

# **Debunking the CUDA Myth**

## **Towards GPU-based AI Systems**

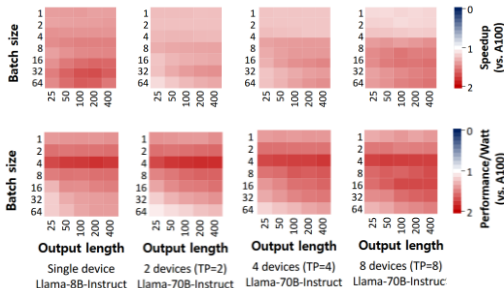
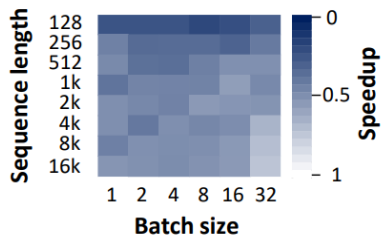
Original Paper by: Yunjae Lee et al. (Prof. Minsoo Rhu's Group, ISCA '25)

Presented by Jongyun Hur

# **Conflict Analysis**

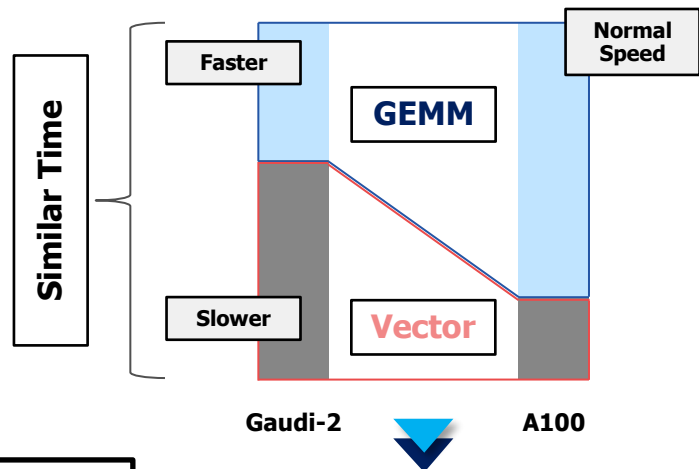
# Conflict Analysis

## Kernel-level weakness vs. End-to-end strength



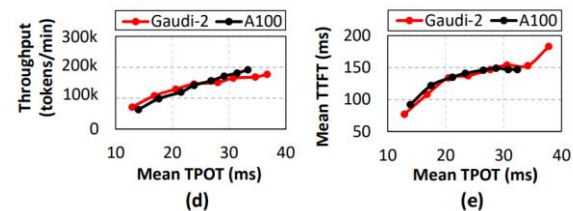
Category	Figure 17(c): PagedAttention (Micro)	Figure 12(a): LLM End-to-End (Macro)
Analysis Scope	Micro Level (Kernel-specific)	Macro Level (System-wide)
Evaluation Target	Throughput of a <b>Single PagedAttention Kernel</b>	Full Llama-3.1 Pipeline (Synthetic Dataset)
Performance vs. A100	~45% of A100 (Inferior)	Avg. 1.47x of A100 (Superior)
Root Cause	<b>Vector &amp; Memory-intensive</b> nature of Attention exposes Gaudi-2's TPC limitations.	<b>GEMM dominates</b> the total workload, where Gaudi-2 gains massive acceleration via MME.

Amdahl's Law:  
Dominant GEMM performance fully offsets minor Vector latency



Difference

vLLM-based Real Serving (Dynamic Dataset)



Similar End-to-End Performance