

# CS 764: Topics in Database Management Systems Lecture 5: Modern Buffer Management

Xiangyao Yu 9/21/2021

### Today's Paper: LeanStore

#### LeanStore: In-Memory Data Management **Beyond Main Memory**

Viktor Leis, Michael Haubenschild\*, Alfons Kemper, Thomas Neumann

Technische Universität München {leis, kemper, neumann}@in.tum.de

Tableau Software\* mhaubenschild@tableau.com\*

Abstract-Disk-based database systems use buffer managers in order to transparently manage data sets larger than main memory. This traditional approach is effective at minimizing the number of I/O operations, but is also the major source of overhead in comparison with in-memory systems. To avoid this overhead, in-memory database systems therefore abandon buffer management altogether, which makes handling data sets larger than main memory very difficult.

In this work, we revisit this fundamental dichotomy and design a novel storage manager that is optimized for modern hardware. Our evaluation, which is based on TPC-C and micro benchmarks, shows that our approach has little overhead in comparison with a pure in-memory system when all data resides in main memory. At the same time, like a traditional buffer manager, weakness: They are not capable of maintaining a replacement it is fully transparent and can manage very large data sets effectively. Furthermore, due to low-overhead synchronization, our implementation is also highly scalable on multi-core CPUs.

#### I. INTRODUCTION

for database systems. Traditional systems cache pages using accesses and transparently loads/evicts pages from/to disk. By and transparently.

lookup in order to translate a logical page identifier into an to design an efficient buffer manager for modern hardware? in-memory pointer. Even worse, in typical implementations the data structures involved are synchronized using multiple designing, implementing, and evaluating a highly efficient latches, which does not scale on modern multi-core CPUs. As storage engine called LeanStore. Our design provides an Fig. 1 shows, traditional buffer manager implementations like abstraction of similar functionality as a traditional buffer BerkeleyDB or WiredTiger therefore only achieve a fraction manager, but without incurring its overhead. As Fig. 1 shows, of the TPC-C performance of an in-memory B-tree.

Hekaton [3], HANA [4], HyPer [5], or Silo [6] eschew buffer is that accessing an in-memory page merely involves a simple, management altogether. Relations as well as indexes are directly well-predicted if statement rather than a costly hash table stored in main memory and virtual memory pointers are used lookup. We also achieve excellent scalability on modern multiinstead of page identifiers. This approach is certainly efficient. core CPUs by avoiding fine-grained latching on the hot path. However, as data sizes grow, asking users to buy more RAM Overall, if the working set fits into RAM, our design achieves or throw away data is not a viable solution. Scaling-out an inmemory database can be an option, but has downsides including systems. At the same time, our buffer manager can transparently hardware and administration cost. For these reasons, at some manage very large data sets on background storage and, using point of any main-memory system's evolution, its designers modern SSDs, throughput degrades smoothly as the working have to implement support for very large data sets.

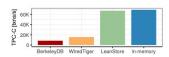


Fig. 1. Single-threaded in-memory TPC-C performance (100 warehouses)

Two representative proposals for efficiently managing largerthan-RAM data sets in main-memory systems are Anti-Caching [7] and Siberia [8], [9], [10]. In comparison with a traditional buffer manager, these approaches exhibit one major strategy over relational and index data. Either the indexes, which can constitute a significant fraction of the overall data size [11], must always reside in RAM, or they require a separate mechanism, which makes these techniques less general and Managing large data sets has always been the raison d'être less transparent than traditional buffer managers.

Another reason for reconsidering buffer managers are the a buffer manager, which has complete knowledge of all page increasingly common PCIe/M2-attached Solid State Drives (SSDs), which are block devices that require page-wise accesses. storing all data on fixed-size pages, arbitrary data structures, These devices can access multiple GB per second, as they including database tables and indexes, can be handled uniformly are not limited by the relatively slow SATA interface. While modern SSDs are still at least 10 times slower than DRAM in While this design succeeds in minimizing the number of I/O terms of bandwidth, they are also cheaper than DRAM by a operations, it incurs a large overhead for in-memory workloads, similar factor. Thus, for economic reasons [12] alone, buffer which are increasingly common. In the canonical buffer pool managers are becoming attractive again. Given the benefits of implementation [1], each page access requires a hash table buffer managers, there remains only one question: Is it possible

In this work, we answer this question affirmatively by LeanStore's performance is very close to that of an in-memory This is why main-memory database systems like H-Store [2], B-tree when executing TPC-C. The reason for this low overhead set starts to exceed main memory.

### Agenda

### **Main-memory DB**

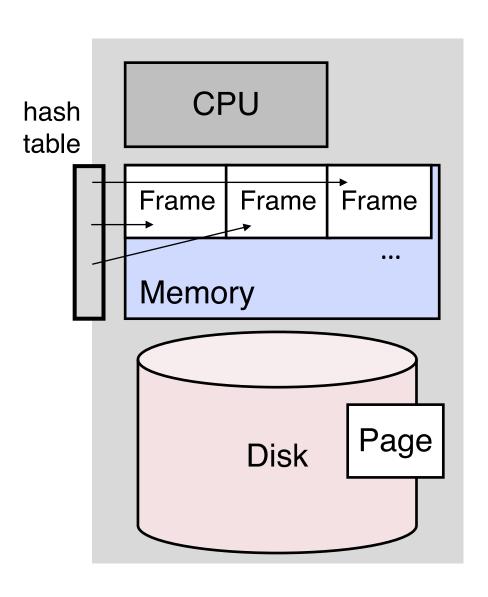
#### LeanStore design

- Pointer swizzling
- Page replacement
- Optimistic latching

#### **Experiments**

Fine-grained in-memory data management

### Conventional DB Architecture



Page granularity: Data managed in page granularity

Indirection: Use page ID to lookup hash table to locate a page

### Conventional DB Performance

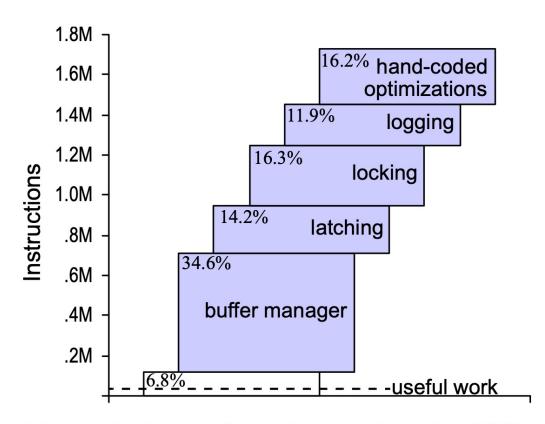
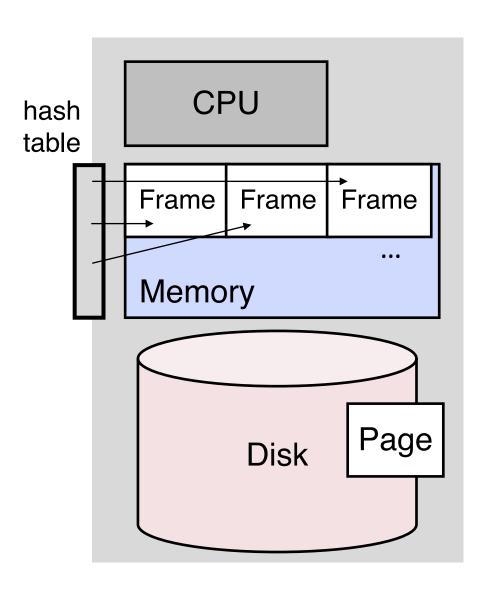


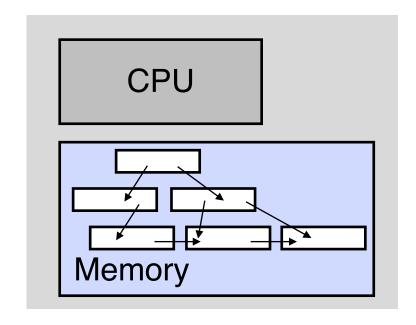
Figure 1. Breakdown of instruction count for various DBMS components for the New Order transaction from TPC-C. The top of the bar-graph is the original Shore performance with a main memory resident database and no thread contention. The bottom dashed line is the useful work, measured by executing the transaction on a no-overhead kernel.

Only a small fraction of instructions execute useful work

Significant instruction count dedicated to buffer management

### Main-Memory DB Architecture

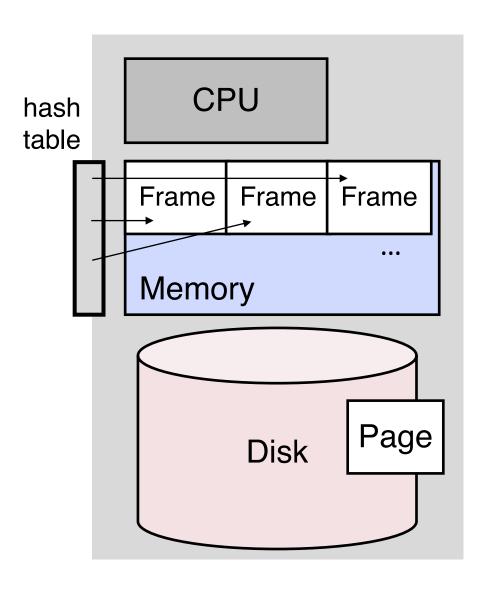


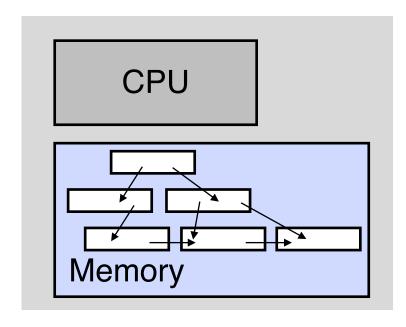


Fine-granularity: Fine-grained (e.g., tuple-level) data management

**No Indirection**: reference data following pointers

### Main-Memory DB Architecture





Fine-granularity: Fine-grained (e.g., tuple-level) data management

No Indirection: reference data following pointers

**⇒** Focus of this paper

### Agenda

#### Main-memory DB

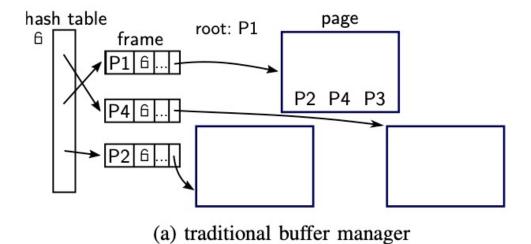
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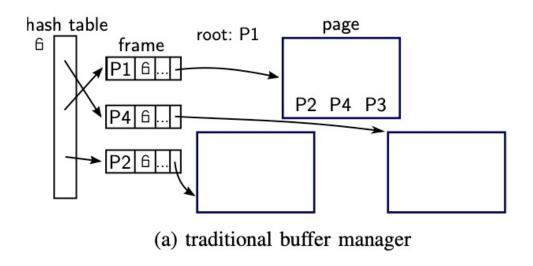
#### **Experiments**

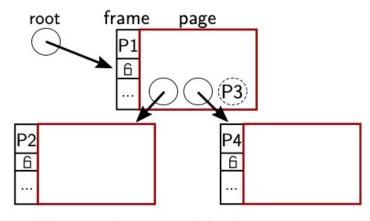
Fine-grained in-memory data management

# Pointer Swizzling



### Pointer Swizzling

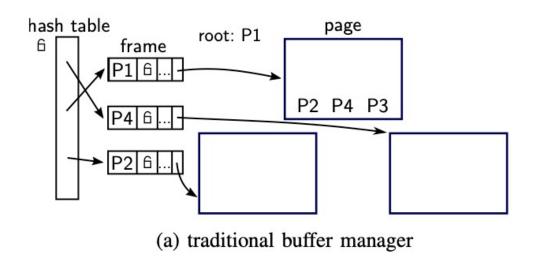


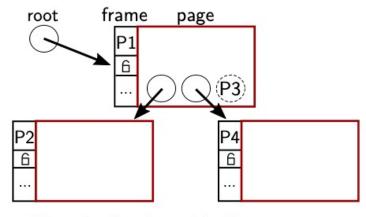


(b) swizzling-based buffer manager

Pages that reside in main memory are directly referenced using virtual memory addresses (i.e., pointers)

### Pointer Swizzling





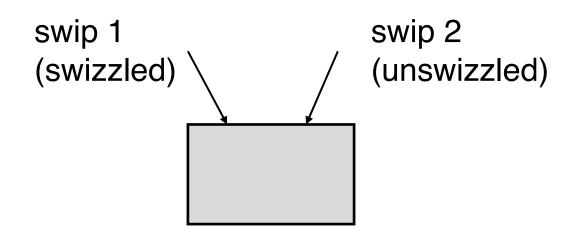
(b) swizzling-based buffer manager

Pages that reside in main memory are directly referenced using virtual memory addresses (i.e., pointers)

Swip: the 8-byte memory location referring to a page

Challenge 1: concurrency problem if a page is referrenced by multiple swips

 All references must be identified and changed atomically if the page is swizzled or unswizzled

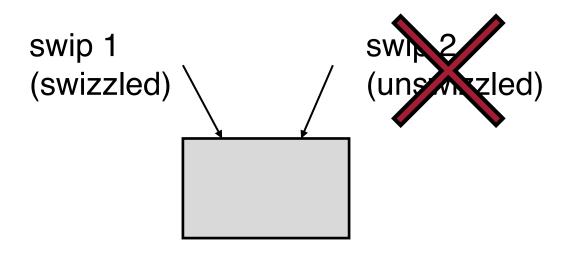


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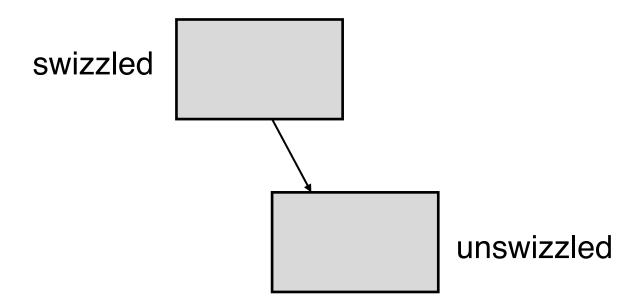
Solution: each page has a single owning swip

- In-memory data structures must be trees or forests



Challenge 2: pages containing memory pointers should not be written to disk

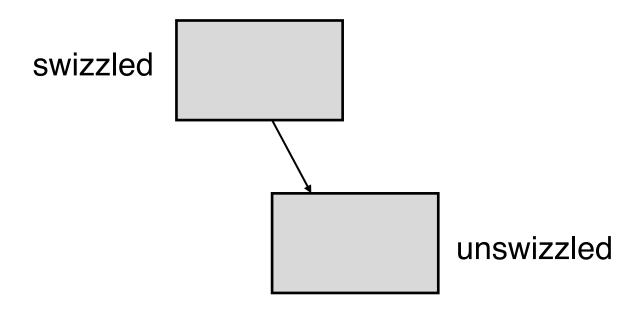
The pointers would not make sense if the system restarts



Challenge 2: pages containing memory pointers should not be written to disk

- The pointers would not make sense if the system restarts

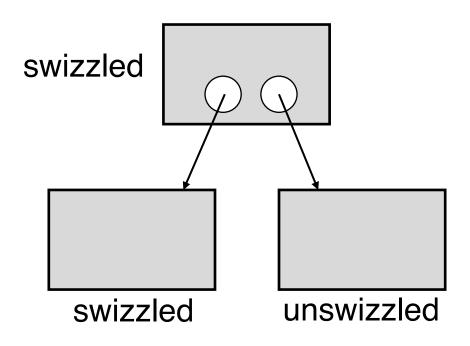
Solution: Never unswizzle a page that has swizzled children



Constraint 1: each page has a single owning swip

Constraint 2: Never unswizzle a page that has swizzled children

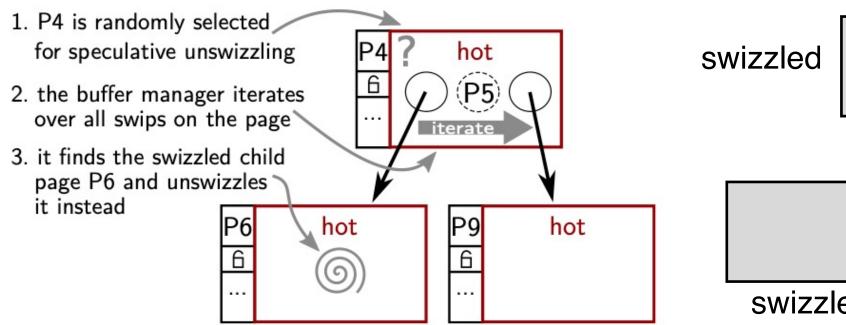
⇒Must be able to iterate over all swips on a page

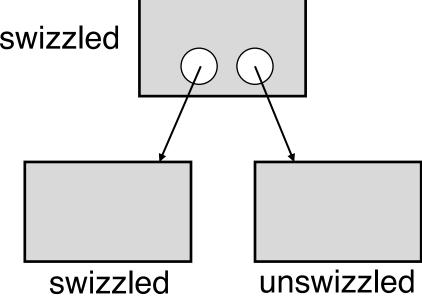


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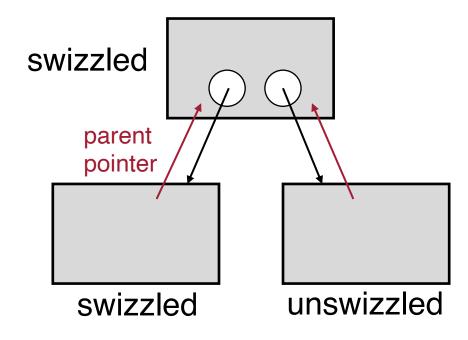




Constraint 1: each page has a single owning swip

Constraint 2: Never unswizzle a page that has swizzled children

- ⇒Must be able to iterate over all swips on a page
- ⇒Must be able to identify parent swip



Constraint 1: each page has a single owning swip

Constraint 2: Never unswizzle a page that has swizzled children

- ⇒Must be able to iterate over all swips on a page
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For example: B+-trees cannot have

link pointer

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LeanStore design

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- Optimistic latching

**Experiments** 

Fine-grained in-memory data management

Least Recent Used (LRU)

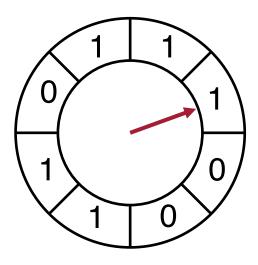
Clock replacement (aka second chance)

An approximation of LRU

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Least Recent Used (LRU)

Clock replacement (aka second chance)

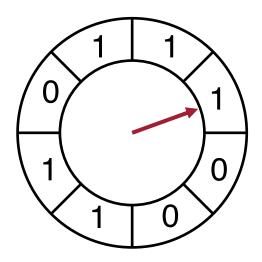
An approximation of LRU

Look for page to replace

If the bit = 0: evict

If the bit = 1: set to 0 and move to next entry

When a page is accessed, set bit to 1



Least Recent Used (LRU)

Clock replacement (aka second chance)

- An approximation of LRU

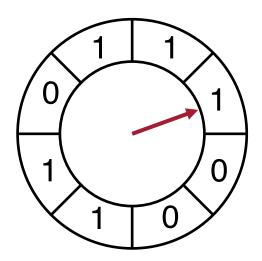
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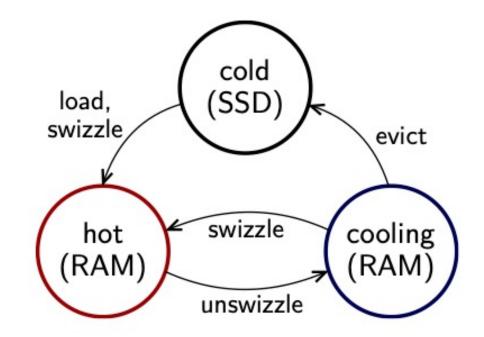
Updating tracking information for each page access is too expensive



### Page Replacement — Cooling

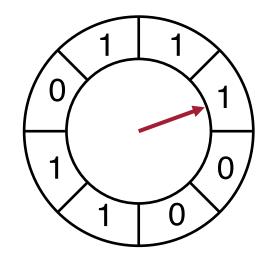
#### Randomly add pages to cooling stage

- Cooling pages are unswizzled but not replaced
- Cooling pages enter a FIFO queue; a page is replaced if it reaches the end of the queue
- Upon an access, a cooling page is swizzled



### Page Replacement Comparison

### Clock replacement



Look for page to replace

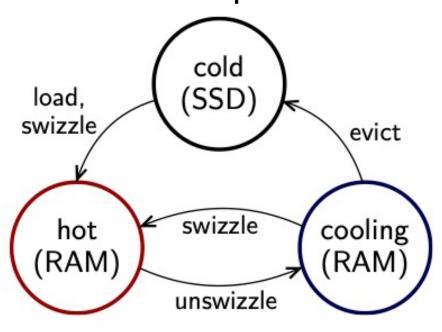
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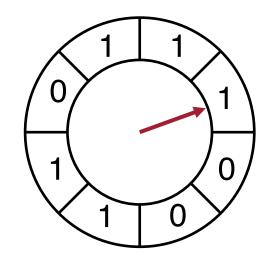
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#### LeanStore replacement



### Page Replacement Comparison

#### Clock replacement



Look for page to replace

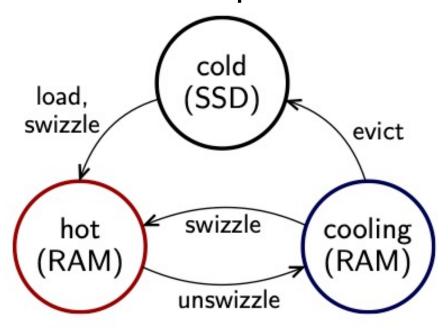
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When a page is accessed, set bit to 1

#### LeanStore replacement



#### **Discussion Question:**

Is clock replacement necessarily worse than cooling replacement?

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### Latching is Expensive

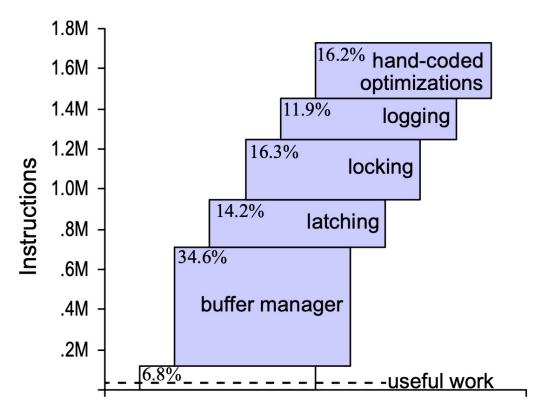


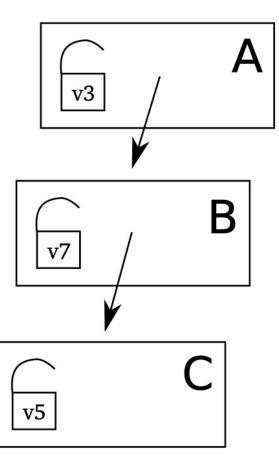
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### Lock Coupling

#### traditional

- 1. lock node A
- 2. access node A

- 3. lock node B
- 4. unlock node A
- 5. access node B
- 6. lock node C
- 7. unlock node B
- 8. access node C
- 9. unlock node C

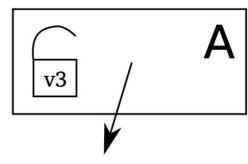


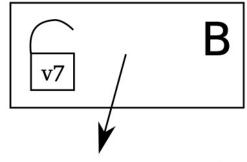
### Optimistic Lock Coupling

#### traditional

- 1. lock node A
- 2. access node A

- 3. lock node B
- 4. unlock node A
- 5. access node B
- 6. lock node C
- 7. unlock node B
- 8. access node C
- 9. unlock node C







#### optimistic

- 1. read version v3
- 2. access node A

- 3. read version v7
- 4. validate version v3
- 5. access node B
- 6. read version v5
- 7. validate version v7
- 8. access node C
- 9. validate version v5

### **Epoch-Based Reclamation**

Problem: reads do not block writes in optimistic locking

- A page is evicted or deleted while another thread is reading the page

### **Epoch-Based Reclamation**

Problem: reads do not block writes in optimistic locking

- A page is evicted or deleted while another thread is reading the page

Solution: Epoch-based reclamation

- Reclaim a page only if all threads have finished reading it

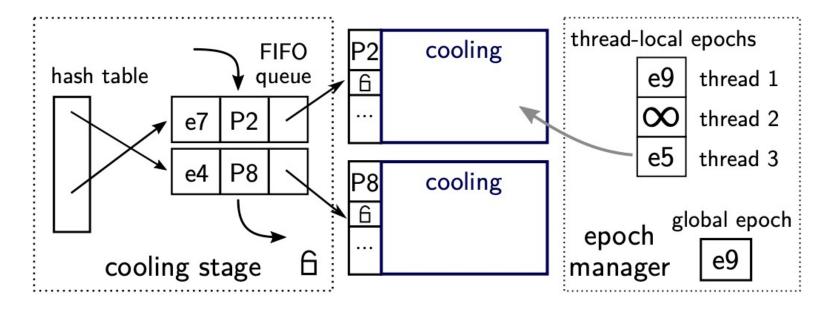


Fig. 6. Epoch-based reclamation.

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### Experiments

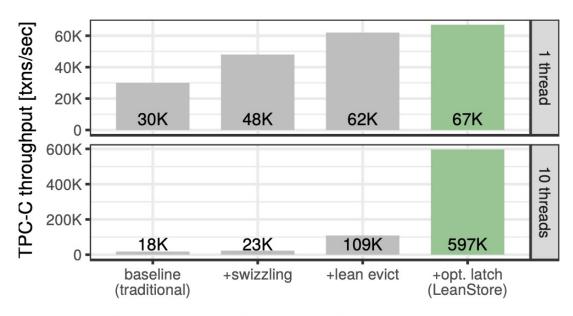


Fig. 7. Impact of the 3 main LeanStore features.

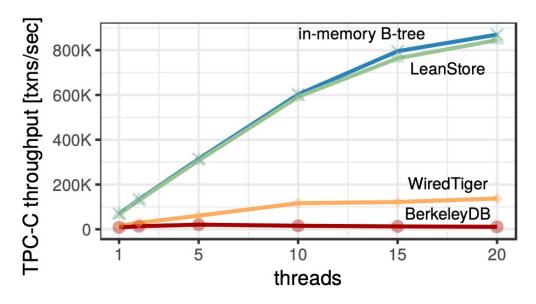


Fig. 8. Multi-threaded, in-memory TPC-C on 10-core system.

### Agenda

Main-memory DB

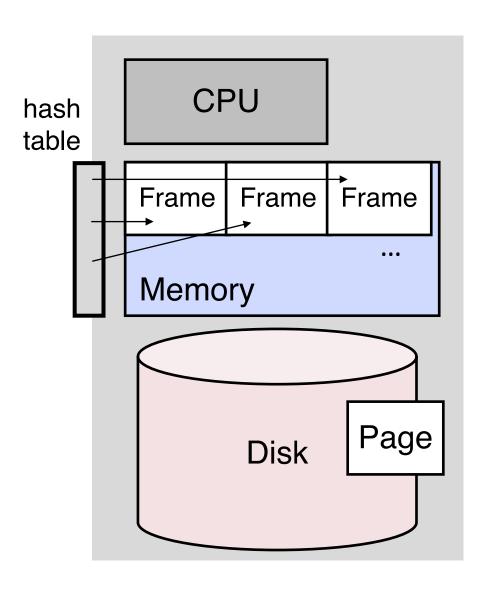
LeanStore design

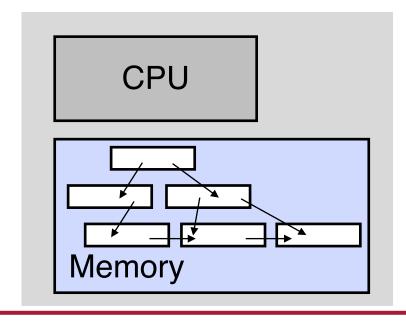
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Fine-grained in-memory data management

### Main-Memory DB Architecture





Fine-granularity: Fine-grained (e.g., tuple-level) data management

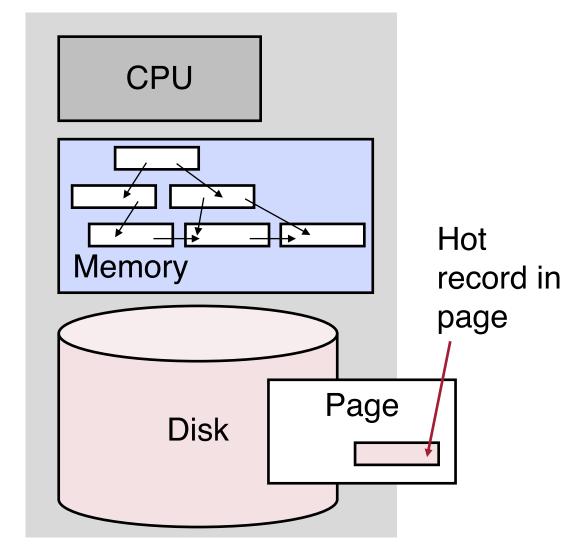
**No Indirection**: reference data following pointers

### Fine-Grained Buffer Management

Migrate tuples, instead of pages, between memory and disk

#### Challenges

- Tracking all data in the system
- Avoid random writes to disk
- Identifying hot/cold data



 <sup>[1]</sup> Justin DeBrabant, et al., <u>Anti-Caching: A New Approach to Database</u>
<u>Management System Architecture</u>. VLDB, 2013
[2] Ahmed Eldawy, et al., <u>Trekking Through Siberia: Managing Cold Data in a Memory-Optimized Database</u>. VLDB 2014

### Q/A - LeanStore

Drawbacks of LeanStore?

– A hot page is constantly unswizzled?

Is scaling out bad in cloud environment?

What recovery guarantees does buffer management provide?

Concurrency control in this paper?

Predict pages for cooling instead of randomly picking?

Why does latching have high overhead?

Is the hash table a bottleneck?

### Wisconsin DB Affiliates Workshop

Time: Thursday, 8:30am-4pm

Location: Northwoods (Union South 3rd Floor)

#### Workshop contents

- Research highlight talk from faculty member
- Research talks from PhD students
- Pitch talks from industry
- Poster session
- Discussion with industry partners including AWS, Databricks, Google, MatrixOrigin, Microsoft, Oracle, Snowflake, TiDB

#### Can also attend on zoom:

https://uwmadison.zoom.us/j/95526978682?pwd=NWxTOXJGSDhiekhwdXBOcG9qMjVKdz09

### Before Next Lecture

Submit review for

Patricia G. Selinger, et al., <u>Access Path Selection in a Relational Database Management System</u>. SIGMOD, 1979