



BY Developers FOR Developers

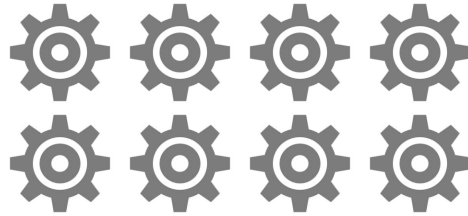
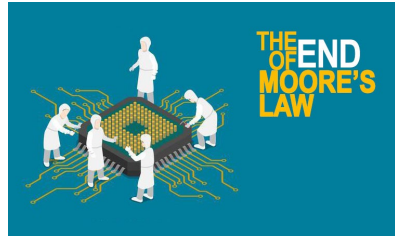
Storage Developer Conference
September 22-23, 2020

The True Value of Storage Drives with Built-in Transparent Compression: Far Beyond Lower Storage Cost

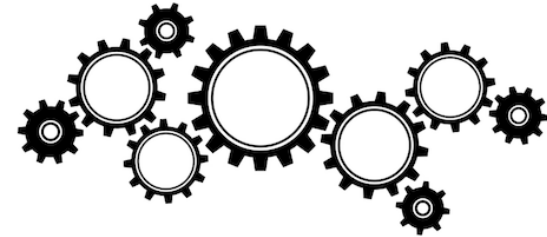
Tong Zhang
ScaleFlux Inc.
San Jose, CA



The Rise of Computational Storage



Homogeneous Computing

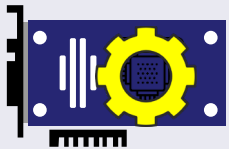


Heterogeneous Computing

Compute



FPGA/GPU/TPU

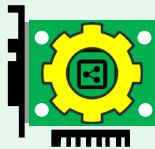


End of Moore's Law

Networking



SmartNICs

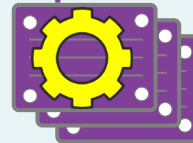


10 → 100-400Gb/s

Storage



Computational



Fast & Big Data Growth

Domain Specific Compute

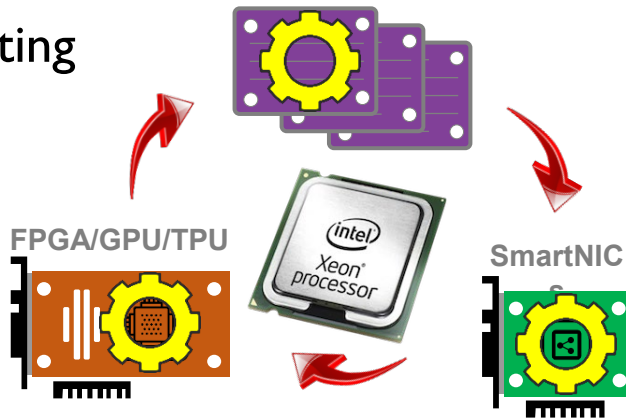
Computational Storage: A Very Simple Idea

❑ End of Moore's Law → heterogeneous computing

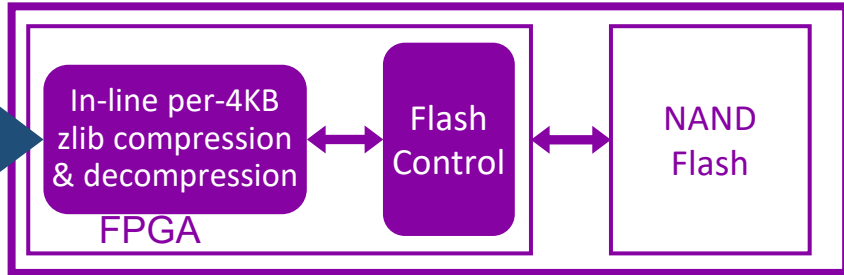
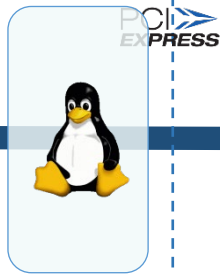
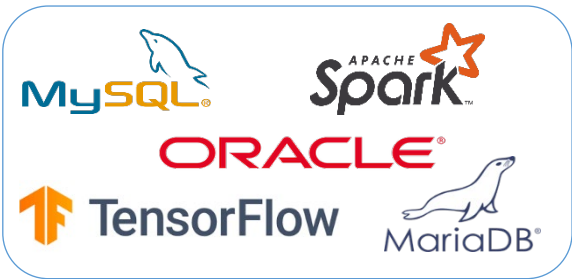


Low-hanging fruits

Computational Storage

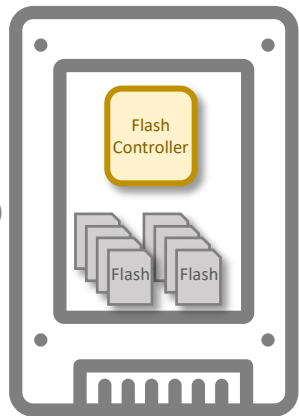
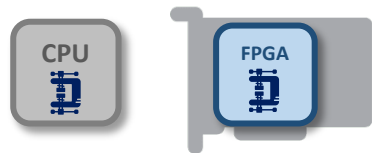


← SW | HW →

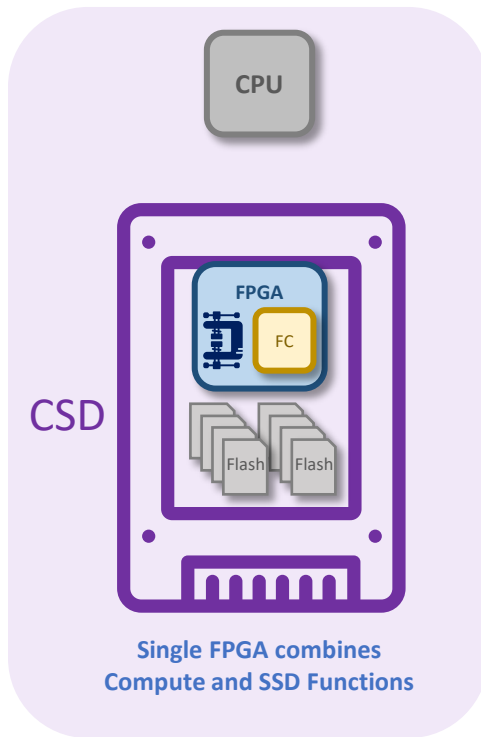


Computational Storage Drive (CSD) with Data Path Transparent Compression

ScaleFlux Computational Storage Drive: CSD 2000



Multiple, discrete components
for Compute and SSD Functions

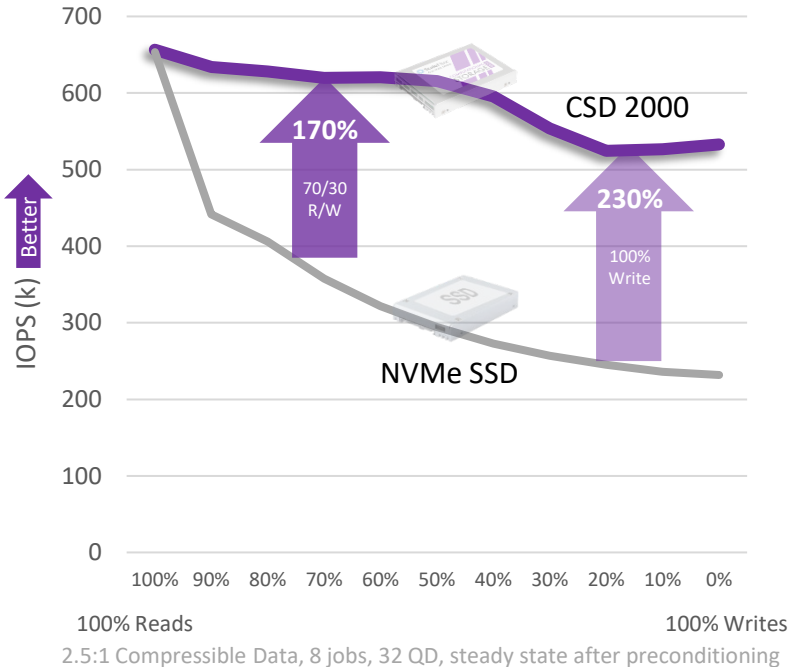


Single FPGA combines
Compute and SSD Functions

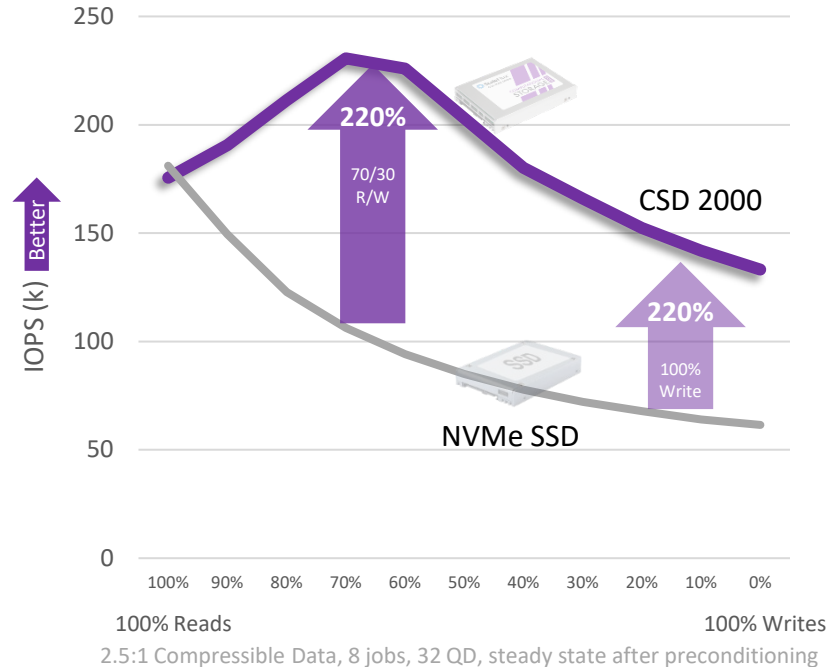
- ✓ Complete, validated solution
 - ✓ Pre-Programmed FPGA
 - ✓ Hardware
 - ✓ Software
 - ✓ Firmware
- ✓ No FPGA knowledge or coding
- ✓ Field upgradeable
- ✓ Standard U.2 & AIC form factors

CSD 2000: Data Path Transparent Compression

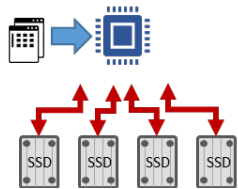
FIO: 4K Random R/W IOPS



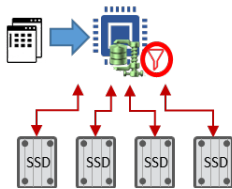
FIO: 16K Random R/W IOPS



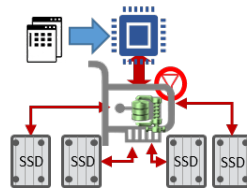
Comparing Compression Options



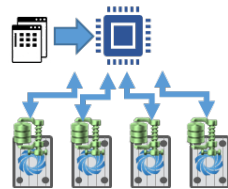
No Compression



Host-Based



Offload Card



CSD 2000

No CPU Overhead



Reduced \$/User GB



Performance scales with capacity



Transparent App Integration



Zero App Latency



No incremental power usage

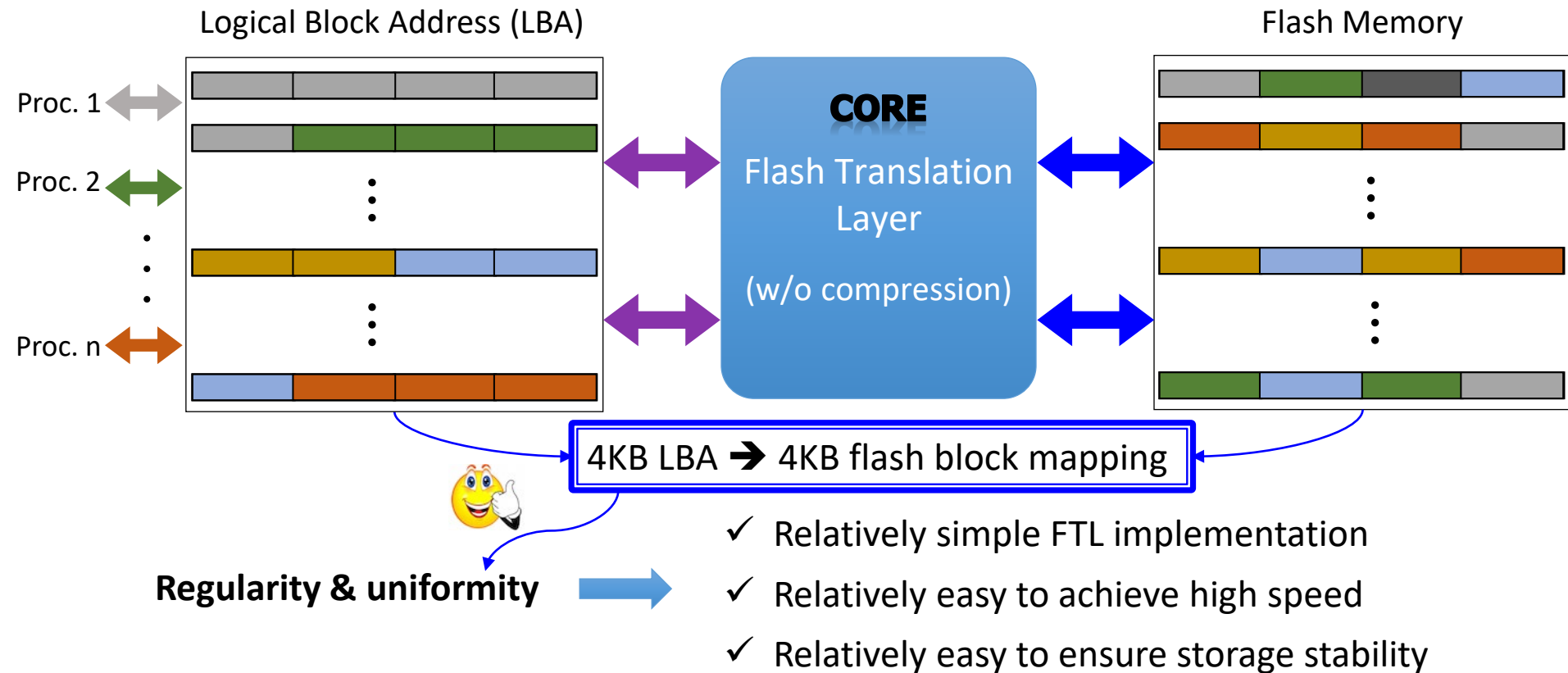


No incremental physical footprint

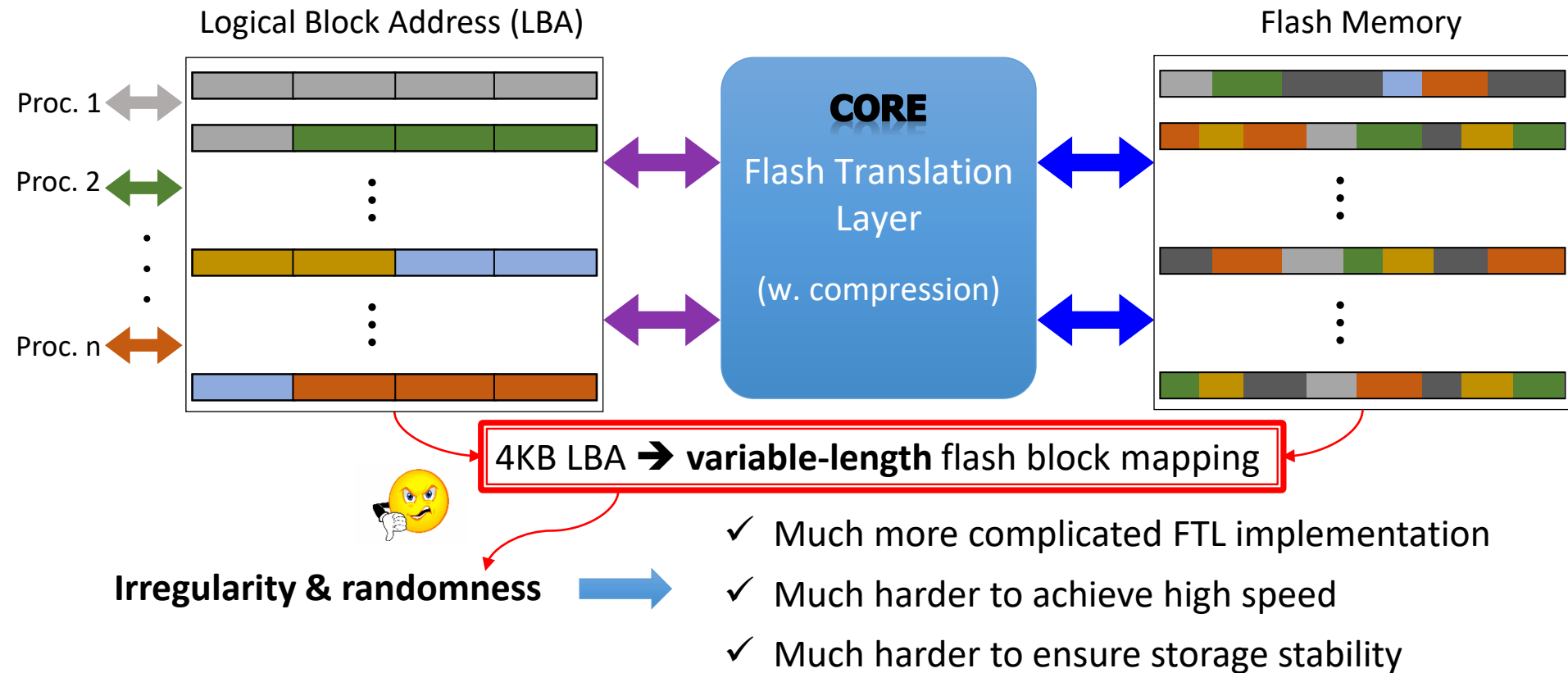


Scalable CSD-based compression reduces Cost/GB without choking the CPU

In-Storage Transparent Compression: Why is It Hard to Build?



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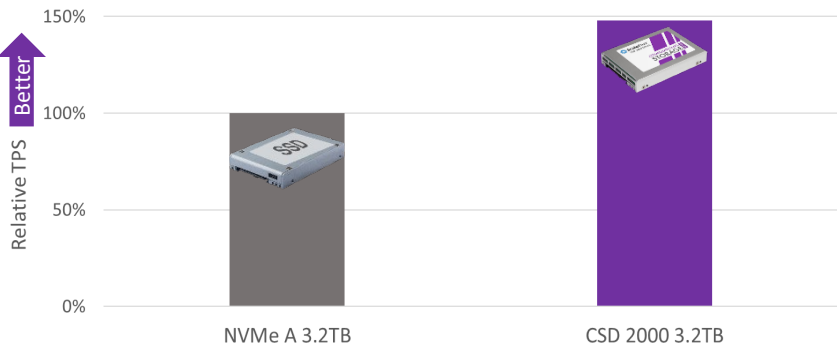
CSD 2000: Highest OLTP TPS, Lowest \$/User GB



- Sysbench (MySQL 5.7.25, InnoDB)
- 50M records, 64 Threads
- 1hr Test run
- Intel(R) Xeon(R) CPU E5-2667 v4 @ 3.20GHz, 256GB DRAM

Performance: 150% TPS

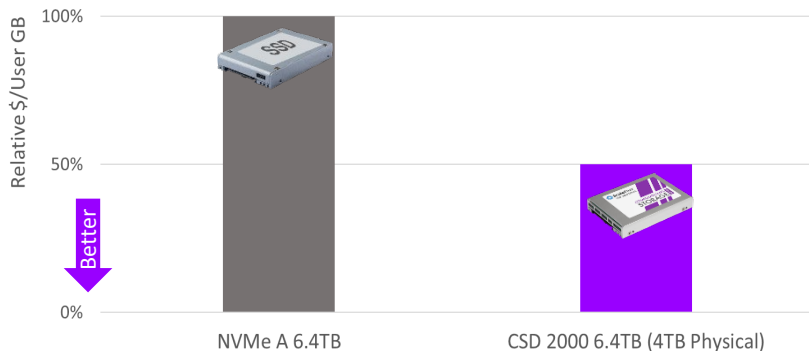
Sysbench (MySQL): Read-Write Transactions per Second (TPS)
2.4TB Dataset



2.4TB Dataset Physical Flash consumed on NVMe A; **0.9TB on CSD 2000**

Cost: 50% Less \$/User GB

Flash Storage \$/User GB Comparison
4.8TB Dataset



4.8TB Dataset Physical Flash consumed on NVMe A; **1.6TB on CSD 2000**
CSD 2000 delivers 30% higher Read-Write TPS in this cost comparison

Flexible Drive Capacity Enables the Best Performance ↔ Cost

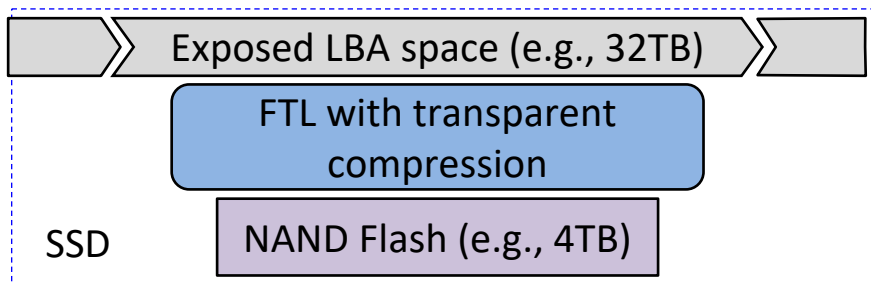
Open a Door for System Innovation

Logical storage space utilization efficiency



Transparent compression

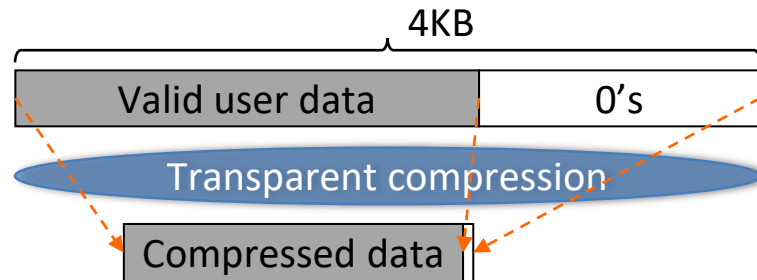
Physical storage space utilization efficiency



Unnecessary to use all the LBAs



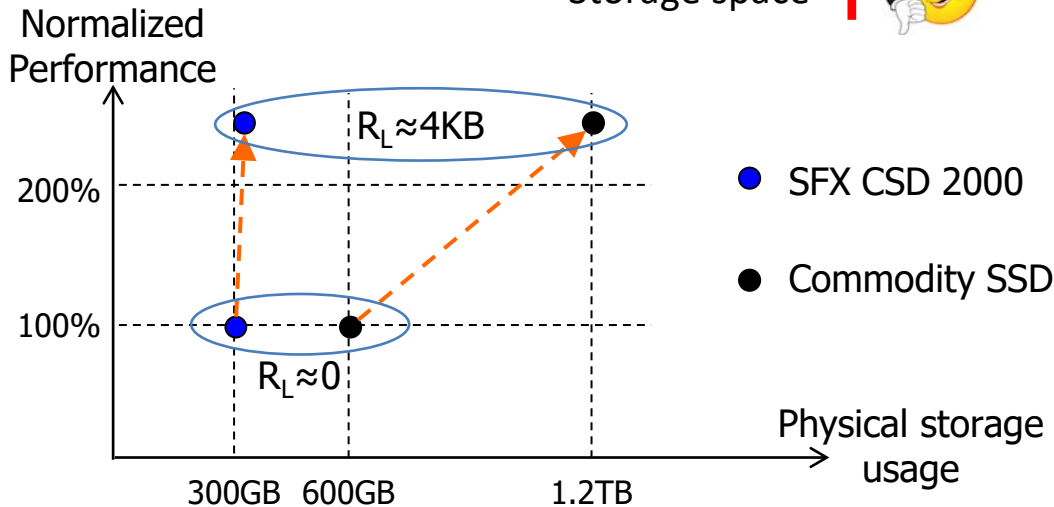
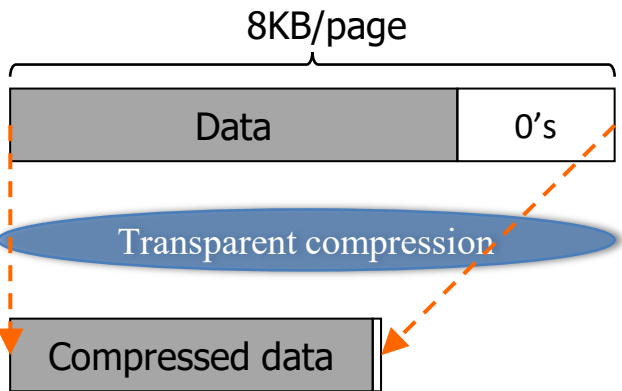
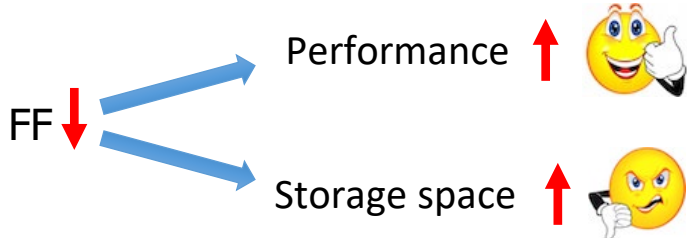
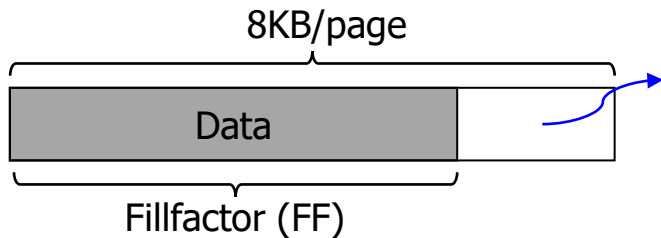
OS/Applications can **purposely waste** logical storage space to gain benefits



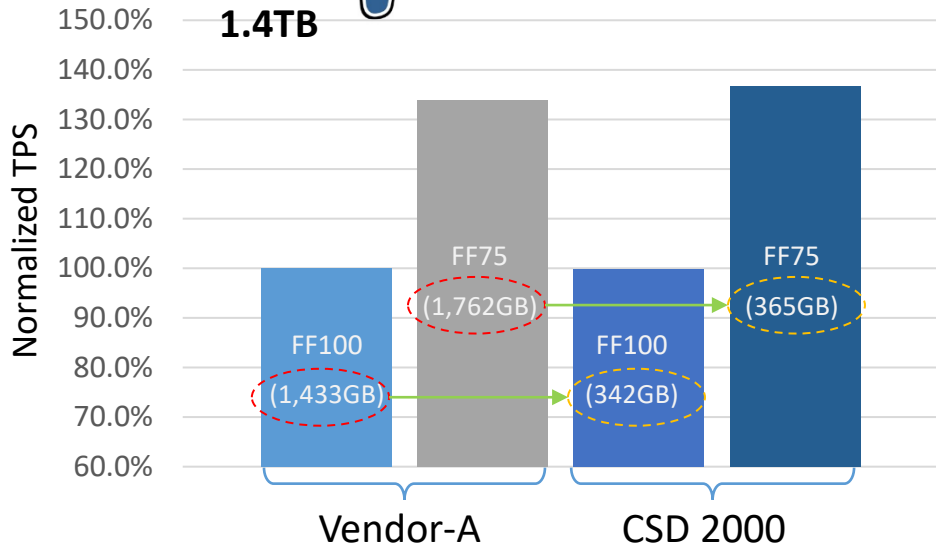
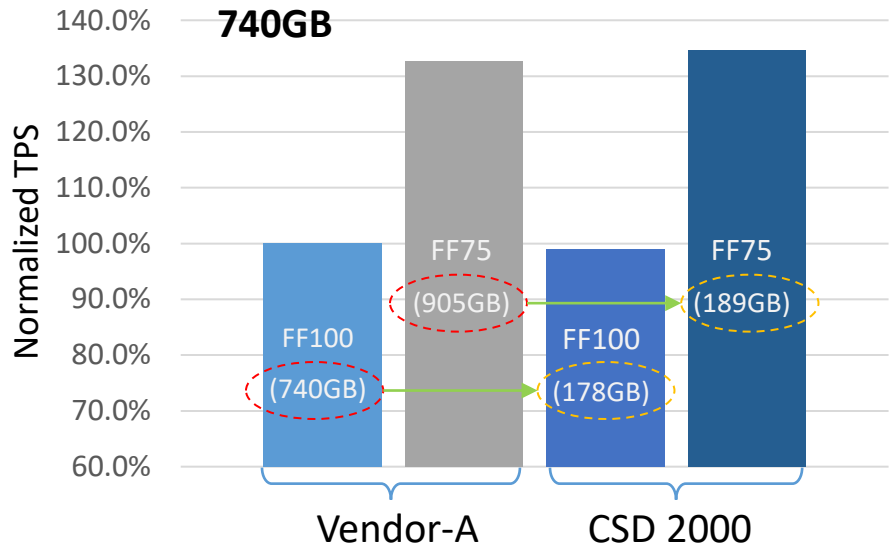
Unnecessary to fill each 4KB sector with user data



Case Study I: PostgreSQL



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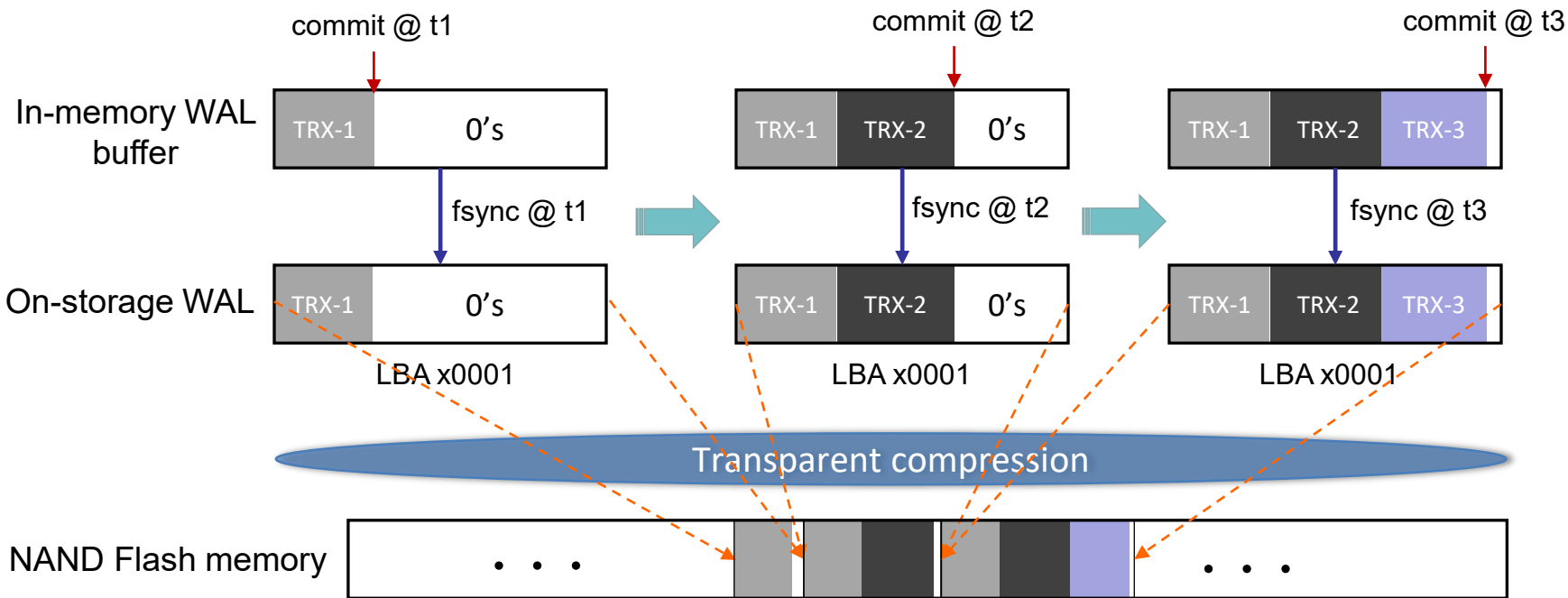
Fillfactor	Drive	Logical size (GB)	Physical size (GB)	Comp Ratio
100	Vendor-A	740	740	1.00
	CSD 2000		178	4.12
75	Vendor-A	905	905	1.00
	CSD 2000		189	4.75

Fillfactor	Drive	Logical size (GB)	Physical size (GB)	Comp Ratio
100	Vendor-A	1,433	1,433	1.00
	CSD 2000		342	4.19
75	Vendor-A	1,762	1,762	1.00
	CSD 2000		365	4.82

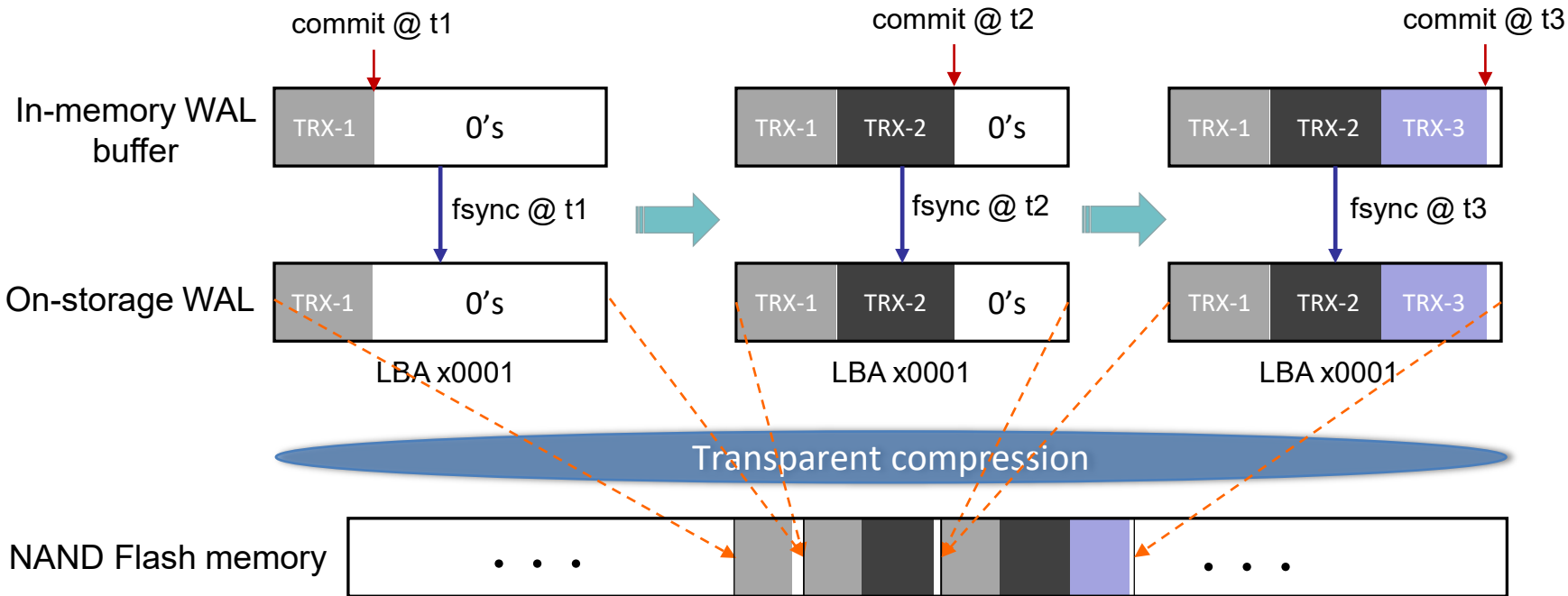
Case Study 2: Sparse Write-Ahead Logging

❑ Write-ahead logging (WAL)

- Universally used by data management systems to achieve atomicity and durability



Case Study 2: Sparse Write-Ahead Logging



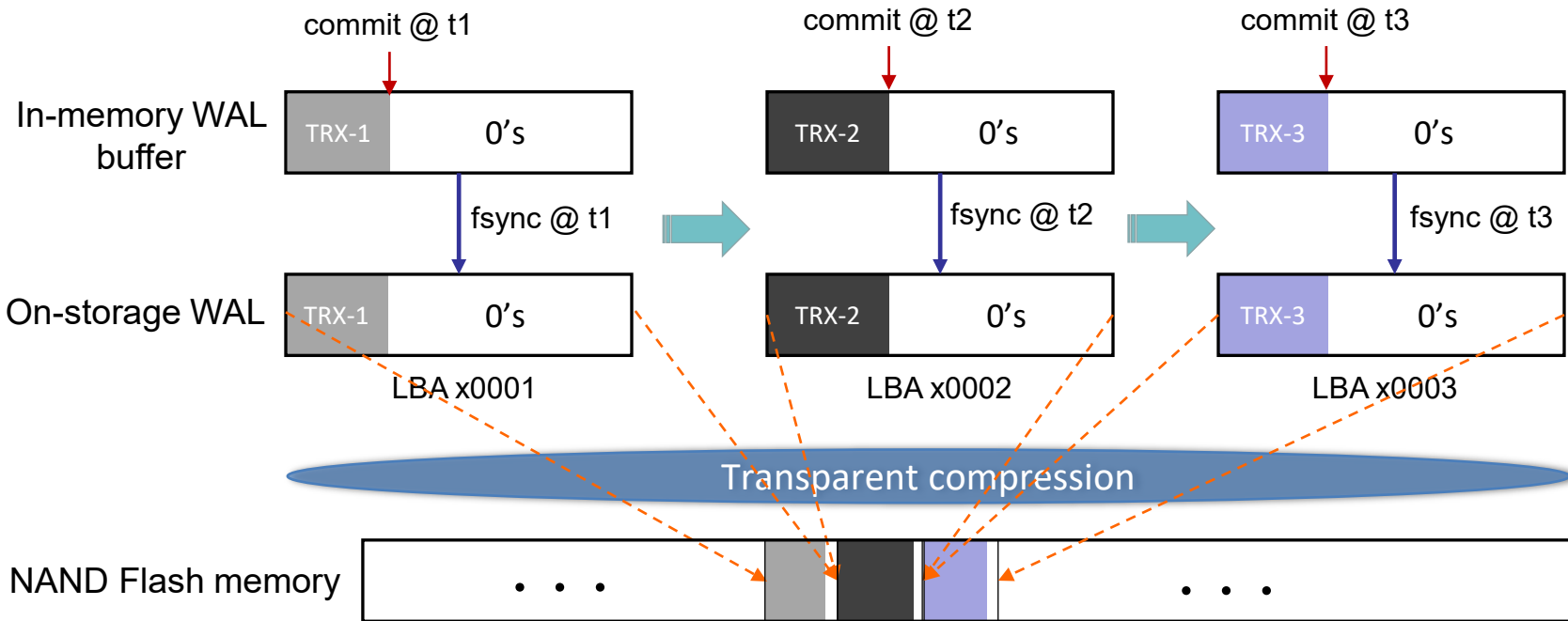
Write amplification

More interference with other IOs ☹️

Shorter NAND flash memory lifetime ☹️

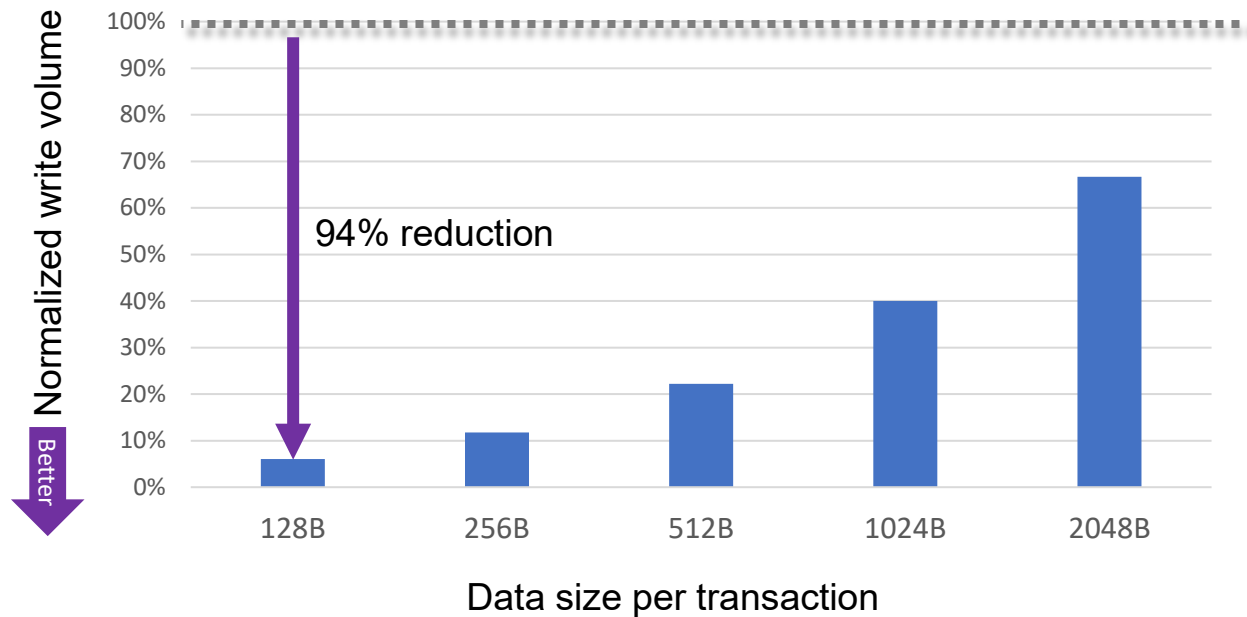
Case Study 2: Sparse Write-Ahead Logging

- ❑ Sparse WAL: Allocate a new 4KB sector per transaction commit
 - ✓ Waste logical storage space → reduce WAL-induced write amplification



Case Study 2: Sparse Write-Ahead Logging

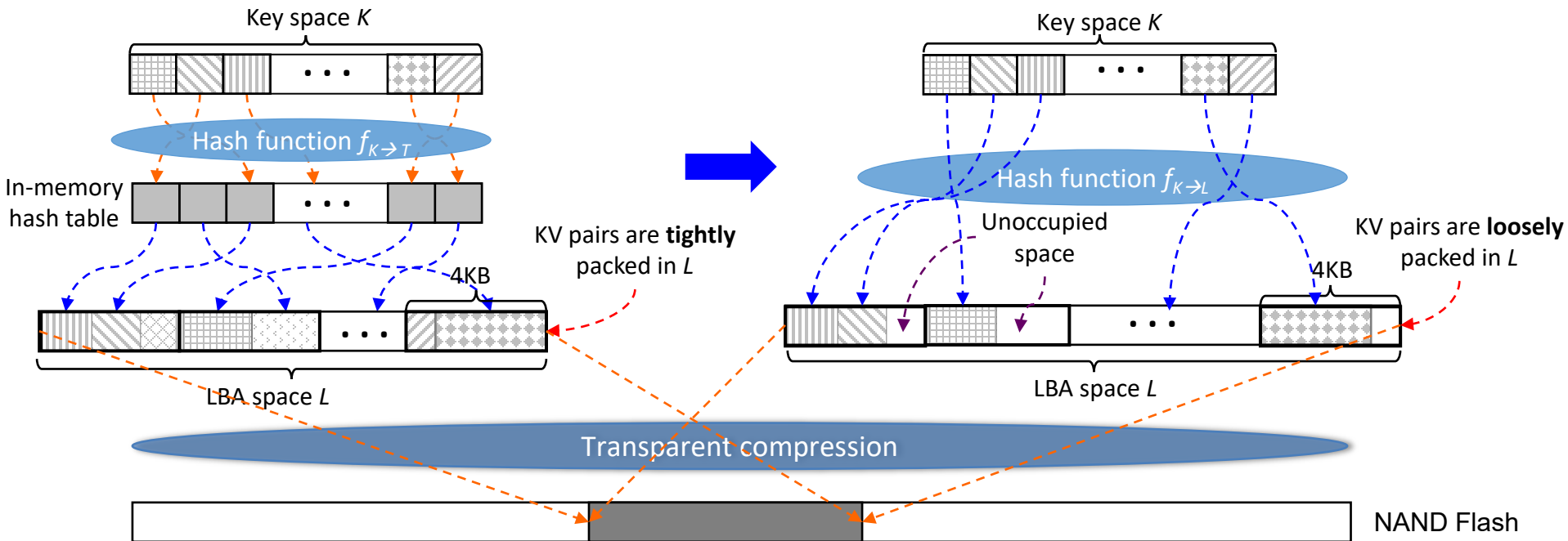
- ❑ Sparse WAL: Allocate a new 4KB sector per transaction commit
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Case Study 3: Table-less Hash-based KV Store

❑ Very simple idea

- Hash *key space* directly onto *logical storage space* → eliminate the in-memory hash table
- Transparent compression eliminates the “unoccupied space” from physical storage space



Case Study 3: Table-less Hash-based KV Store

❑ Eliminate in-memory hash table

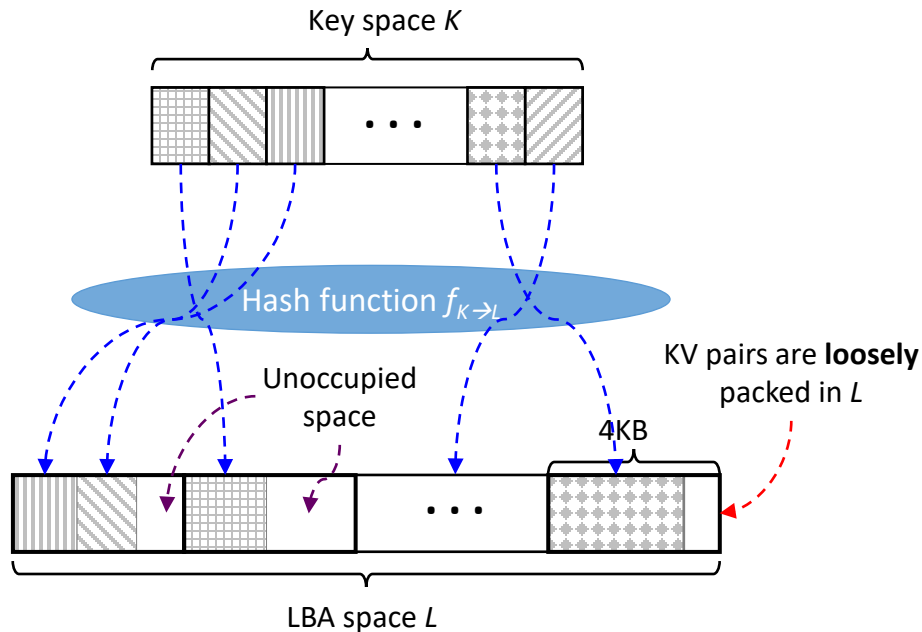
- ✓ Very small memory footprint
- ✓ High operational parallelism
- ✓ Short data access data path
- ✓ Very simple code base

❑ Under-utilize logical storage space

- ✓ Obviate frequent background operations (e.g., GC and compaction)



High performance, low memory cost, and low CPU usage



Case Study 3: Table-less Hash-based KV Store

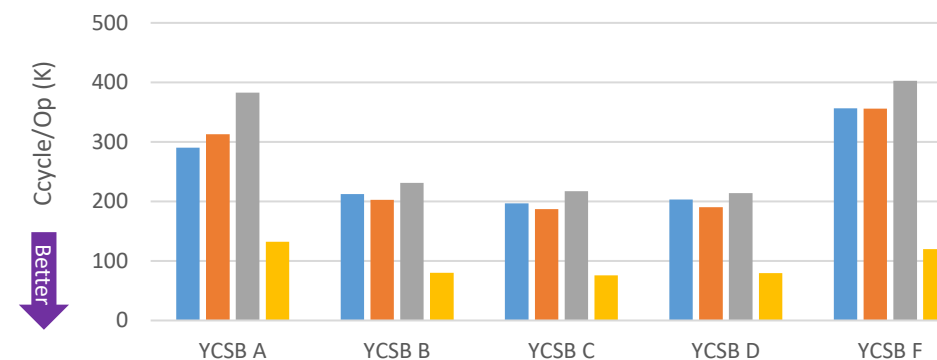
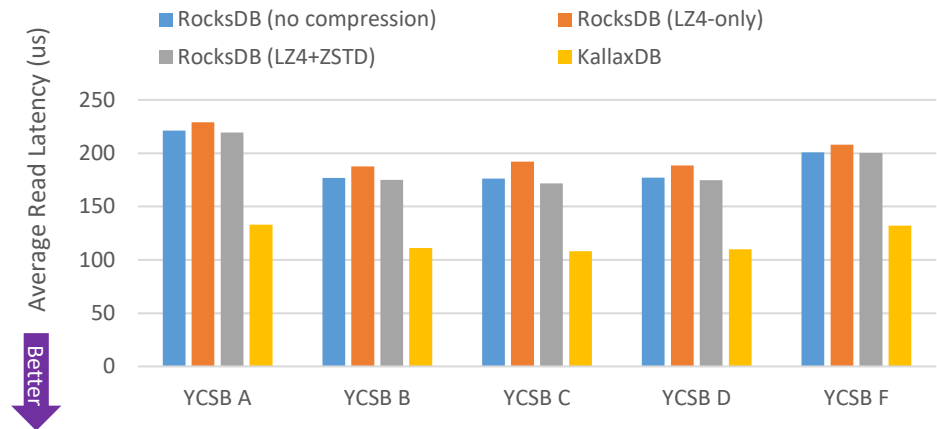
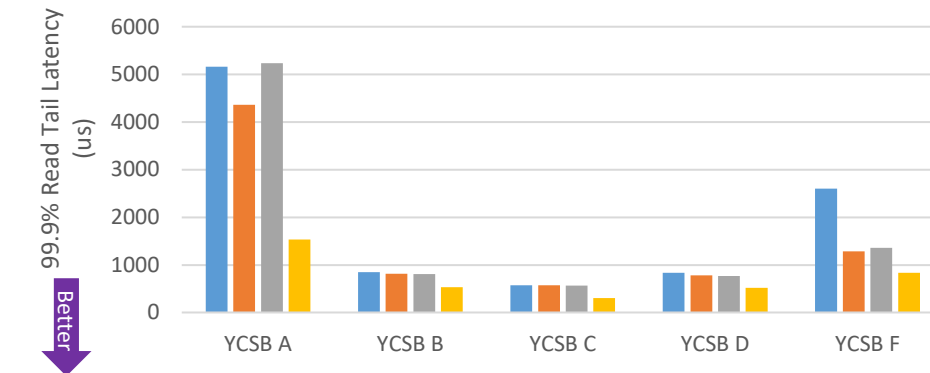
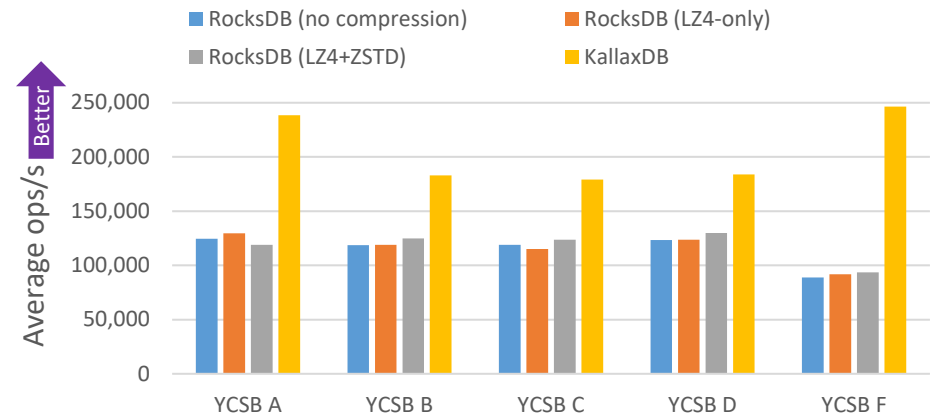
❑ Experimental Setup

- 24-core 2.6GHz Intel CPU, 32GB DDR4 DRAM, and a 3.2TB SFX CSD2000
- RocksDB 6.10 (12 compaction threads and 4 flush threads)
- 400-byte KV pair size, 1 billion KVs ➔ 400GB raw data
- Memory usage: RocksDB (5GB), KallaxDB (600MB)

YCSB A	50% reads, 50% updates
YCSB B	95% reads, 5% updates
YCSB C	100% reads
YCSB D	95% reads, 5% inserts
YCSB F	50% reads, 50% read-modify-writes

	Storage Usage
RocksDB (no compression)	428GB
RocksDB (LZ4-only)	235GB
RocksDB (LZ4+ZSTD)	201GB
KallaxDB	216GB

Case Study 3: Experimental Results (24 clients)



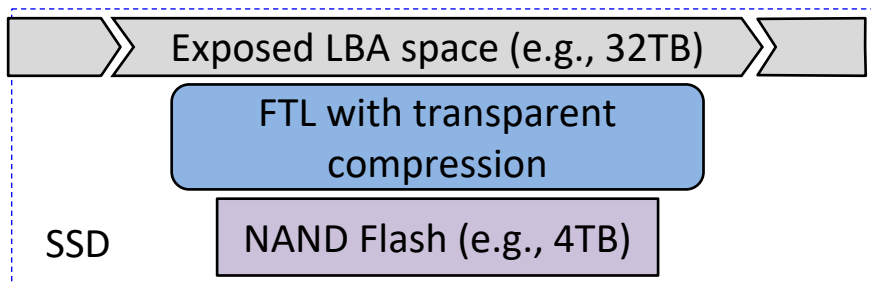
Open a Door for System Innovation

Logical storage space utilization efficiency



Transparent compression

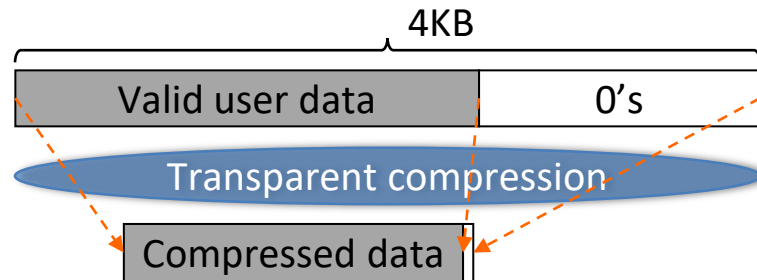
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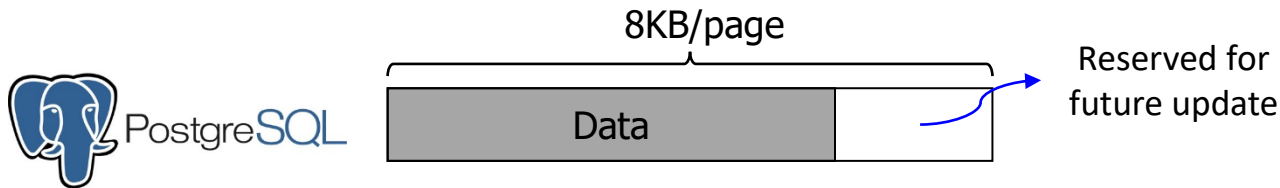
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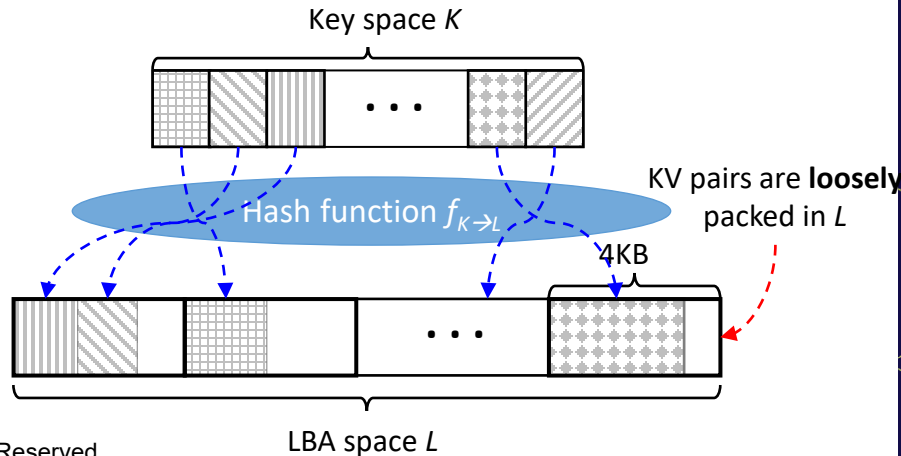
Reserve more space for future update to improve performance @ zero storage overhead

Sparse WAL ➡ Reduce WAL-induced write amplification @ zero storage overhead

Table-less hash-based KV store



High performance, low memory/CPU usage
@ zero storage overhead





Thank You

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