

Carnegie Mellon University

# ADVANCED DATABASE SYSTEMS

In-Memory Databases

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# TODAY'S AGENDA

Background
In-Memory DBMS Architectures
Early Notable In-Memory DBMSs
Peloton Overview
Project #1



# BACKGROUND

Much of the history of DBMSs is about dealing with the limitations of hardware.

Hardware was much different when the original DBMSs were designed:

- → Uniprocessor (single-core CPU)
- $\rightarrow$  RAM was severely limited.
- $\rightarrow$  The database had to be stored on disk.
- $\rightarrow$  Disk is slow. No seriously, I mean really slow.



# BACKGROUND

But now DRAM capacities are large enough that most databases can fit in memory.

- → Structured data sets are smaller.
- → Unstructured or semi-structured data sets are larger.

So why not just use a "traditional" disk-oriented DBMS with a really large cache?



# DISK-ORIENTED DBMS

The primary storage location of the database is on non-volatile storage (e.g., HDD, SSD).

→ The database is organized as a set of fixed-length blocks called **slotted pages**.

The system uses an in-memory (volatile) buffer pool to cache blocks fetched from disk.

→ Its job is to manage the movement of those blocks back and forth between disk and memory.



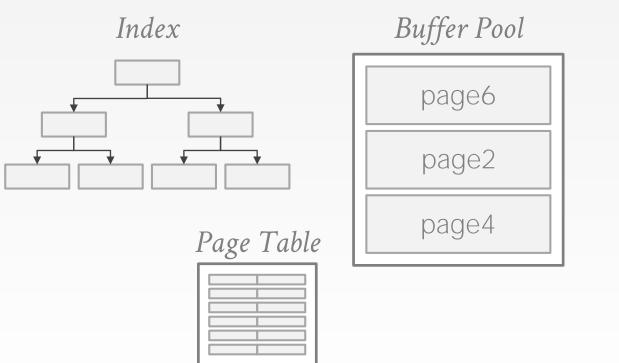
# BUFFER POOL

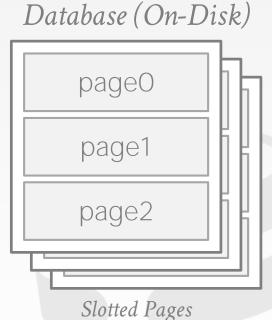
When a query accesses a page, the DBMS checks to see if that page is already in memory:

- → If it's not, then the DBMS has to retrieve it from disk and copy it into a frame in its buffer pool.
- $\rightarrow$  If there are no free frames, then find a page to evict.
- → If the page being evicted is dirty, then the DBMS has to write it back to disk.

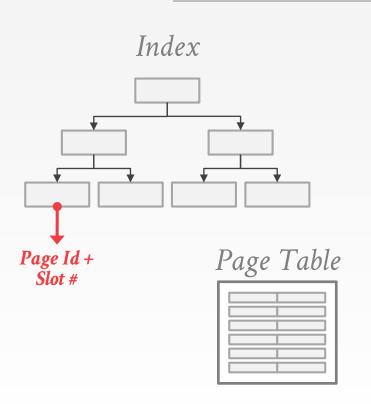
Once the page is in memory, the DBMS translates any on-disk addresses to their in-memory addresses.







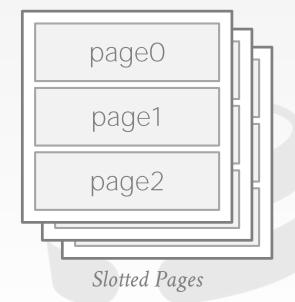




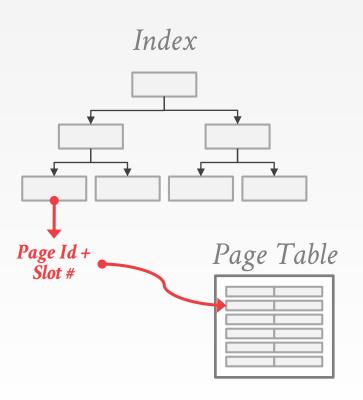
Page6
page2

page4

Database (On-Disk)



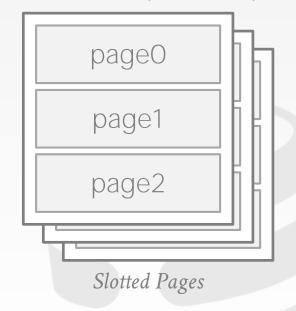


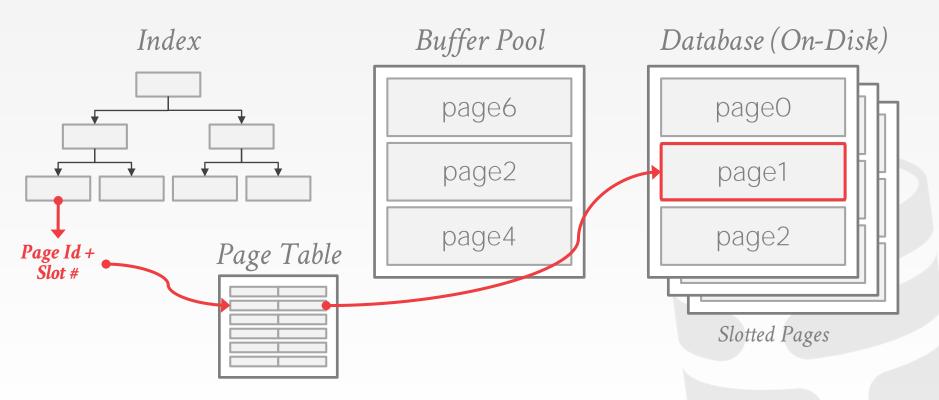


# Buffer Pool

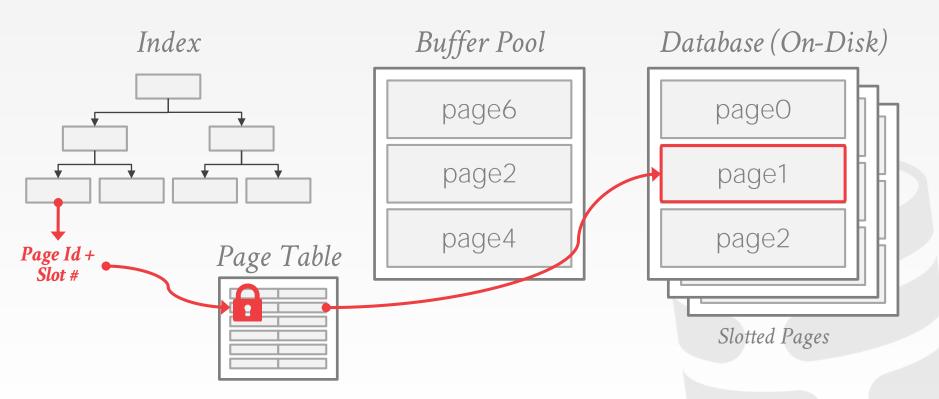


#### Database (On-Disk)

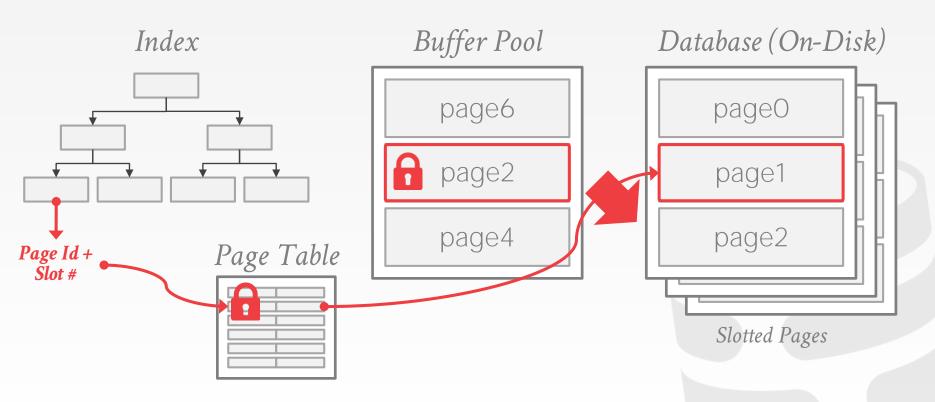




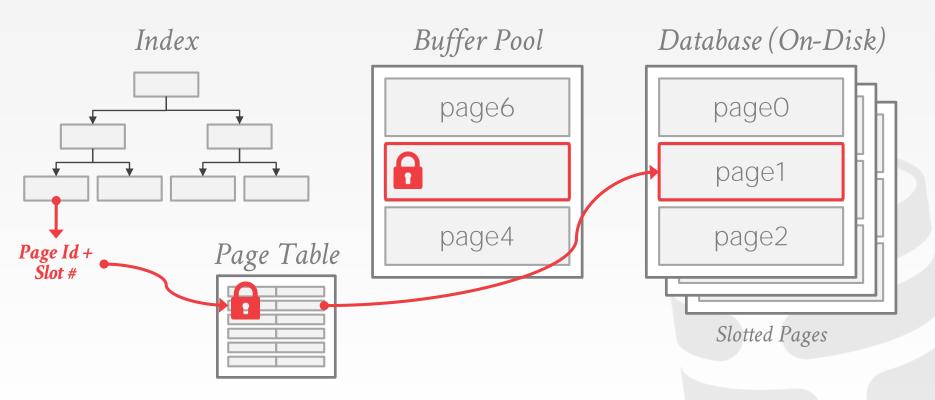




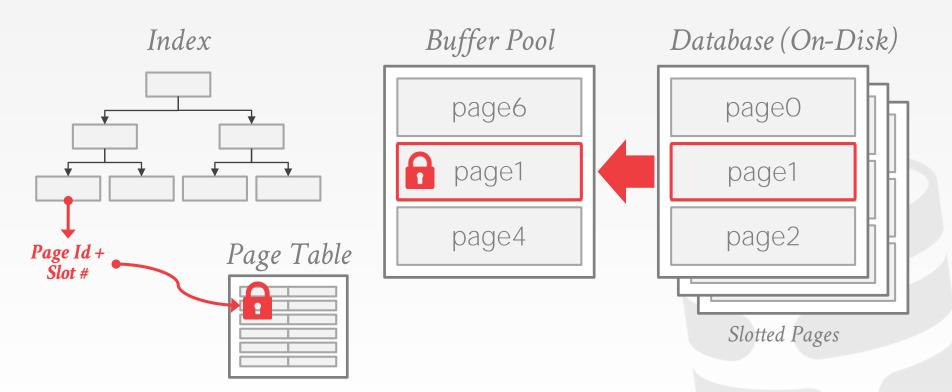




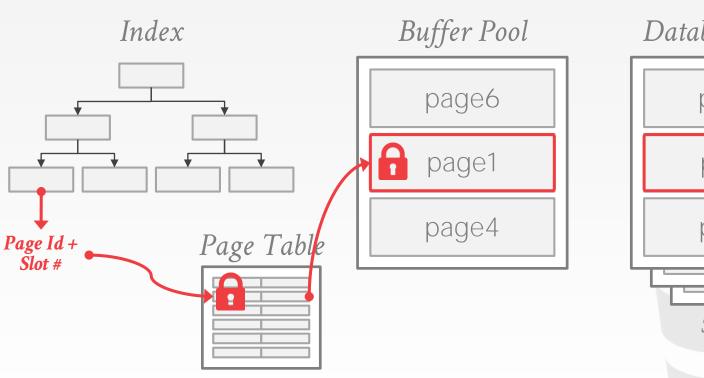


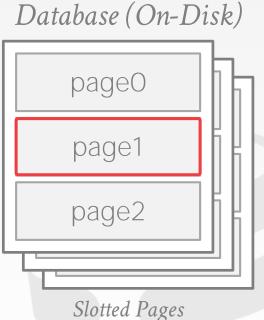




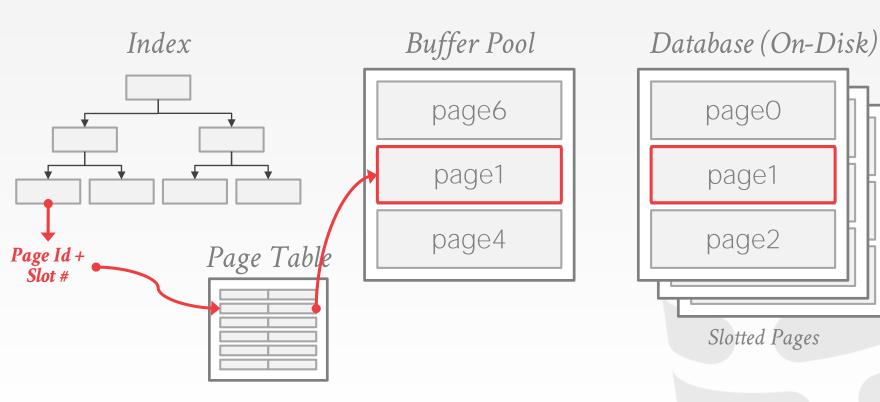














# BUFFER POOL

Every tuple access has to go through the buffer pool manager regardless of whether that data will always be in memory.

- → Always have to translate a tuple's record id to its memory location.
- → Worker thread has to **pin** pages that it needs to make sure that they are not swapped to disk.



# CONCURRENCY CONTROL

In a disk-oriented DBMS, the systems assumes that a txn could stall at any time when it tries to access data that is not in memory.

Execute other txns at the same time so that if one txn stalls then others can keep running.

- → Has to set locks and latches to provide ACID guarantees for txns.
- → Locks are stored in a separate data structure to avoid being swapped to disk.



# LOCKS VS. LATCHES

#### Locks

- → Protects the database's logical contents from other txns.
- $\rightarrow$  Held for txn duration.
- $\rightarrow$  Need to be able to rollback changes.

#### Latches

- → Protects the critical sections of the DBMS's internal data structures from other threads.
- $\rightarrow$  Held for operation duration.
- $\rightarrow$  Do not need to be able to rollback changes.



# LOCKS VS. LATCHES

	Locks	Latches	
Separate	User transactions	Threads	
Protect	Database Contents	In-Memory Data Structures	
During	Entire Transactions	Critical Sections	
Modes	Shared, Exclusive, Update, Intention	Read, Write	
Deadlock	Detection & Resolution	Avoidance	
by	Waits-for, Timeout, Aborts	Coding Discipline	
Kept in	Lock Manager	Protected Data Structure	



# LOGGING & RECOVERY

Most DBMSs use **STEAL** + **NO-FORCE** buffer pool policies, so all modifications have to be flushed to the WAL before a txn can commit.

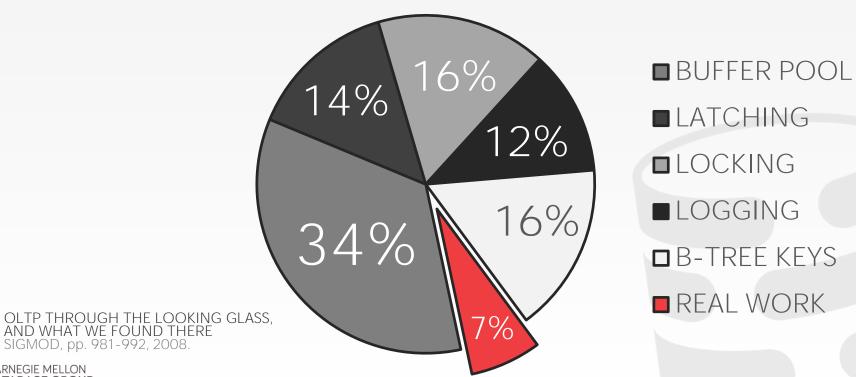
Each log entry contains the before and after image of record modified.

Lots of work to keep track of LSNs all throughout the DBMS.



# DISK-ORIENTED DBMS OVERHEAD

#### Measured CPU Instructions



SIGMOD, pp. 981-992, 2008.

# IN-MEMORY DBMSS

Assume that the primary storage location of the database is **permanently** in memory.

Early ideas proposed in the 1980s but it is now feasible because DRAM prices are low and capacities are high.



# BOTTLENECKS

If I/O is no longer the slowest resource, much of the DBMS's architecture will have to change account for other bottlenecks:

- → Locking/latching
- → Cache-line misses
- → Pointer chasing
- → Predicate evaluations
- → Data movement & copying
- → Networking (between application & DBMS)



# STORAGE ACCESS LATENCIES

_		L3	DRAM	SSD	HDD
	Read Latency	~20 ns	60 ns	25,000 ns	10,000,000 ns
	Write Latency	~20 ns	60 ns	300,000 ns	10,000,000 ns

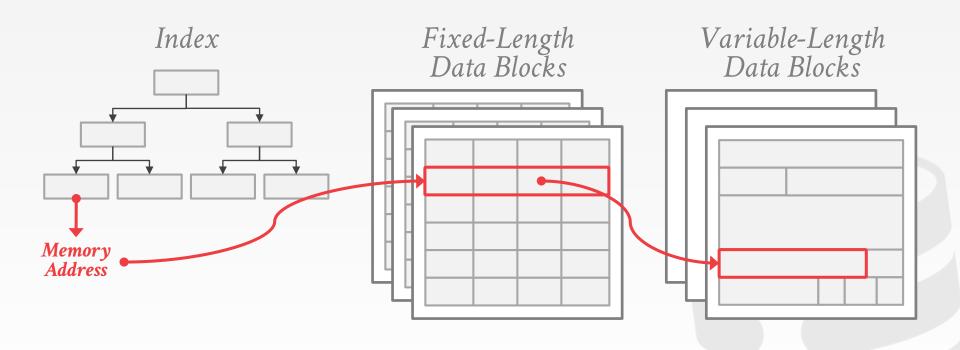


An in-memory DBMS does not need to store the database in slotted pages but it will still organize tuples in blocks/pages:

- → Direct memory pointers vs. record ids
- → Fixed-length vs. variable-length data pools
- → Use checksums to detect software errors from trashing the database.

The OS organizes memory in pages too. We will cover this later.







# WHY NOT MMAP?

Memory-map (mmap) a database file into DRAM and let the OS be in charge of swapping data in and out as needed.

Use **madvise** and **msync** to give hints to the OS about what data is safe to flush.

# Notable mmap DBMSs:

- → MongoDB (pre WiredTiger)
- $\rightarrow$  MonetDB
- $\rightarrow$  LMDB



# WHY NOT MMAP?

Using mmap gives up fine-grained control on the contents of memory.

- → Cannot perform non-blocking memory access.
- → The "on-disk" representation has to be the same as the "in-memory" representation.
- → The DBMS has no way of knowing what pages are in memory or not.
- → Various mmap-related syscalls are not portable.

A well-written DBMS always knows best.



# CONCURRENCY CONTROL

Observation: The cost of a txn acquiring a lock is the same as accessing data.

In-memory DBMS may want to detect conflicts between txns at a different granularity.

- → <u>Fine-grained locking</u> allows for better concurrency but requires more locks.
- → <u>Coarse-grained locking</u> requires fewer locks but limits the amount of concurrency.



# CONCURRENCY CONTROL

The DBMS can store locking information about each tuple together with its data.

- $\rightarrow$  This helps with CPU cache locality.
- → Mutexes are too slow. Need to use CAS instructions.

New bottleneck is contention caused from txns trying access data at the same time.



#### INDEXES

Specialized main-memory indexes were proposed in 1980s when cache and memory access speeds were roughly equivalent.

But then caches got faster than main memory:

→ Memory-optimized indexes performed worse than the B+trees because they were not cache aware.

Indexes are usually rebuilt in an in-memory DBMS after restart to avoid logging overhead.



# QUERY PROCESSING

The best strategy for executing a query plan in a DBMS changes when all of the data is already in memory.

→ Sequential scans are no longer significantly faster than random access.

The traditional **tuple-at-a-time** iterator model is too slow because of function calls.

→ This problem is more significant in OLAP DBMSs.



# LOGGING & RECOVERY

The DBMS still needs a WAL on non-volatile storage since the system could halt at anytime.

- → Use **group commit** to batch log entries and flush them together to amortize **fsync** cost.
- → May be possible to use more lightweight logging schemes (e.g., only store redo information).

But since there are no "dirty" pages, there is no need to maintain LSNs all throughout the system.



# LOGGING & RECOVERY

The system also still takes checkpoints to speed up recovery time.

# Different methods for checkpointing:

- → Old idea: Maintain a second copy of the database in memory that is updated by replaying the WAL.
- → Switch to a special "copy-on-write" mode and then write a dump of the database to disk.
- → Fork the DBMS process and then have the child process write its contents to disk.



# LARGER-THAN-MEMORY DATABASES

DRAM is fast, but data is not accessed with the same frequency and in the same manner.

- → Hot Data: OLTP Operations
- → Cold Data: OLAP Queries

We will study techniques for how to bring back disk-resident data without slowing down the entire system.



#### NON-VOLATILE MEMORY

Emerging hardware that is able to get almost the same read/write speed as DRAM but with the persistence guarantees of an SSD.

- → Also called *storage class memory*
- → Examples: Phase-Change Memory, Memristors

It's not clear how to build a DBMS to operate on this kind memory.

Again, we'll cover this topic later.



# NOTABLE IN-MEMORY DBMSs

Oracle TimesTen

Dali / DataBlitz

<u>Altibase</u>

P\*TIME

**SAP HANA** 

VoltDB / H-Store

Microsoft Hekaton

**Harvard Silo** 

TUM HyPer

<u>MemSQL</u>

IBM DB2 BLU

Apache Geode



#### TIMESTEN

Originally SmallBase from HP Labs in 1995.

Multi-process, shared memory DBMS.

- → Single-version database using two-phase locking.
- $\rightarrow$  Dictionary-encoded columnar compression.

Bought by Oracle in 2005.

Can work as a cache in front of Oracle DBMS.





#### DALI / DATABLITZ

Developed at AT&T Labs in the early 1990s.

Multi-process, shared memory storage manager using memory-mapped files.

Employed additional safety measures to make sure that erroneous writes to memory do not corrupt the database.

- → Meta-data is stored in a non-shared location.
- → A page's checksum is always tested on a read; if the checksum is invalid, recover page from log.



#### P\*TIME

Korean in-memory DBMS from the 2000s.

Performance numbers are still impressive.

Lots of interesting features:

- $\rightarrow$  Uses differential encoding (XOR) for log records.
- $\rightarrow$  Hybrid storage layouts.
- → Support for larger-than-memory databases.

Sold to SAP in 2005. Now part of HANA.



P\*TIME: HIGHLY SCALABLE OLTP DBMS FOR MANAGING UPDATE-INTENSIVE STREAM WORKLOAD VLDB, pp. 1033-1044, 2004.



# S PELOTON DBMS

#### CMU's in-memory hybrid relational DBMS

- → Latch-free Multi-version concurrency control.
- → Latch-free Bw-Tree Index
- → LLVM-based Execution Engine
- → Tile-based storage manager.
- → Multi-threaded architecture.
- → Write-Ahead Logging + Checkpoints
- → Cascades-style Query Optimizer
- → Zone Maps
- → PL/pgSQL UDFs (preliminary)

Currently supports **some** of SQL-92.



# PROJECT #1

Implement the SQL String Functions

- → UPPER()
- → LOWER()
- → CONCAT()

You only need to support two arguments for the **CONCAT** function.

This project is meant to help you understand how our query code generation engine works.



#### PROJECT #1

Step #1 – Implement the String Functions

Step #2 – Register the Functions in the Catalog

Step #3 – Add Function Proxies



# PROJECT #1 - TESTING

We are providing you with a basic C++ unit test for you check your implementation.

We also have a SQL batch script that will execute a couple different queries.

We strongly encourage you to do your own additional testing.



# PROJECT #1 - GRADING

We will run additional tests beyond what we provided you for grading.

We will also use gcc sanitize when testing your code.

All source code must pass ClangFormat syntax formatting checker.

→ See Peloton <u>documentation</u> for formatting guidelines



# DEVELOPMENT ENVIRONMENT

Peloton only builds on 64-bit Linux and OSX.

You can also do development on a VM.

→ We have a <u>Vagrant</u> config file to automatically create a development Ubuntu 14.04 VM for you.

This is CMU so I'm going to assume that each of you are capable of getting access to a machine.



# PROJECT #1

**Due Date:** January 29<sup>th</sup> @ 11:59pm

Projects will be turned in using Autolab.

**Recitation:** Tuesday January 23<sup>rd</sup> @ 5:00pm in GHC 9115.

Full description and instructions:

http://15721.courses.cs.cmu.edu/spring2018/project1.html



#### PARTING THOUGHTS

Disk-oriented DBMSs are a relic of the past.

→ Most databases fit entirely in DRAM on a single machine.

The world has finally become comfortable with inmemory data storage and processing.

Never use **mmap** for your DBMS.



# NEXT CLASS

Query Code Generation + Compilation

