

Studying state-of-the-art HTAP systems

- Satya Sai Bharath Vemula
- Rwitam Bandyopadhyay
- Shubham Pandey

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Introduction (~4 min)

- History of OLAP & OLTP
- Challenges of HTAP Systems
- Architectural taxonomy of present HTAP Systems

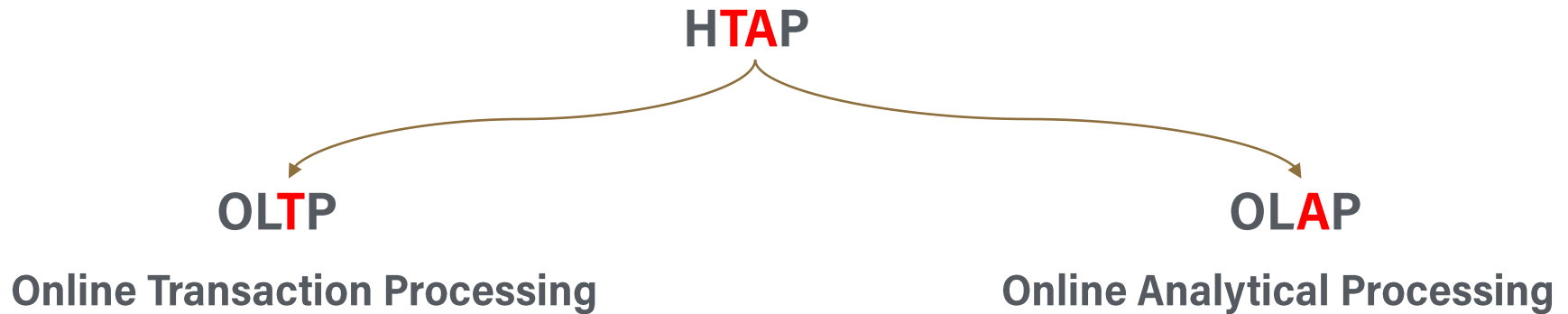
Oracle Dual-Format Database (~8 min)

SAP HANA (~8 min)

TiDB (~9 min)

Project Plan (~1 min)

Introduction

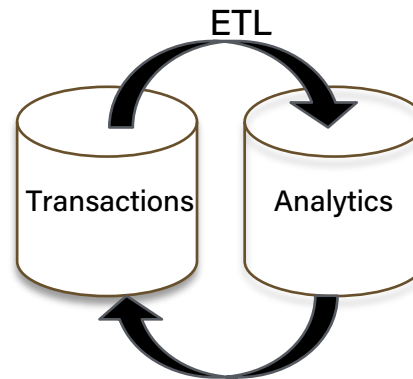


Challenges of HTAP Systems

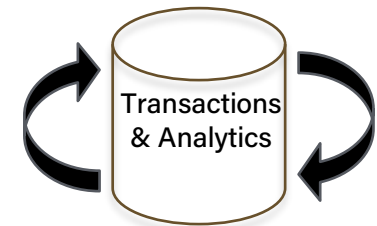
What is it not easy to imagine both systems in one?

- Data modeling complexity
- Performance optimization
- Consistency and correctness
- Scalability
- Cost

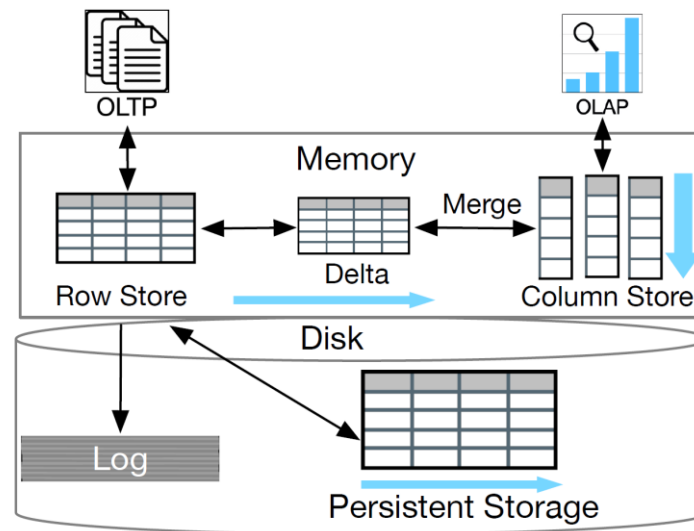
Traditional Systems



HTAP Systems

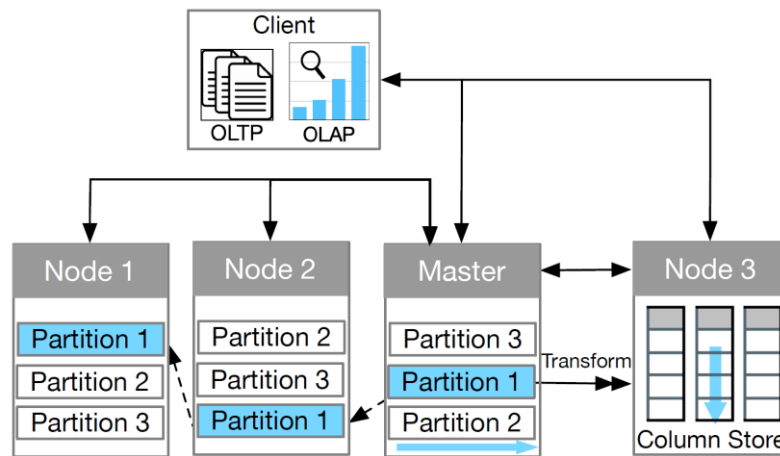


Storage Architectures of HTAP Databases



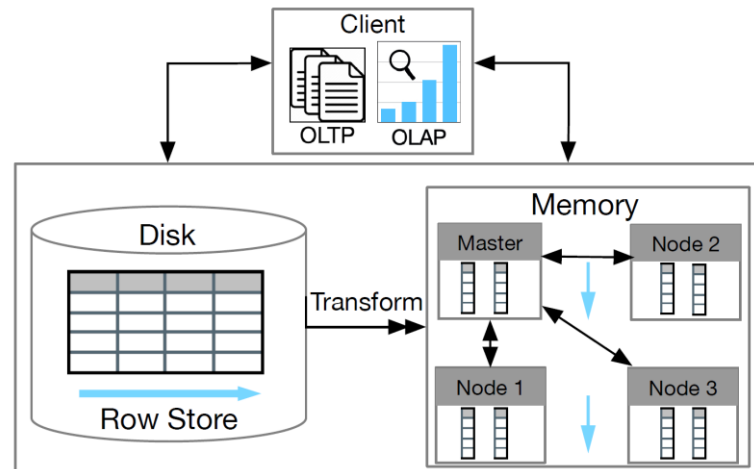
Primary Row Store + In-Memory Column Store

Storage Architectures of HTAP Databases



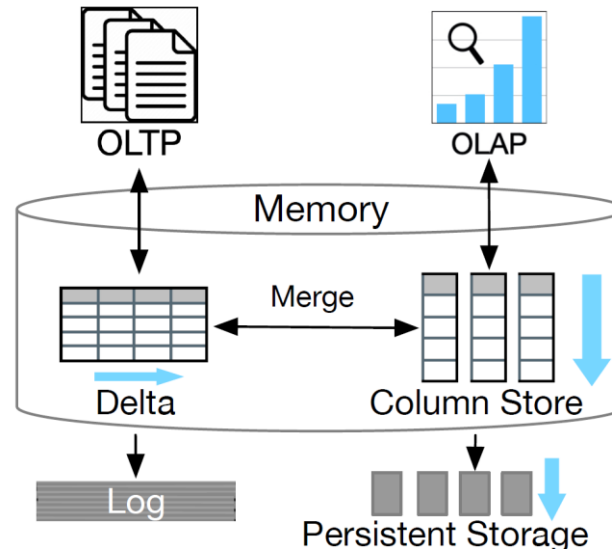
Distributed Row Store + Column Store Replica

Storage Architectures of HTAP Databases



Disk Row Store + Distributed Column Store

Storage Architectures of HTAP Databases



Primary Column Store + Delta Row Store

Oracle Dual Format

Disagg1

Presented by Rwitam Bandyopadhyay

Motivation

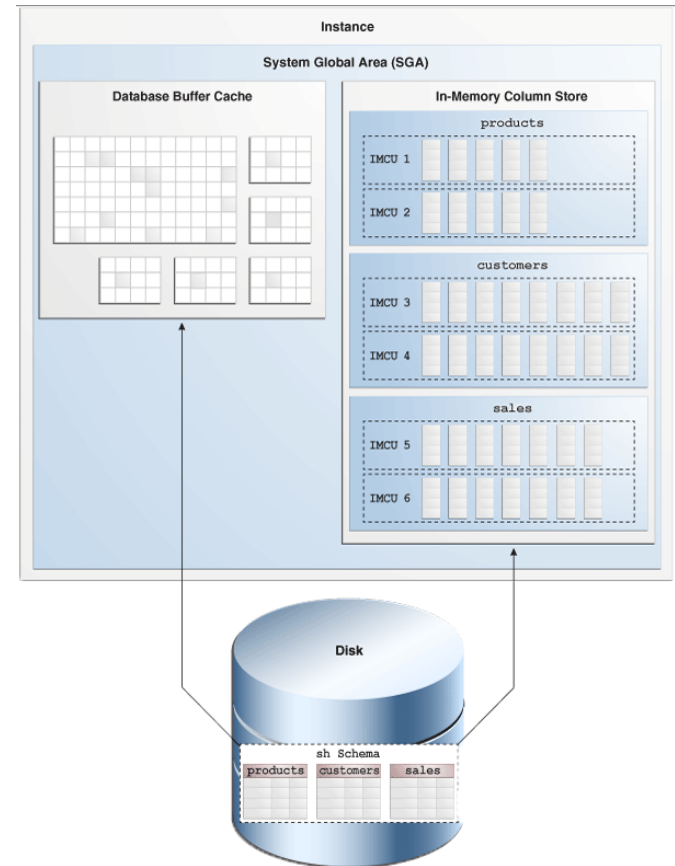
Oracle introduced the Database In-Memory option in 2014 as the industry's first dual-format, in-memory RDBMS

- No single data format is ideal for all type of workloads
- “Primary Row Store + In-Memory Column Store” architectural style
- The new columnar format is a pure in-memory format with no impact to the disk representation
- Tables required for fast analytics can be populated into the In-Memory column store
- Can be plugged in to any existing system with zero changes

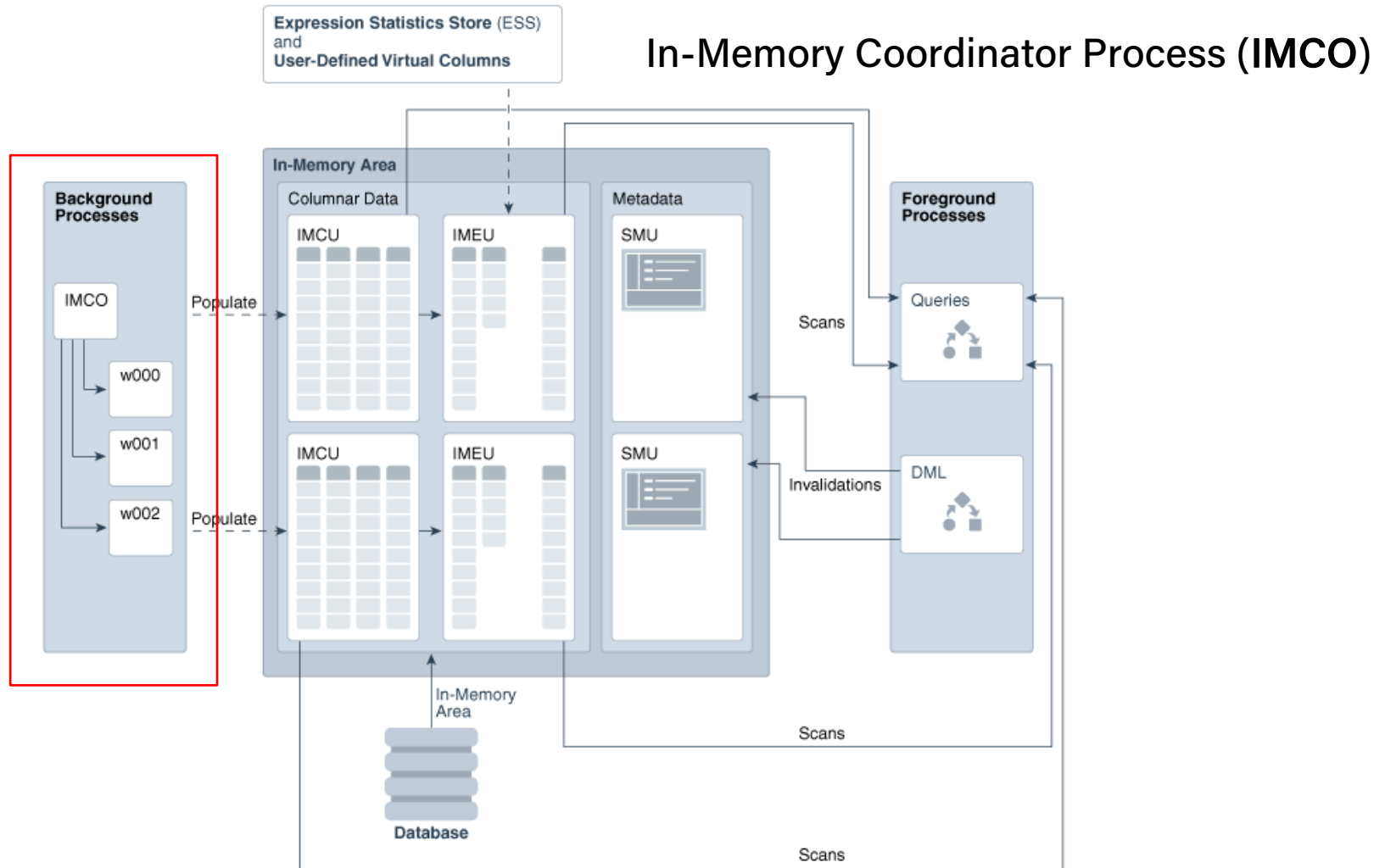
Dual-Format, Dual Memory?

Dual format representation doesn't double memory requirements!

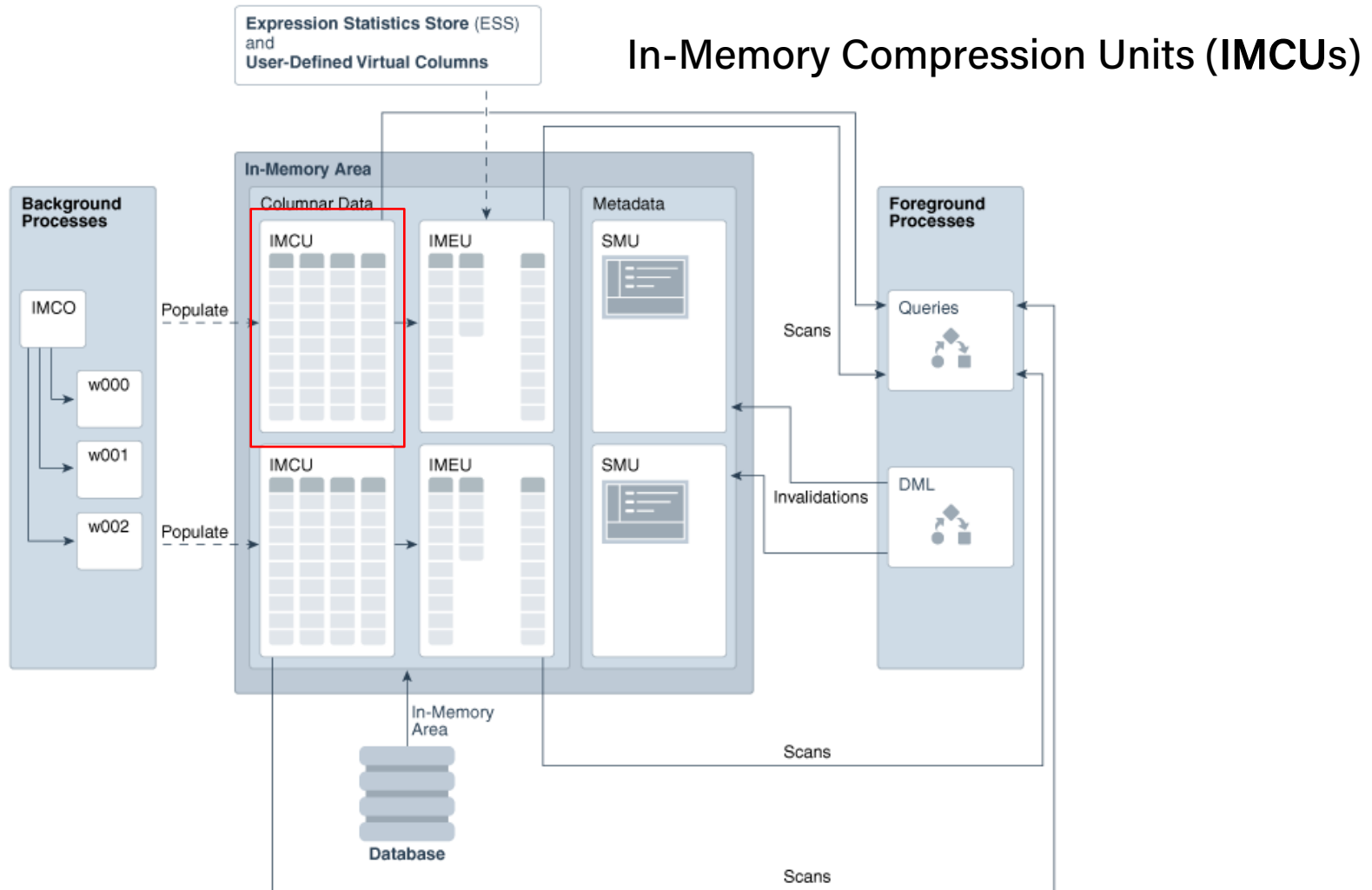
- Tables required for fast analytics
→ In-Memory Store
- OLTP updates/ highly selective lookups
→ Buffer Cache !
- Analytic Indexes can be dropped from the DB
- The IM column store is built into the DB data access layer



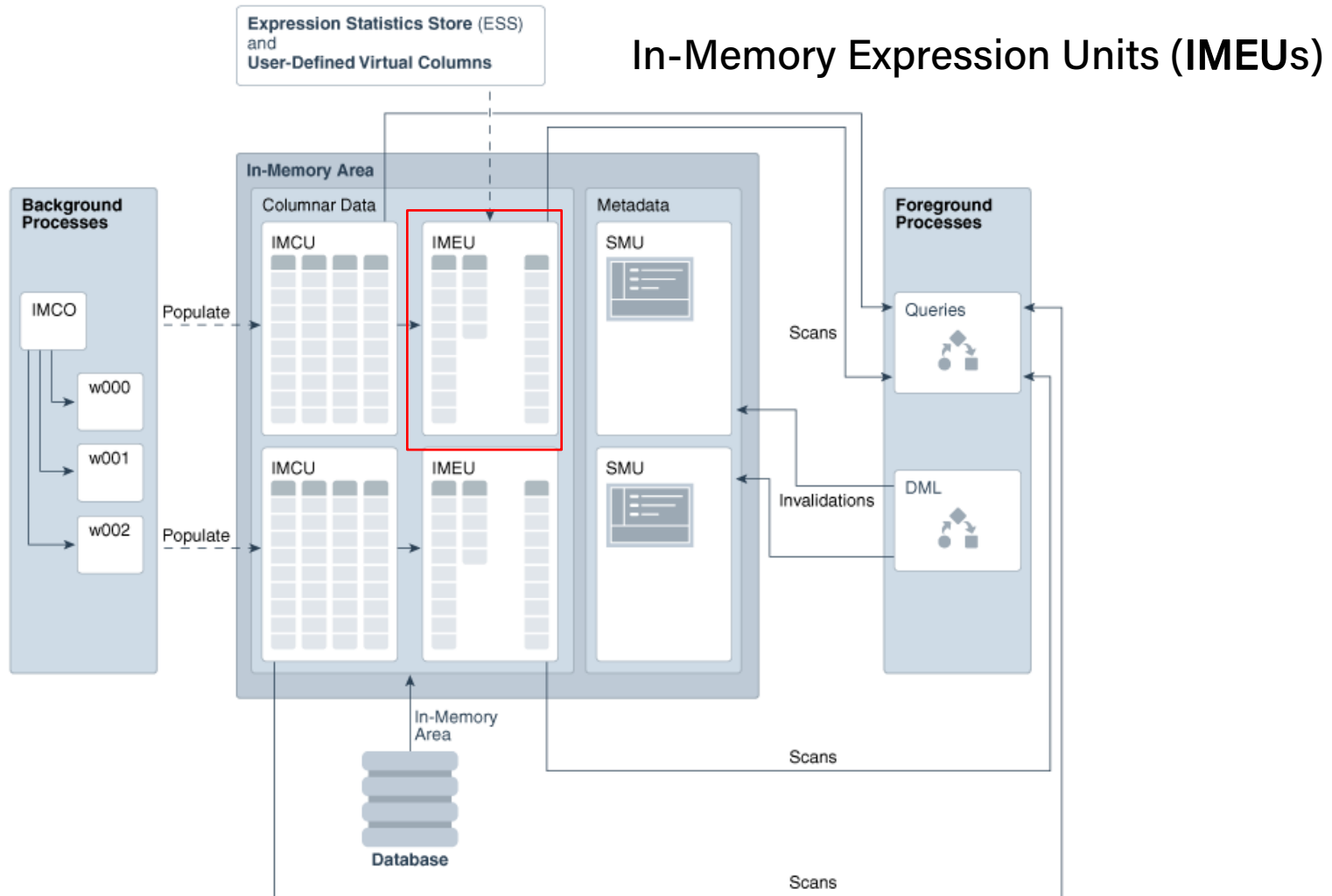
In-Memory Columnar Format (Building Blocks)



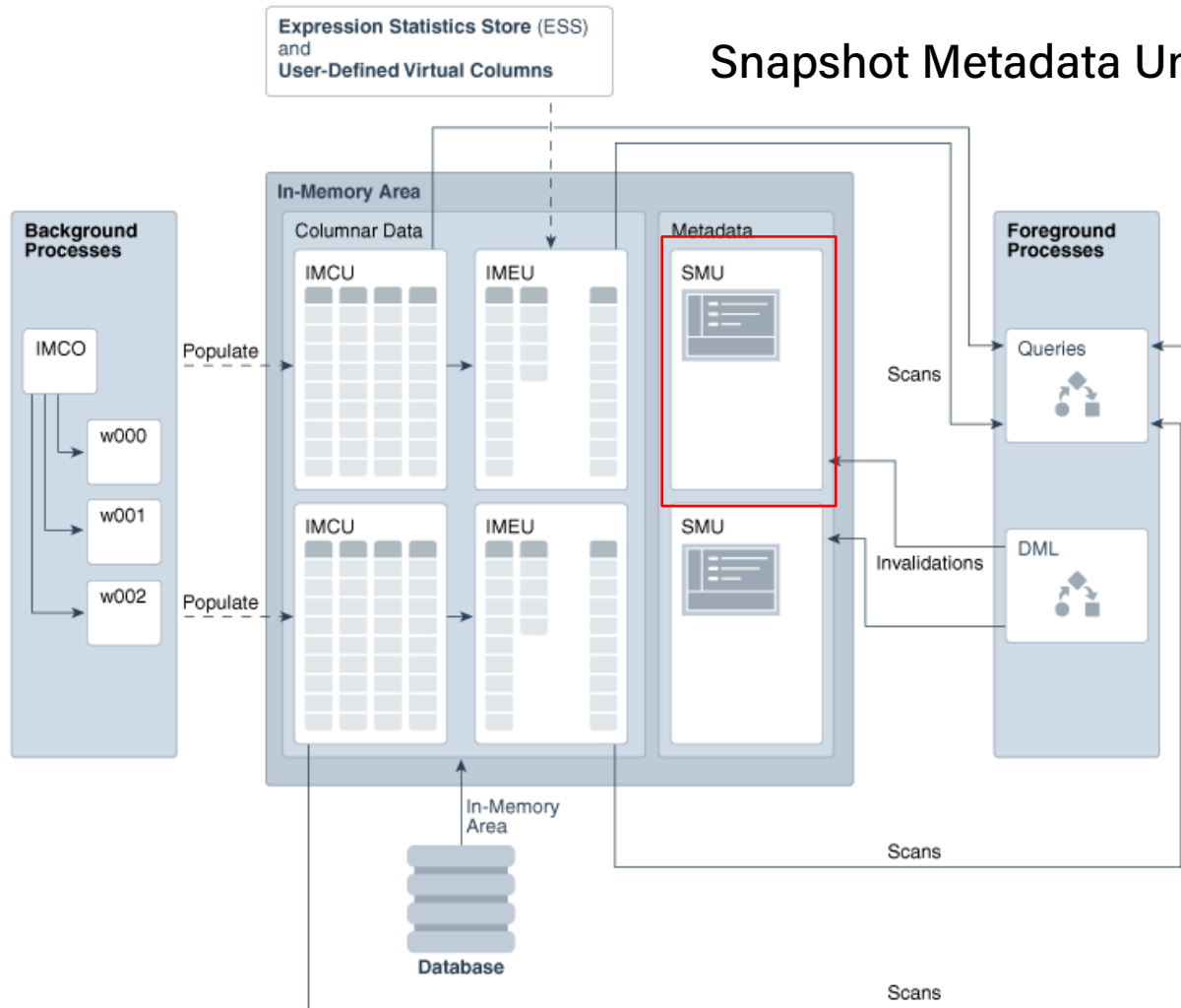
In-Memory Columnar Format (Building Blocks)



In-Memory Columnar Format (Building Blocks)



In-Memory Columnar Format (Building Blocks)



In-Memory Columnar Format (Data Population)

It is possible to selectively populate the IM column store with a chosen subset of the database

- IM column store is populated using a pool of background server processes
- No application downtime during table population since it continues to be accessible via the buffer cache.
- Most other in-memory databases require a wait time for all objects to be brought in-memory

Priority	Description
CRITICAL	Object is populated immediately after the database is opened.
HIGH/ MEDIUM/ LOW	Objects are populated after CRITICAL in order of priority.
NONE	Object is only populated after it is scanned for the first time (Default).

In-Memory Columnar Format (Data Compression)

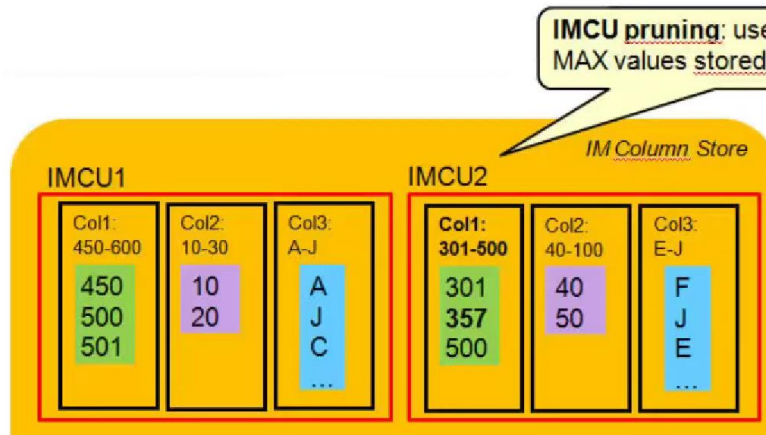
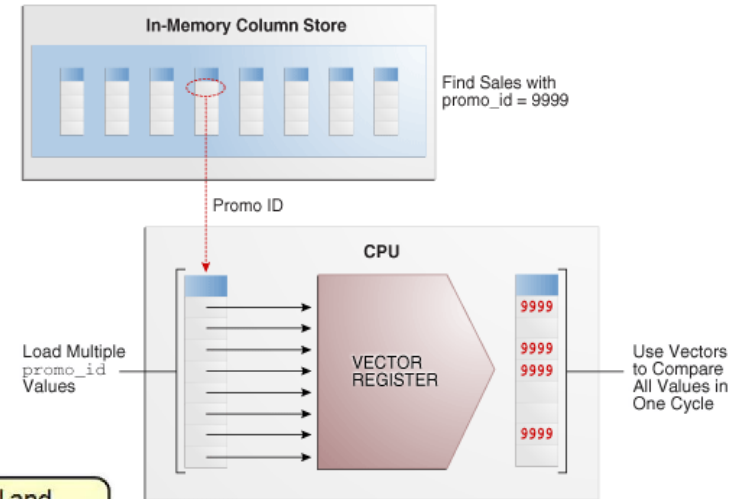
It is possible to selectively populate the IM column store with a chosen subset of the database

- In-memory storage is populated in contiguously allocated units called **In-Memory Compression Units**.
- It is possible to use different compression levels for different partitions within the same table. For example, a SALES table might have →
 - Current week partition - DML Compression
 - Earlier year partitions - Query Compression
 - Decade old partitions - Capacity Compression

MEMCOMPRESS	Description
DML	Minimal compression optimized for DML Performance
QUERY LOW/HIGH	Optimized for fastest query performance (Default)
CAPACITY LOW/HIGH	Optimized for space saving

Query Processing

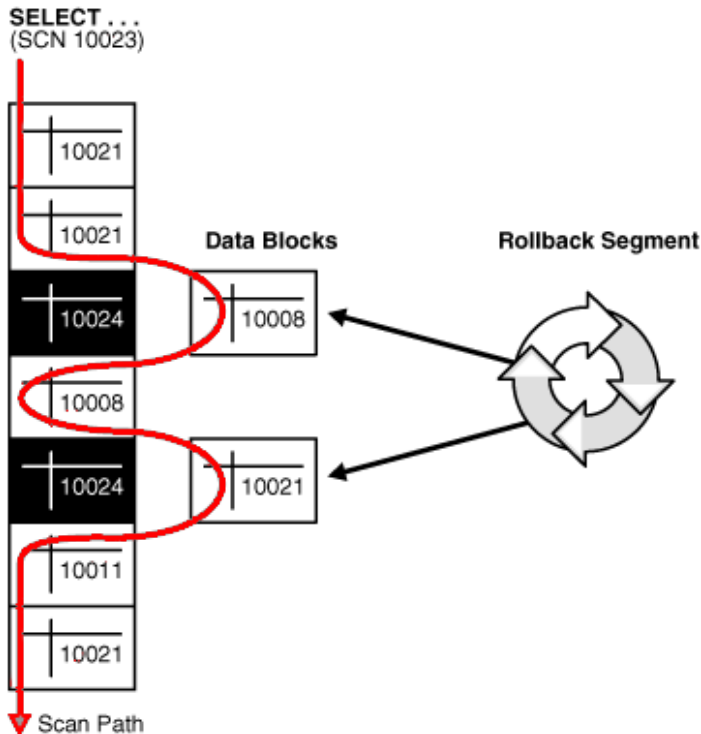
- In-Memory Scans utilize **SIMD** Vector Processing
- In-Memory Storage Indexes are automatically created and maintained on each of the columns.



These allow data pruning to occur based on the filter predicates supplied in a SQL statement.

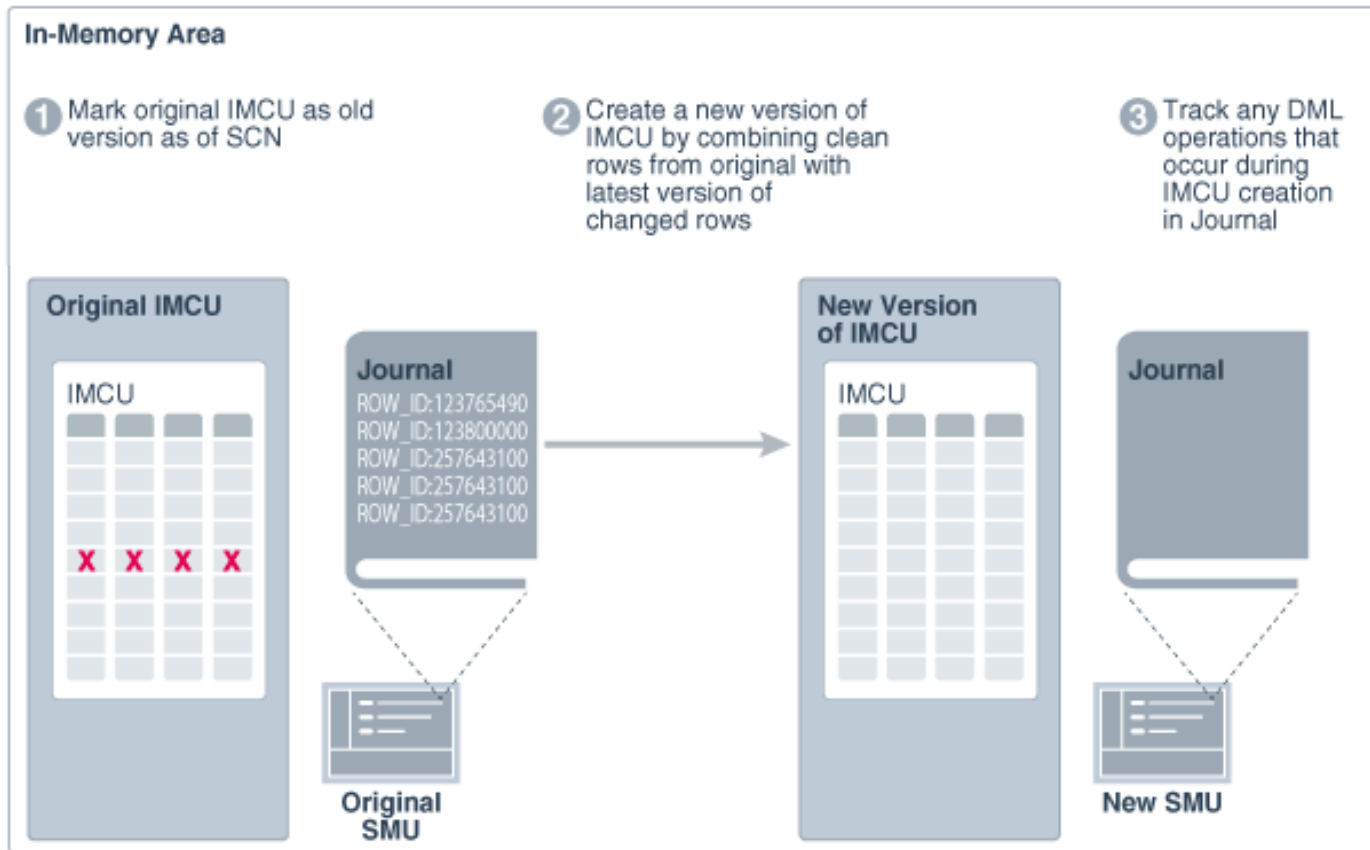
Transactional Consistency

The default isolation level of Oracle DB is “Consistent Read”



- As a query enters the execution stage, the current system change number (SCN) is determined.
- Each query returns all committed data with respect to the SCN recorded at the time that query execution began.
- The In-memory Column store follows similar semantics!

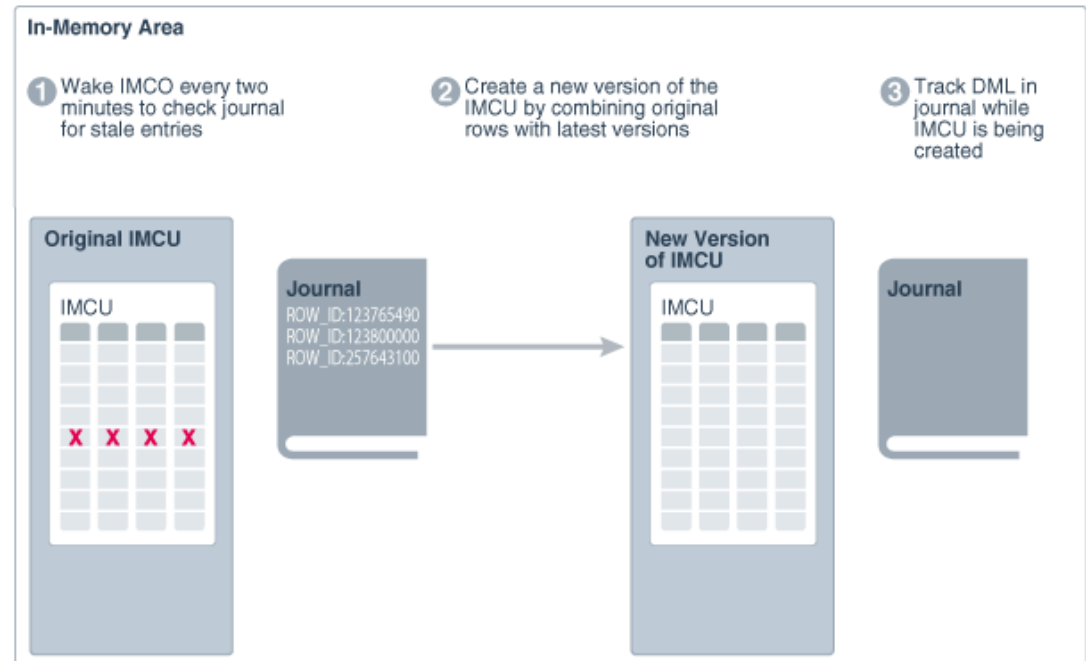
Transactional Consistency (Double Buffering)



Transactional Consistency (Repopulation Overhead)

There are two types of strategies for data repopulation:

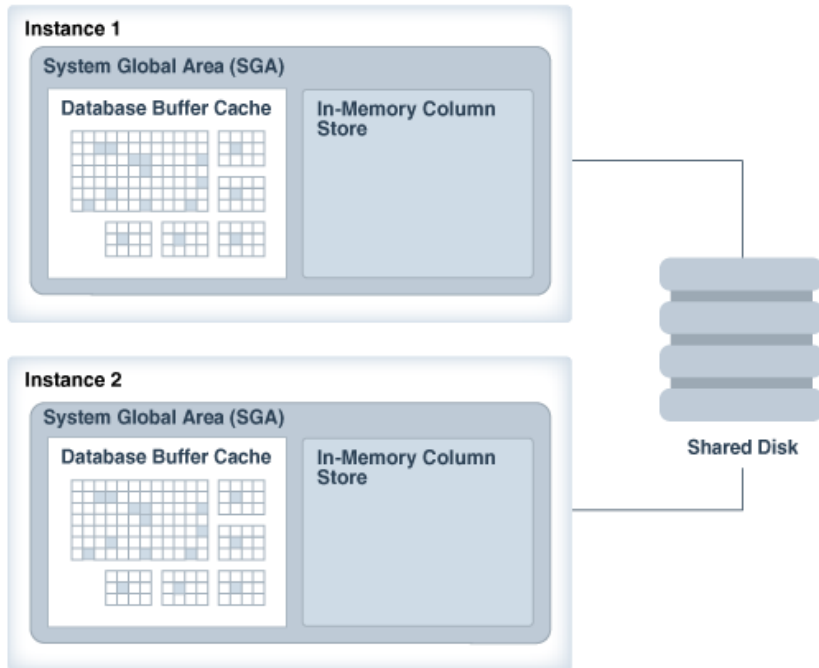
- **Threshold Repopulate**
- **Trickle Repopulate**



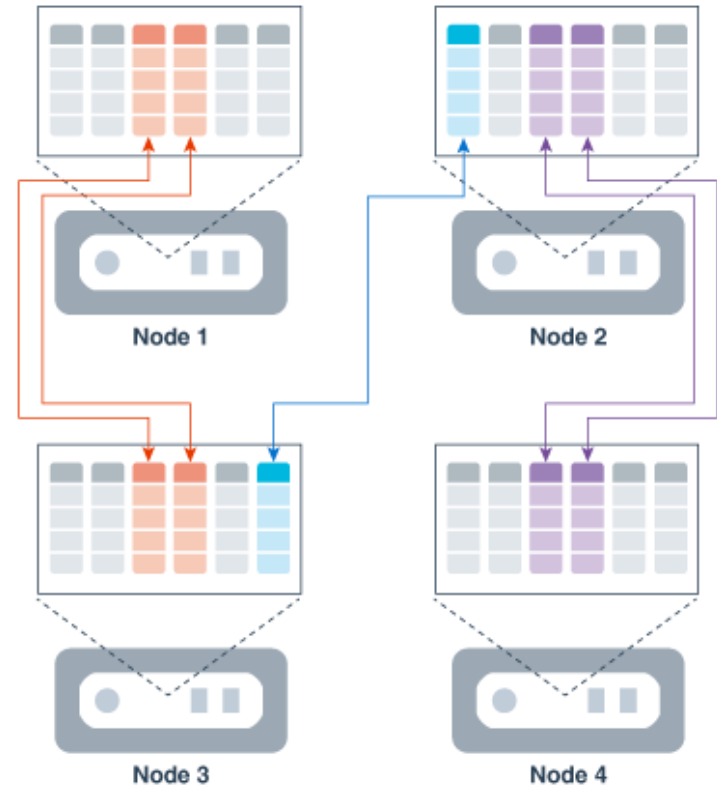
At most $\frac{1}{10}^{th}$ of a single CPU core is dedicated to trickle repopulate

In-Memory Column Store Scale Out

Distribution

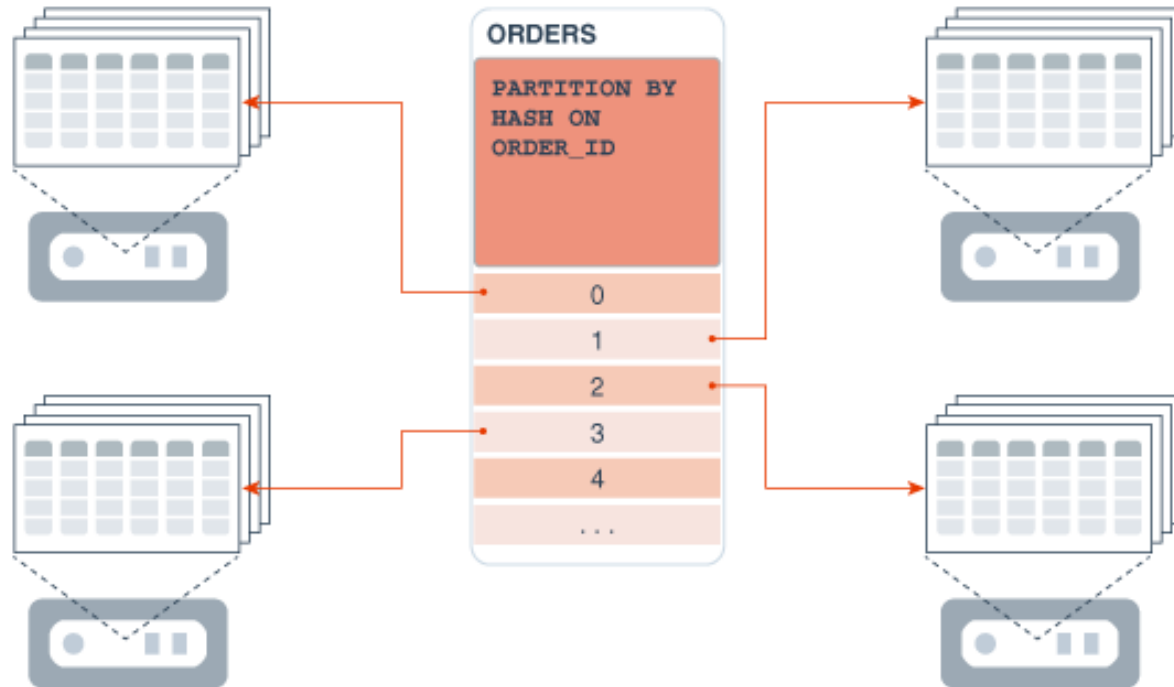


Duplication



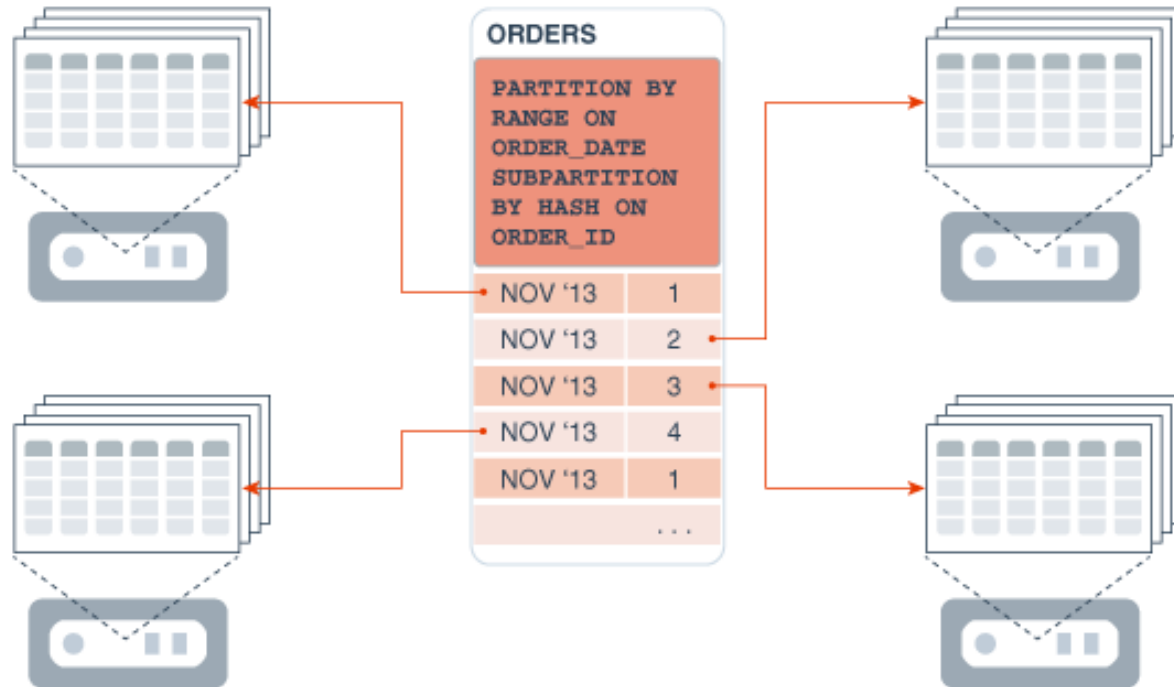
In-Memory Column Store Scale Out (Distribution)

Distributing Partitions by Hash



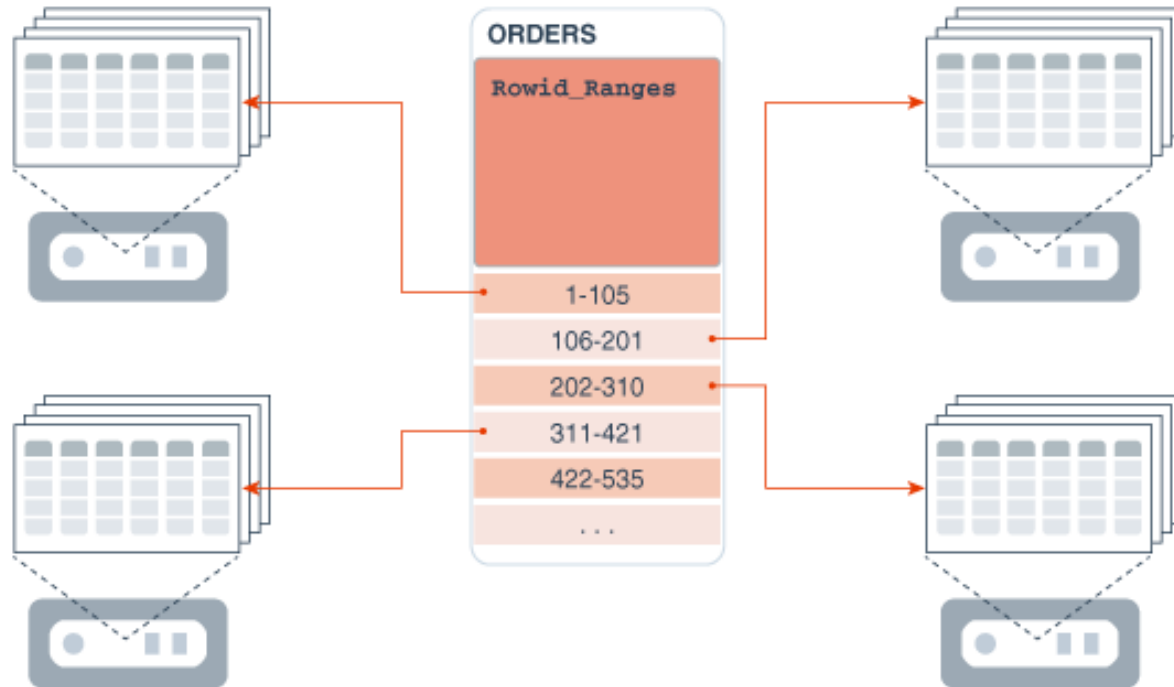
In-Memory Column Store Scale Out (Distribution)

Distributing Partitions by Range and Sub partitions by Hash



In-Memory Column Store Scale Out (Distribution)

Distribution by Rowid Range



Handling Node Failures

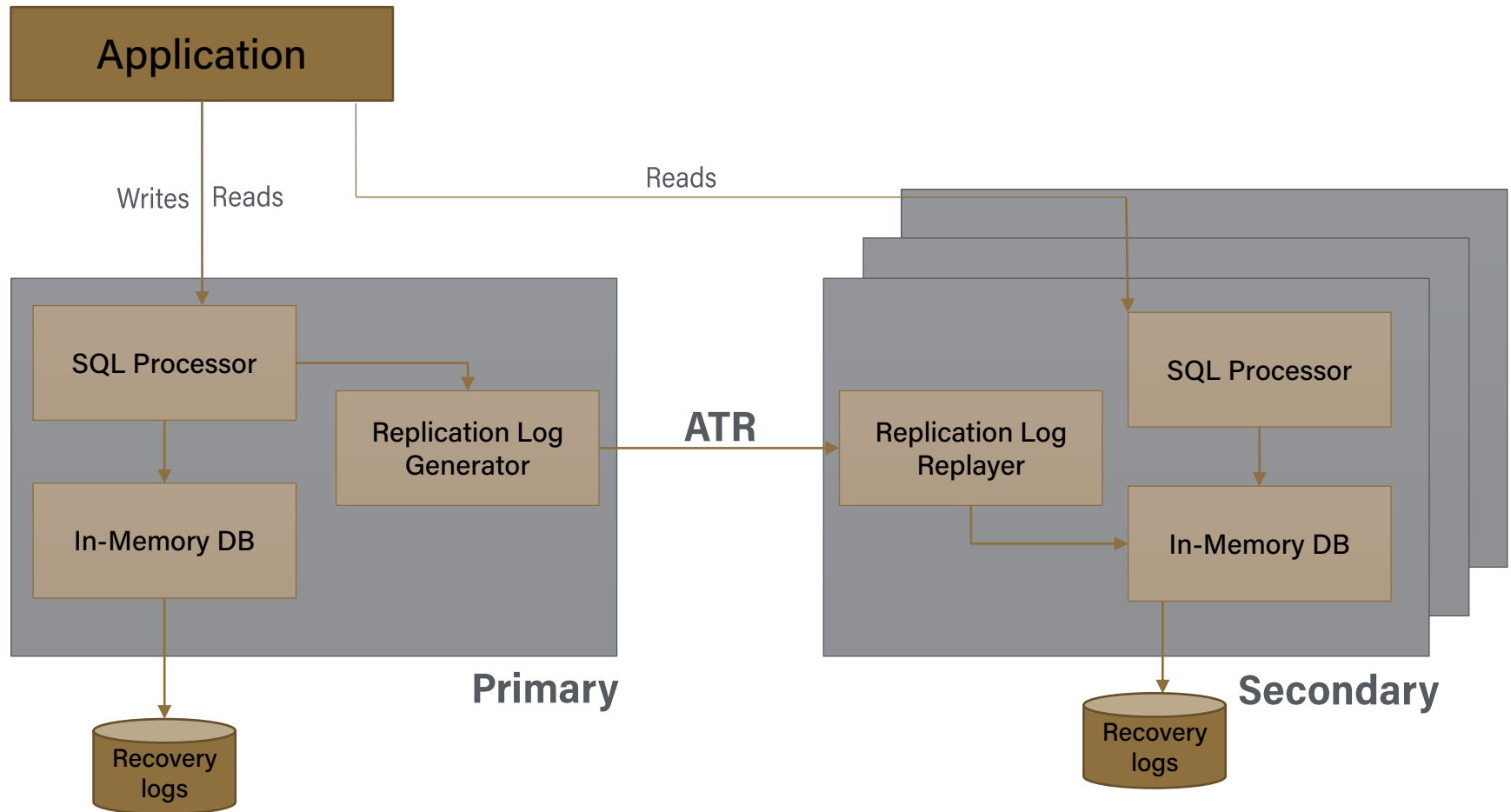
- If an Oracle RAC instance fails, then the IMCUs on the failed instance are unavailable.
- Consequently, a query that needs data stored in the inaccessible IMCUs must read it from somewhere else: the database buffer cache, flash storage, disk, or mirrored IMCUs in other IM column stores.
- Duplication helps to provide fault tolerance because if one node fails, then the mirrored columnar data is accessible from a different node.
- Worst case, queries issued against missing data do not fail. Instead, queries access the data either from the database buffer cache or permanent storage, which may just negatively affect performance.

SAP HANA

Disaggl

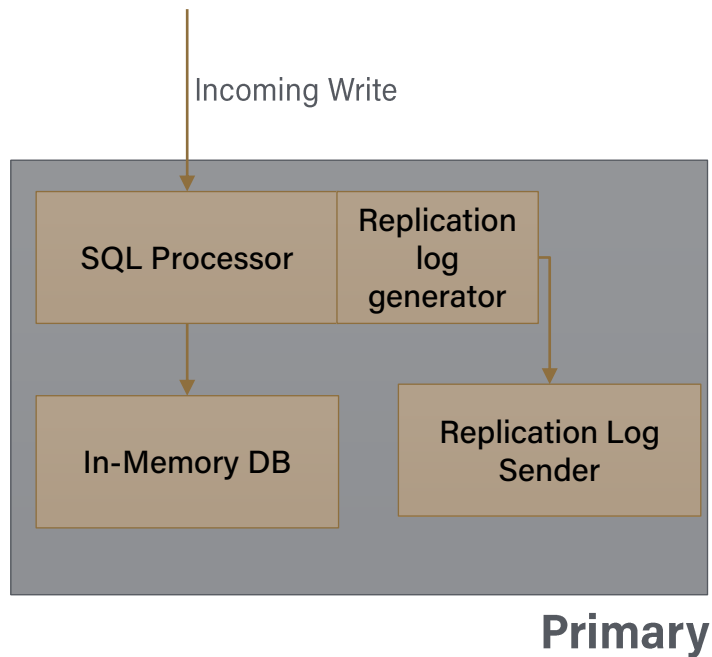
Presented by
Shubham Pandey

Architecture



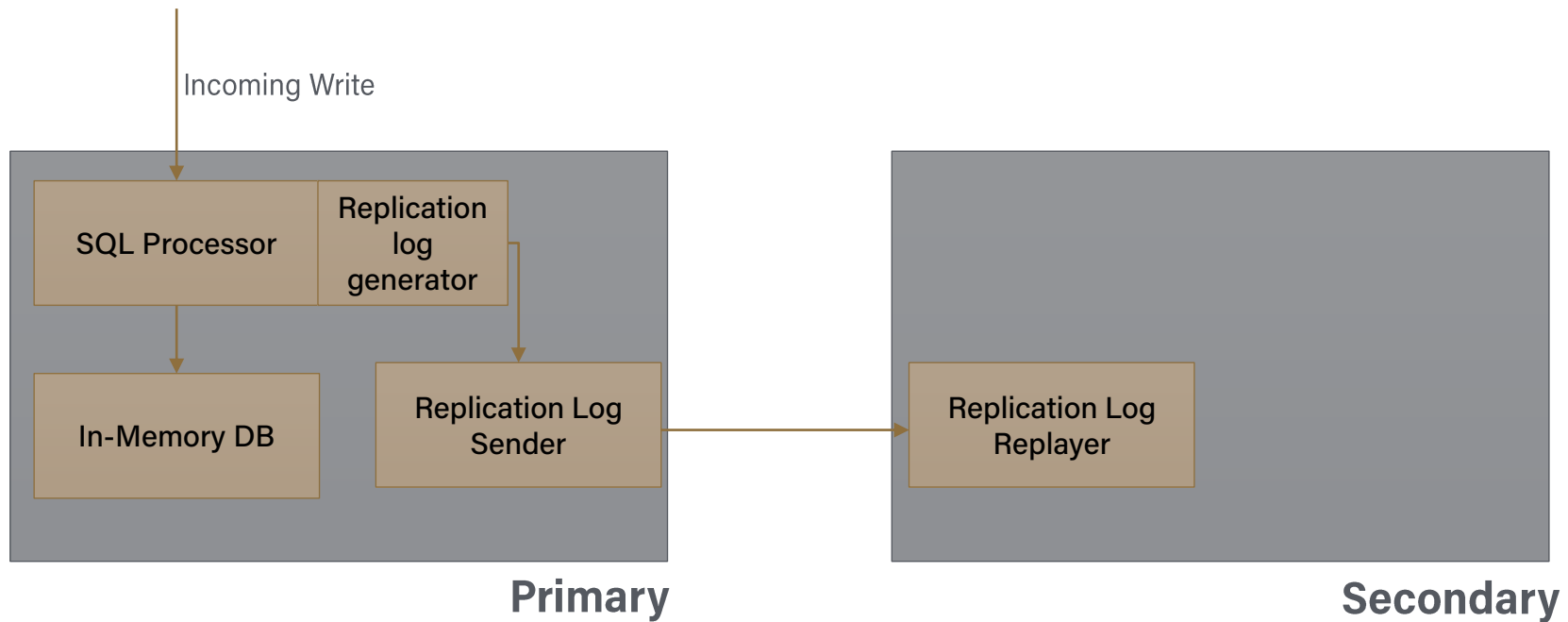
Design Decisions

In-Database Replication



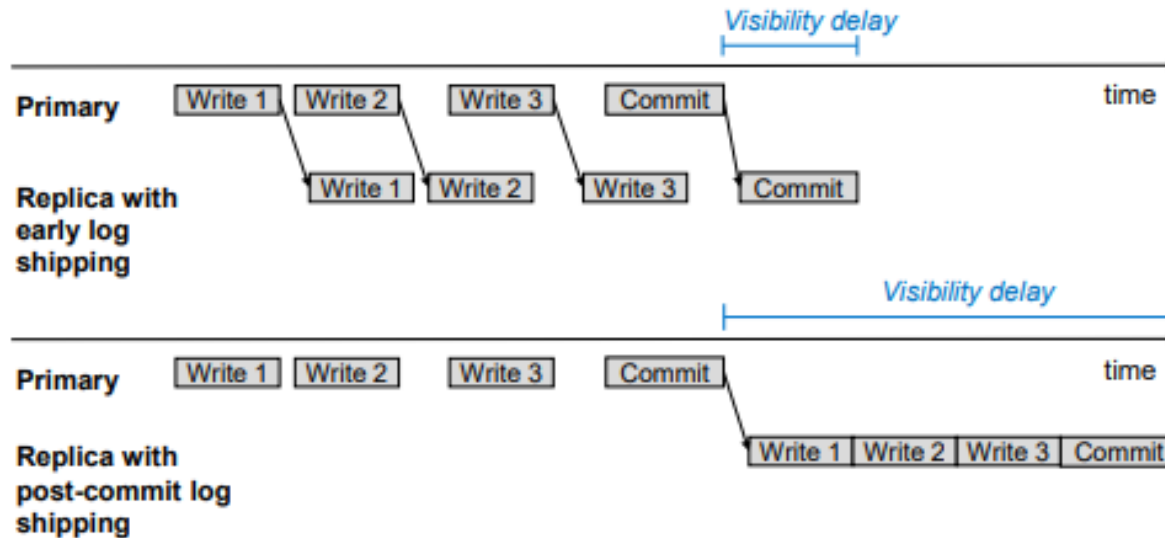
Design Decisions

Asynchronous Replication



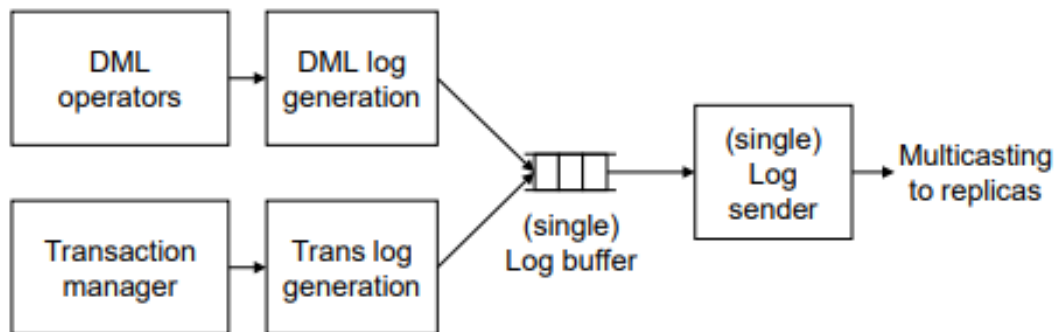
Design Decisions

Early Log Shipping



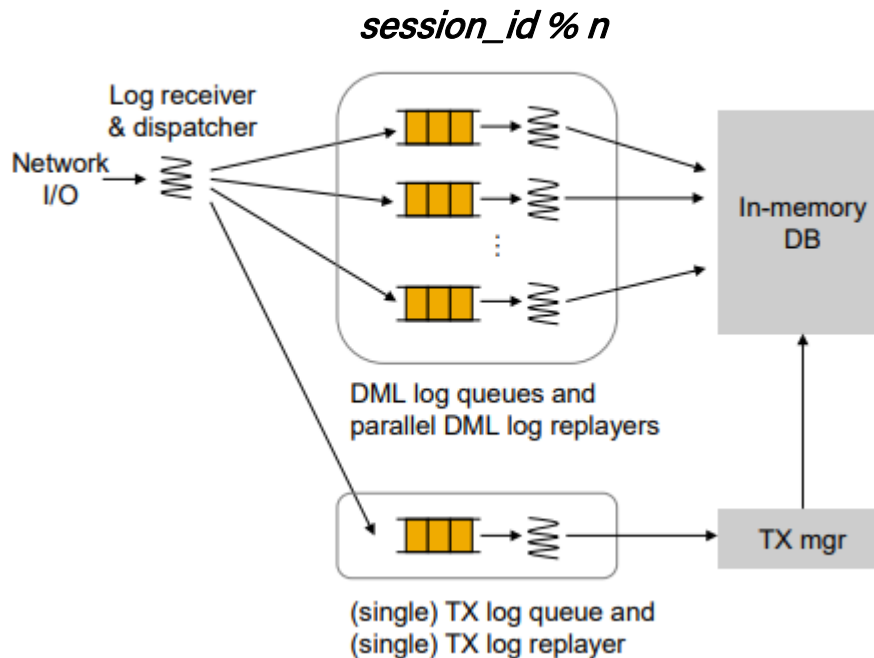
Design Decisions

Log generator and Sender

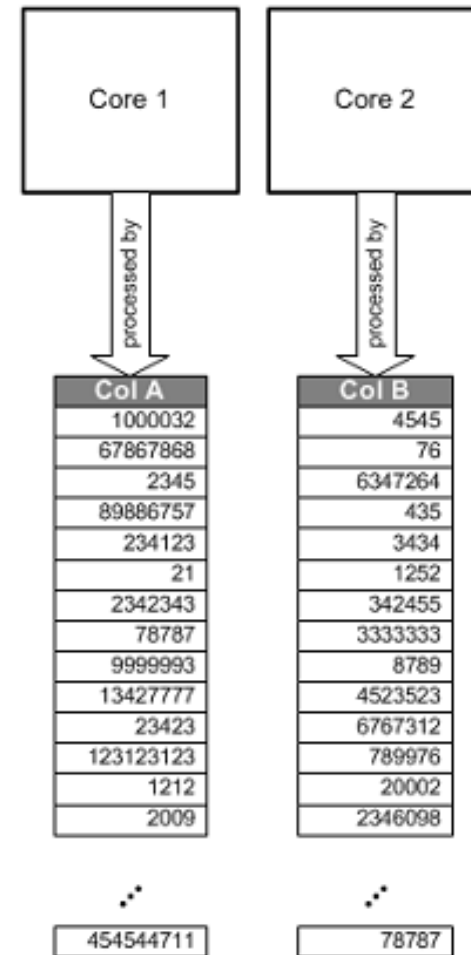
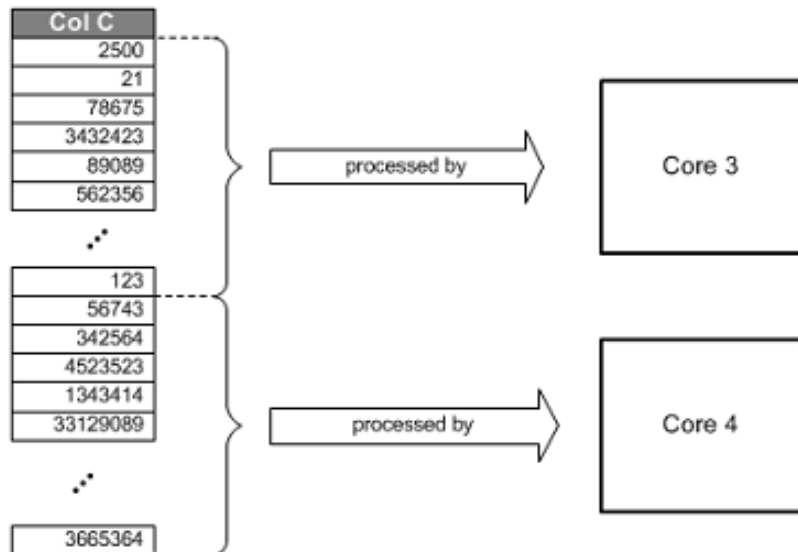


Design Decisions

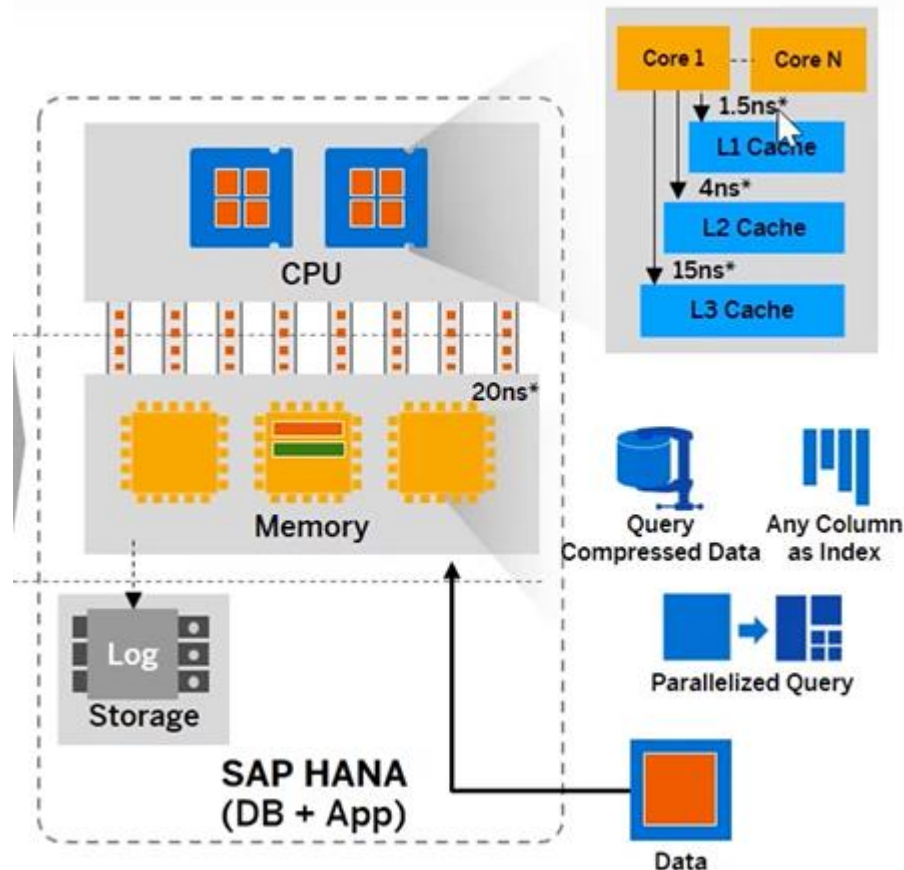
Parallel Log Replay



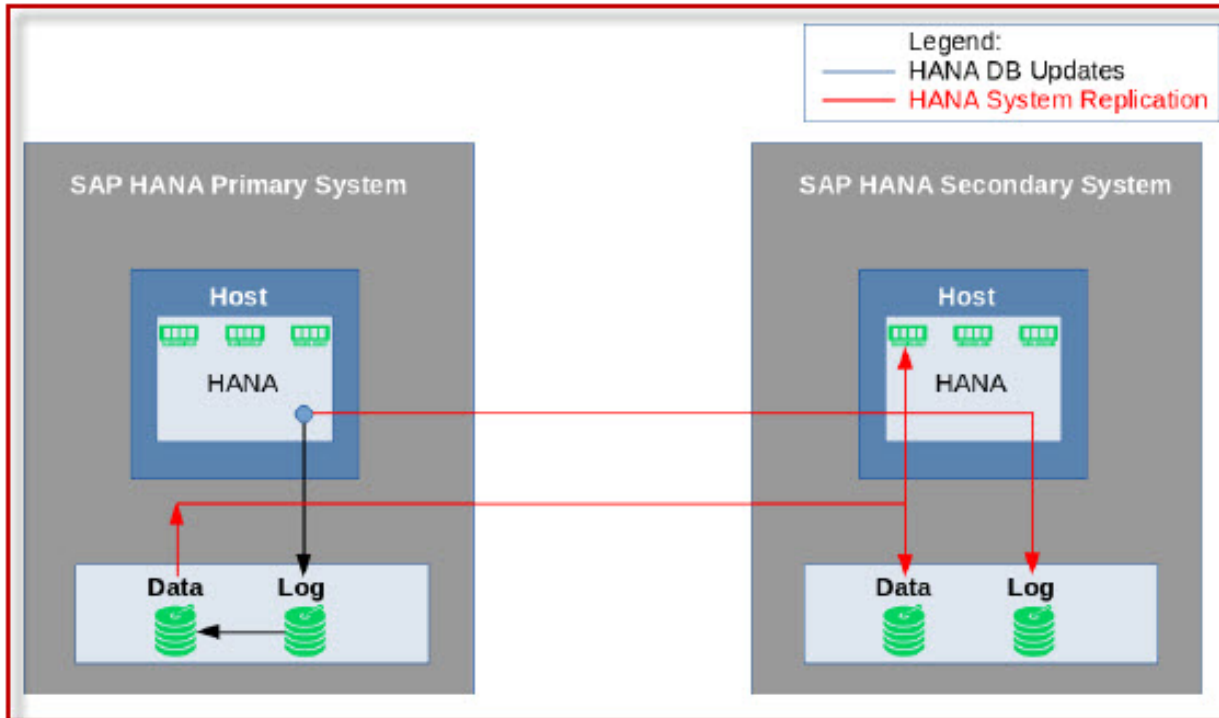
Parallel Processing



Exploiting Modern Hardware



Fault Tolerance



- Continuous synchronization of HANA DB to secondary location
- Secondary location can be in same data center or a different one

TiDB: A Raft Based HTAP Database

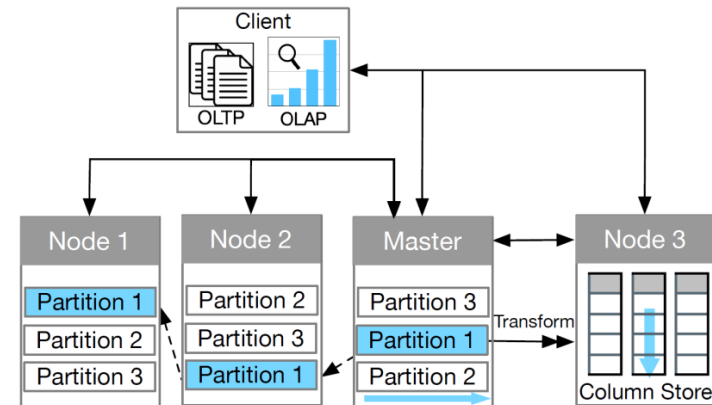
Disaggl

Presented by
Satya Sai Bharath Vemula

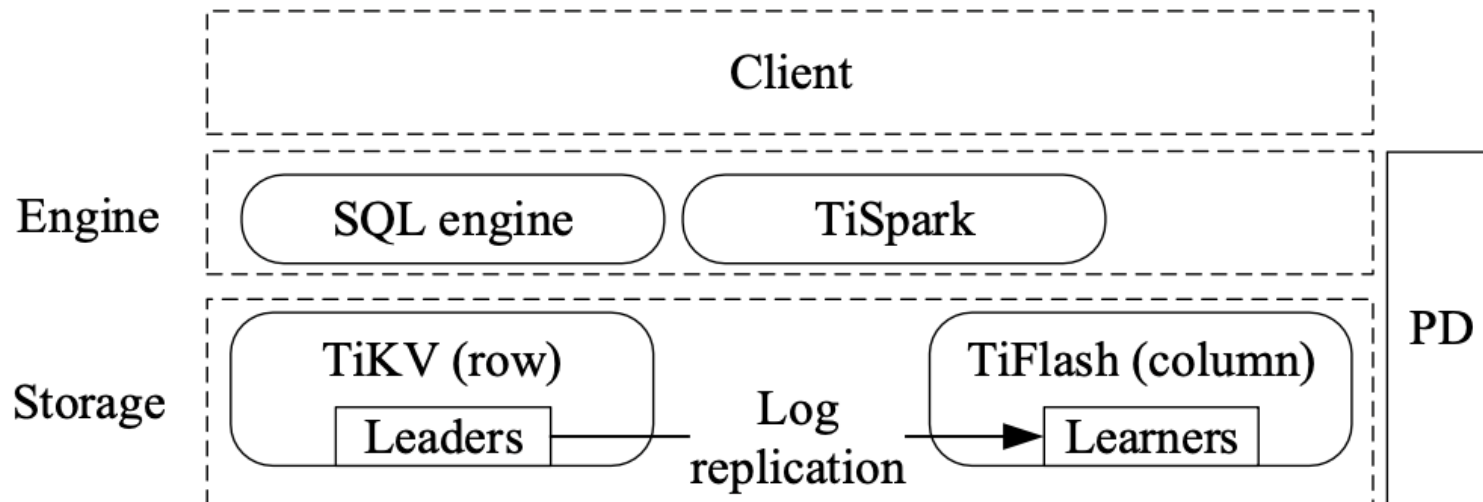
System Goals

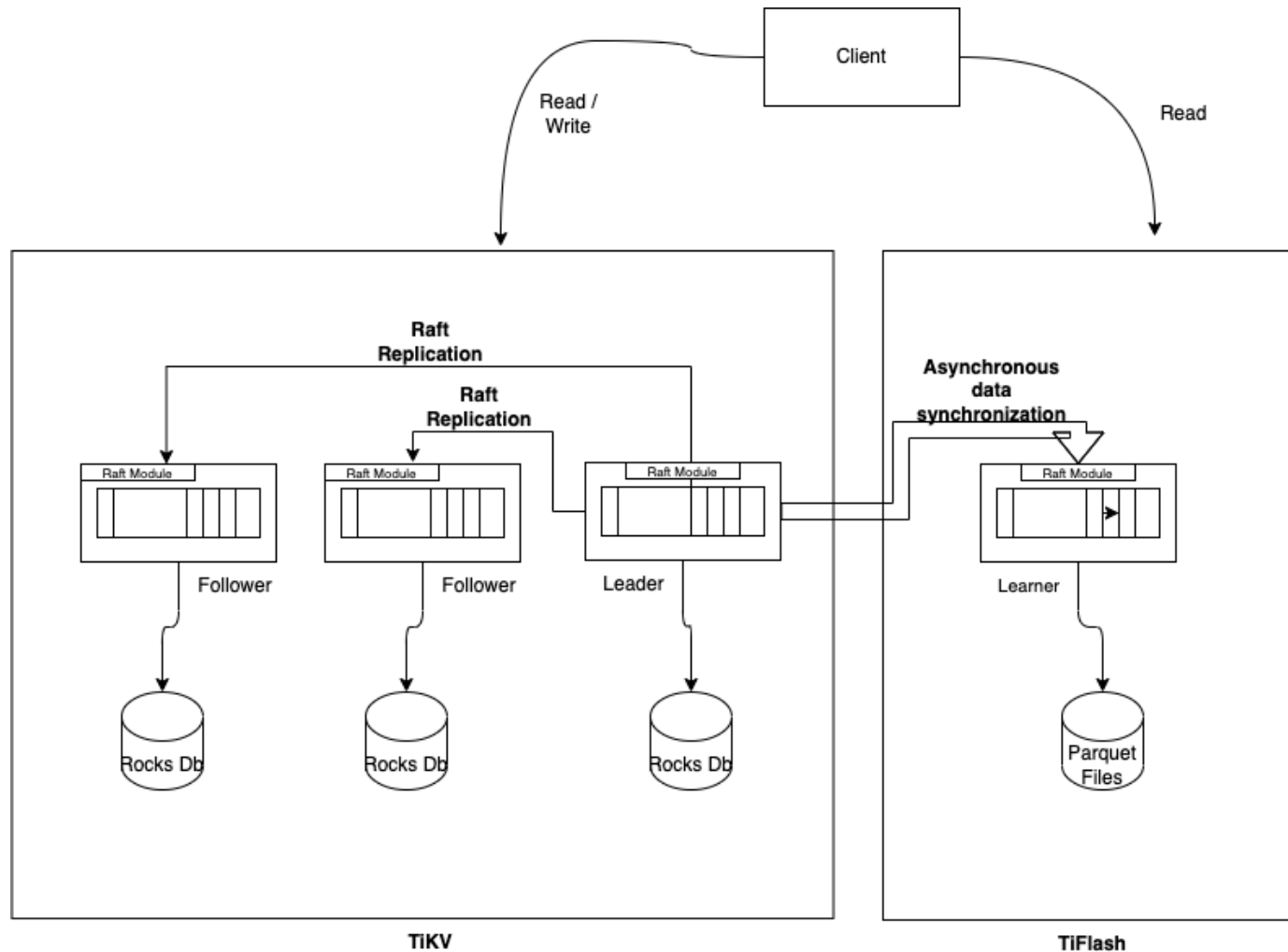
System on Higher Level

- Provide Isolation for OLTP and OLAP Queries
 - Providing separate data stores (Row-based and Column-based) for OLTP and OLAP
- Providing Consistent View of Data / Real time data synchronization between the two data stores by extending using state machine-based consensus algorithm (*RAFT*)
- Highly Scalable and Available System with efficient query processing

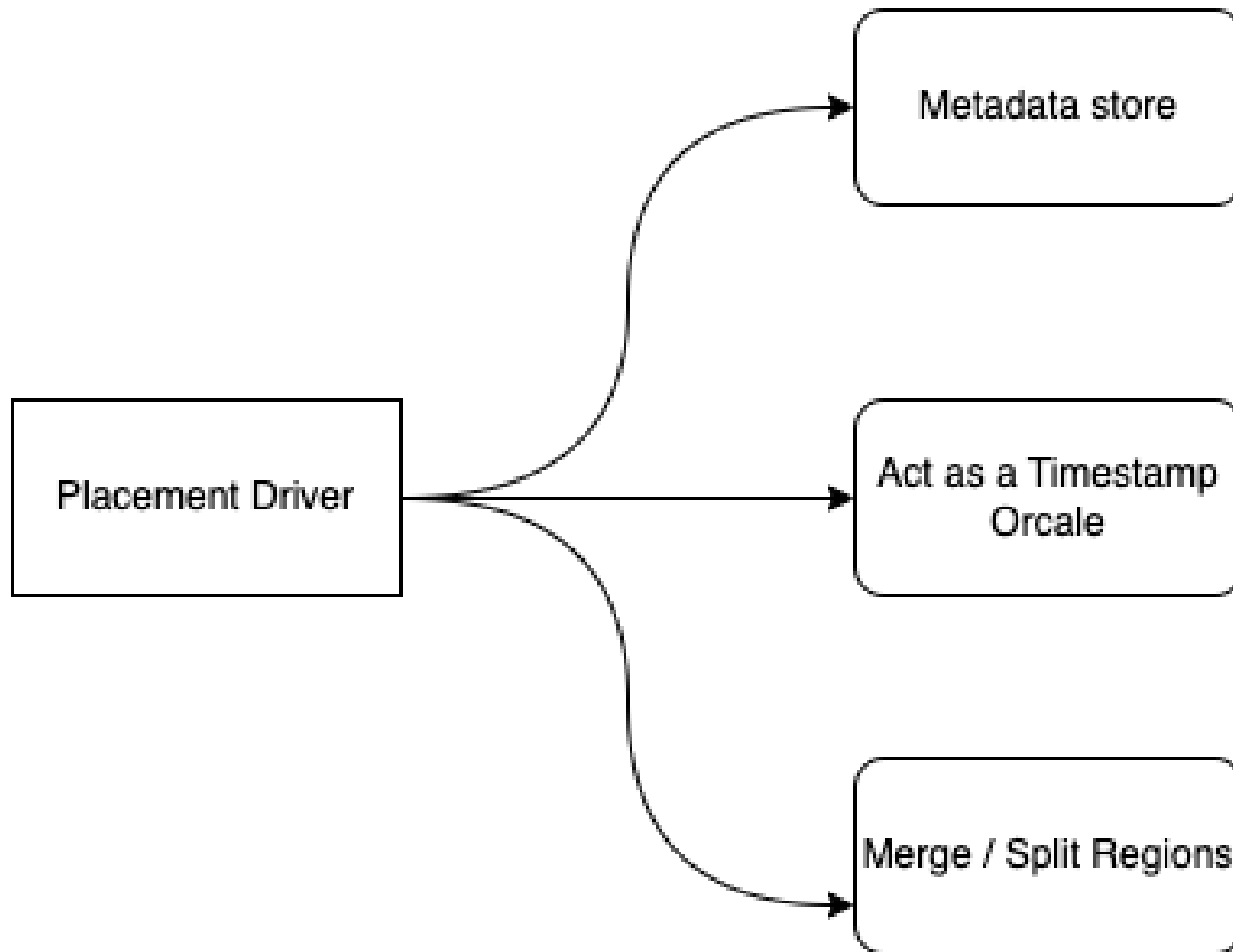


TiDB Architecture

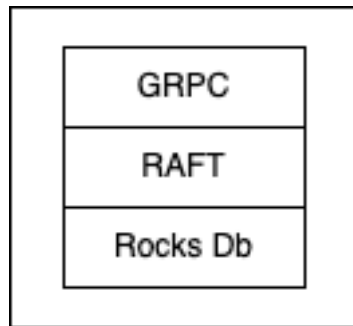




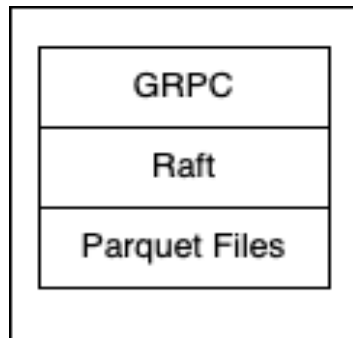
Placement Driver



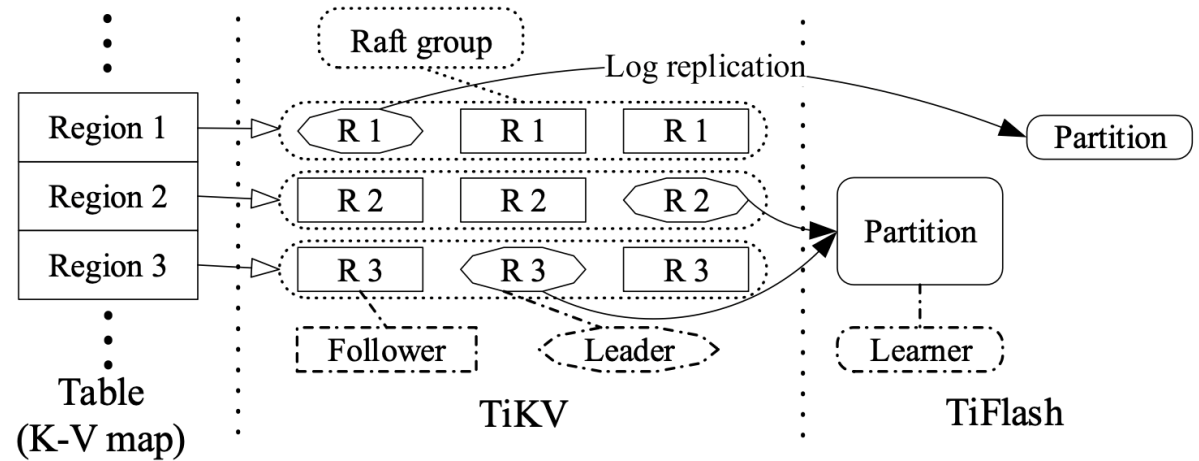
TiDB Scaling



Tikv Node



TiFlash Node



Availability / Failure Tolerance

Failure Tolerance in both TiKV and TiFlash is maintained through replicas

- For each TiKV region if there are $2*m + 1$ nodes there will be no observed down time if $m + 1$ nodes are functional in the same network partition.
 - This Property is because of RAFT Availability Semantics
 - OLTP queries for a region will not be functional if more than $m + 1$ nodes in a region have failed.
- TiFlash supports multiple replicas for the same data
 - If there are N replicas for a single TiFlash Node, The system can bear a failure of $N - 1$ replicas
 - OLAP queries can still be functional if all the TiFlash Nodes are down as the system can also read data from TiKV nodes for serving analytical queries.

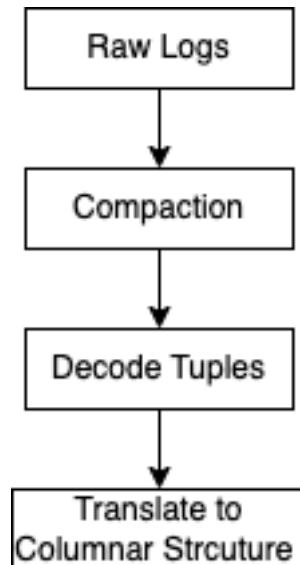
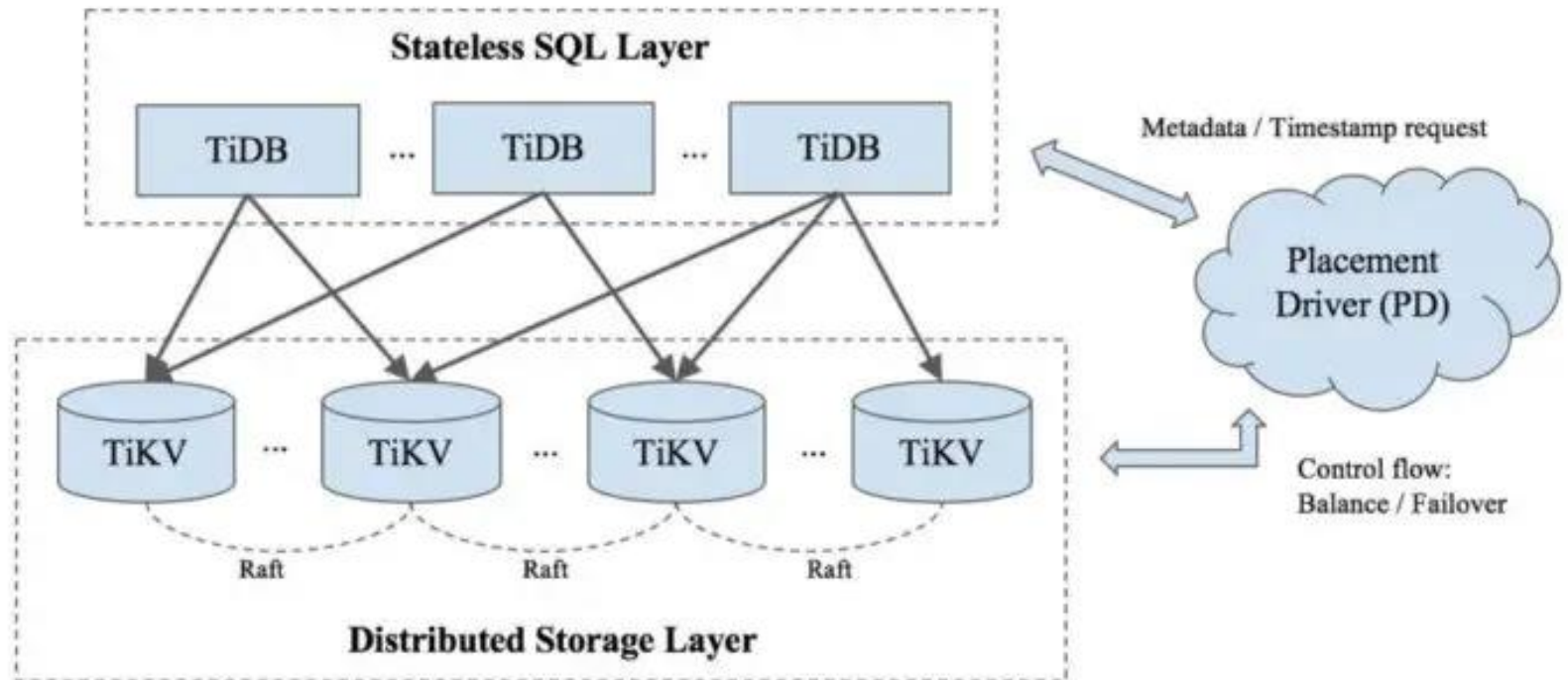


Table 1: Log replaying and decoding

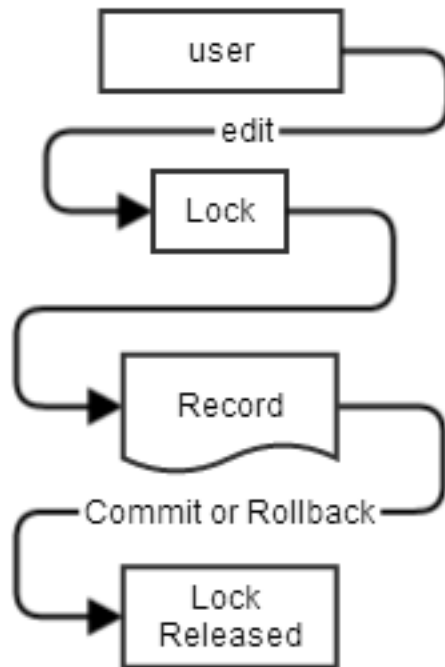
Raw logs	$\{1\}\{\text{insert}\}\{\text{prewritten@1}\}\{k1 \rightarrow (a1, b1)\}$ $\{2\}\{\text{insert}\}\{\text{prewritten@2}\}\{k2 \rightarrow (a2, b2)\}$ $\{3\}\{\text{update}\}\{\text{prewritten@3}\}\{k3 \rightarrow (a3, b3)\}$ $\{1\}\{\text{insert}\}\{\text{rollbacked@1}\}$ $\{2\}\{\text{insert}\}\{\text{committed\#4}\}$ $\{3\}\{\text{update}\}\{\text{committed\#5}\}$ $\{4\}\{\text{delete}\}\{\text{prewritten@6}\}\{k4\}$ $\{4\}\{\text{delete}\}\{\text{committed\#7}\}$
Compacted logs	$\{2\}\{\text{insert}\}\{\text{prewritten@2}\}\{k2 \rightarrow (a2, b2)\}$ $\{3\}\{\text{update}\}\{\text{prewritten@3}\}\{k3 \rightarrow (a3, b3)\}$ $\{2\}\{\text{insert}\}\{\text{committed\#4}\}$ $\{3\}\{\text{update}\}\{\text{committed\#5}\}$ $\{4\}\{\text{delete}\}\{\text{prewritten@6}\}\{k4\}$ $\{4\}\{\text{delete}\}\{\text{committed\#7}\}$
Decoded tuples	$\{\text{insert}\}\{\#4\}\{k2 \rightarrow (a2, b2)\}$ $\{\text{update}\}\{\#5\}\{k3 \rightarrow (a3, b3)\}$ $\{\text{delete}\}\{\#7\}\{k4\}$
Columnar data	$\{\text{insert,update,delete,}\}$ $\{\#4,\#5,\#7,\}$ $\{k2,k3,k4,\}$ $\{a2,a3,,\}$ $\{b2,b3,,\}$

OLTP Query Processing

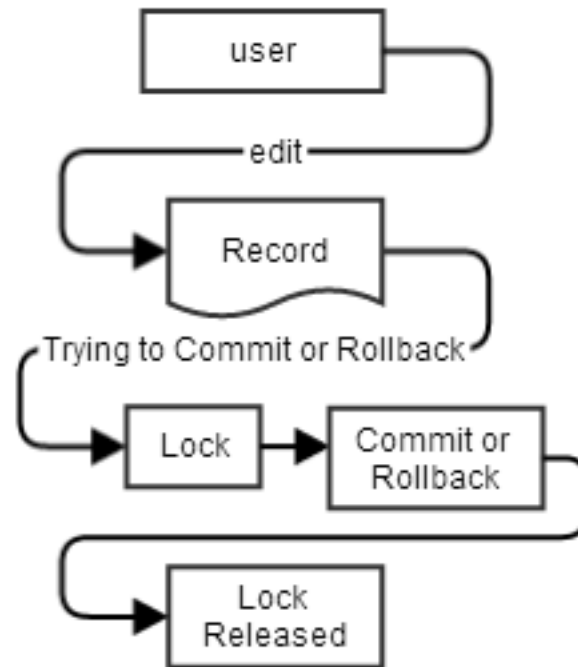


OLTP Distributed Transaction

www.adfjavacodes.blogspot.in

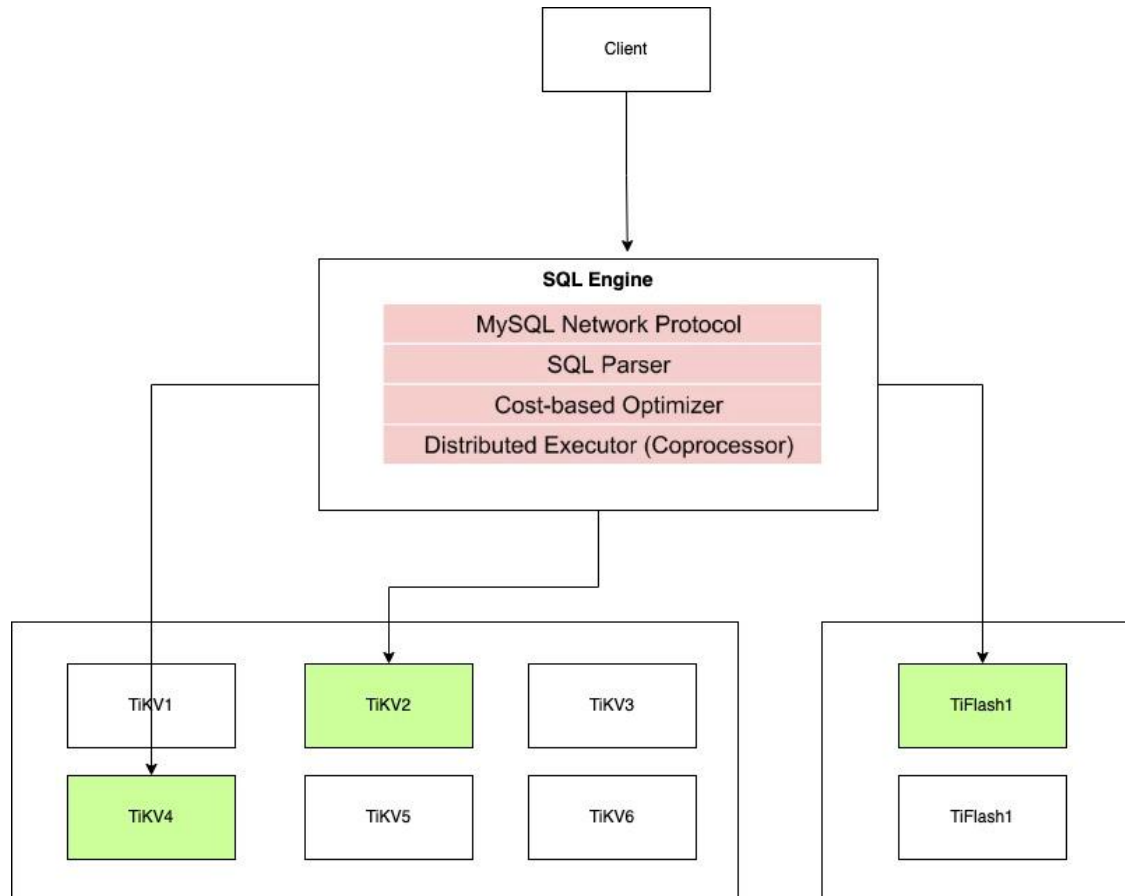


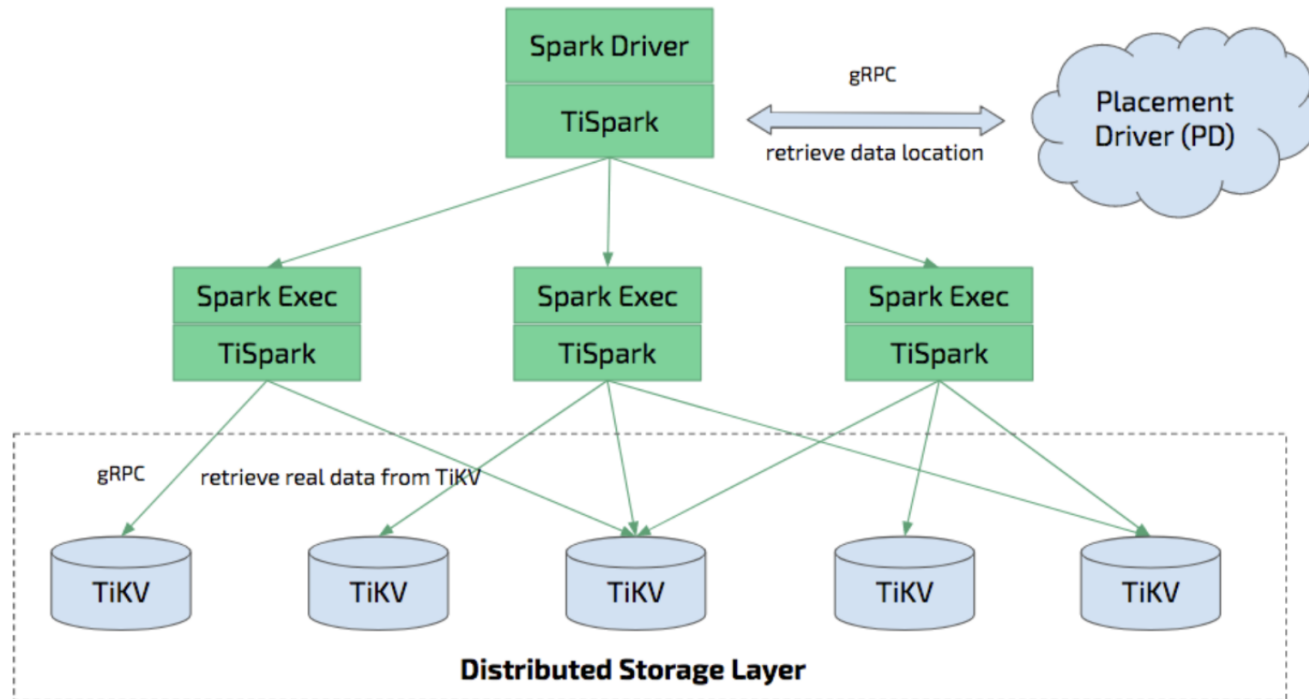
Pessimistic Locking



Optimistic Locking

OLAP Query Processing





- **Study HTAP systems with diverse architectures (e.g. data organisation and synchronization)**
 - Compare and Contrast such systems based on their design decisions
 - Assess the strengths and weakness of these systems (in terms of AP/TP throughput, scalability, among others)
- **Study the systems that have evolved from OLTP to HTAP (e.g. MemSQL, IBM dashDB, and more)**

APPENDIX

DETAILED DESCRIPTION

Replay Algorithms - DML log entry

Require: A DML log entry L .

```

1: Find the transaction object  $T$  for  $L.TransactionID$ .
2: if  $T$  is empty then
3:   Create a transaction object for  $L.TransactionID$ .
4: end if
5: if  $L.OperationType = Insert$  then
6:   Insert  $L.Data$  into the table  $L.\tau$ .
7:   Set the inserted record's  $RVID$  as  $L.\alpha$ .
8: else if  $L.OperationType = Delete$  then
9:   while true do
10:    Find the record version  $R$  whose  $RVID$  equals
11:    to  $L.\beta$  in the table  $L.\tau$ .
12:    if  $R$  is not empty then
13:      Delete  $R$ . return
14:    end if
15:  end while
16: else if  $L.OperationType = Update$  then
17:   while true do
18:    Find the record version  $R$  whose  $RVID$  equals
19:    to  $L.\beta$  in the table  $L.\tau$ .
20:    if  $R$  is not empty then
21:      Update  $R$  with  $L.Data$  and  $L.\alpha$ . return
22:    end if
23:  end while
24: end if

```

τ – Table ID

β – Before update $RVID$

α – After update $RVID$

Replay Algorithms

Replay a precommit log entry

Require: A precommit log entry L .

- 1: Find the transaction object T for $L.TransactionID$.
- 2: Mark T 's state as *precommitted*.

Replay an abort log entry

Require: An abort log entry L .

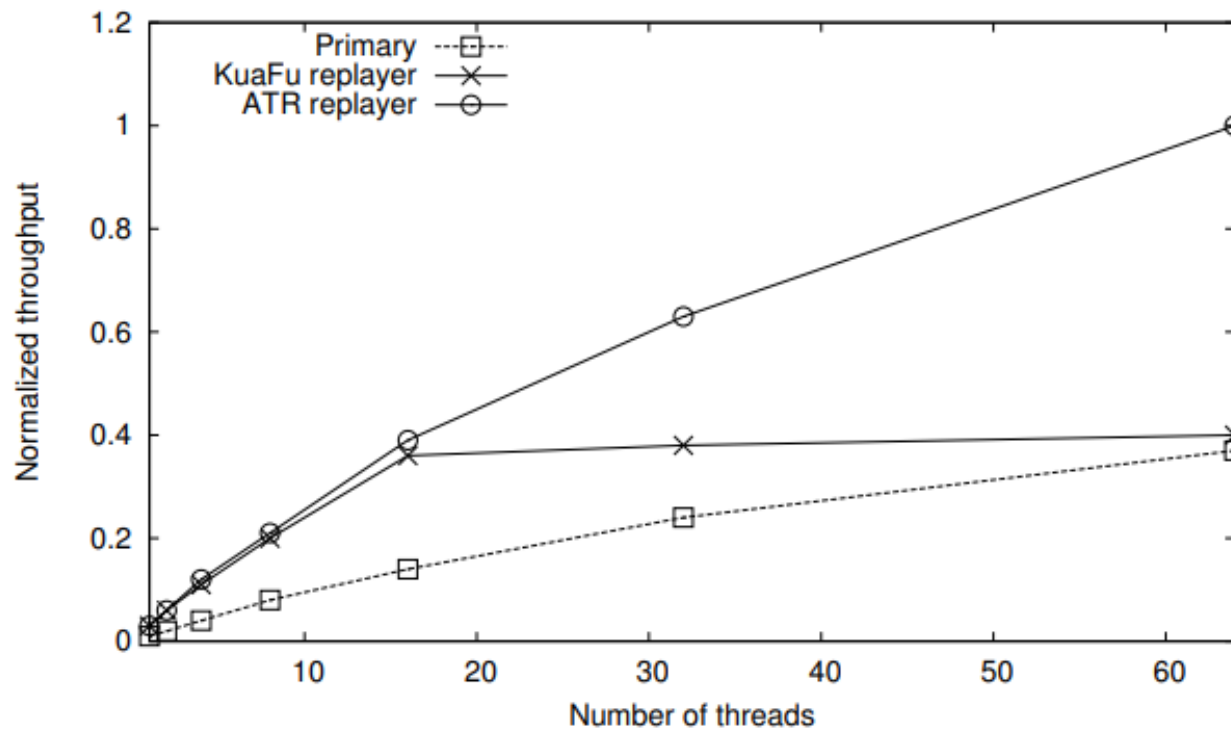
- 1: Find the transaction object T for $L.TransactionID$.
- 2: Abort T .

Replay a commit log entry

Require: A commit log entry L .

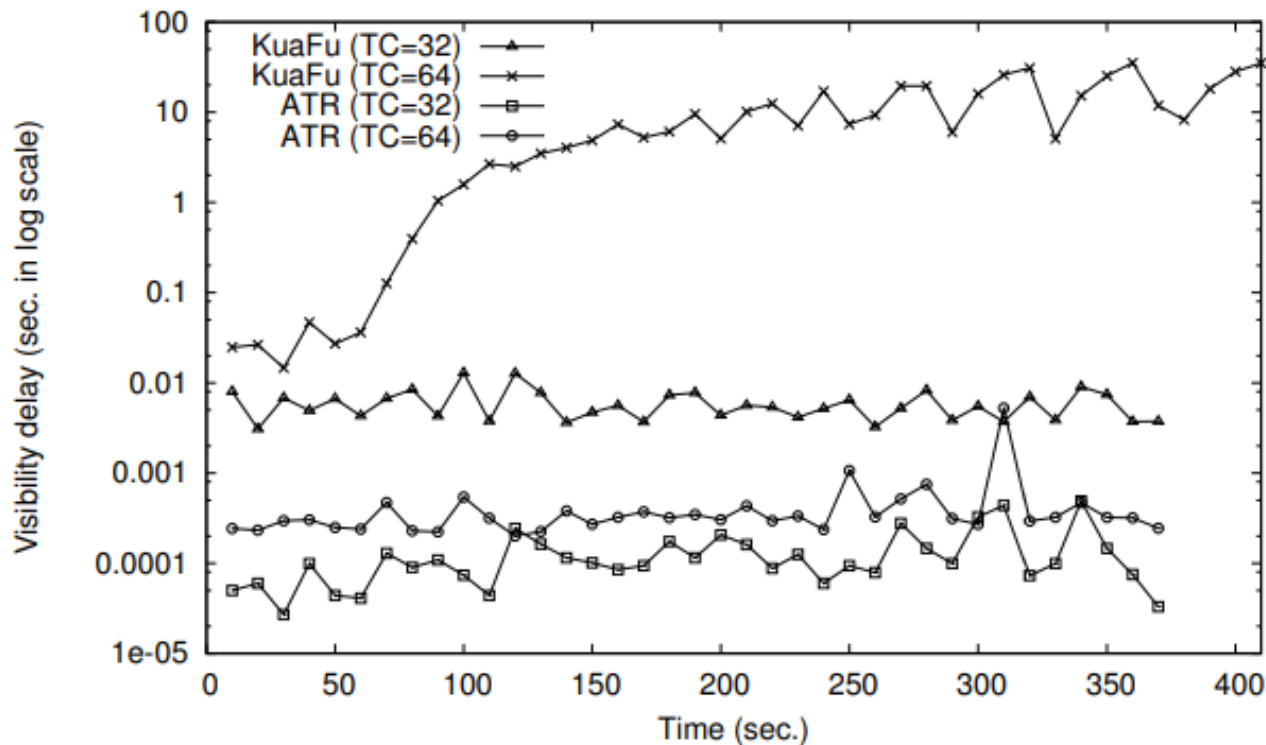
- 1: Find the transaction object T for $L.TransactionID$.
- 2: Wait until T 's state becomes *precommitted*.
- 3: Increment the transaction commit timestamp of the replica server by marking the T 's generated record versions with a new commit timestamp value.

Multi-core Scalability with Parallel Log Replay



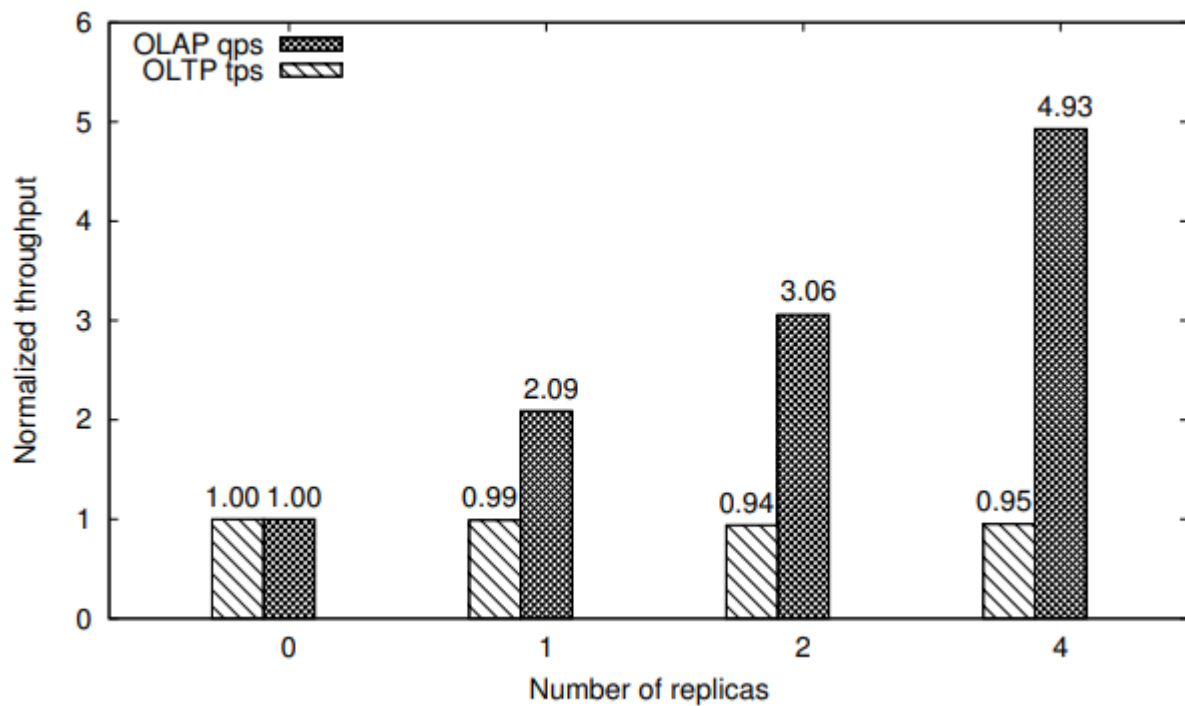
TPC-C benchmark

Visibility Delay

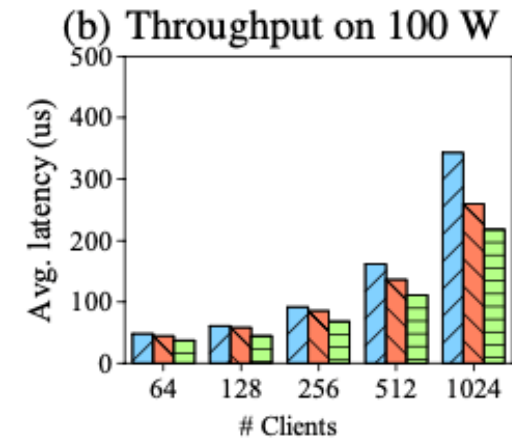
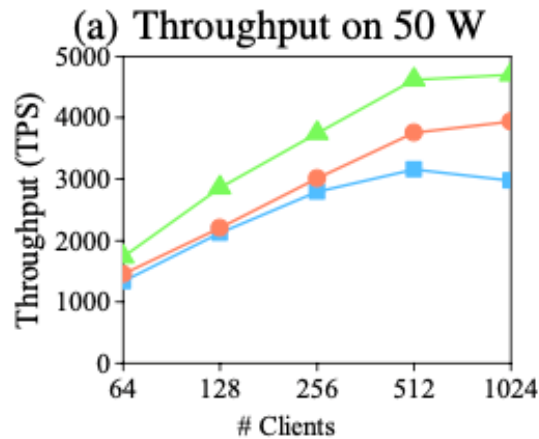
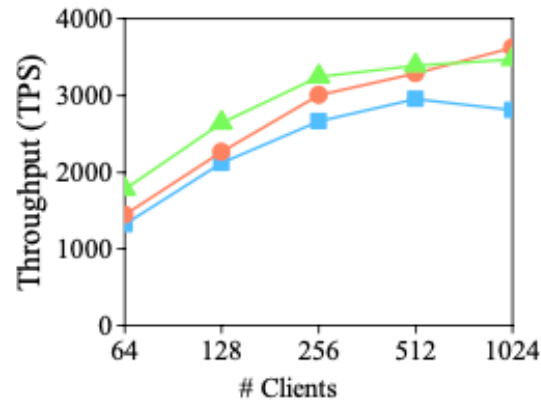
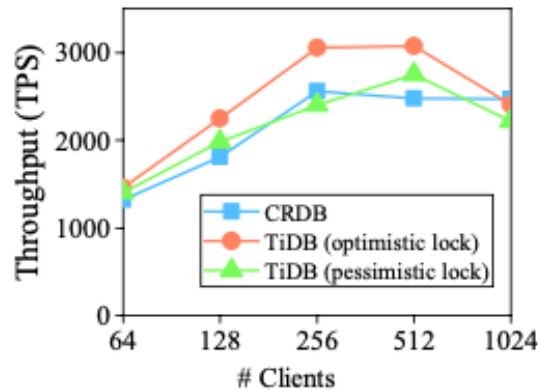


Visibility delay at replica while running TPC-C benchmark at primary

Multi Replica scalability under mixed (OLTP+OLAP) workload



TiDB OLTP Performance Comparison



(c) Throughput on 200 W

(d) Latency on 200 W

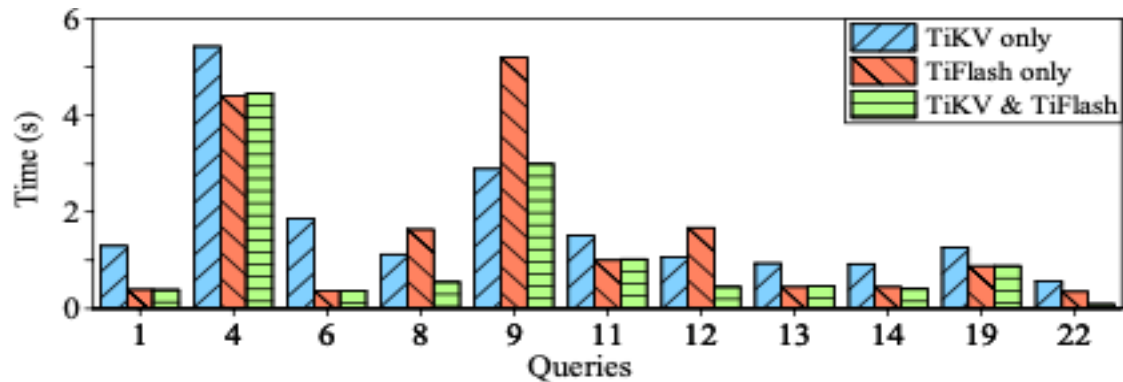


Figure 8: Choice of TiKV or TiFlash for analytical queries

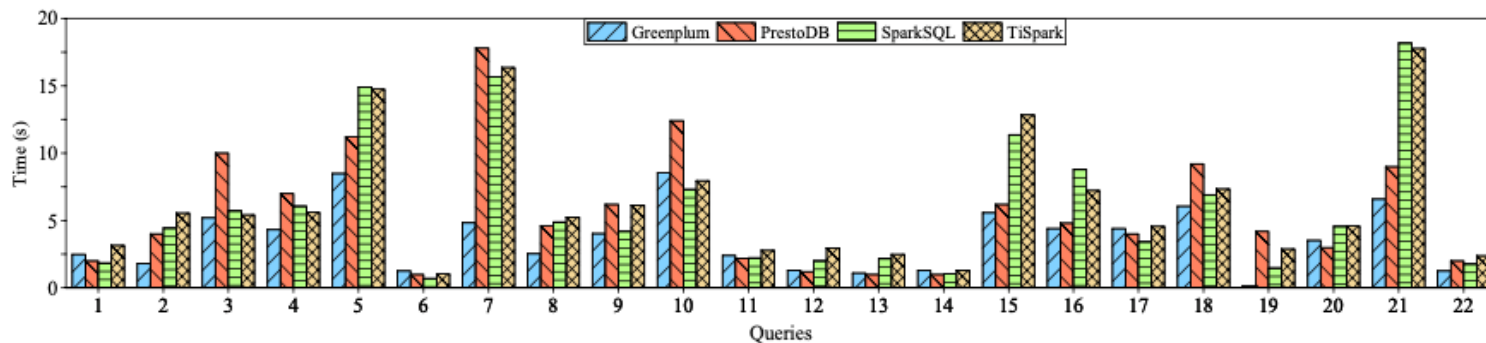


Figure 9: Performance comparison of CH-benCHmark analytical queries

HTAP Performance Results

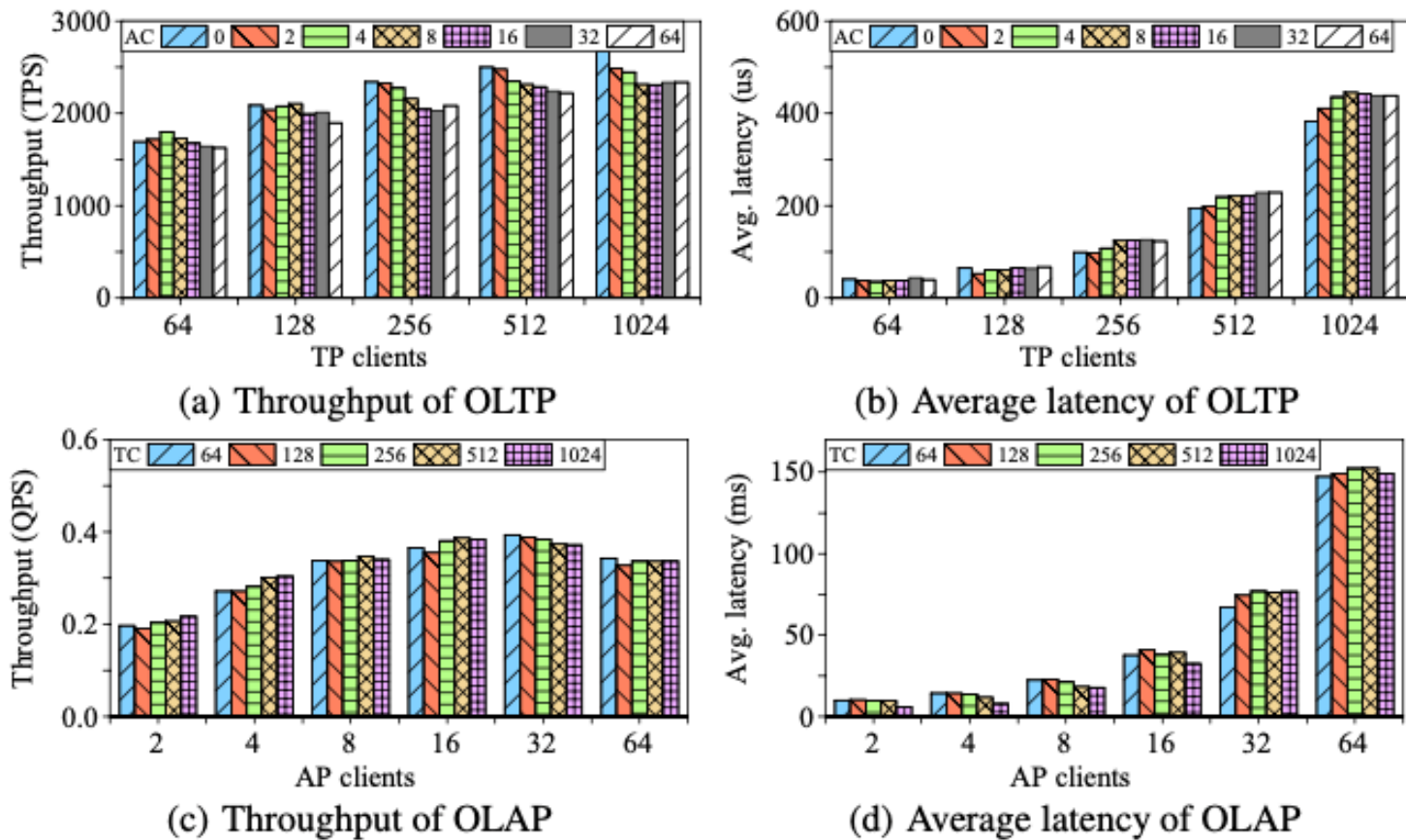


Figure 10: HTAP performance of TiDB

Raft Learner

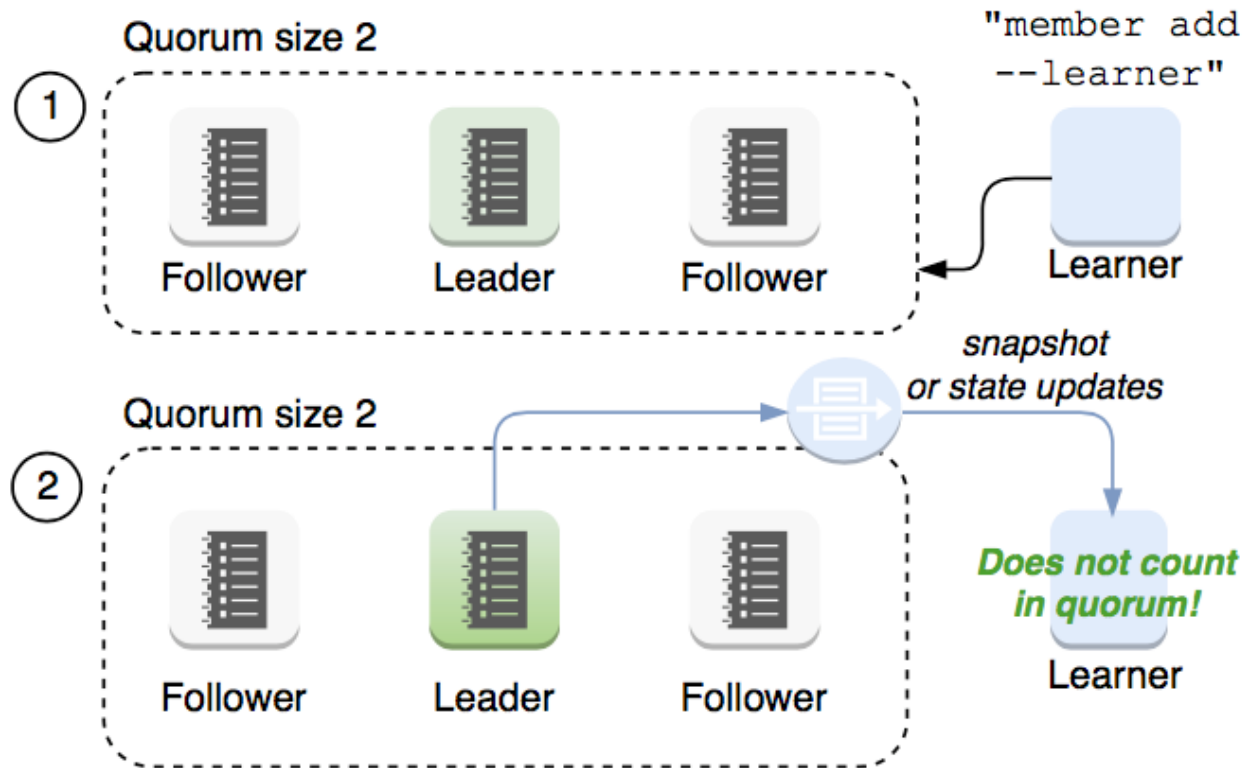
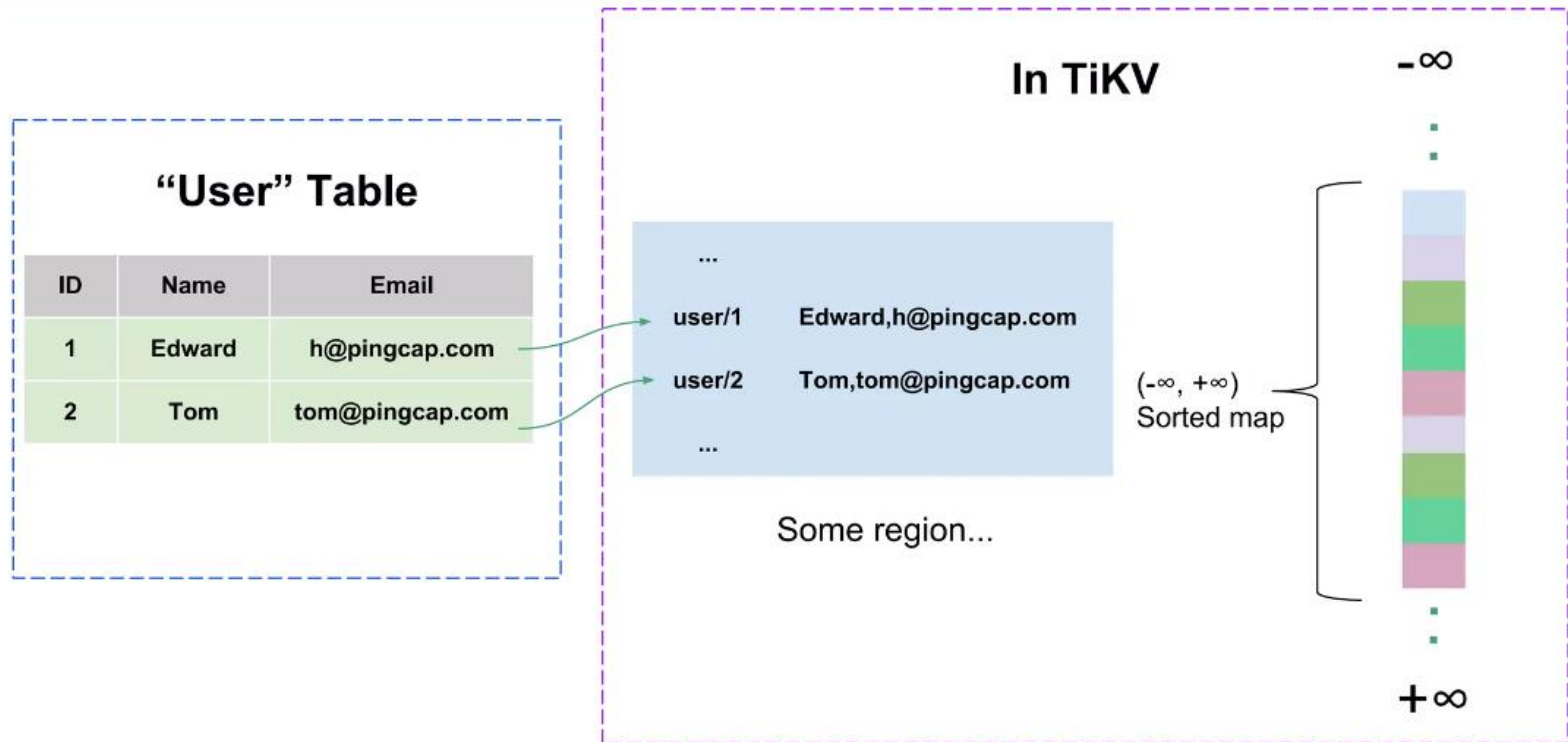


Figure 10. Add a learner node as a non-voting member. Wait until learner node catches up to leader's logs. Until then, learner node neither votes nor counts towards quorum.

TiKV Table Storage



Appendix (OLAP Optimization)

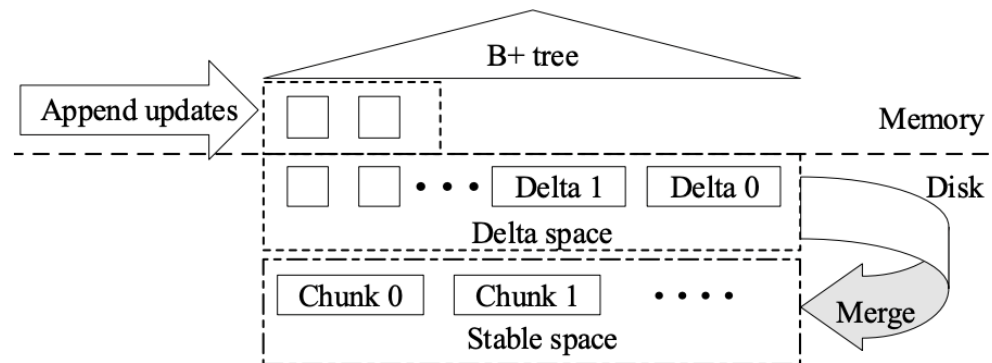
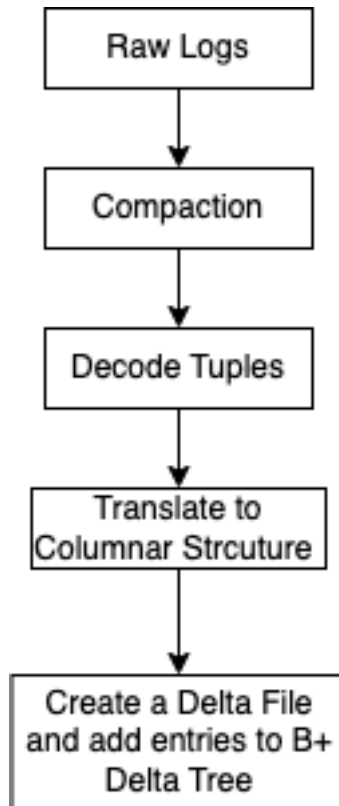


Figure 4: The columnar delta tree

Appendix (Cost Modeling)

$$C_{opt_scan} = \min(C_{col_scan}, C_{row_scan}, C_{index_scan})$$

$$C_{row_scan} = S_{tuple} \cdot N_{tuple} \cdot f_{scan} + N_{reg} \cdot f_{seek}$$

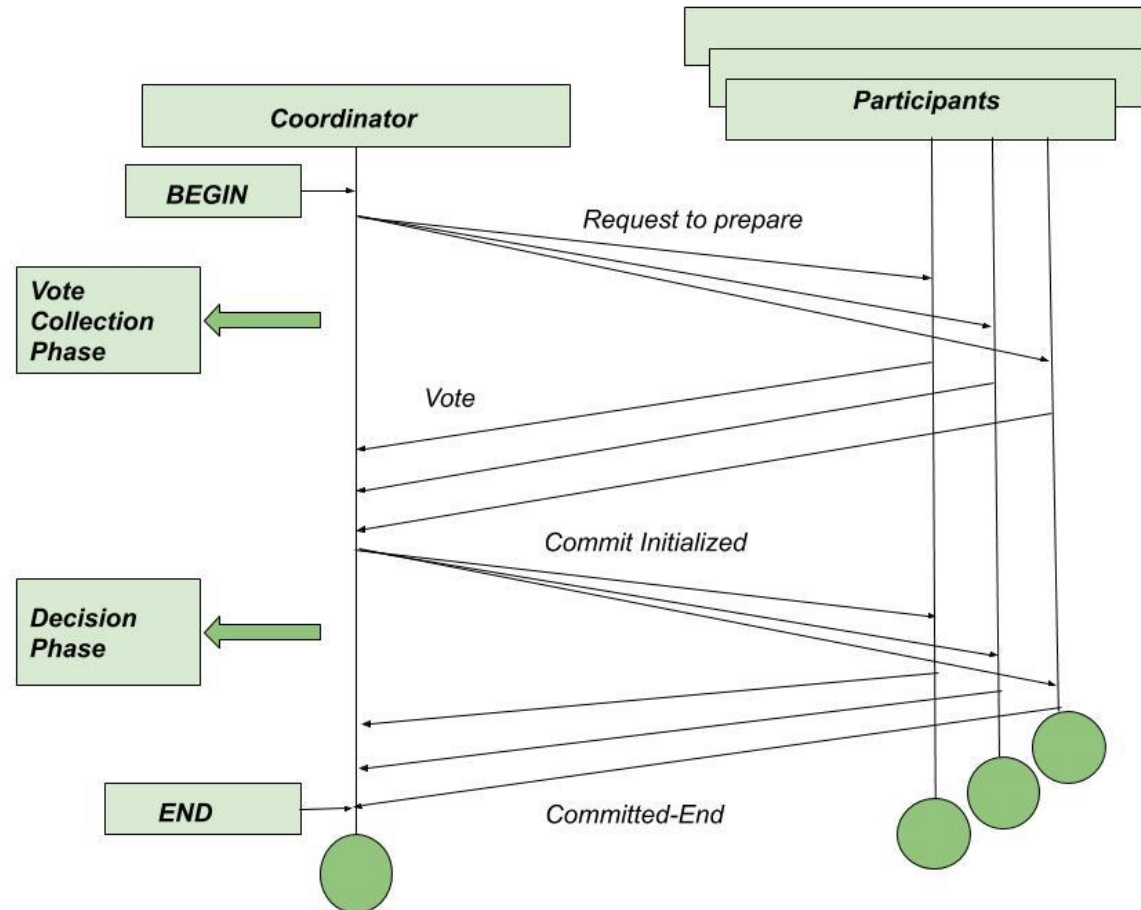
$$C_{col_scan} = \sum_{j=1}^m (S_{col_j} \cdot N_{tuple} \cdot f_{scan} + N_{reg_j} \cdot f_{seek})$$

$$C_{index_scan} = S_{index} \cdot N_{tuple} \cdot f_{scan} + N_{reg} \cdot f_{seek} + C_{double_read}$$

$$C_{double_read} = \begin{cases} 0 & \text{(if without double read)} \\ S_{tuple} \cdot N_{tuple} \cdot f_{scan} + N_{tuple} \cdot f_{seek} \end{cases}$$

C -> Cost
S -> Size
F -> Seek / Scan Cost
N -> Number

Appendix (General Two-Phase Commit)



Appendix (TiDB Two-Phase Commit)

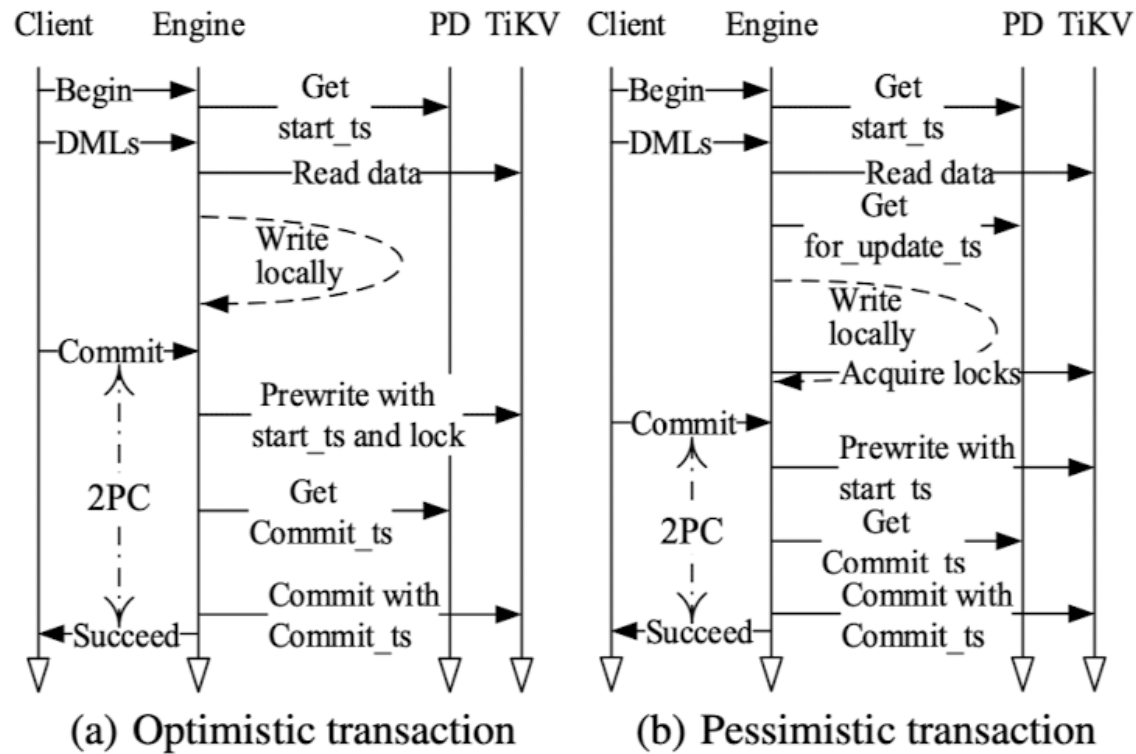


Figure 5: The process of optimistic and pessimistic transaction