LeaFTL: A Learning-based Flash Translation Layer for Solid-State Drives

Jinghan Sun, Shaobo Li, Yunxin Sun, Chao Sun, Dejan Vucinic, and Jian Huang. 2023.

In Proceedings of the 28th ACM International Conference on Architectural Support for Programming Languages and Operating Systems, Volume 2 (ASPLOS 2023).

2024. 08. 07

Presentation by Yeongyu Choi

choiyg@dankook.ac.kr

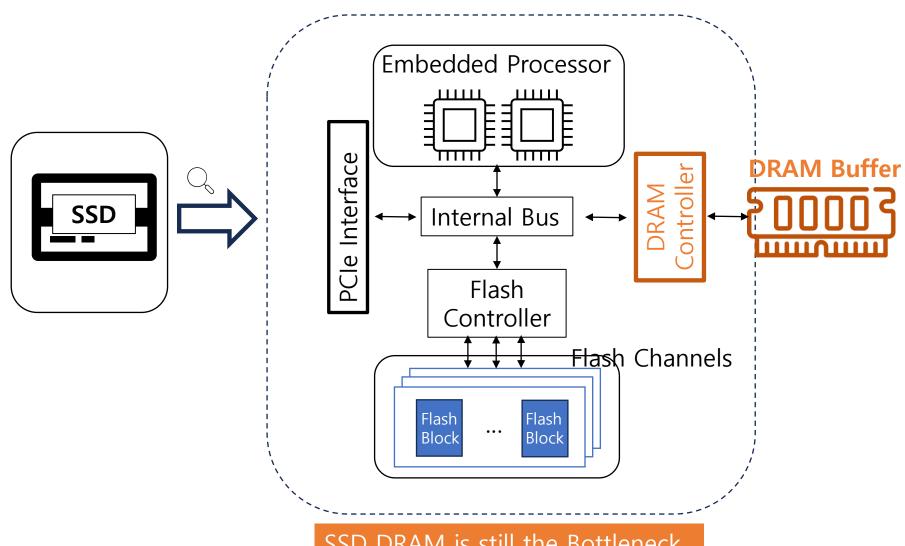




Contents

- 1. Introduction
- 2. Background
- 3. LeaFTL
- 4. Evaluation
- 5. Conclusion

Introduction



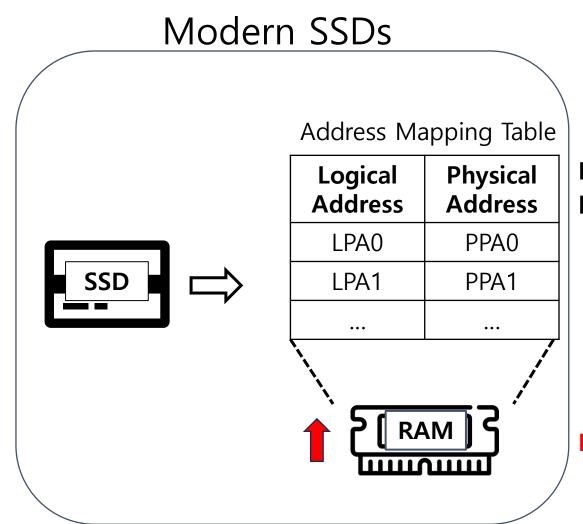
Less Scalable

- High Cost
- Limited Capacity

SSD DRAM is still the Bottleneck



Introduction



Direct LPA-PPA mappings require 8 Bytes per mapping

DRAM Capacity can't keep up with SSD Capacity

Approaches on Mappings

Page-Level Mapping:

Direct LPA-PPA mapping for fast lookup

Entire mapping table requires large storage space



Reduce mapping table size by storing via blocks

Overhead in Lookup in flash page

Hybrid Mapping:

Takes advantage of a mix of Pagelevel and Block-level Mappings

Incurs in significant GC Overhead







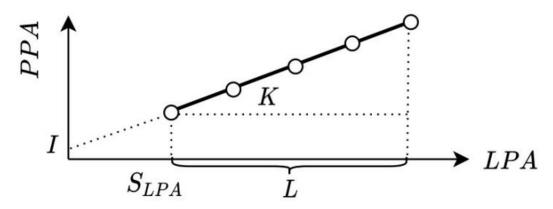
Instead of 1:1 mapping in Page-level Mapping

Exploit learning techniques to identify various LPA-PPA mapping patterns

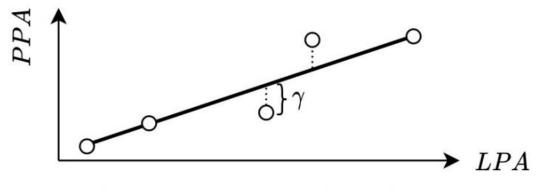




Learning-Based Techniques

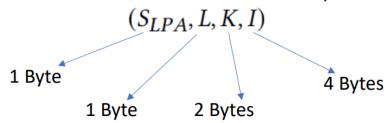


(a) Precise Linear Approximation



(b) Inaccurate Linear Approximation

Learned index are encoded in 8 bytes:



$$PPA = f(LPA) = \lceil K * LPA + I \rceil$$

where $LPA \in [S_{LPA}, S_{LPA} + L]$

Learned Index Segment

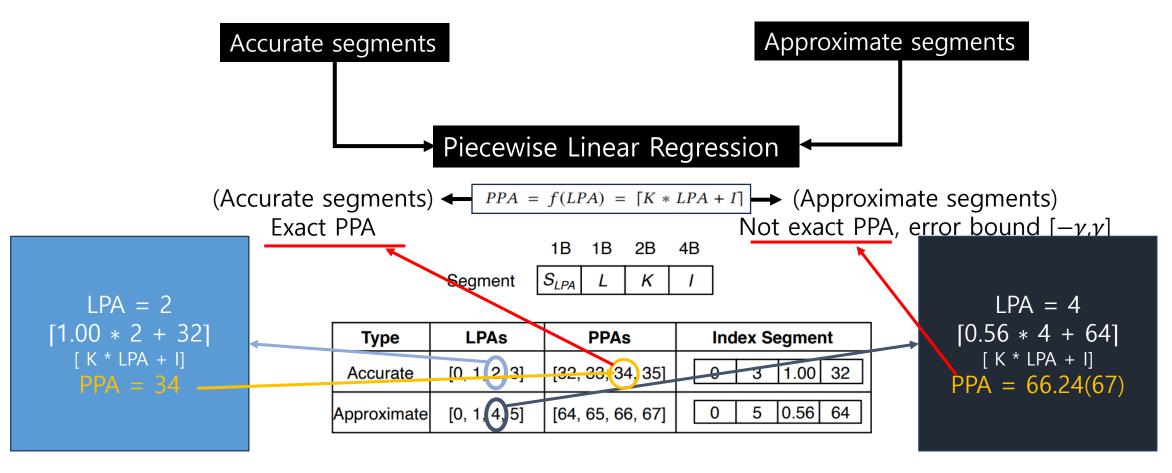
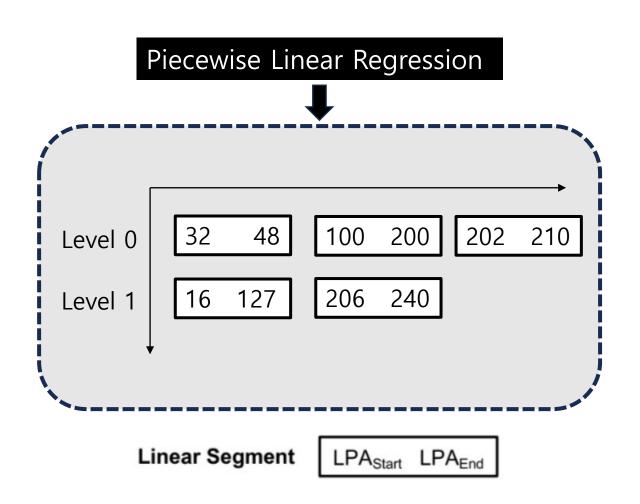


Figure 6: Types of learned segments in LeaFTL.

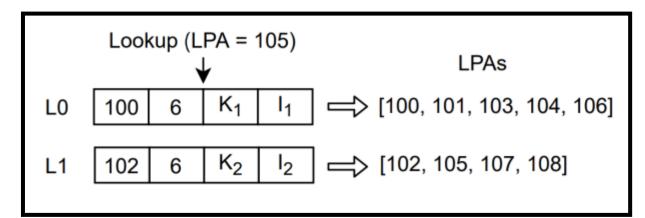


Log-Structed Address Mapping Table



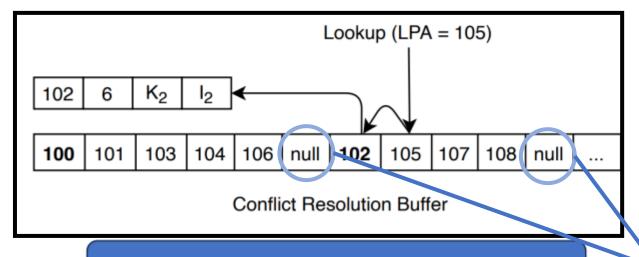
- Non-overlapping segments are sorted by their LPA ranges
- Overlapping segments are allowed across levels

Conflict Resolution Buffer



Possible to get inaccurate PPA

- Affect to segment compaction



- CRB is a nearly sorted list, follows its insertion deletion and lookup are fast

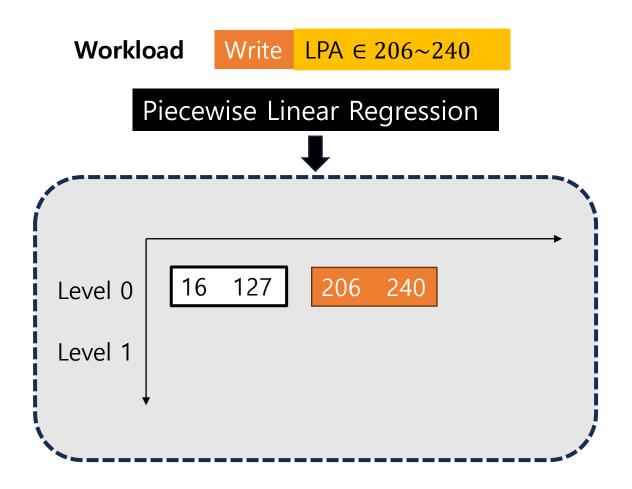
- Each LPA takes only 1 byte and its structured guarantees no redundant LPAs

Very space efficient with at most 256 entries

Separate segments

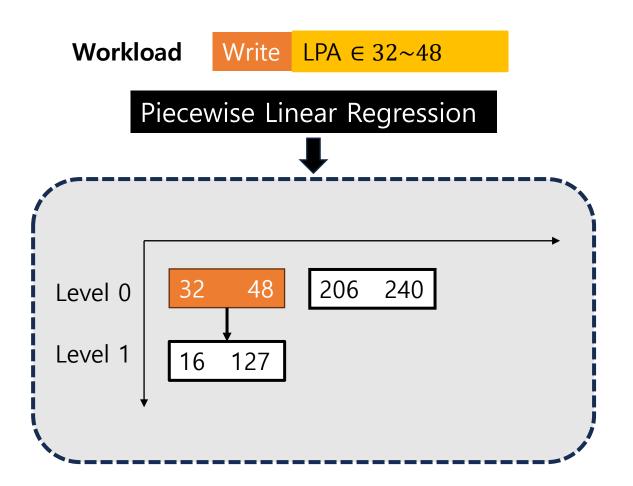






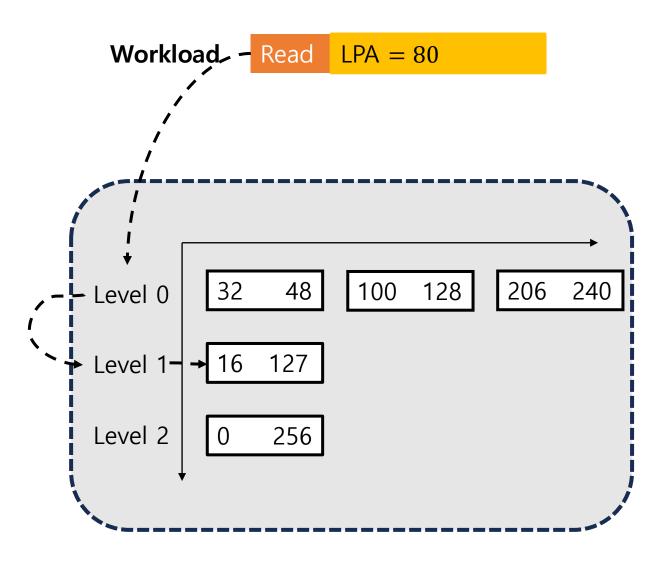
New segments are appended to the topmost level

If it is approximate it is added to CRB

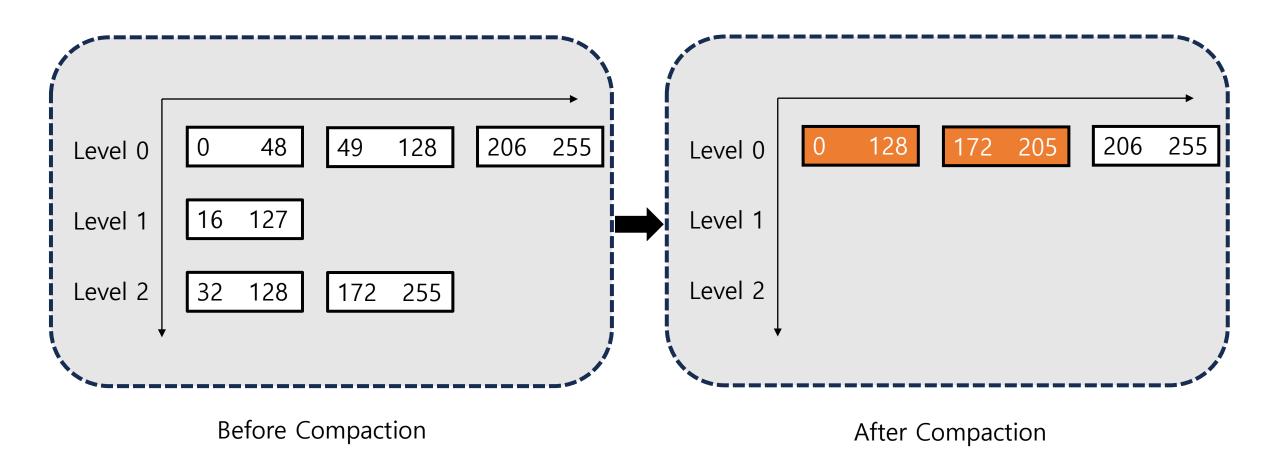


New segments are appended to the topmost level

If it is approximate it is added to CRB

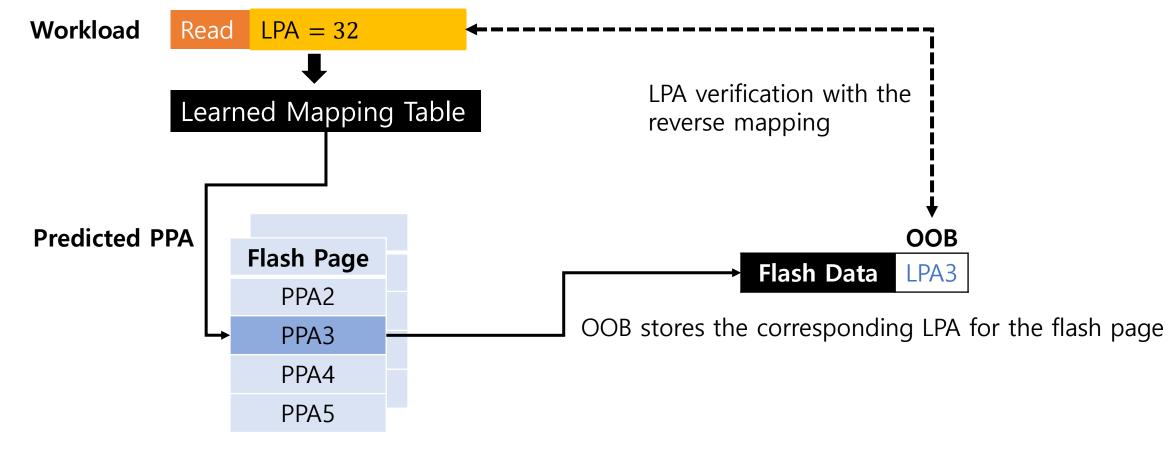


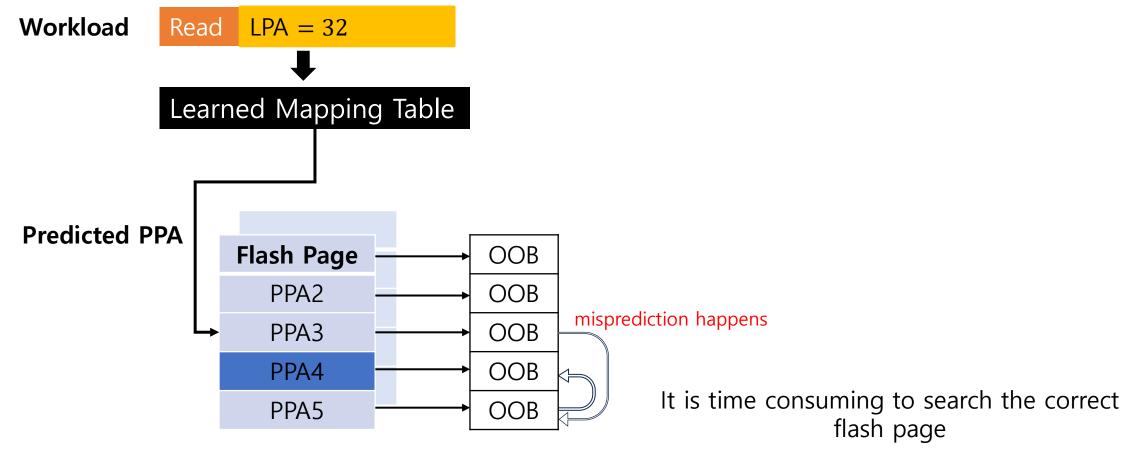
Search from the topmost level, and always find the first matching segment

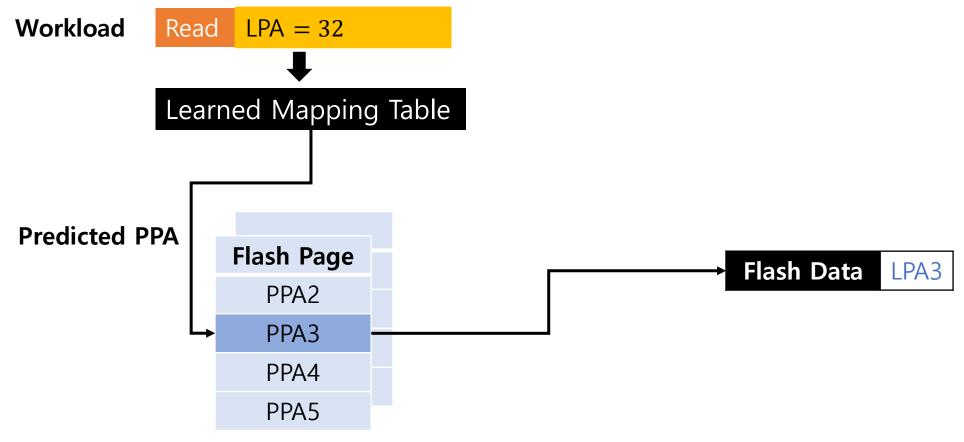


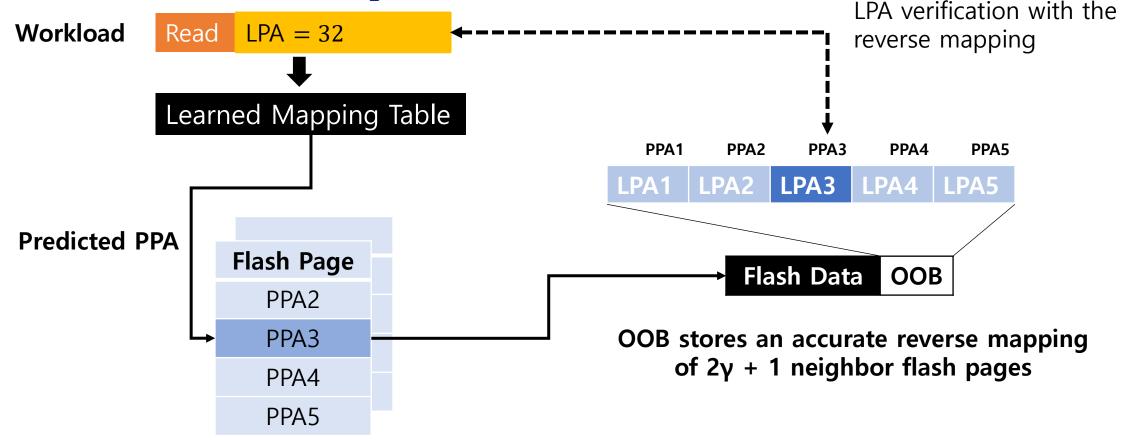


Misprediction Verification with OOB Metadata (OOB:Out-of-Band)

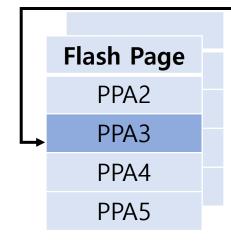






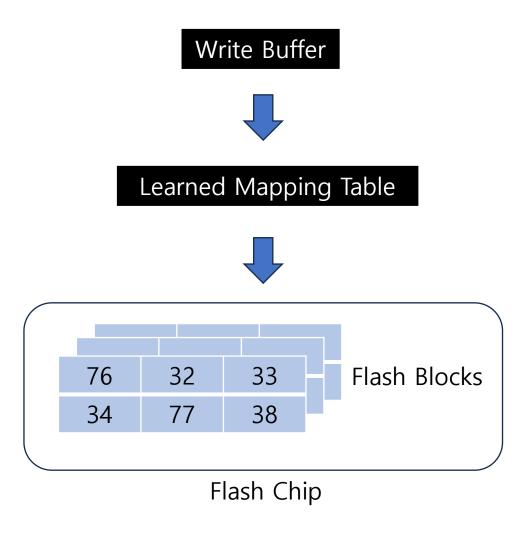






Reducing the misprediction penalty from $log(\gamma)$ flash page reads to only one flash read

Optimized Flash Allocation



Flushed flash pages

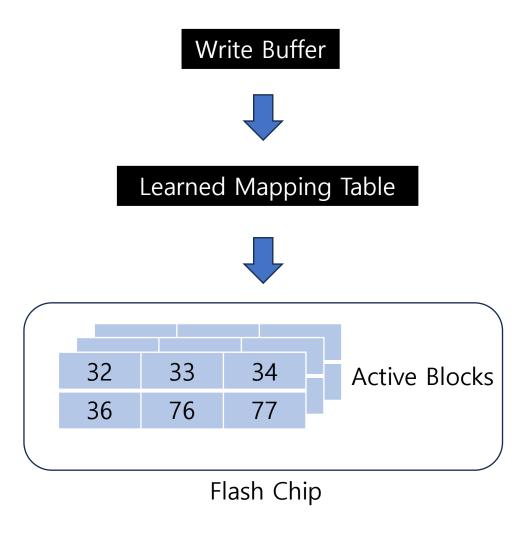


Four new linear segments are created



Flushing flash pages from the write buffer directly is less optimal

Optimized Flash Allocation



Flushed flash pages

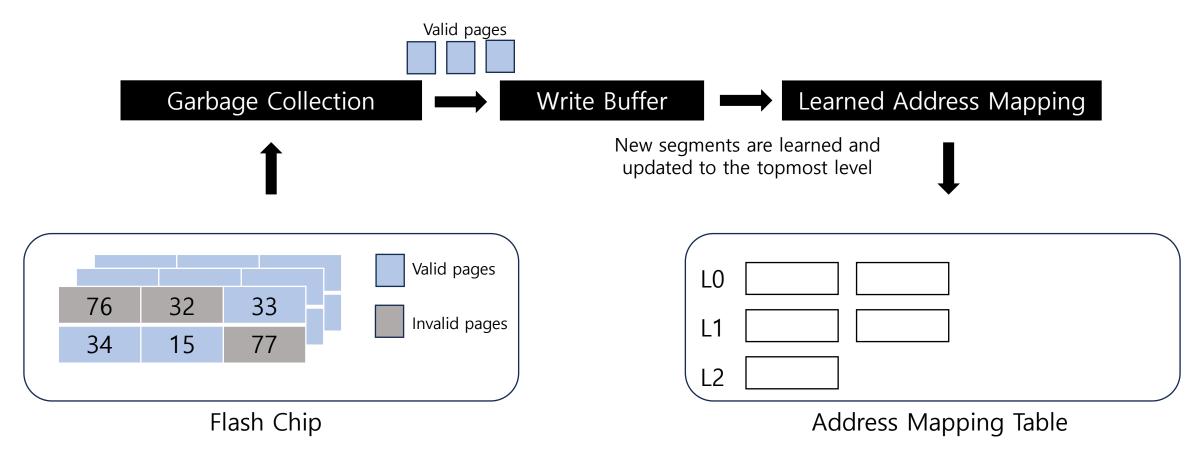


Fewer linear segments are created

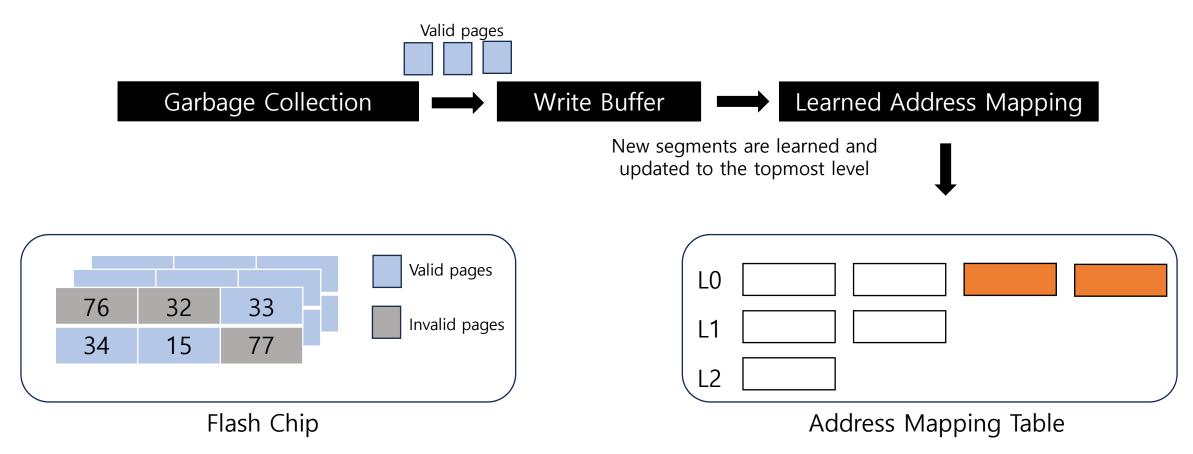


Optimization: Flash pages are sorted by their LPAs before flushed from the write buffer

Coordinated Garbage Collection



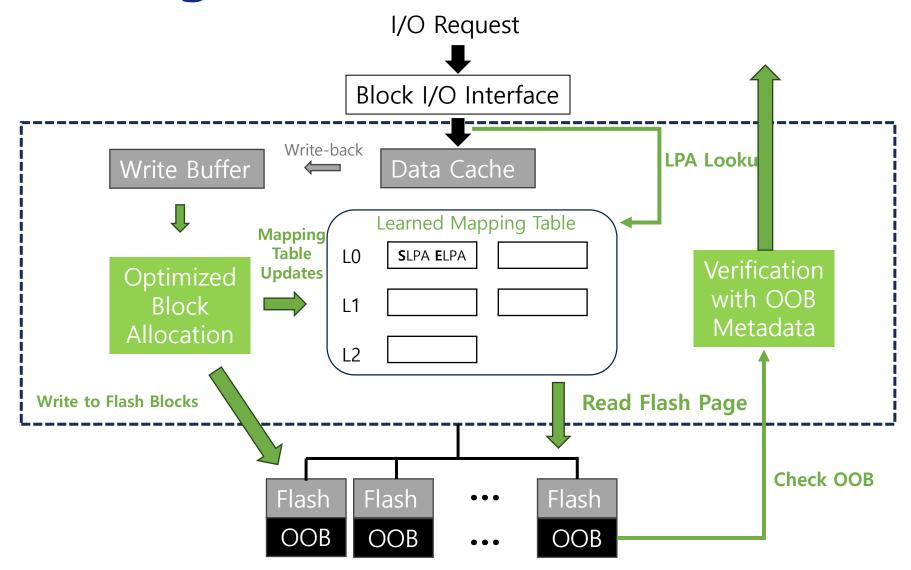
Coordinated Garbage Collection



LeaFTL learns new segments to avoid messing up existing segments in GC



Put It All Together



Implementation Details

- WiscSim Simulator
- A Real 1TB Open-Channel SSD with 16 channels

SSD configuration

Parameter	Value	Parameter	Value
Capacity	2TB	#Channels	16
Page size	4KB	OOB Size	128B
DRAM size	1GB	Pages/block	256
Read latency	20us	Write latency	200us
Erase	1.5ms	Overprovisioning ratio	20%

Workloads

- Block I/O Traces: from enterprise servers and university servers
- Data-Intensive Applications: FileBench, BenchBase

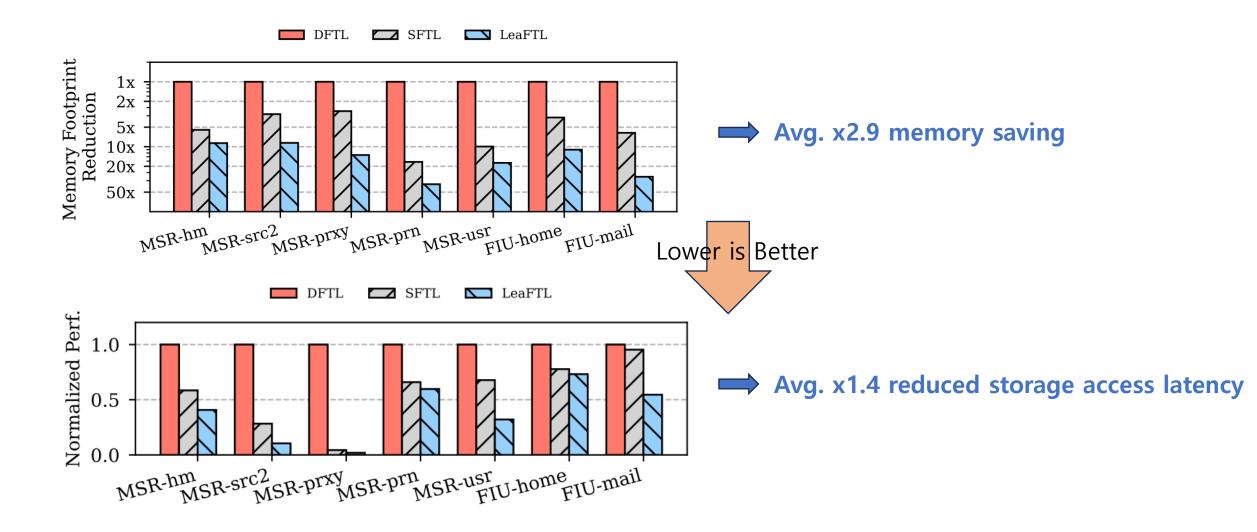
Comparison

- DFTL: Demand-based caching FTL
- S-FTL: Compresses sequential LPA-PPA entries
- LeaFTL with different error bounds (0, 1, 4, 8, 16)

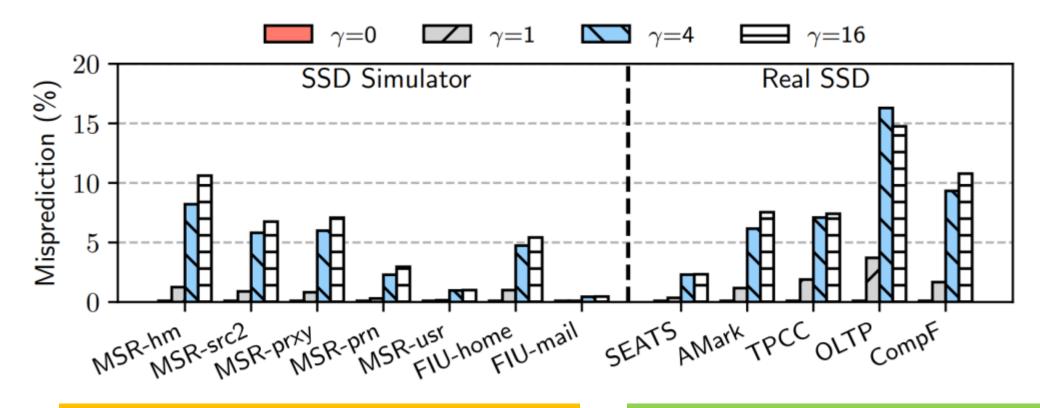




Memory Footprint Analysis & Performance Analysis



Misprediction

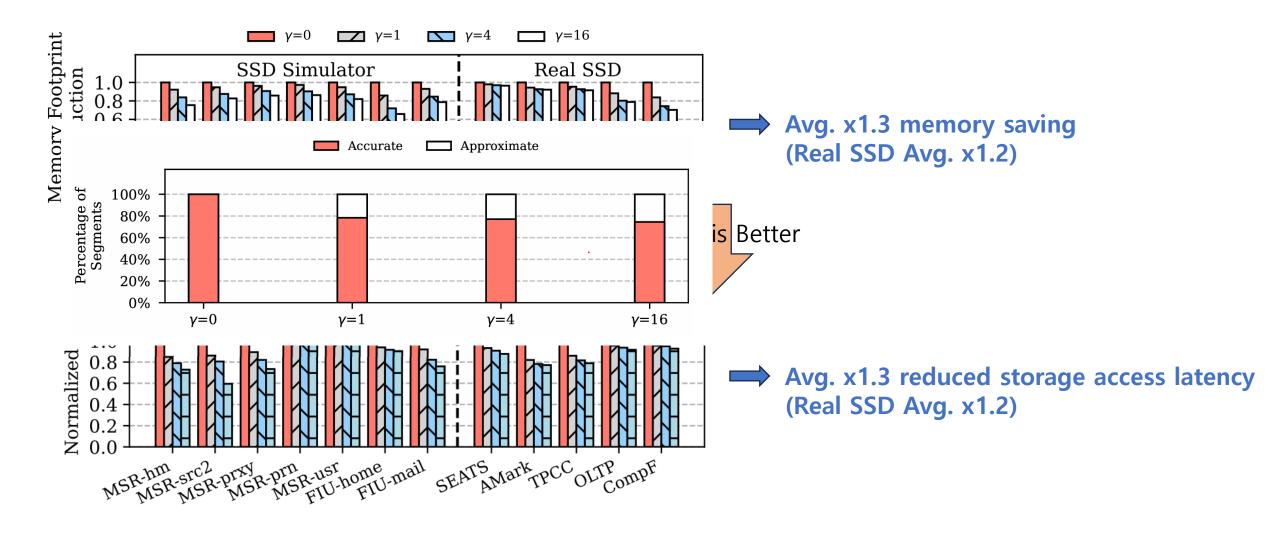


Error bound $\gamma = 16$, misprediction ratio < 10% Avg.

Every misprediction incurs in only one additional flash read



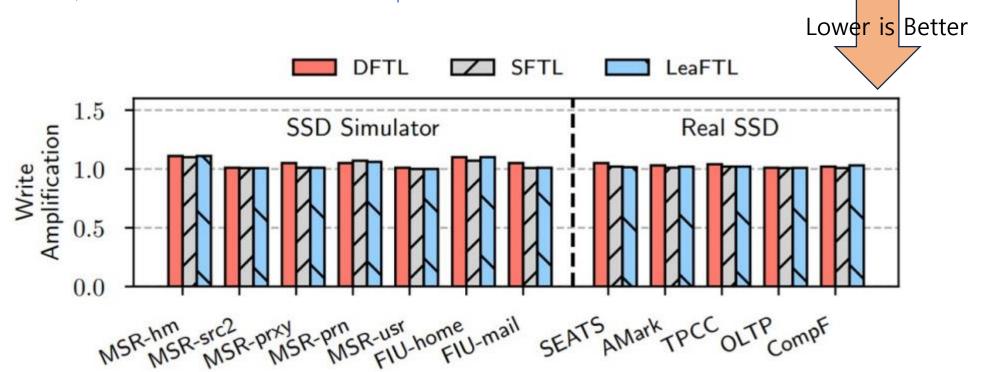
Memory Footprint Analysis on Different γ





Write Amplification Factor Analysis

WAF, the ratio between the actual and requested flash writes



Conclusion

 LeaFTL uses a simple but effective learning-based technique to reduce memory consumption

 LeaFTL stores learned segments in a log-structured manner to avoid relearning

LeaFTL uses OOB metadata to verify its address translation

LeaFTL consumes 2.9x less memory and improves performance by 1.4x

Thank you



