Ambit: In-Memory Accelerator for Bulk Bitwise Operations Using Commodity DRAM Technology

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Moritz Knüsel

Seminar on Computer Architecture

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 - less than 1% area overhead
- Results compared to state-of-the-art:
 - 32x performance improvement, 35x energy reduction averaged across 7 bulk bitwise operations
 - 3x-7x performance improvement for real-world data intensive workloads

Background, Problem & Goal

• Applications that rely on bulk bitwise operations include:

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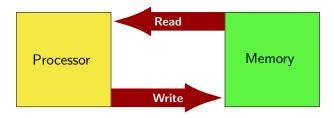
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- Bulk bitwise operations become problematic if
 - The bitvectors involved are very large
 - They cannot be combined with other processing

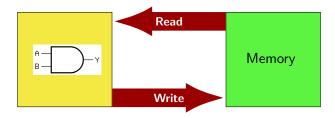
Problem

Memory bandwidth is a bottleneck

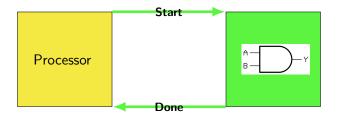


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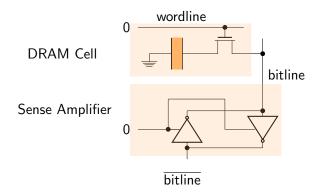
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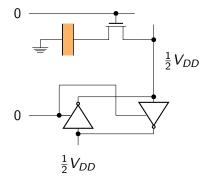


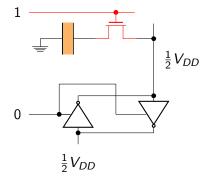
Idea: Perform bitwise operations directly in DRAM

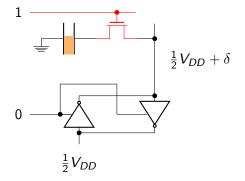


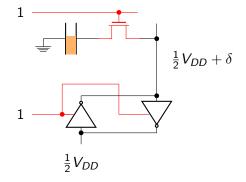
Mechanism

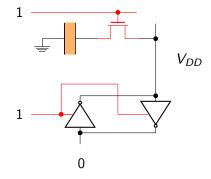












• Ambit-AND-OR relies on analog charge sharing.

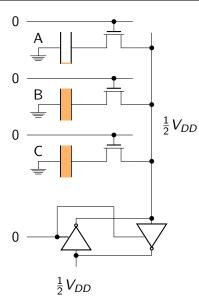
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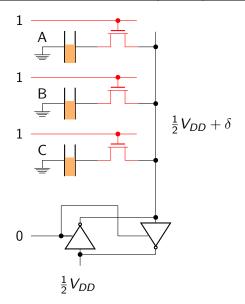
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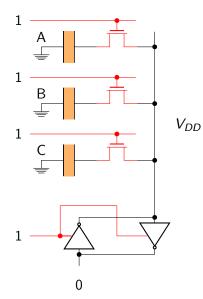
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• Key observation: By activating three rows at once, the bitwise majority of the three rows is computed







С	A	В	Bitwise Majority
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	1

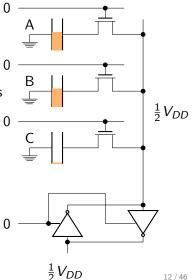
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0	0	1	0	
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0	1	1	1	
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1	0	1	1	
1	1	0	1	
1	1	1	1	

С	А	В	Bitwise Majority	
0	0	0		0
0	0	1	A AND B	0
0	1	0		0
0	1	1		1
1	0	0		0
1	0	1	A OR B	1
1	1	0		1
1	1	1		1

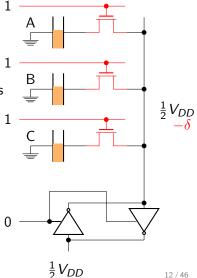
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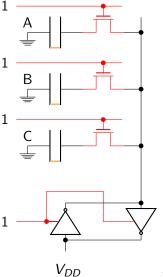
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- All rows in a TRA have been refreshed recently

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- Ambit relies on RowClone for copying data and initializing rows

Vivek Seshadri, Yoongu Kim, Chris Fallin, Donghyuk Lee, Rachata Ausavarungnirun, Gennady Pekhimenko, Yixin Luo, Onur Mutlu, Michael A. Kozuch, Phillip B. Gibbons, and Todd C. Mowry **RowClone: Fast and Energy-Efficient In-DRAM Bulk**

Data Copy and Initialization

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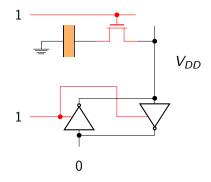
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- How can we transmit a command to activate three rows without sending and decoding 3 addresses?
- The solution: We use a **reserved address** to communicate a TRA on the designated rows
- Also, we can split up the row decoder into two parts, which leads to a simpler design and better performance

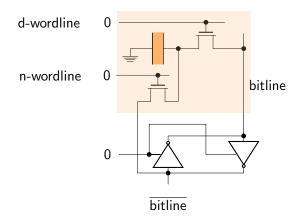
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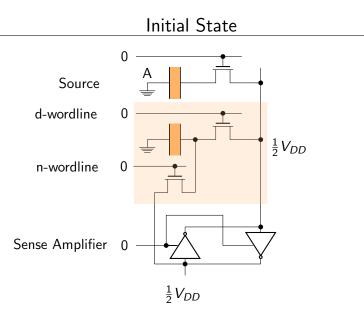
Ambit-NOT

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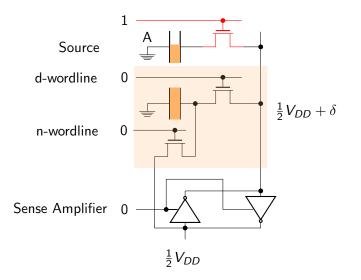
Notice that, during normal DRAM operation, the voltage level of bitline is the negation of the value in the cell.

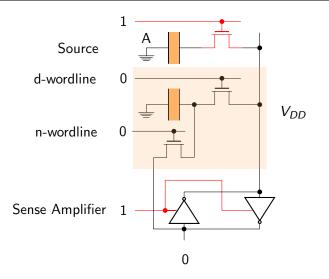


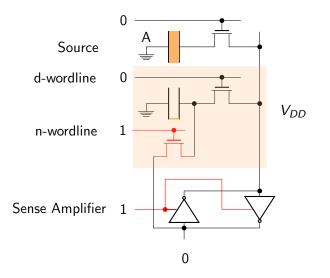




After Charge Sharing



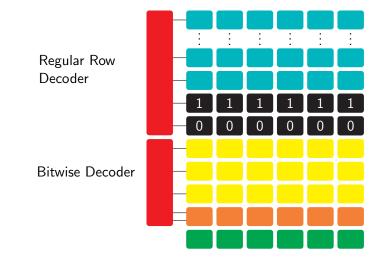




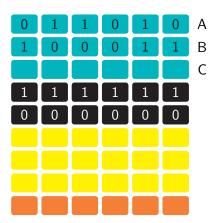
Implementation

Regular DRAM cells ÷ 2 ŝ ÷ . Pre-Initialized Rows Designated Rows for TRA Dual Contact Cells Sense Amplifiers

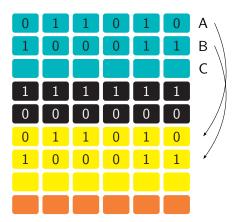
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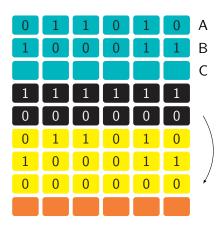
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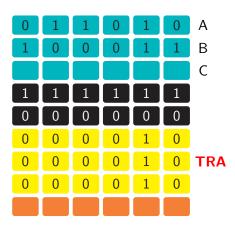
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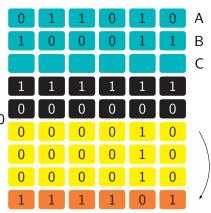
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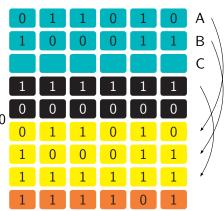
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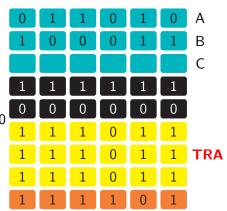
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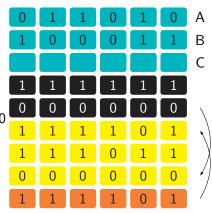
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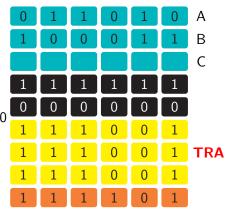
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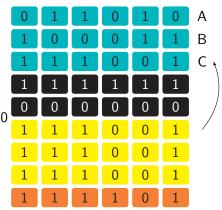
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- Directly plugged onto the memory bus
 - Makes sense since Ambit uses the same interface as regular DRAM
 - However, this requires additional support such as a new CPU instruction

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- The CPU checks the constraints. If they are met, Ambit is used to complete the operation. Otherwise, the CPU does the operation normally.

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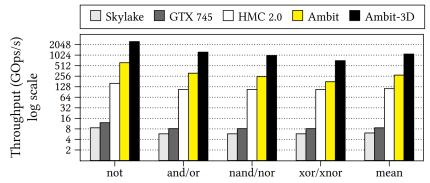
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 - Alternatively, the bbop instruction could implicitly manage the caches as well

Key Results: Methodology & Evaluation

Microbenchmarks with 32MB input vectors



- Energy is estimated for DDR3-1333 using the Rambus power model
- Energy numbers include only the DRAM and channel energy, and not the energy consumed by the processor

DIAM & Channel Lifergy (11/ KD)						
Design	not	and/or	nand/nor	xor/xnor		
DDR3	93.7	137.9	137.9	137.9		
Ambit	1.6	3.2	4.0	5.5		
Energy reduction	59.5 x	43.9x	35.1x	25.1x		

DRAM & Channel Energy (nJ/KB)

Evaluations were carried out using the Gem5 full-system simulator. Major simulation parameters:

Processor	x86, 8-wide, out-of-order, 4 Ghz		
	64-entry instruction queue		
L1 cache	32 KB D-cache, 32 KB I-cache, LRU policy		
L2 cache	2 MB, LRU policy, 64 B cache line size		
Memory Controller	8 KB row size, FR-FCFS scheduling		
Main memory	DDR4-2400, 1-channel, 1-rank, 16 banks		

Workloads:

- Set Operations Comparing to bitvectors and red-black trees
- Bitmap Indices
- Bitweaving Column scans using bulk bitwise operations

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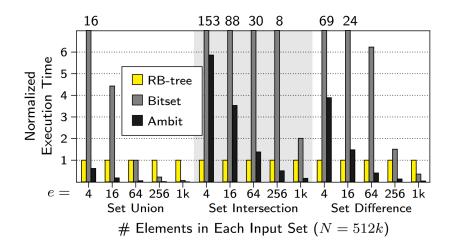
F

a tree

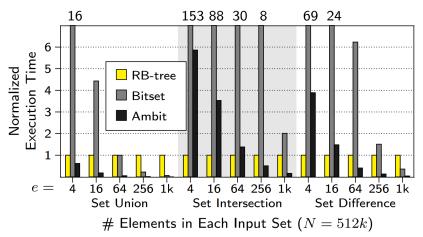
В

or a bitvector 011001

• Set operations on trees scale with the number of elements in the set, whereas bitvectors also have to process elements that are not in the set



• Ambit shifts the balance heavily in favor of bitvectors



• Using bitsets to build database indices

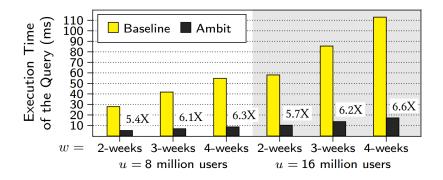
- Using bitsets to build database indices
- As an example, consider this table:

USER_ID	REGION	INCOME_LEVEL
101	east	bracket_1
102	central	$bracket_1$
103	west	bracket_2
104	east	bracket_2

 A bitmap for the REGION and INCOME_LEVEL columns might look like this:

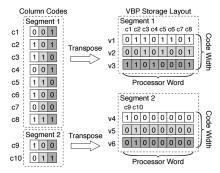
REGION			INCOME	
east	central	west	$bracket_1$	bracket_2
1	0	0	1	0
0	1	0	1	0
0	0	1	0	1
1	0	0	0	1

Running an analytic query using a bitmap index. Each query takes O(w) bulk bitwise operations, each of which takes O(u) time.



Workload: Bitweaving

- Used to speed up predicate evaluation in databases
- Tables are stored columnwise, but at the bitlevel

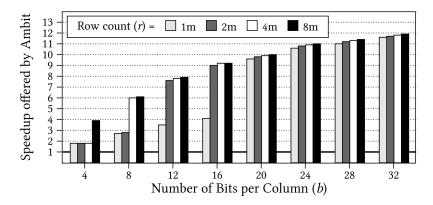


 For details on how it works, refer to BitWeaving: Fast
 Scans for Main Memory Data Processing by Yinan Li and Jignesh M. Patel

Bitweaving Performance

• Evaluated Query:

select count(*) from T where c1 <= val <= c2



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- 43 citations over the last year

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• DRISA: A DRAM-based Reconfigurable In-Situ Accelerator (2017) Shuangchen Li, Dimin Niu, Krishna T. Malladi, Hongzhong Zheng, Bob

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Shuangchen Li, Dimin Niu, Krishna T. Malladi, Hongzhong Zheng, Bob Brennan, Yuan Xie

- Adds more mechanism for data movement, such as bit shifting

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- Add more processing capabilites to DRAM
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• Processing in other kinds of memory

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- In the cache:
 - Compute Caches (2017)

Shaizeen Aga, Supreet Jeloka, Arun Subramaniyan, Satish Narayanasamy, David Blaauw, Reetuparna Das

Related Work

- Processing in other kinds of memory
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 - **Compute Caches** (2017) Shaizeen Aga, Supreet Jeloka, Arun Subramaniyan, Satish Narayanasamy, David Blaauw, Reetuparna Das
 - Neural Cache: Bit-Serial In-Cache Acceleration of Deep Neural Networks (2018)

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• In non-volatile memory:

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- In non-volatile memory:
 - Pinatubo: A Processing-in-Memory Architecture for Bulk Bitwise Operations in Emerging Non-volatile Memories (2016)
 Shuangchen Li, Cong Xu, Qiaosha Zou, Jishen Zhao, Yu Lu, Yuan Xie

Questions?

How could we better support bitvectors whose length is not a multiple of the row size?

• A bitmask could be used to preserve part of a row

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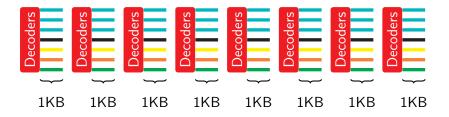
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- But: Even if mask creation is cheap, it still requires 4
 additional operations to mask off a single useful operation

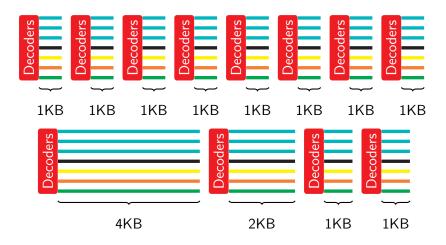
Add a few smaller subarrays



8KB

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What kind of changes thoughout the system might be necessary to make Ambit useful for applications?

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- Applications need to deal with vectors with bad lengths