

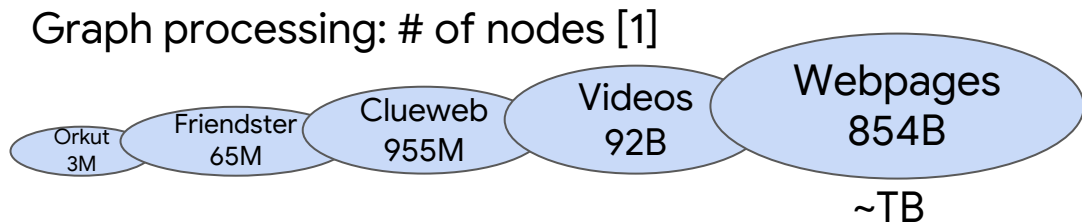
Carbink: Fault-Tolerant Far Memory

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Memory-Intensive Applications in Data Centers



VOLTDB [2]

Memory provisioning is hard, as memory is limited by server physical boundary

- Over-provisioning memory for peak usage → 40%-60% memory utilization [3]
- Growing data in one process even exceeds single-server memory limit

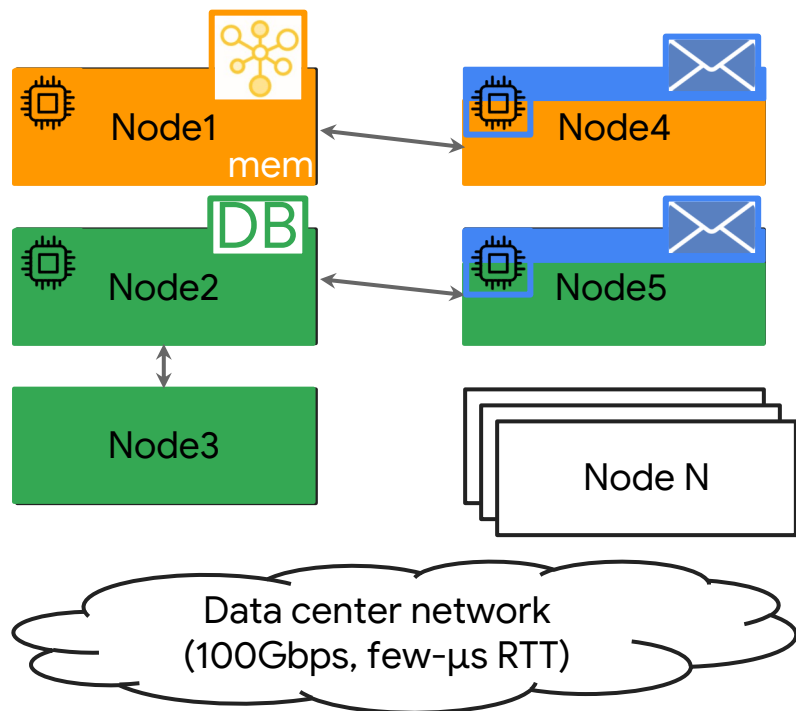
Can applications dynamically utilize the unused memory on other servers?

[1] Łącki, Jakub, et al. "Connected components at scale via local contractions." arXiv preprint 2018

[2] Stonebraker, Michael, et al. "The VoltDB Main Memory DBMS." IEEE Data Eng. Bull 2013

[3] Tirmazi, Muhammad, et al. "Borg: the next generation." EuroSys 2020

Background: Far Memory on Commodity Servers [1,2,3,...]



Benefits of far memory:

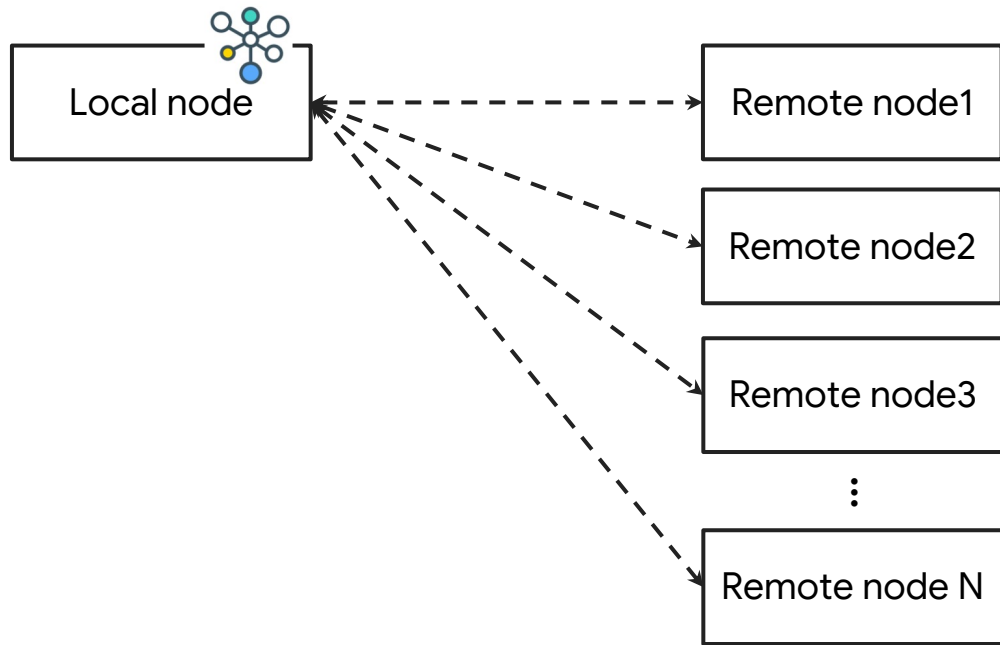
- Dynamically provisioning unused memory to memory-intensive apps
- Apps can use much more memory than single-machine limit

[1] Gu, Juncheng, et al. "Efficient memory disaggregation with infiniswap." NSDI 2017

[2] Aguilera, Marcos K., et al. "Remote regions: a simple abstraction for remote memory." ATC 2018

[3] Amaro, Emmanuel, et al. "Can far memory improve job throughput?." EuroSys 2020

Background: Far Memory on Commodity Servers [1,2,3,...]

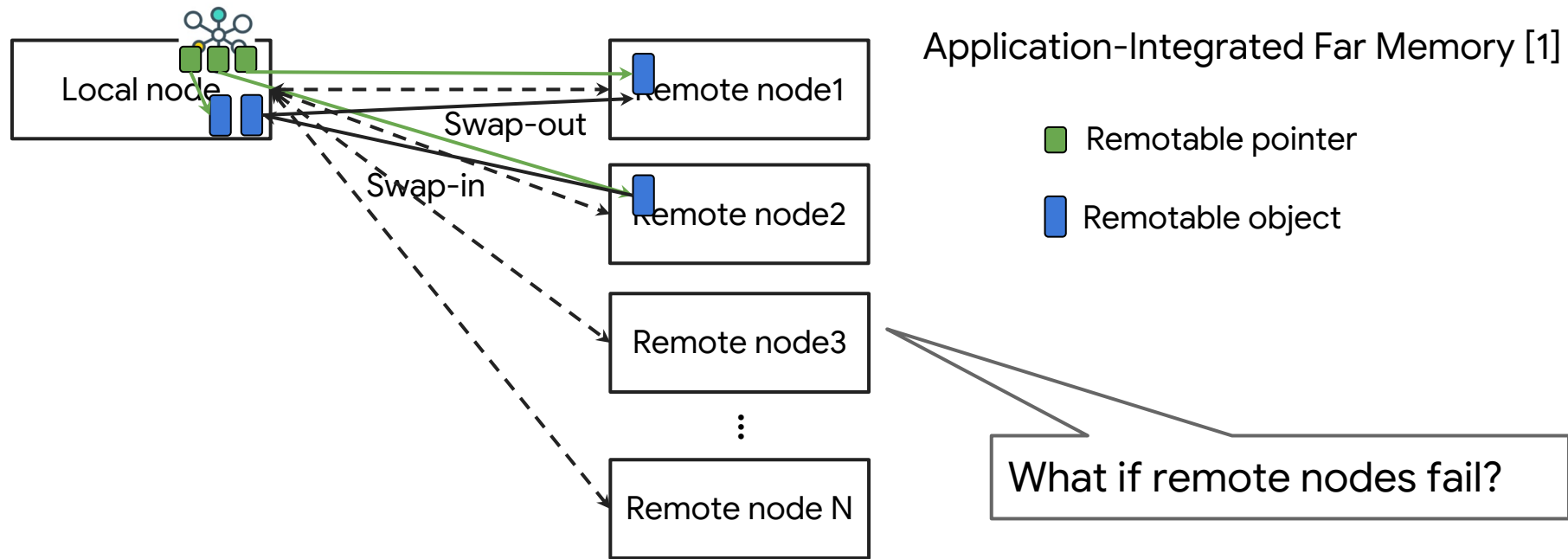


[1] Gu, Juncheng, et al. "Efficient memory disaggregation with infiniswap." NSDI 2017

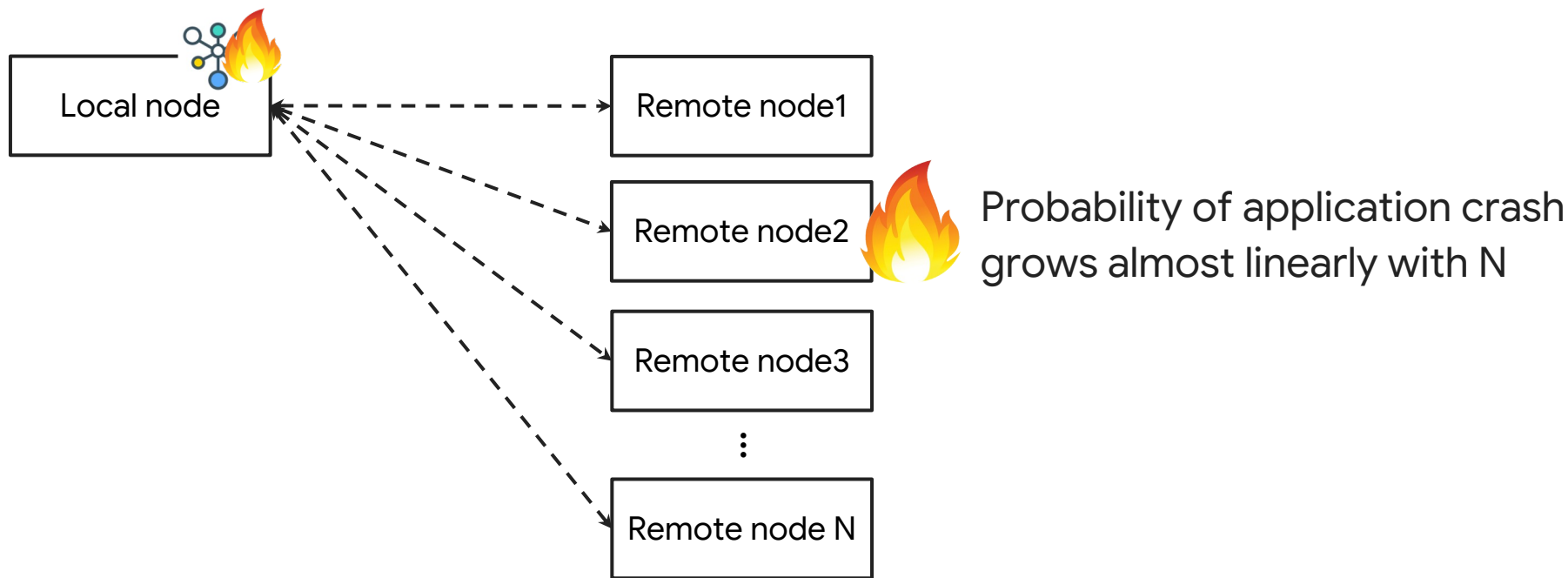
[2] Aguilera, Marcos K., et al. "Remote regions: a simple abstraction for remote memory." ATC 2018

[3] Amaro, Emmanuel, et al. "Can far memory improve job throughput?." EuroSys 2020

Application Interface: Remotable Pointers



The Must-Have Feature: Fault Tolerance



How to build a fault-tolerant far memory system?

... assume fail-stop faults and no partial network failures

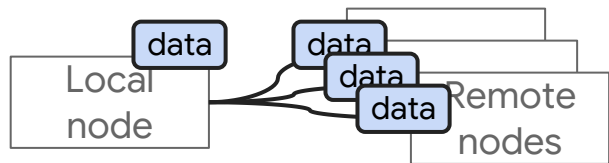
Talk Outline

Direction: in-memory erasure coding for fault tolerance

Carbink: making erasure coding work in practice

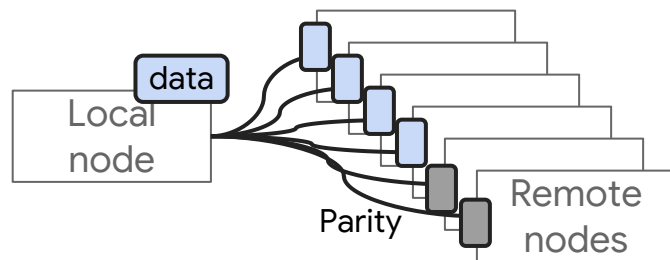
Evaluation: performance and cost of Carbink

Replication vs. Erasure Coding



Replication

- High memory overheads (3x)



Erasure coding (EC)

- Much smaller memory usage (1.5x)
- Single core achieves 4GB/s encoding tput [1]

SSD vs. Memory

SSD would become bottleneck during bursty workloads or failure recovery [1]

In-memory erasure coding



Talk Outline

Direction: in-memory erasure coding for fault tolerance

- High performance & low memory usage

Carbink: making erasure coding work in practice

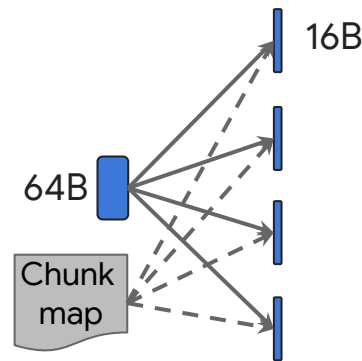
Evaluation: performance and cost of Carbink

Challenge 1: Remotable Objects Have Different Sizes

Erasure coding **irregular**-sized objects is hard



Padding: objects aligned but wasting memory

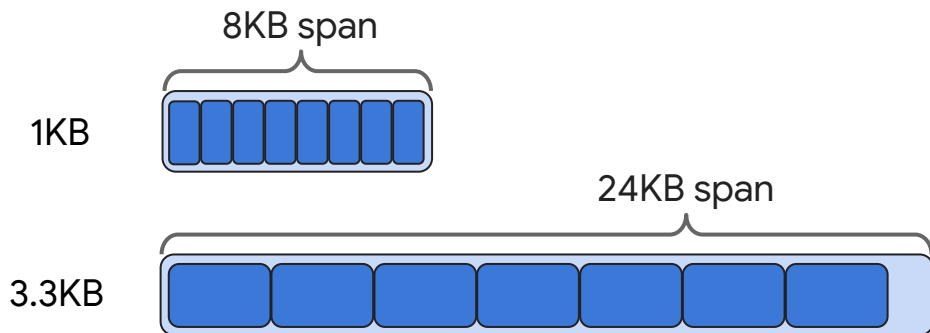


Splitting: small objects incurs large metadata

Carbink approach: grouping similar-sized objects into spans (like TCMalloc [1])

- Spans are page-aligned and **regular-sized**

Grouping Similar-Sized Objects into Spans

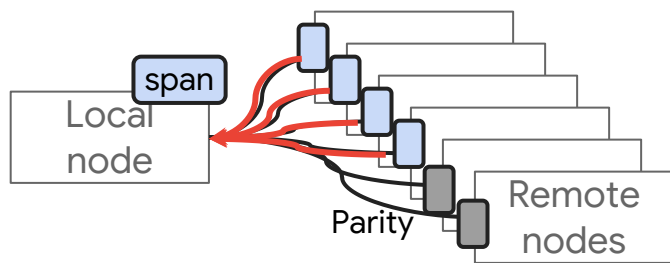


Span-centric memory pooling

- Applying spans to object management and data swapping
- Spans are page-aligned, and never end with a partial object

Challenge 2: Efficient Swapping under Erasure Coding

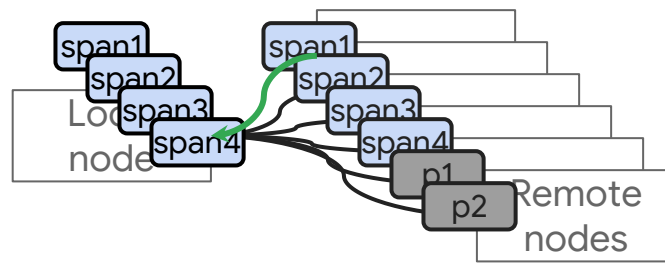
EC-Split (Hydra [1]):
erasure codes individual spans



Multiple network IOs to swap-in/out a span

- Stressing network stack → slow swapping
- Stragglers → high tail latency

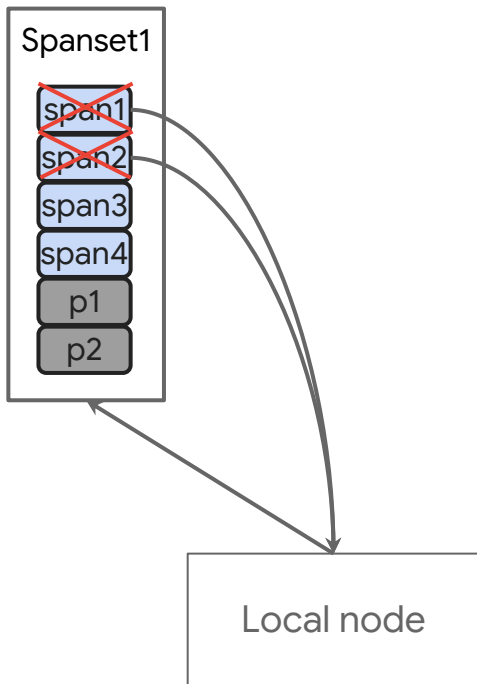
EC-Batch (Carbink):
erasure codes **spansets**



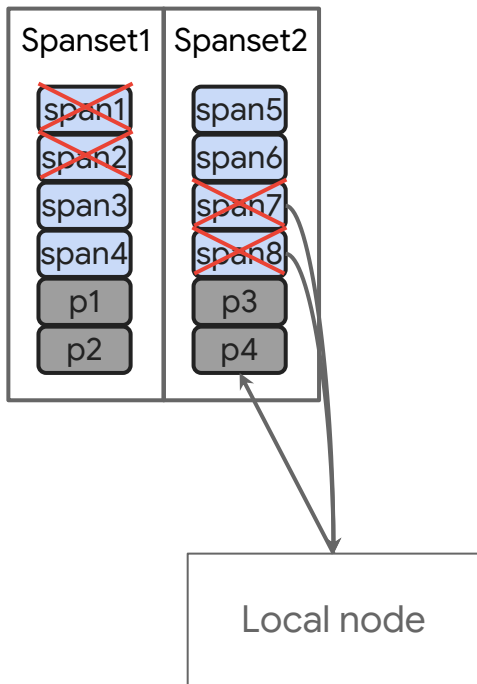
Single network IO to swap-in a span

- Fast swapping and low tail latency

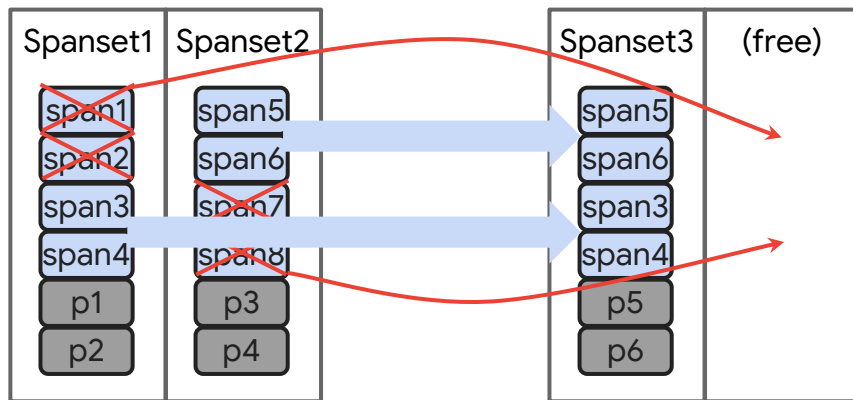
Swap-In&Out Granularity Mismatch → Remote Fragmentation



Swap-In&Out Granularity Mismatch → Remote Fragmentation



Remote Compaction for Defragmentation



No impacts on span swapping perf: off the critical path of swap-ins/outs

Penalty: may consume more memory; dead spans not compacted immediately

Spanset map

~~Spanset1: span1,2,3,4~~

~~Spanset2: span5,6,7,8~~

Spanset3: span5,6,3,4

Zero-copy span merging

Local node

Talk Outline

Direction: in-memory erasure coding for fault tolerance

- High performance & low memory usage

Carbink: making erasure coding work in practice

- Span-centric memory pooling → managing arbitrary-sized objects
- Erasure coding spansets → achieving swapping efficiency

Evaluation: performance and cost of Carbink

Evaluation Overview

Workloads:

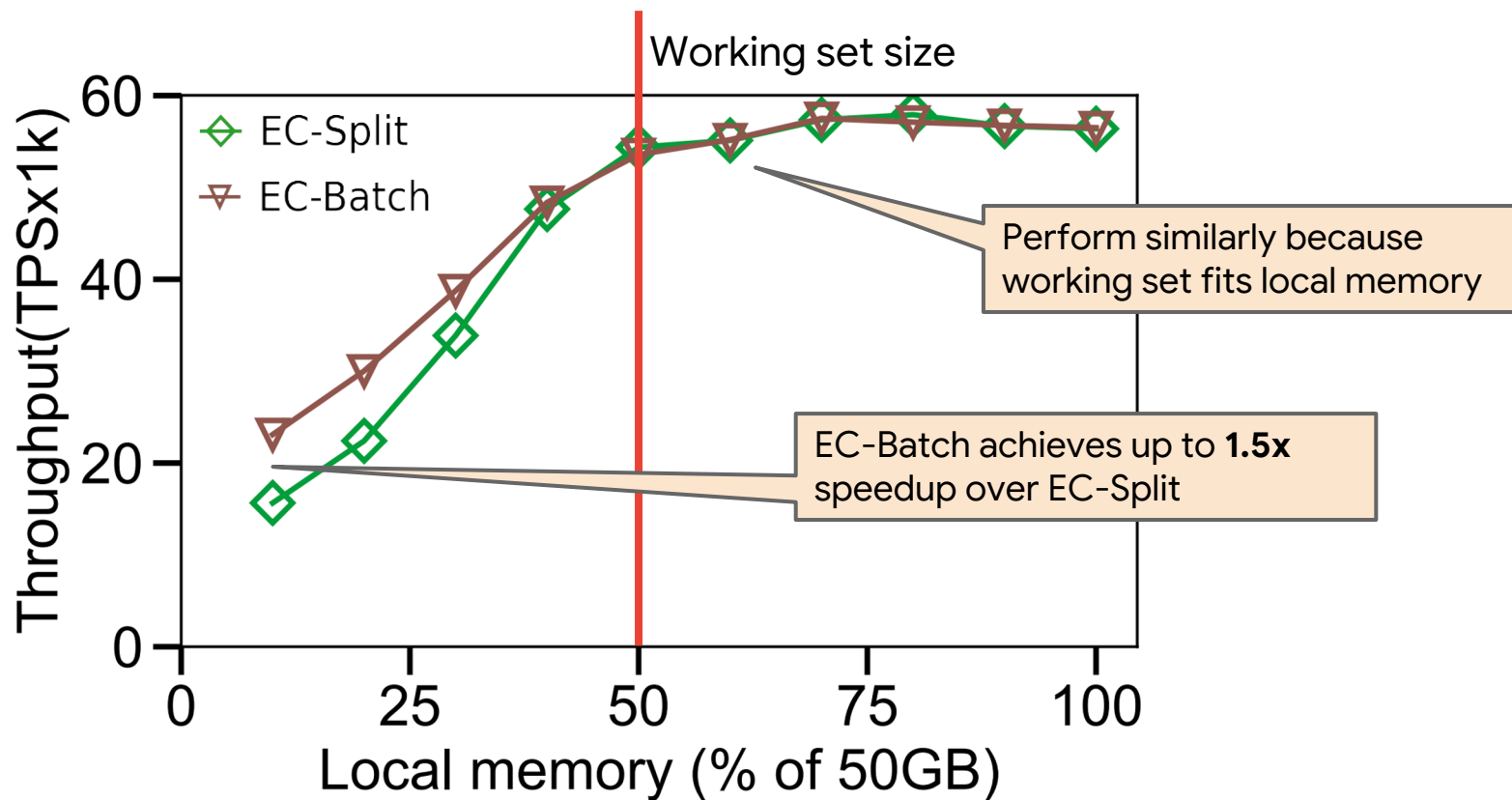
- An internal transactional KV-store doing TPC-A transactions
- Graph connected components (skipped here due to time limit)
- A microbenchmark dereferencing remotable objects

Metrics: throughput, tail latency, memory usage

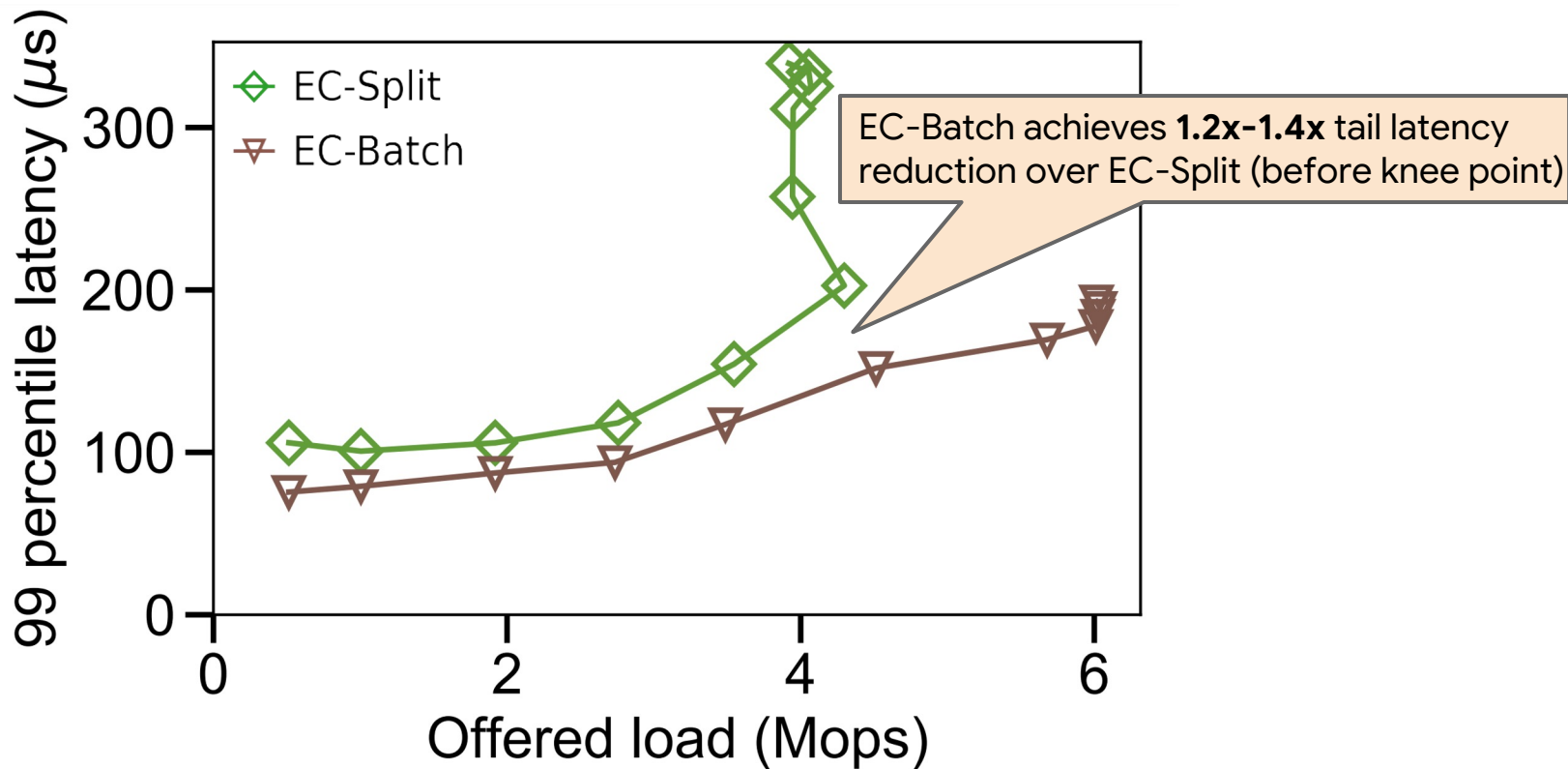
Testbed:

- Servers with 50 Gbps NIC and PonyExpress [1] user-space network stacks
- One-sided RMAs for span swapping; RPCs for remote compaction

Throughput (KV-store)



Tail Latency (Microbenchmark)



Other Results

Remote memory usage:

- EC-Batch consumes at most 35% more memory than EC-Split
- ... but still only $\frac{2}{3}$ of replication memory usage

More in the paper!

- Remote compaction resource usage
- Failure recovery times
- AIFM (swapping individual objects) vs. Carbink

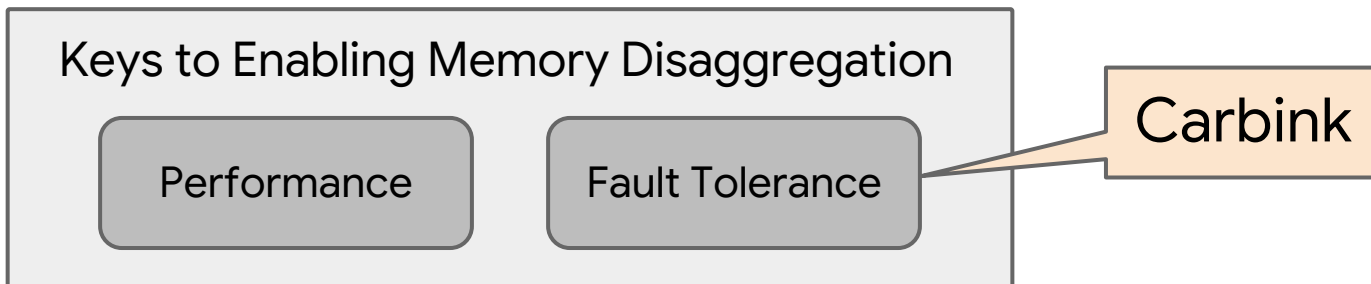
Carbink Summary

Fault tolerance is a must-have feature for applications to use far memory

Carbink: making erasure coding FT work in practice for far memory system

- Grouping objects into spans → handle arbitrary-sized objects
- Erasure coding spansets → single network IO data-fetch

Up to 1.5x application speedup and 1.4x tail latency reduction with up to 35% more memory usage (compared to state-of-the-art EC-Split)



Thank You!

Carbink: making erasure coding FT work in practice for far memory system

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