XI HAN

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EDUCATION

Department of Computer Science, Stony Brook University, New York, United States

Aug 2019 – Present

Ph.D. in Computer Science (In progress, expected by Spring 2026) | GPA: 3.9/4.0

Department of Computer Science and Technology, Tsinghua University, Beijing, China

Aug 2015 – Jul 2019

B.E. in Computer Science and Technology | GPA: 3.25/4.0

PUBLICATIONS

- Xi Han, Fei Hou and Hong Qin, "UGrid: An Efficient-And-Rigorous Neural Multigrid Solver for Linear PDEs", In Proceedings of the 41st International Conference on Machine Learning, pp. 17354 – 17373, July 2024.
- Song-Hai Zhang, Ruilong Li, Xin Dong, Paul Rosin, Zixi Cai, Xi Han, Dingcheng Yang, Hao-Zhi Huang and Shi-Min Hu, "Pose2Seg: Detection Free Human Instance Segmentation", In 2019 IEEE/CVF Conference on Computer Vision and Pattern *Recognition (CVPR)*, pp. 889 – 898, June 2019.

WORK EXPERIENCE

Computer Graphics Lab

Stony Brook University, New York, United States | Research Assistant & Teaching Assistant Advisor: Hong Qin, Professor at Department of Computer Science, Stony Brook University

Aug 2019 – Present

- Conducted research in computer graphics (intelligent physics-based modeling). Involved concepts: Differentiable PDE-based vector graphics, data-driven neural PDE solvers, etc. Implemented multiple advanced research projects related to graphics and numerical analysis (Differentiable PDE solvers with customized CUDA operators).
- Cooperates with Computer Vision lab on training/inference efficiency optimization for AI models. Involved techniques: CUDA kernel fusing, performance profiling, and customized cache-friendly differentiable AI operators such as differentiable Monte-Carlo integrator, fused GEMM, 2D mamba scanner, etc.
- Hosted lectures on OpenGL programming with C++/Python, the implementation details of computer graphics applications and algorithms, and the state-of-the-art research topics on graphics and physics-based modeling.

Computer Graphics and Animation Lab

University of Texas at Dallas, Texas, United States | Research Assistant

Sep 2018 – Nov 2018

Advisor: Xiaohu Guo, Professor at Department of Computer Science, University of Texas at Dallas

- Worked on the 3D face reconstruction project with a local Samsung research lab. Also constructed a human face model dataset for further research purposes.
- Configured a Linux workstation for deep learning purposes from zero and deployed neural network models on it.

Graphics and Geometric Computing Group

Tsinghua University, Beijing, China | Research Assistant

Jan 2017 – Jul 2019

Advisor: Song-Hai Zhang, Professor at Department of Computer Science and Technology, Tsinghua University

- Deployed a MobileNet module on IOS platform with Apple's CoreML framework, and delivered an IOS app for a human segmentation (in Swift and Objective C++).
- Optimized the model used in the app (increased accuracy and added key point recognition) and achieved 10x speedup in FPS.

SKILLS

	Numer	rical analysis, high-performance computing, computer graphics, machine learning, and Linux system skills.
		Expertise in computer graphics and numerical analysis: Neural PDE solvers, customized CUDA-level operators with back-propagation capability.
		Expertise in AI/HPC: Customized AI operators, AI model training/inference efficiency optimization. Involved topics: PyTorch C++/CUDA extensions, kernel profiling, fine-tuning, operator fusing, cache optimization, etc.
		Expertise in programming languages: C/C++ (OOP, STL, Metaprogramming and Concurrency), CUDA (including PTX) and Python.
		Expertise in tools: PyTorch Profiler, CUDA-GDB, Nsight Compute and NVIDIA Compute Sanitizer.
		Expertise in frameworks: PyTorch, OpenGL and Qt.
		Other proficiencies: Bash, CMake, Assembly, MATLAB, Java, Objective C/C++ and Swift.
Language Proficiencies:		

Chinese (Mandarin) (Native speaker);

		English (Proficient for working scenarios. TOEFL: 106/120; GRE: 324/340 + Writing 3.5);		
		Japanese (Sufficient for basic working scenarios. JLPT: N1 173/180, N2 169/180).		
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	ECTED P			
		Efficient-And-Rigorous Neural Multigrid Solver for Linear PDEs		
		UGrid is a neural solver for Partial Differential Equations (PDEs) with convergence guarantee.		
>	to 20x s	soon the combination of the U-Net architecture and the legacy MultiGrid PDE solver, provides users with high speed (up speedup against legacy solvers, high precision (relative residual as low as 1e-5), high robustness (against irregular and aput), high generalization power (to irregular boundary geometries and topology), and high scalability (without need for ng).		
>	Involved techniques: Numerical analysis on convergence, customized AI operators (Python and CUDA based). Implements a customized CUDA convolution module to save computation for specific-shaped convolution kernels used in PDE solvers.			
\triangleright	Code as	vailable at https://github.com/AXIHIXA/UGrid.		
2D-	Mamba:	Hardware-aware 2D Parallel Mamba Scanner		
>	<u>TL;DR</u> : 2D-Mamba scanner extends 1D Mamba into 2D while maintaining its modeling capabilities, high parallelism, and memory efficiency.			
>	Involved techniques: 2D tiling and caching, HBM access optimization, CUDA kernel and AI model profiling, and PyTorch CUDA extension encapsulation.			
\triangleright	Compared to a naïve implementation, achieves a throughput of 10x, while the GPU memory consumption is only 10%.			
>	Code av	ailable at https://github.com/AtlasAnalyticsLab/2DMamba (CUDA extension part).		
CUI	DA Basel	line Experiments		
>	TL;DR:	Implements and fine-tunes multiple CUDA baseline algorithms.		
>	Implem	ented baselines and their optimizations:		
		Parallel reduction (with loop unrolling and warp shuffle primitives);		
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CUDA

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	Parallel reduction (with loop unrolling and warp shuffle primitives);
	Histogram and Copy-If (with atomic primitives);
	Parallel scan (WarpScan and Raking variants);
	Fused Biased-Mask-Scale-Add (fp32 and fp16, for fp16, with half precision primitives likehadd2);
	SGEMM an GEMV (loop unrolling, SMEM padding, warp tiling and double buffer optimizations, reaching 90% throughput of cuBLAS);
	Dropout (with cuRAND APIs);
	Fused SoftMax/LayerNorm/RMSNorm, Im2Col, Matrix transpose, etc.
Code ove	piloble at https://github.com/AVIHIVA/CudeDemo