

Digital Design & Computer Arch.

Lecture 26b: Epilogue

Prof. Onur Mutlu

ETH Zürich

Spring 2021

4 June 2021

**No Required Readings
(for this lecture)**

Recall: Major High-Level Goals of This Course

- In Digital Circuits & Computer Architecture
- Understand the basics
- Understand the principles (of design)
- Understand the precedents
- Based on such understanding:
 - learn how a modern computer works underneath
 - evaluate tradeoffs of different designs and ideas
 - implement a principled design (a simple microprocessor)
 - learn to systematically debug increasingly complex systems
 - Hopefully enable you to develop novel, out-of-the-box designs
- The focus is on basics, principles, precedents, and how to use them to create/implement good designs

Recall: Why These Goals?

- Because you are here for a Computer Science degree
- **Regardless of your future direction**, learning the principles of digital design & computer architecture will be useful to
 - design better hardware
 - design better software
 - design better systems
 - make better tradeoffs in design
 - understand why computers behave the way they do
 - solve problems better
 - think “in parallel”
 - think critically
 - ...

We Have Come A Long Way

- Started from Transistor as the Building Block
- Logic Design (Combinational, Sequential, Timing)
- Execution Model, ISA and Microarchitecture
- Many Key Processing Paradigms
- The Memory System and the Memory Hierarchy
- Built up to Virtual Memory & System Software Mechanisms

- Takeaway 1: All we covered is real and used in real systems
→ and, it is increasingly important
- Takeaway 2: Principles we covered apply broadly
- Takeaway 3: Tradeoff analysis and critical thinking that you are exposed to apply even more broadly

Recall: We Are **Done** With This...

- Single-cycle Microarchitectures
- Multi-cycle and Microprogrammed Microarchitectures
- Pipelining
- Issues in Pipelining: Control & Data Dependence Handling, State Maintenance and Recovery, ...
- Out-of-Order Execution
- Other Execution Paradigms

Approaches to (Instruction-Level) Concurrency

- Pipelining
- Fine-Grained Multithreading
- Out-of-order Execution
- Dataflow (at the ISA level)
- Superscalar Execution
- VLIW
- Systolic Arrays
- Decoupled Access Execute
- SIMD Processing (Vector and Array processors, GPUs)

**Now you are very familiar with
many processing paradigms**

Approaches to (Instruction-Level) Concurrency

- Pipelining
- Fine-Grained Multithreading
- Out-of-order Execution
- Dataflow (at the ISA level)
- Superscalar Execution
- VLIW
- Systolic Arrays
- Decoupled Access Execute
- SIMD Processing (Vector and Array processors, GPUs)

Food for thought:
tradeoffs of these different processing paradigms

We Are **Also Done** With This...

- Memory Organization & Technology
- Memory Hierarchy & Caches
- Advanced Caches
- Prefetching
- Virtual Memory

**Now you are very familiar with
memory systems**

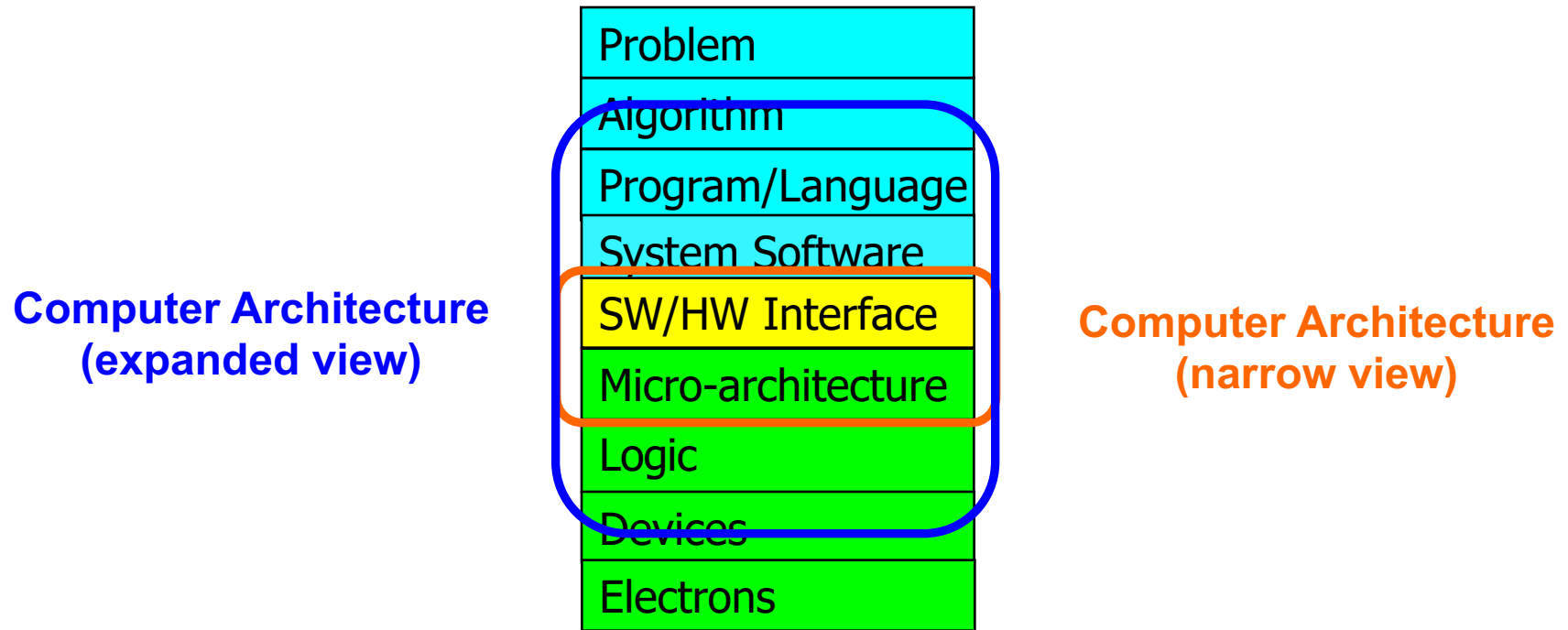
We Are **Also Done** With This...

- Memory Organization & Technology
- Memory Hierarchy & Caches
- Advanced Caches
- Prefetching
- Virtual Memory

Food for thought:

tradeoffs of many different memory system designs & ideas

The Transformation Hierarchy



Food for thought:
how do tradeoffs span and affect the hierarchy

The Best Way to Approach This Course

- Take it as **a learning and growth experience**... all of it
- What we saw changed the world & endured the test of time...
- And, it will be more important...
- You may not all be future architects, but...
 - your development and thinking can greatly benefit from the concepts, tradeoffs, principles, critical thinking...
- **Focus on understanding, learning, critical analysis**
 - these are the agents for your growth
 - the course is designed to activate these agents (lifelong)

What We Did Not Cover

Computer Architecture is Very Rich

- Many ideas, much creativity, many tradeoffs and problems
- As scaling, performance, energy, reliability, security issues become worse in circuits and in software, computer architecture will be more and more important
- Already obvious in
 - AI/ML accelerators
 - Hardware security issues
 - Novel execution paradigms: Processing in memory
 - ...
- See lecture: [Intelligent Architectures for Intelligent Machines](https://www.youtube.com/watch?v=5YKvtNM6XzY)
 - <https://www.youtube.com/watch?v=5YKvtNM6XzY>

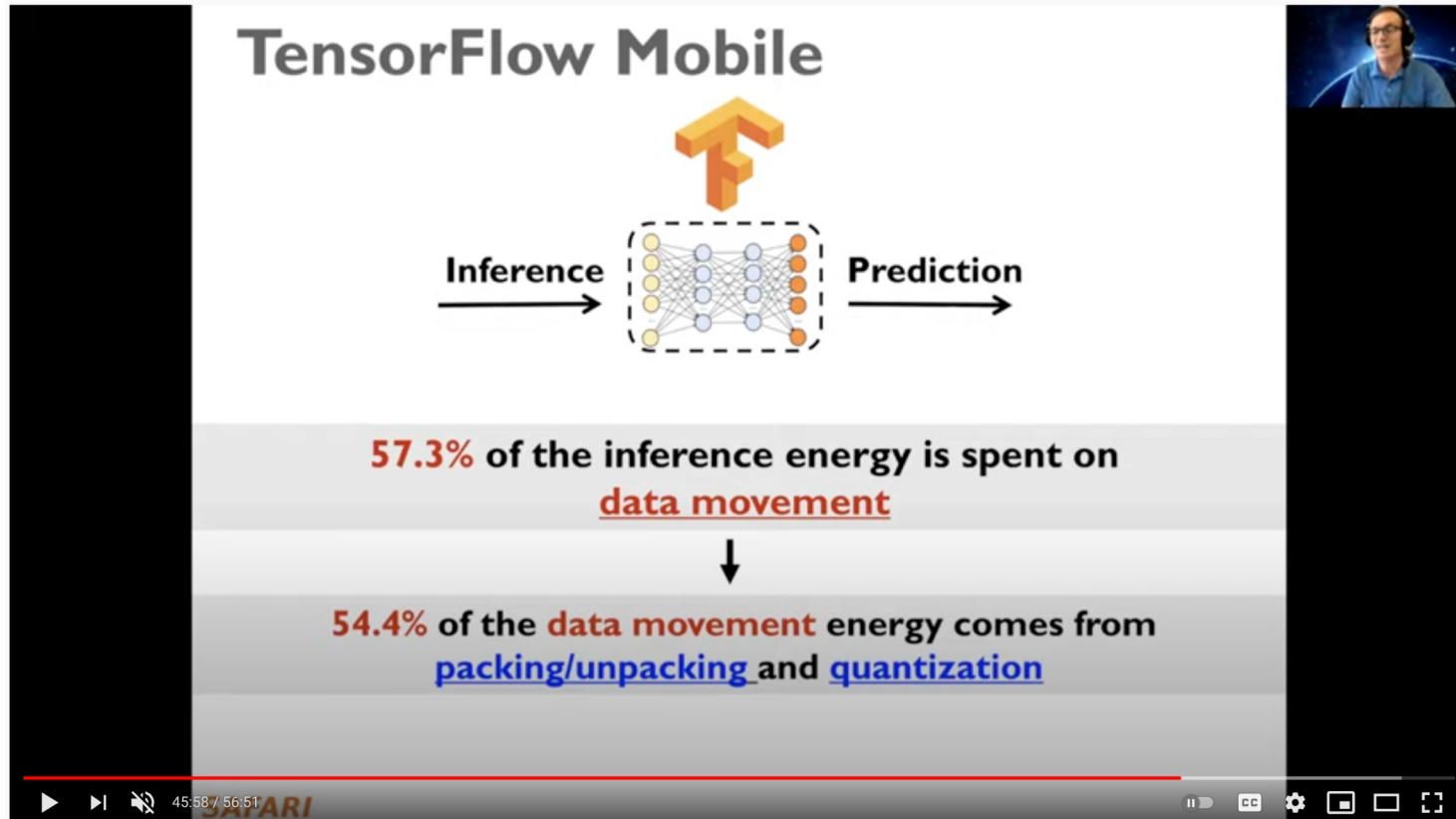
Intelligent Architectures for Intelligent Machines

Onur Mutlu,

["Intelligent Architectures for Intelligent Machines"](#)

Opening Talk at [TU Vienna Mondays in Memory Webinar Series \(MiM\)](#), Virtual, 3 May 2021.

[[Slides \(pptx\)](#)] [[pdf](#)] [[Talk Video](#) (57 minutes)]



MiM Webinar Series Opening Talk - Intelligent Architectures for Intelligent Machines - Onur Mutlu

789 views • Premiered May 23, 2021

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Onur Mutlu Lectures
16.7K subscribers

ANALYTICS

EDIT VIDEO

<https://www.youtube.com/watch?v=5YKvtNM6XzY>

Extra Assignment: Lecture Video Review

- **Intelligent Architectures for Intelligent Machines**
- **Watch this recent lecture and analyze it critically**
 - <https://www.youtube.com/watch?v=5YKvtNM6XzY> (May 2021)
- **Optional Assignment – for 1% extra credit**
 - **Write a 1-page summary** of your analysis
 - What are your key takeaways?
 - What did you learn?
 - What did you like or dislike?
 - Use past review guidelines
 - Submit your summary to [Moodle](#) – Deadline: June 20

A Tutorial on Memory-Centric Systems

- Onur Mutlu,

"Memory-Centric Computing Systems"

Invited Tutorial at *66th International Electron Devices 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>

Memory-Centric Computing Systems



Onur Mutlu

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

12 December 2020

IEDM Tutorial

SAFARI

ETH zürich

Carnegie Mellon



0:06 / 1:51:05



IEDM 2020 Tutorial: Memory-Centric Computing Systems, Onur Mutlu, 12 December 2020

1,641 views • Dec 23, 2020

48 0 SHARE SAVE ...



Onur Mutlu Lectures
13.9K subscribers

ANALYTICS

EDIT VIDEO

<https://www.youtube.com/onurmutlulectures>

Data-centric

Data-driven

Data-aware

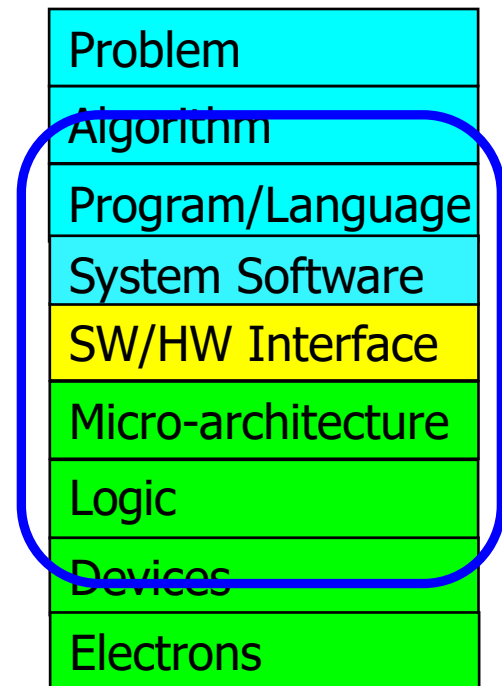


Recall: Takeaways

- It is an exciting time to be understanding and designing computing architectures
- Many challenging and exciting problems in platform design
 - That no one has tackled (or thought about) before
 - That can have huge impact on the world's future
- Driven by huge hunger for data (Big Data), new applications (ML/AI, graph analytics, genomics), ever-greater realism, ...
 - We can easily collect more data than we can analyze/understand
- Driven by significant difficulties in keeping up with that hunger at the technology layer
 - Five walls: Energy, reliability, complexity, security, scalability

State of the Art

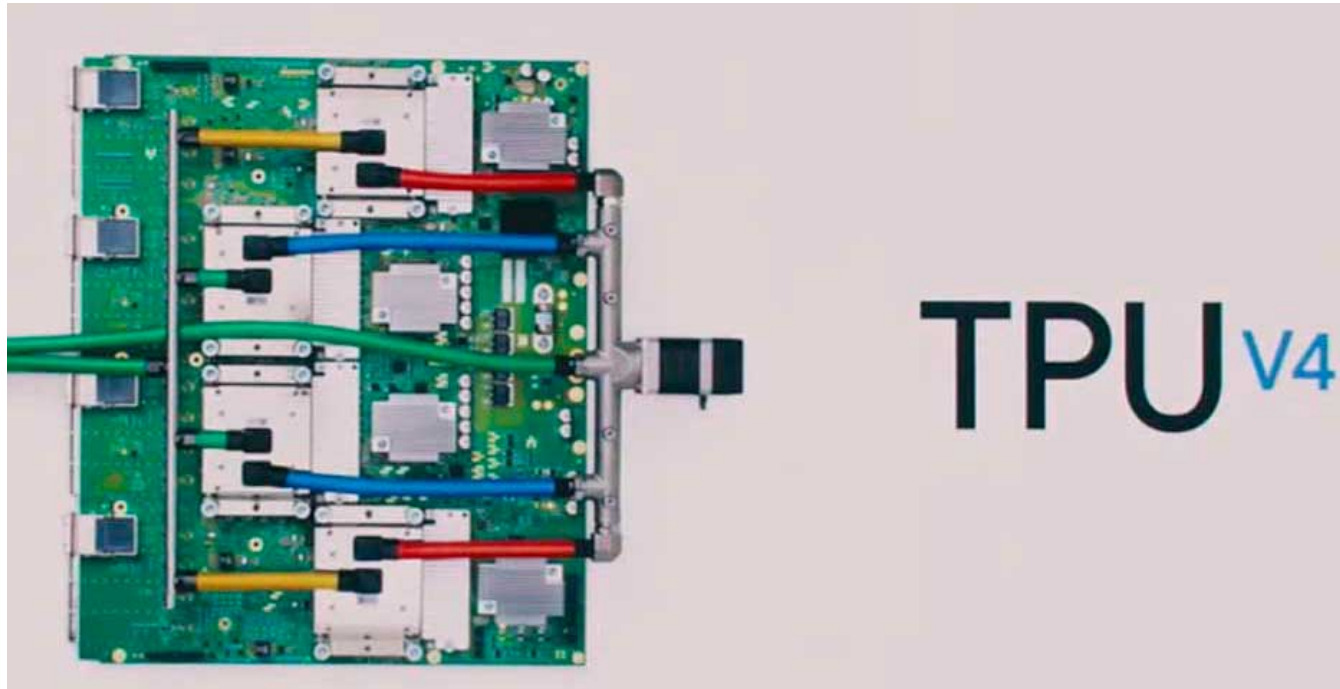
- This is a great time to be a computer architect
- Circuits strained
- Applications ever more demanding
- Multiple possible emerging technologies
- Many requirements, many systems
- Many, many security, reliability issues
- Many big problems waiting to be solved
 - All across the hierarchy
- ...



Many big innovations require computer architecture

Many Interesting Things Are Happening Today in Computer Architecture

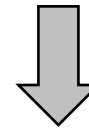
New Generation of Google TPU v4 (2021)



New ML applications (vs TPU3):

- Computer vision
- Natural Language Processing (NLP)
- Recommender system
- Reinforcement learning that plays Go

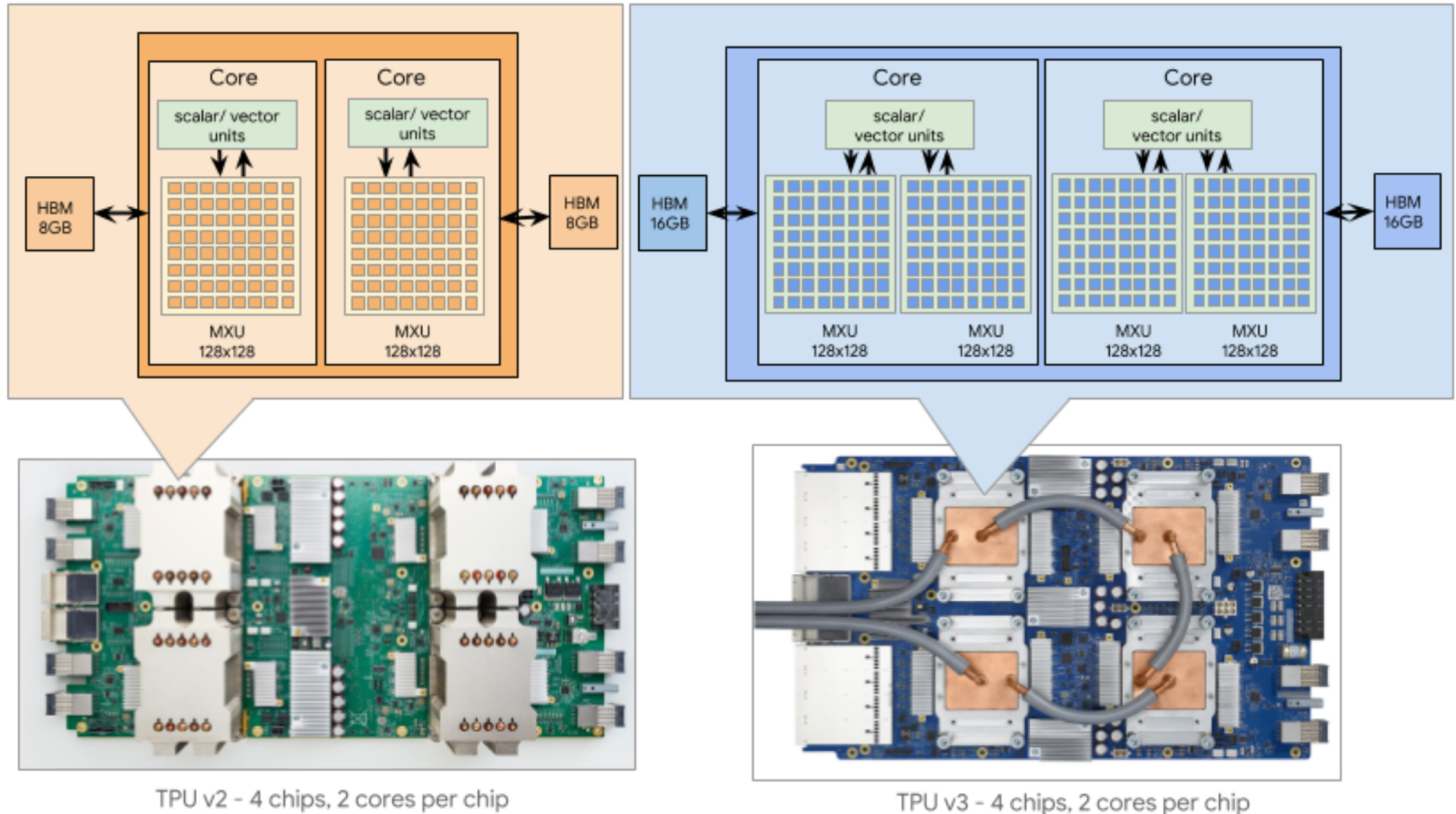
250 TFLOPS per chip
vs 90 TFLOPS in TPU3



1 ExaFLOPS per board

<https://spectrum.ieee.org/tech-talk/computing/hardware/heres-how-googles-tpu-v4-ai-chip-stacked-up-in-training-tests>

An Example Modern Systolic Array: TPU3

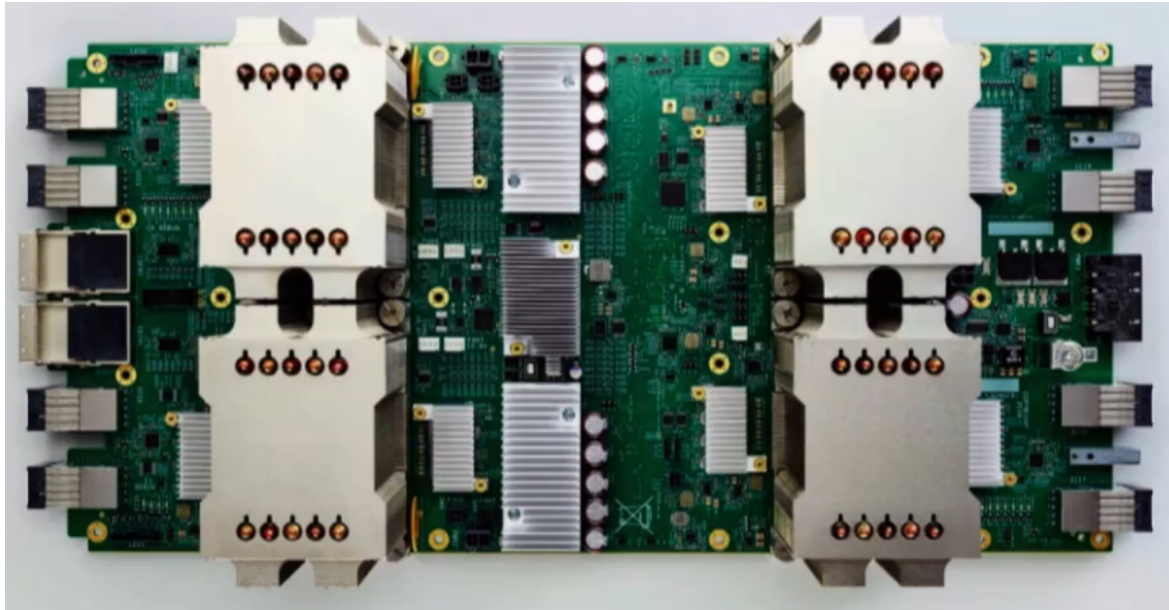


32GB HBM per chip
vs 16GB HBM in TPU2

4 Matrix Units per chip
vs 2 Matrix Units in TPU2

90 TFLOPS per chip
vs 45 TFLOPS in TPU2

An Example Modern Systolic Array: TPU2



<https://www.nextplatform.com/2017/05/17/first-depth-look-googles-new-second-generation-tpu/>

4 TPU chips
vs 1 chip in TPU1

High Bandwidth Memory
vs DDR3

Floating point operations
vs FP16

45 TFLOPS per chip
vs 23 TOPS

Designed for **training**
and **inference**
vs only inference

An Example Modern Systolic Array: TPU (I)

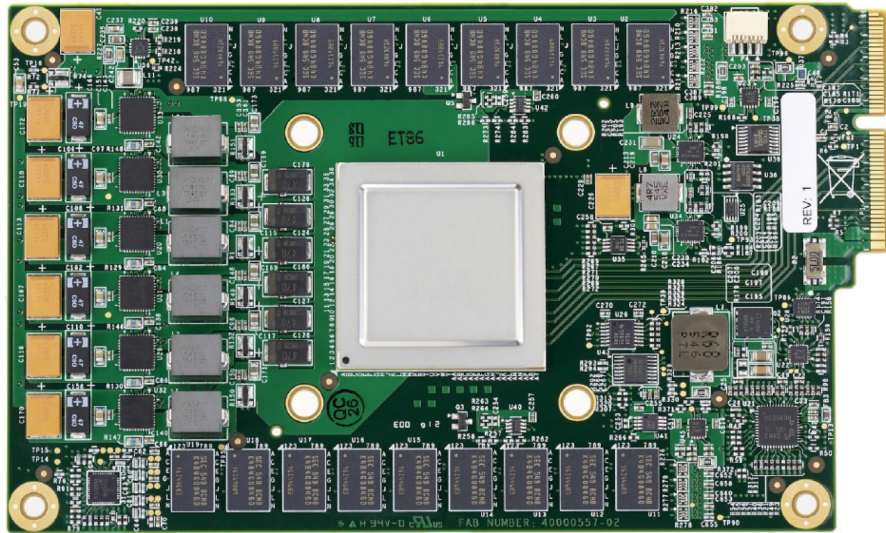


Figure 3. TPU Printed Circuit Board. It can be inserted in the slot for an SATA disk in a server, but the card uses PCIe Gen3 x16.

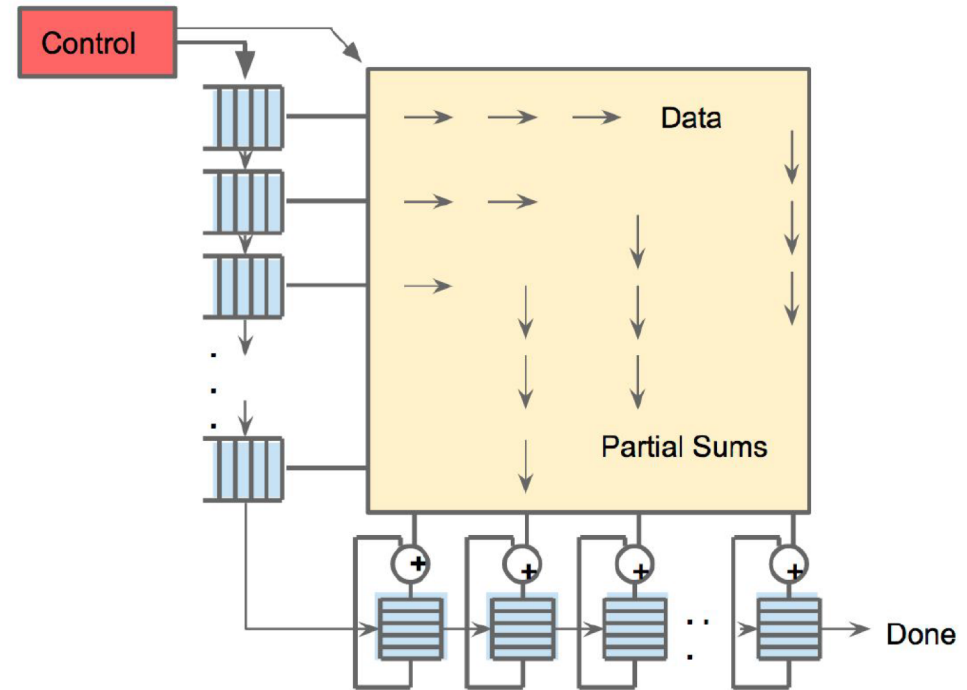
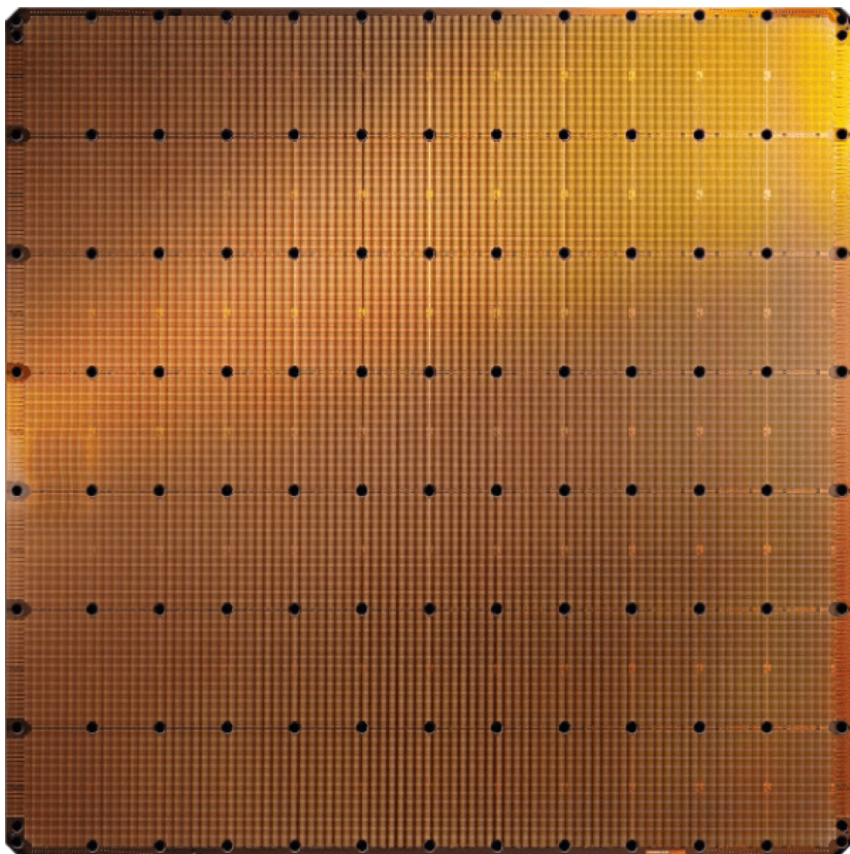


Figure 4. Systolic data flow of the Matrix Multiply Unit. Software has the illusion that each 256B input is read at once, and they instantly update one location of each of 256 accumulator RAMs.

Jouppi et al., “[In-Datcenter Performance Analysis of a Tensor Processing Unit](#)”, ISCA 2017.

Cerebras's Wafer Scale Engine-2 (2021)



Cerebras WSE-2

2.6 Trillion transistors

46,225 mm²

<https://cerebras.net/product/#overview>

- The largest ML accelerator chip
- 850,000 cores
- **40 GB of on-chip memory**
- **20 PB/s memory bandwidth**



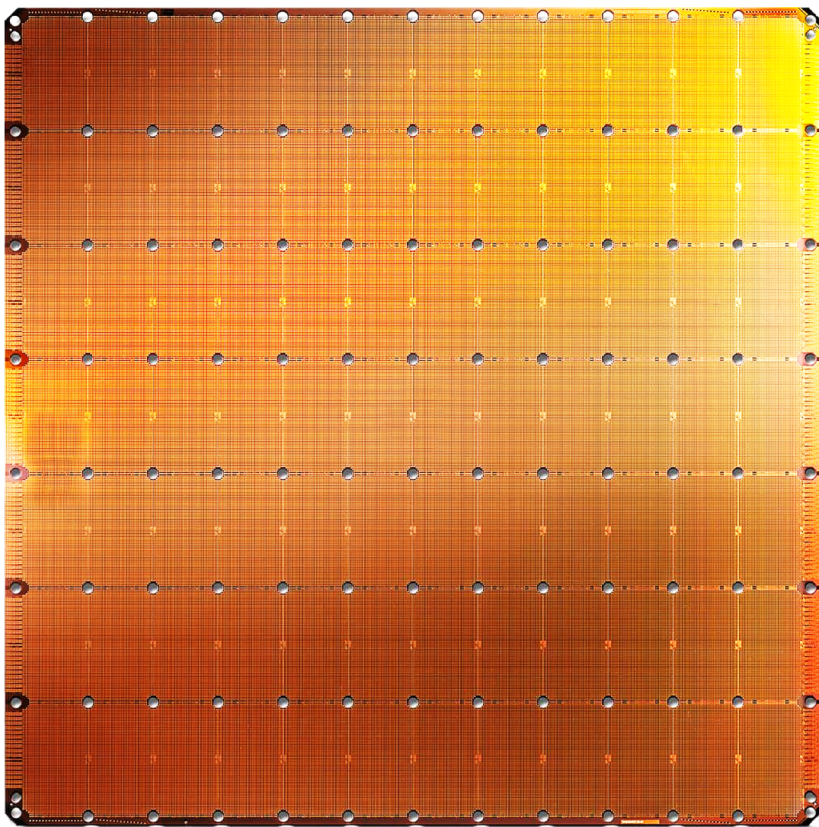
Largest GPU

54.2 Billion transistors

826 mm²

NVIDIA Ampere GA100

Cerebras's Wafer Scale Engine (2019)



Cerebras WSE

1.2 Trillion transistors

46,225 mm²

- The largest ML accelerator chip
- 400,000 cores
- **18 GB of on-chip memory**
- **9 PB/s memory bandwidth**



Largest GPU

21.1 Billion transistors

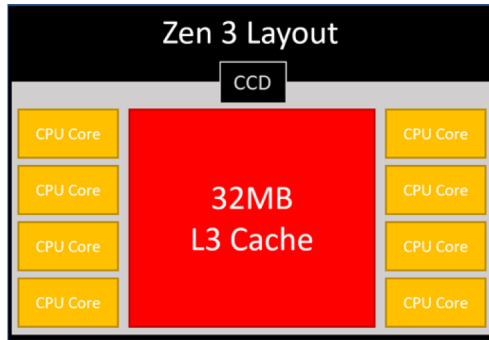
815 mm²

NVIDIA TITAN V

<https://www.anandtech.com/show/14758/hot-chips-31-live-blogs-cerebras-wafer-scale-deep-learning>

<https://www.cerebras.net/cerebras-wafer-scale-engine-why-we-need-big-chips-for-deep-learning>

AMD's 3D Last Level Cache (2021)

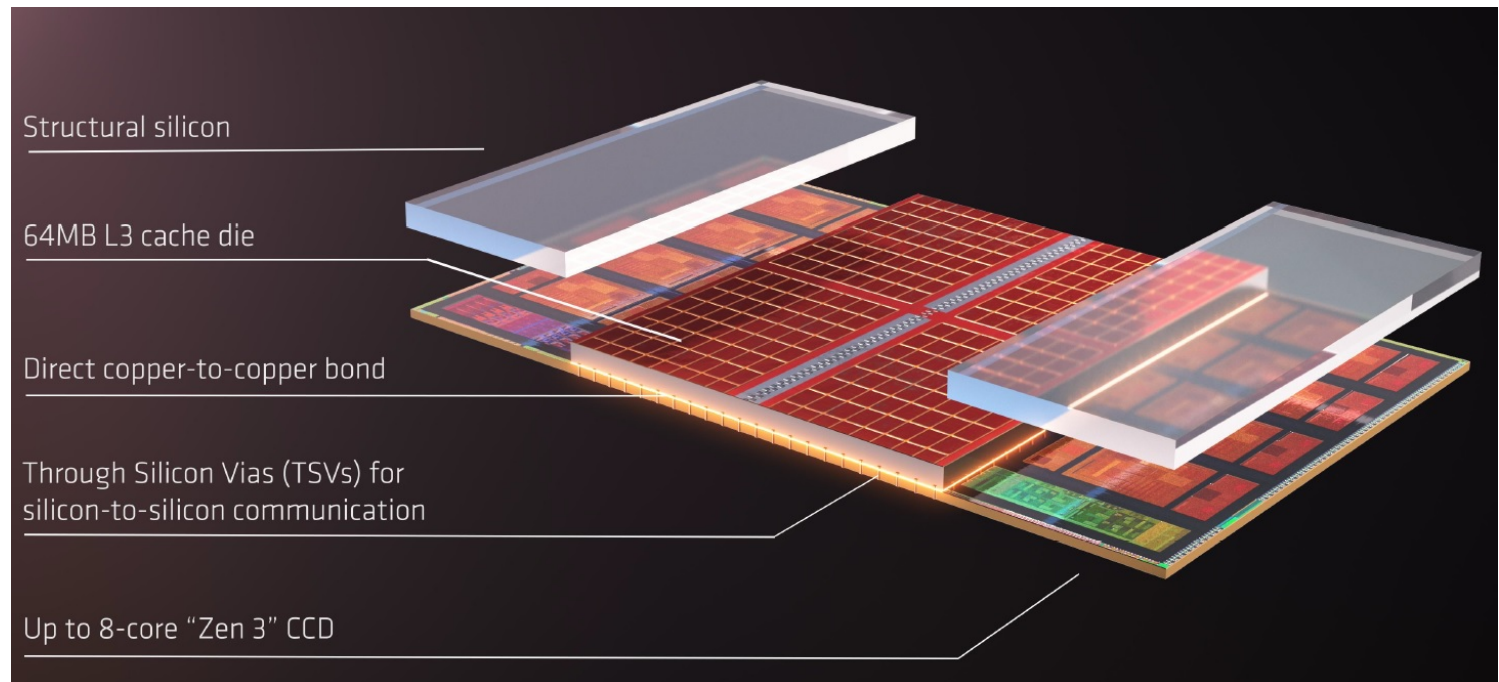


<https://community.microcenter.com/discussion/5134/comparing-zen-3-to-zen-2>

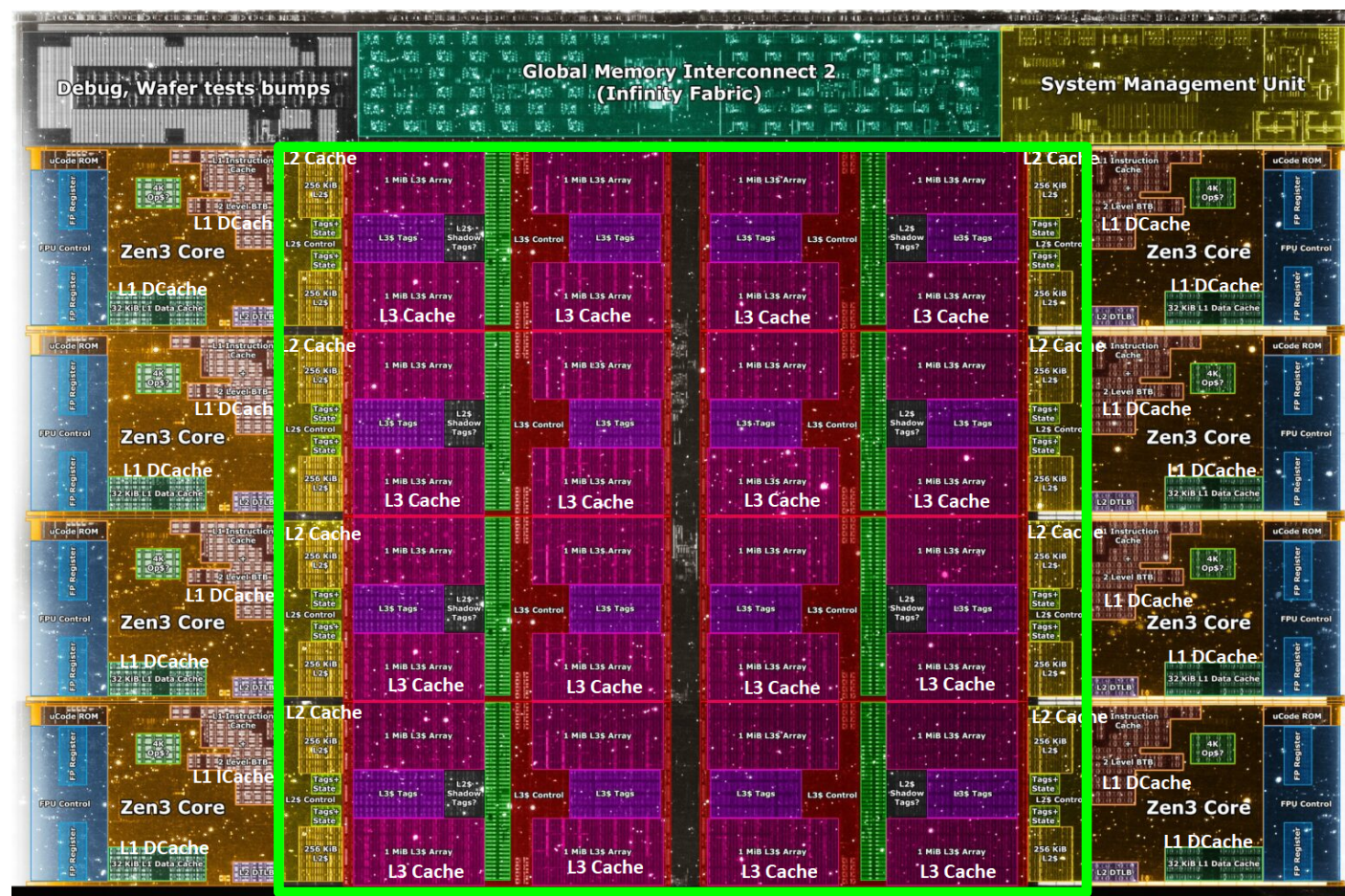
AMD increases the L3 size of their 8-core Zen 3 processors from 32 MB to 96 MB

Additional 64 MB L3 cache die
stacked on top of the processor die

- Connected using Through Silicon Vias (TSVs)
- Total of 96 MB L3 cache



Recall: Deeper and Larger Cache Hierarchies



Core Count:

8 cores/16 threads

L1 Caches:
32 KB per core

L2 Caches:
512 KB per core

L3 Cache:
32 MB shared

AMD Ryzen 5000, 2020

Many Other Ideas and Topics

- Advanced memory systems
- Self-optimizing architectures (using machine learning principles)
- Processing-in-memory systems
- Emerging memory technologies
- Solid state disks and storage, I/O
- Interconnection networks, communication networks
- Many issues in multiprocessing and multithreading
- Distributed architectures
- Heterogeneous systems: Heterogeneous CPU-GPU-FPGA-HWAcc architectures
- QoS and predictable performance
- Reliable and secure architectures
- Programmability, portability
- Better HW/SW interfaces
- Reconfigurable computing
- Specialized and domain-specific architectures – genomics, medicine, health, AI/ML
- Unconventional architectures – nature-inspired, quantum, molecular, ...
- ...

Oculus, New York City

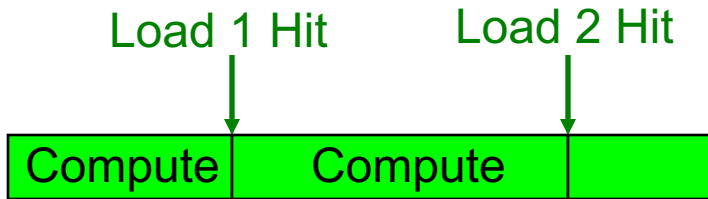


Recall:

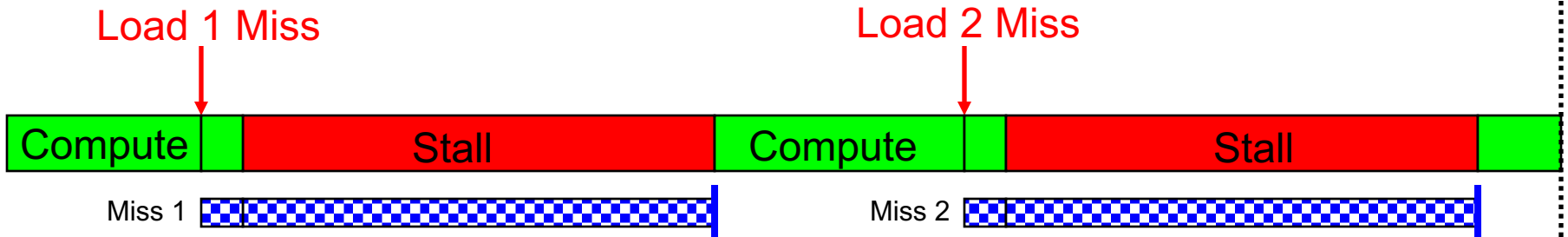
Runahead Execution

Runahead Execution Example

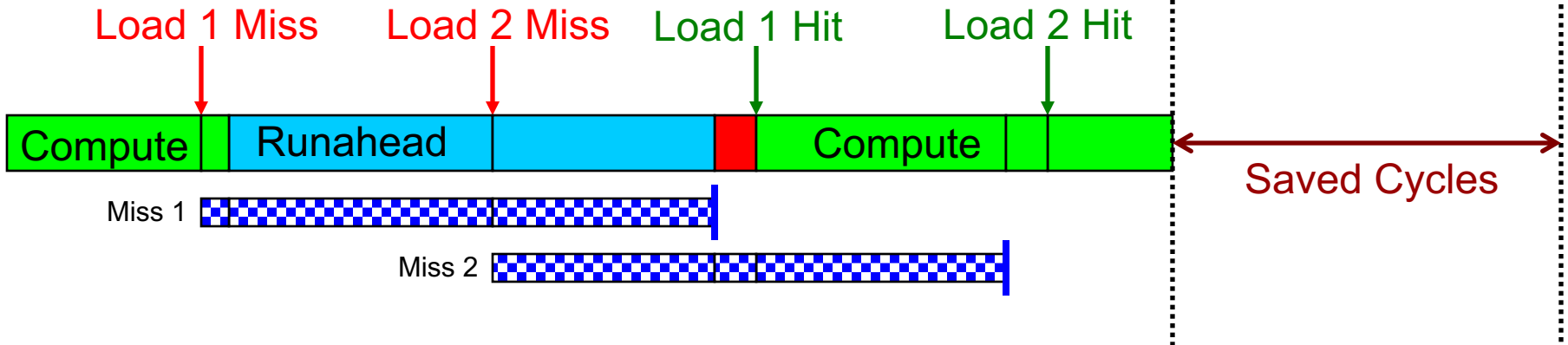
Perfect Caches:



Small OoO Instruction Window:

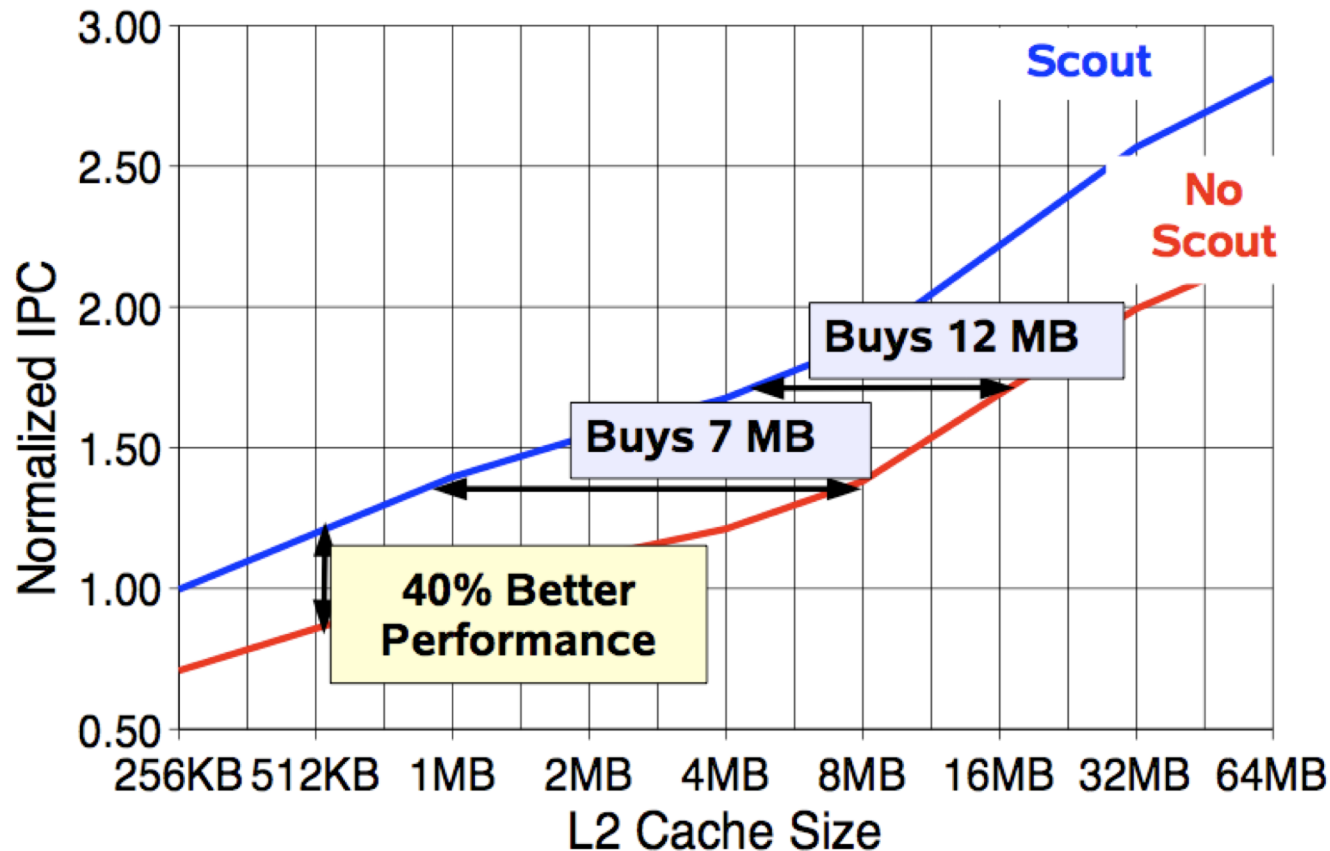


Runahead:



Effect of Runahead Prefetching in Sun ROCK

- Shailender Chaudhry talk, Aug 2008.



Effective prefetching can both improve performance and reduce hardware cost

More on Runahead Execution

- Onur Mutlu, Jared Stark, Chris Wilkerson, and Yale N. Patt,
"Runahead Execution: An Alternative to Very Large Instruction Windows for Out-of-order Processors"

Proceedings of the 9th International Symposium on High-Performance Computer Architecture (HPCA), pages 129-140, Anaheim, CA, February 2003. [Slides \(pdf\)](#)

One of the 15 computer arch. papers of 2003 selected as Top Picks by IEEE Micro. HPCA Test of Time Award (awarded in 2021).

[\[Lecture Slides \(pptx\) \(pdf\)\]](#)

[\[Lecture Video \(1 hr 54 mins\)\]](#)

[\[Retrospective HPCA Test of Time Award Talk Slides \(pptx\) \(pdf\)\]](#)

[\[Retrospective HPCA Test of Time Award Talk Video \(14 minutes\)\]](#)

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

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

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The University of Texas at Austin

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†Microprocessor Research

Intel Labs

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‡Desktop Platforms Group

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HIGH-PERFORMANCE THROUGHPUT COMPUTING

THROUGHPUT COMPUTING, ACHIEVED THROUGH MULTITHREADING AND MULTICORE TECHNOLOGY, CAN LEAD TO PERFORMANCE IMPROVEMENTS THAT ARE 10 TO 30× THOSE OF CONVENTIONAL PROCESSORS AND SYSTEMS. HOWEVER, SUCH SYSTEMS SHOULD ALSO OFFER GOOD SINGLE-THREAD PERFORMANCE. HERE, THE AUTHORS SHOW THAT HARDWARE SCOUTING INCREASES THE PERFORMANCE OF AN ALREADY ROBUST CORE BY UP TO 40 PERCENT FOR COMMERCIAL BENCHMARKS.

More on Runahead in Sun ROCK

Simultaneous Speculative Threading: A Novel Pipeline Architecture Implemented in Sun's ROCK Processor

Shailender Chaudhry, Robert Cypher, Magnus Ekman, Martin Karlsson,
Anders Landin, Sherman Yip, Håkan Zeffer, and Marc Tremblay
Sun Microsystems, Inc.
4180 Network Circle, Mailstop SCA18-211
Santa Clara, CA 95054, USA
{shailender.chaudhry, robert.cypher, magnus.ekman, martin.karlsson,
anders.landin, sherman.yip, haakan.zeffer, marc.tremblay}@sun.com

Runahead Execution in IBM POWER6

Runahead Execution vs. Conventional Data Prefetching in the IBM POWER6 Microprocessor

Harold W. Cain

Priya Nagpurkar

IBM T.J. Watson Research Center
Yorktown Heights, NY
{tcain, pnagpurkar}@us.ibm.com

Cain+, “Runahead Execution vs. Conventional Data Prefetching
in the IBM POWER6 Microprocessor,” ISPASS 2010.

Runahead Execution in IBM POWER6

Abstract

After many years of prefetching research, most commercially available systems support only two types of prefetching: software-directed prefetching and hardware-based prefetchers using simple sequential or stride-based prefetching algorithms. More sophisticated prefetching proposals, despite promises of improved performance, have not been adopted by industry. In this paper, we explore the efficacy of both hardware and software prefetching in the context of an IBM POWER6 commercial server. Using a variety of applications that have been compiled with an aggressively optimizing compiler to use software prefetching when appropriate, we perform the first study of a new runahead prefetching feature adopted by the POWER6 design, evaluating it in isolation and in conjunction with a conventional hardware-based sequential stream prefetcher and compiler-inserted software prefetching.

We find that the POWER6 implementation of runahead prefetching is quite effective on many of the memory intensive applications studied; in isolation it improves performance as much as 36% and on average 10%. However, it outperforms the hardware-based stream prefetcher on only two of the benchmarks studied, and in those by a small margin.

When used in conjunction with the conventional prefetching mechanisms, the runahead feature adds an additional 6% on average, and 39% in the best case (GemsFDTD).

DENVER: NVIDIA'S FIRST 64-BIT ARM PROCESSOR

NVIDIA'S FIRST 64-BIT ARM PROCESSOR, CODE-NAMED DENVER, LEVERAGES A HOST OF NEW TECHNOLOGIES, SUCH AS DYNAMIC CODE OPTIMIZATION, TO ENABLE HIGH-PERFORMANCE MOBILE COMPUTING. IMPLEMENTED IN A 28-NM PROCESS, THE DENVER CPU CAN ATTAIN CLOCK SPEEDS OF UP TO 2.5 GHZ. THIS ARTICLE OUTLINES THE DENVER ARCHITECTURE, DESCRIBES ITS TECHNOLOGICAL INNOVATIONS, AND PROVIDES RELEVANT COMPARISONS AGAINST COMPETING MOBILE PROCESSORS.

Boggs+, "[Denver: NVIDIA's First 64-Bit ARM Processor](#)," IEEE Micro 2015.

Runahead Execution in NVIDIA Denver

Reducing the effects of long cache-miss penalties has been a major focus of the micro-architecture, using techniques like prefetching and run-ahead. An aggressive hardware prefetcher implementation detects L2 cache requests and tracks up to 32 streams, each with complex stride patterns.

Run-ahead uses the idle time that a CPU spends waiting on a long latency operation to discover cache and DTLB misses further down the instruction stream and generates prefetch requests for these misses.¹ These prefetch requests warm up the data cache and DTLB well before the actual execution of

the instructions that require the data. Run-ahead complements the hardware prefetcher because it's better at prefetching nonstrided streams, and it trains the hardware prefetcher faster than normal execution to yield a combined benefit of 13 percent on SPECint2000 and up to 60 percent on SPECfp2000.

The core includes a hardware prefetch unit that Boggs describes as “aggressive” in preloading the data cache but less aggressive in preloading the instruction cache. It also implements a “run-ahead” feature that continues to execute microcode speculatively after a data-cache miss; this execution can trigger additional cache misses that resolve in the shadow of the first miss. Once the data from the original miss returns, the results of this speculative execution are discarded and execution restarts with the bundle containing the original miss, but run-ahead can preload subsequent data into the cache, thus avoiding a string of time-wasting cache misses. These and other features help Denver outscore Cortex-A15 by more than 2.6x on a memory-read test even when both use the same SoC framework (Tegra K1).

Boggs+, “Denver: NVIDIA’s First 64-Bit ARM Processor,” IEEE Micro 2015.

Gwennap, “NVIDIA’s First CPU is a Winner,” MPR 2014.

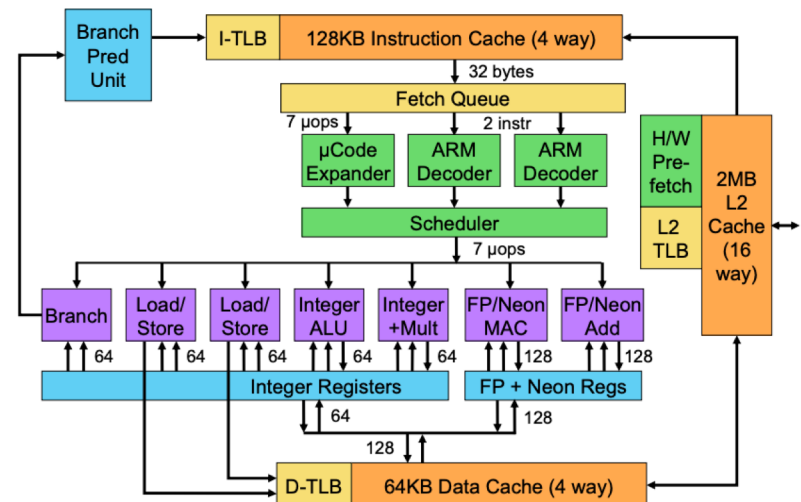


Figure 3. Denver CPU microarchitecture. This design combines a fairly

Runahead Readings

■ Required

- ❑ Mutlu et al., “Runahead Execution”, HPCA 2003, Top Picks 2003.

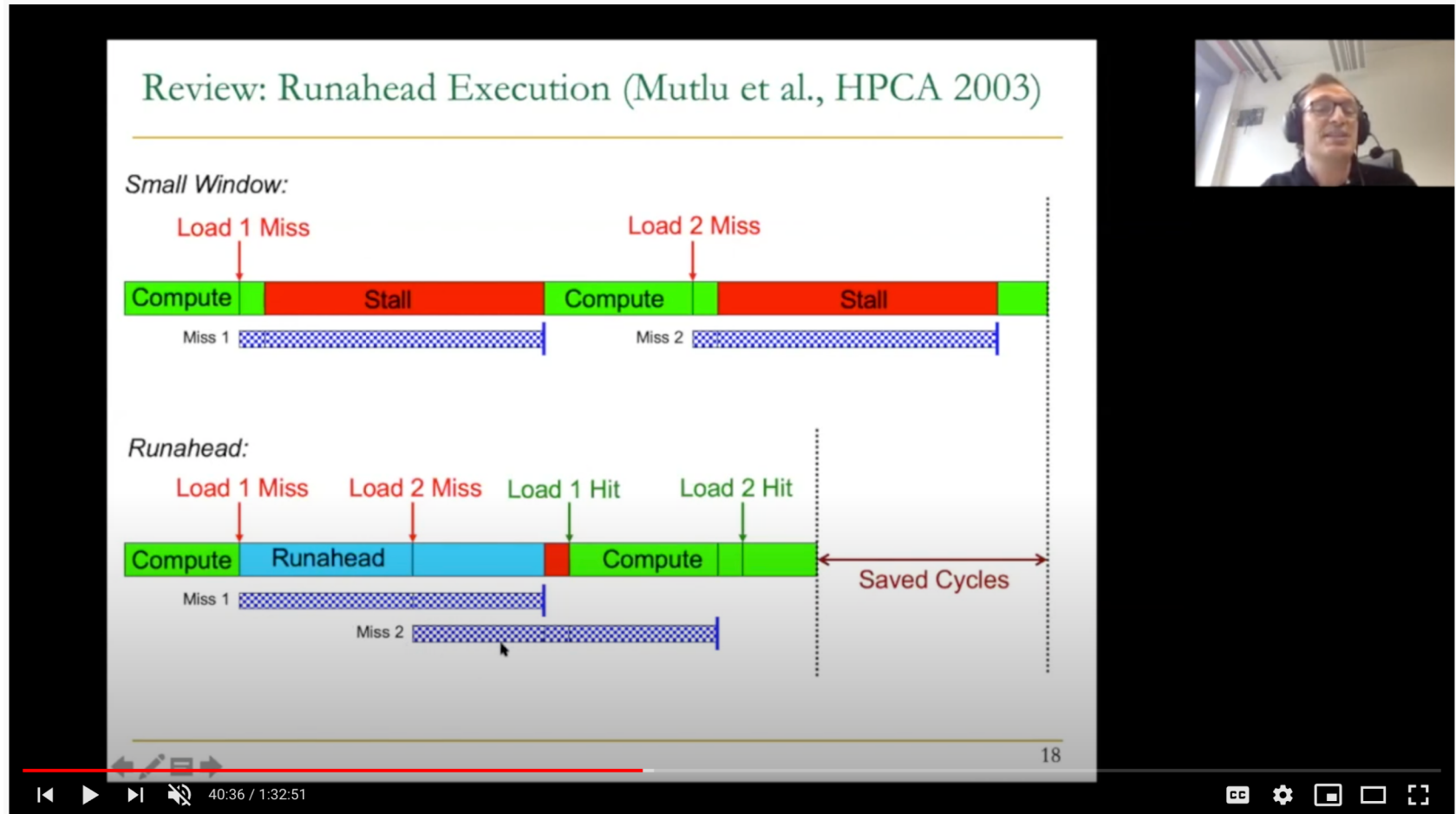
■ Recommended

- ❑ Mutlu et al., “Efficient Runahead Execution: Power-Efficient Memory Latency Tolerance,” ISCA 2005, IEEE Micro Top Picks 2006.
- ❑ Mutlu et al., “Address-Value Delta (AVD) Prediction,” MICRO 2005.
- ❑ Armstrong et al., “Wrong Path Events,” MICRO 2004.

More on Runahead Execution

- Lecture video from Fall 2020, Computer Architecture:
 - https://www.youtube.com/watch?v=zPewo6IaJ_8
- Lecture video from Fall 2017, Computer Architecture:
 - <https://www.youtube.com/watch?v=Kj3relihGF4>
- Onur Mutlu,
"Efficient Runahead Execution Processors"
Ph.D. Dissertation, HPS Technical Report, TR-HPS-2006-007, July 2006. [Slides \(ppt\)](#)
Nominated for the ACM Doctoral Dissertation Award by the University of Texas at Austin.

More on Runahead Execution (I)



Computer Architecture - Lecture 19a: Execution-Based Prefetching (ETH Zürich, Fall 2020)

395 views • Nov 29, 2020

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Onur Mutlu Lectures
16.5K subscribers

ANALYTICS

EDIT VIDEO

https://www.youtube.com/watch?v=zPewo6laJ_8&list=PL5Q2soXY2Zi9xidylgBxUz7xRPS-wisBN&index=34

Runahead Execution in NVIDIA Denver

Run-ahead uses the idle time that a CPU spends waiting on a long latency operation to discover cache and DTLB misses further down the instruction stream and generates prefetch requests for these misses.¹ These prefetch requests warm up the data cache and DTLB well before the actual execution of the instructions that require the data. Run-ahead complements the hardware prefetcher because it's better at prefetching nonstrided streams, and it trains the hardware prefetcher faster than normal execution to yield a combined benefit of 13 percent on SPECint2000 and up to 60 percent on SPECfp2000.

Gwennap, "NVIDIA's First CPU is a Winner," MPR 2014.

```

graph TD
    BP[Branch Pred Unit] --> ITLB[I-TLB]
    BP --> L2[2MB L2 Cache 16 way]
    L2 --> BP
    L2 --> HWP[H/W Pre-fetch]
    HWP --> L2
    HWP --> DTLB[D-TLB]
    DTLB --> L2
    DTLB --> DC[64KB Data Cache 4 way]
    DC --> DTLB
    DC --> FP[FP + Neon Reqs]
    DC --> IR[Integer Registers]
    FP -- 128 --> FP_ADD[FP/Neon Add]
    FP -- 128 --> FP_MAC[FP/Neon MAC]
    FP -- 128 --> INT_MULT[Integer +Mult]
    FP -- 128 --> INT_ALU[Integer ALU]
    FP -- 128 --> LS[Load/Store]
    FP -- 128 --> BR[Branch]
    INT_MULT -- 64 --> IR
    INT_ALU -- 64 --> IR
    LS -- 64 --> IR
    BR -- 64 --> IR
    IR -- 64 --> S[Scheduler]
    S -- 7 pipes --> AD[ARM Decoder]
    S -- 2 instr --> UC[μCode Expander]
    AD -- 32 bytes --> IC[128KB Instruction Cache 4 way]
    IC --> ITLB
    IC --> FQ[Fetch Queue]
    FQ -- 7 pipes --> UC
    FQ -- 7 pipes --> AD
    FQ -- 7 pipes --> INT_MULT
    FQ -- 7 pipes --> INT_ALU
    FQ -- 7 pipes --> LS
    FQ -- 7 pipes --> BR
  
```

▶ ▶ 🔊 6:18 / 14:27



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[EDIT VIDEO](#)

My Suggestions to You

Suggestion to Researchers: Principle: Passion

Follow Your Passion
(Do not get derailed
by naysayers)

Principle: Build Infrastructure

Build Infrastructure to
Enable Your Passion

Principle: Work Hard

Work Hard to
Enable Your Passion

Suggestion to Researchers: Principle: Resilience

Be Resilient

Principle: Learning and Scholarship

Focus on
learning and scholarship

Principle: Learning and Scholarship

The quality of your work
defines your impact

Principle: Good Mindset, Goals & Focus

You can make a
good impact
on the world

Recall: This



Recall: That



Recall: A Key Question

- How was Calavatra able to design especially his key buildings?
- Can have many guesses
 - (Very) hard work, perseverance, dedication (over decades)
 - Experience
 - Creativity, Out-of-the-box thinking
 - A good understanding of past designs
 - Good judgment and intuition
 - Strong skill combination (math, architecture, art, engineering, ...)
 - Funding (\$\$\$\$), luck, initiative, entrepreneurialism
 - Strong understanding of and commitment to fundamentals
 - Principled design
 - ...
- You will be exposed to and hopefully develop/enhance many of these skills in this course

Processing in Memory

The Problem

Data access is the major performance and energy bottleneck

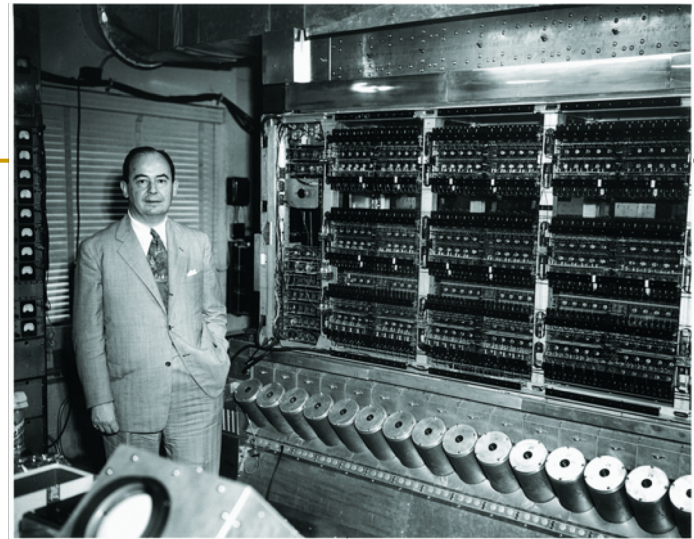
Our current
design principles
cause great energy waste
(and great performance loss)

The Problem

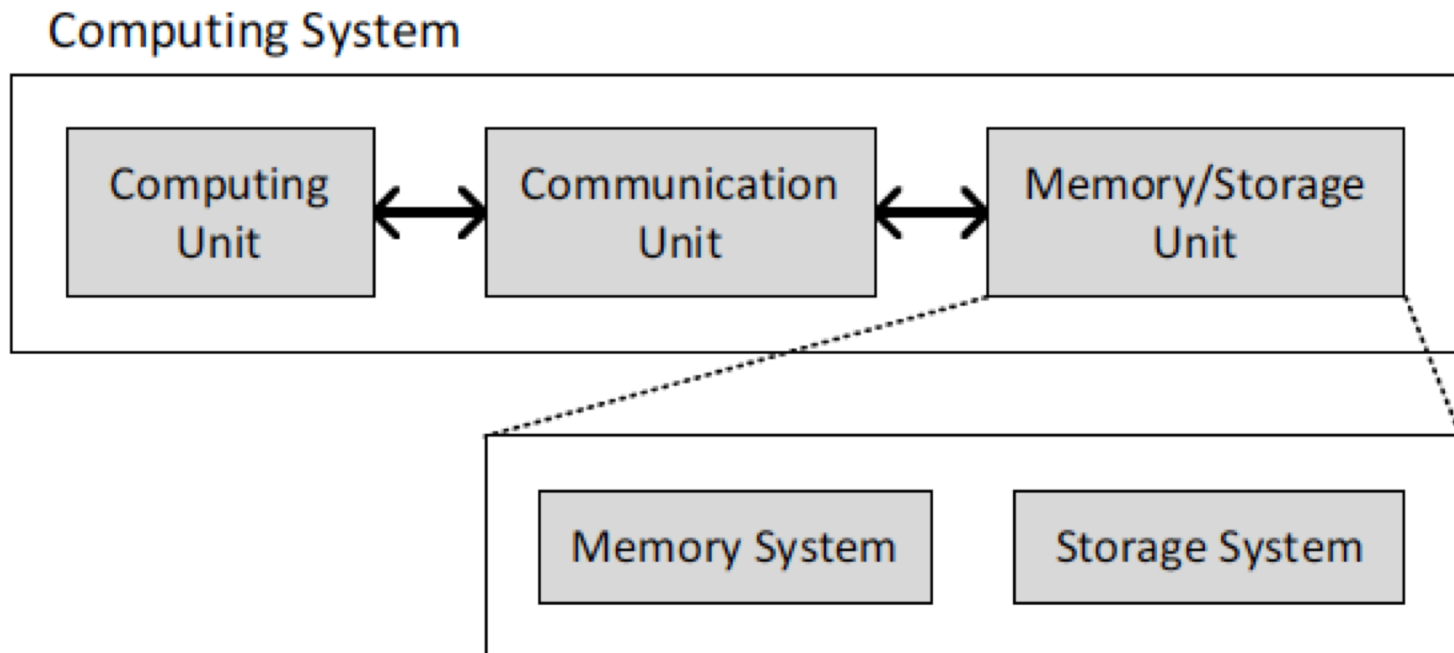
Processing of data
is performed
far away from the data

A Computing System

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

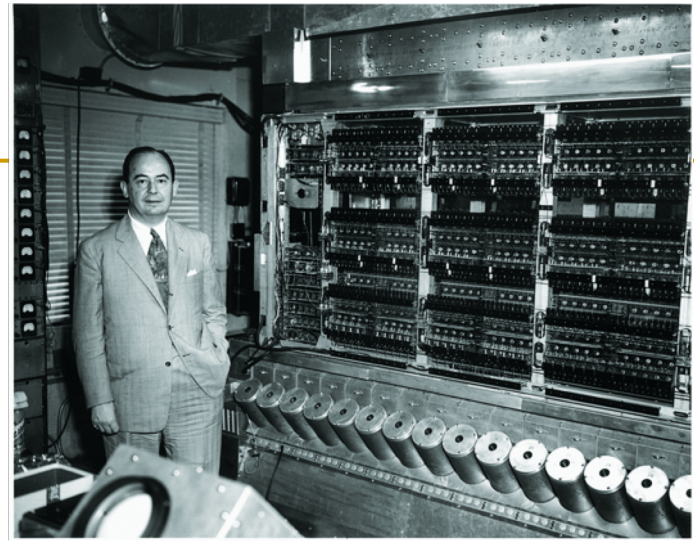


Burks, Goldstein, von Neumann, "Preliminary discussion of the logical design of an electronic computing instrument," 1946.



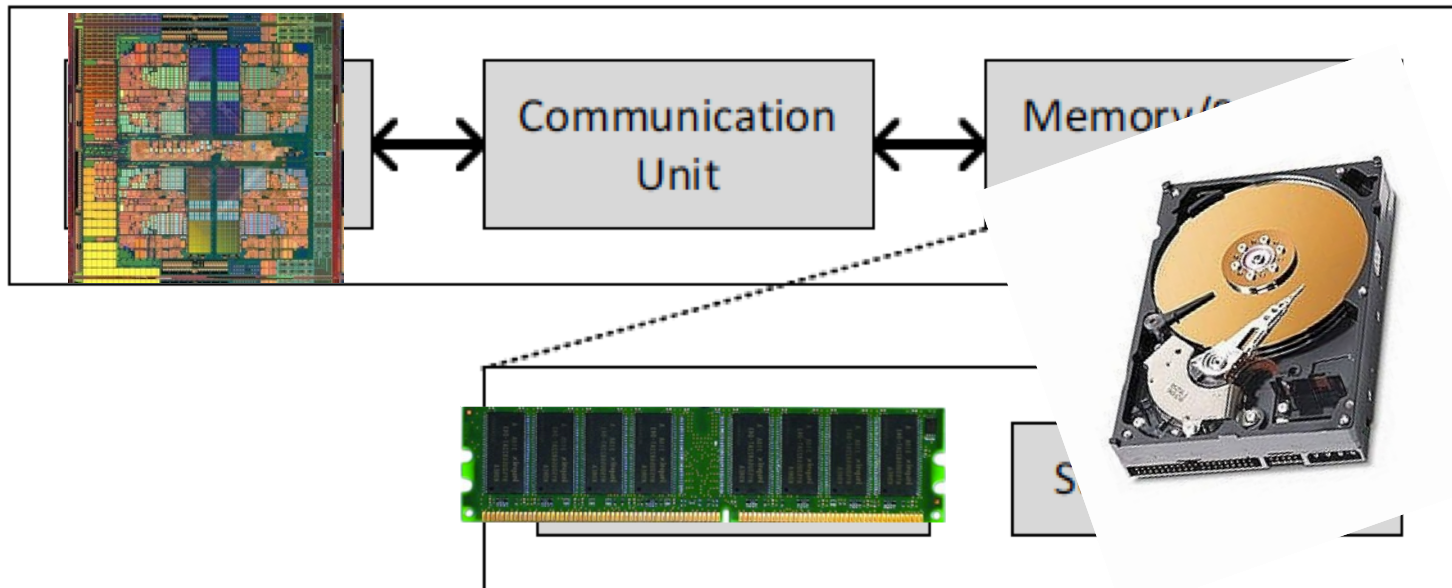
A Computing System

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



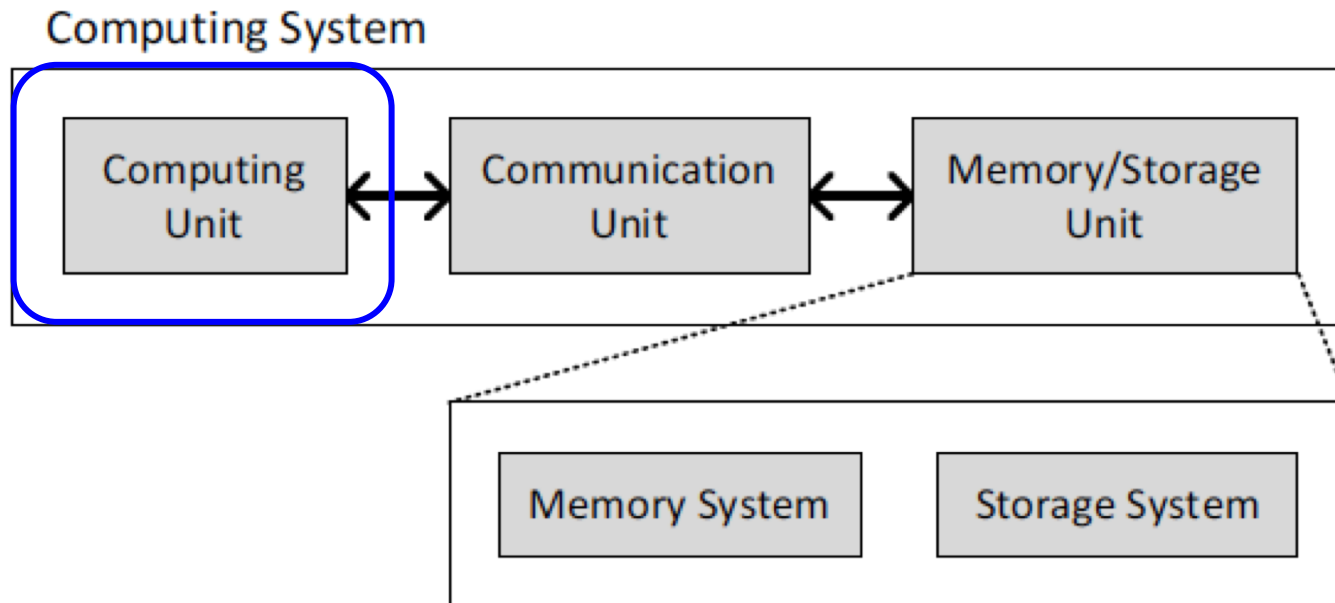
Burks, Goldstein, von Neumann, "Preliminary discussion of the logical design of an electronic computing instrument," 1946.

Computing System



Today's Computing Systems

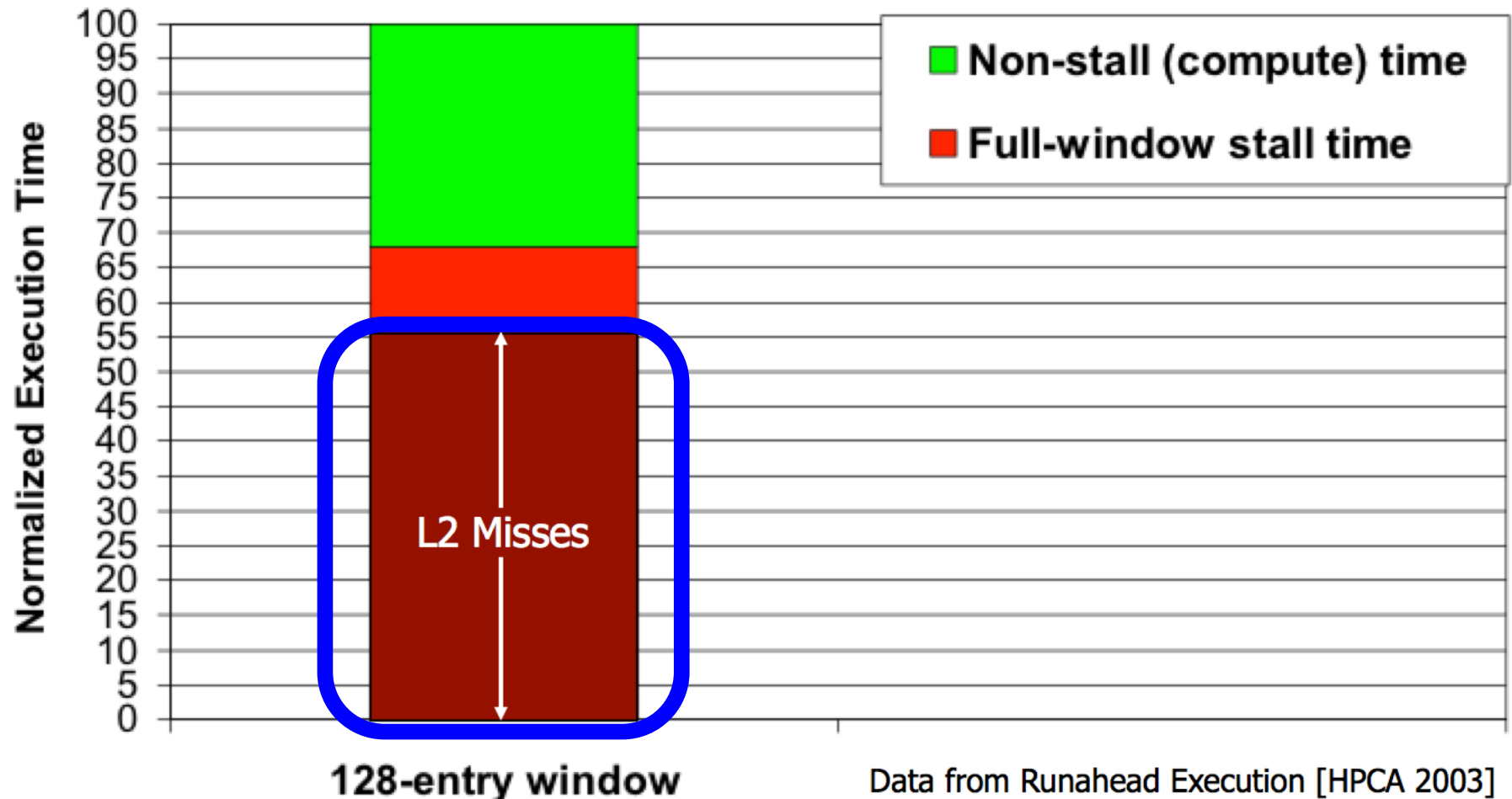
- Are overwhelmingly processor centric
- All data processed in the processor → at great system cost
- Processor is heavily optimized and is considered the master
- Data storage units are dumb and are largely unoptimized (except for some that are on the processor die)



Yet ...

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

- **“It’s the Memory, Stupid!”** (Richard Sites, MPR, 1996)



The Performance Perspective

- Onur Mutlu, Jared Stark, Chris Wilkerson, and Yale N. Patt,
"Runahead Execution: An Alternative to Very Large Instruction Windows for Out-of-order Processors"
Proceedings of the 9th International Symposium on High-Performance Computer Architecture (HPCA), pages 129-140, Anaheim, CA, February 2003. [Slides \(pdf\)](#)
One of the 15 computer arch. papers of 2003 selected as Top Picks by IEEE Micro. HPCA Test of Time Award (awarded in 2021).

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

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

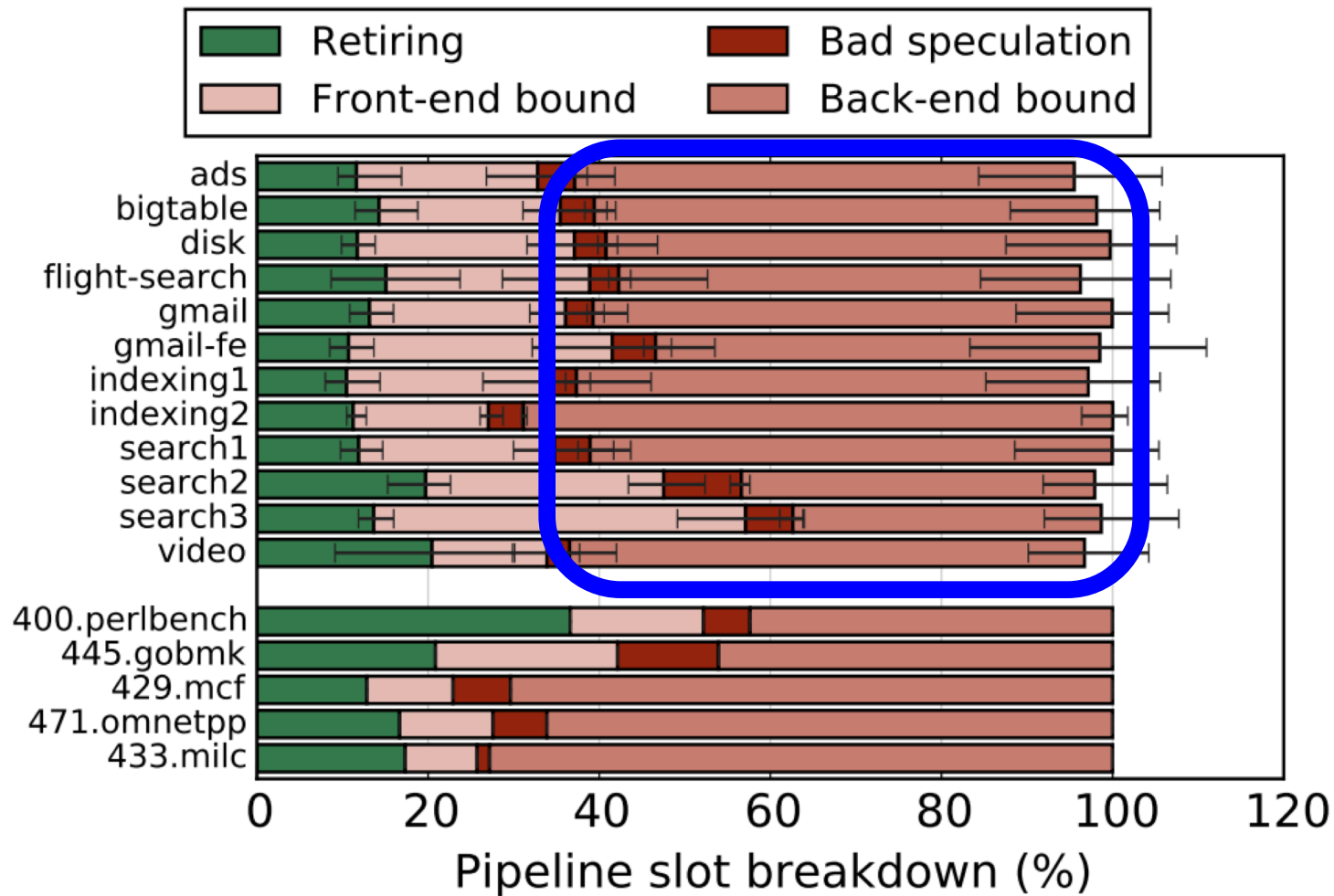
§ECE Department
The University of Texas at Austin
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†Microprocessor Research
Intel Labs
jared.w.stark@intel.com

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

The Performance Perspective (Today)

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



The Performance Perspective (Today)

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

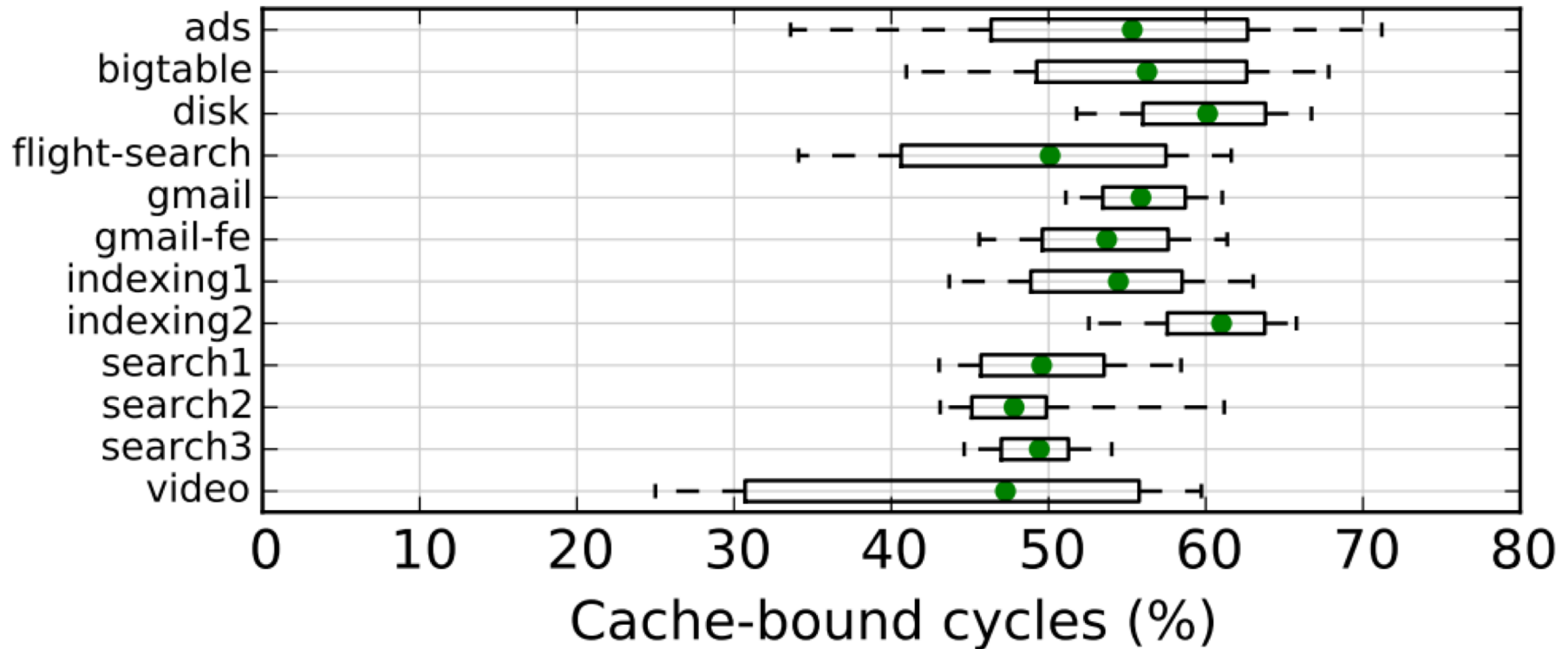


Figure 11: Half of cycles are spent stalled on caches.

Perils of Processor-Centric Design

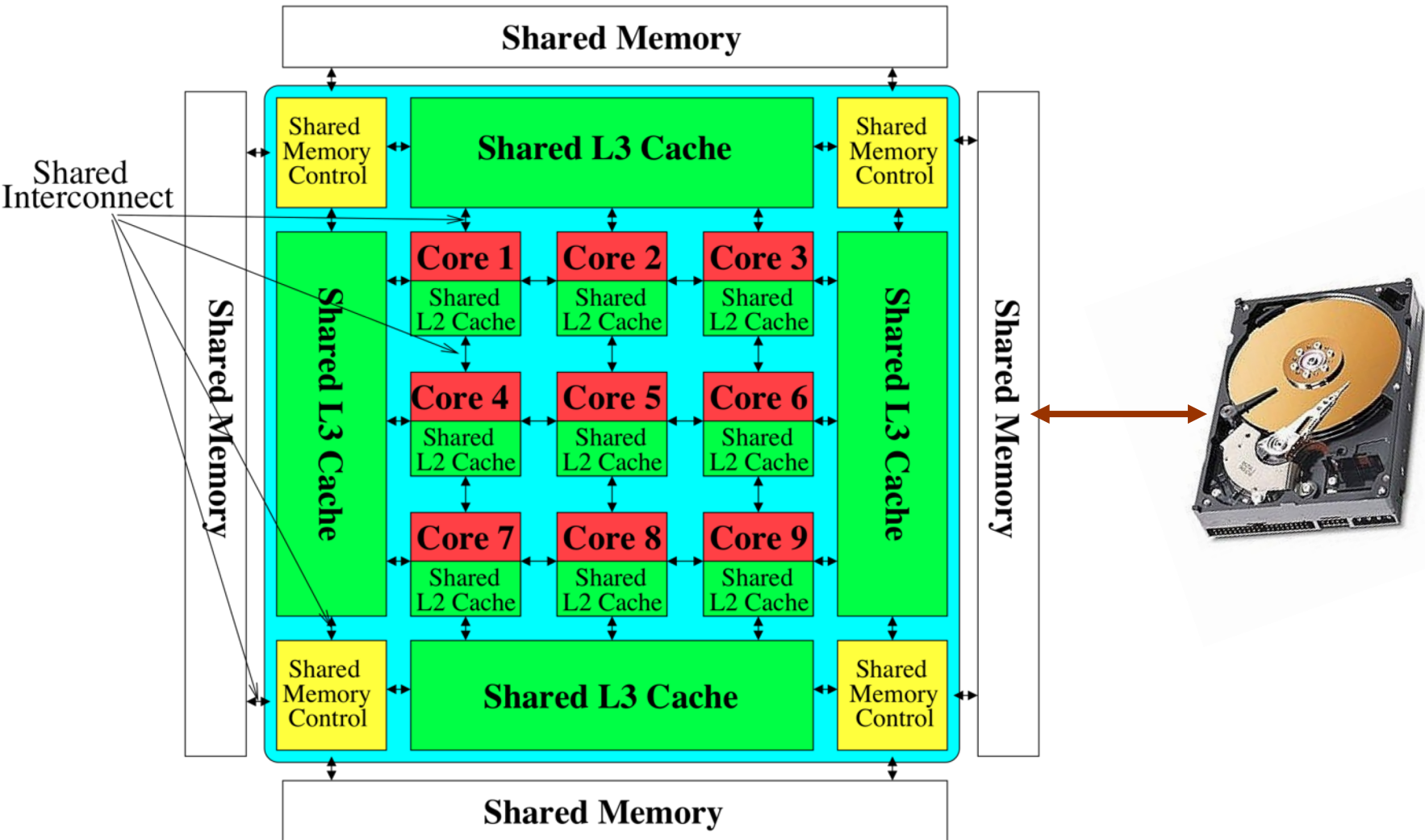
■ Grossly-imbalanced systems

- ❑ Processing done only in **one place**
- ❑ Everything else just stores and moves data: **data moves a lot**
 - Energy inefficient
 - Low performance
 - Complex

■ Overly complex and bloated processor (and accelerators)

- ❑ To tolerate data access from memory
- ❑ Complex hierarchies and mechanisms
 - Energy inefficient
 - Low performance
 - Complex

Perils of Processor-Centric Design

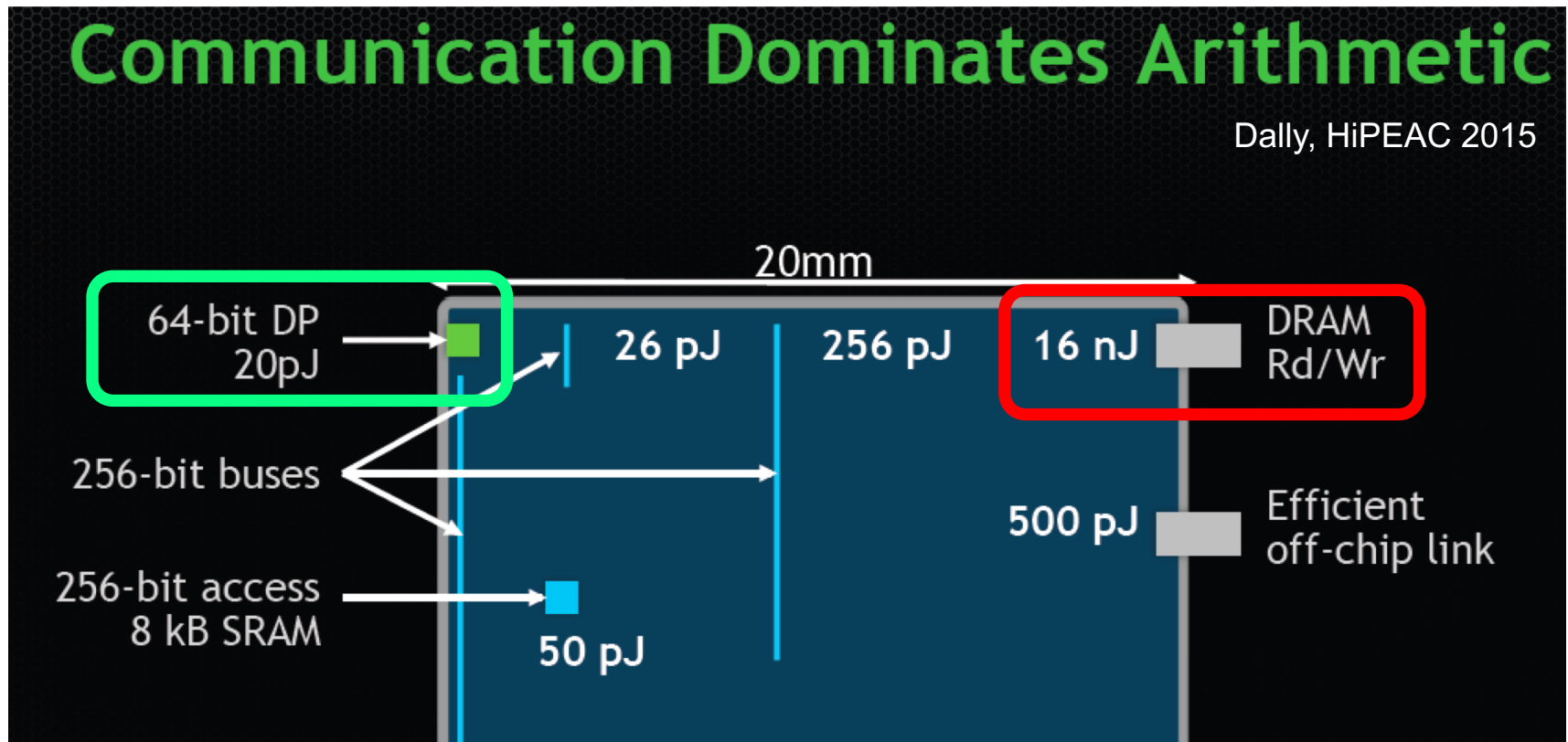


Most of the system is dedicated to storing and moving data

Data Movement vs. Computation Energy

Communication Dominates Arithmetic

Dally, HiPEAC 2015



A memory access consumes $\sim 100\text{-}1000\times$ the energy of a complex addition

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 23rd International Conference on Architectural Support for Programming Languages and Operating Systems (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

Saugata Ghose¹

Youngsok Kim²

Rachata Ausavarungnirun¹

Eric Shiu³

Rahul Thakur³

Daehyun Kim^{4,3}

Aki Kuusela³

Allan Knies³

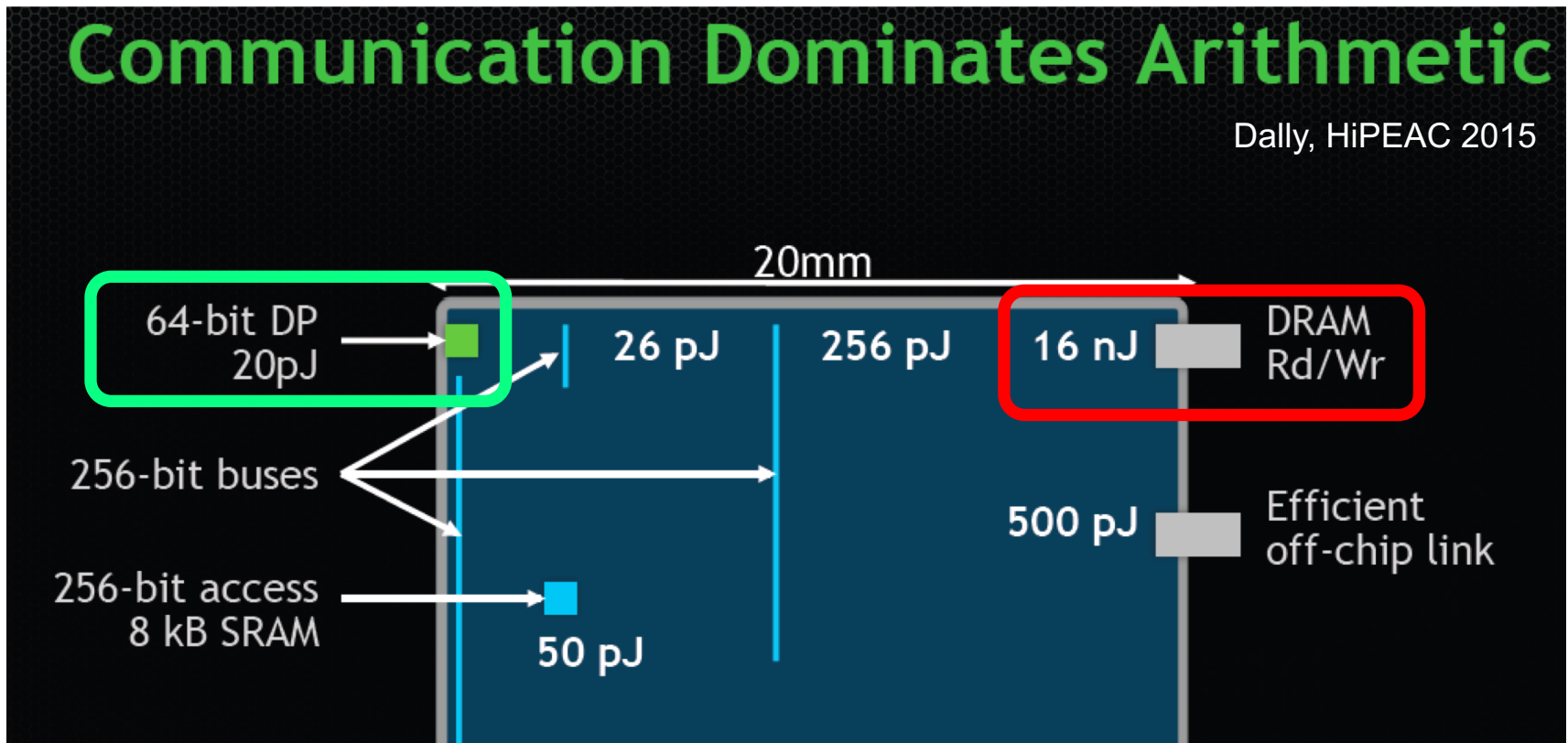
Parthasarathy Ranganathan³

Onur Mutlu^{5,1}

We Do Not Want to Move Data!

Communication Dominates Arithmetic

Dally, HiPEAC 2015

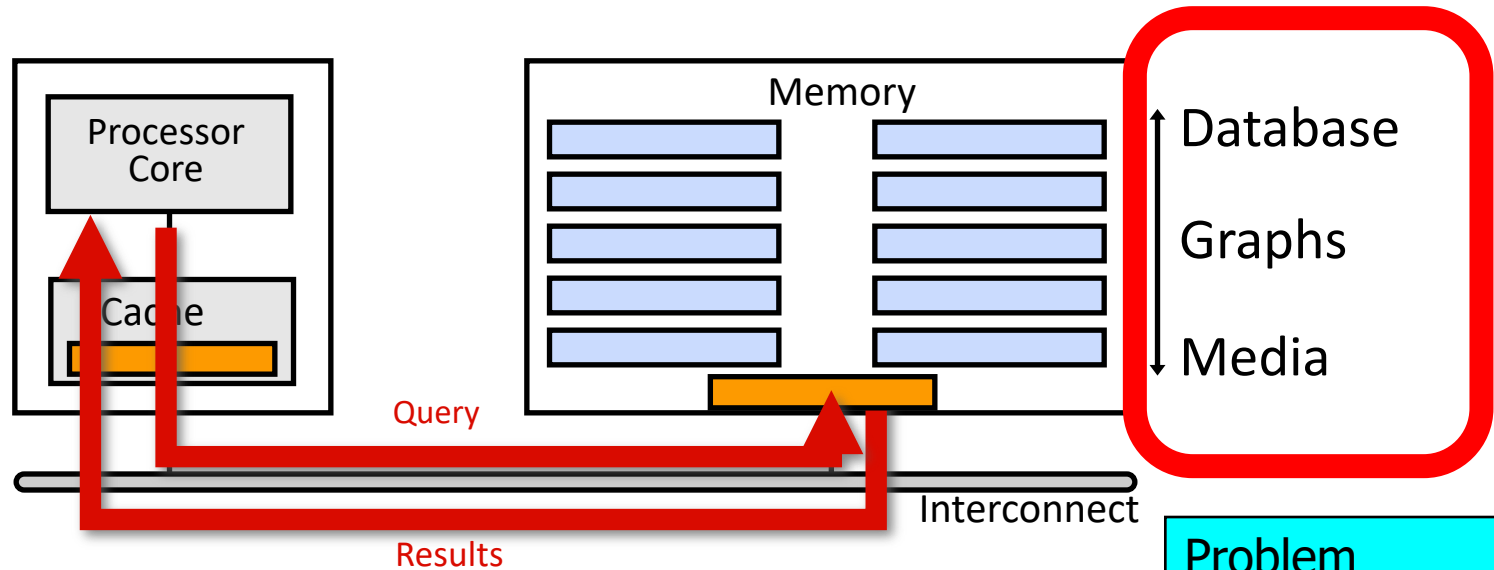


A memory access consumes $\sim 100\text{-}1000\times$ the energy of a complex addition

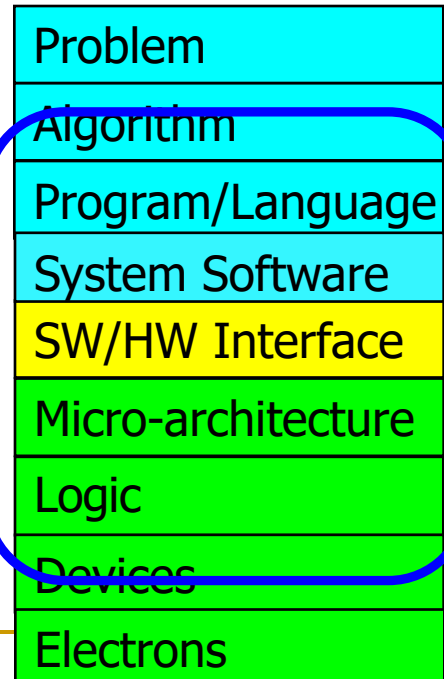
We Need A Paradigm Shift To ...

- Enable computation with minimal data movement
- Compute where it makes sense (where data resides)
- Make computing architectures more data-centric

Goal: Processing Inside Memory



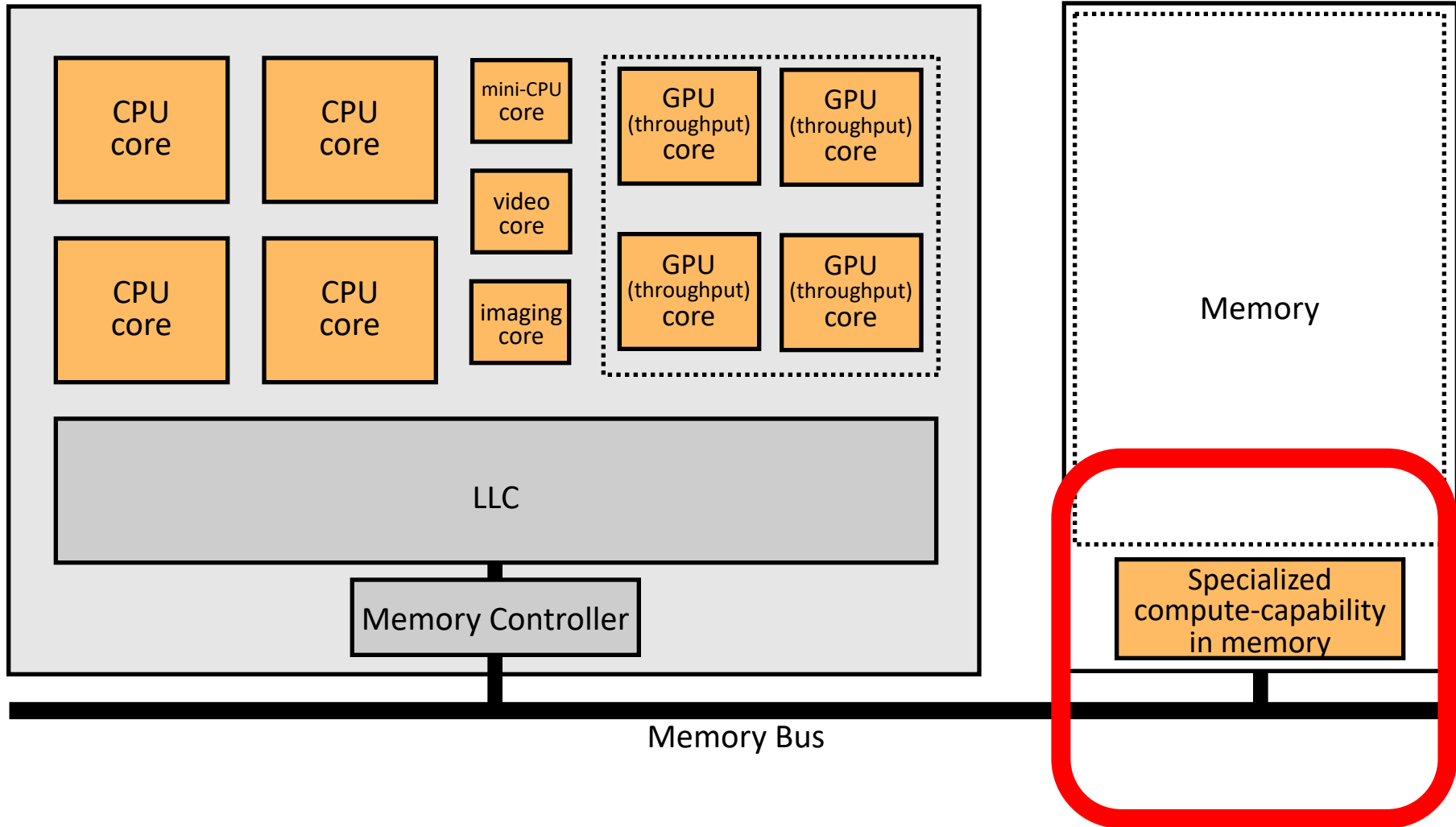
- Many questions ... How do we design the:
 - ❑ compute-capable memory & controllers?
 - ❑ processor chip and in-memory units?
 - ❑ software and hardware interfaces?
 - ❑ system software, compilers, languages?
 - ❑ algorithms and theoretical foundations?



Processing in Memory: Two Approaches

1. Processing using Memory
2. Processing near Memory

Memory as an Active Accelerator



Memory similar to a “conventional” accelerator

In-Memory Bulk Bitwise Operations

- We can support in-DRAM COPY, ZERO, AND, OR, NOT, MAJ
 - At low cost
 - Using **analog computation capability** of DRAM
 - Idea: activating multiple rows performs computation
 - **30-74X 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
-

Ambit [MICRO'17]

- Vivek Seshadri et al., “[Ambit: In-Memory Accelerator for Bulk Bitwise Operations Using Commodity DRAM Technology](#),” MICRO 2017.

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

Vivek Seshadri^{1,5} Donghyuk Lee^{2,5} Thomas Mullins^{3,5} Hasan Hassan⁴ Amirali Boroumand⁵
Jeremie Kim^{4,5} Michael A. Kozuch³ Onur Mutlu^{4,5} Phillip B. Gibbons⁵ Todd C. Mowry⁵

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

In-DRAM Bulk Bitwise Execution

- Vivek Seshadri and Onur Mutlu,
"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

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

¹ETH Zürich

²Simon Fraser University

³University of Illinois at Urbana–Champaign

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 24th International Symposium on High-Performance Computer Architecture (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 25th International Symposium on High-Performance Computer Architecture (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^{§‡}

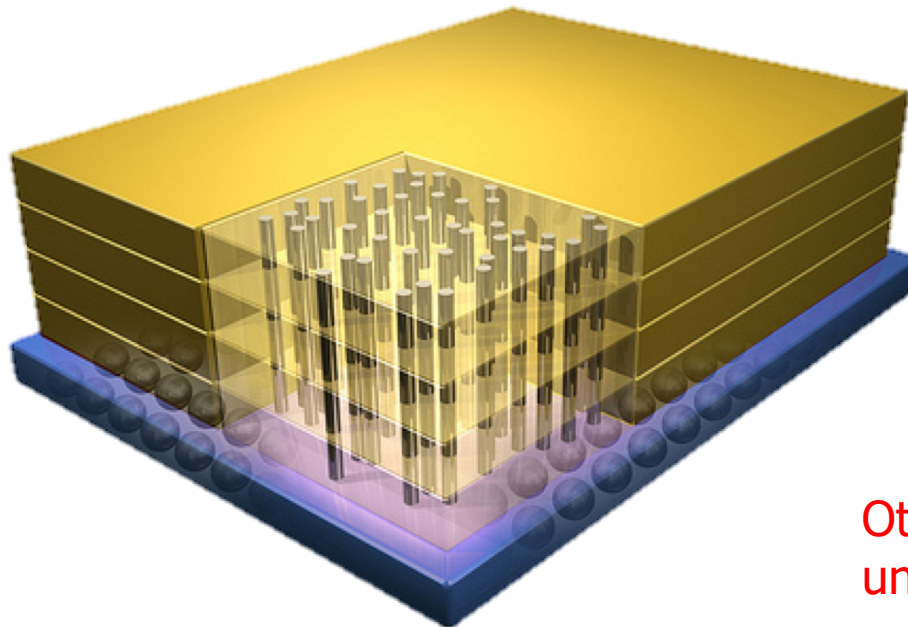
[‡]Carnegie Mellon University

[§]ETH Zürich

Opportunity: 3D-Stacked Logic+Memory



Hybrid Memory Cube
C O N S O R T I U M



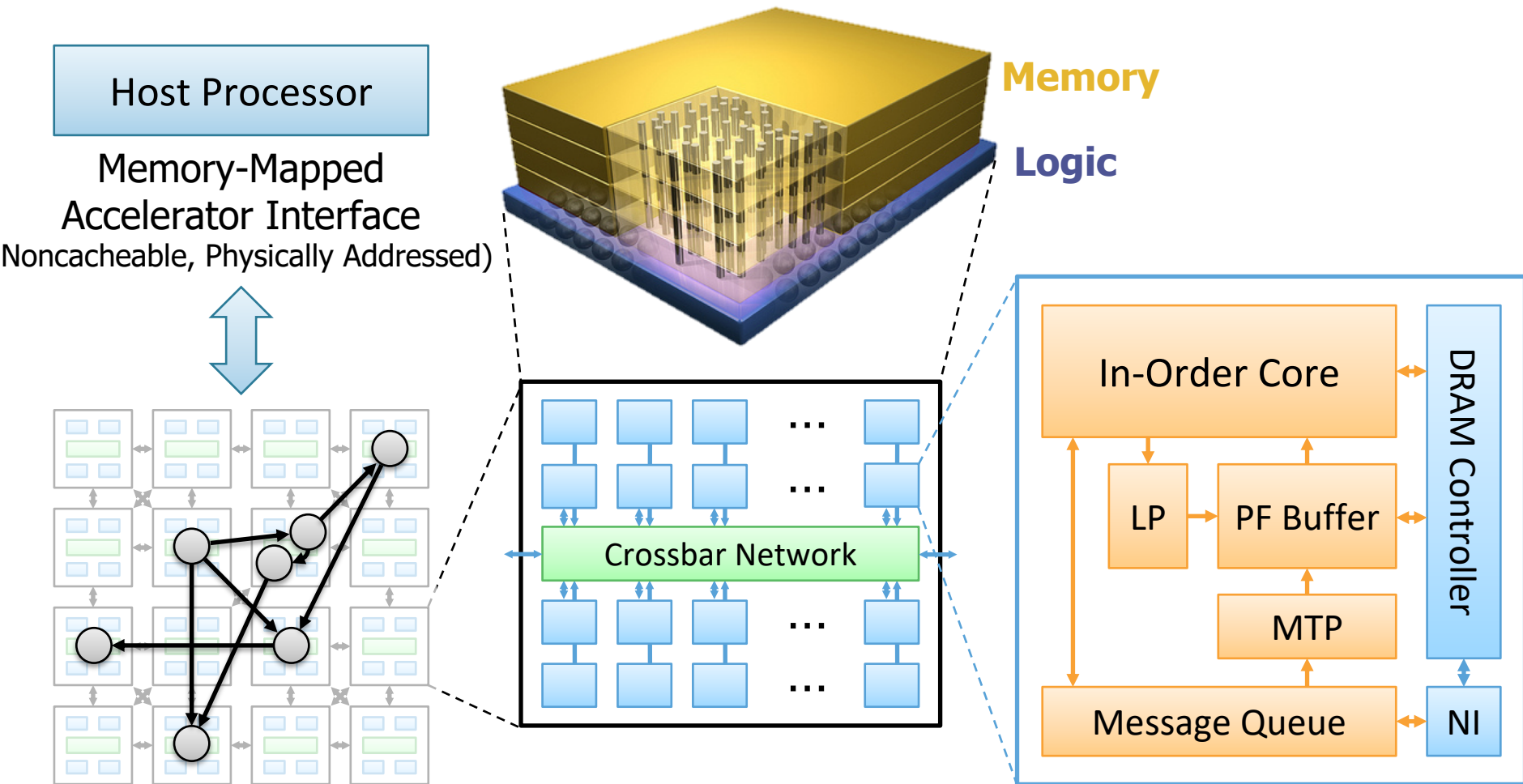
Memory

Logic

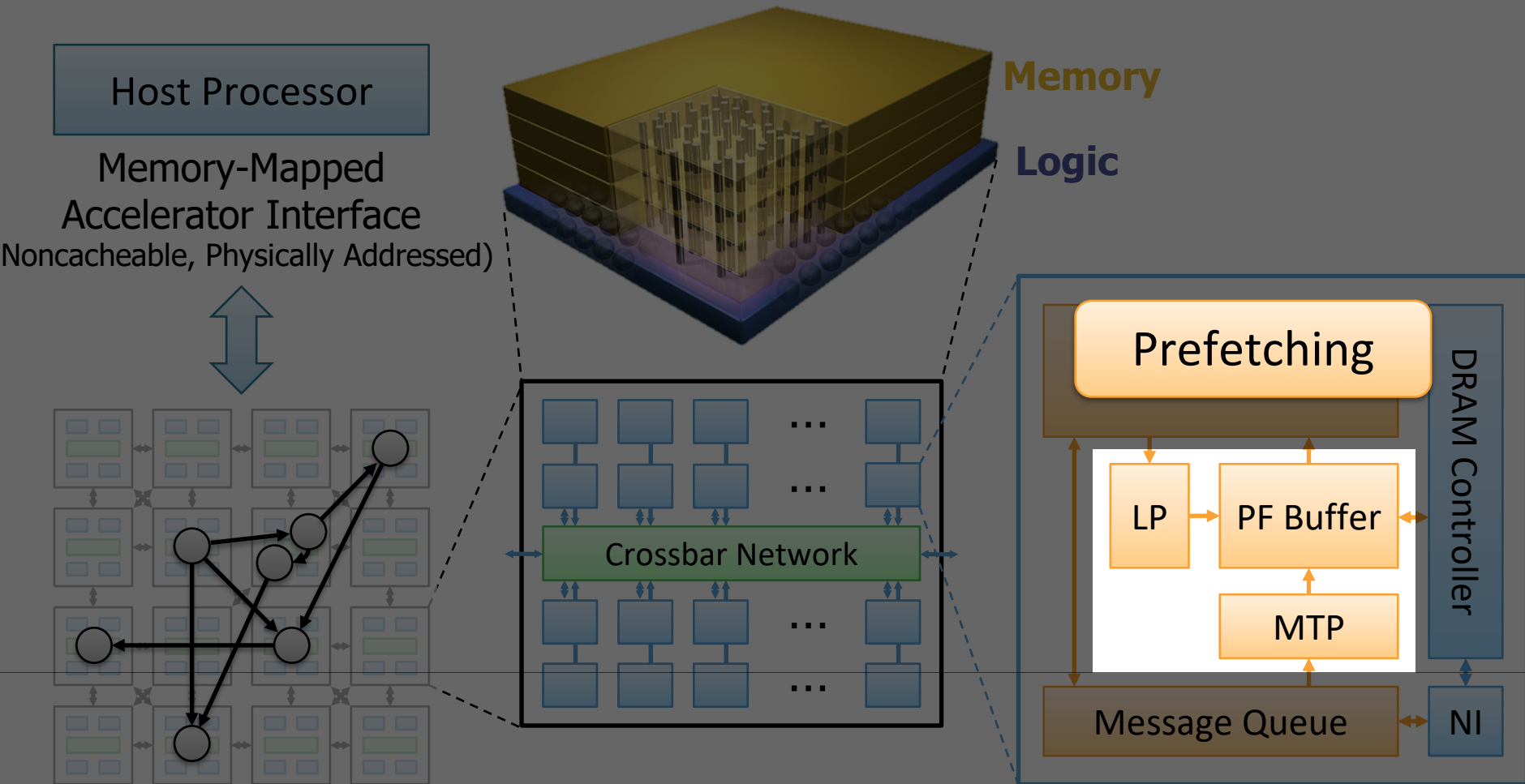
Other "True 3D" technologies
under development

Tesseract System for Graph Processing

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

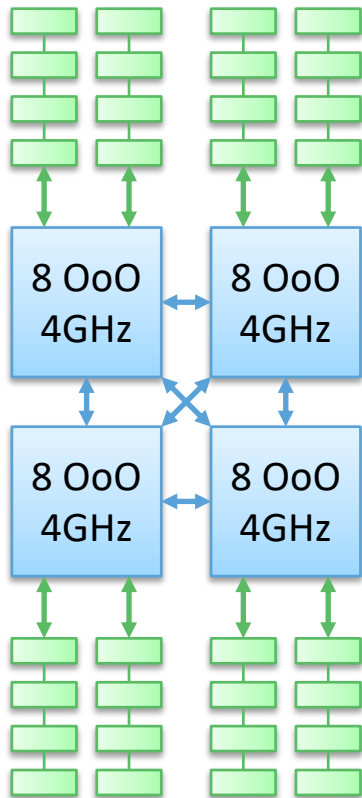


Tesseract System for Graph Processing



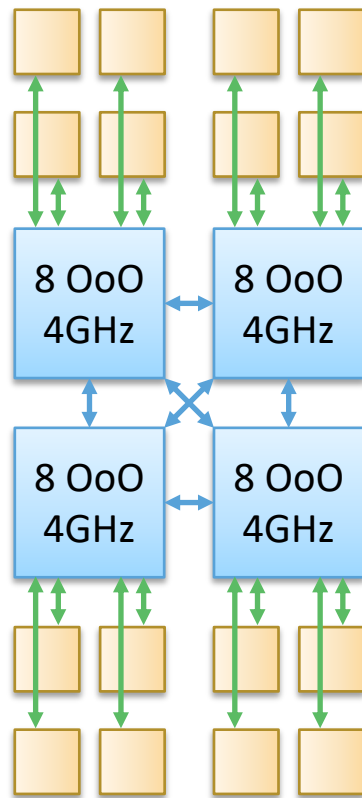
Evaluated Systems

DDR3-OoO



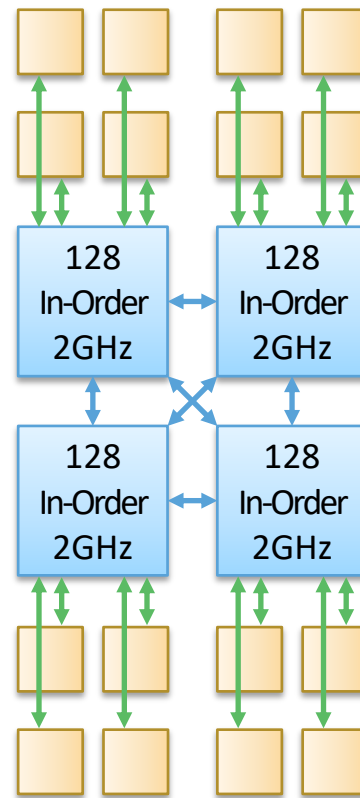
102.4GB/s

HMC-OoO



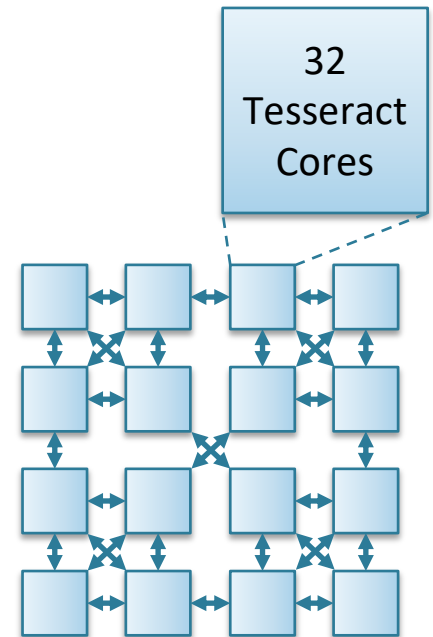
640GB/s

HMC-MC



640GB/s

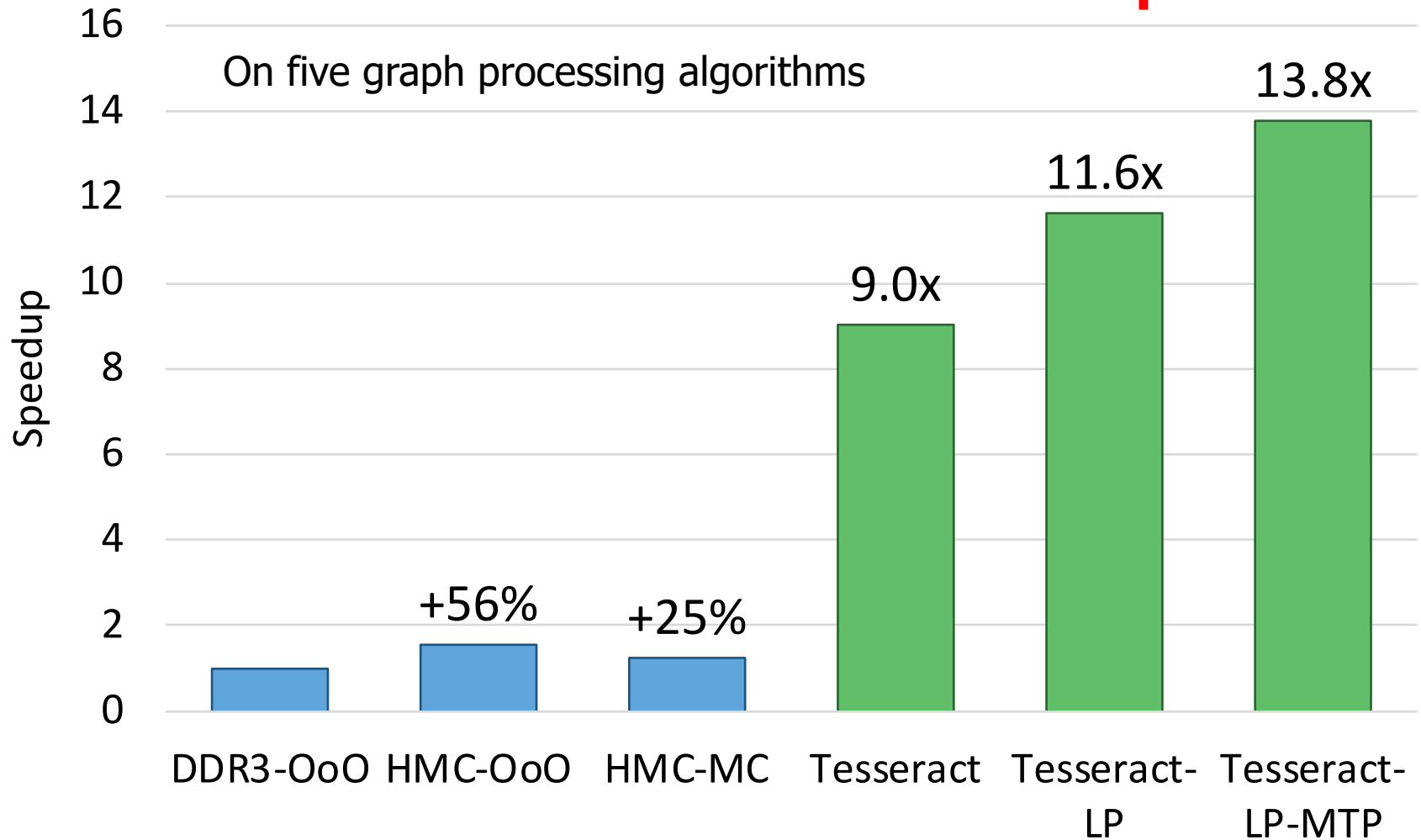
Tesseract



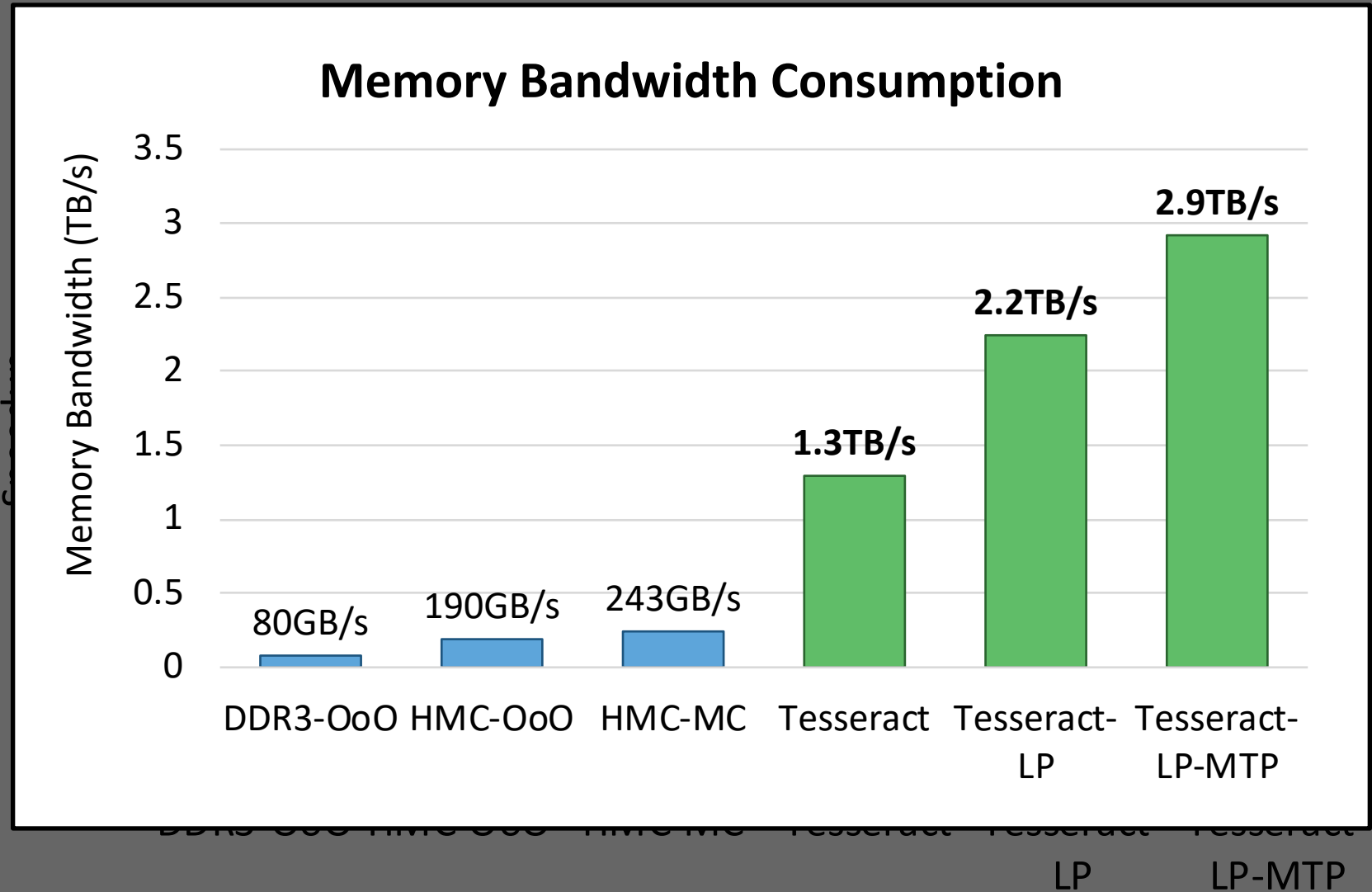
8TB/s

Tesseract Graph Processing Performance

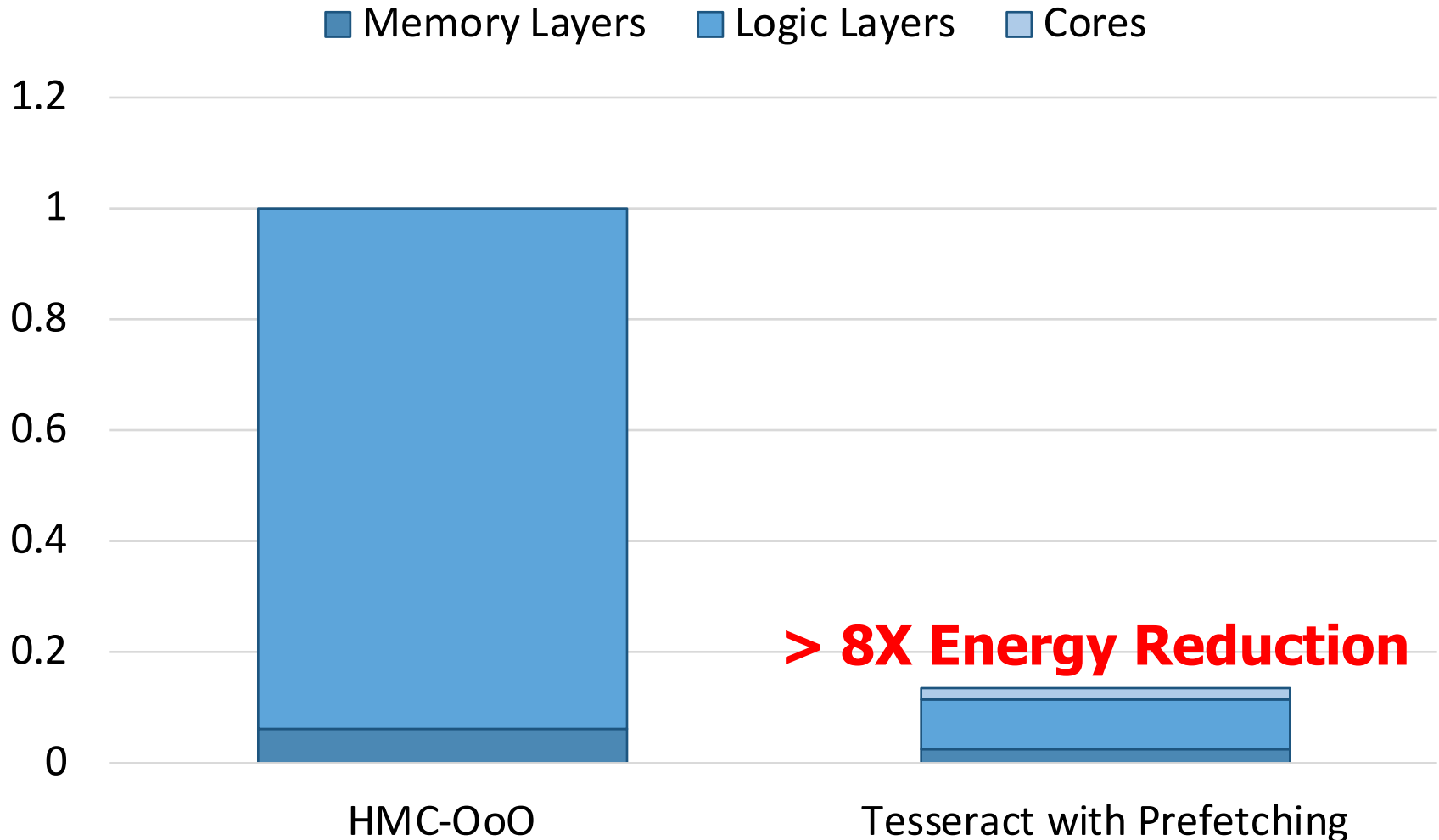
>13X Performance Improvement



Tesseract Graph Processing Performance



Tesseract Graph Processing System Energy



More on Tesseract

- Junwhan Ahn, Sungpack Hong, Sungjoo Yoo, Onur Mutlu, and Kiyoungh Choi,
"A Scalable Processing-in-Memory Accelerator for Parallel Graph Processing"
Proceedings of the 42nd International Symposium on Computer Architecture (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[†] Kiyoungh 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

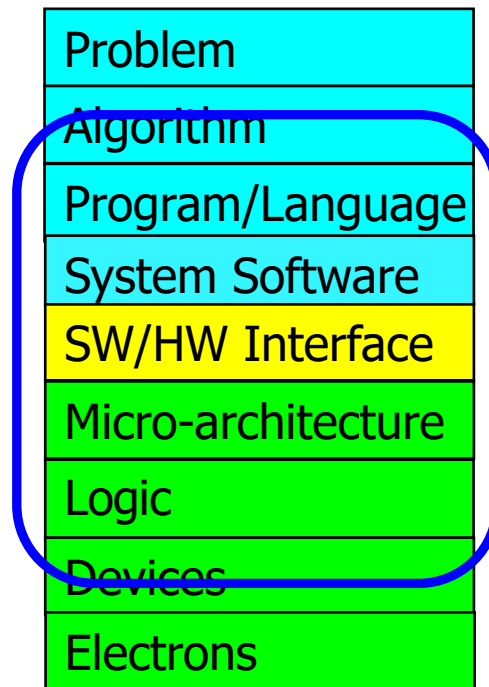
How to Enable Adoption of Processing in Memory

Barriers to Adoption of PIM

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

All can be solved with change of mindset

We Need to Revisit the Entire Stack



We can get there step by step

Computing Architectures with Minimal Data Movement

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

Fundamentally High-Performance **(Data-Centric)** Computing Architectures

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

^a*ETH Zürich*

^b*Carnegie Mellon University*

^c*University of Illinois at Urbana-Champaign*

^d*King 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,

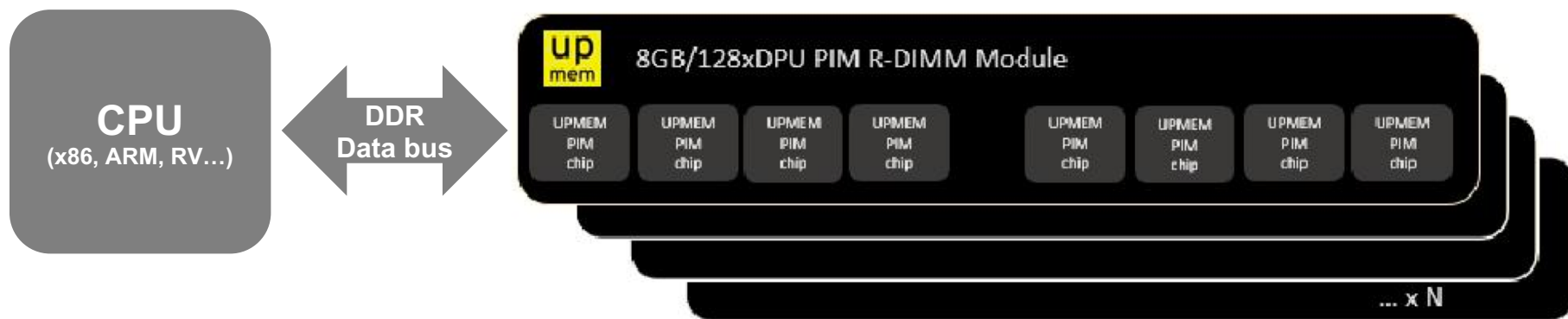
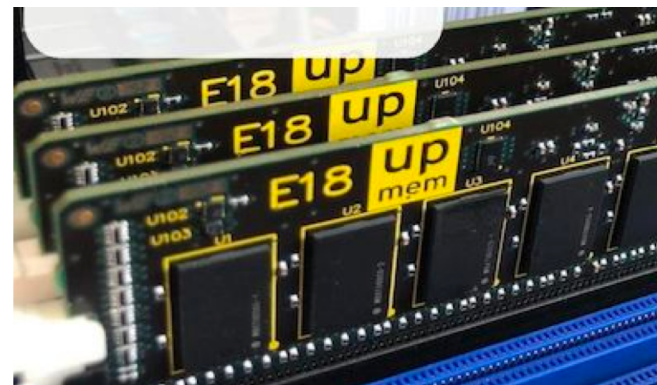
"Processing-in-Memory: A Workload-Driven Perspective"

Invited Article in IBM Journal of Research & Development, Special Issue on Hardware for Artificial Intelligence, to appear in November 2019.

[Preliminary arXiv version]

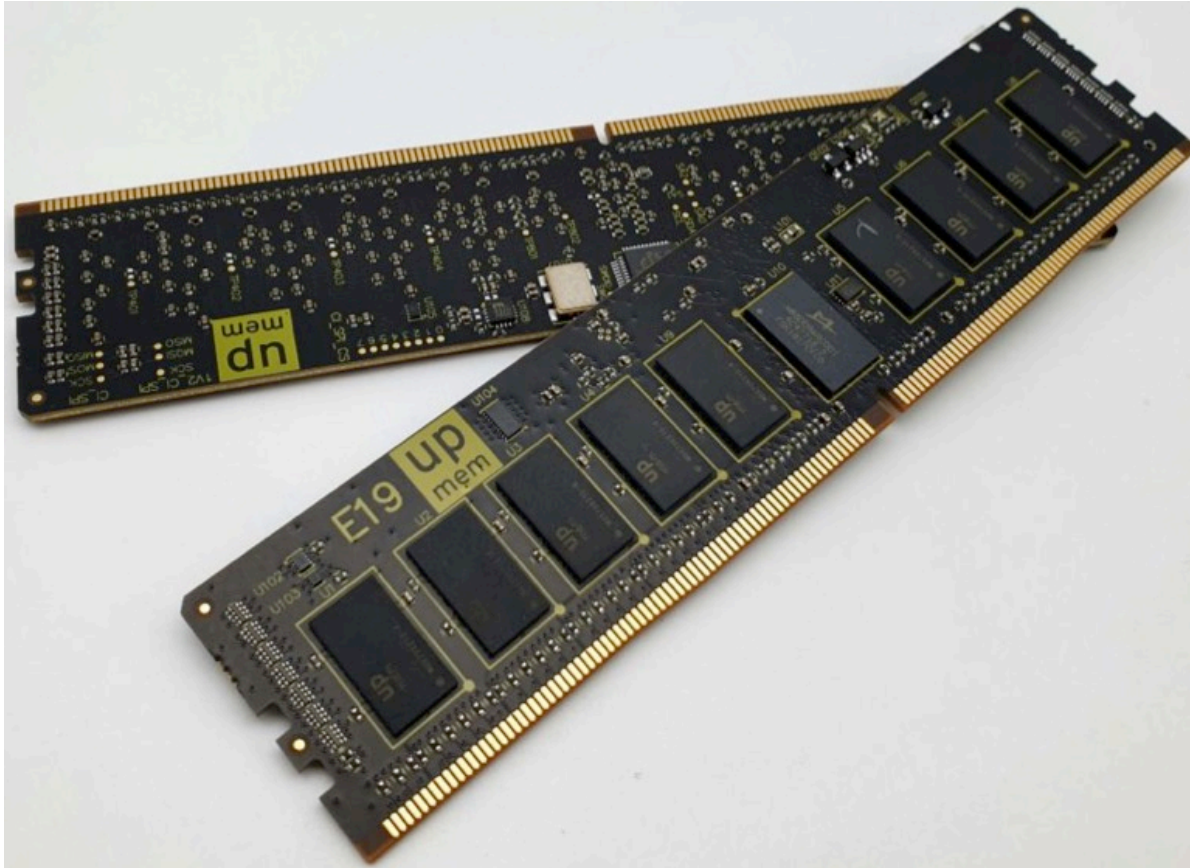
UPMEM Processing-in-DRAM Engine (2019)

- **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



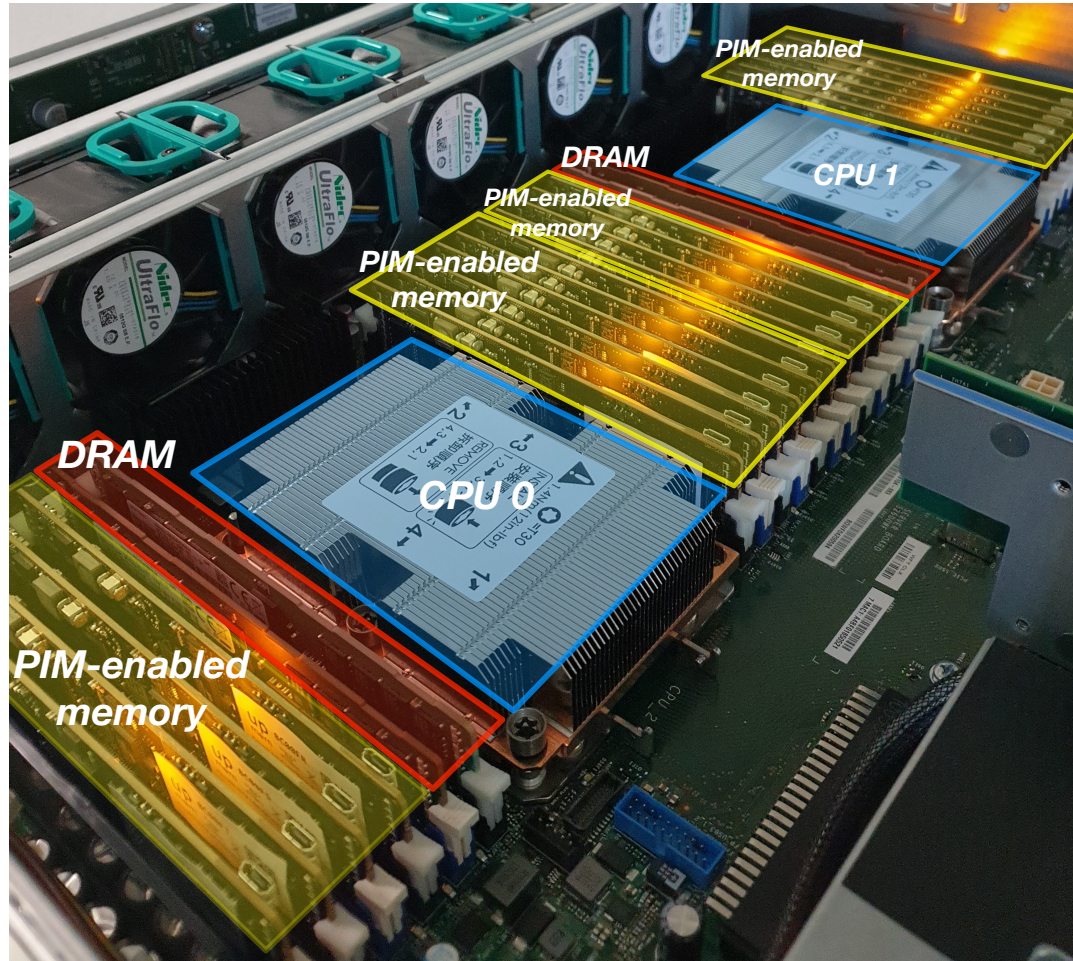
UPMEM Memory Modules

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

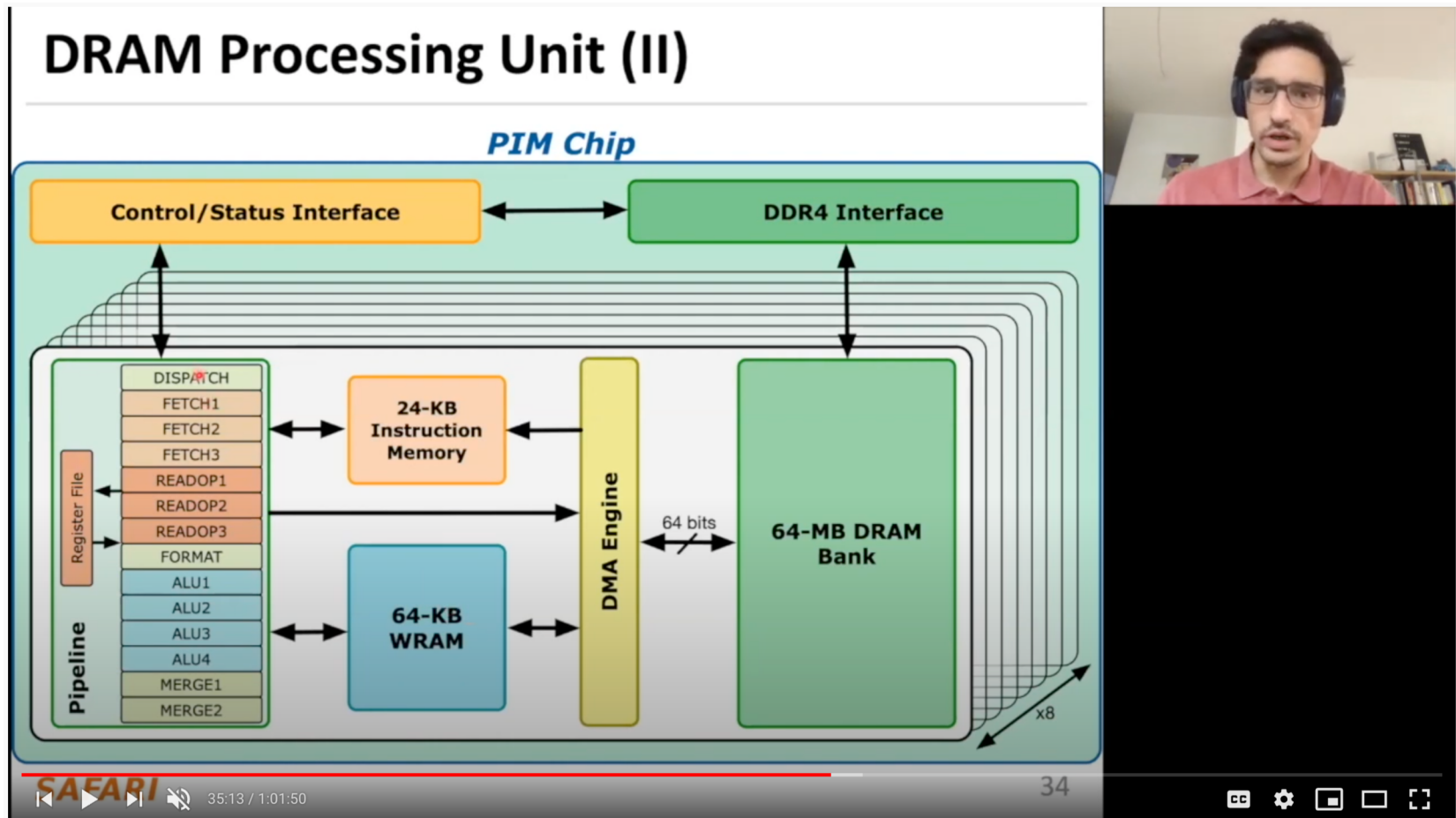


PIM System Organization

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



More on the UPMEM PIM System



ETH ZÜRICH HAUPTGEBÄUDE

Computer Architecture - Lecture 12d: Real Processing-in-DRAM with UPMEM (ETH Zürich, Fall 2020)

1,120 views • Oct 31, 2020

30 0 SHARE SAVE ...



Onur Mutlu Lectures
16.7K subscribers

ANALYTICS

EDIT VIDEO

<https://www.youtube.com/watch?v=Sscy1Wrr22A&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-wisBN&index=26>

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.

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>.

Samsung Function-in-Memory DRAM (2021)



Samsung Develops Industry's First High Bandwidth Memory with AI Processing Power

Korea on February 17, 2021

Audio



Share



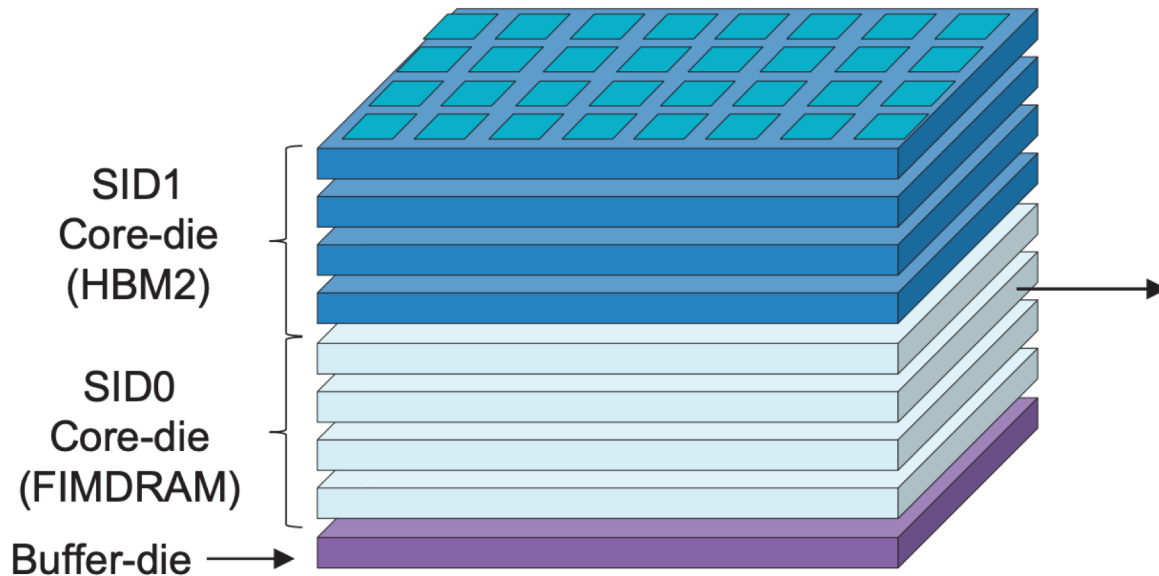
The new architecture will deliver over twice the system performance and reduce energy consumption by more than 70%

Samsung Electronics, the world leader in advanced memory technology, today announced that it has developed the industry's first High Bandwidth Memory (HBM) integrated with artificial intelligence (AI) processing power – the HBM-PIM. The new processing-in-memory (PIM) architecture brings powerful AI computing capabilities inside high-performance memory, to accelerate large-scale processing in data centers, high performance computing (HPC) systems and AI-enabled mobile applications.

Kwangil Park, senior vice president of Memory Product Planning at Samsung Electronics stated, "Our groundbreaking HBM-PIM is the industry's first programmable PIM solution tailored for diverse AI-driven workloads such as HPC, training and inference. We plan to build upon this breakthrough by further collaborating with AI solution providers for even more advanced PIM-powered applications."

Samsung Function-in-Memory DRAM (2021)

■ FIMDRAM based on HBM2



[3D Chip Structure of HBM with FIMDRAM]

Chip Specification

128DQ / 8CH / 16 banks / BL4

32 PCU blocks (1 FIM block/2 banks)

1.2 TFLOPS (4H)

**FP16 ADD /
Multiply (MUL) /
Multiply-Accumulate (MAC) /
Multiply-and- Add (MAD)**

ISSCC 2021 / SESSION 25 / DRAM / 25.4

25.4 A 20nm 6GB Function-In-Memory DRAM, Based on HBM2 with a 1.2TFLOPS Programmable Computing Unit Using Bank-Level Parallelism, for Machine Learning Applications

Young-Cheon Kwon¹, Suk Han Lee¹, Jaehoon Lee¹, Sang-Hyuk Kwon¹, Je Min Ryu¹, Jong-Pil Son¹, Seongil O¹, Hak-Soo Yu¹, Haesuk Lee¹, Soo Young Kim¹, Youngmin Cho¹, Jin Guk Kim¹, Jongyoon Choi¹, Hyun-Sung Shin¹, Jin Kim¹, BengSeng Phuah¹, HyoungMin Kim¹, Myeong Jun Song¹, Ahn Choi¹, Daeho Kim¹, SooYoung Kim¹, Eun-Bong Kim¹, David Wang², Shinhaeng Kang¹, Yuhwan Ro³, Seungwoo Seo³, JoonHo Song³, Jaeyoun Youn¹, Kyomin Sohn¹, Nam Sung Kim¹

¹Samsung Electronics, Hwaseong, Korea

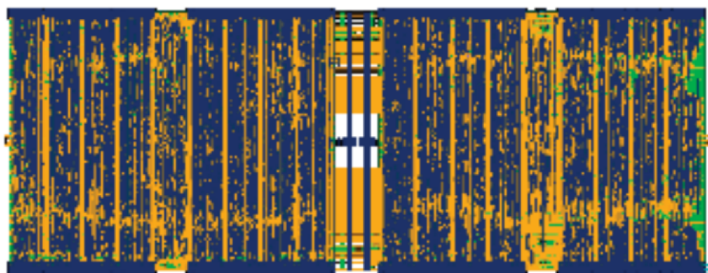
²Samsung Electronics, San Jose, CA

³Samsung Electronics, Suwon, Korea

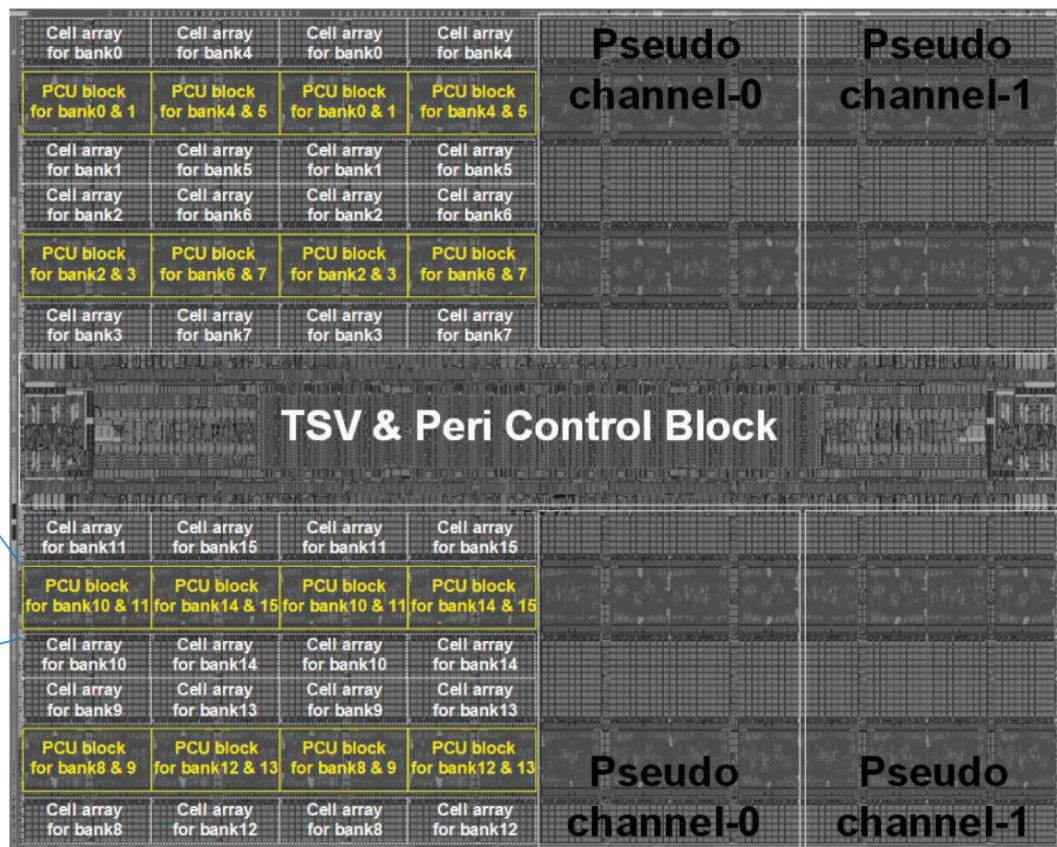
Samsung Function-in-Memory DRAM (2021)

Chip Implementation

- Mixed design methodology to implement FIMDRAM
 - Full-custom + Digital RTL



[Digital RTL design for PCU block]



ISSCC 2021 / SESSION 25 / DRAM / 25.4

25.4 A 20nm 6Gb Function-In-Memory DRAM, Based on HBM2 with a 1.2TFLOPS Programmable Computing Unit Using Bank-Level Parallelism, for Machine Learning Applications

Young-Cheon Kwon¹, Suk Han Lee¹, Jaehoon Lee¹, Sang-Hyuk Kwon¹, Je Min Ryu¹, Jong-Pil Son¹, Seongil O¹, Hak-Soo Yu¹, Haesuk Lee¹, Soo Young Kim¹, Youngmin Cho¹, Jin Guk Kim¹, Jongyeon Choi¹, Hyun-Sung Shim¹, Jin Kim¹, BengSeng Phuah¹, HyounMin Kim¹, Myeong Jun Song¹, Ahn Chai¹, Daeho Kim¹, SooYoung Kim¹, Eun-Bong Kim¹, David Wang¹, Shinhaeng Kang¹, Yuhwan Ro¹, Seungwoo Seo¹, JoonHo Song¹, Jaeyoun Yoon¹, Kyomin Sohn¹, Nam Sung Kim¹

¹Samsung Electronics, Hwaseong, Korea

²Samsung Electronics, San Jose, CA

³Samsung Electronics, Suwon, Korea

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=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-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=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-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=AIE1rD9G_YU&list=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-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=PL5Q2soXY2Zi9xidyIgBxUz7xRPS-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 Processing in Memory

- Onur Mutlu,

"Memory-Centric Computing Systems"

Invited Tutorial at *66th International Electron Devices 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>

Memory-Centric Computing Systems



Onur Mutlu

omutlu@gmail.com

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

12 December 2020

IEDM Tutorial

SAFARI

ETH zürich

Carnegie Mellon



0:06 / 1:51:05



IEDM 2020 Tutorial: Memory-Centric Computing Systems, Onur Mutlu, 12 December 2020

1,641 views • Dec 23, 2020

48 0 SHARE SAVE ...



Onur Mutlu Lectures
13.9K subscribers

ANALYTICS

EDIT VIDEO

<https://www.youtube.com/onurmutlulectures>

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

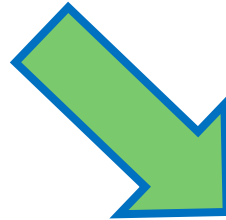
Senol Cali+, “**Nanopore Sequencing Technology and Tools for Genome Assembly: Computational Analysis of the Current State, Bottlenecks and Future Directions**,” *Briefings in Bioinformatics*, 2018.

[[Preliminary arxiv.org version](#)]

Future of Genome Sequencing & Analysis



MinION from ONT



SmidgION from ONT

Accelerating Genome Analysis: Overview

- Mohammed Alser, Zülal Bingöl, 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 [Technion](#), Virtual, 26 January 2021.
[[Slides \(pptx\)](#) ([pdf](#))]
[[Talk Video](#) (1 hour 37 minutes, including Q&A)]
[[Related Invited Paper \(at IEEE Micro, 2020\)](#)]

Insight: Shifting a String Helps Similarity Search

7 matches 1 mismatch

ISTANBUL

ISTNBUL

ISTNBUL

81

46:08 / 1:37:37

Onur Mutlu - Invited Lecture @Technion: Accelerating Genome Analysis: A Primer on an Ongoing Journey

566 views · Premiered Feb 6, 2021

31 0 SHARE SAVE ...

Onur Mutlu Lectures
13.9K subscribers

ANALYTICS 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=PL5Q2soXY2Zi9E2bBVAgCqLgwiDRQDTyId>

If You Want More of This...

Multiple Future Options

- Take the **Bachelor's Seminar in Computer Architecture**
 - ❑ Offered **every Fall and Spring**
 - ❑ https://safari.ethz.ch/architecture_seminar
- Take the Master's-level **Computer Architecture** course
 - ❑ Offered **every Fall**
 - ❑ <https://safari.ethz.ch/architecture>
- Take our **P&S Courses in Computer Architecture**
 - ❑ Offered **every Fall and Spring**
 - ❑ https://safari.ethz.ch/projects_and_seminars
- **Do research with me and my SAFARI research group**
 - ❑ **Bachelor's/Master's theses, semester projects, internships**
 - ❑ Opportunity to be a part of a vibrant, cutting-edge group
 - ❑ <https://safari.ethz.ch/>

SAFARI Research Group

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safari.ethz.ch

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-january-2021/>



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<https://safari.ethz.ch>

SAFARI Newsletter April 2020 Edition

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



[View in your browser](#)

Think Big, Aim High



Dear SAFARI friends,

2019 and the first three months of 2020 have been very positive eventful times for SAFARI.

SAFARI Newsletter January 2021 Edition

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



SAFARI
SAFARI Research Group

Newsletter
January 2021

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



Dear SAFARI friends,

Happy New Year! We are excited to share our group highlights with you in this second edition of the SAFARI newsletter (You can find the first edition from April 2020 [here](#)). 2020 has

Bachelor's Seminar in Computer Architecture

- Fall 2021
- 2 credit units
- **Rigorous seminar on fundamental and cutting-edge topics in computer architecture**
- **Critical presentation, review, and discussion of major works in computer architecture**
 - We will cover many ideas & issues, analyze their tradeoffs, perform **critical thinking**, discussion and brainstorming
- Participation, presentation, synthesis report
- You can register for the course online
- https://safari.ethz.ch/architecture_seminar

Computer Architecture Course

- Fall 2021
- 8 credit units




- **Covers many cutting-edge topics in memory systems, interconnects, multiprocessors, heterogeneous systems, PIM, and more...**

- Many topics we could not cover in this course

- Exam, lab assignments, homeworks
- You can register for the course online
- <https://safari.ethz.ch/architecture/fall2020/doku.php?id=schedule>

Labs

Lab 1: Caching (Due: Friday, October 9 at midnight)

-  Handout
-  Code (Zip)
-  Moodle Submission Page



Lab 2: Memory Hierarchy (Due: Sun. 25.10)

-  Handout
-  Code (Zip)
-  Submission

Lab 3: Memory Scheduling (Due: Tue. 17.11)

-  Handout
-  Code (Zip)
-  Submission

Lab 4: Prefetching (Due: Mon. 21.12)

-  Handout
-  Submission

Lab 5: Multicore and Cache Coherence (BONUS) (Due: Fri. 29.1)

-  Handout
-  Code (Zip)
-  Submission

Projects & Seminars (P&S) Courses

- Fall 2021

- **P&S Courses in Computer Architecture**

- Offered **every Fall and Spring**
- **Short lectures + Hands-on project**
- Lightweight but deep courses on specific topics you can choose from
 - **Memory**
 - Understanding and Improving Modern **DRAM Performance, Reliability, and Security** with Hands-On Experiments
 - Designing and Evaluating **Memory Systems and Modern Software Workloads** with Ramulator
 - **Storage**
 - Understanding and Designing Modern **NAND Flash-Based Solid-State Drives (SSDs)** by Building a Practical SSD Simulator
 - **Heterogenous Computing: GPU, FPGA, PIM**
 - Hands-on Acceleration on **Heterogeneous Computing Systems**
 - Exploring the **Processing-in-Memory** Paradigm for Future Computing Systems
 - **Applications: Bioinformatics & Machine Learning**
 - **Accelerating Genome Analysis** with FPGAs, GPUs, and New Execution Paradigms
 - **Genome Sequencing on Mobile** Devices

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

Doing Research with Us

- If you are interested in **learning more** and **doing research** in Computer Architecture, three suggestions:
 - **Email me** with your interest (CC: Juan)
 - Take **the seminar course or the “Computer Architecture” course**
 - **Do readings** and assignments on your own

- There are **many exciting projects and research positions**, e.g.:
 - Memory systems
 - Hardware security
 - GPUs, FPGAs, heterogeneous systems, ...
 - New execution paradigms (e.g., in-memory computing)
 - Security-architecture-reliability-energy-performance interactions
 - Architectures for medical/health/genomics
 - A limited list is here: <https://safari.ethz.ch/theses/>

How To Do Projects with SAFARI?

- **Be excited about topics we work on**
- **Get in touch with me & my students/postdocs**
- Browse projects here (not all projects are listed)
 - <https://safari.ethz.ch/theses/>
- Take our courses and enjoy them 😊
- Read our papers and get in touch with authors
 - <https://people.inf.ethz.ch/omutlu/projects.htm>
- **Email us & apply here:** <https://safari.ethz.ch/work-with-us/>

Food for Thought: Two Quotes

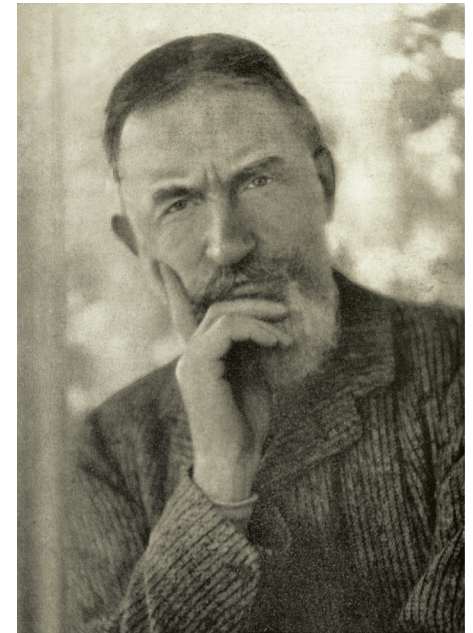
The reasonable man adapts himself to the world;

The unreasonable one persists in trying adapt the world to himself.

Therefore, all progress depends on the unreasonable man.

George Bernard Shaw

Progress is impossible without change,
and those who cannot change their minds
cannot change anything.



Digital Design & Computer Arch.

Lecture 26b: Epilogue

Prof. Onur Mutlu

ETH Zürich

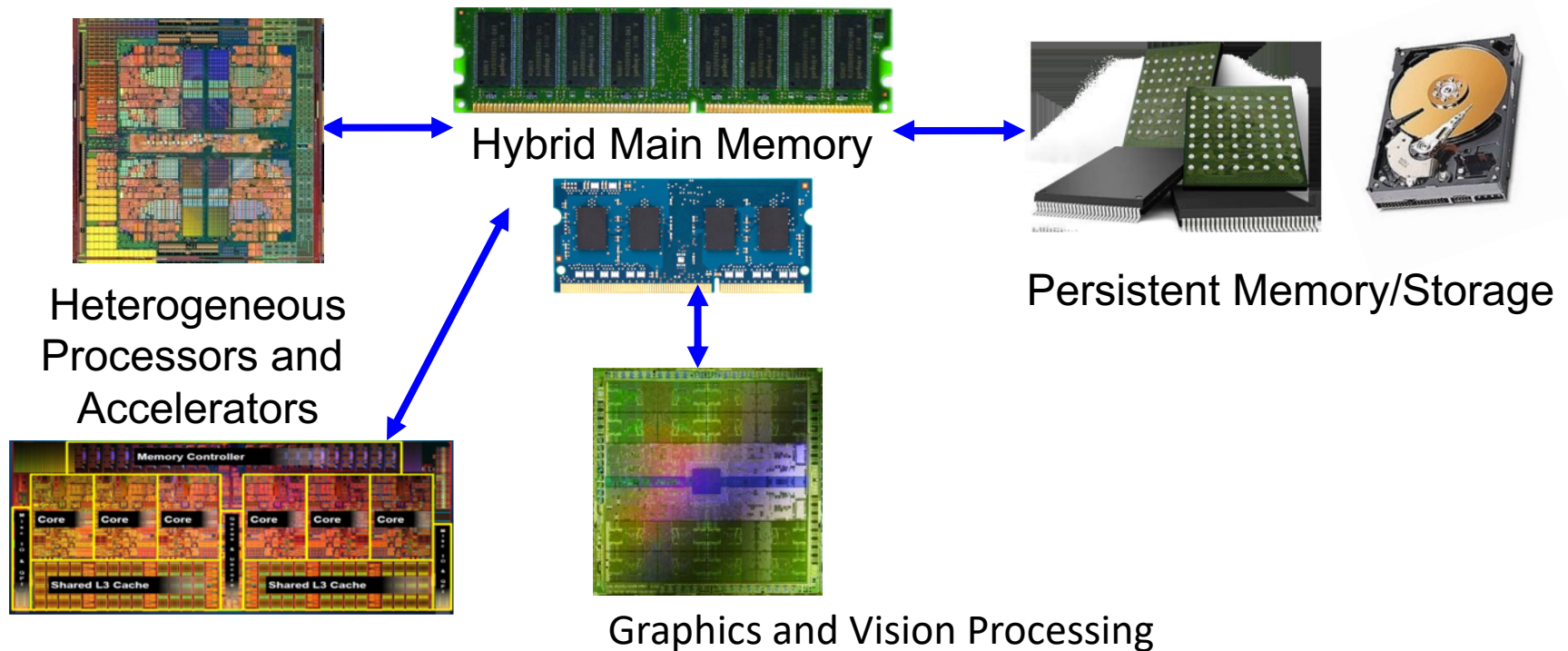
Spring 2021

4 June 2021

Research in Computer Architecture

Current Research Mission

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



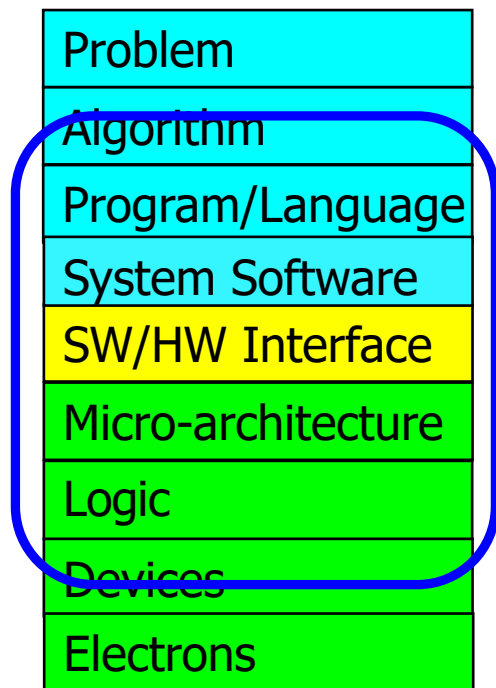
Build fundamentally better architectures

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**

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

Research & Teaching: Some Overview Talks

<https://www.youtube.com/onurmutlulectures>

■ Future Computing Architectures

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

■ Enabling In-Memory Computation

- https://www.youtube.com/watch?v=njX_14584Jw&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=16

■ Accelerating Genome Analysis

- https://www.youtube.com/watch?v=r7sn41IH-4A&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=41

■ Rethinking Memory System Design

- https://www.youtube.com/watch?v=F7xZLNMIY1E&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=3

■ Intelligent Architectures for Intelligent Machines

- https://www.youtube.com/watch?v=c6_LgzuNdkw&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=25

■ The Story of RowHammer

- https://www.youtube.com/watch?v=sgd7PHQQ1AI&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=39

An Interview on Research and Education

- Computing Research and Education (@ ISCA 2019)
 - https://www.youtube.com/watch?v=8ffSEKZhmvo&list=PL5Q2soXY2Zi_4oP9LdL3cc8G6NIjD2Ydz
- Maurice Wilkes Award Speech (10 minutes)
 - https://www.youtube.com/watch?v=tcQ3zZ3JpuA&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJl&index=15

More Thoughts and Suggestions

- Onur Mutlu,
["Some Reflections \(on DRAM\)"](#)
*Award Speech for [ACM SIGARCH Maurice Wilkes Award](#), at the **ISCA** Awards Ceremony, Phoenix, AZ, USA, 25 June 2019.*
[\[Slides \(pptx\) \(pdf\)\]](#)
[\[Video of Award Acceptance Speech \(Youtube; 10 minutes\) \(Youku; 13 minutes\)\]](#)
[\[Video of Interview after Award Acceptance \(Youtube; 1 hour 6 minutes\) \(Youku; 1 hour 6 minutes\)\]](#)
[\[News Article on "ACM SIGARCH Maurice Wilkes Award goes to Prof. Onur Mutlu"\]](#)

- Onur Mutlu,
["How to Build an Impactful Research Group"](#)
*[57th Design Automation Conference Early Career Workshop \(**DAC**\)](#), Virtual, 19 July 2020.*
[\[Slides \(pptx\) \(pdf\)\]](#)

Referenced Papers, Talks, Artifacts

- All are available at

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

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

<https://www.youtube.com/onurmutlulectures>

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

Some Basics of Research

How To Do Research & Advanced Dev.

- We will talk a lot about this in this course
- Learning **by example**
 - Reading and evaluating strong and seminal papers & designs
- Learning **by doing**
 - Semester-long research/design projects, masters' projects, PhD thesis
- Learning **by open, critical discussions**
 - Paper reading groups, frequent brainstorming and discussions
 - Design sessions
 - Collaborations

What Is The Goal of Research?

- To generate new insight
 - that can enable what previously did not exist
- Research is a hunt for insight that can eventually impact the world

Some Basic Advice for Good Research

- Choose great problems to solve: Have great taste
 - Difficult
 - Important
 - High impact
- Read heavily and critically
- Think big (out of the box)
 - Do not restrain yourself to tweaks or constraints of today
 - Yet, think about adoption issues
- Aim high
- Write and present extremely well



Looking here for lost keys

Lost keys here

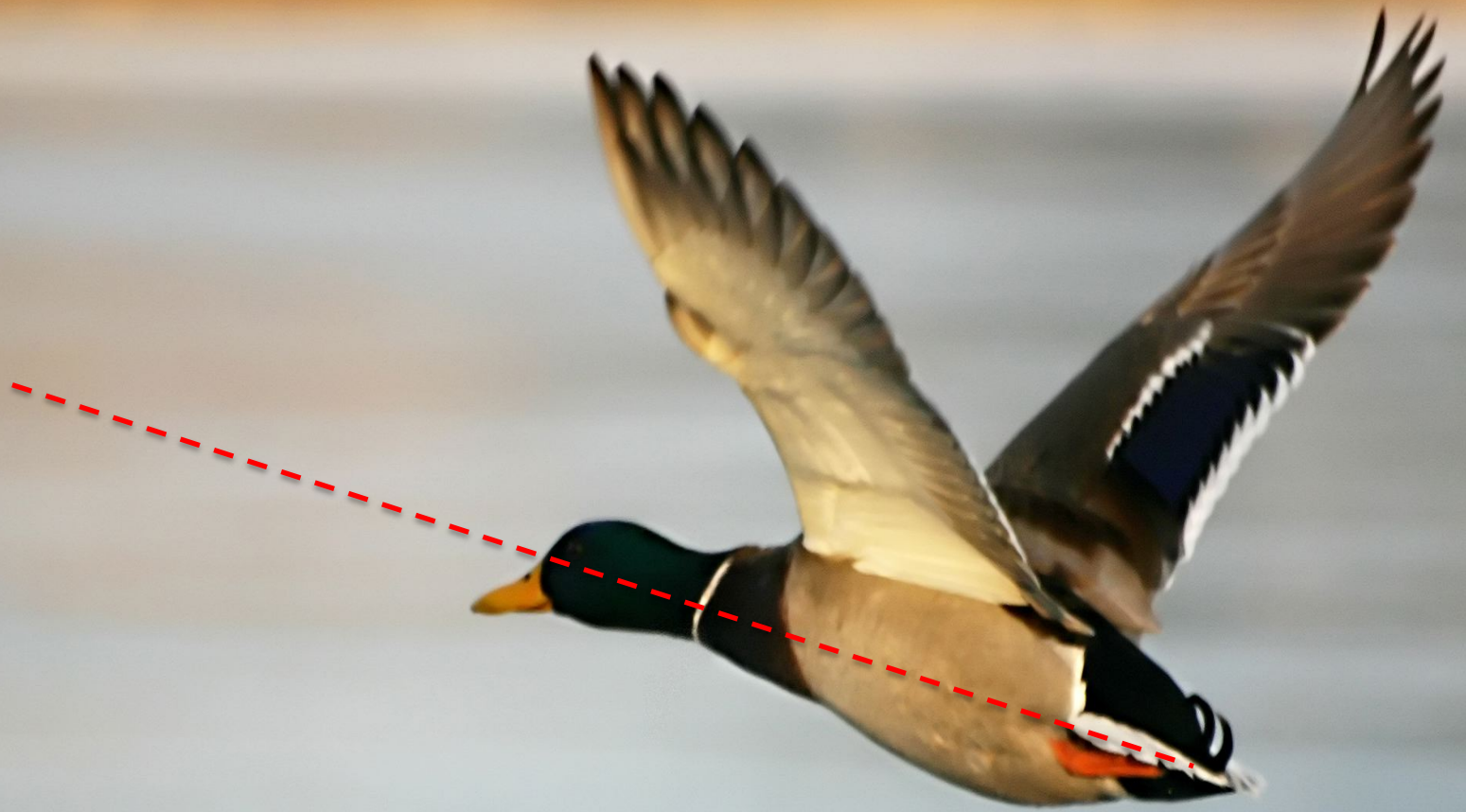


Looking here

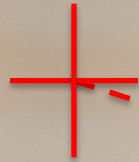
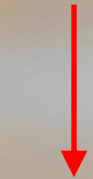




Current Architecture Practice

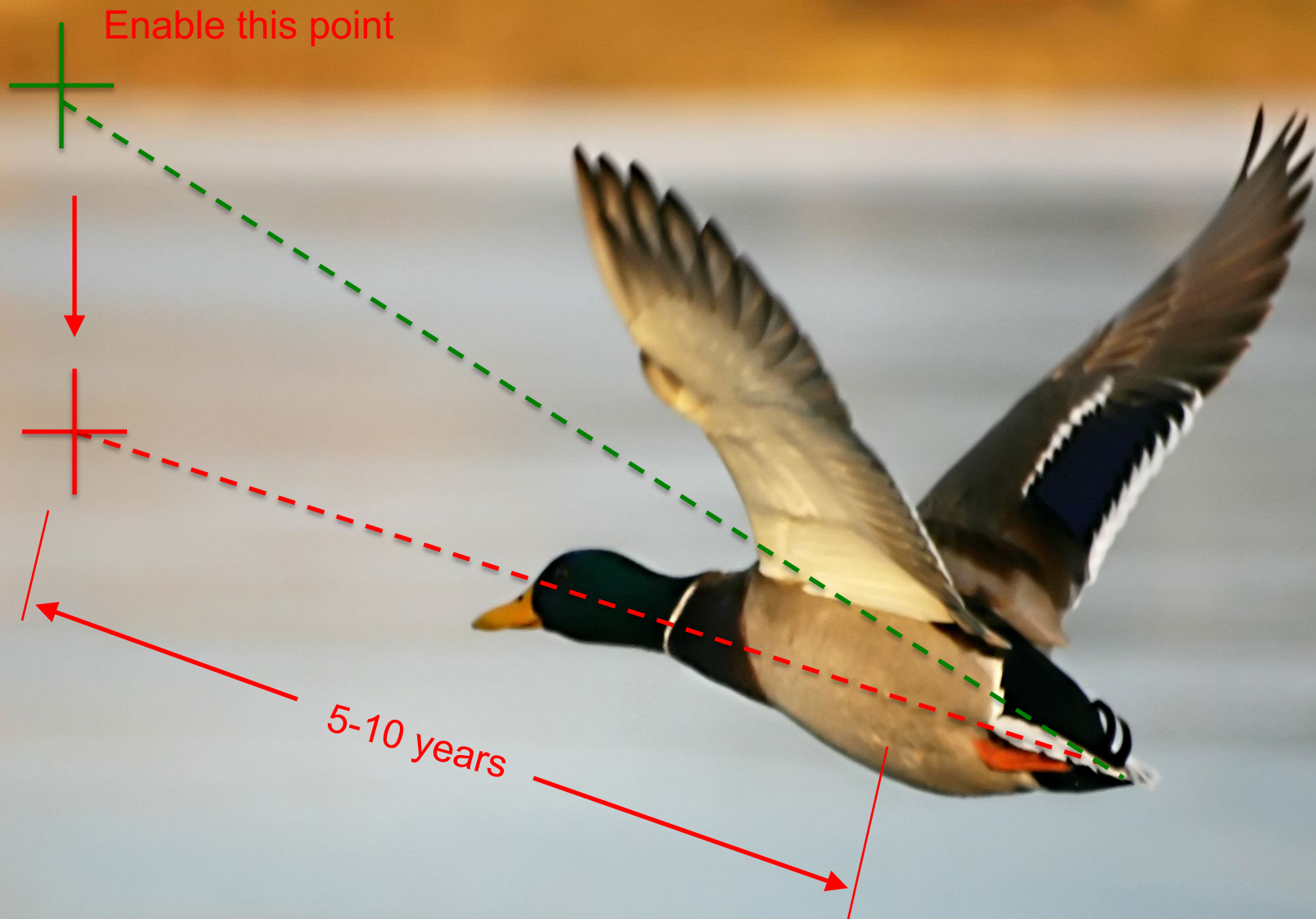


Aim Here



5-10 years





The Research Formula



$$ROI = \frac{reward}{risk \times effort}$$

Reward

If you are wildly successful, what difference will it make?

$$ROI = \frac{\text{reward}}{\text{risk} \times \text{effort}}$$

Effort

Learn as much as possible with as little work as possible

$$ROI = \frac{reward}{risk \times effort}$$

Effort

Do the minimum analysis and experimentation necessary to make a point

$$ROI = \frac{reward}{risk \times effort}$$

Research is a
hunt for insight

Need to get off the beaten
path to find new insights



Recommended Talk

- Bill Dally, Moving the needle: Effective Computer Architecture Research in Academy and Industry
ISCA 2010 Keynote Talk.
- Acknowledgment: Past few slides are from this talk

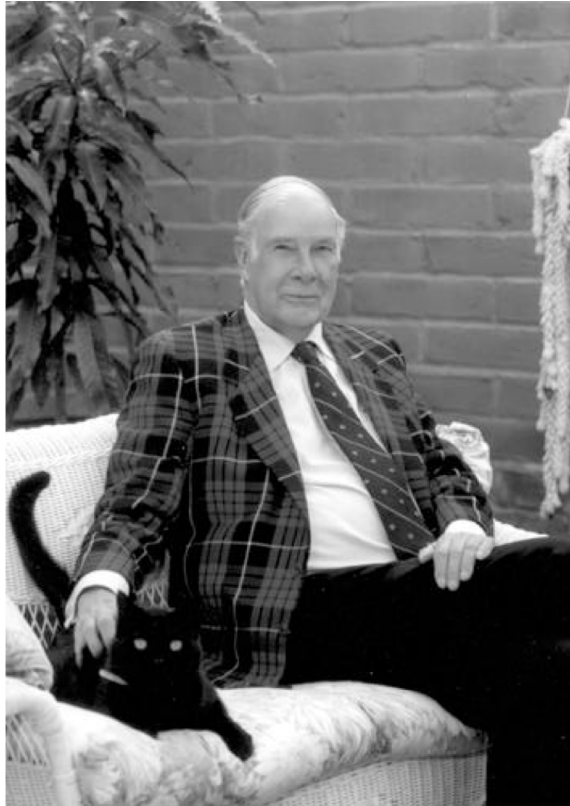
What transfers is *insight*

Not academic design

Not performance numbers



More Good Advice



“The purpose of computing is
insight, not numbers”

Richard Hamming

Some Personal Examples

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>

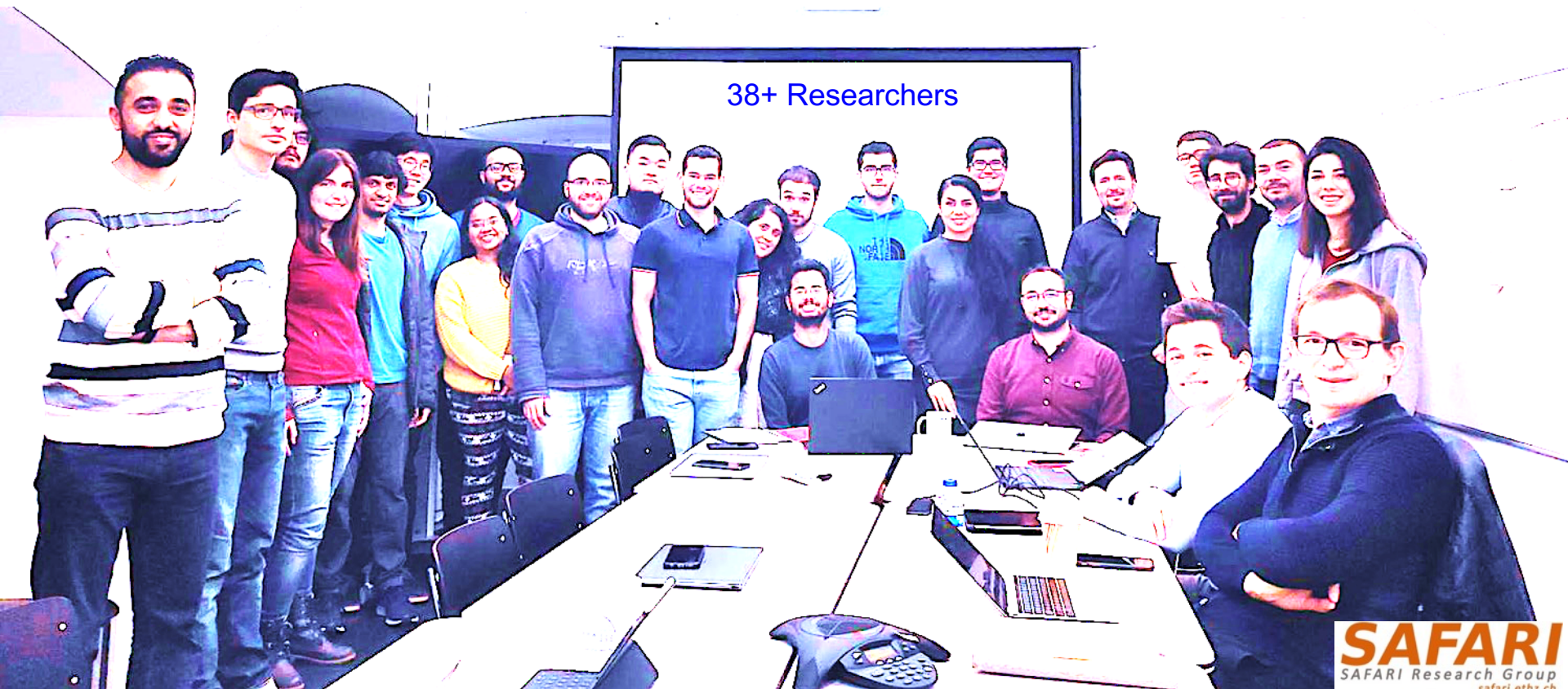
■ Research and Teaching in:

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

Onur Mutlu's SAFARI Research Group

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

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



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<https://safari.ethz.ch>

SAFARI Newsletter January 2021 Edition

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



SAFARI
SAFARI Research Group

Newsletter
January 2021

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



Dear SAFARI friends,

Happy New Year! We are excited to share our group highlights with you in this second edition of the SAFARI newsletter (You can find the first edition from April 2020 [here](#)). 2020 has

Principle: Teaching and Research

...

Teaching drives Research

Research drives Teaching

...

Principle: Insight and Ideas

Focus on Insight

Encourage New Ideas

Principle: Learning and Scholarship

Focus on
learning and scholarship

Principle: Environment of Freedom

Create an environment
that values

free exploration,
openness, collaboration,
hard work, creativity

Principle: Learning and Scholarship

The quality of your work
defines your impact

Research & Teaching: Some Overview Talks

<https://www.youtube.com/onurmutlulectures>

■ Future Computing Architectures

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

■ Enabling In-Memory Computation

- https://www.youtube.com/watch?v=njX_14584Jw&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=16

■ Accelerating Genome Analysis

- https://www.youtube.com/watch?v=r7sn41IH-4A&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=41

■ Rethinking Memory System Design

- https://www.youtube.com/watch?v=F7xZLNMIY1E&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=3

■ Intelligent Architectures for Intelligent Machines

- https://www.youtube.com/watch?v=c6_LgzuNdkw&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=25

■ The Story of RowHammer

- https://www.youtube.com/watch?v=sgd7PHQQ1AI&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=39

An Interview on Research and Education

- Computing Research and Education (@ ISCA 2019)
 - https://www.youtube.com/watch?v=8ffSEKZhmvo&list=PL5Q2soXY2Zi_4oP9LdL3cc8G6NIjD2Ydz

- Maurice Wilkes Award Speech (10 minutes)
 - https://www.youtube.com/watch?v=tcQ3zZ3JpuA&list=PL5Q2soXY2Zi8D_5MGV6EnXEJHnV2YFBJI&index=15

More Thoughts and Suggestions

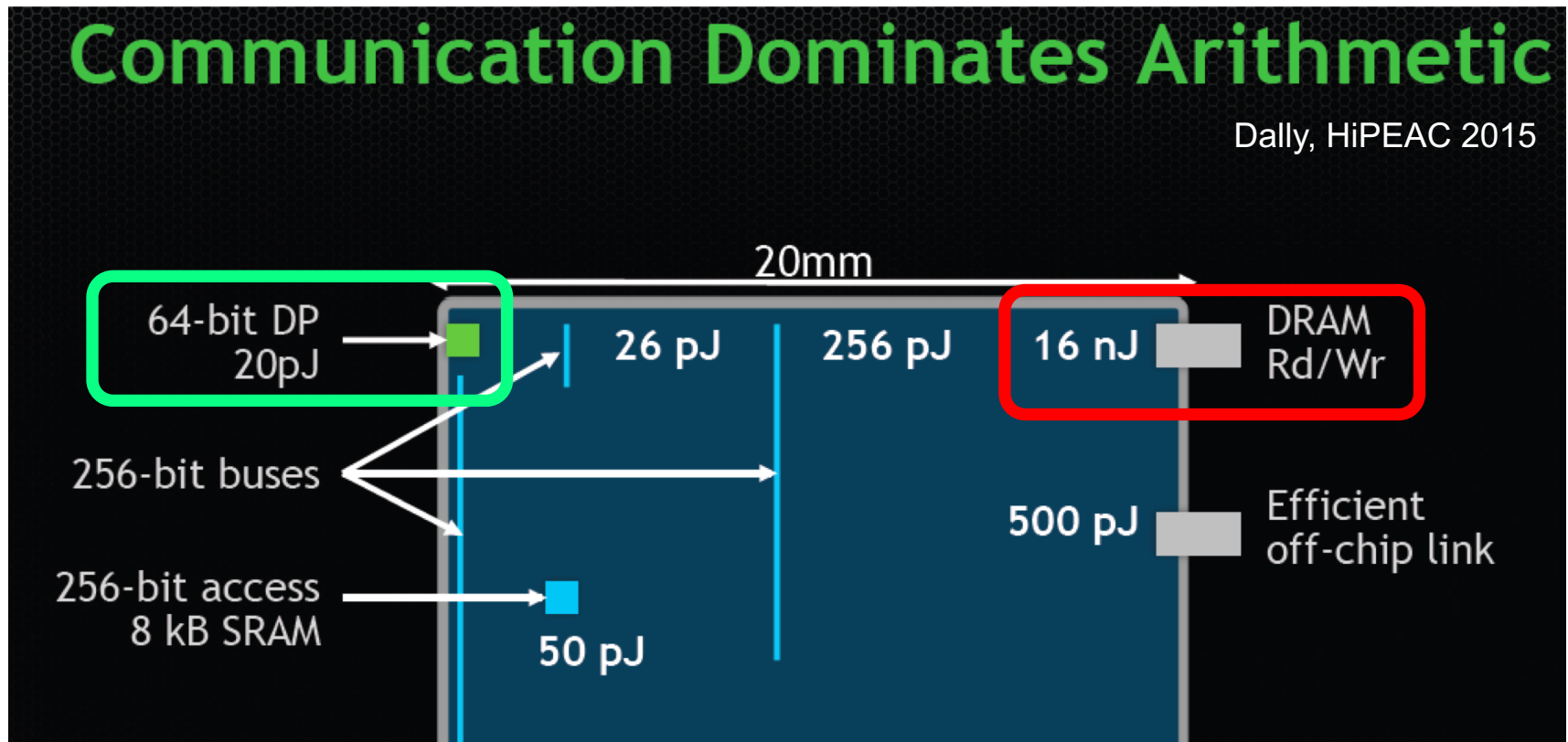
- Onur Mutlu,
["Some Reflections \(on DRAM\)"](#)
*Award Speech for [ACM SIGARCH Maurice Wilkes Award](#), at the **ISCA** Awards Ceremony, Phoenix, AZ, USA, 25 June 2019.*
[\[Slides \(pptx\) \(pdf\)\]](#)
[\[Video of Award Acceptance Speech \(Youtube; 10 minutes\) \(Youku; 13 minutes\)\]](#)
[\[Video of Interview after Award Acceptance \(Youtube; 1 hour 6 minutes\) \(Youku; 1 hour 6 minutes\)\]](#)
[\[News Article on "ACM SIGARCH Maurice Wilkes Award goes to Prof. Onur Mutlu"\]](#)

- Onur Mutlu,
["How to Build an Impactful Research Group"](#)
*[57th Design Automation Conference Early Career Workshop \(**DAC**\)](#), Virtual, 19 July 2020.*
[\[Slides \(pptx\) \(pdf\)\]](#)

Data Movement vs. Computation Energy

Communication Dominates Arithmetic

Dally, HiPEAC 2015



A memory access consumes $\sim 100\text{-}1000\times$ the energy of a complex addition