



Research Track - Distributed Transaction I

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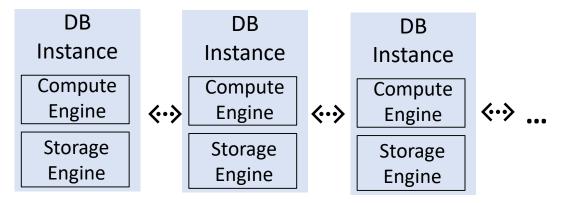
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Northeastern University, China

Huawei Technology Co., Ltd

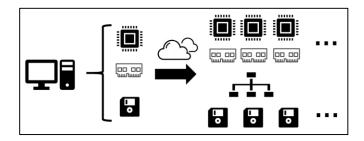
Tsinghua University, China

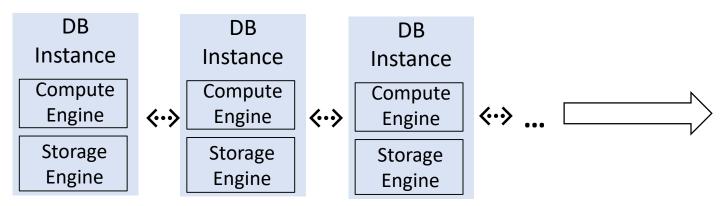
Compute-Storage disaggregated two-layer architecture



Distributed Database System Architecture

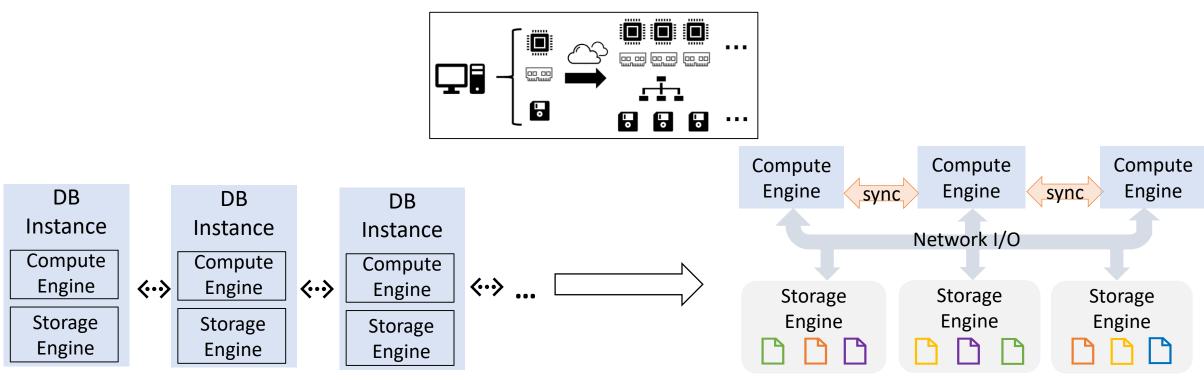
Compute-Storage disaggregated two-layer architecture





Distributed Database System Architecture

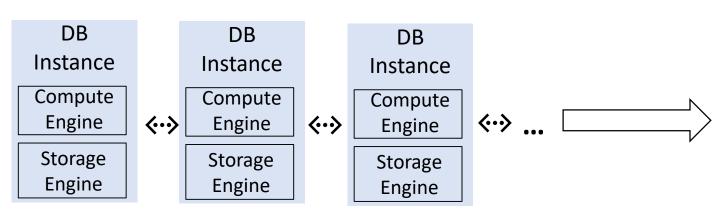
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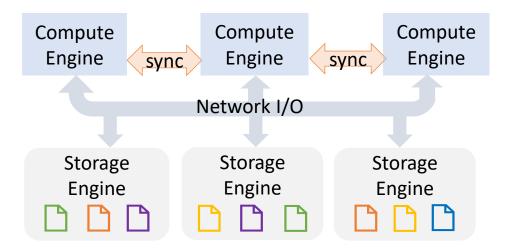


Distributed Database System Architecture

Compute-Storage disaggregated Architecture

- Compute-Storage disaggregated two-layer architecture
- > Benefits:
 - Scalability and Elasticity
 - Cost-effectiveness
 - High Availability





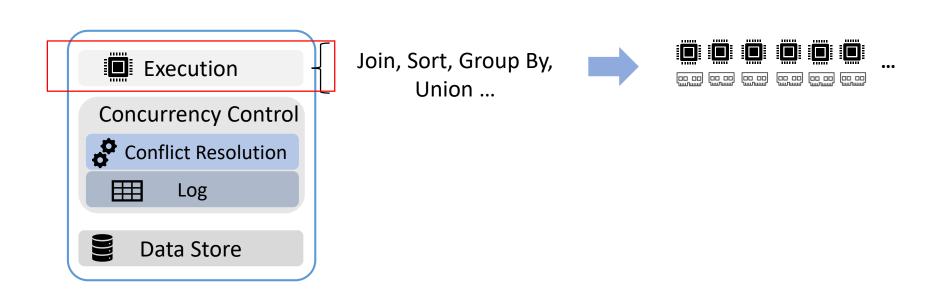
Distributed Database System Architecture

Compute-Storage disaggregated Architecture

• Each module has specific functionality and distinct resource requirements

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 - Execution: High computation tasks, free scale

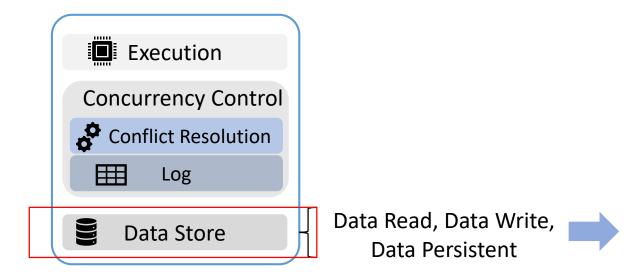
Database System



Each module has specific functionality and distinct resource requirements

- Execution: High computation tasks, free scale
- Data Storage: Large storage space

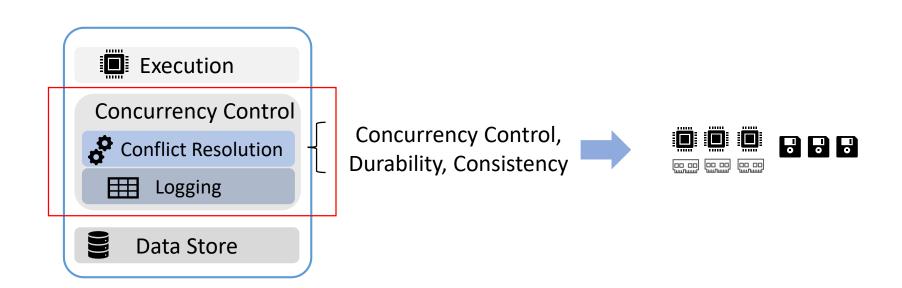
Database System



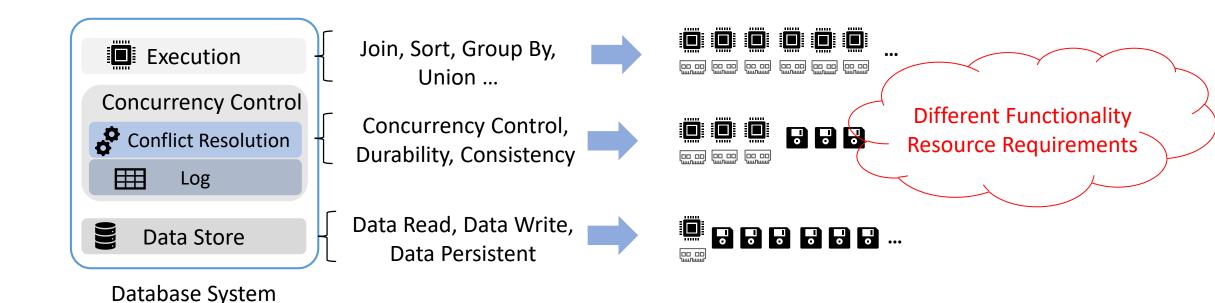
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Database System

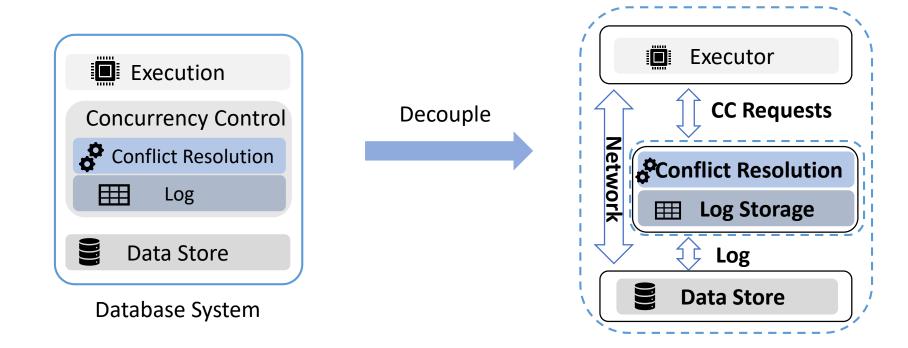
Concurrency Control: high concurrency, data consistency, log durability



- Each module has specific functionality and distinct resource requirements
 - Execution: high computation tasks, free scale
 - Data Storage: large storage space
 - Concurrency Control: high concurrency, data consistency, log durability

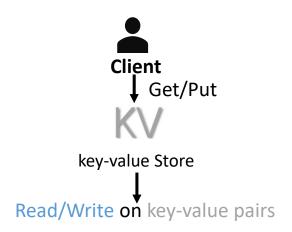


- Each module has specific functionality and distinct resource requirements
 - Decoupling based on functions and resource requirements
 - fully utilize the heterogeneous resources, avoid resource wasting.
 - scale each module independently, high flexibility.

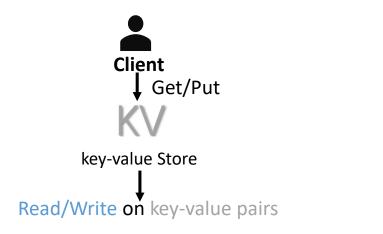


• Databases with different data models process different types of requests:

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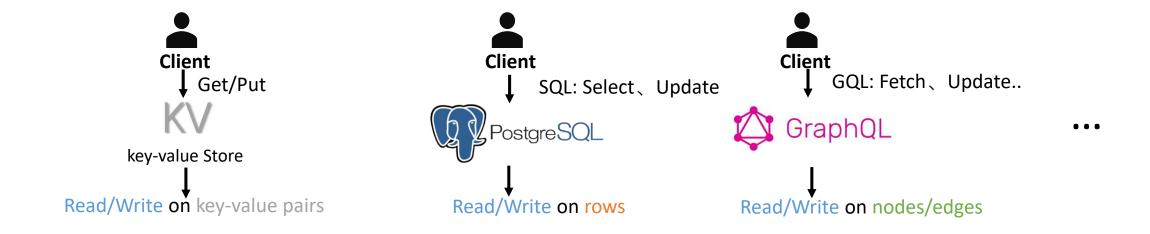


Databases with different data models process different types of requests:

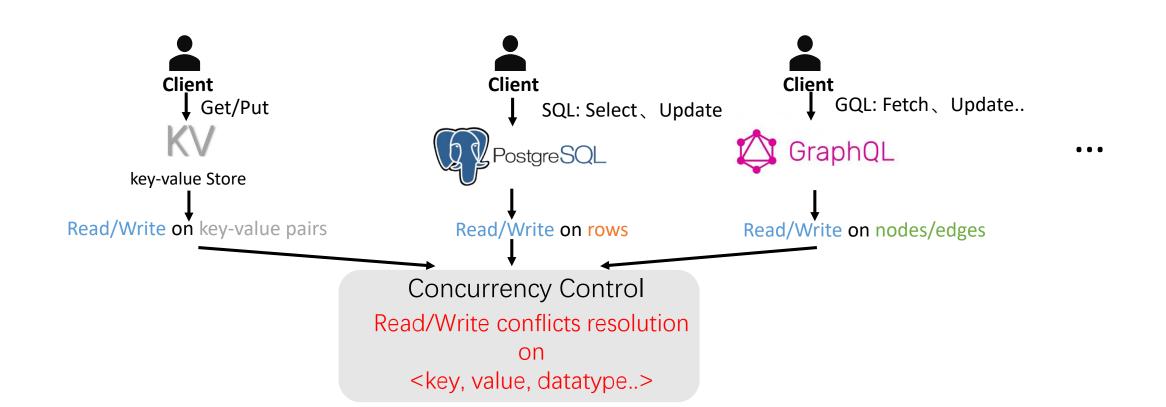




Databases with different data models process different types of requests:

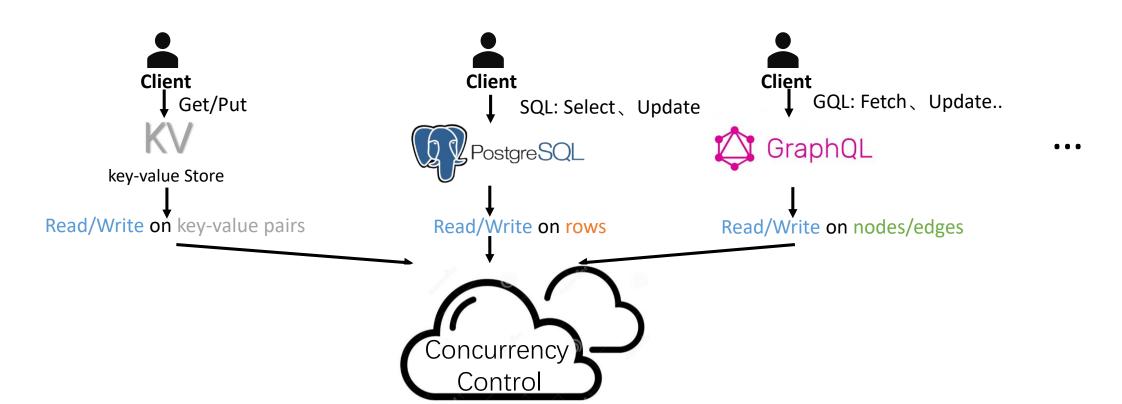


- Databases with different data models process different types of requests:
 - The core of Concurrency Control is handling concurrent read/write conflicts



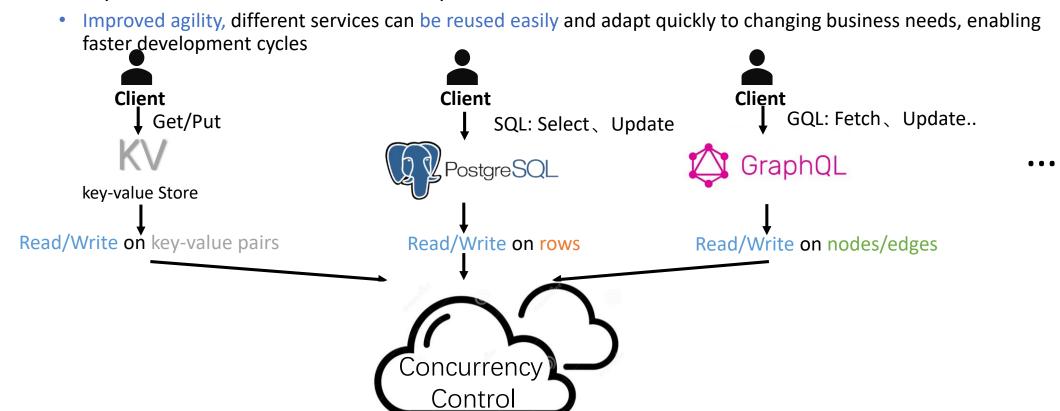
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 - Decoupled function modules

 independent services

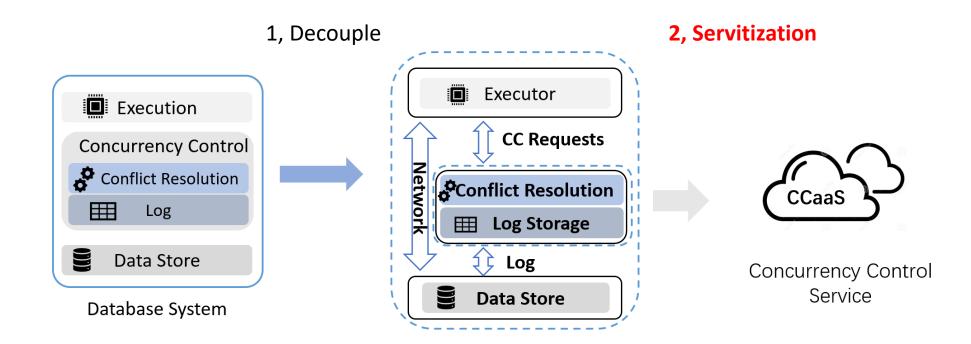


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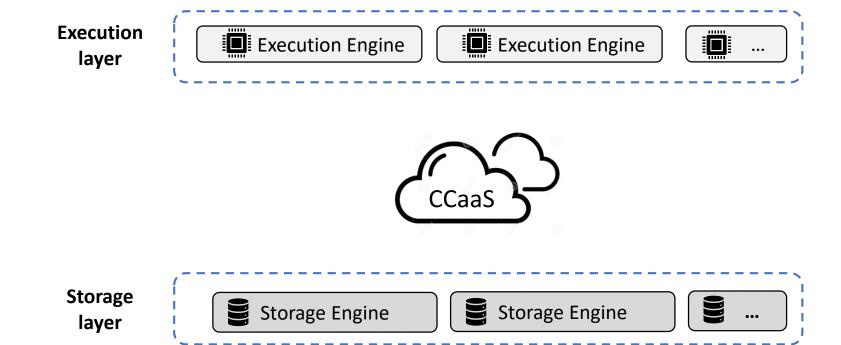
 independent services



- ➤ **Decouple** the concurrency control module high scalability and resource utilization
- Move a step forward: **concurrency control service**development *agility*, different databases can *reuse easily*



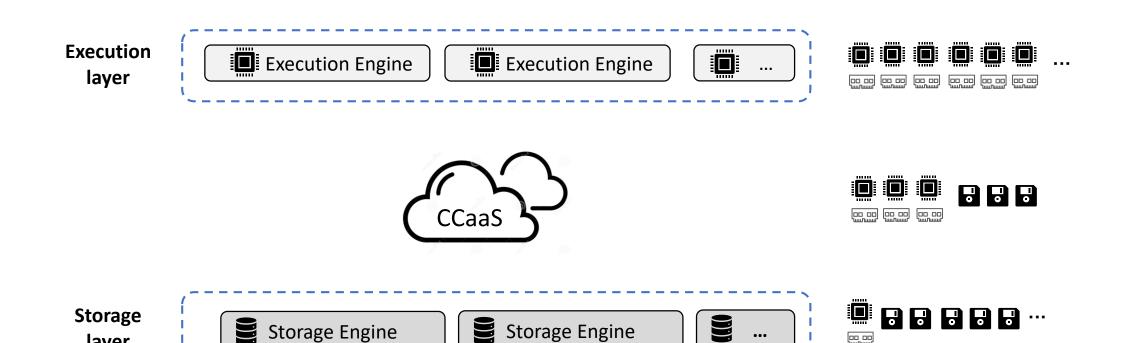
- Decoupled concurrency control module
 - Execution-Concurrency Control-Storage three layer architecture



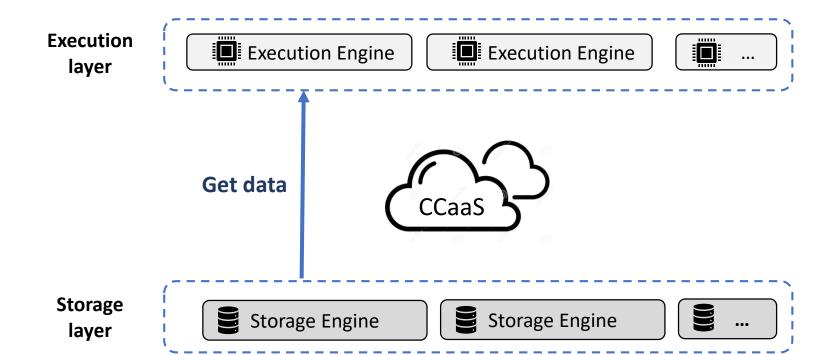
Decoupled concurrency control module

layer

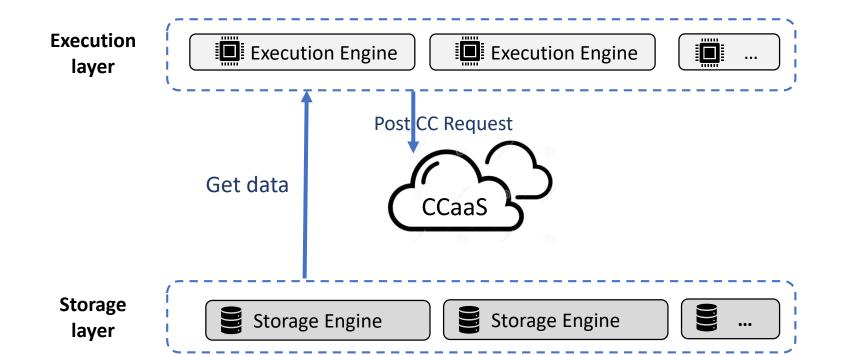
- Execution-Concurrency Control-Storage three layer architecture
 - high scalability and resource utilization



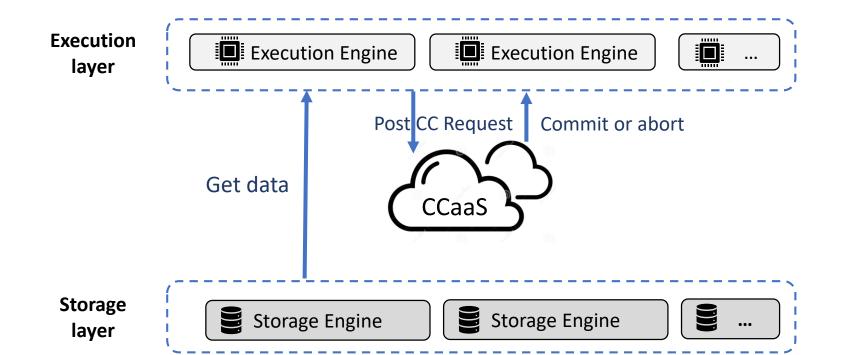
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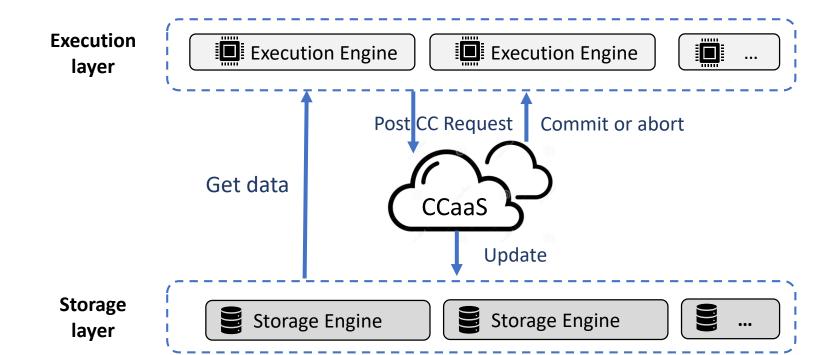
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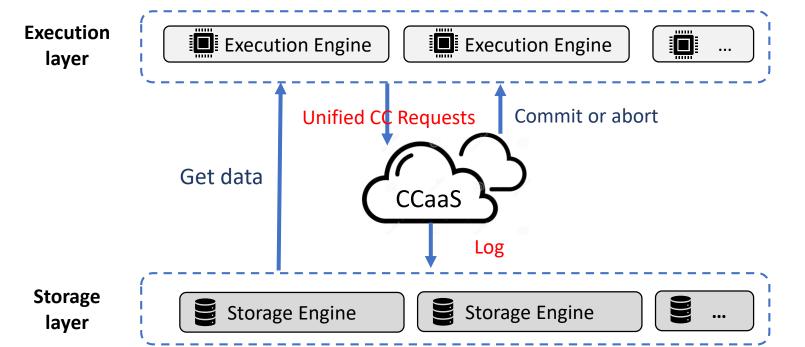
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- Decoupled concurrency control module
 - Execution-Concurrency Control-Storage three layer architecture
 - high scalability and resource utilization
 - concurrency control service



Interfaces for the execution layer:

Begin(ExecutionInfo)

TxnCommit(TxnID, RS, WS)

Lock(TxnID, Key[], Value[],OpType[])

Commit(Txn/D)

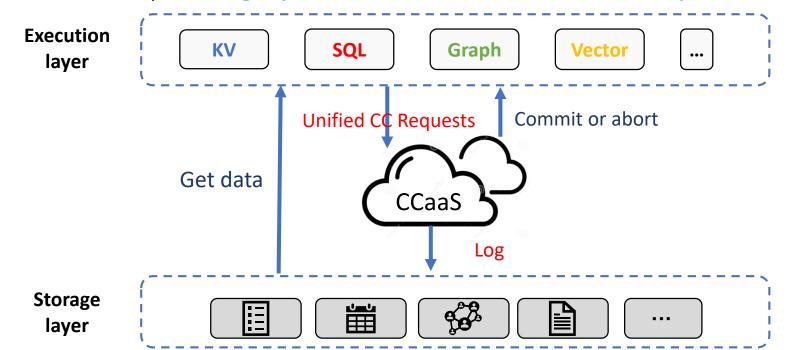
Abort(Txn/D)

Interfaces for the storage layer:

LogPush(LogAdaptorID)
LogPull(LSN∏)

CCaaS Interfaces

- Decoupled concurrency control module
 - Execution-Concurrency Control-Storage three layer architecture
 - high scalability and resource utilization
 - concurrency control service
 - development *agility*, different databases can *reuse easily*



Interfaces for the execution layer:

Begin(ExecutionInfo)

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Commit(Txn/D)

Abort(TxnID)

Interfaces for the storage layer:

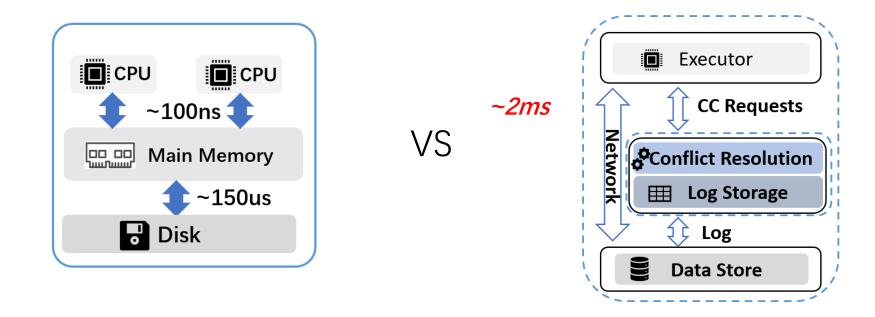
LogPush(LogAdaptorID)
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CCaaS Interfaces

Problem1: Decoupled CC brings more Network Communication

Decoupled CC increasing latency

Communication with hardware VS communication with network.

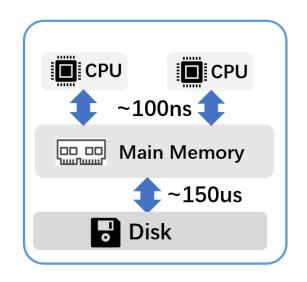


Problem1: Decoupled CC brings more Network Communication

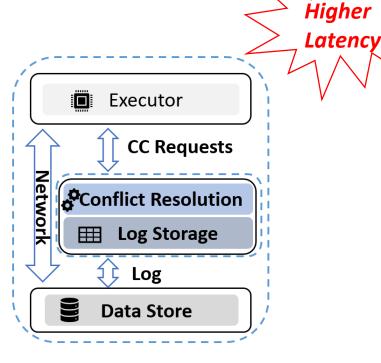
Decoupled CC increasing latency

Communication with hardware VS communication with network.

• Higher latency, not suitable for OLTP workloads.

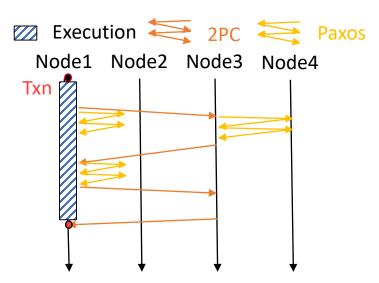






Problem2: High Communication in CC

- Decoupled CC increasing latency
 - Communication with hardware VS communication with network.
 - Higher latency, not suitable for OLTP workloads.
- High Coordination overhead



Communications in Distributed Concurrency Control

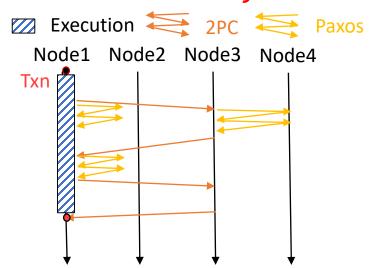
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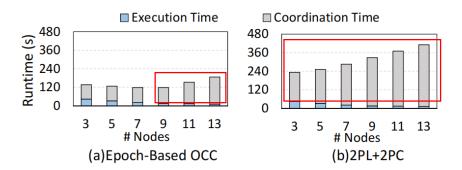
Decoupled CC increasing latency

- Communication with hardware VS communication with network.
- Higher latency, not suitable for OLTP workloads.

High Coordination overhead

- the coordination overhead increases when adding CC nodes
- Limited scalability



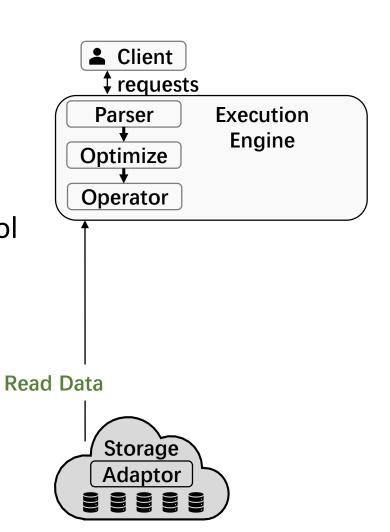


Breakdown of total processing time for distributed transactions with a changing number of nodes.

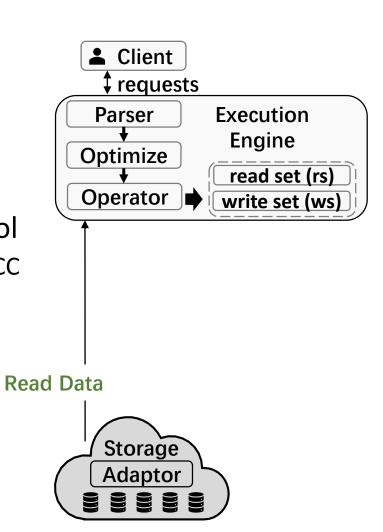
Communications in Distributed Concurrency Control

- Decoupled CC increasing latency
 - Higher latency
- High Coordination overhead
 - Limited scalability
- ➤ Sharded Multi-Write Optimistic Concurrency Control (SM-OCC)

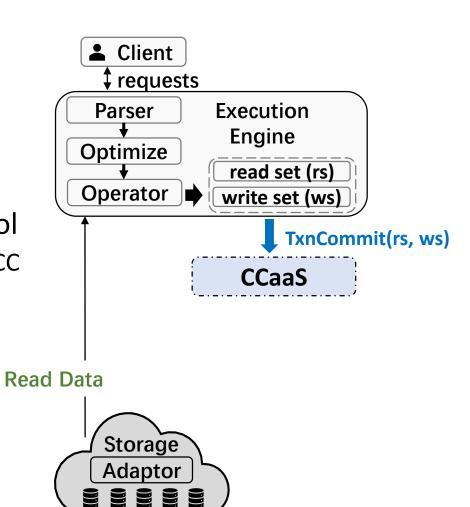
- Decoupled CC increasing latency
 - Higher latency
- High Coordination overhead
 - Limited scalability
- Sharded Multi-Write Optimistic Concurrency Control
 - > Optimistic for fewer communication between execution-CC



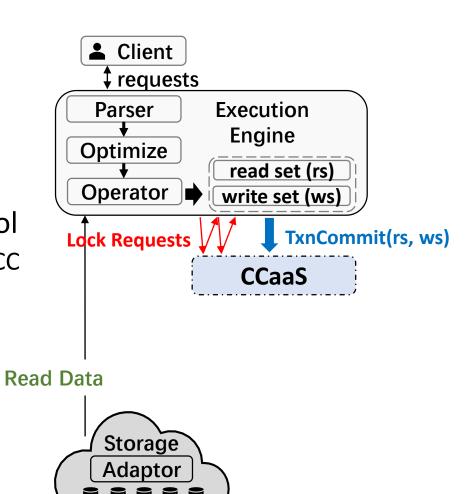
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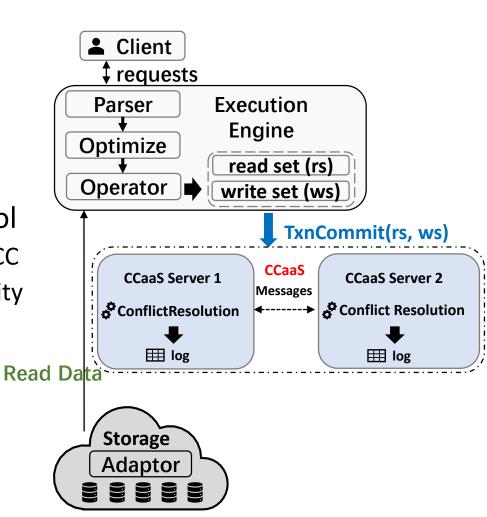
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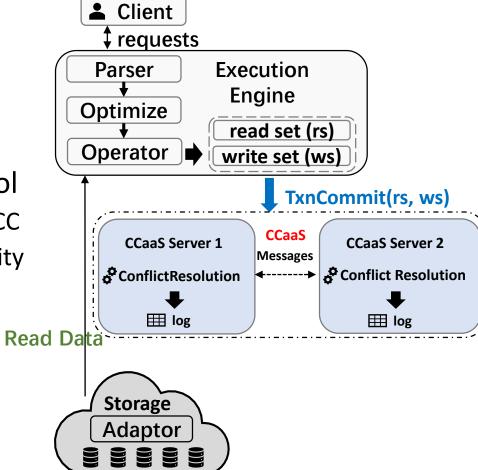
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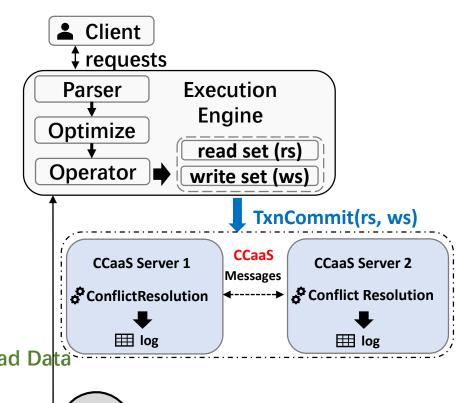
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 - > Multi-Write (Multi-Master) architecture for High availability



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 - > Optimistic for fewer communication between execution-CC
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 - > Sharding for high scalability



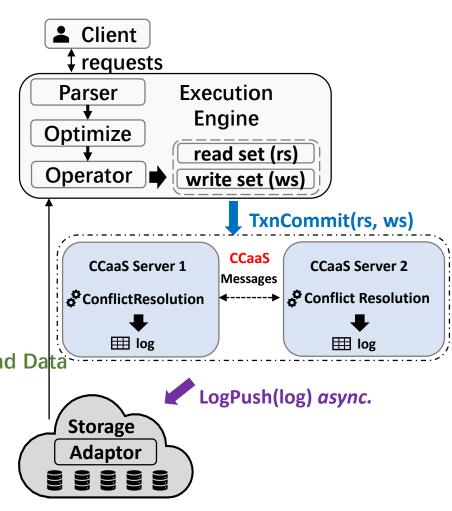
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 - ➤ Epoch-Based & Deterministic Conflict Resolving for fewer Read Data communication between CCaaS nodes



Storage

Adaptor

- Decoupled CC increasing latency
 - Higher latency
- High Coordination overhead
 - Limited scalability
- Sharded Multi-Write Optimistic Concurrency Control
 - > Optimistic for fewer communication between execution-CC
 - > Multi-Write (Multi-Master) architecture for High availability
 - > Sharding for high scalability
 - ➤ Epoch-Based & Deterministic Conflict Resolving for fewer Read Data communication between CCaaS nodes
 - > Async Log push



Snapshot

Each node maintains *sharded* meta-data, achieving high scalability.

CCaaS Node1

CCaaS Node2

Snapshot X=1, ,Z=1

Snapshot

Different nodes maintains some of the same shard, building a multi-master arc.

CCaaS Node1

CCaaS Node2

Snapshot

X=1

,Z=1

```
Snapshot
```

```
X=1,Y=1
```

↓ T1 : R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5

CCaaS Node1



CCaaS Node2



Snapshot X=1, ,Z=1

```
Snapshot
```

X=1,Y=1



 \blacksquare T1: R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5

Note: csn = local time : node_id

CCaaS Node1

①Tag Commit Info {csn = 1:1, cen = 1}

CCaaS Node2



Snapshot X=1, ,Z=1

```
Snapshot X=1,Y=1

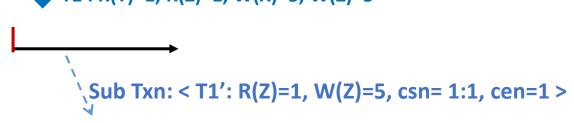
T1: R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5

(CCaaS Node1 Sub Txn: <T1': R(Z)=1, W(Z)=5, csn= 1:1, cen=1>
```

```
Snapshot X=1,Y=1

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CCaaS Node1



Snapshot X=1, ,Z=1

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T1: R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5

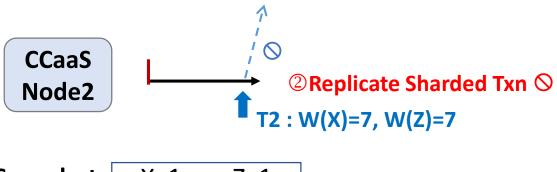
CCaaS
Node1

Sub Txn: <T1': R(Z)=1, W(Z)=5, csn= 1:1, cen=1, >
```

```
CCaaS
Node2

Tag Commit Info {csn = 2:2, cen = 1}
T2: W(X)=7, W(Z)=7

Snapshot X=1, ,Z=1
```



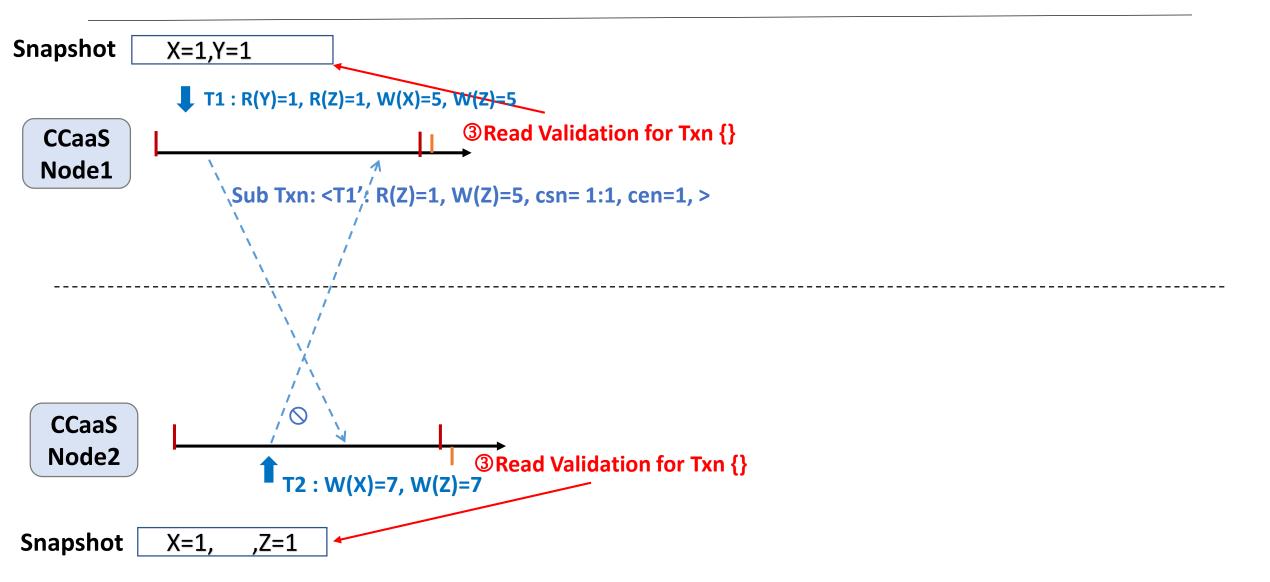
Snapshot X=1, ,Z=1

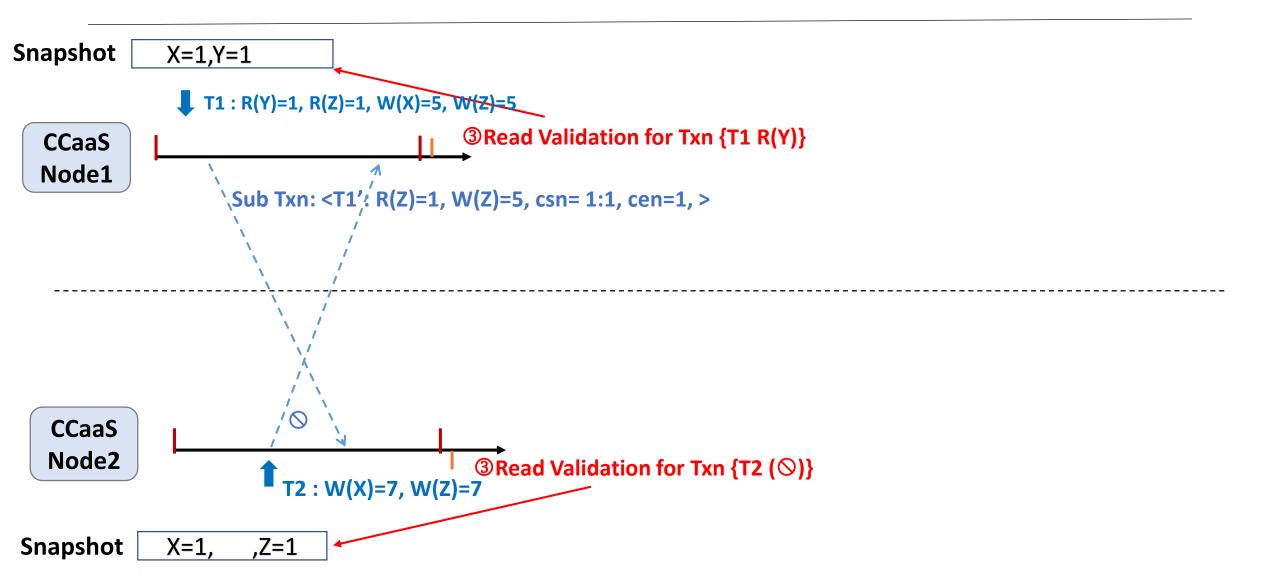
Snapshot

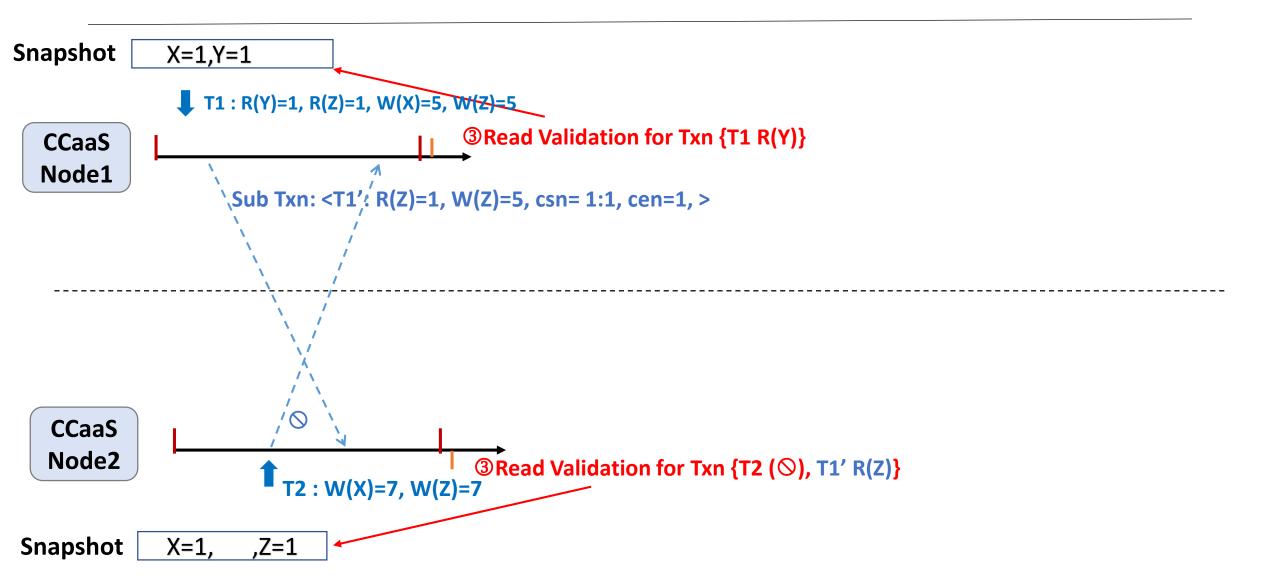
X=1,

,Z=1

```
Snapshot
              X=1,Y=1
               ■ T1 : R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5
   CCaaS
  Node1
                   \Sub Txn: <T1'. R(Z)=1, W(Z)=5, csn= 1:1, cen=1, >
   CCaaS
   Node2
                         T2: W(X)=7, W(Z)=7
```





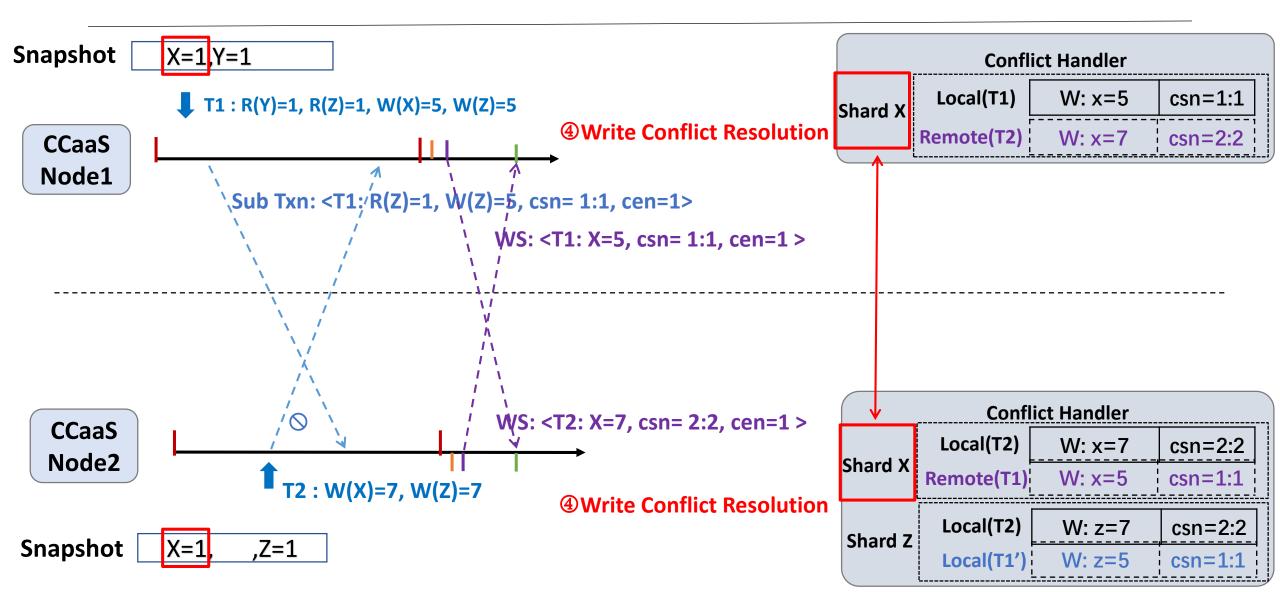


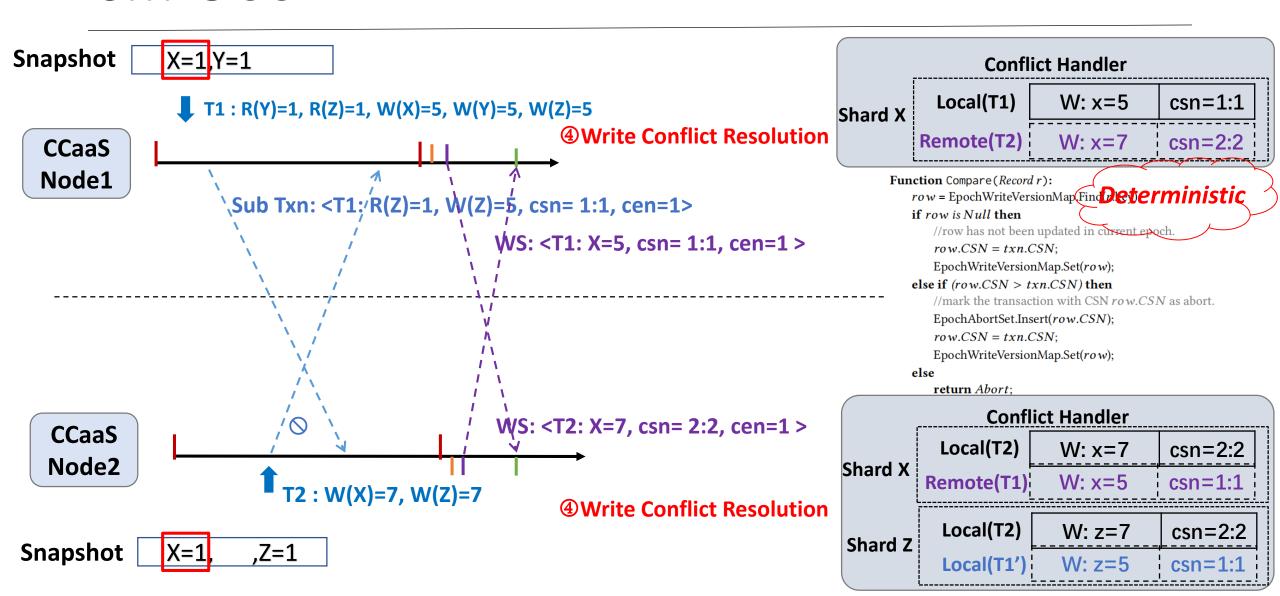
Snapshot

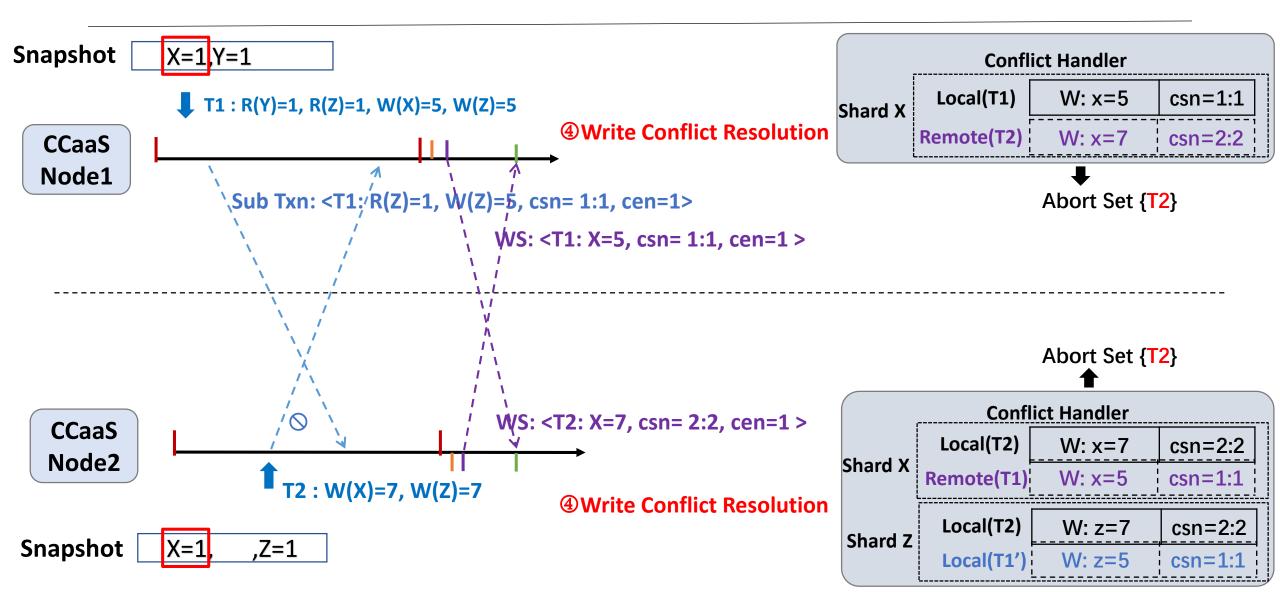
X=1

,Z=1

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Snapshot
               X=1,Y=1
                \blacksquare T1 : R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5
   CCaaS
  Node1
                     \Sub Txn: <T1;'R(Z)=1, W(Z)=5, csn=1:1, cen=1>
                                                WS: <T1: X=5, csn= 1:1, cen=1 >
                                              / WS: <T2: X=7, csn= 2:2, cen=1 >
   CCaaS
   Node2
                           T2: W(X)=7, W(Z)=7
```







Snapshot

X=5,

,Z=5

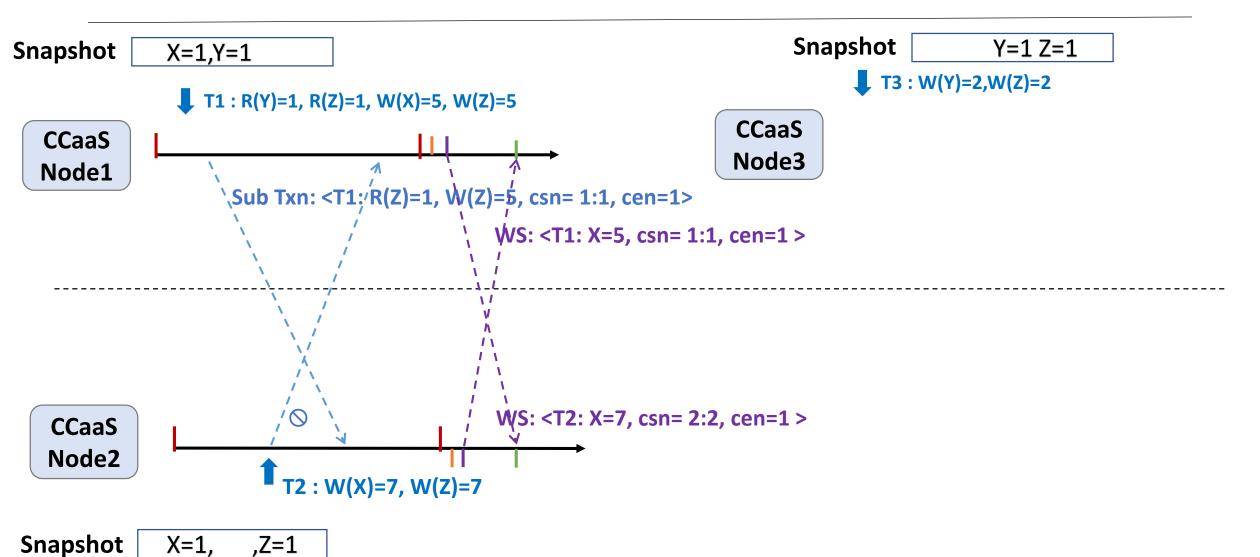
```
Snapshot
               X=5,Y=1
                \blacksquare T1 : R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5
  CCaaS
                                                       ⑤Epoch Logging T1} && Update Snapshot
  Node1
                    \Sub Txn: <T1; R(Z)=1, W(Z)=5, csn= 1:1, cen=1>
                                              WS: <T1: X=5, csn= 1:1, cen=1 >
                                               WS: <T2: X=7, csn= 2:2, cen=1 >
   CCaaS
   Node2
                                                      ⑤Epoch Logging T1} && Update Snapshot
                          T2: W(X)=7, W(Z)=7
```

Snapshot

X=5,

,Z=5

```
Snapshot
               X=5,Y=1
                \blacksquare T1 : R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5
   CCaaS
                                                         ©Log Push to the storage layer
  Node1
                    \Sub Txn: <T1; R(Z)=1, W(Z)=5, csn= 1:1, cen=1>
                                                WS: <T1: X=5, csn= 1:1, cen=1 >
                                                W/S: <T2: X=7, csn= 2:2, cen=1 >
   CCaaS
   Node2
                           T2:W(X)=7,W(Z)=7
                                                          ©Log Push to the storage layer
```



Snapshot

 \blacksquare T1 : R(Y)=1, R(Z)=1, W(X)=5, W(Z)=5

CCaaS Node1



Snapshot

Y=1,Z=1

T3: W(Y)=2,W(Z)=2

CCaaS Node3 **Conflict Handler**

Shard Y

Local(T3) Remote(T2)

Shard Z

Remote(T1') Remote(T2) Local(T3)

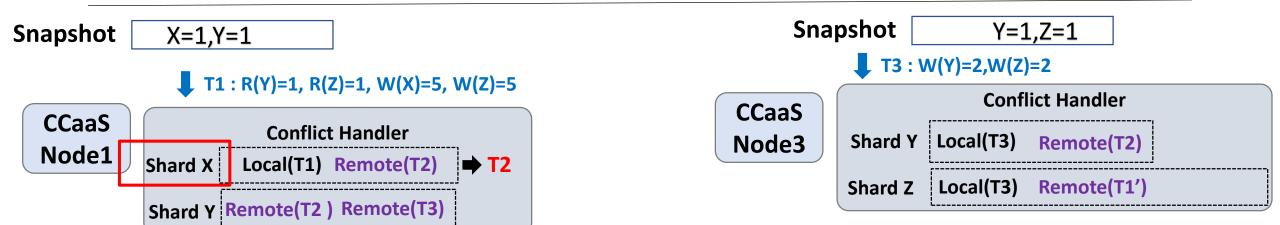
CCaaS Node2

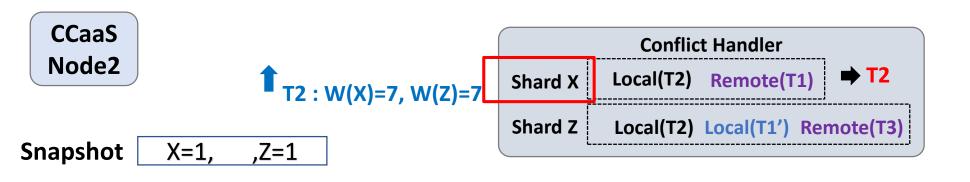
T2: W(X)=7, W(Z)=7

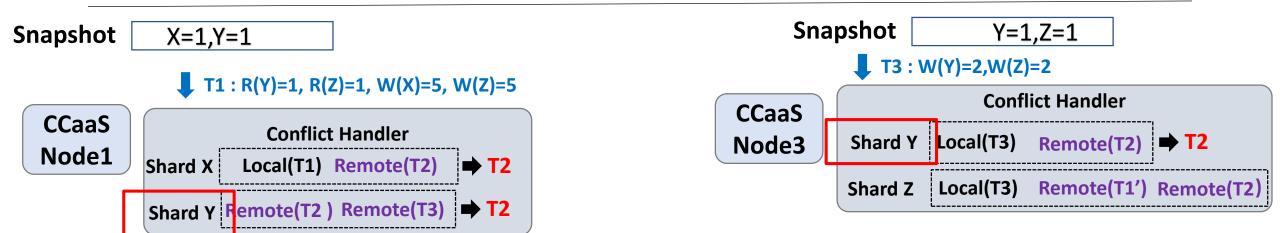
Conflict Handler Local(T2) **Shard X Shard Z** Local(T2) Local(T1') Remote(T3

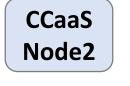
Snapshot

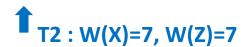
X=1, ,Z=1









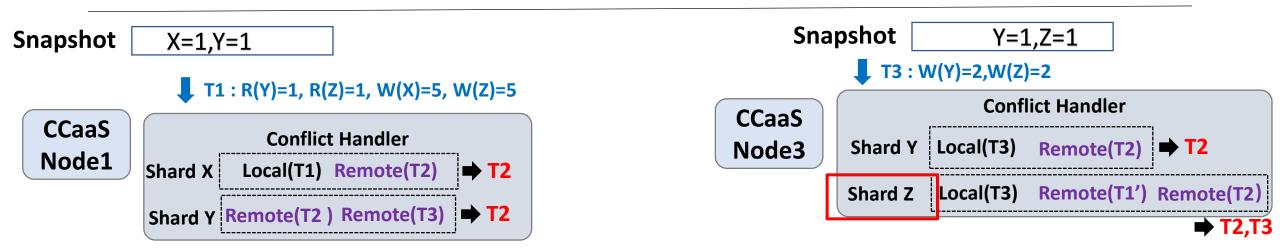


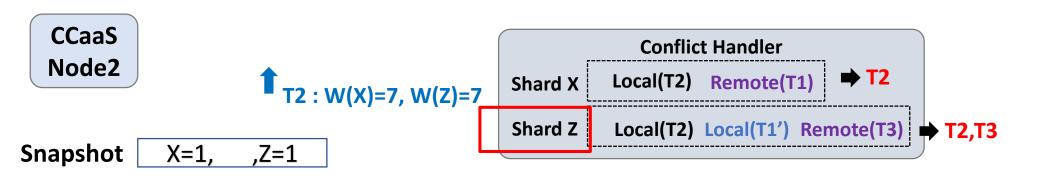
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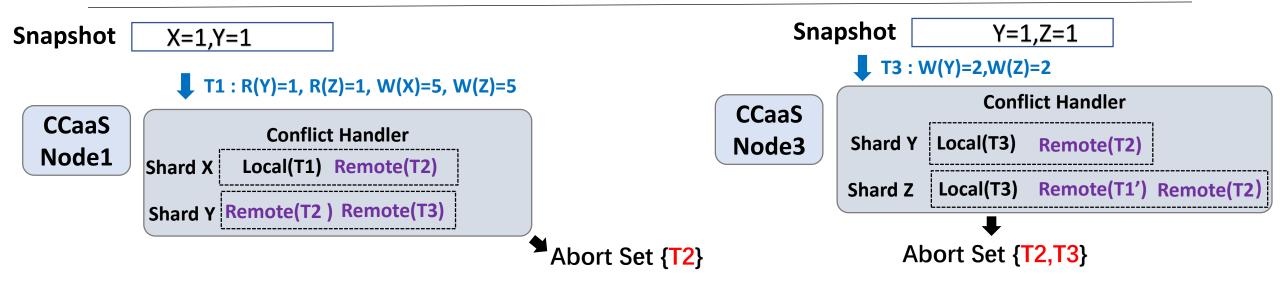
Shard X Local(T2) Remote(T1) → T2

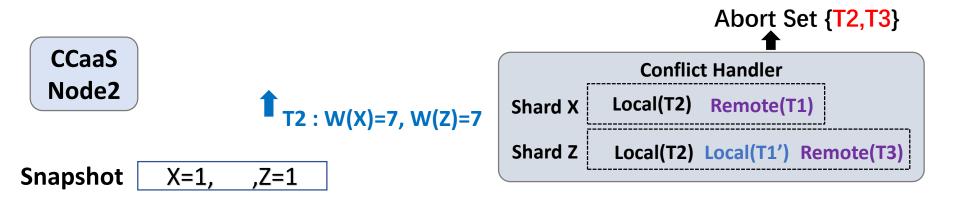
Shard Z Local(T2) Local(T1') Remote(T3)

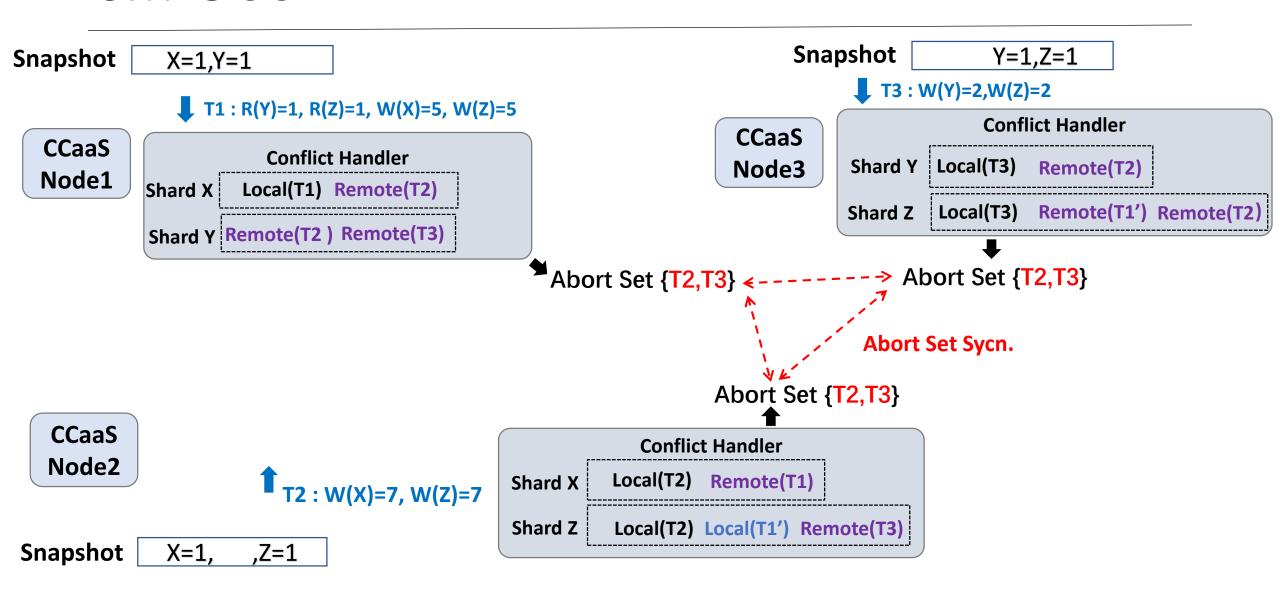
Snapshot X=1, ,Z=1









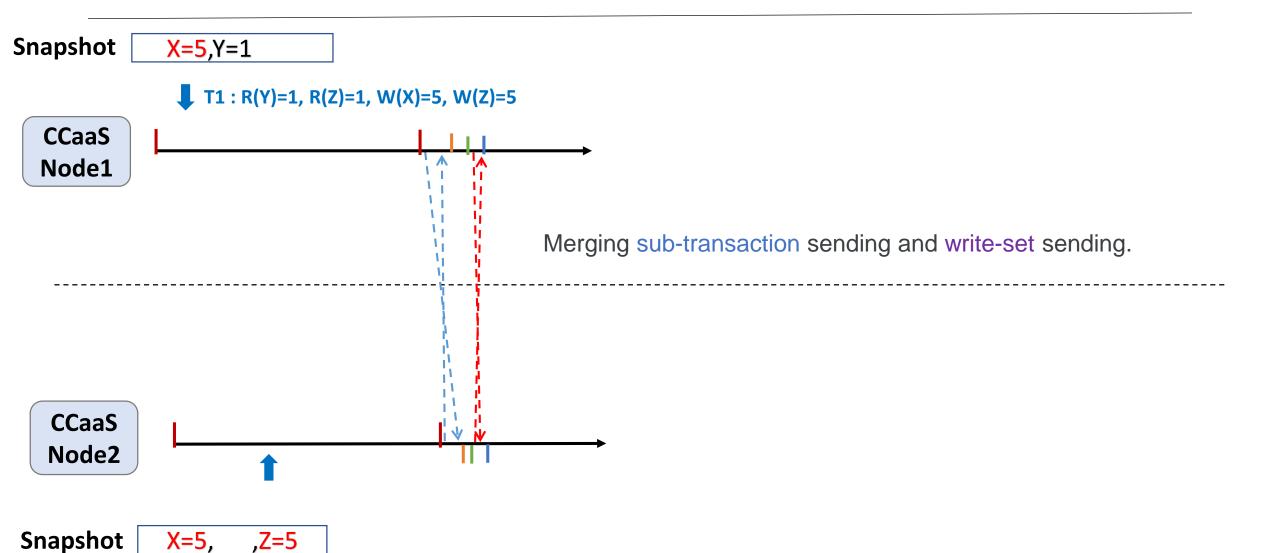


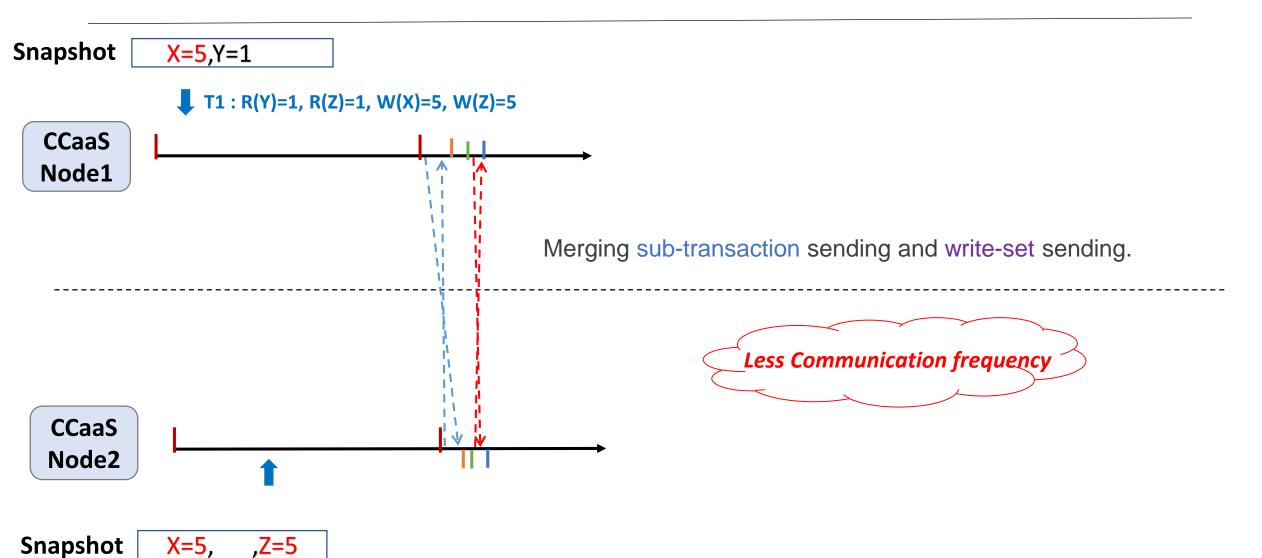
Snapshot

X=5,

,Z=5

```
Snapshot
               X=5,Y=1
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  CCaaS
                                                       ⑤Epoch Logging T1} && Update Snapshot
  Node1
                    \Sub Txn: <T1; R(Z)=1, W(Z)=5, csn= 1:1, cen=1>
                                               WS: <T1: X=5, csn= 1:1, cen=1 >
                                                   Abort Set Sycn.
                                               W$: <T2: X=7, csn= 2:2, cen=1 >
   CCaaS
   Node2
                                                      ⑤Epoch Logging T1} && Update Snapshot
                          T2: W(X)=7, W(Z)=7
```

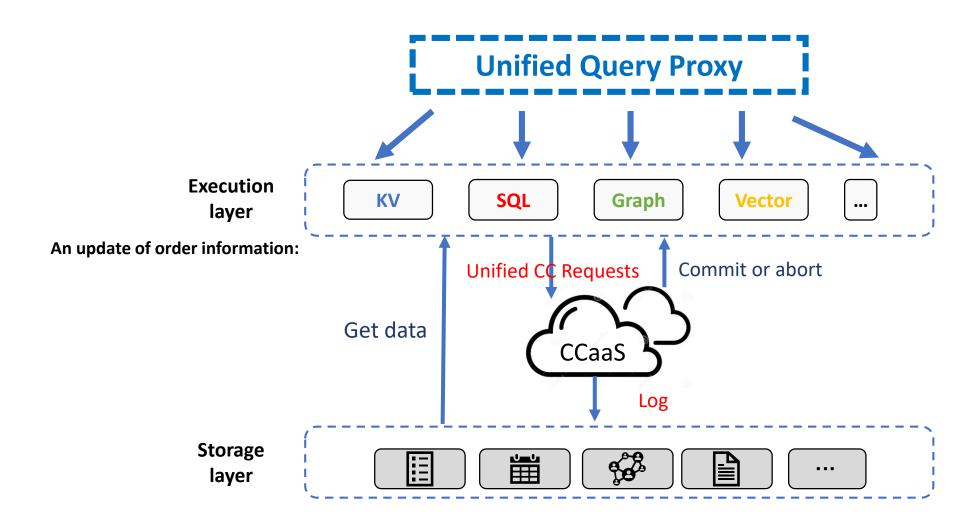




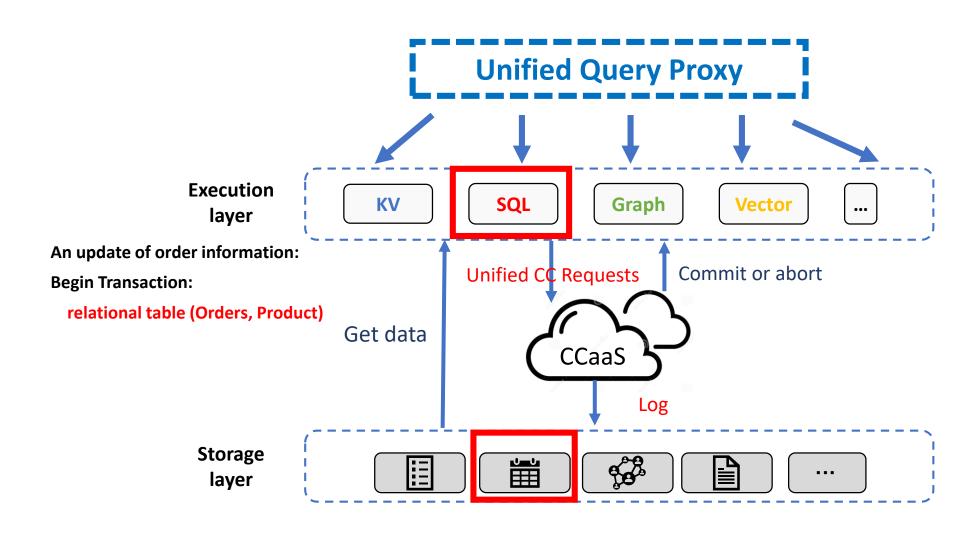
Case Studies

Supporting Cross-Model Transactions

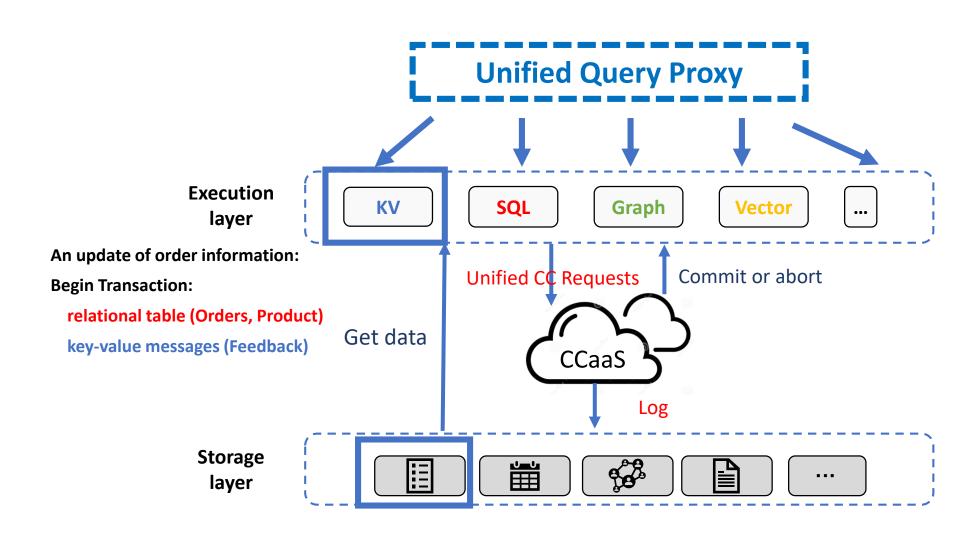
Cross-Model Transaction with CCaaS



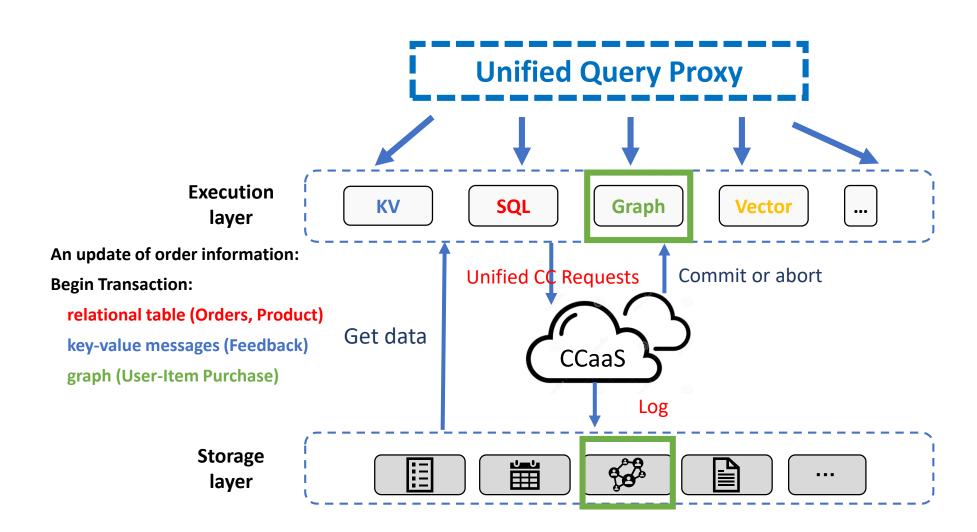
Cross-Model Transaction with CCaaS



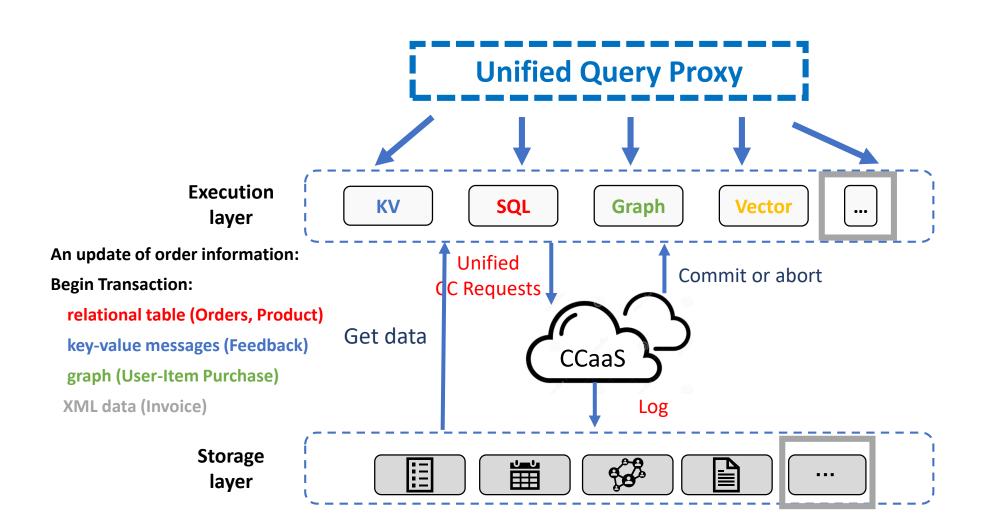
Cross-Model Transaction with CCaaS



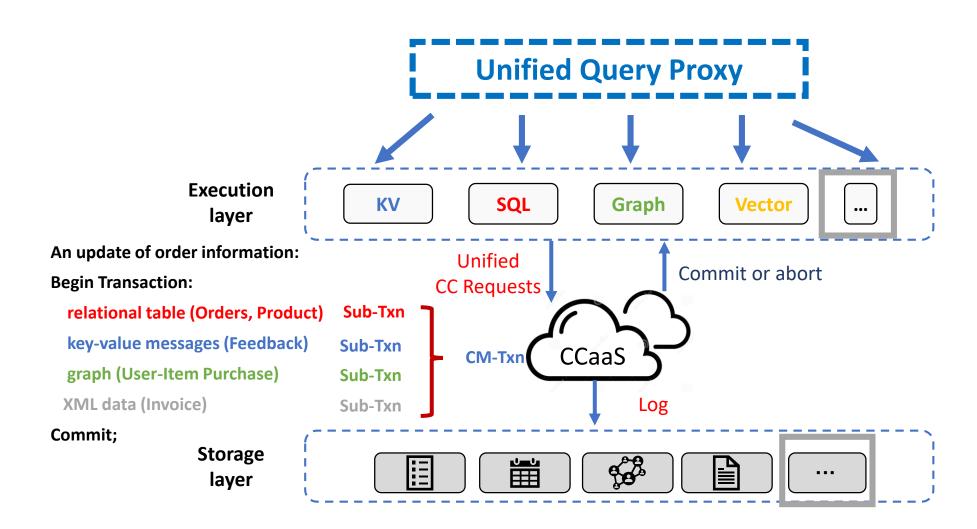
Cross-Model Transaction with CCaaS



Cross-Model Transaction with CCaaS



Cross-Model Transaction with CCaaS



Case Studies

- Supporting Cross-Model Transactions
- Empowering NoSQL DBs with TP Capability
- Making Standalone TP Distributed

• ...

More Details

• More:

- Sharding
- Isolation
- Logging
- Fault Tolerance
- Cross-Model



Concurrency Control as a Service

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Existing disaggregated databases separate execution and storage layers, enabling independent and elastic scaling of resources. In most cases, this design makes transaction concurrency control (CC) a critical bottleneck, which demands significant computing resources for concurrent conflict management and struggles to scale due to the coordination overhead for concurrent conflict resolution Coupling CC with execution or storage limits performance and elasticity, as CC's resource needs do not align with the free scaling of the transaction execution layer or the storage-bound data layer

This paper proposes Concurrency Control as a Service (CCaaS), which decouples CC from databases, building an execution-CC-storage three-layer decoupled database, allowing independent scaling and upgrades for improved elasticity, resource utilization, and development agility. However, adding a new layer increases latency address this, we propose a Sharded Multi-Write OCC (SM-OCC) algorithm with an asynchronous log push-down mechanism to minimize network communications overhead and transaction latency Additionally, we implement a multi-write architecture with a deministic conflict resolution method to reduce coordination over head in the CC layer, thereby improving scalability, CCaaS is designed to be connected by a variety of execution and storage engines. Existing disaggregated databases can be revolutionized with CCaaS to achieve high elasticity, scalability, and high performance. Results show that CCaaS achieves 1.02-3.11× higher throughput and 1.11-2.75× lower latency than SoTA disaggregated databases.

Weixing Zhou, Yanfeng Zhang, Xinji Zhou, Zhiyou Wang, Zeshun Peng, Yang Ren, Sihao Li, Huanchen Zhang, Guoliang Li, and Ge Yu.
Concurrency Control as a Service. PVLDB, 18(9): 2761 - 2774, 2025. doi:10.14778/3746405.3746406

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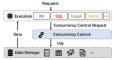


Figure 1: An execution-CC-storage three-layer decoupled

PVLDB Artifact Availability:

The source code, data, and/or other artifacts have been made available at https://github.com/iDC-NEU/CCaaS.

1 INTRODUCTION

Database systems are evolving to a compute-storage disaggregated architecture [24, 33, 48, 52, 73, 80, 83, 87], such as Amazon Aurora [20], Socrates [23], PolarDB [28], and AlloyDB [1]. These databases typically decouple the system into an execution layer, which requires substantial computational resources, and a storage laver, which necessitates significant storage capacity. Compared to traditional databases where execution and storage are bundled together, these two-layer databases allow compute and storage resources to be scaled independently, thereby providing greater elasticity in the cloud environment, which are also called cloud-native databases. A set of works [43, 80, 83, 86] are proposed to improve these cloud-native databases from various aspects. As more and more enterprises move their applications to the cloud, these disaggregated

databases are gaining wide popularity.

The spirit of cloud-native architecture is decoupling. A system should be decoupled into independent function modules, each with specific resource requirements. Cloud provides the elasticity of different decoupled resources (e.g., computation, memory, and storage), allowing the growing or shrinking of resource capacity to adjust to changing demands. Each decoupled function module can

- Cluster Set Up:
 - 3 nodes in each layer,
 - 16 vCPUs, 32G DRAM, Ubuntu 22.04 LTS,
 - 2.5Gbps, ~2ms latency
- Benchmark:
 - YCSB¹
 - YCSB-A (50% read,50 Write, θ = 0.99)
 - YCSB-B (95% read, 5% write, θ = 0.99)
 - TPC-C²

1: YCSB 10 op/txn

2: 50% New-Order – 50% Payment in GeoGauss[SIGMOD 2023]

- Scalability & Elasticity
 - Coupling CC with the execution VS Decoupled CC
 - Dynamically adjust the number of execution and CC nodes

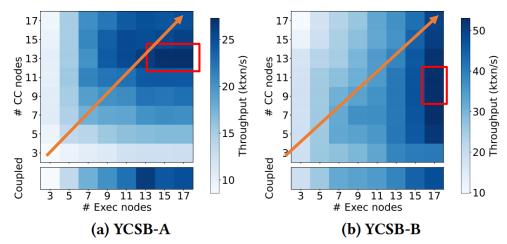


Figure 15: Throughput when scaling the number of execution nodes and CC nodes in CCaaS.

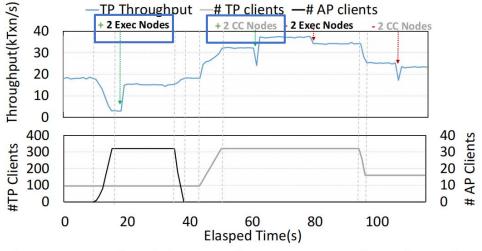


Figure 16: Elasticity performance under changing workloads.

- Performance
 - Throughput 1.02-3.11 X
 - Latency 1.11-2.75 X

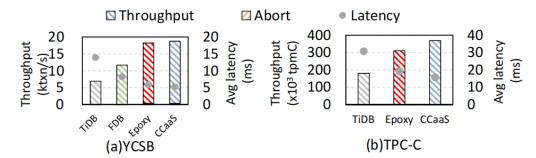


Figure 13: Experiment results with competitors.

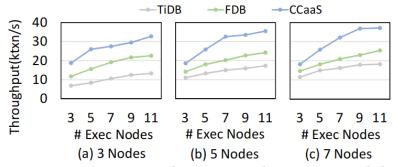


Figure 14: Comparison with existing disaggregated databases under YCSB-B workload.

Servitization

- Empowering TP Capability to NoSQL DBs
- Distributed TP with standalone TP engines
- Cross-Model Transactions

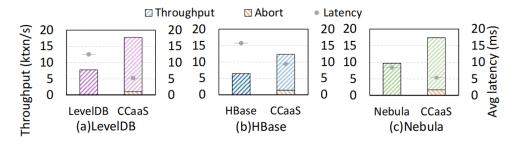


Figure 17: Performance of NoSQL DBs empowered with ACID TP capability.

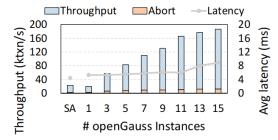


Figure 18: Performance of serving standalone TP engines.

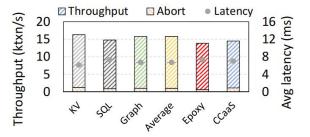


Figure 19: Performance of cross-model (CM) transactions.

Concurrency Control as a Service

- > Decoupled execution-transaction-storage three-layer cloud-native database architecture
 - High scalability, elasticity
 - High development agility
- Concurrency Control Service
 - Empowering NoSQL DBs with TP Capability
 - Making Standalone TP Distributed
 - Supporting Cross-Model Transactions
- > Sharded Multi-Write Optimistic Concurrency Control Algorithm
 - High availability
 - Low communication costs



Thank you! Q&A

Concurrency Control as a Service

■ Scalability

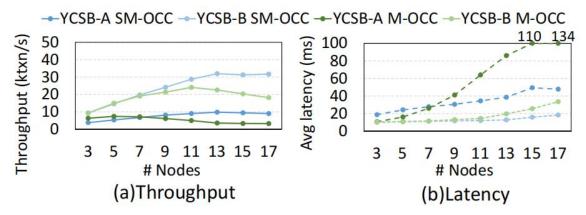


Figure 16: Scaling performance of CCaaS.

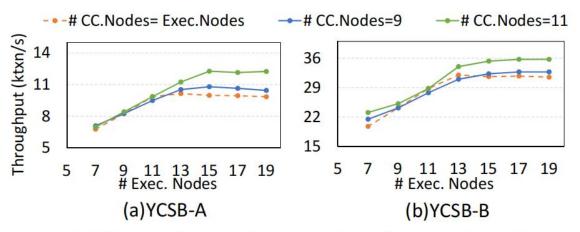
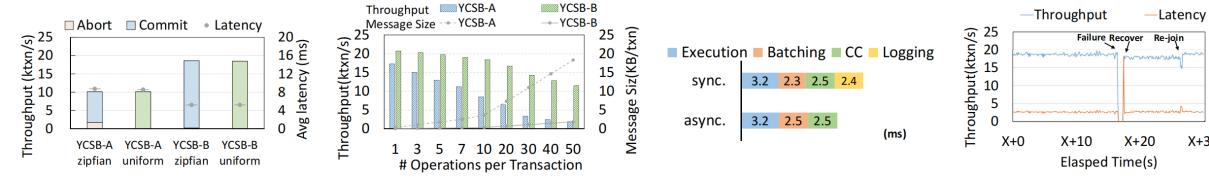


Figure 17: Throughput when varying the number of execution nodes and CCaaS nodes.

Concurrency Control as a Service

■ Breakdown



different contention.

varying number of operations.

Figure 20: Performance under Figure 21: Performance when Figure 22: Runtime break- Figure 23: Performance varidown (sync./async. logging). ation when system recovery.

20

X+30

