

DORA: Dataflow-Instruction Orchestration Architecture for DNN Acceleration

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<https://peipeizhou-eecs.github.io/>

<https://github.com/arc-research-lab/DORA.git>



U.S. DEPARTMENT
of ENERGY

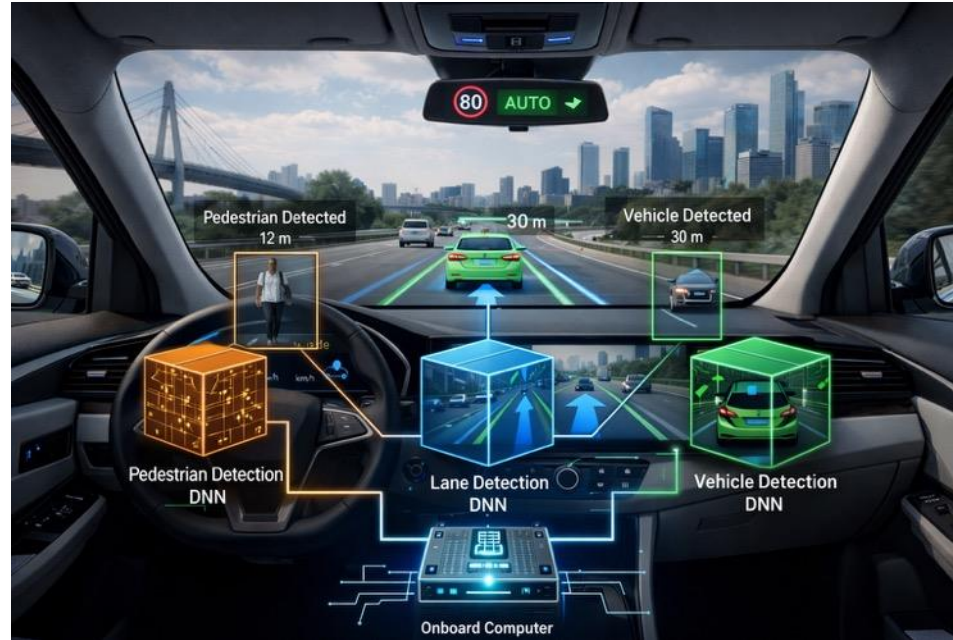


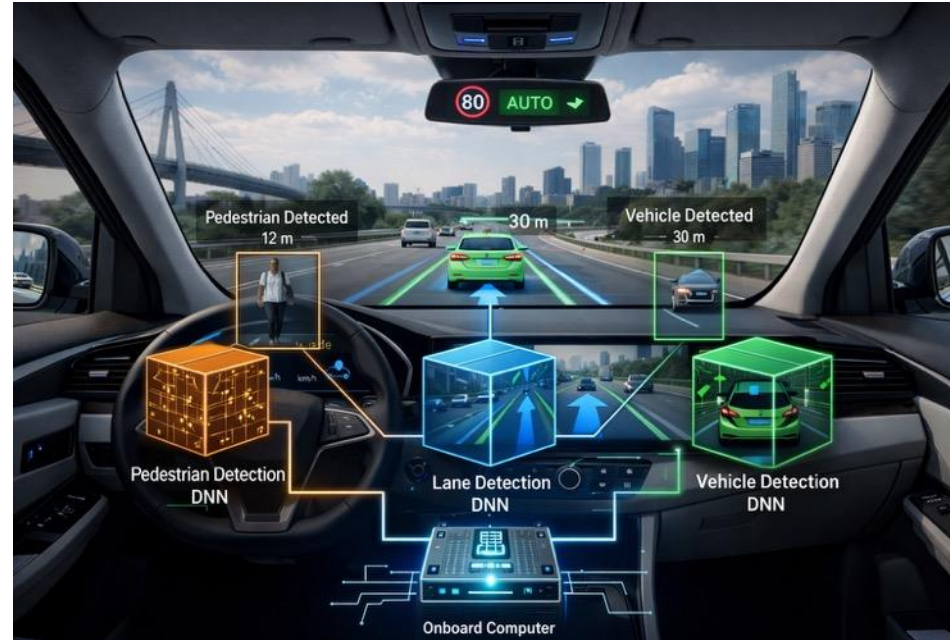
BROWN



WAYNE STATE
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DELAWARE





Tasks behind modern autonomous driving systems.



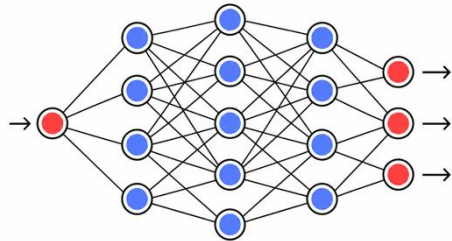
Challenge 1

- Autonomous driving system requires **diverse** DNN models.

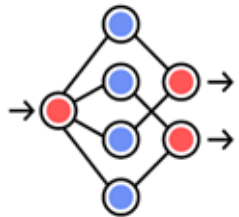
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ADS applications



Large DNNs

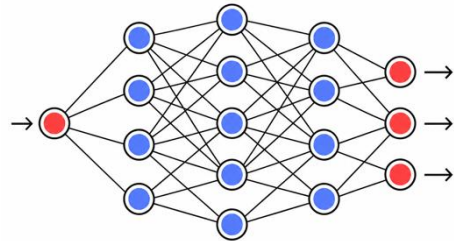


Small DNNs

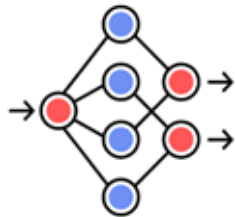
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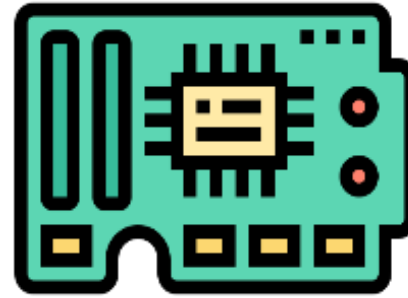
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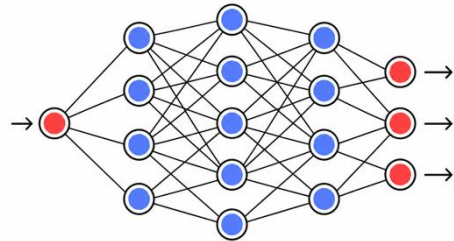
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- Good for large DNNs;
- Bad for small DNNs.

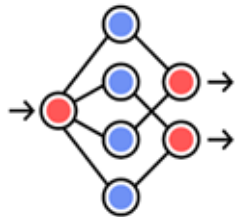
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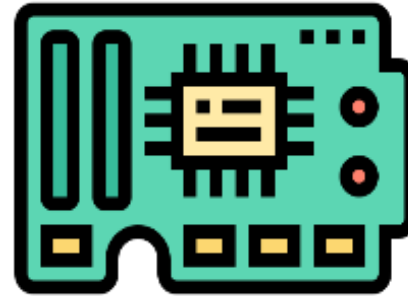
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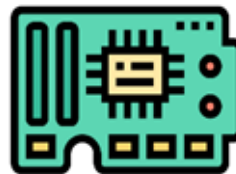


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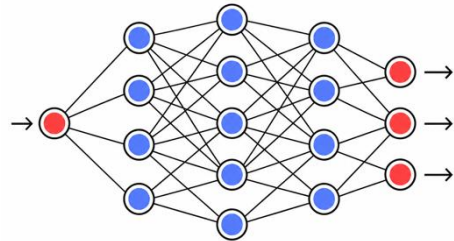
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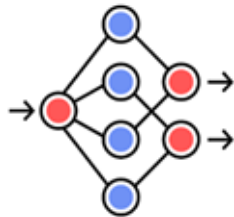
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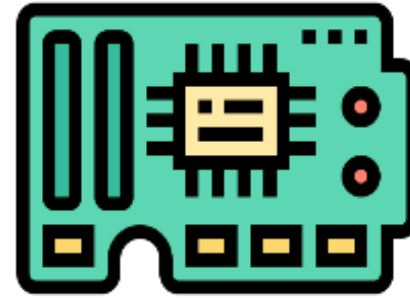
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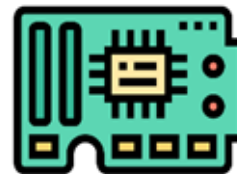


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Switching bitstream requires ~20ms.



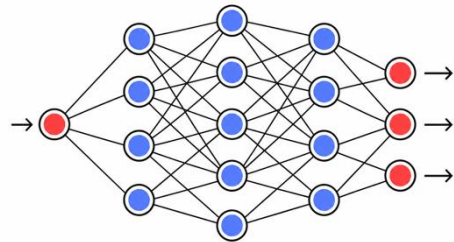
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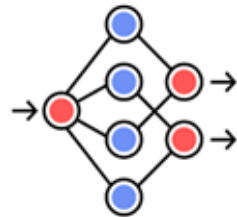
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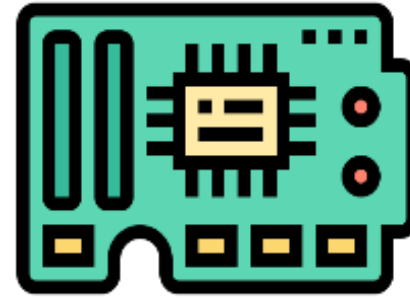
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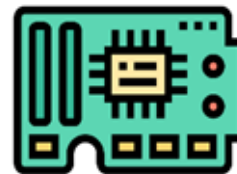


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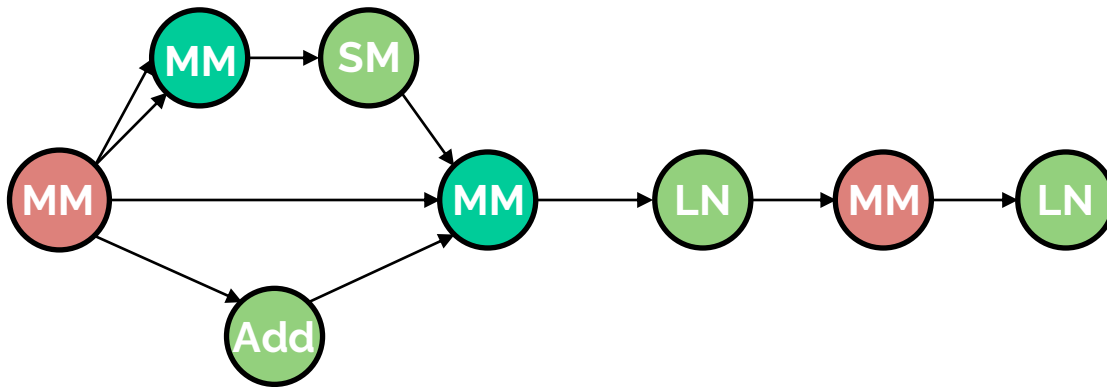
- How to **efficiently** handle diverse DNN models in **one bitstream**?

Challenge 2

- Complicated applications lead to a **huge design space**.

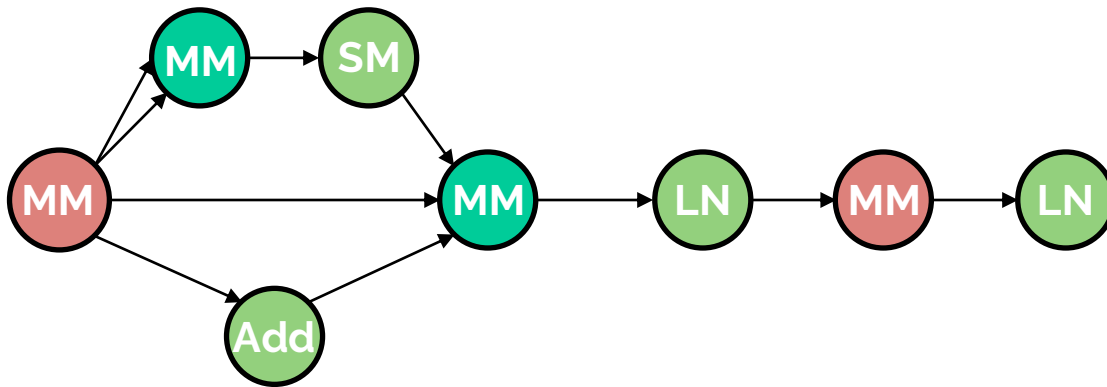
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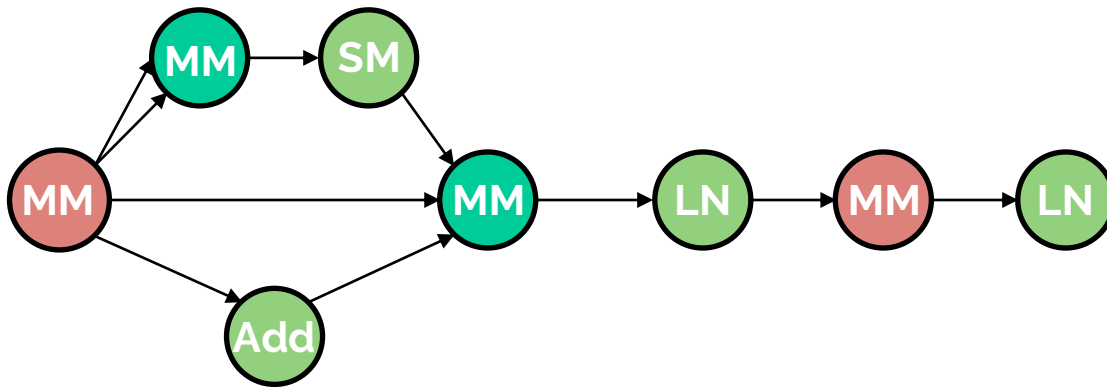
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950x The U.S. GDP in 2025.

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 - Brute-force search induces redundant search time.
 - Heuristic search loses many opportunities for better mapping solutions.
- How to maintain the **optimality** while preserving the **efficiency**?

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Challenges

- **Diverse operations exist in modern DNN models.**
- **Complicated applications lead to a huge design space.**
- **Balance between optimality and scalability.**

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- Diverse operations exist in modern DNN models.
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Solutions

- **DORA proposes an overlay architecture as well as an ISA abstraction for diverse workloads.**
- **DORA is equipped with an automatic framework that can analyze the mapping solutions at a fine-grained layer level.**
- **DORA has a two-stage design space exploration that combines brute-force and heuristic search.**

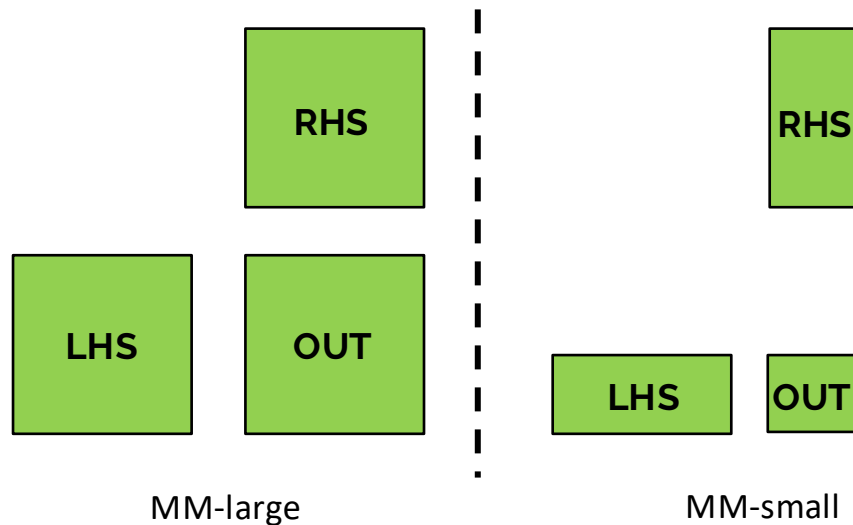
DORA Architecture Design

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- Enable flexible **computation tile size** by **parallelism management**.

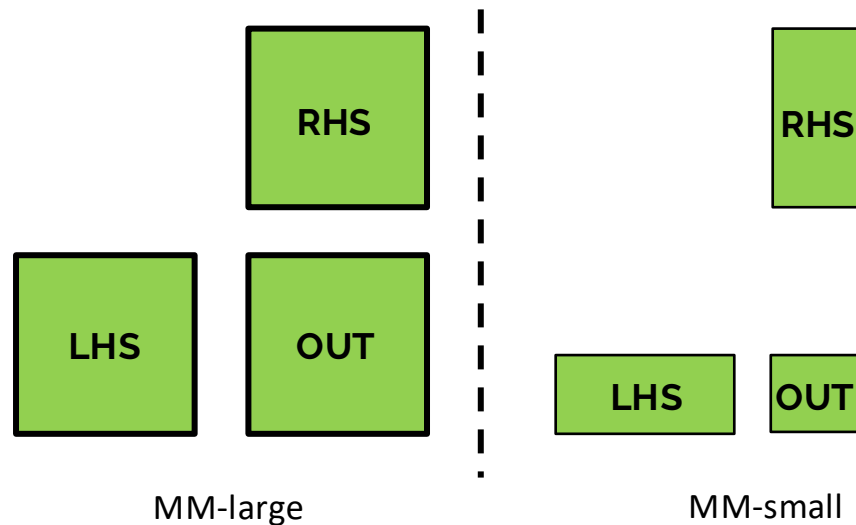
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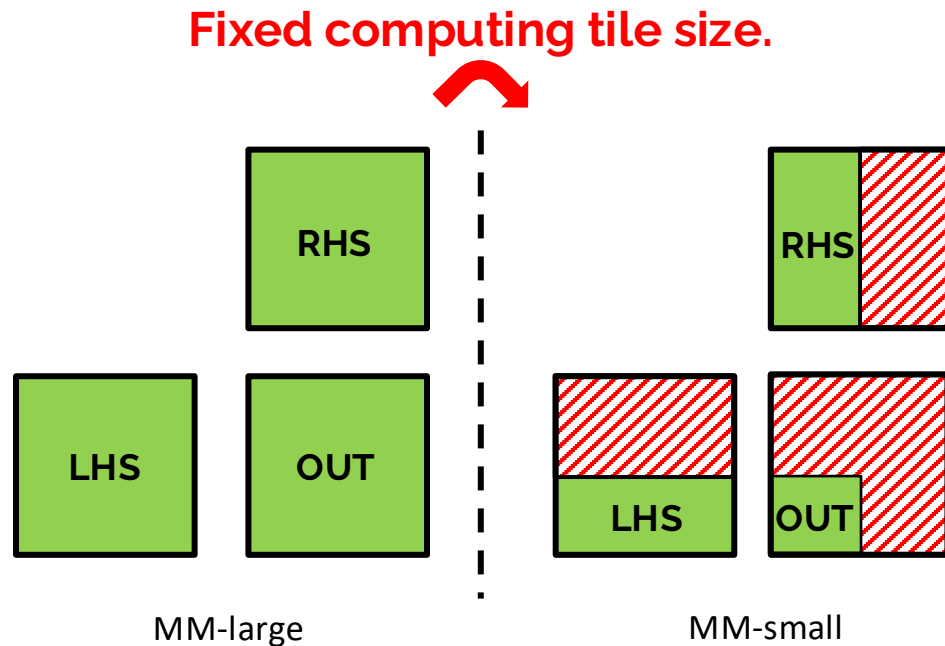
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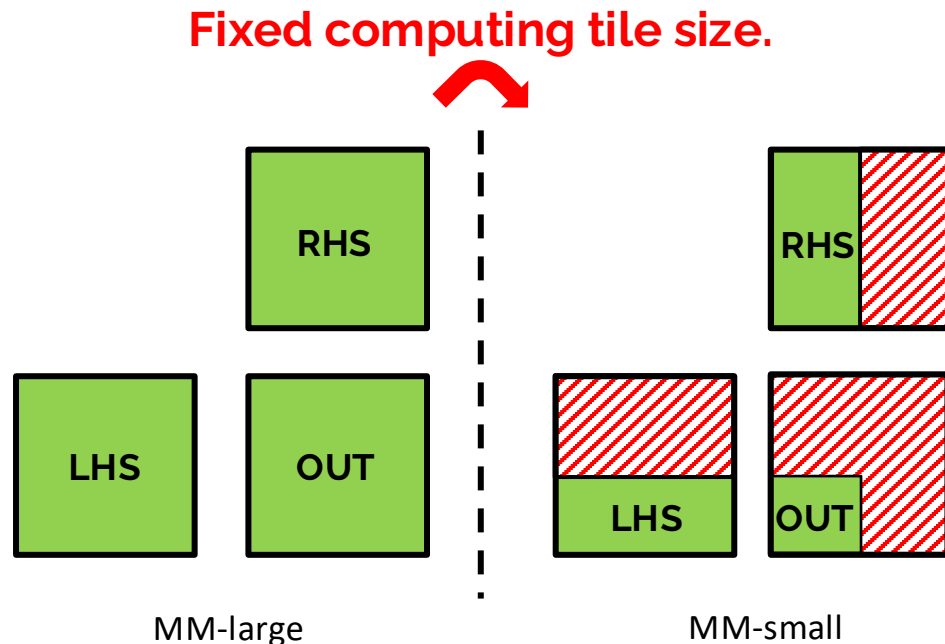
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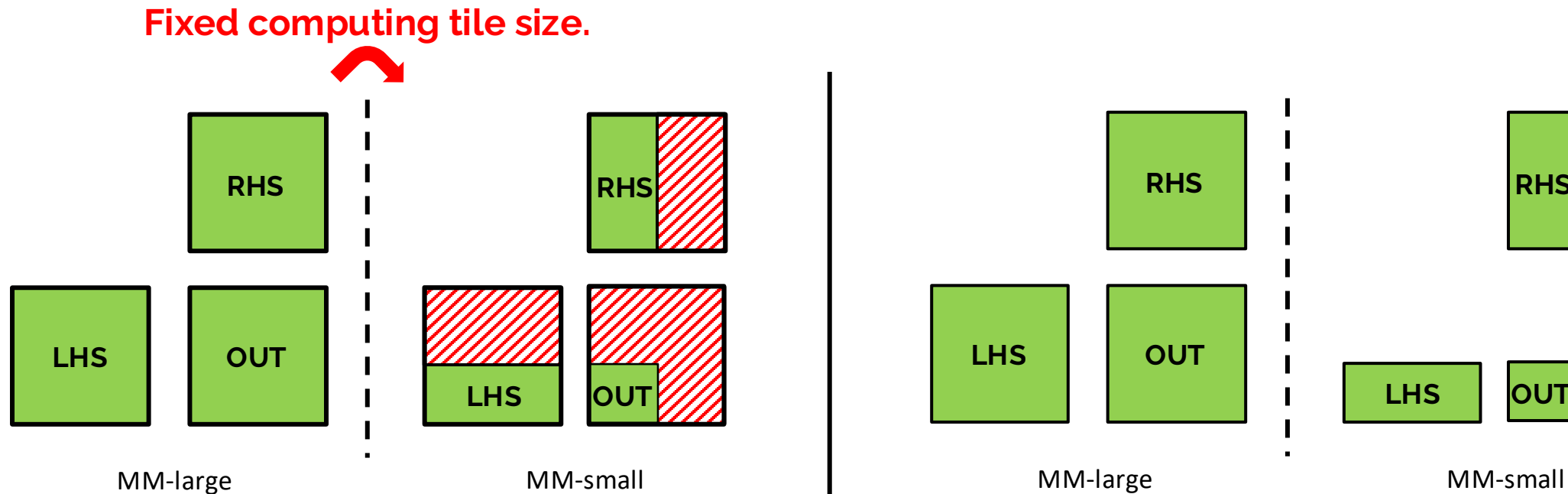
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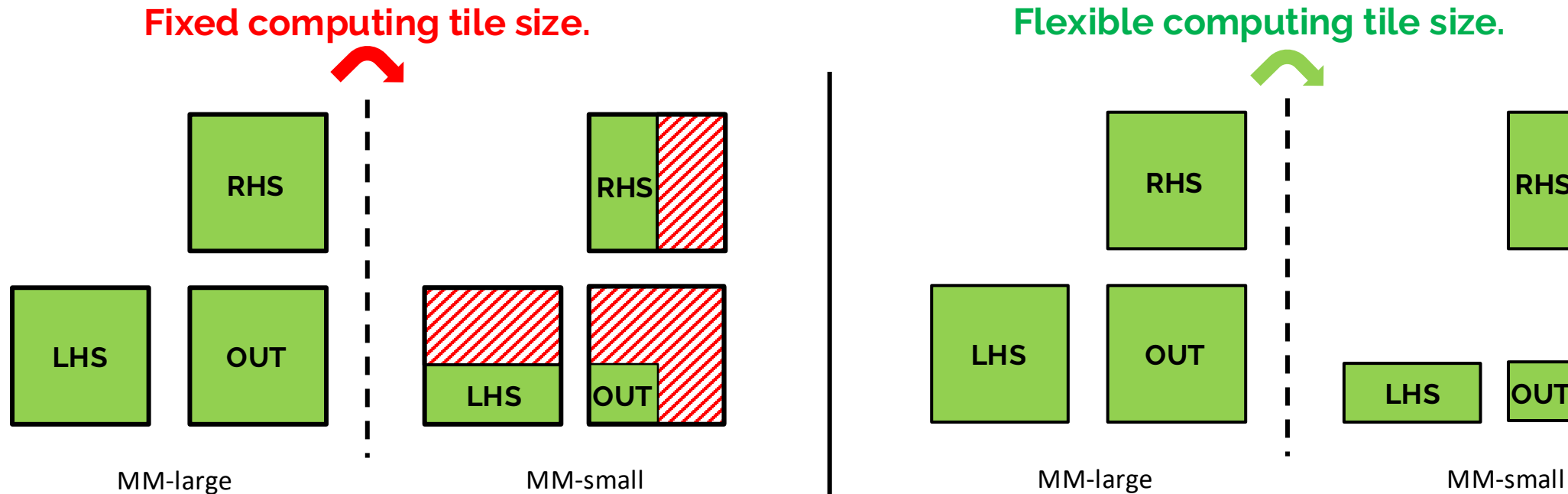
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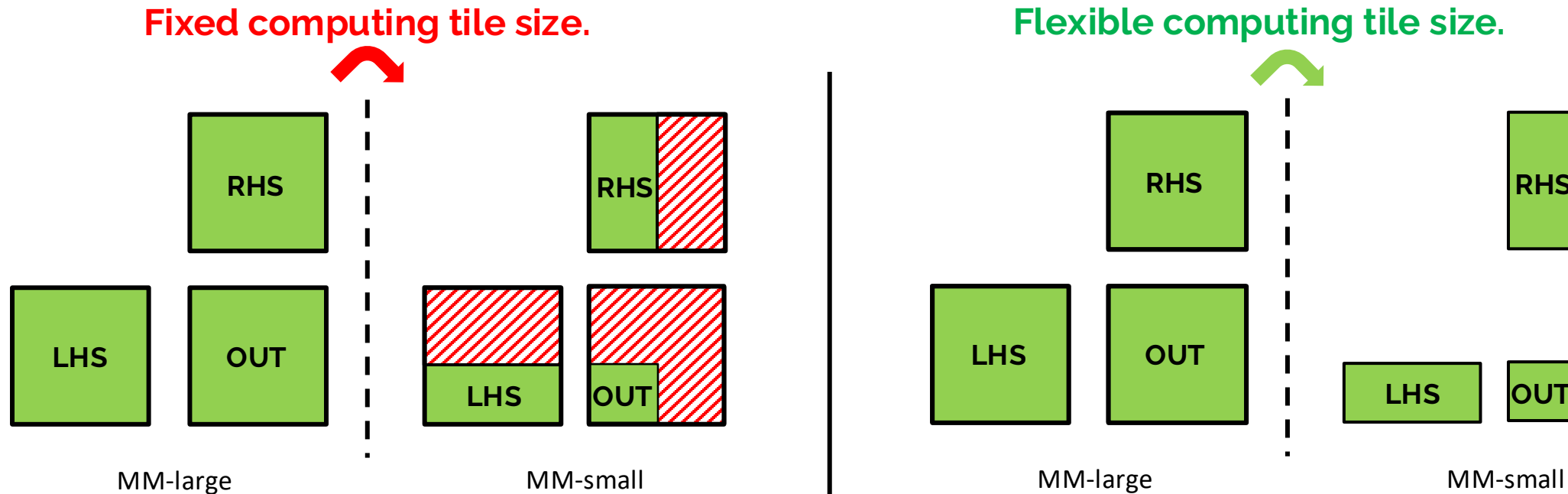
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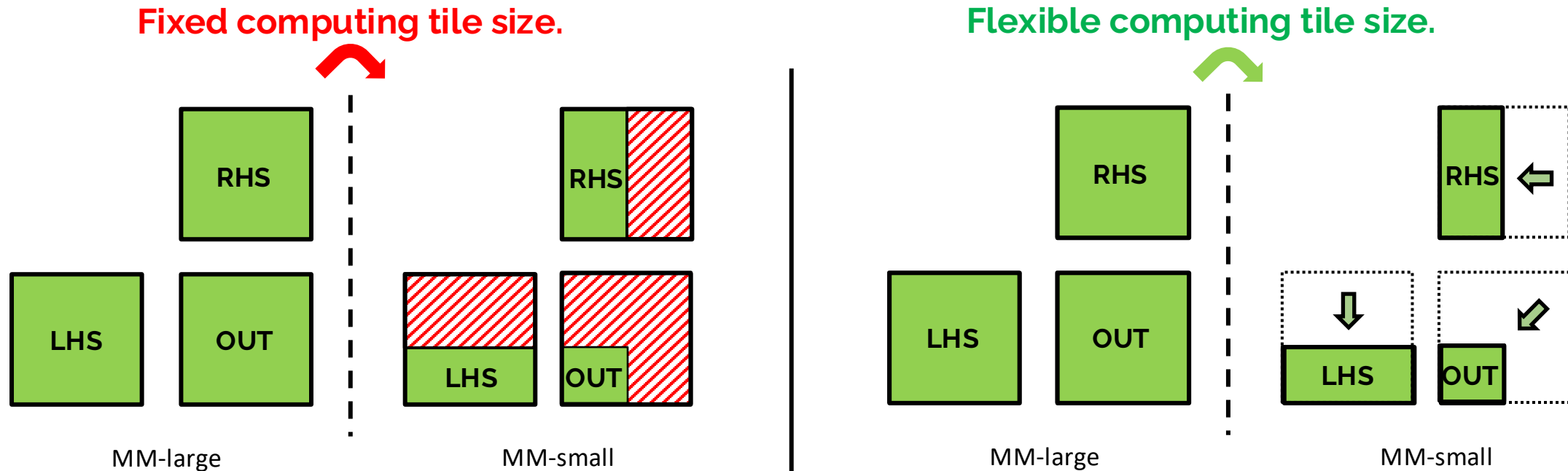
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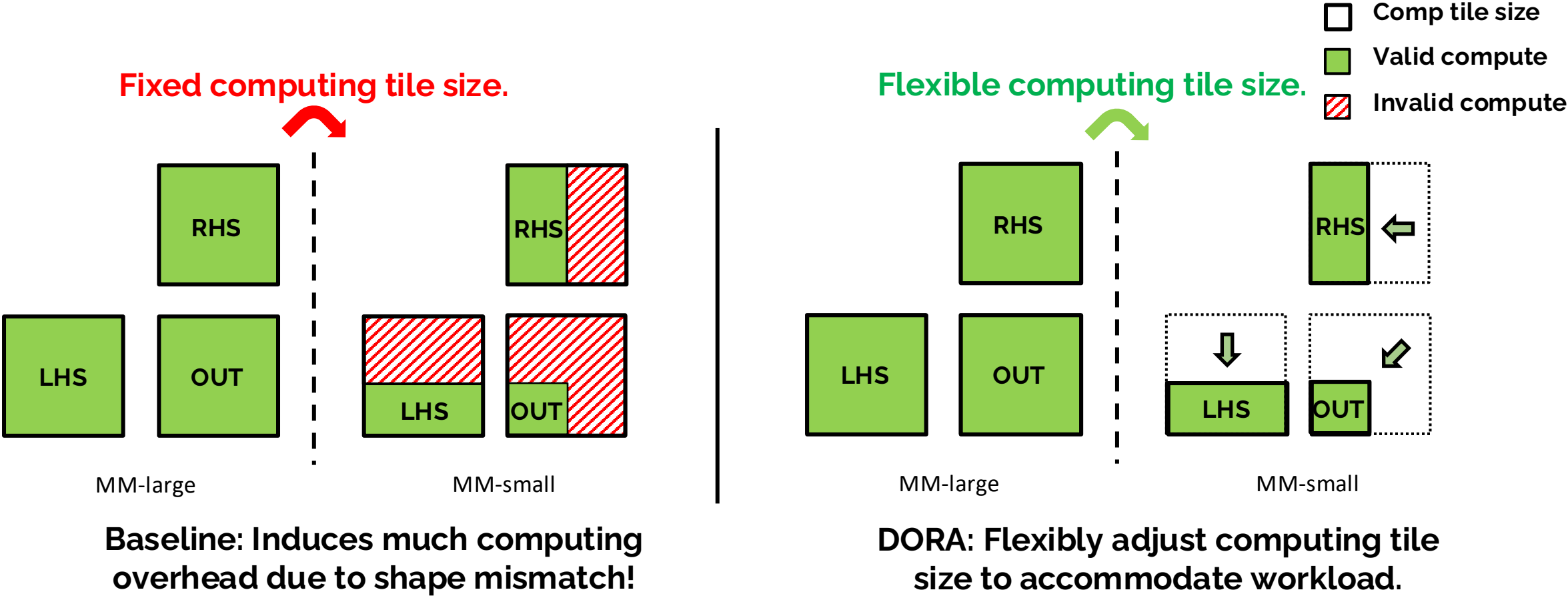
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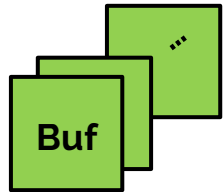
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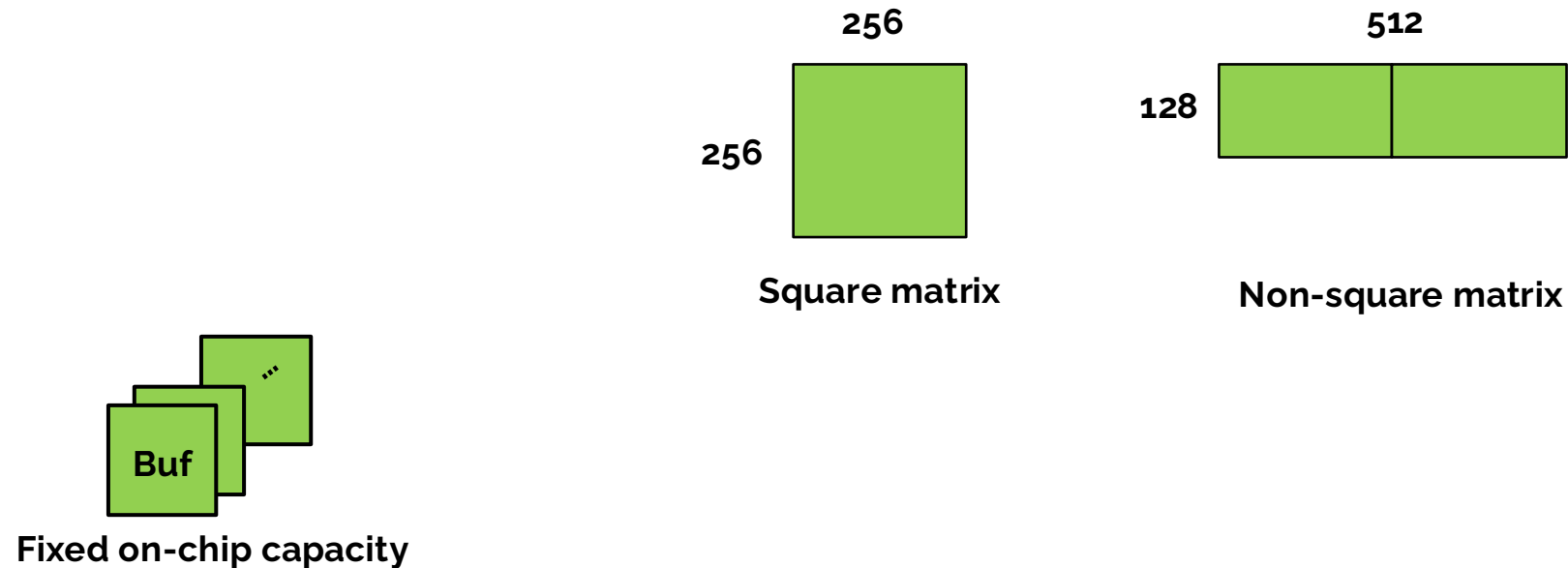
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Fixed on-chip capacity

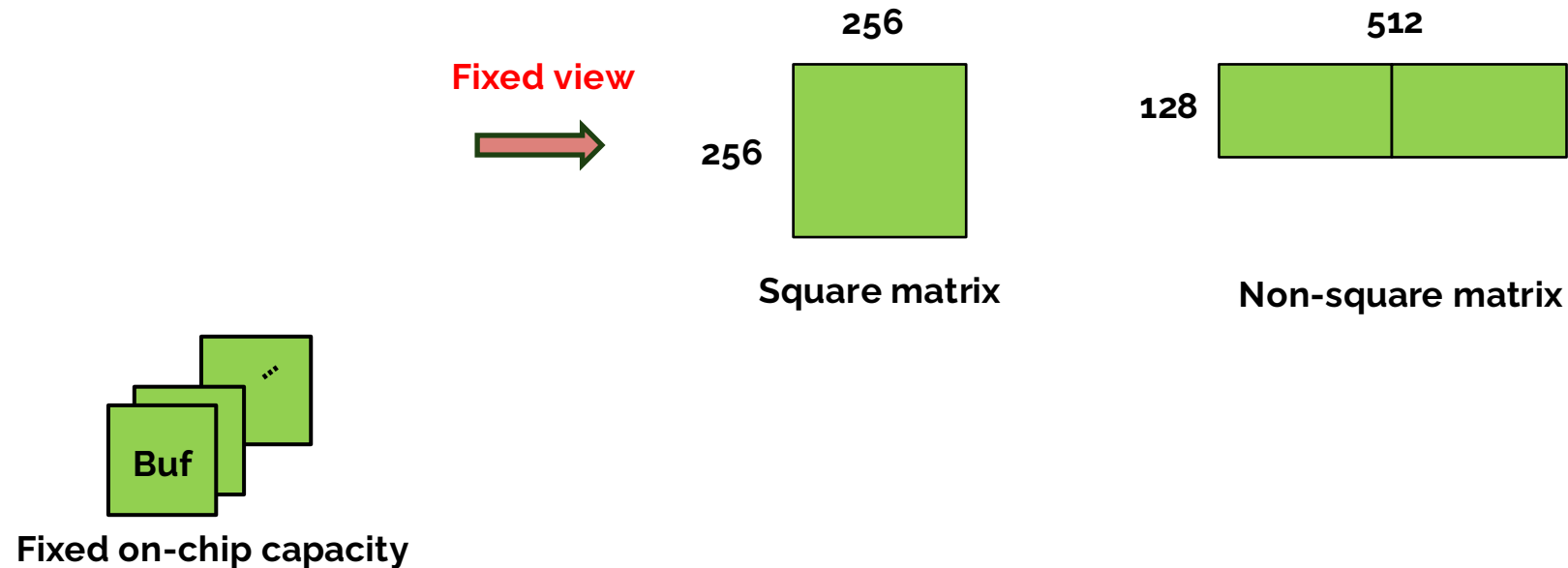
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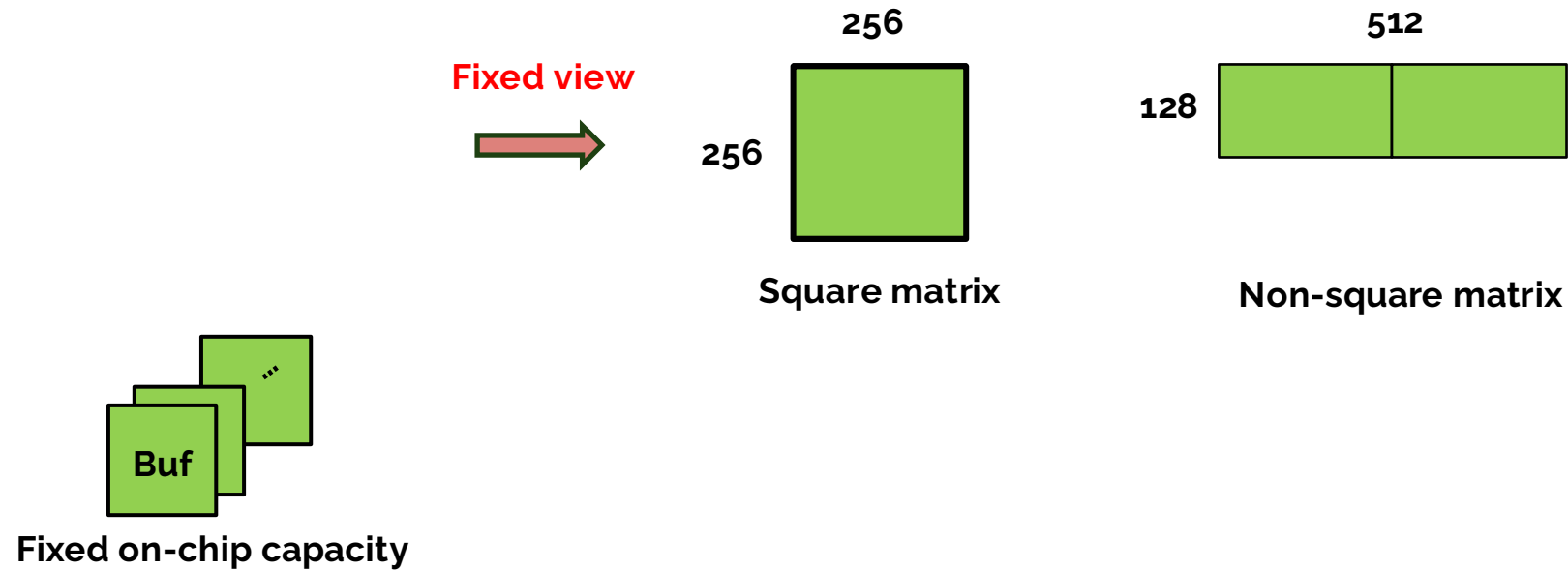
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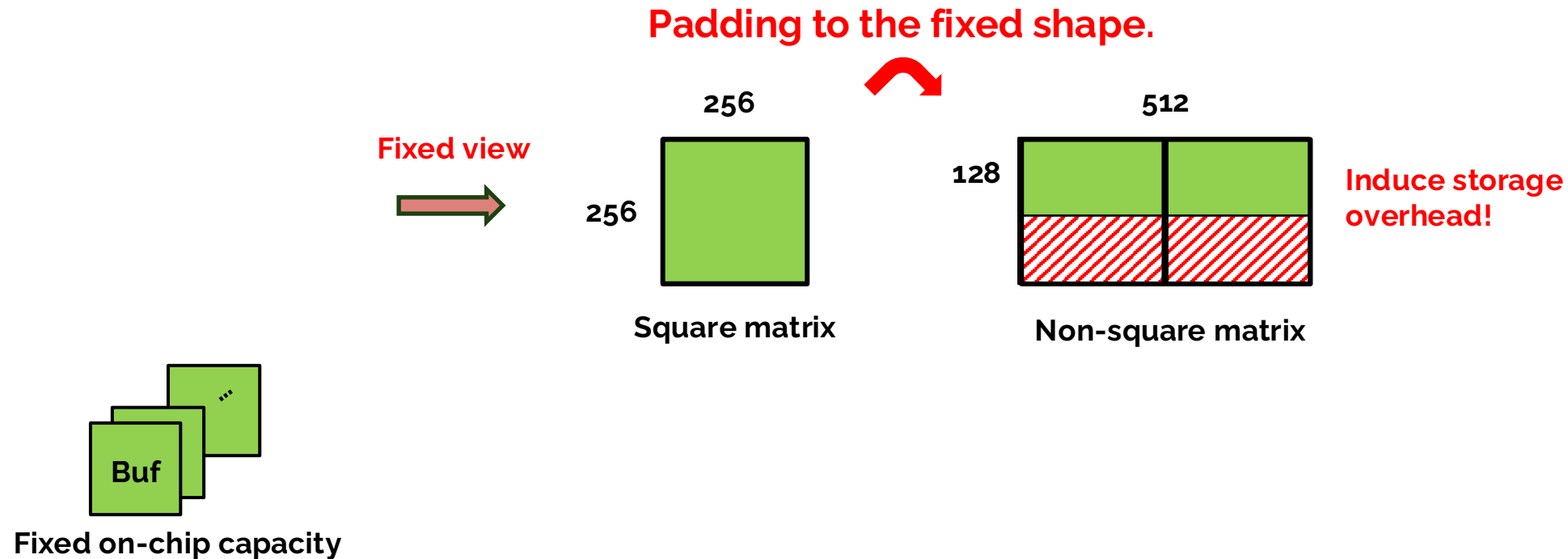
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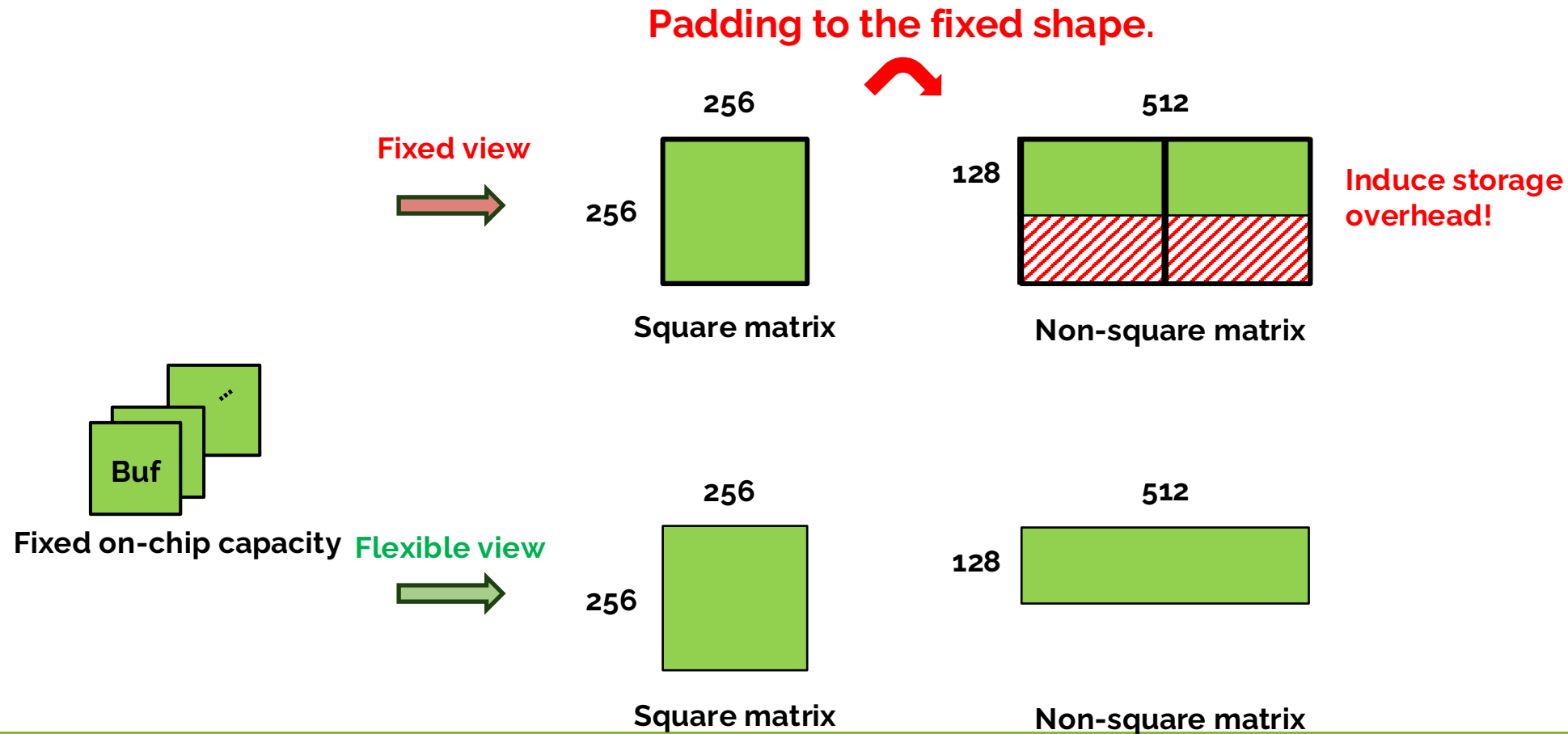
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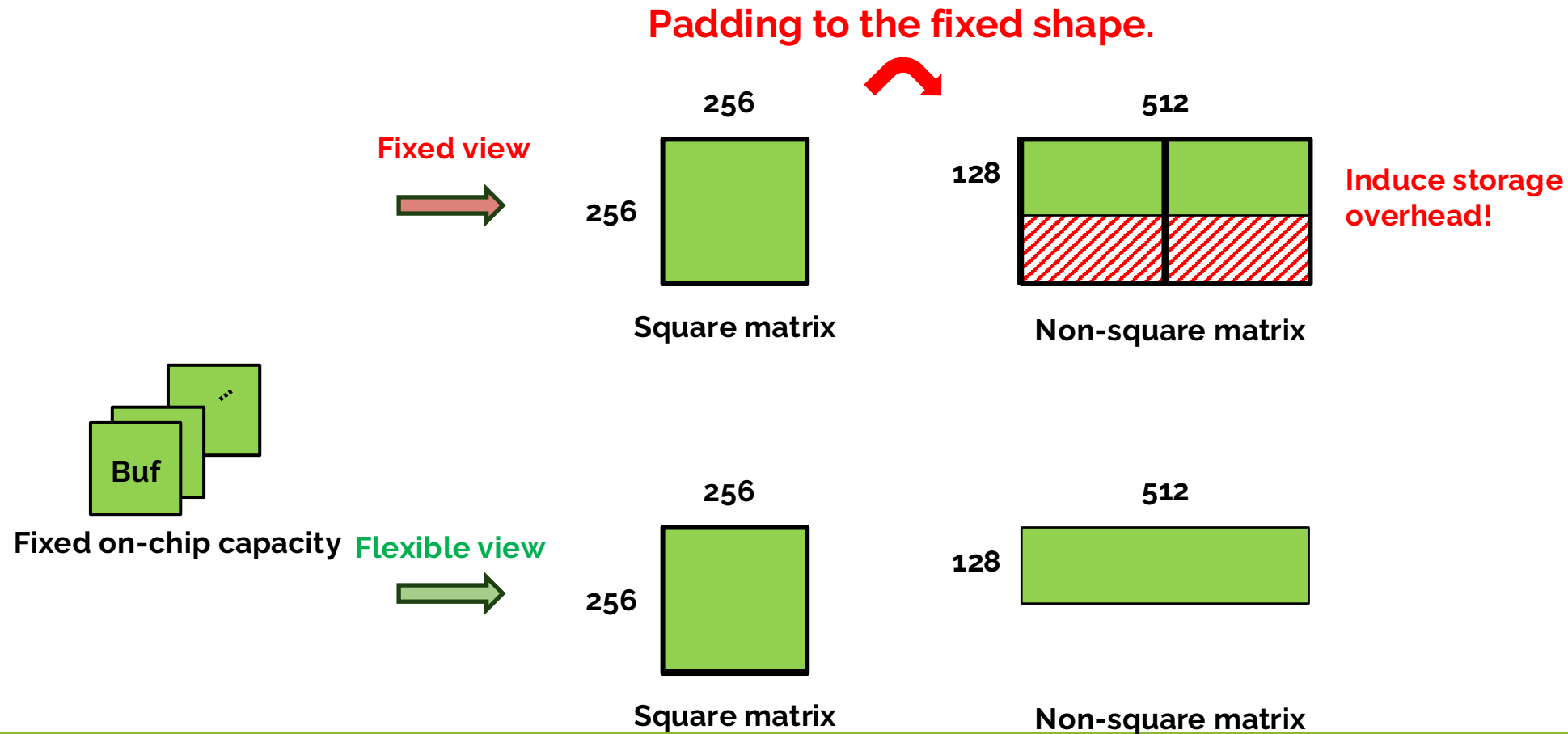
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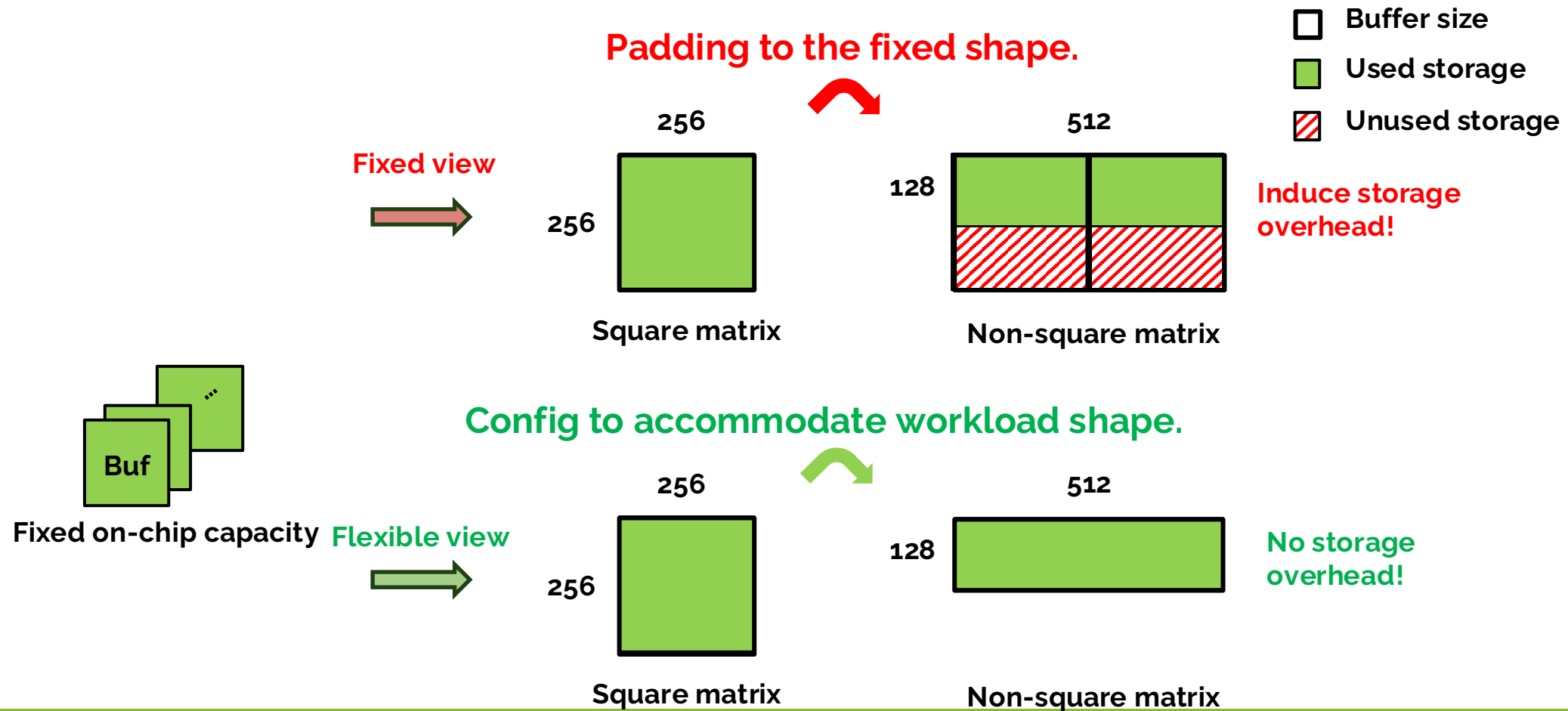
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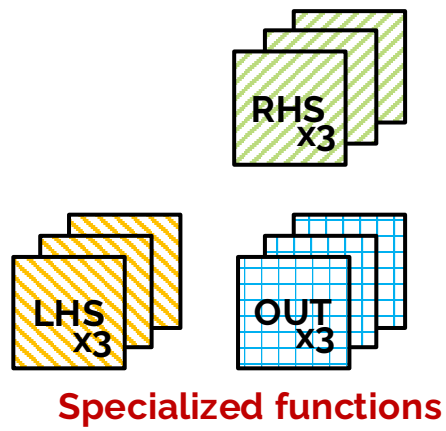
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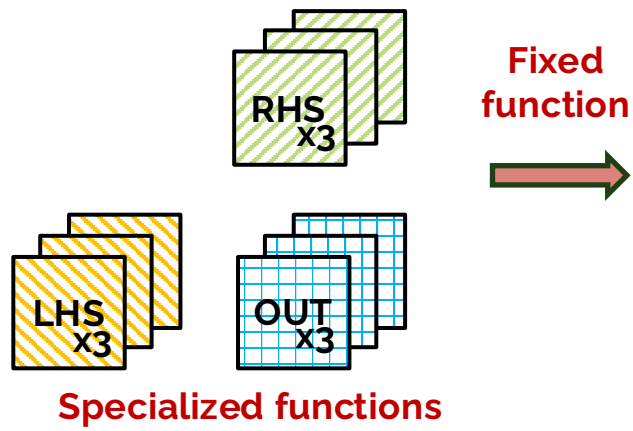
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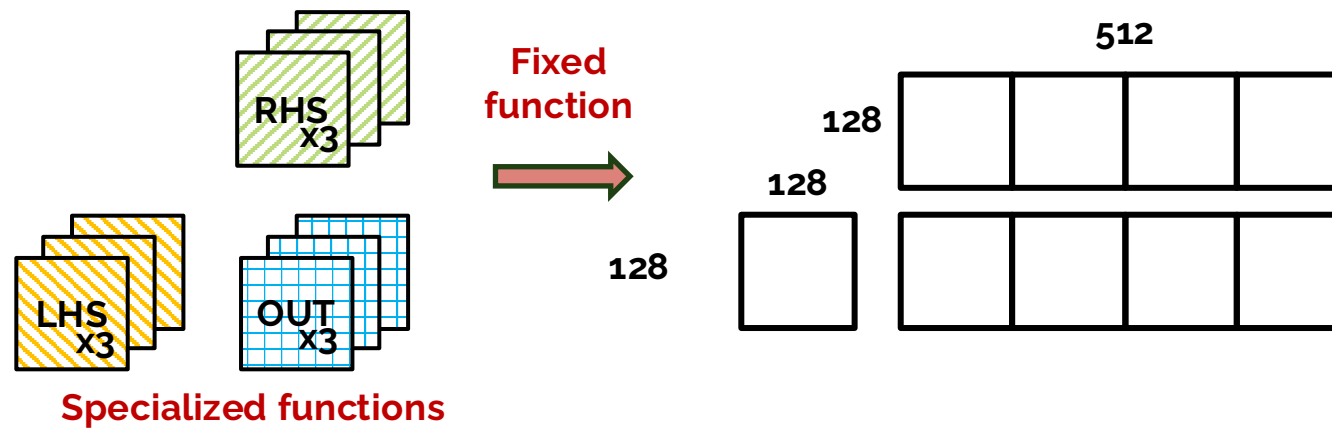
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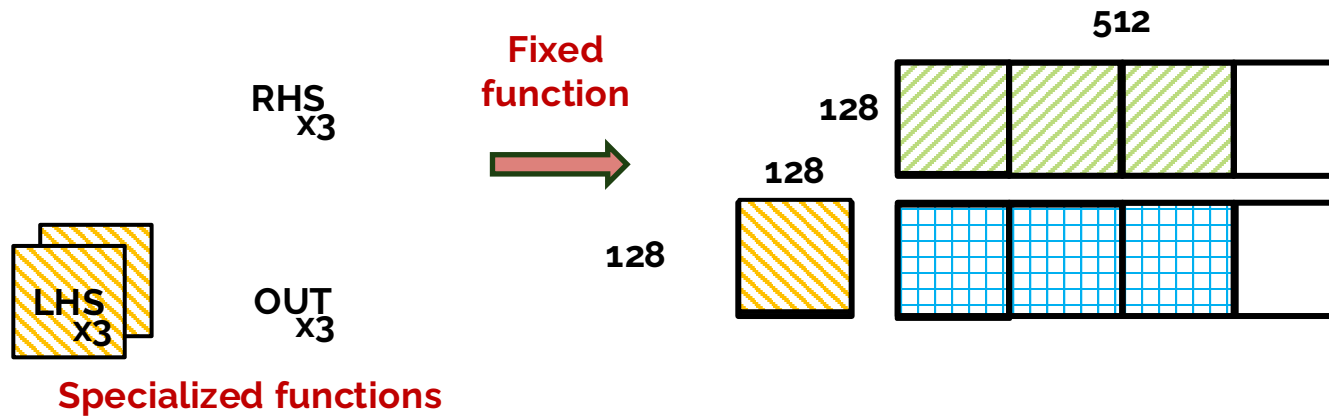
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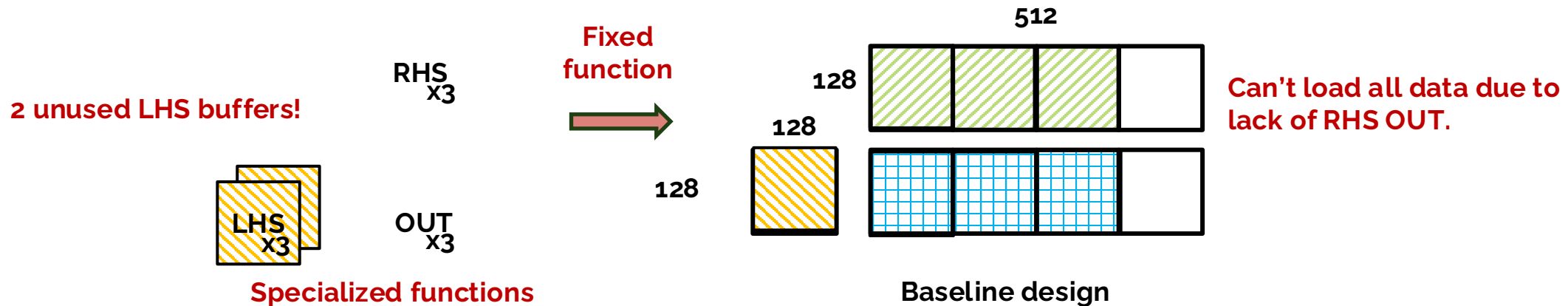
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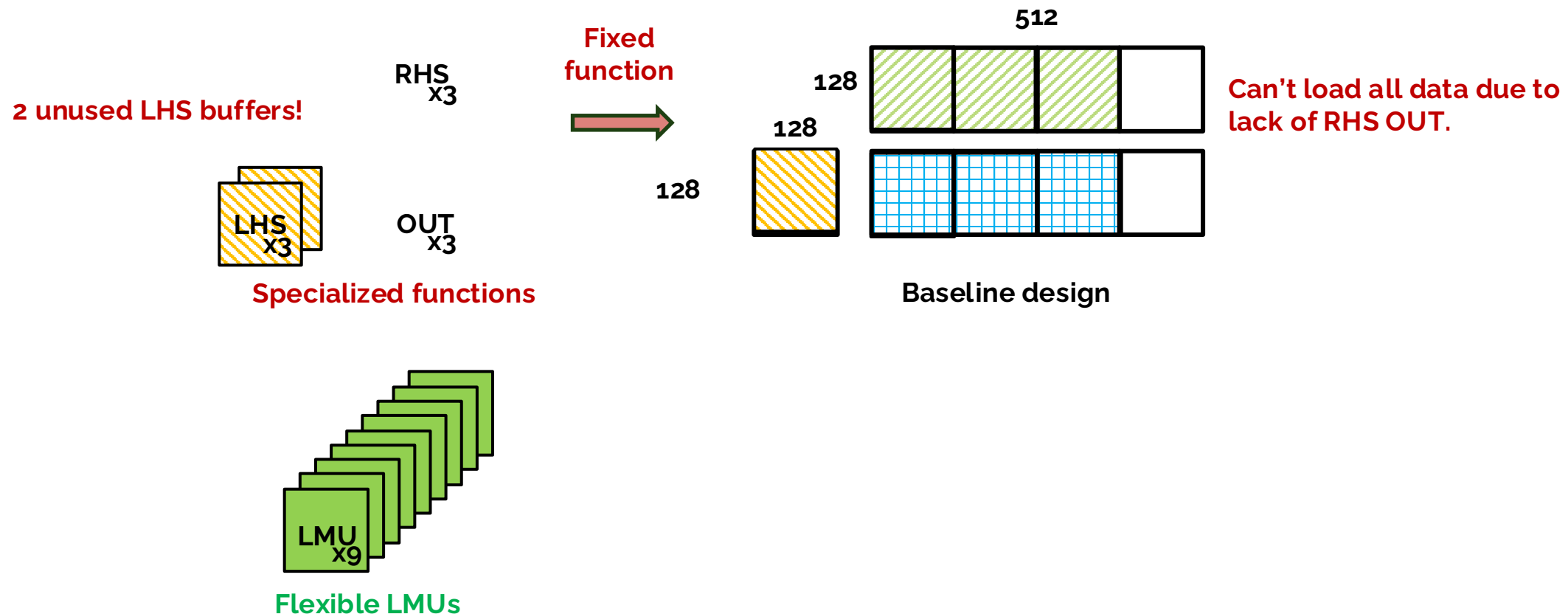
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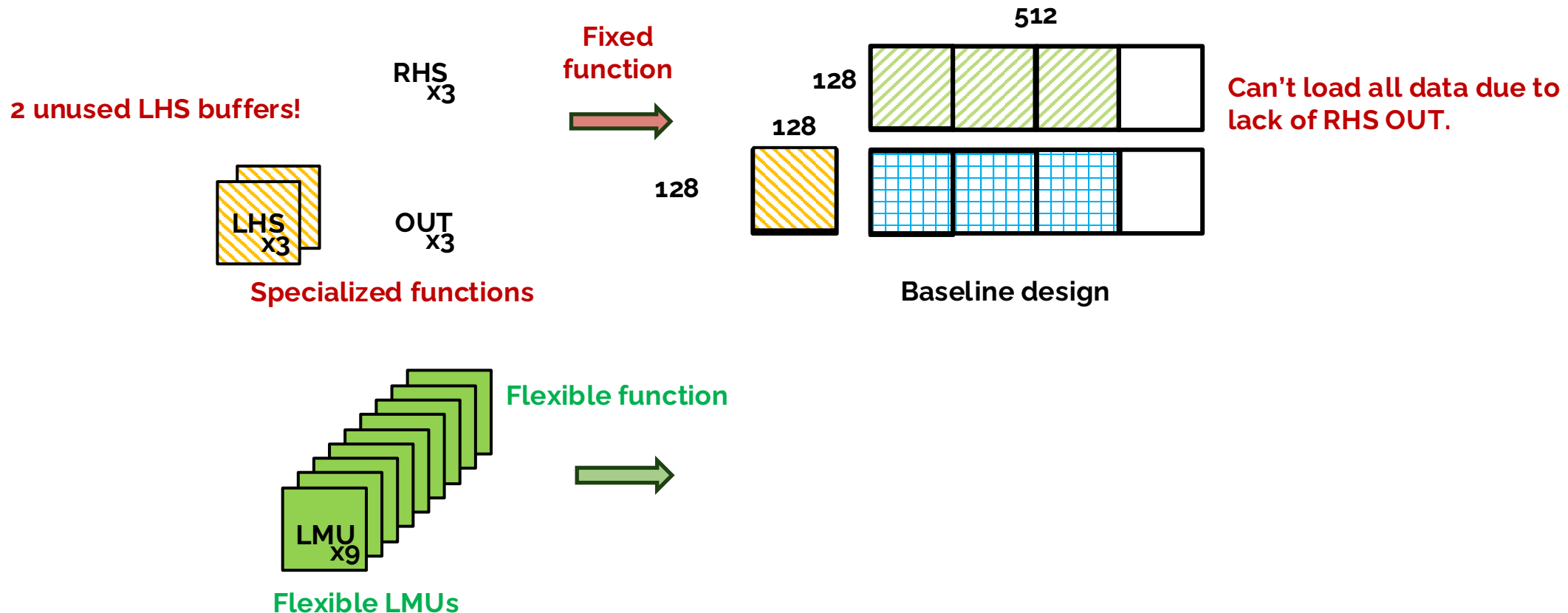
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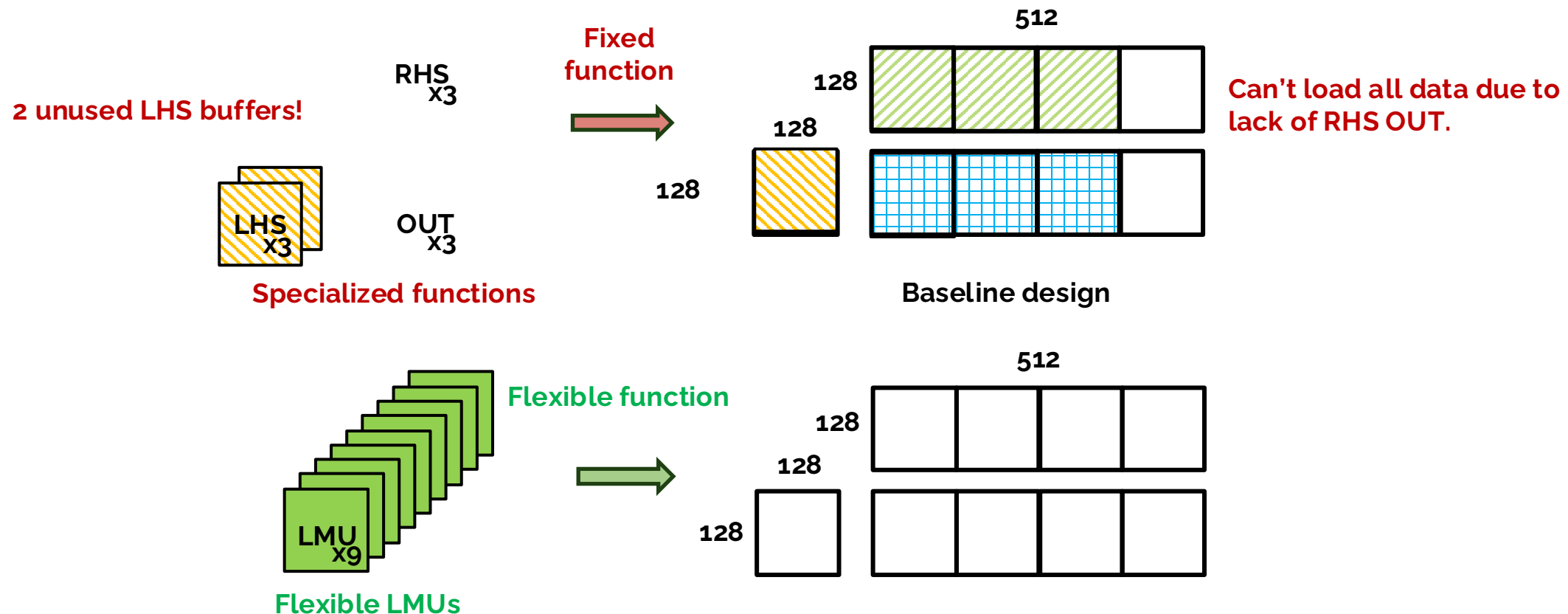
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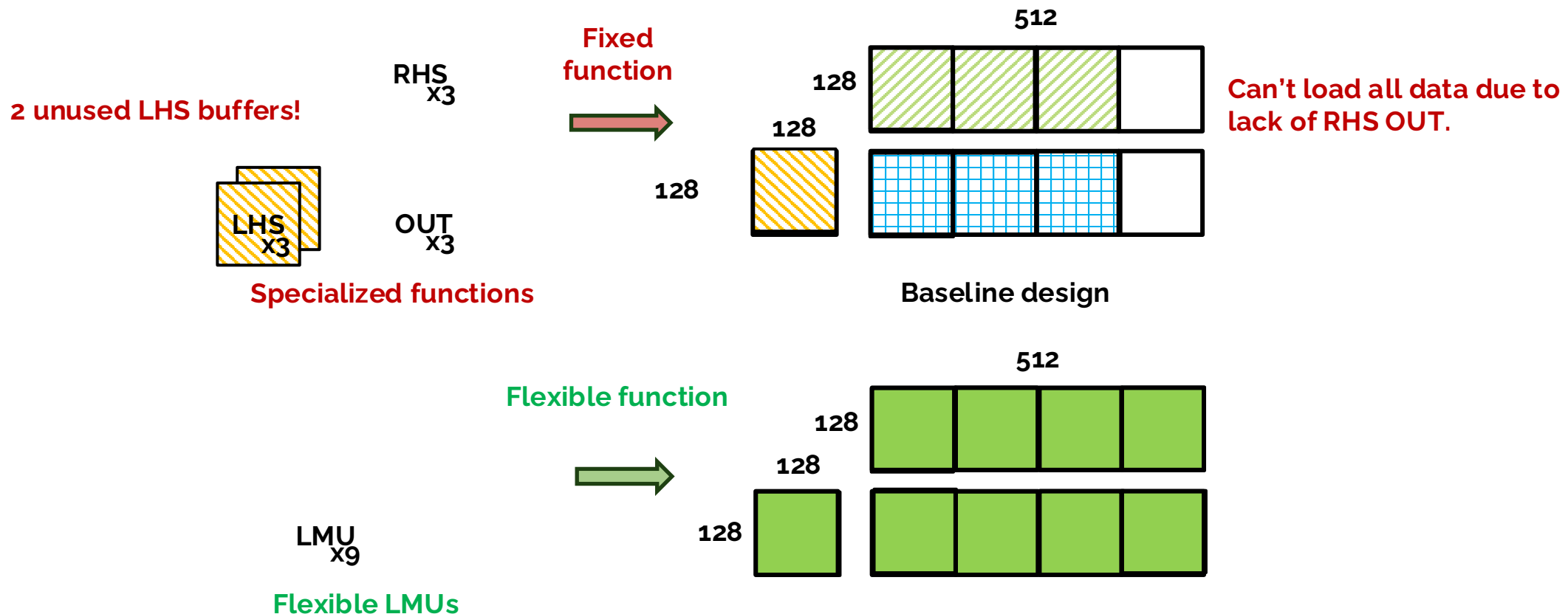
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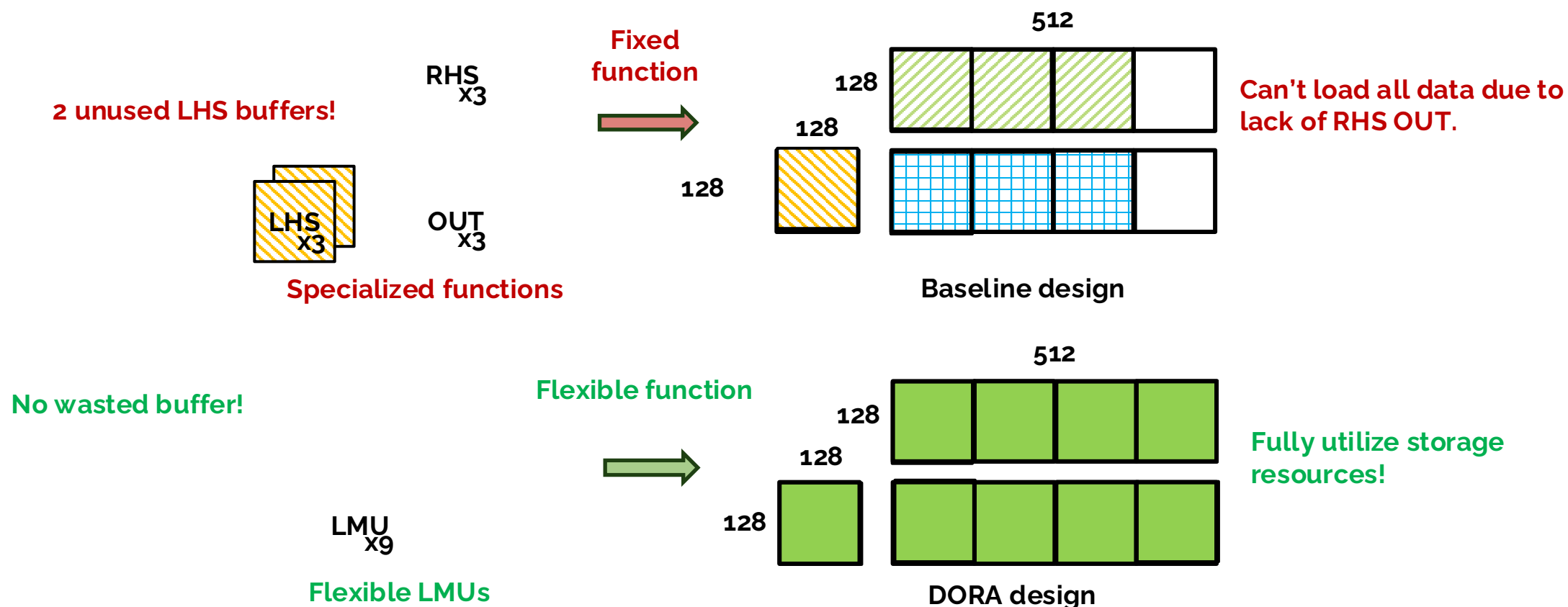
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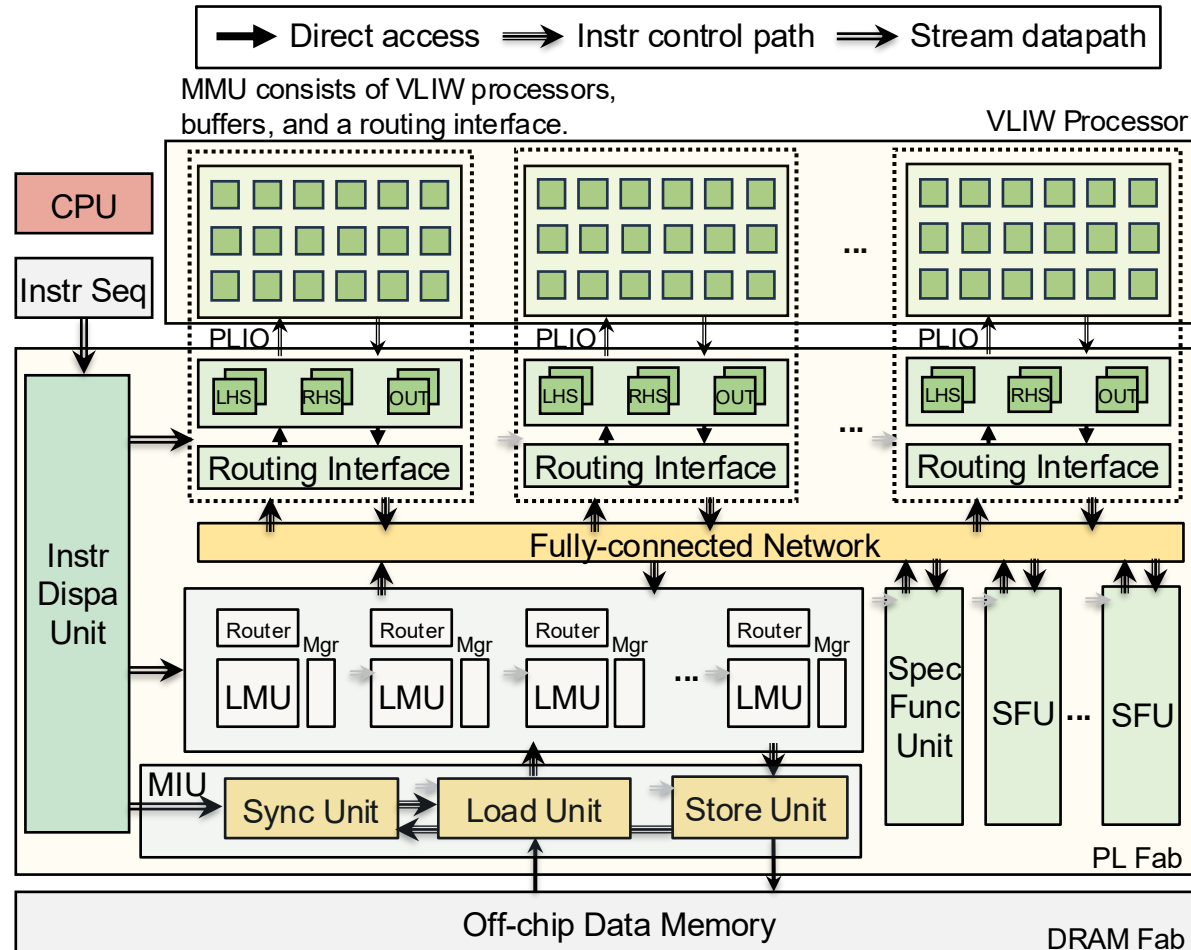
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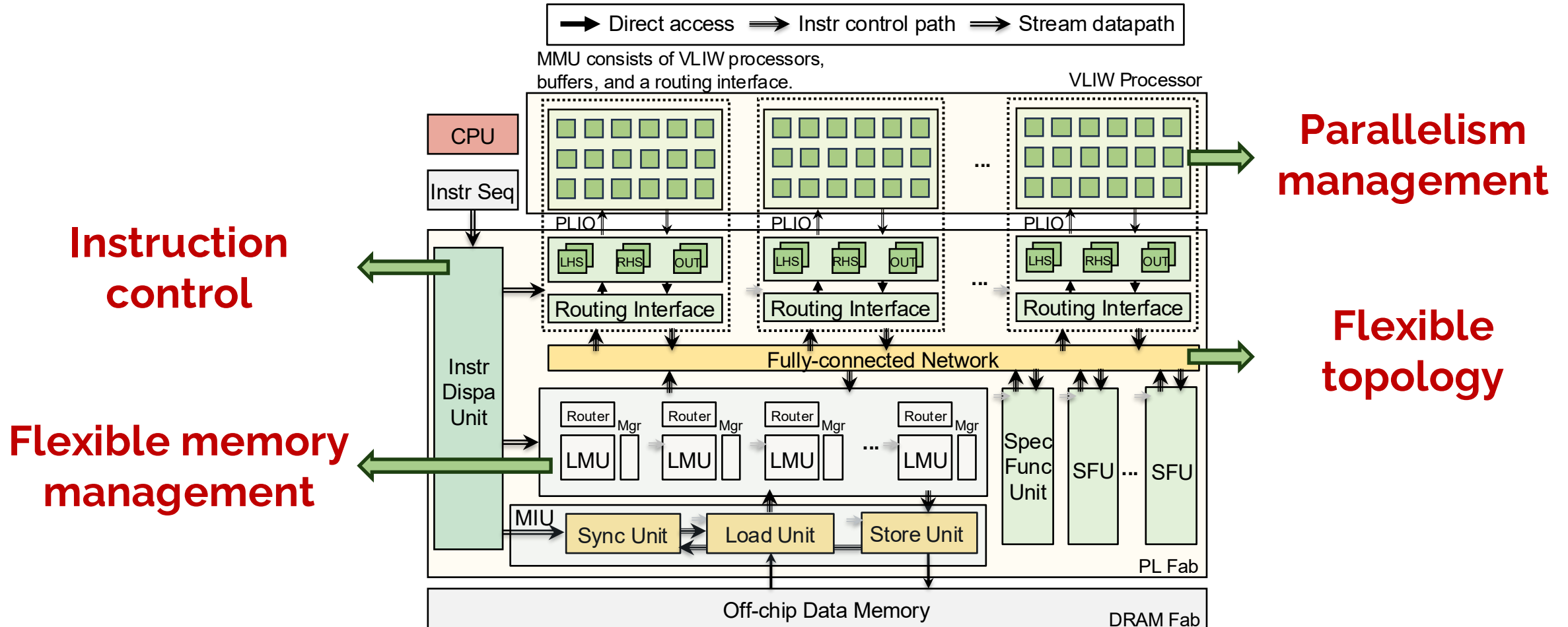
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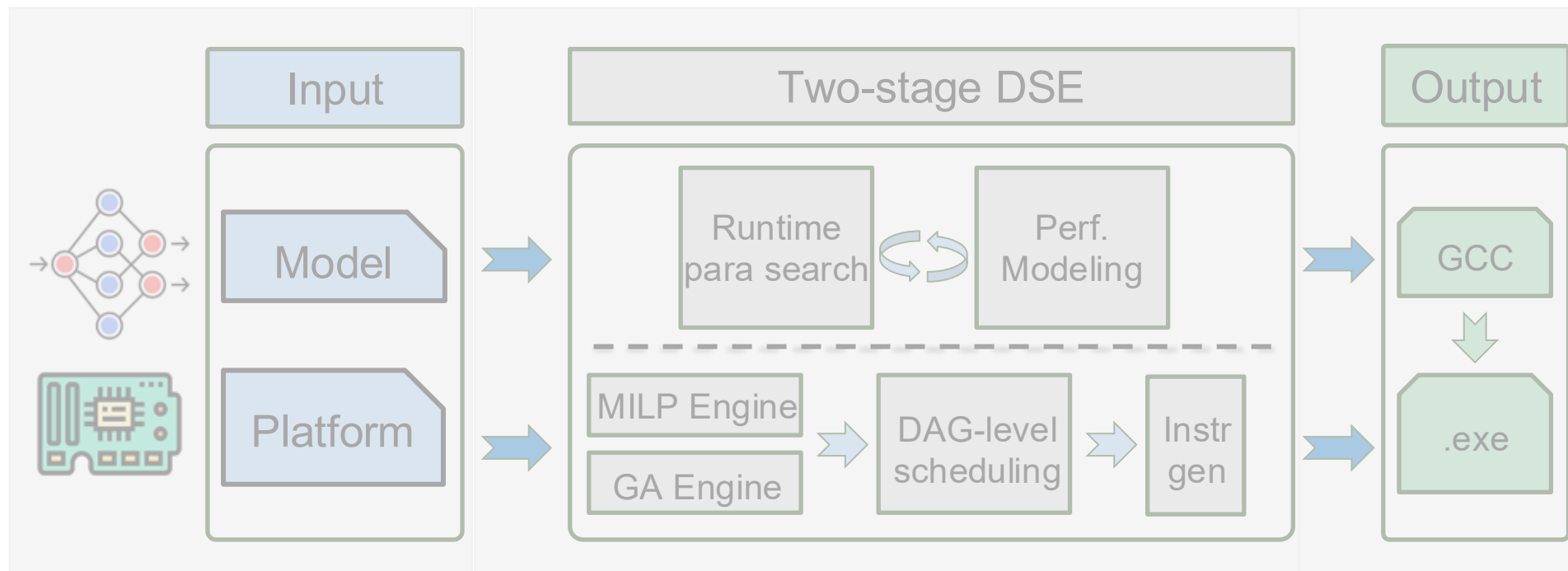
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How to **efficiently** program DORA overlay for **software developers**?



DORA Framework

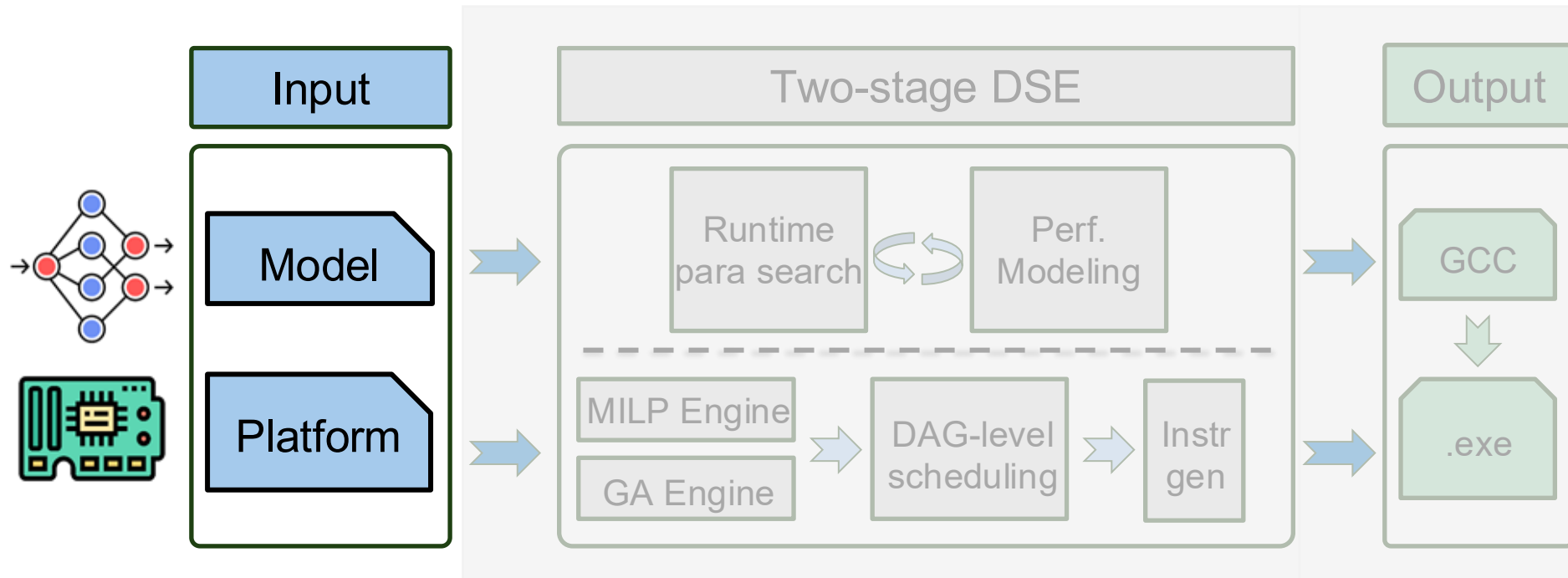
DORA Framework



DORA Framework

Input:

- 1) MM-based AI Model;
- 2) Hardware platforms information.



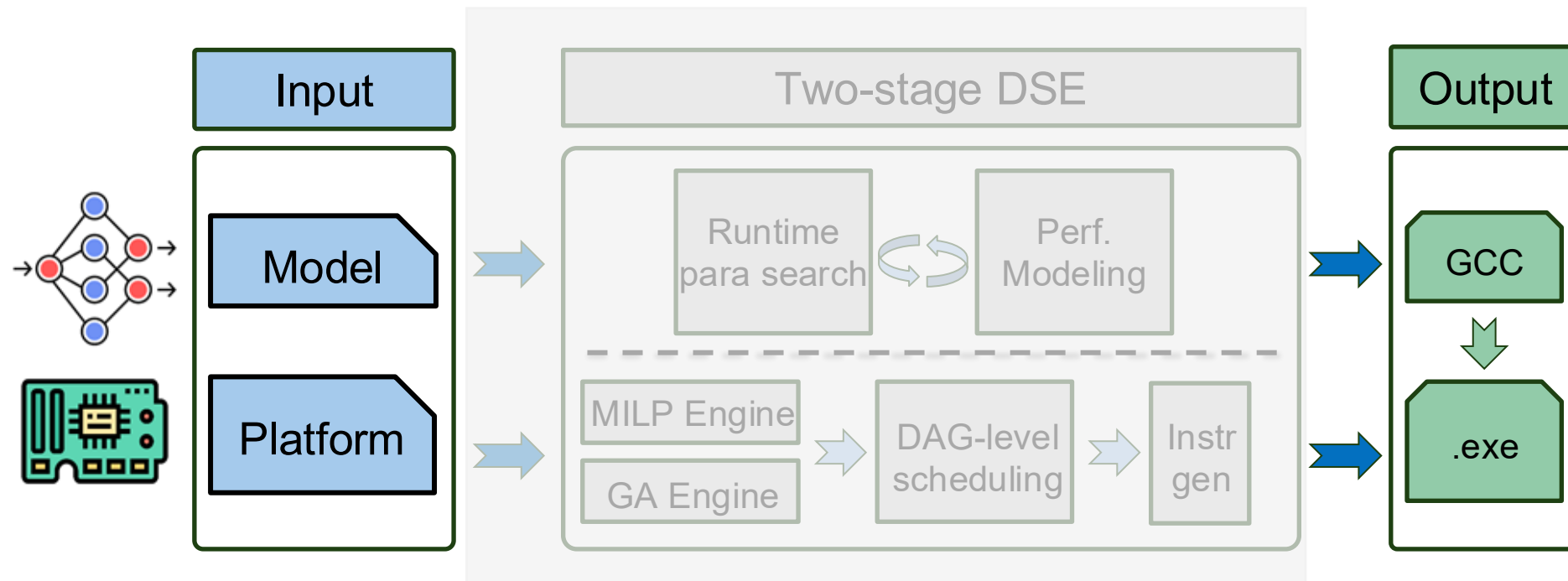
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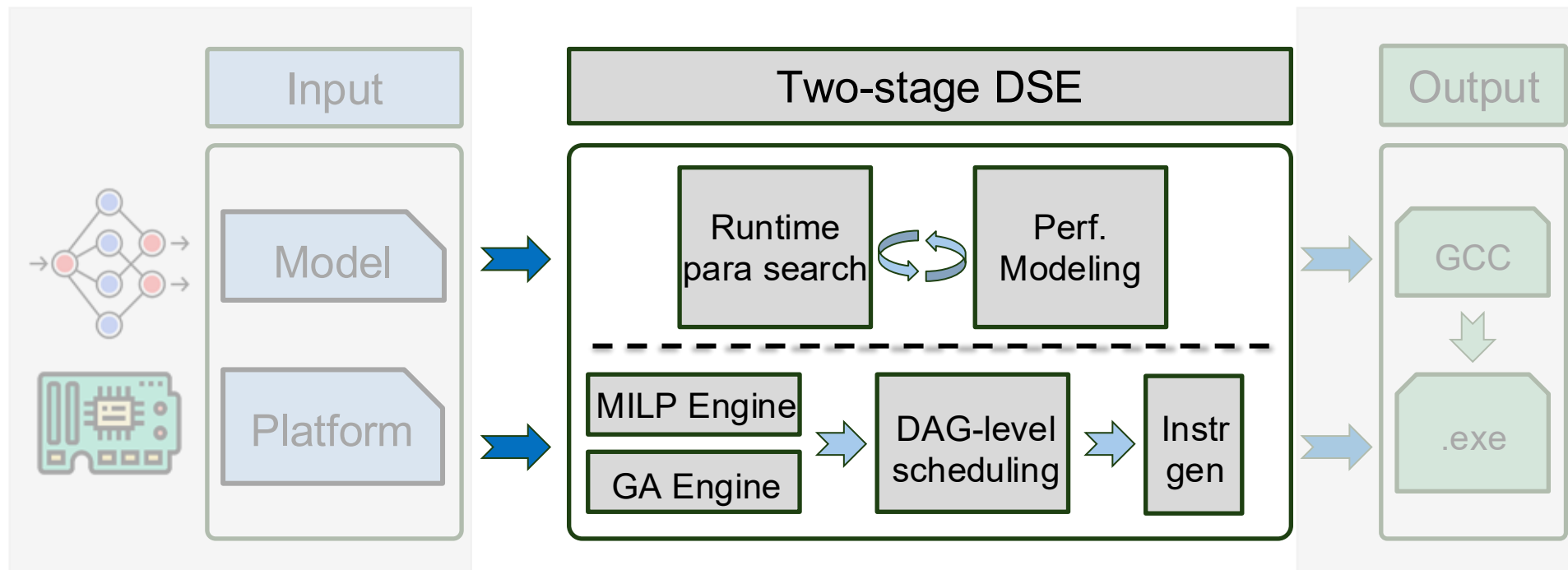
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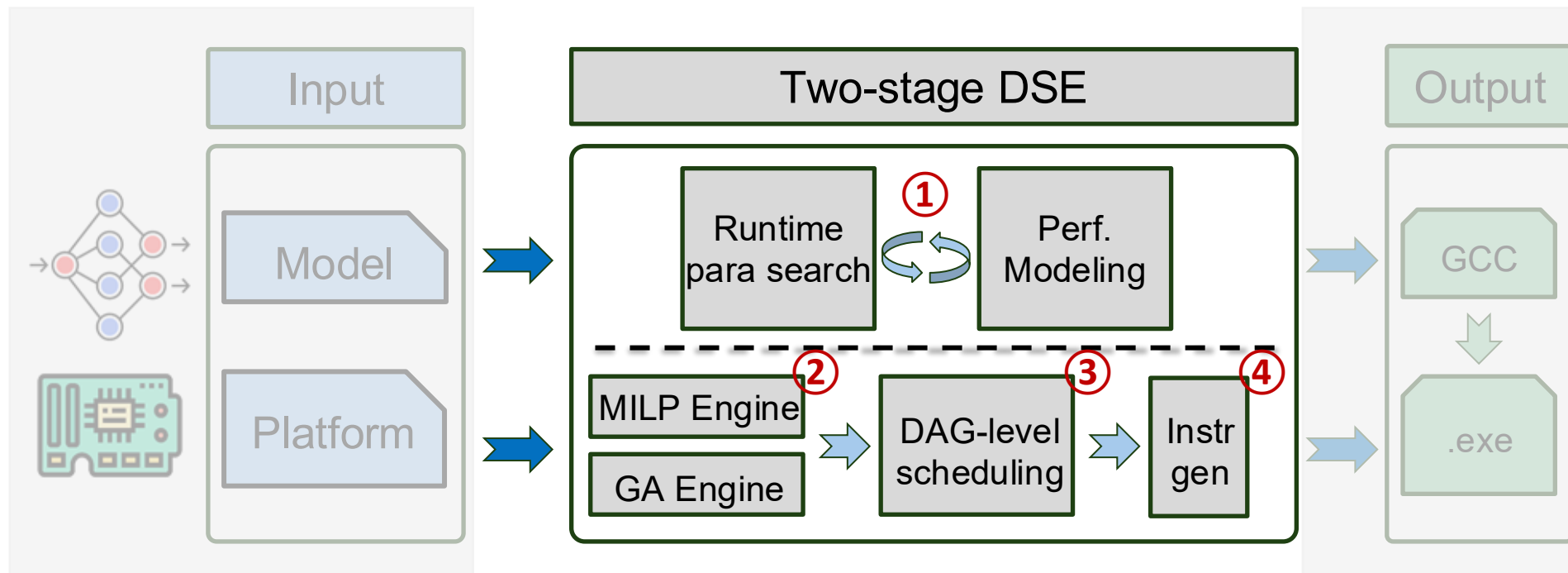
- 1) Bitstream running on the AIE and PL;
- 2) Executable file running on ARM CPU.



DORA Framework

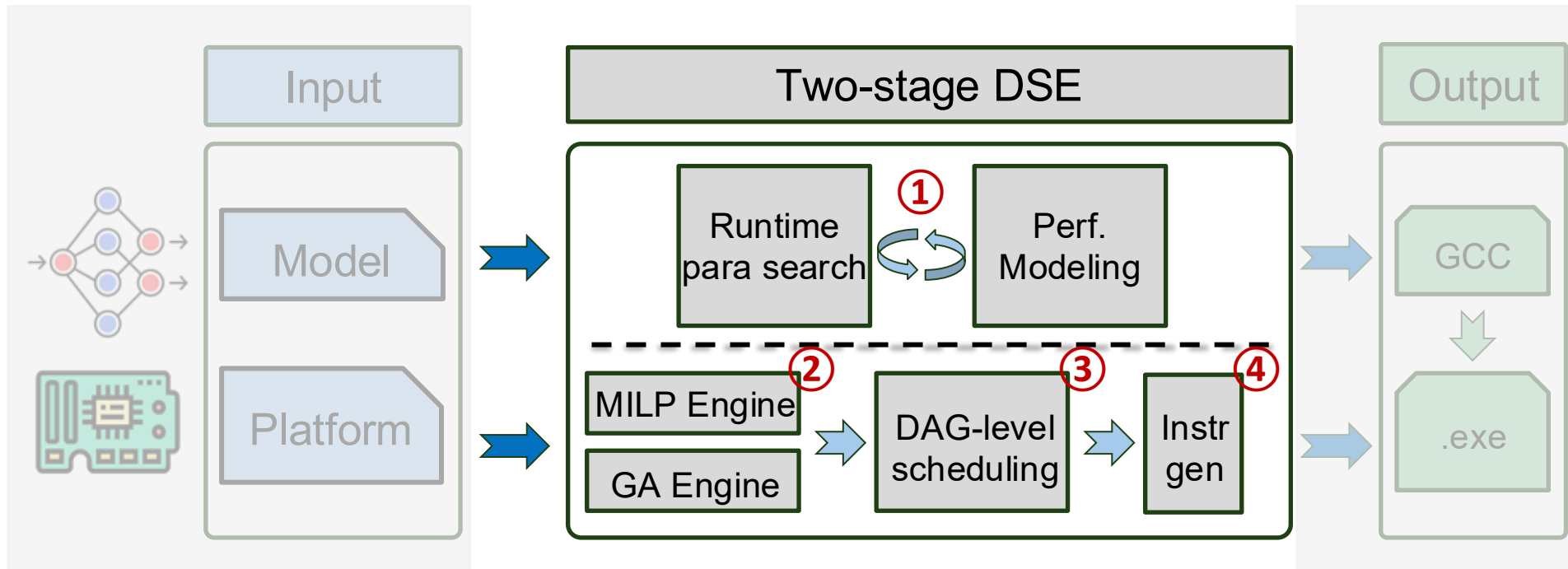


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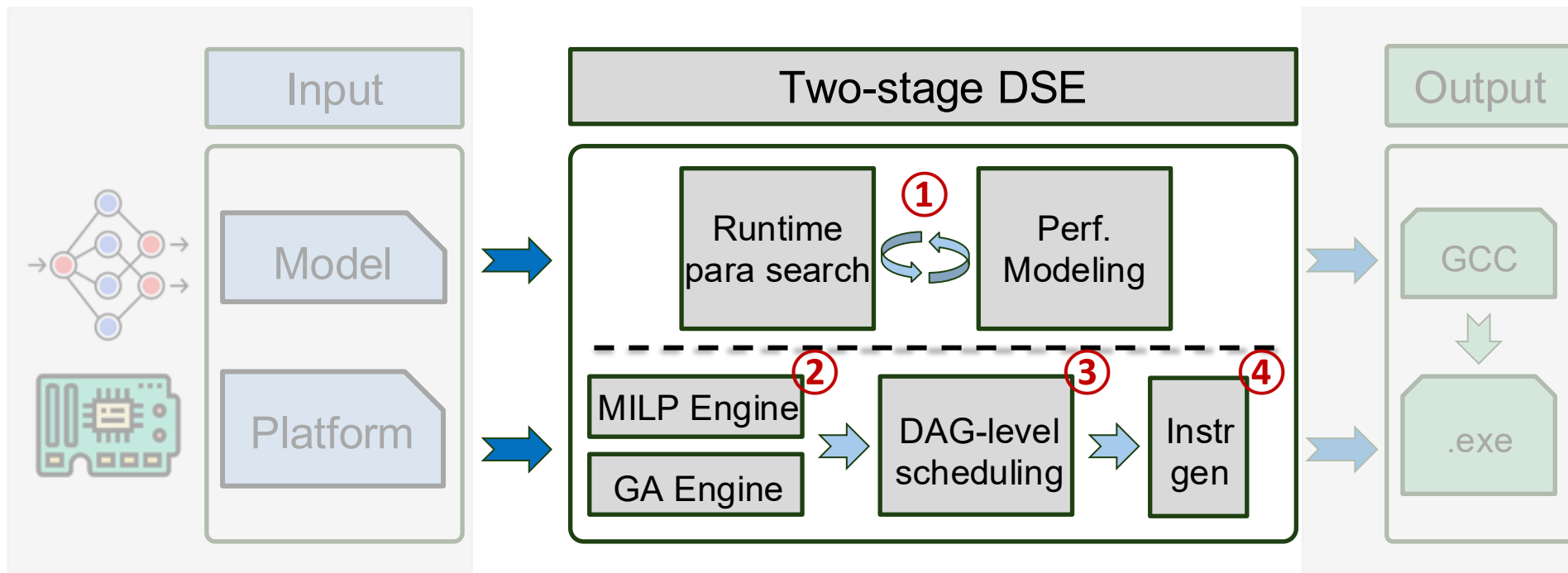
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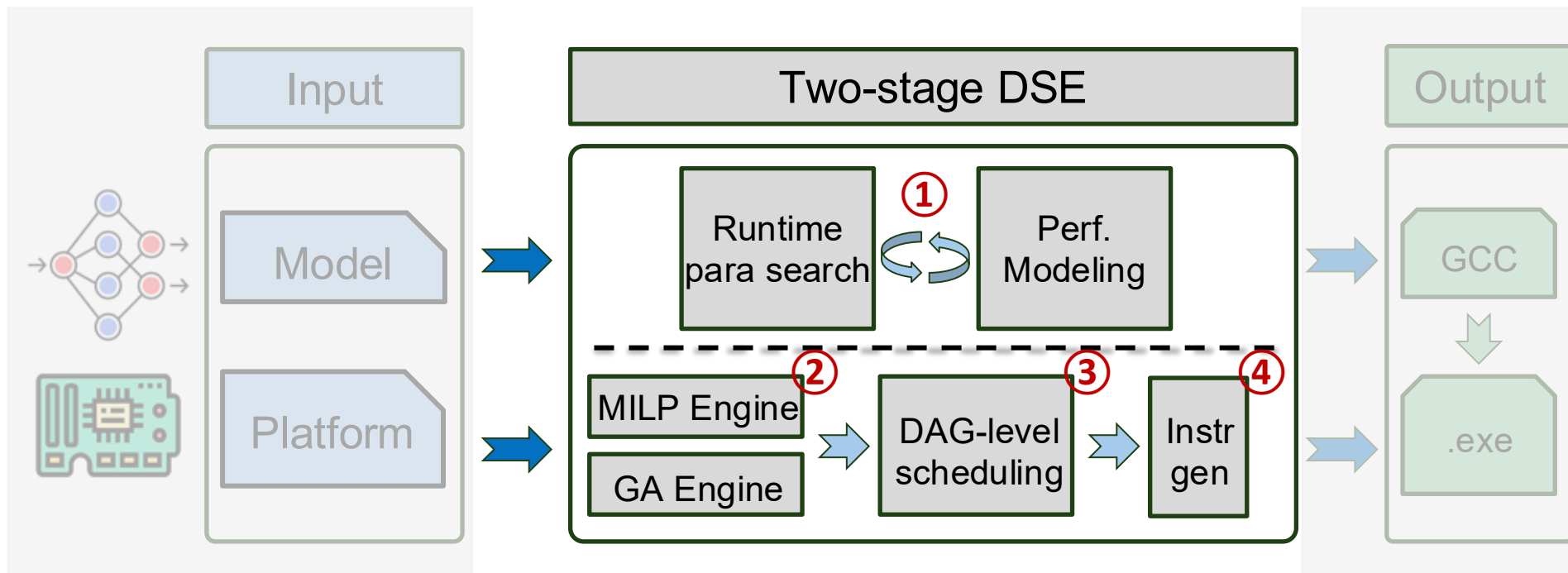
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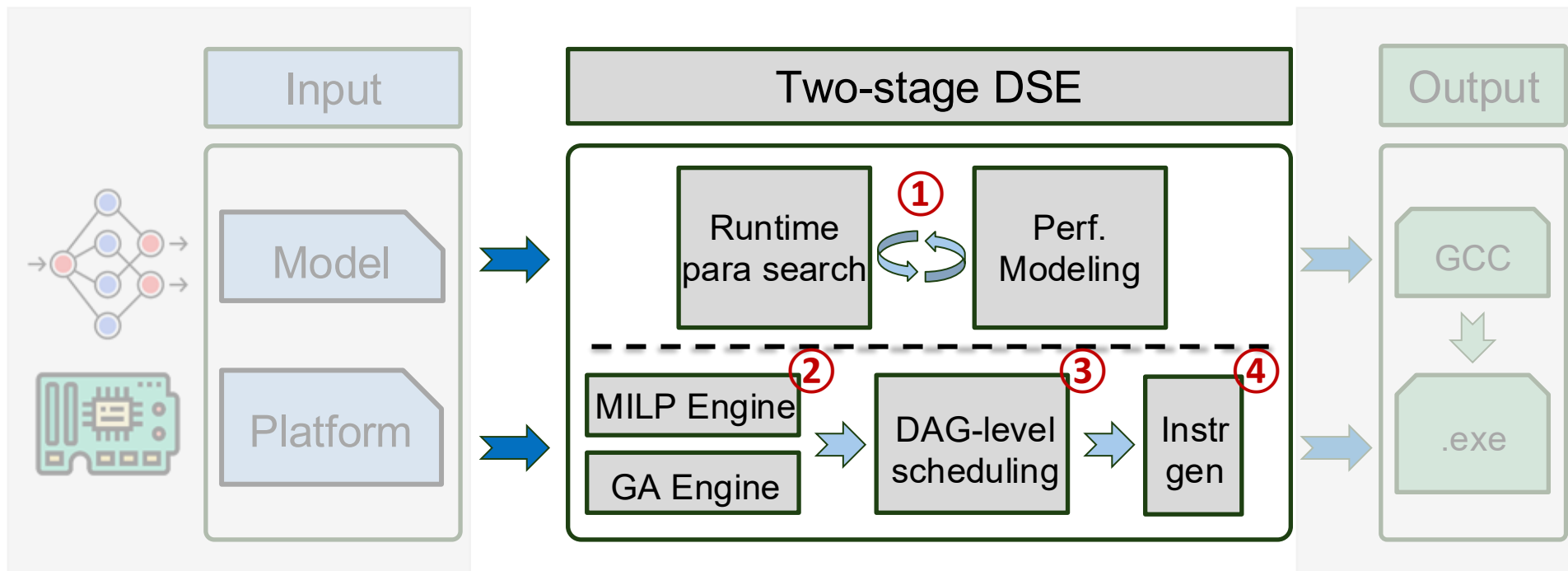
① Single-layer brute-force design space exploration.

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- Store the candidate execution table for the second stage.



DORA Framework

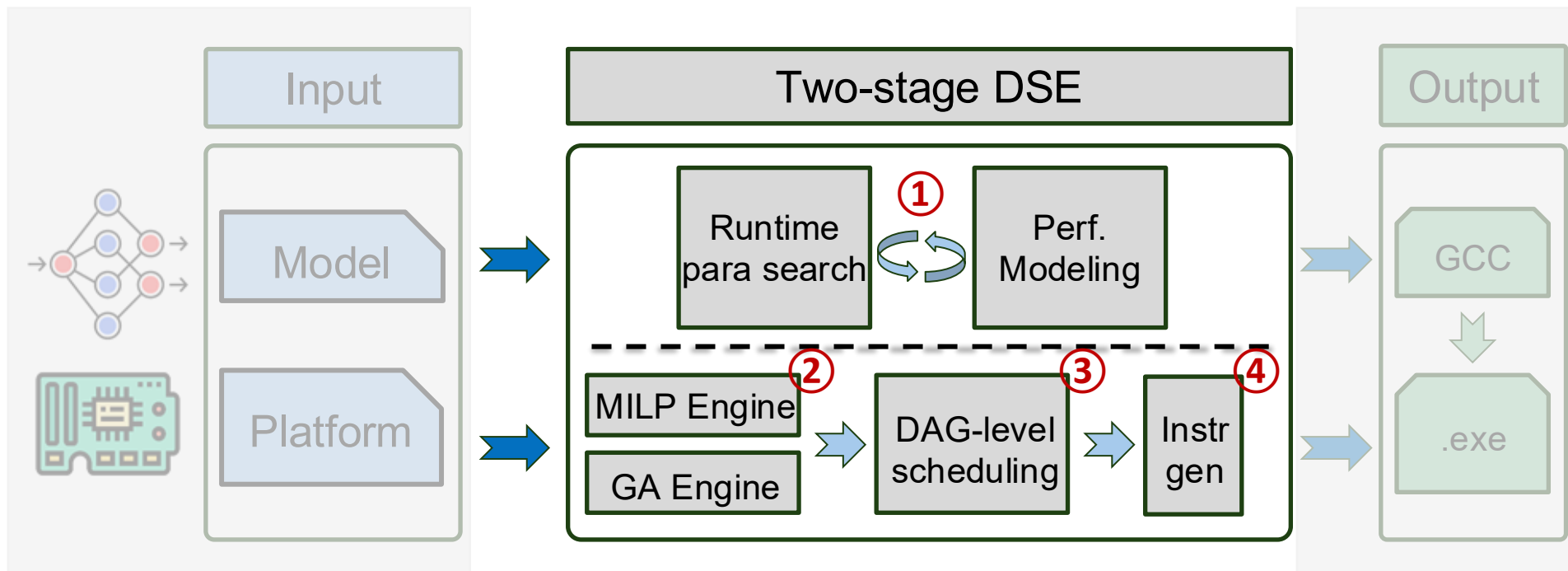
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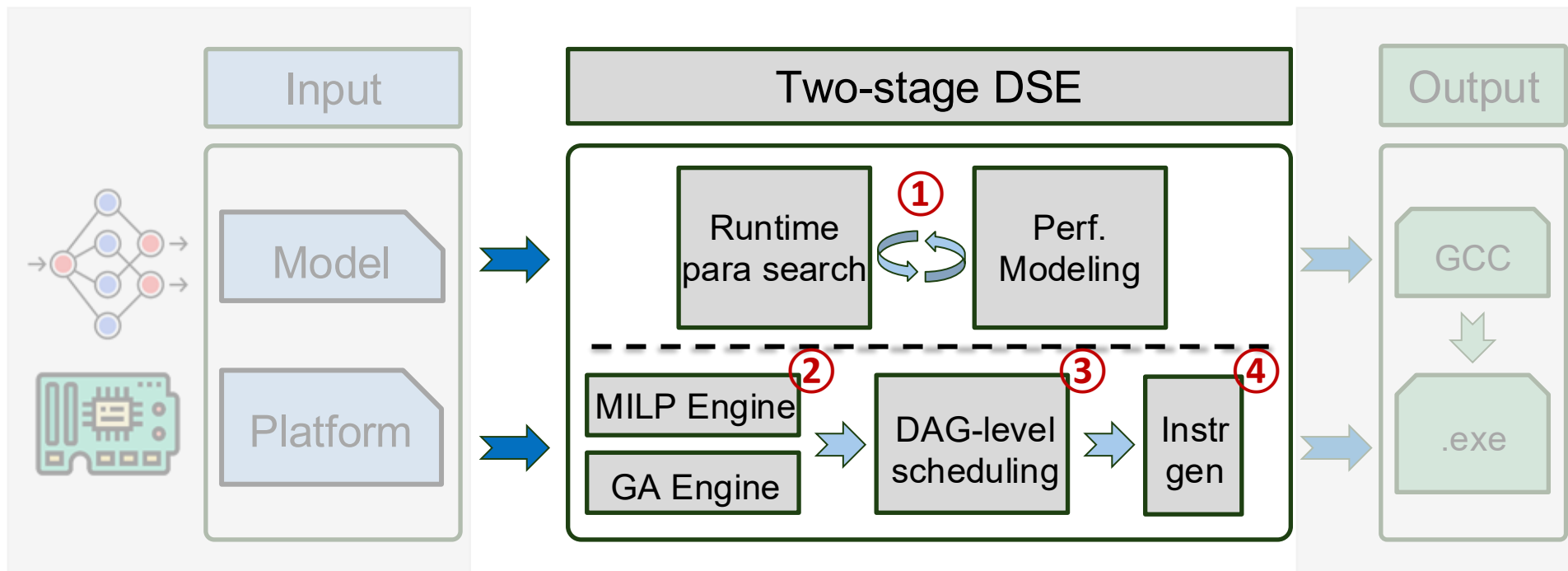
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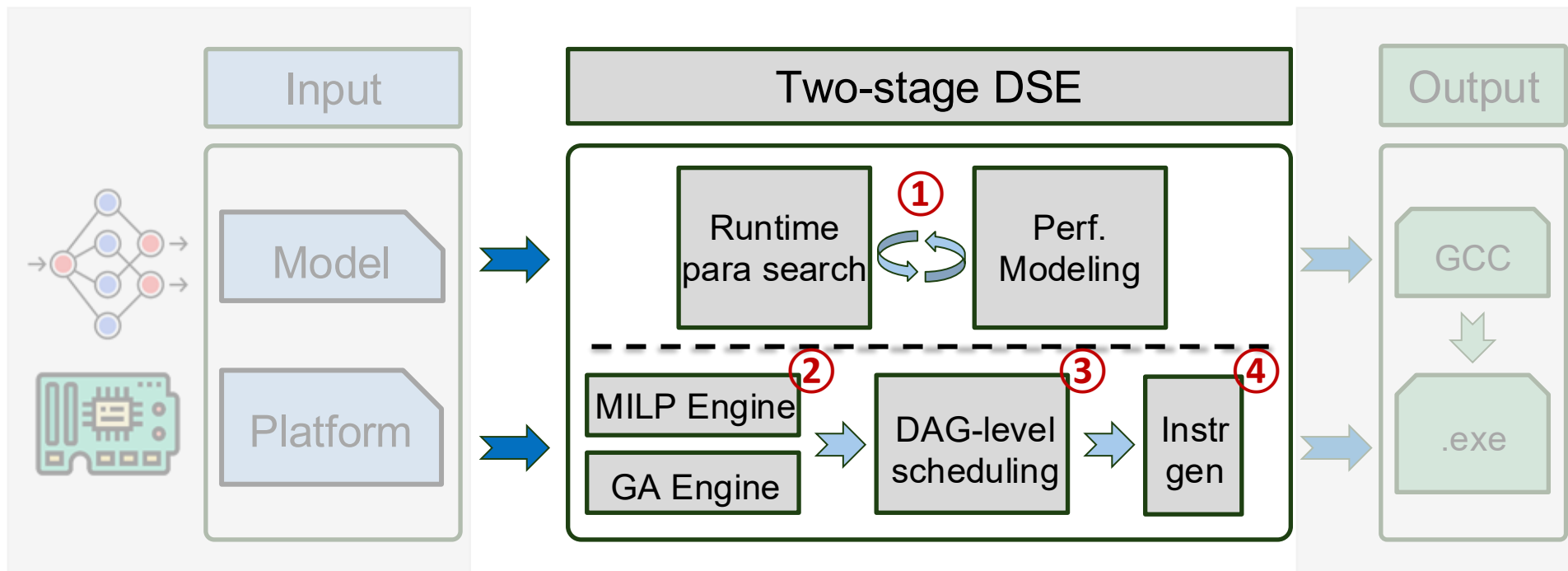
② Scheduling engines at the graph level for whole workloads.

- Mixed Integer Linear Programming (MILP) engine can guarantee the optimal scheduling.
- Genetic Algorithm (GA) engine can achieve high efficiency.



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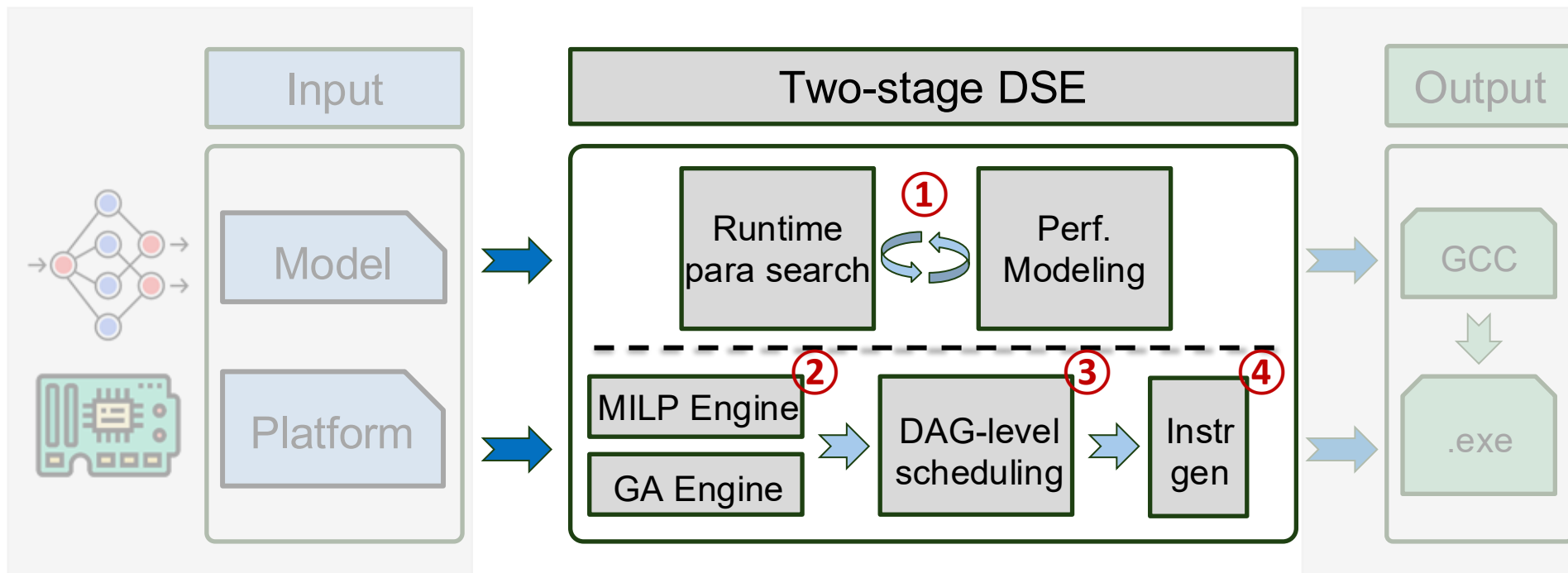
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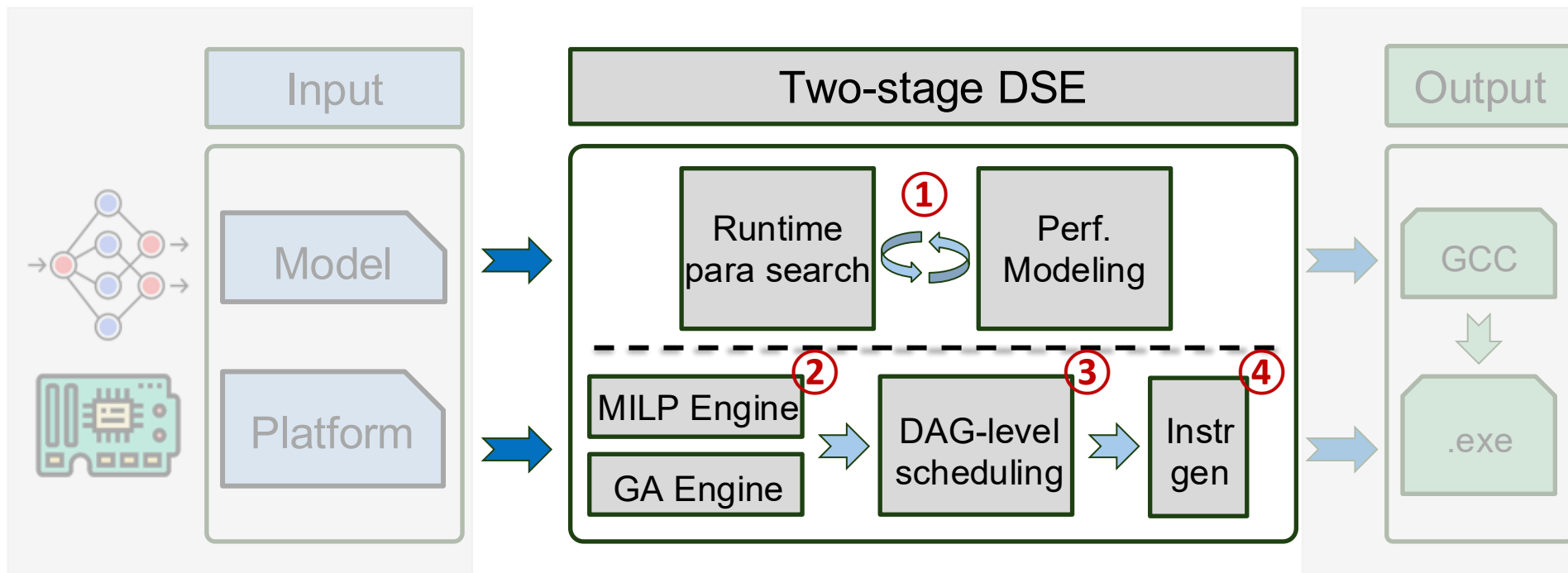
- Analyze the dataflow for each functional unit;



DORA Framework

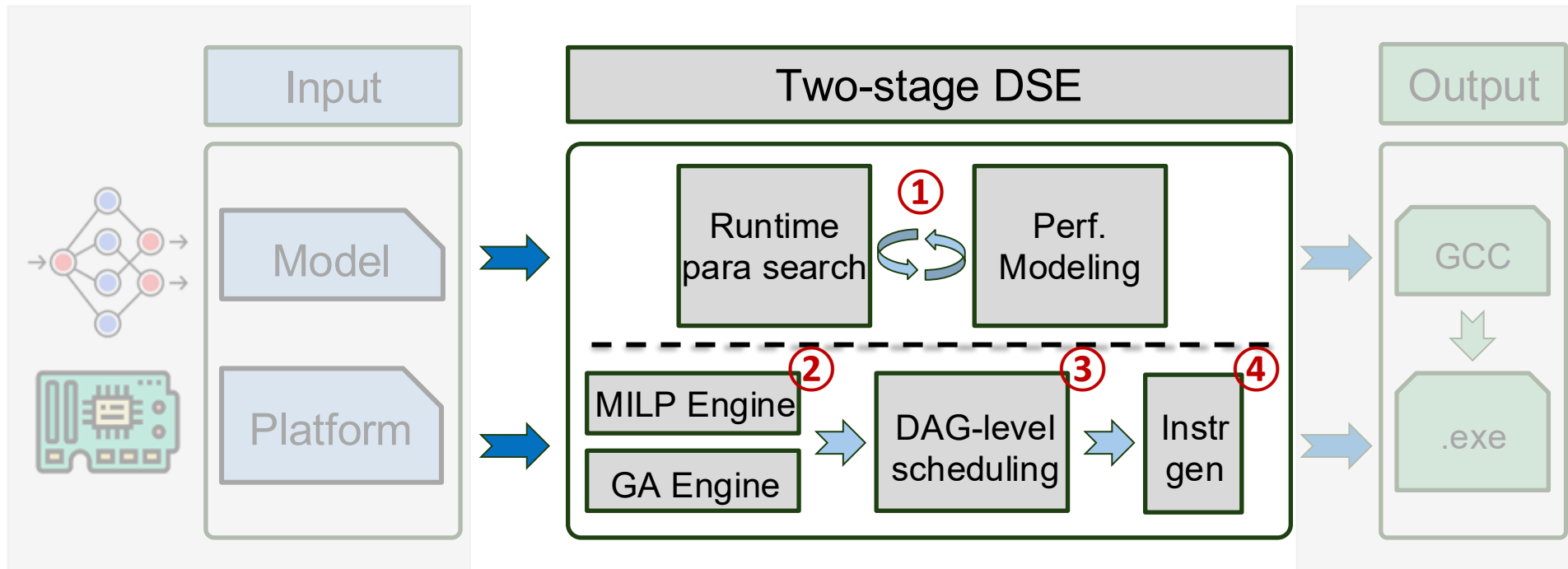
③ Generate execution flow for each functional unit.

- Analyze the dataflow for each functional unit;
- Prepare instruction sequence at FU level.



DORA Framework

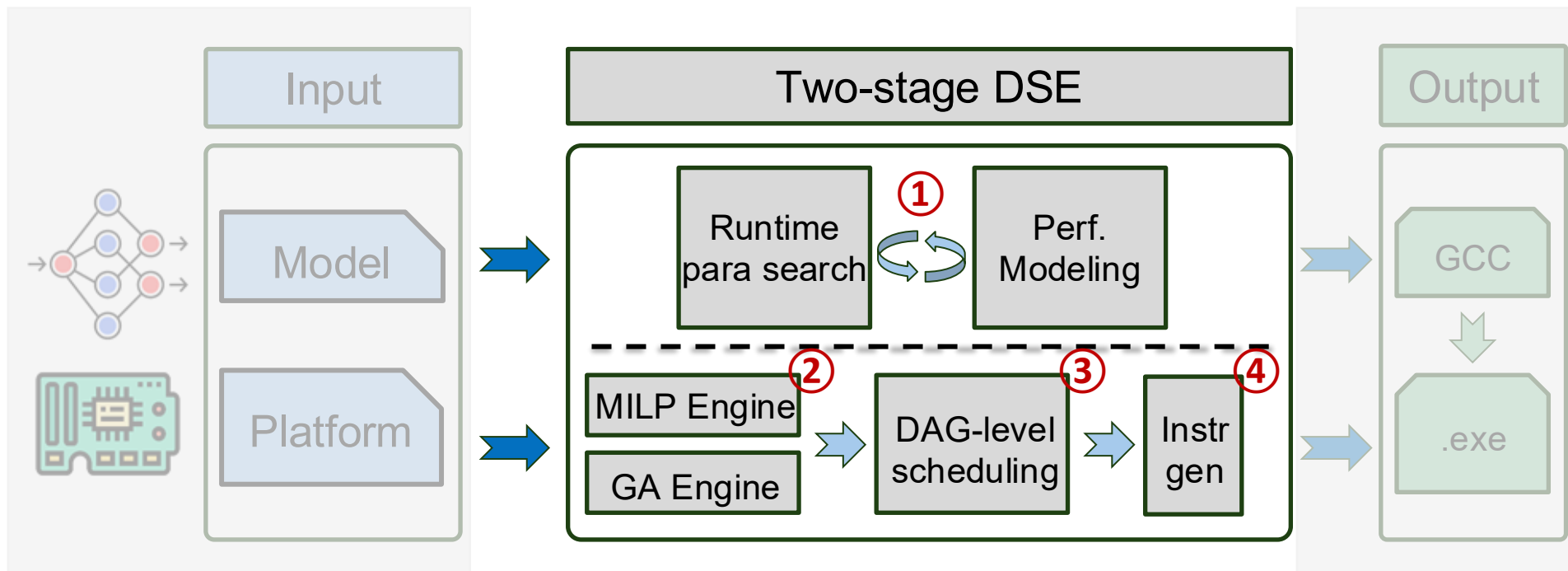
④ Instruction sequence generation.



DORA Framework

④ Instruction sequence generation.

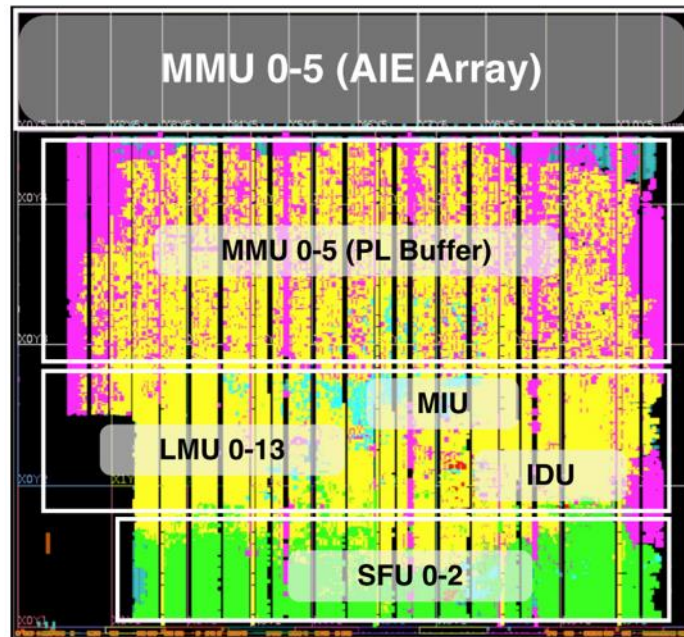
- Generate the instruction sequence and store them into host DRAM.



Experiment Setup

Experiment Setup

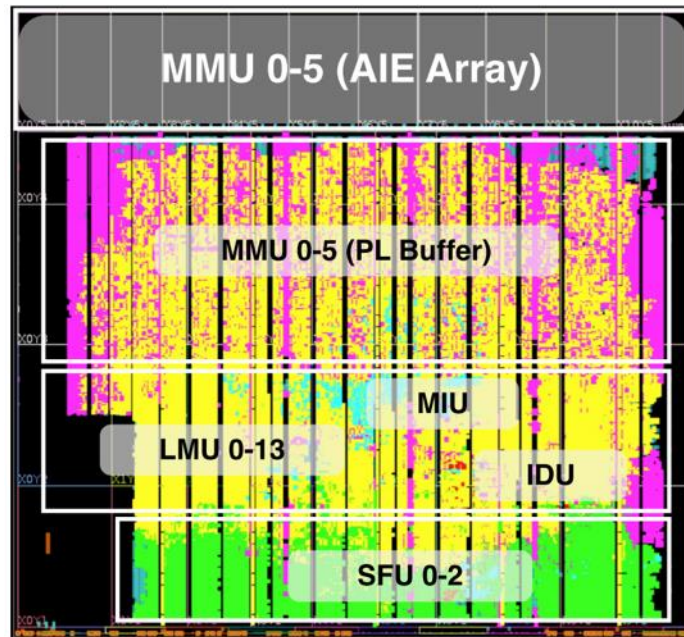
- We prototype DORA architecture on AMD Versal VCK190 platform.



DORA implementation layout

Experiment Setup

- We prototype DORA architecture on AMD Versal VCK190 platform.
- Instruction-related modules only consume REG and LUT, which are usually not bottlenecks in accelerator design.



DORA implementation layout

Unit	REG	LUTLogic	LUTMem	BRAM	URAM	DSP	AIE
MMU	110598 (6.12%)	103890 (11.52%)	0	576 (60%)	0	6 (0.3%)	384 (96%)
LMU	64358 (3.64%)	174538 (19.46%)	43358 (9.66%)	0	448 (96%)	280 (14.28%)	0
SFU	72000 (4%)	100974 (11.22%)	2328 (0.5%)	48 (4.95%)	0	96 (4.89%)	0
IDU	613 (0.03%)	415 (0.05%)	0	0	0	0	0
MIU	1810 (0.09%)	3913 (0.44%)	28 (<0.1%)	0	0	6 (0.3%)	0
Fully-con Network	264768 (14.7%)	68208 (7.58%)	0	0	0	0	0
Others	55111 (3%)	65504 (7.28%)	1419 (0.3%)	23 (2.4%)	0	0	0
Total	569258 (31.5%)	517442 (57.4%)	47133 (10.5%)	647 (67.4%)	448 (96%)	388 (19.8%)	384 (96%)

Resource utilization on VCK190

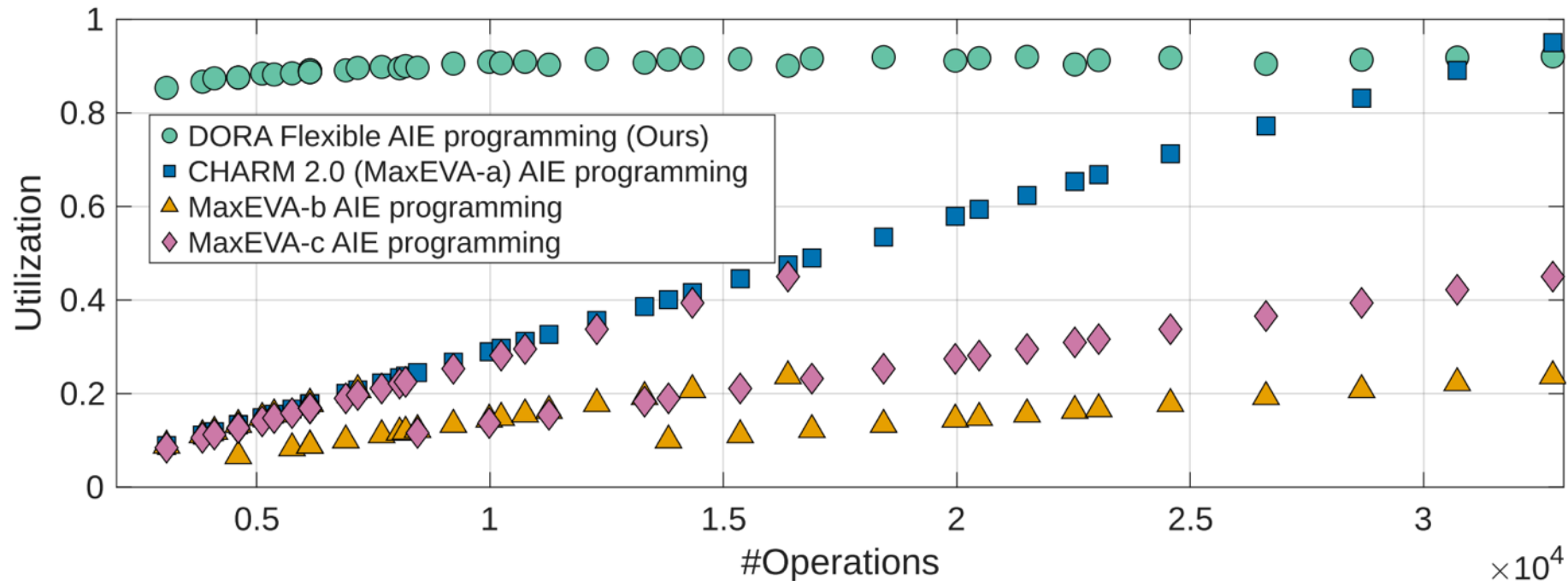
Experiment Results

Experiment Results

- **Flexibility on a single AI Engine core for diverse matrix multiplication.**

Experiment Results

- Flexibility on a single AI Engine core for diverse matrix multiplication.
- Populate MM operations with different sizes, and DORA parallelism management efficiently handles diverse MMs.



Experiments on a single AIE tile for diverse matrix multiplications.

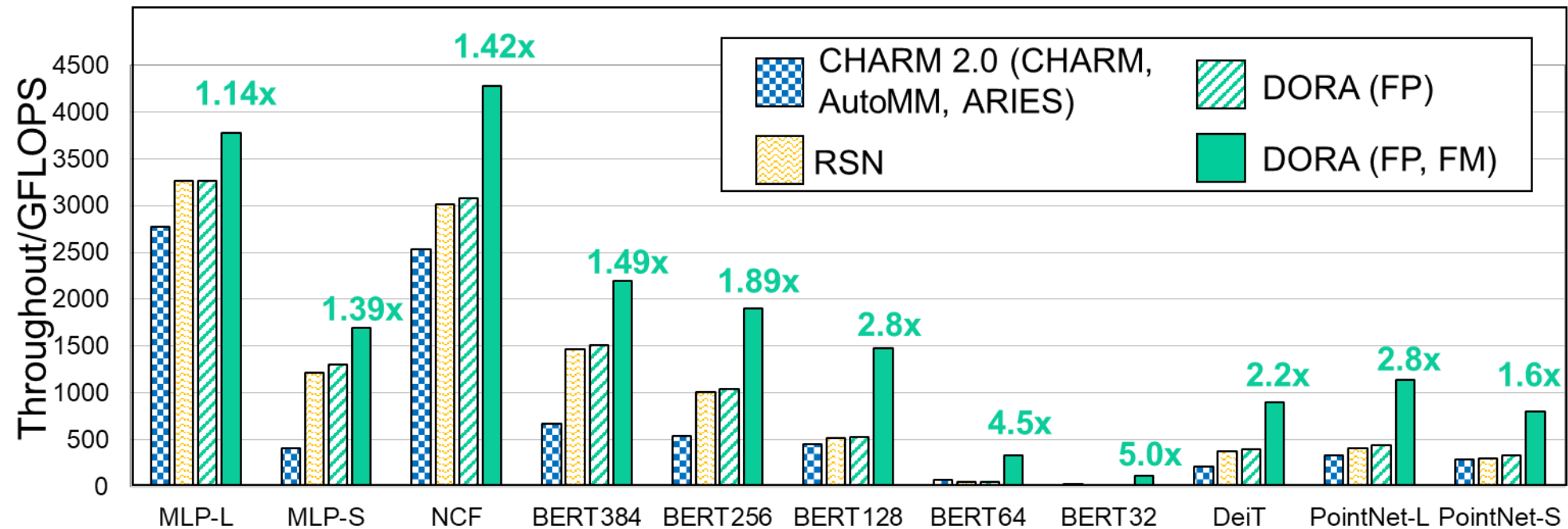
Experiment Results

Experiment Results

- Efficiency on end-to-end DNN inference.

Experiment Results

- Efficiency on end-to-end DNN inference.
- DORA always outperforms customized architecture design and existing overlay design.



Experiments on end-to-end DNN inference for diverse models.

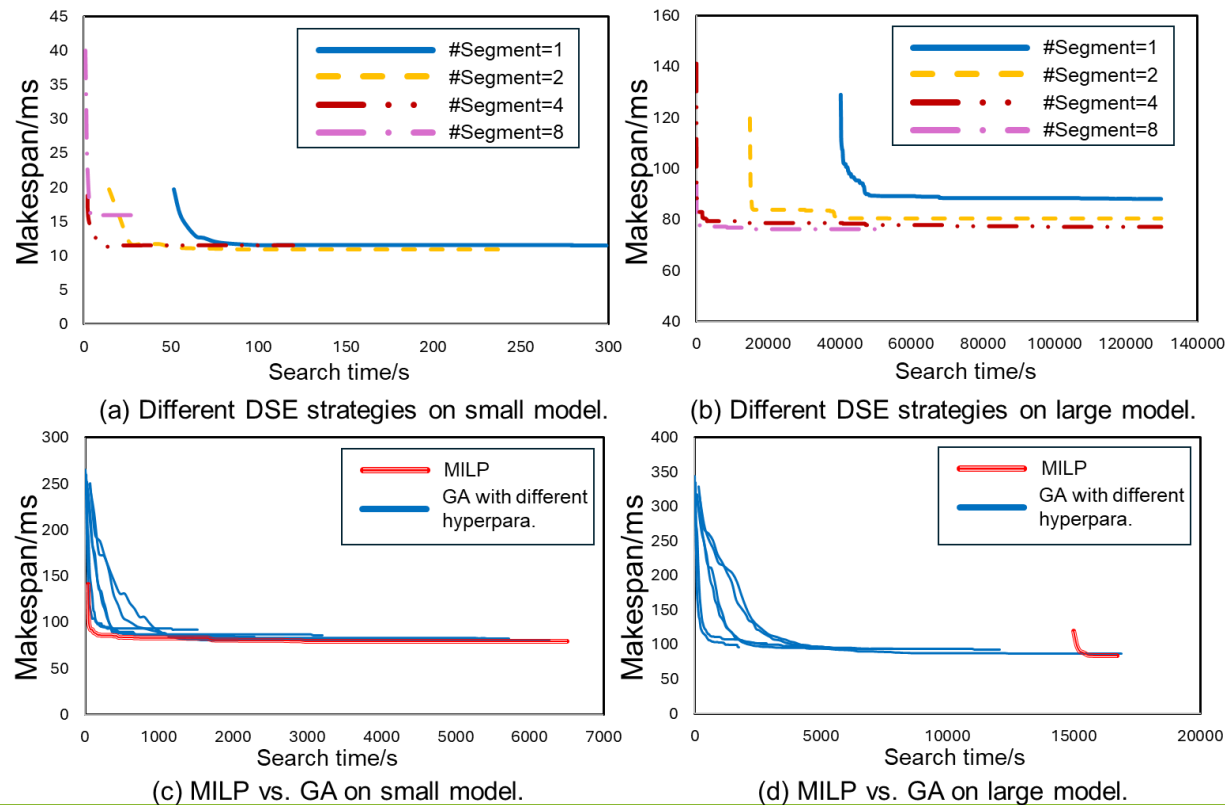
Experiment Results

Experiment Results

- **Optimality and efficiency on MILP and GA formulations.**

Experiment Results

- Optimality and efficiency on MILP and GA formulations.
 - MILP guarantees the **optimality** of medium problem size.
 - GA provides the **scalability** and **efficiency** for large problem size.



Key Takeaways

Key Takeaways

- DORA proposes a **flexible overlay architecture** that orchestrates dataflow with instructions for **diverse workloads**.
- DORA features an **automatic framework** with 2-stage DSE that can take applications and platform information to generate the **ready-to-run executable files and bitstreams**.
- DORA has a good **generality** and **scalability**, especially on large problem sizes.
- DORA open-source GitHub repo:
<https://github.com/arc-research-lab/DORA.git>

Thank You & Welcome to questions!

DORA: Dataflow-Instruction Orchestration Architecture for DNN Acceleration

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