

# Consistency in Transactional Distributed Databases: Protocols and Testing

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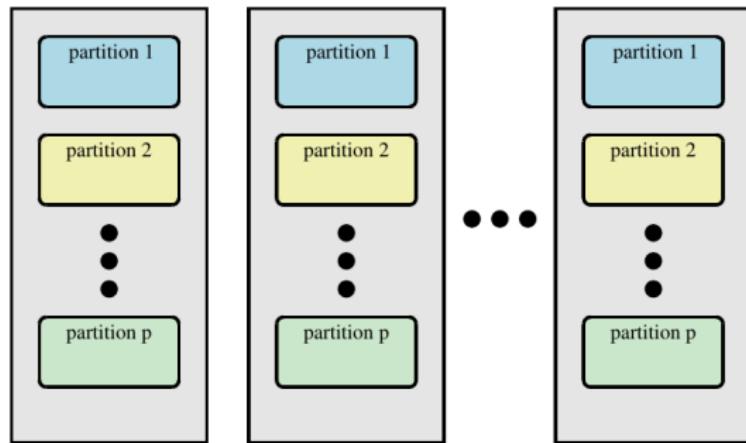
# Background

Centralized Databases *vs.* Distributed Databases



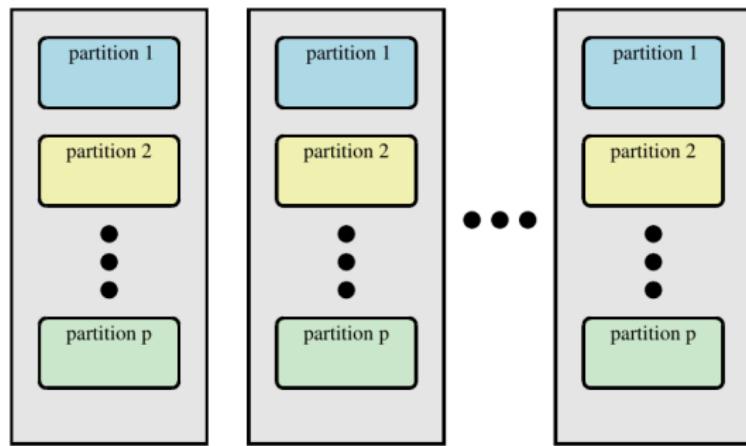
# Data Consistency Problem

The Classic “Data Partition + Data Replication” Architecture



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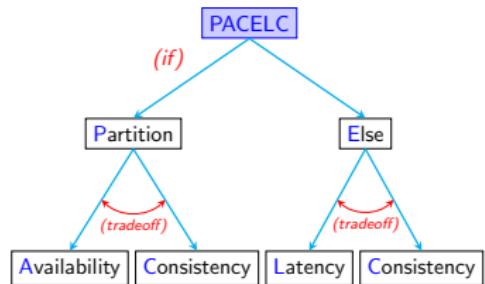
(Distributed) Data Consistency Problem

# Data Consistency Problem

(Strong) Consistency, Availability, Latency、Partition tolerance



CAP Theorem  
(Brewer@PODC2000)



PACELC Tradeoff  
(Abadi@Computer2012)

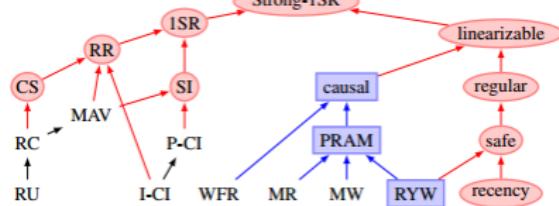
# Data Consistency Problem



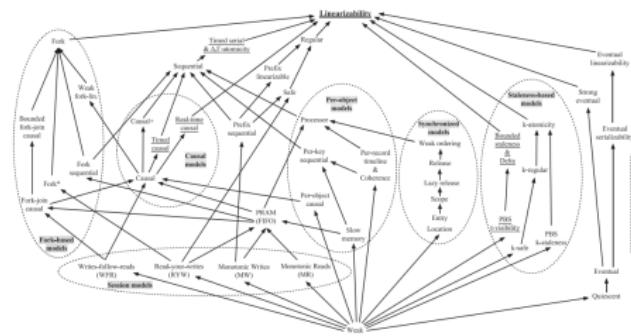
没有一劳永逸的解决方案

## Consistency Models

Use various consistency models to capture these tradeoffs



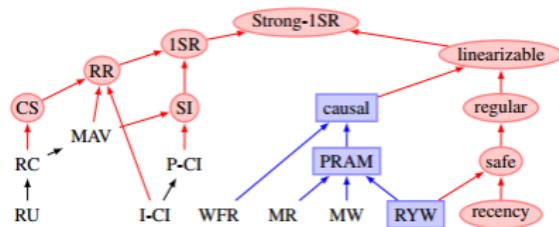
(Bailis@VLDB2012)



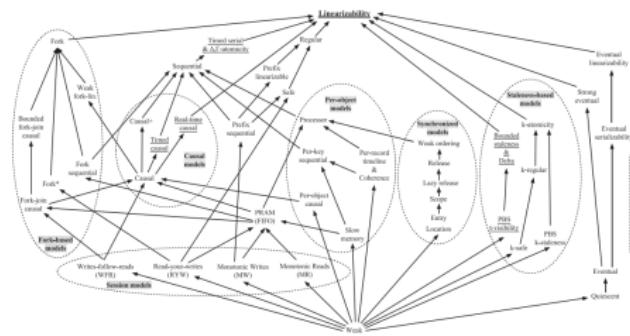
(Viotti@CSUR2016)

## Consistency Models

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(Bailis@VLDB2012)



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# My Researches

On the data consistency theory around consistency models:

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Computability: What is possible or impossible?

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On the data consistency theory around consistency models:

Computability: What is possible or impossible?

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Testing: What is the complexity?

How to design efficient testing algorithms?

Classic problems with the ever-changing requirements

# My Researches (I)

On read/write registers<sup>a</sup> ( $\geq 2014$ )



Distributed Non-transactional Key-Value Stores

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<sup>a</sup> 读写寄存器，也就是读写变量，虽然最初与计算机系统中的“寄存器”概念相关，但已慢慢解耦。

# My Researches (I)

Hengfeng Wei, Marzio De Biasi, Yu Huang, Jiannong Cao, and Jian Lu.  
“Verifying Pipelined-RAM consistency over read/write traces of data replicas”.  
In: *IEEE Trans. Parallel Distrib. Syst. (TPDS’2016)* 27.5 (May 2016),  
pp. 1511–1523. DOI: [10.1109/TPDS.2015.2453985](https://doi.org/10.1109/TPDS.2015.2453985)

Hengfeng Wei, Yu Huang, and Jian Lu. “Probabilistically-Atomic 2-Atomicity:  
enabling almost strong consistency in distributed storage systems”. In: *IEEE  
Trans. Comput. (TC’2017)* 66.3 (Mar. 2017), pp. 502–514. DOI:  
[10.1109/TC.2016.2601322](https://doi.org/10.1109/TC.2016.2601322)

Kaile Huang, Yu Huang, and Hengfeng Wei. “Fine-Grained Analysis on Fast  
Implementations of Distributed Multi-Writer Atomic Registers”. In: *Proceedings  
of the 39th Symposium on Principles of Distributed Computing (PODC’2020)*.  
Association for Computing Machinery, 2020, pp. 200–209. DOI:  
[10.1145/3382734.3405698](https://doi.org/10.1145/3382734.3405698)

# My Researches (II)

On replicated data types<sup>b</sup> ( $\geq 2017$ )



(a) Google Docs



(b) Apache Wave



(c) Wikipedia



(d) LaTeX Editor

<sup>b</sup> 复制数据类型，是经典数据类型的分布式版本，如列表，集合，队列等。

# My Researches (II)

Hengfeng Wei, Yu Huang, and Jian Lu. “Brief Announcement: Specification and Implementation of Replicated List: The Jupiter Protocol Revisited”. In: *Proceedings of the 2018 ACM Symposium on Principles of Distributed Computing*. PODC ’18. ACM, 2018, pp. 81–83

Hengfeng Wei, Yu Huang, and Jian Lu. “Specification and Implementation of Replicated List: The Jupiter Protocol Revisited”. In: *22nd International Conference on Principles of Distributed Systems, OPODIS 2018*. 2018, 12:1–12:16

Xue Jiang, Hengfeng Wei\*, and Yu Huang. “A Generic Specification Framework for Weakly Consistent Replicated Data Types”. In: *International Symposium on Reliable Distributed Systems (SRDS’2020)*. 2020, pp. 143–154. doi: 10.1109/SRDS51746.2020.00022

Ye Ji, Hengfeng Wei\*, Yu Huang, and Jian Lu. “Specifying and verifying CRDT protocols using TLA<sup>+</sup>”. In: *Journal of Software (软件学报 JOS’2020) 31.5 (2020)*, pp. 1332–1352. doi: 10.13328/j.cnki.jos.005956

# My Researches (III)

On distributed transactions<sup>c</sup> ( $\geq 2020$ )



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<sup>c</sup>分布式事务。每个事务由一组操作构成，这些操作要么都成功，要么都失败。 ↩ ↪ ↵

# My Researches (III)

Manuel Bravo, Alexey Gotsman, Borja de Régil, and Hengfeng Wei. “UniStore: A fault-tolerant marriage of causal and strong consistency”. In: *2021 USENIX Annual Technical Conference (USENIX ATC’2021)*. USENIX Association, July 2021, pp. 923–937

# Overview of the Work on UNISTORE

## UNISTORE: A fault-tolerant marriage of causal and strong consistency

Manuel Bravo

Alexey Gotsman

*IMDEA Software Institute*

Borja de Régil

Hengfeng Wei \*

*Nanjing University*

ATC'2021 (CCF A)

UNISTORE is the first **fault-tolerant** and scalable **transactional** data store that **combines** weak and strong consistency.

# Overview of the Work on UNISTORE

Partial Order-Restrictions (PoR) Consistency

$$\text{CC} < \text{PoR} < \text{SER}$$

CC: CAUSALCONSISTENCY; SER: SERIALIZABILITY

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Challenges (I): To ensure **liveness** of the system in the presence of data center failures

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Challenges (I): To ensure **liveness** of the system in the presence of data center failures

Challenges (II): To provide a rigorous correctness **proof** of the protocol which is rather complicated

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ATC'2021 (CCF A)

I am fully responsible for developing a rigorous correctness proof:

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<sup>d</sup>One of these bugs also exists in the well-known Granola protocol proposed by James Cowling and Barbara Liskov, something that had gone unnoticed for 10 years.

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ATC'2021 (CCF A)

I am fully responsible for developing a rigorous correctness proof:

- ▶ Finished a proof of 20 pages contained in the arXiv version
- ▶ Identified several nontrivial bugs in the early versions of the protocol<sup>d</sup>, and proposed solutions to fix them

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# What is UNISTORE?

UNISTORE is a **fast**, **scalable**, and **fault-tolerant**  
**transactional** distributed key-value store  
that supports a combination of weak and strong consistency.

# Why UNI-?

**Weak Consistency:** low latency, high availability,  
but unable to preserve critical application invariants



**Strong Consistency:** easy to preserve critical application invariants,  
but require global synchronization among data centers

# The UNI- Approach

- ▶ Multiple consistency levels coexist in a store

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- ▶ Multiple consistency levels coexist in a store
- ▶ Take **weak consistency** as the default and the baseline
- ▶ Programmers can choose the transactions that should be executed under **strong consistency**
  - ▶ e.g., if the execution of a set of transactions may violate the application violations

# The UNI- Approach

Partial Order-Restrictions (PoR) Consistency  
(Li@OSDI'2012, Li@ACT'2018)

# The UNI- Approach

## Partial Order-Restrictions (PoR) Consistency (Li@OSDI'2012, Li@ACT'2018)

- ▶ PoR allows programmers to classify transactions as either causal or strong.

# The UNI- Approach

## Partial Order-Restrictions (PoR) Consistency (Li@OSDI'2012, Li@ACT'2018)

- ▶ PoR allows programmers to classify transactions as either causal or strong.
- ▶ Causal transactions satisfy CAUSALCONSISTENCY:
  - ▶ Clients see updates in an order that respects the potential causality between them.
  - ▶ Causally independent transactions can be executed concurrently.

# The UNI- Approach

## Partial Order-Restrictions (PoR) Consistency (Li@OSDI'2012, Li@ACT'2018)

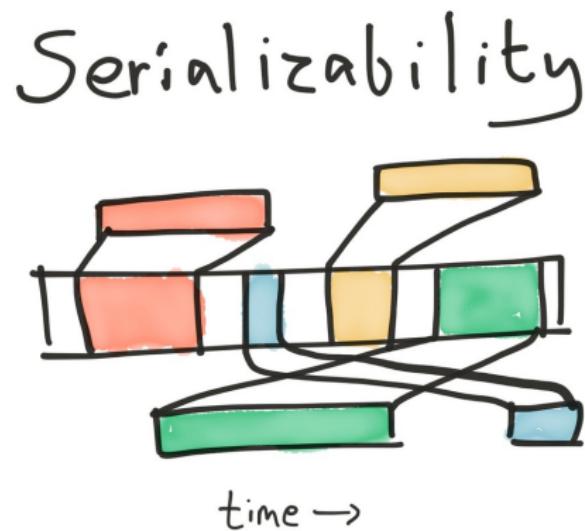
- ▶ PoR allows programmers to classify transactions as either **causal** or **strong**.
- ▶ **Causal** transactions satisfy **CAUSALCONSISTENCY**:
  - ▶ Clients see updates in an order that respects the **potential causality** between them.
  - ▶ **Causally independent** transactions can be executed concurrently.
- ▶ Programmers use **strong** transactions to enforce orders between some causally independent transactions.

# CAUSAL CONSISTENCY

TODO: +fig Clients see updates in an order that respects the potential causality between them.

Causally independent transactions can be executed concurrently.

# SERIALIZABILITY



All transactions seem to be executed in some **sequential** order.

# A Banking Application

DEPOSIT

WITHDRAW

QUERY

INTEREST



Invariant:  $\text{balance} \geq 0$

# A Banking Application

DEPOSIT    WITHDRAW    QUERY    INTEREST



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DEPOSIT operations can be executed under CAUSALCONSISTENCY.

# A Banking Application

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DEPOSIT(50)

# A Banking Application

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However, CAUSALCONSISTENCY also allows two causally independent **WITHDRAW** to execute concurrently, without knowing each other.

# A Banking Application

DEPOSIT    WITHDRAW    QUERY    INTEREST



Invariant:  $\text{balance} \geq 0$

However, CAUSALCONSISTENCY also allows two causally independent **WITHDRAW** to execute concurrently, without knowing each other.

WITHDRAW(60)

# The PoR Approach

- ▶ The programmer provides a symmetric **conflict relation**  $\bowtie$  on transactions.
- ▶ Any transaction involved in the conflict relation is marked **strong**.
- ▶ PoR ensures that conflicting transactions are executed serially.

# A Banking Application

DEPOSIT    WITHDRAW    QUERY    INTEREST



Invariant:  $\text{balance} \geq 0$

Only **WITHDRAW** are marked **strong**.

Declaring that strong transactions

including **WITHDRAW** on the same account conflict.

# Related Work

non-transactional

Hagit Attiya and Roy Friedman. “A Correctness Condition for High-Performance Multiprocessors”. In: *SIAM Journal on Computing* 27.6 (1998), pp. 1637–1670

Hagit Attiya, Soma Chaudhuri, Roy Friedman, and Jennifer L. Welch. “Shared Memory Consistency Conditions for Nonsequential Execution: Definitions and Programming Strategies”. In: *SIAM Journal on Computing* 27.1 (1998), pp. 65–89



# Related Work

not fault-tolerant

Valter Balegas, Sérgio Duarte, Carla Ferreira, Rodrigo Rodrigues, Nuno Preguiça, Mahsa Najafzadeh, and Marc Shapiro. “Putting Consistency Back into Eventual Consistency”. In: *Proceedings of the Tenth European Conference on Computer Systems*. EuroSys ’15. 2015

Cheng Li, Daniel Porto, Allen Clement, Johannes Gehrke, Nuno Preguiça, and Rodrigo Rodrigues. “Making Geo-replicated Systems Fast As Possible, Consistent when Necessary”. In: *Proceedings of the 10th USENIX Conference on Operating Systems Design and Implementation*. OSDI ’12. 2012, pp. 265–278

Cheng Li, Nuno Preguiça, and Rodrigo Rodrigues. “Fine-Grained Consistency for Geo-Replicated Systems”. In: *Proceedings of the 2018 USENIX Conference on Usenix Annual Technical Conference*. USENIX ATC ’18. 2018, pp. 359–371

# Liveness of UNISTORE



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How to ensure **liveness** despite data center failures?



# Liveness of UNISTORE

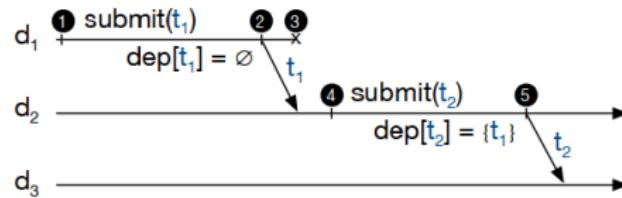
How to ensure **liveness** despite data center failures?



A transaction  $t \in T$  that is either **strong** or originates at a correct data center eventually become **visible** at all correct data centers.

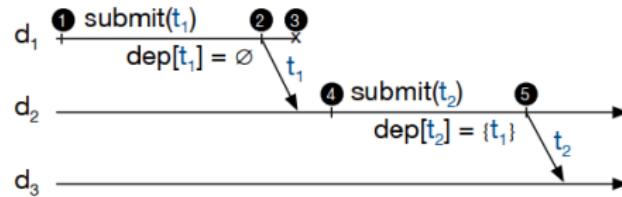
# Liveness of Causal Transactions

$d_1$  crashes before  $t_1$  is replicated to  $d_3$ .



# Liveness of Causal Transactions

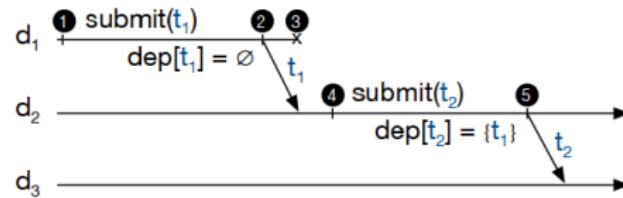
$d_1$  crashes before  $t_1$  is replicated to  $d_3$ .



Transaction  $t_2$  may never become visible at  $d_3$ .

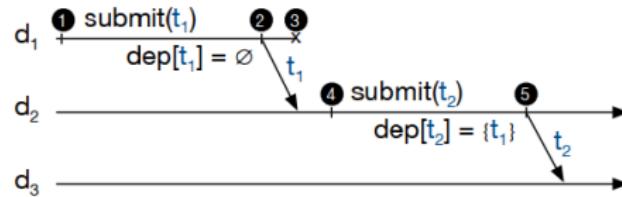
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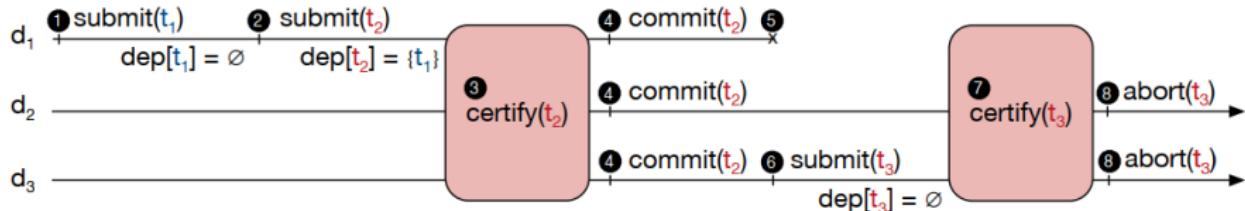
$d_1$  crashes before  $t_1$  is replicated to  $d_3$ .



$d_2$  need to **forward** causal transactions to other data centers.

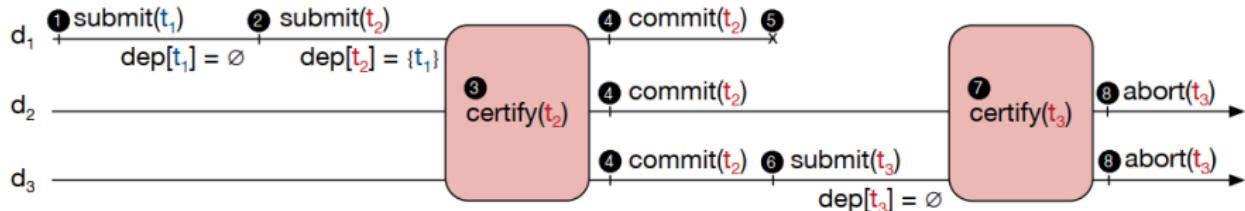
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# Liveness of Strong Transactions

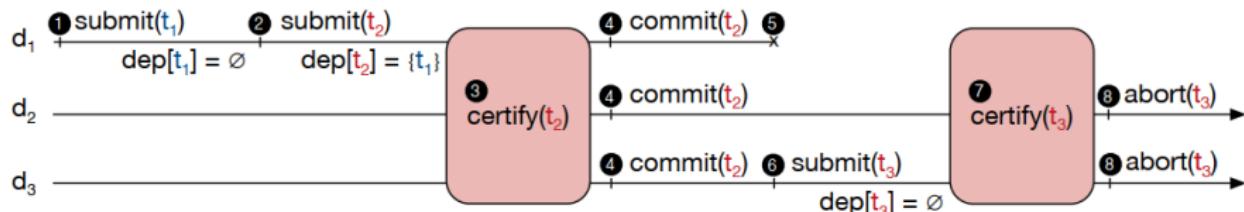
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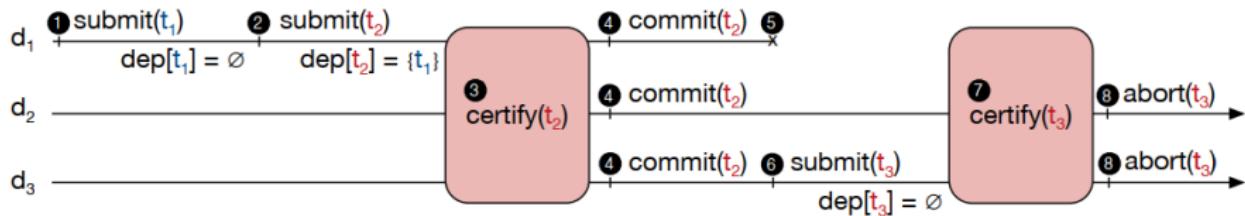


$t_2$  will never be visible at  $d_3$ .

No transaction  $t_3$  conflicting with  $t_2$  can commit.

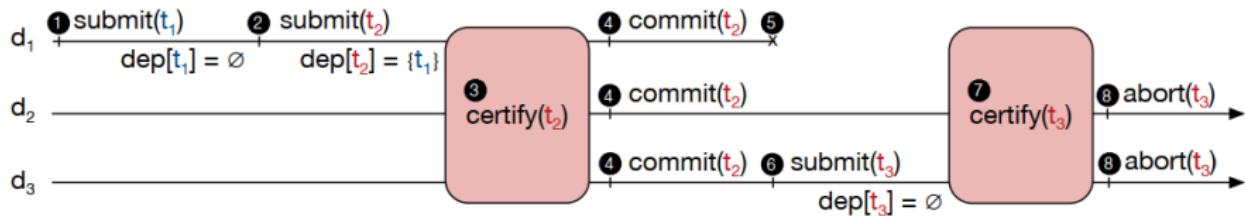
# Liveness of Strong Transactions

A **strong** transaction should wait for all its **causal dependencies** to become **uniform** before committing.  
(eventually become visible at all correct data centers)



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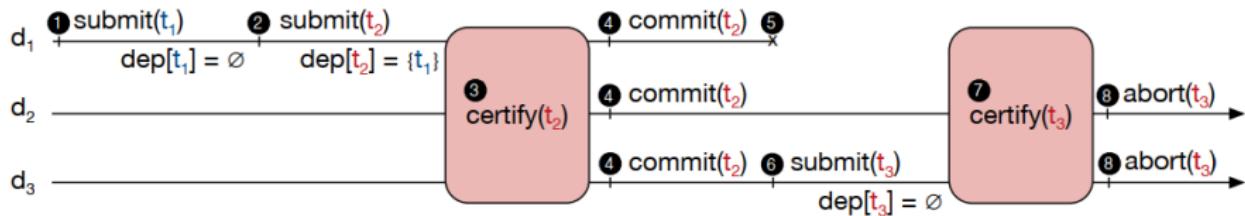
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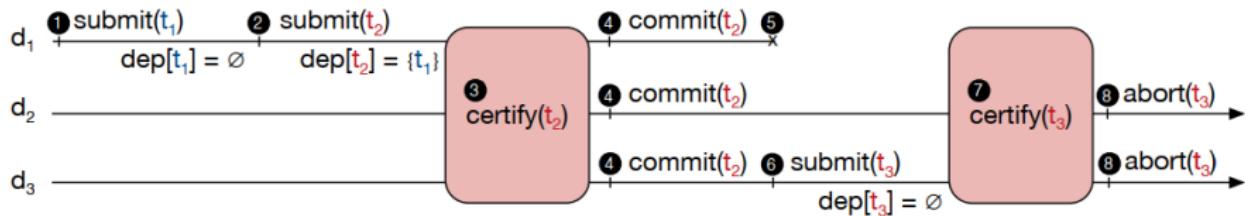


t<sub>1</sub> will eventually be visible at d<sub>3</sub>.

t<sub>2</sub> will eventually be visible at d<sub>3</sub>.

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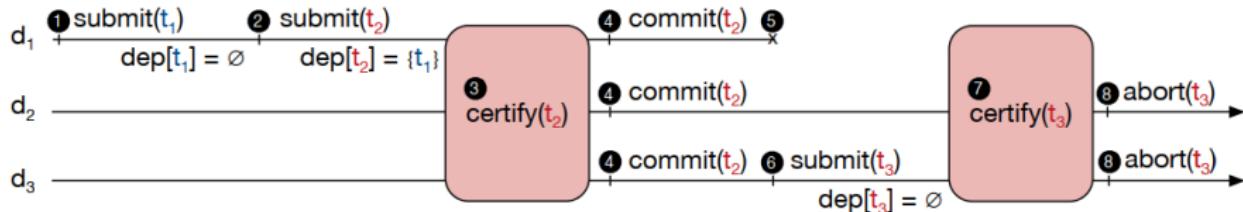
$t_1$  will eventually be visible at  $d_3$ .

$t_2$  will eventually be visible at  $d_3$ .

$t_3$  may be committed at  $d_3$ .

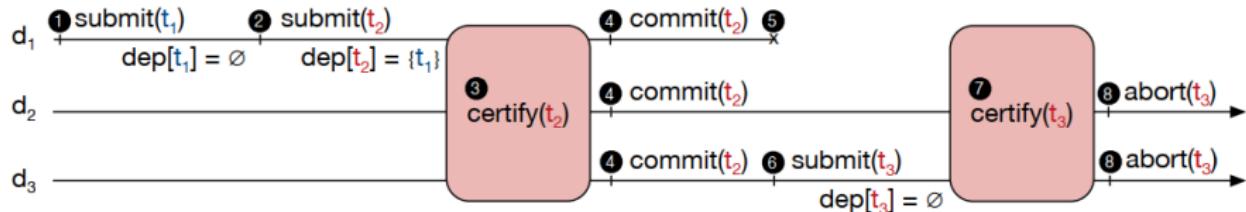
# Minimizing the Latency of Strong Transactions

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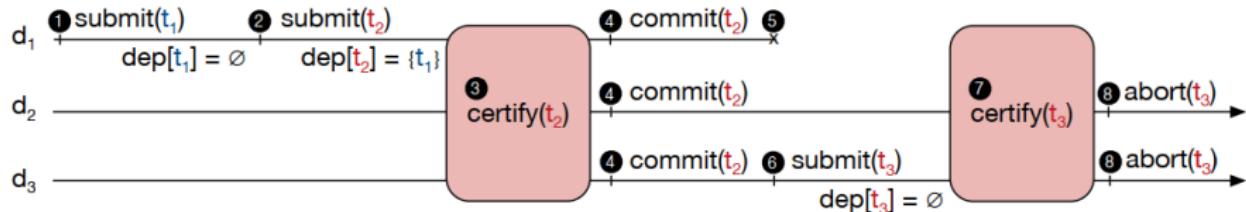
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$t_2$  waits for local  $t_1$  to become uniform before committing.

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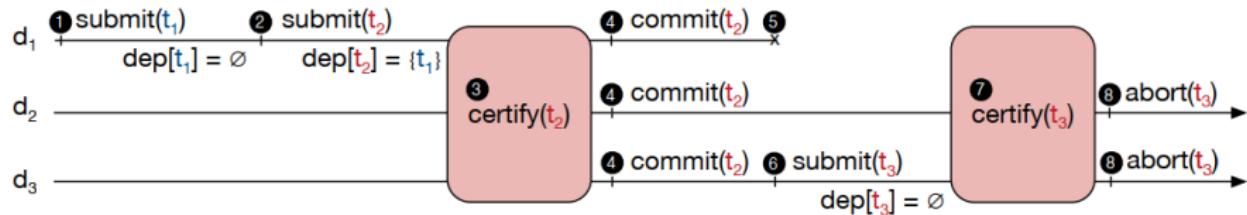


$t_2$  waits for **local**  $t_1$  to become uniform before committing.

However, it may cost too much to wait for **remote causal dependencies** to become uniform.

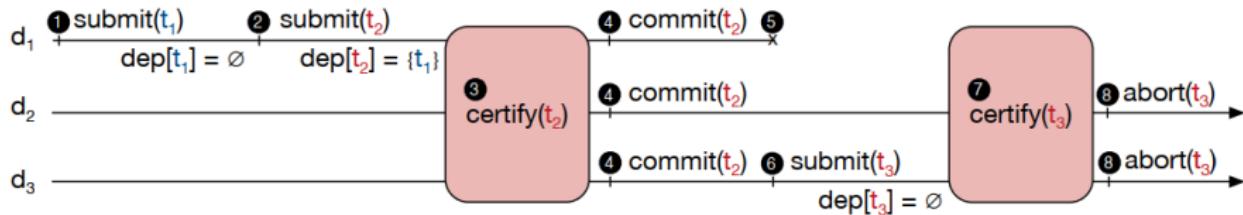
# Minimizing the Latency of Strong Transactions

Let **causal transactions** be executed on an (almost) **uniform snapshot** that may be slightly in the past.



# Minimizing the Latency of Strong Transactions

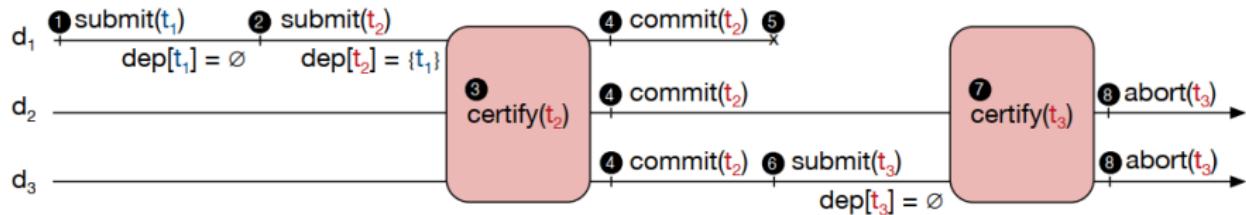
Let **causal transactions** be executed on an (almost) **uniform snapshot** that may be slightly in the past.



Make a **remote causal transaction** visible only **after it is uniform**.

# Minimizing the Latency of Strong Transactions

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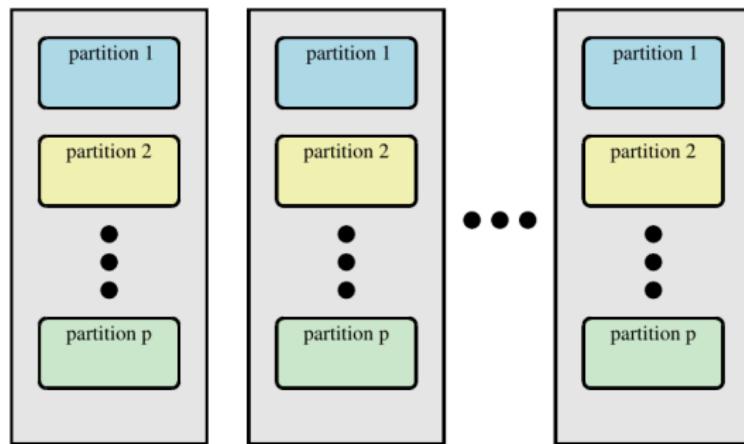
Make a **remote causal transaction** visible only **after it is uniform**.

A **strong** transaction does *not* need to wait for **remote causal dependencies** to become uniform before committing.

# System Model and Notations

$\mathcal{D} = \{1, \dots, D\}$  : the set of data centers

$\mathcal{P} = \{1, \dots, N\}$  : the set of (logical) partitions



# Fault-tolerant Causal Consistency Protocol

UNISTORE relies on various vector clocks  $\in [\mathcal{D} \rightarrow \mathbb{N}]$  for

- ▶ Tracking causality (*commitVec*)
- ▶ Representing causally consistent snapshots (*snapshotVec*)
- ▶ Tracking what is replicated where

# Tracking What is Replicated Where

Make a **remote causal transaction** visible only **after it is uniform**.

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## Definition (Uniform)

A transaction is **uniform** if both the transaction and its causal dependencies are guaranteed to be eventually replicated at all correct data centers.

# Tracking What is Replicated Where

Make a **remote causal transaction** visible only **after it is uniform**.

## Definition (Uniform)

A transaction is **uniform** if both the transaction and its causal dependencies are guaranteed to be eventually replicated at all correct data centers.

A transaction is considered **uniform**  
once it is visible at  $f + 1$  data centers.

# Tracking What is Replicated Where

Each replica maintains three vectors:

**knownVec:** track the set of transactions replicated at this replica

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# Tracking What is Replicated Where

Each replica maintains three vectors:

**knownVec**: track the set of transactions replicated at this replica

**stableVec**: track the set of transactions replicated at its local data center

**uniformVec**: track the set of transactions replicated at  $f + 1$  data centers

# Fault-tolerant Causal Consistency Protocol

- ▶ Causal transactions are executed on a uniform snapshot.
- ▶ Read-only transactions returns immediately.
- ▶ Update transactions commit using a variant of 2PC protocol.

# Adding Strong Transactions

PoR ensures that conflicting transactions are executed serially.

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Assigning to each **strong transaction** a scalar *strong timestamp*.

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$$[\mathcal{D} \rightarrow \mathbb{N}] \quad vs. \quad [\mathcal{D} \cup \{\text{strong}\} \rightarrow \mathbb{N}]$$

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PoR ensures that conflicting transactions are executed serially.

Assigning to each **strong transaction** a scalar *strong timestamp*.

$$[\mathcal{D} \rightarrow \mathbb{N}] \quad vs. \quad [\mathcal{D} \cup \{\text{strong}\} \rightarrow \mathbb{N}]$$

$$\text{commitVec} \in [\mathcal{D} \cup \{\text{strong}\} \rightarrow \mathbb{N}]$$

# Adding Strong Transactions

Each replica maintains three vectors:

`knownVec[strong]`: track the set of **strong** transactions replicated at this replica

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We do not extend `uniformVec`, since the commit protocol for `strong` transactions automatically guarantees their uniformity.

# Adding Strong Transactions

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- ▶ A **strong** transaction waits for local causal dependencies to become uniform before committing.
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  - ▶ First *speculatively* execute **strong** transactions locally
  - ▶ Then certify them globally using a *transaction certification service* (TCS)

# Strong Consistency Protocol: Commit

## Multi-Shot Distributed Transaction Commit

Gregory Chockler

Royal Holloway, University of London, UK

Alexey Gotsman<sup>1</sup>

IMDEA Software Institute, Madrid, Spain

## White-Box Atomic Multicast

Alexey Gotsman  
IMDEA Software Institute

Anatole Lefort  
Télécom SudParis

Gregory Chockler  
Royal Holloway, University of London

- ▶ 2PC across partitions
- ▶ Paxos among replicas of each partition
- ▶ Uses white-box optimizations that minimize the commit latency

# Proof of Correctness

We use the  $(vis, ar)$  formal specification framework.

## A Framework for Transactional Consistency Models with Atomic Visibility

Andrea Cerone, Giovanni Bernardi, and Alexey Gotsman

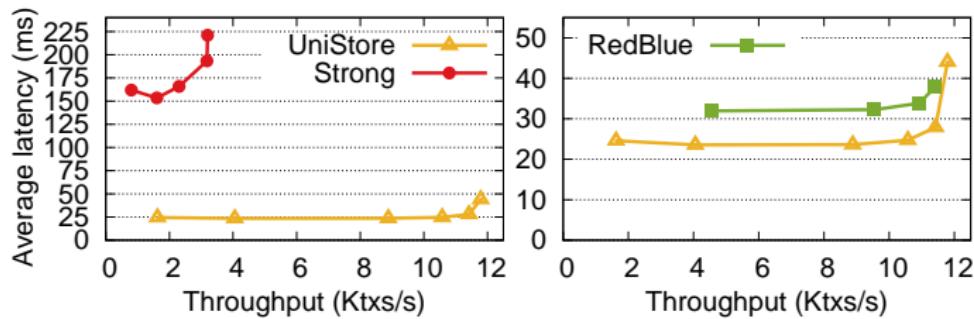
IMDEA Software Institute, Madrid, Spain

# Proof of Correctness

The key is to **construct** the appropriate “visibility” (*vis*) relation and the “arbitration” (*ar*) relation between transactions such that

- ▶ The desired properties of various vector clocks hold.
- ▶ The **liveness** property holds (hard).

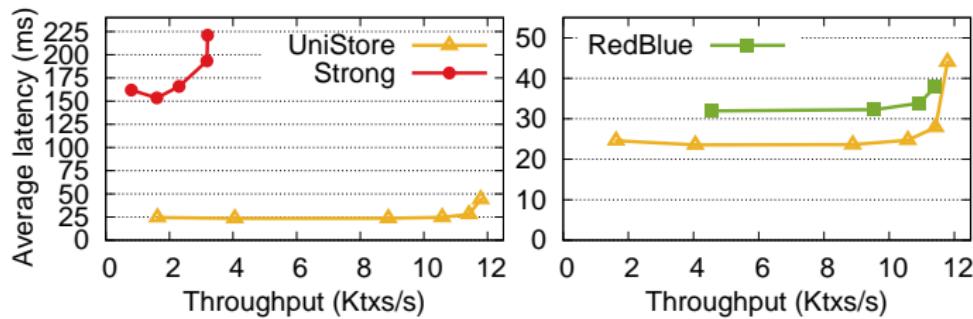
# *Q1 : Does UNISTORE Combine Causal and Strong Consistency Effectively?*



RUBiS benchmark: throughput vs. average latency.

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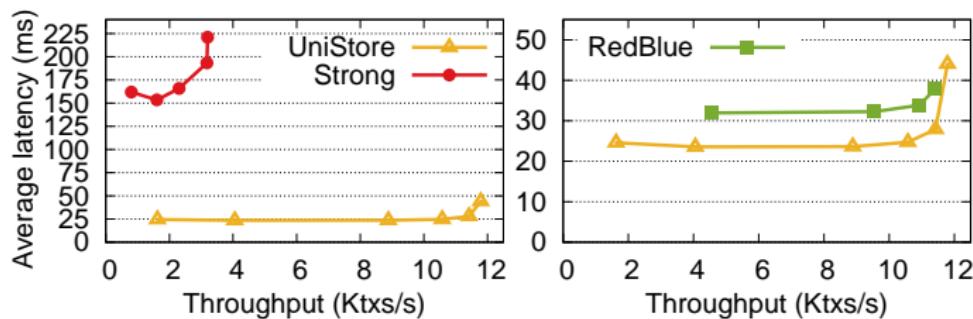
Throughput: 72% and 183% higher than REDBLUE and STRONG,  
45% lower than CAUSAL



RUBiS benchmark: throughput vs. average latency.

# *Q1 : Does UNISTORE Combine Causal and Strong Consistency Effectively?*

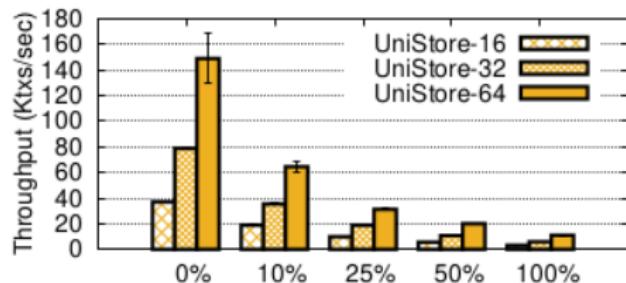
Throughput: 72% and 183% higher than REDBLUE and STRONG,  
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RUBiS benchmark: throughput vs. average latency.

Average latency: 16.5ms of UNISTORE *vs.* 80.4ms of STRONG,  
comparable to that of REDBLUE

## *Q<sub>2</sub>* : How does UNISTORE Scale with the Number of Machines?

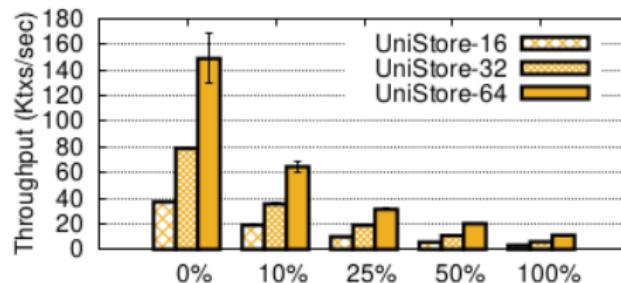


Scalability when varying the ratio of strong transactions with uniform data access  
(top) and under contention (bottom).

## *Q<sub>2</sub>* : How does UNISTORE Scale with the Number of Machines?

Scales almost linearly (top):

9.76% throughput drop compared to the optimal scalability.

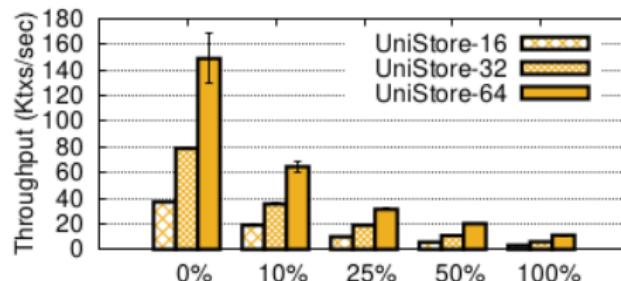


Scalability when varying the ratio of strong transactions with uniform data access (top) and under contention (bottom).

## *Q<sub>2</sub>* : How does UNISTORE Scale with the Number of Machines?

Scales almost linearly (top):

9.76% throughput drop compared to the optimal scalability.



Scalability when varying the ratio of strong transactions with uniform data access (top) and under contention (bottom).

Under contention (bottom): 17.15% throughput drop

## *Q<sub>3</sub>* : What is the Cost of Uniformity?

Throughput penalty of tracking  
uniformity

7.97% throughput drop on average

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Update visibility delay (ms)

5ms ~ 92ms at the 90<sup>th</sup> percentile

# Conclusion

UNISTORE is a **fast**, **scalable**, and **fault-tolerant**  
**transactional** distributed key-value store  
that supports a **combination of weak and strong consistency**.

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UNISTORE is a **fast**, **scalable**, and **fault-tolerant**  
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“We expect the key ideas in UNISTORE to pave the way  
for **practical systems** that combine causal and strong consistency.”



Hengfeng Wei (hfwei@nju.edu.cn)

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