Two is Better Than One: The Case for 2-Tree for Skewed Data Set

Xinjing Zhou, Xiangyao Yu, Goetz Graefe, Michael Stonebraker

MIT University of Google MIT Wisconsin-M adison

Background

- Tree data structures in data systems
 - Access method and storage
 - Buffer-managed





Buffer Pool Utilization Issue: B-tree

- Buffer pool memory utilization issue on **skewed workloads** for B-tree
- A few hot records in a B-tree page with many cold ones
- Records keyed on domains with no spatial locality could spread across the key space
 - e.g., user-id



Buffer Pool Utilization Issue: LSM-tree

- Partially mitigates the memory utilization issue
- Frequently-updated data tend to cluster around the top of tree hierarchy
 - Good buffer pool utilization on skewed reads Ο
- Stationary data get migrated down to the bottom
 - Poor utilization on skewed reads \bigcirc



How to improve the memory utilization of buffer pool for tree structures on skewed workloads?

This Work

- Principles
 - Multi-structuring to physically separate hot data from cold data
 - Actively migrate data at record-level in both direction
- Applications to B-tree and LSM-tree

1. 2B-tree for Disk-based DBMS

- Compose two b-trees
 - Exposing single tree interface logically
- Record migration between two trees
- Size the hot tree to be close to buffer pool capacity
- Intuition: vast majority of the accesses go to the hot b-tree => increased utilization



Uownward Data Migration

- Goal: evict cold records with **low overhead**
- Trigger: hot tree fills up
- Leverage efficient range scans to approximate a clock replacement



Over Content of Conte

- Goal: Migrate only **warm** records upwards to the hot tree to reduce churns
- Probabilistic approach
 - Migrate a sampled set of accesses to the cold tree
 - Intuition: warm records will be more likely to be sampled



Experimental Setup

- 2B-tree implemented using LeanStore buffer pool/B-tree
- 1GB buffer pool and 5GB data set
 - 20M records, 256-byte each.
 - 16KB page size.
- Hot tree sized to be about 90% of the buffer pool capacity
- Probabilistic sampling rate: 0.5
- Workloads:
 - YCSB hotspot
 - Vary the working set size
 - YCSB zipfian
 - Working set covers all data with zipfian access distribution

Point Operations



Range Scan Operations

• On-par with single b+tree throughput for range scan.

--- B+tree -E- 2B+tree



Summary: 2B+tree Improves upon Point Operations

	Point Read	Point Update	Range Scan
B+tree	*	*	***
2B+tree	***	***	***

2. Generalize to a N-Tree using LSM-tree

- A specific N-tree design using LSM-tree
 - Write-optimized Ο
 - Better skew reads Ο
- Augment LSM-tree with upward record migration for reads
 - Actively bring stationary but warm records Ο closer to the top of hierarchy Upward

Record



Upward Migration in LSM-tree

- How to identify stationary and warm records?
- Two heuristics:
 - 1. Upon block cache miss: Likely in the bottom of the hierarchy
 - 2. Apply the probabilistic upward migration to identify warm records

Experimental Setup

- We call our proposal UpLSM-tree, built on RocksDB
- Comparison using 1 GB memory budget
 - UpLSM-tree: 1GB block cache
 - Vanilla LSM-tree: 1GB block cache
 - LSM-tree with in-memory row cache for level files on disk
 - 90% memory allocated to row cache
 - 10% memory allocated to block cache

Point Operations

YCSB-Hotspot



YCSB-Zipfian

17

Range Scan Operations

- On-par with vanilla LSM-tree
- Much better than LSM-tree with row caching
 - Row cache only helps point read operations
 - Block caching is more versatile



Summary: UpLSM-tree Dominates both Baselines

	Point Read	Point Update	Range Scan
LSM-tree	*	***	***
LSM-tree w/ RC	***	*	*
UpLSM-tree	***	***	***

Future Work

- Extend the architecture to non-tree-based structures
 - Hashing
 - Heap file with secondary indexes

Conclusion

- We studied improving memory utilization for tree data structures in data systems
 - Multi-structuring
 - Record-level migration
- Applications
 - 2B+tree for Traditional DBMS
 - A better LSM-tree
- Source Code: https://github.com/zxjcarrot/2-Tree

BACKUP SLIDES

Durability and Recovery of Migration

- No logical data changes
- Migration operation uses lightweight systems transaction
 - does not need to force log records to disk
 - Log persistence piggybacked on user commits
 - Analogy: btree split

2. Optimizing 2-Tree for Main-Memory DBMS

- Hot tree could stay completely in main-memory
 - Spill cold records out to disk
 - Serve larger-than-memory dataset
- A better Anti-Caching implementation
 - Scale to larger data set
 - Efficient range scan
- More details in the paper



Comparison

- Hot tree implemented using an in-memory btree, no buffer pool overhead
- Baseline: Original Anti-Caching implementation
 - LRU-based record eviction
 - Records stored unordered on disk

Point Operations

YCSB-Hotspot



YCSB-Zipfian

26

Range Scan Operations

• Significantly outperformed Anti-Caching original design

