The HV-tree: a Memory Hierarchy Aware Version Index

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What is Versioned data

- Data describing objects that
 - has some attributes that change over time
 - we want to keep track of ALL the changes
 - (implicitly) has an attribute that never changes: identifier

• E.g.

- Bank account: account number, balance
- Stock: stock number, price
- Wikipedia entry: name, contents
- Sales record: item id, sales of a certain day
- Star positions: star name, location
- Every changed value is called a "version" and has a "version number"
- "Versioned data" also called "temporal data", and "version number" becomes "timestamp"



What to do on Versioned Data

- Point query (query with single key and single time)
 - What was the position of star #1234 on 1 Jan 2010?
- (Key-slice) Time-range query
 - Show the trajectory of star #1234 between 1 Jan 2010 and today.
- (Time-slice) Key-range query
 - Show the positions of all the stars on 1 Jan 2010
- Time-range key-range query (relatively rare)
 - Show the trajectories of the stars #1000~2000 between 1 Jan 2010 and today

Why do we Care about it Now

- Very large versioned databases
 - Sales record: Wal-Mart's data warehouse was 70 TB in 2001
 - Star positions: The Sloan Digital Sky Survey project receives 70 gigabytes of images every night
 - Existing version indexes do not scale well
- Improvement on Hardware
 - CPU/cache speed doubles every two years
 - Memory size increases at a similar rate
 - Existing version indexes do not take advantage of main memory techniques





Outline



- Existing Work
 - Version indexes, especially TSB-tree
 - Main memory techniques
- Design for multiple levels of memory hierarchy
 - Principles
 - Straightforward approaches
- HV-tree
- Experimental results

Existing Version Indexes

- Ones that move new data to new nodes
 - Write-once B-tree
 - Multi-version B-tree
- One that moves old data to new nodes
 - Time Split B-tree (TSB-tree)
 - Unique feature of progressively migrate old data to a new medium – a larger medium like hard disk or tape
 - Leaving current data on high-speed medium like the main memory



Time Split B-tree (TSB-tree)



key, time, data



Time Split B-tree (TSB-tree)





time 4



- Time split or key split depends on the portion of current entries in the node β : key split if β is greater than a threshold T
- Search: follow key-time range

Main Memory Indexing Techniques

- Key techniques
 - Aligning node size with block size
 - CSS-tree, CSB+-tree: use the cache block size (typically 32B or 64B) as tree node size
 - Later study shows that the actual optimal node size for the CSB+-tree is much larger than the cache block size
 - We assume the optimal node size S_{cache} is known
 - Pointer elimination
 - Hard to apply to a complicated structure like TSB-tree

Design for Multiple Levels of Memory Hierarchy



- Facts: big latency difference between adjacent levels of memories in the hierarchy, usually 1000 times.
- Principles
 - Tailor the index's structure to suits the characteristics of each level of the memory hierarchy.
 - Keep as much as possible frequently accessed data in higher levels of the memory hierarchy.

Straightforward Adaptions of TSB-tree



TSB-small

- use S_{cache} as the node size
- let the operation system deal with caching and paging
- Worse than TSB-tree because of bad paging behavior

• TSB-cond

- use S_{cache} as the node size initially
- expand/condense to node of size S_{djsk} as historical pages are created and moved to disk
- Worse than TSB-tree because of overhead caused by condensation

The HV-tree: memory <u>Hierarchy aware Version tree</u>



- Node size adjustable to the level of the memory the node resides in (Principle 1)
 - Key: Gradual change of size
- Delayed data migration (Principle 2)

The HV-tree Structure

- Allowable node sizes
 - S_{cache} , $2S_{cache}$, ..., S_{disk} (S_{disk} is a power of 2 times S_{cache})
 - E.g. $S_{cache} = 1K$, $S_{cache} = 4K$, so allowable node sizes are: 1K, 2K, 4K.
- Some additional pointers maintained for data migration



• Start with the no Start Node full? smallest allowable yes node size Calculate: $\beta = R_c \div F_{node}(size)$ $\beta < T$ $\beta \ge T$ Compare: β, Τ • When node is full Compare: key split S_{large}, size size < Slarge Key split time split size \geq Slarge or node expansion Time split Expand • Choice of T Contract current node S_{cache} / S_{disk} Out of main no

memory?

Do Migration

yes

End

Add historical node

to migration chain

The HV-tree Insertion

HV-tree Data Migration

- Upon creation of a historical node
 - Do not move to disk immediately
 - Added to a *Migration Chain*
 - Migrate when out of memory





Experimental Setup



- Generated datasets with updates, search and mixed workloads
- Sizes: 500MB, 1000MB
- Queries follow Zipfian distribution, with varying skewness
- Hardware:
 - 3GHz CPU, 1GB memory, 80GB disk
 - L1 cache: <8K, 64B, 1>, L2 cache: <512KB, 64B, 8>

•
$$S_{cache} = 1K, S_{cache} = 4K$$

Results: Updates and Point Queries





Results: Key-Range Queries, Time-Range Queries







Experiments

• Finding S_{cache}



• Validation of T





Conclusions and Future Work

- First index design optimizing performance for multiple levels of memory hierarchy, achieving a highly scalable and efficient version index.
- Key techniques
 - Difference node sizes
 - Gradual change of node sizes
 - Data migration chain
- Performance
 - Several times faster for updates and point queries
 - 1000 times faster for key/time range queries
- Future work
 - Other data structures
 - Multi-core machine