10. Capacity Optimization

Special Topics in Computer Systems:

Modern Storage Systems (IC820-01)

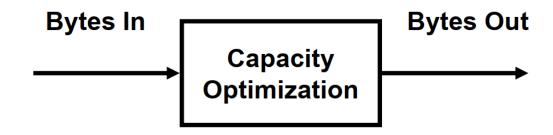
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Capacity Optimization

- There are two popular methods to improve storage capacity
 - Method 1: Data Deduplication
 - The replacement of multiple copies of data with references to a shared copy in order to save storage space
 - Method 2: Data Compression
 - The encoding of data to reduce its storage requirements
- Widely used in modern storage systems and devices
 - Not only improve storage capacity, but I/O performance and energy efficiency

Space Reduction Ratio & Percent



Ratio =
$$\left(\frac{1}{1-\%}\right)$$

$$\% = 1 - \left(\frac{1}{\text{Ratio}}\right)$$

Space Reduction Ratio & Percent (Cont.)

Space Reduction Ratio	Space Reduction Percent		
2:1	1/2 = 50%		
5:1	4/5 = 80%		
10:1	9/10 = 90%		
20:1	19/20 = 95%		
100:1	99/100 = 99%		
500: I	499/500 = 99.8%		

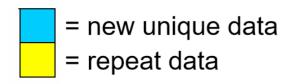
- Ratios can meaningfully be compared only under the same set of assumptions
- Relatively low space reduction ratios provide significant space savings

Outline

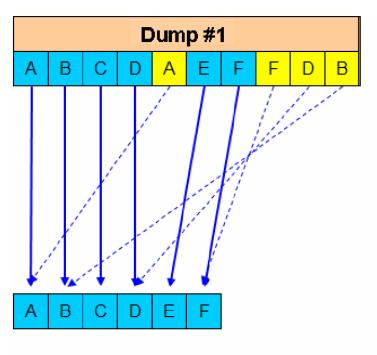
- Data Deduplication
- **■** Data Compression
- **Case Studies**

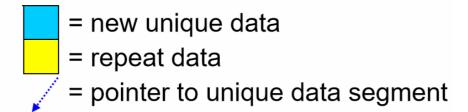
Data Deduplication Simplified



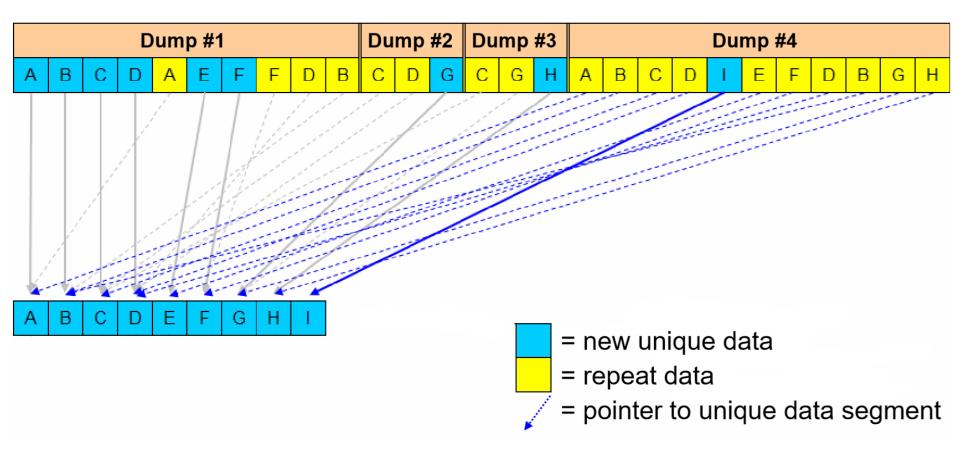


Data Deduplication Simplified (Cont.)

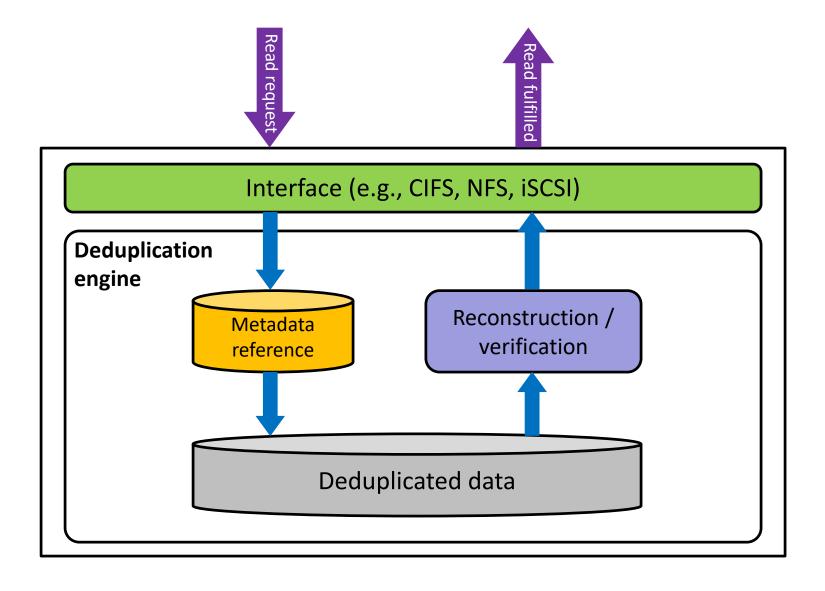




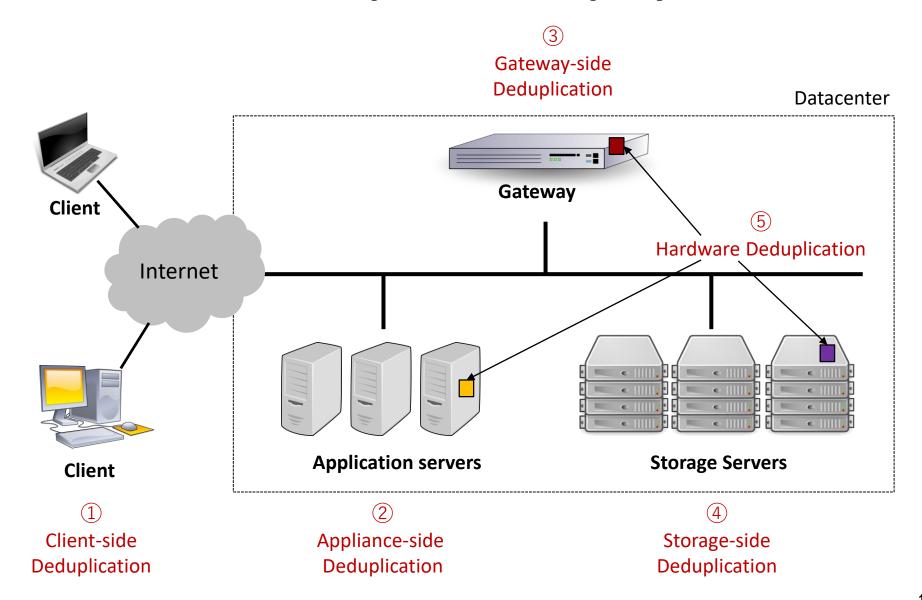
Data Deduplication Simplified (Cont.)



Reading Duplicated Data



Where is Dededupliation Deployed?



Source and Target

■ Source Deduplication

- Identifies duplicate data at the client
- Transfers unique segments to a central repository
- Separate client and server components

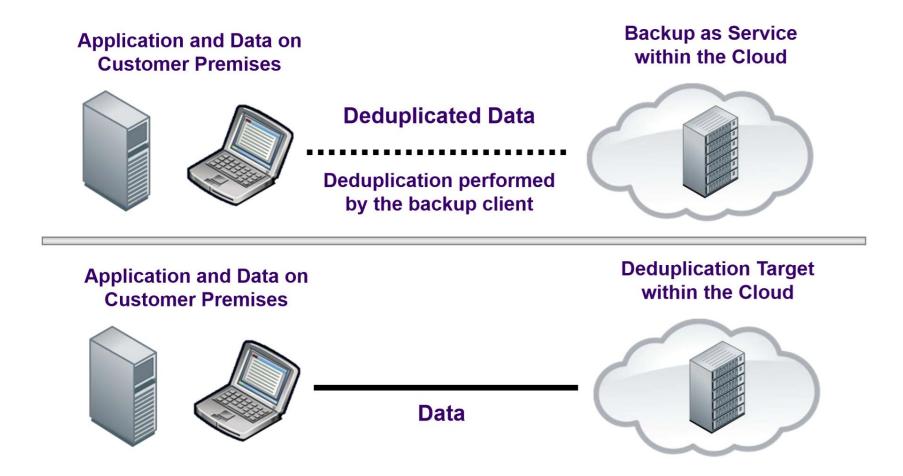
■ Target Deduplication

- Identifies duplicate data where the data is being stored
- Stores unique segments
- Standalone systems

Consideration

- Neither approach enables a greater or lesser space savings
- Scope of data deduplication may vary by implementation

Source and Target (Cont.)



Inline or Post-Process

■ Inline Deduplication

- Data deduplication performed before writing the duplicate data
- May improve I/O performance
- Low deduplication ratio

■ Post-process Deduplication

- Data deduplication performed after the data to be deduplicated has been initially stored
- May degrade I/O performance
- High deduplication ratio

Fixed or Variable Size Segment

■ Subfile Data Deduplication

- Fixed-length Segment Deduplication
 - Evaluation of data includes a fixed reference window used to look at segments of data during deduplication process
 - Provide fixed granularity (e.g., 4 KB, 8 KB, or 128 KB)
- Variable-length Segment Deduplication
 - Evaluation of data uses a variable length window to find duplicate data in stream or volume of data processed
 - Provide variable granularity

■ Single Instance Deduplication

Operate at a granularity of an entire file or data object

Outline

- Data Deduplication
- **■** Data Compression
- **Case Studies**

Data Compression

Lossless versus lossy compression

- Lossless compression means that no information is lost when a file is compressed and then uncompressed
- Lossy compression usually results in better compression ratio, but some information is lost

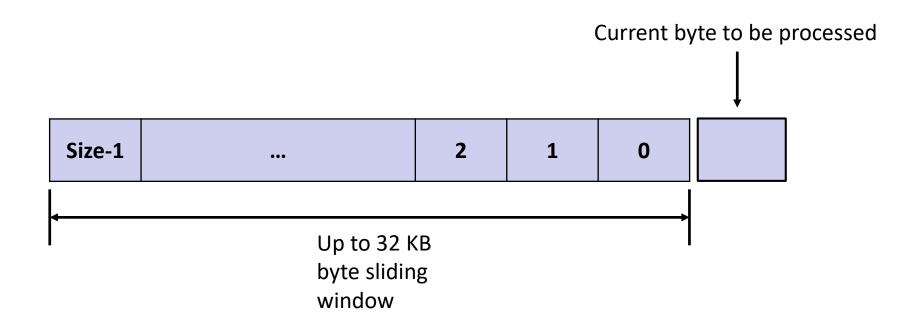
For data storage, lossless compression is mainly used

- LZ1-Based algorithm combined with Huffman encoding
 - LZ1 algorithm to identify matches in a sliding window history buffer
 - A post encoder to Huffman encode the matches and literals
- Hardware and software implementations

LZ1 Architecture

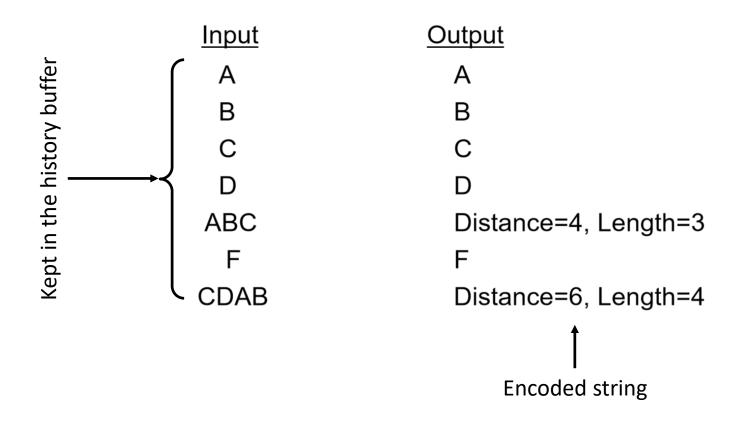
- A String-Matcher searches a History-Buffer to find repeating strings of bytes
- A sliding-window *History-Buffer* adds one new byte and drops off one byte from the back end of the buffer
- A Post-Encoder is a prefix encoder
 - Use statistics to encode the most common string matches with a smaller number of bits

LZ1 Algorithm and History Buffer



LZ1 String Matching

Input String: ABCDABCFCDAB.....



Huffman Encoding

Input String: A B B C B A D B

Probability of Occurrence

Input character	<u>Probability</u>		
Α	0.25		
В	0.5		
С	0.125		
D	0.125		

Huffman Encoding (Cont.)

	\land				
0	1				
/	\				
В	/	\			
	0	1			
	/		١		
Α		/	,	\	
		0		1	
		/			\
	(2			D

<u>Symbol</u>	<u>Code</u>	<u>Pr</u>
Α	10	0.25
В	0	0.5
С	110	0.125
D	111	0.125

Reduction =
 ½[0.25(2) + 0.5(1) + 0.125(3) + .125(3)]
 = 0.875

 Reduction in data size due to Huffman encoding.

Data Dependent

- Random data provides poor compression ratio performance
- Data with repeating byte strings, 2-byte or longer provides grater compression ratio performance
- Compression ratio greater than 100:1 are possible
- May expand if attempting to compress previously compressed data, but a system could detect this and send original data without compression

Algorithm Dependent

- Size of sliding window
- Static or dynamic Huffman encoding
- Number of matches tracked
- Length of matches the algorithm will search for

Hardware Implementation

 Software compression may result in the degradation of I/O throughput and latency as well as energy consumption

Hardware compression

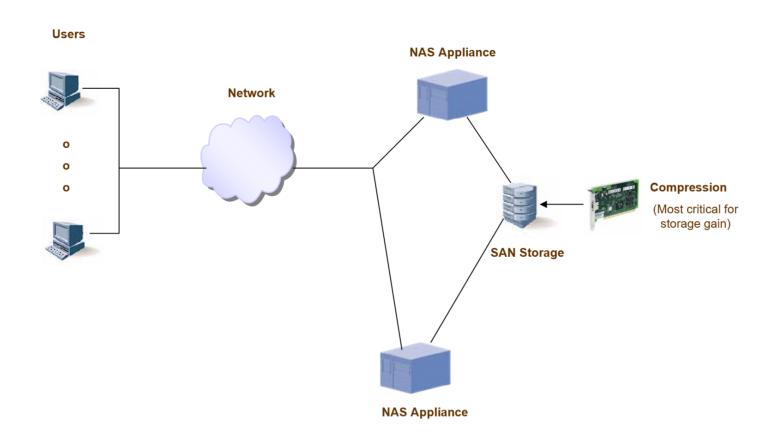
- (+) Higher data rate throughput (10x)
- (+) Free up valuable CPU bandwidth
- (+) reduce power consumption
- (+) Speed up a network link by sending shorter files
- (–) Lower compression ratio

■ Several hardware-specific compression algorithms are available

- Hardware implementation of GZIP algorithms
- X-Match PRO: Real-time compression / decompression

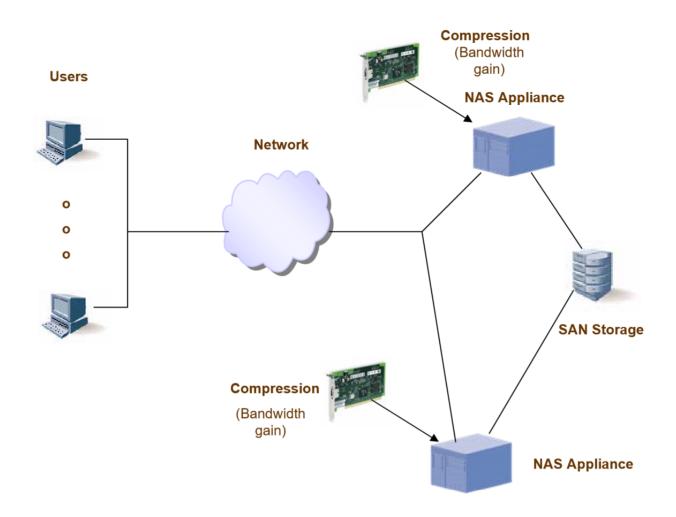
Deployment of Compression

■ Similar to deduplication; combined with deduplication



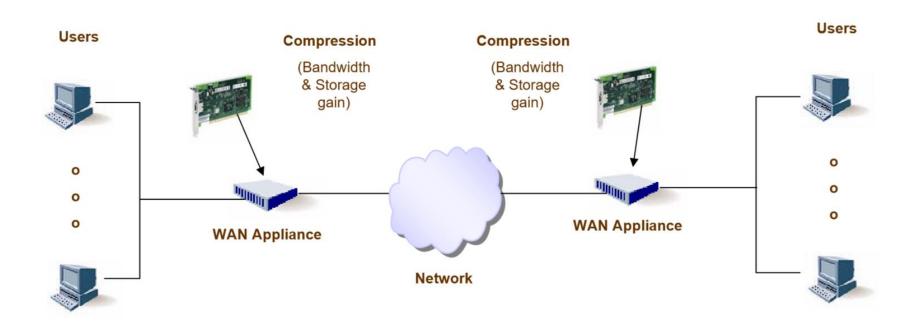
Deployment of Compression (Cont.)

Similar to deduplication; combined with deduplication



Deployment of Compression

■ Similar to deduplication; combined with deduplication



Data Deduplication vs Data Compression

Data deduplication

- (+) Large space saving is possible for exactly matched segments or files
- (–) Even a single-bit difference makes deduplication ineffective
- (–) Unable to get rid of repeated bit patterns of segments or files

Data compression

- (+) Effectively remove repeated bit patterns from input data streams
- (–) Do not provide a high compression ratio all the times

Common design practice

- Use deduplication for input data streams
- Use lossless compression for storage of duplicate data and remaining meta data

Outline

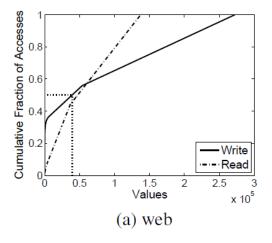
- Data Deduplication
- **■** Data Compression
- Case Studies
 - CA-SSD: Content-aware Solid-state Drives
 - BlueZIP: Hardware-accelerated Compression
 - DAC: Dedup-assisted Compression

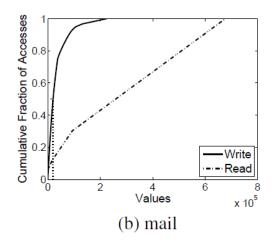
Value Locality

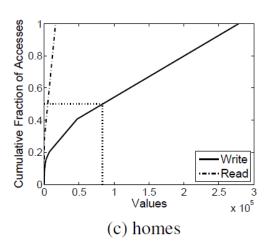
- To improve the reliability and performance of SSD, many FTL schemes exploit temporal/spatial locality
 - Temporal locality: buffering writes to eliminate duplicate writes
 - Spatial locality: coalescing multiple sub page writes into fewer page writes
- However, another form of locality, Value Locality, has not been completely unexplored
 - Value locality: certain content is accessed preferentially

Value Popularity (VP)

- Value popularity: the number of occurrences of each unique value
- Value popularity in real-word workloads





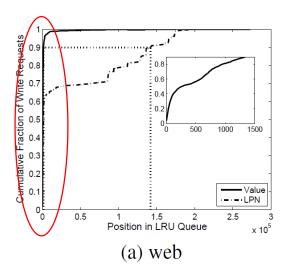


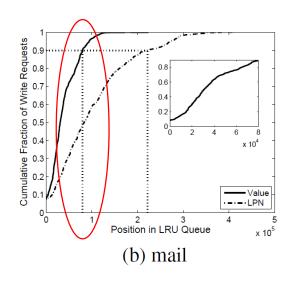
Workload statistics

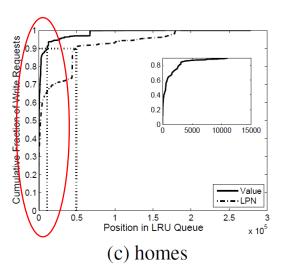
	Size	%	Req.	Unique Request (%)		Seq.
Workload	(GB)	Writes	(mill.)	Write	Read	%
web	1.95	77.01	3.8	42.35	32.05	83.8
mail	4.22	77.32	3.6	7.83	80.85	94.7
homes	3.02	96.76	4.4	66.37	80.75	70.8

Temporal Value Locality (TVL)

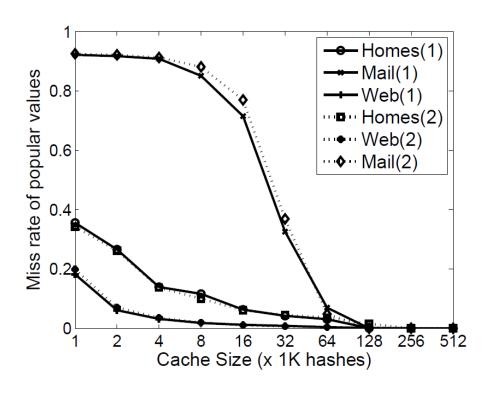
- **Temporal Value Locality**: If a certain value is accessed now, it is likely to be accessed again in the near future
- **■** Temporal value locality in real-world workloads
 - The metadata cache is managed as a queue with an LRU eviction policy
 - CDFs of number of writes of the value at the (i+1)st location in the LRU queue







Cache Miss Rate for Popular Values

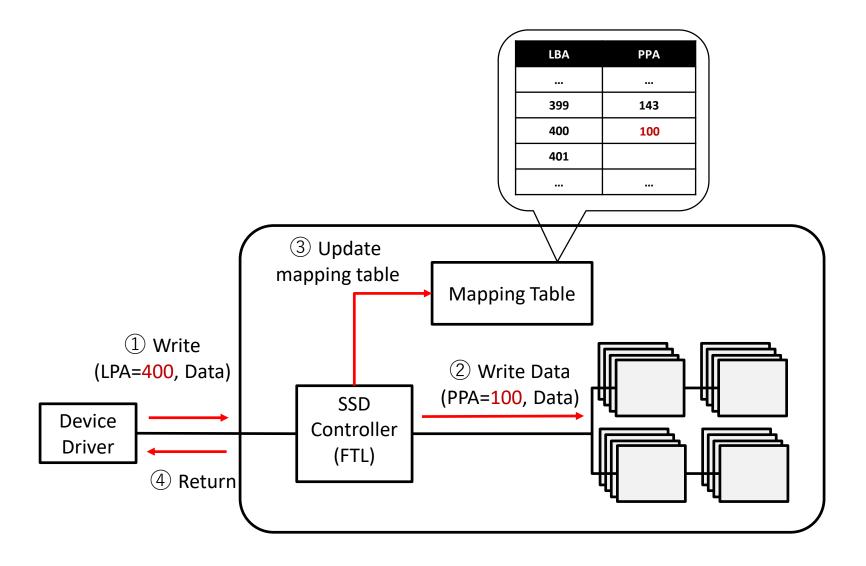


- Popular values represent the minimum number of values which account for 50% of accesses
- All the three traces achieve about 90% hit rate with 1.75 MB metadata cache (64 K hashes X 28 B = 1.75 MB)

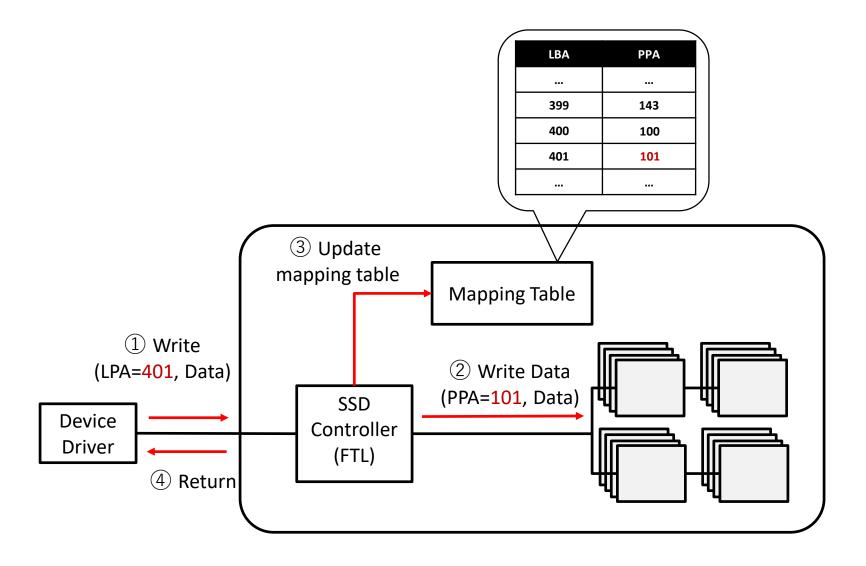
Content-Aware SSD

- The presence of value locality in a workload means that it preferentially accesses certain content over others
- This property facilitates data de-duplication inside an SSD
 - Store only a non-intersecting chunk having a unique hash value
 - Other data blocks with the same content point to the unique block

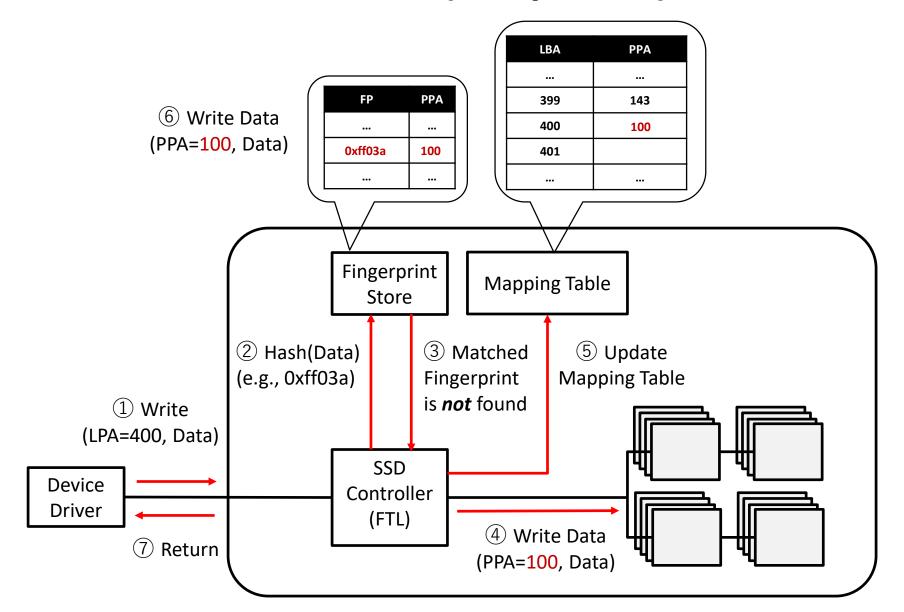
How a Conventional SSD Works?



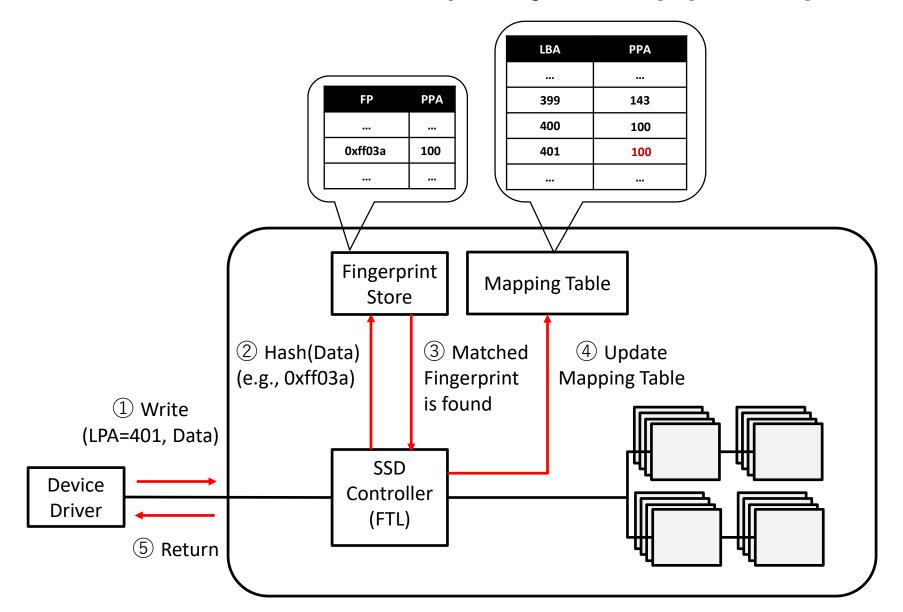
How a Conventional SSD Works?



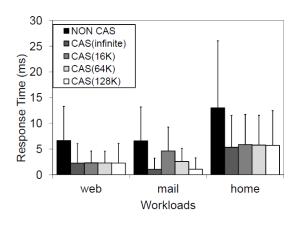
How CA-SSD Work? (Simplified)

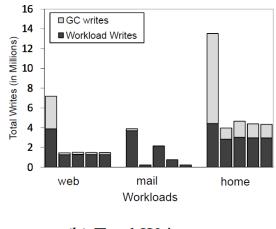


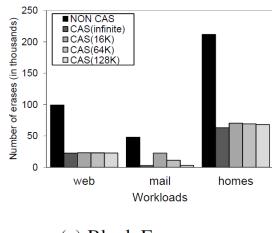
How CA-SSD Work? (Simplified) (Cont.)



Experimental Results







(a) Response Time

(b) Total Writes

- (c) Block Erases
- The reduction in write traffic: 77%, 93%, and 70% for web, mail, and home
- The write reduction benefits directly translate into the reduced response time and the reduced block erases
- CAS(128K) provides the performance close to the CA(infinite)
 - mail shows lower TVL and requires a larger metadata cache

Outline

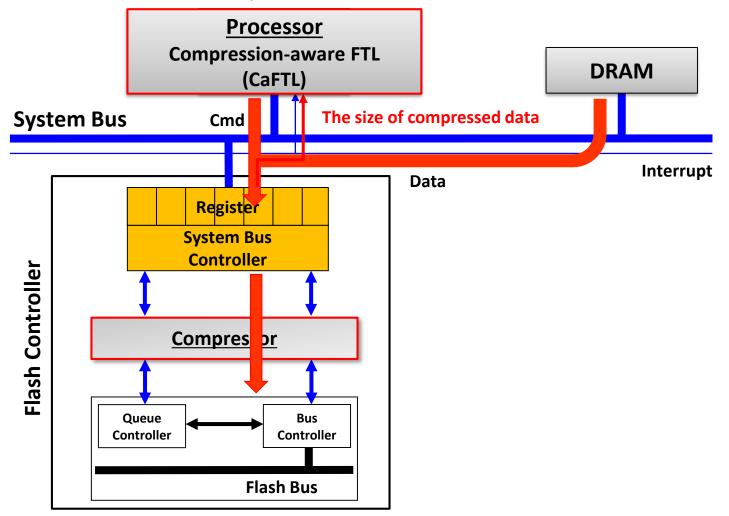
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BlueZIP: Hardware-accelerated Compression

- Requested data contain lots of repeated bit patterns
 - Lossless compression helps to eliminate such repeated bit patterns.
- BlueZIP: A hardware-accelerated compression technique,
 - Compress requested data at runtime
 - Use a hardware-accelerated compression module
 - Provide software support for maximizing the benefits of hardwareaccelerated compression
 - Improve the lifetime and performance of storage devices
 - Reduce the amount of data written to flash memory
 - Reduce the time taken to transfer data between a host system and flash memory

Overall Architecture of BlueZIP

Implement the hardware compression module inside the flash controller to reduce system bus traffic



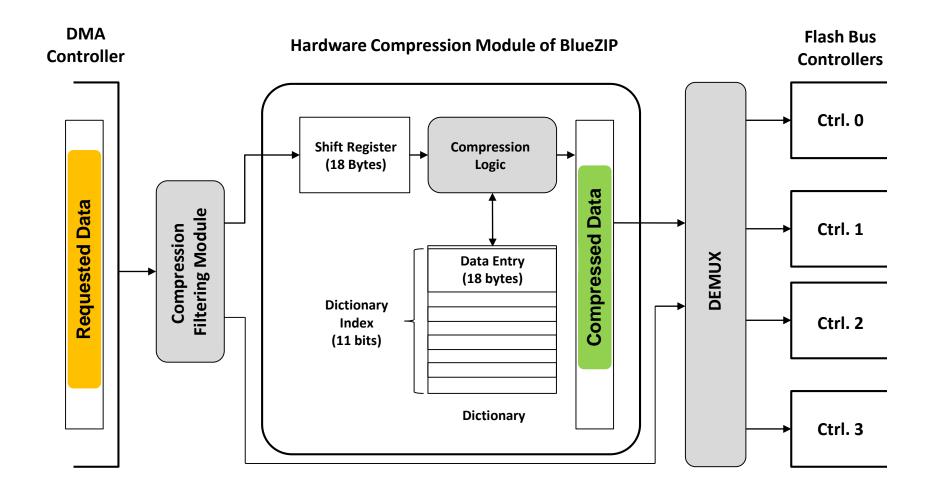
HW: Compression Algorithm

■ Use the LZRW3 compression algorithm

- Relatively high compression ratio
- Easy hardware implementation

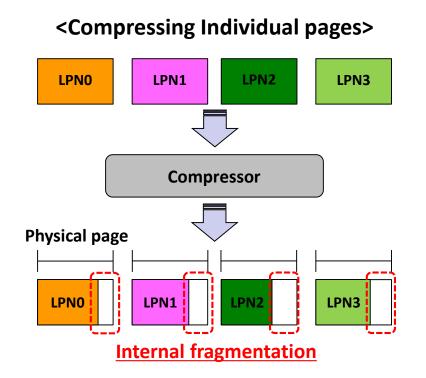
Compression Algorithm	Hardware Complexity	Performance (Cycles)	Compression Ratio	Description
X-Match	Low	1,024 cycles / 4 KB (20.48 us@ 50 MHz)	Low (e.g., 20%)	Hardware-Based Memory Compression
LZ77	High	4,096 cycles > / 4 KB	High (e.g., 50%)	File Compression
LZRW3	Middle	4,096 cycles / 4 KB (81.92 us@ 50 MHz)	High (e.g., 40%)	File Compression

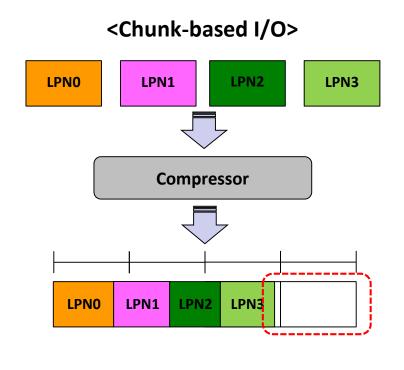
HW: Implementation



HW-SW: Compression Unit

- Chunk-based I/O
 - Compress several logical pages into several physical pages
 - Mitigate the internal fragmentation problem
- Read performance penalty
 - Use a read buffer that holds previously decompressed pages





SW: Compression-Aware FTL

Write Buffer



Page Mapping Table

Data Chunk Table

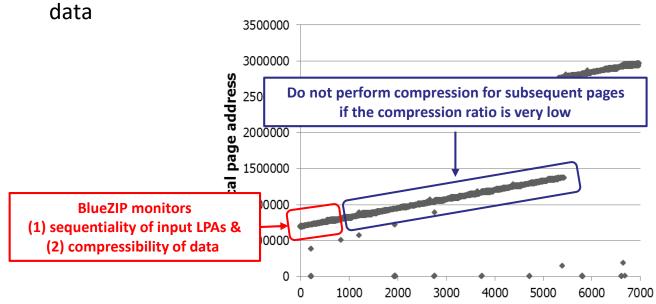
Logical Page Address	Physical Page Address		Physical Page Address		Valid ge Counter	No. of Physical Pag		mpression ndicator
:	:		÷		÷	:		:
301	2311		2311		4	3		1
302	2311	_ <u> </u>	2312		4	3		1
<i>5</i> 07	2244	_	2313		4	3		1
507	2311							
508	2311				:	:		:
Compression Chunk (3 pages) Physical Page Address: 2311 2312 2313								
		301	302	507	508			

SW: Selective Compression

- The data size expansion problem
 - The size of compressed data > the size of original data
 - Degrade storage performance and lifetime
 - Usually observed in writing multimedia files
- Selective compression

Detect poorly compressed sequential writes in advance and do not compress those

Time (sec)



Experimental Settings

Benchmarks

Benchmark	Description	Compression Ratio
SENSOR	A set of sensor data files which were collected during a semiconductor fabrication process	Very high
LINUX	A subset of the Linux kernel 2.6.32 source files	high
DOCUMENT	A set of documents and image files	medium
MP3	A set of MP3 files already highly compressed	Very low

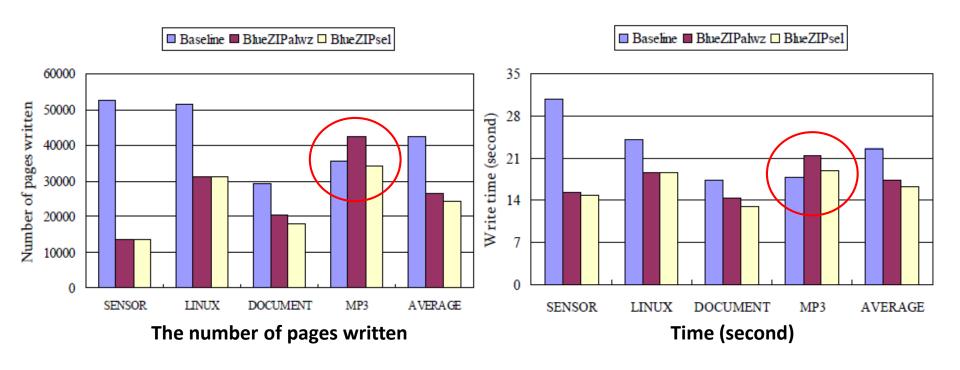
SSD Configurations

Configurations	Description	
Baseline	Use no lossless compression	
BlueZIPalwz	Use lossless compression all the time	
BlueZIPsel	Use selective compression	



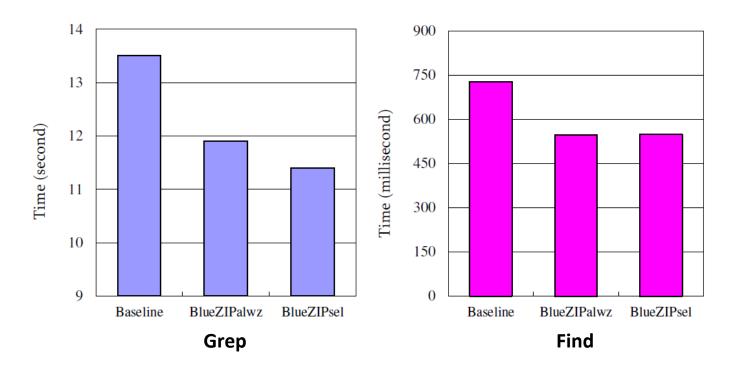
<BlueZIP prototype>

Experimental Results



- BlueZIPsel writes 38% less data over Baseline.
 - The amount of written data is increased with BlueZIPalwz due to the data expansion problem of lossless compression.
- BlueZIPsel achieves 17%-50% higher performance than Baseline

Read Performance



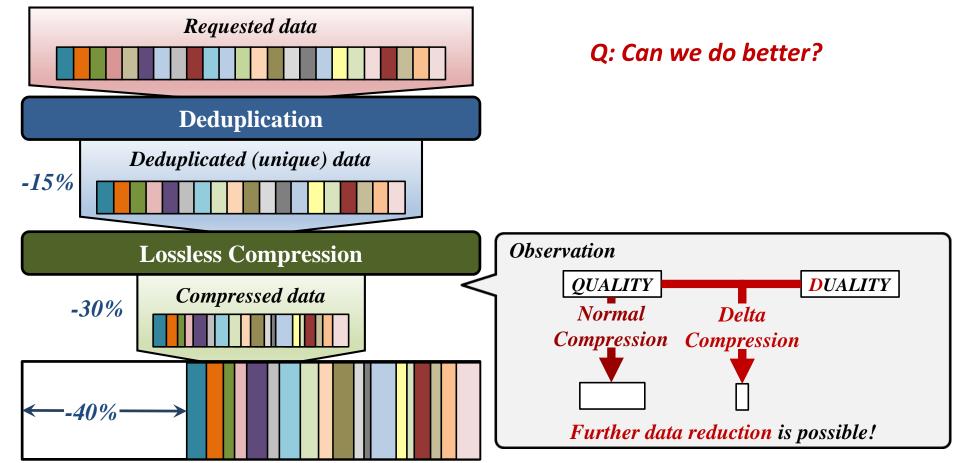
- BlueZIPsel improves the overall read performance by 20% on average.
- The reduction in the number of pages sufficiently offsets the decompression overhead.

Outline

- Data Deduplication
- **Data Compression**
- Case Studies
 - CA-SSD: Content-aware Solid-state Drives
 - BlueZIP: Hardware-accelerated Compression
 - DAC: Dedup-assisted Compression

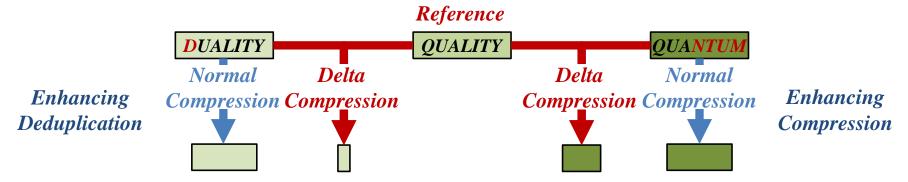
Integration of Capacity-optimization Techniques

- **■** Maximizing the data compression ratio
 - Improvements can be accumulated

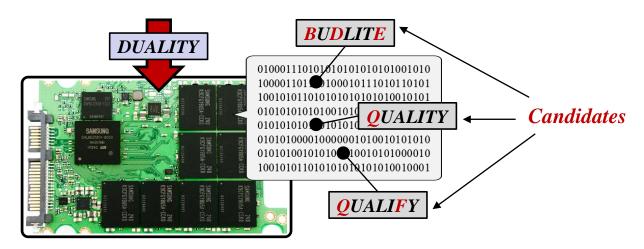


DAC: Dedup-Assisted Compression

- A novel integrated data reduction technique
 - Exploiting data similarity to further increase data reduction ratio



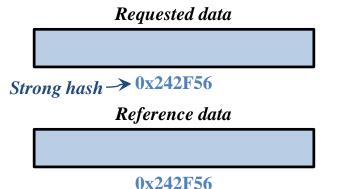
Reference search is not easy



DAC: Reference Data Search

Exploiting partial fingerprints (hashes) of data

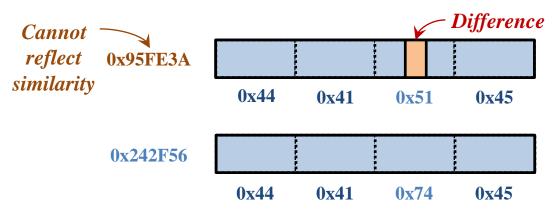
Generated by the strong hash module for deduplication (Dedup-Assisted)



Fingerprint (FP) Store

Page FP	Page Addr		
•••	•••		
0x141250	0x022		
0x242F56	0x0F1		
0x44516F	0x011		
0x6DFF25	0x1CB		
•••	•••		

Deduplication



Sub-Page Fingerprint Store

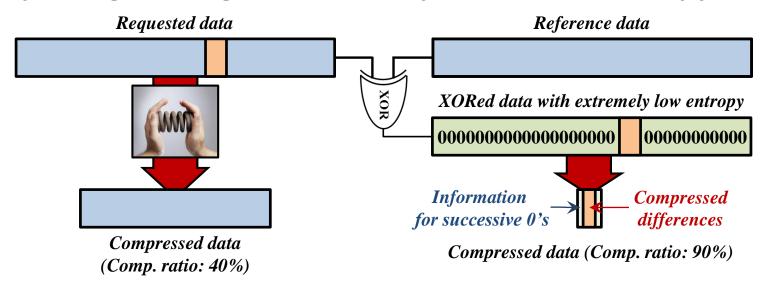
	1st FP	2nd FP	3rd FP	4th FP	Page Addr
	•••	•••	•••	•••	•••
<i>50%</i>	0x44	0x2F	0x51	0x7E	0x02F
<i>75%</i>	0x44	0x41	0x74	0x45	0x0F1
25%	0x44	0x5E	0x3C	0x7A	0x012
0%	0x45	0x0D	0x45	0x6D	0x13B
	•••	•••	•••	•••	•••

Dedup-Assisted Compression

Wide searching range, constant searching time, and quantified data similarity

DAC: Data Compression with a Reference

Exploiting XOR logic to extremely reduce data entropy



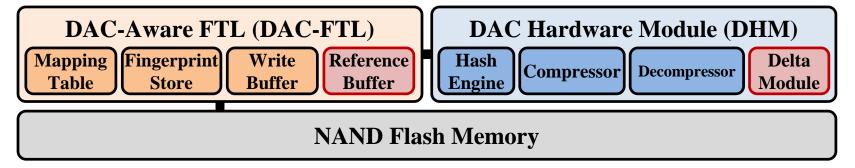
Lossless compression algorithm: XmatchPRO

- Employing both the run-length and dictionary-based data encoding
- Easily implementable in H/W
- High performance and comparable compression ratio

Higher compression efficiency w/o an additional delta-aware data encoding

Experimental Environment

- Target SSD Architecture
 - Recent high-end SSDs employing deduplication and compression

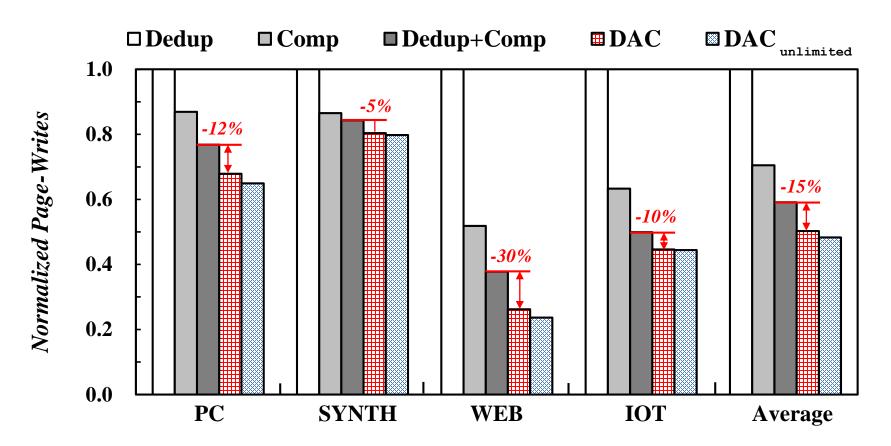


- Evaluation on a storage emulation environment
 - Hardware modules for a strong hash function (MD5) and lossless compression (XmatchPRO) were emulated by software.
- Benchmarks: block I/O traces collected form a high-end PC

Benchmark	PC	SYNTH	WEB	IOT
Data Redundancy	Moderate	Low	Very High	High
Data Compressibility	Moderate	Low	Very High	High

Evaluation Result

- Compared with a combination of deduplication and lossless compression,
 DAC reduces write traffic by up to 30% and 15% on average
- DAC can further increase data reduction whether data entropy is high or not



End of Chapter 10