SAFARI Research Group Introduction & Research

Onur Mutlu

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

19 October 2021

EFCL Huawei Day





Carnegie Mellon

Brief Self Introduction



Onur Mutlu

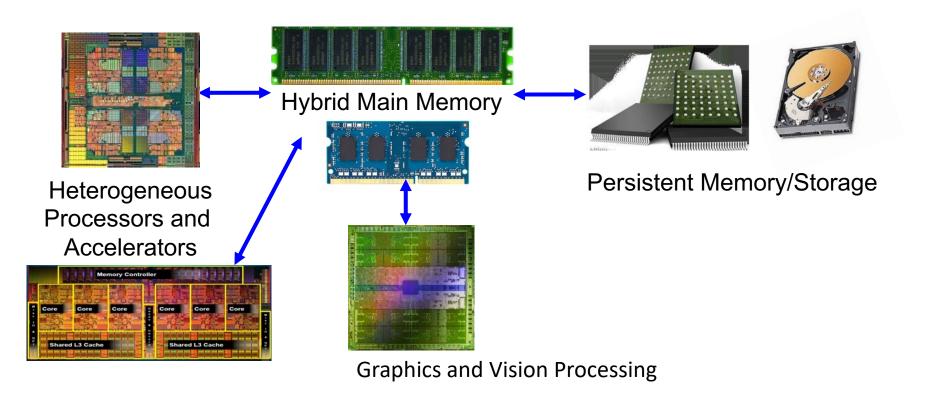
- Full Professor @ ETH Zurich ITET (INFK), since September 2015
- Strecker Professor @ Carnegie Mellon University ECE/CS, 2009-2016, 2016-...
- PhD from UT-Austin, worked at Google, VMware, Microsoft Research, Intel, AMD
- https://people.inf.ethz.ch/omutlu/
- omutlu@gmail.com (Best way to reach me)
-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

Research and Teaching in:

- Computer architecture, computer systems, hardware security, bioinformatics
- Memory and storage systems
- Hardware security, safety, predictability
- Fault tolerance
- Hardware/software cooperation
- Architectures for bioinformatics, health, medicine
- **...**

Current Research Mission

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



Build fundamentally better architectures

Four Key Current Directions

Fundamentally Secure/Reliable/Safe Architectures

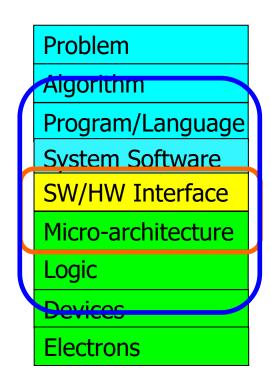
- Fundamentally Energy-Efficient Architectures
 - Memory-centric (Data-centric) Architectures

Fundamentally Low-Latency and Predictable Architectures

Architectures for AI/ML, Genomics, Medicine, Health, ...

The Transformation Hierarchy

Computer Architecture (expanded view)



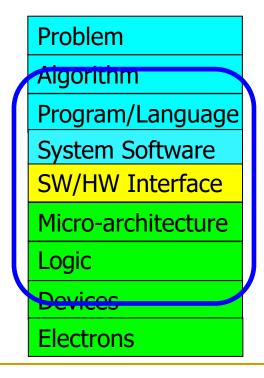
Computer Architecture (narrow view)

Axiom

To achieve the highest energy efficiency and performance:

we must take the expanded view

of computer architecture

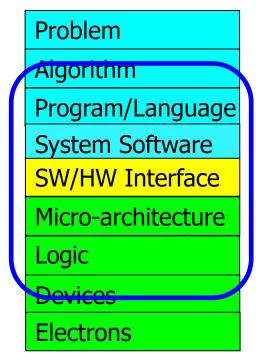


Co-design across the hierarchy:
Algorithms to devices

Specialize as much as possible within the design goals

Current Research Mission & Major Topics

Build fundamentally better architectures



Broad research spanning apps, systems, logic with architecture at the center

- Data-centric arch. for low energy & high perf.
 - Proc. in Mem/DRAM, NVM, unified mem/storage
- Low-latency & predictable architectures
 - Low-latency, low-energy yet low-cost memory
 - QoS-aware and predictable memory systems
- Fundamentally secure/reliable/safe arch.
 - Tolerating all bit flips; patchable HW; secure mem
- Architectures for ML/AI/Genomics/Health/Med
 - Algorithm/arch./logic co-design; full heterogeneity
- Data-driven and data-aware architectures
 - ML/AI-driven architectural controllers and design
 - Expressive memory and expressive systems

SAFARI Research Group



Think BIG, Aim HIGH!

https://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/



Think BIG, Aim HIGH!

SAFARI

https://safari.ethz.ch

SAFARI Newsletter January 2021 Edition

https://safari.ethz.ch/safari-newsletter-january-2021/





Newsletter January 2021

Think Big, Aim High, and Have a Wonderful 2021!



Dear SAFARI friends,

SAFARI PhD and Post-Doc Alumni

https://safari.ethz.ch/safari-alumni/

- Damla Senol Cali (Bionano Genomics)
- Nastaran Hajinazar (ETH Zurich)
- Gagandeep Singh (ETH Zurich)
- Amirali Boroumand (Stanford Univ → Google)
- Jeremie Kim (ETH Zurich)
- Nandita Vijaykumar (Univ. of Toronto, Assistant Professor)
- Kevin Hsieh (Microsoft Research, Senior Researcher)
- Justin Meza (Facebook)
- Mohammed Alser (ETH Zurich)
- Yixin Luo (Google)
- Kevin Chang (Facebook)
- Rachata Ausavarungnirun (KMUNTB, Assistant Professor)
- Gennady Pekhimenko (Univ. of Toronto, Assistant Professor)
- Vivek Seshadri (Microsoft Research)
- Donghyuk Lee (NVIDIA Research, Senior Researcher)
- Yoongu Kim (Software Robotics → Google)
- Lavanya Subramanian (Intel Labs → Facebook)
- Samira Khan (Univ. of Virginia, Assistant Professor)
- Saugata Ghose (Univ. of Illinois, Assistant Professor)
- Jawad Haj-Yahya (Huawei Research Zurich, Principal Researcher)

Principle: Teaching and Research

Teaching drives Research Research drives Teaching

. . .

Research & Teaching: Some Overview Talks

https://www.youtube.com/onurmutlulectures

- Future Computing Architectures
 - https://www.youtube.com/watch?v=kqiZISOcGFM&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=r7sn41lH-4A&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=41
- 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=c6_LgzuNdkw&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=25
- The Story of RowHammer
 - https://www.youtube.com/watch?v=sgd7PHQQ1AI&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=39

Online Courses & Lectures

First Computer Architecture & Digital Design Course

- Digital Design and Computer Architecture
- Spring 2021 Livestream Edition:
 https://www.youtube.com/watch?v=LbC0EZY8yw4&list=PL5Q2soXY2Zi_uej3aY39YB5pfW4SJ7LIN

Advanced Computer Architecture Course

- Computer Architecture
- □ Falll 2020 Edition:

https://www.youtube.com/watch?v=c3mPdZA-Fmc&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN

1:22:29

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VIDEOS

▶ PLAY ALL

1:33:25

PLAYLISTS

COMMUNITY

CHANNELS

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Q

Popular uploads





TTT TT 2:24:11 Computer Architecture -

Computer Architecture -

ML accelerator: 260 mm², 6 billion transist 600 GFLOPS GPU, 12 ARM 2.2 GHz CPUs.



Design of Digital Circuits -Lecture 1: Introduction and...

Design of Digital Circuits



Computer Architecture -Lecture 2: Fundamentals....

17K views • 3 years ago

Digital Design & Computer Architecture: Lecture 1:...

49K views • 1 year ago

Computer Architecture -Lecture 1: Introduction and...

36K views • 3 years ago

Lecture 1: Introduction and...

31K views • 1 year ago

Includes standard DIMM modu number of DPU processors co

30K views • 8 months ago

Lecture 1: Introduction and...

22K views • 2 years ago

First Course in Computer Architecture & Digital Design 2021-2013













Livestream - Digital Design and Digital Design & Computer Computer Architecture - ETH...

Onur Mutlu Lectures VIEW FULL PLAYLIST Architecture - ETH Zürich...

Onur Mutlu Lectures VIEW FULL PLAYLIST

Design of Digital Circuits - ETH Zürich - Spring 2019

Onur Mutlu Lectures VIEW FULL PLAYLIST

Design of Digital Circuits - ETH Zürich - Spring 2018

Onur Mutlu Lectures VIEW FULL PLAYLIST

Digital Circuits and Computer Architecture - ETH Zurich -...

Onur Mutlu Lectures VIEW FULL PLAYLIST

Spring 2015 -- Computer Architecture Lectures --...

Carnegie Mellon Computer Architec... VIEW FULL PLAYLIST

Advanced Computer Architecture Courses 2020-2012















Computer Architecture - ETH Zürich - Fall 2020

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Computer Architecture - ETH Zürich - Fall 2019

Onur Mutlu Lectures VIEW FULL PLAYLIST

Computer Architecture - ETH Zürich - Fall 2018

Onur Mutlu Lectures VIEW FULL PLAYLIST

Computer Architecture - ETH Zürich - Fall 2017

Onur Mutlu Lectures VIEW FULL PLAYLIST

Fall 2015 - 740 Computer Architecture

Carnegie Mellon Computer Architec... VIEW FULL PLAYLIST

Fall 2013 - 740 Computer Architecture - Carnegie Mellon Carnegie Mellon Computer Architec...

VIEW FULL PLAYLIST

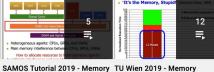
Special Courses on Memory Systems













Memory Technology Lectures Onur Mutlu Lectures

VIEW FULL PLAYLIST

Champéry Winter School 2020 - Perugia NiPS Summer School Memory Systems and Memory... 2019

Onur Mutlu Lectures VIEW FULL PLAYLIST

Onur Mutlu Lectures VIEW FULL PLAYLIST

Systems Onur Mutlu Lectures

VIEW FULL PLAYLIST

Systems and Memory-Centric...

Onur Mutlu Lectures VIEW FULL PLAYLIST

ACACES 2018 Lectures --Memory Systems and Memory...

Onur Mutlu Lectures VIEW FULL PLAYLIST



DDCA (Spring 2021)

- https://safari.ethz.ch/digitaltechnik/ spring2021/doku.php?id=schedule
- https://www.youtube.com/watch?v =LbC0EZY8yw4&list=PL5Q2soXY2Zi uej3aY39YB5pfW4SJ7LIN
- Bachelor's course
 - 2nd semester at ETH Zurich
 - Rigorous introduction into "How Computers Work"
 - Digital Design/Logic
 - Computer Architecture
 - 10 FPGA Lab Assignments



Recent Changes Media Manager Sitemap

schedule

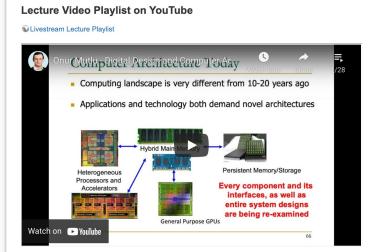
Trace: - schedule

Announcements

- Lectures/Schedule
- Lecture Buzzwords
- Readings Ontional HWs
- Extra Assignments
- Technical Docs

Exams

- Secondary Computer Architecture (CMU)
- SS15: Lecture Videos Computer Architecture (CMU)
- SS15: Course Website
- Spigitaltechnik SS18: Lecture Spigitaltechnik SS18: Course
- Website Specified in the second of the
- Digitaltechnik SS19: Course
- Website Digitaltechnik SS20: Lecture
- Videos Spigitaltechnik SS20: Course
- Website
- Moodle Moodle



Recorded Lecture Playlist



Spring 2021 Lectures/Schedule

Week	Date	Livestream	Lecture	Readings	Lab	HW	
W1	25.02 Thu.	You Tube Live	L1: Introduction and Basics	Required Suggested Mentioned			
	26.02 Fri.	You Tube Live	L2a: Tradeoffs, Metrics, Mindset	Required			
			L2b: Mysteries in Computer Architecture (PDF) (PPT)	Required Mentioned			
W2	04.03 Thu.	You Tube Live	L3a: Mysteries in Computer Architecture II	Required Suggested			



https://www.youtube.com/watch?v=c3 mPdZA-Fmc&list=PL5Q2soXY2Zi9xidyIqBxUz7x **RPS-wisBN**

- Master's level course
 - Taken by Bachelor's/Masters/PhD students
 - Cutting-edge research topics + fundamentals in Computer Architecture
 - 5 Simulator-based Lab Assignments
 - Potential research exploration
 - Many research readings



Q Recent Changes Media Manager Sitemap

schedule

Trace: · start · schedule

Announcements

Materials

- Lectures/Schedule
- Lecture Buzzwords

- Exams Related Courses

- Computer Architecture FS19 Course Webpage
- Computer Architecture FS19:
- Lecture Videos Digitaltechnik SS20: Course
- Webpage Digitaltechnik SS20: Lecture Videos
- Moodle Moodle
- Piazza (Q&A)
- **S** HotCRP
- Verilog Practice Website

Lecture Video Playlist on YouTube



Fall 2020 Lectures & Schedule

Week	Date	Lecture	Readings	Lab	HW
W1	17.09 Thu.	L1: Introduction and Basics (PDF) (PPT) You Video	Described Suggested		HW 0
	18.09 Fri.	L2a: Memory Performance Attacks (PDF) (PPT) Voulton Video	Described Suggested	Lab 1 Out	
		L2b: Data Retention and Memory Refresh (PDF) (PPT) Vou Video	Described Suggested		
		L2c: Course Logistics (PDF) (PPT) You the Video			
W2	24.09 Thu.	L3a: Introduction to Genome Sequence Analysis (PDF) (PPT) (Volume Video	Described Suggested		HW 1 Out
		L3b: Memory Systems: Challenges and Opportunities (PDF) (PPT) Vou Video	Described Suggested		
	25.09 Fri.	L4a: Memory Systems: Solution Directions (PDF) (PPT) You Video	Described Suggested		
		L4b: RowHammer (PDF) (PPT) Vou Video	Described Suggested		
W3	01.10 Thu.	L5a: RowHammer in 2020: TRRespass (PDF) (PPT) Vou Video	Described Suggested		
		L5b: RowHammer in 2020: Revisiting RowHammer (PDF) (PPT) Vou Video	Described Suggested		
		L5c: Secure and Reliable Memory	Described		

Comp Arch (Current)

https://safari.ethz.ch/architecture/fall20 21/doku.php?id=schedule

Youtube Livestream:

https://www.youtube.com/watch?v=4yfk M 5EFgo&list=PL5Q2soXY2Zi-Mnk1PxjEIG32HAGILkTOF

Master's level course

- Taken by Bachelor's/Masters/PhD students
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- 5 Simulator-based Lab Assignments
- Potential research exploration
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schedule

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Trace: · readings · start · schedule

Home

Announcements

Materials

- Lectures/Schedule
- Lecture Buzzwords
- Readings
- HWs
- LabsExams
 - Related Courses
- Tutorials

Pecaurees

- Course Webpage
- Computer Architecture FS20:
- Digitaltechnik SS21: Course
- Digitaltechnik SS21: Lecture Videos
- Moodle
- MotCRP
- S Verilog Practice Website (HDLBits)

Lecture Video Playlist on YouTube



Recorded Lecture Playlist

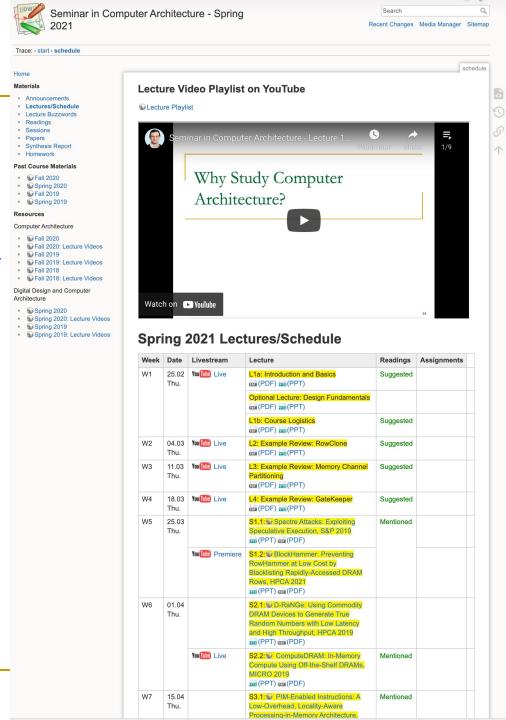


Fall 2021 Lectures & Schedule

Week	Date	Livestream	Lecture	Readings	Lab	HW
W1	30.09 Thu.	You Tube Live	L1: Introduction and Basics	Required Mentioned	Lab 1 Out	HW 0 Out
	01.10 Fri.	You Tube Live	L2: Trends, Tradeoffs and Design Fundamentals (PDF) (PPT)	Required Mentioned		
W2	07.10 Thu.	You Tube Live	L3a: Memory Systems: Challenges and Opportunities	Described Suggested		HW 1 Out
			L3b: Course Info & Logistics			
			L3c: Memory Performance Attacks	Described Suggested		
	08.10 Fri.	You Tube Live	L4a: Memory Performance Attacks (PDF) (PPT)	Described Suggested	Lab 2 Out	
			L4b: Data Retention and Memory Refresh (PDF) (PPT)	Described Suggested		
			L4c: RowHammer	Described Suggested		

Seminar (Spring'21)

- <u>https://safari.ethz.ch/architecture_seminar/spring2021/doku.php?id=schedule</u>
- https://www.youtube.com/watch?v=t3m 93ZpLOyw&list=PL5Q2soXY2Zi awYdjm WVIUegsbY7TPGW4
- Critical analysis course
 - Taken by Bachelor's/Masters/PhD students
 - Cutting-edge research topics + fundamentals in Computer Architecture
 - 20+ research papers, presentations, analyses



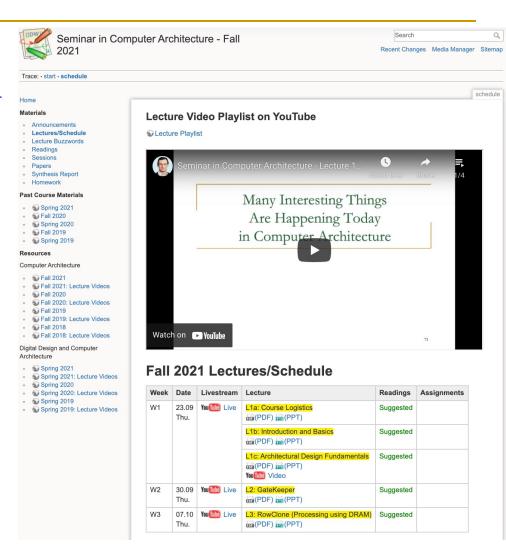


Seminar (Current)

https://safari.ethz.ch/architecture_semin ar/fall2021/doku.php?id=schedule

Youtube Livestream:

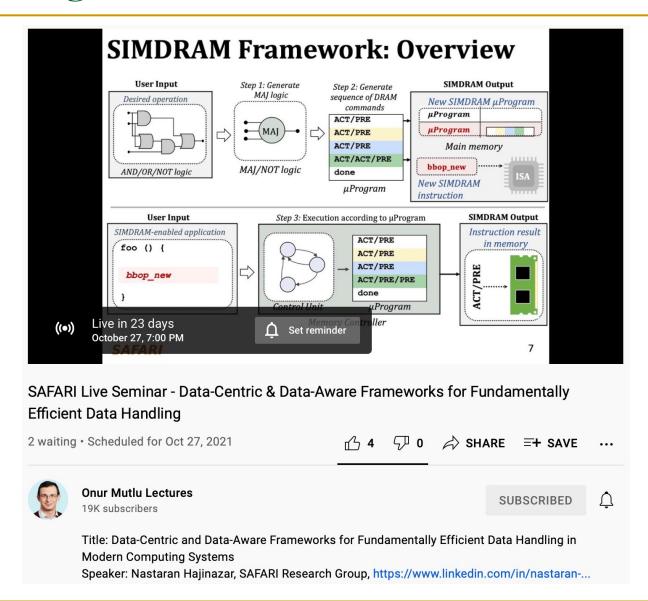
- https://www.youtube.com/watch?v=4TcP 297mdsI&list=PL5Q2soXY2Zi 7UBNmC9B 8Yr5JSwTG9yH4
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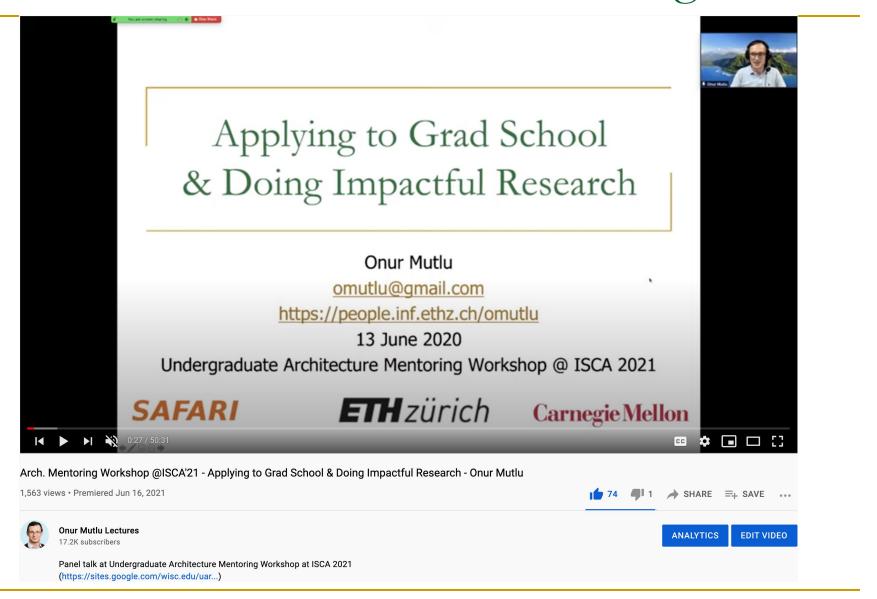
SAFARI Live Seminars



Upcoming SAFARI Live Seminar: Oct 27



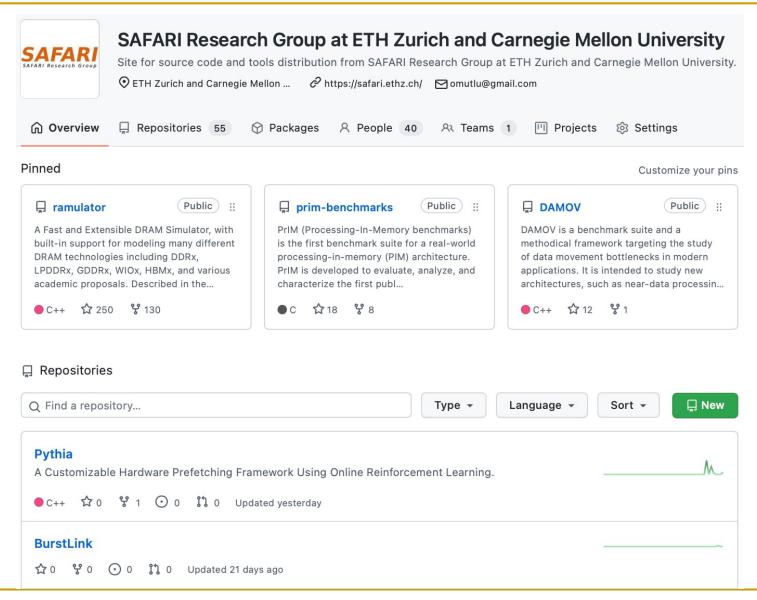
More on Our Research & Teaching

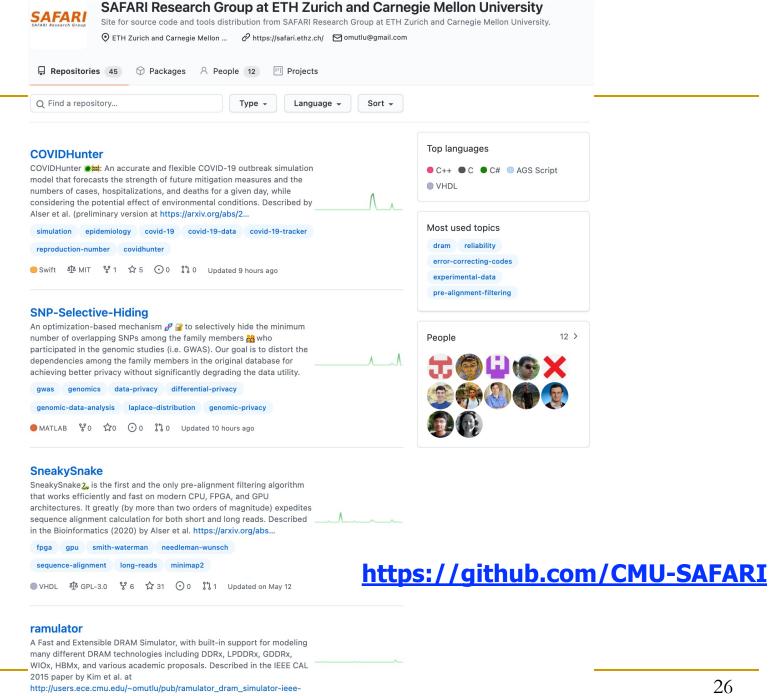


Open-Source Artifacts

https://github.com/CMU-SAFARI

Open Source Tools: SAFARI GitHub





SAFARI

Papers, Talks, Videos, Artifacts

All are available at

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

https://www.youtube.com/onurmutlulectures

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

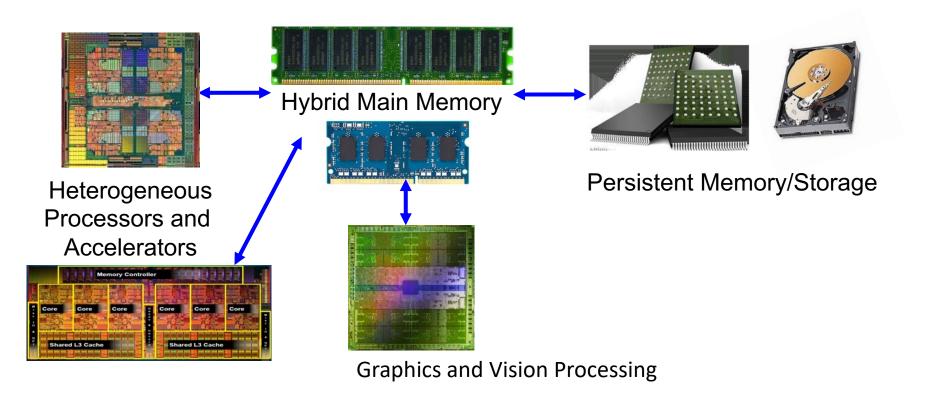
Funding Acknowledgments

- 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

Example Research Topics: Quick Overview

Current Research Mission

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



Build fundamentally better architectures

Four Key Issues in Future Platforms

Fundamentally Secure/Reliable/Safe Architectures

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

Fundamentally Low-Latency and Predictable Architectures

Architectures for AI/ML, Genomics, Medicine, Health

Architectures for Intelligent Machines

Data-centric

Data-driven

Data-aware

Current EFCL Projects

- "A New Methodology and Open-Source Benchmark Suite for Evaluating Data Movement Bottlenecks: A Processing-in-Memory Case Study"
 - Data-centric

- "Machine-Learning-Assisted Intelligent Microarchitectures to Reduce Memory Access Latency"
 - Data-driven

- "Cross-layer Hardware/Software Techniques to Enable Powerful Computation and Memory Optimizations"
 - Data-aware

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]



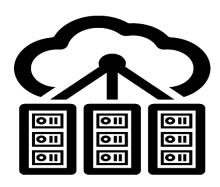
In-Memory Data Analytics

[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' 15]

Data is Key for Future Workloads



Chrome

Google's web browser



TensorFlow Mobile

Google's machine learning framework



Google's video codec



Google's video codec

Data Overwhelms Modern Machines





TensorFlow Mobile

Data → performance & energy bottleneck

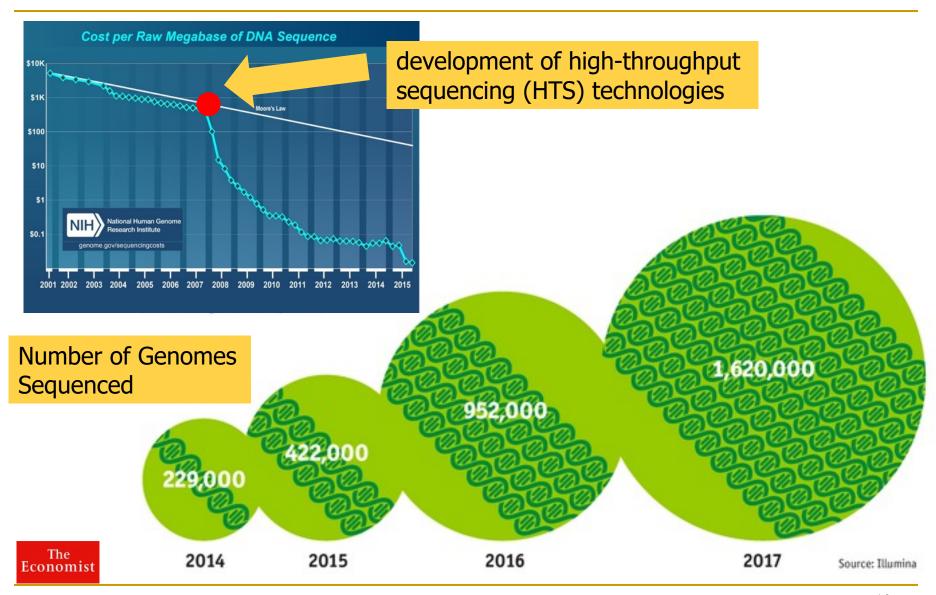
VP9
VouTube
Video Playback

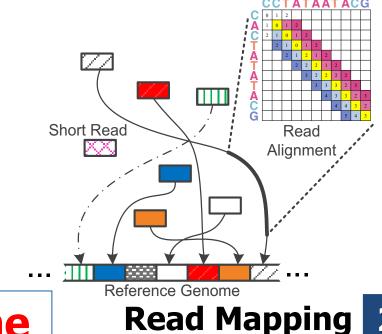
Google's video codec



Google's video codec

Data is Key for Future Workloads





Sequencing

Genome Analysis

Data → performance & energy bottleneck

reau4: CGCTTCCAT

read5: CCATGACGC read6: TTCCATGAC



Scientific Discovery

Variant Calling

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/bby017

Published: 02 April 2018 Article 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

Damla Senol Cali ™, Jeremie S Kim, Saugata Ghose, Can Alkan, Onur Mutlu

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 [IEEE MICRO 2020]

 Mohammed Alser, Zulal Bingol, Damla Senol Cali, Jeremie Kim, Saugata Ghose, Can Alkan, and Onur Mutlu,

"Accelerating Genome Analysis: A Primer on an Ongoing Journey"

IEEE Micro (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

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 Acceleration Framework for Genome Sequence Analysis"
Proceedings of the 53rd International Symposium on Microarchitecture (MICRO), Virtual, October 2020.

[<u>Lighting Talk Video</u> (1.5 minutes)]
[<u>Lightning Talk Slides (pptx) (pdf)</u>]
[<u>Talk Video</u> (18 minutes)]
[<u>Slides (pptx) (pdf)</u>]

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

45

FPGA-based Processing Near Memory

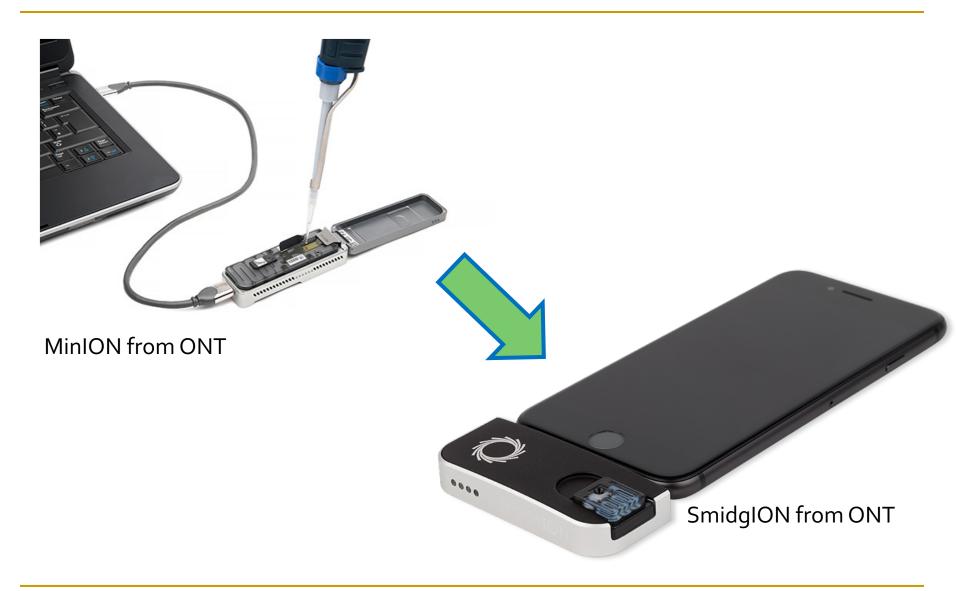
Gagandeep Singh, Mohammed Alser, Damla Senol Cali, Dionysios
 Diamantopoulos, Juan Gómez-Luna, Henk Corporaal, and Onur Mutlu,
 "FPGA-based Near-Memory Acceleration of Modern Data-Intensive
 Applications"
 IFFE Micro (IEEE MICRO), 2021.

FPGA-based Near-Memory Acceleration of Modern Data-Intensive Applications

Gagandeep Singh[⋄] Mohammed Alser[⋄] Damla Senol Cali[⋈]
Dionysios Diamantopoulos[▽] Juan Gómez-Luna[⋄]
Henk Corporaal[⋆] Onur Mutlu^{⋄⋈}

[⋄]ETH Zürich [⋈] Carnegie Mellon University *Eindhoven University of Technology [▽]IBM Research Europe

Future of Genome Sequencing & Analysis



More on Fast & Efficient Genome Analysis ...

Onur Mutlu,

"Accelerating Genome Analysis: A Primer on an Ongoing Journey"

Invited Lecture at <u>Technion</u>, Virtual, 26 January 2021.

[Slides (pptx) (pdf)]

[Talk Video (1 hour 37 minutes, including Q&A)]

[Related Invited Paper (at IEEE Micro, 2020)]





EDIT VIDEO

Detailed Lectures on Genome Analysis

- Computer Architecture, Fall 2020, Lecture 3a
 - Introduction to Genome Sequence Analysis (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=CrRb32v7SJc&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=5
- Computer Architecture, Fall 2020, Lecture 8
 - Intelligent Genome Analysis (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=ygmQpdDTL7o&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=14
- Computer Architecture, Fall 2020, Lecture 9a
 - □ **GenASM: Approx. String Matching Accelerator** (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=XoLpzmN Pas&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=15
- Accelerating Genomics Project Course, Fall 2020, Lecture 1
 - Accelerating Genomics (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=rgjl8ZyLsAg&list=PL5Q2soXY2Zi9E2bBVAgCqL gwiDRQDTyId

Data Overwhelms Modern Machines ...

Storage/memory capability

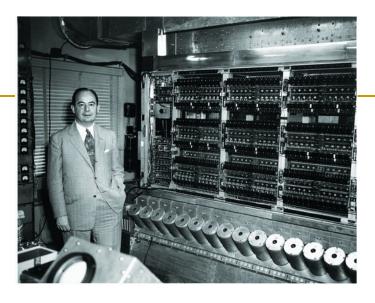
Communication capability

Computation capability

Greatly impacts robustness, energy, performance, cost

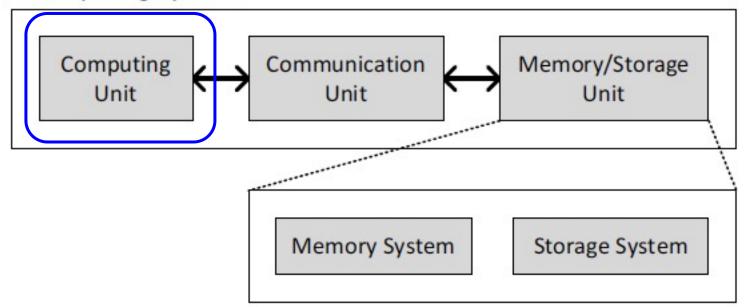
A Computing System

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



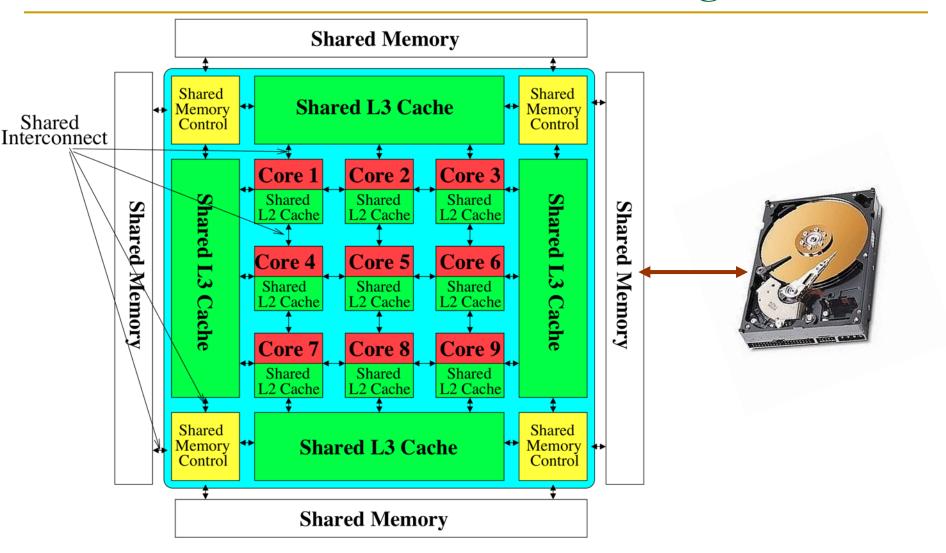
Burks, Goldstein, von Neumann, "Preliminary discussion of the logical design of an electronic computing instrument," 1946.

Computing System



51

Perils of Processor-Centric Design



Most of the system is dedicated to storing and moving data

Data Overwhelms Modern Machines





TensorFlow Mobile

Data → performance & energy bottleneck

VP9
VouTube
Video Playback

Google's video codec



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 Boroumand¹ Rachata Ausavarungnirun¹ Aki Kuusela³ Allan Knies³

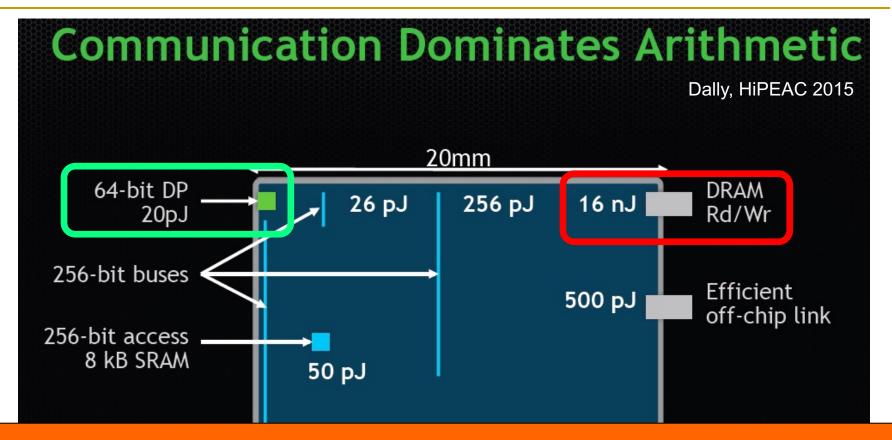
Saugata Ghose¹ Youngsok Kim²

Eric Shiu³ Rahul Thakur³ Daehyun Kim^{4,3}

Parthasarathy Ranganathan³ Onur Mutlu^{5,1}



Data Movement vs. Computation Energy



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

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
- 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

The Problem

Data access is the major performance and energy bottleneck

Our current design principles cause great energy waste

(and great performance loss)

Processing of data is performed far away from the data

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

Challenge and Opportunity for Future

Computing Architectures with Minimal Data Movement

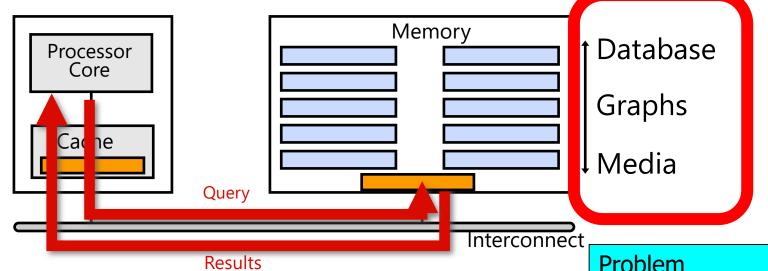
Challenge and Opportunity for Future

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

Challenge and Opportunity for Future

Fundamentally High-Performance (Data-Centric) Computing Architectures

Goal: Processing Inside Memory



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

Problem

Aigorithm

Program/Language

System Software

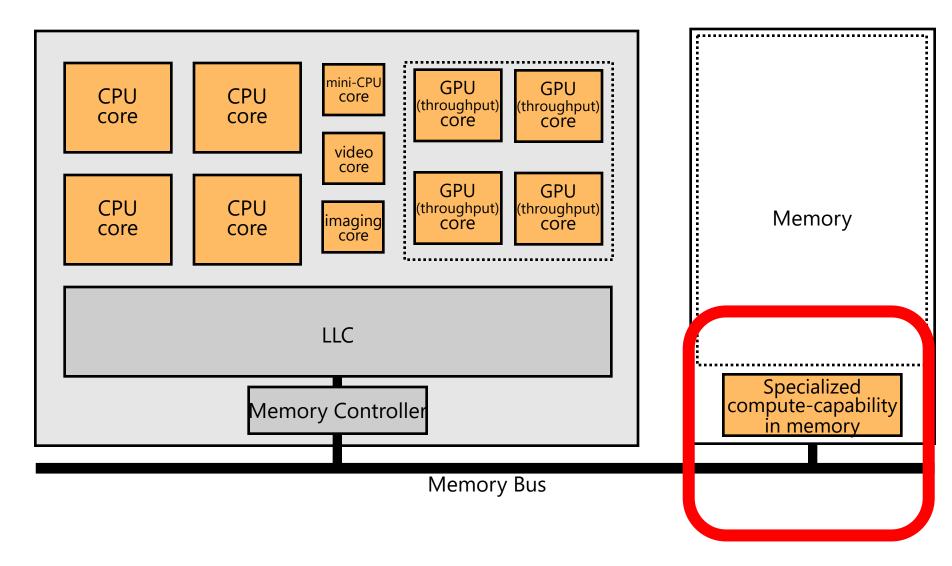
SW/HW Interface

Micro-architecture

Logic

Electrons

Mindset: Memory-Centric Computing



Memory similar to a "conventional" accelerator

PIM Review and Open Problems

A Modern Primer on Processing in Memory

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

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^dKing Mongkut's University of Technology North Bangkok

Onur Mutlu, Saugata Ghose, Juan Gomez-Luna, and Rachata Ausavarungnirun,

"A Modern Primer on Processing in Memory"

Invited Book Chapter in <u>Emerging Computing: From Devices to Systems -</u>

Looking Beyond Moore and Von Neumann, Springer, to be published in 2021.

A Modern Primer on Processing in Memory

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

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Abstract

Modern computing systems are overwhelmingly designed to move data to computation. This design choice goes directly against at least three key trends in computing that cause performance, scalability and energy bottlenecks: (1) data access is a key bottleneck as many important applications are increasingly data-intensive, and memory bandwidth and energy do not scale well, (2) energy consumption is a key limiter in almost all computing platforms, especially server and mobile systems, (3) data movement, especially off-chip to on-chip, is very expensive in terms of bandwidth, energy and latency, much more so than computation. These trends are especially severely-felt in the data-intensive server and energy-constrained mobile systems of today.

At the same time, conventional memory technology is facing many technology scaling challenges in terms of reliability, energy, and performance. As a result, memory system architects are open to organizing memory in different ways and making it more intelligent, at the expense of higher cost. The emergence of 3D-stacked memory plus logic, the adoption of error correcting codes inside the latest DRAM chips, proliferation of different main memory standards and chips, specialized for different purposes (e.g., graphics, low-power, high bandwidth, low latency), and the necessity of designing new solutions to serious reliability and security issues, such as the RowHammer phenomenon, are an evidence of this trend.

This chapter discusses recent research that aims to practically enable computation close to data, an approach we call processing-in-memory (PIM). PIM places computation mechanisms in or near where the data is stored (i.e., inside the memory chips, in the logic layer of 3D-stacked memory, or in the memory controllers), so that data movement between the computation units and memory is reduced or eliminated. While the general idea of PIM is not new, we discuss motivating trends in applications as well as memory circuits/technology that greatly exacerbate the need for enabling it in modern computing systems. We examine at least two promising new approaches to designing PIM systems to accelerate important data-intensive applications: (1) processing using memory by exploiting analog operational properties of DRAM chips to perform massively-parallel operations in memory, with low-cost changes, (2) processing near memory by exploiting 3D-stacked memory technology design to provide high memory bandwidth and low memory latency to in-memory logic. In both approaches, we describe and tackle relevant cross-layer research, design, and adoption challenges in devices, architecture, systems, and programming models. Our focus is on the development of in-memory processing designs that can be adopted in real computing platforms at low cost. We conclude by discussing work on solving key challenges to the practical adoption of PIM.

Keywords: memory systems, data movement, main memory, processing-in-memory, near-data processing, computation-in-memory, processing using memory, processing near memory, 3D-stacked memory, non-volatile memory, energy efficiency, high-performance computing, computer architecture, computing paradigm, emerging technologies, memory scaling, technology scaling, dependable systems, robust systems, hardware security, system security, latency, low-latency computing

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Conclusion and Future Outlook

1. Introduction

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Main memory, built using the Dynamic Random Access Memory (DRAM) technology, is a major component in nearly all computing systems, including servers, cloud platforms, mobile/embedded devices, and sensor systems. Across all of these systems, the data working set sizes of modern applications are rapidly growing, while the need for fast analysis of such data is increasing. Thus, main memory is becoming an increasingly significant bottleneck across a wide variety of computing systems and applications [1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16]. Alleviating the main memory bottleneck requires the memory capacity, energy, cost, and performance to all scale in an efficient manner across technology generations. Unfortunately, it has become increasingly difficult in recent years, especially the past decade, to scale all of these dimensions [1, 2, 17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49], and thus the main memory bottleneck has been worsening.

A major reason for the main memory bottleneck is the high energy and latency cost associated with data movement. In modern computers, to perform any operation on data that resides in main memory, the processor must retrieve the data from main memory. This requires the memory controller to issue commands to a DRAM module across a relatively slow and power-hungry off-chip bus (known as the memory channel). The DRAM module sends the requested data across the memory channel, after which the data is placed in the caches and registers. The CPU can perform computation on the data once the data is in its registers. Data movement from the DRAM to the CPU incurs long latency and consumes a significant amount of energy [7, 50, 51, 52, 53, 54]. These costs are often exacerbated by the fact that much of the data brought into the caches is not reused by the CPU [52, 53, 55, 56], providing little benefit in return for the high latency and energy cost.

The cost of data movement is a fundamental issue with the processor-centric nature of contemporary computer systems. The CPU is considered to be the master in the system, and computation is performed only in the processor (and accelerators). In contrast, data storage and communication units, including the main memory, are treated as unintelligent workers that are incapable of computation. As a result of this processor-centric design paradigm, data moves a lot in the system between the computation units and communication/ storage units so that computation can be done on it. With the increasingly data-centric nature of contemporary and emerging appli-

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]

Processing in Memory: Two Approaches

- 1. Processing using Memory
- 2. Processing near Memory

Processing using Memory

- We can support in-DRAM 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

Ambit: Bulk-Bitwise in-DRAM Computation

 Vivek Seshadri, Donghyuk Lee, Thomas Mullins, Hasan Hassan, Amirali Boroumand, Jeremie Kim, Michael A. Kozuch, Onur Mutlu, Phillip B. Gibbons, and Todd C. Mowry,

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

Proceedings of the <u>50th International Symposium on</u>

Microarchitecture (MICRO), Boston, MA, USA, October 2017.

[Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)] [Poster (pptx) (pdf)]

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

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

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

In-DRAM Bulk Bitwise Execution Paradigm

Vivek Seshadri and Onur Mutlu,
 "In-DRAM Bulk Bitwise Execution Engine"
 Invited Book Chapter in Advances in Computers, 2020.
 [Preliminary arXiv version]

In-DRAM Bulk Bitwise Execution Engine

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SIMDRAM Framework

Nastaran Hajinazar, Geraldo F. Oliveira, Sven Gregorio, Joao Dinis Ferreira, Nika Mansouri Ghiasi, Minesh Patel, Mohammed Alser, Saugata Ghose, Juan Gomez-Luna, and Onur Mutlu, "SIMDRAM: An End-to-End Framework for Bit-Serial SIMD Computing in DRAM" Proceedings of the 26th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), Virtual, March-April 2021.

[2-page Extended Abstract]

[Short Talk Slides (pptx) (pdf)]

[Talk Slides (pptx) (pdf)]

[Short Talk Video (5 mins)]

[Full Talk Video (27 mins)]

SIMDRAM: A Framework for Bit-Serial SIMD Processing using DRAM

*Nastaran Hajinazar^{1,2}
Nika Mansouri Ghiasi¹

*Geraldo F. Oliveira¹
Minesh Patel¹
Juan Gómez-Luna¹

Sven Gregorio¹ Mohammed Alser¹ Onur Mutlu¹

João Dinis Ferreira¹ Saugata Ghose³

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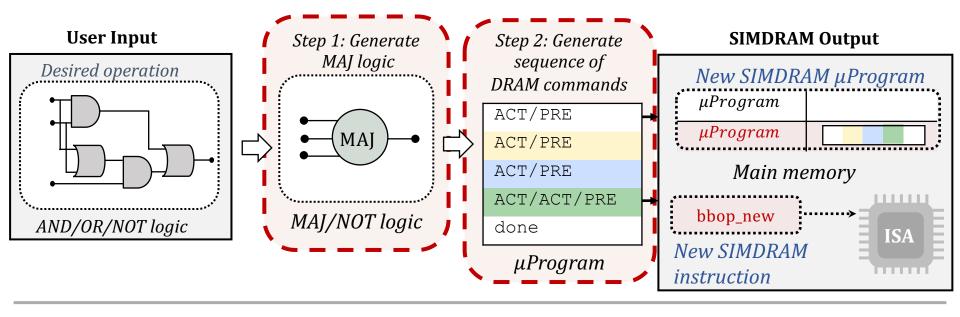
²Simon Fraser University

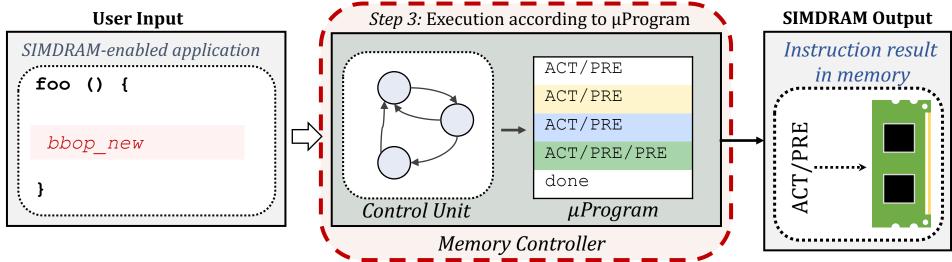
³University of Illinois at Urbana-Champaign

SIMDRAM Key Idea

- **SIMDRAM:** An end-to-end processing-using-DRAM framework that provides the programming interface, the ISA, and the hardware support for:
 - Efficiently computing complex operations in DRAM
 - Providing the ability to implement **arbitrary** operations as required
 - Using an **in-DRAM massively-parallel SIMD substrate** that requires **minimal** changes to DRAM architecture

SIMDRAM Framework: Overview





SIMDRAM Key Results

Evaluated on:

- 16 complex in-DRAM operations
- 7 commonly-used real-world applications

SIMDRAM provides:

- 88× and 5.8× the throughput of a CPU and a high-end GPU, respectively, over 16 operations
- 257× and 31× the energy efficiency of a CPU and a high-end GPU, respectively, over 16 operations
- 21× and 2.1× the performance of a CPU an a high-end GPU, over seven real-world applications

SIMDRAM Conclusion

• SIMDRAM:

- Enables efficient computation of a flexible set and wide range of operations in a PuM massively parallel SIMD substrate
- Provides the hardware, programming, and ISA support, to:
 - Address key system integration challenges
 - Allow programmers to define and employ new operations without hardware changes

SIMDRAM is a promising PuM framework

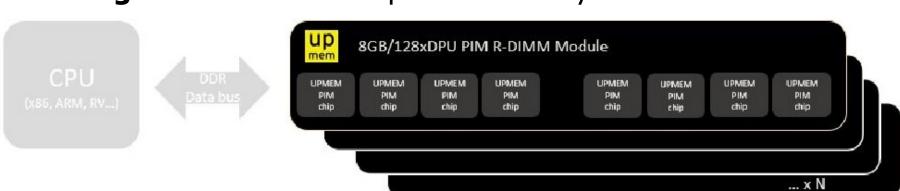
- Can ease the adoption of processing-using-DRAM architectures
- Improves the performance and efficiency of processingusing-memory architectures

Processing in Memory: Two Approaches

- 1. Processing using Memory
- 2. Processing near Memory

UPMEM Processing-in-DRAM Engine (2019)

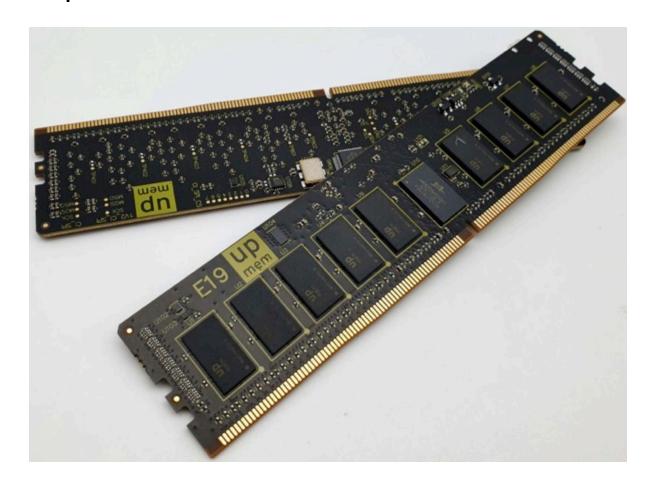
- Processing in DRAM Engine
- Includes standard DIMM modules, with a large number of DPU processors combined with DRAM chips.
- Replaces standard DIMMs
 - DDR4 R-DIMM modules
 - 8GB+128 DPUs (16 PIM chips)
 - Standard 2x-nm DRAM process
 - Large amounts of compute & memory bandwidth





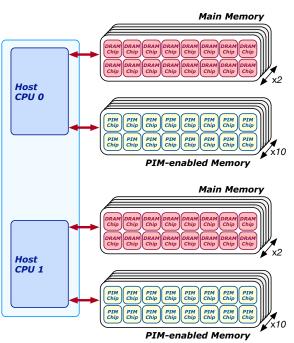
UPMEM Memory Modules

- E19: 8 chips DIMM (1 rank). DPUs @ 267 MHz
- P21: 16 chips DIMM (2 ranks). DPUs @ 350 MHz





2,560-DPU Processing-in-Memory System



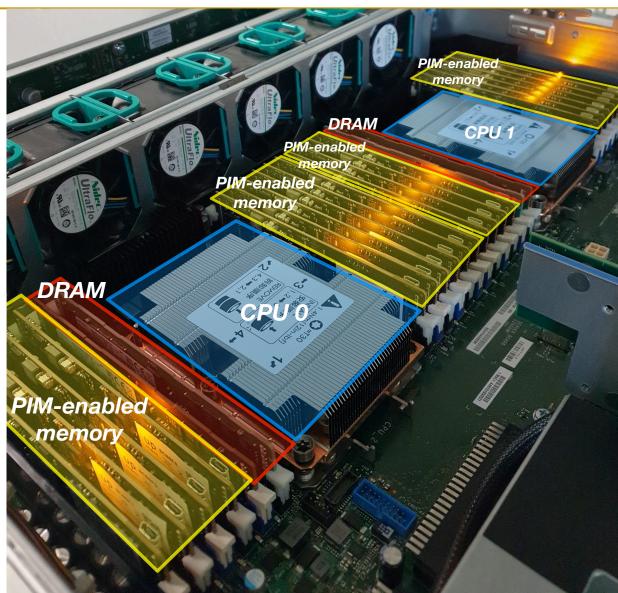
Benchmarking a New Paradigm: An Experimental Analysis of a Real Processing-in-Memory Architecture

JUAN GÓMEZ-LUNA, ETH Zürich, Switzerland
IZZAT EL HAJJ, American University of Beirut, Lebanon
IVAN FERNANDEZ, ETH Zürich, Switzerland and University of Malaga, Spain
CHRISTINA GIANNOULA, ETH Zürich, Switzerland and NTUA, Greece
GERALDO F. OLIVEIRA, ETH Zürich, Switzerland
ONUR MUTLU, ETH Zürich, Switzerland

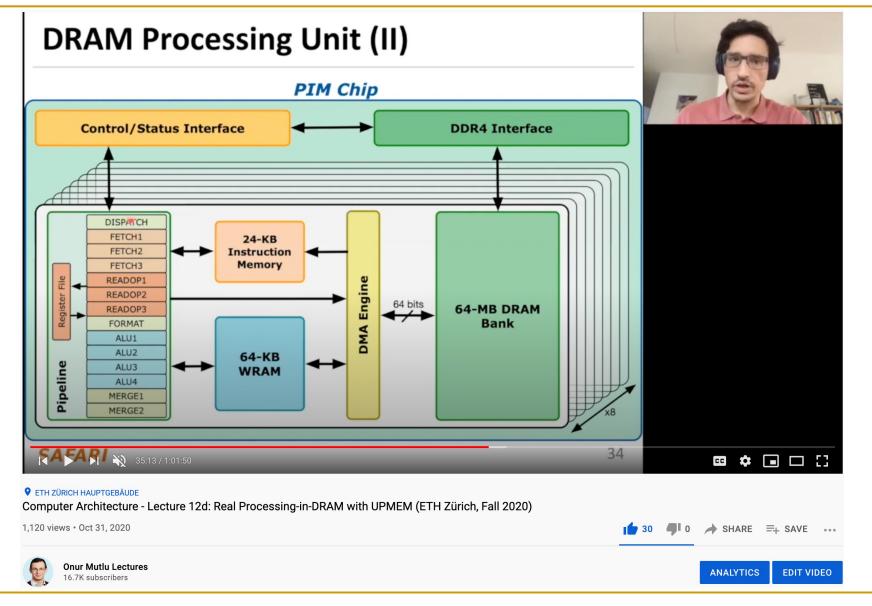
Many modern workloads, such as neural networks, databases, and graph processing, are fundamentally memory-bound. For such workloads, the data novement between main memory and CPU cores imposes a significant overhead in terns of both latency and energy. A major reason is that this communication happens through a narrow bus with high latency and limited bandwidth, and the low data reuse in memory-bound workloads is insufficient to amortize the cost of main memory access. Fundamentally addressing this data movement bottleneck requires a paradigm where the memory system assumes an active role in computing by integrating processing capabilities. This paradigm is known as processing—in-memory (PM).

Recent research explores different forms of PIM architectures, motivated by the emergence of new 3Dstacked memory technologies that integrate memory with a logic layer where processing elements can be
easily placed. Past works evaluate these architectures in simulation or, at best, with simplified hardware
prototypes. In contrast, the UPMEM company has designed and manufactured the first publicly-available
real-world PIM architecture. The UPMEM PIM architecture combines traditional DRAM memory arrays with
general-purpose in-order cores, called DRAM Processing Units (DPUs), integrated in the same chip.

This paper provides the first comprehensive analysis of the first publicly-available real-world PIM architecture. We make two key contributions. First, we conduct an experimental characterization of the UPMEM-based PIM system using microbenchmarks to assess various architecture limits such as compute throughput and memory bandwidth, yielding new insights. Second, we present PIM (Processing,-bendumpy) benchmarks), a benchmark suite of 16 workloads from different application domains (e.g., dense/sparse linear algebra, databases, data analytics, graph processing, which we identify as memory-bound. We evaluate the performance and scaling characteristics of PIM benchmarks on the UPMEM PIM architecture, and compare their performance and energy consumption to their state-of-the-art CPU and CPU counterparts. Our extensive evaluation conducted on two real UPMEM-based PIM systems with 460 and 25.50 DPUs provides new insights about suitability of different workloads to the PIM systems you for the programming recommendations for software designers, and suggestions and hints for hardware and architecture designers of future PIM systems.



More on the UPMEM PIM System



Experimental Analysis of the UPMEM PIM Engine

Benchmarking a New Paradigm: An Experimental Analysis of a Real Processing-in-Memory Architecture

JUAN GÓMEZ-LUNA, ETH Zürich, Switzerland IZZAT EL HAJJ, American University of Beirut, Lebanon IVAN FERNANDEZ, ETH Zürich, Switzerland and University of Malaga, Spain CHRISTINA GIANNOULA, ETH Zürich, Switzerland and NTUA, Greece GERALDO F. OLIVEIRA, ETH Zürich, Switzerland ONUR MUTLU, ETH Zürich, Switzerland

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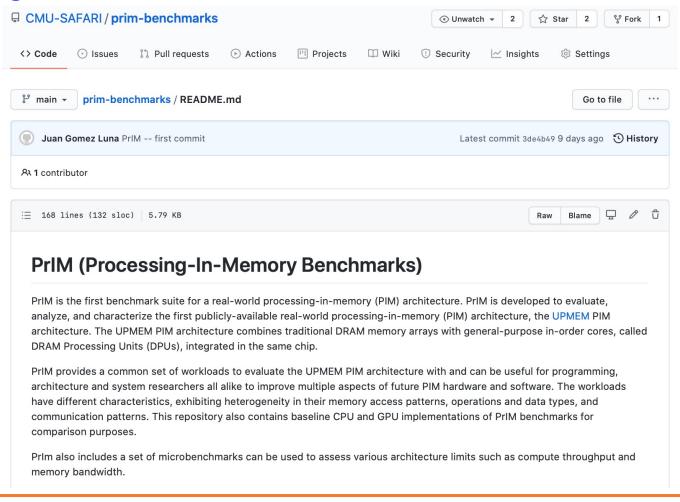
https://arxiv.org/pdf/2105.03814.pdf

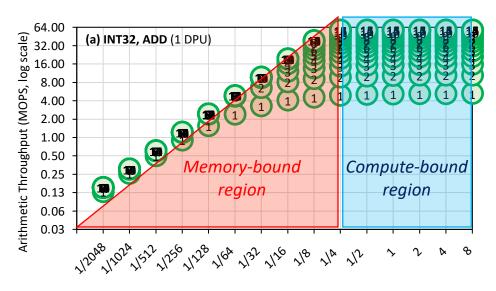
PrIM Benchmarks: Application Domains

Domain	Benchmark	Short name
Dense linear algebra	Vector Addition	VA
	Matrix-Vector Multiply	GEMV
Sparse linear algebra	Sparse Matrix-Vector Multiply	SpMV
Databases	Select	SEL
	Unique	UNI
Data analytics	Binary Search	BS
	Time Series Analysis	TS
Graph processing	Breadth-First Search	BFS
Neural networks	Multilayer Perceptron	MLP
Bioinformatics	Needleman-Wunsch	NW
Image processing	Image histogram (short)	HST-S
	Image histogram (large)	HST-L
Parallel primitives	Reduction	RED
	Prefix sum (scan-scan-add)	SCAN-SSA
	Prefix sum (reduce-scan-scan)	SCAN-RSS
	Matrix transposition	TRNS

PrIM Benchmarks are Open Source

- All microbenchmarks, benchmarks, and scripts
- https://github.com/CMU-SAFARI/prim-benchmarks



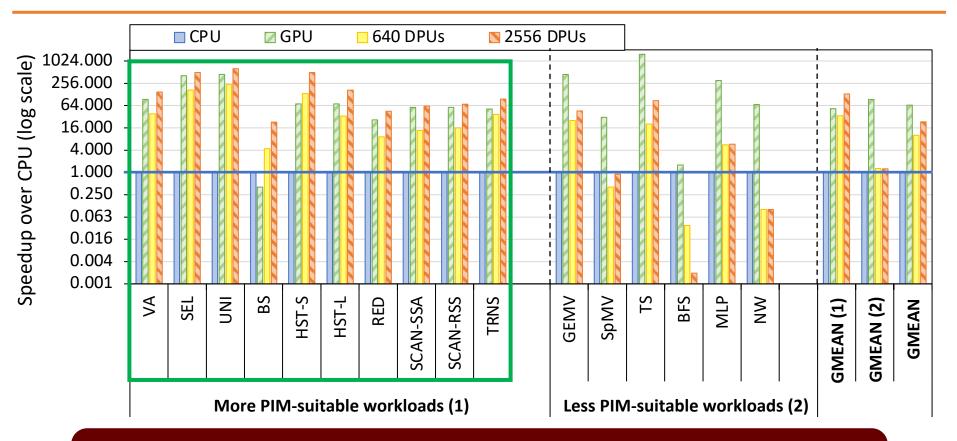


The throughput saturation point is as low as ¼ OP/B, i.e., 1 integer addition per every 32-bit element fetched

Operational Intensity (OP/B)

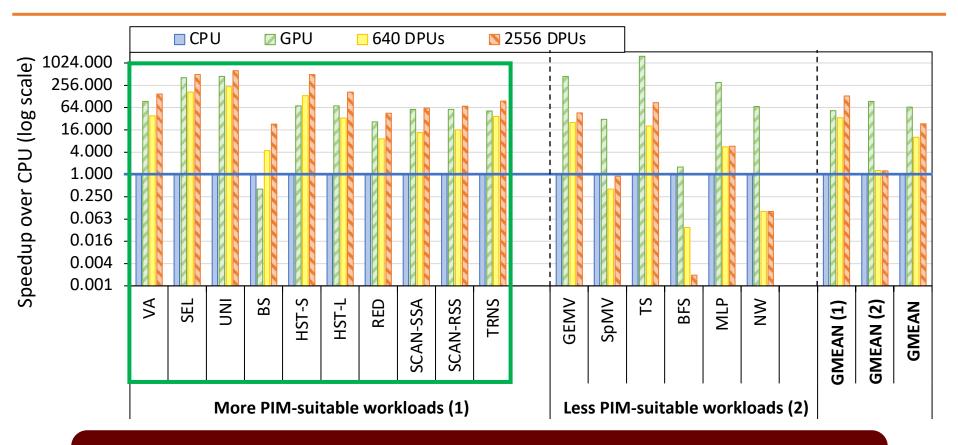
KEY TAKEAWAY 1

The UPMEM PIM architecture is fundamentally compute bound. As a result, the most suitable workloads are memory-bound.



KEY TAKEAWAY 2

The most well-suited workloads for the UPMEM PIM architecture use no arithmetic operations or use only simple operations (e.g., bitwise operations and integer addition/subtraction).



KEY TAKEAWAY 3

The most well-suited workloads for the UPMEM PIM architecture require little or no communication across DPUs (inter-DPU communication).

KEY TAKEAWAY 4

- UPMEM-based PIM systems outperform state-of-the-art CPUs in terms of performance and energy efficiency on most of PrIM benchmarks.
- UPMEM-based PIM systems **outperform state-of-the-art GPUs on a majority of PrIM benchmarks**, and the outlook is even more positive for future PIM systems.
- UPMEM-based PIM systems are more energy-efficient than stateof-the-art CPUs and GPUs on workloads that they provide performance improvements over the CPUs and the GPUs.

More on UPMEM System & Analysis

Juan Gomez-Luna, Izzat El Hajj, Ivan Fernandez, Christina Giannoula, Geraldo F. Oliveira, and Onur Mutlu, "Benchmarking a New Paradigm: An Experimental Analysis of a Real Processing-in-Memory Architecture"

Preprint in arXiv, 9 May 2021.

[arXiv preprint]

PrIM Benchmarks Source Code

[Slides (pptx) (pdf)]

[Long Talk Slides (pptx) (pdf)]

[Short Talk Slides (pptx) (pdf)]

[SAFARI Live Seminar Slides (pptx) (pdf)]

[SAFARI Live Seminar Video (2 hrs 57 mins)]

[Lightning Talk Video (3 minutes)]

[Short Talk Video (21 minutes)]

[<u>1-hour Talk Video</u> (58 minutes)]

Understanding a Modern Processing-in-Memory Architecture: Benchmarking and Experimental Characterization

Juan Gómez-Luna 1 Izzat El Hajj 2 Ivan Fernandez 1,3 Christina Giannoula 1,4 Geraldo F. Oliveira 1 Onur Mutlu 1

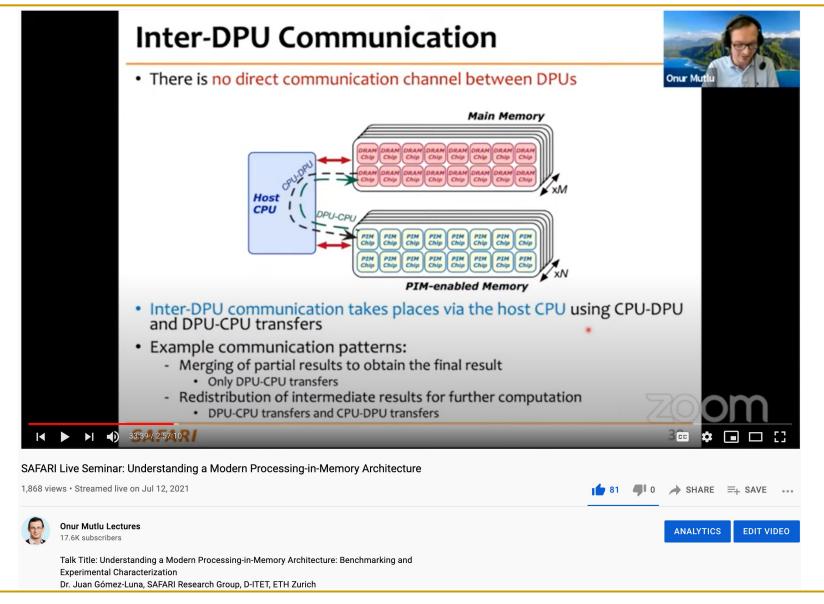
¹ETH Zürich ²American University of Beirut ³University of Malaga ⁴National Technical University of Athens



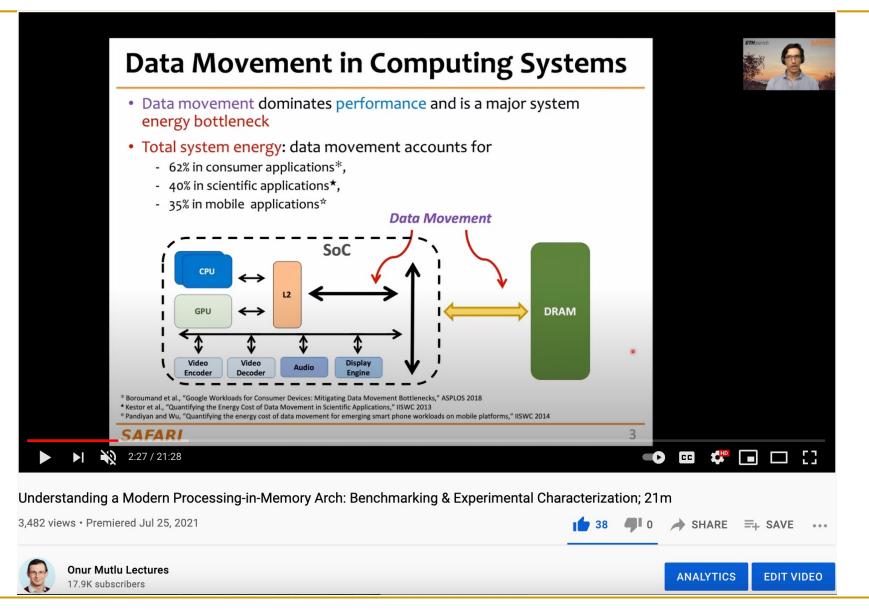
Understanding a Modern PIM Architecture



More on Analysis of the UPMEM PIM Engine



More on Analysis of the UPMEM PIM Engine



FPGA-based Processing Near Memory

Gagandeep Singh, Mohammed Alser, Damla Senol Cali, Dionysios
Diamantopoulos, Juan Gómez-Luna, Henk Corporaal, and Onur Mutlu,

"FPGA-based Near-Memory Acceleration of Modern Data-Intensive

Applications"

IEEE Micro (IEEE MICRO), to appear, 2021.

FPGA-based Near-Memory Acceleration of Modern Data-Intensive Applications

Gagandeep Singh[⋄] Mohammed Alser[⋄] Damla Senol Cali[⋈]
Dionysios Diamantopoulos[▽] Juan Gómez-Luna[⋄]
Henk Corporaal[⋆] Onur Mutlu^{⋄⋈}

[⋄]ETH Zürich [⋈] Carnegie Mellon University *Eindhoven University of Technology [▽]IBM Research Europe

DAMOV Analysis Methodology & Workloads

DAMOV: A New Methodology and Benchmark Suite for Evaluating Data Movement Bottlenecks

GERALDO F. OLIVEIRA, ETH Zürich, Switzerland
JUAN GÓMEZ-LUNA, ETH Zürich, Switzerland
LOIS OROSA, ETH Zürich, Switzerland
SAUGATA GHOSE, University of Illinois at Urbana-Champaign, USA
NANDITA VIJAYKUMAR, University of Toronto, Canada
IVAN FERNANDEZ, University of Malaga, Spain & ETH Zürich, Switzerland
MOHAMMAD SADROSADATI, Institute for Research in Fundamental Sciences (IPM), Iran & ETH
Zürich, Switzerland
ONUR MUTLU, ETH Zürich, Switzerland

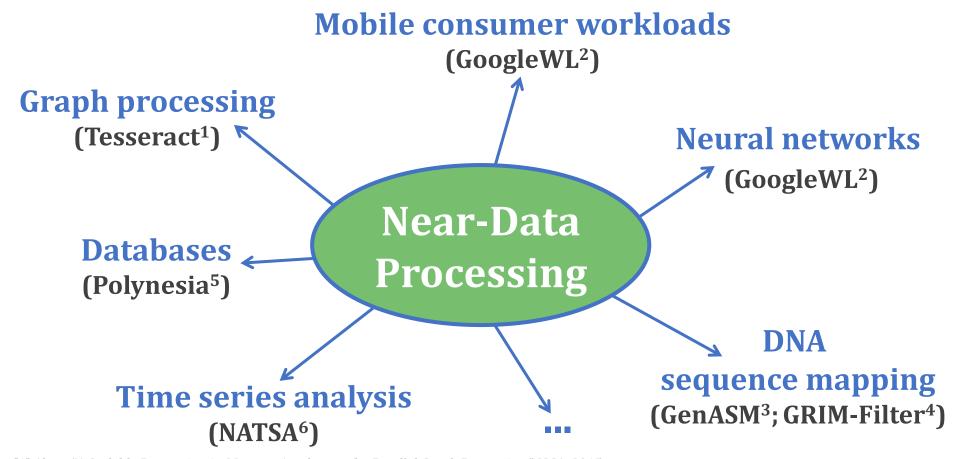
Data movement between the CPU and main memory is a first-order obstacle against improving performance, scalability, and energy efficiency in modern systems. Computer systems employ a range of techniques to reduce overheads tied to data movement, spanning from traditional mechanisms (e.g., deep multi-level cache hierarchies, aggressive hardware prefetchers) to emerging techniques such as Near-Data Processing (NDP), where some computation is moved close to memory. Prior NDP works investigate the root causes of data movement bottlenecks using different profiling methodologies and tools. However, there is still a lack of understanding about the key metrics that can identify different data movement bottlenecks and their relation to traditional and emerging data movement mitigation mechanisms. Our goal is to methodically identify potential sources of data movement over a broad set of applications and to comprehensively compare traditional compute-centric data movement mitigation techniques (e.g., caching and prefetching) to more memory-centric techniques (e.g., NDP), thereby developing a rigorous understanding of the best techniques to mitigate each source of data movement.

With this goal in mind, we perform the first large-scale characterization of a wide variety of applications, across a wide range of application domains, to identify fundamental program properties that lead to data movement to/from main memory. We develop the first systematic methodology to classify applications based on the sources contributing to data movement bottlenecks. From our large-scale characterization of 77K functions across 345 applications, we select 144 functions to form the first open-source benchmark suite (DAMOV) for main memory data movement studies. We select a diverse range of functions that (1) represent different types of data movement bottlenecks, and (2) come from a wide range of application domains. Using NDP as a case study, we identify new insights about the different data movement bottlenecks and use these insights to determine the most suitable data movement mitigation mechanism for a particular application. We open-source DAMOV and the complete source code for our new characterization methodology at https://github.com/CMU-SAFARI/DAMOV.

SAFARI

https://arxiv.org/pdf/2105.03725.pdf

When to Employ Near-Data Processing?



- [1] Ahn+, "A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing," ISCA, 2015
- [2] Boroumand+, "Google Workloads for Consumer Devices: Mitigating Data Movement Bottlenecks," ASPLOS, 2018
- [3] Cali+, "GenASM: A High-Performance, Low-Power Approximate String Matching Acceleration Framework for Genome Sequence Analysis," MICRO, 2020
- [4] Kim+, "GRIM-Filter: Fast Seed Location Filtering in DNA Read Mapping Using Processing-in-Memory Technologies," BMC Genomics, 2018
- [5] Boroumand+, "Polynesia: Enabling Effective Hybrid Transactional/Analytical Databases with Specialized Hardware/Software Co-Design," arXiv:2103.00798 [cs.AR], 2021

[6] Fernandez+, "NATSA: A Near-Data Processing Accelerator for Time Series Analysis," ICCD, 2020

Key Approach

- New workload characterization methodology to analyze:
 - data movement bottlenecks
 - suitability of different data movement mitigation mechanisms
- Two main profiling strategies:

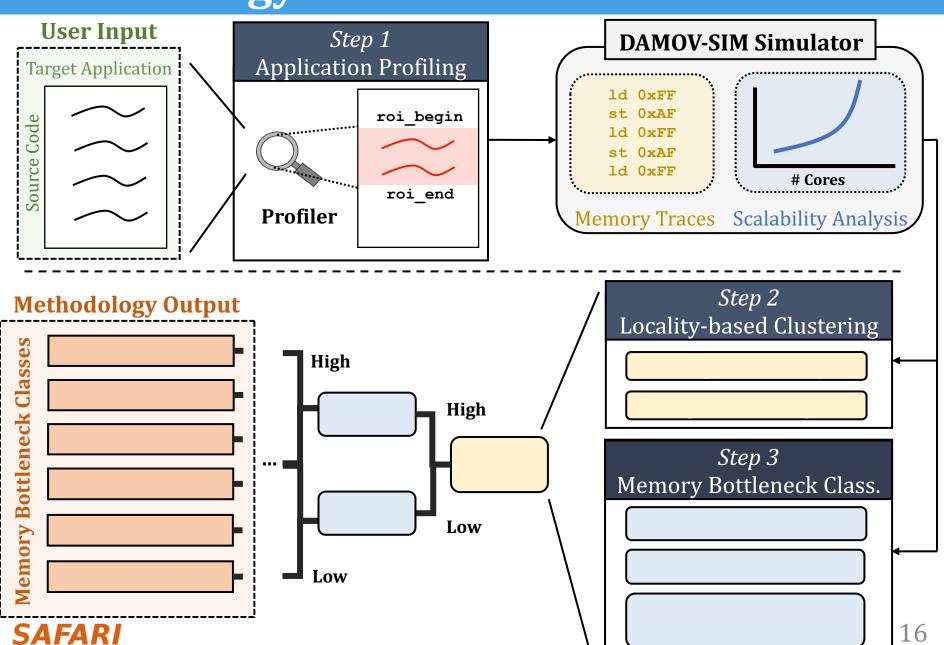
Architecture-independent profiling:

characterizes the memory behavior independently of the underlying hardware

Architecture-dependent profiling:

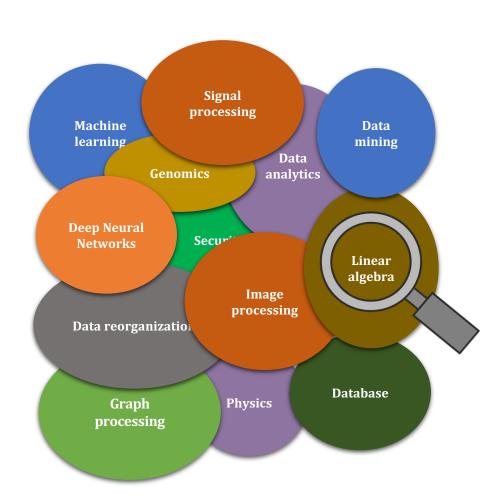
evaluates the impact of the system configuration on the memory behavior

Methodology Overview

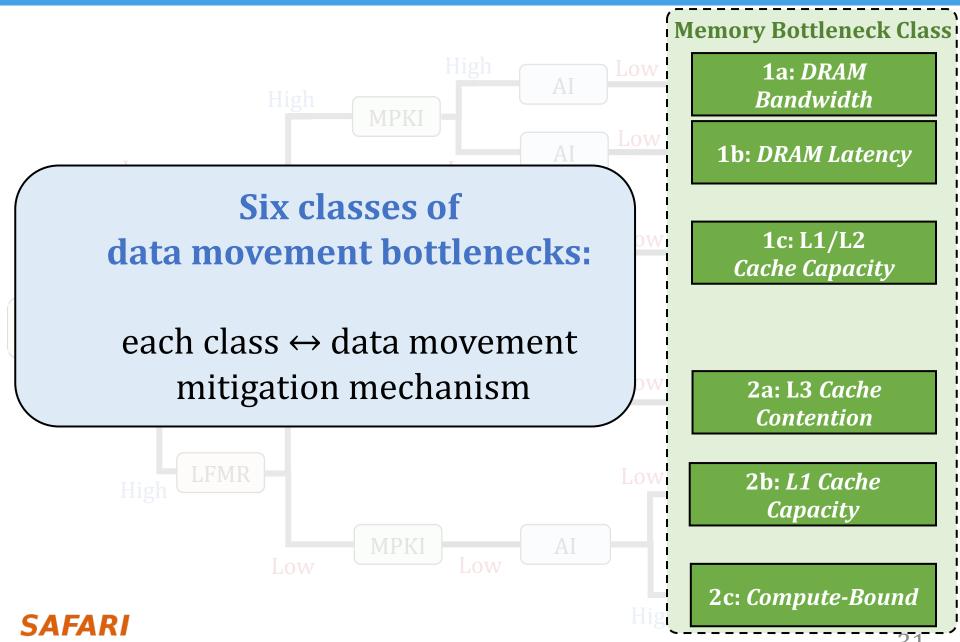


Step 1: Application Profiling

- We analyze 345 applications from distinct domains:
- Graph Processing
- Deep Neural Networks
- Physics
- High-Performance Computing
- Genomics
- Machine Learning
- Databases
- Data Reorganization
- Image Processing
- Map-Reduce
- Benchmarking
- Linear Algebra

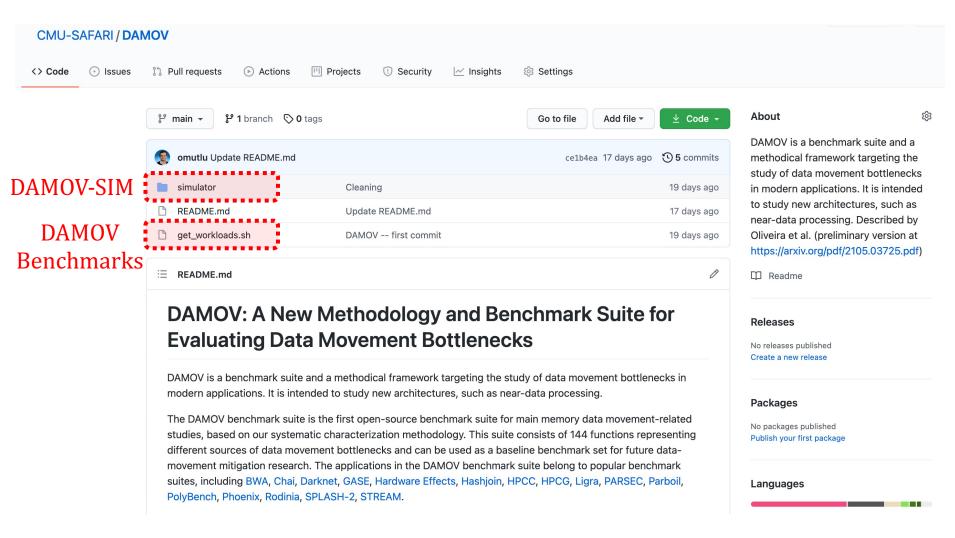


Step 3: Memory Bottleneck Analysis



DAMOV is Open Source

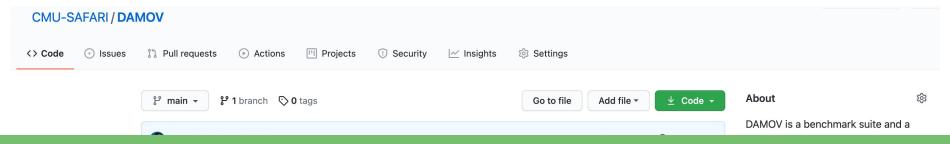
We open-source our benchmark suite and our toolchain





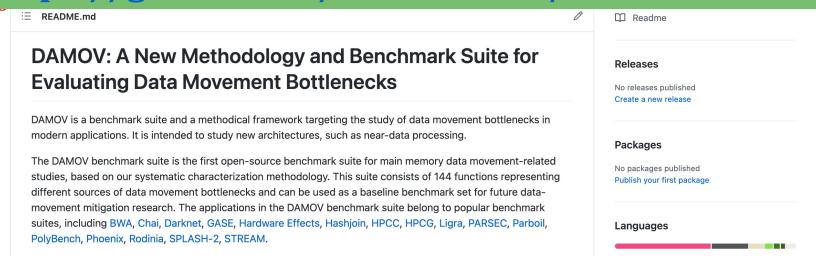
DAMOV is Open Source

We open-source our benchmark suite and our toolchain



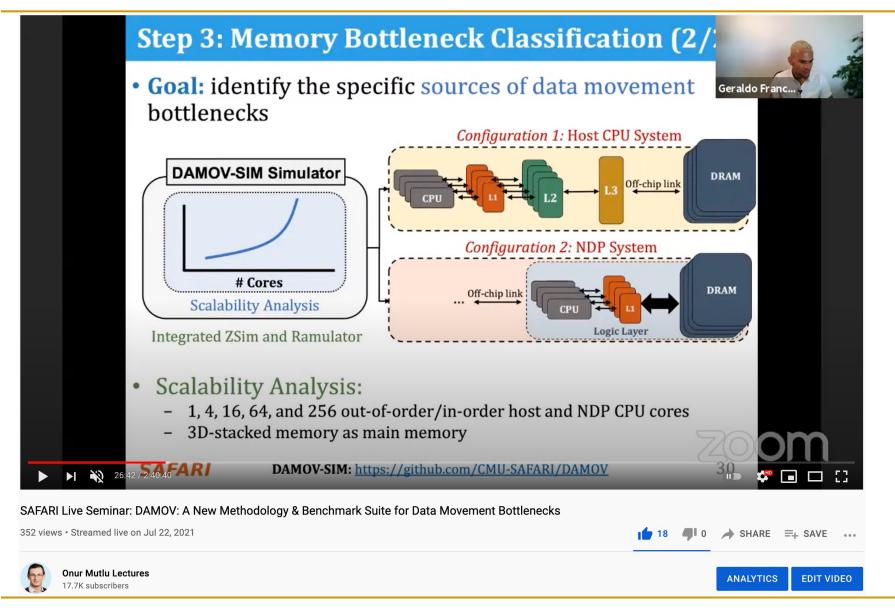
Get DAMOV at:

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





More on DAMOV Analysis Methodology & Workloads



More on DAMOV

Geraldo F. Oliveira, Juan Gomez-Luna, Lois Orosa, Saugata Ghose, Nandita Vijaykumar, Ivan fernandez, Mohammad Sadrosadati, and Onur Mutlu, "DAMOV: A New Methodology and Benchmark Suite for Evaluating Data Movement Bottlenecks"

Preprint in <u>arXiv</u>, 8 May 2021.

[arXiv preprint]

[DAMOV Suite and Simulator Source Code]

[SAFARI Live Seminar Video (2 hrs 40 mins)]

ONUR MUTLU, ETH Zürich, Switzerland

[Short Talk Video (21 minutes)]

DAMOV: A New Methodology and Benchmark Suite for Evaluating Data Movement Bottlenecks

GERALDO F. OLIVEIRA, ETH Zürich, Switzerland
JUAN GÓMEZ-LUNA, ETH Zürich, Switzerland
LOIS OROSA, ETH Zürich, Switzerland
SAUGATA GHOSE, University of Illinois at Urbana-Champaign, USA
NANDITA VIJAYKUMAR, University of Toronto, Canada
IVAN FERNANDEZ, University of Malaga, Spain & ETH Zürich, Switzerland
MOHAMMAD SADROSADATI, ETH Zürich, Switzerland

Eliminating the Adoption Barriers

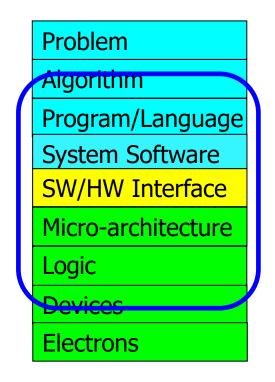
How to Enable Adoption of Processing in Memory

Potential Barriers to Adoption of PIM

- 1. Functionality and applications & software for PIM
- 2. Ease of **programming** (interfaces and compiler/HW support)
- 3. **System** support: coherence, synchronization, 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



We can get there step by step

Challenge and Opportunity for Future

Data-Driven (Self-Optimizing) Computing Architectures

Challenge and Opportunity for Future

Data-Aware (Expressive) Computing Architectures

More Info in This Tutorial...

Onur Mutlu,

"Memory-Centric Computing Systems"

Invited Tutorial at <u>66th International Electron Devices</u>

Meeting (IEDM), Virtual, 12 December 2020.

[Slides (pptx) (pdf)]

[Executive Summary Slides (pptx) (pdf)]

[Tutorial Video (1 hour 51 minutes)]

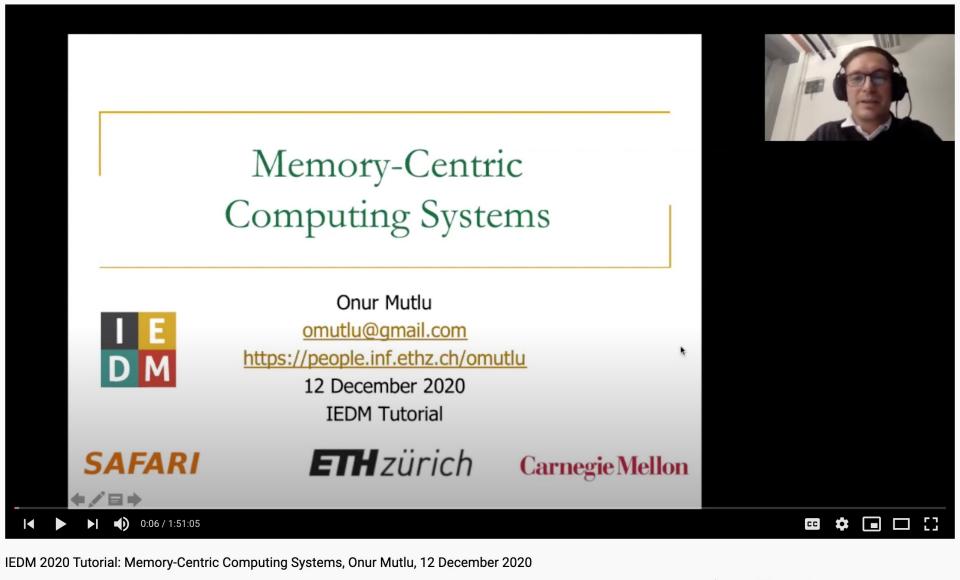
[Executive Summary Video (2 minutes)]

[Abstract and Bio]

[Related Keynote Paper from VLSI-DAT 2020]

[Related Review Paper on Processing in Memory]

https://www.youtube.com/watch?v=H3sEaINPBOE



1,641 views • Dec 23, 2020 → SHARE =+ SAVE •



ANALYTICS

EDIT VIDEO

Architectures for Intelligent Machines

Data-centric

Data-driven

Data-aware





SAFARI Research Group Introduction & Research

Onur Mutlu

omutlu@gmail.com

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

19 October 2021

EFCL Huawei Day





Carnegie Mellon

More Detailed Research Overview

Slides from ISCA 2021 Mentoring Workshop Panel

Onur Mutlu,

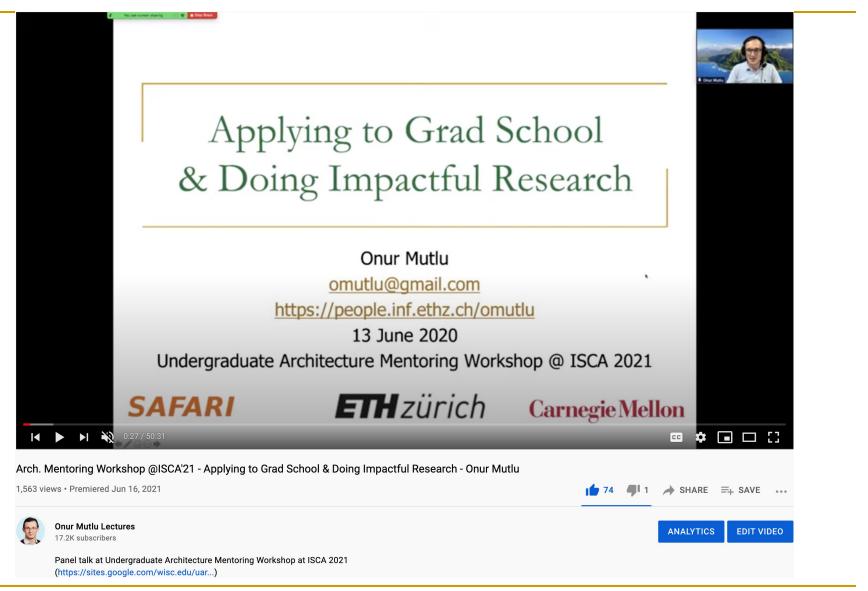
"Applying to Graduate School & Doing Impactful Research"

Invited Panel Talk at the 3rd Undergraduate Mentoring Workshop, held with the 48th International Symposium on Computer Architecture (ISCA), Virtual, 18 June 2021.

[Slides (pptx) (pdf)]

[Talk Video (50 minutes)]

A Talk on Impactful Research & Teaching





Example Research Topics: Quick Overview

Challenge and Opportunity for Future

High Performance

(to solve the **toughest** & **all** problems)

Challenge and Opportunity for Future

Personalized and Private

(in every aspect of life: health, medicine, spaces, devices, robotics, ...)

Accelerating Genome Analysis

 Mohammed Alser, Zulal Bingol, Damla Senol Cali, Jeremie Kim, Saugata Ghose, Can Alkan, and Onur Mutlu,

"Accelerating Genome Analysis: A Primer on an Ongoing Journey"

IEEE Micro (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

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 Acceleration Framework for Genome Sequence Analysis"
Proceedings of the 53rd International Symposium on Microarchitecture (MICRO), Virtual, October 2020.

[<u>Lighting Talk Video</u> (1.5 minutes)]
[<u>Lightning Talk Slides (pptx) (pdf)</u>]
[<u>Talk Video</u> (18 minutes)]
[<u>Slides (pptx) (pdf)</u>]

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

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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/bby017

Published: 02 April 2018 Article history ▼

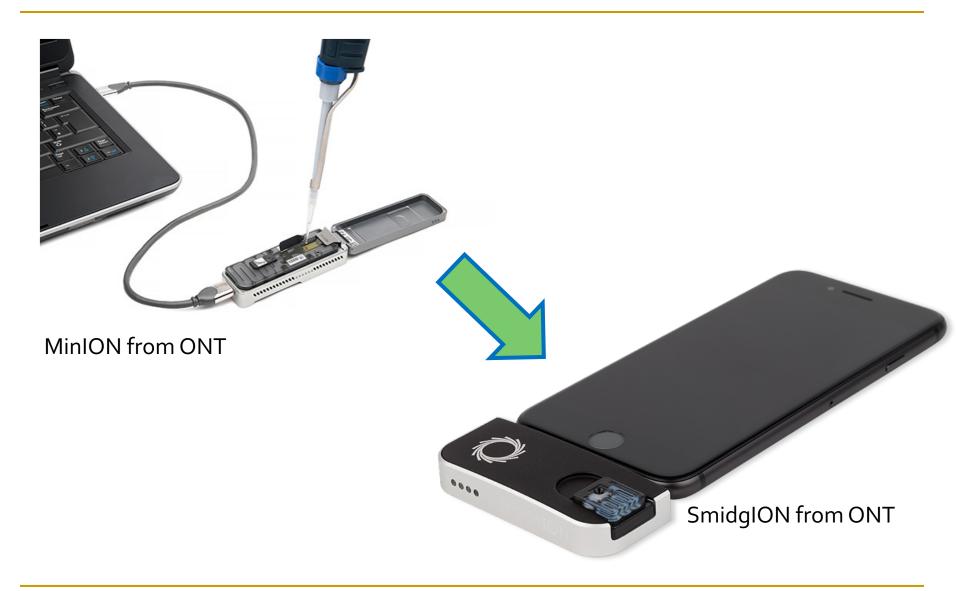


Oxford Nanopore MinION

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

[Preliminary arxiv.org version]

Future of Genome Sequencing & Analysis



More on Fast & Efficient Genome Analysis

Onur Mutlu,

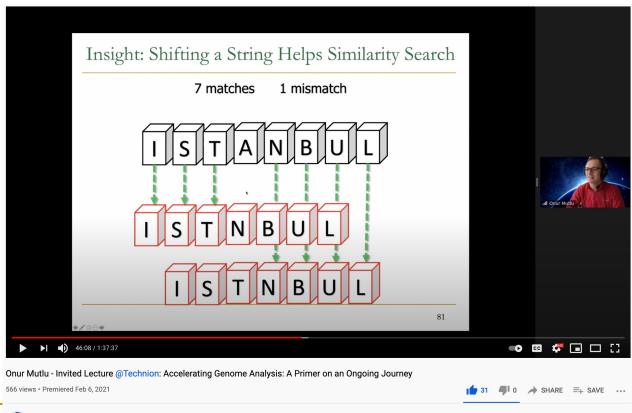
"Accelerating Genome Analysis: A Primer on an Ongoing Journey"

Invited Lecture at <u>Technion</u>, Virtual, 26 January 2021.

[Slides (pptx) (pdf)]

[Talk Video (1 hour 37 minutes, including Q&A)]

[Related Invited Paper (at IEEE Micro, 2020)]





ANALYTICS

Detailed Lectures on Genome Analysis

- Computer Architecture, Fall 2020, Lecture 3a
 - Introduction to Genome Sequence Analysis (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=CrRb32v7SJc&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=5
- Computer Architecture, Fall 2020, Lecture 8
 - □ **Intelligent Genome Analysis** (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=ygmQpdDTL7o&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=14
- Computer Architecture, Fall 2020, Lecture 9a
 - □ **GenASM: Approx. String Matching Accelerator** (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=XoLpzmN Pas&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=15
- Accelerating Genomics Project Course, Fall 2020, Lecture 1
 - Accelerating Genomics (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=rgjl8ZyLsAg&list=PL5Q2soXY2Zi9E2bBVAgCqL gwiDRQDTyId

Computing is Bottlenecked by Data

Modern Systems are Bottlenecked by

Data Storage and Movement

Modern Systems are Bottlenecked by Memory

An "Early" Overview Paper...

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

Memory Scaling: A Systems Architecture Perspective

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

Challenge and Opportunity for Future

Fundamentally Secure, Reliable, Safe Computing Architectures

Infrastructures to Understand Such Issues



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

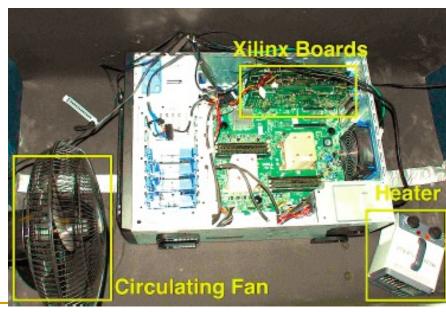
Adaptive-Latency DRAM: Optimizing DRAM
Timing for the Common-Case (Lee et al.,
HPCA 2015)

AVATAR: A Variable-Retention-Time (VRT)

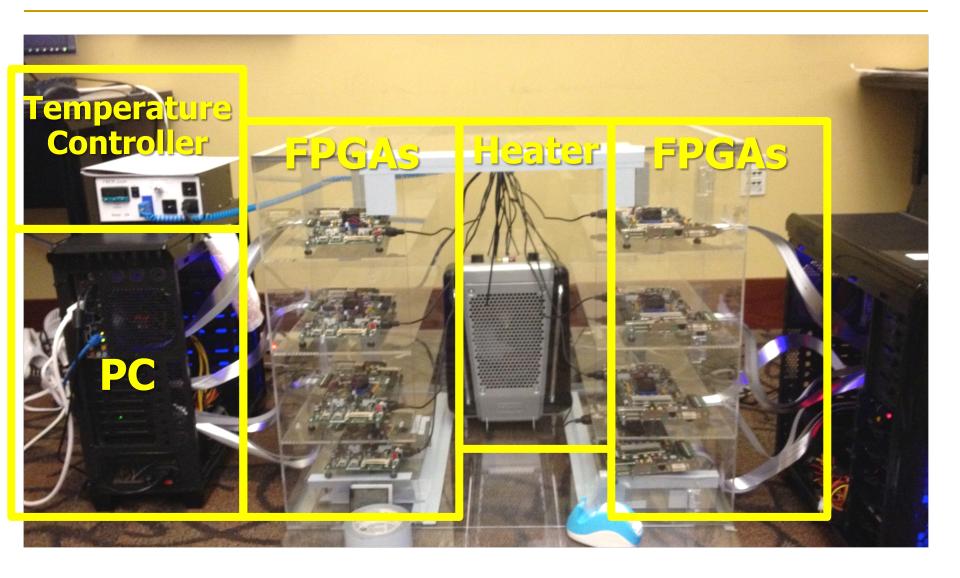
Aware Refresh for DRAM Systems (Qureshi et al., DSN 2015)

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

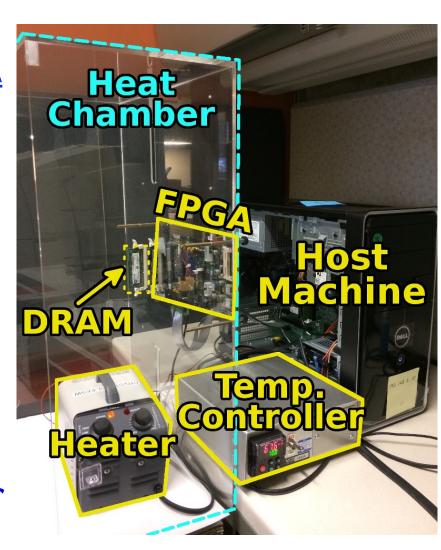


SoftMC: Open Source DRAM Infrastructure

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



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



SoftMC

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

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

```
 Hasan Hassan Nandita Vijaykumar Samira Khan Saugata Ghose Kevin Chang Gennady Pekhimenko Donghyuk Lee Gennady Pekhimenko Onur Mutlu Nandita Vijaykumar Samira Khan Saugata Ghose Kevin Chang Gennady Pekhimenko Onur Mutlu Nandita Vijaykumar Samira Khan Saugata Ghose Nandita Vijaykumar Samira Khan Saugata Ghose Nandita Vijaykumar Samira Khan Saugata Ghose Kevin Chang Gennady Pekhimenko Onur Mutlu Nandita Vijaykumar Samira Khan Saugata Ghose Nandita Vijaykumar Nandita Vija
```

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

A Curious Discovery [Kim et al., ISCA 2014]

One can predictably induce errors in most DRAM memory chips

DRAM RowHammer

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



Forget Software—Now Hackers Are Exploiting Physics

BUSINESS CULTURE DESIGN GEAR SCIENCE

SHARE





ANDY GREENBERG SECURITY 08.31.16 7:00 AM

FORGET SOFTWARE—NOW HACKERS ARE EXPLOITING PHYSICS

One Can Take Over an Otherwise-Secure System

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

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

Project Zero

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

News and updates from the Project Zero team at Google

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

Monday, March 9, 2015

Exploiting the DRAM rowhammer bug to gain kernel privileges



First RowHammer Analysis

Yoongu Kim, Ross Daly, Jeremie Kim, Chris Fallin, Ji Hye Lee, Donghyuk Lee, Chris Wilkerson, Konrad Lai, and Onur Mutlu, "Flipping Bits in Memory Without Accessing Them: An Experimental Study of DRAM Disturbance Errors"

Proceedings of the 41st International Symposium on Computer Architecture (ISCA), Minneapolis, MN, June 2014.

[Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)] [Source Code and Data]

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

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

Carnegie Mellon University ²Intel Labs

Future of Memory Reliability/Security

Onur Mutlu,
 "The RowHammer Problem and Other Issues We May Face as Memory Becomes Denser"
 Invited Paper in Proceedings of the Design, Automation, and Test in 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

A More Recent RowHammer Retrospective

Onur Mutlu and Jeremie Kim,

"RowHammer: A Retrospective"

<u>IEEE Transactions on Computer-Aided Design of Integrated Circuits and Systems</u> (**TCAD**) Special Issue on Top Picks in Hardware and Embedded Security, 2019.

[Preliminary arXiv version]

[Slides from COSADE 2019 (pptx)]

[Slides from VLSI-SOC 2020 (pptx) (pdf)]

[Talk Video (30 minutes)]

RowHammer: A Retrospective

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

RowHammer in 2020

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^{\S \dagger} Minesh Patel^{\S} A. Giray Yağlıkçı^{\S} Hasan Hassan^{\S} Roknoddin Azizi^{\S} Lois Orosa^{\S} Onur Mutlu^{\S \dagger} ^{\S} ETH Zürich ^{\dagger} Carnegie Mellon University
```

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 Privacy</u> (**S&P**), San Francisco, CA, USA, May 2020.

[Slides (pptx) (pdf)]

[Lecture Slides (pptx) (pdf)]

[Talk Video (17 minutes)]

[Lecture Video (59 minutes)]

[Source Code]

[Web Article]

Best paper award.

Pwnie Award 2020 for Most Innovative Research. Pwnie Awards 2020

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

¶Oualcomm Technologies Inc.

RowHammer is still an open problem

Security by obscurity is likely not a good solution

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> <u>Privacy</u> (**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

BlockHammer Solution in 2021

 A. Giray Yaglikci, Minesh Patel, Jeremie S. Kim, Roknoddin Azizi, Ataberk Olgun, Lois Orosa, Hasan Hassan, Jisung Park, Konstantinos Kanellopoulos, Taha Shahroodi, Saugata Ghose, and Onur Mutlu,

"BlockHammer: Preventing RowHammer at Low Cost by Blacklisting Rapidly-Accessed DRAM Rows"

Proceedings of the <u>27th International Symposium on High-Performance</u> <u>Computer Architecture</u> (**HPCA**), Virtual, February-March 2021.

[Slides (pptx) (pdf)]

[Short Talk Slides (pptx) (pdf)]

[Talk Video (22 minutes)]

[Short Talk Video (7 minutes)]

BlockHammer: Preventing RowHammer at Low Cost by Blacklisting Rapidly-Accessed DRAM Rows

A. Giray Yağlıkçı¹ Minesh Patel¹ Jeremie S. Kim¹ Roknoddin Azizi¹ Ataberk Olgun¹ Lois Orosa¹ Hasan Hassan¹ Jisung Park¹ Konstantinos Kanellopoulos¹ Taha Shahroodi¹ Saugata Ghose² Onur Mutlu¹

¹ETH Zürich ²University of Illinois at Urbana–Champaign

Detailed Lectures on RowHammer

- Computer Architecture, Fall 2020, Lecture 4b
 - RowHammer (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=KDy632z23UE&list=PL5Q2soXY2Zi9xidyIgBxUz 7xRPS-wisBN&index=8
- Computer Architecture, Fall 2020, Lecture 5a
 - RowHammer in 2020: TRRespass (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=pwRw7QqK_qA&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=9
- Computer Architecture, Fall 2020, Lecture 5b
 - RowHammer in 2020: Revisiting RowHammer (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=gR7XR-Eepcg&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=10
- Computer Architecture, Fall 2020, Lecture 5c
 - Secure and Reliable Memory (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=HvswnsfG3oQ&list=PL5Q2soXY2Zi9xidyIgBxUz 7xRPS-wisBN&index=11

The Story of RowHammer Lecture ...

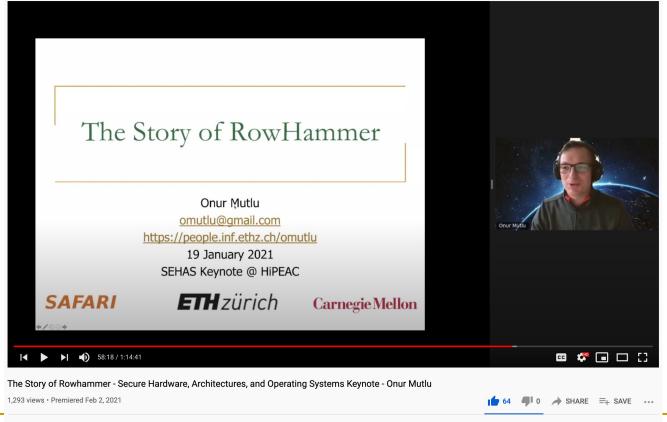
Onur Mutlu,

"The Story of RowHammer"

Keynote Talk at <u>Secure Hardware, Architectures, and Operating Systems</u>
<u>Workshop</u> (**SeHAS**), held with <u>HiPEAC 2021 Conference</u>, Virtual, 19 January 2021.

[Slides (pptx) (pdf)]

[Talk Video (1 hr 15 minutes, with Q&A)]







Understanding Flash Memory & SSD Reliability



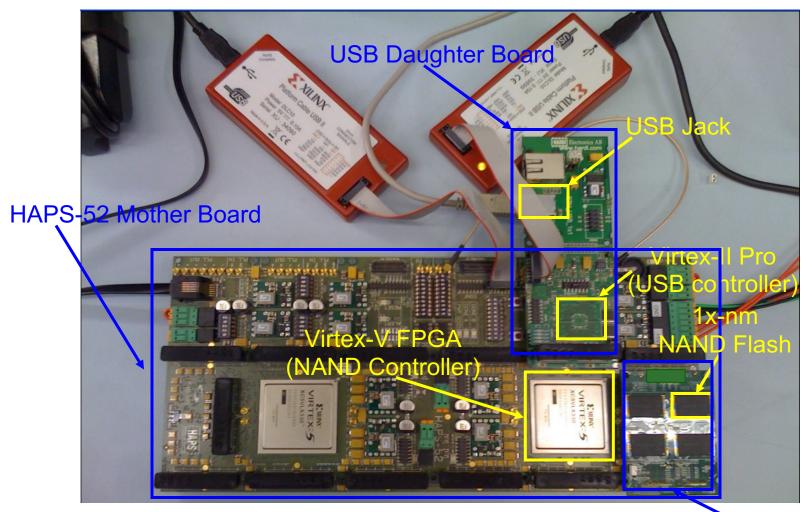
Proceedings of the IEEE, Sept. 2017

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

Understand and Model with Experiments (Flash)



[DATE 2012, ICCD 2012, DATE 2013, ITJ 2013, ICCD 2013, SIGMETRICS 2014, HPCA 2015, DSN 2015, MSST 2015, JSAC 2016, HPCA 2017, DFRWS 2017, PIEEE 2017, HPCA 2018, SIGMETRICS 2018]

NAND Daughter Board

Main Memory Needs Intelligent Controllers

Another Challenge and Opportunity

High Performance, Energy Efficient, Sustainable

Processing of data is performed far away from the data

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 Boroumand¹ Rachata Ausavarungnirun¹ Aki Kuusela³ Allan Knies³

Saugata Ghose¹ Youngsok Kim²

Eric Shiu³ Rahul Thakur³ Daehyun Kim^{4,3}

Parthasarathy Ranganathan³ Onur Mutlu^{5,1}



The Problem

Data access is the major performance and energy bottleneck

Our current design principles cause great energy waste

(and great performance loss)

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

Challenge and Opportunity for Future

Computing Architectures with Minimal Data Movement

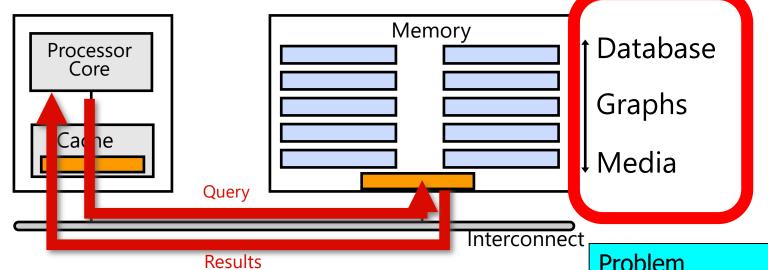
Challenge and Opportunity for Future

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

Challenge and Opportunity for Future

Fundamentally High-Performance (Data-Centric) Computing Architectures

Goal: Processing Inside Memory



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

Problem

Aigorithm

Program/Language

System Software

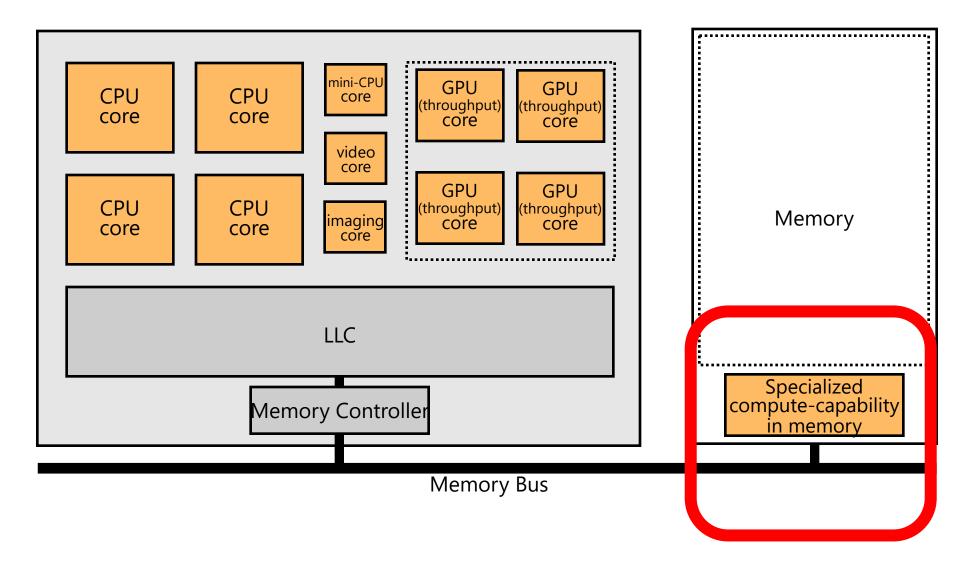
SW/HW Interface

Micro-architecture

Logic

Electrons

Memory as an Accelerator



Memory similar to a "conventional" accelerator

Processing in Memory: Two Approaches

- 1. Processing using Memory
- 2. Processing near Memory

PIM Review and Open Problems

A Modern Primer on Processing in Memory

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

SAFARI Research Group

^aETH Zürich

^bCarnegie Mellon University

^cUniversity of Illinois at Urbana-Champaign

^dKing Mongkut's University of Technology North Bangkok

Onur Mutlu, Saugata Ghose, Juan Gomez-Luna, and Rachata Ausavarungnirun,

"A Modern Primer on Processing in Memory"

Invited Book Chapter in Emerging Computing: From Devices to Systems
Looking Beyond Moore and Von Neumann, Springer, to be published in 2021.

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]

More on Processing in Memory

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

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

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

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

More on Processing in Memory

 Vivek Seshadri and Onur Mutlu, "In-DRAM Bulk Bitwise Execution Engine"

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

More on Processing in Memory (II)

Nastaran Hajinazar, Geraldo F. Oliveira, Sven Gregorio, Joao Dinis Ferreira, Nika Mansouri Ghiasi, Minesh Patel, Mohammed Alser, Saugata Ghose, Juan Gomez-Luna, and Onur Mutlu, "SIMDRAM: An End-to-End Framework for Bit-Serial SIMD Computing in DRAM" Proceedings of the 26th International Conference on Architectural Support for Programming Languages and Operating Systems (ASPLOS), Virtual, March-April 2021.

[2-page Extended Abstract]

[Short Talk Slides (pptx) (pdf)]

[Talk Slides (pptx) (pdf)]

[Short Talk Video (5 mins)]

[Full Talk Video (27 mins)]

SIMDRAM: A Framework for Bit-Serial SIMD Processing using DRAM

*Nastaran Hajinazar^{1,2} Nika Mansouri Ghiasi¹ *Geraldo F. Oliveira¹
Minesh Patel¹
Juan Gómez-Luna¹

Sven Gregorio¹ Mohammed Alser¹ Onur Mutlu¹

João Dinis Ferreira¹ Saugata Ghose³

¹ETH Zürich

²Simon Fraser University

³University of Illinois at Urbana–Champaign

More on Processing in Memory (III)

 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

More on Processing in Memory (IV)

 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 Boroumand¹ Saugata Ghose¹ Youngsok Kim² Rachata Ausavarungnirun¹ Eric Shiu³ Rahul Thakur³ Daehyun Kim^{4,3} Aki Kuusela³ Allan Knies³ Parthasarathy Ranganathan³ Onur Mutlu^{5,1}

More on Processing in Memory (V)

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

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

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

Seoul National University [†]Carnegie Mellon University

In-DRAM Physical Unclonable Functions

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

Proceedings of the <u>24th International Symposium on High-Performance Computer</u> <u>Architecture</u> (**HPCA**), Vienna, Austria, February 2018.

[Lightning Talk Video]

[Slides (pptx) (pdf)] [Lightning Session Slides (pptx) (pdf)]

[Full Talk Lecture Video (28 minutes)]

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

In-DRAM True Random Number Generation

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 Computer</u> <u>Architecture</u> (**HPCA**), Washington, DC, USA, February 2019.

[Slides (pptx) (pdf)]

[Full Talk Video (21 minutes)]

[Full Talk Lecture Video (27 minutes)]

Top Picks Honorable Mention by IEEE Micro.

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

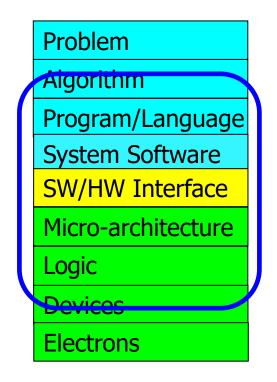
Jeremie S. Kim^{‡§} Minesh Patel[§] Hasan Hassan[§] Lois Orosa[§] Onur Mutlu^{§‡} [‡]Carnegie Mellon University [§]ETH Zürich

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Eliminating the Adoption Barriers

How to Enable Adoption of Processing in Memory

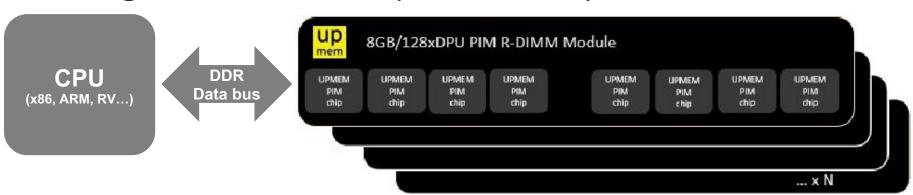
We Need to Revisit the Entire Stack



We can get there step by step

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





UPMEM Memory Modules

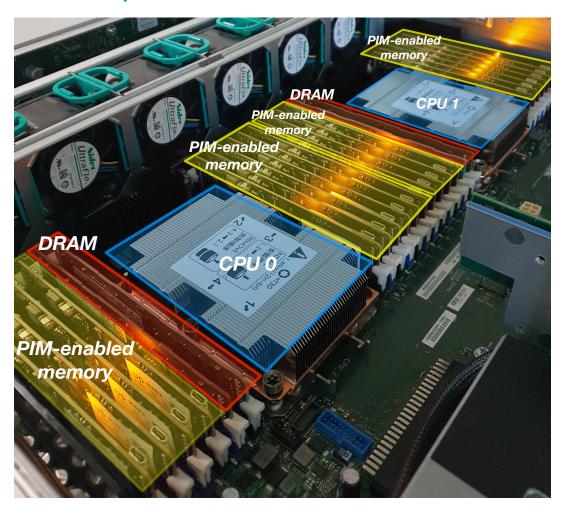
- E19: 8 chips DIMM (1 rank). DPUs @ 267 MHz
- P21: 16 chips DIMM (2 ranks). DPUs @ 350 MHz





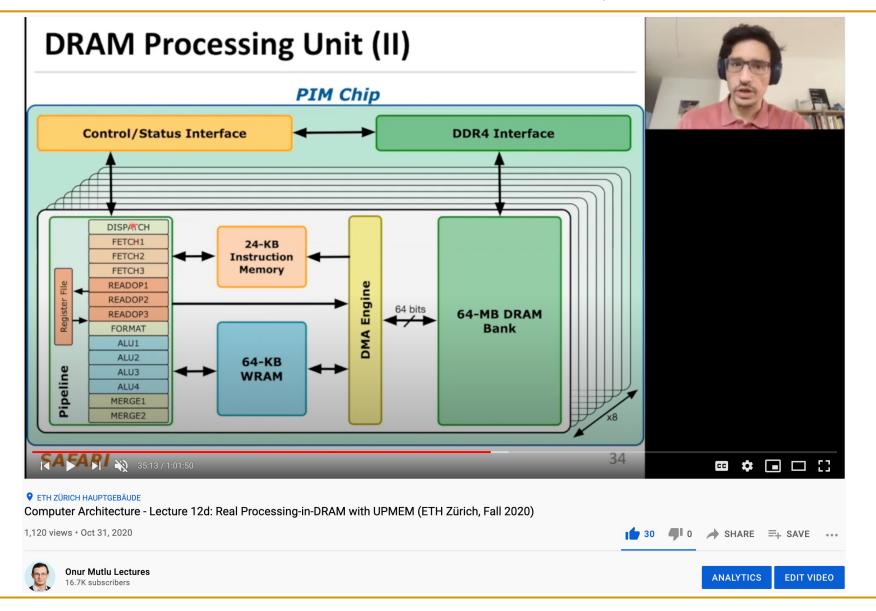
PIM System Organization

 UPMEM-based PIM system with 20 UPMEM memory modules of 16 chips each (40 ranks) → 2560 DPUs





More on the UPMEM PIM System



Experimental Analysis of the UPMEM PIM Engine

Benchmarking a New Paradigm: An Experimental Analysis of a Real Processing-in-Memory Architecture

JUAN GÓMEZ-LUNA, ETH Zürich, Switzerland IZZAT EL HAJJ, American University of Beirut, Lebanon IVAN FERNANDEZ, ETH Zürich, Switzerland and University of Malaga, Spain CHRISTINA GIANNOULA, ETH Zürich, Switzerland and NTUA, Greece GERALDO F. OLIVEIRA, ETH Zürich, Switzerland ONUR MUTLU, ETH Zürich, Switzerland

Many modern workloads, such as neural networks, databases, and graph processing, are fundamentally memory-bound. For such workloads, the data movement between main memory and CPU cores imposes a significant overhead in terms of both latency and energy. A major reason is that this communication happens through a narrow bus with high latency and limited bandwidth, and the low data reuse in memory-bound workloads is insufficient to amortize the cost of main memory access. Fundamentally addressing this *data movement bottleneck* requires a paradigm where the memory system assumes an active role in computing by integrating processing capabilities. This paradigm is known as *processing-in-memory (PIM)*.

Recent research explores different forms of PIM architectures, motivated by the emergence of new 3D-stacked memory technologies that integrate memory with a logic layer where processing elements can be easily placed. Past works evaluate these architectures in simulation or, at best, with simplified hardware prototypes. In contrast, the UPMEM company has designed and manufactured the first publicly-available real-world PIM architecture. The UPMEM PIM architecture combines traditional DRAM memory arrays with general-purpose in-order cores, called *DRAM Processing Units* (*DPUs*), integrated in the same chip.

This paper provides the first comprehensive analysis of the first publicly-available real-world PIM architecture. We make two key contributions. First, we conduct an experimental characterization of the UPMEM-based PIM system using microbenchmarks to assess various architecture limits such as compute throughput and memory bandwidth, yielding new insights. Second, we present *PrIM* (*Processing-In-Memory benchmarks*), a benchmark suite of 16 workloads from different application domains (e.g., dense/sparse linear algebra, databases, data analytics, graph processing, neural networks, bioinformatics, image processing), which we identify as memory-bound. We evaluate the performance and scaling characteristics of PrIM benchmarks on the UPMEM PIM architecture, and compare their performance and energy consumption to their state-of-the-art CPU and GPU counterparts. Our extensive evaluation conducted on two real UPMEM-based PIM systems with 640 and 2,556 DPUs provides new insights about suitability of different workloads to the PIM system, programming recommendations for software designers, and suggestions and hints for hardware and architecture designers of future PIM systems.

https://arxiv.org/pdf/2105.03814.pdf

DAMOV Methodology & Workloads

DAMOV: A New Methodology and Benchmark Suite for Evaluating Data Movement Bottlenecks

GERALDO F. OLIVEIRA, ETH Zürich, Switzerland
JUAN GÓMEZ-LUNA, ETH Zürich, Switzerland
LOIS OROSA, ETH Zürich, Switzerland
SAUGATA GHOSE, University of Illinois at Urbana-Champaign, USA
NANDITA VIJAYKUMAR, University of Toronto, Canada
IVAN FERNANDEZ, University of Malaga, Spain & ETH Zürich, Switzerland
MOHAMMAD SADROSADATI, Institute for Research in Fundamental Sciences (IPM), Iran & ETH
Zürich, Switzerland
ONUR MUTLU, ETH Zürich, Switzerland

Data movement between the CPU and main memory is a first-order obstacle against improving performance, scalability, and energy efficiency in modern systems. Computer systems employ a range of techniques to reduce overheads tied to data movement, spanning from traditional mechanisms (e.g., deep multi-level cache hierarchies, aggressive hardware prefetchers) to emerging techniques such as Near-Data Processing (NDP), where some computation is moved close to memory. Prior NDP works investigate the root causes of data movement bottlenecks using different profiling methodologies and tools. However, there is still a lack of understanding about the key metrics that can identify different data movement bottlenecks and their relation to traditional and emerging data movement mitigation mechanisms. Our goal is to methodically identify potential sources of data movement over a broad set of applications and to comprehensively compare traditional compute-centric data movement mitigation techniques (e.g., caching and prefetching) to more memory-centric techniques (e.g., NDP), thereby developing a rigorous understanding of the best techniques to mitigate each source of data movement.

With this goal in mind, we perform the first large-scale characterization of a wide variety of applications, across a wide range of application domains, to identify fundamental program properties that lead to data movement to/from main memory. We develop the first systematic methodology to classify applications based on the sources contributing to data movement bottlenecks. From our large-scale characterization of 77K functions across 345 applications, we select 144 functions to form the first open-source benchmark suite (DAMOV) for main memory data movement studies. We select a diverse range of functions that (1) represent different types of data movement bottlenecks, and (2) come from a wide range of application domains. Using NDP as a case study, we identify new insights about the different data movement bottlenecks and use these insights to determine the most suitable data movement mitigation mechanism for a particular application. We open-source DAMOV and the complete source code for our new characterization methodology at https://github.com/CMU-SAFARI/DAMOV.

SAFARI

https://arxiv.org/pdf/2105.03725.pdf

Detailed Lectures on PIM (I)

- Computer Architecture, Fall 2020, Lecture 6
 - Computation in Memory (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=oGcZAGwfEUE&list=PL5Q2soXY2Zi9xidyIgBxUz 7xRPS-wisBN&index=12
- Computer Architecture, Fall 2020, Lecture 7
 - Near-Data Processing (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=j2GIigqn1Qw&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=13
- Computer Architecture, Fall 2020, Lecture 11a
 - Memory Controllers (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=TeG773OgiMQ&list=PL5Q2soXY2Zi9xidyIgBxUz 7xRPS-wisBN&index=20
- Computer Architecture, Fall 2020, Lecture 12d
 - Real Processing-in-DRAM with UPMEM (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=Sscy1Wrr22A&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=25

Detailed Lectures on PIM (II)

- Computer Architecture, Fall 2020, Lecture 15
 - Emerging Memory Technologies (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=AlE1rD9G_YU&list=PL5Q2soXY2Zi9xidyIgBxUz 7xRPS-wisBN&index=28
- Computer Architecture, Fall 2020, Lecture 16a
 - Opportunities & Challenges of Emerging Memory Technologies
 (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=pmLszWGmMGQ&list=PL5Q2soXY2Zi9xidyIgBx Uz7xRPS-wisBN&index=29
- Computer Architecture, Fall 2020, Guest Lecture
 - In-Memory Computing: Memory Devices & Applications (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=wNmqQHiEZNk&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=41

A Tutorial on PIM

Onur Mutlu,

"Memory-Centric Computing Systems"

Invited Tutorial at <u>66th International Electron Devices</u>

Meeting (IEDM), Virtual, 12 December 2020.

[Slides (pptx) (pdf)]

[Executive Summary Slides (pptx) (pdf)]

[Tutorial Video (1 hour 51 minutes)]

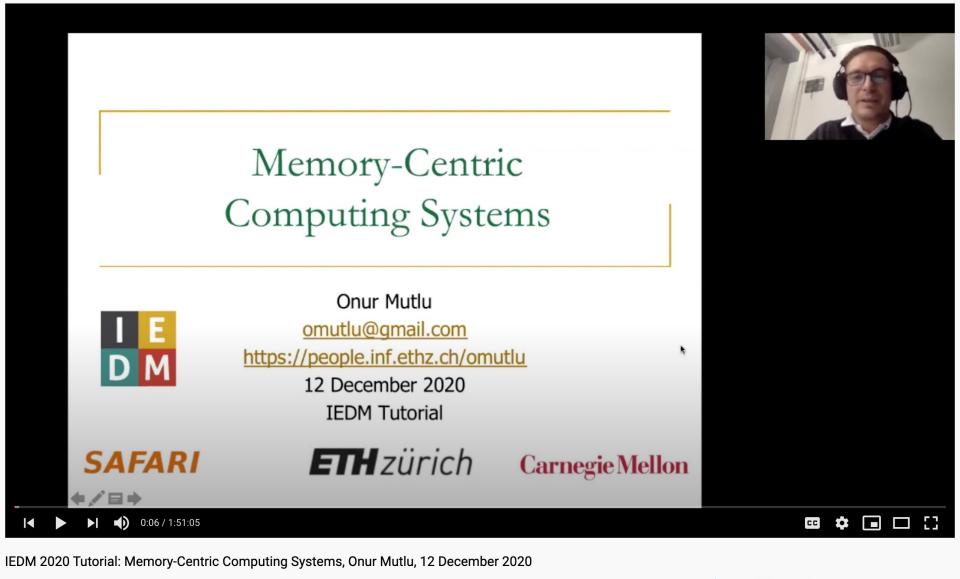
[Executive Summary Video (2 minutes)]

[Abstract and Bio]

[Related Keynote Paper from VLSI-DAT 2020]

[Related Review Paper on Processing in Memory]

https://www.youtube.com/watch?v=H3sEaINPBOE



1,641 views • Dec 23, 2020 ♣ SHARE =+ SAVE •



ANALYTICS

EDIT VIDEO

PIM Can Enable New Medical Platforms

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/bby017

Published: 02 April 2018 Article history ▼

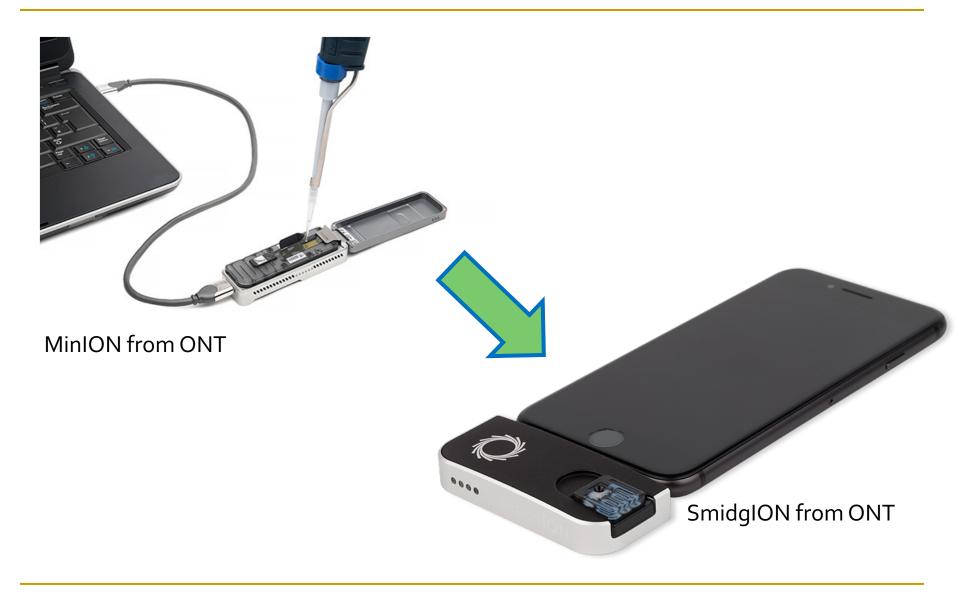


Oxford Nanopore MinION

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

[Preliminary arxiv.org version]

Future of Genome Sequencing & Analysis



Accelerating Genome Analysis: Overview

 Mohammed Alser, Zulal Bingol, Damla Senol Cali, Jeremie Kim, Saugata Ghose, Can Alkan, and Onur Mutlu,

"Accelerating Genome Analysis: A Primer on an Ongoing Journey"

IEEE Micro (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

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

More on Fast Genome Analysis ...

Onur Mutlu,

"Accelerating Genome Analysis: A Primer on an Ongoing Journey" Invited Lecture at <u>Technion</u>, Virtual, 26 January 2021.

[Slides (pptx) (pdf)]

[Talk Video (1 hour 37 minutes, including Q&A)]

[Related Invited Paper (at IEEE Micro, 2020)]





Detailed Lectures on Genome Analysis

- Computer Architecture, Fall 2020, Lecture 3a
 - Introduction to Genome Sequence Analysis (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=CrRb32v7SJc&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=5
- Computer Architecture, Fall 2020, Lecture 8
 - Intelligent Genome Analysis (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=ygmQpdDTL7o&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=14
- Computer Architecture, Fall 2020, Lecture 9a
 - □ **GenASM: Approx. String Matching Accelerator** (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=XoLpzmN Pas&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=15
- Accelerating Genomics Project Course, Fall 2020, Lecture 1
 - Accelerating Genomics (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=rgjl8ZyLsAg&list=PL5Q2soXY2Zi9E2bBVAgCqL gwiDRQDTyId

Challenge and Opportunity for Future

Fundamentally Low-Latency Computing Architectures

Truly Reducing Memory Latency

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 <u>19th International Symposium on High-</u> <u>Performance Computer Architecture</u> (**HPCA**), Shenzhen, China, February 2013. <u>Slides (pptx)</u>

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

Adaptive-Latency DRAM

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

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

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

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

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

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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> Modeling of Computer Systems (SIGMETRICS), Antibes Juan-Les-Pins, France, June 2016.

[Slides (pptx) (pdf)]

Source Code

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

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

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

Design-Induced Latency Variation in DRAM

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

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

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

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

Solar-DRAM: Putting It Together

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 36th IEEE International Conference on
 Computer Design (ICCD), Orlando, FL, USA, October 2018.
 [Slides (pptx) (pdf)]
 [Talk Video (16 minutes)]

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

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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> Architecture (ISCA), Valencia, Spain, June 2020.

[Slides (pptx) (pdf)]

[Lightning Talk Slides (pptx) (pdf)]

[Talk Video (20 minutes)]

[Lightning Talk Video (3 minutes)]

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

Taha Shahroodi[§] Hasan Hassan[§] Haocong Luo§† Minesh Patel§ A. Giray Yağlıkçı[§] Lois Orosa[§] Jisung Park[§] Onur Mutlu[§]

§ETH Zürich †ShanghaiTech University



Low-Latency Solid-State Drives (SSDs)

 Jisung Park, Myungsuk Kim, Myoungjun Chun, Lois Orosa, Jihong Kim, and Onur Mutlu,

"Reducing Solid-State Drive Read Latency by Optimizing Read-Retry"

Proceedings of the <u>26th International Conference on Architectural Support for Programming Languages and Operating Systems</u> (ASPLOS), Virtual, March-April 2021.

[2-page Extended Abstract]

[Short Talk Slides (pptx) (pdf)]

[Full Talk Slides (pptx) (pdf)]

[Short Talk Video (5 mins)]

[Full Talk Video (19 mins)]

Reducing Solid-State Drive Read Latency by Optimizing Read-Retry

Jisung Park¹ Myungsuk Kim^{2,3} Myoungjun Chun² Lois Orosa¹ Jihong Kim² Onur Mutlu¹

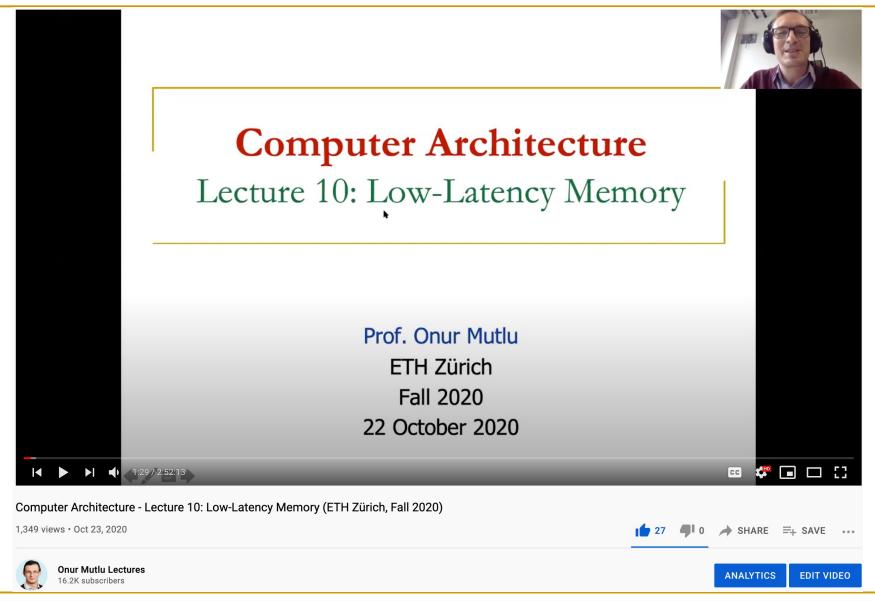
¹ETH Zürich Switzerland ²Seoul National University Republic of Korea

³Kyungpook National University Republic of Korea

Lectures on Low-Latency Memory

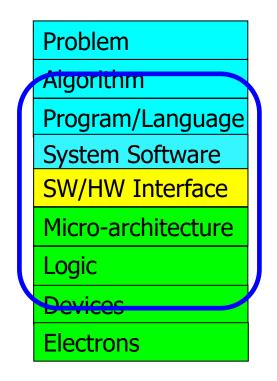
- Computer Architecture, Fall 2020, Lecture 10
 - Low-Latency Memory (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=vQd1YgOH1Mw&list=PL5Q2soXY2Zi9xidyIgBx Uz7xRPS-wisBN&index=19
- Computer Architecture, Fall 2020, Lecture 12b
 - Capacity-Latency Reconfigurable DRAM (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=DUtPFW3jxq4&list=PL5Q2soXY2Zi9xidyIgBxUz 7xRPS-wisBN&index=23
- Computer Architecture, Fall 2019, Lecture 11a
 - DRAM Latency PUF (ETH Zürich, Fall 2019)
 - https://www.youtube.com/watch?v=7gqnrTZpjxE&list=PL5Q2soXY2Zi-DyoI3HbqcdtUm9YWRR_z-&index=15
- Computer Architecture, Fall 2019, Lecture 11b
 - DRAM True Random Number Generator (ETH Zürich, Fall 2020)
 - https://www.youtube.com/watch?v=Y3hPv1I5f8Y&list=PL5Q2soXY2Zi-DyoI3HbqcdtUm9YWRR_z-&index=16

A Tutorial on Low-Latency Memory



https://www.youtube.com/onurmutlulectures

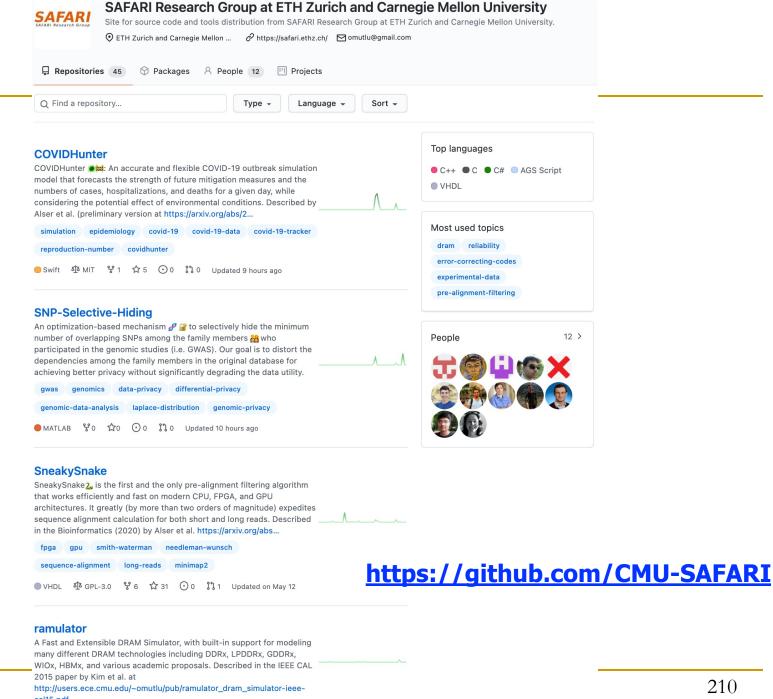
We Need to Revisit the Entire Stack



We can get there step by step

Open-Source Artifacts

https://github.com/CMU-SAFARI



SAFARI

Some Open Source Tools (I)

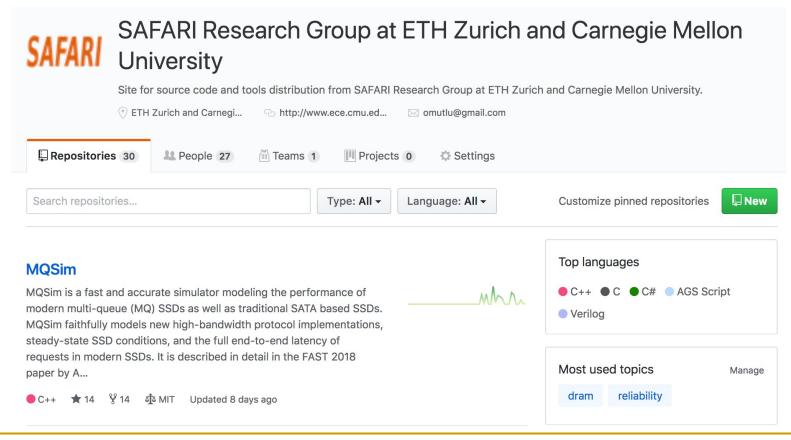
- Rowhammer Program to Induce RowHammer Errors
 - https://github.com/CMU-SAFARI/rowhammer
- Ramulator Fast and Extensible DRAM Simulator
 - https://github.com/CMU-SAFARI/ramulator
- MemSim Simple Memory Simulator
 - https://github.com/CMU-SAFARI/memsim
- NOCulator Flexible Network-on-Chip Simulator
 - https://github.com/CMU-SAFARI/NOCulator
- SoftMC FPGA-Based DRAM Testing Infrastructure
 - https://github.com/CMU-SAFARI/SoftMC
- Other open-source software from my group
 - https://github.com/CMU-SAFARI/
 - http://www.ece.cmu.edu/~safari/tools.html

Some Open Source Tools (II)

- MQSim A Fast Modern SSD Simulator
 - https://github.com/CMU-SAFARI/MQSim
- 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
 - https://github.com/CMU-SAFARI/HWASim
- Other open-source software from my group
 - https://github.com/CMU-SAFARI/
 - http://www.ece.cmu.edu/~safari/tools.html

More Open Source Tools (III)

- A lot more open-source software from my group
 - https://github.com/CMU-SAFARI/



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...

●C++ គ្ MIT ೪ 54 ☆62 ①10 រឿ 1 Updated on May 15

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...

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++ № MIT ೪ 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

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 4 GPL-3.0 ♀ 3 ☆ 11 ① 0 ; 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 ម្0 ☆3 ①0 រុេ្ 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

End of Slides on More Detailed Research Overview