

# Thien Le

550 Memorial Drive, #16A1  
Cambridge, MA 02139  
USA

Email: thienle@mit.edu  
Github: steven-le-thien  
Linkedin: thienle2

## EDUCATION

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**Massachusetts Institute of Technology (MIT), EECS-CSAIL Jegelka Group** Sep 2019 - May 2024  
*Master of Science (SM) and Doctor of Philosophy (PhD) in EECS*  
GPA: 5.0/5.0  
Master thesis: Training invariance and the low-rank phenomenon: beyond linear networks

**University of Illinois at Urbana-Champaign (UIUC)** Sep 2016 - May 2019  
*Bachelor of Science in Mathematics & Computer Science*  
GPA: 3.93/4.00  
Awarded 'Most Outstanding Undergraduate Major Award in Mathematics & CS'

## RESEARCH INTERESTS

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Deep Learning Theory. Optimization. Computational Biology. Inference on Biological Structure.

## PROJECTS (MIT)

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*Graduate researcher under Stefanie Jegelka*

**Learning with complex (non-manifold) biological structure** Sep 2019 - Ongoing

- Studying the algebraic geometry and combinatorics of tree space (a non-manifold moduli space)
- Implemented a new algorithm in Haskell for tree inference based on continuous optimization of that space
- Gave a seminar talk at VinAI research day

**Bounds for reproducing kernel Hilbert space (RKHS) norms** Apr 2020 - Apr 2021

- Studied the Fourier-Stieltjes transform on LCA groups and Young's inequality for convolutional integrals
- Proved an upper bound for RKHS norm of product of functions beyond the Gaussian kernel with application to generalization bounds and optimization algorithm for kernel ridge regression.

**Implicit regularization of deep ReLU neural network** Sep 2020 - Oct 2021

- Generalizing theoretical results on margin and low rank bias in linear network to deep ReLU architectures
- Established a family of invariants during late-stage training of said architectures under gradient flow/descent
- Publish to ICLR2022 (Spotlight)
- Extend framework to characterize implicit bias of mean field limit for deep neural networks (beyond NTK regime)

**Learning theory for invariant/equivariant concept classes** Jul 2022 - Ongoing

- Study learning theory of deep neural network architectures designed to be invariant/equivariant under certain group actions such as CNN, DeepSets, SignNets, etc
- Show learnability of DeepSets under correlated statistical queries (CSQ) model via classical symmetric functions
- Under preparation for to NeurIPS 2023

**Transferability of graph neural network via graph limits (graphops)** October 2022-Ongoing

- Study limits, approximation and transferability of finite graph neural networks via theory of graph limit
- Improve on results for graphons to include graph sequences that are not dense (grid, stars,...)
- Submitted to COLT 2023

## PROJECTS (UIUC)

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**Statistical Phylogeny Gene Tree Estimation** Jun 2018 - Ongoing

- Implementing a scalable phylogeny estimation algorithm with strong statistical guarantees
- Implemented parallelism and algorithmic optimization (e.g. LCA in  $O(1)$  query and  $O(n)$  preprocessing)
- Maintaining 5000 lines of C code at `constraint_inc`, tested with BlueWaters supercomputer
- Three papers published in conference and journal, in submission to Nature (2022)

## TEACHING AND INTEREST

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- Appointed course staff for both core algorithm classes in UIUC CS Department
- Reviewer for ICLR, AISTATS, ICML, NeurIPS, ECML-KPDD.
- Graduate math courses @MIT: Stochastic Calc., Geom. of Manifolds, AG I, Intro to Representation Thr.

## TECHNICAL SKILLS

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Programming: C, C++, Python, Pytorch, R, Haskell