

Tutorial on Memory-Centric Computing: Programming PNM Systems

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<https://geraldofojunior.github.io>

PPoPP 2025

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Agenda

- Processing-Near-Memory Systems: Developments from Academia & Industry
- **Programming Processing-Near-Memory Systems**
- Coffee Break
- Processing-Using-Memory Systems for Bulk Bitwise Operations
- **Ataberk Olgun:**
Infrastructure for Processing-Using-Memory Research
- **Invited Talk by Prof. John Kim:**
Is it Memory-Centric or Communication-Centric?

Eliminating the Adoption Barriers

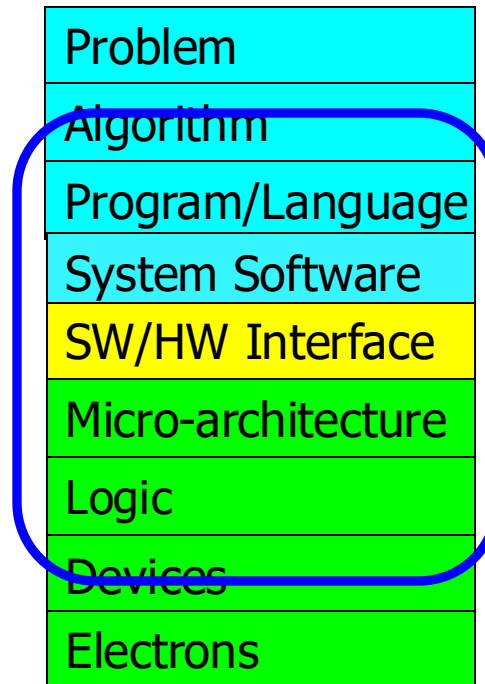
How to Enable Adoption of Processing in Memory

Potential Barriers to Adoption of PIM

1. **Applications & software** for PIM
2. Ease of **programming** (interfaces and compiler/HW support)
3. **System** and **security** support: coherence, synchronization, virtual memory, isolation, communication interfaces, ...
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

Adoption: How to Keep It Simple?

- Junwhan Ahn, Sungjoo Yoo, Onur Mutlu, and Kiyoun Choi, **"PIM-Enabled Instructions: A Low-Overhead, Locality-Aware Processing-in-Memory Architecture"** *Proceedings of the 42nd International Symposium on 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

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Adoption: How to Ease Programmability? (I)

- Geraldo F. Oliveira, Alain Kohli, David Novo, Juan Gómez-Luna, Onur Mutlu,
“DaPPA: A Data-Parallel Framework for Processing-in-Memory Architectures,”
in *PACT SRC Student Competition*, Vienna, Austria, October 2023.

DaPPA: A Data-Parallel Framework for Processing-in-Memory Architectures

Geraldo F. Oliveira*

Alain Kohli*

David Novo[‡]

Juan Gómez-Luna*

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Adoption: How to Ease Programmability? (II)

- Jinfan Chen, Juan Gómez-Luna, Izzat El Hajj, YuXin Guo, and Onur Mutlu,
"SimplePIM: A Software Framework for Productive and Efficient Processing in Memory"
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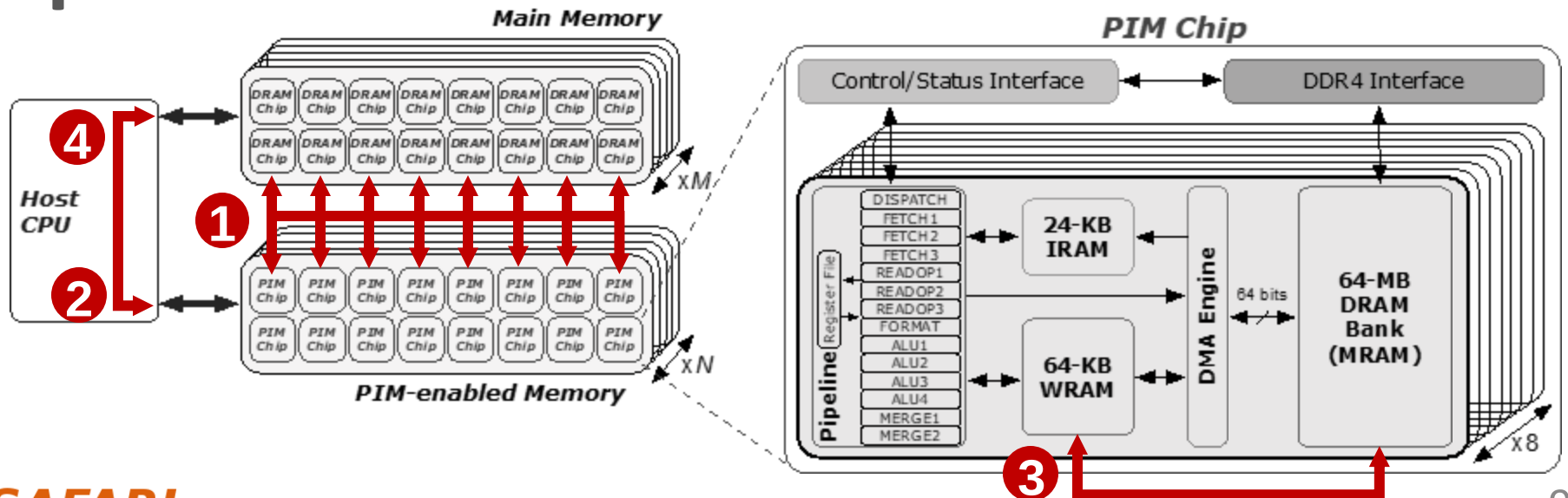
SimplePIM: A Software Framework for Productive and Efficient Processing-in-Memory

Jinfan Chen¹ Juan Gómez-Luna¹ Izzat El Hajj² Yuxin Guo¹ Onur Mutlu¹
¹ETH Zürich ²American University of Beirut

The Programmability Barrier: Overview

Programming the UPMEM-based system requires:

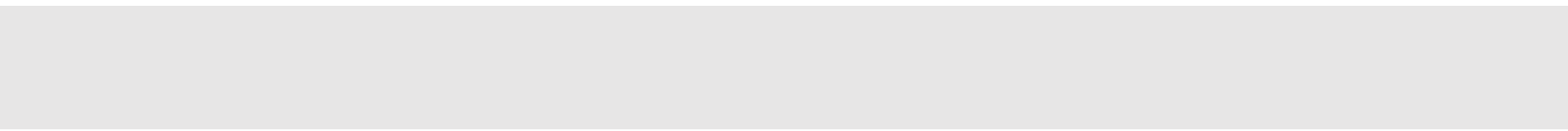
- 1 **Splitting** input data and computation across PIM chips
- 2 **Transferring input data** from main memory to PIM chips
- 3 **Manually handling caching** in PIM's scratchpad memory
- 4 **Transferring output data** from PIM chips to main memory



The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:



The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

```
const unsigned int input_size = 1073741824; // Example value
const unsigned int input_size_8bytes =
    ((input_size * sizeof(T)) % 8) != 0
    ? roundup(input_size, 8) : input_size;

const unsigned int input_size_dpu = divceil(input_size, nr_of_dpus);
const unsigned int input_size_dpu_8bytes =
    ((input_size_dpu * sizeof(T)) % 8) != 0
    ? roundup(input_size_dpu, 8) : input_size_dpu;
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

Collect
parameters

```
unsigned int kernel = 0;
dpu_arguments_t input_arguments[NR_DPUS];
for(i=0; i<nr_of_dpus-1; i++) {
    input_arguments[i].size = input_size_dpu_8bytes * sizeof(T);
    input_arguments[i].transfer_size = input_size_dpu_8bytes * sizeof(T);
    input_arguments[i].kernel = kernel;
}
input_arguments[nr_of_dpus-1].size =
    (input_size_8bytes - input_size_dpu_8bytes * (NR_DPUS-1)) * sizeof(T);
input_arguments[nr_of_dpus-1].transfer_size =
    input_size_dpu_8bytes * sizeof(T);
input_arguments[nr_of_dpus-1].kernel = kernel;
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

Collect
parameters

Distribute
parameters

```
DPU_ASSERT(dpu_push_xfer(dpu_set, DPU_XFER_TO_DPU, "DPU_INPUT_ARGUMENTS",
                        0, sizeof(input_arguments[0]), DPU_XFER_DEFAULT));
DPU_FOREACH(dpu_set, dpu, i) {
    DPU_ASSERT(dpu_prepare_xfer(dpu, bufferA + input_size_dpu_8bytes * i));
}
DPU_ASSERT(dpu_push_xfer(dpu_set, DPU_XFER_TO_DPU, DPU_MRAM_HEAP_POINTER_NAME,
                        0, input_size_dpu_8bytes * sizeof(T), DPU_XFER_DEFAULT));
DPU_FOREACH(dpu_set, dpu, i) {
    DPU_ASSERT(dpu_prepare_xfer(dpu, bufferB + input_size_dpu_8bytes * i));
}
DPU_ASSERT(dpu_push_xfer(dpu_set, DPU_XFER_TO_DPU, DPU_MRAM_HEAP_POINTER_NAME,
                        input_size_dpu_8bytes * sizeof(T),
                        input_size_dpu_8bytes * sizeof(T), DPU_XFER_DEFAULT));
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

Collect
parameters

Distribute
parameters

Launch
computation

```
DPU_ASSERT(dpu_launch(dpu_set, DPU_SYNCHRONOUS));
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

Collect
parameters

Distribute
parameters

Launch
computation

Collect
results

```
i = 0;
DPU_FOREACH(dpu_set, dpu, i) {
    DPU_ASSERT(dpu_prepare_xfer(dpu,
                                bufferC + input_size_dpu_8bytes * i));
}
DPU_ASSERT(dpu_push_xfer(dpu_set, DPU_XFER_FROM_DPU,
                          DPU_MRAM_HEAP_POINTER_NAME,
                          input_size_dpu_8bytes * sizeof(T),
                          input_size_dpu_8bytes * sizeof(T),
                          DPU_XFER_DEFAULT));
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align data	Collect parameters	Distribute parameters	Launch computation	Collect results	Manage scratchpad
---------------	-----------------------	--------------------------	-----------------------	--------------------	----------------------

```
barrier_wait(&my_barrier);
uint32_t input_size_dpu_bytes = DPU_INPUT_ARGUMENTS.size;
uint32_t input_size_dpu_bytes_transfer =
    DPU_INPUT_ARGUMENTS.transfer_size;
uint32_t base_tasklet = tasklet_id << BLOCK_SIZE_LOG2;
uint32_t mram_base_addr_A = (uint32_t)DPU_MRAM_HEAP_POINTER;
uint32_t mram_base_addr_B = (uint32_t)(DPU_MRAM_HEAP_POINTER
    + input_size_dpu_bytes_transfer);
T *cache_A = (T *) mem_alloc(BLOCK_SIZE);
T *cache_B = (T *) mem_alloc(BLOCK_SIZE);
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

Collect
parameters

Distribute
parameters

Launch
computation

Collect
results

Manage
scratchpad

Orchestrate
computation

```
for (int byte_index = base_tasklet; byte_index < input_size_dpu_bytes;
     byte_index += BLOCK_SIZE * NR_TASKLETS){
uint32_t l_size_bytes = (byte_index + BLOCK_SIZE >=
                         input_size_dpu_bytes)
                         ? (input_size_dpu_bytes - byte_index) : BLOCK_SIZE;
mram_read((__mram_ptr void const*)(mram_base_addr_A + byte_index),
           cache_A, l_size_bytes);
mram_read((__mram_ptr void const*)(mram_base_addr_B + byte_index),
           cache_B, l_size_bytes);
vector_addition(cache_B, cache_A, l_size_bytes >> DIV);
mram_write(cache_B, (__mram_ptr void*)(mram_base_addr_B + byte_index)
           l_size_bytes);
}
```

The Programmability Barrier:

Vector Addition Example

Programmer's Tasks:

Align
data

Collect
parameters

Distribute
parameters

Launch
computation

Collect
results

Manage
scratchpad

Orchestrate
computation

Goal:

Just write
my kernel

```
static void vector_addition(T *bufferB, T *bufferA, int l_size){  
    for (unsigned int i = 0; i < l_size; i++){  
        bufferB[i] += bufferA[i];  
    }  
}
```

The Programmability Barrier: Summary

Programmer's Tasks:

Align
data

Collect
parameters

Distribute
parameters

Launch
computation

Collect
results

Manage
scratchpad

Orchestrate
computation

Goal:

Just write
my kernel

Problem

Programming the UPMEM system
leads to non-trivial effort → requires
knowledge of the underlying hardware and
manual fine-grained data movement handling

Our Goal

Goal

To ease programmability for the UPMEM system,
allowing a programmer to write
efficient PIM-friendly code
without the need to
explicitly manage hardware resources

Outline

1

Introduction

2

The Programmability Barrier

3

SimplePIM Overview

Management, Communication & Processing Interfaces

Evaluation Results

4

DaPPA Overview

DaPPA Main Components

Evaluation Results

5

Conclusion

SimplePIM:

A Software Framework for Productive and Efficient Processing in Memory

- Jinfan Chen, Juan Gómez-Luna, Izzat El Hajj, Yuxin Guo, and Onur Mutlu,
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[[Slides \(pptx\)](#)] [[pdf](#)]
[[SimplePIM Source Code](#)]

SimplePIM: A Software Framework for Productive and Efficient Processing-in-Memory

Jinfan Chen¹ Juan Gómez-Luna¹ Izzat El Hajj² Yuxin Guo¹ Onur Mutlu¹
¹ETH Zürich ²American University of Beirut

SimplePIM Programming Framework: Overview

SimplePIM provides **standard abstractions** to **build** and **deploy** applications on PIM systems

1 **Management interface**
→ Metadata for PIM-resident arrays

2 **Communication interface**
→ Abstractions for host-PIM and PIM-PIM communication

3 **Processing interface**
→ Iterators (`map`, `reduce`, `zip`) to implement workloads

SimplePIM Programming Framework: Management Interface

- **Metadata for PIM-resident arrays**
 - `array_meta_data_t` describes a PIM-resident array
 - `simple_pim_management_t` for managing PIM-resident arrays

- **lookup:** Retrieves all relevant information of an array

```
array_meta_data_t* simple_pim_array_lookup(const char* id,  
simple_pim_management_t* management);
```

- **register:** Registers the metadata of an array

```
void simple_pim_array_register(array_meta_data_t* meta_data,  
simple_pim_management_t* management);
```

- **free:** Removes the metadata of an array

```
void simple_pim_array_free(const char* id, simple_pim_management_t* management);
```

SimplePIM Programming Framework: Communication Interface (I)

- **SimplePIM Host-to-CPU Broadcast**

- Transfers a host array to all PIM cores in the system

```
void simple_pim_array_broadcast(char* const id, void* arr, uint64_t len,  
uint32_t type_size, simple_pim_management_t* management);
```



SimplePIM Programming Framework: Communication Interface (II)

- **Host-to-PIM SimplePIM Scatter**

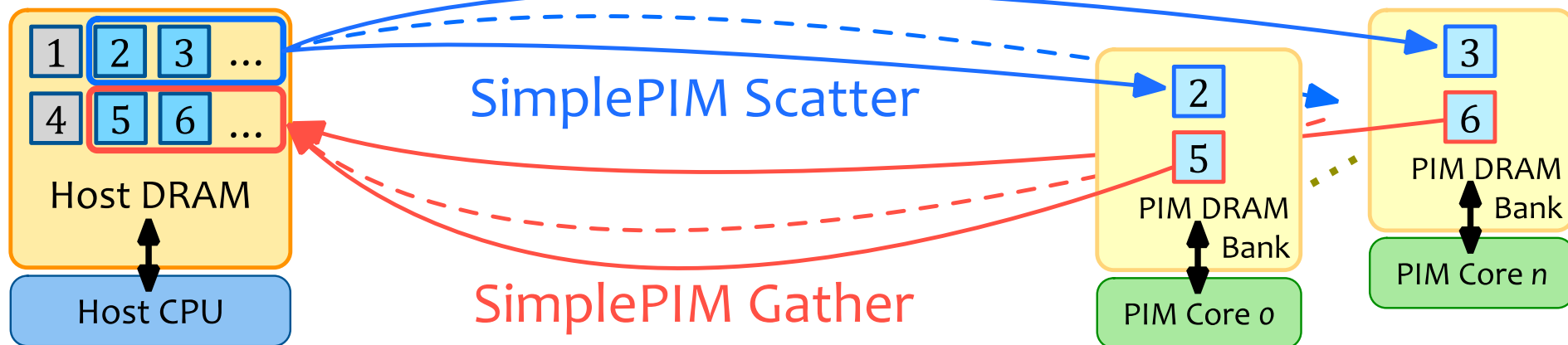
- Distributes an array to PIM DRAM banks

```
void simple_pim_array_scatter(char* const id, void* arr, uint64_t len,  
uint32_t type_size, simple_pim_management_t* management);
```

- **Host-to-PIM SimplePIM Gather**

- Collects portions of an array from PIM DRAM banks

```
void* simple_pim_array_gather(char* const id, simple_pim_management_t*  
management);
```



SimplePIM Programming Framework: Communication Interface (III)

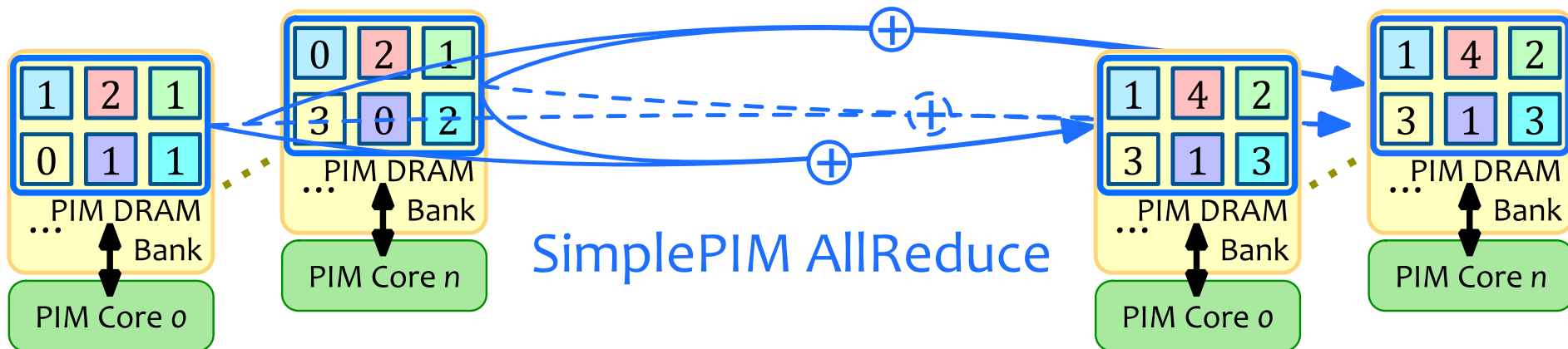
• PIM-to-PIM Communication: AllReduce

- Used for algorithm synchronization
- The programmer specifies an accumulative function

```
void simple_pim_array_allreduce(char* const id, handle_t* handle,  
simple_pim_management_t* management);
```

Before PIM-PIM communication

After PIM-PIM communication

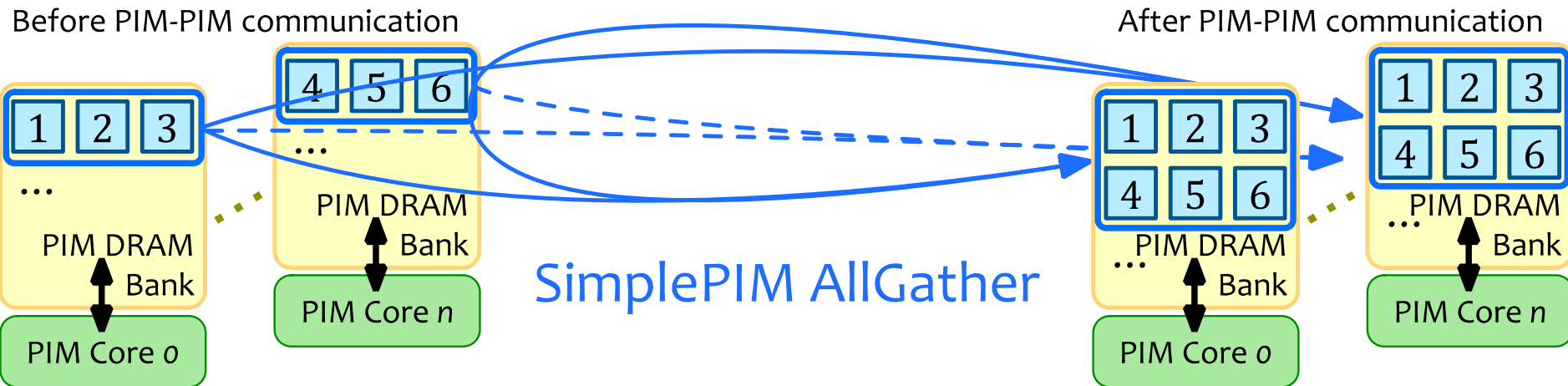


SimplePIM Programming Framework: Communication Interface (IV)

- **PIM-to-PIM Communication: AllGather**

- Combines array pieces and distributes the complete array to all PIM cores

```
void simple_pim_array_allgather(char* const id, char* new_id,  
simple_pim_management_t* management);
```

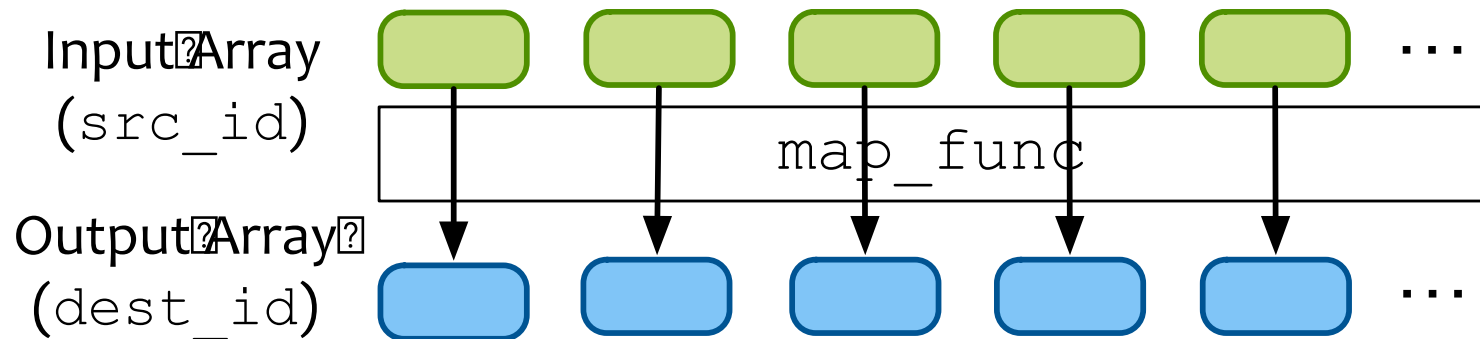


SimplePIM Programming Framework: Processing Interface (I)

- **Array Map**

- Applies `map_func` to every element of the data array

```
void simple_pim_array_map(const char* src_id, const char* dest_id,  
uint32_t output_type, handle_t* handle, simple_pim_management_t* management);
```

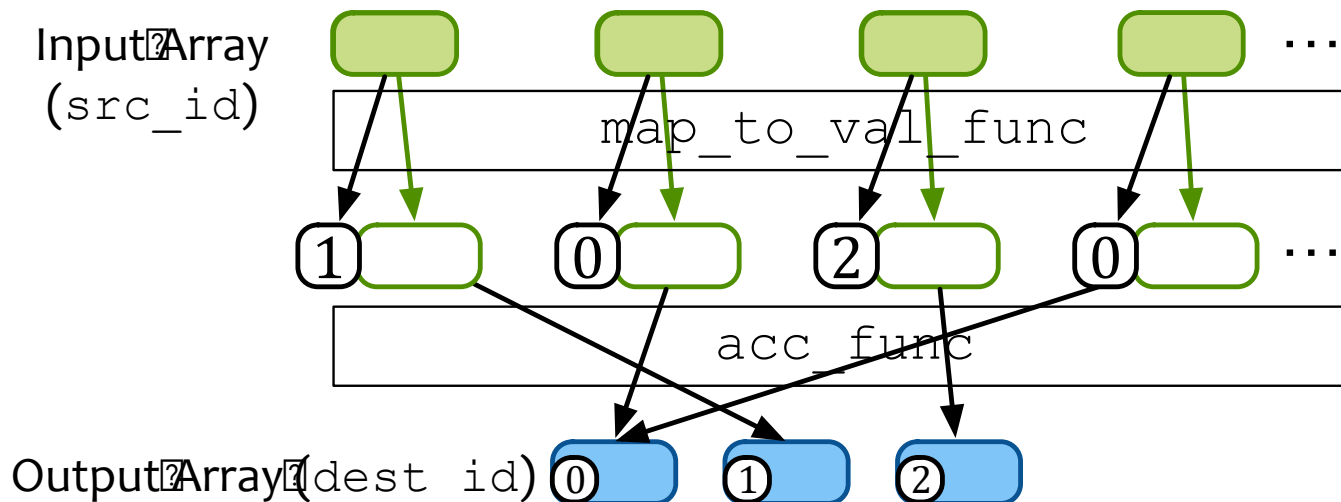


SimplePIM Programming Framework: Processing Interface (II)

• Array Reduction

- The `map_to_val_func` function transforms an input element to an output value and an output index
- The `acc_func` function accumulates the output values onto the output array

```
void simple_pim_array_red(const char* src_id, const char* dest_id,  
uint32_t output_type, uint32_t output_len, handle_t* handle,  
simple_pim_management_t* management);
```

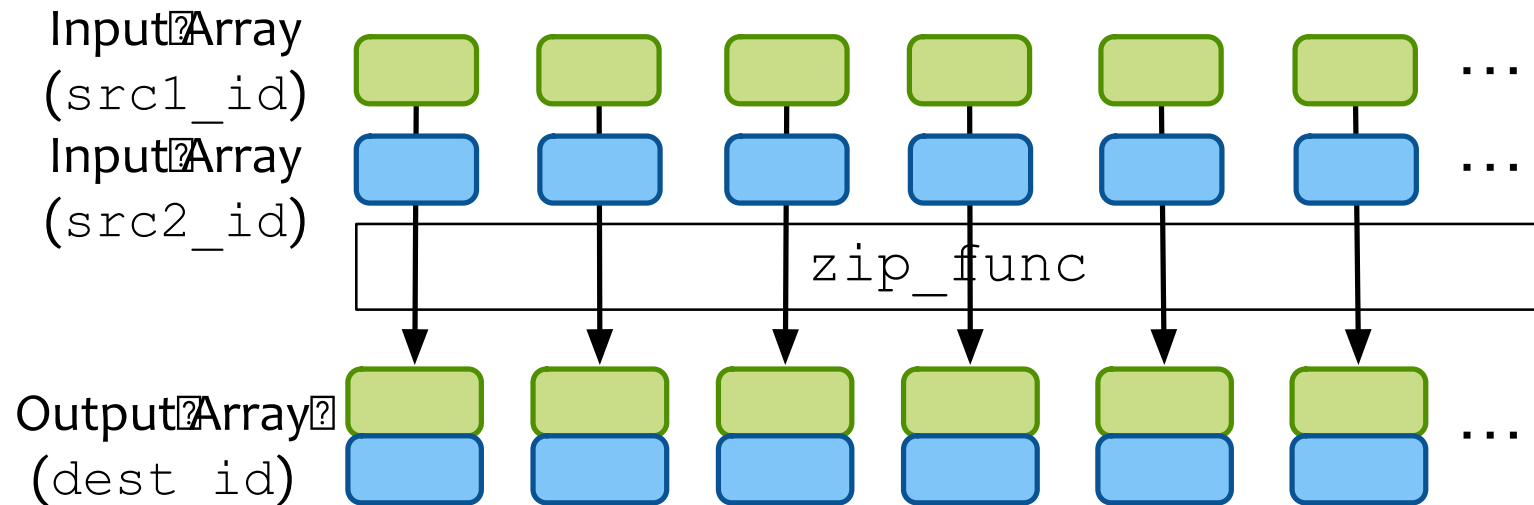


SimplePIM Programming Framework: Processing Interface (III)

- **Array Zip**

- Takes two input arrays and combines their elements into an output array

```
void simple_pim_array_zip(const char* src1_id, const char* src2_id,  
const char* dest_id, simple_pim_management_t* management);
```



SimplePIM Programming Framework:

General Code Optimizations

- Strength reduction
- Loop unrolling
- Avoiding boundary checks
- Function inlining
- Adjustment of data transfer sizes

Evaluation Results:

Evaluation Methodology

- **Evaluated system**
 - UPMEM PIM system with 2,432 PIM cores with 159 GB of PIM DRAM
- **Real-world Benchmarks**
 - Vector addition
 - Reduction
 - Histogram
 - K-Means
 - Linear regression
 - Logistic regression
- Comparison to hand-optimized codes in terms of programming productivity and performance

Evaluation Results:

Productive Improvement (I)

- Example: Hand-optimized histogram with UPMEM SDK

```
... // Initialize global variables and functions for histogram
int main_kernel() {
    if (tasklet_id == 0)
        mem_reset(); // Reset the heap
    ... // Initialize variables and the histogram
    T *input_buff_A = (T*)mem_alloc(2048); // Allocate buffer in scratchpad memory

    for (unsigned int byte_index = base_tasklet; byte_index < input_size; byte_index += stride) {
        // Boundary checking
        uint32_t l_size_bytes = (byte_index + 2048 >= input_size) ? (input_size - byte_index) : 2048;
        // Load scratchpad with a DRAM block
        mram_read((const __mram_ptr void*) (mram_base_addr_A + byte_index), input_buff_A, l_size_bytes);
        // Histogram calculation
        histogram(hist, bins, input_buff_A, l_size_bytes/sizeof(uint32_t));
    }
    ...
    barrier_wait(&my_barrier); // Barrier to synchronize PIM threads
    ... // Merging histograms from different tasklets into one histo_dpu
    // Write result from scratchpad to DRAM
    if (tasklet_id == 0)
        if (bins * sizeof(uint32_t) <= 2048)
            mram_write(histo_dpu, (__mram_ptr void*)mram_base_addr_histo, bins * sizeof(uint32_t));
        else
            for (unsigned int offset = 0; offset < ((bins * sizeof(uint32_t)) >> 11); offset++) {
                mram_write(histo_dpu + (offset << 9), (__mram_ptr void*) (mram_base_addr_histo +
                    (offset << 11)), 2048);
            }
    return 0;
}
```

Evaluation Results:

Productive Improvement (II)

- Example: SimplePIM histogram

```
// Programmer-defined functions in the file "histo_filepath"
void init_func (uint32_t size, void* ptr) {
    char* casted_value_ptr = (char*) ptr;
    for (int i = 0; i < size; i++)
        casted_value_ptr[i] = 0;
}

void acc_func (void* dest, void* src) {
    *(uint32_t*)dest += *(uint32_t*)src;
}

void map_to_val_func (void* input, void* output, uint32_t* key) {
    uint32_t d = *(uint32_t*)input;
    *(uint32_t*)output = 1;
    *key = d * bins >> 12;
}

// Host side handle creation and iterator call
handle_t* handle = simple_pim_create_handle("histo_filepath", REDUCE, NULL, 0);

// Transfer (scatter) data to PIM, register as "t1"
simple_pim_array_scatter("t1", src, bins, sizeof(T), management);

// Run histogram on "t1" and produce "t2"
simple_pim_array_red("t1", "t2", sizeof(T), bins, handle, management);
```

Evaluation Results:

Productive Improvement (III)

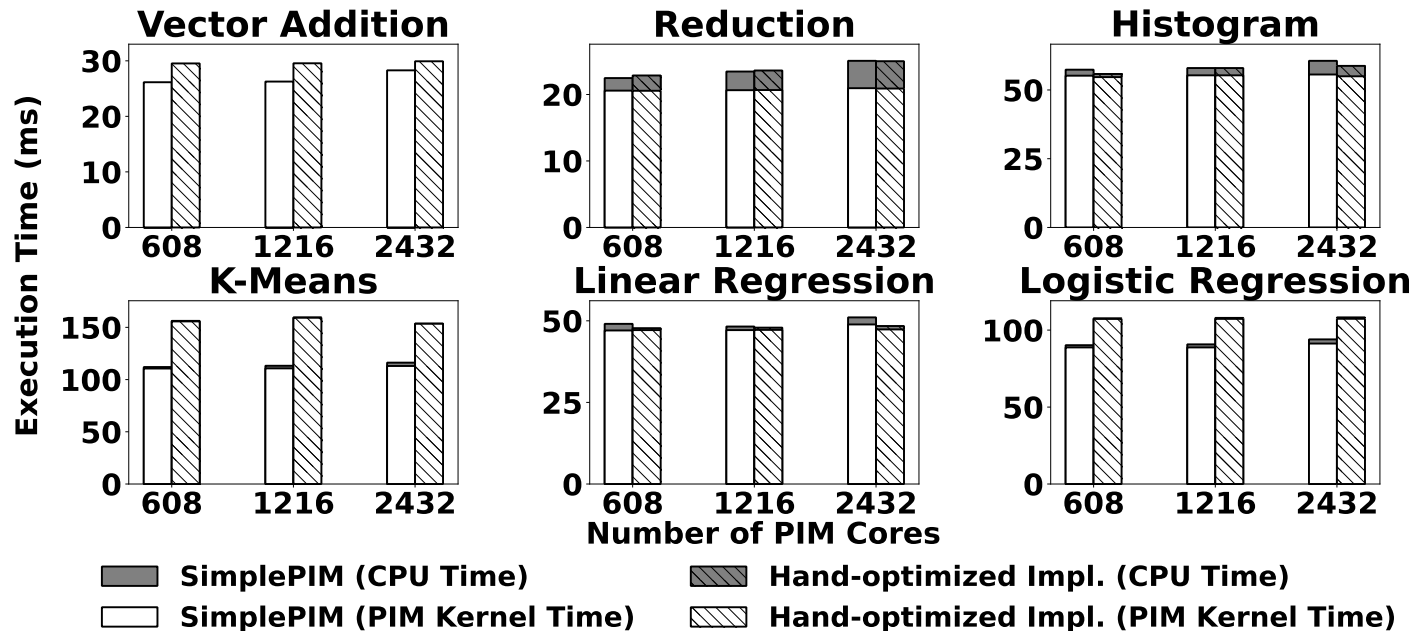
- Lines of code (LoC) reduction

	SimplePIM	Hand-optimized	LoC Reduction
Reduction	14	83	5.93×
Vector Addition	14	82	5.86×
Histogram	21	114	5.43×
Linear Regression	48	157	3.27×
Logistic Regression	59	176	2.98×
K-Means	68	206	3.03×

SimplePIM **reduces** the number of lines of effective code by a factor of **2.98×** to **5.93×**

Evaluation Results:

Weak Scaling Analysis

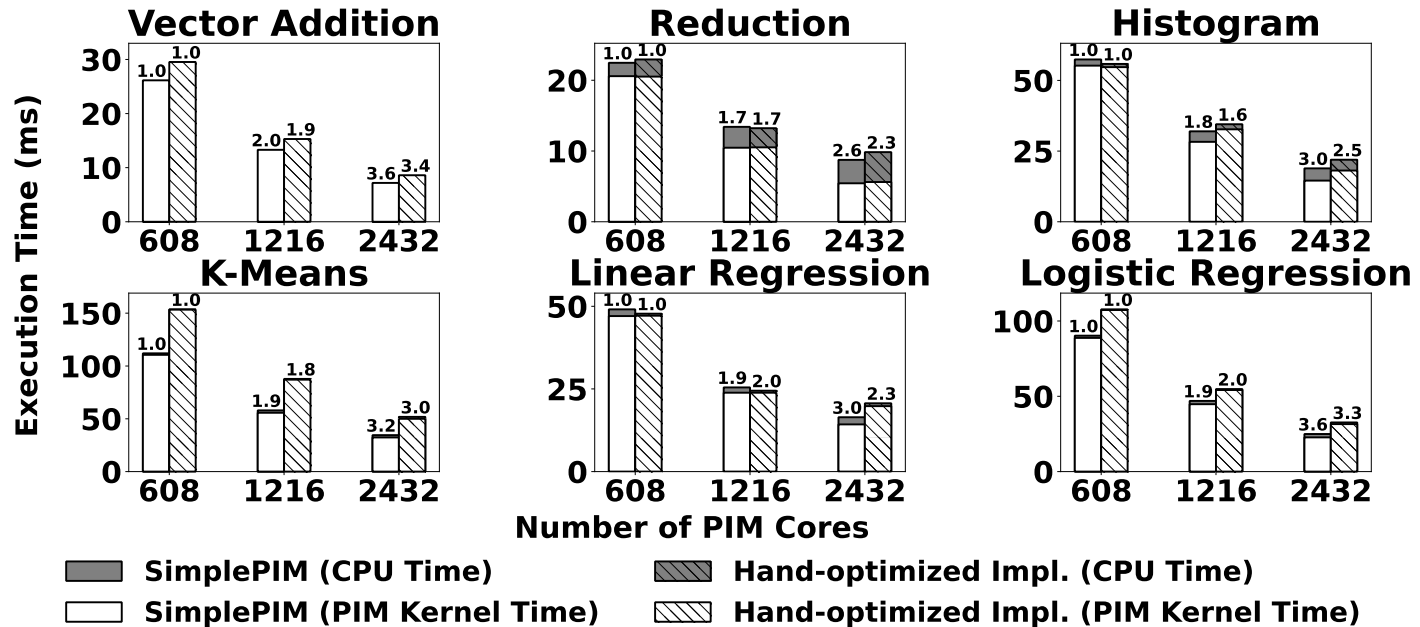


SimplePIM achieves **comparable performance** for reduction, histogram, and linear regression

SimplePIM **outperforms hand-optimized implementations** for vector addition, logistic regression, and k-means by **10%-37%**

Evaluation Results:

Strong Scaling Analysis



SimplePIM scales better than hand-optimized implementations for reduction, histogram, and linear regression

SimplePIM outperforms hand-optimized implementations for vector addition, logistic regression, and k-means by 15%-43%

Source Code

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

The screenshot shows the GitHub repository page for CMU-SAFARI/SimplePIM. The repository is public and has 13 commits. The file list includes benchmarks, lib, .gitignore, LICENSE, and README.md. The README is selected, showing the title "SimplePIM: A Software Framework for Productive and Efficient In-Memory Processing" and a description of the project. The right sidebar contains links to the README, MIT license, activity, custom properties, stars (18), watchers (6), forks (4), releases, packages, and contributors (3).

CMU-SAFARI / SimplePIM Public

Notifications F

<> Code Issues Pull requests Actions Projects Security Insights

main Branches Tags Code

13 Commits

- benchmarks
- lib
- .gitignore
- LICENSE
- README.md

README MIT license

SimplePIM: A Software Framework for Productive and Efficient In-Memory Processing

This project implements SimplePIM, a software framework for easy and efficient in-memory-hardware programming. The code is implemented on UPMEM, an actual, commercially available PIM hardware that combines traditional DRAM memory with general-purpose in-order cores inside the same chip. SimplePIM processes arrays of arbitrary elements on a PIM device by calling iterator functions from the host and provides primitives for communication among PIM cores and between PIM and the host system.

We implement six applications with SimplePIM on UPMEM:

Readme MIT license Activity Custom properties 18 stars 6 watching 4 forks Report repository Releases No releases published Packages No packages published Contributors 3

SimplePIM:

A Software Framework for Productive and Efficient Processing in Memory

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[\[SimplePIM Source Code\]](#)

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DaPPA:

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- Geraldo F. Oliveira, Alain Kohli, David Novo, Juan Gómez-Luna, Onur Mutlu

["DaPPA: A Data-Parallel Framework for Processing-in-Memory Architectures,"](#)
[arXiv:2310.10168 \[cs.AR\]](#)

2nd Place ACM Student Research Competition at the [32nd International Conference on Parallel Architectures and Compilation Techniques \(PACT\)](#),
Vienna, Austria, October 2023.

DaPPA: A Data-Parallel Framework for Processing-in-Memory Architectures

Geraldo F. Oliveira^{*}

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1. Motivation & Problem

The increasing prevalence and growing size of data in modern applications have led to high costs for computation in traditional *processor-centric computing* systems. To mitigate these costs, the *processing-in-memory* (PIM) [1–6] paradigm moves computation closer to where the data resides, reducing the need to move data between memory and the processor. Even though the concept of PIM has been first proposed in the 1960s [7, 8], real-world PIM systems have only recently

face [15, 16] that abstracts the hardware components of the UPMEM system. Using this key idea, DaPPA transforms a data-parallel pattern-based application code into the appropriate UPMEM-target code, including the required APIs for data management and code partition, which can then be compiled into a UPMEM-based binary *transparently* from the programmer. While generating UPMEM-target code, DaPPA implements several code optimizations to improve end-to-end performance.

DaPPA:

Key Idea & Overview

Key Idea

Leverage an intuitive
data-parallel pattern-based interface for
PIM programming



DaPPA , a Data-Parallel PIM Architecture that
automatically distributes input and gathers output data, handles
memory management, and
parallelizes work across PIM cores

DaPPA is composed of three main components:

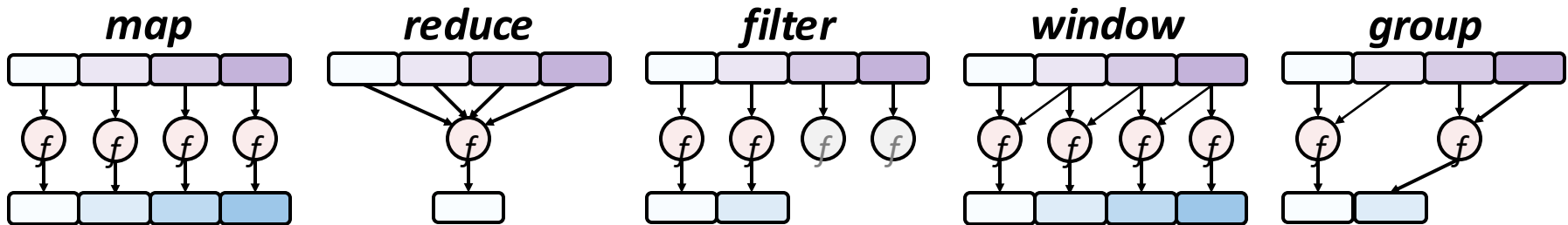
- 1** Data-Parallel Pattern APIs
- 2** Dataflow Programming Interface
- 3** Dynamic Template-Based Compilation

DaPPA:

Data-Parallel Pattern APIs

Pre-defined functions that implement high-level
data-parallel pattern primitives

- Skeleton and pattern-based parallel programming are a **common abstraction for parallel architectures**
 - M. Cole, “*Bringing Skeletons Out of the Closet: A Pragmatic Manifesto for Skeletal Parallel Programming*,” Parallel Computing, 2004
- DaPPA supports **five primary** data-parallel patterns

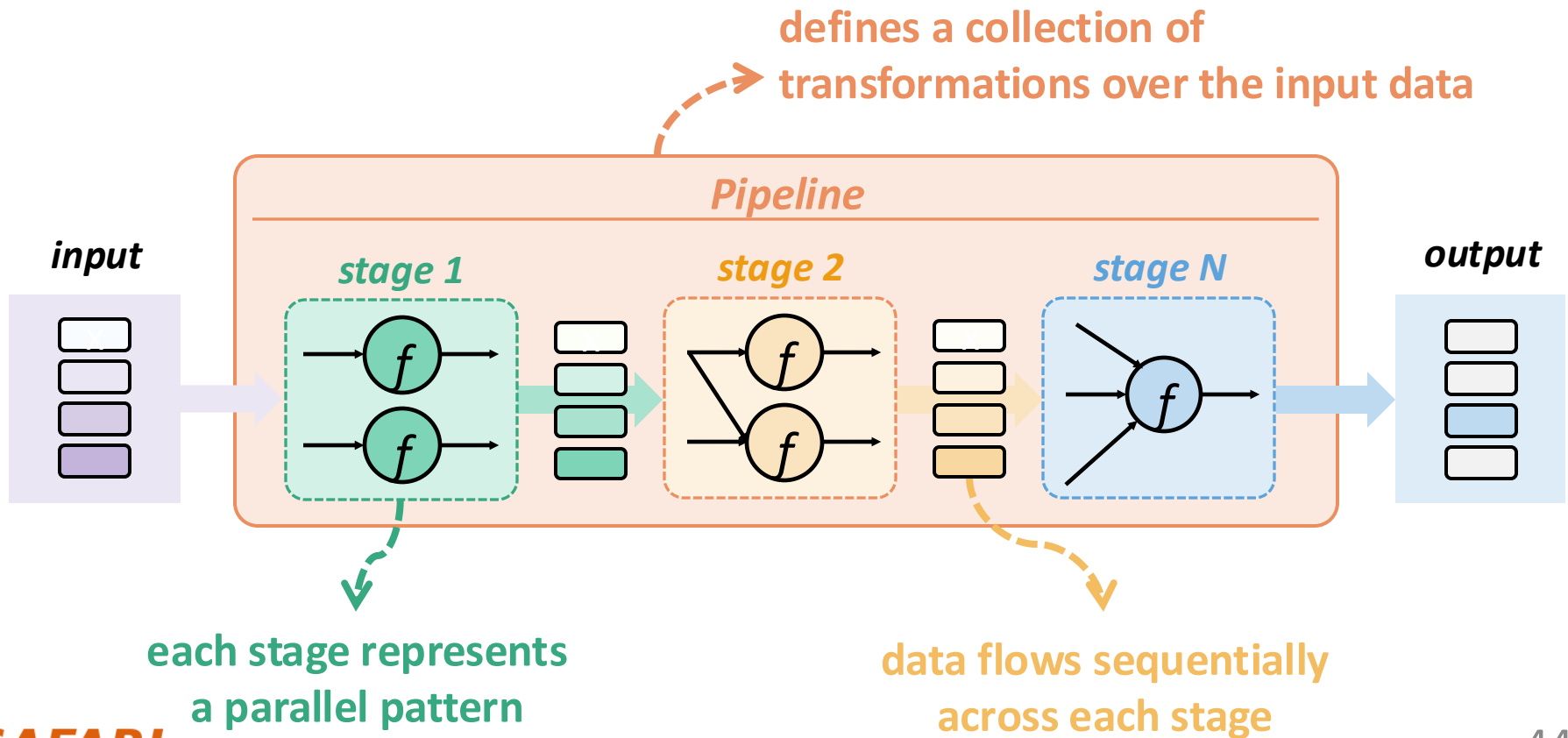


The user can combine all five data-parallel primitives
to describe complex data transformations

DaPPA:

Dataflow Programming Interface

DaPPA exposes to the user a
dataflow-based programming interface



DaPPA:

Dynamic Template-Based Compilation

DaPPA uses a dynamic template-based compilation to generate PIM code **in two main steps**

1 Templating: DaPPA creates a base UPMEM code based on a *basic skeleton* of a UPMEM application

- We use the Inja C++ templating engine

2 Optimizations: DaPPA uses a series of transformations to

- *extract* data required by the UPMEM code template
- calculate the *memory offsets* for MRAMs and WRAMs
- *divide computation* between CPU and PIM cores

DaPPA compiles and executes each stage in a Pipeline per time →
allows for runtime optimizations

DaPPA:

Putting All Together

Example of DaPPA's implementation of a
vector dot product operation

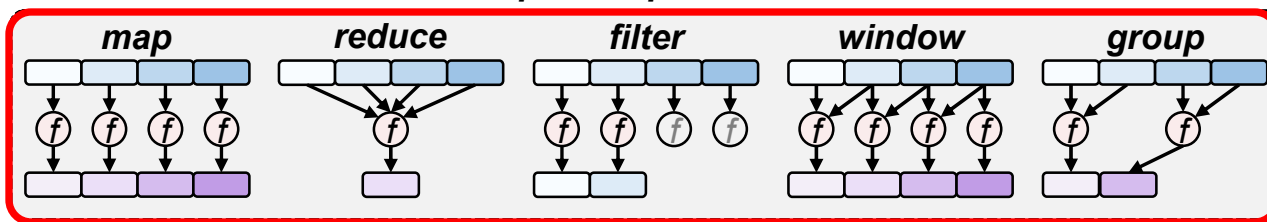
reduce
 $C = A_0B_0 +$
map $A_1B_1 +$
 $A_2B_2 +$
 A_3B_3
target
computation

DaPPA:

Putting All Together

Example of DaPPA's implementation of a
vector dot product operation

❶ *data-parallel pattern APIs*



reduce

$$C = A_0B_0 +$$

map

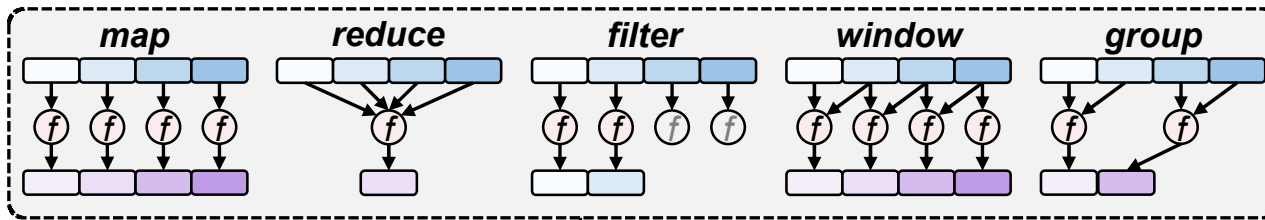
$$A_1B_1 +$$
$$A_2B_2 +$$
$$A_3B_3$$

*target
computation*

DaPPA: Putting All Together

Example of DaPPA's implementation of a
vector dot product operation

① data-parallel pattern APIs



reduce
 $C = A_0B_0 +$
map $A_1B_1 +$
 $A_2B_2 +$
 A_3B_3
target
computation

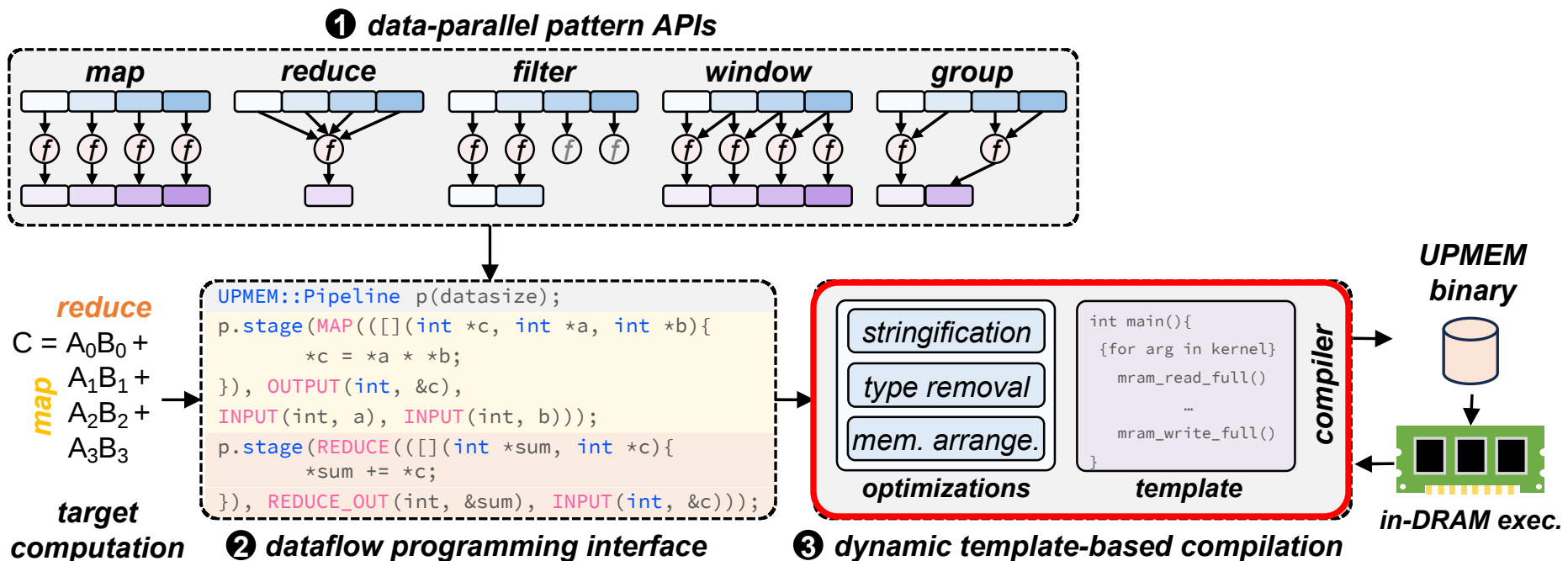
```
UPMEM::Pipeline p(datasize);  
p.stage(MAP)(([])(int *c, int *a, int *b){  
    *c = *a * *b;  
}), OUTPUT(int, &c),  
INPUT(int, a), INPUT(int, b));  
p.stage(REDUCE)(([])(int *sum, int *c){  
    *sum += *c;  
}), REDUCE_OUT(int, &sum), INPUT(int, &c));
```

② dataflow programming interface

DaPPA:

Putting All Together

Example of DaPPA's implementation of a
vector dot product operation



Evaluation:

Methodology Overview

- **Evaluation Setup**

- **Host CPU:** 2-socket Intel® Xeon Silver 4110 CPU
- **PIM Cores:** 20 UPMEM PIM DIMMs (160 GB PIM memory)
- **2560 DPUs** in total

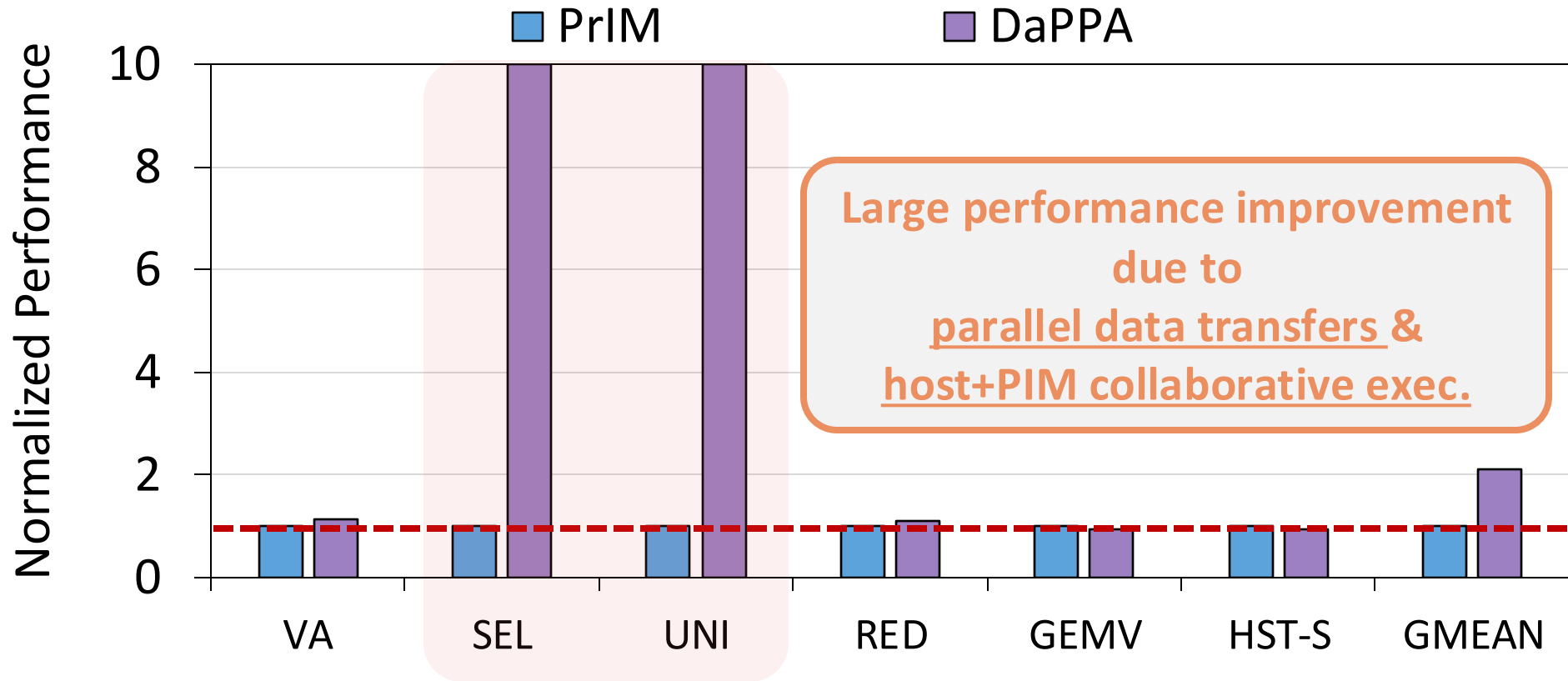
- **Workloads:** 6 workloads from the PRIM benchmark suite

- Vector addition (**VA**); Select (**SEL**); Unique (**UNI**); Reduce (**RED**); General matrix-vector multiply (**GEMV**); Histogram small (**HST-S**)

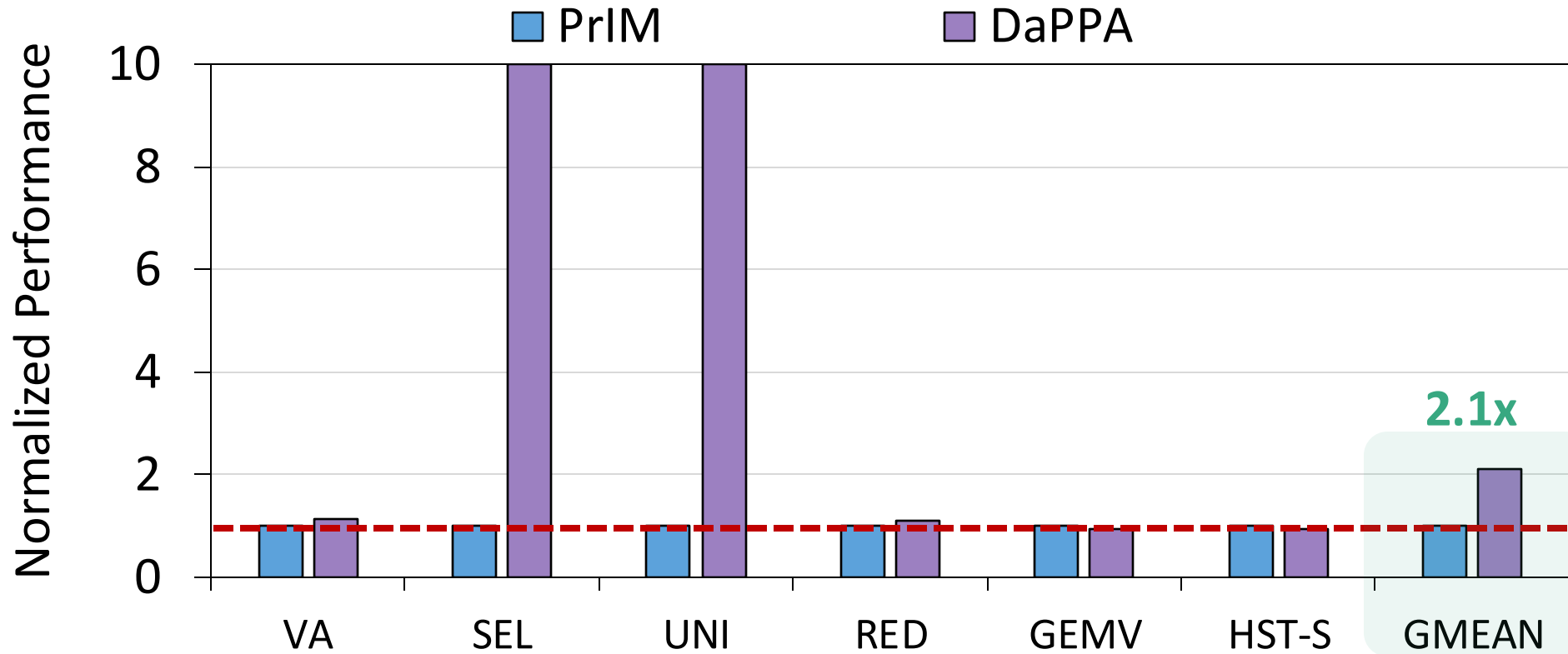
- **Metrics**

- **End-to-end execution time** (average of 10 runs)
- **Programming complexity** (in lines of code)

Evaluation: Performance Analysis



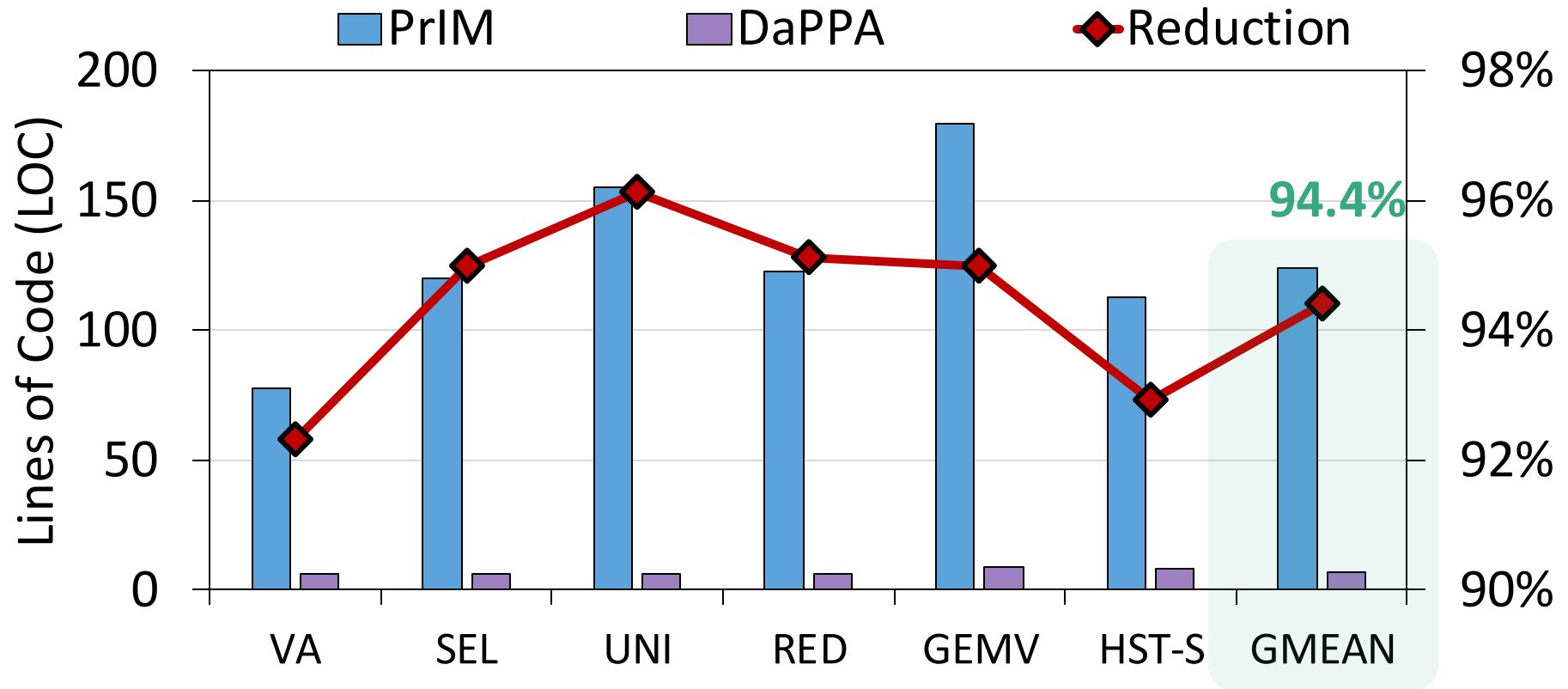
Evaluation: Performance Analysis



DaPPA significantly improves end-to-end performance
compared to hand-tuned implementations

Evaluation:

Programming Complexity Analysis



DaPPA significantly reduces programming complexity by abstracting hardware components

Evaluation:

Comparison to State-of-the-Art

SimplePIM [Chen+, PACT'23]: a framework that uses
(1) iterator functions and (2) primitives for communication
to aid PIM programmability

Compared to SimplePIM, DaPPA provides three key benefits

1. **Higher abstraction level** → The programmer does *not* need to manually specify communication patterns used during computation
2. **Support for more parallel patterns** → DaPPA supports two more parallel primitives (window and group), and allows the mixing of parallel patterns
3. **Further execution optimizations** → DaPPA allows using idle host resources for collaborative execution

**DaPPA improves state-of-the-art frameworks for
PIM programmability**

DaPPA:

A Data-Parallel Framework for Processing-in-Memory Architectures

- Geraldo F. Oliveira, Alain Kohli, David Novo, Juan Gómez-Luna, Onur Mutlu

["DaPPA: A Data-Parallel Framework for Processing-in-Memory Architectures,"](#)
[arXiv:2310.10168 \[cs.AR\]](#)

2nd Place ACM Student Research Competition at the [32nd International Conference on Parallel Architectures and Compilation Techniques \(PACT\)](#), Vienna, Austria, October 2023.

DaPPA: A Data-Parallel Framework for Processing-in-Memory Architectures

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1. Motivation & Problem

The increasing prevalence and growing size of data in modern applications have led to high costs for computation in traditional *processor-centric computing* systems. To mitigate these costs, the *processing-in-memory* (PIM) [1–6] paradigm moves computation closer to where the data resides, reducing the need to move data between memory and the processor. Even though the concept of PIM has been first proposed in the 1960s [7, 8], real-world PIM systems have only recently

face [15, 16] that abstracts the hardware components of the UPMEM system. Using this key idea, DaPPA transforms a data-parallel pattern-based application code into the appropriate UPMEM-target code, including the required APIs for data management and code partition, which can then be compiled into a UPMEM-based binary *transparently* from the programmer. While generating UPMEM-target code, DaPPA implements several code optimizations to improve end-to-end performance.

Tutorial on Memory-Centric Computing: Programming PNM Systems

Geraldo F. Oliveira

<https://geraldofojunior.github.io>

PPoPP 2025

1st March 2025

Agenda

- Processing-Near-Memory Systems: Developments from Academia & Industry
- Programming Processing-Near-Memory Systems
- Coffee Break
- Processing-Using-Memory Systems for Bulk Bitwise Operations
- **Ataberk Olgun:**
Infrastructure for Processing-Using-Memory Research
- **Invited Talk by Prof. John Kim:**
Is it Memory-Centric or Communication-Centric?