Future Computing Architectures Challenges and Opportunities

(with a Large Focus on Memory)

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Euromicro PDP 2018 Keynote Talk





Brief Self Introduction



Onur Mutlu

- Full Professor @ ETH Zurich CS, since September 2015 (officially May 2016)
- Strecker Professor @ Carnegie Mellon University ECE/CS, 2009-2016, 2016-...
- PhD from UT-Austin, worked at Google, VMware, Microsoft Research, Intel, AMD
- https://people.inf.ethz.ch/omutlu/
- omutlu@gmail.com (Best way to reach me)
- https://people.inf.ethz.ch/omutlu/projects.htm

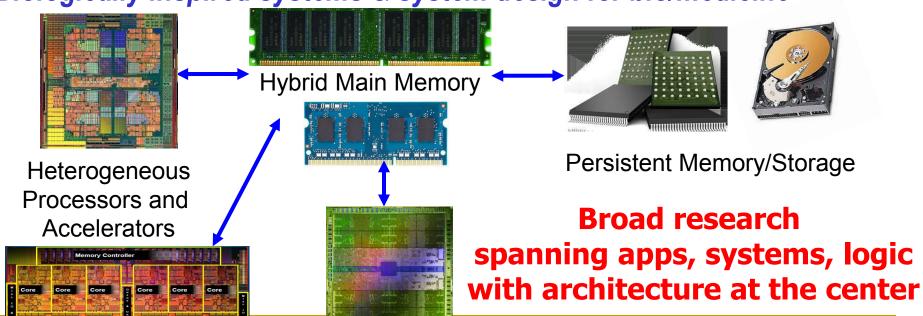
Research and Teaching in:

- Computer architecture, computer systems, security, bioinformatics
- Memory and storage systems
- Hardware security
- Fault tolerance
- Hardware/software cooperation
- ...

Current Research Focus Areas

Research Focus: Computer architecture, HW/SW, bioinformatics

- Memory and storage (DRAM, flash, emerging), interconnects
- · Heterogeneous & parallel systems, GPUs, systems for data analytics
- System/architecture interaction, new execution models, new interfaces
- Hardware security, energy efficiency, fault tolerance, performance
- Genome sequence analysis & assembly algorithms and architectures
- Biologically inspired systems & system design for bio/medicine



Graphics and Vision Processing

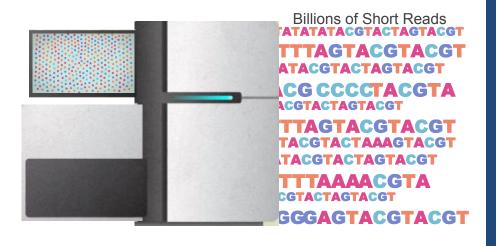
Four Key Directions

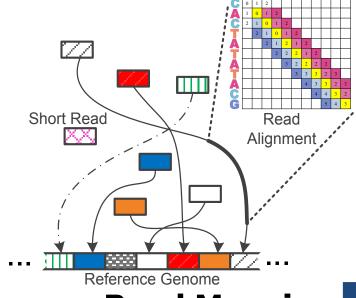
Fundamentally Secure/Reliable/Safe Architectures

- Fundamentally Energy-Efficient Architectures
 - Memory-centric (Data-centric) Architectures

Fundamentally Low-Latency Architectures

Architectures for Genomics, Medicine, Health

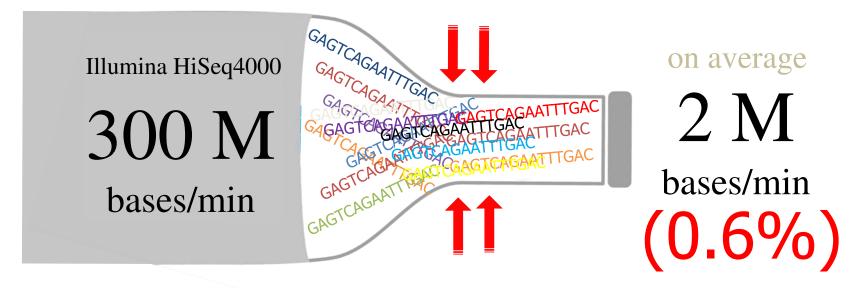




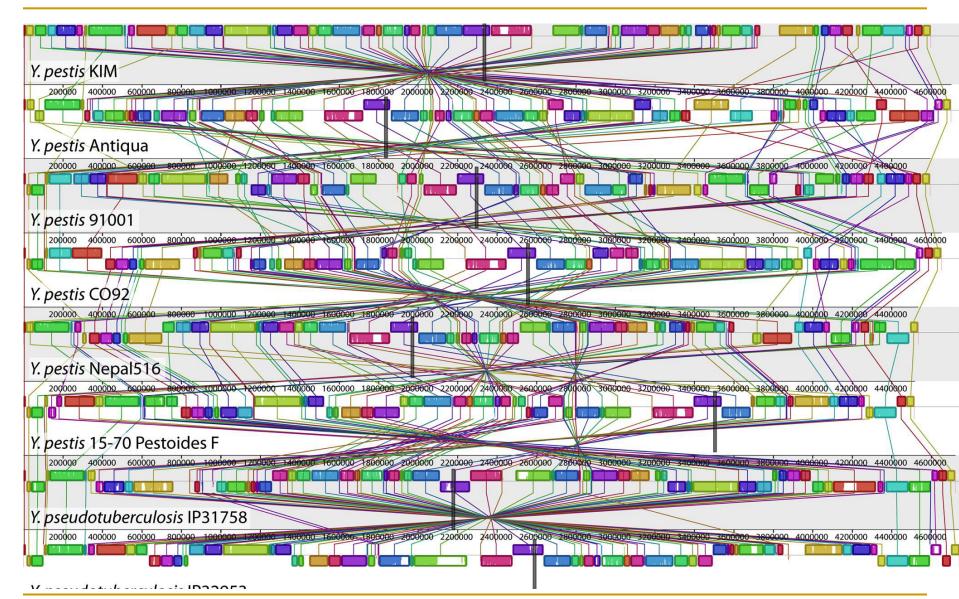
Read Mapping

Sequencing

Bottlenecked in Mapping!!



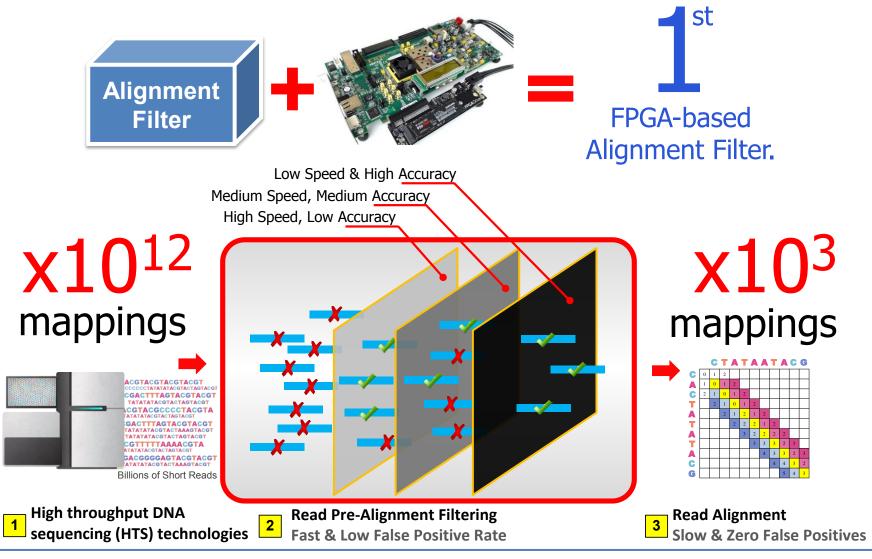
Genome Sequence Alignment: Example



Filter fast before you align

Minimize costly "approximate string comparisons"

An Example Solution: GateKeeper



FPGA-Based Alignment Filtering

 Mohammed Alser, Hasan Hassan, Hongyi Xin, Oguz Ergin, Onur Mutlu, and Can Alkan

"GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in DNA Short Read Mapping" Bioinformatics, [published online, May 31], 2017.

Source Code

Online link at Bioinformatics Journal

GateKeeper: a new hardware architecture for accelerating pre-alignment in DNA short read mapping

Mohammed Alser ™, Hasan Hassan, Hongyi Xin, Oğuz Ergin, Onur Mutlu ™, Can Alkan ™

Bioinformatics, Volume 33, Issue 21, 1 November 2017, Pages 3355–3363,

https://doi.org/10.1093/bioinformatics/btx342

Published: 31 May 2017 Article history ▼

In-Memory DNA Sequence Analysis

Jeremie S. Kim, Damla Senol Cali, Hongyi Xin, Donghyuk Lee, Saugata Ghose, Mohammed Alser, Hasan Hassan, Oguz Ergin, Can Alkan, and Onur Mutlu, "GRIM-Filter: Fast Seed Location Filtering in DNA Read Mapping Using Processing-in-Memory Technologies" to appear in <u>BMC Genomics</u>, 2018. to also appear in Proceedings of the <u>16th Asia Pacific Bioinformatics</u> <u>Conference</u> (APBC), Yokohama, Japan, January 2018. arxiv.org Version (pdf)

GRIM-Filter: Fast Seed Location Filtering in DNA Read Mapping Using Processing-in-Memory Technologies

Jeremie S. Kim^{1,6*}, Damla Senol Cali¹, Hongyi Xin², Donghyuk Lee³, Saugata Ghose¹, Mohammed Alser⁴, Hasan Hassan⁶, Oguz Ergin⁵, Can Alkan*⁴, and Onur Mutlu*^{6,1}

Key Principles and Results

- Two key principles:
 - Exploit the structure of the genome to minimize computation
 - Morph and exploit the structure of the underlying hardware to maximize performance and efficiency
- Algorithm-architecture co-design for DNA read mapping
 - Speeds up read mapping by 20-200X
 - Improves accuracy of read mapping in the presence of errors

Xin et al., "Accelerating Read Mapping with FastHASH," BMC Genomics 2013.

Xin et al., "Shifted Hamming Distance: A Fast and Accurate SIMD-friendly Filter to Accelerate Alignment Verification in Read Mapping," Bioinformatics 2015.

Alser et al., "GateKeeper: A New Hardware Architecture for Accelerating Pre-Alignment in DNA Short Read Mapping," Bioinformatics 2017.

Kim et al., "Genome Read In-Memory (GRIM) Filter," BMC Genomics 2018.

New Genome Sequencing Technologies

Nanopore Sequencing Technology and Tools for Genome Assembly: Computational Analysis of the Current State, Bottlenecks, and Future Directions

Damla Senol Cali 1,*, Jeremie Kim 1,3, Saugata Ghose 1, Can Alkan 2* and Onur Mutlu 3,1*

Senol Cali+, "Nanopore Sequencing Technology and Tools for Genome Assembly: Computational Analysis of the Current State, Bottlenecks and Future Directions," to appear in Briefings in Bioinformatics, 2018.

[Preliminary arxiv.org version]

¹Department of Electrical and Computer Engineering, Carnegie Mellon University, Pittsburgh, PA, USA

²Department of Computer Engineering, Bilkent University, Bilkent, Ankara, Turkey

³Department of Computer Science, Systems Group, ETH Zürich, Zürich, Switzerland

Nanopore Genome Assembly Pipeline

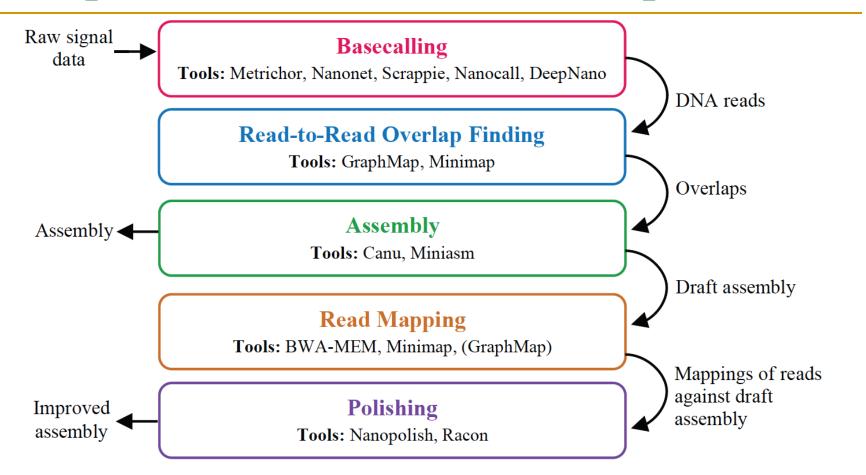


Figure 1. The analyzed genome assembly pipeline using nanopore sequence data, with its five steps and the associated tools for each

___step.

Senol Cali+, "Nanopore Sequencing Technology and Tools for Genome Assembly" to appear in Briefings in Bioinformatics, 2018.

More on Genome Analysis: Another Talk

Accelerating Genome Analysis

A Primer on an Ongoing Journey

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March 8, 2018

ETH HAML Seminar

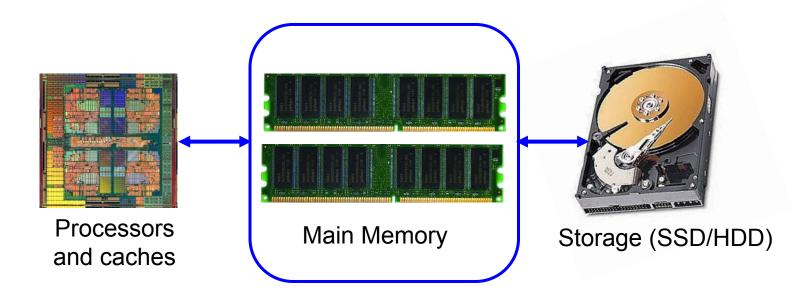






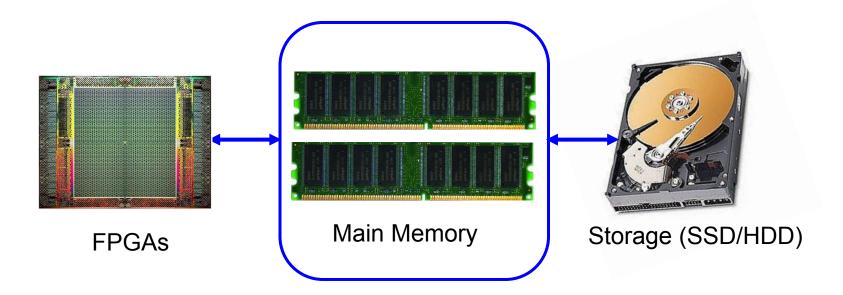
Memory & Storage

The Main Memory System



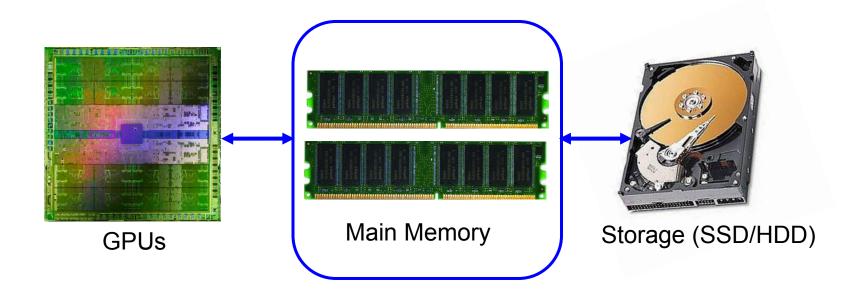
- Main memory is a critical component of all computing systems: server, mobile, embedded, desktop, sensor
- Main memory system must scale (in size, technology, efficiency, cost, and management algorithms) to maintain performance growth and technology scaling benefits

The Main Memory System



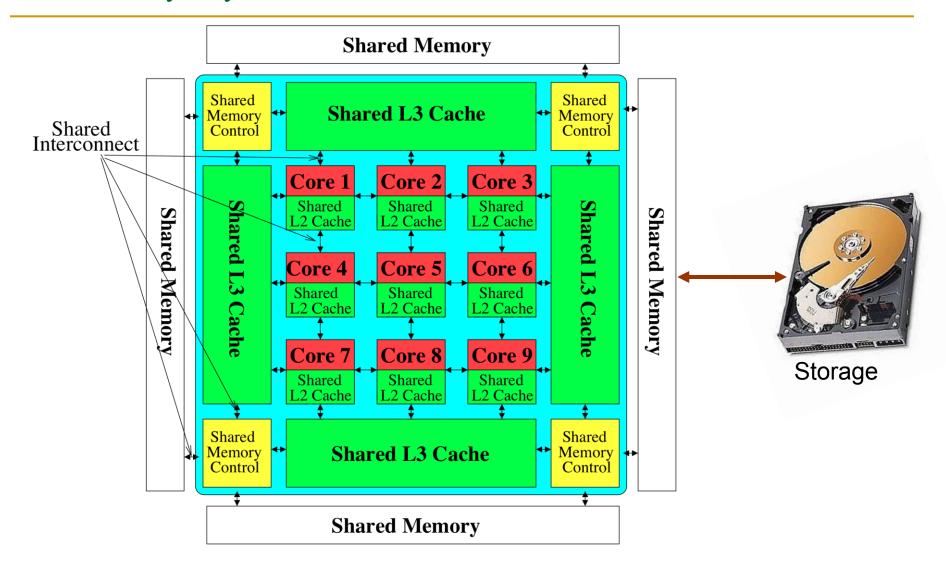
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Memory System: A *Shared Resource* View



Most of the system is dedicated to storing and moving data

State of the Main Memory System

- Recent technology, architecture, and application trends
 - lead to new requirements
 - exacerbate old requirements
- DRAM and memory controllers, as we know them today, are (will be) unlikely to satisfy all requirements
- Some emerging non-volatile memory technologies (e.g., PCM) enable new opportunities: memory+storage merging
- We need to rethink the main memory system
 - to fix DRAM issues and enable emerging technologies
 - to satisfy all requirements

Agenda

- Major Trends Affecting Main Memory
- The Memory Scaling Problem and Solution Directions
 - New Memory Architectures
 - Enabling Emerging (NVM) Technologies
- Cross-Cutting Principles
- Summary

Major Trends Affecting Main Memory (I)

Need for main memory capacity, bandwidth, QoS increasing

Main memory energy/power is a key system design concern

DRAM technology scaling is ending

Major Trends Affecting Main Memory (II)

- Need for main memory capacity, bandwidth, QoS increasing
 - Multi-core: increasing number of cores/agents
 - Data-intensive applications: increasing demand/hunger for data
 - Consolidation: cloud computing, GPUs, mobile, heterogeneity

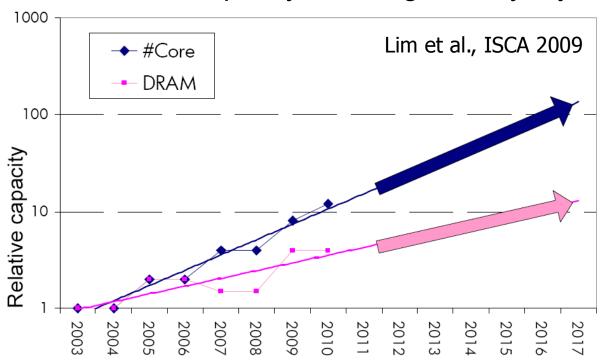
Main memory energy/power is a key system design concern

DRAM technology scaling is ending

Example: The Memory Capacity Gap

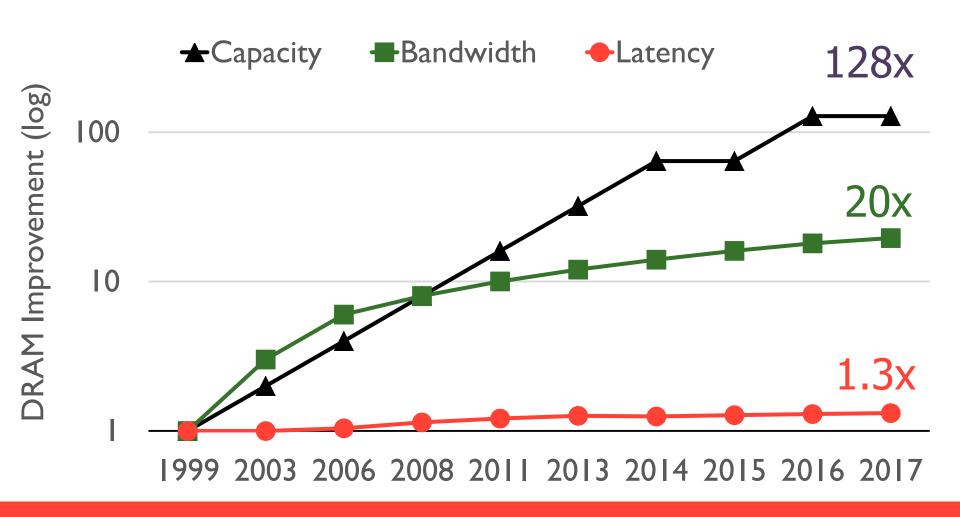
Core count doubling ~ every 2 years

DRAM DIMM capacity doubling ~ every 3 years



- Memory capacity per core expected to drop by 30% every two years
- Trends worse for memory bandwidth per core!

DRAM Capacity, Bandwidth, Latency Trends



Memory latency remains almost constant

DRAM Latency Is Critical for Performance



In-memory Databases

[Mao+, EuroSys'12; Clapp+ (Intel), IISWC'15]



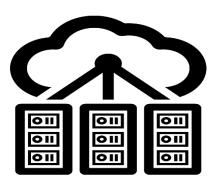
In-Memory Data Analytics

[Clapp+ (Intel), IISWC'15; Awan+, BDCloud'15]



Graph/Tree Processing

[Xu+, IISWC'12; Umuroglu+, FPL'15]



Datacenter Workloads

[Kanev+ (Google), ISCA' I 5]

DRAM Latency Is Critical for Performance



In-memory Databases



Graph/Tree Processing

Long memory latency → performance bottleneck



In-Memory Data Analytics

[Clapp+ (Intel), IISWC'15; Awan+, BDCloud'15]



Datacenter Workloads

[Kanev+ (Google), ISCA' I 5]



Major Trends Affecting Main Memory (III)

Need for main memory capacity, bandwidth, QoS increasing

- Main memory energy/power is a key system design concern
 - □ ~40-50% energy spent in off-chip memory hierarchy [Lefurgy, IEEE Computer'03] >40% power in DRAM [Ware, HPCA'10][Paul,ISCA'15]
 - DRAM consumes power even when not used (periodic refresh)
- DRAM technology scaling is ending

Major Trends Affecting Main Memory (IV)

Need for main memory capacity, bandwidth, QoS increasing

Main memory energy/power is a key system design concern

- DRAM technology scaling is ending
 - ITRS projects DRAM will not scale easily below X nm
 - Scaling has provided many benefits:
 - higher capacity (density), lower cost, lower energy

Major Trends Affecting Main Memory (V)

- DRAM scaling has already become increasingly difficult
 - Increasing cell leakage current, reduced cell reliability, increasing manufacturing difficulties [Kim+ ISCA 2014], [Liu+ ISCA 2013], [Mutlu IMW 2013], [Mutlu DATE 2017]
 - Difficult to significantly improve capacity, energy
- Emerging memory technologies are promising

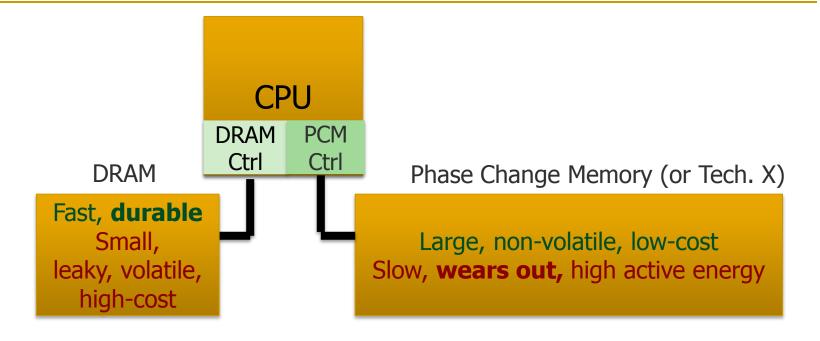
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Major Trends Affecting Main Memory (V)

- DRAM scaling has already become increasingly difficult
 - Increasing cell leakage current, reduced cell reliability, increasing manufacturing difficulties [Kim+ ISCA 2014], [Liu+ ISCA 2013], [Mutlu IMW 2013], [Mutlu DATE 2017]
 - Difficult to significantly improve capacity, energy
- Emerging memory technologies are promising

3D-Stacked DRAM	higher bandwidth	smaller capacity
Reduced-Latency DRAM (e.g., RL/TL-DRAM, FLY-RAM)	lower latency	higher cost
Low-Power DRAM (e.g., LPDDR3, LPDDR4, Voltron)	lower power	higher latency higher cost
Non-Volatile Memory (NVM) (e.g., PCM, STTRAM, ReRAM, 3D Xpoint)	larger capacity	higher latency higher dynamic power lower endurance

Major Trend: Hybrid Main Memory



Hardware/software manage data allocation and movement to achieve the best of multiple technologies

Meza+, "Enabling Efficient and Scalable Hybrid Memories," IEEE Comp. Arch. Letters, 2012. Yoon+, "Row Buffer Locality Aware Caching Policies for Hybrid Memories," ICCD 2012 Best Paper Award.



Main Memory Needs Intelligent Controllers

Agenda

- Major Trends Affecting Main Memory
- The Memory Scaling Problem and Solution Directions
 - New Memory Architectures
 - Enabling Emerging (NVM) Technologies
- Cross-Cutting Principles
- Summary

Four Key Issues in Future Platforms

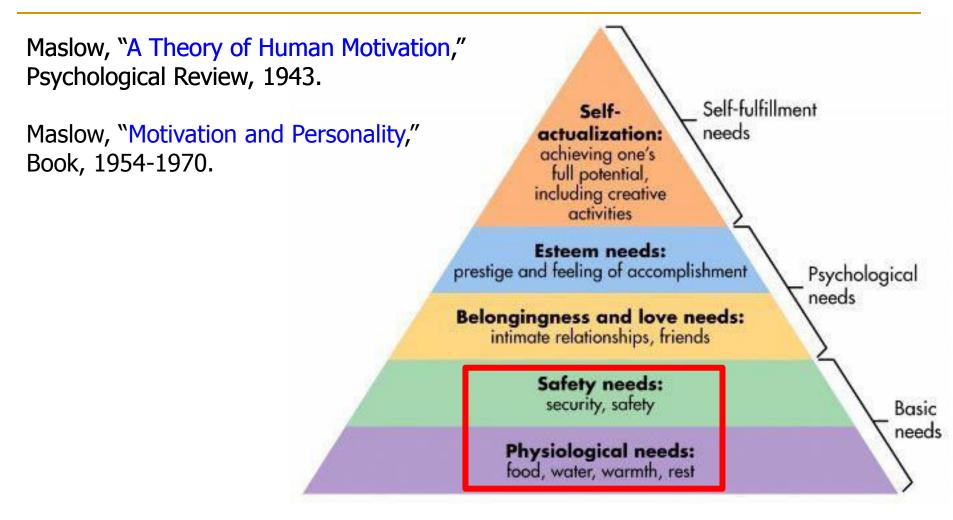
Fundamentally Secure/Reliable/Safe Architectures

- Fundamentally Energy-Efficient Architectures
 - Memory-centric (Data-centric) Architectures

Fundamentally Low-Latency Architectures

Architectures for Genomics, Medicine, Health

Maslow's (Human) Hierarchy of Needs

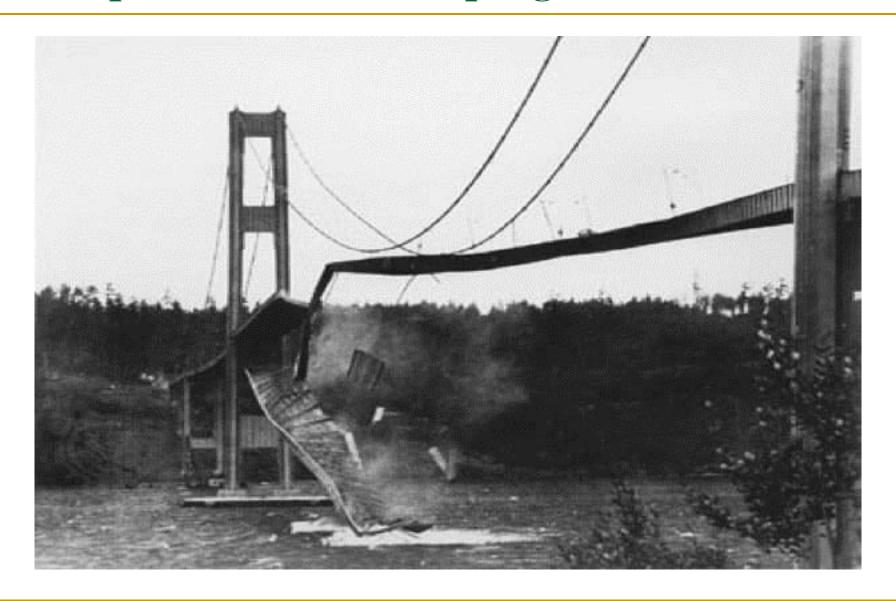


We need to start with reliability and security...

How Reliable/Secure/Safe is This Bridge?



Collapse of the "Galloping Gertie"



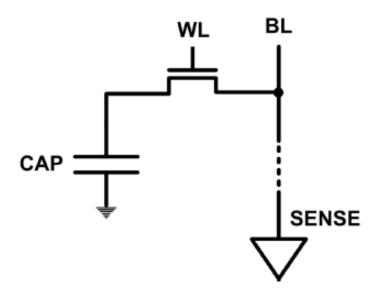
How Secure Are These People?



Security is about preventing unforeseen consequences

The DRAM Scaling Problem

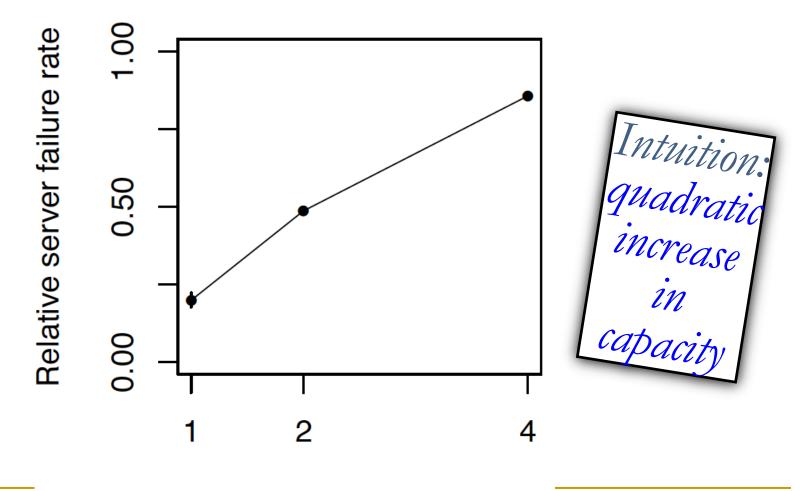
- DRAM stores charge in a capacitor (charge-based memory)
 - Capacitor must be large enough for reliable sensing
 - Access transistor should be large enough for low leakage and high retention time
 - Scaling beyond 40-35nm (2013) is challenging [ITRS, 2009]



DRAM capacity, cost, and energy/power hard to scale

As Memory Scales, It Becomes Unreliable

- Data from all of Facebook's servers worldwide
- Meza+, "Revisiting Memory Errors in Large-Scale Production Data Centers," DSN'15.



Large-Scale Failure Analysis of DRAM Chips

- Analysis and modeling of memory errors found in all of Facebook's server fleet
- Justin Meza, Qiang Wu, Sanjeev Kumar, and Onur Mutlu, "Revisiting Memory Errors in Large-Scale Production Data Centers: Analysis and Modeling of New Trends from the Field" Proceedings of the 45th Annual IEEE/IFIP International Conference on Dependable Systems and Networks (DSN), Rio de Janeiro, Brazil, June 2015.

[Slides (pptx) (pdf)] [DRAM Error Model]

Revisiting Memory Errors in Large-Scale Production Data Centers: Analysis and Modeling of New Trends from the Field

Justin Meza Qiang Wu* Sanjeev Kumar* Onur Mutlu Carnegie Mellon University * Facebook, Inc.

SAFARI

Infrastructures to Understand Such Issues



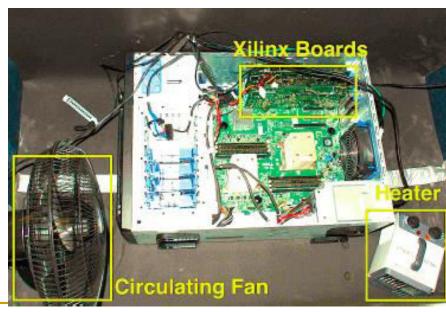
Flipping Bits in Memory Without Accessing
Them: An Experimental Study of DRAM
Disturbance Errors (Kim et al., ISCA 2014)

Adaptive-Latency DRAM: Optimizing DRAM
Timing for the Common-Case (Lee et al.,
HPCA 2015)

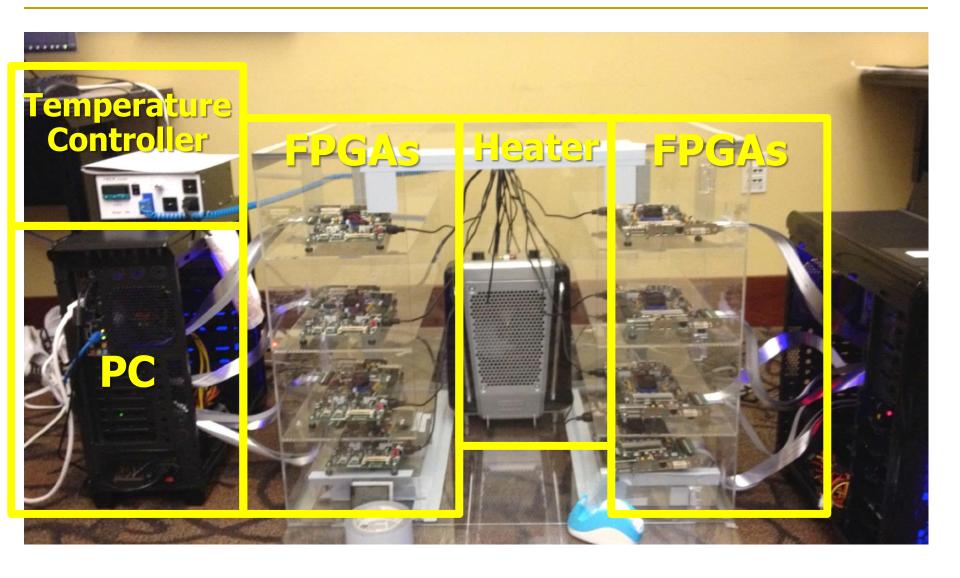
<u>AVATAR: A Variable-Retention-Time (VRT)</u>
<u>Aware Refresh for DRAM Systems</u> (Qureshi et al., DSN 2015)

An Experimental Study of Data Retention
Behavior in Modern DRAM Devices:
Implications for Retention Time Profiling
Mechanisms (Liu et al., ISCA 2013)

The Efficacy of Error Mitigation Techniques for DRAM Retention Failures: A Comparative Experimental Study (Khan et al., SIGMETRICS 2014)



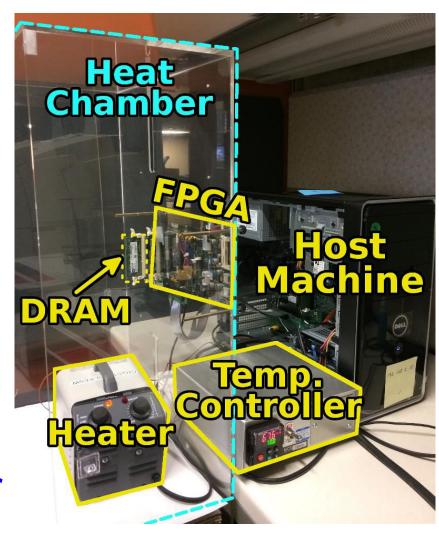
Infrastructures to Understand Such Issues



SoftMC: Open Source DRAM Infrastructure

Hasan Hassan et al., "SoftMC: A
 Flexible and Practical Open Source Infrastructure for
 Enabling Experimental DRAM
 Studies," HPCA 2017.

- Flexible
- Easy to Use (C++ API)
- Open-source github.com/CMU-SAFARI/SoftMC



SoftMC

https://github.com/CMU-SAFARI/SoftMC

SoftMC: A Flexible and Practical Open-Source Infrastructure for Enabling Experimental DRAM Studies

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 Hasan Hassan Nandita Vijaykumar Samira Khan Saugata Ghose Kevin Chang Gennady Pekhimenko Donghyuk Lee Gennady Pekhimenko Onur Mutlu Nandita Vijaykumar Samira Khan Saugata Ghose Kevin Chang Gennady Pekhimenko Onur Mutlu Nandita Vijaykumar Onur Mutlu Nandita Vijaykumar Samira Khan Saugata Ghose Nandita Vijaykumar Onur Mutlu Nandita Vijaykumar Onur Nandita Vijaykum
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<sup>1</sup>ETH Zürich <sup>2</sup>TOBB University of Economics & Technology <sup>3</sup>Carnegie Mellon University <sup>4</sup>University of Virginia <sup>5</sup>Microsoft Research <sup>6</sup>NVIDIA Research
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A Curious Discovery [Kim et al., ISCA 2014]

One can predictably induce errors in most DRAM memory chips

DRAM RowHammer

A simple hardware failure mechanism can create a widespread system security vulnerability



Forget Software—Now Hackers Are Exploiting Physics

BUSINESS CULTURE DESIGN GEAR SCIENCE



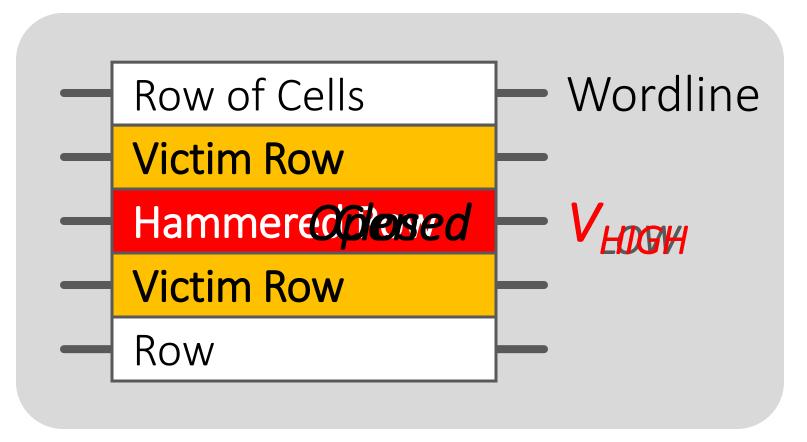




ANDY GREENBERG SECURITY 08.31.16 7:00 AM

FORGET SOFTWARE—NOW HACKERS ARE EXPLOITING PHYSICS

Modern DRAM is Prone to Disturbance Errors



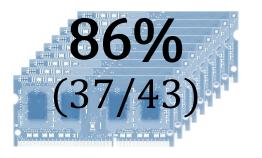
Repeatedly reading a row enough times (before memory gets refreshed) induces disturbance errors in adjacent rows in most real DRAM chips you can buy today

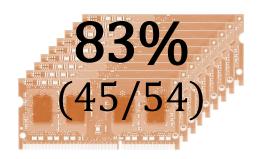
Most DRAM Modules Are at Risk

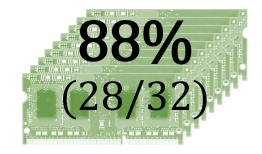
A company

B company

C company







Up to

1.0×10⁷
errors

Up to 2.7×10⁶ errors

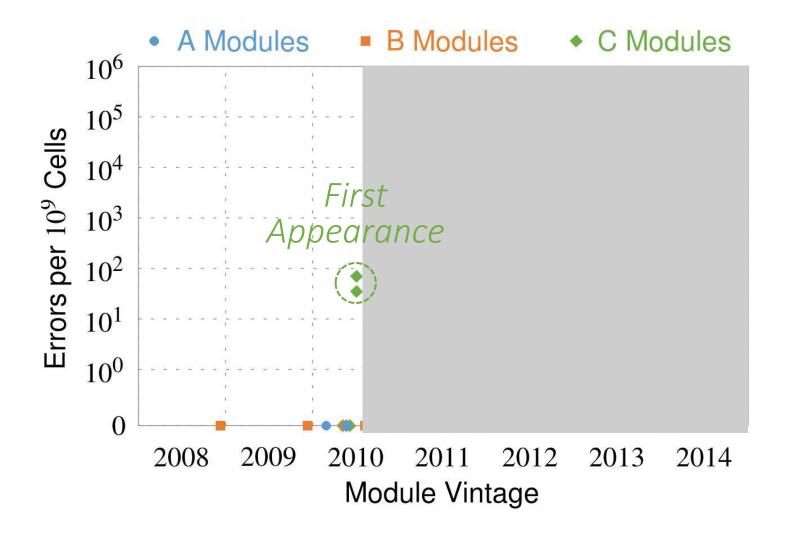
Up to

3.3×10⁵
errors

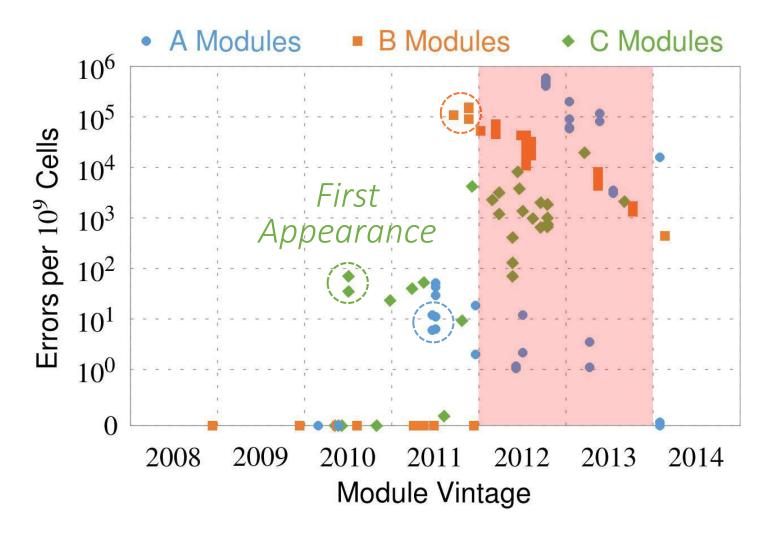
Recent DRAM Is More Vulnerable



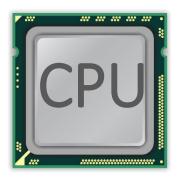
Recent DRAM Is More Vulnerable

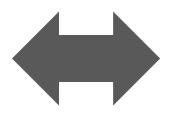


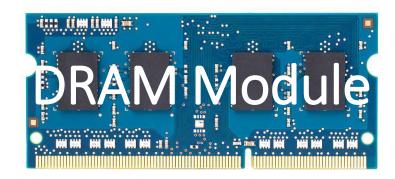
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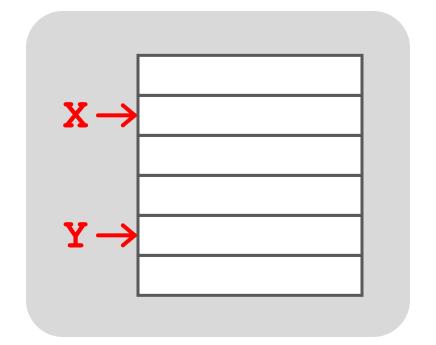
All modules from 2012–2013 are vulnerable

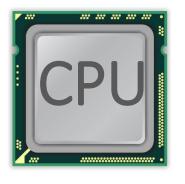


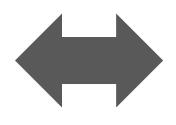




```
loop:
  mov (X), %eax
  mov (Y), %ebx
  clflush (X)
  clflush (Y)
  mfence
  jmp loop
```

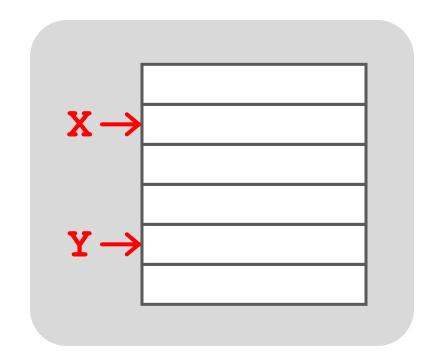


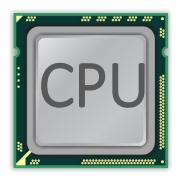


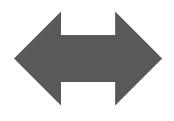


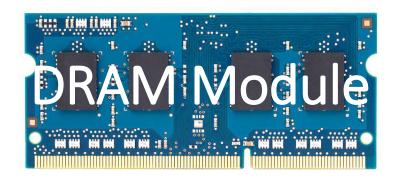


- 1. Avoid *cache hits*
 - Flush X from cache
- 2. Avoid *row hits* to X
 - Read Y in another row

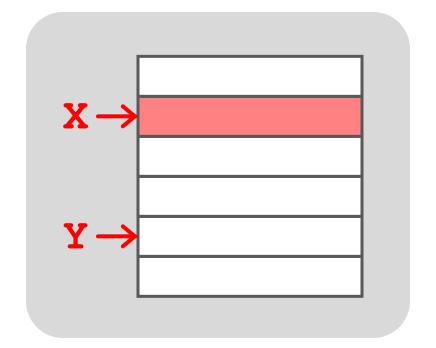


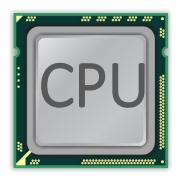


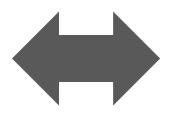


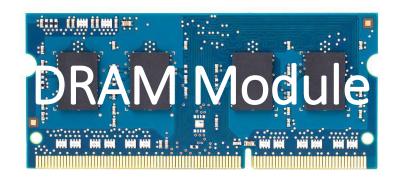


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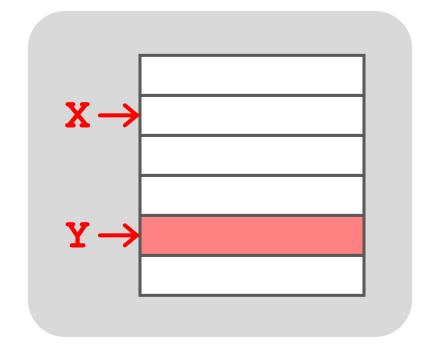


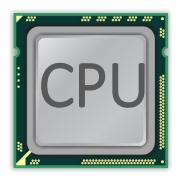


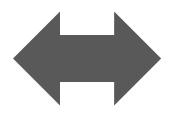


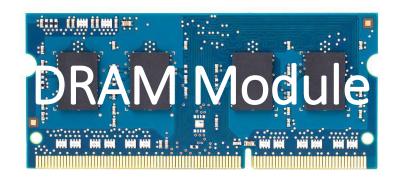


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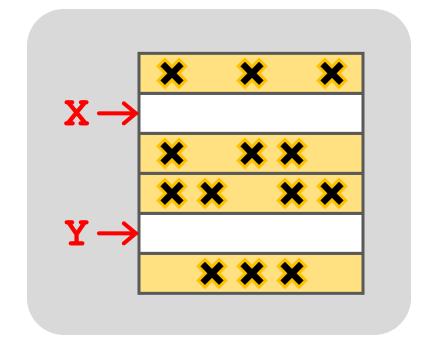








```
loop:
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  mov (Y), %ebx
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  clflush (Y)
  mfence
  jmp loop
```



Observed Errors in Real Systems

CPU Architecture	Errors	Access-Rate
Intel Haswell (2013)	22.9K	12.3M/sec
Intel Ivy Bridge (2012)	20.7K	11.7M/sec
Intel Sandy Bridge (2011)	16.1K	11.6M/sec
AMD Piledriver (2012)	59	6.1M/sec

A real reliability & security issue

One Can Take Over an Otherwise-Secure System

Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors

Abstract. Memory isolation is a key property of a reliable and secure computing system — an access to one memory address should not have unintended side effects on data stored in other addresses. However, as DRAM process technology

Project Zero

Flipping Bits in Memory Without Accessing Them:
An Experimental Study of DRAM Disturbance Errors
(Kim et al., ISCA 2014)

News and updates from the Project Zero team at Google

Exploiting the DRAM rowhammer bug to gain kernel privileges (Seaborn+, 2015)

Monday, March 9, 2015

Exploiting the DRAM rowhammer bug to gain kernel privileges

RowHammer Security Attack Example

- "Rowhammer" is a problem with some recent DRAM devices in which repeatedly accessing a row of memory can cause bit flips in adjacent rows (Kim et al., ISCA 2014).
 - Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors (Kim et al., ISCA 2014)
- We tested a selection of laptops and found that a subset of them exhibited the problem.
- We built two working privilege escalation exploits that use this effect.
 - Exploiting the DRAM rowhammer bug to gain kernel privileges (Seaborn+, 2015)
- One exploit uses rowhammer-induced bit flips to gain kernel privileges on x86-64 Linux when run as an unprivileged userland process.
- When run on a machine vulnerable to the rowhammer problem, the process was able to induce bit flips in page table entries (PTEs).
- It was able to use this to gain write access to its own page table, and hence gain read-write access to all of physical memory.

Security Implications



Security Implications



It's like breaking into an apartment by repeatedly slamming a neighbor's door until the vibrations open the door you were after

More Security Implications

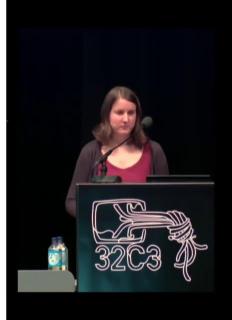
"We can gain unrestricted access to systems of website visitors."

www.iaik.tugraz.at

Not there yet, but ...



ROOT privileges for web apps!





Daniel Gruss (@lavados), Clémentine Maurice (@BloodyTangerine), December 28, 2015 — 32c3, Hamburg, Germany

Rowhammer.js: A Remote Software-Induced Fault Attack in JavaScript (DIMVA'16)

64

More Security Implications

"Can gain control of a smart phone deterministically" Hammer And Root Millions of Androids

Drammer: Deterministic Rowhammer Attacks on Mobile Platforms, CCS'16 65

More Security Implications?



Apple's Patch for RowHammer

https://support.apple.com/en-gb/HT204934

Available for: OS X Mountain Lion v10.8.5, OS X Mavericks v10.9.5

Impact: A malicious application may induce memory corruption to escalate privileges

Description: A disturbance error, also known as Rowhammer, exists with some DDR3 RAM that could have led to memory corruption. This issue was mitigated by increasing memory refresh rates.

CVE-ID

CVE-2015-3693 : Mark Seaborn and Thomas Dullien of Google, working from original research by Yoongu Kim et al (2014)

HP, Lenovo, and other vendors released similar patches

Our Solution to RowHammer

PARA: <u>Probabilistic Adjacent Row Activation</u>

Key Idea

- After closing a row, we activate (i.e., refresh) one of its neighbors with a low probability: p = 0.005

Reliability Guarantee

- When p=0.005, errors in one year: 9.4×10^{-14}
- By adjusting the value of p, we can vary the strength of protection against errors

More on RowHammer Analysis

Yoongu Kim, Ross Daly, Jeremie Kim, Chris Fallin, Ji Hye Lee, Donghyuk Lee, Chris Wilkerson, Konrad Lai, and Onur Mutlu,
 "Flipping Bits in Memory Without Accessing Them: An
 Experimental Study of DRAM Disturbance Errors"
 Proceedings of the 41st International Symposium on Computer
 Architecture (ISCA), Minneapolis, MN, June 2014.
 [Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)] [Source Code and Data]

Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors

Yoongu Kim¹ Ross Daly* Jeremie Kim¹ Chris Fallin* Ji Hye Lee¹ Donghyuk Lee¹ Chris Wilkerson² Konrad Lai Onur Mutlu¹

¹Carnegie Mellon University ²Intel Labs

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Future of Memory Reliability/Security

Onur Mutlu,

"The RowHammer Problem and Other Issues We May Face as Memory Becomes Denser"

Invited Paper in Proceedings of the <u>Design, Automation, and Test in</u> <u>Europe Conference</u> (**DATE**), Lausanne, Switzerland, March 2017. [Slides (pptx) (pdf)]

The RowHammer Problem and Other Issues We May Face as Memory Becomes Denser

Onur Mutlu
ETH Zürich
onur.mutlu@inf.ethz.ch
https://people.inf.ethz.ch/omutlu

Industry Is Writing Papers About It, Too

DRAM Process Scaling Challenges

Refresh

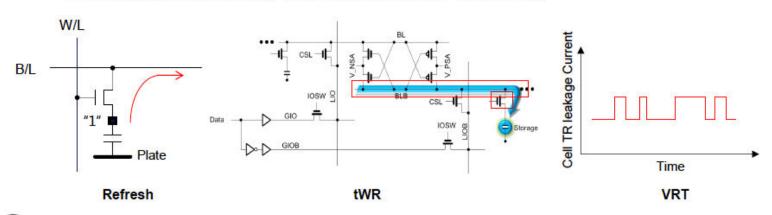
- Difficult to build high-aspect ratio cell capacitors decreasing cell capacitance
- · Leakage current of cell access transistors increasing

tWR

- Contact resistance between the cell capacitor and access transistor increasing
- · On-current of the cell access transistor decreasing
- Bit-line resistance increasing

VRT

· Occurring more frequently with cell capacitance decreasing









Call for Intelligent Memory Controllers

DRAM Process Scaling Challenges

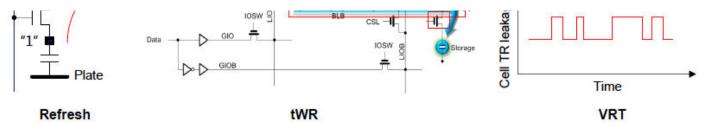
Refresh

Difficult to build high-aspect ratio cell capacitors decreasing cell capacitance
 THE MEMORY FORUM 2014

Co-Architecting Controllers and DRAM to Enhance DRAM Process Scaling

Uksong Kang, Hak-soo Yu, Churoo Park, *Hongzhong Zheng, **John Halbert, **Kuljit Bains, SeongJin Jang, and Joo Sun Choi

Samsung Electronics, Hwasung, Korea / *Samsung Electronics, San Jose / **Intel







Solution Direction: Principled Designs

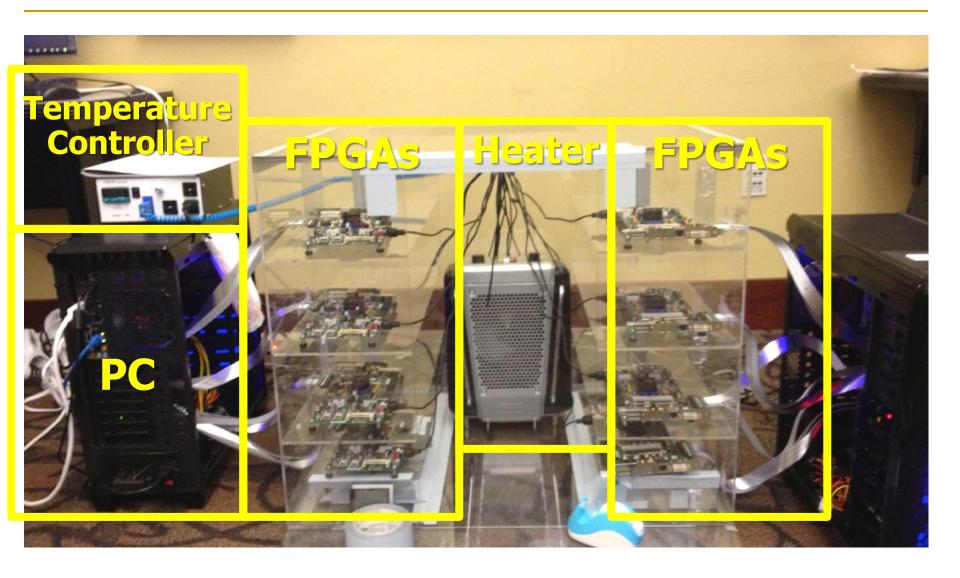
Design fundamentally secure computing architectures

Predict and prevent such safety issues

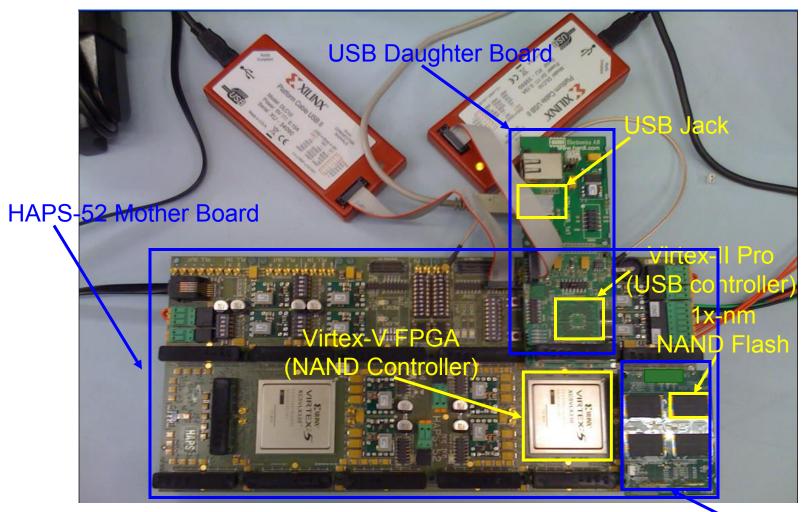
Architecting for Security

- Understand: Methods for vulnerability modeling & discovery
 - Modeling and prediction based on real (device) data and analysis
 - Understanding vulnerabilities
 - Metrics for secure architectures
- Architect: Principled architectures with security as key concern
 - Good partitioning of duties across the stack
 - Cannot give up performance and efficiency
 - Patch-ability in the field
- Design & Test: Principled design, automation, (online) testing
 - Design for security
 - High coverage and good interaction with system reliability methods

Understand and Model with Experiments (DRAM)



Understand and Model with Experiments (Flash)



[DATE 2012, ICCD 2012, DATE 2013, ITJ 2013, ICCD 2013, SIGMETRICS 2014, HPCA 2015, DSN 2015, MSST 2015, JSAC 2016, HPCA 2017, DFRWS 2017, PIEEE'17, HPCA'18]

NAND Daughter Board

There are Two Other Solution Directions

New Technologies: Replace or (more likely) augment DRAM with a different technology

Non-volatile memories

Embracing Un-reliability:

Design memories with different reliability and store data intelligently across them

Problem Algorithm Program/Language System Software SW/HW Interface Micro-architecture Logic Dovices Electrons

Fundamental solutions to security require co-design across the hierarchy

Fundamentally Secure, Reliable, Safe Computing Architectures

Main Memory Needs Intelligent Controllers

Four Key Issues in Future Platforms

Fundamentally Secure/Reliable/Safe Architectures

- Fundamentally Energy-Efficient Architectures
 - Memory-centric (Data-centric) Architectures

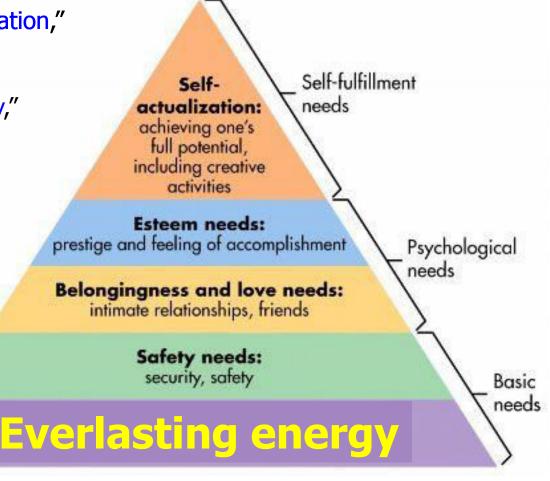
Fundamentally Low-Latency Architectures

Architectures for Genomics, Medicine, Health

Maslow's (Human) Hierarchy of Needs, Revisited

Maslow, "A Theory of Human Motivation," Psychological Review, 1943.

Maslow, "Motivation and Personality," Book, 1954-1970.

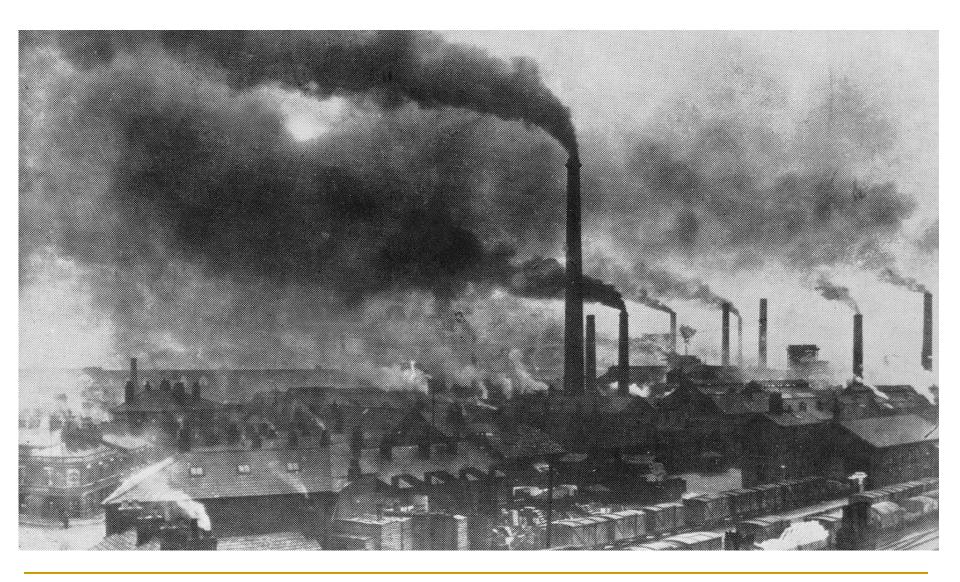


Do We Want This?





Or This?



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Challenge and Opportunity for Future

Sustainable and Energy Efficient

The Problem

Data access is the major performance and energy bottleneck

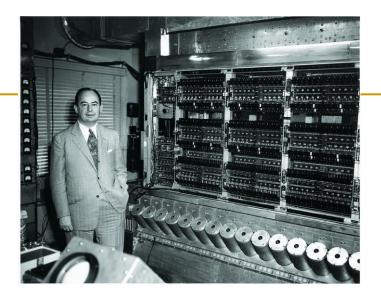
Our current design principles cause great energy waste

(and great performance loss)

Processing of data is performed far away from the data

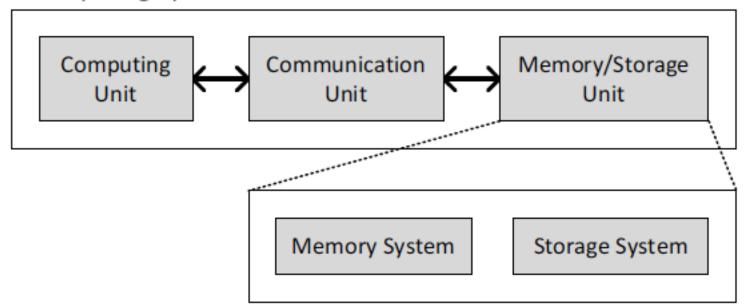
A Computing System

- Three key components
- Computation
- Communication
- Storage/memory



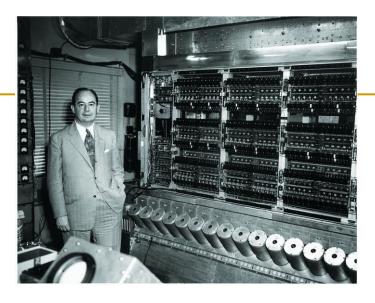
Burks, Goldstein, von Neumann, "Preliminary discussion of the logical design of an electronic computing instrument," 1946.

Computing System



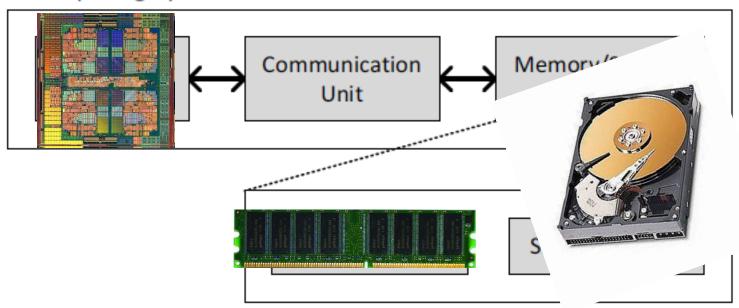
A Computing System

- Three key components
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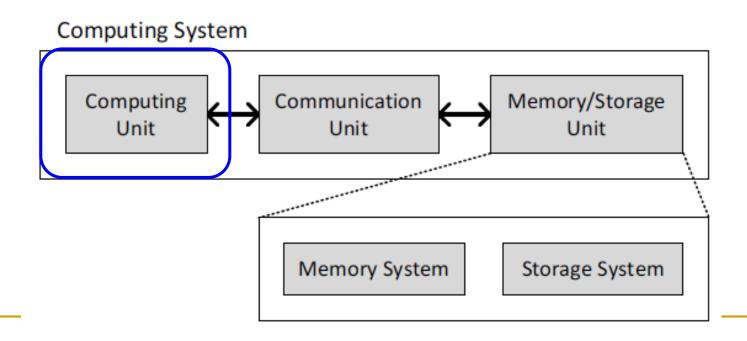
Burks, Goldstein, von Neumann, "Preliminary discussion of the logical design of an electronic computing instrument," 1946.

Computing System



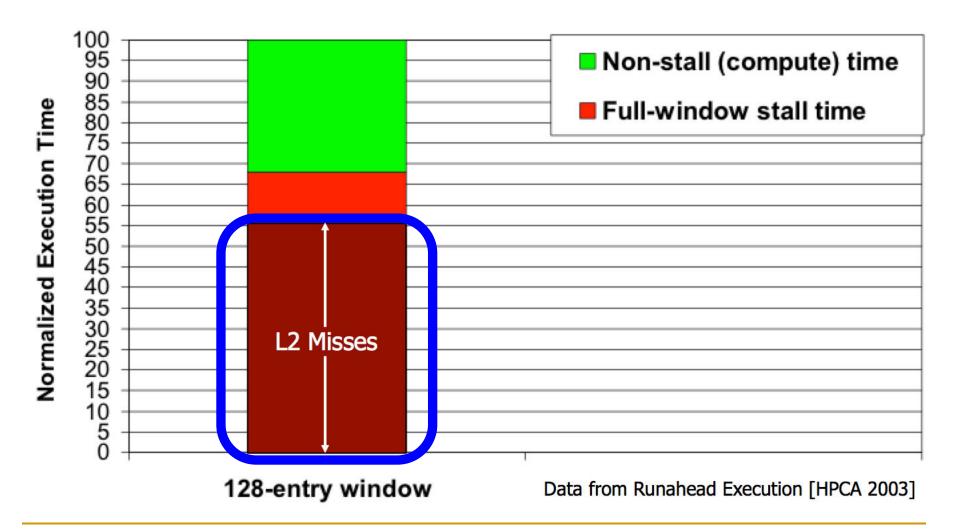
Today's Computing Systems

- Are overwhelmingly processor centric
- All data processed in the processor → at great system cost
- Processor is heavily optimized and is considered the master
- Data storage units are dumb and are largely unoptimized (except for some that are on the processor die)



Yet ...

"It's the Memory, Stupid!" (Richard Sites, MPR, 1996)



The Performance Perspective

Onur Mutlu, Jared Stark, Chris Wilkerson, and Yale N. Patt,
 "Runahead Execution: An Alternative to Very Large Instruction
 Windows for Out-of-order Processors"
 Proceedings of the <u>9th International Symposium on High-Performance</u>
 <u>Computer Architecture</u> (HPCA), pages 129-140, Anaheim, CA, February
 2003. <u>Slides (pdf)</u>

Runahead Execution: An Alternative to Very Large Instruction Windows for Out-of-order Processors

Onur Mutlu § Jared Stark † Chris Wilkerson ‡ Yale N. Patt §

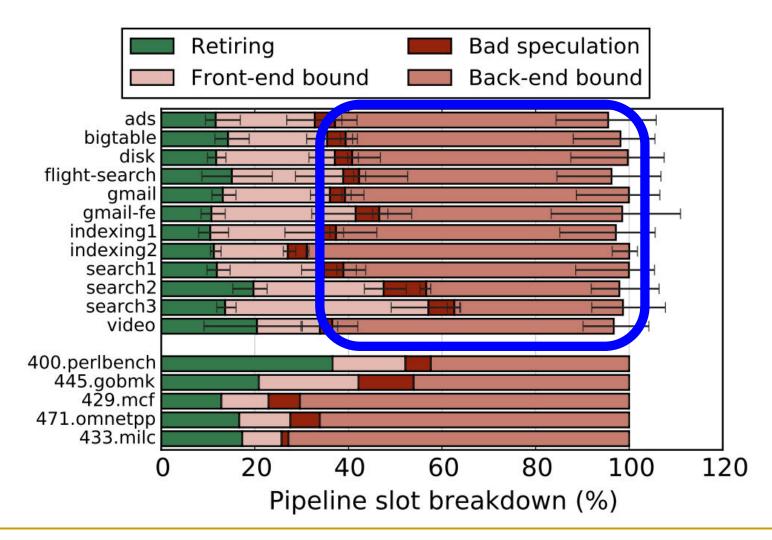
§ECE Department
The University of Texas at Austin
{onur,patt}@ece.utexas.edu

†Microprocessor Research Intel Labs jared.w.stark@intel.com

‡Desktop Platforms Group Intel Corporation chris.wilkerson@intel.com

The Performance Perspective (Today)

All of Google's Data Center Workloads (2015):



The Performance Perspective (Today)

All of Google's Data Center Workloads (2015):

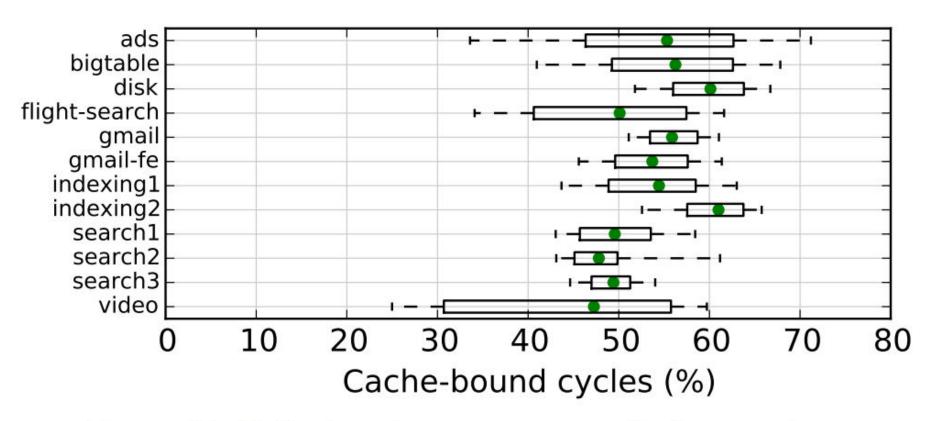
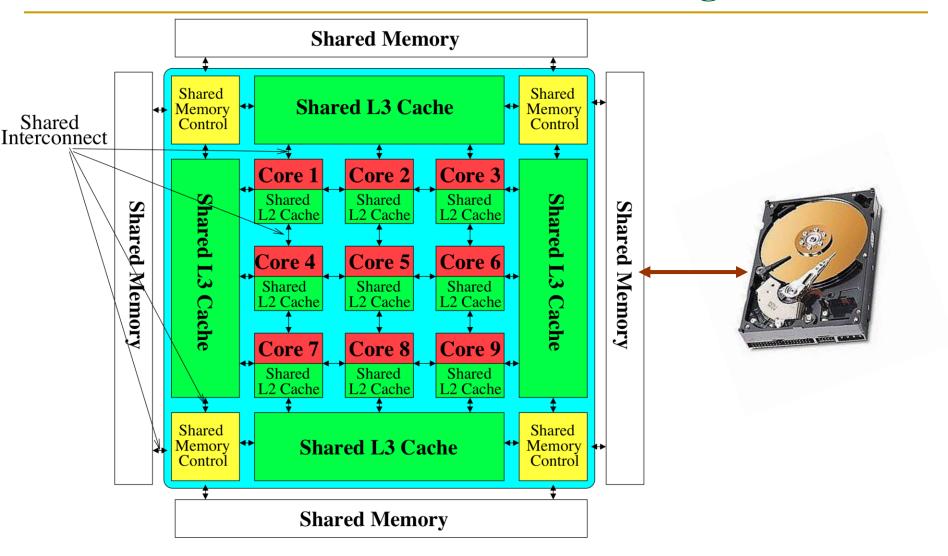


Figure 11: Half of cycles are spent stalled on caches.

Perils of Processor-Centric Design

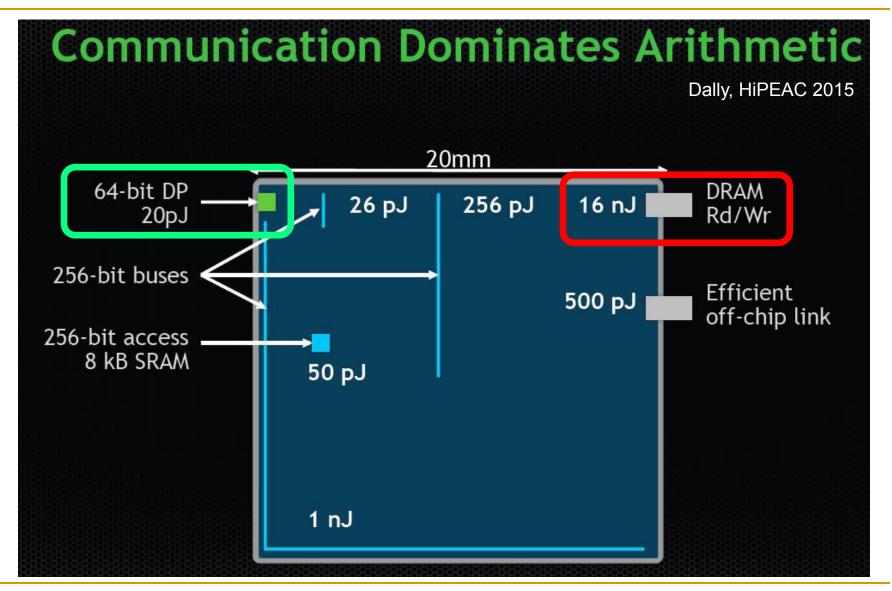
- Grossly-imbalanced systems
 - Processing done only in one place
 - Everything else just stores and moves data: data moves a lot
 - → Energy inefficient
 - → Low performance
 - → Complex
- Overly complex and bloated processor (and accelerators)
 - To tolerate data access from memory
 - Complex hierarchies and mechanisms
 - → Energy inefficient
 - → Low performance
 - → Complex

Perils of Processor-Centric Design

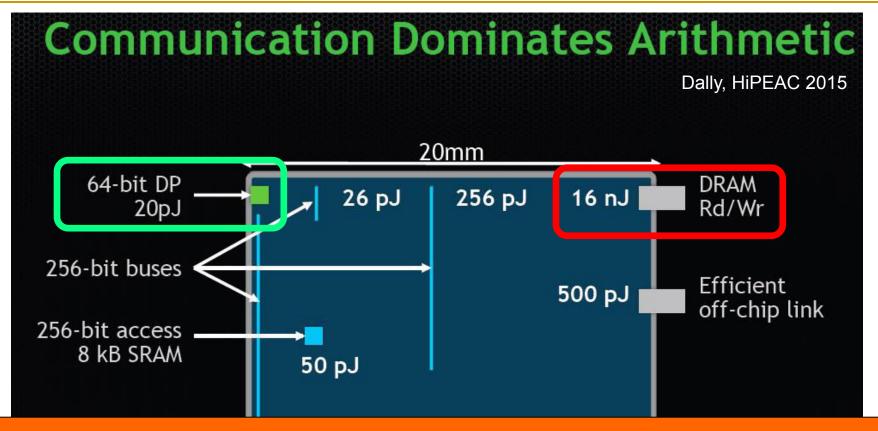


Most of the system is dedicated to storing and moving data

The Energy Perspective

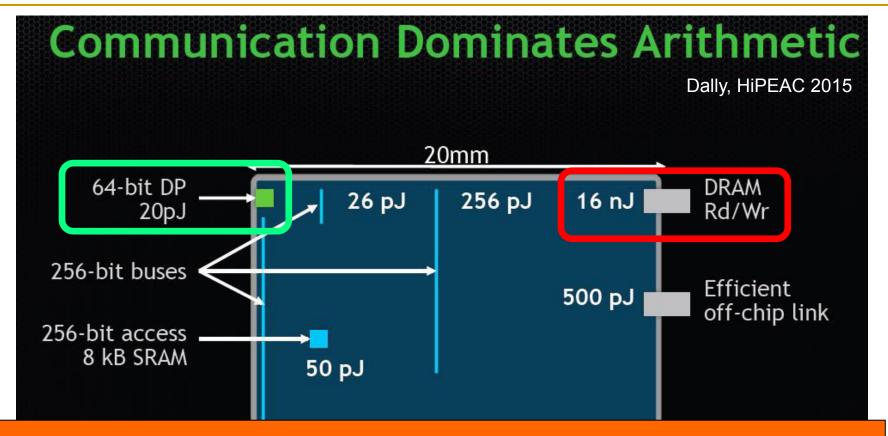


Data Movement vs. Computation Energy



A memory access consumes ~1000X the energy of a complex addition

We Do Not Want to Move Data!



A memory access consumes ~1000X the energy of a complex addition

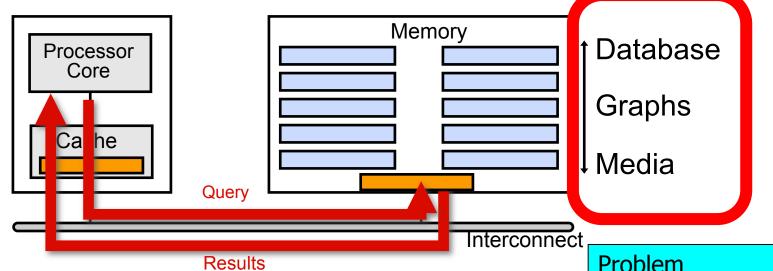
We Need A Paradigm Shift To ...

Enable computation with minimal data movement

Compute where it makes sense (where data resides)

Make computing architectures more data-centric

Goal: Processing Inside Memory



- Many questions ... How do we design the:
 - compute-capable memory & controllers?
 - processor chip?
 - software and hardware interfaces?
 - system software and languages?
 - algorithms?

Problem

Aigorithm

Program/Language

System Software

SW/HW Interface

Micro-architecture

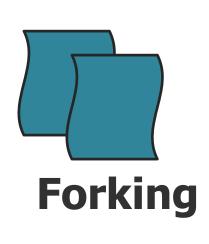
Logic

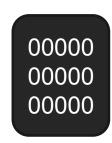
Davicac

Electrons

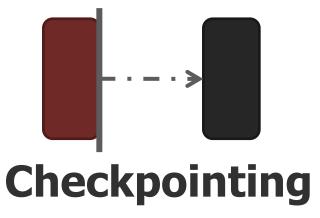
Starting Simple: Data Copy and Initialization

memmove & memcpy: 5% cycles in Google's datacenter [Kanev+ ISCA'15]







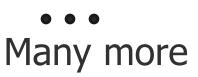




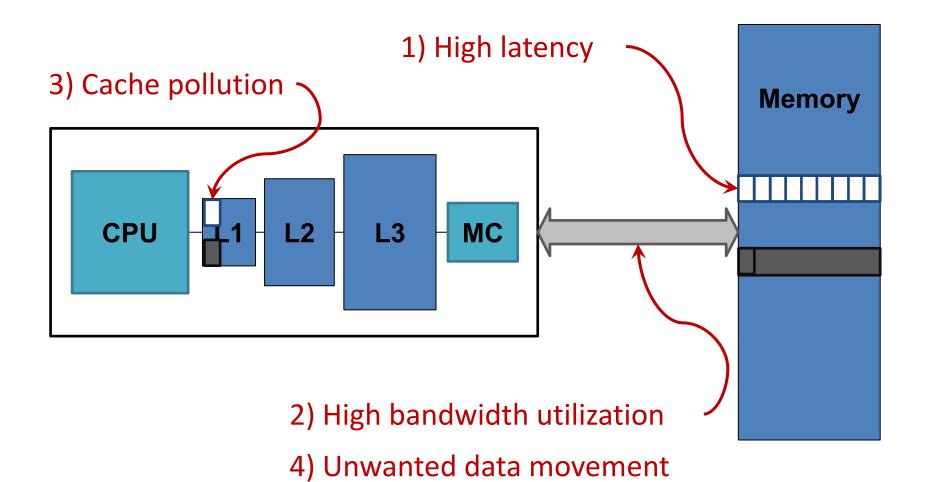




Page Migration

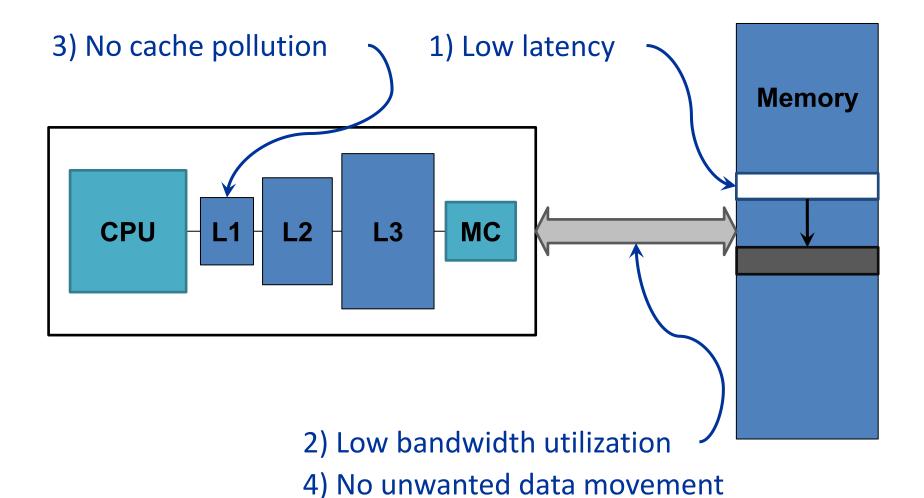


Today's Systems: Bulk Data Copy



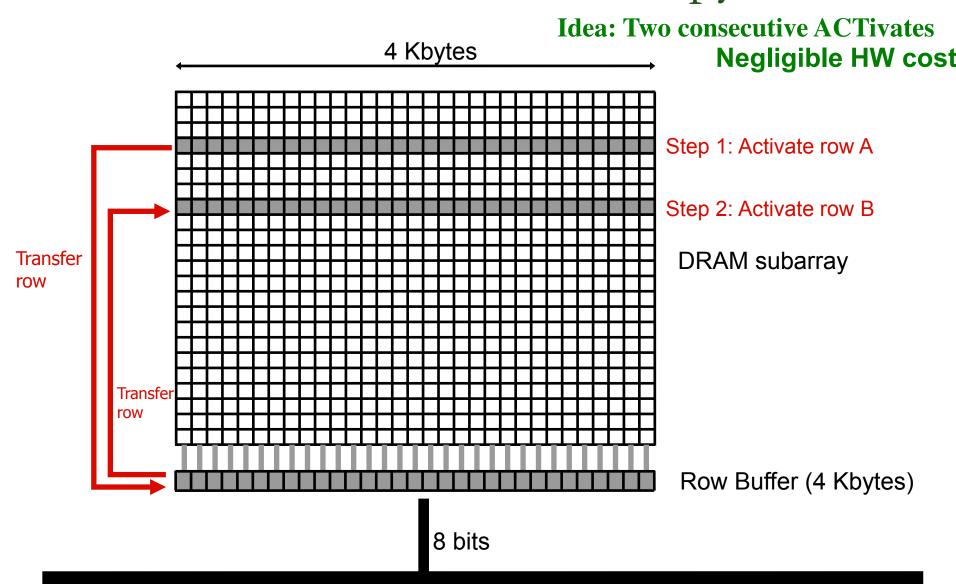
1046ns, 3.6uJ (for 4KB page copy via DMA)

Future Systems: In-Memory Copy

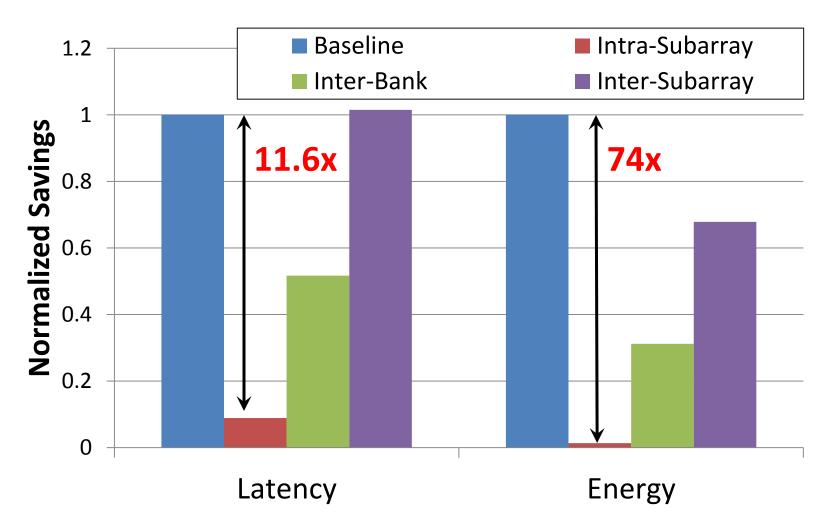


- 1046ns, 3.6uJ
- → 90ns, 0.04uJ

RowClone: In-DRAM Row Copy



RowClone: Latency and Energy Savings



Seshadri et al., "RowClone: Fast and Efficient In-DRAM Copy and Initialization of Bulk Data," MICRO 2013.

More on RowClone

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"

Proceedings of the <u>46th International Symposium on Microarchitecture</u> (**MICRO**), Davis, CA, December 2013. [<u>Slides (pptx) (pdf)</u>] [<u>Lightning Session Slides (pptx) (pdf)</u>] [<u>Poster (pptx) (pdf)</u>]

RowClone: Fast and Energy-Efficient In-DRAM Bulk Data Copy and Initialization

Vivek Seshadri Yoongu Kim Chris Fallin* Donghyuk Lee vseshadr@cs.cmu.edu yoongukim@cmu.edu cfallin@c1f.net donghyuk1@cmu.edu

Rachata Ausavarungnirun Gennady Pekhimenko Yixin Luo rachata@cmu.edu gpekhime@cs.cmu.edu yixinluo@andrew.cmu.edu

Onur Mutlu Phillip B. Gibbons† Michael A. Kozuch† Todd C. Mowry onur@cmu.edu phillip.b.gibbons@intel.com michael.a.kozuch@intel.com tcm@cs.cmu.edu

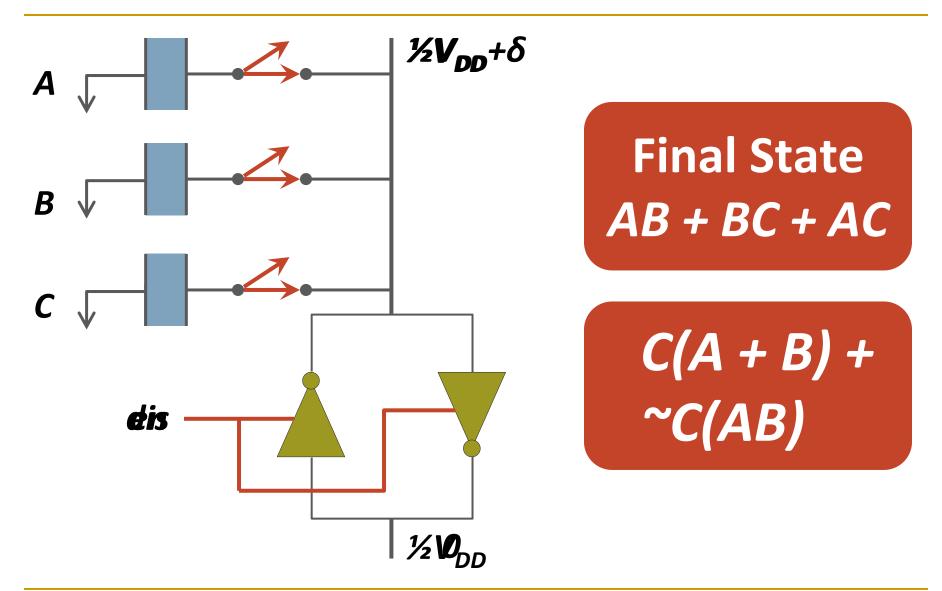
Carnegie Mellon University †Intel Pittsburgh

In-Memory Bulk Bitwise Operations

- We can support in-DRAM COPY, ZERO, AND, OR, NOT, MAJ
- At low cost
- Using analog computation capability of DRAM
 - Idea: activating multiple rows performs computation
- 30-60X performance and energy improvement
 - Seshadri+, "Ambit: In-Memory Accelerator for Bulk Bitwise Operations Using Commodity DRAM Technology," MICRO 2017.

- New memory technologies enable even more opportunities
 - Memristors, resistive RAM, phase change mem, STT-MRAM, ...
 - Can operate on data with minimal movement

In-DRAM AND/OR: Triple Row Activation



In-DRAM NOT: Dual Contact Cell

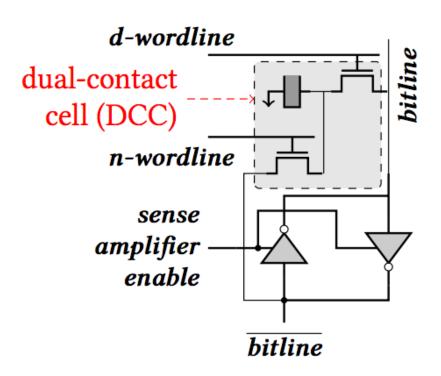
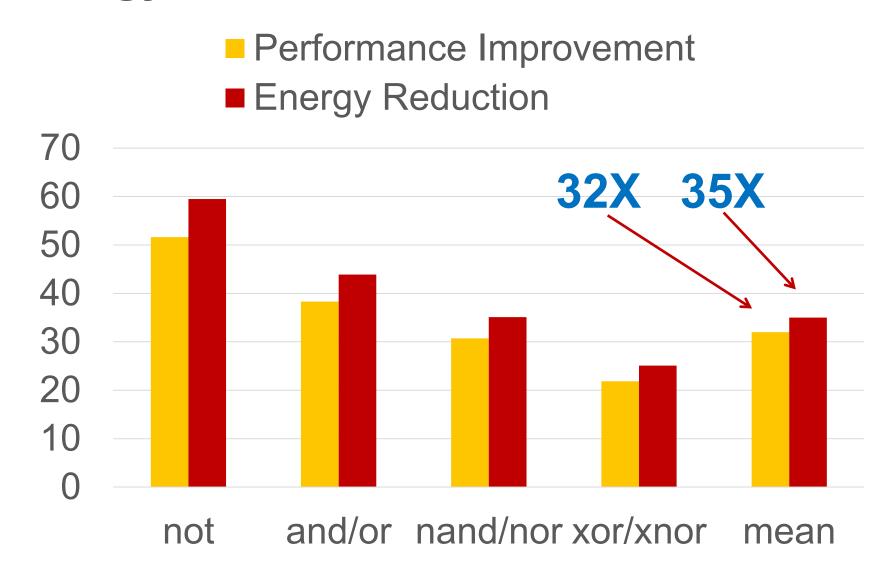


Figure 5: A dual-contact cell connected to both ends of a sense amplifier

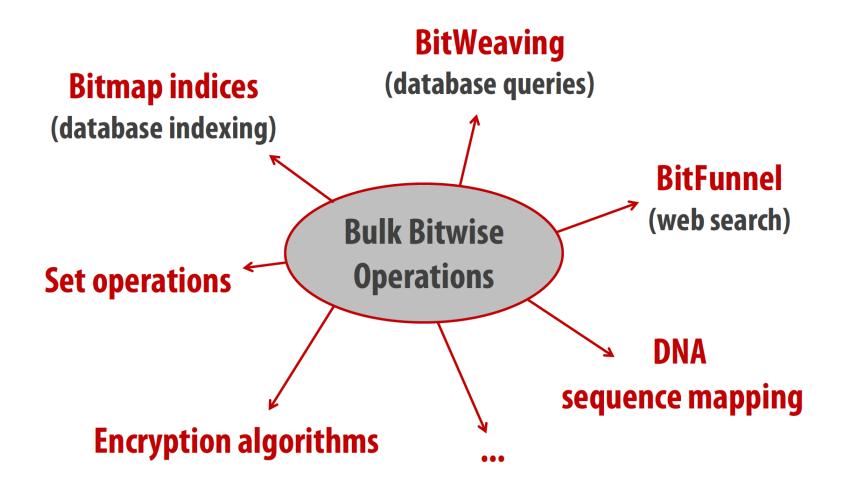
Idea:
Feed the
negated value
in the sense amplifier
into a special row

Seshadri+, "Ambit: In-Memory Accelerator for Bulk Bitwise Operations using Commodity DRAM Technology," MICRO 2017.

Ambit vs. DDR3: Performance and Energy



Bulk Bitwise Operations in Workloads



Performance: Bitmap Index on Ambit

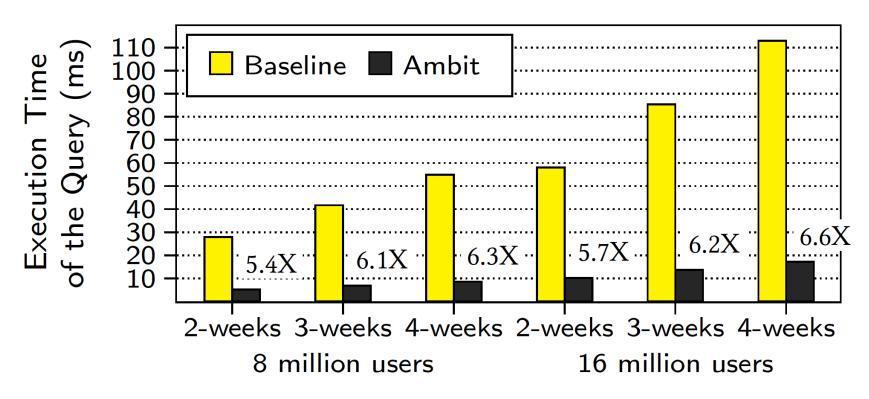


Figure 10: Bitmap index performance. The value above each bar indicates the reduction in execution time due to Ambit.

Seshadri+, "Ambit: In-Memory Accelerator for Bulk Bitwise Operations using Commodity DRAM Technology," MICRO 2017.

Performance: BitWeaving on Ambit

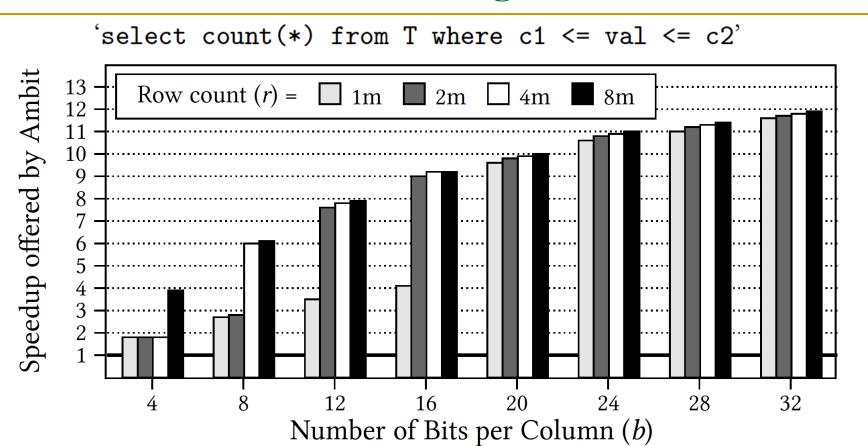


Figure 11: Speedup offered by Ambit over baseline CPU with SIMD for BitWeaving

Seshadri+, "Ambit: In-Memory Accelerator for Bulk Bitwise Operations using Commodity DRAM Technology," MICRO 2017.

More on In-DRAM Bitwise Operations

 Vivek Seshadri et al., "<u>Ambit: In-Memory Accelerator</u> for Bulk Bitwise Operations Using Commodity DRAM <u>Technology</u>," MICRO 2017.

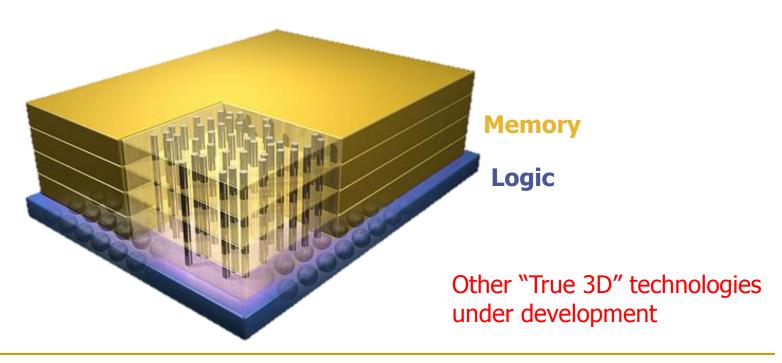
Ambit: In-Memory Accelerator for Bulk Bitwise Operations
Using Commodity DRAM Technology

```
Vivek Seshadri^{1,5} Donghyuk Lee^{2,5} Thomas Mullins^{3,5} Hasan Hassan^4 Amirali Boroumand^5 Jeremie Kim^{4,5} Michael A. Kozuch^3 Onur Mutlu^{4,5} Phillip B. Gibbons^5 Todd C. Mowry^5
```

 1 Microsoft Research India 2 NVIDIA Research 3 Intel 4 ETH Zürich 5 Carnegie Mellon University

Opportunity: 3D-Stacked Logic+Memory





Another Example: In-Memory Graph Processing

Large graphs are everywhere (circa 2015)



36 Million Wikipedia Pages



1.4 Billion Facebook Users

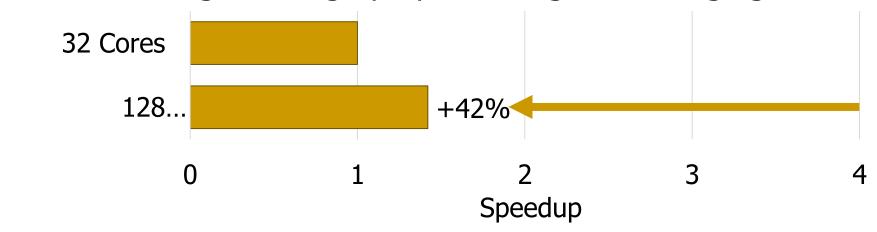


300 Million Twitter Users



30 Billion Instagram Photos

Scalable large-scale graph processing is challenging

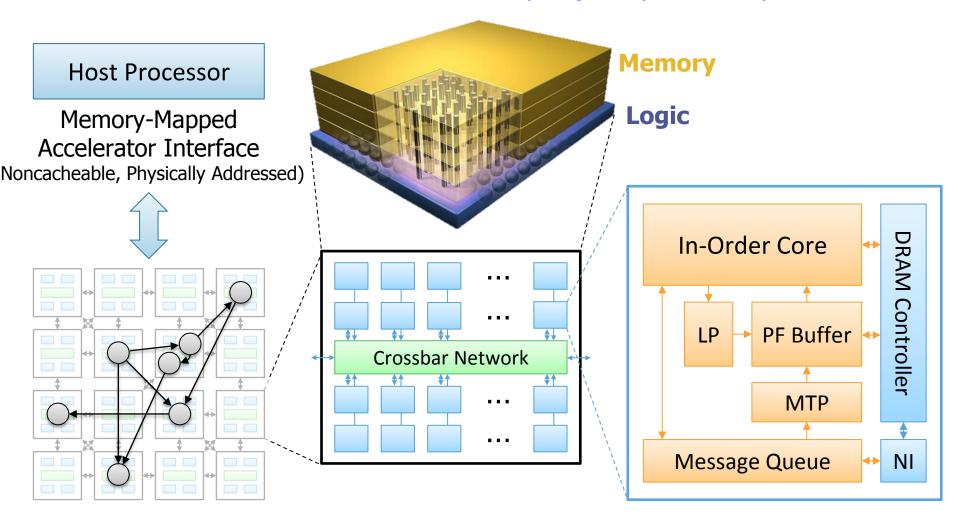


Key Bottlenecks in Graph Processing

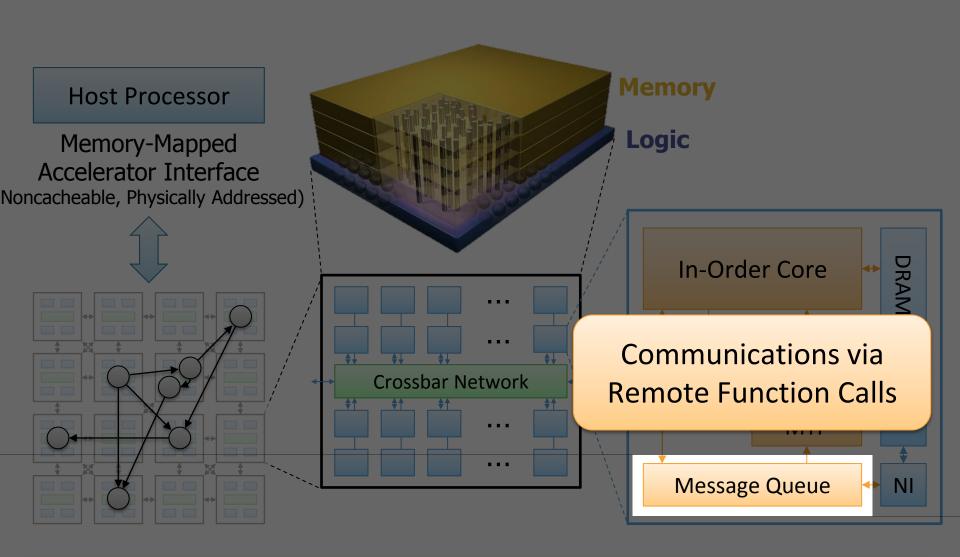
```
for (v: graph.vertices) {
     for (w: v.successors) {
       w.next rank += weight * v.rank;
                       1. Frequent random memory accesses
                                   &w
            V
 w.rank
w.next rank
                              weight * v.rank
 w.edges
            W
                              2. Little amount of computation
```

Tesseract System for Graph Processing

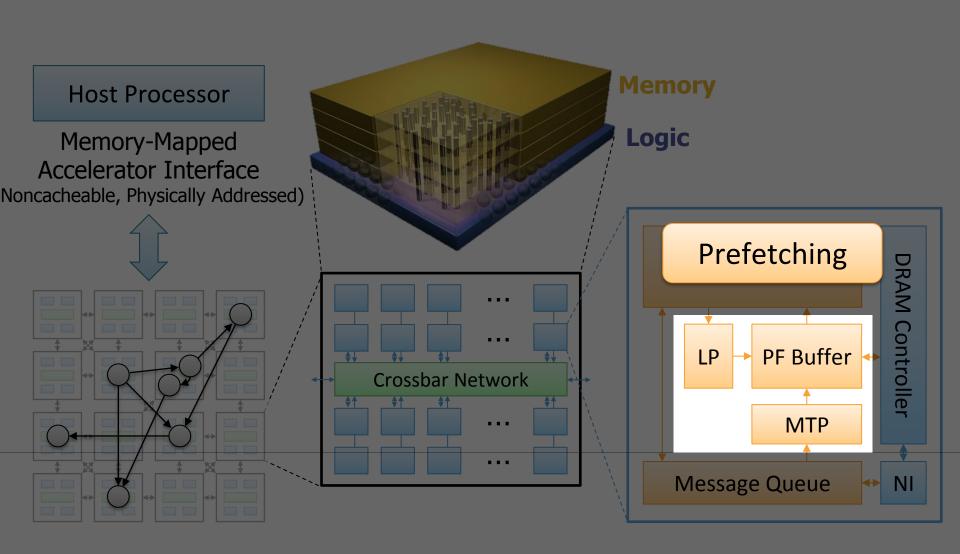
Interconnected set of 3D-stacked memory+logic chips with simple cores



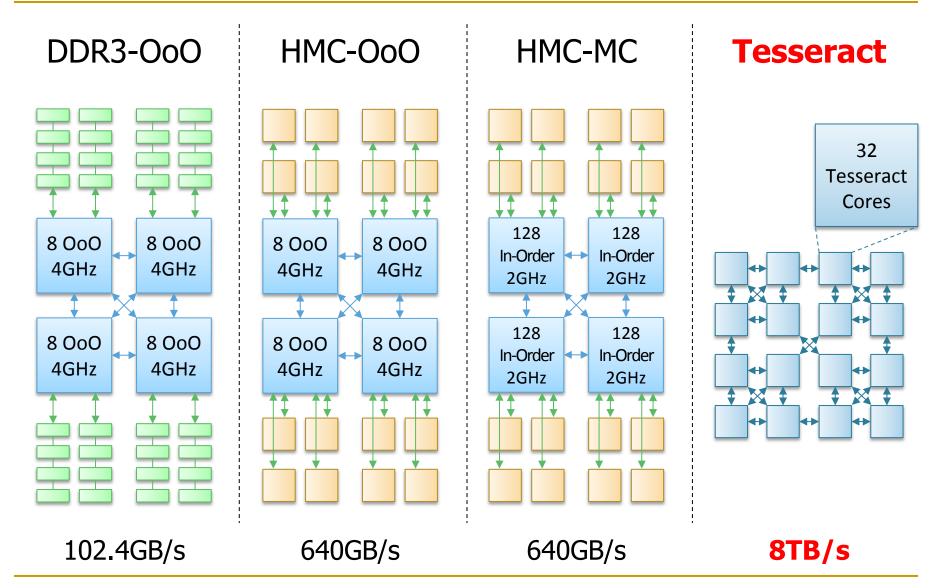
Tesseract System for Graph Processing



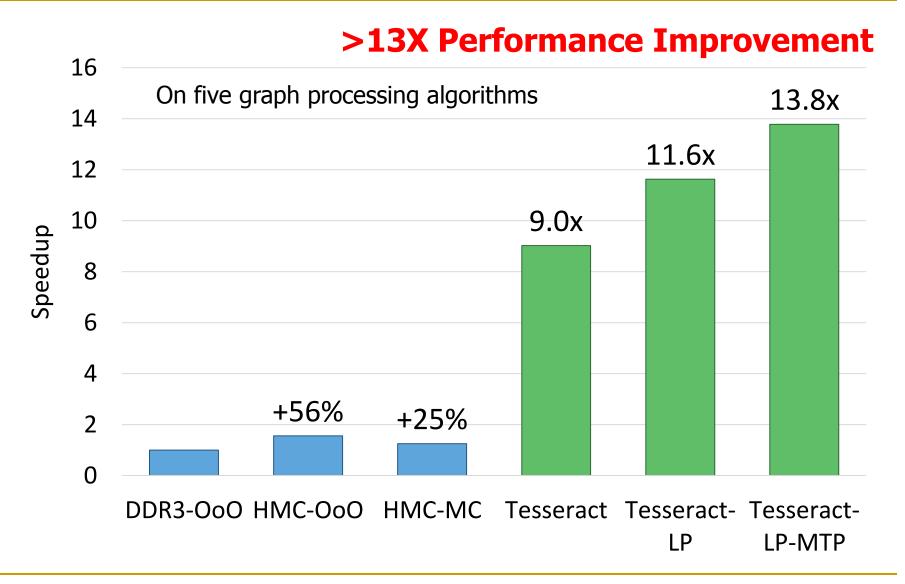
Tesseract System for Graph Processing



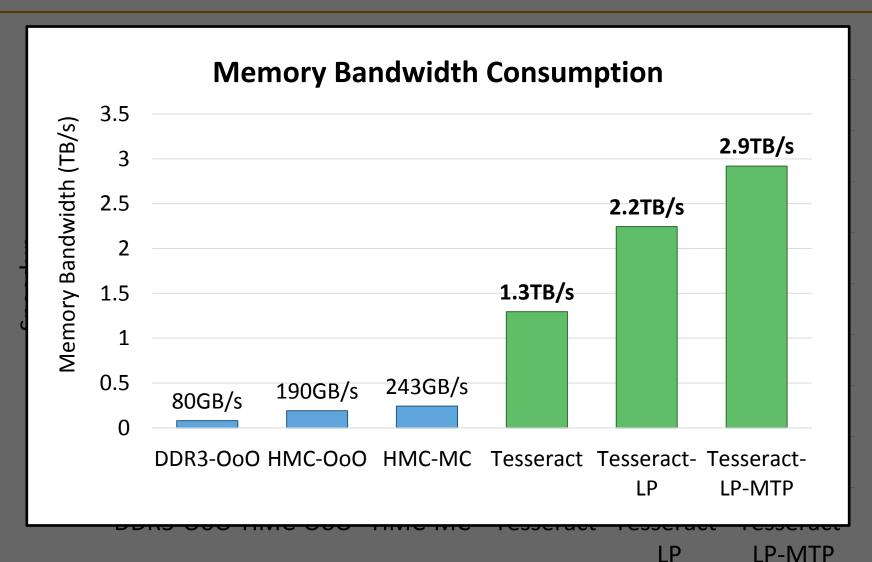
Evaluated Systems



Tesseract Graph Processing Performance

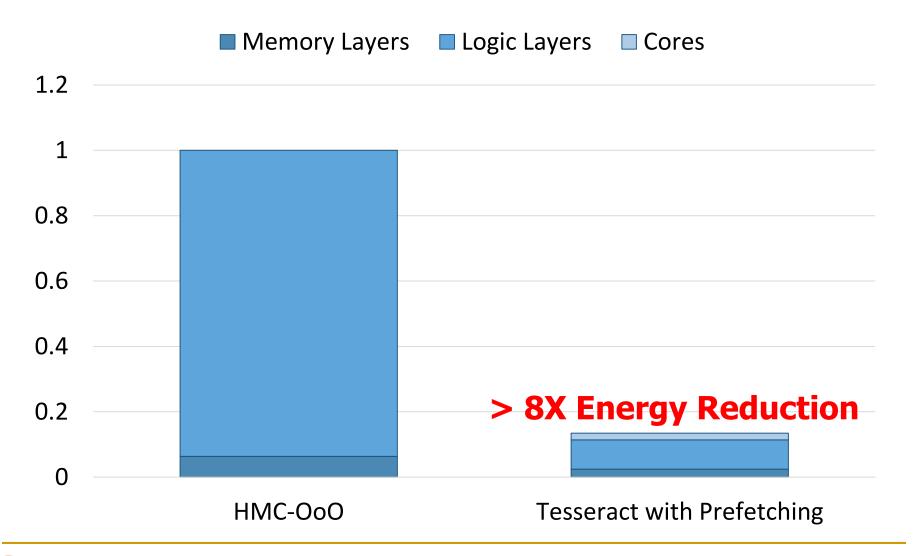


Tesseract Graph Processing Performance



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Tesseract Graph Processing System Energy



SAFARI Ahn+, "A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing" ISCA 2015.

More on Tesseract

 Junwhan Ahn, Sungpack Hong, Sungjoo Yoo, Onur Mutlu, and Kiyoung Choi,

"A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing"

Proceedings of the <u>42nd International Symposium on</u> <u>Computer Architecture</u> (**ISCA**), Portland, OR, June 2015. [Slides (pdf)] [Lightning Session Slides (pdf)]

A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing

Junwhan Ahn Sungpack Hong[§] Sungjoo Yoo Onur Mutlu[†] Kiyoung Choi junwhan@snu.ac.kr, sungpack.hong@oracle.com, sungjoo.yoo@gmail.com, onur@cmu.edu, kchoi@snu.ac.kr Seoul National University [§]Oracle Labs [†]Carnegie Mellon University

Simpler PIM: PIM-Enabled Instructions

Junwhan Ahn, Sungjoo Yoo, Onur Mutlu, and Kiyoung Choi,
 "PIM-Enabled Instructions: A Low-Overhead,
 Locality-Aware Processing-in-Memory Architecture"
 Proceedings of the <u>42nd International Symposium on</u>
 Computer Architecture (ISCA), Portland, OR, June 2015.
 [Slides (pdf)] [Lightning Session Slides (pdf)]

PIM-Enabled Instructions: A Low-Overhead, Locality-Aware Processing-in-Memory Architecture

Junwhan Ahn Sungjoo Yoo Onur Mutlu[†] Kiyoung Choi junwhan@snu.ac.kr, sungjoo.yoo@gmail.com, onur@cmu.edu, kchoi@snu.ac.kr

Seoul National University [†]Carnegie Mellon University

SAFARI

Challenge and Opportunity for Future

Fundamentally **Energy-Efficient** (Data-Centric) Computing Architectures

Computing Architectures with Minimal Data Movement

Eliminating the Adoption Barriers

How to Enable Adoption of Processing in Memory

Four Key Issues in Future Platforms

Fundamentally Secure/Reliable/Safe Architectures

- Fundamentally Energy-Efficient Architectures
 - Memory-centric (Data-centric) Architectures

Fundamentally Low-Latency Architectures

Architectures for Genomics, Medicine, Health





Maslow's Hierarchy of Needs, A Third Time

Maslow, "A Theory of Human Motivation," Psychological Review, 1943. Self-fulfillment Selfneeds Maslow, "Motivation and Personality," actualization: Book, 1954-1970. **Speed** prestige c Speed Psychological needs intim Speed ends ends Belongi **Speed** Basic needs Speed st

See Backup Slides for Latency...

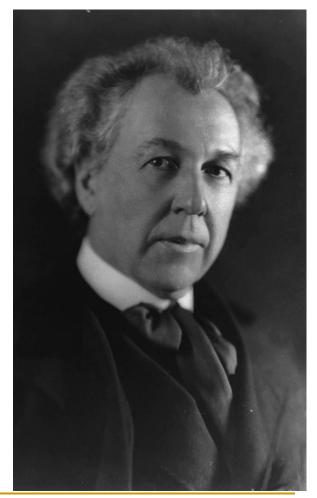
Challenge and Opportunity for Future

Fundamentally Low-Latency Computing Architectures

Concluding Remarks

A Quote from A Famous Architect

"architecture [...] based upon principle, and not upon precedent"



Precedent-Based Design?

"architecture [...] based upon principle, and not upon precedent"

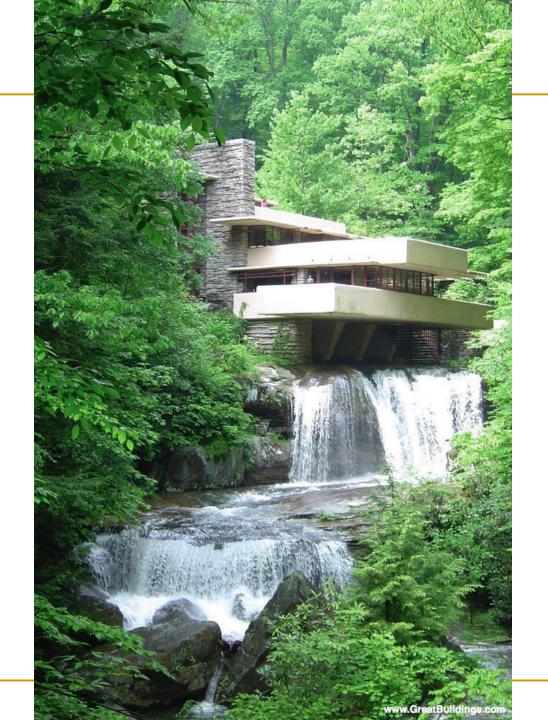


Principled Design

"architecture [...] based upon principle, and not upon precedent"



138



The Overarching Principle

Organic architecture

From Wikipedia, the free encyclopedia

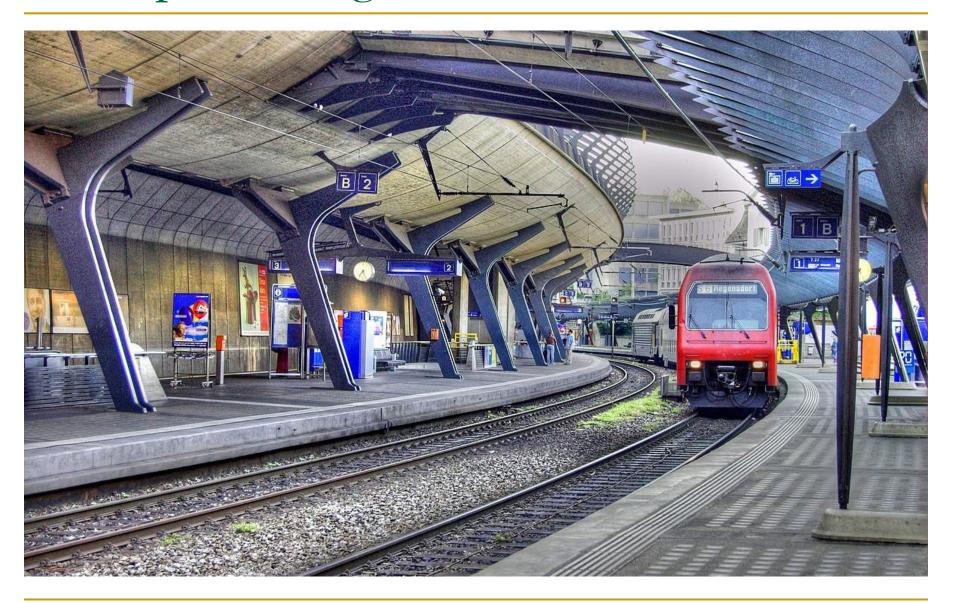
Organic architecture is a philosophy of architecture which promotes harmony between human habitation and the natural world through design approaches so sympathetic and well integrated with its site, that buildings, furnishings, and surroundings become part of a unified, interrelated composition.

A well-known example of organic architecture is Fallingwater, the residence Frank Lloyd Wright designed for the Kaufmann family in rural Pennsylvania. Wright had many choices to locate a home on this large site, but chose to place the home directly over the waterfall and creek creating a close, yet noisy dialog with the rushing water and the steep site. The horizontal striations of stone masonry with daring cantilevers of colored beige concrete blend with native rock outcroppings and the wooded environment.

Another Example: Precedent-Based Design



Principled Design



Another Principled Design



Principle Applied to Another Structure





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Source: By 準建築人手札網站 Forgemind ArchiMedia - Flickr: IMG_2489.JPG, CC BY 2.0, Source: https://www.dezeen.gom/2016/08/29/saptiage-gelatingeness/www.dezeen.gom/2016/08/29/saptiage-gelatingeness/www.dezeen.gom/2016/08/29/saptiage-gelatingeness/www.sald-tradeness/saptiage-gelatingeness/saptiage-gelatin

The Overarching Principle

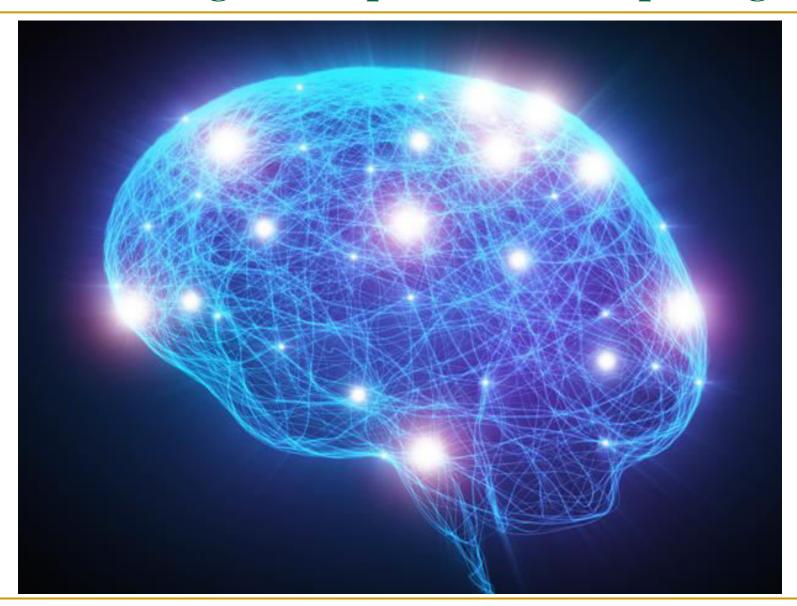
Zoomorphic architecture

From Wikipedia, the free encyclopedia

Zoomorphic architecture is the practice of using animal forms as the inspirational basis and blueprint for architectural design. "While animal forms have always played a role adding some of the deepest layers of meaning in architecture, it is now becoming evident that a new strand of biomorphism is emerging where the meaning derives not from any specific representation but from a more general allusion to biological processes."^[1]

Some well-known examples of Zoomorphic architecture can be found in the TWA Flight Center building in New York City, by Eero Saarinen, or the Milwaukee Art Museum by Santiago Calatrava, both inspired by the form of a bird's wings.^[3]

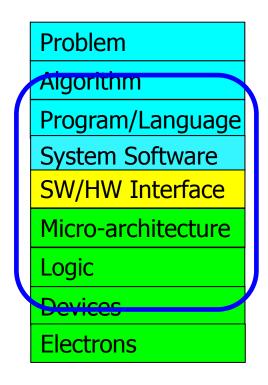
Overarching Principles for Computing?



Concluding Remarks

- It is time to design principled system architectures to solve the memory problem
- Discover design principles for fundamentally secure and reliable computer architectures
- Design complete systems to be balanced and energy-efficient,
 i.e., data-centric (or memory-centric) and low latency
- Enable new and emerging memory architectures
- This can
 - Lead to orders-of-magnitude improvements
 - Enable new applications & computing platforms
 - **-** ...

We Need to Think Across the Stack



If In Doubt, See Other Doubtful Technologies

- A very "doubtful" emerging technology
 - for at least two decades



Proceedings of the IEEE, Sept. 2017

Error Characterization, Mitigation, and Recovery in Flash-Memory-Based Solid-State Drives

This paper reviews the most recent advances in solid-state drive (SSD) error characterization, mitigation, and data recovery techniques to improve both SSD's reliability and lifetime.

By Yu Cai, Saugata Ghose, Erich F. Haratsch, Yixin Luo, and Onur Mutlu



Future Computing Architectures Challenges and Opportunities

(with a Large Focus on Memory)

Onur Mutlu

omutlu@gmail.com

https://people.inf.ethz.ch/omutlu

March 21, 2018

Euromicro PDP 2018 Keynote Talk





Acknowledgments

My current and past students and postdocs

 Rachata Ausavarungnirun, Abhishek Bhowmick, Amirali Boroumand, Rui Cai, Yu Cai, Kevin Chang, Saugata Ghose, Kevin Hsieh, Tyler Huberty, Ben Jaiyen, Samira Khan, Jeremie Kim, Yoongu Kim, Yang Li, Jamie Liu, Lavanya Subramanian, Donghyuk Lee, Yixin Luo, Justin Meza, Gennady Pekhimenko, Vivek Seshadri, Lavanya Subramanian, Nandita Vijaykumar, HanBin Yoon, Jishen Zhao, ...

My collaborators

 Can Alkan, Chita Das, Phil Gibbons, Sriram Govindan, Norm Jouppi, Mahmut Kandemir, Mike Kozuch, Konrad Lai, Ken Mai, Todd Mowry, Yale Patt, Moinuddin Qureshi, Partha Ranganathan, Bikash Sharma, Kushagra Vaid, Chris Wilkerson, ...

Funding Acknowledgments

- NSF
- GSRC
- SRC
- CyLab
- AMD, Google, Facebook, HP Labs, Huawei, IBM, Intel, Microsoft, Nvidia, Oracle, Qualcomm, Rambus, Samsung, Seagate, VMware

Some Open Source Tools

- Rowhammer
 - https://github.com/CMU-SAFARI/rowhammer
- Ramulator Fast and Extensible DRAM Simulator
 - https://github.com/CMU-SAFARI/ramulator
- MemSim
 - https://github.com/CMU-SAFARI/memsim
- NOCulator
 - https://github.com/CMU-SAFARI/NOCulator
- DRAM Error Model
 - http://www.ece.cmu.edu/~safari/tools/memerr/index.html
- Other open-source software from my group
 - https://github.com/CMU-SAFARI/
 - http://www.ece.cmu.edu/~safari/tools.html

Slides Not Covered But Could Be Useful

Performance: In-DRAM Bitwise Operations

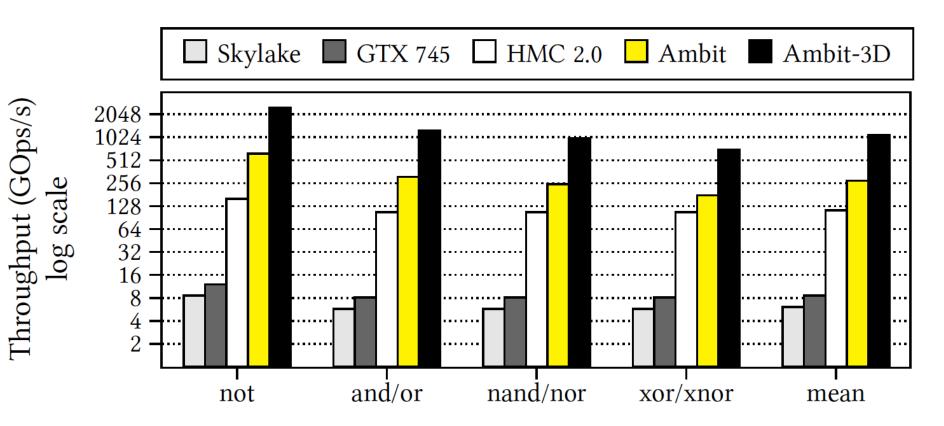


Figure 9: Throughput of bitwise operations on various systems.

Energy of In-DRAM Bitwise Operations

	Design	not	and/or	nand/nor	xor/xnor
DRAM &	DDR3	93.7	137.9	137.9	137.9
Channel Energy	Ambit	1.6	3.2	4.0	5.5
(nJ/KB)	(\downarrow)	59.5X	43.9X	35.1X	25.1X

Table 3: Energy of bitwise operations. (\downarrow) indicates energy reduction of Ambit over the traditional DDR3-based design.

Seshadri+, "Ambit: In-Memory Accelerator for Bulk Bitwise Operations using Commodity DRAM Technology," MICRO 2017.

Open Problems

For More Open Problems, See (I)

Onur Mutlu and Lavanya Subramanian,
 "Research Problems and Opportunities in Memory
 Systems"

Invited Article in <u>Supercomputing Frontiers and Innovations</u> (**SUPERFRI**), 2014/2015.

Research Problems and Opportunities in Memory Systems

Onur Mutlu¹, Lavanya Subramanian¹

For More Open Problems, See (II)

Onur Mutlu,

"The RowHammer Problem and Other Issues We May Face as Memory Becomes Denser"

Invited Paper in Proceedings of the <u>Design, Automation, and Test in</u> <u>Europe Conference</u> (**DATE**), Lausanne, Switzerland, March 2017. [Slides (pptx) (pdf)]

The RowHammer Problem and Other Issues We May Face as Memory Becomes Denser

Onur Mutlu
ETH Zürich
onur.mutlu@inf.ethz.ch
https://people.inf.ethz.ch/omutlu

For More Open Problems, See (III)

Onur Mutlu,
 "Memory Scaling: A Systems Architecture
 Perspective"

Technical talk at <u>MemCon 2013</u> (**MEMCON**), Santa Clara, CA, August 2013. [Slides (pptx) (pdf)]
[Video] [Coverage on StorageSearch]

Memory Scaling: A Systems Architecture Perspective

Onur Mutlu
Carnegie Mellon University
onur@cmu.edu
http://users.ece.cmu.edu/~omutlu/

For More Open Problems, See (IV)

 Yu Cai, Saugata Ghose, Erich F. Haratsch, Yixin Luo, and Onur Mutlu,

"Error Characterization, Mitigation, and Recovery in Flash Memory Based Solid State Drives"

to appear in **Proceedings of the IEEE**, 2017.

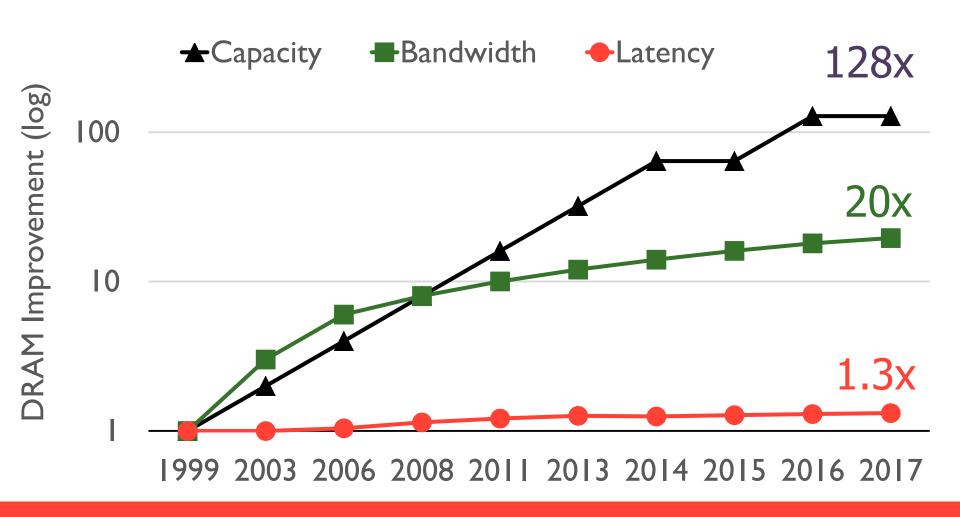
[Preliminary arxiv.org version]

Error Characterization, Mitigation, and Recovery in Flash Memory Based Solid State Drives

Yu Cai, Saugata Ghose, Erich F. Haratsch, Yixin Luo, and Onur Mutlu

Reducing Memory Latency

Main Memory Latency Lags Behind



Memory latency remains almost constant

A Closer Look ...

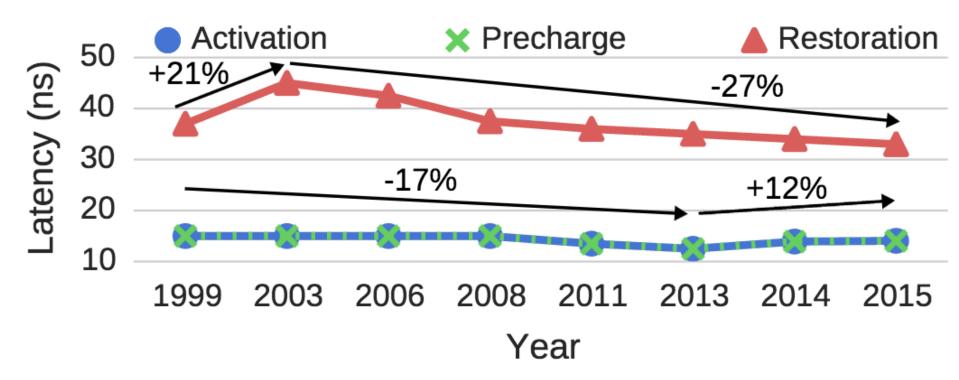


Figure 1: DRAM latency trends over time [20, 21, 23, 51].

Chang+, "<u>Understanding Latency Variation in Modern DRAM Chips: Experimental</u> <u>Characterization, Analysis, and Optimization</u>"," SIGMETRICS 2016.

DRAM Latency Is Critical for Performance



In-memory Databases

[Mao+, EuroSys'12; Clapp+ (Intel), IISWC'15]



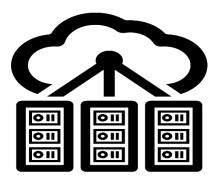
In-Memory Data Analytics

[Clapp+ (Intel), IISWC'15; Awan+, BDCloud'15]



Graph/Tree Processing

[Xu+, IISWC'12; Umuroglu+, FPL'15]



Datacenter Workloads

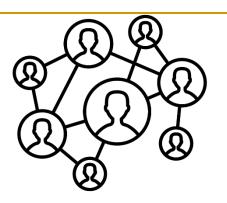
[Kanev+ (Google), ISCA' I 5]



DRAM Latency Is Critical for Performance



In-memory Databases



Graph/Tree Processing

Long memory latency → performance bottleneck



In-Memory Data Analytics

[Clapp+ (Intel), IISWC'15; Awan+, BDCloud'15]



Datacenter Workloads

[Kanev+ (Google), ISCA' 15]

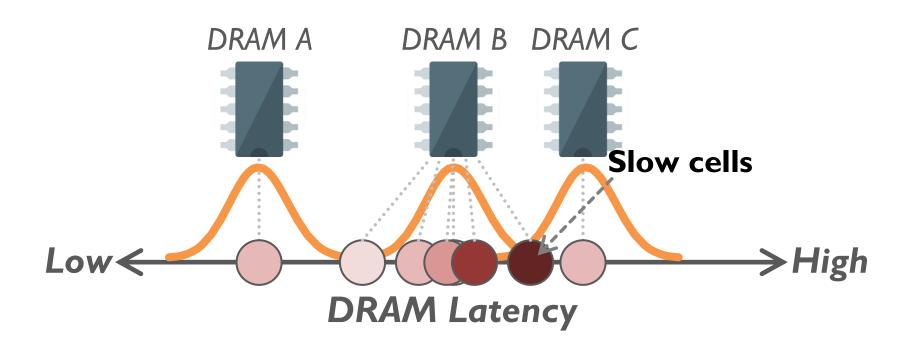


Why the Long Latency?

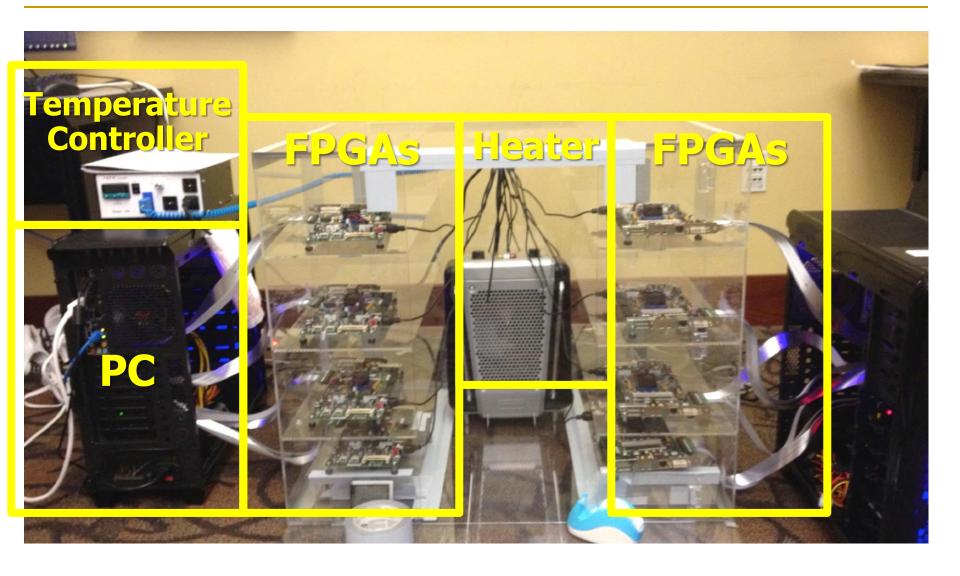
- Design of DRAM uArchitecture
 - Goal: Maximize capacity/area, not minimize latency
- "One size fits all" approach to latency specification
 - Same latency parameters for all temperatures
 - Same latency parameters for all DRAM chips (e.g., rows)
 - Same latency parameters for all parts of a DRAM chip
 - Same latency parameters for all supply voltage levels
 - Same latency parameters for all application data
 - **...**

Latency Variation in Memory Chips

Heterogeneous manufacturing & operating conditions → latency variation in timing parameters



DRAM Characterization Infrastructure

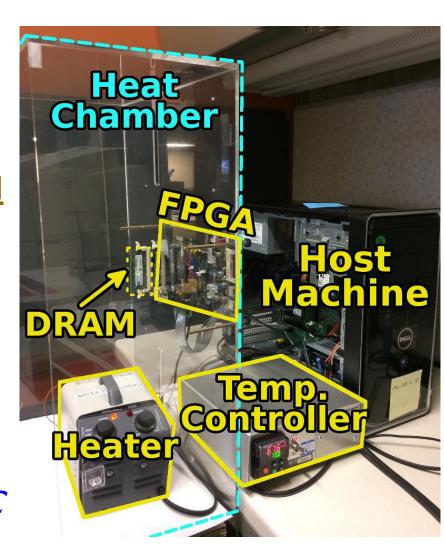


DRAM Characterization Infrastructure

 Hasan Hassan et al., <u>SoftMC: A</u>
 Flexible and Practical Open Source Infrastructure for
 Enabling Experimental DRAM
 Studies, HPCA 2017.



- Easy to Use (C++ API)
- Open-source github.com/CMU-SAFARI/SoftMC



SoftMC: Open Source DRAM Infrastructure

https://github.com/CMU-SAFARI/SoftMC

SoftMC: A Flexible and Practical Open-Source Infrastructure for Enabling Experimental DRAM Studies

```
 Hasan Hassan Nandita Vijaykumar Samira Khan Saugata Ghose Kevin Chang Gennady Pekhimenko Donghyuk Lee^{6,3} Oguz Ergin Onur Mutlu Onur Mutlu
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<sup>1</sup>ETH Zürich <sup>2</sup>TOBB University of Economics & Technology <sup>3</sup>Carnegie Mellon University <sup>4</sup>University of Virginia <sup>5</sup>Microsoft Research <sup>6</sup>NVIDIA Research
```

Tackling the Fixed Latency Mindset

- Reliable operation latency is actually very heterogeneous
 - Across temperatures, chips, parts of a chip, voltage levels, ...
- Idea: Dynamically find out and use the lowest latency one can reliably access a memory location with
 - Adaptive-Latency DRAM [HPCA 2015]
 - Flexible-Latency DRAM [SIGMETRICS 2016]
 - Design-Induced Variation-Aware DRAM [SIGMETRICS 2017]
 - Voltron [SIGMETRICS 2017]
 - **-** ...
- We would like to find sources of latency heterogeneity and exploit them to minimize latency

Adaptive-Latency DRAM

- Key idea
 - Optimize DRAM timing parameters online
- Two components
 - DRAM manufacturer provides multiple sets of reliable DRAM timing parameters at different temperatures for each DIMM
 - System monitors DRAM temperature & uses appropriate DRAM timing parameters



Latency Reduction Summary of 115 DIMMs

- Latency reduction for read & write (55°C)
 - Read Latency: 32.7%
 - Write Latency: 55.1%
- Latency reduction for each timing parameter (55°C)
 - Sensing: 17.3%
 - Restore: 37.3% (read), 54.8% (write)
 - Precharge: 35.2%

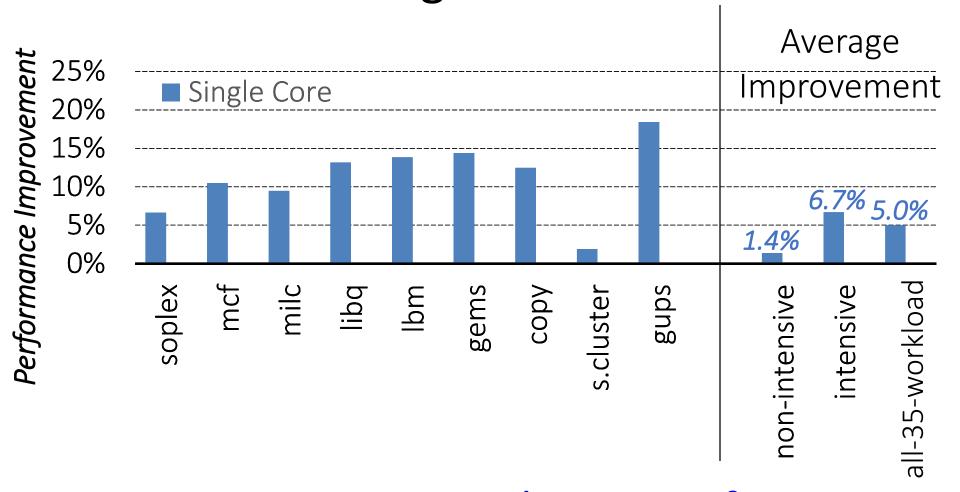


AL-DRAM: Real System Evaluation

- System
 - CPU: AMD 4386 (8 Cores, 3.1GHz, 8MB LLC)

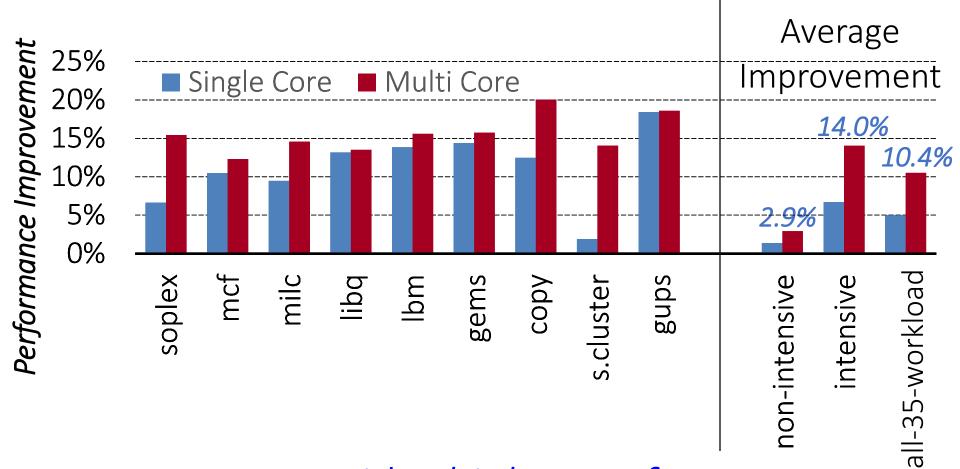
D18F2x200 dct[0] mp[1:0] DDR3 DRAM Timing 0 Reset: 0F05 0505h. See 2.9.3 [DCT Configuration Registers]. Description Bits Reserved. 31:30 29:24 Tras: row active strobe. Read-write. BIOS: See 2.9.7.5 [SPD ROM-Based Configuration]. Specifies the minimum time in memory clock cycles from an activate command to a precharge command, both to the same chip select bank. Bits Description 07h-00h Reserved 2Ah-08h <Tras> clocks 3Fh-2Bh Reserved 23:21 Reserved 20:16 Trp: row precharge time. Read-write. BIOS: See 2.9.7.5 [SPD ROM-Based Configuration]. Specifies the minimum time in memory clock cycles from a precharge command to an activate command or auto refresh command, both to the same bank.

AL-DRAM: Single-Core Evaluation



AL-DRAM improves single-core performance on a real system

AL-DRAM: Multi-Core Evaluation



AL-DRAM provides higher performance on

multi-programmed & multi-threaded workloads

Reducing Latency Also Reduces Energy

- AL-DRAM reduces DRAM power consumption by 5.8%
- Major reason: reduction in row activation time

More on Adaptive-Latency DRAM

 Donghyuk Lee, Yoongu Kim, Gennady Pekhimenko, Samira Khan, Vivek Seshadri, Kevin Chang, and Onur Mutlu,
 "Adaptive-Latency DRAM: Optimizing DRAM Timing for the Common-Case"

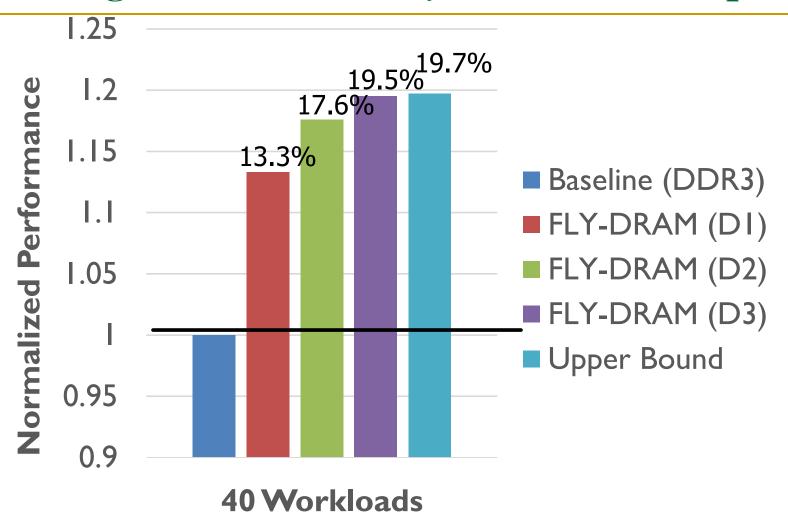
Proceedings of the <u>21st International Symposium on High-</u> <u>Performance Computer Architecture</u> (**HPCA**), Bay Area, CA, February 2015.

[Slides (pptx) (pdf)] [Full data sets]

Adaptive-Latency DRAM: Optimizing DRAM Timing for the Common-Case

Donghyuk Lee Yoongu Kim Gennady Pekhimenko Samira Khan Vivek Seshadri Kevin Chang Onur Mutlu Carnegie Mellon University

Heterogeneous Latency within A Chip



Chang+, "<u>Understanding Latency Variation in Modern DRAM Chips: Experimental</u>
<u>Characterization, Analysis, and Optimization</u>"," SIGMETRICS 2016.

Analysis of Latency Variation in DRAM Chips

Kevin Chang, Abhijith Kashyap, Hasan Hassan, Samira Khan, Kevin Hsieh, Donghyuk Lee, Saugata Ghose, Gennady Pekhimenko, Tianshi Li, and Onur Mutlu,

"Understanding Latency Variation in Modern DRAM Chips: **Experimental Characterization, Analysis, and Optimization**

Proceedings of the <u>ACM International Conference on Measurement and</u> Modeling of Computer Systems (SIGMETRICS), Antibes Juan-Les-Pins, France, June 2016.

[Slides (pptx) (pdf)]

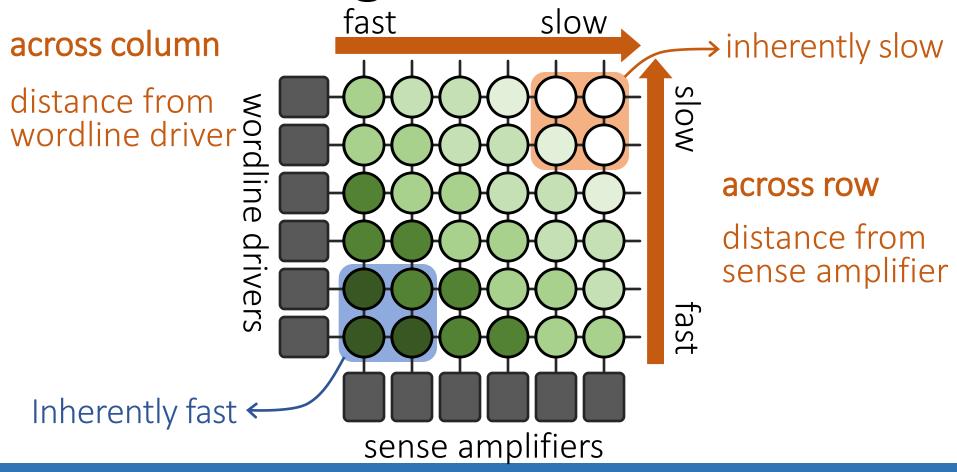
Source Code

Understanding Latency Variation in Modern DRAM Chips: Experimental Characterization, Analysis, and Optimization

Kevin K. Chang¹ Abhijith Kashyap¹ Hasan Hassan^{1,2} Saugata Ghose¹ Kevin Hsieh¹ Donghyuk Lee¹ Tianshi Li^{1,3} Gennady Pekhimenko¹ Samira Khan⁴ Onur Mutlu^{5,1}

¹Carnegie Mellon University ²TOBB ETÜ ³Peking University ⁴University of Virginia ⁵ETH Zürich SAFARI

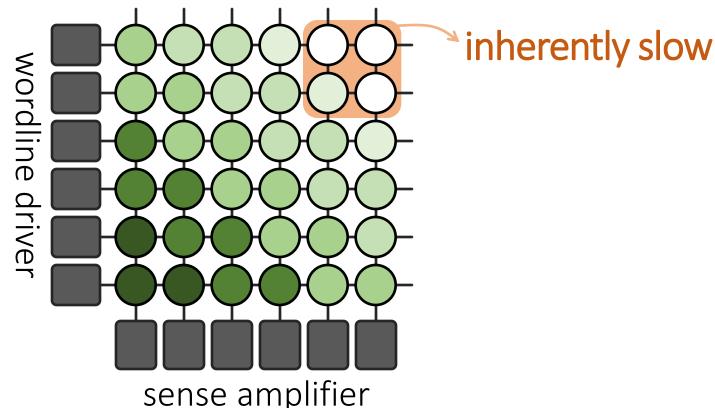
What Is Design-Induced Variation?



Systematic variation in cell access times caused by the **physical organization** of DRAM

DIVA Online **Profiling**

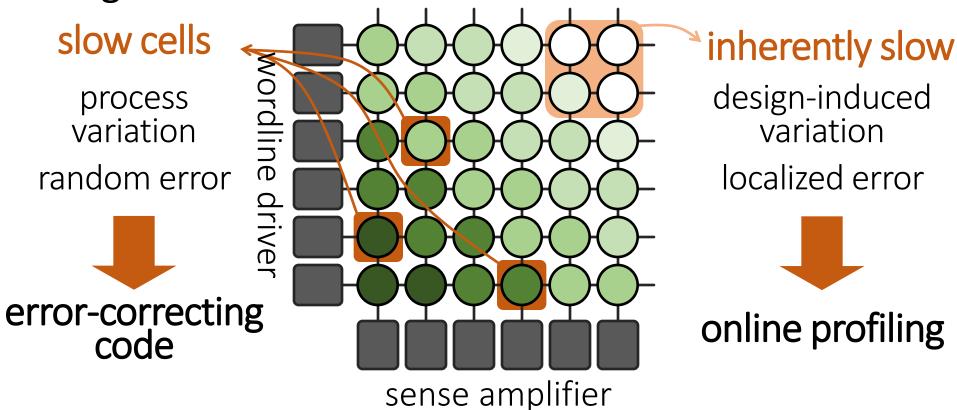
Design-Induced-Variation-Aware



Profile *only slow regions* to determine min. latency Dynamic & low cost latency optimization

DIVA Online **Profiling**

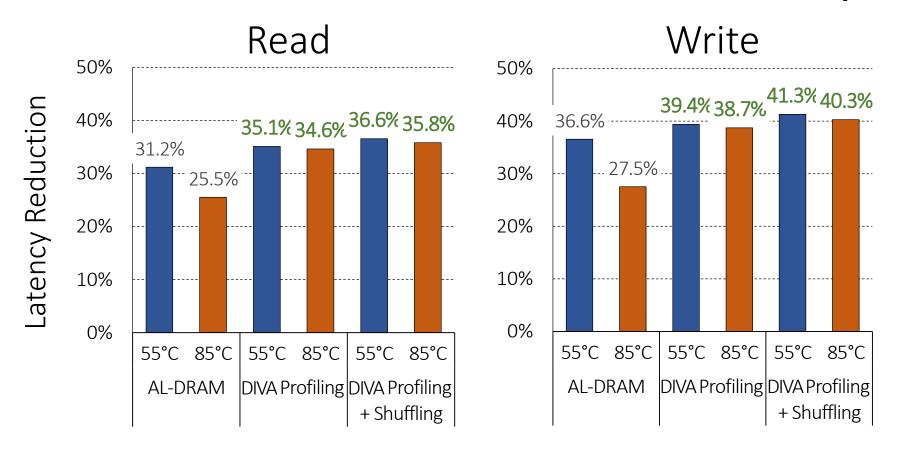
Design-Induced-Variation-Aware



Combine error-correcting codes & online profiling

Reliably reduce DRAM latency

DIVA-DRAM Reduces Latency



DIVA-DRAM *reduces latency more aggressively* and uses ECC to correct random slow cells



Design-Induced Latency Variation in DRAM

 Donghyuk Lee, Samira Khan, Lavanya Subramanian, Saugata Ghose, Rachata Ausavarungnirun, Gennady Pekhimenko, Vivek Seshadri, and Onur Mutlu,

"Design-Induced Latency Variation in Modern DRAM Chips:
Characterization, Analysis, and Latency Reduction Mechanisms"
Proceedings of the ACM International Conference on Measurement and
Modeling of Computer Systems (SIGMETRICS), Urbana-Champaign, IL,
USA, June 2017.

Design-Induced Latency Variation in Modern DRAM Chips: Characterization, Analysis, and Latency Reduction Mechanisms

Donghyuk Lee, NVIDIA and Carnegie Mellon University
Samira Khan, University of Virginia
Lavanya Subramanian, Saugata Ghose, Rachata Ausavarungnirun, Carnegie Mellon University
Gennady Pekhimenko, Vivek Seshadri, Microsoft Research
Onur Mutlu, ETH Zürich and Carnegie Mellon University

Voltron: Exploiting the Voltage-Latency-Reliability Relationship

Executive Summary

- DRAM (memory) power is significant in today's systems
 - Existing low-voltage DRAM reduces voltage conservatively
- Goal: Understand and exploit the reliability and latency behavior of real DRAM chips under aggressive reduced-voltage operation
- Key experimental observations:
 - Huge voltage margin -- Errors occur beyond some voltage
 - Errors exhibit spatial locality
 - Higher operation latency mitigates voltage-induced errors
- Voltron: A new DRAM energy reduction mechanism
 - Reduce DRAM voltage without introducing errors
 - Use a regression model to select voltage that does not degrade performance beyond a chosen target → 7.3% system energy reduction

Analysis of Latency-Voltage in DRAM Chips

 Kevin Chang, A. Giray Yaglikci, Saugata Ghose, Aditya Agrawal, Niladrish Chatterjee, Abhijith Kashyap, Donghyuk Lee, Mike O'Connor, Hasan Hassan, and <u>Onur Mutlu</u>,

"Understanding Reduced-Voltage Operation in Modern DRAM Devices: Experimental Characterization, Analysis, and Mechanisms"

Proceedings of the <u>ACM International Conference on Measurement and</u> <u>Modeling of Computer Systems</u> (**SIGMETRICS**), Urbana-Champaign, IL, USA, June 2017.

Understanding Reduced-Voltage Operation in Modern DRAM Chips: Characterization, Analysis, and Mechanisms

Kevin K. Chang[†] Abdullah Giray Yağlıkçı[†] Saugata Ghose[†] Aditya Agrawal[¶] Niladrish Chatterjee[¶] Abhijith Kashyap[†] Donghyuk Lee[¶] Mike O'Connor^{¶,‡} Hasan Hassan[§] Onur Mutlu^{§,†}

†Carnegie Mellon University [¶]NVIDIA [‡]The University of Texas at Austin [§]ETH Zürich

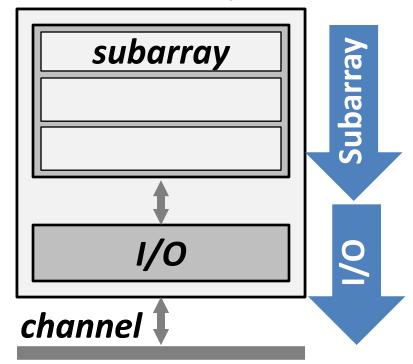
And, What If ...

... we can sacrifice reliability of some data to access it with even lower latency?

Tiered Latency DRAM

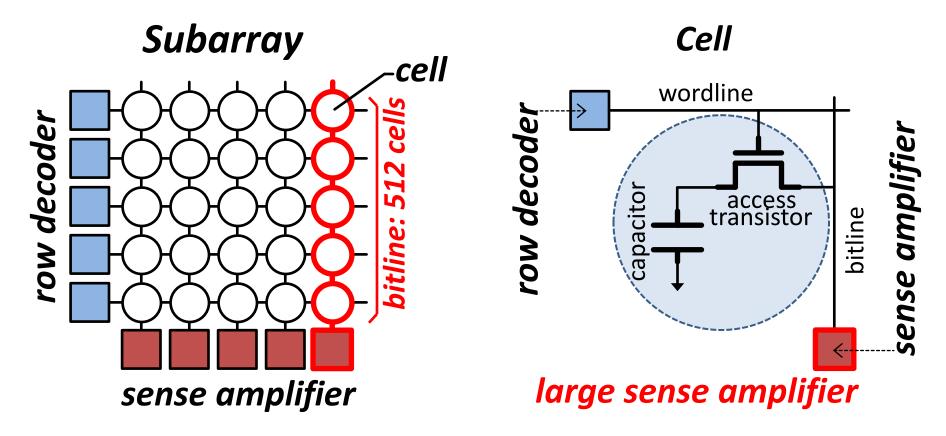
What Causes the Long Latency?

DRAM Chip





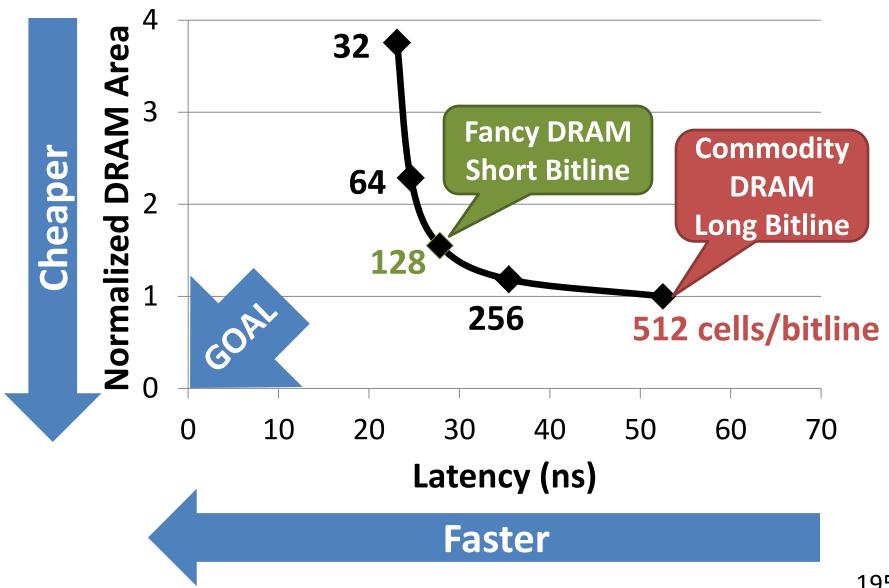
Why is the Subarray So Slow?



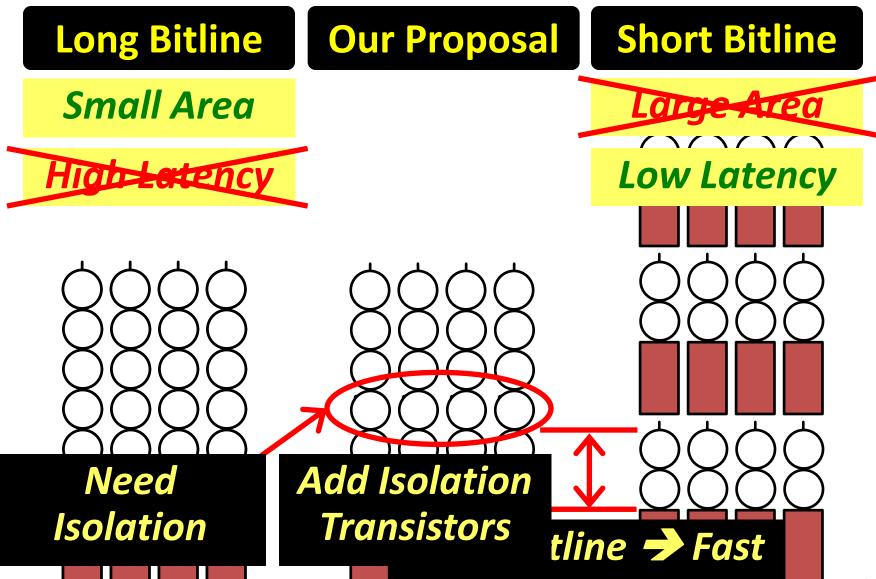
- Long bitline
 - Amortizes sense amplifier cost → Small area
 - Large bitline capacitance → High latency & power

Trade-Off: Area (Die Size) vs. Latency **Long Bitline Short Bitline Faster Smaller** Trade-Off: Area vs. Latency

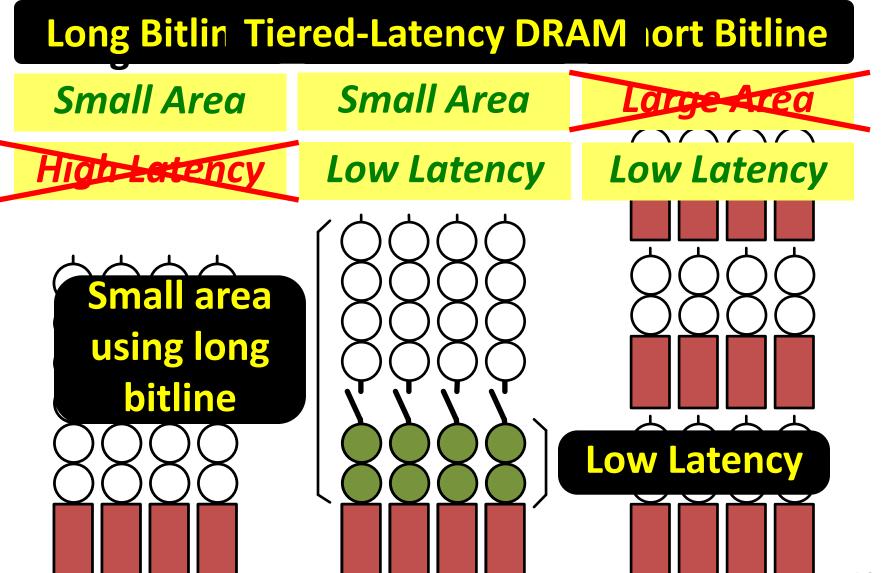
Trade-Off: Area (Die Size) vs. Latency



Approximating the Best of Both Worlds

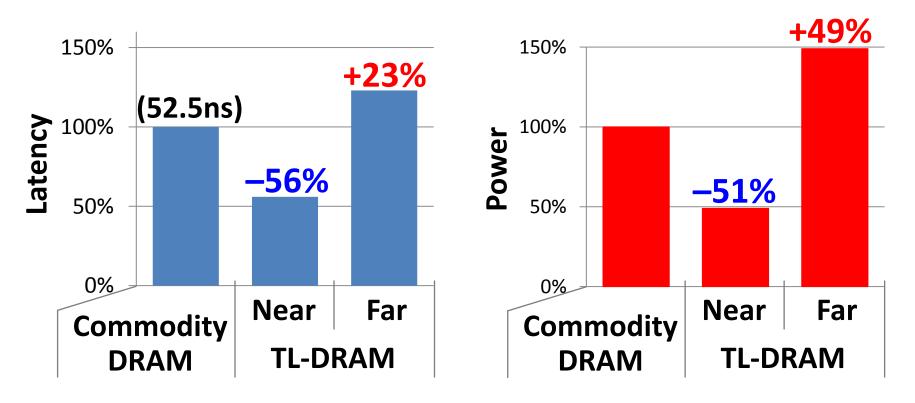


Approximating the Best of Both Worlds



Commodity DRAM vs. TL-DRAM [HPCA 2013]

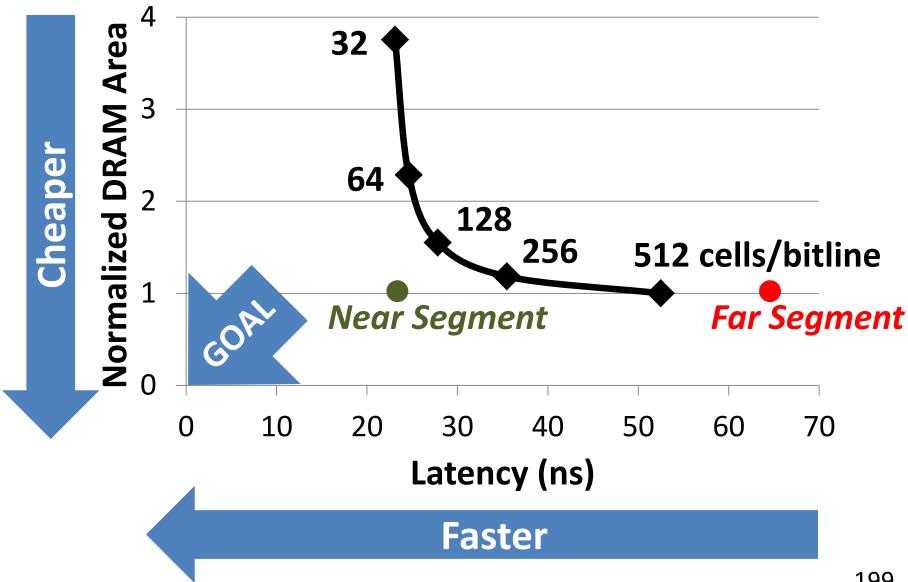
DRAM Latency (tRC)
 DRAM Power



DRAM Area Overhead

~3%: mainly due to the isolation transistors

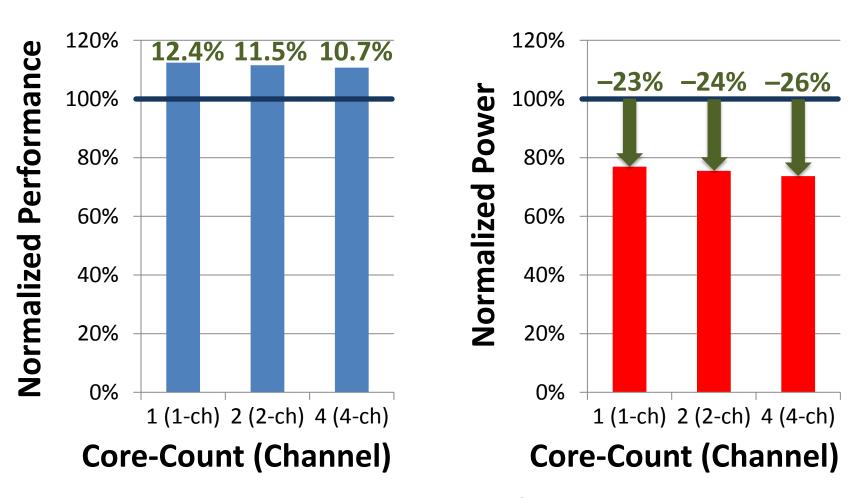
Trade-Off: Area (Die-Area) vs. Latency



Leveraging Tiered-Latency DRAM

- TL-DRAM is a substrate that can be leveraged by the hardware and/or software
- Many potential uses
 - 1. Use near segment as hardware-managed *inclusive* cache to far segment
 - 2. Use near segment as hardware-managed *exclusive* cache to far segment
 - 3. Profile-based page mapping by operating system
 - 4. Simply replace DRAM with TL-DRAM

Performance & Power Consumption



Using near segment as a cache improves performance and reduces power consumption

More on PIM

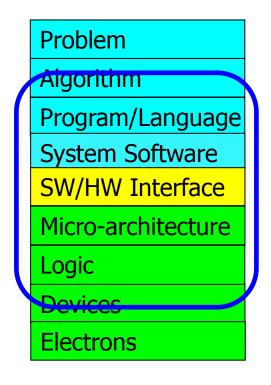
Eliminating the Adoption Barriers

How to Enable Adoption of Processing in Memory

Barriers to Adoption of PIM

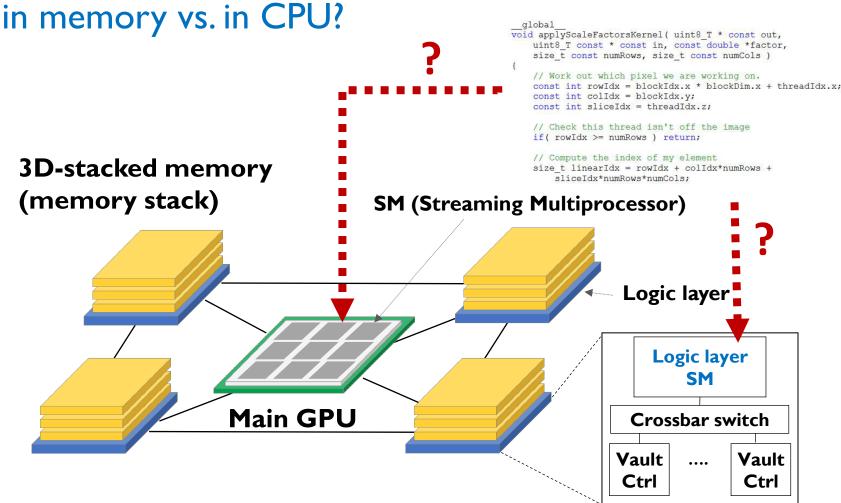
- 1. Functionality of and applications for PIM
- 2. Ease of programming (interfaces and compiler/HW support)
- 3. System support: coherence & virtual memory
- 4. Runtime systems for adaptive scheduling, data mapping, access/sharing control
- 5. Infrastructures to assess benefits and feasibility

We Need to Revisit the Entire Stack



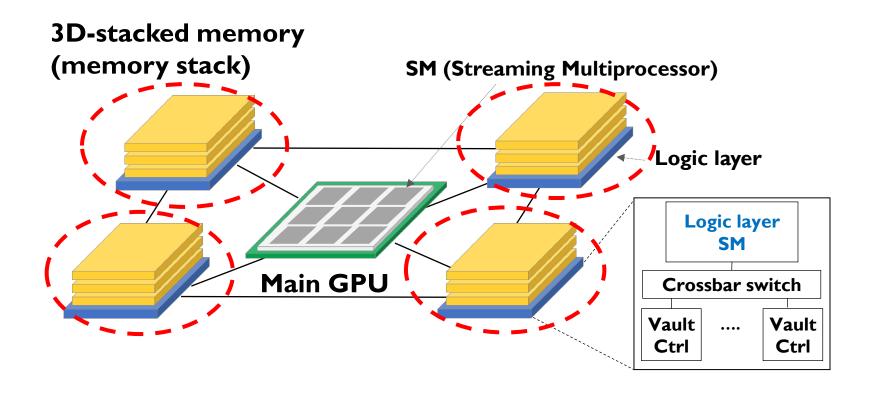
Key Challenge 1: Code Mapping

• Challenge 1: Which operations should be executed in mamory vs. in CDLD



Key Challenge 2: Data Mapping

• Challenge 2: How should data be mapped to different 3D memory stacks?



How to Do the Code and Data Mapping?

Kevin Hsieh, Eiman Ebrahimi, Gwangsun Kim, Niladrish Chatterjee, Mike O'Connor, Nandita Vijaykumar, Onur Mutlu, and Stephen W. Keckler, "Transparent Offloading and Mapping (TOM): Enabling Programmer-Transparent Near-Data Processing in GPU Systems"

Proceedings of the <u>43rd International Symposium on Computer</u>
<u>Architecture</u> (**ISCA**), Seoul, South Korea, June 2016.
[Slides (pptx) (pdf)]

[Lightning Session Slides (pptx) (pdf)]

Transparent Offloading and Mapping (TOM): Enabling Programmer-Transparent Near-Data Processing in GPU Systems

Kevin Hsieh[‡] Eiman Ebrahimi[†] Gwangsun Kim* Niladrish Chatterjee[†] Mike O'Connor[†] Nandita Vijaykumar[‡] Onur Mutlu^{§‡} Stephen W. Keckler[†] [‡]Carnegie Mellon University [†]NVIDIA *KAIST [§]ETH Zürich

How to Schedule Code?

Ashutosh Pattnaik, Xulong Tang, Adwait Jog, Onur Kayiran, Asit K.
 Mishra, Mahmut T. Kandemir, Onur Mutlu, and Chita R. Das,
 "Scheduling Techniques for GPU Architectures with Processing-In-Memory Capabilities"

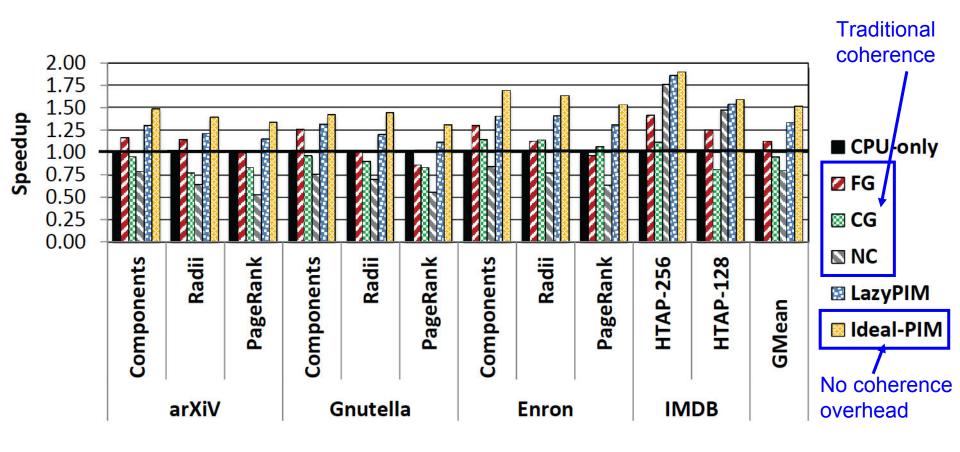
Proceedings of the <u>25th International Conference on Parallel</u>
<u>Architectures and Compilation Techniques</u> (**PACT**), Haifa, Israel,
September 2016.

Scheduling Techniques for GPU Architectures with Processing-In-Memory Capabilities

Ashutosh Pattnaik¹ Xulong Tang¹ Adwait Jog² Onur Kayıran³ Asit K. Mishra⁴ Mahmut T. Kandemir¹ Onur Mutlu^{5,6} Chita R. Das¹

¹Pennsylvania State University ²College of William and Mary ³Advanced Micro Devices, Inc. ⁴Intel Labs ⁵ETH Zürich ⁶Carnegie Mellon University

Challenge: Coherence for Hybrid CPU-PIM Apps



How to Maintain Coherence?

Amirali Boroumand, Saugata Ghose, Minesh Patel, Hasan Hassan, Brandon Lucia, Kevin Hsieh, Krishna T. Malladi, Hongzhong Zheng, and Onur Mutlu,
 "LazyPIM: An Efficient Cache Coherence Mechanism for Processing-in-Memory"
 IEEE Computer Architecture Letters (CAL), June 2016.

LazyPIM: An Efficient Cache Coherence Mechanism for Processing-in-Memory

Amirali Boroumand[†], Saugata Ghose[†], Minesh Patel[†], Hasan Hassan[†], Brandon Lucia[†], Kevin Hsieh[†], Krishna T. Malladi^{*}, Hongzhong Zheng^{*}, and Onur Mutlu^{‡†}

† Carnegie Mellon University * Samsung Semiconductor, Inc. § TOBB ETÜ [‡] ETH Zürich

How to Support Virtual Memory?

Kevin Hsieh, Samira Khan, Nandita Vijaykumar, Kevin K. Chang, Amirali Boroumand, Saugata Ghose, and Onur Mutlu,
 "Accelerating Pointer Chasing in 3D-Stacked Memory:
 Challenges, Mechanisms, Evaluation"
 Proceedings of the 34th IEEE International Conference on Computer
 Design (ICCD), Phoenix, AZ, USA, October 2016.

Accelerating Pointer Chasing in 3D-Stacked Memory: Challenges, Mechanisms, Evaluation

Kevin Hsieh[†] Samira Khan[‡] Nandita Vijaykumar[†] Kevin K. Chang[†] Amirali Boroumand[†] Saugata Ghose[†] Onur Mutlu^{§†} [†] Carnegie Mellon University [‡] University of Virginia [§] ETH Zürich

How to Design Data Structures for PIM?

Zhiyu Liu, Irina Calciu, Maurice Herlihy, and Onur Mutlu,
 "Concurrent Data Structures for Near-Memory Computing"
 Proceedings of the <u>29th ACM Symposium on Parallelism in Algorithms</u>
 and Architectures (SPAA), Washington, DC, USA, July 2017.
 [Slides (pptx) (pdf)]

Concurrent Data Structures for Near-Memory Computing

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Simulation Infrastructures for PIM

- Ramulator extended for PIM
 - Flexible and extensible DRAM simulator
 - Can model many different memory standards and proposals
 - Kim+, "Ramulator: A Flexible and Extensible DRAM Simulator", IEEE CAL 2015.
 - https://github.com/CMU-SAFARI/ramulator

Ramulator: A Fast and Extensible DRAM Simulator

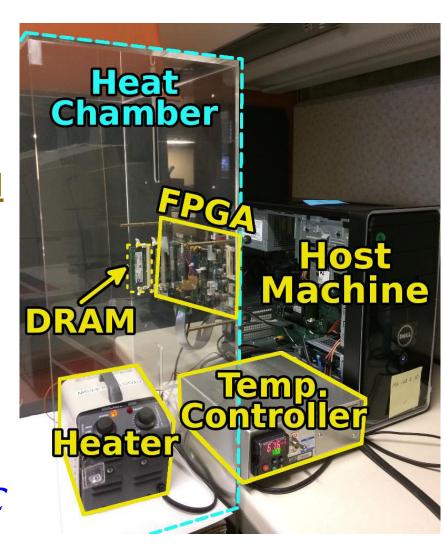
Yoongu Kim¹ Weikun Yang^{1,2} Onur Mutlu¹ Carnegie Mellon University ²Peking University

An FPGA-based Test-bed for PIM?

 Hasan Hassan et al., <u>SoftMC: A</u>
 Flexible and Practical Open Source Infrastructure for
 Enabling Experimental DRAM
 Studies HPCA 2017.



- Easy to Use (C++ API)
- Open-source github.com/CMU-SAFARI/SoftMC



Industry Is Writing Papers About It, Too

DRAM Process Scaling Challenges

Refresh

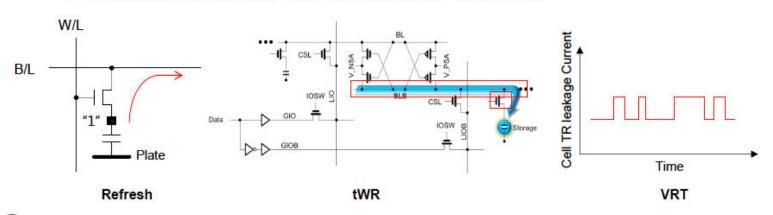
- Difficult to build high-aspect ratio cell capacitors decreasing cell capacitance
- · Leakage current of cell access transistors increasing

tWR

- Contact resistance between the cell capacitor and access transistor increasing
- · On-current of the cell access transistor decreasing
- Bit-line resistance increasing

VRT

· Occurring more frequently with cell capacitance decreasing









Industry Is Writing Papers About It, Too

DRAM Process Scaling Challenges

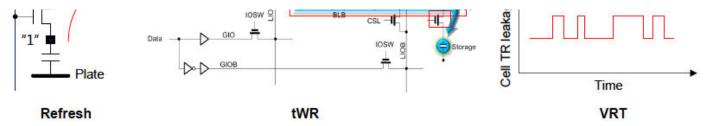
Refresh

Difficult to build high-aspect ratio cell capacitors decreasing cell capacitance
 THE MEMORY FORUM 2014

Co-Architecting Controllers and DRAM to Enhance DRAM Process Scaling

Uksong Kang, Hak-soo Yu, Churoo Park, *Hongzhong Zheng, **John Halbert, **Kuljit Bains, SeongJin Jang, and Joo Sun Choi

Samsung Electronics, Hwasung, Korea / *Samsung Electronics, San Jose / **Intel







How Do We Keep Systems Secure?

- Understand: Methodologies for failure modeling and discovery
 - Modeling and prediction based on real (device) data

- Architect: Principled co-architecting of system and memory
 - Good partitioning of duties across the stack

- Design & Test: Principled design, automation, testing
 - High coverage and good interaction with system reliability methods

Ramulator: A Fast and Extensible DRAM Simulator [IEEE Comp Arch Letters'15]

Ramulator Motivation

- DRAM and Memory Controller landscape is changing
- Many new and upcoming standards
- Many new controller designs
- A fast and easy-to-extend simulator is very much needed

Segment	DRAM Standards & Architectures
Commodity	DDR3 (2007) [14]; DDR4 (2012) [18]
Low-Power	LPDDR3 (2012) [17]; LPDDR4 (2014) [20]
Graphics	GDDR5 (2009) [15]
Performance	eDRAM [28], [32]; RLDRAM3 (2011) [29]
3D-Stacked	WIO (2011) [16]; WIO2 (2014) [21]; MCDRAM (2015) [13]; HBM (2013) [19]; HMC1.0 (2013) [10]; HMC1.1 (2014) [11]
Academic	SBA/SSA (2010) [38]; Staged Reads (2012) [8]; RAIDR (2012) [27]; SALP (2012) [24]; TL-DRAM (2013) [26]; RowClone (2013) [37]; Half-DRAM (2014) [39]; Row-Buffer Decoupling (2014) [33]; SARP (2014) [6]; AL-DRAM (2015) [25]



Ramulator

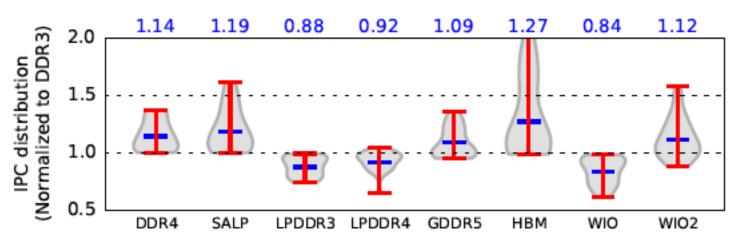
- Provides out-of-the box support for many DRAM standards:
 - DDR3/4, LPDDR3/4, GDDR5, WIO1/2, HBM, plus new proposals (SALP, AL-DRAM, TLDRAM, RowClone, and SARP)
- ~2.5X faster than fastest open-source simulator
- Modular and extensible to different standards

Simulator	Cycles (10 ⁶)		Runtime (sec.)		Reg/sec (10 ³)		Memory	
(clang -03)	Random	Stream	Random	Stream	Random	Stream	(MB)	
Ramulator	652	411	752	249	133	402	2.1	
DRAMSim2	645	413	2,030	876	49	114	1.2	
USIMM	661	409	1,880	750	53	133	4.5	
DrSim	647	406	18,109	12,984	6	8	1.6	
NVMain	666	413	6,881	5,023	15	20	4,230.0	

Table 3. Comparison of five simulators using two traces

Case Study: Comparison of DRAM Standards

Standard	Rate (MT/s)	Timing (CL-RCD-RP)	Data-Bus (Width×Chan.)	Rank-per-Chan	BW (GB/s)
DDR3	1,600	11-11-11	64-bit × 1	1	11.9
DDR4	2,400	16-16-16	64 -bit $\times 1$	1	17.9
SALP [†]	1,600	11-11-11	64 -bit $\times 1$	1	11.9
LPDDR3	1,600	12-15-15	64 -bit $\times 1$	1	11.9
LPDDR4	2,400	22-22-22	32 -bit $\times 2^*$	1	17.9
GDDR5 [12]	6,000	18-18-18	64 -bit $\times 1$	1	44.7
HBM	1,000	7-7-7	128 -bit \times 8 *	1	119.2
WIO	266	7-7-7	128 -bit $\times 4^*$	1	15.9
WIO2	1,066	9-10-10	128 -bit \times $8*$	1	127.2



Across 22 workloads, simple CPU model

Figure 2. Performance comparison of DRAM standards



Ramulator Paper and Source Code

- Yoongu Kim, Weikun Yang, and Onur Mutlu,
 "Ramulator: A Fast and Extensible DRAM Simulator"
 IEEE Computer Architecture Letters (CAL), March 2015.
 [Source Code]
- Source code is released under the liberal MIT License
 - https://github.com/CMU-SAFARI/ramulator

End of Backup Slides