# Jiayi Tian

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### Focus on efficient training and inference of LLMs (Low-rank decomposition, Pruning, Quantization, Knowledge Distillation)

### **EDUCATION**

University of California, Santa Barbara, Ph.D. in Computer Engineering | CA, USA 3.9/4.0 Nanjing University, B.Eng. in VLSI Design & System Integration | China 4.5/5.0

Fall 2023 - ongoing Fall 2019 - Jun 2023

### INDUSTRIAL EXPERIENCE \_

### **Intel Corporation,** Research Intern | Portland, OR

June. 2024 - Sep. 2024

- Proposed a tensor-compressed LLM training accelerator using FPGA with optimized compute ordering, dataflow, and memory allocation.
- Achieved up to 48× memory efficiency and 3.6× energy efficiency compared to Nvidia RTX 3090.
- Resulting paper under review at IEEE TCAD.

## **AMD-Xilinx Technology,** *Co-Op/Intern* | Beijing, China

June 2023 - Sep 2023

- Developed a C++/HLS Transformer training framework with custom tensorized linear layers and nonlinear operations for LLM acceleration.
- Achieved  $30 \times \sim 52 \times$  saving in model size for end-to-end Transformer training.

### SKILLS & RESEARCH INTERESTS \_

Languages & Tools

Python, PyTorch, TensorFlow, Huggingface, C/C++, High-level Synthesis (HLS), Vivado/Vitis/XRT

Large Language Models (LLMs), Efficient Training/Inference (Model Compression, Pruning, Low-rank decomposition, Distillation, Quantization)

### PUBLICATIONS & PREPRINTS \_

## FLAT-LLM: Fine-grained Low-rank Activation Space Transformation for Large Language Model Compression

Jiayi Tian, Ryan Solgi, Jinming Lu, Yifan Yang, Hai Li, Zheng Zhang, under review at ARR May, 2025. arXiv preprint.

## FETTA: Flexible and Efficient Hardware Accelerator for Tensorized Neural Network Training

Jinming Lu, Jiayi Tian, Hai Li, Ian Young, Zheng Zhang, under review at IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems. arXiv preprint.

### Ultra Memory-Efficient On-FPGA Training of Transformers via Tensor-Compressed Optimization

Jiayi Tian, Jinming Lu, Hai Li, Xiangwei Wang, Cong (Callie) Hao, Ian Young, Zheng Zhang, under review at IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems. arXiv preprint.

## **BEBERT: Efficient and robust binary ensemble BERT**

Tian, Jiayi, Chao Fang, Haonan Wang, and Zhongfeng Wang, ICASSP 2023-2023 IEEE International Conference on Acoustics, Speech and Signal Processing (ICASSP). IEEE, 2023.

### RESEARCH PROJECTS \_

## Structural Pruning for Efficient LLM Inference via Low-rank Decomposition

Aug. 2024 - Current

- Developed a training-free, fine-grained compression method that leverages the low-rank structure of the activation space to transform and compress model weights.
- Proposed a novel rank selection algorithm based on component importance, enabling seamless integration with existing low-rank LLM compression pipelines.
- Achieved strong performance on LLaMA-2 and Mistral models with minimal calibration overhead (within minutes), validated across language modeling and downstream tasks.

## **Training Accelerator Design for Tensor-Compressed Transformer Models**

Sep. 2023 - Dec. 2024

- Designed a tensor-compressed training scheme for Transformer models that reduces model size by  $30 \sim 52 \times$ .
- Introduced bidirectional tensor contraction to enhance memory and compute efficiency, especially in long-sequence training and inference.
- Built an HLS-based Transformer training engine achieving up to 48× memory efficiency and 3.6× energy efficiency compared with Nvidia RTX 3090.

## Binary-Quantized Ensemble LLM for Fast and Robust Language Model Inference

Apr. 2021 - June. 2023

- Developed BEBERT, a novel quantization-ensemble strategy enabling efficient and accurate 1-bit BERT inference.
- Leveraged efficient knowledge distillation strategy for high training efficiency.
- Achieved  $13 \times$  model size reduction and  $15 \times$  compute savings over standard BERT with minimal accuracy loss.
- Proposed early-exit inference variant, further cutting compute by  $20\% \sim 40\%$  on GLUE benchmark.