Gan Luo

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Research Interest

- **Optimization**: Federated Learning, Distributed Optimization, SGD Analysis
- **Machine Learning and Deep Learning**: Efficient LLM, Machine Learning System, Deep Learning Theory

Education

- **Peking University** — **Beijing, China**
  B.A. in Mathematics, School of Mathematical Sciences — *Sep. 2022–Jul. 2026 (Expected)*
  - **Mathematics Courses**: Mathematical Analysis, Advanced Algebra, Complex Analysis, Probability Theory, Mathematical Statistics, Stochastic Processes, Stochastic Analysis, Basic Numerical Method, Basic Optimization Method.
  - **Computer Science Courses**: Basic Artificial Intelligence and Deep Learning, Parallel and Distributed Computing, Introduction to Computer Vision, Introduction to Multi-model(auditor).
  - **Honors and Awards**: Applied Mathematics Elite Program (The program accepted only 15 people this year)

Research Experience

- **Convergence and Speedup Analysis of Distributed Optimization Algorithms** — **Peking University**
  Undergraduate Research Assistant — *Nov. 2023 – Present*
  - **Advisor**: Kun Yuan
  - **Project 1: Analysis of Push-Pull Algorithm**
    - **Push-Pull Algorithm Provably Achieves Linear Speedup Over Arbitrary Network Topologies**
      Liyuan Liang*, Gan Luo*, Kun Yuan (*Equal contribution) [coming soon to Arxiv]
    - **Currently finalizing this paper for submission to SIOPT.**
    - Conducted research on the Push-Pull Algorithm, focusing on convergence and linear speedup properties in non-convex and stochastic settings on arbitrary topology.
    - First to prove convergence and linear speedup properties of the Push-Pull Algorithm under non-convex and stochastic settings on arbitrary topology.
    - Validated the proposed theoretical results by conducting distributed optimization numerical experiments on the MNIST and CIFAR10 datasets.
  - **Project 2: Analysis on Decentralized Optimization over Row-stochastic Networks**
    - **Achieving Linear Speedup and Optimal Complexity for Decentralized Optimization over Row-stochastic Networks**
      Liyuan Liang*, Xinyi Chen*, Gan Luo*, Kun Yuan (*Equal contribution) [Arxiv]
    - **Acceped to ICML2025, Spotlight**
    - Introduced effective metrics to capture the influence of row-stochastic mixing matrices
    - Established the first convergence lower bound for decentralized learning over row-stochastic networks
    - Incorporated a multi-step gossip (MG) protocol, to attain the lower bound, achieving optimal complexity.
    - Proposed a novel analysis framework demonstrating that PULL-DIAG-GT achieves linear speedup, which is the first such result for row-stochastic decentralized optimization.
    - Conducted numerical experiments to validate theoretical results.
- **Online Scheduling on LLM Inference** — **Massachusetts Institute of Technology**
  Undergraduate Research Assistant — *Oct. 2024 – Present*
  - **Advisor**: David Simchi-Levi
  - **Project 1: Online Batching Algorithm on LLM Inference**
    - **Optimizing LLM Inference: Fluid-Guided Online Scheduling with Memory Constraints**
      (α-β) Ruicheng Ao*, Gan Luo*, David Simchi-Levi, Xinshang Wang, available on [SSRN] and [Arxiv]
    - **This paper is submitted to Operations Research.**
    - Formulated the LLM inference as a multi-stage online scheduling task with stochastic queueable requests.
    - Proposed a novel online batching algorithm for LLM inference.
    - Proved that the algorithm achieves near-optimal throughput while controlling latency and Time to First Token (TTFT).
    - Conducted numerical experiments on synthetic and real-world datasets with Llama-7B on A100 GPU to validate theoretical results.