Intelligent Architectures for Intelligent Machines

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2 December 2020

Huawei Compute and Storage Technology Workshop Keynote

SAFARI

ETH zürich

Carnegie Mellon



Computing is Bottlenecked by Data



Data is Key for AI, ML, Genomics, ...

Important workloads are all data intensive

 They require rapid and efficient processing of large amounts of data

- Data is increasing
 - □ We can generate more than we can process

Data is Key for Future Workloads



In-memory Databases

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

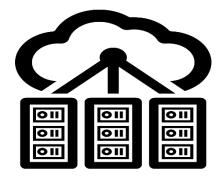


In-Memory Data Analytics

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



Graph/Tree Processing [Xu+, IISWC'12; Umuroglu+, FPL'15]



Datacenter Workloads [Kanev+ (**Google**), ISCA'15]

Data Overwhelms Modern Machines





In-memory Databases

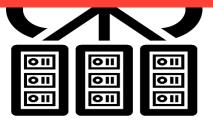
Graph/Tree Processing

Data → performance & energy bottleneck



In-Memory Data Analytics

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



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

Data is Key for Future Workloads



Chrome

Google's web browser



TensorFlow Mobile

Google's machine learning framework



Google's video codec



Data Overwhelms Modern Machines



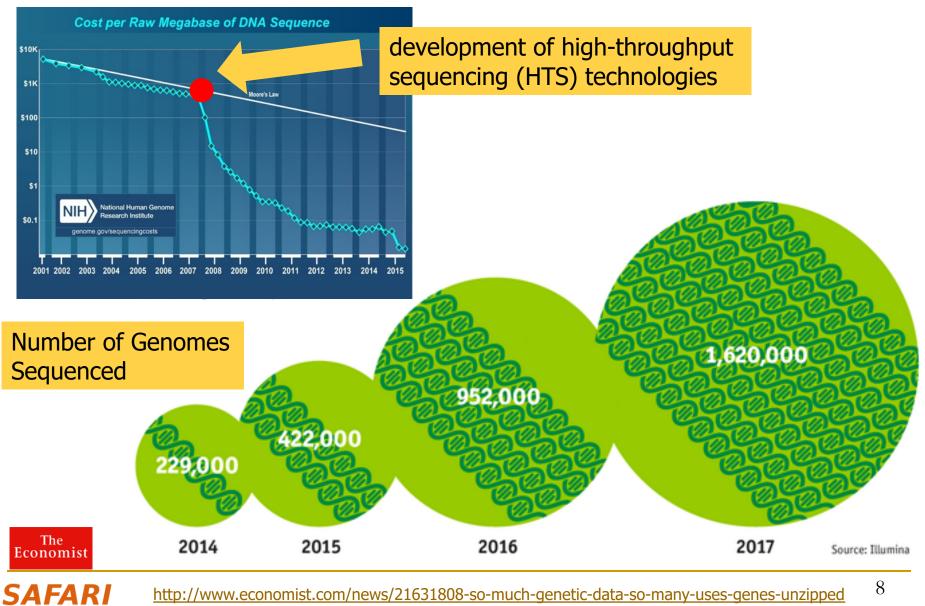
Data → performance & energy bottleneck



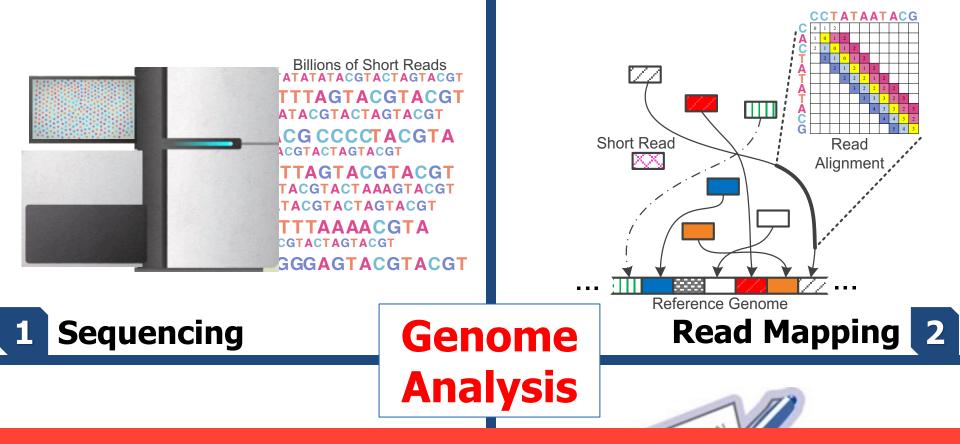
Google's video codec



Data is Key for Future Workloads



http://www.economist.com/news/21631808-so-much-genetic-data-so-many-uses-genes-unzipped



Data → performance & energy bottleneck

read4: CGCTTCCAT read5: CCATGACGC read6: TTCCATGAC

Variant Calling

3

Scientific Discovery 4

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 🖾, Jeremie S Kim, Saugata Ghose, Can Alkan, Onur Mutlu

Briefings in Bioinformatics, bby017, https://doi.org/10.1093/bib/bby017Published:02 April 2018Article history ▼



Oxford Nanopore MinION

Senol Cali+, "Nanopore Sequencing Technology and Tools for Genome Assembly: Computational Analysis of the Current State, Bottlenecks and Future Directions," Briefings in Bioinformatics, 2018. [Open arxiv.org version]

New Genome Sequencing Technologies

Nanopore sequencing technology and tools for genome assembly: computational analysis of the current state, bottlenecks and future directions

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Briefings in Bioinformatics, bby017, https://doi.org/10.1093/bib/bby017 Published: 02 April 2018 Article history ▼



Oxford Nanopore MinION

Data → performance & energy bottleneck

Accelerating Genome Analysis

 Mohammed Alser, Zulal Bingol, Damla Senol Cali, Jeremie Kim, Saugata Ghose, Can Alkan, and Onur Mutlu,
 <u>"Accelerating Genome Analysis: A Primer on an Ongoing Journey"</u> <u>IEEE Micro</u> (*IEEE MICRO*), Vol. 40, No. 5, pages 65-75, September/October 2020.
 [Slides (pptx)(pdf)]
 [Talk Video (1 hour 2 minutes)]

Accelerating Genome Analysis: A Primer on an Ongoing Journey

Mohammed Alser ETH Zürich

Zülal Bingöl Bilkent University

SAFAR

Damla Senol Cali Carnegie Mellon University

Jeremie Kim ETH Zurich and Carnegie Mellon University Saugata Ghose University of Illinois at Urbana–Champaign and Carnegie Mellon University

Can Alkan Bilkent University

Onur Mutlu ETH Zurich, Carnegie Mellon University, and Bilkent University

GenASM Framework [MICRO 2020]

Damla Senol Cali, Gurpreet S. Kalsi, Zulal Bingol, Can Firtina, Lavanya Subramanian, Jeremie S. Kim, Rachata Ausavarungnirun, Mohammed Alser, Juan Gomez-Luna, Amirali Boroumand, Anant Nori, Allison Scibisz, Sreenivas Subramoney, Can Alkan, Saugata Ghose, and Onur Mutlu, "GenASM: A High-Performance, Low-Power Approximate String Matching
 <u>Acceleration Framework for Genome Sequence Analysis"</u>
 *Proceedings of the <u>53rd International Symposium on Microarchitecture</u> (MICRO), Virtual, October 2020.
 [Lighting Talk Video (1.5 minutes)]
 [Lightning Talk Slides (pptx) (pdf)]
 [Talk Video (18 minutes)]
 [Slides (pptx) (pdf)]*

GenASM: A High-Performance, Low-Power Approximate String Matching Acceleration Framework for Genome Sequence Analysis

Damla Senol Cali[†][™] Gurpreet S. Kalsi[™] Zülal Bingöl[▽] Can Firtina[◊] Lavanya Subramanian[‡] Jeremie S. Kim^{◊†} Rachata Ausavarungnirun[⊙] Mohammed Alser[◊] Juan Gomez-Luna[◊] Amirali Boroumand[†] Anant Nori[™] Allison Scibisz[†] Sreenivas Subramoney[™] Can Alkan[▽] Saugata Ghose^{*†} Onur Mutlu^{◊†▽} [†]Carnegie Mellon University [™]Processor Architecture Research Lab, Intel Labs [¬]Bilkent University [◊]ETH Zürich [‡]Facebook [⊙]King Mongkut's University of Technology North Bangkok ^{*}University of Illinois at Urbana–Champaign 13

Data Overwhelms Modern Machines ...

Storage/memory capability

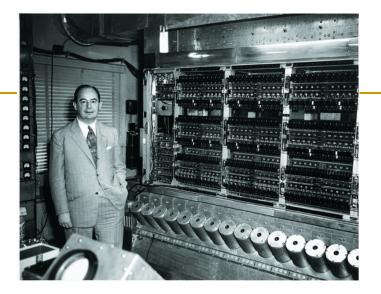
Communication capability

Computation capability

Greatly impacts robustness, energy, performance, cost

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

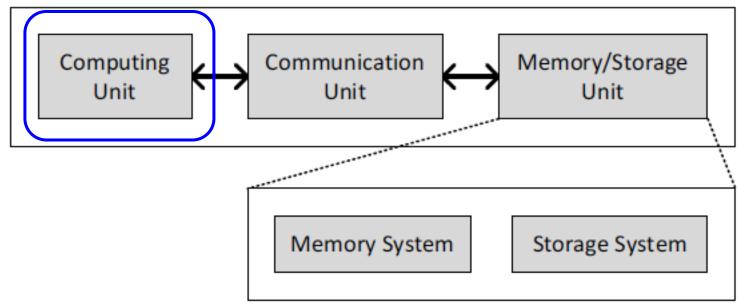
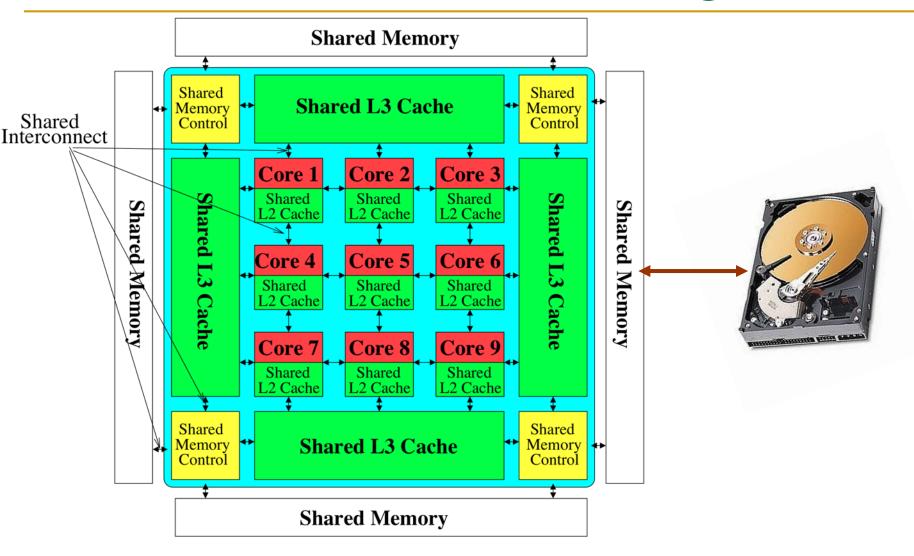


Image source: https://lbsitbytes2010.wordpress.com/2013/03/29/john-von-neumann-roll-no-15/

Perils of Processor-Centric Design



Most of the system is dedicated to storing and moving data

Data Overwhelms Modern Machines



Data → performance & energy bottleneck



Google's video codec



Data Movement Overwhelms Modern Machines

 Amirali Boroumand, Saugata Ghose, Youngsok Kim, Rachata Ausavarungnirun, Eric Shiu, Rahul Thakur, Daehyun Kim, Aki Kuusela, Allan Knies, Parthasarathy Ranganathan, and Onur Mutlu, "Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks" Proceedings of the <u>23rd International Conference on Architectural Support for Programming</u> <u>Languages and Operating Systems</u> (ASPLOS), Williamsburg, VA, USA, March 2018.

62.7% of the total system energy is spent on data movement

Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks

Amirali Boroumand¹Saugata Ghose¹Youngsok Kim²Rachata Ausavarungnirun¹Eric Shiu³Rahul Thakur³Daehyun Kim^{4,3}Aki Kuusela³Allan Knies³Parthasarathy Ranganathan³Onur Mutlu^{5,1}18



An Intelligent Architecture Handles Data Well



How to Handle Data Well

- Ensure data does not overwhelm the components
 - via intelligent algorithms
 - via intelligent architectures
 - via whole system designs: algorithm-architecture-devices

Take advantage of vast amounts of data and metadata
 to improve architectural & system-level decisions

Understand and exploit properties of (different) data
 to improve algorithms & architectures in various metrics

Corollaries: Architectures Today ...

- Architectures are terrible at dealing with data
 - Designed to mainly store and move data vs. to compute
 - □ They are processor-centric as opposed to **data-centric**
- Architectures are terrible at taking advantage of vast amounts of data (and metadata) available to them
 - Designed to make simple decisions, ignoring lots of data
 - □ They make human-driven decisions vs. **data-driven** decisions
- Architectures are terrible at knowing and exploiting different properties of application data
 - Designed to treat all data as the same
 - They make component-aware decisions vs. data-aware

Data-Centric (Memory-Centric) Architectures

Data-Centric Architectures: Properties

Process data where it resides (where it makes sense)

Processing in and near memory structures

Low-latency and low-energy data access

- Low latency memory
- □ Low energy memory

Low-cost data storage and processing

High capacity memory at low cost: hybrid memory, compression

Intelligent data management

Intelligent controllers handling robustness, security, cost

Processing Data Where It Makes Sense

Processing in/near Memory: An Old Idea

Stone, "A Logic-in-Memory Computer," IEEE TC 1970.

A Logic-in-Memory Computer

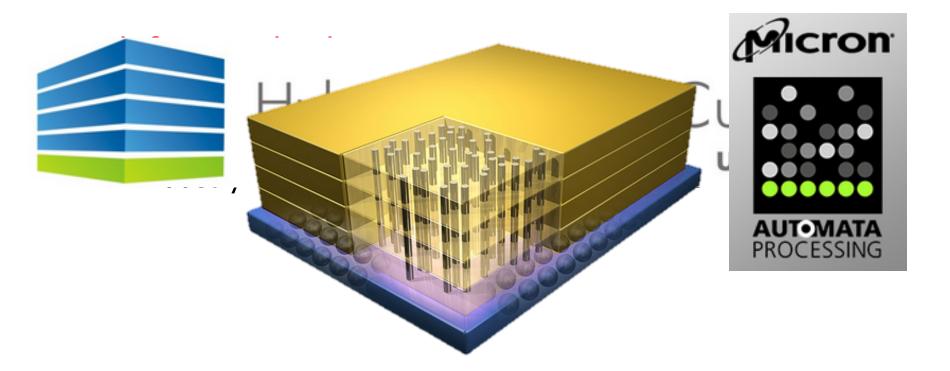
HAROLD S. STONE

Abstract—If, as presently projected, the cost of microelectronic arrays in the future will tend to reflect the number of pins on the array rather than the number of gates, the logic-in-memory array is an extremely attractive computer component. Such an array is essentially a microelectronic memory with some combinational logic associated with each storage element.

Why In-Memory Computation Today?

- Push from Technology
 - DRAM Scaling at jeopardy
 - \rightarrow Controllers close to DRAM
 - \rightarrow Industry open to new memory architectures

Why In-Memory Computation Today?



Memory Scaling Issues Were Real

 Onur Mutlu,
 "Memory Scaling: A Systems Architecture Perspective"
 Proceedings of the <u>5th International Memory</u> Workshop (IMW), Monterey, CA, May 2013. <u>Slides</u> (pptx) (pdf)
 EETimes Reprint

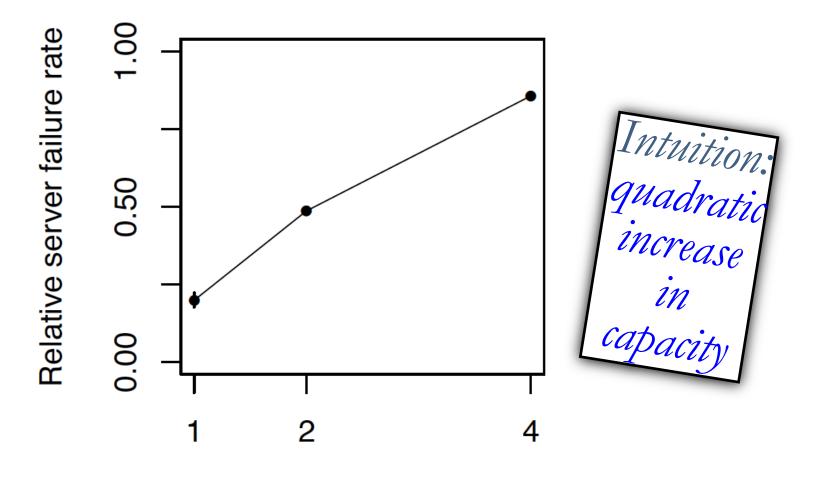
Memory Scaling: A Systems Architecture Perspective

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

https://people.inf.ethz.ch/omutlu/pub/memory-scaling_memcon13.pdf

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.



Chip density (Gb)

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 <u>45th Annual IEEE/IFIP International Conference on</u> 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.

Infrastructures to Understand Such Issues

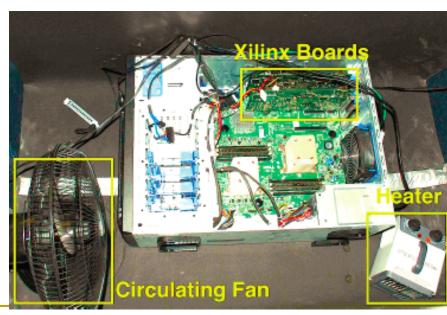


Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors (Kim et al., ISCA 2014)

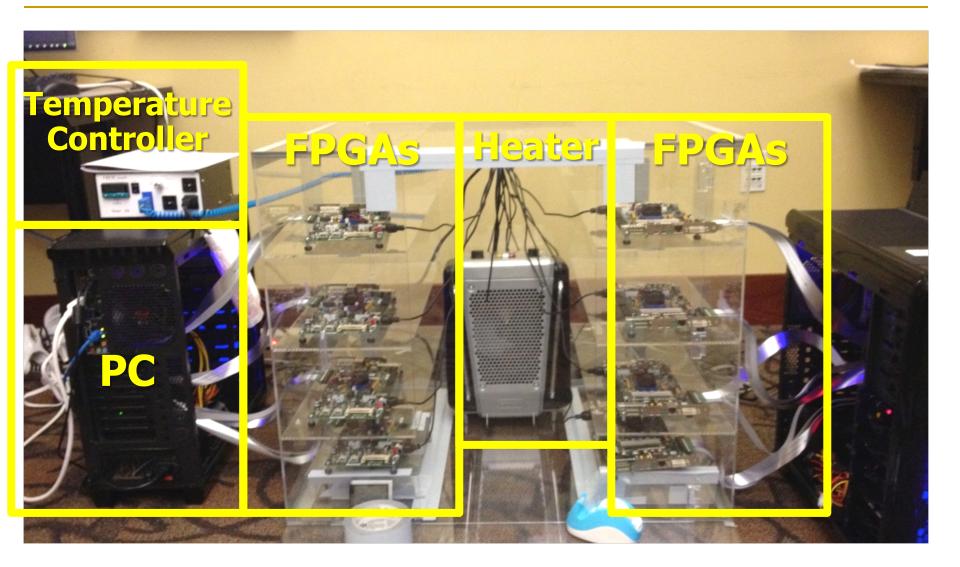
Adaptive-Latency DRAM: Optimizing DRAM Timing for the Common-Case (Lee et al., HPCA 2015)

AVATAR: A Variable-Retention-Time (VRT) Aware Refresh for DRAM Systems (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



SAFARI

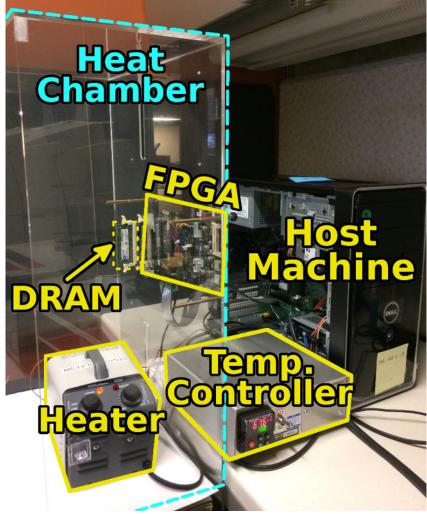
Kim+, "Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors," ISCA 2014.

SoftMC: Open Source DRAM Infrastructure

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

- Flexible
- Easy to Use (C++ API)
- Open-source

github.com/CMU-SAFARI/SoftMC





<u>https://github.com/CMU-SAFARI/SoftMC</u>

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

Hasan Hassan^{1,2,3} Nandita Vijaykumar³ Samira Khan^{4,3} Saugata Ghose³ Kevin Chang³ Gennady Pekhimenko^{5,3} Donghyuk Lee^{6,3} Oguz Ergin² Onur Mutlu^{1,3}

¹ETH Zürich ²TOBB University of Economics & Technology ³Carnegie Mellon University ⁴University of Virginia ⁵Microsoft Research ⁶NVIDIA Research A Curious Discovery [Kim et al., ISCA 2014]

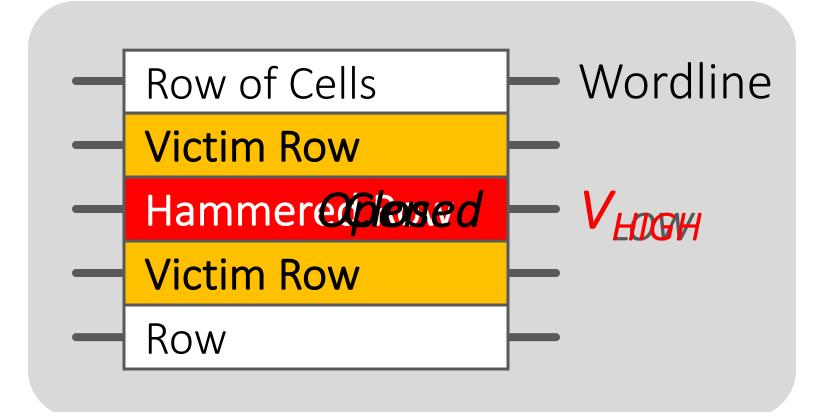
One can predictably induce errors in most DRAM memory chips

The Story of RowHammer

- One can predictably induce bit flips in commodity DRAM chips
 >80% of the tested DRAM chips are vulnerable
- First example of how a simple hardware failure mechanism can create a widespread system security vulnerability



Modern DRAM is Prone to Disturbance Errors

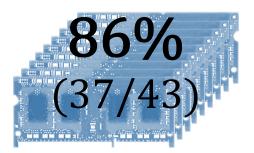


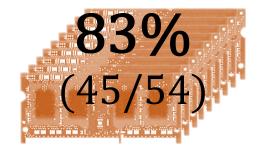
Repeatedly reading a row enough times (before memory gets refreshed) induces disturbance errors in adjacent rows in most real DRAM chips you can buy today

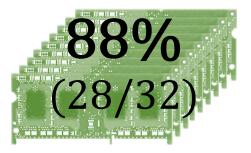
Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors, (Kim et al., ISCA 2014)

Most DRAM Modules Are Vulnerable

A company B company





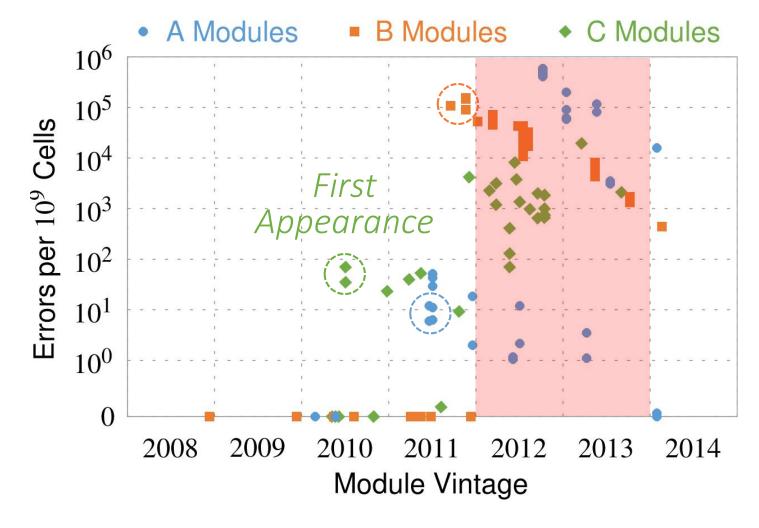


C company

Up to 1.0×10⁷	Up to 2.7×10 ⁶	Up to 3.3×10⁵

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

Recent DRAM Is More Vulnerable



All modules from 2012–2013 are vulnerable

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

<u>Flipping Bits in Memory Without Accessing Them:</u> <u>An Experimental Study of DRAM Disturbance Errors</u> (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

The Push from Circuits and Devices

Main Memory Needs Intelligent Controllers

Memory Scaling Issues Are Real

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 <u>41st International Symposium on Computer</u>
 <u>Architecture</u> (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

Memory Scaling Issues Are Real

Onur Mutlu and Jeremie Kim,
 "RowHammer: A Retrospective"
 IEEE Transactions on Computer-Aided Design of Integrated
 Circuits and Systems (TCAD) Special Issue on Top Picks in
 Hardware and Embedded Security, 2019.
 [Preliminary arXiv version]

RowHammer: A Retrospective

Onur Mutlu^{§‡} Jeremie S. Kim^{‡§} [§]ETH Zürich [‡]Carnegie Mellon University

RowHammer in 2020 (I)

 Jeremie S. Kim, Minesh Patel, A. Giray Yaglikci, Hasan Hassan, Roknoddin Azizi, Lois Orosa, and Onur Mutlu,
 "Revisiting RowHammer: An Experimental Analysis of Modern Devices and Mitigation Techniques"
 Proceedings of the <u>47th International Symposium on Computer</u> <u>Architecture</u> (ISCA), Valencia, Spain, June 2020.
 [Slides (pptx) (pdf)]
 [Lightning Talk Slides (pptx) (pdf)]
 [Talk Video (20 minutes)]
 [Lightning Talk Video (3 minutes)]

Revisiting RowHammer: An Experimental Analysis of Modern DRAM Devices and Mitigation Techniques

Jeremie S. Kim^{§†} Minesh Patel[§] A. Giray Yağlıkçı[§] Hasan Hassan[§] Roknoddin Azizi[§] Lois Orosa[§] Onur Mutlu^{§†} [§]ETH Zürich [†]Carnegie Mellon University

Key Takeaways from 1580 Chips

- Chips of newer DRAM technology nodes are more vulnerable to RowHammer
- There are chips today whose weakest cells fail after only 4800 hammers
- Chips of newer DRAM technology nodes can exhibit RowHammer bit flips 1) in more rows and 2) farther away from the victim row.
- Existing mitigation mechanisms are not effective

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RowHammer in 2020 (II)

 Pietro Frigo, Emanuele Vannacci, Hasan Hassan, Victor van der Veen, Onur Mutlu, Cristiano Giuffrida, Herbert Bos, and Kaveh Razavi, "TRRespass: Exploiting the Many Sides of Target Row Refresh" Proceedings of the <u>41st IEEE Symposium on Security and</u> <u>Privacy</u> (S&P), San Francisco, CA, USA, May 2020.
 [Slides (pptx) (pdf)]
 [Talk Video (17 minutes)]
 [Source Code]
 [Web Article]
 Best paper award.

TRRespass: Exploiting the Many Sides of Target Row Refresh

Pietro Frigo^{*†} Emanuele Vannacci^{*†} Hasan Hassan[§] Victor van der Veen[¶] Onur Mutlu[§] Cristiano Giuffrida^{*} Herbert Bos^{*} Kaveh Razavi^{*}

*Vrije Universiteit Amsterdam

[§]ETH Zürich

[¶]Qualcomm Technologies Inc.

RowHammer in 2020 (III)

 Lucian Cojocar, Jeremie Kim, Minesh Patel, Lillian Tsai, Stefan Saroiu, Alec Wolman, and Onur Mutlu,
 "Are We Susceptible to Rowhammer? An End-to-End Methodology for Cloud Providers"
 Proceedings of the <u>41st IEEE Symposium on Security and</u> Privacy (S&P), San Francisco, CA, USA, May 2020.
 [Slides (pptx) (pdf)]
 [Talk Video (17 minutes)]

Are We Susceptible to Rowhammer? An End-to-End Methodology for Cloud Providers

Lucian Cojocar, Jeremie Kim^{§†}, Minesh Patel[§], Lillian Tsai[‡], Stefan Saroiu, Alec Wolman, and Onur Mutlu^{§†} Microsoft Research, [§]ETH Zürich, [†]CMU, [‡]MIT

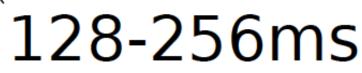
The Push from Circuits and Devices

Main Memory Needs Intelligent Controllers

Data Retention in Memory [Liu et al., ISCA 2013]

Retention Time Profile of DRAM looks like this:

64-128ms >256ms



Location dependent Stored value pattern dependent Time dependent

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More on DRAM Refresh (I)

 Jamie Liu, Ben Jaiyen, Richard Veras, and Onur Mutlu, "RAIDR: Retention-Aware Intelligent DRAM Refresh" Proceedings of the <u>39th International Symposium on</u> <u>Computer Architecture</u> (ISCA), Portland, OR, June 2012. <u>Slides (pdf)</u>

RAIDR: Retention-Aware Intelligent DRAM Refresh

Jamie Liu Ben Jaiyen Richard Veras Onur Mutlu Carnegie Mellon University

More on DRAM Refresh (II)

chris.wilkerson@intel.com

 Jamie Liu, Ben Jaiyen, Yoongu Kim, Chris Wilkerson, and <u>Onur Mutlu</u>,
 "An Experimental Study of Data Retention Behavior in Modern DRAM Devices: Implications for Retention Time Profiling Mechanisms" Proceedings of the <u>40th International Symposium on Computer Architecture</u> (ISCA), Tel-Aviv, Israel, June 2013. <u>Slides (ppt)</u> <u>Slides (pdf)</u>

An Experimental Study of Data Retention Behavior in Modern DRAM Devices: Implications for Retention Time Profiling Mechanisms

Jamie Liu* Ben Jaiyen^{*} Yoongu Kim Carnegie Mellon University Carnegie Mellon University Carnegie Mellon University 5000 Forbes Ave. 5000 Forbes Ave. 5000 Forbes Ave. Pittsburgh, PA 15213 Pittsburgh, PA 15213 Pittsburgh, PA 15213 bjaiyen@alumni.cmu.edu jamiel@alumni.cmu.edu yoonguk@ece.cmu.edu Chris Wilkerson Onur Mutlu Intel Corporation Carnegie Mellon University 2200 Mission College Blvd. 5000 Forbes Ave. Santa Clara, CA 95054 Pittsburgh, PA 15213

onur@cmu.edu

More on DRAM Refresh (III)

Samira Khan, Donghyuk Lee, Yoongu Kim, Alaa Alameldeen, Chris Wilkerson, and Onur Mutlu, "The Efficacy of Error Mitigation Techniques for DRAM Retention Failures: A Comparative Experimental Study" Proceedings of the <u>ACM International Conference on Measurement and</u> Modeling of Computer Systems (SIGMETRICS), Austin, TX, June 2014. [Slides] (pptx) (pdf)] [Poster (pptx) (pdf)] [Full data sets]

The Efficacy of Error Mitigation Techniques for DRAM **Retention Failures: A Comparative Experimental Study**

Samira Khan[†]* samirakhan@cmu.edu

Donghyuk Lee[†] donghyuk1@cmu.edu

Chris Wilkerson*

Yoongu Kim[†] yoongukim@cmu.edu

Alaa R. Alameldeen* alaa.r.alameldeen@intel.com chris.wilkerson@intel.com

Onur Mutlu[†] onur@cmu.edu

*Intel Labs [†]Carnegie Mellon University

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More on DRAM Refresh (IV)

 Moinuddin Qureshi, Dae Hyun Kim, Samira Khan, Prashant Nair, and Onur Mutlu, "AVATAR: A Variable-Retention-Time (VRT) Aware Refresh for DRAM <u>Systems</u>" Proceedings of the <u>45th Annual IEEE/IFIP International Conference on</u> <u>Dependable Systems and Networks</u> (DSN), Rio de Janeiro, Brazil, June 2015. [Slides (pptx) (pdf)]

AVATAR: A Variable-Retention-Time (VRT) Aware Refresh for DRAM Systems

Moinuddin K. Qureshi[†] Dae-Hyun Kim[†] [†]Georgia Institute of Technology {*moin, dhkim, pnair6*}@*ece.gatech.edu* Samira Khan[‡]

Prashant J. Nair[†] Onur Mutlu[‡] [‡]Carnegie Mellon University

{samirakhan, onur}@cmu.edu

More on DRAM Refresh (V)

 Samira Khan, Donghyuk Lee, and Onur Mutlu, "PARBOR: An Efficient System-Level Technique to Detect Data-Dependent Failures in DRAM" Proceedings of the <u>45th Annual IEEE/IFIP International Conference on</u> Dependable Systems and Networks (DSN), Toulouse, France, June 2016. [Slides (pptx) (pdf)]

PARBOR: An Efficient System-Level Technique to Detect Data-Dependent Failures in DRAM

Samira Khan^{*} Donghyuk Lee^{†‡} Onur Mutlu^{*†} *University of Virginia [†]Carnegie Mellon University [‡]Nvidia *ETH Zürich

More on DRAM Refresh (VI)

 Samira Khan, Chris Wilkerson, Zhe Wang, Alaa R. Alameldeen, Donghyuk Lee, and Onur Mutlu,
 "Detecting and Mitigating Data-Dependent DRAM Failures by Exploiting Current Memory Content"
 Proceedings of the <u>50th International Symposium on Microarchitecture</u> (MICRO), Boston, MA, USA, October 2017.
 [Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)] [Poster (pptx) (pdf)]

Detecting and Mitigating Data-Dependent DRAM Failures by Exploiting Current Memory Content

Samira Khan^{*} Chris Wilkerson[†] Zhe Wang[†] Alaa R. Alameldeen[†] Donghyuk Lee[‡] Onur Mutlu^{*} ^{*}University of Virginia [†]Intel Labs [‡]Nvidia Research ^{*}ETH Zürich

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More on DRAM Refresh (VII)

- Minesh Patel, Jeremie S. Kim, and Onur Mutlu,
 "The Reach Profiler (REAPER): Enabling the Mitigation of DRAM Retention Failures via Profiling at Aggressive Conditions"
 Proceedings of the <u>44th International Symposium on Computer</u> Architecture (ISCA), Toronto, Canada, June 2017.
 [Slides (pptx) (pdf)]
 [Lightning Session Slides (pptx) (pdf)]
- First experimental analysis of (mobile) LPDDR4 chips
- Analyzes the complex tradeoff space of retention time profiling
- Idea: enable fast and robust profiling at higher refresh intervals & temperatures

The Reach Profiler (REAPER): Enabling the Mitigation of DRAM Retention Failures via Profiling at Aggressive Conditions

Minesh Patel^{§‡} Jeremie S. Kim^{‡§} Onur Mutlu^{§‡} [§]ETH Zürich [‡]Carnegie Mellon University

More on DRAM Refresh (VIII)

 Minesh Patel, Jeremie S. Kim, Hasan Hassan, and Onur Mutlu, "Understanding and Modeling On-Die Error Correction in Modern DRAM: An Experimental Study Using Real Devices" Proceedings of the <u>49th Annual IEEE/IFIP International Conference on</u> <u>Dependable Systems and Networks</u> (DSN), Portland, OR, USA, June 2019. [Slides (pptx) (pdf)] [Talk Video (26 minutes)] [Full Talk Lecture (29 minutes)] [Source Code for EINSim, the Error Inference Simulator] Best paper award.

Understanding and Modeling On-Die Error Correction in Modern DRAM: An Experimental Study Using Real Devices

Minesh Patel[†] Jeremie S. Kim^{$\ddagger \dagger$} Hasan Hassan[†] Onur Mutlu^{$\dagger \ddagger$}

[†]*ETH Zürich* [‡]*Carnegie Mellon University*

More on DRAM Refresh (IX)

 Minesh Patel, Jeremie S. Kim, Taha Shahroodi, Hasan Hassan, and Onur Mutlu, "Bit-Exact ECC Recovery (BEER): Determining DRAM On-Die ECC Functions by Exploiting DRAM Data Retention Characteristics" Proceedings of the <u>53rd International Symposium on</u> <u>Microarchitecture (MICRO)</u>, Virtual, October 2020. [Slides (pptx) (pdf)] [Lightning Talk Slides (pptx) (pdf)] [Talk Video (15 minutes)]
 [Lightning Talk Video (1.5 minutes)]
 [Lightning Talk Video (1.5 minutes)]

Best paper award.

Bit-Exact ECC Recovery (BEER): Determining DRAM On-Die ECC Functions by Exploiting DRAM Data Retention Characteristics

Minesh Patel[†] Jeremie S. Kim^{‡†} Taha Shahroodi[†] Hasan Hassan[†] Onur Mutlu^{†‡} [†]ETH Zürich [‡]Carnegie Mellon University

The Push from Circuits and Devices

Main Memory Needs Intelligent Controllers

Industry Is Writing Papers About It, Too

DRAM Process Scaling Challenges

Refresh

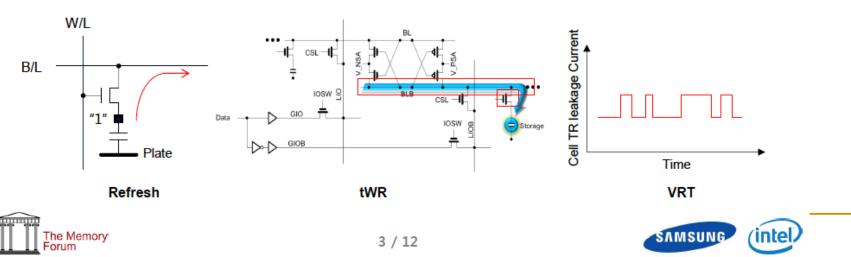
- · Difficult to build high-aspect ratio cell capacitors decreasing cell capacitance
- · Leakage current of cell access transistors increasing

✤ 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

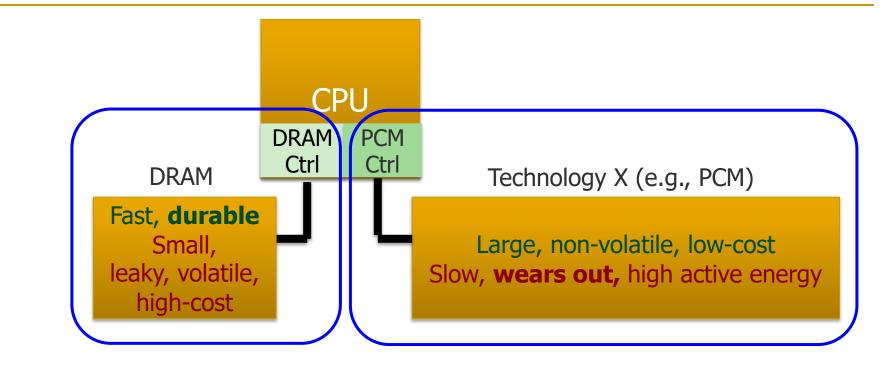
Uksong Kang, Hak-soo Yu, Churoo Park, *Hongzhong Zheng, **John Halbert, **Kuljit Bains, SeongJin Jang, and Joo Sun Choi



61

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

Hybrid Memory Systems



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, Meza et al., "Row Buffer Locality Aware Caching Policies for Hybrid Memories," ICCD 2012 Best Paper Award.

SAFARI

The Push from Circuits and Devices

Main Memory Needs Intelligent Controllers

Why In-Memory Computation Today?

- Push from Technology
 - DRAM Scaling at jeopardy
 - \rightarrow Controllers close to DRAM
 - \rightarrow Industry open to new memory architectures

Pull from Systems and Applications

- Data access is a major system and application bottleneck
- Systems are energy limited
- Data movement much more energy-hungry than computation

SAFARI

1. Data access is a major bottleneck

Applications are increasingly data hungry

2. Energy consumption is a key limiter

3. Data movement energy dominates compute

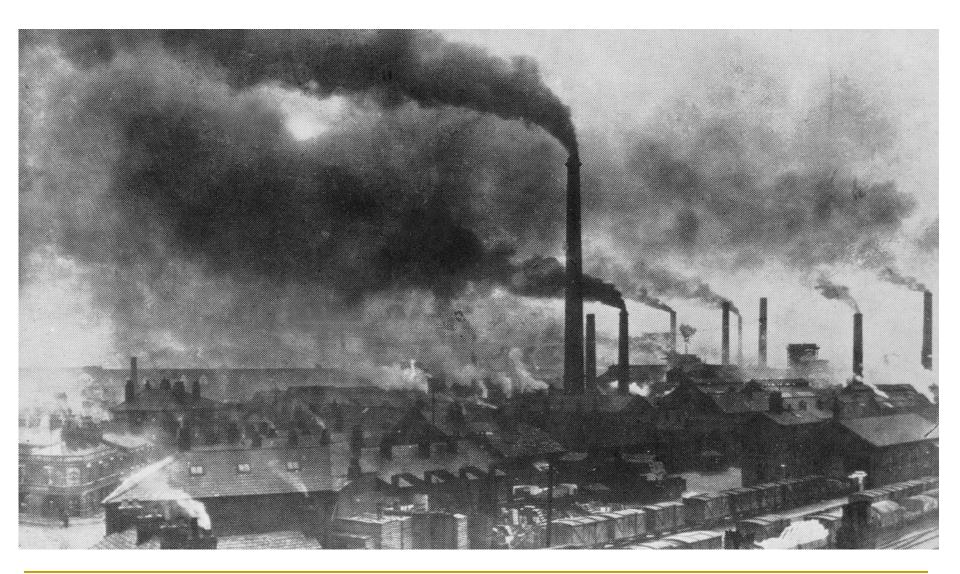
Especially true for off-chip to on-chip movement

Do We Want This?



SAFARI Source

Or This?



Challenge and Opportunity for Future

High Performance, Energy Efficient, Sustainable

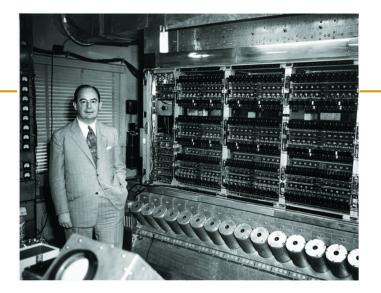
Data access is the major performance and energy bottleneck

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

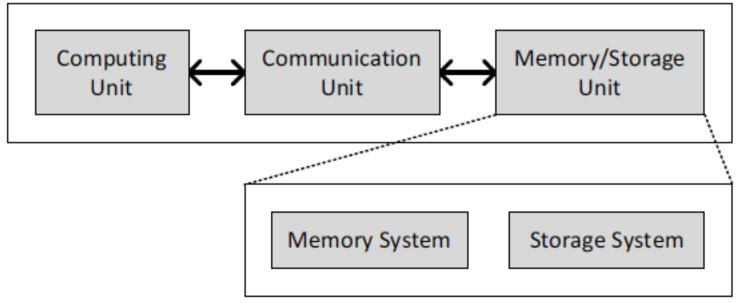
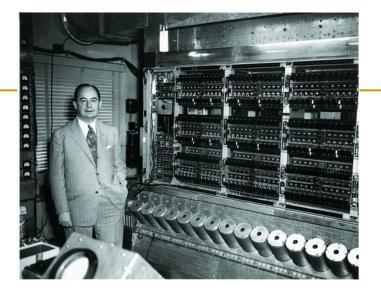


Image source: https://lbsitbytes2010.wordpress.com/2013/03/29/john-von-neumann-roll-no-15/

A Computing System

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Computing System

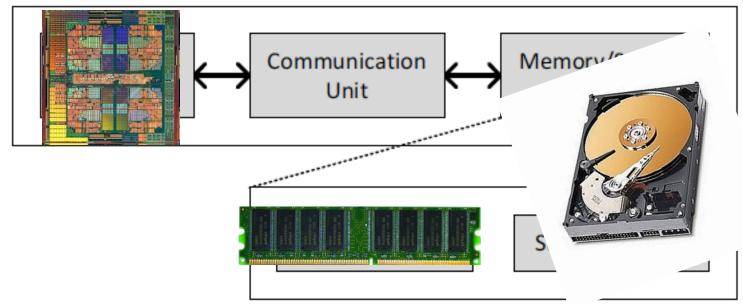
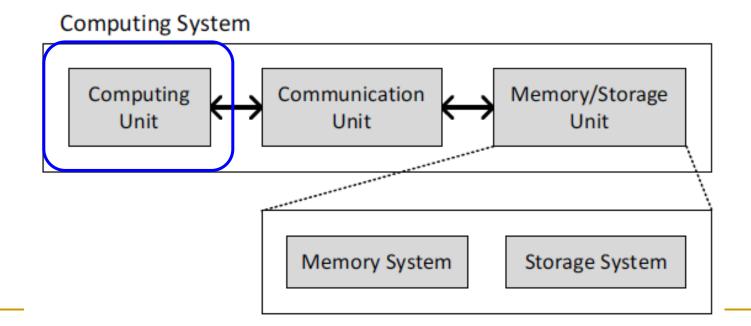


Image source: https://lbsitbytes2010.wordpress.com/2013/03/29/john-von-neumann-roll-no-15/

Today's Computing Systems

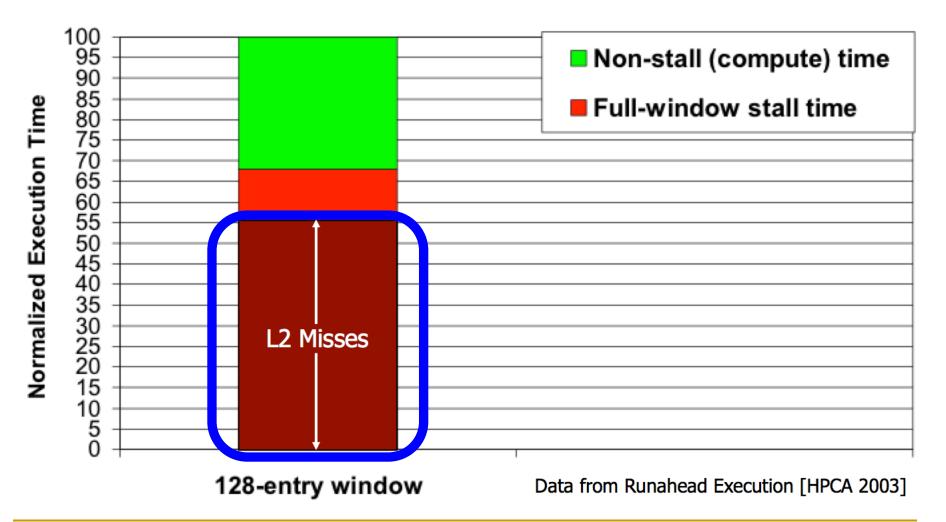
- Are overwhelmingly processor centric
- All data processed in the processor \rightarrow at great system cost
- Processor is heavily optimized and is considered the master
- Data storage units are dumb and are largely unoptimized (except for some that are on the processor die)





I expect that over the coming decade memory subsystem design will be the *only* important design issue for microprocessors.

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



Mutlu+, "Runahead Execution: An Alternative to Very Large Instruction Windows for Out-of-Order Processors," HPCA 2003.

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

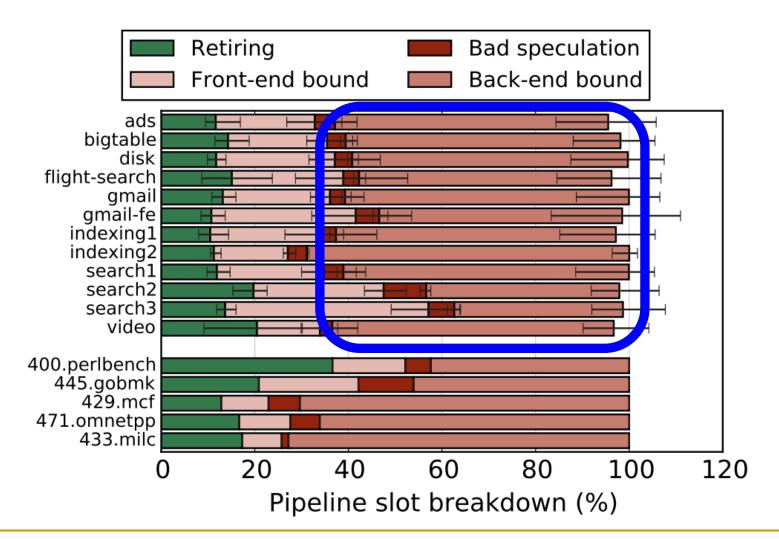
Onur Mutlu § Jared Stark † Chris Wilkerson ‡ Yale N. Patt §

§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):



Kanev+, "Profiling a Warehouse-Scale Computer," ISCA 2015.

The Performance Perspective (Today)

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

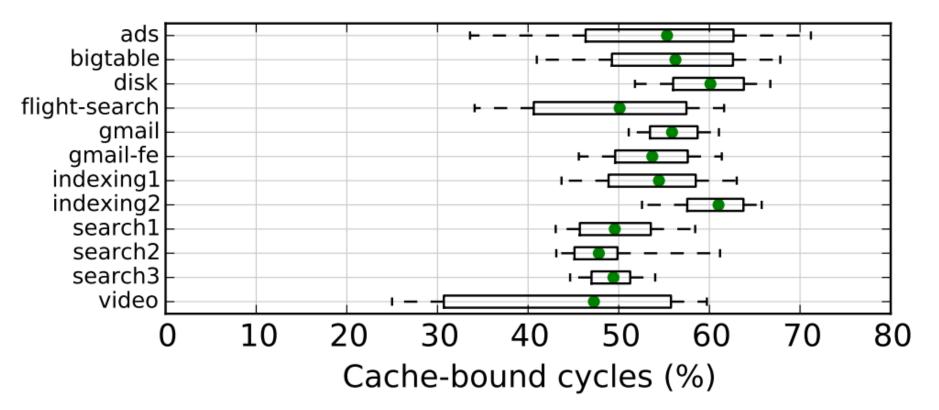


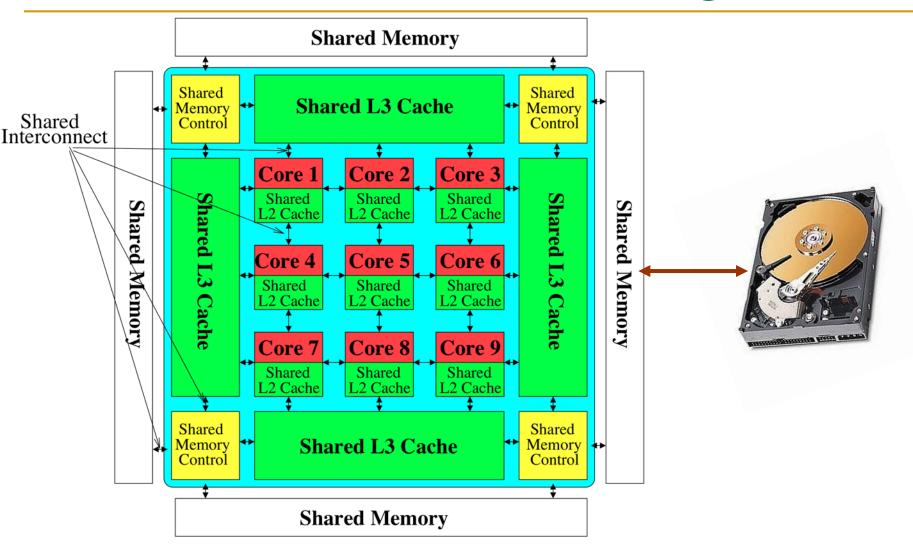
Figure 11: Half of cycles are spent stalled on caches.

Perils of Processor-Centric Design

Grossly-imbalanced systems

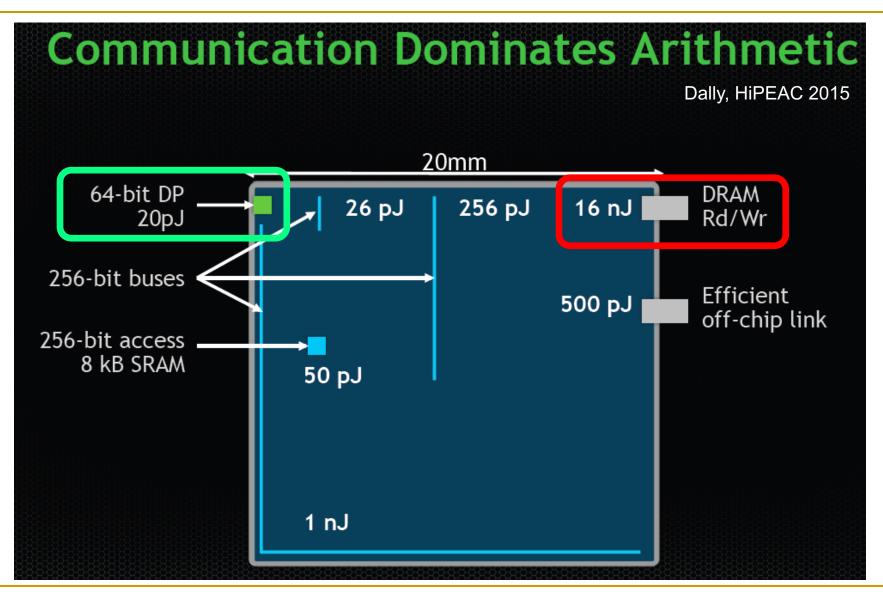
- Processing done only in **one place**
- Everything else just stores and moves data: data moves a lot
- \rightarrow Energy inefficient
- \rightarrow Low performance
- \rightarrow Complex
- Overly complex and bloated processor (and accelerators)
 - To tolerate data access from memory
 - Complex hierarchies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

Perils of Processor-Centric Design

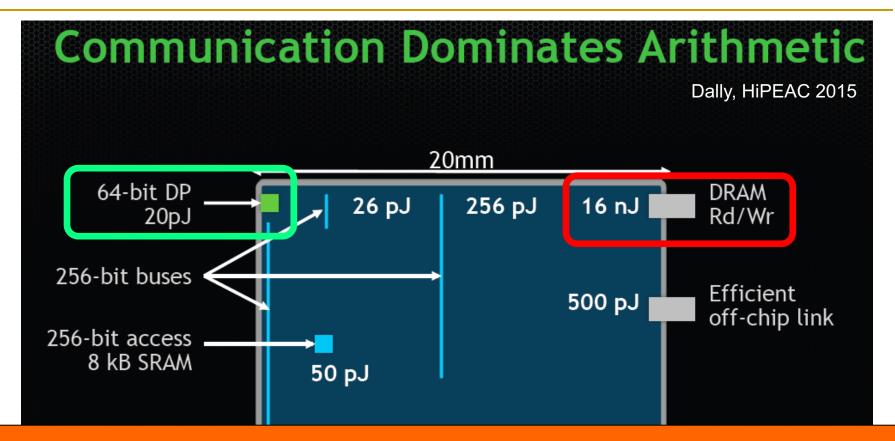


Most of the system is dedicated to storing and moving data

The Energy Perspective



Data Movement vs. Computation Energy

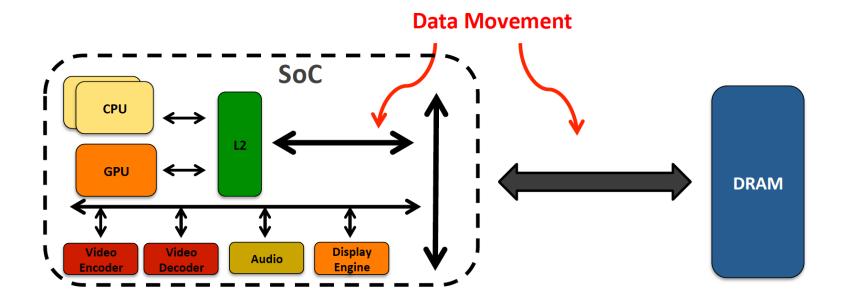


A memory access consumes ~100-1000X the energy of a complex addition

Data Movement vs. Computation Energy

Data movement is a major system energy bottleneck

- Comprises 41% of mobile system energy during web browsing [2]
- Costs ~115 times as much energy as an ADD operation [1, 2]



[1]: Reducing data Movement Energy via Online Data Clustering and Encoding (MICRO'16)

[2]: Quantifying the energy cost of data movement for emerging smart phone workloads on mobile platforms (IISWC'14)

Energy Waste in Mobile Devices

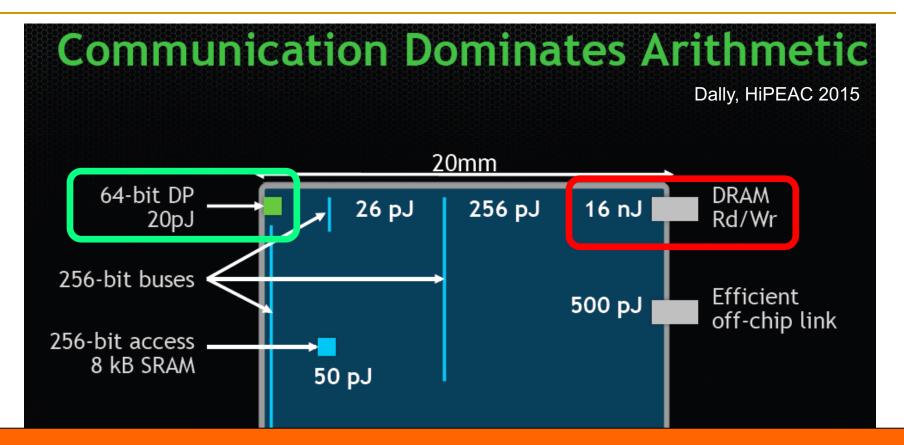
 Amirali Boroumand, Saugata Ghose, Youngsok Kim, Rachata Ausavarungnirun, Eric Shiu, Rahul Thakur, Daehyun Kim, Aki Kuusela, Allan Knies, Parthasarathy Ranganathan, and Onur Mutlu, "Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks" Proceedings of the <u>23rd International Conference on Architectural Support for Programming</u> <u>Languages and Operating Systems</u> (ASPLOS), Williamsburg, VA, USA, March 2018.

62.7% of the total system energy is spent on data movement

Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks

Amirali Boroumand¹Saugata Ghose¹Youngsok Kim²Rachata Ausavarungnirun¹Eric Shiu³Rahul Thakur³Daehyun Kim^{4,3}Aki Kuusela³Allan Knies³Parthasarathy Ranganathan³Onur Mutlu^{5,1}SAFARI83

We Do Not Want to Move Data!



A memory access consumes ~100-1000X the energy of a complex addition

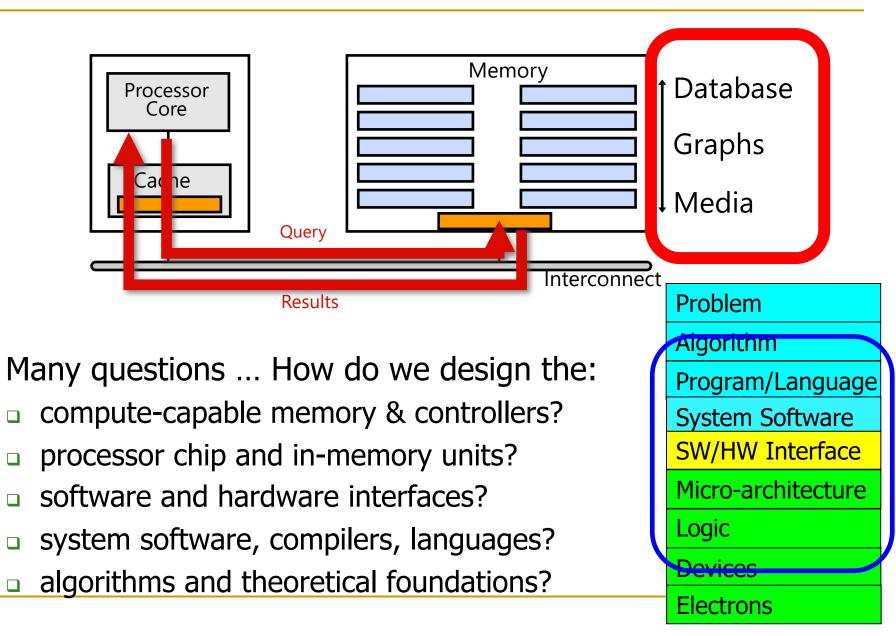
We Need A Paradigm Shift To ...

Enable computation with minimal data movement

Compute where it makes sense (where data resides)

Make computing architectures more data-centric

Goal: Processing Inside Memory



Processing in Memory: Two Approaches

Minimally changing memory chips
 Exploiting 3D-stacked memory

Approach 1: Minimally Changing Memory

- DRAM has great capability to perform bulk data movement and computation internally with small changes
 - Can exploit internal connectivity to move data
 - Can exploit analog computation capability

• Examples: RowClone, In-DRAM AND/OR, Gather/Scatter DRAM

- <u>RowClone: Fast and Efficient In-DRAM Copy and Initialization of Bulk Data</u> (Seshadri et al., MICRO 2013)
- □ Fast Bulk Bitwise AND and OR in DRAM (Seshadri et al., IEEE CAL 2015)
- <u>Gather-Scatter DRAM: In-DRAM Address Translation to Improve the Spatial</u> <u>Locality of Non-unit Strided Accesses</u> (Seshadri et al., MICRO 2015)
- <u>"Ambit: In-Memory Accelerator for Bulk Bitwise Operations Using Commodity</u> <u>DRAM Technology</u>" (Seshadri et al., MICRO 2017)

SAFARI

Starting Simple: Data Copy and Initialization

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





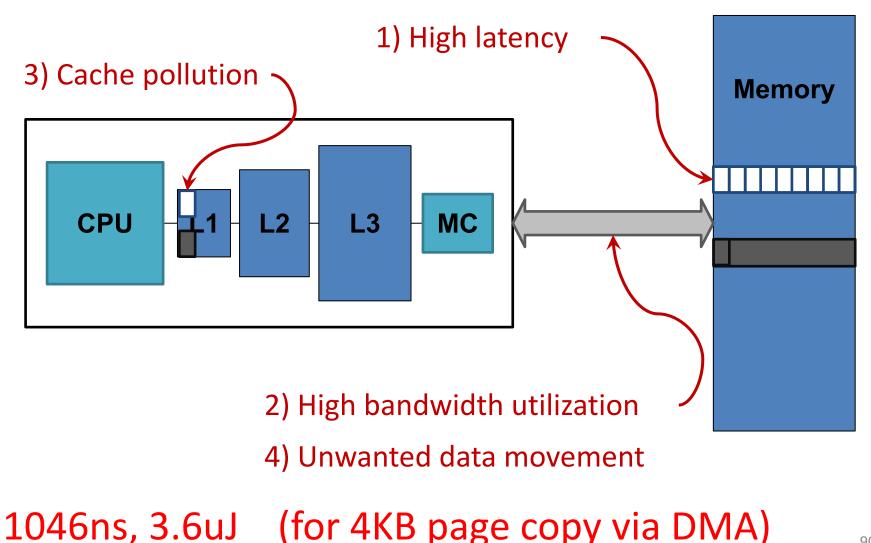
VM Cloning Deduplication



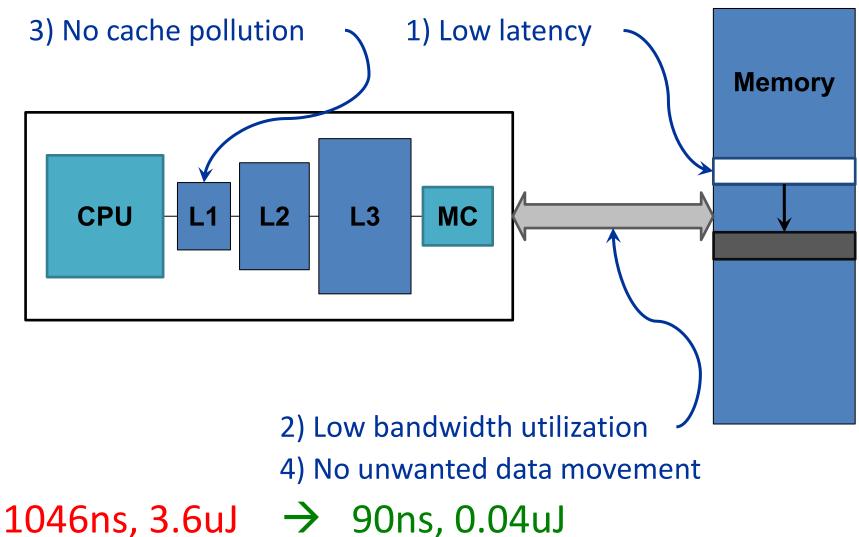
Many more

Page Migration

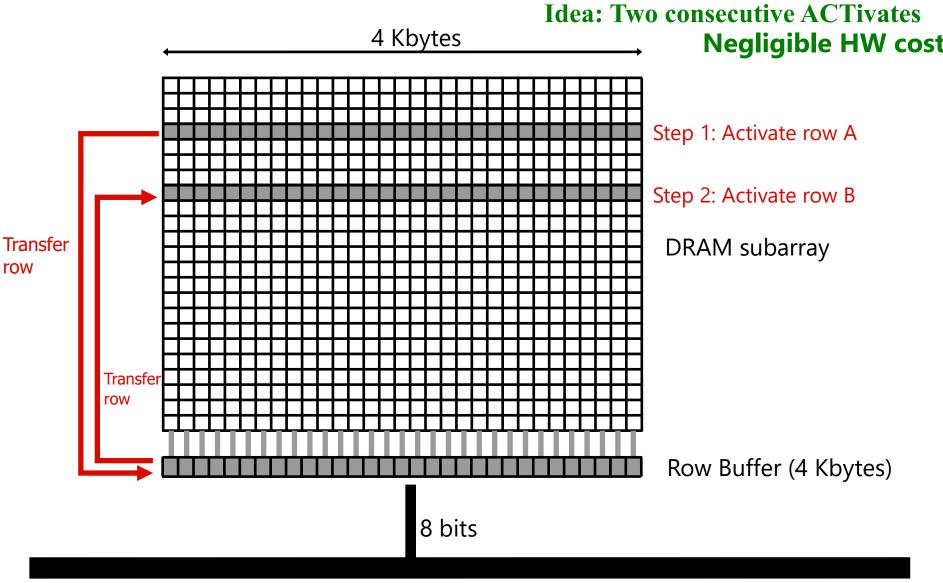
Today's Systems: Bulk Data Copy



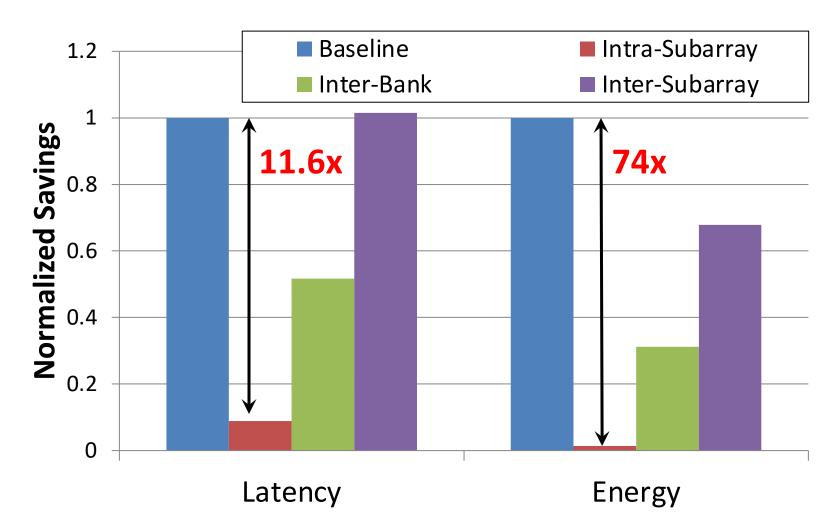
Future Systems: In-Memory Copy



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

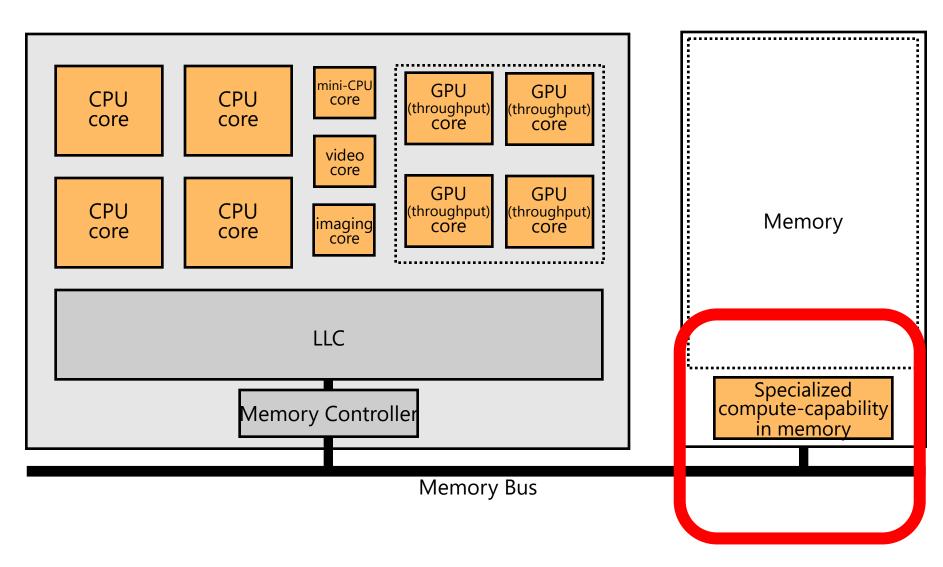
 Vivek Seshadri, Yoongu Kim, Chris Fallin, Donghyuk Lee, Rachata Ausavarungnirun, Gennady Pekhimenko, Yixin Luo, Onur Mutlu, Michael A. Kozuch, Phillip B. Gibbons, and Todd C. Mowry,
 "RowClone: Fast and Energy-Efficient In-DRAM Bulk Data Copy and Initialization" Proceedings of the <u>46th International Symposium on Microarchitecture</u>

(*MICRO*), Davis, CA, December 2013. [<u>Slides (pptx) (pdf)</u>] [<u>Lightning Session</u> <u>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

Memory as an Accelerator



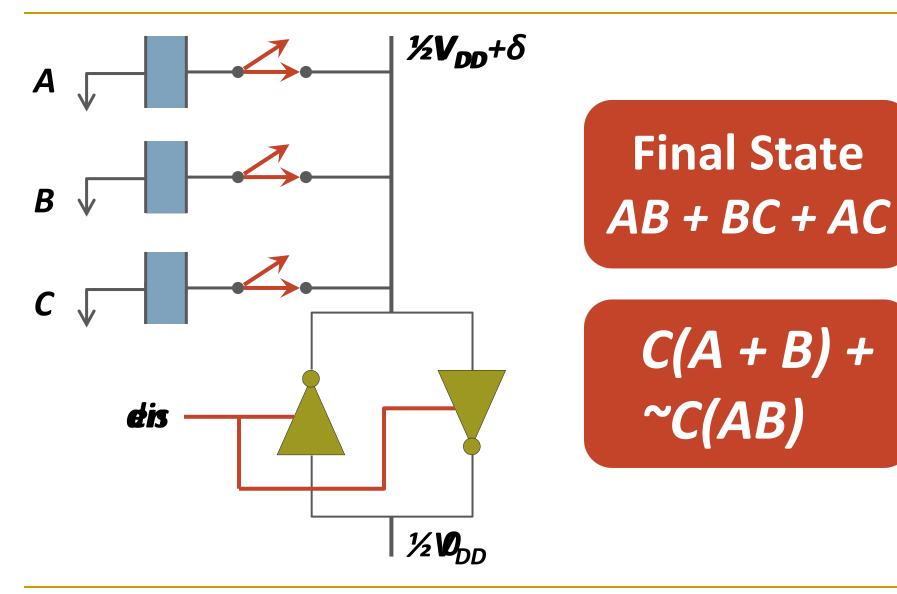
Memory similar to a "conventional" accelerator

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 Bulk Bitwise AND/OR Operation

- BULKAND A, $B \rightarrow C$
- Semantics: Perform a bitwise AND of two rows A and B and store the result in row C
- R0 reserved zero row, R1 reserved one row
- D1, D2, D3 Designated rows for triple activation
- 1. RowClone A into D1
- 2. RowClone B into D2
- 3. RowClone R0 into D3
- 4. ACTIVATE D1,D2,D3
- 5. RowClone Result into C

In-DRAM NOT: Dual Contact Cell

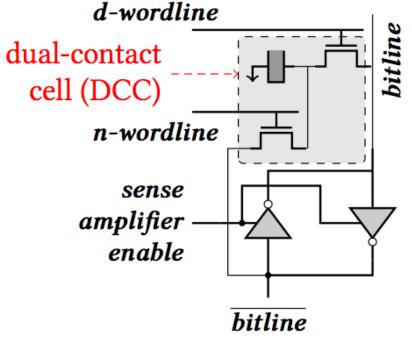


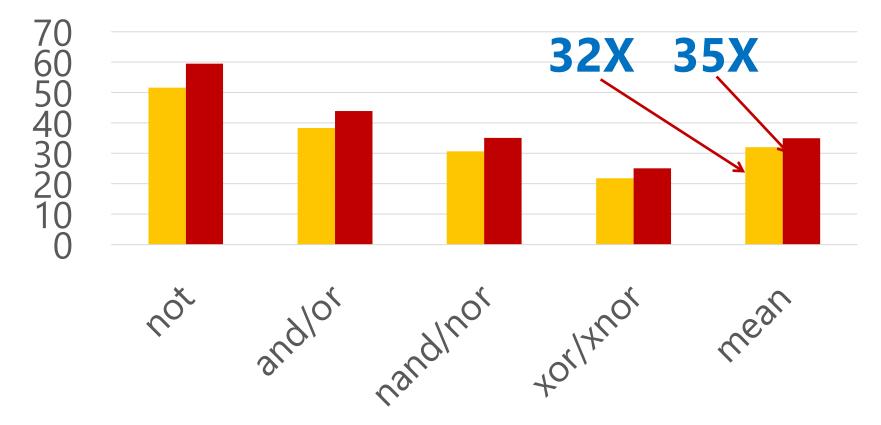
Figure 5: A dual-contact cell connected to both ends of a sense amplifier Idea: Feed the negated value in the sense amplifier into a special row

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

Ambit vs. DDR3: Performance and Energy

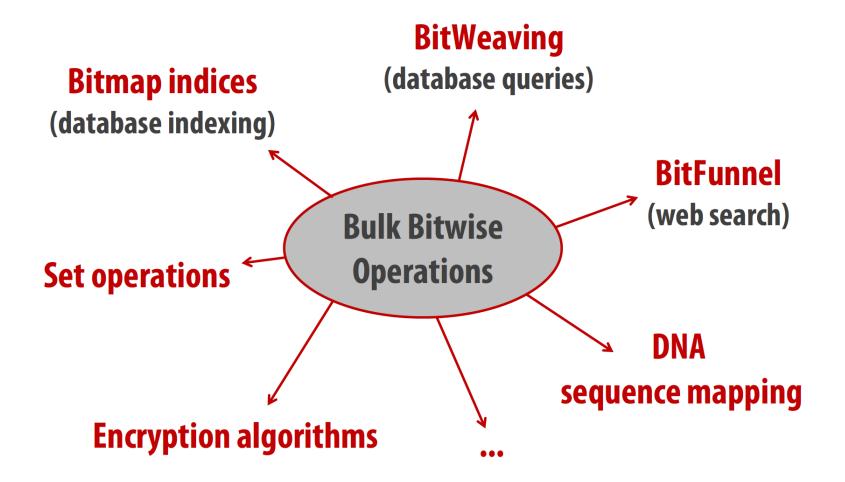
Performance Improvement

Energy Reduction



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

Bulk Bitwise Operations in Workloads



[1] Li and Patel, BitWeaving, SIGMOD 2013[2] Goodwin+, BitFunnel, SIGIR 2017

Performance: Bitmap Index on Ambit

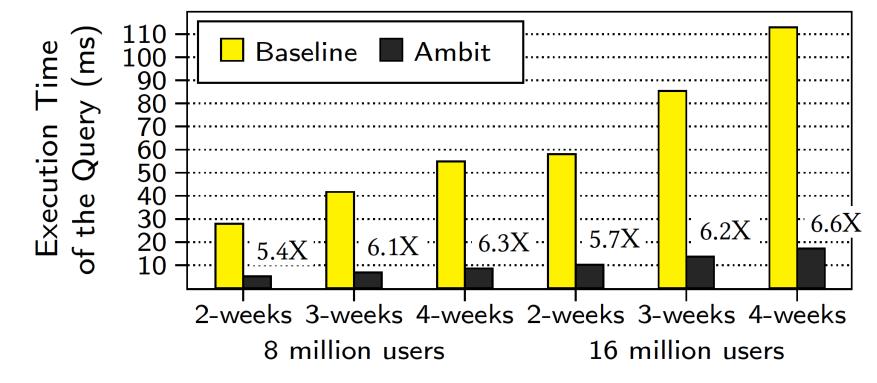


Figure 10: Bitmap index performance. The value above each bar indicates the reduction in execution time due to Ambit.

>5.4-6.6X Performance Improvement

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



Performance: BitWeaving on Ambit

`select count(*) from T where c1 <= val <= c2`</pre>

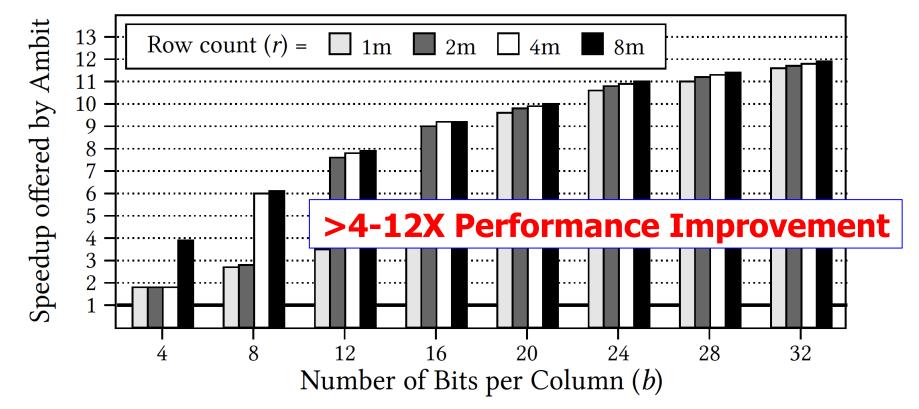


Figure 11: Speedup offered by Ambit over baseline CPU with SIMD for BitWeaving

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

More on In-DRAM Bulk AND/OR

 Vivek Seshadri, Kevin Hsieh, Amirali Boroumand, Donghyuk Lee, Michael A. Kozuch, Onur Mutlu, Phillip B. Gibbons, and Todd C. Mowry,
 <u>"Fast Bulk Bitwise AND and OR in DRAM"</u> <u>IEEE Computer Architecture Letters</u> (CAL), April 2015.

Fast Bulk Bitwise AND and OR in DRAM

Vivek Seshadri*, Kevin Hsieh*, Amirali Boroumand*, Donghyuk Lee*, Michael A. Kozuch[†], Onur Mutlu*, Phillip B. Gibbons[†], Todd C. Mowry* *Carnegie Mellon University [†]Intel Pittsburgh

More on Ambit

 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 Hassan⁴ Amirali Boroumand⁵ Jeremie Kim^{4,5} Michael A. Kozuch³ Onur Mutlu^{4,5} Phillip B. Gibbons⁵ Todd C. Mowry⁵

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

In-DRAM Bulk Bitwise Execution

 Vivek Seshadri and Onur Mutlu,
 <u>"In-DRAM Bulk Bitwise Execution Engine"</u> *Invited Book Chapter in Advances in Computers*, to appear in 2020.
 [Preliminary arXiv version]

In-DRAM Bulk Bitwise Execution Engine

Vivek Seshadri Microsoft Research India visesha@microsoft.com Onur Mutlu ETH Zürich onur.mutlu@inf.ethz.ch

ComputeDRAM: In-Memory Compute Using Off-the-Shelf DRAMs

Fei Gao feig@princeton.edu Department of Electrical Engineering Princeton University Georgios Tziantzioulis georgios.tziantzioulis@princeton.edu Department of Electrical Engineering Princeton University David Wentzlaff wentzlaf@princeton.edu Department of Electrical Engineering Princeton University

SAFARI

https://parallel.princeton.edu/papers/micro19-gao.pdf

Pinatubo: A Processing-in-Memory Architecture for Bulk Bitwise Operations in Emerging Non-volatile Memories

Shuangchen Li¹; Cong Xu², Qiaosha Zou^{1,5}, Jishen Zhao³, Yu Lu⁴, and Yuan Xie¹

University of California, Santa Barbara¹, Hewlett Packard Labs² University of California, Santa Cruz³, Qualcomm Inc.⁴, Huawei Technologies Inc.⁵ {shuangchenli, yuanxie}ece.ucsb.edu¹

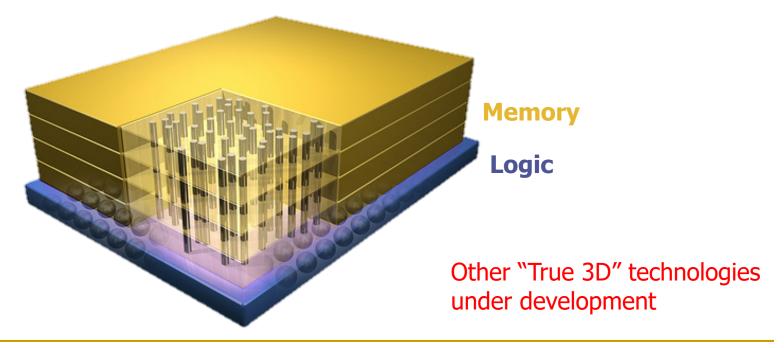
108

Processing in Memory: Two Approaches

Minimally changing memory chips
 Exploiting 3D-stacked memory

Opportunity: 3D-Stacked Logic+Memory





DRAM Landscape (circa 2015)

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]

Table 1. Landscape of DRAM-based memory

Kim+, "Ramulator: A Flexible and Extensible DRAM Simulator", IEEE CAL 2015.

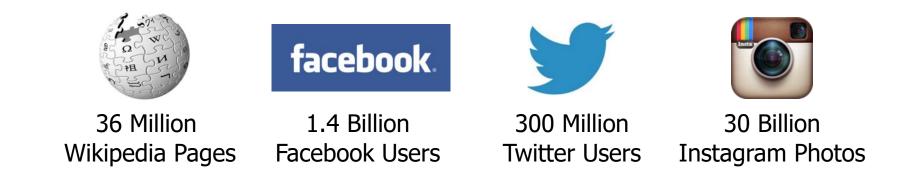
Two Key Questions in 3D-Stacked PIM

- What are the performance and energy benefits of using 3D-stacked memory as a coarse-grained accelerator?
 - By changing the entire system
 - By performing simple function offloading

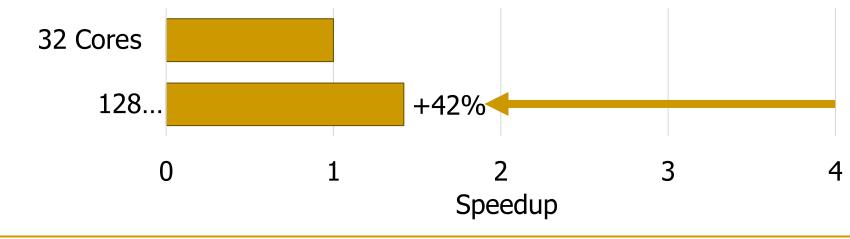
- What is the minimal processing-in-memory support we can provide?
 - With minimal changes to system and programming

Graph Processing

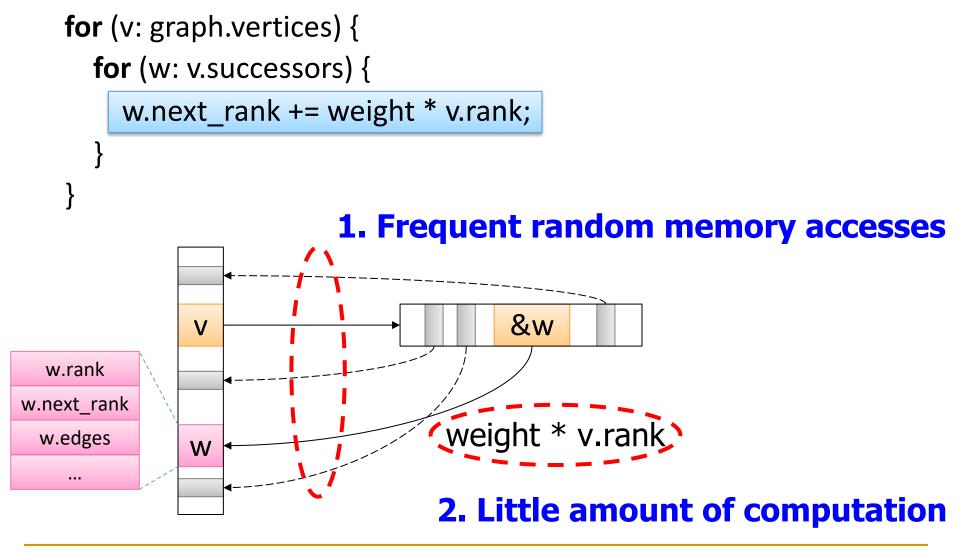
Large graphs are everywhere (circa 2015)



Scalable large-scale graph processing is challenging

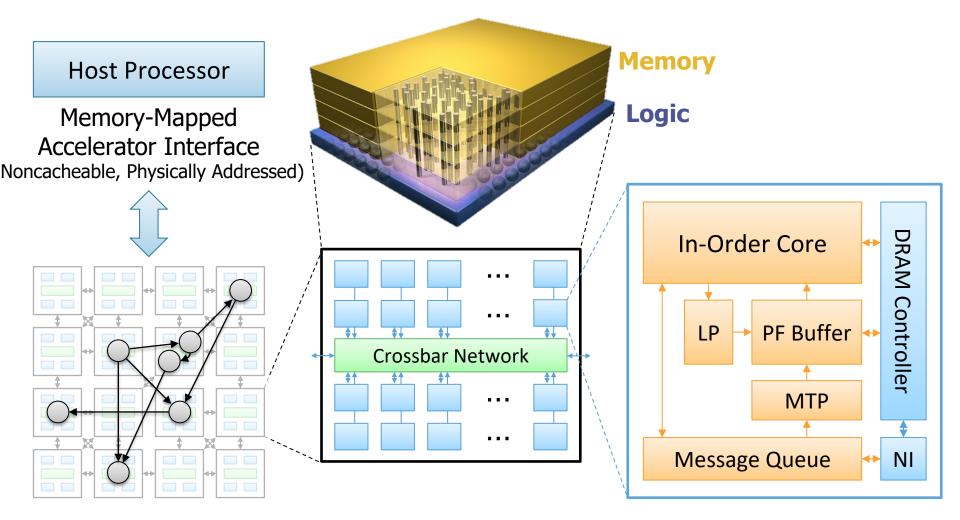


Key Bottlenecks in Graph Processing



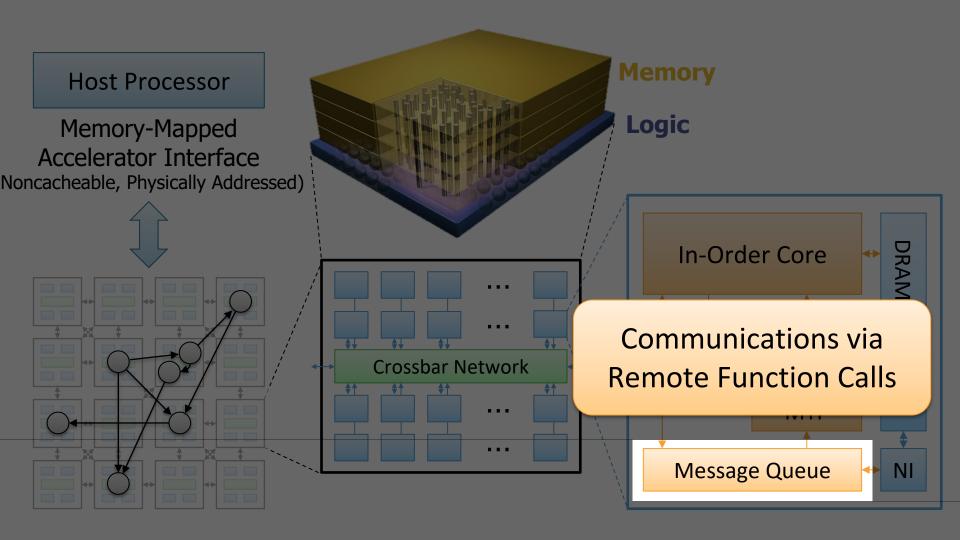
Tesseract System for Graph Processing

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

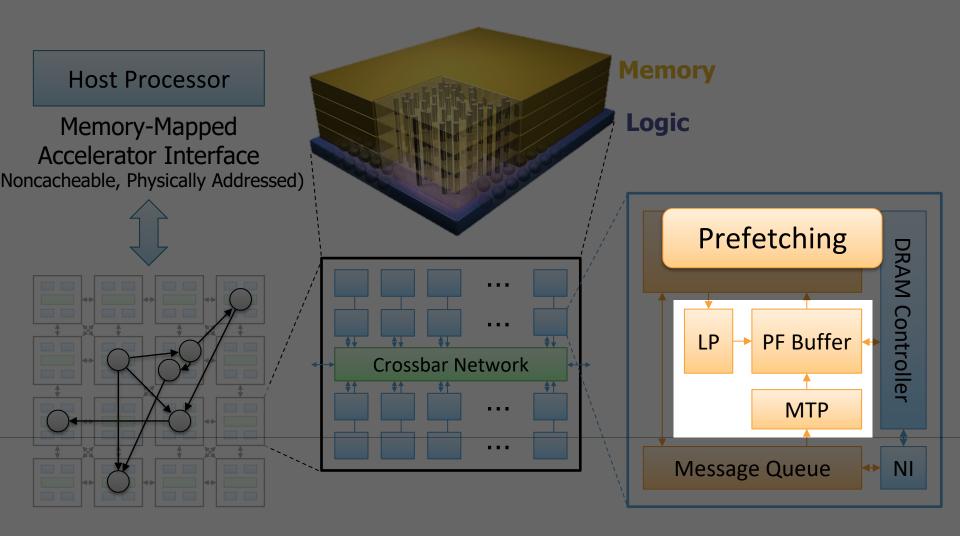


SAFARI Ahn+, "A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing" ISCA 2015.

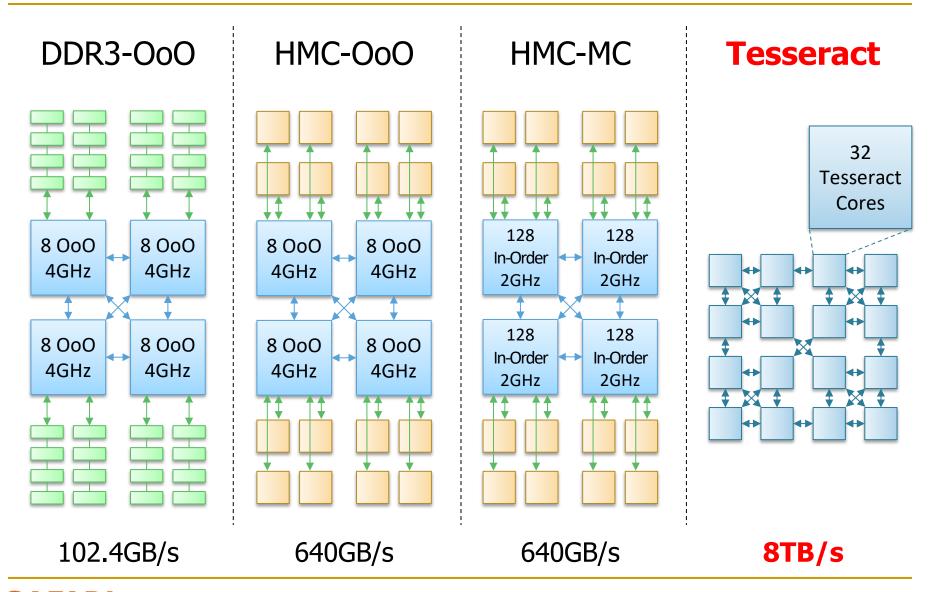
Tesseract System for Graph Processing



Tesseract System for Graph Processing



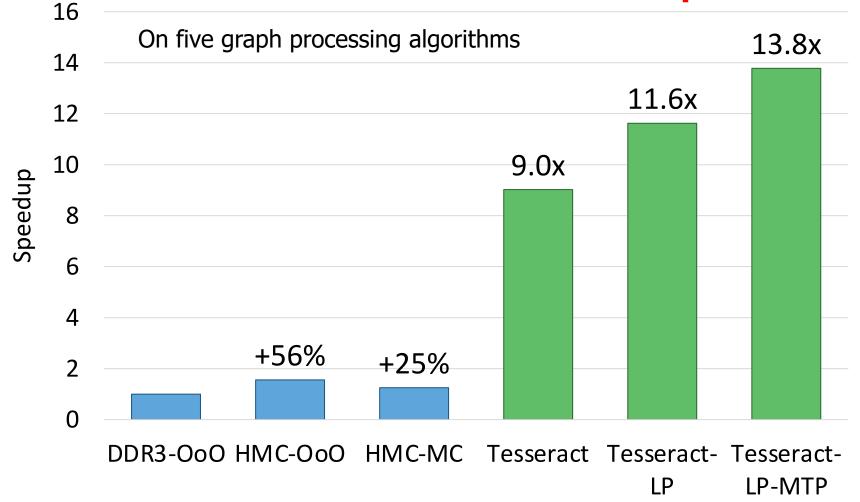
Evaluated Systems



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

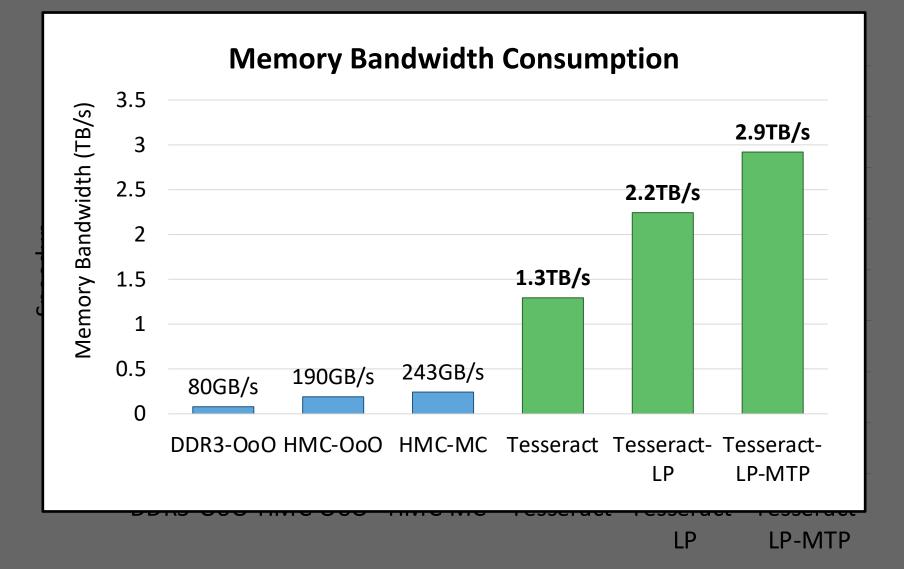
Tesseract Graph Processing Performance

>13X Performance Improvement

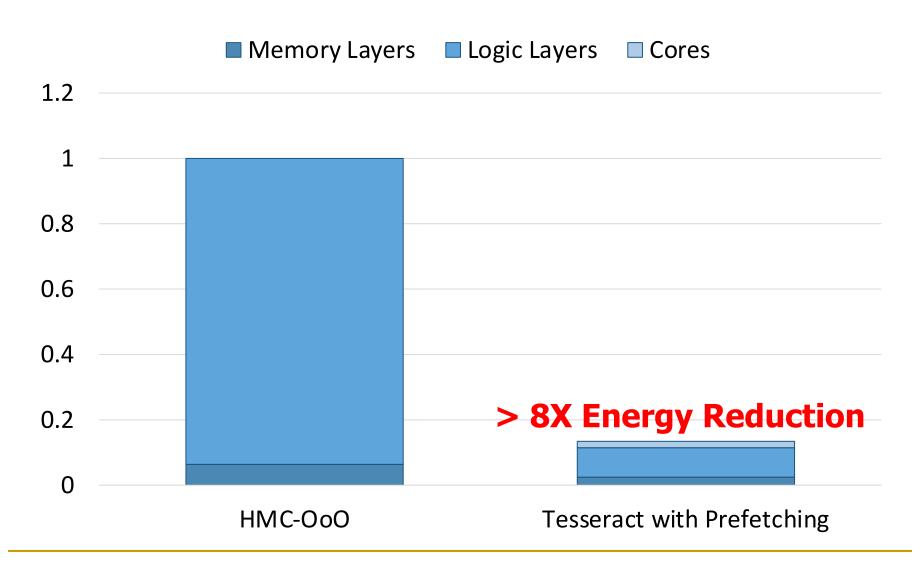


SAFARI Ahn+, "A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing" ISCA 2015.

Tesseract Graph Processing Performance



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

Two Key Questions in 3D-Stacked PIM

- What are the performance and energy benefits of using 3D-stacked memory as a coarse-grained accelerator?
 - By changing the entire system
 - By performing simple function offloading

- What is the minimal processing-in-memory support we can provide?
 - With minimal changes to system and programming

PIM on Mobile Devices

 Amirali Boroumand, Saugata Ghose, Youngsok Kim, Rachata Ausavarungnirun, Eric Shiu, Rahul Thakur, Daehyun Kim, Aki Kuusela, Allan Knies, Parthasarathy Ranganathan, and Onur Mutlu, "Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks"

Proceedings of the <u>23rd International Conference on Architectural</u> <u>Support for Programming Languages and Operating</u> <u>Systems</u> (**ASPLOS**), Williamsburg, VA, USA, March 2018.

Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks

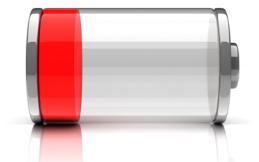
Amirali Boroumand1Saugata Ghose1Youngsok Kim2Rachata Ausavarungnirun1Eric Shiu3Rahul Thakur3Daehyun Kim4,3Aki Kuusela3Allan Knies3Parthasarathy Ranganathan3Onur Mutlu^{5,1}

Consumer Devices



Consumer devices are everywhere!

Energy consumption is a first-class concern in consumer devices



Four Important Workloads





Google's web browser



TensorFlow Mobile

Google's machine learning framework

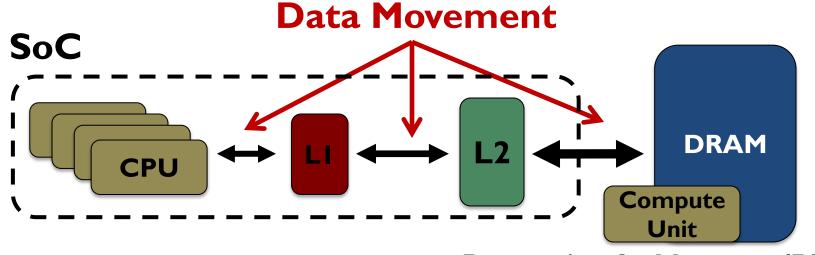


Google's video codec



Energy Cost of Data Movement

Ist key observation: 62.7% of the total system energy is spent on data movement



Processing-In-Memory (PIM)

Potential solution: move computation close to data

Challenge: limited area and energy budget

Simple PIM on Mobile Workloads

2nd key observation: a significant fraction of the data movement often comes from simple functions

We can design lightweight logic to implement these <u>simple functions</u> in <u>memory</u>

Small embedded low-power core



Small fixed-function accelerators



Offloading to PIM logic reduces energy and execution time, on average, by 55.4% and 54.2%

Workload Analysis





Chrome Google's web browser



TensorFlow Mobile

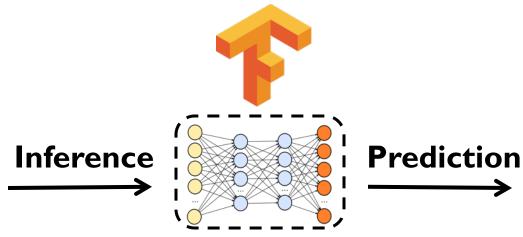
Google's machine learning framework



Google's video codec



TensorFlow Mobile

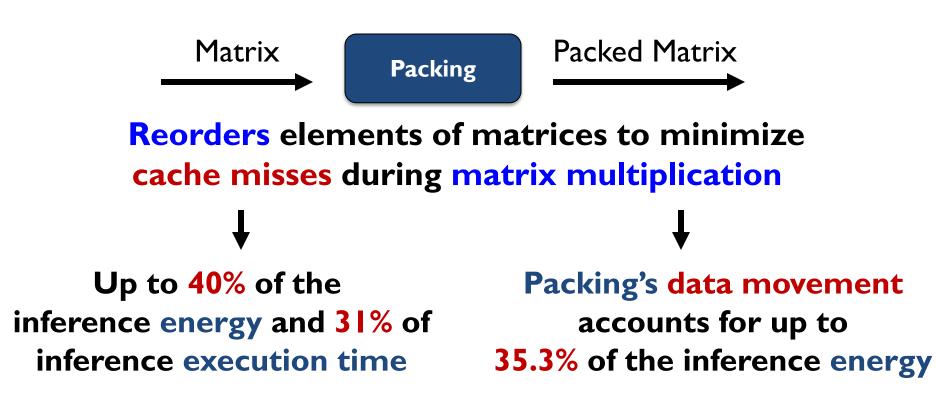


57.3% of the inference energy is spent on data movement

54.4% of the data movement energy comes from packing/unpacking_and quantization

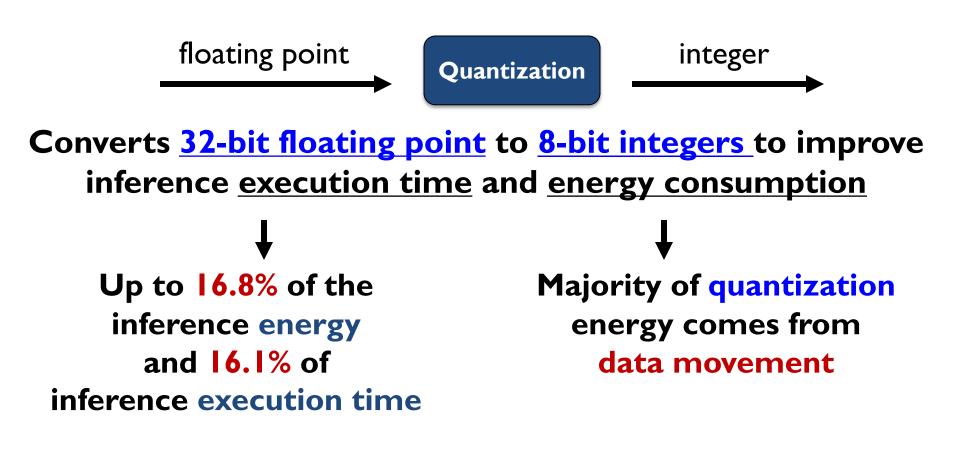


Packing



A simple data reorganization process that requires simple arithmetic

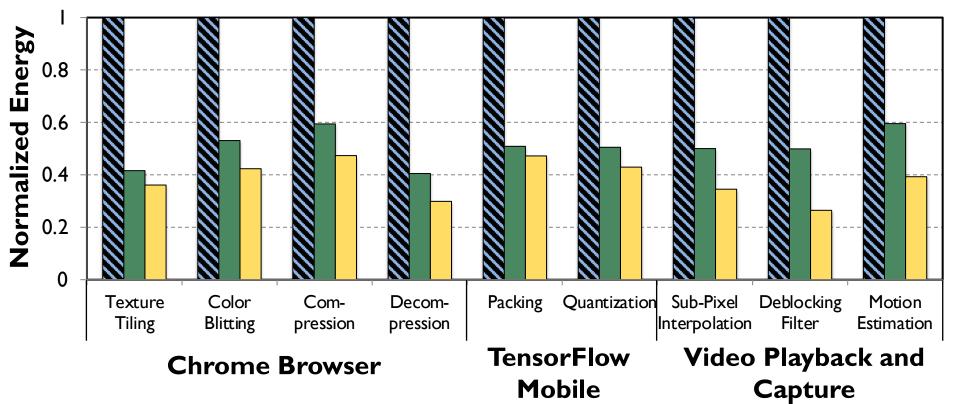
Quantization



A simple data conversion operation that requires shift, addition, and multiplication operations

Normalized Energy

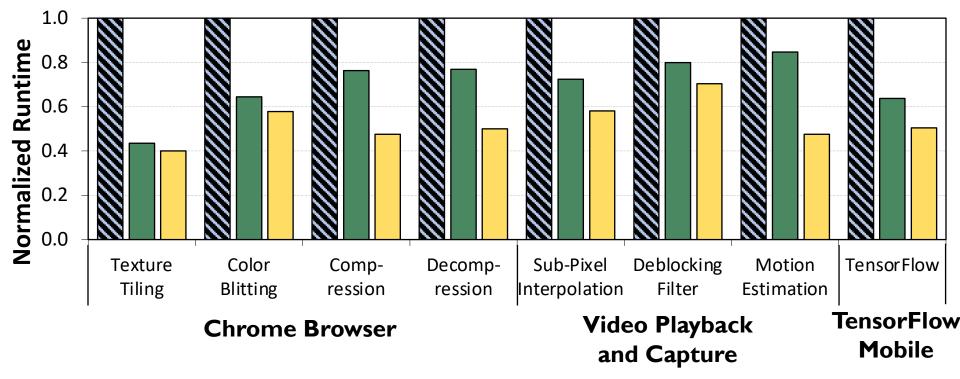




PIM core and PIM accelerator reduce <u>energy consumption</u> on average by 49.1% and 55.4% SAFARI

Normalized Runtime

S CPU-Only ■ PIM-Core ■ PIM-Acc



Offloading these kernels to PIM core and PIM accelerator reduces program runtime on average by 44.6% and 54.2%

More on PIM for Mobile Devices

 Amirali Boroumand, Saugata Ghose, Youngsok Kim, Rachata Ausavarungnirun, Eric Shiu, Rahul Thakur, Daehyun Kim, Aki Kuusela, Allan Knies, Parthasarathy Ranganathan, and Onur Mutlu,

"Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks"

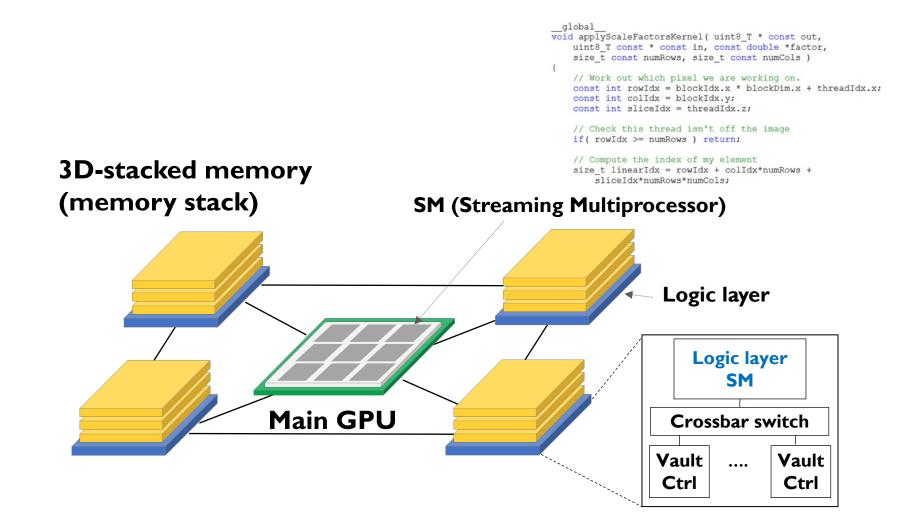
Proceedings of the <u>23rd International Conference on Architectural Support for</u> <u>Programming Languages and Operating Systems</u> (**ASPLOS**), Williamsburg, VA, USA, March 2018. [<u>Slides (pptx) (pdf)</u>] [<u>Lightning Session Slides (pptx) (pdf)</u>] [<u>Poster (pptx) (pdf)</u>] [<u>Lightning Talk Video</u> (2 minutes)]

[Full Talk Video (21 minutes)]

Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks

Amirali Boroumand¹Saugata Ghose¹Youngsok Kim²Rachata Ausavarungnirun¹Eric Shiu³Rahul Thakur³Daehyun Kim^{4,3}Aki Kuusela³Allan Knies³Parthasarathy Ranganathan³Onur Mutlu^{5,1}135

Truly Distributed GPU Processing with PIM?



Accelerating GPU Execution with PIM (I)

 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>

Architecture (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

Accelerating GPU Execution with PIM (II)

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

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

Accelerating Linked Data Structures

 Kevin Hsieh, Samira Khan, Nandita Vijaykumar, Kevin K. Chang, Amirali Boroumand, Saugata Ghose, and Onur Mutlu, <u>"Accelerating Pointer Chasing in 3D-Stacked Memory:</u> <u>Challenges, Mechanisms, Evaluation"</u> *Proceedings of the <u>34th IEEE International Conference on Computer</u> <u>Design</u> (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

Accelerating Dependent Cache Misses

 Milad Hashemi, Khubaib, Eiman Ebrahimi, Onur Mutlu, and Yale N. Patt, "Accelerating Dependent Cache Misses with an Enhanced <u>Memory Controller"</u> *Proceedings of the <u>43rd International Symposium on Computer</u> <i>Architecture (ISCA)*, Seoul, South Korea, June 2016. [Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)]

Accelerating Dependent Cache Misses with an Enhanced Memory Controller

Milad Hashemi^{*}, Khubaib[†], Eiman Ebrahimi[‡], Onur Mutlu[§], Yale N. Patt^{*}

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

Accelerating Runahead Execution

 Milad Hashemi, Onur Mutlu, and Yale N. Patt, "Continuous Runahead: Transparent Hardware Acceleration for <u>Memory Intensive Workloads"</u> *Proceedings of the <u>49th International Symposium on</u> <u>Microarchitecture</u> (<i>MICRO*), Taipei, Taiwan, October 2016. [Slides (pptx) (pdf)] [Lightning Session Slides (pdf)] [Poster (pptx) (pdf)]

Continuous Runahead: Transparent Hardware Acceleration for Memory Intensive Workloads

Milad Hashemi*, Onur Mutlu[§], Yale N. Patt*

* The University of Texas at Austin §ETH Zürich

Accelerating Climate Modeling

 Gagandeep Singh, Dionysios Diamantopoulos, Christoph Hagleitner, Juan Gómez-Luna, Sander Stuijk, Onur Mutlu, and Henk Corporaal, "NERO: A Near High-Bandwidth Memory Stencil Accelerator for Weather Prediction Modeling" Proceedings of the <u>30th International Conference on Field-Programmable Logic</u> and Applications (FPL), Gothenburg, Sweden, September 2020.
 [Slides (pptx) (pdf)]
 [Lightning Talk Slides (pptx) (pdf)]
 [Talk Video (23 minutes)] Nominated for the Stamatis Vassiliadis Memorial Award,

NERO: A Near High-Bandwidth Memory Stencil Accelerator for Weather Prediction Modeling

Gagandeep Singh^{*a,b,c*} Dionysios Diamantopoulos^{*c*} Christoph Hagleitner^{*c*} Juan Gómez-Luna^{*b*} Sander Stuijk^{*a*} Onur Mutlu^{*b*} Henk Corporaal^{*a*} ^{*a*}Eindhoven University of Technology ^{*b*}ETH Zürich ^{*c*}IBM Research Europe, Zurich

Accelerating Approximate String Matching

Damla Senol Cali, Gurpreet S. Kalsi, Zulal Bingol, Can Firtina, Lavanya Subramanian, Jeremie S. Kim, Rachata Ausavarungnirun, Mohammed Alser, Juan Gomez-Luna, Amirali Boroumand, Anant Nori, Allison Scibisz, Sreenivas Subramoney, Can Alkan, Saugata Ghose, and Onur Mutlu, "GenASM: A High-Performance, Low-Power Approximate String Matching Acceleration Framework for Genome Sequence Analysis" *Proceedings of the <u>53rd International Symposium on Microarchitecture</u> (<i>MICRO*), Virtual, October 2020.
 [Lighting Talk Video (1.5 minutes)]
 [Lighting Talk Slides (pptx) (pdf)]
 [Talk Video (18 minutes)]
 [Slides (pptx) (pdf)]

GenASM: A High-Performance, Low-Power Approximate String Matching Acceleration Framework for Genome Sequence Analysis

Damla Senol Cali[†][™] Gurpreet S. Kalsi[™] Zülal Bingöl[▽] Can Firtina[◊] Lavanya Subramanian[‡] Jeremie S. Kim^{◊†} Rachata Ausavarungnirun[⊙] Mohammed Alser[◊] Juan Gomez-Luna[◊] Amirali Boroumand[†] Anant Nori[™] Allison Scibisz[†] Sreenivas Subramoney[™] Can Alkan[▽] Saugata Ghose^{*†} Onur Mutlu^{◊†▽} [†]Carnegie Mellon University [™]Processor Architecture Research Lab, Intel Labs [¬]Bilkent University [◊]ETH Zürich [‡]Facebook [⊙]King Mongkut's University of Technology North Bangkok ^{*}University of Illinois at Urbana–Champaign 143

Accelerating Time Series Analysis

 Ivan Fernandez, Ricardo Quislant, Christina Giannoula, Mohammed Alser, Juan Gómez-Luna, Eladio Gutiérrez, Oscar Plata, and Onur Mutlu,
 <u>"NATSA: A Near-Data Processing Accelerator for Time Series Analysis"</u> *Proceedings of the <u>38th IEEE International Conference on Computer</u> <i>Design (ICCD)*, Virtual, October 2020.

NATSA: A Near-Data Processing Accelerator for Time Series Analysis

Ivan FernandezRicardo QuislantChristina GiannoulaMohammed AlserJuan Gómez-LunaEladio GutiérrezOscar PlataOnur Mutlu§University of Malaga†National Technical University of Athens‡ETH Zürich

Two Key Questions in 3D-Stacked PIM

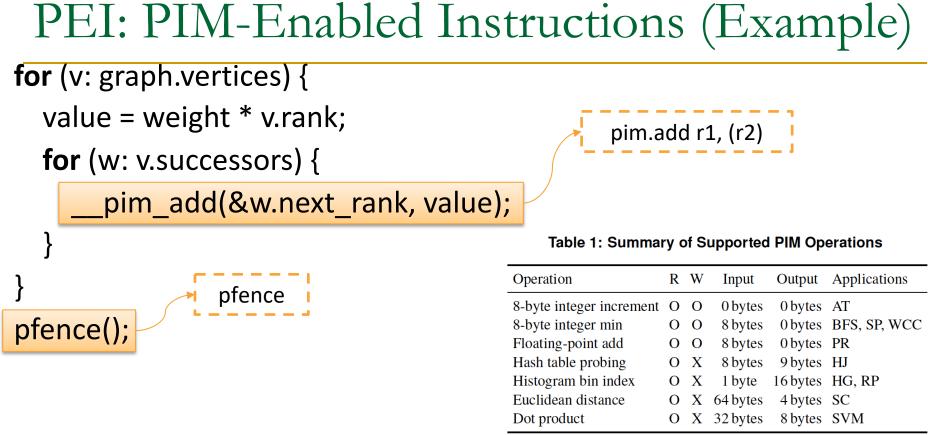
- What are the performance and energy benefits of using 3D-stacked memory as a coarse-grained accelerator?
 - By changing the entire system
 - By performing simple function offloading

What is the minimal processing-in-memory support we can provide?

With minimal changes to system and programming

PEI: PIM-Enabled Instructions (Ideas)

- Goal: Develop mechanisms to get the most out of near-data processing with minimal cost, minimal changes to the system, no changes to the programming model
- Key Idea 1: Expose each PIM operation as a cache-coherent, virtually-addressed host processor instruction (called PEI) that operates on only a single cache block
 - e.g., __pim_add(&w.next_rank, value) \rightarrow pim.add r1, (r2)
 - No changes sequential execution/programming model
 - No changes to virtual memory
 - Minimal changes to cache coherence
 - No need for data mapping: Each PEI restricted to a single memory module
- Key Idea 2: Dynamically decide where to execute a PEI (i.e., the host processor or PIM accelerator) based on simple locality characteristics and simple hardware predictors
 - Execute each operation at the location that provides the best performance



- Executed either in memory or in the processor: dynamic decision
 - Low-cost locality monitoring for a single instruction
- Cache-coherent, virtually-addressed, single cache block only
- Atomic between different PEIs
- Not atomic with normal instructions (use pfence for ordering)
 SAFARI

PEI: Initial Evaluation Results

Initial evaluations with 10 emerging data-intensive workloads

- Large-scale graph processing
- In-memory data analytics
- Machine learning and data mining
- Three input sets (small, medium, large) for each workload to analyze the impact of data locality

Table 2: Baseline Simulation Configuration

Component	Configuration
Core	16 out-of-order cores, 4 GHz, 4-issue
L1 I/D-Cache	Private, 32 KB, 4/8-way, 64 B blocks, 16 MSHRs
L2 Cache	Private, 256 KB, 8-way, 64 B blocks, 16 MSHRs
L3 Cache	Shared, 16 MB, 16-way, 64 B blocks, 64 MSHRs
On-Chip Network	Crossbar, 2 GHz, 144-bit links
Main Memory	32 GB, 8 HMCs, daisy-chain (80 GB/s full-duplex)
HMC	4 GB, 16 vaults, 256 DRAM banks [20]
– DRAM	FR-FCFS, $tCL = tRCD = tRP = 13.75 \text{ ns} [27]$
 Vertical Links 	64 TSVs per vault with 2 Gb/s signaling rate [23]

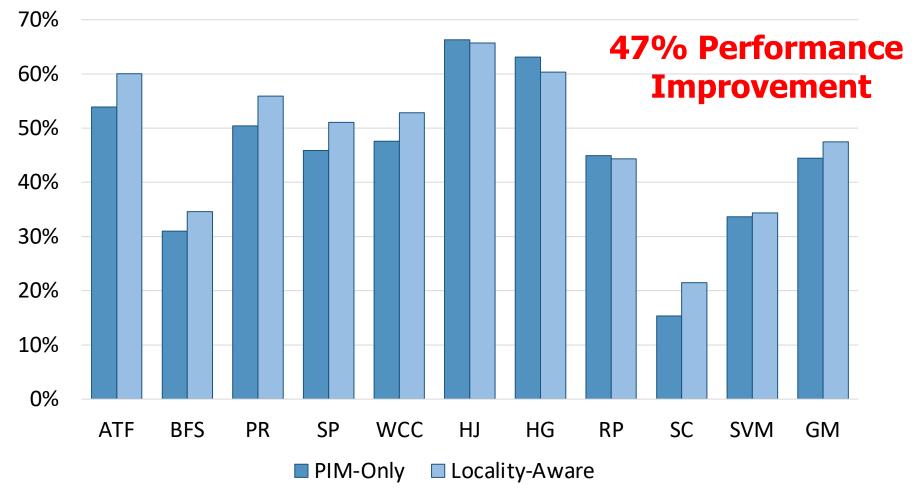
Pin-based cycle-level x86-64 simulation

Performance Improvement and Energy Reduction:

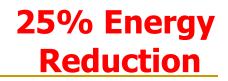
- 47% average speedup with large input data sets
- 32% speedup with small input data sets
- 25% avg. energy reduction in a single node with large input data sets

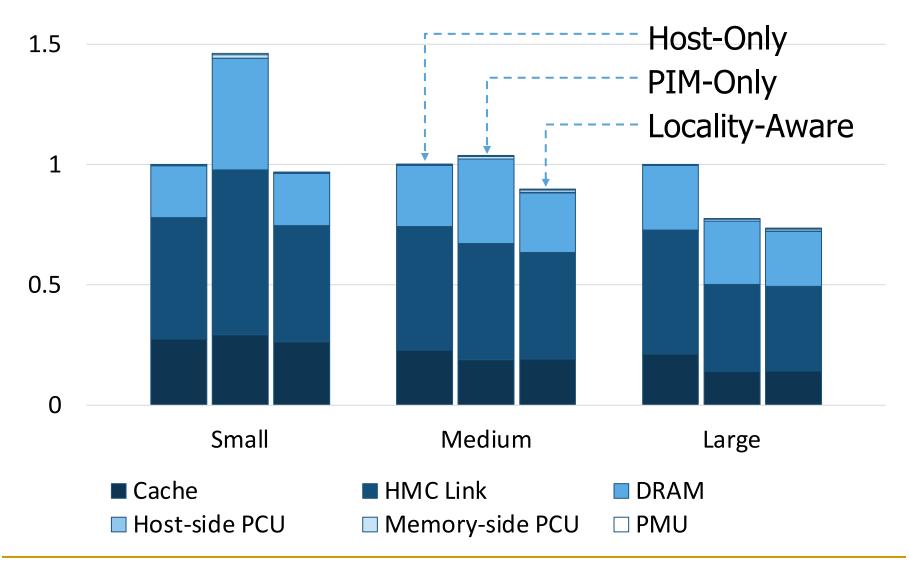
PEI Performance Delta: Large Data Sets





PEI Energy Consumption





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> <u>Computer Architecture</u> (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

Eliminating the Adoption Barriers

How to Enable Adoption of Processing in Memory

Barriers to Adoption of PIM

1. Functionality of and applications & software for PIM

- 2. Ease of programming (interfaces and compiler/HW support)
- 3. System support: coherence & virtual memory

4. Runtime and compilation systems for adaptive scheduling, data mapping, access/sharing control

5. Infrastructures to assess benefits and feasibility

All can be solved with change of mindset

We Need to Revisit the Entire Stack

Problem	,
Aigorithm	
Program/Language	
System Software	
SW/HW Interface	
Micro-architecture	
Logic	
Devices	
Electrons	

We can get there step by step

PIM Review and Open Problems

Processing Data Where It Makes Sense: Enabling In-Memory Computation

Onur Mutlu^{a,b}, Saugata Ghose^b, Juan Gómez-Luna^a, Rachata Ausavarungnirun^{b,c}

^aETH Zürich ^bCarnegie Mellon University ^cKing Mongkut's University of Technology North Bangkok

Onur Mutlu, Saugata Ghose, Juan Gomez-Luna, and Rachata Ausavarungnirun, "Processing Data Where It Makes Sense: Enabling In-Memory Computation" Invited paper in Microprocessors and Microsystems (MICPRO), June 2019. [arXiv version]

https://arxiv.org/pdf/1903.03988.pdf

PIM Review and Open Problems (II)

A Workload and Programming Ease Driven Perspective of Processing-in-Memory

Saugata Ghose†Amirali Boroumand†Jeremie S. Kim†§Juan Gómez-Luna§Onur Mutlu§††Carnegie Mellon University§ETH Zürich

Saugata Ghose, Amirali Boroumand, Jeremie S. Kim, Juan Gomez-Luna, and Onur Mutlu, "Processing-in-Memory: A Workload-Driven Perspective" *Invited Article in IBM Journal of Research & Development, Special Issue on Hardware for Artificial Intelligence*, to appear in November 2019. [Preliminary arXiv version]

SAFARI

https://arxiv.org/pdf/1907.12947.pdf

UPMEM Processing-in-DRAM Engine (2019)

Processing in DRAM Engine

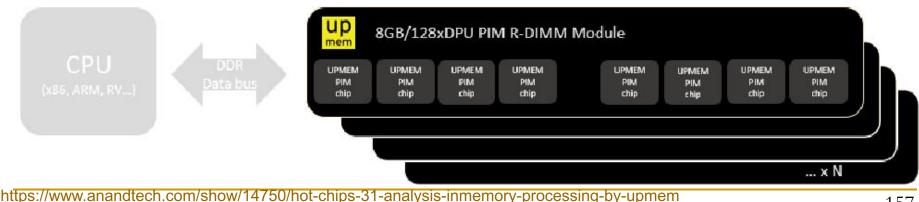
 Includes standard DIMM modules, with a large number of DPU processors combined with DRAM chips.

Replaces standard DIMMs

- DDR4 R-DIMM modules
 - 8GB+128 DPUs (16 PIM chips)
 - Standard 2x-nm DRAM process



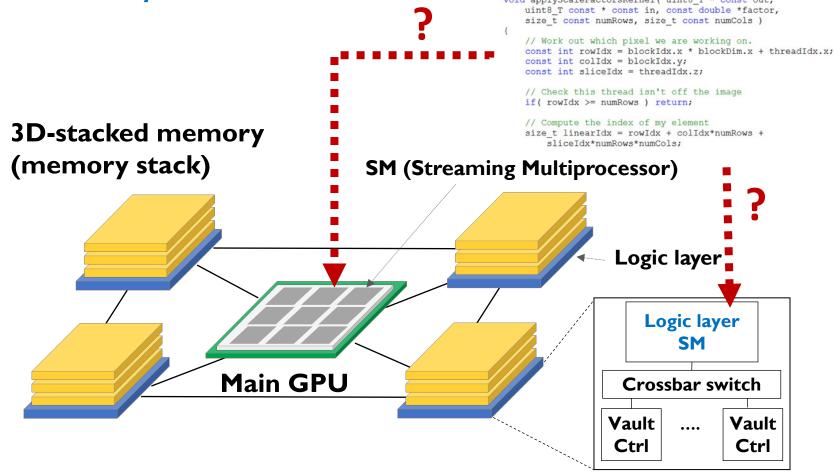
Large amounts of compute & memory bandwidth



https://www.upmem.com/video-upmem-presenting-its-true-processing-in-memory-solution-hot-chips-2019/

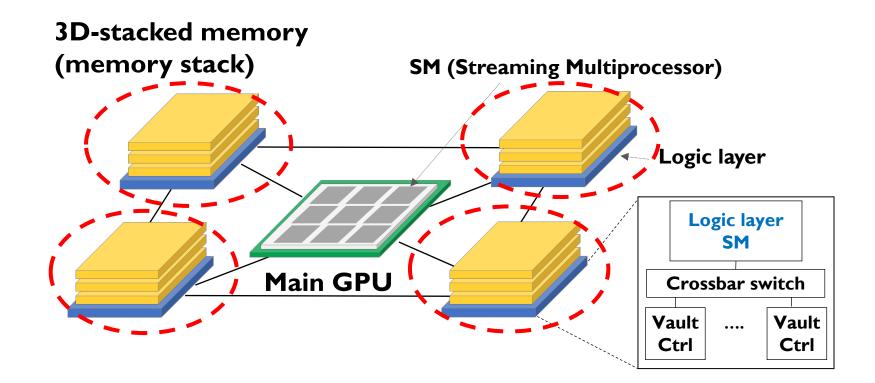
Key Challenge 1: Code Mapping

• Challenge 1: Which operations should be executed in memory vs. in CPU?



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, <u>"Transparent Offloading and Mapping (TOM): Enabling</u> <u>Programmer-Transparent Near-Data Processing in GPU</u> <u>Systems"</u> *Proceedings of the <u>43rd International Symposium on Computer</u>*

Architecture (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? (I)

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

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

How to Schedule Code? (II)

 Milad Hashemi, Khubaib, Eiman Ebrahimi, Onur Mutlu, and Yale N. Patt, "Accelerating Dependent Cache Misses with an Enhanced <u>Memory Controller"</u> *Proceedings of the <u>43rd International Symposium on Computer</u> <i>Architecture (ISCA)*, Seoul, South Korea, June 2016. [Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)]

Accelerating Dependent Cache Misses with an Enhanced Memory Controller

Milad Hashemi^{*}, Khubaib[†], Eiman Ebrahimi[‡], Onur Mutlu[§], Yale N. Patt^{*}

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

How to Schedule Code? (III)

 Milad Hashemi, Onur Mutlu, and Yale N. Patt, "Continuous Runahead: Transparent Hardware Acceleration for <u>Memory Intensive Workloads"</u> *Proceedings of the <u>49th International Symposium on</u> <u>Microarchitecture</u> (<i>MICRO*), Taipei, Taiwan, October 2016. [Slides (pptx) (pdf)] [Lightning Session Slides (pdf)] [Poster (pptx) (pdf)]

Continuous Runahead: Transparent Hardware Acceleration for Memory Intensive Workloads

Milad Hashemi*, Onur Mutlu[§], Yale N. Patt*

* The University of Texas at Austin §ETH Zürich

How to Maintain Coherence? (I)

 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"

<u>IEEE Computer Architecture Letters</u> (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 Maintain Coherence? (II)

Amirali Boroumand, Saugata Ghose, Minesh Patel, Hasan Hassan, Brandon Lucia, Kevin Hsieh, Krishna T. Malladi, Hongzhong Zheng, and Onur Mutlu, "CoNDA: Efficient Cache Coherence Support for Near-**Data Accelerators**"

Proceedings of the <u>46th International Symposium on Computer</u> Architecture (ISCA), Phoenix, AZ, USA, June 2019.

CoNDA: Efficient Cache Coherence Support for Near-Data Accelerators

Saugata Ghose[†] Minesh Patel^{*} Hasan Hassan^{*} Amirali Boroumand[†] Brandon Lucia[†] Rachata Ausavarungnirun^{†‡} Kevin Hsieh[†] Nastaran Hajinazar^{\phi†} Krishna T. Malladi[§] Hongzhong Zheng[§] Onur Mutlu^{*†} [†]Carnegie Mellon University *****ETH Zürich [‡]KMUTNB

*Simon Fraser University *Samsung Semiconductor, Inc.

How to Support Virtual Memory?

 Kevin Hsieh, Samira Khan, Nandita Vijaykumar, Kevin K. Chang, Amirali Boroumand, Saugata Ghose, and Onur Mutlu, <u>"Accelerating Pointer Chasing in 3D-Stacked Memory:</u> <u>Challenges, Mechanisms, Evaluation"</u> *Proceedings of the <u>34th IEEE International Conference on Computer</u> <u>Design</u> (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> <u>and Architectures</u> (SPAA), Washington, DC, USA, July 2017. [Slides (pptx) (pdf)]

Concurrent Data Structures for Near-Memory Computing

Zhiyu Liu Computer Science Department Brown University zhiyu_liu@brown.edu

Maurice Herlihy Computer Science Department Brown University mph@cs.brown.edu Irina Calciu VMware Research Group icalciu@vmware.com

Onur Mutlu Computer Science Department ETH Zürich onur.mutlu@inf.ethz.ch

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.
 - <u>https://github.com/CMU-SAFARI/ramulator-pim</u>
 - <u>https://github.com/CMU-SAFARI/ramulator</u>
 - Source Code for Ramulator-PIM

Ramulator: A Fast and Extensible DRAM Simulator

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

Performance & Energy Models for PIM

- Gagandeep Singh, Juan Gomez-Luna, Giovanni Mariani, Geraldo F. Oliveira, Stefano Corda, Sander Stujik, <u>Onur Mutlu</u>, and Henk Corporaal, <u>"NAPEL: Near-Memory Computing Application Performance</u> <u>Prediction via Ensemble Learning"</u> *Proceedings of the <u>56th Design Automation Conference</u> (<i>DAC*), Las Vegas, NV, USA, June 2019.
 [Slides (pptx) (pdf)]
 [Poster (pptx) (pdf)]
 - Source Code for Ramulator-PIM

NAPEL: Near-Memory Computing Application Performance Prediction via Ensemble Learning

Gagandeep Singh^{*a,c*} Juan Gómez-Luna^{*b*} Stefano Corda^{*a,c*} Sander Stuijk^{*a*} ^{*a*}Eindhoven University of Technology ^{*b*}ET

Giovanni MarianiGeraldo F. OliveiraOnur MutluHenk Corporaal b ETH Zürich c IBM Research - Zurich

Challenge and Opportunity for Future

Computing Architectures with

Minimal Data Movement



Corollaries: Architectures Today ...

- Architectures are terrible at dealing with data
 - Designed to mainly store and move data vs. to compute
 - They are processor-centric as opposed to data-centric
- Architectures are terrible at taking advantage of vast amounts of data (and metadata) available to them
 - Designed to make simple decisions, ignoring lots of data
 - □ They make human-driven decisions vs. **data-driven** decisions
- Architectures are terrible at knowing and exploiting different properties of application data
 - Designed to treat all data as the same
 - They make component-aware decisions vs. data-aware

Exploiting Data to Design Intelligent Architectures

System Architecture Design Today

- Human-driven
 - Humans design the policies (how to do things)
- Many (too) simple, short-sighted policies all over the system
- No automatic data-driven policy learning
- (Almost) no learning: cannot take lessons from past actions

Can we design fundamentally intelligent architectures?

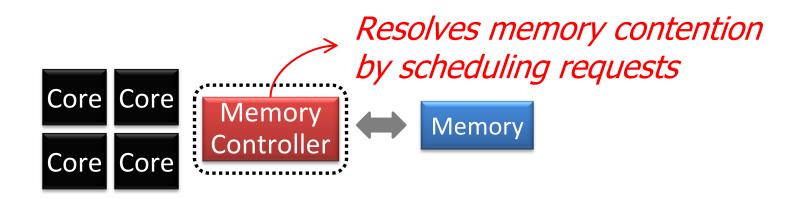
An Intelligent Architecture

- Data-driven
 - Machine learns the "best" policies (how to do things)
- Sophisticated, workload-driven, changing, far-sighted policies
- Automatic data-driven policy learning
- All controllers are intelligent data-driven agents

How do we start?

Self-Optimizing Memory Controllers

Memory Controller



How to schedule requests to maximize system performance?

Why are Memory Controllers Difficult to Design?

Need to obey DRAM timing constraints for correctness

- □ There are many (50+) timing constraints in DRAM
- tWTR: Minimum number of cycles to wait before issuing a read command after a write command is issued
- tRC: Minimum number of cycles between the issuing of two consecutive activate commands to the same bank

• ...

- Need to keep track of many resources to prevent conflicts
 - Channels, banks, ranks, data bus, address bus, row buffers, ...
- Need to handle DRAM refresh
- Need to manage power consumption
- Need to optimize performance & QoS (in the presence of constraints)
 - Reordering is not simple
 - Fairness and QoS needs complicates the scheduling problem

Many Memory Timing Constraints

Latency	Symbol	DRAM cycles	Latency	Symbol	DRAM cycles
Precharge	^{t}RP	11	Activate to read/write	^{t}RCD	11
Read column address strobe	CL	11	Write column address strobe	CWL	8
Additive	AL	0	Activate to activate	^{t}RC	39
Activate to precharge	^{t}RAS	28	Read to precharge	^{t}RTP	6
Burst length	^{t}BL	4	Column address strobe to column address strobe	^{t}CCD	4
Activate to activate (different bank)	^{t}RRD	6	Four activate windows	^{t}FAW	24
Write to read	^{t}WTR	6	Write recovery	^{t}WR	12

Table 4. DDR3 1600 DRAM timing specifications

 From Lee et al., "DRAM-Aware Last-Level Cache Writeback: Reducing Write-Caused Interference in Memory Systems," HPS Technical Report, April 2010.

Many Memory Timing Constraints

- Kim et al., "A Case for Exploiting Subarray-Level Parallelism (SALP) in DRAM," ISCA 2012.
- Lee et al., "Tiered-Latency DRAM: A Low Latency and Low Cost DRAM Architecture," HPCA 2013.

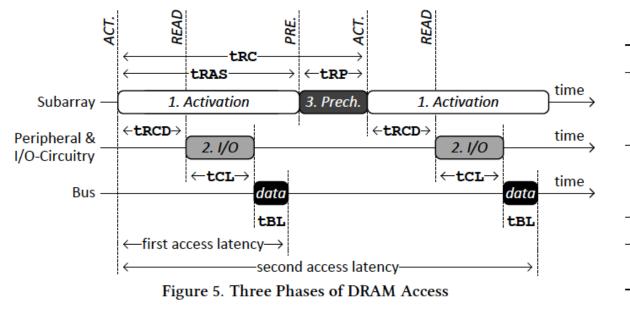
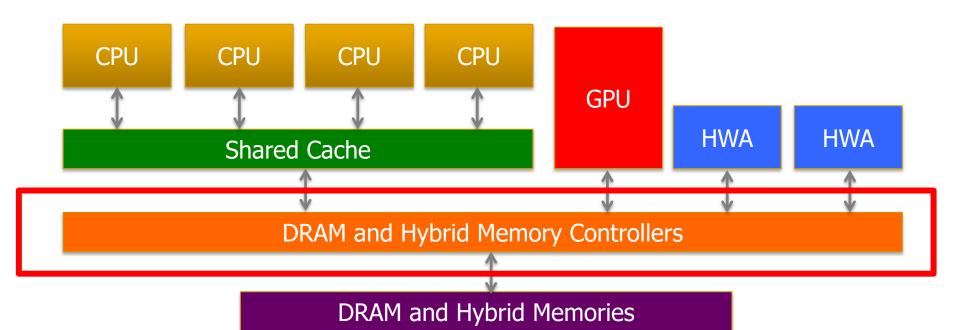


Table 2. Timing Constraints (DDR3-1066) [43]

Phase	Commands	Name	Value
1	$\begin{array}{l} \text{ACT} \rightarrow \text{READ} \\ \text{ACT} \rightarrow \text{WRITE} \end{array}$	tRCD	15ns
	$\mathrm{ACT} \to \mathrm{PRE}$	tRAS	37.5ns
2	$\begin{array}{l} \text{READ} \rightarrow \textit{data} \\ \text{WRITE} \rightarrow \textit{data} \end{array}$	tCL tCWL	15ns 11.25ns
	data burst	tBL	7.5ns
3	$\text{PRE} \rightarrow \text{ACT}$	tRP	15ns
1&3	$ACT \rightarrow ACT$	tRC (tRAS+tRP)	52.5ns

Memory Controller Design Is Becoming More Difficult



- Heterogeneous agents: CPUs, GPUs, and HWAs
- Main memory interference between CPUs, GPUs, HWAs
- Many timing constraints for various memory types
- Many goals at the same time: performance, fairness, QoS, energy efficiency, ...

Reality and Dream

- Reality: It difficult to design a policy that maximizes performance, QoS, energy-efficiency, ...
 - Too many things to think about
 - Continuously changing workload and system behavior

Dream: Wouldn't it be nice if the DRAM controller automatically found a good scheduling policy on its own?

- Problem: DRAM controllers are difficult to design
 - It is difficult for human designers to design a policy that can adapt itself very well to different workloads and different system conditions
- Idea: A memory controller that adapts its scheduling policy to workload behavior and system conditions using machine learning.
- Observation: Reinforcement learning maps nicely to memory control.
- Design: Memory controller is a reinforcement learning agent
 - It dynamically and continuously learns and employs the best scheduling policy to maximize long-term performance.

Ipek+, "Self Optimizing Memory Controllers: A Reinforcement Learning Approach," ISCA 2008.

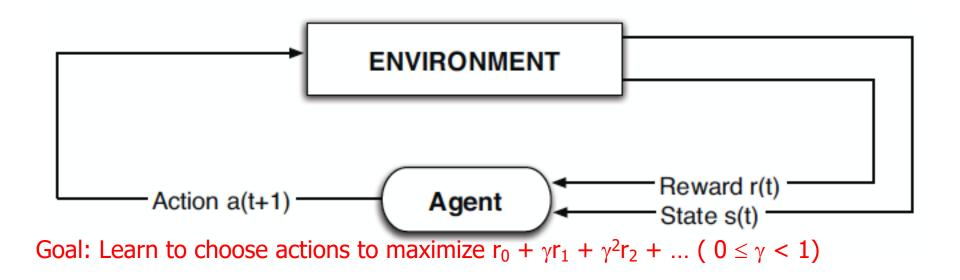
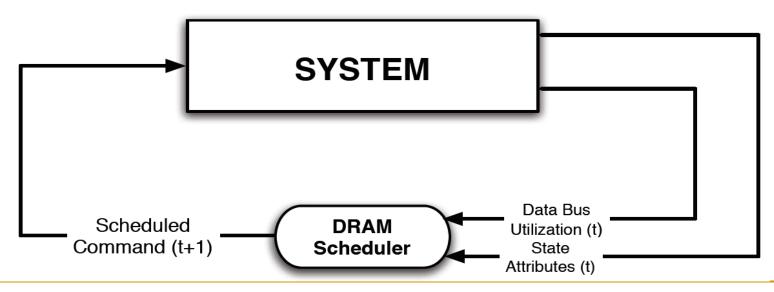


Figure 2: (a) Intelligent agent based on reinforcement learning principles;

- Dynamically adapt the memory scheduling policy via interaction with the system at runtime
 - Associate system states and actions (commands) with long term reward values: each action at a given state leads to a learned reward
 - Schedule command with highest estimated long-term reward value in each state
 - Continuously update reward values for <state, action> pairs based on feedback from system



 Engin Ipek, Onur Mutlu, José F. Martínez, and Rich Caruana, <u>"Self Optimizing Memory Controllers: A Reinforcement Learning</u> <u>Approach</u>"

Proceedings of the <u>35th International Symposium on Computer Architecture</u> (<i>ISCA), pages 39-50, Beijing, China, June 2008.

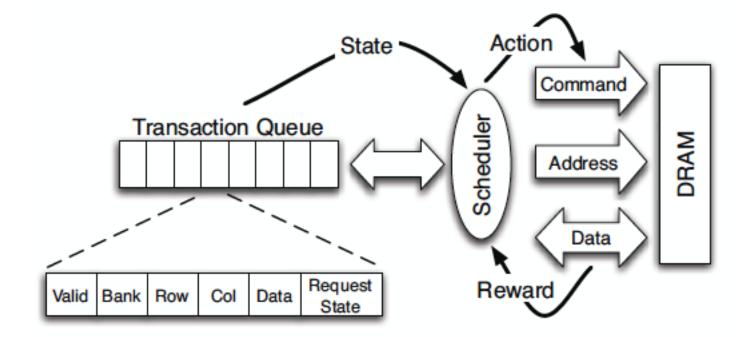


Figure 4: High-level overview of an RL-based scheduler.

States, Actions, Rewards

Reward function

- +1 for scheduling Read and Write commands
- 0 at all other times
- Goal is to maximize long-term data bus utilization

State attributes

- Number of reads, writes, and load misses in transaction queue
- Number of pending writes and ROB heads waiting for referenced row
- Request's relative ROB order

Actions

- Activate
- Write
- Read load miss
- Read store miss
- Precharge pending
- Precharge preemptive
- NOP

Performance Results

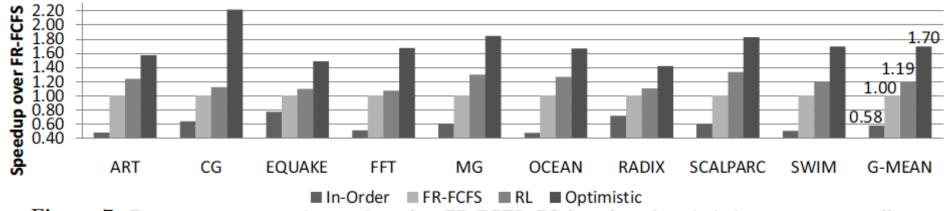


Figure 7: Performance comparison of in-order, FR-FCFS, RL-based, and optimistic memory controllers

Large, robust performance improvements over many human-designed policies

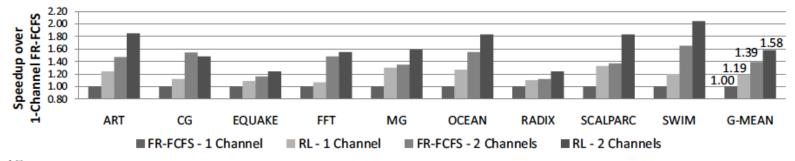


Figure 15: Performance comparison of FR-FCFS and RL-based memory controllers on systems with 6.4GB/s and 12.8GB/s peak DRAM bandwidth

+ Continuous learning in the presence of changing environment

+ Reduced designer burden in finding a good scheduling policy. Designer specifies:

1) What system variables might be useful

2) What target to optimize, but not how to optimize it

-- How to specify different objectives? (e.g., fairness, QoS, ...)

-- Hardware complexity?

-- Design **mindset** and flow

More on Self-Optimizing DRAM Controllers

 Engin Ipek, Onur Mutlu, José F. Martínez, and Rich Caruana,
 "Self Optimizing Memory Controllers: A Reinforcement Learning Approach"
 Proceedings of the <u>35th International Symposium on Computer Architecture</u> (ISCA), pages 39-50, Beijing, China, June 2008.

Self-Optimizing Memory Controllers: A Reinforcement Learning Approach

Engin İpek^{1,2} Onur Mutlu² José F. Martínez¹ Rich Caruana¹

¹Cornell University, Ithaca, NY 14850 USA

² Microsoft Research, Redmond, WA 98052 USA

An Intelligent Architecture

- Data-driven
 - Machine learns the "best" policies (how to do things)
- Sophisticated, workload-driven, changing, far-sighted policies
- Automatic data-driven policy learning
- All controllers are intelligent data-driven agents

We need to rethink design (of all controllers)

Challenge and Opportunity for Future

Self-Optimizing (Data-Driven) Computing Architectures

Corollaries: Architectures Today ...

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SAFARI

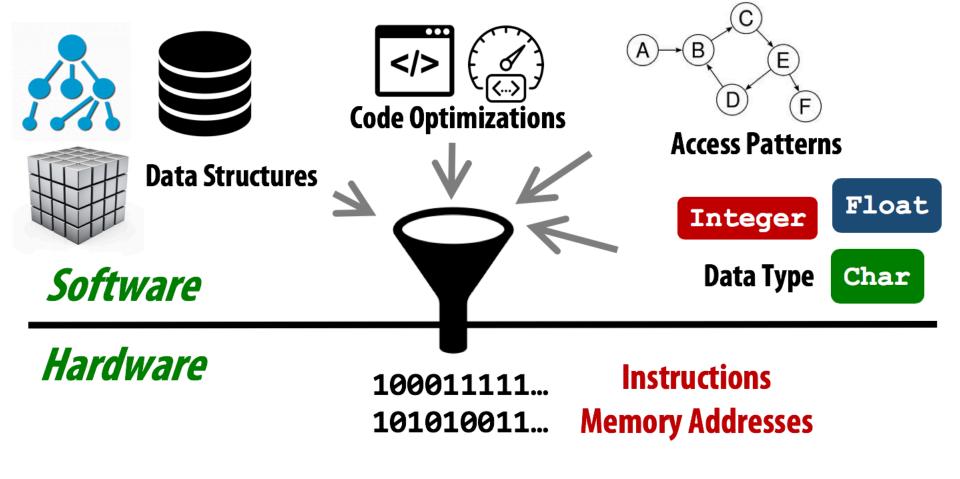
Data-Aware Architectures

- A data-aware architecture understands what it can do with and to each piece of data
- It makes use of different properties of data to improve performance, efficiency and other metrics
 - Compressibility
 - Approximability
 - Locality
 - Sparsity
 - Criticality for Computation X
 - Access Semantics

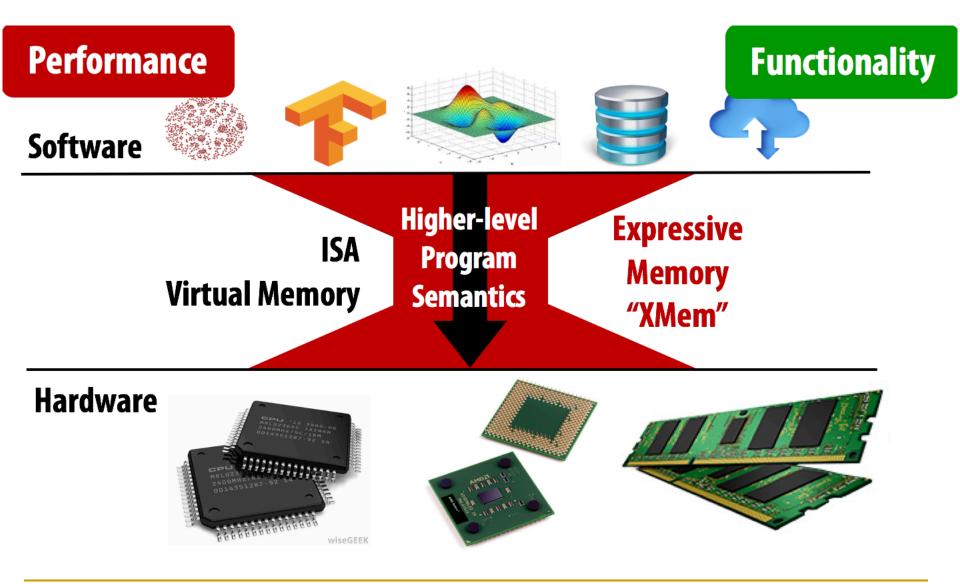
• ...

One Problem: Limited Expressiveness

Higher-level information is not visible to HW



A Solution: More Expressive Interfaces



Expressive (Memory) Interfaces

 Nandita Vijaykumar, Abhilasha Jain, Diptesh Majumdar, Kevin Hsieh, Gennady Pekhimenko, Eiman Ebrahimi, Nastaran Hajinazar, Phillip B. Gibbons and Onur Mutlu, "A Case for Richer Cross-layer Abstractions: Bridging the Semantic Gap with Expressive Memory" Proceedings of the <u>45th International Symposium on Computer Architecture</u> (ISCA), Los Angeles, CA, USA, June 2018.
 [Slides (pptx) (pdf)] [Lightning Talk Slides (pptx) (pdf)] [Lightning Talk Video]

A Case for Richer Cross-layer Abstractions: Bridging the Semantic Gap with Expressive Memory

Nandita Vijaykumar^{†§} Abhilasha Jain[†] Diptesh Majumdar[†] Kevin Hsieh[†] Gennady Pekhimenko[‡] Eiman Ebrahimi[&] Nastaran Hajinazar[∔] Phillip B. Gibbons[†] Onur Mutlu^{§†}

[†]Carnegie Mellon University [‡]University of Toronto [&]NVIDIA ⁺Simon Fraser University [§]ETH Zürich

X-MeM Aids Many Optimizations

	Table	1: Sum	mary of th	e exam	nple me	emoi	ry optimizations that XMem aids.
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Memory optimization	Example semantics provided by XMem (described in §3.3)	Example Benefits of XMem
Cache management	(<i>i</i>) Distinguishing between data structures or pools of similar data; (<i>ii</i>) Working set size; (<i>iii</i>) Data reuse	Enables: (<i>i</i>) applying different caching policies to different data structures or pools of data; (<i>ii</i>) avoiding cache thrashing by <i>knowing</i> the active working set size; (<i>iii</i>) bypassing/prioritizing data that has no/high reuse. (§5)
Page placement in DRAM e.g., [23, 24]	(i) Distinguishing between data structures; (ii) Access pattern; (iii) Access intensity	Enables page placement at the <i>data structure</i> granularity to (<i>i</i>) isolate data structures that have high row buffer locality and (<i>ii</i>) spread out concurrently-accessed irregular data structures across banks and channels to improve parallelism. (§6)
Cache/memory compression e.g., [25–32]	(i) Data type: integer, float, char; (ii) Data properties: sparse, pointer, data index	Enables using a <i>different compression algorithm</i> for each data structure based on data type and data properties, e.g., sparse data encodings, FP-specific compression, delta-based compression for pointers [27].
Data prefetching e.g., [33–36]	(<i>i</i>) Access pattern: strided, irregular, irregular but repeated (e.g., graphs), access stride; (<i>ii</i>) Data type: index, pointer	Enables (<i>i</i>) <i>highly accurate</i> software-driven prefetching while leveraging the benefits of hard- ware prefetching (e.g., by being memory bandwidth-aware, avoiding cache thrashing); (<i>ii</i>) using different prefetcher <i>types</i> for different data structures: e.g., stride [33], tile-based [20], pattern- based [34–37], data-based for indices/pointers [38, 39], etc.
DRAM cache management e.g., [40–46]	(i) Access intensity; (ii) Data reuse; (iii) Working set size	<i>(i)</i> Helps avoid cache thrashing by knowing working set size [44]; <i>(ii)</i> Better DRAM cache management via reuse behavior and access intensity information.
Approximation in memory e.g., [47–53]	<i>(i)</i> Distinguishing between pools of similar data; <i>(ii)</i> Data properties: tolerance towards approximation	Enables (<i>i</i>) each memory component to track how approximable data is (at a fine granularity) to inform approximation techniques; (<i>ii</i>) data placement in heterogeneous reliability memories [54].
Data placement: NUMA systems e.g., [55, 56]	(<i>i</i>) Data partitioning across threads (i.e., relating data to threads that access it); (<i>ii</i>) Read-Write properties	Reduces the need for profiling or data migration <i>(i)</i> to co-locate data with threads that access it and <i>(ii)</i> to identify Read-Only data, thereby enabling techniques such as replication.
Data placement: hybrid memories e.g., [16, 57, 58]	(i) Read-Write properties (Read-Only/Read-Write); (ii) Access intensity; (iii) Data structure size; (iv) Access pattern	Avoids the need for profiling/migration of data in hybrid memories to (<i>i</i>) effectively manage the asymmetric read-write properties in NVM (e.g., placing Read-Only data in the NVM) [16, 57]; (<i>ii</i>) make tradeoffs between data structure "hotness" and size to allocate fast/high bandwidth memory [14]; and (<i>iii</i>) leverage row-buffer locality in placement based on access pattern [45].
Managing NUCA systems e.g., [15, 59]	(<i>i</i>) Distinguishing pools of similar data; (<i>ii</i>) Access intensity; (<i>iii</i>) Read-Write or Private-Shared properties	<i>(i)</i> Enables using different cache policies for different data pools (similar to [15]); <i>(ii)</i> Reduces the need for reactive mechanisms that detect sharing and read-write characteristics to inform cache policies.

Expressive (Memory) Interfaces for GPUs

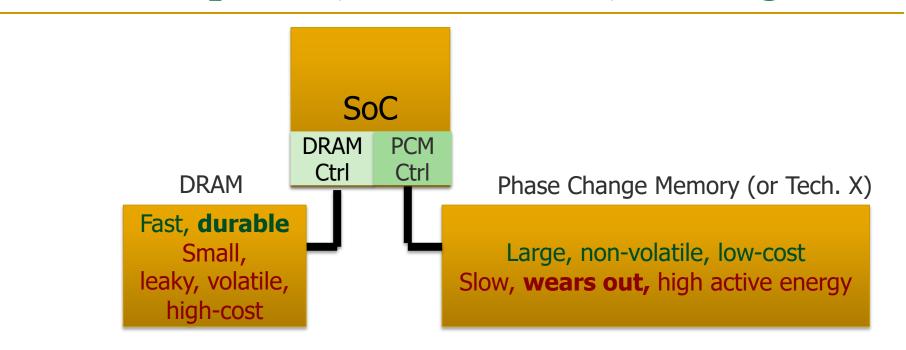
 Nandita Vijaykumar, Eiman Ebrahimi, Kevin Hsieh, Phillip B. Gibbons and Onur Mutlu, "The Locality Descriptor: A Holistic Cross-Layer Abstraction to Express Data Locality in GPUs" Proceedings of the <u>45th International Symposium on Computer Architecture</u> (ISCA), Los Angeles, CA, USA, June 2018. [Slides (pptx) (pdf)] [Lightning Talk Slides (pptx) (pdf)] [Lightning Talk Video]

The Locality Descriptor:

A Holistic Cross-Layer Abstraction to Express Data Locality in GPUs

Nandita Vijaykumar^{†§} Eiman Ebrahimi[‡] Kevin Hsieh[†] Phillip B. Gibbons[†] Onur Mutlu^{§†} [†]Carnegie Mellon University [‡]NVIDIA [§]ETH Zürich

An Example: Hybrid Memory Management



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.

SAFARI

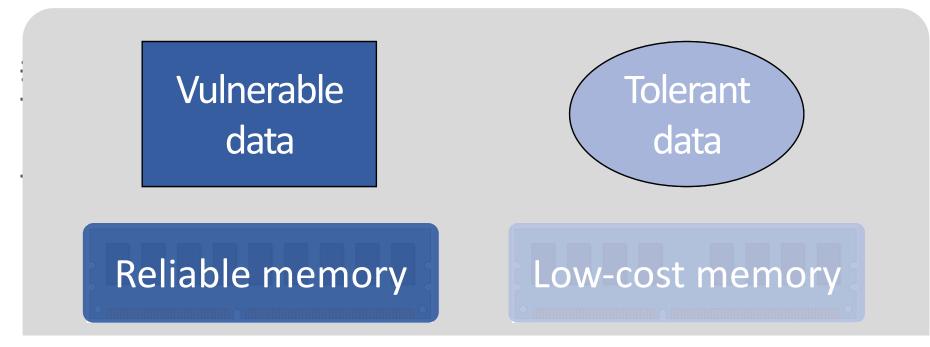
An Example: Heterogeneous-Reliability Memory

Yixin Luo, Sriram Govindan, Bikash Sharma, Mark Santaniello, Justin Meza, Aman Kansal, Jie Liu, Badriddine Khessib, Kushagra Vaid, and Onur Mutlu,
 <u>"Characterizing Application Memory Error Vulnerability to Optimize</u>
 <u>Data Center Cost via Heterogeneous-Reliability Memory"</u>
 *Proceedings of the <u>44th Annual IEEE/IFIP International Conference on</u>
 <u>Dependable Systems and Networks (DSN</u>), Atlanta, GA, June 2014. [Summary]
 [Slides (pptx) (pdf)] [Coverage on ZDNet]*

Characterizing Application Memory Error Vulnerability to Optimize Datacenter Cost via Heterogeneous-Reliability Memory

Yixin Luo Sriram Govindan^{*} Bikash Sharma^{*} Mark Santaniello^{*} Justin Meza Aman Kansal^{*} Jie Liu^{*} Badriddine Khessib^{*} Kushagra Vaid^{*} Onur Mutlu Carnegie Mellon University, yixinluo@cs.cmu.edu, {meza, onur}@cmu.edu *Microsoft Corporation, {srgovin, bsharma, marksan, kansal, jie.liu, bkhessib, kvaid}@microsoft.com

Exploiting Memory Error Tolerance with Hybrid Memory Systems



On Microsoft's Web Search workload Reduces server hardware cost by 4.7 % Achieves single server availability target of 99.90 % Heterogeneous-Reliability Memory [DSN 2014]

More on Heterogeneous-Reliability Memory

Yixin Luo, Sriram Govindan, Bikash Sharma, Mark Santaniello, Justin Meza, Aman Kansal, Jie Liu, Badriddine Khessib, Kushagra Vaid, and Onur Mutlu,
 <u>"Characterizing Application Memory Error Vulnerability to Optimize</u>
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Characterizing Application Memory Error Vulnerability to Optimize Datacenter Cost via Heterogeneous-Reliability Memory

Yixin Luo Sriram Govindan^{*} Bikash Sharma^{*} Mark Santaniello^{*} Justin Meza Aman Kansal^{*} Jie Liu^{*} Badriddine Khessib^{*} Kushagra Vaid^{*} Onur Mutlu Carnegie Mellon University, yixinluo@cs.cmu.edu, {meza, onur}@cmu.edu *Microsoft Corporation, {srgovin, bsharma, marksan, kansal, jie.liu, bkhessib, kvaid}@microsoft.com

Another Example: EDEN for DNNs

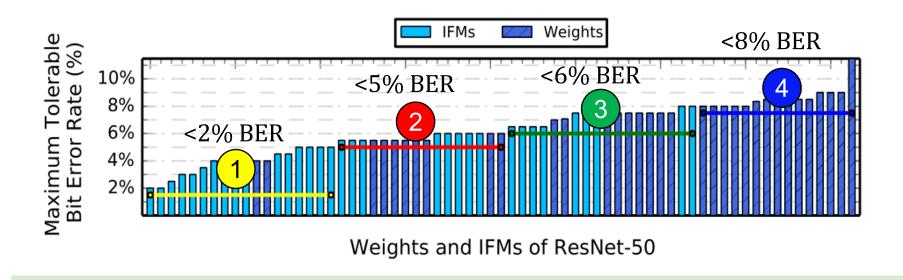
- Deep Neural Network evaluation is very DRAM-intensive (especially for large networks)
- 1. Some data and layers in DNNs are very tolerant to errors
- 2. Reduce DRAM latency and voltage on such data and layers

3. While still achieving a user-specified DNN accuracy target by making training DRAM-error-aware

Data-aware management of DRAM latency and voltage for Deep Neural Network Inference

Example DNN Data Type to DRAM Mapping

Mapping example of ResNet-50:



Map more error-tolerant DNN layers to DRAM partitions with lower voltage/latency

4 DRAM partitions with different error rates

EDEN: Data-Aware Efficient DNN Inference

 Skanda Koppula, Lois Orosa, A. Giray Yaglikci, Roknoddin Azizi, Taha Shahroodi, Konstantinos Kanellopoulos, and Onur Mutlu, "EDEN: Enabling Energy-Efficient, High-Performance Deep Neural Network Inference Using Approximate DRAM" Proceedings of the 52nd International Symposium on Microarchitecture (MICRO), Columbus, OH, USA, October 2019.
 [Lightning Talk Slides (pptx) (pdf)]
 [Lightning Talk Video (90 seconds)]

EDEN: Enabling Energy-Efficient, High-Performance Deep Neural Network Inference Using Approximate DRAM

Skanda Koppula Lois Orosa A. Giray Yağlıkçı Roknoddin Azizi Taha Shahroodi Konstantinos Kanellopoulos Onur Mutlu ETH Zürich

SMASH: SW/HW Indexing Acceleration

- Konstantinos Kanellopoulos, Nandita Vijaykumar, Christina Giannoula, Roknoddin Azizi, Skanda Koppula, Nika Mansouri Ghiasi, Taha Shahroodi, Juan Gomez-Luna, and Onur Mutlu,
 - "SMASH: Co-designing Software Compression and Hardware-Accelerated Indexing for Efficient Sparse Matrix Operations"

Proceedings of the <u>52nd International Symposium on</u> <u>Microarchitecture</u> (**MICRO**), Columbus, OH, USA, October 2019. [Slides (pptx) (pdf)] [Lightning Talk Slides (pptx) (pdf)] [Poster (pptx) (pdf)] [Lightning Talk Video (90 seconds)] [Full Talk Lecture (30 minutes)]

SMASH: Co-designing Software Compression and Hardware-Accelerated Indexing for Efficient Sparse Matrix Operations

Konstantinos Kanellopoulos¹ Nandita Vijaykumar^{2,1} Christina Giannoula^{1,3} Roknoddin Azizi¹ Skanda Koppula¹ Nika Mansouri Ghiasi¹ Taha Shahroodi¹ Juan Gomez Luna¹ Onur Mutlu^{1,2} ¹ETH Zürich ²Carnegie Mellon University ³National Technical University of Athens

Data-Aware Virtual Memory Framework

Nastaran Hajinazar, Pratyush Patel, Minesh Patel, Konstantinos Kanellopoulos, Saugata Ghose, Rachata Ausavarungnirun, Geraldo Francisco de Oliveira Jr., Jonathan Appavoo, Vivek Seshadri, and Onur Mutlu, "The Virtual Block Interface: A Flexible Alternative to the Conventional Virtual Memory Framework" *Proceedings of the <u>47th International Symposium on Computer Architecture</u> (ISCA), Virtual, June 2020. [Slides (pptx) (pdf)] [Lightning Talk Slides (pptx) (pdf)] [ARM Research Summit Poster (pptx) (pdf)] [Talk Video (26 minutes)] [Lightning Talk Video (3 minutes)]*

The Virtual Block Interface: A Flexible Alternative to the Conventional Virtual Memory Framework

Nastaran Hajinazar^{*†} Pratyush Patel[⋈] Minesh Patel[★] Konstantinos Kanellopoulos[★] Saugata Ghose[‡] Rachata Ausavarungnirun[⊙] Geraldo F. Oliveira[★] Jonathan Appavoo[◊] Vivek Seshadri[▽] Onur Mutlu^{*‡}

*ETH Zürich [†]Simon Fraser University \bowtie University of Washington [‡]Carnegie Mellon University \odot King Mongkut's University of Technology North Bangkok \diamond Boston University \bigtriangledown Microsoft Research India

Challenge and Opportunity for Future

Data-Aware (Expressive) Computing Architectures

Concluding Remarks

Recap: Corollaries: Architectures Today

- Architectures are terrible at dealing with data
 - Designed to mainly store and move data vs. to compute
 - They are processor-centric as opposed to data-centric
- Architectures are terrible at taking advantage of vast amounts of data (and metadata) available to them
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Concluding Remarks

- It is time to design principled system architectures to solve the data handling (i.e., memory/storage) problem
- Design complete systems to be truly balanced, highperformance, and energy-efficient → intelligent architectures
 Data-centric, data-driven, data-aware
- Enable computation capability inside and close to memory
- This can
 - □ Lead to **orders-of-magnitude** improvements
 - Enable new applications & computing platforms
 - Enable better understanding of nature

Architectures for Intelligent Machines

Data-centric

Data-driven

Data-aware





SAFARI

Source: http://spectrum.ieee.org/image/MjYzMzAyMg.jpeg

We Need to Revisit the Entire Stack

	Problem						
	Aigorithm						
	Program/Language						
	System Software						
	SW/HW Interface						
	Micro-architecture						
	Logic						
	Devices						
	Electrons						

We can get there step by step

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We Need to Exploit Good Principles

- Data-centric system design
- All components intelligent
- Better cross-layer communication, better interfaces
- Better-than-worst-case design
- Heterogeneity
- Flexibility, adaptability



PIM Review and Open Problems

Processing Data Where It Makes Sense: Enabling In-Memory Computation

Onur Mutlu^{a,b}, Saugata Ghose^b, Juan Gómez-Luna^a, Rachata Ausavarungnirun^{b,c}

^aETH Zürich ^bCarnegie Mellon University ^cKing Mongkut's University of Technology North Bangkok

Onur Mutlu, Saugata Ghose, Juan Gomez-Luna, and Rachata Ausavarungnirun, "Processing Data Where It Makes Sense: Enabling In-Memory Computation" Invited paper in Microprocessors and Microsystems (MICPRO), June 2019. [arXiv version]

https://arxiv.org/pdf/1903.03988.pdf

PIM Review and Open Problems (II)

A Workload and Programming Ease Driven Perspective of Processing-in-Memory

Saugata Ghose†Amirali Boroumand†Jeremie S. Kim†§Juan Gómez-Luna§Onur Mutlu§††Carnegie Mellon University§ETH Zürich

Saugata Ghose, Amirali Boroumand, Jeremie S. Kim, Juan Gomez-Luna, and Onur Mutlu, "Processing-in-Memory: A Workload-Driven Perspective" *Invited Article in <u>IBM Journal of Research & Development</u>, Special Issue on Hardware for Artificial Intelligence*, to appear in November 2019. [Preliminary arXiv version]

https://arxiv.org/pdf/1907.12947.pdf

A Longer Version of This Talk

 Onur Mutlu,
 "Intelligent Architectures for Intelligent Machines"
 Distinguished Lecture at <u>HKUST Engineering and HKSTP</u> <u>Distinguished Speaker Series</u>, Virtual, 7 October 2020.
 [Slides (pptx) (pdf)]
 [Talk Video (1 hour 24 minutes)]
 [Related Keynote Paper from VLSI-DAT 2020]

https://www.youtube.com/watch?v=CmuhFB-QEWw

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

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, ...

Acknowledgments

SAFARI Research Group safari.ethz.ch



Onur Mutlu's SAFARI Research Group

Computer architecture, HW/SW, systems, bioinformatics, security, memory

https://safari.ethz.ch/safari-newsletter-april-2020/



SAFARI Newsletter April 2020 Edition

https://safari.ethz.ch/safari-newsletter-april-2020/





View in your browser

Think Big, Aim High



Dear SAFARI friends,

Intelligent Architectures for Intelligent Machines

Onur Mutlu <u>omutlu@gmail.com</u> <u>https://people.inf.ethz.ch/omutlu</u>

2 December 2020

Huawei Compute and Storage Technology Workshop Keynote

SAFARI

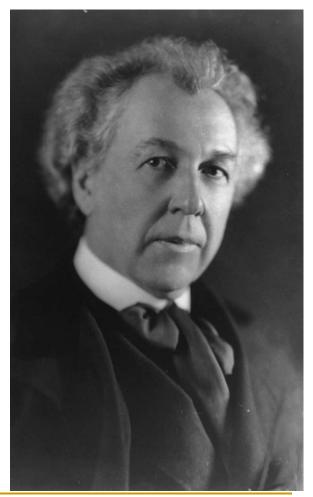
ETH zürich

Carnegie Mellon

Backup Slides

A Quote from A Famous Architect

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



Precedent-Based Design?

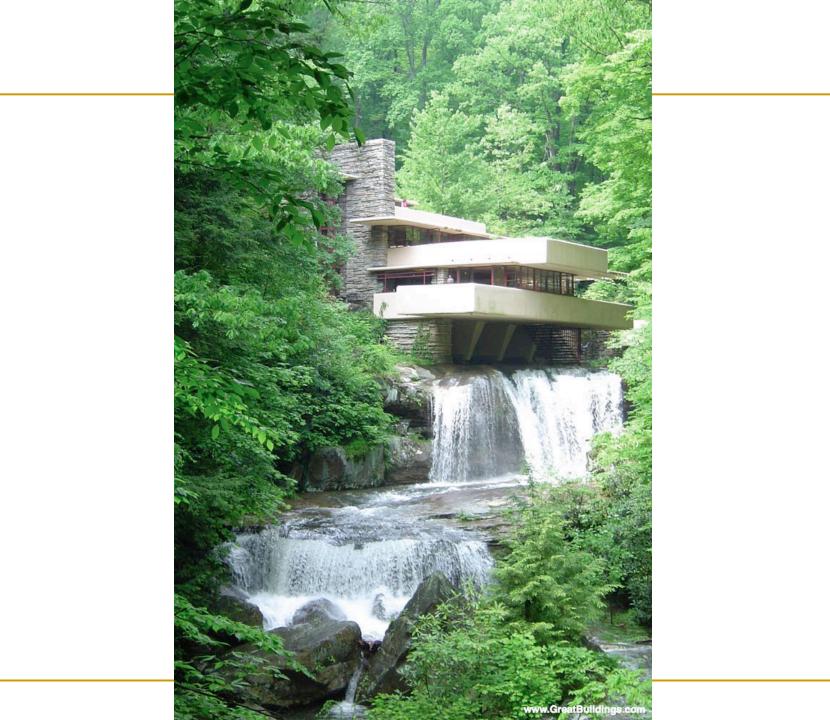
"architecture [...] based upon principle, and not upon precedent"



Principled Design

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





The Overarching Principle

Organic architecture

From Wikipedia, the free encyclopedia

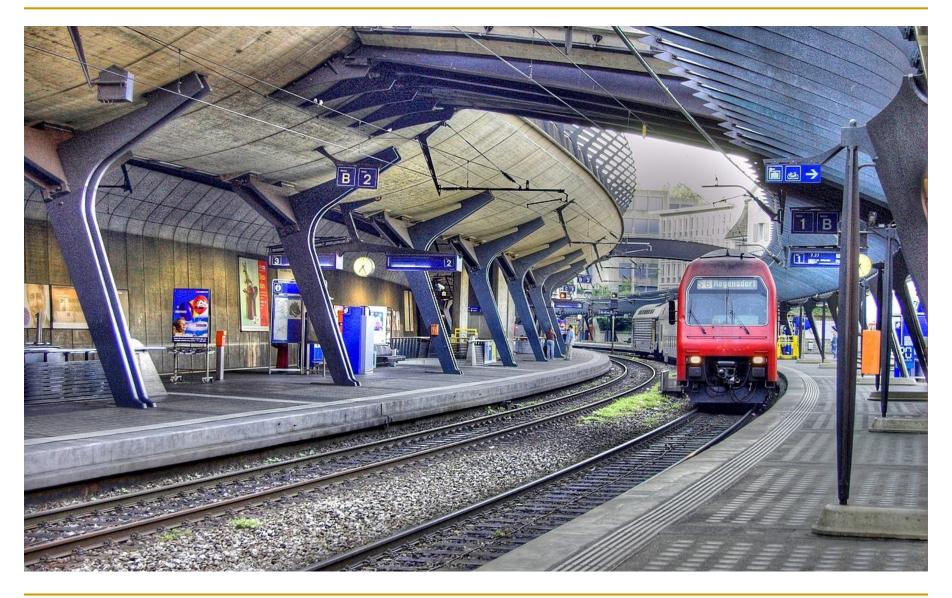
Organic architecture is a philosophy of architecture which promotes harmony between human habitation and the natural world through design approaches so sympathetic and well integrated with its site, that buildings, furnishings, and surroundings become part of a unified, interrelated composition.

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.

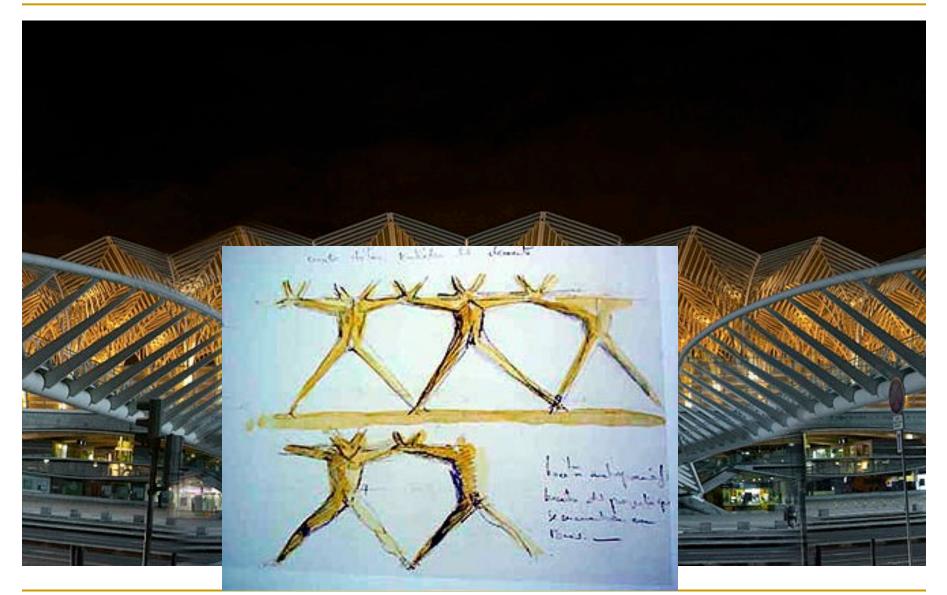
Another Example: Precedent-Based Design



Principled Design



Another Principled Design



Source: By Martín Gómez Tagle - Lisbon, Portugal, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=13764903 Source: http://www.arcspace.com/exhibitions/unsorted/santiago-calatrava/

Another Principled Design



Principle Applied to Another Structure



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?



Source: http://spectrum.ieee.org/image/MjYzMzAyMg.jpeg

Readings, Videos, Reference Materials

- Onur Mutlu
 - Full Professor @ ETH Zurich, since September 2015
 - Strecker Professor @ Carnegie Mellon University ECE/CS, 2009-2016, 2016-...
 - □ PhD from UT-Austin, worked at Google, VMware, Microsoft Research, Intel, AMD
 - <u>https://people.inf.ethz.ch/omutlu/</u>
 - omutlu@gmail.com (Best way to reach me)
 - https://people.inf.ethz.ch/omutlu/projects.htm
- Research and Teaching in:
 - Computer architecture, computer systems, hardware security, bioinformatics
 - Memory and storage systems
 - Hardware security, safety, predictability
 - Fault tolerance
 - Hardware/software cooperation
 - Architectures for bioinformatics, health, medicine

• ...

https://www.youtube.com/onurmutlulectures

Future Computing Architectures

https://www.youtube.com/watch?v=kgiZISOcGFM&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=1

Enabling In-Memory Computation

https://www.youtube.com/watch?v=njX 14584Jw&list=PL5Q2soXY2Zi8D 5MGV6EnXEJHnV2YFBJl&index=16

Accelerating Genome Analysis

https://www.youtube.com/watch?v=hPnSmfwu2-A&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=9

Rethinking Memory System Design

https://www.youtube.com/watch?v=F7xZLNMIY1E&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=3

Intelligent Architectures for Intelligent Machines

https://www.youtube.com/watch?v=n8Aj_A0WSg8&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=22

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Accelerated Memory Course (~6.5 hours)

ACACES 2018

- Memory Systems and Memory-Centric Computing Systems
- Taught by Onur Mutlu July 9-13, 2018
- ~6.5 hours of lectures

Website for the Course including Videos, Slides, Papers

- https://people.inf.ethz.ch/omutlu/acaces2018.html
- https://www.youtube.com/playlist?list=PL5Q2soXY2Zi-HXxomthrpDpMJm05P6J9x
- All Papers are at:
 - <u>https://people.inf.ethz.ch/omutlu/projects.htm</u>
 - Final lecture notes and readings (for all topics)

Longer Memory Course (~18 hours)

Tu Wien 2019

- Memory Systems and Memory-Centric Computing Systems
- Taught by Onur Mutlu June 12-19, 2019
- ~18 hours of lectures
- Website for the Course including Videos, Slides, Papers
 - https://safari.ethz.ch/memory_systems/TUWien2019
 - https://www.youtube.com/playlist?list=PL5Q2soXY2Zi_gntM55
 VoMIKIw7YrXOhbl
- All Papers are at:
 - https://people.inf.ethz.ch/omutlu/projects.htm
 - Final lecture notes and readings (for all topics)

An Interview on Research and Education

- Computing Research and Education (@ ISCA 2019)
 - https://www.youtube.com/watch?v=8ffSEKZhmvo&list=PL5Q2 soXY2Zi_4oP9LdL3cc8G6NIjD2Ydz

- Maurice Wilkes Award Speech (10 minutes)
 - https://www.youtube.com/watch?v=tcQ3zZ3JpuA&list=PL5Q2 soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=15

Reference Overview Paper I

ΔΓΔΠ

Processing Data Where It Makes Sense: Enabling In-Memory Computation

Onur Mutlu^{a,b}, Saugata Ghose^b, Juan Gómez-Luna^a, Rachata Ausavarungnirun^{b,c}

^aETH Zürich ^bCarnegie Mellon University ^cKing Mongkut's University of Technology North Bangkok

Onur Mutlu, Saugata Ghose, Juan Gomez-Luna, and Rachata Ausavarungnirun, "Processing Data Where It Makes Sense: Enabling In-Memory Computation" Invited paper in Microprocessors and Microsystems (MICPRO), June 2019. [arXiv version]

https://arxiv.org/pdf/1903.03988.pdf

Reference Overview Paper II

A Workload and Programming Ease Driven Perspective of Processing-in-Memory

Saugata Ghose†Amirali Boroumand†Jeremie S. Kim†§Juan Gómez-Luna§Onur Mutlu§††Carnegie Mellon University§ETH Zürich

Saugata Ghose, Amirali Boroumand, Jeremie S. Kim, Juan Gomez-Luna, and Onur Mutlu, "Processing-in-Memory: A Workload-Driven Perspective" *Invited Article in IBM Journal of Research & Development, Special Issue on Hardware for Artificial Intelligence*, to appear in November 2019. [Preliminary arXiv version]

https://arxiv.org/pdf/1907.12947.pdf

Reference Overview Paper III

Enabling the Adoption of Processing-in-Memory: Challenges, Mechanisms, Future Research Directions

SAUGATA GHOSE, KEVIN HSIEH, AMIRALI BOROUMAND, RACHATA AUSAVARUNGNIRUN

Carnegie Mellon University

ONUR MUTLU ETH Zürich and Carnegie Mellon University

Saugata Ghose, Kevin Hsieh, Amirali Boroumand, Rachata Ausavarungnirun, Onur Mutlu, "Enabling the Adoption of Processing-in-Memory: Challenges, Mechanisms, Future Research Directions" Invited Book Chapter, to appear in 2018. [Preliminary arxiv.org version]

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Reference Overview Paper IV

 Onur Mutlu and Lavanya Subramanian, <u>"Research Problems and Opportunities in Memory</u> <u>Systems"</u> *Invited Article in <u>Supercomputing Frontiers and Innovations</u> (SUPERFRI), 2014/2015.*

Research Problems and Opportunities in Memory Systems

Onur Mutlu¹, Lavanya Subramanian¹

https://people.inf.ethz.ch/omutlu/pub/memory-systems-research_superfri14.pdf

Reference Overview Paper V

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> <i>Europe Conference (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

https://people.inf.ethz.ch/omutlu/pub/rowhammer-and-other-memory-issues_date17.pdf

Reference Overview Paper VI

Onur Mutlu,
 <u>"Memory Scaling: A Systems Architecture</u>
 <u>Perspective"</u>
 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/

https://people.inf.ethz.ch/omutlu/pub/memory-scaling_memcon13.pdf

Reference Overview Paper VII



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

https://arxiv.org/pdf/1706.08642

Reference Overview Paper VIII

Onur Mutlu and Jeremie Kim,
 "RowHammer: A Retrospective"
 <u>IEEE Transactions on Computer-Aided Design of Integrated</u>
 <u>Circuits and Systems</u> (TCAD) Special Issue on Top Picks in
 Hardware and Embedded Security, 2019.
 [Preliminary arXiv version]

RowHammer: A Retrospective

Onur Mutlu^{§‡} Jeremie S. Kim^{‡§} [§]ETH Zürich [‡]Carnegie Mellon University

Related Videos and Course Materials (I)

- <u>Undergraduate Digital Design & Computer</u>
 <u>Architecture Course Lecture</u>
 <u>Videos (2020, 2019, 2018, 2017, 2015, 2014, 2013)</u>
- Undergraduate Digital Design & Computer Architecture Course Materials (2020, 2019, 2018, 2015, 2014, 2013)
- Graduate Computer Architecture Course Lecture Videos (2019, 2018, 2017, 2015, 2013)
- Graduate Computer Architecture Course Materials (2019, 2018, 2017, 2015, 2013)
- Parallel Computer Architecture Course Materials (Lecture Videos)

Related Videos and Course Materials (II)

- Seminar in Computer Architecture Course Lecture
 Videos (Spring 2020, Fall 2019, Spring 2019, 2018)
- <u>Seminar in Computer Architecture Course</u>
 <u>Materials (Spring 2020, Fall 2019, Spring 2019, 2018)</u>
- Memory Systems Course Lecture Videos (Sept 2019, July 2019, June 2019, October 2018)
- Memory Systems Short Course Lecture Materials (Sept 2019, July 2019, June 2019, October 2018)
- ACACES Summer School Memory Systems Course Lecture Videos (2018, 2013)
- ACACES Summer School Memory Systems Course Materials (2018, 2013)

Some Open Source Tools (I)

- Rowhammer Program to Induce RowHammer Errors
 - <u>https://github.com/CMU-SAFARI/rowhammer</u>
- Ramulator Fast and Extensible DRAM Simulator
 - <u>https://github.com/CMU-SAFARI/ramulator</u>
- MemSim Simple Memory Simulator
 - https://github.com/CMU-SAFARI/memsim
- NOCulator Flexible Network-on-Chip Simulator
 - <u>https://github.com/CMU-SAFARI/NOCulator</u>
- SoftMC FPGA-Based DRAM Testing Infrastructure
 - https://github.com/CMU-SAFARI/SoftMC
- Other open-source software from my group
 - https://github.com/CMU-SAFARI/

<u>http://www.ece.cmu.edu/~safari/tools.html</u>
SAFARI

Some Open Source Tools (II)

- MQSim A Fast Modern SSD Simulator
 - <u>https://github.com/CMU-SAFARI/MQSim</u>
- Mosaic GPU Simulator Supporting Concurrent Applications
 - https://github.com/CMU-SAFARI/Mosaic
- IMPICA Processing in 3D-Stacked Memory Simulator
 - https://github.com/CMU-SAFARI/IMPICA
- SMLA Detailed 3D-Stacked Memory Simulator
 - https://github.com/CMU-SAFARI/SMLA
- HWASim Simulator for Heterogeneous CPU-HWA Systems
 <u>https://github.com/CMU-SAFARI/HWASim</u>
- Other open-source software from my group
 - https://github.com/CMU-SAFARI/

<u>http://www.ece.cmu.edu/~safari/tools.html</u>
SAFARI

More Open Source Tools (III)

- A lot more open-source software from my group
 - https://github.com/CMU-SAFARI/
 - http://www.ece.cmu.edu/~safari/tools.html

SAFARI Research Group at ETH Zurich and Carnegie Mellon University									
				and Carnegie Mellon University.					
Repositories (30)	H Zurich and Carnegi	Teams 1 IProjects	 omutlu@gmail.com omutlu@gmail.com Settings 						
Search repositories		Type: All -	Language: All -	Customize pinned repositories					
modern multi-queue (M MQSim faithfully model	IQ) SSDs as well as tr s new high-bandwidtl	eling the performance of aditional SATA based SSDs. h protocol implementations, d-to-end latency of	M^	Top languages ● C++ ● C ● C# ● AGS Script ● Verilog					
steady-state SSD conditions, and the full end-to-end latency of requests in modern SSDs. It is described in detail in the FAST 2018 paper by A ● C++ ★ 14				Most used topics Manage					

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ramulator-pim

A fast and flexible simulation infrastructure for exploring general-purpose processing-in-memory (PIM) architectures. Ramulator-PIM combines a widely-used simulator for out-of-order and in-order processors (ZSim) with Ramulator, a DRAM simulator with memory models for DDRx, LPDDRx, GDDRx, WIOx, HBMx, and HMCx. Ramulator is described in the IEEE ...

● C++ ♀ 11 ☆ 29 ① 6 ☎ 0 Updated 19 days ago

SMASH

SMASH is a hardware-software cooperative mechanism that enables highly-efficient indexing and storage of sparse matrices. The key idea of SMASH is to compress sparse matrices with a hierarchical bitmap compression format that can be accelerated from hardware. Described by Kanellopoulos et al. (MICRO '19) https://people.inf.ethz.ch/omutlu/pub/SMA...

●C ♀1 ☆6 ①0 ₥0 Updated on May 17

MQSim

MQSim is a fast and accurate simulator modeling the performance of modern multi-queue (MQ) SSDs as well as traditional SATA based SSDs. MQSim faithfully models new high-bandwidth protocol implementations, steady-state SSD conditions, and the full end-to-end latency of requests in modern SSDs. It is described in detail in the FAST 2018 paper by A...

Apollo

Apollo is an assembly polishing algorithm that attempts to correct the errors in an assembly. It can take multiple set of reads in a single run and polish the assemblies of genomes of any size. Described in the Bioinformatics journal paper (2020) by Firtina et al. at https://people.inf.ethz.ch/omutlu/pub/apollotechnology-independent-genome-asse...

●C++ 亟 GPL-3.0 撃1 ☆12 ①0 カュ0 Updated on May10

ramulator

A Fast and Extensible DRAM Simulator, with built-in support for modeling many different DRAM technologies including DDRx, LPDDRx, GDDRx, WIOx, HBMx, and various academic proposals. Described in the IEEE CAL 2015 paper by Kim et al. at http://users.ece.cmu.edu/~omutlu/pub/ramulator_dram_ simulator-ieee-cal15.pdf

● C++ 콰 MIT ♀ 93 ☆ 170 ① 37 \$\$ 2 Updated on Apr 13

Shifted-Hamming-Distance

Source code for the Shifted Hamming Distance (SHD) filtering mechanism for sequence alignment. Described ______ in the Bioinformatics journal paper (2015) by Xin et al. at http://users.ece.cmu.edu/~omutlu/pub/shiftedhamming-distance_bioinformatics15_proofs.pdf

●C 亟 GPL-2.0 ೪5 ☆20 ① 0 沿1 Updated on Mar 29

SneakySnake

The first and the only pre-alignment filtering algorithm that works on all modern high-performance computing architectures. It works efficiently and fast on CPU, FPGA, and GPU architectures and that greatly (by more than two orders of magnitude) expedites sequence alignment calculation. Described by Alser et al. (preliminary version at https://a...

● VHDL 亟 GPL-3.0 学3 ☆11 ① 0 カカ 0 Updated on Mar 10

AirLift

AirLift is a tool that updates mapped reads from one reference genome to another. Unlike existing tools, It accounts for regions not shared between the two reference genomes and enables remapping across all parts of the references. Described by Kim et al. (preliminary version at http://arxiv.org/abs/1912.08735)

●C ♀0 ☆3 ①0 カゥ0 Updated on Feb 19

GPGPUSim-Ramulator

The source code for GPGPUSim+Ramulator simulator. In this version, GPGPUSim uses Ramulator to simulate the DRAM. This simulator is used to produce some of the

Referenced Papers and Talks

All are available at

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

http://scholar.google.com/citations?user=7XyGUGkAAAAJ&hl=en

https://www.youtube.com/onurmutlulectures



An Interview on Research and Education

- Computing Research and Education (@ ISCA 2019)
 - https://www.youtube.com/watch?v=8ffSEKZhmvo&list=PL5Q2 soXY2Zi_4oP9LdL3cc8G6NIjD2Ydz

- Maurice Wilkes Award Speech (10 minutes)
 - https://www.youtube.com/watch?v=tcQ3zZ3JpuA&list=PL5Q2 soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=15

Low-Latency Memory

Workload-DRAM Interaction Analysis

Saugata Ghose, Tianshi Li, Nastaran Hajinazar, Damla Senol Cali, and Onur Mutlu,

"Demystifying Workload–DRAM Interactions: An Experimental Study"

Proceedings of the <u>ACM International Conference on Measurement and</u> Modeling of Computer Systems (SIGMETRICS), Phoenix, AZ, USA, June 2019. Preliminary arXiv Version

[Abstract]

[Slides (pptx) (pdf)]

Demystifying Complex Workload–DRAM Interactions: An Experimental Study

Tianshi Li † Saugata Ghose[†] Nastaran Hajinazar^{‡†} Damla Senol Cali[†] Onur Mutlu^{§†}

[†]Carnegie Mellon University [‡]Simon Fraser University

[§]ETH Zürich

Why Study Workload–DRAM Interactions?

- Manufacturers are developing many new types of DRAM
 - **DRAM limits performance, energy improvements:** new types may overcome some limitations
 - Memory systems now serve a very diverse set of applications: can no longer take a one-size-fits-all approach

• So which DRAM type works best with which application?

- Difficult to understand intuitively due to the complexity of the interaction
- Can't be tested methodically on real systems: new type needs a new CPU
- We perform a wide-ranging experimental study to uncover the combined behavior of workloads and DRAM types
 - 115 prevalent/emerging applications and multiprogrammed workloads
 - 9 modern DRAM types: DDR3, DDR4, GDDR5, HBM, HMC, LPDDR3, LPDDR4, Wide I/O, Wide I/O 2

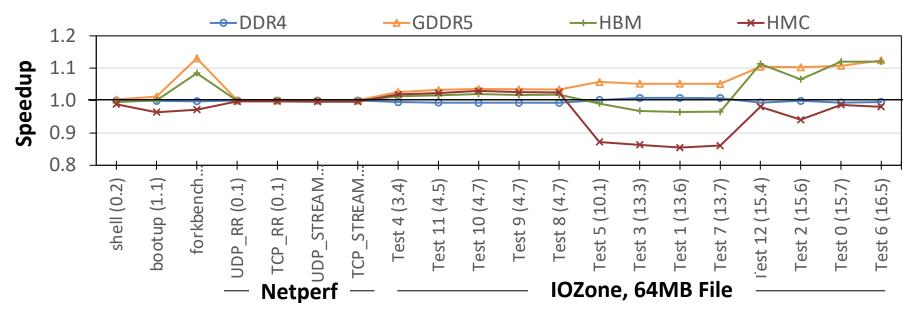
Modern DRAM Types: Comparison to DDR3

DRAM Type	Banks per Rank	Bank Groups	3D- Stacked	Low- Power	Bank groups Bank Group Bank Group Bank Group Bank Group
DDR3	8				Bank Bank Bank Bank
DDR4	16	✓ _	increased	latency	(/ (/
GDDR5	16	🗸 🚺	creased are	ea/power	memory channel
HBM High- Bandwidth Memory	16		√		3D-stacked DRAM high bandwidth with
HMC Hybrid Memory Cube	256 narrower rows, higher latency ✓				Through-Silicon Vias (TSVs)
Wide I/O	4		\checkmark	\checkmark	
Wide I/O 2	8		\checkmark	\checkmark	
LPDDR3	8			\checkmark	dedicated Logic Layer
LPDDR4	16			\checkmark	Page 263 of 25

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4. Need for Lower Access Latency: Performance SAFARI

- New DRAM types often increase access latency in order to provide more banks, higher throughput
- Many applications can't make up for the increased latency
 - Especially true of common OS routines (e.g., file I/O, process forking)



• A variety of desktop/scientific, server/cloud, GPGPU applications

Several applications don't benefit from more parallelism

- 1. DRAM latency remains a critical bottleneck for many applications
- 2. Bank parallelism is not fully utilized by a wide variety of applications
- 3. Spatial locality continues to provide significant performance benefits if it is exploited by the memory subsystem
- 4. For some classes of applications, low-power memory can provide energy savings without sacrificing significant performance

Conclusion

- Manufacturers are developing many new types of DRAM
 - **DRAM limits performance, energy improvements:** new types may overcome some limitations
 - Memory systems now serve a very diverse set of applications: can no longer take a one-size-fits-all approach
 - Difficult to intuitively determine which DRAM-workload pair works best
- We perform a wide-ranging experimental study to uncover the combined behavior of workloads, DRAM types
 - 115 prevalent/emerging applications and multiprogrammed workloads
 - 9 modern DRAM types

12 key observations on DRAM-workload behavior
 Open-source tools: <u>https://github.com/CMU-SAFARI/ramulator</u>
 Full paper: <u>https://arxiv.org/pdf/1902.07609</u>

The Memory Latency Problem

- High memory latency is a significant limiter of system performance and energy-efficiency
- It is becoming increasingly so with higher memory contention in multi-core and heterogeneous architectures
 - Exacerbating the bandwidth need
 - Exacerbating the QoS problem
- It increases processor design complexity due to the mechanisms incorporated to tolerate memory latency

Retrospective: Conventional Latency Tolerance Techniques

- Caching [initially by Wilkes, 1965]
 - Widely used, simple, effective, but inefficient, passive
 - Not all applications/phases exhibit temporal or spatial locality

Prefetching [initially in IBM 360/91 1967]

None of These Fundamentally Reduce Memory Latency

ongoing research effort

- Out-of-order execution [initially by Tomasulo, 1967]
 - Tolerates cache misses that cannot be prefetched
 - Requires extensive hardware resources for tolerating long latencies



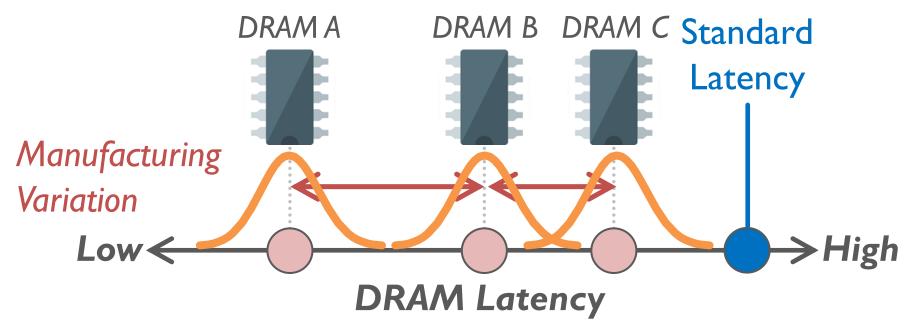
Two Major Sources of Latency Inefficiency

- Modern DRAM is **not** designed for low latency
 Main focus is cost-per-bit (capacity)
- Modern DRAM latency is determined by worst case conditions and worst case devices
 - Much of memory latency is unnecessary

Our Goal: Reduce Memory Latency at the Source of the Problem

Why is Memory Latency High?

- DRAM latency: Delay as specified in DRAM standards
 - Doesn't reflect true DRAM device latency
- Imperfect manufacturing process \rightarrow latency variation
- High standard latency chosen to increase yield

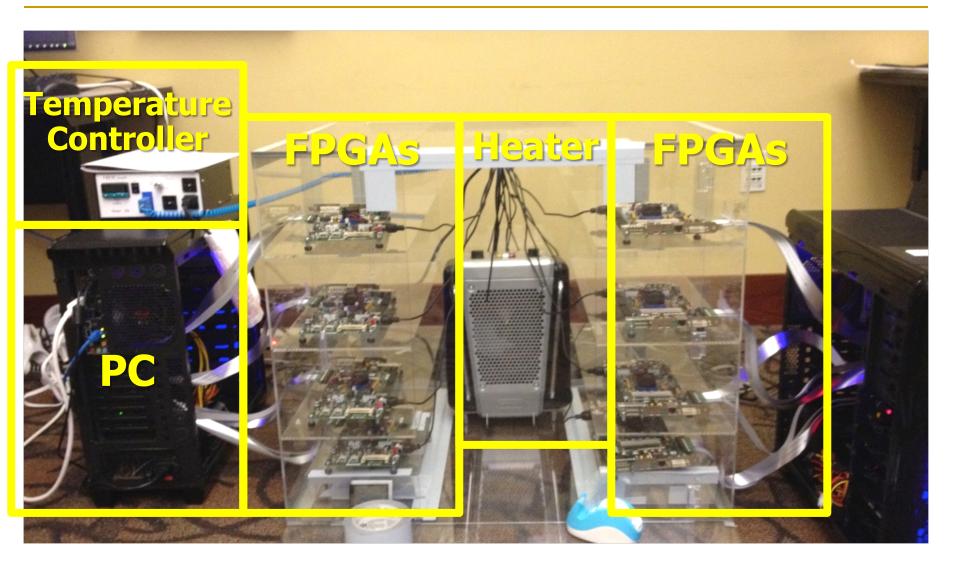


Adaptive-Latency DRAM

- 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

SAFARI Lee+, "Adaptive-Latency DRAM: Optimizing DRAM Timing for the Common-Case," HPCA 271 2015.

Infrastructures to Understand Such Issues



SAFARI

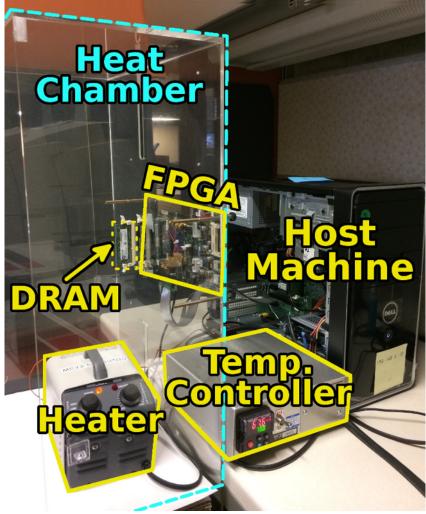
Kim+, "Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors," ISCA 2014.

SoftMC: Open Source DRAM Infrastructure

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

- Flexible
- Easy to Use (C++ API)
- Open-source

github.com/CMU-SAFARI/SoftMC





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

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

Hasan Hassan^{1,2,3} Nandita Vijaykumar³ Samira Khan^{4,3} Saugata Ghose³ Kevin Chang³ Gennady Pekhimenko^{5,3} Donghyuk Lee^{6,3} Oguz Ergin² Onur Mutlu^{1,3}

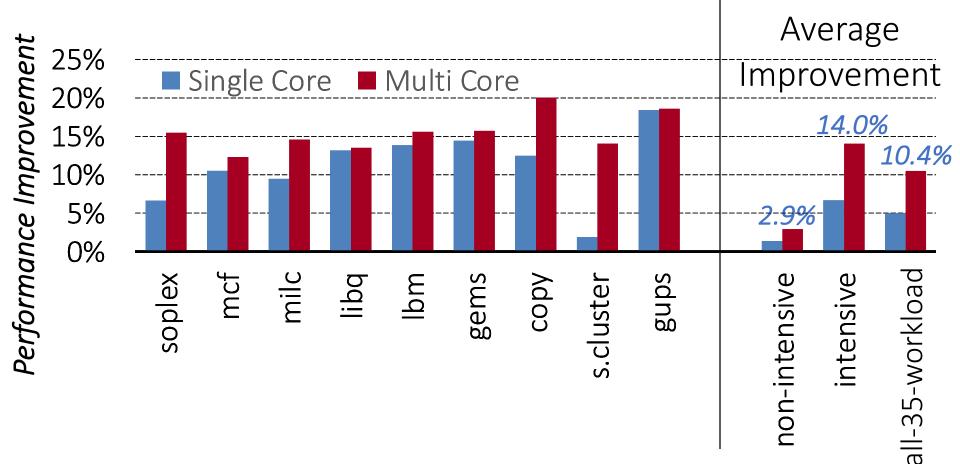
¹ETH Zürich ²TOBB University of Economics & Technology ³Carnegie Mellon University ⁴University of Virginia ⁵Microsoft Research ⁶NVIDIA Research

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%**

SAFARI Lee+, "Adaptive-Latency DRAM: Optimizing DRAM Timing for the Common-Case," HPCA 275 2015.

AL-DRAM: Real-System Performance



AL-DRAM provides high performance on memory-intensive workloads

Reducing Latency Also Reduces Energy

- AL-DRAM reduces DRAM power consumption
- 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,
 <u>"Adaptive-Latency DRAM: Optimizing DRAM Timing for</u> <u>the Common-Case"</u>
 Proceedings of the 21st International Symposium on High-Performance Computer Architecture (HPCA), Bay Area, CA, February 2015.
 [Slides (pptx) (pdf)] [Full data sets]

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

Donghyuk LeeYoongu KimGennady PekhimenkoSamira KhanVivek SeshadriKevin ChangOnur Mutlu

Carnegie Mellon University

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]
 - DRAM Latency PUF [HPCA 2018]
 - DRAM Latency True Random Number Generator [HPCA 2019]

•••

 We would like to find sources of latency heterogeneity and exploit them to minimize latency (or create other benefits)
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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>

<u>Modeling of Computer Systems</u> (**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

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 <u>ACM International Conference on Measurement and</u> <u>Modeling of Computer Systems</u> (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

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Solar-DRAM: Exploiting Spatial Variation

 Jeremie S. Kim, Minesh Patel, Hasan Hassan, and Onur Mutlu, "Solar-DRAM: Reducing DRAM Access Latency by Exploiting the Variation in Local Bitlines" Proceedings of the <u>36th IEEE International Conference on Computer</u> Design (ICCD), Orlando, FL, USA, October 2018.

Solar-DRAM: Reducing DRAM Access Latency by Exploiting the Variation in Local Bitlines

Jeremie S. Kim^{‡§} Minesh Patel[§] Hasan Hassan[§] Onur Mutlu^{§‡} [‡]Carnegie Mellon University [§]ETH Zürich

DRAM Latency PUFs

 Jeremie S. Kim, Minesh Patel, Hasan Hassan, and <u>Onur Mutlu</u>, <u>"The DRAM Latency PUF: Quickly Evaluating Physical Unclonable</u> <u>Functions by Exploiting the Latency-Reliability Tradeoff in</u> <u>Modern DRAM Devices</u>"

Proceedings of the <u>24th International Symposium on High-Performance</u> <u>Computer Architecture</u> (**HPCA**), Vienna, Austria, February 2018. [Lightning Talk Video] [Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)]

The DRAM Latency PUF:

Quickly Evaluating Physical Unclonable Functions by Exploiting the Latency-Reliability Tradeoff in Modern Commodity DRAM Devices

> Jeremie S. Kim^{†§} Minesh Patel[§] Hasan Hassan[§] Onur Mutlu^{§†} [†]Carnegie Mellon University [§]ETH Zürich

DRAM Latency True Random Number Generator

 Jeremie S. Kim, Minesh Patel, Hasan Hassan, Lois Orosa, and Onur Mutlu, "D-RaNGe: Using Commodity DRAM Devices to Generate True Random Numbers with Low Latency and High Throughput" Proceedings of the <u>25th International Symposium on High-Performance</u> <u>Computer Architecture</u> (HPCA), Washington, DC, USA, February 2019.

D-RaNGe: Using Commodity DRAM Devices to Generate True Random Numbers with Low Latency and High Throughput

Jeremie S. Kim^{‡§} Minesh Patel[§] Hasan Hassan[§] Lois Orosa[§] Onur Mutlu^{§‡} [‡]Carnegie Mellon University [§]ETH Zürich

ChargeCache: Exploiting Access Patterns

 Hasan Hassan, Gennady Pekhimenko, Nandita Vijaykumar, Vivek Seshadri, Donghyuk Lee, Oguz Ergin, and Onur Mutlu, "ChargeCache: Reducing DRAM Latency by Exploiting Row Access Locality"

Proceedings of the <u>22nd International Symposium on High-</u>

Performance Computer Architecture (HPCA), Barcelona, Spain, March 2016. [Slides (pptx) (pdf)]

[Source Code]

ChargeCache: Reducing DRAM Latency by Exploiting Row Access Locality

Hasan Hassan[†]*, Gennady Pekhimenko[†], Nandita Vijaykumar[†] Vivek Seshadri[†], Donghyuk Lee[†], Oguz Ergin^{*}, Onur Mutlu[†]

[†]*Carnegie Mellon University*

* TOBB University of Economics & Technology

Exploiting Subarray Level Parallelism

 Yoongu Kim, Vivek Seshadri, Donghyuk Lee, Jamie Liu, and Onur Mutlu,
 <u>"A Case for Exploiting Subarray-Level Parallelism</u>

(SALP) in DRAM"

Proceedings of the <u>39th International Symposium on</u> <u>Computer Architecture</u> (**ISCA**), Portland, OR, June 2012. <u>Slides (pptx)</u>

A Case for Exploiting Subarray-Level Parallelism (SALP) in DRAM

Yoongu Kim

Vivek Seshadri Donghyuk Lee

Jamie Liu Onur Mutlu

Carnegie Mellon University

Tiered-Latency DRAM

 Donghyuk Lee, Yoongu Kim, Vivek Seshadri, Jamie Liu, Lavanya Subramanian, and Onur Mutlu,
 "Tiered-Latency DRAM: A Low Latency and Low Cost
 DRAM Architecture"
 Proceedings of the 19th International Symposium on High-Performance Computer Architecture (HPCA), Shenzhen, China, February 2013. Slides (pptx)

Tiered-Latency DRAM: A Low Latency and Low Cost DRAM Architecture

Donghyuk Lee Yoongu Kim Vivek Seshadri Jamie Liu Lavanya Subramanian Onur Mutlu Carnegie Mellon University

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LISA: Low-cost Inter-linked Subarrays

 Kevin K. Chang, Prashant J. Nair, Saugata Ghose, Donghyuk Lee, Moinuddin K. Qureshi, and Onur Mutlu, "Low-Cost Inter-Linked Subarrays (LISA): Enabling Fast Inter-Subarray Data Movement in DRAM" Proceedings of the <u>22nd International Symposium on High-</u> Performance Computer Architecture (HPCA), Barcelona, Spain, March 2016.
 [Slides (pptx) (pdf)] [Source Code]

Low-Cost Inter-Linked Subarrays (LISA): Enabling Fast Inter-Subarray Data Movement in DRAM

Kevin K. Chang[†], Prashant J. Nair^{*}, Donghyuk Lee[†], Saugata Ghose[†], Moinuddin K. Qureshi^{*}, and Onur Mutlu[†] [†]Carnegie Mellon University ^{*}Georgia Institute of Technology

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The CROW Substrate for DRAM

 Hasan Hassan, Minesh Patel, Jeremie S. Kim, A. Giray Yaglikci, Nandita Vijaykumar, Nika Mansourighiasi, Saugata Ghose, and Onur Mutlu,
 "CROW: A Low-Cost Substrate for Improving DRAM Performance, Energy Efficiency, and Reliability" Proceedings of the <u>46th International Symposium on Computer</u> <u>Architecture</u> (ISCA), Phoenix, AZ, USA, June 2019.

CROW: A Low-Cost Substrate for Improving DRAM Performance, Energy Efficiency, and Reliability

Hasan Hassan[†]Minesh Patel[†]Jeremie S. Kim^{†§}A. Giray Yaglikci[†]Nandita Vijaykumar^{†§}Nika Mansouri Ghiasi[†]Saugata Ghose[§]Onur Mutlu^{†§}

[†]*ETH Zürich* [§]*Carnegie Mellon University*

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CROW: The Copy Row Substrate [ISCA 2019]

Challenges of DRAM Scaling



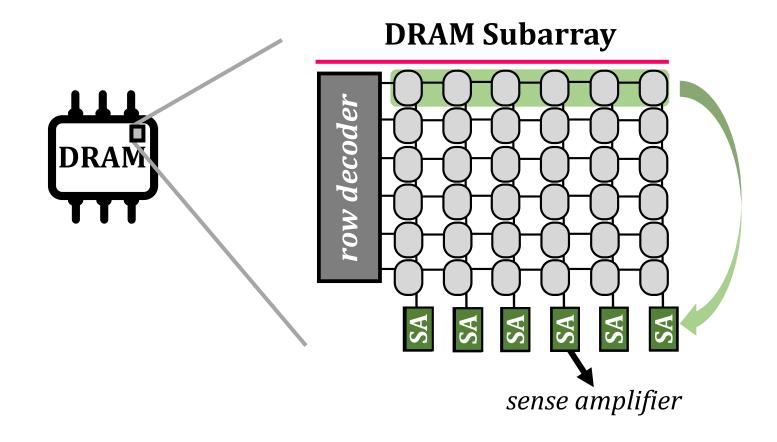


3 exposure to vulnerabilities



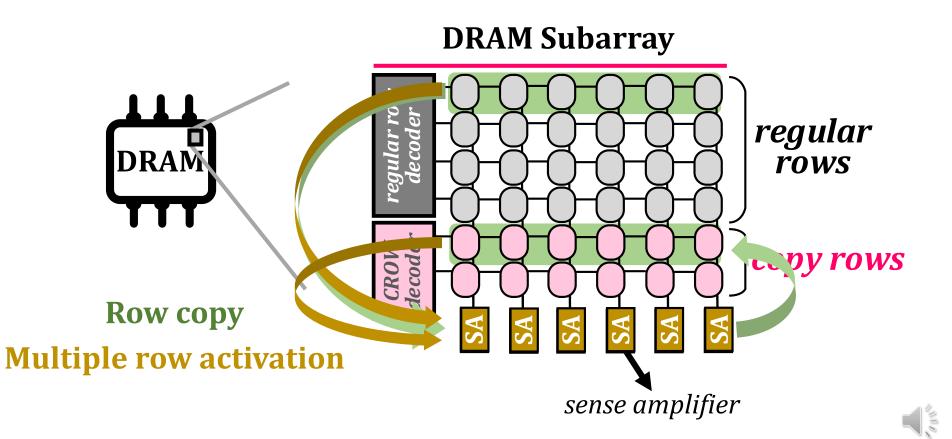
DRAM

Conventional DRAM





Copy Row DRAM (CROW)





Use Cases of CROW

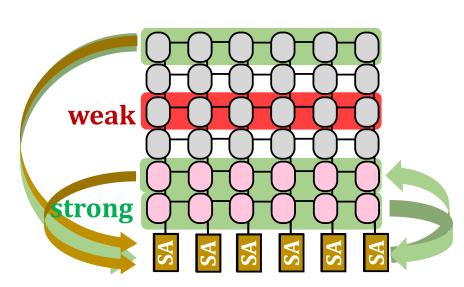
➤ CROW-cache ✓ reduces access latency

≻CROW-ref

✓ reduces DRAM *refresh* overhead

➢A mechanism for protecting against *RowHammer*





Key Results

CROW-cache + CROW-ref

- •20% speedup
- 22% less DRAM energy

Hardware Overhead

- •0.5% DRAM chip area
- •1.6% DRAM capacity
- •11.3 KiB memory controller storage



More on CROW

 Hasan Hassan, Minesh Patel, Jeremie S. Kim, A. Giray Yaglikci, Nandita Vijaykumar, Nika Mansourighiasi, Saugata Ghose, and Onur Mutlu,

"CROW: A Low-Cost Substrate for Improving DRAM Performance, Energy Efficiency, and Reliability"

Proceedings of the <u>46th International Symposium on Computer Architecture</u> (**ISCA**), Phoenix, AZ, USA, June 2019. [<u>Slides (pptx) (pdf)</u>] [<u>Lightning Talk Slides (pptx) (pdf)</u>] [<u>Poster (pptx) (pdf)</u>] [<u>Lightning Talk Video</u> (3 minutes)] [<u>Full Talk Video</u> (16 minutes)] [<u>Source Code for CROW</u> (Ramulator and Circuit Modeling)]

CROW: A Low-Cost Substrate for Improving DRAM Performance, Energy Efficiency, and Reliability

Hasan Hassan[†]Minesh Patel[†]Jeremie S. Kim^{†§}A. Giray Yaglikci[†]Nandita Vijaykumar^{†§}Nika Mansouri Ghiasi[†]Saugata Ghose[§]Onur Mutlu^{†§}

[†]*ETH Zürich* [§]*Carnegie Mellon University*

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CLR-DRAM: Capacity-Latency Reconfigurability

Haocong Luo, Taha Shahroodi, Hasan Hassan, Minesh Patel, A. Giray Yaglikci, Lois Orosa, Jisung Park, and Onur Mutlu,
 "CLR-DRAM: A Low-Cost DRAM Architecture Enabling
 Dynamic Capacity-Latency Trade-Off"
 Proceedings of the <u>47th International Symposium on Computer</u>
 <u>Architecture</u> (ISCA), Valencia, Spain, June 2020.
 [Slides (pptx) (pdf)]
 [Lightning Talk Slides (pptx) (pdf)]
 [Lightning Talk Video (3 minutes)]

CLR-DRAM: A Low-Cost DRAM Architecture Enabling Dynamic Capacity-Latency Trade-Off

Haocong Luo^{§†} Taha Shahroodi[§] Hasan Hassan[§] Minesh Patel[§] A. Giray Yağlıkçı[§] Lois Orosa[§] Jisung Park[§] Onur Mutlu[§] [§]ETH Zürich [†]ShanghaiTech University

CLR-DRAM: Capacity-Latency Reconfigurable DRAM [ISCA 2020]

CLR-DRAM: A Low-Cost DRAM Architecture Enabling Dynamic Capacity-Latency Trade-off

Haocong Luo Taha Shahroodi Hasan Hassan Minesh Patel A. Giray Yaglıkçı Lois Orosa Jisung Park Onur Mutlu

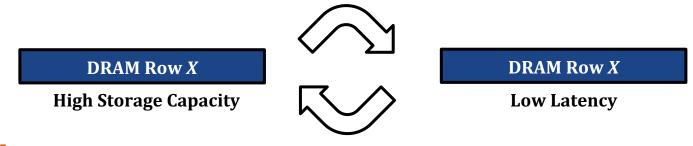






Motivation & Goal

- Workloads and systems have varying main memory capacity and latency demands.
- Existing commodity DRAM makes static capacity-latency trade-off at design time.
- Systems miss opportunities to improve performance by adapting to changes in main memory capacity and latency demands.
- <u>Goal</u>: Design a low-cost DRAM architecture that can be dynamically configured to have high capacity or low latency at a fine granularity (i.e., at the granularity of a row).



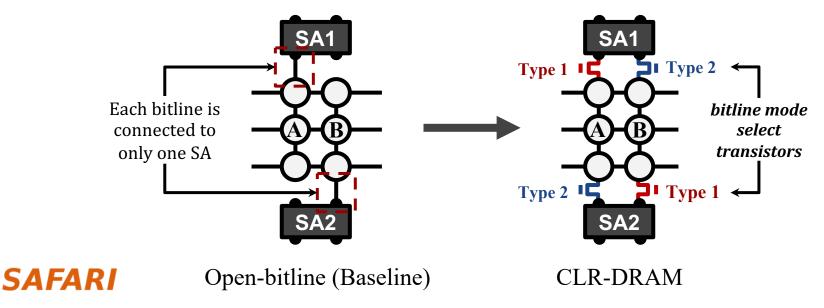
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CLR-DRAM (<u>Capacity-Latency-Reconfigurable DRAM</u>)

- CLR-DRAM (Capacity-Latency-Reconfigurable DRAM):
 - A low cost DRAM architecture that enables a single DRAM row to *dynamically* switch between **max-capacity mode** or **high-performance mode**.

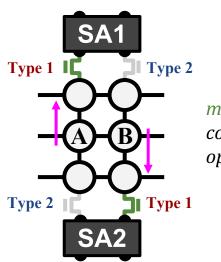
• Key Idea:

Dynamically configure the connections between DRAM cells and sense amplifiers in the density-optimized open-bitline architecture.

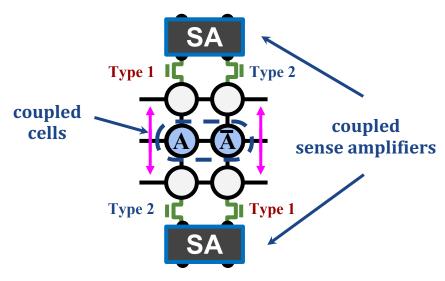


CLR-DRAM (<u>Capacity-Latency-Reconfigurable DRAM</u>)

• Max-capacity mode



mimics the cell-to-SA connections as in the open-bitline architecture • High-performance mode



The same storage capacity as the conventional open-bitline architecture

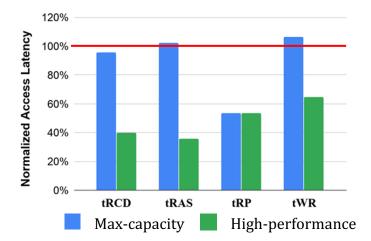
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Reduced latency and refresh overhead via coupled cell/SA operation

Key Results

• DRAM Latency Reduction:

- Activation latency (tRCD) by 60.1%
- Restoration latency (tRAS) by 64.2%
- Precharge latency (**tRP**) by **46.4%**
- Write-recovery latency (tWR) by 35.2%



<u>System-level Benefits</u>:

- Performance improvement: **18.6%**
- DRAM energy reduction: **29.7%**
- DRAM refresh energy reduction: **66.1%**

We hope that CLR-DRAM can be exploited to develop more flexible systems that can adapt to the diverse and changing DRAM capacity and latency demands of workloads.

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More on CLR-DRAM

Haocong Luo, Taha Shahroodi, Hasan Hassan, Minesh Patel, A. Giray Yaglikci, Lois Orosa, Jisung Park, and Onur Mutlu,
 "CLR-DRAM: A Low-Cost DRAM Architecture Enabling
 Dynamic Capacity-Latency Trade-Off"
 Proceedings of the <u>47th International Symposium on Computer</u>
 <u>Architecture</u> (ISCA), Valencia, Spain, June 2020.
 [Slides (pptx) (pdf)]
 [Lightning Talk Slides (pptx) (pdf)]
 [Lightning Talk Video (3 minutes)]

CLR-DRAM: A Low-Cost DRAM Architecture Enabling Dynamic Capacity-Latency Trade-Off

Haocong Luo^{§†} Taha Shahroodi[§] Hasan Hassan[§] Minesh Patel[§] A. Giray Yağlıkçı[§] Lois Orosa[§] Jisung Park[§] Onur Mutlu[§] [§]ETH Zürich [†]ShanghaiTech University

Reducing Refresh Latency

 Anup Das, Hasan Hassan, and Onur Mutlu, "VRL-DRAM: Improving DRAM Performance via Variable Refresh Latency" Proceedings of the <u>55th Design Automation</u> <u>Conference</u> (DAC), San Francisco, CA, USA, June 2018.

VRL-DRAM: Improving DRAM Performance via Variable Refresh Latency

Anup Das Drexel University Philadelphia, PA, USA anup.das@drexel.edu Hasan Hassan ETH Zürich Zürich, Switzerland hhasan@ethz.ch Onur Mutlu ETH Zürich Zürich, Switzerland omutlu@gmail.com

Parallelizing Refreshes and Accesses

 Kevin Chang, Donghyuk Lee, Zeshan Chishti, Alaa Alameldeen, Chris Wilkerson, Yoongu Kim, and Onur Mutlu,
 "Improving DRAM Performance by Parallelizing Refreshes with Accesses"
 Proceedings of the <u>20th International Symposium on High-Performance</u> Computer Architecture (HPCA), Orlando, FL, February 2014.

[Summary] [Slides (pptx) (pdf)]

Reducing Performance Impact of DRAM Refresh by Parallelizing Refreshes with Accesses

Kevin Kai-Wei Chang Donghyuk Lee Zeshan Chishti† Alaa R. Alameldeen† Chris Wilkerson† Yoongu Kim Onur Mutlu Carnegie Mellon University †Intel Labs

Eliminating Refreshes

 Jamie Liu, Ben Jaiyen, Richard Veras, and Onur Mutlu, "RAIDR: Retention-Aware Intelligent DRAM Refresh" Proceedings of the <u>39th International Symposium on</u> <u>Computer Architecture</u> (ISCA), Portland, OR, June 2012. <u>Slides (pdf)</u>

RAIDR: Retention-Aware Intelligent DRAM Refresh

Jamie Liu Ben Jaiyen Richard Veras Onur Mutlu Carnegie Mellon University

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 Onur Mutlu,
 <u>"Understanding Reduced-Voltage Operation in Modern DRAM</u> <u>Devices: Experimental Characterization, Analysis, and</u> <u>Mechanisms"</u> *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

VAMPIRE DRAM Power Model

 Saugata Ghose, A. Giray Yaglikci, Raghav Gupta, Donghyuk Lee, Kais Kudrolli, William X. Liu, Hasan Hassan, Kevin K. Chang, Niladrish Chatterjee, Aditya Agrawal, Mike O'Connor, and Onur Mutlu,

"What Your DRAM Power Models Are Not Telling You: Lessons from a Detailed Experimental Study"

Proceedings of the <u>ACM International Conference on Measurement and Modeling of</u> Computer Systems (**SIGMETRICS**), Irvine, CA, USA, June 2018.

[Abstract]

[POMACS Journal Version (same content, different format)]

Slides (pptx) (pdf)

[VAMPIRE DRAM Power Model]

What Your DRAM Power Models Are Not Telling You: Lessons from a Detailed Experimental Study

Saugata Ghose[†] Abdullah Giray Yağlıkçı^{‡†} Raghav Gupta[†] Donghyuk Lee[§]
 Kais Kudrolli[†] William X. Liu[†] Hasan Hassan[‡] Kevin K. Chang[†]
 Niladrish Chatterjee[§] Aditya Agrawal[§] Mike O'Connor^{§¶} Onur Mutlu^{‡†}
 [†]Carnegie Mellon University [‡]ETH Zürich [§]NVIDIA [¶]University of Texas at Austin



We Can Reduce Memory Latency with Change of Mindset





Main Memory Needs Intelligent Controllers to Reduce Latency

