optimization algorithms to maximize the signal-to-noise ratio while in operation
- Helped to write and test telemetry system using VERILOG and XILINX.

### **University of Illinois**

Center for Complex Systems Research

RESEARCH ASSISTANT

SISTANT 2011 — 2015

- Actively maintained the legacy code base (C++ / MATLAB) for our research group and was in charge of quality analysis of new contributions
- Co-inventor on a patent **✓**(pending) titled "Energy Storage in Quantum Resonators"
- · Constructed a model capable of describing the energy density and self-discharge time of nanoscale capacitors

# Publications & Talks

- S. Foreman, invited talk on Accelerated Sampling Methods for Lattice Gauge Theory at BNL-HET & RBRC Joint Workshop "DWQ @ 25", Dec 2021
- S. Foreman, invited talk on Training Topological Samplers for Lattice Gauge Theories at ML for HEP, on and off the Lattice ECT\*-Trento, Sep 2021
- S. Foreman, X.Y. Jin, & J.C. Osborn, LeapFrogLayers: A Trainable Framework for Effective Topological Sampling, Lattice, 2021
- S. Foreman, L. Jin, X.Y. Jin, A. Tomiya, J.C. Osborn, & T. Izubuchi, HMC with Normalizing Flows, 38th Intl. Symposium on Lattice Field Theory, 2021
- S. Foreman, invited talk on 12hmc-qcd at the MIT Lattice Group Seminar, 2021
- S. Foreman, invited talk on Deep Learning HMC for Improved Gauge Generation to the Machine Learning Techniques in Lattice QCD Workshop, 2021
- S. Foreman, X.Y. Jin & J.C. Osborn, Deep Learning Hamiltonian Monte Carlo, SimDL Workshop ICLR, 2021
- S. Foreman, X.Y. Jin, & J.C Osborn, Machine Learning and Neural Networks for Field Theory SnowMass, 2020
- S. Foreman, invited talk on Machine Learning for Lattice QCD at the University of Iowa, 2020
- S. Foreman, contributed talk Machine learning inspired analysis of the Ising model transition to 36th Intl. Symposium on Lattice Field Theory, 2018
- S. Foreman, Y. Meurice, J. Giedt & J. Unmuth-Yockey, Examples of renormalization group transformations for image sets Physical Review E., 2018
- S. Foreman, invited talk on Machine Learning Analysis of Ising Worms at Brookhaven National Laboratory, 2017
- S. Foreman, J. Giedt, Y. Meurice & J. Unmuth-Yockey, RG inspired Machine Learning for lattice field theory arXiv:1710.02079, 2017
- A. Hubler, S. Foreman, J. Liu, & L. Wortsmann, Large Energy Density in Three-Plate Nanocapacitors due to Coulomb Blockade J. Appl. Phys, 2018

