# Jiamin Li

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Ph.D. Candidate ( $3^{rd}$  year)  $\diamond$  Computer Science  $\diamond$  City University of Hong Kong

## RESEARCH INTERESTS

- Distributed machine learning system: accelerating large-scale distributed training and inference tasks
- Simulation: accurate performance simulator of DNN training workloads
- Resource scheduling in GPU clusters: schedule DNN tasks with maximum efficiency

#### **EDUCATION**

## City University of Hong Kong

Sep. 2019 - Present

Ph.D. Candidate, Department of Computer Science.

Supervisors: Prof. Hong Xu (CUHK), Prof. Cong Wang (CityU)

## City University of Hong Kong

Aug. 2015 - Jul. 2019

B.S., Department of Computer Science (First Class Honours).

Advisor: Dr. Shiqi Wang

Dissertation Title: "Mobile Face Anti-spoofing with Deep Learning"

## University of Missouri, Columbia, MO, United States

Jul. 2017 – Aug. 2017

Big Data Analysis Summer Program, Department of Computer Science.

## **EXPERIENCES**

## Research Intern, MLSys - AI Lab, ByteDance

May. 2019 – May. 2021

Supervisor: Dr. Yibo Zhu

We build a task scheduler for training and inference GPU clusters. The key idea is to exploit cluster-level elasticity by loaning idle inferences servers for training and job-level elasticity by scaling jobs to better utilize the dynamic resource pool.

#### Part-time Research Assistant, City University of Hong Kong

Jun. 2018 – Jun. 2019

Supervisor: Prof. Hong Xu

In backpropagation of DNN training, the update of some DNN model gradients is tiny across multiple iterations. We use a threshold to adaptively control the gradients that need to be sent each time so that the transfer time of communication operations can be reduced.

## Backend Developer Intern, Jardine Matheson & Co. Limited

May. 2017 – May. 2018

Design and develop web services to facilitate employee recruitment for the Group Human Resources department.

## **PUBLICATIONS**

## **Preprints**

- [P3] **Jiamin Li**, Cheng Luo, Ziyue Yang, Lei Qu, Peng Cheng, Cong Wang, Hong Xu, "Merak: An Analytical Performance Simulator for Large Scale Distributed Training", under review.
- [P2] **Jiamin Li**, Yimin Jiang, Yibo Zhu, Cong Wang, Hong Xu, "Lina: Accelerating Distributed Training of Large Sparsely Activated Models".
- [P1] **Jiamin Li**, Hong Xu, Yibo Zhu, Zherui Liu, Chuanxiong Guo, Cong Wang, "Lyra: Elastic Cluster Scheduling for Deep Learning", under review.

## Conference paper

[C1] Kaiwei Mo, Chen Chen, Jiamin Li, Hong Xu, Chun Jason Xue, "Two-Dimensional Learning Rate Decay: Towards Accurate Federated Learning with Non-IID Data", IEEE International Joint Conference on Neural Networks (IJCNN), 2021.

## Journal paper

- [J2] Libin Liu, Hong Xu, Zhixiong Niu, Jingzong Li, Wei Zhang, Peng Wang, Jiamin Li, Jason Xue Chun, Cong Wang, "ScaleFlux: Efficient Stateful Scaling in NFV", accepted by IEEE Transactions on Parallel and Distributed Systems, 2022 (TPDS).
- [J1] Libin Liu, Chengxi Gao, Peng Wang, Hongming Huang, Jiamin Li, Hong Xu, Wei Zhang, "Bottleneck-Aware Non-Clairvoyant Coflow Scheduling with Fai", accepted by IEEE Transactions on Cloud Computing, 2021 (TCC).

## **MAJOR PROJECTS**

## Accelerating distributed training of MoE models

Collaborating with ByteDance

Distributed training of MoE models is prone to low efficiency, mainly due to the interleaved all-to-all communication during model computation. In Lina, we propose a new communication scheduling scheme based on tensor partitioning that prioritizes the all-to-all operations over other communication.

## Performance Simulator for Large-scale Distributed DNN Training

Collaborating with MSRA

We design a performance simulator to accurately predict the step time of large-scale distributed DNN training tasks. Merak introduces an analytical formulation to compute the all-reduce kernel running time. We also build an ML model to predict the running time slowdown caused by concurrent execution in wait-free backpropagation.

## **Elastic Cluster Scheduler for DNN jobs**

Completed at ByteDance

Organizations often build separate training and inference clusters for deep learning. This leads to low utilization in inference clusters and long queuing for jobs in training clusters. Lyra is designed and built to address these problems by cluster-level capacity loaning and job-level elastic scaling.

#### **SELECTED AWARDS**

Full Postgraduate Studentship	CityU	2019 - 2023
Dean's List (College of Engineering)	CityU	2015 - 2019

#### PROFESSIONAL SERVICE

## **Artifact Evaluation Committe**

ACM CoNEXT 2022

## **TECHNICAL SKILLS**

Languages C++, Python, Bash, Go, LATEX

Operating Systems Linux/UNIX

MLSys PyTorch, MXNet, DeepSpeed, NCCL, CUDA