

An OLTP Topic

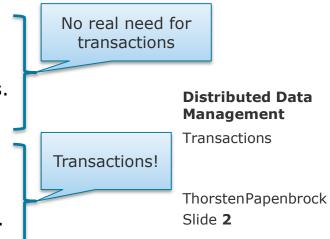


Motivation

- Most database interactions consist of multiple, coherent operations.
- Interactions can be affected by other interfering interactions and errors.
 - \triangleright Database must ensure that interactions work correctly (\rightarrow transactions).

OLAP vs. OLTP

- OLAP systems ...
 - prepare the data once.
 - send complex but individual, ungrouped read-queries.
 - resend failed queries and do not interfere.
- OLTP systems ...
 - change the data frequently.
 - send coherent operations with mixed read/write load.
 - must ensure that interactions succeed consistently.



Transactions Definition

See lecture "Database Systems I" by Prof. Naumann



Transaction

- A sequence of database operations (read/write) that carry a database from one state into another (possibly changed) state.
- Transactions operate in different items (multi-object operations).
- Transactions succeed (commit) or fail (abort/rollback).
- The ACID safety guarantees must be satisfied:
 - Atomicity: A transaction is executed entirely or not at all.
 - Consistency: A transaction carries the database from a consistent state into a consistent state (consistent = logically and technically sound).
 - Isolation: A transaction does not contend with other transactions.
 Contentious access to data is moderated by the database so that transactions appear to run sequentially.
 - Durability: A transaction causes, if successful, a persistent change to the database.

Most distributed DBMSs do not support transactions and stick to the BASE consistency model

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Transactions

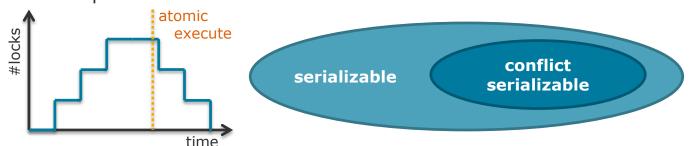
Achieving Isolation

See lecture "Database Systems I" by Prof. Naumann



Locking

- Block an item (row, document, ...) for exclusive reads/writes of one transaction.
- Two-Phase Locking:
 - All locks in one transaction are set before the first lock is given up.
 - Technique to ensure conflict-serializable execution of transactions.



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Transactions

Scheduling

- Creating an execution order for transaction operations.
- See: serial schedule, serializable schedule, legal schedule

Locking is an issue if data is replicated!

Causal Ordering (recap)



Linearizable (and Total Order Broadcast)

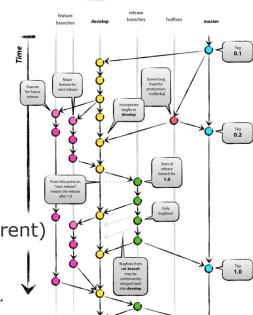
- Imposes a total order:
 - All events can be compared.
 - For one object, only the newest event is relevant.
- Implies causality:
 - A linear order is always also a causal order of the events.
- Is expensive (due to global order enforcement)

Causal ordering

- Imposes a partial order:
 - Some events are comparable (causal), others are not (concurrent)
 - For many events some partial order is just fine:
 - Order of writes, side-channel messages, transactions ...
- Is cheaper (order enforcement only for related events)

Thinking: timelines that branch/merge; events compare only along lines





Ordering Guarantees

Causal Consistency



Causal ordering:

- Example: reads and writes in transactional systems
 - Reads and writes are causally unrelated unless they ...
 - target the same object or
 - connect through transactions.
- A system that guarantees causal ordering is causal consistent.

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Consistency and Consensus

Transactions Inconsistencies





Dirty Read: (write-read conflict)

- Reading a inconsistent value
- Example: $W_1(A) r_2(A) W_1(A)$ **A** was not finished and never supposed to be read.

Non-Repeatable Read: (read-write conflict)

- Reading an outdated value
- Example: $\mathbf{r_1}(\mathbf{A})$ $\mathbf{w_2}(\mathbf{A})$ $\mathbf{r_1}(\mathbf{A})$ Re-reading **A** resulted in a different, inconsistent value.

Lost Update: (write-write conflict)

- Losing a written value
- Example: $\mathbf{w_1}(\mathbf{A})$ $\mathbf{w_2}(\mathbf{A})$ $\mathbf{r_1}(\mathbf{A})$ Update of **A** is lost during the transaction.

Phantom Read: (read-write and write-read conflict)

- Reading/writing of inconsistent values
- Example: $\mathbf{r_1}(\mathbf{A})$ $\mathbf{w_2}(\mathbf{B})$ $\mathbf{r_1}(\mathbf{B})$ $\mathbf{w_2}(\mathbf{A})$ Either A's or B's value is a phantom (should not be there).

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Isolation

See lecture "Database Systems I" by Prof. Naumann



Isolation levels

- To ensure ACID, transactions must be serializable.
 - Very costly, but any weaker level breaks isolation.

Usually default

Isolations-Level	Lost Update	Dirty Reads	Non-Repeatable Reads	Phantom Rea
READ_UNCOMMITTED	prevented	possible	possible	possible
READ_COMMITTED	prevented	prevented	possible	possible
REPEATABLE_READ	prevented	prevented	prevented	possible
SERIALIZABLE	prevented	prevented	prevented	prevented

- READ_COMMITTED:
 - Read only committed values (remember local logical UNDO/REDO logs).
 - No dirty reads, because only consistent values are committed.
 - Still non-repeatable reads, because transactions interleave.

Isolation



occur concurrently)

Isolation levels

Causally related operations are ordered (unrelated operations still

- Snapshot isolation: "readers don't block writers and vice versa"
 - Transactions see only data that was committed when they started.

Isolations-Level	Lost Update	Dirty Reads	Non-Repeatable Reads	Phantom Reads
READ_UNCOMMITTED	prevented	possible	possible	possible
READ_COMMITTED	prevented	prevented	possible	possible
REPEATABLE_READ	prevented	prevented	prevented	possible
SERIALIZABLE	prevented	prevented	prevented	prevented

- Uncommitted transactions may read old values;
 hence, causal consistency but no linearizability!
- Is expensive, because it not only orders the events for the same object but also for an entire transaction!
- Implementations:

Keep both old and new value until commit; let others read the old value

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shared/exclusive locks or multi-version concurrency control (MVCC)

Snapshot Isolation via MVCC



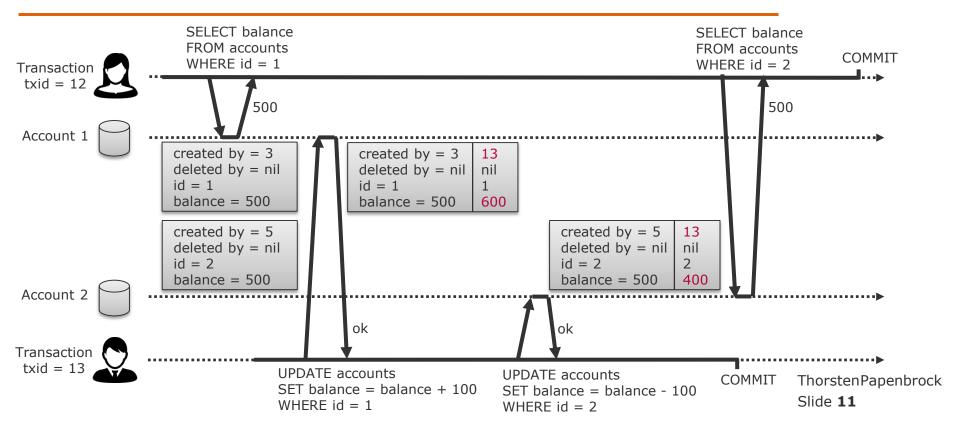
Snapshot Isolation via MVCC

- For each entry (row, key-value pair, ...) store created by and deleted by fields.
- Instead of changing entries directly, always append new versions.
- Transactions can now operate on consistent snapshots (= changes up to a fixed version).
- Algorithm:
 - At transaction start, make a list of all yet un-committed transactions.
 - During execution, ignore all changes made by ...
 - a) un-committed transactions from the start;
 - b) aborted transactions;
 - c) newer transactions (i.e. transactions with higher transaction id).
 - At transaction end, commit all changes; if write conflicts exist, rollback.
- MVCC is an optimistic approach that performs well if transactions do not collide frequently but causes many rollbacks otherwise.

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Snapshot Isolation via MVCC





Isolation



> SFRIALIZABLE

Isolation levels

Snapshot isolation: "readers don't block writers and vice versa"

Isolations-Level	Lost Update	Dirty Reads	Non-Repeatable Reads	Phan	ntom Reads
READ_UNCOMMITTED	prevented	possible	possible	possi	hla
READ_COMMITTED	prevented	prevented	possible	possi	
REPEATABLE_READ	prevented	prevented	prevented	possi	avoid this proble because all rea
SERIALIZABLE	prevented	prevented	prevented	preve	applicants are
 Although it avoid 	locked for writing				

- - Write Skew:
 - Same reads lead to different, non-conflicting but inconsistent writes.
 - Example: Two transactions scan a list of job applicants. Both see that all no applicant was hired, yet, and mark one applicant as hired. ThorstenPapenbrock Slide 13 If they hire different applicants, no conflict is created but the table is inconsistent (two hires for one job).

Consensus for Transaction Commits



Two-Phase Commit (2PC)

"Let's be ACID conform!"

- Goal:
 - Ensure that all nodes consistently commit or abort a transaction.
 - Consensus = "all agree"
- Requirements:
 - One node that acts as a coordinator for a transaction (e.g. leader).
 - Coordinator must be able to generate unique IDs for transactions.
- Steps: (coordinator view)
 - Writing: Send the data to all nodes.
 - Phase 1: Upon global success, send prepare requests to all nodes.
 - Phase 2: Upon global success, send commit request to all nodes.
 - > 2PC transaction commits are blocking operations.

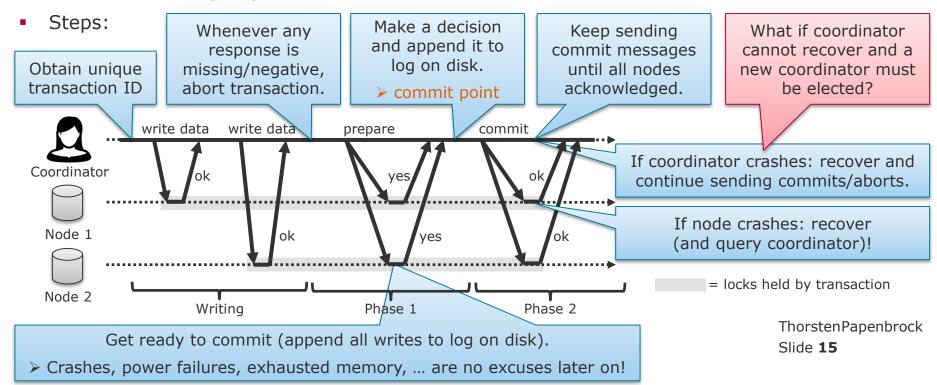
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Consistency and Consensus

Consensus for Transaction Commits



Two-Phase Commit (2PC)



Consensus for Transaction Commits



Three-Phase Commit (3PC)

- Extension of the 2PC protocol that safely handles unrecoverable coordinators
- Idea: Spit the commit phase into two rounds.
- Steps: (coordinator view)
 - Writing: Send the data to all nodes.
 - Phase 1: Upon global success, send prepare requests to all nodes.
 - Phase 2: Upon global success, send pre-commit request to all nodes.
 - Phase 3: Upon global success, send commit request to all nodes.
- If the coordinator dies:
 - The new coordinator asks all nodes for their state.
 - If at least one node is in pre-commit phase, the new coordinator knows that the decision to commit was made and continues to push pre-commit (and then commit) messages.

Without the pre-commit phase, a new coordinator cannot know if a commit+close was already done by some node.

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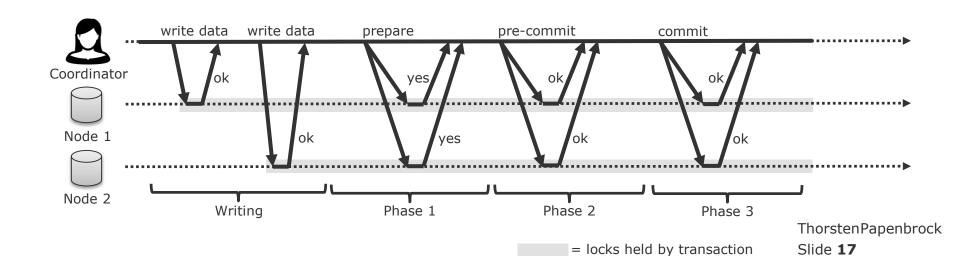
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Consensus for Transaction Commits



Three-Phase Commit (3PC)

Steps:



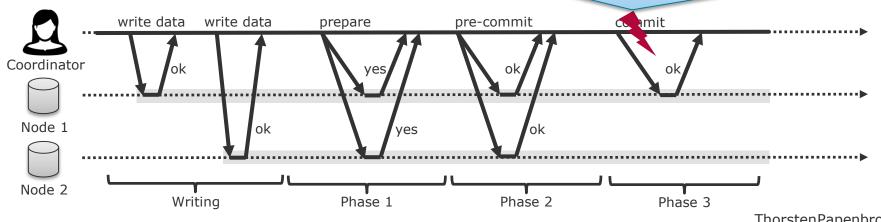
Consensus for Transaction Commits



Three-Phase Commit (3PC)

Steps:

New coordinator sees node 2 in pre-commit state; it re-sends pre-commit to node 1 and continues pushing commit messages to node 1 and 2 (node 1 simply plays the protocol as it knows the transaction is already closed locally).



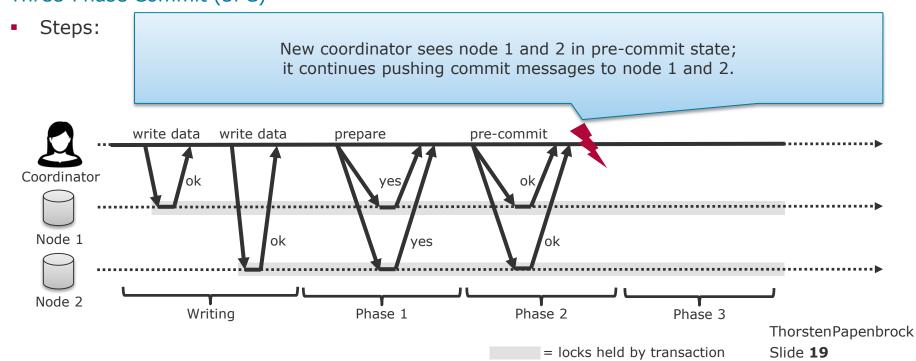
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= locks held by transaction

Consensus for Transaction Commits



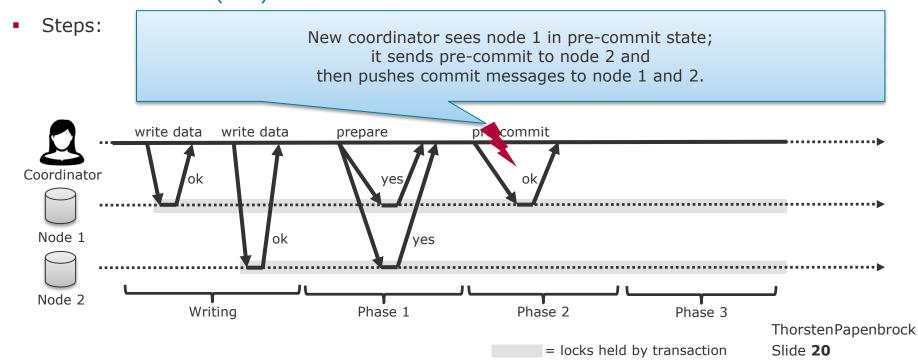
Three-Phase Commit (3PC)



Consensus for Transaction Commits



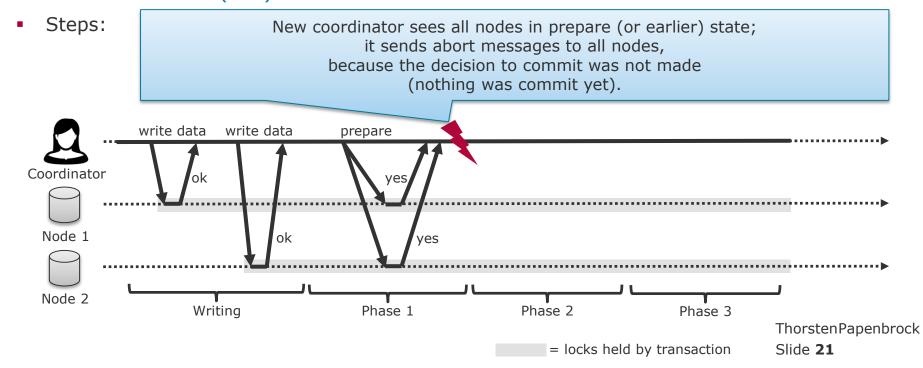
Three-Phase Commit (3PC)



Consensus for Transaction Commits



Three-Phase Commit (3PC)

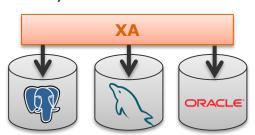


Consensus for Transaction Commits



Two/Three-Phase Commit (2PC / 3PC)

- What if the distributed database is a combination of different DBMS systems?
- eXtended Architecture (XA):
 - Standard for implementing 2PC across multiple DBMSs
 - Implemented as C API with bindings to e.g. Java:
 - Java Transaction API (JTA) supported by various drivers for ...
 - databases, i.e., Java Database Connectivity (JDBC) and
 - message brokers, i.e., Java Message Service (JMS).
 - Used in:
 - Databases: PostgreSQL, MySQL, DB2, SQL Server, Oracle, ...
 - Message Broker: ActiveMQ, HornetQ, MSMQ, IBM MQ, ...



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Consistency and Consensus

Consensus for Transaction Commits



Two/Three-Phase Commit (2PC / 3PC)

- Evaluation 2PC:
 - Expensive: about 10 times slower than single-node transactions in MySQL

2PC is no good consensus protocol for **non-transactional votings.**

- Risky: locks are held indefinitely long if coordinator is lost
- Evaluation 3PC:
 - Expensive: even more expensive than 2PC
 - Blocking: locks are held for long times
 - Complex: automatically electing a new leader if the first failed is consensus voting inside a consensus protocol!

Both are merely used in practical implementations.

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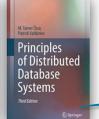
Consistency and Consensus

Summary



- Transaction support costs memory resources:
 - Additional fields (lock or changed/deleted), versions, temporary lists ...
- Transaction support costs computing resources:
 - Setting and checking locks, searching and cleaning versions ...
- Transaction support scales badly in distributed systems:
 - Many actions require voting and/or change propagation.
- Transaction support is an open research area:
 - Achieving consistency for individual values in distributed systems is challenging; achieving the same for sequences of changes is even harder!





If you like to read more about distributed transaction handling, have a look at these two books!

