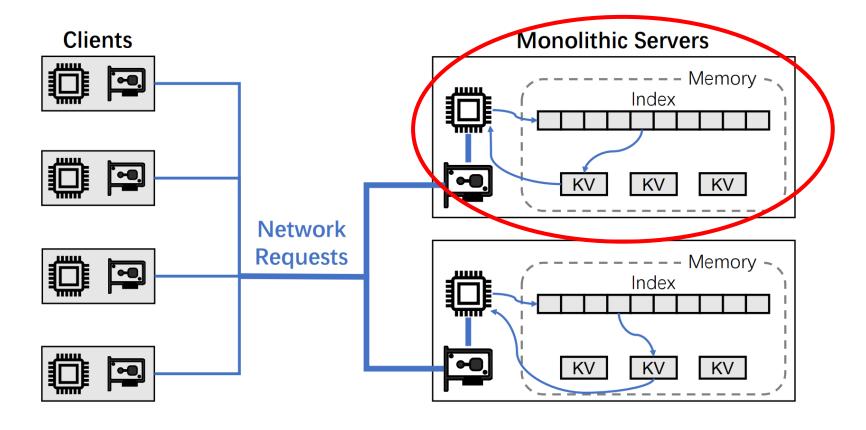
FUSEE: A Fully Memory-Disaggregated Key-Value Store

Jiacheng Shen, Pengfei Zuo, Xuchuan Luo, Tianyi Yang , Yuxin Su, Yangfan Zhou, $\alpha\nu\delta$ Michael . Lyu

Speaker wrl

FAST 2023

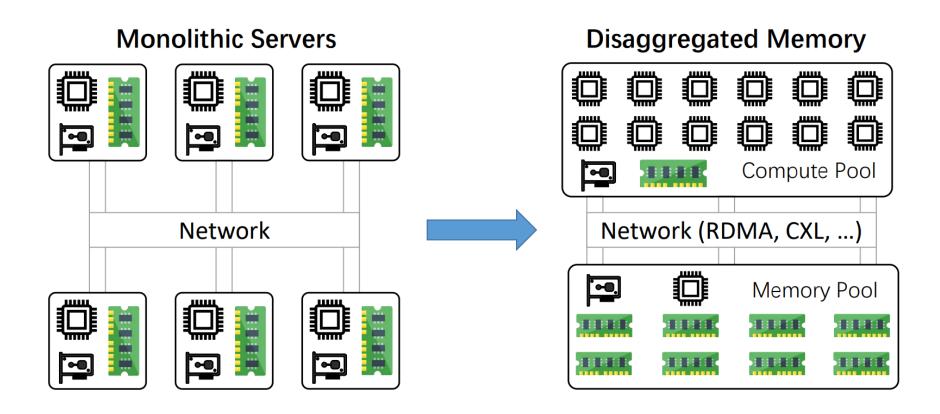
In-Memory Key-Value Stores



Existing Problem:

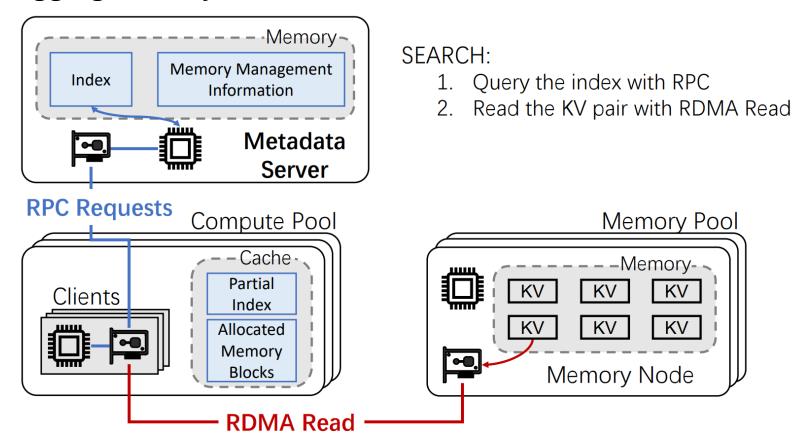
Coupled CPU and Memory will be accompanied by low Resource efficiency and Elasticity.

Disaggregated Memory



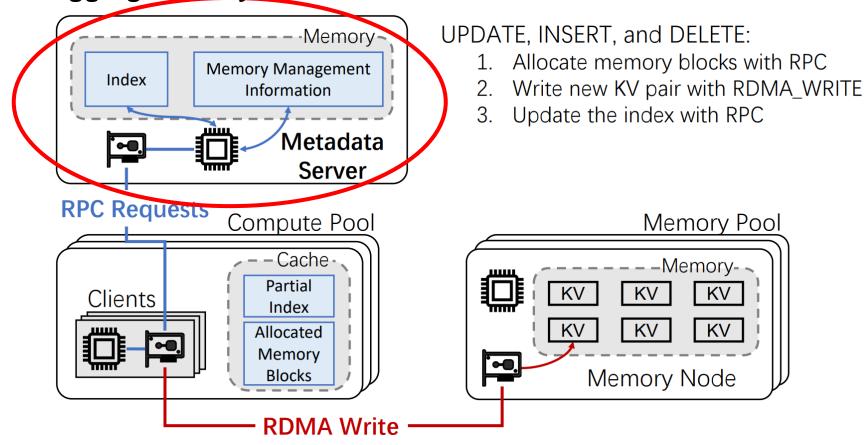
- Compute Pool: Strong computing power and weak storage capacity
- Memory Pool: Weak computing power and Strong storage capacity

Existing Semi-Disaggregated Key-Value Stores



MetadataServer: Used for managing metadata (address indexing)

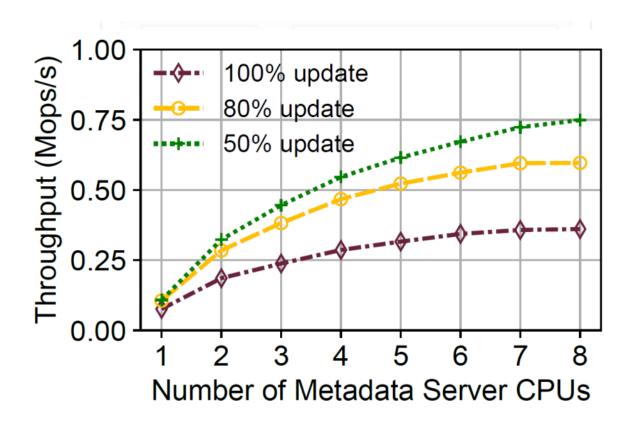
Existing Semi-Disaggregated Key-Value Stores



MetadataServer: Used for managing metadata (address indexing)

Problem

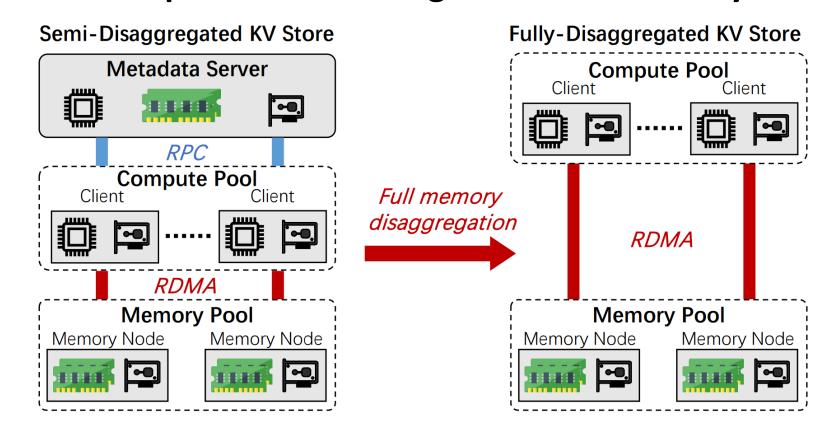
The problem: The metadata management still relies on monolithic servers, cannot fully exploit the resource efficiency and elasticity benefits of DM



Additional resource required:
At least 6 CPU cores and 1 RNIC

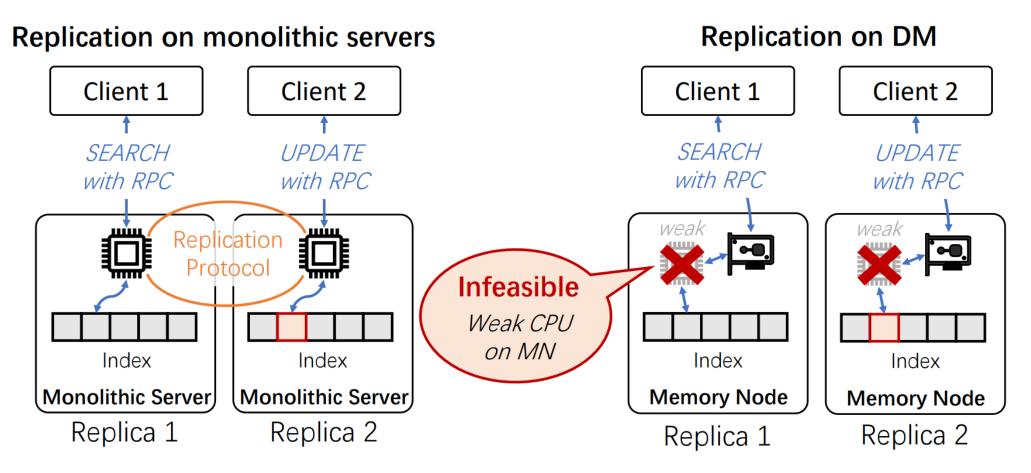
Motivation&Key Idea

- The motivation: Can the metadata server be removed to achieve complete decoupling?
- The key-idea: Is it possible to manage metadata directly with clients?



Challenge 1: Client-Centric Index Replication

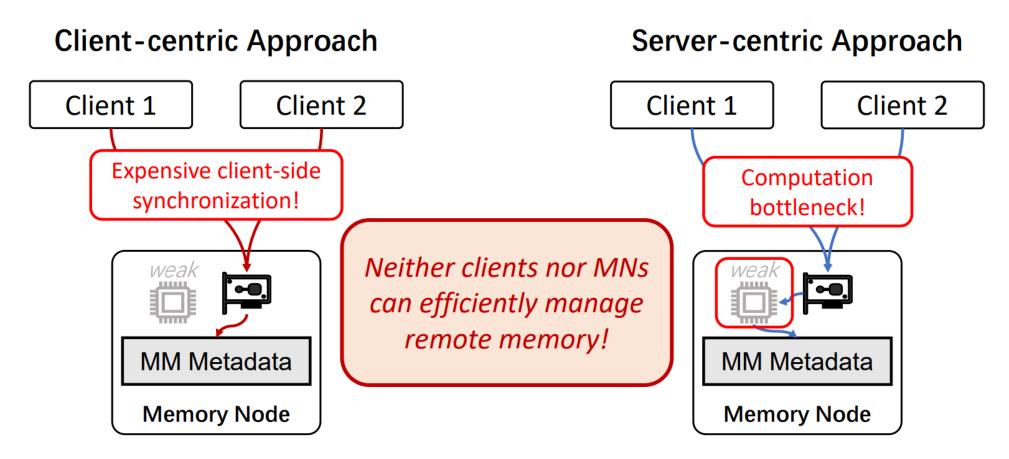
- Replication on monolithic servers: Replication protocols
- Replication on DM: Weak compute power



Weak MN-side compute power cannot execute replication protocols

Challenge 2: Remote Memory Allocation

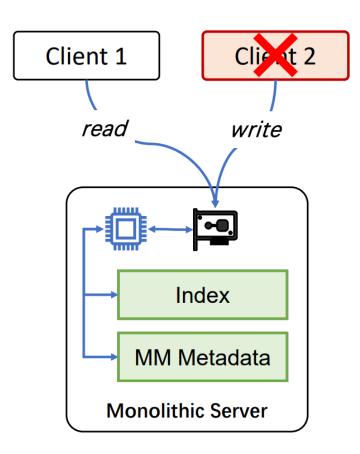
- Client-centric Approach: compute on Client
- Server-centric Approach: compute on Server



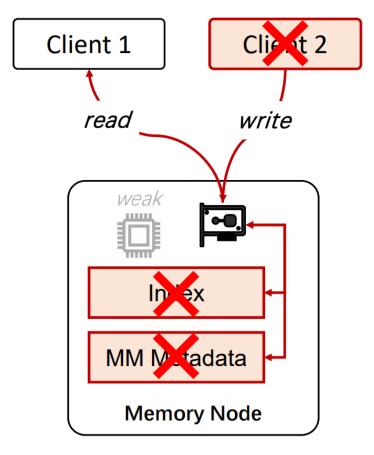
➤ Weak MN-side compute power cannot efficiently allocate memory spaces

Challenge 3: Metadata Corruption

- monolithic-server-based KV stores:
- CPUs of the KV store server exclusively access and modify metadata

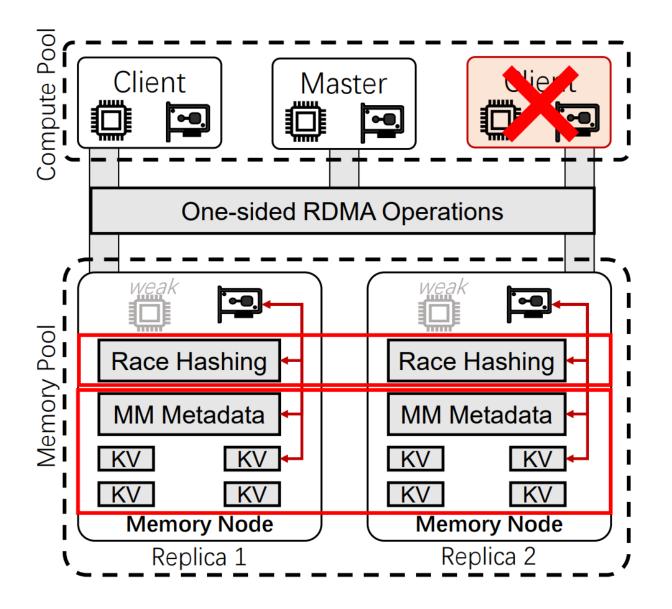


- > fully memory-disaggregated KV stores:
- Clients directly access and modify metadata with one-sided RDMA verbs



> Client-failures compromises the correctness of the entire KV store

FUSEE Overview

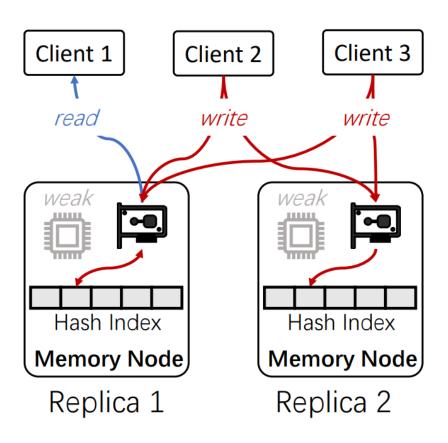


➤ Metadata corruption: Embedded operation log

- Client-centric index replication:
 The SNAPSHOT protocol
- Remote memory allocation:
 Two-level memory allocation



Index Replication on DM

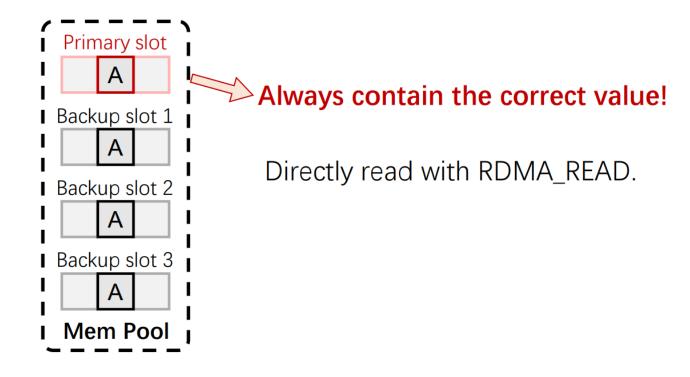


Main problem

- How to protect readers from reading incomplete writes?
- > How to efficiently resolve write-write conflicts among clients?

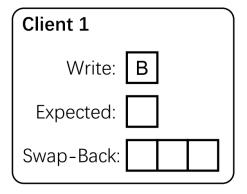
Design —Client-Centric Index Replication

- > How to protect readers from reading incomplete values?
 - Separate index replicas into primary and backups
 - Resolve write-write conflicts on backup replicas



Design —Client-Centric Index Replication

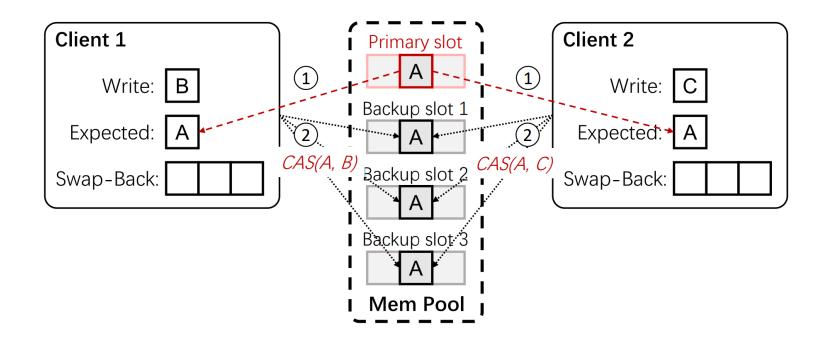
- > How to efficiently resolve write-write conflicts on clients?
 - Out-of-place KV modification -> Conflict clients write different values
 - Last-writer-wins conflict resolution -> Simple rules on client sides



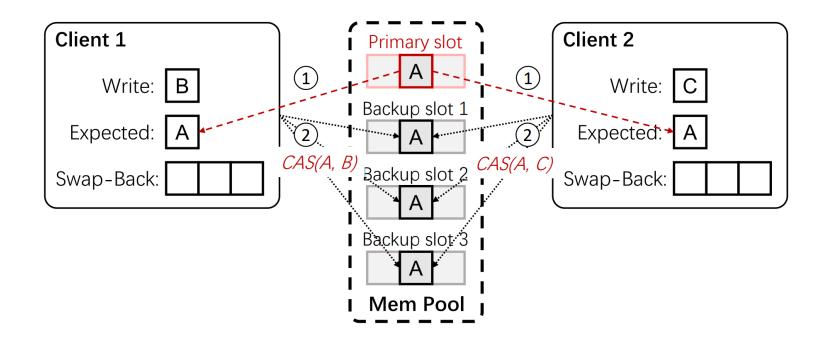
/	Prim	nary	slot	` \ 	
l I		Α		ı	
Backup slot 1					
I I		Α			
[Back	up s	lot 2	· ·	
¦ 		Α			
 -	3ack	up s	lot 3	}	
i		Α			
L	Mem Pool				

Client 2	
Write:	С
Expected:	
Swap-Back:	

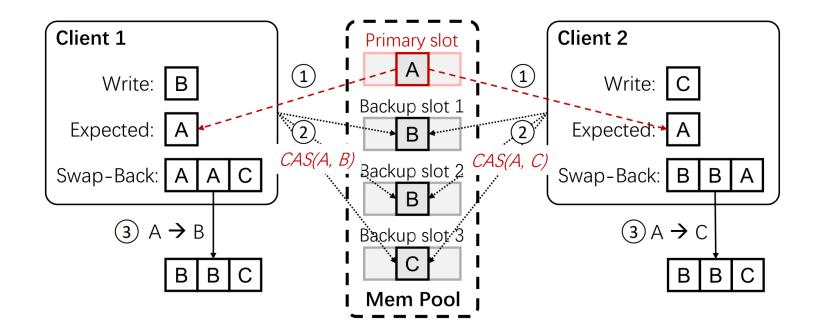
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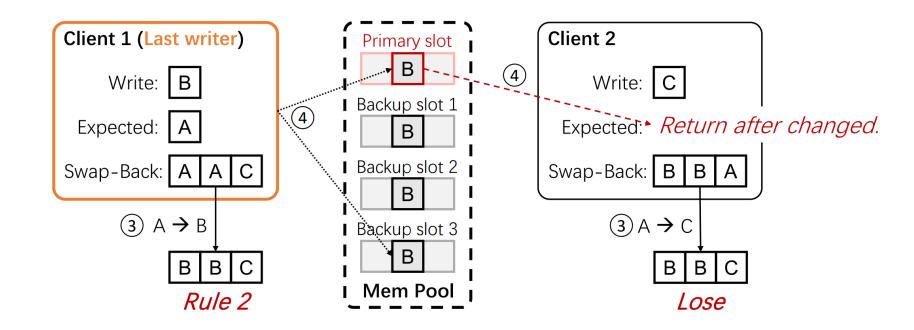


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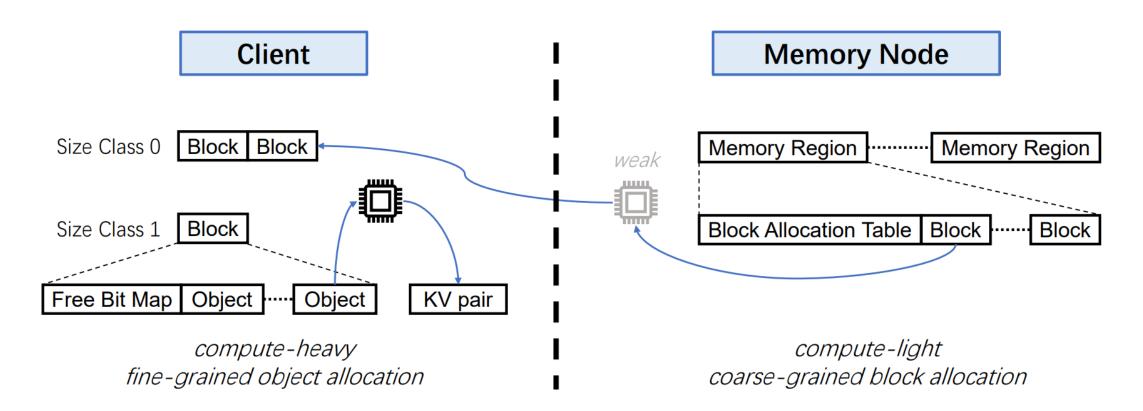
Design —Client-Centric Index Replication

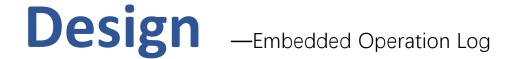
- > A client is a last writer if:
- Successfully modified all backup slots
- Or, modified a majority of backup slots
- Or, writes the smallest value when there is no majority



➤ Challenge: Where to execute memory management logic?

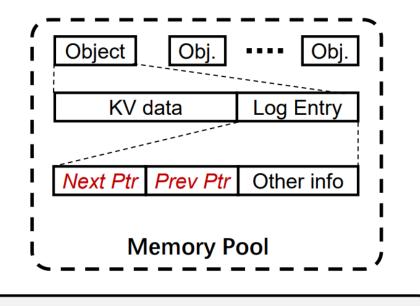
Key idea: Separate into compute-light and compute-heavy tasks



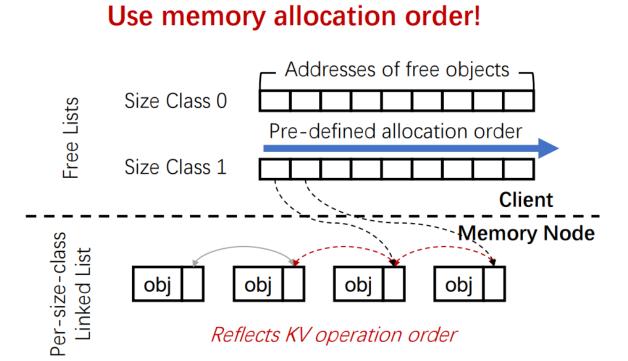


> Challenge: An additional RTT to create a log entry before operation starts

Key idea: Embed operation logs into KV pairs to reduce the additional RTT



How to construct operation sequence?



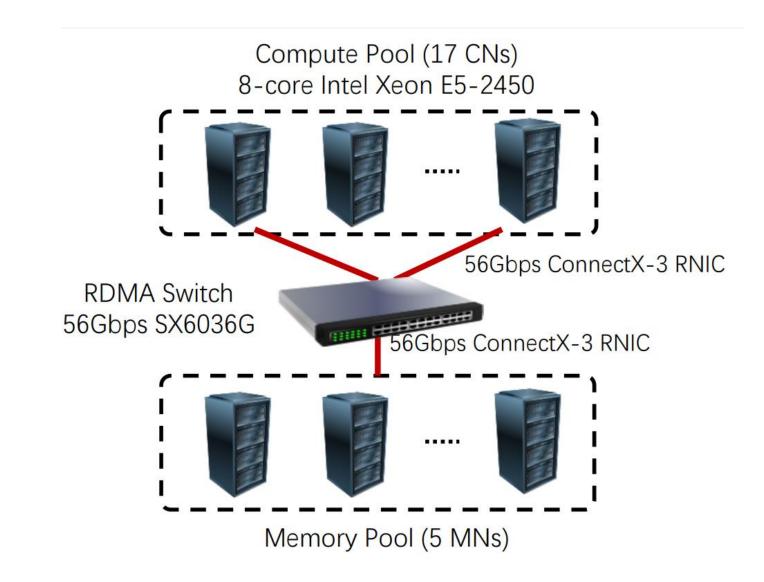
Evaluation

Benchmarks:

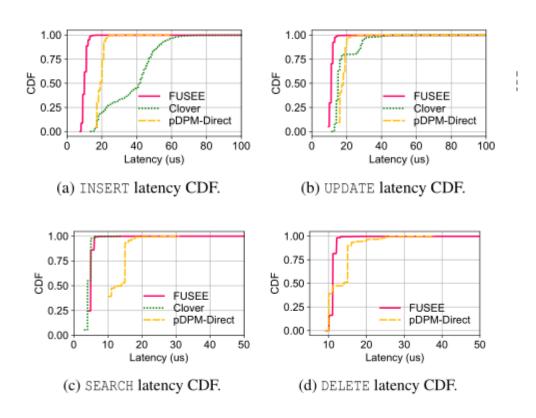
- YCSB hybrid benchmarks
- Microbenchmark

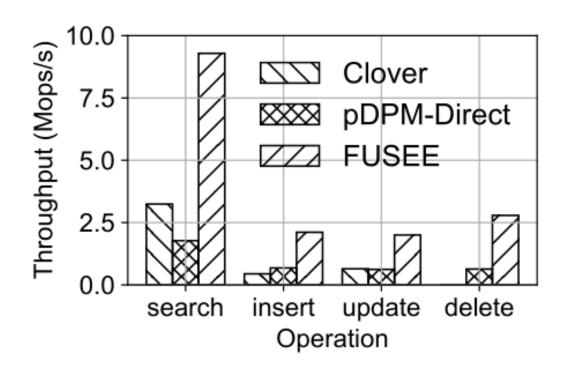
> Comparisons:

- Clover [ATC 20]
- pDPM-Direct [ATC 20]
- FUSEE-NC (no cache)



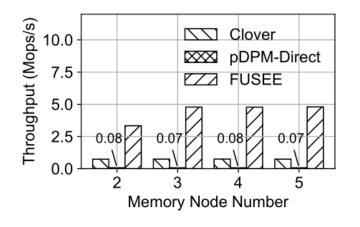
Evaluation —Latency&Throughput

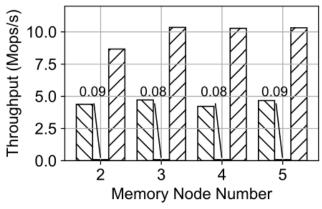


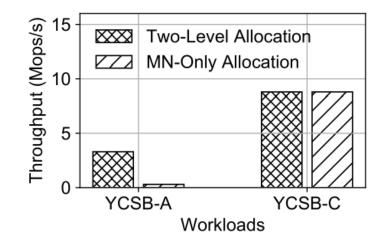


- FUSEE performs the best on INSERT and UPDATE, since the SNAPSHOT replication protocol has bounded RTTs.
- FUSEE has a little higher SEARCH latency than Clover since FUSEE reads the hash index and the KV pair in a single RTT, which is slower than only reading the KV pair in Clover.
- FUSEE has the Highest Throughput.

Evaluation —YCSB-Throughput & Two-level memory allocation performance.







(a) YCSB-A throughput.

(b) YCSB-C throughput.

- As for FUSEE, the throughput improves as the number of memory nodes increases from 2 to 3. There is no further throughput improvement because the total throughput is limited by the number of compute nodes.
- The YCSB-A throughput drops 90.9% due to the limited compute power on MNs, while the YCSB-C throughput remains the same since no memory allocation is involved in the read-only setting.

Evaluation —Recover from Crashed Clients.

Table 1: Client recovery time breakdown.

Step	Time (ms)	Percentage
Recover connection & MR	163.1	92.1%
Get Metadata	0.3	0.2%
Traverse Log	3.5	2.0%
Recover KV Requests	3.5	2.0%
Construct Free List	6.6	3.7%
Total	177.0	100%

• The log traversal and KV request recovery only account for 4% of the recovery time, which implies the affordable overhead of log traversal.

Conclusion

