

Qiaobo Li

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Graduation year: May 2027

EDUCATION

University of Illinois at Urbana-Champaign <i>Siebel School of Computing and Data Science</i>	08/2022–present
• Ph.D., Major in Computer Science, Overall GPA (till now): 3.95/4	
• Programming & Software: Python, Pytorch, C, C++, MATLAB, R; Latex	
• Research interests: Federated Learning, Trustworthy Machine Learning, Deep Learning Theory, High-dimensional statistics.	
<i>Main Course and Score</i>	
Deep Learning Theory	4/4
Adv Topics in Sec, Priv and ML	4/4
Deep Generative Models	4/4
Trustworthy ML	4/4
Adv Topics in NLP	4/4
Deep Learning with Graphs	4/4
PEKING UNIVERSITY	09/2018–06/2022
<i>School of Mathematical Sciences</i>	
• Undergraduate, Major in Statistics and Probability, Overall GPA: 3.79/4 ; Major GPA: 3.80/4	
• Programming & Software: C, C++, Python, MATLAB, R; Latex	
• English Proficiency: TOEFL 110, GRE Verbal 157, Quantitative 170, Writing 3.5	
<i>Awards and Honors</i>	
Academic Excellence Award, Peking University	11/2020&11/2021
Annual Scholarship, Peking University	11/2020&11/2021
3 rd Prize in Chinese Mathematical Olympiad	01/2018
2 nd Prize in Chinese Mathematics Competition	11/2019

RESEARCH EXPERIENCE

Understanding Learning/Generalization Landscape for GD on Multi-task/Single-task Learning	07/2021-08/2023
<i>Supervised by Prof. Quanquan Gu, Department of Computer Science, UCLA</i>	
Trustworthy Machine Learning	02/2023-08/2023
<i>Supervised by Prof. Han Zhao, Siebel School of Computing and Data Science, UIUC</i>	
Differential Privacy and Convergence Analysis in Federated Learning	09/2023-present
<i>Supervised by Prof. Arindam Banerjee, Siebel School of Computing and Data Science, UIUC</i>	

WORK EXPERIENCE

Machine Learning Engineer Intern	05/2025-08/2025
<i>Responsible Recommendation System, TikTok Inc.</i>	
Key Contributions:	
• Select state-of-the-art vision–language models and appropriate fine-tuning methods and aim to achieve acceptable accuracy on the given datasets.	
• Explore strategies for imbalanced datasets, such as combining class-specific fine-tuned models and rebalancing the dataset through resampling or reweighting.	
• Compare and analyze how different frame-extraction strategies affect video classification performance of vision–language models.	
• Design and study a router that dispatches video inputs between larger and smaller vision–language models to balance accuracy and computational cost.	
Recognition: Exceeds Expectations (E) performance rating.	

PUBLICATIONS

- **Sketched Gaussian Mechanism for Private Federated Learning**
Qiaobo Li, Zhijie Chen, Arindam Banerjee
NeurIPS Spotlight, 2025

- **Sketched Adaptive Federated Deep Learning: A Sharp Convergence Analysis**
Zhijie Chen, Qiaobo Li, Arindam Banerjee
 NeurIPS, 2025
- **On the Power of Multitask Representation Learning with Gradient Descent**
Qiaobo Li, Zixiang Chen, Yihe Deng, Yiwén Kou, Yuan Cao, Quanquan Gu
 AISTATS, 2025
- **Loss Gradient Gaussian Width based Generalization and Optimization Guarantees**
Arindam Banerjee, Qiaobo Li, Yingxue Zhou
 AISTATS Oral, 2025
- **Differentially Private Post-Processing for Fair Regression**
Ruicheng Xian, Qiaobo Li, Gautam Kamath, Han Zhao
 ICML, 2024
- **Sketching for Distributed Deep Learning: A Sharper Analysis**
Mayank Shrivastava, Berivan Isik, Qiaobo Li, Sanmi Koyejo, Arindam Banerjee
 NeurIPS, 2024
- **Beyond Johnson-Lindenstrauss: Uniform Bounds for Sketched Bilinear Forms**
Rohan Deb, Qiaobo Li, Mayank Shrivastava, Arindam Banerjee
 ArXiv Preprint, 2025
- **Sketched Gaussian Mechanism on Matrix for Private Federated LoRA**
Qiaobo Li, Zhijie Chen, Arindam Banerjee
 Under Review, 2025

Coding Projects

- **Federated Learning with Communication-Efficient and Private Algorithms**
 - **Objective:** Advance federated learning by developing a novel algorithm and benchmarking it against state-of-the-art communication-efficient and differentially private methods for both vision and language tasks.
 - **Key Contributions:**
 - Designed and implemented different communication-efficient and private federated learning algorithms, leveraging sketching techniques to enhance communication efficiency.
 - Conducted rigorous evaluations of the algorithm alongside **FetchSGD** and **1bit-ADAM**, benchmarking performance on vision and language tasks:
 - Trained **ResNet101** and fine-tuned **ViT-based model** on **CIFAR-10** for vision tasks.
 - Fine-tuned **BERT** on **SST2** from the GLUE benchmark for sentiment analysis.
 - Investigated spectral properties of optimization dynamics to gain insights into model optimization behavior during training and fine-tuning by analyzing **loss Hessian** and **predictor Hessian spectrum structures** using the **SLQ algorithm**, revealing trends in curvature and stability.
 - Demonstrated improvements in **accuracy**, **convergence speed**, and **resource utilization**, while maintaining competitive communication overhead across diverse server optimizers, including **SGD**, **SGD with momentum**, **ADAM**, and **AdaClip**
- **Fair and Private Post-Processing for Regressors**
 - **Objective:** Developed and conducted a fair and private post-processing algorithm to process regressors on the **Communities & Crime** and **Law School** datasets.
 - **Key Contributions:**
 - Investigated and quantified trade-offs among **test error**, **fairness violation**, and **differential privacy** by implementing **differential privacy mechanisms** and fairness-aware evaluation metrics.
 - Enhanced model performance while balancing **privacy guarantees** and **fairness constraints** through systematic analysis of the interplay between privacy loss (ϵ), fairness measures, and predictive accuracy.
- **Loss Gradient Gaussian Width: Generalization and Optimization Guarantees**
 - **Objective:** Analyze generalization and optimization trends during training by investigating gradient and featurizer properties across models and datasets.
 - **Key Contributions:**
 - Trained **ResNet18**, **Feedforward Networks (FFN)**, and **CNN** on **CIFAR-10**, **MNIST**, and **Fashion-MNIST** using **SGD**, **ADAM**, and **ADAMW**.
 - Conducted comprehensive gradient analyses, identifying trends in **Gradient Domination Ratio**, **L2 norm of gradients**, and **L0/L1 norms of featurizers**, and examined gradient coordinate structures to understand their relationship to generalization performance.
 - Assessed the interplay between model architectures, optimizers, and dataset characteristics, providing insights into how these factors influence optimization dynamics and generalization behavior.