

Basic row-based storage									
	8	32	4	2	4	= 50			
	ld BIGINT	Name CHAR(32)	Age INT	Gender SMALLINT	Birthday DATE				
						2			

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	8	32	4	2	4	= 50				
0	ld BIGINT	Name CHAR(32)	Age INT	Gender SMALLINT	Birthday DATE					
50	ld BIGINT	Name CHAR(32)	Age INT	Gender SMALLINT	Birthday DATE					
100	ld BIGINT	Name CHAR(32)	Age INT	Gender SMALLINT	Birthday DATE					
150	ld BIGINT	Name CHAR(32)	Age INT	Gender SMALLINT	Birthday DATE	-				
						3				

How to efficiently find data?

- Types of queries
 - Exact match: id = 139856151
 - Predicate scans: All entries with age > 80
- Option 1: Full scan
- Option 2: Use an index!

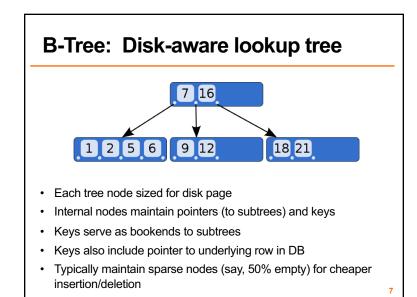
4

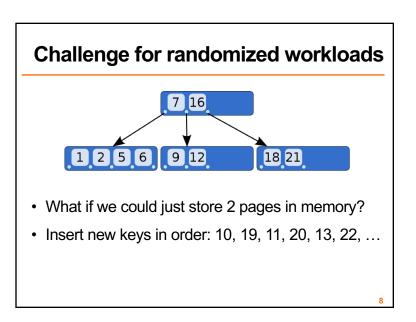
Use a balanced binary tree?

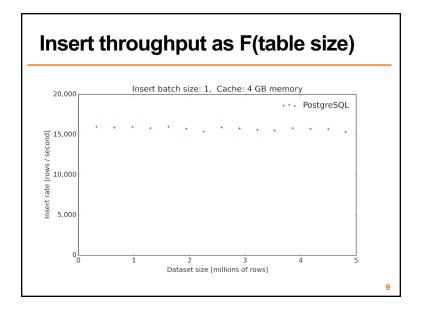
- · Easy to do in memory
- Except how do you store on disk?
 - Let's say 1M entries
 - 20 levels of a binary tree
 - Don't want 20 random disk seeks (say, ~200ms)

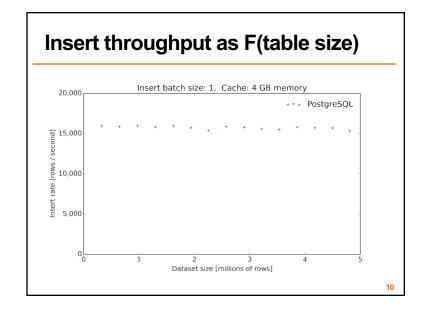
Strawman: Leverage locality on disk

- Can collocate many pointers on disk
 - Databases typically read data in "page" granularities (4-16KB per page)
- But keeping a binary tree "balanced" requires many rotations (e.g., red-black tree rotations)









LSM Trees + LevelDB

- Goal #1: High-throughput writes (inserts/updates/deletes)
- Goal #2: Support index size >> memory size
- Observation: Writes to disk should be batched for high throughput (either SSD or HDD)
- · Observation: Sorting/indexing in memory is fast
- Main insight: Don't try to maintain single data structure with in-place ordering

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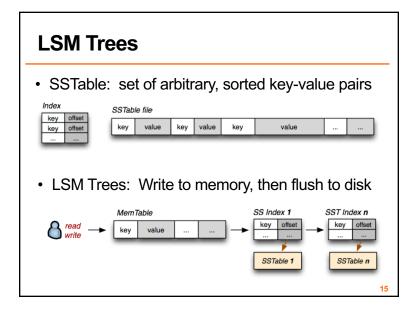
Collect writes and batch in memory

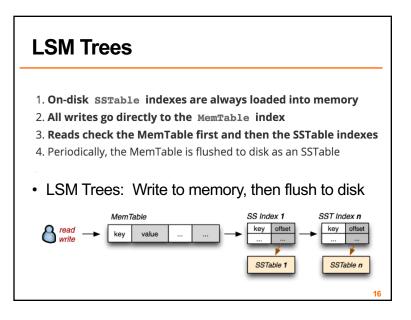
- Collect writes in memory
 - Can maintain sorted list in memory
 - Can update in-place (overwrite, delete, etc.)
- Disk is immutable
 - Once written to disk, not modified in-place
 - Queries will need to find all records and merge
 - Deletes are simply "tombstone records"

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Spill From Memory To Disk

- As memory budget fills up, spill them to disk
 - Write out entire sorted string table
 - Write out a subtree, then remove and prune it in memory
- Each dump forms a "run" ordered by write time





Problem & solution

- Index lookups can traverse many SSIndexes
 - Esp. with range scan vs. exact-match lookup
 - Other optimizations trade-off memory for additional disk lookups, e.g., bloom filter vs. SSIndex
- Idea
 - Merge and compact SSTables in background

Log-structured MERGE Tree

- Merge updates disk data structures in background
 - Input: K overlapping sorted SSTables from A-Z at level L
 - Output: Disjoint sorted SSTables files at level L+1
- Compaction also occurs as part of the merges
 - Merges multiple updates into one
 - Deletes tombstoned records
 - Recovers storage from merged updates and deleted values
- Can occur very efficiently by sequential reads from each SSTable and writing to output files (why? Hint: all sorted))

