

REMIX: Efficient Range Query for LSM-trees

Wenshao Zhong* Chen Chen* Xingbo Wu* Song Jiang†

*University of Illinois at Chicago

†University of Texas at Arlington

LSM-Tree



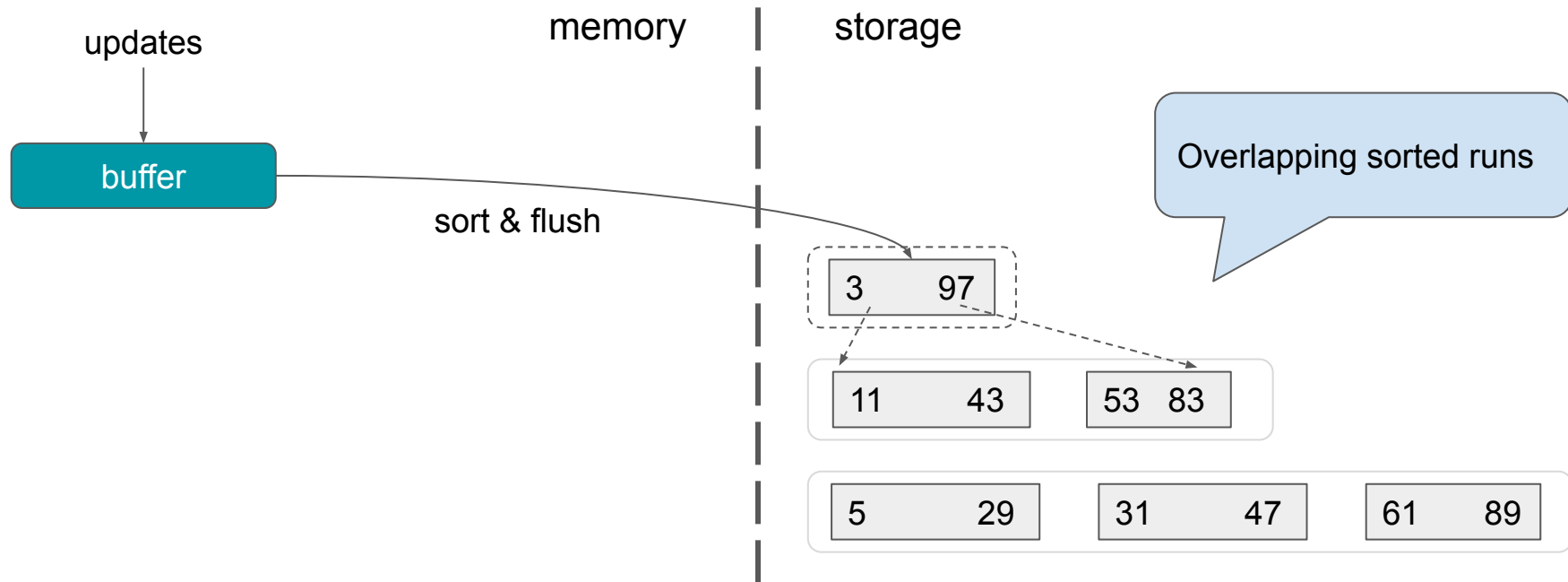
LSM-trees are the backbone of many database systems:

- Key-value stores
- Relational databases
- Time-series databases

They are the building blocks of social media, e-commerce, real-time analytics...

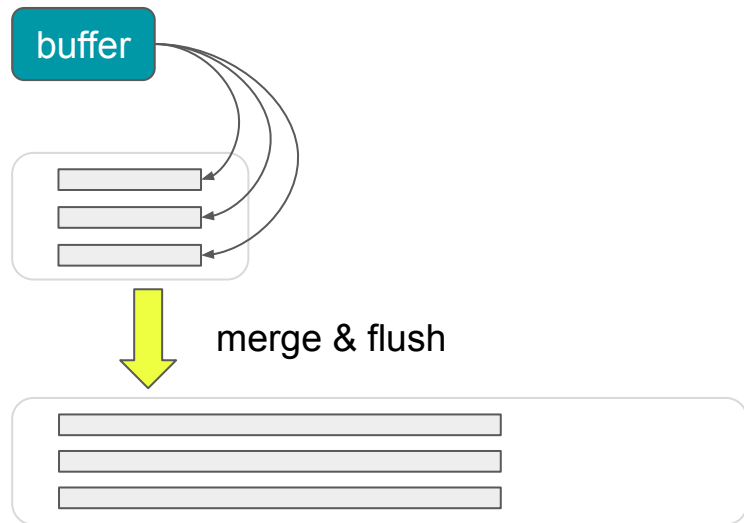
LSM-tree: **L**og-**s**tructured **M**erge-tree, a write-optimized data structure

An LSM-Tree



Compaction Policies

Tiered compaction



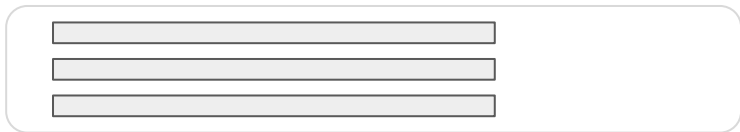
Leveled compaction



Compaction Policies

Tiered compaction

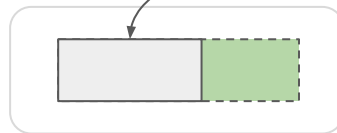
buffer



Lower insert cost
More sorted runs

Leveled compaction

buffer

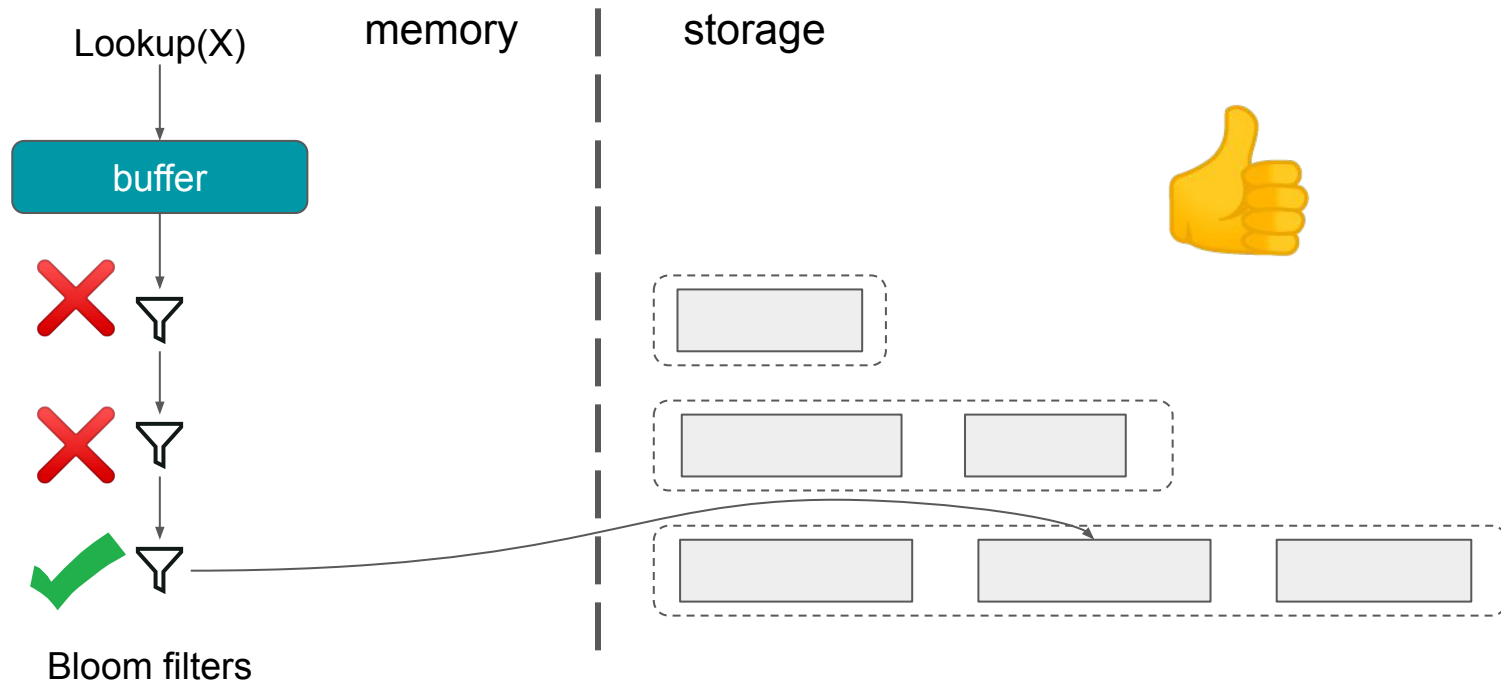


flush & merge

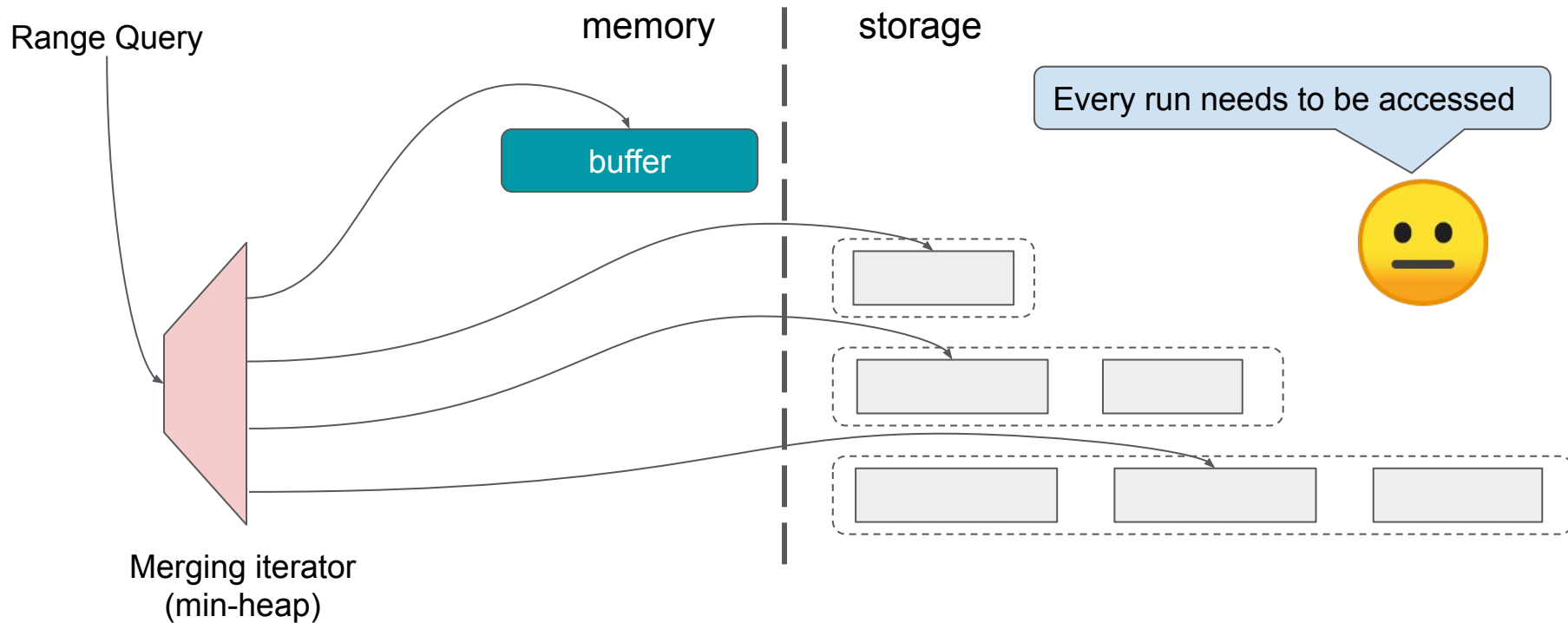


Higher insert cost
Fewer sorted runs

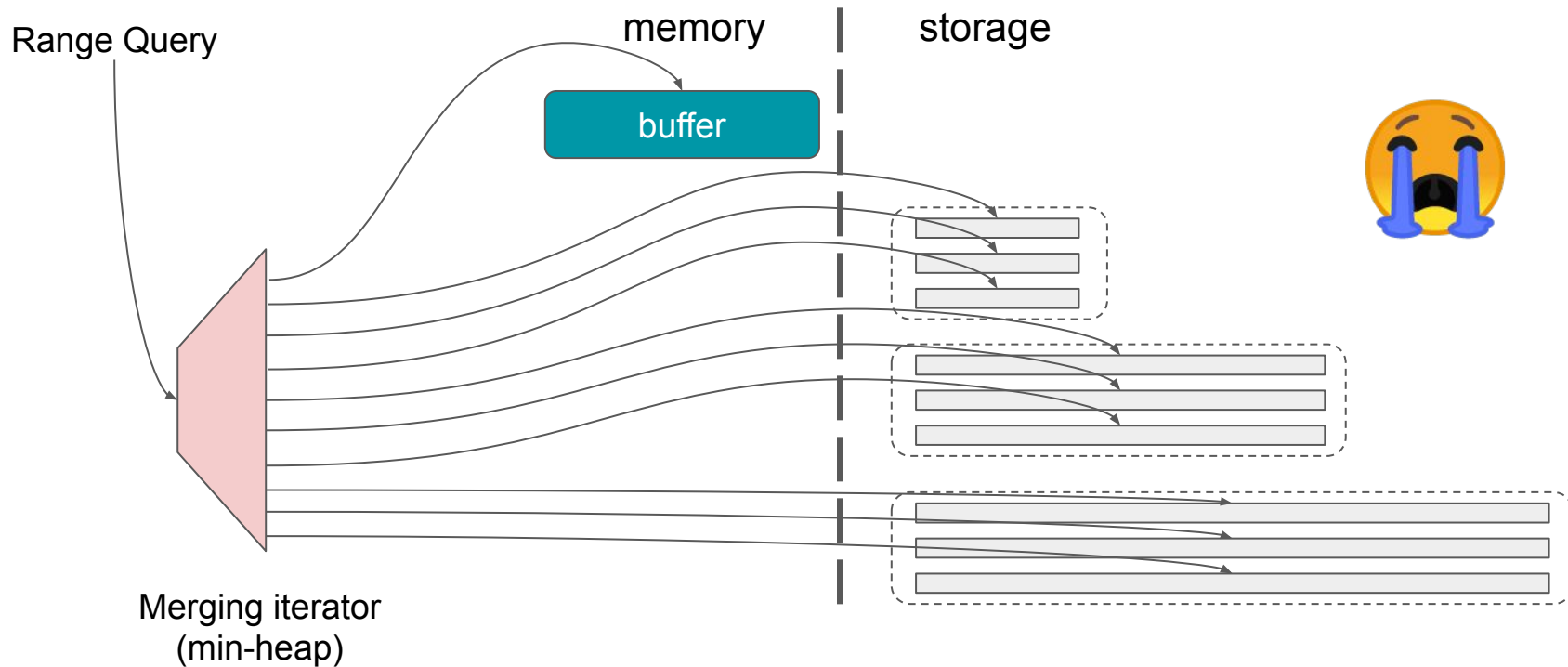
Point Query



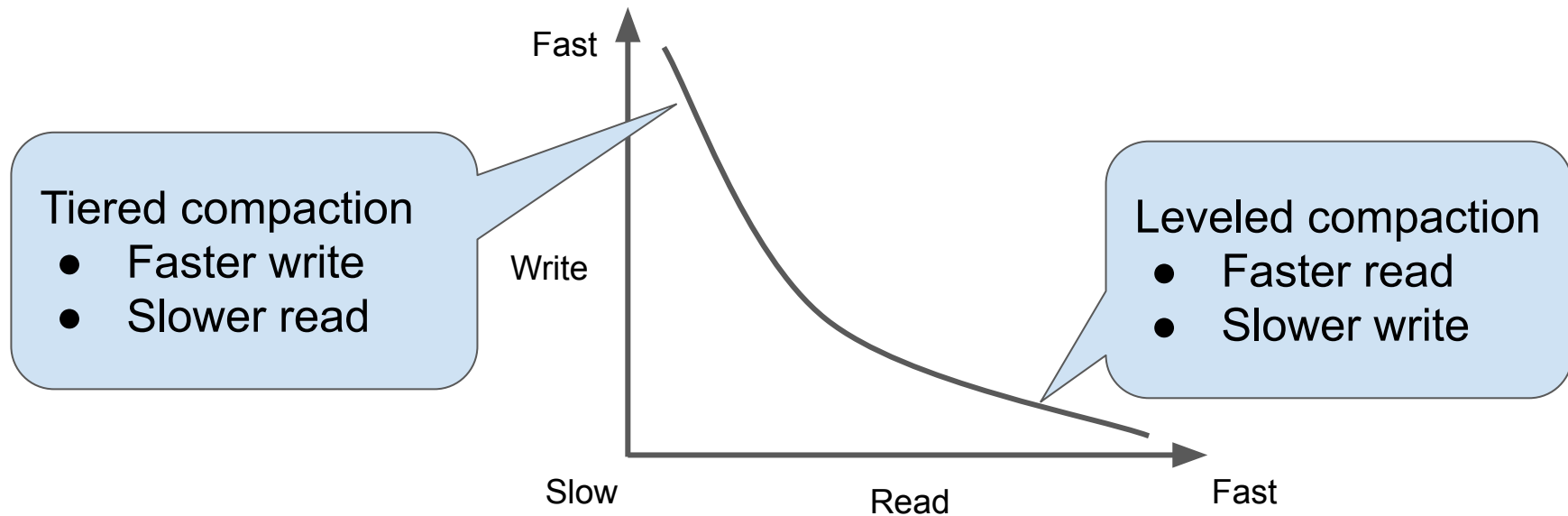
Range Query---Leveled Compaction



Range Query---Tiered Compaction



Trade-Off



Can we achieve fast read and write at the same time?

Tiered Compaction

- Accumulates and merges runs in batch --- efficient writes 😄
- Maintains multiple runs --- slow range queries 😞

Can we achieve fast range query with multiple runs?

- Let us understand why a range query can be expensive.

Range Query using a Min-Heap

```
range_scan(min_key, max_key)  
Example: range_scan(min="30", max="50")
```

A min-heap

cursor_offset: 3
(key 71)

cursor_offset: 4
(key 73)

cursor_offset: 1
(key 31)

Outputs: _____

(1) Seek: find the smallest key
>= start_key on each run

R₀

2	11	23	71	91
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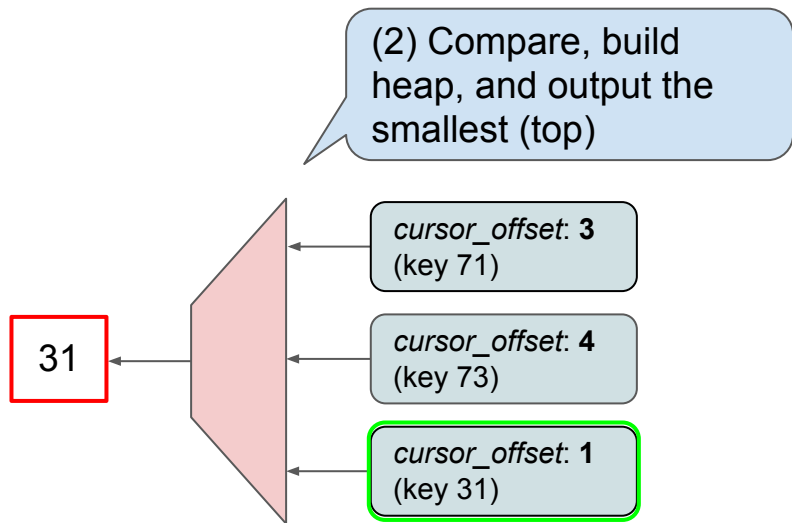
R₁

6	7	17	29	73
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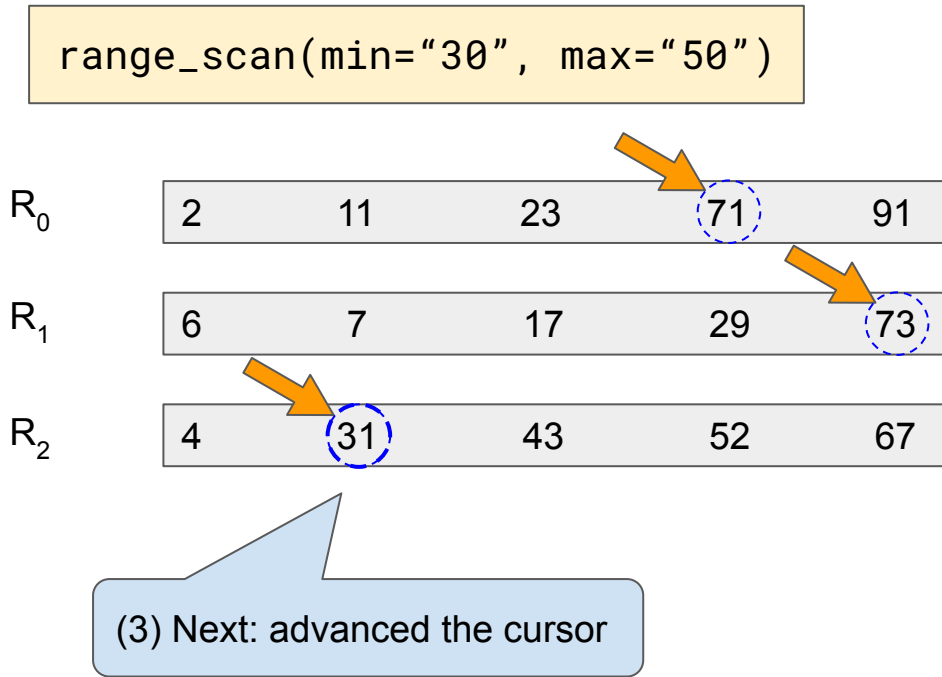
R₂

4	31	43	52	67
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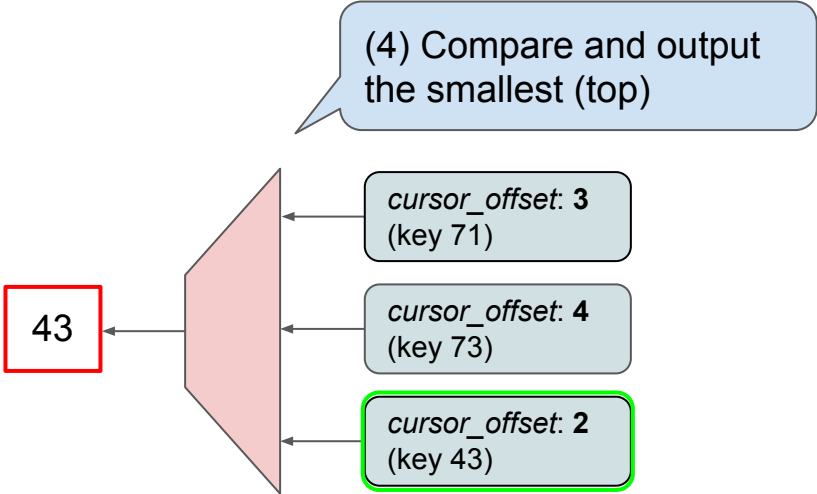
Range Query using a Min-Heap



Outputs: 31

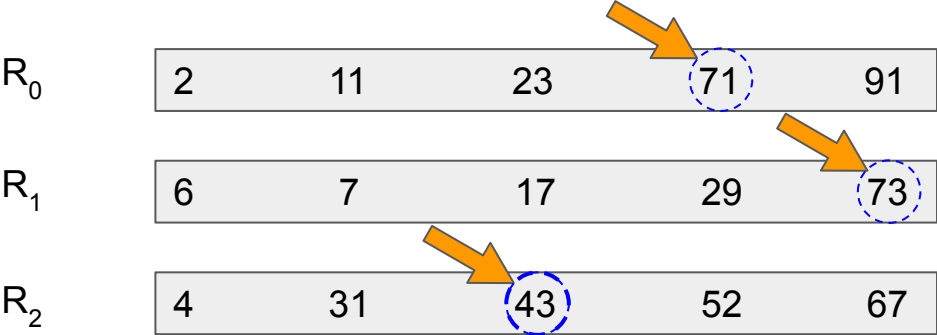


Range Query using a Min-Heap



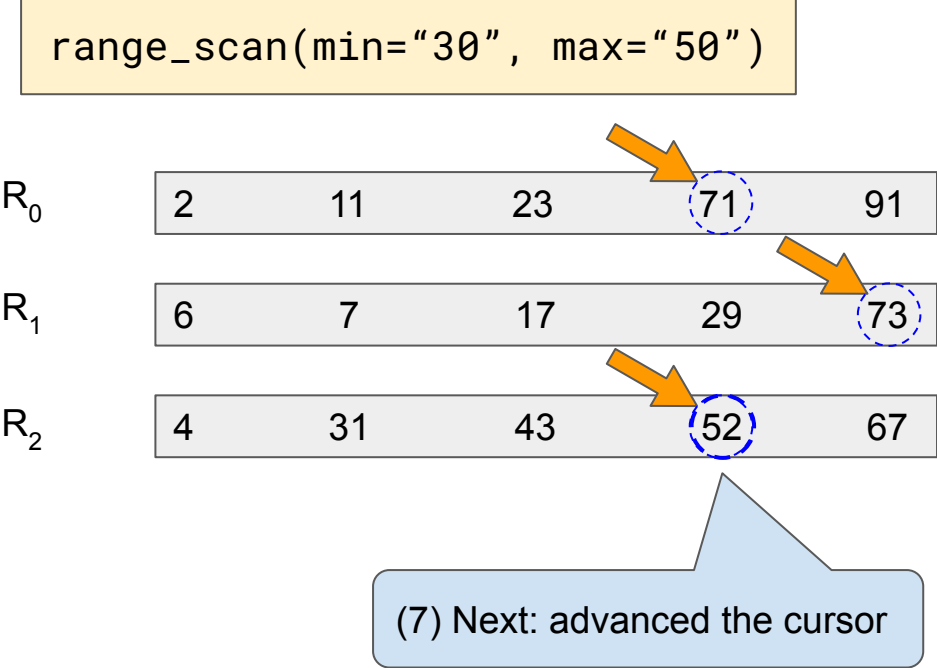
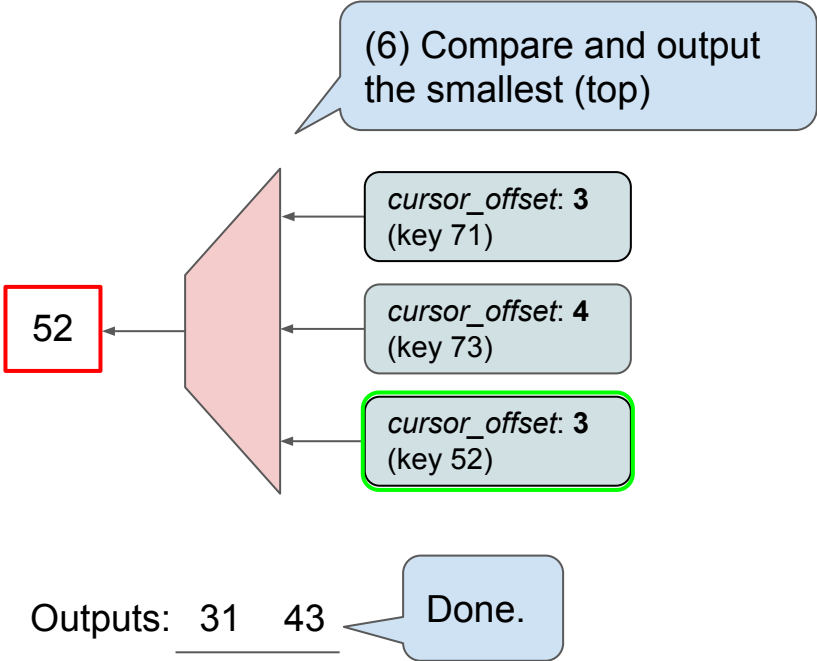
Outputs: 31 43

range_scan(min="30", max="50")



(5) Next: advanced the cursor

Range Query using a Min-Heap

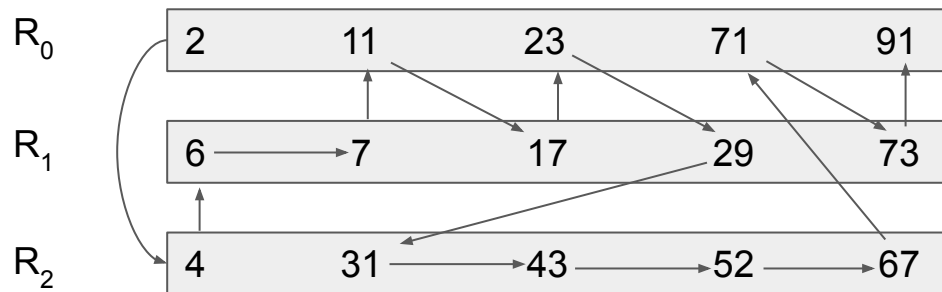


Break Down of the Range Query Cost

- Requires a **binary search on every run** to find the starting point
- Uses multiple **key comparisons** to advance the iterator
- Must access every run even if some do **NOT** cover keys in the range

R_0	<table><tr><td>2</td><td>11</td><td>23</td><td>71</td><td>91</td></tr></table>	2	11	23	71	91
2	11	23	71	91		
R_1	<table><tr><td>6</td><td>7</td><td>17</td><td>29</td><td>73</td></tr></table>	6	7	17	29	73
6	7	17	29	73		
R_2	<table><tr><td>4</td><td>31</td><td>→ 43</td><td>→ 52</td><td>→ 67</td></tr></table>	4	31	→ 43	→ 52	→ 67
4	31	→ 43	→ 52	→ 67		

Observation #1: A Stable Sorted View



Sorted view: 2 4 6 7 11 17 23 29 31 43 52 67 71 73 91

Runs: R₀ R₂ R₁ R₁ R₀ R₁ R₀ R₁ R₂ R₂ R₂ R₂ R₀ R₁
R₀

The sorted view stays the same as long as the three runs stay unchanged.

Observation #2: Hardware Trend

- Disk

- Poor random access performance
- I/O dominates the query cost

- SSD

- Better random access performance
- CPU is becoming the bottleneck (NVM, Optane)

Observations:

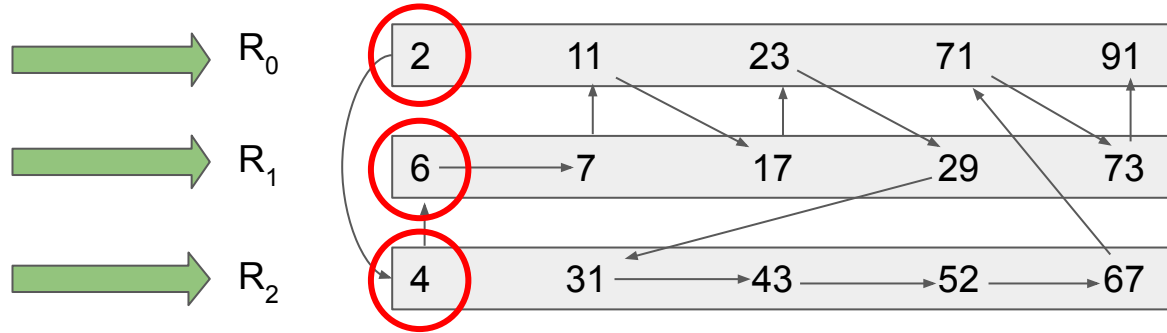
- Merging iterator builds **temporary** sorted view, then **discarded** afterwards
- Modern SSDs have better random read performance

Our Solution:

- ~~Discard Record~~ and **reuse** sorted views
- Sort-merge the runs ~~physically~~ **virtually**

REMIX: **R**ange-query-**E**fficient **M**ulti-table **I**nde**X**.

REMUX Data Structure



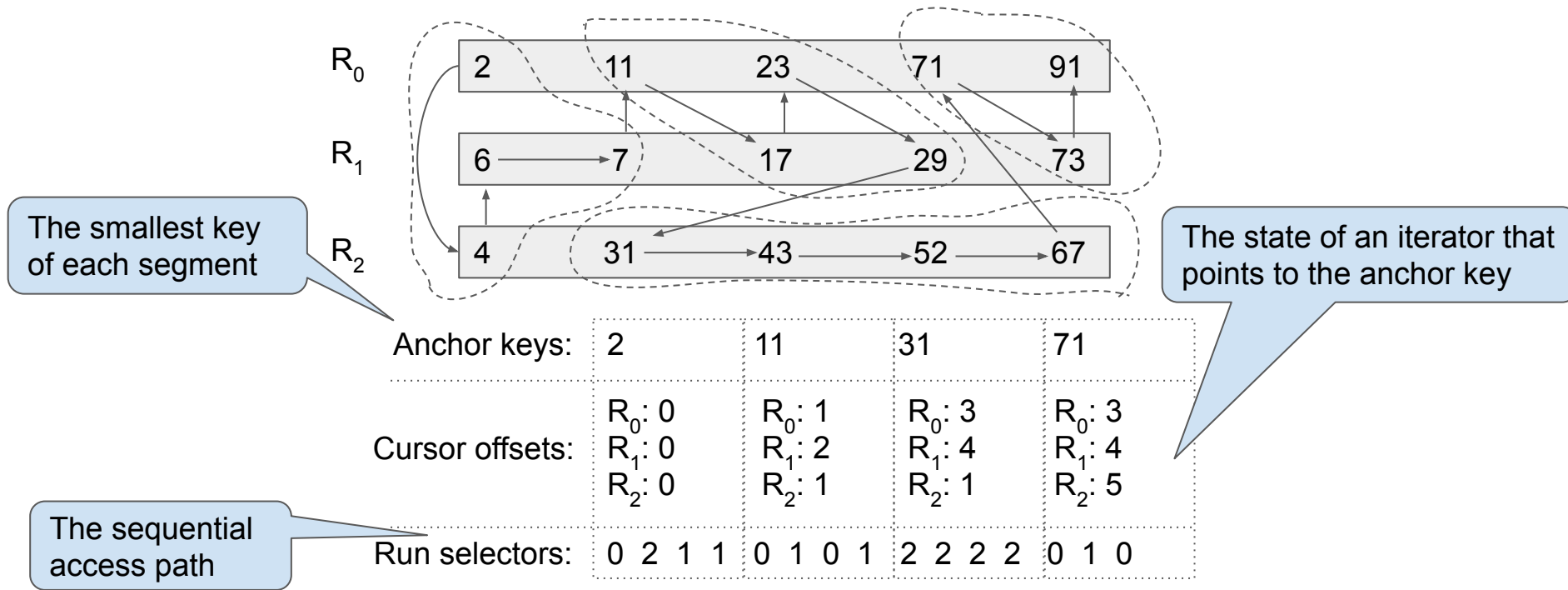
The sequential
access path

Run selectors:

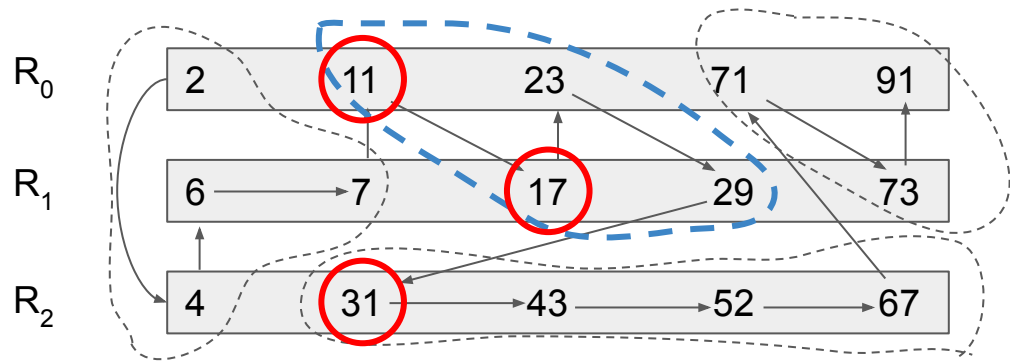
0 2 1 1 0 1 0 1 2 2 2 2 0 1 0



REMIX Data Structure

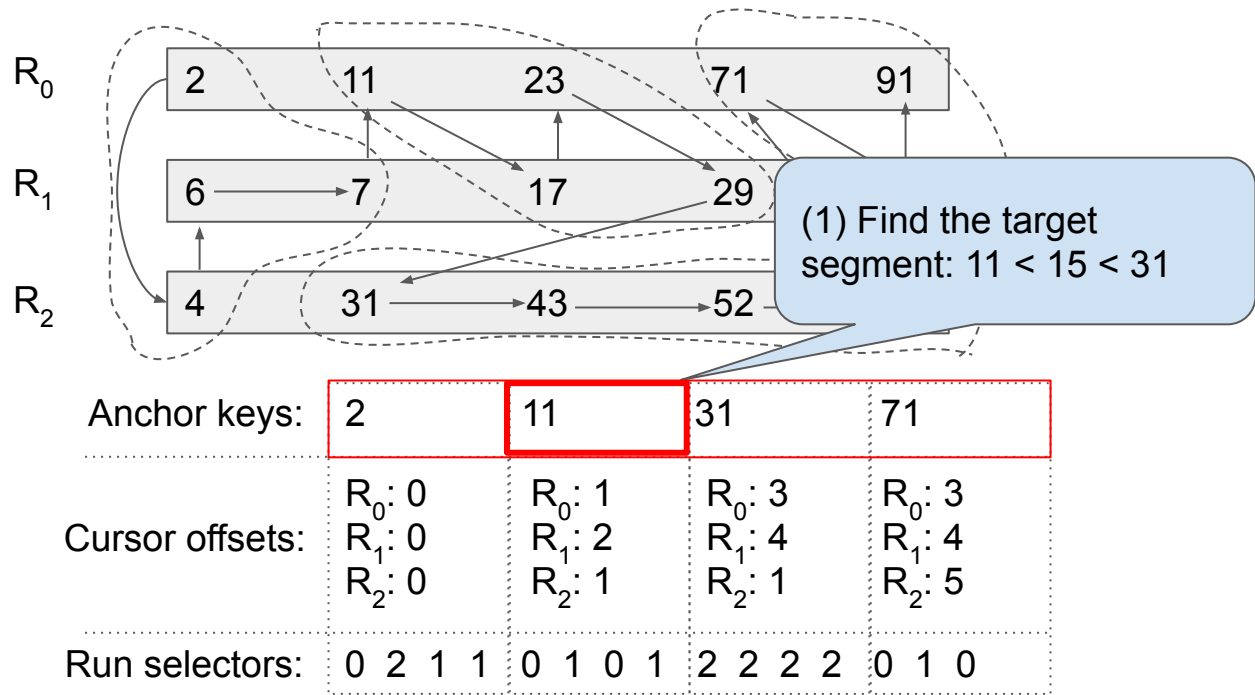


REMIX Data Structure



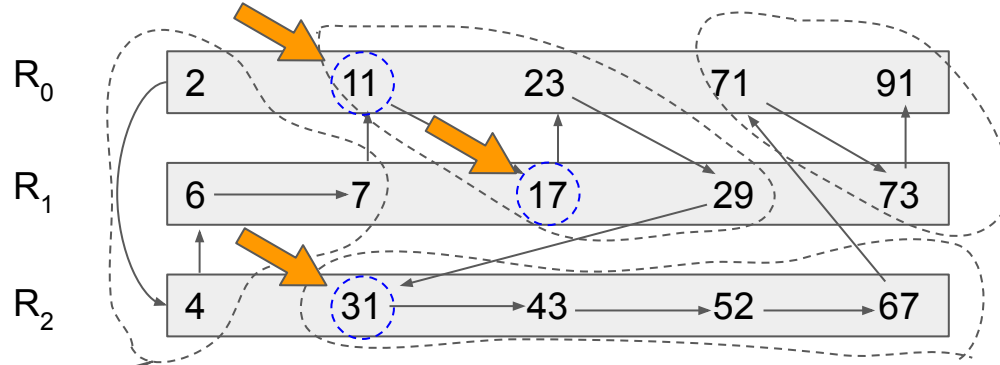
Anchor keys:	2	11	31	71
Cursor offsets:	$R_0: 0$	$R_0: 1$	$R_0: 3$	$R_0: 3$
	$R_1: 0$	$R_1: 2$	$R_1: 4$	$R_1: 4$
	$R_2: 0$	$R_2: 1$	$R_2: 1$	$R_2: 5$
Run selectors:	0 2 1 1	0 1 0 1	2 2 2 2	0 1 0

Example: Scan(min="15", max="28")



Outputs: _____

Example: Scan(min="15", max="28")

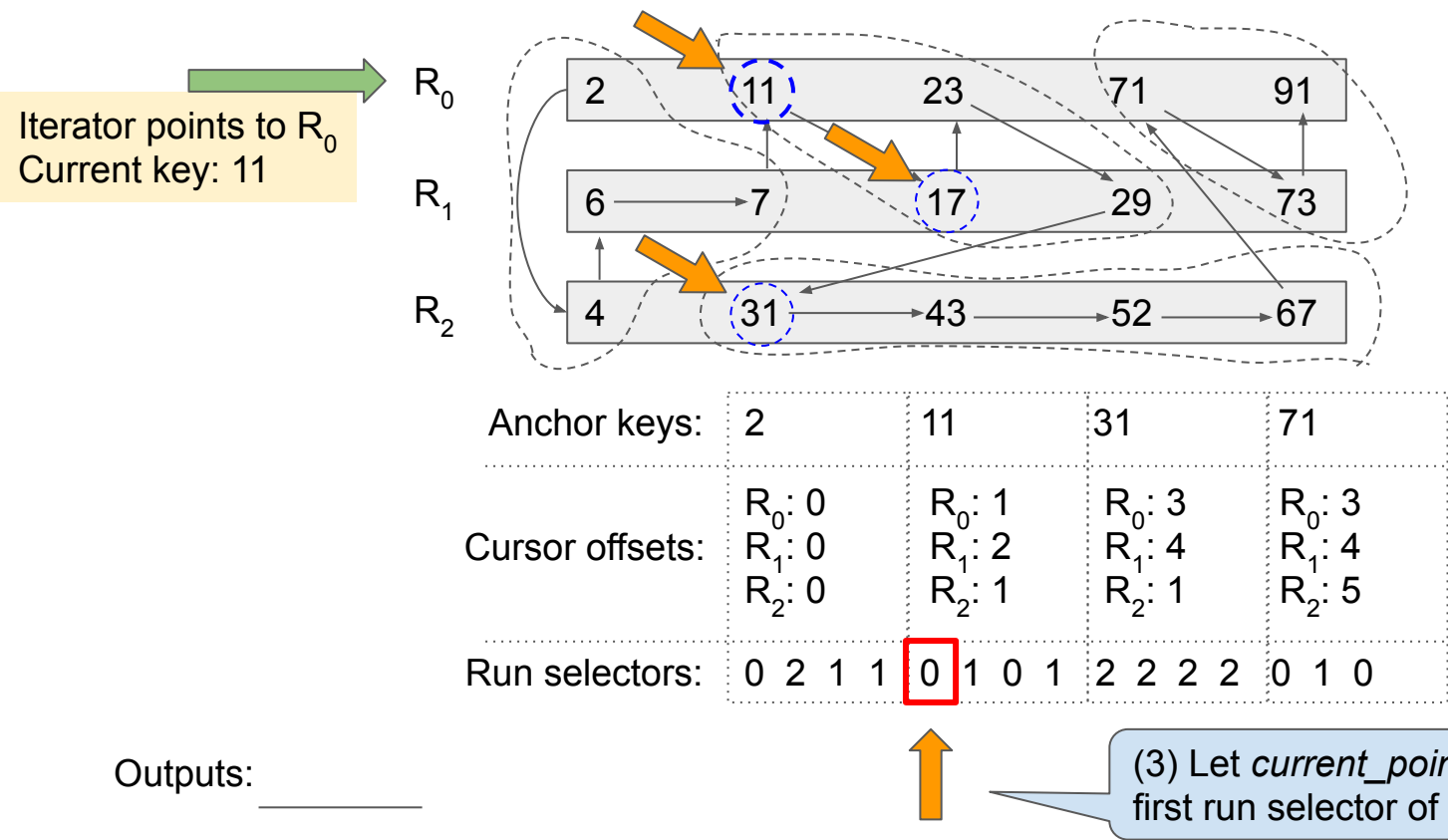


(2) Initialize the *cursors*

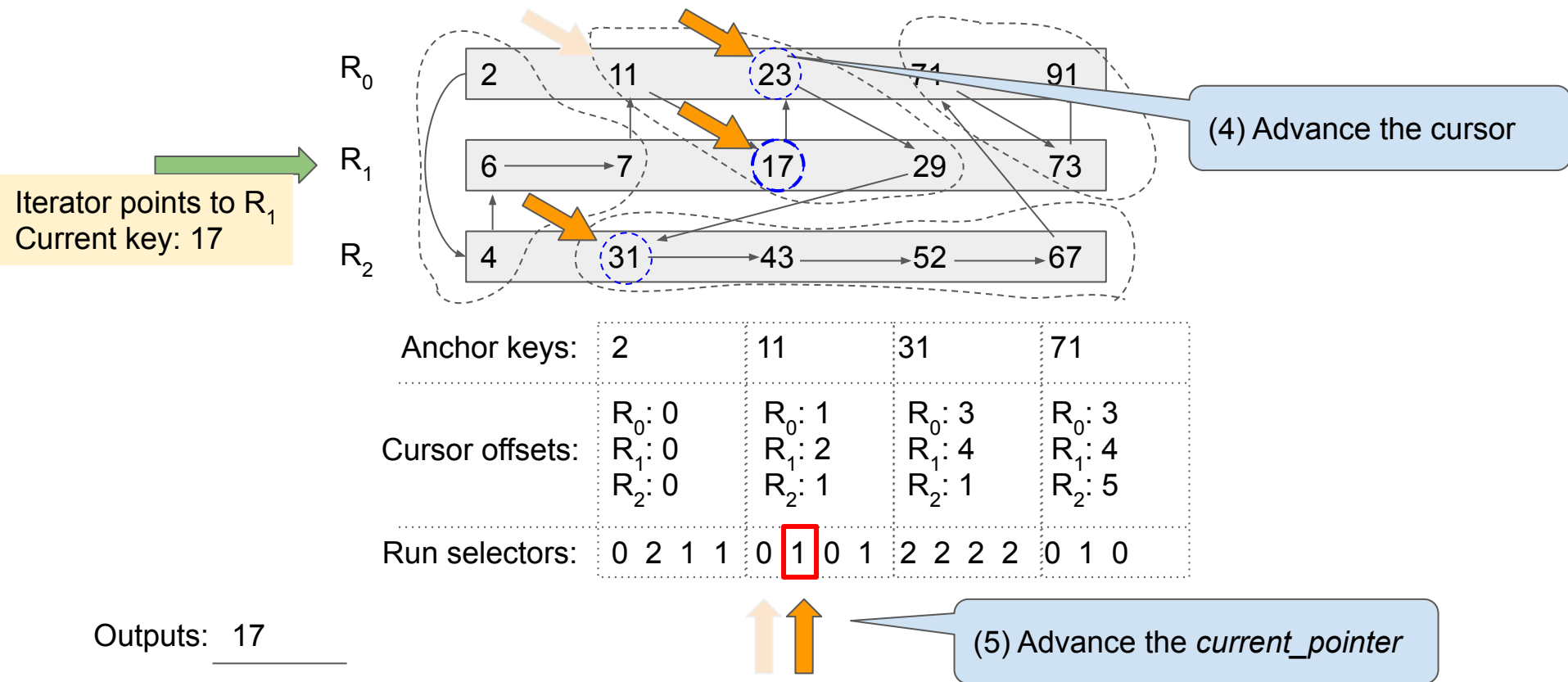
Anchor keys:	2	11	31	71
Cursor offsets:	R_0 : 0	R_0 : 1	R_0 : 3	R_0 : 3
	R_1 : 0	R_1 : 2	R_1 : 4	R_1 : 4
	R_2 : 0	R_2 : 1	R_2 : 1	R_2 : 5
Run selectors:	0 2 1 1	0 1 0 1	2 2 2 2	0 1 0

Outputs: _____

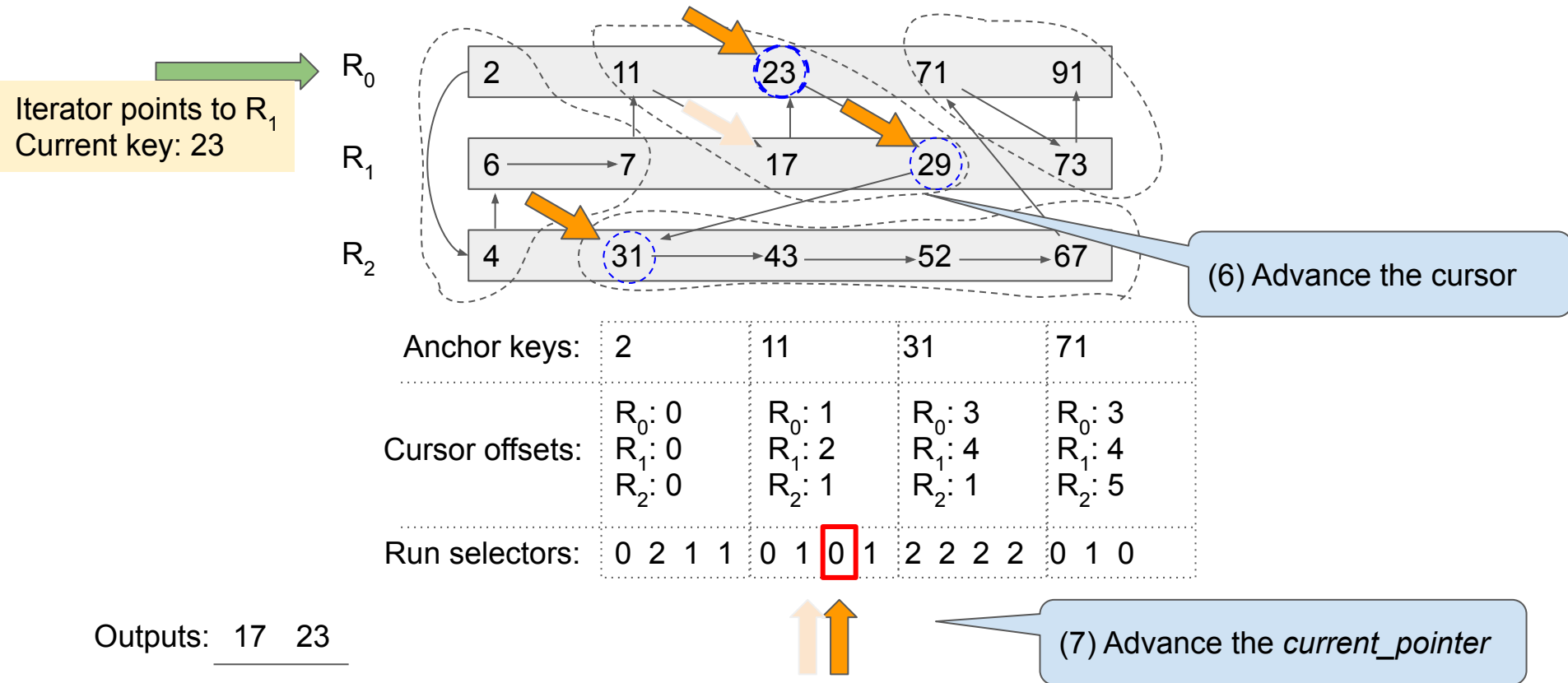
Example: Scan(min="15", max="28")



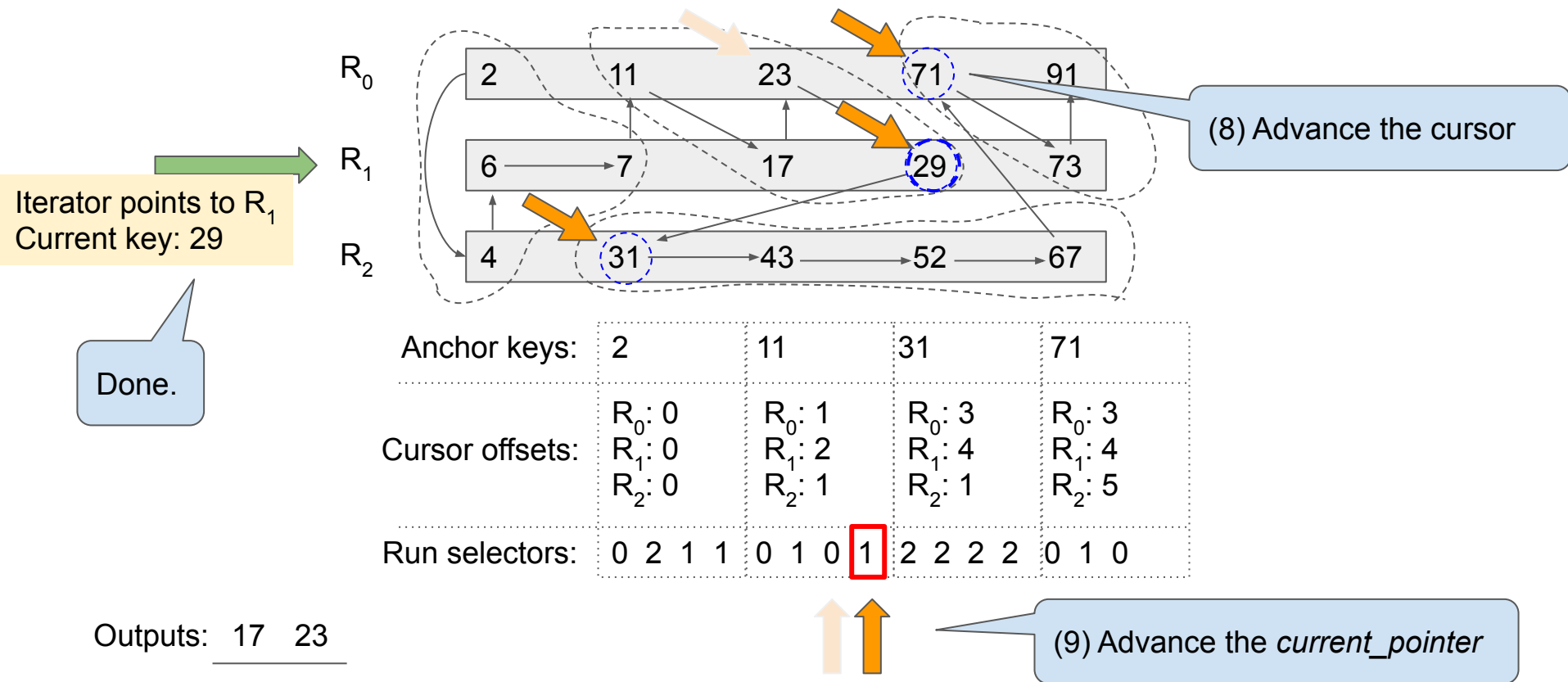
Example: Scan(min="15", max="28")



Example: Scan(min="15", max="28")



Example: Scan(min="15", max="28")



REMIX Summary

- Initialize an iterator with **one** binary search
- Advance an iterator **without** key comparisons
- **Skip** runs that are **NOT** on the search path
- Support binary search in a segment *
- Space efficient *

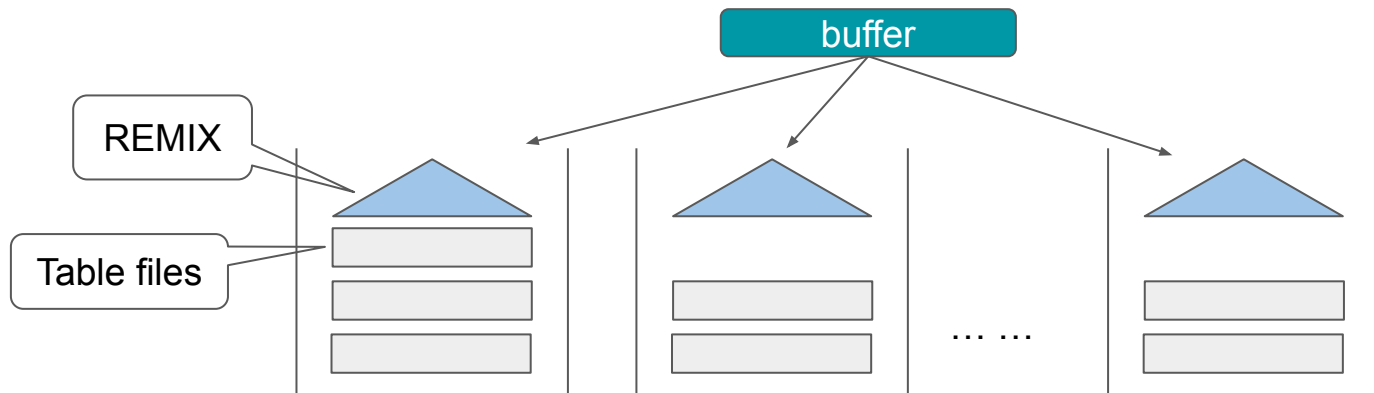
* See details in the paper

RemixDB

- Single-level partitioned layout
 - Efficient under real-world workloads
- Tiered compaction
- REMIX-indexed table files

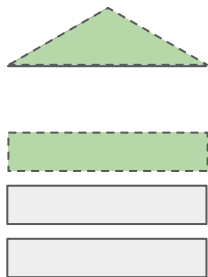
✓ write

✓ read

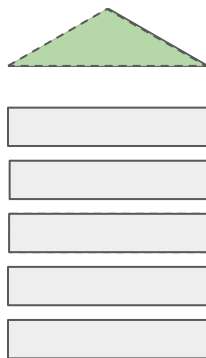


Compactions

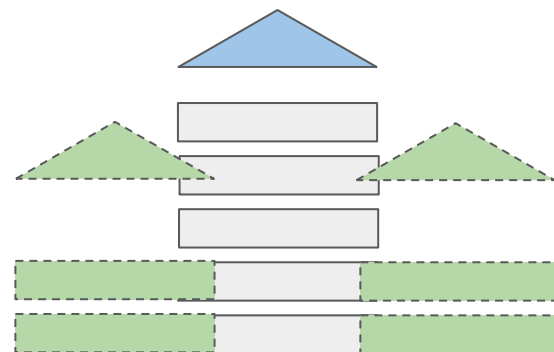
Minor Compactions



Major Compactions



Split Compactions



Evaluation

- Microbenchmarks: REMIX range query performance
- YCSB: RemixDB performance
- Setup
 - Intel Xeon 4210 CPU
 - 64G memory
 - Intel 905P Optane PCIe SSD (960GB)

Microbenchmark

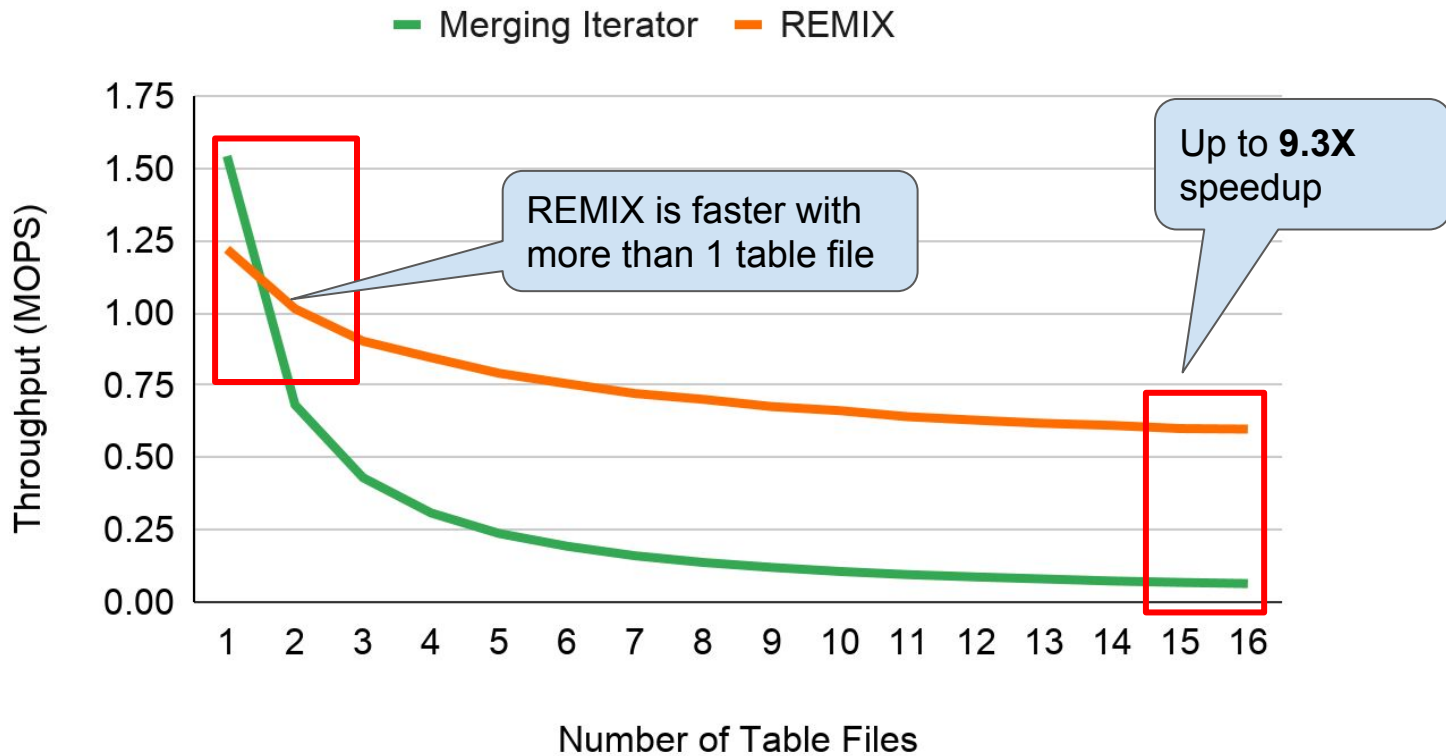
REMIX vs. Merging Iterator

- REMIX-Indexed table files
- Regular SSTables with merging iterator (min-heap)

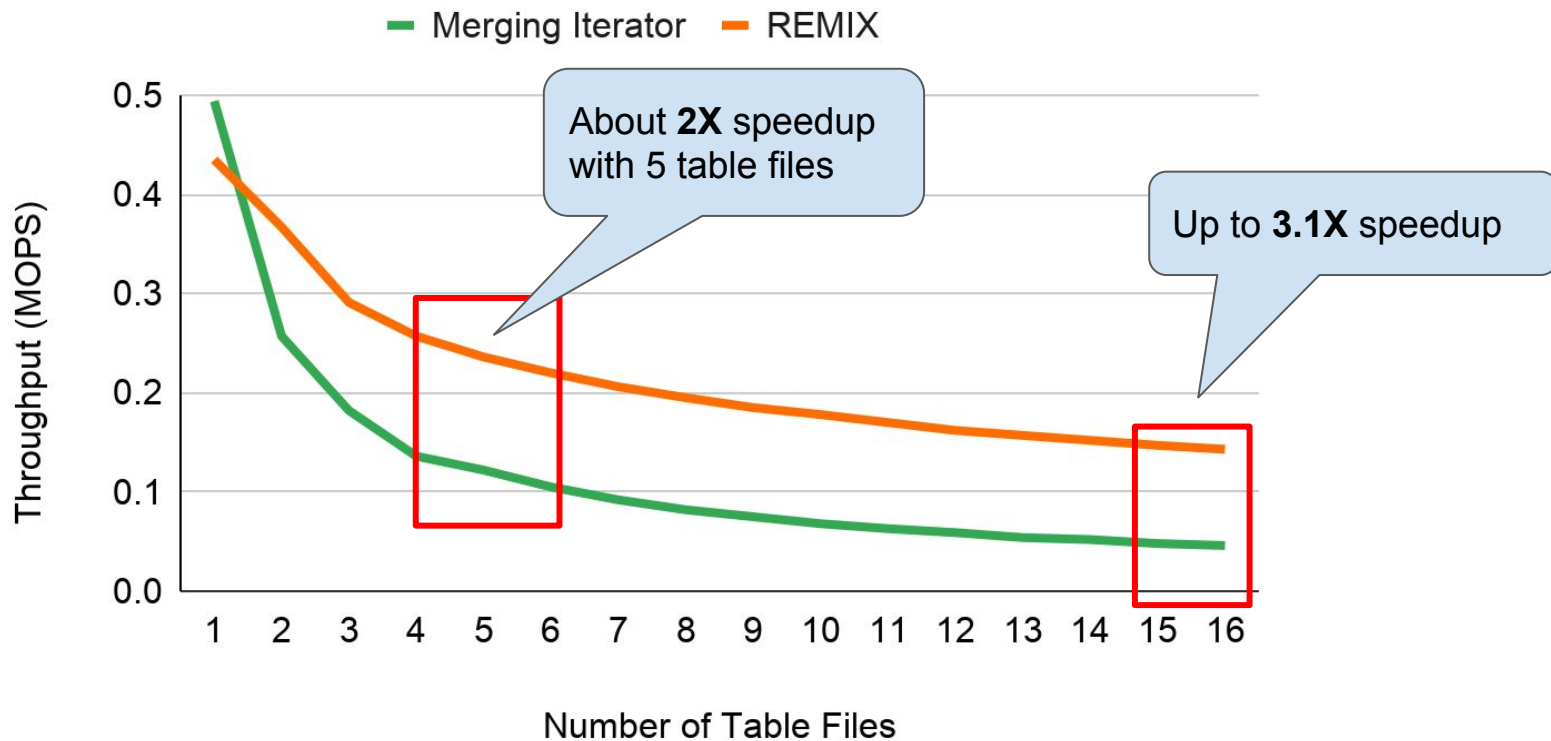
Configuration

- Segment size: 32
- Each table file contains 64 MB KV-pairs
- Key size: 16 B
- Value size: 100 B
- KVs are randomly assigned to different tables

Microbenchmark: Seek



Microbenchmark: Seek and Scan 50 Keys



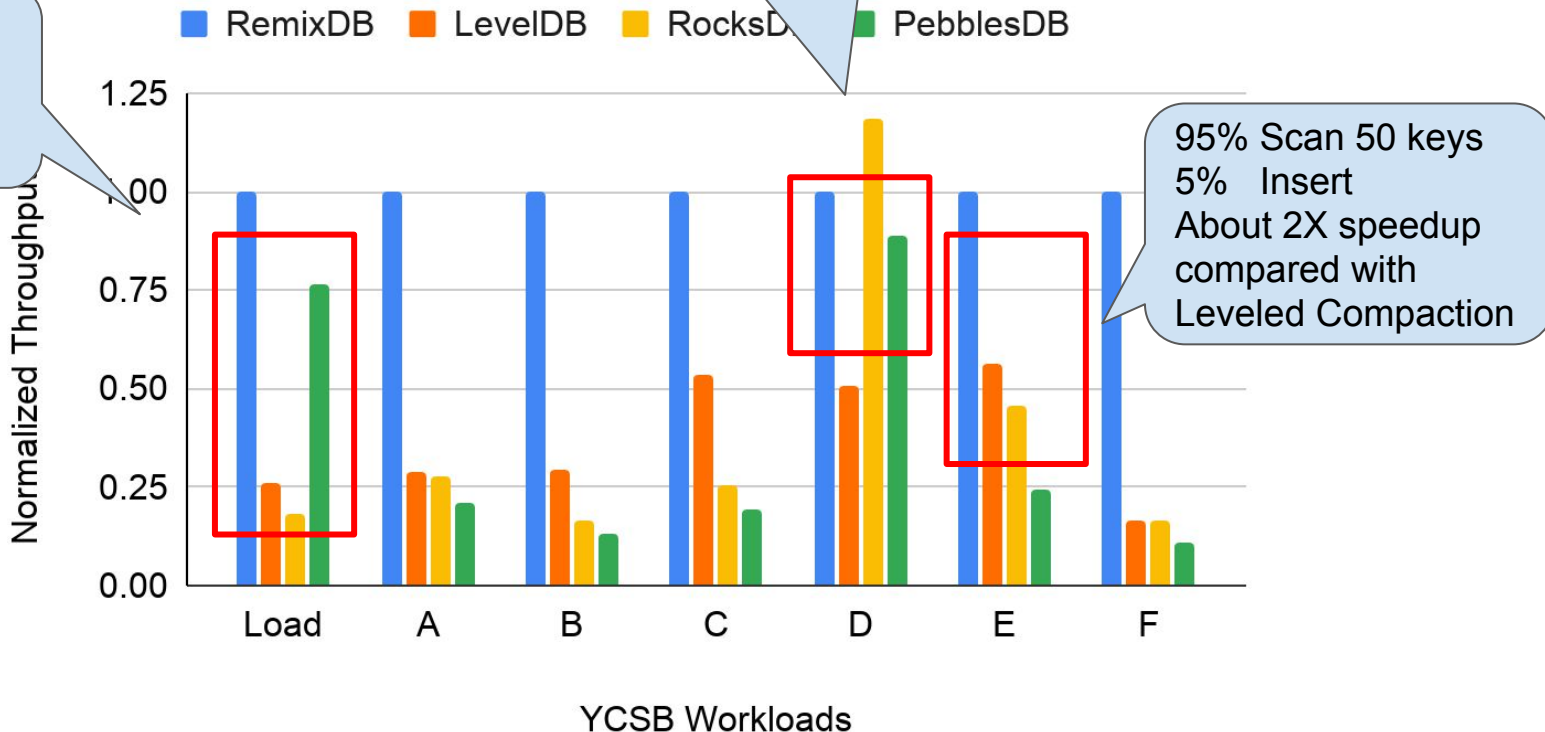
YCSB

- Dataset
 - Key size: 16 B
 - Value size: 120 B
 - 256GB Stores
- Baseline systems: LevelDB, RocksDB, PebblesDB
- Configuration
 - 4 worker threads
 - 4GB user-space block cache
- Load a store in randomly shuffled order
- YCSB workloads A-F

RemixDB---YCSB

30% faster than
write-optimized
PebblesDB

95% Read recent updates
5% Sequential insert
Mostly served in memory



Conclusion

- REMIX
 - Record and reuse sorted view
 - Fast range query on multiple sorted runs
- RemixDB
 - Single-level partitioned layout
 - Efficient read and write

Thank you!

