

### Discovering the In-Memory Kernels of 3D Dot-Product Engines

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## Outline

- Preliminaries
- 3D DPE Level
  - Sharing of hardware
  - Hardware mapping
- Application level
  - Methodology
- Evaluation
- Summary

#### Why In-Memory Computing?





#### NVM Technology and In-Memory Computing







ReRAM

Matrix Vector Multipication<sup>3</sup>



Pro

- In-situ computation
- Extremely energy efficient

- 1. Ghazi Sarwat, Syed, et al. "Projected Mushroom Type Phase-Change Memory." Advanced Functional Materials 31.49 (2021): 2106547.
- 2. Seyedzadeh, Seyed Mohammad & Maddah, Rakan & Jones, Alex & Melhem, Rami. (2016). Leveraging ECC to Mitigate Read Disturbance, False Reads and Write Faults in STT-RAM.
- 3. 1. Wang, Feng, Guangyu Sun, and Guojie Luo. "SSR: A Skeleton-based Synthesis Flow for Hybrid Processing-in-RRAM Modes." 2021 ICCAD. IEEE, 2021.

## 2D vs 3D DPE



## **Previous Work**

#### Table 1: Architectural comparison of dot-product engines

Work in	Form of DPE	# Metal layers
[9, 14, 18]	2D	2
[1, 4, 6, 13]	3D	2-3
This work	3D	2-8

- Previous works focused on mitigation of parasitics on the algorithmic level
- DAC and ADC operations are time multiplexed
- Efficient utilization of 3D ReRAM crossbars with more than three layers have not been explored

[1] G. C. Adam et al. 3d reram arrays and crossbars: Fabrication, characterization and applications. In 2017 leee-Nano, pages 844–849. IEEE, 2017

[4] B. Chakrabarti et al. A multiplyadd engine with monolithically integrated 3d memristor crossbar/cmos hybridcircuit. *Scientific reports*, 7(1):1–10, 2017.

[6] F. Chen et al., A novel zero-free dataflow accelerator for generative adversarial networks in 3d reram. In Proceedings of the 56<sup>th</sup> DAC 2019, pages 1–6, 2019.

[9] B. Feinberg et. Al, Enabling scientific computing on memristive accelerators. In 2018 ACM/IEEE 45th ISCA, pages 367–382. IEEE, 2018.

[13] M. A. Lastras-Montano et. Al, 3d-dpe: A 3d high-bandwidth dot-product engine for high-performance neuromorphic computing. In 2017 DATE, pages 1257–1260. IEEE, 2017.

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## MVM in In-memory DPE



Hardware constraint: Maximum Crossbar Dimension





## **Overhead Comparison**

#### Table 2: Overhead comparison of 2D vs 3D architecture.

	2D Architecture	3D Architecture
Crossbar footprint on chip	4X	1X
# DACs	4X	2X
ADC timesteps	4X	2X

### Problem 1: Hardware Mapping



#### **3D Hardware**

### Shareability Opportunity

 Two adjacent crossbars in a 3D stack can share a set of DACs if the mapped matrix segments are in the same column i.e., to be multiplied with the same input vector



 The dot product output if two adjacent crossbars in a 3D stack can combined if the mapped matrix segments are in the same row i.e., contributing to the same output vector



#### Problem 2: Discovering In-memory 3D DPE Kernel







## **Discovering In-memory 3D DPE Kernel**





## **Discovering In-memory 3D DPE Kernel**



Kernel 2



## **Discovering In-memory 3D DPE Kernel**





### Library of In-memory 3D DPE Kernel



Kernel 1Kernel 2Kernel 3

## **Overhead Comparison**



#### Application level



Problem:

 Matrix dimension might exceed crossbar dimension

Solution:

 Partitioning Matrix into segments



### **Greedy Kernel Cover of Sparse System**



Target:

• Covering non-zero blocks using the kernel library



# Evaluation

#### Area-power cost of architectural components

Component	Parameter	Specs	Area	Power
Crossbar	Size	$128 \times 128$	$25 \ \mu \mathrm{m}^2$	0.3 mW
DAC	Resolution	1 bit	$0.17 \ \mu \mathrm{m}^2$	0.004 mW
ADC	Resolution	8 bits	$0.0012 \text{ mm}^2$	2 mW
Sample+Hold	# Unit	1	$0.04~\mu\mathrm{m}^2$	10 nW
Shift+Add	# Unit	1	$60 \ \mu \mathrm{m}^2$	0.05 mW
	# Crossbar	1		
2D Crossbar	# DACs	128	51.25 $\mu m^2$	0.80 mW
	# Sample+Hold	128		
	# Crossbar	7		
<b>3D Crossbar</b>	# DACs	$4 \times 128$	122.28 $\mu \mathbf{m}^2$	4.10 mW
	# Sample+Hold	$2 \times 128$		
eDRAM Buffer	Size	128 KB	$0.17 \text{ mm}^2$	41.4 mW
IR	Size	4 KB	4200 $\mu m^2$	2.48 mW
OR	Size	512 B	$1500  \mu \mathrm{m}^2$	0.46 mW
Bus	Bandwidth	128-bits	15.7 mm <sup>2</sup>	13 mW

#### Table 4: Overview of benchmarks from the SuiteSparse Matrix Collection.

Applications	Systems	Matrix Dimensions	#Non-zeros
bcsstk34	Structural Problem	$588 \times 588$	21418
eris1176	Power Network Problem	$1176 \times 1176$	18552
coater1	Computational Fluid Dynamics	$1348 \times 1348$	19457
cegb2919	Structural Problem	$2919 \times 2919$	321543
mycielskian12	Undirected Graph	$3071 \times 3071$	407200
raefsky1	Computational Fluid Dynamics	$3242 \times 3242$	293409
crystk01	Materials Problem	$4875 \times 4875$	315891
fxm3_6	Optimization Problem	$5026 \times 5026$	94026
Na5	Theoretical/Quantum Chemistry	$5832 \times 5832$	305630
EX5	Combinatorial Problem	$6545 \times 6545$	295680
fp	Electromagnetics Problem	$7548 \times 7548$	834222
ex40	Computational Fluid Dynamics	$7740 \times 7740$	456188
benzene	Theoretical/Quantum Chemistry	8219 × 8219	242669
bcsstk33	Structural Problem	8738 × 8738	591904
graham1	Computational Fluid Dynamics	$9035 \times 9035$	335472
net25	Optimization Problem	$9520 \times 9520$	401200
bundle1	Computer Graphics/Vision	$10581 \times 10581$	770811
Si10H16	Theoretical/Quantum Chemistry	$17077 \times 17077$	875923
Goodwin_040	Computational Fluid Dynamics	$17922 \times 17922$	561677
pkustk06	Structural Problem	$43164 \times 43164$	2571768



Improvements:

- Area by 2.02X
- Latency by 2.45X
- Energy by 2.37X

# Summary

# Thank You

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