

## 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

Nov. 2023 - Present

- Undergraduate Research Assistant

   Advisor: Kun Yuan
  - o Project 1: Analysis of Push-Pull Algorithm
    - \* Push-Pull Algorithm Provably Achieves Linear Speedup Over Arbitrary Network Topologies Liyuan Liang\*, <u>Gan Luo</u>\*, Kun Yuan (\*Equal contribution) [Arxiv]
    - \* This paper is submitted to 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.
  - o 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\*, <u>Gan Luo</u>\*, 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

Oct. 2024 - Present

- Undergraduate Research Assistant

  o Advisor: David Simchi-Levi
  - o Project 1: Online Batching Algorithm on LLM Inference
    - \* Optimizing LLM Inference: Fluid-Guided Online Scheduling with Memory Constraints  $(\alpha-\beta)$  Ruicheng Ao\*, <u>Gan Luo</u>\*, 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.