

Da Long

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EDUCATION	The University of Utah, Salt Lake City, Utah Ph.D. student in Computer Science, GPA: 3.89, 2021 - Present • Advisor: Shandian Zhe • Expected: May 2026 The University of Arizona, Tucson, Arizona B.S. in Computer Science, GPA: 4.0, 2019 - 2021 B.S. in Mathematics, GPA: 4.0, 2019 - 2021
WORK EXPERIENCE	Meta, Menlo Park, CA Research Scientist Intern, May 2024 - Aug. 2024 • Integrated reinforcement learning algorithms, including Deep Q-Network (DQN) and Advantage Actor-Critic (A2C), into Meta's recommendation foundation model. Lawrence Berkeley National Laboratory, Berkeley, CA Student Researcher, Aug. 2024 - Dec. 2024 • Designed a hierarchical spatio-temporal Fourier transformer (StFT) for long-term forecasting, improving prediction stability and accuracy with meaningful uncertainty quantification.
SKILLS	Technical: Python, PyTorch, JAX, MATLAB, LaTeX
RESEARCH EXPERIENCE	Spatio-temporal Fourier Transformer for Long-term Dynamics Prediction • Designed a hierarchical Fourier transformer for multi-scale and multi-physics long-term spatio-temporal forecasting, with a generative residual refiner for further improvement and uncertainty quantification. Arbitrarily-Conditioned Multi-Functional Diffusion for Multi-Physics Emulation • Proposed a flexible diffusion model framework for multi-physics systems to simulate diverse physical processes, addressing arbitrary conditional tasks using a single, unified model. Solving Forward and Inverse Problems via an Invertible Fourier Neural Operator • Invented an invertible Fourier neural operator to solve both PDE forward and inverse problems. Toward Efficient Kernel-Based Solvers for Nonlinear PDEs • Developed a kernel learning framework that efficiently and effectively solves nonlinear, high-dimensional PDEs. Learning High-frequent and Multi-scale Solutions via Gaussian Process • Solving high-frequent and multi-scale PDEs by selecting and learning high-frequent components through a spectral mixture kernel. Gaussian Process for Solving ODE/PDEs • Developed a Gaussian process framework to solve ODE/PDEs while quantifying uncertainties in solutions through variational inference.
RESEARCH INTERESTS	Spatio-temporal Forecasting, Diffusion Model for Operator Learning, Foundation Model for Physics, Scientific Large Language Model, Surrogate Modeling, Gaussian Process, Reinforcement Learning
PUBLICATIONS	* indicates equal contribution. • Long D., Xu Z., Yuan Q., Yang Y., & Zhe S., Invertible Fourier Neural Operators for Tackling Both Forward and Inverse Problems. In <i>International Conference on Artificial Intelligence and Statistics (AISTATS 2025)</i>.

- **Long D.**, Xing W., Krishnapriyan A., Kirby R., Zhe S., & Mahoney M., Equation Discovery with Bayesian Spike-and-Slab Priors and Efficient Kernels. In *International Conference on Artificial Intelligence and Statistics (AISTATS 2024)*.
- Fang S.*, Cooley M.*, **Long D.***, Li S., Kirby R., & Zhe S., Solving High Frequency and Multi-Scale PDEs with Gaussian Processes. In *International Conference on Learning Representations (ICLR 2024)*.
- **Long D.**, Mrvaljevic N., Zhe S., & Hosseini B., A Kernel Approach for PDE Discovery and Operator Learning. In *Physica D: Nonlinear Phenomena*.
- **Long D.**, Wang Z., Krishnapriyan A., Kirby R., Zhe S., & Mahoney M. (2022). AutoIP: A United Framework to Integrate Physics into Gaussian Processes. In *International Conference on Machine Learning (ICML 2022)*.

PAPERS IN
SUBMISSION

- **Long D.**, Zhe S., Williams S., Olier L., & Bai Z., Spatio-temporal Fourier Transformer (StFT) for Long-term Dynamics Prediction.
- **Long D.**, Xu Z., Yang G., Narayan A., & Zhe S., Arbitrarily-Conditioned Multi-Functional Diffusion for Multi-Physics Emulation.
- Xu Z.*, **Long D.***, Xu Y., Yang G., Zhe S., & Owhadi H., Toward Efficient Kernel-Based Solvers for Nonlinear PDEs.
- Li Y., Chen K., **Long D.**, Xu Z., Xing W., Hochhalter J., & Zhe S., Pseudo Physics-Informed Neural Operators.