Lecture 23: Vectorized Execution & Course Retrospective



Logistics

In-class presentations on Nov 21



Recap

- Columnar Storage
- Compression
- Compressed Columnar Storage



Lecture Overview

- Vectorized Execution
- Course Retrospective



Vectorized Execution



Limitations of Tuple-at-a-time Processing

- Each tuple incurs the cost of:
 - Function calls between operators.
 - Deserializing, interpreting, and processing the tuples.
 - **Doesn't leverage the CPU's ability to process batches of data efficiently.** •
 - Results in frequent pipeline stalls and cache misses.



Vector-at-a-time Processing

- Process a vector of tuples at a time to reduce overhead of function calls etc.
- Use SIMD (Single Instruction, Multiple Data) instructions
- A single instruction operates on multiple data points simultaneously.

18	40	25	15	
0	1	1	1	

d of function calls etc. ons simultaneously.





SIMD in Query Execution

SIMD Use Cases:

- Filtering (e.g., select rows within a range).
- Aggregations (e.g., sum, average).
- Compression (e.g., decoding bit-packed data).

Key SIMD Operations

- Vector Loads: Load multiple data points.
- Masks: Apply conditions to filter data.
- Horizontal Reduction: Sum vector elements.



Filter timestamps using SIMD

- Load data in batches using vld1q_s32.
- Perform vectorized comparisons using vcgeq_s32 and vcleq_s32.
- Combine results with a bitwise AND using vandq_u32.

```
int32x4_t ts_vec = vld1q_s32(&data.timestamps[i]); // Load timestamps
uint32x4_t mask = vandq_u32(vcgeq_s32(ts_vec, 1),
                            vcleq_s32(ts_vec, end)); // Mask for range
```



Filter timestamps using SIMD

vld1q_s32	18	40	25
vcgeq_s32			
	0	1	1
vcleq_s32			
	1	0	1
vandq_u32			
	0	0	1





Scalar vs SIMD Execution

- Scalar Execution:
 - Processes one data element per cycle.
 - Repeated instruction fetch, decode, and execute for each element.
- SIMD Execution:
 - Processes multiple data elements per cycle by leveraging wide registers.
 - Executes the same operation on an entire vector (batch) with a single instruction.



Benefits of SIMD Execution

- Instruction-Level Efficiency
 - Fewer instructions due to vectorized operations.
 - Scalar processing requires N instructions for N data points.
 - SIMD requires N / W instructions, where W is the width of the SIMD vector.
- Cache and Memory Efficiency
 - Contiguous memory access aligns with cache lines.
 - SIMD operates on contiguous memory (columnar layouts align well with SIMD).
 - Cache lines are fully utilized, reducing memory latency.



Benefits of SIMD Execution

- Minimized Control Overhead:
 - SIMD minimizes branching by applying the same operation to all elements in a vector.
 - With scalar execution, pipeline stalls if the branch prediction is incorrect.
 - With SIMD execution, masks handle conditional operations, avoiding pipeline stalls.
- Hardware Support:
 - Modern CPUs have dedicated SIMD execution units optimized for throughput.



History of SIMD

- 1960s-1970s: Supercomputers and Scientific Computing
 - SIMD first appeared in systems like ILLIAC IV for scientific workloads.
 - Allowed simultaneous operations on multiple data points (e.g., matrix rows).
- 1980s-1990s: Multimedia Processing
 - MMX (Intel, 1996): Integer operations for multimedia.
 - Example: Increase the brightness of an image's pixels by a constant value.

Pixels	100	120	140	
	120	140	160	

160

+ 20

180



Modern SIMD in General-Purpose Computing

- 2000s: Integration in General-Purpose CPUs
 - SIMD became a standard in consumer CPUs.
 - SSE/AVX (Intel): Wide registers for floats and integers.
 - NEON (ARM): Optimized for embedded and mobile devices.
- 2010s-Present: Acceleration for Analytical Workloads
 - SIMD now powers modern databases and big data systems.
 - Vectorized query execution (e.g., Apache Arrow).



Sensor Data Analysis

5

};

- Designed with SoA (Structure of Arrays) layout for SIMD.
- Ensures contiguous memory access for efficient vectorized operations.

struct SensorData { int* timestamps; // Contiguous array of timestamps float* temperatures; // Contiguous array of temperatures

```
SensorData(size_t count) {
    timestamps = static_cast<int*>(aligned_alloc(16, sizeof(int) * count));
    temperatures = static_cast<float*>(aligned_alloc(16, sizeof(float) * count));
```



Sensor Data Generation

}

Create synthetic data with timestamps and temperatures.

void generateData(SensorData& data, int count) {
 std::random_device rd;
 std::mt19937 gen(rd());
 std::normal_distribution<float> temp_dist(25.0, 5.0);
 std::uniform_int_distribution<int> time_dist(1, 5);

```
int timestamp = START_TIMESTAMP;
for (int i = 0; i < count; ++i) {
    data.timestamps[i] = timestamp;
    data.temperatures[i] = temp_dist(gen);
    timestamp += time_dist(gen); // Increment timestamp</pre>
```



SIMD Query

• Task: Calculate the average temperature within a timestamp range

Timestamps	1	2	5
Temperature	25.5	26.0	27.2
> 2	0	1	1
< 6	1	1	1
> 2 AND < 6	0	1	1
Masked Temps	0	26.0	27.2
Sum	53.2		

7
28.3
0
0
0
0



SIMD Query: Loading Data

Load 4 consecutive timestamps and temperatures into SIMD registers.

int32x4_t ts_vec = vld1q_s32(&data.timestamps[i]); // Load 4 timestamps float32x4_t temp_vec = vld1q_f32(&data.temperatures[i]); // Load 4 temperatures



SIMD Query: Filtering Timestamps

- Filter timestamps within the query range.

- Compare for Lower Bound: vcgeq_s32: Compares timestamps with startTimestamp. Compare for Upper Bound: vcleq_s32: Compares timestamps with endTimestamp. Combine Results: vandq_u32: Combines the two masks with a bitwise AND.

uint32x4_t in_range_mask = vandq_u32(vcgeq_s32(ts_vec, vdupq_n_s32(startTimestamp)), vcleq_s32(ts_vec, vdupq_n_s32(endTimestamp)));





SIMD Query: Mask Application

- Apply the mask to the temperatures and sum up the valid values.
- vmulq_f32: Multiplies the mask with the temperature vector.
- Keeps valid temperatures, zeros out invalid ones.
- vaddq_f32: Adds the valid temperatures to the running sum.

float32x4_t masked_temps = vmulq_f32(temp_vec, vcvtq_f32_u32(in_range_mask)); sum_vec = vaddq_f32(sum_vec, masked_temps);



SIMD Query: Final Steps

- Horizontal Reduction: Sum up all elements in the SIMD vector.
- Handle Remaining Scalar Elements: Process leftover elements not divisible by the SIMD width.

```
float total_sum = vaddvq_f32(sum_vec);
for (int i = count - (count % 4); i < count; ++i) {</pre>
    if (data.timestamps[i] >= startTimestamp && data.timestamps[i] <= endTimestamp) {</pre>
        total_sum += data.temperatures[i];
    }
}
```



SIMD Instruction Naming

- SIMD instruction names are structured to reflect:
 - Operation type: Add, multiply, compare, etc.
 - Data type: Integer, floating-point, or specialized formats.
 - Vector width: Number of data elements processed.

vaddq_f32	vaddq_f32 Addition Operation	
vcgeq_s32	Compare greater than or equal to	Operates on sig 32-bit integer

point

gned rs. Quad-word (4 floats) 64-bit vector (2 integers)



Evolution of SIMD Widths Over Time

- Wider SIMD registers enable higher parallelism.
- 64-bit SIMD:
 - Technologies: MMX (Intel, 1996).
 - Operated on 64-bit registers (e.g., 4 integers).
- 128-bit SIMD:
 - Technologies: SSE (Intel), NEON (ARM).
 - Supported 4x32-bit floats or 2x64-bit doubles.



Evolution of SIMD Widths Over Time

- 256-bit SIMD:
 - Technologies: AVX (Intel, 2010).
 - Doubled vector width for 8x32-bit floats or 4x64-bit doubles.
- 512-bit SIMD:
 - Technologies: AVX-512 (Intel, 2017).
 - Supported 16x32-bit floats or 8x64-bit doubles in a single operation.



Advanced SIMD: Hash Join Algorithm

- Hash computation and comparison in parallel using SIMD masks.
- Hash table lookups may involve non-contiguous memory accesses

// Load probe keys int32x4_t probe_keys = vld1q_s32(&probe_table.keys[i]); // Compute hash int32x4_t hash_vec = vmodq_s32(vmulq_s32(probe_keys, hash_multiplier), hash_table_size); // Gather hash table values int32x4_t hash_table_vals = vld1q_s32(&hash_table[hash_vec]); // Compare keys uint32x4_t match_mask = vceqq_s32(probe_keys, hash_table_vals);



Course Retrospective



Takeaways from the Course

- Let's take a step back and reflect on what you've accomplished.
- Systems programming is challenging.
 - Delving into internals teaches attention to detail. •
 - It forces understanding of how things work under the hood. ٠
- Foundational systems knowledge beyond databases.
 - Threading, memory management, and I/O.
 - You now have tools to approach any system-level problem.
 - Reflect on how much you've learned and grown as a programmer.



Big Ideas from the Course

- Database Systems Are Awesome
 - They solve real-world problems elegantly.
 - But they are not magic.
- Abstractions Are Key
 - Elegant abstractions are the "magic" enabling usability and performance.
- Declarativity Rules
 - Declarative query models make complex systems usable.
 - Taken to the extreme -- Google search



Big Ideas from the Course

- Building Systems Is More Than Hacking:
 - It's about design, principles, and reusability.
- Recurring Patterns:
 - Recognizing motifs like parallelism, caching, and transactions.
- Intellectual Contributions:
 - Computer Science is evolving, and you can contribute to its history.



Looking Ahead: What's Next?

- CS 6423 (Advanced Database Implementation)
 - Query Optimization
 - Concurrency Control
 - Logging and Recovery
- Building on this course
 - Deeper insights into how databases optimize for efficiency.
 - Expanding your knowledge of system-level guarantees.



Looking Ahead: What's Next?

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Logging and Recovery

• Example: A system crash during a transfer:

UPDATE accounts SET balance = balance - 100 WHERE id = 1; UPDATE accounts SET balance = balance + 100 WHERE id = 2;



Logging and Recovery

- Mechanisms to restore the database to a consistent state after crashes.
- Write-Ahead Logging (WAL)
 - Log changes before applying them.
- Checkpointing and Crash Recovery
 - Periodically save the database state to reduce recovery time.
 - Redo/Undo logs to reconstruct committed transactions and roll back uncommitted ones.
- ARIES Algorithm
 - Advanced Recovery Algorithm (Analysis, Redo, Undo).



Concurrency Control

Example: Two users simultaneously trying to update the same row

UPDATE accounts SET balance = balance - 100 WHERE id = 1;



Concurrency Control

- Ensuring correctness and consistency when multiple users access the database simultaneously.
- Locks and Latches:
 - Types of locks (shared, exclusive); Deadlock detection and prevention.
- Multi-Version Concurrency Control (MVCC)
 - Readers don't block writers; writers don't block readers.
- Isolation Levels:
 - Read Committed, Repeatable Read, Serializable.
- Distributed Transactions:
 - Two-phase commit (2PC), distributed locking.



Query Optimization

 Example: Push filters before the join. Use an indexed join if available.

SELECT * FROM orders JOIN customers ON orders.customer_id = customers.id WHERE customers.city = 'Atlanta';



Query Optimization

- Selecting the best execution plan for a query.
- Goal: Minimize execution cost (e.g., time, memory, I/O).
- Execution Plans
 - Logical vs. physical plans.
- Cost Models
 - Estimating costs for different plans.
- Heuristics and Rules
 - Simplifying query trees and reordering joins.
- Advanced Techniques
 - Dynamic Programming (e.g., System R algorithm).
 - Cardinality Estimation



Feedback and Project Presentations

- Please share your feedback via CIOS
- +1% extra credit for entire class if we get 80%+ participation
- In-class presentations on Nov 21
- Tentatively prepare a 5-minute presentation



Conclusion

- Vectorized Execution
- Course Retrospective

