VU-ANH LE

Greater Chicago Area • levuanh@mit.edu Google Scholar • ORCID • LinkedIn

PROFESSIONAL INTERESTS & SKILLS

Professional Interests: Algorithms, Optimization, Operations Research

Programming Languages and Software: Python, C++, Java, MATLAB, Haskell, Lisp, Coq, Isabelle, Lean, Git

EDUCATION

Beloit College
Bachelor of Science, Mathematics, GPA: 3.74/4.00

Beloit, Wisconsin Aug 2021 - May 2025

Relevant Coursework: Mathematical Statistics, Differential Equations, Complex Analysis, Topology, Algorithm Design and Analysis, Data Mining

ARTICLES AND PREPRINTS

- 1. Le, Vu Anh and Dik, Mehmet, "Topology-Preserving Scaling in Data Augmentation," in arXiv, Nov 2024
- **2. Le, Vu Anh** and Dik, Mehmet, "The Stability of Persistence Diagrams Under Non-Uniform Scaling," in *arXiv*, Nov 2024
- **3. Le, Vu Anh**, and Dik, Mehmet, "How Analysis Can Teach Us the Optimal Way to Design Neural Operators," in *Proceedings of International Mathematical Sciences*, Nov 2024
- **4. Le, Vu Anh**, and Dik, Mehmet, "A Mathematical Analysis of Neural Operator Behaviors," in *arXiv*, Oct 2024. (Accepted as a chapter to the book issue Advances in Quantum Calculus and Functional Analysis, CRC Press, Taylor & Francis Group)

RESEARCH EXPERIENCE

Google Research
Student Researcher, B.S.

Aug 2024 - Present

Remote

- Research Advisors: Jake Garrison (Google Research) and Prof. Mehmet Dik (Beloit College).
- Developed a mathematical framework to analyze the behaviors of neural operators, focusing on these aspects:
 - Stability: Established bounds for neural operators using Lipschitz continuity conditions.
 - Convergence: Proved exponential convergence via the Banach Fixed Point Theorem.
 - Clustering Behavior: Analyzed long-term solution dynamics through gradient flow interpretation.
 - Universality: Extended the Universal Approximation Theorem and Stone-Weierstrass Theorem
 to demonstrate the approximation capabilities of neural operators.
- Provided theoretical guarantees on stability, exponential convergence, and generalization. Detailed results published in the following papers 1 and 2.
- Applied the proposed framework in designing a case study model for solving complex partial differential equations. Compared with the state-of-the-art works e.g. DeepONet, it requires 15% fewer epochs.

Massachusetts Institute of Technology

Intern, MIT Summer Research Program - General

Cambridge, Massachusetts

June 2024 - Aug 2024

- Research Advisor: Prof. Haruko Murakami Wainwright.
- Integrated two machine learning models, namely Random Forest and Bidirectional LSTM, into the Python-based computational chemistry library PyLEnM to monitor the behavior of toxic analytes.
- Achieved 97.7% accuracy in predicting the time taken for analyte concentration to drop to safety levels. Outperformed baseline models e.g. linear regression and univariate LSTMs with accuracies between 70–85%.