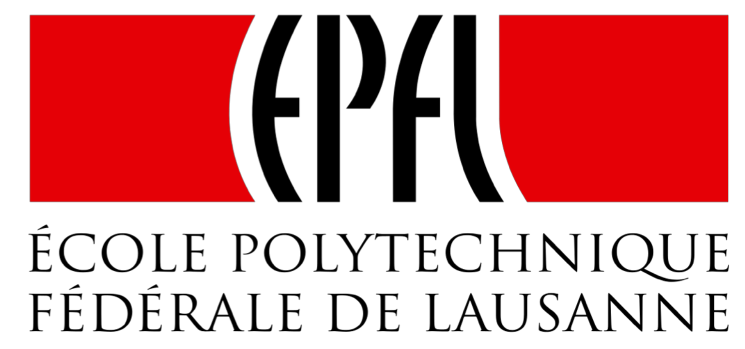
Size-aware sharding for improving tail latencies in in-memory key-value stores

Diego Didona (EPFL), Willy Zwaenepoel (University of Sydney)



Contributions

(1) **Size-aware sharding**

Improve tail latencies of in-memory key-value stores with heterogeneous item sizes

(2) **Minos in-memory key-value store**

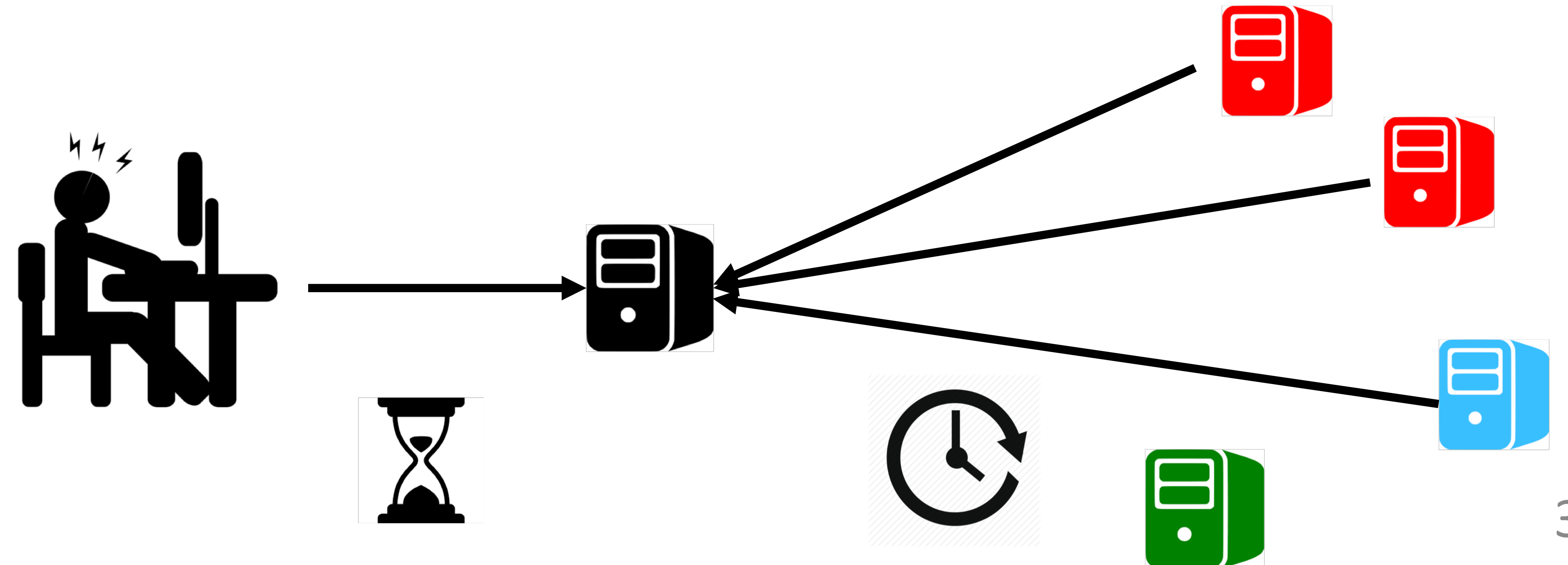
Order-of-magnitude lower 99th percentile latency

BACKGROUND

Tail latencies in high fan-out applications

Slowest reply determines request latency

SLO: N-th percentile of resp. time < X

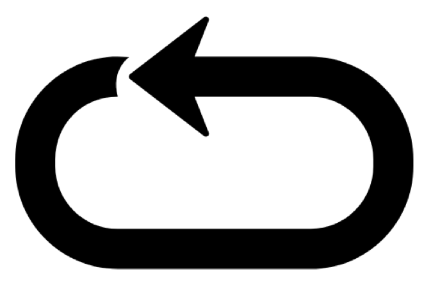


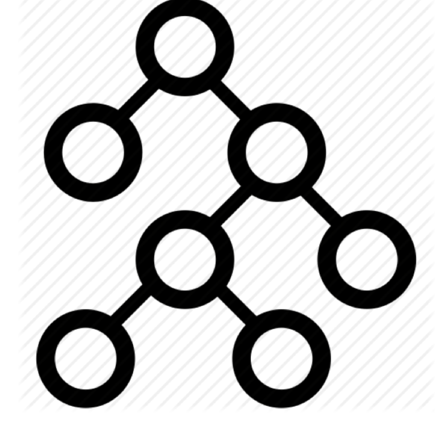
In-memory key-value stores (KV)

- Widespread solution to deliver low latency

- Caches / non-persistent data repositories

State-of-the-art KVs: design





High-bandwidth, multi-queue NICs
+
Kernel-bypassing network stacks

Run-to-completion model
+
Ad hoc data structures and CC

State-of-the-art KVs: performance

µsec-scale latencies @ several Mops/sec

State-of-the-art KVs: performance

μsec-scale latencies @ several Mops/sec

But high tail latencies with heterogeneous item sizes

Heterogeneous item sizes are common

Facebook [SIGMETRICS12]

Wikipedia [ISCA13]

Flickr [ISCA13]

Memcachier [NSDI19]

Heterogeneous item sizes are common

Facebook [SIGMETRICS12]

Wikipedia [ISCA13]

Flickr [ISCA13]

Memcachier [NSDI19]

Heavy tail: few large requests but very costly

Why high tail latencies?

1. Head-of-line blocking

2. Convoy effect

Head-of-line blocking

Small requests enqueued behind a large

Convoy effect

Burst of large requests may take most (or all) cores

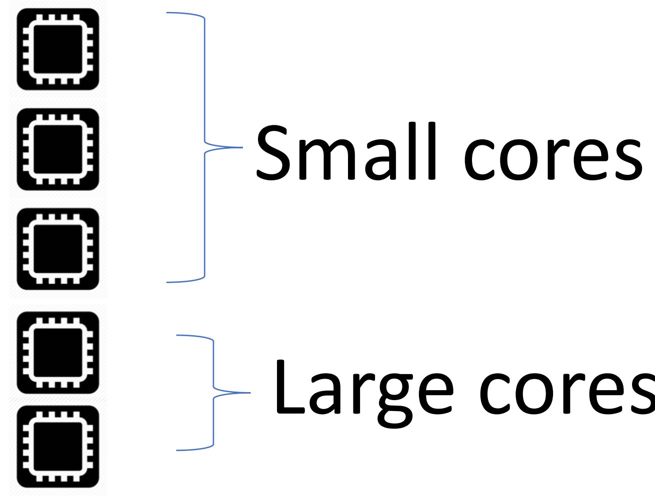
SIZE-AWARE SHARDING

Size-aware sharding

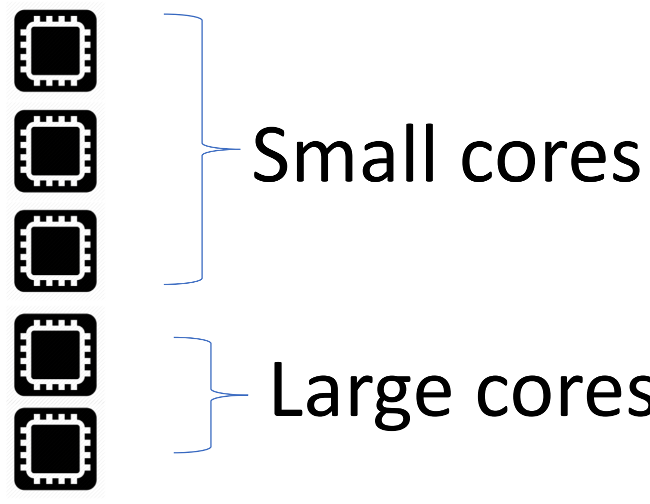
1. Small and large requests on disjoint sets of cores
   ➔Avoid head-of-line blocking

2. Reserve some cores for small requests
   ➔Avoid convoy effect

Size-aware sharding in operation



Size-aware sharding in operation



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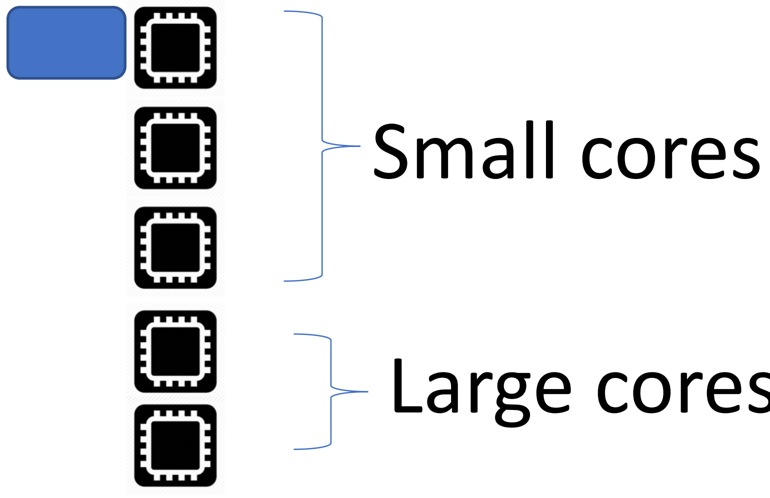

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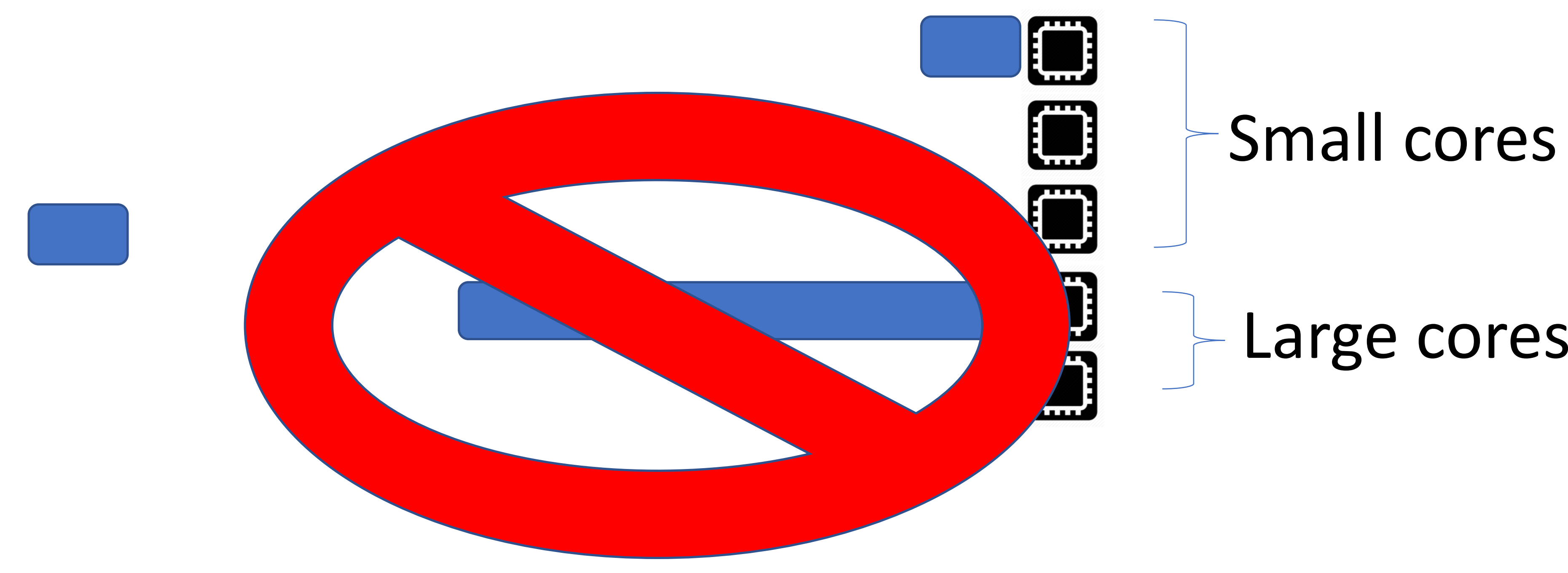

Size-aware sharding in operation

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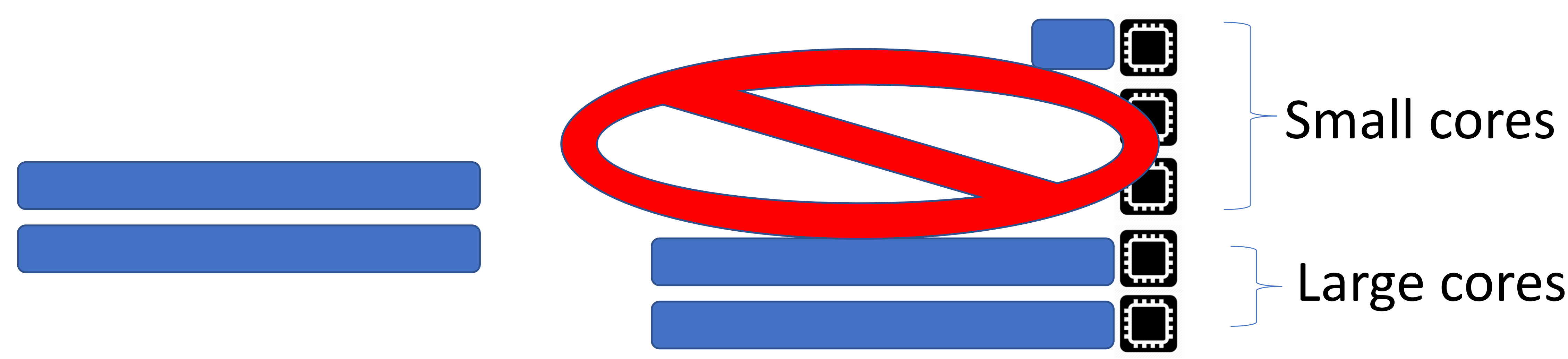

Size-aware sharding in operation

Trade-off: large requests take longer

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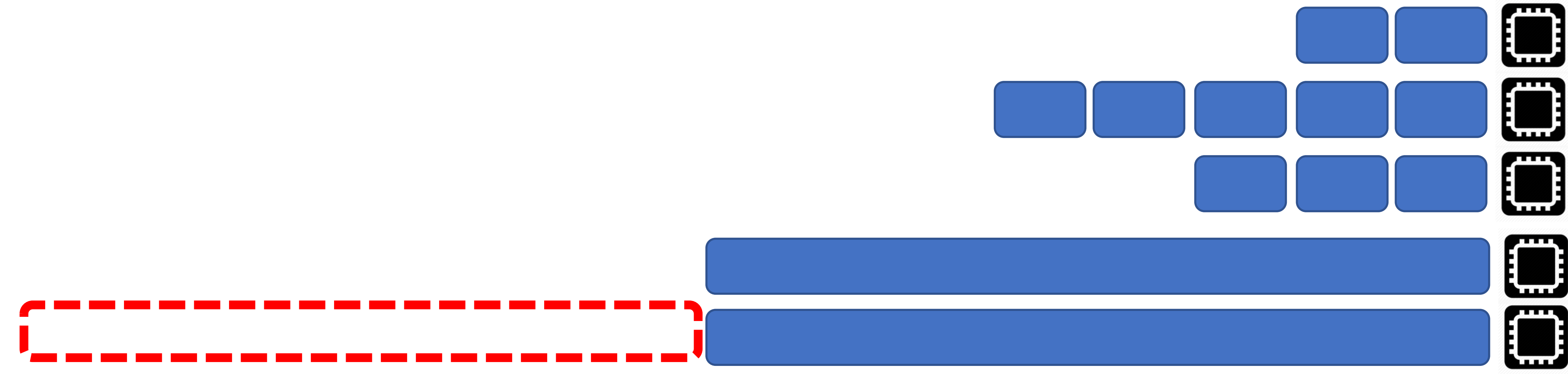
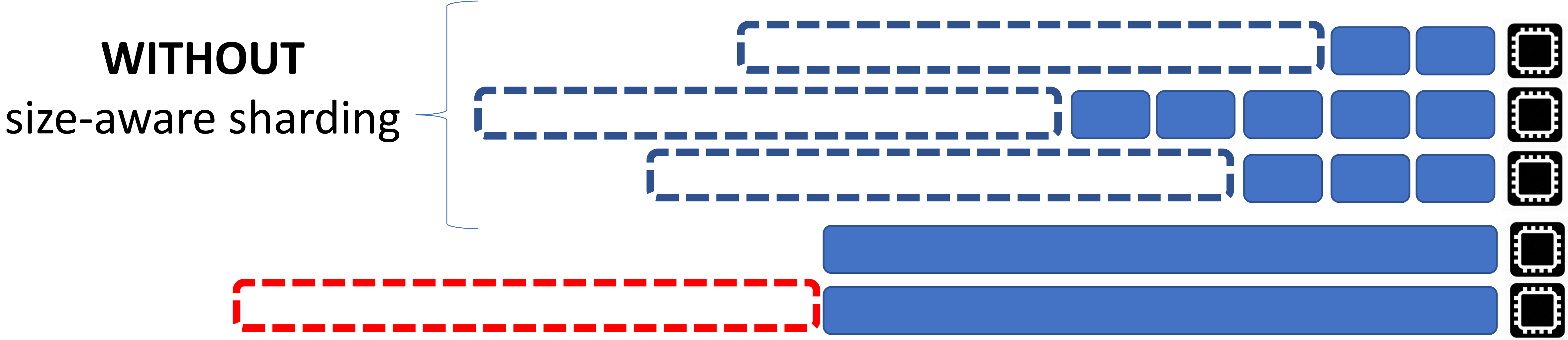


Trade-off: large requests take longer



MINOS IN-MEMORY KV

Implementation

- Single-node, PUT-GET
- Commodity hardware
- No data durability

Minos design challenges

Small vs large threshold

Core partitioning

Request dispatch

Minos design challenges

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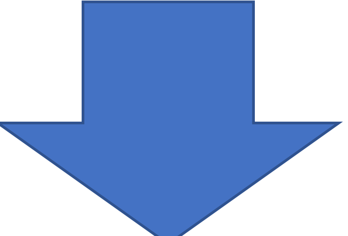
Main insight

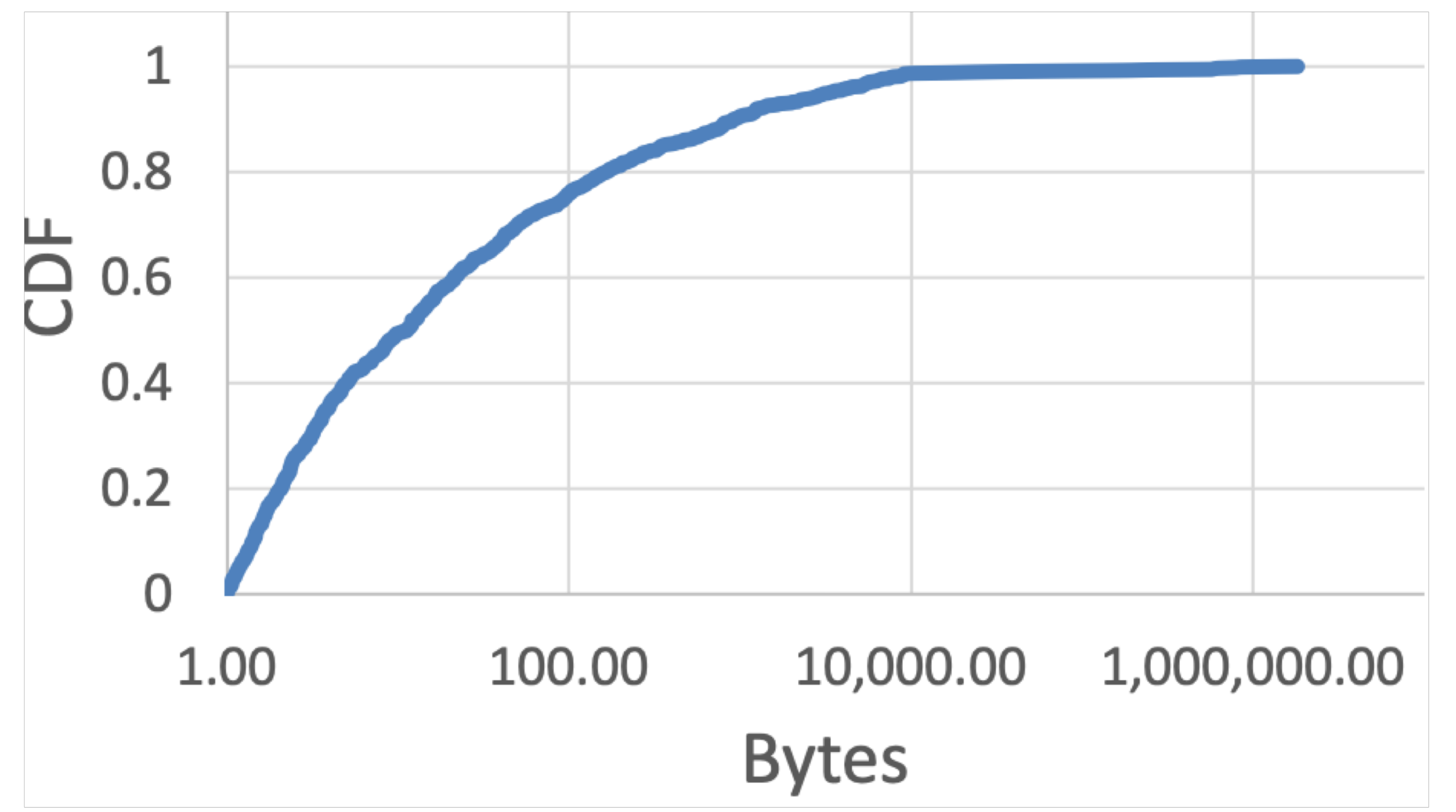
Improve **N**-th percentile of latencies



Improve latencies of **N**% smallest requests

Example with 99th percentile

Obtain at runtime the CDF of the sizes of accessed items



Example with 99th percentile

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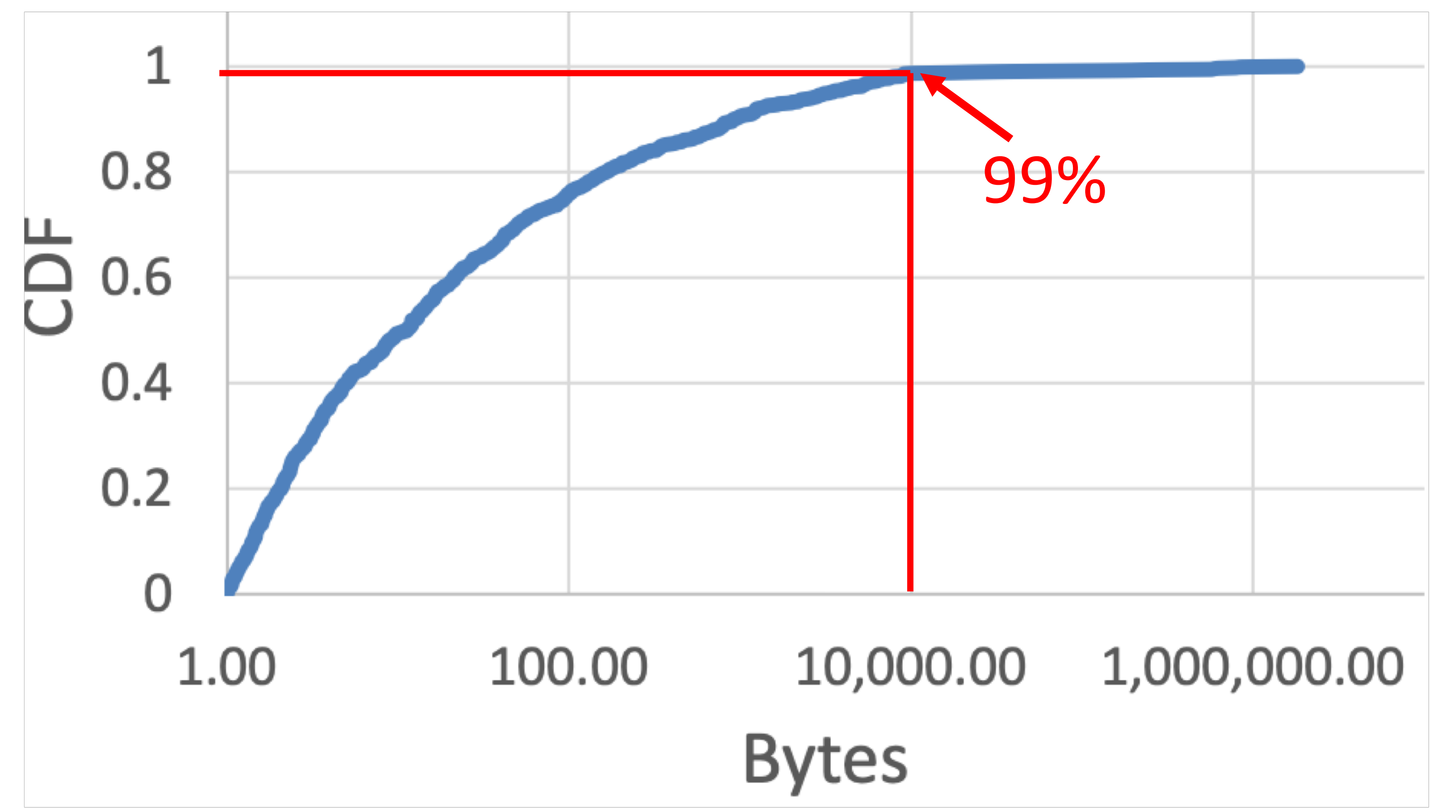

Example with 99th percentile

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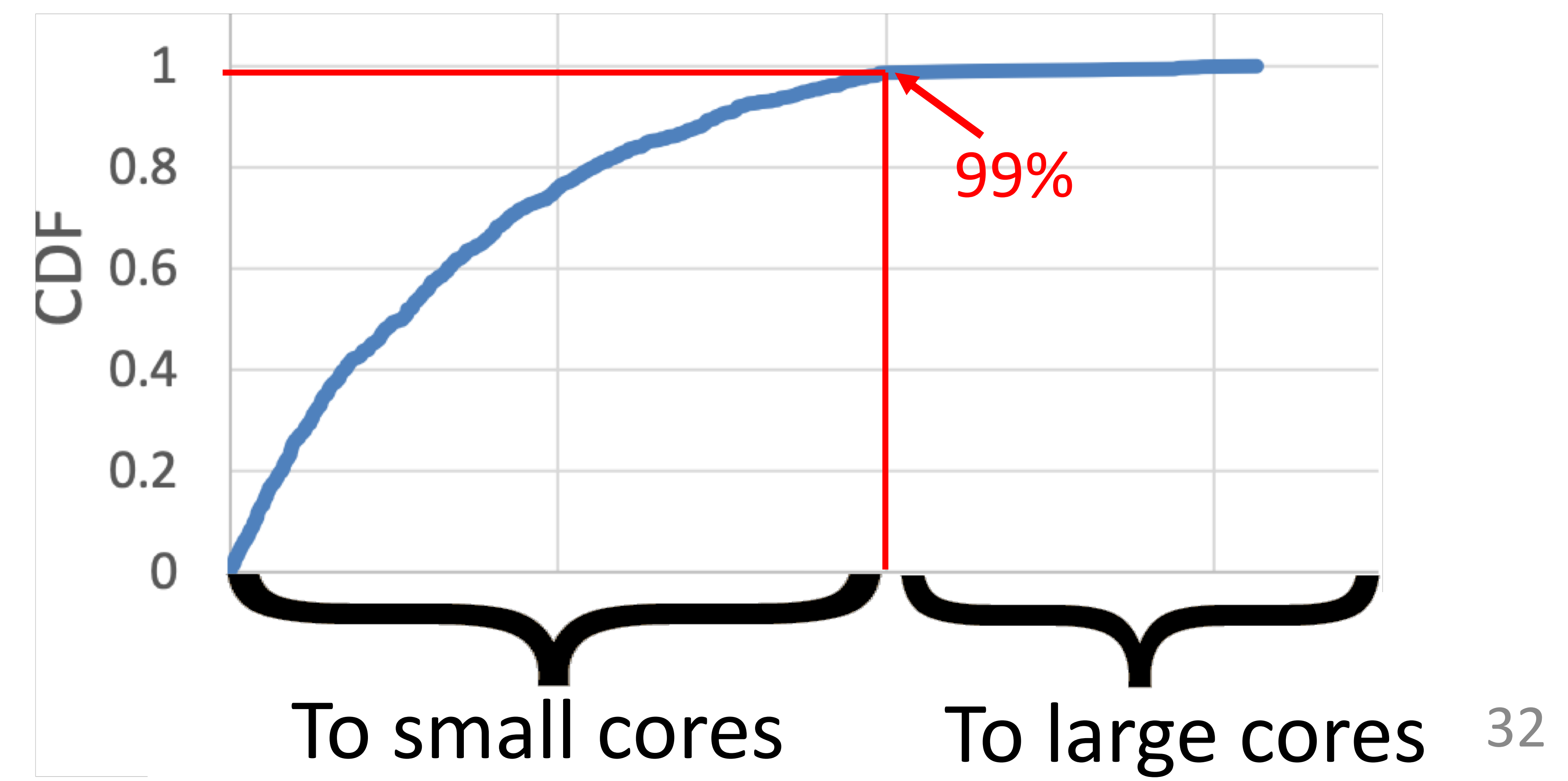


Minos design challenges

Small vs large threshold

Core partitioning

Request dispatch

Goal

Improve small requests

Avoid overloading large cores

Load-proportional core allocation

K% of the load for small requests

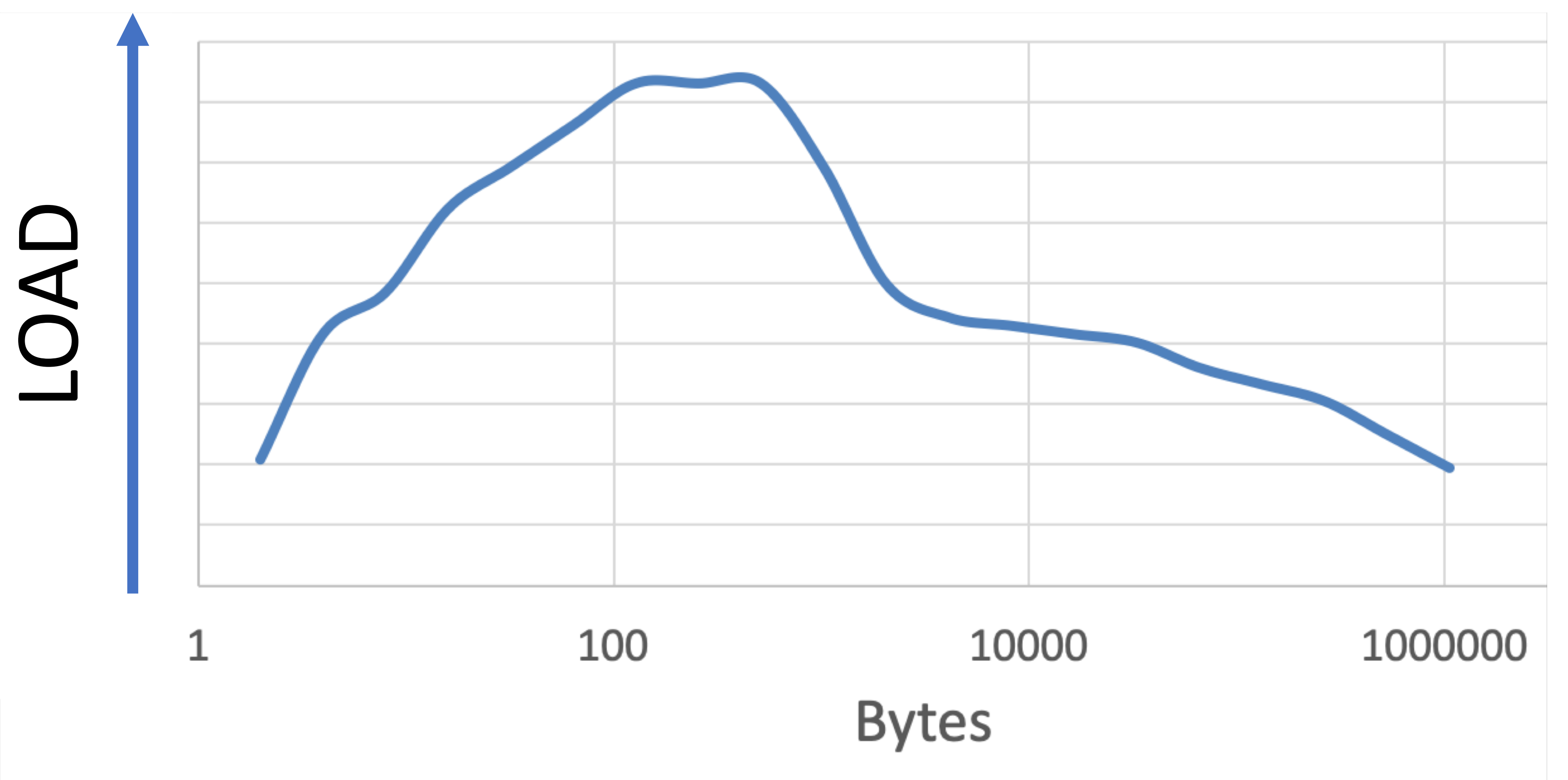
K% of small cores

Measuring the load of a request

Load of a request = # processed network packets

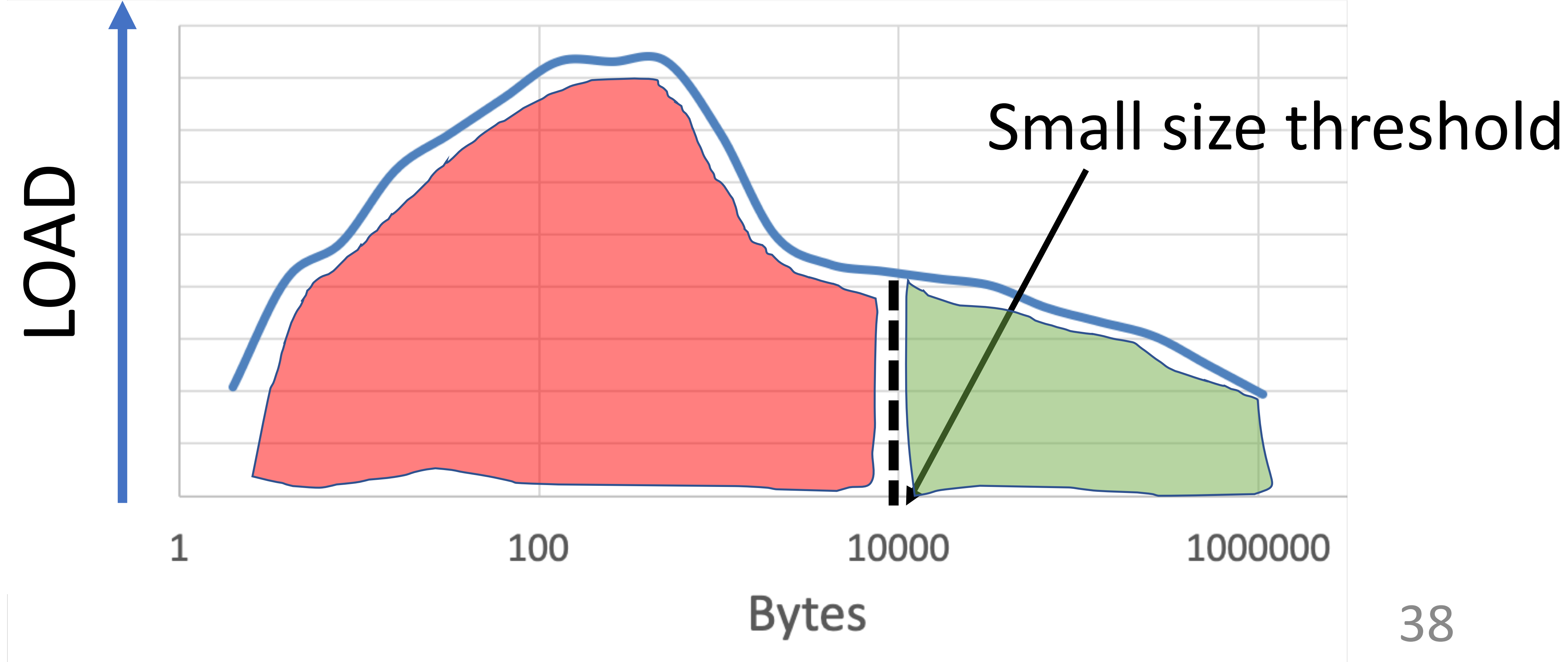
Dynamic core allocation

1. Obtain at runtime the load of requests of different sizes



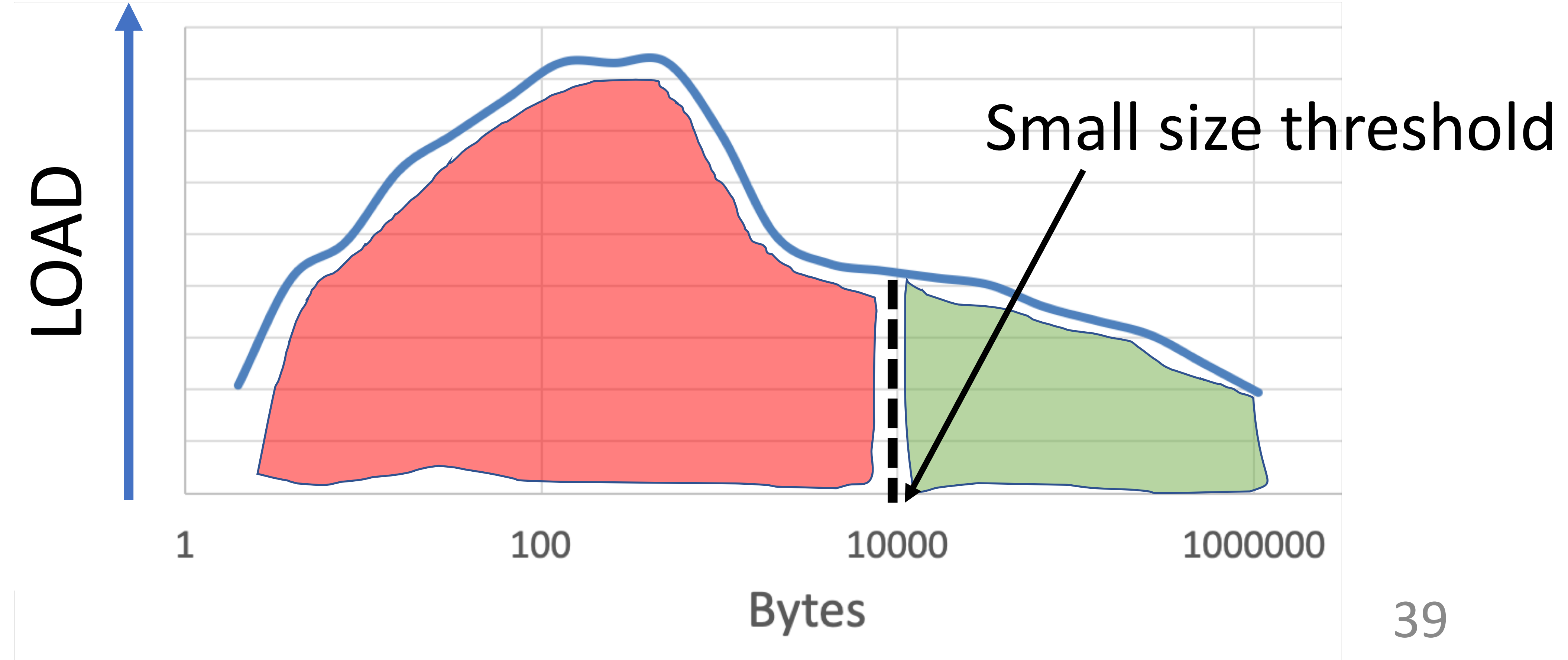
Dynamic core allocation

2. Fraction of small request load = [red] / ([green] + [red])



Dynamic core allocation

3. # Small cores = ceiling (small load * # total cores)



Minos design challenges

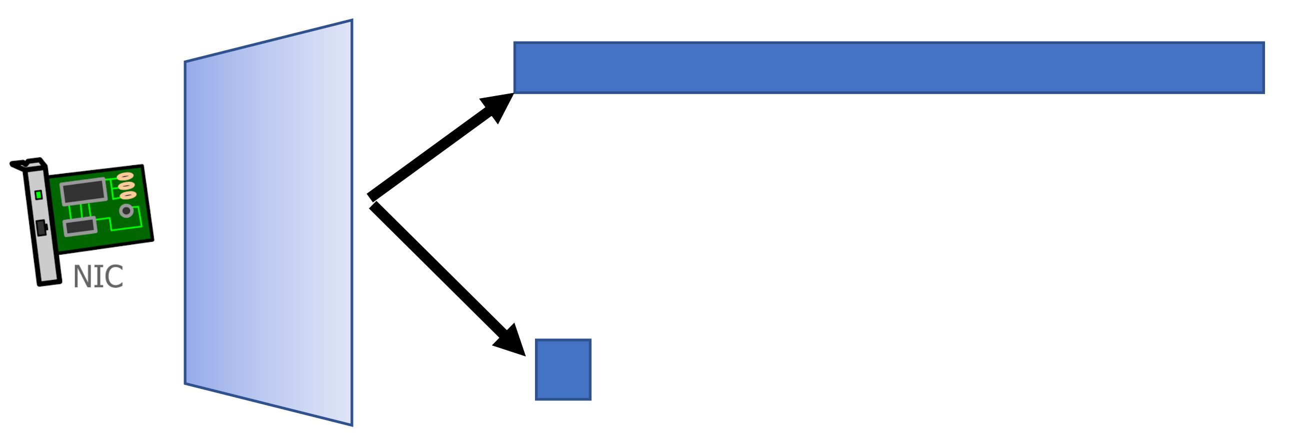
Small vs large threshold

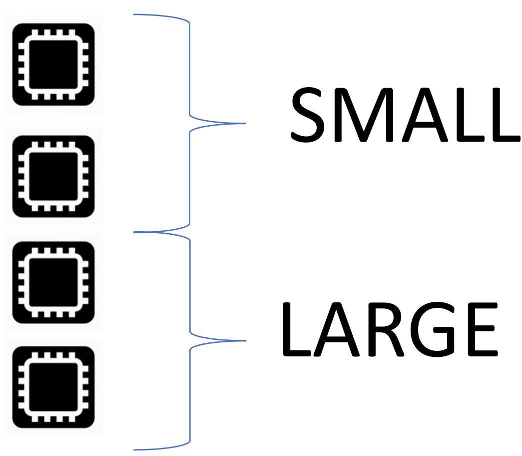
Core partitioning

Request dispatch

Request size unknown *a priori*

RX queues





Reduce software dispatch

Only small cores read from the NIC

1. From its own RX queue

2. From large cores’ RX queues

Operation of small cores on a request

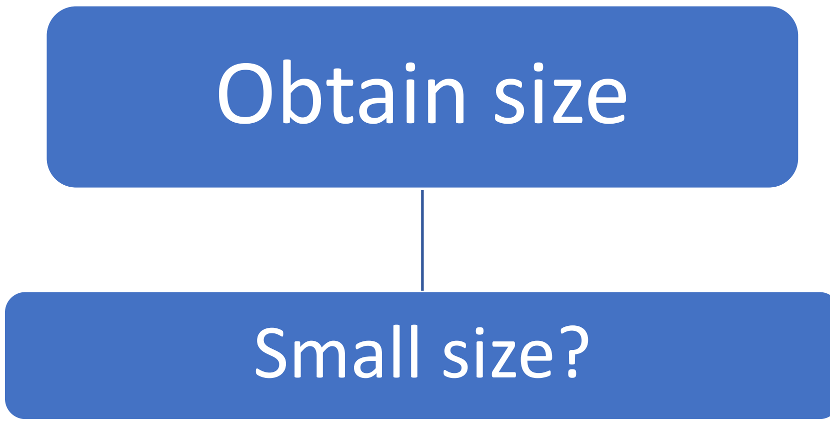
Obtain size

Operation of small cores on a request

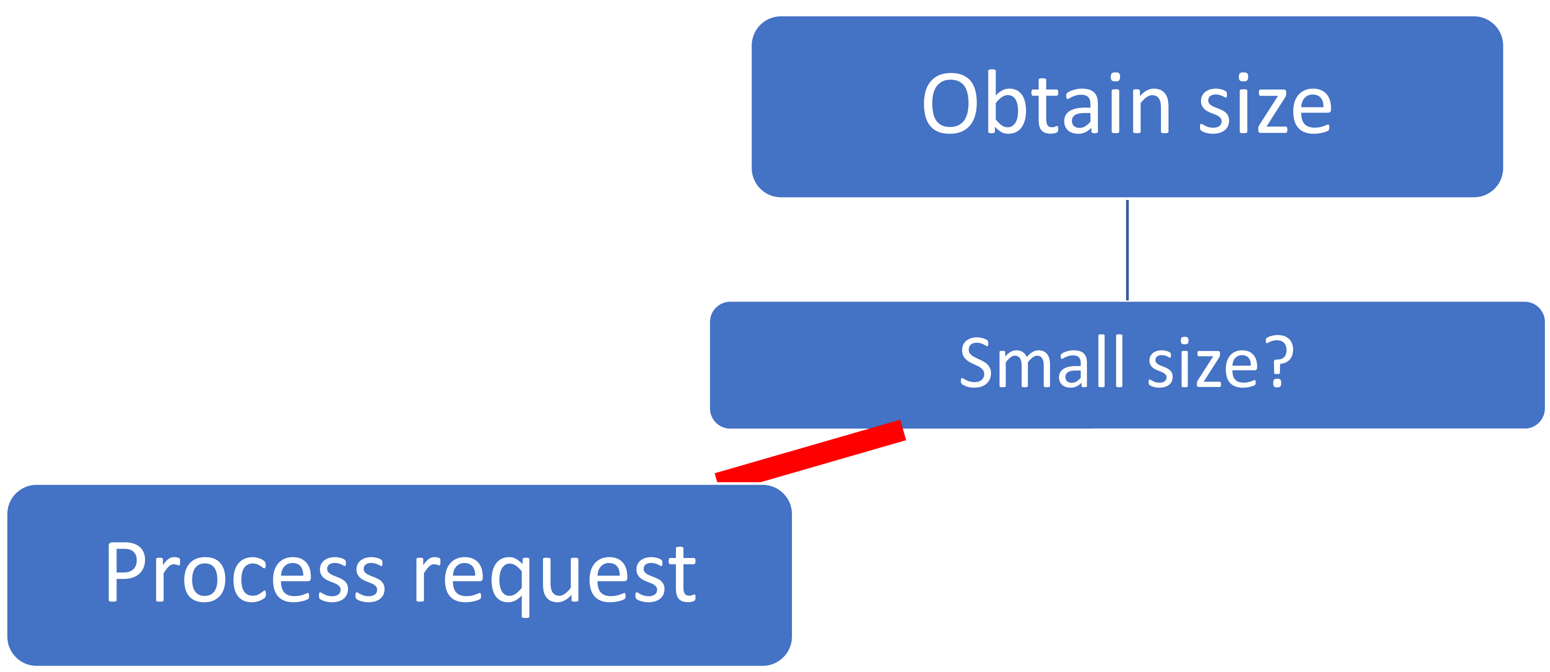
Obtain size

GET: do lookup
PUT: in header

Operation of small cores on a request



Operation of small cores on a request



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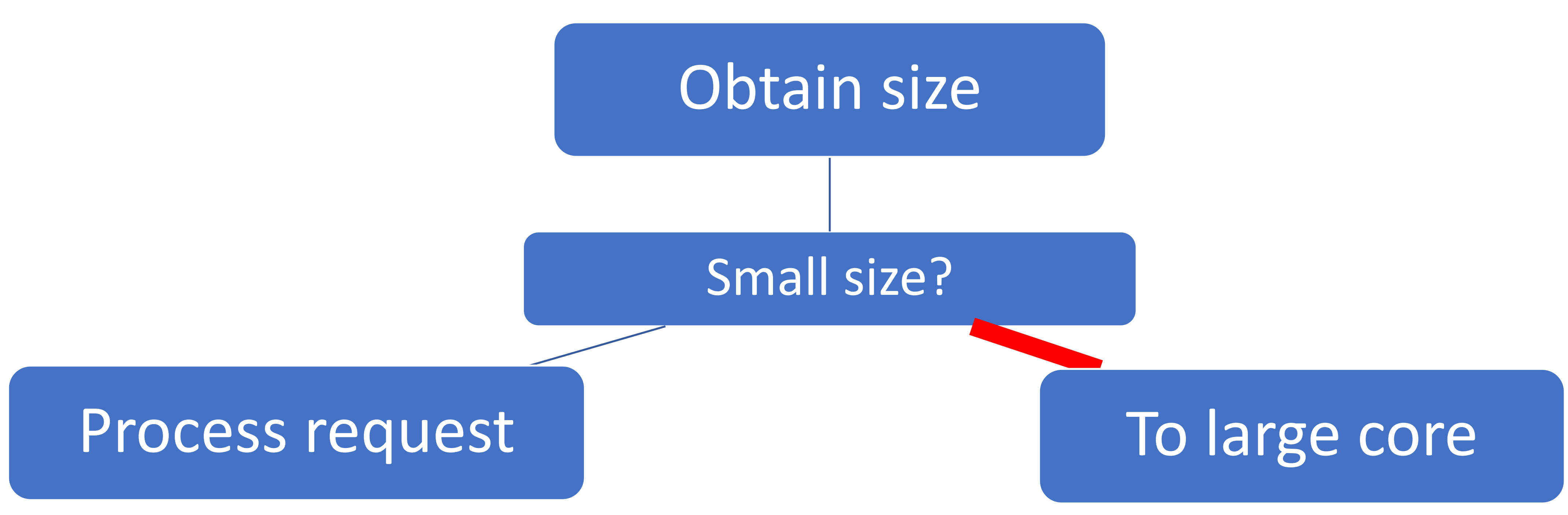
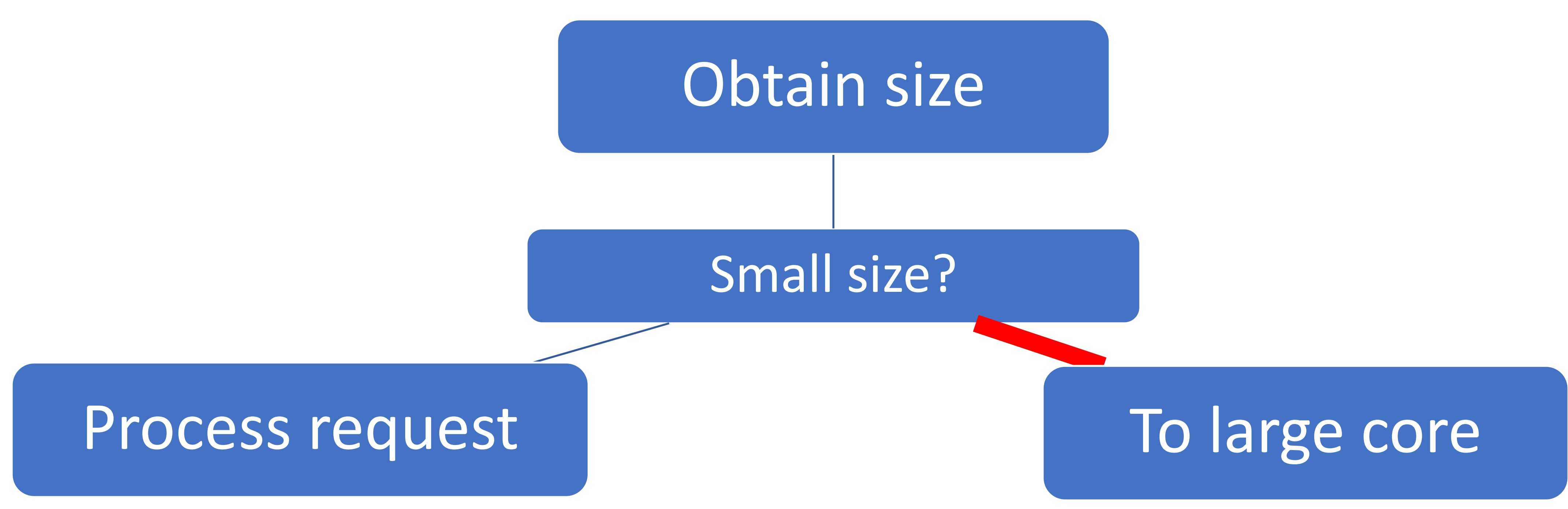


Operation of small cores on a request



SOFTWARE DISPATCH ONLY FOR FEW LARGE

EVALUATION

Test-bed

- Server: 8 cores, 40Gbps NIC, DPDK stack
- Wkld ~ ETC Facebook [SIGMETRICS12]
  - < 1 % large requests [1.5, 500] KB
- 95:5 GET:PUT ratio
- Skewed accesses (zipf 0.99)

Competitors

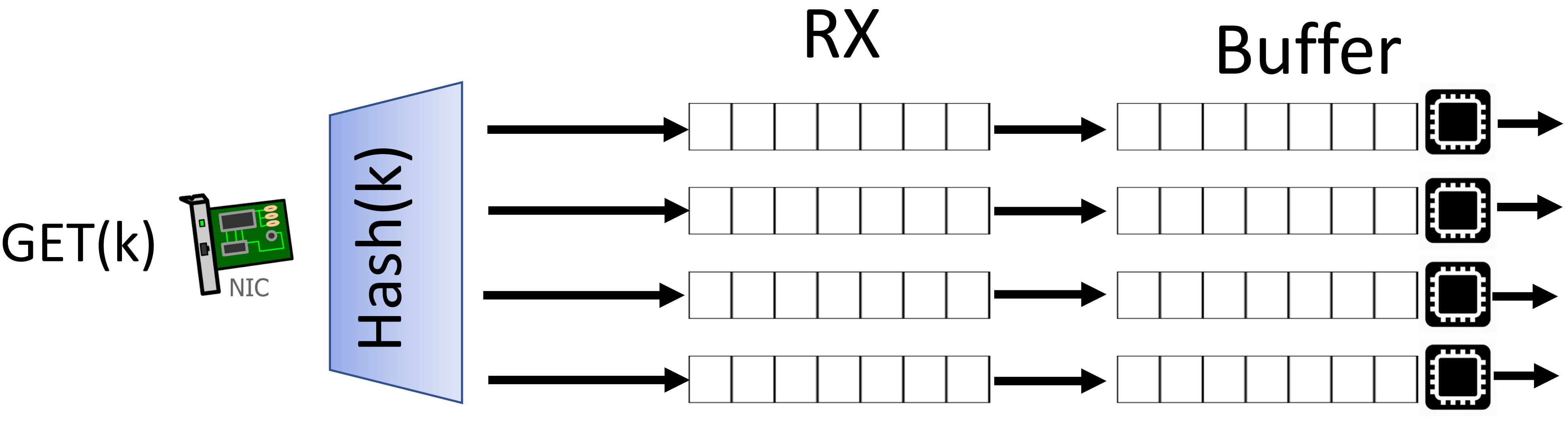
1. Early binding (~ MICA [NSDI14])

2. Early binding + stealing (~ ZygOS [SOSP17])

3. Late binding (~RAMCloud [TOCS15])

Early binding (MICA, NSDI2014)

Request -> core based on key-hash of target item



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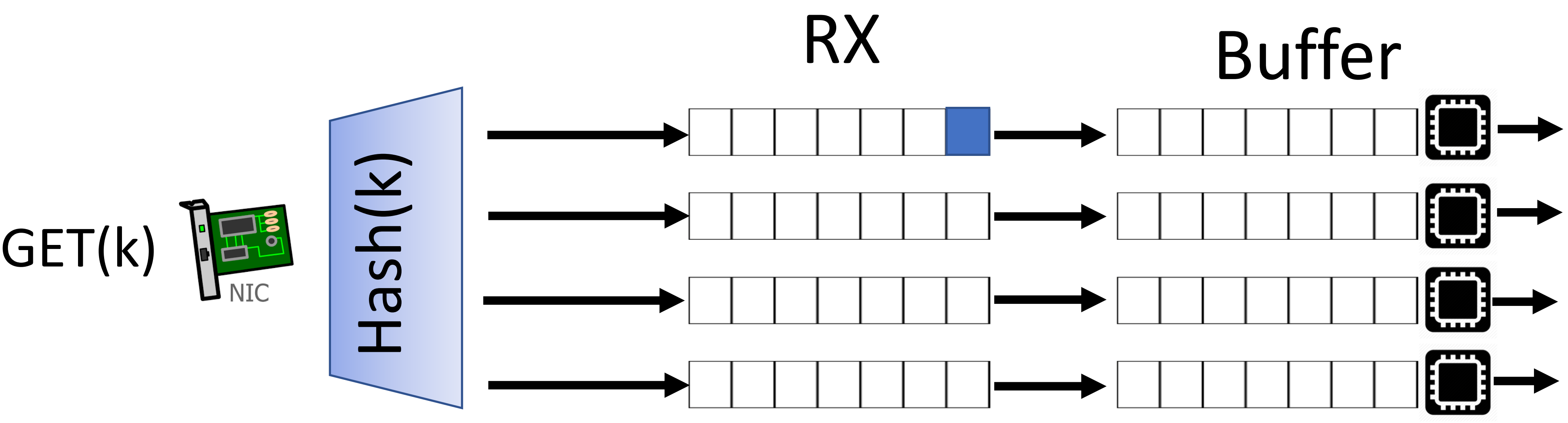

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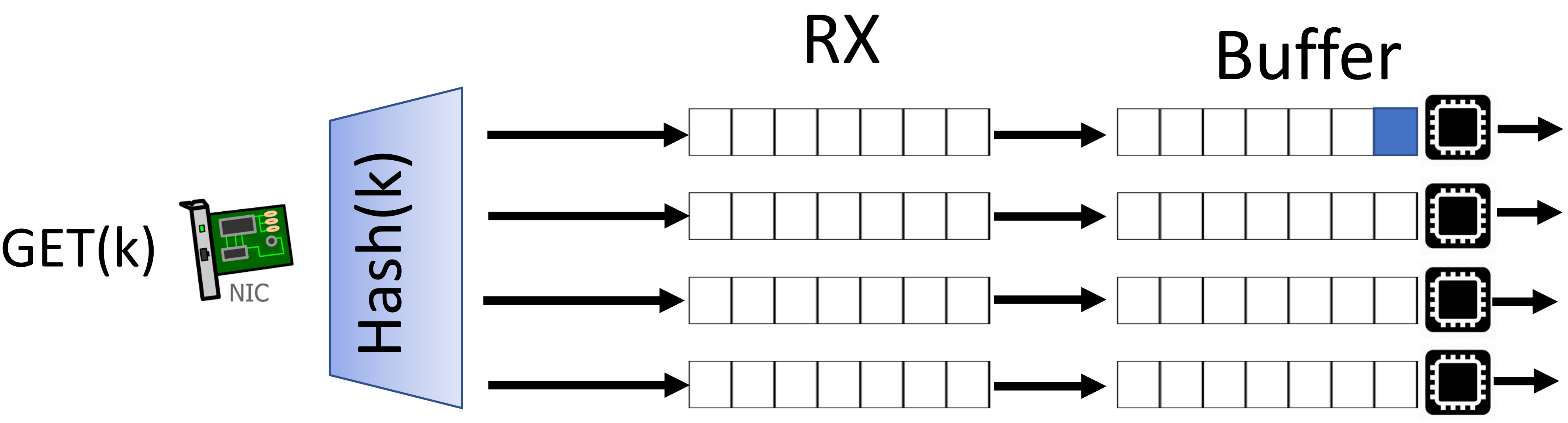
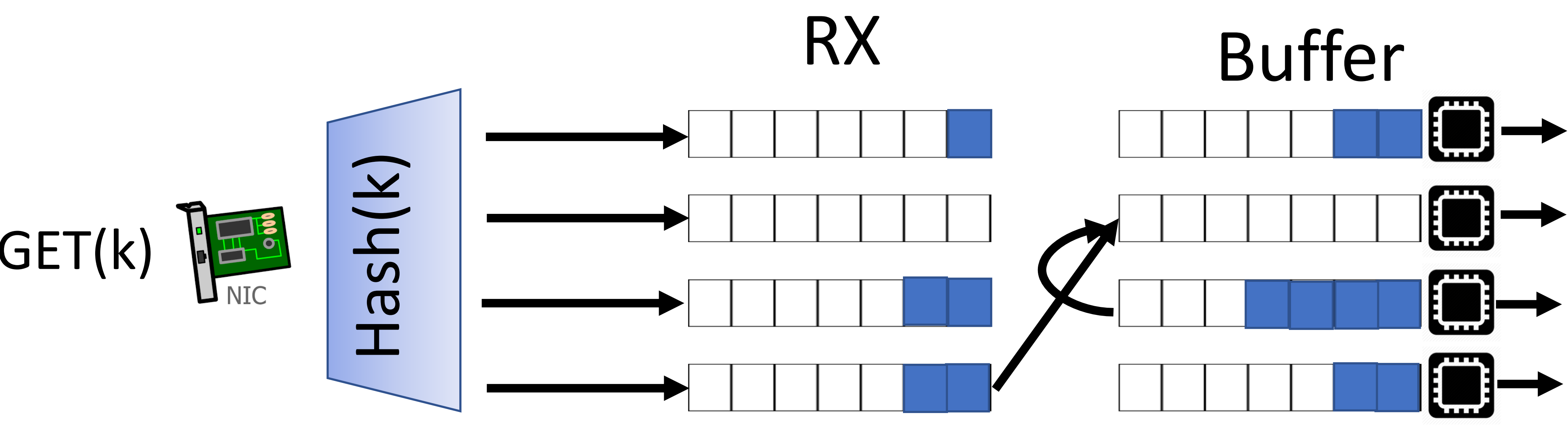


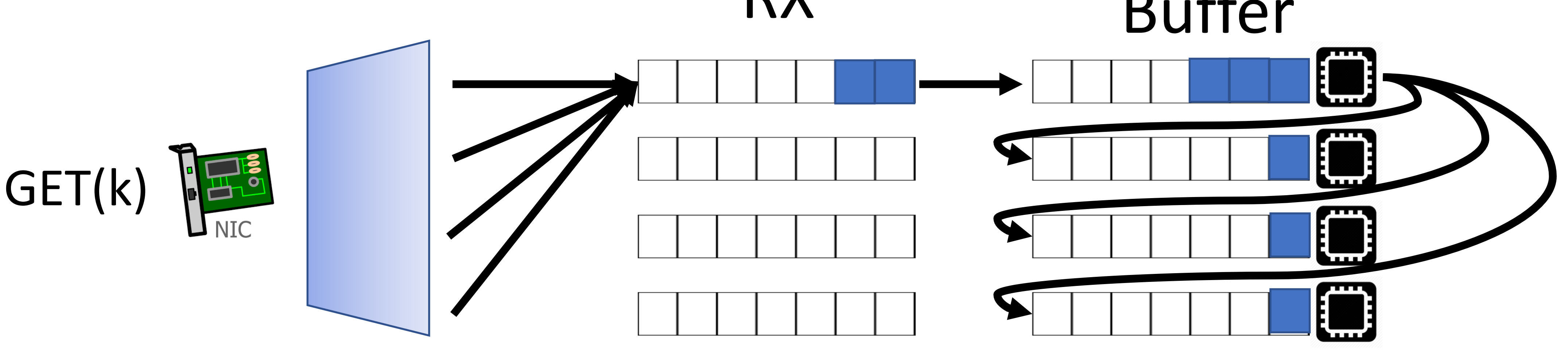
Early binding + stealing (ZygOS SOSP17)

Idle cores steal requests from other queues/buffers

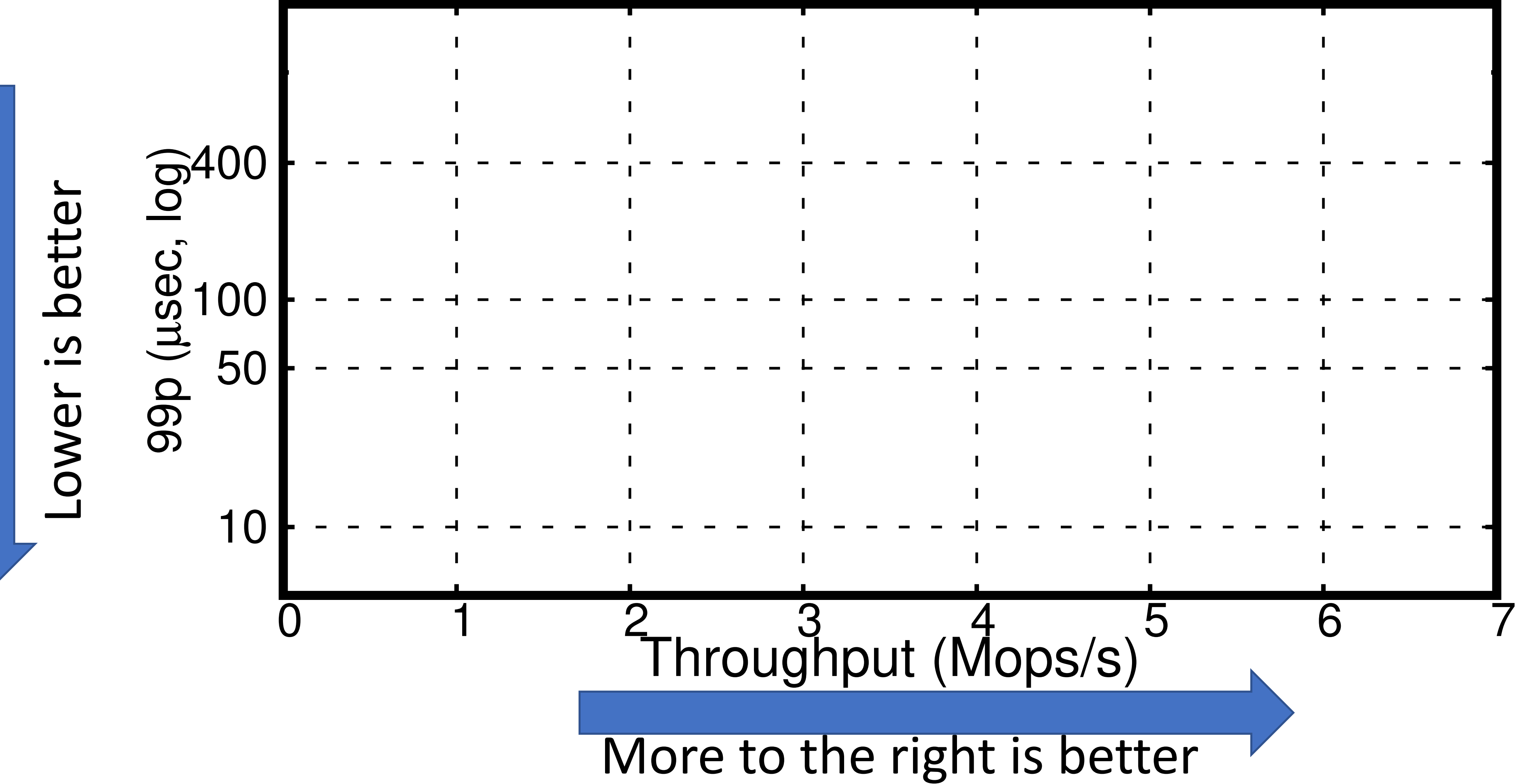


Late binding (RAMCloud TOCS15)

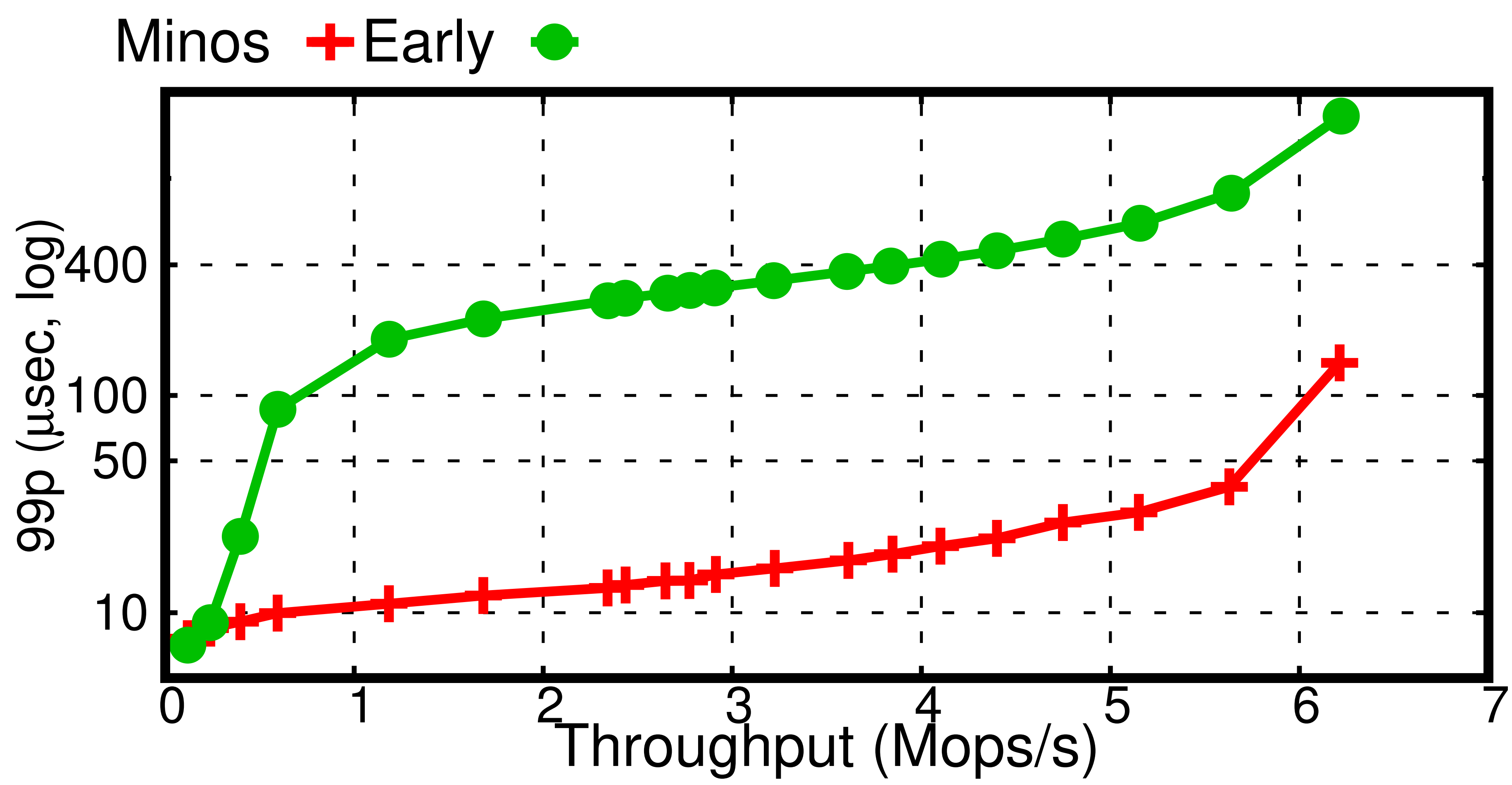
One core receives all requests
dispatches them to idle cores



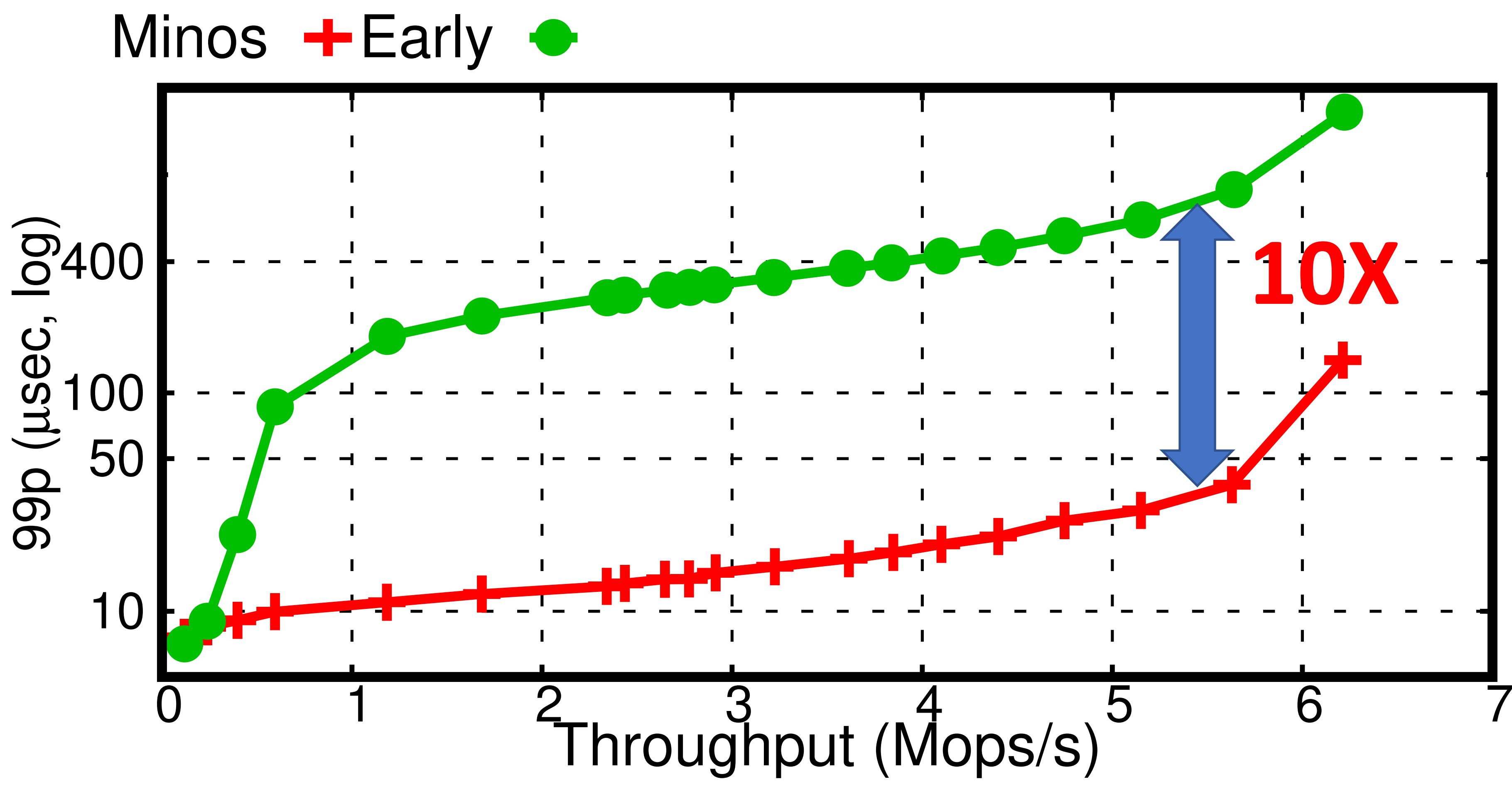
Throughput vs overall 99th latency



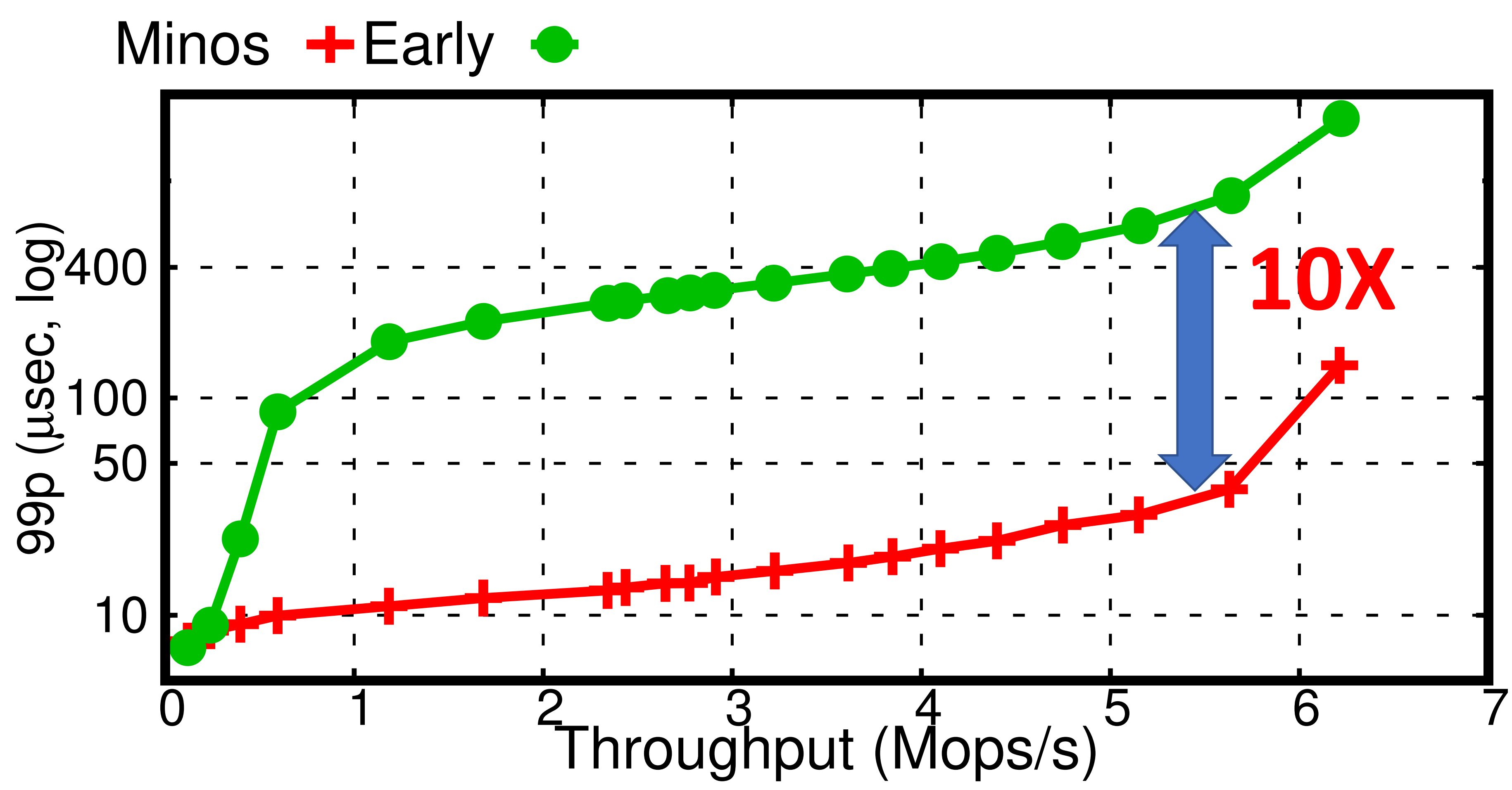
Minos vs early binding



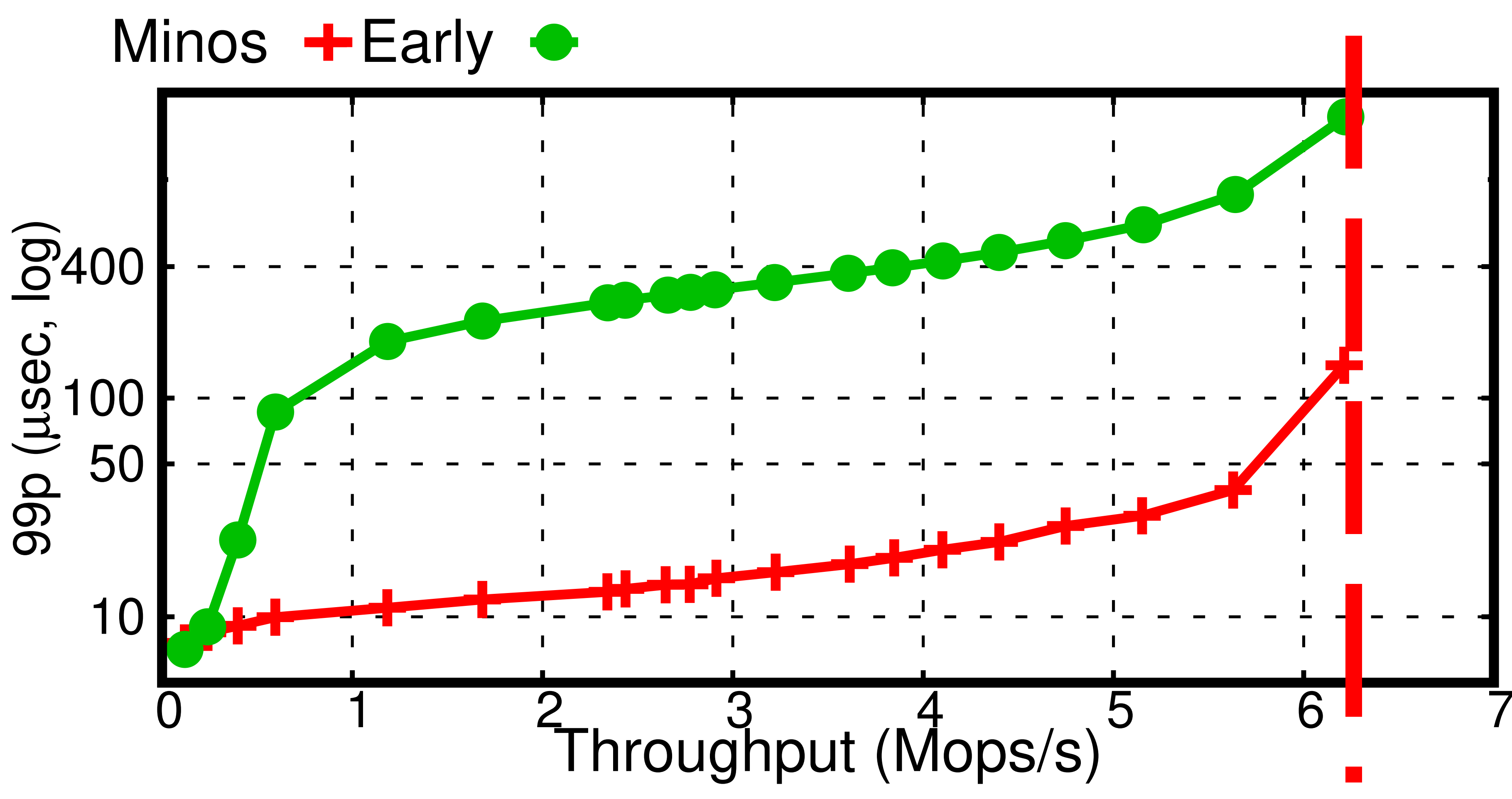
Lower latency



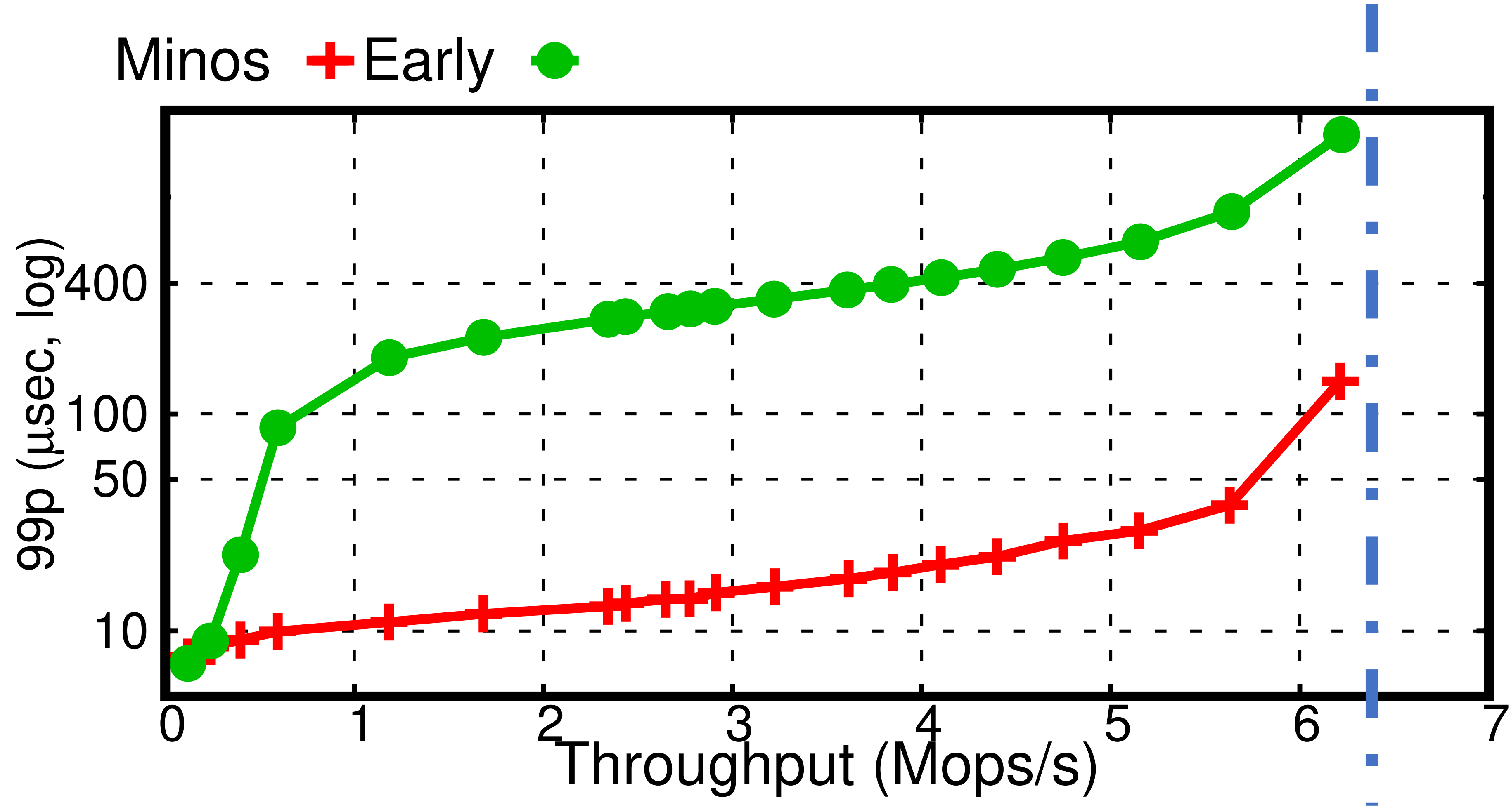
Why? No head-of-line blocking



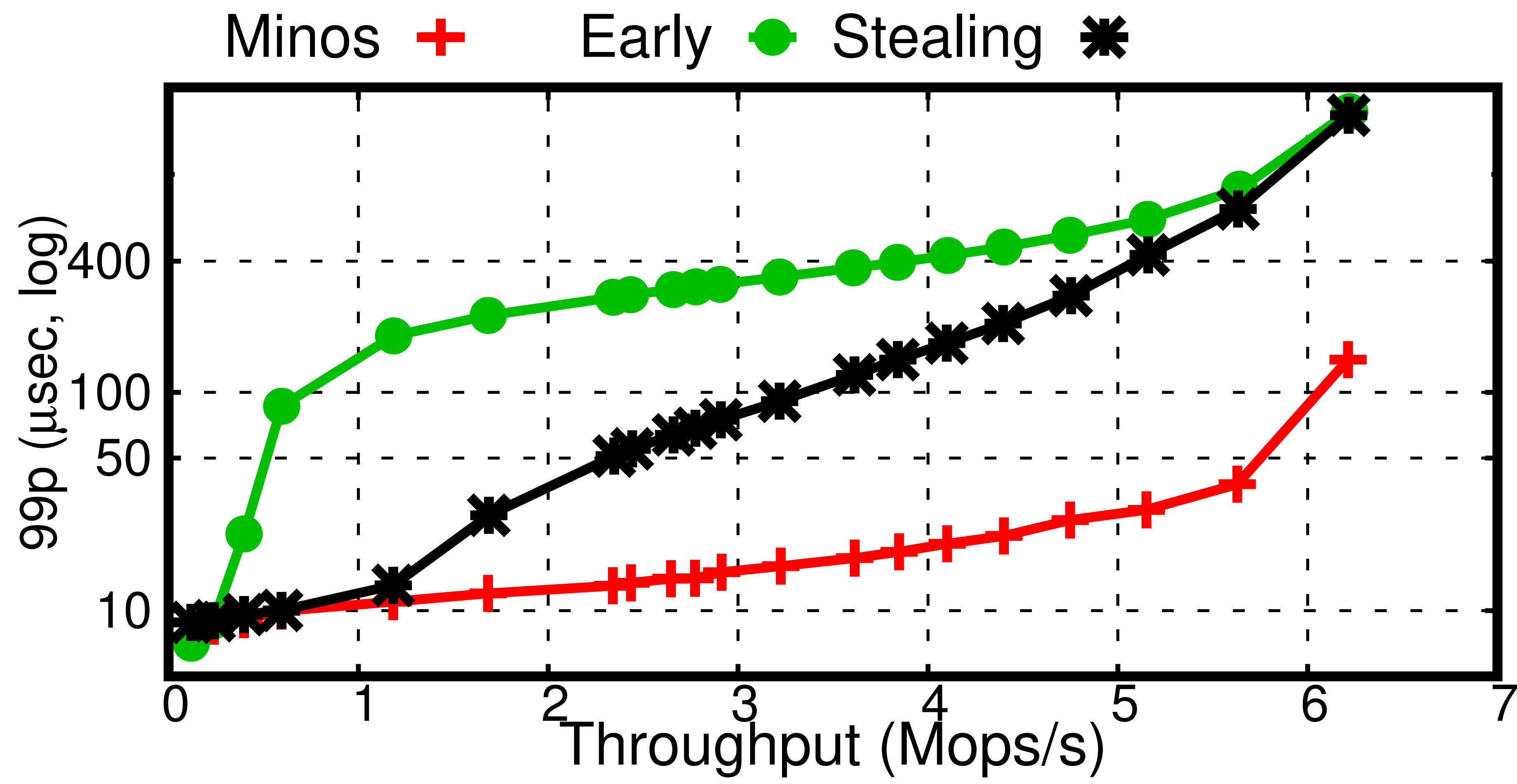
Same maximum throughput



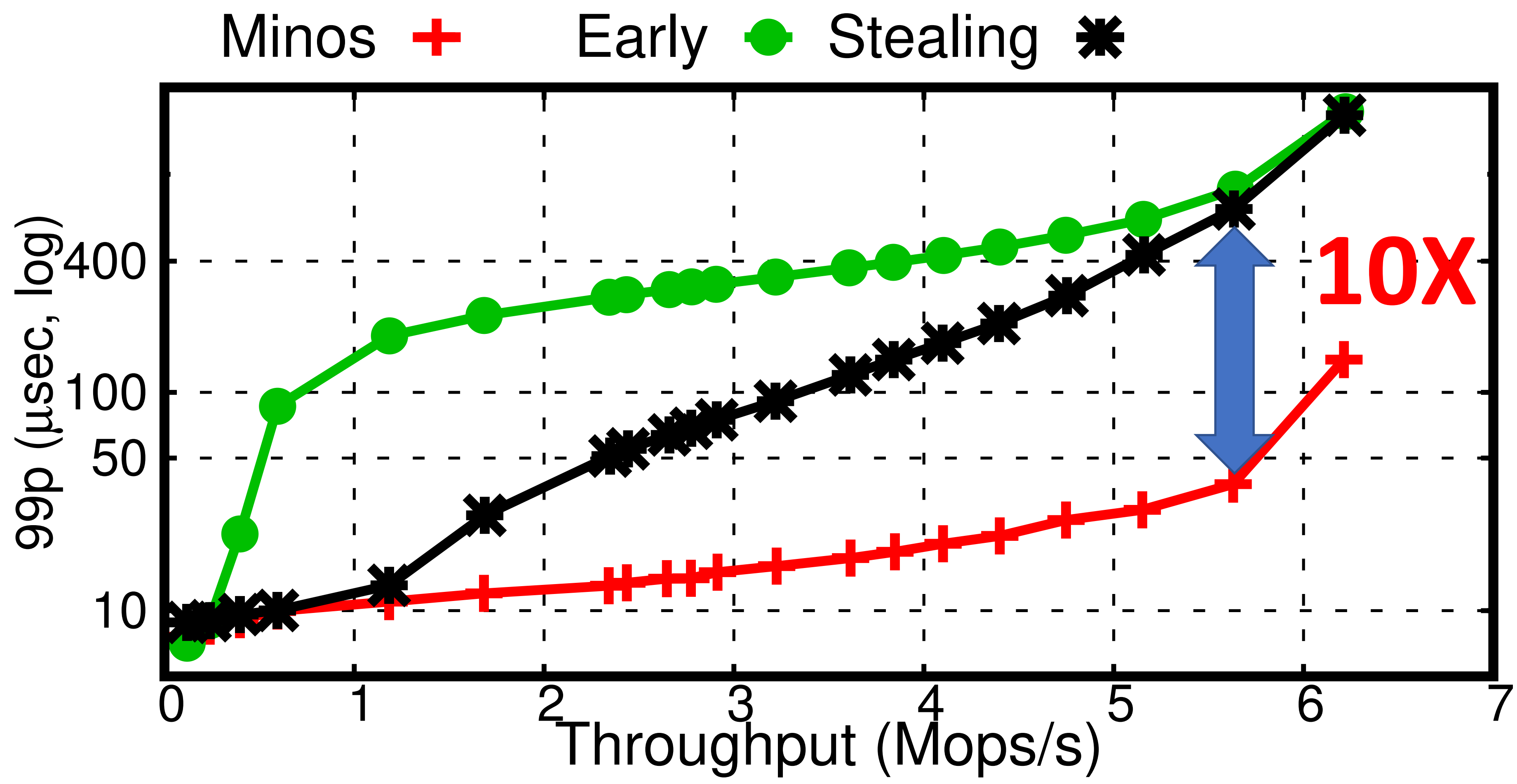
Why? Low dispatch overhead



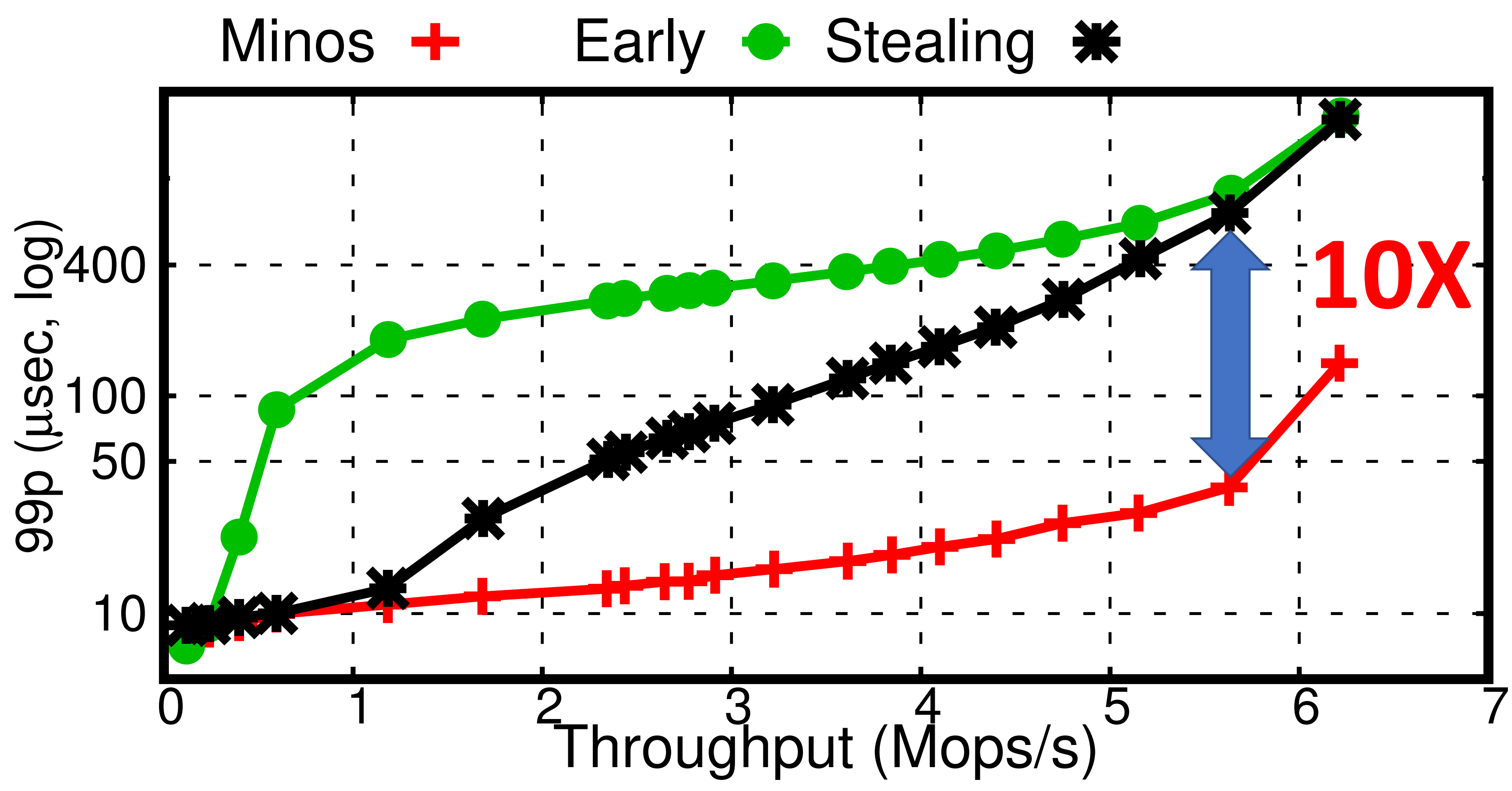
Minos vs stealing



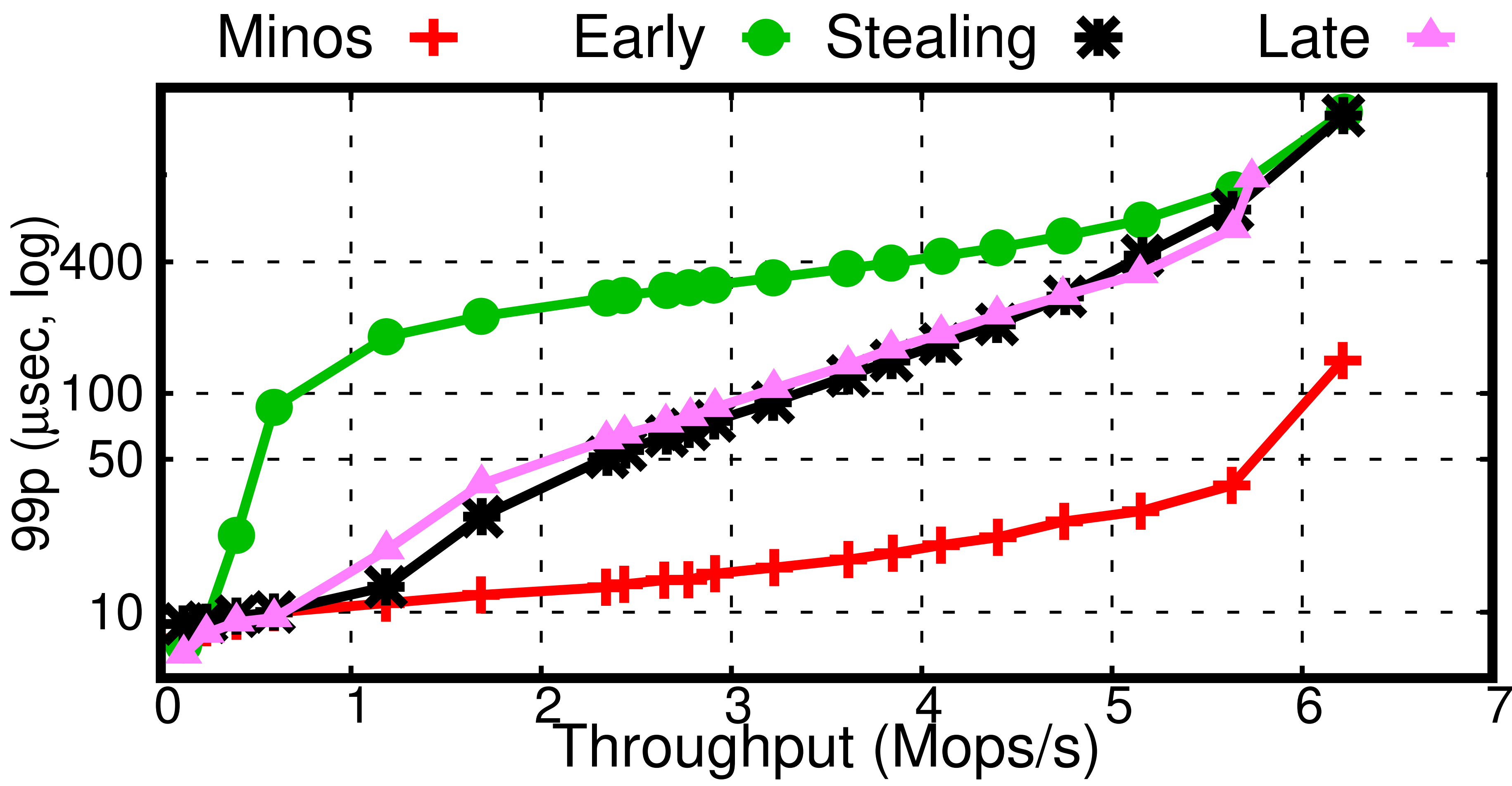
Lower latency



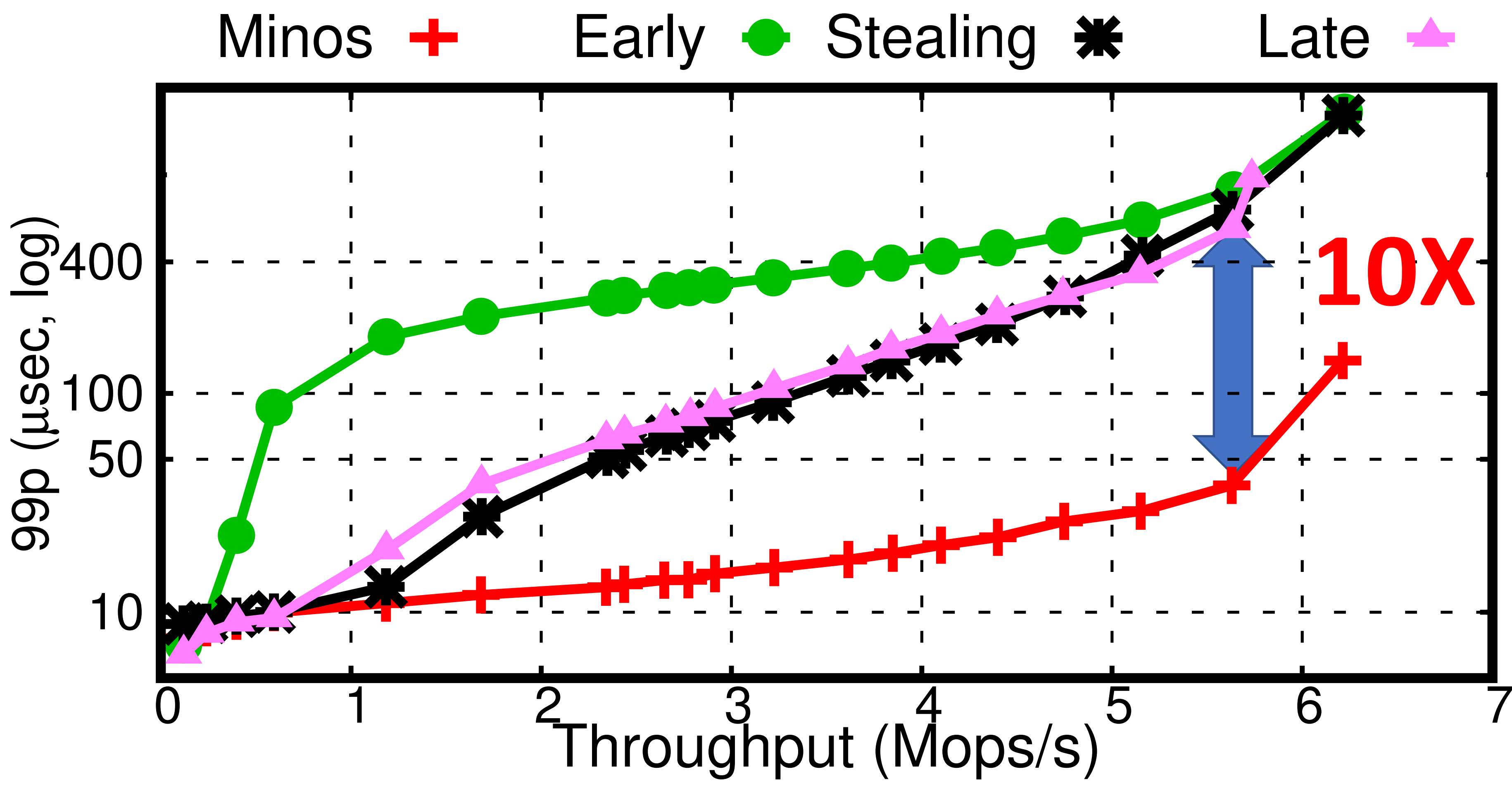
Why? Higher load→ lower stealing



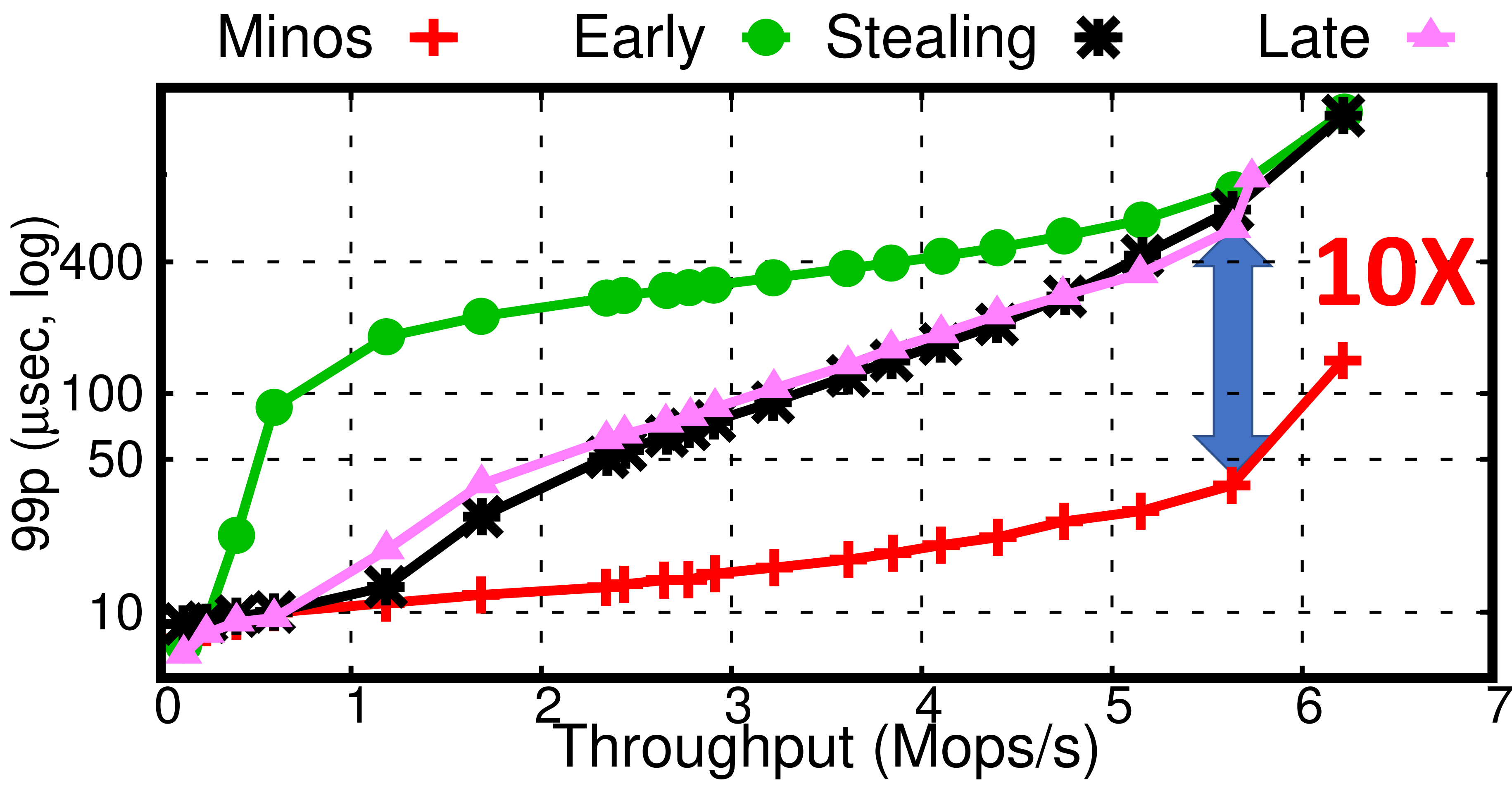
Minos vs late binding



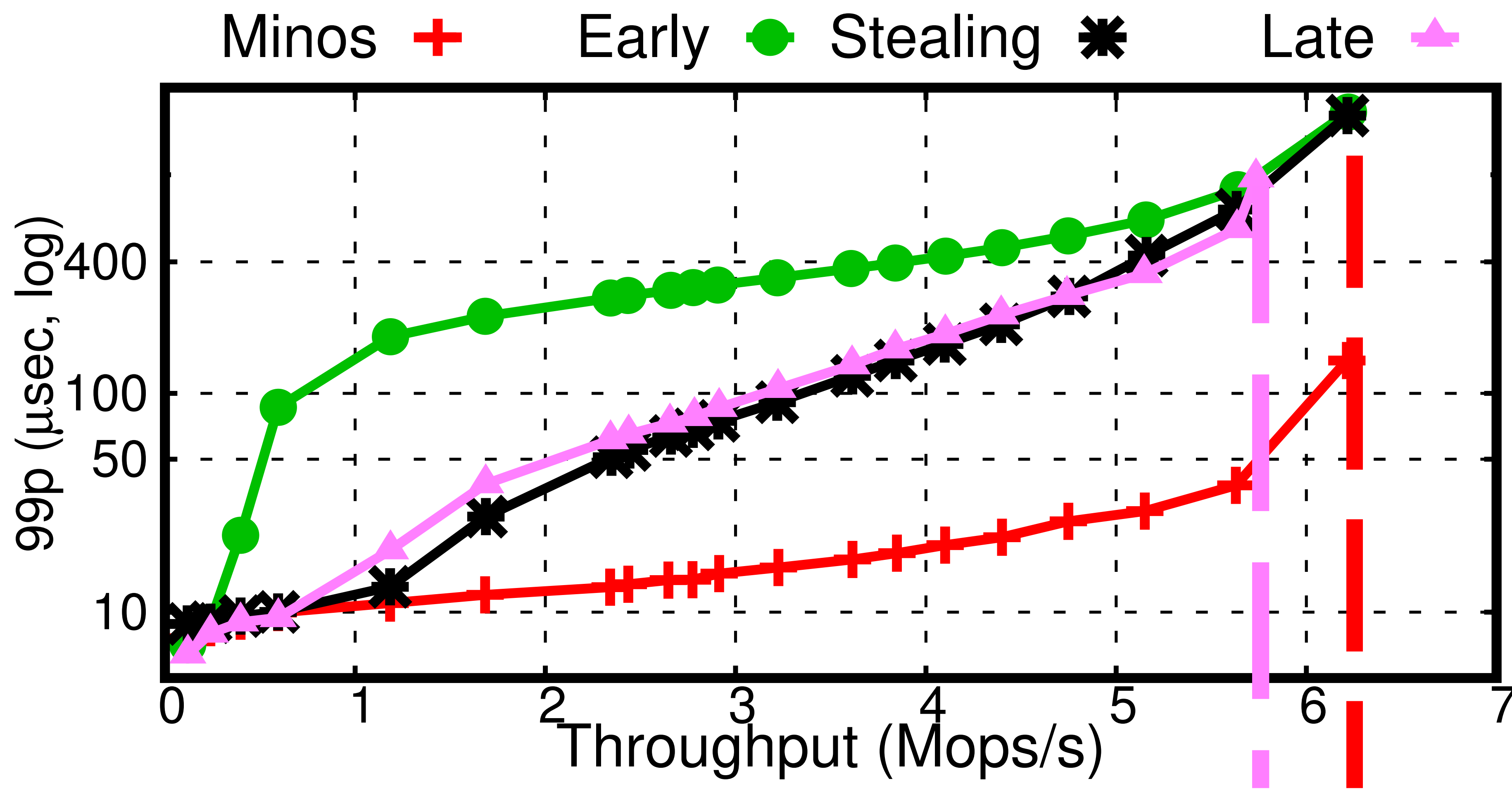
Lower latency



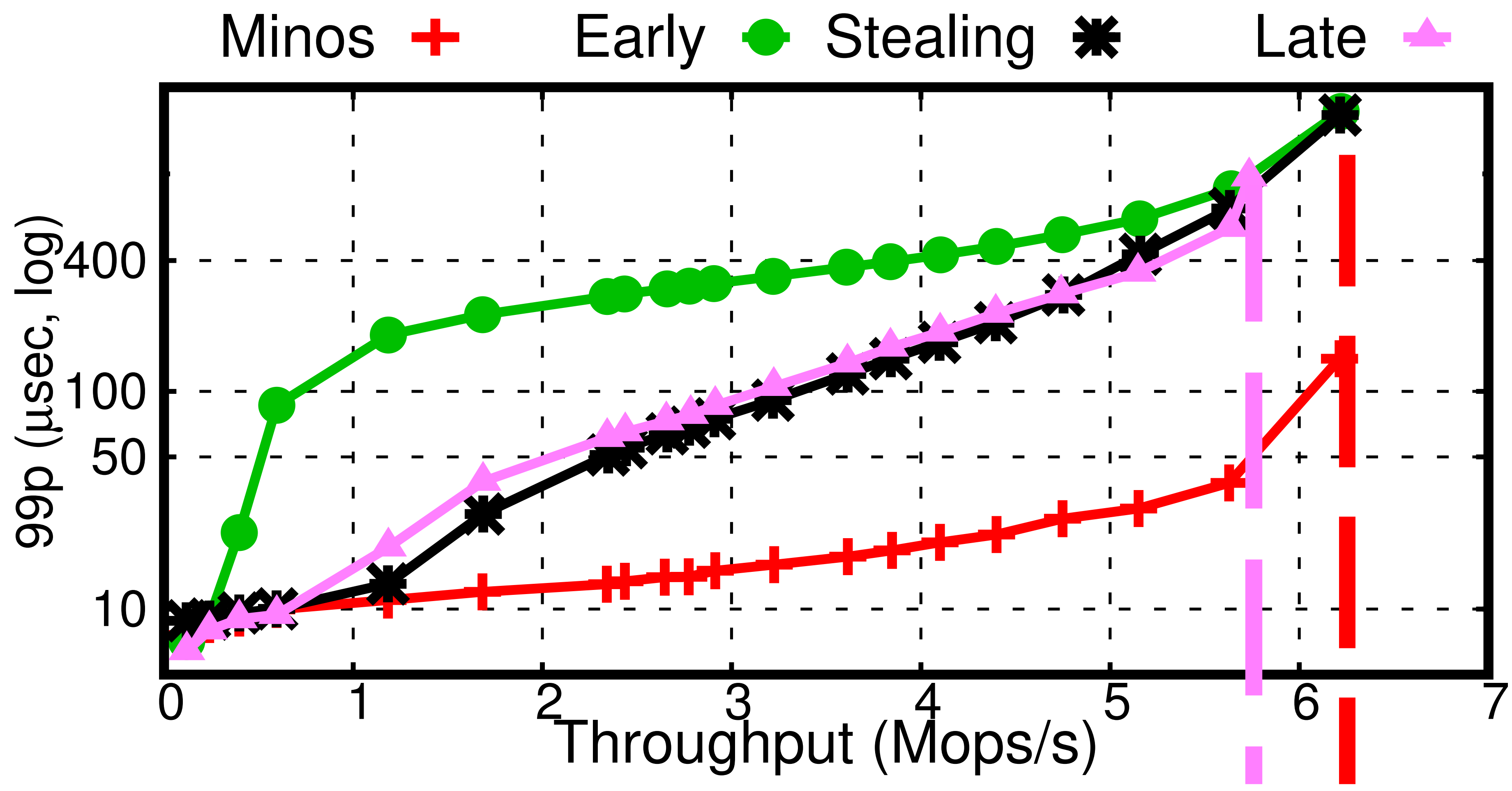
Why? No convoy effect



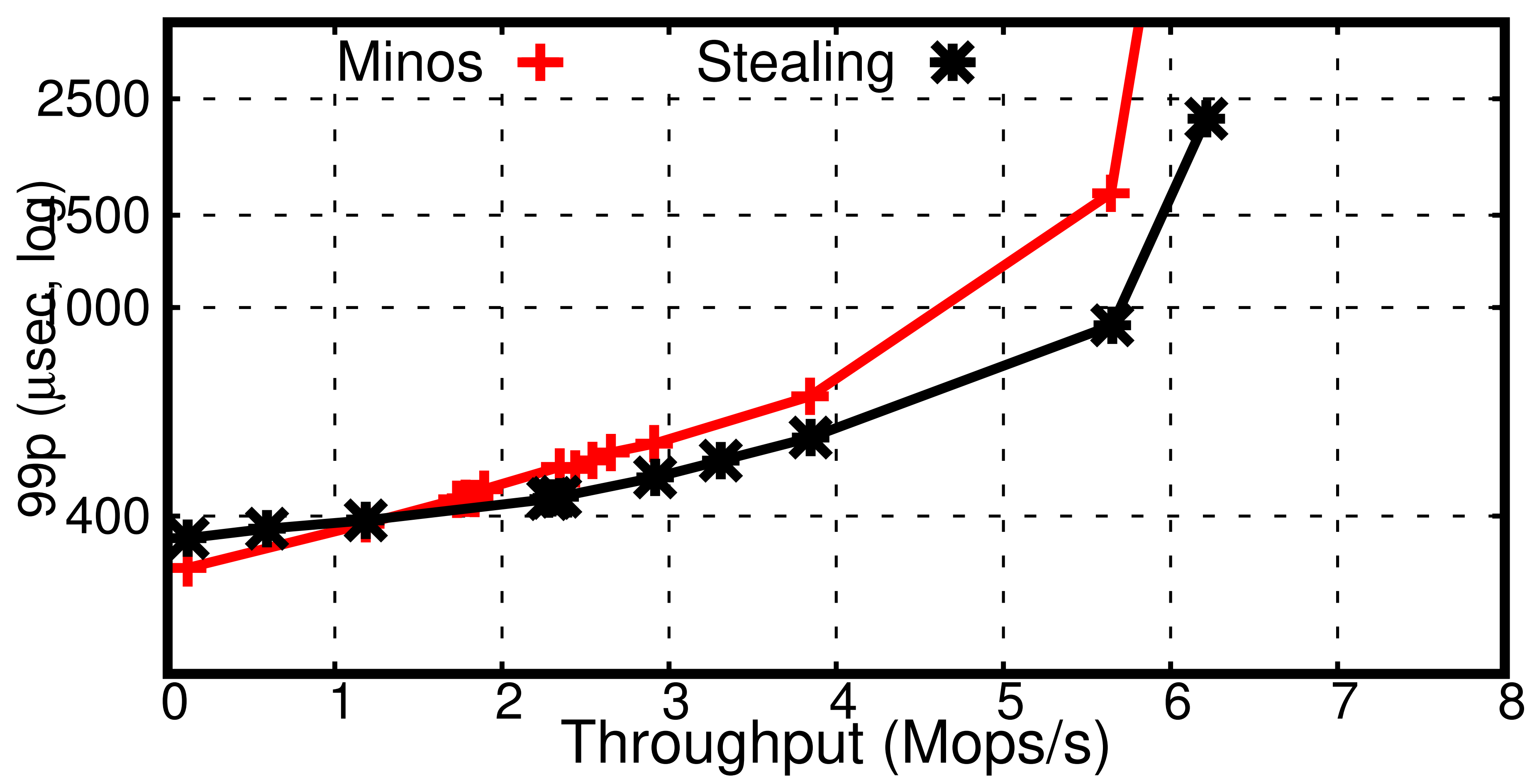
Higher throughput



Why? No dispatch bottleneck



Trade-off: 99th latency of **large** operations



Other results in the paper

- More item size distributions
- Dynamic workload
- Write intensive workload
- Scalability

Conclusion: size-aware sharding

Improve tail latency in in-memory key-value stores

Serve small and large requests on different cores

- Minos in-memory KV: 10x lower 99th percentile latency

THANK YOU

ANY QUESTIONS?