

# Da Long

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| EDUCATION            | <p><b>The University of Utah</b>, Salt Lake City, Utah<br/>Ph.D. student in Computer Science, GPA: 3.88, Expected May 2026<br/>• Advisor: <a href="#">Shandian Zhe</a></p> <p><b>The University of Arizona</b>, Tucson, Arizona<br/>B.S. in Computer Science, B.S. in Mathematics, GPA: 4.0, Dec. 2020</p>   |
| RESEARCH INTERESTS   | <p><b>Probabilistic Learning:</b> Bayesian Modeling, Uncertainty Quantification, Approximate Inference, Gaussian Process</p> <p><b>AI for Scientific Discovery:</b> Surrogate Modeling, Physics Informed Machine Learning, Operator Learning, Equation Discovery</p>   |
| SKILLS               | <p><b>Technical:</b> Python, Matlab, C, Java, Pytorch, Jax, LaTeX</p>  |
| RESEARCH EXPERIENCE  | <p><b>Gaussian Process for ODE/PDEs Discovery via Spike-and-slab</b></p> <ul style="list-style-type: none"><li>• Succeeded in discovering underlying equations in noisy and sparse datasets using a probabilistic approach while state-of-the-art methods failed</li></ul> <p><b>Kernel Method for Operator Learning</b></p> <ul style="list-style-type: none"><li>• Outperformed leading methods such as FNO and DeepONet in noisy and sparse datasets</li></ul> <p><b>Learning High-frequent and Multi-scale Solutions via Gaussian Process</b></p> <ul style="list-style-type: none"><li>• Selected and learned high-frequent components through a spectral mixture kernel</li></ul> <p><b>Gaussian Process for Solving ODE/PDEs</b></p> <ul style="list-style-type: none"><li>• Developed a Gaussian process framework to solve ODE/PDEs and quantified solution uncertainties through variational inference</li></ul> <p><b>Physics Informed Neural Networks for Learning High-frequent and Multi-scale Solutions</b></p> <ul style="list-style-type: none"><li>• Enhanced physics informed neural networks (PINNs) with Fourier bases, achieved best accuracy compared to state-of-the-art methods</li></ul> |
| ACADEMIC SERVICES    | <p><b>Conference Reviewer</b></p> <ul style="list-style-type: none"><li>• AISTATS 2023, ICML 2022</li></ul>  |
| TEACHING EXPERIENCE  | <p><b>The University of Utah</b></p> <p>Teaching Mentorships</p> <ul style="list-style-type: none"><li>• CS 6350 Machine Learning (Fall 2022)</li><li>• CS 6190 Probabilistic Machine Learning (Spring 2023)</li></ul>   |
| PUBLICATIONS         | <ul style="list-style-type: none"><li>• <b>Long D.</b>, Wang Z., Krishnapriyan A., Kirby R., Zhe S., &amp; Mahoney M. (2022). AutoIP: A United Framework to Integrate Physics into Gaussian Processes. In <i>International Conference on Machine Learning (ICML 2022)</i>.</li></ul>   |
| PAPERS IN SUBMISSION | <ul style="list-style-type: none"><li>• <b>Long D.</b>, Xing W., Krishnapriyan A., Kirby R., Zhe S., &amp; Mahoney M., Equation Discovery with Bayesian Spike-and-Slab Priors and Efficient Kernels.</li><li>• <b>Long D.</b>, Mrvaljevic N., Zhe S., &amp; Hosseini B., A Kernel Approach for PDE Discovery and Operator Learning.</li><li>• Fang S., Cooley M., <b>Long D.</b>, Li S., Kirby R., &amp; Zhe S., Solving High Frequency and Multi-Scale PDEs with Gaussian Processes.</li><li>• Cooley M., <b>Long D.</b>, Kirby R., &amp; Zhe S., Fourier PINNs: From Strong Boundary Conditions to Adaptive Fourier Bases.</li></ul>   |