Yuhao Zhang

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Education

University of California, San Diego, La Jolla, California USA

Ph.D. Student, Computer Science, September 2017 - Spring 2023
- Dissertation Topic: "High-throughput Data Systems for Deep Learning Workloads"
- Advisor: Prof. Arun Kumar

Nankai University, Tianjin, China

B.S., Theoretical Physics, June 2016
- Advisor: Prof. Xueqian Li and Prof. Yangang Miao

Research Interests

My research is primarily on machine learning systems and large-scale data analytics systems, including systems powered by applied ML that enable novel applications and systems designed for ML to make data science easier and faster. Past and ongoing projects include systems for: distributed deep learning model selection and training, distributed in-database deep learning, distributed large-scale graph neural network training, and video analytics and querying.

Projects

Distributed High-Throughput Graph Neural Network Training System (Lotan):

Recent advances in Graph Neural Networks (GNNs) have changed the landscape of modern graph analytics. The complexity of GNN training and the challenges of GNN scalability has also sparked interest from the system community, with efforts to build systems that provide higher efficiency, better memory management, and schemes to reduce costs. However, many of these systems reinvent the wheel by reiterating years of research and development on advanced graph data systems. Further, they often couple the scalability challenges of graph data processing with those of GNN training, resulting in entangled complex problems and systems that need help to handle either scalability challenge. Lotan is our highly scalable data system for full-batch GNN training with the separation of graph and neural network. Lotan provides separation of the graph and neural network computation, free individual scaling for each, and bridges existing graph data systems and deep learning frameworks. Lotan offers a series of technical innovations, including execution plan rewriting, highly-efficient data movement between systems, a GNN-centric graph partitioning method and the corresponding gradients backpropagation scheme, and GNN model batching. By taking several different workloads, we demonstrate the system's capability of training large GNN models that prior art fails to handle. Our system can surpass the state-of-art systems' training throughput by over 40x and can beat a naively implemented in-data-system GNN training framework by 76x. We believe our system can increase efficiency for existing workloads and open new possibilities for future GNN algorithmic research.

Distributed Deep Learning on Data Systems (Cerebro-DS):

Deep learning (DL) is growing in popularity for many data analytics applications, including among enterprises. Large business-critical datasets in such settings typically reside in RDBMSs or other data systems. The DB community has long aimed to bring machine learning (ML) to DBMS-resident data. Given past lessons from in-DBMS ML and recent advances in scalable DL systems, DBMS and cloud vendors are increasingly interested in adding more DL support for DB-resident data. In this paper, we show that there is no single "best" approach to achieve that goal and an

interesting tradeoff space of approaches exists. We explain four canonical approaches, compare them analytically on multiple criteria (e.g., runtime efficiency and ease of governance) and compare them empirically with large-scale DL workloads. Our experiments and analyses show that it is non-trivial to meet all practical desiderata well and there is a Pareto frontier; for instance, some approaches are 3x-6x faster but fare worse on governance and portability. Our results and insights can help DBMS and cloud vendors design better DL support for DB users. The system is based on top of Postgres, Apache Spark, TensorFlow, and PyTorch.
Project homepage: `https://adalabucsd.github.io/cerebro.html`
Open-sourced release: `https://github.com/makemebitter/cerebro-ds`

Resource-efficient Distributed Deep Learning Model Selection and Training System (Cerebro):

Cerebro is a layered data platform for scalable deep learning. One of the biggest challenges for scalable deep learning is model selection, which is difficult and resource-heavy. This system provides high-level APIs for deep learning model selection and optimizes distributed deep learning training for model selection workloads. It adopts a new form of parallelism called Model Hopper Parallelism (MOP) and is the resource-optimal choice. It can expedite distributed deep learning training by 3x-10x compared to Horovod and TensorFlow Parameter Server. There are also subsequent works about bridging data systems (such as Apache Spark, Distributed databases, etc.) with distributed DL systems.
Project homepage: `https://adalabucsd.github.io/cerebro.html`
Open-sourced release: `https://adalabucsd.github.io/cerebro-system`

Video Querying System under Unbounded Vocabulary Setting (Panorama):

Panorama is a video querying system for object detection and classification. It is designed to tackle the life-long learning or bounded vocabulary issues. Computer-vision-based video analytics systems often require *update* of the vocabulary as new classes/labels need to be added from time-to-time. Such updates involve re-training of the deep learning model, which requires expertise and is very time-consuming. Panorama is built to partially avoid and automate this process while expediting inference speed at the same time.
Project homepage: `https://adalabucsd.github.io/panorama.html`
Open-sourced release: `https://github.com/makemebitter/Panorama-UCSD`

| RESEARCH IMPACT | Cerebro and MOP integrated into Greenplum Database and shipped by VMware | 2019 |
| | Code of Cerebro integrated into Apache MADlib project | 2019 |

| ACADEMIC EXPERIENCE | **University of California, San Diego**, La Jolla, California USA | |
| | *PhD Student* | **Sept 2017 - present** |

Includes current Ph.D. research, Ph.D. and Masters level coursework and research/consulting projects.
Courses taken: Machine Learning, Deep Learning, Data Mining & Analytics, Advanced Data Analytics, Computer Vision, Database Systems, Advanced Algorithms, Advanced Compilers, Principles of Programming Languages, Introduction to Robotics

Collaborator for project DeepPostures: a deep learning library for identifying human postures from wearable devices data   **Spring 2022 - Spring 2023**
Built and maintained project website (https://adalabucsd.github.io/DeepPostures/). Wrote documentation and made demos about the project. Provided software engineering support for other public health researchers. Conducted tests and data analysis for the project.

Teaching Assistant for CSE234: Data Systems for Machine Learning **Winter 2021**
Helped designing and supervising a research-oriented course on machine learning systems. Mentored 12 master students with their course projects ranging from advanced implementation of cutting-edge research, to evaluation and surveying the state-of-art work, and to open-ended research.

Teaching Assistant for DSC102: Systems for Scalable Analytics **Winter 2020**
Developed the first edition of course assignments and auto-grading programs with Python and Bash. The assignments involve Python Dask, Spark, AWS EC2/S3/EBS, and Kubernetes. These assignments have been adopted by the course ever since and used by 500+ students.

Texas A&M University, College Station, Texas USA

Research Intern **Summer 2015**
Worked on vision-based object tracking, modeling, and data analytics.

Institute of Physics, Chinese Academy of Sciences, Beijing, China

Research Intern **Summer 2014**
Theoretical physics research. Particle physics.

Nankai University, Tianjin, China

Research Assistant **2013 - 2016**
Theoretical physics research. Black holes, gravity, and neutrinos.

PROFESSIONAL EXPERIENCE

Microsoft Gray System Lab, California USA

Research Intern **Summer 2021**
Worked on machine learning system research, focusing on factorized in-DBMS machine learning. Built a highly scalable system on top of Apache Spark that can outperform its MLlib by orders of magnitude.

VMware, Palo Alto, California USA

Software Engineer Intern **Summer 2019**
Worked on the first in-DBMS deep learning system, allowing training and inference of deep learning models with TensorFlow on database-resident data. Integrated my research project, Cerebro, into the deep learning training infrastructure of Greenplum Database, boosting efficiency by over 10x. Contributed to the Apache MADlib project in Python and SQL. Lead the development of a major release. This project has been incorporated into Greenplum and production-ready for VMware's customers.

Opera Solutions, San Diego, California USA

Data Scientist Intern **Summer 2018**
Worked on a theatre scheduling & recommender system. Proposed new models and optimized the existing system.

PUBLICATIONS

Lotan: a highly scalable system for GNN training via the separation of graph and neural network
Yuhao Zhang and Arun Kumar
Under submission

Some Damaging Delusions of Deep Learning Practice (and How to Avoid Them)
Arun Kumar, Supun Nakandala, Yuhao Zhang
KDD 2021 Deep Learning Day

Distributed Deep Learning on Data Systems: A Comparative Analysis of Approaches
Yuhao Zhang, Arun Kumar, Frank McQuillan, Nandish Jayaram, Nikhil Kak, Ekta Khanna, Orhan

Kislal, and Domino Valdano
VLDB 2021

Cerebro: A Layered Data Platform for Scalable Deep Learning
Arun Kumar, Supun Nakandala, Yuhao Zhang, Side Li, Advitya Gemawat, and Kabir Nagrecha
CIDR 2021

Cerebro: A Data System for Optimized Deep Learning Model Selection
Supun Nakandala, Yuhao Zhang, and Arun Kumar
VLDB 2020

Panorama: A Data System for Unbounded Vocabulary Querying over Video
Yuhao Zhang and Arun Kumar
VLDB 2020

Cerebro: Efficient and Reproducible Model Selection on Deep Learning Systems
Supun Nakandala, Yuhao Zhang, and Arun Kumar
ACM SIGMOD 2019 DEEM Workshop

Three-generation neutrino oscillations in curved spacetime
Yuhao Zhang and Xueqian Li
Nuclear Physics B, 2016

Self-regular black holes quantized by means of an analogue to hydrogen atoms
Chang Liu, Yangang Miao, Yumei Wu, and Yuhao Zhang
Advances in High Energy Physics, 2016

PRESENTATIONS
High-throughput Data Systems for Deep Learning Workloads
UCSD Database Seminar   **Spring 2022**

Distributed Deep Learning on Data Systems
VLDB 2021   **August 2021**

A Layered Data Platform for Scalable Deep Learning
UCSD Database Seminar   **October 2020**
UCSD CNS Research Review   **October 2020**

Resource-Efficient Deep Learning Model Selection on Apache Spark
Spark+AI Summit 2020   **June 2020**
UCSD Database Seminar   **May 2020**

Cerebro: A Data System for Optimized Deep Learning Model Selection
VLDB 2020   **September 2020**
UCSD Database Seminar   **June 2020**

Panorama: Unbounded Vocabulary Queries on Video
VLDB 2020   **September 2020**
UCSD Database Seminar   **June 2020**
Poster presentation, UCSD CSE Research Open House   **January 2019**
UCSD Database Seminar   **November 2019**
Poster presentation, SoCal DB Day 2018   **October 2018**

Apache MADlib 1.17 Showcase
VMware and online   **September 2019**

SERVICE   **External reviewer**: SIGMOD 2020, SIGMOD 2021, VLDB 2022

               Reviewer: JMLR MLOSS 2022, SIGMOD 2024

MISC   SIGMOD 2021 *Students and Postdocs in DB Panel Discussion*   **June 2021**

HONORS AND   Best Thesis Award, School of Physics, Nankai University, 2016
AWARDS   Wang Kechang Scholarship for Academic Distinction, 2014
               Gong-Neng Scholarship, 2013
               Poling Academy Scholarship, 2012-2016

TECHNICAL SKILLS
- Programming languages: Python, C/C++, Scala, Java, SQL, R
- Data platforms: Spark, Dask, PostgreSQL, Greenplum Database, Hadoop, Hive
- Machine learning frameworks: TensorFlow, PyTorch, Keras, xgBoost, LightGBM, Scikit-learn
- Other data analytics packages: Pandas, NumPy, Matplotlib, OpenCV, Scikit-image, Pillow/PIL
- Cloud/Cluster tools: AWS EC2/S3/EMR, Google Cloud, Azure, Kubernetes, Docker
- Miscs: MPI, MATLAB, Mathematica, CERN ROOT, CERN GEANT4