

Yuhao Zhang

CONTACT INFORMATION

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EDUCATION

University of California, San Diego, La Jolla, California USA
Ph.D. Student, Computer Science, September 2017 - December 2022 (anticipated)

- Advisor: Prof. Arun Kumar

Nankai University, Tianjin, China
B.S., Theoretical Physics, June 2016

- Advisor: Prof. Xueqian Li and Prof. Yangang Miao

RESEARCH INTERESTS

Machine learning systems, including systems powered by applied ML that enable novel applications, and systems designed for ML to make data science easier and faster

HONORS AND AWARDS

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

PROJECTS

Distributed Deep Learning on Data Systems:

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.

Open-sourced release: <https://github.com/makemebitter/cerebro-ds>

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>

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, Data Mining & Analytics, Advanced Data Analytics, Computer Vision, Database Systems, Advanced Algorithms, Advanced Compilers, Principles of Programming Languages, Introduction to Robotics

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, to open-ended research questions.

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 and data analysis & modeling.

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

Research Intern

Summer 2014

Research in theoretical physics.

Nankai University, Tianjin, China

Research Assistant

2013 - 2016

Research in theoretical physics.

PROFESSIONAL EXPERIENCE

Microsoft Gray System Lab, California USA

Research Intern

Summer 2021

Worked on machine learning system research, focusing factorized in-DBMS machine learning.

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

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

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

<i>Resource-Efficient Deep Learning Model Selection on Apache Spark</i>	June 2020
Spark+AI Summit 2020	
UCSD Database Seminar	May 2020
<i>Cerebro: A Data System for Optimized Deep Learning Model Selection</i>	
VLDB 2020	September 2020
UCSD Database Seminar	June 2020
<i>Panorama: Unbounded Vocabulary Queries on Video</i>	
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
<i>Apache MADlib 1.17 Showcase</i>	
VMware and online	September 2019

SERVICE	External reviewer: SIGMOD 2020, SIGMOD 2021, VLDB 2022, JMLR MLOSS 2022
MISC	SIGMOD 2021 <i>Students and Postdocs in DB Panel Discussion</i> June 2021
TECHNICAL SKILLS	<ul style="list-style-type: none"> • 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