I'm a Ph.D. candidate at the University of Michigan, advised by Prof. Satish Narayanasamy, working on confidential computing and trusted hardware, with a strong background in GPU architecture and ML systems. I'm seeking a research internship position for the spring or summer of 2025.

My research seeks to advance confidential computing solutions for enabling privacy-preserving data analytics solutions ranging from population-scale genomic analysis to generative AI. My approach involves developing trustworthy hardware to efficiently guarantee privacy across system components, excluding the operating system and system administrators from the trust base.

I have made four key contributions. First, I invented the **Toleo**. Today, trusted processors (Intel SGX) support only a few hundred MBs of secure memory space. Toleo is an innovative solution that expands trust to intelligent memory and scales secure memory space to tens of TeraBytes, which is a million times larger than what is feasible today (ASPLOS'24).

Secondly, in collaboration with Meta, I built FlexDecoding, the inference backend for **FlexAttention**. FlexDecoding addresses the lack of flexibility in supporting new attention variants in today's LLM infrastructure and combines the flexibility of the PyTorch compiler with the performance of expert-tuned decoding kernels. FlexAttention (initially launched with only the training backend in Aug 2024) received 170k views on X. FlexDecoding is set to launch in Oct 2024.

Thirdly, I contributed to the development of **SECRET-GWAS**, a privacy-preserving genome-wide association study platform built on Microsoft Azure's confidential computing platform. SECRET-GWAS scales to over 1000 cores and we demonstrated for the first time that regression analysis on large genomic datasets from multiple institutions can be performed in a few seconds, without exposing data to cloud service provider (under review for Nature Computational Science).

Lastly, in collaboration with AMD, I developed new techniques to accelerate long-read genome sequencing (mm2-gb, published in ACM BCB'24). mm2-gb advances computing for genome mapping and alignment, removing computational bottlenecks in the sequencing pipeline to keep up with the ultra-long read trend. Our artifact has gained significant community attention and will soon be released as part of AMD Research Open-source Project.

Additionally, I am contributing to **Timelocked Storage**, a project aimed at minimizing the Trusted Code Base (TCB) for ransomeware defense and excluding human administrators, operating systems, even the main processor from the TCB.

Looking ahead, I plan to further advance confidential computing to address privacy and safety concerns of generative AI.

The release of NVidia's Hopper confidential computing feature brings GPUs into the trusted hardware family, and offers an exciting opportunity to impact how ML models are trained and used. Current GPU TEE solutions heavily relies on software drivers and limits key optimizations, such as DMA for CPU/GPU communication. I aim to explore solutions that integrate GPU and CPU into a unified TEE, leveragin finer granularity on data movement based by low-level hardware primitives (such as through CXL-IDE feature (Integrity Data Encryption)) to construct a unified trusted CPU-GPU memory. I'm also open to exploring other challenges related to integrating GPUs into the TEE system.

Please see the next page for my CV and detailed background.

Juechu Dong

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SUMMARY

Juechu (Joy) Dong is a Ph.D. candidate at the University of Michigan, advised by Prof. Satish Narayanasamy. Her research focuses on confidential computing, computing for biotechnologies and GPU architecture. Her work seeks to advance confidential computing solutions for enabling privacy-preserving data analytics solutions ranging from population scale genomic analysis to generative AI.

EDUCATION

University of Michigan - Ann Arbor

 $(\exp.) 2027$

Computer Science and Engineering, PhD

Topics: Computer Architecture, Confidential Computing, Computing for Biotechnologies

Advisor: Prof. Satish Narayanasamy

University of Michigan - Shanghai Jiao Tong University Joint Institute

Aug 2022

Computer Engineering, Bachelor of Science

University of Michigan - Ann Arbor

Apr 2022

Computer Engineering, Bachelor of Science in Engineering, Summa Cum Lauda

SELECTED HONORS

Meta 2024 Internship Project Spotlight: FlexDecoding

2024

Awarded to 3 Internship project picked by CEO Mark Zuckerberg as spotlight of the year

Rackham International Student Fellowship

2023-2024

Awarded to 25 outstanding students among all international PhD and MS students in the university

PUBLICATION

Toleo: Scaling Freshness to Tera-scale Memory Using CXL and PIM

ASPLOS'24

J. Dong, J. Rosenblum, S. Narayanasamy

Accepted

- Scale trusted memory size from hundreds of MB to tens of TB by expanding the span of trusted from a single trusted processor to an entire platform including intelligent memories.
- Design a new scheme of freshness protection that reduces the space requirement by 50x.
- Reduce deployment cost by spacing sharing one intelligent memory device among multiple CPUs.

mm2-gb: GPU Accelerated Minimap2 for Long Read DNA Mapping

ACM BCB'24

J. Dong, X. Liu, H. Sadasivan, S. Sitaraman, S. Narayanasamy

Accepted

- Accelerate computational intensive chaining step in the state-of-art long sequence mapping tool minimap2 using AMD GPU by 2.57-5.33x.
- Optimize towards ultra long reads of 100k+ to accommodate genome sequencing technology trend.
- Develop adaptive GPU scheduling algorithm to balance highly heterogeneous workload.

SECRET-GWAS: Confidential Computing for Population-Scale GWAS

Nature Comp. Sci.

J. Rosenblum, J. Dong, S. Narayanasamy

under review

- Develop a thousand-core platform on Azure Confidential Computing to conduct multi-institutional GWAS on millions of patients in less than a minute.
- Adapt Spark-based Hail genomic analysis framework to run on TEE under obliviousness requirement.
- Parallelize GWAS computation on 1k cores to achieve near linear speedup.

PROJECTS IN PROGRESS

FlexAttention: Flexibility of PyTorch with the Performance of FlashAttn

PyTorch Blog

Joy Dong, Driss Guessous, Yanbo Liang, Boyuan Feng, Horace He

- Flexible: An API that allows implementing many attention variants in a few lines of idiomatic PyTorch code.
- Performant: Provides 1.1x 1.3x speedup compared to FlashAttn2 by lowering customizable attention into a fast Triton kernel + taking advantage of sparsity.
- Self Adaptive: optimized to efficiently support decoding, GQA and PagedAttention.

Timelocked Storage for Ransomware Defense

- J. Rosenblum, J. Dong, S. Narayanasamy
 - Shrink the Trusted Code Base (TCB) for randomware defense to a formally verified timelock gate-keeper in front of disks.
 - Compatible with different flavors of versioning file system designs.
 - Achieve near zero disk access overhead.

INDUSTRY EXPERIENCE

PyTorch group, Meta Inc.

May 2024 - Aug. 2024

Rearch Scientist Intern

- Build the fast, efficient and flexible attention API for decoding and GQA.
- Develop new techniques in PyTorch compiler with a focus on GPU performance optimization.
- Conduct performance analysis and optimizations on attention kernels.

NVIDIA May 2022 - Aug. 2022

GPU Deep Learning Architect Intern

- Model and analyze new memory features on next-gen GPUs such as distributed shared memory, asynchronous transaction barrier, etc.
- Analyze and optimize multi-GPU data movement for deep learning workloads using Tensor Memory Accelerator (TMA).

TEACHING

Instructional Aide & Graduate Student Instructor

2021 - 2024

EECS470 Comp Arch; EECS471 Applied GPU Prog; EECS570 Parallel Comp Arch

SERVICE

University of Michigan - Shanghai Jiao Tong University Dean Search Committee

2024

Committee Member, Alumni Representative

University of Michigan Computer Engineering Lab Reading Group

2022 - 2024

Coordinator

UM-SJTU Joint Institute Alumni Association

2022 - present

Founder & Vice President

SKILLS

Programming Languages: C/C++, CUDA, HIP, Triton, (system) verilog

Technologies/Frameworks:

ML Stack: PyTorch (TorchInductor, TorchDynamo)

GPU Tuning: nsight-compute/nsight-sys, omniperf/omnitrace/rocprof

Simulation: SniperSim, DRAMSim, pinplay

Confidential Computing: Open Enclave SDK, Intel SGX

Architectures: AMD CDNA2 Instinct GPU, NVIDIA Hopper GPU, Intel Xeon Phi, Out-of-order CPU