

PinK: High-speed In-storage Key-value Store with Bounded Tails

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Key-Value Store is Everywhere!

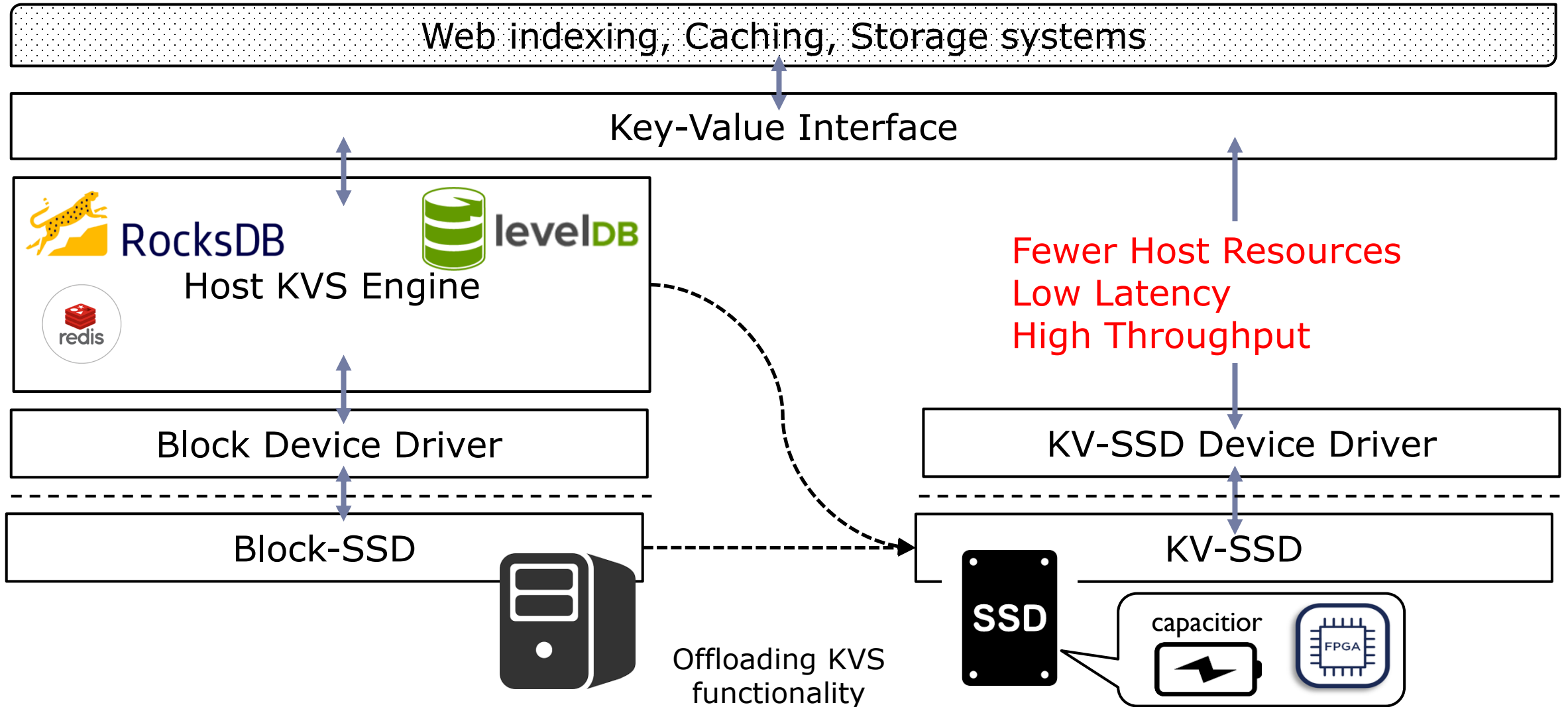
- ▶ Key-Value store (KVS) has become a necessary infrastructure

Web indexing, Caching, Storage systems

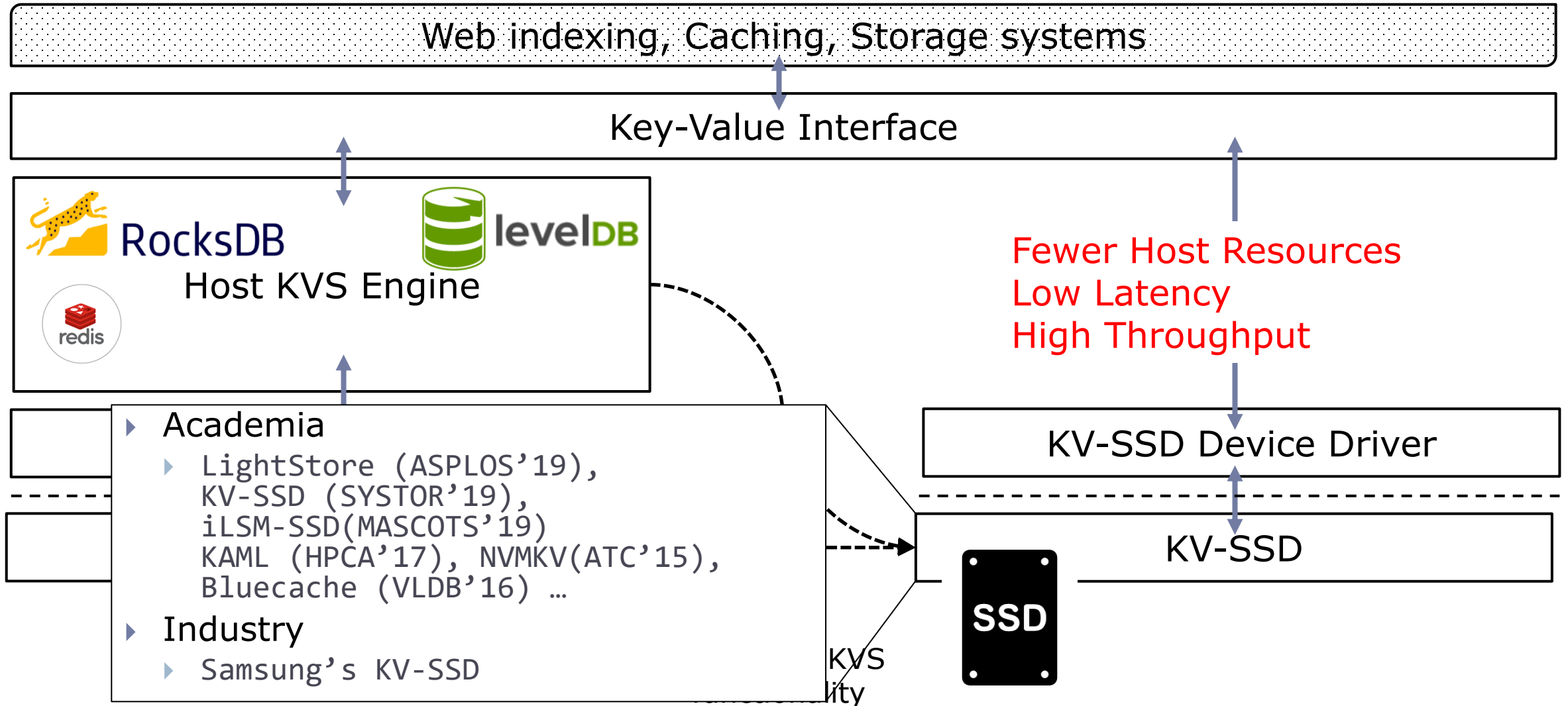


- ▶ Algorithm
 - ▶ SILK (ATC'19),
 - ▶ Dostoevsky (SIGMOD'18)
 - ▶ Monkey (SIGMOD'17) ...
- ▶ System
 - ▶ FlashStore (VLDB'10)
 - ▶ Wiskey (FAST'16)
 - ▶ LOCS (Eurosys'14) ...
- ▶ Architecture
 - ▶ Bluecache (VLDB'16) ...

Key-Value (KV) Storage Device



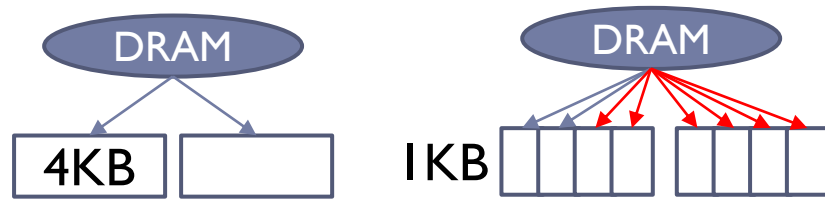
Key-Value (KV) Storage Device



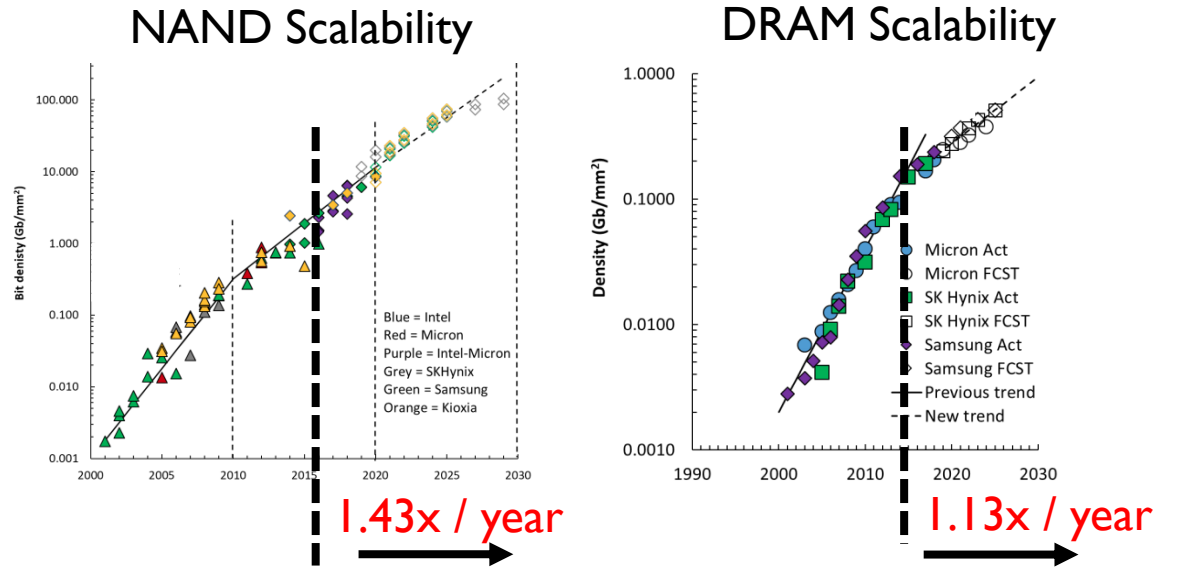
Key Challenges of Designing KV-SSD

▶ 1. Limited DRAM resource

- ▶ SSDs usually have DRAM as much as 0.1% of NAND for indexing!
 - ▶ Logical block: 4KB > KV-pair: 1KB on average



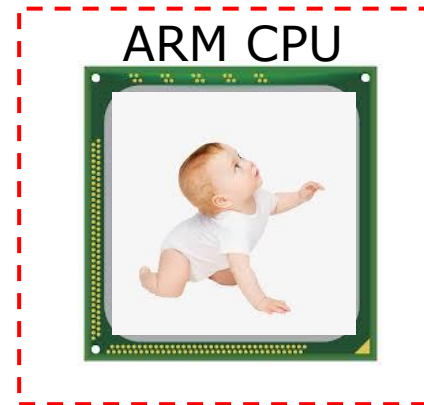
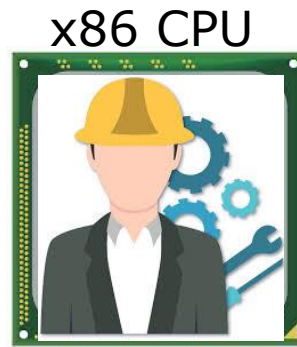
- ▶ DRAM scalability slower than NAND!



Technology and Cost Trends at Advanced Nodes, 2020,
<https://semiwiki.com/wp-content/uploads/2020/03/Lithovision-2020.pdf>

Key Challenges of Designing KV-SSD (Cont.)

- ▶ 2. Limited CPU performance
 - ▶ SSDs have low power CPU (ARM based)

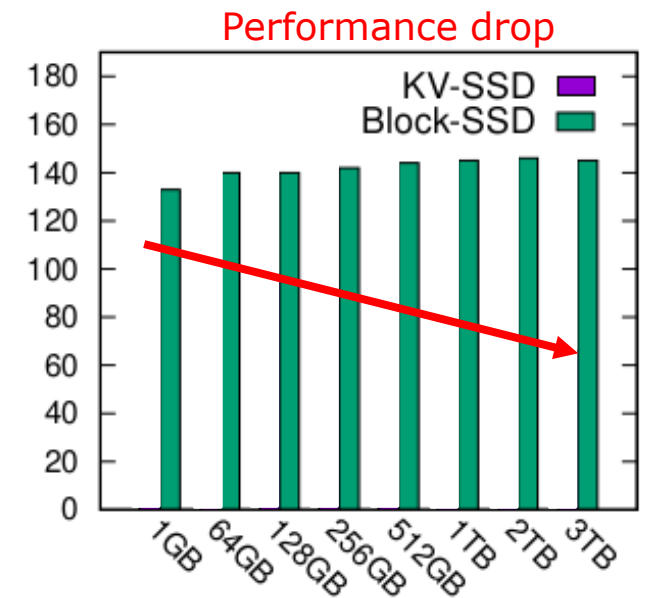
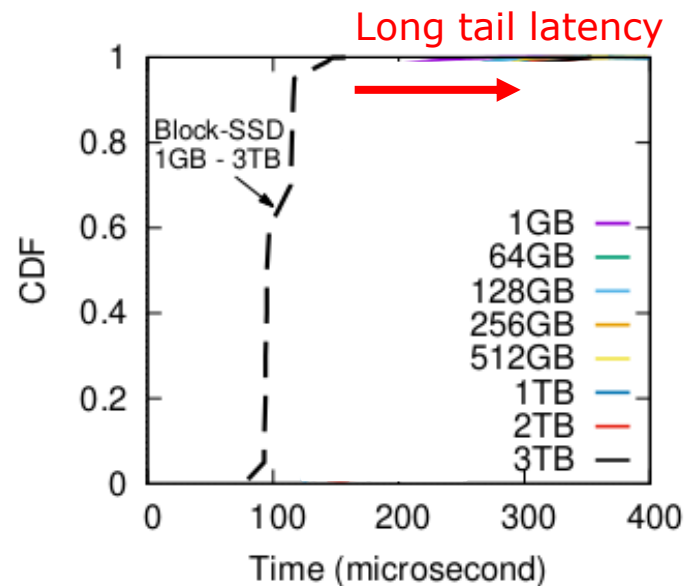


Which algorithm is better for KV-SSD with these limitations,
Hash or Log-structured Merge-tree (LSM-tree)?

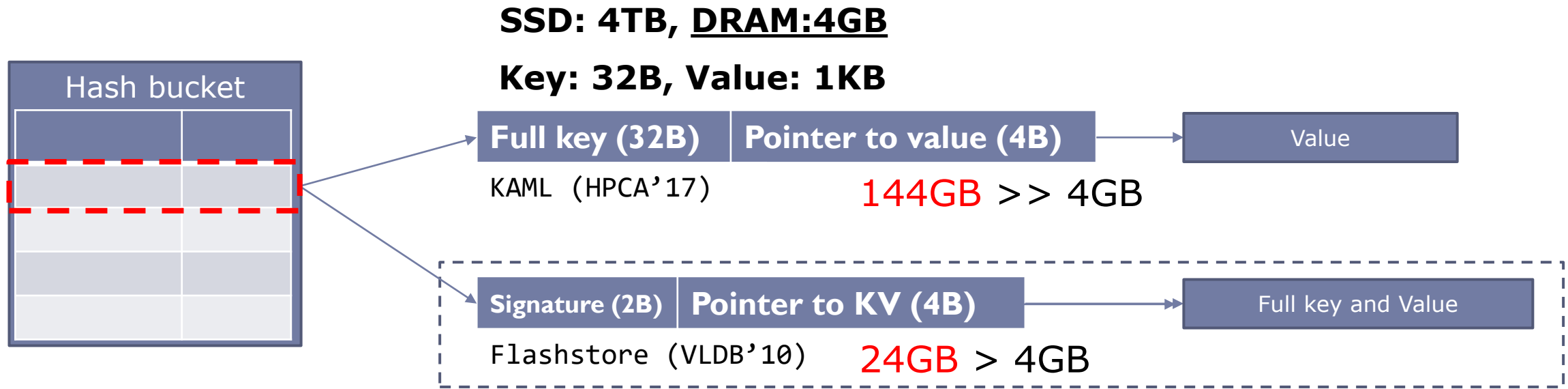
Experiments using Hash-based KV-SSD

- ▶ Samsung KV-SSD prototype
 - ▶ hash-based KV-SSD*
- ▶ Benchmark
 - ▶ KV-SSD: KV Bench**, 32B key and 1KB value read request
 - ▶ Block-SSD: FIO, 1KB read request

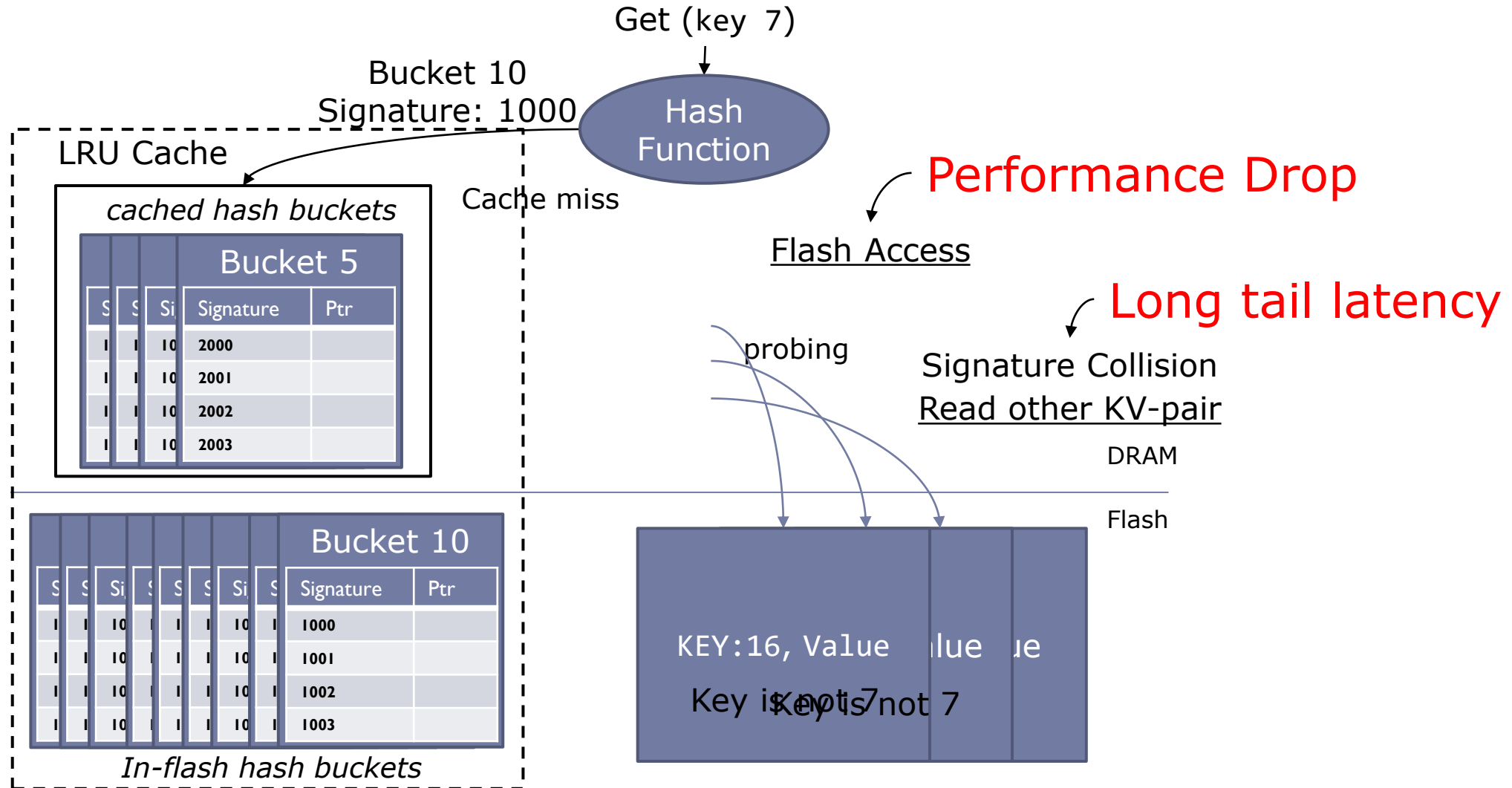
What is the reason?



Problem of Hash-based KV-SSD



Problem of Hash-based KV-SSD



LSM-tree?

- ▶ Another Option “LSM-tree”
 - ▶ Low DRAM requirement
 - ▶ No collision
 - ▶ Easy to serve range query

Is the LSM-tree really good enough?

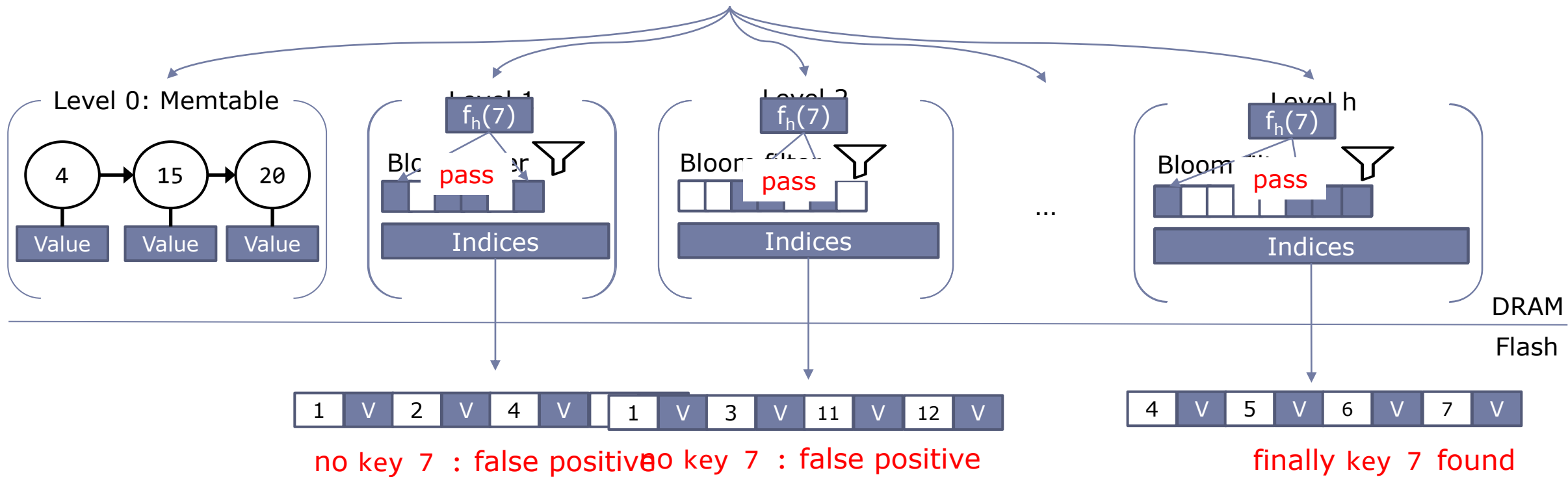
Problem of LSM-tree-based KV-SSD

▶ 1. Long tail latency!

In the worst case,
 $h-1$ flash accesses for 1 KV
 (h = height of LSM-tree)

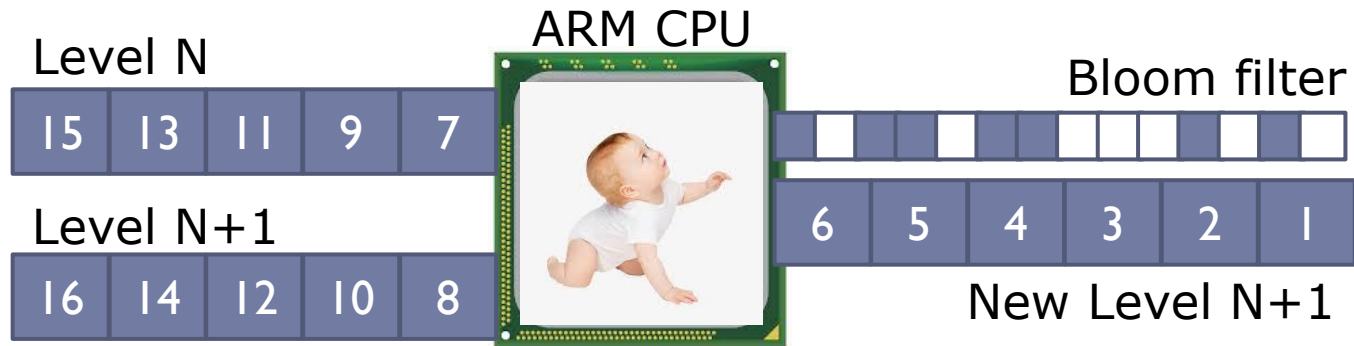


Get (key 7)



Problem of LSM-tree-based KV-SSD

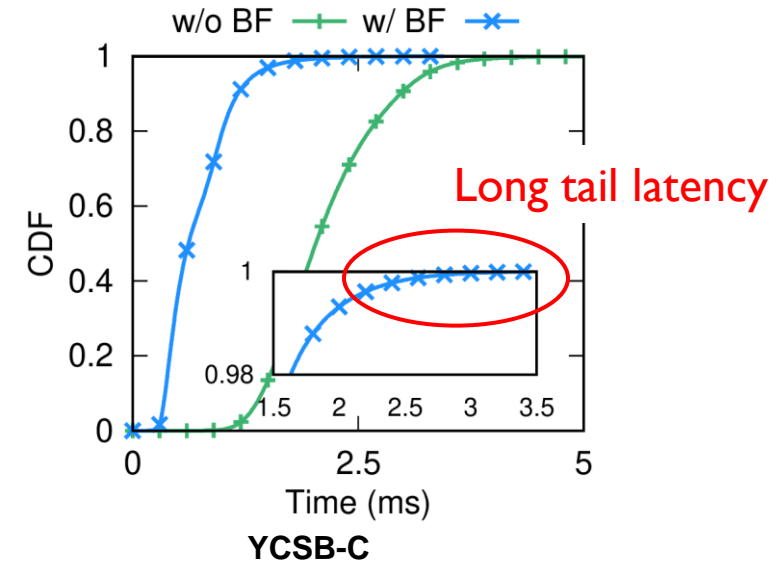
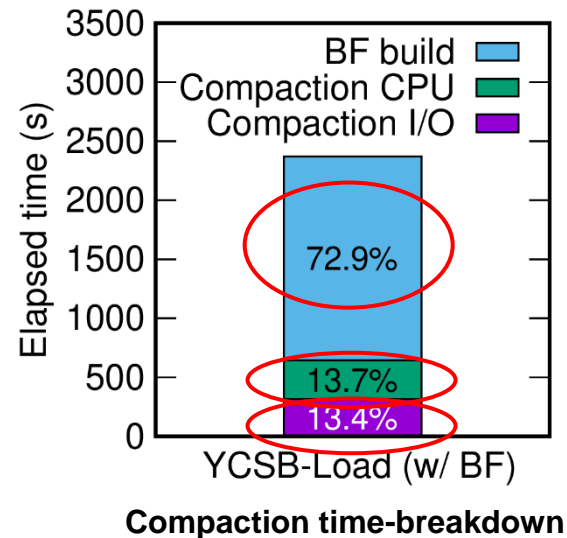
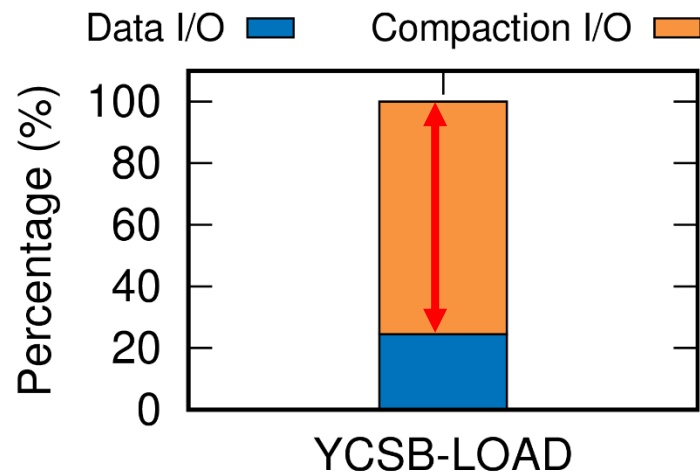
- ▶ 2. CPU overhead!
 - ▶ Merge sort in compaction
 - ▶ Building bloom filters



- ▶ 3. I/O overhead!
 - ▶ Compaction I/O added by LSM-tree

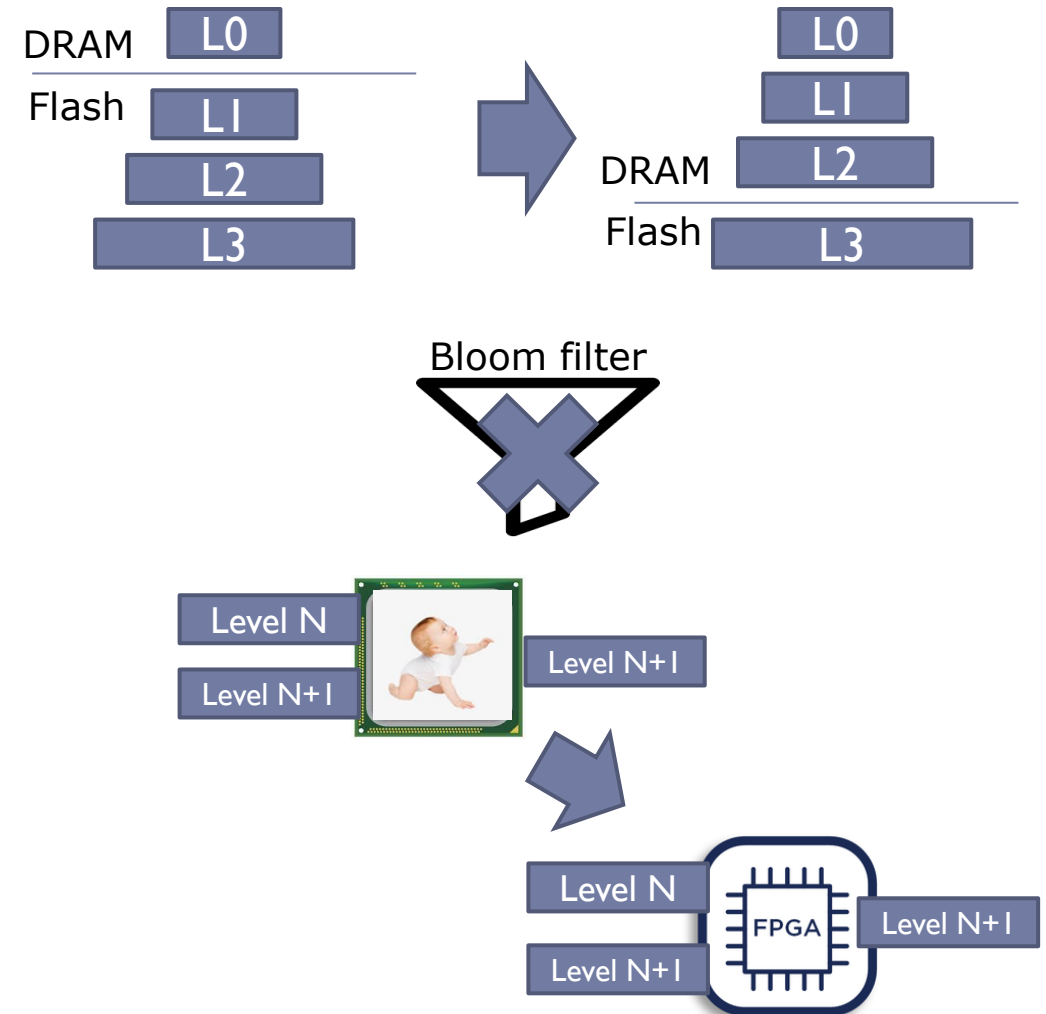
Experiments using LSM-tree-based KV-SSD

- ▶ Lightstore*: LSM-tree-based KV-SSD
 - ▶ Key-value separation (Wisckey**) and Bloom filter (Monkey***)
- ▶ Benchmark
 - ▶ Lightstore: YCSB-LOAD and YCSB-C (Read only), 32B key and 1KB value



PinK : New LSM-tree-based KV-SSD

- ▶ Long tail latency?
 - ▶ Using “Level-pinning”
- ▶ CPU overhead?
 - ▶ “No Bloom filter”
 - ▶ “HW accelerator” for compaction
- ▶ I/O overhead?
 - ▶ Reducing compaction I/O by level-pinning
 - ▶ Optimizing GC by reinserting valid data to LSM-tree



Introduction

PinK

Overview of LSM-tree in PinK

Bounding tail latency

Memory requirement

Reducing search overhead

Reducing compaction I/O

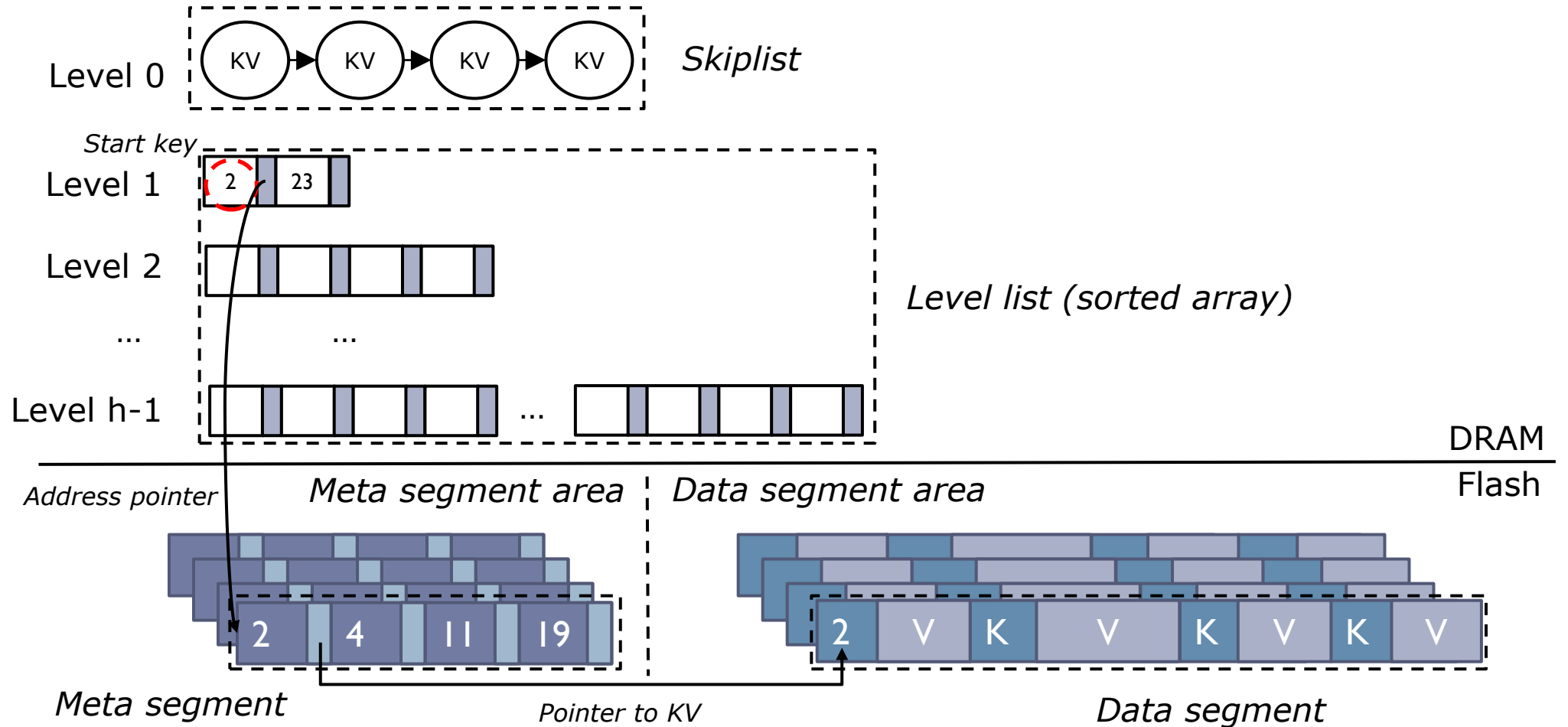
Reducing sorting time

Experiments

Conclusion

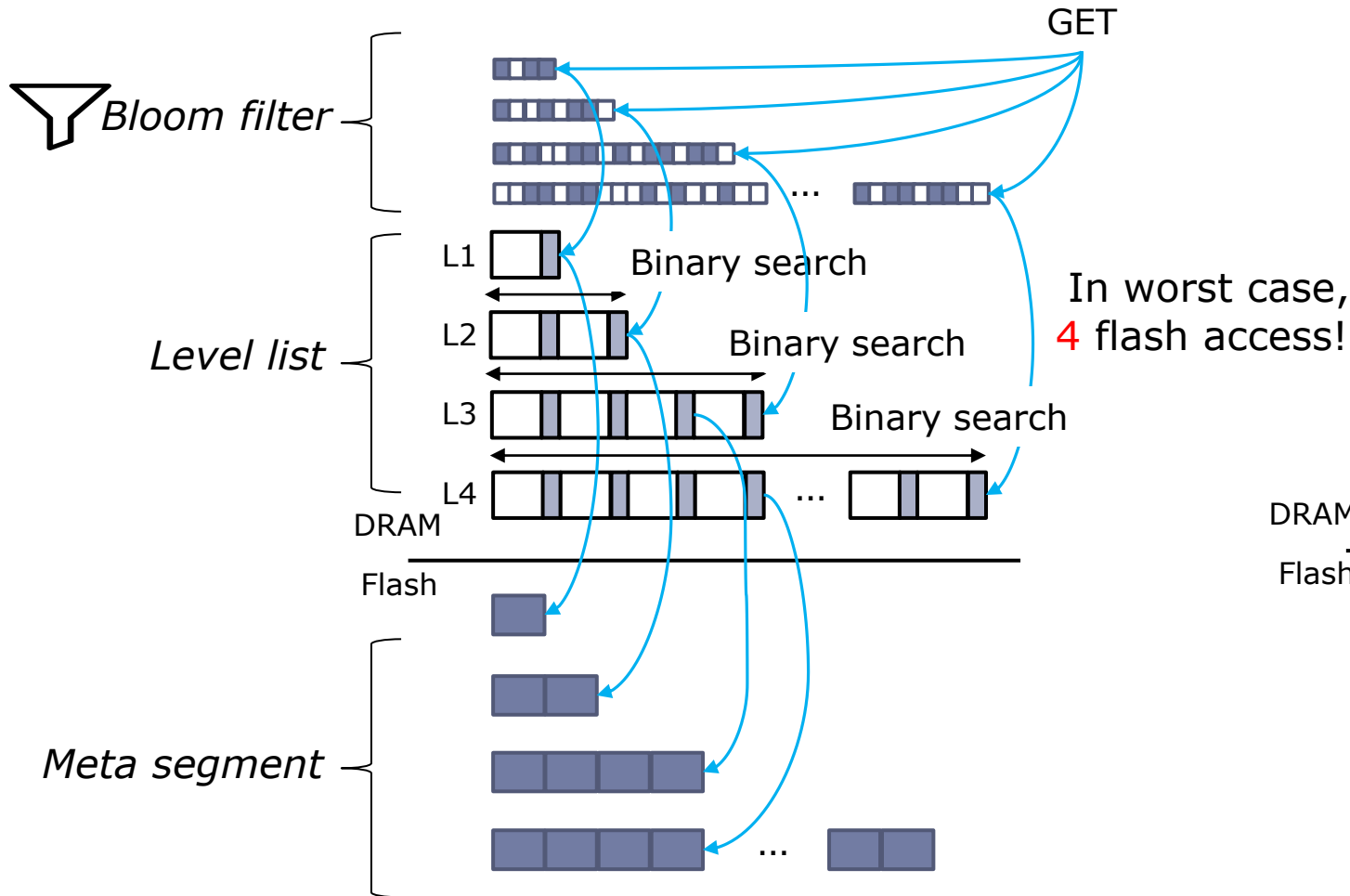
Overview of LSM-tree in PinK

- ▶ PinK is based on key-value separated LSM-tree

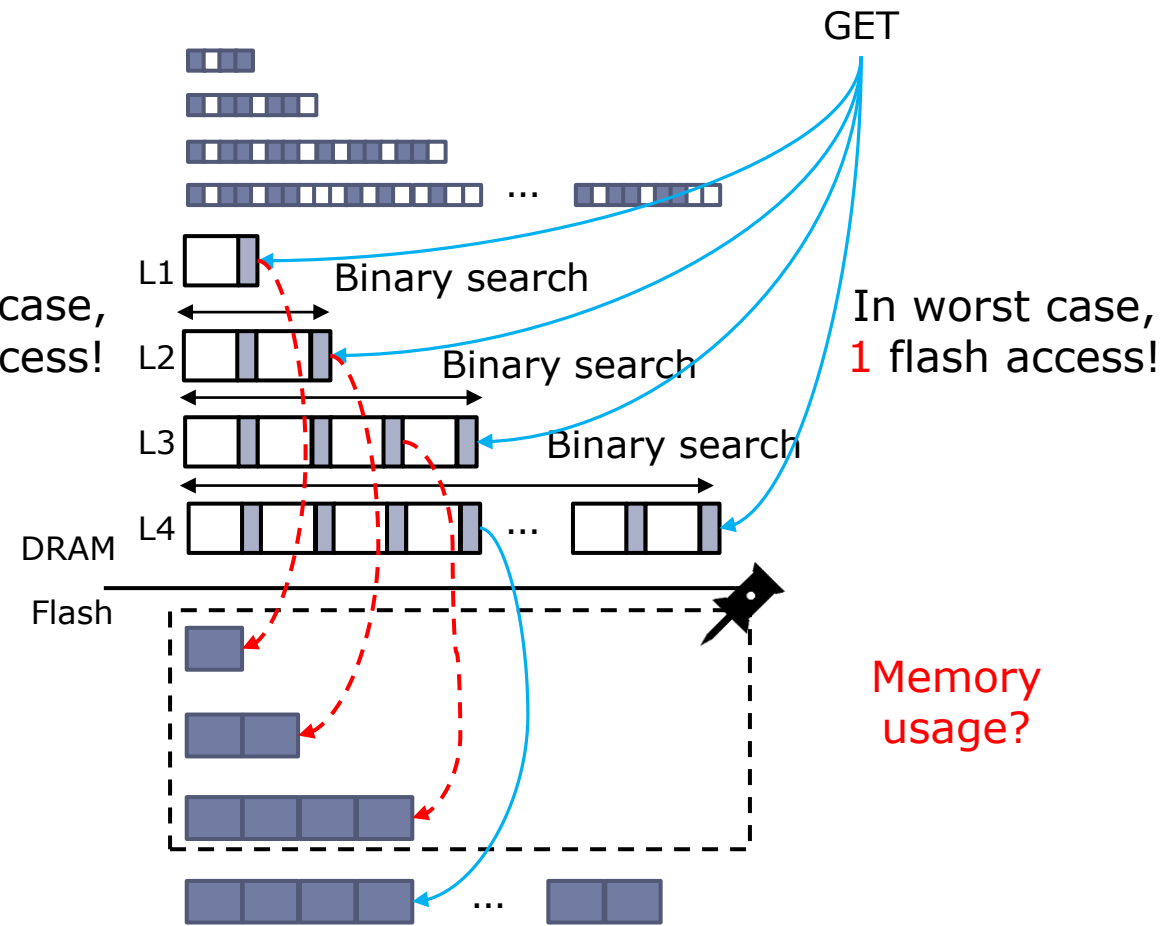


Bounding Tail Latency

LSM-tree with bloom filter

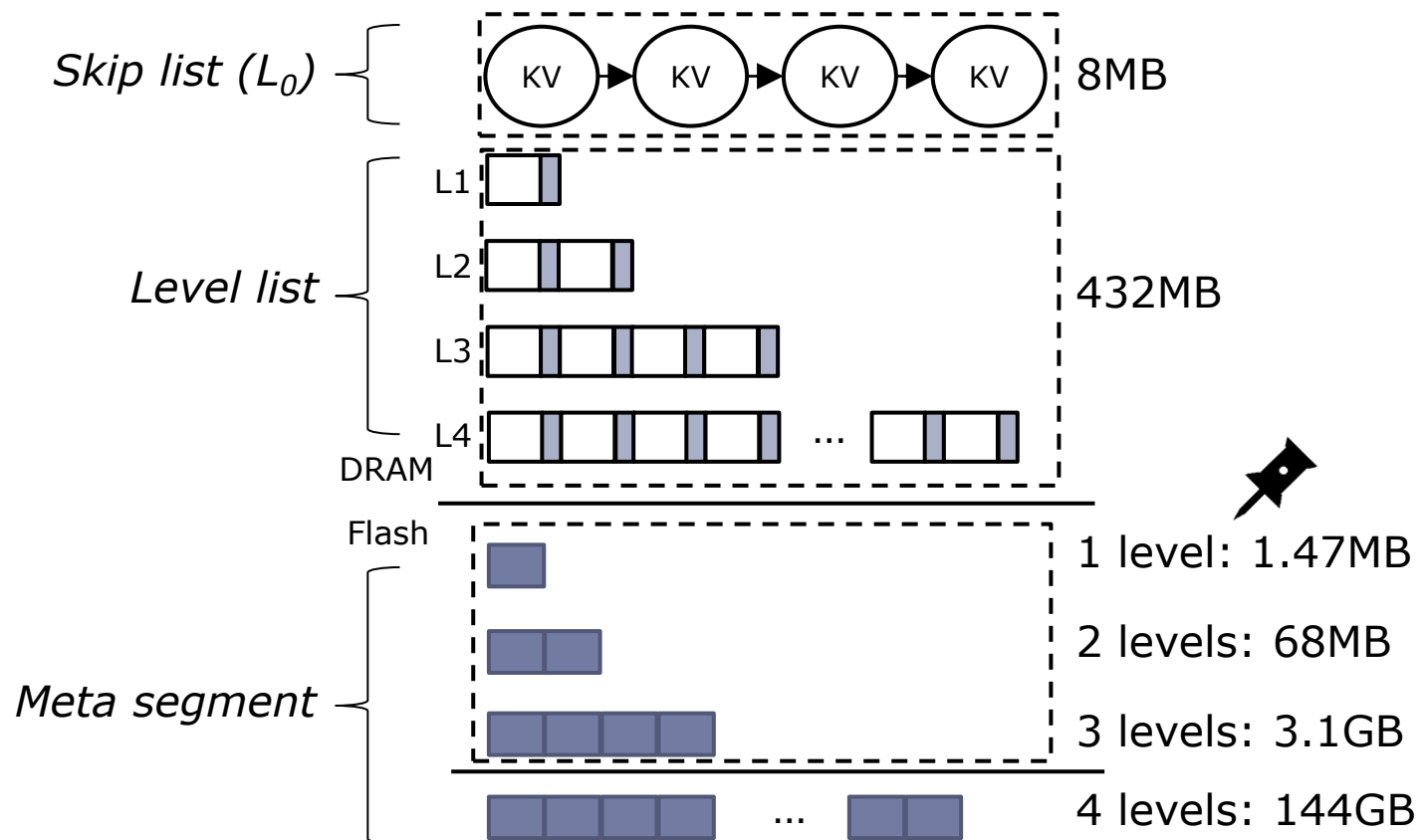


PinK



Memory Requirement

- ▶ 4TB SSD, 4GB DRAM (32B key, 1KB value)
 - ▶ Total # of levels: 5

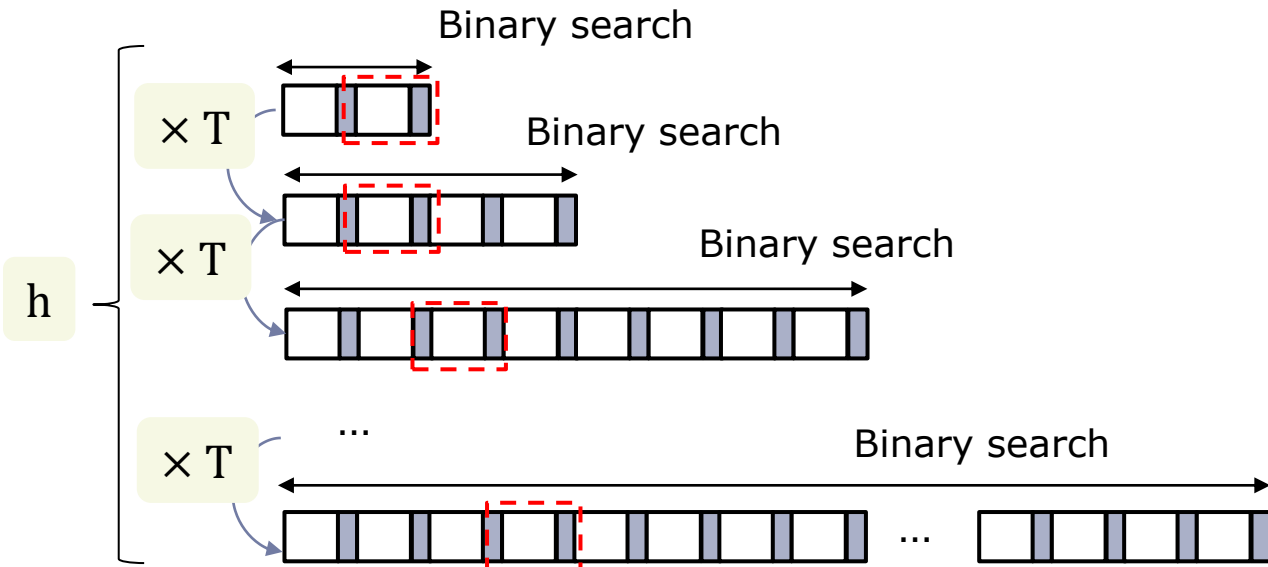


3.5 GB < 4GB

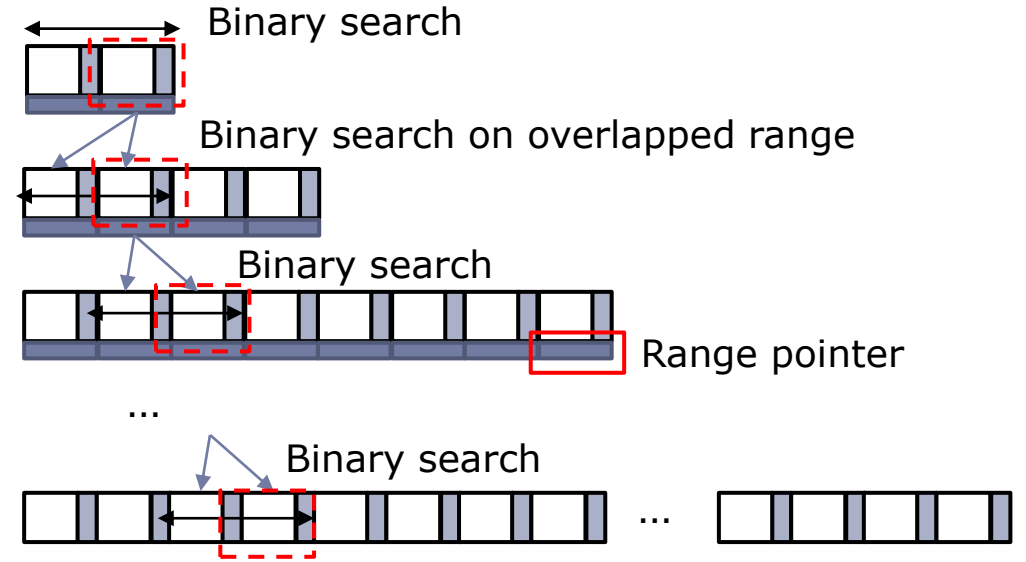
Only **one** flash access for indexing

Reducing Search Overhead

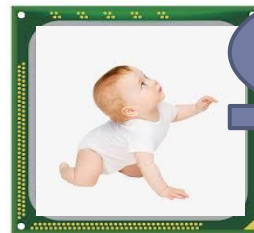
▶ Fractional cascading



search complexity is $O(h^2 \log(T))$



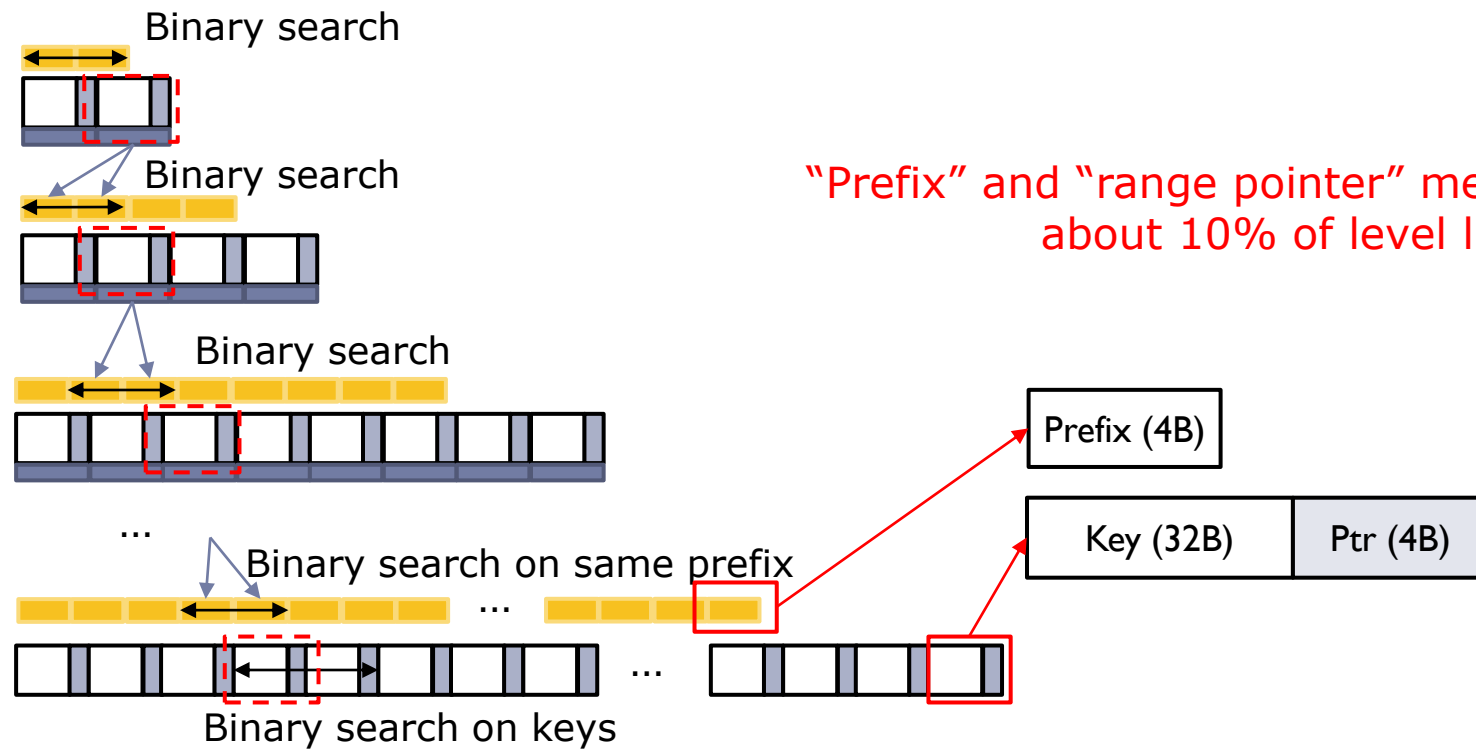
$O(h \log(T))$



Burdensome!

Reducing Search Overhead

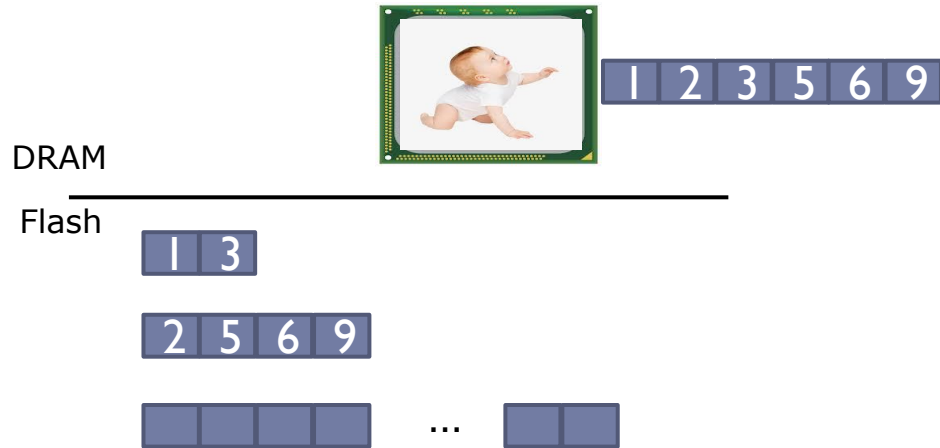
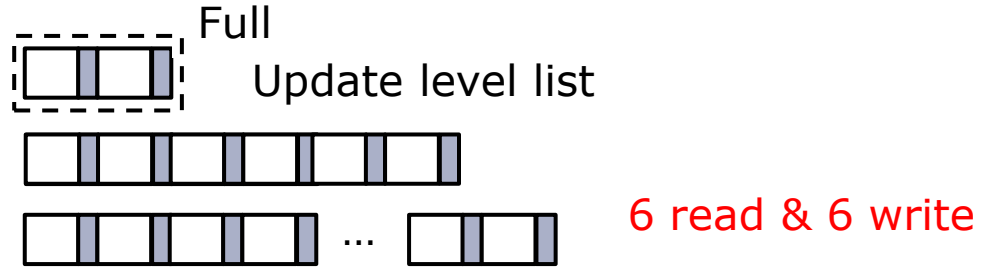
- ▶ Prefix
 - ▶ Less compare overhead
 - ▶ Cache efficient search



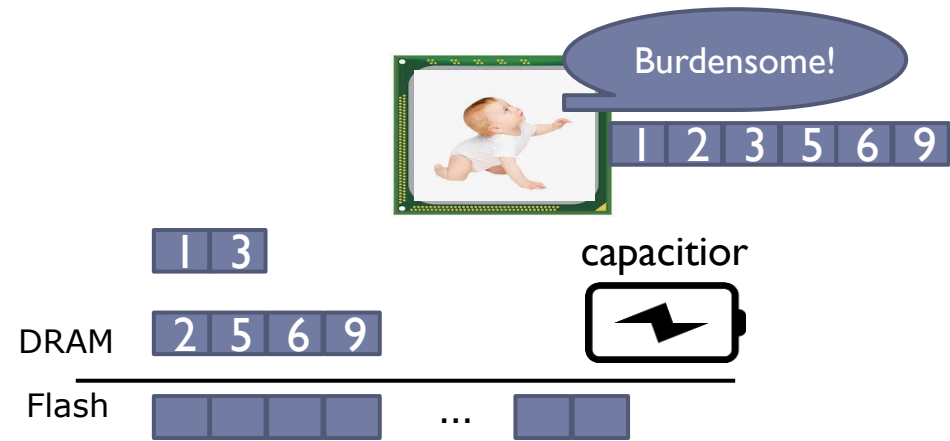
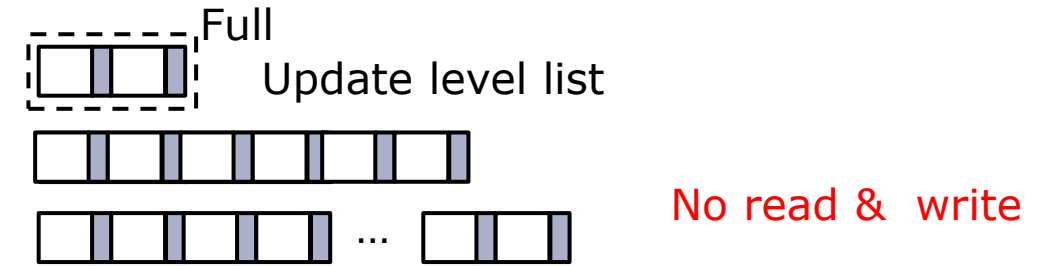
“Prefix” and “range pointer” memory usage:
about 10% of level list

Reducing Compaction I/O

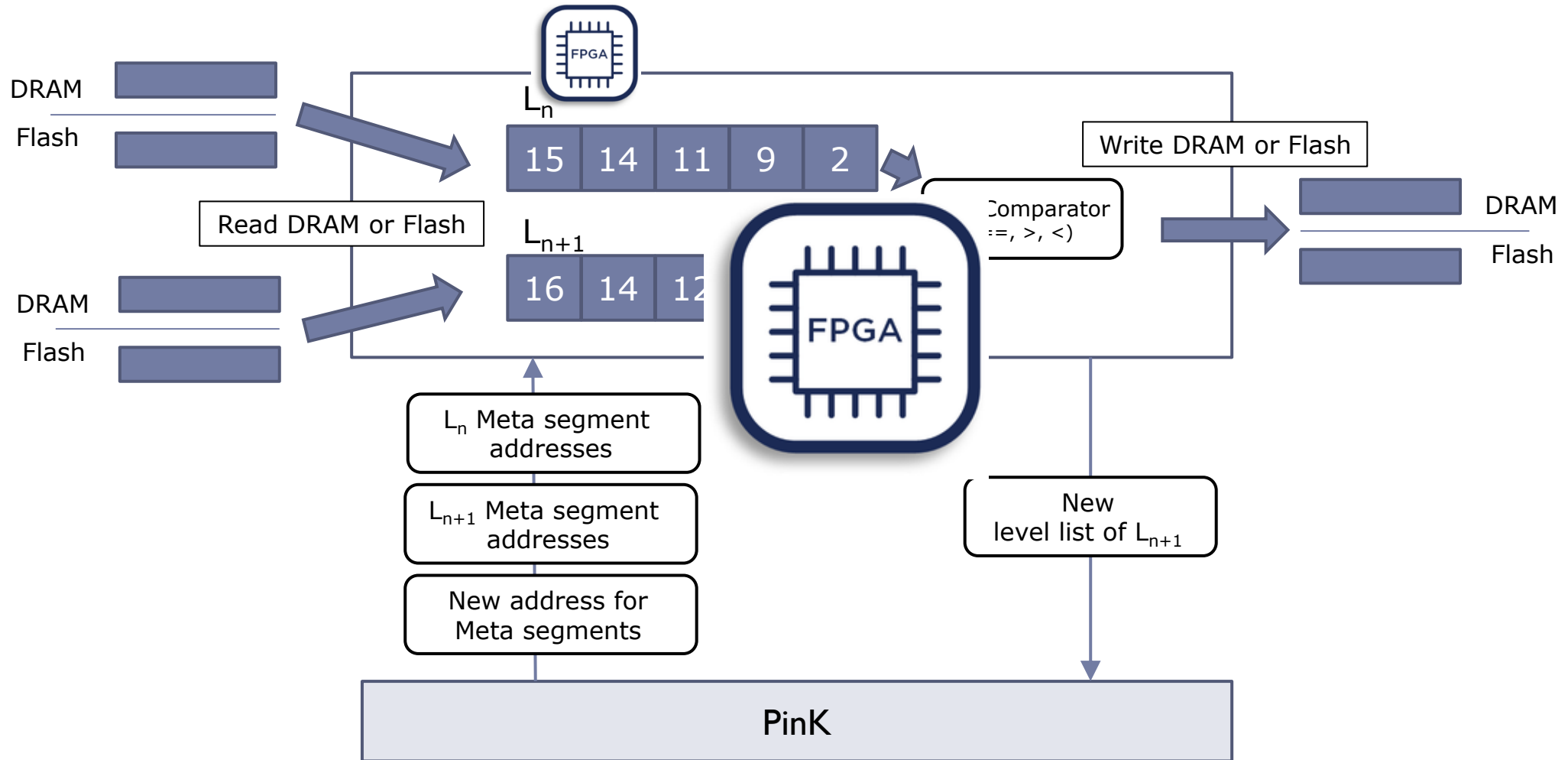
PinK without level-pinning



PinK with level-pinning



Reducing Sorting Time

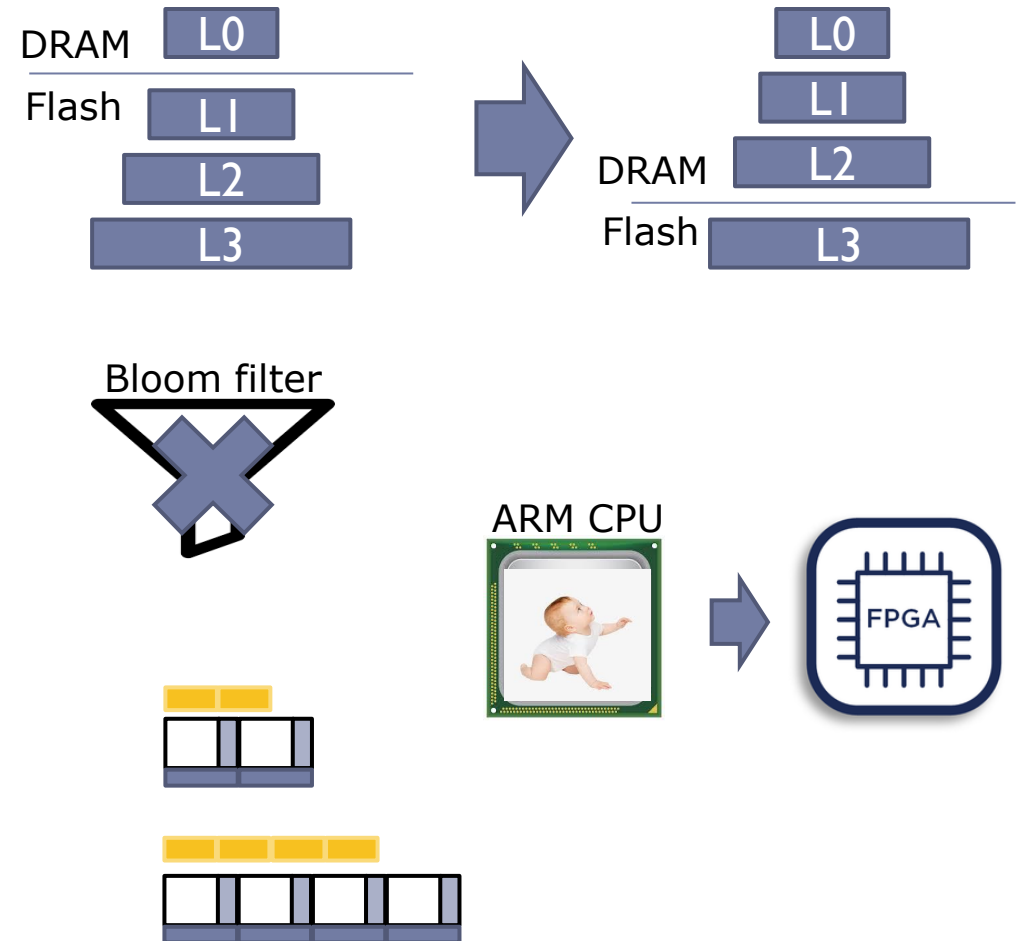


PinK Summary

- ▶ Long tail latency?
Using level-pinning
- ▶ CPU overhead?
Removing Bloom filter
Optimizing binary search
Adopting HW accelerator
- ▶ I/O overhead?
Reducing compaction I/O

Optimizing GC by
reinserting valid data to LSM-tree

Please refer to the paper!



Introduction

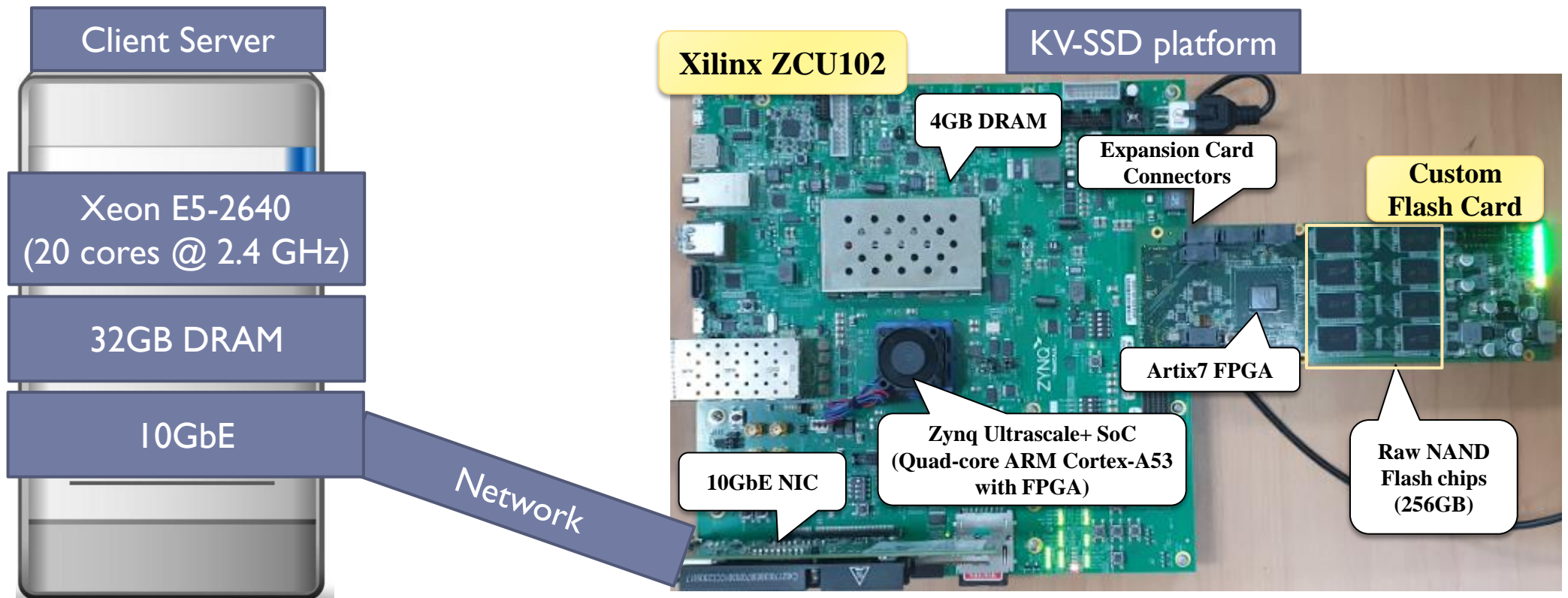
PinK

Experiments

Conclusion

Custom KV-SSD Prototype and Setup

- ▶ All algorithms for KV-SSD were implemented on ZCU102 board
 - ▶ For fast experiments: 64GB SSD, 64 MB DRAM (0.1% of NAND capacity)



Benchmark Setup

- ▶ YCSB: 32B key, 1KB value

	Load	A	B	C	D	E	F
R:W ratio	0:100	50:50	95:5	100:0	95:5	95:5	50:50(RMW)
Query type	Point					Range read	Point
Request distribution	Uniform	Zipfian			Latest (Highest locality)	Zipfian	

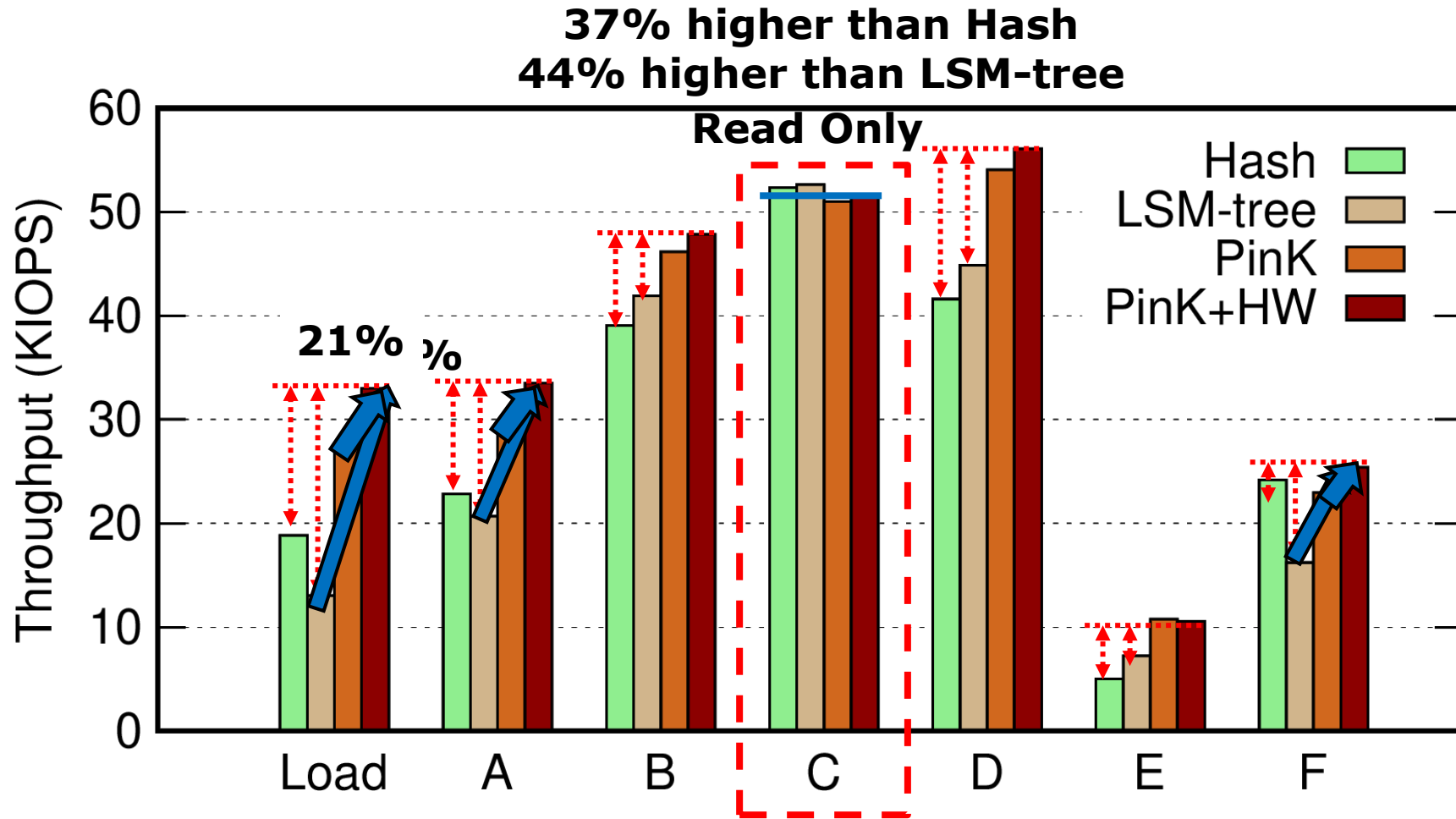
- ▶ Two phases
 - ▶ Load: issue unique 44M KV pairs (44GB, 70% of total SSD)
 - ▶ Run: issue 44M KV pairs following workload description

Testing Algorithms

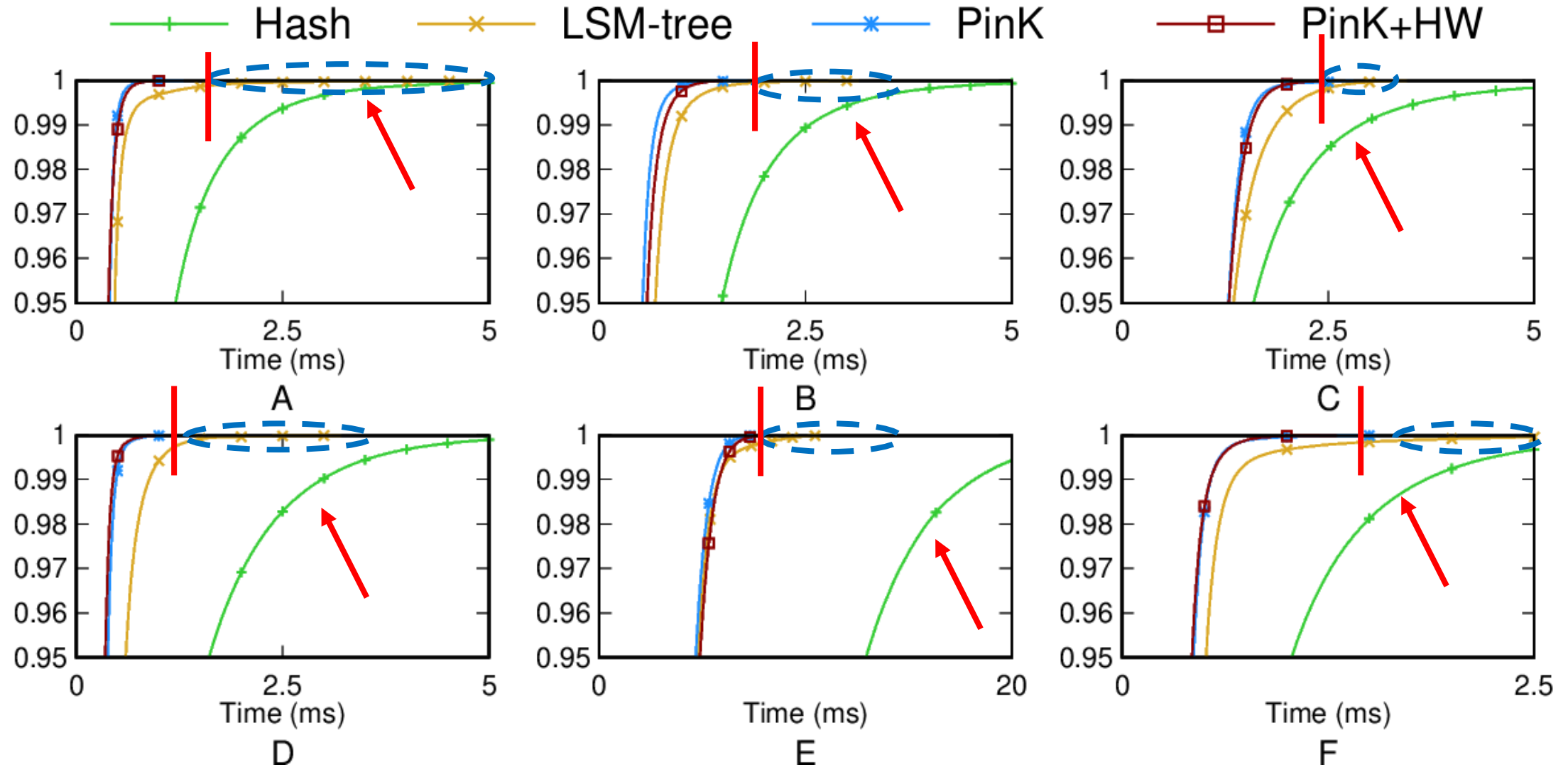
- ▶ Hash
 - ▶ 8-bit signature: total 320MB buckets
- ▶ LSM-tree
 - ▶ The conventional LSM-tree implementation based on Lightstore*
 - ▶ Total 5 levels (1~4 level in flash)
- ▶ PinK
 - ▶ Total 5 levels (pinning top 3 levels, one level in flash)
- ▶ PinK+HW
 - ▶ Using HW accelerator for compaction based on PinK

	Hash	LSM-tree	PinK, PinK+HW
64MB DRAM	LRU bucket caching (64MB)	Level list (9MB) Bloom filter (55MB)	Level list + prefix, range (10MB) Level-pinning (54MB)

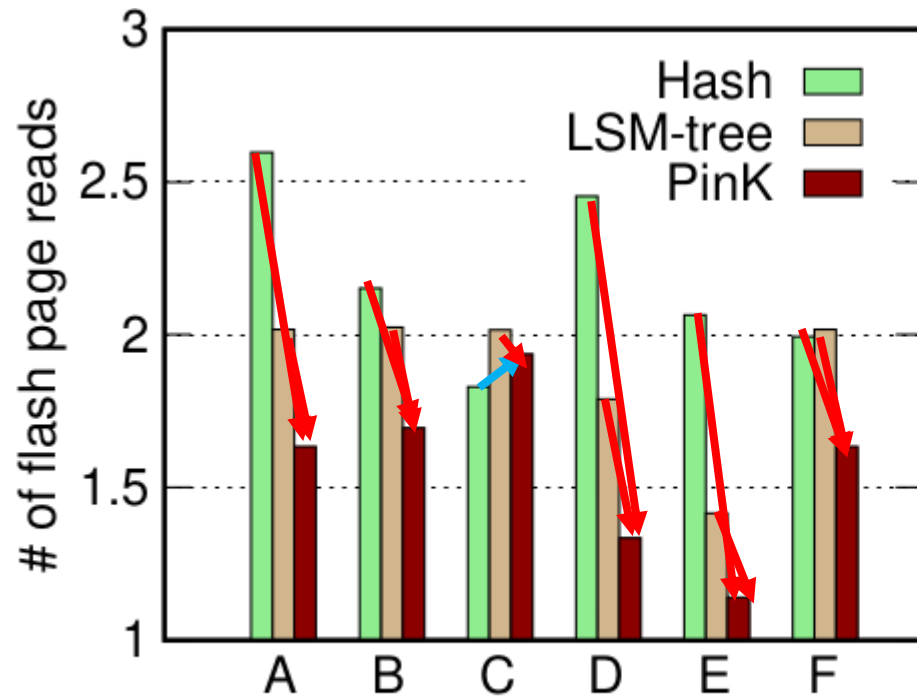
Experiment: Throughput



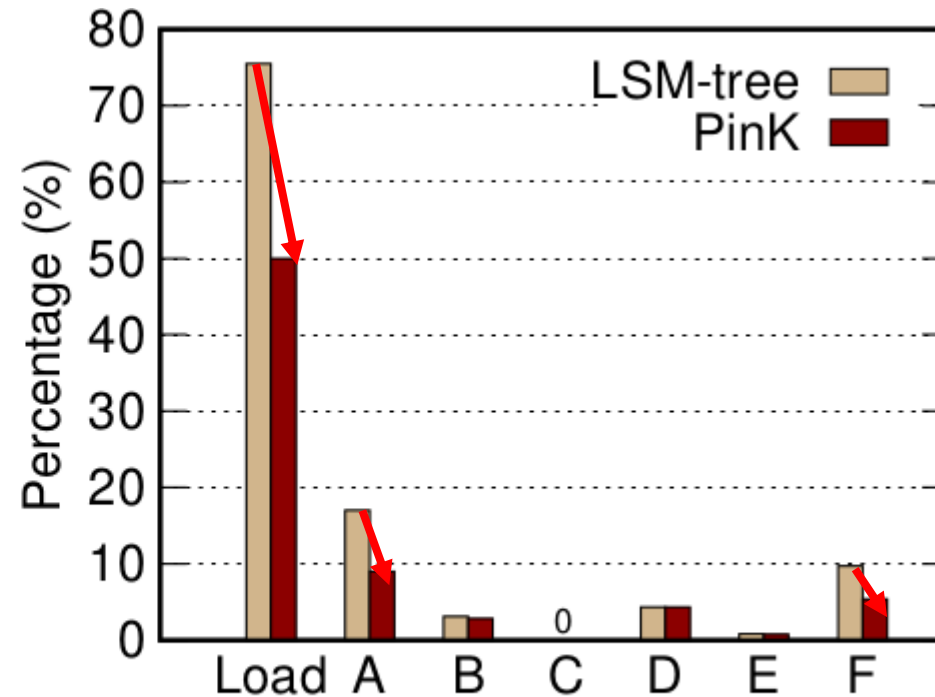
Experiment: Latency



Experiment: Impact of Level-pinning



(a) Flash page reads per query

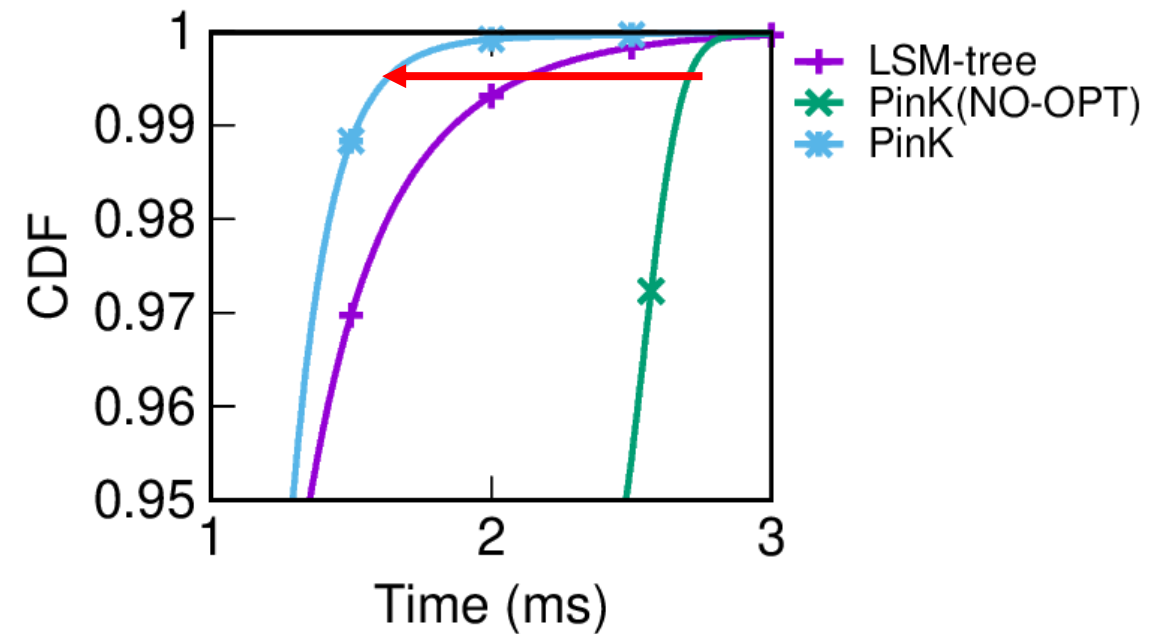
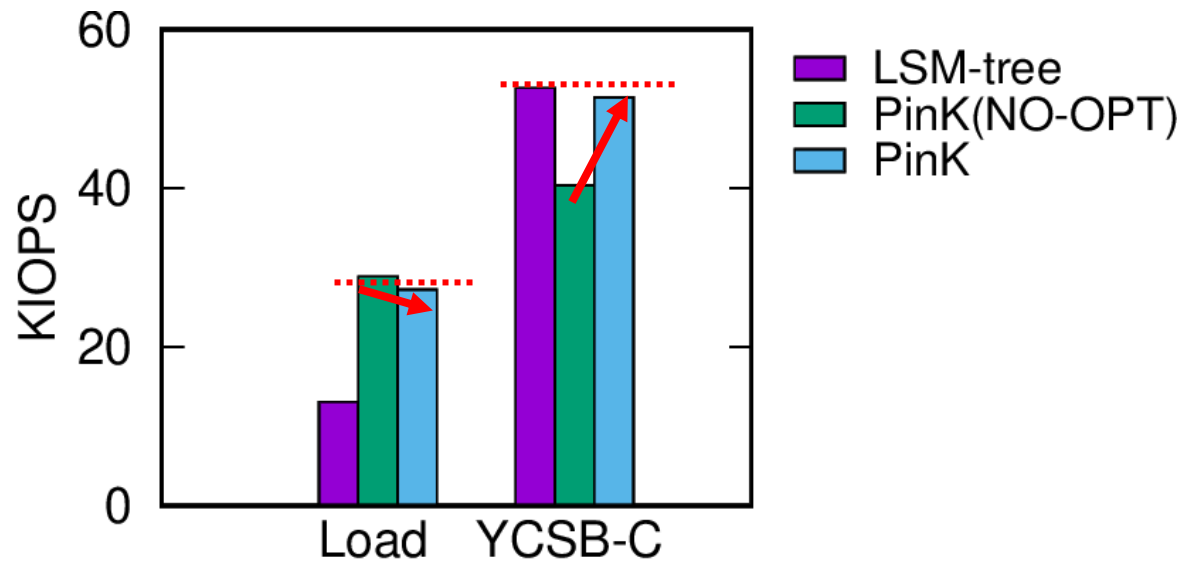


(b) Compaction I/Os

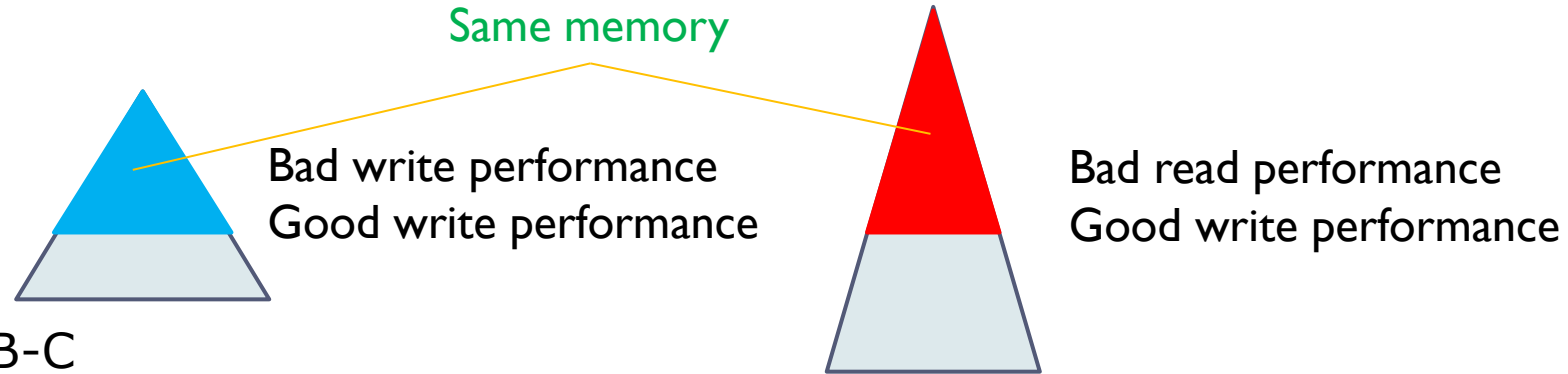
Experiment: Search Optimization

▶ Settings

- ▶ PinK (NO-OPT): PinK without prefix and range pointer
- ▶ Benchmark: YCSB-Load and YCSB-C

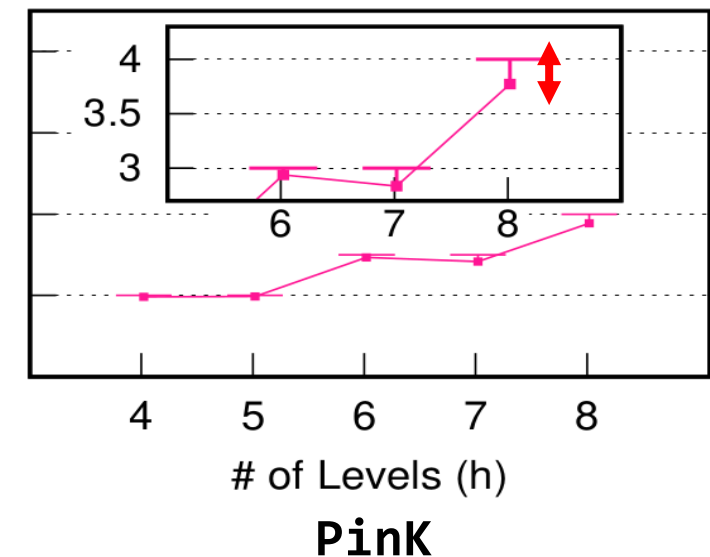
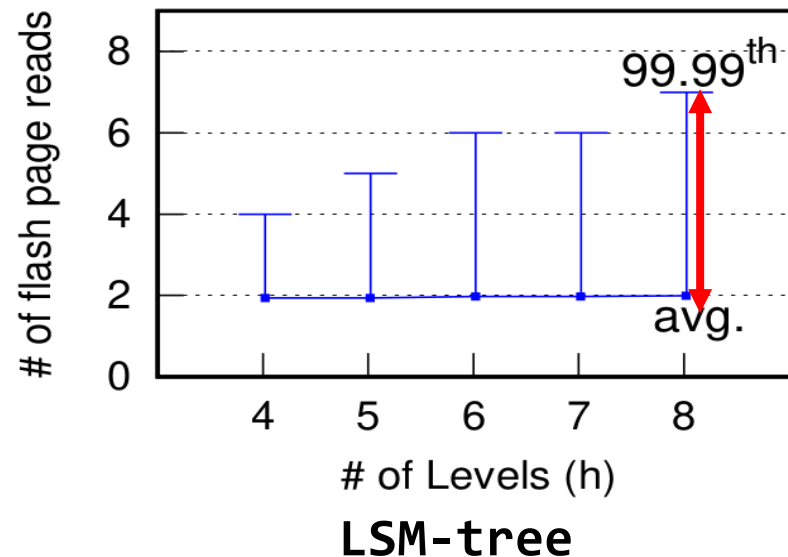


Experiment: Level-pinning on Higher LSM-tree



- ▶ Benchmark: YCSB-C
 - ▶ # of total levels: 4 ~ 8

- ▶ PinK
 - # of levels: 4, 5
 - ▶ Unpinned-level: 1
 - # of levels: 6, 7
 - ▶ Unpinned-level: 2
 - # of levels: 8
 - ▶ Unpinned-level: 3



Introduction

PinK

Experiments

Conclusion

Conclusion

- ▶ Since the conventional KV-SSD's algorithms did not consider the embedded system's limitations well, they have suffered from long tail latency and throughput degradation
- ▶ PinK
 - ▶ Pinning KV indices of top levels of LSM-tree to DRAM to reduce latency
 - ▶ Using HW accelerator for compaction sorting
- ▶ Benefits
 - ▶ 99 percentile tail latency: 73%
 - ▶ Average latency: 42%
 - ▶ Throughput : 37%

Thank You !

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