HielM: Highly Flexible In-Memory Computing Using STT MRAM

Deliang Fan

Assistant Professor

dfan@ucf.edu

http://www.eecs.ucf.edu/~dfan/



Department of Electrical and Computer Engineering, University of Central Florida, Orlando, FL

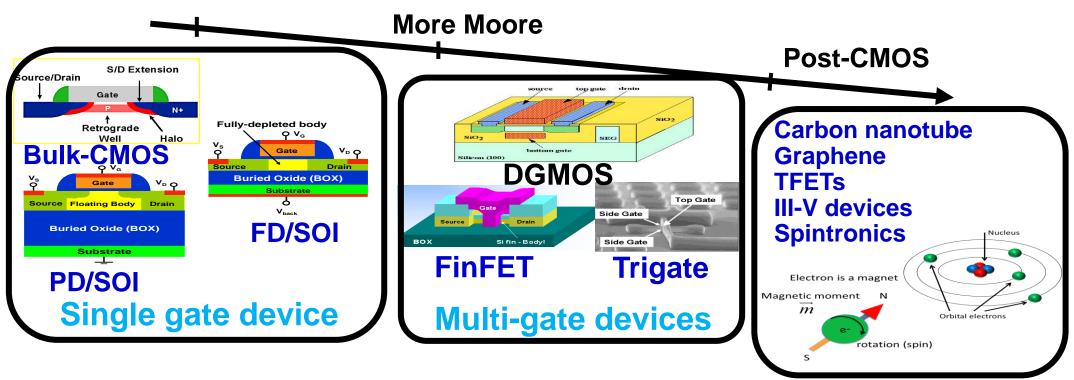




- Motivation
- Post-CMOS Spintronic Devices
- In-Memory Processing Platform based on STT-MRAM
- Performance Evaluation
- Case Study I: In-memory Bulk Bitwise Vector Operation
- Case Study II: In-memory Data Encryption Engine

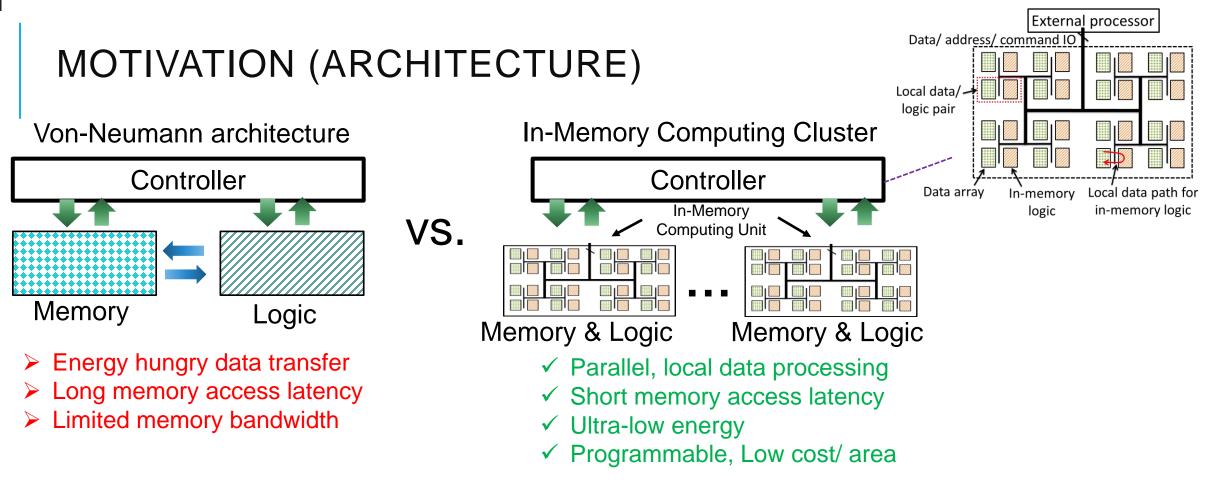


MOTIVATION (DEVICE)-TECHNOLOGY TREND



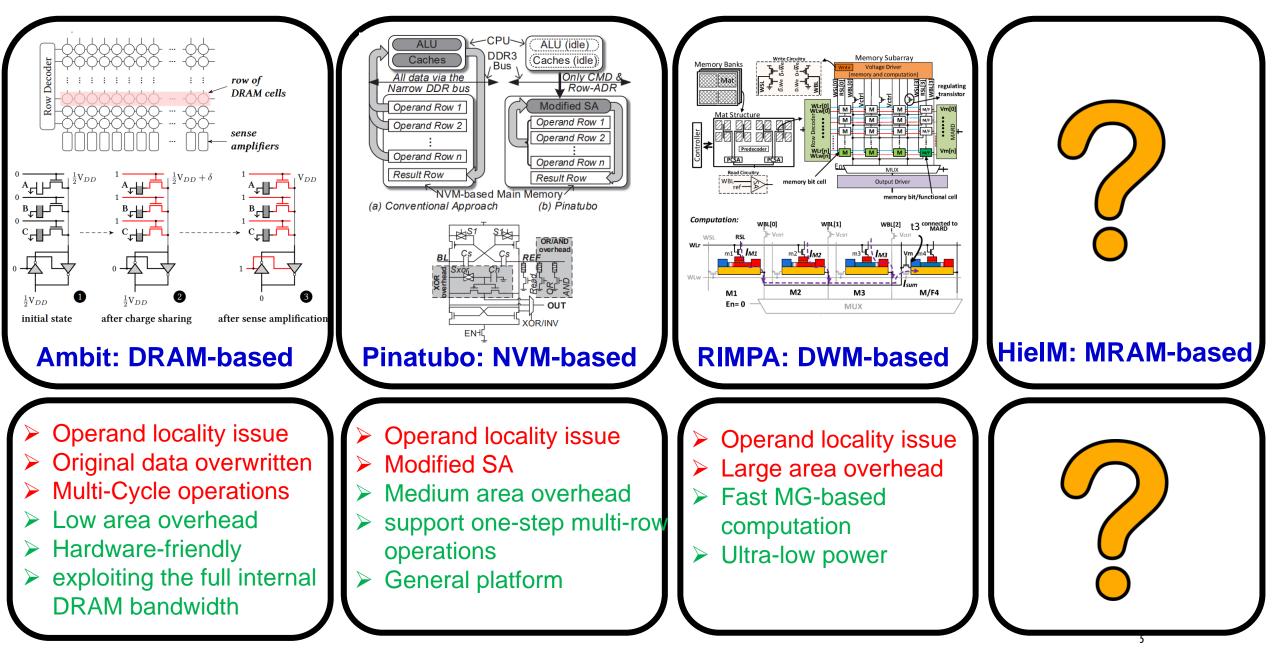
- Energy efficient and high performance computing hardware development is beginning to stall fundamentally due to limitations in both *devices* and *architectures*.
- First, the current computing platforms primarily depend on Complementary Metal Oxide Semiconductor (CMOS) technology, which is reaching its power wall





• There is an urgent need to investigate fundamentally different devices and architectures for information processing and data storage with the ability to continuously deliver energy efficient and high performance computing solutions.

RECENT IN-MEMORY COMPUTING PLATFORMS



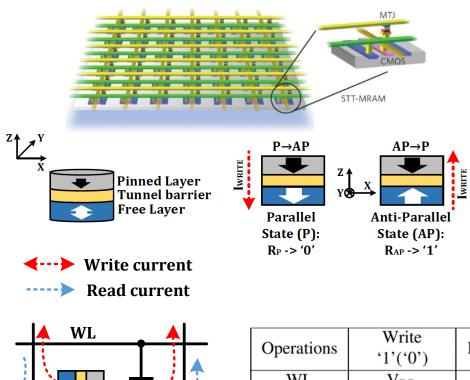


- Motivation
- Post-CMOS Spintronic Devices
- In-Memory Processing Platform based on STT-MRAM
- Performance Evaluation
- Case Study I: In-memory Bulk Bitwise Vector Operation
- Case Study II: In-memory Data Encryption Engine



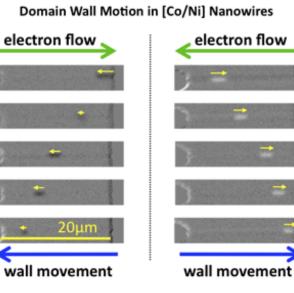
POST-CMOS SPINTRONIC DEVICES

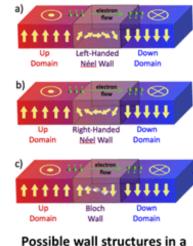
STT-MRAM



Operations	Write '1'('0')	Read
WL	V _{DD}	V _{DD}
BL	GND (V _{DD})	Isense
SL	V _{DD} (GND)	GND

Domain Wall Motion Device





magnetic nanowire

7



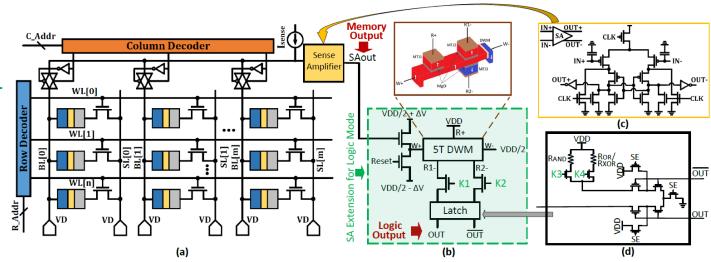
- Motivation
- Post-CMOS Spintronic Devices
- In-Memory Processing Platform based on STT-MRAM
- Performance Evaluation
- Case Study I: In-memory Bulk Bitwise Vector Operation
- Case Study II: In-memory Data Encryption Engine



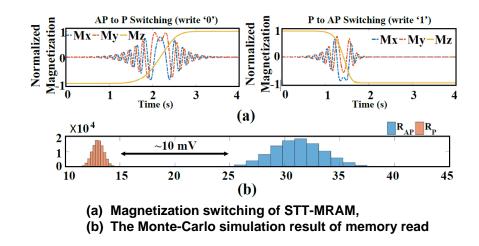
Operations	Write '1'('0')	Read
WL	V _{DD}	V _{DD}
BL	GND (V _{DD})	Isense
SL	V _{DD} (GND)	GND

IN-MEMORY PROCESSING PLATFORM

- Dual mode architecture that perform both memory readwrite and in-memory logic (AND/NAND, OR/NOR, XOR/XNOR).
- Memory Write: To write data in a memory cell, the corresponding WL is activated using the row decoder. Then appropriate voltage difference is applied to the corresponding BL and SL using the voltage drivers.
- Memory Read: The corresponding WL is activated using the row decoder and the corresponding BL is connected to the sense amplifier (SA) using the column decoder.
- Computing Mode: We propose a sensing circuit design using 5T DWM device [1], as an extension to SA of memory array, to implement complete Boolean logic functions between any two cells in the memory array.

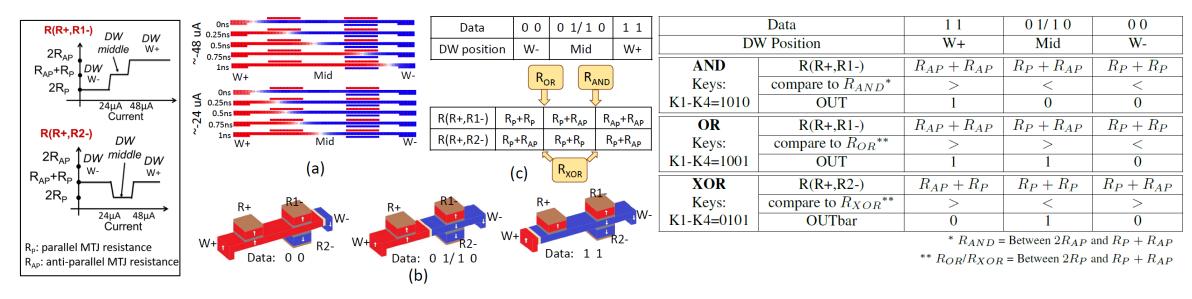


(a) HielM, (b) proposed sensing scheme, (c) Memory sense amplifier, (d) Differential Latch.

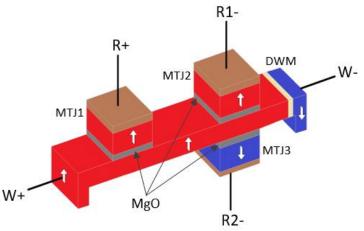




IN-MEMORY PROCESSING PLATFORM



- For a complete Boolean operation, the SA extension needs 3 subsequent stages-Reset, Compute and Sense.
- In Reset stage (Reset=1), the reset transistor is turned on for 1ns. A current of 48uA flows from W- to W+ terminals, which sets the DW back to its initial position at W- side.
- In Compute stage, two operands stored in the memory array are read in two consecutive cycles using the SA and applied to DWM device.
- In Sense stage, a small sensing current is injected through DWM device from R+ to R1or from R+ to R2- terminals based on required logic implementation.





- Motivation
- Post-CMOS Spintronic Devices
- In-Memory Processing Platform based on STT-MRAM
- Performance Evaluation
- Case Study I: In-memory Bulk Bitwise Vector Operation
- Case Study II: In-memory Data Encryption Engine

PERFORMANCE EVALUATION

Device to System Level Simulations:

Device Level:

Verilog-A model of 5T DWM device was developed to co-simulate with the interface CMOS circuits in SPICE to validate the functionality and evaluate performance of the proposed design. The STT-MRAM is simulated by solving LLG equation to model dynamics of MTJ free layer.

Circuit Level:

45nm North Carolina State University (NCSU) Product Development Kit (PDK) [1] library is used in SPICE to verify the proposed design and evaluate the performance.

System Level:

We employ the modified self-consistent NVSim [2] along with an in-house developed C++ code to verify the performance of memory.

[1] www:eda.ncsu.edu/wiki/FreePDK45

[2] X. Dong et al., "NVSim: A circuit-level performance, energy, and area model for emerging non-volatile memory," Springer, 2014, pp. 15-50.



PERFORMANCE EVALUATION

Memory Mode:

- The proposed STT-MRAM memory model shows the least write dynamic energy in comparison to other designs.
- It reduces the total leakage power compared to SRAM.
- It shows longer average latency compared to SRAM due to the longer write latency of magnetic memory storage.
- Its area overhead is 29.1% more than DRAM but still 37.51% less than SRAM design.

SRAM, DRAM AND PROPOSED STT-MRAM MEMORY MODEL VALIDATION AND COMPARISON FOR A SAMPLE 4MB MEMORY

	SRAM	STT-MRAM
Metrics W	Write Read	d Write Read
ge Latency (ns)	1.76	2.67
nic Energy (pJ) 1	1213 1483	7 826.15 870.042
ge Power (mW)	5316	830.847
rea (mm^2)	10.613	6.632

[27] K. Huang et al. Magnetic domain-wall racetrack memory-based nonvolatile logic for low-power computing and fast run-timereconfiguration. 2016 [28] K. Huang et al. Stt-mram based low power synchronous non-volatile logic with timing demultiplexing. In NANOARCH, pages 31–36. ACM, 2014

[29] H.-P. Trinh et al. Magnetic adder based on racetrack memory. IEEE TCAS I, 60(6):1469–1477, 2013.

[30] A. Roohi et al. A tunable majority gate-based full adder using currentinduced domain wall nanomagnets. IEEE Trans. Magn., 52(8):1–7, 2016.

[31] Y. Zhang et al. Giant spin hall effect (gshe) logic design for lowpower application. In DATE, pages 1000–1005, 2015.

Computing Mode:

- The in-memory AND operation shows 65.3% and 81.32% lower energy consumption than Domain-Wall (DW) Racetrack based and MTJ based in-memory non-volatile AND gate implementations.
- Our design requires longer latency to compute the logic result than other designs

PERFORMANCE EVALUATION AND GATES

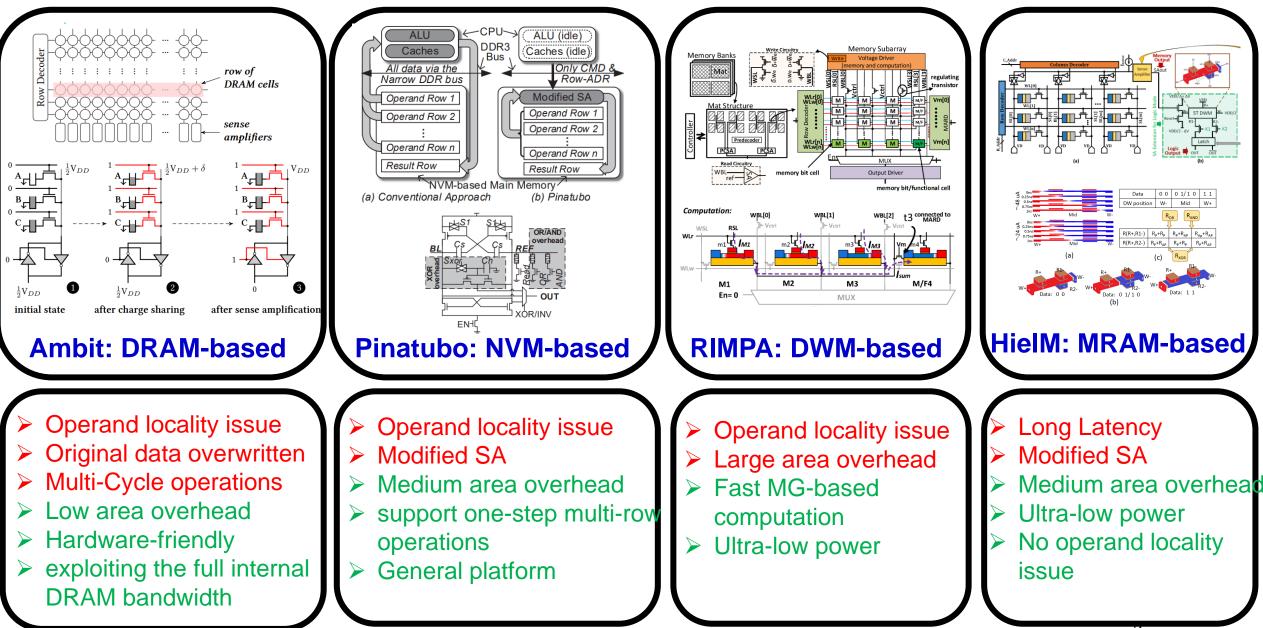
Performance	HieIM	DW Racetrack based [27]	MTJ based [28]	DW Racetrack based [29]	CMOS
Energy (fJ)	23.5	67.72	125.85	504.36	6.69
Speed (ns)	4	1.12	1.18	2.14	0.062

 The in-memory FA implementation is comparable to that of LPM based FA design. However, our design requires longer delay due to the read-and-write-back overhead of the intermediate results.

PERFORMANCE EVALUATION OF FA CELLS

	Parameters	HieIM	HSM [30]	LPM [30]	Diode- GSHE [31]	CMOS [31]	
	Average Power (μW)	91.93	1354	85	15.6	49.4	
201	Delay (ps)	26,000	269	877	10,000	1000	

LETS FILL IT



4

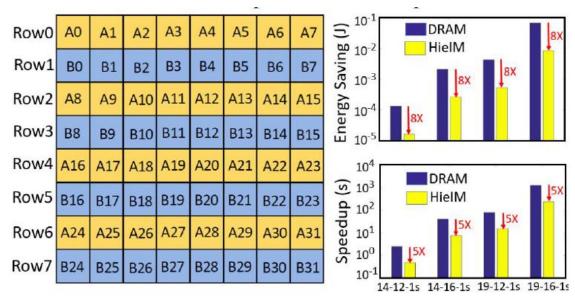


- Motivation
- Post-CMOS Spintronic Devices
- In-Memory Processing Platform based on STT-MRAM
- Performance Evaluation
- Case Study I: In-memory Bulk Bitwise Vector Operation
- Case Study II: In-memory Data Encryption Engine



CASE STUDY I: IN-MEMORY BULK BITWISE VECTOR OPERATION

- Four different vector datasets [1] have been used. Here, a dataset '19-16-1s' refers to a vector dataset with vector length= 2^{19} , number of vectors= 2^{16} and AND/OR operation is done between 2^{1} rows.
- Each compute (AND/OR) operation has been carried out using 4 consecutive clock cycles (1ns each).
- HielM offers ~8× energy saving and ~5× speed up compared to that using Ambit-DRAM based in-memory computing platform [2].



Data mapping for performing vector operation between two 32 bit vectors using an 8*8 STT-MRAM array



- Motivation
- Post-CMOS Spintronic Devices
- In-Memory Processing Platform based on STT-MRAM
- Performance Evaluation
- Case Study I: In-memory Bulk Bitwise Vector Operation
- Case Study II: In-memory Data Encryption Engine

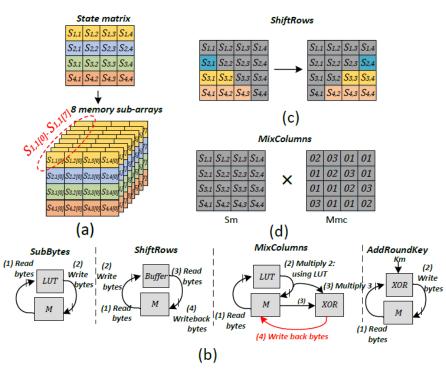


CASE STUDY II: IN-MEMORY DATA ENCRYPTION ENGINE

- Advanced Encryption Standard (AES) has been used to employ in-memory data encryption engine using HielM.
- HielM can achieve 51.5% and 68.9% lower energy consumption compared to CMOS-ASIC and CMOL based implementations, respectively.
- HielM occupies ~3.5× less area compared to baseline DW-AES.
- Note that, Baseline DW AES [36] requires lower number of cycles due to intrinsic shift operation and multi-bit data storage of DWM racetrack devices.

Platforms	Energy (nJ)	Cycles	Area (μm^2)
GPP [37]	460	2309	2.5e+6
ASIC [41]	6.6	336	4400
CMOL[42]	10.3	470	320
Baseline DW [36]	2.4	1022	78
Pipelined DW [36]	2.3	2652	83
Multi-issue DW [36]	2.7	1320	155
HieIM	3.2	1620	21.8

AES PERFORMANCE



(a) Data Organization, (b) Data Mapping of four AES transformations, (c) ShiftRows transformation, (d) MixColumn transformation.

[36] Y. Wang et al. Dw-aes: a domain-wall nanowire-based aes for high throughput and energy-efficient data encryption in non-volatile memory. IEEE TIFS, 11(11):2426–2440, 2016.

[37] K Malbrain. Byte-oriented-aes: a public domain byte-oriented implementation of aes in c, 2009.

[41] S. Mathew et al. 340 mv-1.1 v, 289 gbps/w, 2090-gate nanoaes hardware accelerator with area-optimized encrypt/decrypt gf (2 4) 2 polynomials in 22 nm tri-gate cmos. IEEE JSSC, 50(4):1048-1058, 2015. [42] Z Abid et al. Efficient cmol gate designs for cryptography applications. IEEE TNANO, 8:315-321, 2009.



CONCLUSION

 In this work, we develop a new in-memory processing architecture based on STT-MRAM called HeilM, which could be used as both non-volatile memory and reconfigurable in-memory logic.

 HeilM offers several significant features as non-volatility, in-memory logic, operation with high data mapping flexibility, low dynamic power consumption and high packing density.

 The in-memory AND operation itself shows 65.3% and 81.32% lower energy consumption than Domain-Wall (DW) Racetrack based and MTJ based in-memory non-volatile AND implementations.

 In-memory bulk bitwise Boolean vector logic (AND/OR) operation for different vector datasets ~8× energy saving and ~5× speed up compared to that using DRAM based in-memory computing platform.

 We further have employed in-memory data encryption engine using AES algorithm, which shows 51.5% and 68.9% lower energy consumption compared to CMOS-ASIC and CMOL-based implementations, respectively.



THANKS