

# Coordination Avoidance in Distributed Databases

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# Outline

- Coordination avoidance and Invariant Confluence principle
- Examples of Invariant Confluence principle application
- Read-atomic multi-partition transactions
- Conclusions

# Transaction serializability

- Transactions - groups of multiple operations over multiple data items
- Traditional way to deal with concurrent transactions - serializability: a database providing serializability guarantees that the result of executing the transactions is equivalent to some serial execution of the transactions
  - Convenient for programmers - no need to reason about concurrency
  - Inconvenient for databases - serializability requires coordination
- Reasons to avoid coordination - allows for greater availability of a distributed database, lower access latency and greater scalability

# Thesis statement

- Key question posed by Bailis - when is it necessary to use coordination to achieve conflict-free parallel execution, and when is it possible to forego coordination without compromising the safety of parallel transactions
- *Thesis Statement: Many semantic requirements of database-backed applications can be efficiently enforced without coordination, thus improving scalability, latency, and availability.*

# Application side: data invariants

- Bailis proposal: instead of reasoning about data consistency on the level of read and write operations (transactions), **consider what coordination is actually required by a given application**
- Make applications define data invariants and explicitly specify what correctness means to these applications
- Example: “each employee record is linked to a department record”
- Consider if **application requires coordination**, by **taking into account both** application invariants **and** the nature of data transformation in transaction

# Invariant confluence test

- Invariant confluence determines whether the result of executing operations on independent copies of data can be combined (or “merged”) into a single, coherent (i.e., convergent) copy of database state.
- Given **a set of operations, a safety property** that we wish to maintain over all copies of database state, **and a merge function**, invariant confluence tells us **whether coordination-free execution is possible**.

# **Application of Invariant Confluence principle**

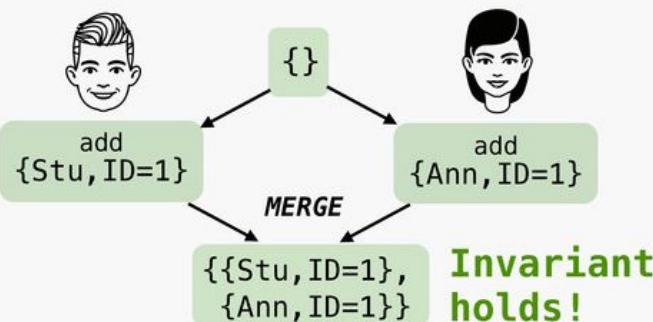
# I-Confluence proof construction

- To show a set of transactions **are not invariant confluent** with respect to an invariant I - **use proof by counterexample**: present two I-T-reachable states with a common ancestor that, when merged, are not I-valid.
- To show a set of transactions **are invariant confluent** with respect to an invariant I **use proof by contradiction**: show that if a state S is not I-valid, merging two I-T-reachable states S1 and S2 with a common ancestor state to produce S implies either one or both of S1 or S2 must not be I-valid.

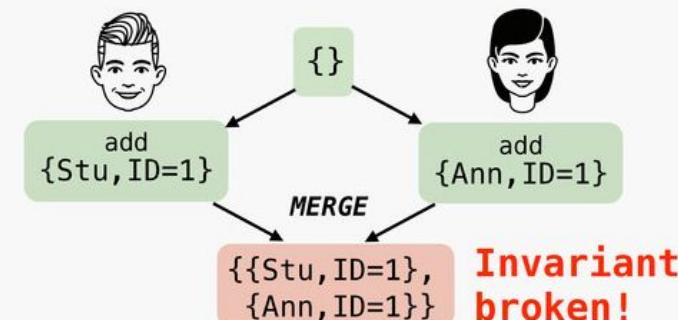
# Invariant confluence test example

- Key question: can invariants be violated by merging independent operations?

**INVARIANT:** User IDs are positive  
**OPERATION:** Save new user  
**MERGE:** Add both records to DB



**INVARIANT:** User IDs are **unique** ←  
**OPERATION:** Save new user  
**MERGE:** Add both records to DB



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# I-Confluence applied to common SQL operations

Invariant	Operation	invariant confluent?	Proof #
Attribute Equality	Any	Yes	1
Attribute Inequality	Any	Yes	2
Uniqueness	Choose specific value	No	3
Uniqueness	Choose some value	Yes	4
AUTO_INCREMENT	Insert	No	5
Foreign Key	Insert	Yes	6
Foreign Key	Delete	No	7
Foreign Key	Cascading Delete	Yes	8
Secondary Indexing	Update	Yes	9
Materialized Views	Update	Yes	10
>	Increment [Counter]	Yes	11
<	Increment [Counter]	No	12
>	Decrement [Counter]	No	13
<	Decrement [Counter]	Yes	14
[NOT] CONTAINS	Any [Set, List, Map]	Yes	15, 16
SIZE=	Mutation [Set, List, Map]	No	17

Table 6.1: Example SQL (top) and ADT invariant confluence along with references to formal proofs in Section 6.2.

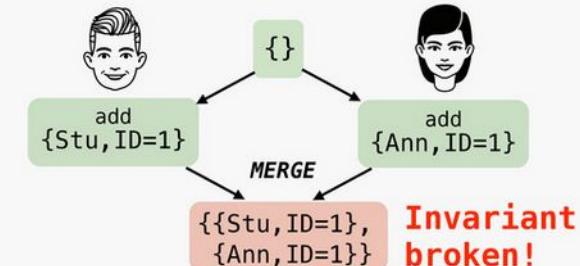
# I-Confluence example 1

Invariant	Operation	invariant confluent?	Proof #
Uniqueness	Choose specific value	No	3
Uniqueness	Choose some value	Yes	4

- Claim: **common uniqueness invariants aren't I-Confluent** (e.g., PRIMARY KEY and UNIQUE constraints).
- Example invariant: user IDs must be unique
- However, **reads and deletions are both invariant confluent** under uniqueness invariants: **reading and removing items cannot introduce duplicates**
- Case 2: the database chooses unique values on behalf of users. If **replicas assign unique IDs within their respective portion of the ID namespace, then merging locally valid states will also be globally valid**

Proof by counterexample:

**INVARIANT:** User IDs are **unique** ←  
**OPERATION:** Save new user  
**MERGE:** Add both records to DB



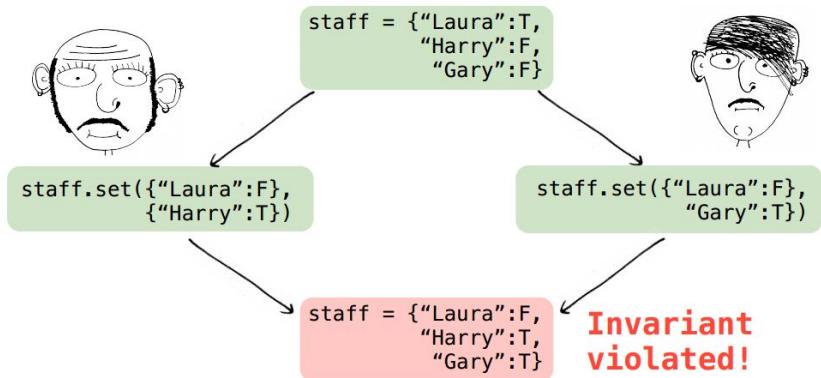
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## Application of I-C principle

# I-Confluence example 2

- Claim 9: writing arbitrary values is not invariant confluent with respect to multi-item uniqueness constraints.
- Proof: by counterexample

Invariant: only one ops on staff at a time  
Operations: change staffing



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# I-Confluence example 3

Invariant	Operation	invariant confluent?	Proof #
Attribute Equality	Any	Yes	1
Attribute Inequality	Any	Yes	2

- Claim: Attributed equality invariants are i-confluent for any operations
- Example: **every user must have a last name assigned**, marking the LNAME column with a NOT NULL constraint
- **Proof by contradiction:** assume two database states  $S_1$  and  $S_2$  are each I-T-reachable under per-record inequality invariant I but that  $I(S_1 \cup S_2)$  is false. Then there must be a  $r \in S_1 \cup S_2$  that violates I (i.e., r has the forbidden value) and such  $r$  must appear in  $S_1$ ,  $S_2$ , or both. But, that would imply that one of  $S_1$  or  $S_2$  is not I-valid under I, a contradiction.

# I-Confluence example 4

Invariant	Operation	invariant confluent?	Proof #
AUTO_INCREMENT	Insert	No	5

Consider the following transactions:

$$T_{1s} := w(x_a = 1); \text{ commit}$$

$$T_{2s} := w(x_b = 3); \text{ commit}$$

and the sequentiality constraint on records:

$$I_s(D) = \{\max(r \in D) - \min(r \in D) = |D| + 1\} \vee \{|D| = 0\}$$

Now,  $I_s$  holds over the empty database ( $I_s(\emptyset) \rightarrow \text{true}$ ), while inserting sequential new records into independent, empty replicas is also valid:

$$T_{1s}(\emptyset) = \{x_a = 1\}, I_u(\{x_a = 1\}) \rightarrow \text{true}$$

$$T_{2s}(\emptyset) = \{x_b = 3\}, I_u(\{x_b = 3\}) \rightarrow \text{true}$$

However, merging these states results in invalid state:

$$I_s(\{x_a = 1\} \sqcup \{x_b = 3\} = \{x_a = 1, x_b = 3\}) \rightarrow \text{false}$$

Therefore,  $\{T_{1s}, T_{2s}\}$  is not invariant confluent under  $I_s$ .

## Overview

# I-Confluence example 5

Invariant	Operation	invariant confluent?	Proof #
>	Increment [Counter]	Yes	11
<	Increment [Counter]	No	12
>	Decrement [Counter]	No	13
<	Decrement [Counter]	Yes	14

**Claim 17** Counter ADT increments are invariant confluent with respect to greater-than constraints .

**Claim 18** Counter ADT increments are not invariant confluent with respect to less-than constraints .

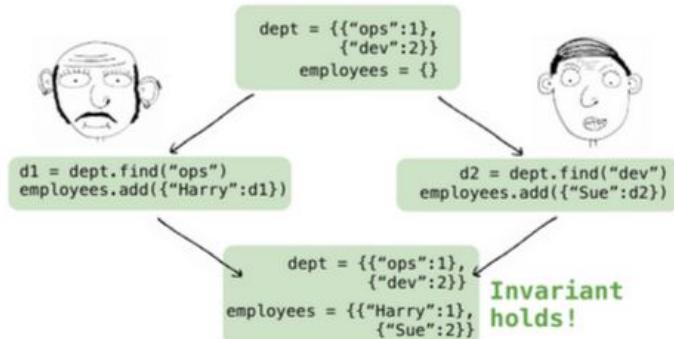
**Claim 19** Counter ADT decrements are not invariant confluent with respect to greater-than constraints .

**Claim 20** Counter ADT decrements are invariant confluent with respect to less-than constraints .

# I-Confluence example 6: foreign keys

- Claim: **Insertions under foreign key constraints are invariant confluent**
- Proof by contradiction:
  - Invalid state: a record missing a corresponding record on the opposite side of the association
  - $S_1$  and  $S_2$  - correct states before the merge (no invalid records)
  - $r$  - invalid record in merged state  $S$
  - As  $S_1$  and  $S_2$  are both valid,  $r$  must have a corresponding foreign key record ( $f$ ) that “disappeared” during merge. Merge (in the current model) does not remove versions, so this is impossible.
- Arbitrary **deletion/modification** of records **is unsafe**: a user might be added to a department that was concurrently deleted

Invariant: each employee is in a department  
Operations: add employees



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# **Read-atomic multi-partition transactions**

# Foreign keys updates

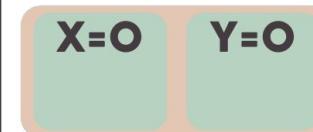
FOREIGN KEY DEPENDENCIES  
NEED  
ATOMIC VISIBILITY  
SEE ALL OF A TXN'S UPDATES, OR NONE OF THEM

I.e. in Facebook graph:  
either we're both friends,  
or neither of us is

HOW TO ACHIEVE ATOMIC VISIBILITY



STRAWMAN: LOCKING



STRAWMAN: LOCKING



STRAWMAN: LOCKING



STRAWMAN: LOCKING



STRAWMAN: LOCKING



"Scalable Atomic Visibility with RAMP Transactions" SIGMOD 2014

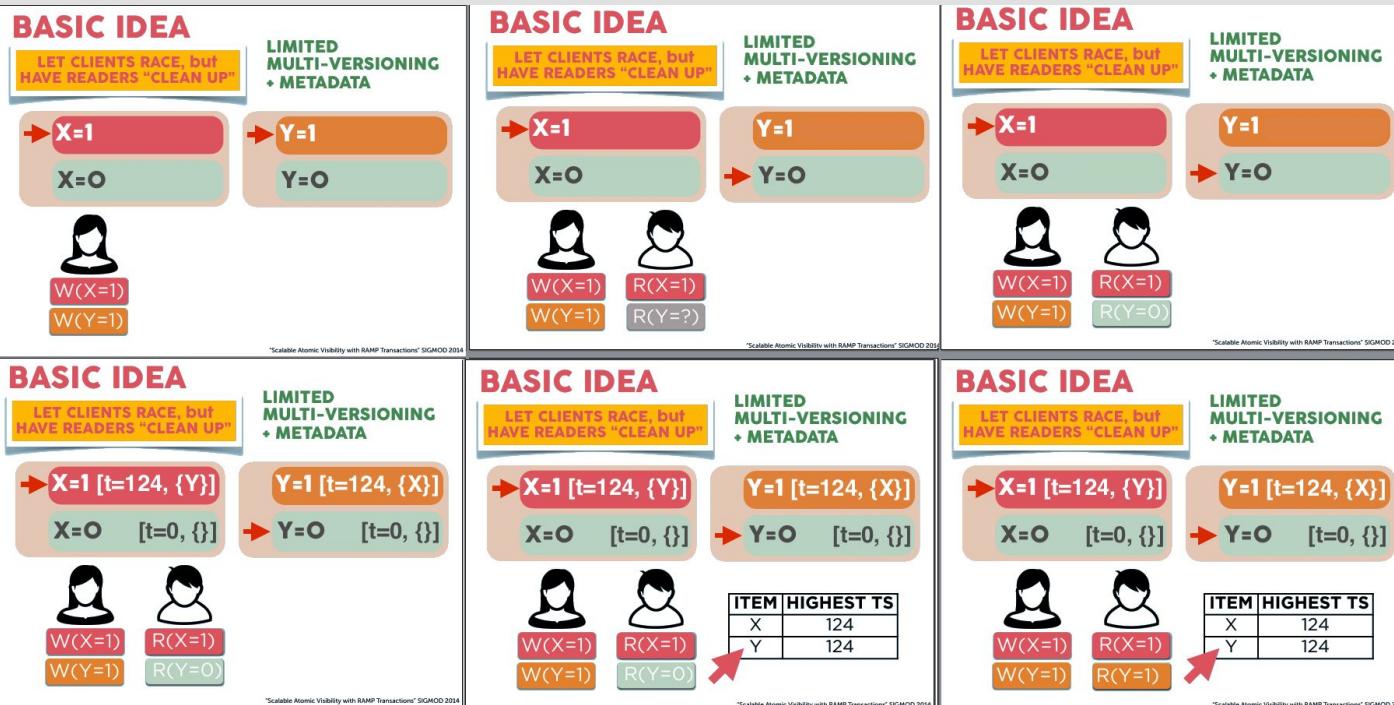
# Read-atomic multi-partition transactions

## RAMP TRANSACTIONS

DECOPLE  
ATOMIC VISIBILITY from  
MUTUAL EXCLUSION

## BASIC IDEA

LET CLIENTS RACE, but  
HAVE READERS "CLEAN UP"



## RAMP transactions

# Another problem solved with RAMP transactions

*Partition by  
primary key (ID)*

*How should we look up by age?*



ID: 123  
AGE: 22



ID: 412  
AGE: 72



ID: 532  
AGE: 42



ID: 892  
AGE: 13



ID: 2345  
AGE: 1



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# Secondary indexing

*Partition by primary key (ID)*



ID: 123  
AGE: 22



ID: 412  
AGE: 72



ID: 532  
AGE: 42



ID: 892  
AGE: 13



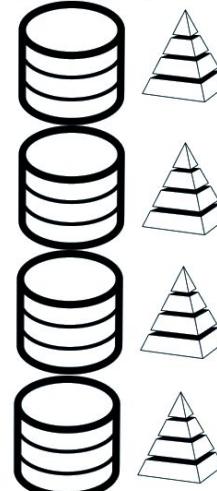
ID: 2345  
AGE: 1



*How should we look up by age?*

## SECONDARY INDEXING

*Partition by primary key (ID)*



*How should we look up by age?*

Option I: Local Secondary Indexing  
Build indexes co-located with primary data

**WRITE ONE SERVER, READ ALL**  
poor scalability

# Secondary indexing

Partition by primary key (ID)

How should we look up by age?



ID: 123  
AGE: 22



ID: 412  
AGE: 72



ID: 532  
AGE: 42



ID: 892  
AGE: 13



ID: 2345  
AGE: 1



## SECONDARY INDEXING

Partition by primary key (ID)

How should we look up by age?



Partition by secondary attribute



Option I: Local Secondary Indexing  
Build indexes co-located with primary data

**WRITE ONE SERVER, READ ALL**  
poor scalability

Option II: Global Secondary Indexing  
Partition indexes by secondary key

**WRITE 2+ SERVERS, READ ONE**  
scalable lookups

Real-world services employ either local secondary indexing (e.g., Espresso [38], Cassandra, and Google Megastore's local indexes [7]) or non-atomic (incorrect) global secondary indexing (e.g., Espresso and Megastore's global indexes, Yahoo! PNUTS's proposed secondary indexes [15]). The former is non-scalable but correct, while the latter is scalable but incorrect.

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# Conclusions

# Conclusions

Traditional database systems suffer from coordination bottlenecks

By understanding application requirements, we can avoid coordination unless *necessary*

We can build systems that actually scale while providing correct behavior

Use of validations (DB constraints) in Rails web-apps:

Name	Occurrences	I-Confluent?
validates_presence_of	1762	Depends
validates_uniqueness_of	440	No
validates_length_of	438	Yes
validates_inclusion_of	201	Yes
validates_numericality_of	133	Yes
validates_associated	39	Depends
validates_email	34	Yes
validates_attachment_content_type	29	Yes
validates_attachment_size	29	Yes
validates_confirmation_of	19	Yes
Other	321	Mixed

Table 6.4: Use of and invariant confluence of built-in validations.

# Thank you!