

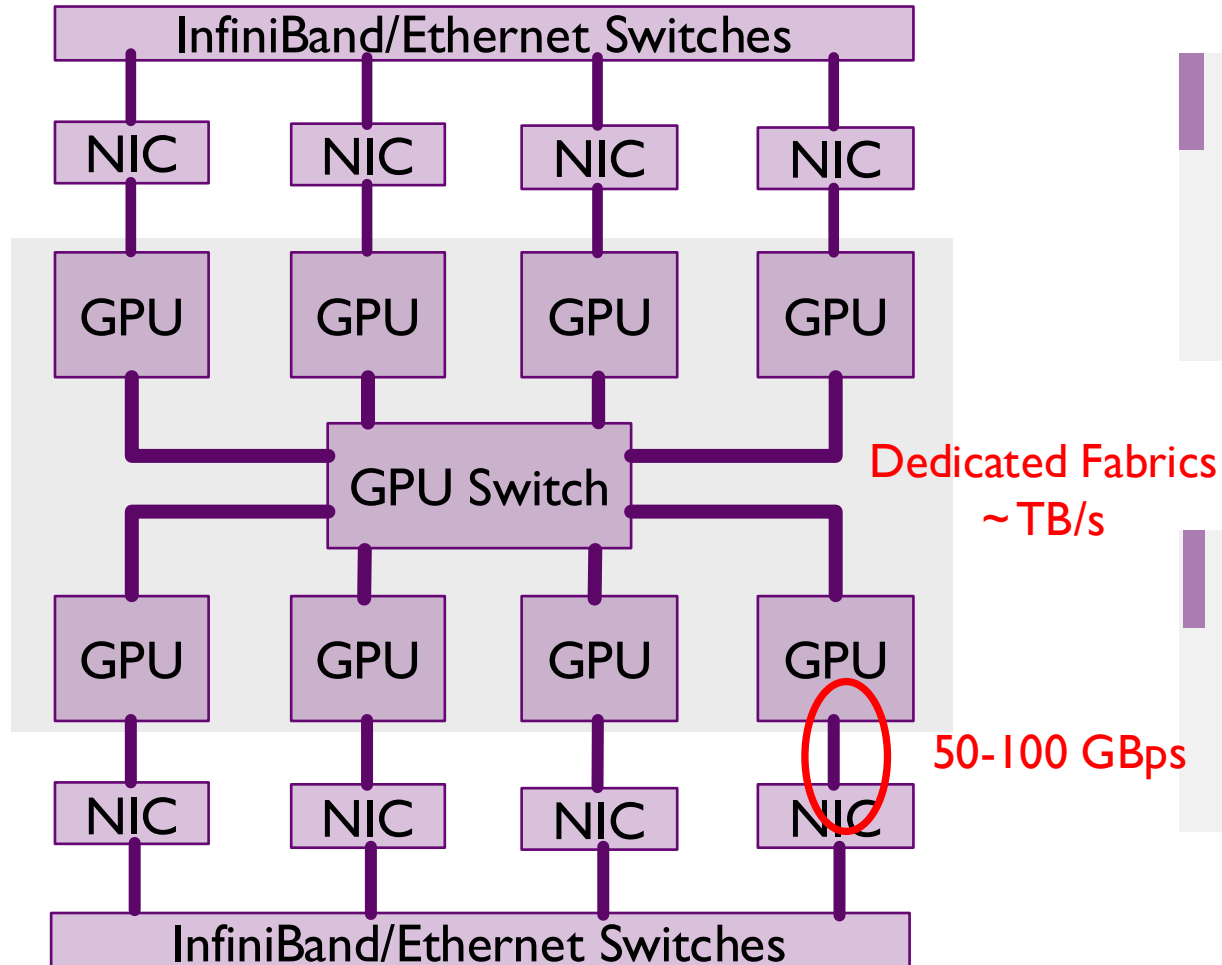
19th USENIX Symposium on Operating Systems Design and Implementation

Enabling Efficient GPU Communication over Multiple NICs with FuseLink

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Advancement of high-bandwidth GPU communication



Dedicated interconnect

- Relatively small groups
- Fast growing bandwidth

Network over general NICs

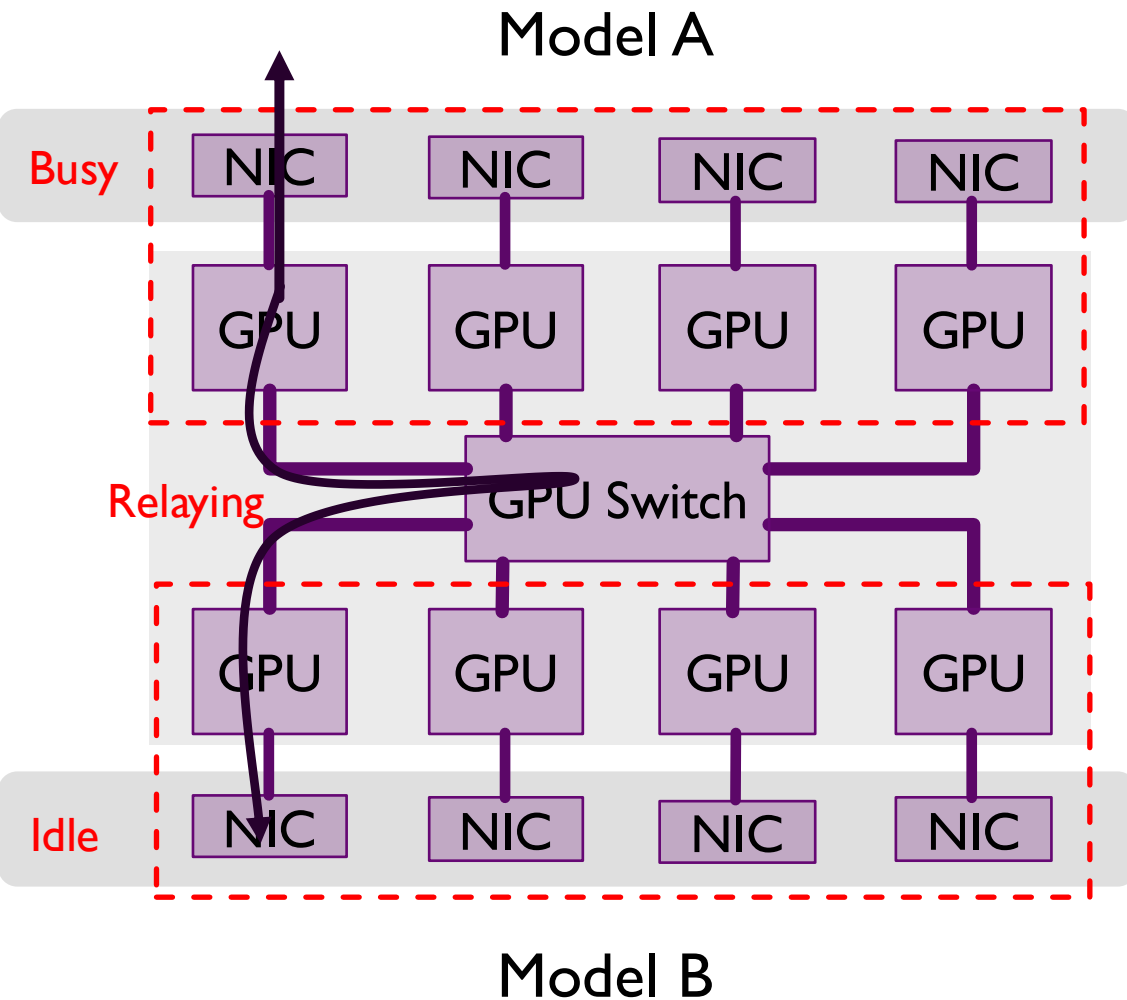
- Large scale
- Slow growing bandwidth

Can be tens of GB for every GPU!
KV Cache/Embeddings/Gradients,...

When can we multiplex the NICs and to accelerate inter-server communication?

With PCIe limitation, can we leverage **dedicated fabrics** to improve bandwidth over **general NICs**?

Observation: traffic imbalance is common



The GPU server has two independent LLM serving deployments with inter-server context (KVCache) transmission.

- Models serve different requests
- Half NICs busy and half idle.

Opportunity: multi-NIC acceleration under imbalanced traffic

Biased Traffic Volume

NICs with less traffic
being idle

Using indirect NICs

Data traverse via
CPU/UPI becomes
bottleneck

Delayed
Communication

Late GPUs increase
transmission tail

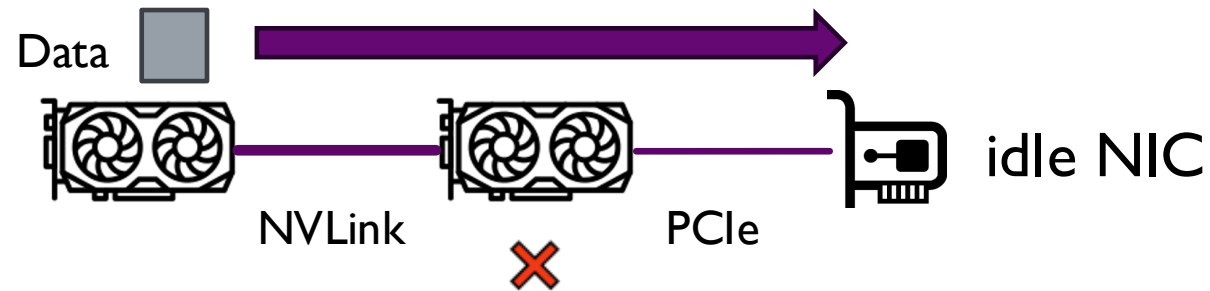
Suboptimal NICs throughput
Accelerate via multi-NIC transmission

Can arbitrary ML workloads effectively leverage all available NICs, aggregating links as a "FuseLink" for inter-server communication ?

Design goals

- Efficiently use idle NICs through dedicated intra-server fabrics
- Seamless integration into existing systems (NCCL, Gloo,...)
- Avoid contention & interruption among GPUs

Challenge: incompatibility of dedicated fabrics and NICs



TX Direction

Hardware-offloaded NIC network stack cannot transmit data across GPU interconnects

RX Direction

NIC traffic cannot be modified to be routed across GPUs

Efficient indirect NIC communication: memory & network combined

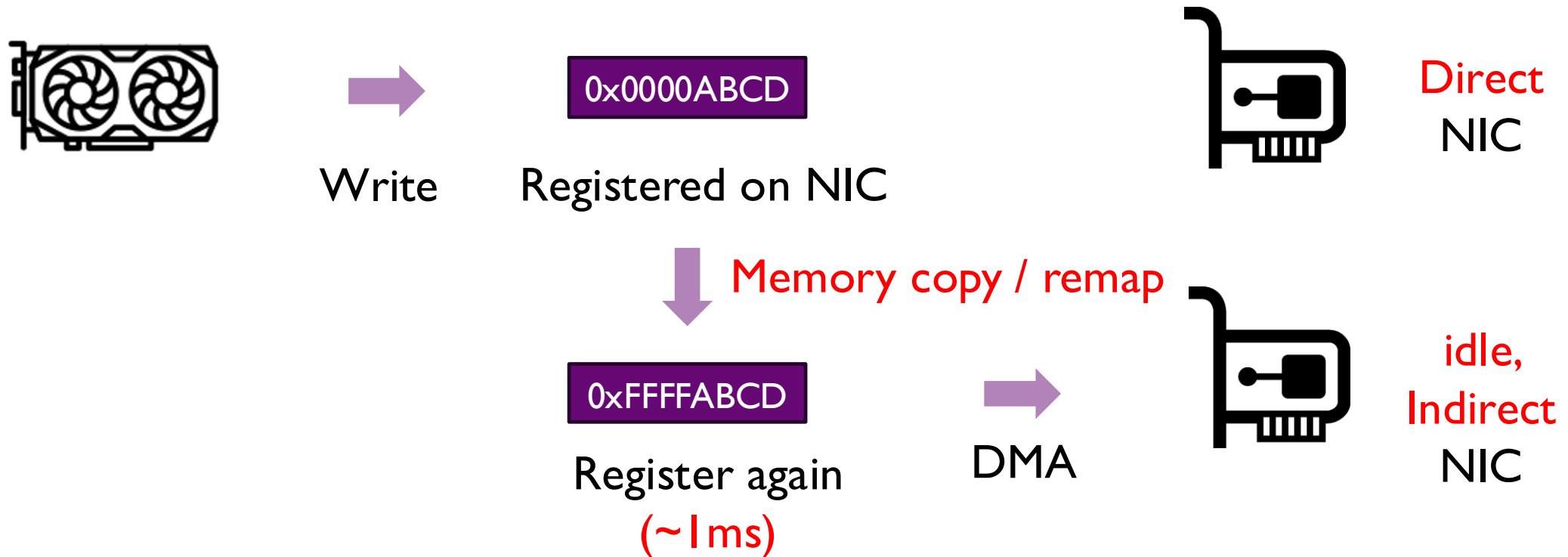


GPU program fills network buffer registered on NICs

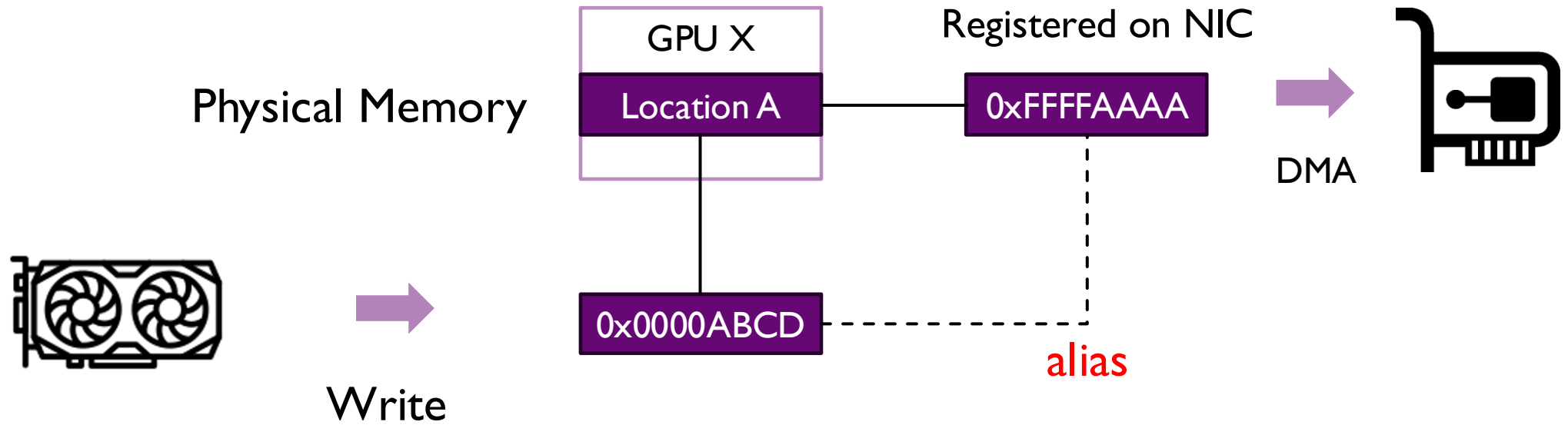
NIC reads registered (pinned) memory through PCIe

Efficient indirect NIC communication: memory & network combined

Straight forward solution: **copy data** to intermediate GPU close to the NIC
Not performant because of **extra copy and frequent CPU interruption**

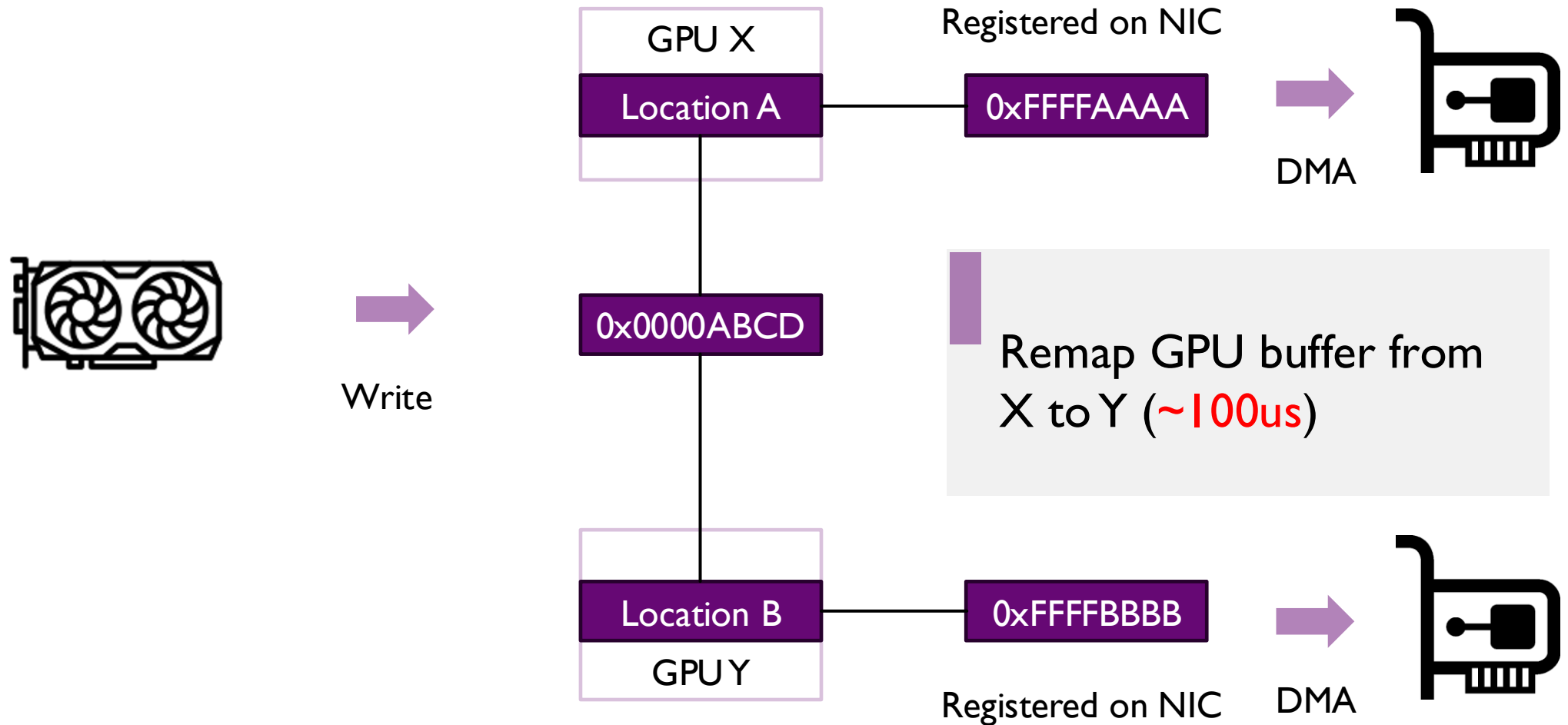


Efficient indirect NIC communication: memory & network combined

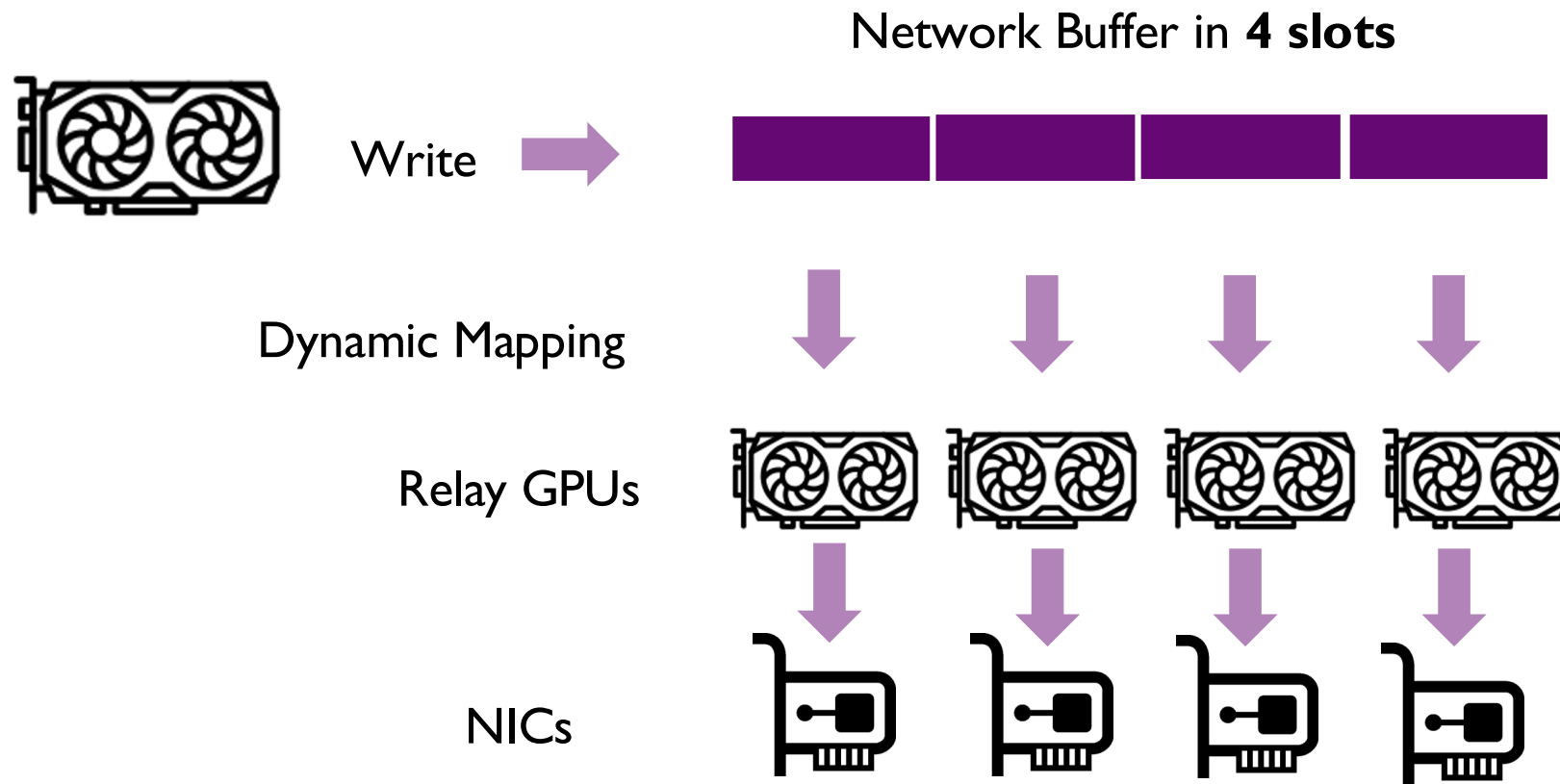


GPUs and NICs has different memory address with the same backend (**alias**)
No need to change memory registration (~1ms)

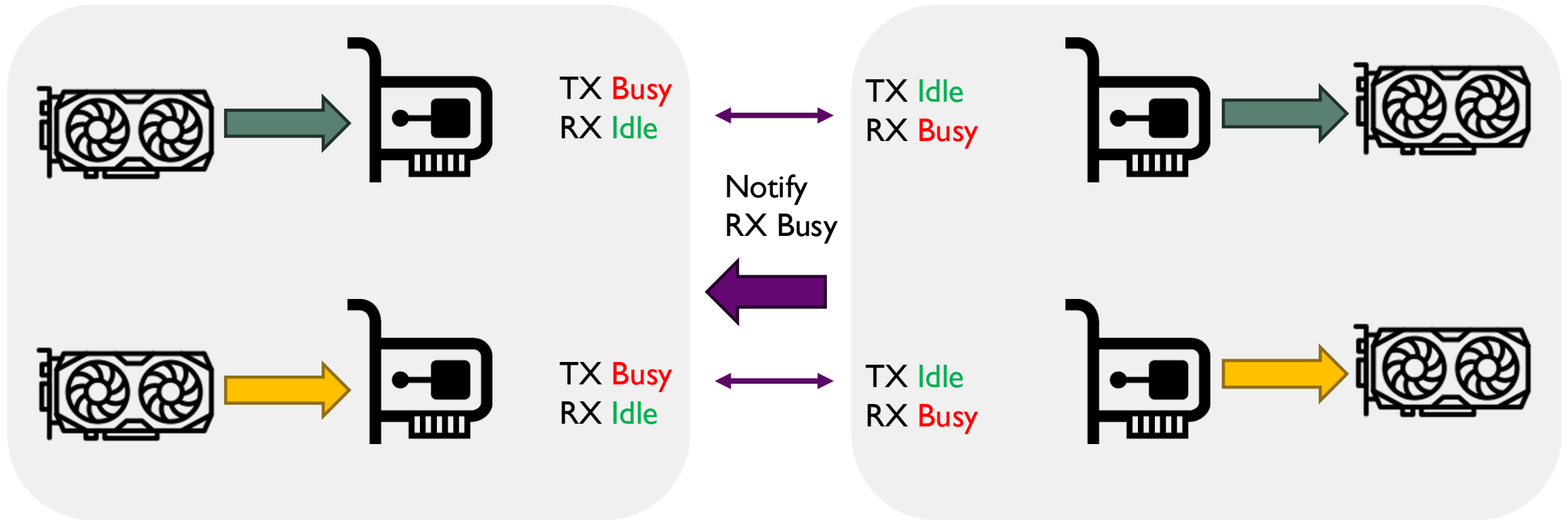
Efficient indirect NIC communication: memory & network combined



Efficient indirect NIC communication: memory & network combined

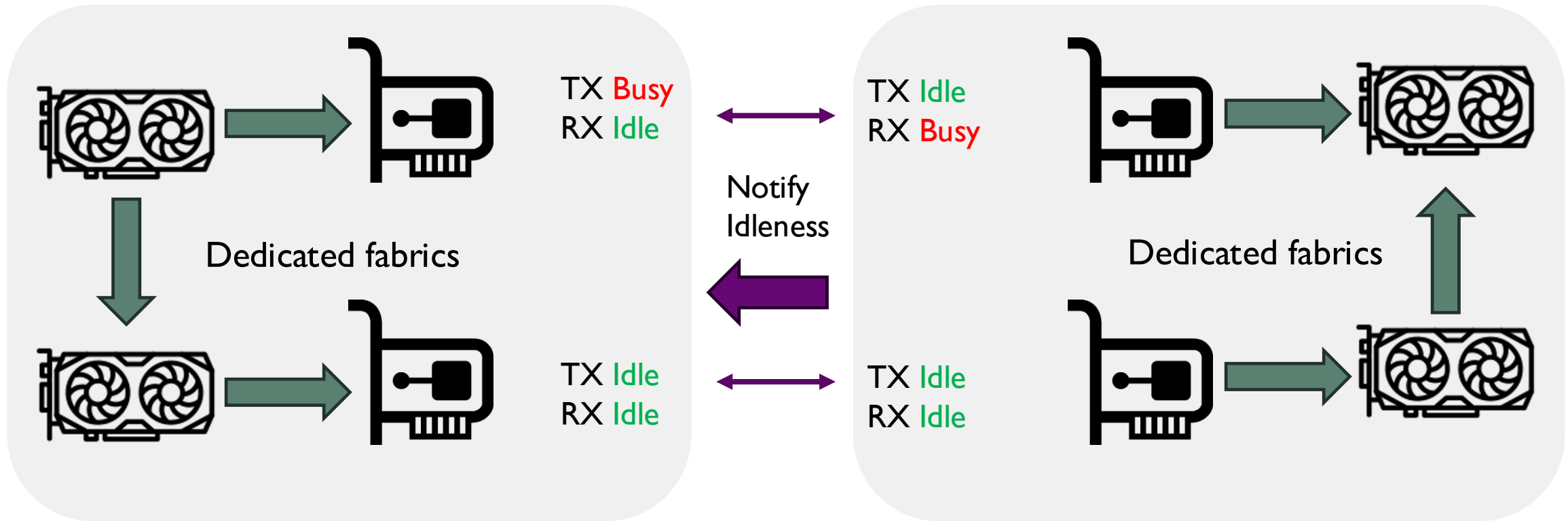


Dynamic load balance without contention & interruption



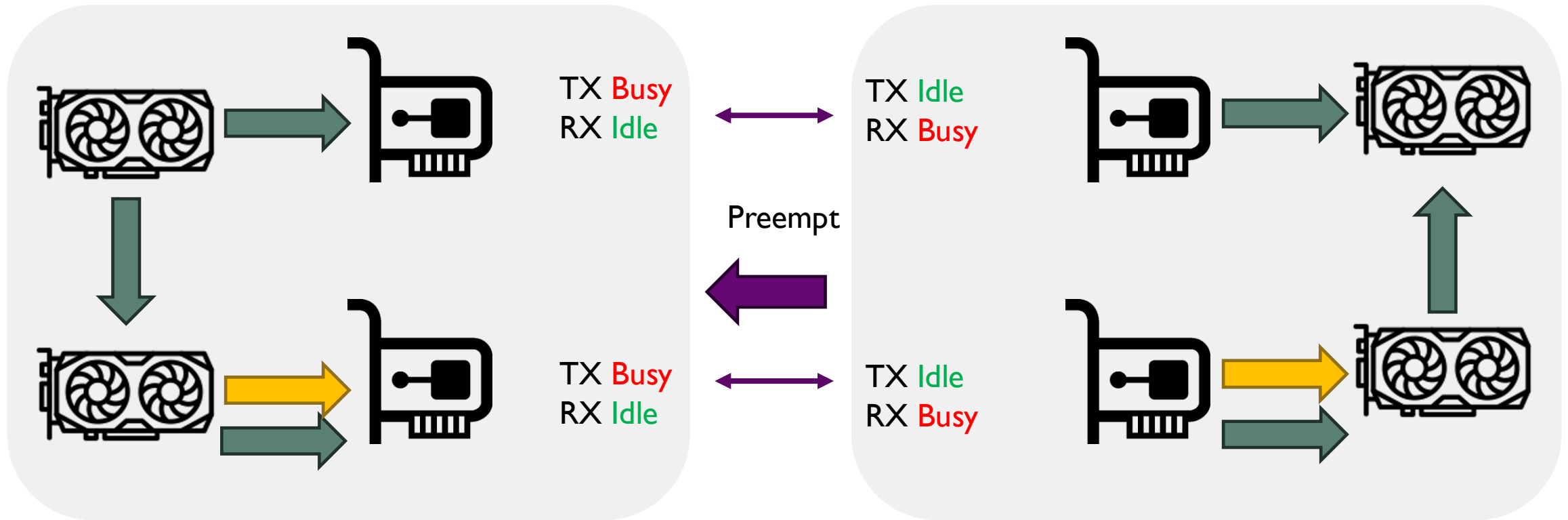
- Record NIC TX/RX workload status **within each server**
- Exchange TX/RX status, **agree** on NIC selection

Dynamic load balance without contention & interruption



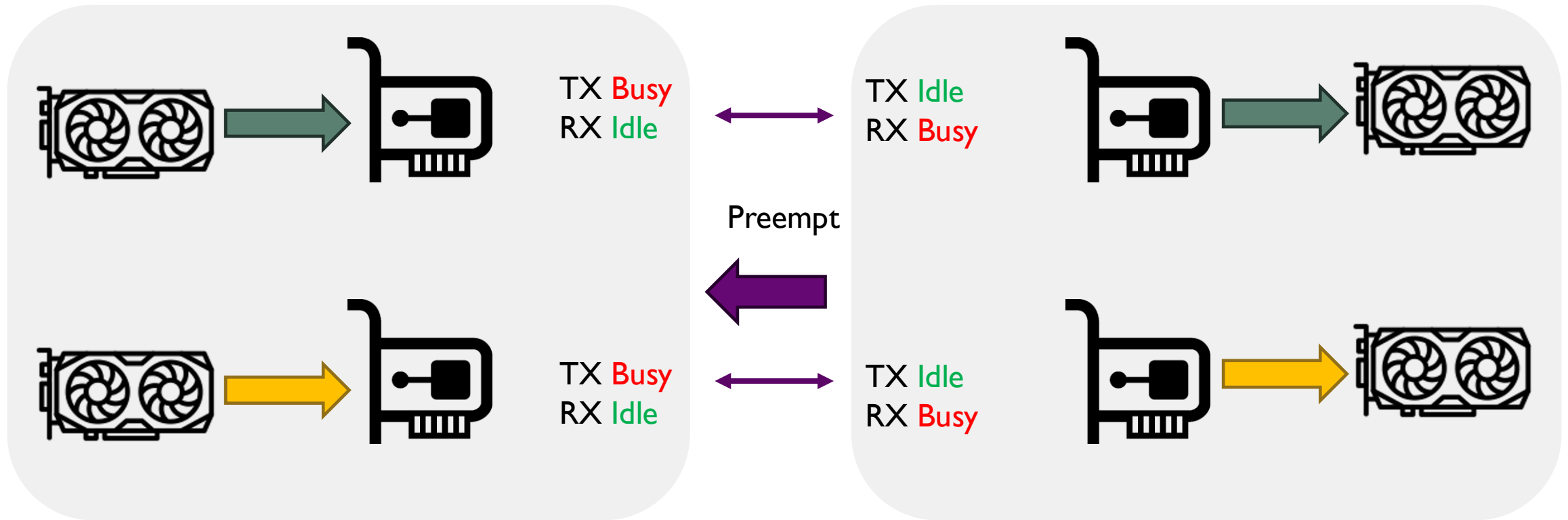
- RX & TX being idle
- Using two NICs through intra-server relaying

Dynamic load balance without contention & interruption



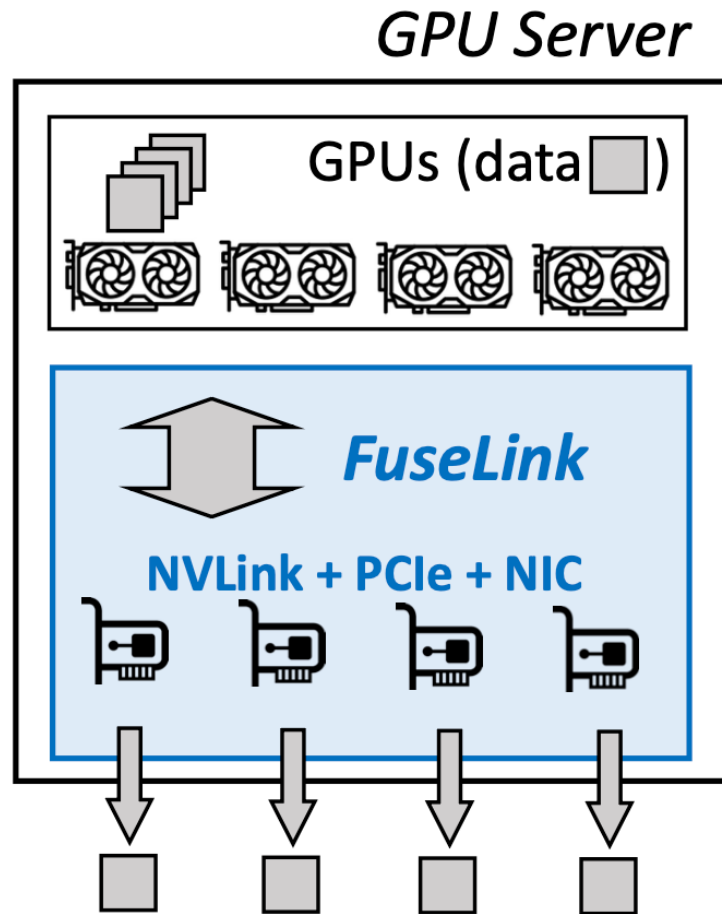
- NIC contention detected
- Preempt relay traffic and fallback to direct NIC

Dynamic load balance without contention & interruption



- NIC contention detected
- Preempt relay traffic and fallback to direct NIC

System Overview

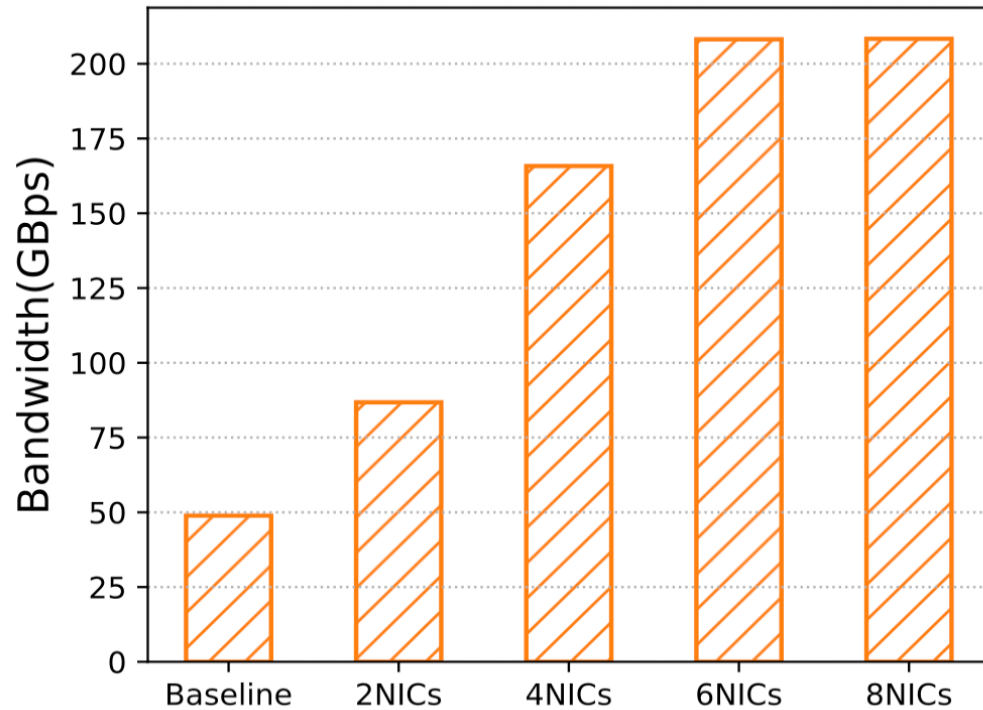


FuseLink system components

- **NIC idleness detection:** mark NICs as idle when nothing is being sent/received.
- **NVLink + NIC transport:** send traffic to intermediate GPUs and transmit through idle NICs efficiently.
- **Contention elimination:** ensure GPUs fully occupy direct NICs during communication.

Evaluation: FuseLink bandwidth over NVLink + NIC

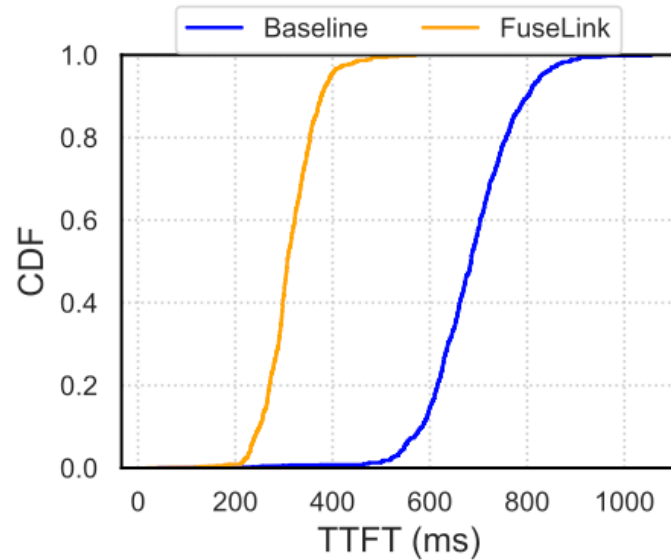
Each server: eight GPU, eight-lane NVLink (~160 GBps) and 400 Gbps (50 GBps) NICs.



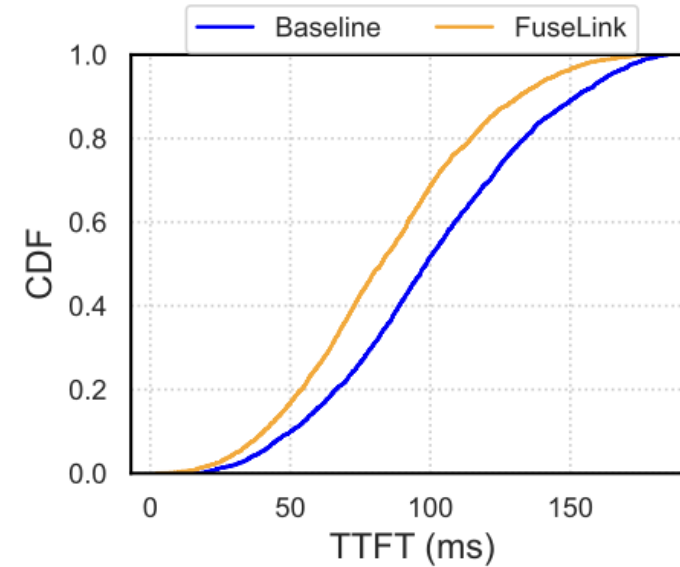
FuseLink bandwidth for **two** GPUs using **different number of NICs**

Bandwidth reaches limit when both **direct NIC** and **NVLink** are fully utilized.

Evaluation: LLM serving of independent instances



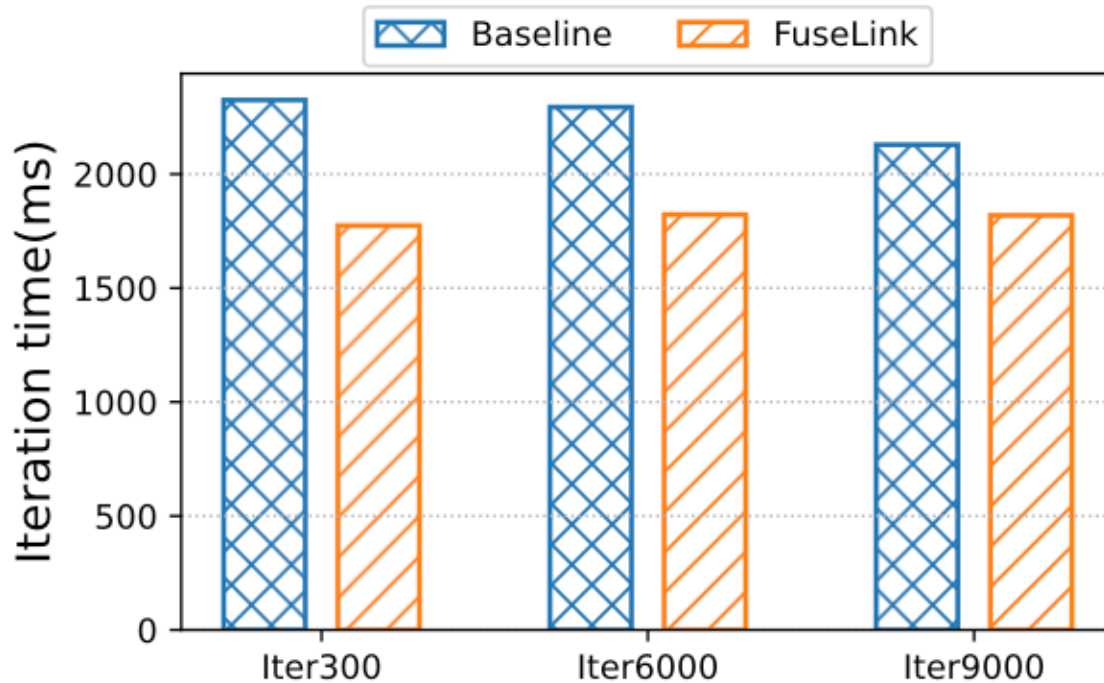
Eight serving instances



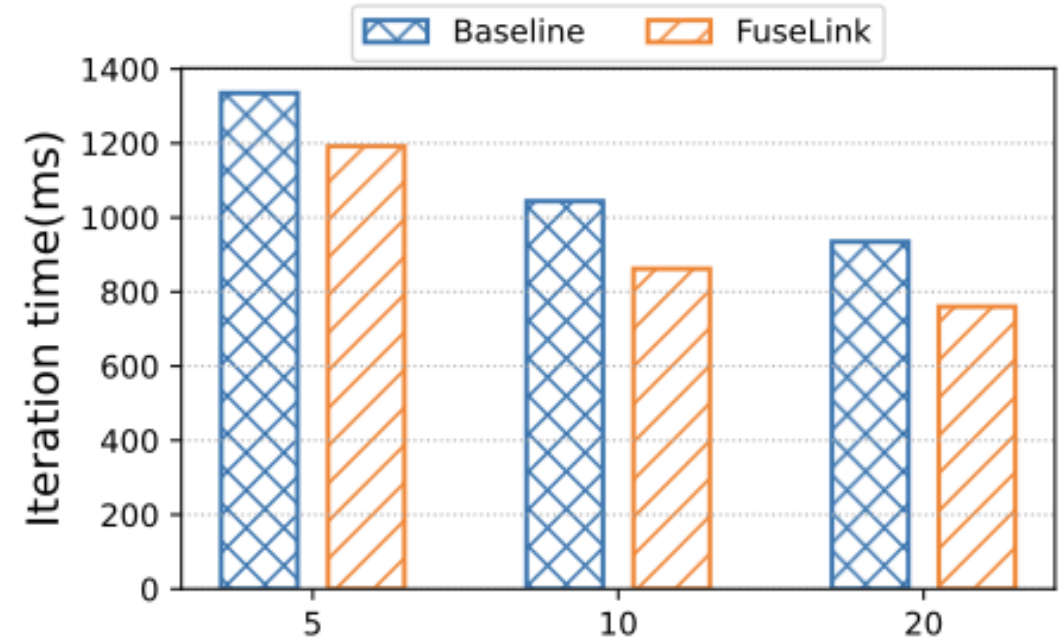
Two serving instances

Improvement: accelerated data transfer & reduced waiting time

Evaluation: expert-parallelism and imbalanced embedding transmission



Each server has two expert shards
Accelerating **imbalanced all-to-all** in
Mixtral 8x22B expert-parallel training



Accelerating **imbalanced embedding transmission** when training DLRM

Limitations and future works

- Applicable to other GPUs?
Yes, only need P2P memory access & virtual memory mapping feature.
- Fine-grained load balancing?
Per-chunk load balancing to per-packet load balancing.



Thank You!

Code: <https://github.com/axio-project/FuseLink>

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